

# **An overview of privacy concepts in computer science**

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\*Joint work with Anubhuti Singh, Malavika Raghavan, Prof. Subodh Sharma and Prof. Subhashis Banerjee

# Agenda

To give an overview of privacy techniques in computer science and evaluate how well they align with the legal principles of privacy

- Only a broad-strokes picture (privacy research almost 4 decades old, dozens of journals and conferences, thousands of PhDs)
- Details and references in Sec. 3 of “*An operational architecture for privacy-by-design in public service applications*” by P. Agrawal, A. Singh, M. Raghavan, S. Sharma and S. Banerjee. (Link: <https://arxiv.org/pdf/2006.04654.pdf>)

# Informational privacy, data security and data protection

- ***Informational privacy:*** The broad concept of individuals' right to be left alone ([Puttaswamy I](#))
- ***Data security:*** The technical safeguards and operations kept in place by entities to protect the data that is collected and stored by them.
- ***Data protection:*** A legal framework for achieving informational privacy, e.g., by preventing unlawful collection and processing by entities.

# Legal principles of privacy (OECD)

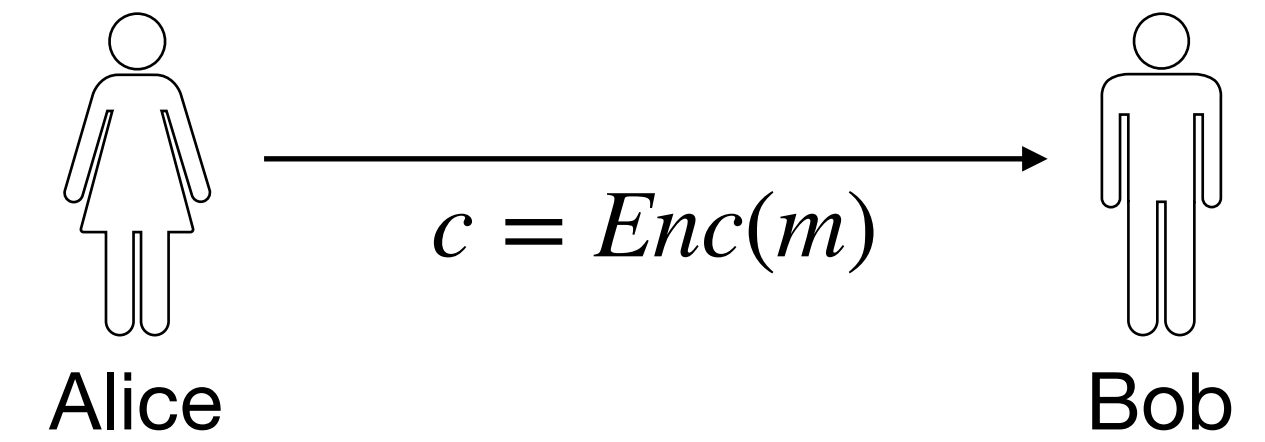
- ***Collection Limitation:*** Collection should occur with the knowledge and consent of the individual
- ***Data Quality:*** Only relevant and necessary information is collected
- ***Purpose Specification:*** Intended purpose of data collection must be specified
- ***Use Limitation:*** Data shouldn't be used for purposes other than those specified at collection
- ***Individual Participation***
- ***Security Safeguards***
- ***Openness***
- ***Accountability***

# Privacy risks

1. Leakage of sensitive data in transit
2. Unauthorised access of information
3. Linking of information shared across multiple databases
4. Re-identification of individuals from anonymised data
5. Post-access manual purpose violation (insider attacks)
  - e.g., data selling, illegal surveillance, misuse of data
6. Post-access automatic purpose violation
  - e.g., illegal profiling or targeting using AI

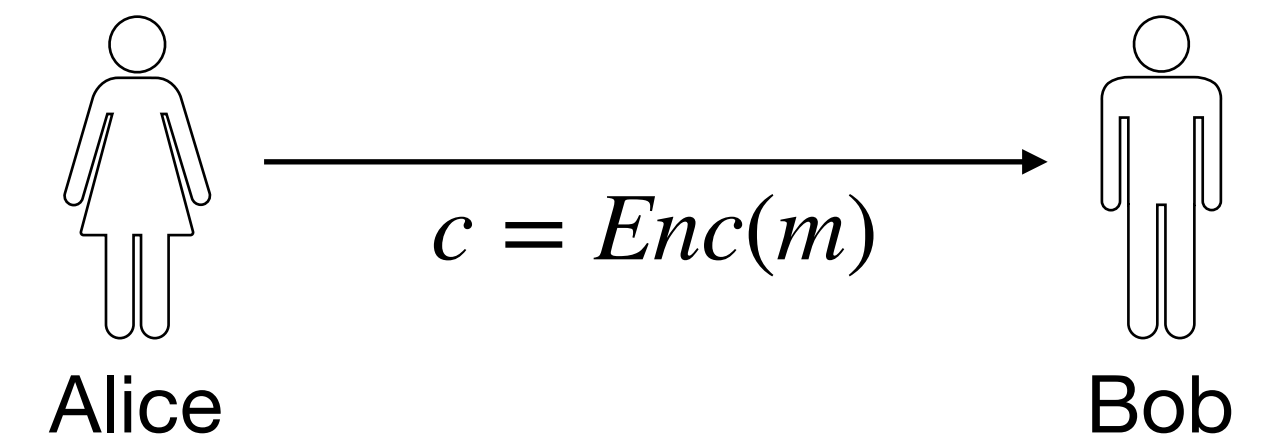
# Encryption

- Protects data in storage & transit
  - given  $c$ , hard to find  $m$  (e.g., hardness of factoring in RSA - [Rivest et al. '78](#))

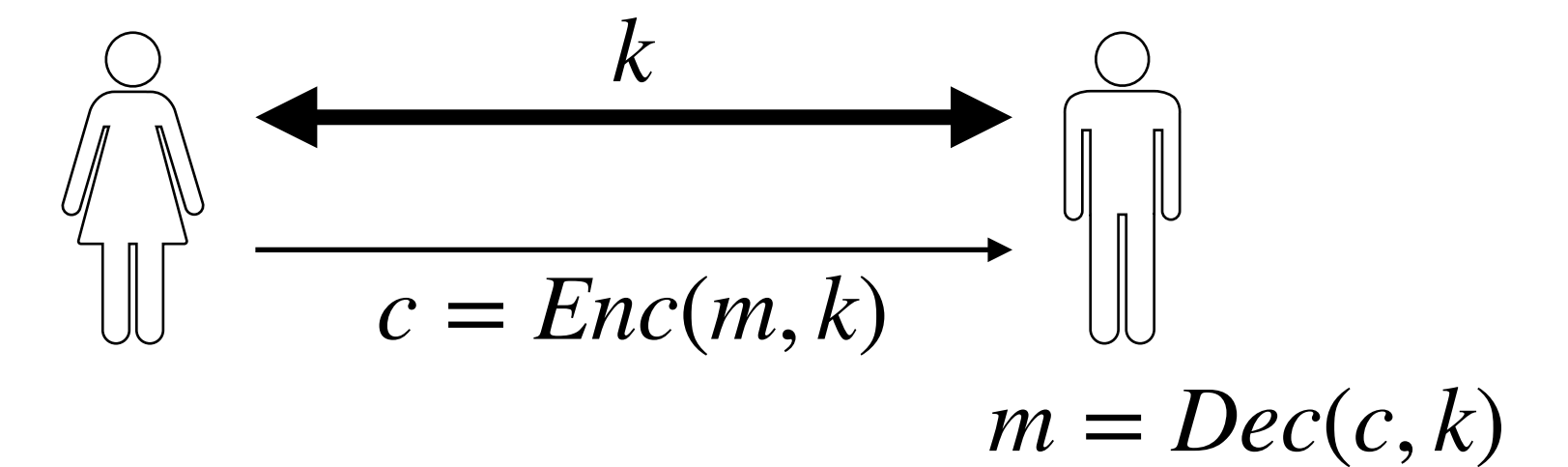


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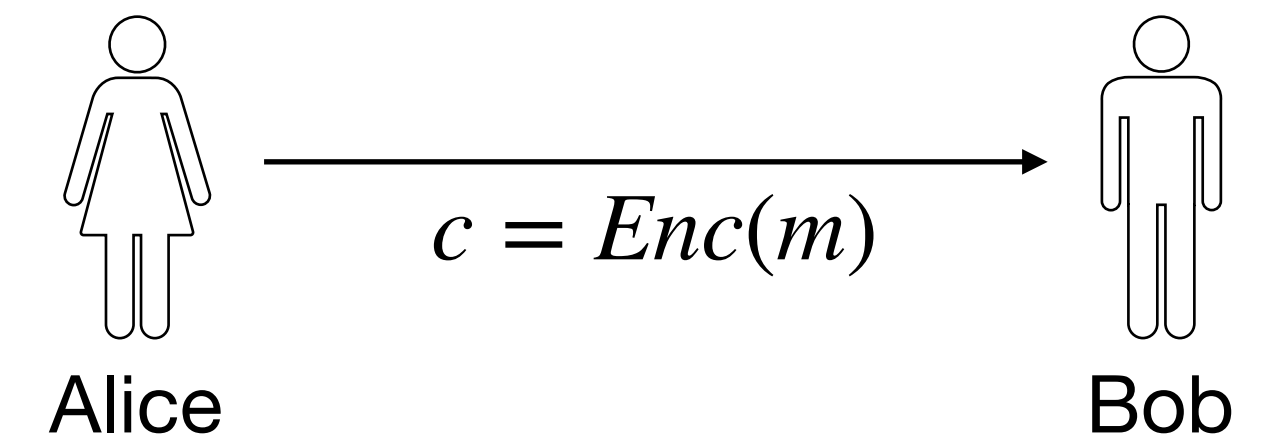


- Symmetric encryption (aka shared-key encryption)

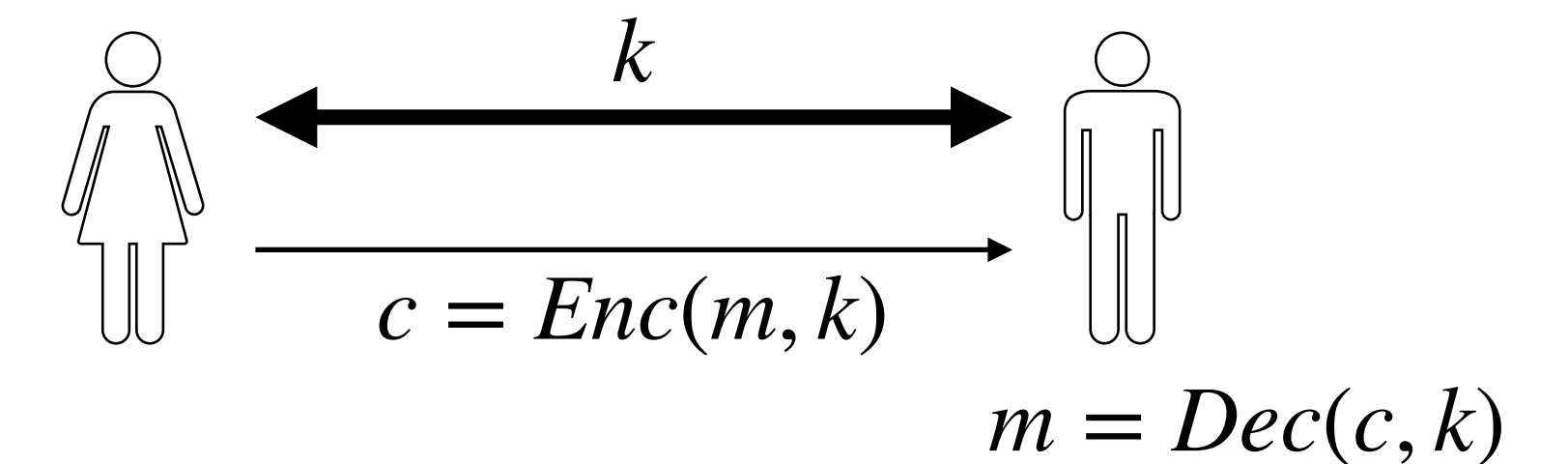


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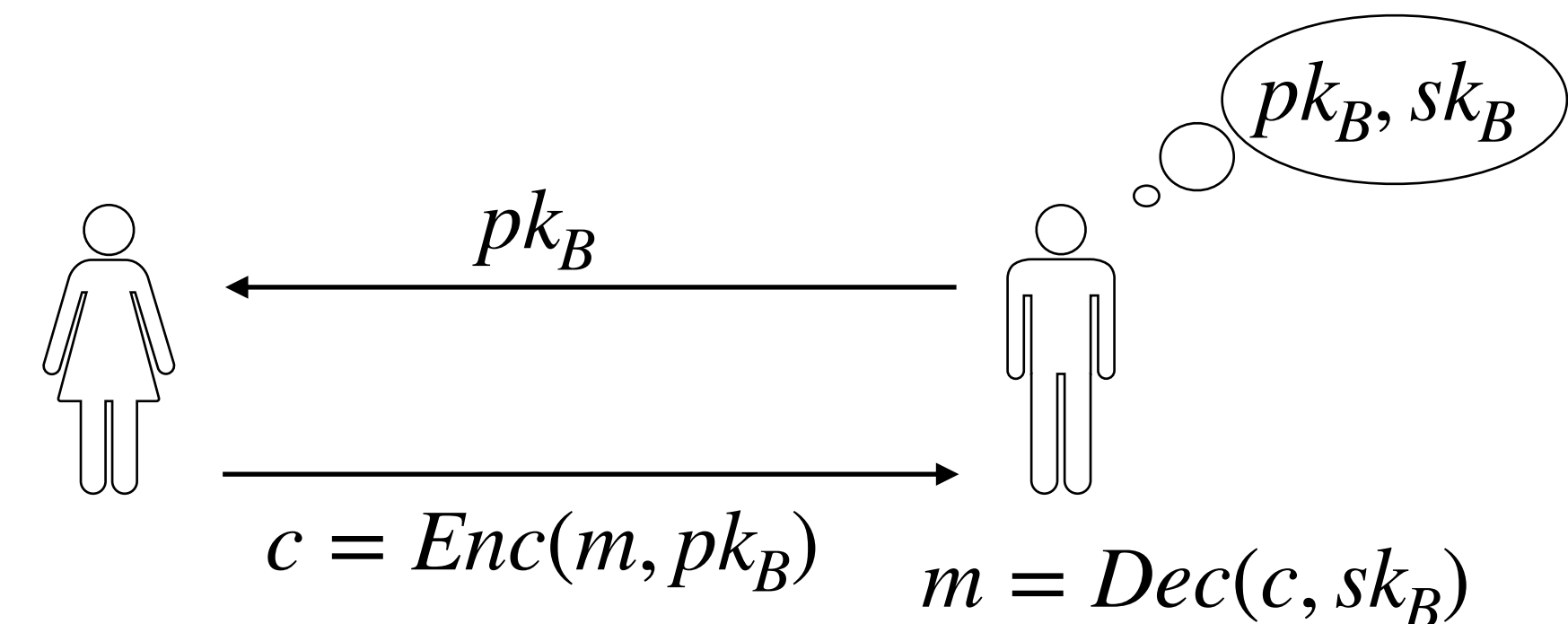
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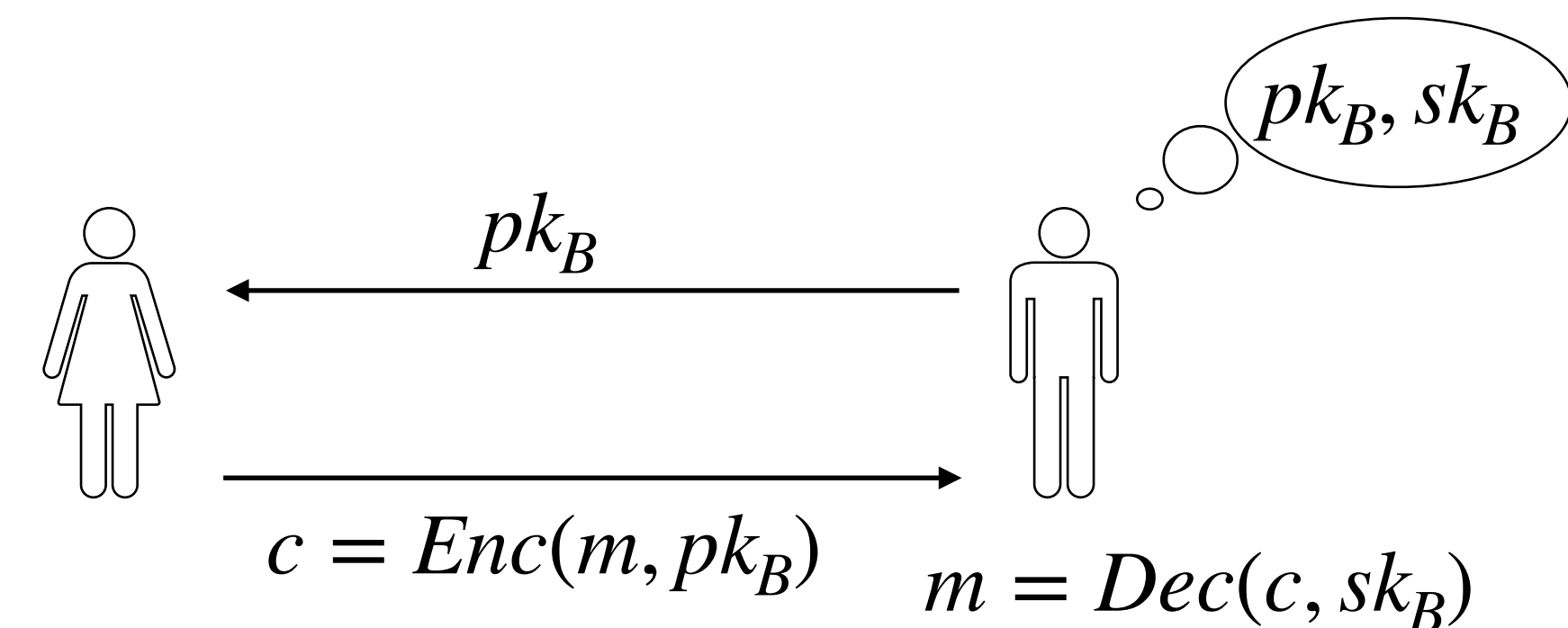
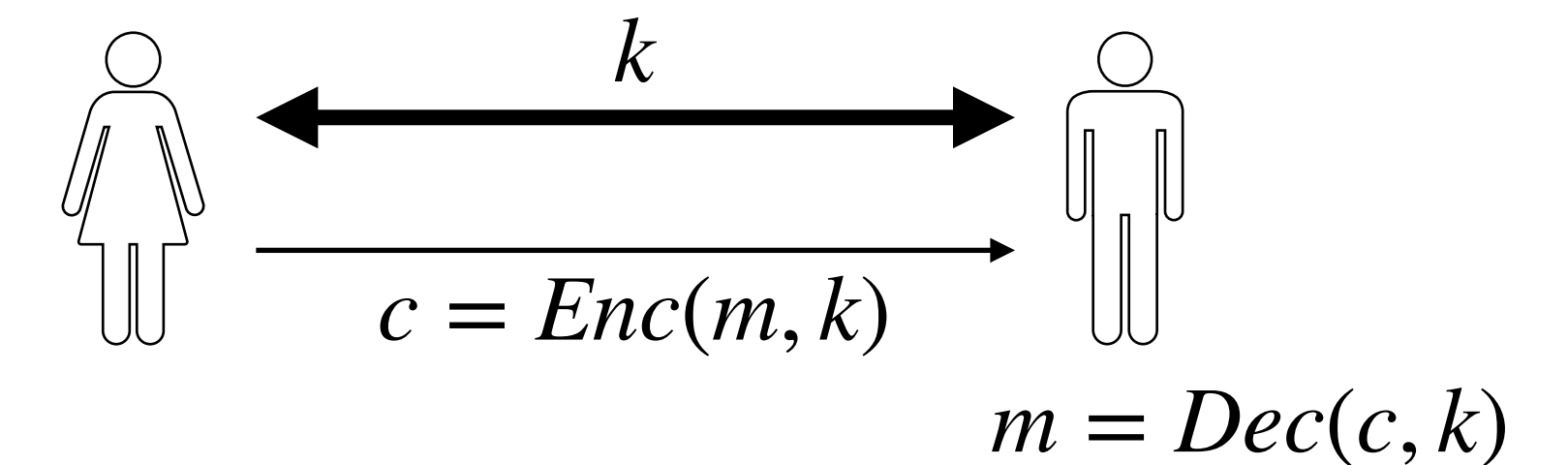
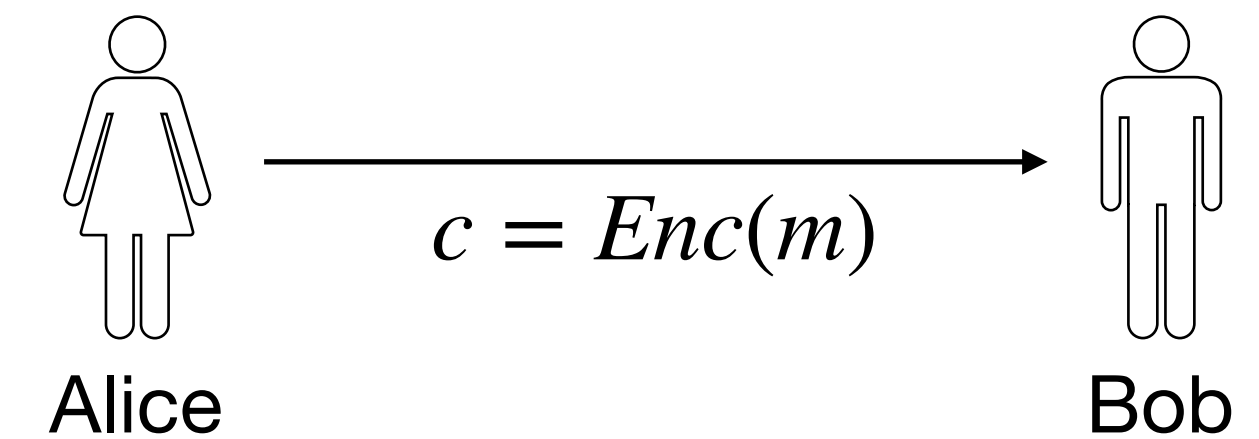
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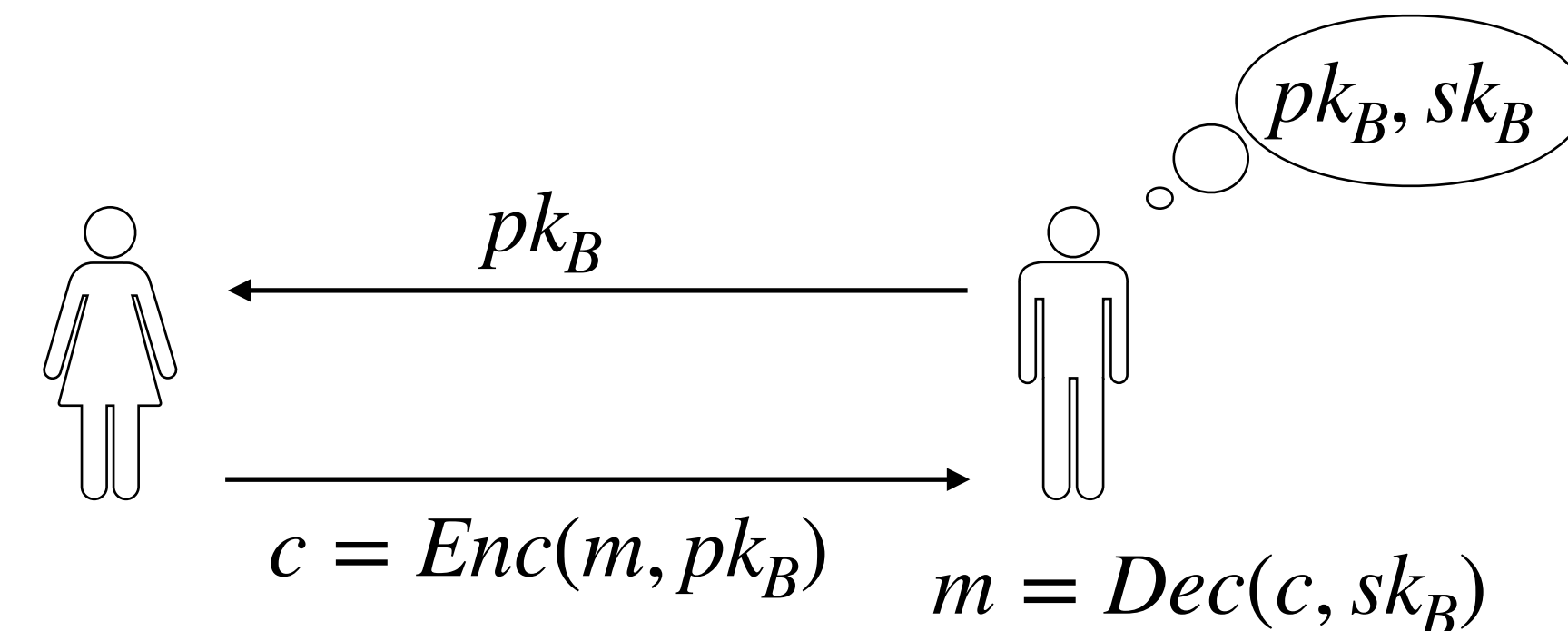
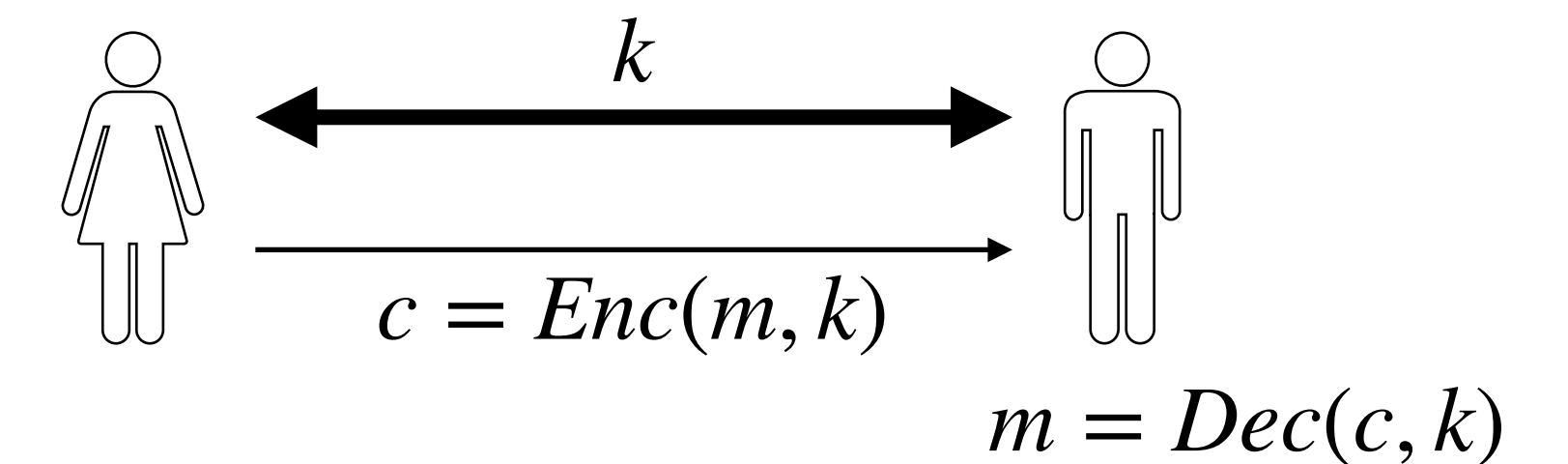
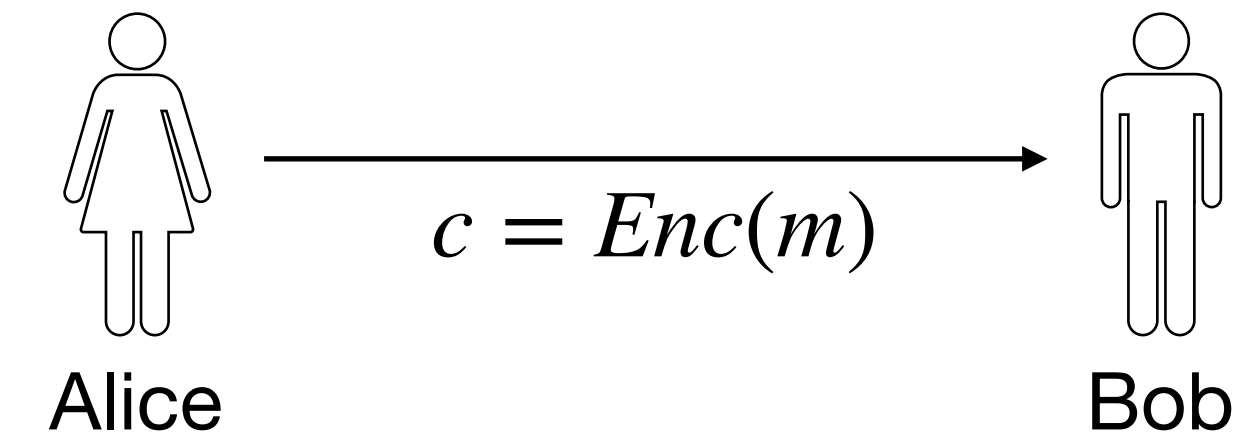
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# Encryption

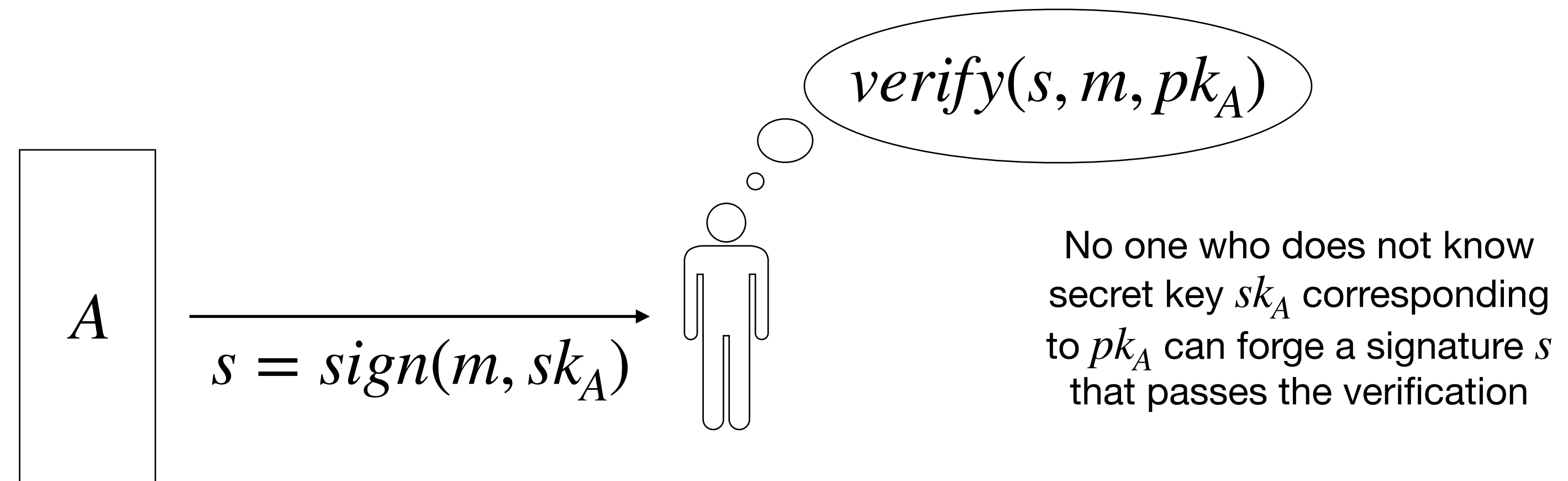
Computation still  
requires decrypting to  
plaintext

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# Digital signatures

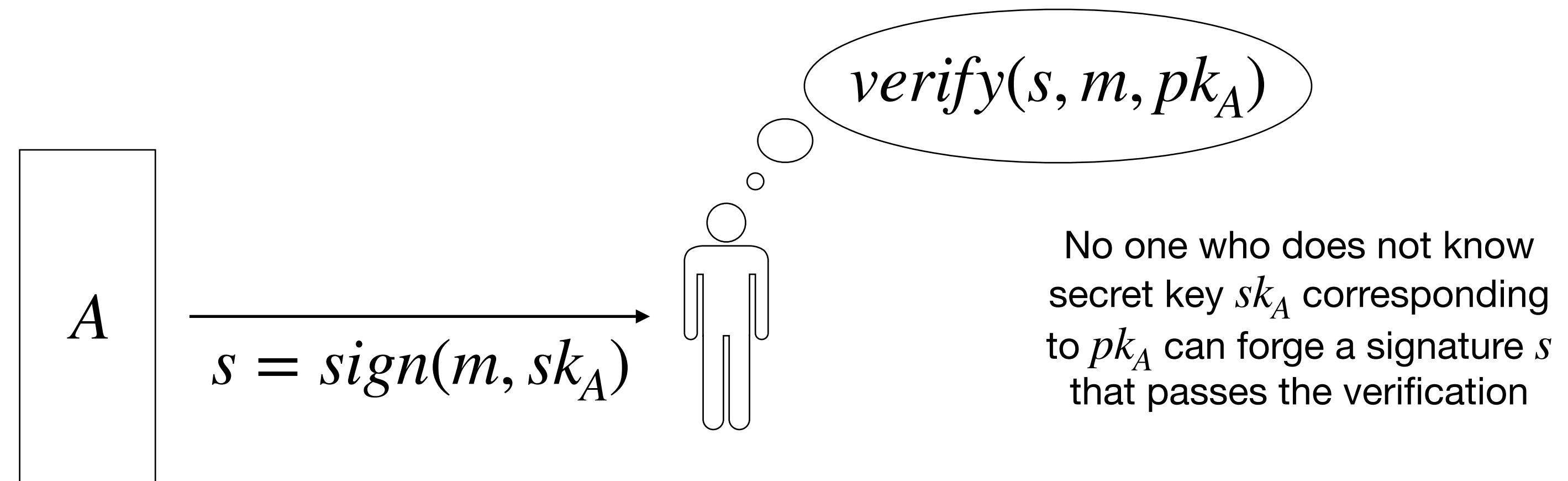
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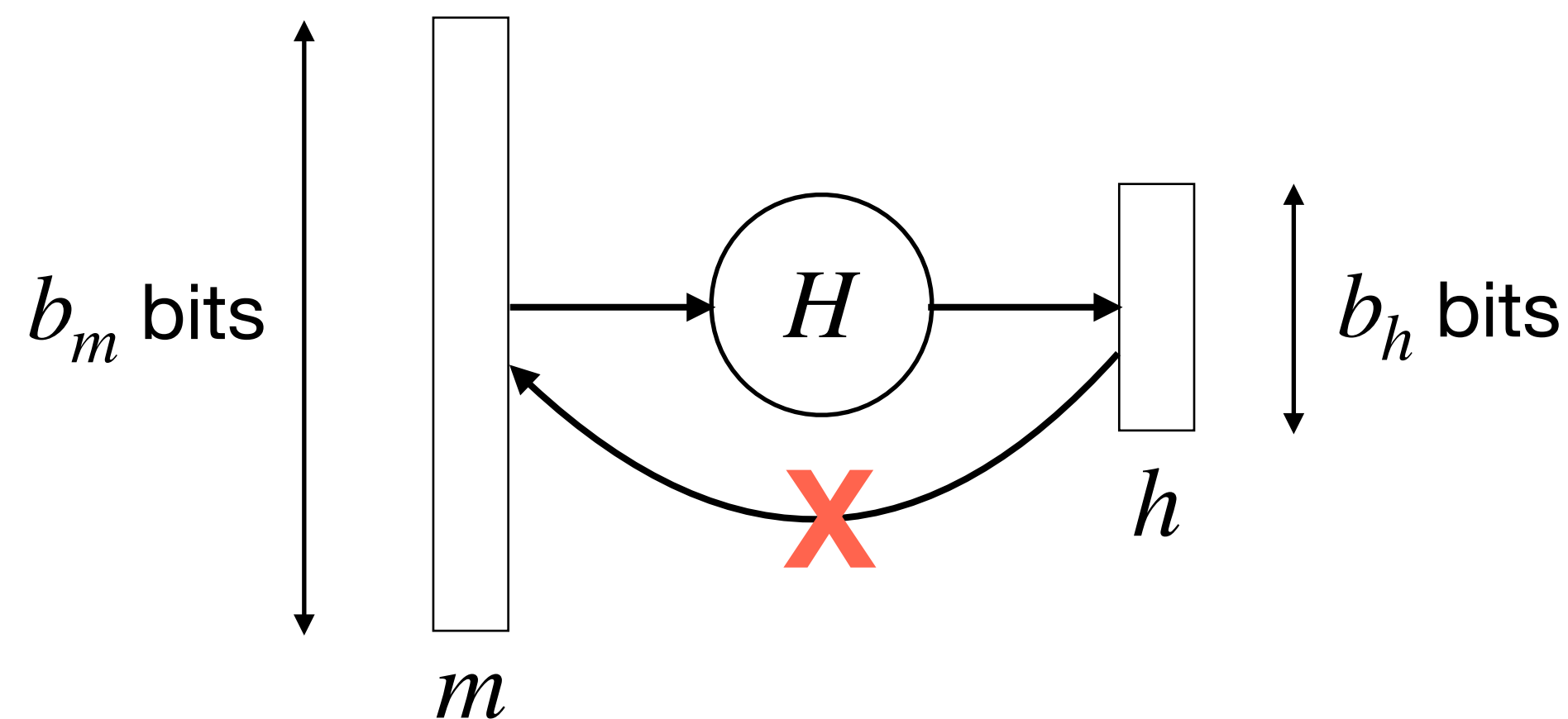


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***Security of encryption and digital signatures crucially depends on who has access to the secret keys!***

# One-way hash functions

- Given message  $m$ , computing hash  $h = H(m)$  is easy, but given  $h$ , it's hard to find which  $m$  might have produced it ( $h$  appears to have been *randomly generated*).



- Collision resistance*: It's even hard to find another  $m'$  such that  $h = H(m')$
- E.g., md5, sha256, etc.

# Data minimisation (DM) (aka “minimum disclosure”)

- Share only the minimum amount of data required for the purpose
  - e.g., to prove that “I am over 18 years” without disclosing my exact DOB or unique ID
  - e.g., sharing only anonymised information for statistical purposes

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- Many techniques and concepts:
  1. Zero-knowledge proofs
  2. Anonymity and unlinkability: virtual identities, anonymous credentials, etc.
  3. Database anonymisation

# DM1: Zero-knowledge proofs

- Techniques to prove a statement *without revealing anything other than the statement itself* ([Goldwasser et al. '89](#))
  - e.g., to prove that “I know the secret key corresponding to a given public key”, without revealing the secret key



# Sudoku in Zero Knowledge

**Goal:** Prover wants to prove to the verifier that it knows the solution\* to the following Sudoku puzzle, without revealing the solution to the verifier.

	1	2	3	4	5	6	7	8	9
A							6	8	
B					7	3			9
C	3		9					4	5
D	4	9							
E	8		3		5		9		2
F								3	6
G	9	6					3		8
H	7			6	8				
I		2	8						

\* For a correct solution to a Sudoku problem, each row, column and box must contain all the numbers from 1-9

# Sudoku in Zero Knowledge

- Step 1 (**Commitment**): Prover writes the solution for each cell in one of the 81 cards placed face down over the Sudoku grid. Since the cards are numbered, the prover is *committed* to the location of each card.



Face up



Face down

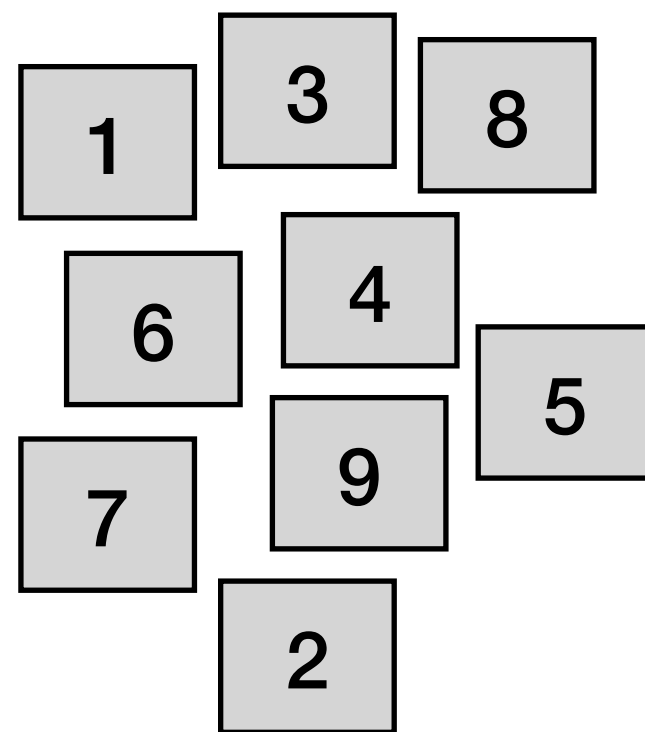
	1	2	3	4	5	6	7	8	9
A	A1	A2	A3	A4	A5	A6	A7	A8	A9
B	B1	B2	B3	B4	B5	B6	B7	B8	B9
C	C1	C2	C3	C4	C5	C6	C7	C8	C9
D	D1	D2	D3	D4	D5	D6	D7	D8	D9
E	E1	E2	E3	E4	E5	E6	E7	E8	E9
F	F1	F2	F3	F4	F5	F6	F7	F8	F9
G	G1	G2	G3	G4	G5	G6	G7	G8	G9
H	H1	H2	H3	H4	H5	H6	H7	H8	H9
I	I1	I2	I3	I4	I5	I6	I7	I8	I9

*Note:* Commitment is both **binding** and **hiding**!

# Sudoku in Zero Knowledge

- Step 2 (**Challenge/Response**): Verifier challenges to open a random column/row/box. Prover shuffles the cards in the requested column/row/box and reveals face up. Verifier checks that all 1-9 are present and all other face-down cards are intact. Repeat  $k$  times.

Shuffled cards of the first column revealed face up!



	1	2	3	4	5	6	7	8	9
A		A2	A3	A4	A5	A6	A7	A8	A9
B		B2	B3	B4	B5	B6	B7	B8	B9
C	3	C2	C3	C4	C5	C6	C7	C8	C9
D	4	D2	D3	D4	D5	D6	D7	D8	D9
E	8	E2	E3	E4	E5	E6	E7	E8	E9
F		F2	F3	F4	F5	F6	F7	F8	F9
G	9	G2	G3	G4	G5	G6	G7	G8	G9
H	7	H2	H3	H4	H5	H6	H7	H8	H9
I		I2	I3	I4	I5	I6	I7	I8	I9

Original problem

	1	2	3	4	5	6	7	8	9
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B					7	3			9
C	3		9					4	5
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F								3	6
G	9	6					3		8
H	7			6	8				
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# Sudoku in Zero Knowledge

- Step 3 (**Final Reveal**): Prover reveals all non-empty cells in the original Sudoku problem to show that he has actually solved the given problem, and not some other problem.

	1	2	3	4	5	6	7	8	9
A	A1	A2	A3	A4	A5	A6	6	8	A9
B	B1	B2	B3	B4	7	3	B7	B8	9
C	3	C2	9	C4	C5	C6	C7	4	5
D	4	9	D3	D4	D5	D6	D7	D8	D9
E	8	E2	3	E4	5	E6	9	E8	2
F	F1	F2	F3	F4	F5	F6	F7	3	6
G	9	6	G3	G4	G5	G6	3	G8	8
H	7	H2	H3	6	8	H6	H7	H8	H9
I	I1	2	8	I4	I5	I6	I7	I8	I9

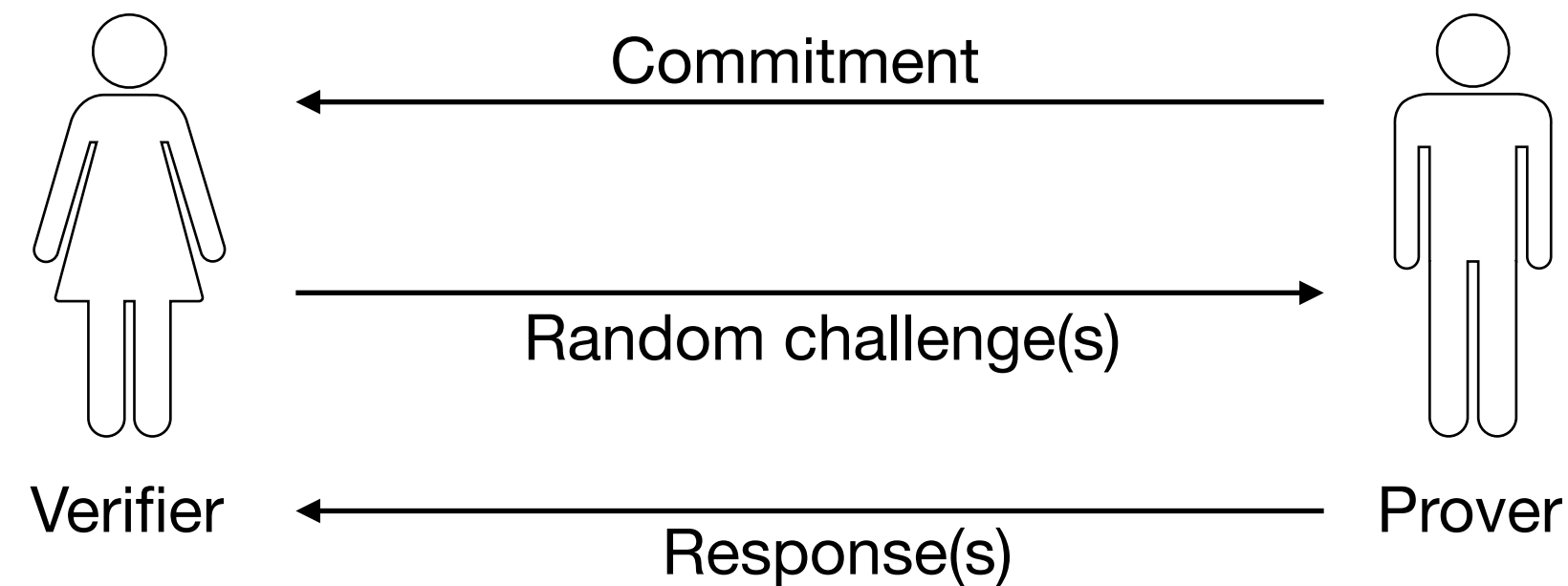
Verifier is convinced that prover knows the solution but does not learn *anything* about the solution.

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# Zero-knowledge proofs: Summary

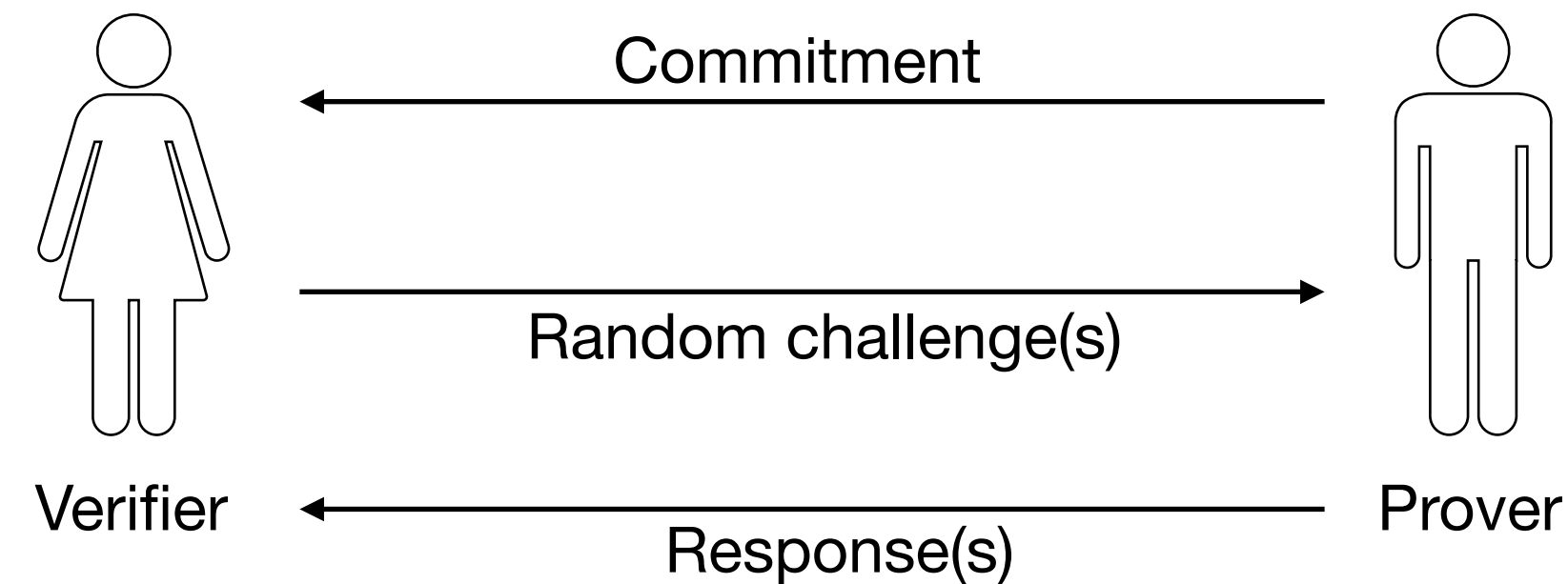
- Typical structure



- Commitment with binding+hiding properties is a cryptographic construct too ([Pedersen '91](#))
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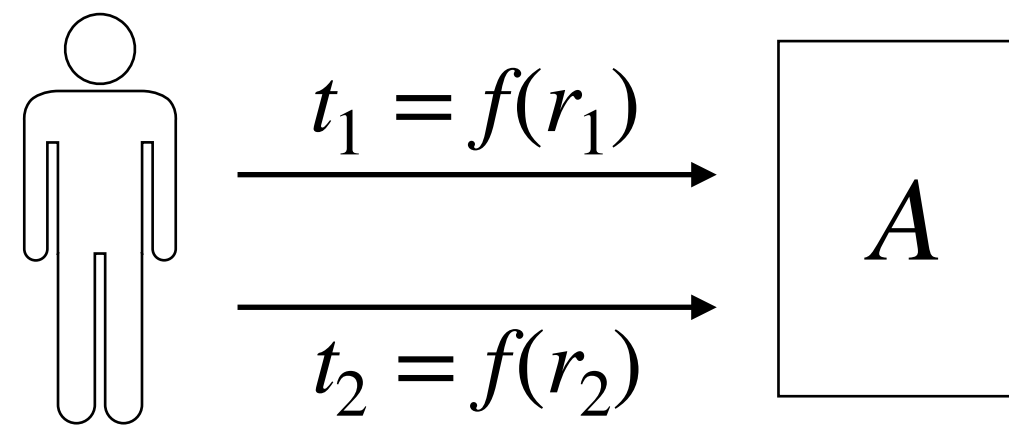


- Commitment with binding+hiding properties is a cryptographic construct too ([Pedersen '91](#))
- Interactive ZKPs can be made non-interactive using one-way hash functions instead of random challenges ([Fiat & Shamir '86](#))
- **All practical statements\*** can be proved in zero knowledge with **overwhelming probability** ([Goldreich et al. '91](#))

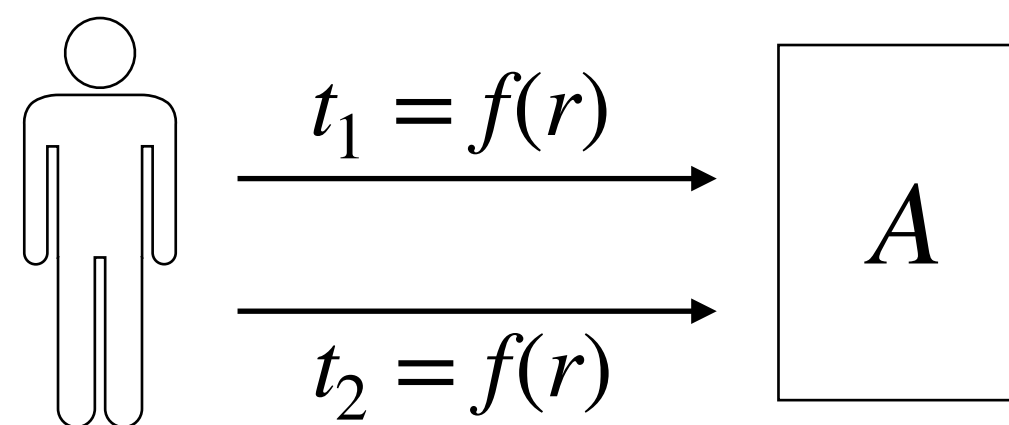
\* All NP statements, i.e., which can be verified in polynomial time

# DM2: Anonymity and unlinkability

- *Anonymity*: The state of not being identifiable in a set of individuals
- *Unlinkable anonymity*: Transactions do not reveal individuals' true identities and even multiple transactions by the same individual are unlinkable

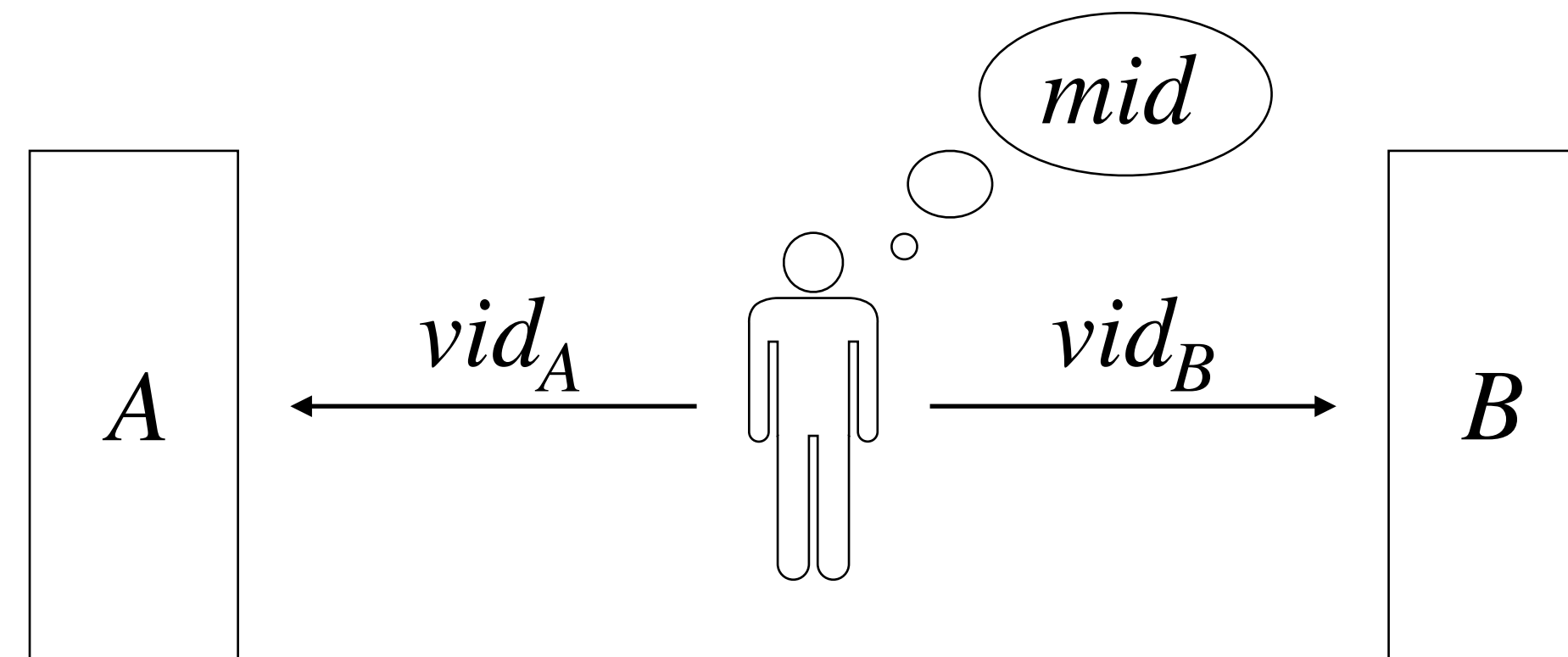


- *Linkable anonymity*: Transactions do not reveal individuals' true identities but multiple transactions by the same individual are linkable



# Virtual identities

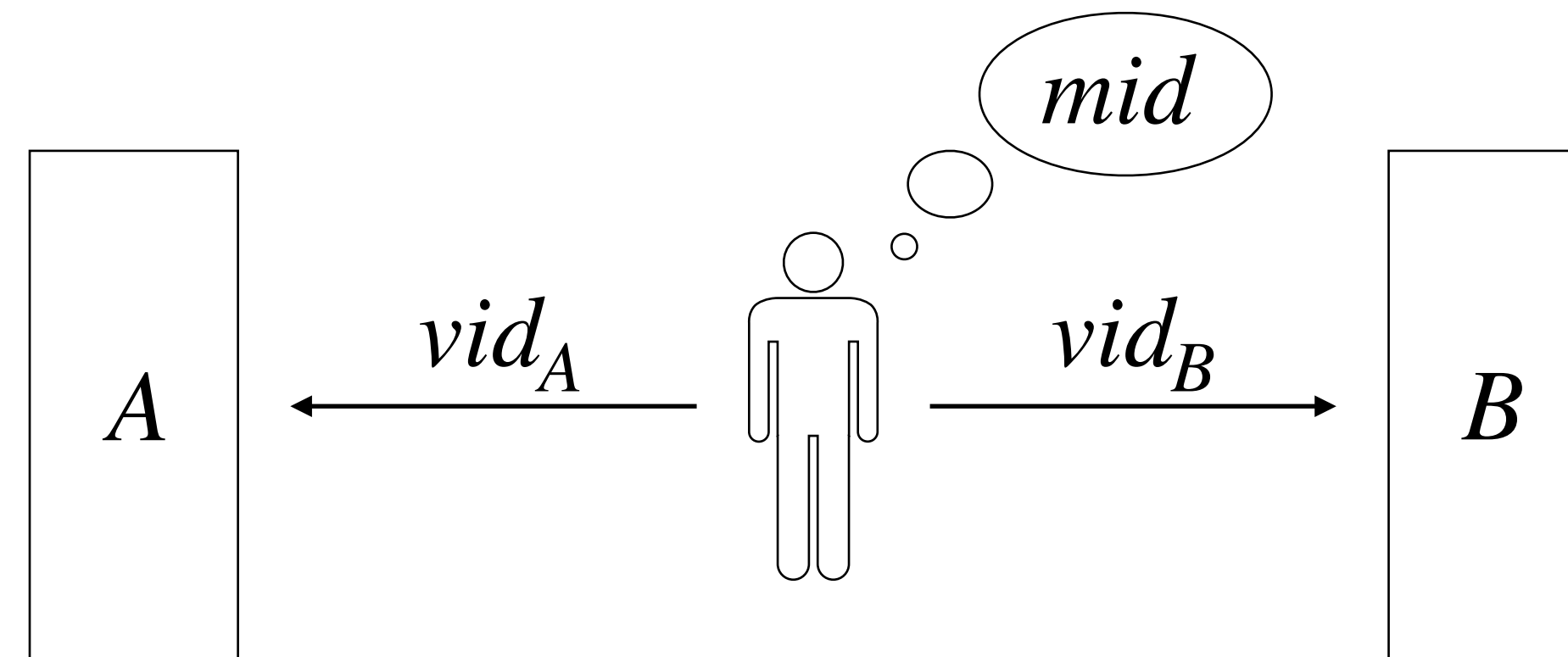
- Individuals own a master identity and generate random looking, completely unlinkable virtual identities for different organisations ([Chaum '85](#))





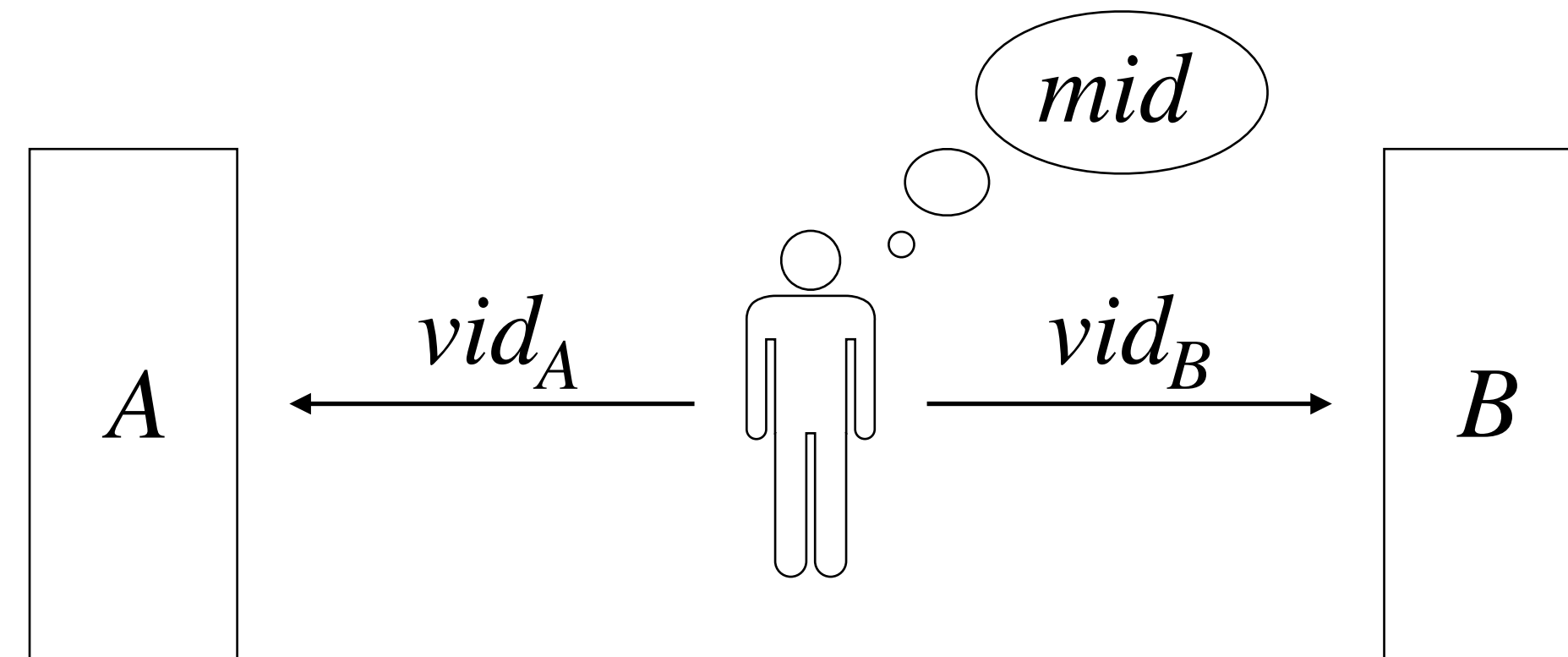
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- Unlinkable anonymity for inter-organisation transactions / linkable or unlinkable anonymity for intra-organisation transactions
- Purpose-limited linkability by a trusted authority (for genuine reasons, accountability)

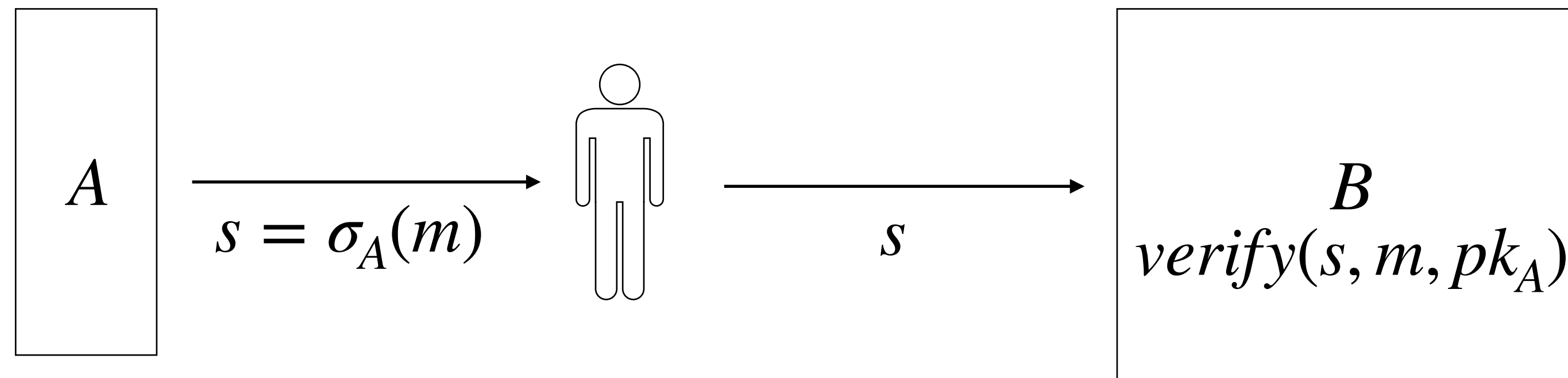
# Anonymous credentials



- What if the individual wants to show some credentials obtained from  $A$  to  $B$ , without allowing  $A$  or  $B$  to link  $vid_A$  and  $vid_B$  (e.g.,  $A$ =college,  $B$ =employer)?

# Anonymous credentials

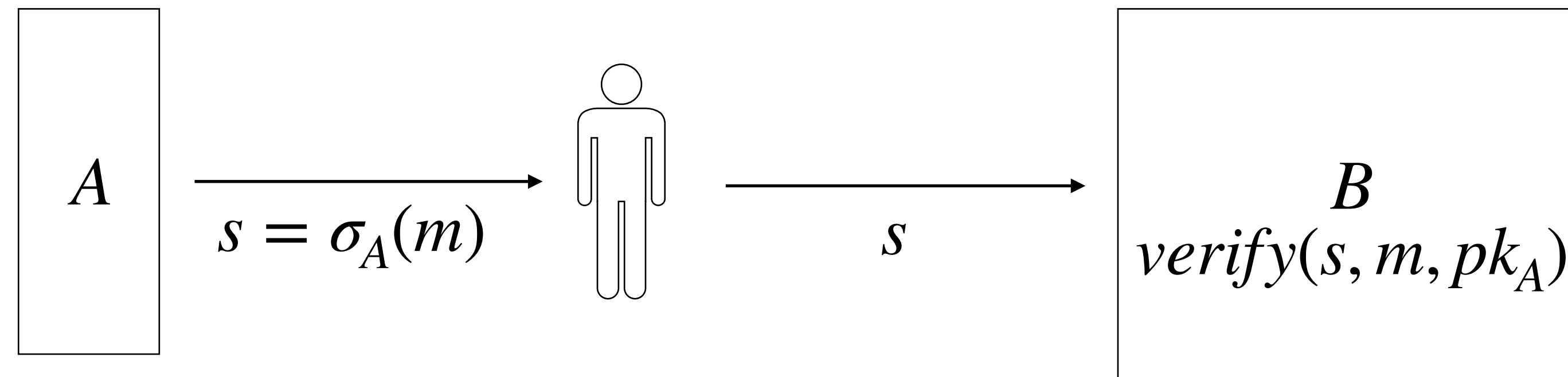
- Regular credentials: Based on *digital signatures* that are unforgeable



No one except  $A$  can forge a signature that passes the verification

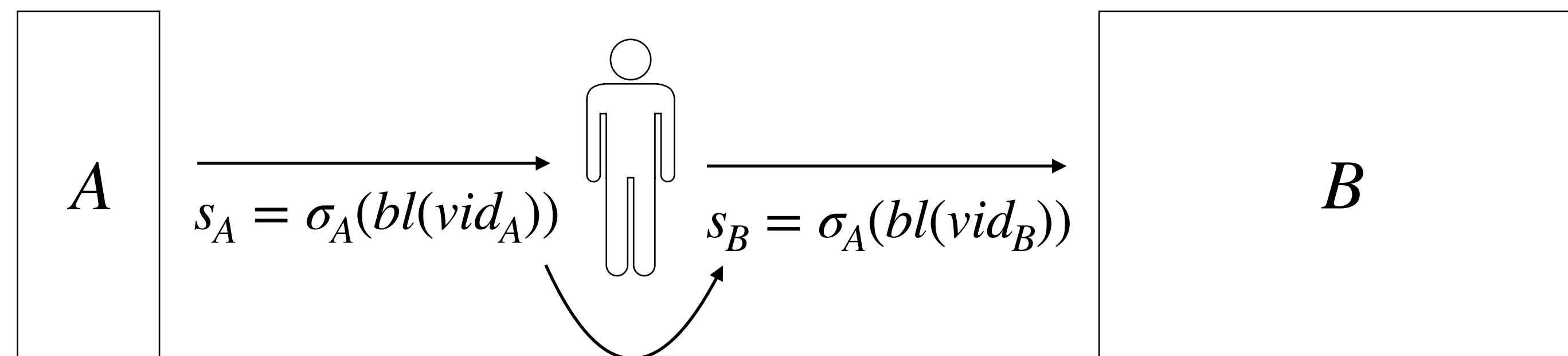
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No one except  $A$  can forge a signature that passes the verification

- Anonymous credentials: Often based on *blind signatures* that are transformable (Chaum '85)



No one can present a signature on a *vid* unless they obtained a signature on another *vid* they own

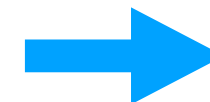
e.g.  $s_A = \sigma_A("vid_A \text{ has degree } X") \longrightarrow s_B = \sigma_A("vid_B \text{ has degree } X")$

# DM3: Database anonymisation

- Data minimisation technique to allow analytics on DBs while preserving anonymity
- Hide personally identifiable information by adding noise, suppressing info or coarsening data

Name	Age	Sex	Height	Weight	Location	HIV
Alice	15	F	5.5	84	Rohila Apartments, Pune	Yes
Bob	28	M	5.1	58	Bldg X, DLF Phase 3, GG	No
Charles	34	M	5.9	65	Sameer Bungalow, Delhi	No
David	43	M	6.1	76	55, Sunset Blvd, Mumbai	No

Medical Database

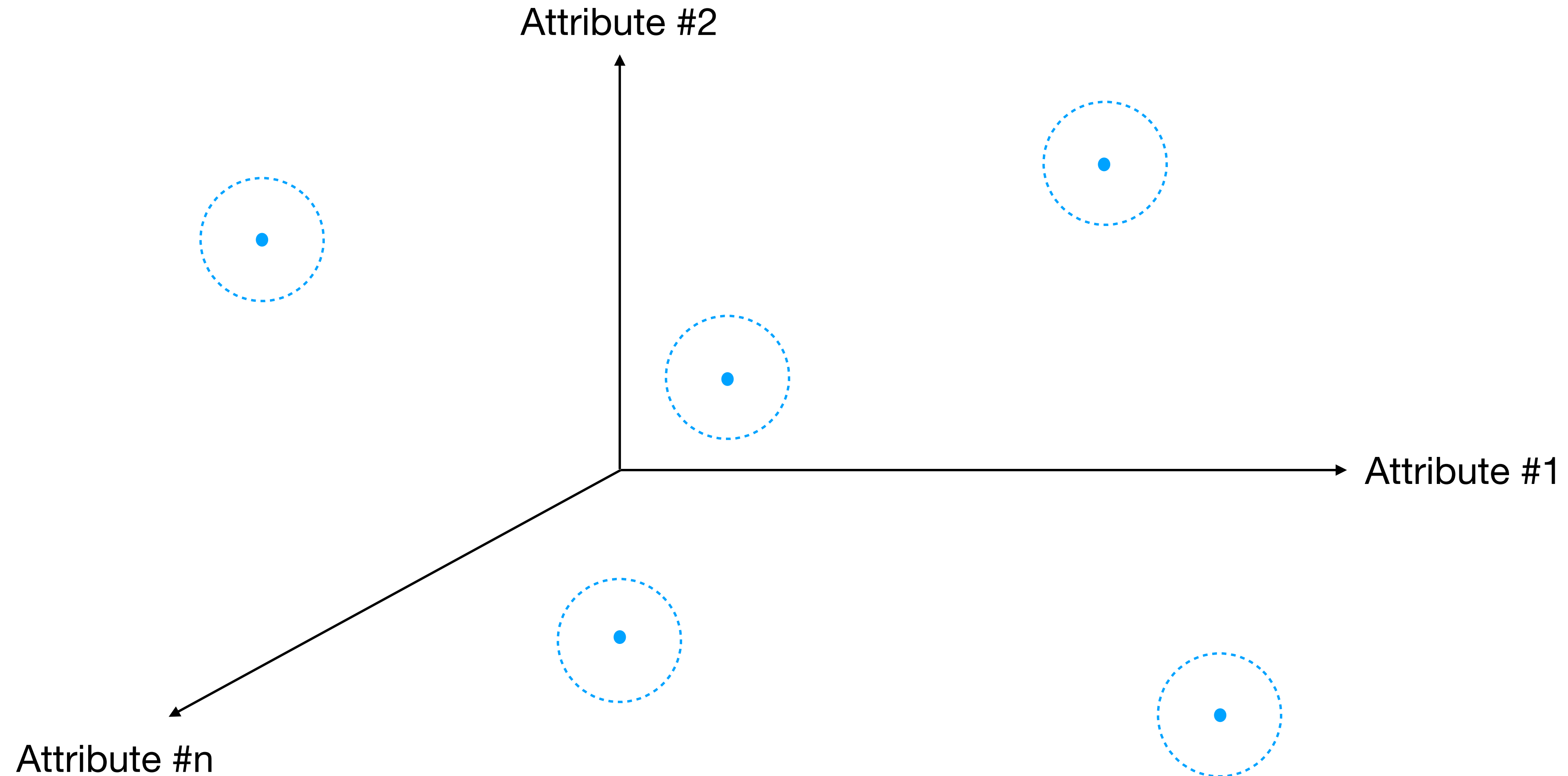


Name	Age	Sex	Height	Weight	Location	HIV Status
*	15-20	F	5.5-6.0	80-85	Pune	Yes
*	25-30	M	5.0-5.5	55-60	Gurgaon	No
*	30-35	M	5.5-6.0	60-65	Delhi	No
*	40-45	M	6.0-6.5	75-80	Mumbai	No

Anonymised Medical Database

- Many notions:  $k$ -anonymity,  $l$ -diversity, etc.

# Anonymisation is a myth



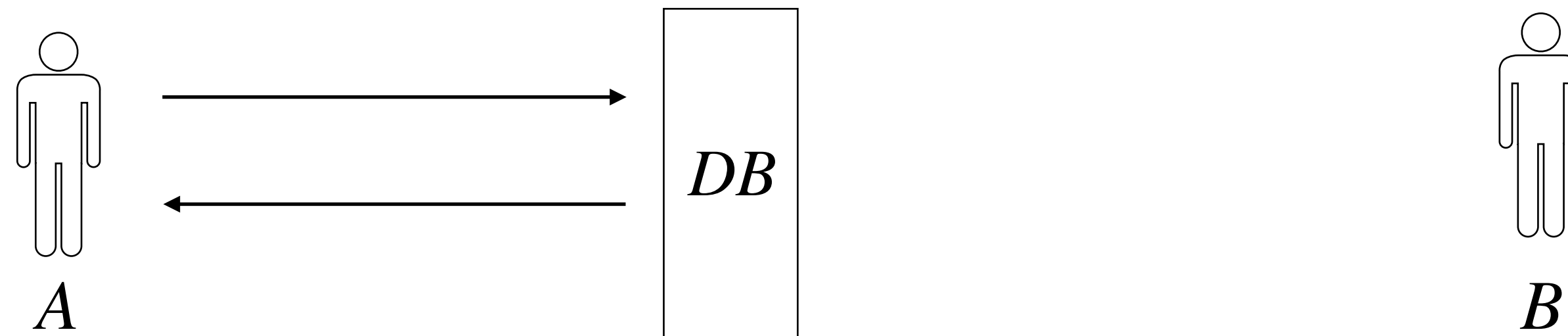
Individuals map to points in a sparse high-dimensional space where they are uniquely identifiable even after adding a lot of noise.

# Anonymisation is a myth

- De-anonymisation attacks on anonymised social network data, location data, writing style, source code, browser history, etc., exist. ([Narayanan et al. '19](#))
- *Theoretical bottlenecks:* Given a database with  $n$  rows, if the adversary is allowed to obtain answers to  $O(n)$  subset-sum queries, it can **reconstruct the entire database** (unless you add an unacceptable amount of noise) ([Dinur & Nissim '03](#))
- *Rough intuition:* Solve a bunch of linear equations to derive individual values from aggregate answers

# Impossibility of absolute privacy

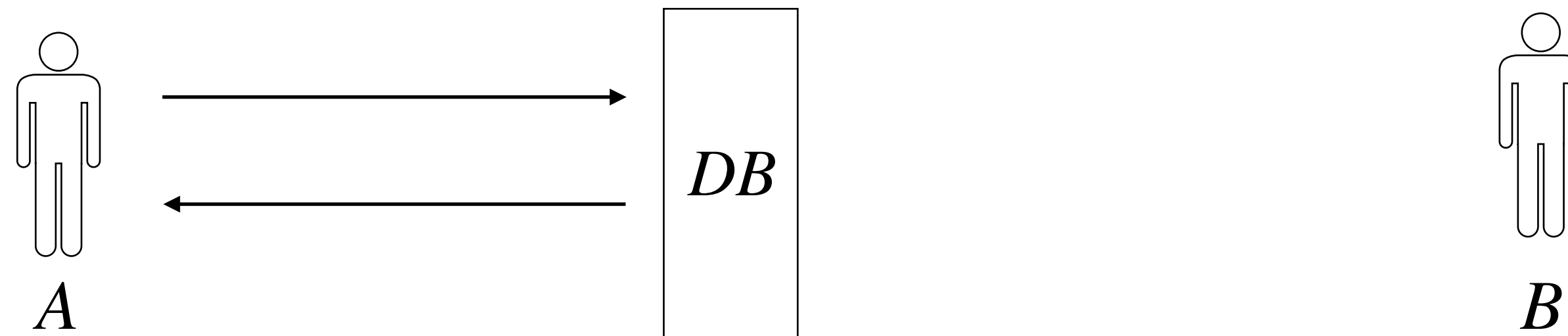
***Absolute privacy goal (aka inferential privacy):***  $A$  should not obtain any information about an individual that  $B$  cannot obtain without access to  $DB$





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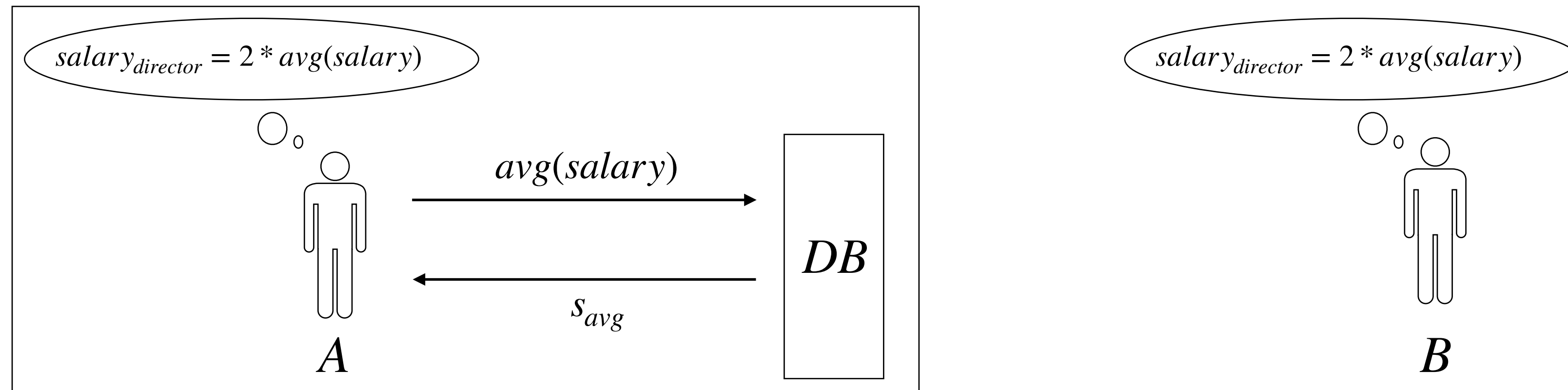
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If the adversary has arbitrary side-information, above absolute privacy goal is impossible to achieve. ([Dwork '05](#))

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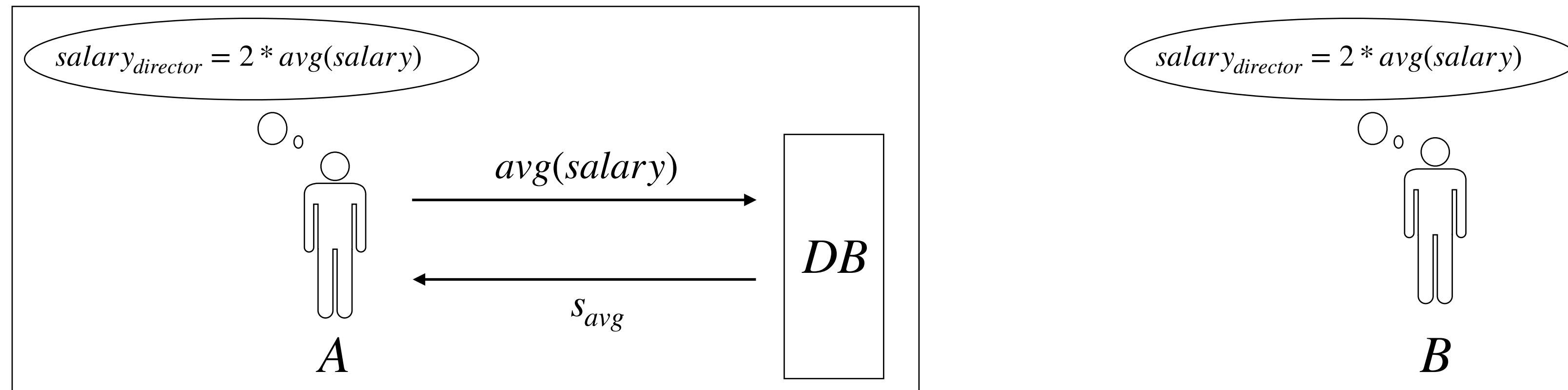
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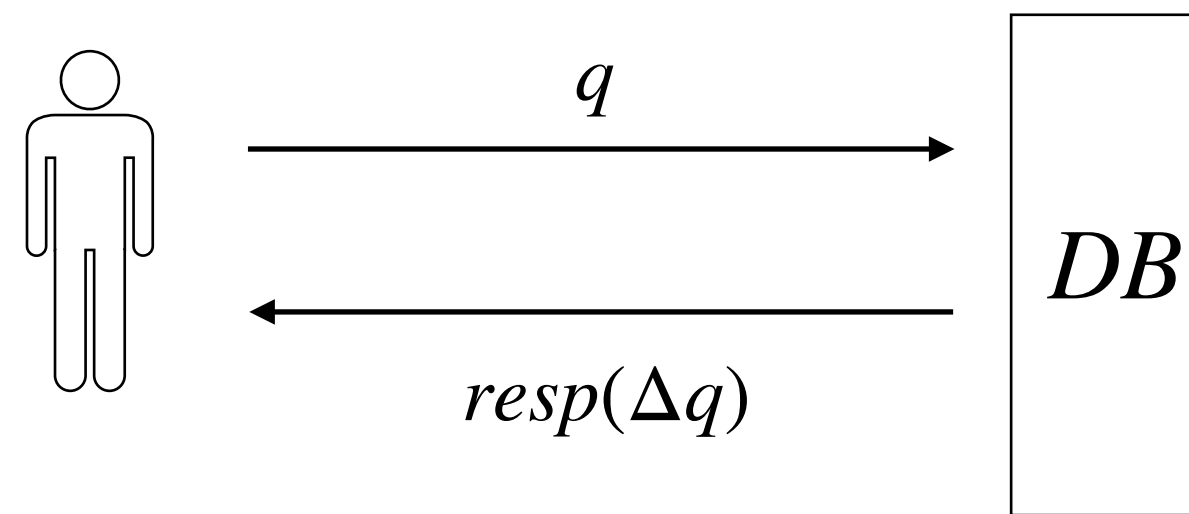


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**Observe: Privacy of director's salary is compromised even if the director is not in the  $DB$**

# Changing the goalpost: differential privacy

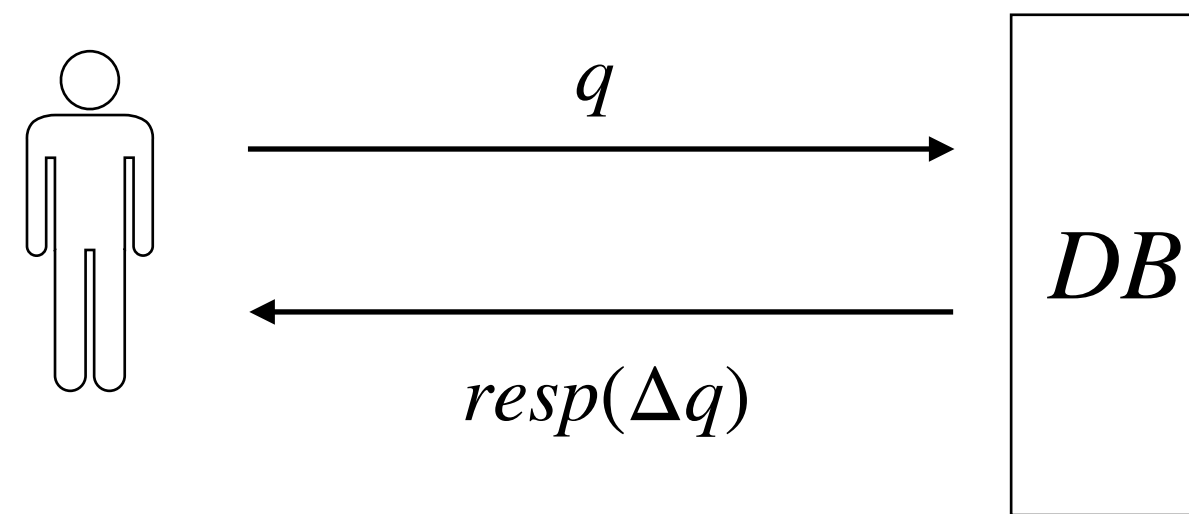
**Differential privacy goal:** Database should answer such that the *additional* privacy risk any individual incurs by participating in the database is minimal.



Query  $q$ 's sensitivity  $\Delta q$  measures how much the answer for  $q$  can change w.r.t. changes in only one row

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**Interactively calibrate noise as per query sensitivity to maintain user's differential privacy**

# Limitations of differential privacy

- As query sensitivity increases, noise increases
- DP mechanisms for different types of queries need to be specially designed
- Cannot easily answer many many queries (although some sophisticated noise addition techniques allow answering many queries of a special type)
- How to prevent community-level profiling (cf. Cambridge Analytica)?
- Not suitable for non-statistical uses

# Necessary conditions for privacy

- Impossibility of absolute privacy suggests that *all illegal data accesses and processing must be prevented in the first place.*
- Data controllers must declare purpose upfront and mechanisms should exist to only allow computations that fulfil the stated purpose.
- Legitimate purpose depends on dynamically changing consent, approvals, authentication, etc.
- Preliminary work on purpose-based privacy policies exists but it mostly assumes a poor proxy for purpose (e.g., role of the data requester)
- Also, data minimisation should be followed as a further defence and whenever data exits the regulatory boundary

# Secure remote execution

*Goal: Data should be secured such that a remote party can only execute a given program on it, and not use it in any other way.*

- **Software solutions (fully secure, but slow):**
  - Homomorphic encryption (computing in ciphertext space without decrypting)
  - Secure multiparty computation and garbled circuits
- **Hardware solutions (depend on trusted hardware, but fast):**
  - Intel SGX (many of today's laptops, desktops, etc.)
  - ARM Trustzone (Android devices)



# Homomorphic encryption (computing without decrypting)

***Additive  
homomorphic  
encryption***

$$Enc(a + b) = Enc(a) \oplus Enc(b)$$

***Multiplicative  
homomorphic  
encryption***

$$Enc(a \times b) = Enc(a) \otimes Enc(b)$$

***Fact:*** All computations can be expressed as a circuit of  $+$  and  $\times$ .

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$$Enc(a + b) = Enc(a) \oplus Enc(b)$$

**Multiplicative  
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$$Enc(a \times b) = Enc(a) \otimes Enc(b)$$

**Fully  
homomorphic  
encryption (FHE)**

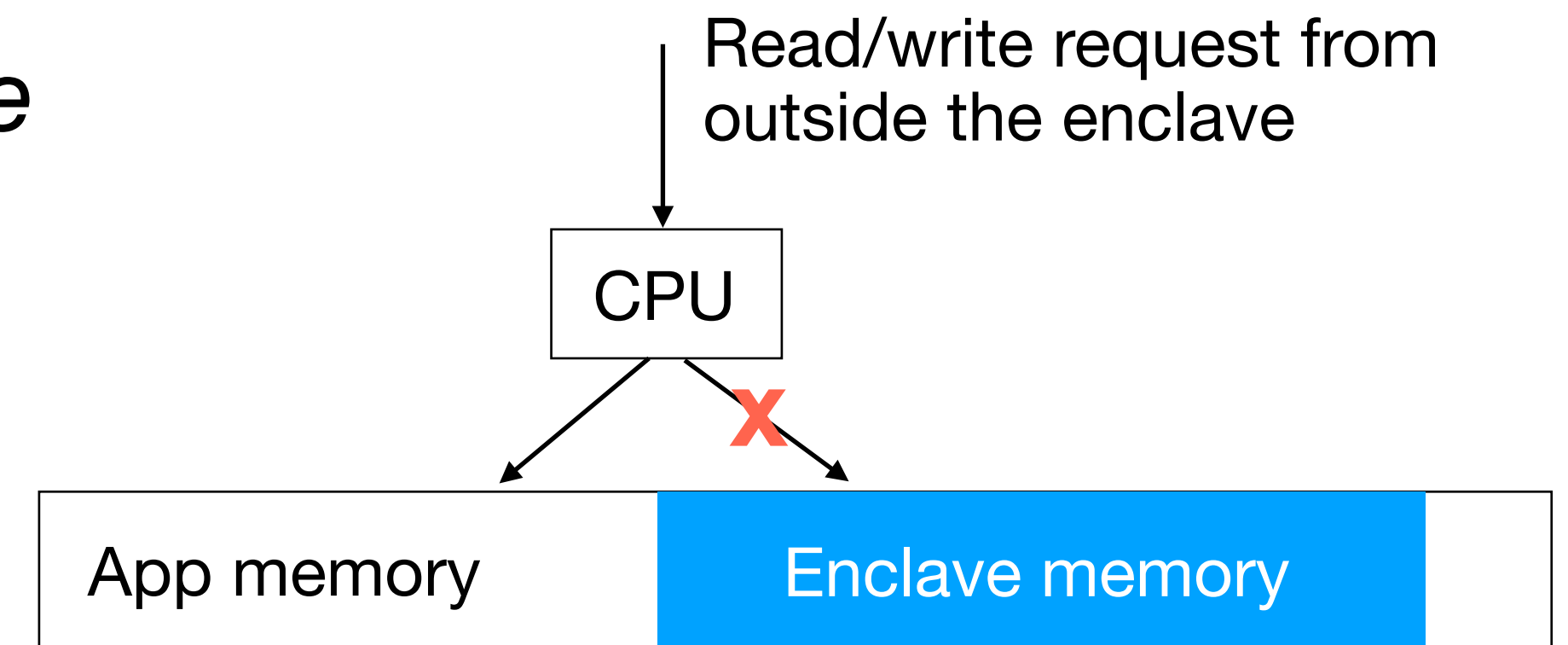
**Fact:** All computations can be expressed as a circuit of  $+$  and  $\times$ .

**Creating an encryption scheme  $Enc$  that satisfies both the above equations is tough. Such schemes exist but are extremely slow. Also, conversion to circuit model isn't practical.**

# Intel SGX

**Confidentiality:** *No one can peek into any state of the program running within the enclave.*

**Integrity:** *No one can modify execution of the program running within the enclave, except through explicit entry points.*

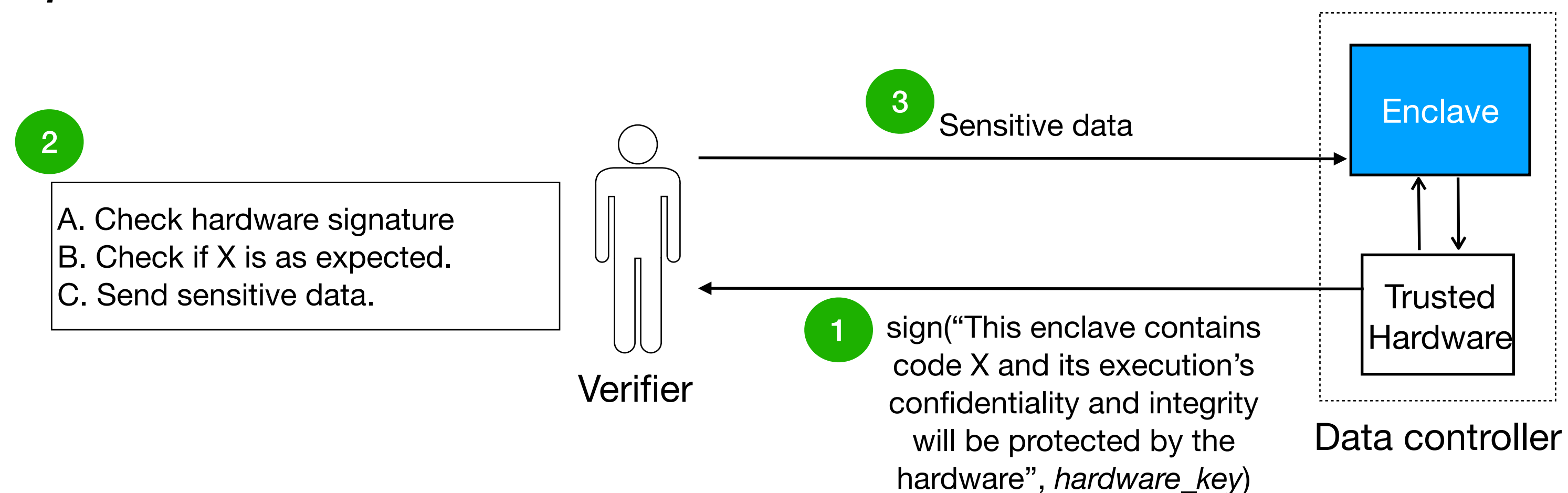
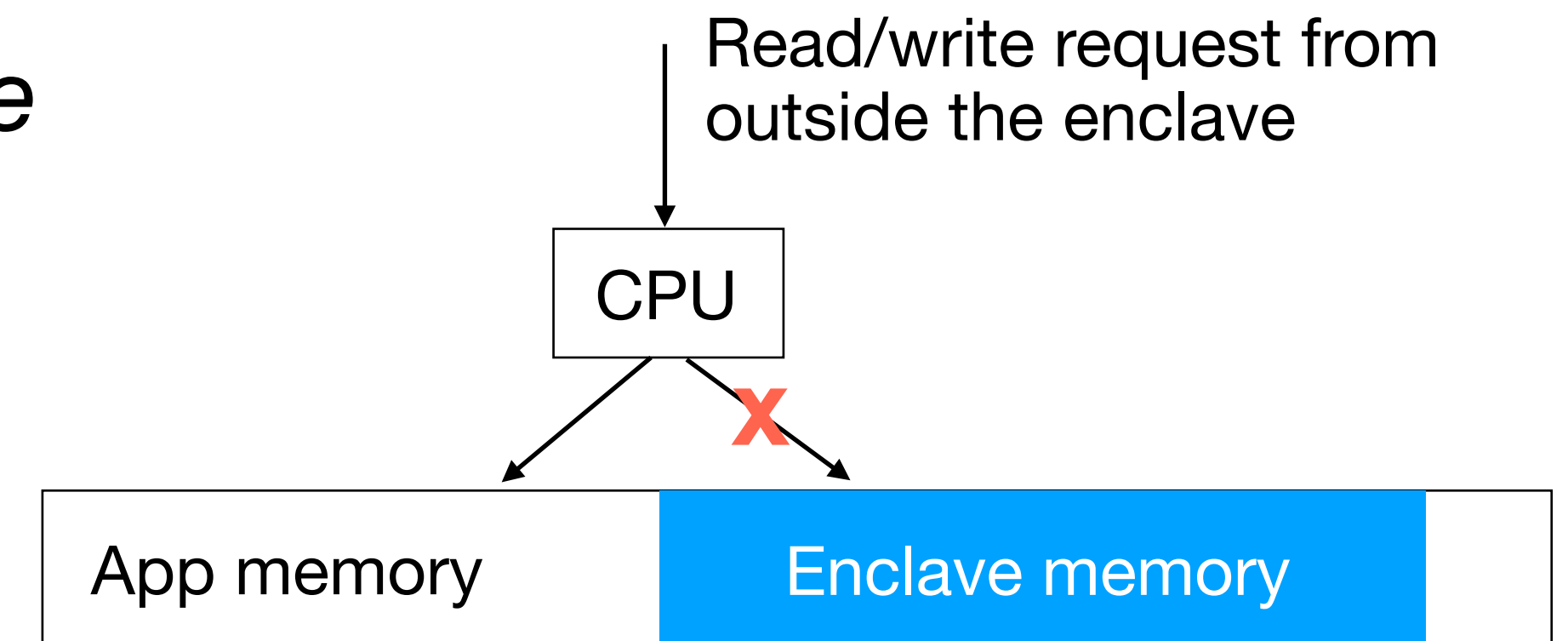


# Intel SGX

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**Remote attestation:** Sensitive data will only be processed as per code X.



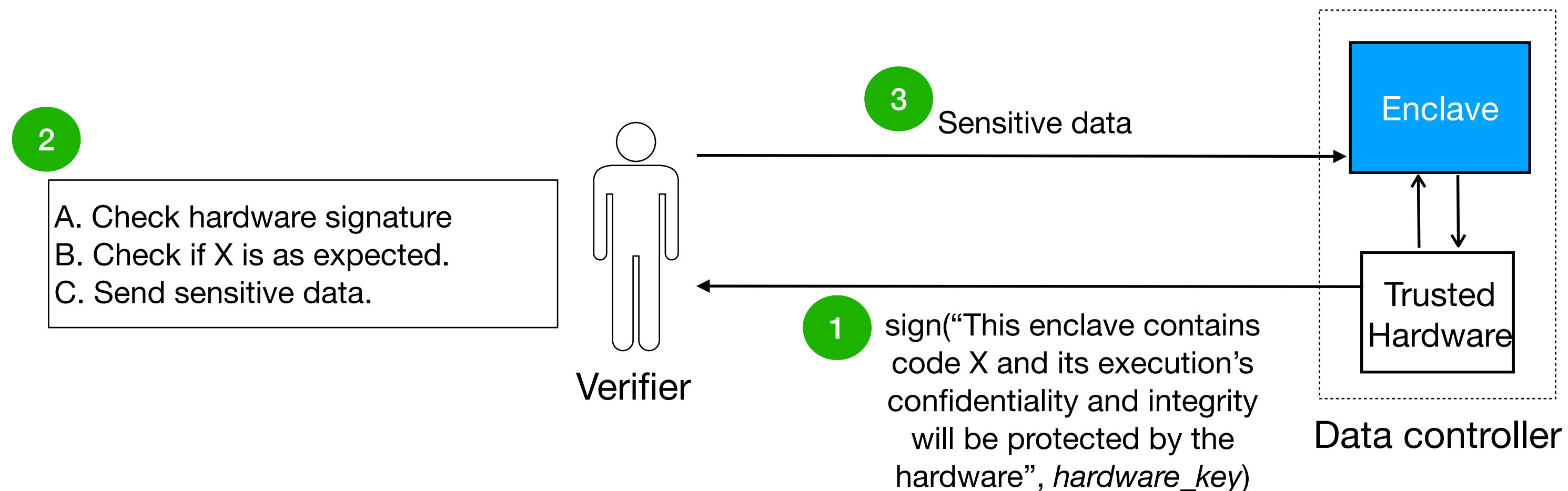
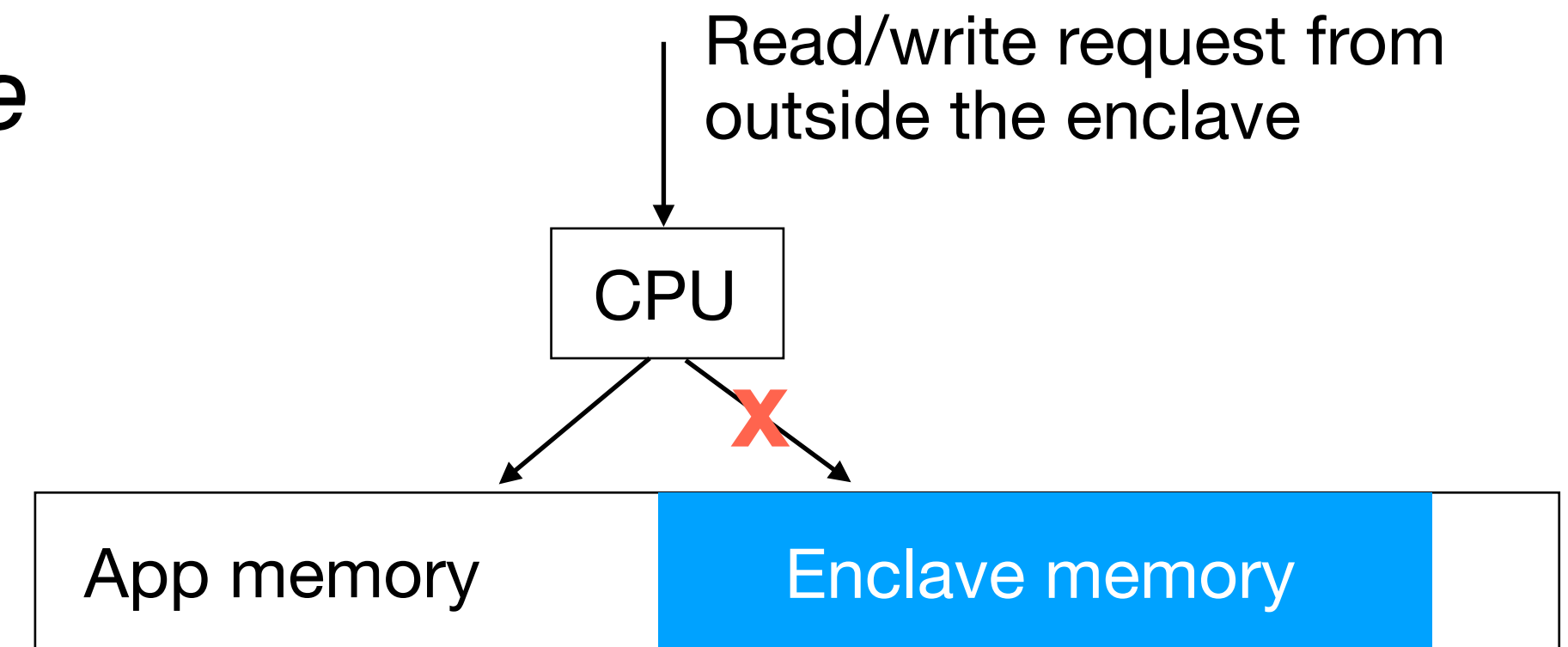
# Intel SGX

Careful about side-channel attacks though!

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# Encrypted DBs

*Practical secure remote execution requires interacting with encrypted DBs*

- **Querying encrypted databases:**
  - Searchable encryption schemes, searchable indices along with encrypted data (susceptible to reconstruction attacks)
  - EnclaveDB: entire database within an enclave
- **Private information retrieval:** only hides access patterns while DB is in cleartext (perhaps not very useful for us)

# Summary

- Privacy requires a multi-pronged approach and many techniques exist for each, but an overarching privacy architecture is missing
- Purpose limitation is a crucial privacy requirement, considering the impossibility of absolute privacy
- Almost no technique talks about an external regulator preventing illegal access and processing of data (tomorrow's talk)
- Issues related to consent and privacy self-management (tomorrow's talk)

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