

Ensuring Correct Composition of Components using Lattice-based Ontologies

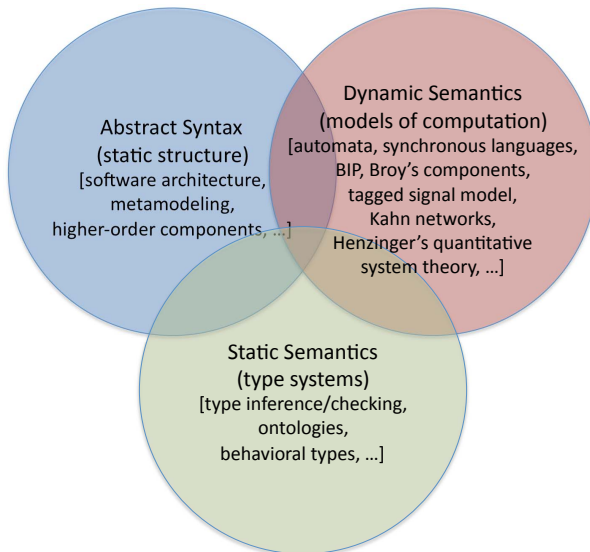
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WFCD - Foundations and Applications of Component-based Design
in Conjunction with ES Week
Grenoble, France, Oct. 11, 2009

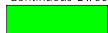
Taxonomy of Modeling Issues



Abstract Syntax: Communicating Hierarchical Components

This model shows a simple adaptive cruise control system, illustrating model-integrated control strategies. A leading car model produces information that is observed with possible flaws by a following car. If the following car detects flaws, it uses a conservative strategy. Otherwise, it tracks the leading car closely.

Continuous Director



- faultStartTime: 50.0
- faultStopTime: 70.0

DriverSimulator



Simulate the driver of the leading car. Output is the driver's desired speed.

Car Model



Simulate a car that matches the desired speed using feedback control with a specified time constant.

CurrentTime



RecordAssembler



PeriodicSampler



Simulate a wireless network that corrupts the data when the fault input is true.

NetworkModel



Simulate a car that attempts to detect faults in communication and adapt its behavior.

FollowingCar



DimensionSystemSolver

Double click to
Resolve Properties

ConstNonconstSolver

Double click to
Resolve Properties

DiscreteClock



Simulate faults.

TimedPlotter



Author: Xiaojun Liu and Edward A. Lee

This example shows a composition of submodels for a cooperative cruise control system.

Static Semantics: Questions to Ask

- Are data types compatible?
- Are units used consistently?
- Are dimensions used consistently?
- Are communication protocols compatible?
- Are components required?
- Are component designs mature?
-

I will describe a framework called **Ptomas** for posing these questions in a domain-specific way and answering them using techniques from type inference.

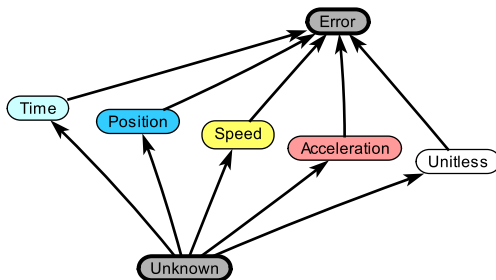
Static Semantics: Questions to Ask

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Dimensions are semantic information associated with the data in the model. A system of such semantic information is called an *ontology*.

Static Semantics: Defining an Ontology



Above is a *lattice* constructed to represent a simple *ontology* in an application domain, such as cruise-control design.

Approach



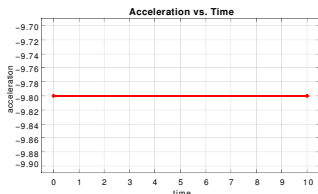
Ptolemy II: Our laboratory for experimenting with modeling, design, and simulation technology

Pthomas: A framework on top of Ptolemy II, allowing analysis of these semantic properties

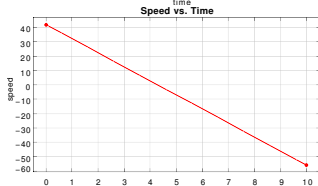
Types of analysis:

- Inference
- Validity Checking

In our Example Ontology, relations given by basic Calculus



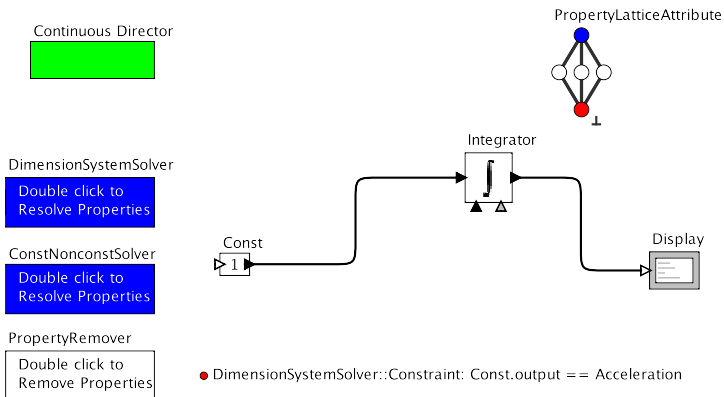
$$\int \text{acceleration}(t) dt = \text{speed}(t)$$



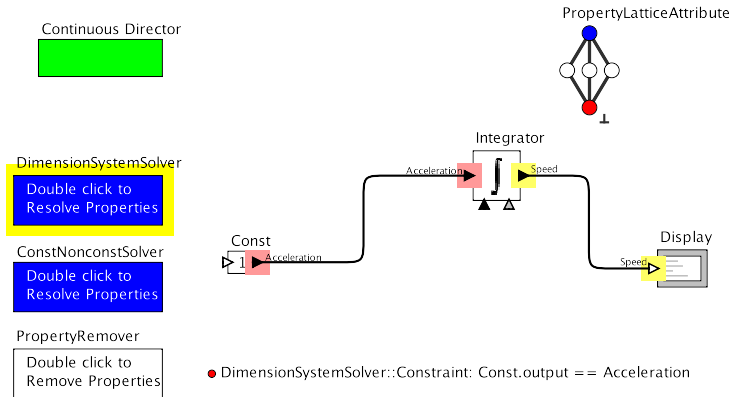
$$\int \text{speed}(t) dt = \text{position}(t)$$



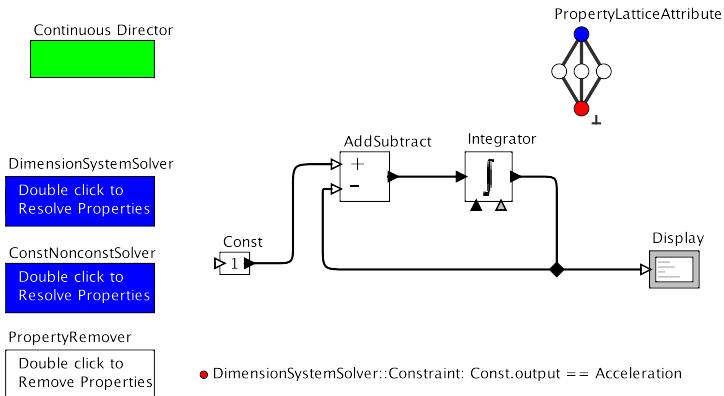
Applying This Ontology to a Model



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Applying This Ontology to a Model

Continuous Director



DimensionSystemSolver

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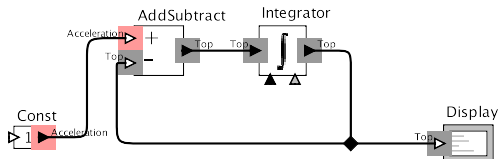
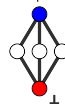
ConstNonconstSolver

Double click to
Resolve Properties

PropertyRemover

Double click to
Remove Properties

PropertyLatticeAttribute



● DimensionSystemSolver::Constraint: Const.output == Acceleration

Applying This Ontology to a Model

Continuous Director



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This model shows a simple adaptive cruise control system, illustrating model-integrated control strategies. A leading car model produces information that is observed with possible flaws by a following car. If the following car detects flaws, it uses a conservative strategy. Otherwise, it tracks the leading car closely.

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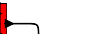


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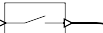
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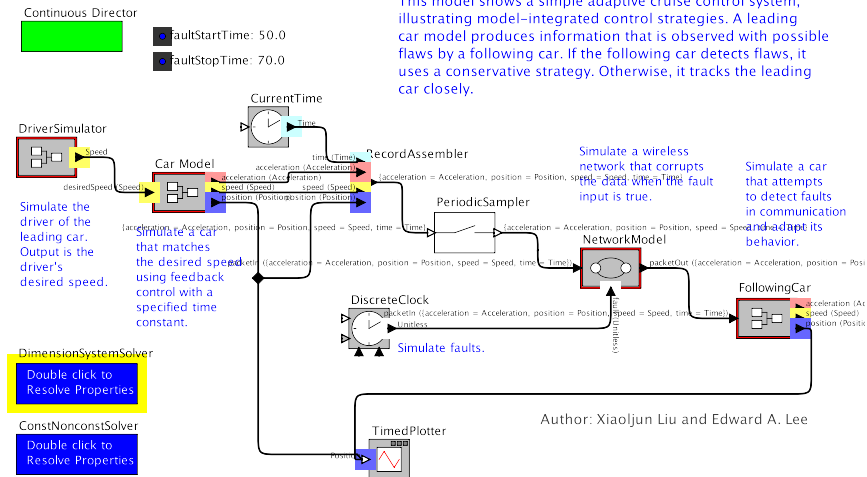
TimedPlotter



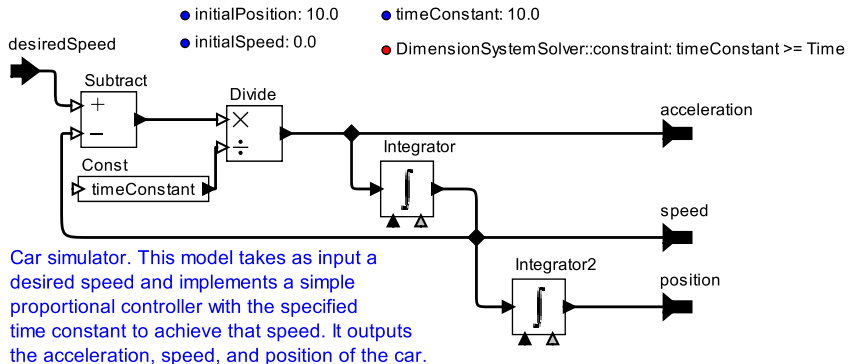
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Applying This Ontology to a Model

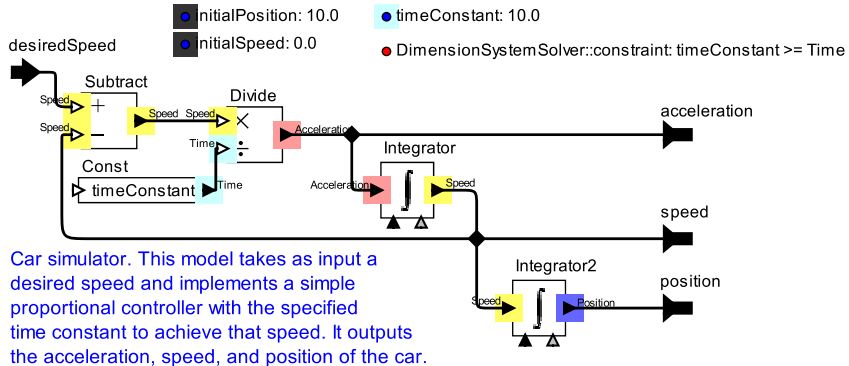
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Applying This Ontology to a Model



Applying This Ontology to a Model



Background

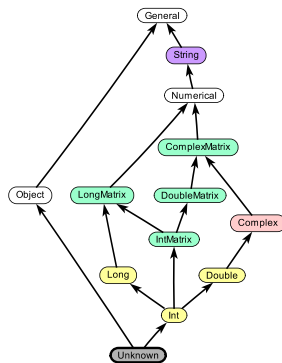
- **Lattice**

A partially ordered set in which every pair of elements has both a least upper bound and a greatest lower bound.

- **Monotonic Function**

A function f for which

$$\vec{x}_1 \leq \vec{x}_2 \implies f(\vec{x}_1) \leq f(\vec{x}_2)$$



Relational Constraint Problem (RCP)

$$RCP : (P, C)$$

P is a partially ordered set, C is a set constraints of the form:

$$r(p_x, p_y, \dots)$$

where r is a relation (e.g. $=, \leq$).

A *solution* is a satisfying assignment to property variables p_x, p_y, \dots .

Definite Monotone Function Problem (DMFP)

Special case of RCP

$$DMFP : (P, C_F)$$

P is a **lattice**, C_F is a set of *definite* inequalities:

$$f(p_y, p_z, \dots) \leq p_x$$

where f is a **monotonic function**.

Here, there is a **unique** *least fixed point* (LFP) solution, efficiently solved by an algorithm given by Rehof and Mogensen (1996).

Problem Statement

Given:

Lattice: P (1)

Constants & Variables: $p_1, p_2, \dots \in P$ (2)

Constraints of the form: $f(p_1, p_2, \dots) \leq p_n$ (f monotonic) (3)

is there a satisfying assignment to variables?

Problem Statement

Given:

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is there a satisfying assignment to variables?

This problem has a linear time algorithm!
(Rehof and Mogensen, 1996)

How to Make this Usable in Practice?

Problem: $|C| \propto |M|$

1 Default Constraints

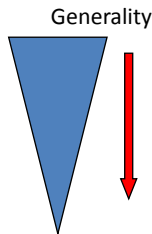
- Set globally by the property solver (actors, connections, etc.)

2 Actor-specific Constraints

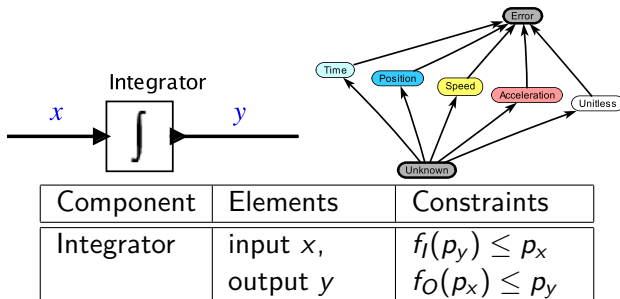
- Uses an *adapter pattern* for actors

3 Instance-specific Constraints

- Specified through model annotations



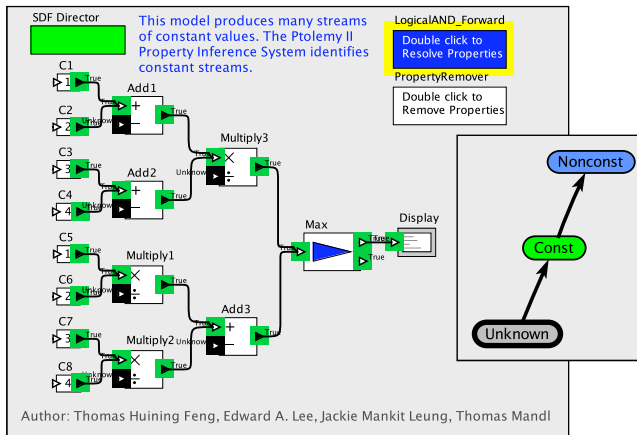
Defining Actor-specific Constraints



$$f_I(p_y) = \begin{cases} \text{Undef.} & \text{if } p_y = \text{Undef.} \\ \text{Speed} & \text{if } p_y = \text{Pos.} \\ \text{Accel.} & \text{if } p_y = \text{Speed} \\ \text{Unitless} & \text{if } p_y = \text{Time} \\ \text{Error} & \text{otherwise} \end{cases}$$

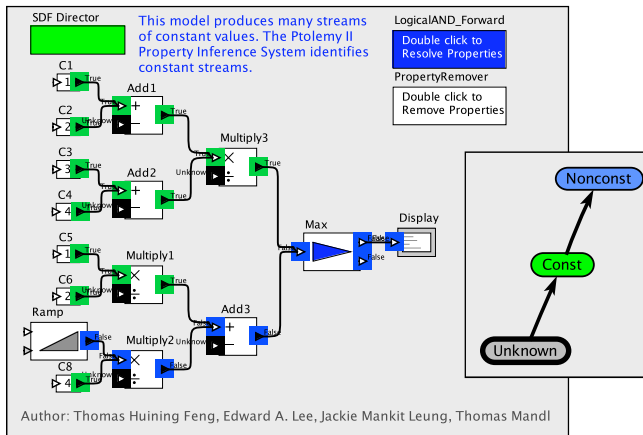
$$f_O(p_x) = \begin{cases} \text{Undef.} & \text{if } p_x = \text{Undef.} \\ \text{Pos.} & \text{if } p_x = \text{Speed} \\ \text{Speed} & \text{if } p_x = \text{Accel.} \\ \text{Time} & \text{if } p_x = \text{Unitless} \\ \text{Error} & \text{otherwise} \end{cases}$$

Another Example



This example illustrates that an ontology can be used to determine in which parts of a model signals vary dynamically.

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Related work

- Constraint Satisfiability (Rehof and Mogensen)
 - Linear time algorithm for the monotone function problem
- Hindley-Milner Type Theory
 - Sound, incomplete static check of programs before run time.
- Web Ontology Language (OWL), Eclipse Modeling Framework (EMF), Object Constraint Language (OCL)
 - Similarities: Ontology frameworks (concepts and relationships)
 - Differences: Expressiveness, Efficiency, Uniqueness of inference

Open Issues

- Usability
 - Defining lattices
 - Giving constraints
- Handling of conflicts
 - Is there a “minimal” set of relaxations to the constraints that makes them satisfiable?
- Extension to infinite lattices
 - Straightforward in theory, but how to make it usable?
- Generality of lattice based ontologies
 - How completely can a unit system be represented?
 - Behavioral types (e.g. using Interface Automata)?
 - ...

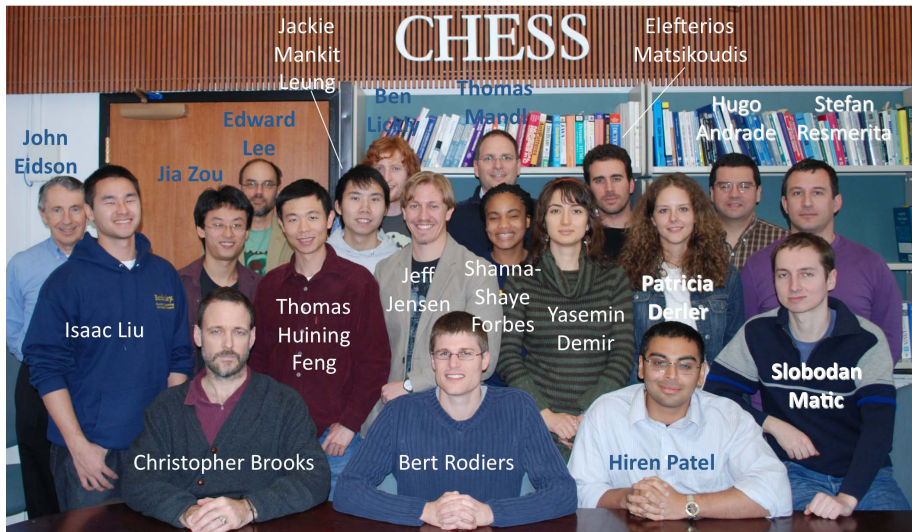
Summary

Our framework (Pthomas) for analyzing model properties:

- ① Customizable for an application domain
(lattice and default constraints)
- ② Requires minimal user specification
(model annotations with specific constraints)
- ③ Infers unspecified properties
- ④ Catches and reports design errors
- ⑤ Scales up efficiently to large models
(leverages Ptolemy II type system implementation by Yuhong Xiong)

The Ptolemy Pteam

See <http://chess.eecs.berkeley.edu/pthomas>.



Algorithm D (Rehof and Mogensen, 1996)

Pseudocode

```
 $C_{var} \leftarrow \{\tau \leq A \in C : A \text{ a variable}\} ;$   
 $C_{const} \leftarrow \{\tau \leq A \in C : A \text{ a constant}\} ;$   
 $p(\beta) = \text{Undef. for all variables } \beta ;$   
while there are unsatisfied constraints in  $C_{var}$  do  
  Let  $\tau \leq \beta$  be one such constraint ;  
   $\beta \leftarrow \beta \vee \tau ;$   
end  
if there are unsatisfied constraints in  $C_{const}$  then  
  Fail: There is no solution ;  
end
```

For a finite lattice, this algorithm takes $O(\text{height}(L) \times |C|)$.

Conflicts

- What is a conflict?
 - **Unsatisfiable** constraints
- How is it detected?

$C_{const} \leftarrow \{\tau \leq A \in C : A \text{ a constant}\} ;$

...

if *there are unsatisfied constraints in C_{const}* **then**

 Fail: There is no solution ;

end