Using the metamodel matching and transformation synthesis tools

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1 Introduction

The Agile UML toolset provides techniques for deducing matchings of classes and features between metamodels. These matchings can be used to derive model transformations in UML-RSDS, ATL, ETL, QVT-O and QVT-R.

The latest version of the tools can be obtained from: https://nms.kcl.ac.uk/kevin.lano/uml2web/. These form part of the Agile UML toolset (https://projects.eclipse.org/projects/modeling.agileuml).

2 Metamodel matching

Metamodels should be loaded using the File menu option Recent (this loads the file output/mm.txt) or $Load\ metamodel$. Classes in the metamodel(s) should be marked as source, ie., with this stereotype, if they are in the source metamodel of the matching, and as target if they are in the target metamodel. Unmarked classes are assumed to be shared (in both metamodels and mapped to themselves). It is convenient to use the Import facility to separate source and target metamodels into separate files, eg:

Import:
classmm.txt

Import:

relationalmm.txt

in mm.txt.

Figure 1 shows the visual representations of the metamodels of the ATL Class2Relational transformation case from the ATL zoo (www.eclipse.org/atl/atlTransformations). The source metamodel MM_1 is Class, on the LHS, the target metamodel MM_2 is Relational, on the RHS.

In KM3 format the source metamodel is written as:

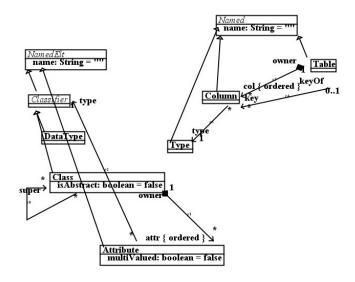


Figure 1: Class and Relational metamodels

```
package Class {
abstract class NamedElt {
attribute name : String;
}
abstract class Classifier extends NamedElt {
class DataType extends Classifier {
}
class Class extends Classifier {
reference super[*] : Class;
reference attr[*] ordered container : Attribute oppositeOf owner;
attribute isAbstract : boolean;
}
class Attribute extends NamedElt {
attribute multiValued : boolean;
reference type : Classifier;
reference owner : Class oppositeOf attr;
```

```
The target metamodel is:

package Relational {

abstract class Named {

attribute name : String;
}

class Table extends Named {

reference col[*] ordered container : Column oppositeOf owner;

reference key[*] : Column oppositeOf keyOf;
}

class Column extends Named {

reference owner : Table oppositeOf col;

reference keyOf[0-1] : Table oppositeOf key;

reference type : Type;
}

class Type extends Named {

}
}
```

Once both metamodels are loaded, select the option $Synthesise\ transformation$ from the Synthesis menu. This provides several options for matching strategies (Table 1). A matching comprises a relation cm between the classes of the two metamodels, and a relation fm between the features.

For DSS either a general matching can be used, or matchings can be restricted to be *inheritance preserving*: a subclass D of source class C is only permitted to map to a class C1 which C maps to, or to a subclass/descendant of such a C1.

For small examples such as the class/relational mapping, the NMS or DSS options are suitable. NMS uses a thesaurus (in output/thesaurus.txt) to match classes, and it is more appropriate if there are some linguistic similarities between the metamodels (such as *NamedElt* and *Named*, or *Class* and *Table*). If the metamodels have quite different terminologies then DSS is more suitable.

The tool will prompt you for the maximum navigation path to be considered on the source and target side. This means the maximum length of feature chains such as *super.isAbstract* or *key.type* (both of length 2). For cases where there is a close structural similarity between the metamodels, such as the Ant/Maven case, the choice of length 1 for source and target is usually adequate.

The results of the matching are shown in the console (Figure 2) and written to *output/forward.tl* for the forward mapping, and *output/reverse.tl* for the reverse mapping.

Measure	Definition	
Data structure	Classes possess similar data in their owned,	
$similarity \ (DSS)$	inherited or composed features [2]	
Graph structural	Class neighbourhoods in the 2 metamodels	
similarity (GSS)	have similar graph structure metrics [7]	
Graph edit	Class reachability graphs in the 2 metamodels	
$similarity \ (GES)$	have low graph edit distance [1]	
Name syntactic	Classes have names	
$similarity \ (NSS)$	with low string edit distances [6]	
Name semantic	Class names are synonymous terms	
similarity (NMS)	or in the same/linked term	
	families according to a thesaurus [3]	
Semantic context	Classes play similar semantic roles	
similarity (SCS)	in the 2 metamodels [8].	

Table 1: Syntactic and semantic similarity measures for classes

For example, the initial forward matchings derived by NMS with maximum source and target navigation length 1 look as follows:

```
\begin{array}{c} NamedElt \longmapsto Named \\ name \longmapsto name \\ Class \longmapsto Table \\ name \longmapsto name \\ attr \longmapsto col \\ Attribute \longmapsto Column \\ name \longmapsto name \\ owner \longmapsto owner \\ type \longmapsto type \\ Classifier \longmapsto Type \\ name \longmapsto name \\ DataType \longmapsto Type \\ name \longmapsto name \\ \end{array}
```

However, this matching is incomplete on both target and source sides (isAbstract, multiValued and super are unused source features, key and keyOf are unused target features). In addition, there is a potential inconsistency in that Class is mapped to Table, but Table is not a specialisation of (or equal to) the image Type of Classifier, even though Class is a specialisation of Classifier.

An interactive process following the matching derivation is used to identify such flaws and to suggest possible resolutions.

Table 2 summarises the different checks which we use.

For the case of feature mapping incompleteness in Class2Relational, because the unused source feature super is a self-association on Class, the system pro-

Issue	Correction
Class mapping	Retarget Sub
$Sub \longmapsto T$	mapping, or add
for Sub subclass of E ,	target splitting map
has T not subclass/or	$Sub \longmapsto F$
equal to F , where	
$E \longmapsto F$	
Two directions of	Modify one
bidirectional association r	feature mapping
not mapped to mutually	to ensure
reverse target features	consistency
Source, target features	Propose modified
have different	mappings
multiplicities	
Unused target subclasses	Introduce
F1 of F , where	condition $F1C$
$E \longmapsto F$	and mapping
	$\{F1C\}E \longmapsto F1$
Unused source	Suggest class or
or target	feature mapping
feature f	that uses f
Feature mapping	Propose concrete
$f \longmapsto r.g$	subclass $RSub$ of
with $r:R$ of abstract	R for instantiation
type/element type	of r .

Table 2: Consistency and completeness checks

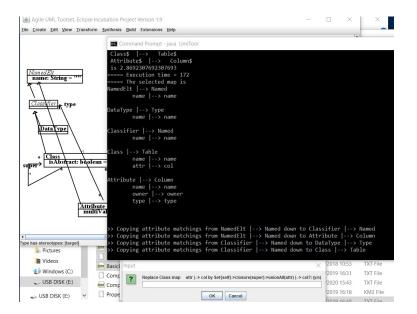


Figure 2: Initial metamodel matching

poses to replace $attr \longmapsto col$ by the mapping

$$Set\{self\} {\rightarrow} closure(\\ super) {\rightarrow} unionAll(attr) \ \longmapsto \ col$$

of all defined attributes of a class to the columns of a table, ie., all attributes of the class itself and of all its ancestors are mapped to columns of the table corresponding to the class (Figure 2).

Because of the inheritance conflict in the targets of the class mappings, the additional class mapping

$$\begin{array}{c} Class \longmapsto Type \\ name \longmapsto name \end{array}$$

is also proposed: this is a 'vertical class splitting' of Class: each Class instance in a source model is represented by both a Type instance and a Table instance in the resulting Relational model¹.

In the final stage of metamodel matching, the details of the matching and any correspondence patterns identified are listed in the console (Figure 3). Warnings are given in cases (such as multiplicity or type narrowing of target features relative to the source) where semantic problems may arise in mapping source models to target models.

The matchings can also be checked against specific source and target models, to identify detailed corrections in feature mappings. This is the option

¹The target classes must have no common MM_2 ancestor which is a type/element type of some $g \in \text{ran}(fm)$

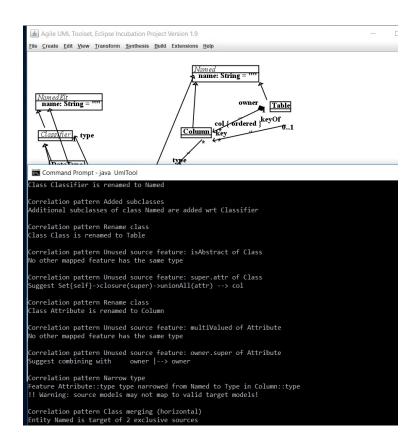


Figure 3: Metamodel matching with correspondence patterns

"Check model wrt TL" on the File menu. The models should be stored in a file output/out.txt. Load the metamodels and the TL transformation (this loads forward.tl), then run the check model option. This checks for numeric functional relationships between numeric features (linear, quadratic and exponential relations), string transformations such as case changes and prefixing/suffixing, and collection transformations such as front/tail/reverse. Eg., the $Class \mapsto Table$ matching could have example model data:

3 Generating transformation specifications

Together with the metamodel matchings, the tool produces files forward.txt and reverse.txt, which contain transformation specifications for the two directions of the matching, in QVT-R, QVT-O, UML-RSDS, ATL and ETL.

While class matchings translate to rules in the MT languages, sometimes multiple class matchings must be combined in a single rule (in ATL), or one class matching is split into several rules (in QVT-R). In ATL and ETL, composite target features in mappings $f \mapsto g.h$ must be implemented using additional lazy/called rules. In QVT-R, QVT-O and ETL rule inheritance is used to remove redundant mappings (in cases where the same feature mappings occur for a class and its superclass).

For example, the synthesised QVT-R of the above case is:

```
transformation tau(source: MM1, target: MM2)
{
  abstract top relation NamedElt2Named
  { checkonly domain source namedelt$x : NamedElt {};
    enforce domain target named$x : Named {};
  }

  abstract top relation Classifier2Type overrides NamedElt2Named
  { checkonly domain source classifier$x : Classifier {};
    enforce domain target type$x : Type {};
```

```
}
 {\tt top\ relation\ DataType2Type\ overrides\ Classifier2Type}
 { checkonly domain source datatype$x : DataType {};
   enforce domain target type$x : Type {};
}
 top relation Class2Table overrides Classifier2Type
 { checkonly domain source class$x : Class {};
   enforce domain target table$x : Table {};
}
 top relation Attribute2Column overrides NamedElt2Named
 { checkonly domain source attribute$x : Attribute {};
   enforce domain target column$x : Column {};
 top relation Class2Type overrides Classifier2Type
 { checkonly domain source classx : Class {};
   enforce domain target typex : Type {};
 top relation MapDataType2Type
 { checkonly domain source
   datatype$x : DataType { name = datatype$x_name$value };
   enforce domain target
   type$x : Type { name = datatype$x_name$value, typeFlag = "DataType" };
   when {
  DataType2Type(datatype$x,type$x) }
}
 top relation MapClass2Table
 { domain source var$0 : Attribute {};
  checkonly domain source
   class$x : Class { name = class$x_name$value }
   { Set{class$x}->closure(super)->unionAll(attr)->includes(var$0) };
   enforce domain target
   table$x : Table { col = table$x_col$x : Column { },
name = class$x_name$value };
   Class2Table(class$x,table$x) and
          Attribute2Column(var$0,table$x_col$x) }
}
 top relation MapAttribute2Column
```

```
checkonly domain source
    attribute$x : Attribute { name = attribute$x_name$value,
      owner = attribute$x_owner$x : Class { },
      type = attribute$x_type$x : Classifier { } };
    enforce domain target
    column$x : Column { name = attribute$x_name$value,
      owner = column$x_owner$x : Table { },
      type = column$x_type$x : Type { } };
    when {
    Attribute2Column(attribute$x,column$x) and
                Class2Table(attribute$x_owner$x,column$x_owner$x) and
           Classifier2Type(attribute$x_type$x,column$x_type$x) }
  }
  top relation MapClass2Type
  { checkonly domain source
    classx : Class { name = classx_name$value };
    enforce domain target
    typex : Type { name = classx_name$value };
    when {
    Class2Type(classx,typex) }
 }
}
```

An alternative mapping, aimed at generating bidirectional (bx) transformations is provided by the "Map TL to bx" option on the *File* menu. This generates QVT-R and UML-RSDS.

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