MODEL DRIVEN DEVELOPMENT OF BUSINESS PROCESS MONITORING AND CONTROL SYSTEMS

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Abstract: This paper describes a model-driven approach in monitoring and controlling the behaviour of business processes. The business-level monitoring and control requirements are first described by a series of policies that can be combined together to construct a Directed Acyclic Graph (DAG), which can be regarded as the Platform Independent Model (PIM) for the high level business solution. PIM provides a convenient and clear way for business users to understand, monitor and control the interactions in the target business process. Then the PIM is transformed to an executable representation (Platform Specific Model, PSM), such as BPEL (Business Process Execution Language for Web Service) by decomposing the DAG into several sub-processes and modelling each sub-process as a BPEL process that will be deployed to runtime.

1 INTRODUCTION

Business process monitoring and control systems provide real-time information about execution status of business process as well as performance evaluation. By having this capability, business users can configure, track and analyze desired key performance indicators (KPI) and take actions. For example, business managers want to identify and resolve business problems such as whether customer orders are delivered promptly, out of stock, etc. Generally, business users that are doing process monitoring and control are divided into three roles: (1) Business Analyst who defines KPIs to be observed. An example of KPI can be the cycle time to process customer order. (2) Data Specialist who defines the data logic required to filter, cleanse, and correlate events. Correlation rules (patterns) are used to specify what event patterns need be caught and data carried in them should be extracted. (3) Operation Manager who defines what business situations (or exceptions) must be monitored or resolved as well as actions to be taken when some situation occurs. E.g. when a server unreachable exception occurs, monitoring system should send a notification to administrator to ask him restart the server.

Business users work together to design a business process monitoring and control system. After system design has been finished, KPIs, correlation patterns and business situations are defined and data sources identified. Traditional development methods require the solution to be developed from scratch. In this paper, we propose using model-driven approach to develop business process monitoring and control systems. Two advantages can be stated for model-driven approach: (1) Save cost and reduce development time: Monitoring and control solution can be defined at the business process level without being burdened by implementation detail of target platform. There is no need of navigating through development lifecycle. (2) Increase the software quality: The transformation algorithm to transform platform-independent model (PIM) to executable representation is similar to language compilers that translate the higher-level instructions into native processor instructions, which can be interpreted by the machine. Once the transformation tool has been well developed and thoroughly tested, it can be reused and the quality of the software generated by it can be guaranteed.

In general, the method of developing such systems consists of 5 steps as shown in Figure 1.

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Step 1: Business Design Phase where different roles define the subset of the monitoring and control solution at the business level (e.g. KPIs, situations, actions, correlation rules).

Step 2: Model Generation Phase where system designers create a set of enforceable policies for the models defined in step 1;

Step 3: Composition Phase where policies defined in Step 2 are composed into a directed acyclic graph (DAG) that represents the platform-independent model (PIM) for the monitoring and control systems.

Step 4: Decomposition Phase where the system decomposes the generated DAGs from Step 3 into several sub-processes that are transformable to executable modules in the target platform;

Step 5: Transform Phase where each sub-process is transformed into an executable module. In this paper, BPEL is used as the example of such modules. Each BPEL process can be wrapped as a service and they can communicate with each other through event bus. By doing so, the PIM (policies) can be transformed into PSM (BPEL) that can be executed by an executable runtime engine.

The rest of this paper is organized as follows. Section 2 introduces some background information and related work. Section 3 talks about MDA and our proposed MDA approach for business process monitoring and control. Section 4 presents the detailed information about process decomposition and transformation. Section 5 gives an account of our experience of using model-driven approach for developing business process monitoring and control systems. Section 6 summarizes the whole paper.

2 RELATED WORK

Model-Driven Architecture (MDA) that has been defined by the Object Management Group (OMG) is a new approach to application design and implementation. It helps computer users solve integration problems by supplying open, platform-neutral interoperability specifications (OMG, 2001). MDA approach is widely used for information and service integration (Siegel, 2002). Our approach provides a domain specific model-driven approach based on MDA with specific target on business process monitoring and control. Many business process monitoring and control tools are developed by different organizations, such as QName! from mqsoftware (MQSoftware, 2004), TransactionVision from Bristol Technology (TransactionVision, 2004). The benefit brought by monitoring and control include lower the process cost and faster the process execution. However, they do not use model-driven approaches. PolarLake Inc. (PolarLake, 2004), provides a technology how to use BPEL and XML to automating business process management. BPEL is simply an example of transformation target in our paper. Our approach can help BPEL solutions in a much simpler way.

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MDA (Soley, 2000) is a framework of software development. Models and model driven software development are the key elements in MDA. A typical model-driven development process includes three steps: (1) Build a model with high level abstraction. This model is a PIM that is independent of any implementation technology. In our business process monitoring and control system, the PIM is the DAG constructed by a set of policies as shown in step 3 of Figure 1; (2) Transform PIM into one or more PSM. A PSM is related to a specific implementation technology, such as J2EE model, EJB model. In our system, we choose BPEL process as target PSM; (3) Transform PSM to code to be executed by machine. This step is usually completed by the implementation technology related to the PSM. In our system, BPEL process is interpreted and executed by a BPEL engine.
3.1 Policy Description

There are three groups of policies defined by business analyst, data specialist and operation manager: KPIs expressions, Correlation rules and Action policies.

A KPI expression is made up of parameters and operators. Based on parameter data types in KPI, there are several categories of operators: Boolean, Arithmetic, Relational, Set, String, Calendar, Vector and etc. Table 1 shows some common operators in each category. Some Examples of KPIs: 

- \( \text{process\_time} = \text{response\_timestamp} - \text{request\_timestamp} \)
- \( \text{Server\_Down} = (\text{Count (server\_down\_events)} > 30) \) within 30 seconds

Business processes interact with one another and the environment through events. Events are captured by business process monitoring and control system. Many of the captured events are meaningless to specific monitoring and control system while others need to be considered in a specific pattern. Correlation rules (patterns) are used to specify the event patterns that need be caught and data carried in them should be extracted according to requirements.

The definition of a correlation rule includes a number of rule-specific parameters (such as threshold, time period), event selection criteria to select events that relevant to the rule and actions should take (defined by action policy) once the rule fires (Bussani, 2003).

We define seven basic correlation rules for our system: Match, Block Duplicates, Update Last, Collection, Threshold, Sequence and Sequence Absence. Match is the only stateless rule, in which event are treated independently. All the others are stateful, in which events rely on previous detected events and they depend on each other. These rules are defined based on IBM Zurich Correlation Engine rule definition (Bussani, 2003). Correlation rules can be defined using XML syntax. The format of rule definition is shown in Figure 2 as follows.

```
<rule id = "rule identification"> --- Rule identifier
<rule type [attributes]> --- Rule Type
<selection criteria> --- Event selection criteria
......
</selection criteria>
</rule type>
<action policy = "policy name"> ... </action>
------------ Actions to take, defined by action policy
</rule>
```

Figure 2: Correlation Rule Definition

Action policy provides policy rules for system behaviour in response to business situation. A simple example of action policy is send notification to administrator once server unreachable event has been detected by correlation rules defined above.

Action policy can also be defined using XML syntax. The definition of action policy includes a number of policy-specific parameters, the target of the policy ---- messages generated by action policy will be sent to the defined target, a series of KPIs used by policy (defined by KPI expression), a set of correlation rules that triggered this policy (defined above) as well as actions will take once the policy has been triggered.

```
<policy name = "policy name">  --- policy name
<parameters> .... </parameters>
--- parameters will be used in policy
<target> .... </target>
--- destination for message generated by action policy
<metrics> .... </metrics> --- metrics used
<correlation rules>
--- correlation rules that trigger the policy
<rule id = "">... </rule>
</correlation rules>
<action list> --- actions will take
<action function = "function name">
....
</action>
</action list>
</policy>
```

Figure 3: Format of action policy definition

3.2 PIM construction

In a business process monitoring and control system, business analyst, data specialist and operation manager define different policies (KPIs, correlation rules and action policies) for a business situation. After all policies have been defined, the system will compose them to form a DAG as the PIM for high level business solution (as shown in step 3 of Figure 1).

The question is: How to combine different policies into a DAG? Which policies should be connected to each other? Each policy has a set of input and output interface definitions describing the message formats it expects to accept and to generate, finding policy pairs is to match an input and an output interface definition of two policies.

This problem is quite similar to the service composition problem in Web service field, where we need to integrate different services into a business process. A lot of researches have been done in semantic web service composition ([9,10]).
Currently only simple interface match checking is used in our system. In following four conditions, we define the condition when two policies’ interfaces match with each other. More matching conditions will be added in our future work. ($AO$: output of policy A, $BI$: input of policy B):

1. If $AO = BI$, match successful; (Figure 4.(a))
2. If $AO \subseteq BI$, match successful; (Figure 4.(b))
3. If $AO \supseteq BI$ and the matching part can be separated out from $AO$, match successful. (Figure 4.(c)).
4. If $AO \cap BI \neq \emptyset$ and the matching part can be separated out from $AO$ and also from $BI$, match successful. (Figure 4.(d))

Figure 4: Policy Interface Matching Condition

Obviously, the similarities of two policies interfaces are decreased from condition 1 to condition 4. If two policies interfaces match successfully, we can add a directed link (from policy A’s output to policy B’s input) between them. Considering each policy as a node, a graph can be constructed by adding links between all the matching policy pairs. And because the special features of business process monitoring and control system, it must be a DAG. If there’s a loop exist in the constructed graph, we can remove the link that has the minimal similarity in the loop and continue this step until there is no loop in the graph. This DAG is the PIM for high level business solutions.

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A composed DAG cannot be transformed into executable modules unless it can be decomposed into sub-graphs each of which is transformable to executable modules in target runtime platform. After the DAG (PIM) has been constructed by connecting policy pairs, we need to transform the PIM into one or more PSM.

There are many kinds of PSM, e.g., BPEL, workflows, Web services. We have been using BPEL as our target PSM. Usually, it is too complicated to present the entire PIM by a single BPEL process. Firstly, the DAG is divided into several parts, each of which is a sub-process and can be transformed into a BPEL process. This task is called model decomposition.

For example, we can decompose the DAG shown in step 3 of Figure 1 into four sub-processes, as shown in step 4. The sub-process 1 contains policy P1, P3 and P8, sub-process 2 contains policy P0 and P2, sub-process 3 contains policy P4, P5, P6, while sub-process 4 contains policy P7.

Several issues need to be considered during the decomposition process: (1) Are there any criteria for establishing a bottom level process component? Which means, how do we decide when to stop the process decomposition? (2) If there exists more than one way to do the decomposition and generate different sets of sub-processes, which one is the best?

The reason of decomposing models into smaller ones is because, in many cases, it’s too difficult (sometimes impossible) to transform the entire complex PIM into one executable modules (for example, a single BPEL process). So the basic criterion for decomposition is to make sure the fact that after decomposition, each sub-process can be presented by a BPEL process and all generated BPEL processes can communicate each other to achieve the original objective. Thereby, model decomposition is stoppable when each sub-part is presentable as PSM.

To show the concepts, we show several basic rules to perform process decomposition as follows.

1. Each policy must be included in at least one sub-process; Some policies can reside in more than one sub-processes;
2. Each sub-process contains at least one correlation rule, one KPI expression and one action policy, as shown in Figure 5. The correlation rule takes inputs from outside or from the output of other sub-processes and passes them to metrics for some calculation. The results produced by metrics expressions will be passed to action policy and take the predefined actions to generate outputs.

Figure 5: Sub-process
decomposition of PIM (i.e. DAG) can be divided according to given correlation rules. Each rule can be considered as a starting point of a sub-process, and the KPIs and action policy related to the correlation rule will be added to generate a complete sub-process.

4. Combine some small simple sub-processes to be a bigger sub-process:
   For all the sub-processes constructed on correlation rules, some of them are very simple. For example, the sub-process that contains Match correlation rule may only filter some special events out and extract the message from the event and assign to another variable, finally, the message will be sent out to another sub-process by action policy.
   
   If there are two sub-processes with Match rules, one takes the other one’s output as its input, we can combine them into a new bigger sub-process that contains two Match rules. Figure 6 shows what the system model looks like after decomposition. Four sub-processes have been generated and each of them will be transformed to a BPEL process.

   There is always a tradeoff between the complexity of each sub-process and the efficiency of the whole system. If the whole system is divided to very fine-grained, each sub-process only performs a simple task but there are lots of sub-processes exist and they must communicate with each other to achieve the original objective. In this case, system efficiency is decreased by large amount of communications and message exchanges among sub-processes.

   On the other hand, if the whole system is divided to a very few large-grained sub-processes, each sub-process need to perform a lot of work and tend to be complicated. Of course, there will be less communications overhead and the system turns to be more efficient. The detailed information and algorithms for model decomposition will be reported in our future work.

   Due to the space limitation, the detail of generated BPEL modules is not presented in this paper. Only an example of using BPEL to model correlation rules is given as follows, where the rule is sequential event patterns (Figure 7).

   This process describes (1) Receiving “start process” command from partner “Process Initializer” (command =1 in Initialization variable) through <Receive> activity defined by BPEL; (2) Keep running to receive events messages from “Monitor” through <Pick>; (3) A <Switch/Case> is used to check whether the received event matches the selection criteria; (4) If event matches the selection criteria, contents of notification will be set by <Assign> activity and sent out to “Admin” through <Invoke>; (5) Go back to Step 2 and repeat Step 2–4 to detect all events that match the selection criteria.

5 LESSONS LEARNED

We have applied this approach to developing supply chain management systems in the domain of microelectronic manufacturing (Jeng, 2004) and transportation management systems. The development time has been greatly reduced to 30-40% of the originally defined development cycle. As long as the policies defined by the business roles are accurate and precise, the software generated is guaranteed to run almost correctly.

As mentioned, we have used BPEL as our target PSM. Our approach lead to a systematic way of developing BPEL modules based on models that are developed in a distributed fashion. The only factor hard to be predicted is the overhead induced by those generated BPEL modules. Since current implementation of BPEL engine heavily depends on database synchronization and our transformation algorithms have not considered concurrency issues, database access deadlock happen frequently. Hence,
generated BPEL modules need manual tuning to achieve satisfied performance. Another issue is tooling.

To make model driven approach to be successful, high level business policies need to be captured in an accurate manner. Due to lack of policy tools, the early phases of development process become disproportionately large.

In general, the contribution of our work lies on:
1. Propose a new approach (MDA approach) to provide solution for business process monitoring and control system, which is faster, more cost-effective and reliable compared to traditional way;
2. Present PIM by a DAG, which is constructed by a series of policies (includes KPIs, correlation rules and action policies) through interface matching;
3. Transform PIM to executable representation (BPEL in this paper) by first decomposing the entire PIM into several sub-processes to increase the feasibility and decrease the complexity of transformation;
4. Show how to model different correlation rules, which is the foundation of PIM, in BPEL process. We are among the first ones that are doing this kind of work.

6 CONCLUSION

In this paper, we have presented the model-driven approach for developing business process monitoring and control systems. The solution is first described by the high level abstract model (PIM), which is independent from platform and implementation technologies. This PIM is presented as a DAG that is constructed by a series of policies described in XML. Then the PIM is decomposed into several sub-processes that can be easily transformed into an executable representation, such as BPEL or JAVA. We use BPEL as the example to show the model transformation.

There is still a lot of work need to be done in the future: (1) Algorithms/rules for PIM construction through interface matching; (2) Algorithms for process decomposition in order to find a optimal division for entire PIM; (3) More correlation rules definition (currently, only 7 rules are defined); (4) Prototype implementation for using BPEL to model correlation rules; and (5) Prototype implementation for business process monitoring and control system to verify our proposed approach.

REFERENCES


