

Red Hat Build of OptaPlanner 8.38

Developing solvers with Red Hat Build of OptaPlanner

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Abstract

This document describes how to develop solvers with Red Hat Build of OptaPlanner to find the optimal solution to planning problems.

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PREFACE

You can use Red Hat Build of OptaPlanner to develop solvers that determine the optimal solution to planning problems. OptaPlanner is a built-in component of Red Hat Build of OptaPlanner. You can use solvers as part of your services in Red Hat Build of OptaPlanner to optimize limited resources with specific constraints.

MAKING OPEN SOURCE MORE INCLUSIVE

Red Hat is committed to replacing problematic language in our code, documentation, and web properties. We are beginning with these four terms: master, slave, blacklist, and whitelist. Because of the enormity of this endeavor, these changes will be implemented gradually over several upcoming releases. For more details, see *our CTO Chris Wright's message*.

PART I. RELEASE NOTES FOR RED HAT BUILD OF OPTAPLANNER 8.38

These release notes list new features and provide upgrade instructions for Red Hat Build of OptaPlanner 8.38.

CHAPTER 1. UPGRADING FROM OPTAPLANNER 8.13 TO RED HAT BUILD OF OPTAPLANNER 8.38

To upgrade from OptaPlanner 8.13 to Red Hat Build of OptaPlanner 8.38, merge the previous versions of OptaPlanner in the following order:

- From 8.13.0.Final to 8.14.0.Final
- From 8.19.0.Final to 8.20.0.Final
- From 8.22.0.Final to 8.23.0.Final
- From 8.27.0.Final to 8.28.0.Final
- From 8.28.0.Final to 8.29.0.Final
- From 8.31.0.Final to 8.32.0.Final
- From 8.33.0.Final to 8.34.0.Final
- From 8.36.0.Final to 8.37.0.Final

Procedure

- 1. Open the OptaPlanner Upgrade Recipe 8 page in a browser.
- 2. Complete the instructions for first version that you want to upgrade, for example **From 8.13.0.Final to 8.14.0.Final**.
- 3. Repeat the instructions until you have upgraded to 8.37.0.Final.

CHAPTER 2. RED HAT BUILD OF OPTAPLANNER 8.38 NEW FEATURES

This section highlights new features in Red Hat Build of OptaPlanner 8.38.



NOTE

Bavet is a feature used for fast score calculation. Bavet is currently only available in the community version of OptaPlanner. It is not available in Red Hat Build of OptaPlanner 8.38.

2.1. PERFORMANCE IMPROVEMENTS IN PILLAR MOVES AND NEARBY SELECTION

OptaPlanner can now auto-detect situations where multiple pillar move selectors can share a precomputed pillar cache and reuse it instead of recomputing the pillar cache for each move selector. If you combine pillar moves, for example, **PillarChangeMove** and **PillarSwapMove**, you should see significant performance inprovements.

This also applies if you use nearby selection. OptaPlanner can now auto-detect situations where a precomputed distance matrix can be shared between multiple move selectors, which saves memory and CPU processing time.

As a consequence of this enhancement, implementations of the following interfaces are expected to be stateless:

- org.optaplanner.core.impl.heuristic.selector.common.nearby.NearbyDistanceMeter
- org.optaplanner.core.impl.heuristic.selector.common.decorator.SelectionFilter
- org.optaplanner.core.impl.heuristic.selector.common.decorator.SelectionProbabilityWeig htFactory
- org.optaplanner.core.impl.heuristic.selector.common.decorator.SelectionSorter
- org.optaplanner.core.impl.heuristic.selector.common.decorator.SelectionSorterWeightFa ctory

In general, if solver configuration asks the user to implement an interface, the expectation is that the implementation will be stateless or not try to include an external state. With the these performance improvements, failing to follow this requirement will result in subtle bugs and score corruption because the solver will now reuse these instances as it sees fit.

2.2. OPTAPLANNER CONFIGURATION IMPROVEMENT

Various configuration classes, such as **EntitySelectorConfig** and **ValueSelectorConfig**, contain new builder methods which make it easier to replace XML-based solver configuration with fluent Java code.

2.3. PLANNINGLISTVARIABLE SUPPORT FOR K-OPT MOVES

A new move selector for list variables, **KOptListMoveSelector**, has been added. **KOptListMoveSelector** selects a single entity, removes **k** edges from its route, and adds **k** new edges from the removed edges' endpoints. **KOptListMoveSelector** can help the solver escape local optima in vehicle routing problems.

2.4. SOLUTIONMANAGER SUPPORT FOR UPDATING SHADOW VARIABLES

SolutionManager (formerly **ScoreManager**) methods such as **explain(solution)** and **update(solution)** received a new overload with an extra argument, **SolutionUpdatePolicy**. This is useful for users who load their solutions from persistent storage (such as a relational database), where these solutions do not include the information carried by shadow variables or the score. By calling these new overloads and picking the right policy, OptaPlanner automatically computes values for all of the shadow variables in a solution or recalculates the score, or both.

Similarly, **ProblemChangeDirector** received a new method called **updateShadowVariables()**, so that you can update shadow variables on demand in real-time planning.

2.5. VALUE RANGE AUTO-DETECTION

In most cases, links between planning variables and value ranges can now be automatically detected. Therefore, **@ValueRangeProvider** no longer needs to provide an ID property. Likewise, planning variables no longer need to reference value range providers through the **valueRangeProviderRefs** property.

No code changes or configuration changes are required. Users who prefer clarity over brevity can continue to explicitly reference value range providers.

PART II. GETTING STARTED WITH RED HAT BUILD OF OPTAPLANNER

As a business rules developer, you can use Red Hat Build of OptaPlanner to find the optimal solution to planning problems based on a set of limited resources and under specific constraints.

Use this document to start developing solvers with OptaPlanner.

CHAPTER 3. INTRODUCTION TO RED HAT BUILD OF OPTAPLANNER

OptaPlanner is a lightweight, embeddable planning engine that optimizes planning problems. It helps normal Java programmers solve planning problems efficiently, and it combines optimization heuristics and metaheuristics with very efficient score calculations.

For example, OptaPlanner helps solve various use cases:

- *Employee/Patient Rosters*: It helps create timetables for nurses and keeps track of patient bed management.
- *Educational Timetables*: It helps schedule lessons, courses, exams, and conference presentations.
- *Shop Schedules*: It tracks car assembly lines, machine queue planning, and workforce task planning.
- *Cutting Stock*: It minimizes waste by reducing the consumption of resources such as paper and steel.

Every organization faces planning problems; that is, they provide products and services with a limited set of constrained resources (employees, assets, time, and money).

OptaPlanner is open source software under the Apache Software License 2.0. It is 100% pure Java and runs on most Java virtual machines (JVMs).

3.1. BACKWARDS COMPATIBILITY

OptaPlanner separates the API and the implementation:

- Public API: All classes in the package namespace org.optaplanner.core.api, org.optaplanner.benchmark.api, org.optaplanner.test.api and org.optaplanner.persistence.api are 100% backwards compatible in future minor and patch releases. In rare instances, if the major version number changes, a few specific classes might have a few backwards incompatible changes, but those changes will be clearly documented in the upgrade recipe.
- XML configuration: The XML solver configuration is backwards compatible for all elements, except for elements that require the use of non-public API classes. The XML solver configuration is defined by the classes in the package namespace **org.optaplanner.core.config** and **org.optaplanner.benchmark.config**.
- Implementation classes: All other classes are *not* backwards compatible. They will change in future major or minor releases. The upgrade recipe describes relevant changes and on how to resolve them when upgrading to a newer version.

3.2. PLANNING PROBLEMS

A *planning problem* has an optimal goal, based on limited resources and under specific constraints. Optimal goals can be any number of things, such as:

- Maximized profits the optimal goal results in the highest possible profit.
- Minimized ecological footprint the optimal goal has the least amount of environmental impact.

• Maximized satisfaction for employees or customers - the optimal goal prioritizes the needs of employees or customers.

The ability to achieve these goals relies on the number of resources available. For example, the following resources might be limited:

- Number of people
- Amount of time
- Budget
- Physical assets, for example, machinery, vehicles, computers, buildings

You must also take into account the specific constraints related to these resources, such as the number of hours a person works, their ability to use certain machines, or compatibility between pieces of equipment.

Red Hat Build of OptaPlanner helps Java programmers solve constraint satisfaction problems efficiently. It combines optimization heuristics and metaheuristics with efficient score calculation.

3.3. NP-COMPLETENESS IN PLANNING PROBLEMS

The provided use cases are *probably* NP-complete or NP-hard, which means the following statements apply:

- It is easy to verify a specific solution to a problem in reasonable time.
- There is no simple way to find the optimal solution of a problem in reasonable time.

The implication is that solving your problem is probably harder than you anticipated, because the two common techniques do not suffice:

- A brute force algorithm (even a more advanced variant) takes too long.
- A quick algorithm, for example in the bin packing problem, putting in the largest items first returns a solution that is far from optimal.

By using advanced optimization algorithms, OptaPlanner finds a good solution in reasonable time for such planning problems.

3.4. SOLUTIONS TO PLANNING PROBLEMS

A planning problem has a number of solutions.

Several categories of solutions are:

Possible solution

A possible solution is any solution, whether or not it breaks any number of constraints. Planning problems often have an incredibly large number of possible solutions. Many of those solutions are not useful.

Feasible solution

A feasible solution is a solution that does not break any (negative) hard constraints. The number of feasible solutions are relative to the number of possible solutions. Sometimes there are no feasible solutions. Every feasible solution is a possible solution.

Optimal solution

Optimal solutions are the solutions with the highest scores. Planning problems usually have a few optimal solutions. They always have at least one optimal solution, even in the case that there are no feasible solutions and the optimal solution is not feasible.

Best solution found

The best solution is the solution with the highest score found by an implementation in a specified amount of time. The best solution found is likely to be feasible and, given enough time, it's an optimal solution.

Counterintuitively, the number of possible solutions is huge (if calculated correctly), even with a small data set.

In the examples provided in the **optaplanner-examples/src** distribution folder, most instances have a large number of possible solutions. As there is no guaranteed way to find the optimal solution, any implementation is forced to evaluate at least a subset of all those possible solutions.

OptaPlanner supports several optimization algorithms to efficiently wade through that incredibly large number of possible solutions.

Depending on the use case, some optimization algorithms perform better than others, but it is impossible to know in advance. Using OptaPlanner, you can switch the optimization algorithm by changing the solver configuration in a few lines of XML or code.

3.5. CONSTRAINTS ON PLANNING PROBLEMS

Usually, a planning problem has minimum two levels of constraints:

- A (*negative*) *hard constraint* must not be broken. For example, one teacher can not teach two different lessons at the same time.
- A (*negative*) soft constraint should not be broken if it can be avoided. For example, Teacher A does not like to teach on Friday afternoons.

Some problems also have positive constraints:

• A *positive soft constraint (or reward)* should be fulfilled if possible. For example, Teacher B likes to teach on Monday mornings.

Some basic problems only have hard constraints. Some problems have three or more levels of constraints, for example, hard, medium, and soft constraints.

These constraints define the *score calculation* (otherwise known as the *fitness function*) of a planning problem. Each solution of a planning problem is graded with a score. With OptaPlanner, score constraints are written in an object oriented language such as Java, or in Drools rules.

This type of code is flexible and scalable.

3.6. EXAMPLES PROVIDED WITH RED HAT BUILD OF OPTAPLANNER

Several OptaPlanner examples are shipped with Red Hat Build of OptaPlanner. You can review the code for examples and modify it as necessary to suit your needs.



NOTE

Red Hat does not provide support for the example code included in the Red Hat Build of OptaPlanner distribution.

Some of the OptaPlanner examples solve problems that are presented in academic contests. The **Contest** column in the following table lists the contests. It also identifies an example as being either *realistic* or *unrealistic* for the purpose of a contest. A *realistic contest* is an official, independent contest that meets the following standards:

- Clearly defined real-world use cases
- Real-world constraints
- Multiple real-world datasets
- Reproducible results within a specific time limit on specific hardware
- Serious participation from the academic and/or enterprise Operations Research community.

Realistic contests provide an objective comparison of OptaPlanner with competitive software and academic research.

Table 3.1. Examples overview

Example	Domain	Size	Contest	Directory name
N queens	1 entity class (1 variable)	Entity ← 256 Value ← 256 Search space ← 10^616	Pointless (cheatable)	nqueens
Cloud balancing	1 entity class (1 variable)	Entity ← 2400 Value ← 800 Search space ← 10^6967	No (Defined by us)	cloudbalancing
Traveling salesman	1 entity class (1 chained variable)	Entity ← 980 Value ← 980 Search space ← 10^2504	Unrealistic TSP web	tsp
Tennis club scheduling	1 entity class (1 variable)	Entity ← 72 Value ← 7 Search space ← 10^60	No (Defined by us)	tennis

Example	Domain	Size	Contest	Directory name
Meeting scheduling	1 entity class (2 variables)	Entity ← 10 Value ← 320 and ← 5 Search space ← 10^320	No (Defined by us)	meetingscheduling
Course timetablin g	1 entity class (2 variables)	Entity ← 434 Value ← 25 and ← 20 Search space ← 10^1171	Realistic ITC 2007 track 3	curriculumCourse
Machine reassignm ent	1 entity class (1 variable)	Entity ← 50000 Value ← 5000 Search space ← 10^184948	Nearly realistic ROADEF 2012	machineReassignm ent
Vehicle routing	1 entity class (1 chained variable) 1 shadow entity class (1 automatic shadow variable)	Entity ← 55 Value ← 2750 Search space ← 10^8380	Unrealistic VRP web	vehiclerouting
Vehicle routing with time windows	All of Vehicle routing (1 shadow variable)	Entity ← 55 Value ← 2750 Search space ← 10^8380	Unrealistic VRP web	vehiclerouting
Project job scheduling	1 entity class (2 variables) (1 shadow variable)	Entity ← 640 Value ← ? and ← ? Search space ← ?	Nearly realistic MISTA 2013	projectjobschedulin g
Task assigning	1 entity class (1 list variable) 1 shadow entity class (1 automatic shadow variable) (1 shadow variable)	Entity ← 20 Value ← 500 Search space ← 10^1168	No Defined by us	taskassigning

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Example	Domain	Size	Contest	Directory name
Exam timetablin g	2 entity classes (same hierarchy) (2 variables)	Entity ← 1096 Value ← 80 and ← 49 Search space ← 10^3374	Realistic ITC 2007 track 1	examination
Nurse rostering	1 entity class (1 variable)	Entity ← 752 Value ← 50 Search space ← 10^1277	Realistic INRC 2010	nurserostering
Traveling tournamen t	1 entity class (1 variable)	Entity ← 1560 Value ← 78 Search space ← 10^2301	Unrealistic TTP	travelingtournament
Conferenc e scheduling	1 entity class (2 variables)	Entity ← 216 Value ← 18 and ← 20 Search space ← 10^552	No Defined by us	conferencescheduli ng
Flight crew scheduling	1 entity class (1 variable) 1 shadow entity class (1 automatic shadow variable)	Entity ← 4375 Value ← 750 Search space ← 10^12578	No Defined by us	flightcrewschedulin g

3.7. N QUEENS

Place *n* number of queens on an *n* sized chessboard so that no two queens can attack each other. The most common *n* queens puzzle is the eight queens puzzle, with n = 8:

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						Ŵ	

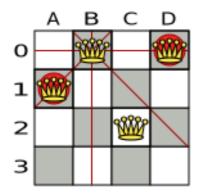
Constraints:

- Use a chessboard of *n* columns and *n* rows.
- Place *n* queens on the chessboard.
- No two queens can attack each other. A queen can attack any other queen on the same horizontal, vertical, or diagonal line.

This documentation heavily uses the four queens puzzle as the primary example.

A proposed solution could be:

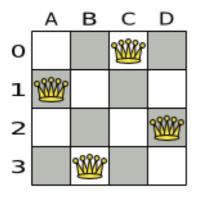
Figure 3.1. A wrong solution for the four queens puzzle



The above solution is wrong because queens **A1** and **B0** can attack each other (so can queens **B0** and **D0**). Removing queen **B0** would respect the "no two queens can attack each other" constraint, but would break the "place *n* queens" constraint.

Below is a correct solution:





All the constraints have been met, so the solution is correct.

Note that most n queens puzzles have multiple correct solutions. We will focus on finding a single correct solution for a specific n, not on finding the number of possible correct solutions for a specific n.

Problem size

4queens has	4 queens with a search space of 256.
8queens has	8 queens with a search space of 10^7.
16queens has	16 queens with a search space of 10^19.
32queens has	32 queens with a search space of 10^48.
64queens has	64 queens with a search space of 10^115.
256queens has	256 queens with a search space of 10 ⁶¹⁶ .

The implementation of the *n* queens example has not been optimized because it functions as a beginner example. Nevertheless, it can easily handle 64 queens. With a few changes it has been shown to easily handle 5000 queens and more.

3.7.1. Domain model for N queens

This example uses the domain model to solve the four queens problem.

• Creating a Domain Model

A good domain model will make it easier to understand and solve your planning problem.

This is the domain model for the *n* queens example:

public class Column {
 private int index;
 // ... getters and setters
}

public class Row {

<pre>private int index; // getters and setters }</pre>
public class Queen {
private Column column; private Row row;
<pre>public int getAscendingDiagonalIndex() {} public int getDescendingDiagonalIndex() {}</pre>
// getters and setters }

• Calculating the Search Space.

A **Queen** instance has a **Column** (for example: 0 is column A, 1 is column B, ...) and a **Row** (its row, for example: 0 is row 0, 1 is row 1, ...).

The ascending diagonal line and the descending diagonal line can be calculated based on the column and the row.

The column and row indexes start from the upper left corner of the chessboard.

public class NQueens {
 private int n;
 private List<Column> columnList;
 private List<Row> rowList;
 private List<Queen> queenList;
 private SimpleScore score;
 // ... getters and setters

• Finding the Solution

}

A single **NQueens** instance contains a list of all **Queen** instances. It is the **Solution** implementation which will be supplied to, solved by, and retrieved from the Solver.

Notice that in the four queens example, the NQueens getN() method will always return four.

Figure 3.3. A solution for Four Queens

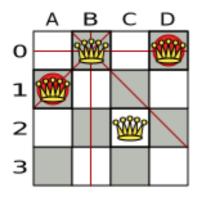


Table 3.2. Details of the solution in the domain model

	columnIndex	rowIndex	ascendingDiagonalIndex (columnIndex + rowIndex)	descendingDiagonalIndex (columnIndex - rowIndex)
A1	0	1	1(**)	-1
BO	1	0 (*)	1 (**)	1
C2	2	2	4	0
DO	3	0 (*)	3	3

When two queens share the same column, row or diagonal line, such as (*) and (**), they can attack each other.

3.8. CLOUD BALANCING

For information about this example, see Red Hat Build of OptaPlanner quick start guides .

3.9. TRAVELING SALESMAN (TSP - TRAVELING SALESMAN PROBLEM)

Given a list of cities, find the shortest tour for a salesman that visits each city exactly once.

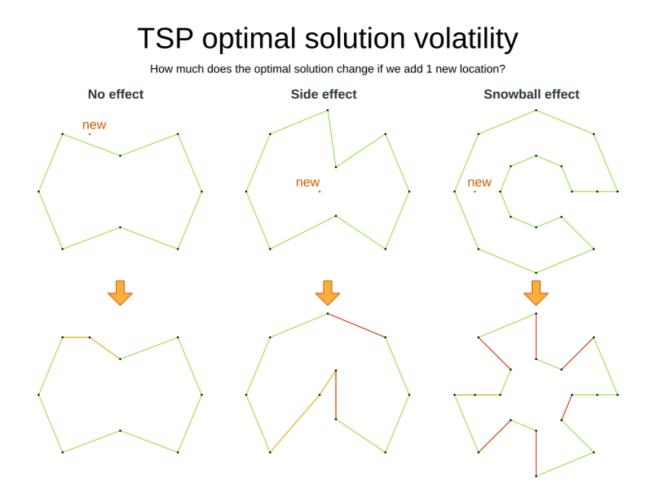
The problem is defined by Wikipedia. It is one of the most intensively studied problems in computational mathematics. Yet, in the real world, it is often only part of a planning problem, along with other constraints, such as employee shift rostering constraints.

Problem size

dj38 has 38 cities with a search space of 10^43. europe40 has 40 cities with a search space of 10^46. st70 has 70 cities with a search space of 10^98. pcb442 has 442 cities with a search space of 10^976. lu980 has 980 cities with a search space of 10^2504.

Problem difficulty

Despite TSP's simple definition, the problem is surprisingly hard to solve. Because it is an NP-hard problem (like most planning problems), the optimal solution for a specific problem dataset can change a lot when that problem dataset is slightly altered:



3.10. TENNIS CLUB SCHEDULING

Every week the tennis club has four teams playing round robin against each other. Assign those four spots to the teams fairly.

Hard constraints:

- Conflict: A team can only play once per day.
- Unavailability: Some teams are unavailable on some dates.

Medium constraints:

• Fair assignment: All teams should play an (almost) equal number of times.

Soft constraints:

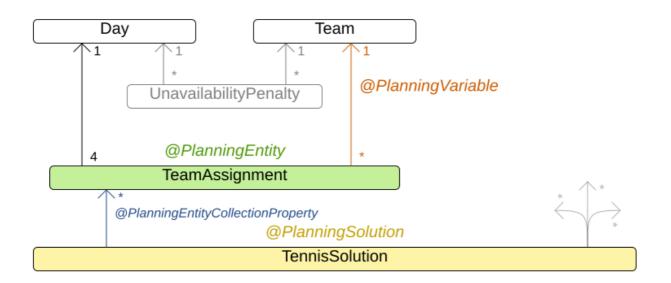
• Evenly confrontation: Each team should play against every other team an equal number of times.

Problem size

munich-7teams has 7 teams, 18 days, 12 unavailabilityPenalties and 72 teamAssignments with a search space of 10⁶⁰.

Figure 3.4. Domain model

Tennis class diagram



3.11. MEETING SCHEDULING

Assign each meeting to a starting time and a room. Meetings have different durations.

Hard constraints:

- Room conflict: Two meetings must not use the same room at the same time.
- Required attendance: A person cannot have two required meetings at the same time.
- Required room capacity: A meeting must not be in a room that doesn't fit all of the meeting's attendees.
- Start and end on same day: A meeting shouldn't be scheduled over multiple days.

Medium constraints:

• Preferred attendance: A person cannot have two preferred meetings at the same time, nor a preferred and a required meeting at the same time.

Soft constraints:

• Sooner rather than later: Schedule all meetings as soon as possible.

- A break between meetings: Any two meetings should have at least one time grain break between them.
- Overlapping meetings: To minimize the number of meetings in parallel so people don't have to choose one meeting over the other.
- Assign larger rooms first: If a larger room is available any meeting should be assigned to that room in order to accommodate as many people as possible even if they haven't signed up to that meeting.
- Room stability: If a person has two consecutive meetings with two or less time grains break between them they better be in the same room.

Problem size

50meetings-160timegrains-5rooms has 50 meetings, 160 timeGrains and 5 rooms with a search space of 10^145.

100meetings-320timegrains-5rooms has 100 meetings, 320 timeGrains and 5 rooms with a search space of 10^320.

200meetings-640timegrains-5rooms has 200 meetings, 640 timeGrains and 5 rooms with a search space of 10^701.

400meetings-1280timegrains-5rooms has 400 meetings, 1280 timeGrains and 5 rooms with a search space of 10^1522.

800meetings-2560timegrains-5rooms has 800 meetings, 2560 timeGrains and 5 rooms with a search space of 10^3285.

3.12. COURSE TIMETABLING (ITC 2007 TRACK 3 - CURRICULUM COURSE SCHEDULING)

Schedule each lecture into a timeslot and into a room.

Hard constraints:

- Teacher conflict: A teacher must not have two lectures in the same period.
- Curriculum conflict: A curriculum must not have two lectures in the same period.
- Room occupancy: Two lectures must not be in the same room in the same period.
- Unavailable period (specified per dataset): A specific lecture must not be assigned to a specific period.

Soft constraints:

- Room capacity: A room's capacity should not be less than the number of students in its lecture.
- Minimum working days: Lectures of the same course should be spread out into a minimum number of days.
- Curriculum compactness: Lectures belonging to the same curriculum should be adjacent to each other (so in consecutive periods).
- Room stability: Lectures of the same course should be assigned to the same room.

The problem is defined by the International Timetabling Competition 2007 track 3.

Problem size

comp01 has 24 teachers, 14 curricula, 30 courses, 160 lectures, 30 periods, 6 rooms and 53 unavailable period constraints with a search space of 10^360.

comp02 has 71 teachers, 70 curricula, 82 courses, 283 lectures, 25 periods, 16 rooms and 513 unavailable period constraints with a search space of 10^736.

comp03 has 61 teachers, 68 curricula, 72 courses, 251 lectures, 25 periods, 16 rooms and 382 unavailable period constraints with a search space of 10^653.

comp04 has 70 teachers, 57 curricula, 79 courses, 286 lectures, 25 periods, 18 rooms and 396 unavailable period constraints with a search space of 10^758.

comp05 has 47 teachers, 139 curricula, 54 courses, 152 lectures, 36 periods, 9 rooms and 771 unavailable period constraints with a search space of 10^381.

comp06 has 87 teachers, 70 curricula, 108 courses, 361 lectures, 25 periods, 18 rooms and 632 unavailable period constraints with a search space of 10^957.

comp07 has 99 teachers, 77 curricula, 131 courses, 434 lectures, 25 periods, 20 rooms and 667 unavailable period constraints with a search space of 10^1171.

comp08 has 76 teachers, 61 curricula, 86 courses, 324 lectures, 25 periods, 18 rooms and 478 unavailable period constraints with a search space of 10^859.

comp09 has 68 teachers, 75 curricula, 76 courses, 279 lectures, 25 periods, 18 rooms and 405 unavailable period constraints with a search space of 10⁷40.

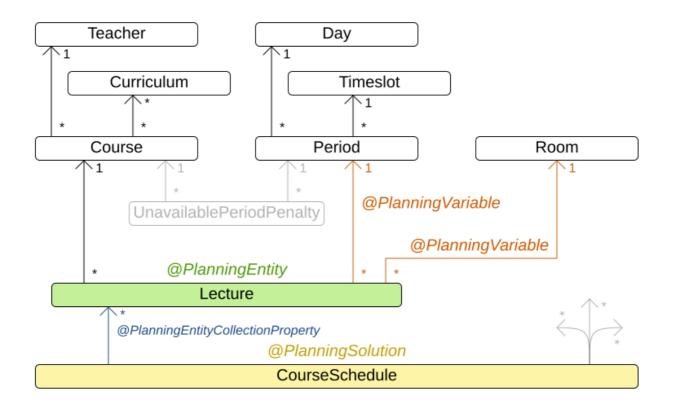
comp10 has 88 teachers, 67 curricula, 115 courses, 370 lectures, 25 periods, 18 rooms and 694 unavailable period constraints with a search space of 10^981.

comp11 has 24 teachers, 13 curricula, 30 courses, 162 lectures, 45 periods, 5 rooms and 94 unavailable period constraints with a search space of 10^381.

comp12 has 74 teachers, 150 curricula, 88 courses, 218 lectures, 36 periods, 11 rooms and 1368 unavailable period constraints with a search space of 10^566.

comp13 has 77 teachers, 66 curricula, 82 courses, 308 lectures, 25 periods, 19 rooms and 468 unavailable period constraints with a search space of 10⁸²⁴.

comp14 has 68 teachers, 60 curricula, 85 courses, 275 lectures, 25 periods, 17 rooms and 486 unavailable period constraints with a search space of 10^722.



Curriculum course class diagram

3.13. MACHINE REASSIGNMENT (GOOGLE ROADEF 2012)

Assign each process to a machine. All processes already have an original (unoptimized) assignment. Each process requires an amount of each resource (such as CPU or RAM). This is a more complex version of the Cloud Balancing example.

Hard constraints:

- Maximum capacity: The maximum capacity for each resource for each machine must not be exceeded.
- Conflict: Processes of the same service must run on distinct machines.
- Spread: Processes of the same service must be spread out across locations.
- Dependency: The processes of a service depending on another service must run in the neighborhood of a process of the other service.
- Transient usage: Some resources are transient and count towards the maximum capacity of both the original machine as the newly assigned machine.

Soft constraints:

• Load: The safety capacity for each resource for each machine should not be exceeded.

- Balance: Leave room for future assignments by balancing the available resources on each machine.
- Process move cost: A process has a move cost.
- Service move cost: A service has a move cost.
- Machine move cost: Moving a process from machine A to machine B has another A-B specific move cost.

The problem is defined by the Google ROADEF/EURO Challenge 2012.

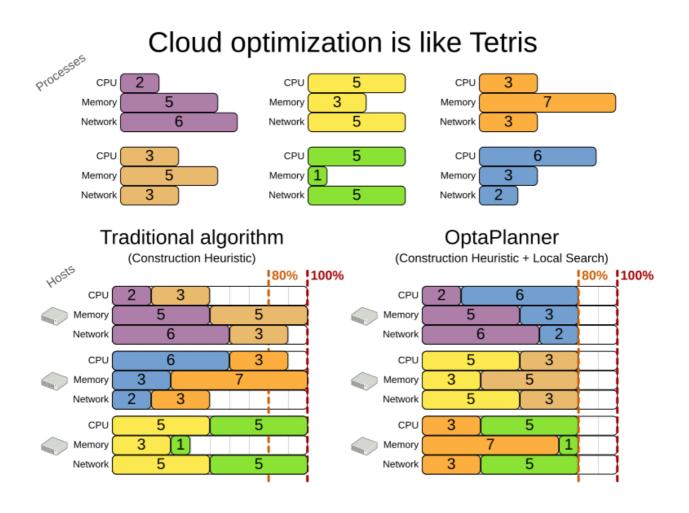
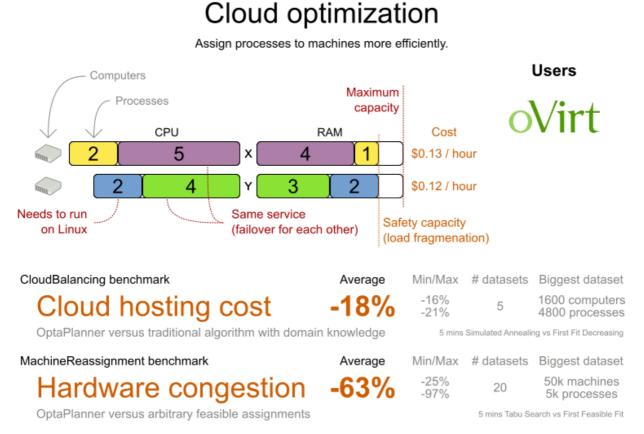


Figure 3.6. Value proposition



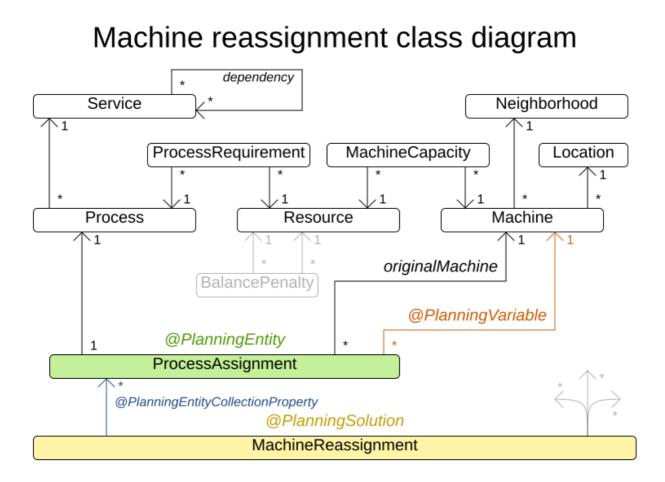
Don't believe us? Run our open benchmarks yourself: http://www.optaplanner.org/code/benchmarks.html

Problem size

model a1 1 has 2 resources, 1 neighborhoods, 4 locations, 4 machines, 79 services, 100 processes and 1 balancePenalties with a search space of 10^60. model a1 2 has 4 resources, 2 neighborhoods, 4 locations, 100 machines, 980 services, 1000 processes and 0 balancePenalties with a search space of 10^2000. model a1 3 has 3 resources, 5 neighborhoods, 25 locations, 100 machines, 216 services, 1000 processes and 0 balancePenalties with a search space of 10^2000. model_a1_4 has 3 resources, 50 neighborhoods, 50 locations, 50 machines, 142 services, 1000 processes and 1 balancePenalties with a search space of 10^1698. model a1 5 has 4 resources, 2 neighborhoods, 4 locations, 12 machines, 981 services, 1000 processes and 1 balancePenalties with a search space of 10^1079. model a2 1 has 3 resources, 1 neighborhoods, 1 locations, 100 machines, 1000 services, 1000 processes and 0 balancePenalties with a search space of 10^2000. model a2 2 has 12 resources, 5 neighborhoods, 25 locations, 100 machines, 170 services, 1000 processes and 0 balancePenalties with a search space of 10^2000. model a2 3 has 12 resources, 5 neighborhoods, 25 locations, 100 machines, 129 services, 1000 processes and 0 balancePenalties with a search space of 10^2000. model_a2_4 has 12 resources, 5 neighborhoods, 25 locations, 50 machines, 180 services, 1000 processes and 1 balancePenalties with a search space of 10^1698. model_a2_5 has 12 resources, 5 neighborhoods, 25 locations, 50 machines, 153 services, 1000 processes and 0 balancePenalties with a search space of 10^1698. model_b_1 has 12 resources, 5 neighborhoods, 10 locations, 100 machines, 2512 services, 5000 processes and 0 balancePenalties with a search space of 10^10000. model_b_2 has 12 resources, 5 neighborhoods, 10 locations, 100 machines, 2462 services, 5000 processes and 1 balancePenalties with a search space of 10^10000.

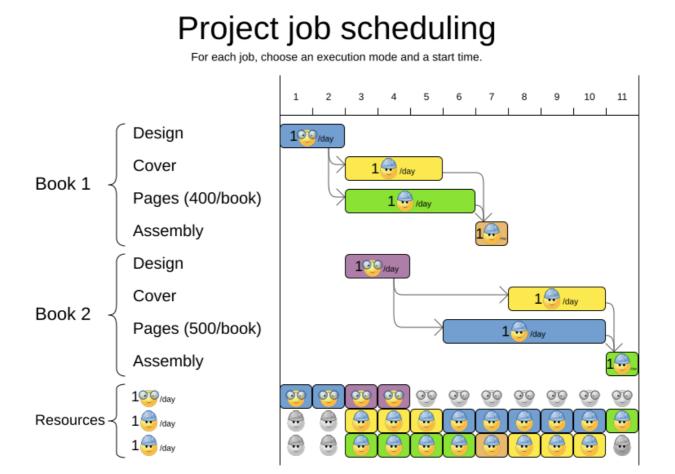
model_b_3 has 6 resources, 5 neighborhoods, 10 locations, 100 machines, 15025 services, 20000 processes and 0 balancePenalties with a search space of 10^40000. model_b_4 has 6 resources, 5 neighborhoods, 50 locations, 500 machines, 1732 services, 20000 processes and 1 balancePenalties with a search space of 10^53979. model_b_5 has 6 resources, 5 neighborhoods, 10 locations, 100 machines, 35082 services, 40000 processes and 0 balancePenalties with a search space of 10^80000. model b 6 has 6 resources, 5 neighborhoods, 50 locations, 200 machines, 14680 services, 40000 processes and 1 balancePenalties with a search space of 10^92041. model b 7 has 6 resources, 5 neighborhoods, 50 locations, 4000 machines, 15050 services, 40000 processes and 1 balancePenalties with a search space of 10¹⁴⁴⁰⁸². model_b_8 has 3 resources, 5 neighborhoods, 10 locations, 100 machines, 45030 services, 50000 processes and 0 balancePenalties with a search space of 10^100000. model_b_9 has 3 resources, 5 neighborhoods, 100 locations, 1000 machines, 4609 services, 50000 processes and 1 balancePenalties with a search space of 10^150000. model_b_10 has 3 resources, 5 neighborhoods, 100 locations, 5000 machines, 4896 services, 50000 processes and 1 balancePenalties with a search space of 10¹⁸⁴⁹⁴⁸.

Figure 3.7. Domain model



3.14. PROJECT JOB SCHEDULING

Schedule all jobs in time and execution mode to minimize project delays. Each job is part of a project. A job can be executed in different ways: each way is an execution mode that implies a different duration but also different resource usages. This is a form of flexible *job shop scheduling*.



Hard constraints:

- Job precedence: a job can only start when all its predecessor jobs are finished.
- Resource capacity: do not use more resources than available.
 - Resources are local (shared between jobs of the same project) or global (shared between all jobs)
 - Resources are renewable (capacity available per day) or nonrenewable (capacity available for all days)

Medium constraints:

• Total project delay: minimize the duration (makespan) of each project.

Soft constraints:

• Total makespan: minimize the duration of the whole multi-project schedule.

The problem is defined by the MISTA 2013 challenge.

Problem size

```
Schedule A-1 has 2 projects, 24 jobs, 64 execution modes, 7 resources and 150 resource requirements.
Schedule A-2 has 2 projects, 44 jobs, 124 execution modes, 7 resources and 420 resource requirements.
```

Schedule A-3 has 2 projects, 64 jobs, 184 execution modes, 7 resources and 630 resource requirements. Schedule A-4 has 5 projects, 60 jobs, 160 execution modes, 16 resources and 390 resource requirements. Schedule A-5 has 5 projects, 110 jobs, 310 execution modes, 16 resources and 900 resource requirements. Schedule A-6 has 5 projects, 160 jobs, 460 execution modes, 16 resources and 1440 resource requirements. Schedule A-7 has 10 projects, 120 jobs, 320 execution modes, 22 resources and 900 resource requirements. Schedule A-8 has 10 projects, 220 jobs, 620 execution modes, 22 resources and 1860 resource requirements. Schedule A-9 has 10 projects, 320 jobs, 920 execution modes, 31 resources and 2880 resource requirements. Schedule A-10 has 10 projects, 320 jobs, 920 execution modes, 31 resources and 2970 resource requirements. Schedule B-1 has 10 projects, 120 jobs, 320 execution modes, 31 resources and 900 resource requirements. Schedule B-2 has 10 projects, 220 jobs, 620 execution modes, 22 resources and 1740 resource requirements. Schedule B-3 has 10 projects, 320 jobs, 920 execution modes, 31 resources and 3060 resource requirements. Schedule B-4 has 15 projects, 180 jobs, 480 execution modes, 46 resources and 1530 resource requirements. Schedule B-5 has 15 projects, 330 jobs, 930 execution modes, 46 resources and 2760 resource requirements. Schedule B-6 has 15 projects, 480 jobs, 1380 execution modes, 46 resources and 4500 resource requirements. Schedule B-7 has 20 projects, 240 jobs, 640 execution modes, 61 resources and 1710 resource requirements. Schedule B-8 has 20 projects, 440 jobs, 1240 execution modes, 42 resources and 3180 resource requirements. Schedule B-9 has 20 projects, 640 jobs, 1840 execution modes, 61 resources and 5940 resource requirements. Schedule B-10 has 20 projects, 460 jobs, 1300 execution modes, 42 resources and 4260 resource requirements.

3.15. TASK ASSIGNING

Assign each task to a spot in an employee's queue. Each task has a duration which is affected by the employee's affinity level with the task's customer.

Hard constraints:

• Skill: Each task requires one or more skills. The employee must possess all these skills.

Soft level 0 constraints:

• Critical tasks: Complete critical tasks first, sooner than major and minor tasks.

Soft level 1 constraints:

- Minimize makespan: Reduce the time to complete all tasks.
 - Start with the longest working employee first, then the second longest working employee and so forth, to create fairness and load balancing.

Soft level 2 constraints:

• Major tasks: Complete major tasks as soon as possible, sooner than minor tasks.

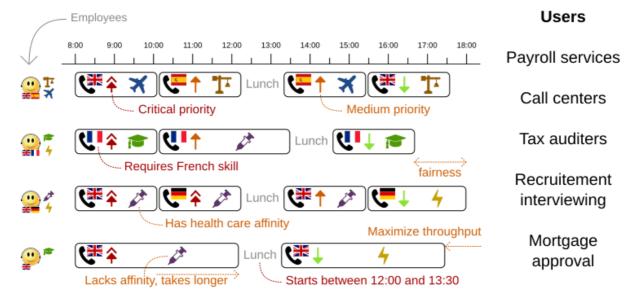
Soft level 3 constraints:

• Minor tasks: Complete minor tasks as soon as possible.

Figure 3.8. Value proposition

Task assigning

Optimize the task queue of every employee by reassigning and reordering tasks.



Problem size

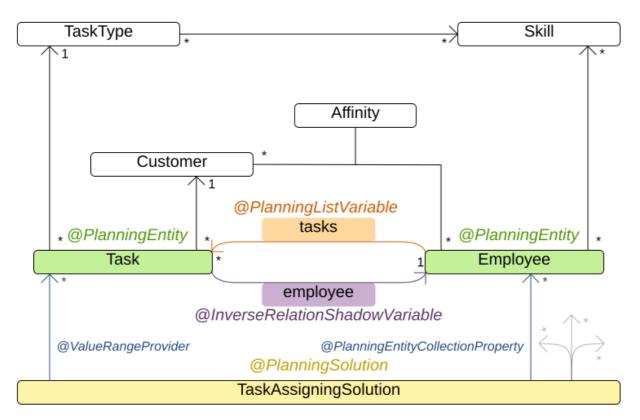
24tasks-8employees has 24 tasks, 6 skills, 8 employees, 4 task types and 4 customers with a search space of 10^30.

50tasks-5employees has 50 tasks, 5 skills, 5 employees, 10 task types and 10 customers with a search space of 10^69.

100tasks-5employees has 100 tasks, 5 skills, 5 employees, 20 task types and 15 customers with a search space of 10¹⁶⁴.

500tasks-20employees has 500 tasks, 6 skills, 20 employees, 100 task types and 60 customers with a search space of 10^1168.

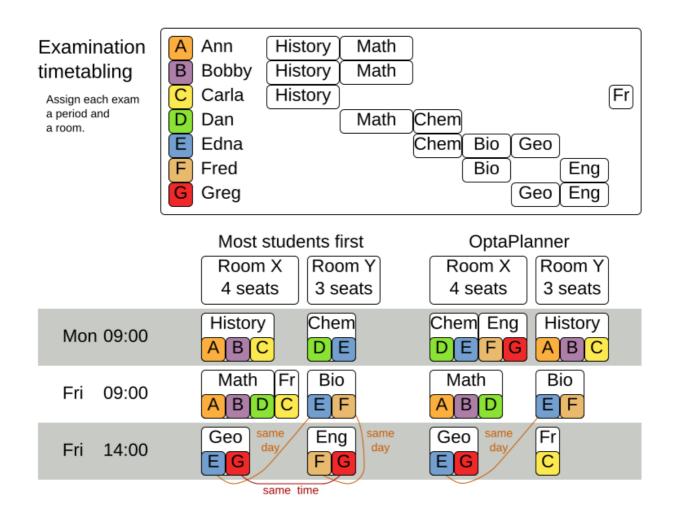
Figure 3.9. Domain model



Task assigning class diagram

3.16. EXAM TIMETABLING (ITC 2007 TRACK 1 - EXAMINATION)

Schedule each exam into a period and into a room. Multiple exams can share the same room during the same period.



Hard constraints:

- Exam conflict: Two exams that share students must not occur in the same period.
- Room capacity: A room's seating capacity must suffice at all times.
- Period duration: A period's duration must suffice for all of its exams.
- Period related hard constraints (specified per dataset):
 - Coincidence: Two specified exams must use the same period (but possibly another room).
 - Exclusion: Two specified exams must not use the same period.
 - After: A specified exam must occur in a period after another specified exam's period.
- Room related hard constraints (specified per dataset):
 - Exclusive: One specified exam should not have to share its room with any other exam.

Soft constraints (each of which has a parametrized penalty):

- The same student should not have two exams in a row.
- The same student should not have two exams on the same day.
- Period spread: Two exams that share students should be a number of periods apart.
- Mixed durations: Two exams that share a room should not have different durations.

- Front load: Large exams should be scheduled earlier in the schedule.
- Period penalty (specified per dataset): Some periods have a penalty when used.
- Room penalty (specified per dataset): Some rooms have a penalty when used.

It uses large test data sets of real-life universities.

The problem is defined by the International Timetabling Competition 2007 track 1. Geoffrey De Smet finished 4th in that competition with a very early version of OptaPlanner. Many improvements have been made since then.

Problem Size

exam_comp_set1 has 7883 students, 607 exams, 54 periods, 7 rooms, 12 period constraints and 0 room constraints with a search space of 10^1564.

exam_comp_set2 has 12484 students, 870 exams, 40 periods, 49 rooms, 12 period constraints and 2 room constraints with a search space of 10^2864.

exam_comp_set3 has 16365 students, 934 exams, 36 periods, 48 rooms, 168 period constraints and 15 room constraints with a search space of 10^3023.

exam_comp_set4 has 4421 students, 273 exams, 21 periods, 1 rooms, 40 period constraints and 0 room constraints with a search space of 10^360.

exam_comp_set5 has 8719 students, 1018 exams, 42 periods, 3 rooms, 27 period constraints and 0 room constraints with a search space of 10^2138.

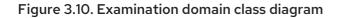
exam_comp_set6 has 7909 students, 242 exams, 16 periods, 8 rooms, 22 period constraints and 0 room constraints with a search space of 10^509.

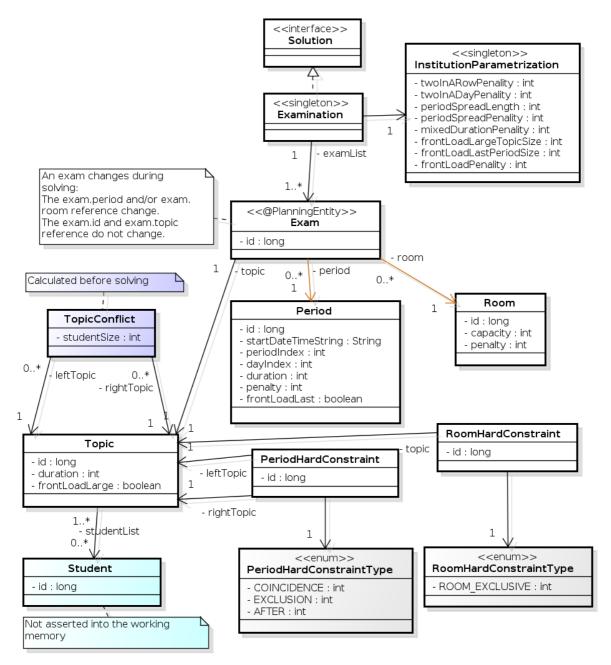
exam_comp_set7 has 13795 students, 1096 exams, 80 periods, 15 rooms, 28 period constraints and 0 room constraints with a search space of 10^3374.

exam_comp_set8 has 7718 students, 598 exams, 80 periods, 8 rooms, 20 period constraints and 1 room constraints with a search space of 10^1678.

3.16.1. Domain model for exam timetabling

The following diagram shows the main examination domain classes:





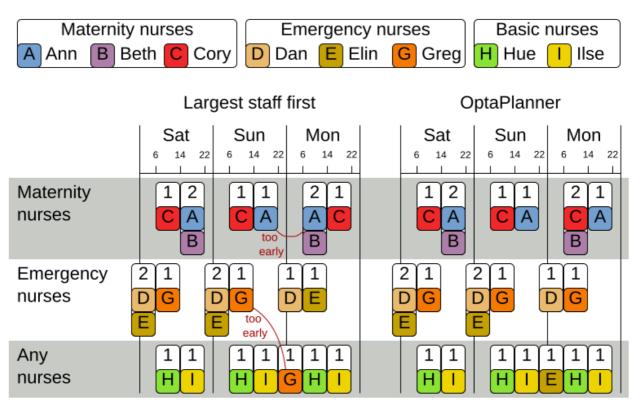
Notice that we've split up the exam concept into an **Exam** class and a **Topic** class. The **Exam** instances change during solving (this is the planning entity class), when their period or room property changes. The **Topic**, **Period** and **Room** instances never change during solving (these are problem facts, just like some other classes).

3.17. NURSE ROSTERING (INRC 2010)

For each shift, assign a nurse to work that shift.



Populate each work shift with a nurse.



Hard constraints:

- No unassigned shifts (built-in): Every shift need to be assigned to an employee.
- Shift conflict: An employee can have only one shift per day.

Soft constraints:

- Contract obligations. The business frequently violates these, so they decided to define these as soft constraints instead of hard constraints.
 - **Minimum and maximum assignments** Each employee needs to work more than x shifts and less than y shifts (depending on their contract).
 - **Minimum and maximum consecutive working days** Each employee needs to work between x and y days in a row (depending on their contract).
 - Minimum and maximum consecutive free days Each employee needs to be free between x and y days in a row (depending on their contract).
 - Minimum and maximum consecutive working weekends Each employee needs to work between x and y weekends in a row (depending on their contract).
 - Complete weekends: Each employee needs to work every day in a weekend or not at all.
 - Identical shift types during weekend Each weekend shift for the same weekend of the same employee must be the same shift type.

-

- **Unwanted patterns:** A combination of unwanted shift types in a row, for example a late shift followed by an early shift followed by a late shift.
- Employee wishes:
 - Day on request: An employee wants to work on a specific day.
 - Day off request: An employee does not want to work on a specific day.
 - Shift on request An employee wants to be assigned to a specific shift.
 - Shift off request: An employee does not want to be assigned to a specific shift.
- Alternative skill: An employee assigned to a skill should have a proficiency in every skill required by that shift.

The problem is defined by the International Nurse Rostering Competition 2010 .

Figure 3.11. Value proposition

Employee rostering Assign shifts to employees more efficiently. Employees Users Thu Fri Sat Tue Wed Sun Mon 14 22 6 14 22 14 22 6 14 22 14 22 14 14 22 Hospitals ≥ 10 hours ÷ Free Free à 📥 Security ≤ 5 consecutive shifts Forward rotation 1 shift per day (enough time to sleep) guard firms * Free Free Requires nurse skill Day off request Call centers Free Free -Police and Requires engineering skill ≥ 48 hours rest fire department Free Free No weekend work Shifts NurseRostering benchmark Average Min/Max # datasets Biggest dataset +19% 752 assignments Employee well-being 26 +85% 50 employees OptaPlanner versus traditional algorithm with domain knowledge 5 mins Tabu Search vs First Fit Decreasing

Don't believe us? Run our open benchmarks yourself: https://www.optaplanner.org/code/benchmarks.html

Problem size

There are three dataset types:

- Sprint: must be solved in seconds.
- Medium: must be solved in minutes.
- Long: must be solved in hours.

toy1 has 1 skills, 3 shiftTypes, 2 patterns, 1 contracts, 6 employees, 7 shiftDates, 35 shiftAssignments and 0 requests with a search space of 10^27.

toy2 has 1 skills, 3 shiftTypes, 3 patterns, 2 contracts, 20 employees, 28 shiftDates, 180 shiftAssignments and 140 requests with a search space of 10^234.

sprint01 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint02 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint03 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint04 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint05 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint06 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint07 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint08 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint09 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint10 has 1 skills, 4 shiftTypes, 3 patterns, 4 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_hint01 has 1 skills, 4 shiftTypes, 8 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_hint02 has 1 skills, 4 shiftTypes, 0 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_hint03 has 1 skills, 4 shiftTypes, 8 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_late01 has 1 skills, 4 shiftTypes, 8 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_late02 has 1 skills, 3 shiftTypes, 4 patterns, 3 contracts, 10 employees, 28 shiftDates, 144 shiftAssignments and 139 requests with a search space of 10^144.

sprint_late03 has 1 skills, 4 shiftTypes, 8 patterns, 3 contracts, 10 employees, 28 shiftDates, 160 shiftAssignments and 150 requests with a search space of 10^160.

sprint_late04 has 1 skills, 4 shiftTypes, 8 patterns, 3 contracts, 10 employees, 28 shiftDates, 160 shiftAssignments and 150 requests with a search space of 10^160.

sprint_late05 has 1 skills, 4 shiftTypes, 8 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_late06 has 1 skills, 4 shiftTypes, 0 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_late07 has 1 skills, 4 shiftTypes, 0 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

sprint_late08 has 1 skills, 4 shiftTypes, 0 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 0 requests with a search space of 10^152.

sprint_late09 has 1 skills, 4 shiftTypes, 0 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 0 requests with a search space of 10^152.

sprint_late10 has 1 skills, 4 shiftTypes, 0 patterns, 3 contracts, 10 employees, 28 shiftDates, 152 shiftAssignments and 150 requests with a search space of 10^152.

medium01 has 1 skills, 4 shiftTypes, 0 patterns, 4 contracts, 31 employees, 28 shiftDates, 608 shiftAssignments and 403 requests with a search space of 10^906.

medium02 has 1 skills, 4 shiftTypes, 0 patterns, 4 contracts, 31 employees, 28 shiftDates, 608 shiftAssignments and 403 requests with a search space of 10^906.

medium03 has 1 skills, 4 shiftTypes, 0 patterns, 4 contracts, 31 employees, 28 shiftDates, 608 shiftAssignments and 403 requests with a search space of 10^906.

medium04 has 1 skills, 4 shiftTypes, 0 patterns, 4 contracts, 31 employees, 28 shiftDates, 608 shiftAssignments and 403 requests with a search space of 10^906.

medium05 has 1 skills, 4 shiftTypes, 0 patterns, 4 contracts, 31 employees, 28 shiftDates, 608 shiftAssignments and 403 requests with a search space of 10^906.

medium_hint01 has 1 skills, 4 shiftTypes, 7 patterns, 4 contracts, 30 employees, 28 shiftDates, 428 shiftAssignments and 390 requests with a search space of 10^632.

medium_hint02 has 1 skills, 4 shiftTypes, 7 patterns, 3 contracts, 30 employees, 28 shiftDates, 428 shiftAssignments and 390 requests with a search space of 10^632.

medium_hint03 has 1 skills, 4 shiftTypes, 7 patterns, 4 contracts, 30 employees, 28 shiftDates, 428 shiftAssignments and 390 requests with a search space of 10^632.

medium_late01 has 1 skills, 4 shiftTypes, 7 patterns, 4 contracts, 30 employees, 28 shiftDates, 424 shiftAssignments and 390 requests with a search space of 10^626.

medium_late02 has 1 skills, 4 shiftTypes, 7 patterns, 3 contracts, 30 employees, 28 shiftDates, 428 shiftAssignments and 390 requests with a search space of 10^632.

medium_late03 has 1 skills, 4 shiftTypes, 0 patterns, 4 contracts, 30 employees, 28 shiftDates, 428 shiftAssignments and 390 requests with a search space of 10^632.

medium_late04 has 1 skills, 4 shiftTypes, 7 patterns, 3 contracts, 30 employees, 28 shiftDates, 416 shiftAssignments and 390 requests with a search space of 10^614.

medium_late05 has 2 skills, 5 shiftTypes, 7 patterns, 4 contracts, 30 employees, 28 shiftDates, 452 shiftAssignments and 390 requests with a search space of 10^667.

long01 has 2 skills, 5 shiftTypes, 3 patterns, 3 contracts, 49 employees, 28 shiftDates, 740 shiftAssignments and 735 requests with a search space of 10^1250.

long02 has 2 skills, 5 shiftTypes, 3 patterns, 3 contracts, 49 employees, 28 shiftDates, 740 shiftAssignments and 735 requests with a search space of 10^1250.

long03 has 2 skills, 5 shiftTypes, 3 patterns, 3 contracts, 49 employees, 28 shiftDates, 740 shiftAssignments and 735 requests with a search space of 10^1250.

long04 has 2 skills, 5 shiftTypes, 3 patterns, 3 contracts, 49 employees, 28 shiftDates, 740 shiftAssignments and 735 requests with a search space of 10^1250.

long05 has 2 skills, 5 shiftTypes, 3 patterns, 3 contracts, 49 employees, 28 shiftDates, 740 shiftAssignments and 735 requests with a search space of 10^1250.

long_hint01 has 2 skills, 5 shiftTypes, 9 patterns, 3 contracts, 50 employees, 28 shiftDates, 740 shiftAssignments and 0 requests with a search space of 10^1257.

long_hint02 has 2 skills, 5 shiftTypes, 7 patterns, 3 contracts, 50 employees, 28 shiftDates, 740 shiftAssignments and 0 requests with a search space of 10^1257.

long_hint03 has 2 skills, 5 shiftTypes, 7 patterns, 3 contracts, 50 employees, 28 shiftDates, 740 shiftAssignments and 0 requests with a search space of 10^1257.

long_late01 has 2 skills, 5 shiftTypes, 9 patterns, 3 contracts, 50 employees, 28 shiftDates, 752 shiftAssignments and 0 requests with a search space of 10^1277.

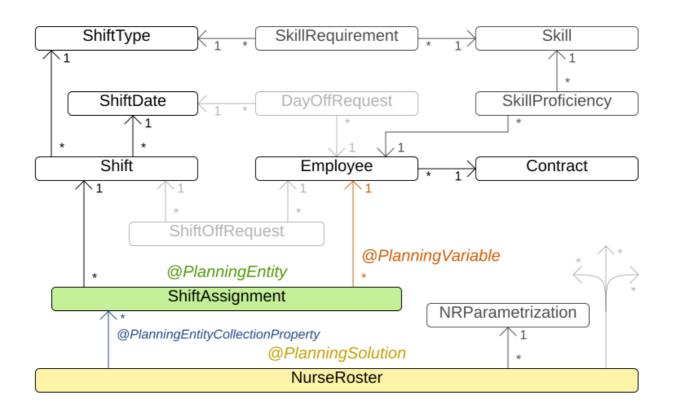
long_late02 has 2 skills, 5 shiftTypes, 9 patterns, 4 contracts, 50 employees, 28 shiftDates, 752 shiftAssignments and 0 requests with a search space of 10^1277.

long_late03 has 2 skills, 5 shiftTypes, 9 patterns, 3 contracts, 50 employees, 28 shiftDates, 752 shiftAssignments and 0 requests with a search space of 10^1277.

long_late04 has 2 skills, 5 shiftTypes, 9 patterns, 4 contracts, 50 employees, 28 shiftDates, 752 shiftAssignments and 0 requests with a search space of 10^1277.

long_late05 has 2 skills, 5 shiftTypes, 9 patterns, 3 contracts, 50 employees, 28 shiftDates, 740 shiftAssignments and 0 requests with a search space of 10^1257.

Figure 3.12. Domain model

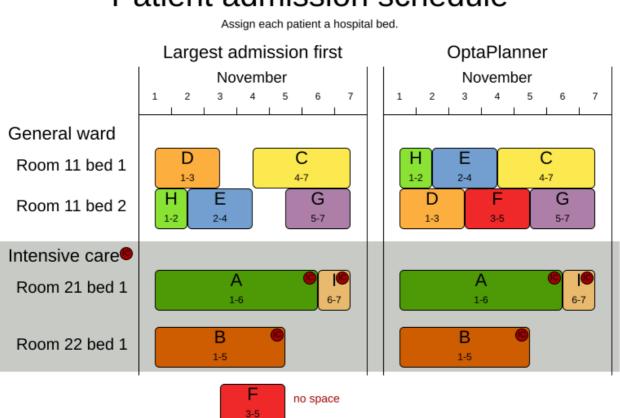


Nurse rostering class diagram

3.18. PATIENT ADMISSION SCHEDULING

Patient admission scheduling (PAS), also known as hospital bed planning, assigns a bed to each patient that is admitted to the hospital. The bed is assigned to the patient for the duration of the patient's scheduled stay. Each bed belongs to a room and each room belongs to a department. The arrival and departure dates of the patients are fixed. You only need to assign a bed.

This problem features overconstrained datasets. When it is not necessary to assign all planning entities, it is preferable to assign as many entities as required without breaking hard constraints. This is called overconstrained planning.



Patient admission schedule

Hard constraints:

- Two patients must not be assigned to the same bed on the same night. Weight: -1000hard * conflictNightCount.
- A room can have a gender limitation: only females, only males, the same gender in the same night or no gender limitation at all. Weight: **-50hard * nightCount**.
- A department can have a minimum or maximum age. Weight: **-100hard * nightCount**.
- A patient can require a room with specific equipment. Weight: **-50hard * nightCount**.

Medium constraints:

• Assign every patient to a bed unless the dataset is overconstrained. Weight: **-1medium** * **nightCount**.

Soft constraints:

- A patient can specify a preference for a maximum room size, for example if the patient wants a single room. Weight: **-8soft * nightCount**.
- A patient is best assigned to a department that specializes in the patient's medical problem. Weight: **-10soft * nightCount**.
- A patient is best assigned to a room that specializes in the patient's mecical problem. Weight: **20soft * nightCount**.
 - The room speciality should be priority 1. Weight: -10soft * (priority 1) * nightCount.

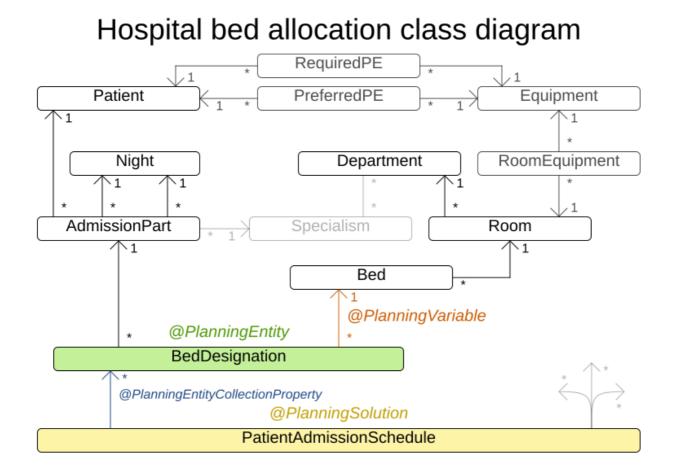
• A patient can specify a preference for a room with specific equipment. Weight: **-20soft** * **nightCount**.

The problem is a variant on Kaho's Patient Scheduling and the datasets come from real world hospitals.

Problem size

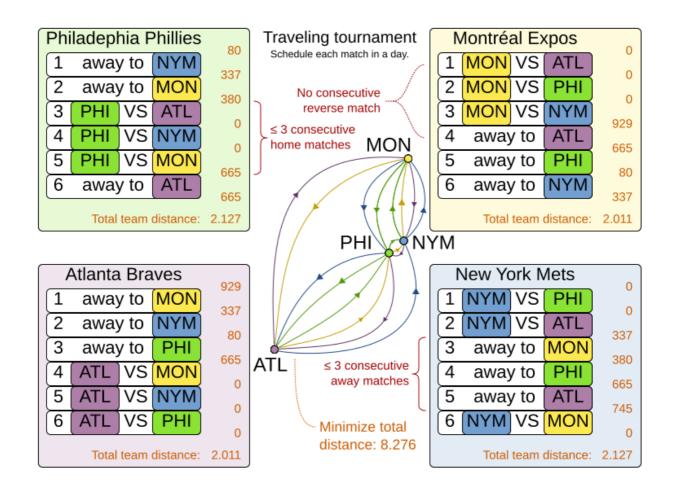
overconstrained01 has 6 specialisms, 4 equipments, 1 departments, 25 rooms, 69 beds, 14 nights, 519 patients and 519 admissions with a search space of 10^958. has 4 specialisms, 2 equipments, 4 departments, 98 rooms, 286 beds, 14 nights, 652 testdata01 patients and 652 admissions with a search space of 10¹⁶⁰³. testdata02 has 6 specialisms, 2 equipments, 6 departments, 151 rooms, 465 beds, 14 nights, 755 patients and 755 admissions with a search space of 10²015. testdata03 has 5 specialisms, 2 equipments, 5 departments, 131 rooms, 395 beds, 14 nights, 708 patients and 708 admissions with a search space of 10¹⁸⁴⁰. has 6 specialisms, 2 equipments, 6 departments, 155 rooms, 471 beds, 14 nights, testdata04 746 patients and 746 admissions with a search space of 10¹⁹⁹⁵. has 4 specialisms, 2 equipments, 4 departments, 102 rooms, 325 beds, 14 nights, testdata05 587 patients and 587 admissions with a search space of 10¹⁴⁷⁶. testdata06 has 4 specialisms, 2 equipments, 4 departments, 104 rooms, 313 beds, 14 nights, 685 patients and 685 admissions with a search space of 10^1711. testdata07 has 6 specialisms, 4 equipments, 6 departments, 162 rooms, 472 beds, 14 nights, 519 patients and 519 admissions with a search space of 10¹³⁸⁹. testdata08 has 6 specialisms, 4 equipments, 6 departments, 148 rooms, 441 beds, 21 nights, 895 patients and 895 admissions with a search space of 10²368. testdata09 has 4 specialisms, 4 equipments, 4 departments, 105 rooms, 310 beds, 28 nights, 1400 patients and 1400 admissions with a search space of 10³490. testdata10 has 4 specialisms, 4 equipments, 4 departments, 104 rooms, 308 beds, 56 nights, 1575 patients and 1575 admissions with a search space of 10³922. testdata11 has 4 specialisms, 4 equipments, 4 departments, 107 rooms, 318 beds, 91 nights, 2514 patients and 2514 admissions with a search space of 10⁶²⁹⁵. has 4 specialisms, 4 equipments, 4 departments, 105 rooms, 310 beds, 84 nights, testdata12 2750 patients and 2750 admissions with a search space of 10^6856. testdata13 has 5 specialisms, 4 equipments, 5 departments, 125 rooms, 368 beds, 28 nights,

907 patients and 1109 admissions with a search space of 10²⁸⁴⁷.



3.19. TRAVELING TOURNAMENT PROBLEM (TTP)

Schedule matches between *n* number of teams.



Hard constraints:

- Each team plays twice against every other team: once home and once away.
- Each team has exactly one match on each timeslot.
- No team must have more than three consecutive home or three consecutive away matches.
- No repeaters: no two consecutive matches of the same two opposing teams.

Soft constraints:

• Minimize the total distance traveled by all teams.

The problem is defined on Michael Trick's website (which contains the world records too).

Problem size

```
1-nl04
        has 6 days, 4 teams and 12 matches with a search space of 10^5.
        has 10 days, 6 teams and 30 matches with a search space of 10<sup>19</sup>.
1-nl06
1-nl08
        has 14 days, 8 teams and 56 matches with a search space of 10^43.
        has 18 days, 10 teams and 90 matches with a search space of 10^79.
1-nl10
        has 22 days, 12 teams and 132 matches with a search space of 10<sup>126</sup>.
1-nl12
1-nl14
        has 26 days, 14 teams and 182 matches with a search space of 10<sup>186</sup>.
        has 30 days, 16 teams and 240 matches with a search space of 10^259.
1-nl16
2-bra24 has 46 days, 24 teams and 552 matches with a search space of 10<sup>692</sup>.
        has 30 days, 16 teams and 240 matches with a search space of 10^259.
3-nfl16
        has 34 days, 18 teams and 306 matches with a search space of 10^346.
3-nfl18
```

```
3-nfl20 has 38 days, 20 teams and 380 matches with a search space of 10^447.
3-nfl22 has 42 days, 22 teams and 462 matches with a search space of 10<sup>562</sup>.
3-nfl24 has 46 days, 24 teams and 552 matches with a search space of 10^692.
3-nfl26 has 50 days, 26 teams and 650 matches with a search space of 10^838.
3-nfl28 has 54 days, 28 teams and 756 matches with a search space of 10^999.
3-nfl30 has 58 days, 30 teams and 870 matches with a search space of 10<sup>1175</sup>.
3-nfl32 has 62 days, 32 teams and 992 matches with a search space of 10<sup>1367</sup>.
4-super04 has 6 days, 4 teams and 12 matches with a search space of 10<sup>5</sup>.
4-super06 has 10 days, 6 teams and 30 matches with a search space of 10<sup>1</sup>9.
4-super08 has 14 days, 8 teams and 56 matches with a search space of 10^43.
4-super10 has 18 days, 10 teams and 90 matches with a search space of 10^79.
4-super12 has 22 days, 12 teams and 132 matches with a search space of 10<sup>126</sup>.
4-super14 has 26 days, 14 teams and 182 matches with a search space of 10<sup>186</sup>.
5-galaxy04 has 6 days, 4 teams and 12 matches with a search space of 10^5.
5-galaxy06 has 10 days, 6 teams and 30 matches with a search space of 10<sup>19</sup>.
5-galaxy08 has 14 days, 8 teams and 56 matches with a search space of 10<sup>4</sup>3.
5-galaxy10 has 18 days, 10 teams and 90 matches with a search space of 10<sup>7</sup>9.
5-galaxy12 has 22 days, 12 teams and 132 matches with a search space of 10<sup>126</sup>.
5-galaxy14 has 26 days, 14 teams and 182 matches with a search space of 10^186.
5-galaxy16 has 30 days, 16 teams and 240 matches with a search space of 10^259.
5-galaxy18 has 34 days, 18 teams and 306 matches with a search space of 10<sup>346</sup>.
5-galaxy20 has 38 days, 20 teams and 380 matches with a search space of 10^447.
5-galaxy22 has 42 days, 22 teams and 462 matches with a search space of 10^562.
5-galaxy24 has 46 days, 24 teams and 552 matches with a search space of 10<sup>692</sup>.
5-galaxy26 has 50 days, 26 teams and 650 matches with a search space of 10^838.
5-galaxy28 has 54 days, 28 teams and 756 matches with a search space of 10^999.
5-galaxy30 has 58 days, 30 teams and 870 matches with a search space of 10<sup>1175</sup>.
5-galaxy32 has 62 days, 32 teams and 992 matches with a search space of 10<sup>1367</sup>.
5-galaxy34 has 66 days, 34 teams and 1122 matches with a search space of 10<sup>1576</sup>.
5-galaxy36 has 70 days, 36 teams and 1260 matches with a search space of 10<sup>1801</sup>.
5-galaxy38 has 74 days, 38 teams and 1406 matches with a search space of 10<sup>2042</sup>.
5-galaxy40 has 78 days, 40 teams and 1560 matches with a search space of 10<sup>2</sup>301.
```

3.20. CHEAP TIME SCHEDULING

Schedule all tasks in time and on a machine to minimize power cost. Power prices differ in time. This is a form of *job shop scheduling*.

Hard constraints:

- Start time limits: Each task must start between its earliest start and latest start limit.
- Maximum capacity: The maximum capacity for each resource for each machine must not be exceeded.
- Startup and shutdown: Each machine must be active in the periods during which it has assigned tasks. Between tasks it is allowed to be idle to avoid startup and shutdown costs.

Medium constraints:

- Power cost: Minimize the total power cost of the whole schedule.
 - Machine power cost: Each active or idle machine consumes power, which infers a power cost (depending on the power price during that time).

- Task power cost: Each task consumes power too, which infers a power cost (depending on the power price during its time).
- Machine startup and shutdown cost: Every time a machine starts up or shuts down, an extra cost is incurred.

Soft constraints (addendum to the original problem definition):

• Start early: Prefer starting a task sooner rather than later.

The problem is defined by the ICON challenge.

Problem size

sample01 has 3 resources, 2 machines, 288 periods and 25 tasks with a search space of 10^53. sample02 has 3 resources, 2 machines, 288 periods and 50 tasks with a search space of 10^114. sample03 has 3 resources, 2 machines, 288 periods and 100 tasks with a search space of 10^226. sample04 has 3 resources, 5 machines, 288 periods and 100 tasks with a search space of 10^266. sample05 has 3 resources, 2 machines, 288 periods and 250 tasks with a search space of 10^584. sample06 has 3 resources, 5 machines, 288 periods and 250 tasks with a search space of 10^673. sample07 has 3 resources, 2 machines, 288 periods and 1000 tasks with a search space of 10^2388. sample08 has 3 resources, 5 machines, 288 periods and 1000 tasks with a search space of 10^2748. sample09 has 4 resources, 20 machines, 288 periods and 2000 tasks with a search space of 10^6668. instance00 has 1 resources, 10 machines, 288 periods and 200 tasks with a search space of 10^595. instance01 has 1 resources, 10 machines, 288 periods and 200 tasks with a search space of 10^599. instance02 has 1 resources, 10 machines, 288 periods and 200 tasks with a search space of 10^599. instance03 has 1 resources, 10 machines, 288 periods and 200 tasks with a search space of 10^591. instance04 has 1 resources, 10 machines, 288 periods and 200 tasks with a search space of 10^590. instance05 has 2 resources, 25 machines, 288 periods and 200 tasks with a search space of 10^667. instance06 has 2 resources, 25 machines, 288 periods and 200 tasks with a search space of 10^660. instance07 has 2 resources, 25 machines, 288 periods and 200 tasks with a search space of 10^662. instance08 has 2 resources, 25 machines, 288 periods and 200 tasks with a search space of 10^651. instance09 has 2 resources, 25 machines, 288 periods and 200 tasks with a search space of 10^659. instance10 has 2 resources, 20 machines, 288 periods and 500 tasks with a search space of 10^1657. instance11 has 2 resources, 20 machines, 288 periods and 500 tasks with a search space of 10^1644.

instance12 has 2 resources, 20 machines, 288 periods and 500 tasks with a search space of 10^1637. instance13 has 2 resources, 20 machines, 288 periods and 500 tasks with a search space of 10^1659. instance14 has 2 resources, 20 machines, 288 periods and 500 tasks with a search space of 10^1643. instance15 has 3 resources, 40 machines, 288 periods and 500 tasks with a search space of 10^1782. instance16 has 3 resources, 40 machines, 288 periods and 500 tasks with a search space of 10^1778. instance17 has 3 resources, 40 machines, 288 periods and 500 tasks with a search space of 10^1764. instance18 has 3 resources, 40 machines, 288 periods and 500 tasks with a search space of 10^1769. instance19 has 3 resources, 40 machines, 288 periods and 500 tasks with a search space of 10^1778. instance20 has 3 resources, 50 machines, 288 periods and 1000 tasks with a search space of 10^3689. instance21 has 3 resources, 50 machines, 288 periods and 1000 tasks with a search space of 10^3678. instance22 has 3 resources, 50 machines, 288 periods and 1000 tasks with a search space of 10^3706. instance23 has 3 resources, 50 machines, 288 periods and 1000 tasks with a search space of 10^3676. instance24 has 3 resources, 50 machines, 288 periods and 1000 tasks with a search space of 10^3681. instance25 has 3 resources, 60 machines, 288 periods and 1000 tasks with a search space of 10^3774. instance26 has 3 resources, 60 machines, 288 periods and 1000 tasks with a search space of 10^3737. instance27 has 3 resources, 60 machines, 288 periods and 1000 tasks with a search space of 10^3744. instance28 has 3 resources, 60 machines, 288 periods and 1000 tasks with a search space of 10^3731. instance29 has 3 resources, 60 machines, 288 periods and 1000 tasks with a search space of 10^3746. instance30 has 4 resources, 70 machines, 288 periods and 2000 tasks with a search space of 10^7718. instance31 has 4 resources, 70 machines, 288 periods and 2000 tasks with a search space of 10^7740. instance32 has 4 resources, 70 machines, 288 periods and 2000 tasks with a search space of 10^7686. instance33 has 4 resources, 70 machines, 288 periods and 2000 tasks with a search space of 10^7672. instance34 has 4 resources, 70 machines, 288 periods and 2000 tasks with a search space of 10^7695. instance35 has 4 resources, 80 machines, 288 periods and 2000 tasks with a search space of 10^7807. instance36 has 4 resources, 80 machines, 288 periods and 2000 tasks with a search space of 10^7814. instance37 has 4 resources, 80 machines, 288 periods and 2000 tasks with a search space of 10^7764. instance38 has 4 resources, 80 machines, 288 periods and 2000 tasks with a search space of 10^7736. instance39 has 4 resources, 80 machines, 288 periods and 2000 tasks with a search space of 10^7783.

instance40 has 4 resources, 90 machines, 288 periods and 4000 tasks with a search space of 10^15976. instance41 has 4 resources, 90 machines, 288 periods and 4000 tasks with a search space of 10^15935. instance42 has 4 resources, 90 machines, 288 periods and 4000 tasks with a search space of 10^15887. instance43 has 4 resources, 90 machines, 288 periods and 4000 tasks with a search space of 10^15896. instance44 has 4 resources, 90 machines, 288 periods and 4000 tasks with a search space of 10^15885. instance45 has 4 resources, 100 machines, 288 periods and 5000 tasks with a search space of 10^20173. instance46 has 4 resources, 100 machines, 288 periods and 5000 tasks with a search space of 10^20132. instance47 has 4 resources, 100 machines, 288 periods and 5000 tasks with a search space of 10^20126. instance48 has 4 resources, 100 machines, 288 periods and 5000 tasks with a search space of 10^20110. instance49 has 4 resources, 100 machines, 288 periods and 5000 tasks with a search space of 10^20078.

3.21. INVESTMENT ASSET CLASS ALLOCATION (PORTFOLIO OPTIMIZATION)

Decide the relative quantity to invest in each asset class.

Hard constraints:

- Risk maximum: the total standard deviation must not be higher than the standard deviation maximum.
 - Total standard deviation calculation takes asset class correlations into account by applying Markowitz Portfolio Theory.
- Region maximum: Each region has a quantity maximum.
- Sector maximum: Each sector has a quantity maximum.

Soft constraints:

• Maximize expected return.

Problem size

de_smet_1 has 1 regions, 3 sectors and 11 asset classes with a search space of 10^4. irrinki_1 has 2 regions, 3 sectors and 6 asset classes with a search space of 10^3.

Larger datasets have not been created or tested yet, but should not pose a problem. A good source of data is this Asset Correlation website.

3.22. CONFERENCE SCHEDULING

Assign each conference talk to a timeslot and a room. Timeslots can overlap. Read and write to and from an ***.xlsx** file that can be edited with LibreOffice or Excel.

Hard constraints:

- Talk type of timeslot: The type of a talk must match the timeslot's talk type.
- Room unavailable timeslots: A talk's room must be available during the talk's timeslot.
- Room conflict: Two talks can't use the same room during overlapping timeslots.
- Speaker unavailable timeslots: Every talk's speaker must be available during the talk's timeslot.
- Speaker conflict: Two talks can't share a speaker during overlapping timeslots.
- Generic purpose timeslot and room tags:
 - Speaker required timeslot tag: If a speaker has a required timeslot tag, then all of his or her talks must be assigned to a timeslot with that tag.
 - Speaker prohibited timeslot tag: If a speaker has a prohibited timeslot tag, then all of his or her talks cannot be assigned to a timeslot with that tag.
 - Talk required timeslot tag: If a talk has a required timeslot tag, then it must be assigned to a timeslot with that tag.
 - Talk prohibited timeslot tag: If a talk has a prohibited timeslot tag, then it cannot be assigned to a timeslot with that tag.
 - Speaker required room tag: If a speaker has a required room tag, then all of his or her talks must be assigned to a room with that tag.
 - Speaker prohibited room tag: If a speaker has a prohibited room tag, then all of his or her talks cannot be assigned to a room with that tag.
 - Talk required room tag: If a talk has a required room tag, then it must be assigned to a room with that tag.
 - Talk prohibited room tag: If a talk has a prohibited room tag, then it cannot be assigned to a room with that tag.
- Talk mutually-exclusive-talks tag: Talks that share such a tag must not be scheduled in overlapping timeslots.
- Talk prerequisite talks: A talk must be scheduled after all its prerequisite talks.

Soft constraints:

- Theme track conflict: Minimize the number of talks that share a theme tag during overlapping timeslots.
- Sector conflict: Minimize the number of talks that share a same sector tag during overlapping timeslots.
- Content audience level flow violation: For every content tag, schedule the introductory talks before the advanced talks.
- Audience level diversity: For every timeslot, maximize the number of talks with a different audience level.
- Language diversity: For every timeslot, maximize the number of talks with a different language.

- Generic purpose timeslot and room tags:
 - Speaker preferred timeslot tag: If a speaker has a preferred timeslot tag, then all of his or her talks should be assigned to a timeslot with that tag.
 - Speaker undesired timeslot tag: If a speaker has an undesired timeslot tag, then none of his or her talks should be assigned to a timeslot with that tag.
 - Talk preferred timeslot tag: If a talk has a preferred timeslot tag, then it should be assigned to a timeslot with that tag.
 - Talk undesired timeslot tag: If a talk has an undesired timeslot tag, then it should not be assigned to a timeslot with that tag.
 - Speaker preferred room tag: If a speaker has a preferred room tag, then all of his or her talks should be assigned to a room with that tag.
 - Speaker undesired room tag: If a speaker has an undesired room tag, then none of his or her talks should be assigned to a room with that tag.
 - Talk preferred room tag: If a talk has a preferred room tag, then it should be assigned to a room with that tag.
 - Talk undesired room tag: If a talk has an undesired room tag, then it should not be assigned to a room with that tag.
- Same day talks: All talks that share a theme tag or content tag should be scheduled in the minimum number of days (ideally in the same day).

Figure 3.14. Value proposition

Conference scheduling Optimize attandee experience when assigning each talk to a room and a timeslot. Used by Rooms 10:00 11:00 12:00 13:00 14:00 15:00 16:00 17:00 Devoxx Belgium 1 Breakout 1 Lunch oVirt £ $\overline{}$ Red Hat Summit large room Shared theme: Large room Artificial Intelligence for speaker Voxxed Zurich Breakout 2 Lunch A ç >:< normal room Same Same technology speaker Lab 1 Lunch £ power plugs room Speaker is only available untill 14:00 Talks

Problem size

18talks-6timeslots-5rooms has 18 talks, 6 timeslots and 5 rooms with a search space of 10²⁶. 36talks-12timeslots-5rooms has 36 talks, 12 timeslots and 5 rooms with a search space of 10⁶⁴. 72talks-12timeslots-10rooms has 72 talks, 12 timeslots and 10 rooms with a search space of 10¹⁴⁹.

108talks-18timeslots-10rooms has 108 talks, 18 timeslots and 10 rooms with a search space of 10^243.

216talks-18timeslots-20rooms has 216 talks, 18 timeslots and 20 rooms with a search space of 10^552.

3.23. ROCK TOUR

Drive the rock bank bus from show to show, but schedule shows only on available days.

Hard constraints:

- Schedule every required show.
- Schedule as many shows as possible.

Medium constraints:

• Maximize revenue opportunity.

- Minimize driving time.
- Visit sooner than later.

Soft constraints:

• Avoid long driving times.

Problem size

47shows has 47 shows with a search space of 10⁵⁹.

3.24. FLIGHT CREW SCHEDULING

Assign flights to pilots and flight attendants.

Hard constraints:

- Required skill: each flight assignment has a required skill. For example, flight AB0001 requires 2 pilots and 3 flight attendants.
- Flight conflict: each employee can only attend one flight at the same time
- Transfer between two flights: between two flights, an employee must be able to transfer from the arrival airport to the departure airport. For example, Ann arrives in Brussels at 10:00 and departs in Amsterdam at 15:00.
- Employee unavailability: the employee must be available on the day of the flight. For example, Ann is on PTO on 1-Feb.

Soft constraints:

- First assignment departing from home
- Last assignment arriving at home
- Load balance flight duration total per employee

Problem size

175flights-7days-Europe has 2 skills, 50 airports, 150 employees, 175 flights and 875 flight assignments with a search space of 10^1904.

700flights-28days-Europe has 2 skills, 50 airports, 150 employees, 700 flights and 3500 flight assignments with a search space of 10⁷616.

875flights-7days-Europe has 2 skills, 50 airports, 750 employees, 875 flights and 4375 flight assignments with a search space of 10¹²⁵⁷⁸.

175flights-7days-US has 2 skills, 48 airports, 150 employees, 175 flights and 875 flight assignments with a search space of 10¹904.

CHAPTER 4. DOWNLOADING AND BUILDING RED HAT BUILD OF OPTAPLANNER EXAMPLES

You can download the Red Hat Build of OptaPlanner examples as a part of the Red Hat Build of OptaPlanner sources package available on the Red Hat Customer Portal.

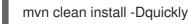


NOTE

Red Hat Build of OptaPlanner has no GUI dependencies. It runs just as well on a server or a mobile JVM as it does on the desktop.

Procedure

- 1. Navigate to the Software Downloads page in the Red Hat Customer Portal (login required), and select the product and version from the drop-down options:
 - Product: Red Hat Build of OptaPlanner
 - Version: 8.38
- 2. Download Red Hat Build of OptaPlanner 8.38 Source Distribution
- 3. Extract the **rhbop-8.38.0-optaplanner-sources.zip** file. The extracted **org.optaplanner.optaplanner-8.38.0.Final-redhat-00004/optaplannerexamples/src/main/java/org/optaplanner/examples** directory contains example source code.
- 4. To build the examples, in the **org.optaplanner.optaplanner-8.38.0.Final-redhat-00004** directory enter the following command:



5. Change to the examples directory:



6. To run the examples, enter the following command:

mvn exec java

CHAPTER 5. GETTING STARTED WITH RED HAT BUILD OF OPTAPLANNER ON THE RED HAT BUILD OF QUARKUS PLATFORM

Red Hat Build of OptaPlanner is integrated with the Red Hat build of Quarkus platform. Versions of platform artifact dependencies, including OptaPlanner dependencies, are maintained in the Quarkus bill of materials (BOM) file, **com.redhat.quarkus.platform:quarkus-bom**. You do not need to specify which dependency versions work together. Instead, you can import the Quarkus BOM file to the **pom.xml** configuration file, where the dependency versions are included in the **<dependencyManagement>** section. Therefore, you do not need to list the versions of individual Quarkus dependencies that are managed by the specified BOM in the **pom.xml** file.

Additional resources

- For instructions about using the Maven plug-in to create an OptaPlanner project on the Quarkus platform, see Section 5.2, "Using the Maven plug-in to create an Red Hat Build of OptaPlanner project on the Quarkus platform".
- For instructions about using the **code.quarkus.redhat.com** website to generate an OptaPlanner project on the Quarkus platform, see Section 5.3, "Using code.quarkus.redhat.com to create an Red Hat Build of OptaPlanner project on the Quarkus platform".
- For instructions about using the CLI to generate an OptaPlanner project on the Quarkus platform, see Section 5.4, "Using the Quarkus CLI to create an Red Hat Build of OptaPlanner project on the Quarkus platform".

5.1. APACHE MAVEN AND RED HAT BUILD OF QUARKUS

Apache Maven is a distributed build automation tool used in Java application development to create, manage, and build software projects. Maven uses standard configuration files called Project Object Model (POM) files to define projects and manage the build process. POM files describe the module and component dependencies, build order, and targets for the resulting project packaging and output using an XML file. This ensures that the project is built in a correct and uniform manner.

Maven repositories

A Maven repository stores Java libraries, plug-ins, and other build artifacts. The default public repository is the Maven 2 Central Repository, but repositories can be private and internal within a company to share common artifacts among development teams. Repositories are also available from third parties.

You can use the online Maven repository with your Quarkus projects or you can download the Red Hat build of Quarkus Maven repository.

Maven plug-ins

Maven plug-ins are defined parts of a POM file that achieve one or more goals. Quarkus applications use the following Maven plug-ins:

- Quarkus Maven plug-in (**quarkus-maven-plugin**): Enables Maven to create Quarkus projects, supports the generation of uber-JAR files, and provides a development mode.
- Maven Surefire plug-in (**maven-surefire-plugin**): Used during the test phase of the build lifecycle to execute unit tests on your application. The plug-in generates text and XML files that contain the test reports.

5.1.1. Configuring the Maven settings.xml file for the online repository

You can use the online Maven repository with your Maven project by configuring your user **settings.xml** file. This is the recommended approach. Maven settings used with a repository manager or repository on a shared server provide better control and manageability of projects.



NOTE

When you configure the repository by modifying the Maven **settings.xml** file, the changes apply to all of your Maven projects.

Procedure

1. Open the Maven ~/.m2/settings.xml file in a text editor or integrated development environment (IDE).



NOTE

If there is not a **settings.xml** file in the ~/.**m2**/ directory, copy the **settings.xml** file from the **\$MAVEN_HOME**/.**m2**/conf/ directory into the ~/.**m2**/ directory.

2. Add the following lines to the **<profiles>** element of the **settings.xml** file:

```
<!-- Configure the Maven repository -->
<profile>
 <id>red-hat-enterprise-maven-repository</id>
 <repositories>
  <repository>
   <id>red-hat-enterprise-maven-repository</id>
   <url>https://maven.repository.redhat.com/ga/</url>
   <releases>
    <enabled>true</enabled>
   </releases>
   <snapshots>
    <enabled>false</enabled>
   </snapshots>
  </repository>
 </repositories>
 <pluginRepositories>
  <pluginRepository>
   <id>red-hat-enterprise-maven-repository</id>
   <url>https://maven.repository.redhat.com/ga/</url>
   <releases>
    <enabled>true</enabled>
   </releases>
   <snapshots>
    <enabled>false</enabled>
   </snapshots>
  </pluginRepository>
 </pluginRepositories>
</profile>
```

3. Add the following lines to the **<activeProfiles>** element of the **settings.xml** file and save the file.

<activeProfile>red-hat-enterprise-maven-repository</activeProfile>

5.1.2. Downloading and configuring the Quarkus Maven repository

If you do not want to use the online Maven repository, you can download and configure the Quarkus Maven repository to create a Quarkus application with Maven. The Quarkus Maven repository contains many of the requirements that Java developers typically use to build their applications. This procedure describes how to edit the **settings.xml** file to configure the Quarkus Maven repository.



NOTE

When you configure the repository by modifying the Maven **settings.xml** file, the changes apply to all of your Maven projects.

Procedure

- 1. Download the Red Hat build of Quarkus Maven repository ZIP file from the Software Downloads page of the Red Hat Customer Portal (login required).
- 2. Expand the downloaded archive.
- 3. Change directory to the ~/.**m2**/ directory and open the Maven **settings.xml** file in a text editor or integrated development environment (IDE).
- Add the following lines to the <profiles> element of the settings.xml file, where QUARKUS_MAVEN_REPOSITORY is the path of the Quarkus Maven repository that you downloaded. The format of QUARKUS_MAVEN_REPOSITORY must be file://\$PATH, for example file://home/userX/rh-quarkus-2.13.8.GA-maven-repository/maven-repository.

<!-- Configure the Quarkus Maven repository --> <profile> <id>red-hat-quarkus-maven-repository</id> <repositories> <repository> <id>red-hat-quarkus-maven-repository</id> <url>QUARKUS MAVEN REPOSITORY</url> <releases> <enabled>true</enabled> </releases> <snapshots> <enabled>false</enabled> </snapshots> </repository> </repositories> <pluginRepositories> <pluginRepository> <id>red-hat-quarkus-maven-repository</id> <url>QUARKUS MAVEN REPOSITORY</url> <releases> <enabled>true</enabled> </releases> <snapshots> <enabled>false</enabled> </snapshots>

</pluginRepository> </pluginRepositories> </profile>

5. Add the following lines to the **<activeProfiles>** element of the **settings.xml** file and save the file.

<activeProfile>red-hat-quarkus-maven-repository</activeProfile>



IMPORTANT

If your Maven repository contains outdated artifacts, you might encounter one of the following Maven error messages when you build or deploy your project, where **ARTIFACT_NAME** is the name of a missing artifact and **PROJECT_NAME** is the name of the project you are trying to build:

- Missing artifact PROJECT_NAME
- [ERROR] Failed to execute goal on project *ARTIFACT_NAME*; Could not resolve dependencies for *PROJECT_NAME*

To resolve the issue, delete the cached version of your local repository located in the ~/.m2/repository directory to force a download of the latest Maven artifacts.

5.2. USING THE MAVEN PLUG-IN TO CREATE AN RED HAT BUILD OF OPTAPLANNER PROJECT ON THE QUARKUS PLATFORM

You can get up and running with an Red Hat Build of OptaPlanner and Quarkus application using Apache Maven and the Quarkus Maven plug-in.

Prerequisites

- OpenJDK 11 or later is installed. Red Hat build of Open JDK is available from the Software Downloads page in the Red Hat Customer Portal (login required).
- Apache Maven 3.8 or higher is installed. Maven is available from the Apache Maven Project website.

Procedure

1. In a command terminal, enter the following command to verify that Maven is using JDK 11 and that the Maven version is 3.8 or higher:

mvn --version

- 2. If the preceding command does not return JDK 11, add the path to JDK 11 to the PATH environment variable and enter the preceding command again.
- 3. To generate a Quarkus OptaPlanner quickstart project, enter the following command, where **redhat-0000x** is the current version of the Quarkus BOM file:

-DprojectArtifactId=optaplanner-quickstart \

-DplatformGroupId=com.redhat.quarkus.platform

-DplatformArtifactId=quarkus-bom

-DplatformVersion=2.13.8.SP1-redhat-0000x \

-DnoExamples

-Dextensions="resteasy,resteasy-jackson,optaplanner-quarkus,optaplanner-quarkus-jackson" $\$

This command create the following elements in the ./optaplanner-quickstart directory:

- The Maven structure
- Example Dockerfile file in src/main/docker
- The application configuration file

Table 5.1. Properties used in the mvnio.quarkus:quarkus-maven-plugin:2.13.8.SP1redhat-0000x:create command

Property	Description
projectGroupId	The group ID of the project.
projectArtifactId	The artifact ID of the project.
extensions	A comma-separated list of Quarkus extensions to use with this project. For a full list of Quarkus extensions, enter mvn quarkus:list-extensions on the command line.
noExamples	Creates a project with the project structure but without tests or classes.

The values of the **projectGroupID** and the **projectArtifactID** properties are used to generate the project version. The default project version is **1.0.0-SNAPSHOT**.

4. To view your OptaPlanner project, change directory to the OptaPlanner Quickstarts directory:



5. Review the **pom.xml** file. The content should be similar to the following example:

```
<?xml version="1.0"?>
<project xsi:schemaLocation="http://maven.apache.org/POM/4.0.0
https://maven.apache.org/xsd/maven-4.0.0.xsd" xmlns="http://maven.apache.org/POM/4.0.0"
xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance">
<modelVersion>4.0.0</modelVersion>
<groupId>com.example</groupId>
<artifactId>optaplanner-quickstart</artifactId>
<version>1.0.0-SNAPSHOT</version>
<properties>
<compiler-plugin.version>3.8.1</compiler-plugin.version>
<maven.compiler.release>11</maven.compiler.release>
<project.build.sourceEncoding>UTF-8</project.build.sourceEncoding>
```

<project.reporting.outputEncoding>UTF-8</project.reporting.outputEncoding> <quarkus.platform.artifact-id>quarkus-bom</quarkus.platform.artifact-id> <quarkus.platform.group-id>com.redhat.quarkus.platform</quarkus.platform.group-id> <quarkus.platform.version>2.13.8.SP1-redhat-0000x</quarkus.platform.version> <skipITs>true</skipITs> <surefire-plugin.version>3.0.0-M7</surefire-plugin.version> </properties> <dependencyManagement> <dependencies> <dependency> <groupId>\${quarkus.platform.group-id}</groupId> <artifactId>\${quarkus.platform.artifact-id}</artifactId> <version>\${quarkus.platform.version}</version> <type>pom</type> <scope>import</scope> </dependency> <dependency> <groupId>\${quarkus.platform.group-id}</groupId> <artifactId>quarkus-optaplanner-bom</artifactId> <version>\${quarkus.platform.version}</version> <type>pom</type> <scope>import</scope> </dependency> </dependencies> </dependencyManagement> <dependencies> <dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-guarkus</artifactId> </dependency> <dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-quarkus-jackson</artifactId> </dependency> <dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-resteasy-jackson</artifactId> </dependency> <dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-resteasy</artifactId> </dependency> <dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-arc</artifactId> </dependency> <dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-junit5</artifactId> <scope>test</scope> </dependency> <dependency> <groupId>io.rest-assured</groupId> <artifactId>rest-assured</artifactId> <scope>test</scope> </dependency>

</dependencies> <repositories> <repository> <releases> <enabled>true</enabled> </releases> <snapshots> <enabled>false</enabled> </snapshots> <id>redhat</id> <url>https://maven.repository.redhat.com/ga</url> </repository> </repositories> <pluginRepositories> <pluginRepository> <releases> <enabled>true</enabled> </releases> <snapshots> <enabled>false</enabled> </snapshots> <id>redhat</id> <url>https://maven.repository.redhat.com/ga</url> </pluginRepository> </pluginRepositories> <build> <plugins> <plugin> <groupId>\${quarkus.platform.group-id}</groupId> <artifactId>quarkus-maven-plugin</artifactId> <version>\${quarkus.platform.version}</version> <extensions>true</extensions> <executions> <execution> <goals> <goal>build</goal> <goal>generate-code</goal> <goal>generate-code-tests</goal> </goals> </execution> </executions> </plugin> <plugin> <artifactId>maven-compiler-plugin</artifactId> <version>\${compiler-plugin.version}</version> <configuration> <compilerArgs> <arg>-parameters</arg> </compilerArgs> </configuration> </plugin> <plugin> <artifactId>maven-surefire-plugin</artifactId> <version>\${surefire-plugin.version}</version> <configuration> <systemPropertyVariables>

<java.util.logging.manager>org.jboss.logmanager.LogManager</java.util.logging.manager> <maven.home>\${maven.home}</maven.home> </systemPropertyVariables> </configuration> </plugin> <plugin> <artifactId>maven-failsafe-plugin</artifactId> <version>\${surefire-plugin.version}</version> <executions> <execution> <goals> <goal>integration-test</goal> <goal>verify</goal> </goals> <configuration> <systemPropertyVariables> <native.image.path>\${project.build.directory}/\${project.build.finalName}runner</native.image.path> <java.util.logging.manager>org.jboss.logmanager.LogManager</java.util.logging.manager> <maven.home>\${maven.home}</maven.home> </systemPropertyVariables> </configuration> </execution> </executions> </plugin> </plugins> </build> <profiles> <profile> <id>native</id> <activation> <property> <name>native</name> </property> </activation> <properties> <skiplTs>false</skiplTs> <quarkus.package.type>native</quarkus.package.type> </properties> </profile> </profiles> </project>

5.3. USING CODE.QUARKUS.REDHAT.COM TO CREATE AN RED HAT BUILD OF OPTAPLANNER PROJECT ON THE QUARKUS PLATFORM

You can use the **code.quarkus.redhat.com** website to generate an Red Hat Build of OptaPlanner Quarkus Maven project and automatically add and configure the extensions that you want to use in your application.

This section walks you through the process of generating an OptaPlanner Maven project and includes the following topics:

- Specifying basic details about your application.
- Choosing the extensions that you want to include in your project.
- Generating a downloadable archive with your project files.
- Using the custom commands for compiling and starting your application.

Prerequisites

• You have a web browser.

Procedure

- 1. Open https://code.quarkus.redhat.com in your web browser:
- 2. Specify details about your project:
- 3. Enter a group name for your project. The format of the name follows the Java package naming convention, for example, **com.example**.
- 4. Enter a name that you want to use for Maven artifacts generated from your project, for example **code-with-quarkus**.
- 5. Select **Build Tool** > **Maven** to specify that you want to create a Maven project. The build tool that you choose determines the items:
 - The directory structure of your generated project
 - The format of configuration files used in your generated project
 - The custom build script and command for compiling and starting your application that **code.quarkus.redhat.com** displays for you after you generate your project



NOTE

Red Hat provides support for using **code.quarkus.redhat.com** to create OptaPlanner Maven projects only. Generating Gradle projects is not supported by Red Hat.

- 6. Enter a version to be used in artifacts generated from your project. The default value of this field is **1.0.0-SNAPSHOT**. Using semantic versioning is recommended, but you can use a different type of versioning if you prefer.
- 7. Enter the package name of artifacts that the build tool generates when you package your project.

According to the Java package naming conventions the package name should match the group name that you use for your project, but you can specify a different name.

- 8. Select the following extensions to include as dependencies:
 - RESTEasy JAX-RS (quarkus-resteasy)
 - RESTEasy Jackson (quarkus-resteasy-jackson)
 - OptaPlanner Al constraint solver(optaplanner-quarkus)

- OptaPlanner Jackson (optaplanner-quarkus-jackson) Red Hat provides different levels of support for individual extensions on the list, which are indicated by labels next to the name of each extension:
 - *SUPPORTED* extensions are fully supported by Red Hat for use in enterprise applications in production environments.
 - *TECH-PREVIEW* extensions are subject to limited support by Red Hat in production environments under the Technology Preview Features Support Scope.
 - *DEV-SUPPORT* extensions are not supported by Red Hat for use in production environments, but the core functionalities that they provide are supported by Red Hat developers for use in developing new applications.
 - DEPRECATED extension are planned to be replaced with a newer technology or implementation that provides the same functionality.
 Unlabeled extensions are not supported by Red Hat for use in production environments.
- 9. Select *Generate your application* to confirm your choices and display the overlay screen with the download link for the archive that contains your generated project. The overlay screen also shows the custom command that you can use to compile and start your application.
- 10. Select **Download the ZIP** to save the archive with the generated project files to your system.
- 11. Extract the contents of the archive.
- 12. Navigate to the directory that contains your extracted project files:



13. Compile and start your application in development mode:

./mvnw compile quarkus:dev

5.4. USING THE QUARKUS CLI TO CREATE AN RED HAT BUILD OF OPTAPLANNER PROJECT ON THE QUARKUS PLATFORM

You can use the Quarkus command line interface (CLI) to create a Quarkus OptaPlanner project.

Prerequisites

• You have installed the Quarkus CLI. For information, see Building Quarkus Apps with Quarkus Command Line Interface.

Procedure

1. Create a Quarkus application:



quarkus create app -P io.quarkus:quarkus-bom:2.13.8.SP1-redhat-0000x

2. To view the available extensions, enter the following command:



This command returns the following extensions:

optaplanner-quarkus optaplanner-quarkus-benchmark optaplanner-quarkus-jackson optaplanner-quarkus-jsonb

3. Enter the following command to add extensions to the project's **pom.xml** file:

quarkus ext add resteasy-jackson quarkus ext add optaplanner-quarkus quarkus ext add optaplanner-quarkus-jackson

4. Open the **pom.xml** file in a text editor. The contents of the file should look similar to the following example:

```
<?xml version="1.0"?>
<project xsi:schemaLocation="http://maven.apache.org/POM/4.0.0"
https://maven.apache.org/xsd/maven-4.0.0.xsd" xmlns="http://maven.apache.org/POM/4.0.0"
xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance">
 <modelVersion>4.0.0</modelVersion>
 <groupId>org.acme</groupId>
 <artifactId>code-with-quarkus-optaplanner</artifactId>
 <version>1.0.0-SNAPSHOT</version>
 <properties>
<compiler-plugin.version>3.8.1</compiler-plugin.version>
<maven.compiler.parameters>true</maven.compiler.parameters>
<maven.compiler.source>11</maven.compiler.source>
<maven.compiler.target>11</maven.compiler.target>
<project.build.sourceEncoding>UTF-8</project.build.sourceEncoding>
<project.reporting.outputEncoding>UTF-8</project.reporting.outputEncoding>
<quarkus.platform.artifact-id>quarkus-bom</quarkus.platform.artifact-id>
<quarkus.platform.group-id>io.quarkus</quarkus.platform.group-id>
<quarkus.platform.version>2.13.8.SP1-redhat-0000x</quarkus.platform.version>
<surefire-plugin.version>3.0.0-M5</surefire-plugin.version>
 </properties>
 <dependencyManagement>
<dependencies>
 <dependency>
   <groupId>${quarkus.platform.group-id}</groupId>
   <artifactId>${quarkus.platform.artifact-id}</artifactId>
   <version>${quarkus.platform.version}</version>
   <type>pom</type>
   <scope>import</scope>
 </dependency>
 <dependency>
   <groupId>io.quarkus.platform</groupId>
   <artifactId>optaplanner-quarkus</artifactId>
   <version>2.2.2.Final</version>
   <type>pom</type>
   <scope>import</scope>
 </dependency>
</dependencies>
 </dependencyManagement>
 <dependencies>
```

<dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-arc</artifactId> </dependency> <dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-resteasy</artifactId> </dependency> <dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-guarkus</artifactId> </dependency> <dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-quarkus-jackson</artifactId> </dependency> <dependency> <groupId>io.guarkus</groupId> <artifactId>quarkus-resteasy-jackson</artifactId> </dependency> <dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-junit5</artifactId> <scope>test</scope> </dependency> <dependency> <groupId>io.rest-assured</groupId> <artifactId>rest-assured</artifactId> <scope>test</scope> </dependency> </dependencies> <build> <plugins> <plugin> <groupId>\${quarkus.platform.group-id}</groupId> <artifactId>quarkus-maven-plugin</artifactId> <version>\${quarkus.platform.version}</version> <extensions>true</extensions> <executions> <execution> <goals> <goal>build</goal> <goal>generate-code</goal> <goal>generate-code-tests</goal> </goals> </execution> </executions> </plugin> <plugin> <artifactId>maven-compiler-plugin</artifactId> <version>\${compiler-plugin.version}</version> <configuration> <parameters>\${maven.compiler.parameters}</parameters> </configuration> </plugin> <plugin>

```
<artifactId>maven-surefire-plugin</artifactId>
   <version>${surefire-plugin.version}</version>
   <configuration>
    <systemPropertyVariables>
<java.util.logging.manager>org.jboss.logmanager.LogManager</java.util.logging.manager>
     <maven.home>${maven.home}</maven.home>
    </systemPropertyVariables>
   </configuration>
  </plugin>
</plugins>
 </build>
 <profiles>
<profile>
 <id>native</id>
 <activation>
   <property>
    <name>native</name>
   </property>
 </activation>
 <build>
   <plugins>
    <plugin>
     <artifactId>maven-failsafe-plugin</artifactId>
     <version>${surefire-plugin.version}</version>
     <executions>
      <execution>
        <goals>
         <goal>integration-test</goal>
         <goal>verify</goal>
        </goals>
        <configuration>
         <systemPropertyVariables>
          <native.image.path>${project.build.directory}/${project.build.finalName}-
run</native.image.path>
<java.util.logging.manager>org.jboss.logmanager.LogManager</java.util.logging.manager>
          <maven.home>${maven.home}</maven.home>
         </systemPropertyVariables>
       </configuration>
      </execution>
     </executions>
    </plugin>
   </plugins>
 </build>
 <properties>
   <quarkus.package.type>native</quarkus.package.type>
 </properties>
</profile>
 </profiles>
</project>
```

PART III. THE RED HAT BUILD OF OPTAPLANNER SOLVER

Solving a planning problem with OptaPlanner consists of the following steps:

- 1. Model your planning problem as a class annotated with the **@PlanningSolution** annotation (for example, the **NQueens** class).
- 2. Configure a Solver (for example a First Fit and Tabu Search solver for any NQueens instance).
- 3. Load a problem data set from your data layer (for example a Four Queens instance). That is the planning problem.
- 4. Solve it with Solver.solve(problem), which returns the best solution found.

Solver configuration Domain (.java, ...) Score function (.drl, ...) Hard constraints: - CPU power capacity - RAM memory capacity Computer Process Network bandwidth capacity Soft constraints: Minimize maintenance cost SolverFactory Problem Computers buildSolver() Solution 7 Each process assigned 6 to a computer Processes 7 х 5 Solver 5 4 6 4 solve(problem)

Input/Output overview

Use 1 SolverFactory per application and 1 Solver per dataset.

CHAPTER 6. CONFIGURING THE RED HAT BUILD OF OPTAPLANNER SOLVER

You can use the following methods to configure your OptaPlanner solver:

- Use an XML file.
- Use the **SolverConfig** API.
- Add class annotations and JavaBean property annotations on the domain model.
- Control the method that OptaPlanner uses to access your domain.
- Define custom properties.

6.1. USING AN XML FILE TO CONFIGURE THE OPTAPLANNER SOLVER

Each example project has a solver configuration file that you can edit. The **<EXAMPLE>SolverConfig.xml** file is located in the **org.optaplanner.optaplanner-8.38.0.Final-redhat-00004/optaplanner-examples/src/main/resources/org/optaplanner/examples/<EXAMPLE>** directory, where **<EXAMPLE>** is the name of the OptaPlanner example project. Alternatively, you can create a **SolverFactory** from a file with **SolverFactory.createFromXmlFile()**. However, for portability reasons, a classpath resource is recommended.

Both a **Solver** and a **SolverFactory** have a generic type called **Solution_**, which is the class representing a planning problem and solution.

OptaPlanner makes it relatively easy to switch optimization algorithms by changing the configuration.

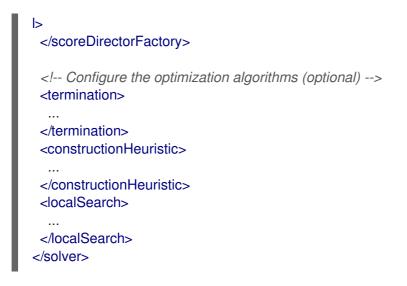
Procedure

- 1. Build a **Solver** instance with the **SolverFactory**.
- 2. Configure the solver configuration XML file:
 - a. Define the model.
 - b. Define the score function.
 - c. Optional: Configure the optimization algorithm. The following example is a solver XML file for the NQueens problem:

<?xml version="1.0" encoding="UTF-8"?> <solver xmlns="https://www.optaplanner.org/xsd/solver" xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance" xsi:schemaLocation="https://www.optaplanner.org/xsd/solver https://www.optaplanner.org/xsd/solver/solver.xsd"> <!-- Define the model --> <solutionClass>org.optaplanner.examples.nqueens.domain.NQueens</solutionClass> <entityClass>org.optaplanner.examples.nqueens.domain.Queen</entityClass>

<!-- Define the score function --> <scoreDirectorFactory>

<scoreDrl>org/optaplanner/examples/nqueens/optional/nQueensConstraints.drl</scoreDr





On some environments, for example OSGi and JBoss modules, classpath resources such as the solver config, score DRLs, and domain classe) in your JAR files might not be available to the default **ClassLoader** of the **optaplanner-core** JAR file. In those cases, provide the **ClassLoader** of your classes as a parameter:

SolverFactory<NQueens> solverFactory = SolverFactory.createFromXmlResource(".../nqueensSolverConfig.xml", getClass().getClassLoader());

3. Configure the **SolverFactory** with a solver configuration XML file, provided as a classpath resource as defined by **ClassLoader.getResource()**:

SolverFasctory<NQueens> solverFactory = SolverFactory.createFromXmlResource("org/optaplanner/examples/nqueens/optional/nqueensSolverConfig.xml"); Solver<NQueens> solver = solverFactory.buildSolver();

6.2. USING THE JAVA API TO CONFIGURE THE OPTAPLANNER SOLVER

You can configure a solver by using the **SolverConfig** API. This is especially useful to change values dynamically at runtime. The following example changes the running time based on system properties before building the **Solver** in the NQueens project:

SolverConfig solverConfig = SolverConfig.createFromXmlResource("org/optaplanner/examples/nqueens/optional/nqueensSolverConfig.xml"); solverConfig.withTerminationConfig(new TerminationConfig() .withMinutesSpentLimit(userInput));

SolverFactory<NQueens> solverFactory = SolverFactory.create(solverConfig); Solver<NQueens> solver = solverFactory.buildSolver();

Every element in the solver configuration XML file is available as a **Config** class or a property on a **Config** class in the package namespace **org.optaplanner.core.config**. These **Config** classes are the Java representation of the XML format. They build the runtime components of the package namespace **org.optaplanner.core.impl** and assemble them into an efficient **Solver**.

To configure a **SolverFactory** dynamically for each user request, build a template **SolverConfig** during initialization and copy it with the copy constructor for each user request. The following example shows how to do this with the NQueens problem:

private SolverConfig template;

public void init() {
 template = SolverConfig.createFromXmlResource(
 "org/optaplanner/examples/nqueens/optional/nqueensSolverConfig.xml");
 template.setTerminationConfig(new TerminationConfig());
}

// Called concurrently from different threads
public void userRequest(..., long userInput) {
 SolverConfig solverConfig = new SolverConfig(template); // Copy it
 solverConfig.getTerminationConfig().setMinutesSpentLimit(userInput);
 SolverFactory<NQueens> solverFactory = SolverFactory.create(solverConfig);
 Solver<NQueens> solver = solverFactory.buildSolver();
 ...

6.3. OPTAPLANNER ANNOTATION

}

You must specify which classes in your domain model are planning entities, which properties are planning variables, and so on. Use one of the following methods to add annotations to your OptaPlanner project:

- Add class annotations and JavaBean property annotations on the domain model. The property annotations must be on the getter method, not on the setter method. Annotated getter methods do not need to be public. This is the recommended method.
- Add class annotations and field annotations on the domain model. Annotated fields do not need to be public.

6.4. SPECIFYING OPTAPLANNER DOMAIN ACCESS

By default, OptaPlanner accesses your domain using reflection. Reflection is reliable but slow compared to direct access. Alternatively, you can configure OptaPlanner to access your domain using Gizmo, which will generate bytecode that directly accesses the fields and methods of your domain without reflection. However, this method has the following restrictions:

- The planning annotations can only be on public fields and public getters.
- io.quarkus.gizmo:gizmo must be on the classpath.



NOTE

These restrictions do not apply when you use OptaPlanner with Quarkus because Gizmo is the default domain access type.

Procedure

To use Gizmo outside of Quarkus, set the **domainAccessType** in the solver configuration:

```
<solver>
<domainAccessType>GIZMO</domainAccessType>
</solver>
```

6.5. CONFIGURING CUSTOM PROPERTIES

In your OptaPlanner projects, you can add custom properties to solver configuration elements that instantiate classes and have documents that explicitly mention custom properties.

Prerequisites

• You have a solver.

Procedure

1. Add a custom property.

For example, if your **EasyScoreCalculator** has heavy calculations which are cached and you want to increase the cache size in one benchmark add the **myCacheSize** property:

<scoreDirectorFactory> <easyScoreCalculatorClass>...MyEasyScoreCalculator</easyScoreCalculatorClass> <easyScoreCalculatorCustomProperties> <property name="myCacheSize" value="1000"/><!-- Override value --> </easyScoreCalculatorCustomProperties> </scoreDirectorFactory>

2. Add a public setter for each custom property, which is called when a **Solver** is built.

```
public class MyEasyScoreCalculator extends EasyScoreCalculator<MySolution,
SimpleScore> {
    private int myCacheSize = 500; // Default value
    @SuppressWarnings("unused")
    public void setMyCacheSize(int myCacheSize) {
        this.myCacheSize = myCacheSize;
    }
    ...
}
```

Most value types are supported, including **boolean**, **int**, **double**, **BigDecimal**, **String** and **enums**.

CHAPTER 7. USING THE OPTAPLANNER SOLVER

A solver finds the best and optimal solution to your planning problem. A solver can only solve one planning problem instance at a time. Solvers are built with the **SolverFactory** method:

```
public interface Solver<Solution_> {
    Solution_ solve(Solution_ problem);
    ...
}
```

A solver should only be accessed from a single thread, except for the methods that are specifically documented in **javadoc** as being thread-safe. The **solve()** method hogs the current thread. Hogging the thread can cause HTTP timeouts for REST services and it requires extra code to solve multiple data sets in parallel. To avoid such issues, use a **SolverManager** instead.

7.1. SOLVING A PROBLEM

Use the solver to solve a planning problem.

Prerequisites

- A **Solver** built from a solver configuration
- An @PlanningSolution annotation that represents the planning problem instance

Procedure

Provide the planning problem as an argument to the **solve()** method. The solver will return the best solution found.

The following example solves the NQueens problem:

NQueens problem = ...; NQueens bestSolution = solver.solve(problem);

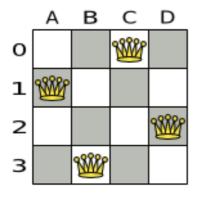
In this example, the **solve()** method will return an **NQueens** instance with every **Queen** assigned to a **Row**.



NOTE

The solution instance given to the **solve(Solution)** method can be partially or fully initialized, which is often the case in repeated planning.

Figure 7.1. Best Solution for the Four Queens Puzzle in 8ms (Also an Optimal Solution)



The **solve(Solution)** method can take a long time depending on the problem size and the solver configuration. The **Solver** intelligently works through the search space of possible solutions and remembers the best solution it encounters during solving. Depending on a number of factors, including problem size, how much time the **Solver** has, the solver configuration, and so forth, the **best** solution might or might not be an **optimal** solution.



NOTE

The solution instance given to the method **solve(Solution)** is changed by the **Solver**, but do not mistake it for the best solution.

The solution instance returned by the methods **solve(Solution)** or **getBestSolution()** is most likely a planning clone of the instance given to the method **solve(Solution)**, which implies it is a different instance.

7.2. SOLVER ENVIRONMENT MODE

The solver environment mode enables you to detect common bugs in your implementation. It does not affect the logging level.

A solver has a single random instance. Some solver configurations use the random instance a lot more than others. For example, the Simulated Annealing algorithm depends highly on random numbers, while **Tabu Search** only depends on it to resolve score ties. The environment mode influences the seed of that random instance.

You can set the environment mode in the solver configuration XML file. The following example sets the **FAST_ASSERT** mode:

```
<solver xmlns="https://www.optaplanner.org/xsd/solver"
xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance"
xsi:schemaLocation="https://www.optaplanner.org/xsd/solver
https://www.optaplanner.org/xsd/solver/solver.xsd">
<environmentMode>FAST_ASSERT</environmentMode>
...
```

</solver>

The following list describes the environment modes that you can use in the solver configuration file:

• **FULL_ASSERT** mode turns on all assertions, for example the assertion that the incremental score calculation is uncorrupted for each move, to fail-fast on a bug in a Move implementation, a

constraint, the engine itself, and so on. This mode is reproducible. It is also intrusive because it calls the method **calculateScore()** more frequently than a non-assert mode. The **FULL_ASSERT** mode is very slow because it does not rely on incremental score calculation.

- NON_INTRUSIVE_FULL_ASSERT mode turns on several assertions to fail-fast on a bug in a Move implementation, a constraint, the engine itself, and so on. This mode is reproducible. It is non-intrusive because it does not call the method **calculateScore()** more frequently than a non assert mode. The NON_INTRUSIVE_FULL_ASSERT mode is very slow because it does not rely on incremental score calculation.
- FAST_ASSERT mode turns on most assertions, such as the assertions that an undoMove's score is the same as before the Move, to fail-fast on a bug in a Move implementation, a constraint, the engine itself, and so on. This mode is reproducible. It is also intrusive because it calls the method calculateScore() more frequently than a non-assert mode. The FAST_ASSERT mode is slow. Write a test case that does a short run of your planning problem with the FAST_ASSERT mode on.
- **REPRODUCIBLE** mode is the default mode because it is recommended during development. In this mode, two runs in the same OptaPlanner version execute the same code in the same order. Those two runs have the same result at every step, except if the following note applies. This enables you to reproduce bugs consistently. It also enables you to benchmark certain refactorings, such as a score constraint performance optimization, fairly across runs.



NOTE

Despite using **REPRODCIBLE** mode, your application might still not be fully reproducible for the following reasons:

- Use of **HashSet** or another **Collection** which has an inconsistent order between JVM runs for collections of planning entities or planning values but not normal problem facts, especially in the solution implementation. Replace it with **LinkedHashSet**.
- Combining a time gradient dependent algorithm, most notably the Simulated Annealing algorithm, together with time spent termination. A sufficiently large difference in allocated CPU time will influence the time gradient values. Replace the Simulated Annealing algorithms with the Late Acceptance algorithm, or replace time spent termination with step count termination.
- **REPRODUCIBLE** mode can be slightly slower than **NON_REPRODUCIBLE** mode. If your production environment can benefit from reproducibility, use this mode in production. In practice, **REPRODUCIBLE** mode uses the default fixed random seed if no seed is specified and it also disables certain concurrency optimizations such as work stealing.
- NON_REPRODUCIBLE mode can be slightly faster than REPRODUCIBLE mode. Avoid using it during development because it makes debugging and bug fixing difficult. If reproducibility isn't important in your production environment, use NON_REPRODUCIBLE mode in production. In practice, this mode uses no fixed random seed if no seed is specified.

7.3. CHANGING THE OPTAPLANNER SOLVER LOGGING LEVEL

You can change the logging level in an OptaPlanner solver to review solver activity. The following list describes the different logging levels:

- error: Logs errors, except those that are thrown to the calling code as a RuntimeException. If an error occurs, OptaPlanner normally fails fast. It throws a subclass of RuntimeException with a detailed message to the calling code. To avoid duplicate log messages, it does not log it as an error. Unless the calling code explicitly catches and eliminates that RuntimeException, a Thread's default `ExceptionHandler will log it as an error anyway. Meanwhile, the code is disrupted from doing further harm or obfuscating the error.
- warn: Logs suspicious circumstances
- info: Logs every phase and the solver itself
- debug: Logs every step of every phase
- trace: Logs every move of every step of every phase



Specifying **trace** logging will slow down performance considerably. However, **trace** logging is invaluable during development to discover a bottleneck.

Even **debug** logging can slow down performance considerably for fast stepping algorithms such as Late Acceptance and Simulated Annealing, but not for slow stepping algorithms such as Tabu Search.

Both trace` and **debug** logging cause congestion in multithreaded solving with most appenders.

In Eclipse, **debug** logging to the console tends to cause congestion with score calculation speeds above 10000 per second. Neither IntelliJ or the Maven command line suffer from this problem.

Procedure

Set the logging level to **debug** logging to see when the phases end and how fast steps are taken.

The following example shows output from debug logging:

INFO Solving started: time spent (3), best score (-4init/0), random (JDK with seed 0). DEBUG CH step (0), time spent (5), score (-3init/0), selected move count (1), picked move (Queen-2 {null -> Row-0}). DEBUG CH step (1), time spent (7), score (-2init/0), selected move count (3), picked move (Queen-1 { $null \rightarrow Row-2$ }). DEBUG CH step (2), time spent (10), score (-1init/0), selected move count (4), picked move (Queen-3 {null -> Row-3}). DEBUG CH step (3), time spent (12), score (-1), selected move count (4), picked move (Queen-0 {null -> Row-1}). INFO Construction Heuristic phase (0) ended: time spent (12), best score (-1), score calculation speed (9000/sec), step total (4). LS step (0), time spent (19), score (-1), DEBUG best score (-1), accepted/selected move count (12/12), picked move (Queen-1 {Row-2 -> Row-3}). LS step (1), time spent (24), score (0), new best score (0), accepted/selected move count DEBUG (9/12), picked move (Queen-3 {Row-3 -> Row-2}). INFO Local Search phase (1) ended: time spent (24), best score (0), score calculation speed (4000/sec), step total (2). INFO Solving ended: time spent (24), best score (0), score calculation speed (7000/sec), phase total (2), environment mode (REPRODUCIBLE).

All time spent values are in milliseconds.

Everything is logged to SLF4J, which is a simple logging facade that delegates every log message to Logback, Apache Commons Logging, Log4j, or java.util.logging. Add a dependency to the logging adaptor for your logging framework of choice.

7.4. USING LOGBACK TO LOG OPTAPLANNER SOLVER ACTIVITY

Logback is the recommended logging frameworkd to use with OptaPlanner. Use Logback to log OptaPlanner solver activity.

Prerequisites

• You have an OptaPlanner project.

Procedure

1. Add the following Maven dependency to your OptaPlanner project's **pom.xml** file:



NOTE

You do not need to add an extra bridge dependency.

<dependency> <groupId>ch.qos.logback</groupId> <artifactId>logback-classic</artifactId> <version>1.x</version> </dependency>

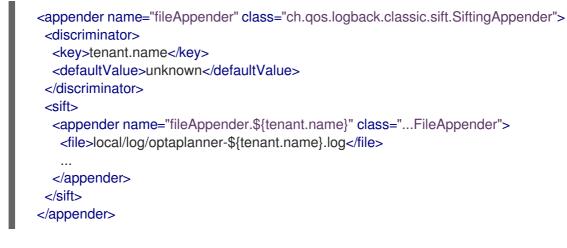
 Configure the logging level on the org.optaplanner package in your logback.xml file as shown in the following example where <LEVEL> is a logging level listed in Section 7.4, "Using Logback to log OptaPlanner solver activity".

<configuration></configuration>
<logger level="<LEVEL>" name="org.optaplanner"></logger>

- 3. Optional: If you have a multitenant application where multiple **Solver** instances might be running at the same time, separate the logging of each instance into separate files:
 - a. Surround the **solve()** call with Mapped Diagnostic Context (MDC):

MDC.put("tenant.name",tenantName); MySolution bestSolution = solver.solve(problem); MDC.remove("tenant.name");

b. Configure your logger to use different files for each **\${tenant.name}**. For example, use a **SiftingAppender** in the **logback.xml** file:





When running multiple solvers or one multithreaded solve, most appenders, including the console, cause congestion with **debug** and **trace** logging. Switch to an async appender to avoid this problem or turn off **debug** logging.

4. If OptaPlanner doesn't recognize the new level, temporarily add the system property - **Dlogback**.*LEVEL*=true to troubleshoot.

7.5. USING LOG4J TO LOG OPTAPLANNER SOLVER ACTIVITY

If you are already using Log4J and you do not want to switch to its faster successor, Logback, you can configure your OptaPlanner project for Log4J.

Prerequisites

- You have an OptaPlanner project
- You are using the Log4J logging framework

Procedure

1. Add the bridge dependency to the project **pom.xml** file:

<dependency> <groupId>org.slf4j</groupId> <artifactId>slf4j-log4j12</artifactId> <version>1.x</version> </dependency>

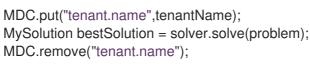
2. Configure the logging level on the package **org.optaplanner** in your **log4j.xml** file as shown in the following example, where **<LEVEL>** is a logging level listed in Section 7.4, "Using Logback to log OptaPlanner solver activity".

<log4j:configuration xmlns:log4j="http://jakarta.apache.org/log4j/">

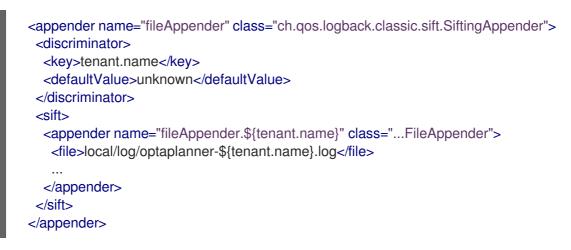
```
<category name="org.optaplanner">
<priority value="<LEVEL>" />
</category>
```

</log4j:configuration>

- 3. Optional: If you have a multitenant application where multiple **Solver** instances might be running at the same time, separate the logging of each instance into separate files:
 - a. Surround the **solve()** call with Mapped Diagnostic Context (MDC):



b. Configure your logger to use different files for each **\${tenant.name}**. For example, use a **SiftingAppender** in the **logback.xml** file:





NOTE

When running multiple solvers or one multithreaded solve, most appenders, including the console, cause congestion with **debug** and **trace** logging. Switch to an async appender to avoid this problem or turn off **debug** logging.

7.6. MONITORING THE SOLVER

OptaPlanner exposes metrics through Micrometer, a metrics instrumentation library for Java applications. You can use Micrometer with popular monitoring systems to monitor the OptaPlanner solver.

7.6.1. Configuring a Quarkus OptaPlanner application for Micrometer

To configure your OptaPlanner Quarkus application to use Micrometer and a specified monitoring system, add the Micrometer dependency to the **pom.xml** file.

Prerequisites

• You have a Quarkus OptaPlanner application.

Procedure

Add the following dependency to your application's **pom.xml** file where
 <MONITORING_SYSTEM> is a monitoring system supported by Micrometer and Quarkus:



NOTE

Prometheus is currently the only monitoring system supported by Quarkus.

<dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-micrometer-registry-<MONITORING_SYSTEM></artifactId> </dependency>

2. To run the application in development mode, enter the following command:



3. To view metrics for your application, enter the following URL in a browser:

http://localhost:8080/q/metrics

7.6.2. Configuring a Spring Boot OptaPlanner application for Micrometer

To configure your Spring Boot OptaPlanner application to use Micrometer and a specified monitoring system, add the Micrometer dependency to the **pom.xml** file.

Prerequisites

• You have a Spring Boot OptaPlanner application.

Procedure

- Add the following dependency to your application's **pom.xml** file where
 <MONITORING_SYSTEM> is a monitoring system supported by Micrometer and Spring Boot:
 - <dependency> <groupId>org.springframework.boot</groupId> <artifactId>spring-boot-starter-actuator</artifactId> </dependency> <dependency> <groupId>io.micrometer</groupId> <artifactId>micrometer-registry-<MONITORING_SYSTEM></artifactId> </dependency>
- 2. Add configuration information to the application's **application.properties** file. For information, see the Micrometer web site.
- 3. To run the application, enter the following command:

mvn spring-boot:run

4. To view metrics for your application, enter the following URL in a browser: http://localhost:8080/actuator/metrics



Use the following URL as the Prometheus scraper path: http://localhost:8080/actuator/prometheus

7.6.3. Configuring a plain Java OptaPlanner application for Micrometer

To configuring a plain Java OptaPlanner application to use Micrometer, you must add Micrometer dependencies and configuration information for your chosen monitoring system to your project's **POM.XML** file.

Prerequisites

• You have a plain Java OptaPlanner application.

Procedure

Add the following dependencies to your application's **pom.xml** file where
 <MONITORING_SYSTEM> is a monitoring system that is configured with Micrometer and
 <VERSION> is the version of Micrometer that you are using:

<dependency> <groupid>io.micrometer</groupid></dependency>
<pre><artifactid>micrometer-registry-<monitoring_system></monitoring_system></artifactid> <version><version></version></version></pre>
<dependency></dependency>
<groupid>io.micrometer</groupid>
<artifactid>micrometer-core</artifactid> <version><version></version></version>

- 2. Add Micrometer configuration information for your monitoring system to the beginning of your project's **pom.xml** file. For information, see the Micrometer web site.
- 3. Add the following line below the configuration information, where **<MONITORING_SYSTEM>** is the monitoring system that you added:

Metrics.addRegistry(<MONITORING_SYSTEM>);

The following example shows how to add the Prometheus monitoring system:

```
PrometheusMeterRegistry prometheusRegistry = new
PrometheusMeterRegistry(PrometheusConfig.DEFAULT);
try {
    HttpServer server = HttpServer.create(new InetSocketAddress(8080), 0);
    server.createContext("/prometheus", httpExchange -> {
        String response = prometheusRegistry.scrape();
        httpExchange.sendResponseHeaders(200, response.getBytes().length);
        try (OutputStream os = httpExchange.getResponseBody()) {
            os.write(response.getBytes());
        }
    });
    new Thread(server::start).start();
} catch (IOException e) {
```

throw new RuntimeException(e);

Metrics.addRegistry(prometheusRegistry);

4. Open your monitoring system to view the metrics for your OptaPlanner project. The following metrics are exposed:



NOTE

The names and format of the metrics vary depending on the registry.

- **optaplanner.solver.errors.total**: the total number of errors that occurred while solving since the start of the measuring.
- optaplanner.solver.solve-length.active-count: the number of solvers currently solving.
- **optaplanner.solver.solve-length.seconds-max**: run time of the longest-running currently active solver.
- **optaplanner.solver.solve-length.seconds-duration-sum**: the sum of each active solver's solve duration. For example, if there are two active solvers, one running for three minutes and the other for one minute, the total solve time is four minutes.

7.6.4. Additional Metrics

For more detailed monitoring, you can configure OptaPlanner in the solver configuration to monitor additional metrics at a performance cost. The following example uses the **BEST_SCORE** and **SCORE_CALCULATION_COUNT** metric :

You can enable the following metrics in this configuration:

- **SOLVE_DURATION** (enabled by default, Micrometer meter ID: **optaplanner.solver.solve.duration**): Measures the duration of solving for the longest active solver, the number of active solvers, and the cumulative duration of all active solvers.
- **ERROR_COUNT** (enabled by default, Micrometer meter ID: **optaplanner.solver.errors**): Measures the number of errors that occurred while solving.
- SCORE_CALCULATION_COUNT (enabled by default, Micrometer meter ID: optaplanner.solver.score.calculation.count): Measures the number of score calculations the OptaPlanner performed.
- BEST_SCORE (Micrometer meter ID: optaplanner.solver.best.score.*): Measures the score

of the best solution that OptaPlanner has found so far. There are separate meters for each level of the score. For instance, for a **HardSoftScore**, there are **optaplanner.solver.best.score.hard.score** and **optaplanner.solver.best.score.soft.score** meters.

- STEP_SCORE (Micrometer meter ID: optaplanner.solver.step.score.*): Measures the score of each step that OptaPlanner takes. There are separate meters for each level of the score. For instance, for a HardSoftScore, there are optaplanner.solver.step.score.hard.score and optaplanner.solver.step.score.soft.score meters.
- **BEST_SOLUTION_MUTATION** (Micrometer meter ID: **optaplanner.solver.best.solution.mutation**): Measures the number of changed planning variables between consecutive best solutions.
- **MOVE_COUNT_PER_STEP** (Micrometer meter ID: **optaplanner.solver.step.move.count**): Measures the number of moves evaluated in a step.
- **MEMORY_USE** (Micrometer meter ID: **jvm.memory.used**): Measures the amount of memory used across the JVM. This metric does not measure the amount of memory used by a solver; two solvers on the same JVM will report the same value for this metric.
- CONSTRAINT_MATCH_TOTAL_BEST_SCORE (Micrometer meter ID: optaplanner.solver.constraint.match.best.score.*): Measures the score impact of each constraint on the best solution that OptaPlanner has found so far. There are separate meters for each level of the score, with tags for each constraint. For instance, for a HardSoftScore for a constraint "Minimize Cost" in package "com.example", there are optaplanner.solver.constraint.match.best.score.hard.score and optaplanner.solver.constraint.match.best.score.soft.score meters with tags "constraint.package=com.example" and "constraint.name=Minimize Cost".
- **CONSTRAINT_MATCH_TOTAL_STEP_SCORE** (Micrometer meter ID: optaplanner.solver.constraint.match.step.score.*): Measures the score impact of each constraint on the current step. There are separate meters for each level of the score, with tags for each constraint. For instance, for a **HardSoftScore** for a constraint "Minimize Cost" in package "com.example", there are

optaplanner.solver.constraint.match.step.score.hard.score and optaplanner.solver.constraint.match.step.score.soft.score meters with tags "constraint.package=com.example" and "constraint.name=Minimize Cost".

- PICKED_MOVE_TYPE_BEST_SCORE_DIFF (Micrometer meter ID: optaplanner.solver.move.type.best.score.diff.*): Measures how much a particular move type improves the best solution. There are separate meters for each level of the score, with a tag for the move type. For instance, for a HardSoftScore and a ChangeMove for the computer of a process, there are optaplanner.solver.move.type.best.score.diff.hard.score and optaplanner.solver.move.type.best.score.diff.soft.score meters with the tag move.type=ChangeMove(Process.computer).
- PICKED_MOVE_TYPE_STEP_SCORE_DIFF (Micrometer meter ID: optaplanner.solver.move.type.step.score.diff.*): Measures how much a particular move type improves the best solution. There are separate meters for each level of the score, with a tag for the move type. For instance, for a HardSoftScore and a ChangeMove for the computer of a process, there are optaplanner.solver.move.type.step.score.diff.hard.score and optaplanner.solver.move.type.step.score.diff.soft.score meters with the tag move.type=ChangeMove(Process.computer).

7.7. CONFIGURING THE RANDOM NUMBER GENERATOR

Many heuristics and metaheuristics depend on a pseudorandom number generator for move selection, to resolve score ties, probability based move acceptance, and so on. During solving, the same random instance is reused to improve reproducibility, performance, and uniform distribution of random values.

A random seed is a number used to initialize a pseudorandom number generator.

Procedure

1. Optional: To change the random seed of a random instance, specify a **randomSeed**:

<solver xmlns="https://www.optaplanner.org/xsd/solver" xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance" xsi:schemaLocation="https://www.optaplanner.org/xsd/solver https://www.optaplanner.org/xsd/solver/solver.xsd"> <randomSeed>0</randomSeed> ... </solver>

 Optional: To change the pseudorandom number generator implementation, specify a value for the randomType property listed in the solver configuration file below, where <RANDOM_NUMBER_GENERATOR> is a pseudorandom number generator:

<solver xmlns="https://www.optaplanner.org/xsd/solver" xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance" xsi:schemaLocation="https://www.optaplanner.org/xsd/solver https://www.optaplanner.org/xsd/solver/solver.xsd"> <randomType><RANDOM_NUMBER_GENERATOR></randomType> ... </solver>

The following pseudorandom number generators are supported:

- JDK (default): Standard random number generator implementation (java.util.Random)
- **MERSENNE_TWISTER**: Random number generator implementation by Commons Math
- WELL512A, WELL1024A, WELL19937A, WELL19937C, WELL44497A and WELL44497B: Random number generator implementation by Commons Math

For most use cases, the value of the **randomType** property has no significant impact on the average quality of the best solution on multiple data sets.

CHAPTER 8. THE OPTAPLANNER SOLVERMANAGER

A **SolverManager** is a facade for one or more **Solver** instances to simplify solving planning problems in REST and other enterprise services.

Unlike the Solver.solve(...) method, a SolverManager has the following characteristics:

- **SolverManager.solve(...)** returns immediately: it schedules a problem for asynchronous solving without blocking the calling thread. This avoids timeout issues of HTTP and other technologies.
- SolverManager.solve(...) solves multiple planning problems of the same domain, in parallel.

Internally, a **SolverManager** manages a thread pool of solver threads, which call **Solver.solve(...)**, and a thread pool of consumer threads, which handle best solution changed events.

In Quarkus and Spring Boot, the **SolverManager** instance is automatically injected in your code. If you are using a platform other than Quarkus or Spring Boot, build a **SolverManager** instance with the **create(...)** method:

SolverConfig solverConfig = SolverConfig.createFromXmlResource(".../cloudBalancingSolverConfig.xml"); SolverManager<CloudBalance, UUID> solverManager = SolverManager.create(solverConfig, new SolverManagerConfig());

Each problem submitted to the **SolverManager.solve(...)** methods must have a unique problem ID. Later calls to **getSolverStatus(problemId)** or **terminateEarly(problemId)** use that problem ID to distinguish between planning problems. The problem ID must be an immutable class, such as **Long**, **String**, or **java.util.UUID**.

The **SolverManagerConfig** class has a **parallelSolverCount** property that controls how many solvers are run in parallel. For example, if the **parallelSolverCount** property` is set to **4** and you submit five problems, four problems start solving immediately and the fifth problem starts when one of the first problems ends. If those problems solve for five minutes each, the fifth problem takes 10 minutes to finish. By default, **parallelSolverCount** is set to **AUTO**, which resolves to half the CPU cores, regardless of the **moveThreadCount** of the solvers.

To retrieve the best solution, after solving terminates normally use SolverJob.getFinalBestSolution():

```
CloudBalance problem1 = ...;

UUID problemId = UUID.randomUUID();

// Returns immediately

SolverJob<CloudBalance, UUID> solverJob = solverManager.solve(problemId, problem1);

...

CloudBalance solution1;

try {

    // Returns only after solving terminates

    solution1 = solverJob.getFinalBestSolution();

} catch (InterruptedException | ExecutionException e) {

    throw ...;

}
```

However, there are better approaches, both for solving batch problems before a user needs the solution as well as for live solving while a user is actively waiting for the solution.

The current **SolverManager** implementation runs on a single computer node, but future work aims to distribute solver loads across a cloud.

8.1. BATCH SOLVING PROBLEMS

Batch solving is solving multiple data sets in parallel. Batch solving is particularly useful overnight:

- There are typically few or no problem changes in the middle of the night. Some organizations enforce a deadline, for example, *submit all day off requests before midnight*.
- The solvers can run for much longer, often hours, because nobody is waiting for the results and CPU resources are often cheaper.
- Solutions are available when employees arrive at work the next working day.

Procedure

To batch solve problems in parallel, limited by **parallelSolverCount**, call **solve(...)** for each data set created the following class:

8.2. SOLVE AND LISTEN TO SHOW PROGRESS

When a solver is running while a user is waiting for a solution, the user might need to wait for several minutes or hours before receiving a result. To assure the user that everything is going well, show progress by displaying the best solution and best score attained so far.

Procedure

1. To handle intermediate best solutions, use **solveAndListen(...)**:

public class TimeTableService {

private SolverManager<TimeTable, Long> solverManager;

// Returns immediately
public void solveLive(Long timeTableId) {



This implementation is using the database to communicate with the UI, which polls the database. More advanced implementations push the best solutions directly to the UI or a messaging queue.

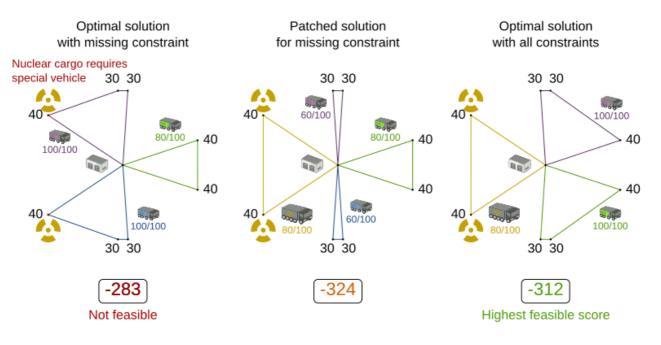
2. When the user is satisfied with the intermediate best solution and does not want to wait any longer for a better one, call **SolverManager.terminateEarly(problemId)**.

PART IV. OPTAPLANNER SCORE CALCULATION

Every **@PlanningSolution** class has a score. The score is an objective way to compare two solutions. The solution with the higher score is better. The solver aims to find the solution with the highest score out of all possible solutions. The *best solution* is the solution with the highest score the solver has encountered during solving, which might be the *optimal solution*.

OptaPlanner cannot automatically know which solution is best for your business, so you must tell it how to calculate the score of a specified **@PlanningSolution** instance according to your business requirements. If you forget or are unable to implement an important business constraint, the solution is probably useless, as illustrated in the following image:

Optimal with incomplete constraints



The optimal solution for a problem that misses a constraint is probably useless.

Note Pinned entities can sometimes offer a temporary workaround for an end-user.

CHAPTER 9. BUSINESS CONSTRAINTS IN OPTAPLANNER

Business constraints are used to limit conditions within a scenario. The conditions might be based on existing business contracts, resource availability, employee preferences, or business rules. To implement a business constraint in OptaPlanner, the business constraint must be formalized as a score constraint. The following score properties available in OptaPlanner provide flexible solutions:

- Score signum: make the constraint type positive or negative
- Score weight: put a cost or profit on a constraint type
- Score level (hard, soft, and so forth) prioritize a group of constraint types

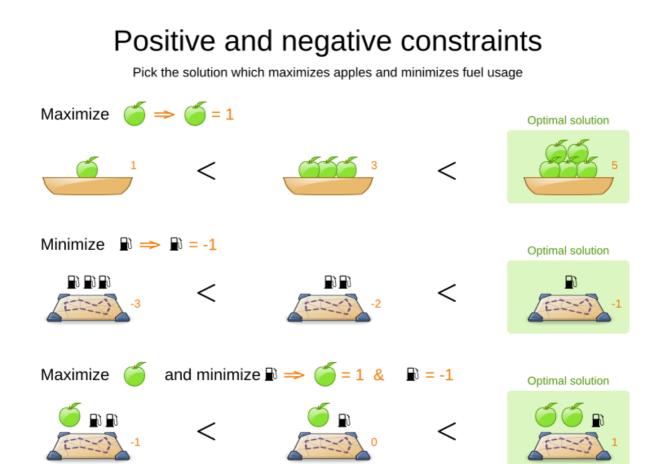


NOTE

Do not presume that your business knows all of its score constraints in advance. Expect score constraints to be added, changed, or removed after the first releases.

9.1. NEGATIVE AND POSITIVE SCORE CONSTRAINTS

All score techniques are based on constraints. A constraint can be a simple pattern such as *Maximize the apple harvest in the solution* or a more complex pattern. A constraint is either negative or positive. A positive constraint is a constraint that you want to maximize. A negative constraint is a constraint that you want to minimize.



This image illustrates that the optimal solution always has the highest score, regardless of whether the constraints are positive or negative.

Most planning problems have only negative constraints and therefore have a negative score. In this case, the score is the sum of the weight of the negative constraints being broken, with a perfect score of 0. For example, in the N Queens problem, the score is the negative of the number of queen pairs which can attack each other. You can combine negative and positive constraints, even in the same score level.

When a constraint activates on a certain planning entity set because the negative constraint is broken or the positive constraint is fulfilled, it is called a *constraint match*.

9.2. SCORE CONSTRAINT WEIGHT

Not all score constraints are equally important. If breaking one constraint once is equally as bad as breaking another constraint multiple times, then those two constraints have different weights even though they are in the same score level.

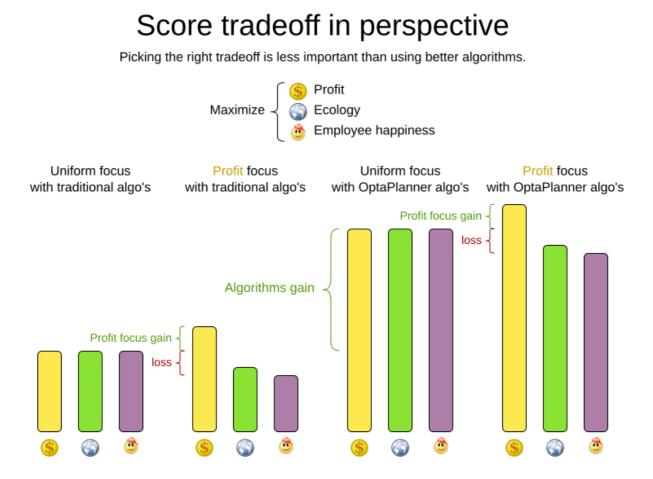
Score weighting is easy in use cases where you can assign a cost to everything. In this case, the positive constraints maximize revenue and the negative constraints minimize expenses and together they maximize profit.

Alternatively, score weighting is also often used to create social fairness. For example, a nurse who requests a free day pays a higher weight on New Years Eve than on a normal day.

The weight of a constraint match can depend on the planning entities involved. For example, in the cloud balancing problem the weight of the soft constraint match for an active computer is the maintenance cost of that computer and this is different for each computer.

Putting a good weight on a constraint is often a difficult analytical decision because it is about making choices and trade-offs against other constraints. Different stakeholders have different priorities.

Do not waste time with constraint weight discussions at the start of an implementation. Instead, add a @constraintConfiguration annotation and allow users to change them through a UI. An inaccurate weight is less damaging than mediocre algorithms as shown in the following illustration:

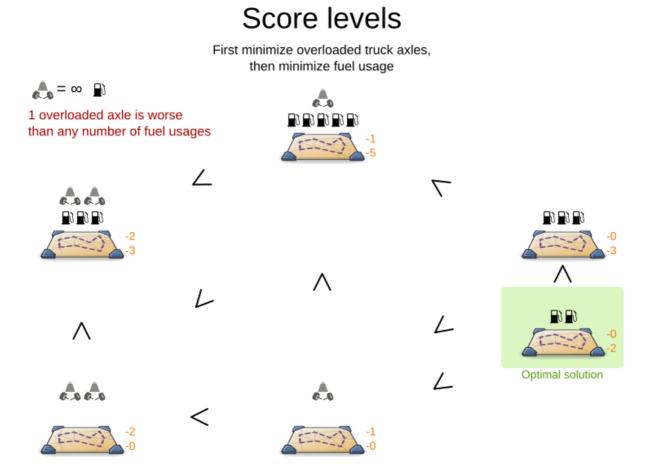


Most use cases use a score with int weights, such as HardSoftScore.

9.3. SCORE CONSTRAINT LEVEL

Sometimes a score constraint outranks another score constraint, no matter how many times the latter is broken. In this case, those score constraints are in different levels. For example, a nurse cannot work two shifts at the same time because of the constraints of physical reality, so this constraint outranks all nurse happiness constraints.

Most use cases have only two score levels, hard and soft. The levels of two scores are compared in order. The first score level is compared first. If the two scores differ, the remaining score levels are ignored. For example, a score that breaks **0** hard constraints and **1000000** soft constraints is better than a score that breaks **1** hard constraint and **0** soft constraints.



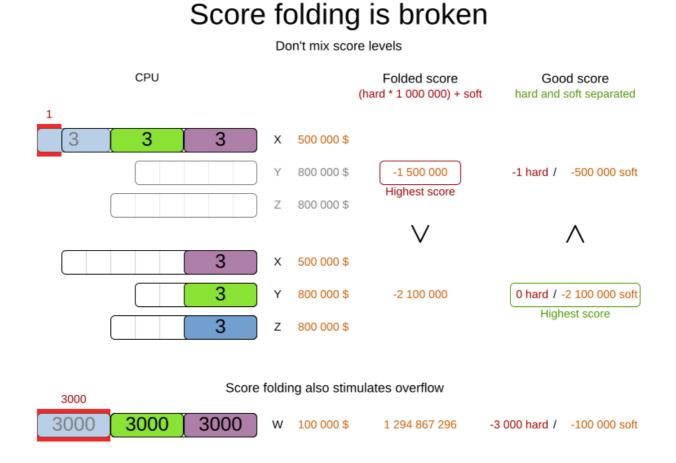
If there are two or more score levels, then a score is *feasible* if no hard constraints are broken.



NOTE

By default, OptaPlanner assigns all planning variables a planning value. If there is no feasible solution, this means that the best solution is infeasible. To leave some of the planning entities unassigned, apply overconstrained planning.

For each constraint, you must pick a score level, a score weight, and a score signum. For example, **-1soft** has a score level of **soft**, a weight of **1**, and a negative signum. Do not use a large constraint weight when your business actually wants different score levels. That workaround, known as *score folding*, is broken:



Your business might tell you that your hard constraints all have the same weight because they cannot be broken and therefore the weight does not matter. This is not true. If no feasible solution exists for a specific data set, the business can use the least infeasible solution to estimate how many business resources they are lacking. For example, in the cloud balancing problem, the least infeasible solution can reveal how many new computers are needed.

Furthermore, if all of your hard constraints have the same weight, you will likely create a score trap. For example, in the cloud balancing problem if a computer has seven too few CPUs for its processes, then it must be weighted seven times as much as if it had only one CPU too few.

OptaPlanner supports three or more score levels as well. For example, a company might decide that profit outranks employee satisfaction, or vice versa, while both constraints are outranked by the constraints of physical reality.



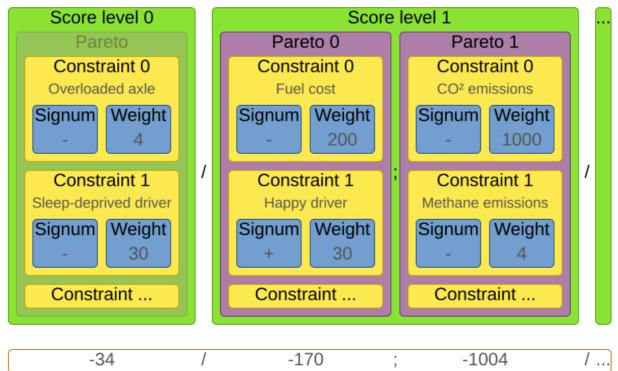
NOTE

To model fairness or load balancing, you do not need to use lots of score levels even though OptaPlanner can handle many score levels.

Most use cases use a score with two or three weights, such as **HardSoftScore** and **HardMediumSoftScore**. You can combine all of these techniques seemlessly:

Score composition

How are the score techniques combined?



Score for 1 solution

CHAPTER 10. THE OPTAPLANNER SCORE INTERFACE

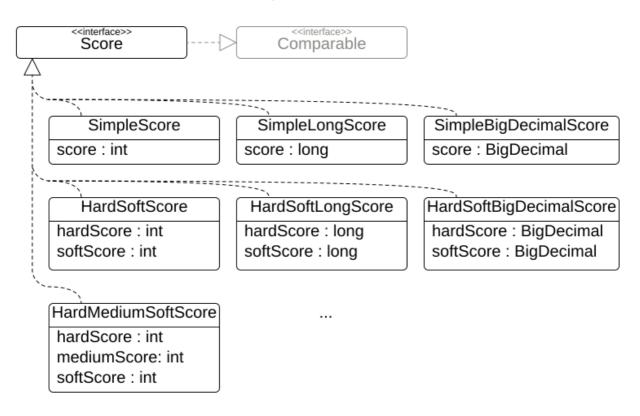
A score is represented by the **Score** interface, which extends the **Comparable** interface:

```
public interface Score<...> extends Comparable<...> {
    ...
}
```

The score implementation to use depends on your use case. Your score might not efficiently fit in a single **long** value. OptaPlanner has several built-in score implementations, but you can implement a custom score as well. Most use cases use the built-in **HardSoftScore** score.

Score class diagram

Choose a Score implementation or write a custom one



All Score implementations also have an **initScore** (which is an **int**). It is mostly intended for internal use in OptaPlanner: it is the negative number of uninitialized planning variables. From a user's perspective, this is **0**, unless a construction heuristic is terminated before it could initialize all planning variables. In this case, **Score.isSolutionInitialized()** returns **false**.

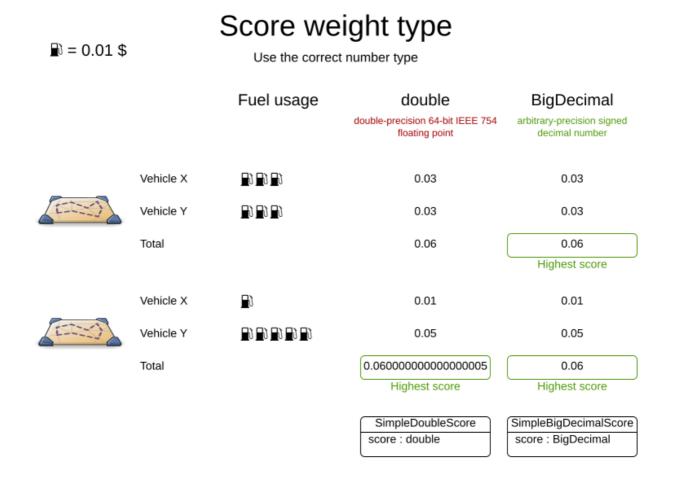
The score implementation (for example **HardSoftScore**) must be the same throughout a solver runtime. The score implementation is configured in the solution domain class:

```
@PlanningSolution
public class CloudBalance {
    ...
    @PlanningScore
```

private HardSoftScore score;

10.1. FLOATING POINT NUMBERS IN SCORE CALCULATION

Avoid the use of the floating point number types **float** or **double** in score calculation. Use **BigDecimal** or scaled **long** instead. Floating point numbers cannot represent a decimal number correctly. For example, a **double** cannot contain the value **0.05** correctly. Instead, it contains the nearest representable value. Arithmetic, including addition and subtraction, that uses floating point numbers, especially for planning problems, leads to incorrect decisions as shown in the following illustration:



Additionally, floating point number addition is not associative:

System.out.println(((0.01 + 0.02) + 0.03) == (0.01 + (0.02 + 0.03))); // returns false

This leads to score corruption.

Decimal numbers (BigDecimal) have none of these problems.



BigDecimal arithmetic is considerably slower than **int**, **long**, or **double** arithmetic. In some experiments, the score calculation takes five times longer.

Therefore, in many cases, it can be worthwhile to multiply *all* numbers for a single score weight by a plural of ten, so the score weight fits in a scaled **int** or **long**. For example, if you multiply all weights by **1000**, a fuelCost of **0.07** becomes a **fuelCostMillis** of **70** and no longer uses a decimal score weight.

10.2. SCORE CALCULATION TYPES

There are several types of ways to calculate the score of a solution:

- Easy Java score calculation Implement all constraints together in a single method in Java or another JVM language. This method does not scale.
- **Constraint streams score calculation** Implement each constraint as a separate constraint stream in Java or another JVM language. This method is fast and scalable.
- Incremental Java score calculation (not recommended): Implement multiple low-level methods in Java or another JVM language. This method is fast and scalable but very difficult to implement and maintain.
- **Drools score calculation (deprecated)**: Implement each constraint as a separate score rule in DRL. This method is scalable.

Each score calculation type can work with any score definition, for example **HardSoftScore** or **HardMediumSoftScore**. All score calculation types are object oriented and can reuse existing Java code.



IMPORTANT

The score calculation must be read-only. It must not change the planning entities or the problem facts in any way. For example, the score calculation must not call a setter method on a planning entity in the score calculation.

OptaPlanner does not recalculate the score of a solution if it can predict it unless an **environmentMode** assertion is enabled. For example, after a winning step is done, there is no need to calculate the score because that move was done and undone earlier. As a result, there is no guarantee that changes applied during score calculation actually happen.

To update planning entities when the planning variable changes, use shadow variables instead.

10.2.1. Implenting the Easy Java score calculation type

The Easy Java score calculation type provides an easy way to implement your score calculation in Java. You can implement all constraints together in a single method in Java or another JVM language.

- Advantages:
 - Uses plain old Java so there is no learning curve

- Provides an opportunity to delegate score calculation to an existing code base or legacy system
- Disadvantages:
 - Slowest calculation type
 - Does not scale because there is no incremental score calculation

Procedure

1. Implement the **EasyScoreCalculator** interface:

```
public interface EasyScoreCalculator<Solution_, Score_ extends Score<Score_>>> {
    Score_ calculateScore(Solution_ solution);
}
```

The following example implements this interface in the N Queens problem:

```
public class NQueensEasyScoreCalculator
  implements EasyScoreCalculator<NQueens, SimpleScore> {
  @Override
  public SimpleScore calculateScore(NQueens nQueens) {
    int n = nQueens.getN();
     List<Queen> queenList = nQueens.getQueenList();
    int score = 0;
    for (int i = 0; i < n; i++) {
       for (int j = i + 1; j < n; j++) {
         Queen leftQueen = queenList.get(i);
         Queen rightQueen = queenList.get(j);
         if (leftQueen.getRow() != null && rightQueen.getRow() != null) {
            if (leftQueen.getRowIndex() == rightQueen.getRowIndex()) {
              score--;
            }
            if (leftQueen.getAscendingDiagonalIndex() ==
rightQueen.getAscendingDiagonalIndex()) {
              score--;
            }
            if (leftQueen.getDescendingDiagonalIndex() ==
rightQueen.getDescendingDiagonalIndex()) {
              score--;
            }
         }
       }
    }
     return SimpleScore.valueOf(score);
  }
```

2. Configure the **EasyScoreCalculator** class in the solver configuration. The following example shows how to implement this interface in the N Queens problem:

<scoreDirectorFactory>

<easyScoreCalculatorClass>org.optaplanner.examples.nqueens.optional.score.NQueensEasy ScoreCalculator</easyScoreCalculatorClass> </scoreDirectorFactory>

3. To configure values of the **EasyScoreCalculator** method dynamically in the solver configuration so that the benchmarker can tweak those parameters, add the **easyScoreCalculatorCustomProperties** element and use custom properties:

<scoreDirectorFactory> <easyScoreCalculatorClass>...MyEasyScoreCalculator</easyScoreCalculatorClass> <easyScoreCalculatorCustomProperties> <property name="myCacheSize" value="1000" /> </easyScoreCalculatorCustomProperties> </scoreDirectorFactory>

10.2.2. Implementing the Incremental Java score calculation type

The Incremental Java score calculation type provides a way to implement your score calculation incrementally in Java.



NOTE

This type is not recommended.

- Advantages:
 - Very fast and scalable. This is currently the fastest type if implemented correctly.
- Disadvantages:
 - Hard to write.
 - A scalable implementation that heavily uses maps, indexes, and so forth.
 - You have to learn, design, write, and improve all of these performance optimizations yourself.
 - Hard to read. Regular score constraint changes can lead to a high maintenance cost.

Procedure

1. Implement all of the methods of the **IncrementalScoreCalculator** interface:

public interface IncrementalScoreCalculator<Solution_, Score_ extends Score<Score_>> {

void resetWorkingSolution(Solution_ workingSolution);

void beforeEntityAdded(Object entity);

void afterEntityAdded(Object entity);

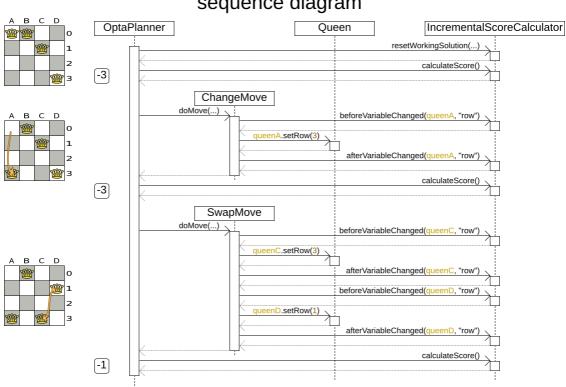
void beforeVariableChanged(Object entity, String variableName);

void afterVariableChanged(Object entity, String variableName);

void beforeEntityRemoved(Object entity);

void afterEntityRemoved(Object entity);

Score_calculateScore();



The following example implements this interface in the N Queens problem:

```
public class NQueensAdvancedIncrementalScoreCalculator
implements IncrementalScoreCalculator<NQueens, SimpleScore> {
    private Map<Integer, List<Queen>> rowIndexMap;
    private Map<Integer, List<Queen>> ascendingDiagonalIndexMap;
    private Map<Integer, List<Queen>> descendingDiagonalIndexMap;
    private int score;
    public void resetWorkingSolution(NQueens nQueens) {
        int n = nQueens.getN();
        rowIndexMap = new HashMap<Integer, List<Queen>>(n);
        ascendingDiagonalIndexMap = new HashMap<Integer, List<Queen>>(n * 2);
        descendingDiagonalIndexMap = new HashMap<Integer, List<Queen>>(n * 2);
        for (int i = 0; i < n; i++) {</pre>
```

IncrementalScoreCalculator sequence diagram

```
rowIndexMap.put(i, new ArrayList<Queen>(n));
       ascendingDiagonalIndexMap.put(i, new ArrayList<Queen>(n));
       descendingDiagonalIndexMap.put(i, new ArrayList<Queen>(n));
       if (i != 0) {
         ascendingDiagonalIndexMap.put(n - 1 + i, new ArrayList<Queen>(n));
         descendingDiagonalIndexMap.put((-i), new ArrayList<Queen>(n));
       }
    }
    score = 0;
    for (Queen queen : nQueens.getQueenList()) {
       insert(queen);
    }
  }
  public void beforeEntityAdded(Object entity) {
    // Do nothing
  }
  public void afterEntityAdded(Object entity) {
     insert((Queen) entity);
  }
  public void beforeVariableChanged(Object entity, String variableName) {
     retract((Queen) entity);
  }
  public void afterVariableChanged(Object entity, String variableName) {
     insert((Queen) entity);
  }
  public void beforeEntityRemoved(Object entity) {
     retract((Queen) entity);
  }
  public void afterEntityRemoved(Object entity) {
    // Do nothing
  }
  private void insert(Queen queen) {
     Row row = queen.getRow();
    if (row != null) {
       int rowIndex = queen.getRowIndex();
       List<Queen> rowIndexList = rowIndexMap.get(rowIndex);
       score -= rowIndexList.size();
       rowIndexList.add(queen);
       List<Queen> ascendingDiagonalIndexList =
ascendingDiagonalIndexMap.get(queen.getAscendingDiagonalIndex());
       score -= ascendingDiagonalIndexList.size();
       ascendingDiagonalIndexList.add(queen);
       List<Queen> descendingDiagonalIndexList =
descendingDiagonalIndexMap.get(queen.getDescendingDiagonalIndex());
       score -= descendingDiagonalIndexList.size();
       descendingDiagonalIndexList.add(queen);
    }
  }
```



2. Configure the **incrementalScoreCalculatorClass** class in the solver configuration. The following example shows how to implement this interface in the N Queens problem:

<scoreDirectorFactory>

<incrementalScoreCalculatorClass>org.optaplanner.examples.nqueens.optional.score.NQuee nsAdvancedIncrementalScoreCalculator</incrementalScoreCalculatorClass> </scoreDirectorFactory>



IMPORTANT

A piece of incremental score calculator code can be difficult to write and to review. Assert its correctness by using an **EasyScoreCalculator** to fulfill the assertions triggered by the **environmentMode**.

3. To configure values of an **IncrementalScoreCalculator** dynamically in the solver configuration so the benchmarker can tweak those parameters, add the **incrementalScoreCalculatorCustomProperties** element and use custom properties:

<scoreDirectorFactory>

<incrementalScoreCalculatorClass>...MyIncrementalScoreCalculator</incrementalScoreCalculatorClass>

<incrementalScoreCalculatorCustomProperties> <property name="myCacheSize" value="1000"/> </incrementalScoreCalculatorCustomProperties> </scoreDirectorFactory>

- 4. Optional: Implement the **ConstraintMatchAwareIncrementalScoreCalculator** interface to facilitate the following goals:
 - Explain a score by splitting it up for each score constraint with

ScoreExplanation.getConstraintMatchTotalMap().

- Visualize or sort planning entities by how many constraints each one breaks with **ScoreExplanation.getIndictmentMap()**.
- Receive a detailed analysis if the **IncrementalScoreCalculator** is corrupted in **FAST_ASSERT** or **FULL_ASSERT environmentMode**.

public interface ConstraintMatchAwareIncrementalScoreCalculator<Solution_, Score_
extends Score<Score_>> {

void resetWorkingSolution(Solution_ workingSolution, boolean constraintMatchEnabled);

Collection<ConstraintMatchTotal<Score_>> getConstraintMatchTotals();

Map<Object, Indictment<Score_>> getIndictmentMap();

}

For example, in machine reassignment create one **ConstraintMatchTotal** for each constraint type and call **addConstraintMatch()** for each constraint match:

```
public class MachineReassignmentIncrementalScoreCalculator
    implements
ConstraintMatchAwareIncrementalScoreCalculator<MachineReassignment,
HardSoftLongScore> {
  . . .
  @Override
  public void resetWorkingSolution(MachineReassignment workingSolution, boolean
constraintMatchEnabled) {
    resetWorkingSolution(workingSolution);
    // ignore constraintMatchEnabled, it is always presumed enabled
  }
  @Override
  public Collection<ConstraintMatchTotal<HardSoftLongScore>>
getConstraintMatchTotals() {
    ConstraintMatchTotal<HardSoftLongScore> maximumCapacityMatchTotal = new
DefaultConstraintMatchTotal<>(CONSTRAINT PACKAGE,
       "maximumCapacity", HardSoftLongScore.ZERO);
    for (MrMachineScorePart machineScorePart : machineScorePartMap.values()) {
      for (MrMachineCapacityScorePart machineCapacityScorePart :
machineScorePart.machineCapacityScorePartList) {
         if (machineCapacityScorePart.maximumAvailable < 0L) {
           maximumCapacityMatchTotal.addConstraintMatch(
                Arrays.asList(machineCapacityScorePart.machineCapacity),
HardSoftLongScore.valueOf(machineCapacityScorePart.maximumAvailable, 0));
         }
      }
    }
    List<ConstraintMatchTotal<HardSoftLongScore>> constraintMatchTotalList = new
ArrayList<>(4);
```

```
constraintMatchTotalList.add(maximumCapacityMatchTotal);
...
return constraintMatchTotalList;
}
@Override
public Map<Object, Indictment<HardSoftLongScore>> getIndictmentMap() {
    return null; // Calculate it non-incrementally from getConstraintMatchTotals()
}
```

The **getConstraintMatchTotals()** code often duplicates some of the logic of the normal **IncrementalScoreCalculator** methods. Constraint Streams and Drools Score Calculation do not have this disadvantage because they are constraint-match aware automatically when needed without any extra domain-specific code.

CHAPTER 11. THE INITIALIZINGSCORETREND CLASS

You can add the **InitializingScoreTrend** class to optimization algorithms to specify how the score changes when additional variables are initialized and the already-initialized variables do not change. Some optimization algorithms, such Construction Heuristics and Exhaustive Search, run faster when this information is available.

You can specify one of the following trends for the score or each score level separately:

- **ANY** (default): Initializing an extra variable can change the score positively or negatively. This trend does not provide a performance gain.
- **ONLY_UP** (rare): Initializing an extra variable can only change the score positively. The **ONLY_UP** trend requires the following conditions:
 - There are only positive constraints.
 - Initializing the next variable cannot unmatch a positive constraint that was matched by a previous initialized variable.
- **ONLY_DOWN**: Initializing an additional variable can only change the score negatively. **ONLY_DOWN** requires the following conditions:
 - There are only negative constraints.
 - Initializing the next variable cannot unmatch a negative constraint that was matched by a previous initialized variable.

Most use cases have only negative constraints. Many of those use cases have an **InitializingScoreTrend** class that only goes down, as shown in the following example:

<scoreDirectorFactory>

<constraintProviderClass>org.optaplanner.examples.cloudbalancing.score.CloudBalancingConstraintPr ovider</constraintProviderClass> <initializingScoreTrend>ONLY_DOWN</initializingScoreTrend> </scoreDirectorFactory>

Alternatively, you can also specify the trend for each score level separately, as shown in the following example:

<scoreDirectorFactory>

<constraintProviderClass>org.optaplanner.examples.cloudbalancing.score.CloudBalancingConstraintPr ovider</constraintProviderClass> <initializingScoreTrend>ONLY_DOWN/ONLY_DOWN</initializingScoreTrend>

</scoreDirectorFactory>

CHAPTER 12. INVALID SCORE DETECTION

If you use the **environmentMode** class and specify the value as **FULL_ASSERT** or **FAST_ASSERT**, the environment mode detects score corruption in the incremental score calculation.

However, doing this will not verify that your score calculator implements your score constraints the way that your business wants. For example, one constraint might consistently match the wrong pattern. To verify the constraints against an independent implementation, configure an **assertionScoreDirectorFactory** class:

<environmentMode>FAST_ASSERT</environmentMode>
...
<scoreDirectorFactory>
<constraintProviderClass>org.optaplanner.examples.nqueens.optional.score.NQueensConstraintProvid

er</constraintProviderClass> <assertionScoreDirectorFactory>

<easyScoreCalculatorClass>org.optaplanner.examples.nqueens.optional.score.NQueensEasyScoreCa lculator</easyScoreCalculatorClass> </assertionScoreDirectorFactory> </scoreDirectorFactory>

In this example, the **NQueensConstraintProvider** implementation is validated by the **EasyScoreCalculator**.



NOTE

This technique works well to isolate score corruption, but to verify that the constraint implements the real business needs, a unit test with a **ConstraintVerifier** is usually better.

CHAPTER 13. SCORE CALCULATION PERFORMANCE TRICKS

Most of the execution time of a solver involves running the score calculation, which is called in the solver's deepest loops. Faster score calculation returns the same solution in less time with the same algorithm. This usually provides a better solution in the same amount of time. Use the following techniques to improve your score calculation performance.

13.1. SCORE CALCULATION SPEED

When you are improving your score calculation, focus on maximizing the score calculation speed instead of maximizing the best score. A big improvement in score calculation can sometimes yield little or no best score improvement, for example when the algorithm is stuck in a local or global optima. If you are watching the calculation speed instead, score calculation improvements are far more visible.

The score calculation speed per second is a reliable measurement of score calculation performance, even though it is affected by non-score calculation execution time. The result depends on the problem scale of the problem data set. Normally, even for high scale problems, the score calculation speed per second is higher than **1000**, unless you are using an **EasyScoreCalculator** class.

By watching the calculation speed, you can remove or add score constraints and compare the latest calculation speed with the original calculation speed.



NOTE

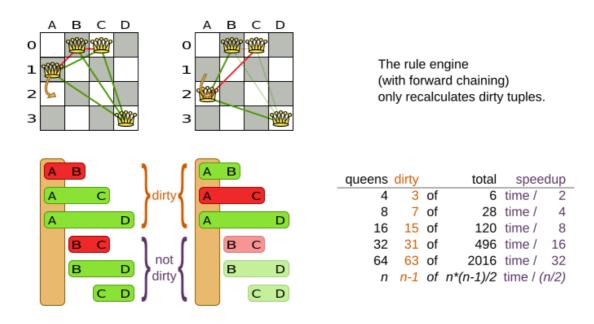
Comparing the best score with the original best score is pointless. It's like comparing apples and oranges.

13.2. INCREMENTAL SCORE CALCULATION

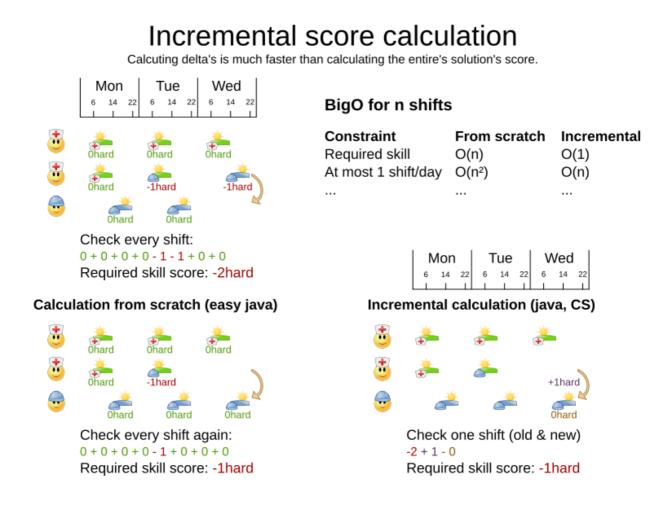
Incremental score calculation is also known as delta-based score calculation. When a solution changes, incremental score calculation finds the new score by evaluating changes between the current state and the previous state instead of recalculating the entire score on every solution evaluation. For example, in the N Queens problem, when queen A moves from row **1** to **2**, the **incrementalScoreCalculation** class does not check whether queen B and C can attack each other, because neither of them changed position, as shown in the following illustration:

Incremental score calculation

Incremental score calculation is much more scalable because only the delta is calculated.



The following example show incremental score calculation for employee rostering:



Incremental score calculation provides a significant performance and scalability gain. Constraint streams or Drools score calculation provides this scalability gain without forcing you to write a complicated incremental score calculation algorithm. Just let the rule engine do the hard work.

Notice that the increase in calculation speed is relative to the size of your planning problem (your *n*). This makes incremental score calculation scalable.

13.3. REMOTE SERVICES

Do not call remote services in your score calculation unless you are bridging the **EasyScoreCalculator** class to a legacy system. The network latency will significantly downgrade your score calculation performance. Cache the results of those remote services if possible.

If some parts of a constraint can be calculated once, when the solver starts, and never change during solving, then turn them into cached problem facts.

13.4. POINTLESS CONSTRAINTS

If you know that a specific constraint can never be broken or that it is always broken, do not write a score constraint for it. For example, in the N Queens problem, the score calculation does not check whether multiple queens occupy the same column because a queen's column never changes and every solution starts with each queen on a different column.



NOTE

Do not overuse this technique. If some data sets do not use a specific constraint but others do, just return out of the constraint as soon as you can. There is no need to dynamically change your score calculation based on the data set.

13.5. BUILT-IN HARD CONSTRAINTS

Instead of implementing a hard constraint, the hard constraint can sometimes be built in. For example, in the school timetabling example, if **Lecture** A should never be assigned to **Room** X, but it uses the **ValueRangeProvider** class on **Solution**, the **Solver** will often try to assign it to **Room** X only to discover that it breaks a hard constraint. Use a **ValueRangeProvider** on the planning entity or filtered selection to define that Lecture A should only be assigned a **Room** different than X.

This can give a good performance gain in some use cases, not just because the score calculation is faster, but because most optimization algorithms will spend less time evaluating infeasible solutions. However, usually this is not a good idea because there is a real risk of trading short term benefits for long term harm:

- Many optimization algorithms rely on the freedom to break hard constraints when changing planning entities, to get out of local optima.
- Both implementation approaches have limitations such as feature compatibility and disabling automatic performance optimizations.

13.6. SCORE TRAPS

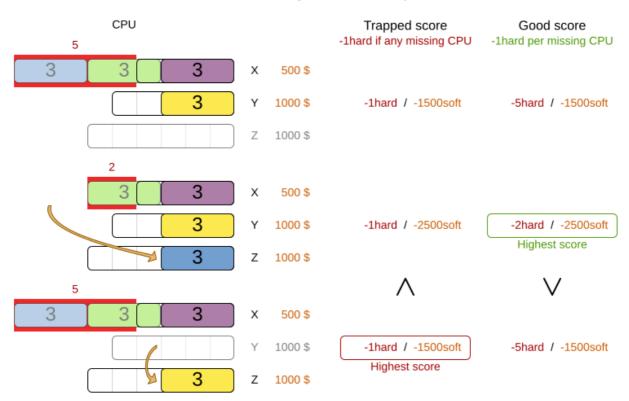
Make sure that none of your score constraints cause a score trap. A trapped score constraint applies the same weight to multiple constraint matches. Doing this groups constraint matches together and creates a flatlined score function for that constraint. This can cause a solution state in which several moves must be done to resolve or lower the weight of that single constraint. The following examples illustrate score traps:

- You need two doctors at each operating table but you are only moving one doctor at a time. The solver has no incentive to move a doctor to a table with no doctors. To fix this, penalize a table with no doctors more than a table with only one doctor in that score constraint in the score function.
- Two exams must be conducted at the same time, but you are only moving one exam at a time. The solver must move one of those exams to another time slot without moving the other in the same move. To fix this, add a coarse-grained move that moves both exams at the same time.

The following illustration shows a score trap:

Score trap

There are degrees of infeasibility



If the blue item moves from an overloaded computer to an empty computer, the hard score should improve. But the trapped score implementation cannot do that. The solver should eventually get out of this trap, but it will take a lot of effort, especially if there are even more processes on the overloaded computer. Before it can do that, it might actually start moving more processes into that overloaded computer, because there is no penalty for doing so.



NOTE

Avoiding score traps does not mean that your score function should be smart enough to avoid local optima. Leave it to the optimization algorithms to deal with the local optima.

Avoiding score traps means to avoid, for each score constraint individually, a flatlined score function.



IMPORTANT

Always specify the degree of infeasibility. The business often says "if the solution is infeasible, it does not matter how infeasible it is." While that is true for the business, it is not true for score calculation because score calculation benefits from knowing how infeasible a solution is. In practice, soft constraints usually do this naturally and it is just a matter of doing it for the hard constraints too.

There are several ways to deal with a score trap:

• Improve the score constraint to make a distinction in the score weight. For example, penalize - **1hard** for every missing CPU instead of just **-1hard** if any CPU is missing.

- If changing the score constraint is not allowed from the business perspective, add a lower score level with a score constraint that makes such a distinction. For example, penalize **-1subsoft** for every missing CPU, on top of **-1hard** if any CPU is missing. The business ignores the subsoft score level.
- Add coarse-grained moves and union-select them with the existing fine-grained moves. A coarse-grained move effectively does multiple moves to directly get out of a score trap with a single move. For example, move multiple items from the same container to another container.

13.7. THE STEPLIMIT BENCHMARK

Not all score constraints have the same performance cost. Sometimes one score constraint can kill the score calculation performance outright. Use the benchmarker to perform a one minute run and then check what happens to the score calculation speed if you comment out all but one of the score constraints.

13.8. FAIRNESS SCORE CONSTRAINTS

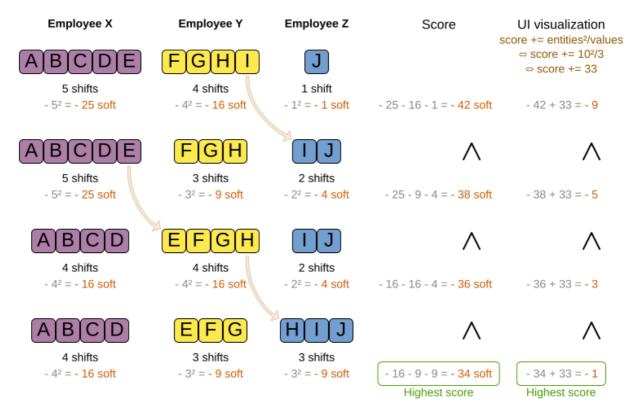
Some use cases have a business requirement to provide a fair schedule, usually as a soft score constraint, for example:

- To avoid envy, fairly distribute the workload among the employees.
- To improve reliability, evenly distribute the workload among assets.

Implementing such a constraint might seem difficult, especially because there are different ways to formalize fairness, but usually the *squared workload* implementation behaves in the most desirable manner. For each employee or asset, specify the workload as \mathbf{w} and subtract \mathbf{w}^2 from the score.

Fairness score constraint

Distribute the shift workload fairly across all employees by squaring the number of their shifts.



The *squared workload* implementation guarantees that if you select two employees from a specified solution and make the distribution between those two employees fairer, then the resulting new solution will have a better overall score. Do not use only the difference from the average workload because that can lead to unfairness, as demonstrated in the following illustration:

Fairness score constraint pitfall Don't use the deviation from the mean. Use the workload squared, variance or standard deviation. 15 shifts for 5 employees: average workload is 3 Employee Employee Employee Employee Employee Bad score Better score γ w х 7 sum(deviationMean) sum(workload²) ⇔ -sum(|workload - 3|) Ο 6 shifts 🙁 5 shifts 🙁 2 shifts 🙂 1 shifts 🙂 1 shift 🙂 - 3 - 2 - 1 - 2 - 2 = - 10 - 36 - 25 - 4 - 1 - 1 = - 67 0 D E н Highest score 5 shifts 🙁 5 shifts 🙁 2 shifts 🙂 1 shift 🙂 -2-2-1-1-2=-8 - 25 - 25 - 4 - 4 - 1 = - 59 2 shifts 😬 G 0 н Highest score Highest score 6 shifts 🙁 3 shifts 🙂 3 shifts 🙂 2 shifts 🙂 1 shift 🙂 -3-0-0-1-2=-6 - 36 - 9 - 9 - 4 -1 = - 59



NOTE

Instead of the *squared workload* implementation, it is also possible to use the variance (squared difference to the average) or the standard deviation (square root of the variance). This has no effect on the score comparison, because the average does not change during planning. It is just more work to implement because the average needs to be known and trivially slower because the calculation takes a little longer.

When the workload is perfectly balanced, users often like to see a **0** score, instead of the distracting **-34soft**. This is shown in the preceding image for the last solution which is almost perfectly balanced. To nullify this, either add the average multiplied by the number of entities to the score or show the variance or standard deviation in the UI.

13.9. OTHER SCORE CALCULATION PERFORMANCE TRICKS

Use the following tips to further improve your score calculation performance:

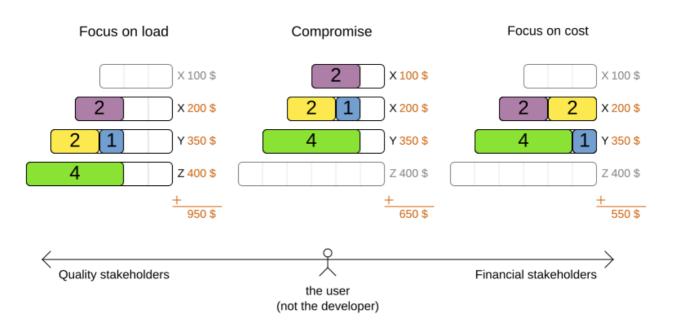
- Verify that your score calculation occurs in the correct number type. For example, if you are adding values of the type **int**, do not store the result as the type **double**, which takes longer to calculate.
- For optimal performance, use the latest Java version. For example, you can achieve a ~10 % performance increase by switching from Java 11 to 17.
- Always remember that premature optimization is very undesireable. Make sure your design is flexible enough to allow configuration-based adjustments.

13.10. CONFIGUING CONSTRAINTS

Deciding the correct weight and level for each constraint is not easy. It often involves negotiating with different stakeholders and their priorities. Furthermore, quantifying the impact of soft constraints is often a new experience for business managers, so they need a number of iterations to get it right. To make this easier, use the **@ConstraintConfiguration** class with constraint weights and parameters. Then, provide a UI so business managers can adjust the constraint weights themselves and visualize the resulting solution, as shown in the following illustration:

Parameterize the score weights

Give the user a UI to change the score weights. He/she tweaks weights by evaluating the impact on the solution.



For example, in the conference scheduling problem, the minimum pause constraint has a constraint weight, but it also has a constraint parameter that defines the length of time between two talks by the same speaker. The pause length depends on the conference: in some large conferences 20 minutes isn't enough time to go from one room to the other and in smaller conferences 10 minutes can be enough time. The pause length is a field in the constraint configuration without a **@ConstraintWeight** annotation.

Each constraint has a constraint package and a constraint name and together they form the constraint ID. The constraint ID connects the constraint weight with the constraint implementation. For each constraint weight, there must be a constraint implementation with the same package and the same name.

- The @ConstraintConfiguration annotation has a constraintPackage property that defaults to the package of the constraint configuration class. Cases with constraint streams normally do not need to specify it.
- The @ConstraintWeight annotation has a value which is the constraint name (for example "Speaker conflict"). It inherits the constraint package from the @ConstraintConfiguration, but it can override that, for example @ConstraintWeight(constraintPackage = "...region.france",

...) to use a different constraint package than some other weights.

So every constraint weight ends up with a constraint package and a constraint name. Each constraint weight links with a constraint implementation, for example in constraint streams:

```
public final class ConferenceSchedulingConstraintProvider implements ConstraintProvider {
  @Override
  public Constraint[] defineConstraints(ConstraintFactory factory) {
     return new Constraint[] {
          speakerConflict(factory),
          themeTrackConflict(factory),
          contentConflict(factory),
          ...
    };
  }
  protected Constraint speakerConflict(ConstraintFactory factory) {
     return factory.forEachUniquePair(...)
          .penalizeConfigurable("Speaker conflict", ...);
  }
  protected Constraint themeTrackConflict(ConstraintFactory factory) {
     return factory.forEachUniquePair(...)
          ...
          .penalizeConfigurable("Theme track conflict", ...);
  }
  protected Constraint contentConflict(ConstraintFactory factory) {
     return factory.forEachUniquePair(...)
          .penalizeConfigurable("Content conflict", ...);
  }
```

Each of the constraint weights defines the score level and score weight of their constraint. The constraint implementation calls **rewardConfigurable()** or **penalizeConfigurable()** and the constraint weight is automatically applied.

If the constraint implementation provides a match weight, that match weight is multiplied with the constraint weight. For example, the *content conflict* constraint weight defaults to **100soft** and the constraint implementation penalizes each match based on the number of shared content tags and the overlapping duration of the two talks:

```
@ConstraintWeight("Content conflict")
private HardMediumSoftScore contentConflict = HardMediumSoftScore.ofSoft(100);
Constraint contentConflict(ConstraintFactory factory) {
```

```
return factory.forEachUniquePair(Talk.class,
overlapping(t -> t.getTimeslot().getStartDateTime(),
t -> t.getTimeslot().getEndDateTime()),
```

```
filtering((talk1, talk2) -> talk1.overlappingContentCount(talk2) > 0))
.penalizeConfigurable("Content conflict",
(talk1, talk2) -> talk1.overlappingContentCount(talk2)
* talk1.overlappingDurationInMinutes(talk2));
```

So when 2 overlapping talks share only 1 content tag and overlap by 60 minutes, the score is impacted by **-6000soft**. But when 2 overlapping talks share 3 content tags, the match weight is 180, so the score is impacted by **-18000soft**.

Procedure

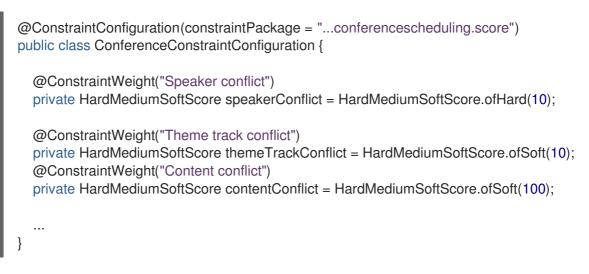
- 1. Create a new class to hold the constraint weights and other constraint parameters, for example **ConferenceConstraintConfiguration**.
- 2. Annotate this class with @ConstraintConfiguration:



3. Add the constraint configuration on the planning solution and annotate that field or property with **@ConstraintConfigurationProvider**:

<pre>@PlanningSolution public class ConferenceSolution {</pre>	
@ConstraintConfigurationProvider private ConferenceConstraintConfiguration constraintConfiguration;	

4. In the constraint configuration class, add a **@ConstraintWeight** property for each constraint and give each constraint weight a default value:



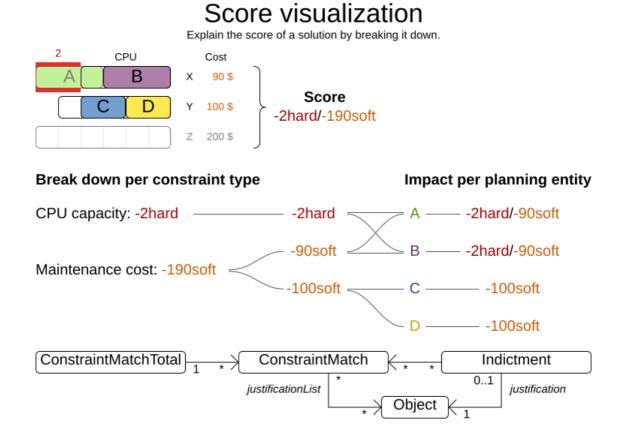
The **@ConstraintConfigurationProvider** annotation automatically exposes the constraint configuration as a problem fact. There is no need to add a **@ProblemFactProperty** annotation. A constraint weight cannot be null.

5. Expose the constraint weights in a UI so business users can tweak the values. The preceding example uses the ofHard(), ofMedium() and ofSoft() methods to do that. Notice how it defaults the content conflict constraint as ten times more important than the theme track conflict constraint. Normally, a constraint weight only uses one score level, but it's possible to use multiple score levels (at a small performance cost).

13.11. EXPLAINING THE SCORE

There are several ways to show how the OptaPlanner score is derived. This is called explaining the score:

- Print the return value of **getSummary()**. This is the easiest way to explain the score during development, but only use this method for diagnostic purposes.
- Use the **ScoreManager** API in an application or web UI.
- Break down the score for each constraint for a more granular view.



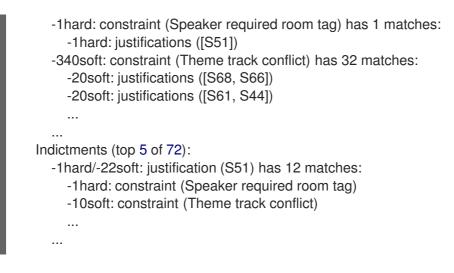
Procedure

- Use one of the following methods to explain the score:
 - Print the return value of getSummary():

System.out.println(scoreManager.getSummary(solution));

The following conference scheduling example prints that talk **S51** is responsible for breaking the hard constraint **Speaker required room tag**:

Explanation of score (-1hard/-806soft): Constraint match totals:





IMPORTANT

Do not attempt to parse this string or use it in your UI or exposed services. Instead use the ConstraintMatch API.

- Use the **ScoreManager** API in an application or web UI.
 - a. Enter code similar to the following example:

ScoreManager<CloudBalance, HardSoftScore> scoreManager = ScoreManager.create(solverFactory); ScoreExplanation<CloudBalance, HardSoftScore> scoreExplanation = scoreManager.explainScore(cloudBalance);

b. Use this code when you need to calculate the score of a solution:

HardSoftScore score = scoreExplanation.getScore();

- Break down the score by constraint:
 - a. Get the ConstraintMatchTotal values from ScoreExplanation:

```
Collection<ConstraintMatchTotal<HardSoftScore>> constraintMatchTotals =
scoreExplanation.getConstraintMatchTotalMap().values();
for (ConstraintMatchTotal<HardSoftScore> constraintMatchTotal :
constraintMatchTotals) {
String constraintName = constraintMatchTotal.getConstraintName();
// The score impact of that constraint
HardSoftScore totalScore = constraintMatchTotal.getScore();
for (ConstraintMatch<HardSoftScore> constraintMatch :
constraintMatchTotal.getConstraintMatchSet()) {
List<Object> justificationList = constraintMatch.getJustificationList();
HardSoftScore score = constraintMatch.getScore();
...
}
```

Each **ConstraintMatchTotal** represents one constraint and has a part of the overall score. The sum of all the **ConstraintMatchTotal.getScore()** equals the overall score.



NOTE

Constraint streams and Drools score calculation support constraint matches automatically, but incremental Java score calculation requires implementing an extra interface.

13.12. VISUALIZING THE HOT PLANNING ENTITIES

Show a heat map in the UI that highlights the planning entities and problem facts that have an impact on the score.

Procedure

• Get the Indictment map from the ScoreExplanation:



Each **Indictment** is the sum of all constraints where that justification object is involved. The sum of all the **Indictment.getScoreTotal()** differs from the overall score because multiple **Indictment** entities can share the same **ConstraintMatch**.



NOTE

Constraint streams and Drools score calculation supports constraint matches automatically, but incremental Java score calculation requires implementing an extra interface.

13.13. SCORE CONSTRAINTS TESTING

Different score calculation types come with different tools for testing. Write a unit test for each score constraint individually to check that it behaves correctly.

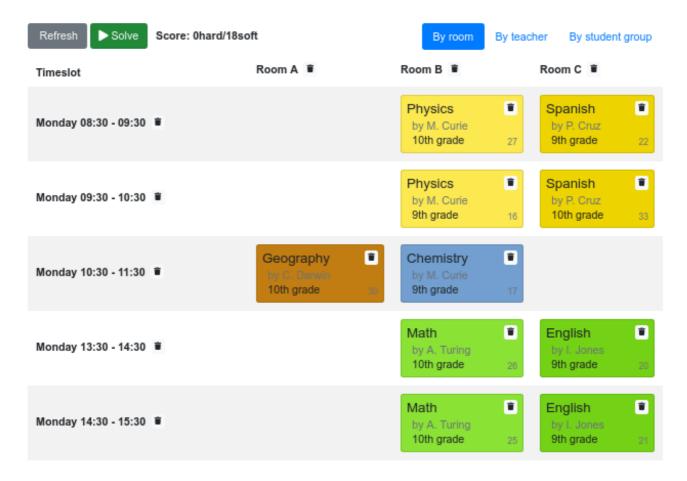
PART V. RED HAT BUILD OF OPTAPLANNER QUICK START GUIDES

Red Hat Build of OptaPlanner provides the following quick start guides to demonstrate how OptaPlanner can integrate with different techologies:

- Red Hat Build of OptaPlanner on the Red Hat build of Quarkus platform: a school timetable quick start guide
- Red Hat Build of OptaPlanner on the Red Hat build of Quarkus platform: a vaccination appointment scheduler quick start guide
- Red Hat Build of OptaPlanner on the Red Hat build of Quarkus platform: an employee scheduler quick start guide
- Red Hat Build of OptaPlanner on Spring Boot: a school timetable quick start guide
- Red Hat Build of OptaPlanner with Java solvers: a school timetable quick start guide

CHAPTER 14. RED HAT BUILD OF OPTAPLANNER ON THE RED HAT BUILD OF QUARKUS PLATFORM: A SCHOOL TIMETABLE QUICK START GUIDE

This guide walks you through the process of creating a Red Hat build of Quarkus application using the Red Hat Build of OptaPlanner constraint solving artificial intelligence (AI). You will build a REST application that optimizes a school timetable for students and teachers



Your service will assign **Lesson** instances to **Timeslot** and **Room** instances automatically by using AI to adhere to the following hard and soft *scheduling constraints*:

- A room can have at most one lesson at the same time.
- A teacher can teach at most one lesson at the same time.
- A student can attend at most one lesson at the same time.
- A teacher prefers to teach in a single room.
- A teacher prefers to teach sequential lessons and dislikes gaps between lessons.

Mathematically speaking, school timetabling is an *NP-hard* problem. That means it is difficult to scale. Simply iterating through all possible combinations with brute force would take millions of years for a non-trivial data set, even on a supercomputer. Fortunately, AI constraint solvers such as Red Hat Build of OptaPlanner have advanced algorithms that deliver a near-optimal solution in a reasonable amount of time. What is considered to be a reasonable amount of time is subjective and depends on the goals of your problem.

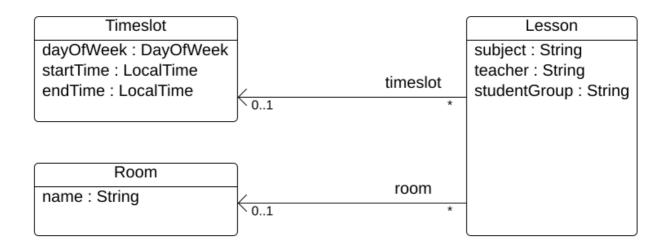
Prerequisites

- OpenJDK 11 or later is installed. Red Hat build of Open JDK is available from the Software Downloads page in the Red Hat Customer Portal (login required).
- Apache Maven 3.8 or higher is installed. Maven is available from the Apache Maven Project website.
- An IDE, such as IntelliJ IDEA, VSCode, or Eclipse is available.
- An Red Hat Build of OptaPlanner Red Hat build of Quarkus project is available. For instruction on creating an Red Hat Build of OptaPlanner Red Hat build of Quarkus project, see "Getting started with OptaPlanner and Quarkus" in the *Getting Started with Red Hat Build of OptaPlanner* section.

14.1. MODEL THE DOMAIN OBJECTS

The goal of the Red Hat Build of OptaPlanner timetable project is to assign each lesson to a time slot and a room. To do this, add three classes, **Timeslot**, **Lesson**, and **Room**, as shown in the following diagram:

Time table class diagram



Timeslot

The **Timeslot** class represents a time interval when lessons are taught, for example, **Monday 10:30 -11:30** or **Tuesday 13:30 - 14:30**. In this example, all time slots have the same duration and there are no time slots during lunch or other breaks.

A time slot has no date because a high school schedule just repeats every week. There is no need for continuous planning. A timeslot is called a *problem fact* because no **Timeslot** instances change during solving. Such classes do not require any OptaPlanner-specific annotations.

Room

The **Room** class represents a location where lessons are taught, for example, **Room A** or **Room B**. In this example, all rooms are without capacity limits and they can accommodate all lessons.

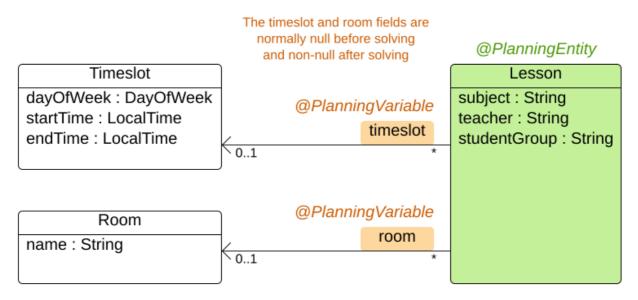
Room instances do not change during solving so **Room** is also a *problem fact*.

Lesson

During a lesson, represented by the **Lesson** class, a teacher teaches a subject to a group of students, for example, **Math by A.Turing for 9th grade** or **Chemistry by M.Curie for 10th grade**. If a subject is taught multiple times each week by the same teacher to the same student group, there are multiple **Lesson** instances that are only distinguishable by **id**. For example, the 9th grade has six math lessons a week.

During solving, OptaPlanner changes the **timeslot** and **room** fields of the **Lesson** class to assign each lesson to a time slot and a room. Because OptaPlanner changes these fields, **Lesson** is a *planning entity*:

Time table class diagram



Most of the fields in the previous diagram contain input data, except for the orange fields. A lesson's **timeslot** and **room** fields are unassigned (**null**) in the input data and assigned (not **null**) in the output data. OptaPlanner changes these fields during solving. Such fields are called planning variables. In order for OptaPlanner to recognize them, both the **timeslot** and **room** fields require an **@PlanningVariable** annotation. Their containing class, **Lesson**, requires an **@PlanningEntity** annotation.

Procedure

1. Create the src/main/java/com/example/domain/Timeslot.java class:

package com.example.domain; import java.time.DayOfWeek; import java.time.LocalTime; public class Timeslot { private DayOfWeek dayOfWeek; private LocalTime startTime; private LocalTime endTime;

private Timeslot() {



Notice the **toString()** method keeps the output short so it is easier to read OptaPlanner's **DEBUG** or **TRACE** log, as shown later.

2. Create the src/main/java/com/example/domain/Room.java class:

3. Create the src/main/java/com/example/domain/Lesson.java class:

```
package com.example.domain;
import org.optaplanner.core.api.domain.entity.PlanningEntity;
import org.optaplanner.core.api.domain.variable.PlanningVariable;
@PlanningEntity
public class Lesson {
  private Long id;
  private String subject;
  private String teacher;
  private String studentGroup;
  @PlanningVariable(valueRangeProviderRefs = "timeslotRange")
  private Timeslot timeslot;
  @PlanningVariable(valueRangeProviderRefs = "roomRange")
  private Room room;
  private Lesson() {
  }
  public Lesson(Long id, String subject, String teacher, String studentGroup) {
    this.id = id;
    this.subject = subject;
    this.teacher = teacher;
    this.studentGroup = studentGroup;
  }
  @Override
  public String toString() {
    return subject + "(" + id + ")";
  }
  // ********
  // Getters and setters
  // *********
  public Long getId() {
    return id;
  }
  public String getSubject() {
    return subject;
  }
```

```
public String getTeacher() {
     return teacher:
  }
  public String getStudentGroup() {
     return studentGroup;
  }
  public Timeslot getTimeslot() {
     return timeslot:
  }
  public void setTimeslot(Timeslot timeslot) {
     this.timeslot = timeslot;
  }
  public Room getRoom() {
     return room;
  }
  public void setRoom(Room room) {
     this.room = room;
  }
}
```

The **Lesson** class has an **@PlanningEntity** annotation, so OptaPlanner knows that this class changes during solving because it contains one or more planning variables.

The **timeslot** field has an **@PlanningVariable** annotation, so OptaPlanner knows that it can change its value. In order to find potential **Timeslot** instances to assign to this field, OptaPlanner uses the **valueRangeProviderRefs** property to connect to a value range provider that provides a **List<Timeslot>** to pick from. See Section 14.3, "Gather the domain objects in a planning solution" for information about value range providers.

The **room** field also has an **@PlanningVariable** annotation for the same reasons.

14.2. DEFINE THE CONSTRAINTS AND CALCULATE THE SCORE

When solving a problem, a score represents the quality of a specific solution. The higher the score the better. Red Hat Build of OptaPlanner looks for the best solution, which is the solution with the highest score found in the available time. It might be the *optimal* solution.

Because the timetable example use case has hard and soft constraints, use the **HardSoftScore** class to represent the score:

- Hard constraints must not be broken. For example: A room can have at most one lesson at the same time.
- Soft constraints should not be broken. For example: A teacher prefers to teach in a single room.

Hard constraints are weighted against other hard constraints. Soft constraints are weighted against other soft constraints. Hard constraints always outweigh soft constraints, regardless of their respective weights.

To calculate the score, you could implement an **EasyScoreCalculator** class:

```
public class TimeTableEasyScoreCalculator implements EasyScoreCalculator<TimeTable> {
  @Override
  public HardSoftScore calculateScore(TimeTable timeTable) {
     List<Lesson> lessonList = timeTable.getLessonList();
     int hardScore = 0;
     for (Lesson a : lessonList) {
       for (Lesson b : lessonList) {
          if (a.getTimeslot() != null && a.getTimeslot().equals(b.getTimeslot())
               && a.getId() < b.getId()) {
            // A room can accommodate at most one lesson at the same time.
            if (a.getRoom() != null && a.getRoom().equals(b.getRoom())) {
               hardScore--;
            }
            // A teacher can teach at most one lesson at the same time.
            if (a.getTeacher().equals(b.getTeacher())) {
               hardScore--;
            }
            // A student can attend at most one lesson at the same time.
            if (a.getStudentGroup().equals(b.getStudentGroup())) {
               hardScore--;
            }
         }
       }
     }
     int softScore = 0;
     // Soft constraints are only implemented in the "complete" implementation
     return HardSoftScore.of(hardScore, softScore);
  }
```

Unfortunately, this solution does not scale well because it is non-incremental: every time a lesson is assigned to a different time slot or room, all lessons are re-evaluated to calculate the new score.

A better solution is to create a **src/main/java/com/example/solver/TimeTableConstraintProvider.java** class to perform incremental score calculation. This class uses OptaPlanner's ConstraintStream APl which is inspired by Java 8 Streams and SQL. The **ConstraintProvider** scales an order of magnitude better than the **EasyScoreCalculator**: O(n) instead of O(n²).

Procedure

Create the following src/main/java/com/example/solver/TimeTableConstraintProvider.java class:

package com.example.solver;

import com.example.domain.Lesson; import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore; import org.optaplanner.core.api.score.stream.Constraint; import org.optaplanner.core.api.score.stream.ConstraintFactory; import org.optaplanner.core.api.score.stream.ConstraintProvider; import org.optaplanner.core.api.score.stream.Joiners;

public class TimeTableConstraintProvider implements ConstraintProvider {

```
@Override
public Constraint[] defineConstraints(ConstraintFactory constraintFactory) {
  return new Constraint[] {
       // Hard constraints
       roomConflict(constraintFactory),
       teacherConflict(constraintFactory),
       studentGroupConflict(constraintFactory),
       // Soft constraints are only implemented in the "complete" implementation
  };
}
private Constraint roomConflict(ConstraintFactory constraintFactory) {
  // A room can accommodate at most one lesson at the same time.
  // Select a lesson ...
  return constraintFactory.forEach(Lesson.class)
       // ... and pair it with another lesson ...
       .join(Lesson.class,
            // ... in the same timeslot ...
            Joiners.equal(Lesson::getTimeslot),
            // ... in the same room ...
            Joiners.equal(Lesson::getRoom),
            // ... and the pair is unique (different id, no reverse pairs)
            Joiners.lessThan(Lesson::getId))
       // then penalize each pair with a hard weight.
       .penalize(HardSoftScore.ONE_HARD)
       .asConstraint("Room conflict");
}
private Constraint teacherConflict(ConstraintFactory constraintFactory) {
  // A teacher can teach at most one lesson at the same time.
  return constraintFactory.forEach(Lesson.class)
       .join(Lesson.class,
            Joiners.equal(Lesson::getTimeslot),
            Joiners.equal(Lesson::getTeacher),
            Joiners.lessThan(Lesson::getId))
            .penalize(HardSoftScore.ONE HARD)
            .asConstraint("Teacher conflict");
}
private Constraint studentGroupConflict(ConstraintFactory constraintFactory) {
  // A student can attend at most one lesson at the same time.
  return constraintFactory.forEach(Lesson.class)
       .join(Lesson.class,
            Joiners.equal(Lesson::getTimeslot),
            Joiners.equal(Lesson::getStudentGroup),
            Joiners.lessThan(Lesson::getId))
            .penalize(HardSoftScore.ONE HARD)
            .asConstraint("Student group conflict");
}
```

14.3. GATHER THE DOMAIN OBJECTS IN A PLANNING SOLUTION

A **TimeTable** instance wraps all **Timeslot**, **Room**, and **Lesson** instances of a single dataset. Furthermore, because it contains all lessons, each with a specific planning variable state, it is a *planning solution* and it has a score:

- If lessons are still unassigned, then it is an *uninitialized* solution, for example, a solution with the score **-4init/0hard/0soft**.
- If it breaks hard constraints, then it is an *infeasible* solution, for example, a solution with the score **-2hard/-3soft**.
- If it adheres to all hard constraints, then it is a *feasible* solution, for example, a solution with the score **0hard/-7soft**.

The **TimeTable** class has an **@PlanningSolution** annotation, so Red Hat Build of OptaPlanner knows that this class contains all of the input and output data.

Specifically, this class is the input of the problem:

- A timeslotList field with all time slots
 - This is a list of problem facts, because they do not change during solving.
- A roomList field with all rooms
 - This is a list of problem facts, because they do not change during solving.
- A lessonList field with all lessons
 - This is a list of planning entities because they change during solving.
 - Of each Lesson:
 - The values of the timeslot and room fields are typically still null, so unassigned. They are planning variables.
 - The other fields, such as subject, teacher and studentGroup, are filled in. These fields are problem properties.

However, this class is also the output of the solution:

- A lessonList field for which each Lesson instance has non-null timeslot and room fields after solving
- A score field that represents the quality of the output solution, for example, **0hard/-5soft**

Procedure

Create the src/main/java/com/example/domain/TimeTable.java class:

package com.example.domain;

import java.util.List;

import org.optaplanner.core.api.domain.solution.PlanningEntityCollectionProperty; import org.optaplanner.core.api.domain.solution.PlanningScore; import org.optaplanner.core.api.domain.solution.PlanningSolution; import org.optaplanner.core.api.domain.solution.ProblemFactCollectionProperty; import org.optaplanner.core.api.domain.valuerange.ValueRangeProvider; import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore;

```
@PlanningSolution
public class TimeTable {
  @ValueRangeProvider(id = "timeslotRange")
  @ProblemFactCollectionProperty
  private List<Timeslot> timeslotList;
  @ValueRangeProvider(id = "roomRange")
  @ProblemFactCollectionProperty
  private List<Room> roomList;
  @PlanningEntityCollectionProperty
  private List<Lesson> lessonList;
  @PlanningScore
  private HardSoftScore score;
  private TimeTable() {
  }
  public TimeTable(List<Timeslot> timeslotList, List<Room> roomList,
       List<Lesson> lessonList) {
    this.timeslotList = timeslotList;
    this.roomList = roomList;
    this.lessonList = lessonList;
  }
  // Getters and setters
  // ********
  public List<Timeslot> getTimeslotList() {
    return timeslotList;
  }
  public List<Room> getRoomList() {
    return roomList;
  }
  public List<Lesson> getLessonList() {
    return lessonList;
  }
  public HardSoftScore getScore() {
    return score:
  }
```

The value range providers

The **timeslotList** field is a value range provider. It holds the **Timeslot** instances which OptaPlanner can pick from to assign to the **timeslot** field of **Lesson** instances. The **timeslotList** field has an **@ValueRangeProvider** annotation to connect those two, by matching the **id** with the

valueRangeProviderRefs of the @PlanningVariable in the Lesson.

Following the same logic, the **roomList** field also has an **@ValueRangeProvider** annotation.

The problem fact and planning entity properties

Furthermore, OptaPlanner needs to know which **Lesson** instances it can change as well as how to retrieve the **Timeslot** and **Room** instances used for score calculation by your **TimeTableConstraintProvider**.

The timeslotList and roomList fields have an @ProblemFactCollectionProperty annotation, so your TimeTableConstraintProvider can select from those instances.

The **lessonList** has an **@PlanningEntityCollectionProperty** annotation, so OptaPlanner can change them during solving and your **TimeTableConstraintProvider** can select from those too.

14.4. CREATE THE SOLVER SERVICE

Solving planning problems on REST threads causes HTTP timeout issues. Therefore, the Quarkus extension injects a SolverManager, which runs solvers in a separate thread pool and can solve multiple data sets in parallel.

Procedure

Create the src/main/java/org/acme/optaplanner/rest/TimeTableResource.java class:

package org.acme.optaplanner.rest;

```
import java.util.UUID;
import java.util.concurrent.ExecutionException;
import javax.inject.Inject;
import javax.ws.rs.POST;
import javax.ws.rs.Path;
```

import org.acme.optaplanner.domain.TimeTable; import org.optaplanner.core.api.solver.SolverJob; import org.optaplanner.core.api.solver.SolverManager;

@Path("/timeTable")
public class TimeTableResource {

@Inject
SolverManager<TimeTable, UUID> solverManager;

@POST @Path("/solve") public TimeTable solve(TimeTable problem) { UUID problemId = UUID.randomUUID(); // Submit the problem to start solving SolverJob<TimeTable, UUID> solverJob = solverManager.solve(problemId, problem); TimeTable solution; try { // Wait until the solving ends solution = solverJob.getFinalBestSolution(); } catch (InterruptedException | ExecutionException e) { throw new IllegalStateException("Solving failed.", e); } }

	}	} return solution;
}		

This initial implementation waits for the solver to finish, which can still cause an HTTP timeout. The complete implementation avoids HTTP timeouts much more elegantly.

14.5. SET THE SOLVER TERMINATION TIME

If your planning application does not have a termination setting or a termination event, it theoretically runs forever and in reality eventually causes an HTTP timeout error. To prevent this from occurring, use the **optaplanner.solver.termination.spent-limit** parameter to specify the length of time after which the application terminates. In most applications, set the time to at least five minutes (**5m**). However, in the Timetable example, limit the solving time to five seconds, which is short enough to avoid the HTTP timeout.

Procedure

Create the src/main/resources/application.properties file with the following content:

quarkus.optaplanner.solver.termination.spent-limit=5s

14.6. RUNNING THE SCHOOL TIMETABLE APPLICATION

After you have created the school timetable project, run it in development mode. In development mode, you can update the application sources and configurations while your application is running. Your changes will appear in the running application.

Prerequisites

• You have created the school timetable project.

Procedure

1. To compile the application in development mode, enter the following command from the project directory:



./mvnw compile quarkus:dev

2. Test the REST service. You can use any REST client. The following example uses the Linux command **curl** to send a POST request:

```
$ curl -i -X POST http://localhost:8080/timeTable/solve -H "Content-Type:application/json" -d
'{"timeslotList":[{"dayOfWeek":"MONDAY","startTime":"08:30:00","endTime":"09:30:00"},
{"dayOfWeek":"MONDAY","startTime":"09:30:00","endTime":"10:30:00"}],"roomList":
[{"name":"Room A"},{"name":"Room B"}],"lessonList":[{"id":1,"subject":"Math","teacher":"A.
Turing","studentGroup":"9th grade"},{"id":2,"subject":"Chemistry","teacher":"M.
Curie","studentGroup":"9th grade"},{"id":3,"subject":"French","teacher":"M.
Curie","studentGroup":"10th grade"},{"id":4,"subject":"History","teacher":"I.
Jones","studentGroup":"10th grade"}]}'
```

After the time period specified in **termination spent time** defined in your **application.properties** file, the service returns output similar to the following example:

HTTP/1.1 200 Content-Type: application/json

•••

{"timeslotList":...,"roomList":...,"lessonList":[{"id":1,"subject":"Math","teacher":"A. Turing","studentGroup":"9th grade","timeslot":

{"dayOfWeek":"MONDAY","startTime":"08:30:00","endTime":"09:30:00"},"room": {"name":"Room A"}},{"id":2,"subject":"Chemistry","teacher":"M. Curie","studentGroup":"9th grade","timeslot": {"dayOfWeek":"MONDAY","startTime":"09:30:00","endTime":"10:30:00"},"room":

{"name":"Room A"}},{"id":3,"subject":"French","teacher":"M. Curie","studentGroup":"10th grade","timeslot":

{"dayOfWeek":"MONDAY","startTime":"08:30:00","endTime":"09:30:00"},"room": {"name":"Room B"}},{"id":4,"subject":"History","teacher":"I. Jones","studentGroup":"10th grade","timeslot":

{"dayOfWeek":"MONDAY","startTime":"09:30:00","endTime":"10:30:00"},"room": {"name":"Room B"}}],"score":"0hard/0soft"}

Notice that your application assigned all four lessons to one of the two time slots and one of the two rooms. Also notice that it conforms to all hard constraints. For example, M. Curie's two lessons are in different time slots.

3. To review what OptaPlanner did during the solving time, review the info log on the server side. The following is sample info log output:

... Solving started: time spent (33), best score (-8init/0hard/0soft), environment mode (REPRODUCIBLE), random (JDK with seed 0).

... Construction Heuristic phase (0) ended: time spent (73), best score (0hard/0soft), score calculation speed (459/sec), step total (4).

... Local Search phase (1) ended: time spent (5000), best score (0hard/0soft), score calculation speed (28949/sec), step total (28398).

... Solving ended: time spent (5000), best score (0hard/0soft), score calculation speed (28524/sec), phase total (2), environment mode (REPRODUCIBLE).

14.7. TESTING THE APPLICATION

A good application includes test coverage. Test the constraints and the solver in your timetable project.

14.7.1. Test the school timetable constraints

To test each constraint of the timetable project in isolation, use a **ConstraintVerifier** in unit tests. This tests each constraint's corner cases in isolation from the other tests, which lowers maintenance when adding a new constraint with proper test coverage.

This test verifies that the constraint **TimeTableConstraintProvider::roomConflict**, when given three lessons in the same room and two of the lessons have the same timeslot, penalizes with a match weight of 1. So if the constraint weight is **10hard** it reduces the score by **-10hard**.

Procedure

Create the src/test/java/org/acme/optaplanner/solver/TimeTableConstraintProviderTest.java class:

```
package org.acme.optaplanner.solver;
import java.time.DayOfWeek;
import java.time.LocalTime;
import javax.inject.Inject;
import io.quarkus.test.junit.QuarkusTest;
import org.acme.optaplanner.domain.Lesson;
import org.acme.optaplanner.domain.Room;
import org.acme.optaplanner.domain.TimeTable;
import org.acme.optaplanner.domain.Timeslot;
import org.junit.jupiter.api.Test;
import org.optaplanner.test.api.score.stream.ConstraintVerifier;
@QuarkusTest
class TimeTableConstraintProviderTest {
  private static final Room ROOM = new Room("Room1");
  private static final Timeslot TIMESLOT1 = new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9,0),
LocalTime.NOON);
  private static final Timeslot TIMESLOT2 = new Timeslot(DayOfWeek.TUESDAY,
LocalTime.of(9,0), LocalTime.NOON);
  @Inject
  ConstraintVerifier<TimeTableConstraintProvider, TimeTable> constraintVerifier;
  @Test
  void roomConflict() {
    Lesson firstLesson = new Lesson(1, "Subject1", "Teacher1", "Group1");
    Lesson conflictingLesson = new Lesson(2, "Subject2", "Teacher2", "Group2");
    Lesson nonConflictingLesson = new Lesson(3, "Subject3", "Teacher3", "Group3");
    firstLesson.setRoom(ROOM);
    firstLesson.setTimeslot(TIMESLOT1);
    conflictingLesson.setRoom(ROOM);
    conflictingLesson.setTimeslot(TIMESLOT1);
    nonConflictingLesson.setRoom(ROOM);
    nonConflictingLesson.setTimeslot(TIMESLOT2);
    constraintVerifier.verifyThat(TimeTableConstraintProvider::roomConflict)
         .given(firstLesson, conflictingLesson, nonConflictingLesson)
         .penalizesBy(1);
  }
```

Notice how **ConstraintVerifier** ignores the constraint weight during testing even if those constraint weights are hardcoded in the **ConstraintProvider**. This is because constraint weights change regularly before going into production. This way, constraint weight tweaking does not break the unit tests.

14.7.2. Test the school timetable solver

This example tests the Red Hat Build of OptaPlanner school timetable project on the Red Hat build of Quarkus platform. It uses a JUnit test to generate a test data set and send it to the **TimeTableController** to solve.

Procedure

1. Create the **src/test/java/com/example/rest/TimeTableResourceTest.java** class with the following content:

package com.exmaple.optaplanner.rest;

import java.time.DayOfWeek; import java.time.LocalTime; import java.util.ArrayList; import java.util.List;

import javax.inject.Inject;

import io.quarkus.test.junit.QuarkusTest; import com.exmaple.optaplanner.domain.Room; import com.exmaple.optaplanner.domain.Timeslot; import com.exmaple.optaplanner.domain.Lesson; import com.exmaple.optaplanner.domain.TimeTable; import com.exmaple.optaplanner.rest.TimeTableResource; import org.junit.jupiter.api.Test; import org.junit.jupiter.api.Timeout;

import static org.junit.jupiter.api.Assertions.assertFalse; import static org.junit.jupiter.api.Assertions.assertNotNull; import static org.junit.jupiter.api.Assertions.assertTrue;

@QuarkusTest
public class TimeTableResourceTest {

@Inject
TimeTableResource timeTableResource;

```
@Test
  @Timeout(600_000)
  public void solve() {
     TimeTable problem = generateProblem();
     TimeTable solution = timeTableResource.solve(problem);
     assertFalse(solution.getLessonList().isEmpty());
    for (Lesson lesson : solution.getLessonList()) {
       assertNotNull(lesson.getTimeslot());
       assertNotNull(lesson.getRoom());
    }
     assertTrue(solution.getScore().isFeasible());
  }
  private TimeTable generateProblem() {
     List<Timeslot> timeslotList = new ArrayList<>();
     timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(8, 30),
LocalTime.of(9, 30)));
     timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9, 30),
LocalTime.of(10, 30)));
```

timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(10, 30), LocalTime.of(11, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(13, 30), LocalTime.of(14, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(14, 30), LocalTime.of(15, 30))); List<Room> roomList = new ArrayList<>(); roomList.add(new Room("Room A")); roomList.add(new Room("Room B")); roomList.add(new Room("Room C")); List<Lesson> lessonList = new ArrayList<>(); lessonList.add(new Lesson(101L, "Math", "B. May", "9th grade")); lessonList.add(new Lesson(102L, "Physics", "M. Curie", "9th grade")); lessonList.add(new Lesson(103L, "Geography", "M. Polo", "9th grade")); lessonList.add(new Lesson(104L, "English", "I. Jones", "9th grade")); lessonList.add(new Lesson(105L, "Spanish", "P. Cruz", "9th grade")); lessonList.add(new Lesson(201L, "Math", "B. May", "10th grade")); lessonList.add(new Lesson(202L, "Chemistry", "M. Curie", "10th grade")); lessonList.add(new Lesson(203L, "History", "I. Jones", "10th grade")); lessonList.add(new Lesson(204L, "English", "P. Cruz", "10th grade")); lessonList.add(new Lesson(205L, "French", "M. Curie", "10th grade")); return new TimeTable(timeslotList, roomList, lessonList); } }

This test verifies that after solving, all lessons are assigned to a time slot and a room. It also verifies that it found a feasible solution (no hard constraints broken).

2. Add test properties to the src/main/resources/application.properties file:

The solver runs only for 5 seconds to avoid a HTTP timeout in this simple implementation.
 # It's recommended to run for at least 5 minutes ("5m") otherwise.
 quarkus.optaplanner.solver.termination.spent-limit=5s

Effectively disable this termination in favor of the best-score-limit %test.quarkus.optaplanner.solver.termination.spent-limit=1h %test.quarkus.optaplanner.solver.termination.best-score-limit=0hard/*soft

Normally, the solver finds a feasible solution in less than 200 milliseconds. Notice how the **application.properties** file overwrites the solver termination during tests to terminate as soon as a feasible solution **(0hard/*soft)** is found. This avoids hard coding a solver time, because the unit test might run on arbitrary hardware. This approach ensures that the test runs long enough to find a feasible solution, even on slow systems. But it does not run a millisecond longer than it strictly must, even on fast systems.

14.8. LOGGING

After you complete the Red Hat Build of OptaPlanner school timetable project, you can use logging information to help you fine-tune the constraints in the **ConstraintProvider**. Review the score calculation speed in the **info** log file to assess the impact of changes to your constraints. Run the

application in debug mode to show every step that your application takes or use trace logging to log every step and every move.

Procedure

- 1. Run the school timetable application for a fixed amount of time, for example, five minutes.
- 2. Review the score calculation speed in the **log** file as shown in the following example:

... Solving ended: ..., score calculation speed (29455/sec), ...

- 3. Change a constraint, run the planning application again for the same amount of time, and review the score calculation speed recorded in the **log** file.
- 4. Run the application in debug mode to log every step that the application makes:
 - To run debug mode from the command line, use the **-D** system property.
 - To permanently enable debug mode, add the following line to the **application.properties** file:

quarkus.log.category."org.optaplanner".level=debug

The following example shows output in the **log** file in debug mode:

... Solving started: time spent (67), best score (-20init/0hard/0soft), environment mode (REPRODUCIBLE), random (JDK with seed 0). ... CH step (0), time spent (128), score (-18init/0hard/0soft), selected move count (15), picked move ([Math(101) {null -> Room A}, Math(101) {null -> MONDAY 08:30}]). ... CH step (1), time spent (145), score (-16init/0hard/0soft), selected move count (15), picked move ([Physics(102) {null -> Room A}, Physics(102) {null -> MONDAY 09:30}]). ...

5. Use **trace** logging to show every step and every move for each step.

14.9. INTEGRATING A DATABASE WITH YOUR QUARKUS OPTAPLANNER SCHOOL TIMETABLE APPLICATION

After you create your Quarkus OptaPlanner school timetable application, you can integrate it with a database and create a web-based user interface to display the timetable.

Prerequisites

• You have a Quarkus OptaPlanner school timetable application.

Procedure

- 1. Use Hibernate and Panache to store **Timeslot**, **Room**, and **Lesson** instances in a database. See Simplified Hibernate ORM with Panache for more information.
- 2. Expose the instances through REST. For information, see Writing JSON REST Services.
- 3. Update the **TimeTableResource** class to read and write a **TimeTable** instance in a single transaction:

package org.acme.optaplanner.rest;

```
import javax.inject.Inject;
import javax.transaction.Transactional;
import javax.ws.rs.GET;
import javax.ws.rs.POST;
import javax.ws.rs.Path;
import io.quarkus.panache.common.Sort;
```

import org.acme.optaplanner.domain.Lesson; import org.acme.optaplanner.domain.Room; import org.acme.optaplanner.domain.TimeTable; import org.acme.optaplanner.domain.Timeslot; import org.optaplanner.core.api.score.ScoreManager; import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore; import org.optaplanner.core.api.solver.SolverManager; import org.optaplanner.core.api.solver.SolverManager;

@Path("/timeTable")
public class TimeTableResource {

```
public static final Long SINGLETON_TIME_TABLE_ID = 1L;
```

@InjectSolverManager<TimeTable, Long> solverManager;@InjectScoreManager<TimeTable, HardSoftScore> scoreManager;

```
// To try, open http://localhost:8080/timeTable
@GET
public TimeTable getTimeTable() {
```

// Get the solver status before loading the solution
// to avoid the race condition that the solver terminates between them
SolverStatus solverStatus = getSolverStatus();

```
TimeTable solution = findById(SINGLETON_TIME_TABLE_ID);
scoreManager.updateScore(solution); // Sets the score
solution.setSolverStatus(solverStatus);
```

return solution;

```
}
```

}

```
@POST
@Path("/solve")
public void solve() {
    solverManager.solveAndListen(SINGLETON_TIME_TABLE_ID,
        this::findById,
        this::save);
}
public SolverStatus getSolverStatus() {
    return solverManager.getSolverStatus(SINGLETON_TIME_TABLE_ID);
```

```
@POST
@Path("/stopSolving")
public void stopSolving() {
    solverManager.terminateEarly(SINGLETON_TIME_TABLE_ID);
```



This example includes a **TimeTable** instance. However, you can enable multi-tenancy and handle **TimeTable** instances for multiple schools in parallel.

The **getTimeTable()** method returns the latest timetable from the database. It uses the **ScoreManager** method, which is automatically injected, to calculate the score of that timetable and make it available to the UI.

The **solve()** method starts a job to solve the current timetable and stores the time slot and room assignments in the database. It uses the **SolverManager.solveAndListen()** method to listen to intermediate best solutions and update the database accordingly. The UI uses this to show progress while the backend is still solving.

4. Update the **TimeTableResourceTest** class to reflect that the **solve()** method returns immediately and to poll for the latest solution until the solver finishes solving:

package org.acme.optaplanner.rest;

import javax.inject.Inject;

import io.quarkus.test.junit.QuarkusTest; import org.acme.optaplanner.domain.Lesson; import org.acme.optaplanner.domain.TimeTable; import org.junit.jupiter.api.Test; import org.junit.jupiter.api.Timeout; import org.optaplanner.core.api.solver.SolverStatus;

import static org.junit.jupiter.api.Assertions.assertFalse;



- 5. Build a web UI on top of these REST methods to provide a visual representation of the timetable.
- 6. Review the quickstart source code.

14.10. USING MICROMETER AND PROMETHEUS TO MONITOR YOUR SCHOOL TIMETABLE OPTAPLANNER QUARKUS APPLICATION

OptaPlanner exposes metrics through Micrometer, a metrics instrumentation library for Java applications. You can use Micrometer with Prometheus to monitor the OptaPlanner solver in the school timetable application.

Prerequisites

- You have created the Quarkus OptaPlanner school timetable application.
- Prometheus is installed. For information about installing Prometheus, see the Prometheus website.

Procedure

1. Add the Micrometer Prometheus dependency to the school timetable **pom.xml** file:

<dependency> <groupId>io.quarkus</groupId> <artifactId>quarkus-micrometer-registry-prometheus</artifactId> </dependency>

- 2. Start the school timetable application:
 - mvn compile quarkus:dev
- 3. Open http://localhost:8080/q/metric in a web browser.

CHAPTER 15. RED HAT BUILD OF OPTAPLANNER ON RED HAT BUILD OF QUARKUS: A VACCINATION APPOINTMENT SCHEDULER QUICK START GUIDE

You can use the OptaPlanner vaccination appointment scheduler quick start to develop a vaccination schedule that is both efficient and fair. The vaccination appointment scheduler uses artificial intelligence (AI) to prioritize people and allocate time slots based on multiple constraints and priorities.

Prerequisites

- OpenJDK 11 or later is installed. Red Hat build of Open JDK is available from the Software Downloads page in the Red Hat Customer Portal (login required).
- Apache Maven 3.8 or higher is installed. Maven is available from the Apache Maven Project website.
- An IDE, such as IntelliJ IDEA, VSCode, or Eclipse is available.
- You have created a OptaPlanner project on the Red Hat build of Quarkus platform project as described in Chapter 5, *Getting Started with Red Hat Build of OptaPlanner on the Red Hat build of Quarkus platform*.

15.1. HOW THE OPTAPLANNER VACCINATION APPOINTMENT SCHEDULER WORKS

There are two main approaches to scheduling appointments. The system can either let a person choose an appointment slot (user-selects) or the system assigns a slot and tells the person when and where to attend (system-automatically-assigns). The OptaPlanner vaccination appointment scheduler uses the system-automatically-assigns approach. With the OptaPlanner vaccination appointment scheduler, you can create an application where people provide their information to the system and the system assigns an appointment.

Characteristics of this approach:

- Appointment slots are allocated based on priority.
- The system allocates the best appointment time and location based on preconfigured planning constraints.
- The system is not overwhelmed by a large number of users competing for a limited number of appointments.

This approach solves the problem of vaccinating as many people as possible by using planning constraints to create a score for each person. The person's score determines when they get an appointment. The higher the person's score, the better chance they have of receiving an earlier appointment.

15.1.1. Red Hat Build of OptaPlanner vaccination appointment scheduler constraints

Red Hat Build of OptaPlanner vaccination appointment scheduler constraints are either hard, medium, or soft:

• Hard constraints cannot be broken. If any hard constraint is broken, the plan is unfeasible and cannot be executed:

- Capacity: Do not over-book vaccine capacity at any time at any location.
- Vaccine max age: If a vaccine has a maximum age, do not administer it to people who at the time of the first dose vaccination are older than the vaccine maximum age. Ensure people are given a vaccine type appropriate for their age. For example, do not assign a 75 year old person an appointment for a vaccine that has a maximum age restriction of 65 years.
- Required vaccine type: Use the required vaccine type. For example, the second dose of a vaccine must be the same vaccine type as the first dose.
- Ready date: Administer the vaccine on or after the specified date. For example, if a person receives a second dose, do not administer it before the recommended earliest possible vaccination date for the specific vaccine type, for example 26 days after the first dose.
- Due date: Administer the vaccine on or before the specified date. For example, if a person receives a second dose, administer it before the recommended vaccination final due date for the specific vaccine, for example three months after the first dose.
- Restrict maximum travel distance: Assign each person to one of a group of vaccination centers nearest to them. This is typically one of three centers. This restriction is calculated by travel time, not distance, so a person that lives in an urban area usually has a lower maximum distance to travel than a person living in a rural area.
- Medium constraints decide who does not get an appointment when there is not enough capacity to assign appointments to everyone. This is called overconstrained planning:
 - Schedule second dose vaccinations: Do not leave any second dose vaccination appointments unassigned unless the ideal date falls outside of the planning window.
 - Schedule people based on their priority rating: Each person has a priority rating. This is typically their age but it can be much higher if they are, for example, a health care worker. Leave only people with the lowest priority ratings unassigned. They will be considered in the next run. This constraint is softer than the previous constraint because the second dose is always prioritized over priority rating.
- Soft constraints should not be broken:
 - Preferred vaccination center: If a person has a preferred vaccination center, give them an appointment at that center.
 - Distance: Minimize the distance that a person must travel to their assigned vaccination center.
 - Ideal date: Administer the vaccine on or as close to the specified date as possible. For example, if a person receives a second dose, administer it on the ideal date for the specific vaccine, for example 28 days after the first dose. This constraint is softer than the distance constraint to avoid sending people halfway across the country just to be one day closer to their ideal date.
 - Priority rating: Schedule people with a higher priority rating earlier in the planning window. This constraint is softer than the distance constraint to avoid sending people halfway across the country. This constraint is also softer than the ideal date constraint because the second dose is prioritized over priority rating.

Hard constraints are weighted against other hard constraints. Soft constraints are weighted against other soft constraints. However, hard constraints always take precedence over medium and soft constraints. If a hard constraint is broken, then the plan is not feasible. But if no hard constraints are

broken then soft and medium constraints are considered in order to determine priority. Because there are often more people than available appointment slots, you must prioritize. Second dose appointments are always assigned first to avoid creating a backlog that would overwhelm your system later. After that, people are assigned based on their priority rating. Everyone starts with a priority rating that is their age. Doing this prioritizes older people over younger people. After that, people that are in specific priority groups receive, for example, a few hundred extra points. This varies based on the prioritized over younger nurses and young nurses are prioritized over people who are not nurses. The following table illustrates this concept:

Table 15.1. Priority rating table

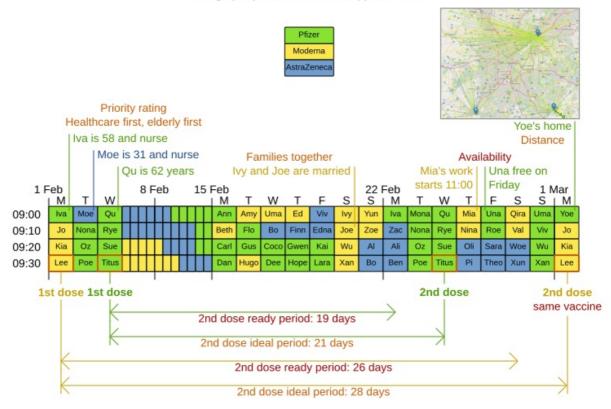
Age	Job	Priority rating
60	nurse	1060
33	nurse	1033
71	retired	71
52	office worker	52

15.1.2. The Red Hat Build of OptaPlanner solver

At the core of OptaPlanner is the solver, the engine that takes the problem data set and overlays the planning constraints and configurations. The problem data set includes all of the information about the people, the vaccines, and the vaccination centers. The solver works through the various combinations of data and eventually determines an optimized appointment schedule with people assigned to vaccination appointments at a specific center. The following illustration shows a schedule that the solver created:

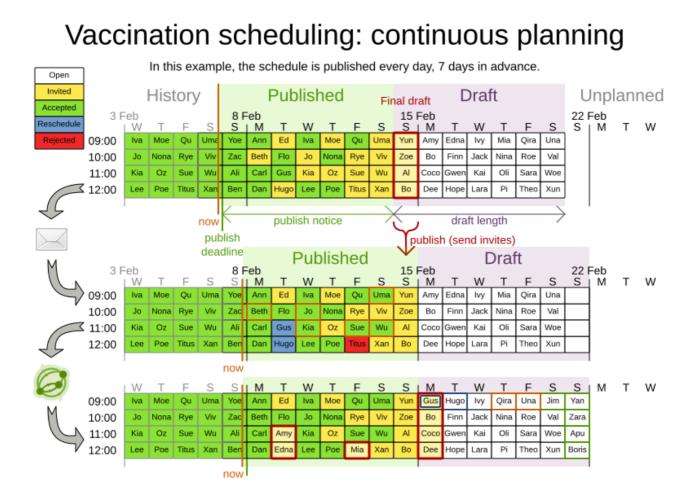
Vaccination scheduling

Assign people to vaccination appointments.



15.1.3. Continuous planning

Continuous planning is the technique of managing one or more upcoming planning periods at the same time and repeating that process monthly, weekly, daily, hourly, or even more frequently. The planning window advances incrementally by a specified interval. The following illustration shows a two week planning window that is updated daily:



The two week planning window is divided in half. The first week is in the published state and the second week is in the draft state. People are assigned to appointments in both the published and draft parts of the planning window. However, only people in the published part of the planning window are notified of their appointments. The other appointments can still change easily in the next run. Doing this gives OptaPlanner the flexibility to change the appointments in the draft part when you run the solver again, if necessary. For example, if a person who needs a second dose has a ready date of Monday and an ideal date of Wednesday, OptaPlanner does not have to give them an appointment for Monday if you can prove OptaPlanner can demonstrate that it can give them a draft appointment later in the week.

You can determine the size of the planning window but just be aware of the size of the problem space. The problem space is all of the various elements that go into creating the schedule. The more days you plan ahead, the larger the problem space.

15.1.4. Pinned planning entities

If you are continuously planning on a daily basis, there will be appointments within the two week period that are already allocated to people. To ensure that appointments are not double-booked, OptaPlanner marks existing appointments as allocated by pinning them. Pinning is used to anchor one or more specific assignments and force OptaPlanner to schedule around those fixed assignments. A pinned planning entity, such as an appointment, does not change during solving.

Whether an entity is pinned or not is determined by the appointment state. An appointment can have five states : **Open**, **Invited**, **Accepted**, **Rejected**, or **Rescheduled**.



NOTE

You do not actually see these states directly in the quick start demo code because the OptaPlanner engine is only interested in whether the appointment is pinned or not.

You need to be able to plan around appointments that have already been scheduled. An appointment with the **Invited** or **Accepted** state is pinned. Appointments with the **Open**, **Reschedule**, and **Rejected** state are not pinned and are available for scheduling.

In this example, when the solver runs it searches across the entire two week planning window in both the published and draft ranges. The solver considers any unpinned entities, appointments with the **Open**, **Reschedule**, or **Rejected** states, in addition to the unscheduled input data, to find the optimal solution. If the solver is run daily, you will see a new day added to the schedule before you run the solver.

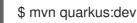
Notice that the appointments on the new day have been assigned and Amy and Edna who were previously scheduled in the draft part of the planning window are now scheduled in the published part of the window. This was possible because Gus and Hugo requested a reschedule. This will not cause any confusion because Amy and Edna were never notified about their draft dates. Now, because they have appointments in the published section of the planning window, they will be notified and asked to accept or reject their appointments, and their appointments are now pinned.

15.2. DOWNLOADING AND RUNNING THE OPTAPLANNER VACCINATION APPOINTMENT SCHEDULER

Download the OptaPlanner vaccination appointment scheduler quick start archive, start it in Quarkus development mode, and view the application in a browser. Quarkus development mode enables you to make changes and update your application while it is running.

Procedure

- 1. Navigate to the Software Downloads page in the Red Hat Customer Portal (login required), and select the product and version from the drop-down options:
 - Product: Red Hat Build of OptaPlanner
 - Version: 8.38
- 2. Download Red Hat Build of OptaPlanner 8.38 Quick Starts
- 3. Extract the **rhbop-8.38.0-optaplanner-quickstarts-sources.zip** file. The extracted **org.optaplanner.optaplanner-quickstarts-8.38.0.Final-redhat-00004/use-cases/vaccination-scheduling** directory contains example source code.
- 4. Navigate to the org.optaplanner.optaplanner-quickstarts-8.38.0.Final-redhat-00004/use-cases/vaccination-scheduling directory.
- 5. Enter the following command to start the OptaPlanner vaccination appointment scheduler in development mode:



6. To view the OptaPlanner vaccination appointment scheduler, enter the following URL in a web browser.

http://localhost:8080/

- -
- 7. To run the OptaPlanner vaccination appointment scheduler, click **Solve**.
- 8. Make changes to the source code then press the F5 key to refresh your browser. Notice that the changes that you made are now available.

15.3. PACKAGE AND RUN THE OPTAPLANNER VACCINATION APPOINTMENT SCHEDULER

When you have completed development work on the OptaPlanner vaccination appointment scheduler in **quarkus:dev** mode, run the application as a conventional jar file.

Prerequisites

• You have downloaded the OptaPlanner vaccination appointment scheduler quick start.

Procedure

- 1. Navigate to the /use-cases/vaccination-scheduling directory.
- 2. To compile the OptaPlanner vaccination appointment scheduler, enter the following command:

\$ mvn package

3. To run the compiled OptaPlanner vaccination appointment scheduler, enter the following command:

\$ java -jar ./target/quarkus-app/quarkus-run.jar



NOTE

To run the application on port 8081, add **-Dquarkus.http.port=8081** to the preceding command.

4. To start the OptaPlanner vaccination appointment scheduler, enter the following URL in a web browser.

http://localhost:8080/

15.4. ADDITIONAL RESOURCES

• Vaccination appointment scheduling video

CHAPTER 16. RED HAT BUILD OF OPTAPLANNER ON RED HAT BUILD OF QUARKUS: AN EMPLOYEE SCHEDULER QUICK START GUIDE

The employee scheduler quick start application assigns employees to shifts on various positions in an organization. For example, you can use the application to distribute shifts in a hospital between nurses, guard duty shifts across a number of locations, or shifts on an assembly line between workers.

Optimal employee scheduling must take a number of variables into account. For example, different skills can be required for shifts in different positions. Also, some employees might be unavailable for some time slots or might prefer a particular time slot. Moreover, an employee can have a contract that limits the number of hours that the employee can work in a single time period.

The Red Hat Build of OptaPlanner rules for this starter application use both hard and soft constraints. During an optimization, the Planner engine may not violate hard constraints, for example, if an employee is unavailable (out sick), or that an employee cannot work two spots in a single shift. The Planner engine tries to adhere to soft constraints, such as an employee's preference to not work a specific shift, but can violate them if the optimal solution requires it.

Prerequisites

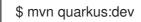
- OpenJDK 11 or later is installed. Red Hat build of Open JDK is available from the Software Downloads page in the Red Hat Customer Portal (login required).
- Apache Maven 3.8 or higher is installed. Maven is available from the Apache Maven Project website.
- An IDE, such as IntelliJ IDEA, VSCode, or Eclipse is available.

16.1. DOWNLOADING AND RUNNING THE OPTAPLANNER EMPLOYEE SCHEDULER

Download the OptaPlanner employee scheduler quick start archive, start it in Quarkus development mode, and view the application in a browser. Quarkus development mode enables you to make changes and update your application while it is running.

Procedure

- 1. Navigate to the Software Downloads page in the Red Hat Customer Portal (login required), and select the product and version from the drop-down options:
 - Product: Red Hat Build of OptaPlanner
 - Version: 8.38
- 2. Download Red Hat Build of OptaPlanner 8.38 Quick Starts
- 3. Extract the rhbop-8.38.0-optaplanner-quickstarts-sources.zip file.
- 4. Navigate to the org.optaplanner.optaplanner-quickstarts-8.38.0.Final-redhat-00004/use-cases/employee-scheduling directory.
- 5. Enter the following command to start the OptaPlanner employee scheduler in development mode:



6. To view the OptaPlanner employee scheduler, enter the following URL in a web browser.

http://localhost:8080/

- 7. To run the OptaPlanner employee scheduler, click **Solve**.
- 8. Make changes to the source code then press the F5 key to refresh your browser. Notice that the changes that you made are now available.

16.2. PACKAGE AND RUN THE OPTAPLANNER EMPLOYEE SCHEDULER

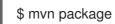
When you have completed development work on the OptaPlanner employee scheduler in **quarkus:dev** mode, run the application as a conventional jar file.

Prerequisites

• You have downloaded the OptaPlanner employee scheduling quick start.

Procedure

- 1. Navigate to the /use-cases/vaccination-scheduling directory.
- 2. To compile the OptaPlanner employee scheduler, enter the following command:



3. To run the compiled OptaPlanner employee scheduler, enter the following command:

\$ java -jar ./target/quarkus-app/quarkus-run.jar



NOTE

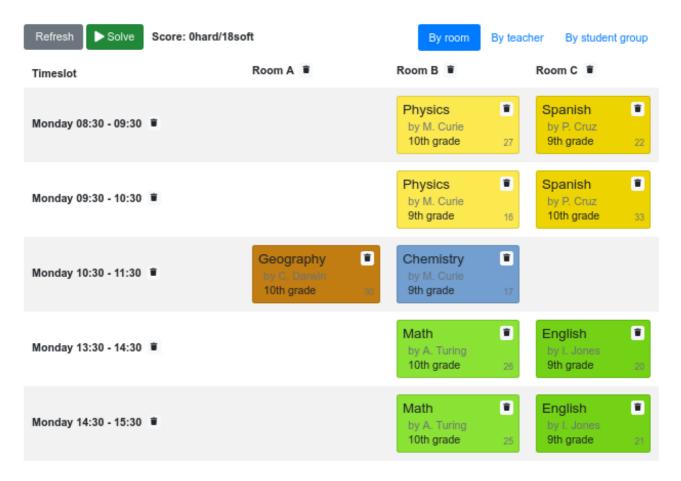
To run the application on port 8081, add **-Dquarkus.http.port=8081** to the preceding command.

4. To start the OptaPlanner employee scheduler, enter the following URL in a web browser.



CHAPTER 17. RED HAT BUILD OF OPTAPLANNER ON SPRING BOOT: A SCHOOL TIMETABLE QUICK START GUIDE

This guide walks you through the process of creating a Spring Boot application with OptaPlanner's constraint solving artificial intelligence (AI). You will build a REST application that optimizes a school timetable for students and teachers.



Your service will assign **Lesson** instances to **Timeslot** and **Room** instances automatically by using AI to adhere to the following hard and soft *scheduling constraints*:

- A room can have at most one lesson at the same time.
- A teacher can teach at most one lesson at the same time.
- A student can attend at most one lesson at the same time.
- A teacher prefers to teach in a single room.
- A teacher prefers to teach sequential lessons and dislikes gaps between lessons.

Mathematically speaking, school timetabling is an *NP-hard* problem. That means it is difficult to scale. Simply iterating through all possible combinations with brute force would take millions of years for a non-trivial data set, even on a supercomputer. Fortunately, AI constraint solvers such as OptaPlanner have advanced algorithms that deliver a near-optimal solution in a reasonable amount of time. What is considered to be a reasonable amount of time is subjective and depends on the goals of your problem.

Prerequisites

- OpenJDK 11 or later is installed. Red Hat build of Open JDK is available from the Software Downloads page in the Red Hat Customer Portal (login required).
- Apache Maven 3.8 or higher is installed. Maven is available from the Apache Maven Project website.
- An IDE, such as IntelliJ IDEA, VSCode, or Eclipse is available.

17.1. DOWNLOADING AND BUILDING THE SPRING BOOT SCHOOL TIMETABLE QUICK START

If you want to see a completed example of the school timetable project for Red Hat Build of OptaPlanner with Spring Boot product, download the starter application from the Red Hat Customer Portal.

Procedure

- 1. Navigate to the Software Downloads page in the Red Hat Customer Portal (login required), and select the product and version from the drop-down options:
 - Product: Red Hat Build of OptaPlanner
 - Version: 8.38
- 2. Download Red Hat Build of OptaPlanner 8.38 Quick Starts
- 3. Extract the **rhbop-8.38.0-optaplanner-quickstarts-sources.zip** file. The extracted **org.optaplanner.optaplanner-quickstarts-8.38.0.Final-redhat-00004/use-cases/school-timetabling** directory contains example source code.
- 4. Navigate to the org.optaplanner.optaplanner-quickstarts-8.38.0.Final-redhat-00004/use-cases/school-timetabling directory.
- 5. Download the Red Hat Build of OptaPlanner 8.38.0 Maven Repositroy(rhbop-8.38.0optaplanner-maven-repository.zip).
- 6. Extract the rhbop-8.38.0-optaplanner-maven-repository.zip file.
- 7. Copy the contents of the **rhbop-8.38.0-optaplanner/maven-repository** subdirectory into the ~/.**m2/repository** directory.
- 8. Navigate to the org.optaplanner.optaplanner-quickstarts-8.38.0.Final-redhat-00004/technology/java-spring-boot directory.
- 9. Enter the following command to build the Spring Boot school timetabling project:

mvn clean install -DskipTests

10. To build the Spring Boot school timetabling project, enter the following command:

mvn spring-boot:run -DskipTests

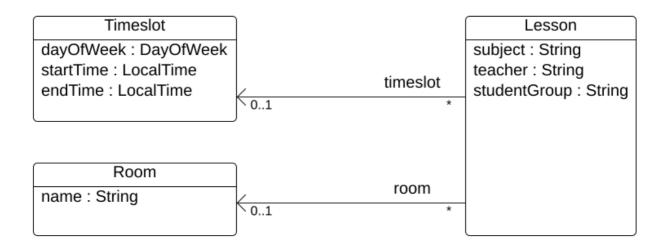
11. To view the project, enter the following URL in a web browser:

http://localhost:8080/

17.2. MODEL THE DOMAIN OBJECTS

The goal of the Red Hat Build of OptaPlanner timetable project is to assign each lesson to a time slot and a room. To do this, add three classes, **Timeslot**, **Lesson**, and **Room**, as shown in the following diagram:

Time table class diagram



Timeslot

The **Timeslot** class represents a time interval when lessons are taught, for example, **Monday 10:30** - **11:30** or **Tuesday 13:30** - **14:30**. In this example, all time slots have the same duration and there are no time slots during lunch or other breaks.

A time slot has no date because a high school schedule just repeats every week. There is no need for continuous planning. A timeslot is called a *problem fact* because no **Timeslot** instances change during solving. Such classes do not require any OptaPlanner-specific annotations.

Room

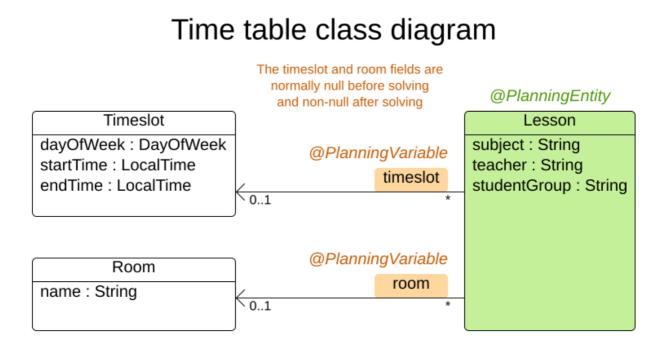
The **Room** class represents a location where lessons are taught, for example, **Room A** or **Room B**. In this example, all rooms are without capacity limits and they can accommodate all lessons.

Room instances do not change during solving so **Room** is also a *problem fact*.

Lesson

During a lesson, represented by the **Lesson** class, a teacher teaches a subject to a group of students, for example, **Math by A.Turing for 9th grade** or **Chemistry by M.Curie for 10th grade**. If a subject is taught multiple times each week by the same teacher to the same student group, there are multiple **Lesson** instances that are only distinguishable by **id**. For example, the 9th grade has six math lessons a week.

During solving, OptaPlanner changes the **timeslot** and **room** fields of the **Lesson** class to assign each lesson to a time slot and a room. Because OptaPlanner changes these fields, **Lesson** is a *planning entity*:



Most of the fields in the previous diagram contain input data, except for the orange fields. A lesson's **timeslot** and **room** fields are unassigned (**null**) in the input data and assigned (not **null**) in the output data. OptaPlanner changes these fields during solving. Such fields are called planning variables. In order for OptaPlanner to recognize them, both the **timeslot** and **room** fields require an **@PlanningVariable** annotation. Their containing class, **Lesson**, requires an **@PlanningEntity** annotation.

Procedure

1. Create the src/main/java/com/example/domain/Timeslot.java class:

```
package com.example.domain;
import java.time.DayOfWeek;
import java.time.LocalTime;
public class Timeslot {
  private DayOfWeek dayOfWeek;
  private LocalTime startTime;
  private LocalTime endTime;
  private Timeslot() {
  }
  public Timeslot(DayOfWeek dayOfWeek, LocalTime startTime, LocalTime endTime) {
    this.dayOfWeek = dayOfWeek;
    this.startTime = startTime;
    this.endTime = endTime;
  }
  @Override
  public String toString() {
    return dayOfWeek + " " + startTime.toString();
```

Notice the **toString()** method keeps the output short so it is easier to read OptaPlanner's **DEBUG** or **TRACE** log, as shown later.

2. Create the src/main/java/com/example/domain/Room.java class:

```
package com.example.domain;
public class Room {
  private String name;
  private Room() {
  }
  public Room(String name) {
    this.name = name;
  }
  @Override
  public String toString() {
    return name;
  }
  // ********
  // Getters and setters
  public String getName() {
    return name;
  }
}
```

3. Create the src/main/java/com/example/domain/Lesson.java class:

package com.example.domain;

```
import org.optaplanner.core.api.domain.entity.PlanningEntity;
import org.optaplanner.core.api.domain.variable.PlanningVariable;
@PlanningEntity
public class Lesson {
  private Long id;
  private String subject;
  private String teacher;
  private String studentGroup;
  @PlanningVariable(valueRangeProviderRefs = "timeslotRange")
  private Timeslot timeslot;
  @PlanningVariable(valueRangeProviderRefs = "roomRange")
  private Room room;
  private Lesson() {
  }
  public Lesson(Long id, String subject, String teacher, String studentGroup) {
    this.id = id;
    this.subject = subject;
    this.teacher = teacher;
    this.studentGroup = studentGroup;
  }
  @Override
  public String toString() {
    return subject + "(" + id + ")";
  }
  // *********
  // Getters and setters
  // ********
  public Long getId() {
    return id;
  }
  public String getSubject() {
    return subject;
  }
  public String getTeacher() {
    return teacher;
  }
  public String getStudentGroup() {
    return studentGroup;
  }
  public Timeslot getTimeslot() {
    return timeslot;
  }
```

```
public void setTimeslot(Timeslot timeslot) {
    this.timeslot = timeslot;
}
public Room getRoom() {
    return room;
}
public void setRoom(Room room) {
    this.room = room;
}
```

The **Lesson** class has an **@PlanningEntity** annotation, so OptaPlanner knows that this class changes during solving because it contains one or more planning variables.

The **timeslot** field has an **@PlanningVariable** annotation, so OptaPlanner knows that it can change its value. In order to find potential **Timeslot** instances to assign to this field, OptaPlanner uses the **valueRangeProviderRefs** property to connect to a value range provider that provides a **List<Timeslot>** to pick from. See Section 17.4, "Gather the domain objects in a planning solution" for information about value range providers.

The room field also has an *@PlanningVariable* annotation for the same reasons.

17.3. DEFINE THE CONSTRAINTS AND CALCULATE THE SCORE

When solving a problem, a score represents the quality of a specific solution. The higher the score the better. Red Hat Build of OptaPlanner looks for the best solution, which is the solution with the highest score found in the available time. It might be the *optimal* solution.

Because the timetable example use case has hard and soft constraints, use the **HardSoftScore** class to represent the score:

- Hard constraints must not be broken. For example: A room can have at most one lesson at the same time.
- Soft constraints should not be broken. For example: A teacher prefers to teach in a single room.

Hard constraints are weighted against other hard constraints. Soft constraints are weighted against other soft constraints. Hard constraints always outweigh soft constraints, regardless of their respective weights.

To calculate the score, you could implement an **EasyScoreCalculator** class:

```
public class TimeTableEasyScoreCalculator implements EasyScoreCalculator<TimeTable> {
```

```
@Override
public HardSoftScore calculateScore(TimeTable timeTable) {
  List<Lesson> lessonList = timeTable.getLessonList();
  int hardScore = 0;
  for (Lesson a : lessonList) {
    for (Lesson b : lessonList) {
        if (a.getTimeslot() != null && a.getTimeslot().equals(b.getTimeslot())
            && a.getId() < b.getId()) {
    }
}
</pre>
```

```
// A room can accommodate at most one lesson at the same time.
          if (a.getRoom() != null && a.getRoom().equals(b.getRoom())) {
            hardScore--;
         }
          // A teacher can teach at most one lesson at the same time.
          if (a.getTeacher().equals(b.getTeacher())) {
            hardScore--;
          }
          // A student can attend at most one lesson at the same time.
          if (a.getStudentGroup().equals(b.getStudentGroup())) {
            hardScore--;
         }
       }
    }
  }
  int softScore = 0;
  // Soft constraints are only implemented in the "complete" implementation
  return HardSoftScore.of(hardScore, softScore);
}
```

Unfortunately, this solution does not scale well because it is non-incremental: every time a lesson is assigned to a different time slot or room, all lessons are re-evaluated to calculate the new score.

A better solution is to create a **src/main/java/com/example/solver/TimeTableConstraintProvider.java** class to perform incremental score calculation. This class uses OptaPlanner's ConstraintStream APl which is inspired by Java 8 Streams and SQL. The **ConstraintProvider** scales an order of magnitude better than the **EasyScoreCalculator**: O(n) instead of O(n²).

Procedure

Create the following src/main/java/com/example/solver/TimeTableConstraintProvider.java class:

package com.example.solver;

```
import com.example.domain.Lesson;
import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore;
import org.optaplanner.core.api.score.stream.Constraint;
import org.optaplanner.core.api.score.stream.ConstraintFactory;
import org.optaplanner.core.api.score.stream.ConstraintProvider;
import org.optaplanner.core.api.score.stream.Joiners;
public class TimeTableConstraintProvider implements ConstraintProvider {
  @Override
  public Constraint[] defineConstraints(ConstraintFactory constraintFactory) {
     return new Constraint[] {
          // Hard constraints
          roomConflict(constraintFactory),
          teacherConflict(constraintFactory),
          studentGroupConflict(constraintFactory),
          // Soft constraints are only implemented in the "complete" implementation
    };
  }
```

```
private Constraint roomConflict(ConstraintFactory constraintFactory) {
  // A room can accommodate at most one lesson at the same time.
  // Select a lesson ...
  return constraintFactory.forEach(Lesson.class)
       // ... and pair it with another lesson ...
       .join(Lesson.class,
            // ... in the same timeslot ...
            Joiners.equal(Lesson::getTimeslot),
            // ... in the same room ...
            Joiners.equal(Lesson::getRoom),
            // ... and the pair is unique (different id, no reverse pairs)
            Joiners.lessThan(Lesson::getId))
       // then penalize each pair with a hard weight.
       .penalize(HardSoftScore.ONE_HARD)
       .asConstraint("Room conflict");
}
private Constraint teacherConflict(ConstraintFactory constraintFactory) {
  // A teacher can teach at most one lesson at the same time.
  return constraintFactory.forEach(Lesson.class)
       .join(Lesson.class,
            Joiners.equal(Lesson::getTimeslot),
            Joiners.equal(Lesson::getTeacher),
            Joiners.lessThan(Lesson::getId))
            .penalize(HardSoftScore.ONE_HARD)
            .asConstraint("Teacher conflict");
}
private Constraint studentGroupConflict(ConstraintFactory constraintFactory) {
  // A student can attend at most one lesson at the same time.
  return constraintFactory.forEach(Lesson.class)
       .join(Lesson.class,
            Joiners.equal(Lesson::getTimeslot),
            Joiners.equal(Lesson::getStudentGroup),
            Joiners.lessThan(Lesson::getId))
            .penalize(HardSoftScore.ONE HARD)
            .asConstraint("Student group conflict");
}
```

17.4. GATHER THE DOMAIN OBJECTS IN A PLANNING SOLUTION

A **TimeTable** instance wraps all **Timeslot**, **Room**, and **Lesson** instances of a single dataset. Furthermore, because it contains all lessons, each with a specific planning variable state, it is a *planning solution* and it has a score:

- If lessons are still unassigned, then it is an *uninitialized* solution, for example, a solution with the score **-4init/Ohard/Osoft**.
- If it breaks hard constraints, then it is an *infeasible* solution, for example, a solution with the score **-2hard/-3soft**.

• If it adheres to all hard constraints, then it is a *feasible* solution, for example, a solution with the score **Ohard/-7soft**.

The **TimeTable** class has an **@PlanningSolution** annotation, so Red Hat Build of OptaPlanner knows that this class contains all of the input and output data.

Specifically, this class is the input of the problem:

- A timeslotList field with all time slots
 - This is a list of problem facts, because they do not change during solving.
- A roomList field with all rooms
 - This is a list of problem facts, because they do not change during solving.
- A lessonList field with all lessons
 - This is a list of planning entities because they change during solving.
 - Of each **Lesson**:
 - The values of the timeslot and room fields are typically still null, so unassigned. They are planning variables.
 - The other fields, such as **subject**, **teacher** and **studentGroup**, are filled in. These fields are problem properties.

However, this class is also the output of the solution:

- A lessonList field for which each Lesson instance has non-null timeslot and room fields after solving
- A score field that represents the quality of the output solution, for example, **0hard/-5soft**

Procedure

Create the src/main/java/com/example/domain/TimeTable.java class:

package com.example.domain;

import java.util.List;

import org.optaplanner.core.api.domain.solution.PlanningEntityCollectionProperty; import org.optaplanner.core.api.domain.solution.PlanningScore; import org.optaplanner.core.api.domain.solution.PlanningSolution; import org.optaplanner.core.api.domain.solution.ProblemFactCollectionProperty; import org.optaplanner.core.api.domain.valuerange.ValueRangeProvider; import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore;

```
@PlanningSolution
public class TimeTable {
```

@ValueRangeProvider(id = "timeslotRange") @ProblemFactCollectionProperty private List<Timeslot> timeslotList;

@ValueRangeProvider(id = "roomRange")

```
@ProblemFactCollectionProperty
private List<Room> roomList;
@PlanningEntityCollectionProperty
private List<Lesson> lessonList;
@PlanningScore
private HardSoftScore score;
private TimeTable() {
}
public TimeTable(List<Timeslot> timeslotList, List<Room> roomList,
    List<Lesson> lessonList) {
  this.timeslotList = timeslotList;
  this.roomList = roomList;
  this.lessonList = lessonList;
}
// *******
// Getters and setters
// ********
public List<Timeslot> getTimeslotList() {
  return timeslotList;
}
public List<Room> getRoomList() {
  return roomList;
}
public List<Lesson> getLessonList() {
  return lessonList;
}
public HardSoftScore getScore() {
  return score;
}
```

The value range providers

The **timeslotList** field is a value range provider. It holds the **Timeslot** instances which OptaPlanner can pick from to assign to the **timeslot** field of **Lesson** instances. The **timeslotList** field has an **@ValueRangeProvider** annotation to connect those two, by matching the **id** with the **valueRangeProviderRefs** of the **@PlanningVariable** in the **Lesson**.

Following the same logic, the **roomList** field also has an **@ValueRangeProvider** annotation.

The problem fact and planning entity properties

Furthermore, OptaPlanner needs to know which **Lesson** instances it can change as well as how to retrieve the **Timeslot** and **Room** instances used for score calculation by your **TimeTableConstraintProvider**.

The timeslotList and roomList fields have an @ProblemFactCollectionProperty annotation, so your TimeTableConstraintProvider can select from those instances.

The **lessonList** has an **@PlanningEntityCollectionProperty** annotation, so OptaPlanner can change them during solving and your **TimeTableConstraintProvider** can select from those too.

17.5. CREATE THE TIMETABLE SERVICE

Now you are ready to put everything together and create a REST service. But solving planning problems on REST threads causes HTTP timeout issues. Therefore, the Spring Boot starter injects a **SolverManager**, which runs solvers in a separate thread pool and can solve multiple datasets in parallel.

Procedure

Create the src/main/java/com/example/solver/TimeTableController.java class:

package com.example.solver;

```
import java.util.UUID;
import java.util.concurrent.ExecutionException;
```

```
import com.example.domain.TimeTable;
import org.optaplanner.core.api.solver.SolverJob;
import org.optaplanner.core.api.solver.SolverManager;
import org.springframework.beans.factory.annotation.Autowired;
import org.springframework.web.bind.annotation.PostMapping;
import org.springframework.web.bind.annotation.RequestBody;
import org.springframework.web.bind.annotation.RequestMapping;
import org.springframework.web.bind.annotation.RequestMapping;
```

```
@RestController
@RequestMapping("/timeTable")
public class TimeTableController {
```

```
@Autowired
private SolverManager<TimeTable, UUID> solverManager;
```

```
@PostMapping("/solve")
public TimeTable solve(@RequestBody TimeTable problem) {
    UUID problemId = UUID.randomUUID();
    // Submit the problem to start solving
    SolverJob<TimeTable, UUID> solverJob = solverManager.solve(problemId, problem);
    TimeTable solution;
    try {
        // Wait until the solving ends
        solution = solverJob.getFinalBestSolution();
    } catch (InterruptedException | ExecutionException e) {
        throw new IllegalStateException("Solving failed.", e);
    }
    return solution;
}
```

In this example, the initial implementation waits for the solver to finish, which can still cause an HTTP timeout. The *complete* implementation avoids HTTP timeouts much more elegantly.

17.6. SET THE SOLVER TERMINATION TIME

If your planning application does not have a termination setting or a termination event, it theoretically runs forever and in reality eventually causes an HTTP timeout error. To prevent this from occurring, use the **optaplanner.solver.termination.spent-limit** parameter to specify the length of time after which the application terminates. In most applications, set the time to at least five minutes (**5m**). However, in the Timetable example, limit the solving time to five seconds, which is short enough to avoid the HTTP timeout.

Procedure

Create the src/main/resources/application.properties file with the following content:

quarkus.optaplanner.solver.termination.spent-limit=5s

17.7. MAKE THE APPLICATION EXECUTABLE

After you complete the Red Hat Build of OptaPlanner Spring Boot timetable project, package everything into a single executable JAR file driven by a standard Java **main()** method.

Prerequisites

• You have a completed OptaPlanner Spring Boot timetable project.

Procedure

1. Create the TimeTableSpringBootApp.java class with the following content:

```
package com.example;
import org.springframework.boot.SpringApplication;
import org.springframework.boot.autoconfigure.SpringBootApplication;
@SpringBootApplication
public class TimeTableSpringBootApp {
    public static void main(String[] args) {
        SpringApplication.run(TimeTableSpringBootApp.class, args);
    }
}
```

- 2. Replace the **src/main/java/com/example/DemoApplication.java** class created by Spring Initializr with the **TimeTableSpringBootApp.java** class.
- 3. Run the TimeTableSpringBootApp.java class as the main class of a regular Java application.

17.7.1. Try the timetable application

After you start the Red Hat Build of OptaPlanner Spring Boot timetable application, you can test the REST service with any REST client that you want. This example uses the Linux **curl** command to send a POST request.

Prerequisites

• The OptaPlanner Spring Boot timetable application is running.

Procedure

Enter the following command:

\$ curl -i -X POST http://localhost:8080/timeTable/solve -H "Content-Type:application/json" -d
'{"timeslotList":[{"dayOfWeek":"MONDAY","startTime":"08:30:00","endTime":"09:30:00"},
{"dayOfWeek":"MONDAY","startTime":"09:30:00","endTime":"10:30:00"}],"roomList":[{"name":"Room
A"},{"name":"Room B"}],"lessonList":[{"id":1,"subject":"Math","teacher":"A. Turing","studentGroup":"9th
grade"},{"id":2,"subject":"Chemistry","teacher":"M. Curie","studentGroup":"9th grade"},
{"id":3,"subject":"French","teacher":"I. Jones","studentGroup":"10th grade"}]}'

After about five seconds, the termination spent time defined in **application.properties**, the service returns an output similar to the following example:

HTTP/1.1 200 Content-Type: application/json

{"timeslotList":...,"roomList":...,"lessonList":[{"id":1,"subject":"Math","teacher":"A. Turing","studentGroup":"9th grade","timeslot":

{"dayOfWeek":"MONDAY","startTime":"08:30:00","endTime":"09:30:00"},"room":{"name":"Room A"}}, {"id":2,"subject":"Chemistry","teacher":"M. Curie","studentGroup":"9th grade","timeslot": {"dayOfWeek":"MONDAY","startTime":"09:30:00","endTime":"10:30:00"},"room":{"name":"Room A"}}, {"id":3,"subject":"French","teacher":"M. Curie","studentGroup":"10th grade","timeslot": {"dayOfWeek":"MONDAY","startTime":"08:30:00","endTime":"09:30:00"},"room":{"name":"Room A"}},

{"dayOfWeek":"MONDAY", "start lime":"08:30:00", "end lime":"09:30:00"}, "room":{"name":"Room B"}}, {"id":4, "subject":"History", "teacher":"I. Jones", "studentGroup":"10th grade", "timeslot":

{"dayOfWeek":"MONDAY","startTime":"09:30:00","endTime":"10:30:00"},"room":{"name":"Room B"}}],"score":"0hard/0soft"}

Notice that the application assigned all four lessons to one of the two time slots and one of the two rooms. Also notice that it conforms to all hard constraints. For example, M. Curie's two lessons are in different time slots.

On the server side, the **info** log shows what OptaPlanner did in those five seconds:

... Solving started: time spent (33), best score (-8init/0hard/0soft), environment mode (REPRODUCIBLE), random (JDK with seed 0).

... Construction Heuristic phase (0) ended: time spent (73), best score (0hard/0soft), score calculation speed (459/sec), step total (4).

... Local Search phase (1) ended: time spent (5000), best score (0hard/0soft), score calculation speed (28949/sec), step total (28398).

... Solving ended: time spent (5000), best score (0hard/0soft), score calculation speed (28524/sec), phase total (2), environment mode (REPRODUCIBLE).

17.7.2. Test the application

A good application includes test coverage. This example tests the Timetable Red Hat Build of OptaPlanner Spring Boot application. It uses a JUnit test to generate a test dataset and send it to the **TimeTableController** to solve.

Procedure

Create the **src/test/java/com/example/solver/TimeTableControllerTest.java** class with the following content:

package com.example.solver;

import java.time.DayOfWeek; import java.time.LocalTime; import java.util.ArrayList; import java.util.List;

import com.example.domain.Lesson; import com.example.domain.Room; import com.example.domain.TimeTable; import com.example.domain.Timeslot; import org.junit.jupiter.api.Test; import org.junit.jupiter.api.Timeout; import org.springframework.beans.factory.annotation.Autowired; import org.springframework.boot.test.context.SpringBootTest;

import static org.junit.jupiter.api.Assertions.assertFalse; import static org.junit.jupiter.api.Assertions.assertNotNull; import static org.junit.jupiter.api.Assertions.assertTrue;

@SpringBootTest(properties = {

"optaplanner.solver.termination.spent-limit=1h", // Effectively disable this termination in favor of the best-score-limit

```
"optaplanner.solver.termination.best-score-limit=0hard/*soft"})
```

public class TimeTableControllerTest {

```
@Autowired
private TimeTableController timeTableController;
```

```
@Test
  @Timeout(600 000)
  public void solve() {
    TimeTable problem = generateProblem();
    TimeTable solution = timeTableController.solve(problem);
    assertFalse(solution.getLessonList().isEmpty());
    for (Lesson lesson : solution.getLessonList()) {
       assertNotNull(lesson.getTimeslot());
       assertNotNull(lesson.getRoom());
    }
    assertTrue(solution.getScore().isFeasible());
  }
  private TimeTable generateProblem() {
    List<Timeslot> timeslotList = new ArrayList<>();
    timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(8, 30), LocalTime.of(9,
30)));
    timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9, 30), LocalTime.of(10,
30)));
    timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(10, 30), LocalTime.of(11,
30)));
    timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(13, 30), LocalTime.of(14,
30)));
```

timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(14, 30), LocalTime.of(15,

```
30)));
     List<Room> roomList = new ArrayList<>();
     roomList.add(new Room("Room A"));
     roomList.add(new Room("Room B"));
     roomList.add(new Room("Room C"));
     List<Lesson> lessonList = new ArrayList<>();
     lessonList.add(new Lesson(101L, "Math", "B. May", "9th grade"));
     lessonList.add(new Lesson(102L, "Physics", "M. Curie", "9th grade"));
     lessonList.add(new Lesson(103L, "Geography", "M. Polo", "9th grade"));
     lessonList.add(new Lesson(104L, "English", "I. Jones", "9th grade"));
     lessonList.add(new Lesson(105L, "Spanish", "P. Cruz", "9th grade"));
     lessonList.add(new Lesson(201L, "Math", "B. May", "10th grade"));
     lessonList.add(new Lesson(202L, "Chemistry", "M. Curie", "10th grade"));
     lessonList.add(new Lesson(203L, "History", "I. Jones", "10th grade"));
    lessonList.add(new Lesson(204L, "English", "P. Cruz", "10th grade"));
     lessonList.add(new Lesson(205L, "French", "M. Curie", "10th grade"));
     return new TimeTable(timeslotList, roomList, lessonList);
  }
```

This test verifies that after solving, all lessons are assigned to a time slot and a room. It also verifies that it found a feasible solution (no hard constraints broken).

Normally, the solver finds a feasible solution in less than 200 milliseconds. Notice how the **@SpringBootTest** annotation's **properties** overwrites the solver termination to terminate as soon as a feasible solution (**0hard**/***soft**) is found. This avoids hard coding a solver time, because the unit test might run on arbitrary hardware. This approach ensures that the test runs long enough to find a feasible solution, even on slow systems. However, it does not run a millisecond longer than it strictly must, even on fast systems.

17.7.3. Logging

After you complete the Red Hat Build of OptaPlanner Spring Boot timetable application, you can use logging information to help you fine-tune the constraints in the **ConstraintProvider**. Review the score calculation speed in the **info** log file to assess the impact of changes to your constraints. Run the application in debug mode to show every step that your application takes or use trace logging to log every step and every move.

Procedure

- 1. Run the timetable application for a fixed amount of time, for example, five minutes.
- 2. Review the score calculation speed in the **log** file as shown in the following example:

... Solving ended: ..., score calculation speed (29455/sec), ...

- 3. Change a constraint, run the planning application again for the same amount of time, and review the score calculation speed recorded in the **log** file.
- 4. Run the application in debug mode to log every step:

- To run debug mode from the command line, use the **-D** system property.
- To change logging in the **application.properties** file, add the following line to that file:

logging.level.org.optaplanner=debug

The following example shows output in the **log** file in debug mode:

... Solving started: time spent (67), best score (-20init/0hard/0soft), environment mode (REPRODUCIBLE), random (JDK with seed 0). ... CH step (0), time spent (128), score (-18init/0hard/0soft), selected move count (15), picked move ([Math(101) {null -> Room A}, Math(101) {null -> MONDAY 08:30}]). ... CH step (1), time spent (145), score (-16init/0hard/0soft), selected move count (15), picked move ([Physics(102) {null -> Room A}, Physics(102) {null -> MONDAY 09:30}]). ...

5. Use **trace** logging to show every step and every move for each step.

17.8. ADD DATABASE AND UI INTEGRATION

After you create the Red Hat Build of OptaPlanner application example with Spring Boot, add database and UI integration.

Prerequisite

• You have created the OptaPlanner Spring Boot timetable example.

Procedure

- 1. Create Java Persistence API (JPA) repositories for **Timeslot**, **Room**, and **Lesson**. For information about creating JPA repositories, see *Accessing Data with JPA* on the Spring website.
- 2. Expose the JPA repositories through REST. For information about exposing the repositories, see *Accessing JPA Data with REST* on the Spring website.
- 3. Build a **TimeTableRepository** facade to read and write a **TimeTable** in a single transaction.
- 4. Adjust the **TimeTableController** as shown in the following example:

package com.example.solver;

import com.example.domain.TimeTable; import com.example.persistence.TimeTableRepository; import org.optaplanner.core.api.score.ScoreManager; import org.optaplanner.core.api.solver.SolverManager; import org.optaplanner.core.api.solver.SolverStatus; import org.springframework.beans.factory.annotation.Autowired; import org.springframework.web.bind.annotation.GetMapping; import org.springframework.web.bind.annotation.PostMapping; import org.springframework.web.bind.annotation.RequestMapping; import org.springframework.web.bind.annotation.RestController;

@RestController

```
@RequestMapping("/timeTable")
public class TimeTableController {
  @Autowired
  private TimeTableRepository timeTableRepository;
  @Autowired
  private SolverManager<TimeTable, Long> solverManager;
  @Autowired
  private ScoreManager<TimeTable> scoreManager;
  // To try, GET http://localhost:8080/timeTable
  @GetMapping()
  public TimeTable getTimeTable() {
    // Get the solver status before loading the solution
    // to avoid the race condition that the solver terminates between them
    SolverStatus solverStatus = getSolverStatus();
    TimeTable solution =
timeTableRepository.findById(TimeTableRepository.SINGLETON TIME TABLE ID);
    scoreManager.updateScore(solution); // Sets the score
    solution.setSolverStatus(solverStatus);
    return solution;
  }
  @PostMapping("/solve")
  public void solve() {
    solverManager.solveAndListen(TimeTableRepository.SINGLETON_TIME_TABLE_ID,
         timeTableRepository::findById,
         timeTableRepository::save);
  }
  public SolverStatus getSolverStatus() {
    return
solverManager.getSolverStatus(TimeTableRepository.SINGLETON TIME TABLE ID);
  }
  @PostMapping("/stopSolving")
  public void stopSolving() {
    solverManager.terminateEarly(TimeTableRepository.SINGLETON_TIME_TABLE_ID);
  }
}
```

For simplicity, this code handles only one **TimeTable** instance, but it is straightforward to enable multi-tenancy and handle multiple **TimeTable** instances of different high schools in parallel.

The **getTimeTable()** method returns the latest timetable from the database. It uses the **ScoreManager** (which is automatically injected) to calculate the score of that timetable so the UI can show the score.

The **solve()** method starts a job to solve the current timetable and store the time slot and room assignments in the database. It uses the **SolverManager.solveAndListen()** method to listen to intermediate best solutions and update the database accordingly. This enables the UI to show progress while the backend is still solving.

5. Now that the **solve()** method returns immediately, adjust the **TimeTableControllerTest** as shown in the following example:

package com.example.solver;

```
import com.example.domain.Lesson;
import com.example.domain.TimeTable;
import org.junit.jupiter.api.Test;
import org.junit.jupiter.api.Timeout;
import org.optaplanner.core.api.solver.SolverStatus;
import org.springframework.beans.factory.annotation.Autowired;
import org.springframework.boot.test.context.SpringBootTest;
import static org.junit.jupiter.api.Assertions.assertFalse;
import static org.junit.jupiter.api.Assertions.assertNotNull;
import static org.junit.jupiter.api.Assertions.assertTrue;
@SpringBootTest(properties = {
     "optaplanner.solver.termination.spent-limit=1h", // Effectively disable this termination in
favor of the best-score-limit
     "optaplanner.solver.termination.best-score-limit=0hard/*soft"})
public class TimeTableControllerTest {
  @Autowired
  private TimeTableController timeTableController;
  @Test
  @Timeout(600 000)
  public void solveDemoDataUntilFeasible() throws InterruptedException {
     timeTableController.solve();
     TimeTable timeTable = timeTableController.getTimeTable();
    while (timeTable.getSolverStatus() != SolverStatus.NOT_SOLVING) {
       // Quick polling (not a Test Thread Sleep anti-pattern)
       // Test is still fast on fast systems and doesn't randomly fail on slow systems.
       Thread.sleep(20L);
       timeTable = timeTableController.getTimeTable();
    }
     assertFalse(timeTable.getLessonList().isEmpty());
    for (Lesson lesson : timeTable.getLessonList()) {
       assertNotNull(lesson.getTimeslot());
       assertNotNull(lesson.getRoom());
    }
     assertTrue(timeTable.getScore().isFeasible());
```

- 6. Poll for the latest solution until the solver finishes solving.
- 7. To visualize the timetable, build an attractive web UI on top of these REST methods.

17.9. USING MICROMETER AND PROMETHEUS TO MONITOR YOUR SCHOOL TIMETABLE OPTAPLANNER SPRING BOOT APPLICATION

OptaPlanner exposes metrics through Micrometer, a metrics instrumentation library for Java applications. You can use Micrometer with Prometheus to monitor the OptaPlanner solver in the school timetable application.

Prerequisites

- You have created the Spring Boot OptaPlanner school timetable application.
- Prometheus is installed. For information about installing Prometheus, see the Prometheus website.

Procedure

- 1. Navigate to the **technology/java-spring-boot** directory.
- 2. Add the Micrometer Prometheus dependencies to the school timetable **pom.xml** file:

<dependency> <groupId>org.springframework.boot</groupId> <artifactId>spring-boot-starter-actuator</artifactId> </dependency> <groupId>io.micrometer</groupId> <artifactId>micrometer-registry-prometheus</artifactId> </dependency>

3. Add the following property to the application.properties file:

management.endpoints.web.exposure.include=metrics,prometheus

4. Start the school timetable application:

mvn spring-boot:run

5. Open http://localhost:8080/actuator/prometheus in a web browser.

CHAPTER 18. RED HAT BUILD OF OPTAPLANNER AND JAVA: A SCHOOL TIMETABLE QUICKSTART GUIDE

This guide walks you through the process of creating a simple Java application with the OptaPlanner constraint solving artificial intelligence (AI). You will build a command-line application that optimizes a school timetable for students and teachers:

INFO Solving ended: time spent (5000), best score (0hard/9soft),
INFO
INFO Room A Room B Room C
INFO
INFO MON 08:30 English Math
INFO I. Jones A. Turing
INFO 9th grade 10th grade
INFO
INFO MON 09:30 History Physics
INFO I. Jones M. Curie
INFO 9th grade 10th grade
INFO
INFO MON 10:30 History Physics
INFO I. Jones M. Curie
INFO 10th grade 9th grade
INFO
INFO

Your application will assign **Lesson** instances to **Timeslot** and **Room** instances automatically by using AI to adhere to hard and soft scheduling *constraints*, for example:

- A room can have at most one lesson at the same time.
- A teacher can teach at most one lesson at the same time.
- A student can attend at most one lesson at the same time.
- A teacher prefers to teach all lessons in the same room.
- A teacher prefers to teach sequential lessons and dislikes gaps between lessons.
- A student dislikes sequential lessons on the same subject.

Mathematically speaking, school timetabling is an *NP-hard* problem. This means it is difficult to scale. Simply brute force iterating through all possible combinations takes millions of years for a non-trivial data set, even on a supercomputer. Fortunately, AI constraint solvers such as OptaPlanner have advanced algorithms that deliver a near-optimal solution in a reasonable amount of time.

Prerequisites

- OpenJDK (JDK) 11 is installed. Red Hat build of Open JDK is available from the Software Downloads page in the Red Hat Customer Portal (login required).
- Apache Maven 3.6 or higher is installed. Maven is available from the Apache Maven Project website.

• An IDE, such as IntelliJ IDEA, VSCode or Eclipse

18.1. CREATING THE MAVEN OR GRADLE BUILD FILE AND ADD DEPENDENCIES

You can use Maven or Gradle for the OptaPlanner school timetable application. After you create the build files, add the following dependencies:

- optaplanner-core (compile scope) to solve the school timetable problem
- optaplanner-test (test scope) to JUnit test the school timetabling constraints
- An implementation such as **logback-classic** (runtime scope) to view the steps that OptaPlanner takes

Procedure

- 1. Create the Maven or Gradle build file.
- 2. Add optaplanner-core, optaplanner-test, and logback-classic dependencies to your build file:
 - For Maven, add the following dependencies to the **pom.xml** file:

<dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-core</artifactId> </dependency>

<groupId>org.optaplanner</groupId>
<artifactId>optaplanner-test</artifactId>
<scope>test</scope>
</dependency>

<dependency> <groupId>ch.qos.logback</groupId> <artifactId>logback-classic</artifactId> <version>1.2.3</version> </dependency>

The following example shows the complete **pom.xml** file.

```
<?xml version="1.0" encoding="UTF-8"?>
<project xmlns="http://maven.apache.org/POM/4.0.0"
xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance"
xsi:schemaLocation="http://maven.apache.org/POM/4.0.0
https://maven.apache.org/xsd/maven-4.0.0.xsd">
<modelCersion="http://maven.apache.org/POM/4.0.0
https://maven.apache.org/xsd/maven-4.0.0.xsd">
<modelCersion="http://maven.apache.org/POM/4.0.0
https://maven.apache.org/xsd/maven-4.0.0.xsd">
<modelCersion="http://maven.apache.org/POM/4.0.0
https://maven.apache.org/xsd/maven-4.0.0.xsd">
<modelCersion="http://maven.apache.org/POM/4.0.0"
<modelCersion>4.0.0</modelCersion>
<groupId>org.acme</groupId>
<artifactId>optapIanner-hello-world-school-timetabling-quickstart</artifactId>
<version>1.0-SNAPSHOT</version>
<properties>
<maven.compiler.release>11</maven.compiler.release>
```

<project.build.sourceEncoding>UTF-8</project.build.sourceEncoding>

<version.org.optaplanner>8.38.0.Final-redhat-00004</version.org.optaplanner> <version.org.logback>1.2.3</version.org.logback>

<version.compiler.plugin>3.8.1</version.compiler.plugin> <version.surefire.plugin>3.0.0-M5</version.surefire.plugin> <version.exec.plugin>3.0.0</version.exec.plugin> </properties>

<dependencyManagement>

<dependencies> <dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-bom</artifactId> <version>\${version.org.optaplanner}</version> <type>pom</type> <scope>import</scope> </dependency>

<dependency> <groupId>ch.qos.logback</groupId> <artifactId>logback-classic</artifactId> <version>\${version.org.logback}</version> </dependency>

</dependencies> </dependencyManagement>

<dependencies>

<dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-core</artifactId> </dependency> <dependency> <groupId>ch.qos.logback</groupId> <artifactId>logback-classic</artifactId> <scope>runtime</scope> </dependency>

<!-- Testing -->

<dependency> <groupId>org.optaplanner</groupId> <artifactId>optaplanner-test</artifactId> <scope>test</scope> </dependency> </dependencies>

<build>

<plugins>

<plugin>

<artifactId>maven-compiler-plugin</artifactId> <version>\${version.compiler.plugin}</version> </plugin>

<plugin>

<artifactId>maven-surefire-plugin</artifactId> <version>\${version.surefire.plugin}</version> </plugin>



• For Gradle, add the following dependencies to the gradle.build file:

dependencies {
 implementation platform("org.optaplanner:optaplanner-bom:\${optaplannerVersion}")
 implementation "org.optaplanner:optaplanner-core"
 testImplementation "org.optaplanner:optaplanner-test"
 runtimeOnly "ch.qos.logback:logback-classic:\${logbackVersion}"

}

The following example shows the completed gradle.build file.

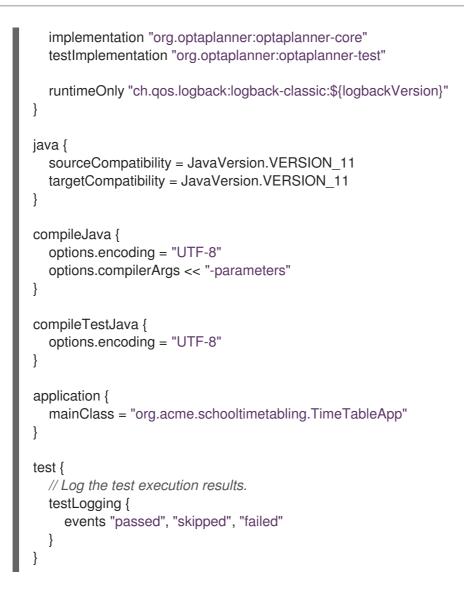
```
plugins {
    id "java"
    id "application"
}

def optaplannerVersion = "{optaplanner-version}"
    def logbackVersion = "1.2.9"

group = "org.acme"
version = "1.0-SNAPSHOT"

repositories {
    mavenCentral()
}

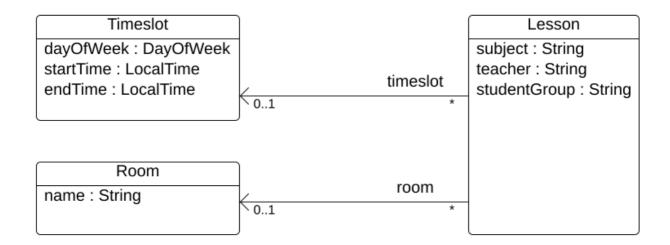
dependencies {
    implementation platform("org.optaplanner:optaplanner-bom:${optaplannerVersion}")
```



18.2. MODEL THE DOMAIN OBJECTS

The goal of the Red Hat Build of OptaPlanner timetable project is to assign each lesson to a time slot and a room. To do this, add three classes, **Timeslot**, **Lesson**, and **Room**, as shown in the following diagram:

Time table class diagram



Timeslot

The **Timeslot** class represents a time interval when lessons are taught, for example, **Monday 10:30 -11:30** or **Tuesday 13:30 - 14:30**. In this example, all time slots have the same duration and there are no time slots during lunch or other breaks.

A time slot has no date because a high school schedule just repeats every week. There is no need for continuous planning. A timeslot is called a *problem fact* because no **Timeslot** instances change during solving. Such classes do not require any OptaPlanner-specific annotations.

Room

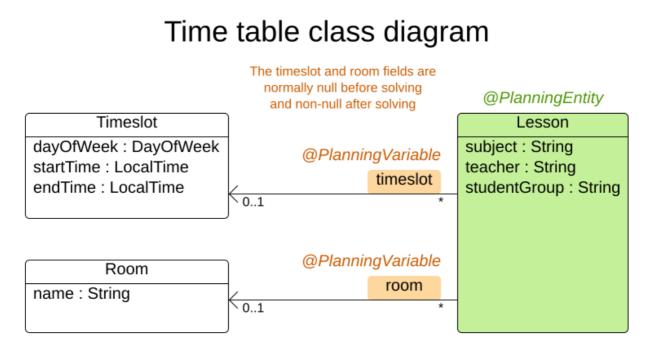
The **Room** class represents a location where lessons are taught, for example, **Room A** or **Room B**. In this example, all rooms are without capacity limits and they can accommodate all lessons.

Room instances do not change during solving so **Room** is also a *problem fact*.

Lesson

During a lesson, represented by the **Lesson** class, a teacher teaches a subject to a group of students, for example, **Math by A.Turing for 9th grade** or **Chemistry by M.Curie for 10th grade**. If a subject is taught multiple times each week by the same teacher to the same student group, there are multiple **Lesson** instances that are only distinguishable by **id**. For example, the 9th grade has six math lessons a week.

During solving, OptaPlanner changes the **timeslot** and **room** fields of the **Lesson** class to assign each lesson to a time slot and a room. Because OptaPlanner changes these fields, **Lesson** is a *planning entity*:



Most of the fields in the previous diagram contain input data, except for the orange fields. A lesson's **timeslot** and **room** fields are unassigned (**null**) in the input data and assigned (not **null**) in the output data. OptaPlanner changes these fields during solving. Such fields are called planning variables. In order for OptaPlanner to recognize them, both the **timeslot** and **room** fields require an **@PlanningVariable** annotation. Their containing class, **Lesson**, requires an **@PlanningEntity** annotation.

Procedure

1. Create the src/main/java/com/example/domain/Timeslot.java class:

```
package com.example.domain;
import java.time.DayOfWeek;
import java.time.LocalTime;
public class Timeslot {
  private DayOfWeek dayOfWeek;
  private LocalTime startTime;
  private LocalTime endTime;
  private Timeslot() {
  }
  public Timeslot(DayOfWeek dayOfWeek, LocalTime startTime, LocalTime endTime) {
    this.dayOfWeek = dayOfWeek;
    this.startTime = startTime;
    this.endTime = endTime;
  }
  @Override
  public String toString() {
    return dayOfWeek + " " + startTime.toString();
```



Notice the **toString()** method keeps the output short so it is easier to read OptaPlanner's **DEBUG** or **TRACE** log, as shown later.

2. Create the src/main/java/com/example/domain/Room.java class:

```
package com.example.domain;
public class Room {
  private String name;
  private Room() {
  }
  public Room(String name) {
    this.name = name;
  }
  @Override
  public String toString() {
    return name;
  }
  // ********
  // Getters and setters
  public String getName() {
    return name;
  }
}
```

3. Create the src/main/java/com/example/domain/Lesson.java class:

package com.example.domain;

```
import org.optaplanner.core.api.domain.entity.PlanningEntity;
import org.optaplanner.core.api.domain.variable.PlanningVariable;
@PlanningEntity
public class Lesson {
  private Long id;
  private String subject;
  private String teacher;
  private String studentGroup;
  @PlanningVariable(valueRangeProviderRefs = "timeslotRange")
  private Timeslot timeslot;
  @PlanningVariable(valueRangeProviderRefs = "roomRange")
  private Room room;
  private Lesson() {
  }
  public Lesson(Long id, String subject, String teacher, String studentGroup) {
    this.id = id;
    this.subject = subject;
    this.teacher = teacher;
    this.studentGroup = studentGroup;
  }
  @Override
  public String toString() {
    return subject + "(" + id + ")";
  }
  // *********
  // Getters and setters
  public Long getId() {
    return id;
  }
  public String getSubject() {
    return subject;
  }
  public String getTeacher() {
    return teacher;
  }
  public String getStudentGroup() {
    return studentGroup;
  }
  public Timeslot getTimeslot() {
    return timeslot;
  }
```

```
public void setTimeslot(Timeslot timeslot) {
    this.timeslot = timeslot;
}
public Room getRoom() {
    return room;
}
public void setRoom(Room room) {
    this.room = room;
}
```

The **Lesson** class has an **@PlanningEntity** annotation, so OptaPlanner knows that this class changes during solving because it contains one or more planning variables.

The **timeslot** field has an **@PlanningVariable** annotation, so OptaPlanner knows that it can change its value. In order to find potential **Timeslot** instances to assign to this field, OptaPlanner uses the **valueRangeProviderRefs** property to connect to a value range provider that provides a **List<Timeslot>** to pick from. See Section 18.4, "Gather the domain objects in a planning solution" for information about value range providers.

The **room** field also has an **@PlanningVariable** annotation for the same reasons.

18.3. DEFINE THE CONSTRAINTS AND CALCULATE THE SCORE

When solving a problem, a score represents the quality of a specific solution. The higher the score the better. Red Hat Build of OptaPlanner looks for the best solution, which is the solution with the highest score found in the available time. It might be the *optimal* solution.

Because the timetable example use case has hard and soft constraints, use the **HardSoftScore** class to represent the score:

- Hard constraints must not be broken. For example: A room can have at most one lesson at the same time.
- Soft constraints should not be broken. For example: A teacher prefers to teach in a single room.

Hard constraints are weighted against other hard constraints. Soft constraints are weighted against other soft constraints. Hard constraints always outweigh soft constraints, regardless of their respective weights.

To calculate the score, you could implement an EasyScoreCalculator class:

```
public class TimeTableEasyScoreCalculator implements EasyScoreCalculator<TimeTable> {
```

```
@Override
public HardSoftScore calculateScore(TimeTable timeTable) {
  List<Lesson> lessonList = timeTable.getLessonList();
  int hardScore = 0;
  for (Lesson a : lessonList) {
    for (Lesson b : lessonList) {
        if (a.getTimeslot() != null && a.getTimeslot().equals(b.getTimeslot())
            && a.getId() < b.getId()) {
    }
}
</pre>
```

```
// A room can accommodate at most one lesson at the same time.
          if (a.getRoom() != null && a.getRoom().equals(b.getRoom())) {
            hardScore--;
         }
          // A teacher can teach at most one lesson at the same time.
          if (a.getTeacher().equals(b.getTeacher())) {
            hardScore--;
          }
          // A student can attend at most one lesson at the same time.
          if (a.getStudentGroup().equals(b.getStudentGroup())) {
            hardScore--;
         }
       }
    }
  }
  int softScore = 0;
  // Soft constraints are only implemented in the "complete" implementation
  return HardSoftScore.of(hardScore, softScore);
}
```

Unfortunately, this solution does not scale well because it is non-incremental: every time a lesson is assigned to a different time slot or room, all lessons are re-evaluated to calculate the new score.

A better solution is to create a **src/main/java/com/example/solver/TimeTableConstraintProvider.java** class to perform incremental score calculation. This class uses OptaPlanner's ConstraintStream APl which is inspired by Java 8 Streams and SQL. The **ConstraintProvider** scales an order of magnitude better than the **EasyScoreCalculator**: O(n) instead of O(n²).

Procedure

Create the following src/main/java/com/example/solver/TimeTableConstraintProvider.java class:

package com.example.solver;

```
import com.example.domain.Lesson;
import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore;
import org.optaplanner.core.api.score.stream.Constraint;
import org.optaplanner.core.api.score.stream.ConstraintFactory;
import org.optaplanner.core.api.score.stream.ConstraintProvider;
import org.optaplanner.core.api.score.stream.Joiners;
public class TimeTableConstraintProvider implements ConstraintProvider {
  @Override
  public Constraint[] defineConstraints(ConstraintFactory constraintFactory) {
     return new Constraint[] {
          // Hard constraints
          roomConflict(constraintFactory),
          teacherConflict(constraintFactory),
          studentGroupConflict(constraintFactory),
          // Soft constraints are only implemented in the "complete" implementation
    };
  }
```

```
private Constraint roomConflict(ConstraintFactory constraintFactory) {
  // A room can accommodate at most one lesson at the same time.
  // Select a lesson ...
  return constraintFactory.forEach(Lesson.class)
       // ... and pair it with another lesson ...
       .join(Lesson.class,
            // ... in the same timeslot ...
            Joiners.equal(Lesson::getTimeslot),
            // ... in the same room ...
            Joiners.equal(Lesson::getRoom),
            // ... and the pair is unique (different id, no reverse pairs)
            Joiners.lessThan(Lesson::getId))
       // then penalize each pair with a hard weight.
       .penalize(HardSoftScore.ONE_HARD)
       .asConstraint("Room conflict");
}
private Constraint teacherConflict(ConstraintFactory constraintFactory) {
  // A teacher can teach at most one lesson at the same time.
  return constraintFactory.forEach(Lesson.class)
       .join(Lesson.class,
            Joiners.equal(Lesson::getTimeslot),
            Joiners.equal(Lesson::getTeacher),
            Joiners.lessThan(Lesson::getId))
            .penalize(HardSoftScore.ONE_HARD)
            .asConstraint("Teacher conflict");
}
private Constraint studentGroupConflict(ConstraintFactory constraintFactory) {
  // A student can attend at most one lesson at the same time.
  return constraintFactory.forEach(Lesson.class)
       .join(Lesson.class,
            Joiners.equal(Lesson::getTimeslot),
            Joiners.equal(Lesson::getStudentGroup),
            Joiners.lessThan(Lesson::getId))
            .penalize(HardSoftScore.ONE HARD)
            .asConstraint("Student group conflict");
}
```

18.4. GATHER THE DOMAIN OBJECTS IN A PLANNING SOLUTION

A **TimeTable** instance wraps all **Timeslot**, **Room**, and **Lesson** instances of a single dataset. Furthermore, because it contains all lessons, each with a specific planning variable state, it is a *planning solution* and it has a score:

- If lessons are still unassigned, then it is an *uninitialized* solution, for example, a solution with the score **-4init/Ohard/Osoft**.
- If it breaks hard constraints, then it is an *infeasible* solution, for example, a solution with the score **-2hard/-3soft**.

• If it adheres to all hard constraints, then it is a *feasible* solution, for example, a solution with the score **Ohard/-7soft**.

The **TimeTable** class has an **@PlanningSolution** annotation, so Red Hat Build of OptaPlanner knows that this class contains all of the input and output data.

Specifically, this class is the input of the problem:

- A timeslotList field with all time slots
 - This is a list of problem facts, because they do not change during solving.
- A **roomList** field with all rooms
 - This is a list of problem facts, because they do not change during solving.
- A lessonList field with all lessons
 - This is a list of planning entities because they change during solving.
 - Of each **Lesson**:
 - The values of the timeslot and room fields are typically still null, so unassigned. They are planning variables.
 - The other fields, such as subject, teacher and studentGroup, are filled in. These fields are problem properties.

However, this class is also the output of the solution:

- A lessonList field for which each Lesson instance has non-null timeslot and room fields after solving
- A score field that represents the quality of the output solution, for example, **0hard/-5soft**

Procedure

Create the src/main/java/com/example/domain/TimeTable.java class:

package com.example.domain;

import java.util.List;

import org.optaplanner.core.api.domain.solution.PlanningEntityCollectionProperty; import org.optaplanner.core.api.domain.solution.PlanningScore; import org.optaplanner.core.api.domain.solution.PlanningSolution; import org.optaplanner.core.api.domain.solution.ProblemFactCollectionProperty; import org.optaplanner.core.api.domain.valuerange.ValueRangeProvider; import org.optaplanner.core.api.score.buildin.hardsoft.HardSoftScore;

@PlanningSolution
public class TimeTable {

@ValueRangeProvider(id = "timeslotRange") @ProblemFactCollectionProperty private List<Timeslot> timeslotList;

@ValueRangeProvider(id = "roomRange")

```
@ProblemFactCollectionProperty
private List<Room> roomList;
@PlanningEntityCollectionProperty
private List<Lesson> lessonList;
@PlanningScore
private HardSoftScore score;
private TimeTable() {
}
public TimeTable(List<Timeslot> timeslotList, List<Room> roomList,
    List<Lesson> lessonList) {
  this.timeslotList = timeslotList;
  this.roomList = roomList;
  this.lessonList = lessonList;
}
// *******
// Getters and setters
// ********
public List<Timeslot> getTimeslotList() {
  return timeslotList;
}
public List<Room> getRoomList() {
  return roomList:
}
public List<Lesson> getLessonList() {
  return lessonList:
}
public HardSoftScore getScore() {
  return score;
}
```

The value range providers

The **timeslotList** field is a value range provider. It holds the **Timeslot** instances which OptaPlanner can pick from to assign to the **timeslot** field of **Lesson** instances. The **timeslotList** field has an **@ValueRangeProvider** annotation to connect those two, by matching the **id** with the **valueRangeProviderRefs** of the **@PlanningVariable** in the **Lesson**.

Following the same logic, the **roomList** field also has an **@ValueRangeProvider** annotation.

The problem fact and planning entity properties

Furthermore, OptaPlanner needs to know which **Lesson** instances it can change as well as how to retrieve the **Timeslot** and **Room** instances used for score calculation by your **TimeTableConstraintProvider**.

The timeslotList and roomList fields have an @ProblemFactCollectionProperty annotation, so your TimeTableConstraintProvider can select from those instances.

The **lessonList** has an **@PlanningEntityCollectionProperty** annotation, so OptaPlanner can change them during solving and your **TimeTableConstraintProvider** can select from those too.

18.5. THE TIMETABLEAPP.JAVA CLASS

After you have created all of the components of the school timetable application, you will put them all together in the **TimeTableApp.java** class.

The **main()** method performs the following tasks:

- 1. Creates the **SolverFactory** to build a **Solver** for each data set.
- 2. Loads a data set.
- 3. Solves it with **Solver.solve()**.
- 4. Visualizes the solution for that data set.

Typically, an application has a single **SolverFactory** to build a new **Solver** instance for each problem data set to solve. A **SolverFactory** is thread-safe, but a **Solver** is not. For the school timetable application, there is only one data set, so only one **Solver** instance.

Here is the completed TimeTableApp.java class:

package org.acme.schooltimetabling;

import java.time.DayOfWeek; import java.time.Duration; import java.time.LocalTime; import java.util.ArrayList; import java.util.Collections; import java.util.List; import java.util.Map; import java.util.stream.Collectors;

import org.acme.schooltimetabling.domain.Lesson; import org.acme.schooltimetabling.domain.Room; import org.acme.schooltimetabling.domain.TimeTable; import org.acme.schooltimetabling.domain.Timeslot; import org.acme.schooltimetabling.solver.TimeTableConstraintProvider; import org.optaplanner.core.api.solver.Solver; import org.optaplanner.core.api.solver.SolverFactory; import org.optaplanner.core.config.solver.SolverConfig; import org.slf4j.Logger; import org.slf4j.LoggerFactory;

public class TimeTableApp {

private static final Logger LOGGER = LoggerFactory.getLogger(TimeTableApp.class);

public static void main(String[] args) {
 SolverFactory<TimeTable> solverFactory = SolverFactory.create(new SolverConfig()
 .withSolutionClass(TimeTable.class)

.withEntityClasses(Lesson.class) .withConstraintProviderClass(TimeTableConstraintProvider.class) // The solver runs only for 5 seconds on this small data set. // It's recommended to run for at least 5 minutes ("5m") otherwise. .withTerminationSpentLimit(Duration.ofSeconds(5))); // Load the problem TimeTable problem = generateDemoData(); // Solve the problem Solver<TimeTable> solver = solverFactory.buildSolver(); TimeTable solution = solver.solve(problem); // Visualize the solution printTimetable(solution); } public static TimeTable generateDemoData() { List<Timeslot> timeslotList = new ArrayList<>(10); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(8, 30), LocalTime.of(9, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9, 30), LocalTime.of(10, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(10, 30), LocalTime.of(11, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(13, 30), LocalTime.of(14, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(14, 30), LocalTime.of(15, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(8, 30), LocalTime.of(9, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(9, 30), LocalTime.of(10, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(10, 30), LocalTime.of(11, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(13, 30), LocalTime.of(14, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(14, 30), LocalTime.of(15, 30))); List<Room> roomList = new ArrayList<>(3); roomList.add(new Room("Room A")); roomList.add(new Room("Room B")); roomList.add(new Room("Room C")); List<Lesson> lessonList = new ArrayList<>(); long id = 0; lessonList.add(new Lesson(id++, "Math", "A. Turing", "9th grade")); lessonList.add(new Lesson(id++, "Math", "A. Turing", "9th grade")); lessonList.add(new Lesson(id++, "Physics", "M. Curie", "9th grade")); lessonList.add(new Lesson(id++, "Chemistry", "M. Curie", "9th grade")); lessonList.add(new Lesson(id++, "Biology", "C. Darwin", "9th grade")); lessonList.add(new Lesson(id++, "History", "I. Jones", "9th grade")); lessonList.add(new Lesson(id++, "English", "I. Jones", "9th grade")); lessonList.add(new Lesson(id++, "English", "I. Jones", "9th grade"));

```
lessonList.add(new Lesson(id++, "Spanish", "P. Cruz", "9th grade"));
     lessonList.add(new Lesson(id++, "Spanish", "P. Cruz", "9th grade"));
     lessonList.add(new Lesson(id++, "Math", "A. Turing", "10th grade"));
     lessonList.add(new Lesson(id++, "Math", "A. Turing", "10th grade"));
     lessonList.add(new Lesson(id++, "Math", "A. Turing", "10th grade"));
     lessonList.add(new Lesson(id++, "Physics", "M. Curie", "10th grade"));
     lessonList.add(new Lesson(id++, "Chemistry", "M. Curie", "10th grade"));
     lessonList.add(new Lesson(id++, "French", "M. Curie", "10th grade"));
    lessonList.add(new Lesson(id++, "Geography", "C. Darwin", "10th grade"));
     lessonList.add(new Lesson(id++, "History", "I. Jones", "10th grade"));
     lessonList.add(new Lesson(id++, "English", "P. Cruz", "10th grade"));
     lessonList.add(new Lesson(id++, "Spanish", "P. Cruz", "10th grade"));
     return new TimeTable(timeslotList, roomList, lessonList);
  }
  private static void printTimetable(TimeTable timeTable) {
     LOGGER.info("");
     List<Room> roomList = timeTable.getRoomList();
     List<Lesson> lessonList = timeTable.getLessonList();
     Map<Timeslot, Map<Room, List<Lesson>>> lessonMap = lessonList.stream()
          .filter(lesson -> lesson.getTimeslot() != null && lesson.getRoom() != null)
          .collect(Collectors.groupingBy(Lesson::getTimeslot,
Collectors.groupingBy(Lesson::getRoom)));
     LOGGER.info("
                          | " + roomList.stream()
          .map(room -> String.format("%-10s", room.getName())).collect(Collectors.joining(" | ")) + "
|");
     LOGGER.info("|" + "------|".repeat(roomList.size() + 1));
     for (Timeslot timeslot : timeTable.getTimeslotList()) {
       List<List<Lesson>> cellList = roomList.stream()
            .map(room -> {
               Map<Room, List<Lesson>> byRoomMap = lessonMap.get(timeslot);
              if (byRoomMap == null) {
                 return Collections.<Lesson>emptyList();
               }
              List<Lesson> cellLessonList = byRoomMap.get(room);
              if (cellLessonList == null) {
                 return Collections.<Lesson>emptyList();
              }
              return cellLessonList;
            })
            .collect(Collectors.toList());
       LOGGER.info("| " + String.format("%-10s",
            timeslot.getDayOfWeek().toString().substring(0, 3) + " " + timeslot.getStartTime()) + " | "
            + cellList.stream().map(cellLessonList -> String.format("%-10s",
                 cellLessonList.stream().map(Lesson::getSubject).collect(Collectors.joining(", "))))
                 .collect(Collectors.joining(" | "))
            + " |");
                               1"
       LOGGER.info("|
            + cellList.stream().map(cellLessonList -> String.format("%-10s",
                 cellLessonList.stream().map(Lesson::getTeacher).collect(Collectors.joining(", "))))
                 .collect(Collectors.joining(" | "))
            + " |");
                               1"
       LOGGER.info("
```

```
+ cellList.stream().map(cellLessonList -> String.format("%-10s",
                 cellLessonList.stream().map(Lesson::getStudentGroup).collect(Collectors.joining(",
"))))
                 .collect(Collectors.joining(" | "))
            + " |");
       LOGGER.info("|" + "------|".repeat(roomList.size() + 1));
     }
     List<Lesson> unassignedLessons = lessonList.stream()
          .filter(lesson -> lesson.getTimeslot() == null || lesson.getRoom() == null)
          .collect(Collectors.toList());
     if (!unassignedLessons.isEmpty()) {
       LOGGER.info("");
       LOGGER.info("Unassigned lessons");
       for (Lesson lesson : unassignedLessons) {
          LOGGER.info(" " + lesson.getSubject() + " - " + lesson.getTeacher() + " - " +
lesson.getStudentGroup());
       }
     }
  }
```

The main() method first creates the SolverFactory:

SolverFactory<TimeTable> solverFactory = SolverFactory.create(new SolverConfig() .withSolutionClass(TimeTable.class) .withEntityClasses(Lesson.class) .withConstraintProviderClass(TimeTableConstraintProvider.class) // The solver runs only for 5 seconds on this small data set. // It's recommended to run for at least 5 minutes ("5m") otherwise. .withTerminationSpentLimit(Duration.ofSeconds(5)));

The **SolverFactory** creation registers the **@PlanningSolution** class, the **@PlanningEntity** classes, and the **ConstraintProvider** class, all of which you created earlier.

Without a termination setting or a **terminationEarly()** event, the solver runs forever. To avoid that, the solver limits the solving time to five seconds.

After five seconds, the **main()** method loads the problem, solves it, and prints the solution:

// Load the problem
TimeTable problem = generateDemoData();

// Solve the problem
Solver<TimeTable> solver = solverFactory.buildSolver();
TimeTable solution = solver.solve(problem);

// Visualize the solution
printTimetable(solution);

The **solve()** method doesn't return instantly. It runs for five seconds before returning the best solution.

OptaPlanner returns *the best solution* found in the available termination time. Due to the nature of NP-hard problems, the best solution might not be optimal, especially for larger data sets. Increase the termination time to potentially find a better solution.

The **generateDemoData()** method generates the school timetable problem to solve.

The **printTimetable()** method prettyprints the timetable to the console, so it's easy to determine visually whether or not it's a good schedule.

18.6. CREATING AND RUNNING THE SCHOOL TIMETABLE APPLICATION

Now that you have completed all of the components of the school timetable Java application, you are ready to put them all together in the **TimeTableApp.java** class and run it.

Prerequisites

• You have created all of the required components of the school timetable application.

Procedure

1. Create the src/main/java/org/acme/schooltimetabling/TimeTableApp.java class:

package org.acme.schooltimetabling;

import java.time.DayOfWeek; import java.time.Duration; import java.time.LocalTime; import java.util.ArrayList; import java.util.Collections; import java.util.List; import java.util.Map; import java.util.stream.Collectors;

import org.acme.schooltimetabling.domain.Lesson; import org.acme.schooltimetabling.domain.Room; import org.acme.schooltimetabling.domain.TimeTable; import org.acme.schooltimetabling.domain.Timeslot; import org.acme.schooltimetabling.solver.TimeTableConstraintProvider; import org.optaplanner.core.api.solver.Solver; import org.optaplanner.core.api.solver.SolverFactory; import org.optaplanner.core.config.solver.SolverConfig; import org.slf4j.Logger; import org.slf4j.LoggerFactory;

public class TimeTableApp {

private static final Logger LOGGER = LoggerFactory.getLogger(TimeTableApp.class);

public static void main(String[] args) {

SolverFactory<TimeTable> solverFactory = SolverFactory.create(new SolverConfig() .withSolutionClass(TimeTable.class) .withEntityClasses(Lesson.class)

.withConstraintProviderClass(TimeTableConstraintProvider.class)

// The solver runs only for 5 seconds on this small data set.

// It's recommended to run for at least 5 minutes ("5m") otherwise.

.withTerminationSpentLimit(Duration.ofSeconds(5)));

// Load the problem

TimeTable problem = generateDemoData(); // Solve the problem Solver<TimeTable> solver = solverFactory.buildSolver(); TimeTable solution = solver.solve(problem); // Visualize the solution printTimetable(solution); } public static TimeTable generateDemoData() { List<Timeslot> timeslotList = new ArrayList<>(10); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(8, 30), LocalTime.of(9, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9, 30), LocalTime.of(10, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(10, 30), LocalTime.of(11, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(13, 30), LocalTime.of(14, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(14, 30), LocalTime.of(15, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(8, 30), LocalTime.of(9, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(9, 30), LocalTime.of(10, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(10, 30), LocalTime.of(11, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(13, 30), LocalTime.of(14, 30))); timeslotList.add(new Timeslot(DayOfWeek.TUESDAY, LocalTime.of(14, 30), LocalTime.of(15, 30))); List<Room> roomList = new ArrayList<>(3); roomList.add(new Room("Room A")); roomList.add(new Room("Room B")); roomList.add(new Room("Room C")); List<Lesson> lessonList = new ArrayList<>(); long id = 0; lessonList.add(new Lesson(id++, "Math", "A. Turing", "9th grade")); lessonList.add(new Lesson(id++, "Math", "A. Turing", "9th grade")); lessonList.add(new Lesson(id++, "Physics", "M. Curie", "9th grade")); lessonList.add(new Lesson(id++, "Chemistry", "M. Curie", "9th grade")); lessonList.add(new Lesson(id++, "Biology", "C. Darwin", "9th grade")); lessonList.add(new Lesson(id++, "History", "I. Jones", "9th grade")); lessonList.add(new Lesson(id++, "English", "I. Jones", "9th grade")); lessonList.add(new Lesson(id++, "English", "I. Jones", "9th grade")); lessonList.add(new Lesson(id++, "Spanish", "P. Cruz", "9th grade")); lessonList.add(new Lesson(id++, "Spanish", "P. Cruz", "9th grade")); lessonList.add(new Lesson(id++, "Math", "A. Turing", "10th grade")); lessonList.add(new Lesson(id++, "Math", "A. Turing", "10th grade")); lessonList.add(new Lesson(id++, "Math", "A. Turing", "10th grade")); lessonList.add(new Lesson(id++, "Physics", "M. Curie", "10th grade"));

```
lessonList.add(new Lesson(id++, "Chemistry", "M. Curie", "10th grade"));
     lessonList.add(new Lesson(id++, "French", "M. Curie", "10th grade"));
    lessonList.add(new Lesson(id++, "Geography", "C. Darwin", "10th grade"));
     lessonList.add(new Lesson(id++, "History", "I. Jones", "10th grade"));
    lessonList.add(new Lesson(id++, "English", "P. Cruz", "10th grade"));
     lessonList.add(new Lesson(id++, "Spanish", "P. Cruz", "10th grade"));
    return new TimeTable(timeslotList, roomList, lessonList);
  }
  private static void printTimetable(TimeTable timeTable) {
     LOGGER.info("");
     List<Room> roomList = timeTable.getRoomList();
     List<Lesson> lessonList = timeTable.getLessonList();
     Map<Timeslot, Map<Room, List<Lesson>>> lessonMap = lessonList.stream()
          .filter(lesson -> lesson.getTimeslot() != null && lesson.getRoom() != null)
          .collect(Collectors.groupingBy(Lesson::getTimeslot,
Collectors.groupingBy(Lesson::getRoom)));
     LOGGER.info("
                            | " + roomList.stream()
          .map(room -> String.format("%-10s", room.getName())).collect(Collectors.joining(" |
")) + " |");
    LOGGER.info("|" + "------|".repeat(roomList.size() + 1));
    for (Timeslot timeslot : timeTable.getTimeslotList()) {
       List<List<Lesson>> cellList = roomList.stream()
            .map(room -> {
               Map<Room, List<Lesson>> byRoomMap = lessonMap.get(timeslot);
               if (byRoomMap == null) {
                 return Collections.<Lesson>emptyList();
               }
               List<Lesson> cellLessonList = byRoomMap.get(room);
               if (cellLessonList == null) {
                 return Collections.<Lesson>emptyList();
               }
               return cellLessonList;
            })
            .collect(Collectors.toList());
       LOGGER.info("| " + String.format("%-10s",
            timeslot.getDayOfWeek().toString().substring(0, 3) + " " +
timeslot.getStartTime()) + " | "
            + cellList.stream().map(cellLessonList -> String.format("%-10s",
cellLessonList.stream().map(Lesson::getSubject).collect(Collectors.joining(", "))))
                 .collect(Collectors.joining(" | "))
            + " |");
       LOGGER.info("
            + cellList.stream().map(cellLessonList -> String.format("%-10s",
cellLessonList.stream().map(Lesson::getTeacher).collect(Collectors.joining(", "))))
                 .collect(Collectors.joining(" | "))
            + " |");
       LOGGER.info("
                              1"
            + cellList.stream().map(cellLessonList -> String.format("%-10s",
cellLessonList.stream().map(Lesson::getStudentGroup).collect(Collectors.joining(", "))))
                 .collect(Collectors.joining(" | "))
```

```
+ " |");
       LOGGER.info("|" + "------|".repeat(roomList.size() + 1));
     }
     List<Lesson> unassignedLessons = lessonList.stream()
          .filter(lesson -> lesson.getTimeslot() == null || lesson.getRoom() == null)
          .collect(Collectors.toList());
    if (!unassignedLessons.isEmpty()) {
       LOGGER.info("");
       LOGGER.info("Unassigned lessons");
       for (Lesson lesson : unassignedLessons) {
          LOGGER.info(" " + lesson.getSubject() + " - " + lesson.getTeacher() + " - " +
lesson.getStudentGroup());
       }
    }
  }
}
```

2. Run the **TimeTableApp** class as the main class of a normal Java application. The following output should result:

INFO Room A Room B Room C
INFO
INFO MON 08:30 English Math
INFO I. Jones A. Turing
INFO 9th grade 10th grade
INFO
INFO MON 09:30 History Physics
INFO I. Jones M. Curie
INFO 9th grade 10th grade

3. Verify the console output. Does it conform to all hard constraints? What happens if you comment out the **roomConflict** constraint in **TimeTableConstraintProvider**?

The **info** log shows what OptaPlanner did in those five seconds:

... Solving started: time spent (33), best score (-8init/0hard/0soft), environment mode (REPRODUCIBLE), random (JDK with seed 0).

... Construction Heuristic phase (0) ended: time spent (73), best score (0hard/0soft), score calculation speed (459/sec), step total (4).

... Local Search phase (1) ended: time spent (5000), best score (0hard/0soft), score calculation speed (28949/sec), step total (28398).

... Solving ended: time spent (5000), best score (0hard/0soft), score calculation speed (28524/sec), phase total (2), environment mode (REPRODUCIBLE).

18.7. TESTING THE APPLICATION

A good application includes test coverage. Test the constraints and the solver in your timetable project.

18.7.1. Test the school timetable constraints

To test each constraint of the timetable project in isolation, use a **ConstraintVerifier** in unit tests. This tests each constraint's corner cases in isolation from the other tests, which lowers maintenance when adding a new constraint with proper test coverage.

This test verifies that the constraint **TimeTableConstraintProvider::roomConflict**, when given three lessons in the same room and two of the lessons have the same timeslot, penalizes with a match weight of 1. So if the constraint weight is **10hard** it reduces the score by **-10hard**.

Procedure

Create the src/test/java/org/acme/optaplanner/solver/TimeTableConstraintProviderTest.java class:

```
package org.acme.optaplanner.solver;
import java.time.DayOfWeek;
import java.time.LocalTime;
import javax.inject.Inject;
import io.quarkus.test.junit.QuarkusTest;
import org.acme.optaplanner.domain.Lesson;
import org.acme.optaplanner.domain.Room;
import org.acme.optaplanner.domain.TimeTable;
import org.acme.optaplanner.domain.Timeslot;
import org.junit.jupiter.api.Test;
import org.optaplanner.test.api.score.stream.ConstraintVerifier;
@QuarkusTest
class TimeTableConstraintProviderTest {
  private static final Room ROOM = new Room("Room1");
  private static final Timeslot TIMESLOT1 = new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9,0),
LocalTime.NOON);
  private static final Timeslot TIMESLOT2 = new Timeslot(DayOfWeek.TUESDAY,
LocalTime.of(9,0), LocalTime.NOON);
  @Inject
  ConstraintVerifier<TimeTableConstraintProvider, TimeTable> constraintVerifier;
  @Test
  void roomConflict() {
    Lesson firstLesson = new Lesson(1, "Subject1", "Teacher1", "Group1");
    Lesson conflictingLesson = new Lesson(2, "Subject2", "Teacher2", "Group2");
    Lesson nonConflictingLesson = new Lesson(3, "Subject3", "Teacher3", "Group3");
    firstLesson.setRoom(ROOM);
    firstLesson.setTimeslot(TIMESLOT1);
    conflictingLesson.setRoom(ROOM);
    conflictingLesson.setTimeslot(TIMESLOT1);
    nonConflictingLesson.setRoom(ROOM);
    nonConflictingLesson.setTimeslot(TIMESLOT2);
    constraintVerifier.verifyThat(TimeTableConstraintProvider::roomConflict)
         .given(firstLesson, conflictingLesson, nonConflictingLesson)
         .penalizesBy(1);
```

Notice how **ConstraintVerifier** ignores the constraint weight during testing even if those constraint weights are hardcoded in the **ConstraintProvider**. This is because constraint weights change regularly before going into production. This way, constraint weight tweaking does not break the unit tests.

18.7.2. Test the school timetable solver

This example tests the Red Hat Build of OptaPlanner school timetable project on the Red Hat build of Quarkus platform. It uses a JUnit test to generate a test data set and send it to the **TimeTableController** to solve.

Procedure

}

1. Create the **src/test/java/com/example/rest/TimeTableResourceTest.java** class with the following content:

package com.exmaple.optaplanner.rest;

import java.time.DayOfWeek; import java.time.LocalTime; import java.util.ArrayList; import java.util.List;

import javax.inject.Inject;

import io.quarkus.test.junit.QuarkusTest; import com.exmaple.optaplanner.domain.Room; import com.exmaple.optaplanner.domain.Timeslot; import com.exmaple.optaplanner.domain.Lesson; import com.exmaple.optaplanner.domain.TimeTable; import com.exmaple.optaplanner.rest.TimeTableResource; import org.junit.jupiter.api.Test; import org.junit.jupiter.api.Timeout;

import static org.junit.jupiter.api.Assertions.assertFalse; import static org.junit.jupiter.api.Assertions.assertNotNull; import static org.junit.jupiter.api.Assertions.assertTrue;

@QuarkusTest
public class TimeTableResourceTest {

@Inject TimeTableResource timeTableResource;

```
@Test
@Timeout(600_000)
public void solve() {
    TimeTable problem = generateProblem();
    TimeTable solution = timeTableResource.solve(problem);
    assertFalse(solution.getLessonList().isEmpty());
    for (Lesson lesson : solution.getLessonList()) {
        assertNotNull(lesson.getTimeslot());
    };
```

assertNotNull(lesson.getRoom()); } assertTrue(solution.getScore().isFeasible()); } private TimeTable generateProblem() { List<Timeslot> timeslotList = new ArrayList<>(); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(8, 30), LocalTime.of(9, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(9, 30), LocalTime.of(10, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(10, 30), LocalTime.of(11, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(13, 30), LocalTime.of(14, 30))); timeslotList.add(new Timeslot(DayOfWeek.MONDAY, LocalTime.of(14, 30), LocalTime.of(15, 30))); List<Room> roomList = new ArrayList<>(); roomList.add(new Room("Room A")); roomList.add(new Room("Room B")); roomList.add(new Room("Room C")); List<Lesson> lessonList = new ArrayList<>(); lessonList.add(new Lesson(101L, "Math", "B. May", "9th grade")); lessonList.add(new Lesson(102L, "Physics", "M. Curie", "9th grade")); lessonList.add(new Lesson(103L, "Geography", "M. Polo", "9th grade")); lessonList.add(new Lesson(104L, "English", "I. Jones", "9th grade")); lessonList.add(new Lesson(105L, "Spanish", "P. Cruz", "9th grade")); lessonList.add(new Lesson(201L, "Math", "B. May", "10th grade")); lessonList.add(new Lesson(202L, "Chemistry", "M. Curie", "10th grade")); lessonList.add(new Lesson(203L, "History", "I. Jones", "10th grade")); lessonList.add(new Lesson(204L, "English", "P. Cruz", "10th grade")); lessonList.add(new Lesson(205L, "French", "M. Curie", "10th grade")); return new TimeTable(timeslotList, roomList, lessonList); } }

This test verifies that after solving, all lessons are assigned to a time slot and a room. It also verifies that it found a feasible solution (no hard constraints broken).

2. Add test properties to the src/main/resources/application.properties file:

The solver runs only for 5 seconds to avoid a HTTP timeout in this simple implementation.
 # It's recommended to run for at least 5 minutes ("5m") otherwise.
 quarkus.optaplanner.solver.termination.spent-limit=5s

Effectively disable this termination in favor of the best-score-limit %test.quarkus.optaplanner.solver.termination.spent-limit=1h %test.quarkus.optaplanner.solver.termination.best-score-limit=0hard/*soft

Normally, the solver finds a feasible solution in less than 200 milliseconds. Notice how the **application.properties** file overwrites the solver termination during tests to terminate as soon as a feasible solution **(0hard/*soft)** is found. This avoids hard coding a solver time, because the unit test

might run on arbitrary hardware. This approach ensures that the test runs long enough to find a feasible solution, even on slow systems. But it does not run a millisecond longer than it strictly must, even on fast systems.

18.8. LOGGING

After you complete the Red Hat Build of OptaPlanner school timetable project, you can use logging information to help you fine-tune the constraints in the **ConstraintProvider**. Review the score calculation speed in the **info** log file to assess the impact of changes to your constraints. Run the application in debug mode to show every step that your application takes or use trace logging to log every step and every move.

Procedure

- 1. Run the school timetable application for a fixed amount of time, for example, five minutes.
- 2. Review the score calculation speed in the **log** file as shown in the following example:



... Solving ended: ..., score calculation speed (29455/sec), ...

- 3. Change a constraint, run the planning application again for the same amount of time, and review the score calculation speed recorded in the **log** file.
- 4. Run the application in debug mode to log every step that the application makes:
 - To run debug mode from the command line, use the **-D** system property.
 - To permanently enable debug mode, add the following line to the **application.properties** file:

quarkus.log.category."org.optaplanner".level=debug

The following example shows output in the **log** file in debug mode:

... Solving started: time spent (67), best score (-20init/0hard/0soft), environment mode (REPRODUCIBLE), random (JDK with seed 0).

... CH step (0), time spent (128), score (-18init/0hard/0soft), selected move count (15), picked move ([Math(101) {null -> Room A}, Math(101) {null -> MONDAY 08:30}]).
... CH step (1), time spent (145), score (-16init/0hard/0soft), selected move count (15), picked move ([Physics(102) {null -> Room A}, Physics(102) {null -> MONDAY 09:30}]).

5. Use **trace** logging to show every step and every move for each step.

18.9. USING MICROMETER AND PROMETHEUS TO MONITOR YOUR SCHOOL TIMETABLE OPTAPLANNER JAVA APPLICATION

OptaPlanner exposes metrics through Micrometer, a metrics instrumentation library for Java applications. You can use Micrometer with Prometheus to monitor the OptaPlanner solver in the school timetable application.

Prerequisites

- You have created the OptaPlanner school timetable application with Java.
- Prometheus is installed. For information about installing Prometheus, see the Prometheus website.

Procedure

1. Add the Micrometer Prometheus dependencies to the school timetable **pom.xml** file where **<MICROMETER_VERSION>** is the version of Micrometer that you installed:

<dependency> <groupId>io.micrometer</groupId> <artifactId>micrometer-registry-prometheus</artifactId> <version><MICROMETER_VERSION></version> </dependency>



NOTE

The **micrometer-core** dependency is also required. However this dependency is contained in the **optaplanner-core** dependency so you do not need to add it to the **pom.xml** file.

2. Add the following import statements to the TimeTableApp.java class.

import io.micrometer.core.instrument.Metrics; import io.micrometer.prometheus.PrometheusConfig; import io.micrometer.prometheus.PrometheusMeterRegistry;

3. Add the following lines to the top of the main method of the **TimeTableApp.java** class so Prometheus can scrap data from **com.sun.net.httpserver.HttpServer** before the solution starts:

```
PrometheusMeterRegistry prometheusRegistry = new
PrometheusMeterRegistry(PrometheusConfig.DEFAULT);
    try {
       HttpServer server = HttpServer.create(new InetSocketAddress(8080), 0);
       server.createContext("/prometheus", httpExchange -> {
         String response = prometheusRegistry.scrape();
         httpExchange.sendResponseHeaders(200, response.getBytes().length);
         try (OutputStream os = httpExchange.getResponseBody()) {
            os.write(response.getBytes());
         }
       });
       new Thread(server::start).start();
    } catch (IOException e) {
       throw new RuntimeException(e);
    }
    Metrics.addRegistry(prometheusRegistry);
    solve();
```

4. Add the following line to control the solving time. By adjusting the solving time, you can see how the metrics change based on the time spent solving.

withTerminationSpentLimit(Duration.ofMinutes(5)));

- 5. Start the school timetable application.
- 6. Open http://localhost:8080/prometheus in a web browser to view the timetable application in Prometheus.
- 7. Open your monitoring system to view the metrics for your OptaPlanner project. The following metrics are exposed:
 - **optaplanner_solver_errors_total**: the total number of errors that occurred while solving since the start of the measuring.
 - **optaplanner_solver_solve_duration_seconds_active_count**: the number of solvers currently solving.
 - **optaplanner_solver_solve_duration_seconds_max**: run time of the longest-running currently active solver.
 - **optaplanner_solver_solve_duration_seconds_duration_sum**: the sum of each active solver's solve duration. For example, if there are two active solvers, one running for three minutes and the other for one minute, the total solve time is four minutes.

APPENDIX A. VERSIONING INFORMATION

Documentation last updated on Friday, July 14, 2023.