# **Henshin:**

# A Model Transformation Language and its Use for Search-Based Model Optimisation in MDEOptimiser

#### Part 2

Daniel Strüber<sup>1</sup>, Alexandru Burdusel<sup>2</sup>, Stefan John<sup>3</sup>, Steffen Zschaler<sup>2</sup>

<sup>1</sup> Universität Koblenz-Landau, <sup>2</sup> King's College London, <sup>3</sup> Philipps-Universität Marburg

> Fachtagung Modellierung February 21, 2018



### Overview

- Part 1: Henshin: A Guided Tour
  - Language
  - In Action (interactive)
  - Features
  - Applications
- Part 2: Henshin in Search-Based Model Optimization
  - Background
  - MDEOptimiser
  - In Action
    - Case 1: Class Responsibility Assignment (interactive)
    - Case 2: SCRUM Planning (interactive)



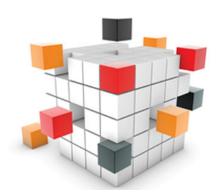
Architecture refactoring



Sprint planning



**Component** deployment







**Sprint** planning



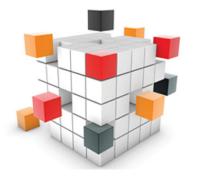
**Component** deployment

Common task: find an optimal solution among a vast number of candidates



#### **Solutions**

**Optimality** 



Architecture refactoring



Sprint planning

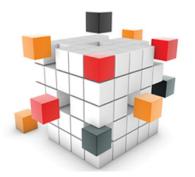


**Component** deployment

**Solutions** 

Assignment Classes→ Packages

**Optimality** 



Architecture refactoring



**Sprint** planning



**Component** deployment

**Solutions** 

Assignment Classes→ Packages

**Optimality** 

max. Cohesion min. Coupling



Architecture refactoring



Sprint planning



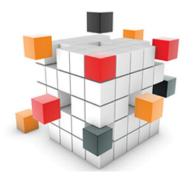
**Component** deployment

**Solutions** 

Assignment Classes→ Packages Assignment
Work Items→
Sprints

**Optimality** 

max. Cohesion min. Coupling



Architecture refactoring



Sprint planning



**Component** deployment

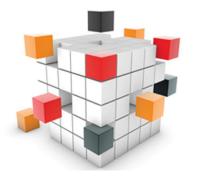
**Solutions** 

Assignment Classes→ Packages Assignment
Work Items→
Sprints

**Optimality** 

max. Cohesion min. Coupling

max. Items/Sprint max. Customer Satisfaction







Sprint planning



**Component** deployment

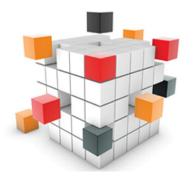
#### **Solutions**

Assignment Classes→ Packages Assignment Work Items→ Sprints Assignment Components→ Hosts

### **Optimality**

max. Cohesion min. Coupling

max. Items/Sprint max. Customer Satisfaction



Architecture refactoring



**Sprint** planning



**Component** deployment

**Solutions** 

Assignment Classes→ Packages Assignment Work Items→ Sprints Assignment Components→
Hosts

**Optimality** 

max. Cohesion min. Coupling

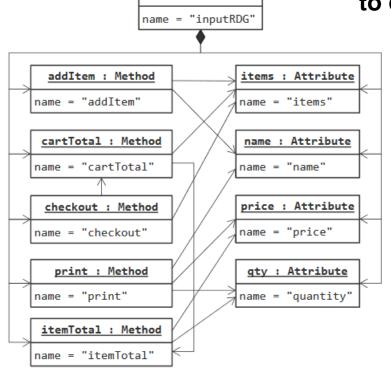
max. Items/Sprint max. Customer Satisfaction

min. Price min. Overhead

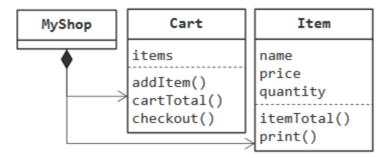
- CRA deals with the creation of high-quality object-oriented models
- For solving a particular CRA problem, one needs to decide where responsibilities, i.e., class operations and attributes, belong
- When do we have to deal with CRA problems?
  - Generating class diagrams: When migrating an application from a procedural language to an object-oriented language
  - Optimizing class diagrams: During the refactoring of an existing objectoriented model
- CRA is a computationally challenging problem
  - Huge search space!
  - Considered as an optimization problem

Input model Task: Assign methods+ attributes rdg : ClassModel to classes name = "inputRDG" addItem : Method items : Attribute name = "addItem" name = "items" cartTotal : Method name : Attribute name = "cartTotal" name = "name" checkout : Method price : Attribute name = "price" name = "checkout" print : Method qty : Attribute name = "print" name = "quantity" itemTotal : Method name = "itemTotal"

# Input model Task: Assign methods+ attributes to classes

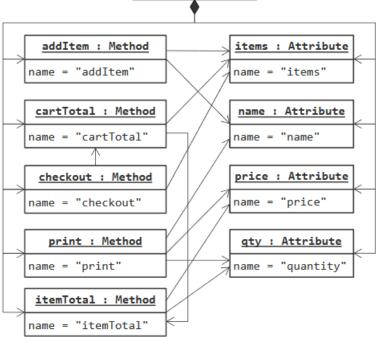


#### **Example solution**

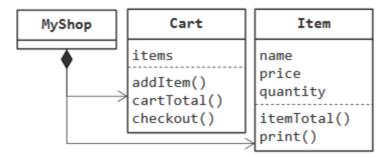


	Cart	Item
$MAI(c_i, c_i)$	3	5
MMI(c <sub>i</sub> , c <sub>i</sub> )	1	0
CohesionRatio	1.1667	0.8333
$MAI(c_i, c_j)$	1	0
$MMI(c_i, c_j)$	1	0
CouplingRatio	0.4444	0
∑ CohesionRatio	2	
∑ CouplingRatio	0.4444	
CRA-Index	1.5556	

# Input model Task: Assign methods+ attributes to classes

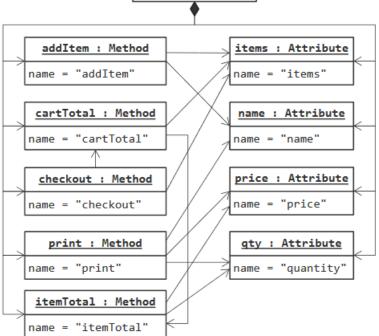


#### **Example solution**

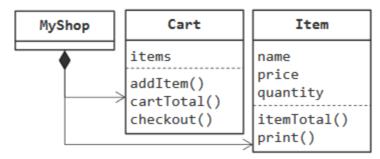


	Cart	Item
$MAI(c_i, c_i)$	3	5
MMI(c <sub>i</sub> , c <sub>i</sub> )	1	0
CohesionRatio	1.1667	0.8333
$MAI(c_i, c_j)$	1	0
$MMI(c_i, c_j)$	1	0
CouplingRatio	0.4444	0
∑ CohesionRatio	2	
∑ CouplingRatio	0.4444	
CRA-Index	1.5556	

# Input model Task: Assign methods+ attributes | rdg : ClassModel | to classes | | name = "inputRDG" |

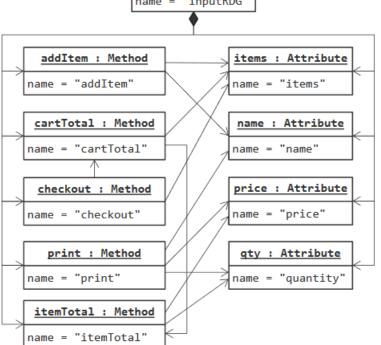


#### **Example solution**

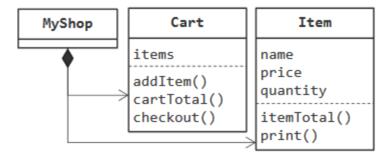


	Cart	Item
$MAI(c_i, c_i)$	3	5
MMI(c <sub>i</sub> , c <sub>i</sub> )	1	0
CohesionRatio	1.1667	0.8333
$MAI(c_i, c_j)$	1	0
$MMI(c_i, c_j)$	1	0
CouplingRatio	0.4444	0
∑ CohesionRatio	2	
∑ CouplingRatio	0.4444	
CRA-Index	1.5556	

# Input model Task: Assign methods+ attributes to classes



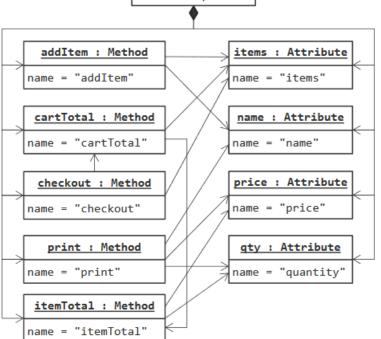
#### **Example solution**



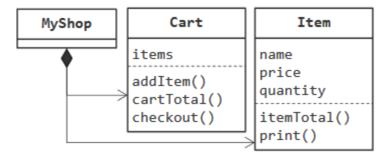
Quality

	Cart	Item
$MAI(c_i, c_i)$	3	5
MMI(c <sub>i</sub> , c <sub>i</sub> )	1	0
CohesionRatio	1.1667	0.8333
$MAI(c_i, c_j)$	1	0
$MMI(c_i, c_j)$	1	0
CouplingRatio	0.4444	0
∑ CohesionRatio	2	
∑ CouplingRatio	0.4444	
CRA-Index	1.5556	

# Input model Task: Assign methods+ attributes to classes



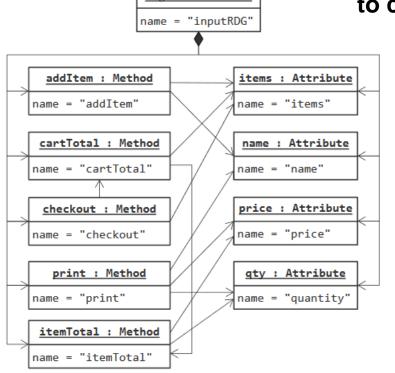
#### **Example solution**



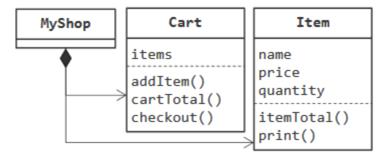
Quality

	Cart	Item
$MAI(c_i, c_i)$	3	5
MMI(c <sub>i</sub> , c <sub>i</sub> )	1	0
CohesionRatio	1.1667	0.8333
$MAI(c_i, c_j)$	1	0
$MMI(c_i, c_j)$	1	0
CouplingRatio	0.4444	0
∑ CohesionRatio	2	
∑ CouplingRatio	0.4444	
CRA-Index	1.5556	

Input model Task: Assign methods+
attributes
to classes



#### **Example solution**



Quality

	Cart	Item
$MAI(c_i, c_i)$	3	5
MMI(c <sub>i</sub> , c <sub>i</sub> )	1	0
CohesionRatio	1.1667	0.8333
$MAI(c_i, c_j)$	1	0
$MMI(c_i, c_j)$	1	0
CouplingRatio	0.4444	0
∑ CohesionRatio	2	
∑ CouplingRatio	0.4444	
CRA-Index	1.5556	

#### Fitness function

$$CRA-Index = CohesionRatio - CouplingRatio$$

$$CohesionRatio = \sum_{c_i \in Classes} \frac{MAI(c_i, c_i)}{|M(c_i)| \times |A(c_i)|} + \frac{MMI(c_i, c_i)}{|M(c_i)| \times |M(c_i) - 1|}$$

$$CouplingRatio = \sum_{\substack{c_i, c_j \in Classes \\ c_i \neq c_j}} \frac{MAI(c_i, c_j)}{|M(c_i)| \times |A(c_j)|} + \frac{MMI(c_i, c_j)}{|M(c_i)| \times |M(c_j) - 1|}$$

$$MMI(c_i, c_j) = \sum_{\substack{m_i \in M(c_i) \\ m_j \in M(c_j)}} DMM(m_i, m_j)$$

$$Courtesy \text{ of Fleck et al. [TTC 2016]}$$

$$DMA(m_i, a_j) = \begin{cases} 1 & \text{if there is a dependency between method } m_i \text{ and attribute } a_j \\ 0 & \text{otherwise} \end{cases}$$

$$DMM(m_i, m_j) = \begin{cases} 1 & \text{if there is a dependency between method } m_i \text{ and } m_j \\ 0 & \text{otherwise} \end{cases}$$

#### Fitness function

 $CRA ext{-}Index = CohesionRatio - CouplingRatio$ 

$$CohesionRatio = \sum_{c_i \in Classes} \frac{MAI(c_i, c_i)}{|M(c_i)| \times |A(c_i)|} + \frac{MMI(c_i, c_i)}{|M(c_i)| \times |M(c_i) - 1|}$$

$$CouplingRatio = \sum_{\substack{c_i, c_j \in Classes \\ c_i \neq c_j}} \frac{MAI(c_i, c_j)}{|M(c_i)| \times |A(c_j)|} + \frac{MMI(c_i, c_j)}{|M(c_i)| \times |M(c_j) - 1|}$$

$$MMI(c_i, c_j) = \sum_{\substack{m_i \in M(c_i) \\ m_j \in M(c_j)}} DMM(m_i, m_j)$$

$$Courtesy \text{ of Fleck et al. [TTC 2016]}$$

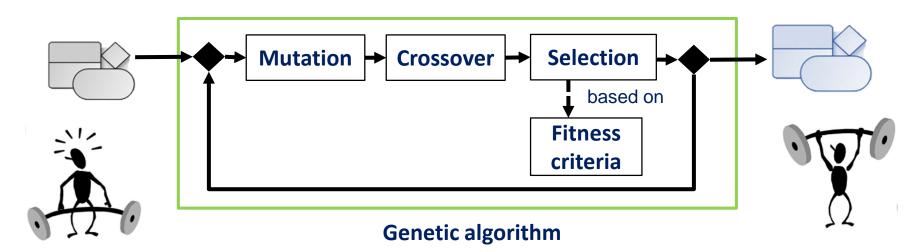
$$MAI(c_i, c_j) = \sum_{\substack{m_i \in M(c_i) \\ a_j \in A(c_j)}} DMA(m_i, a_j)$$

$$DMA(m_i, a_j) = \begin{cases} 1 & \text{if there is a dependency between method } m_i \text{ and attribute } a_j \\ 0 & \text{otherwise} \end{cases}$$

$$DMM(m_i, m_j) = \begin{cases} 1 & \text{if there is a dependency between method } m_i \text{ and } m_j \\ 0 & \text{otherwise} \end{cases}$$

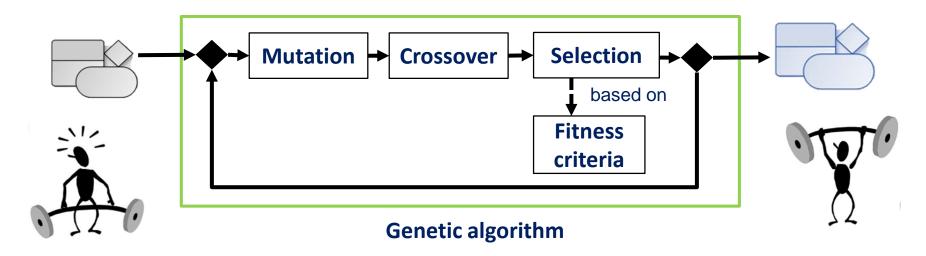
# Search-based software engineering

- Problem: Search space usually too large to enumerate all solutions
- **Solution**: *Guided search* can explore space more efficiently than humans



## Search-based software engineering

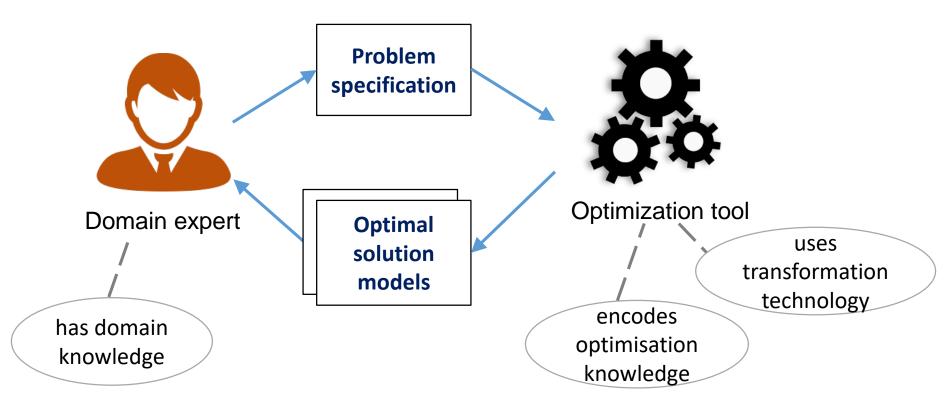
- Problem: Search space usually too large to enumerate all solutions
- **Solution**: *Guided search* can explore space more efficiently than humans



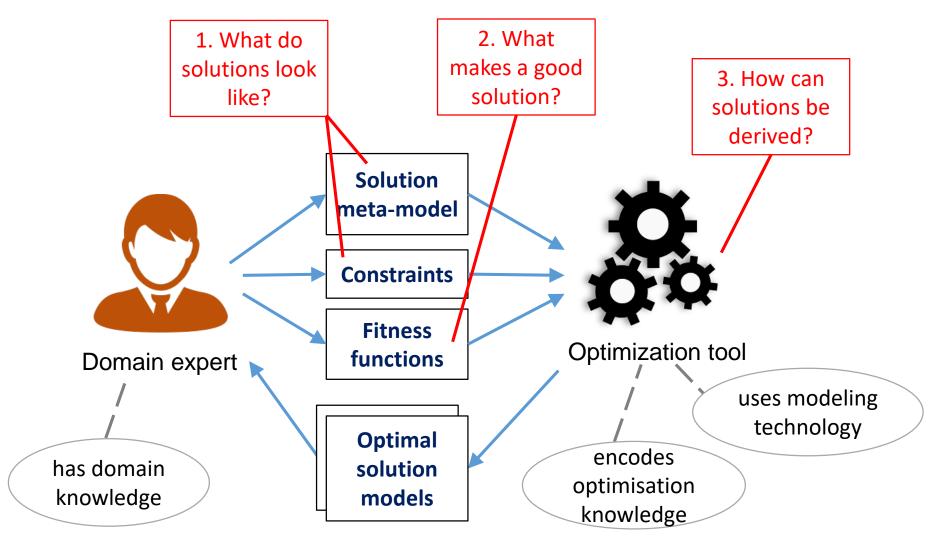
 Cost: Search algorithms need to be customized to problem at hand; substantial expertise required

## Solution: Search-based model optimisation

- Use models to describe solutions
- Standard manipulations available (model transformations!)
- Move optimisation knowledge from humans to tools



## Search-based model optimisation: what's needed?

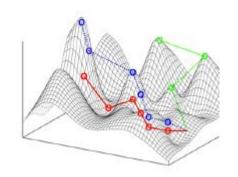


### Overview

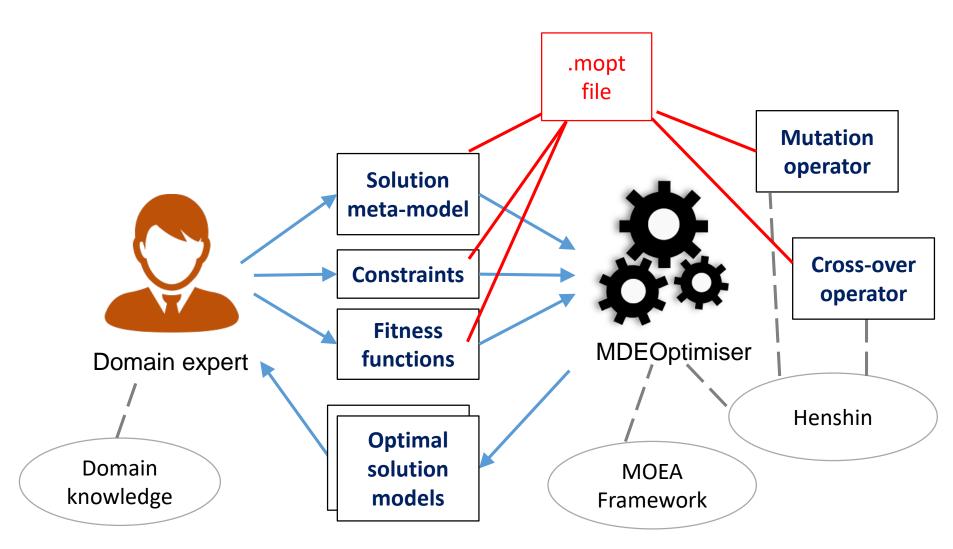
- Part 1: Henshin: A Guided Tour
  - Language
  - In Action (interactive)
  - Features
  - Applications
- Part 2: Henshin in Search-Based Model Optimization
  - Background
  - MDEOptimiser
  - In Action
    - Case 1: Class Responsibility Assignment (interactive)
    - Case 2: SCRUM Planning (interactive)

## MDE Optimiser

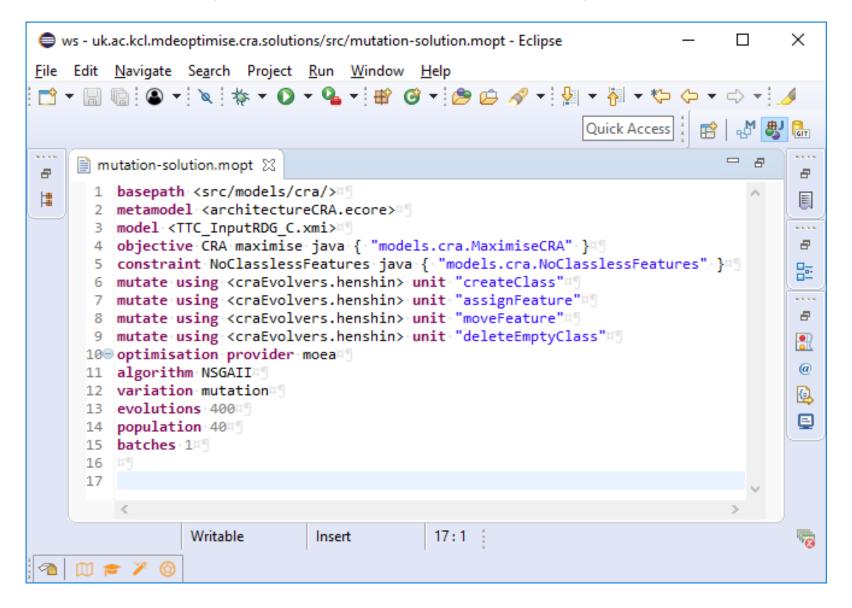
- Multi-objective optimization
- Directly over models (no separate solution encoding)
- Specification language + kernel
- Uses Henshin to specify evolutionary operators



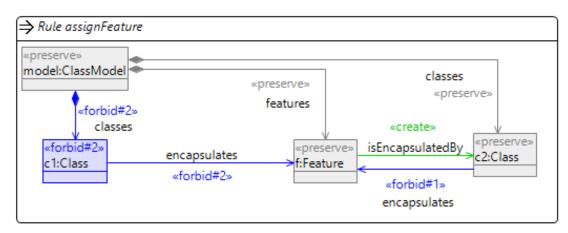
## MDE Optimiser

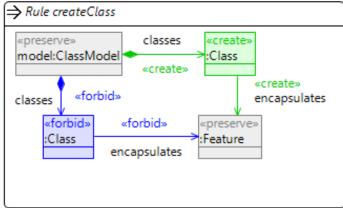


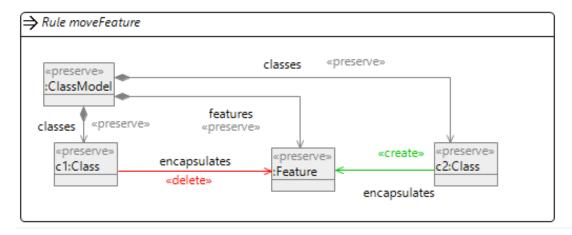
## Problem specification in MDEOptimiser: CRA

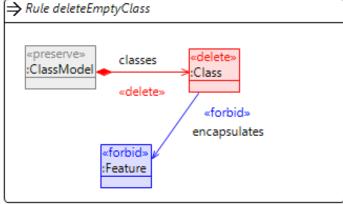


## Pre-defined mutation rules for CRA









## Pre-defined objective function for CRA

```
1 package models.cra

   2 119
   3⊕ import org.eclipse.emf.ecore.EObject
   5 119
   6 class MaximiseCRA extends AbstractModelQueryFitnessFunction {□
   7 ¤¶
        override double computeFitness(EObject model) { [1]
             val cohesion = calculateCohesionRatio(model);
             val coupling = calculateCouplingRatio(model);
  11 49
             return (cohesion - coupling) * -149
  12 >>
  13 »
  14 ¤¶
  15⊕ >>
         def double calculateCohesionRatio(EObject classModel) {[]
         def double calculateCouplingRatio(EObject classModel) {[]
  52⊕ >>
         def double calculateCouplingRatio(EObject classSource, EObject classModel) {[]
  64⊕ >>
         def mai(EObject classSource, EObject classTarget) {[]
 112⊕ >>
         def mmi(EObject classSource, EObject classTarget) {[]
 129⊕ >>
             override getName() { [4]
△146⊜ »
             return "Maximise CRA" [4]
 147 >>
         }¤¶
 148 >>
 149 >>
```

### Pre-defined constraint for CRA

```
1 package models.cra
 3⊕ import org.eclipse.emf.common.util.EList[
 6 class NoClasslessFeatures extends AbstractModelQueryFitnessFunction {□
△ 8⊖ » override computeFitness(EObject model) {

[*]
   >> var fitness = (model.getFeature("features") as EList<EObject>).

   10
   » » println("Classless features: " + fitness)

 12 » » return fitness; I
 13 » }¤¶
 14 » ¤¶
△15⊖ » override getName() {¤¶
          return "No classless features" "
16 »
 17 >>
      }¤9
 18 >>
 20
```

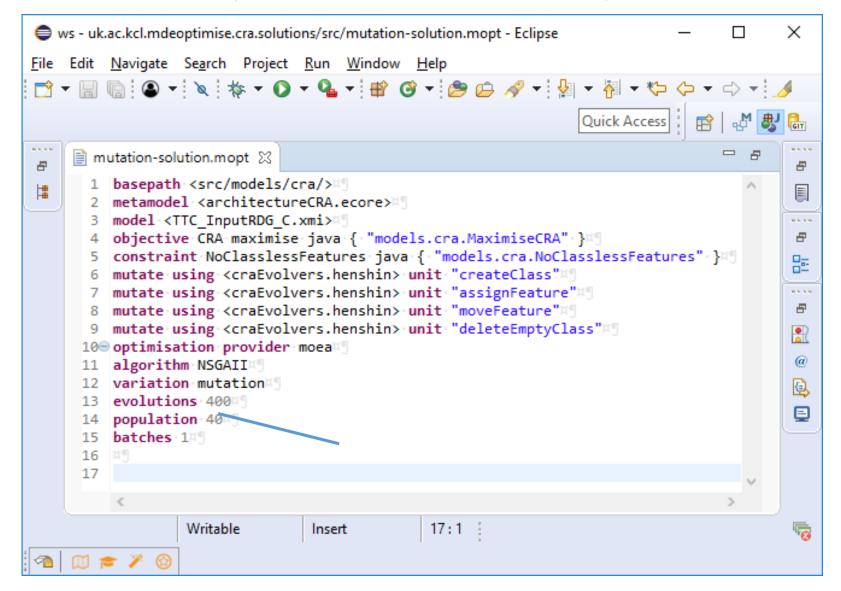
### Pre-defined constraint for CRA

```
NoClasslessFeatures,xtend ⋈
  1 package models.cra

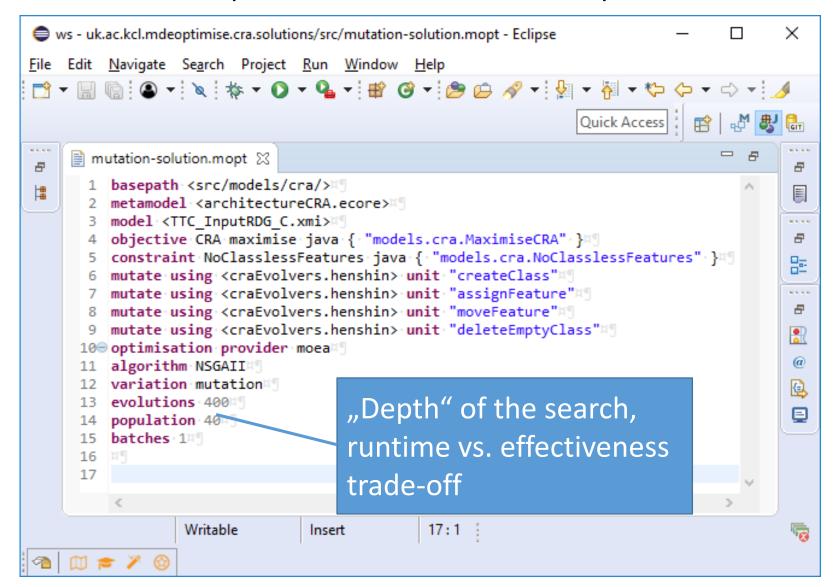
  3⊕ import org.eclipse.emf.common.util.EList[
  6@ class NoClasslessFeatures extends AbstractModelQueryFitnessFunction { [4]
△ 8⊖ » override computeFitness(EObject model) { [1]
   >> var fitness = (model.getFeature("features") as EList<EObject>).

   11 » » println("Classless features: " + fitness) []
 12 » » return fitness; [1]
 13 » }¤¶
 14 » III
△15⊖ » override getName() { [4]
          return "No classless features" 49
 16 »
 17 >>
       }¤¶
 18 >>
                     Solution's "fitness" w.r.t. a constraint:
 20
                     How far is the solution away from
                     fulfilling the constraint?
```

## Problem specification in MDEOptimiser: CRA



## Problem specification in MDEOptimiser: CRA



### Overview

- Part 1: Henshin: A Guided Tour
  - Language
  - In Action (interactive)
  - Features
  - Applications
- Part 2: Henshin in Search-Based Model Optimization
  - Background
  - MDEOptimiser
  - In Action
    - Case 1: Class Responsibility Assignment (interactive)
    - Case 2: SCRUM Planning (interactive)

- 1. Import projects
- 2. View the specification of CRA case
- 3. Set up and apply run configuration
- 4. View created results
- 5. Design a good mutation operator

## 1. Import projects

- 2. View the specification of CRA case
- 3. Set up and apply run configuration
- 4. View created results
- 5. Design a good mutation operator

## Import projects

- In Eclipse, do File → Import... → General → Existing Projects Into Workspace → Next
- Do Select Archive File → Choose optimization-cases.zip
- By now, you should be an expert on importing projects. :-)

- 1. Import projects
- 2. View the specification of CRA case
- 3. Set up and apply run configuration
- 4. View created results
- 5. Design a good mutation operator

## View the specification of CRA case

In the Package Explorer, navigate to project
 uk.ac.kcl.mdeoptimise.cra.solutions, folder src/models.cra

Have a look at the files:

MDEOptimiser spec: cra-solution.mopt

meta-model: architectureCRA.ecore

five input models: TTC\_InputRDG\_<A-E>.xmi

objective function: MaximiseCRA.xtend

constraint: NoClasslessFeatures.xtend

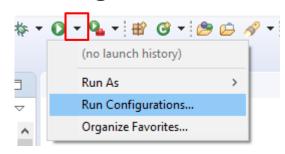
mutation operators: craEvolvers.henshin\_diagram

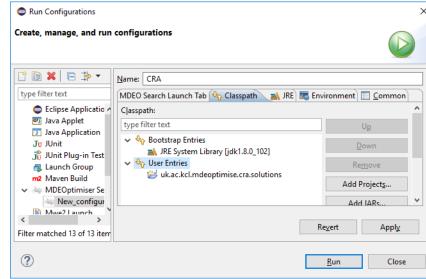
- 1. Import projects
- 2. View the specification of CRA case
- 3. Set up and apply run configuration
- 4. View created results
- 5. Design a good mutation operator

# Set up and apply run configuration

To execute **cra-solution.mopt**, create a new run configuration:

- Click on the triangle next to the Run Icon, select Run Configurations...
- Right click on MDEOptimiser Search, select New. As Name for the configuration, enter: CRA
- Do Browse Workspace -> Select Source cra-solution.mopt
- In the Classpath tab, add the current project as a User Entry, using Add Projects...
- Hit Apply and Run. If everything works correctly, you will see console output like on the right.





Calculated CRA: 1.3707386363636362 Classless features: 0 Calculated CRA: 0.5790719696969693 Classless features: 0

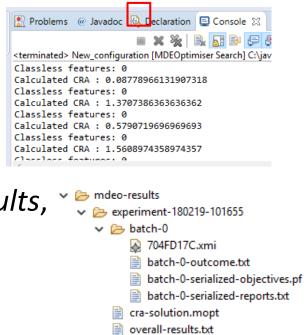
- 1. Import projects
- 2. View the specification of CRA case
- 3. Set up and apply run configuration
- 4. View created results
- 5. Design a good mutation operator

## View created results

After approximately 15 seconds, the run is finished: No new console outputs, "Terminate" switch has turned gray.

Results are in a newly created folder:
In the Package Explorer, click on *mde-results*, hit F5 (refresh) and find the folder.

Open **overall-results.txt**. This shows the execution time and an overview of the best solution with its CRA index.



- 1. Import projects
- 2. View the specification of CRA case
- 3. Set up and apply run configuration
- 4. View created results
- 5. Design a good mutation operator

## Design a good mutation operator

**Problem**: Search is time-consuming.

**Solution**: Improve the search by designing good evolutionary operators.

A mutation operator m1 is *better* than a mutation operator m2, if the solutions found using m1 are better than those found using m2 (assuming an otherwise equal configuration).

**Task**: Design a better mutation operator for the CRA case than the given one.

Reference values:

Model	Α	В	С	D	E
CRA	1.6	2.2	1.8	2.4	-11.6

**Hint**: What are desirable structures from the perspective of the objective function, and how to create them?

## Overview

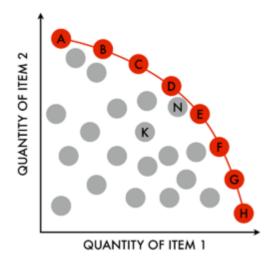
- Part 1: Henshin: A Guided Tour
  - Language
  - In Action (interactive)
  - Features
  - Applications
- Part 2: Henshin in Search-Based Model Optimization
  - Background
  - MDEOptimiser
  - In Action
    - Case 1: Class Responsibility Assignment (interactive)
    - Case 2: SCRUM Planning (interactive)

## **SCRUM Planning**

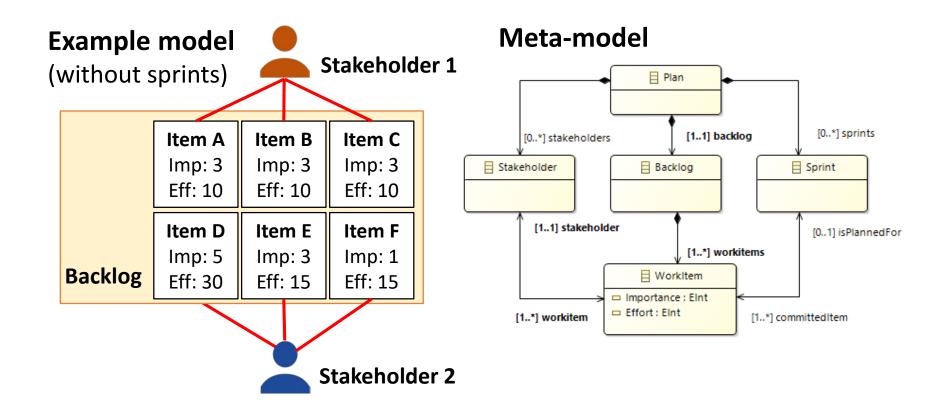
 Have a backlog of work items for different customers.
 Need to allocate work items to sprints.



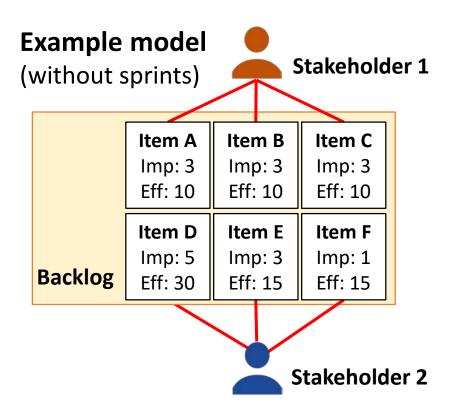
- Goals: Be fast and keep customers satisfied.
- Multi-objective optimisation problem: pareto front instead of one best solution



# **SCRUM Planning**



# SCRUM Planning: Objectives and constraints



**Objective 1:** Sprints are as balanced as possible.

**Objective 2:** *Maximize customer satisfaction -> next slide* 

**Constraint 1:** *Number of sprints is below a given maximum* 

**Constraint 2:** Every work item is assigned to a sprint.

# SCRUM Planning: Customer satisfaction index











Stakeholder 1 Stakeholder 2 Stakeholder 3

Stakeholder 4

Stakeholder 5

Sprint 1 Capacity: 28

WorkItem 2

WorkItem 4

WorkItem 5

WorkItem 7

WorkItem 8

WorkItem 10

Sprint 2 Capacity: 27

WorkItem 1

WorkItem 3

WorkItem 6

WorkItem 11

WorkItem 9

#### Good solution

Customer Satisfaction Index

ilidex

0.80

#### Sprint 1 Capacity: 28

WorkItem 1

WorkItem 2

WorkItem 3

WorkItem 4

WorkItem 8

WorkItem 10

#### Sprint 2 Capacity: 27

WorkItem 7

WorkItem 5

WorkItem 6

WorkItem 11

WorkItem 9

#### Bad solution

Customer Satisfaction Index

5.39

## Low means good:

#### Indicates low standard deviation

## Backlog

workitem 1		
Stakeholder:	1 Importance: 3	Effort: 8

WorkItem 2

Stakeholder: 1 Importance: 8 Effort: 5

WorkItem 3

Stakeholder: 1 Importance: 8 Effort: 8

WorkItem 4

Stakeholder: 2 Importance: 1 Effort: 1

WorkItem 5

Stakeholder: 2 Importance: 1 Effort: 8

WorkItem 6

Stakeholder: 3 Importance: 5 Effort: 5

WorkItem 7

Stakeholder: 3 Importance: 3 Effort: 8

WorkItem 8

Stakeholder: 3 Importance: 3  $\,$  Effort: 1

WorkItem 9

Stakeholder: 4 Importance: 1 Effort: 3

WorkItem 10

Stakeholder: 5 Importance: 5 Effort: 5

WorkItem 11

Stakeholder: 5 Importance: 5 Effort: 3

# SCRUM Planning: Customer satisfaction index











Stakeholder 1 Stakeholder 2 Stakeholder 3

Stakeholder 4

Stakeholder 5

Sprint 1 Capacity: 28

WorkItem 2

WorkItem 5

WorkItem 7

WorkItem 8

WorkItem 10

WorkItem 4

WorkItem 6

WorkItem 11

WorkItem 1

WorkItem 3

Sprint 2 Capacity: 27

WorkItem 9

Good solution

Customer Satisfaction Index

0.80

## Satisfaction of customer c in sprint s

$$sat(c,s) = \sum_{i \in s.items,i.cust=c} i.importance$$

## **Customer satisfaction index of plan p**

```
csi(p) = std.dev_{c \in p.customers}(std.dev_{s \in p.sprints}(sat(c,s)))
```

#### Backlog WorkItem 1 Stakeholder: 1 Importance: 3 Effort: 8 WorkItem 2 Stakeholder: 1 Importance: 8 Effort: 5 WorkItem 3 Stakeholder: 1 Importance: 8 Effort: 8 WorkItem 4 Stakeholder: 2 Importance: 1 Effort: 1 WorkItem 5 Stakeholder: 2 Importance: 1 Effort: 8 WorkItem 6 Stakeholder: 3 Importance: 5 Effort: 5 WorkItem 7 Stakeholder: 3 Importance: 3 Effort: 8 WorkItem 8 Stakeholder: 3 Importance: 3 Effort: 1 WorkItem 9 Stakeholder: 4 Importance: 1 Effort: 3 WorkItem 10 Stakeholder: 5 Importance: 5 Effort: 5 WorkItem 11 Stakeholder: 5 Importance: 5 Effort: 3

## Problem specification in MDEOptimiser: SCRUM

```
metamodel <planning.ecore>!!
model <sprint-planning-model-5-stakeholders-119-items.xmi>□
objective MinimiseCustomerSatisfactionIndex minimise java 🖑
 -{ "models.scrum.MinimiseCustomerSatisfactionIndex" } "
|objective MinimiseSprintEffortDeviation minimise java | 19
 -{ "models.scrum.MinimiseSprintEffortDeviation" -} # 
| constraint HasNoUnassignedWorkItems java □
 -{-"models.scrum.HasNoUnassignedWorkItems"-}#¶
·{·"models.scrum.HasTheAllowedMaximalNumberOfSprints"·}¤¶
mutate using <sprint-repair.henshin> unit "createSprint" [4]
mutate using <sprint-repair.henshin> unit "addItemToSprint" [4]
mutate using <sprint-repair.henshin> unit "createSprint lb repair" [4]
mutate using <sprint-repair.henshin> unit "move item between sprints" [4]
mutate using <sprint-repair.henshin> unit "deleteSprint lb repair" [4]
optimisation provider moea 🛒
algorithm NSGAII I
variation mutation [9]
evolutions 500
population 30×9
batches 114
```

- 1. View the specification of SCRUM case
- 2. Set up and apply run configuration
- 3. View created results

# View the specification of the SCRUM case

 In the Package Explorer, navigate to project uk.ac.kcl.mdeoptimise.scrum.planning.solutions, folder src/models.scrum

Have a look at the files:

MDEOptimiser spec: scrum-planning.mopt

meta-model: planning.ecore

input models: input directory

objective functions: MinimiseSprintEffortDeviation.xtend

and MinimiseCustomerSatisfactionIndex.xtend

constraints: HasTheAllowedMaximalNumberOfSprints.xtend

and HasNoUnassignedWorkItems.xtend

mutation operators: sprint-repair.henshin\_diagram

- 1. View the specification of SCRUM case
- 2. Set up and apply run configuration
- 3. View created results

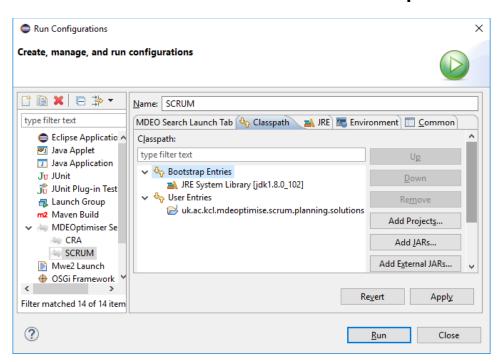
# Set up and apply run configuration

Similar to the CRA case, create a new configuration

Name: SCRUM

Source: scrum-planning.mopt

Make sure to add the corresponding classpath entry:



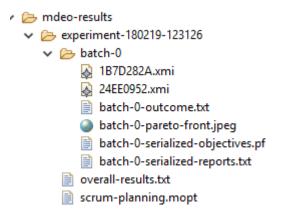
- 1. View the specification of SCRUM case
- 2. Set up and apply run configuration
- 3. View created results

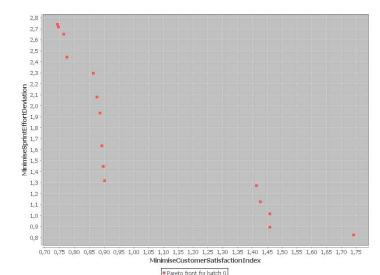
## View created results

Results are, again, in a newly created folder: In the Package Explorer, click on *mde-results*, hit F5 (refresh) and find the folder.

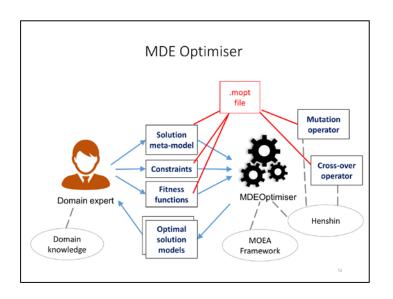
Open **overall-results.txt**. This shows the execution time and an overview of the best solutions with their objective values.

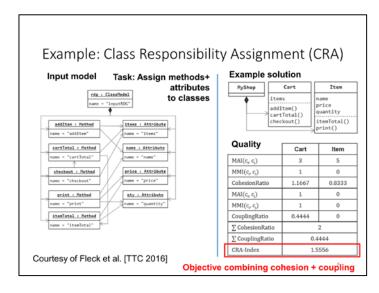
Since we have multiple solutions in general, an image file **batch-0-pareto-front.jpeg** is generated showing the pareto front.





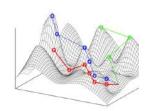
# Summary of Part 2





# Have a backlog of work items for different customers. Need to allocate work items to sprints. Goals: Be fast and keep customers satisfied. Multi-objective optimisation problem: pareto front instead of one best solution





Further information: www.eclipse.org/henshin mde-optimiser.github.io