

# **Privacy-preserving methods: multimedia and biometrics perspectives**

**S. Voloshynovskiy**

<http://sip.unige.ch/education/multimedia-security/>

# Lecture Outline

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- Introduction: privacy and security in biometrics and multimedia
- Cryptographic approach
- Signal processing approach
- Hybrid approach

# Information security

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- Personal Identifier Number (PIN): to control access to service systems and information



- Subscriber Identity Module (SIM): access to mobile phones, to protect against illegal use of network



- Secure Socket Layer (SSL): to prevent eavesdropping on information exchanged between two terminals



# Traditional cryptography

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**Cryptography**: develops tools for protecting information and establishing secure interactions.

- Encryption: provides confidentiality of information
- Authentication: verifies the identity

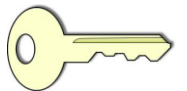
Note: The algorithms are public, therefore, the security of the schemes depends on the secrecy of a key owned by the legitimate parties.

## **Practical cryptography issues in our modern society:**

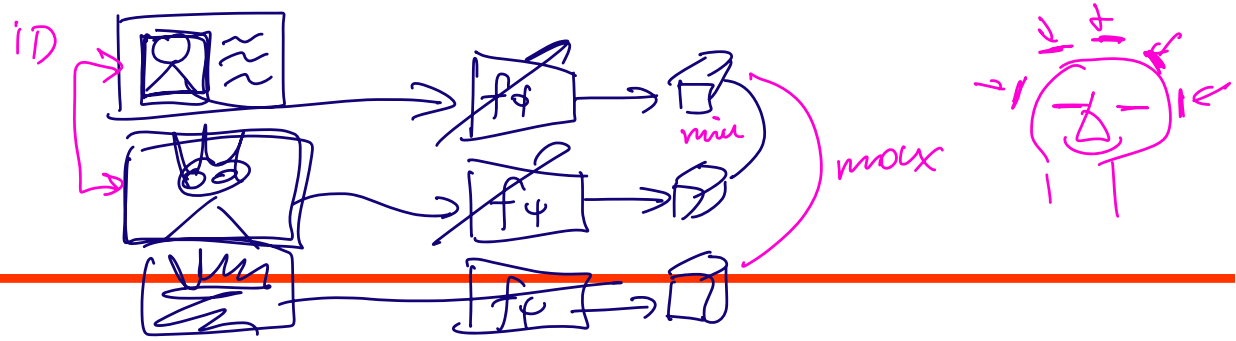
1 Key generation and distribution

2 Safely storage: (mobile phones, bank tokens, smart cards, : : :)

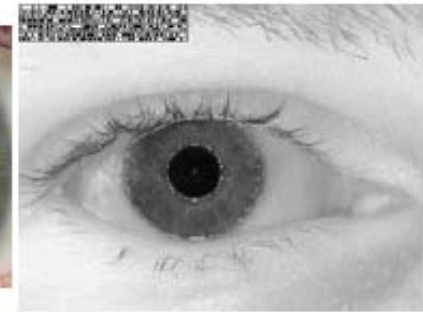
3 Noisy observation: authentication based on something closely linked to physiological characters



# Biometrics

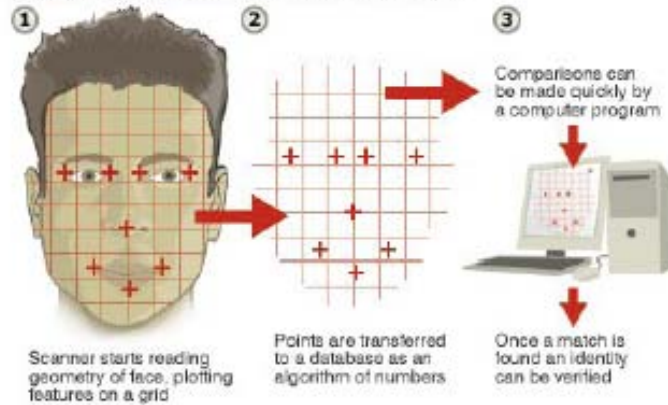


Fingerprint



Iris

## HOW 2D FACIAL SCANNERS RECORD IDENTITIES

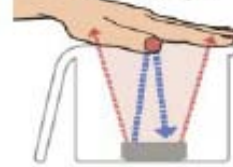


Facial

## Hands Down

How palm-scanning identification works:

The scanner emits infrared light. Hemoglobin in the veins absorbs the light...



...creating an image of the vein pattern that is reflected back and captured by the scanner.

Source: Reuters

The scan is stored in a database. A returning user's vein pattern is compared against the database to determine if there is a match.



Palm vein

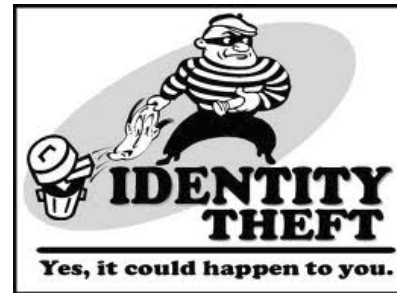
# Biometrics: why to be protected?

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A biometric template is a representation of a unique characteristic of an individual. It might also contain some information about diseases. Therefore, it contains privacy-sensitive information.

To abuse database:

- impersonation
- identity theft
- cross-database matching
- detectable pathologies
- ... yet undiscovered attacks



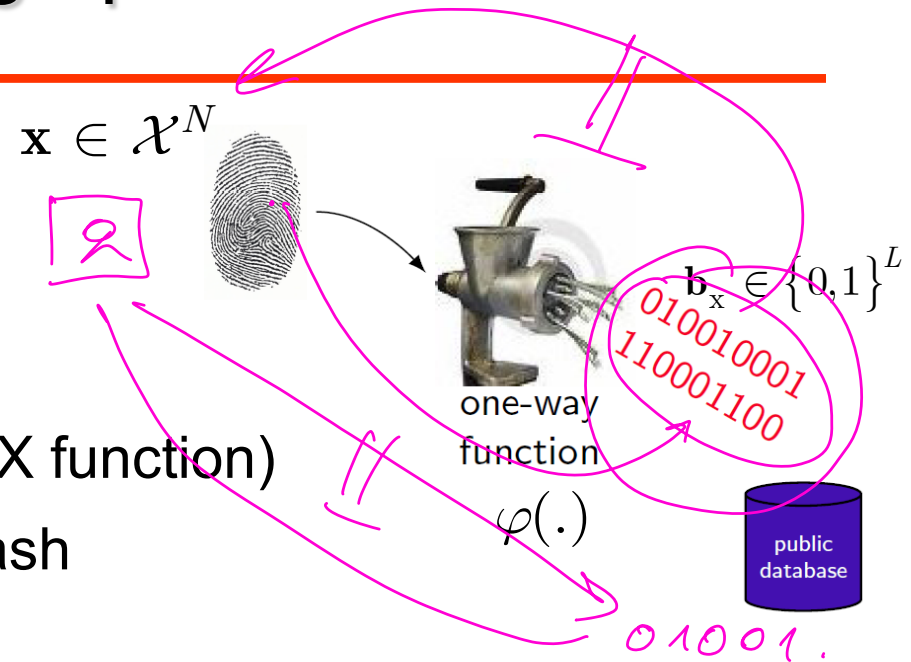
**Note:** biometric templates can not be “cancelable” because people have only ten fingers, two eyes, etc.

15:10.

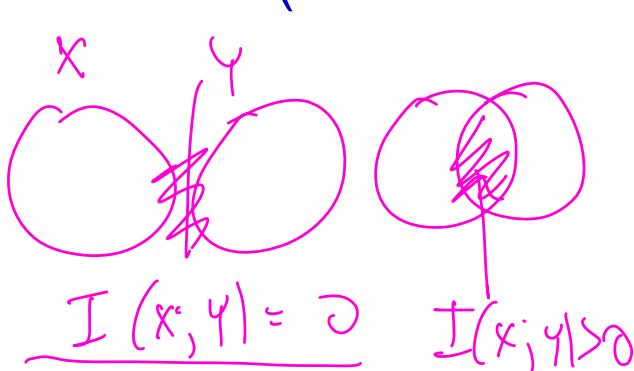
# Multimedia security: cryptographic solution

## ■ Main idea:

- do not store data itself
- store a one-way hash  
(non-invertible function like UNIX function)
- for the same data, the same hash



Remark: the hash  $\varphi(\cdot)$  is assumed to be uninformative  $\Rightarrow$  non-invertible.  
(which is not true in information-theoretic sense).



$$I(X; h(X)) = 0$$

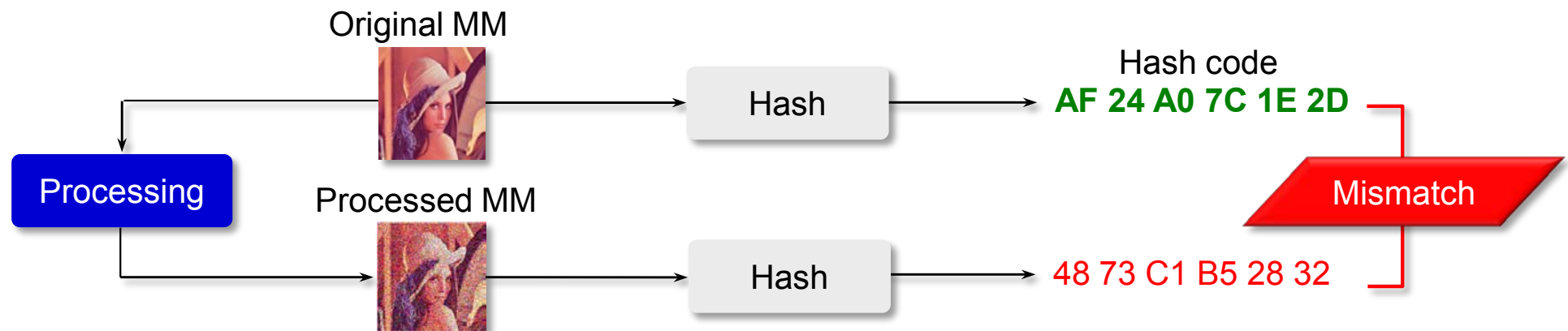
This is not true, since  $h(X)$  is a function of  $X$ !

$b_x = h(x)$

# Multimedia security: cryptographic solution

## Why are traditional security tools not suitable for multimedia?

- Traditional *security*:
  - Cryptographic encryption (confidentiality of information)
  - Cryptographic hashes (authentication, trust, access control)
- Main concerns of classical crypto-based algorithms:
  - Sensitivity to noise and unintentional distortions in input data
  - Data handling in the encrypted domain





# Multimedia privacy: data «degradation»

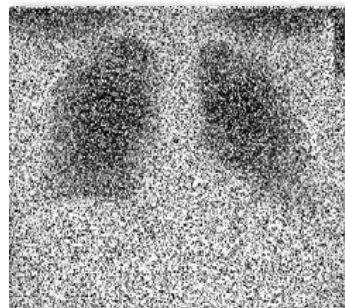
## Why are traditional privacy tools not suitable for multimedia?

- Traditional *privacy protection* [1]:
  - Data owner (protect data): based on data degradation and randomization (noise addition, lossy compression, data removal, dim. reduction...)
  - Data user (protect requests): based on anonymization, randomized rules
- Main concerns of classical privacy preserving algorithms:
  - Reduction of accuracy (data utility)
  - Not very efficient against experienced attackers (sensitivity analysis)

Original



50% removed based on key



Median 9x9



Adaptive recovery



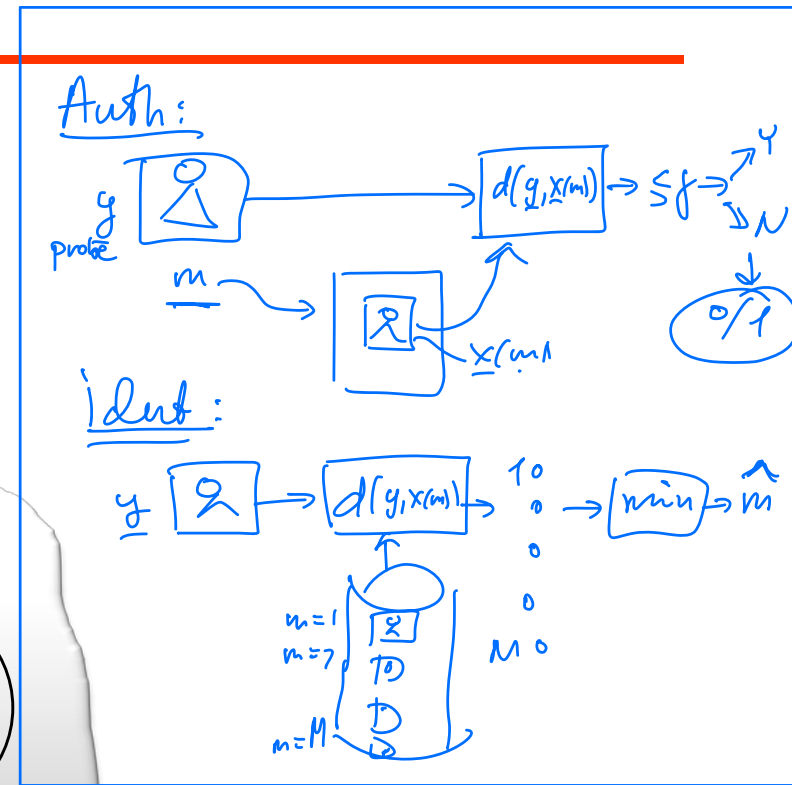
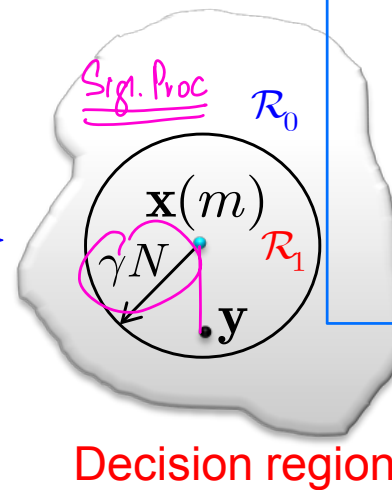
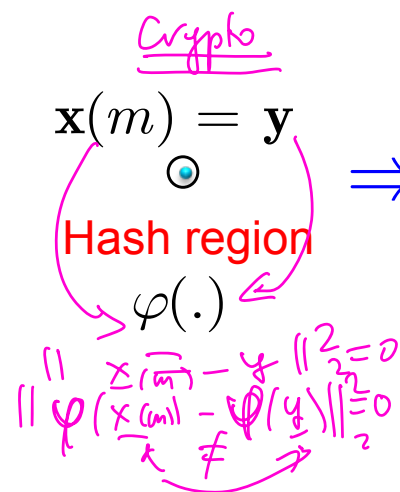
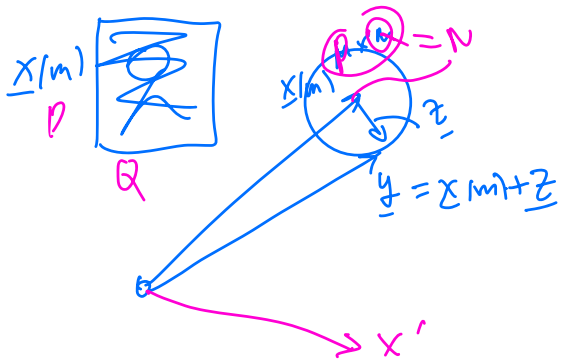
[1] C. Aggarwal and P. Yu, Privacy-Preserving Data Mining: Models and Algorithms, Springer, 2008.

# Signal processing approach: authentication

- Use hypothesis testing
- Handle noisy data  $\Rightarrow$  extended decision region

$$\begin{cases} H_0 : \mathbf{y} = \mathbf{x}' + \mathbf{z}, \\ H_1 : \mathbf{y} = \mathbf{x}(m) + \mathbf{z}, \end{cases}$$

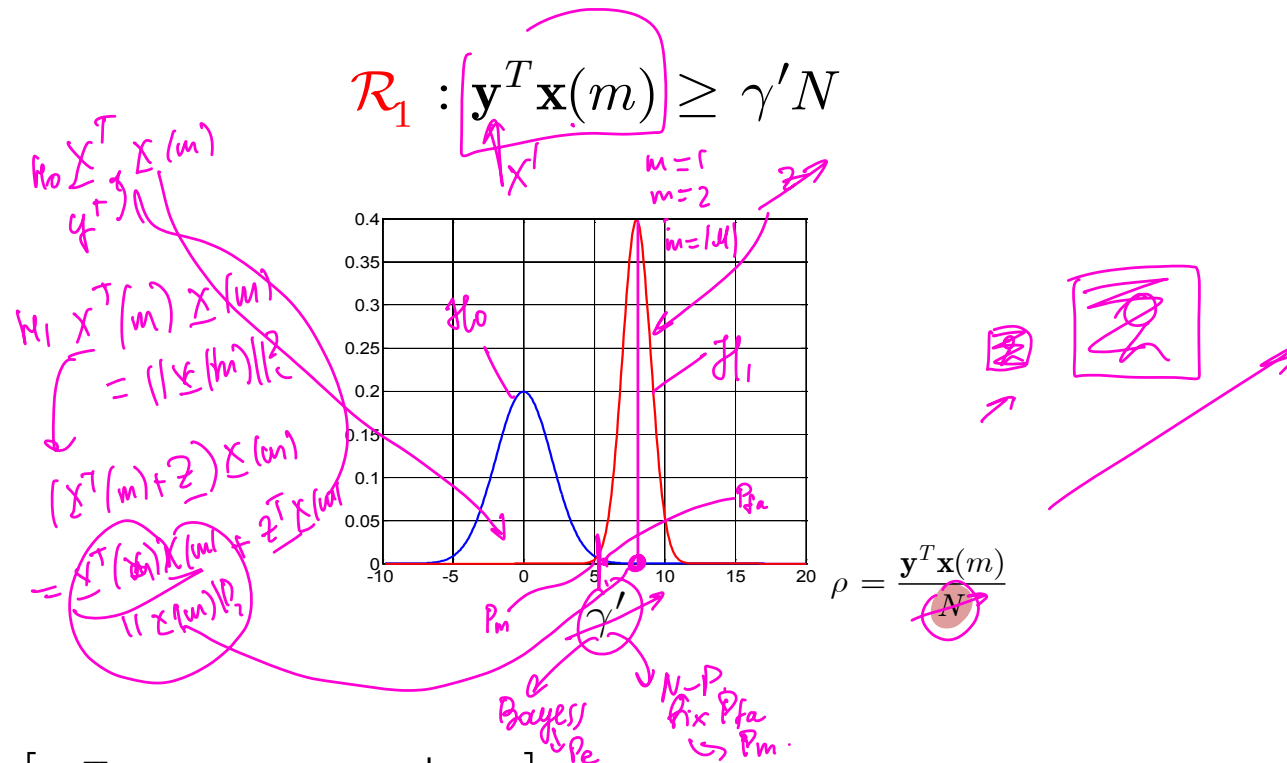
$\mathbf{x}' \neq \mathbf{x}(m), \text{ any } 1 \leq m \leq |\mathcal{M}|$



Gaussian assumption  $\Rightarrow$  Euclidian distance  $\equiv$  Sphere region

$$\mathcal{R}_1 : d^E(\mathbf{y}, \mathbf{x}(m)) = \|\mathbf{y} - \mathbf{x}(m)\|^2 \leq \gamma N \Rightarrow \mathcal{R}_1 : \mathbf{y}^T \mathbf{x}(m) \geq \gamma' N$$

# Signal processing approach: authentication



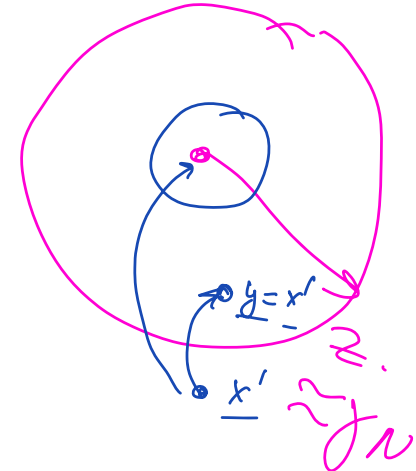
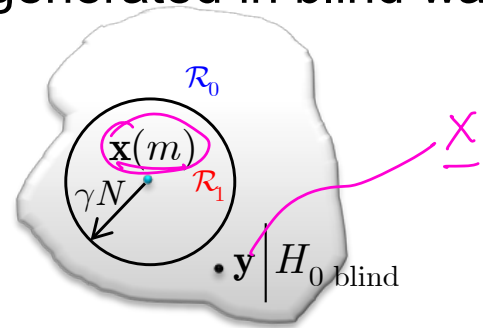
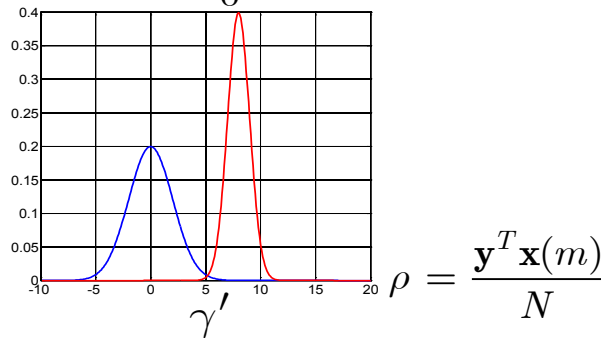
$$P_f = \Pr[\mathbf{y}^T \mathbf{x}(m) \geq \gamma' N | H_0] - \text{probability of false acceptance}$$

$$P_m = \Pr[\mathbf{y}^T \mathbf{x}(m) < \gamma' N | H_1] - \text{probability of miss}$$

# Signal processing approach: authentication

## Blind attacks

- $\mathbf{x}'$  under  $H_0$  was assumed to be generated in blind way

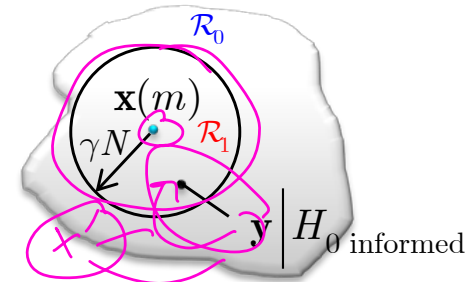


## Informed attacks

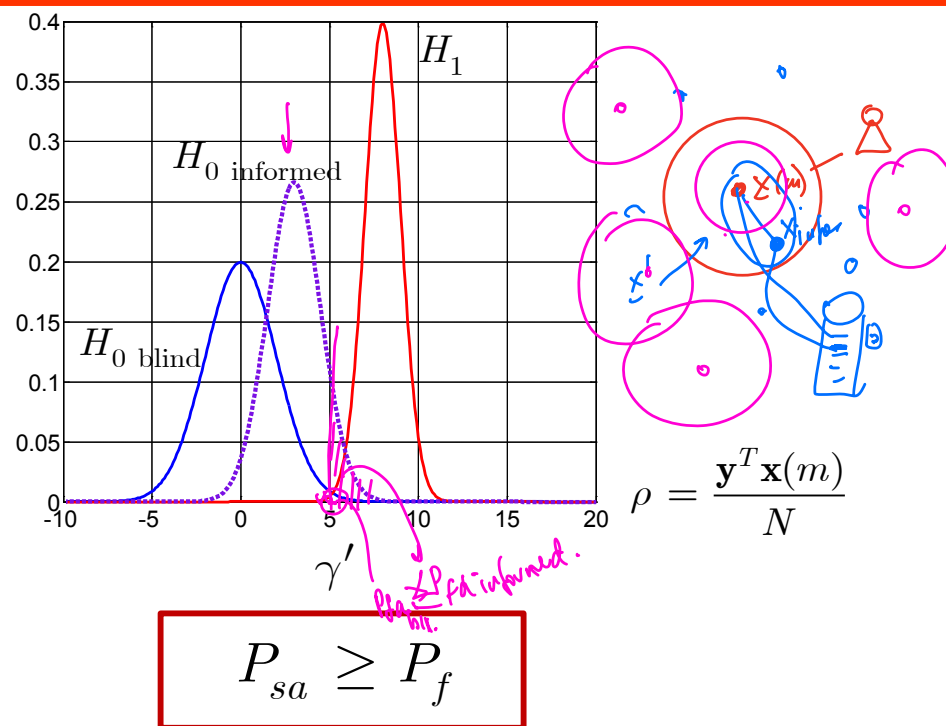
- Attacker can generate  $\mathbf{x}'$  as close as possible to  $\mathbf{x}(m)$ , if it is disclosed!

$$P_f = \Pr[\mathbf{y}^T \mathbf{x}(m) \geq \gamma' N | H_0 \text{ blind}]$$

$$P_{sa} = \Pr[\mathbf{y}^T \mathbf{x}(m) \geq \gamma' N | H_0 \text{ informed}] - \text{prob. of succ. attack}$$



# Signal processing approach: authentication

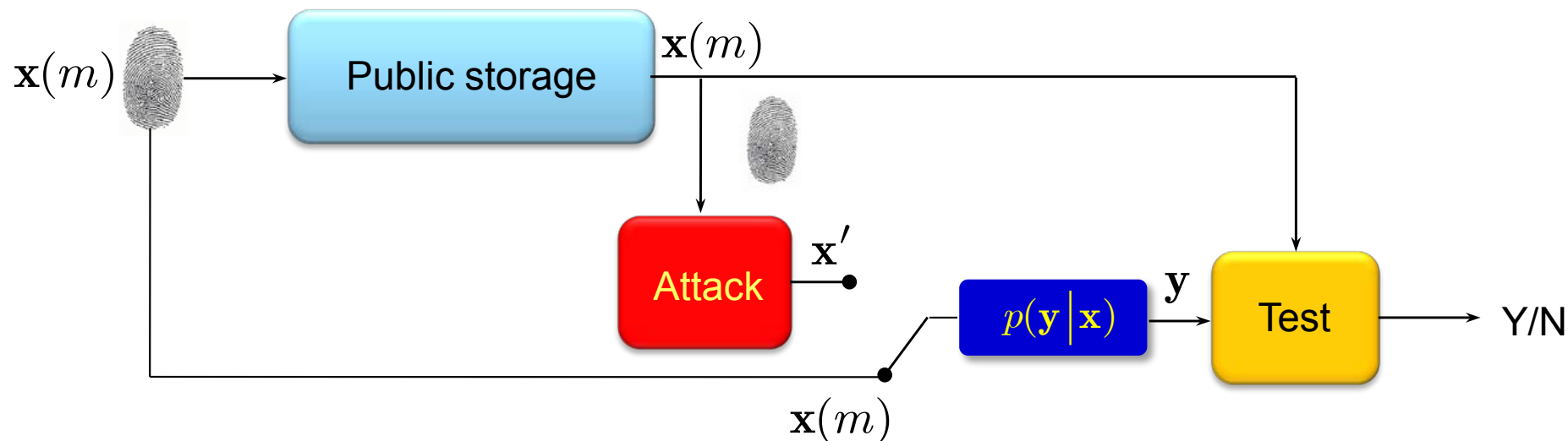


## Conclusion:

- the disclosure of  $\mathbf{x}(m)$  is dangerous for the performance (security).

# Signal processing approach: authentication

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## Open issues:

- $x(m)$  is stored in the public domain
  - no security
  - no privacy
- }  $x(m)$  can be used for various attacks

# Signal processing approach: identification

## ■ M-ary hypothesis testing

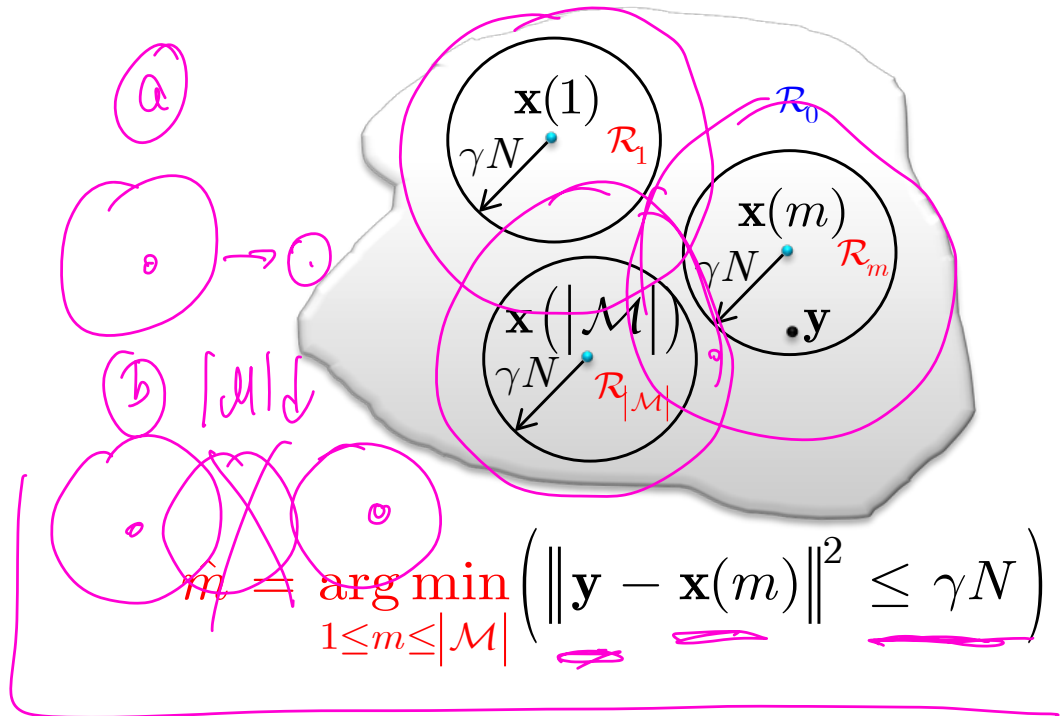
$$\begin{cases} H_0 : \mathbf{y} = \mathbf{x}' + \mathbf{z}, \\ H_1 : \mathbf{y} = \mathbf{x}(1) + \mathbf{z}, \\ \vdots \\ H_{|\mathcal{M}|} : \mathbf{y} = \mathbf{x}(|\mathcal{M}|) + \mathbf{z}, \end{cases}$$

$\mathbf{x}' \neq \mathbf{x}(m), \text{ any } 1 \leq m \leq |\mathcal{M}|$

$$P_f = \Pr \left[ \bigcup_{m=1}^{|\mathcal{M}|} \mathbf{y}^T \mathbf{x}(m) \geq \gamma' N \mid H_{0 \text{ blind}} \right]$$

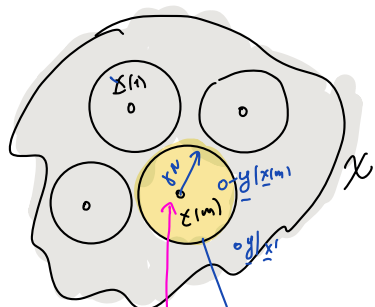
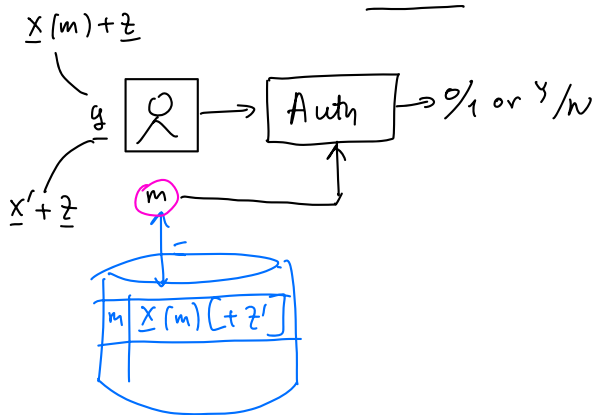
$$P_{sa} = \Pr \left[ \bigcup_{m=1}^{|\mathcal{M}|} \mathbf{y}^T \mathbf{x}(m) \geq \gamma' N \mid H_{0 \text{ informed}} \right]$$

$$P_{ic} = \Pr \left[ \underbrace{\mathbf{y}^T \mathbf{x}(m) < \gamma' N}_{\text{miss}} \bigcup \underbrace{\bigcup_{m' \neq m}^{|\mathcal{M}|} \mathbf{y}^T \mathbf{x}(m') \geq \gamma' N}_{\text{missacceptance}} \mid H_m \right]$$



$\hat{m}$   
 $\rightarrow \emptyset$

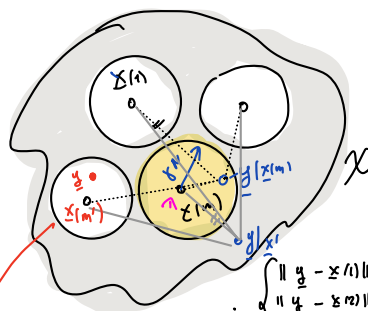
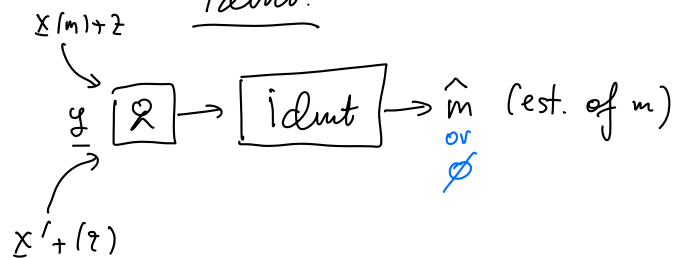
Auth



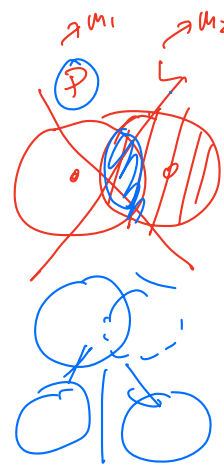
Test of auth:  
 $\|y - x(m)\|_2^2 \leq \gamma N$

$$\begin{aligned} x(m) &\in \mathbb{R}^N \\ y &\in \mathbb{R}^N \end{aligned}$$

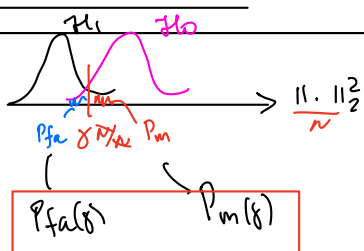
Ident.



$$\min \begin{cases} \|y - x(1)\|_2^2 \leq \gamma N \\ \|y - x(2)\|_2^2 \leq \gamma N \\ \vdots \\ \|y - x(M)\|_2^2 \leq \gamma N \end{cases} \Rightarrow \hat{m}$$



Conclusions:



$$(M-1) \cdot P_{fa}(\gamma) \quad P_m(\gamma)$$

Performance

Complexity

$\mathcal{O}(1)$

$$\|y - x(m)\|_2^2$$

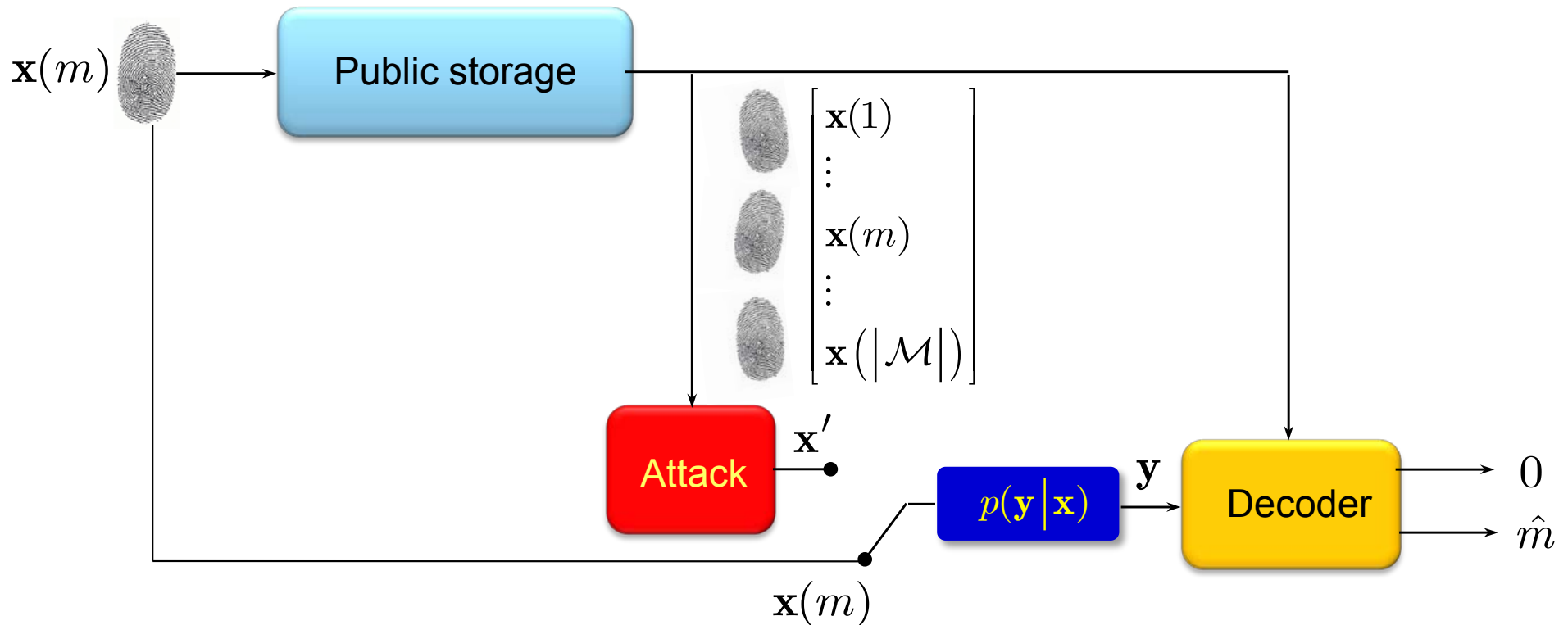
one-to-one  
 $y$  vs  $x(m)$

$\mathcal{O}(M) \sim A-N$   
 'approx

one-to-many  
 $y$  vs  $\{x(m)\}_{m=1}^M$



# Signal processing approach: identification



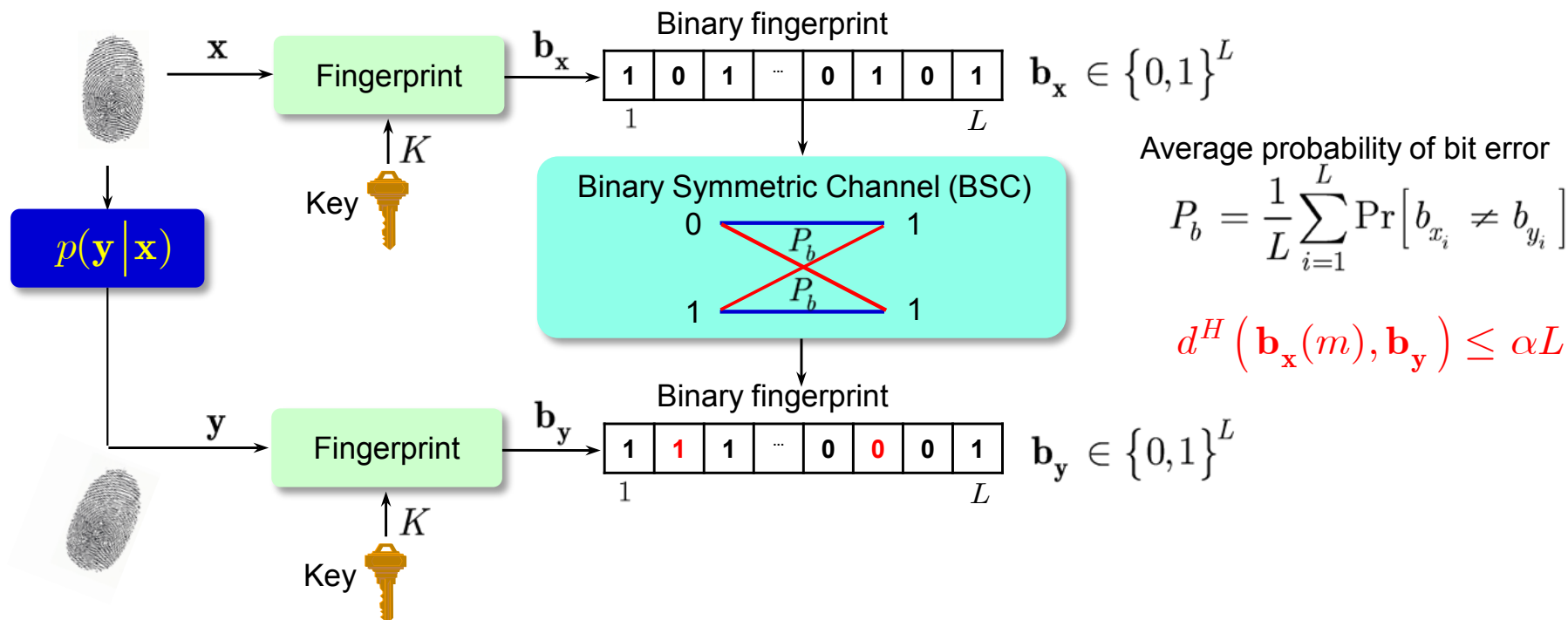
## Open issues:

- all  $\{\mathbf{x}(m)\}$  are stored in the public domain
  - no security
  - no privacy
- }  $\mathbf{x}(m)$  can be used for various attacks

# Signal processing approach: towards «robust hashing»

$x(m)$   $H$   $N$   $f_g$   $b_x(m)$   $ML \leq AE$   $Classif$  (see Lecture 1).  $SSL$   $HC: \mathcal{O}(W_{x(m)})$  (see lecture on robust perc. hashing).  
 What for: (a) complexity ( $\downarrow$ ). (O/HW) vs  $\mathcal{O}(L)$ .  
 (b) privacy.

- If  $x(m)$  is open, the attacker can: (a) **deduce privacy** and (b) **design attacks**
- **Robust hashing**: an attempt to mimic crypto hashing but stay robust to signal processing (blur, noise, lossy compression, etc.) and geometric (affine, projective) distortions.



# Signal processing approach: towards «robust hashing»

- Main assumptions:

- Robust hashing is robust to some minor modifications
- Robust hashing is non-invertible under

- (a) unknown key

- (b) known key

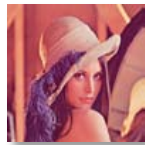
such that  $d^E(\mathbf{x}, \hat{\mathbf{x}}) \geq \gamma N$

This is possible due to loss of information

$$\mathbf{X} \rightarrow \mathbf{B}_x$$

$$\mathbf{Y} \rightarrow \mathbf{B}_y$$

Original/Query



$\mathbf{x}$

Fingerprint

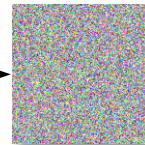
$K$

$\mathbf{b}_x$

Estimate

$K$

Estimated MM



$\hat{\mathbf{x}}$

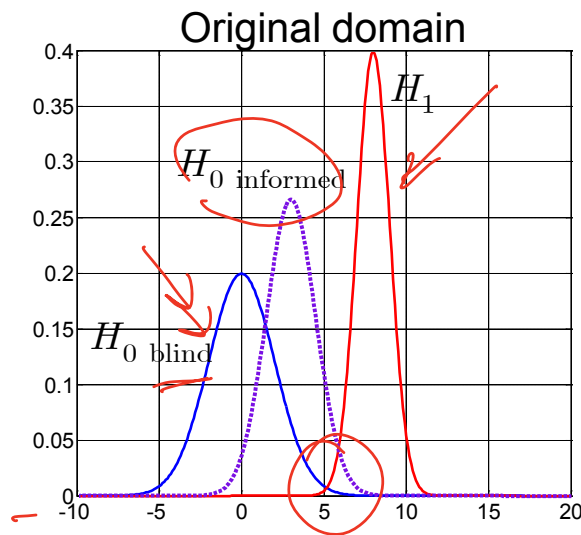
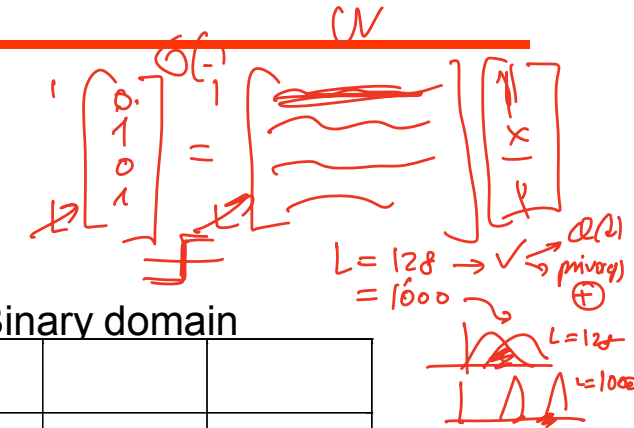


**Remark:** authentication/identification based on  $(\mathbf{B}_x, \mathbf{B}_y)$  is less accurate than based on  $(\mathbf{X}, \mathbf{Y})$

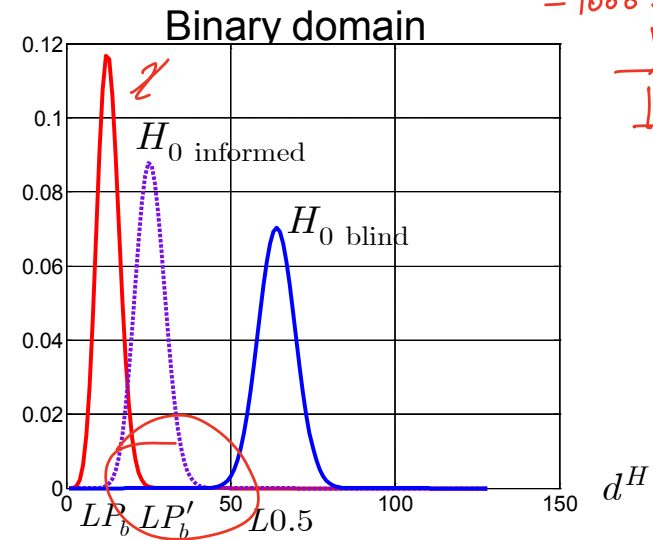
# Signal processing approach: towards «robust hashing»

- Impact of reduction of dimensionality and binarization:

- FP:  $\mathcal{R}^N \times \mathcal{K} \rightarrow \{0,1\}^L$



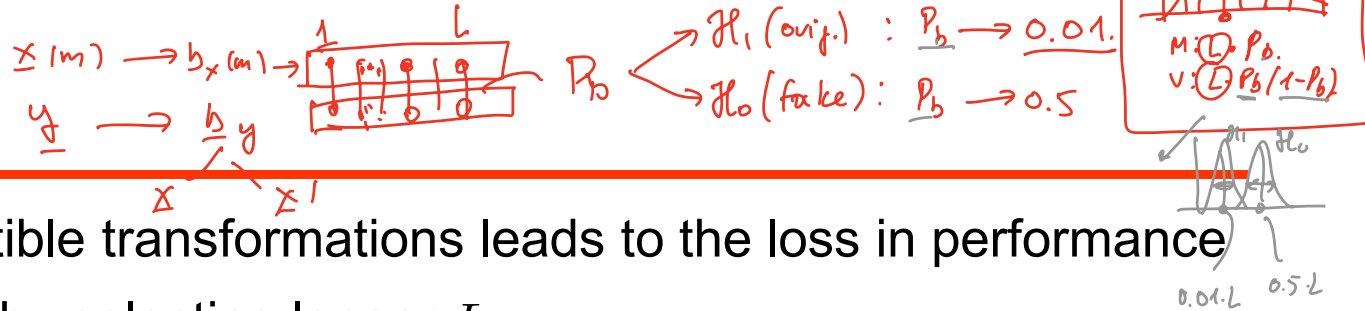
Fingerprinting  
 $\Rightarrow$   
 Loss of information



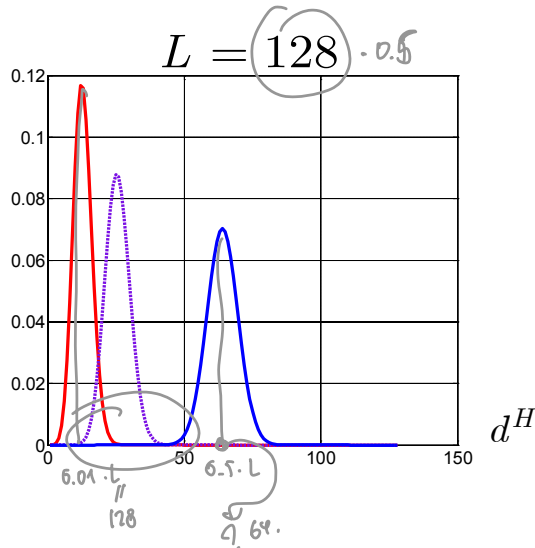
Handwritten notes and equations:

- $\rho = \frac{\mathbf{y}^T \mathbf{x}(m)}{N}$
- $\min \left[ \frac{1}{N} \|\mathbf{y} - \mathbf{x}(m)\|_2^2 \right] < \gamma$
- $\|\mathbf{y}\|_2^2 = 2\mathbf{y}^T \mathbf{x}(m) + \|\mathbf{x}(m)\|_2^2$
- $d^H(\mathbf{b}_y, \mathbf{b}_{x(m)})$

# Hybrid solution

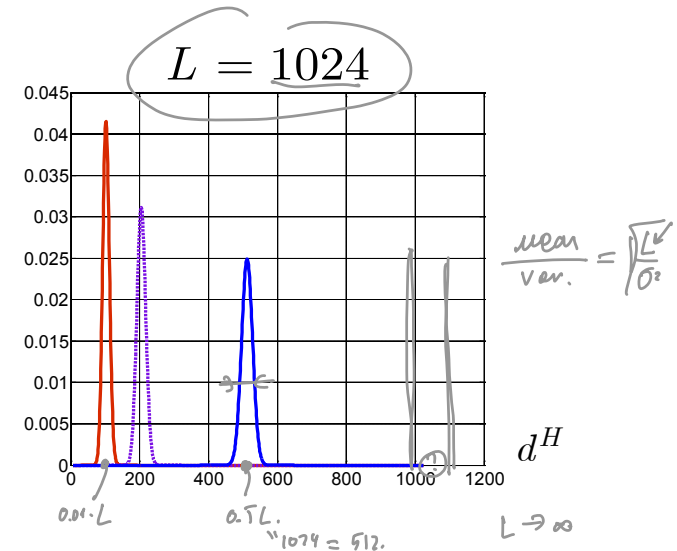


- Application of non-invertible transformations leads to the loss in performance
- One can compensate it by selecting longer  $L$



Increase  $L$

$\Rightarrow$



- In turn, longer  $L$  leads to better chances for image reconstruction!
- Problem:** can one trade-off security/privacy – performance?

$\Rightarrow$  **Solution:** crypto + signal processing + coding

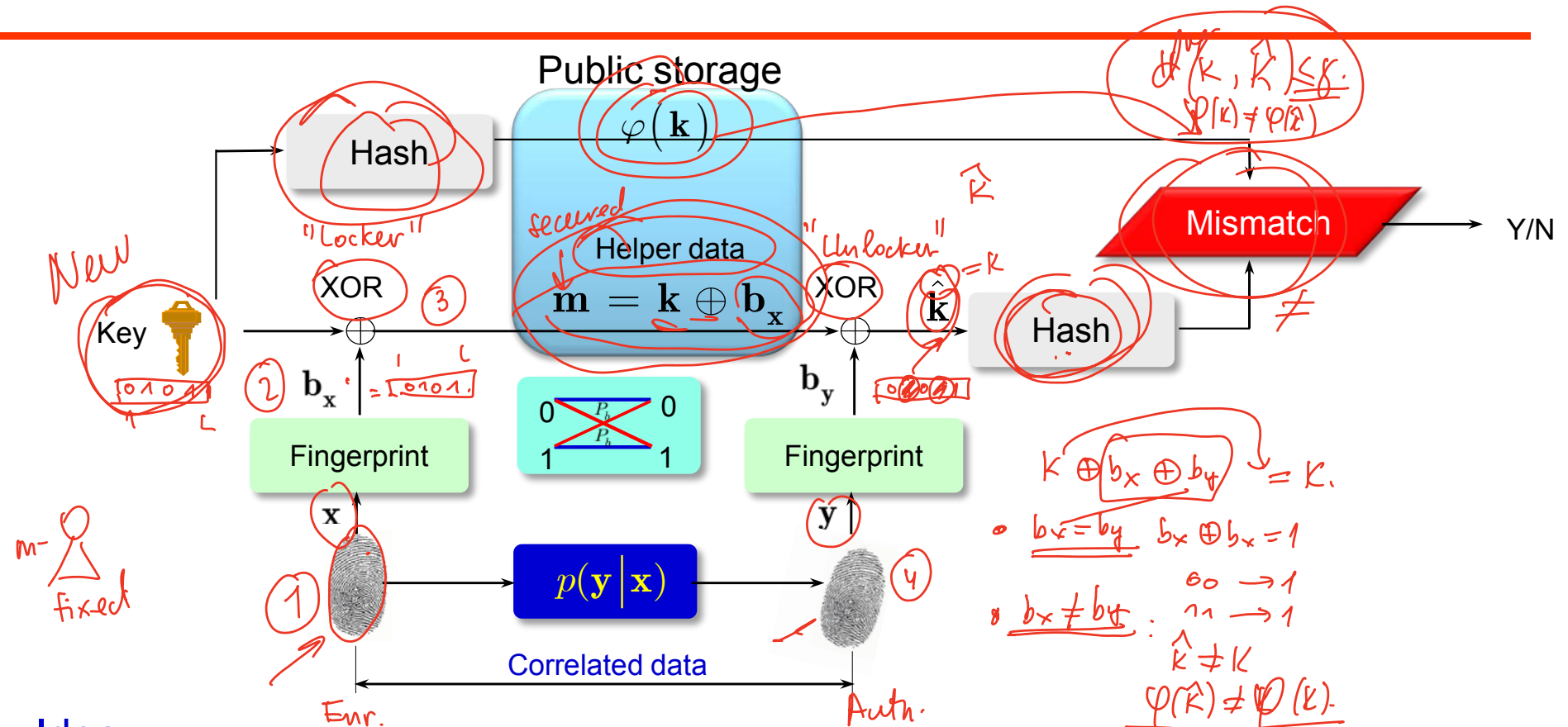
B. Škorić

# Hybrid solution: strategies

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- Several methods (based on common idea):
  - fuzzy commitment
  - helper data
- Main strategy:
  - use additional helper data to correct errors
  - use a fact that  $\mathbf{X}$  and  $\mathbf{Y}$  are correlated as well as  $\mathbf{B}_x$  and  $\mathbf{B}_y$  for the hypothesis  $H_1$  in authentication and the hypothesis  $H_m$  in identification

# Hybrid solution: fuzzy commitment

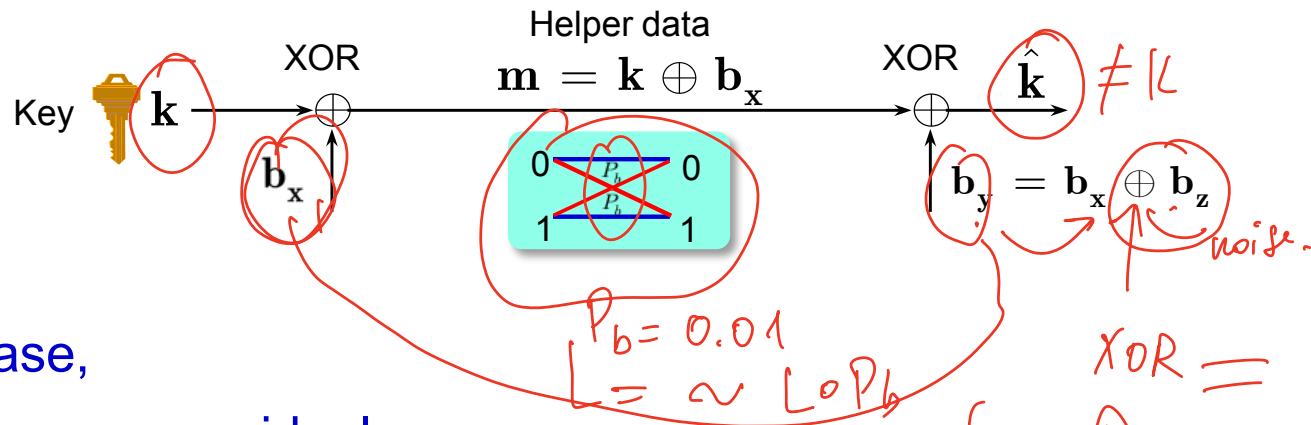


## ■ Idea:

- $k$  is used for every new session (i.i.d.) while  $b_x$  remains the same (non-renewable)
- $b_x$  is used to lock the key (even, if it is not i.i.d.)  $\Rightarrow$  analog of one-time-pad

# Hybrid solution: fuzzy commitment

- Open issues:



- In general case,

- $b_x \neq b_y \Rightarrow$  residual error

$$\hat{k} = k \oplus \underbrace{b_x \oplus b_y}_{b_z} = k \oplus b_z \Rightarrow \varphi(k) \neq \varphi(\hat{k})$$

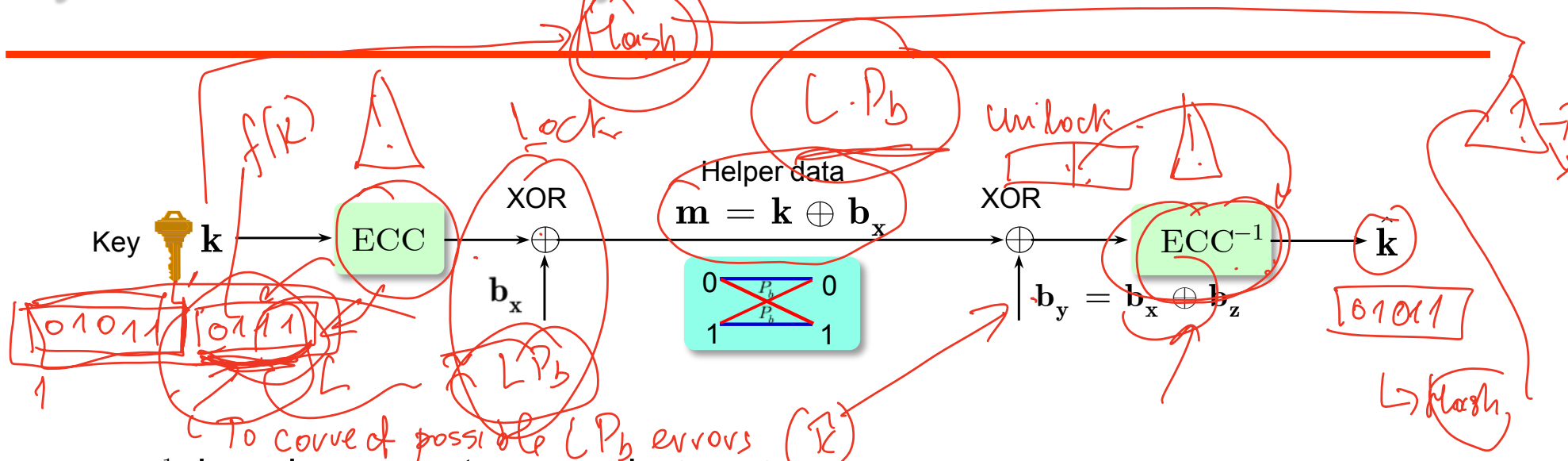
- Idea:

- add redundancy to  $k$  using error correction code to correct errors
- redundancy is proportional to  $\approx LH_2(P_b)$

$$\sqrt{U_{oc}} \approx \sqrt{LP_b + 3 \sqrt{LP_b(1-P_b)}}$$

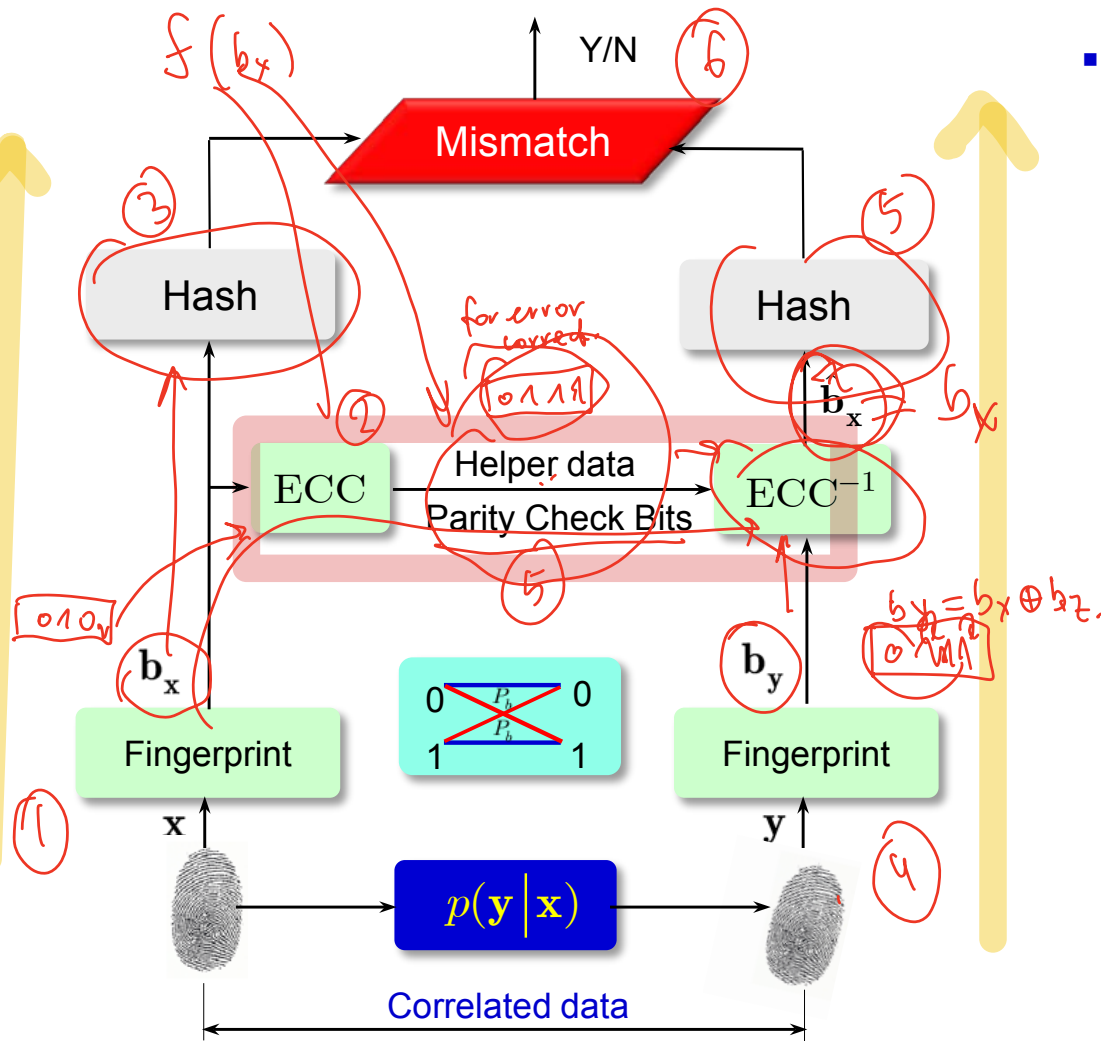


# Hybrid solution: fuzzy commitment



- $ECC^{-1}$  decoder corrects errors in  $k \oplus b_z$
- Open issues:
  - $ECC(k)$  is not anymore i.i.d.
    - ⇒ smaller entropy: easier to predict and attack
  - In addition  $b_x$  is not i.i.d., the attacks are possible against both!

# Hybrid solution: helper data based systems



## Open issues:

- redundancy is proportional to  $\approx LH_2(P_b)$
- However, these methods represent the state-of-the-art!
- It is up to you to find something better.