

Subsampling Suffices for

Adaptive Data Analysis

Focus today

Using Samples to understand the Population

What pets do people like?



$$p(\text{dog}) = 5/6$$

$$p(\text{cat}) = 5/6$$

$$p(\text{turtle}) = 3/6$$

$$p(\text{bird}) = 3/6$$

Using Samples to understand the Population

Fact: With a sample of size

$$n \geq \Omega\left(\frac{\log q}{\varepsilon^2}\right)$$

q -many $p()$ queries will be within $\pm\varepsilon$ of their values in the overall population.

Proof: Union bound over q queries each with failure probability $\ll 1/q$.

For a single query, value of p_{sample} is mean of n independent $\text{Ber}(p_{\text{population}})$. By Hoeffding's inequality,

$$\Pr\left[|p_{\text{sample}} - p_{\text{population}}| \geq \varepsilon\right] \leq 2e^{-2\varepsilon^2 n}$$

$$p(\text{dog}) = 5/6 \pm \varepsilon$$

$$p(\text{cat}) = 5/6 \pm \varepsilon$$

$$p(\text{turtle}) = 3/6 \pm \varepsilon$$

$$p(\text{bird}) = 3/6 \pm \varepsilon$$

Using Samples to understand the Population

Other examples:

1. What fraction of patients on medication X experience remission?
2. What fraction of people will vote for candidate Y?
3. What fraction of concepts does a student understand?

Adaptive data analysis

What pets do people like?



$$p(\text{dog}) = 5/6$$

$$p(\text{cat}) = 5/6$$

$$\text{dog} + \text{cat} ?$$

$$p_{\text{sample}}(\text{dog}, \text{cat}) = 4/6$$

$p_{\text{population}}$ close?

Adaptive Data Analysis

How can we guarantee results are representative of the population even when the queries are chosen adaptively?

Proposed by [Dwork, Feldman, Hardt, Pitassi, Reingold, and Roth 15]

Why adaptive data analysis is hard

Non-adaptive: Sample of size $n \geq \Omega\left(\frac{\log q}{\varepsilon^2}\right)$ suffices

Adaptive counterexample: Population distribution,

$$\mathcal{D} := \text{Uniform}(\{1, 2, \dots, 2n\})$$

Given sample $\mathcal{S} \sim \mathcal{D}^n$, for each $i \in \{1, 2, \dots, 2n\}$, ask query:

$$y_i = p_{\mathcal{S}}(x \mapsto \mathbf{1}[x = i])$$

Track $T := \{i \text{ where } y_i > 0\}$

After receiving response, ask query $x \mapsto \mathbf{1}[x \in T]$.

1. $p_{\mathcal{S}}(x \mapsto \mathbf{1}[x = i]) = 1$
2. $p_{\mathcal{D}}(x \mapsto \mathbf{1}[x = i]) \leq \frac{1}{2}$

Adaptive: With $q = 2n + 1$, can force error $\varepsilon \geq 1/2$.

Adaptive Data Analysis

How can we guarantee results are representative of the population even when the queries are chosen adaptively?

Proposed by [Dwork, Feldman, Hardt, Pitassi, Reingold, and Roth 15]

Any ideas?

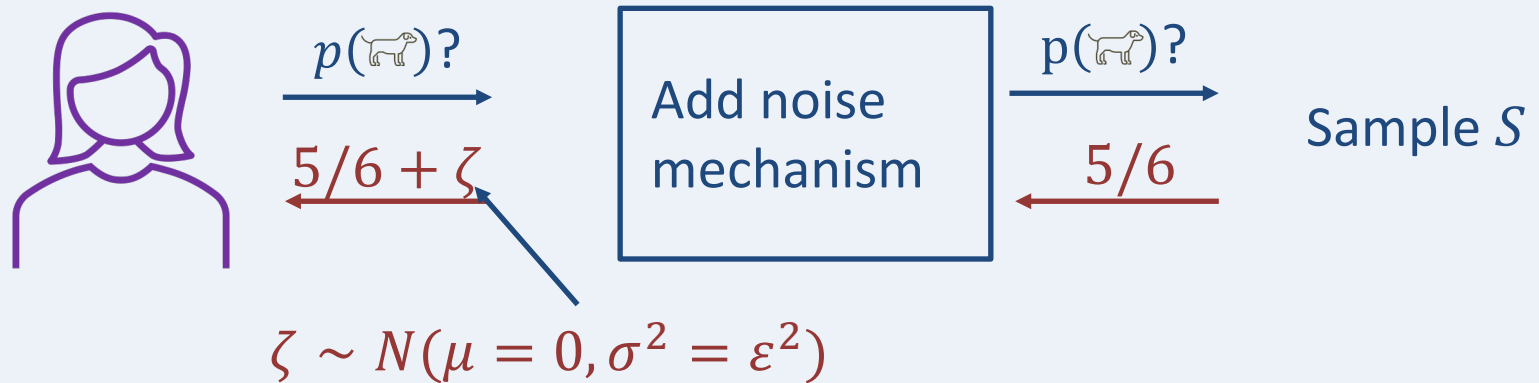
A simple solution

Take a fresh batch of $\approx 1/\varepsilon^2$ samples for each query.

Requires sample size of

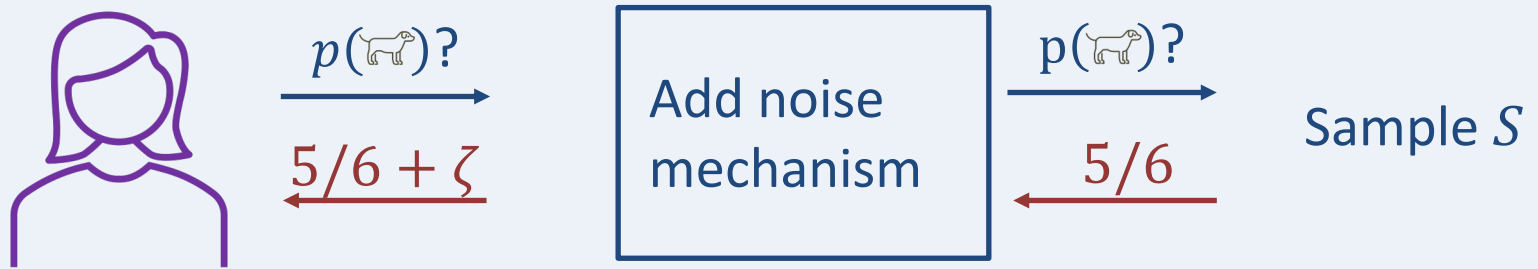
$$n \approx \frac{q}{\varepsilon^2}.$$

A better mechanism



Needs only $n = \tilde{O}\left(\frac{\sqrt{q}}{\varepsilon^2}\right)$ samples to answer q queries
[DFHPR15, BNSSSU16].

Why is adding noise good?



Intuition: To ask a bad query,  must have lots of information about S .

Adding noise “hides” information about S . Formally, it ensures the query responses are **differentially private**.

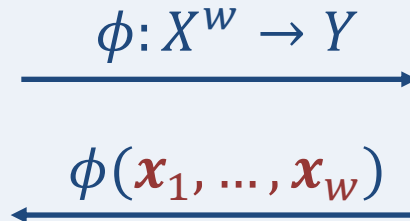
Very cool active area of research on how to quantify private algorithms.

My research question

What minimal assumptions can we make about the queries to guarantee the results generalize, even **without an explicit mechanism?**

My solution: Sufficient for each query to take as input a **random subsample** and **outputs few bits.**

Subsampling queries



Sample $S \in X^n$

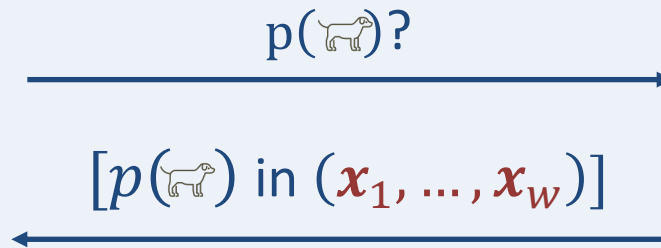
$\mathbf{x}_1, \dots, \mathbf{x}_w$ chosen uniformly without replacement from S

Theorem (informal): If each $|Y|$ is small, results will be representative for q queries as long as the sample size satisfies

$$n \geq \Omega(w\sqrt{q}).$$

Compare to $n \geq wq$ required if we use a separate sample for each query.

Example application #1: Fraction queries

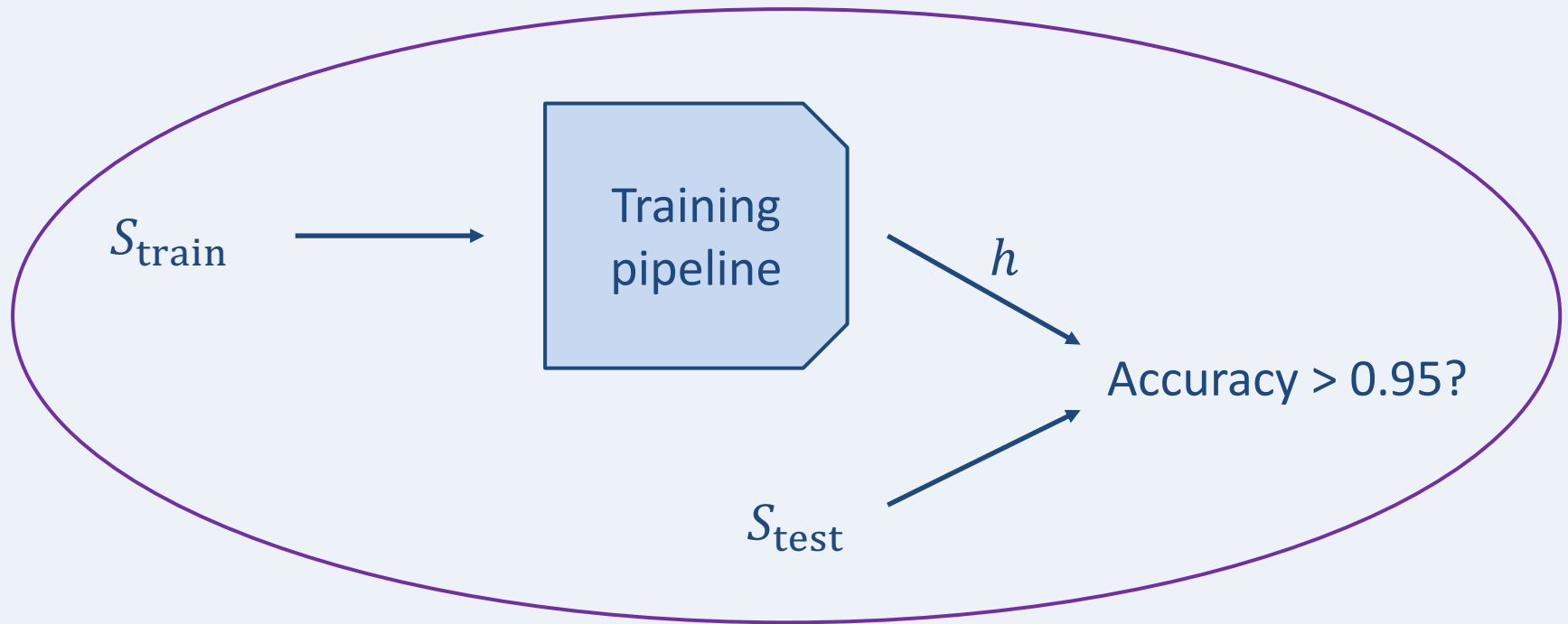


Sample $S \in X^n$

$\mathbf{x}_i \sim \text{Unif}(S)$

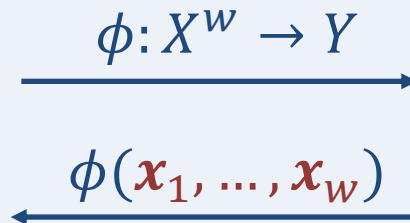
This simple mechanism gives state of the art sample size-accuracy tradeoff.

Example application #2: Is a training pipeline accurate?



All one subsampling query with $w = |S_{\text{train}}| + |S_{\text{test}}|$ and $Y = \{0,1\}$

Questions?



Sample $S \in X^n$

$\mathbf{x}_1, \dots, \mathbf{x}_w$ chosen uniformly without replacement from S

Theorem (informal): If each $|Y|$ is small, results will be representative for q queries if

$$n \geq \Omega(w\sqrt{q}).$$