Model-Driven Productivity Evaluation for Self-Adaptive Context-Oriented Software Development

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Model-Driven Productivity Evaluation for Self-Adaptive Context-Oriented Software Development

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Abstract—Anticipating context changes using a model-based approach requires a formal procedure for analysing and modelling their context-dependent functionality, and a stable description of the architecture which supports dynamic decision-making and architecture evolution. This article demonstrates the capabilities of the context-oriented component-based application-model-driven architecture (COCA-MDA) to support the development of self-adaptive applications; we describe a state-of-the-art case study and evaluate the development effort involved in adopting the COCA-MDA in constructing the application. An intensive analysis of the application requirements simplified the process of modelling the application’s behavioural model; therefore, instead of modelling several variation models, the developers modelled an extra-functionality model. COCA-MDA reduces the development effort because it maintains a clear separation of concerns and employs a decomposition mechanism to produce a context-oriented component model which decouples the applications’ core functionality from the context-dependent functionality. Estimating the MDA approach’s productivity can help the software developers to select the best MDA-based methodology from the available solutions proposed in the literature. Thus, counting the source line of code is not adequate for evaluating the development effort of the MDA-based methodology. Quantifying the maintenance adjustment factor of the new, adapted, and reused code is a better estimate of the development effort of the MDA approaches.

Keywords-model-driven architecture, MDA productivity, COCOMO ii, self-adaptive application.

I. INTRODUCTION

Mobile computing environments are heterogeneous and dynamic. Everything from the devices used and resources available to network bandwidths and user context can change drastically at runtime [1]. This presents the software developers with the challenge of tailoring behavioural variations both to each specific user need and to the context information. Context-dependent behavioural variations can be seen as a collaboration of individual features expressed in requirements, design, and implementation. Before encapsulating the crosscutting context-dependent behaviours into a software module, the developers must first identify them both in the requirements document and in the software model. This is difficult to achieve because, by their nature, context-dependent behaviours are entangled with other behaviours, and are likely to be included in multiple parts (scattered) of the software modules. Using intuition or even domain knowledge is not necessarily sufficient for identifying the behavioural variations; instead, it requires a formal analysis procedure for the software requirements and a separation of their individual concerns. Moreover, a formal procedure for modelling these variations is needed. This kind of analysis and modelling procedure can reduce the complexity in modelling self-adaptive applications and encapsulate the context-dependent part of the distinct architecture module (component).

A context oriented component model (COCA-component) [2] is used to encapsulate behavioural variations and decouple them from the application’s core functionality. In this way, dynamic component composition is achieved. Additionally, from the software developer’s perspective, it is vital to know the productivity of the development paradigm which might be used in constructing the self-adaptive application. Productivity evaluation of model-driven approaches can assist the developers in selecting among the proposed methodologies in the literature which achieve dynamic behavioural variations of self-adaptive software. Context-oriented component-based application-model-driven architecture (COCA-MDA) emerged as a development paradigm which facilitates the development of self-adaptive context-oriented software [3], [4].

This article evaluates the development effort involved in adopting COCA-MDA when constructing a self-adaptive application for an indoor wayfinding application (IWayfinder) developed for individuals with cognitive impairments. The development effort of COCA-MDA is compared to other model-driven approaches proposed in the literature.

The remainder of the article is structured as follows. Section II provides a comparative analysis of related studies. Section III demonstrates a case study. The COCA-MDA phases are described in Section IV. Section V provides a COCA-MDA evaluation using constructive cost model II (COCOMO II). Section VI summarizes the research findings and describes directions for future work.

II. RELATED WORK

In the literature, there are several MDA approaches which target the development of self-adaptive applications for mobile
computing environments which produce component-based applications; this study borrows from the following methodologies: MUSIC, proposed by [5]; U-MUSIC [6]; and Pasapallis MDA [7].

The MUSIC development methodology [5] adopts a model-driven approach to constructing the application variability model. The applications are built as a component framework with component types as variation points. Middleware is used to resolve the variation points, which involves the election of a concrete component as a realization of the component type. Using this method, a number of application variants can automatically be derived.

The U-MUSIC methodology, proposed by [6], adopts the model-driven approach to constructing self-adaptive applications and enabling dynamic unanticipated adaptation based on a component model. The U-MUSIC system enables the developers to specify the application variability model, context elements, and data structure. The developers are able to model the component functionalities and quality of service (QoS) properties using an abstract, platform-independent model (PIM).

Pasapallis [7] proposes another MDA-based methodology which considers the context providers for the application. For each context provider, a plug-in is proposed during the design phase. At runtime, a utility function is used to consider the context state and perform decision-making. Once the plug-in is selected (to be load into the application), middleware support performs dynamic runtime loading of the plug-in. However, it is impossible to consider all the context providers which might produce context information at runtime.

In MUSIC, U-MUSIC, and Pasapallis approaches, dynamic decision making is supported by a utility function. The utility function is defined as the weighted sum of the different objectives based on user preferences and QoS. However, this approach suffers from a number of drawbacks. First, it is well known that correct identification of the weight of each goal is a major difficulty. Second, the approach hides conflicts among multiple goals in a single, aggregate objective function, rather than exposing the conflicts and reasoning about them. At runtime, a utility function is used to select the best application variant; this is the so-called ‘adaptation plan’. Realistically, it is impossible for the developer to predict all possible variations of the application when unanticipated conditions could arise. In addition, mobile computing devices have limited resources for evaluating the many application variations at runtime and can consume significant amounts of device resources. As an outcome, the benefit gained from the adaptation is negated by the overhead required to achieve the adaptation. Because of the above issue, it is impossible to use MUSIC to provide unanticipated adaptation in a self-adaptive application. Moreover, modeling the application using U-MUSIC, MUSIC, and Pasapallis’s MDA produces an architecture with a tight coupling between the architecture and the target platform.

Lewis et al. [8] have evaluated the impact of MDA on the development effort and the learning curve of the MDA-based development tools based on their own experiences. The authors concluded that the real potential behind MDA is not completely employed either by current tools or by the proposed MDA approaches in the literature. In addition, the developers have to modify the generated code such that it is suitable for the target platform. The degree to which the generated code needs modification is affected by the MDA tools used. In the same way, the developer’s understanding of the MDA tasks and familiarity with the target platform have direct impacts on MDA productivity. Constructive cost model II (COCOMO II) [9] emerged as a software cost estimation model which considers the development methodology productivity. The productivity evaluates the quality of benefits derived from using the development methodology, in terms of its impact on the development time, complexity of implementation, code quality, and cost effectiveness [10]. COCOMO II allows estimation of the effort, time, and cost required for software development. The main advantage of this model is that COCOMO II is an open model with various parameters which affect the estimation of the development effort. Moreover, the COCOMO II model allows estimation of the software application development effort in both person-months (PM) and time to develop (TDEV). A set of inputs such as software scale factors (SFs) and 17 effort multipliers is needed. A full description of these parameters can be found in [9]. An example of an evaluation of MDA approaches with (COCOMO II) can be found in [11].

III. Self-Adaptive Indoor Wayfinding Application for Individuals with Cognitive Impairments

IWayFinder provides distributed cognition support for indoor navigation to persons with cognitive disabilities. RFID tags and QR-codes are placed at decision points such as hallway intersections, exits, elevators, and entrances to stairways. After reading the encoded URL in the QR-codes, the Cisco engine provides the required navigation information and instructs the user. The proposed self-adaptive application uses an augmented reality browser (ARB) to display the navigation directions. The browser displays the directions on the physical display of the tool’s camera. The application is able to provide the user with time-based events such as the opening hours of the building, lunch time, closing hours of the offices, location access rights which control the entrance of users to certain locations, and any real-time alarm events. Moreover, the infrastructure support allows several persons to monitor and collaborate with the user on route. The IWayFinder application and the COCA-MDA development methodology were fully described in our previous work [3], [4]. This article focuses on describing an evaluation of the cost effectiveness when adopting the COCA-MDA (among other MDA approaches) in developing the IWayFinder application. Assuming that the context information is delivered by the Cisco infrastructure, we propose the following anticipation scenarios:

A1: Self-tuning
The application must track the user’s path inside the building. When decision points (DPs) are reached, the application places a marker for each DP the user passed. If the user is unable to locate a decision point in the building, the
application must be able to guide the user towards a safe exit. The route directions can be delivered to the user in several output formats: video, still images, and voice commands. The application should change the direction output while also adapting to the device resources and the level of cognitive impairment of the individual.

**A2: Self-recovering** Assuming that the user is trapped in a lift with no GPRS connection (or in the case of a fire), the fire alarm is raised, the application is notified, and the application adopts the shortest path to the nearest fire exit. In both cases, the application submits the user’s current coordinates and an emergency help message to the emergency number, parents, career team, and security staff. The communication is achieved using the available connection, regardless of the resource cost, to alert any nearby devices to the emergent need for help. If no connection is made, the device emits an alarm sound and increases the device volume to maximum. The security staff or fire fighters receive the emergency message and can view the CCTV video to identify the floor on which the user is trapped. When the CCTV system locates the user, full information about the user is displayed, including a personal and health profile. At the same time, the application guides the user to a safe exit using a preloaded path (in case the CCTV camera is disabled and the services engine is off). Fire fighters can use the received message to locate the user within the building.

IV. COCA-MDA DEVELOPMENT APPROACH

The COCA-MDA follows the principles of the object management group (OMG) model-driven architecture. The design of a context-aware application according to the COCA-MDA approach generally involves the six phases shown in Figure 1. Modelling IWayfinder using COCA-MDA can be summarised as shown in Figure 2. The figure summarises the modelling tasks using the associated UML diagrams. The developer can start the analysis of an application scenario to capture the requirements.

**Analysis:** The requirements of the system are modelled in a computation-independent model (CIM), thus describing the situation in which the application will be used and predicting the exact behaviour of the application as a result of runtime context changes. This phase is accomplished by performing the following three tasks.

**Task 1. Requirement capturing by textual analysis:** In this task, the developer identifies the candidate requirements for the illustration scenario using a textual analysis of the application scenario. It is recommended that the developer identify the candidate actors, use-cases, classes, and activities. This can be achieved by creating a table which lists the results of the analysis.

**Task 2. Identifying the extra-functional requirements and relating them to the middleware functionality:** The requirement is classified in the requirements’ diagram, based on its type and whether it comes from a context provider or a consumer. The next level of requirements classification is to classify the requirements based on their anticipation level: foreseeable, foreseen, or unforeseen. This classification allows the developer to model the application behaviour as much as possible and to plan for the adaptation actions. However, to facilitate this classification framework, a UML profile is designed to support the requirements analysis and to be used by the software designer, as shown in Figure 3. For example, displaying the direction output in the camera browser is a functional requirement which drives the extra-functional requirement number 4, ’utilise the resources’, which requires a middleware functionality to manage the context changes and take the adaptation actions which satisfy it. This requirement is classified as the foreseeable anticipation level.

**Task 3. Capturing user requirements:** This task is combined with the previous requirements diagram. This task focuses on capturing the user’s requirements as a subset of the functional requirements, as shown in the UML profile in Figure 3. This task allows the developers to analyse the main functions of the application which achieve specific goals or objectives. Normally, this kind of requirement is expressed by ’The user must be able to do ...’.
Modelling and design: COCA-MDA has adopted the component collaboration architecture (CCA) [12] at the PIV phase by partitioning the software into two views: the structure view and the behaviour view. The structure view focuses on the core components of the self-adaptive application and hides the context-driven components. The behaviour viewpoint focuses on modelling the context-driven behaviour of the component, which may be invoked in the application execution at runtime. To achieve this function, the following three tasks are performed.

Task 4. Resources and context entity model: Resources and context model refers to a generic overview of the underlying device’s resources, sensors, and logical context providers. This diagram models the engagement between the resources and the application under development. It facilitates the developer in understanding the relationship between the context providers and their dependency.

Task 5. Use cases: In this phase, the requirements diagram is combined into a use-case model. The use-cases describe the interactions between the software system and the actor. The system-dependent and environment-dependent behaviours are modelled as extensions of the functional use-cases. The functional use-cases are modelled in a class diagram describing the application core functions. The extended use-cases are modelled as another object diagram which describes the application’s behavioural view. For example, the ‘adapt the direction output’ use-case is a contextually driven use-case which extends the application functionality to utilise the devices’ resources so as to provide a route to the nearest fire exit.

Task 6. Modelling the application core structure: In this task, a classical class diagram models the components which provide the application’s core functions. These functions are identified in the use-case diagram in the previous task. However, the class diagram is modelled independently of the variations in the context information. In this scenario, some classes, such as ‘Displaying POI’s’, ‘Route-Planning UI, CameraUI, MapUI, and User Interface’, are classified to be in the application core. These classes provide the core functions for the user during their tour. Figure 5 shows the core-structure class model ignoring any interaction with the context environment or the middleware.

Task 7. Identifying Application Variant Behaviour (Behaviour View): The use-case diagram is split into two distinct object diagrams. The first diagram describes the basic application components which are executed regardless of the execution context. The core structure is integrated with the extra-functional class model in the final architecture model. The extra-functionality class diagram provides a detailed view of the application COCA-component and the COCA-middleware. In addition, these diagrams model the desired behaviour, which can be used to anticipate context changes. Figure 6 shows a COCA-component designed to anticipate the ‘direction output’. The COCA-component implements delegate objects and sub layers; each layer implements a specific context-dependent function. The COCA-middleware [2], [13], uses this delegate object to redirect the execution among the sub layers based on the context condition.

The application behavioural model is used to demonstrate the decision points in the execution which might be reached whenever internal or external variables are found. This decision point requires several parameter inputs to make the correct choice at this critical time. Using the activity diagram, the developers can extract numerous decision polices. Each policy must be modelled in a state diagram: textbfPolicy: Direction output. This policy is attached to the ‘direction output’ COCA-
component in Figure 6. The policy syntax can be described by the code shown in List 1.

Listing 1. Decision Policy 2

<table>
<thead>
<tr>
<th>Direction</th>
<th>Memory Level</th>
<th>CPU Throughput</th>
<th>Battery Level</th>
</tr>
</thead>
<tbody>
<tr>
<td>Provided</td>
<td>&gt;= 50</td>
<td>&lt;= 89</td>
<td>&gt;= 50</td>
</tr>
<tr>
<td>Available memory</td>
<td>&gt;= 50</td>
<td>&lt;= 89</td>
<td>&lt;= 50</td>
</tr>
<tr>
<td>then {PlayVideo(); displayImage(); VoiceCommand( );}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>elif (BatteryLevel &lt; 50)</td>
<td></td>
<td>memory level &lt; 50</td>
<td></td>
</tr>
<tr>
<td>then {displayImage(); VoiceCommand( );}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>else if (BatteryLevel &lt; 20)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>then VoiceCommand();</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

The variant behaviour model is supported by a state-machine model which describes the application decision policies. The three models of the application are used as input for the next phase, model-to-model transformation.

Model-to-model transformation: The platform-independent model and behavioural model are translated into architecture description language (COCA-ADL). This phase includes model-to-model transformation and model verification for the application’s structure and behaviour views. The COCA-ADL is implemented by extending the xADL schema (an extensible XML language). ArchStudio is an environment of integrated tools for modelling, visualizing, analysing, and implementing software and systems architectures. The ArchStudio provides Archipelago as the graphical editor used to model the architecture. Archipelago was used to extend the xADL by implementing the COCA-ADL meta-model. The ArchStudio editor enables the developer to model their application using three distinct models: structure, state machine, and activity diagram [14].

Testing and validating: Tests the model and verifies its fitness for achieving the application goals and objectives.

Platform-specific model: The platform-specific model produced by the transformation is a model of the same system specified by the PIM (it also specifies how that system makes use of the chosen platform). A PSM may provide more or fewer details, depending on its purpose. A PSM will be an implementation if it provides all the information needed to construct a system and to put it into use. Alternatively, it may act as a PIM used to further refine the PSM so that it can be directly implemented.

Code generation: Model-to-text includes model-to-text transformation deployment and execution verification. The COCA-ADL XMI code is transformed into the implementation language.

V. MDA Evaluation

The IWayFinder application has been selected to determine the development effort required by COCA-MDA compared to that required by three MDA approaches which have been proposed in the literature: U-MUSIC-MDA, proposed by [6]; Paspallis’s MDA, proposed in [7]; and MUSIC-MDA, proposed in the IST-MUSIC project deliverables in [5]. The enterprise architecture tool (EA) [15] was used to develop the IWayFinder application using the four MDAs (COCA, MUSIC, U-MUSIC, and Paspallis’s). Each MDA phase was carried out separately. COCOMO II [9] was used to find the development effort in person-months for each MDA. There are two COCOMO II models, i.e. the post-architecture model and the early design model. The post-architecture model is a detailed model used once the project is ready to develop and sustain a fielded system. The early design model is a high-level model which is used to explore alternative architectures or incremental development strategies [9]. Based on the above considerations, we selected the post-architecture model to evaluate the four MDAs: COCA, MUSIC, U-MUSIC, and Paspallis’s.

<table>
<thead>
<tr>
<th>Phase</th>
<th>Sizing Method</th>
<th>Results</th>
</tr>
</thead>
<tbody>
<tr>
<td>CIM</td>
<td>Counting Unadjusted Function Points (UFP)</td>
<td>Relating UFP into SLOC</td>
</tr>
<tr>
<td>PIM</td>
<td>UFP</td>
<td>UFP into SLOC</td>
</tr>
<tr>
<td>PSM</td>
<td>Quantifying the Maintenance Adjustment Factor (MAF)</td>
<td>(Size) PM</td>
</tr>
<tr>
<td>Transformation</td>
<td>Quantifying the Maintenance Change Factor (MCF)</td>
<td>(Size) PM</td>
</tr>
<tr>
<td>Final code integration</td>
<td>Source Line of Code</td>
<td>SLOC = Final SLOC - Generated SLOC</td>
</tr>
<tr>
<td>Deployment integration</td>
<td>Quantifying the Maintenance Change Factor (MCF)</td>
<td>SLOC</td>
</tr>
</tbody>
</table>

TABLE I MDA PHASES AND SIZE FACTORS

Based on the COCOMO II model, the sizing of new and reused code can be estimated via three major methods, as described in [9]. These methods are counting source lines of code (SLOC); counting unadjusted function points (UFPs); and aggregating new, adapted, and reused code, i.e. the adapted source lines of code (ASLOC).

In general, MDA-based approaches must apply a computation-independent model (CIM), a platform-independent model (PIM), a platform-specific model (PSM), transformation, deployment, and code generation. For each phase in the MDA, a sizing method was adopted for estimating the development effort, as shown in Table I. The code which is directly generated from the MDA tool (EA) is excluded from the development effort, but is used as an input to measure the software maintenance effort by
counting the UFPs. In addition, the middleware code has to
be adopted and maintained, or even configured, to suit the
new application platform. Therefore, a third sizing measure
is used to evaluate the middleware code.

The following equations describe the effort (PM) and
the time to develop (TDEV), taking into consideration the
aforementioned inputs, as shown in Equation 1. The primary
equation in 1 denotes the effort in person-months derived
from the software size defined in thousands of source lines
of code (SLOC). The exponent E defines the sum of the scale
factors (SFs), i.e. the Cartesian product of the effort
multipliers (EMs) and the constant value A. A value was
determined based on several software projects which were
surveyed in [9]. The second equation in Equation 2 depicts
the time required to develop the software, derived from the
nominal effort (PM), the sum of SFs, and the constant values
based on several software projects evaluated by COCOMO II.
The rating scale factors and the effort multipliers were used in
this work to derive the effort and the time required to develop
the IWayFinder application using COCA-MDA.

Based on the above discussion, Equation 3 is used to
calculate the sizing of code maintenance [9]. The initial
maintenance size estimate is adjusted using a maintenance ad-
justment factor (MAF). This relationship can both estimate the
level of effort using the full time equivalent software personnel
FSPM, given TM (as found in annual maintenance estimates,
as shown in Equation 4, where TM = 12 months), and,
given a fixed maintenance staff level, FSPM, determine the
necessary time, TM, to complete the effort [9]. To estimate the
adopted code, the COCOMO II model uses an additional set of
equations to calculate the final count for source instructions,
related costs, and schedule. The equations in 3, 4, and 5 use
the following values as parameters.

- Adapted source lines of code (ASLOC). The number of
  source lines of code adapted from existing software used
  in developing the new product.
- Percentage of design modification (DM). The percentage of
  the adapted software’s design which received mod-
  ifications to fulfill the objectives and accommodate the
  environment of the new product.
- Percentage of code modification (CM). The percentage of
  the adapted software’s code which receives modifications
  to fulfill the objectives and accommodate the environment
  of the new product.
- Percentage of integration required for modified software
  (IM). The percentage of effort needed to integrate and test
  the adapted software to combine it into the new product.
- Percentage of reuse effort resulting from software under-
  standing (SU). Percentage of reuse effort resulting from
  assessment and assimilation (AA) and programmer unfa-
  miliarity with the software domain (UNFM). [9] provides
  a rating scale for programmer unfamiliarity (UNFM).

\[
PM = A \times (\text{Size})^E \times \prod_{i=1}^{17} EM_i, \quad (1)
\]

\[
TDEV = C \times (PM)^F,
\]

\[
MAF = 1 + \left( \frac{SU}{100} \times UNFM \right), \quad (3)
\]

\[
PM_M = T_M - FSP_M,
\]

\[
PM = A \times (\text{Size})^E + \left[ \frac{ASLOC(\text{AT})}{ATPROD} \right] \quad (5)
\]

Figure 8 provides the estimated efforts for the four MDAs.
Figure 7 shows the total size (SLOC) for the IWayFinder
application after it has been developed using each MDA. The
figure shows that COCA-MDA requires less effort in person-
months, despite the fact that the total SLOC is greater than for
Paspallis’s MDA. In Paspallis’s MDA, each context provider
requires a separate plug-in architecture, which requires new
software engineering to build the plug-in. The MDA tool does
not generate the required code for the plug-in, but instead
leaves the required code to be composed and configured in
the deployment stage. This increases the effort required to
configure and maintain the plug-in architecture. This effort is
captured using UFP analysis. The total effort for Paspallis’s
MDA is one of the highest because the ratio of the mainte-
nance adjustment factor is very high. These facts demonstrate
the accuracy obtained using COCOMO II in estimating self-
adaptive software development methodology. In addition, the
figure shows that the effort in MUSIC is the greatest; the
reason for this is a lower ratio of adaptive and reused code
in MUSIC compared to that in its extensions, U-MUSIC and
Paspallis’s MDA.

Figure 9 provides more information on each MDA regarding
the estimated cost per MDA phase. As shown in the figure,
the cost of performing the PIM was high for all MDAs. The
reason for this is that all MDAs focus more on modelling
the application variation model through the PIM. The cost
of adopting the PIM in MUSIC is the largest because of the
complexity of adopting the MUSIC PIM tasks; this requires
the developer to produce more UML models than in the others.
For the same stage, Paspallis’s MDA has a lower cost. In Paspallis’s MDA, the time spent by the developers to build the context-provider plug-ins is greater than the effort required to build the architecture itself. This is why Paspallis’s MDA is second, after U-MUSIC, in terms of the evaluation of the PSM phase.

On the other hand, Paspallis’s MDA increases the effort required for software maintenance in the transformation and deployment phases. Specifically, COCA-MDA and U-MUSIC reduce the effort needed to implement new or reused context providers (i.e., integrating a new sensor into the platform). This result reflects the benefits gained from employing the COCA-ADL for architecture deployment in several platforms. It is worth noting here that the ‘labour rate per month’ has been given the same value for all the MDAs throughout the evaluation.

VI. CONCLUSIONS AND FUTURE WORK

Self-adaptability requirement, modelling, architecture, implementation, and assurance approaches require a systematic solution which inter-relates all aspects on a single platform. Requirements analysis can provide a great deal of information about the extra-functionalities of the self-adaptive system. In the same way, requirements analysis can facilitate and simplify architecture reflection by providing the information required by the software to manage itself. Moreover, COCA-MDA can reduce the complexity of self-adaptive engineering through mapping requirements to actor-, system-, and environment-dependent behaviours. This study shows how COCA-MDA reduces the required development effort compared to other MDAs. It also demonstrates how COCA-MDA reduces the software maintenance ratio through the architecture deployment and transformation.

The COCA-MDA requires improvement before it can support requirements reflection and modelling requirements as runtime entities. The requirements reflection mechanism requires support at the modelling level and architecture level. Requirements reflection can be used to anticipate the evolution of both functional and non-functional requirements.

REFERENCES