# Shannon's Theory of Secure Communication

CSG 252 Fall 2006

Riccardo Pucella

#### Introduction

- Last time, we have seen various cryptosystems, and some cryptanalyses
- How do you ascertain the security of a cryptosystem?
- Some reasonable ideas:
  - Computational Security: best alg takes a long time
    - This is impossible to achieve
    - © Can be done against specific attacks (brute-force search)
  - Provable Security: reduce the security of a cryptosystem to a problem believed (or known) to be hard
  - Unconditional Security: Cryptosystem cannot be broken even with infinite computation power

# Review of Probability Theory

- Security generally expressed in terms of probability
  - Because an attacker can always guess the key!
  - This is true of any cryptosystem, and unavoidable

We only need discrete probabilities for now

## Probability Distributions

- $\circ$  Probability space:  $(\Omega,Pr)$ 
  - $\odot$   $\Omega$ , the sample space, is a finite set of possible worlds (or possible outcomes)
  - $\circ$  Pr is a function  $P(\Omega) \rightarrow [0,1]$  such that
    - $Pr(\Omega) = 1$
    - $Pr(\emptyset) = 0$
  - Pr is called a probability distribution, a probability measure, or just a probability
- $\odot$  Because of additivity, Pr determined by Pr( $\{a\}$ )  $\forall a$

## Examples

- Single dice:
  - $\Omega = \{1,2,3,4,5,6\}$
  - $Pr({4}) = 1/6$
  - $Pr(\{1,3,5\}) = 3/6 = 1/2$
- Pair of dice:
  - $\Omega = \{(1,1),(1,2),(1,3),(1,4),...,(6,5),(6,6)\}$
  - $Pr (\{(1,1)\}) = 1/36$
  - $Pr(\{(1,a) \mid a=1,2,3,4\}) = 4/36 = 1/9$

#### Joint Probabilities

- $\odot$  Suppose  $(\Omega_1, Pr_1)$  is a probability space
- $\odot$  Suppose  $(\Omega_2, Pr_2)$  is a probability space
- © Can create the joint probability space ( $\Omega_1 \times \Omega_2$ ,Pr) by taking:
  - $Pr({a,b}) = Pr_1({a})Pr_2({b})$
  - Extending by additivity

# Conditional Probability

- - Only defined if Pr(B)>0
- More easily understood with a picture...

Bayes' Theorem: Pr (B | A) = Pr (A | B) Pr(B) / Pr(A)

#### Random Variables

- A random variable is a function from worlds to some set of values
- Given probability space and a random variable X, the probability that the random variable X takes value x is:

$$Pr ( \{ w \mid X(w) = x \} )$$

- This is often written Pr(X=x) YUCK or Pr[x] YUCK<sup>2</sup>
- The probability space is often left implicit
- © Conditional probabilities:
  Pr (X=x | Y=y) = Pr ({w | X(w)=x} | {w | Y(w)=y})
- $\bullet$  X and Y are independent if  $P(X=x \cap Y=y) = Pr(X=x) Pr(Y=y) \forall x,y$

# Application to Cryptography

- $\odot$  Suppose a probability space ( $\Omega$ ,Pr) with:
  - Random variable K (=key)
  - Random variable P (=plaintext)
  - K and P are independent random variables
    - Simple example: worlds are (key,plaintext) pairs
- Key probability is Pr(K=k)
- Plaintext probability is Pr(P=x)

# Ciphertext Probability

This induces a probability over ciphertexts:

$$Pr(C = y) = \sum_{x,k \bullet e_k(x) = y} Pr(P = x)Pr(K = k)$$

Can compute conditional probabilities:

$$Pr(C = y \cap P = x) = Pr(P = x) \sum_{k \bullet e_k(x) = y} Pr(K = k)$$
 $Pr(C = y \mid P = x) = \sum Pr(K = k)$ 

 $k \bullet e_k(x) = y$ 

$$Pr(P=x \mid C=y) = \frac{Pr(P=x) \sum_{k \bullet e_k(x)=y} Pr(K=k)}{\sum_{x',k \bullet e_k(x')=y} Pr(P=x') Pr(K=k)}$$

# Perfect Secrecy

We say a cryptosystem has perfect secrecy if

$$Pr(P=x \mid C=y) = Pr(P=x)$$
 for all x,y

- The probability that the plaintext is x given that you have observed ciphertext y is the same as the probability that the plaintext is x (without seeing the ciphertext)
- Depends on key probability and plaintext probability

# Characterizing Perfect Secrecy

Theorem: The shift cipher, where all keys have probability 1/26, has perfect secrecy if we use the key only once, for any plaintext probability.

© Can we characterize those cryptosystems with perfect secrecy?

Theorem: Let (P,C,K,E,D) be a cryptosystem with |K| = |P| = |C|. This cryptosystem has perfect secrecy if and only if all keys have the same probability 1/|K| and

 $\forall x \in P \ \forall y \in C \ \exists k \in K \bullet e_k(x) = y$ 

## Vernam Cipher

Also know as the one-time pad

- P = C = K =  $(Z_2)^n$ Strings of bits of length n
- If K=(k<sub>1</sub>, ..., k<sub>n</sub>):
   Ø e<sub>K</sub> (x<sub>1</sub>, ..., x<sub>n</sub>) = (x<sub>1</sub>+k<sub>1</sub> (mod 2), ..., x<sub>n</sub>+k<sub>n</sub> (mod 2))<br/>
   Ø d<sub>K</sub> (x<sub>1</sub>, ..., x<sub>n</sub>) = (x<sub>1</sub>-k<sub>1</sub> (mod 2), ..., x<sub>n</sub>-k<sub>n</sub> (mod 2))
- To encrypt a string of length N, choose a one-time pad of length N

#### Conclusions

- If ciphertexts are short (same length as key), can get perfect security
  - Approach still used for very sensitive data (embassies, military, etc)
- But keys get very long for long messages
- And there is the whole key distribution problem
- Modern cryptosystems: one key used to encrypt long plaintext (by breaking it into pieces)
  - We will see more of these next time
- Need to be able to reason about reusing keys

## 10 minutes break

## A Detour: Entropy

- Entropy: measure of uncertainty (in bits) introduced by Shannon in 1948
  - Foundation of Information Theory
- Intuition
  - Suppose a random variable that takes value {1,...,n} with some nonzero probability
  - Consider the string of values generated by that probability distribution
  - What is the most efficient way (in number of bits) to encode every value to minimize how many bits it take to encode a random string?
  - Example: {1,...,8}, where 8 is much more likely than others

# Definition of Entropy

Let random variable take values in finite set V

$$H(X) = -\sum_{v \in V} Pr(X = v) \log_2 Pr(X = v)$$

Weighted average of -log<sub>2</sub> Pr (X=v)

Theorem: Suppose X is a random variable taking n values with nonzero probability, then

$$H(X) \leq log_2(n)$$

When do we have equality?

# Huffman Encoding

Algorithm to get a  $\{0,1\}$  encoding that takes less than H(X)+1 bits on average

- 1. Start with a table of letter probabilities
- 2. Create a list of trees, initially all trees with only a letter and associated probability
- 3. Iteratively:
  - a. Pick the two trees  $T_1$ ,  $T_2$  with smallest probabilities from the list
  - b. Create a small tree with edge 0 leading to  $\mathsf{T}_1$  and edge 1 leading to  $\mathsf{T}_2$
  - c. Add that tree back to the list, with probability the sum of the original probabilities
- 4. Stop when you get a single tree giving the encoding

# Conditional Entropy

- Let X and Y be random variables
- Fix a value y of Y
- The Define the random variable X|y such that  $Pr(X|y = x) = Pr(X=x \mid Y=y)$

$$H(X \mid y) = -\sum_{v \in V} Pr(X = v \mid Y = y) \log_2 Pr(X = v \mid Y = y)$$

Conditional entropy, written H(X|Y):

$$H(X \mid Y) = \sum_{y} Pr(Y = y)H(X \mid y)$$

Intuition: average amount of information about X that remains after observing Y

# Application to Cryptography

 ⊗ Key equivocation H(K | C): amount of uncertainty of the key that remains after observing the ciphertext

Theorem: 
$$H(K \mid C) = H(K) + H(P) - H(C)$$

- A spurious key is a possible key, but incorrect
  - © E.g., shift cipher, with ciphertext WNAJW
  - Possible keys: k=5 (RIVER) or k=22 (ARENA)
- Many spurious keys Good!

# How Many Spurious Keys?

- Question: how long of a message can we permit before the number of spurious keys is 0?
  - That is, before the only key that is possible is the right one?
- This depends on the underlying language in which plaintexts are taken
- © Cf: cryptanalysis, where we took advantage that not all letters have equal probability in English messages

## Entropy of a Language

- H<sub>L</sub> = number of information bits per letter in language L
- Example:
  - If all letters have the same probability, a first approximation would be 4.7
  - For English, based on probabilities of plaintexts (letters), a first approximation is 4.19
  - For pairs of letters? Triplets of letters? ...
- Entropy of L:

$$H_L = \lim_{n \to \infty} \frac{H(P^n)}{n}$$

Redundancy of L:

$$R_L = 1 - \frac{H_L}{\log_2|P|}$$

# Unicity Distance

Theorem: Suppose (P,C,K,E,D) is a cryptosystem with |C| = |P| and keys are chosen equiprobably, and let L be the underlying language. Given a ciphertext of length n (sufficiently large), the expected number of spurious keys  $s_n$  satisfies

$$s_n \ge \frac{|K|}{|P|^{nR_L}} - 1$$

- The unicity distance of a cryptosystem is the value  $n_0$  after which the number expected number of spurious keys is 0.
  - Average amount of ciphertext required for an adversary to be able to compute the key (given enough time)
- Substitution cipher:  $n_0 = 25$ 
  - So have a chance to recover the key if encrypted message is longer than 25 characters