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Knowledge Sharing and Inconsistency Checking on Multiple Enterprise Models

by

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Abstract :

The approach of Multi-Perspective Enterprise Modelling is now more commonly accepted and used in practice than ever before. However, the concept of applying multiple modelling languages to describe the same domain may still sound frightening to many. In addition to the cost, time and complexity involved, problems such as knowledge sharing between multiple models and achieving and maintaining integrity between them are also important. We argue that Multi-Perspective Enterprise Modelling is helpful and in some situations necessary. This paper gives examples of how formal methods, such as logical languages, can provide assistance in making such an approach more appealing and transparent. We suggest that the MPM approach is valuable in representing, understanding and analysing a complex domain, but that much automated support is needed.

Keywords : Multi-Perspective Modelling, Business Modelling, BSDM, Enterprise Modelling, Process Modelling, Knowledge-Based Support Tool, Business Process Re-Engineering, I

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Knowledge Sharing and Inconsistency Checking on Multiple Enterprise Models

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Abstract The approach of *Multi-Perspective Enterprise Modelling* is now more commonly accepted and used in practice than ever before. However, the concept of applying multiple modelling languages to describe the same domain may still sound frightening to many. In addition to the cost, time and complexity involved, problems such as knowledge sharing between multiple models and achieving and maintaining integrity between them are also important. We argue that *Multi-Perspective Enterprise Modelling* is helpful and in some situations necessary. This paper gives examples of how formal methods, such as logical languages, can provide assistance in making such an approach more appealing and transparent. We suggest that the *MPM* approach is valuable in representing, understanding and analysing a complex domain, but that much automated support is needed.

Key-words Knowledge Sharing, Multi-Perspective Modelling, Business Modelling, BSDM, Enterprise Modelling, Process Modelling, Knowledge-Based Support Tool, Business Process Re-Engineering, Role Activity and Communication Diagram.

1 Introduction

Today's economy is often called a *knowledge economy* indicating that stake-holders of the right kind of knowledge may gain competitive advantage. This knowledge may be roughly divided into two types of knowledge: internal and external knowledge. Internal knowledge is the *corporate knowledge* within an organisation and external knowledge is about the economic environment the organisation operates in.

A modern enterprise is often a virtual entity which consists of many sub-organisations which are distributed across different geographical areas, each possessing different expertise and specialising in certain functions. This complicates the task of treating *corporate knowledge* as a whole and making effective use of it. Furthermore, the economy in which an organisation operates is very dynamic which requires an organisation to

react appropriately and promptly — in adapting their goals and processes. This paper focuses on the capture of *corporate knowledge* using *Multi-Perspective Modelling* (*MPM*) techniques. We call the created models *Enterprise Models*.

Multi-Perspective Modelling (MPM) techniques make use of multiple modelling languages which allow one to present and analyse organisational knowledge from different points of view, which in turn allows the knowledge to be used for different purposes. The *MPM* approach is necessary because organisational knowledge is often so complicated and heterogeneous types that normally no single modelling method can capture all of the important aspects and present them clearly and appropriately. Thus a Multi-Perspective modelling approach makes use of multiple modelling languages which compliment each other and work as a whole to describe the enterprise knowledge better.

Furthermore, the required uses of *Enterprise Models* are often diverse: businesses often need to examine different but specific aspects of the knowledge for different purposes as a part of their decision making process. This knowledge may come from the same body of knowledge, but just examining the concerned sides of it at a time. This demands a diversified but specialised presentation of the *corporate knowledge* which allows analysis to be performed on the desired aspects of the knowledge for each specific purpose.

The components of domain knowledge of an enterprise model, however, can be highly inter-dependent if they are not structured and presented appropriately. It is therefore important to have a mechanism which allows the relevant information to be gathered and presented in a clear, concise and structured way which is not overburdened with other irrelevant information. Since a multi-perspective modelling approach uses several different modelling languages, each language provides a specialised presentation to the knowledge domain which also allows insight analysis into the specific aspects of the domain.

This approach is already used in research and practice: Common KADS methodology [14] embodies several modelling languages to help understand and capture domain knowledge and to help the design of knowledge based systems; Booch, Rumbaugh and Jacobson [1] fully embrace this approach and have offered a suite of inter-supportive modelling notations as part of the *Unified Modelling Language*, for gathering requirements and development of software systems; Frank[6] advocates this approach based on which multiple notations have been used and a multi-perspective knowledge management system (MEMO) was developed; in Zachman's[16] Framework for Enterprise Architecture, it suggests using a variety of modelling languages to capture and describe the different aspects of a domain. The importance and benefits of using multiple and complimentary modelling languages to represent a complex knowledge body is wellrecognised and adopted more frequently than before.

During the Air Operation Enterprise Modelling project[10], a Multi-Perspective Modelling approach was taken. The domain of (military) Air Operations is complex. A main source of knowledge regarding Air Operations was provided to the initiative in an *IDEF0* model¹. It consists of 290 functions, 307 inputs (data types which provide input information for the functions), 294 outputs (data types or results which are pro-

¹The Air Operations IDEF model was developed by Larry Tonneson, Zel Technologies, LLC, USA.

duced by the functions), and 45 controls (data types which provide principles, guidance and information for executing the functions). In addition, documents written in natural language, informal diagrams and tables are provided — describing different parts of the air operations. This above information is aided with fact-to-fact explanation and email correspondence with domain experts. Several aspects are considered: the infrastructures used during the operations, the operations to be carried out, people involved and their actions and the interactions between them, policies that are followed, resources and information needed, and issues such as timing for cooperation during the operation.

To illustrate these aspects, three types of models are initially chosen and built: a *Meta-Model* provides a taxonomic structure to capture all the high-level and fundamental concepts that are important to the air operation, a *Business Model* to capture the infrastructure involved in the operations and the detailed concepts that are used in the context of air operations, a *Role Activity and Communication Model* to identify the type of actors who are involved in the operations, their individual operations and the interactions between them.

Although these modelling languages were appropriate to the needs of the project, we commonly faced the problem, as any other *MPM* initiatives would, of keeping track of information that is distributed and shared between different models and to make sure that this is consistently represented in all models. Furthermore, on the higher level of abstraction, the "principles of business operations" that are described, subscribed and implied in all models must also be consistent with each other. To obtain and maintain such consistency is a highly labour-intensive task and can be error-prone when no computing aid is available.

Examples of how logical methods can provide support to obtain and maintain such consistency are described in detail in the later sections. We firstly give an overview of our multi-perspective modelling approach.

2 Overview of Multi-Perspective Modelling Framework

To maximise the advantages of the *MPM* approach proposed in this paper, a few principles are followed. Firstly, all of the modelling languages that are chosen as a part of the enterprise model set must be suitable to describe the chosen problem domain and appropriate to achieve the modelling objectives. Secondly, the chosen modelling languages should be complimentary of each other so that all of the concerned knowledge are described among them. Thirdly, these modelling languages should be "compatible" with each other, i.e. their modelling principles are sufficiently similar to each other so that the built model can achieve a consistent and coherent view of the domain.

It is also important that all of the models are built based at the same (or at least similar) level of abstraction: if one (or more) of the modelling language allows multiple levels of abstraction, e.g. the modelling languages of *IDEF0*[12] or *IDEF3*[11], appropriate guidelines must be established to determine which level of abstraction is mapped to the other non-decomposable modelling languages. Discussions about modelling approach and concept mapping principles between models in a MPM initiative are given in more detail in [3].

Figure 1 shows our *MPM* approach. As mentioned earlier, three models are used: *IBM BSDM's Business Model (BM)*[9]²[8], *Meta-Model (MM)*[2], and *Role Activity and Communication Diagram (RACD)*[10].³ Each of the three circles represents the domain knowledge that is covered by each model. The overlapping areas denote the common knowledge that is covered in different models, although it is presented in different forms (i.e. using the specialised model primitives) in each model. The area that is covered by only one model denotes the specialisation of the particular modelling language that describes the kind of knowledge that cannot be captured by any other models. An example of such specialised knowledge is the type of "role" that people play in an air operation as well as its responsibilities and operations which is only covered by *RACD*.



Figure 1: Overview of Multi-Perspective Modelling Approach



Figure 2: MPM using Meta-Model as a Backbone

Figure 2 depicts how the *Meta-Model (MM)* has been used as a backbone, a *light-weight ontology*, for the *MPM* approach. It provides a taxonomic structure to store the fundamental and important knowledge of the domain. Two main types of knowledge have been captured: the higher-level classification information about the domain itself and the (lower-level) model concepts that are often represented as model primitives in other models. Typically, an instantiation of a lower-level model concept has a direct correspondence to objects of that described domain area.

²BSDM stands for Business System Development Method.

 $^{{}^{3}}RACD$ was developed by the author of the paper specifically to meet requirements for the AOEM project. It was adopted from the *Role Activity Diagram*[13] with its process notations extended with influences from *IDEF3*.

Because the information stored in the *MM* is common and sharable between different models, it forms a natural communication media for knowledge transfer and translation. Model concepts that are described in one model are mapped to *MM* which are then mapped to model concepts that are captured in another model. Given the mapping of concepts, the appropriate knowledge can be mapped, shared and translated between models. This approach has been implemented in a rule-based system, *KBST-EM*[3], as a part of the *AOEM* project. More details of the semantics of a *Meta-Model* and how model primitives and concepts are mapped between multiple models through *MM* are given in [2] and [3].

3 Obtaining and Maintaining Consistency and Integrity

Enterprise models are domain-specific because they are often used to describe a business domain that is within the operational context of an organisation; they are also method-specific, i.e. they are often described using a chosen modelling language following its standard practice to achieve a desired goal. Researchers have provided guidelines and mechanisms in the past to obtain and maintain the integrity and consistency among data.

The *Entity Relational Modelling* method often includes tasks of capturing domainspecific integrity constraints as a part of its *Data Modelling* exercises in which the integrity constrains are checked against the design of schema before the actual database is implemented [5]. Thalheim[15] classifies such constraints in a three-layered hierarchical structure and describes them in a formal algebra. Domain-specific integrity constraints are sometimes called "business rules" which indicate the directives and policies that are followed in a business operation. To obtain the consistency between business rules, Grosof[7] suggests a prioritised conflict-handling mechanism for business rules based on logical inferencing techniques. Various types of BSDM's business model method-specific consistency checking and critiquing are described in detailed in [4].

All of the above research describes the capturing and handling of integrity constraints for only one modelling method and of the same "aspect" of the information. This paper focuses on identifying generic integrity rules that are applicable across models (each model may be using different modelling languages) and that each of the models may cover different perspectives of the same knowledge body. The rationale is that although these models cover different perspectives of the domain, since they are describing the same domain, they should together present a consistent view of that domain.

The issues to be considered here are two fold: the similarity of model concepts that have been captured in different models and the compatibility of model primitives used in different models. It is the case that the more similar the model concepts and the more compatible the modelling primitives of different models are the more plausible it is that integrity rules can be constructed and applied. The concept mapping mechanism that is described in the previous section tackles the issues of matching concepts between different models. Based on this mechanism the integrity-checking can be identified and performed on the mapped concepts. We use $A \cong B$ to denote that model concept A is (conceptually) *fully equivalent* to model concept B, where A and B may be captured in different models and represented in different forms, i.e. using the appropriate model primitives in that modelling language. We use $A \rightleftharpoons B$ to denote that model primitive A is *compatible* to model primitive B, where A and B are used in different modelling languages. Two types of inference operators of different strength of enforcement have been deployed: $A \Rightarrow B$ means that if A is true then B *must be* true, $A \triangleright B$ means that if A is true then B *must be* true.

A set of fundamental consistency rules can therefore be defined below:

(1) Consistent Representation of Information:

 $\begin{array}{l} \forall T1, M1, T2, M2, O1. \\ model_primitive_of(T1, M1) \rightleftharpoons model_primitive_of(T2, M2) \land \\ object_type((O1, T1), M1) \cong object_type((O2, T2), M2) \land \\ object_attribute_in_model((Value, Att), (O1, T1), M1) \\ \Rightarrow \\ object_attribute_in_model((Value, Att), (O2, T2), M2) \end{array}$

where $model_primitive_of(T1, M1)$ indicates that T1 is a model primitive (type) of model M1; $object_type((O1, T1), M1)$ defines that the model object O1 is of model primitive type T1 in Model M1; $object_attribute_in_model((Value, Att), (O1, T1), M1)$ stores the attribute value in Value for attribute Att for model object O1 in model M1. This formula indicates that **if** two model primitives, T1 and T2, in models, M1 and M2, are compatible and that the model objects, O1 and O2, of model primitive (type), T1 and T2, are fully equivalent, **then** the corresponding attributes of O1 and O2 must be the same.⁴

This consistency rule ensures the information that is shared across models is consistently represented in relevant models.

(2) Correct Conceptualisation of Information:

 $\begin{array}{l} \forall T1, M1, T2, M2, O1, O2, S1. \\ model_primitive_of(T1, M1) \rightleftharpoons model_primitive_of(T2, M2) \land \\ object_type((O1, T1), M1) \cong object_type((O2, T2), M2) \land \\ sub_concept(S1, O1, M1) \\ \Rightarrow \\ \neg \exists S2.object_type((S1, T1), M1) \cong object_type((S2, T2), M2) \land \\ sub_concept(O2, S2, M2)) \end{array}$

where $sub_concept(S, O, M)$ denotes that model concept S is a sub-concept or specialisation of model concept O in model M. This formula imposes a consistent definition on the specialisation of concepts between models, i.e. **if** two model primitives, T1 and

 $^{^{4}\}mathrm{The}$ same attribute may also be given a different name in different models. We simplify this in the formula.

T2, in models, M1 and M2, are compatible and model objects, O1 and O2, in model M1 and M2, are fully equivalent, and that the model objects S1 is the sub-concept of O1, **then** it *must not* be the case that another concept S2 is found in model M2 which is conceptually fully equivalent to S1 and is the super-concept of O2.

(3) Consistent Application of Dependencies

 $\begin{array}{l} \forall T1, M1, T2, M2, O1, D1. \\ model_primitive_of(T1, M1) \rightleftharpoons model_primitive_of(T2, M2) \land \\ object_type((O1, T1), M1) \cong object_type((O2, T2), M2) \land \\ depends_on(O1, D1, M1) \\ \rhd \\ \neg \exists D2.object_type((D1, T1), M1) \cong object_type((D2, T2), M2) \land \\ depends_on(D2, O2, M2)) \end{array}$

where $depends_on(O, D, M)$ indicates that information that is represented in model concept O depends upon the "existence" of information that is represented in model concept D in the model M. The above rule states **if** two model primitives, T1 and T2, in models, M1 and M2, are compatible and model objects O1 and O2, in model M1and M2, are fully equivalent, and that the information that is represented in the model objects O1 is depended upon information that is described in model objects D1, **then** it *should NOT* be the case that another concept D2 is found in model M2 which is conceptually fully equivalent to D1 and conceptually depends on the information that is stored in model concept O2. Because the type of "dependency" that is indicated here is generic, we use the weak inference operator to leave flexibility for model design.



Figure 3: Example Dependencies in Different Models

The above rule enforces a consistent business practice that is described in different models (or indeed different parts of the same model). Figure 3 gives examples of where dependencies can be derived. Three examples are given: dependencies may be derived from a process model, data model and a business model, each are captured in the specialised model primitives. Figure (a) abstracts a structure that is commonly seen in a process model⁵ where the execution of process 2 is proceed by process 1 which produces information D1 that is used to generate information O1 by process 2. Based on such an architecture, one can derive that model object O1 depends upon model object D1, since instances of O1 can not be created unless instance(s) of D1 has (have) been created.

Figure (b) shows an example data model. In this case, we have used the notation from an *Entity-Relational* data model, but similar dependency relationships can be found in other types of data models. In the data model, a special type of relationship, *Total Participation*, has been used to indicate an *Existence Dependency*[5] which gives the constraint that every instances of entity O1' must be related to at least one instance of entity D1' in the relationship Rel'. Figure (c) specifies a similar dependency relationship that model object O1'' is depended on D1'' in a dependency link⁶

These above examples suggest that dependency is a common property in many different modelling languages and that it can be extracted and generalised. Given the mapping of model objects between different models, the consistent application of dependencies can be checked across models.

The above consistency rule generalises all types of dependencies that may be described in modelling methods. Therefore, a weak inference (may be) has been used to allow modelling flexibility. In fact, there are many different types of dependencies, each may indicate a different level of "dependability". When application-domain-specific or method-dependent dependencies which have a strong implication on the described models are identified, the strong inference operator, \Rightarrow (must be), can be used.

(4) Detecting Incompleteness

 $\begin{array}{l} \forall R1, M1, R2, M2, T1, T2, O1, P1, O2, P2. \\ relationship_type(R1, M1) \rightleftharpoons relationship_type(R2, M2) \land \\ model_primitive_of(T1, M1) \rightleftharpoons model_primitive_of(T2, M2) \land \\ object_type((O1, T1), M1) \cong object_type((O2, T2), M2) \land \\ object_type((P1, T1), M1) \cong object_type((P2, T2), M2) \land \\ in_relation(R1, O1, P1, M1) \\ \end{array}$

 $in_relation(R2, O2, P2, M2)$

where the predicate $in_relation(R1, O1, P1, M1)$ specifies that model object O1 is associated with P1 in the relationship type R1 in model M1. The predicate $in_relation$ is generic that it includes any kind of relationships that may be described between two

⁵Example processes are those in the *IDEF0[12]*, *IDEF3[11]*, *RACD[3]* models.

⁶This notation is taken from *BSDM's Business Model[9]*.

model objects. This above rule states that **if** the relationships, R1 and R2, are compatible, and that model primitives T1 and T2 are compatible, and that model object O1is fully equivalent to model object O2, and that model object P1 is fully equivalent to model object P2, and that model object O1 is associated with model object P1 in the relationship type R1 in model M1, **then** it may be the case that model object O2 is associated with model object P2 in the relationship type R2 in model M2.

Based on the appropriate mapping of relationships and model primitives between two models, this rule suggests a relationship, R2, in the second model M2 based on observations made on the first model. Since the predicate $in_relation$ includes any relationships which makes the above rule quite generic, a weak inference, \triangleright (may be), is therefore used. The rule is an example of what can be done to detect and *suggest* missing information in a model which contributes to the completeness analysis as a part of the consistency checking process. Similar principles can be applied to infer more specific types of information and more specific results may be concluded from that.

(5) Transferable Property of Full Equivalence

 $\begin{array}{l} \forall O1, T1, M1, O2, T2, M2, O3, T3, M3. \\ model_primitive_of(T1, M1) \rightleftharpoons model_primitive_of(T2, M2) \land \\ object_type((O1, T1), M1) \cong object_type((O2, T2), M2) \land \\ model_primitive_of(T2, M2) \rightleftharpoons model_primitive_of(T3, M3) \land \\ object_type((O2, T2), M2) \cong object_type((O3, T3), M3) \\ \Rightarrow \\ object_type((O1, T1), M1) \cong object_type((O3, T3), M3) \end{array}$

This rule states that **if** model primitive T1 (in model M1) is compatible with model primitive T2 (in model M2), and that model object O1 is *fully equivalent* to model object O2, and that model primitive T2 is compatible with model primitive T3, and that model object O2 is *fully equivalent* to model object O3, **then** model object O1 *must be fully equivalent* to model object O3 (note that the compatibility between model primitives are not transferable). The "transferable" property of "full equivalence" allows knowledge that is common and sharable among different models to be transfered and communicated between models. It also provides a basis for (automatic and semiautomatic) support for obtaining and maintaining consistency between models.

All of the above rules are generic and can be used to check any two model concepts that are captured in any two models. We propose a systematic and incremental way of deploying the above rules using a *Meta-Model (MM)* as a mapping medium. The *Global Consistency* can be reached among all models by exhaustively determining *Pair-wise Consistency* between all models assuming that *Local Consistency* has been reached within each model. This is the three-tier framework which uses incremental efforts to allow models (each described in different modelling languages) to be gradually added to the "consistent set of models" to achieve the *Global Consistency*. This framework is described in detail in [3].

The process of achieving *Global Consistency* is an iterative one. It sometimes requires a revisit of the model design phase as (new) information has been discovered and added to the model. Since every model needs to maintain consistency with every other model, in the worst scenario, the checking and updating activities may continue infinitely and become computationally intractable (even to achieve the local consistency of a model may not necessarily be computed in general). However, in our experience so far, such occasions rarely occur if the modelling languages have been chosen to be compatible with each other and the models have been carefully built. Typically, when an update does trigger a few other updates it does not trigger an infinite loop.

Conclusion 4

The Multi-Perspective Modelling approach has been adopted to describe a complex domain, Air Operation. We found this approach suitable and often necessary when such a complicated domain must be captured and understood. Although the MPM approach is valuable in describing and prescribing the context and operations of an organisation, one important issue is to ensure the quality of the built models is high. We propose a framework which makes use of a light-weight ontology, a *Meta-Model*, as the underlying concept sharing mechanism to allow knowledge sharing, and obtaining and maintaining consistency across models which are described in different modelling languages. This work suggests how formal logical methods can be used to provide a foundation to support a framework that is independent of modelling language and application domain knowledge. As a result, it enhances the process of model qualityassurance.

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