#### CSE 152 Mid-Term Examination

# University of California San Diego Oct 29, 2019

Name:	
UCSD ID:	@ucsd.edu

Question	Points	Points Earned
1	10	
2	10	
3	10	
4	10	
5	10	
6	10	
7	10	
8	10	
9	10	
Total	90	

#### **Instructions:**

- 1. This examination contains 12 pages, including this page.
- 2. You have **one** (1) hour to complete the examination. As a courtesy to your classmates, we ask that you not leave during the last fifteen minutes.
- 3. Write your answers in this booklet. We scan this into Gradescope, so **please try to avoid writing** on the backs of pages. If you must do so, please indicate **very** clearly on the front of the page that you have written on the back of the page.
- 4. You may use two (2) double-sided  $8.5" \times 11"$  pages with notes that you have prepared. You may not use any other resources, including lecture notes, books, other students or other engineers.
- 5. You may use a calculator. You may not share a calculator with anyone. If you didn't bring a calculator, you may use your phone, **but** you must put it on **flight mode** and clear all visible notifications **before** the examination starts, and you must not open any applications other than the calculator and a timer.

## Question 1: Basic Linear Algebra

$$A = \begin{bmatrix} 1 & 0 & 1 \\ \lambda & 1 & 1 \\ 1 & 1 & \lambda \end{bmatrix}$$

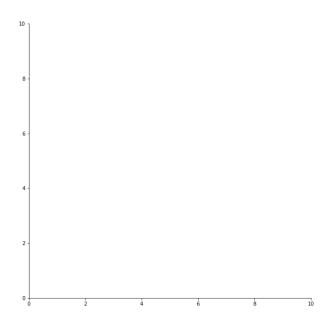
- (a) Calculate the rank of A when  $\lambda=1.$  Is A invertible? [2 pts]
- (b) What's the null space of A when  $\lambda = 1$ ? What's the dimension of its null space? [3 pts]
- (c) Give the condition of  $\lambda$  when A is a full rank matrix. What is the dimension of A's null space for such  $\lambda$ 's? [5 pts]

# Question 2: Least Squares Problem

Question 3: K-Means

Point	X Coord.	Y Coord.
1	6	2
2	7	1
3	8	3
4	7	2
5	3	7
6	1	9
7	2	7
8	2	5

(a) Plot the points on the graph and draw the circles around the two clusters of data. [2 pts]



b) The first iteration of your k-means algorithm initialized the cluster centers at  $C_1 = (2,6)$  and  $C_2 = (6,2)$ . Calculate the distance between Point 2 and it's cluster center. [3 pts]

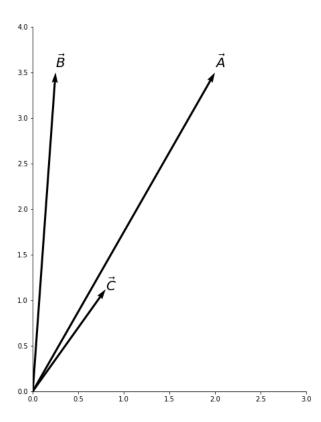
c) Using the work you have done in part a), draw the (rough) new locations for clusters  $C_1$  and  $C_2$  for the next iteration of the algorithm. [5 pts]

# Question 4: Bag of Visual Words

Creating and implementing a Bag of Words model involves the following steps:

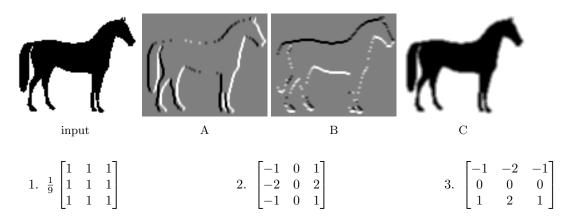
Extract Features  Learn Visual Vocab  Quantize Features  Represent Image by Frequency of Visual Words
a) Explain what occurs at each stage of the process. [8 pts]
Extract Features:
Learn Visual Vocab:
Quantize Features:
Represent Image by Frequency of Visual Words:

b) We utilize the cosine similarity metric to compare the visual words distributions between images. In the image below, indicate which vector has the higher cosine similarity to  $\vec{A}$  by circling the vector letter. [2 pts]



#### Question 5: Filters

(a) An input image shown below is filtered by some image filters to produce A, B, and C. Briefly explain the effect of each filter. [6 pts]



Explain.

(b) Given a gray-scale image I with shape (H, W). Let  $\operatorname{Filter}(K, I)$  be the operation that filters image I with kernel K. In this operation, the filtered result has shape (H, W) and the original image is padded with 0s if necessary. Prove

$$\operatorname{Filter}\left(\begin{bmatrix}1 & 1 & 1\\ 1 & 1 & 1\\ 1 & 1 & 1\end{bmatrix}, I\right) = \operatorname{Filter}\left(\begin{bmatrix}1\\1\\1\end{bmatrix}, \operatorname{Filter}(\begin{bmatrix}1 & 1 & 1\end{bmatrix}, I)\right)$$
 [4 pts]

### Question 6: Principal Component Analysis

- (a) Draw the direction of the first principal component in figure (a). [1 pt]
- (b) Draw the direction of the second principal component in figure (b). [1 pt]



(c) Suppose that we have already obtained the SVD result for some centered data matrix X (each row is the vector of a data point), which satisfies  $X = U\Sigma V^{\top}$ . By this SVD decomposition, what are the principal components and eigenvalues of PCA? [4 pts]

(d) Can you explain PCA from the optimization's perspective? What is the goal/objective of PCA [2 pts]? Is there any intuitive explanation? [2 pts]

#### Question 7: Harris Corner

(a) Here is the second moment matrix for Harris corner detection. [6 pts]

$$\sum_{x,y} w(x,y) \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix}$$

When evaluated at 3 patches A, B, C of an image, the eigenvalues of the matrices are

$$A: \lambda_1 = 0.9, \lambda_2 = 0.01; B: \lambda_1 = 0.012, \lambda_2 = 0.014, C: \lambda_1 = 0.9, \lambda_2 = 0.92$$

If we know that among A, B, C, there is a corner, an edge, and a flat region. What are A, B, C respectively? Fill in the blanks below with A, B, C.

Flat region: \_\_\_\_\_. Corner: \_\_\_\_. Edge: \_\_\_\_.

(b) Fill in the the blank space below in the pseudo-code for Harris corner response. Given the second moment matrix M, the Harris corner response is computed by

$$\det(M) - 0.06(\operatorname{tr}(M))^2$$

where

$$\det\left(\begin{bmatrix} a & b \\ c & d \end{bmatrix}\right) = ad - bc, \operatorname{tr}\left(\begin{bmatrix} a & b \\ c & d \end{bmatrix}\right) = a + d$$

You can assume addition, subtraction, multiplication, division, and square operations are performed elementwise on multi-dimensional arrays. [4 pts]

function HarrisCorner(I)

 $I_x \leftarrow x\_gradient(I)$ 

 $I_y \leftarrow y\_gradient(I)$ 

$$W \leftarrow \begin{bmatrix} \cdots & \cdots & \cdots \\ \cdots & \cdots & \cdots \\ \cdots & \cdots & \cdots \end{bmatrix}$$

 $\triangleright$  W is a 3x3 box filter

 $A \leftarrow \operatorname{convolve}(W, I_x I_x)$ 

 $B \leftarrow \operatorname{convolve}(W, \ldots)$ 

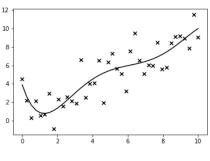
 $C \leftarrow \operatorname{convolve}(W, \ldots)$ 

 $response \leftarrow \_\_\_\_$ 

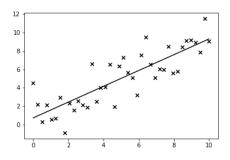
return response end function

#### Question 8: Bias and Variance

(a) The following images show two type of models for a regression task. [5 pts]



model A (K-Nearest Neighbor)



model B (Linear Regression)

Fill in the blanks (circle the correct answer) below.

Compared to Model B, Model A has \_\_\_\_\_ (lower/higher) bias and \_\_\_\_ (lower/higher) variance. Model A has \_\_\_\_ (better/worse) generalizability and is prone to \_\_\_\_ (overfitting/underfitting). Model A is \_\_\_\_ (more/less) sensitive to changes in the training data.

(b) Explain what "hyperparameter" for a model is [1 pt].

(c) Give an example of "hyperparameter" (e.g., a hyperparameter of Harris corner, KNN, K-Means, or Neural network)  $[1\,\mathrm{pt}]$ .

(d) Describe the procedure to choose hyperparameters with cross-validation. (You need to list the steps in cross-validation.)[3 pts]

## Question 9: Logistic Regression

Class	$S = f(x_i; w)$	Normalized Probability	$P(Y = k   X = x_i)$
Cat	3.4		
Dog	5.9		
Plane	5		

a) Explain the core function of the Softmax classifier in one to two sentences. [4 pts]

b) Given the model scores in the first column, what steps must you take in order to calculate class probability in the third column? (Explain process, do not do perform calculations) [4 pts]

c) Which class has the highest probability? [2 pts]