

- **Continuous pdf:**

$$f_X(x) : \mathcal{X} \rightarrow [0, \infty), f_X(x) = \lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} \cdot \Pr[x \leq X \leq x + \epsilon]$$

- **Continuous cdf:**

$$F_X(x) := \Pr[X \leq x] \text{ where } \Pr[a \leq X \leq b] = \int_a^b f_X(x) dx$$

- **Uniform distribution on  $\mathcal{X} = [u, v]$ :**

$$f_X(x) = 1/(v - u) \text{ for } x \in \mathcal{X}$$

- **Normal distribution:**

$$f_X(x) = (2\pi)^{-1/2} \exp(-x^2/2)$$

- **Dirac delta property:**

$$\int_{-\infty}^{\infty} dx b(x) \delta(x - a) = b(a)$$

- **Expectation value of  $g(X)$  where  $X \in \mathcal{X}$ :**

$$\mathbb{E}[g(X)] = \sum_{x \in \mathcal{X}} \Pr[X = x] g(x), \text{ k'th moment of } X = \mathbb{E}[X^k]$$

- **Statistical distance of  $X, Y \in \mathcal{X}$ :**

$$\Delta(X, Y) = \frac{1}{2} \sum_{x \in \mathcal{X}} |\mathbb{P}(x) - \mathbb{Q}(x)|$$

- **Covariance matrix  $K$ :**

$$K_{i,j} = \mathbb{E}[X_i X_j] - \mathbb{E}[X_i] \cdot \mathbb{E}[X_j]$$

Zero covariance ( $K_{1,2} = K_{2,1} = 0$ ) does not imply  $X_1$  and  $X_2$  are independent.

- **Marginal distribution for  $X$  when  $(X, Y) \sim \mathbb{P}$ :**

$$\Pr[X = x] = \sum_y \mathbb{P}(x, y)$$

- **Conditional probability for  $(X, Y) \sim \mathbb{P}$ :**

$$\Pr[X = x | Y = y] = \frac{\Pr[X=x, Y=y]}{\Pr[Y=y]} = \frac{\mathbb{P}(x, y)}{\mathbb{P}_2(y)}$$

- **(Shannon) Entropy rules:**

1. **Additivity:** inf of a set of indep. RVs must be the sum of indiv. inf. contents
2. **Sub-additivity:** Total inf. content of two jointly distrib. RVs cannot exceed sum of separate infs.
3. **Expansibility:** Adding extra outcome of prob. 0 does not affect inf.
4. **Normalization:** The distrib  $(1/2, 1/2)$  has inf. of 1 bit.
5. The distrib  $(p, 1 - p)$  for  $p \rightarrow 0$  has zero inf.

- **Shannon entropy:**

Lower bound on the avg length of the shortest desc of  $X$ .

$$H(X) = \sum_{x \in \mathcal{X}} p_x \log_2 \frac{1}{p_x}$$

$$H(X, Y) = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} p_{xy} \log_2 \frac{1}{p_{xy}}$$

- **Renyi entropy:**

TODO!

- **Binary entropy function: 2 outcomes with prob.  $p$  and  $1 - p$ :**

$$h(p) = p \log \frac{1}{p} + (1 - p) \log \frac{1}{1-p}$$

- **Differential entropy for continuous RV  $X \sim \rho$ :**

$$h_{\text{diff}}(X) = - \int dx \rho(x) \log \rho(x) = \mathbb{E}_x \log \frac{1}{\rho(x)}$$

- **Relative entropy (Kullback-Leibler distance):**

$$D(\mathbb{P} || \mathbb{Q}) = \sum_{x \in \mathcal{X}} \mathbb{P}(x) \log \frac{\mathbb{P}(x)}{\mathbb{Q}(x)}$$

- **Entropy of jointly distrib. RVs:  $H(X, Y)$  or  $H(XY)$ :**

$$H(X, Y) = \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} p_{xy} \log \frac{1}{p_{xy}}$$

- **Conditional entropy:**

$$H(X|Y) = \mathbb{E}_y [H(X|Y = y)] = - \sum_{x \in \mathcal{X}} p_x \sum_{y \in \mathcal{Y}} p_{x|y} \log p_{x|y}$$

$$H(X|Y) = H(X, Y) - H(Y)$$

- **Mutual information:**

$$\mathbf{I}(X; Y) = H(X) - H(X|Y) = H(Y) - H(Y|X)$$

$$\mathbf{I}(X; Y) = H(X, Y) - H(X|Y) - H(Y|X)$$

$$\mathbf{I}(X; Y) = H(X) + H(Y) - H(X, Y)$$

$$\mathbf{I}(X; Y|Z) = \mathbb{E}_z \mathbf{I}(X|Z = z; Y|Z = z)$$

- **Min entropy:**

$$H_{\min}(X) = - \log \max_{x \in \mathcal{X}} p_x = - \log p_{\max}$$

$$H_{\min}(X|Y) = - \log \mathbb{E}_y \max_{x \in \mathcal{X}} p_{x|y}$$

- **Linear binary codes:**

Maps  $k$ -bit msg  $x$  to  $n$ -bit ( $n > k$ ) codeword  $c_x \in \mathcal{C}$ . Perceived string  $z = c_x \oplus e$ . Minimum distance of code:  $d = \min_{c, c' \in \mathcal{C}} \text{HammingWeight}(c \oplus c')$ . Receiver determines which  $c_{\hat{x}}$  is closest to  $z$  and decodes it into  $\hat{x}$ . Error correcting capability  $t = \lfloor \frac{d-1}{2} \rfloor$ .

- **Generator ( $G$  is  $k \times n$ ) and parity check ( $H$  is  $(n - k) \times n$ ) matrix:**

$c_x = xG$ .  $G = (\mathbf{1}_k | A)$ .  $H = (-A^T | \mathbf{1}_{n-k})$ .  $GH^T = 0$ ,  $cH^T = 0$ . All  $k$  rows of  $G$  are linearly independent.

- **Syndrome decoding ( $s(z) \in \{0, 1\}^{n-k}$ ):**

$$s(z) = zH^T = (c_x + e)H^T = eH^T$$

Syndrome depends only on the error pattern, not on the message.

- **Hamming bound: Binary code of length  $n$  that can correct  $t$  errors:**

$$2^k \leq 2^n / \sum_{j=0}^t \binom{n}{j}. \text{ Approx } \log n \text{ bits of redundancy per bit error.}$$

- **Channel capacity:**

Inf. content error free:  $k \leq \mathbf{I}(C; Z)$

$$\text{BSC capacity (per bit): } \frac{k}{n} \leq \mathbf{I}(C_j; Z_j) = H(Z_j) - H(Z_j|C_j)$$

This is called the BSC code rate:  $\text{BSC CODE RATE} \leq 1 - h(\epsilon)$

Following the rule of thumb:  $h(\epsilon) = -\epsilon \log \epsilon + \mathcal{O}(\epsilon)$

- **Uniformly random bits from continuous source:**

TODO

- **Randomness sources:**

Ring oscillators (odd number of inverters causes jitter in period, gaussian), noisy resistors (no voltage applied, noise amplitude gaussian distribution), radioactive decay (poisson)

- **The von Neumann alg:**

Given  $(b_1, b_2)$ , if  $b_1 = b_2$  then no output, else output  $b_1$ .

- **Piling-up lemma:**

Let  $X_1, \dots, X_n \in \{0, 1\}$  be independent with biases  $\Pr[X_i = 1] - \Pr[X_i = 0] = \alpha_i$ . Construct  $Y = X_1 \oplus X_2 \oplus \dots \oplus X_n$ . The bias of  $Y$  is  $\Pr[Y = 1] - \Pr[Y = 0] = (-1)^{n-1} \prod_{i=1}^n \alpha_i$ . Thus by xoring many bits together the bias gets reduced.

- **Resilient function:**

A function  $\Psi : \{0, 1\}^n \rightarrow \{0, 1\}^m$  is  $(n, m, t)$ -resilient of, for any  $t$  coords  $i_1, \dots, i_t \in [n]$ , any  $a_1, \dots, a_t \in \{0, 1\}$  and any  $y \in \{0, 1\}^m$  it holds that:  $\Pr[\Psi(X) = y | x_{i_1} = a_1, \dots, x_{i_t} = a_t] = 2^{-m}$   
i.e.: Knowledge of  $t$  values of the input does not give inf. that would help in guessing the output. ECC example  $([n, k, d]$  code):  $\Psi : \{0, 1\}^n \rightarrow \{0, 1\}^k$ .  $\Psi = xG^T$ . Then  $\Psi$  is an  $(n, k, d - 1)$ -resilient fun.

- **Strong extractor  $\text{Ext} : \{0, 1\}^n \times \{0, 1\}^* \rightarrow \{0, 1\}^l$ :**

Takes  $n$ -bit string  $X$  and randomness  $R$  and outputs an  $l$ -bit string ( $l < n$ ).  $Z = \text{Ext}(X, R)$ .  $\text{Ext}$  is a strong extractor for source min-entropy  $m$ , output length  $l$  and nonuniformity  $\epsilon$  if for all distrib of  $X$  with  $H_{\infty}(X) \geq m$  it holds that  $\Delta(ZR; U_l R) \leq \epsilon$ .  $U_l$  is an RV uniform on  $\{0, 1\}^l$ . Also true:  $\mathbb{E}_r \Delta(Z|R = r; U_l) \leq \epsilon$

- **Extractable randomness:**

TODO

- **Universal hash functions:**

Let  $\mathcal{R}, \mathcal{X}$  and  $\mathcal{T}$  be finite sets. Let  $\{\Phi_r\}_{r \in \mathcal{R}}$  be a family of hash functions from  $\mathcal{X}$  to  $\mathcal{T}$ . The family is called universal iff, for  $R$  drawn uniformly from  $\mathcal{R}$ , it holds that:  $\Pr[\Phi_R(x) = \Phi_R(x')] \leq 1/|\mathcal{T}|$ . It's called  $\eta$ -almost universal if it holds that:  $\Pr[\Phi_R(x) = \Phi_R(x')] \leq \eta$  (for all  $x, x' \in \mathcal{X}$  with  $x \neq x'$ ).

- **Leftover hash lemma:**

TODO

- **Noisy broadcast channel and no return channel:**

Secret capacity  $C_s$  of the broadcast channel  $P_{Y|Z|X}$  can be bounded:  $C_s(P_{Y|Z|X}) \geq \max_{P_X} [\mathbf{I}(X; Y) - \mathbf{I}(X; Z)] = \max_{P_X} [H(X|Z) - H(X|Y)]$ . Condition: if Eve's reception quality is better than Bob's ( $\mathbf{I}(X; Y) < \mathbf{I}(X; Z)$ ) then the secrecy capacity is zero. Secrecy capacity of BSC with error rates  $\epsilon$  and  $\delta$  is:  $h(\delta) - h(\epsilon)$  if  $\delta > \epsilon$ , and 0 otherwise.

- **Noisy broadcast channel plus public return channel:**

$\hat{C}_s(P_{Y|Z|X}) \leq \min\{\max_{P_X} \mathbf{I}(X; Y), \max_{P_X} \mathbf{I}(X; Y|Z)\}$   
 $\hat{C}_s(\epsilon, \delta) = h(\epsilon * \delta) - h(\epsilon)$ .

- **Satellite scenario:**

TODO

- **PUF types and their properties:**

General properties:

- The object can be subjected to a large number of diff challenges that yield an unpredictable response
- The object is very hard to clone physically
- Mathematical modeling of the challenge-response physics is very difficult
- Opaqueness: It is hard to characterize the physical structure of the object in a non-destructive way.

Types:

- **Coating PUF:** random layer of conductors and insulators: probes result in binary string. Used for secure key storage.
- **Optical PUF:** 3D optical structure produces speckle pattern. Challenge: props of laser beam: angle of incidence, focal dist.
- **Silicon PUF:** variations in IC from manufacturing. Challenge: pulsed time signal to certain part. Response: delay times of various wires and logic devices.
- **SRAM PUF:** Undefined state of RAM cells. Challenge: memory address, response: returned start-up values.
- **Randomly positioned glass fibers:** Challenge: ordinary beam of light lighting up part of the layer. Response: Certain fibers light up.

Uncontrolled PUF: reader interacts directly with PUF structure, trusted reader. Controlled PUF (CPUF): interaction through a *control layer*, PUF and *cl* are bound together, separation will damage the PUF. Result: attacker has no direct access to PUF. Example: secure key storage where control layer performs zk-protocol to prove knowledge of the key (called *Physically Obscured Key (POK)*).

- **PUF math:**

Information revealed by a noisy measurement outcome  $U'$  where  $m \in \mathcal{M}(m(K) = U)$ :  $\mathbf{I}(U'; K) = \mathbf{I}(U'; U)$ . Noiseless case:  $\mathbf{I}_m$

Measurable entropy of PUF (space  $\mathcal{K}$  and  $\mathcal{M}$ ):

$$\mathbf{I}_{\mathcal{P}, \mathcal{M}}^{\text{meas}} = \max_{m \in \mathcal{M}} H(m(k))$$

Security param of bare PUF: min num of C-R measurements required to reveal all measurable info of the PUF:  $S_{\mathcal{P}, \mathcal{M}_0}$

- **Fuzzy extractor:**

Gen and Rep algorithms.  $(S_x, W_x) = \text{Gen}(X)$ .  $S'_x = \text{Rep}(X', W_x)$   
 Must satisfy the following properties:

- **Correctness:** The prob that  $S'_x = S_x$  must be close to 1.
- **Security:** The RV  $S_x$  must be close to uniform, given knowledge of  $W_x$ .

- **Secure Sketch (for discrete src space  $\mathcal{X}$ ):**

$SS : x \mapsto w_x, \text{Rec} : (x', w_x) \mapsto \hat{x}$  with:

- **Correctness:** The prob that  $\hat{X} = X$  must be close to 1.

– **Security:**  $X$  given  $W_X$  must have high entropy.

- **When to use FE and SS:**

FE: reliably extract a cryptographic key from noisy data. SS: reliably extract a string with sufficient (min-)entropy. Easier to construct SS than FE, in general: SS extracts more (min-)entropy than a FE from the same source.

- **Code offset method (COM):**

Enroll (Gen):  $s \in \{0, 1\}^k, c_s = \text{Enc}(s)$ .  $w = c_s \oplus x$ . Output  $s$  as secret and  $w$  as helper data. Reconstruct (Rep):  $\hat{s} = \text{Dec}(x' \oplus w)$ .

- **Zero leakage FE (for continuous RV) based on partitions**

$\Pr[S = s | W = w] = 1/n$ . Enroll: determine which partition the measured val  $x$  is located in:  $\mathcal{A}_{ij}$ . Set  $s_x = i, w_x = j$ . Reconstruct:  $X'$  is measured, determine for which  $s'$  the interval  $\mathcal{A}_{s', w_x}$  is closest to  $x'$ . This  $s'$  is the reconstructed key.  $H_\infty(S|W = w) = \log n$ .  $H_\infty(S|W) = \log n$ .

- **Helper data schemes for specific PUF types:**

TODO

- **Distance bounding principles (and fraud types):**

**Mafia fraud:** challenge is relayed to different location where a legit. device is tricked into giving a response, response is relayed back to verifier. **Terrorist fraud:** legit. device cooperates with attacker, however auth secrets are not shared with the attacker.

(Light travels about 300m every  $\mu s$ ). If a device repeatedly correctly responds to an *unpredictable* challenge within time  $\Delta t$ , then the location where the response is computed cannot be further away than  $x = c\Delta t/2$ .

- **Brands-chaum protocol:**

- **Swiss knife protocol**

- **Analog impl.:**

- **Quantum stuff:**