CS 591: Formal Methods in Security and Privacy

Approximate probabilistic relational Hoare Logic

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Q&A

To increase interactivity, I will ask more question to each one of you.

It is not a test, you can always answer "pass!"



Everyone should have a project now.

Please, don't hesitate in contacting us if there is some issue with your project.



- This is a reminder that we will record the class and we will post the link on Piazza.
- This is also a reminder to myself to start recording!

From the previous classes

An example

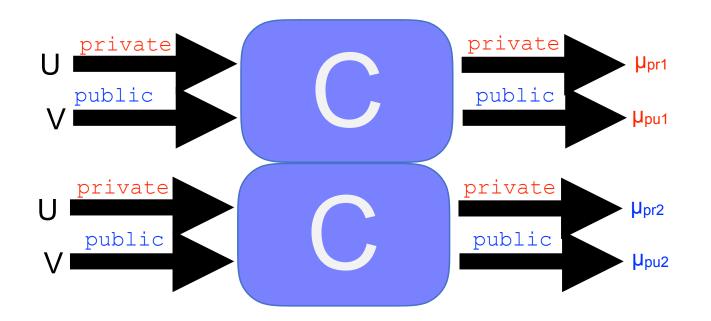
OneTimePad(m : private msg) : public msg
key :=\$ Uniform({0,1}ⁿ);
cipher := msg xor key;
return cipher

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

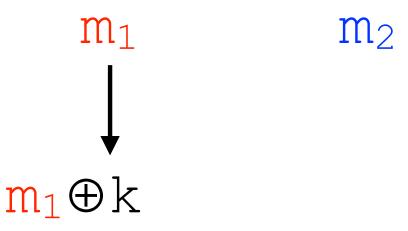
Probabilistic Noninterference as a Relational Property

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

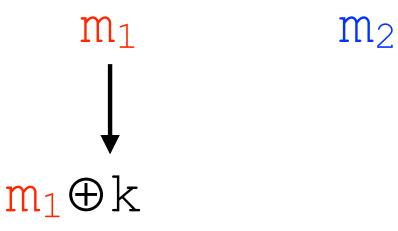
 ${c}_{m1}=\mu_1 \text{ and } {c}_{m2}=\mu_2 \text{ implies } \mu_1 \sim_{low} \mu_2$



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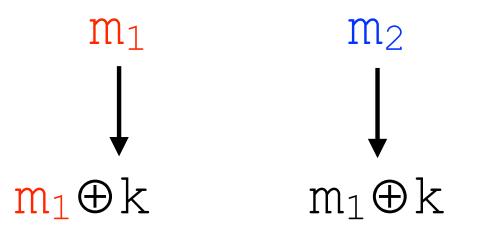


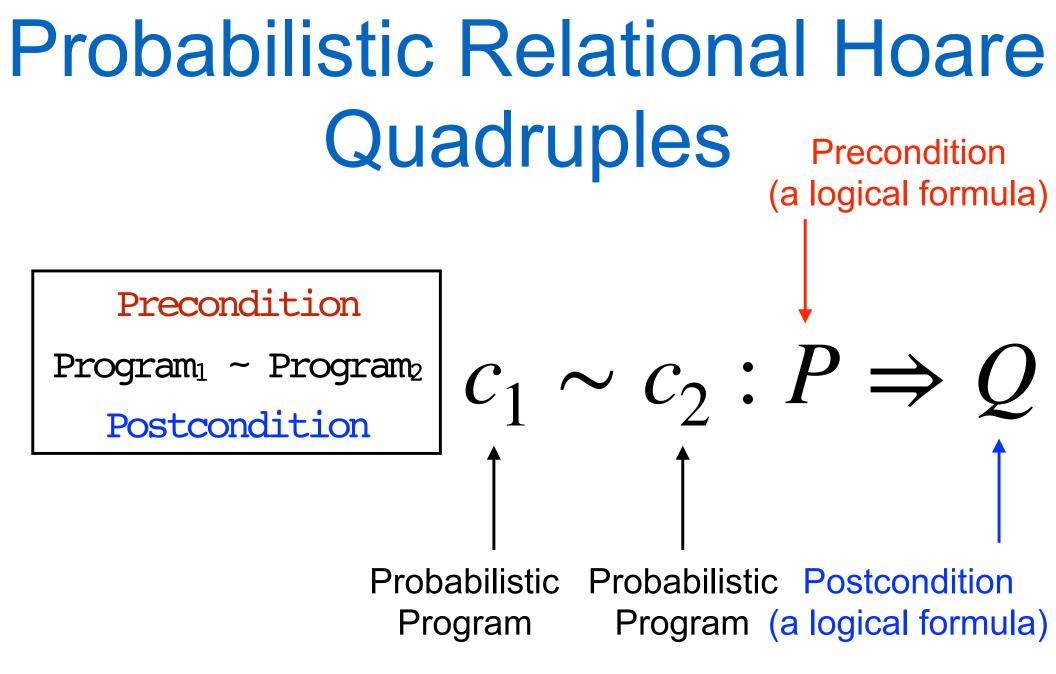
Suppose we can now chose the key for m₂. What could we choose?

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

 $\begin{array}{cccc} m_1 & m_2 & & Suppose we \\ \downarrow & \downarrow & \downarrow \\ m_1 \oplus k & m_2 \oplus (m_1 \oplus k \oplus m_2) \end{array} & \begin{array}{c} \text{Suppose we} \\ \text{can now} \\ \text{chose the key} \\ \text{for } m_2. \end{array} \\ \begin{array}{c} \text{What} \\ \text{could we} \\ \text{choose?} \end{array}$

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R-Coupling

Given two distributions $\mu_1 \in D(A)$, and $\mu_2 \in D(B)$, an *R*-coupling between them, for $R \subseteq AxB$, is a joint distribution $\mu \in D(AxB)$ such that:

- the marginal distributions of µ are µ₁ and µ₂, respectively,
- 2) the support of μ is contained in R. That is, if $\mu(a,b)>0$, then $(a,b)\in R$.

Validity of Probabilistic 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: $\{c_1\}_{m1}=\mu_1$ and $\{c_2\}_{m2}=\mu_2$ implies $Q^*(\mu_1, \mu_2)$.

Consequences of Coupling

Given the following pRHL judgment

$$\vdash c_1 \sim c_2 : \mathsf{True} \Rightarrow Q$$

We have that:

if $Q \Rightarrow (R\langle 1 \rangle \iff S\langle 2 \rangle)$, then $\Pr[c_1 : R] = \Pr[c_2 : S]$

if $Q \Rightarrow (R\langle 1 \rangle \Rightarrow S\langle 2 \rangle)$, then $\Pr[c_1 : R] \le \Pr[c_2 : S]$

A sufficient condition for R-Coupling

- Given two distributions $\mu_1 \in D(A)$, and $\mu_2 \in D(B)$, and a relation $R \subseteq AxB$, if there is a mapping h:A \rightarrow B such that:
 - h is a bijective map between elements in supp(µ1) and supp(µ2),
 - 2) for every $a \in \text{supp}(\mu_1)$, $(a,h(a)) \in \mathbb{R}$

3)
$$Pr_{x \sim \mu 1}[x=a] = Pr_{x \sim \mu 2}[x=h(a)]$$

Then, there is an R-coupling between μ_1 and μ_2 . We write $h \triangleleft (\mu_1, \mu_2)$ in this case.

Probabilistic Relational Hoare Logic Random Assignment

h ⊲ ({d₁}, {d₂}) P=∀v, v∈supp({d₁}) $\Rightarrow Q[v/x_1 < 1>, h(v)/x_2 < 2>]$

 $\vdash x_1 :=$ $d_1 \sim x_2 :=$ $d_2 : P \Rightarrow Q$

Back to our example

h (k) = (m<1>⊕k⊕m<2>) ⊲ ({d₁}, {d₂}) P=∀k, k∈{0,1}ⁿ

⇒ $m < 1 > \oplus k_1 < 1 > = m < 2 > \oplus k_2 < 2 > [v/k_1 < 1 >, h(v)/k_2 < 2 >] =$

 $m < 1 > \oplus k = m < 2 > \oplus (m < 1 > \oplus k \oplus m < 2 >)$

 $\vdash k_1 := \$Uniform(\{0,1\}^n) \sim k_2 := \$Uniform(\{0,1\}^n) :$ True \Rightarrow m<1> $\oplus k_1 < 1 >= m<2> \oplus k_2 < 2>$

Back to our example

h (k) = (m<1>⊕k⊕m<2>) ⊲ ({d₁}, {d₂}) P=∀k, k∈{0,1}ⁿ

⇒ $m < 1 > \oplus k_1 < 1 > = m < 2 > \oplus k_2 < 2 > [v/k_1 < 1 >, h(v)/k_2 < 2 >] =$

 $m < 1 > \bigoplus k = m < 2 > \bigoplus (m < 1 > \bigoplus k \oplus m < 2 >)$

 $\vdash k_1 := \$ \text{Uniform} (\{0, 1\}^n) \sim k_2 := \$ \text{Uniform} (\{0, 1\}^n) :$ True $\Rightarrow m < 1 > \bigoplus k_1 < 1 > = m < 2 > \bigoplus k_2 < 2 >$

Using the assignment rule, we can conclude.

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. **Completeness?**

Today: approximate probabilistic noninterference

One time pad

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

What are the drawbacks of one time pads?

A more realistic example

StreamCipher(m : private msg[n]) : public msg[n]
 pkey :=\$ PRG(Uniform({0,1}*));
 cipher := msg xor pkey;
 return cipher

What guarantees do we want from a PRG?

Properties of PRG

We would like the PRG to increase the number of random bits but also to guarantee the result to be (almost) random.

We can express this as:

PRG:
$$\{0,1\}^k \rightarrow \{0,1\}^n$$
 for $n > k$

 $PRG(Uniform(\{0,1\}^k) \approx Uniform(\{0,1\}^n)$

How can we measure the similarity between the result of PRG and the uniform distribution?

Statistical distance

We say that two distributions $\mu_1, \mu_2 \in D(A)$, are at statistical distance δ if and only if:

 $\Delta(\mu 1, \mu 2) = \max_{E \subseteq A} | Pr_{x \sim \mu 1}[x \in E] - Pr_{x \sim \mu 2}[x \in E] | = \delta$

Properties of PRG

We would like the PRG to increase the number of random bits but also to guarantee the result to be (almost) random.

We can express this as:

PRG: $\{0,1\}^k \rightarrow \{0,1\}^n$ for n > k

 $\Delta(\mathsf{PRG}(\mathsf{Uniform}(\{0,1\}^k),\mathsf{Uniform}(\{0,1\}^n) \le 2^{-n})$

In fact this is a too strong requirement - usually we require that every polynomial time adversary cannot distinguish the two distributions in statistical distance

```
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          : public msg[n]
          key :=$ Uniform({0,1}<sup>n</sup>);
          cipher := msg xor key;
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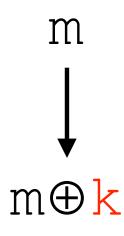
M

m

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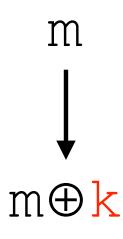
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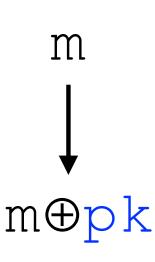
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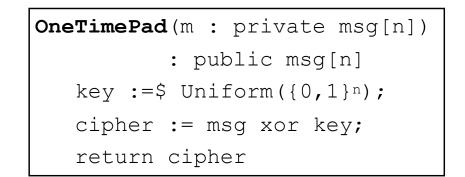


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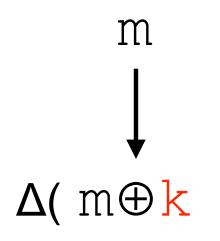
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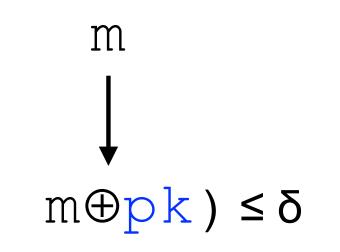






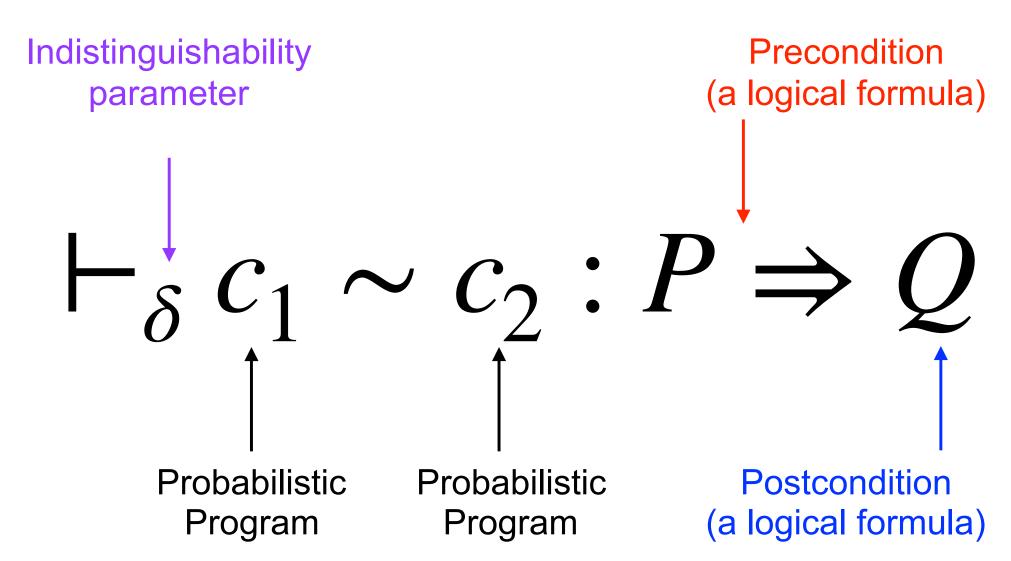
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How to reason formally about this formally?

Approximate Probabilistic Relational Hoare Logic



How can we define validity?

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