

Lecture Notes 2:**Review of Probability & Classical Ciphers****Reading.**

- Katz–Lindell §A.3, Ch. 1.
- Cormen, Leiserson, Rivest, Stein. *Introduction to Algorithms* (2nd ed), Appendix C & Ch. 5.

1 Review of Probability**1.1 Probability spaces**

A *probability space* is a finite or countable set S together with a function $\Pr : S \rightarrow [0, 1]$ such that $\sum_{x \in S} \Pr[x] = 1$. In this course, the probability space will not always be specified explicitly. Consider the following example:

- Alice flips 100 fair coins $A \in \{0, 1\}^{100}$.
- Bob flips 100 fair coins $B \in \{0, 1\}^{100}$.
- Carol chooses with probability $3/4$ Alice's coin tosses ($C = A$), with probability $1/4$ Bob's coin tosses ($C = B$).
- Eve gets $E = A \oplus B$ (bitwise XOR).

Here, the underlying probability space is $S = \{0, 1\}^{100} \times \{0, 1\}^{100} \times \{a, b\}$. For any triplet (x, y, z) ,

$$\Pr[(x, y, z)] =$$

The source of the randomness is all the coin tosses of the involved parties or the random choices made.

An *event* is a subset of the probability space. The probability of an event T is defined to be $\Pr[T] \stackrel{\text{def}}{=} \sum_{x \in T} \Pr[x]$, but often can be computed more directly.

Example:

$$\Pr[\text{Alice has 7 zeroes}] =$$

1.2 Random variables

Random variables are functions, not necessarily real-valued, on the probability space. In our example, we can consider the following random variables:

- A , Alice's coin tosses (which is just the first coordinate for an element of the probability space)

- Z_A , the number of zeroes obtained by Alice
- $Z_A + Z_B$, the number of zeroes obtained by Alice and Bob together

The random variables X and Y are said to be *independent* if :

$$\forall x, y, \Pr[X = x \ \& \ Y = y] = \Pr[X = x] \cdot \Pr[Y = y].$$

Examples:

- A and B ?
- Z_A and Z_B ?
- Z_A and Z_C ?

Random variables X_1, \dots, X_k are *independent* if

$$\forall x_1, \dots, x_k, \Pr[X_1 = x_1 \ \& \ X_2 = x_2 \ \& \ \dots \ \& \ X_k = x_k] = \Pr[X_1 = x_1] \cdot \Pr[X_2 = x_2] \cdot \dots \cdot \Pr[X_k = x_k].$$

Not the same as pairwise independence!

Example: Three random variables from above experiment that are pairwise independent but not independent are...

1.3 Expectation of a random variable

The *expectation* of a real-valued random variable X is defined as:

$$\mathbb{E}[X] \stackrel{\text{def}}{=} \sum_s \Pr[s] \cdot X(s) = \sum_{x: \Pr[X=x] > 0} \Pr[X = x] \cdot x,$$

where the second equality holds if $\mathbb{E}[|X|]$ is finite (in particular, for finite probability spaces), but may not hold in general (due to convergence issues).

We have the property of linearity:

$$\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y]$$

$$\mathbb{E}[cX] = c \cdot \mathbb{E}[X] \text{ for any constant } c$$

Note that $\mathbb{E}[XY] = \mathbb{E}[X] \cdot \mathbb{E}[Y]$ if X and Y are independent, *but not in general*.

Examples:

- $\mathbb{E}[Z_A] =$
- $\mathbb{E}[Z_A^2] =$

1.4 Markov's inequality

If X is a non-negative real-valued random variable, we have:

$$\Pr[X \geq t] \leq \frac{\mathbb{E}[X]}{t}$$

If X has a small expectation, we have a bound on how often the random variable can get large.

Example: $\Pr[Z_A \geq 70] \leq$

1.5 Chernoff Bound

This is a form of the Law of Large Numbers, which says that the average of random variables over many independent trials will be close to the expectation (with high probability).

Let X_1, \dots, X_n be independent $[0, 1]$ -valued random variables, $X = (1/n) \cdot \sum_i X_i$ be the average of the X_i 's, and $\mu = \mathbb{E}[X]$. The *Chernoff Bound* states that

$$\Pr \left[\frac{1}{n} \sum_{i=1}^n X_i \geq \mu + \varepsilon \right] \leq e^{-2\varepsilon^2 n}$$

and

$$\Pr \left[\frac{1}{n} \sum_{i=1}^n X_i \leq \mu - \varepsilon \right] \leq e^{-2\varepsilon^2 n}.$$

Example: $\Pr[Z_A \geq 70] \leq$

1.6 Conditioning

Let E and F be events. We define the probability of E occurring given that F occurs as:

$$\Pr[E|F] = \frac{\Pr[E \cap F]}{\Pr[F]}$$

Bayes' Law states that:

$$\Pr[E|F] = \frac{\Pr[F|E] \cdot \Pr[E]}{\Pr[F]}$$

Example: How to compute $\Pr[Z_A \text{ is even} | Z_C \text{ is even}]$?

2 Private-Key Encryption: Classical Ciphers

- The setting for private-key encryption is the following: two parties share a *secret key* and want to exchange messages *privately* over “insecure channel”. For now, we will not worry about how they came to share the secret key.
- Kerckhoffs's Principle: Assume encryption/decryption algorithms are known to adversary. Only thing secret is the *key*.
- For now, “insecure channel” means that adversary can *listen* to all messages sent, but cannot inject/alter messages, i.e. *passive* rather than *active*.
- **Definition 1** A (private-key) encryption scheme *consists of three algorithms* (Gen, Enc, Dec), *as follows*:

- The key generation algorithm **Gen** is a randomized algorithm that returns a key $k \in \mathcal{K}$; we write $k \xleftarrow{R} \text{Gen}$.
- The encryption algorithm **Enc** is a randomized algorithm that takes a key $k \in \mathcal{K}$ and a plaintext (aka message) $m \in \mathcal{M}$ and outputs a ciphertext $c \in \mathcal{C}$; we write $c \xleftarrow{R} \text{Enc}_k(m)$.
- The decryption algorithm **Dec** is a deterministic algorithm that takes a key $k \in \mathcal{K}$ and a ciphertext $c \in \mathcal{C}$ and returns a plaintext $m \in \mathcal{M}$.

The *message space* \mathcal{M} is often the set of strings of a given length. The ciphertext space \mathcal{C} does not have to equal the plaintext space. We require $\text{Dec}_k(\text{Enc}_k(m)) = m$ for all $m \in \mathcal{M}$.

- The definition describes the functionalities of the encryption scheme but does not take security into account yet. For example:

- **Examples:**

- Shift cipher (cf. Caesar cipher). The key is a random number: $k \xleftarrow{R} \{0, \dots, 25\}$, the message space is $\mathcal{M} = \{A, \dots, Z\}^\ell$ (strings of length ℓ over the English alphabet) so we can see the message as $m \in \{0, \dots, 25\}^\ell$. $\text{Enc}_k(m_1 m_2 \dots m_\ell) = c_1 c_2 \dots c_\ell$, where $c_i = m_i + k \pmod{26}$.
- Substitution cipher. The key k is a random permutation of $\{0, \dots, 25\}$. $\text{Enc}_k(m_1 m_2 \dots m_\ell) = k(m_1) k(m_2) \dots k(m_\ell)$.
- One-time pad. The message space consists of binary strings of length ℓ and the key k is a random element of $\{0, 1\}^\ell$. $\text{Enc}_k(m) = m \oplus k$ (bitwise XOR). The decryption is $\text{Dec}_k(c) = c \oplus k$.
- Vigenère cipher. The key is a string $k = k_0 k_1 \dots k_{t-1}$ in $\{0, \dots, 25\}^t$ for some length parameter t , and now a (possibly long) message $m = m_0 \dots m_\ell$ is encrypted to a ciphertext $c = c_0 \dots c_\ell$ by using a shift cipher with key $k_{i \bmod t}$ to encrypt message symbol m_i .
- Are any of these “secure”?