Java Constraint Programming with JSR-331

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Outline

- Introduction to Constraint Programming (CP)
- JSR-331: oncoming Java CP API standard
  - allow a user to switch between different CP Solvers without changing a line in the application code
- Examples of practical use of Constraint Programming for Java-based decision support applications
  - Demonstrate how CP gives Java developers unprecedented power to define and solve complex constraint satisfaction and optimization problems
  - Integration of Constraint Solvers with Rule Engines and Machine Learning tools
Introduction to CP

Constraint Programming (CP) is a very powerful problem solving paradigm with strong roots in Operation Research and AI:

- Handbook of Constraint Programming (Elsevier, 2006)
- Association for CP - http://slash.math.unipd.it/acp/
- Cork Constraint Computation Centre - http://www.4c.ucc.ie/

CP is a proven optimization technology introduced to the business application development at the beginning of 1990s.

During the 90s ILOG Solver became the most popular optimization tool that was widely used by commercial C++ developers. Being implemented not as a specialized language but rather as an API for the mainstream language of that time, ILOG Solver successfully built a bridge between the academic and business worlds.

Nowadays Optimization technology is quickly coming back to the business application development world as an important component of the Enterprise Decision Management (EDM).
Optimization technology helps organizations make better plans and schedules.

A model captures your complex planning or scheduling problem. Then a mathematical engine applies the model to a scenario find the best possible solution.

When optimization models are embedded in applications, planners and operations managers can perform what-if analysis, and compare scenarios.

Equipped with intelligent alternatives, you make better decisions, dramatically improving operational efficiency.
Examples of Constraints

Constraints represent conditions which restrict our freedom of decision making:

- The meeting must start no later than 3:30PM
- Glass components cannot be placed in the same bin with copper components
- The job requires Joe or Jim but cannot use John
- Mary prefers not to work on Wednesday
- The portfolio cannot include more than 15% of technology stocks unless it includes at least 7% of utility stocks
There are 3 integers X, Y, Z defined from 0 to 10. Constraints: \( X < Y \) and \( X + Y = Z \). Find all feasible values of X, Y, and Z

**Simple Java Solution:**

```java
for(int x=0; x<11; x++)
    for(int y = 0; y<11; y++)
        for(int z=0; z<11; z++)
            if (x < y && z == x+y)
                System.out.println("X="+x+" Y="+y+" Z="+z);
```

**“Optimized” Java Solution:**

```java
for(int x=0; x<11; x++)
    for(int y = x+1; y<11; y++)
        if (x+y < 11)
            System.out.println("X="+x+" Y="+y+" Z="+(x+y));
```

**What’s wrong with this “solution”?**

- Readability, Extensibility, Performance,...
Simple Solution with Java CP API (JSR-331):

// Problem Definition
Problem problem = new Problem("XYZ");
Var x = problem.var("X", 0, 10);
Var y = problem.var("Y", 0, 10);
Var z = problem.var("Z", 0, 10);
x.lt(y).post(); // X < Y
x.add(y).eq(z).post(); // X + Y = Z

// Problem Resolution
Solution[] solutions = problem.getSolver().findAllSolutions();
for(Solution solution : solutions)
    solution.log();
How the constraint “X < Y” works

Let’s assume X and Y are defined on the domain [0,10]

Initial constraint propagation after posting X<Y constraint:

X[0;9]  
Y[1;10]

Changes in X cause the changes in Y

X>3  =>  Y > 4

Changes in Y cause the changes in X

Y<=8  =>  X<=7

Bi-Directional constraint propagation
Constraint Satisfaction Problem - CSP

- CP clearly separates “What” from “How”

- Problem Definition (WHAT):
  - Constrained Variables with all possible values
    - Integer, Boolean, Real, and Set variables
  - Constraints on the variables
    - Basic arithmetic and logical constraints and expressions
    - Global constraints (AllDifferent, Cardinality, ElementAt, ...)

- Problem Resolution (HOW):
  - Find Solution(s) that defines a value for each variable such that all constraints are satisfied
    - Find a feasible solution
    - Find an optimal solution
    - Find (iterate through) all solutions
  - Search Strategies
Constraint Satisfaction Environment

- Predefined classes for Constrained Variables, Constraints, and Search Strategies
- Domain representations for major constrained objects
- Generic reversible environment
  - “Try/Fail/Backtrack” capabilities
  - Powerful customizable event management mechanism
  - Constraints use events to control states of all constrained objects
- Constraint propagation mechanisms
- Ability to create problem-specific constraints and search strategies
Constraint Programming: a bridge between academia and biz

- CP is especially successful dealing with real-world scheduling, resource allocation, and complex configuration problems:
  - CP clearly separates problem definition from problem resolution bringing declarative programming to the real-world
  - CP made different optimization techniques handily available to normal software developers (without PhDs in Operation Research)

- A few real world CP application examples from my consulting practice:
  - Financial Portfolio Balancing for a Wall Street Wealth Management System
  - Grain Train Scheduling for a Canadian R/R company
  - Truck Loading and Routing system
  - Data Centre Capacity Management
  - Workforce/Workload Scheduling system for a Utility company
Real-world example: Workforce/Workload Management

Field Service Scheduling for the Long Island Gas and Electric Utility

- More than 1 million customers in Long Island, NY
- More than 5000 employees
- Service territory 1,230 square miles
- Hundreds jobs per day
- Job requires a mix of people skills, vehicles and equipment

Multi-objective Work Planning and Scheduling:

- Travel time minimization
- Resource load levelization
- Skill utilization (use the least costly skills/equipment)
- Schedule jobs ASAP
- Honor user-defined preferences
Some Popular CP Tools

CP Modeling Languages
- ILOG OPL from IBM ILOG (www.ilog.com)
- MiniZinc from G12 group, Australia (http://www.g12.cs.mu.oz.au)
- Comet, Brown University (www.dynadec.com)
- Prolog (ECLiPSe, SICStus)

C++ API
- ILOG CP – Commercial from IBM ILOG
- Gecode – Open Source (www.gecode.org)

Java API
- Choco - Open Source
- ILOG JSolver – Commercial
- Constrainer - Open Source

20+ other CP Solvers:  http://slash.math.unipd.it/cp/
CP Solvers are usually well integrated with other optimization tools (LP, MIP)
JSR-331 – Java Specification Request

❖ JSR-331 - Java Constraint Programming API under the roof of the Java Community Process www.jcp.org

❖ JSR-331 covers key concepts and design decisions related to the standard representation and resolution of constraint satisfaction and optimization problems

❖ JSR-331 Early Draft is now available for public review www.cpstandards.org
Key Standardization Objectives

- Make Constraint Programming more accessible for business software developers
- Allow a Java business application developer to easily switch between different solver implementations without any(!) changes in the application code
- Assist CP vendors in creating practical JSR-331 implementations
Standard is Oriented to Application Developers yet allowing CP Vendors to provide their own implementations.
A map-coloring problem involves choosing colors for the countries on a map in such a way that at most 4 colors are used and no two neighboring countries have the same color.

We will consider six countries: Belgium, Denmark, France, Germany, Netherlands, and Luxembourg.

The colors are red, green, blue, yellow.
Example “Map Coloring”: problem variables

```
static final String[] colors = { "red", "green", "blue", "yellow" };
```

Problem  $p = \text{new Problem("Map-coloring");}$

```
// Define Variables
Var Belgium       = $p$.var("Belgium“,0, 3);
Var Denmark       = $p$.var("Denmark“,0, 3);
Var France         = $p$.var("France“,0, 3);
Var Germany       = $p$.var("Germany“,0, 3);
Var Netherlands = $p$.var("Netherlands“,0, 3);
Var Luxemburg      = $p$.var("Luxemburg“,0, 3);
```

Each country is represented as a variable that corresponds to an unknown color: 0,1,2, or 3
// Define Constraints
France.neq(Belgium).post();
France.neq(Luxemburg).post();
France.neq(Germany).post();
Luxemburg.neq(Germany).post();
Luxemburg.neq(Belgium).post();
Belgium.neq(Netherlands).post();
Belgium.neq(Germany).post();
Germany.neq(Netherlands).post();
Germany.neq(Denmark).post();

// We actually create a constraint and then post it
Constraint c = Germany.neq(Denmark);
c.post();
"Map Coloring": solution search

// Solve
Goal goal = p.goalGenerate();
Solution solution = p.getSolved().findSolution();
if (solution != null) {
    for (int i = 0; i < p.getVars().length; i++) {
        Var var = p.getVars()[i];
        p.log(var.getName() + " - " + colors[var.getValue()]);
    }
}

// Solution:
Belgium – red
Denmark – red
France – green
Germany – blue
Netherlands – green
Luxemburg - yellow
In real-world many problems are over-constrained. If this is a case, we may want to find a solution that minimizes the total constraint violation.

Consider a map coloring problem when there are no enough colors, e.g. only three colors:

- Coloring violations may have different importance for France – Belgium and France – Germany
- Find a solution that minimizes total constraint violations
Constraint “softening” rules:

Coloring constraint violations have different importance on the scale 0-9999:
- Luxemburg– Germany (9043)
- France – Luxemburg (257)
- Luxemburg – Belgium (568)

We want to find a solution that minimizes the total constraint violation
Map Coloring with Hard and Soft Constraints

// Hard Constraints
France.neq(Belgium).post();
France.neq(Germany).post();
Belgium.neq(Netherlands).post();
Belgium.neq(Germany).post();
Germany.neq(Denmark).post();
Germany.neq(Netherlands).post();

// Soft Constraints
Var[] weightVars = {
    Luxemburg.eq(Germany).asBool().mul(9043),
    France.eq(Luxemburg).asBool().mul(257),
    Luxemburg.eq(Belgium).asBool().mul(568)
};
Var weightedSum = p.sum(weightVars);
// Optimal Solution Search

Solution solution = p.getSolver().getOptimalSolution(weightedSum);

if (solution == null)
    p.log("No solutions found");
else
    solution.log();

Solution:
Scheduling problems usually deals with:

- **Activities** with yet unknown start times and known durations (not always)
- **Resources** with limited capacities varying over time
- **Constraints:**
  - Between activities (e.g. Job2 starts after the end of Job1)
  - Between activities and resources (e.g. Job1 requires a welder, where Jim and Joe both have a welder skills)

- There are multiple scheduling objectives (e.g. minimize the makespan, utilize resources, etc.)
How we may create a CP-based Scheduler?

- Precedence Constraints:
  - “starts after”
  - “starts before”
  - “starts at”
  - “and before”

- Resource Constraints:
  - “requires”
  - “consumes”
  - “produces”
  - “provides”

- Var start
- Var duration
- Var end

// Alternative resource requirements
activity1.requires(resource2, varReq2).post();
activity1.requires(resource3, varReq3).post();
varReq2.ne(varReq3).post();
Oven - job scheduling with one resource

There is an oven in which we can fire batches of bricks. There are five orders to fire $X$ batches during $Y$ days. Schedule all orders to be done in no more than 11 days taking into consideration the following oven availability:

# Batches

<table>
<thead>
<tr>
<th>Days</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
<th>10</th>
<th>11</th>
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<tbody>
<tr>
<td>Batches</td>
<td>1</td>
<td>2</td>
<td>3</td>
<td>2</td>
<td>1</td>
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<td>1</td>
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</tr>
</tbody>
</table>

Global capacity of the oven

- A: 2 batches, 1 day
- B: 1 batch, 4 days
- C: 1 batch, 4 days
- D: 1 batch, 2 days
- E: 2 batches, 4 days

5 Activities
Problem problem = new Problem("Oven Scheduling Example");
Schedule schedule = problem.addSchedule(0, 11);
Activity A = schedule.addActivity(1, "A");
Activity B = schedule.addActivity(4, "B");
Activity C = schedule.addActivity(4, "C");
Activity D = schedule.addActivity(2, "D");
Activity E = schedule.addActivity(4, "E");
Resource oven = schedule.addResource(3, "oven");
oven.setCapacityMax(0, 2);
oven.setCapacityMax(1, 1);
oven.setCapacityMax(2, 0);
oven.setCapacityMax(3, 1);
oven.setCapacityMax(4, 1);
oven.setCapacityMax(10, 1);
// Resource Constraints
A.requires(oven, 2).post();
B.requires(oven, 1).post();
C.requires(oven, 1).post();
D.requires(oven, 1).post();
E.requires(oven, 2).post();
// Find Solution
schedule.scheduleActivities();
schedule.displayActivities();

SOLUTION:
A[5 -- 1 --> 6) requires oven[2]
B[3 -- 4 --> 7) requires oven[1]
C[7 -- 4 --> 11) requires oven[1]
D[0 -- 2 --> 2) requires oven[1]
E[6 -- 4 --> 10) requires oven[2]
Business Rules could be used to define and modify business objects

Rule Engine can generate a related constraint satisfaction problem/subproblem representing it in terms of constrained variables and constraints

CP Solver can solve the optimization problems and return the results to the Rules Engine for further analysis
Notorious CSP “SUDOKU”
Sudoku Constraints in OpenRules Excel Rules Table

<table>
<thead>
<tr>
<th>Array Name</th>
<th>x00</th>
<th>x01</th>
<th>x02</th>
<th>x03</th>
<th>x04</th>
<th>x05</th>
<th>x06</th>
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</table>

Rules void postSudokuConstraints(CpProblem p)

CpVariable[] array = p.addArray(name,vars);
p.allDiff(array).post();

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<th>Array Name</th>
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Online Decision Support: modeling and solving constraint satisfaction problems

Typical Online Systems with CP-based Solvers:
- Online Reservation systems (hotels, tours, vacations, ..)
- Event Scheduling (both business and personal events in social networks)
- Field Service Scheduling, Advertisement Scheduling, and more

Traditional Approach:
- “Fat” Problem Solver tuned for all possible problem states
- Complexity grows over time – hard to create and maintain
Online Decision Support: CP + BR
adding Rule Engine to find the “best” strategy

CSP (New State)

CP Solver
(the “best” strategy)

New Change Request

Selected Model & Strategy

Rule Engine

Predefined Models and Strategies

State Analysis and Strategy Selection Rules

IT Guru

Business Analysts
Online Decision Support: CP + BR + ML
adding Rule Learner to find the “best” strategy

CSP (Current State) → CP Solver (the “best” strategy) → New Change Request

CP

Predefined Models and Strategies

BR

Rule Engine

ML

Rule Learner

Selected Model & Strategy

State Analysis and Strategy Selection Rules

Historical CSP States

State Analyzer

Positive & Negative Samples
Summary

- Constraint Programming empowers application developers with sophisticated decision-support (optimization) capabilities

- Proven CP + BR methodology and supporting open source and commercial tools are available in a vendor-neutral way (JSR-331)

- Online decision support may be done with

  - **CP or BR only**: Hard to create and maintain “fat” Solvers controlled by IT

  - **CP + BR**: Rule Engine recommends a CSP model and search strategy based on business rules controlled by business analysts

  - **CP + BR + ML**: Rule Learner discovers model/strategy selection rules based on historical Solver runs – “Ever-learning” decision support!