Exam 3

Last Name	First Name	Student ID #
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Honor Code: This exam represents only my own work. I did not give or receive help.

Signature:

Partial Credit: The most important issue is knowing how to approach a particular problem. Therefore, there will be partial credit for good solution outlines even if not all the mathematical manipulations are completed correctly. Be sure to attempt every problem!

- You have exactly 2 hours to complete this exam.
- No devices are allowed including no phones and no calculators.
- Unless indicated otherwise, you only need to setup up integrals correctly for full credit, which includes the correct limits and case-by-case conditions.
- You can use the provided formula sheet handouts no extra materials are allowed.
- No form of collaboration is allowed.
- There are 5 problems in total, each worth 20 points.

*** GOOD LUCK! ***

Problem	Points earned	out of	Problem	Points earned	out of
Problem 1		20	Problem 4		20
Problem 2		20	Problem 5		20
Problem 3		20			
			Total		100

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Problem 1 (Detection)

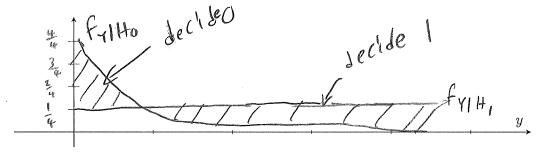
20 points

(This is a two-page problem with a single scenario.) Consider the following detection problem. $\mathbb{P}[H_0] = 4/5$ and $\mathbb{P}[H_1] = 1/5$. Under H_0 , Y is Exponential (1). Under H_1 , Y is Uniform (0,4).

(a) Determine the ML rule. Simplify your expression as much as you can.

$$\int_{|Y|H_0}^{|Y|H_0} = e^{-\frac{1}{2}} \left(\frac{1}{Y|H_0} + \frac{1}{4} \right) \\
\int_{|Y|H_0}^{|Y|H_0} = e^{-\frac{1}{2}} \left(\frac{1}{Y|H_0} + \frac{1}{2} \right) \\
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(b) Sketch the conditional PDFs $f_{Y|H_0}(y)$ and $f_{Y|H_1}(y)$ below. Clearly indicate the regions where the ML rule will decide 0 and where it will decide 1.



(c) Determine the probability of error for the ML rule.

$$P_{e} = P_{FA} P[H_{o}] + P_{mp} P[H_{i}]$$

$$P_{FA} = P_{Y|H_{o}}(o) + P_{Y|H_{o}}(i) = \frac{1}{4} + \frac{1}{4} = \frac{1}{16}$$

$$P_{mo} = P_{Y|H_{o}}(z) + P_{Y|H_{o}}(3) + P_{Y|H_{o}}(4) = e^{z} + e^{z} + e^{z} = e^{-q}$$

$$P_{mo} = P_{Y|H_{o}}(z) + P_{Y|H_{o}}(3) + P_{Y|H_{o}}(4) = e^{z} + e^{z} + e^{z} = e^{-q}$$

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$$P_{mo} = P_{Y|H_{o}}(z) + P_{Y|H_{o}}(3) + P_{Y|H_{o}}(4) = e^{z} + e^{z} + e^{z} + e^{z} + e^{z} + e^{z} = e^{-q}$$

$$P_{mo} = P_{Y|H_{o}}(z) + P_{Y|H_{o}}(3) + P_{Y|H_{o}}(4) = e^{z} + e^$$

(CONTINUED ON NEXT PAGE)

(d) Determine the MAP rule. Simplify your expression as much as you can.

$$D^{MAP}(x) = \begin{cases} 1, & \ln(4) + y \ge \ln(4) \\ 0, & \ln(4+y) \le \ln(4) \end{cases}$$

$$\begin{cases} 1, & y \ge \ln(4) - \ln(4) \\ 0, & y \le \ln(4) - \ln(4) \end{cases} = \begin{cases} 1, & y \ge \ln(8) \\ 0, & y \le \ln(4) - \ln(4) \end{cases}$$

$$\begin{cases} 1, & y \ge \ln(8) \\ 0, & y \le \ln(4) - \ln(4) \end{cases} = \begin{cases} 1, & y \ge \ln(8) \\ 0, & y \le \ln(8) \end{cases}$$

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$$\begin{cases} 1, & y \ge \ln(4) - \ln(4) \\ 0, & y \le \ln(8) \end{cases}$$

(e) Determine the probability of error for the MAP rule.

$$P_{FA} = P_{Y|A_0}(0) + P_{Y|A_0}(1) = \frac{1}{4} + \frac{1}{4} = \frac{1}{16}$$

$$P_{MD} = P_{Y|A_1}(2) + P_{Y|A_1}(3) + P_{Y|A_1}(4) = e^{-4}$$

$$P_{CC} = \frac{1}{16}(\frac{1}{5}) + e^{-4}(\frac{1}{5}) = \frac{1}{20} + \frac{1}{5e^{-4}}$$

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Problem 2 (Estimation)

20 points

(This is a two-page problem with two different scenarios.)

Scenario 1: Y = X + Z where X and Z are independent random variables with

$$\mathbb{E}[X] = 3 \qquad \mathbb{E}[Z] = 0 \qquad \operatorname{Var}[X] = 4 \qquad \operatorname{Var}[Z] = 2$$

(a) Determine
$$\mathbb{E}[Y]$$
, $Var[Y]$, and $Cov[X,Y]$.

$$E[Y] = E[X + 2] = E[X] + E[2] = 3 + 0 = \boxed{3}$$

$$Var[Y] = Var[x+z] = 4+2$$

 $E[xY] = E[x^2 + xz] = 9 + 0 = 9$

$$Cov[x, Y] = Gov[x, x+2] = Var[x] + Var[Y]$$

$$Cov[x, Y] = 9 - (3)(3) = 0$$

(b) We would like to find a linear estimator of X of the form aY + b that minimizes the mean-squared error. Determine the optimal values of a and b.

$$\hat{x}_{\text{MMSE}}(y) = E[(x - \hat{x}(Y))^2]$$

(c) Determine the mean-squared error of your estimator from part (b).

Scenario 2: $Y_1 = X_1 + Z_1$ where X_1, X_2, Z_1, Z_2 are independent random variables with $Y_2 = 2X_2 + Z_2$

$$\mathbb{E}[X_1] = \mathbb{E}[X_2] = 0 \qquad \mathbb{E}[Z_1] = \mathbb{E}[Z_2] = 0 \qquad \mathsf{Var}[X_1] = \mathsf{Var}[X_2] = 4 \qquad \mathsf{Var}[Z_1] = \mathsf{Var}[Z_2] = 2$$

(d) Determine $\Sigma_{\underline{Y}} = \begin{bmatrix} \mathsf{Var}[Y_1] & \mathsf{Cov}[Y_1,Y_2] \\ \mathsf{Cov}[Y_2,Y_1] & \mathsf{Var}[Y_2] \end{bmatrix}$ and $\Sigma_{\underline{X},\underline{Y}} = \begin{bmatrix} \mathsf{Cov}[X_1,Y_1] & \mathsf{Cov}[X_1,Y_2] \\ \mathsf{Cov}[X_2,Y_1] & \mathsf{Cov}[X_2,Y_2] \end{bmatrix}$.

(Hint: Note that $g(X_1, Z_1)$ and $h(X_2, Z_2)$ are independent for any functions g and h.)

$$Var[Y_1] = Var[X_1+2_1] = 4+2 = 6$$

 $Var[Y_2] = Var[2x_2+2_2] = 2^{2}(4)+2 = 18$

(e) We would like to find a linear estimator of $\begin{bmatrix} X_1 \\ X_2 \end{bmatrix}$ of the form $\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$ that minimizes the mean-squared error. Determine the optimal values of $a_{11}, a_{12}, a_{21}, a_{22}, b_1, b_2$.

You have been asked to evaluate the performance of two new stores. The table below summarizes how many online reviews each store received for a given star count (from 1 to 5 stars).

	1 Star	2 Stars	3 Stars	4 Stars	5 Stars
Store A Review Count	0	, 1	0	1	0
Store B Review Count	0	0	0	2_r	. 2

You may find the following table useful. Recall that $F_{T_m}(t)$ is the CDF for a t-distribution with m degrees-of-freedom and $F_{T_m}^{-1}(\beta)$ is its inverse.

m	1	2^{m}	3	4	5	6	7	8	9	10
$F_{T_m}^{-1}(0.025)$	-12.71	-4.30	-3.18	-2.78	-2.57	-2.45	-2.36	-2.31	-2.62	-2.23
$F_{T_m}^{-1}(0.05)$	l .	-2.92	i	t .						
$F_{T_m}^{-1}(0.1)$	-3.08	-1.89	-1.64	-1.53 (-1.48	-1.44	-1.41	-1.40	-1.38	-1.37

(a) Determine the sample mean and sample variance for Store A as well as for Store B.

$$\mathcal{A}_{A} = \frac{1}{2} \left(2 + 4 \right) = 3$$

$$\mathbf{A}_{A} = \frac{1}{2} \left(2 + 4 \right) = 3$$

$$\mathbf{A}_{A} = \frac{1}{2} \left(2 + 3 \right)^{2} + \left(4 - 3 \right)^{2} = 1 + 1 - \left[2 \right]$$

$$\hat{W}_{B} = \frac{1}{4} \left(4 + 4 + 5 + 5 \right) = \frac{18}{4} \left[\frac{9}{2} \right] V_{NB} = \frac{1}{4 - 1} \left(4 - \frac{9}{2} \right) + \left(4 - \frac{9}{2} \right)^{2} + \left(5 - \frac{9}{2} \right)^{$$

(b) Construct a confidence interval for the Store A average review with confidence level 0.9.

$$T = \sqrt{2} \left(3 - \frac{31}{2} \right)$$

$$P = \sqrt{2} \left(3 - \frac{31}{2} \right)$$

$$= -12.62$$

$$\left[3 \pm 12.62 \right]$$

(c) You have good reason to believe that the review variance is equal across stores. Use this new information to calculate the pooled sample variance. 2 summle $\frac{1}{1}$ test $\frac{1}{1}$ $\frac{1}{1}$

$$= \left(\left(2 - 1 \right) \left(2 \right) + \left(4 - 1 \right) \left(\frac{1}{3} \right) \right) / \left(2 + 4 - 2 \right) = \frac{2 + 1}{4} = \left(\frac{3}{4} \right)$$

(d) You would like to evaluate whether gap between the average review for Store A and Store B is statistically significant. Assuming the review variance is equal across stores, what kind of significance test should vou use?

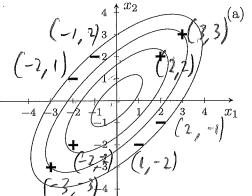
(e) Should we reject the null hypothesis at a significance level of 0.1? Justify your answer.

$$T = \frac{(N_{n_1} - M_{n_2})}{\sqrt{2}} = \frac{-\frac{3}{2}}{\sqrt{2}} = \frac{-\frac{3}{2$$

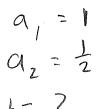
Problem 4 (Machine Learning)

20 points

You are given the 8 training data points on the figure, denoted by + and - symbols. The ellipses represent a contour plot for a vector Gaussian distribution fit to the entire training dataset. You will use PCA dimensionality reduction to create a one-dimensional version of this training dataset. Each part can be solved mainly with plots and illustrations.



(a) The PCA transform is of the form: $z = a_1x_1 + a_2x_2 + b$ Determine the values of a_1, a_1, b .



(b) Sketch the reduced one-dimensional dataset on the plot at the bottom of the page. (You do not need to exactly evaluate the one-dimensional coordinates or label your axes, but

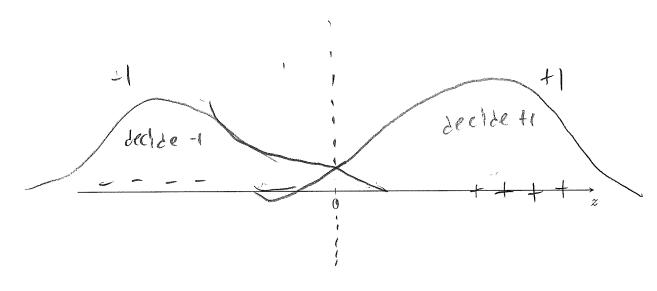
(c) For the reduced one-dimensional dataset, determine the training error rate for the closest average classifier. Justify your answer.

the relative spacing of the 8 points should be correct.)

2 error rate any linear line through the origin would give half thand half -1 on eitner 5'ide of the line

(d) Using dashed lines, sketch decision boundaries below that will result in 0 training errors.

(e) For this reduced dataset, it turns out the QDA classifier has 0 training errors. Below, sketch the likelihoods of the two Gaussian distributions used to determine these decision boundaries. No calculations are necessary, just an approximate sketch.

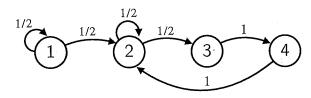


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Problem 5 (Markov Chains)

20 points

Consider the following discrete-time Markov chain. X_0 is equally likely to be 1, 2, 3, or 4.



(a) List the communicating classes. For each communicating class, determine the period and whether it is transient or recurrent.

and whether it is transient or recurrent. $C_1 = \{1\}$ feriod = 1 transfer t $C_2 = \{2,3,4\}$ period = 1 recurrent

(b) Determine
$$\mathbb{P}[X_2 = 1 | X_0 = 4]$$
. $P[X_2 = 1 | X_0 = 4] = P[X_0 = 4 | X_2 = 1] P[X_1 = 1]$

(c) Determine $\mathbb{P}[X_2 = 1]$.

(d) Does a unique limiting state probability vector π exist? If so, argue why and solve for it.
 If not, argue why.
 A unique limiting state probability vector π exists.

 $\begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow P^{T} \Pi = \Pi = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \Pi_{1} \\ \Pi_{2} \\ \Pi_{3} \\ \Pi_{4} \end{bmatrix} = \begin{bmatrix} \Pi_{1} \\ \Pi_{2} \\ \Pi_{3} \\ \Pi_{4} \end{bmatrix}$

$$\frac{1}{2}\pi_{1} = \pi_{1}$$

$$\frac{1}{2}\pi_{1} + \frac{1}{2}\pi_{2} = \pi_{2}$$

$$\frac{1}{2}\pi_{2} = \pi_{3}$$

$$\pi_{3} = \pi_{4}$$

(e) Given that the Markov chain starts in state 3, find the expected number of steps until it returns to state 3.

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