







Reasoning about Feature Models in Higher-Order Logic

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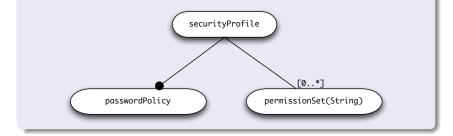
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Feature Oriented Domain Analysis

Feature Models

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



Why Formalize?

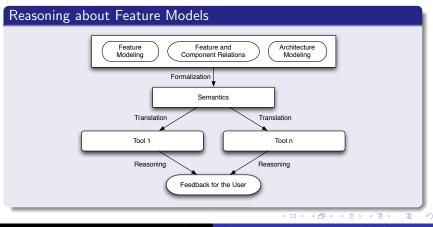
Disambiguation

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

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Disambiguation

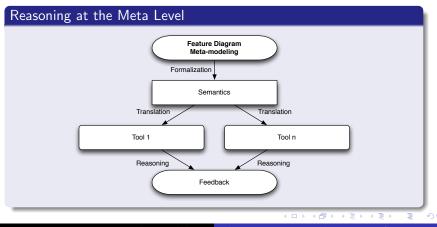
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Mechanization of the Formalization

PVS

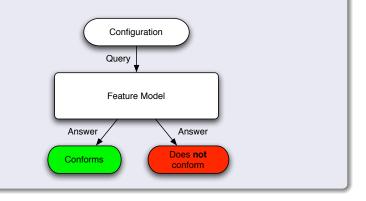
- proof assistant widely used in computer science
- typed higher-order logic language

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-assistantsome tasks might be tedious

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





$$\label{eq:Feature} \begin{split} \text{Feature} & \to \mathcal{P}(\text{AttributeIdentifier}) \\ \text{AttributeIdentifier} \to \mathrm{Type} \end{split}$$



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Feature Configurations



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Feature Configurations



• value assignment function assigns values to attributes

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



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Feature Configurations



• value assignment function assigns values to attributes

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

• selection function determines the selected features

 $\textbf{select} \equiv \textsf{Feature} \rightarrow \operatorname{Boolean}$

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

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

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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:

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

• restriction functions can be combined:

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

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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))

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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))
 combine two given restriction functions:

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

Specialization of a Feature Model via Restriction Functions

specialization?(restr₁, restr₂ : restr) $\equiv \forall s :$ select; $a : A \bullet restr_1(s, a) \Rightarrow restr_2(s, a)$

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Higher-Order Functions on Restriction Functions

assignment to an attribute value:

assign-value(r : restr) $\equiv \lambda s : select, a : A \bullet r(s, a) \land (a(f_1)(version) = 3)$

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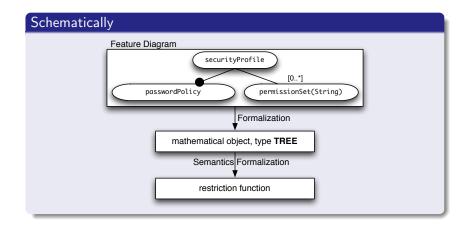
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Reasoning

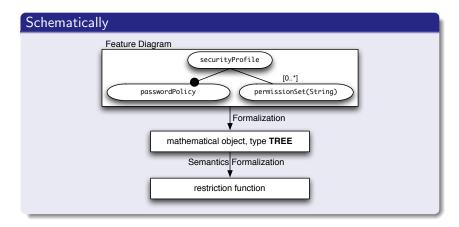
the function assign-value returns a specialization:

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

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)

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Baking Restriction Functions

Modeling Gradual Specialization of Restriction Function

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

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• compose the functions defining each specialization:

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Bringing Specializations Together

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

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