

Belief in Information Flow

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1 Introduction

- Information Flow
- Techniques for Information Flow Security
- Quantitative Information Flow Security

2 Informal Reasoning

- Initial State
- Experiment 1
- Experiment 2
- The Uncertainty Reduction Principle
- The Proposed Principle

3 Formal Reasoning

- Basics
- Attacker-System Interaction
- The Proposed Measure

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- **Information flow security** establishes bounds on information leakage

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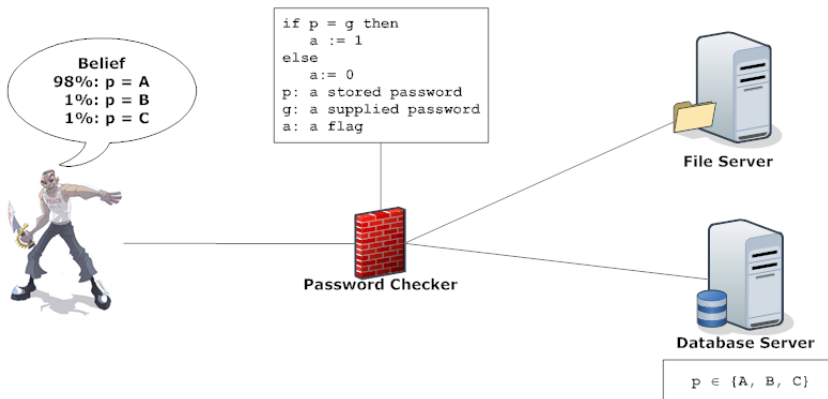
- Treat a program's execution as a channel for transmitting messages
~> Compute the capacity of this channel
- Set bounds on the values of the entropy of input distributions
- Assume that the program input values are independently and uniformly chosen
- Fix a probability distribution on a program's secret inputs
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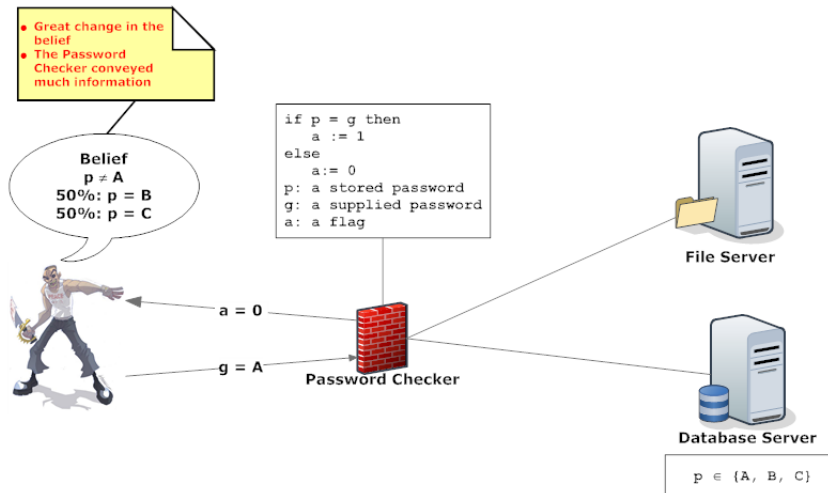
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Initial State



Experiment 2 \rightsquigarrow p is C

The Uncertainty Reduction Principle

- A measure of information flow proposed by Denning in the eighties
 - \uparrow in uncertainty \rightsquigarrow information has flowed
 - \downarrow in uncertainty \rightsquigarrow information has not flowed
- This principle is unsuitable when input distributions represent attacker beliefs
 - In the initial state \rightsquigarrow attacker is **almost certain**
 - After experiment 2 \rightsquigarrow attacker is **somewhat uncertain**
 - This is \uparrow in uncertainty \rightsquigarrow information has not flowed
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The Proposed Principle

- Information flow corresponds to an **improvement** in the accuracy of an attacker's belief
 - \uparrow in accuracy \rightsquigarrow attacker was informed \rightsquigarrow information has flowed
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- We suppose 4 sets
 - *Var* set of variables
 - *Val* set of values
 - *State* set of program states
 - *Dist* set of distributions
- A state $\sigma \in State$ is an assignment in $Var \rightarrow Val$
- A distribution $\delta \in Dist$ is an assignment in $State \rightarrow \mathbb{R}^+$
- A state mass $\dot{\sigma}$ is a probability distribution that maps σ to 1
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- We use confidentiality labels to identify secret data
 - $L \rightsquigarrow$ low-confidentiality public data
 - $H \rightsquigarrow$ high-confidentiality secret data
- $\sigma \upharpoonright L \rightsquigarrow$ low projection of the state $\sigma \rightsquigarrow$ the part of σ visible to the attacker
- $\sigma \upharpoonright H \rightsquigarrow$ high projection of the state $\sigma \rightsquigarrow$ the part of σ not visible to the attacker
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- PWC: if $p_H = g_L$ then $a_L := 1$ else $a_L := 0$
- The attacker chooses a pre-belief $b_H = (0.98, 0.01, 0.01)$
- The system chooses $\sigma_H = (p \rightarrow A)$
- The attacker chooses $\sigma_L = (g \rightarrow A, a \rightarrow 0)$
- The input to PWC is $\sigma'_L \otimes \sigma'_H$
- PWC executes once
- The output is a frequency distribution $\delta' = [PWC](\sigma'_L \otimes \sigma'_H)$
from which one state is chosen $\sigma' = (p \rightarrow A, g \rightarrow A, a \rightarrow 1)$
- The attacker observes $o = \sigma' \upharpoonright L = (g \rightarrow A, a \rightarrow 1)$

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- The attacker generates a **prediction** of getting authenticated
 $\delta'_A = [PWC](\sigma_L \otimes b_H)$
- To incorporate the information in o , The attacker **conditions**
 $\delta'_A|o$

p	g	a	δ'_A	$\delta'_A o$
A	A	0	0	0
A	A	1	0.98	1
B	A	0	0.01	0
B	A	1	0	0
C	A	0	0.01	0
C	A	1	0	0

- The attacker **projects** on the high state to obtain her post-belief $b'_H = (\delta'_A|o) \upharpoonright H = (1, 0, 0)$
- This matches with the informal reasoning

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The Proposed Measure

- Information flow corresponds to an **improvement** in the accuracy of an attacker's belief
 - Accuracy of attacker's pre-belief b_H is $D(b_H \rightarrow \dot{\sigma}_H)$ (Kullback–Leibler divergence)
 - Accuracy of attacker's post-belief b'_H is $D(b'_H \rightarrow \dot{\sigma}_H)$
 - $\Delta = D(b_H \rightarrow \dot{\sigma}_H) - D(b'_H \rightarrow \dot{\sigma}_H)$
 - $= \dot{\sigma}_H \bullet \log \frac{\dot{\sigma}_H}{b_H(\dot{\sigma}_H)} - \dot{\sigma}_H \bullet \log \frac{\dot{\sigma}_H}{b'_H(\dot{\sigma}_H)}$ (Kullback–Leibler)
 - $= 1 \bullet \log \frac{1}{b_H(\dot{\sigma}_H)} - 1 \bullet \log \frac{1}{b'_H(\dot{\sigma}_H)}$ (Definition of state mass)
 - $= -\log b_H(\dot{\sigma}_H) + \log b'_H(\dot{\sigma}_H)$
 - $= -\log b_H(\dot{\sigma}_H) + \log b_H(\dot{\sigma}_H) \bullet \frac{\delta_S(o)}{\delta_A(o)}$ (proved in the paper)
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 - $= -\log \delta_A(o) + \log \delta_S(o)$
 - $= -\log \Pr_{\delta_A}(o) + \log \Pr_{\delta_S}(o)$
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Flow in Experiment 1

- $\Delta_1 = -\log Pr_{\delta_A}(o_1) + \log Pr_{\delta_S}(o_1) = -\log 0.98 + \log 1 = 0.0291$ bit.

Flow in Experiment 2

- $\Delta_2 = -\log Pr_{\delta_A}(o_2) + \log Pr_{\delta_S}(o_2) = -\log 0.02 + \log 1 = 5.6439$ bit.

- Thus the flow in Experiment 2 is larger than it is in Experiment 1
- Again, this matches with the informal reasoning

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- The **Uncertainty** Reduction Principle cannot satisfactorily explain information flow when input distributions represent attacker beliefs
- **Accuracy** is the appropriate measure for information flow in the presence of attacker beliefs

Thank you!