



Hybrid monitoring of Attacker Knowledge

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Dynamic Information Flow analysis

Assign "high-low (H-L)" security labels to

- values (track confidential information)
- program counter (track side effects under secret control)

Enforcement: halt execution before output of H values.

```
l = 0;  
if h then l = 1;  
return l;
```

No-sensitive-upgrade

Just tracking labels is not enough:

```
int secret s; // s ∈ {0,1}
int public p,q; p := 0; q := 1;
                    s=0          s=1
                    p=0, q=1     p=0, q=1

if s == 0 then
  q := 0;           p=0, q=0     skip
if q == 1 then
  p := 1;           skip        p=1, q=1
                    p=0         p=1
```

No upgrade of security label under secret control

NSU is (too) restrictive

If the upgraded value is not read afterwards

```
l = 0;  
if h then l = 1;  
l = 2;  
return l;
```

If the upgraded value is the same in two branches.

```
if h then l = 1 else l = 1;  
return l;
```

Attacker knowledge

Compute a description of the knowledge that an attacker obtains by observing a given output.

Attacker model:

- "Gadget" attacker P : a program,
- Can see/set initial low values,
- Can observe low output.

Attacker knowledge

The set of initial environments E that lead to a given final output v after execution of P .

Definition of knowledge

Let \downarrow denote the semantics of the language

- $(P,E) \downarrow v$

Let $[E]$ denote the low-equivalent environments of E .

The **knowledge K** of an attacker P that observes v when running P on initial environment E is defined by

$$K(E,v) = \{ E' \in [E] : (P,E') \downarrow v \}.$$

```
if h1 then l = 1;  
if h2 then l = 1;  
return l;
```

Knowledge about E
when seeing 1
with $E(l) = 0$
is
 $E(h1)$ or $E(h2)$ is true.

Non-interference

Characterisation of non-interference in terms of knowledge
(Askarov & Sabelfeld):

A program is (termination-insensitive) non-interferent if attacker learns nothing more by observing a given output v than by just observing non-interference.

Formally, for all initial environments E ,

$$K(E, v) = \{ E' \in [E] \mid (P, E') \downarrow \}.$$

Approximating knowledge

Goal: compute attacker knowledge by program analysis.

Not always computable - have to accept safe approximations.

A safe approximation of attacker knowledge **over-estimates** the knowledge obtained by an attacker.

More knowledge \approx smaller set of initial environments.

Eg. safe to approximate

h1 or h2

by

h1 and h2.

```
if h1 then l = 1;  
if h2 then l = 1;  
return l;
```


Hybrid monitor of knowledge

Compute an estimation of knowledge in each variable.

State: (E, K) where

- E is an environment of values
- K is an environment of knowledge

Hybrid monitor

- $(P, (E, K)) \Downarrow (E', K')$
- executes the program
- computes the knowledge stored in each variable

Knowledge representation

Knowledge domain:

$$\text{Know} : \text{Env} \rightarrow \text{Val} \cup \{\top, \perp\}.$$

Environment of knowledge:

$$K : \text{Var} \rightarrow \text{Know}.$$

Intuition:

$K(x)$ = a function that maps initial env. E to the value of x that E will lead to.

Hybrid monitor

A static and a dynamic part

- dynamic part executes program on concrete values,
- static part analyses code when value are not available.

For conditionals,

- use dynamic part to execute the taken branch,
- use static part to analyse non-executed branch, in order to refine knowledge computation.

Hybrid monitor rules

Static rules are used when no concrete values are available.

$$\text{ASSIGNDYN} \frac{[[e]]_{\rho} = v \quad (e)_{\kappa} = e^{\#}}{(x := e, ([\rho], \kappa)) \Downarrow ([\rho[x \mapsto v]], \kappa[x \mapsto e^{\#}])}$$

$$\text{ASSIGNSTAT} \frac{(e)_{\kappa} = e^{\#}}{(x := e, (\perp, \kappa)) \Downarrow (\perp, \kappa[x \mapsto e^{\#}])}$$

Hybrid monitor rules

Conditionals

$$\text{IFDYN} \frac{C[[e]]_{\rho} = \alpha \quad (c_{\alpha}, (\lfloor \rho \rfloor, \kappa)) \Downarrow (\rho', \kappa_{\alpha}) \quad (c_{\bar{\alpha}}, (\perp, \kappa)) \Downarrow (\perp, \kappa_{\bar{\alpha}})}{(\text{if } e \text{ then } c_{tt} \text{ else } c_{ff}, (\lfloor \rho \rfloor, \kappa)) \Downarrow (\rho', \text{IFF}(C[(e)_{\kappa}], \kappa_{tt}, \kappa_{ff}))}$$

$$\text{IFSTAT} \frac{(c_{tt}, (\perp, \kappa)) \Downarrow (\perp, \kappa_{tt}) \quad (c_{ff}, (\perp, \kappa)) \Downarrow (\perp, \kappa_{ff})}{(\text{if } e \text{ then } c_{tt} \text{ else } c_{ff}, (\perp, \kappa)) \Downarrow (\perp, \text{IFF}(C[(e)_{\kappa}], \kappa_{tt}, \kappa_{ff}))}$$

where

$$\text{IFF}(c, \kappa_1, \kappa_2)(x) = IF(c(x), \kappa_1(x), \kappa_2(x))$$

$$\kappa(x) \quad \lfloor n \rfloor_{\kappa} = \lambda \rho. \lfloor n \rfloor \quad \lfloor e_1 \oplus e_2 \rfloor_{\kappa} = \lfloor e_1 \rfloor_{\kappa} \oplus^{\#} \lfloor e_2 \rfloor_{\kappa}$$

Example

```
1 if h then  
2   z := x + y  
3 else  
4   z := y - x;  
5 output z
```

Program 5

$$\kappa(z) = \lambda\rho. \text{if}(C[[h]]_\rho, [[x + y]]_\rho, [[y - x]]_\rho)$$

Results

The monitor is

- sound wrt the standard semantics,
- transparent (no change of semantics),
- can enforce non-interference,
- can be to other dynamic monitors
 - first use knowledge then use the dynamic monitor
 - track knowledge in shadow variables
 - applied to improve NSU.