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Detecting Common Errors in Event-Driven Process Chains by Label Analysis

In this article, we discuss several classes of error patterns that can frequently be found in Event-Driven Process Chains (EPC). Instances of these patterns can be detected by using a pattern-matching approach: The model is translated into a set of Prolog rules and potential modelling problems are located by querying the Prolog fact base for certain problem patterns. In particular, this article presents patterns for problems that can be detected by analysing the labels of events and functions in EPCs. To make reasoning about the contents of the labels possible, the labels are transformed into a normal form. By taking synonyms, antonyms and negating words (no, not) into account, we locate labels that contradict each other. This leads to the detection of some classes of errors like modelling of an event and its negation occurring at the same time. Our method has been applied to 1253 EPC models in German language. We have been able to detect a large number of errors in those models that remain undetected using traditional approaches for EPC validation.

1 Introduction

According to Esswein et al. (2004), two stances of semantics of a model must be separated: First, semantics includes the rules for automated interpretation of a model (for EPC models represented by rules defining how the state space of all possible executions is calculated). Second, each modelling construct represents some real-world entities which is another aspect of model semantics. In EPC models, this second aspect of semantics is expressed by the labels of functions and events.

Pfeiffer and Niehaves (2005) call the graphical arrangements of model elements and their connections among each other model element structure. The actual meaning of the modelling constructs in the application domain is called terminological structure. Pfeiffer and Niehaves stress that an essential part of the semantics of a conceptual model descents from a reference to the language of the application domain. Hence, model validation should take into account both aspects of semantics.

If we look at approaches that have been published for the validation of EPC model semantics (for example van der Aalst 1997, Rump 1999, and Mendling 2007), we realise that the majority of them focuses on the first aspect of semantics. They allow to find control-flow errors like deadlocks but abstract away from the actual meaning of the events and functions in EPCs.

Only a few authors (for example Awad et al. 2008a whose approach will be discussed in Sect. 5) tackled the problem of detecting modelling problems related to the actual meaning of function and event labels.

When reviewing real-world EPCs from various sources, we categorised some frequent error patterns that can only be detected by analysing the labels of functions and events. The aim of our work was to automate this analysis such that a modelling tool can instantly give feedback about the occurrence of a problem. This helps to improve the model quality at the time when a model is created.

Our method has been implemented into the open-source modelling tool *bflow* Toolbox. Now this tool gives feedback not only about syntactical and control-flow problems as described in Gruhn
et al. (2009), but also about contradictions found in the usage of labels.

The remainder of this paper is structured as follows: The error patterns are categorised in Sect. 2. In Sect. 3, we describe our algorithm for locating instances of those patterns in a model. The results of a validation of our approach, using 1253 EPC models from various sources, can be found in Sect. 4. In Sect. 5, we discuss the results and compare our approach with previously published methods.

2 Error Patterns

When analysing EPC models, we identified several error patterns that remain undetected by current control-flow analysis tools. Table 1 gives an overview about the error patterns.

While we use the term ‘error patterns’ for all patterns in this article, we have to stress that not every occurrence of such a pattern necessarily indicates an actual error in the model. The severity of the occurrence of the patterns is discussed in Tab. 1 and in the subsections of this section. Some patterns indicate actual modelling errors (for example pattern B and J, see below), while other patterns (for example patterns C and D) show a possibility to improve the readability of the model.

In general, the occurrence of a pattern alerts the modeller to double-checked the part of the model where the pattern has been found.

In this section, we discuss the patterns and give examples of the occurrence of those patterns in real-world models. For the purpose of this article, these model fragments have been translated from German into English language.

2.1 Pattern A: Identical Events Before a Join

In an EPC, the semantics of events that are directly connected to a join is that the events model different conditions that can occur during the execution of the business process. For this reason, two events with the same meaning both having an arc to the same join can be an indication of a modelling problem.

In some cases, the reason for the occurrence of this pattern is that the business process has been modelled in a wrong way. For example, in the model fragment in Fig. 1 (taken from Grohmann et al. 2007) the two events with the same label ‘mark as unsuccessful’ are the result of a modelling error: Most likely, the left branch should depict the situation ‘to be marked as successful’.

However, in most cases, the multiple occurrence of the same event just means that the EPC can be simplified. An example for such a case is shown in Fig. 2. The left model does not fulfill the ‘minimality’ property as required by Weber (1997), i.e. the requirement to restrict the number of modelling elements to a minimum in order to improve readability.

When searching for identical events, we have to disregard trivial texts such as ‘OK’ or ‘processed successfully’. As there is no description what has been processed successfully, both events can depict different managerial situations. Our algorithm disregards such trivial texts. Anyway, this pattern can cause occasional ‘false alarms’ (see Fig. 3 (taken from König 2006) for an example).

Furthermore, we have to consider that events which are associated to different organisational units always have to be considered as different - even if they have the same label. For example, an event ‘audit successful’ associated to the accounting department is something else as an event ‘audit successful’ associated to the legal department.
Pattern A  Identical Events Before a Join

\[ \text{Definition: A join is preceded by two events with the same meaning.} \]
\[ \text{Consequences: In the most cases the model can be simplified by placing the event after the join.} \]

Pattern B  Identical Events After a Split

\[ \text{Definition: A split is followed by two events with the same meaning.} \]
\[ \text{Consequences: Such a construct contradicts the semantical meaning of events which should show different postconditions that can occur after the split.} \]

Pattern C  Identical Start Events Precede Same Join

\[ \text{Definition: There are two start events } S_1 \text{ and } S_2 \text{ with the same meaning. There is a path } p_1 \text{ from } S_1 \text{ to a join } J \text{ and a path from } S_2 \text{ to the same join } J \text{ such that the only common node of } p_1 \text{ and } p_2 \text{ is } J. \]
\[ \text{Consequences: The modeller should double-check whether the model can be simplified.} \]

Pattern D  Identical End Events Follow Same Split

\[ \text{Definition: There are two end events } E_1 \text{ and } E_2 \text{ with the same meaning. There is a path } p_1 \text{ from a split } S \text{ to } E_1 \text{ and a path from the same split } S \text{ to } E_2 \text{ such that the only common node of } p_1 \text{ and } p_2 \text{ is } S. \]
\[ \text{Consequences: The modeller should consider closing the control-flow block started by } S \text{ by adding a corresponding join node in order to make the model easier to read.} \]

Pattern E  Contradicting Events at an AND/OR-Connector

\[ \text{Definition: An AND- or OR-connector has two events which contradict each other as direct successor or predecessor. This is a logical error in the model.} \]
\[ \text{Consequences: The modeller should consider to replace the connector by an XOR-connector.} \]

Pattern F  Contradicting Start Events Precede Same AND/OR-Join

\[ \text{Definition: There are two start events } E_1 \text{ and } E_2 \text{ which contradict each other. There is a path } p_1 \text{ from } S_1 \text{ to a join } J \text{ of type AND or OR. There is also a path from } S_2 \text{ to the same join } J \text{ such that the only common node of } p_1 \text{ and } p_2 \text{ is } J. \]
\[ \text{Consequences: This pattern can indicate a logical error in the model.} \]

Pattern G  Contradicting End Events Follow Same AND/OR-Split

\[ \text{Definition: There are two end events } E_1 \text{ and } E_2 \text{ which contradict each other. There is a path } p_1 \text{ from a split } S \text{ of type AND or OR to } E_1 \text{. There is also a path from the same split } S \text{ to } E_2 \text{ such that the only common node of } p_1 \text{ and } p_2 \text{ is } S. \]
\[ \text{Consequences: This pattern can indicate a logical error in the model.} \]

Pattern H  XOR Connector Preceded or Followed by Events \( A \), \( \neg A \) and \( B \)

\[ \text{Definition: An XOR-connector is directly followed or preceded by at least three events. For two of these events } (A \text{ and } \neg A) \text{ it holds that one is the negation of the other one.} \]
\[ \text{Consequences: The purpose of a third event is questionable.} \]

Pattern I  Comparison Between Two Values Does not Include the Case “Equality”

\[ \text{Definition: Events after a split describe a comparision of values like } "x < y" / '"x > y" or changes of a value like "value x has increased" / "value x has decreased" \]
\[ \text{Consequences: It could be the case that the modeller has forgotten the case "Equality" ("x = y") or "Constancy" (value x remained unchanged).} \]

Pattern J  AND- or OR-Split After a yes/no-Question

\[ \text{Definition: A question that can be answered by either "yes" or "no" is followed by an AND- or OR-split.} \]
\[ \text{Consequences: The modeller should consider to replace the split by an XOR-split.} \]
2.2 Pattern B: Identical Events After a Split

In an EPC, events that follow a split are used to model the different postconditions that can occur after the processing of the split. There is no good reason for having modelled the same event twice after a split. If such a situation is found anyway in a model, this indicates an error that should be corrected. An example model we have identified with this problem is shown in Fig. 4 (which is a model fragment from the EPC Financial Accounting-Consolidation-Preparations for Consolidation from the SAP R/3 reference model).

2.3 Pattern C/D: Identical Start Events Precede Same Join/Identical End Events Follow Same Split

These patterns are similar to pattern A and pattern B: The purpose of start events and end events in an EPC is to model different situations that can occur before starting/after finishing the execution of a business process. If the same start event/end event is modelled multiple times before the same join/after the same split, this often means that the EPC can be modelled in a more structured way as shown in Fig. 5.

2.4 Pattern E: Contradicting Events at an AND/OR-Connector

If an AND- or OR-split has an arc both to an event $E_1$ as to another event $E_2$, the semantics of the split means that both events can take place at the same point of time. If $E_1$ and $E_2$ contradict each other, we can assume that there is an error.
in the model. Analogously, we can assume an error being in an EPC where contradicting events $E_1$ and $E_2$ both have an arc to the same AND- or OR-join.

Such modelling errors are quite common for novice modellers. In particular, they may be misled by a natural-language specification such as ‘After the processing of $X$, $E_1$ or $E_2$ can occur’. While described by the word ‘or’ in natural language, the correct connector to use in an EPC diagram is the XOR connector. An example of such a modelling error (found in a diploma thesis) is shown in Fig. 6.

2.5 Pattern F/G: Contradicting Start Events Precede Same AND/OR-Join/Contradicting End Events Follow Same AND/OR-Split

These patterns refer to models where an event $E_1$ and its negation $E_2 = \neg E_1$ are modelled such that they can occur together either as start events or as end events. Such a construction seems to be erroneous. As we did not find a remarkable number of instances of these patterns in our model repository, we do not discuss them in detail. The formal definition of the patterns can be found in Tab. 1.

2.6 Pattern H: XOR Connector Preceded or Followed by Events $A$, $\neg A$ and $B$

In this pattern, an XOR-connector is directly followed or preceded by an event $A$, its negation $\neg A$ and at least one additional event $B$. As $A$ is the negation of $\neg A$, either $A$ or $\neg A$ should become true. For this reason, the purpose of a third event is questionable (*Tertium non datur*).

A modelling construct like the one shown in Fig. 7 (taken from Giesa and Kopfer 2000), will be at least difficult to understand\(^1\): Obviously, one of

\(^1\)We are unable to judge about the correctness of the modelling fragment, because we are not familiar with the modelled domain.
the leftmost events has to occur so that there seems to be no good reason for having the other events in the model.

Interestingly, this pattern gave rise to some false alarms in cases like the one shown in Fig. 8. In this model, the statement ‘CR is available and (in)complete’ has been abbreviated by ‘CR is (in)complete’. Although this is easy to understand by a human reader, such a construct caused our algorithm to report a possible modelling problem.

While it is in fact impossible to prevent false alarms in cases like the one shown in Fig. 8, we can prevent false alarms in another case that can frequently be found in EPCs: Often A, ¬A and B depict the results of a decision such as ‘yes’/’no’/’possibly’. For this reason, our algorithm does not signal the occurrence of this pattern if the label of event B contains restricting words such as ‘partially’, ‘possibly’, ‘conditionally’, etc.

2.7  Pattern I: Comparison Between Two Values Does not Include the Case ‘Equality’

Often, the label of an an event in an EPC is based on the comparison of two values.

2.8  Pattern J: AND- or OR-Split After a yes/no-Question

Usually, a function that is labelled with a question that can be answered by either ‘yes’ or ‘no’ should be followed by an XOR-split and two events (one for the positive reply, another one for the negative reply). The same holds for functions with labels of the form ‘Test whether x or y’. Our
algorithm identifies this kind of questions that are wrongly followed by an AND- or OR-split instead of an XOR-split.

An example where our algorithm can locate an error is shown in Fig. 10. In this construct, the OR should be replaced by an XOR. Note that in this case, pattern E will be found as well, but this has not to be necessarily the case.

3 Algorithm for Detecting Error Pattern Instances

Our algorithm for detecting instances of the described patterns uses the approach published in Gruhn and Laue (2007): The information contained in an EPC model is transformed into a set of Prolog facts, and Prolog queries are used for locating errors in the model. In previous papers we have shown how this method can be used for checking syntactical correctness (Gruhn and Laue 2006), absence of control-flow errors (Gruhn and Laue 2009a) and unnecessarily complicated modelling structures (Gruhn and Laue 2009b).

The necessary steps for analysing an EPC using this Prolog-based approach are as follows:

1. The information contained in the EPC model is translated into Prolog facts like 
   \( \text{event}(i_3) \) (there is an event with an ID \( i_3 \)) or
   \( \text{elementname}(i_3, "Create Invoice") \) (the node with ID \( i_3 \) has the label ‘Create Invoice’).

2. The labels are transformed into a textual normal form such that labels with the same meaning will be transformed into the same normal form.

3. The patterns described in this article are searched by querying the Prolog system.

3.1 Step 1: Translating the EPC into Prolog Facts

The translation of the information within an EPC model into Prolog facts is done by a simple XSLT transformation. Details can be found in Gruhn and Laue (2006).

3.2 Step 2: Transforming the Labels into a Textual Normal Form

In the second step, the labels of events and functions are transformed into a textual normal form. The purpose of this normal form is that labels with the same meaning should be transformed into the same normal form. For this purpose, we take into account a list of stop words, synonyms and antonyms. In our prototypical implementation, we have included 210 synonyms and 70 antonyms of German terms that can frequently be found in business process models.

For the purpose of populating our catalogue of synonyms and antonyms, we counted the frequency of words in 1154 EPC models. These models contained 66,088 words (among them 7,599 different ones). Among the most frequently found words we grouped terms with the same meaning into classes of synonyms.

For generating the normal form of a label, our algorithm repeatedly substitutes substrings of the label such that terms that belong to the same class of synonyms will be replaced by the same string that symbolises the normal form for this class. For example, in the labels ‘The claim is allowed’ and ‘Claim has been accepted’, at first stop words such as the, is and has been will be replaced by the empty string. Afterwards, the terms ‘allowed’ and ‘accepted’ (which both belong to the
same class of synonyms) will be replaced by the same normal form ‘NF_ACCEPTED’. As the result, both labels will be transformed into the same normal form ‘claim NF_ACCEPTED’.

In the same way, we deal with terms that are included into the list of antonyms. In this case, we also set a flag that signals the fact that the normal form expresses the opposite of the original label. This way, the label ‘claim dismissed’ will also be transformed into the normal form ‘claim NF_ACCEPTED’. The mentioned flag tells us that the label expresses the negation of the normal form.

Labels that have the form $x \diamond y$ where $\diamond \in \{<, >, =, \leq, \geq\}$ and labels such as ‘x has increased’, ‘x has decreased’, ‘x has been incremented’, ‘x has been decremented’, ‘x remained constant’, etc. are processed in a special way. Instead of describing the details here, we give a few examples. Our algorithm realises that

- ‘$x > 1000$’ and ‘1000 is greater than $x$’ contradicts to ‘$x < 1000$’ or ‘$x$ is equal to 1000’.
- ‘x has increased’ contradicts to ‘x remained unchanged’ and ‘x has decreased’.
- The propositions ‘x is smaller than y’ and ‘x is greater than y’ describe two of the three possible cases ‘smaller than/greater than/equal to’.

### 3.3 Step 3: Querying the Prolog System

For identifying the patterns, we use Prolog rules. An example for such a rule which queries for pattern A looks as shown in listing 1.

From the description of the problem patterns in Sect. 2 we see that we need Prolog rules for reasoning about the following situations:

1. **An event is a direct predecessor or a direct successor of a connector:**
   This fact is represented by the existence of an arc in the EPC. In the Prolog representation of the EPC, this is represented by a fact `arc(x,y)`.

2. **Two events have the same meaning:**
   As discussed in the explanation to step 2, this will cause both events to have the same normal form.

3. **Two labels A and B contradict each other:**
   This is the case if A and B have been transformed into the same normal form, but a pair of words that is included in the list of antonyms has been used in the process of replacements. This would be signalised by setting the corresponding flag to true.

Another case for labels A and B contradicting each other is when A results from B by inserting a negating substring (‘non-’ or ‘not’).

Finally, label A contradicts label B if they represent two of the three cases ‘smaller than/greater than/equal to’ or ‘increased/decreased/unchanged’.
4. A label expresses the fact that an event occurs only partially or only under certain circumstances:
   This situation is assumed if the label contains a word that restrict its proposition. Such words are ‘possibly’, ‘partially’, ‘potentially’, etc.

5. The label is a question that can be answered by either ‘yes’ or ‘no’:
   This situation is assumed if the label contains the substring ‘whether’ or if it ends with a question mark but does not start with an interrogative pronoun. This way, the question ‘Has the claimant the authority to withdraw the claim?’ is regarded as a yes/no-question while ‘Which products should be offered’ is not.

4 Validation

We have searched for the pattern described above in a repository of 1253 EPC models in German language that we have collected from various sources:
- 591 models from the SAP R/3 reference model, a widespread business reference model
- 127 models from textbooks
- 48 models from PhD theses
- 70 models from published scientific papers
- 252 models from bachelor and diploma theses and term papers
- 84 models from real-world projects
- 22 models from university lecture notes
- 13 models from technical manuals

The remaining 37 models were EPCs published on the Internet for which we were unable to assign them into one of the above categories. Among the models in our repository, there is a great variation in size of the models, purpose of modelling, business domain and experience of the modelers. For this reason, we think that the models represent a reasonable good sample of real-world models.

Because of the large number of models and the great variation in the modeled domain, we have been unable to perform a manual analysis of all models. For this reason, we are unable to decide whether there were any instances of the discussed error patterns that have not been found by our algorithm, i.e. we cannot quantify the Recall value for our algorithm.

However, we have analysed the Precision value as follows: All problems that have been signalised by our tool have been checked manually in order to decide whether the model in fact should be improved with respect to this problem pattern. In some cases, such a decision was impossible for us due to lack of domain knowledge (in particular for pattern H). If there were any doubts, we categorised a reported problem as ‘(possible) false alarm’.

This way, we got the following numbers of legitimate problem alerts and false alarms:

Altogether, our tool has located 129 cases in 99 EPCs where indeed an improvement of the model seemed to be advisable. On the other hand, there were 46 false alarms in 44 models. The large percentage of false alarms for the patterns C, D, F, G and H gives raise to the assumption that it might be less useful to look for these patterns. For the patterns A, B, I and J, almost all instances found of these patterns pointed to a possibility to correct or improve the model. It would be useful for a modeller to run these checks at modelling time in order to avoid the occurrence of the problem patterns.

There is a remarkable coherence between the origin of the EPCs and the classes of problems found in them. For instance, the error pattern E is very typical for novice modellers: Instead of an XOR-split, they use an OR-split which would be suggested by the specification of a business process in natural language. While error pattern E was found very often in students’ papers (and as well in models from a modelling project in the media domain), there was no instance of this pattern in the SAP R/3 reference model. This shows
that the detection of problems by our approach would be more useful for beginners.

5 Comparison of our Approach with Related Work

The results in the previous chapter show that by analysing EPC labels, we are able to identify a remarkable number of modelling problems.

While the method has been implemented for German-language models, we see no reason why it should be difficult to adapt it to models in other languages as well. The main part of the work to achieve this adaption would be to build the synonym/antonym catalogue.

Our approach does not force the modeller to restrict the language used for the labels. Labels of events and functions can contain any phrase in natural language which is common practice in EPC modelling. By restricting the language for describing events and functions, it should be possible to improve the effectiveness of the algorithm considerably.

Such a consolidated usage of words that can occur in labels can be achieved by using the tool Semtalk (Fillies and Weichhardt 2005) which makes use of ontologies. By using such an ontology-based concept, it is even possible to assess the correctness of a business process model with regard to the application domain. Fillies and Weichhardt (2003) give an example where two business process models ‘incoming order’ and ‘order processing’ are analysed. An example for a business rule that can be checked for those models is that ‘no order should be processed unless it is confirmed’. A very similar example is used in Thomas and Fellmann (2007) where EPC models are used together with ontologies of the business domain.

We are convinced that combining business process modelling and ontologies can lead to very powerful tools. Such tools can be used for checking consistency between models, for validating the compliance with business rules and for querying model repositories. Current work on this topic is described in Weber et al. (2008) and Governatori et al. (2008).

However, we are also convinced that such ontology-based methods will become very difficult to implement in practical modelling projects. The reason is that (at least during the introduction stage), the cost and complexity of such a method will be much higher than with traditional modelling. The approaches described in the literature demand to build an ontology of the business domain in addition to the usual modelling task (see for example the description on how to construct a the so-called semantical model in Filipowska et al. 2008).

<table>
<thead>
<tr>
<th>Pattern</th>
<th>pattern instances found</th>
<th>(possible) false alarms</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pattern A</td>
<td>70 in 54 models</td>
<td>1 in 1 model</td>
</tr>
<tr>
<td>Pattern B</td>
<td>1 in 1 model</td>
<td>none</td>
</tr>
<tr>
<td>Pattern C</td>
<td>22 in 21 models</td>
<td>12 in 11 models</td>
</tr>
<tr>
<td>Pattern D</td>
<td>21 in 21 models</td>
<td>17 in 17 models</td>
</tr>
<tr>
<td>Pattern E</td>
<td>31 in 19 models</td>
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</tr>
<tr>
<td>Pattern F</td>
<td>2 in 2 models</td>
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</tr>
<tr>
<td>Pattern G</td>
<td>2 in 2 models</td>
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</tr>
<tr>
<td>Pattern H</td>
<td>21 in 18 models</td>
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</tr>
<tr>
<td>Pattern J</td>
<td>2 in 2 models</td>
<td>none</td>
</tr>
</tbody>
</table>

Table 2: Patterns found in the 1253 EPCs of our repository
Our method works without the need to construct an ontology of the application domain. We only have to fill the lists of synonyms and antonyms which can be regarded as a very rudimental part of a domain ontology. In its current form, the catalogue of synonyms and antonyms is yet far from being complete. For this reason we cannot claim to detect all occurrences of the problem patterns described in this article.

While this incompleteness is clearly a disadvantage, we see an advantage in the fact that our method is easy to implement and easy to use. Our validation has shown that even a rather small catalogue of synonyms and antonyms is sufficient for finding a remarkable number of modelling problems. Other than it is the case for ontology-based approaches, our approach delivers information about the modelling problems without adding to the complexity of the modelling method.

The pattern-based analysis of EPC models has already been integrated into the open-source modelling tool bflow* Toolbox (see Fig. 11). The analysis can be started ‘at the push of a button’. This adds more power to the business process modelling with immediate feedback in the bflow* Toolbox as described in Gruhn et al. (2009).

Awad et al. (2008b) use information retrieval techniques for defining a measure of similarity between labels within a BPMN diagram. This approach uses the general-purpose lexical database WordNet (Fellbaum 1998), which makes it possible to compare the similarity of labels without having to restrict the use of the language. Combining Awad et al. (2008b) with previous work on BPMN-Q (Awad et al. 2008a) makes it possible to validate business rules like ‘An account must not be opened until certain checks have been completed’.

A similar approach is described in Koschmider and Oberweis (2007). However, the main purpose of this work is not the validation of business process models but to detect model variants. An important difference between the method described in our article and the approaches given by Awad et al. (2008b) and Koschmider and Oberweis (2007) is that we are able to locate a substantial number of modelling problems with a very small catalogue of synonyms and antonyms that can be used very efficiently. It will be an interesting subject for further studies how the results of our pattern search can be improved by using more advanced algorithms (as described in Awad et al. 2008b, Koschmider and Oberweis 2007 or Gervasi and Zowghi 2005) for detecting labels with the same meaning or labels that contradict each other.

It is another interesting research field to validate whether naming conventions for events and functions have been followed. Bögl et al. (2008) show how lexical databases and semantical patterns for labels can be used for this purpose. Other authors have suggested somewhat different solutions for the same problem: Both Leopold et al. (2009) and Becker et al. (2009) make use of linguistic syntax parsing methods for assuring the compliance to naming conventions.
An approach to measure the label quality of EPC models has been suggested by Friedrich (2009). In particular, this method can find labels with a high chance of ambiguity.

It is our intention to research more patterns that can signalise the possibility to improve a model. Researchers and practitioners are invited to use and to improve the tests in the bflow* Toolbox. The most current release can be downloaded from www.bflow.org.

References


The Prolog rules that are responsible for the tests described in this article can be found in the plugin org.bflow.toolbox.mitamm.prolog. The list of synonyms and antonyms we have used for the purpose of our validation can be found at the same place.
Detecting Common Errors in Event-Driven Process Chains by Label Analysis


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