

## 1 A Priori and A Posteriori Probabilities

### Model Scenario

Consider

- a finite set  $M_0 \subseteq M$  of possible plaintexts—for example all plaintexts of length  $r$  or of length  $\leq r$ ,
- a finite set  $K$  of keys,
- a cipher  $F = (f_k)_{k \in K}$  with  $f_k: M \rightarrow \Sigma^*$ .

The restriction to a finite set  $M_0$  allows us to handle probabilities in the naive way. It is no real restriction since plaintexts of lengths  $> 10^{100}$  are extremely unlikely in this universe that has at most  $10^{80}$  elementary particles.

### Motivating Example

For English plaintexts of length 5 we potentially know exact a priori probabilities, say from a lot of countings. A small excerpt from the list is

Plaintext	Probability
hello	$p > 0$
fruit	$q > 0$
xykph	0
...	...

Now assume we see a monoalphabetically encrypted English text XTJJA. Without knowing the key—that is in a situation where all keys have the same probability—and without further context information we nevertheless assign to the single plaintexts different “a posteriori probabilities”:

Plaintext	Probability
hello	$p_1 \gg p$
fruit	0
xykph	0
...	...

Thus knowledge of the ciphertext alone (and knowledge of the encryption method) changed our information on the plaintext.

A “BAYESian” approach gives a general model of this observation.

### Model

**The probability of plaintexts** is given as a function

$$P: M_0 \rightarrow [0, 1] \quad \text{where} \quad P(a) > 0 \quad \text{for all } a \in M_0$$

$$\text{and} \quad \sum_{a \in M_0} P(a) = 1.$$

(This is the a priori probability of plaintexts.)

**The probability of keys** is likewise given as a function

$$P: K \longrightarrow [0, 1] \quad \text{such that} \quad \sum_{k \in K} P(k) = 1.$$

(By abuse of notation denoted by the same letter  $P$ .) In general we assume a uniform distribution  $P(k) = 1/\#K$  for all  $k \in K$ .

**The probability of ciphertexts** derives from the probabilities of plaintexts and keys, implicitly assumed as independently chosen:

$$P: \Sigma^* \longrightarrow [0, 1], \quad P(c) := \sum_{a \in M_0} \sum_{k \in K_{ac}} P(a) \cdot P(k),$$

where  $K_{ac} := \{k \in K \mid f_k(a) = c\}$  is the set of all keys that transform  $a$  to  $c$ .

**Remark 1** Only finitely many  $c \in \Sigma^*$  have  $P(c) \neq 0$ . These form the set

$$C_0 := \{c \in \Sigma^* \mid P(c) > 0\}$$

of “possible ciphertexts”.

**Remark 2** We have

$$\begin{aligned} \sum_{c \in \Sigma^*} P(c) &= \sum_{c \in \Sigma^*} \sum_{a \in M_0} \sum_{k \in K_{ac}} P(a) \cdot P(k) \\ &= \sum_{a \in M_0} \sum_{k \in K} P(a) \cdot P(k) \\ &= \sum_{a \in M_0} P(a) \cdot \sum_{k \in K} P(k) \\ &= 1. \end{aligned}$$

**The conditional probability for a ciphertext** to stem from a given plaintext  $a \in M_0$  is modeled by the function

$$P(\bullet|a): \Sigma^* \longrightarrow [0, 1], \quad P(c|a) := \sum_{k \in K_{ac}} P(k).$$

**Remark 3**  $\sum_{c \in \Sigma^*} P(c|a) = \sum_{k \in K} P(k) = 1$ .

**Remark 4**  $P(c) = \sum_{a \in M_0} P(a) \cdot P(c|a)$ .

## A Posteriori Probabilities of Plaintexts

The cryptanalyst is interested in the converse, the conditional probability  $P(a|c)$  of a plaintext  $a \in M_0$  if the ciphertext  $c \in \Sigma^*$  is given.

First we describe the probability of the simultaneous occurrence of  $a$  and  $c$  as

$$P: M_0 \times \Sigma^* \longrightarrow [0, 1], \quad P(a, c) := P(a) \cdot P(c|a).$$

**Remark 5** Then

$$\sum_{a \in M_0} P(a, c) = \sum_{a \in M_0} P(a) \cdot P(c|a) = P(c).$$

**The conditional probability of a plaintext** is given by a function  $P(\bullet|c)$  with  $P(a, c) = P(c) \cdot P(a|c)$  by the BAYESIAN formula

$$P(a|c) := \begin{cases} \frac{P(a) \cdot P(c|a)}{P(c)} & \text{if } P(c) \neq 0, \\ 0 & \text{if } P(c) = 0. \end{cases}$$

**Remark 6**  $\sum_{c \in \Sigma^*} P(c) \cdot P(a|c) = \sum_{c \in \Sigma^*} P(a) \cdot P(c|a) = P(a)$  by Remark 3.