

# K-RANDOMIZATION

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## OUTLINE OF THE PROCEDURE

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### 1. THEORETICAL SETUP

In the following we work with data in the form of bit vectors. A **bit vector** is a vector  $v \in \{0, 1\}^L$ .

First we define the randomization procedure we will be applying.

**Definition.** The randomization procedure  $R$  with **lie probability**  $0 < q < 1/2$  flips a bit with probability  $q$ , and leaves it as-is with probability  $1 - q$ . In other words, for a bit  $b \in \{0, 1\}$ ,

$$R(b) = R(b; X) = (1 - b) \cdot X + b \cdot (1 - X) \quad \text{where } X \sim \text{Ber}(q).$$

When applied to a vector, each bit is randomized independently:

$$R(v) = R(v; (X_1, \dots, X_L)) = (R(v_1; X_1), \dots, R(v_L; X_L)) \quad \text{where } X \stackrel{\text{iid}}{\sim} \text{Ber}(q).$$

**Remark.** The randomization  $R$  reports the original bit value with probability  $1 - q > q$ , and lies with probability  $q$ . This is equivalent to the randomized response procedure where the value is reported as-is with probability  $1 - f$ , and with probability  $f$  the reported value is the outcome of the toss of a fair coin. In this case,  $q = f/2$ .

**Remark.** If  $q = 1/2$ , then  $R(0) \stackrel{d}{=} R(1)$ , and the reported value is “completely” randomly generated, i.e., independently of the original value.

Distribution of  $R(v)$ .

For a bit  $b$ , the randomization lies iff  $R(b) \neq b$ :

$$P[R(b) = s] = q^{\mathbf{1}_{\{b \neq s\}}} (1 - q)^{\mathbf{1}_{\{b = s\}}}$$

Hence, for a bit vector  $v$ ,

$$P[R(v) = s] = q^{\sum \mathbf{1}_{\{b_i \neq s_i\}}} (1 - q)^{\sum \mathbf{1}_{\{b_i = s_i\}}} = q^{L - m(v, s)} (1 - q)^{m(v, s)},$$

where  $m(v, s) = |\{i : v_i = s_i\}|$ . Note that this probability is maximized when  $m(v, s) = L$  (the reported vector  $s$  is identical to the original vector  $v$ ), and minimized when  $m(v, s) = 0$ . In other words, the most likely outcome of randomizing a bit vector is obtaining an identical vector.

For a collection  $T$ ,

$$P[s \in R(T)] = 1 - P[s \notin R(T)] = 1 - \prod_{v \in T} P[R(v) \neq s] = 1 - \prod_{v \in T} [1 - q^{L - m(v, s)} (1 - q)^{m(v, s)}].$$

## 2. DIFFERENTIAL PRIVACY

Consider a collection  $T$  of bit vectors, and write  $T_v = T \setminus \{v\}$ . The randomization procedure  $R$  is  $\epsilon$ -differentially private if

$$\log \left( \frac{P[R(T) \in S]}{P[R(T_v) \in S]} \right) \leq \epsilon$$

for any set of bit vectors  $S$ .

Anonymity:

$$A_p = \min_{v \in T, s \in \{0,1\}^L} \frac{P[s \in R(T_v)]}{P[s = R(v)]}$$