

# **CSC380: Principles of Data Science**

#### **Final review**

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#### Announcements 5/7

- We will continue to hold office hours until final!
  - Office hour next Tuesday 5/13 moved to 10am

- Project submission
  - We created a 'project code' entry in gradescope
  - Submit your data there as well

# Logistics

- Time: May 13 3:30-5:30pm
- Place: WSEL200W (here)
- You are welcome to:
  - Bring calculators
  - Bring a letter-size "cheatsheet" (formulas, examples, pictures)
- Scope:
  - topics covered after midterm

# Logistics

- Problem types:
  - True/False questions (8 questions, 16pts)
    - We also ask you to provide brief justifications
  - Multiple-choice questions (6 questions, 24pts)
    - There can be multiple correct choices
    - Select all correct choices to get full credit
  - Free-form questions (about 9 questions, total 60 pts)
    - We don't expect them to be calculation heavy (mainly assessing understanding of basic concepts / methods)
    - We expect answers with justifications (steps)

# **General tips**

- Review:
  - Lecture slides
  - Homeworks
  - Quizzes
  - Practice problem set
- We will go over some example questions below
  - We will also release answer keys soon
- Let us know if anything is unclear
  - We are here to help (in office hours or online)

#### 1. Suppose that expectation of a random variable X is E(X) = 5. What is E(3X - 5)?

Q1

Linearity of expectation

• 
$$E[3X - 5] = E[3X] - 5 = 3 E[X] - 5 = 10$$

4. Suppose that a random X can take each of the five values -2, 0, 1, 3, 4 with equal probability. Compute the variance of X using the formula  $Var(X) = E(X^2) - (E(X))^2$ . Also compute the variance of Y = 2X - 10.

#### ... after some work, we found that Var(X) = 4.56

How can we find 
$$Var(Y) = Var(2X - 10)$$
?

$$Var(2X - 10) = Var(2X)$$
  
= 2<sup>2</sup> Var(X) = 18.24

Variance is preserved with constant shifts

$$Var(aX) = a^2 Var(X)$$

- 3. Suppose  $X_1, \ldots, X_n \sim N(\mu, \sigma^2)$ . Let  $\hat{\mu} := \frac{1}{n} \sum_{i=1}^n X_i$ . What distribution does  $\hat{\mu}$  follow? Fully specify the parameters of the distribution. Show your work and reasoning.
- Actually, this problem needs some extra conditions to be solvable..
- We need to assume that  $X_1, \ldots, X_n$  are independent
- What is the type of distribution of  $\hat{\mu}$ ?
  - Gaussian
- What specific Gaussian distribution does  $\hat{\mu}$  follow?
  - Figure out its mean and variance
  - See quiz 7

- 5. Suppose  $X \sim \text{Binomial}(10, 0.6)$ , and  $Y \sim \text{Binomial}(8, 0.6)$ , and X and Y are independent. Find the distribution of Z = X + Y. What is its variance? (Hint: it may be useful to think of X as a sum of independent Bernoulli random variables.)
- This is a tricky question.. (such questions are rare in exam)
- Idea 1:
  - Figure out all values Z can take (0, 1, ..., 18)

• Find 
$$P(Z = 0) = P(X = 0, Y = 0)$$
,  
 $P(Z = 1) = P(X = 0, Y = 1) + P(X = 1, Y = 0)$ 

Tons of calculations

- - -

- 5. Suppose  $X \sim \text{Binomial}(10, 0.6)$ , and  $Y \sim \text{Binomial}(8, 0.6)$ , and X and Y are independent. Find the distribution of Z = X + Y. What is its variance? (Hint: it may be useful to think of X as a sum of independent Bernoulli random variables.)
- This is a tricky question.. (such questions are rare in exam)
- Idea 2:
  - Think about these random variables *physically*
  - X = #shots made in the first 10 trials (with success prob. 0.6)
  - Y = #shots made in the next 8 trials (with success prob. 0.6)
  - X+Y = #shots made in a total of 18 trials (with success prob. 0.6)
  - Thus, ~ Binomial(18, 0.6)

#### $O_{C}$

9. Suppose we have placed two advertisements next to each other in a website. A user can either click both, click one of them, or not click at all. Let  $A \in \{1,0\}$  and  $B \in \{1,0\}$  be the random variables indicating whether each ad is clicked (1) or not(0). They follow the following joint probability table.

	B = 1	B = 0
A = 1	1/8	3/8
A = 0	3/8	1/8

(i) Which distribution does the random variable A follow? Specify the parameter of the distribution as well.

(ii) Are A and B independent? Justify your answer.

1. From joint distribution to marginal distribution: marginalization

2. How to tell if A and B are independent?

4 equalities need to hold: P(A = 0, B = 0) = P(A = 0)P(B = 0)...

19. Associate each plot below with one of the correlation coefficient values: 1, 0.8, 0.4, 0, -0.4, -0.8, -1.



Figure 1: 7 samples of (X, Y) with different joing distributions.

- Correlation coefficient  $\rho$  is always in [-1, 1]
- $\rho = -1 / +1 : X, Y$  are perfectly negatively / positively correlated

- True or false: expectation of product of two random variables *X*, *Y* is equal to the product of expectations of *X*, *Y*
- Is E[XY] = E[X] E[Y] ?
- Not necessarily the above is equivalent to Cov(X, Y) = 0Cov(X, Y) = E[XY] - E[X] E[Y]
- this is false when X, Y are positively / negatively correlated

- 12. Suppose that we observe 4 data points  $S = (x_1, x_2, x_3, x_4) = (3, 0, 1, 2)$  from Binomial $(3, \theta)$  and wish to estimate  $\theta$ .
  - (a) Compute the log-likelihood function  $\ln L(\theta)$ .

(b) Compute the value of the maximum likelihood estimator for  $\theta$ . You can use the fact that  $\ln L(\theta)$ 's only stationary point is its maximizer.

- Again this is a harder question (maybe better for HWs)
- Step 1: write down the log-likelihood

$$\ln L(\theta) = \sum_{i=1}^{4} \ln f(x_i; \theta)$$

 $f(x_i; \theta)$ : PMF of Binomial(3, $\theta$ ) on example  $x_i$ 

Taking the natural log, it is  $\ln \begin{pmatrix} 3 \\ x_i \end{pmatrix} + x_i \ln \theta + (3 - x_i) \ln(1 - \theta) = \begin{pmatrix} 3 \\ x_i \end{pmatrix} \theta^{x_i} (1 - \theta)^{3 - x_i}$ 

• Step 2: simplify the log-likelihood

$$\ln L(\theta) = \sum_{i=1}^{4} x_i \ln \theta + (3 - x_i) \ln(1 - \theta) + \ln \binom{3}{x_i}$$
$$= 6 \ln \theta + 6 \ln(1 - \theta) + \text{constant}$$

Step 3: find the parameter that maximize the log-likelihood as given by the hint, finding stationary point of  $\ln L(\theta)$  suffices

#### An example multiple-choice question

Q2: In least-squares linear regression, which of the following are true about the trained model?

- It fits a linear function to input-output pairs
- O It always achieves zero training loss

○ It minimizes the average square loss on the training set

It assumes output labels are binary

- Answer: 1, 3
- Points will be calculated based on the correctness of each check box

14. A random sample of n items is to be taken from a distribution with mean  $\mu$  and standard deviation  $\sigma$ . Use the central limit theorem to determine the smallest number of items n that must be taken in order to satisfy

$$P(|\overline{X}_n - \mu| \le \frac{\sigma}{4}) \ge 0.99$$

approximately. You may make use of the following outputs from scipy.stats: norm.ppf(0.995) = 2.58, norm.ppf(0.95) = 1.64, norm.cdf(0.995) = 0.84, norm.cdf(0.95) = 0.83.

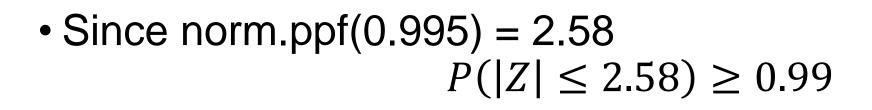
- General methodology: represent the probability using some RV whose distribution we know about (e.g. N(0,1))
- Here, we can consider the Z-statistic:

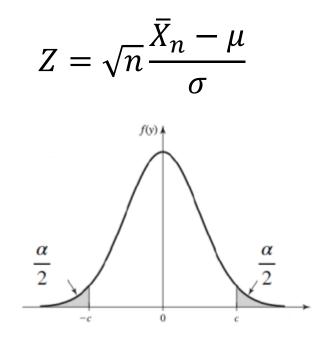
$$Z = \sqrt{n} \frac{\overline{X}_n - \mu}{\sigma} \sim N(0, 1)$$
 Holds approximately, by CLT

$$P(|\overline{X}_n - \mu| \le \frac{\sigma}{4}) \ge 0.99$$

 $P\left(|Z| \le \frac{\sqrt{n}}{4}\right) \ge 0.99$ 

• Can we write it in terms of *Z*?





• We should pick *n* such that  $\frac{\sqrt{n}}{4} \ge 2.58 \Rightarrow n \ge 106.5$ 

- 24. Compute the accuracy and F-score values for the following scenario: 10% of the items are positive and the rest are negative. Suppose we are using a random classifier that classifies the items as positive with 0.6 probability.
  - f(x): classifier's prediction y: example's true label
- Random classifier: f(x) and y and independent

Accuracy: P(f(x) = y) = P(f(x) = 1, y = 1) + P(f(x) = 0, y = 0)

independence, = 0.6 \* 0.1 independence, = 0.4 \* 0.9

= 0.06 + 0.36 = 0.42

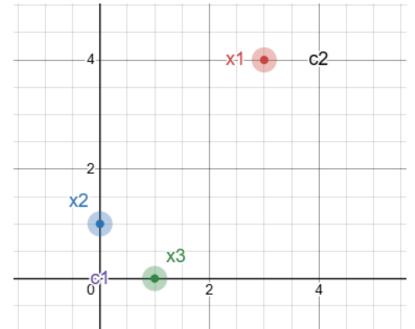
24. Compute the accuracy and F-score values for the following scenario: 10% of the items are positive and the rest are negative. Suppose we are using a random classifier that classifies the items as positive with 0.6 probability.

f(x): classifier's prediction y: example's true label Random classifier: f(x) and y and independent F-score:  $\frac{2}{Recall^{-1}+Precision^{-1}}$ 

Recall: P(f(x) = 1 | y = 1) = P(f(x) = 1) = 0.6Precision: P(y = 1 | f(x) = 1) = P(y = 1) = 0.1

F-score = 0.171

- 29. Suppose that we have the following 3 data points  $x_1 = (3, 4), x_2 = (0, 1), x_3 = (1, 0)$  Starting from the initial centroids  $c_1 = (0, 0)$  and  $c_2 = (4, 4)$ , run the k-means clustering algorithm until the centroids don't move anymore. State your final clustering result (i.e., state which points are in the same cluster)
- First, we always recommend drawing a 2D picture for such geometric problems



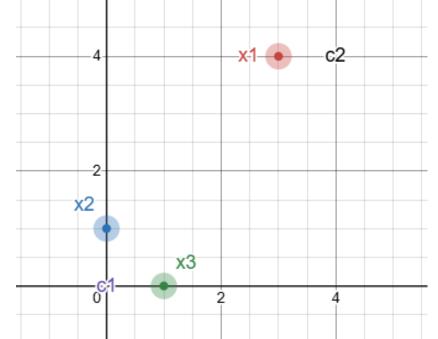
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#### k-means repeat:

step 1. re-assigning the clusters step 2. update the centroids

Iteration 1:

step 1:  $x_2, x_3$  goes to cluster 1,  $x_1$  goes to cluster 2 why?

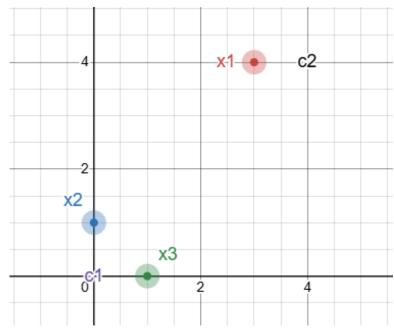


### ration 1.

Iteration 1:

step 1:  $x_2, x_3$  goes to cluster 1,  $x_1$  goes to cluster 2  $x_2$  is closer to  $c_1$  than  $c_2$ How to verify this rigorously?  $||x_2 - c_1|| = ||(0,1) - (0,0)|| = ||(0,1)||$   $= \sqrt{0^2 + 1^2} = 1$  $||x_2 - c_2|| = ||(0,1) - (4,4)|| = 5$ 

Also useful in nearest neighbor classification step 2: new centroids:  $c_1$  updated to (0.5,0.5),  $c_2$  updated to (3,4)

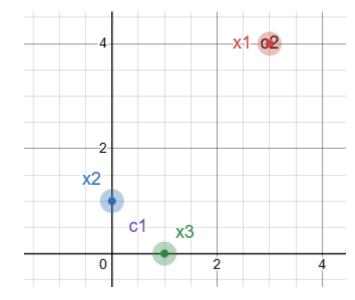


29. Suppose that we have the following 3 data points  $x_1 = (3, 4), x_2 = (0, 1), x_3 = (1, 0)$  Starting from the initial centroids  $c_1 = (0, 0)$  and  $c_2 = (4, 4)$ , run the k-means clustering algorithm until the centroids don't move anymore. State your final clustering result (i.e., state which points are in the same cluster)

#### Iteration 2:

- step 1:  $x_2$ ,  $x_3$  goes to cluster 1,  $x_1$  goes to cluster 2
- step 2: new centroids:  $c_1$  updated to (0.5,0.5),  $c_2$  updated to (3,4)

The centroids don't move any more. We are done!



# Hope you all excel in finals!