

# Image-based PUFs for Anti-Counterfeiting

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TELE Lab  
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# Outline

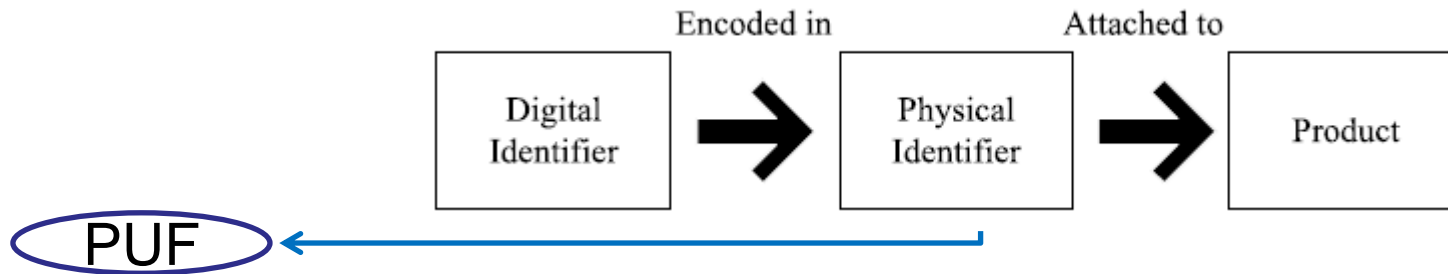
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- Anti-counterfeiting and unclonability
- Image-based PUFs
  - Laser-Written PUF
  - Crystal PUF
- Image-based Physical Function System
  - Image-based Physical Function
  - Image-based Extraction
- Experiments
  - Image-based Extraction
  - Robustness
  - Physical Unclonability



# Anti-counterfeiting

## State of the art

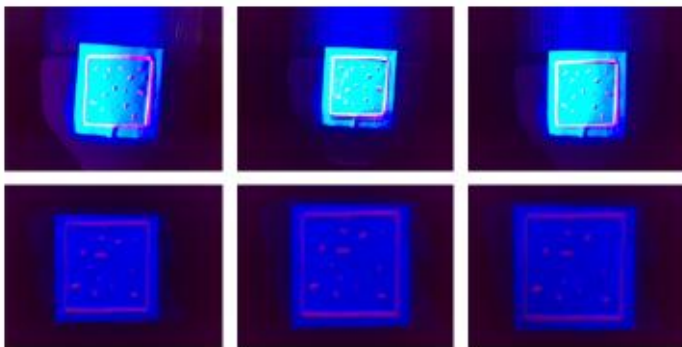
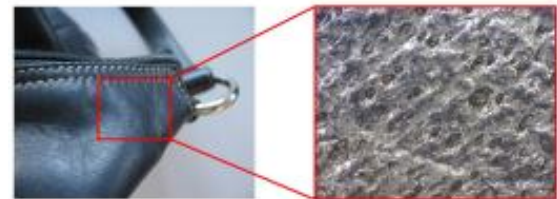
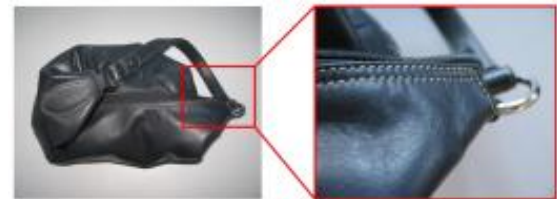
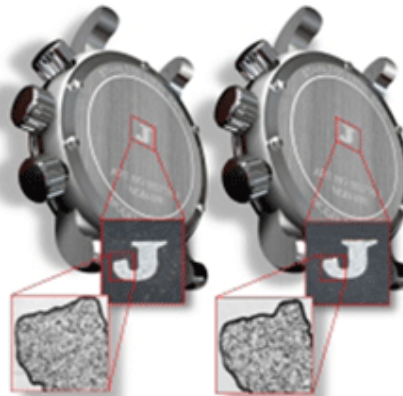
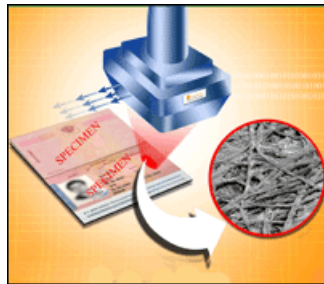


- Two types of physical identifiers:
  - Overt: No need to a specific device
    - Can be easily cloned
  - Covert: Retrieving digital identifier needs the specific device

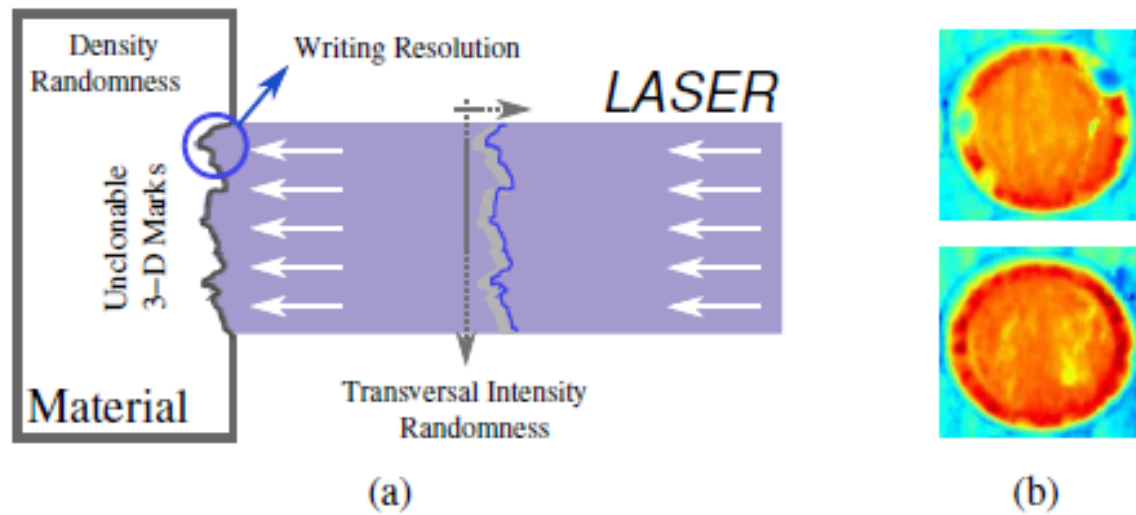


# Image-based PUFs

- Using Random features : Hardly-cloneable
- Random features characterized by imaging methods.



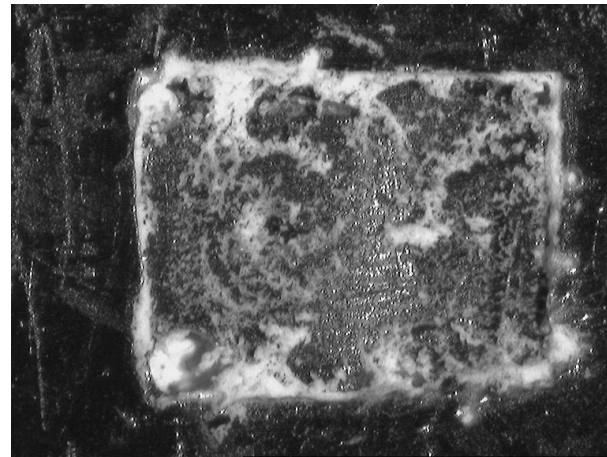
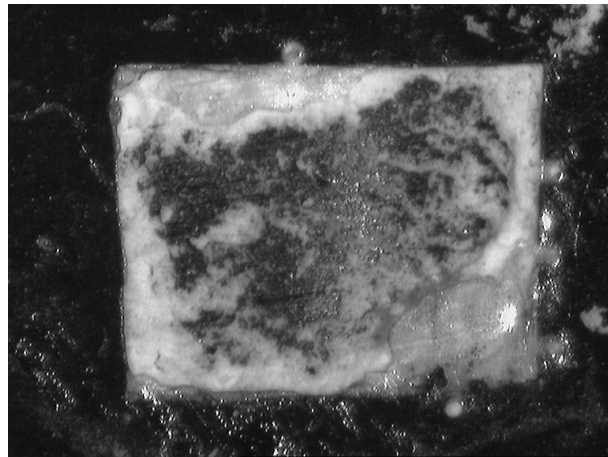
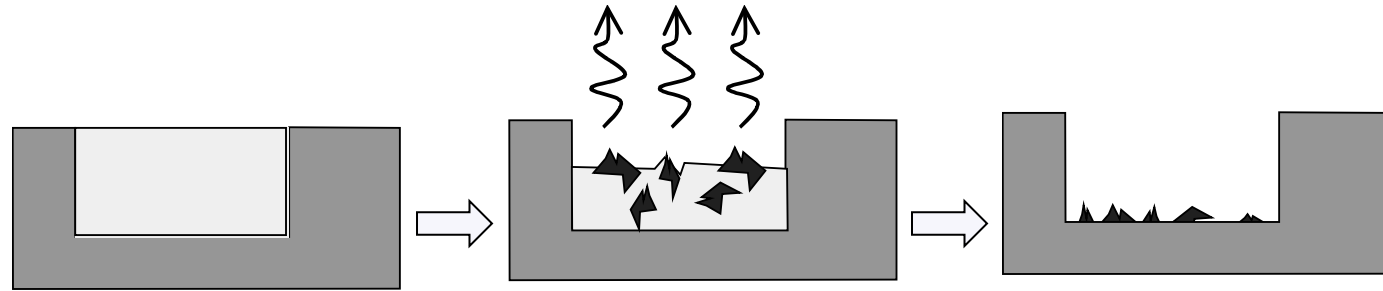
# Laser-Written PUF



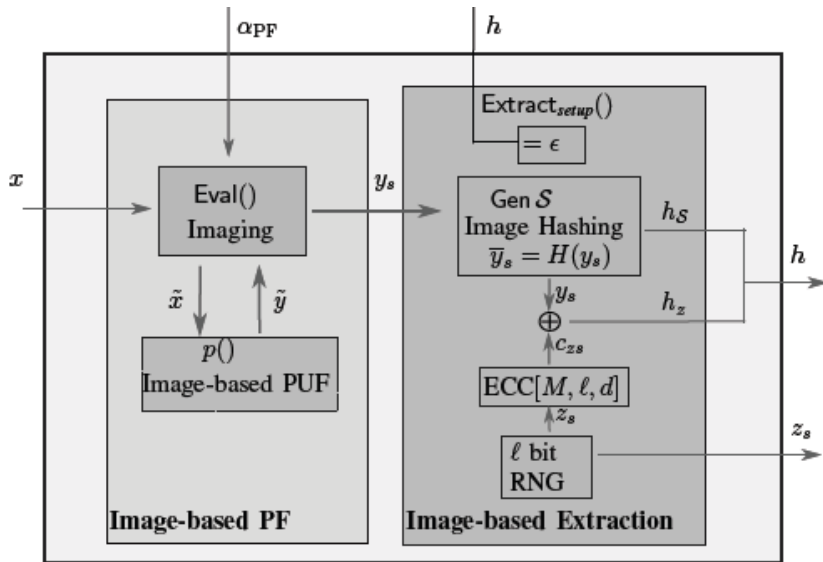
(a) Laser engraving principle

(b) Two laser marks

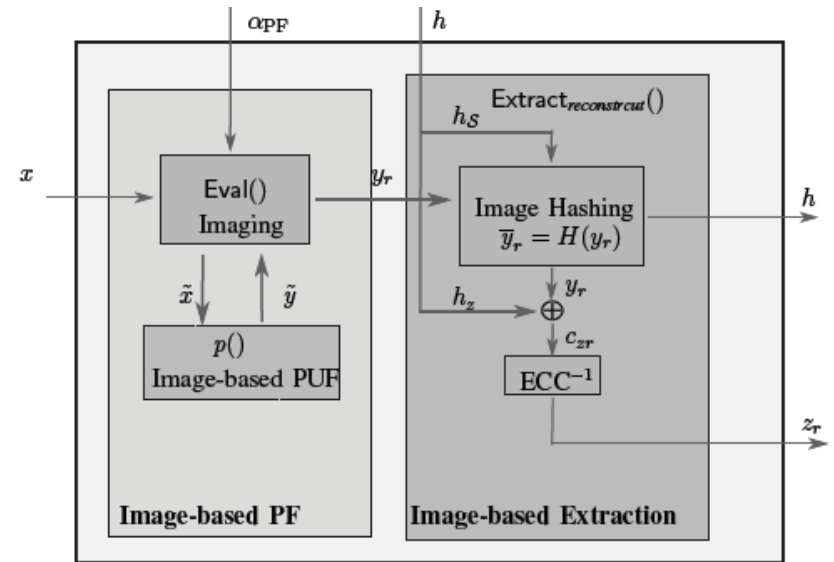
# Crystal PUF



# Image-based PF System Model



(a) Setup mode.



(b) Reconstruction mode.

# Image-based Extraction: Challenges

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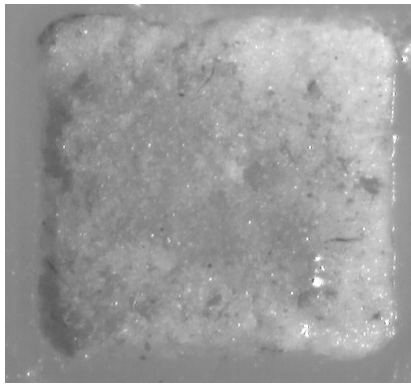
- Short Binary identifier instead of a big Image.
- Being robust to observation noise
- Not loosing the randomness present between images.



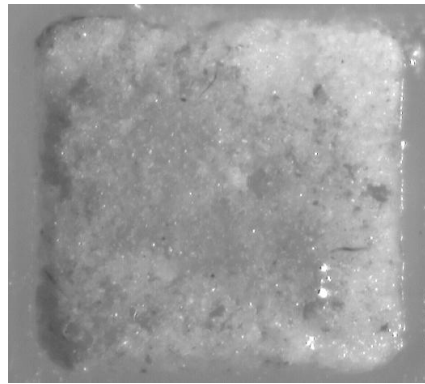


# Image-based Extraction

PUF1



PUF1+noise



PUF2

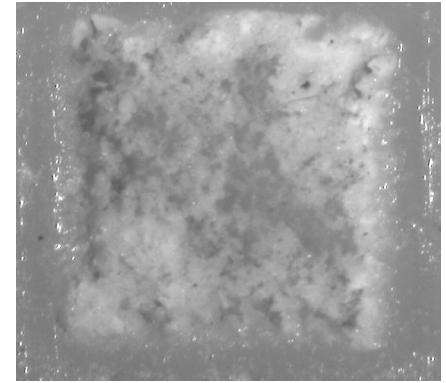


Image Hashing

Fuzzy Extraction

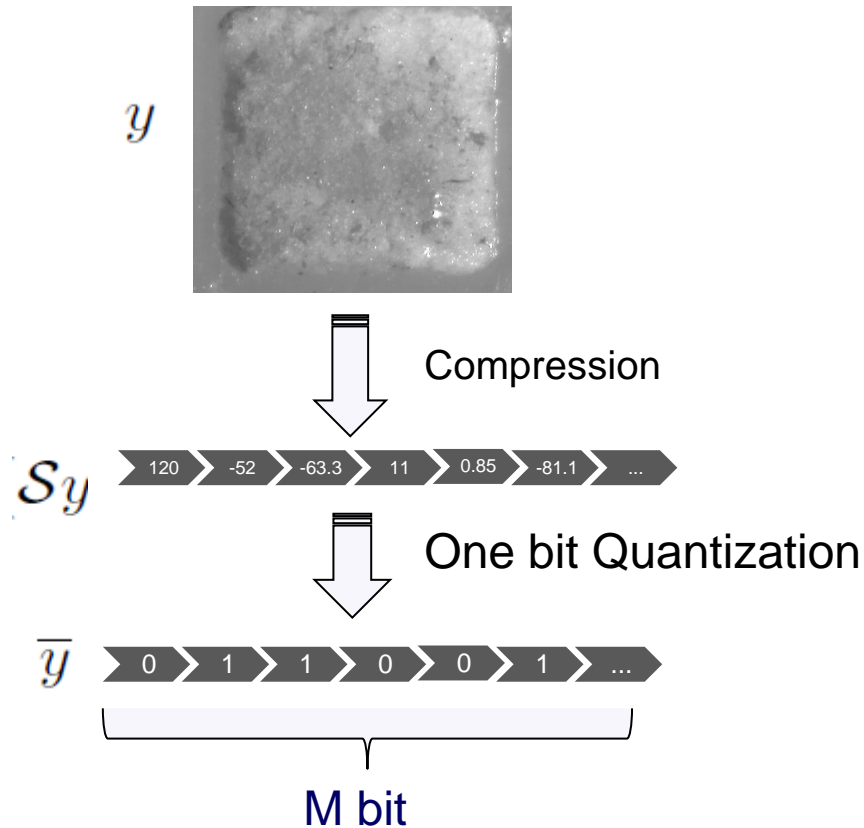
Fingerprint

Identifier



# Image Hashing

$$h : \mathbb{R}^N \rightarrow \mathcal{B}^M, y \mapsto \bar{y} = H(y) = \text{sign}_b(\mathcal{S}y),$$



# Adaptive Image Hashing-Theory

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- Let  $y = \sum_i \alpha_i \psi_i = \Psi \alpha$  be sparse in a certain basis (Fourier, Wavelet, Gabor, Curvelet),
- By sparse it means that the entries of  $\alpha = [\alpha_1, \alpha_2, \dots, \alpha_N]$  when sorted from largest to smallest, decay rapidly to zero.
- Adaptive compressing of the signal accounts for a  $K$ -term approximation, consisting of the terms  $\alpha$  with the  $K$  largest magnitudes.



# Non-adaptive Image Hashing--Theory

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- A signal that is sparse in one basis can be recovered from  $M = O(K \log N/K)$  non-adaptive linear projections onto a basis  $\Phi$  that is incoherent with the sparsity basis.

- So the signal is approximated by

$$y_M = \Phi y$$

- For instance, random matrices that are incoherent with most known basis functions have most typically used to perform the non-adaptive Compressed Sensing.



# Random Binary Hashing

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- A random matrix  $\Phi = (\Phi_{ij}) \in \mathbb{R}^{M \times N}$  such as Gaussian matrix  $\Phi_{ij} \sim_{\text{iid}} \mathcal{N}(0, 1/M)$  is used as the sensing matrix  $\mathcal{S}$ .

$$\mathbb{P}_{\Phi} \left[ \left| \frac{1}{M} \text{dist}_{\text{H}}(\bar{y}, \bar{w}) - \text{dist}_{\angle}(y, w) \right| \geq \beta \right] \leq 2e^{-2\beta^2 M}.$$

A fast and efficient hashing of the images.

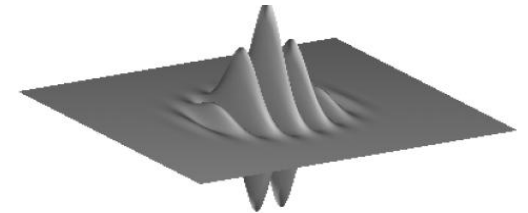


# Gabor Binary Hashing

- Sparsity Basis : Gabor
- Gabor Basis : a complete family of localized, oriented, bandpass atoms.

$$g_{\psi}(t) = \frac{1}{a\sqrt{2\pi}} \sin(\nu \cdot (t - k)) \exp\left(-\frac{1}{4a^2} \|t - k\|^2\right),$$

$$\langle g_{\psi}, y \rangle = \sum_{i=1}^N g_{\psi_i} y_i \cong G_{\psi}^T y,$$



$$\Psi(a, \nu_0, \Delta) = \left\{ (a, \nu_f, b_m) : \right. \\ \left. \begin{aligned} \nu_f &= \nu_0(\sin \theta_f, \cos \theta_f), \quad \theta_f = 2\pi f/F, \quad 0 \leq f < F, \\ b_m &= (m_1\Delta, m_2\Delta), \quad 0 \leq m_i < \lfloor N_i/\Delta \rfloor, \quad i \in \{1, 2\} \end{aligned} \right\},$$

# Gabor Binary Hashing

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- Let  $\psi^{(i)}$  be the parameter vector pointing out the strongest value of Gabor coefficients.
- Sensing matrix is defined:

$$S_G = [G_{\psi^{(1)}}, G_{\psi^{(2)}}, \dots, G_{\psi^{(M)}}]^T$$



# Experiments -LPUF

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- Robustness  $\rho_{\text{PFS}} = \Pr[z_r = z_s],$

- Database: R=20, Q=60

$$\mathcal{S}_1 = \{y_{rq}, 1 \leq r \leq R, 1 \leq q \leq Q\} \subset \mathbb{R}^N$$

- Physical Unclonability

$$PU_{ex\text{PFS}} = \Pr[z_r = z_s : p \neq p'].$$

- Database: R=1000

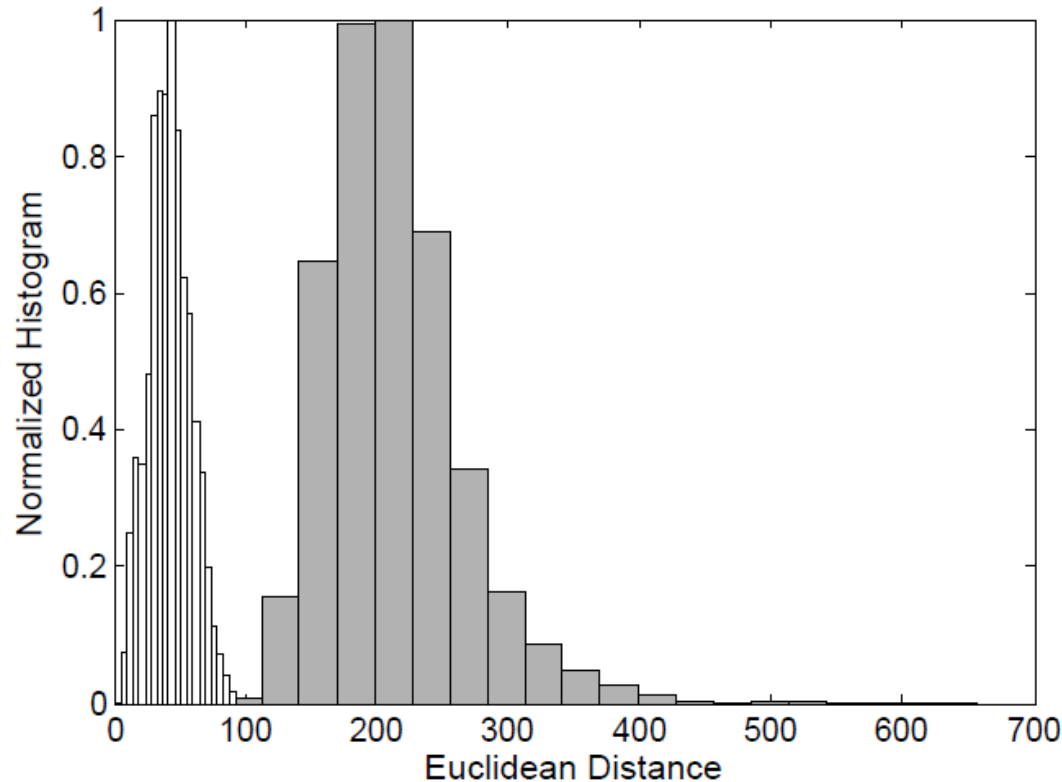
$$\mathcal{S}_2 = \{y_r, 1 \leq r \leq R\} \subset \mathbb{R}^N$$



# Preliminary analysis of database

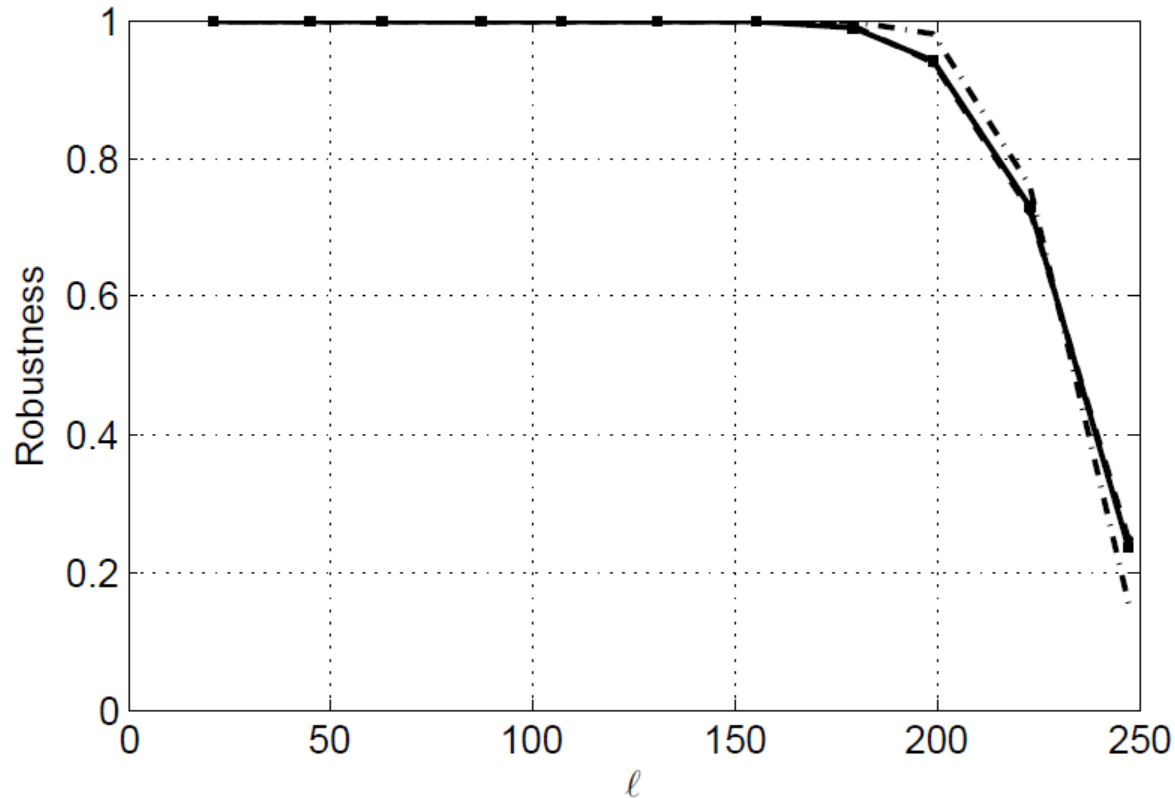
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## Intra and Inter-class Histogram of



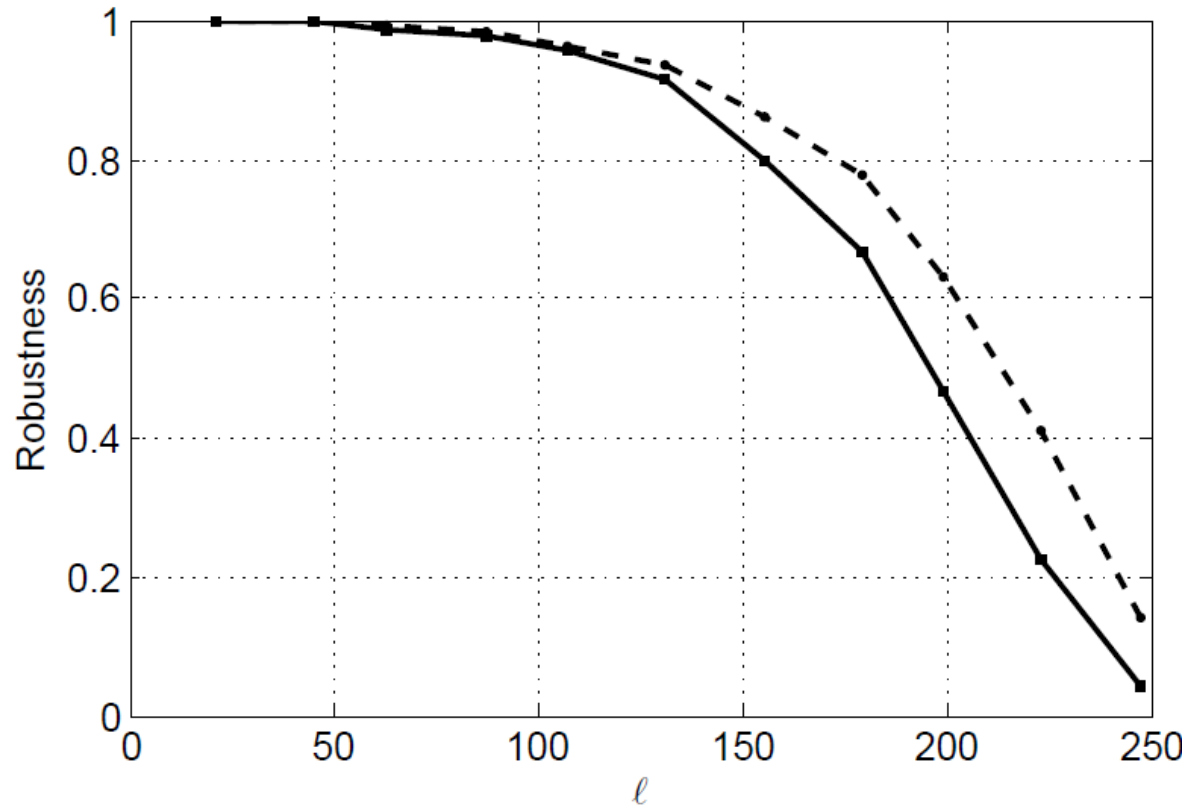
# Robustness vs. Length of Identifier

RbEX



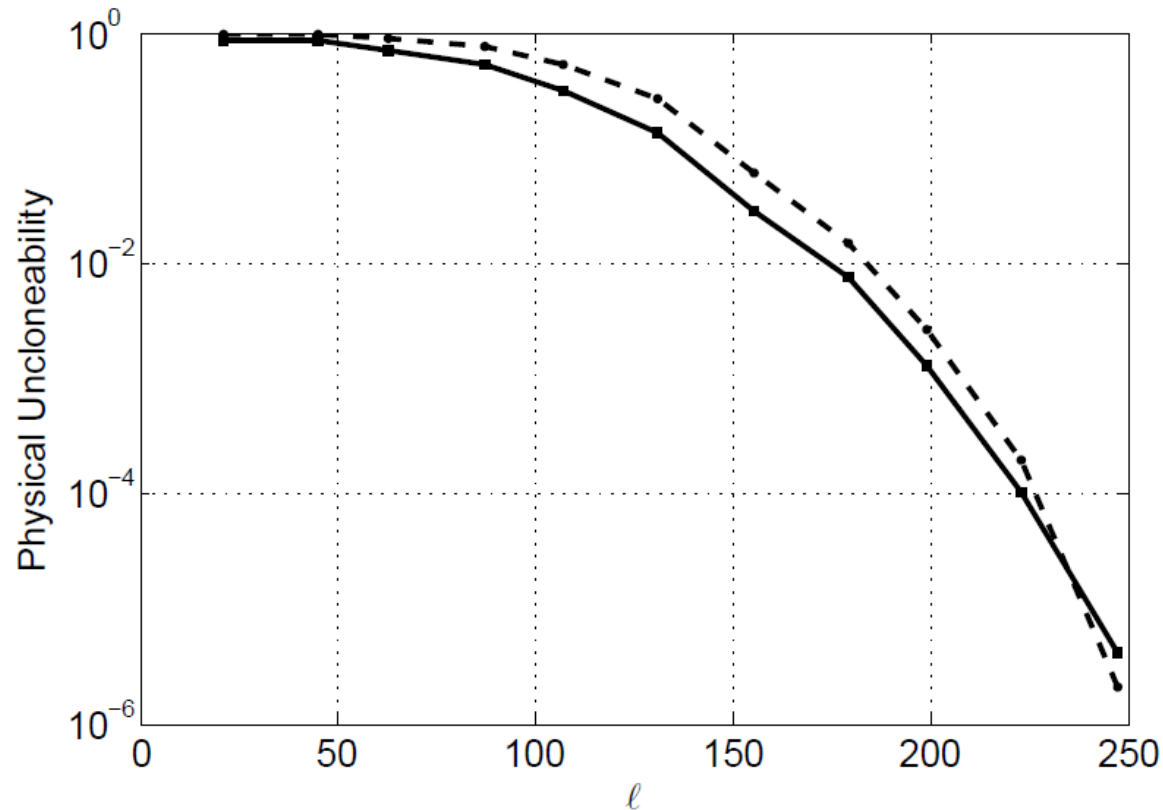
# Robustness vs. Length of Identifier

GbEX



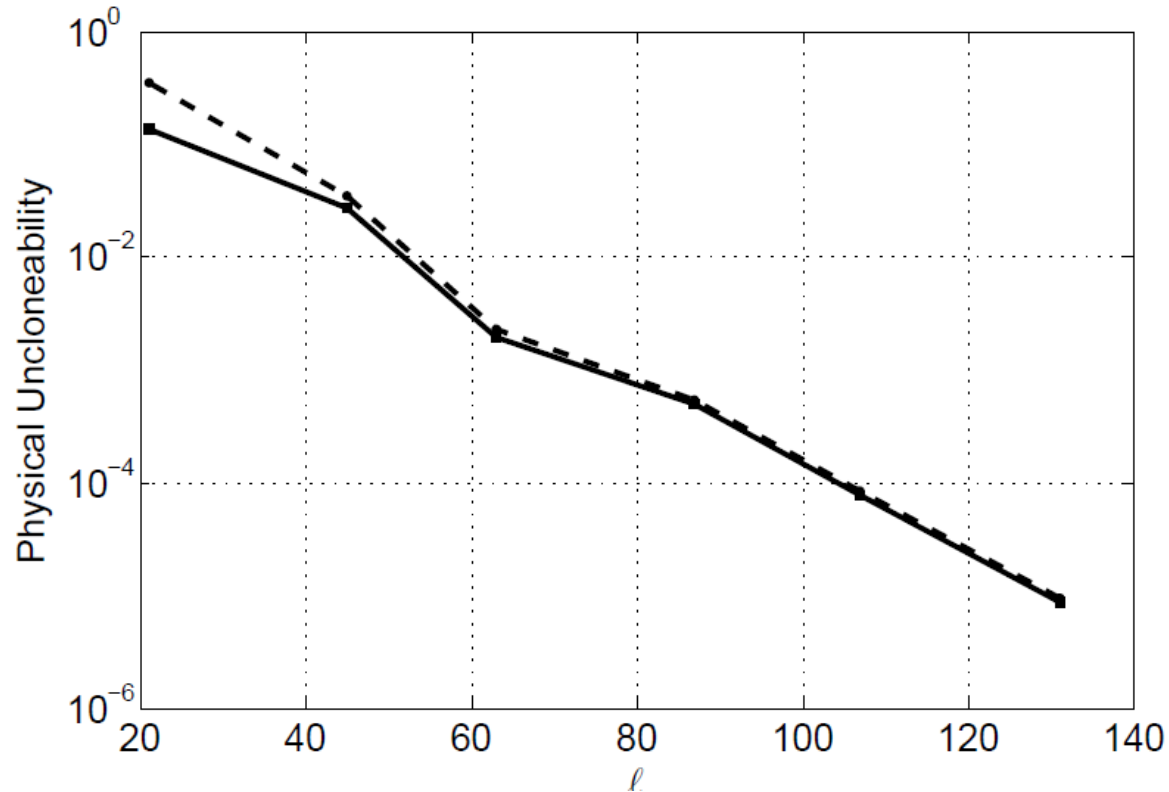
# Physical Unclonability vs. Length of Identifier

RbEX



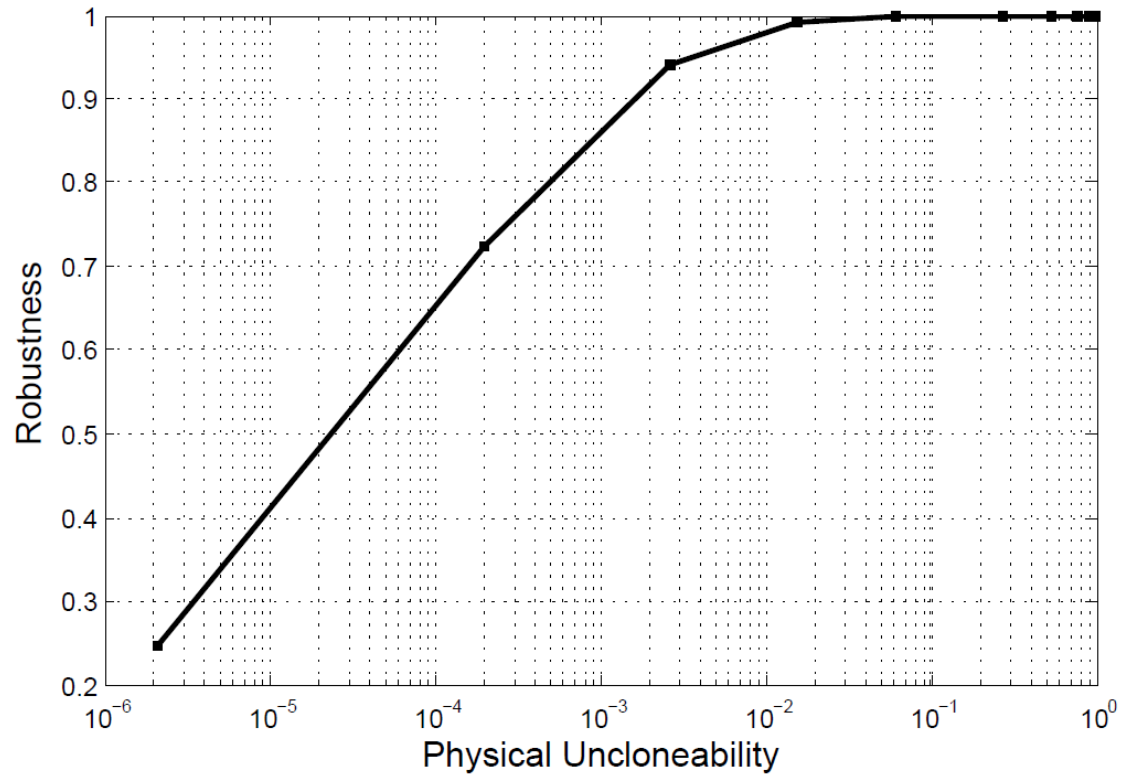
# Physical Unclonability vs. Length of Identifier

GbEX



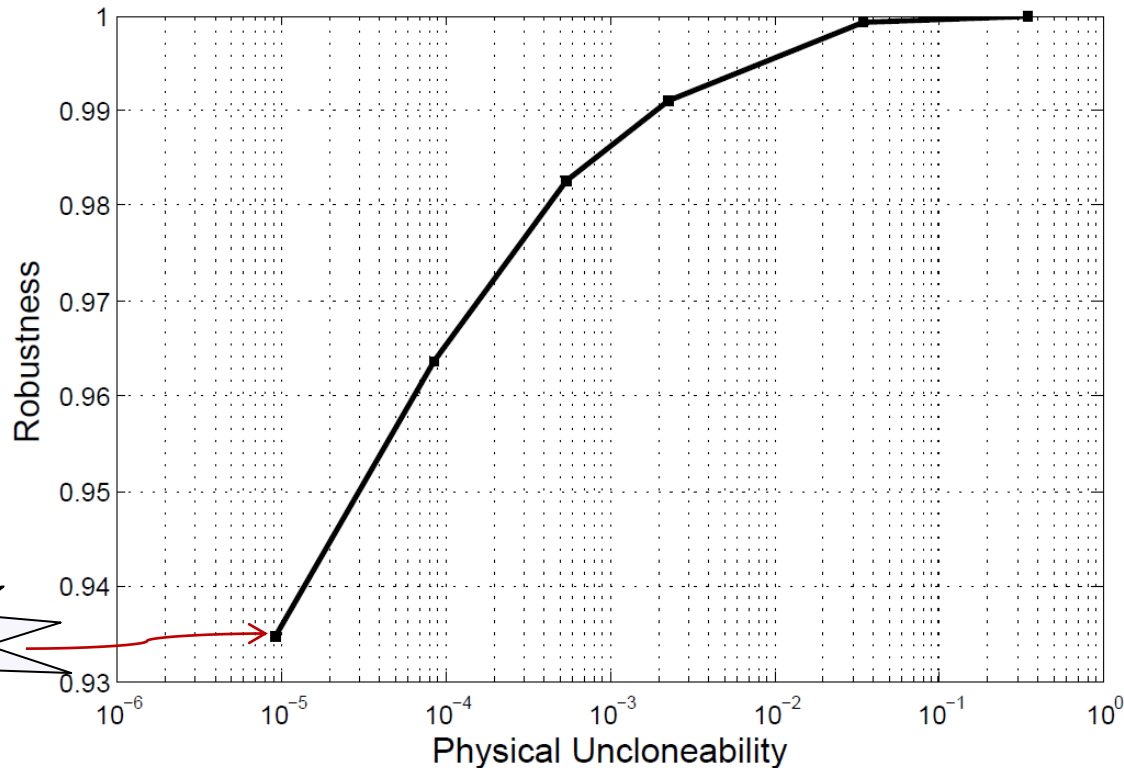
# Robustness vs. Physical Unclonability

RbEX



# Robustness vs. Physical Unclonability

GbEX



0.94,  $10^{-5}$



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Thank you

