

1 Syntax

1.1 V-AST Syntax

An abstract syntax tree with variability (V-AST) can be represented using the syntax given in (1). A tree vertex (or an object) *obj* can be either a presence condition (*pc*), a language element (*lang*) or a generic node (*node*). All V-AST types have corresponding types in the Java/XTend code as shown in Tbl. 1.

$$\begin{aligned} obj &::= pc \mid lang \mid node \\ node &::= (name, pair) \\ pair &::= \epsilon \mid obj :: pair \end{aligned} \tag{1}$$

Objects. Objects are represented only as identifiers $o_{1?}, o_{2?}$.

Nodes. A *node* is a tuple of two elements: a string *name* and a *pair* which contains the children. It can be represented as an identifier $n_{1\circ}, n_{2\circ}$, as a tuple (Conditional, $p_{()}$) or by using the *name* Conditional_o. The *name* of the *node* can be accessed through the function *name* (e.g. Conditional_o.*name*). The *node* type also has an index accessor for its children (e.g. (Conditional, $p_{()}$)[*i*] is accessing child *i* as ordered in the *pair* $p_{()}$).

Pair. A *pair* can either be empty (ϵ) or it can be formed of a head of type *obj* and a tail of type *pair*. The head and tail can be accessed through their respective functions: $p_{1()}.head$ and $p_{1()}.tail$.

V-AST	Java/XTend
<i>obj</i>	Object
<i>pc</i>	PresenceCondition
<i>lang</i>	Language<CTag>
<i>node</i>	GNode
<i>pair</i>	Pair<Object>

Table 1: Mapping from V-AST vertices to Java classes.

type	representations
<i>obj</i>	$o_{1?}, o_{2?}$
<i>pc</i>	$pc_{1\phi}, pc_{2\phi}, \phi_1, \psi_1, true_{\phi}, false_{\phi}, \phi \wedge false$
<i>lang</i>	$ln_{\@}$
<i>node</i>	$n_{\circ}, (Conditional, p_{()})$, Conditional _o
<i>pair</i>	$p_{0()}$

Table 2: Various representations for vertices.

1.2 Rule Syntax

A rule has a name, an input pattern, an output patten and an algorithm.

REMONERULE		
Input $pair_{()}()$	Output $_{()}()$	Algorithm
$\left[\begin{array}{l} \text{Conditional}_o \\ \left \begin{array}{l} \text{true}_\phi \\ \text{children}_{()}() \\ \text{tail}_{()}() \end{array} \right. \end{array} \right]$	$\left[\begin{array}{l} \text{children}_{()}() \\ \text{tail}_{()}() \end{array} \right]$	<pre> preconditions : pair₍₎ != ε pair₍₎.head : (Conditional, _) pair₍₎.head.filter(pc).size = 1 pair₍₎.head[0] = true_φ do : return pair₍₎.head .getChildrenGuardedBy(true_φ) .append(pair₍₎.tail) </pre>
REMZERORULE		
Input $pair_{()}()$	Output $_{()}()$	Algorithm
$\left[\begin{array}{l} \text{Conditional}_o \\ \left \begin{array}{l} \text{false}_\phi \\ \text{children}_{()}() \\ \text{tail}_{()}() \end{array} \right. \end{array} \right]$	$\text{tail}_{()}()$	<pre> preconditions : pair₍₎ != ε pair₍₎.head : (Conditional, _) pair₍₎.head.filter(pc).size = 1 pair₍₎.head[0] = false_φ do : return pair₍₎.tail </pre>
SPLITCONDITIONALRULE		
Input $pair_{()}()$	Output $_{()}()$	Algorithm
$\left[\begin{array}{l} \text{Conditional}_o \\ \left \begin{array}{l} \phi_1 \\ \text{children}_{1}{}_{()}() \\ \phi_2 \\ \text{children}_{2}{}_{()}() \\ \vdots \\ \phi_n \\ \text{children}_{n}{}_{()}() \\ \text{tail}_{()}() \end{array} \right. \end{array} \right]$	$\left[\begin{array}{l} \text{Conditional}_o \\ \left \begin{array}{l} \phi_1 \\ \text{children}_{1}{}_{()}() \\ \text{Conditional}_o \\ \left \begin{array}{l} \phi_2 \\ \text{children}_{2}{}_{()}() \end{array} \right. \\ \text{Conditional}_o \\ \left \begin{array}{l} \phi_n \\ \text{children}_{n}{}_{()}() \end{array} \right. \\ \text{tail}_{()}() \end{array} \right. \end{array} \right]$	<pre> preconditions : pair₍₎ != ε pair₍₎.head : (Conditional, _) pair₍₎.head.filter(pc).size >= 2 do : newPair₍₎ = ε for (φ_i in [φ₁..φ_n]) newPair₍₎ = newPair₍₎.append((Conditional, φ_i::pair₍₎.head.getChildrenGuardedBy(φ_i))) return pair₍₎.tail </pre>

CONSTRAINNESTEDCONDITIONALSRULE		
Input $node_o$	Output $_{o}$	Algorithm
<pre> ancestors : (Conditional, $\psi_1 :: _()$) (Conditional, $\psi_2 :: _()$) (Conditional, $\psi_n :: _()$) Conditional_o ├ ϕ_1 └ children_o </pre>	<pre> Conditional_o ├ constrain($\phi_1, \psi_1 \wedge \psi_2 \wedge \psi_n$) └ children_o </pre>	<pre> preconditions : node_o.name = Conditional node_o.filter(pc).size = 1 do : simpl_φ = constrain(ϕ, node_o.presenceCondition) if (simpl_φ != ϕ) return (Conditional, simpl_φ :: node_o.toPair.tail) else return node_o </pre>
CONDITIONPUSHDOWNRULE		
Input $pair_{()}()$	Output $_{()}()$	Algorithm
<pre> ├ Conditional_o │ ├── ψ_1 │ └ Conditional_o │ ├── ϕ_{11} │ └ children₁₁() ├ ψ_2 └ Conditional_o ├── ϕ_{21} ├── children₂₁() ├── ϕ_{22} └ children₂₂() ├ Conditional_o │ ├── ϕ_{31} │ └ children₃₁() ├ ψ_n └ Conditional_o ├── ϕ_{n1} └ children_{n1}() └ tail() </pre>	<pre> ├ Conditional_o │ ├── $\phi_{11} \wedge \psi_1$ │ └ children₁₁() ├ Conditional_o │ ├── $\phi_{21} \wedge \psi_2$ │ ├── children₂₁() │ ├── $\phi_{22} \wedge \psi_2$ │ └ children₂₂() ├ Conditional_o │ ├── $\phi_{31} \wedge \psi_2$ │ └ children₃₁() ├ Conditional_o │ ├── $\phi_{n1} \wedge \psi_n$ │ └ children_{n1}() └ tail() </pre>	<pre> preconditions : pair₍₎ != ϵ pair₍₎.head : (Conditional, $_()$) pair₍₎.head.forall [it : $_()$ \vee it : (Conditional, $_()$)] do : pair₍₎.head.filter(cond).map [node_o (Conditional, node_o.map [child_? if (child_? : $_()$) child_φ \wedge pcOf(node_o) else child_?])] . append(tail()) </pre>

MERGESEQUENTIALMUTEXCONDITIONALRULE		
Input $pair_{\circ}$	Output $_{\circ}$	Algorithm
$\left[\begin{array}{l} \text{Conditional}_{\circ} \\ \quad \left[\begin{array}{l} \phi_1 \\ \quad children_{1\circ} \end{array} \right] \\ \text{Conditional}_{\circ} \\ \quad \left[\begin{array}{l} \phi_2 \\ \quad children_{2\circ} \end{array} \right] \\ \quad tail_{\circ} \end{array} \right]$	$\left[\begin{array}{l} \text{Conditional}_{\circ} \\ \quad \left[\begin{array}{l} \phi_1 \vee \phi_2 \\ \quad children_{1\circ} \end{array} \right] \\ \quad tail_{\circ} \end{array} \right]$	<pre> preconditions : pair_∘ != ε pair_∘.size >= 2 pair_∘.head : (Conditional,_) pair_∘.head.filter(cond).size == 1 pair_∘.tail.head : (Conditional,_) pair_∘.tail.head.filter(cond).size == 1 areMutex(φ₁,φ₂) structurallyEquals(children_{1∘},children_{2∘}) do : (Conditional, φ₁ ∨ φ₂ :: children_{1∘}) :: tail_∘ </pre>

1.3 Functions

Constrain / Generalized cofactor. The constrain function (2), a.k.a. the generalized cofactor, constrains a presence condition ϕ by another presence condition ψ that has already been decided to be true and eliminates any redundant variables.

$$\begin{aligned} \text{models}(\phi) &= \{M \mid M \implies \phi\} \\ \text{constrain}(\phi, \psi) &= \phi' \left| \begin{array}{l} \phi' \wedge \psi \implies \phi \\ \text{models}(\phi') = \text{models}(\phi) \setminus \text{models}(\psi) \end{array} \right. \quad \text{and} \quad (2) \end{aligned}$$

For example, $\text{constrain}(A \wedge C, A \wedge B) = C$ shows that, having decided that $A \wedge B$ holds, C is the minimal proposition required to show that $A \wedge C$ also holds: $C \wedge A \wedge B \rightarrow A \wedge C$.

Another example: $\text{constrain}(B \vee C, A \wedge B) = \text{true}$ shows that, having decided that $A \wedge B$ holds, true is the minimal proposition required to show that $B \vee C$ also holds: $\text{true} \wedge A \wedge B \rightarrow B \vee C$.