# CS 599: Formal Methods in Security and Privacy RHL and probabilistic computations

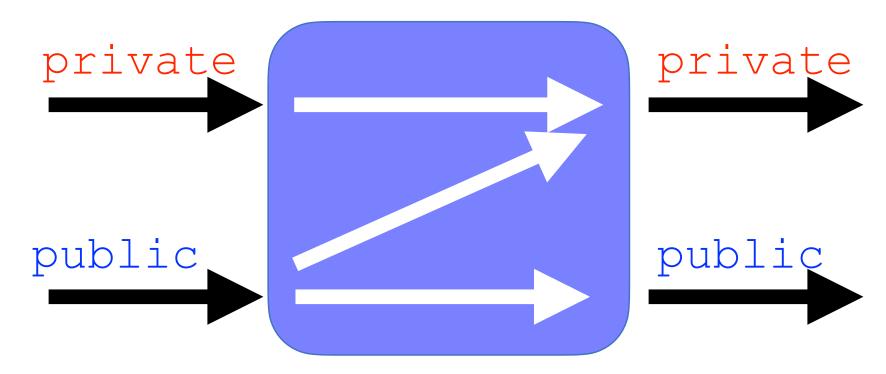
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## From the previous classes

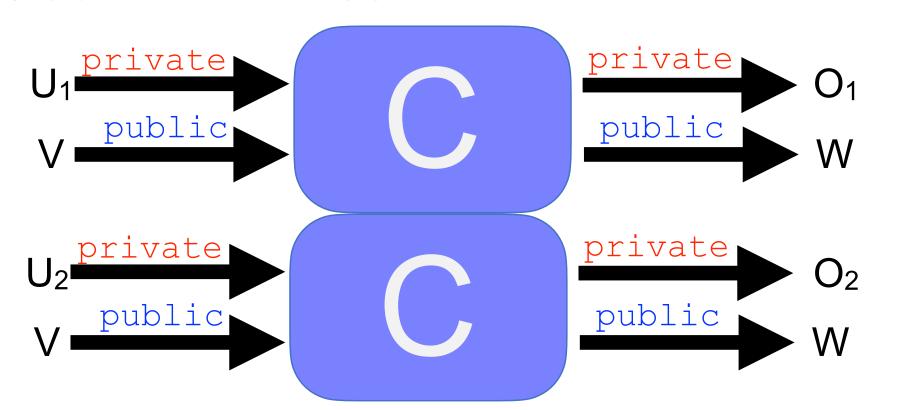
### Information Flow Control

We want to guarantee that confidential information do not flow in what is considered nonconfidential.

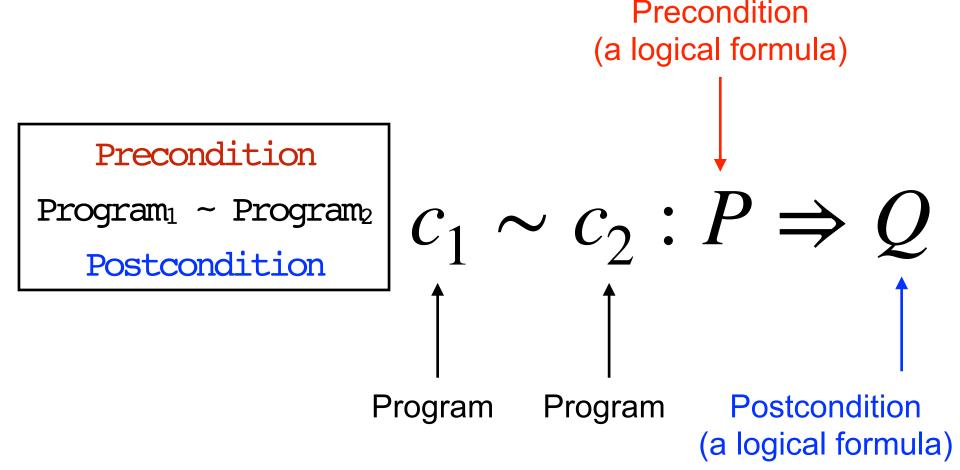


NonInterference
In symbols, c is noninterferent if and only if for every  $m_1 \sim_{low} m_2$ :

- 1)  $\{c\}_{m1} = \bot$  iff  $\{c\}_{m2} = \bot$
- 2)  $\{c\}_{m1}=m_1'$  and  $\{c\}_{m2}=m_2'$  implies  $m_1' \sim_{low} m_2'$



## Relational Hoare Logic - RHL



# More Relational Hoare Logic

## How can we prove this?

```
s1:public
s2:private
r:private
i:public
proc Compare (s1:list[n] bool, s2:list[n] bool)
i := 0;
r := 0;
while i<n do
 if not(s1[i]=s2[i]) then
    r := 1
 i := i + 1
: n>0 /\ =low \Rightarrow =low
```

# Rules of Relational Hoare-Logic One-sided Rules

What do we do if our two programs have different forms? There are three pairs of *one-sided* rules.

One pair for if.

### Assignment — left

```
\vdash x := e \sim skip:
P[e < 1 > / x < 1 >] \rightarrow P
```

### Assignment — right

$$⊢skip ~ x := e :$$

$$P[e<2>/x<2>] ⇒ P$$

Also pair of one-sided rules for while — we'll ignore for now

### Rules of Relational Hoare Logic Program Equivalence Rule

```
\models P: c_1' \equiv c_1
\models P: c_2' \equiv c_2 \qquad \vdash c_1' \sim c_2' : P \Rightarrow Q
\vdash c_1 \sim c_2 : P \Rightarrow Q
```

```
\models P: c_1 \equiv c_2 \text{ means } \{c_1\}_m = \{c_2\}_m for all m such that P (m)
```

# Rules of Relational Hoare Logic Program Equivalences

```
\models P : skip; c \equiv c
```

 $\models P : c; skip = c$ 

$$\models P: (c1; c2); c3 = c1; (c2; c3)$$

. . .

# Rules of Relational Hoare Logic Combining Composition and Equivalence

We can combine the Composition and Program Equivalence Rules to split commands where we like:

```
\vdash c_1; c_2 \sim c_1': P \Rightarrow R
\vdash c_3 \sim c_2'; c_3': R \Rightarrow Q
```

```
\vdash c_1; c_2; c_3 \sim c_1'; c_2'; c_3': P \Rightarrow Q
```

# Rules of Relational Hoare Logic Combining Composition and Equivalence

 $\vdash C_1; C_2 \sim C_1': P \Rightarrow O$ 

# Rules of Relational Hoare Logic Combining Composition and Equivalence

```
\vdash c_1 \sim c_1': P \Rightarrow R
\vdash c_2 \sim \text{skip: } R \Rightarrow Q
```

$$\vdash c_1; c_2 \sim c_1'; skip: P \Rightarrow Q$$

$$\vdash c_1; c_2 \sim c_1' : P \Rightarrow Q$$

## Relational Hoare Logic in EasyCrypt

- EasyCrypt's implementation of Relational Hoare Logic has much in common with its implementation of Hoare Logic.
- Look for the pRHL tactics in Section 3.4 of the EasyCrypt Reference Manual (the "p" stands for "probabilistic", but ignore that for now).

### Soundness

If we can derive  $\vdash c_1 \sim c_2 : P \Rightarrow Q$  through the rules of the logic, then the quadruple  $c_1 \sim c_2 : P \Rightarrow Q$  is valid.

## Validity of Hoare quadruple

We say that the quadruple  $c_1 \sim c_2 : P \rightarrow Q$  is valid if and only if for every pair of memories  $m_1, m_2$  such that  $P(m_1, m_2)$  we have:

```
1) \{c_1\}_{m1} = \bot iff \{c_2\}_{m2} = \bot
```

```
2) \{c_1\}_{m1}=m_1' and \{c_2\}_{m2}=m_2' implies Q(m_1', m_2').
```

How do we check this?

## Relative Completeness

If a quadruple  $c_1 \sim c_2 : P \Rightarrow Q$  is valid, and we have an oracle to derive all the true statements of the form  $P \Rightarrow S$  and of the form  $R \Rightarrow Q$ , then we can derive  $\vdash c_1 \sim c_2 : P \Rightarrow Q$  through the rules of the logic.

# Soundness and completeness with respect to Hoare Logic

Under the assumption that we can partition the memory adequately, and that we have termination.

## Possible projects

#### In Easycrypt

- Look at how to guarantee trace-based noninterference.
- Look at how to guarantee side-channel free noninterference.
- Look at the relations between self-composition and relational logic.

#### Not related to Easycrypt

- Look at type systems for non-interference.
- Look at other methods for relational reasoning
- Look at declassification

## Probabilistic Language

## An example

```
OneTimePad(m : private msg) : public msg
  key :=$ Uniform({0,1}n);
  cipher := msg xor key;
  return cipher
```

Learning a ciphertext does not change any a priori knowledge about the likelihood of messages.

## Probabilistic While (PWhile)

d<sub>1</sub>, d<sub>2</sub>, ... probabilistic expressions

## Probabilistic Expressions

We extend the language with expression describing probability distributions.

$$d::= f(e_1, ..., e_n, d_1, ..., d_k)$$

Where f is a distribution declaration

Some expression examples

```
uniform (\{0,1\}^n) gaussian (k,\sigma) laplace (k,b)
```

## Semantics of Probabilistic Expressions

We would like to define it on the structure:

```
\{f(e_1,...,e_n,d_1,...,d_k)\}_m = \{f\}(\{e_1\}_m,...,\{e_n\}_m,\{d_1\}_m,...,\{d_k\}_m)\}
```

but is the result just a value?

### Probabilistic Subdistributions

A discrete subdistribution over a set A is a function

$$\mu: A \rightarrow [0, 1]$$
  
such that the mass of  $\mu$ ,  
 $|\mu| = \sum_{a \in A} \mu(a)$   
verifies  $|\mu| \le 1$ .

The support of a discrete subdistribution  $\mu$ , supp( $\mu$ ) = {a  $\in$  A |  $\mu$ (a) > 0} is necessarily countable, i.e. finite or countably infinite.

We will denote the set of sub-distributions over A by D(A), and say that  $\mu$  is of type D(A) denoted  $\mu$ :D(A) if  $\mu \in D(A)$ .

### Probabilistic Subdistributions

We call a subdistribution with mass exactly 1, a distribution.

We define the probability of an event E⊆A with respect to the subdistribution µ:D(A) as

$$\mathbb{P}_{\mu}[E] = \sum_{a \in E} \mu(a)$$

### Probabilistic Subdistributions

Let's consider  $\mu \in D(A)$ , and  $E \subseteq A$ , we have the following properties

$$\mathbb{P}_{\mu}[\emptyset] = 0$$

$$\mathbb{P}_{\mu}[A] \le 1$$

$$0 \le \mathbb{P}_{\mu}[E] \le 1$$

$$\mathsf{E} \subseteq \mathsf{F} \subseteq \mathsf{A} \text{ implies } \mathbb{P}_{\mu}[E] \leq \mathbb{P}_{\mu}[F]$$

$$E \subseteq A$$
 and  $F \subseteq A$  implies  $\mathbb{P}_{\mu}[E \cup F] \leq \mathbb{P}_{\mu}[E] + \mathbb{P}_{\mu}[F] - \mathbb{P}_{\mu}[E \cap F]$ 

We will denote by  $\mathbf{O}$  the subdistribution  $\mu$  defined as constant 0.

# Operations over Probabilistic Subdistributions

Let's consider an arbitrary a∈A, we will often use the distribution unit(a) defined as:

$$\mathbb{P}_{\mathsf{unit}(a)}[\{b\}] = \begin{cases} 1 \text{ if a=b} \\ 0 \text{ otherwise} \end{cases}$$

We can think about unit as a function of type unit:  $A \rightarrow D(A)$ 

# Operations over Probabilistic Subdistributions

Let's consider a distribution  $\mu \in D(A)$ , and a function M:A  $\to D(B)$  then we can define their composition by means of an expression let  $a = \mu$  in M a defined as:

$$\mathbb{P} \text{let a} = \mu \text{ in M a}^{[E]} = \sum_{a \in \text{supp}(\mu)} \mathbb{P}_{\mu}[\{a\}] \cdot \mathbb{P}_{(Ma)}[E]$$

## Semantics of Probabilistic Expressions - revisited

We would like to define it on the structure:

```
\{f(e_1, ..., e_n, d_1, ..., d_k)\}_m = \{f\}(\{e_1\}_m, ..., \{e_n\}_m, \{d_1\}_m, ..., \{d_k\}_m)
```

With input a memory m and output a subdistribution  $\mu \in D(A)$  over the corresponding type A. E.g.

```
{uniform(\{0,1\}^n)}<sub>m</sub>\inD(\{0,1\}^n)} {gaussian(k,\sigma)}<sub>m</sub>\inD(Real)
```

# Semantics of PWhile Commands

What is the meaning of the following command?

```
k := \$ uniform(\{0,1\}^n); z := x mod k;
```

We can give the semantics as a function between command, memories and subdistributions over memories.

Cmd \* Mem 
$$\rightarrow$$
 D (Mem)

We will denote this relation as:

$$\{c\}_{m}=\mu$$

### **Semantics of Commands**

This is defined on the structure of commands:

```
\{abort\}_m = \mathbf{O}
     \{skip\}_m = unit(m)
     \{x := e\}_m = unit(m[x \leftarrow \{e\}_m])
     \{x:=\$ d\}_m = let a = \{d\}_m in unit(m[x \leftarrow a])
   \{c;c'\}_{m} = let m' = \{c\}_{m} in \{c'\}_{m'}
{if e then c_t else c_f}<sub>m</sub> = {c_t}<sub>m</sub> If {e}<sub>m</sub>=true
{if e then c_t else c_f}<sub>m</sub> = {c_f}<sub>m</sub> | if {e}<sub>m</sub>=false
```

## Revisiting the example

```
OneTimePad(m : private msg) : public msg
  key :=$ Uniform({0,1}n);
  cipher := msg xor key;
  return cipher
```

Learning a ciphertext does not change any a priori knowledge about the likelihood of messages.

How do we formalize this?

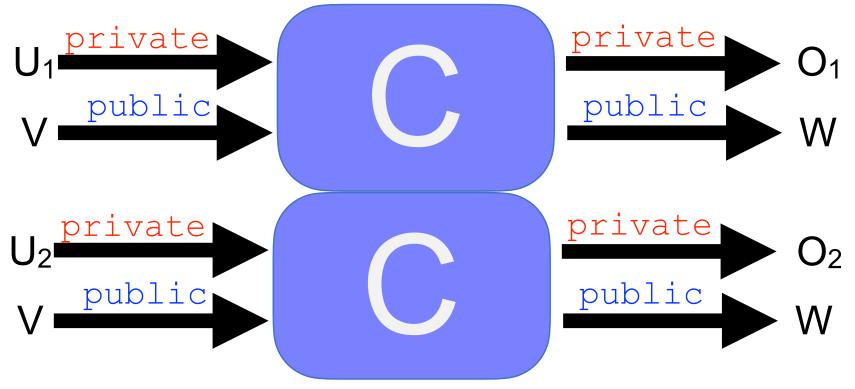
### Probabilistic Noninterference

A program prog is probabilistically noninterferent if and only if, whenever we run it on two low equivalent memories m<sub>1</sub> and m<sub>2</sub> we have that the probabilistic distributions we get as outputs are the same on public outputs.

### Noninterference as a Relational Property

In symbols, c is noninterferent if and only if for every  $m_1 \sim_{low} m_2$ :

 $\{c\}_{m1}=\mu_1 \text{ and } \{c\}_{m2}=\mu_2 \text{ implies } \mu_1 \sim_{low} \mu_2$ 



## Revisiting the example

```
OneTimePad(m : private msg) : public msg
  key :=$ Uniform({0,1}n);
  cipher := msg xor key;
  return cipher
```

How can we prove that this is noninterferent?