Model-to-Model based Approach for Software Component Allocation in Embedded Systems

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Abstract: Due to the popularity and heterogeneity of embedded systems, the problem of software component (SW-component) allocation in such systems is receiving increasing attention. Addressing this problem using a graphical modeling language such as Ecore will enable system designers to better and more easily allocate their components. However, the existing Ecore models do not address the problem of SW-component allocation in heterogeneous embedded systems. Because of Ecore informal semantics, Ecore models cannot be analyzed using mathematical tools. On the other hand, an approach based on colored Petri nets (CPNs) was proposed for the modeling and analysis of the software component allocation problem. The approach was shown to be applicable in the field not only with respect to the cost optimization problem, but also because it takes nonfunctional requirements into consideration. In this paper, we propose an approach for the automated transformation of an Ecore model into an equivalent CPN model, which will help the modeler use the power of a formal modeling language by only modeling the system using a simple Ecore-based modeling language.

1 INTRODUCTION

Model Driven Engineering (MDE) uses models as a cornerstone throughout the software lifecycle, which indicates a shift from everything is an object paradigm to everything is a model paradigm (Wimmer et al., 2011). In this context, choosing a good model will lead to a better system. Ecore is a powerful, expressive and rich language used to define metamodels conformed to EMF (Eclipse Modeling Framework)\(^1\). The semantics of Ecore is not formally expressed. Thus models defined in Ecore are difficult to analyze. On the other hand, CPN (Petri and Reisig, 2008) is a mathematical modeling language offering detailed view of the modeled system with a graphical representation. Also, models by CPN can be analyzed using powerful tools (such as CPN Tools\(^2\)). These properties have made CPN popular in modeling, analyzing, and optimizing complex systems.

SW-component allocation is the process of mapping SW-component to the computational units. Recently, many existing research studies focus on the modeling of these components and verifying these models in an effective way. The SW-component allocation has not been modeled neither in Ecore (Biermann et al., 2008) nor in UML (Force, 2010).

As known, model transformation is a key process in MDE. For that, in this research we aim to introduce an automated transformation technique to automatically generate a CPN model that addresses the SW-component allocation problem specified in a given Ecore model in order to help the modeler access the full benefits of the formal modeling language without necessarily creating the CPN model. The contributions of the paper are summarized as follows:

1. We create a new Ecore model for the SW-component allocation.
2. We automatically transform the Ecore model into an equivalent CPN's model
3. We automatically re-write the CPN model using a Java parser program that obtains the output from ATL and transforms it into the required CPN Tools format.

The organization of the paper is as follows. In Section 2, we define the preliminaries used in our paper. We illustrate our approach in Section 3 and evaluate it in Section 4. The related work is discussed in Section

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\(^{1}\)http://wiki.eclipse.org/EMF
\(^{2}\)http://cpntools.org/
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5. Section 6 concludes the paper and outlines future work.

2 PRELIMINARIES

2.1 CPN

Colored Petri nets (CPN) are an extension of the concept known as Petri nets (Petri and Reisig, 2008). CPNs sustain the useful properties of Petri nets and extend them with initial formalism to allow differences between tokens (Jensen, 2013). CPN is a formal, general-purpose modeling language. In CPNs, tokens are stored in places that carry data (or colors) of a given type. CPN is a discrete-event modeling language combining the capabilities of Petri nets with those of a high-level programming language. A system model in CPN contains both states and action-oriented behaviors. The CPN model describes the states of a system and the events (transitions) that can cause the system to change its state (Jensen and Kristensen, 2009).

2.2 Software Component Allocation for Embedded Systems

An embedded system is a computer system and its related software built into a piece of equipment (Carlson et al., 2010). Recently, significant effort has been done on developing and maintaining embedded systems, including SW-component allocation. SW-component allocation is the process of mapping SW-component to the computational units. Consider a software system (ˇSvogor et al., 2014a) consisting of component to the computational units. Each component must be assigned to a computational unit on a hardware platform consisting of computational units. The computational units offer a number of resources \( l \). The component resource consumption matrix

\[
T = [t_{ijk}]_{(n \times m \times l)}
\]

defines the resources required by each component.

The computational unit resource capacity matrix \( R \)

\[
R = [r_{jk}]_{(m \times l)}
\]

defines the resources that each computational unit can provide. The component allocation problem seeks to find an allocation \( \{p_1, \ldots, p_n\} \), where component \( i \) is allocated to computational unit \( p_i \), which is both feasible and optimal. The feasibility condition can be stated as follows: Given an allocation \( \{p_1, \ldots, p_n\} \), for all computational units \( j \):

\[
\sum_{i,p=j} t_{ip,k} \leq r_{jk} \quad (1)
\]

for all resources \( k \).

Another set of constraints must be satisfied by a feasible allocation. These constraints are system architectural constraints. In a feasible allocation, the architect may require that a particular component should (or should not) be allocated to a particular computational unit.

Given an allocation \( \{p_1, \ldots, p_n\} \), its cost can be computed using the following cost function:

\[
w = \sum_{k=1}^{l} \sum_{i=1}^{n} f_k t_{ip,k} \quad (2)
\]

where \( f_k \) represents a tradeoff factor whose purpose is to specify the weight of each resource. The allocation with the smallest \( w \) (greater than 0) is an optimal allocation.

Therefore, to solve the component allocation problem, we must find an allocation that satisfies (1) and achieve the smallest cost (greater than 0) where cost is defined by (2).

We use the same system in (Al-Azzoni, 2015) to demonstrate a sample component allocation problem. The system is taken from (ˇSvogor et al., 2014b).

The system consists of the following:

- \( n = 11 \) components, as follows:
  1. UI User Interface
  2. CH Communication Handler
  3. MP Message Parser
  4. MD Manual Drive
  5. MM Mission Manager
  6. MC Movement Control
  7. V Vision
  8. AC Actuator Control
  9. SI Sensors Layer 1
  10. S2 Sensors Layer 2
  11. SF Stream Filtering

- \( m = 4 \) computational units, as follows:
  1. mCPU Multicore CPU
  2. FPGA FPGA I
  3. FPGA FPGA II
  4. GPU

- \( l = 3 \) resources, which are as follows:
  1. CPU : average execution time
  2. Memory
  3. Power : average energy consumption

The resource capacity for each computational unit is given by the following matrix:

\[
R = \begin{bmatrix}
100 & 256 & 50 \\
150 & 640 & 25 \\
150 & 640 & 25 \\
100 & 256 & 15 \\
\end{bmatrix}
\]
The resource consumption for the CPU, memory, and power resources is given by the following metrics:

\[
\begin{array}{cccc}
10 & 90 & 90 & 55 \\
50 & 20 & 20 & 72 \\
30 & 20 & 20 & 72 \\
10 & 40 & 40 & 72 \\
20 & 40 & 40 & 72 \\
20 & 50 & 50 & 55 \\
90 & 20 & 20 & 15 \\
20 & 10 & 10 & 70 \\
20 & 10 & 10 & 70 \\
20 & 15 & 15 & 70 \\
90 & 10 & 10 & 33
\end{array}
\]

\[
\begin{array}{cccc}
48 & 256 & 256 & 128 \\
128 & 256 & 256 & 148 \\
64 & 256 & 256 & 148 \\
48 & 168 & 168 & 148 \\
64 & 168 & 168 & 148 \\
64 & 168 & 168 & 64 \\
168 & 128 & 128 & 64 \\
148 & 96 & 96 & 148 \\
48 & 32 & 32 & 148 \\
48 & 32 & 32 & 148 \\
168 & 64 & 64 & 96
\end{array}
\]

\[\begin{bmatrix}
2 & 18 & 18 & 11 \\
10 & 4 & 4 & 14 \\
6 & 4 & 4 & 14 \\
2 & 8 & 8 & 14 \\
4 & 8 & 8 & 14 \\
4 & 10 & 10 & 11 \\
18 & 4 & 4 & 3 \\
4 & 2 & 2 & 14 \\
4 & 2 & 2 & 14 \\
4 & 3 & 3 & 14 \\
18 & 2 & 2 & 7
\end{bmatrix}
\]

\[\begin{pmatrix}
90 & 10 & 10 & 33 \\
20 & 50 & 50 & 55 \\
10 & 20 & 20 & 15 \\
30 & 20 & 20 & 15 \\
20 & 10 & 10 & 70 \\
20 & 10 & 10 & 70 \\
20 & 15 & 15 & 70 \\
90 & 10 & 10 & 33 \\
50 & 20 & 20 & 15 \\
20 & 50 & 50 & 55 \\
10 & 20 & 20 & 15 \\
30 & 20 & 20 & 15 \\
20 & 10 & 10 & 70 \\
20 & 10 & 10 & 70 \\
20 & 15 & 15 & 70 \\
90 & 10 & 10 & 33
\end{pmatrix}
\]

In this allocation problem, two constraints are defined:

- Component V should be allocated in the computational unit
- Component MD should not be allocated in computational unit
- In the component resource consumption matrix:
  - GPU
  - mCPU
- Finally, the trade-off vector is as follows:
  \[F = (0.1557, 0.0856, 0.7095)\]

### 2.3 CPN Model for The Component Allocation Problem

The CPN model of the SW-component allocation problem presented in Section 2.2 was developed in (Al-Azzoni, 2015) and is shown in Figure 1. It contains six places. Place Components, carries tokens for each component; and Place CompUnits, holds tokens for each computational unit, and each token includes the available resources for the computational unit. The Place ResConsumption holds tokens containing the resource consumption for each component using the matrix \(T\), which defines the amount of resources needed by each component based on the component resource consumption matrix:

\[T = [t_{ijk}]_{(n \times m \times l)}\]

The element \(t_{ijk}\) represents the amount of the \(k\)-th resources required by the \(i\)-th component when allocated to the \(j\)-th computational unit.

Another place in the CPN model is the Allocations place, which also holds tokens that represent the allocation of the components to the computational units. The place NextSend is responsible for controlling which component will be allocated next and is used to reduce the state space by allocating the components in order of their numbers. Finally, the place Cost holds an individual token which records the total cost of the allocation. A single transition is proposed in the CPN model, the name of which is allocate; it corresponds to the authorization of one component to one computational unit. Furthermore, color sets are defined in the CPN model as follows:

- \(\text{colset UNIT = unit;}\)
- \(\text{colset INT = int;}\)
- \(\text{colset REAL = real;}\)
- \(\text{colset BOOL = bool;}\)
- \(\text{colset STRING = string;}\)
- \(\text{colset Component = int;}\)
- \(\text{colset CompUnit = product } \text{INT} \times \text{INT} \times \text{INT} \times \text{INT};\)
- \(\text{colset Allocation = product } \text{INT} \times \text{INT} \times \text{INT} \times \text{INT} \times \text{INT} \times \text{INT};\)
- \(\text{colset ResCons = product } \text{INT} \times \text{INT} \times \text{INT} \times \text{INT} \times \text{INT} \times \text{INT};\)

And the variables are declared as:

\[\begin{align*}
\text{var } & c, cu : \text{INT;} \\
\text{var } & co : \text{REAL;} \\
\text{var } & a-cpu, a-mem, a-pwr : \text{INT;} \\
\text{var } & r-cpu, r-mem, r-pwr : \text{INT;}
\end{align*}\]

Moreover, the proposed approach addresses the user allocation constraints by using the guard of transition allocate.

### 3 SW-COMPONENT ALLOCATION PROBLEM TO CPN TRANSFORMATION

ATL is a transformation language and toolkit that transforms elements between source and target models according to defined rules based on their metamodels. An ATL program includes two metamodels and composes a set of rules that define how a source model element will be matched in the target model. ATL is applied in the context of the transformation architecture shown in Figure 2. In this architecture, a source model ComponentAllocation.xmi conforming to a metamodel ComponentAllocation.ecore is transformed into a target model CPN.xmi. This target model conforms to CPN.ecore according to the transformation definition ComponentAllocation2CPN.atl written in ATL language. The transformation definition is a model conforming to the ATL metamodel with a transformation architecture consisting of three levels. M3 is the metamodel level, and in our
case, is represented by Ecore, the metametamodel for the entire transformation engine and is defined by itself. M2, the metamodel level is represented here by ComponentAllocation.ecore and CPN.ecore, which is defined by M3. M1 is the model level and is defined by M2. Figure 3 describes the component allocation metamodel used in the scope of our transformation. The Component class contains ComponentName, while the CompUnit class contains the CompUnitName and the available resources for each one. The ResConsumption class represents the amount of resources consumed by component X while it is allocated to CompUnit Y. The user can add the ResConsumption to all components individually with any CompUnit. If there is any constraint in the allocation, the user can represent the constraints using AllocationConstraint or AntiAllocationConstraint classes; the TradeOffVector class will indicate if there are any trade-offs in the allocation problem.

Figure 1: CPN model (extracted from (Al-Azzoni, 2015)).

Figure 2: ATL model transformation process.

Figure 4 describes the CPN metamodel used in the scope of our transformation. The CompToken class contains CompTokenName attribute taken from Component class in the ComponentAllocation metamodel, while the CompUnitToken class contains CompUnit attribute that holds CompUnitName for each CompUnit in the ComponentAllocation metamodel. The ResConsToken class contains ResCons attribute for
the amount of resources consumed by a component when its allocated to a computational unit. The component allocation constraints are stored in AllocationConstraint and AntiAllocationConstraint classes. The TradeoffVectorToken class contains a string defining the tradeoff vectors of the allocation problem. The ATL code for the ComponentAllocation2CPN.Atl transformation consists of six rules.

-- First Rule
rule Component2Token {
from
A: ComponentAllocation!Component
to
Tokent1: CPN!CompToken
(CompTokenName <- A.CompName)
}

-- Second Rule
rule CompUnit2Token { 
from
B: ComponentAllocation!Compunit
to
Token2: CPN!CompUnitToken
(CompUnit <- B.CompUnitName+','+B.CPUAvail+','
+B.MemoryAvail+','+B.PowerAvail)
}

-- Third Rule
rule TradeOffVector2Token{ 
from
C: ComponentAllocation!TradeOffVector
to
Token3: CPN!TradeOffVectorToken
(TradeOffVector <- 'co+'+C.CPUVector+'*'
(Real.fromInt r_cpu)+'
+C.MemoryVector+'*(Real.fromInt r_mem)+'
+C.PowerVector+'*(Real.fromInt r_pwr)')

Figure 3: ComponentAllocation Ecore metamodel.

Figure 4: CPN metamodel.
One of the main challenges in this research was how to make this transformation as powerful as possible while retaining the full benefits of the CPN Tools in the SW-component allocation problem. All file formats in the CPN Tools are XML based. For that, the CPN Tools can not load any file unless it is XML based. This issue was a challenge for our transformation tool because the ATL transformation generates xmi files. The format of the XML files required by CPN Tools is described in DTD with multiple restrictions provided in 3. This file format is different from the ATL's generated file format. We overcame this obstacle by developing a Java parser program that obtains the output from ATL and writes it in the required CPN Tools format. The supported tool reads xmi files and obtains data from specific elements to add characters and generate a string that is written into an output XML file for use by the CPN Tools in creating a model. In order to achieve this goal, we decided to use an XML parser, which provided a way to access and modify data in an XML document. Because xmi is special form of XML, we can use the XML parser in Java without any complications.

4 EVALUATION

In order to ensure the correctness of our transformation approach we evaluated it using static and dynamic analysis.

4.1 Static Analysis - anATLyzer

A transformation rule is written based on its source and target metamodels, specifically to match between elements from the source model and generate its corresponding elements in the target model. It describes the structure of models manipulated by the transformation including the allowed types, relations, features and its constraints. anATLyzer (Sanchez Cuadrado et al., 2014) is a tool integrated with ATL as an Eclipse plug-in which analyzes the transformation to discover its potential problems and suggest many ways to fix them. anATLyzer can identify problems by combining static analysis and constraints solving. Static analysis detects statements of the transformation that contain errors or may cause potential problems when executing the transformation. If the error can not be detected in the transformation statement, the constraints solver will find an input model that cause the error occur to be better understood it by the developer. We integrate anATLyzer with our ATL tool

3http://cpntools.org/cpn2000/sgml_vs_xml
and use it to analyze our transformation rules. Our rules passed the rule conflict analysis. The only warning returned by the anATLyzer tool is that our components in the target model are not connected. This is fine since we declared each class to store separate tokens in order to reuse these tokens in the XMI parser.

4.2 Dynamic Analysis

The authors of (Selim et al., 2012) formulate the model transformation testing into four phases. The first phase is test case generation which involves generating test models conforming to the source metamodel using some criteria to ensure that each test model covers the source metamodel. The second phase is assessing the generated test suite. The third phase is building the oracle function which is used to compare the actual transformation output with the expected output to evaluate the correctness of the transformation. The fourth phase involves running the generated test models and comparing the output with the oracle function results.

4.2.1 Test Model Generation - Black-Box Test Model Generation Based on Metamodel Coverage

The input scope for each test model is based on the source metamodel. For that, each parameter in the source metamodel is an input for the transformation. The source metamodel used in this transformation is an Ecore class diagram. For this purpose, we reused adequacy criteria for UML class diagrams to ensure the coverage for the test models. We used representative values for the Association-End Multiplicity (AEM) and the Class Attribute (CA) since the possible values of multiplicities and attributes are infinite. We used default partitioning to extract values for each attribute taking into consideration the structure or type of the data. For a string attribute this would be null, “something”, and for an [0..1] association end it would be 0, 1. This policy suggests choosing boundary values and in addition values outside the boundaries if the goal is to check the robustness of the transformation.

4.2.2 Building the Oracle Function

In this phase, after generating the test models an oracle function is used to compare the actual output with the expected output to validate the transformation correctness. The oracle function used to test our approach was developed and presented in (Al-Dakheel, L and Al-Azzoni, I, 2016). It is implemented as a Java program that takes a component allocation problem and generates an optimal allocation. It implements full state enumeration that evaluates the cost for all allocations and chooses an allocation that has the minimum cost.

4.2.3 Running the Transformation on the Generated Test Models

Eight test models were used in our ATL tool. We first created an instance for each model, then we loaded the output into the XMI parser, which in turn converted the output to an XML format accepted by the CPN Tools. The goal was to enable the use of CPN Tools by generating the required XML format.

4.2.4 Results

We achieved the results shown in Table 1 after loading the XMI parser output to the CPN Tools as a net and generating the state space to obtain an optimal allocation. The table shows the optimal allocation cost for using the oracle function and using our model-based approach. This shows perfect accuracy of our approach.

5 RELATED WORK

In the field of Model-to-Model Transformation (MMT), the authors of (André et al., 2014) proposed and implemented an approach to translate a UML state machine diagram (SMD), into CPN in order to use its formalization in analysis and verification. The authors of (Hei et al., 2011) introduced an automated transformation tool that transforms UML state charts into Petri nets via extensible markup language (XML) to analyze and verify the original UML model and check its correctness in order to ensure safety analysis for systems using the transformation rules they defined in an XML file.

The authors of (Choppy et al., 2011) proposed a technique for the automated translation of UML state diagrams into CPNs. The CPN model generated from their approach was used in a CPN-AMI Petri net-based case environment to check the correctness of the Petri net model. Furthermore, the authors of (Guo et al., 2014) discussed the absence of models in the field of embedded systems and the limitations of semi-formal models such as UML in addressing this problem. To address these limitations, they proposed an automated technique for the transformation of the UML state machine model into the Simulink Stateflow model, because Simulink is a formal semantic modeling language that uses ATL as model-transformation architecture.
Table 1: Results of the oracle function VS. our approach. TM stands for Test Model.

<table>
<thead>
<tr>
<th></th>
<th>TM1</th>
<th>TM2</th>
<th>TM3</th>
<th>TM4</th>
<th>TM5</th>
<th>TM6</th>
<th>TM7</th>
<th>TM8</th>
</tr>
</thead>
<tbody>
<tr>
<td>Oracle Function Results</td>
<td>176.61</td>
<td>159.77</td>
<td>No feasible allocation</td>
<td>186.16</td>
<td>196.31</td>
<td>208.47</td>
<td>184.77</td>
<td>148.87</td>
</tr>
<tr>
<td>Our Approach Results</td>
<td>176.61</td>
<td>159.77</td>
<td>No feasible allocation</td>
<td>186.16</td>
<td>196.31</td>
<td>208.47</td>
<td>184.77</td>
<td>148.87</td>
</tr>
</tbody>
</table>

* TM = Test Model

In order to identify the optimal SW-component allocation, the authors of (Švogor et al., 2014a) applied a genetic algorithm (GA) to identify optimal solutions to the component allocation problem; they also applied an analytical hierarchical process to address the problem of different measurement units in the calculation of trade-off factors.

However, although GA usually finds a good solution, there is no guarantee that these solutions will be optimal. Another method for solving the component allocation problem was presented in (Wang et al., 2004). The method uses branch-and-bound and forward checking mechanisms. The method was implemented in the Automatic Integration of Reusable Embedded Software (AIRS) toolkit 4.

Obviously, there is still a clear gap on the modeling language that addressed the problem of SW-component allocation on embedded systems. The author of (Al-Azzoni, 2015) was the first to present a model-based approach for modeling and solving the SW-component allocation problem using CPN. He used CPN as a modeling language and described the use of CPN Tools in analyzing the CPN model and solving the component problem. Using CPN to address the SW-component allocation problem will not only optimize the cost function, but will also optimize allocation for other types of nonfunctional analysis, including security and dependability analysis.

6 CONCLUSION AND FUTURE WORK

In this paper, we have described an automated model-to-model transformation approach for software component allocation in embedded systems, which was developed to benefit from the mathematical power of CPN. We have demonstrated full accuracy in our results regarding finding the optimal cost for the problem of SW-component allocation in embedded systems using only a simple Ecore diagram followed by its transformation into the corresponding CPN model. We concluded that the MMT tools have deservedly received a great deal of attention from researchers because they lead to satisfactory results. However, we investigated the existing work related to identifying the optimal SW-component allocation and determined there is a clear gap in the modeling languages and model transformations as they pertain to the SW-component allocation problem in heterogeneous embedded systems. This study is the first to close the gap in this field by identifying a new model that not only addresses the problem of SW-component allocation but transforms it into a formal and analyzable CPN model automatically. We accomplished our transformation in two primary steps. The first was transforming the input model into CPN tokens in an XMI file followed by rewriting the tokens as a CPN model in an XML file using a supported XMI parser.

As a future work, several research activities open directly from the contributions of this paper. First, we plan on combining our ATL transformation technique and the XMI parser into a stand-alone application. To do so, we need to first run the ATL transformation into a programming environment and then combine it with the XMI parser. Second, although we obtained full accuracy in our test results, more test models need to be conducted with a variety of problem sizes in order to further validate our transformation approach.

REFERENCES


4http://kabru.eecs.umich.edu/bin/view/Main/AIRES


