Homework 2: More on Differential Privacy

Instructor: Adam Smith

Collaboration and Honesty Policy Reminder: Collaboration in the form of discussion is allowed. However, all forms of cheating (copying parts of a classmate's assignment, plagiarism from papers or old posted solutions) are NOT allowed. A rough rule of thumb: you should be able to walk away from a discussion of a homework problem with no notes at all and write your solution on your own. Finding answers to problems on the Web or from other outside sources (these include anyone not enrolled in the class) is forbidden.

- You must write up each problem solution by yourself without assistance, even if you collaborate with others to solve the problem.
- You must identify your collaborators. If you did not work with anyone, you should write "Collaborators: none."
- Asking and answering questions in every forum the class provides (on Piazza, in class, and in office hours) is encouraged!
- Even though look up answers is forbidden, using the web to find alternative explanations of concepts you need for the homework *is* allowed, and encouraged. For example, you can look up background on probability and linear algebra, documentation for particular programming languages, etc.

1 Medians

Suppose we want to find the median of a list of real numbers $\mathbf{x} = (x_1, ..., x_n)$ that lie in the set $\{1, ..., R\}$. Consider an instantiation of the exponential mechanism based on the following score function: For every $y \in \{1, ..., R\}$, let

$$q_1(y; \mathbf{x}) = |rank(y) - n/2| = -\left| \sum_{i=1}^n sign(y - x_i) \right|.$$

where

$$sign(z) = \begin{cases} 1 & \text{if } z > 0, \\ 0 & \text{if } z = 0, \\ -1 & \text{if } z < 0. \end{cases}$$

Note that this score is 0 exactly when y is a valid median for \mathbf{x} .

- 1. Show that q has sensitivity at most 1 (when neighboring data sets are allowed to differ by the insertion or deletion of one entry).
- 2. Let A_{ϵ} be the algorithm one gets by instantiating the exponential mechanism with score q, parameter ϵ and output set $\mathcal{Y} = \{1, ..., R\}$. Show that there is a constant c > 0 such that: for every data set \mathbf{x} , for every R and $\epsilon < 1$, and for every $\beta \in (0, 1)$, the probability that $A_{\epsilon}(\mathbf{x})$ samples a value y with $|rank_{\mathbf{x}}(y) n/2| > c \cdot \frac{\ln(R) + \ln(1/\beta)}{\epsilon}$ is at most β . Here $rank_{\mathbf{x}}(y) \in \{0, 1, ..., n\}$ is the position y would have in the sorted order of \mathbf{x} .

[*Hint*: How does $rank_{\mathbf{x}}(\cdot)$ relate to $q(\cdot;_{\mathbf{x}})$? Look at the ratio between the probability mass of a true median and the probability mass of an element with very low or high rank.]

2 Histograms

Show that for any domain \mathcal{X} , the following algorithm is (ϵ, δ) -differentially private when neighboring data sets are allowed to differ by the insertion or deletion of one value.

Algorithm 1: Stable Histogam($\mathbf{x}; \epsilon, \delta$)

```
1 for every z \in \mathcal{X} that appears in \mathbf{x} do
2 | \tilde{c}_z = \#\{i : x_i = z\} + \operatorname{Lap}(1/\epsilon);
```

3 Release the set of pairs $\{(z, \tilde{c}_z) : \tilde{c}_z > \tau\}$ where $\tau = 1 + \frac{\ln(1/\delta)}{\epsilon}$.

Hint: The delicate part of this result is that we add noise only to counts of non-empty bins. (For example, if we were counting how many people live on each square mile of land in Alaska, most of the bins would be empty, but others would have lots of people.) There are two kinds of adjacent data sets: those where the set of nonempty bins changes, and those where it does not. You may need the following simple concentration bound for Laplace random variables: If $Y \sim \text{Lap}(\lambda)$, then for every t > 0, we have $\Pr(Y > \lambda t) \leq \frac{1}{2} \exp(-t)$.

For extra credit: Prove that the Stable Histograms algorithm is not $(\epsilon', 0)$ differentially private for any finite positive value ϵ' .