

# Homomorphic Signatures for Polynomial Functions

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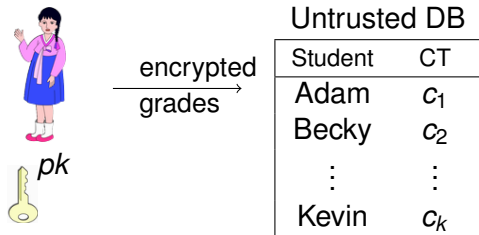


Untrusted DB



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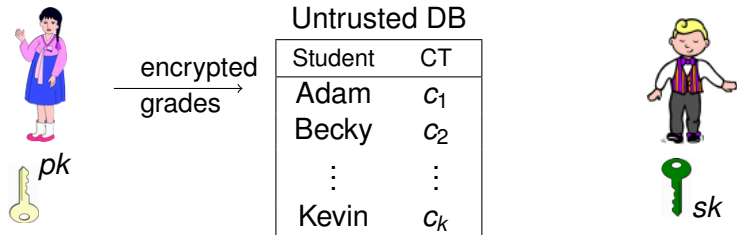
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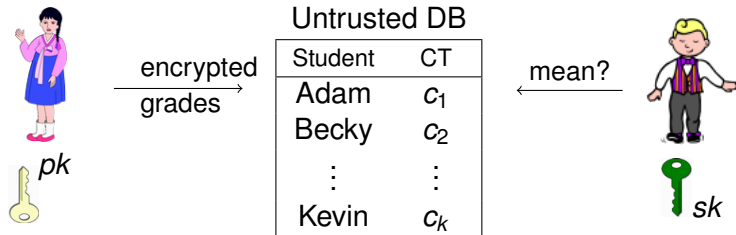
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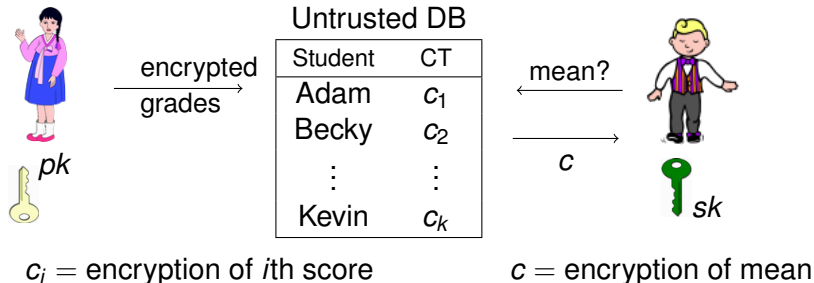
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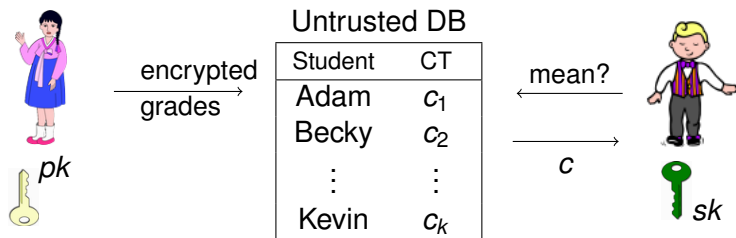
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$c_i$  = encryption of  $i$ th score

$c$  = encryption of mean

- Validity:  $c$  decrypts to the correct mean.
- Security: no adversary can obtain any info about scores.
- Length efficiency:  $c$  is short.
- Privacy: decrypted mean reveals nothing else about data.



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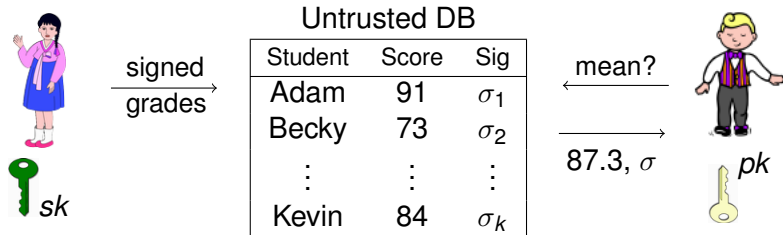
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- 4 **Privacy**:  $\sigma$  reveals nothing about data other than the mean.

# More generally: $\mathcal{F}$ -homomorphic signatures

- $\mathcal{F}$  is a set of “admissible” functions on messages.
- $\tau$  is a “tag” tying together data from the same set.  
(like a filename)
  - prevents mixing of data from different sets
- Given  $pk$ , admissible function  $f \in \mathcal{F}$ , and **signatures** on data

$$m_1, \dots, m_k,$$

anyone can compute a valid signature on

$$(\tau, f(m_1, \dots, m_k), \omega(f)),$$

where  $\omega(f)$  is an “encoding” or “digest” of the function  $f$ .

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Specifically, we construct secure, length-efficient,  $\mathcal{F}$ -homomorphic signatures for

$$\mathcal{F} = \{\text{polynomials of bounded degree with small coefficients}\}$$

## Application: Least Squares Fits

# Least squares fits — the basics

For a data set  $\{(x_i, y_i)\}_{i=1}^k$ , the **degree  $d$  least squares fit** is a polynomial

$$f(x) = c_0 + c_1x + \cdots + c_dx^d$$

that “best” approximates the  $y$  values.

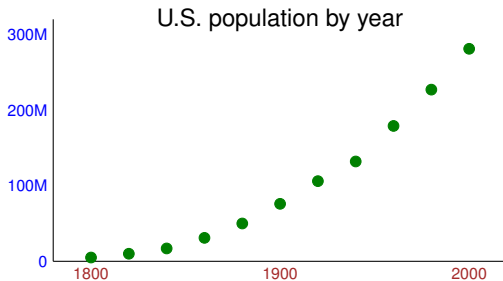


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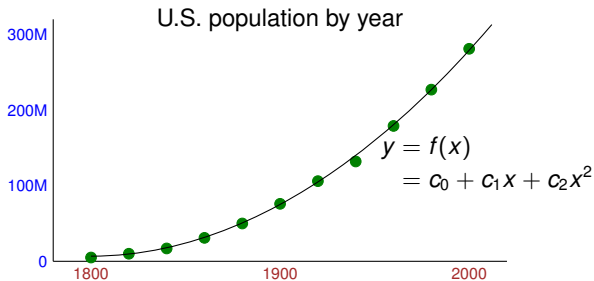


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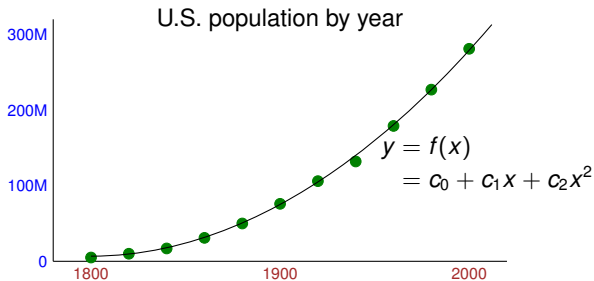


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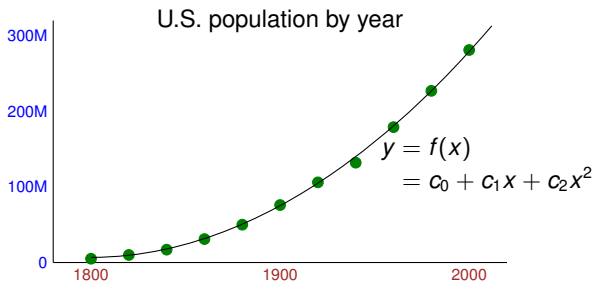


Formula:

$$\vec{c} = (X^tX)^{-1}X^t\vec{y}$$

$\vec{c}$  = vector of coefficients of  $f(x)$ ,  
 $X$  = Vandermonde matrix of  $x$  values,  
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# Authenticating a least-squares fit ( $x$ -values only)

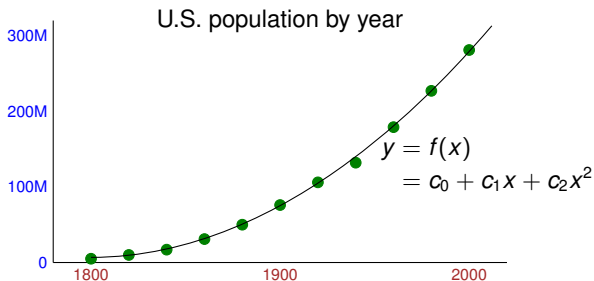


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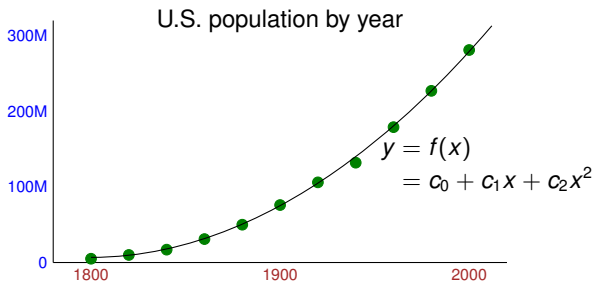
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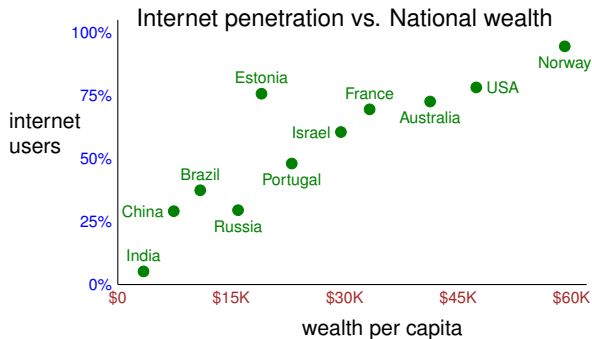
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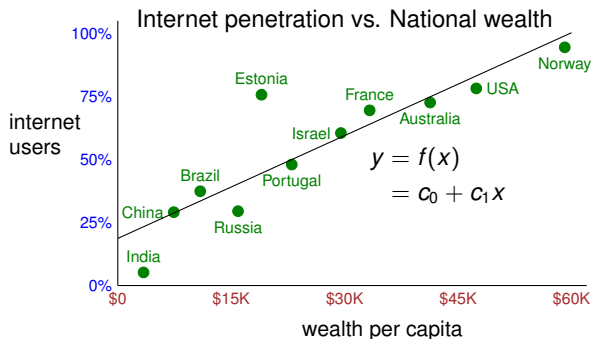
If  $x$  values are **fixed**, then  $\vec{c}$  is **linear** function of  $y$  values.

- Census bureau stores signed population counts on server using **linearly homomorphic** signature.
- Server can authenticate coefficients of least-squares fit.

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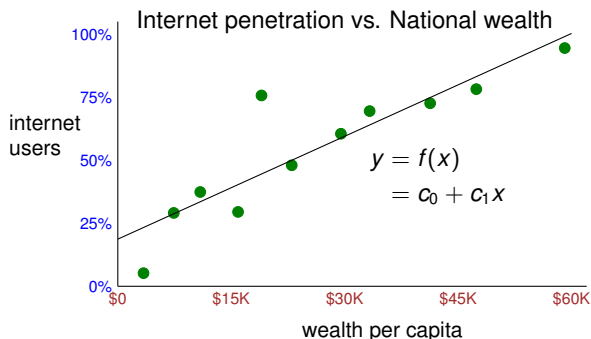
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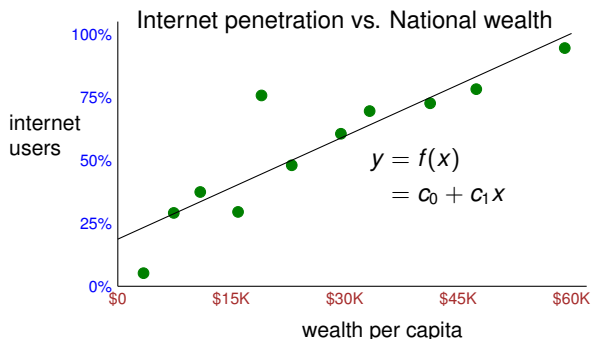
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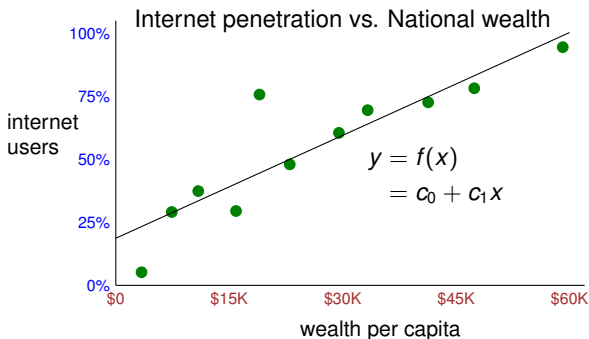
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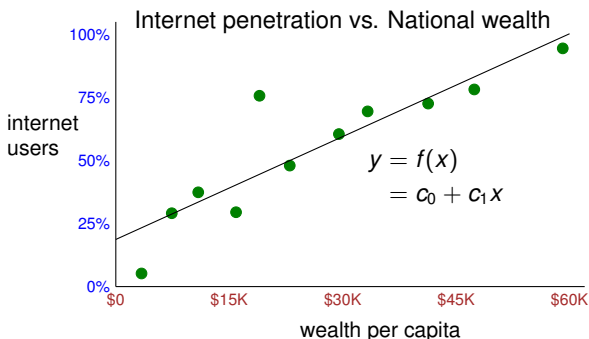
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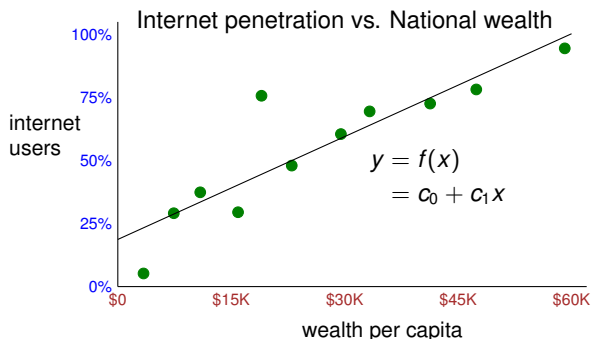
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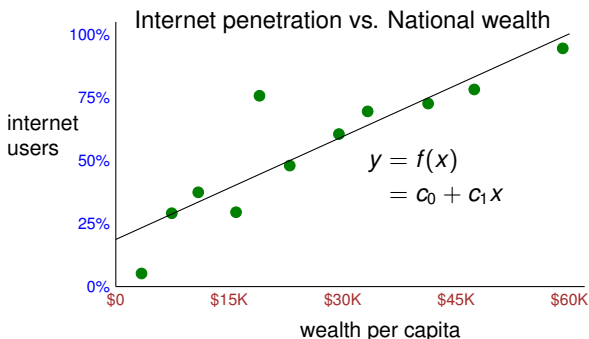
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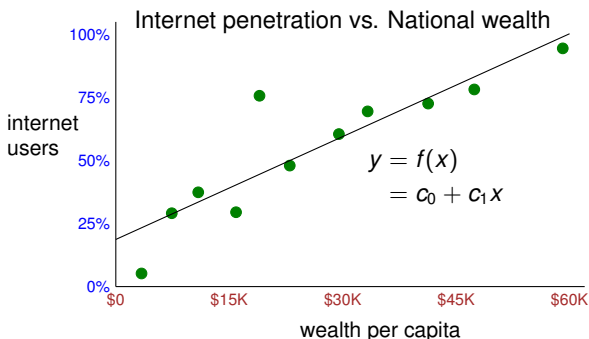
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- User can compute least-squares fit from server's values.
- Linear fit can be computed using degree 3 polynomials.

# Linearly Homomorphic Signatures



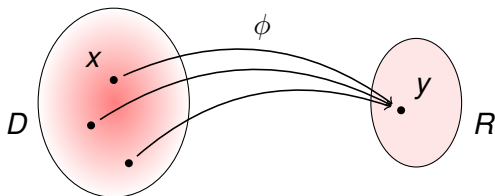
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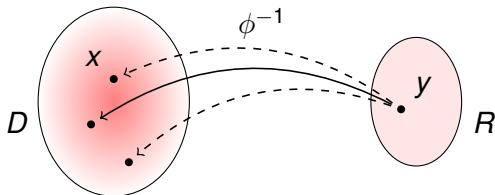
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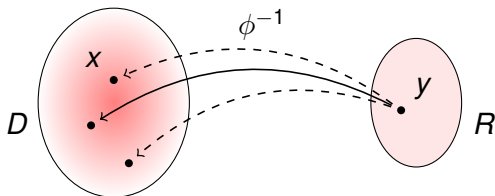
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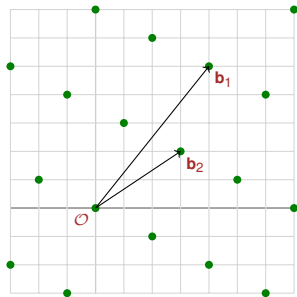
- “Hash and sign:”  $pk = \phi$ ,  $sk = \phi^{-1}$ , hash  $H: \{0, 1\}^* \rightarrow R$

$$\text{Sign}(m) := \phi^{-1}(H(m))$$

$$\text{Verify}(\sigma) : \phi(\sigma) \stackrel{?}{=} H(m)$$

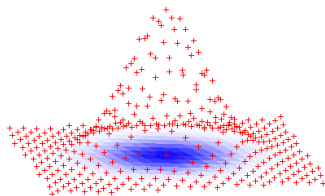
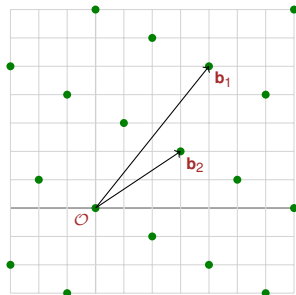
# GPV Signatures, concretely

- $\Lambda \subset \mathbb{Z}^n$  a lattice (full-rank additive subgroup), defined by basis.



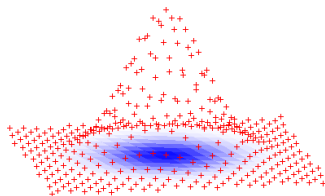
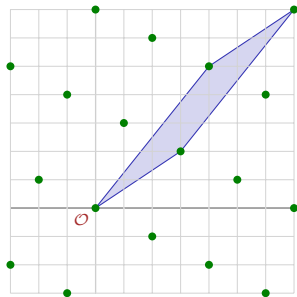
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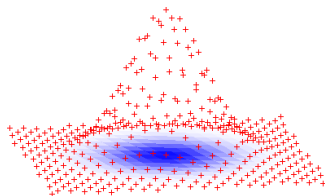
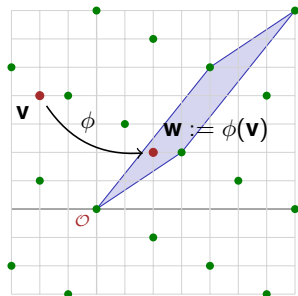
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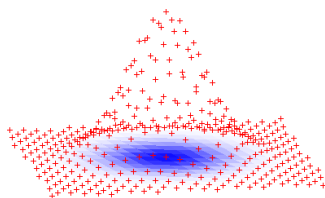
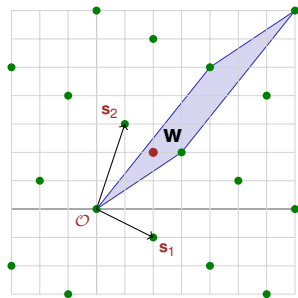
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- Trapdoor function  $\phi: \mathbf{v} \mapsto (\mathbf{v} \bmod \Lambda)$  i.e., move  $\mathbf{v}$  into a fundamental parallelepiped.





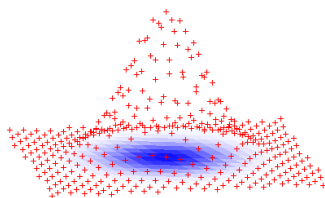
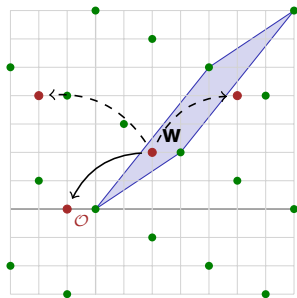
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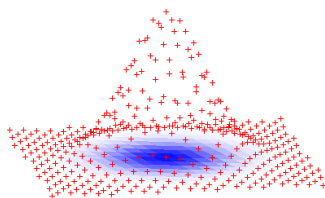
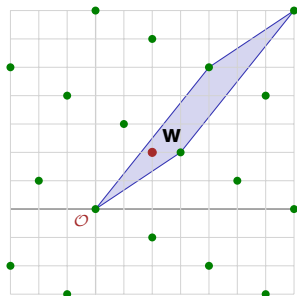
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- Sampling from  $\Lambda + \mathbf{w}$  **without** short basis is hard. (How hard depends on Gaussian parameter.)



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Idea: instead of hashing the messages to  $R = \mathbb{Z}^n / \Lambda$ ,  
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- For  $a, b \in \mathbb{Z}$ , define signature on  $am_1 + bm_2$  to be  
 $\sigma := a\sigma_1 + b\sigma_2$ .  
 $\Rightarrow \sigma$  is short (if  $a, b$  small),  $\sigma \bmod \Lambda = am_1 + bm_2$ .

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Valid signature doesn't imply function was computed correctly.

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signed  
grades →

Untrusted DB

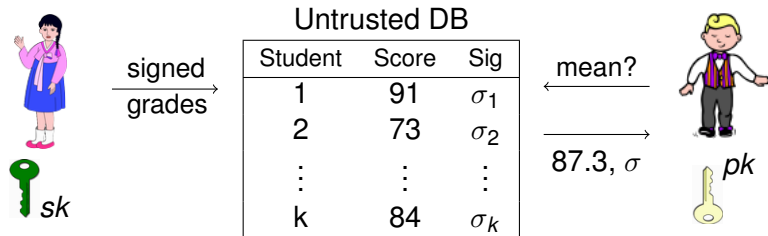
Student	Score	Sig
1	91	$\sigma_1$
2	73	$\sigma_2$
$\vdots$	$\vdots$	$\vdots$
k	84	$\sigma_k$

← mean?



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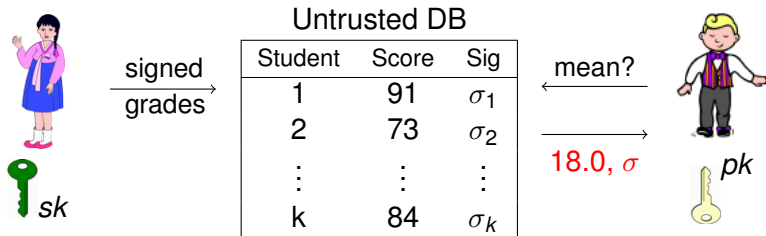
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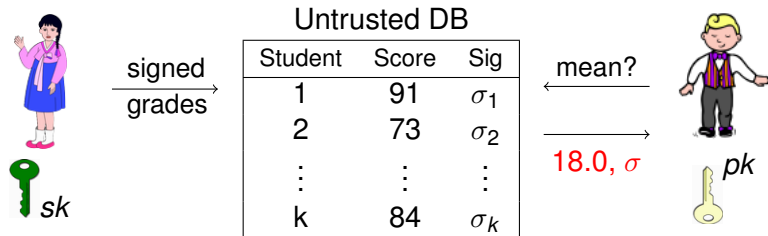
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- $\sigma$  authenticates 18, but 18 is not the mean!

# Linearly Homomorphic Signatures: Key Idea #2

Use a **second lattice** to authenticate **functions**:

- $\Lambda_2 \subset \mathbb{Z}^n$  distinct from  $\Lambda_1 := \Lambda$ .
  - require  $\Lambda_1 + \Lambda_2 = \mathbb{Z}^n$
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“Encode” functions  $f$  as elements  $\omega(f) \in \mathbb{Z}^n/\Lambda_2$ .

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If “encoding”  $\omega(\cdot)$  is linear, (i.e.,  $\omega(f) + \omega(g) = \omega(f + g)$ )  
then signature is a linear operator on the space of functions.

# How do we encode functions?

Ingredients:

- $k :=$  number of messages input to a function.
- $\tau :=$  “tag” that ties together messages in same data set.
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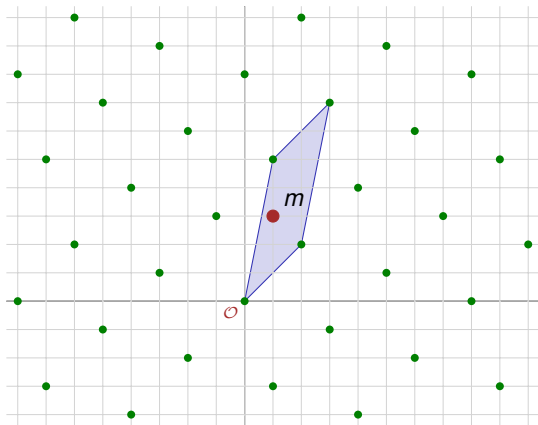
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- $c_i$  are small integers.
- “encoding”  $\omega(f)$  much shorter than description of  $f$ .

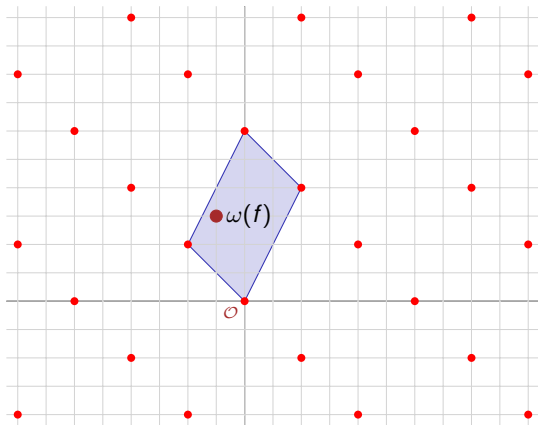
# “Intersection method” binds messages to functions

- Messages  $m \in \mathbb{Z}^n / \Lambda_1$ .



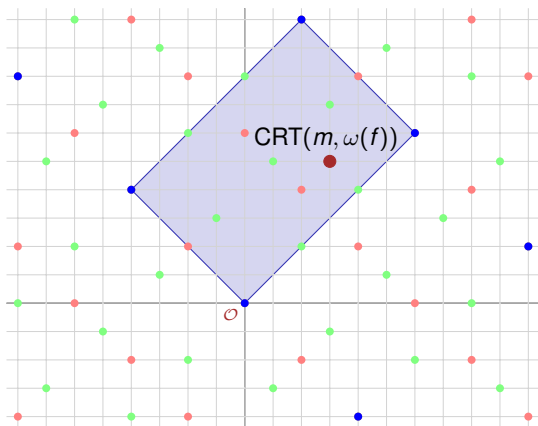
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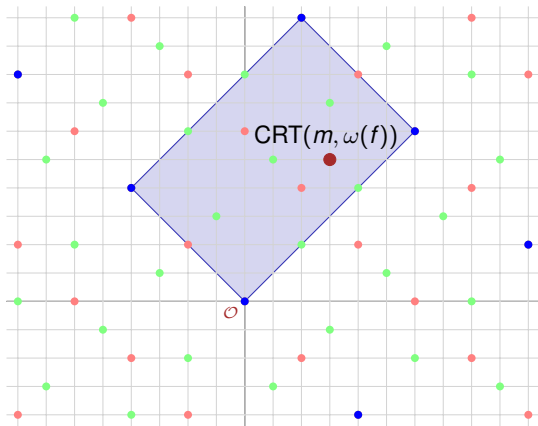


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# Signing a message-function pair

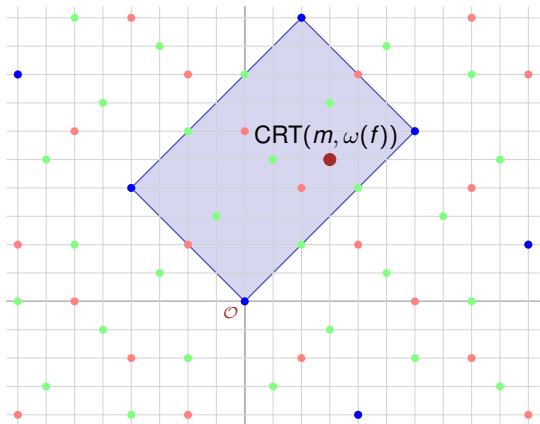
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$\text{Sign}(m) :=$  short vector in  $(\Lambda_1 \cap \Lambda_2) + \text{CRT}(m, \omega(f))$

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and  $\sigma$  is short

# Linearly homomorphic signature scheme

- **KeyGen**( $n$ ):
  - $pk$  = Lattices  $\Lambda_1, \Lambda_2 \subset \mathbb{Z}^n$ , Gaussian parameter  $\beta$
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  - 1  $\sigma \bmod \Lambda_1 = m$ ,
  - 2  $\sigma \bmod \Lambda_2 = \omega(f) = \sum c_i \alpha_i$ ,
  - 3  $\sigma$  sufficiently short.

# Concrete example

- Lattices:

 $\Lambda_1$  $\Lambda_2$  $\Lambda_1 \cap \Lambda_2$

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$$\Lambda_1 = p\mathbb{Z}^n$$

$p$  small prime

$$\Lambda_2$$

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Same functionality as *network coding* signatures [BFKW09,GKKR10], except  $p$  can be small (even  $p = 2$ ).

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Chall.



Adversary





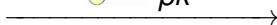
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$pk$

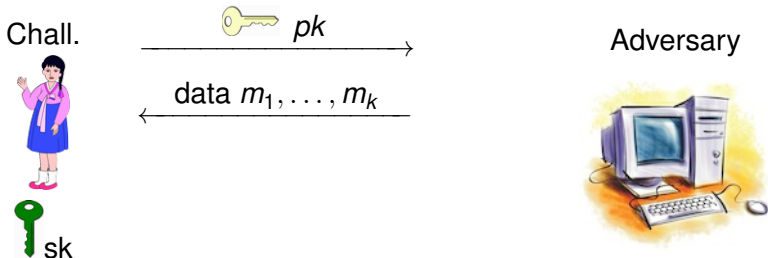


Adversary



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- Forgery is a valid signature on  $(\tau, m^*, f)$  with  $m^* \neq f$  (messages with tag  $\tau$ ).



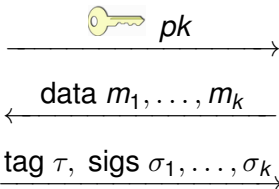
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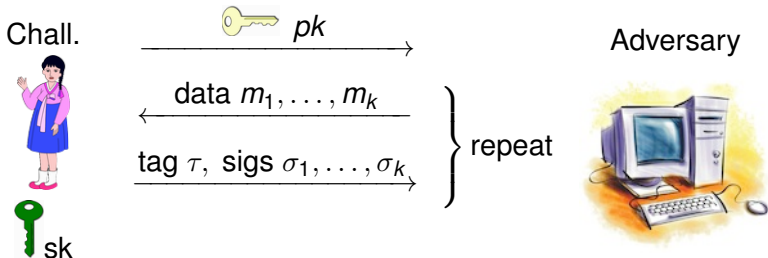


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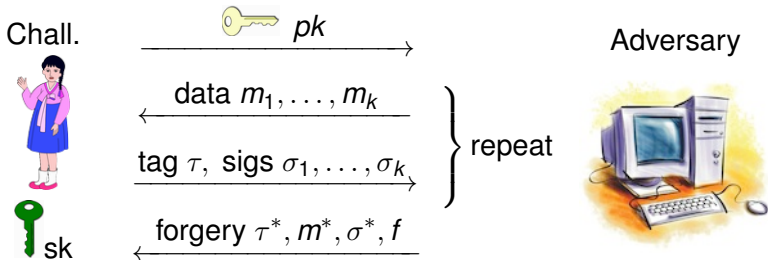
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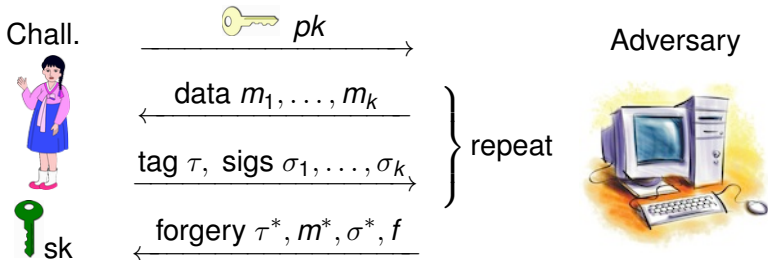
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Adversary wins if  $f$  admissible,  $\sigma^*$  verifies for  $(\tau^*, m^*, f)$ , and

- 1  $\tau^*$  not obtained in response to a query, **or**
- 2  $\tau^* = \tau$  for query  $(m_1, \dots, m_k)$ , and  $m^* \neq f(m_1, \dots, m_k)$ .

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## Proof outline

- 1 Given a “challenge”  $\Lambda_2$ , answer signature queries **without** a basis of  $\Lambda_1 \cap \Lambda_2$ .
- 2 Use adversary’s forgery to produce a short vector in  $\Lambda_2$ .

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  - 1 Use basis of  $\Lambda_1$  to compute short vectors  $\sigma_i \in \Lambda_1 + m_i$ ;
  - 2 Set  $\alpha_i := \sigma_i \bmod \Lambda_2 \in \mathbb{Z}^n / \Lambda_2$ .
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- Simulation is indistinguishable from real system.

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**Conclusion:**  $\sigma^* - \sigma$  is (1) nonzero, (2) in  $\Lambda_2$ , (3) short.

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**Conclusion:**  $\sigma^* - \sigma$  is (1) nonzero, (2) in  $\Lambda_2$ , (3) short.

- If  $\tau^*$  not obtained from a query, sign random messages  $m_i$  and perform same analysis.

Privacy property: derived signature on  $f(m_1, \dots, m_k)$  reveals nothing about  $m_1, \dots, m_k$  beyond value of  $f$ .



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Specifically: given data sets

$$\vec{m} = (m_1, \dots, m_k), \quad \vec{m}' = (m'_1, \dots, m'_k)$$

and admissible function  $f$  with

$$f(\vec{m}) = f(\vec{m}'),$$

even **unbounded** adversary cannot distinguish derived signature on  $f(\vec{m})$  from derived signature on  $f(\vec{m}')$ .

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- Distribution of derived signature on  $f(\vec{m})$  depends only on  $f$  and  $f(\vec{m})$ , **not on  $\vec{m}$** .
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Key technical fact [BF11]: distribution of linear combination of discrete Gaussian samples is **also discrete Gaussian**.

- Sigs on  $m_i$  sampled from discrete Gaussian distribution, derived sigs are linear combinations.

# Polynomially Homomorphic Signatures from Ideal Lattices

# Extending the system

Linearly homomorphic scheme: messages in  $\mathbb{Z}^n/\Lambda_1$ , functions “encoded” in  $\mathbb{Z}^n/\Lambda_2$ , signatures are short vectors in  $\mathbb{Z}^n$ .

$\phi_i: \mathbb{Z}^n \rightarrow \mathbb{Z}^n/\Lambda_i$  given by  $\mathbf{v} \mapsto (\mathbf{v} \bmod \Lambda_i)$  is a **linear map**, so we can add either before or after applying  $\phi_i$ .

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New idea: what if  $\mathbb{Z}^n$  has a **ring** structure and  $\Lambda_1, \Lambda_2$  are **ideals**?

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Then  $\phi$  is a **ring homomorphism**, so we can add **or multiply** either before or after applying  $\phi$ .

- Can authenticate polynomial functions on messages.



# Setup for polynomial system [G09]

Fix monic, irreducible  $F(x) \in \mathbb{Z}[x]$  of degree  $n$ .

- $R := \mathbb{Z}[x]/(F(x))$  gives a ring structure on  $\mathbb{Z}^n$ :

(coordinates of vectors)  $\leftrightarrow$  (coefficients of polynomials mod  $F$ )

$$(a_0, \dots, a_{n-1}) \leftrightarrow a_0 + a_1x + \dots + a_{n-1}x^{n-1}$$

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$\Lambda_2 =$  prime ideal  $\mathfrak{q}$ ; polynomials “encoded” in  $R/\mathfrak{q} = \mathbb{F}_q$ :

- Hash function  $H: \{0, 1\}^* \rightarrow \mathbb{F}_q^k$  maps  $\tau \mapsto (\alpha_1, \dots, \alpha_k)$ .
- “Encode”  $f$  by  $\omega(f) := f(\alpha_1, \dots, \alpha_k)$ .  
(think of coefficients of  $f$  as small integers).

# The polynomial system, concretely

- **KeyGen( $n$ ):**
  - $F(x) \in \mathbb{Z}[x]$  degree  $n$   
 $\Rightarrow$  ring structure on  $\mathbb{Z}^n \cong R := \mathbb{Z}[x]/(F(x))$ .
  - $pk =$  prime ideals  $\mathfrak{p}, \mathfrak{q} \subset R$ , Gaussian parameter  $\beta$
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# Products of short elements

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[G09,G10]: parameter  $\gamma_F$  measures how much multiplication in  $R$  increases length:

$$\gamma_F := \sup_{u,v \in R} \frac{\|u \cdot v\|}{\|u\| \cdot \|v\|}.$$

- Product of  $d$  elements of length  $< \beta$  has length  $< \gamma_F^{d-1} \beta^d$ .
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Lots of  $F(x)$  have small  $\gamma_F$ :

- e.g., cyclotomic polynomials  $\Phi_\ell(x)$ ,  $\ell$  prime or  $\ell = 2^a 3^b$ .

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- Repeat to get a second prime ideal  $\mathfrak{q} = v \cdot R$ .
- $uv \cdot R = \mathfrak{p} \cdot \mathfrak{q}$ , and

$$\mathbf{B} := \{uv, uv \cdot x, uv \cdot x^2, \dots, uv \cdot x^{n-1}\}.$$

spans  $\mathfrak{p} \cdot \mathfrak{q}$  and consists of short elements:

$$\|uv \cdot x^i\| \leq \|u\| \cdot \|v\| \cdot \gamma_F^2.$$

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Operation	Length expansion
Evaluate degree- $d$ monomial	$l \mapsto l^d \cdot \gamma_F^{d-1}$
Multiply by coefficient in $[-y, y]$	$l \mapsto l \cdot y$
Sum of $m$ monomials of length $l$	$l \mapsto l \cdot m$

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Sum of $m$ monomials of length $\ell$	$\ell \mapsto \ell \cdot m$

- Signatures on original messages  $m_i$  have length  $< \beta$   
 $\Rightarrow$  signature on  $f(m_1, \dots, m_k)$  has length  $< \beta^d \cdot \gamma_F^{d-1} \cdot y \cdot \binom{k+d}{d}$ .

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Define admissible function set  $\mathcal{F}$  to be polynomials in  $\mathbb{F}_p[x_1, \dots, x_k]$  of degree  $\leq d$  with coefficients in  $[-y, y]$ .

Operation	Length expansion
Evaluate degree- $d$ monomial	$\ell \mapsto \ell^d \cdot \gamma_F^{d-1}$
Multiply by coefficient in $[-y, y]$	$\ell \mapsto \ell \cdot y$
Sum of $m$ monomials of length $\ell$	$\ell \mapsto \ell \cdot m$

- Signatures on original messages  $m_i$  have length  $< \beta$   
 $\Rightarrow$  signature on  $f(m_1, \dots, m_k)$  has length  $< \beta^d \cdot \gamma_F^{d-1} \cdot y \cdot \binom{k+d}{d}$ .
- If  $\beta, \gamma_F, k, y \in \text{poly}(n)$  and  $d = O(1)$ , then derived signature length is  $\text{poly}(n)$ . ( $p$  is exponential in  $n$ )

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- For fixed  $n$ , bit length of derived signatures is linear in  $d$ , logarithmic in  $k$ .

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  - Producing a short generator of arbitrary principal  $\mathfrak{q}$  is a classical problem in algorithmic number theory.
- Distribution of Smart-Vercauteren  $\mathfrak{q}$  not well understood.
  - Want  $\mathfrak{q}$  in distribution that admits a worst-case reduction.

# Open questions

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## Thank you!

Thanks also to Chris Peikert for help with graphics.