Class starts after this song

Kygo – Stranger Things (2017) requested by Ahbab Abeer (TA-of-CM8)

Hello, I am an avid clash royale player and love terraria. I failed my elementary school entrance exam. My favorite music genre is EDM



Duke

CS230 Spring 2024 EM B: Probability Applications in Privacy



Poll (Not a PI)

- Disclaimer:
 - You can feel free to answer this poll honestly; no consequences will result from answering this poll



4

Plausible Deniability

• Theoretically,

you should feel less uncomfortable being told to answer the version with coin flips because you now have *plausible deniability:*

"I answered YES just because I got a head in my coin flip"

Randomized response mechanism

1/2

 $P + (l-p) + \frac{1}{2}$

Let's formalize the probability model (for each user):

- Pr(reports cheating|have cheated) =
- Pr(reports no cheating|have cheated) = 0
- Pr(reports cheating|have not cheated) =
- Pr(reports no cheating|have not cheated) =

Pr (have cheated)= P Pr (report cheaty) =

Aggregation

 Assume all students followed the instructions correctly. If p (proportion, so p ∈ [0,1]) of students have cheated, what is the expected proportion of YES responses (i.e., reports cheating) that Shao-Heng sees in the Canvas backend?

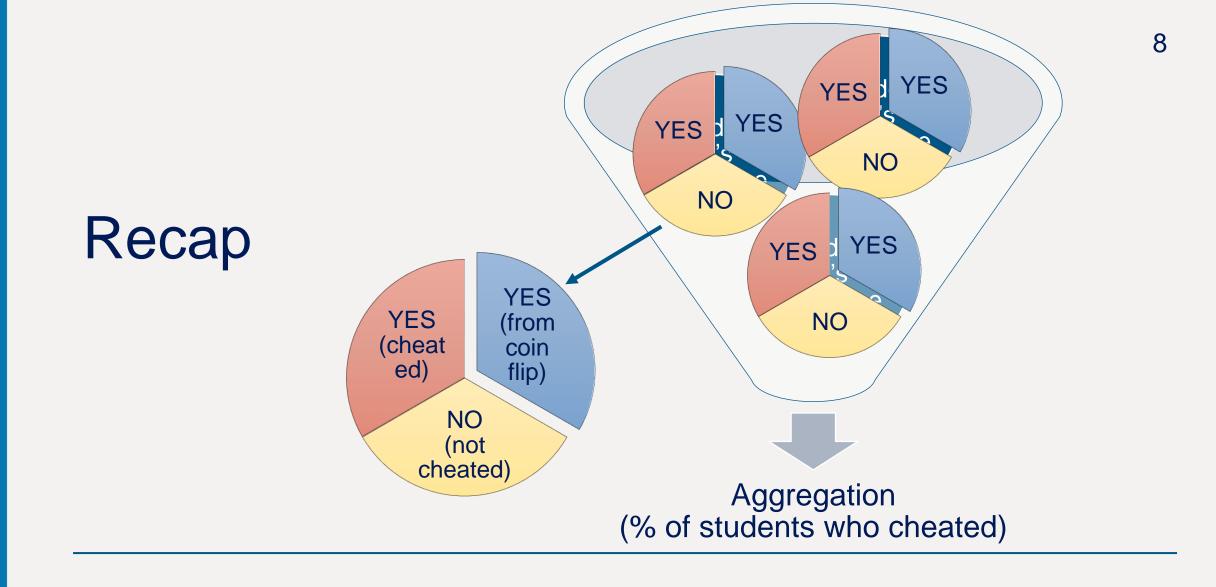


7

Inferring about underlying *p*

- What Shao-Heng is interested in this (hypothetical) scenario is exactly p
- p is unknown to Shao-Heng, remains unknown after poll
- Given any p, the proportion Shao-Heng observes (from the mechanism) has expectation $\frac{1+p}{2}$ and is centralized at $\frac{1+p}{2}$
 - In other words, $\frac{1+p}{2}$ is the most likely outcome that Shao-Heng observes, if the true proportion was p
 - Shao-Heng can then "estimate" what p is, treating the observation as $\frac{1+p}{2}$

Duke





Discussion: What was wrong in the poll

• The poll was not private. It had a serious design flaw. What is it?

9

A general notion of differential privacy

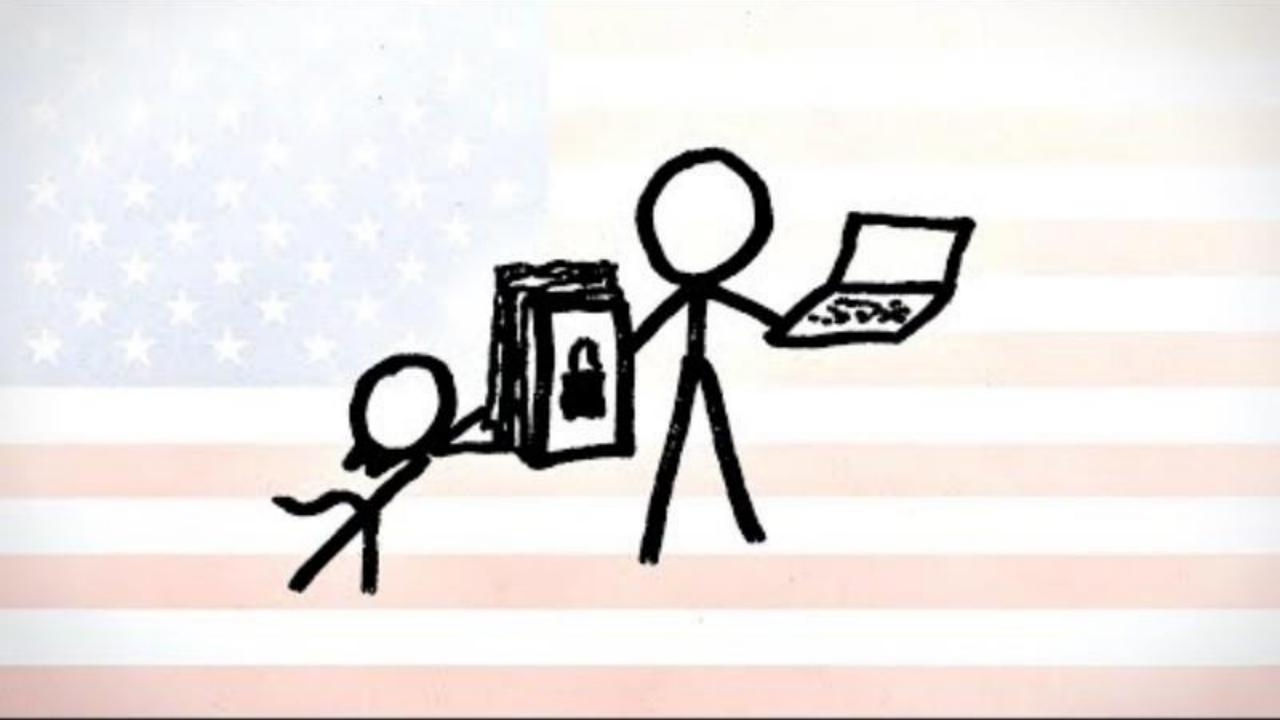
- Datasets: $D \in \mathcal{D}$
 - all respondents' responses to a survey ("the truth")
 - binary relation on \mathcal{D} (neighboring): whether two sets of "truths" are close to each other
- Queries: $q \in Q$
 - what's the proportion of responses who said X?
- Mechanisms: M(D,q)
 - given a dataset ("truth") and a query ("question"), how do we answer the query?
- Outputs: $S \subseteq S$
 - a "solution" or "statistics" that the mechanism outputs

A general notion of differential privacy

• A mechanism *M* is ε -differentially private (ε -DP) if $\Pr[M(D,q) \in S] \leq \underbrace{e^{\varepsilon}} \Pr[M(D',q) \in S], \forall S \subseteq S, q \in Q$

for all *neighboring* datasets $D, D' \in D$





Disclaimers

- Many different notions/models of differential privacy exist
 - local, central, ε -DP, (ε, δ) -DP...
 - who ensures privacy? users? researchers? both?
- Most of the DP literature needs continuous probability
 - we won't go there in CS230
 - but there's a whole reading list in Canvas for interested

