

Physical Unclonable Functions (PUFs): Signal Processing and Information-Theoretic Aspects

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Outline

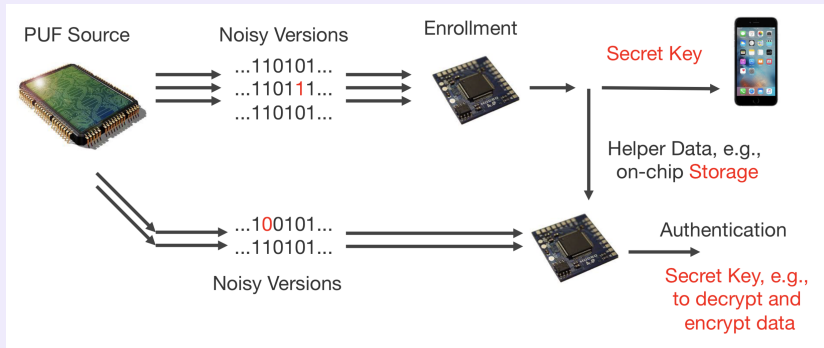
- **Part I:** Signal Processing for PUFs with Security Guarantees
- **Part II:** Optimal Error-Correcting Code Designs for PUFs

- **Signal Processing for PUFs with Security Guarantees**

Motivations for PUFs

Digital Secrecy and Privacy Example

- Digital devices have to be protected/secured/authenticated/identified, similar to individuals, to provide security for Things and their Internet (in order to protect device owners and businesses):
⇒ Hardware “Fingerprint” via a PUF

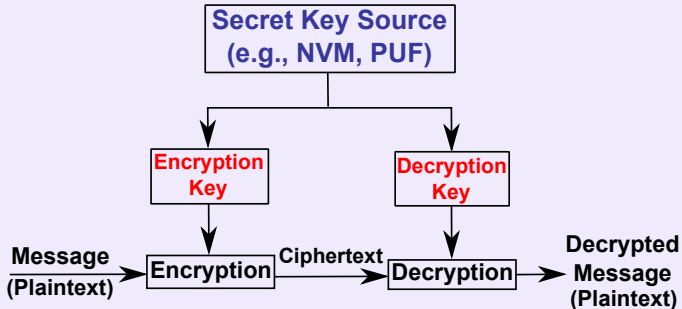


Motivations for PUFs II

Digital Secrecy and Privacy Example II

- Encryption/Decryption with PUFs

NVM= Non-Volatile Memory



Motivations for PUFs III

Other Applications of PUFs

- ▶ **6G mobile devices that use SRAM outputs**, available in mobile devices, as a PUF to extract secret keys;
- ▶ A **PUF in a USB token to encrypt user data** before uploading it to the cloud;
- ▶ System developers want to **mutually authenticate an FPGA chip and the IP components** in the chip, while IP developers want to **protect the IP**. One symmetric cipher and one PUF can achieve these.

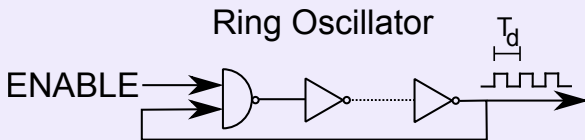
A Brief Definition of PUF

- ▶ A PUF is a challenge-response-mapping embodied by a physical device (e.g., digital circuit outputs) such that it is
 - ▶ **easy and fast** for the physical device **to evaluate** the PUF response;
 - ▶ **hard** for an attacker **to determine the PUF response** to a randomly chosen challenge, even if the attacker has access to a set of challenge-response pairs.
- ▶ PUFs are significantly **cheaper and safer alternatives** to storing keys in a non-volatile memory and they are the **only alternative** that fits perfectly to the hardware requirements of **IoT networks** to provide security to all digital devices, as well as to the whole network.
- ▶ PUFs can be used to also secure **6G networks**, **Cyber devices**, **Digital twins**, **Metaverse**, etc.

A Simple PUF with Continuous-Valued Outputs

Ring Oscillator (RO) PUFs

- A delay-based intrinsic PUF scheme uses the **random variations in the oscillation frequencies** of ROs to generate a secret key.

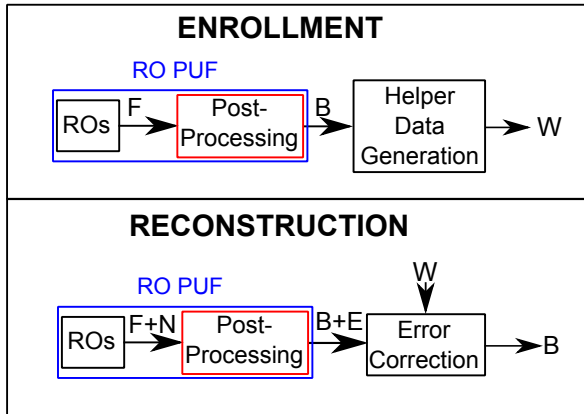


A Simple PUF with Continuous-valued Outputs II

RO PUFs

- ▶ **Source of randomness:** Uncontrollable silicon process variations on digital components' delays;
- ▶ **Hard macro designs** are used for each RO: identical implementations;
- ▶ **Temperature and voltage effects** are orders of magnitude greater than the random variations in RO outputs;
- ▶ **Correlations** in RO outputs decrease entropy in the extracted bit sequence;
- ▶ There is **noise** in every measurement of digital circuits.

Secret Key Generation with RO PUFs



F: Real-valued Oscillation Frequencies

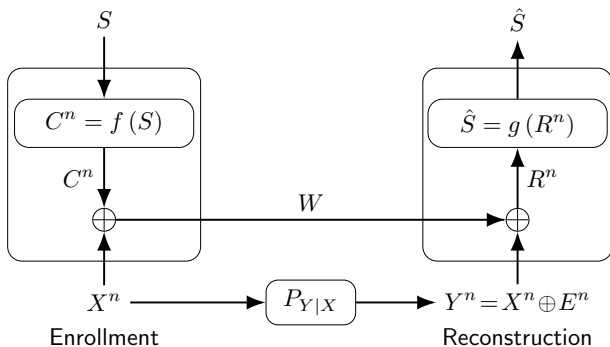
B: Uniform Bit Sequence

W: Side Information

N: Noise

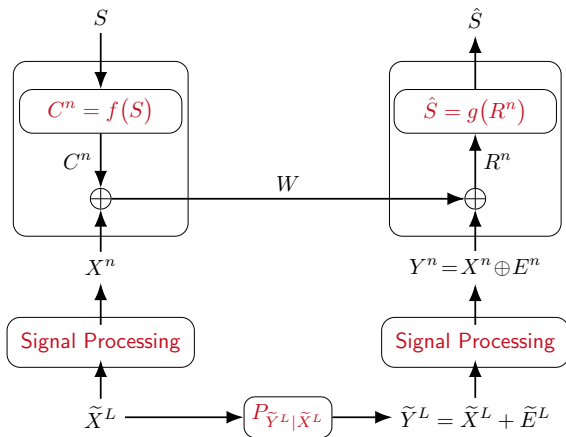
E: Error Vector

Fuzzy Commitment Scheme



- ▶ Secret key S and helper data W have to be independent,
- ▶ Block error probability should satisfy, e.g., $P_B = \Pr[S \neq \hat{S}] \leq 10^{-9}$;
- ▶ S should be uniformly random with entropy, e.g., $H(S) \geq 256$ bits.

Main Components to Design

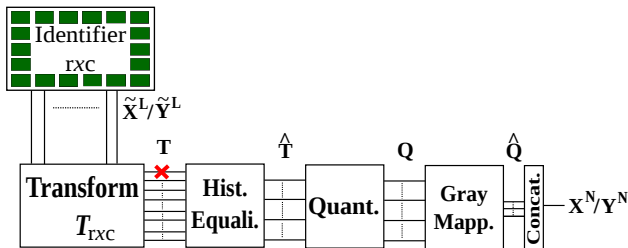


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How to Solve the 3 Problems Simultaneously

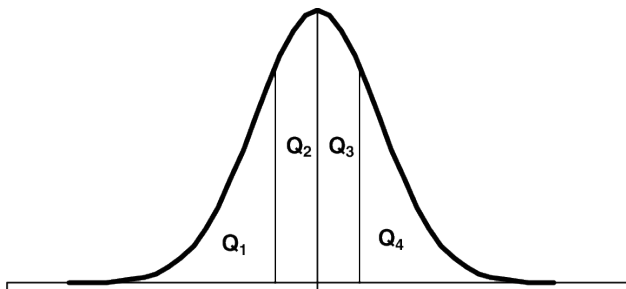


- Apply a **transform** $T_{rxc}(\cdot)$ to decorrelate \tilde{X}^L ;
- Histogram equalization converts each transform-coefficient T output into a **standard (Gaussian) random variable**;

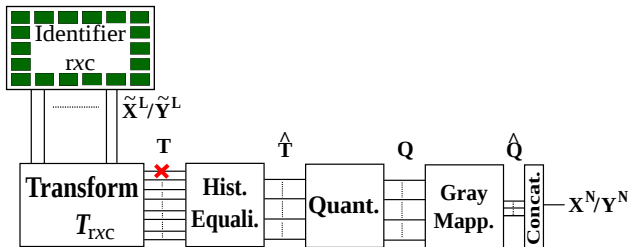
How to Solve the 3 Problems Simultaneously II

- Apply **scalar quantizers** that satisfy the **uniformity** property:

$$\Pr[\text{Quant}(\widehat{T}_i) = (q_1, q_2, \dots, q_{K_i})] = \frac{1}{2^{K_i}} \quad \text{for} \quad i = 1, 2, \dots, L \quad (1)$$



How to Solve the 3 Problems Simultaneously III



- The noise components have zero mean, so use Gray mapping to extract bit sequences from quantized outputs;
- Concatenate all extracted bits to obtain X^N/Y^N ;
- Error symbols $E_i = X_i \oplus Y_i$ need not be independent or identically distributed.

PUFs with Security Guarantees

- ▶ Model transform coefficients T as, e.g., **truncated Gaussian distributions**. Then, quantization boundaries, when $m \geq 1$ bits are extracted from a transform coefficient, are

$$b_k = Q^{-1} \left(Q(b_0) \cdot \left(1 - \frac{k}{2^m} \right) + Q(b_{2^m}) \cdot \frac{k}{2^m} \right) \quad \text{for } k = 1, 2, \dots, (2^m - 1) \quad (2)$$

- ▶ Consider any transform coefficient realization $T = t$ that is on the quantization boundary
 - ⇒ Error probability is 0.5, so one cannot guarantee secrecy for all devices!!
 - ⇒ **Current PUF products** only provide security guarantees for the **average over all devices**, so millions of intelligent IoT devices are likely **vulnerable** to malicious attacks!!

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PUFs with Security Guarantees II

- ▶ We propose to eliminate realizations

$$\bar{t} \in ((b_k - \delta/2), (b_k + \delta/2)] \quad (3)$$

for all transform coefficients and for some fixed $\delta \geq 0$ that is called the **Quality-of-Security-Service (QoSS) parameter** for all PUF outputs that are used for secure and private device authentication.

PUFs with Security Guarantees III

- Denote the **ratio of eliminated realizations vs. all realizations** as

$$\gamma(\delta) = \frac{\sum_{k=1}^{(2^m-1)} \left(Q\left(b_k - \frac{\delta}{2}\right) - Q\left(b_k + \frac{\delta}{2}\right) \right)}{Q(b_0) - Q(b_{2^m})}. \quad (4)$$

- The **percentage of realizations \bar{t} used for secret key agreement**, i.e., **manufacturing yield**, is

$$\beta(\delta) = 100 \times (1 - \gamma(\delta)) \quad (5)$$

which **decreases for increasing QoSS parameter δ**

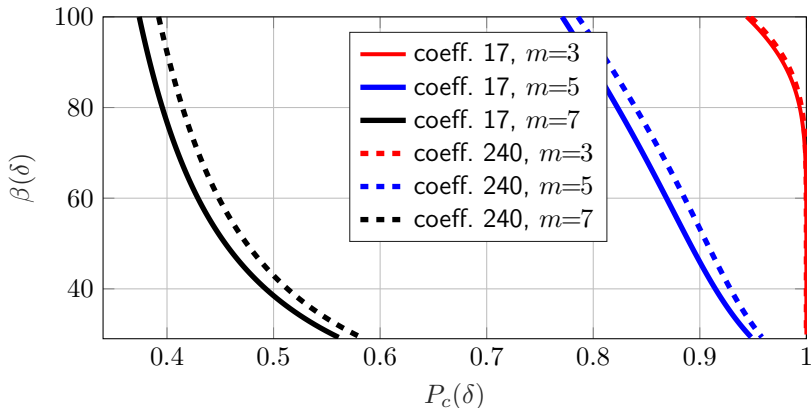
⇒ **Stringent trade-off to be optimized!!**

PUFs with Security Guarantees IV

- Define a reliability metric called **correctness probability** P_c that measures the probability that all bits are correct.
- P_c **increases for increasing** δ .

PUF Results from an RO Dataset

- ▶ $P_c(\delta)$ vs. $\beta(\delta)$ when a selected low-complexity transform with m -bit uniform quantization is applied to 16×16 RO arrays.
- ▶ We have $\beta = 100$ and P_c at its minimum when $\delta = 0$.



- **Optimal Error-Correcting Code Designs for PUFs**

Main Contributions

- We propose **binning**-based code constructions that are **Pareto optimal** and **improve on all** existing methods [GIS+'19].
- **Polar codes** designed for RO and SRAM PUFs **achieve rate tuples that cannot be achieved by existing methods**.
- **Significant performance improvements** are illustrated for
 - Multiple PUF measurements [GKS'15, GK'18],
 - Adaptive PUF measurements [KGS+'16, GKS+'18],
 - Multiple PUF enrollments (uses) [KGW'18],
 - Multiple rounds of communication [GGK'18], etc.

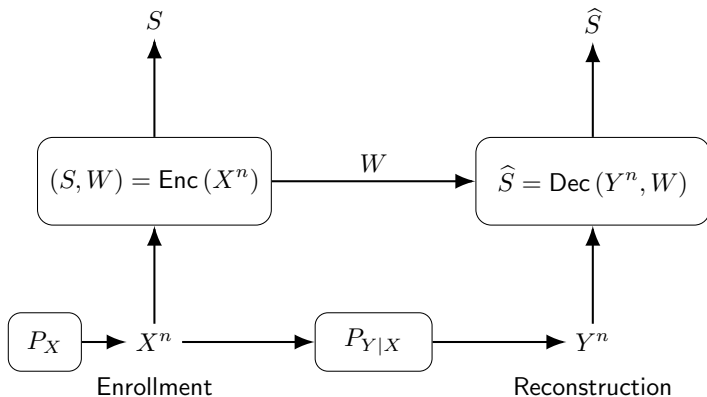
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Problem Formulation – Generated-Secret (GS) Model



Notation

- ▶ The Shannon entropy measures the uncertainty, i.e.,

$$H(X) = - \sum_{x \in \mathcal{X}} P_X(x) \log(P_X(x)).$$

- ▶ Mutual information measures statistical dependency, i.e.,

$$I(X;Y) = H(X) - H(X|Y).$$

- ▶ Note that min-entropy is relevant to but different from the Shannon entropy, and we can discuss why the Shannon entropy is the right metric for secret key generation from PUFs/biometrics with nested codes!

Key-Leakage-Storage Region

Definition

A key-leakage-storage tuple (R_s, R_ℓ, R_w) is *achievable* if, given any $\epsilon > 0$, there is some $n \geq 1$ for which $R_s = \frac{\log |\mathcal{S}|}{n}$ and

$$\Pr[\hat{S} \neq S] \leq \epsilon \quad (\text{reliability}) \quad (6)$$

$$\frac{1}{n} I(S; W) \leq \epsilon \quad (\text{secrecy leakage}) \quad (7)$$

$$\frac{1}{n} H(S) \geq R_s - \epsilon \quad (\text{key uniformity}) \quad (8)$$

$$\frac{1}{n} I(X^n; W) \leq R_\ell + \epsilon \quad (\text{privacy leakage}) \quad (9)$$

$$\frac{1}{n} \log |\mathcal{W}| \leq R_w + \epsilon \quad (\text{public storage}). \quad (10)$$

Key-Leakage-Storage Region (Cont'd)

Theorem (Ignatenko and Willems'09)

The key-leakage-storage region for the GS model is

$$\begin{aligned} \mathcal{R}_{gs} = \bigcup_{P_{U|X}} \{ & (R_s, R_\ell, R_w) : 0 \leq R_s \leq I(U; Y), \\ & R_\ell \geq I(U; X) - I(U; Y), \\ & R_w \geq I(U; X) - I(U; Y) \text{ for} \\ & P_{UXY} = P_{U|X} P_X P_{Y|X} \}. \end{aligned} \quad (11)$$

Main Existing Methods

- *Code-offset fuzzy extractors* (COFE) [Dodis et al.'08] for the GS model,
- *Fuzzy-commitment scheme* (FCS) [Juels and Wattenberg'99] for the chosen-secret (CS) model,
- *Syndrome-based Polar Code Construction* [Chen et al.'17] for the GS model.

Existing Methods (Cont'd)

- ▶ COFE and FCS result in a storage rate of 1 bit/symbol since they apply **one-time padding**.
- ▶ Syndrome-based polar code construction
 - ▶ improved on existing methods because it is a **Slepian-Wolf coding (i.e., lossless compression)** construction,
 - ▶ achieves only a single point on the region \mathcal{R}_{gs} boundary.
- ▶ We now show that our **Wyner-Ziv (WZ)-coding (i.e., lossy compression) constructions** [Shamai et al.'98, Korada et al.'10] are **Pareto-optimal**.

WZ-Coding with Random Linear Codes (RLCs)

Assume

- ▶ $X^n \sim \text{Bern}^n\left(\frac{1}{2}\right)$, i.e., $\Pr[X_i = 1] = \Pr[X_i = 0] = 0.5$ for all $i = 1, 2, \dots, n$.
- ▶ $P_{Y|X}$ is a binary symmetric channel (**BSC**) with crossover probability p_A , i.e., $\Pr[Y \neq X] = p_A$.

WZ-Coding with RLCs (Cont'd)

- ▶ Choose **uniformly at random** full-rank parity-check matrices \mathbf{H}_1 , \mathbf{H}_2 , and \mathbf{H} as

$$\mathbf{H} = \begin{bmatrix} \mathbf{H}_1 \\ \mathbf{H}_2 \end{bmatrix} \quad (12)$$

- ▶ $\mathbf{H}_1 \in \{0, 1\}^{m_1 \times n}$ defines a binary linear code \mathcal{C}_1 with parameters $(n, n - m_1)$, meaning that the codewords have size n and there are $2^{n - m_1}$ codewords.
- ▶ $\mathbf{H} \in \{0, 1\}^{(m_1 + m_2) \times n}$ defines a binary linear code \mathcal{C} with generator matrix \mathbf{G} and parameters $(n, n - m_1 - m_2)$,
- ▶ Codes are **nested**, i.e., $\mathcal{C} \subseteq \mathcal{C}_1$.

WZ-Coding with RLCs (Cont'd)

- ▶ Impose the conditions, for some $q \in [0, 0.5]$ and $\delta > 0$ (not to be confused with the QoS parameter),

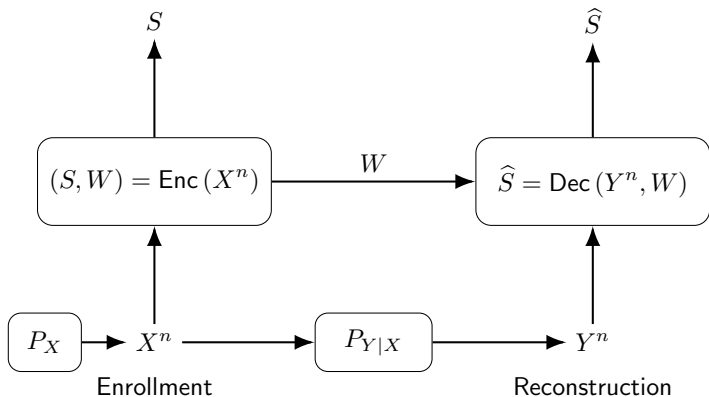
$$\frac{k_1}{n} \triangleq \frac{n - m_1}{n} = 1 - H_b(q) - \delta, \quad (13)$$

$$\frac{k}{n} \triangleq \frac{n - m_1 - m_2}{n} = 1 - H_b(q * p_A) - 2\delta \quad (14)$$

where

- ▶ $H_b(q) = -q \log q - (1 - q) \log(1 - q)$,
- ▶ $q * p_A = q(1 - p_A) + (1 - q)p_A$.

Problem Formulation (Recall)



WZ-Coding with RLCs (Cont'd)

Enrollment:

- ▶ Observe X^n and find the codeword $X_q^n \in \mathcal{C}_1$ such that

$$X_q^n = \arg \min_{C^n \in \mathcal{C}_1} d_H(X^n, C^n) \quad (15)$$

where $d_H(\cdot)$ is the Hamming distance,

- ▶ Error sequence $X^n \oplus X_q^n \triangleq E_q^n \sim \text{Bern}^n(q)$ when $n \rightarrow \infty$,
- ▶ Assign $W = X_q^n \mathbf{H}_2^T$ as **helper data** since $X_q^n \mathbf{H}^T = [0 \ W]$,

WZ-Coding with RLCs (Cont'd)

Enrollment (Cont'd):

- ▶ Sum X_q^n with the sequence L_W^n that is in the same coset as X_q^n and that has the minimum Hamming weight. The sum is $X_q^n \oplus L_W^n = X_c^n \in \mathcal{C}$,
- ▶ Assign the secret key S such that $X_c^n = SG$,

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Enrollment (Cont'd):

- ▶ Codewords in blue and green belong to \mathcal{C}_1 ,
- ▶ Codewords in green belong to \mathcal{C} .

	$W = 0$	$W = 1$...	$W = 2^{m_2} - 1$
$S = 0$	$C_0^n = L_0^n = 00..0$	$L_1^n = 00..1$...	$L_{2^{m_2}-1}^n = 01..1$
\vdots	\vdots	\vdots	\ddots	\vdots
$S = s$	C_s^n	$C_s^n \oplus L_1^n$...	$C_s^n \oplus L_{2^{m_2}-1}^n$
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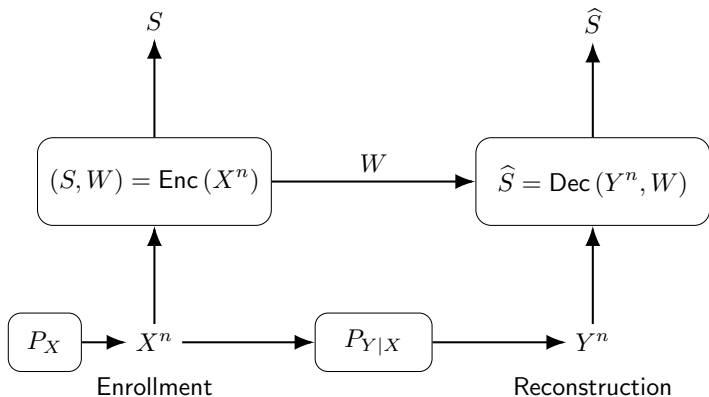
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Problem Formulation (Recall)



WZ-Coding with RLCs (Cont'd)

Reconstruction:

- ▶ The channel $P_{Y^n|X_q^n} \sim \text{Bern}^n(q * p_A)$ when $n \rightarrow \infty$,
- ▶ \mathcal{C} can correct errors in $P_{Y^n|X_q^n}$ with high probability to estimate X_q^n ,
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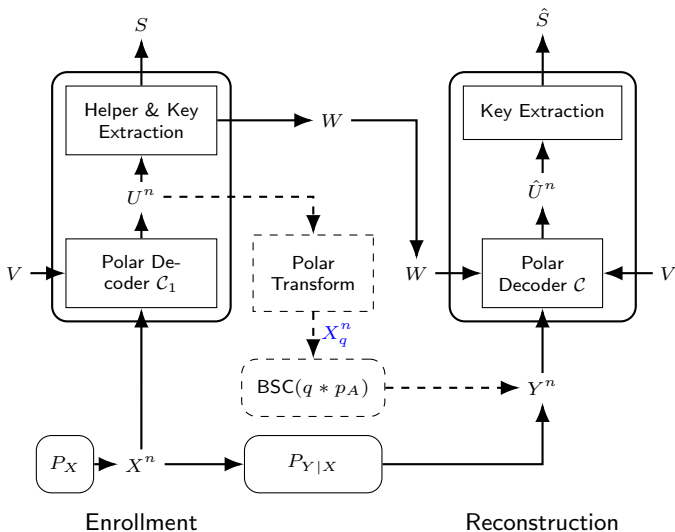
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WZ Polar Code Construction

Polar Codes (used in the 5G wireless communications standard)

- A polar transform converts an input sequence U^n with frozen and unfrozen bits to a codeword X^n .
- Polar codes [Arıkan'08] rely on **converting** the physical channel $P_{Y|X}^n$ **into virtual channels** $P_{Y^n U^{i-1}|U_i}$.

WZ Polar Code Construction (Cont'd)



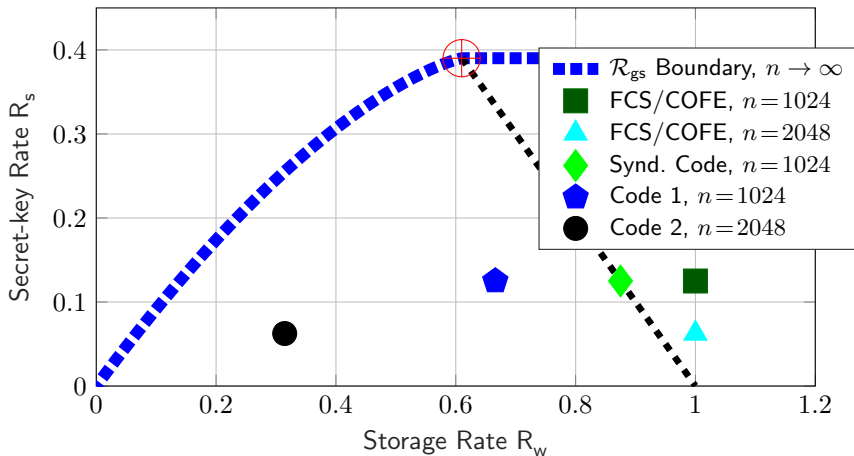
Rate-Tuple Comparisons

- ▶ Key length 128 bits,
- ▶ Block error probability $P_B = 10^{-6}$,
- ▶ $P_{Y|X} \sim \text{BSC}(p_A = 0.15)$.
- ▶ Design nested polar codes in combination with successive cancellation list (SCL) decoders with list size 8.

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- ▶ Block error probability $P_B = 10^{-6}$,
- ▶ $P_{Y|X} \sim \text{BSC}(p_A = 0.15)$.
- ▶ Design **nested polar codes** in combination with successive cancellation list (**SCL**) decoders with **list size** 8.

Rate-tuple Comparisons (Cont'd)



Conclusion

- We proved that **security guarantees** can be given (**unlike some of the current PUF products**) to **every digital device with a PUF**;
- Illustrated that **by removing a small percentage** of manufactured PUF circuits, it is possible to **significantly simplify the error-correcting code design** due to increased reliability;
- Conversely, we proved that **without removing any PUF** circuit output, **no security guarantee per PUF**;
- We illustrated that the proposed nested linear block code constructions are **optimal for key extraction from PUFs**;
- We estimated the information leakage of nested polar codes by using neural estimators to showcase that **practical PUF designs with our proposed code constructions leak only a negligible amount of information**.

THANK YOU!

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