

# Efficient Private Matching and Set Intersection

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# A Story...

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Is there any chance we might be compatible?

We could see if we have similar interests?

Have you heard of “secure function evaluation” ?

Maybe...

I don't really like to give personal information

I don't want to waste my entire night...

# Making SFE more efficient...

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1. Improvements to generic primitives (SFE, OT)

2. Improvements in specific protocol examples

We could see if we have similar interests?

Have you heard of “secure function evaluation” ?

I don't want to waste my entire night...

# Secure Function Evaluation

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X



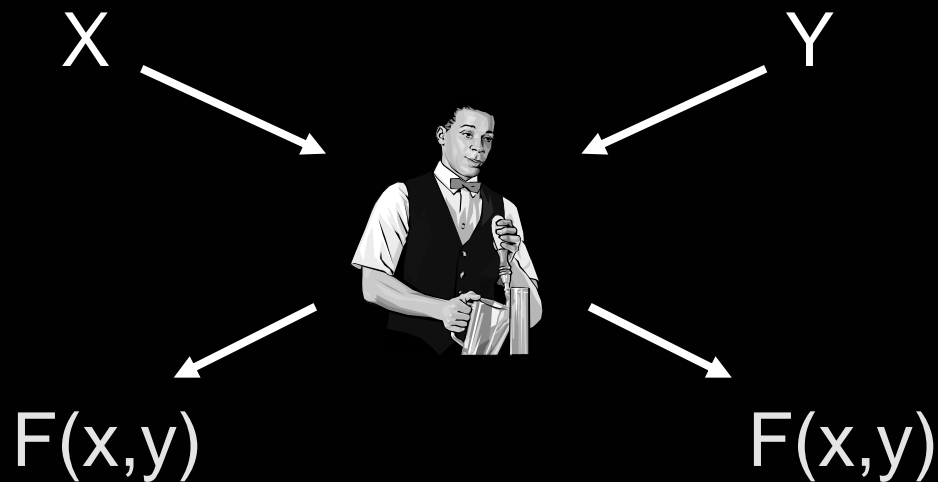
Y

Input:

Output:

$F(x,y)$  and nothing else

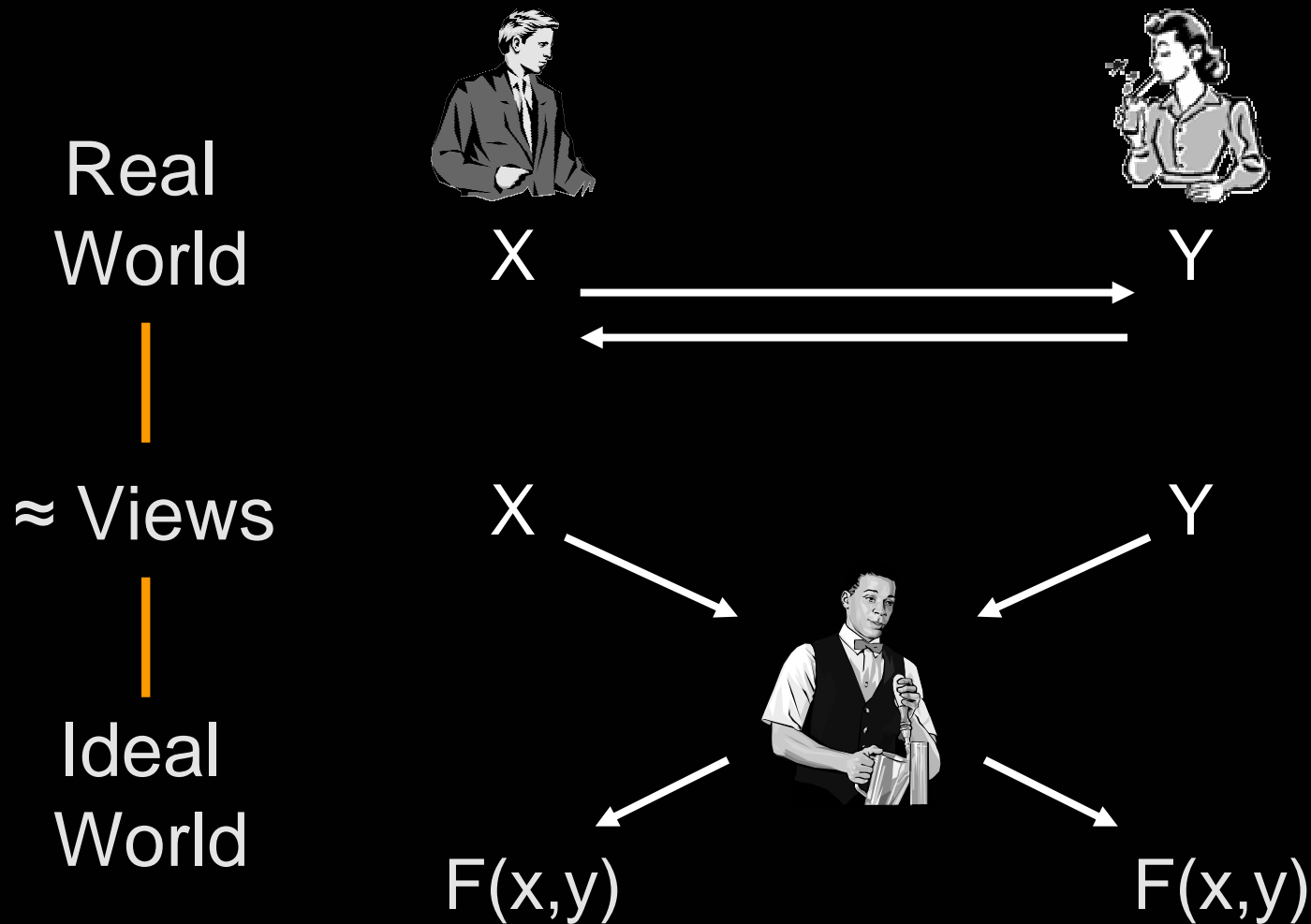
As if...



What if such trustworthy barkeepers don't exist?

# Proving SFE Protocols...

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Can consider semi-honest and malicious models

# Our Specific Scenario

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Client



Server

Input:  $X = x_1 \dots x_k$

$Y = y_1 \dots y_k$

Output:  $X \cap Y$  only

nothing

- Shared interests (research, music)
- Dating, genetic compatibility, etc.
- Credit card rating
- Terrorist watch list (CAPS II)

# Related work

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- Generic constructions [Yao,GMW,BGW,CCD]
  - Represent function as a circuit with combinatorial gates
  - Concern is size of circuit (as communication is  $O(|C|)$ )
  - Simplest uses  $k^2$  comparisons
- Diffie-Hellman based solutions [FHH99, EGS03]
  - Insecure against malicious adversaries
  - Considered in the “random oracle” model
- Our work:  $O(k \ln \ln k)$  overhead.
  - “Semi-honest” adversaries – in standard model
  - “Malicious” adversaries – in RO model

# This talk...

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- Overview
- Basic protocol in semi-honest model
- Efficient Improvements
- Extending protocol to malicious model
- Other results...



# Basic tool: Homomorphic Encryption

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- Semantically-secure public-key encryption
- Given  $\text{Enc}(M1)$ ,  $\text{Enc}(M2)$  can compute, without knowing decryption key,
  - $\text{Enc}(M1+M2) = \text{Enc}(M1) \cdot \text{Enc}(M2)$
  - $\text{Enc}(c \cdot M1) = [\text{Enc}(M1)]^c$ , for any constant  $c$
- Examples: El Gamal variant, Paillier, DJ

# The Protocol

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- Client (C) defines a polynomial of degree  $k$  whose roots are his inputs  $x_1, \dots, x_k$

$$P(y) = (x_1 - y)(x_2 - y) \dots (x_k - y) = a_0 + a_1 y + \dots + a_k y^k$$

- C sends to server (S) homomorphic encryptions of polynomial's coefficients

$$\text{Enc}(a_0), \dots, \text{Enc}(a_k)$$

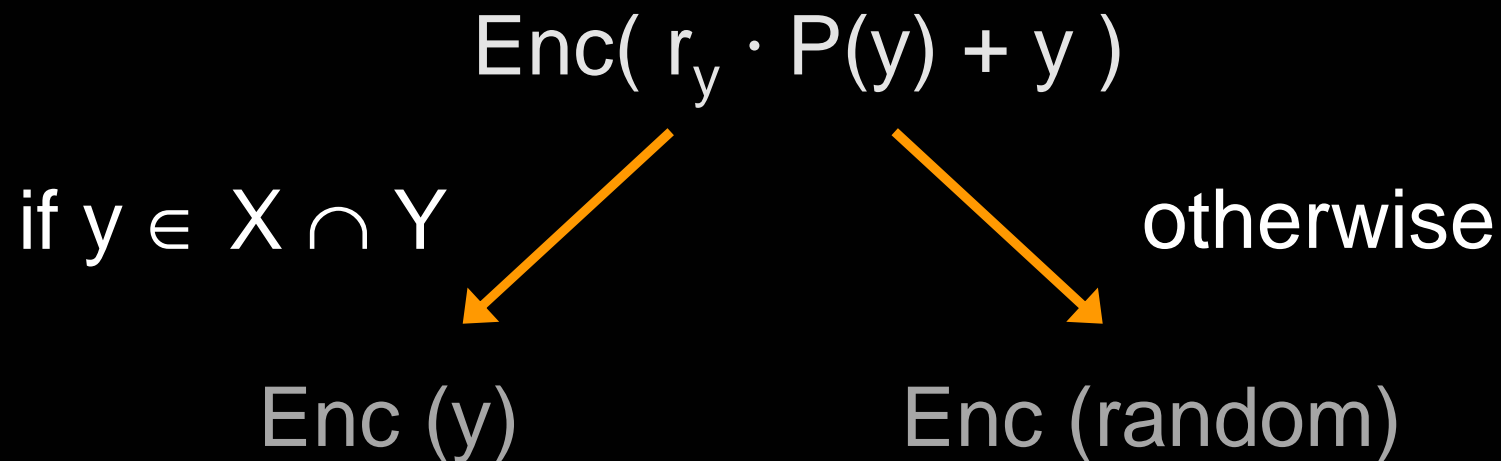
$$\text{Enc}( P(y) ) = \text{Enc}( a_0 + a_1 \cdot y^1 + \dots + a_k \cdot y^k )$$

$$\text{Enc}(a_0) \cdot \text{Enc}(a_1)^{y^1} \cdot \dots \cdot \text{Enc}(a_k)^{y^k}$$

# The Protocol

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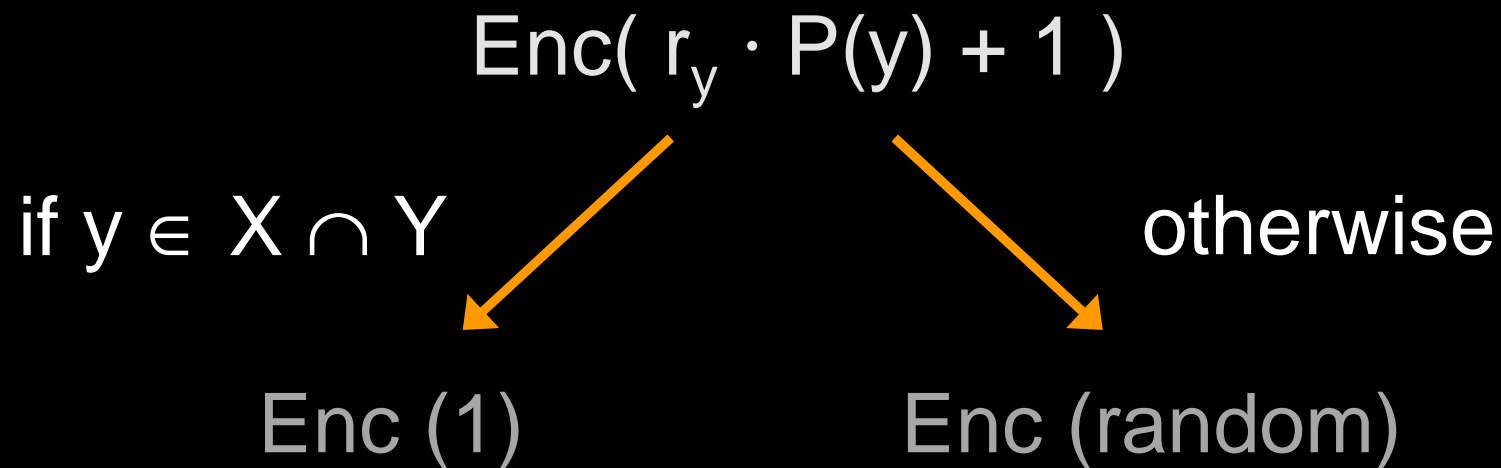
- S uses homomorphic properties to compute,  
 $\forall y, r_y \leftarrow \text{random}$



- S sends (permuted) results back to C
- C decrypts results, identifies  $y$ 's

# Variant protocols...cardinality

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- Computes size of intersection: # Enc (1)
- Others... Output 1 iff  $|X \cap Y| > t$

# Security (semi-honest case)

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- Client's privacy
  - S only sees semantically-secure enc's
  - Learning about C's input = breaking enc's
- Server's privacy (proof via simulation)
  - Client gets  $X \cap Y$  in ideal (TTP) model
  - Given that, can compute  $E(y)$ 's and  $E(\text{rand})$ 's and thus simulate real model

# Efficiency

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- Communication is  $O(k)$ 
  - ✓ C sends  $k$  coefficients
  - ✓ S sends  $k$  evaluations on polynomial
- Computation
  - ✓ Client encrypts and decrypts  $k$  values
  - ✗ Server:
    - $\forall y \in Y$ , computes  $\text{Enc}(r_y \cdot P(y) + y)$ , using  $k$  exponentiations
    - Total  $O(k^2)$  exponentiations

# Improving Efficiency (1)

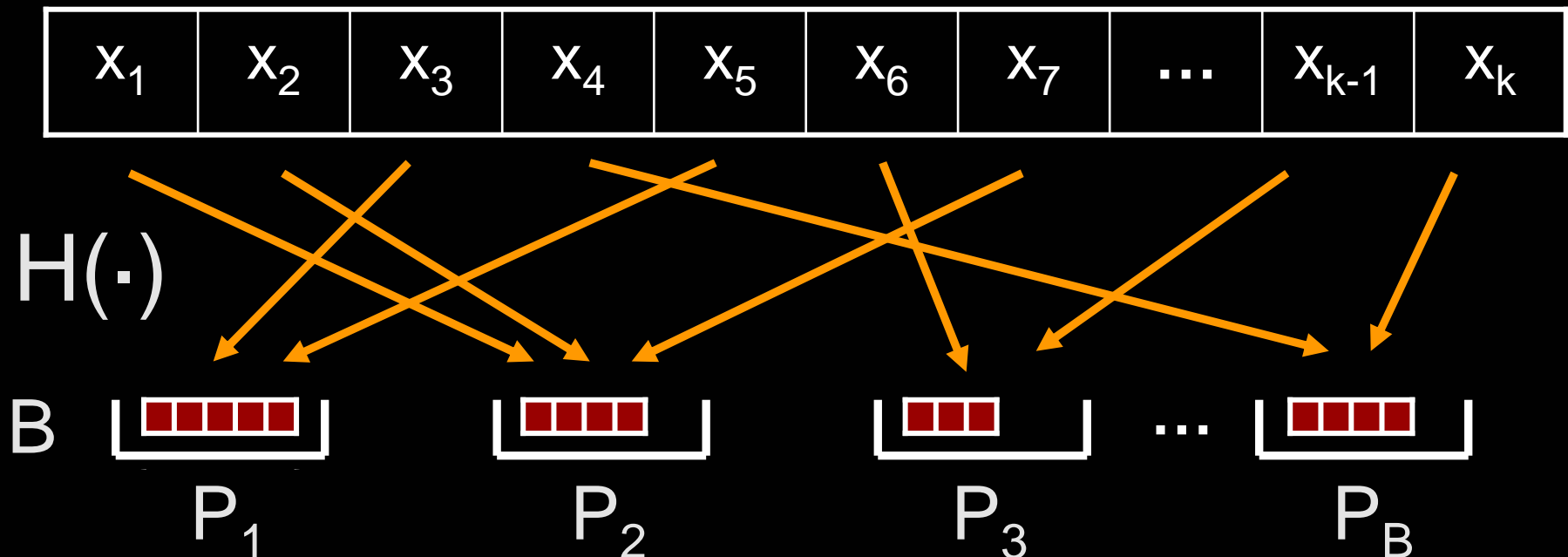
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- Inputs typically from a “small” domain of  $D$  values. Represented by  $\log D$  bits (...20)
- Use Horner’s rule

$$P(y) = a_0 + y (a_1 + \dots y (a_{n-1} + ya_n) \dots)$$

- That is, exponents are only  $\log D$  bits
  - Overhead of exponentiation is linear in  $| \text{exponent} |$
- “Improve” by factor of  $| \text{modulus} | / \log D$   
e.g.,  $1024 / 20 \approx 50$

# Improving Efficiency (2): Hashing

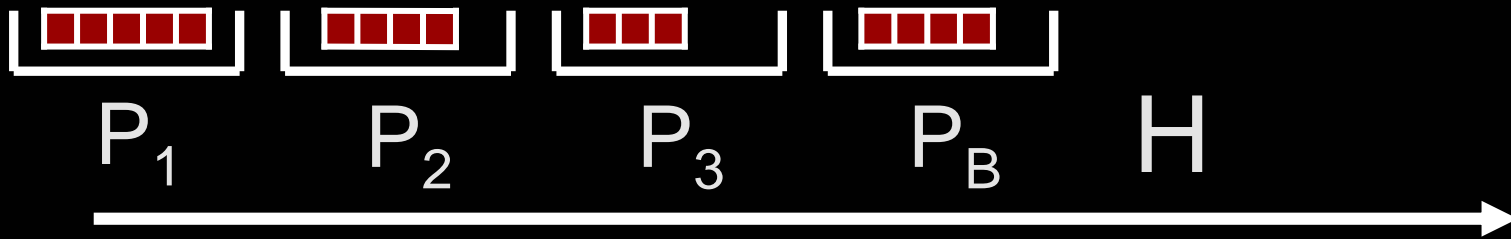


- C uses PRF  $H(\cdot)$  to hash inputs to  $B$  bins
- Let  $M$  bound max # of items in a bin
- Client defines  $B$  polynomials of deg  $M$ . Each poly encodes  $x$ 's mapped to its bin + enough "other" roots



# Improving Efficiency (2): Hashing

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- Client sends  $B$  polynomials and  $H$  to server.
- For every  $y$ ,  $S$  computes  $H(y)$  and evaluates the single corresponding poly of degree  $M$

# Overhead with Hashing

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- Communication:  $B \cdot M$  (# bins · # items per)
- Server:  $k \cdot M$  short exp's,  $k$  full exp's  
(  $P_i(y)$  ) (  $r_y \cdot P_i(y) + y$  )
- How to make  $M$  small as possible?
- Choose most balanced hash function

# Balanced allocations [ABKU]

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- H: Choose two bins, map to emptier bin

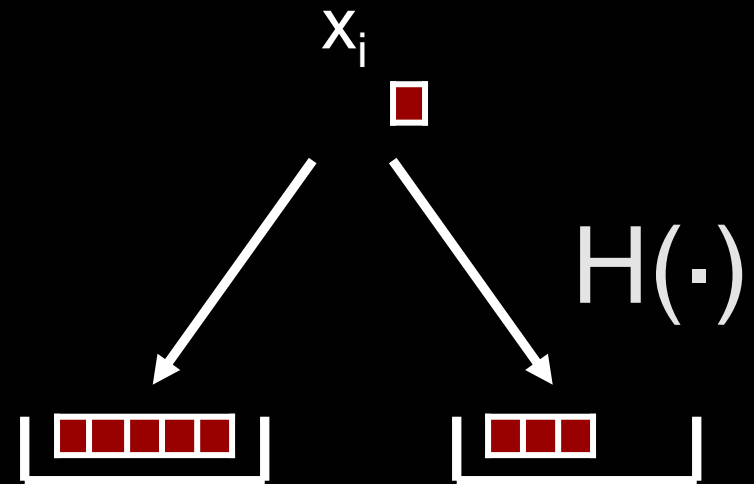
- $B = k / \ln \ln k$

$$\rightarrow M = O(\ln \ln k)$$

$$M \leq 5 \text{ [BM]}$$

- Communication:  $O(k)$

- Server:  $k \ln \ln k$  short exp,  $k$  full exp in practice



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# Malicious Adversaries

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- Malicious clients
  - Without hashing is trivial: Ensure  $a_0 \neq 0$
  - With hashing
    - Verify that total # of roots (in all B poly's) is k
    - Solution using cut-and-choose
    - Exponentially small error probability
  - Still standard model
- Malicious servers
  - Privacy...easy:  
S receives semantically-secure encryptions

# Security against Malicious Server

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- Correctness: Ensure that there is an input of  $k$  items corresponding to  $S$ 's actions
- Problem: Server can compute  $r_y \cdot P(y) + y'$
- Solution: Server uses RO to commit to seed, then uses resulting randomness to “prove” correctness of encryption

# Security against Malicious Server

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$$\forall y, s \leftarrow \text{rand}, r \leftarrow H_1(s)$$

$$[e, h] \leftarrow [ \text{Enc} ( r_1 \cdot P(y) + s ), H_2 ( r_2, y ) ]$$

Deterministic

$$s^* \leftarrow \text{Dec} ( e ), r^* \leftarrow H_1(s^*)$$

?  $\exists x, \text{s.t.}$

$$e = \text{Enc} ( r_1^* \cdot P(x) + s^* ) \quad \wedge \quad h = H_2 ( r_2^*, x )$$

# Other results and open problems

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- Approximating size of intersection (scalar product)
  - Requires  $\Omega(k)$  communication
  - Provide secure approximation protocol
- PM protocol extends efficiently to multiple parties
- Malicious-party protocol in standard model?
- Fuzzy Matching?
  - Databases are not always accurate or full
  - Report iff entries match in  $t$  out of  $V$  “attributes”



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Questions?