Math 342W / 642 / RM742 Spring 2025 Midterm Examination One

Professor Adam Kapelner March 20, 2025

to the success of this ed member of the college co	ademic community, its fundamental purpose is the pursuit of knowledge. Essential ucational mission is a commitment to the principles of academic integrity. Every mmunity is responsible for upholding the highest standards of honesty at all times. the community, are also responsible for adhering to the principles and spirit of the mic Integrity.
Activities that have	the effect or intention of interfering with education, pursuit of knowledge, or fair performance are prohibited. Examples of such activities include but are not limited
or other academic work	ttempting to use unauthorized assistance, material, or study aids in examinations or preventing, or attempting to prevent, another from using authorized assistance, Example: using an unauthorized cheat sheet in a quiz or exam, altering a graded t for a better grade, etc.
I acknowledge and agree	to uphold this Code of Academic Integrity.

Instructions

Full Name _____

This exam is 110 minutes (variable time per question) and closed-book. You are allowed **two** pages (front and back) of a "cheat sheet", blank scrap paper (provided by the proctor) and a graphing calculator (which is not your smartphone). Please read the questions carefully. Within each problem, I recommend considering the questions that are easy first and then circling back to evaluate the harder ones. No food is allowed, only drinks.

date

signature

Problem 1 This question relates to conceptual issues.

- (a) [8 pt / 8 pts] Circle the letters of all the following that are **true**. The notation below should be assumed the same from class. c d f i j k m n
 - (a) A model g for a real-world, non-man-made phenomenon g which is a function of real-world measurements x_1, \ldots, x_p could be absolutely true, (i.e. always have zero error when predicting in for future units).
 - (b) A model g for phenomenon y can be proved true analytically (using a mathematical deductive proof).
 - (c) A null model g_0 outputs a value in the set \mathcal{Y} .
 - (d) A null model g_0 can still be "useful" to the model's user.
 - (e) Any model g which is a function of real-world measurements x_1, \ldots, x_p always beats g_0 on predictive performance.
 - (f) A model g for phenomenon g which is of type nominal categorical with possible values within the set C_1, C_2, \ldots, C_L must return a value within the set C_1, C_2, \ldots, C_L .
 - (g) A model g for phenomenon g which is of type nominal categorical with possible values within the set C_1, C_2, \ldots, C_L must have a proxy x_j which is also type nominal categorical with possible values within the set C_1, C_2, \ldots, C_L .
 - (h) If x_j is of type nominal categorical with possible values within the set C_1, C_2, \ldots, C_L , then it is reasonable when modeling that those categorical values should be replaced by $1, 2, \ldots, L$.
 - (i) If x_j is of type nominal categorical with possible values within the set C_1, C_2, \ldots, C_L , then it is reasonable when modeling that those categorical values should be replaced by L-1 binary variables indicating the presence of absence of categories C_2, \ldots, C_L .
 - (j) If x_j is of type ordinal with possible values within the set C_1, C_2, \ldots, C_L , assuming that set is ordered from smallest to largest, then it is reasonable when modeling that those categorical values can be replaced by $1, 2, \ldots, L$.
 - (k) If x_j is of type ordinal with possible values within the set C_1, C_2, \ldots, C_L , then it is reasonable when modeling that those categorical values can be replaced by L-1 binary variables indicating the presence of absence of categories C_2, \ldots, C_L .
 - (l) In-sample performance metrics gives you an honest idea about how accurate your model is when using it for future prediction *always*.
 - (m) In-sample performance metrics gives you an honest idea about how accurate your model is when using it for future prediction only when n is much larger than p.
 - (n) Out-of-sample validation gives you an idea about how accurate your model is when using it for future prediction *always*.
 - (o) Out-of-sample validation gives you an idea about how accurate your model is when using it for future prediction only when n is much larger than p.
 - (p) Out-of-sample validation gives you an idea about how accurate your model is when using it for future prediction *only* when $\mathcal{Y} \subseteq \mathbb{R}$.

Problem 2 Consider the following data frame \mathbb{D} . Below is some R code (lines beginning with > are commands that were entered, other lines are output from those commands).

```
> skimr :: skim (D)
     - Data Summary
                                  Values
  Name
                                 D
  Number of rows
                                  592
  Number of columns
                                  4
  Key
                                 NULL
  Column type frequency:
    character
                                  2
    numeric
12
  Group variables
                                 None
13
14

    Variable type: factor

    skim_variable n_missing complete_rate ordered n_unique top_counts
16
                                              1 FALSE
                                                                  4 Bro: 220, Blu: 215,
  1 eye_color
                              0
      Haz: 93, Gre: 64
                              0
  2 hair_color
                                              1 FALSE
                                                                  4 Bro: 286, Blo: 127,
      Bla: 108, Red: 71
19
     – Variable type: numeric
    skim_variable n_missing complete_rate
                                                  mean
                                                           \operatorname{sd}
                                                              p0 p25 p50 p75 p100
                                                       0.500
22 1 is_male
                              0
                                              1 \ 0.471
                                                                0
                                                                    0
                                                                         0
                                                                              1
  2 is not male
                              0
                                              1 \ 0.529 \ 0.500
                                                                    0
23
24
  > D$ is _not _male = 1 - D$ is _male
  > D[sample(1 : nrow(D), 10),]
27
       is_male eye_color hair_color is_not_male
28 71
                     Brown
                                 Brown
                                                    0
              1
  446
              0
                     Hazel
                                  Brown
                                                    1
29
              0
                     Green
                                  Brown
  474
                                                    1
30
                     Brown
                                  Brown
  75
              1
                                                    0
31
              0
32 489
                     Brown
                                    Red
                                                    1
зз 107
              1
                     Brown
                                 Brown
                                                    0
34 | 473
              0
                     Green
                                 Brown
                                                    1
35 69
              1
                     Brown
                                 Brown
                                                    0
36 447
              0
                     Hazel
                                  Brown
                                                    1
  137
              1
                      Blue
                                  Brown
                                                    0
```

- (a) [1 pt / 9 pts] Circle one letter: the output on lines 28–37 is ...
 - A) the first 10 rows of data frame \mathbb{D}
 - B) a random 10 rows of data frame \mathbb{D}
 - C) the last 10 rows of data frame \mathbb{D}

- (b) [1 pt / 10 pts] What type of variable is eye_color? Be as specific as you can. nominal (or unordered) categorical (or factor)
 - Let y be defined as the variable is_male and all other variables be considered features.
- (c) [2 pt / 12 pts] What is the g_0 function equal to explicitly? $g_0 = 0$
- (d) [3 pt / 15 pts] Circle all of the following modeling procedures that would be appropriate for this modeling setting considering the response space.
 - A) OLS
 - B) Perceptron
 - C) SVM
 - D) SVM with Vapnik objective function (and a numeric choice for hyperparameter λ)
 - E) KNN with the Euclidean distance function
 - F) KNN with a non-Euclidean, but still valid distance function
- (e) [3 pt / 18 pts] Regardless of your answer in (c), how many columns will the full-rank design matrix X have? Hint: make sure you include the $\mathbf{1}_n$ as the first column in X.
 - 8 = 1 for intercept + 3 for eye_color + 3 for hair_color + 1 for is_not_male
- (f) [2 pt / 20 pts] Regardless of your answer in (c), consider using the SVM to fit g. What would the in-sample misclassification error for g be?

0

(g) [2 pt / 22 pts] Regardless of your answer in (c), consider using the OLS to fit g. What would the in-sample misclassification error for R^2 be?

1

Now we will change the modeling target. Let y be defined as the variable eye_color and all other variables be considered features.

- (h) [2 pt / 24 pts] What is the g_0 function equal to explicitly? $g_0 = \text{Brown}$
- (i) [3 pt / 27 pts] Circle all of the following modeling procedures that would be appropriate for this modeling setting considering the response space.
 - A) OLS
 - B) Perceptron
 - C) SVM
 - D) SVM with Vapnik objective function (and a numeric choice for hyperparameter λ)
 - E) KNN with the Euclidean distance function
 - F) KNN with a non-Euclidean, but still valid distance function
- (j) [3 pt / 30 pts] Regardless of your answer in (g), how many columns will the full-rank design matrix X have? Hint: make sure you include the $\mathbf{1}_n$ as the first column in X.
 - 5 = 1 for intercept + 3 for hair_color + 1 for is_male (this is_not_male is dropped as it's a linear combination of the intercept and is_male)

Now we will change the modeling target again. Let y be defined as the variable hair_color and we only use the feature eye_color.

(k) [5 pt / 35 pts] Let the reference level of eye_color be Green. For the subset of the data shown on rows 28–37 of the code, provide the same 10-row subset of the full-rank design matrix X. Name the columns of X appropriately.

(intercept)	eye_color_is_brown	eye_color_is_blue	eye_color_is_hazel
1	1	0	0
1	0	0	1
1	0	0	0
1	1	0	0
1	1	0	0
1	1	0	0
1	0	0	0
1	1	0	0
1	0	0	1
1	0	1	0

(l) [2 pt / 37 pts] Assume you use KNN with the Euclidean distance function to fit g. Of the three sources of error, which is likely the largest source of error in g's predictions? ignorance

Problem 3 Here are some theoretical questions. For all notation used, assume it has the same as it did in class.

(a) [4 pt / 41 pts] Assume $\mathcal{A} = \text{OLS}$ (linear regression). Prove that $\boldsymbol{b} = R^{-1}Q^{\mathsf{T}}\boldsymbol{y}$.

$$\mathbf{b} = (X^{\top}X)^{-1}X^{\top}\mathbf{y}$$

$$= ((QR)^{\top}QR)^{-1}(QR)^{\top}\mathbf{y}$$

$$= (R^{\top}Q^{\top}QR)^{-1}R^{\top}Q^{\top}\mathbf{y}$$

$$= (R^{\top}R)^{-1}R^{\top}Q^{\top}\mathbf{y}$$

$$= R^{-1}(R^{\top})^{-1}R^{\top}Q^{\top}\mathbf{y}$$

$$= R^{-1}Q^{\top}\mathbf{y}$$

- (b) [10 pt / 51 pts] Assume $\mathcal{A} = \text{OLS}$ (linear regression). For each of the following, if the operation is illegal, write "illegal". If the operation is legal and the expression allows for simplification, simplify as best you can after an "=" sign. If you cannot simplify, then leave the line *blank*. If your answer is the zero or one vector, please indicate the dimension as a subscript to get full credit.
 - \bullet XQ illegal
 - *XR*
 - $X\mathbf{b} = \hat{\mathbf{y}}$
 - HQ = Q

- QH illegal
- *Qb*
- $\bullet \ \mathbf{y}^{\top} H = \hat{\mathbf{y}}^{\top}$
- $\bullet \ e^{\top}H = \mathbf{0}_n^{\top}$
- $\bullet \ QQ^{\top}\mathbf{1}_n = \mathbf{1}_n$
- $\operatorname{rank}[H]/\operatorname{rank}[X] = 1$

Problem 4 Consider the Galton dataset we examined in lab. Below is some R code (lines beginning with > are commands that were entered, other lines are output from those commands).

```
| > skimr::skim(HistData::Galton)
  — Data Summary –
 Name
                               HistData :: Galton
 Number of rows
                               928
 Number of columns
                               2
 Column type frequency:
   numeric
                               2

    Variable type: numeric -

   skim_variable n_missing complete_rate mean
                                           1\ 68.3\ 1.79\ 64
 1 parent
                           0
                                                              67.5 \ 68.5 \ 69.5 \ 73
 2 child
                           0
                                                  2.52
                                             68.1
                                                       61.7 66.2
                                                                   68.2
                                                                        70.2
```

(a) [1 pt / 52 pts] What is the sample size of this dataset? 928 As in the lab, consider the dependent variable child which measures the height of a child and consider the independent variable parent which measures the average height of the mother and father of the child.

- (b) [1 pt / 53 pts] What is p in this modeling context?
- (c) [3 pt / 56 pts] "The unit is one ___" (fill in the blank below by describing the unit). father-mother-child trio
- (d) [2 pt / 58 pts] Regardless of the \mathcal{A} employed to create a model, which will likely be the greatest of the three sources of error?

 ignorance (since there's only one feature and the phenomenon of interest is complicated)
- (e) [2 pt / 60 pts] If $\mathcal{A} = \text{OLS}$, what are the contents of the set \mathcal{H} ?

$$\mathcal{H} = \{w_0 + w_1 x : w_0, w_1 \in \mathbb{R}\} = \{[1 \ x] \boldsymbol{w} : \boldsymbol{w} \in \mathbb{R}^2\}$$

(f) [3 pt / 63 pts] If $\mathcal{A} = \text{OLS}$, provide one specific $\langle x, y \rangle$ point on the OLS line (i.e. the g model).

since $\langle \bar{x}, \bar{y} \rangle$ is always on the OLS line, the only point you could know about without more output is $\langle 68.3, 68.1 \rangle$

(g) [5 pt / 68 pts] Prove numerically that there is "regression to the mean" for the phenomenon of interest in this modeling setting. Assume that the correlation coefficient between the two variables is 0.46.

Regression to the mean means that we predict a child to have a smaller height than their parents for parents with above-average height and we predict a child to have a larger height than their parents for parents with below-average height. This is equivalent to showing $b_1 < 1$. We now calculate b_1 and show it is less than 1.

$$b_1 = r \frac{s_y}{s_x} = 0.46 \cdot \frac{2.52}{1.79} = 0.65 < 1$$

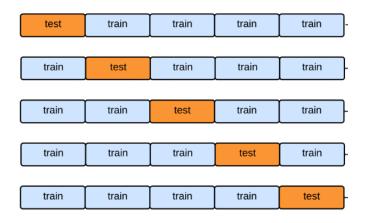
(h) [5 pt / 73 pts] If A = OLS, calculate the RMSE to two decimals.

$$SST = (n-1)s_y^2 = 927 \cdot 2.52^2 = 5886.821$$

$$R^2 = 1 - SSE/SST \Rightarrow SSE = SST(1 - R^2) = 5886.821(1 - .46^2) = 4650.588$$

$$RMSE = \sqrt{\frac{SSE}{n - (p+1)}} = \sqrt{\frac{4650.588}{926}} = 2.24$$

Problem 5 Consider the following illustration of cross-validation for \mathbb{D} with n=200.



- (a) [1 pt / 74 pts] What is the value of K?
- (b) [1 pt / 75 pts] How many units will be in \mathbb{D}_{train} in the first split? $n \frac{1}{K}n = 160$
- (c) [1 pt / 76 pts] If you calculate oos residuals in the second split, how many oos residuals will there be?

It's the same as the size of \mathbb{D}_{test} which is 40.

(d) [2 pt / 78 pts] If g was overfit on \mathbb{D}_{train} , then how would oosRMSE be expected to compare with in-sample RMSE?

oosRMSE > in-sample RMSE

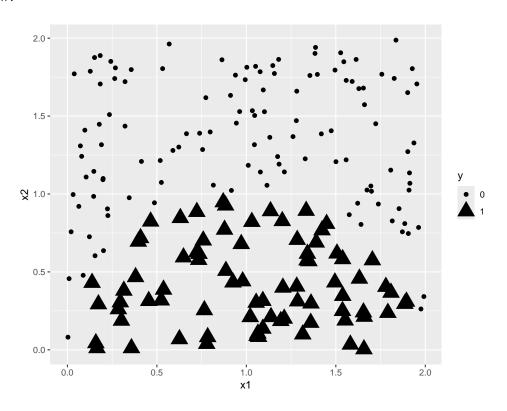
- (e) [3 pt / 81 pts] If you calculate the oosRMSE from the residuals in (c), this oosRMSE is expected to be greater than, less than or equal to the oosRMSE you expect in the future when predicting on g_{final} which is fit with all \mathbb{D} ?
- (f) [3 pt / 84 pts] If you aggregate all the residuals from all K folds together and calculate the oosRMSE, this oosRMSE is expected to be greater than, less than or equal to the oosRMSE you expect in the future when predicting on g_{final} which is fit with all \mathbb{D} ? greater than

Problem 6 Consider the following $\mathbb D$ illustrated within the questions below.

(a) [2 pt / 86 pts] Examine the plot of \mathbb{D} in the following question. If y is the response variable and the other two variables are considered features, what is the formal name of the type of modeling scenario most likely depicted?

binary classification

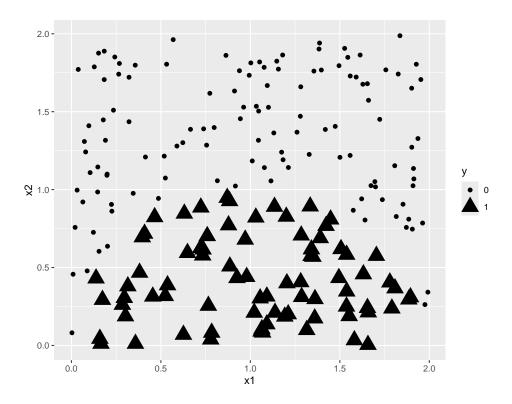
(b) [2 pt / 88 pts] If $\mathcal{A}=$ perceptron with the \mathcal{H} we considered thus far in class, draw a possible g that outputs from this algorithm after 10,000 iterations on the illustration below:



any straight line is acceptable

(c) [1 pt / 89 pts] If \mathcal{A} = perceptron with the \mathcal{H} we considered thus far in class, will the algorithm eventually converge? Circle one: yes / no

(d) [4 pt / 93 pts] If $\mathcal{A} = \text{SVM}$ with the \mathcal{H} we considered thus far in class and the Vapnik function with hinge errors and $\lambda = 0.001$, draw the g that most likely outputs from this algorithm on the illustration below:



With the hyperparameter set to such a low value, we're really just minimizing sum of hinge errors which would likely be the horizontal line at $x_2 \approx 0.75$.

(e) [3 pt / 96 pts] If $\mathcal{A} = \text{SVM}$ with the \mathcal{H} we considered thus far in class and the Vapnik function with hinge errors and $\lambda = 0.001$, what is the most likely of the three sources of error in g?

Misspecification since we can clearly see a semicircle threshold with diameter running across $x_2 = 0$ can obtain near zero misclassification error.

(f) [2 pt / 98 pts] If $\mathcal{A} = \text{KNN}$ with the Euclidean distance function and K = 1, what would be g(0,0)?

The closest point to <0,0> has y=0 so thus $\hat{\boldsymbol{y}}=0.$

(g) [2 pt / 100 pts] If A = KNN with the Euclidean distance function and K = 10, what would be g(0,0)?

Draw a circle arc from the y-axis to the x-axis with < 0, 0 > at its center such that 10 points are included. It is clear that there are more triangles than dots so that the modal value $\hat{y} = 1$.