# Statistical and Computational Security

## **Statistical (Total Variation) Distance**

• Let  $\mathcal X$  and  $\mathcal Y$  be two distributions over  $\{0,1\}^m$  The statistical distance between  $\mathcal X$  and  $\mathcal Y$ , denoted  $\Delta(\mathcal X,\mathcal Y)$  is

$$\max_{T\subseteq\{0,1\}^m} |\Pr[\mathcal{X}\in T] - \Pr[\mathcal{Y}\in T]|$$

$$\mathcal{X} \qquad \text{If } \Delta(\mathcal{X},\mathcal{Y}) \leq \varepsilon \text{ we write}$$

$$\mathcal{X} \equiv_{\varepsilon} \mathcal{Y}$$

#### **Statistical Security**

A symmetric encryption scheme is said to be ε-statistically secure, if for any two plaintexts  $P_1, P_2$  distributions  $E_k(P_1), E_k(P_2)$  are ε-equivalent

#### Theorem.

Let (K,E,D) be a SES with m-bit messages and m-1 –bit keys. Then there are plaintexts  $P_1, P_2$  with  $\Delta(E_k(P_1), E_k(P_2)) \ge \frac{1}{2}$ 

## **Statistical Security (cntd)**

Observation.

If  $\mathbb{E}[X] \le \mu$  then  $\Pr[X \le \mu] > 0$ .

Proof (of the theorem).

Let  $P_1 = 0^m$  and  $S = \{E_k(0^m) \mid k \in \{0,1\}^{m-1}\}$  Then  $|S| \le 2^{m-1}$  Experiment:

Choose a random plaintext  $P \in \{0,1\}^m$  define  $2^{m-1}$  random variables: for every  $k \in \{0,1\}^{m-1}$  we set

 $T_k(P) = 1$  if  $E_k(P) \in S$  and 0 otherwise

For every k,  $E_k$  is one-to-one, hence,  $\Pr[T_k = 1] \le \frac{1}{2}$ 

Therefore  $\mathbb{E}[T_k] \leq \frac{1}{2}$ 

#### **Statistical Security (cntd)**

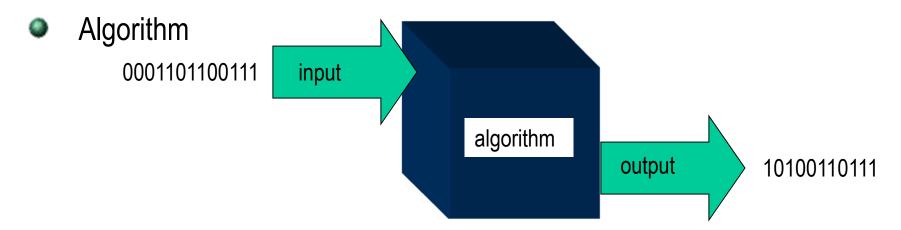
Proof (cntd)
Set  $T = \sum_{k} T_{k}$  Then  $\mathbb{E}[T] = \mathbb{E}\left[\sum_{k} T_{k}\right] = \sum_{k} \mathbb{E}[T_{k}] \le \frac{2^{m-1}}{2}$ 

By Observation,  $\Pr[T \leq \frac{2^{m-1}}{2}] > 0$ , or in other words, there exists P such that  $\sum_k T_k(P) \leq \frac{2^{m-1}}{2}$ 

For such P at most half of the keys satisfy  $E_k(P) \in S$  or, equivalently,  $\Pr[E_k(P) \in S] \leq \frac{1}{2}$ 

Since  $\Pr[E_k(0^m) \in S] = 1$ , we get  $\Delta(E_k(0^m), E_k(P)) \ge \frac{1}{2}$ 

#### **Algorithms**



- Algorithm performs a sequence of `elementary steps' that can be:
  - arithmetic operations
  - bit operations
  - Turing machine moves
  - ..... (but not quantum computing!!)
- We allow probabilistic algorithms, that is flipping coins is permitted

## Complexity

- The time complexity of algorithm A is function f(n) that is equal to the number of elementary steps required to process the most difficult input of length n
- We do not distinguish algorithms of complexity 2n² and 100000n²
- A computational problem has time complexity at most f(n) if there is an algorithm that solves the problem and has complexity O(f(n))
  - problem solvable in linear time: there is an algorithm that on input of length n performs at most Cn steps
  - problem solvable in quadratic time: there is an algorithm that on input of length n performs at most Cn² steps

## **Complexity (cntd)**

- Polynomial time solvable problems:
  - There is a polynomial p(n) such that the problem is solvable in time O(p(n))
- P class of problems solvable in poly time by a deterministic algorithm
- BPP class of problems solvable in poly time by a probabilistic algorithm
- An algorithm is superpolynomial if its time complexity f(n) is not in O(p(n)) for any polynomial p(n)
- A function  $\varepsilon: \mathbb{N} \to [0,1]$  is polynomially bounded if  $\varepsilon(n) \ge \frac{1}{p(n)}$  for some polynomial p(n)

#### **Computational Security**

Let (K,E,D) be a SES that uses n-bit keys to encrypt m(n)-bit messages. It is computationally secure if for any polynomial time algorithm Eve:  $\{0,1\}^* \rightarrow \{0,1\}$ , any polynomially bounded  $\epsilon$ :  $\{0,1\}^* \rightarrow [0,1]$ , n, and  $P_1,P_2 \in \{0,1\}^{m(n)}$ 

$$|\Pr[\mathsf{Eve}(E_{U_n}(P_1)) = 1] - \Pr[\mathsf{Eve}(E_{U_n}(P_2)) = 1]| < \varepsilon(n)$$

#### Conjecture.

A computationally secure SES exists for  $m(n) = n^{100}$  (may be even for  $m(n) = 2^{0.9n}$ )

## **Computational Indistinguishability: Difficulties**

- It is useful to define computational security in a similar way as statistical one: define distance or equivalence of distributions and then say that  $E_{U_n}(P_1)$  and  $E_{U_n}(P_2)$  are 'equivalent'. However, there are problems
- For computational definitions we need algorithms, not events Solution: Instead of saying  $X \in S$  we use the characteristic function f of S. So we say f(X) = 1 instead.

Distance between distributions can then be defined as

$$\max_{f} \big| \Pr[f(X) = 1] - \Pr[f(Y) = 1] \big|$$
 over all 'easily' computable functions f

Computational complexity does not make sense for fixed distributions.

Solution: Use collections or sequences of random variables

## **Computational Indistinguishability: Definition**

Let T(n) and  $\varepsilon(n)$  be functions on natural numbers. Collections of random variables  $\{X_n\}$  and  $\{Y_n\}$  such that  $X_n, Y_n \in \{0,1\}^n$  are said to be computationally  $(T,\varepsilon)$ -indistinguishable, if for any probabilistic algorithm. Alg with time complexity at most T(n)

$$\big|\Pr[\mathsf{Alg}(X_n)=1] - \Pr[\mathsf{Alg}(Y_n)=1]\big| \leq \mathcal{E}(n)$$
   
 Denoted  $\{X_n\} \approx_{T,\mathcal{E}} \{Y_n\}$ 

For example:

Let (K,E,D) be a SES that uses n-bit keys to encript m(n)-bit messages. It is computationally secure if for any  $P_1,P_2\in\{0,1\}^{m(n)}$  distributions  $E_{U_n}(P_1)$  and  $E_{U_n}(P_2)$  are  $(T,\epsilon)$ -indistinguishable for any polynomial T and any polynomially bounded  $\epsilon$