Privacy

Data minimisation

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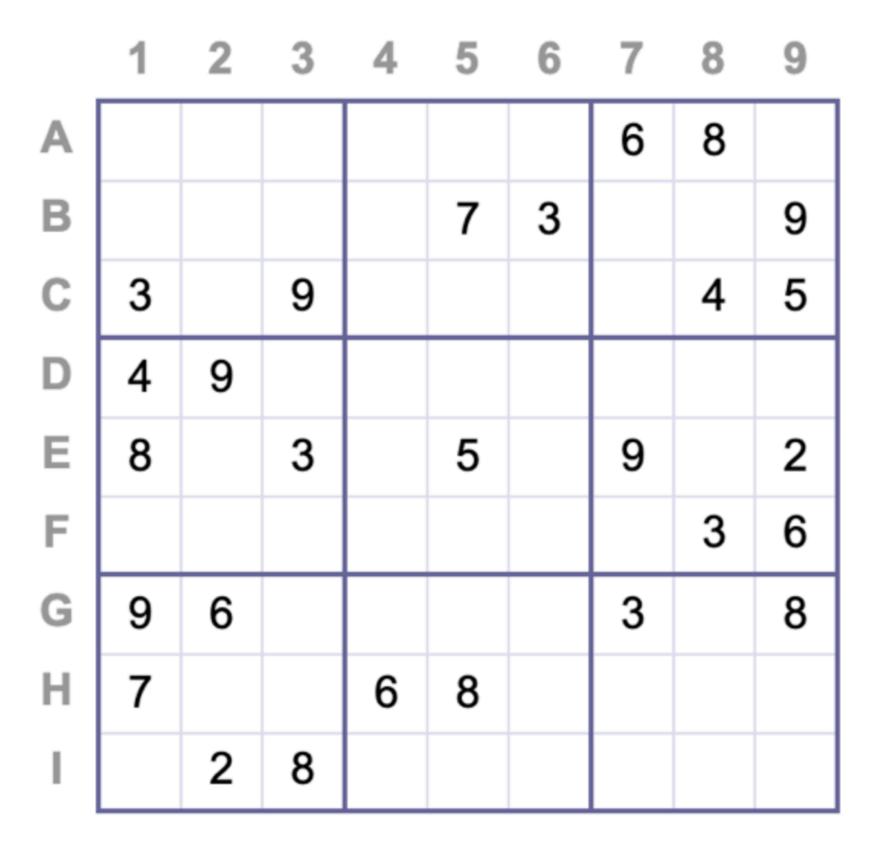
Data minimisation (DM) (aka "minimum disclosure")

- 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
- 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

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.

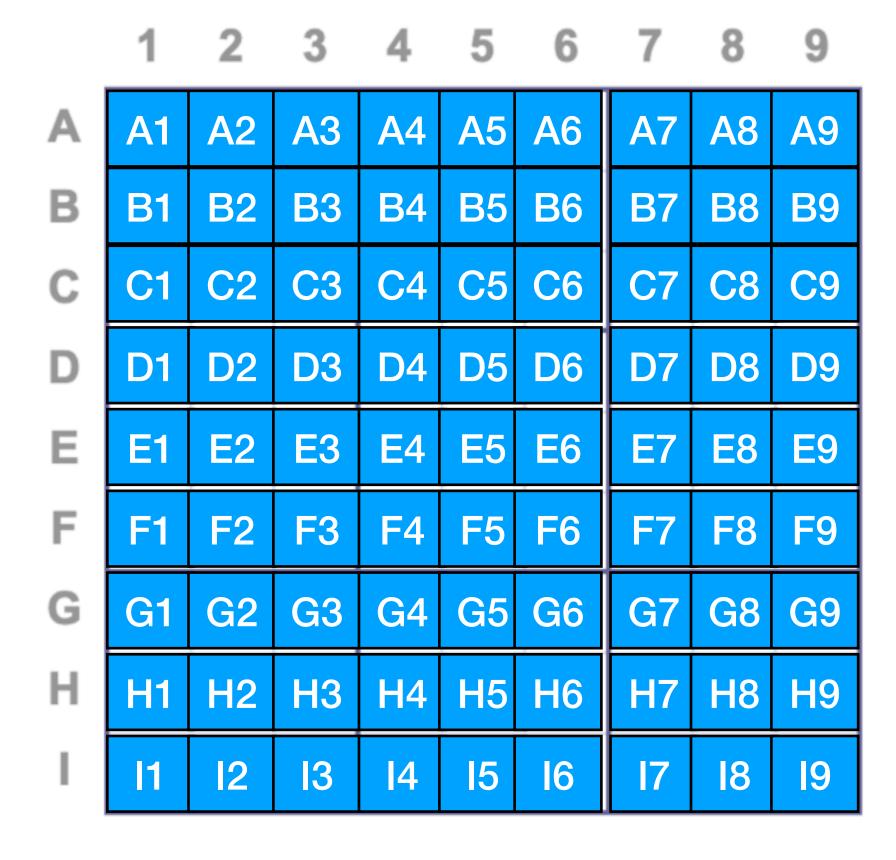


^{*} For a correct solution to a Sudoku problem, each row, column and box must contain all the numbers from 1-9

• 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.

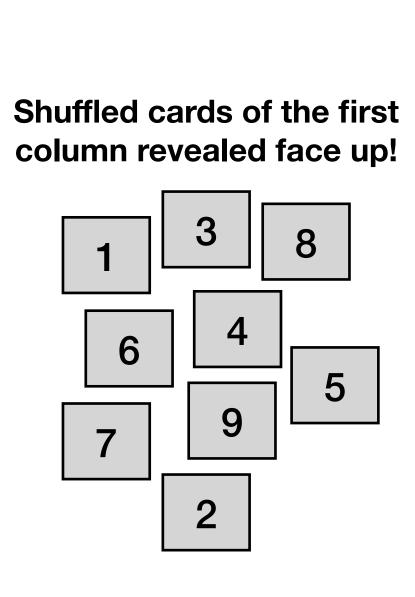


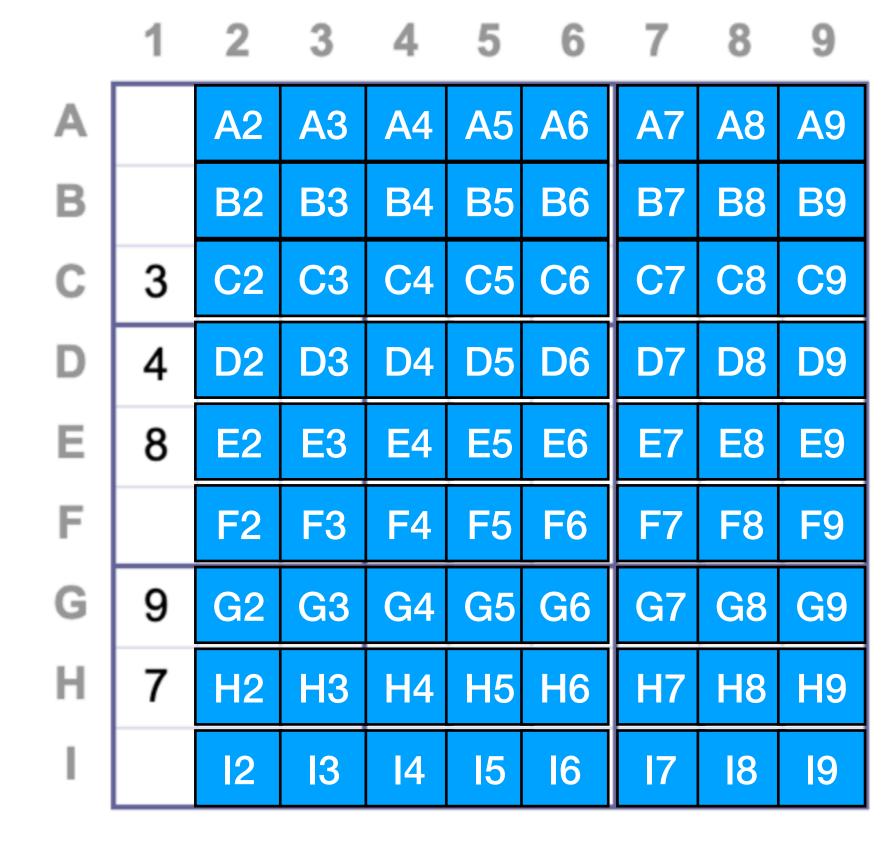
C3 Face down



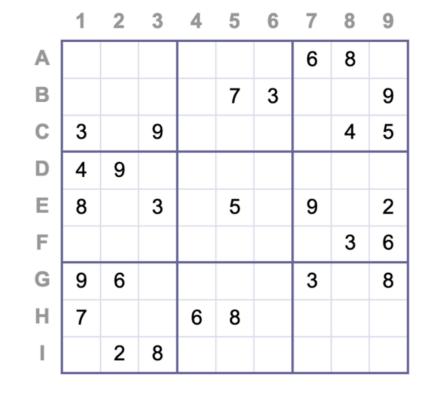
Note: Commitment is both binding and hiding!

• 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.



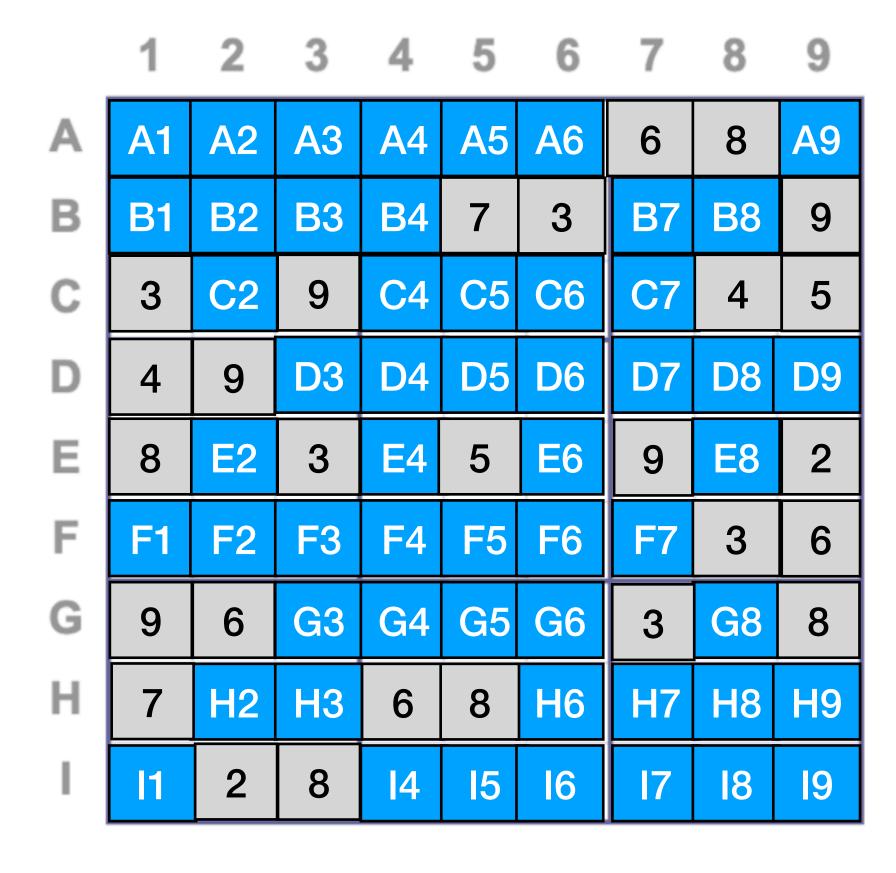


Original problem

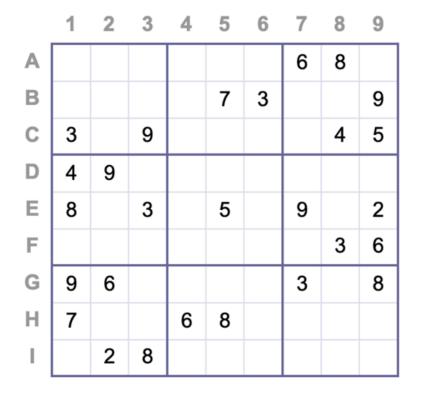


• Step 3 (**Final Reveal**): Prover reveals all non-empty cells in the original Sudoku problem to show that he has actually solved the given problem, and not some other problem.

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

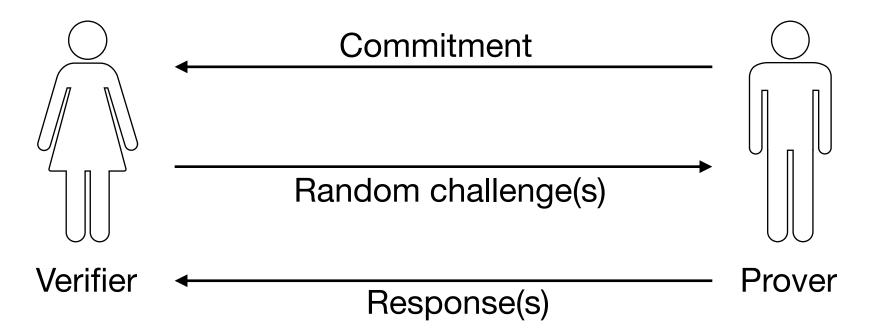


Original problem



Zero-knowledge proofs: Summary

Typical structure



- 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

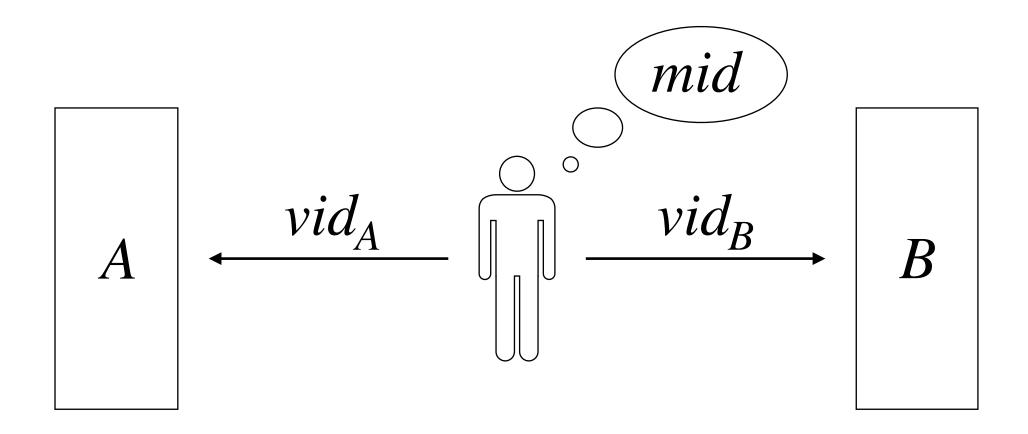
- Anonymity: The state of not being identifiable in a set of individuals
- Unlinkable anonymity: Transactions do not reveal individuals' true identities and even multiple transactions by the same individual are unlinkable

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

$$\begin{array}{c|c}
 & t_1 = f(r) \\
\hline
 & t_2 = f(r)
\end{array}$$

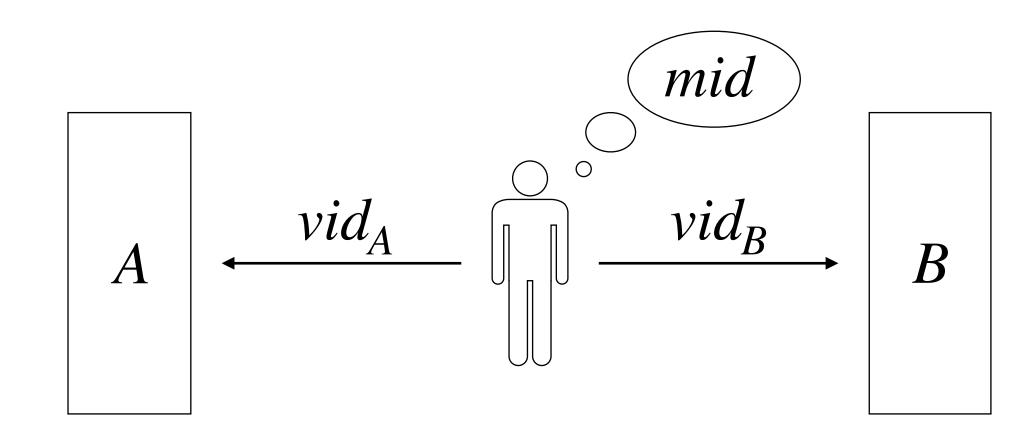
Virtual identities

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



- 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

• Anonymous credentials: Often based on blind signatures that are transformable

(Chaum '85)

$$A \qquad \overrightarrow{s_A = \sigma_A(bl(vid_A))} \qquad \overrightarrow{s_B = \sigma_A(bl(vid_B))} \qquad B$$

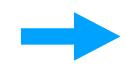
No one can present a signature on a *vid* unless they obtained a signature on another *vid* they own

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	М	5.1	58	Bldg X, DLF Phase 3, GG	No
Charles	34	М	5.9	65	Sameer Bungalow, Delhi	No
David	43	М	6.1	76	55, Sunset Blvd, Mumbai	No



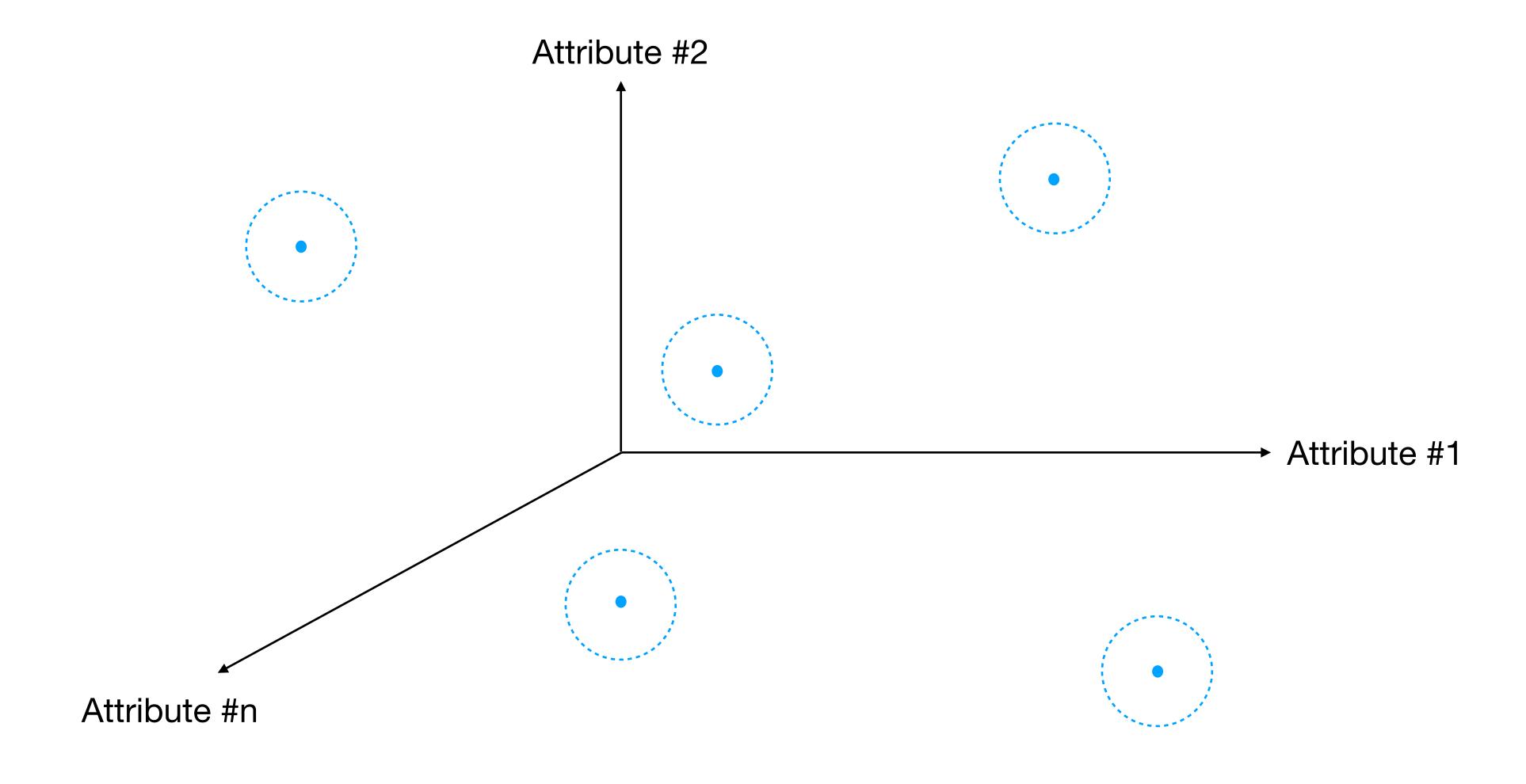
Height Weight **HIV Status** Location 15-20 5.5-6.0 80-85 Pune Yes 25-30 5.0-5.5 55-60 Gurgaon No Delhi 5.5-6.0 60-65 No 40-45 6.0-6.5 75-80 Mumbai No

Medical Database

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.

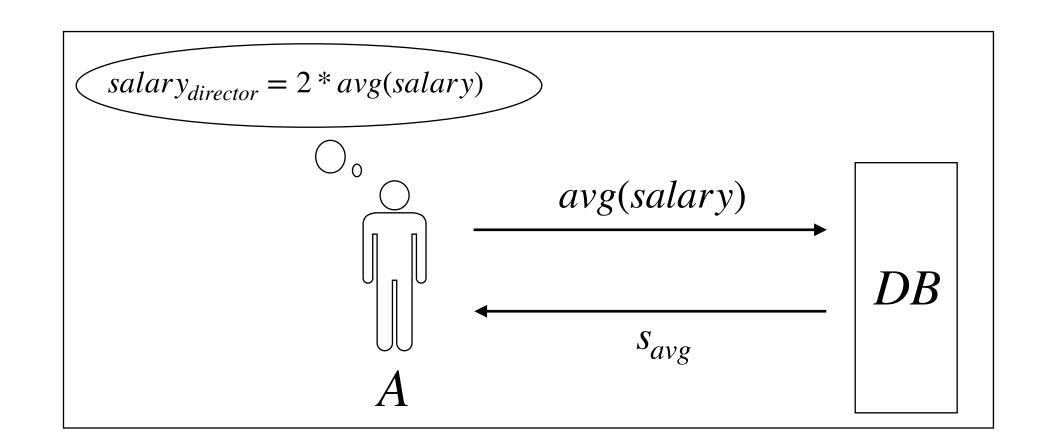
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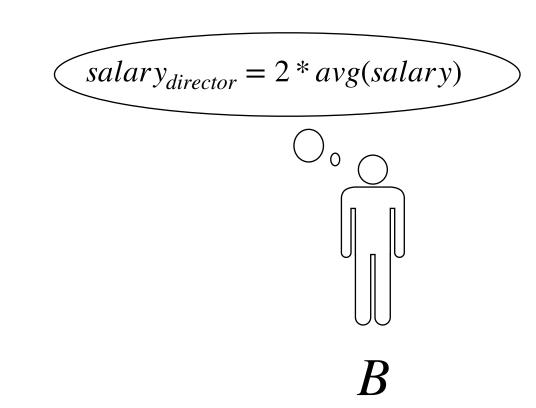
- De-anonymisation attacks on anonymised social network data, location data, writing style, source code, browser history, etc., exist. (Narayanan et al. '19)
- Theoretical bottlenecks: Given a database with n rows, if the adversary is allowed to obtain answers to O(n) subset-sum queries, it can **reconstruct** 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



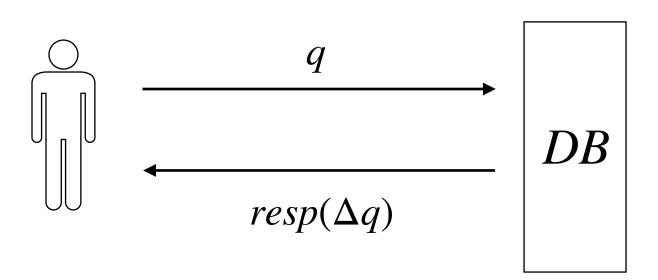


If the adversary has arbitrary side-information, above absolute privacy goal is impossible to achieve. (Dwork '05)

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 Δq measures how much the answer for q can change w.r.t. changes in only one row

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