

Semi-automated Business Process Model Matching and Merging Considering Advanced Modeling Constraints

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Abstract: Model merging helps to manage the combination and coevolution of business processes. Combining models (semi-)automatically can be a helpful technique in manifold areas and has been investigated since decades by the scientific community. The rising complexity of (business-) processes in shifting environments demands for a more differentiated view on model matching and merging techniques. In this domain, we identified the problem of considering additional constraints in the matching and merging process and suggest an approach by adapting state of the art solutions correspondingly. In addition, we state necessary reduction rules and discuss their suitability. Moreover we provide a prototypical implementation of a matching and merging tool, which allows further investigations of the approach concerning quality, usefulness and efficiency.

1 INTRODUCTION

Combining various kinds of models automatically is still a prominent research field. Depending on the problem domain, manifold values might occur. (La Rosa et al., 2013) declare that “the purpose of merged models is to allow analysts to view the commonalities and differences between multiple variants of a business process and to manage their coevolution and convergence”. (Nejati et al., 2007) conclude that model matching and merging play a central role in supporting distribution and coordination of modeling tasks. Moreover, “merged models are intended for analysts who want to create a model that subsumes a collection of process models with the aim of replacing variants with the merged model” (La Rosa et al., 2013).

But whereas nowadays, advanced modeling languages are extensive tools, offering manifold operators and elements to model complex systems, model merging does not cover every modeling aspect yet. In conjunction with shifting business environments, changing firm structures or IT-system integration scenarios, business processes might be modified, complemented or optimized. Coping with a continuous change, improvement and re-engineering of business processes with the help of modeling techniques has been investigated by the scientific community since decades. One example that clarifies the problem is

project Mobility Broker (Beutel et al., 2015), (Kluth et al., 2015), where similar service processes have to be integrated and consolidated.

Pre-eminently, we want to investigate the suitability of semi-automated, graph based matching and merging of business process models in complex scenarios. As a consequence, we consider additional modeling constraints. Therefore, we review and extend existing solutions. On hand of a specific use case, we analyze how additional constraints affect the merging results.

For example, incorporating different roles, represented by swimlanes in the Business Process Model and Notation (BPMN) language into the merging process widens the scope. This effort goes beyond involving interfaces into model merging, but incorporates roles, their (sub-)processes and their messages flow types. In fact, this affects the whole merging process fundamentally.

Within this work, we present an approach for model merging, allowing to incorporate multiple advanced modeling constraints. In addition, we show an initial prototypical tool and discuss its suitability.

The remainder of this work is structured as follows. Section 2 describes the relevant theoretical basis. Afterwards, Section 3 describes the merging approach considering additional modeling constraints. Section 4 presents the technical realization in form

of a merging tool. Section 5 shows the outcome of our solution based on an exemplary use case. Finally, Section 6 reflects and concludes the work.

2 FOUNDATIONS

This section describes the fundamentals concerning business modeling, matching and merging.

2.1 Business Process Modeling Languages

Process modeling is supposed to be an instrument for coping with the complexity of process planning, with importance for many purposes besides the development of software (Becker et al., 2002). Business process models help to improve and to re-engineer processes. Vital for this purpose are models which help to identify process weaknesses and allow an automated comparison of new scenarios (Becker et al., 2003).

BPMN is a modeling language, using events and activities to visualize (business-) processes. It describes the logic of process flows with the help of gateway operators such as AND, OR and XOR¹. Sequence flow, message flow and association elements are used to describe the connection between certain process objects (Kocian, 2011). In addition, swimlanes represent different roles, which allow to visualize role-specific processes and their interdependencies between each other in parallel.

Another prominent way of modeling processes is Event-driven Process Chain (EPC). An EPC diagram is a flowchart based diagram, used for resource planning and identifying possible improvements of a business process (Object Management Group, 2016). There are various other modeling techniques as well. For the following explanations, we refer to the BPMN 2.0 standard.

2.2 Model Matching and Merging

In general, model matching and merging had already been investigated in various modeling areas. For example, (Brunet et al., 2006) show merging approaches of entity relationship diagrams or state machines. Moreover, (Melnik, 2004) investigates the matching and merging of conceptual database schemata, (Mandelin et al., 2006) proposes a technique for matching system architecture diagrams, using machine learning and (Nejati et al., 2007) provide an approach

¹Exclusive or: True only, when one input is true and the other is false.

for matching and merging statecharts specifications. In general, (Brunet et al., 2006) describe model merging as an exploratory process, in which the goal is to discover the exact nature of relationship between models, as much as to combine them. Model merging is facilitated by a number of related operations on models, such as comparing, checking their consistency and finding matches between them.

For our descriptions, we distinguish the operators of matching and merging and refer to a business process graph setting. Therefore, we use the following definitions:

- A **business process graph** G is a set of pairs of process model notes – each pair denoting a direct edge. A node v of G is a tuple $(id_G(v), \lambda_G(v), \tau_G(v))$ consisting of a unique identifier $id_G(v)$ within G , a label $\lambda_G(v)$ and a type $\tau_G(v)$ (La Rosa et al., 2013).
- Given process graphs G_1, G_2 and nodes $v \in G_1, w \in G_2$, process **model matching** is defined as finding an injective mapping $f(v, w) : v \rightarrow w$, where $f(v, w)$ is a scoring function and its value is within the interval $[0, 1] \subset \mathbb{R}$. This operation is used to find commonalities between models (Brunet et al., 2006).
- Given n process graphs, process **model merging** describes the reduction to k graphs ($k < n$), such that the behavior and correctness is preserved.

In the specific field of interest of business processes some valuable research has been done already. (Kuester et al., 2008) suggest a process merging tool that detects and resolves changes between process models, which has limited applicability concerning recent modeling languages. In addition, they focus more on the user-friendly design of their solution. (La Rosa et al., 2013) provide a merging algorithm for business process graphs. They explicitly abstract from any specific notation, which increases the applicability of the approach.

We base our studies on these findings and adapt the algorithm correspondingly to our specific area of interest.

3 APPROACH

We convert BPMN models to graphs. The approach for merging the graphs consists of three main steps (Figure 1):

1. *Matching*: Discovering similar parts of the graphs and building a similarity map, which is passed on to the next step.

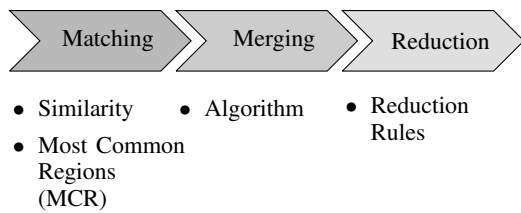


Figure 1: Merging process steps.

2. *Merging*: Traversal of the two graphs and based on the similarity map, addition of the nodes to the resulting merged graph.
3. *Reduction*: A clean-up of the merged graph since there might be unnecessary/obsolete nodes, which can be left out.

3.1 Matching

A business process is modeled as a directed graph $G = (V, E)$, where V is the set of nodes and E is the set of edges such that for $v, w \in V$ there is an edge $e = (v, w) \in E$. A node can represent, for example, a task or an event. An edge can represent, for example, a message or sequence flow. Both nodes and edges may hold any number of additional *properties*, which are modeled as key-value pairs (Figure 2).

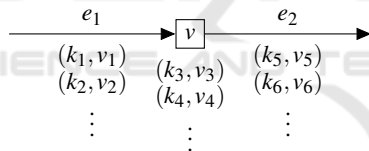


Figure 2: A graph node with incoming and outgoing edges.

Example. Consider a *check availability* process in a sales company of goods. This process takes the product identifier as input and returns the number of items that are in stock. This can be modeled as depicted in Figure 3. In this example, the process becomes the node which contains the property (title, Check Availability). The incoming edge contains the property (input, product id), the outgoing edge contains the property (output, number of items).

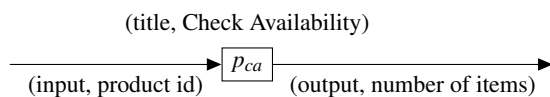


Figure 3: Modeling example for a check availability process

Definition 1 (Element). An element is the building block of a graph and can be either a node or an edge.

Definition 2 (Element similarity). Two elements are similar, if

1. they hold the same property keys,
2. for every property key, the corresponding property values are similar according to a similarity function.

Definition 3 (Matching). Given process graphs G_1, G_2 and nodes $v \in V(G_1), w \in V(G_2)$; we claim v and w are matching, if

- v and w are similar,
- there is a bijection between the sets of the incoming edge(s) of v and w , where the corresponding elements are similar,
- there is a bijection between the sets of the outgoing edge(s) of v and w , where the corresponding elements are similar.

Considering the example in Figure 3, if a different business process graph has a check availability process, where the node contains the property (title, Check Availability), incoming edge the property (input, product id) and the outgoing edge the property (output, number of items), then we say that both check availability processes are *matching*.

Similarity Function

To identify the similarity between the values of two properties with the same key, a similarity function is required. Since the property keys might have varying importance depending on the application domain, we assign a similarity threshold to each key, which is used to check whether the result of the similarity function is acceptable or not. We conclude that the two properties are similar, only if the result is above this particular threshold. This requirement affects the modeling of properties as key-value pairs, since every key k becomes a pair $k = (k_{id}, k_t)$, such that k_{id} uniquely identifies the property and k_t denotes the threshold value for the similarity score.

Moreover, each key might dictate a different kind of similarity function. For example, a key with a textual value might require an *exact* or a *partial* match. Therefore, we extend the key representation to express this information as well, and k becomes a tuple $k = (k_{id}, k_t, k_{sim})$, such that k_{sim} denotes the similarity function to be used for this key.

For properties with textual value that require an exact match, we look at the equality of both values. For example, in the context of BPMN, the nodes contain a *type* parameter, which is assigned to one of {event, gateway, task, role}, and the edges contain a *type* parameter, which is assigned to one of

{message_flow, sequence_flow}. Both of these *type* parameters require an exact match.

For properties with textual value that allow partial match, the similarity function is based on the *edit distance*. The standard edit distance takes two single words as input and returns a score. But in our case, a textual value might be a phrase which consists of multiple words. In order to avoid lower similarity score due to word formations, we changed the standard approach such that word-wise similarity check is performed: The first and second phrases are tokenized into m and n number of words, respectively. We measure the edit distances between a word of the first phrase and all words in the second phrase; and remember the highest similarity score for this word. We repeat this process for all the words in the first phrase. Then, we sum over all the highest similarity scores, and average the sum over $\max(m, n)$. The result gives us the similarity score between the two phrases.

Advanced Similarity Measure: SimRank

Definition 3 takes only the *local* environment of a node into account. It is not concerned about any predecessors or where the nodes are located in the graph. In some advanced cases this is not sufficient, since we might need the *global* context of the nodes, i. e. if they are referenced by similar nodes. Here, we employ an adaptation of the SimRank algorithm (Jeh and Widom, 2002). The first extension concerns the input: Instead of searching for similarity within one graph, we extend it to pairwise comparison $SimRank(G_1, G_2)$. The second extension is the condition for the initial SimRank score: With two nodes v and w , the original paper (Jeh and Widom, 2002) considers the equality of nodes ($v = w$). Instead of that, we employ $sim(v, w) > t$ as depicted in Equation (1), where $v \in G_1$, $w \in G_2$, sim is the similarity function, and t is the similarity threshold.

$$s_0(v, w) = \begin{cases} 1 & \text{Reduction if } sim(v, w) > t \\ 0 & \text{Reduction else} \end{cases} \quad (1)$$

The starting point of SimRank is the local environment as described in Definition 3. But it goes beyond that by looking at predecessor nodes and incorporating their similarity scores. The result is a similarity matrix between the nodes of G_1 and G_2 , where each entry denotes the final similarity score. Within a column the entry with the highest score indicates that the corresponding row is the most similar.

Matching Output and Maximum Common Regions

Both similarity approaches are used in the comparison of BPMN models: for connected nodes who have a name property (e. g. tasks) the similarity function is applied, where as for the nodes without name property (e. g. events and gateways) we apply SimRank. The results are combined in a map consisting of pairs of matched nodes and their respective similarity score. In order to derive the bigger picture from individual matching nodes i. e. to find out groups of connected matching nodes, the similarity map is used to construct the set of all Maximum Common Regions (MCRs). A Maximum Common Region (MCR) is the largest connected subgraph consisting only of matching nodes (La Rosa et al., 2013). The MCRs serve as basis for applying the merging algorithm.

3.2 Merging

The merging step traverses the two graphs, which represent two BPMN 2.0 models, and constructs another graph, taking the similarity/match information determined from the previous step into account. The resulting graph represents a BPMN 2.0 model containing swimlanes and messages.

Merging of Roles

As mentioned before, swimlanes in BPMN represent roles which have identifiers and names. If two swimlanes do completely match, i. e. their processes are exactly the same (i. e. all the nodes and edges exhibit the same characteristics), then the swimlanes are merged. Otherwise, both swimlanes are added in the merged graph.

Merging of Nodes

The merging process is done by iterating every node v_1 from G_1 and checking whether it belongs to a MCR. Next, each successor s_1 of v_1 is also checked if it is a part of a MCR. This produces four outcomes:

- Both v_1 and s_1 are part of a MCR. Since they are connected, they are part of the same MCR. As the behavior is the same for both graphs, connect them directly in the result graph.
- Neither v_1 nor s_1 are part of a MCR. This particular behavior is specific only for G_1 . Similar to the previous case, we connect them directly in the result graph.

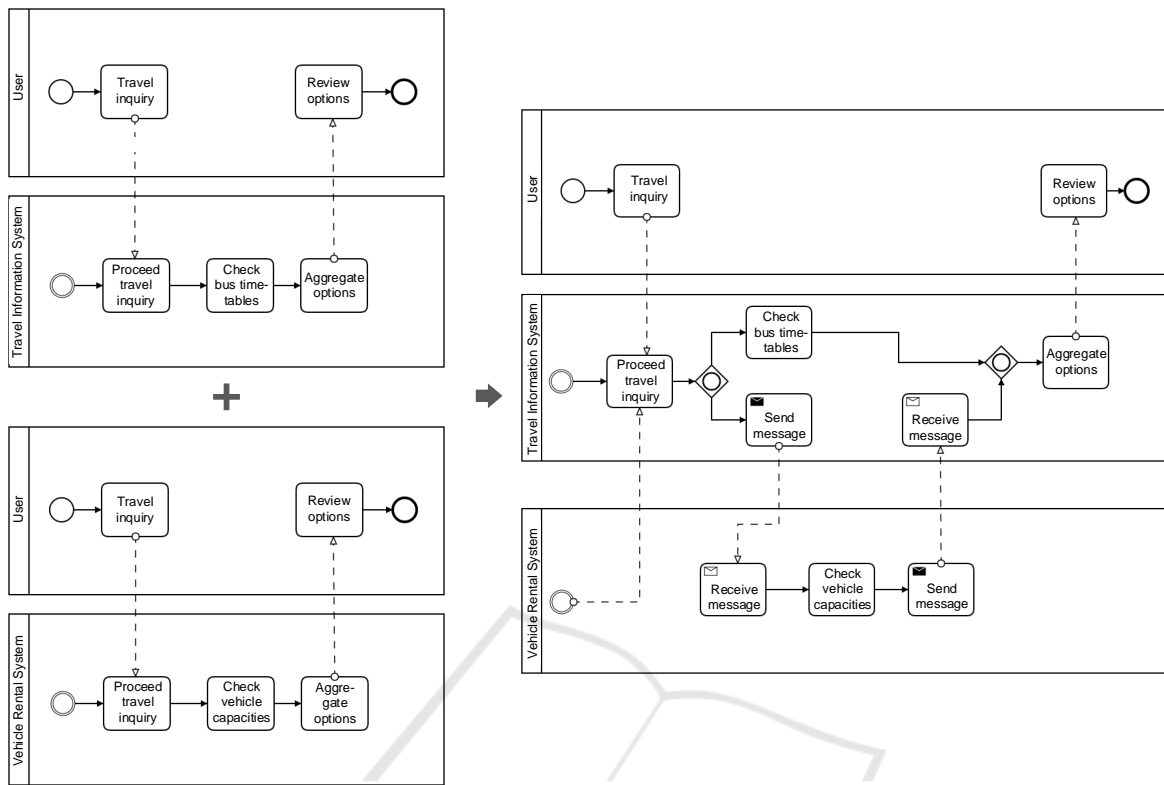


Figure 4: Simplified use case.

- v_1 belongs to a MCR, but s_1 does not. In this case, we add a gateway with node v_1 as input. The output of the gateway is s_1 and the successors of v_2 , which is the mapped node of v_1 .
- v_1 does not belong to a MCR, but s_1 does. In this case, we insert a gateway: as input of we assign v_1 and the predecessors of the mapped node of v_1 , where as the gateway output we assign node v_1 .

The type of inserted gateways depends on the predefined choice for merging. If the inclusive configuration is selected, then the inserted gateways are of inclusive type, otherwise they are of exclusive type.

Next, we iterate every node v_2 from G_2 that is not part of any MCR (the leftover nodes). We check the successors for any missing edges in the final graph. If an edge between v_2 and a successor s_2 in G_2 exists, we connect v_2 and the mapped node of s_2 in the result graph as well.

If two similar textual properties have different values, then in the merged graph the textual value from the first graph is considered. For example, the property (name, Travel Inquiry) of a task node from G_1 and (name, Inquiry) of a task node from G_2 , are merged into resulting property (name, Travel Inquiry).

The merging process does not reposition any tasks

or events in other swimlane. As mentioned earlier, BPMN edges and nodes contain a *type* property. This concept is so fundamental for BPMN, that the merging strategies are specific to node types. The node-by-node inspection allows us to apply different merging behavior for different node types during the merging process. For example, merging events and its successors require retaining the events in both lanes, as shown in Figure 4.

3.3 Post-merge operations

Reduction

The end result of the merging can contain obsolete behavior when merging gateway nodes. We identified that applying reduction rules can remove the unnecessary gateways by inspecting whether the added gateways from the merge process and the original gateways have common successors. There are many possibilities for reduction since the gateway type has to be considered. We inspected a scenario given in figure 5, which eliminates an unnecessary XOR gateway.

The process of reduction becomes more challenging when gateways of different types occur. Identi-

ifying if a reduction is possible and its application is part of our future research.

Send/Receive Tasks

As final step we identify if any communication between gateways and tasks from different roles occurred after merging. Such behavior is not allowed in BPMN 2.0. Therefore, we add nodes to preserve correctness in the merged model. For each such message flow, we introduce a pair of *Send* and *Receive* tasks - activities with limited scope that explicitly define that a message should be sent and received respectively. The *Send Task* node is placed in the same swimlane as the node from which the message originates and the *Receive Task* in the same swimlane as the recipient node of the message. Instead of direct communication, the message flow between gateways and tasks from different swimlanes now happens through the message tasks.

4 IMPLEMENTATION

The approach was implemented in Java as a standalone application that takes as an input two BPMN 2.0 graphs and outputs a merged graph as result.

For creating BPMN 2.0 models, we used the Camunda Open Source platform². For the underlying graph models, we used the JGraphT³. It also provides an adapter to JGraph⁴, which offers elementary visualizations as a Java Applet.

The merging tool provides a graph model as output that serves as a reference for building the models in Camunda. The deserialization process from the graph model to BPMN model is done manually.

5 USE CASE: DIFFERENT ROLES AND FLOW TYPES

We applied the approach in a use case which incorporates different roles. In BPMN, different roles are represented as pools. As a consequence, we have to consider different flow types, represented by the edges of a node. In more detail, we have to distinguish between sequence flows and messages flows between tasks of different roles. Figure 4 visualizes the merging output. Initially, we have three different roles:

²<https://camunda.org>

³<http://jgraph.org>

⁴<https://jgraph.com>

The user, Travel Information System (TIS) and Vehicle Rental System (VRS). The role user is the same in both input processes, because of which it can be entirely merged. The tasks of the remaining roles are similar, but to prevent information loss of participating roles, a messages flow points to the merged task.

Figure 5 shows the application of a reduction rule concerning gateway operators. To reduce complexity, we decided to replace doubled gateways and preferring the more heavily weighted gateway (in this case XOR).

6 DISCUSSION

We investigated the suitability of matching and merging techniques in complex scenarios, considering advanced modeling constraints. Therefore, we presented a graph matching and merging approach especially intended for business processes using BPMN. We based our work on state-of-the-art approaches, combined them partially and provided solutions for open merging problems in the specific domain. We showed the underlying idea of considering multiple restrictions, e. g. different edge types and roles and investigated the resulting effects on the merging results regarding a specific use case. Moreover, our approach is suitable to consider additional constraints as well.

At the current state of the approach, there is a limited suitability in complex scenarios because here are additional constraints that have to be considered as well. Nevertheless, first results indicate the importance to consider advanced modeling constraints. In addition, it seems to be beneficial to have more room for manual adjustments by the user to create useful merging results. We showed that the consideration of advanced modeling constraints and operations affect the merging results fundamentally. In more detail, considering roles and their specific modeling operators, including different flow types prevents loss of relevant information and increases merging accuracy.

To reflect our work in more detail, we want to base a discussion on three requirements proposed by (La Rosa et al., 2013):

1. *Behavior-Preservation* describes that the behavior of the merged model should encompass that of the input models.

Our merging process does not change the behavior of the message flow. Furthermore, the nodes are kept within the same lanes. Additional interface nodes are added to circumvent the constraints originating from BPMN 2.0.

2. *Traceability* means that, given an element in the

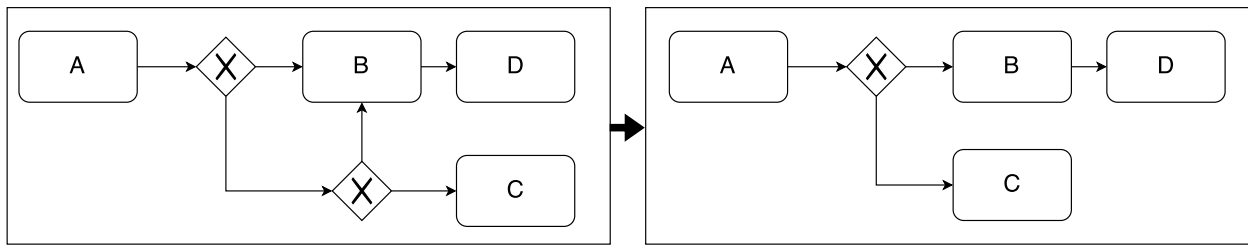


Figure 5: Applying reduction rules.

merged model, analysts should be able to trace back the source process model from which the element originates.

Our work achieves partial traceability: Nodes that are inserted (gateways and interface nodes) are assigned generated ids and behavior. However, there is not a clear distinction between nodes which are specific to one model and nodes that belong to both of the graphs. Since merged properties take the value from G_1 , it is not clear whether a node is specific to G_1 or for both the graphs. For nodes specific to G_2 , that is not the case. This is only an implementation issue which we became aware of later in the development after the need for traceability arose. It can be solved by storing an additional property in each graph element that points/references to the input model which it came from. In case the element occurs in multiple models, the property value can be a list of references.

3. *Reversibility* describes the possibility to derive the input process models from the merged model.

Our work does not focus on reversibility. After merging, the result model cannot be decomposed into the original input models. Nevertheless, since reversibility bases on traceability, once the traceability feature is implemented, it would be trivial to extend the merging tool to support reversibility.

In fact, more fine-grained and precise matching and merging approaches might have the potential to widen the area of application and to create advanced results for complex scenarios.

6.1 Limitations

In our initial realization we did not consider some modeling operations, including for example:

- The communication between pools
- Full set of BPMN 2.0 constraints
- Advanced modeling operators, e. g. nested subprocesses or specified sequence flows

In addition, we did not consider more than two input graphs.

6.2 Future Work

Primarily, this approach has to be evaluated, using different use cases from other domains. Hence, there might be some room for different optimizations and extensions, e. g. concerning the algorithm or included operators. Further evaluation efforts should focus on qualitative feedback from experts to estimate the suitability in more detail. Especially the investigation of the optimal degree of automation seems to be interesting. Moreover, measurements concerning efficiency issues might be valuable as well.

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