

Design Requirements Iterative Process (DRIP) Tool Demonstration

Concurrent Engineering of Design, Requirements and Knowledge

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Abstract

Requirements definition and design decisions are highly coupled for mechatronic systems, and heavily influenced by prior knowledge. Upfront engineering of requirements and design is often addressed by inefficient, ad-hoc iterative methods. We propose a methodology to perform concurrent engineering of high level requirements and design along with prior knowledge by using a “common constraint framework” to describe the requirements, design and knowledge precisely. Then an upfront symbolic simultaneous analysis of all the constraints allows us to identify infeasibilities.

Next we define a design architecture that can be used to extend the above constraint framework to include temporal aspects leading to an ability to define low level requirements and test scenarios under which these requirements can be verified. Importantly the low level requirements and test scenarios can be specified independent of the final implementation. These provide the critical link between upfront requirements engineering process and downstream implementation verification.

Finally we define “mappings” between implementations and the design architecture that allows definitions of executable tests in the implementation environment that in turn can be used to verify the low level requirements.

We demonstrate the above methodologies using the tool DRIP.

General Terms Algorithms, Management, Documentation, Performance, Design, Reliability, Experimentation, Languages, Verification.

Keywords *concurrent engineering of requirements and design with knowledge; iterative engineering of requirements and design with knowledge; constraint analysis framework; infeasibility analysis; architecture; simulation based verification; Domain Specific Language; Projectional Editor; Language Work Bench;*

1. Introduction

Cyber-physical systems (cyber systems coupled with physical systems) have become ubiquitous, spanning many fields such as auto-

otive, aerospace, medical devices, and consumer devices. Consequently, poor performance or failure in these systems can have very large impact on the society. However, simultaneously, the complexity and size of these systems have also been growing exponentially, making analysis and verification of such systems a huge challenge. The social and economic costs of discovering faulty behaviour of the system increases exponentially through the development process and after start of production. Many development tools and methodologies have evolved to enable discovery of such faults earlier in the development process. These tools have primarily evolved in two categories: (i) Downstream Design and Engineering (DE) tools. Examples include physical system and control system modelling tools, desktop and hardware-in-the-loop simulation tools, calibration tools, etc. and (ii) Upstream Requirements authoring and management (RM) tools. The DE tools typically capture the behaviour of the system in machine-understandable models that are then used to drive the system development. The RM tools are focused on eliciting and documenting information in human-understandable form, and are often not analysable. This leads to a significant gap in the development process during a transition from the upstream requirements and concept definition stage to the design and implementation stage. This fundamental gap between upstream and downstream tool chains is the motivation for the development of the tool DRIP that is discussed here. Similar research initiatives are happening, example [1].

The coupling between requirements and design decisions is much stronger in cyber-physical systems compared to purely software systems. Design decisions drive requirements as much as requirements drive design. Complex cyber-physical systems are never built from scratch; there is a huge knowledge base that becomes the foundation of any new development. Further this knowledge base is evolving driven by emergent requirements and design explorations. This strong linkage between requirements, design and knowledge is not comprehended in today’s tool chains. Instead we rely on ad-hoc processes to iteratively engineer the requirements and design of a mechatronic system, while trying to map it back to a “linear V Process” through traceability linkages. These linkages become incomprehensible and minimally useful to the user as the system goes through multiple iterations. DRIP is developed to address this issue, by providing a framework to concurrently engineer requirements and design in the context of prior knowledge.

In this paper, we will describe some key methodologies that we believe are instrumental in bridging the gap between upstream and downstream engineering processes. We will illustrate an implementation of these methodologies using DRIP.

2. Key Challenges to Bridging Requirements, Design and Knowledge

2.1 Ambiguous Information

While the coupling between requirements and design rationale is relatively well-understood by practicing mechatronics engineers, there are often no systematic methods to address this because of a fundamental challenge: High Level requirements (often, even low level requirements), design rationale and prior knowledge are all documented in Natural Language (NL). NL inherently has ambiguity. This lack of precision leads to multiple challenges:

1. Multiple interpretations (especially when happening over space, time, people) lead to erroneous analysis.
2. Inability to identify inconsistencies in the requirements.
3. Inability to do simultaneous engineering of requirements, design and knowledge. Instead engineers do ad-hoc iterations until reasonable convergence
4. Inability to “hand-off” high level requirements and design rationale to downstream detailed engineering and implementation processes.

2.2 Information Overload and need for Abstraction

Sometimes knowledge is captured in precise ways through “detailed models” or “general purpose programming” environments. Examples include physics based simulatable models, Simulink [6] or UML based control models, excel based production schedules, Finite Element Models, C-code algorithms, etc. But each of these requires a unique set of technical expertise, that they are meaningful only for separate and small sets of engineers. In particular, they become somewhat irrelevant to the “systems engineer” who is trying to perform upstream concurrent engineering of requirements and design in the context of knowledge.

2.3 Implementation dependent definitions of test scenarios and low level requirements

In an effort to do automated regression testing towards higher quality and verification of requirements, low level requirements and test scenarios are developed in the same framework in which the implementations are done. For example, if the controller is in C, tests are developed in C code in the context of the interfaces specified within the C code. Similarly if the controller is in Simulink, tests are developed in the context of the Simulink model. However if the implementation evolves and there are changes in the interface, or if the platform of implementation changes, the tests need to be re-done. This is very time consuming and prone to human error.

3. Proposed Solution Approach

3.1 Common Constraint Framework

Our approach to providing precision is to use a common constraint framework for representing requirements. The fundamental idea of using constraints as a way to abstract information also allows us to have a framework that captures not only requirements, but also the design rationale as well as existing knowledge in a common fashion. For example, consider the simple case of a room heater.

1. 25 deg C <= room temperature
2. power_when_on == 500 W
3. heat_loss = 100 W/deg C * (room_temp – outside_temp)

The first item represents an end user requirement, the second constraint represents the constraints that are a consequence of design decisions made, and the third constraint is a capture of knowledge acquired over time through testing or analysis. But all these constraints have a similar formulation, enabling a combined analysis, or concurrent engineering of requirements, design and knowledge.

The constraints are essentially a logical combination of predicates that are in turn created based on the behaviour of different system variables of interest. Recognizing that such precise constraints evolve in clarity over time, we propose two layers of the constraint specification:

3.1.1 Early Stage Constraints (Invariants):

Initially we consider those constraints that can be defined to be universally true, and treated as invariants. (e.g. vehicle speed shall never be greater than 300 mph; the PRNDL switch cannot be shifted to “R” when the vehicle speed is greater than 10 mph). The form of a predicate would be:

$$\phi_I: f(x, x^D, \theta) \leq 0 \quad (1a)$$

Where x is a real variable vector of size n , x^D is a enumerated vector of size m that can take on a finite set of values and θ is vector of real parameters of size p :

$$\begin{aligned} x &\in X \subseteq \mathbb{R}^n & (1b) \\ x^D &\in \{s_1, s_2, \dots, s_D\} \\ s_i &\in S \subseteq \mathbb{R}^m \forall i \in \{1, 2, \dots, D\} \\ \theta &\in \Theta \subseteq \mathbb{R}^p \end{aligned}$$

$f(\dots)$ is a piecewise continuous function in its arguments. More complex predicate constraints could be constructed using the conjunction (AND), disjunction (OR), negation (NOT) and implication operators.

3.1.2 Downstream Constraints (Temporal Constraints):

As the understanding and design of the system evolves, the constraints can be more nuanced in identifying the temporal nature of the constraints. (e.g. The peak to peak vehicle acceleration shall be less than 1g within 500 ms after the PRNDL switch is shifted to “R”)

In this case, the formulation of the constraint predicates are inspired by Signal Temporal Logic (STL). However, we do not strictly adopt STL, instead make modifications that we feel provides the needed usability and expressivity.

We first extend the basic predicate to be a function of time as below, along with existence and universality qualifiers:

$$\phi_{\forall}: f(x(t), x^{FSM}(t), \theta) \leq 0 \quad (2a)$$

$$\forall t \ni \{t: g(x(t), x^{FSM}(t), \theta, t) \leq 0\}$$

$$\phi_{\exists}: \exists t_0 \in \{t: g(x(t), x^{FSM}(t), \theta, t) \leq 0\} \quad (2b)$$

$$\ni f(x(t_0), x^{FSM}(t_0), \theta) \leq 0$$

Here x is a time-dependent vector, and is in fact associated with the behaviour of a system that is defined in the form of an architecture (defined later below in 3.4). x^{FSM} is a vector state of a finite state machine with vector-length m , also defined in the context of an architecture. t represents time.

As with the invariant constraints, more complex predicate constraints could be constructed using the conjunction (AND), disjunction (OR), negation (NOT) and implication operators.

The above mathematical representation of the constraints, both in the simpler invariant form and in the temporal constraint form, are sufficient to cover a large range of requirements, design and knowledge relevant for mechatronics system development.

Specifically, the invariant constraints are directly used for performing solution infeasibility analysis, and the temporal constraints are used for performing simulation based requirement verification analysis.

3.2 Multi-Notation Representation

While the common constraint framework above is powerful in providing the mechanism for analysis, its biggest drawback comes from the fact that it is too mathematical to be of interest to a large number of practicing engineers – both in terms of understanding as well as usability.

We address this in DRIP explicitly through the use of technologies (see section 4.2 for specific enabling technologies) that facilitate multiple representations of the same object. Every concept within DRIP is captured by an explicit “model” of the concept, which can then be “projected” through multiple editors, providing multiple user-experiences for the end user. In particular, the constraints are concepts that can be projected either as mathematical expressions, or can be projected as text that provides the same meaning, but is more readable. In this paper we refer to this projection as “Controlled Natural Language” (CNL).

3.3 Upfront (In)Feasibility Analysis

A fundamental question that needs to be addressed when the requirements are considered in the context of some design choices and existing prior knowledge is: Is the solution set for the system consisting of the requirements, design and knowledge feasible? This is a question that is difficult to answer precisely at early stages of the system development. However, the following reverse question could be posed: Is the solution set for the system consisting of the requirements, design and knowledge “infeasible”? This is in fact a solvable problem for the case of Invariants in equations (1).

DRIP will perform the mathematical analysis to identify the domains $\{X, S, \Theta\}$ that satisfies equations (1). If $X = \{ \}$, $S = \{ \}$ or $\Theta = \{ \}$ is true, then we declare the problem infeasible.

As we concurrently consider the requirements, design and prior knowledge we avoid having to iterate to find a solution.

An important by-product of such an analysis is the identification of the available design space Θ , as this can be useful in understanding the robustness of the final solution, as well as enhance any downstream optimization performed over the design space.

3.4 Design Architecture, LLRs and Test Scenarios

3.4.1 Design Architecture

In order to meaningfully tie the upstream high level requirements and design engineering efforts into the downstream more detailed design and implementation, we use the concept of the Design Architecture (DA). The DA is intended to be a high level representation of the system, that is independent of the implementation; however it has enough richness to enable precise specification of the low level requirements for the system, as well as the test scenarios that shall be used to verify that the system satisfies the requirements.

The DA consists of the definition of the topology of the system – in terms of the hierarchical decomposition of the system, the interfaces, and the connectivity. The DA can also specify high level state behaviour through finite state machine definitions. The DA can also specify stimulating actions on the architecture.

3.4.2 Low Level Requirement (LLR)

Once a DA is defined, low level requirements can be defined for that DA. To do this, we use the temporal constraints as in equations

(2), with the vectors x and x^{FSM} directly associated with the DA. The use of the constraints enables the LLR to be analysed.

3.4.3 Test Scenario

A Test Scenario (TS) is also defined for a specific DA. A TS is defined as a sequence of steps to be taken by one or more actors, during which the LLRs associated with the DA will be evaluated for conformance. Appropriate methods to define transitions between steps, including going back to previous steps (loops) are provided for within DRIP. Essentially, DRIP has a language to define the TS precisely, and thus the TS can be used in downstream analysis.

3.5 Simulation Based Verification of Implementation

3.5.1 Mapping an Implementation to Design Architecture

An important consideration within DRIP is the idea of an “upstream” definition of the design, independent of the implementation. This translates to definitions of a Design Architecture that is independent of the implementation, and definitions of Requirements and Test Scenarios, all tied to the DA, but independent of the implementation. This is described in section 3.4 above. However, when an implementation needs to be verified against the requirements, we need a mechanism to connect the Design Architecture with the Implementation.

This is done in DRIP through the idea of “Architecture Mapping”. (AM). First, we extract an “Implementation Architecture (IA) from the implementation. This will include critical information such as the interfaces, the stimulating functions, parameters, important internal signals, etc. Then we define a language within DRIP to associate individual elements of the IA to the elements of the DA. This association is the AM.

When there are multiple implementations for a given design architecture, we simply create multiple AMs, one each for every implementation.

3.5.2 Verification Test Definition

Once we have an AM defined, we can define the tests to be performed:

A Test is defined for an Implementation model, for a specific AM. This is then interpreted as executing the Implementation model, for specific test scenarios that are defined for the Design Architecture corresponding to the AM. During such execution, all the requirements defined for the Design Architecture shall be verified for conformance.

Within DRIP a special language is built to describe the Test Definitions.

3.5.3 Simulation Based Verification

The final step in the verification process is actual execution of the Test. DRIP currently provides an interface with Simulink, so implementations in Simulink can be tested automatically from DRIP. A Test Harness is automatically created in Simulink around the original simulation. The Test Harness includes the auto-generated test scenarios and verification criteria.

When the Test Harness is simulated, the requirements are assessed for any violations. The results of the simulation are then reverted back to DRIP for recording and display of successful verification or failure.

In order to do a thorough verification we would need multiple test scenarios, and specialized analysis to understand the coverage of the test scenarios in the context of the degrees of freedom for the

system. However such analysis is beyond the scope of DRIP currently and so will not be discussed here.

4. Enabling Technologies

4.1 Domain Specific Languages and Language Workbenches

One of the fundamental challenges that we set out to address is the balance between ambiguous information when using NL to express information, and information overload and lack of usability when using detailed modelling or general purpose programming languages.

Our solution hinges on identifying key concepts that are relevant to the systems engineer – concepts that help bridge the upstream requirements/design engineering process and the downstream design and implementation processes. These concepts then need to be precisely defined, and evolved in to a “custom language” that is not only meaningful to the practicing engineer, but also machine readable and analysable by a computer.

While function interfaces and scripting mechanisms provide one way to develop such a language, we decided to formalize this approach leveraging the many advances made in “Domain Specific Languages” [2]. Specifically we have paid attention to making these custom languages not ad-hoc constructs, but a set of constructs that have well defined semantics, type definitions and rules, clear relationships and constraints with respect to other concepts, etc.

These languages essentially allow every aspect of the processes supported by DRIP to be “modelled” explicitly. The concepts are all relatable to each other, and thus enables the paradigm that we call “Extreme Model Based Development” (XMBD) [3]. Such an XMBD approach is critically important as we link information across the development processes. For example, having a common variable definition that is used in the upfront analysis as also downstream as part of the simulation based verification.

The value of a DSL is ultimately defined by the end-user experience when performing operations such as creating, editing, searching, re-factoring, etc. From an implementation development perspective, this can often be very time consuming and expensive, without any unique value-add. However, this effort is dramatically reduced by leveraging “Language Workbenches” which allow the developer to define the language concepts at a high level of abstraction, and auto-generate the user-experience aspects with relatively little effort. The Language Workbench that is leveraged for the development of DRIP is MPS [4].

Well defined DSLs have the advantage that concepts can be cumulatively integrated, enabling significant leverage of prior work. DRIP is also built by leveraging the large volume of work already done developing useful DSLs as part of the mbeddr project [5].

4.2 Projectional Editing based Language Workbench

Another fundamental challenge that we wanted to address was to provide a more user-friendly experience in defining the constraints of the common constraint framework. The precise way of describing the constraints is important from the point of view of analysis – both the upfront infeasibility analysis using symbolic computer algebra, as well as the downstream verification analysis using simulation of the implementations. However, this mathematical notation of the constraints is tailored for mathematicians and control theoretic engineers, and is often unnatural for many practicing engineers. In DRIP we wanted to provide alternate user experiences. Specifically, we provide “Controlled Natural Language” (CNL) expressions for the mathematical constraints, that are similar to NL, however retain the precision of the underlying mathematical constraints. This is achieved by maintaining the constraints

using the structure of the underlying model of the constraints (as an Abstract Syntax Tree), but using different “projections” to provide different user experiences when viewing or editing the constraints. Language Editors that support such multiple projections of a common concept are called “Projectional Editors”, and MPS is a prime example of projectional editing technology.

5. Illustration of methodology with DRIP

We will illustrate the above concepts by applying the tool DRIP on a safety-critical application.

5.1 Collision Warning System (CWS)

A safety critical system is often presumed to be one where the system safety is directly attributable to the performance of the software and the hardware it is controlling. While this is most often the case, with more and more closer interaction with humans, safety critical systems often include human-in-the-loop. To highlight this, we choose the Collision Warning System (CWS) installable on an automobile as the application for illustration. On the surface, the CWS is just a warning system and does not directly impact the vehicle performance and safety. However, depending on how the driver responds there is a definite consequence. If the driver were to internalize the CWS with certain expectations on what it would do, then we have a closed system where the appropriate performance of the CWS becomes important.

A high-level schematic of a CWS is shown in Figure 1. There is an obstacle detection sensor that measures the distance from an object in the path of the car, and depending on the speed of the car and the distance, provides a warning to the driver. The expectation is that the driver will react to such a warning by pressing on the brakes and avoid a collision in time.

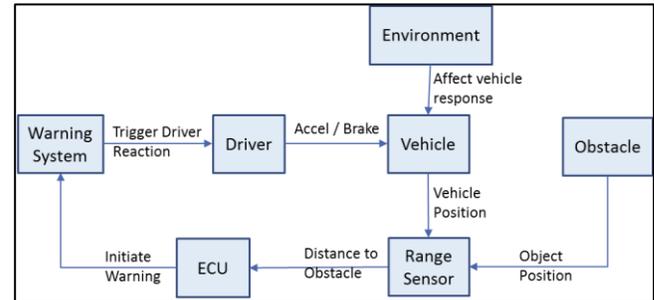


Figure 1: High Level Schematic of the CWS System

5.1.1 (High Level) Requirements for a CWS

1. The CWS should issue warnings enabling collisions being avoided.

This is a key and basic requirement, and can be captured as a constraint:

```

1.1 | Collision Avoidance
    | ERICA /functional: tags
    | created by Admin at 14 Sep, 2016 8:44:12 AM (31 hours ago)
    |
    | In order to avoid a collision, we need the stopping distance to be smaller
    | than the distance at which the obstacle was detected.
    |
    | enforce enforce_0:
    |     stoppingDistafterWarning < obstacleDist
  
```

Figure 2: Basic Requirement of the CWS System

The variables in the above expression are not text. Rather the above expression is a textual projection of an underlying model. This allows this expression to be analysable. We can navigate

to the definitions of the variables, which can be as rich as we deem valuable.

In Figure 3 below, we see that the variables can be defined with units, data type, and a domain that can be continuous or discrete (finite or infinite).

```

CommonDefinitions
doc config: DefaultDocConfig imports:
Variables:
double frictionCoeff
double/m/ detectionRange
double/kph/ brakeStartVehicleSpeed : range 80 kph .. 150 kph
double/m/ obstacleDist
double/m/ stoppingDistafterWarning
double/m/ perceptionReactionDist
double/m/ brakingDist
double sensorCost
double warnSysCost
double T2M
double/s/ driverReactionTime
double/mps2/ decel : range 0 mps2 .. 14.7 mps2

```

Figure 3: Declaration of Variables used in the Requirements, Design and Prior Knowledge

2. Reuse of existing sensors and warning systems.

This is a business requirement that puts constraints on the design choices that can be made. There is also a presumption that there is a knowledge base (KB) where existing sensors and warning systems would be discovered. The KB is discussed in section 5.1.2 below.

The corresponding design constraint is discussed in section 5.1.3 below.

3. Cost Constraint

This is another business requirements that can be explicitly captured as a constraint:

```

1.2 Cost
ERICost /functional: tags
created by swami at 5 Apr, 2016 (14 months ago)

A critical business requirement to make this initiative successful is that the cost of the overall system be kept to acceptable limits. The overall system cost is considered in terms of three major constituents: The cost of the sensing system, the cost of the warning system, and the equivalent cost associated with time-to-market related to any solution.

enforce enforce_0:
T2M * 0.9 + warnSysCost * 1.12 + sensorCost <= 10.0
design explored --- DesignAndEngineering.CWSDesign.DsgnExp

```

Figure 4: Cost constraint for the CWS

5.1.2 Knowledge Base

The Knowledge Base (KB) is simply a compendium of all knowledge relevant for the design-requirements engineering process. Within DRIP knowledge is abstracted in the form of constraints. For example, the knowledge about a sensor might be abstracted and captured mathematically as in Figure 5.

```

1 Sensor DataBase
SensData /abstract systems knowledge: tags
created by Admin at 11 Mar, 2016 (16 months ago)

All known sensors are included in this database

1.1 High Cost High Range Sensor
HCHRSens /abstract systems knowledge: tags
created by Admin at 11 Mar, 2016 (16 months ago)

This is an existing Sensor typically used as a primary sensor in the company's ADAS products, with a detection range of 150m

fact fact_0:
detectionRange = 150 m
fact fact_1:
sensorCost = 5
fact fact_2:
T2M = 1.5

1.2 Low Cost Low Range Sensor
LCLRSens /abstract systems knowledge: tags
created by Admin at 11 Mar, 2016 (16 months ago)

This is an existing Sensor typically used as a secondary sensor in the company's ADAS products, with a detection range of 100m

fact fact_0:
detectionRange = 100 m
fact fact_1:
sensorCost = 2.5
fact fact_2:
T2M = 3

```

Figure 5: Sensors in the Knowledge Base

Along with knowledge about products, we can also institutionalize practical knowledge associated with the products. For example, we could embed knowledge about typical reaction times of human beings for different warning systems in the KB as in Figure 6.

```

2 Warning Systems Database
WSysdB /abstract systems knowledge: tags
created by swami at 5 Apr, 2016 (14 months ago)

A primary characterization of the warning system is the associated reaction time for the driver to respond - specifically from the onset of the warning to the completion of any corrective action by the driver. This is database of known warning systems

2.1 Tactile Warning System
TactileWarnSys /abstract systems knowledge: tags
created by swami at 5 Apr, 2016 (14 months ago)

This is the warning system that uses vibratory motors embedded inside the seats to provide warning to the driver.

enforce enforce_0:
driverReactionTime >= 1.8 s
fact fact_1:
warnSysCost = 5

2.2 Aural Warning System
AuralWarnSys /abstract systems knowledge: tags
created by swami at 5 Apr, 2016 (14 months ago)

This is the warning system that uses the built-in speakers to issue a warning to the driver.

enforce enforce_0:
driverReactionTime >= 2 s
enforce enforce_1:
warnSysCost = 3.3

2.3 Visual Warning System
VisualWarnSys /abstract systems knowledge: tags
created by swami at 5 Apr, 2016 (14 months ago)

This is the warning system that uses the existing display unit to provide the warning.

enforce enforce_0:
driverReactionTime >= 2.2 s
enforce enforce_1:
warnSysCost = 1.1

```

Figure 6: Warning Systems and Driver Reaction Times in the Knowledge Base

Further, constraints arising from laws of Physics are captured inside the Knowledge Base.

```

3 | Vehicle Stopping Performance
  | VehStopPerf /abstract systems knowledge: tags
  | created by swami at 5 Apr, 2016 (14 months ago)

Based on the capabilities of the brake systems in the current production vehicles,
the stopping distance is a function of the speed at the start of the emergency maneuver
and the road conditions. The stopping distance is the sum of perception-reaction distance
and braking distance

fact fact_0:
  stoppingDistAfterWarning = perceptionReactionDist + brakingDist

3.1 | Perception-Reaction Distance
    | PRDist /abstract systems knowledge: tags
    | created by Admin at 14 Sep, 2016 8:52:53 AM (31 hours ago)

This is the distance travelled by the car before the driver realize the warning
and press the brakes

fact fact_0:
  perceptionReactionDist = convert[brakeStartVehicleSpeed -> mps] * driverReactionTime

3.2 | Braking distance
    | BrakeDist /abstract systems knowledge: tags
    | created by Admin at 14 Sep, 2016 8:54:23 AM (32 hours ago)

This is the distance travelled by the car after the brakes are actuated. Also the
deceleration of the vehicle depends on the road-tyre friction

fact fact_0:
  brakingDist = (convert[brakeStartVehicleSpeed -> mps] *
  convert[brakeStartVehicleSpeed -> mps] / (decel * 2))

fact fact_1:
  decel < frictionCoeff * 9.8 mps2
  
```

Figure 7: Physics Based Inferences in Knowledge Base

5.1.3 Initial Design Constraints

The Design constraints are derived from the requirement to reuse sensors and warning systems. This is captured in Figure 8 below:

```

1.2 | Design Exploration
    | DsgnExp /design: tags
    | created by Admin at 5 Apr, 2016 (14 months ago)

The two important subsystems that need to be designed are the Sensor System
and the Warning System. These subsystems also decide the overall cost of the
design. However, there is a need to find the combination of sensor and warning
system that satisfy the cost and functional requirements.

choice for SensData -> TBD
choice for WSysdB -> TBD
  
```

Figure 8: Design Constraint driven by requirement to Reuse

Combining the design constraint with the KB, we can conclude that there are two choices for the sensors (high cost and low cost sensor) and three choices for the warning system (aural, tactile and visual).

5.2 Upfront concurrent engineering of requirements, design and prior knowledge for the CWS

The development process for any system typically starts from a rather high level requirement. However any further elaboration will be preceded by making critical design choices. In this example, we need to start making design choices on the sensing and warning systems. This requires a simultaneous consideration of the requirements, design and prior knowledge simultaneously.

Since we are using a common constraint framework, for every possible set of design choices, DRIP assembles all the constraints together and symbolically analyses them together to identify infeasibility. This analysis leverages a commercial symbolic analysis engine that is embedded inside DRIP.

Figure 9 shows how the constraints from different sources across a project are collected for analysis. Such an analysis is performed for every permutation of possible design choices to upfront identify infeasible design choices. This is seen in Figure 10.

```

Name: Analysis
TestName: Test
Solution for Project CWSProject
Module Selection: All Modules are loaded

Design Choices
Choice for SensData in DsgnExp --> TBD
Choice for WSysdB in DsgnExp --> TBD

Design Parameters:
<< ... >>

Design Solution Variables:
frictionCoeff : there should be a feasible solution for every values in the range [ 0.4 , 0.9 ]
Executed by Admin at 15 Sep, 2016 5:10:04 PM

Equations Used:
enforce on ERICost : stoppingDistAfterWarning < obstacleDist
enforce on ERICost : T2M * 0.9 + warnSysCost * 1.12 + sensorCost <= 10.0
enforce on DistReq : (obstacleDist - stoppingDistAfterWarning) > 3 m
enforce on decelRange : decel < 4.9 mps2
fact on VehStopPerf : stoppingDistAfterWarning = perceptionReactionDist + brakingDist
fact on PRDist : perceptionReactionDist = convert[brakeStartVehicleSpeed -> mps] * driverReactionTime
fact on BrakeDist : brakingDist = (convert[brakeStartVehicleSpeed -> mps] *
  convert[brakeStartVehicleSpeed -> mps] / (decel * 2))
fact on CollAvoid : obstacleDist <= detectionRange
  
```

Figure 9: Collecting Equations for Analysis

```

Name: Analysis
Test
Design Choices
Choice for SensData in DsgnExp --> TBD
Choice for WSysdB in DsgnExp --> TBD

Test_1 : Infeasible
Test_2 : Infeasible
Test_3 : Infeasible
Test_4 : Infeasible

TestName: Test_5
Solution for Project CWSProject
Module Selection: All Modules are loaded

Design Choices
Choice for SensData in DsgnExp --> HCHRSens
Choice for WSysdB in DsgnExp --> VisualWarnSys

Design Parameters:
<< ... >>

Design Solution Variables:
frictionCoeff : there should be a feasible solution for every values in the range [ 0.4 , 0.9 ]
Executed by Admin at 15 Sep, 2016 5:18:19 PM

Equations Used:
enforce on ERICost : stoppingDistAfterWarning < obstacleDist
enforce on ERICost : T2M * 0.9 + warnSysCost * 1.12 + sensorCost <= 10.0
enforce on DistReq : (obstacleDist - stoppingDistAfterWarning) > 3 m
enforce on decelRange : decel < 4.9 mps2
fact on VehStopPerf : stoppingDistAfterWarning = perceptionReactionDist + brakingDist
fact on PRDist : perceptionReactionDist = convert[brakeStartVehicleSpeed -> mps] * driverReactionTime
fact on BrakeDist : brakingDist = (convert[brakeStartVehicleSpeed -> mps] *
  convert[brakeStartVehicleSpeed -> mps] / (decel * 2))
fact on BrakeDist : decel < frictionCoeff * 9.8 mps2
enforce on CollAvoid : obstacleDist <= detectionRange
fact on HCHRSens : detectionRange = 150 m
fact on HCHRSens : sensorCost = 5
fact on HCHRSens : T2M = 1.5
enforce on VisualWarnSys : driverReactionTime >= 2.2 s
enforce on VisualWarnSys : warnSysCost = 1.1

Test_6 : Infeasible
Infeasibility Check : Feasible (Up-to-date)
  
```

Figure 10: Combinatorial Infeasibility Analysis for each of the Design Choices

Once we perform such an infeasibility analysis, we can make specific design choices based on such analysis. When a design decision is made, we have a clear traceability from the design decision to the specific analysis that is the basis for such a decision. (Figure 11)

```

1.2 | Design Exploration
    | DsgnExp /design: tags
    | created by Admin at 5 Apr, 2016 (14 months ago)

The two important subsystems that need to be designed are the Sensor System
and the Warning System. These subsystems also decide the overall cost of the
design. However, there is a need to find the combination of sensor and warning
system that satisfy the cost and functional requirements.

choice for SensData -> HCHRSens : High Cost High Range Sensor
choice for WSysdB -> VisualWarnSys : Visual Warning System
decision decision_2:
  "High Range sensor and Visual warning system are chosen for the CWS"
  based on the set of result Analysis
  
```

Figure 11: Tracing Analysis Results Back to Design Decision

Once specific design choices are made, a more detailed analysis can be performed in order to understand the feasible solution space.

For this example, even though the system is feasible, the domain for feasible values of brakeStartVehicleSpeed is diminished.

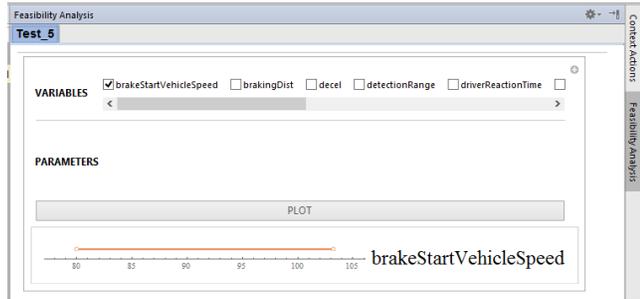


Figure 12: Detailed Analysis - reduced Feasible Domain

With further analysis we can identify the reduced design space. This can be captured through different visualizations, such as in Figure 13 and Figure 14.

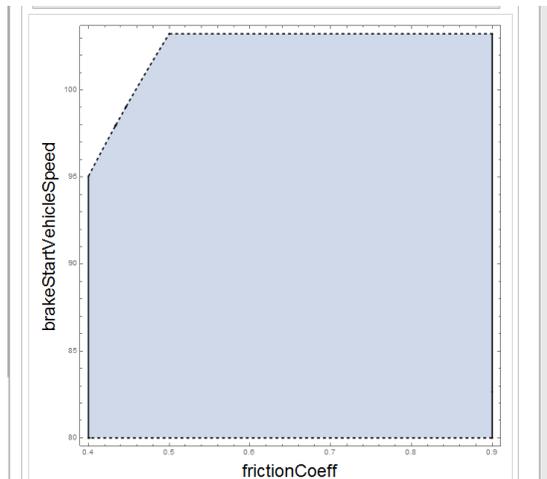


Figure 13: Identifying the feasible Design Space - a

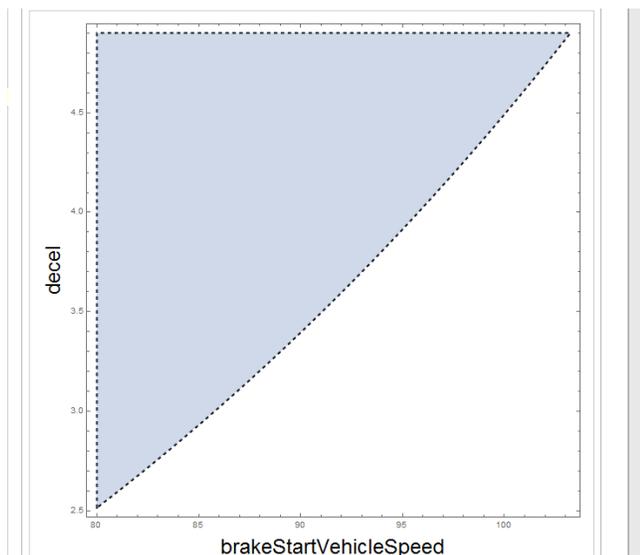


Figure 14: Identifying the feasible design space - b

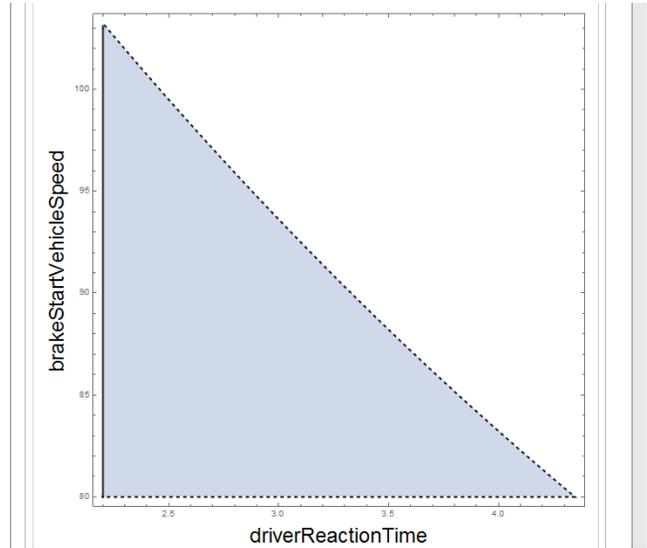


Figure 15: Identifying Feasible Design Space - c

Such analysis can also yield important starting points for control strategy for downstream implementation. For example, we can identify the relation between the vehicle speed, the distance to the obstacle and issuance of warning. Figure 16 shows the acceptable vehicle speeds where a warning can be issued meaningfully for different obstacle or braking distances. The left extreme of the curve is a constraint based on the maximum deceleration, which can be correlated to road and weather conditions. The right extreme of the curve is a constraint coming from the reaction times of the driver and the sensor range.

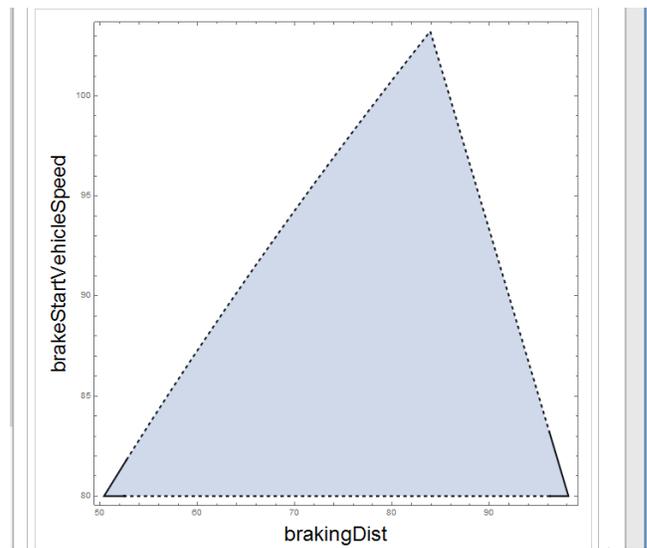


Figure 16: Downstream Control Strategy Guidance

5.3 Implementation-independent specification of system architecture, low-level requirements, and test scenarios

The Design Architecture for the CWS can be relatively simple, as shown in the example below.

```

exported part CWSsystem
  use subjectVehicle as subject
  use obstacle as target

  requires subjectBrake: driverBrake
  requires targetDeceleration: obstacleDeceleration
  requires subjectThrottle: driverThrottle
  requires targetAcceleration: obstacleAcceleration

  provides subjectVehicleSpeed: signal
  provides subjectVehiclePosition: position
  provides targetVehicleSpeed: signal
  provides targetVehiclePosition: position

  provides obstacleDetection: detectObstacleandWarnDriver

  delegate logical subject.brake ↑ subjectBrake
  delegate logical target.brake ↑ targetDeceleration

  delegate logical subject.throttle ↑ subjectThrottle
  delegate logical target.throttle ↑ targetAcceleration

  delegate logical subject.vehiclePosition ↑ subjectVehiclePosition
  delegate logical subject.vehicleSpeed ↑ subjectVehicleSpeed
  delegate logical target.position ↑ targetVehiclePosition
  delegate logical target.speed ↑ targetVehicleSpeed

  delegate logical subject.obstacleDetection ↑ obstacleDetection

  connect logical target.position -> subject.obstaclePosition

```

Figure 17: Top Level CWS Architecture

```

exported part subjectVehicle
  variables
    continuous [state] position = 0 m
    continuous [state] reactionTime = 0.21 s
    continuous [state] velocity = 0 mps
  requires brake: driverBrake
  requires throttle: driverThrottle
  requires obstaclePosition: position

  provides vehiclePosition: position
  provides vehicleSpeed: signal
  provides obstacleDetection: detectObstacleandWarnDriver

  use CWS_controller as controller
  part vehicleDynamics
    requires brake: driverBrake
    requires throttle: driverThrottle
    requires obstacleDetection: detectObstacleandWarnDriver

    provides vehiclePosition: position
    provides vehicleSpeed: signal

  connect logical vehicleDynamics.vehiclePosition -> controller.vehiclePosition
  connect logical vehicleDynamics.vehicleSpeed -> controller.vehicleSpeed
  connect logical controller.warnDriver -> vehicleDynamics.obstacleDetection

  delegate logical vehicleDynamics.brake ↑ brake
  delegate logical vehicleDynamics.throttle ↑ throttle
  delegate logical vehicleDynamics.vehiclePosition ↑ vehiclePosition
  delegate logical vehicleDynamics.vehicleSpeed ↑ vehicleSpeed
  delegate logical controller.obstaclePosition ↑ obstaclePosition

part CWS_controller
  requires vehiclePosition: position
  requires vehicleSpeed: signal
  requires obstaclePosition: position

  provides warnDriver: detectObstacleandWarnDriver

```

Figure 18: Second Level CWS Architecture

Once the architecture is defined, we can now define the requirements in the context of the architecture (lower level requirements). For the CWS, this will translate to stating that under the assumption that the human driver presses the brake firmly within a reasonable time after an obstacle-detected warning is issued, the vehicle with the CWS should not hit an obstacle. We make this precise as below:

```

behaviour for CWSsystem
  REQUIRED [3] ∃t0 ∈ (∅ < obstacleDetection.obstacleDetected >, lowerCondition +
    0.22 s] | ∀t ∈ [t0, Tend], subjectBrake.pressBrake is always true -> [V] ∀t ∈ [0 s,
    Tend], subjectVehiclePosition.val < targetVehiclePosition.val - 2 m is always true

```

Figure 19: Low Level Requirement as a temporal constraint, with mathematical projection

We could also project the requirement in more readable “Controlled Natural Language” format as below:

```

behaviour for CWSsystem
  REQUIRED
    [3] At some point in time after obstacleDetection.obstacleDetected until 0.22 s later,
    subjectBrake.pressBrake shall become and remain true
  IMPLIES
    [V] For all time,
    subjectVehiclePosition.val shall be less than targetVehiclePosition.val - 2 m

```

Figure 20: Low Level Requirements, projected in Controlled Natural Language - English

We can define a typical test scenario that can be used to verify if the requirement above is satisfied by any implementation. For the CWS, one typical scenario would be when the subject car is moving at a certain speed, it discovers an obstacle in its path. The obstacle could be stationary, or moving. The driver will respond to any warning from the CWS, and the driver’s response may or may not be within the expected reaction times.

We want to capture the above scenario in precise terms, so it can be incorporated into an analysis in the context of any implementation to be tested.

We capture this inside DRIP as below:

```

testcase TestCase1 for CWSsystem {
  setup
    set state target.position = 100 m
    set state target.velocity = 10 s
    set state subject.position = 0 m
    set state subject.velocity = 50 s

  actor Subject
    step 1
      start subjectThrottle.pressThrottle
      wait for obstacleDetection.obstacleDetected
    step 2
      wait for stepClock > 0.21
    step 3
      stop subjectThrottle.pressThrottle
      start subjectBrake.pressBrake

  actor Target
    step 1
      start targetAcceleration.do
      wait for globalClock > 10
    step 2
      start targetDeceleration.do
      stop targetAcceleration.do

  assertions
    << assertions >>
}

```

Figure 21: An implementation-independent Test Scenario Definition

5.4 Downstream verification of a CWS implementation against low level requirements

Once the architecture, low level requirements and test scenarios have been defined the process moves downstream to implementation. There are many great tools to support the design and implementation of a CWS. In our example we use Simulink to develop such a system, where both the physical system (consisting of the subject car, the sensors, the driver and the obstacle – which will be a target car) as well as the control strategy are developed using Simulink. The development is driven by the upstream activities. Once such development is completed we would like to assess the performance of the resulting system relative to the original requirements.

In order to do this, we first “map” the implementation to the design architecture. This is done by first extracting an “implementation architecture” of the Simulink implementation, and then associating the subsystems, interfaces and functions of the implementation to the ports and sub-parts of the design architecture. The mapping for one such implementation is shown in Figure 22.

Once such a mapping is defined we are ready to define actual verification “tests”, that relate the low level requirements to be tested, the test scenario to be used on the specific implementation.

These tests can now be automatically executed from DRIP. First, DRIP automatically creates a “test harness” in Simulink that wraps the original implementation with appropriate stimulation to create the test scenarios and appropriate verification criteria to assess the performance of the implementation relative to the requirements. Figure 24 shows such a test harness.

Design Architecture	Implementation Architecture
part CWSsystem	system
use subject	<no simulink>
use target	<no simulink>
requires subjectBrake	subjectBrake
action continuous pressBrake	pressPedal.pressPedal
	Implementation Architecture Design Architecture
	OutPort Command subjectBrake
requires targetDeceleration	targetBrake
action continuous do	pressPedal.pressPedal
	Implementation Architecture Design Architecture
	OutPort Command targetDeceleration
requires subjectThrottle	subjectThrottle
action continuous pressThrottle	pressPedal.pressPedal
	Implementation Architecture Design Architecture
	OutPort Command subjectThrottle
requires targetAcceleration	targetThrottle
action continuous do	pressPedal.pressPedal
	Implementation Architecture Design Architecture
	OutPort Command targetAcceleration
provides subjectVehicleSpeed	subjectVehicleSpeed
signal val	<no mapping>
provides subjectVehiclePosition	subjectVehiclePosition
signal val	<no mapping>
provides targetVehicleSpeed	targetVehicleSpeed
signal val	<no mapping>
provides targetVehiclePosition	targetVehiclePosition
signal val	<no mapping>
provides obstacleDetection	obstacleDetection
action command obstacleDetected	obstacle

Figure 22: Mapping between Design Architecture and an Implementation Architecture in Simulink

```

makes use of Scenario TestCase1 via mapping system_mapping
Variant Selection: TargetVehicleFM -> TargetVehicleCar
Init Script: I:/external/simulink_models/scenario/initialize.m
PostSim Script: I:/external/simulink_models/scenario/assess.m
  
```

Figure 23: Executable Test Definition for an Implementation

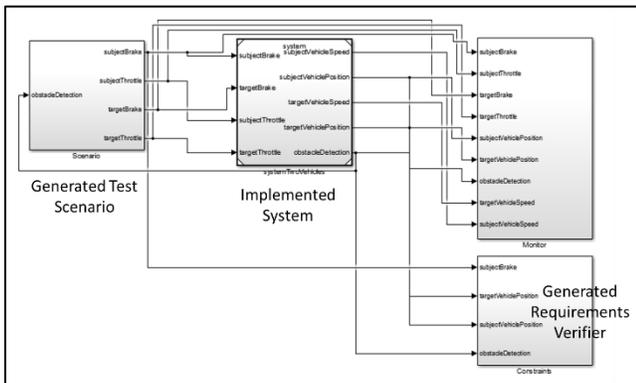


Figure 24: Test Harness Auto-Generated in Simulink

The test harness is simulated in Simulink. The assessment results are then flowed-back to DRIP, where the satisfaction of a requirement is captured as a “success” or “failure” flag, as shown in Figure 25.

```

makes use of Scenario TestCase1 via mapping system_mapping
Variant Selection: TargetVehicleFM -> TargetVehicleCar
Init Script: I:/external/simulink_models/scenario/initialize.m
PostSim Script: I:/external/simulink_models/scenario/assess.m
Constraint_1 : Success
  
```

Figure 25: Verification results from Implementation flow back into DRIP

6. Conclusions

We present a methodology to bridge the gap between upstream requirements and design engineering processes with the downstream design and implementation processes. The methodology involves

1. Definition of a common constraint framework to precisely define requirements, design and prior knowledge, so they can be analysed using computer algebra systems. In particular, use this framework to do upfront infeasibility analysis on solutions, considering requirements, design and prior knowledge simultaneously.
2. Definition of an implementation independent Design Architecture with enough structure to precisely define lower level requirements and test scenarios for the Design Architecture. The low level requirements are defined using an extension of the constraint framework to include temporal aspects of the constraints.
3. Definition of a “Mapping” between the DA and an implementation that needs to be tested. Tests are then defined using the mapping and the associated requirements and test scenarios. These tests are finally used to auto-generate test harnesses for the implementation in a simulation environment to complete the simulation based verification of the implementation against the requirements.

The methodology is illustrated through its implementation in the tool DRIP.

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