

KTH-AAR CTF

- May 6, 14.00 - 18.00
- Novice-friendly CTF across two Language-Based Security courses
 - KTH (Stockholm)
 - Aarhus
- Teams up to 6 people

Quantitative security

Language-Based Security 2025
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Security metrics

- Noninterference is a good theoretical basis, BUT not sufficient in practice
 - (cf. declassifies in programming assignment)
- Quantitative approach:
 - “how much” information is leaked through program execution.
 - If we have such an approach then noninterference is just a special case where “how much == nothing”.

Idea behind quantitative approach

- Given a program. There are
 - some secret input H
 - some public output L
- Adversary observes L , tries to deduce something about H .
- Can we quantify the information content using *numbers*:
 - how much information there is in H ?
 - how much of that information about H is leaked to L ?
 - how much information about H remains unlearned?
- Intuition
 - “initial uncertainty = information leaked + remaining uncertainty”
- Whatever these numbers are they need to be meaningful:
 - Sanity checks
 - for NI, the leakage should be 0
 - for non-zero case the numbers should have value proportional to their actual security significance

The formal setting

- Simple imperative deterministic programming language
 - a priori known distribution of inputs
- We'll look at two definitions:
 - one based on Shannon entropy: see that it satisfies only one of the sanity checks but not the other
 - a definition based on min-entropy
- We will focus only on the *semantic definition* of security, not how to enforce it, but this is important:
 - If the semantic definition is broken, a sound/precise analysis w.r.t. that definition is useless.

Example

Bitwise AND



- Consider program
 $L := H \& (11111)_2$
- This program copies the last 5 bits of H into L
- Suppose H is a 32-bit integer
 - This program leaks 5 out of 32 bits of H

Shannon entropy

- Suppose that X is a random variable
 - example: a coin, a dice, etc
- let \mathbf{X} be the set of possible values of X
 - for coin: $\mathbf{X} = \{\text{head, tails}\}$
 - for a dice: $\mathbf{X} = \{1, \dots, 6\}$
- Write $p(x)$ for probability that r.v. X has value x
- Definition: *entropy* of the r.v. X is a measure of *uncertainty* in X

$$\mathbf{H}(X) = - \sum_{x \in \mathbf{X}} p(x) \cdot \log p(x)$$

(Convention: $0 \cdot \log 0 = 0$)

Example: a coin

- Suppose we have a fair coin
 - $p(\text{heads}) = 1/2$
 - $p(\text{tails}) = 1/2$
- Q: what's the entropy of the fair coin?
 - A: $-((0.5 * \log_2 0.5) + (0.5 * \log_2 0.5))$
- Obs: this is independent of the values in **X**.

Conditional entropy

- Suppose we have two r.v.s. X and Y :
- Conditional entropy: how much uncertainty there is about X given Y
- (expected value of the entropies of the conditional distribution)

Conditional random variable

$$\mathbf{H}(X|Y) = \sum_{y \in Y} p(y) \cdot \mathbf{H}(X|Y = y)$$

probability of $X = x$, given $Y = y$

where $\mathbf{H}(X|Y = y) = - \sum_{x \in X} p(x|y) \cdot \log p(x|y)$

If X is determined by Y , then $\mathbf{H}(X|Y) = 0$

Mutual information

- Amount of information shared between r.v.s X and Y
 - $I(X; Y) = H(X) - H(X|Y)$
 - Turns out to be symmetric: $I(X; Y) = I(Y; X)$
- Slogan:
 - “initial uncertainty = information leaked + remaining uncertainty”
- OR:
 - “information leaked = initial uncertainty - remaining uncertainty”
- If H and L are r.v.s. define leakage to be $H(H) - H(H | L)$, or simply $I(H; L)$

Deterministic programs

- L is determined by H, in this case $\mathbf{H}(L|H) = 0$
- $\mathbf{I}(H; L)$
= $\mathbf{I}(L; H)$
= $\mathbf{H}(L) - \mathbf{H}(L | H)$
= $\mathbf{H}(L)$
- Suppose H is a uniformly distributed 32-bit int.

Program	$H(H)$	$I(H; L)$	$H(H L)$
L := 0	32	0	32
L := H	32	32	0
L := H & 037	32	5	27

What about sanity checks? OK for NI (leakage 0), but what about non-zero leakage?

Examples

Suppose H is uniformly distributed $8k$ bit integer

$$0 \leq H \leq 2^{8k}$$

$$\mathbf{H}(H) = 8k$$

Example1

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if H mod 8 = 0 then
  L := H
else
  L := 1
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Example2

$$L := H \ \& \ 0^{7k-1} 1^{k+1}$$

Proposal: vulnerability

Definition 1. Given a random variable X with space of possible values \mathcal{X} , the vulnerability of X , denoted $V(X)$, is given by

$$V(X) = \max_{x \in \mathcal{X}} P[X = x].$$

Intuition: worst-case probability that adversary could guess the value of X correctly in one try.

m guesses: at most $m V(x)$

Definition 2. The min-entropy of X , denoted $H_\infty(X)$, is given by

$$H_\infty(X) = \log \frac{1}{V(X)}.$$

log, because we want a metric in bits ;)

Conditional vulnerability

Definition 3. Given (jointly distributed) random variables X and Y , the conditional vulnerability $V(X|Y)$ is

$$V(X|Y) = \sum_{y \in \mathcal{Y}} P[Y = y] V(X|Y = y)$$

where

$$V(X|Y = y) = \max_{x \in \mathcal{X}} P[X = x|Y = y].$$

Definition 4. The conditional min-entropy $H_\infty(X|Y)$ is

$$H_\infty(X|Y) = \log \frac{1}{V(X|Y)}.$$

Back to the slogan:

- initial uncertainty = $H_\infty(H)$
- remaining uncertainty = $H_\infty(H|L)$
- information leaked = $H_\infty(H) - H_\infty(H|L)$

Theorems

Theorem 1. *If c is deterministic and H is uniformly distributed, then the information leaked is $\log |\mathcal{L}|$.*

Theorem 2. *If c is deterministic and H is uniformly distributed, then the information leaked, $\log |\mathcal{L}|$, is equal to the channel capacity of c .*

channel capacity is the max value of $H(L)$

Dangers of scientific bandwagons

“While we feel that information theory is indeed a valuable tool in providing fundamental insights into the nature of communication problems and will continue to grow in importance, it is certainly no panacea for the communications engineer or, a fortiori, for anyone else. Seldom do more than a few of nature’s secrets give way at one time.”

C. E. Shannon, “The bandwagon,” IRE Transactions on Information Theory, vol. 2, no. 1, p. 3, **1956**.

Conclusion

- Security definitions are tricky business ;-)
- Quantitative security: measuring leakage of a program (system) in bits
 - Min-entropy is the consensus definition these days
 - Applications: timing leaks, where the timing behavior of the program is the observable output depending on secrets
 - Research trends: synthesis of adaptive attacks for timing leaks from source code analysis