DS 100/200: Principles and Techniques of Data Science Date: August 12, 2019

Exam Review

Name:

EDA & Visualization

- 1. For each of the following scenarios, determine which plot type is *most* appropriate to reveal the distribution of and/or the relationships between the following variable(s). For each scenario, select only one plot type. Some plot types may be used multiple times.
 - A. histogram B. pie chart C. bar plot D. line plot
 - E. side-by-side boxplots F. scatter plot G. stacked bar plot H. overlaid line plots
 - (a) Sale price and number of bedrooms for houses sold in Berkeley in 2010.
 - (b) Sale price and date of sale for houses sold in Berkeley between 1995 and 2015.
 - (c) Infant birth weight (grams) for babies born at Alta Bates hospital in 2016.
 - (d) Mother's education-level (highest degree held) for students admitted to UC Berkeley in 2016.
 - (e) SAT score and HS GPA of students admitted to UC Berkeley in 2016.
 - (f) The percentage of female student admitted to UC Berkeley each year from 1950 to 2000.
 - (g) SAT score for males and females of students admitted to UCB from 1950 to 2000

Exam Review 2

Optimization

2. Fix the following buggy Python implementation of gradient descent:

```
def grad_descent(X, Y, theta0, grad_function, max_iter = 1000):
"""X: A 2D array, the feature matrix.
Y: A 1D array, the response vector.
theta0: A 1D array, the initial parameter vector.
grad_function: Maps a parameter vector, a feature matrix, and a
    response vector to the gradient of some loss function at the
    given parameter value. The return value is a 1D array."""
theta = theta0
for t in range(1, max_iter+1):
    grad = grad_function(theta, X, Y)
    theta = theta0 + t * grad
return grad
```

3. Suppose you are given a dataset $\{(x_i, y_i)\}_{i=1}^n$ where $x_i \in \mathbb{R}$ is a one dimensional feature and $y_i \in \mathbb{R}$ is a real-valued response. You use f_θ to model the data where θ is the model parameter. You choose to use the following regularized loss:

$$L(\theta) = \frac{1}{n} \sum_{i=1}^{n} (y_i - f_{\theta}(x_i))^2 + \lambda \theta^2$$

- (a) This regularized loss is best described as:
 - (a) Average absolute loss with L^2 regularization.
 - (b) Average squared loss with L^1 regularization.
 - (c) Average squared loss with L^2 regularization.
 - (d) Average Huber loss with λ regularization.
- (b) Suppose you choose the model $f_{\theta}(x_i) = \theta x_i^3$. Using the above objective derive the loss minimizing estimate for θ .

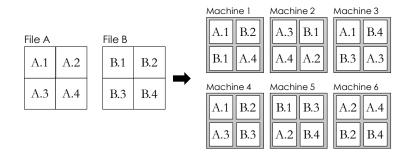
Exam Review 3

Inference

4. **True or False.** Determine whether the following statements are true or false.

- (a) Suppose we have 100 samples drawn independently from a population. If we construct a 95% confidence interval for each sample, we expect 95 of them to include the **sample** mean.
- (b) We often prefer a pseudo-random number generator because our simulations results can be exactly reproduced by controlling the seed.
- 5. Suppose we have a Pandas Series called **thePop** which contains a census of **25000** subjects. We also have a simple random sample of **400** individuals saved in the Series **theSample**. We are interested in studying the behavior of the bootstrap procedure on the simple random sample. Fill in the blanks in the code below to construct **10000** bootstrapped estimates for the median.

6. Consider the following layout of the files A and B onto a distributed file-system of 6 machines.



Assume that all blocks have the same file size and computation takes the same amount of time.

- (a) (1 point) If we wanted to load file A in parallel which of the following sets of machines would give the best load performance:
 - A. M1, M2 B. M1, M2, M3 C. M2, M4, M5, M6
- (b) (1 point) If we were to lose machines M1, M2, and M3 which of the following file or files would we lose (select all that apply).
 - A. File A B. File B C. We would still be able to load both files.
- (c) (1 point) If each of the six machines fail with probability p, what is the probability that we will lose block B.1 of file B.?
 - A. 3p B. p^3 C. $(1-p)^3$ D. $1-p^3$