

Assignment week 7

1. When calculating the sensitivity in ϵ -Differential Privacy where the values to be derived from the data points is a d-dimension vector, identify the normalisation technique. (Notations are the same as used in the lecture)
 - a. **Manhattan normalisation**
 - b. Euclidian normalisation
 - c. Max normalisation
 - d. Min-max normalisation
 - e. Sigmoid normalisation
2. In (ϵ, δ) -Differential privacy what does $\delta=0$ imply? (Notations are the same as used in the lecture)
 - a. The equation $(P(M(x) \in S) \leq e^\epsilon (P(M(x') \in S))$ should hold for some of the subsets S
 - b. The equation $(P(M(x) \in S) \leq e^\epsilon (P(M(x') \in S))$ should hold for most of the subsets S
 - c. **The equation $(P(M(x) \in S) \leq e^\epsilon (P(M(x') \in S))$ should hold for all of the subsets S**
 - d. The equation $(P(M(x) \in S) \leq e^\epsilon (P(M(x') \in S))$ should hold for none of the subsets S
3. How do the utilities vary in the Laplacian mechanism vs the Gaussian mechanism in a higher dimension differential privacy setting?
 - a. As the dimension increases, the Gaussian mechanism requires quadratically more amount of noise than the Laplacian mechanism, decreasing the utility
 - b. As the dimension increases, the Gaussian mechanism requires quadratically lesser amount of noise than the Laplacian mechanism, decreasing the utility
 - c. **As the dimension increases, the Gaussian mechanism requires quadratically lesser amount of noise than the Laplacian mechanism, increasing the utility**
 - d. As the dimension increases, the Gaussian mechanism requires quadratically more amount of noise than the Laplacian mechanism, increasing the utility
4. _____ property ensures that a function applied on the privacy-protected data _____ its privacy aspect after applying a function over it.
 - a. **i. Post-processing ii. Retains**
 - b. i. Post-processing ii. Loses
 - c. i. Composition ii. Retains
 - d. i. Composition ii. Loses

5. After using k mechanisms for getting k (ϵ, δ) - differentially private data variations for a dataset, the combined leakage that is observed from these k mechanisms can be minimized by:
 - a. Using Laplacian Mechanism
 - b. **Using Gaussian Mechanism**
 - c. Using Uniform Mechanism
 - d. Using Exponential Mechanism

6. In a buyer-seller problem, given n buyers and n valuations by the buyers, what is the total **revenue** given a price p .
 - a. $p \sum_{i=1}^n A$ where $A = 1$ if $v_i \geq p$ and $A = 0$ if $v_i \leq p$
 - b. $p \sum_{i=1}^n A$ where $A = 0$ if $v_i \geq p$ and $A = 1$ if $v_i \leq p$
 - c. pn
 - d. $p(n - 1)$
 - e. $p(1/n)$

7. In the exponential mechanism to calculate the price to maximize the revenue, identify the correct statement in the scenario where 2 unequal prices result in the same revenue:
 - a. Both prices have an unequal probability of being selected
 - b. **Both prices have an equal probability of being selected**
 - c. A higher price has a higher probability of being chosen due to normalisation
 - d. A lower price has a higher probability of being chosen due to normalisation

8. In a classification problem, if a data point lies on a hyperplane that perfectly separates the two classes, the probability of the data point belonging to class A is:
 - a. 25%
 - b. **50%**
 - c. 75%
 - d. 100 %

9. In a vanilla Principle Component Analysis method, the reconstruction loss of a protected group is _____ than the remaining data before resampling and _____ than the remaining data after resampling.
 - a. **Higher, higher**
 - b. Higher, lower
 - c. Lower, higher
 - d. Lower, lower

10. The goal of a Fair PCA is to find a PCA solution U where $U=[U_a, U_b]$ such that reconstruction loss of the two groups A and B where A is the protected group is:

- a. Equal
- b. Unequal
- c. The protected group has a lower reconstruction loss
- d. The protected group has a higher reconstruction loss

11. In an ideal situation where the models are completely fair, the different parity values are:

- a. Approach 0
- b. 1
- c. Approach 1
- d. 0

12. Match the following:

i. $P(M(x) = 1 \mid x \text{ in } C) - P(M(x) = 1)$
regression

ii. $P(M(x) = 1 \mid y = 1 \text{ and } C) - P(M(x) = 1 \mid y = 1)$

iii. $P(M(x) = 1 \mid C = 1) - P(M(x) = 1 \mid C = 0)$

Opportunity

- a. i. - a, ii. - b, iii. - c
- b. i. - b, ii. - a, iii. - c
- c. i. - c, ii. - a, iii. - b
- d. i. - b, ii. - c, iii. - a

a. Fair Logistic

b. Statistical Parity

c. Equality of