



IBM Research

Performance evaluation of cancelable biometrics

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***Joint work with members of biometrics research team**

Cancelable Biometrics

- Intentional **repeatable** distortion
 - Generates a similar signal each time for the same user
- Compromised scenario:
 - a new **distortion** creates a new biometrics
- Comparison scenario:
 - **different** distortions for different accounts
- **Backwards compatibility**
 - Representation is not changed.



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Cancelability requirements of the transform

1. The intrinsic strength (individuality) of the biometric should not be reduced after transformation. (Constraint on FAR)

$$D(x_1, x_2) > t \Rightarrow D(T(x_1), T(x_2)) > t$$

2. The transformation should be tolerant to intra-user variation (Constraint on FRR)

$$D(x_1, x_2) < t \Rightarrow D(T(x_1), T(x_2)) < t$$

3. The original should not match with the transform,

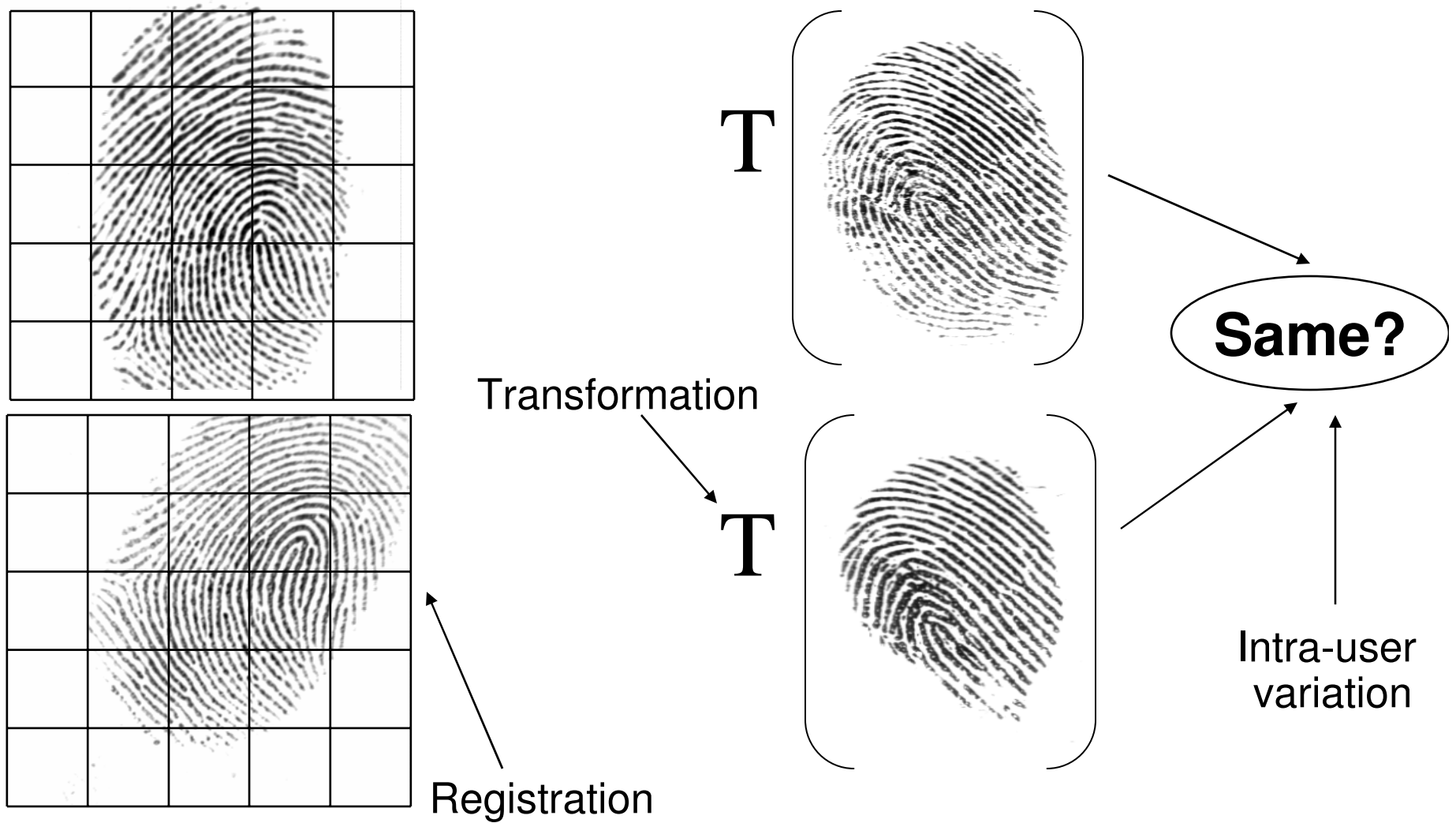
$$D(x, T(x)) > t$$

4. Different transforms of the same user should not match with each other

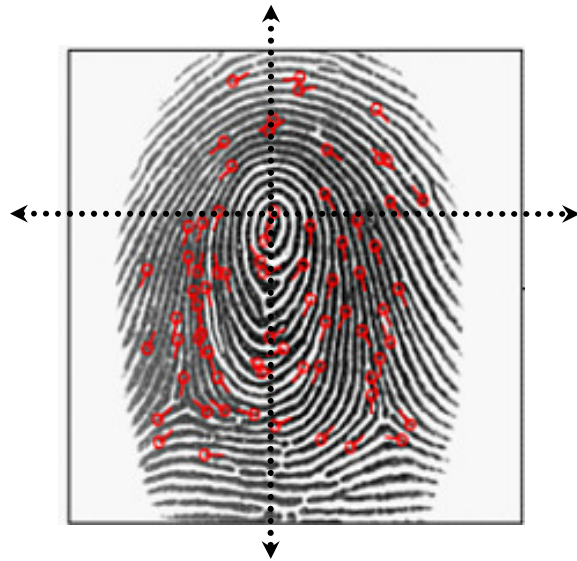
$$D(T_1(x), T_2(x)) > t$$

Registration based

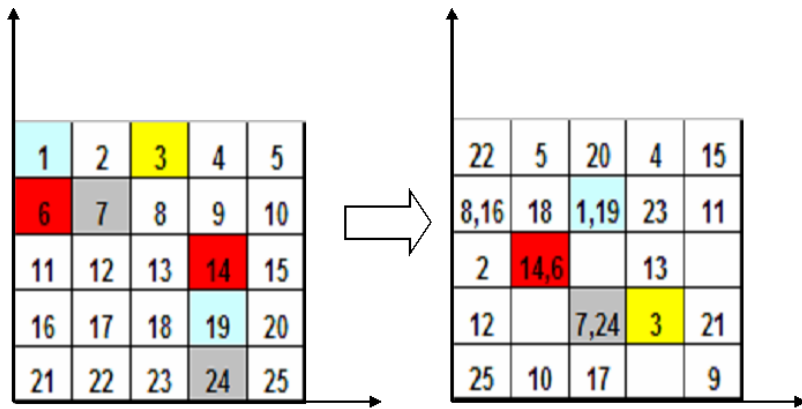
Challenges



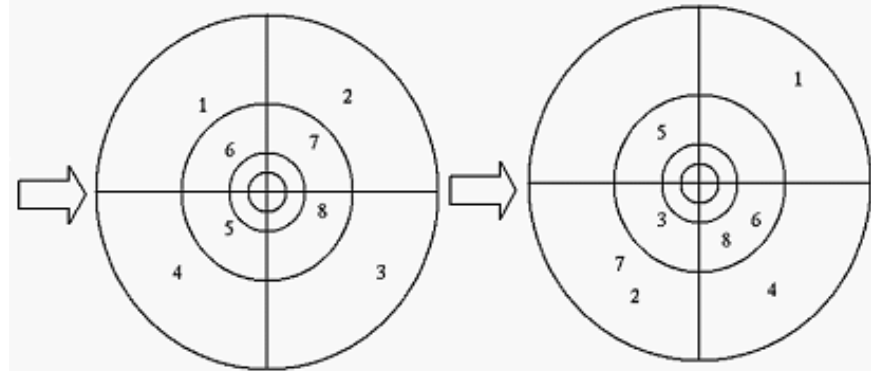
Feature Domain Transformation



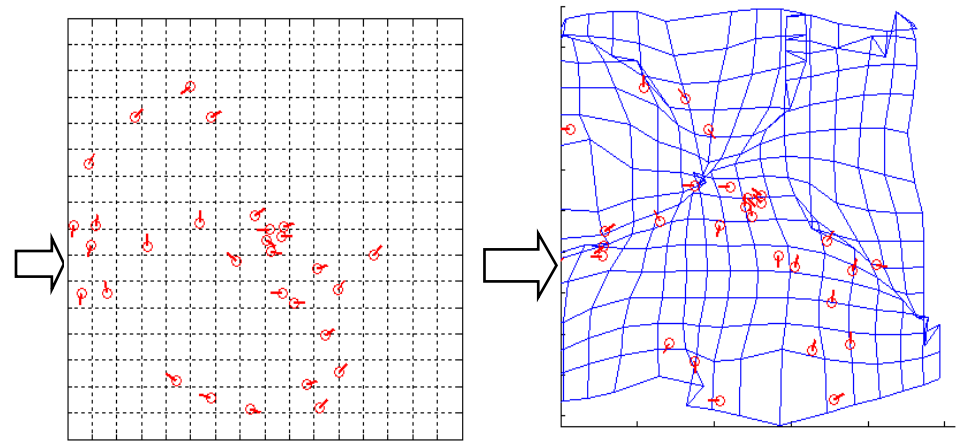
Feature Extraction



Cartesian Transformation



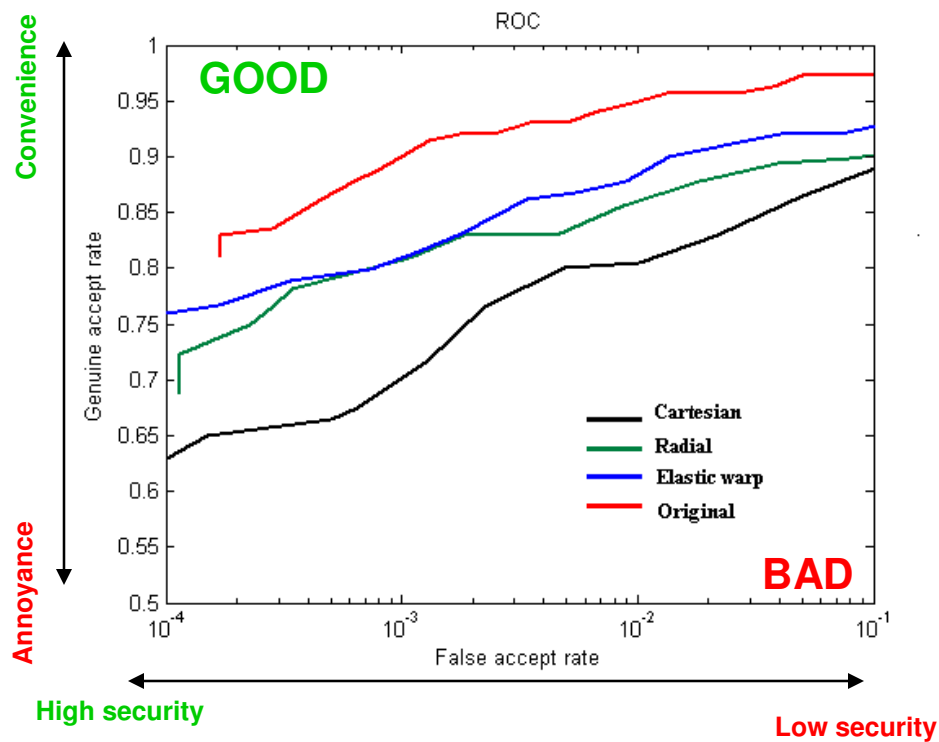
Polar Transformation



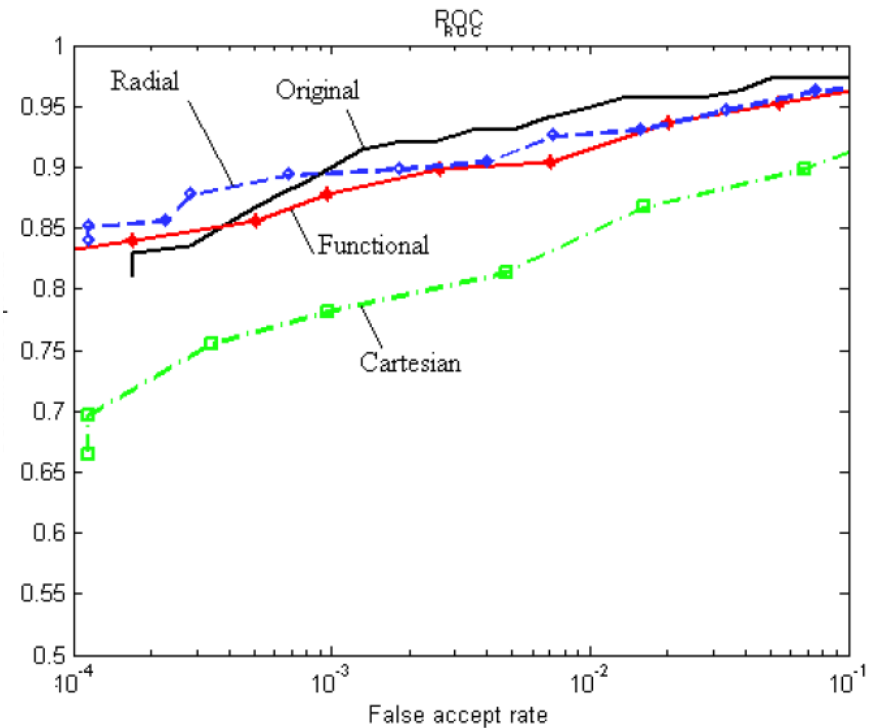
Surface Folding Transformation

How does it affect accuracy?

Same transform for all users



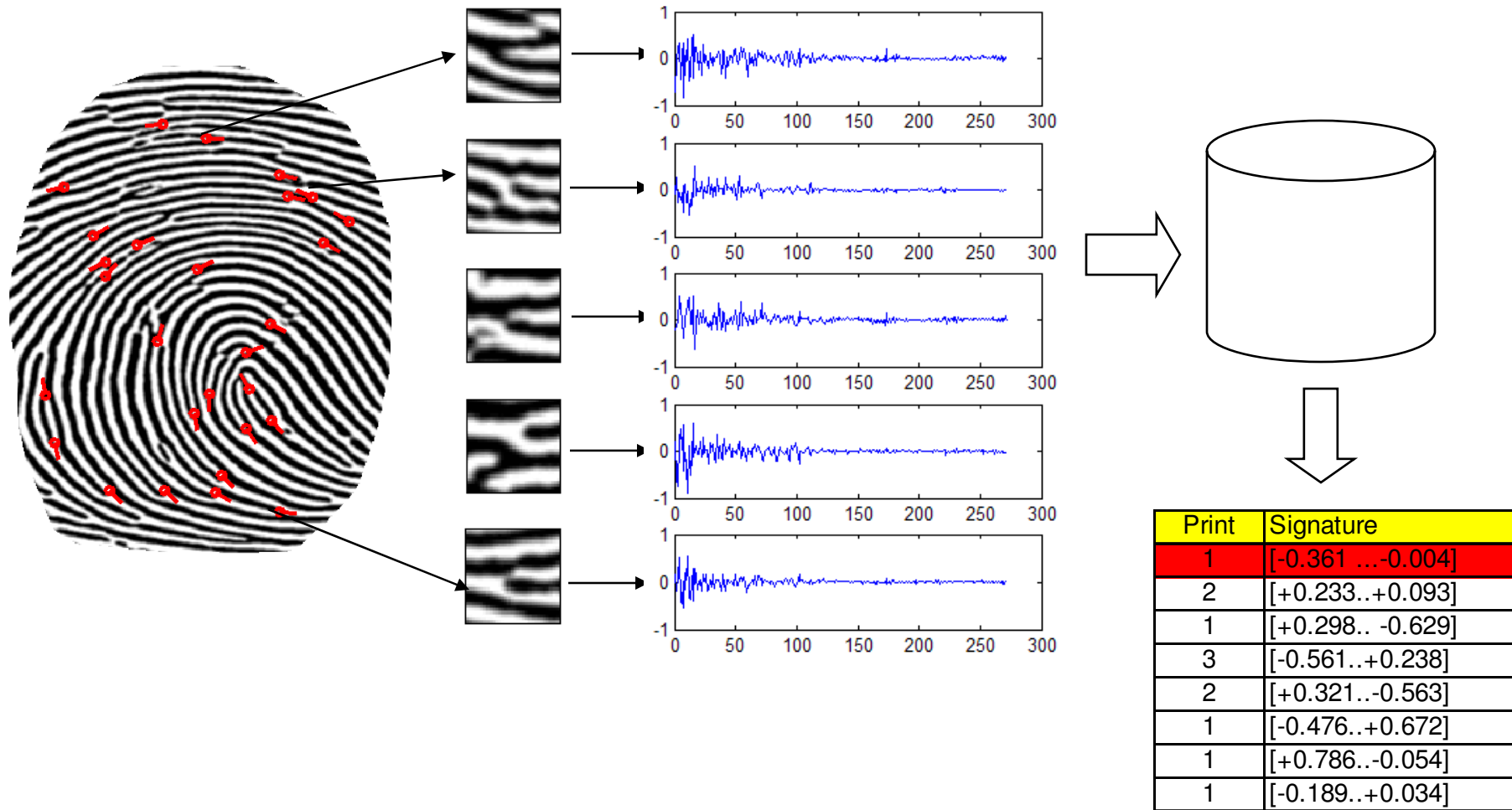
Different transforms for different users



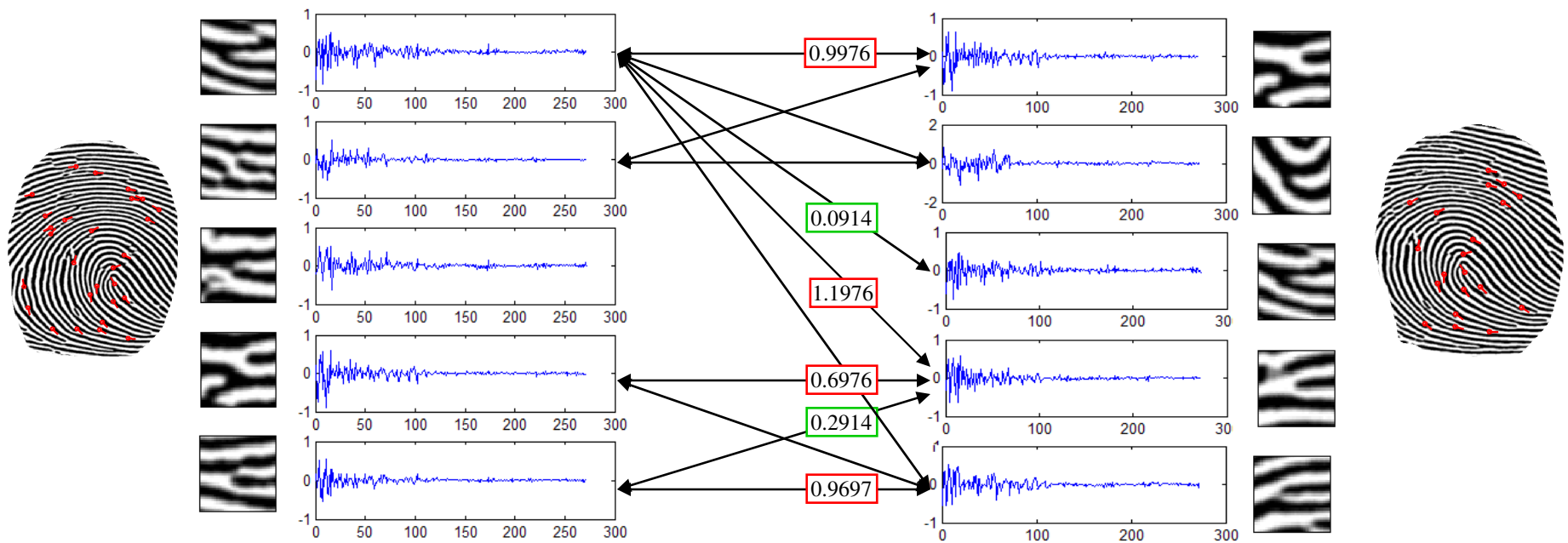
- Results reported in
 - “Cancelable biometrics: A case study in Fingerprints”, ICPR 06
 - “Generating cancelable fingerprint templates”, IEEE PAMI

Registration free

Enrollment

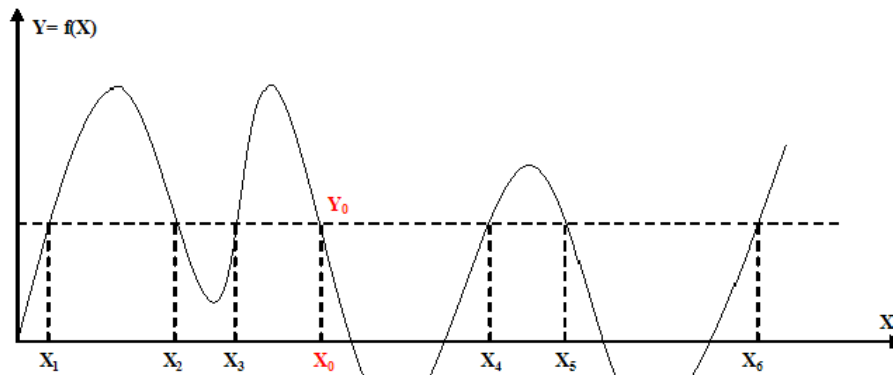


Verification



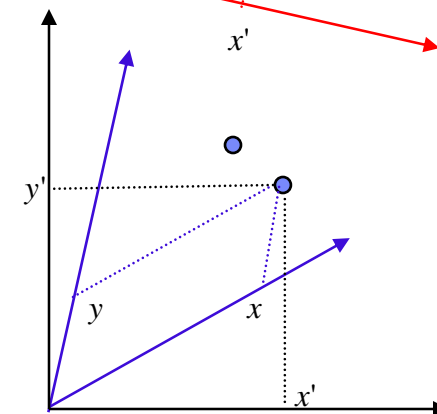
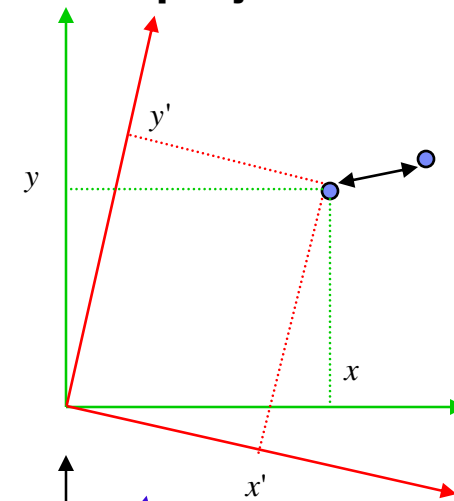
Cancelable methods

- Can we **avoid storing the original patch signatures?**
- Ways to transform/hide the feature vector
 - Encryption - representation too unstable for encryption
 - Polynomial transformation
 - Random projection- **fits well with NDP distance**



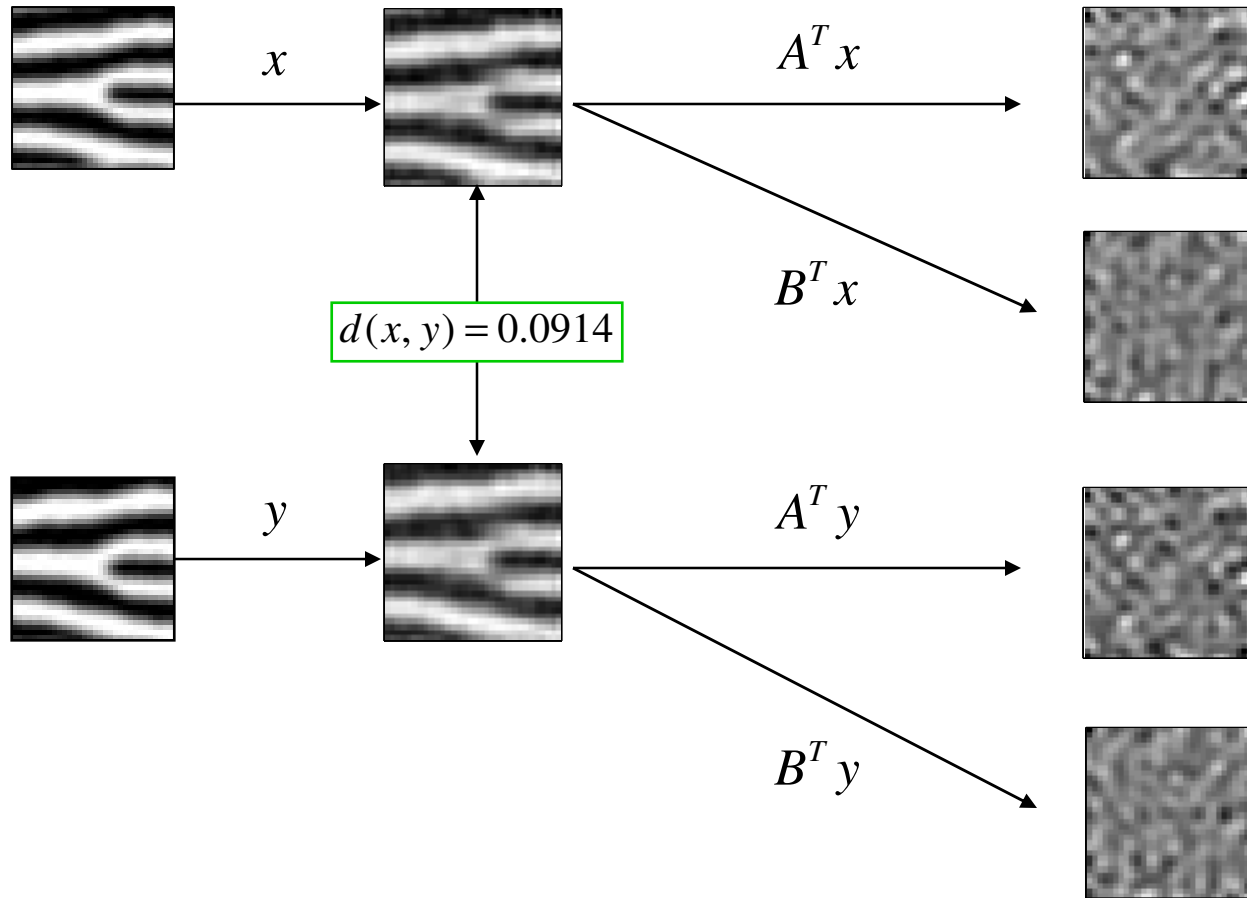
Polynomial transformation

Preferred: Ortho normal projections



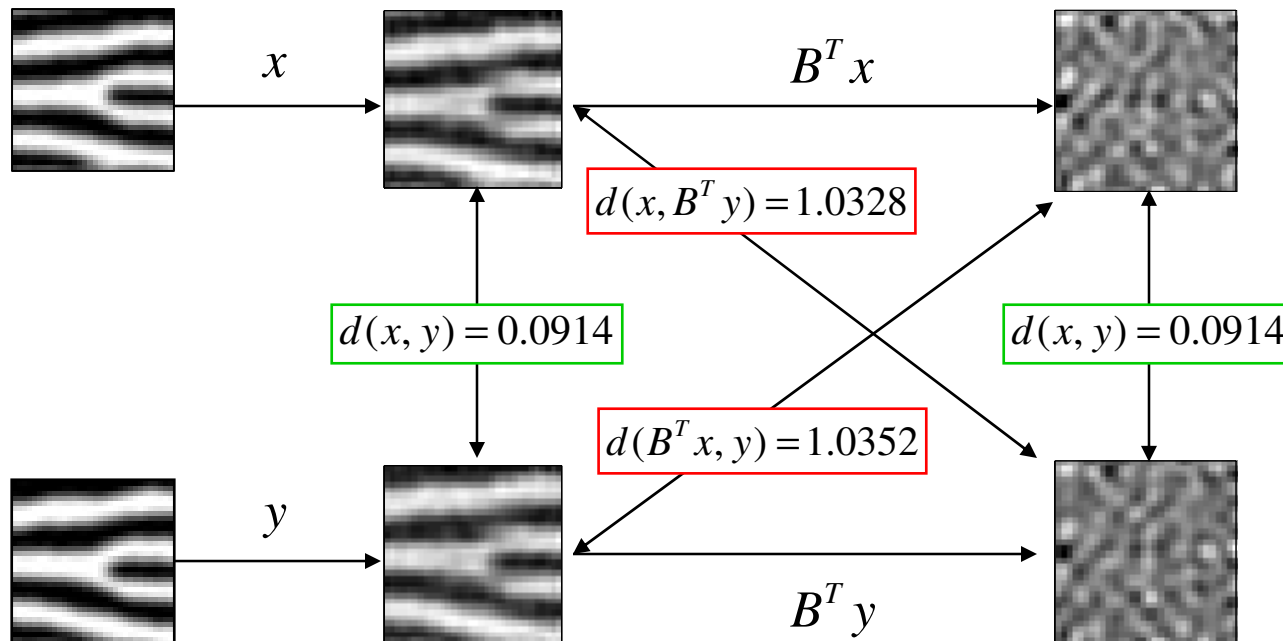
Random Projections

Cancelability (2)



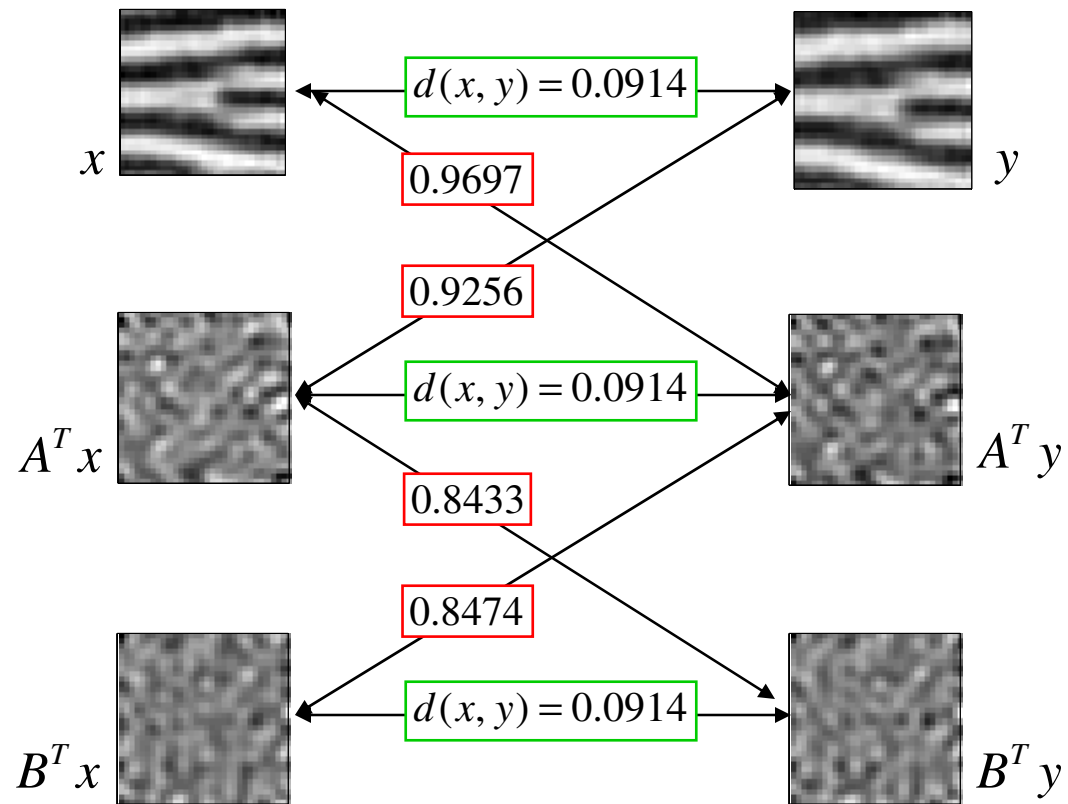
- Each patch can be used to produce **multiple transforms**

→ Cancelability (3)



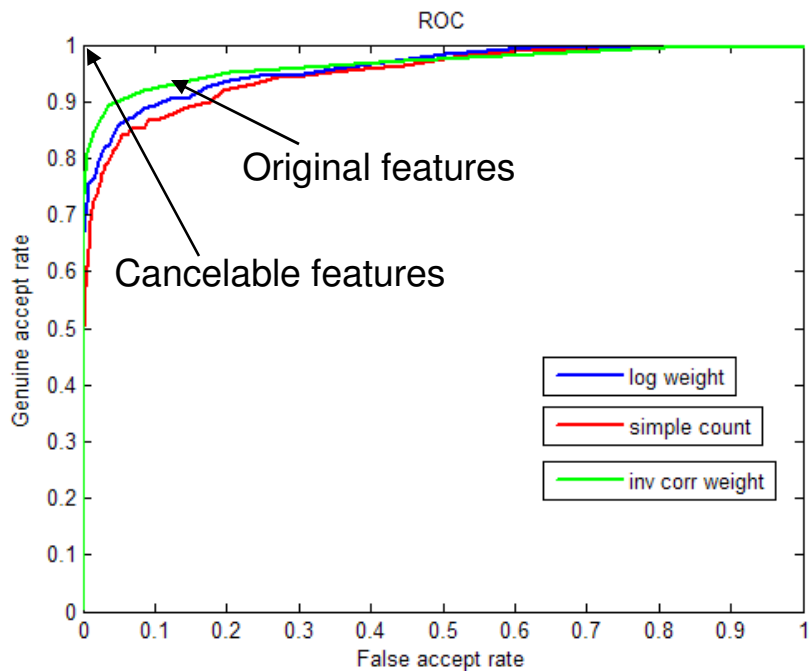
- Original **match** among themselves
- Transforms **match** among themselves
- Transform **does not match** with original

Cancelability (4)



- Score more than **0.5** is a mismatch
- Different Transforms **don't match** with each other

Empirical Results (1)



Patch based verification

- Performance is less than geometry based matchers (62% GAR at 0.01% FAR)

Cancelability

- Complete separation** (100% GAR, 0% FAR) achieved by having **separate** transforms for **separate** individuals

Diversity of key space

- Complete separation** (100% GAR, 0% FAR) achieved for **separate** (188) transforms of the **same** individual.

Non invertibility

- Complete separation** (100% GAR, 0% FAR) achieved for **non-invertible** construction as well

- **Perfect performance because uses entropy from key also**
- **If everyone uses the same key performance will not change because distances are preserved**

Increasing security: Two factor transformation

- The current construction **is invertible**

If we have the projecting matrix B , and the transform $T(x) = B^T x$

$x = BT(x) = BB^T x$, can be recovered

- **Can we increase security?**

- Two factor transformation

- The projection matrix B is constructed using **two** orthonormal matrices U, V

$$B = UV^T$$

$$UU^T = U^T U = VV^T = V^T V = I$$

$$BB^T = (UV^T)VU^T = U(V^T V)U^T = I$$

U, V are obtained by performing SVD on a random matrix $R = USV^T$

S is not recorded anywhere in the system.

U, V do not leak information about each other

- U and V can be separately stored **separately** (e.g. split between user and application?)
- **Symmetric key, public key comparison**

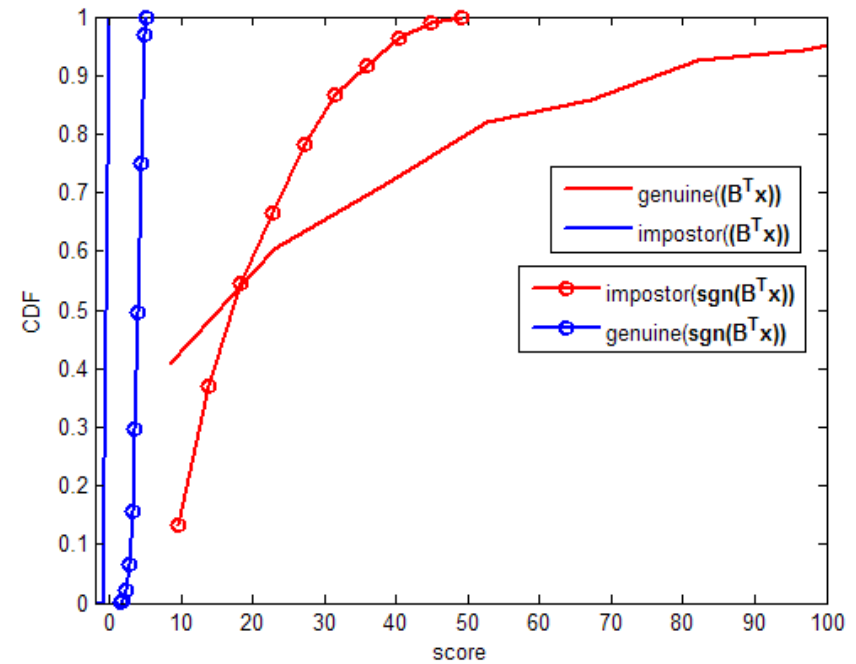
More security: Non-invertibility

- We can make the construction non-invertible by introducing some **non-linearity**

Define,

$$T(x) = \begin{cases} 1 & \text{if } B^T x > 0, (B = UV^T) \\ 0 & \text{otherwise} \end{cases}$$

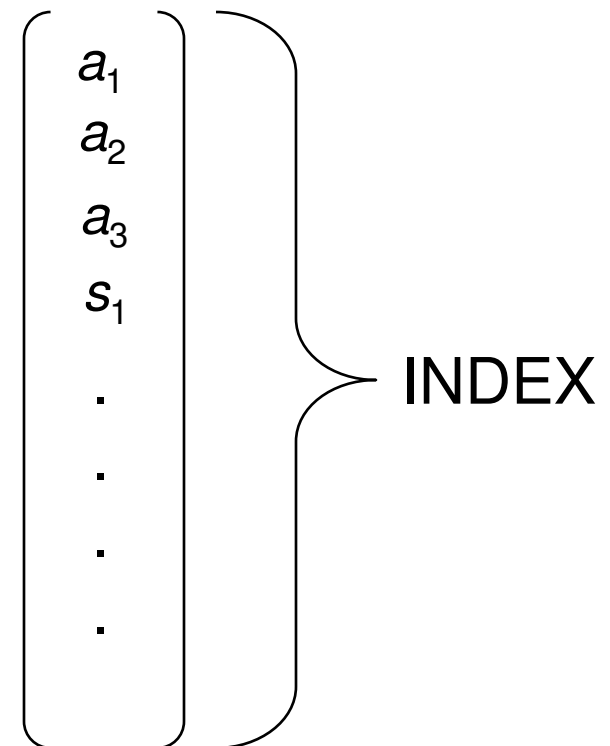
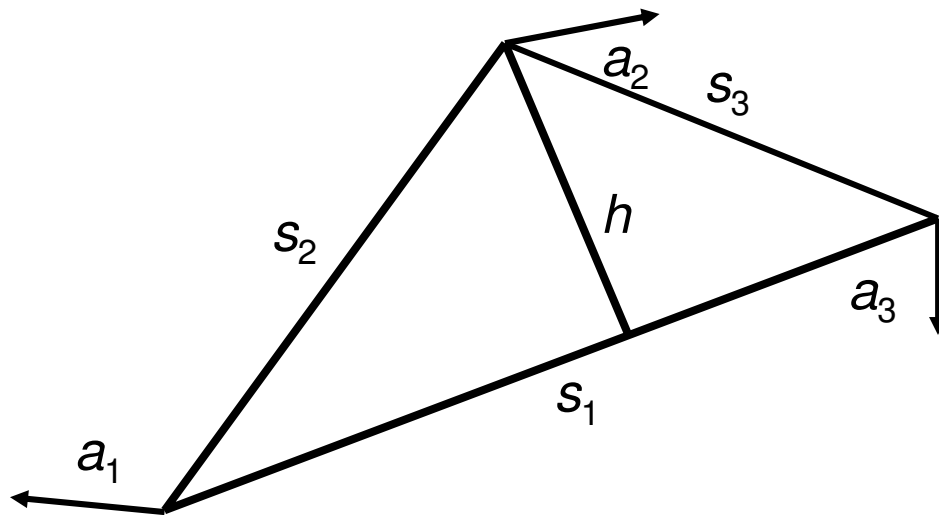
- Thus, even if U , V , $T(x)$ are known, it is impossible to recover x from $T(x)$
- Advantages:
 - The construction is **non-invertible**
- Disadvantages
 - Brute force attack is easier. (More pre-images of $B^T x$ produce the same sign)



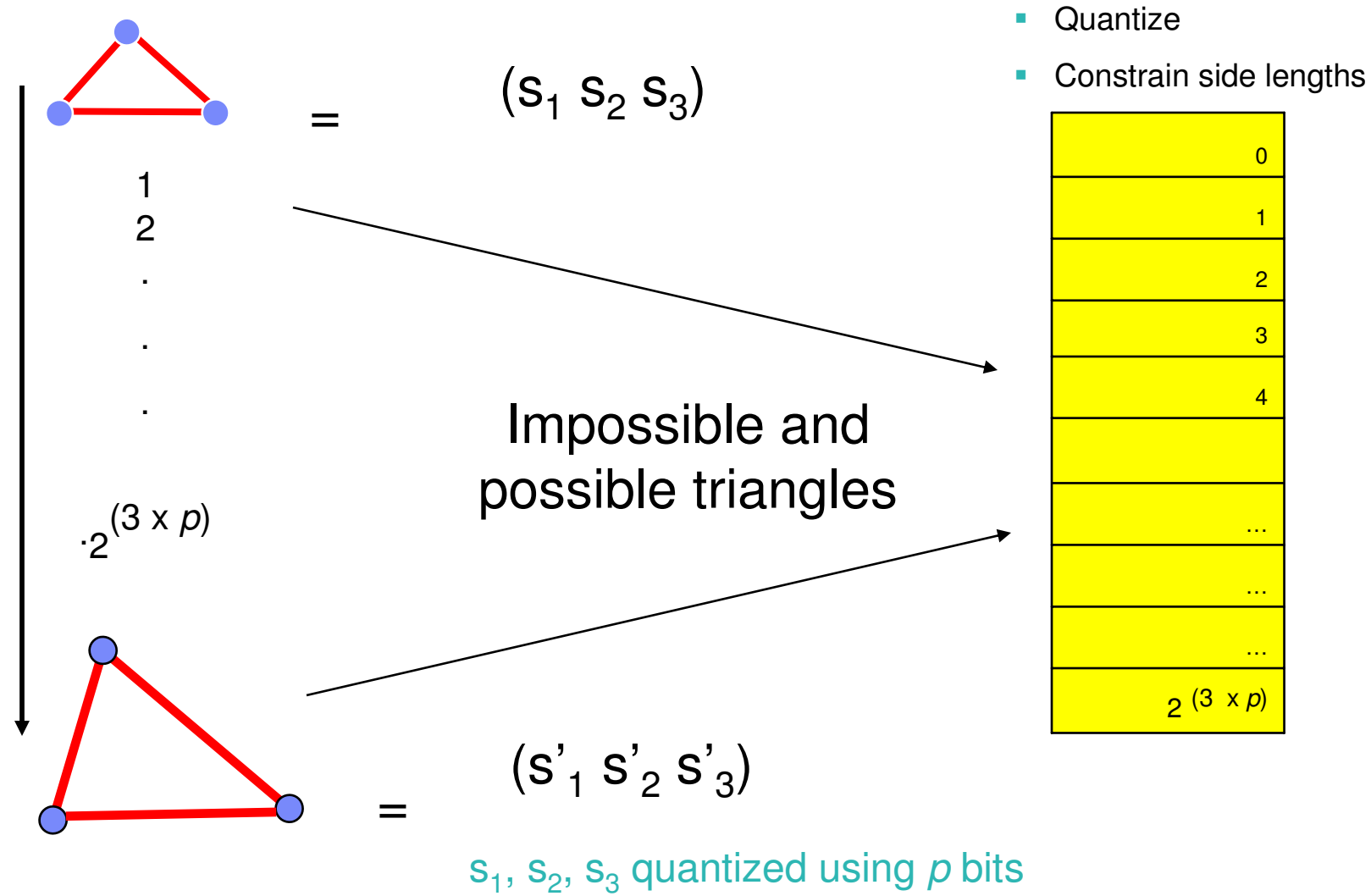
Score distributions for invertible and non-invertible construction

Invariant features

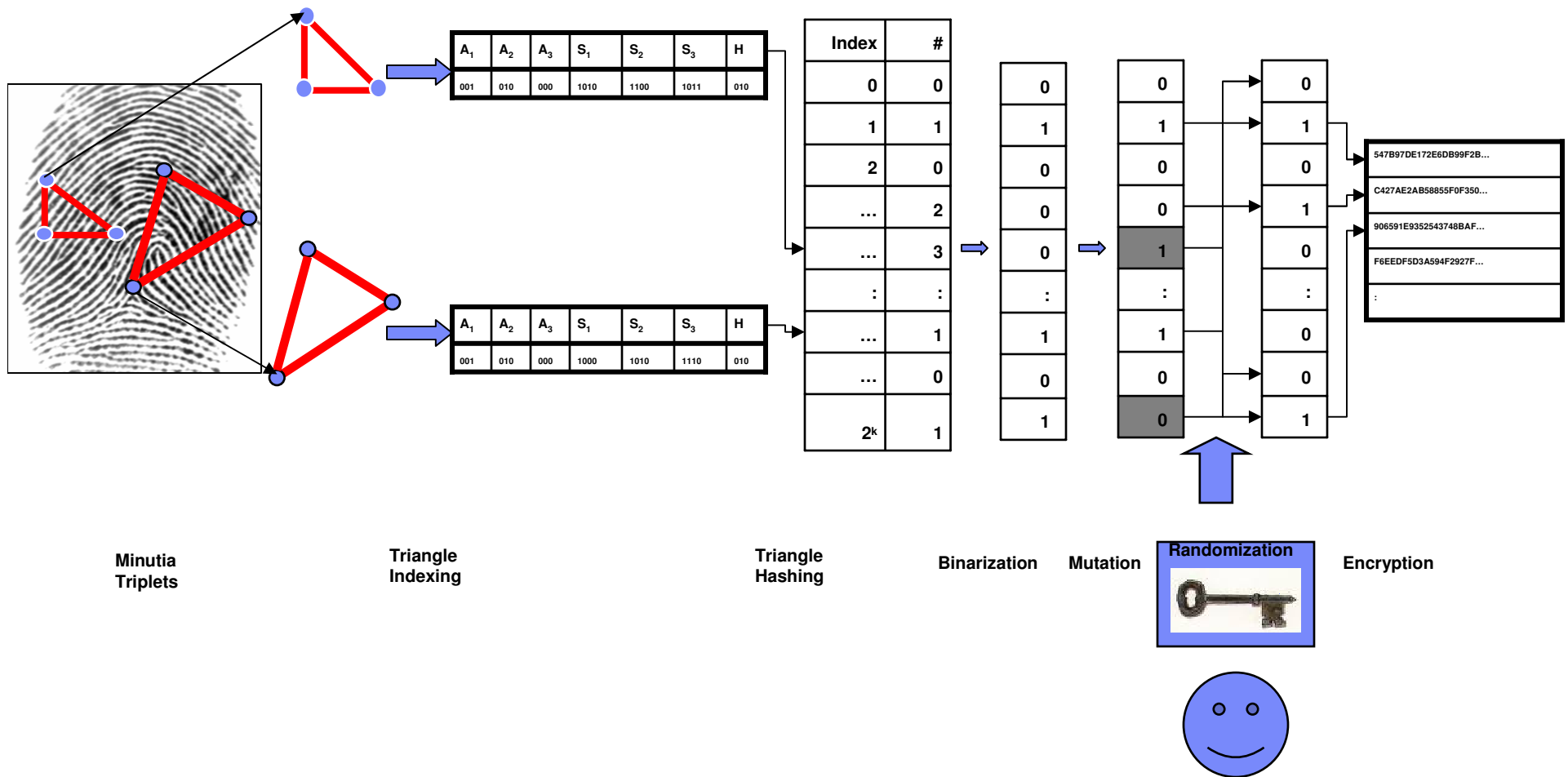
- Independent triangle features
 - The sides
- Dependent triangle feature
 - Height at largest side
- Fingerprint features
 - Minutiae angles with respect to triangle



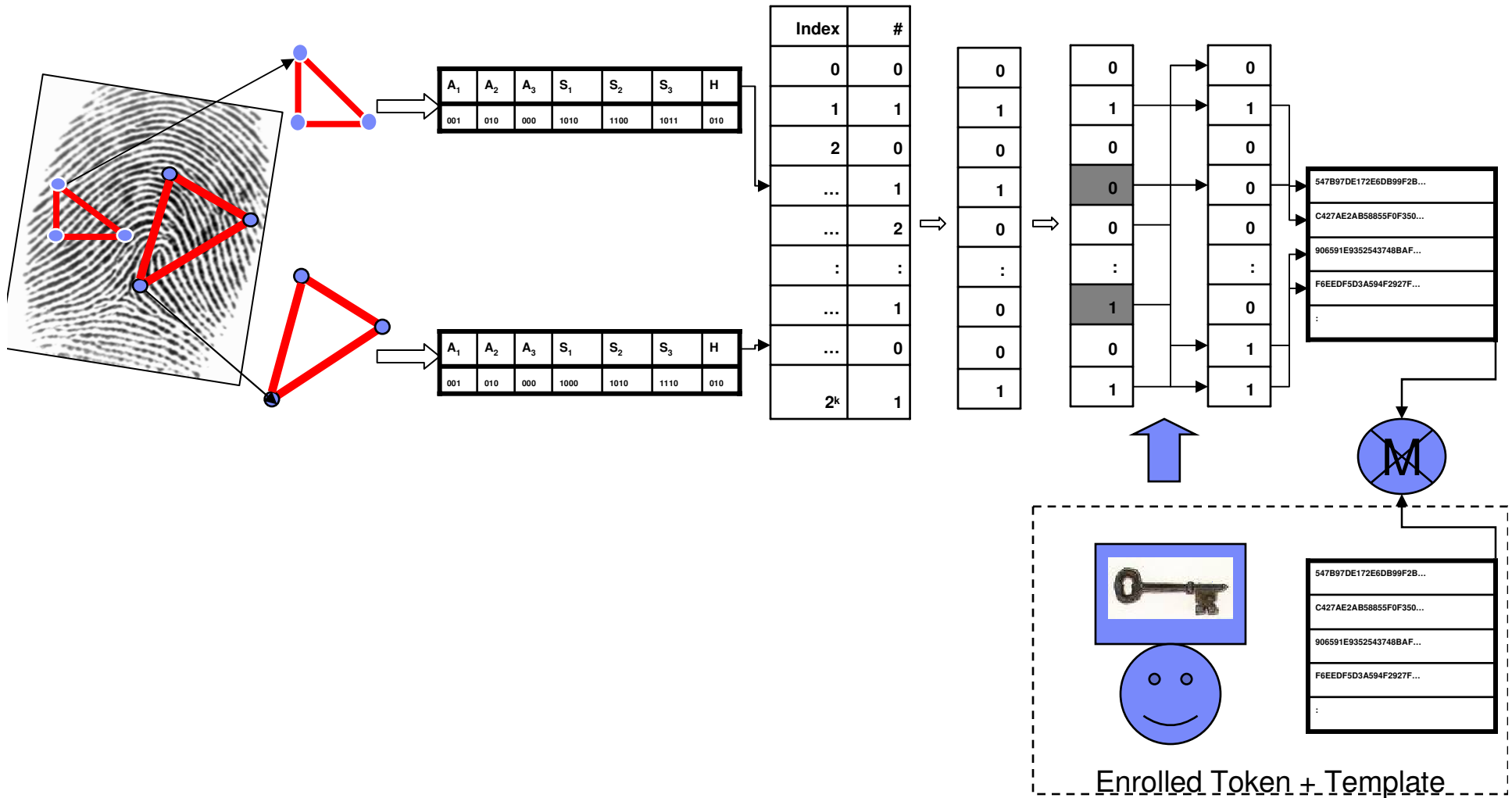
Triangles can be enumerated



Enrolment

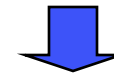
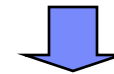
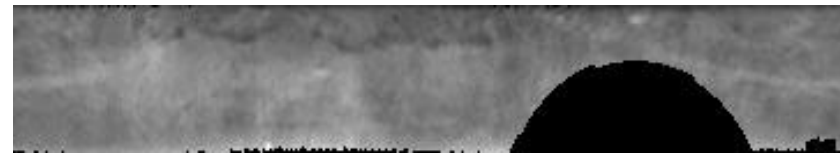
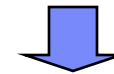
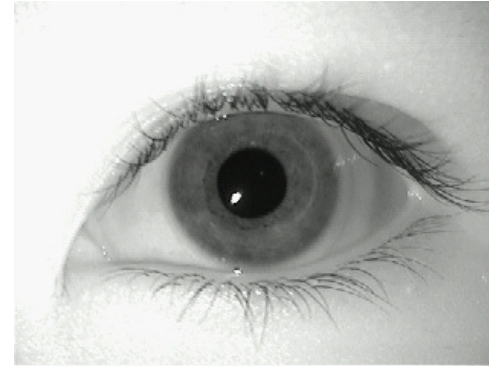


Verification



Steps in building a cancelable iris system

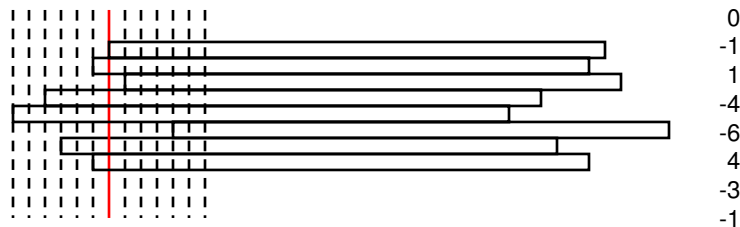
- Segmentation
- Feature extraction
- Cancelable techniques ♦



Method 1: GRAY COMBO

- **template based row shift and combination**

- Step 1: for each row shift circularly:



- Step 2: combine two rows together to get a new one:

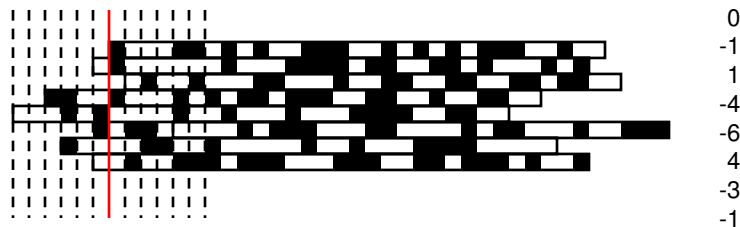
- Intensity +, -
- One row can be used more than once
- Easy methods: odd+even, fold like a mirror

Combine rows 1, 3 to the new 1st row
 Combine rows 2, 8 to the new 2nd row
 Combine rows 4, 6 to the new 3rd row
 Combine rows 5, 7 to the new 4th row

Method 2: BIN COMBO

- code based row shift and combination

- Step 1: for each row shift circularly:



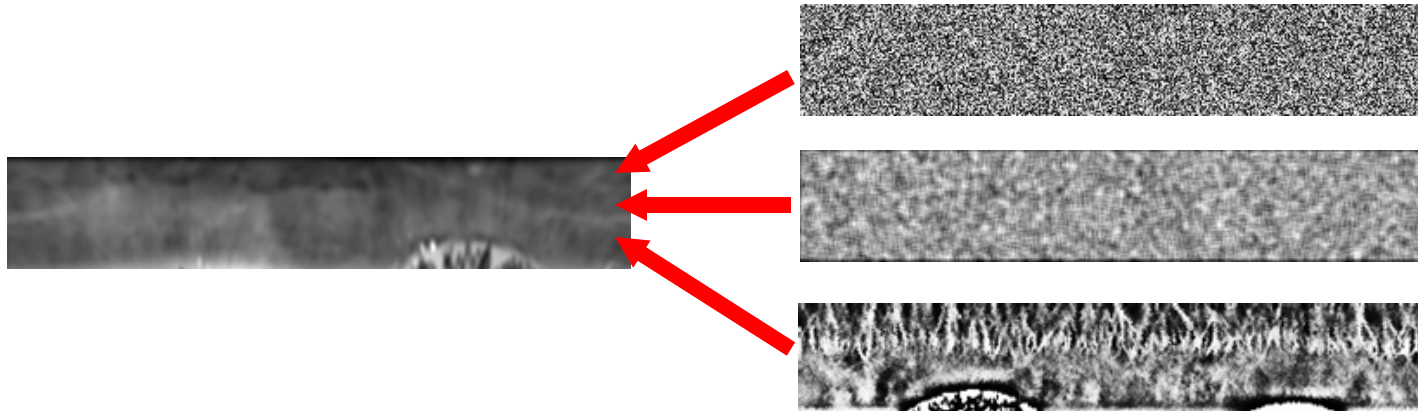
- Step 2: combine two rows together to get a new one:

- Binary XOR, or NXOR
- One row can be used more than once
- Easy methods: odd+even, fold like a mirror

Combine rows 1, 3 to the new 1st row
 Combine rows 2, 8 to the new 2nd row
 Combine rows 4, 6 to the new 3rd row
 Combine rows 5, 7 to the new 4th row

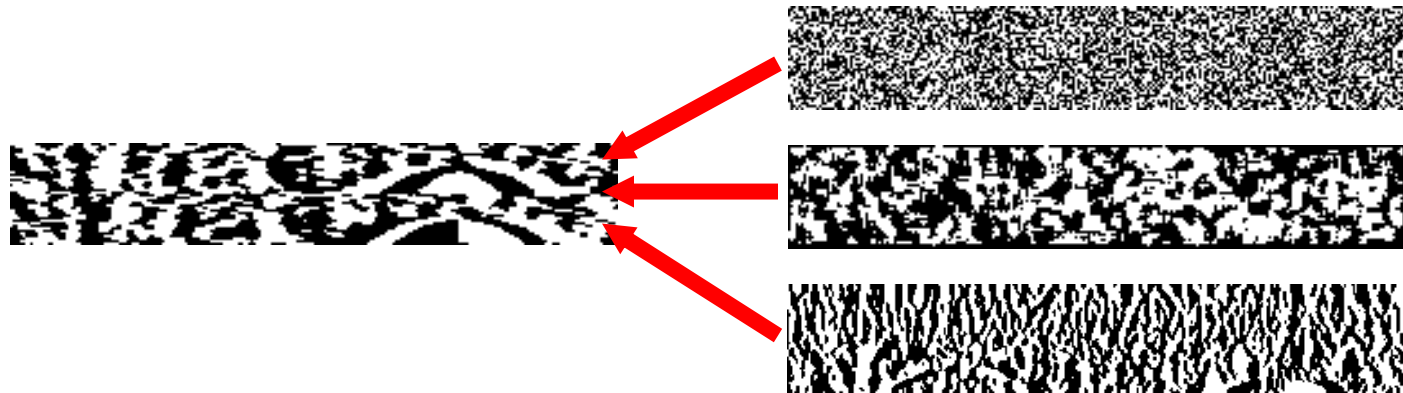
Method 3: GRAY SALT

- **template based salty noise**
 - Just plus a unique pattern --- random noise, random pattern or random synthetic iris texture
 - Generate new code according to the new texture



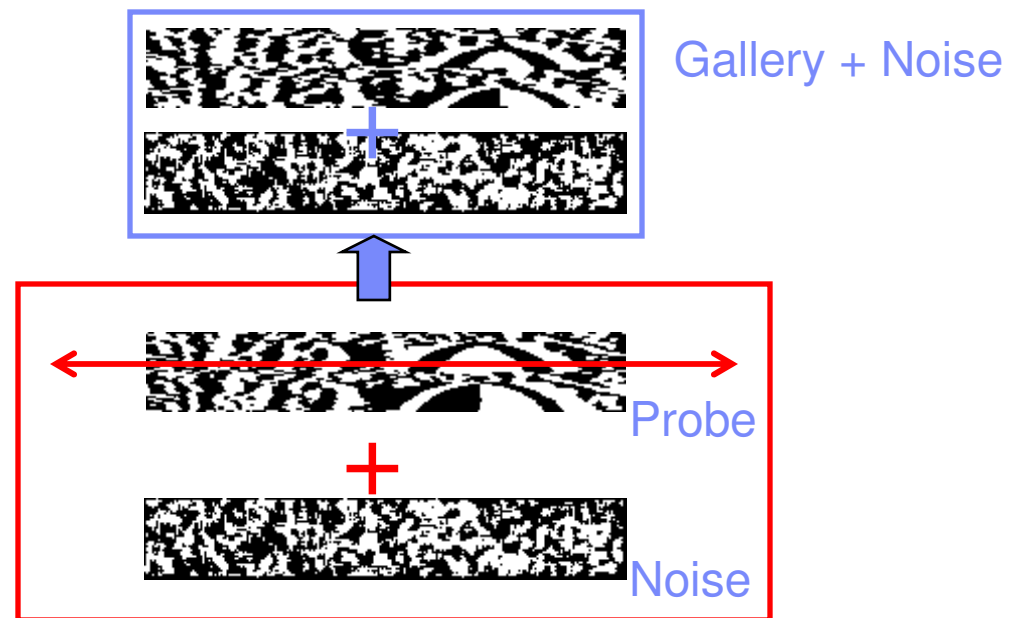
Method 4: BIN SALT

- **code based salty noise**
 - Just plus a unique binary pattern --- random noise , random pattern or random synthetic iris code



Matcher

- Assume head tilt is not heavy
- Matching algorithm need to be modified:



Key performance metrics

- **Accuracy**
 - How do the error rates change?
 - Same transform vs. different transform
- **Transform space**
 - How many transforms are possible?
 - Brute force non-invertible strength of the transform
- **Backward compatibility**
- **Impact on speed**

Thank you