

# Lecture 10

## Beyond Global Sensitivity Local Sensitivity

$$\Delta_{GS} = \max_{x, x'} \|f(x) - f(x')\|$$

$$f: X^n \rightarrow \mathbb{R}$$

$$\Delta_{LS}^{(f)}(x) = \max_{x' \text{ nbrs to } x} |f(x) - f(x')|$$

Before:  $f(x) + \text{Lap}\left(\frac{\Delta_{GS}}{\epsilon}\right) \checkmark$  (2,0)-DP

Now:  $f(x) + \text{Lap}\left(\frac{\Delta_{LS}(x)}{\epsilon}\right) X$

$f(x)$ : dist btw 2 closest points in  $X$   $X \in \mathbb{R}^n$

$$\Delta_{GS} = \infty \quad X = \{0, 0\}, \quad X' = \{0, \infty\}$$

$$X = \{0, 0, 0\} \quad X' = \{0, 0, \infty\}$$

$$\Delta_{LS}(x) = 0$$

$$f(x) + \text{Lap}(0) = f(x) \Rightarrow 0 \text{ deterministically}$$

$$X' = \{0, 0, 10000\}$$

$$\Delta_{LS}(x') = 10000$$

$$f(x') = \text{Lap}\left(\frac{10000}{\epsilon}\right) = \text{Lap}\left(\frac{10000}{\epsilon}\right) \quad \downarrow f(x')$$

# Propose-Test-Release Dwork - Lei '09

1. Propose: bound on LS
2. Test: Is it?
3. Release: If yes,  $f(x) + \text{Lap}(\frac{LS}{\epsilon})$

1. Propose a bound  $\beta$  on LS at  $X$
2. Compute  $\gamma = \min_x d(x, \tilde{x})$ , where  $\Delta_{LS}(\tilde{x}) \geq \beta$ 
  - Compute  $LS: n/|X|$
  - $\leftarrow |X| \leftarrow \text{slow}$
  - Hamming dist, # of pts to change
3.  $\hat{\gamma} = \gamma + \text{Lap}(1/\epsilon)$
4. If  $\hat{\gamma} \leq \ln(1/\delta)/\epsilon$ , return  $\perp$
5. If  $\hat{\gamma} > \ln(1/\delta)/\epsilon$ , return  $f(x) + \text{Lap}(\beta/\epsilon)$ .

Theorem: PTR is  $(2\epsilon, \delta)$ -DP

Proof:  $\perp$

$$\Pr[M(x) = \perp] \in [e^{-\epsilon}, e^{\epsilon}] \cdot \Pr[M(x') = \perp]$$

Case 1:  $\Delta_{LS}(x) \geq \beta$

$$\gamma = 0. \Pr[\hat{\gamma} > \ln(1/\delta)/\epsilon] \leq \delta$$

$$\Rightarrow \Pr[M(x) \neq \perp] \leq \delta$$

$$T \subseteq \mathbb{R} \cup \{\perp\}$$

$$\begin{aligned} \Pr[M(x) \in T] &= \Pr[M(x) \in T \cap \{\perp\}] + \Pr[M(x) \in T \cap \mathbb{R}] \\ &\leq e^{\epsilon} \Pr[M(x') \in T \cap \{\perp\}] + \Pr[M(x) \neq \perp] \\ &\leq e^{\epsilon} \Pr[M(x') \in T] + \delta \end{aligned}$$

$(\epsilon, \delta)$ -DP

Case 2:  $\Delta_{LS}(x) \leq \beta$

1. Release  $\hat{y}$ :  $\epsilon$ -DP
2. Release  $f(x) + \text{Lap}(\beta/\epsilon)$ :  $\epsilon$ -DP

$(2\epsilon, 0)$ -DP

Application: Histograms (modal element)

Laplace Histogram:  $L_\infty$  error  $\approx \log |X|$  error  $\epsilon$ -DP

Relax to approx  $\rightarrow$  no dependence on  $|X|$ .

Problem:  $X \in \mathcal{X}^n$ . Compute most freq elt.

$X = \{v_1, v_2, \dots, v_n\}$ , How many pts of  $x$  to change before mode changes?

$\approx \frac{1}{2} (\text{count of most freq elt} - \text{count of 2nd most})$

- $(\epsilon, \delta)$ -DP
1. Propose  $LS \leq 0$ .
  2.  $\hat{y} =$
  3.  $\hat{y} = \hat{y} + \text{Lap}(1/\epsilon)$ .
  4. If  $\hat{y} \leq \log(1/\delta)/\epsilon$ , 1.
  5. If  $\hat{y} \geq \log(1/\delta)/\epsilon$ , output  $\text{mode}(x)$ .
- $\Pr[\text{Lap}(1/\epsilon) > \ln(1/\delta)/\epsilon] \leq \delta$ . Want:  $\gamma \geq 2 \log(1/\delta)/\epsilon$ .
- Need: diff #1 and #2 elt  $\geq 4 \ln(1/\delta)/\epsilon$ .

**Theorem 3.** There exists an  $(\epsilon, \delta)$ -differentially private algorithm which identifies the most frequent element from an arbitrary dataset with probability at least  $1 - \delta$ , as long as the gap between the count of the most frequent and the second most frequent element is at least  $4 \ln(1/\delta)/\epsilon$ .

Thm 3.5 Vadhan '17

**Theorem 4.** There exists an  $(\epsilon, \delta)$ -differentially private algorithm which can, with high probability, output the count of every item in a dataset up to additive  $O(\log(1/\delta)/\epsilon)$ .

$\epsilon$ -DP  $O(\log |X| / \epsilon)$

## Privately Bounding Local Sensitivity

Before: Guessed LS bound

Now: Est LS.  $n(|X|-1)$

1. Compute  $\hat{\gamma} = \Delta_{LS}^{(f)}(X) + \text{Lap}\left(\frac{\Delta_{GS}^{(f)}}{\epsilon}\right) + \ln(1/\delta)/\epsilon$

2. Output  $f(x) + \text{Lap}(\hat{\gamma}/\epsilon)$ .

# Smooth Sensitivity Nissim Raskhodnikova Smith

$$\Delta_{SS}^{(f)}(X) = \max_{\tilde{X} \in X^n} \left\{ \Delta_{LS}^{(f)}(\tilde{X}) e^{-\epsilon d(x, \tilde{X})} \right\}$$

↑ Hamming.

~~$f(x) + \text{Lap}\left(\frac{\Delta_{SS}(x)}{\epsilon}\right)$~~       $f(x) + \text{Cauchy}\left(\frac{\Delta_{SS}(x)}{\epsilon}\right)$

↑  $(\epsilon, \delta)$ -DP

←  $\epsilon$ -DP

