# DPB: A Benchmark for Design Pattern Detection tools

Francesca Arcelli Fontana, Andrea E. F. Caracciolo, Marco Zanoni







30/03/2012 - CSMR 2012 - Szeged, Hungary

#### **Main Goal**

# Define a system allowing users to **compare** the **quality** of Design Pattern Detection (DPD) tools results

#### Who cares?

- End users: to be able to choose a tool
- Researchers: compare existing techniques/ reuse valid techniques

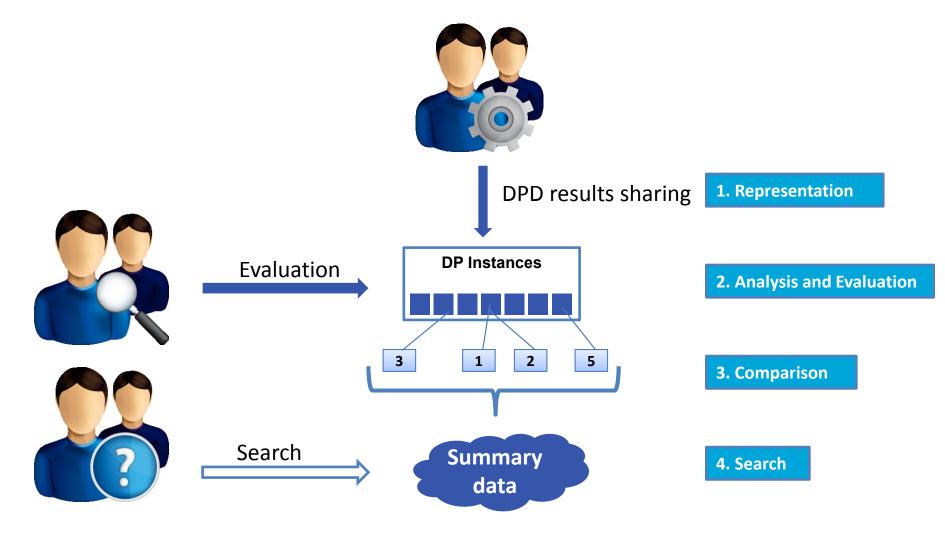
#### **Related works**

- DEEBEE [Fülöp et al., 2008]
  - x Usability
  - x Data model

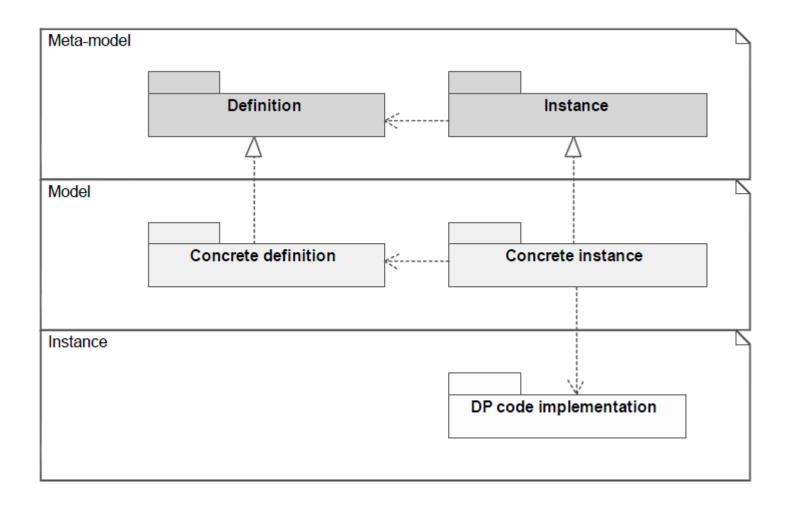
- ✓ Open web application
- ✓ Interesting choice of functionalities

- P-MARt [Guéhéneuc, 2007]
  - x No support for discussion
  - x No way to measure reliability
- ✓ Pattern instances identified by experts

# **Proposed Solution**

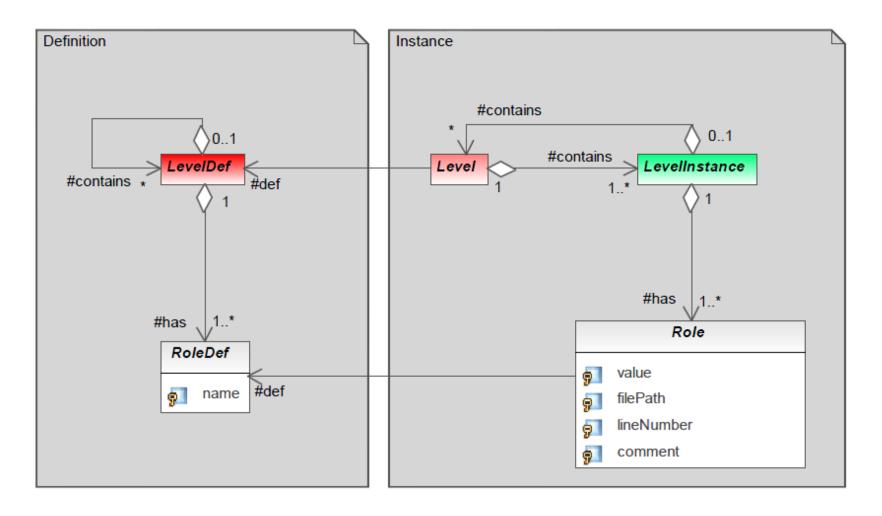


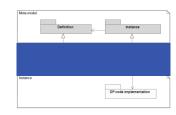
## Representation (panoramic)



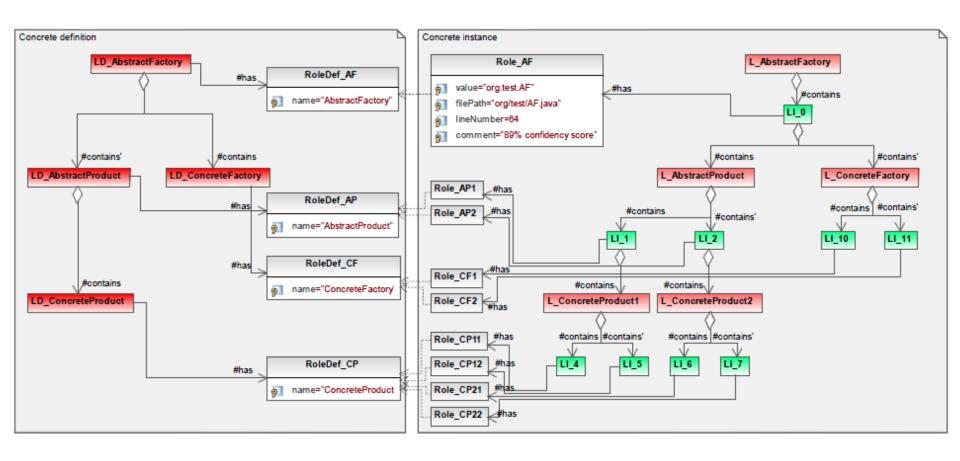
# Model Concrete definition Concrete instance Instance DP code implementation

## Representation (meta-model)

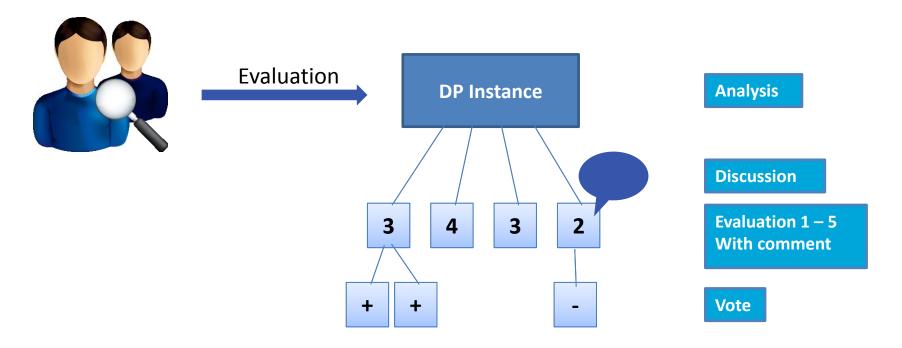




## Representation (model)



#### **Analysis and Evaluation**

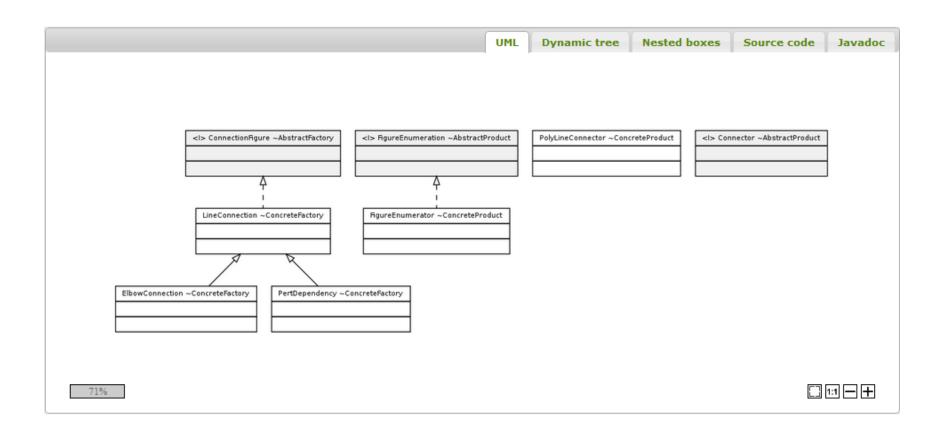


$$rating(instance) = \frac{\sum_{i=1}^{|evals|} eval_i \cdot votesBalance_i}{\sum_{i=1}^{|evals|} votesBalance_i}$$

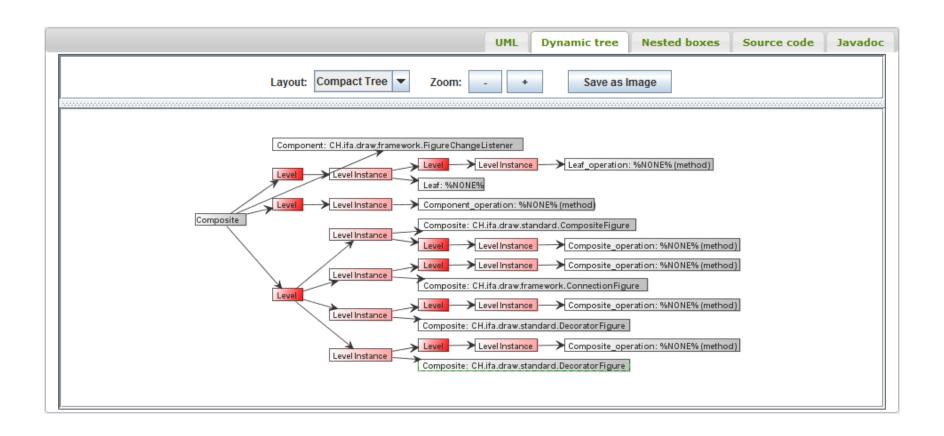
**Instance score** 

$$votesBalance_i = max(def + votes_i^+ - votes_i^-, 0)$$

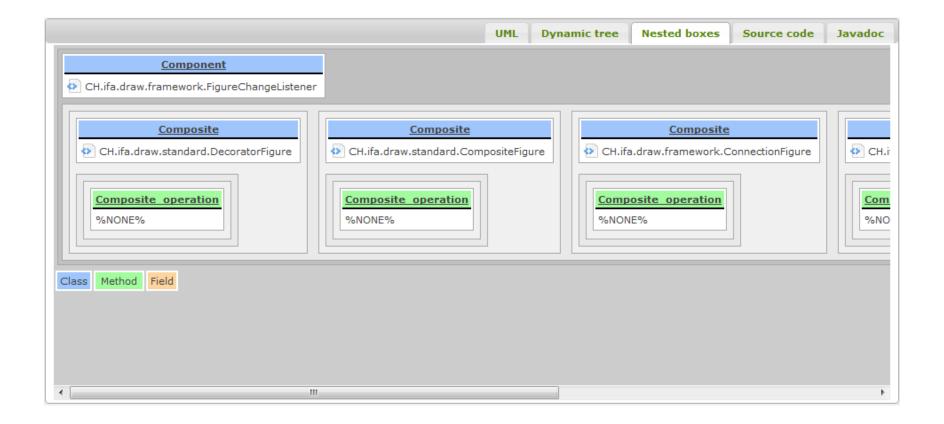
## Analysis and Evaluation (UML diagram)



#### Analysis and Evaluation (structural view)



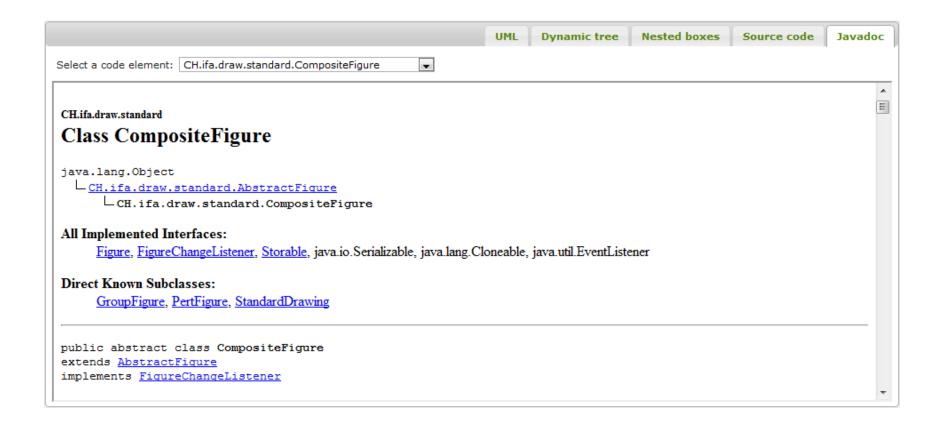
#### Analysis and Evaluation (structural view 2)



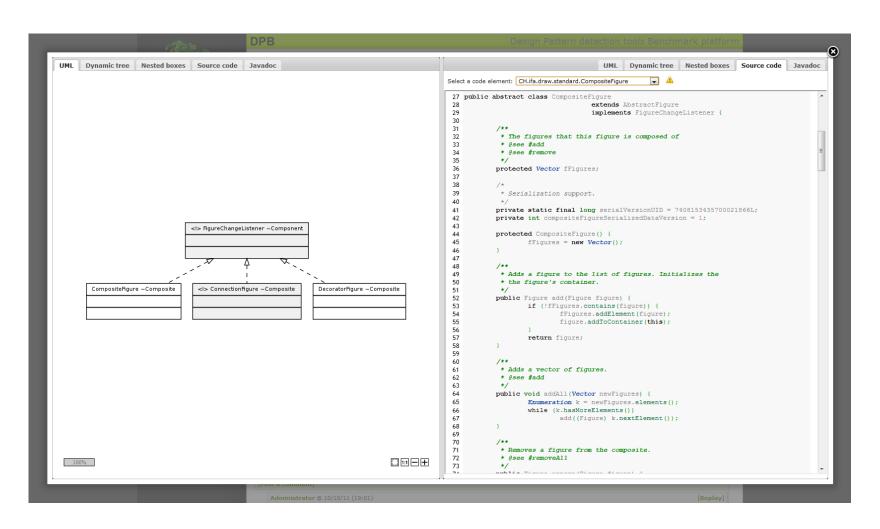
#### Analysis and Evaluation (source code)

```
UML
                                                                         Dynamic tree
                                                                                        Nested boxes
                                                                                                       Source code
                                                                                                                     Javadoc
Select a code element: CH.ifa.draw.standard.DecoratorFigure
     public abstract class DecoratorFigure
                                       extends AbstractFigure
  40
  41
                                       implements FigureChangeListener {
  42
              /**
  43
               * The decorated figure.
  44
  45
              protected Figure fComponent;
  46
  47
  48
  49
               * Serialization support.
  50
  51
              private static final long serialVersionUID = 8993011151564573288L;
              private int decoratorFigureSerializedDataVersion = 1;
  52
  53
  54
              public DecoratorFigure() { }
  55
              /**
  56
  57
               * Constructs a DecoratorFigure and decorates the passed in figure.
  58
  59
              public DecoratorFigure(Figure figure) {
  60
                      decorate (figure);
```

#### Analysis and Evaluation (javadoc)



## Analysis and Evalutation (side by side view)



#### **Analysis and Evaluation** (evaluations)



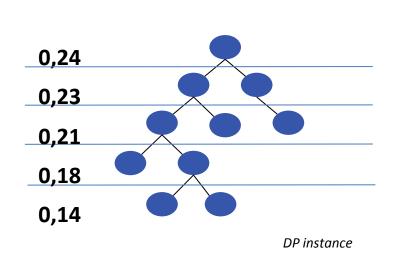
# Comparison (system analysis comparison)

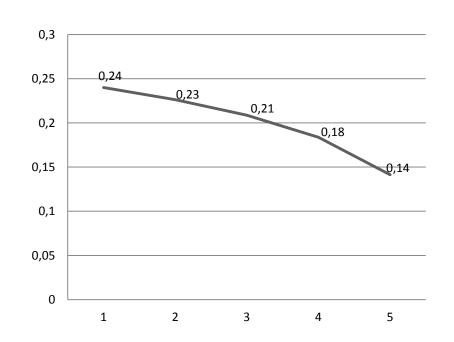
Compare two Analyses				
Project: MapperXML 1.9.7 ▶				
Analysis 1: Analysis #8 (tool: DPD Tool 4.5)				
Analysis 2: Analysis #42 (tool: P-MARt )				
Design Pattern: Singleton				
	Analysis #8			
	<u>#498</u>	<u>#513</u>	<u>#517</u>	
<u>#1237</u>	67%	33%	33%	
Analysis #42 <u>#1242</u>	33%	67%	33%	
<u>#1247</u>	33%	33%	67%	
•				
View Options:				
View layout: Table ▼				
Hide rows/columns below 0 % ▼				
Color scheme: Blue gradient 💌				
Highlight highest value of each: Row				

#### Comparison (algorithm: weights setting)

$$depthScore_{depth} = \log_{10}(treeHeight - depth) + 1$$

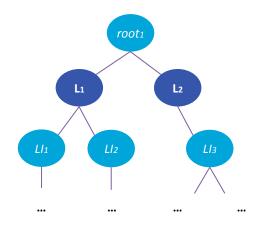
$$weight_i = \frac{depthScore_i}{\sum_{j=0}^{treeH} depthScore_i \cdot |levels_i|}$$

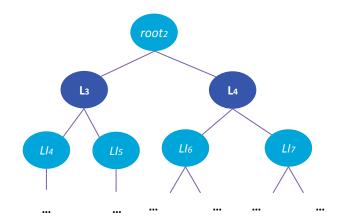




$$sim(inst_1, inst_2) = \begin{cases} simLI(root_1, root_2) \cdot weight_0 \\ + \sum_{i=1}^n simL(subL_{1,i}, subL_{2,i}, 1) & \text{se } simLI(root_1, root_2) > 0 \\ \\ 0 & \text{altrimenti} \end{cases}$$

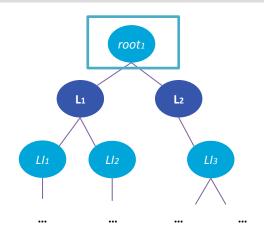
#### Similarity = $simL(root_1, root_2) * weight_0 + simL(L_1, L_3) + simL(L_2, L_4) + ...$

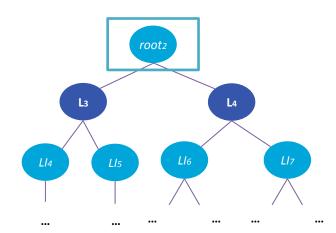




$$sim(inst_1, inst_2) = \begin{cases} simLI(root_1, root_2) \cdot weight_0 \\ + \sum_{i=1}^n simL(subL_{1,i}, subL_{2,i}, 1) & \text{se } simLI(root_1, root_2) > 0 \\ \\ 0 & \text{altrimenti} \end{cases}$$

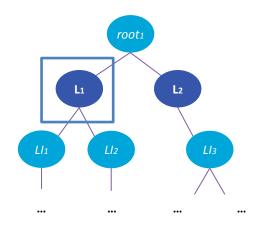
Similarity =  $simLl(root_1, root_2)$  \* weight<sub>0</sub> +  $simL(L_1, L_3)$  +  $simL(L_2, L_4)$  + ...

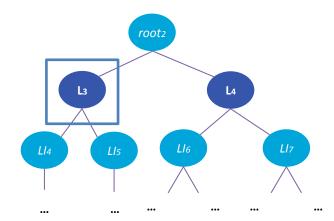




$$sim(inst_1, inst_2) = \begin{cases} simLI(root_1, root_2) \cdot weight_0 \\ + \sum_{i=1}^n simL(subL_{1,i}, subL_{2,i}, 1) & \text{se } simLI(root_1, root_2) > 0 \\ \\ 0 & \text{altrimenti} \end{cases}$$

Similarity =  $simL(root_1, root_2) * weight_0 + simL(L_1, L_3) + simL(L_2, L_4) + ...$ 

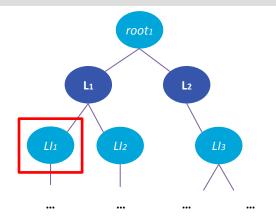


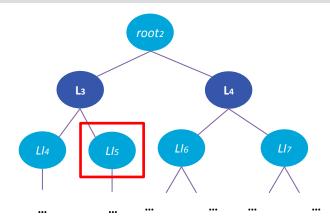


$$simL(l_1, l_2, depth) = \sum_{i=1}^{n} simLI(subLi_{1,i}, mostSim(subLi_{1,i}, subLi_{2,i}))$$

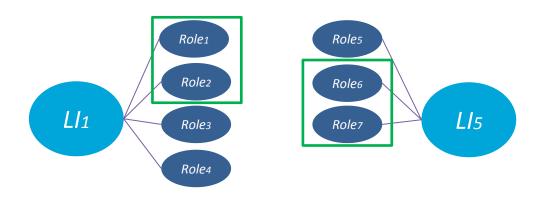
$$\cdot \frac{weight_{depth}}{n} + \sum_{i=1}^{m} simL(subL_{1,i}, subL_{2,i}, depth + 1)$$

Similarity =  $simL(root_1, root_2)$  \* weight<sub>0</sub> +  $simL(L_1, L_3)$  +  $simL(L_2, L_4)$  + ...  $simL(L_1, L_3)$  = [  $simL(L_1, L_1, L_2)$  +  $simL(L_1, L_1, L_2)$  +  $simL(L_1, L_2, L_3)$  ] \* weight<sub>1</sub> / 2



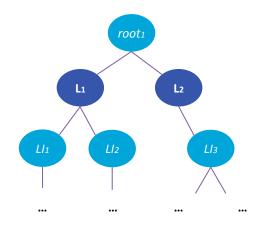


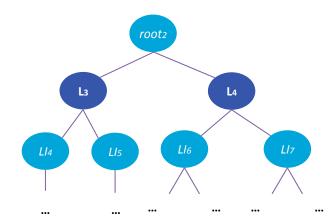
$$simLI(li_1, li_2) = \frac{|sharedRoles|}{\max(|subRoles_1|, |subRoles_2|)}$$



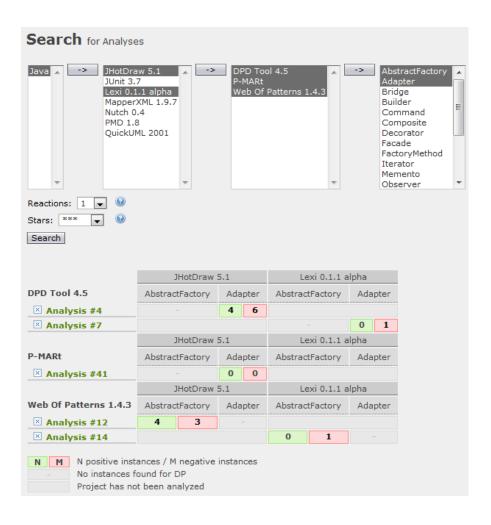
$$sim(inst_1, inst_2) = \begin{cases} simLI(root_1, root_2) \cdot weight_0 \\ + \sum_{i=1}^{n} simL(subL_{1,i}, subL_{2,i}, 1) & \text{se } simLI(root_1, root_2) > 0 \\ \\ 0 & \text{altrimenti} \end{cases}$$

Similarity =  $simL(root_1, root_2) * weight_0 + simL(L_1, L_3) + simL(L_2, L_4) + ...$ 



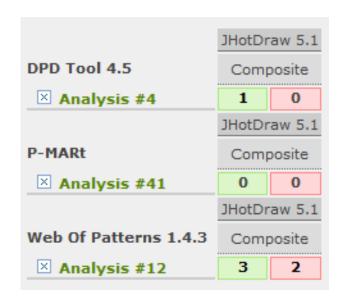


#### Search (example)

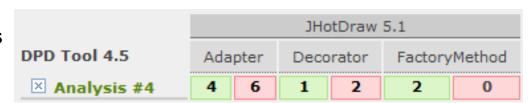


#### Search (results analysis)

**Comparison for the same context** 



**Comparison respect to patterns** 



## Collaboration, beta-testing and feedback

Günter Kniesel and Alex Binun (Universität Bonn, Germany)



Nikos Tsantalis (University of Alberta, Canada)

Yann-Gaël Guéhéneuc (École Polytechnique de Montréal, Canada)

#### **Conclusions**

- A benchmark for DPD tools
  - Specific meta-model for DP representation
  - A new algorithm for DP instances comparison
  - Largely Experimented

www.essere.disco.unimib.it/DPB

#### **Future work**

- Simplify the results importing process
  - Compatibility extension for other meta-models
  - Web service for results upload
- Add statistical analyses
- Think at new interaction types
  - Eclipse plug-in

# QQA

#### **Statistics**

- The platform is currently poplated with:
  - 2 DPD tools (WOP and DPD-tool(Tsantalis))
  - 1 verified instances dataset (P-Mart)
  - 20+ system analysis
  - 700+ DP instances.
  - 160+ evaluations.
- There are 36 registered users.
- Access statistics:
  - 900+ visits e 360 unique users.
  - 13.000+ page visualization.
  - 15 minutes of average spent time on the web site

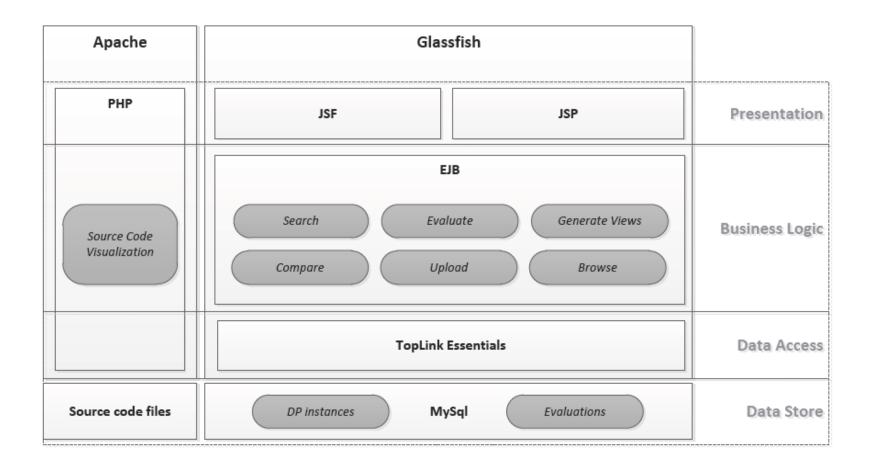
#### **Meta-model requirements**

- 1. Minimum effort to understand how to define a new DP instance
- 2. Compact representation (to make data store and elaboration faster).
- 3. Support for DP instances having multi-value roles.
- 4. Flexible enough to support any DP definition
- Requisiti soddisfatti
  - DPB: all ☺
  - DPDX: only 3 and 4
    - Quite big and too generic in many cases
    - Models code is not very readable
    - The lack of a shared set of Schema meta-models does not allow to make the models reallly interoperable
  - KDM: needs extension
  - FAMIX, Dagstuhl, Marple, other: only code representation

## Principles for the definition of the specification

- Multiplicity principle: Given the level A and B, having respectively the associed roles (A1,A2, ...,An) and (B1,B2, ...,Bn), it is possible to state that B is sublevel of A if (and only if) for each instance of any role associated to level A, at least one instance exists of each role associated to level B. In other words, the multiplicity rate between the number of instances of any role Ai (belonging to A) and any role Bj (belonging to B) is always 1:1 or 1:many.
- Coupling principle: Two roles A1 and A2 are associated to the same levele, if every time an element playing a role A1 is present it is possible to observe one and only one element playing role A2.

# **Technologies**



# **Example: DP instance scoring**

#### evaluations:

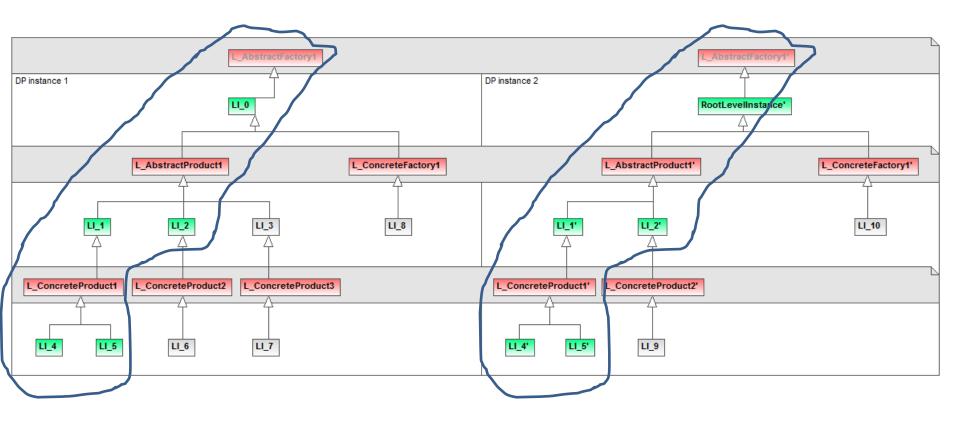
- evaluation 1: 4 stars (3 agreements / 1 disagreement)
- evaluation 2: 3 stars (8 agreements / 0 agreements)
- evaluation 3: 1 stars (0 agreements / 8 agreements)
- evaluation 4: 4 stars (1 agreement / 4 disagreements)
- formula applications brings these results:
  - votesBalance1 = 3 + 3 1 = +5
  - votesBalance2 = 3 + 8 0 = +11
  - votesBalance3 = 3 + 0 8 = -5 (< 0, => votesBalance3 = 0)
  - votesBalance4 = 3 + 1 4 = 0 (< 0, => votesBalance4 = 0)
- Result:

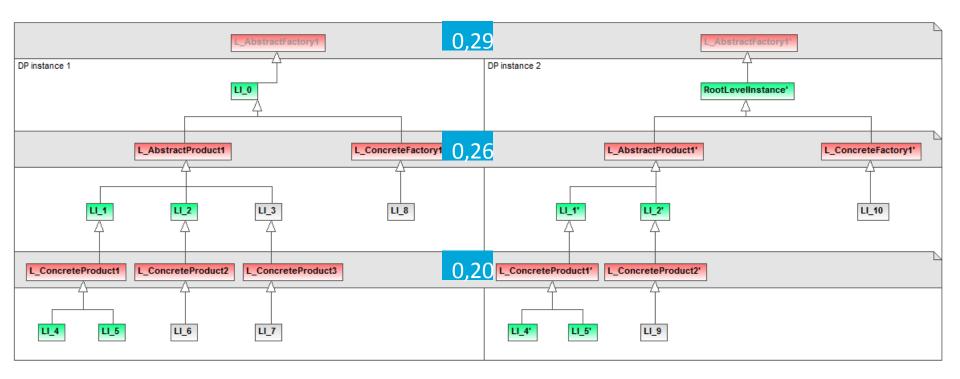
$$- rating(instance) = (4 * 5 + 3 * 11 + 1 * 0 + 4 * 0) / (5 + 11 + 0 + 0)$$
$$= (20 + 33) / 16$$
$$= 3.31$$

#### **Online examples**

- System analysis:
  - http://essere.disco.unimib.it:8080/DPBWeb/faces/Analysis.jsp?id=12
- Instance:
  - http://essere.disco.unimib.it:8080/DPBWeb/faces/ViewDP.jsp?id=692&dpa=83
- Search:
  - http://essere.disco.unimib.it:8080/DPBWeb/faces/Search.jsp?new=1
  - Java JHotDraw+QuickUML DPD+WOP AbstractFactory+Adapter+Bridge
- Comparison:
  - http://essere.disco.unimib.it:8080/DPBWeb/faces/Compare.jsp?new=1
  - JHotDraw #4 #41 Strategy
  - 64%
- Definition:
  - http://essere.disco.unimib.it:8080/DPBWeb/faces/Doc\_DpDef.jsp?id=28&name=AbstractFactory
- Browse:
  - http://essere.disco.unimib.it:8080/DPBWeb/faces/Browse.jsp

# Similarity algorithm – Example





Detection tools

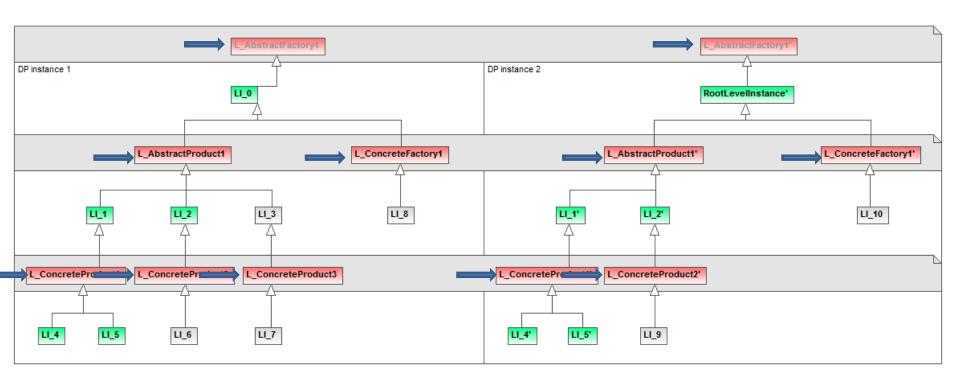
#### **Calculate weights** (based on definition's structure; see slide 4):

```
# # depthScore_1: log10(3-1)+1 = 1,3
# depthScore_2: log10(3-2)+1 = 1

Sum(depthScore_i * numLevels_i) = 1,48 * 1 + 1,3 * 2 + 1 * 1 = 5,08
# weight_0 = 1,48/5,08 = 0,29
# # weight_1 = 1,3/5,08 = 0,26
# weight_2 = 1/5,08 = 0,20

DPB: A Benchmark for Design Pattern
```

depthScore 0: log10(3-0)+1 = 1,48



```
similarity = 1 * weight_0 + simL(L_AP1, L_AP1',1) + simL(L_CF1, L_CF1',1)

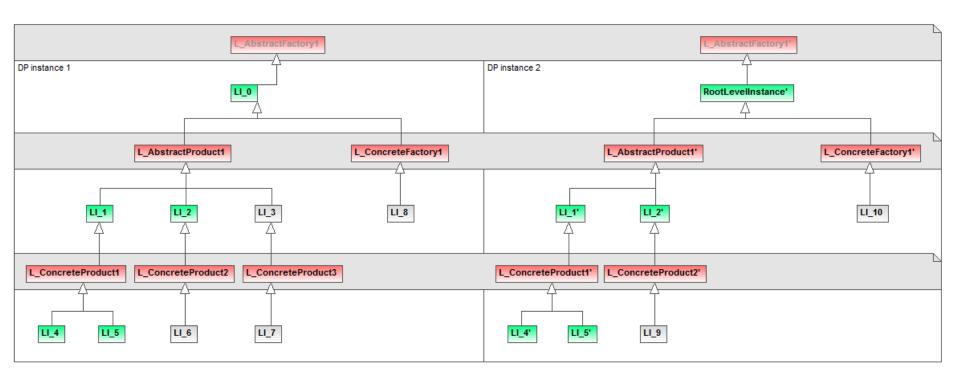
simL(L_AP1,L_AP10, 1) = (simLl(LI_1, LI_1') + simLl(LI_2, LI_2') + simLl(LI_3, null)) * weight_1/3 +(simL(L_CP1,L_CP1', 2) + simL(L_CP2,L_CP2', 2) + simL(L_CP3, null, 2))

simL(L_CP1,L_CP1', 2) = (simLl(LI_4, LI_4') + simLl(LI_5, LI_5')) * weight_2/2 = (1+1) * 0.2 / 2 = 0.2

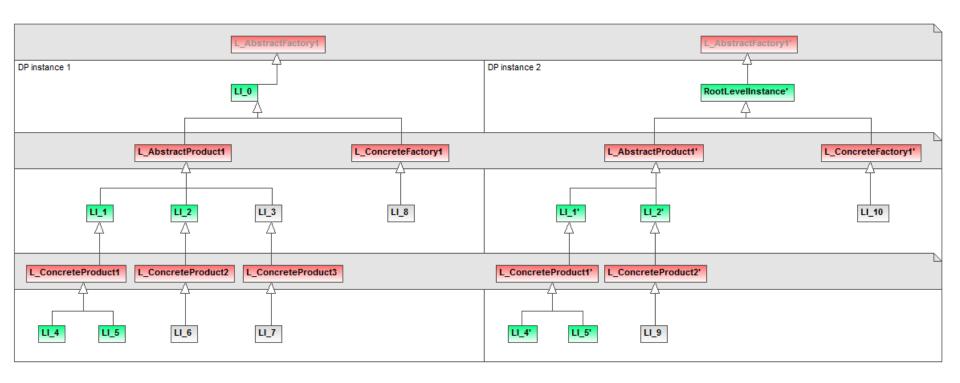
simL(L_CP2,L_CP2', 2) = (simLl(LI_6, LI_9)) * weight_2/1 = 0 * 0.2 = 0

simL(L_CP3, null, 2) = 0

simL(L_CF1,L_CF1',1) = simLl(LI_8,LI_10) * weight_1/1
```



$$similarity = 1 * weight_0 + simL(L_AP1, L_AP1', 1) + simL(L_CF1, L_CF1', 1) \\ simL(L_AP1, L_AP10, 1) = (1 + 1 + 0) * 0.26 / 3 + (0.2 + 0 + 0) = 0.37 \\ simL(L_CP1, L_CP1', 2) = (simLl(Ll_4, Ll_4') + simLl(Ll_5, Ll_5')) * weight_2 / 2 = (1 + 1) * 0.2 / 2 = 0.2 \\ simL(L_CP2, L_CP2', 2) = (simLl(Ll_6, Ll_9)) * weight_2 / 1 = 0 * 0.2 = 0 \\ simL(L_CP3, null, 2) = 0 \\ simL(L_CF1, L_CF1', 1) = 0 * 0.26 / 1 = 0$$



similarity = 1 \* 0,29 + 0,37 + 0 = 0.66 => 
$$66\%$$
  
simL(L\_AP1,L\_AP10, 1) = (1+ 1 + 0) \* 0.26 / 3 +(0.2 + 0 + 0) =  $0.37$   
simL(L\_CP1,L\_CP1', 2) = (simLl(Ll\_4, Ll\_4') + simLl(Ll\_5, Ll\_5')) \* weight\_2 / 2 = (1+1) \* 0.2 / 2 =  $0.2$   
simL(L\_CP2,L\_CP2', 2) = (simLl(Ll\_6, Ll\_9)) \* weight\_2 / 1 = 0 \* 0.2 =  $0$   
simL(L\_CP3, null, 2) =  $0$