

- Cryptography - Fri. 12. May. 2017

$$X = \{x_1, \dots, x_n\} \quad \mathbb{P}(X=x_i) = P_i$$

$H(X) \rightarrow ?$

$$(X=x_i) = E_i^{\mathbb{P}} \longrightarrow \underbrace{-\log \mathbb{P}(X=x_i)}_{\substack{\text{Single letter} \\ \text{information}}} \left\{ \begin{array}{l} \mathbb{P}(X=x_i)=1 \\ \quad \downarrow \\ \quad 0 \\ \mathbb{P}(X=x_i)=0 \\ \quad \downarrow \\ \quad +\infty \end{array} \right.$$

$$\left\{ \begin{array}{l} X=x_1 \quad k \text{ bits} \longrightarrow -\log_2 \mathbb{P}(X=x_1) \\ \vdots \\ X=x_n \quad k' \text{ bits} \longrightarrow -\log_2 \mathbb{P}(X=x_n) \end{array} \right.$$

$$n=2^k \quad \underbrace{\log n}_{\text{bits}} \quad \underbrace{-\sum_{i=1}^n \mathbb{P}(X=x_i) \log \mathbb{P}(X=x_i)}_{i=x_i=1}$$

$$\mathbb{P}(X=x_i)=0 \Rightarrow \mathbb{P}(X=x_i) \log \mathbb{P}(X=x_i)=0$$

$$\rightarrow H(X) = -\sum_{x_i=1}^n \mathbb{P}(X=x_i) \log \mathbb{P}(X=x_i)$$

$$H(X|Y) = \sum_{x,y} -\mathbb{P}(X=x, Y=y) \log \mathbb{P}(X=x|Y=y)$$

$$H(X|Y) \leq H(X)$$

\*  $H(X)$  : uncertainty we have about  $X$  in average

→ Shannon proved that  $H(X)$  is a lower bound for the average codeword length of any uniquely decodable code.

$$\rightarrow X = \{x_1, \dots, x_n\} \quad \mathbb{P}(X=x_i) = p_i$$

the worst case is when you have uniform distribution

$$\mathbb{P}(X=x_i) = \frac{1}{|X|} = \frac{1}{n} \Rightarrow H(X) = \log n$$

→ Hence  $R = 1 - \frac{H(X)}{\log n}$  is called redundancy of a source.

#### 4.2 Perfect secrecy

Consider the cryptosystem  $(\mathcal{M}, \mathcal{K}, \mathcal{C}, d, e)$

$\mathcal{M} = \{M_1, \dots, M_m\}$  message space

$\mathcal{K} = \{K_1, \dots, K_k\}$  key space

$\mathcal{C} = \{C_1, \dots, C_n\}$  ciphertext space.

$$e(M_i, K_j) = C_{ij}$$

Assume that  $\hat{M}, \hat{K}$  are independent RV with their support  $\mathcal{M}, \mathcal{K}$ .

$$\mathbb{P}(\hat{M}=M_i) = P_i \quad \text{and} \quad \mathbb{P}(\hat{K}=K_j) = q_j$$

these prob. distributions model the occurrence of messages and keys.

\* Encryption:  $\hat{C} = e(\hat{M}, \hat{K})$ :

$$\mathbb{P}(\hat{C} = c_\ell) = \sum_{e(M_i, K_j) = c_\ell} \mathbb{P}(M_i, K_j)$$

$$e(M_i, K_j) = c_\ell$$

$$= \sum_{e(M_i, K_j) = c_\ell} p_i \cdot q_j$$

$$e(M_i, K_j) = c_\ell$$

$$H(\hat{M}) = -\sum_{i=1}^m p_i \log p_i \quad H(\hat{K}) = -\sum_{j=1}^k q_j \log q_j$$

$$H(\hat{C}) = -\sum_{\ell} \mathbb{P}(\hat{C} = c_\ell) \log \mathbb{P}(\hat{C} = c_\ell)$$

$H(\hat{K} | \hat{C})$  is called key equivocation

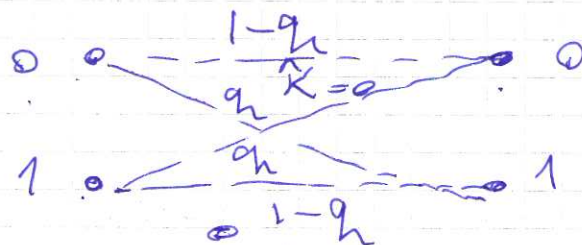
$H(\hat{M} | \hat{C})$  is called message equivocation.

Example Let  $M = K = C = \{0, 1\}$

x  $\hat{M} \sim \text{Bern}(p) \rightarrow \mathbb{P}(\hat{M}=1) = p$  and  $\mathbb{P}(\hat{M}=0) = 1-p$

x  $\hat{K} \sim \text{Bern}(q) \rightarrow \mathbb{P}(\hat{K}=1) = q$  and  $\mathbb{P}(\hat{K}=0) = 1-q$

$$\hat{C} = e(\hat{M}, \hat{K}) = (\hat{M} + \hat{K}) \bmod 2$$





$$\rightarrow \hat{C} = (\hat{M} + \hat{K}) \text{ mod } 2$$

1) Probability distributions

$$\hat{M}, \hat{K}$$

$$\mathbb{P}(\hat{C}=0) \quad \mathbb{P}(\hat{C}=1)$$

$$\begin{aligned} \mathbb{P}(\hat{C}=0) &= \mathbb{P}(\hat{C}=0, \hat{M}=0) + \mathbb{P}(\hat{C}=0, \hat{M}=1) \\ &= \mathbb{P}(\hat{C}=0 | \hat{M}=0) \mathbb{P}(\hat{M}=0) \\ &\quad + \mathbb{P}(\hat{C}=0 | \hat{M}=1) \mathbb{P}(\hat{M}=1) \end{aligned}$$

$$\mathbb{P}(\hat{C}=0 | \hat{M}=0) = \mathbb{P}(\hat{K}=0) = 1 - q_h$$

$$\mathbb{P}(\hat{C}=0 | \hat{M}=1) = \mathbb{P}(\hat{K}=1) = q_h$$

$$\mathbb{P}(\hat{C}=0) = q_h \cdot p + (1-p)(1-q_h)$$

$$\mathbb{P}(\hat{C}=1) = \cancel{(1-p)} q_h + \cancel{p} (1-q_h)$$

$$\begin{array}{ccc} 1-p & \circ & \frac{1-q_h}{q_h} \circ \circ \\ & \diagdown & \diagup \\ p & \circ & \frac{q_h}{1-q_h} \circ \circ \end{array}$$

$$\left. \begin{aligned} \mathbb{P}(\hat{C}=0, \hat{M}=0) &= (1-q_h) \cdot (1-p) \\ \mathbb{P}(\hat{M}=0) &= 1-p \\ \mathbb{P}(\hat{C}=0) &= q_h p + (1-p)(1-q_h) \end{aligned} \right\}$$

In general  $\hat{M}$  and  $\hat{C}$  are not statistically independent  $\mathbb{P}(\hat{C}=0, \hat{M}=0) \neq \mathbb{P}(\hat{C}=0) \mathbb{P}(\hat{M}=0)$

$$q_n = \frac{1}{2} \Rightarrow \mathbb{P}(\hat{C}=0) = P \cdot \frac{1}{2} + (1-P) \cdot \frac{1}{2} = \frac{1}{2}$$

$$\mathbb{P}(\hat{C}=0, \hat{M}=0) = \frac{1}{2} \times (1-P)$$

$$\mathbb{P}(\hat{M}=0) = 1-P$$

$$\Rightarrow \mathbb{P}(\hat{C}=0, \hat{M}=0) = \mathbb{P}(\hat{C}=0) \times \mathbb{P}(\hat{M}=0)$$

$q_n = \frac{1}{2} \rightarrow$  the key is uniformly distributed.

$$H(\hat{M}|\hat{C}) \leq H(\hat{M}) \Rightarrow H(\hat{M}|\hat{C}) = H(\hat{M})$$

$\hat{M}$  and  $\hat{C}$  are  
stat. independent.

\* Definition 4.9) a cryptosystem  $(M, K, C, e, d)$  is said to have perfect secrecy if  $H(\hat{M}|\hat{C}) = H(\hat{M})$

Interpretation: The knowledge of cryptosystem  $C$  does not decrease uncertainty about  $\hat{M}$  or does not increase information about  $\hat{M}$ .

Corollary 4.11 A cryptosystem has perfect secrecy.

$\Leftrightarrow \hat{M}$  and  $\hat{C}$  are stochastically independent.

$$\Leftrightarrow \mathbb{P}(\hat{M} = M_i | \hat{C} = C_\ell) = \mathbb{P}(\hat{M} = M_i) \quad \forall M_i \in \mathcal{M} \quad \mathbb{P}(\hat{C} = C_\ell) > 0$$

$$\Leftrightarrow \mathbb{P}(\hat{C} = C_\ell | \hat{M} = M_i) = \mathbb{P}(\hat{C} = C_\ell) \quad \forall M_i \in \mathcal{M} \text{ with } \mathbb{P}(\hat{M} = M_i) > 0$$

The above conditions are all tedious to check, easy sufficient conditions are given in the following.

Theorem 4.14. A cryptosystem  $(\mathcal{M}, \mathcal{K}, \mathcal{E}, e, d)$  has perfect secrecy if:

(i)  $\mathbb{P}(\hat{K} = K) = \frac{1}{|\mathcal{K}|}$  for all  $K \in \mathcal{K}$

(ii) for all  $M \in \mathcal{M}$  and  $C \in \mathcal{E}$ , there is a unique  $K \in \mathcal{K}$  s.t.  $e(M, K) = C$ .

Proof:

$$\begin{aligned} \mathbb{P}(\hat{C} = C | \hat{M} = M) &\stackrel{(ii)}{=} \frac{\mathbb{P}(e(\hat{M}, \hat{K}) = C, \hat{M} = M)}{\mathbb{P}(\hat{M} = M)} \\ &= \frac{\mathbb{P}(e(M, \hat{K}) = C, \hat{M} = M)}{\mathbb{P}(\hat{M} = M)} \end{aligned}$$



$$\begin{aligned}
 \text{(ii) } \exists K & \\
 K = K(M, C) &= \frac{\mathbb{P}(e(M, \hat{K}) = C) \mathbb{P}(\hat{M} = M)}{\mathbb{P}(\hat{M} = M)} \\
 &= \mathbb{P}(e(M, \hat{K}) = C) \\
 &= \mathbb{P}(\hat{K} = K(M, C)) = \frac{1}{|K|}
 \end{aligned}$$

$$\begin{aligned}
 \mathbb{P}(\hat{C} = C) &= \sum \mathbb{P}(\hat{C} = C | \hat{M} = M) \mathbb{P}(\hat{M} = M) \\
 &= \sum \frac{1}{|K|} \mathbb{P}(\hat{M} = M) = \frac{1}{|K|}
 \end{aligned}$$

$$\Rightarrow \mathbb{P}(\hat{C} = C | \hat{M} = M) = \mathbb{P}(\hat{C} = C)$$

Hence perfect secrecy  $\square$

## Vernam Ciphers

$$\mathcal{X} = \{0, \dots, m-1\} \quad M_N = e_N = K_N = \mathcal{X}^N$$

$$e(M, K) = ((a_1 + s_1) \bmod m, \dots, (a_N + s_N) \bmod m)$$

$$M = (a_1, \dots, a_N)$$

$$K = (s_1, \dots, s_N)$$

$\hat{M}_N$  r.v. with  $\text{supp. } M_N$

$\hat{K}_N$  i.i.d.  $\mathbb{P}(\hat{K}_j = i) = \frac{1}{m}$

$i = 1, \dots, m$

Theorem 4.15 The vernal cipher has perfect secrecy.

Proof) Thm. 4.14,

$$\begin{aligned} \text{i) } \mathbb{P}(\hat{K}_N = K) &= \mathbb{P}(\hat{K}_1 = S_1, \dots, \hat{K}_N = S_N) \\ &= \frac{1}{m} \times \dots \times \frac{1}{m} = \frac{1}{m^N} \\ &= \frac{1}{|K_N|} \end{aligned}$$

ii)  $M, C \rightarrow$  unique  $\underline{K}$

$M \in \mathcal{M}_N$  and  $C \in \mathcal{E}_N$

$$C = (c_1, \dots, c_N)$$

$$M = (a_1, \dots, a_N)$$

therefor

$$c_i = (a_i + s_i) \bmod m$$

$$s_i = (c_i - a_i) \bmod m$$

$$K = ((c_1 - a_1) \bmod m, \dots, (c_N - a_N) \bmod m)$$

unique

□



## 5) Fast Block Ciphers.

### 5.1) The Data Encryption Standard (DES)

15 May 1973: The national Bureau of standards (NBS), national Institut of standards and Technology (NIST) solicited proposals for a cryptosystem.

An algorithm proposed by IBM was chosen.

- Based on a predecessor LUCIFER

x 17 March 19~~75~~<sup>77</sup>: DES was published and publications started.

x 15 Jan 1979: DES was adopted for unclassified applications.

Reviewed every 5 years

- Last official renewal 1999,

19.5.2005 NIST suspended DES as a standard.