### COPING WITH THE STOCHASTIC BIOMETRICS

Tomáš Rosa Raiffeisenbank, a.s.

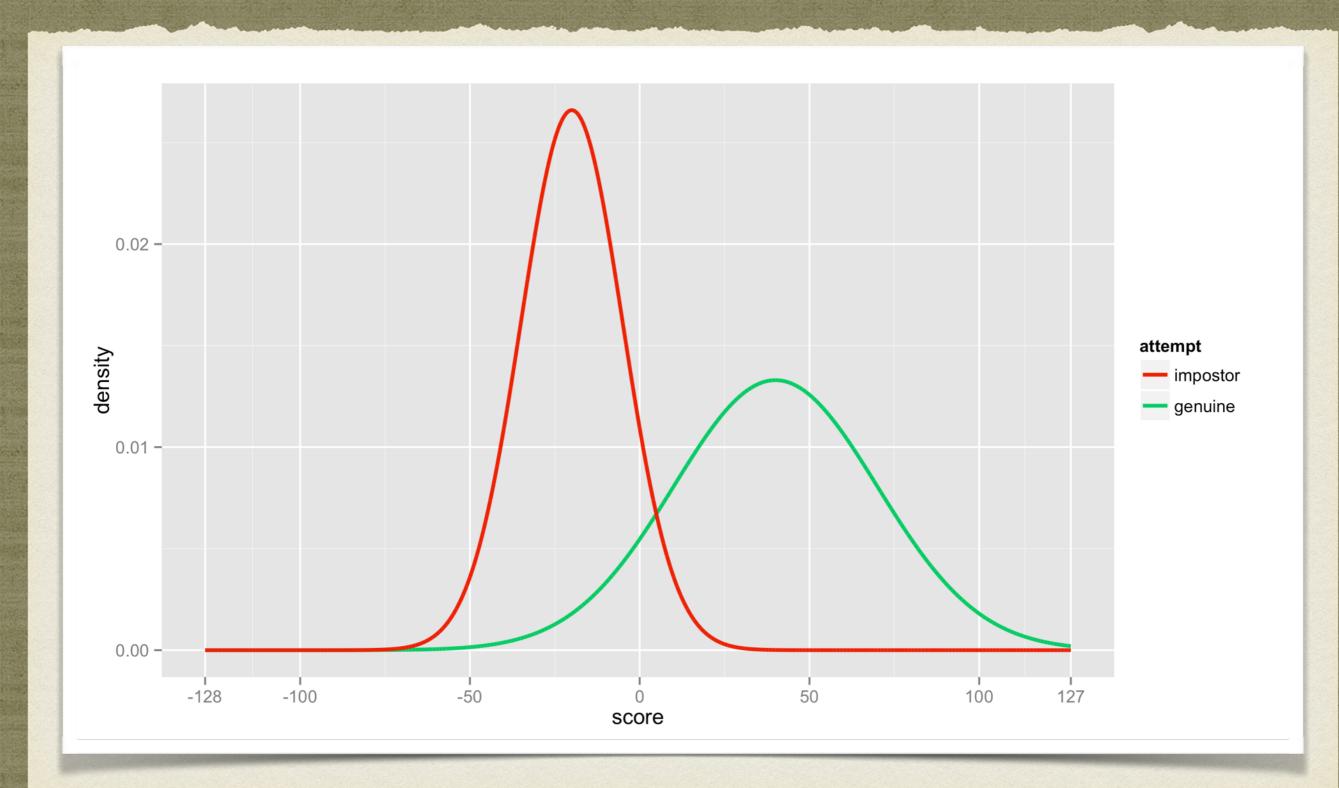
#### SIGNALS PRIMER

- Let a signal be any measurable space-time varying quantity conveying information about a physical phenomena.
- Signal detection is then an ability to discern between information-bearing patterns (signals) and random patterns (noise) that distract from the information.

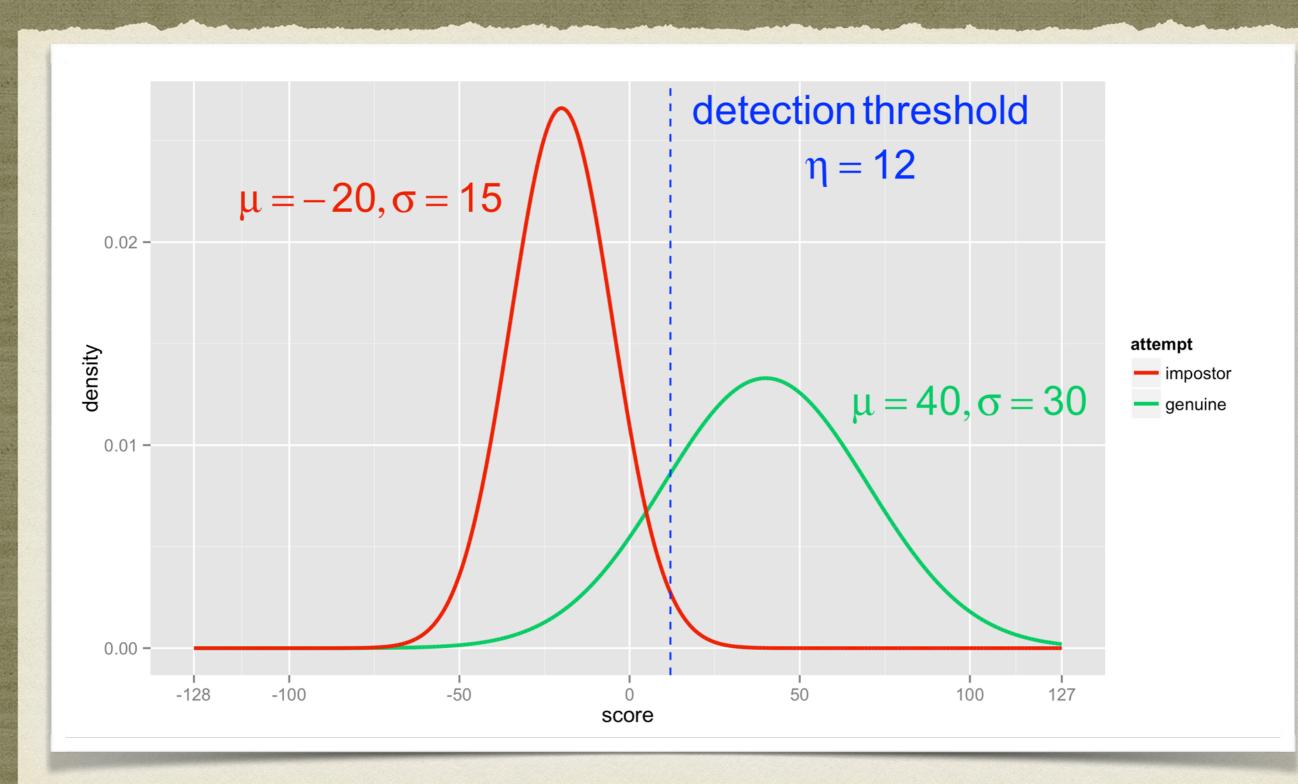
#### MATCH SCORE

- It would be nice if we had a simple true-false result.
  - As in conventional crypto.
  - But we cannot...
- All we have is a value of random variable *X* that follows two conditional distributions.
  - f(x | impostor)
  - f(x | genuine)

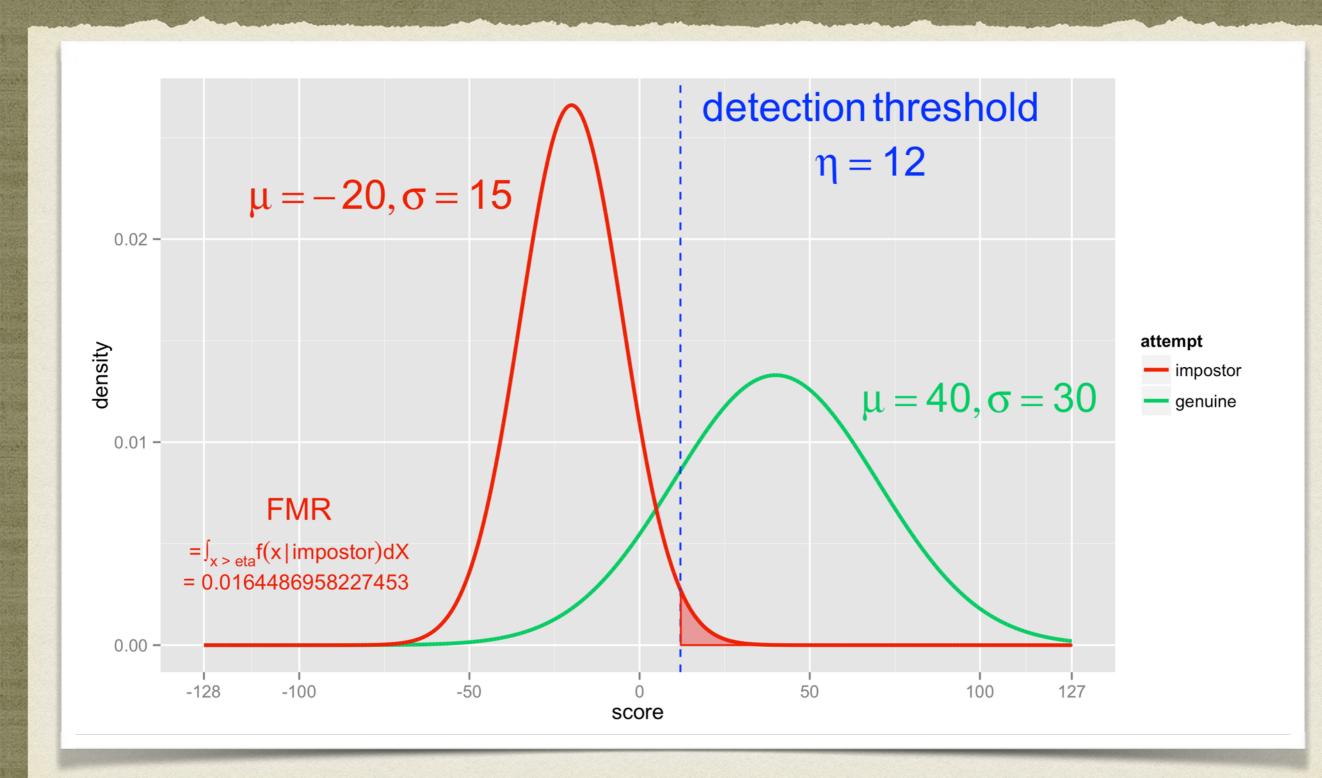
### BASE "CAMEL" GRAPH



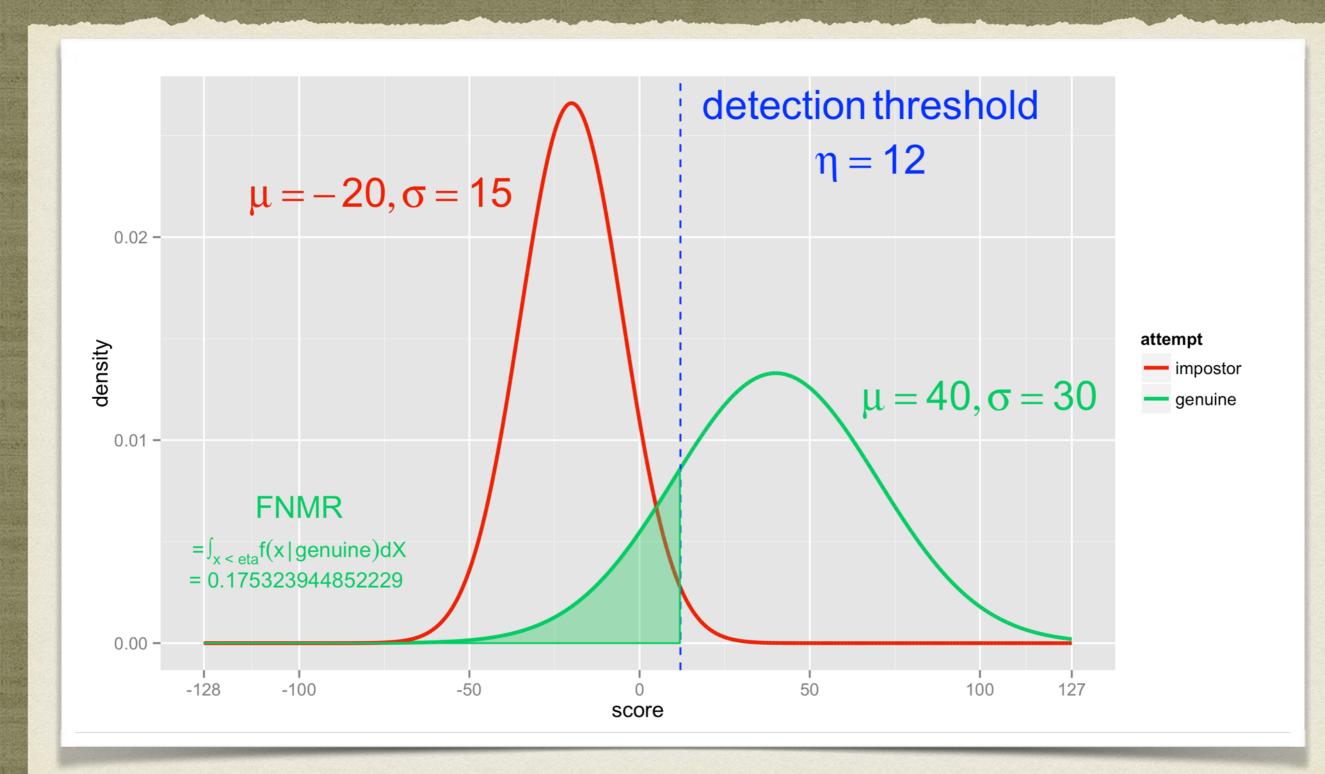
## SIGNAL DETECTION APPROACH



#### FALSE MATCH RATE



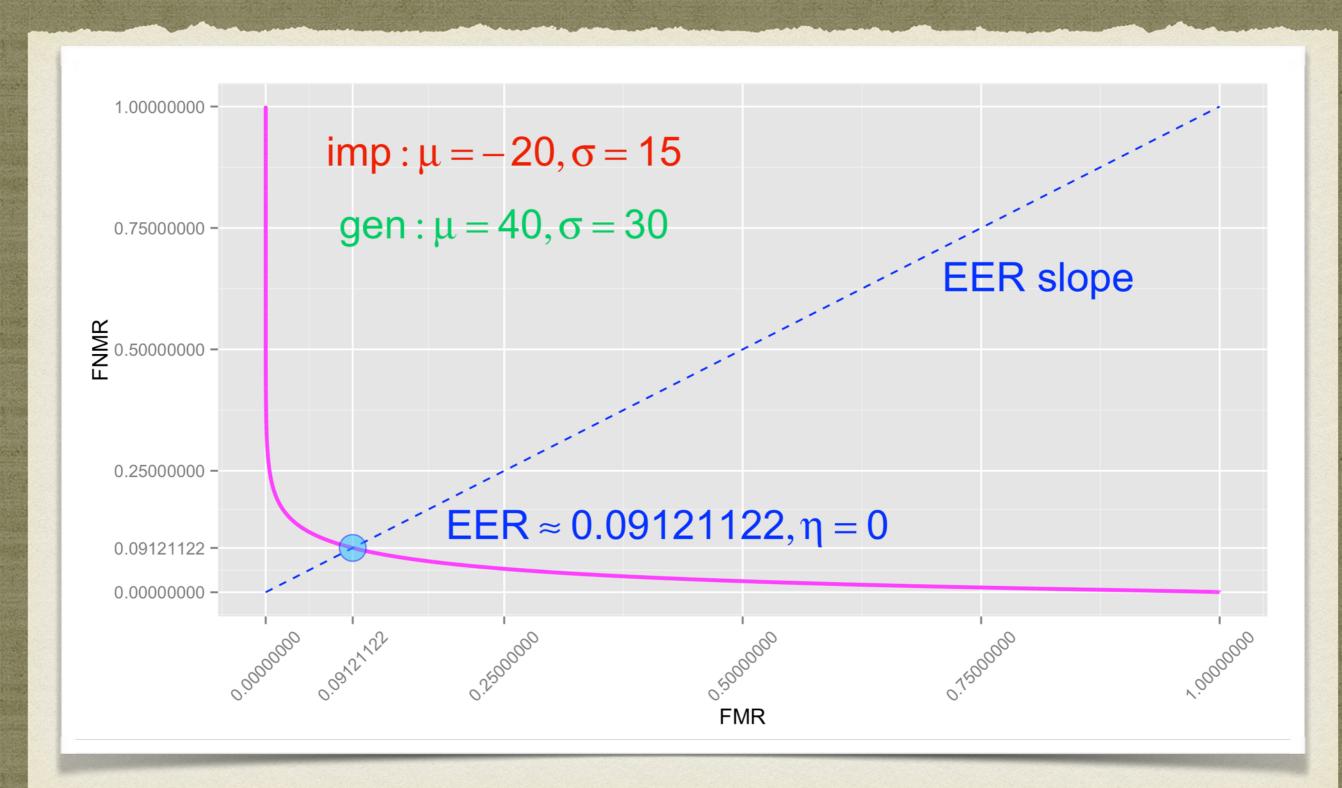
#### FALSE NON-MATCH RATE



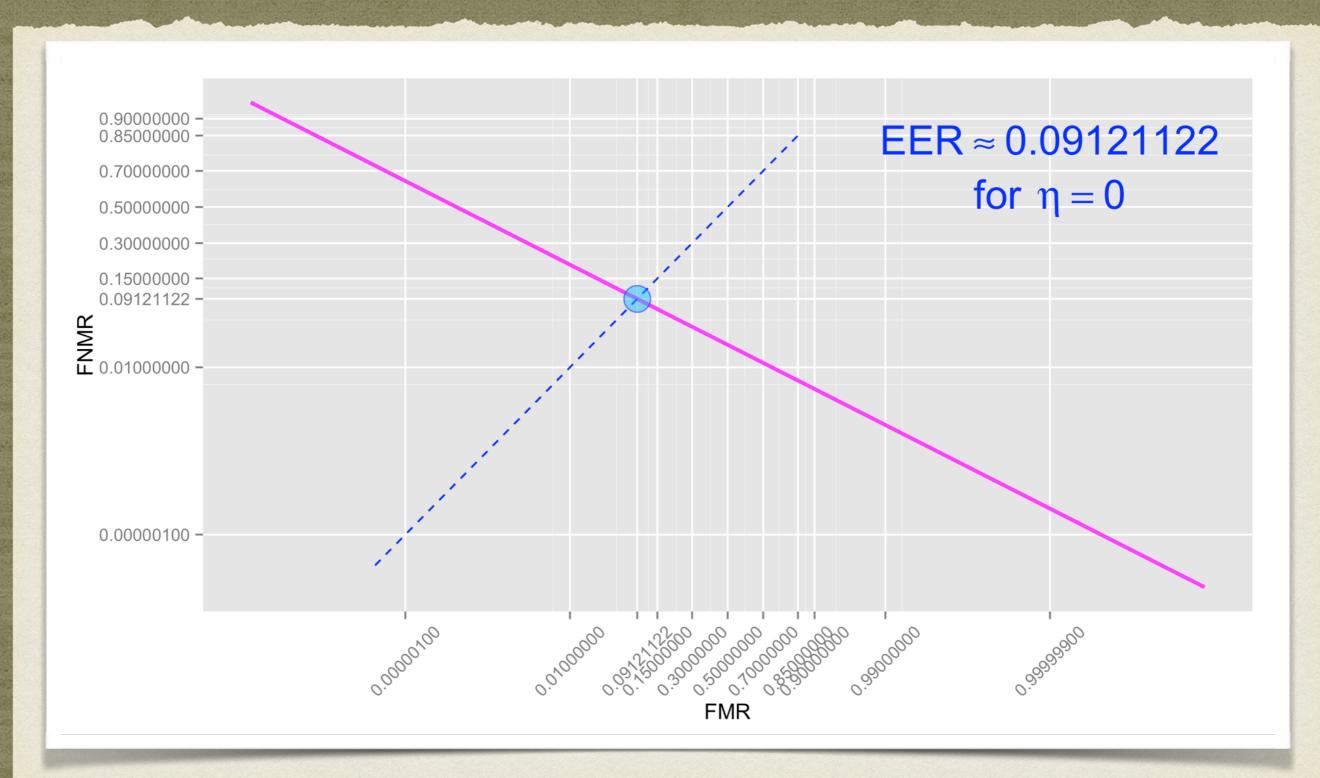
## ERROR DISTRIBUTION FUNCTIONS



### RECEIVER OPERATING CHARACTERISTICS



## DETECTION ERROR TRADE-OFF



### ISO/IEC 19795

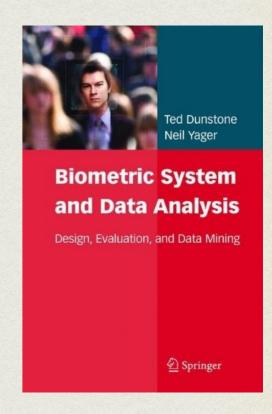
- Performance test methodologies for different lifecycle phases:
  - technology evaluation
  - scenario evaluation
  - operational evaluation
- We get comparable results with plausible confidence intervals.

#### BUNCH OF PARAMETERS

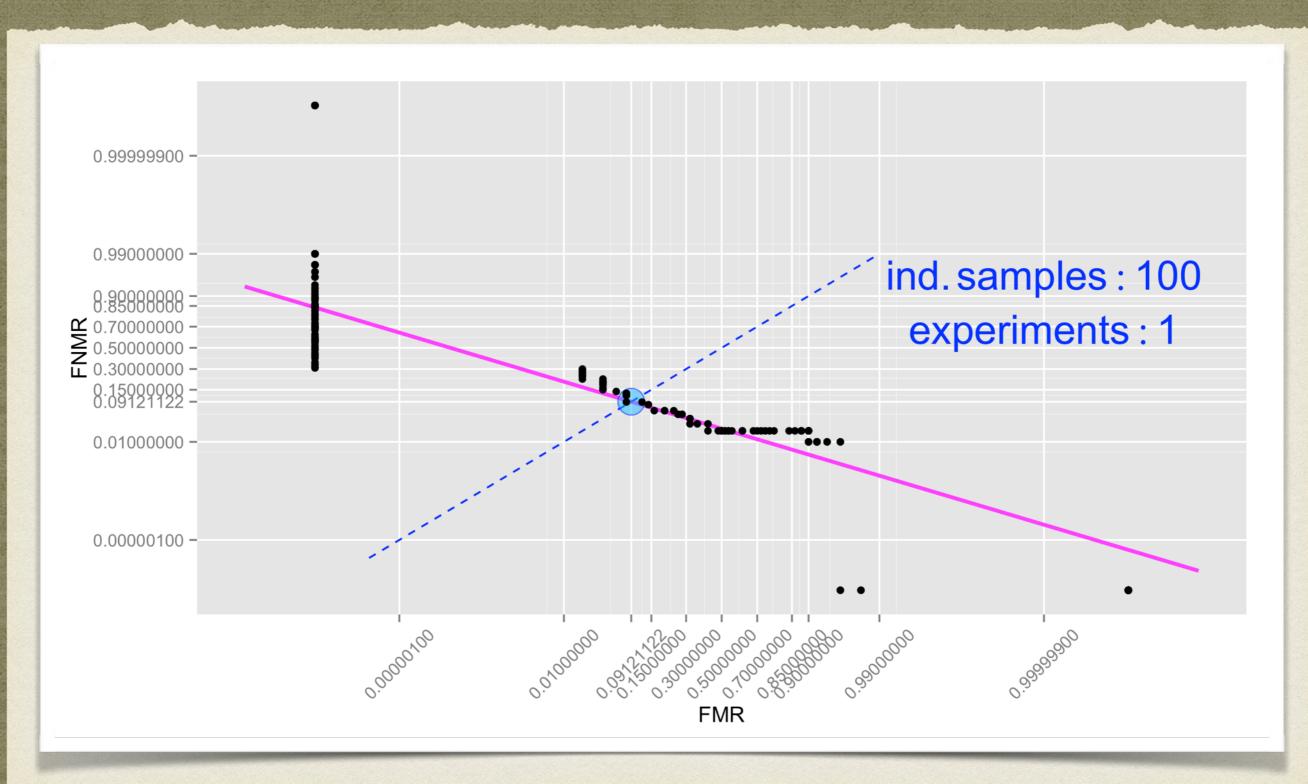
- False Match Rate / False Non-Match Rate
  - attempt oriented
- False Acceptance Rate / False Rejection Rate
  - transactional version of FMR/FNMR
- Failure To Acquire
- Failure To Enroll
  - both attempt and txn-oriented versions

### BIOMETRIC DATA MINING

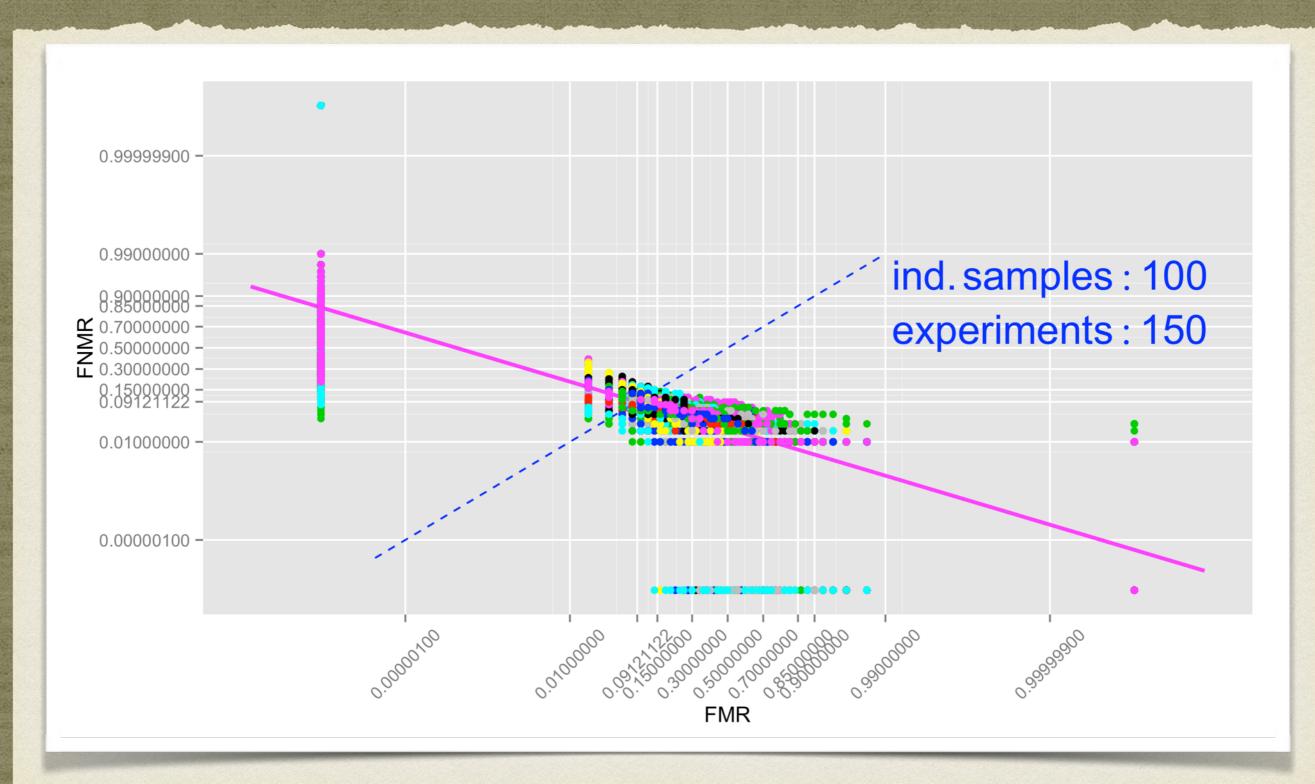
- In any life-cycle phase, we shall gather as much data as we can to estimate the performance or check we are still operating in expected margins.
- Anomalies may indicate a component malfunction or even a fraud.
- Again, be careful about confidence.
- Misleading statistics can be worse than none!



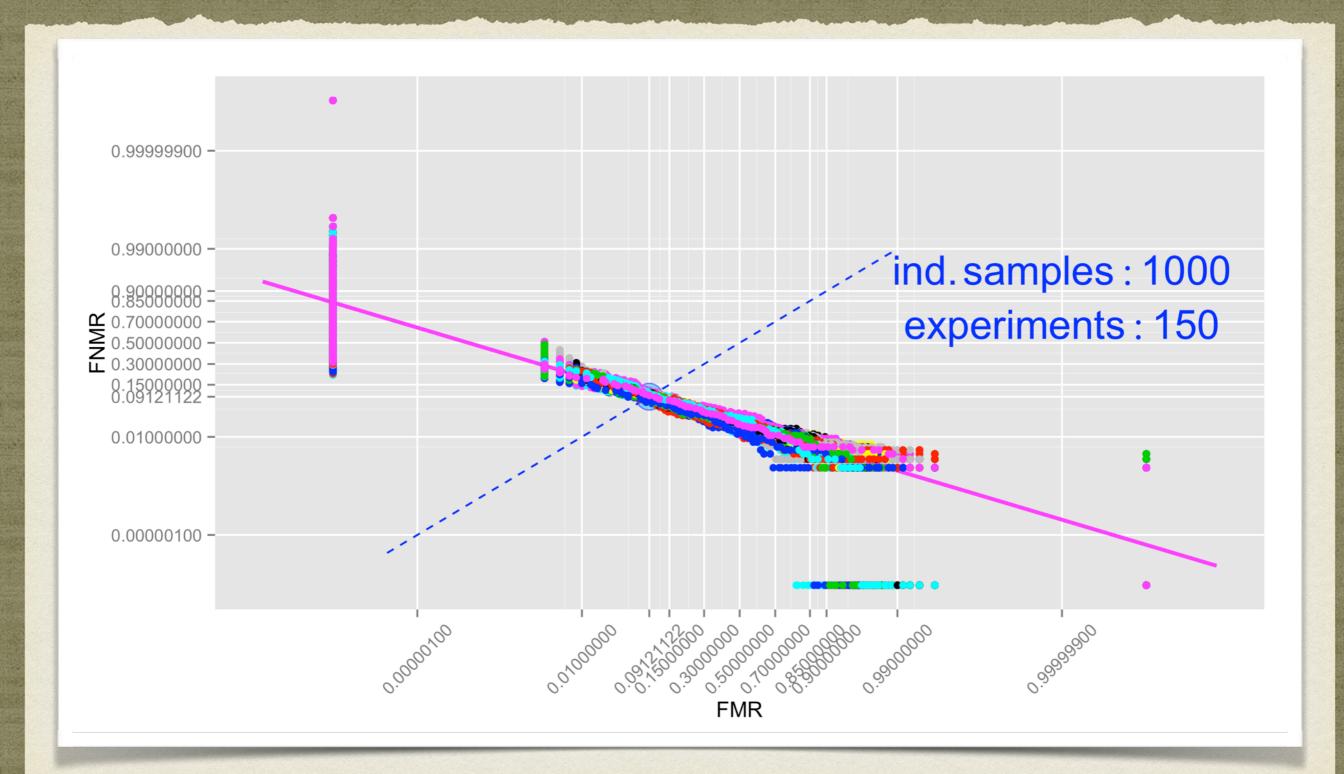
## DET ESTIMATION SIMULATION



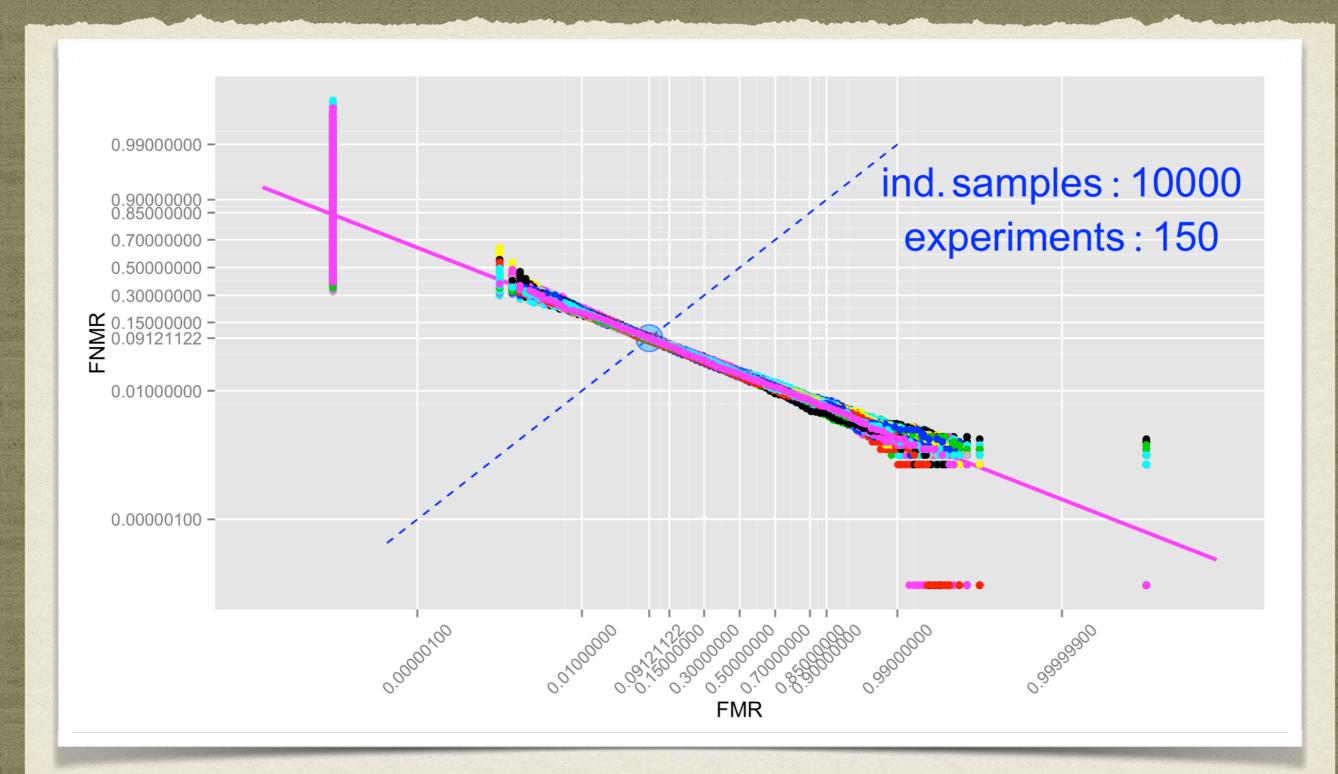
## CONFIDENCE INTERVALS?!



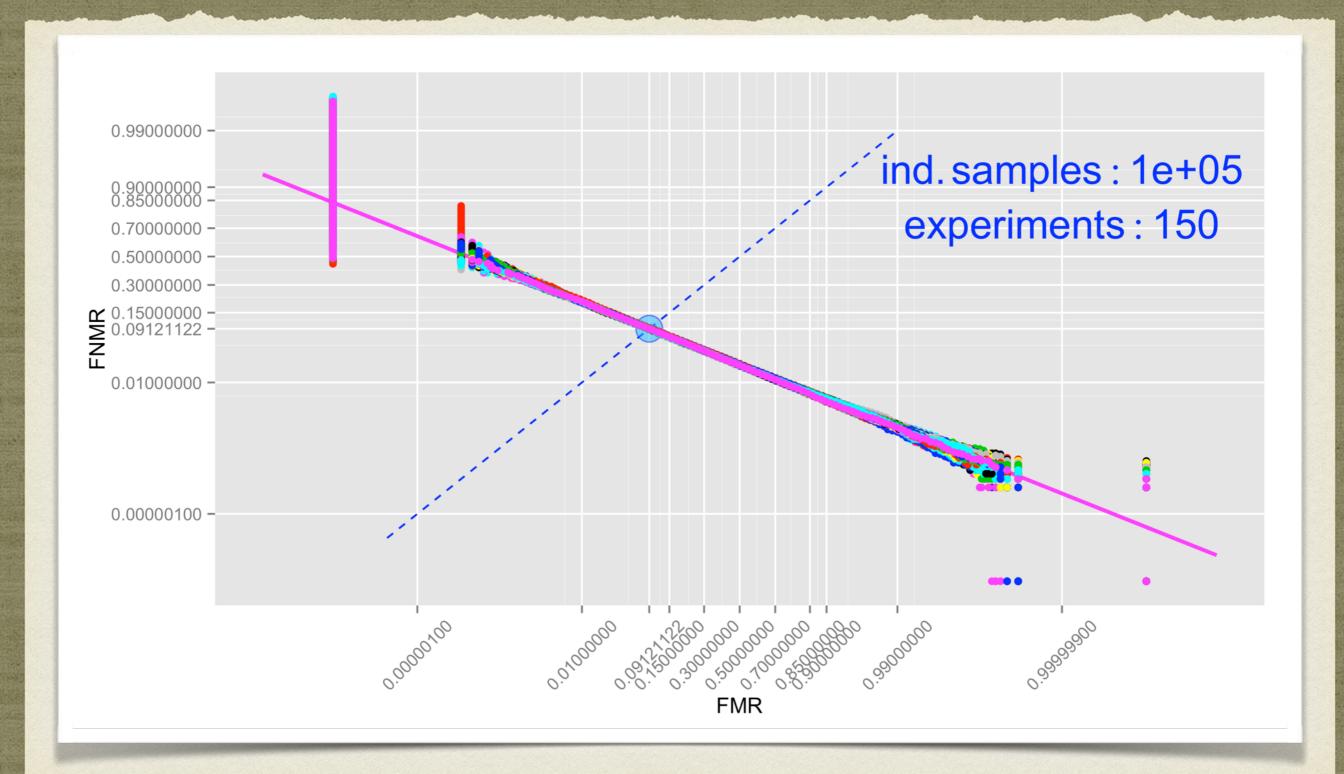
### ANY CONFIDENCE, YET?



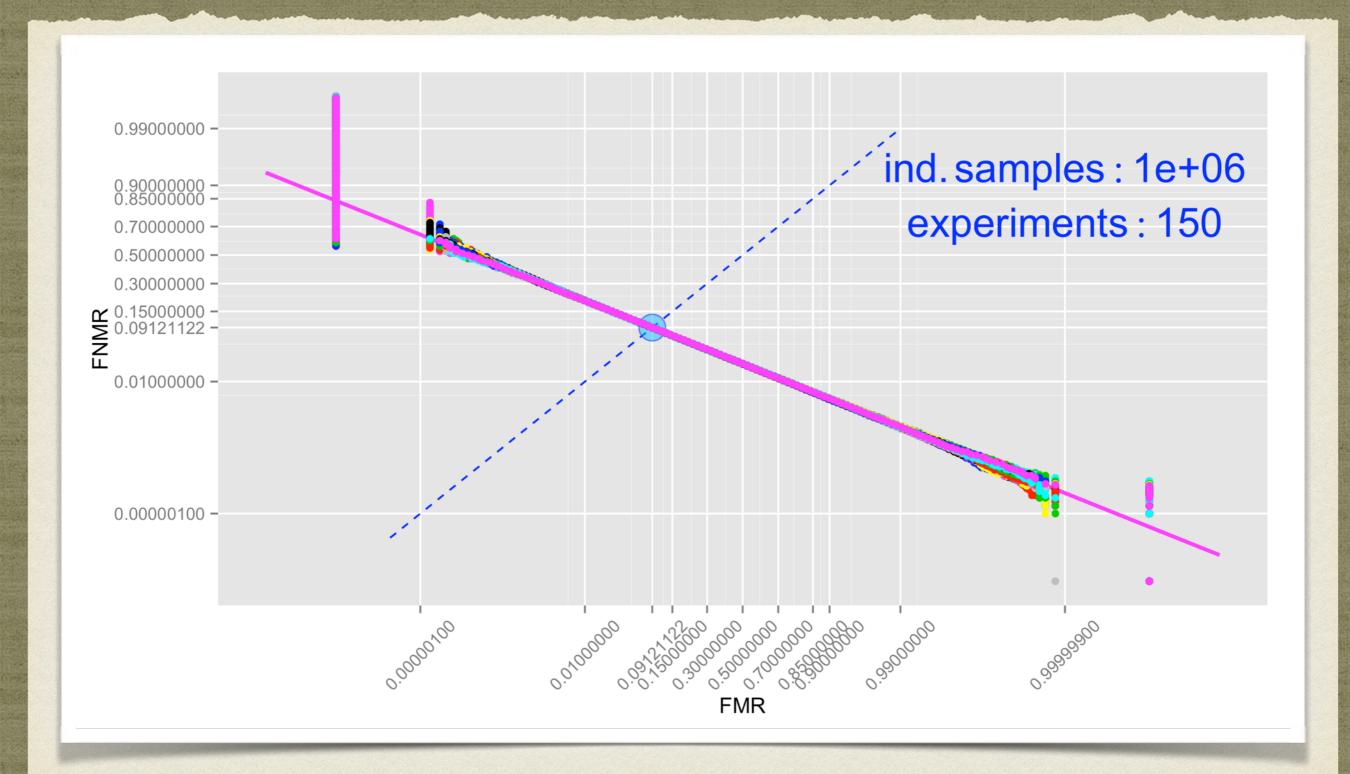
### FAIR CONFIDENCE



#### WE CAN BE PROUD



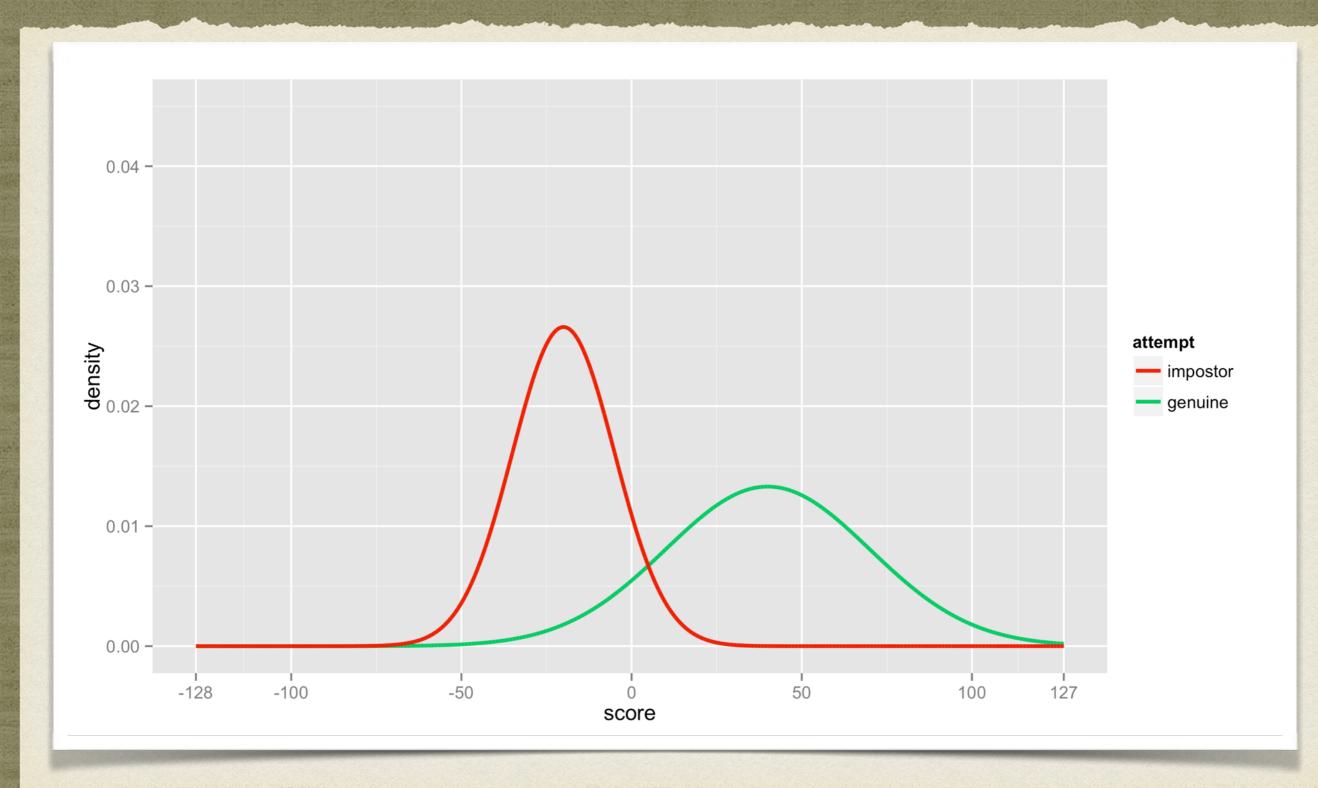
### JUST A DREAM...



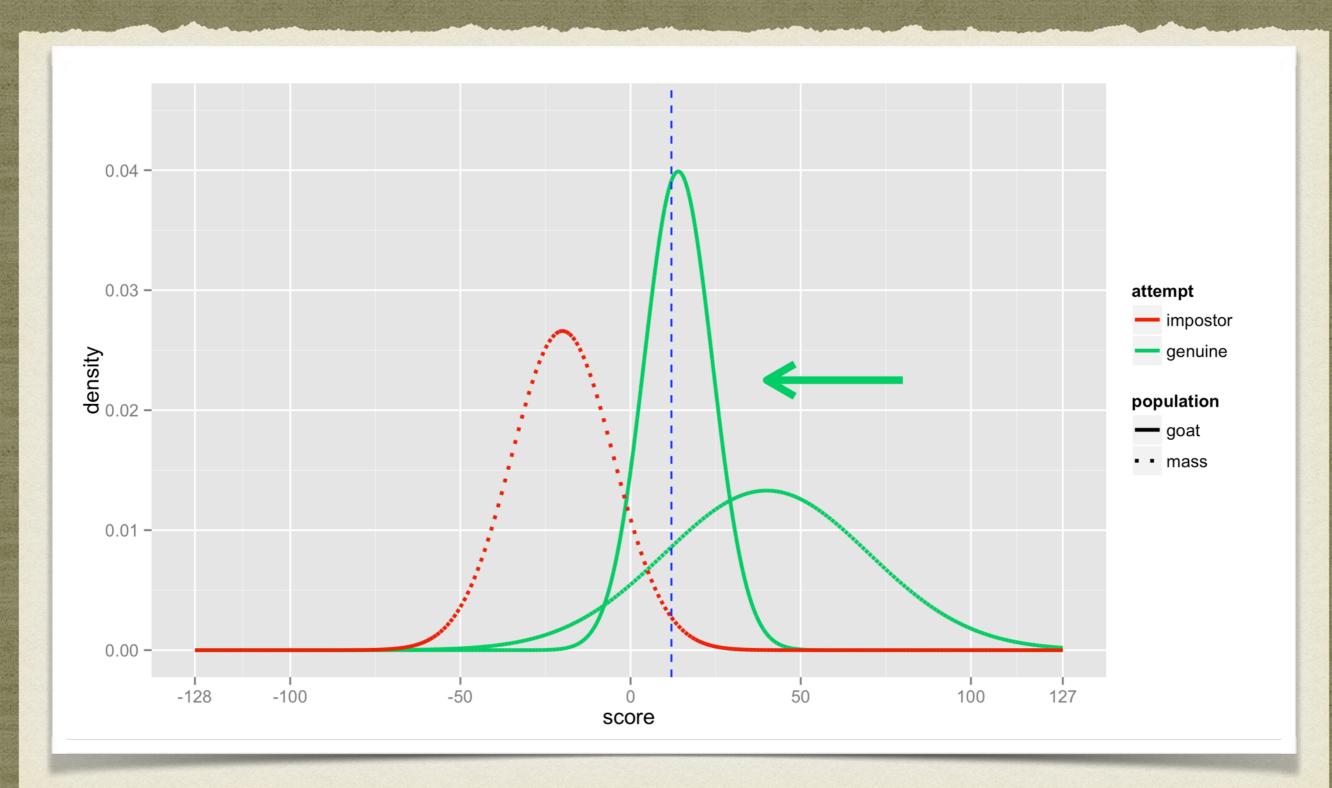
#### BIOMETRIC MENAGERIE

- To further complicate biometrics testing, those score distributions are usually not person-independent.
  - That means the performance is not the same for all people.
- There are plenty of anomalies out there we shall be aware of to interpret the system behaviour correctly.

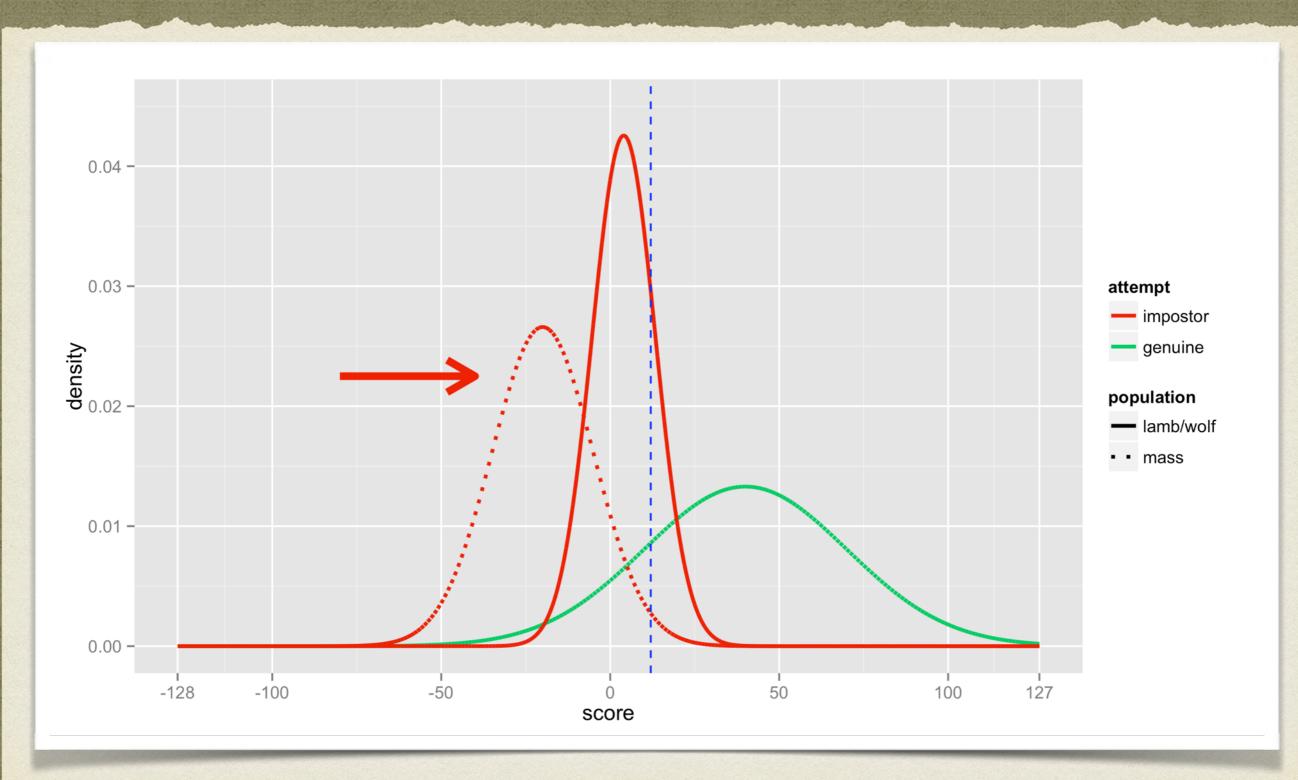
# SHEEP: AN ORDINARY USER



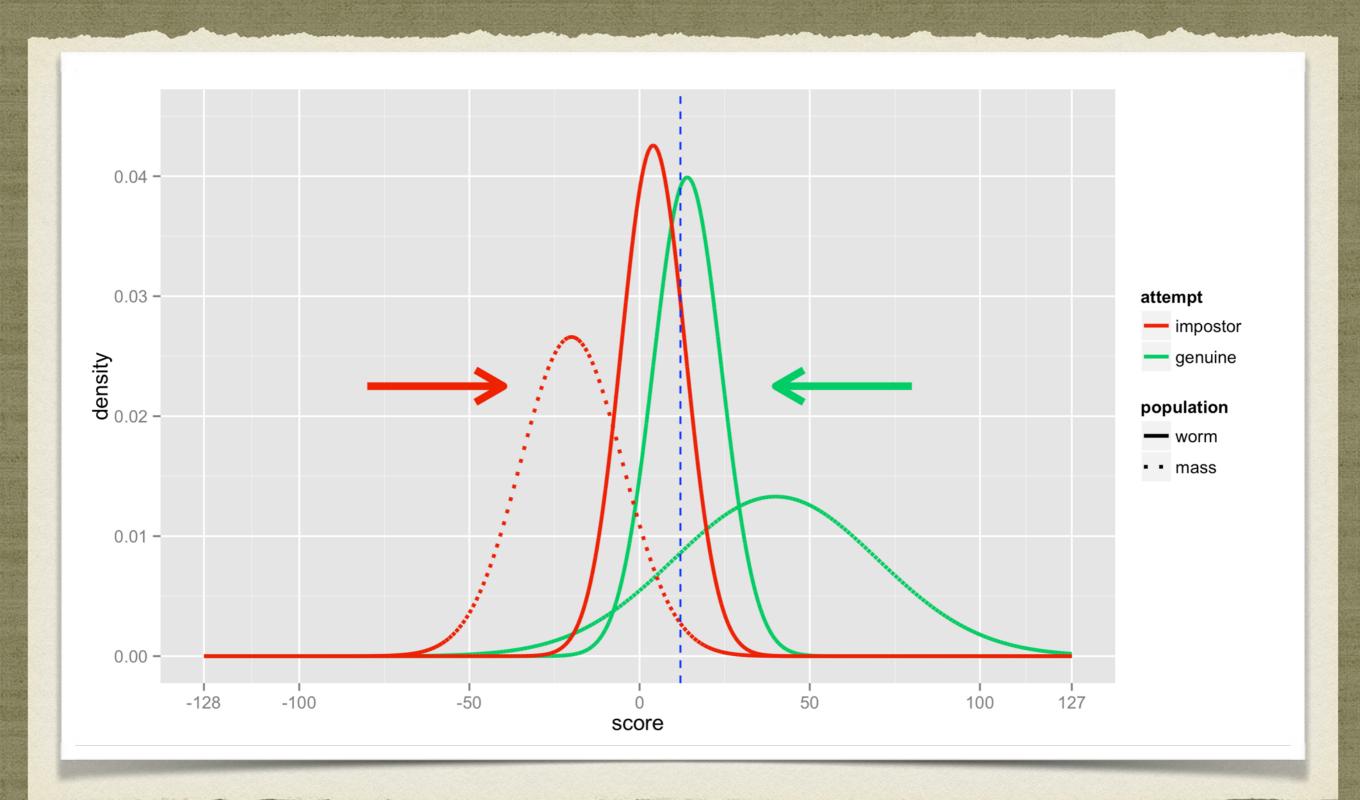
## GOAT: PROBLEMATIC FNMR



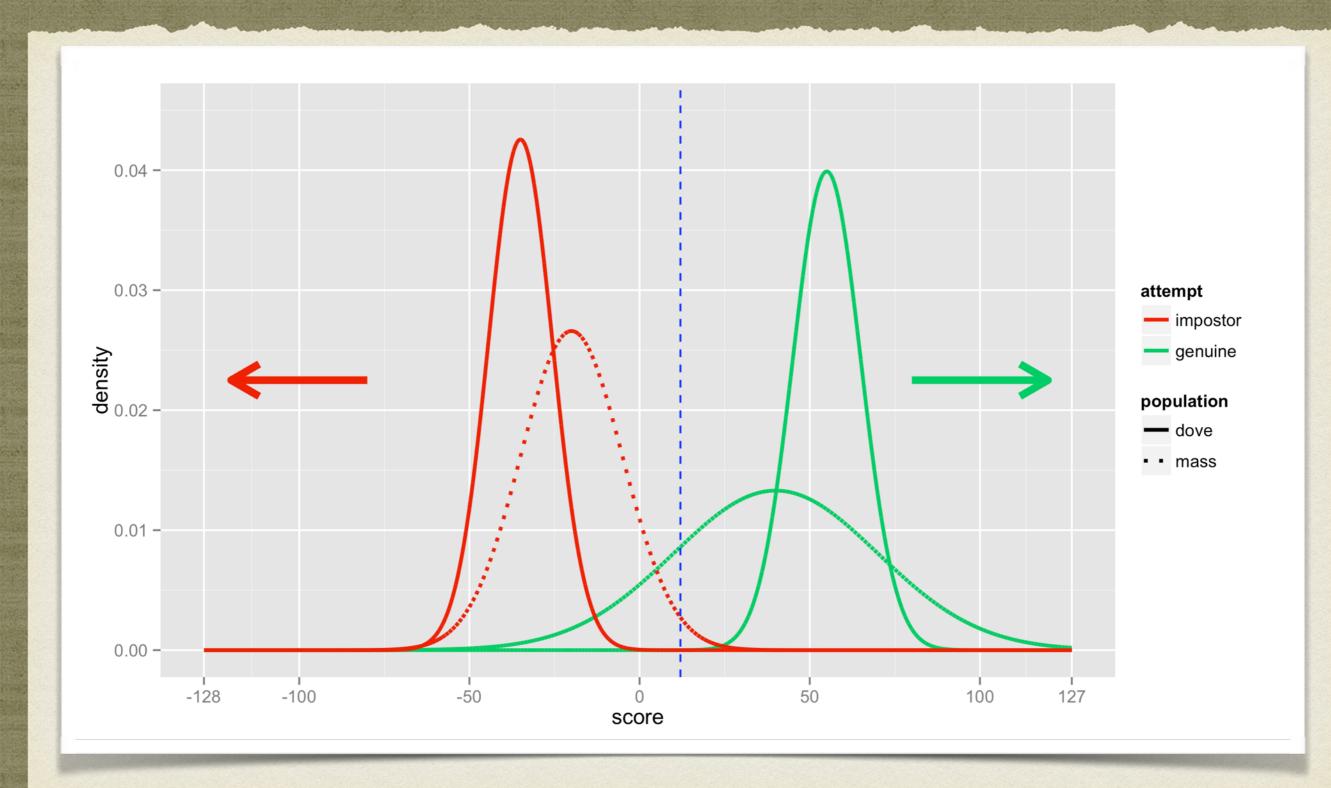
#### LAMB/WOLF: EASY TARGET AND-OR EFFECTIVE PREDATOR



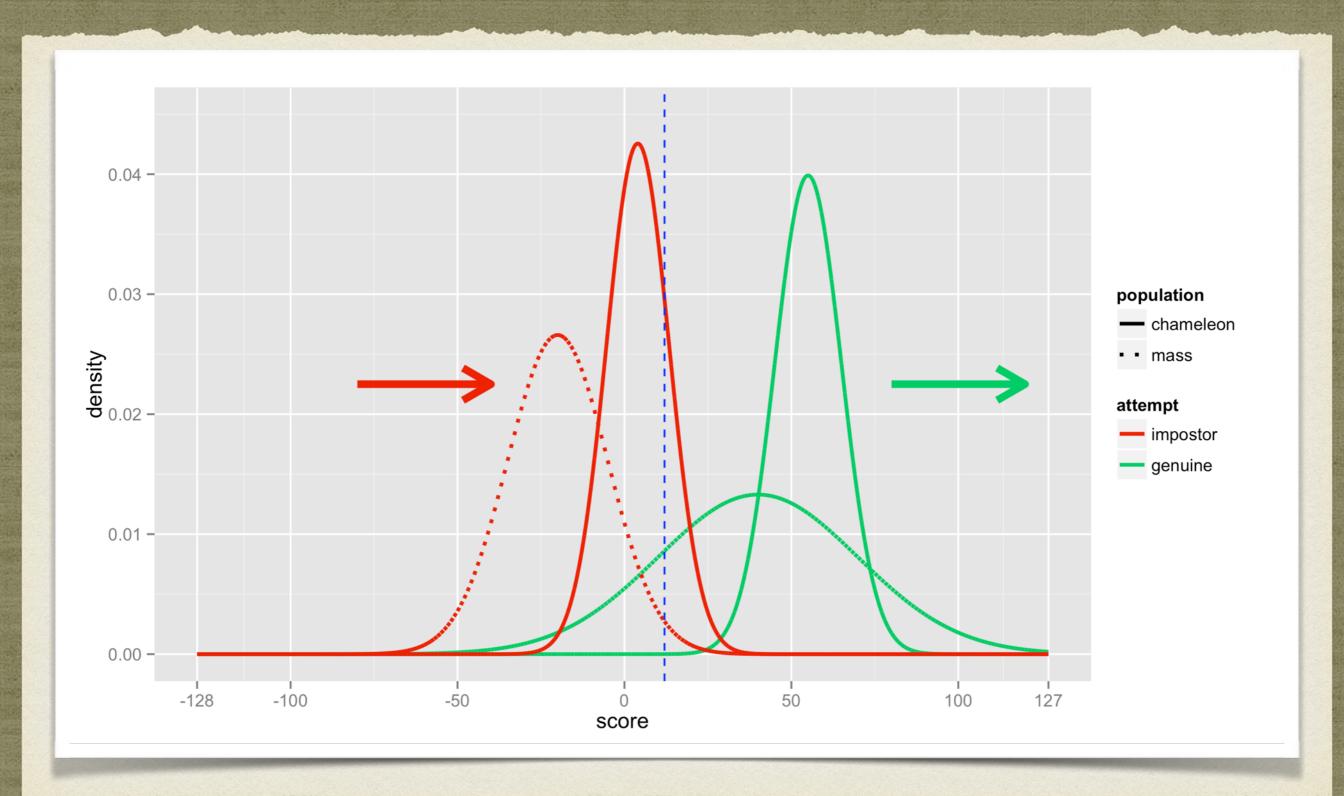
#### WORMS: BOTH FNMR AND FMR INCREASED



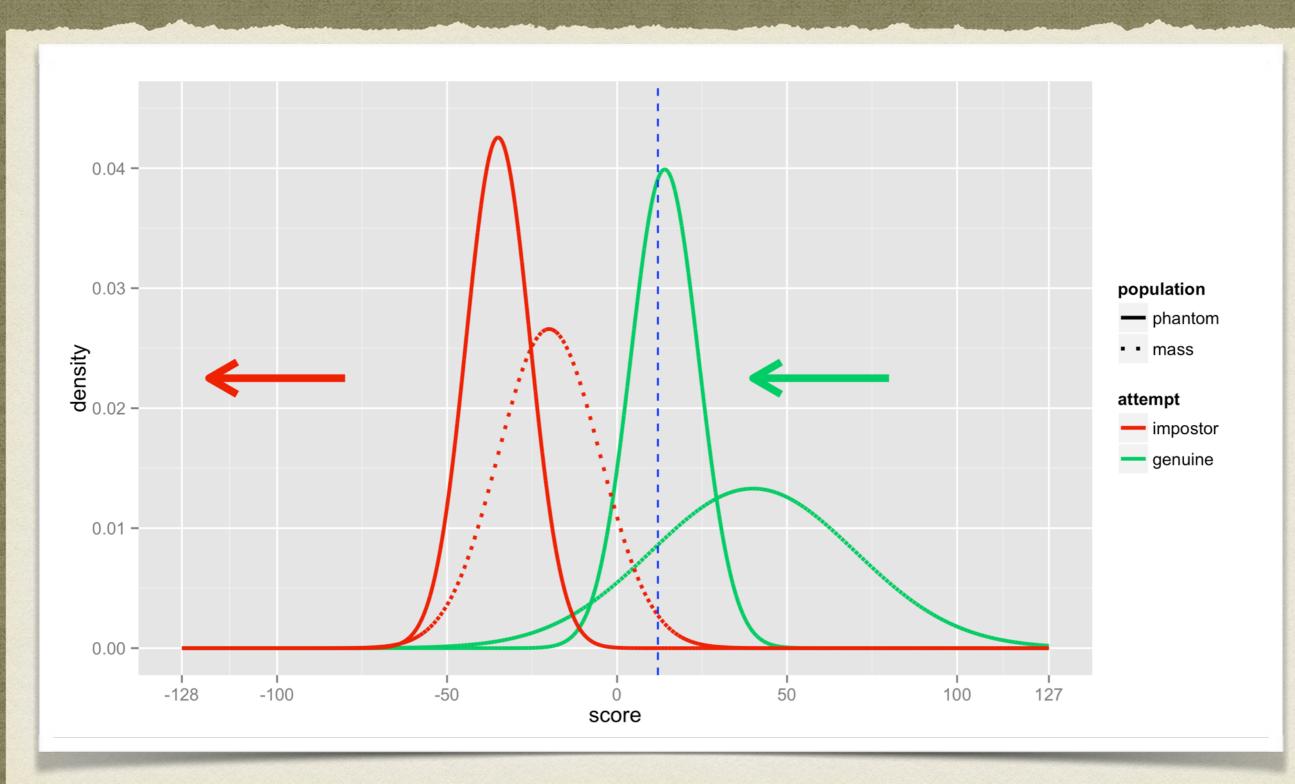
### DOVE: EXCELLENT USER



#### CHAMELEON: EXCELLENT SCORES, ANYWAY(!)



#### PHANTOM: PROBLEMATIC MATCHING, ANYWAY





### BIO BRUTE FORCE ATTACK

- Randomly generate plausible circa 1/FMR samples and put them to the test.
  - Also termed "Zero-Effort", denoting that the attacker makes no special effort to imitate the original person characteristic.
- Synthetic samples generation is quite feasible today.

#### BIOMETRIC INVERSE PROBLEMS

Svetlana N. Yanushkevich Adrian Stoica Vlad P. Shmerko Denis V. Popel



## CRYPTANALYSIS-LIKE ATTACKS

- Masquerade attacks, can be a variant of "Hill-Climbing" denoting the attacker iteratively improves the BIO sample data based on:
  - scoring feedback (side channels)
  - stolen template (pre-image attacks)
  - independent template trained from intercepted BIO samples (correlation attacks)
  - known scoring anomaly (differential analysis)
  - implementation faults (general hacking)

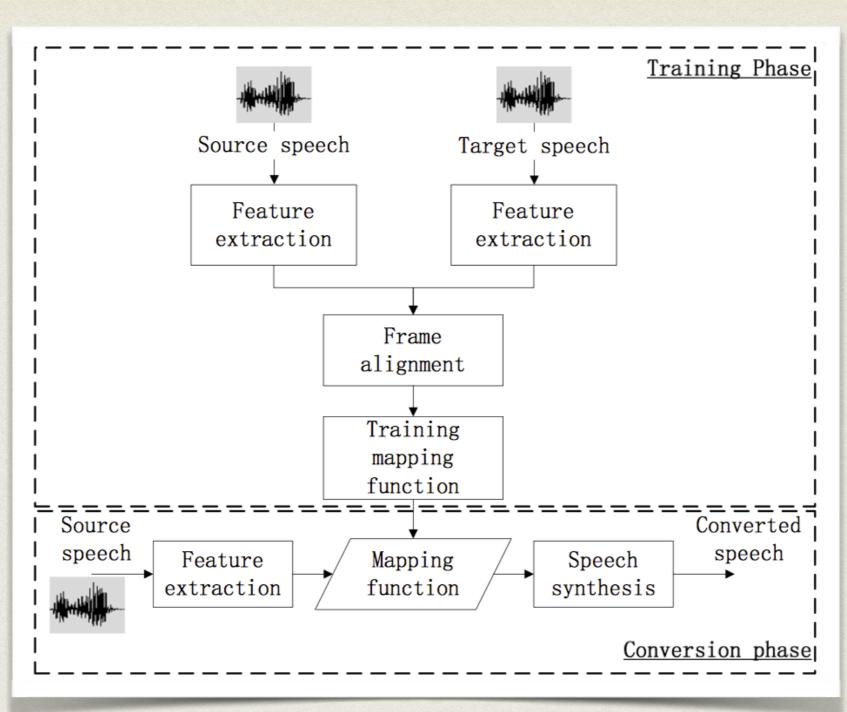
#### SPOOFING

- The process of defeating a biometric system through the introduction of fake biometric samples.
  - (Schuckers, Adler et al., 2010)
- Particular modus operandi on how to deploy the attacking data vectors.
  - Can be seen as being orthogonal to the aforementioned ways of gaining fake samples.

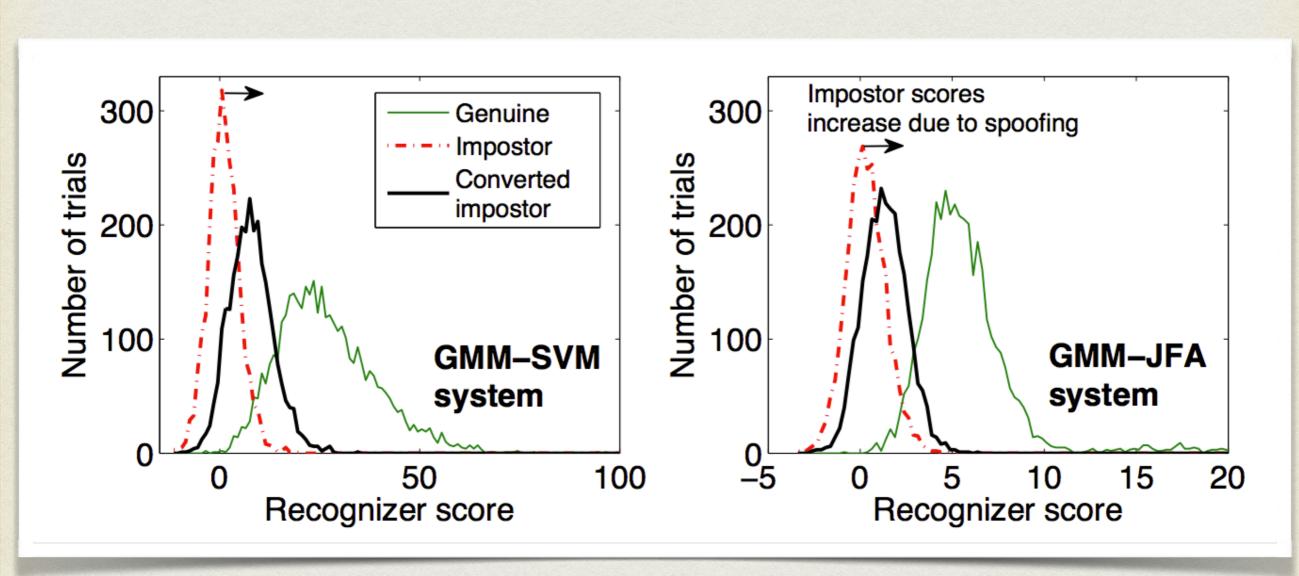
#### SENSOR-BYPASS ATTACKS

- Do not expose API service for unrestricted automated sample verification!
  - Recall the zero-effort attack complexity is often trivial.
  - Furthermore, masquerade attacks can shift FMR significantly.

### CONVERSION ATTACK EXAMPLE

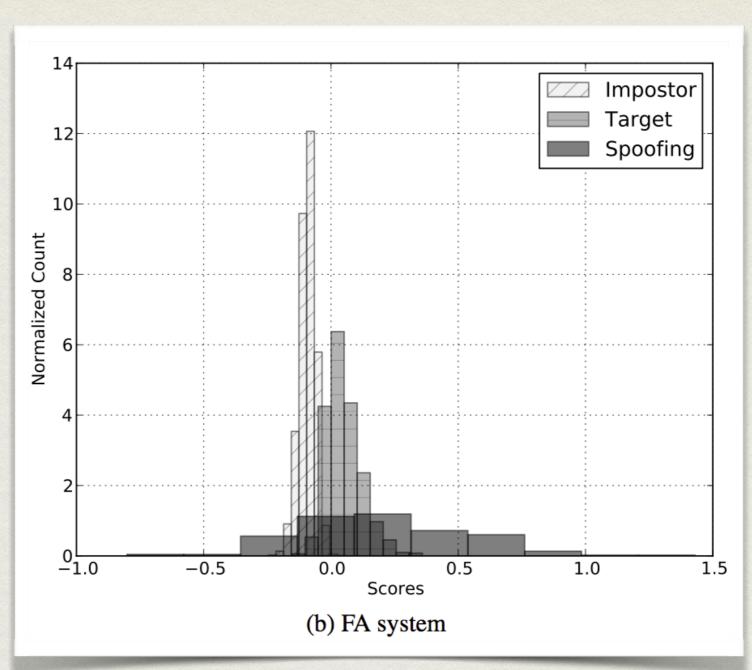


## REPORTING ATTACK IMPACT



Kinnunen et al., ICASSP 2012

## ARTIFICIAL SIGNALS IMPACT

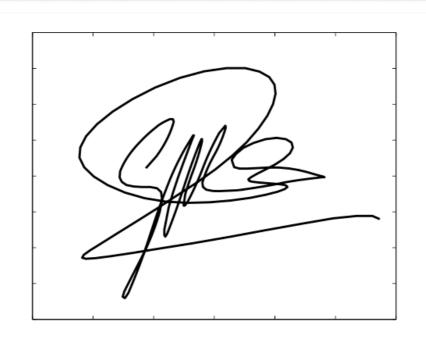


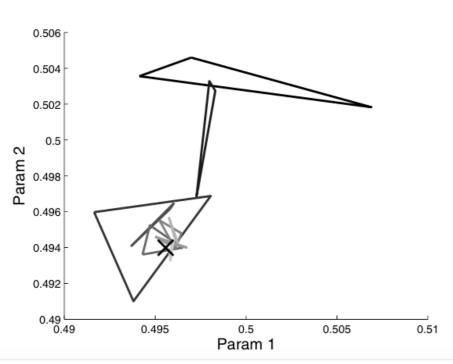
### BIOMETRIC SIGNATURE MASQUERADE

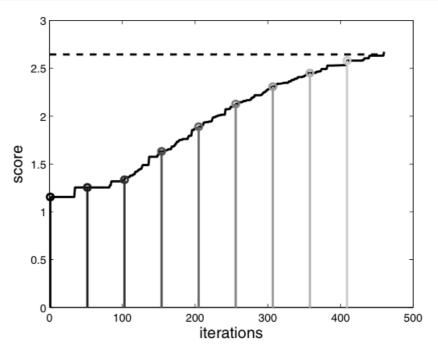
- Hill-Climbing attack
   based on the Uphill
   Simplex algorithm and
   its application to
   signature verification
  - Gomez-Barrero, M., Galbally, J., Fierrez, J., and Garcia, J.-O. at BioID 2011

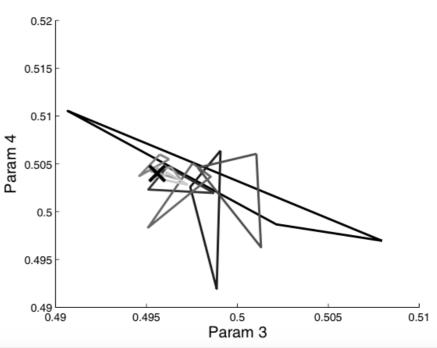
FMR o-effort	φ(#trials) o-effort	FMR' masq.	φ(#iters) masq.
0.05%	2 000	91.76%	1 556
0.01%	10 000	89.58%	1 678
0.0025 %	40 000	87.82%	1805

# SUBSPACE CONVERGENCE ILLUSTRATED









### X-TALK SIGNAL LEAKAGE

- Furthermore, there is a certain link in between online (sign-pad made) and offline (pen-and-paper made) signatures.
  - Btw., we also hope to exploit this link should it come to a trial.
  - On the other hand, the amount of information being cross-transferred in between these two signal forms is yet to be discovered!

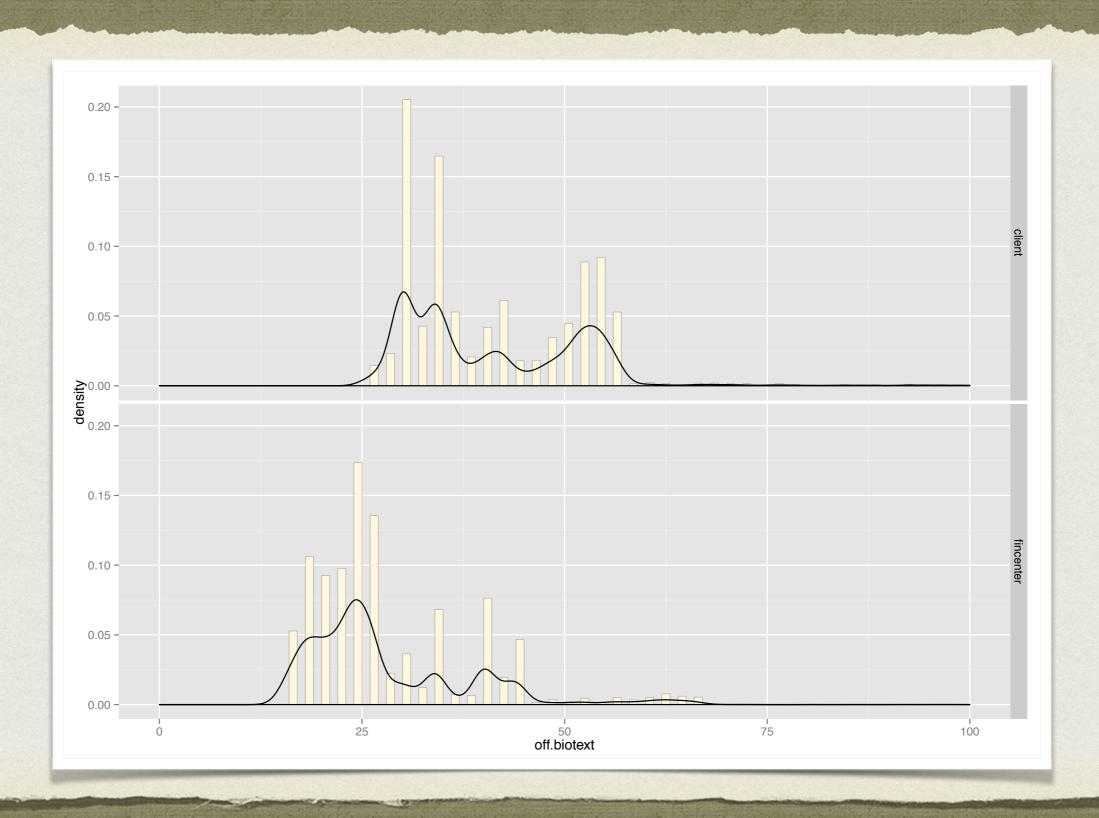
## PDF SIGNATURE LEAKAGE

- When signing a PDF using online signature data, we often put a human readable picture into the PDF annotation.
  - This is just to make the technology more userfriendly.
- This is, however, usually an offline plaintext projection of the (encrypted) online signature data.
  - How much information is leaking this way?

# OFFLINE PROJECTION EXAMPLE



### OFFLINE SIGNAL BRIEF -THERE IS SOMETHING!



### ISO/IEC 24745 REQUIREMENTS

- Renewability
  - allows multiple independent biometric references created ad hoc
  - a particular leaked template does not compromise the other ones (provably!)
- Revocability
  - user can revoke the ability of being successfully verified by a particular template from now on
- Biocryptography is an effective way on how to achieve these goals.



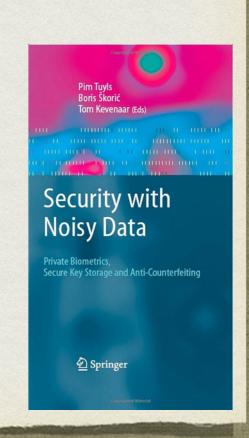
### CRYPTOGRAPHY EXACTNESS

Let 
$$y = AES_K(x)$$
 for a random  $K$ .  
Then  $AES_K^{-1}(y) = x$ , while  $AES_{K\oplus 1}^{-1}(y) \neq x$  (probability  $\approx 1$ ).

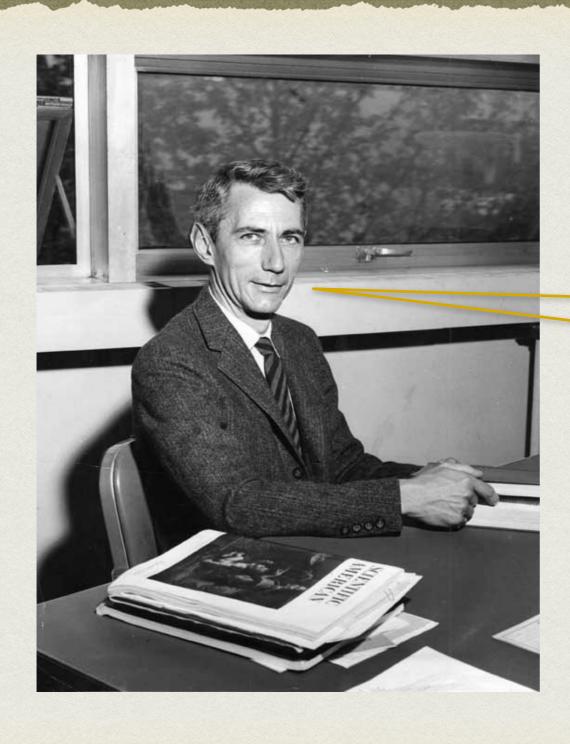
• The better the algorithm is the more randomized response we get for even one-bit error.

### BIOMETRICS FUZZINESS

- We seldom get the same data in the subsequent scans of the very same person.
  - Actually, this is usually a clear sign of a spoofed sample.
- To overcome this (intra-class) variability, we can employ the *biometric cryptography*.



### BACK TO THE ORIGIN



- 1. analyse the entropy gain from inter-class variation
- 2. use an errorcorrection code to cope with intra-class noise

Claude Elwood Shannon, 1948-49

## ERROR-CORRECTING CODE C

Let  $(F, \rho)$  be a metric space,  $\rho: F \times F \to [0, \infty)$ .

translation invariant metric:  $\rho(x,y) = \rho(0,x-y)$ 

Error correcting code is  $C \subset F$ ,  $C = \{c_1, c_2, ...\}$ .

 $decode: F \rightarrow C$ 

t-error correction capability:

Let  $\rho(c_i, y) \le t$ , then  $decode(c_i) = decode(y) = c_i$ .

We assume *decode*() always returns a (possibly wrong) codeword.

#### ENROLMENT

- i) randomly choose  $c_{key} \in C \subset F$
- ii) get BIO features vector  $w \in \mathbf{F}$
- iii) let  $\xi = w c_{key}$
- iv) let  $BIO\_key = hash(c_{key})$
- v) template =  $(\xi)$

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More involved entropy extractors can be used here...

#### VERIFICATION

- i) get BIO features vector  $w' \in F$
- ii) let  $y = w' \xi$
- iii) let  $c_{key}$ ' = decode(y)
- iv) let  $BIO\_key' = hash(c_{key'})$
- v) use BIO\_key' in the upper-layer protocol

#### VERIFICATION

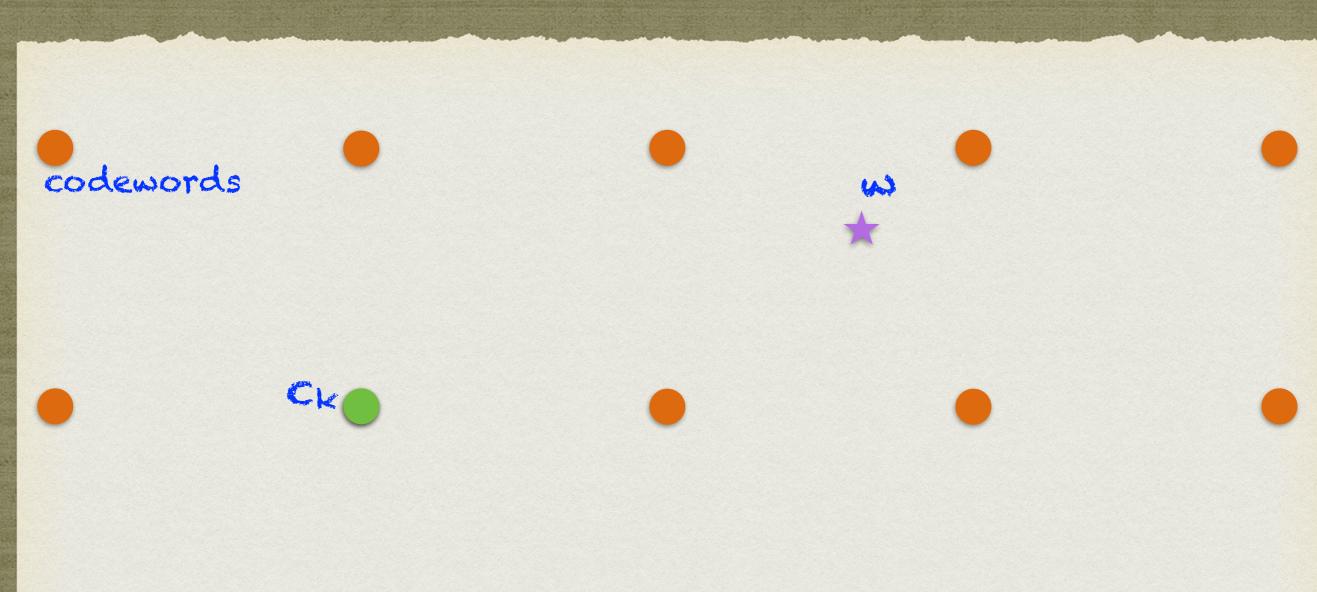
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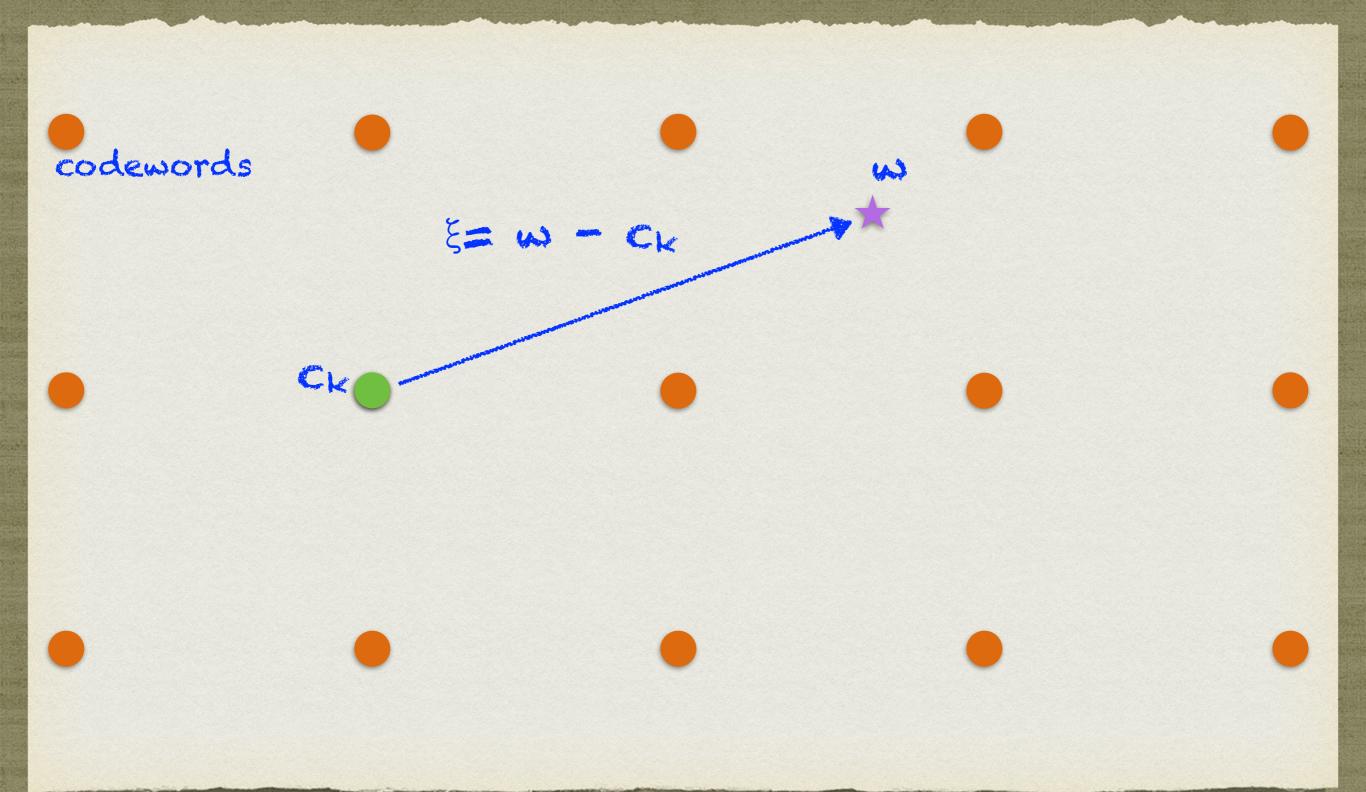
We have an ordinary crypto key, now...

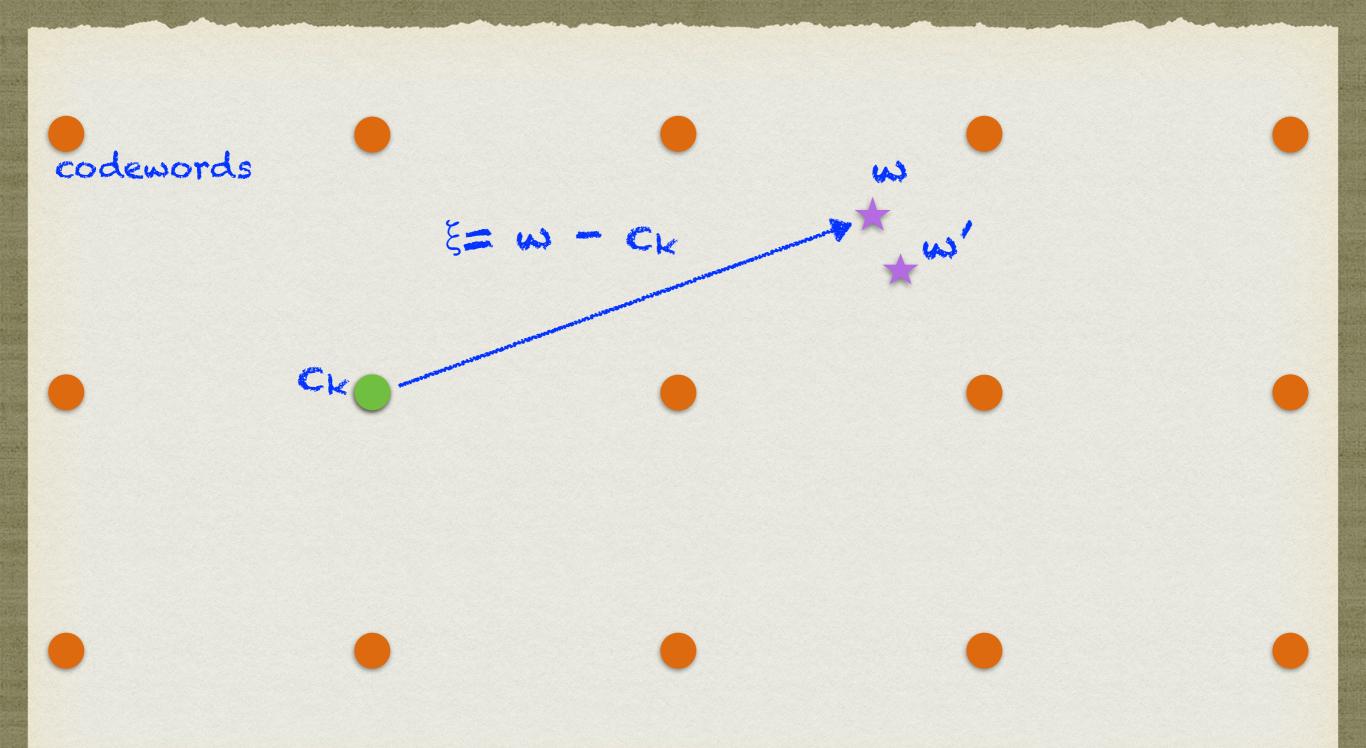
codewords

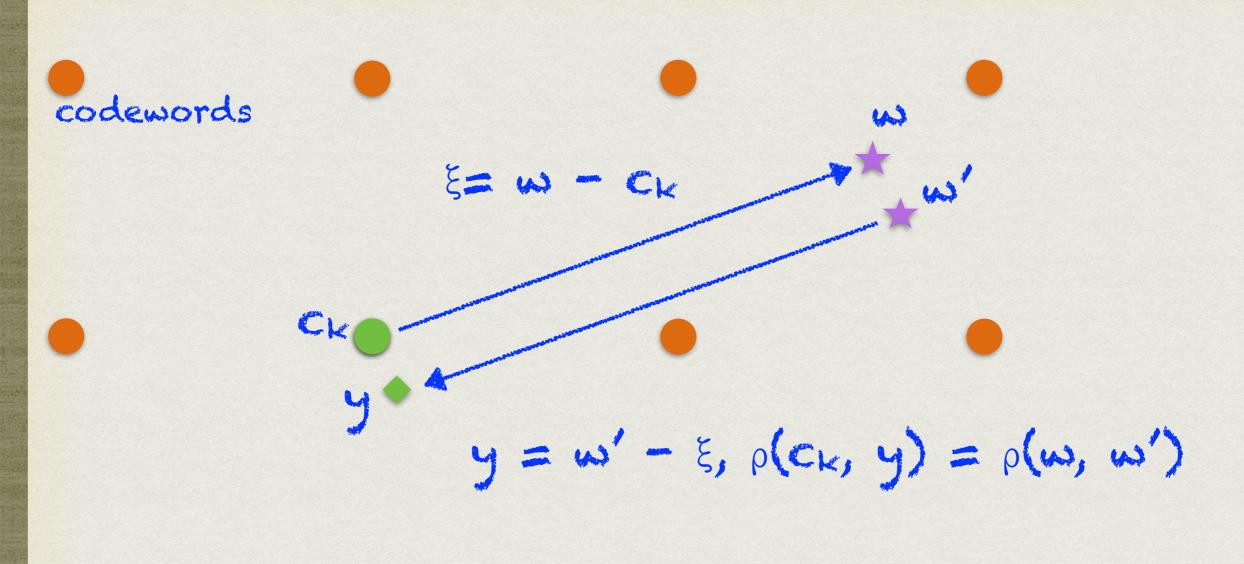
codewords

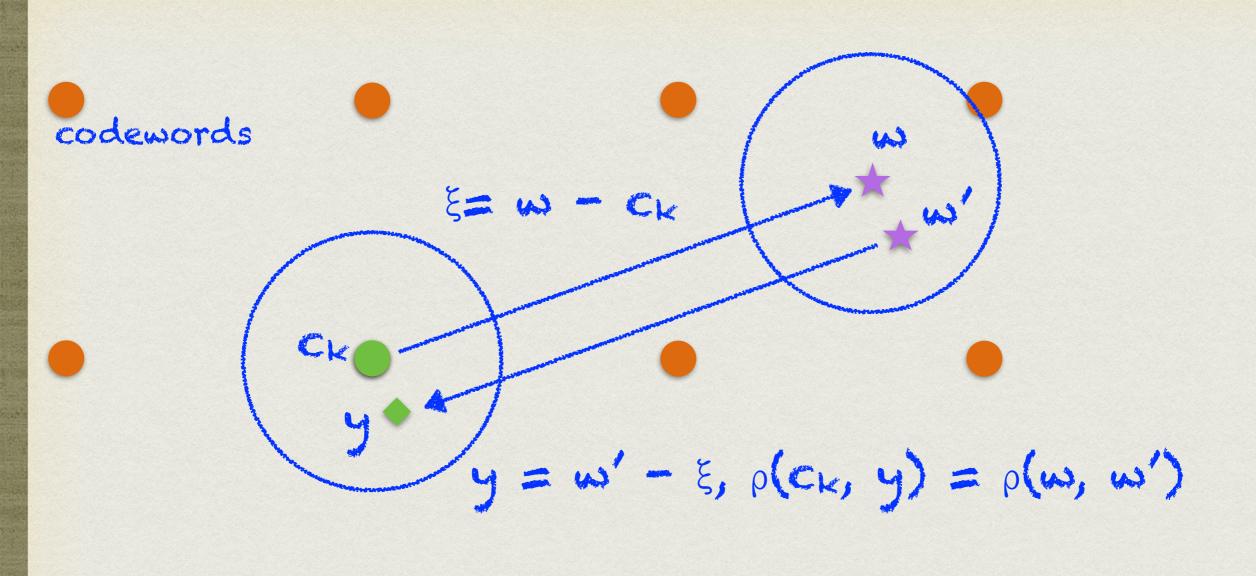
CK

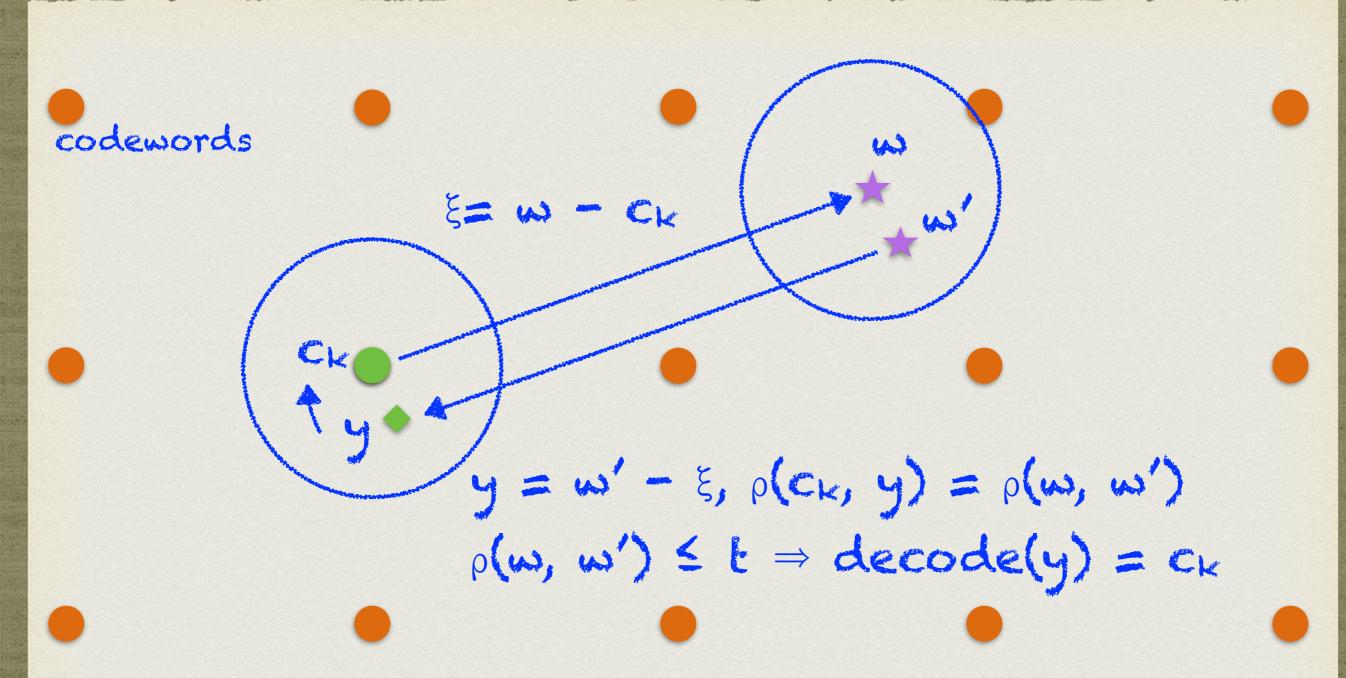












### IS IT ENOUGH?

- Template protection in contemporary systems is often quite questionable (to be polite).
- On the other hand, is it the only one problem?
  - No. We shall not push the concept of bio-keys too hard anyway.

#### BIO-SKIMMING

- Once biometric systems become ubiquitous, this will be a fruitful attack vector.
  - Attackers use a fake sensor (or hack into an original one) to skim the "bio-master-key".
  - At the end of the day, how many eyes, fingers, faces, vocal tracts (etc.) do we have?
  - It is like having few master-keys for a whole life.
  - Furthermore, we prove the master-key possession by simply handing it over to almost any device that asks so (again, again, ...and again).

## SPOOFING STILL MATTERS!

- That said, liveness detection will be always important!
  - Remember, biometrics is a signal detection.
  - It all works as long as we can assume the signal is coming from a particular human being!
    - Apparently, the biometric signal detector output shall be just one out of many inputs into an authentication process (itself being another multidimensional signal detection problem).

# TAMPER-RESISTANT SENSOR

- It signs the biometric signal samples with its private key to indicate it already has sampled that signal from a living individual.
  - Furthermore, the sample shall be then processed as soon as possible.
  - Otherwise, we have to mitigate the risk of a sensor compromise in the intermediate time by a further time-stamping: Long Term Validation of bio-samples.
  - This concept is all too often neglected in the emerging handwritten signature biometrics!



#### CONCLUSION

- We shall require ISO 19795 methodology during biometric application selection, comparison, and operational testing.
- Use an independent penetration test to verify:
  - zero-effort attack complexity
    - -beware of automated APIs!
  - masquerade attacks
  - spoofing possibilities
  - template security
  - system security in general
    - -threshold settings, template tampering