

On the Relation Between Differential Privacy, Identifiability and Mutual-Information Privacy

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DIFFERENTIAL PRIVACY

- ◆ A mechanism satisfies ϵ -differential privacy if for any neighboring x, x' and any y

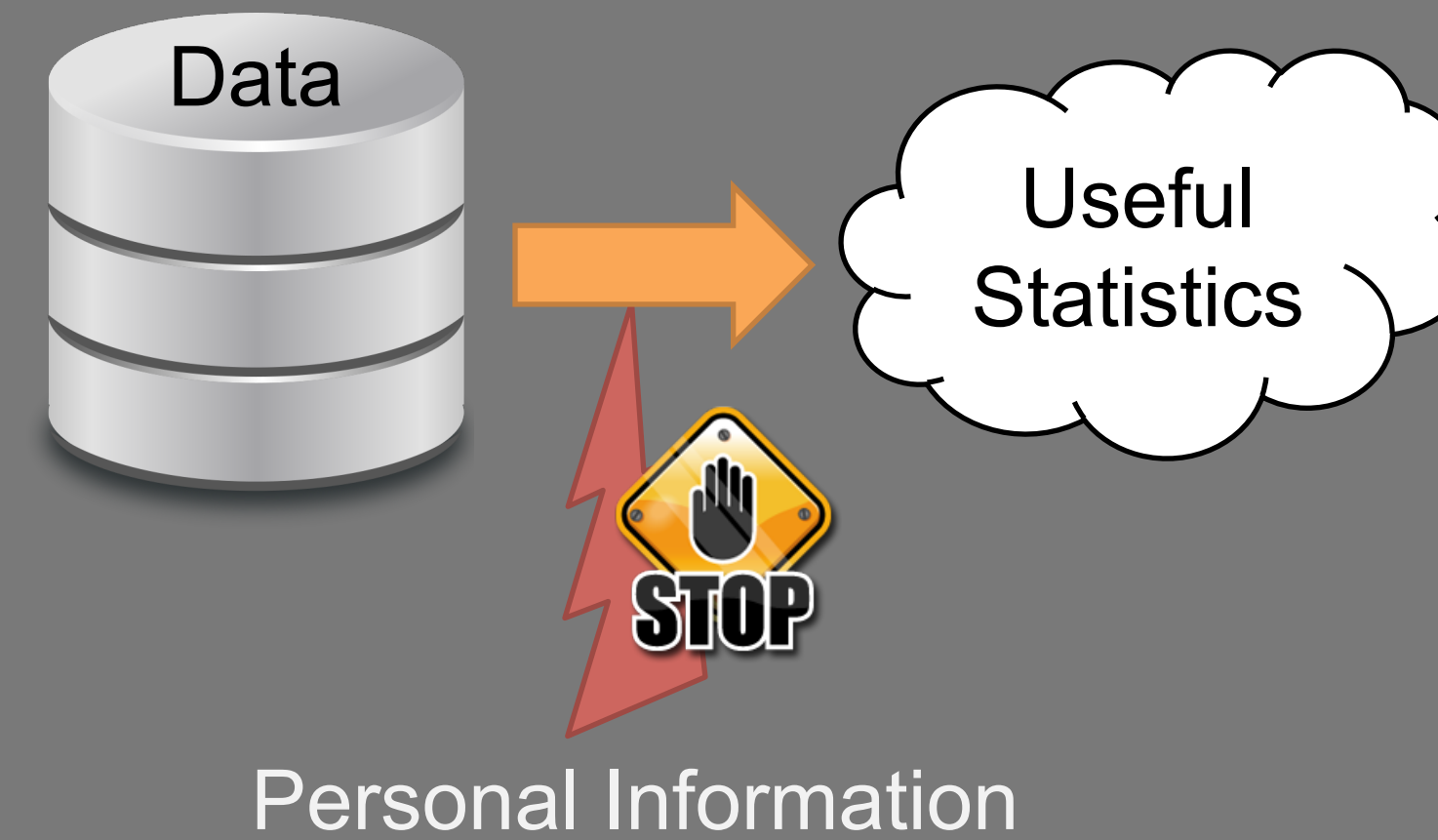
$$p_{Y|X}(y | x) \leq e^\epsilon p_{Y|X}(y | x').$$

- Indistinguishability between pairwise likelihoods.
- Limited additional information leakage.

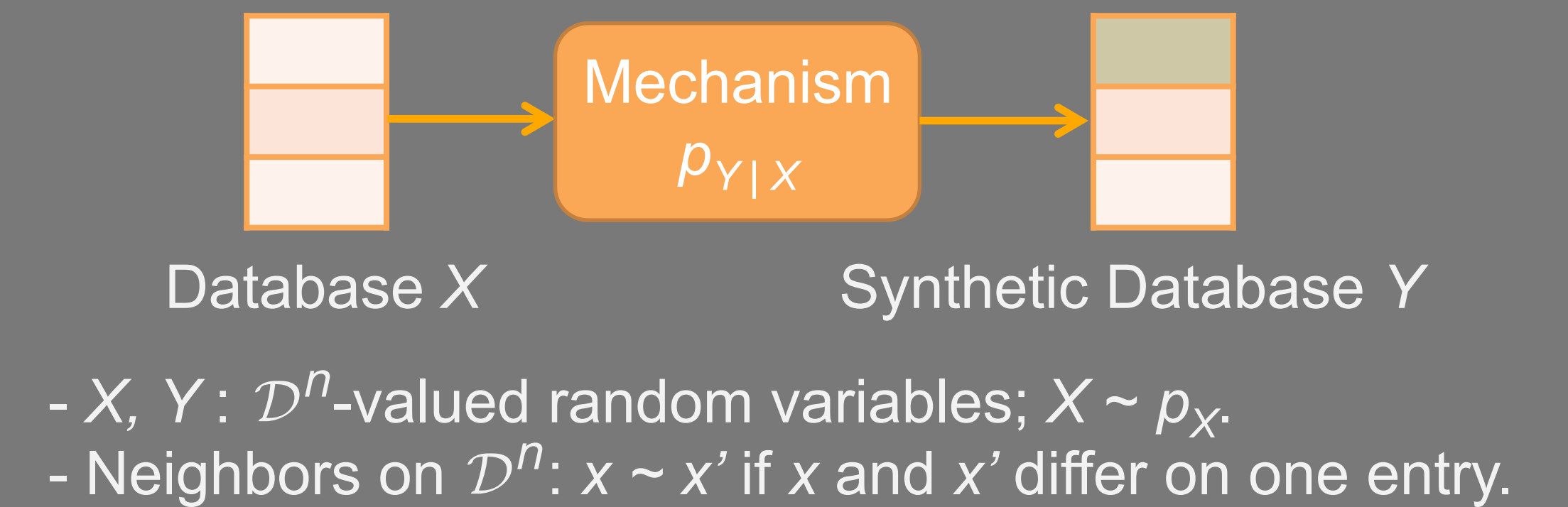
- ◆ The privacy–distortion problem under differential privacy (PD-DP):

$$\begin{aligned} & \min_{p_{Y|X}} \sum_{x \in \mathcal{D}^n} \sum_{y \in \mathcal{D}^n} p_X(x) p_{Y|X}(y | x) d(x, y) \\ & \text{subject to } p_{Y|X}(y | x) \leq e^{\epsilon_d} p_{Y|X}(y | x'), \\ & \quad \forall x, x' \in \mathcal{D}^n: x \sim x', \forall y \in \mathcal{D}^n, \\ & \quad p_{Y|X} \text{ is valid.} \end{aligned}$$

PRIVACY-PRESERVING DATA ANALYSIS



SYNTHETIC DATABASE RELEASING



Privacy–Distortion Tradeoff

$$\epsilon^*(D) = \inf \{ \epsilon : \epsilon\text{-privacy level is achievable with } \mathbb{E}[d(X, Y)] \leq D \},$$

where d is the Hamming distance.

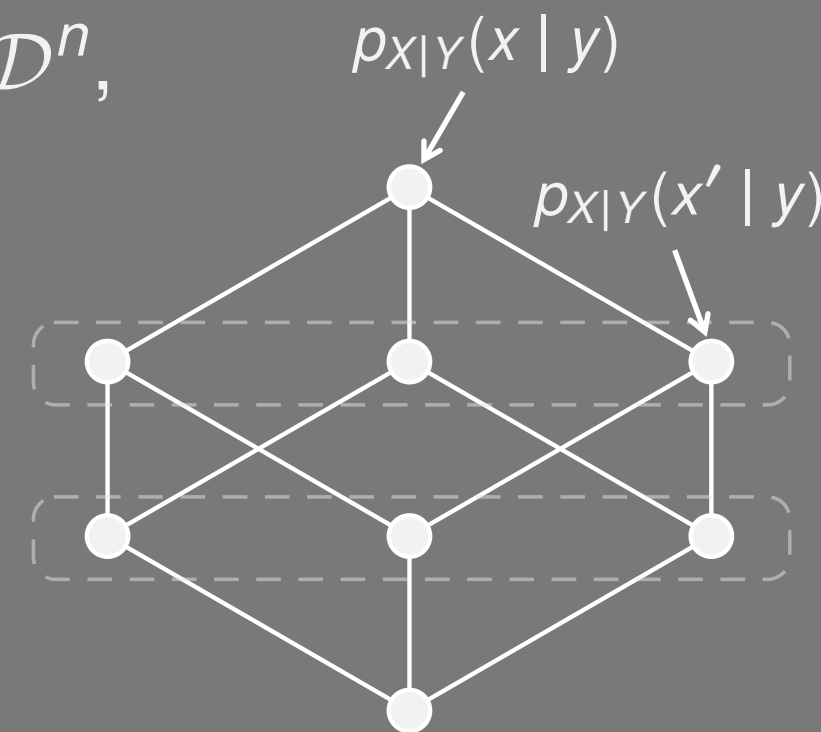
Relaxed to the same problem:

$$\begin{aligned} & \min_{p_{X|Y}, p_Y} \sum_{x \in \mathcal{D}^n} \sum_{y \in \mathcal{D}^n} p_Y(y) p_{X|Y}(x | y) d(x, y) \\ & \text{subject to } p_{X|Y}(x | y) \leq e^\epsilon p_{X|Y}(x' | y), \\ & \quad \forall x, x' \in \mathcal{D}^n: x \sim x', \forall y \in \mathcal{D}^n, \\ & \quad p_{X|Y}, p_Y \text{ are valid.} \end{aligned}$$

Optimal Solution:

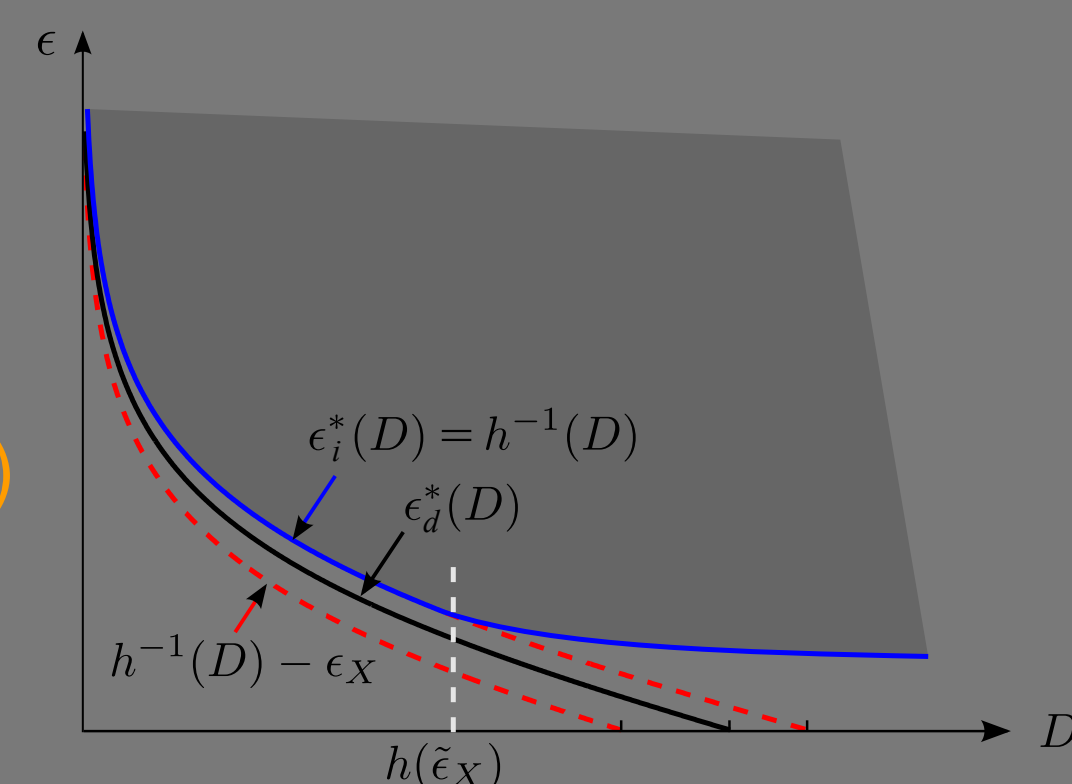
$$p_{X|Y}(x | y) = \frac{e^{-\epsilon d(x, y)}}{(1 + (|\mathcal{D}| - 1)e^{-\epsilon})^n}$$

$$D^*(\epsilon) = h(\epsilon) \triangleq \frac{n}{1 + \frac{e^\epsilon}{|\mathcal{D}| - 1}}$$



Main Result 1:

$$\epsilon_i^*(D) - \epsilon_X \leq \epsilon_d^*(D) \leq \epsilon_i^*(D)$$



IDENTIFIABILITY

- ◆ A mechanism satisfies ϵ -identifiability if for any neighboring x, x' and any y

$$p_{X|Y}(x | y) \leq e^\epsilon p_{X|Y}(x' | y).$$

- Indistinguishability between pairwise posteriors.
- Absolute guarantee.

- ◆ The privacy–distortion problem under identifiability (PD-I):

$$\begin{aligned} & \min_{p_{X|Y}, p_Y} \sum_{x \in \mathcal{D}^n} \sum_{y \in \mathcal{D}^n} p_Y(y) p_{X|Y}(x | y) d(x, y) \\ & \text{subject to } p_{X|Y}(x | y) \leq e^{\epsilon_i} p_{X|Y}(x' | y), \\ & \quad \forall x, x' \in \mathcal{D}^n: x \sim x', \forall y \in \mathcal{D}^n, \\ & \quad p_{X|Y}, p_Y \text{ are valid, consistent with } p_X. \end{aligned}$$

MUTUAL-INFORMATION PRIVACY

- ◆ A mechanism satisfies ϵ -mutual-information privacy if $I(X; Y) \leq \epsilon$.

- Average guarantee.

- ◆ The privacy–distortion problem under mutual-information privacy (PD-MIP):

$$\begin{aligned} & \min_{p_{Y|X}} I(X; Y) \\ & \text{subject to } \sum_{x \in \mathcal{D}^n} \sum_{y \in \mathcal{D}^n} p_X(x) p_{Y|X}(y | x) d(x, y) \leq D, \\ & \quad p_{Y|X} \text{ is valid.} \end{aligned}$$

□ Rate–distortion function.

Main Result 2:

Given a distortion requirement, the optimal mechanisms are the same.

