

Propose Test Release

Today

- Gaussian Mean Estimation
- Propose-Test-Release
- Gaussian Mean Estimation w/PTR

Task: given $x_1, \dots, x_n \stackrel{iid}{\sim} \mathcal{N}(\mu, \Sigma)$, estimate μ
with a DP algorithm (as always, privacy is worst-case)

What should be our target error?

Need to hide difference between

$$\left(\frac{1}{n} \sum_{i=1}^n x_i \right) - \left(\frac{1}{n} \sum_{j=2}^{n+1} x_j \right) = \frac{1}{n} (x_1 - x_{n+1}) \sim \mathcal{N}\left(0, \frac{\Sigma}{n^2}\right)$$

Need Gaussian noise like $\mathcal{N}\left(0, \sigma_{\epsilon}^2 \frac{d}{n^2} \Sigma\right)$

Then might have:

$$\begin{aligned}\|u - \tilde{u}\|_2 &\leq \|u - \hat{u}\|_2 + \|\hat{u} - \tilde{u}\|_2 \\ &\leq \|\mathcal{N}(0, \frac{1}{n}\mathbb{I})\|_2 + \|\mathcal{N}(0, \sigma_{\epsilon, \delta}^2 \frac{d}{n^2}\mathbb{I})\|_2 \\ &= \frac{1}{\sqrt{n}} \|\mathcal{N}(0, \mathbb{I})\|_2 + \frac{\sigma_{\epsilon, \delta} \sqrt{d}}{n} \|\mathcal{N}(0, \mathbb{I})\|_2 \\ &\leq \sqrt{\frac{d}{n}} + \sigma_{\epsilon, \delta} \frac{d}{n} \\ &\quad \hookrightarrow \approx \frac{\sqrt{\log \frac{1}{\delta}}}{\epsilon}\end{aligned}$$

Happy: privacy noise $\rightarrow 0$ faster

Issue: global sensitivity of mean = $+\infty$

"Textbook Solution" Assume ^{or force} all i have $\|x_i\|_2 \leq R$.

$$\text{Then } \Delta = \max_{x, x'} \|\hat{u}_x - \hat{u}_{x'}\|_2 \leq \frac{2R}{n} \quad x_i \in \mathcal{B}(0, R)$$

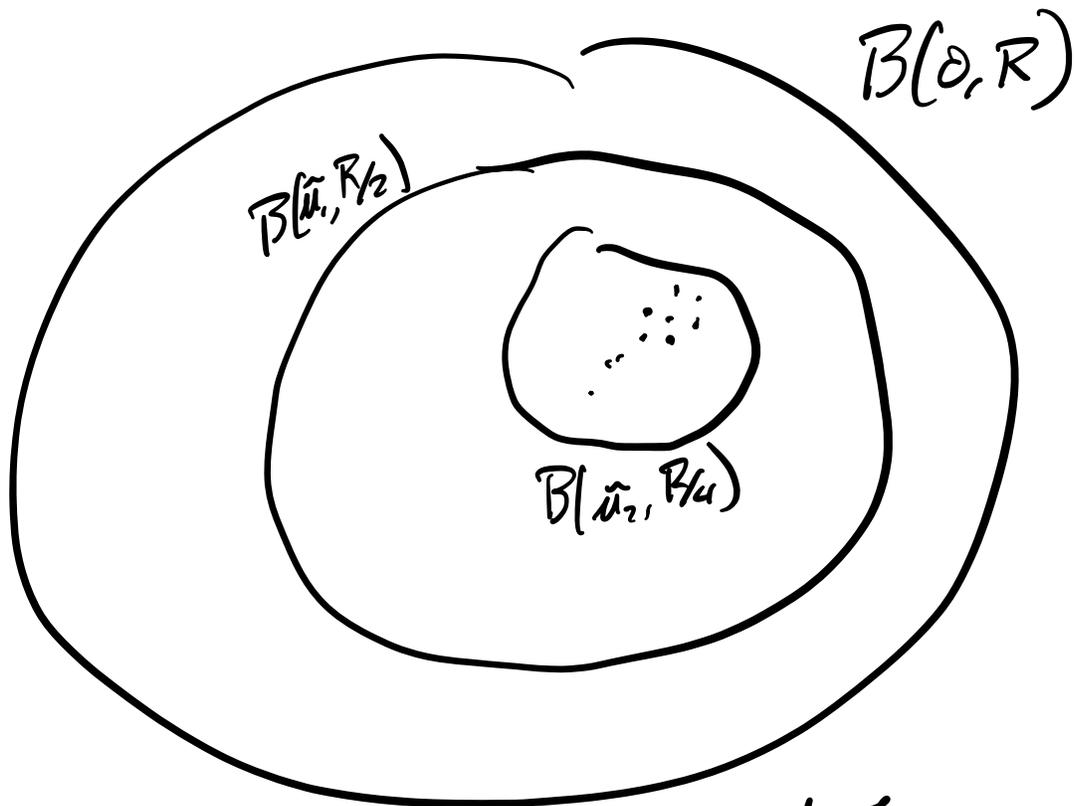
If all data live in $\mathcal{B}(0, R)$, then

$$\text{release } \hat{u} + \mathcal{N}(0, \frac{R^2}{n^2} \sigma_{\epsilon, \delta}^2 \mathbb{I})$$

Error like $\|u - \tilde{x}\|_2 \leq \sqrt{\frac{d}{n}} + \frac{\sigma_{\varepsilon, \delta} R \sqrt{d}}{n}$

What if R is really big?

KLSO17 show how to iterate:



error like $\|u - \tilde{u}\|_2 \leq \sqrt{\frac{d}{n}} + \frac{\sigma_{\varepsilon, \delta} \sqrt{105} R^2 d}{n}$

What if R is really big? $R = \infty$?

PTR.

Algorithm 1

Propose-Test-Release

Input: $\epsilon, \delta \in (0, 1)$, $x \in \mathcal{X}^n$, $A: \mathcal{X}^n \rightarrow \mathcal{Y}$,
"safety oracle" $\mathcal{O}_{\epsilon, \delta}^A: \mathcal{X}^n \rightarrow \mathbb{N}$

Output: $\tilde{y} \in \mathcal{Y}$ or \perp

1) $k \leftarrow \mathcal{O}_{\epsilon, \delta}^A(x)$

2) $\tilde{k} \leftarrow k + \text{Lap}(1/\epsilon)$

3) If $\tilde{k} \geq \frac{\log(1/\delta)}{\epsilon}$; release $\tilde{y} \sim A(x)$

4) Else; release \perp

Assumption 1 Fix A, ϵ, δ . Let

$$\text{SAFE} = \text{SAFE}_{\epsilon, \delta}^A = \left\{ x \in \mathcal{X}^n : \forall x' \sim \pi, A(x) \approx_{(\epsilon, \delta)} A(x') \right\}.$$

There exists $S \subseteq \text{SAFE}$ such that, for all $x \in \mathcal{X}^n$,

$$\mathcal{O}_{\epsilon, \delta}^A(x) = \min_{x' \notin S} d(x, x')$$

Draw picture

Claim If Assumption 1 holds, then Algorithm 1

is $(2\epsilon, e^2\delta)$ -DP.

Proof Fix x, x' and output $E \subseteq \mathcal{Y} \cup \{\perp\}$.
Write Alg 1 as $B(x)$.

Steps!

1) $\Pr[B(x) = \perp]$ & $\Pr[B(x) \neq \perp]$ are e^ϵ ,
b/c Laplace & low-sensitivity

2) write

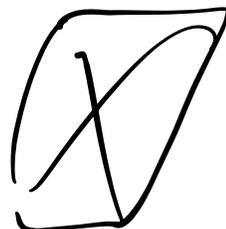
$$\Pr[B(x) \in E] = \Pr[B(x) \in E \mid B(x) \neq \perp] \Pr[B(x) \neq \perp] \\ + \Pr[B(x) \in E \mid B(x) = \perp] \Pr[B(x) = \perp]$$

3) Two cases: $x \in S \subseteq \text{SAFE}$

$x \notin S$

$$t > 0, \Pr[Z > t] = \frac{1}{2} \exp\left(-\frac{t}{b}\right)$$

$Z \sim \text{Lap}(b)$



Back to mean estimation

Instantiate Alg 1 with ... require $n \geq \frac{2 \log^{1/5}}{\epsilon}$

Alg 2

Input: $\epsilon, \delta \in (0, 1)$, $x \in \mathbb{R}^{n \times d}$

Return: $\tilde{\mu} \in \mathbb{R}^d$

- 1) $c_1 \leftarrow 3\sqrt{d}$; $c_2 \leftarrow c_2(\epsilon, \delta, d, n)$
- 2) $\mu_0 \leftarrow \underset{\mu}{\operatorname{argmax}} \# \{i \in [n] : x_i \in B(\mu, c_1)\}$
- 3) Project all n points to $B(\mu_0, c_1)$
- 4) Release $\tilde{\mu} \leftarrow \mu_x + \mathcal{N}(0, c_2 \mathbb{I})$

set this later

Safety Oracle | Define $S \subseteq \mathcal{X}^n$ as:

$$x \in S \text{ iff } \exists \mu_0 \text{ s.t. } \# \{i \in [n] : x_i \in B(\mu_0, c_1)\} \geq n - \frac{\log^{1/5}}{\epsilon}$$

$$\text{Let } m(x) = \max_{\mu} \# \{i \in [n] : x_i \in B(\mu, c_1)\}$$

$$\mathcal{O}_{\epsilon, \delta}^A(x) = \begin{cases} 0 & \text{if } m(x) \geq n - \frac{\log^{1/5}}{\epsilon} \\ n - \frac{\log^{1/5}}{\epsilon} - m(x) & \text{otherwise.} \end{cases}$$

What do we need to prove?

1) Does this oracle compute the distance to some fixed set S ?

2) Is this $S \subseteq \text{SAFE}_{\varepsilon, \delta}^A$?

3) Accuracy: when $x \stackrel{\text{iid}}{\sim} \mathcal{N}(\mu, \Pi)$, whp have $x \in S$, pass the check whp, and return $\hat{\mu} + \mathcal{N}(0, c_2^2 \Pi)$

$$c_2 = c_2(\varepsilon, \delta, d, n) \approx \frac{\sqrt{d} (\log 1/\delta)^{3/2}}{\varepsilon^2 n}$$