[mock] 2024-02-27 Mid-term Exam

This is a partial exam of the Data Science course. There are 25 multiple-choice questions. All the questions have the same weight. There is only one correct or best answer for each question. Below are the general instructions:

- You will have a maximum of 2 hours to complete the exam. Students who have applied for special accommodation receive 30 minutes extra time.
- You are not allowed to leave the exam room during the first 30 minutes and last 15 minutes of the normal exam schedule.
- You can use the Jupyter Notebook to perform computations.
- You can use the "?" feature in the Jupyter Notebook to check the documentation of functions.
- You cannot use any reference materials (e.g., smartphone, books) except your one-page double-sided A4 cheatsheet.
- Read each question and all options very carefully.
- Double-check your answers before submission. We suggest answering all questions. We also suggest that you prioritize the questions that you are confident in and then strategically guess the answers for others that you are not confident in.
- Avoid spending too much time on a difficult question. It is better to complete other questions first and then go back to finish the difficult ones.
- You must keep the exam content confidential and are not allowed to copy or distribute the content in any form.

We have defined several terms in the table below. You can refer back to them when the questions mention these terms. The feature (i.e., the input) is x. The ground truth (i.e., the true output) is y. The predicted output is z.

Term	Equation
Identity activation function $f(x)$	$f(x) = x$
Sigmoid activation function $f(x)$	$f(x) = 1/(1 + e^{-x})$
Tanh activation function $f(x)$	$f(x) = (e^x - e^{-x})/(e^x + e^{-x})$
Hinge loss function $f(y, z)$	$f(y, z) = \max(1 - yz, 0)$
Squared error loss function $f(y, z)$	$f(y, z) = (y - z)^2$
Perceptron loss function $f(y, z)$	$f(y, z) = \max(-yz, 0)$
Logistic loss function $f(y, z)$	$f(y, z) = \log(1 + e^{-yz})$

Binary cross-entropy loss function $f(y,z)$	$f(y,z) = (1 - y)\log_2(1-z) - y\log_2(z)$
Softmax activation function $f(x)$, where x_i means the ith element in array x, and n means the total number of elements in array x	$f(x_i) = e^{x_i} / \sum_{j=1}^{n} e^{x_j}$
Entropy H for a coin with two sides (one side has probability p_1, and another side has probability p_2)	$H = p_1 \log_2(1/p_1) + p_2 \log_2(1/p_2)$

General Data Science Questions

1.0p **1** Given the following tables A and B, what will be the result after performing an outer join by User ID?

Table A

User ID	Name
1	Alice
2	Bob

Table B:

User ID	Age
2	25
3	30

User ID	Name	Age
2	Bob	25
3		30

User ID	Name	Age

○	2	Bob	25

	User ID	Name	Age
○	1	Alice	
	2	Bob	25

	User ID	Name	Age
	1	Alice	
○	2	Bob	25
	3		30

1.0p **2** Which of the following would be the best way to deal with missing data when you do not have a lot of data available when building a regression model, and you know that your data is MNAR (Missing Not At Random)?

○ Remove all observations with missing data

○ Replace the missing values with value -1

○ Find more data about the causes for the MNAR missingness

○ Use imputation techniques to model and fill in the missing values

Regression and Classification Questions

1.0p **3** Which of the following statements about overfitting is **FALSE**?

○ Using cross-validation can help us tackle overfitting.

○ We can combat overfitting by increasing the size of the test set.

○ Overfitting means fitting the training data extremely well but performing poorly on the validation or test data.

○ Overfitting usually happens when using a very complex model.

1.0p 4 Which of the following pairs is used to calculate recall?

○ True Positive and False Negative

○ True Positive and False Positive

○ True Negative and False Positive

○ True Negative and False Negative

1.0p 5 Suppose we fit a binary classification model to predict if a bad smell event will happen in the city for the next 6 hours. The positive label means that there will be smell events, and the negative label means no events. Which of the following statements is **FALSE**?

○ If the recall of the model is very high, it means that the model catches almost all the smell events and does not miss them.

○ If the model predicts that there would be an event, but it turns out there is nothing happening, this is called a False Positive.

○ If the precision of the model is very low, it means that the model is not very reliable when it predicts that there would be smell events in the future.

○ If the model predicts that there would be NO event, but it turns out that the model made a wrong prediction, this is called a True Negative.

1.0p 6 What is the purpose of using a loss function in classification or regression?

○ To measure the error between the prediction and the ground truth

○ To reduce the number of features

○ To increase the accuracy of the model on the training set

○ To regularize the model parameters

1.0p 7 You already have a linear classifier $f(x_1, x_2)$ that can predict a binary label y (yes/no) using two features, x_1 and x_2. Which of the following is the best way to use the linear classifier to determine the label of a new incoming data point with the values of these two features? For example, these two features could be the number of special characters and digits in an email message, and the labels could be whether the email is spam or not.

○ Calculate $f(x_1, x_2)$ and check if it is larger or smaller than the average of x_1 and x_2.

○ Calculate $f(x_1, x_2)$ and check if it is larger or smaller than zero.

○ Calculate $f(x_1, 0)$ and $f(0, x_2)$ and check which value is larger.

○ Calculate the average of $f(x_1, 0)$ and $f(0, x_2)$, and then check if the average is larger or smaller than zero.

1.0p **8** What is the goal of minimizing the sum of squared errors in linear regression?

○ To ensure that the residuals (the differences between predicted and true values) are normally distributed.

○ To find the line that best fits the data by minimizing the distance between the predicted and true values.

○ To find the optimal values of the input variables (features) that minimize the output variable (responses).

○ To reduce the variance in the model and prevent overfitting.

1.0p **9** Which of the following is the order that best describes the process of training a linear classifier for detecting whether an email is spam or not?
- A: Initialize a linear decision boundary randomly (e.g., a line, plane, or hyperplane)
- B: Take a separate set of emails with labels
- C: Update the decision boundary iteratively using an optimization algorithm with an error metric
- D: Transform the spam and non-spam messages into feature vectors
- E: Normalize the feature vectors to ensure they are on a similar scale
- F: Evaluate the classifier's performance using precision, recall, and F1 score
- G: Randomly shuffle the labels to test the classifier's robustness against noise

○ DEACF

○ DAEGC

○ ADGCBF

○ DEACBF

1.0p **10** Suppose we fit a simple linear regression model F using the training data with one feature X and one true response Y. We then use the model to output prediction Z=F(X). The coefficient of determination (R-squared) is equal to the square of:

○ The coefficient of feature X.

○ Pearson correlation coefficient between the true response Y and the predictions Z.

○ Pearson correlation coefficient between the predictions Z and the feature X.

○ Pearson correlation coefficient between the true response Y and the feature X.

1.0p **11** Suppose that we fit a binary classification model in identifying spam and ham (i.e., non-spam). Spam is the positive label, and ham is the negative label. The following shows the evaluation result of the model. What is the recall of the model based on the evaluation result?
- 70 samples are predicted as ham, but it turns out that they are spam in reality
- 80 samples are predicted as ham, and they are indeed ham in reality
- 30 samples are predicted as spam, and they are indeed spam in reality
- 20 samples are predicted as spam, but it turns out that they are not spam in reality

○ 0.3

○ 0.55

○ 0.4

○ 0.6

Structured Data Processing Question

1.0p **12** What is the main purpose of computing feature importance?

○ To prevent overfitting of the model

○ To determine which features are most important in making predictions

○ To increase the number of features used in the model

○ To improve the accuracy of the model on the training set

1.0p **13** Which of the following statements about Random Forest is **TRUE**?

○ Random Forest contains multiple Decision Tree models, and the best tree is used for performing the task.

○ Random Forest is more likely to overfit the data than the Decision Tree model.

○ Random Forest contains multiple Decision Tree models that are trained identically using the same set of features.

○ Random Forest uses randomly selected features and bootstrapped samples (i.e., sample with replacement).

1.0p **14** Suppose we flip a coin (with two sides) many times and we compute the entropy. Which of the following statements is **TRUE**?

○ Entropy is always one in this case because the coin has only two sides.

○ Entropy intuitively means the averaged surprise when we flip the coin.

○ Entropy reaches the minimum when the coin is fair, meaning two sides have equal probability.

○ If we change the probability of one side of the coin (to make it appear more frequently or less frequently), entropy is not sensitive to this change in probabilities.

1.0p **15** Suppose we want to classify if an image contains a banana or mango by using a Decision Tree model. We have 3 green banana images, 3 yellow banana images, 1 green mango image, and 1 yellow mango image. The Decision Tree uses entropy as the node-splitting strategy. Recall that Information Gain is the difference between the parent node's entropy and the leaf nodes' averaged entropy. What is the Information Gain after we ask the question, "is the fruit color yellow or not"?

○ 0.3333

○ 1

○ 0

○ 0.5

1.0p **16** What is the main advantage of using an ensemble of decision trees, such as a Random Forest, over a single decision tree in a classification or regression problem? Recall that errors of the model that we trained can be decomposed into bias, variance, and noise.

○ Reduced variance in the error

○ Reduced bias in the error

○ There is not really an advantage

○ Reduced noise in the error

1.0p **17** Which of the following best describes the reason why using misclassification error as the node-splitting strategy when training a decision tree model is not a good idea?

○ Using misclassification error as the node splitting strategy can lead to overfitting the training data, as it always creates more complex trees with a higher number of splits.

○ Misclassification error directly increases the computational complexity of training decision trees, making the process significantly slower compared to other methods such as entropy.

○ Misclassification error is less sensitive to changes after splitting a node than other measures such as entropy, making it less effective when deciding which feature to use.

○ Misclassification error is more prone to being affected by imbalanced datasets, leading to biased trees that favor the majority class.

1.0p **18** Suppose we want to train a Decision Tree based on the following dataset to predict whether Alex will go out or not. We use the misclassification error as the strategy when splitting a node. Which feature will the Decision Tree pick to split the first node?

Weather	Feeling	Wind	Time	Going out?
sunny	cold	calm	daytime	yes
rainy	warm	calm	nighttime	no
rainy	warm	windy	daytime	yes
sunny	warm	windy	daytime	no
rainy	warm	calm	daytime	no

○ Time

○ Wind

○ Weather

○ Feeling

Coding Question

1.0p **19** Which of the following best explains the purpose of the following Python Pandas code?

```
pandas.merge_ordered(A, B, on=A.index.name, how="outer", fill_method=None)
```

○ The code merges two pandas DataFrames A and B in order by the index of A, using an outer join, without filling missing values.

○ The code concatenates A and B vertically, aligning rows based on the index name of A, using the specified fill_method to handle None values.

○ The code performs an inner join on A and B based on their indexes, using the specified fill_method to handle None values.

○ The code sorts A based on its index and then appends B to it, without filling missing values.

1.0p **20** Suppose we have a pandas data frame D with 100 rows and two columns (C1 and C2). Column C1 has no missing data, and column C2 has 25% missing data. We want to sum up all valid items in column C2 and get a single integer (not a pandas.Series or pandas.DataFrame). Which

of the following code produces the desired output? For example, if D looks like the table below, the code should output only one single integer 27.

C1	C2
3	NaN
15	3
2	4
7	20

○ D.drop("C2", axis=1).sum().iloc[0]

○ D.dropna().sum()["C2"]

○ D.sum()["C1"]

○ D.dropna().iloc[2].sum()

1.0p **21** Suppose we have a pandas data frame D with the index column and also one value column. The value column has the name "smell". The index column contains pandas datetime objects. The "smell" column contains ratings of how bad the smell is at the corresponding timestamp. What is the following code doing?

```
D.resample("120Min", label="right").mean()
```

○ Compute the total smell values from the future two hours.

○ Compute the average smell values from the future two hours.

○ Compute the total smell values from the previous two hours.

○ Compute the average smell values from the previous two hours.

1.0p **22** You want to extract the year from a text column in a table using Python Pandas, and you want to use regular expressions (i.e., the pandas.Series.str.extract function). Which of the following regular expressions can achieve the task?

○ {4}[0123456789]

○ [0-9]{4}

○ [a-zA-Z]{4}

○ {4}[1-10]

1.0p **23** Given a pandas series object S with integers or floats. The index of each row means the time steps. What is the following code doing?

> S.rolling(3, min_periods=1, closed="right").sum()

○ For each row in series S, compute the sum of the current value and the previous two values from the two preceding time steps. Then, store the sum in the current row.

○ Compute the cumulative sum of series S, where the value in each row is the sum of all values up to and including that value.

○ Replace each value in series S with the sum of the current value and the maximum value in a window of size 3. The window is moved one position ahead at a time.

○ For each row in series S, compute the difference between each value and the average of the previous two values from the two preceding time steps.

Deep Learning Questions

1.0p **24** What is the main purpose of using an activation function in a neural network?

○ To regularize the model parameters

○ To introduce non-linearity into the model

○ To scale the input features to a consistent range

○ To reduce the dimensionality of the input data

1.0p **25** Given an error function $f(x^3)$, what is the value of x after making three gradient descent updates ($x_0 \rightarrow x_1 \rightarrow x_2 \rightarrow x_3$) using the initial point $x_0 = 10$ and learning rate 0.1? The derivative of x^3 is $3x^2$.

○ -38598736831290

○ 236379867649310

○ 81640

○ -6020