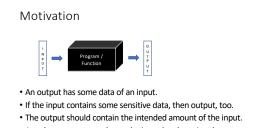
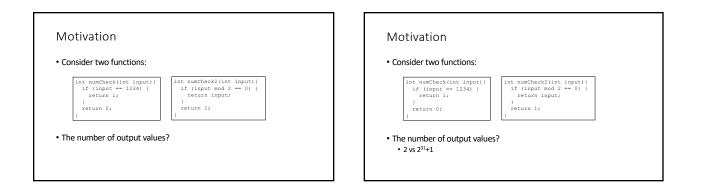
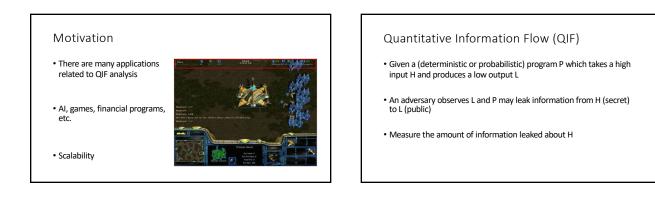
# Quantitative Information Flow Analysis

CSCI 5271 Guest Lecture Seonmo (Sean) Kim



• An adversary wants to know the input by observing the output.





# Early models of QIF

- Used the Shannon mutual information I(X;Y)
- Uncertainty
  - I(H; L) = H(H) H(H | L)
  - information leaked = initial uncertainty remaining uncertainty
     the adversary's initial uncertainty before observing L
     the adversary's remaining uncertainty after observing L
  - H(H) I(H; L) = H(H | L)

- Shannon entropy: initial uncertainty
- $H(X) = -\sum_{x \in \mathcal{X}} \Pr[X=x] \cdot \log_2 \Pr[X=x]$
- If H is a 32-bit integer and L := H  $Pr[H=x]=1/2^{32}$ ,  $log_2 Pr[H=x] = log_2 2^{-32} = -32$   $H(H) = -2^{32}(1/2^{32})(-32) = 32$

## Shannon entropy: information leaked

- I(X;Y) = H(X) H(X | Y) = H(X) + H(Y) H(X,Y)• If X is determined by Y, then H(X|Y)=0.
- I(H; L) = I(L; H) = H(L) H(L | H) = H(L)

## • If H is a 32-bit integer and L := H

- $\begin{array}{l} \text{In } \text{Is } \text{32-bit miggs}, \text{ since } \text{I}, \text{II} \in \mathbb{R}^{3}, \\ \text{I}(H_{1}) = H(L) = H(L) = H(H) = 32 \\ \text{Pr}[\text{H}=x] = 1/2^{32}, \log_{2}\Pr[\text{H}=x] = \log_{2}2^{-32} = -32 \\ \text{I}(H) = -\sum_{x \in X}\Pr[\text{X}=x] \cdot \log_{2}\Pr[\text{X}=x] = -2^{32}(1/2^{32})(-32) = 32 \end{array}$
- Remaining uncertainty: H(H|L) = 32 32 = 0

# Shannon entropy

- $H(X) = -\sum_{x \in \mathcal{X}} \Pr[X=x] \cdot \log_2 \Pr[X=x]$  If H is a 32-bit integer and L := H, H(H) = 32
- $H(X \mid Y) = H(X) I(X;Y)$
- If H is a 32-bit integer and L := H, H(H | L) = 0 • I(X;Y) = I(Y;X) = H(Y) - H(Y | X) = H(Y), if Y is determined by X

I (H ; L)

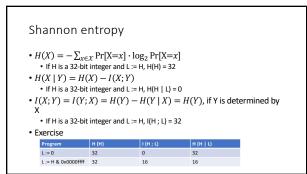
H (H | L)

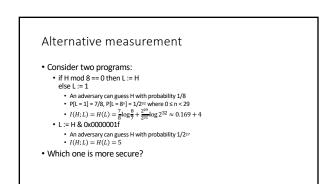
• If H is a 32-bit integer and L := H, I(H ; L) = 32

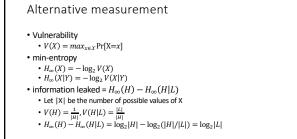
H (H)

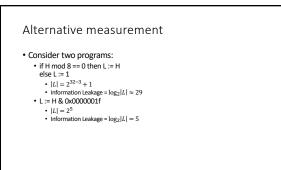
 Exercise Program L := 0

L := H & 0x0000ffff



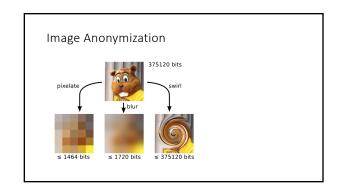


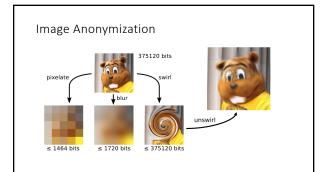


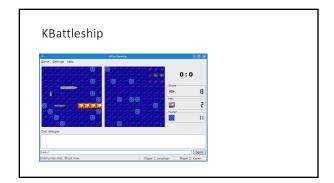


# Applications

- Image anonymization and Kbattleship (PLDI 2008)
   Computing a maximum flow of information
- Error reporting system (ASPLOS 2008)
- Heartbleed (VMCAI 2018)
   Using the model counting technique to measure the leakage







# Flowcheck

Dynamic analysis tool to measure an upper-bound estimate of the amount of information leaked

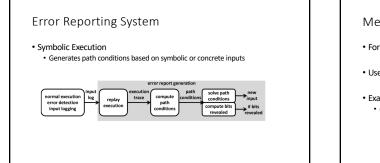
a)

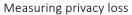
C

(b)

- Dynamic tainting
- Static control-flow regions
- c = d = a + b



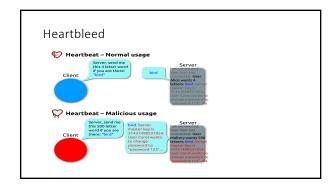


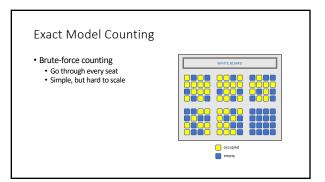


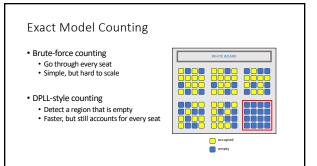
- For each condition (op f(.) g(.)), compute a summary for f and g
- Use a set of rules to compute the bound given the summaries

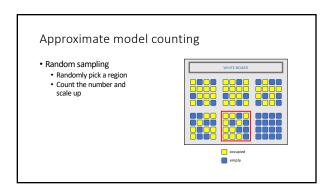
### Example

- (add (bitwise-and x 1) 3)
   (bitwise-and x 1) -> 0 or 1
  - (add (bitwise-and x 1) 3) -> 3 or 4









#### Approximate model counting • Random sampling Randomly pick a region Count the number and scale up • Random hashing(AAAI 2006) Everyone flips a coin k times Leave if a tail is ever shown • Count the persons $\boldsymbol{n}$ empty • Approximately $2^k \cdot n$ persons

Q & A		
	Thank You:)	