







# Reasoning about Feature Models in Higher-Order Logic

## Mikoláš Janota Joseph Kiniry

Systems Research Group, University College Dublin, Ireland

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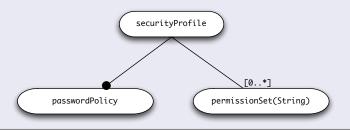
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# Feature Oriented Domain Analysis

#### Feature Models

- capture variability and commonality of a product line
- features represent the building blocks



# Why Formalize?

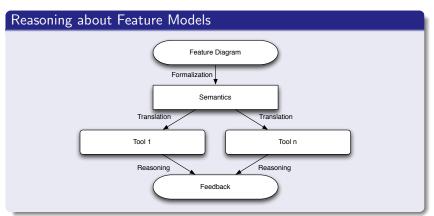
## Disambiguation

- informal explanation of the meaning might be ambiguous
- for example, absolute vs. relative meaning of mandatory

# Why Formalize?

## Disambiguation

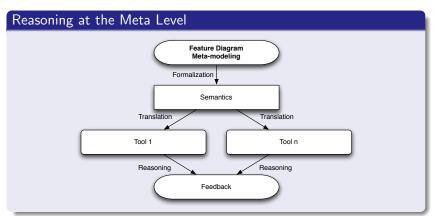
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# Why Formalize?

## Disambiguation

- informal explanation of the meaning might be ambiguous
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# Mechanization of the Formalization

#### **PVS**

- proof assistant widely used in computer science
- typed higher-order logic language

#### Pros and Cons

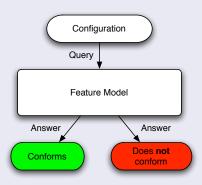
- reason about feature-models that have infinite number of configurations (e.g., feature cloning, attributes)
- express and reason about constraints expressible in HOL
- high level of trustworthiness of the formalization as proofs are checked by a computer
- requires expertise in using a HOL proof-assistant
- some tasks might be tedious



# Concepts

#### Feature Models as Oracles

- the set of selected features and values of their attributes constitute a *configuration*
- a configuration either does or does not conform to the model



#### Features and Attributes

Feature name : String size : Integer

 $\mathsf{Feature} \to \mathcal{P}(\mathsf{Attributeldentifier})$ 

 $\mathsf{Attributeldentifier} \to \mathsf{Type}$ 

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## Feature Configurations

Feature name : String

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Feature
name : String
memoryRequirement : Memory

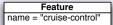
#### Features and Attributes

#### Feature

name : String size : Integer Feature  $\rightarrow \mathcal{P}(\mathsf{Attributeldentifier})$ Attributeldentifier  $\rightarrow \mathsf{Type}$ 

## Feature Configurations

#### Feature name = "air-bag"



Feature
name = "crash-detection"
memoryRequirement = 100MB

• value assignment function assigns values to attributes

 $\mathbb{A} \equiv \mathsf{Feature} \to (\mathsf{AttributeIdentifier} \to \mathsf{AttributeValues})$ 

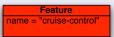
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## Feature Configurations





# Feature name = "crash-detection" memoryRequirement = 100MB

- value assignment function assigns values to attributes
  - $\mathbb{A} \equiv \mathsf{Feature} \to \mathsf{(Attributeldentifier} \to \mathsf{AttributeValues)}$
- selection function determines the selected features

 $select \equiv Feature \rightarrow Boolean$ 



#### Feature Models as Restriction Functions

a *restriction function* determines whether the given feature selection and attributes' values conform to the model

 $\textbf{restr} \equiv \textbf{select} \times \mathbb{A} \to \operatorname{Boolean}$ 

#### Feature Models as Restriction Functions

a restriction function determines whether the given feature selection and attributes' values conform to the model

$$restr \equiv select \times \mathbb{A} \rightarrow Boolean$$

#### **Examples of Restriction Functions**

•  $f_1$  requires  $f_2$ :

$$r_1(s:\mathbf{select},a:\mathbb{A})\equiv s(f_1)\Rightarrow s(f_2)$$

•  $f_2$  requires  $f_3$  with a specific version:

$$r_2(s : \mathbf{select}, a : \mathbb{A}) \equiv s(f_2) \Rightarrow (s(f_3) \land a(f_3)(\text{version}) = 7)$$

restriction functions can be combined:

$$r_3(s : \mathbf{select}, a : \mathbb{A}) \equiv r_1(s, a) \wedge r_2(s, a)$$



#### Feature Models as Restriction Functions

a *restriction function* determines whether the given feature selection and attributes' values conform to the model

 $restr \equiv select \times \mathbb{A} \rightarrow Boolean$ 

#### More Examples in PVS Notation

• a restriction function that corresponds to a requires relation:

```
require(requiree, required: FEATURE) : RESTRICTION =
  LAMBDA (select: SELECT, da: DOMAIN_ASSIGNMENT):
        (select(requiree) IMPLIES select(required))
```

#### Feature Models as Restriction Functions

a *restriction function* determines whether the given feature selection and attributes' values conform to the model

 $restr \equiv select \times \mathbb{A} \rightarrow Boolean$ 

#### More Examples in PVS Notation

- a restriction function that corresponds to a requires relation:
- combine two given restriction functions:

```
intersect(r1, r2: RESTRICTION) : RESTRICTION =
   LAMBDA (select: SELECT, da: DOMAIN_ASSIGNMENT):
   r1(select, da) AND r2(select, da)
```



# Meta-Level Property Example

## Specialization of a Feature Model via Restriction Functions

```
specialization?(restr_1, restr_2 : restr) \equiv \\ \forall s : select; a : \mathbb{A} \bullet restr_1(s, a) \Rightarrow restr_2(s, a)
```

# Meta-Level Property Example

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#### Higher-Order Functions on Restriction Functions

assignment to an attribute value:

```
assign-value(r : restr) ≡ \lambda s : select, a : A \bullet r(s, a) \land (a(f_1)(\text{version}) = 3)
```

# Meta-Level Property Example

## Specialization of a Feature Model via Restriction Functions

$$specialization?(restr_1, restr_2 : restr) \equiv \\ \forall s : select; a : \mathbb{A} \bullet restr_1(s, a) \Rightarrow restr_2(s, a)$$

### Higher-Order Functions on Restriction Functions

assignment to an attribute value:

assign-value(
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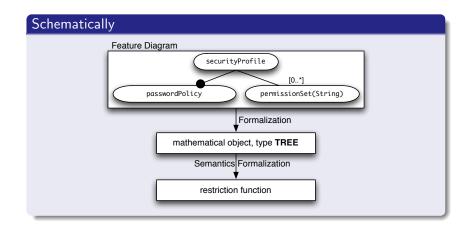
#### Reasoning

the function assign-value returns a specialization:

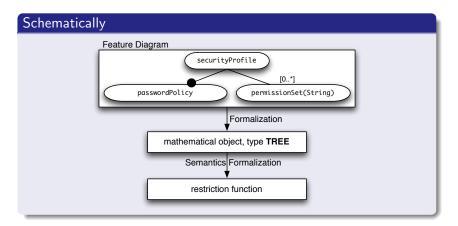
$$\forall r \bullet specialized?(assign-value(r), r)$$



# From Feature Diagrams to Restriction Functions



# From Feature Diagrams to Restriction Functions



## A Function From Diagram to Restriction Function

 $getRestriction : TREE \rightarrow (select \times \mathbb{A} \rightarrow Boolean)$ 



# **Baking Restriction Functions**

## Modeling Gradual Specialization of Restriction Function

• obtain a restriction function, e.g., from a feature diagram

```
r_0 \equiv \text{getRestriction}(tree)
```

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• compose the functions defining each specialization:

$$r_1 \equiv spec_1(r_0)$$
  
 $r_2 \equiv spec_2(r_1)$   
 $\dots$   
 $r_n \equiv spec_n(r_{n-1})$ 

# **Baking Restriction Functions**

## Modeling Gradual Specialization of Restriction Function

• obtain a restriction function, e.g., from a feature diagram

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## Bringing Specializations Together

$$r_n = spec_n(\dots(spec_1(getRestriction(tree)))\dots)$$



# Summary

#### Feature Models as Oracles

- the oracle is an important characteristic of the feature model
- enables unified mathematical approach
  - meta-model level, e.g., what is specialization
  - model level, e.g., record constraints in mathematical notation
- oracles are compositional