

# Exam I

**Due** No due date      **Points** 10      **Questions** 50**Available** Oct 10 at 11:30am - Oct 10 at 1pm about 2 hours**Time Limit** 75 Minutes

## Instructions

This exam is 50 questions and you only have 75 minutes to complete so please pace appropriately.

Also, questions are shown sequentially one at a time, and your answers are locked after you submit each question, so be sure of your answer to a question before you move on to the next one.

This quiz was locked Oct 10 at 1pm.

## Attempt History

	Attempt	Time	Score
LATEST	<a href="#">Attempt 1</a>	29 minutes	9.5 out of 10

Score for this quiz: **9.5** out of 10

Submitted Oct 10 at 12:01pm

This attempt took 29 minutes.

**Correct!**

### Question 1

**0.2 / 0.2 pts**

Select all scenarios that are examples of supervised learning



Predicting a buyer's chance of clicking on an online advertisement based on the previous behavior of similar online shoppers.



Using the nucleotide sequences on a region of non-coding DNA shared among species to estimate a phylogenetic tree.



An infant, unable to speak, but forming concepts of 'r' or 'l' sounds based on the grouping of similar sounds over time.

**Correct!**



Netflix using their database of user ratings to predict how you would rate a movie you haven't seen

## Question 2

0.2 / 0.2 pts

Select all examples of semi-supervised learning (as opposed to pure supervised or unsupervised learning examples)



Making stock predictions for a high-frequency trading company



Determining a taxonomy (tree-like classification) for animals based solely on their features

**Correct!**



Collecting constant GPS data, automatically clustering repeated locations, then having a personal label those clusters as "home" or "work" with the goal of having the setup detect whenever the wearer is at home or at work.

**Correct!**



Your learning of music genres, especially your ability to ask questions about a category of music that you notice as particularly distinct.

## Question 3

0.2 / 0.2 pts

Weather forecasters in Denton decided to build a model that predicts tomorrow's high temperature from the previous 30 day's high temperatures.

To do this, they used the past year's weather data to train the model. They had perfect accuracy in predicting when using last year's data for testing. However, when they applied the same model to predict the weather the next day, they found it was off by 10 degrees.

Select all statements that are likely to apply to their model.

Correct!



They should have used separate sets of data for training and for testing to pick the right model



It's Denton. It's not possible to predict the weather. They should just give up.

Correct!



The model is overfitting (it is too complex, too many variables)



The model is overgeneralizing (it is too simple)

#### Question 4

0.2 / 0.2 pts

Select all below that are an example of model overfitting (instead of overgeneralization)



You have an idea. You are often correct with your ideas. therefore since you came up with the idea, it must be right.



Fitting a U-shaped curve with a straight line

Correct!



This often happens when your model has too many parameters for too little data

Correct!



Fitting a U-shaped curve with a 20 degree polynomial, that fits all 21 points perfectly.

**Question 5****0.2 / 0.2 pts**

If k-nearest neighbors was your model of how you make decisions, which value of k would be more likely to be superstitious (lead to poor generalization, fit to the noise, be “too complex” of a model...)

- ☐ k=100
- ☒ k=1
- ☐ k as it approaches infinity
- ☐ k=5

**Correct!****Question 6****0.2 / 0.2 pts**

If you were trying to create a system to predict whether a picture is of a city or country scene, match the corresponding concept in machine learning terminology.

**Correct!****Sample**

a picture ▼

**Correct!****Feature**

the percentage of blue pix ▼

**Correct!****Target**

a label of 'City' or 'Country' ▼

**You Answered**

**Correct Answer****Question 7****0.2 / 0.2 pts**

Select all the ways you may be able to improve the accuracy of your machine learning prediction.

**Correct!**☒ Pick a different learning model (kNN, SVM, ...)**Correct!**☒ Derive/predict new features from current features in your data set**Correct!**☒ Add/remove features**Correct!**☒ Change your hyperparameter (e.g. k in kNN)**Correct!**☒ Get more samples**Question 8****0.2 / 0.2 pts**

What is the best rule of thumb for the number of samples needed for good machine model learning.

**Correct!**☐ features < samples☒ samples > feature<sup>2</sup>☐ samples > features☐ features < samples<sup>2</sup>

**Question 9****0.2 / 0.2 pts**

Match the advanced machine learning concept to an example of it

**Correct!****Clustering**

babies learning that 'r' and ▼

**Correct!****Feature engineering**

creating 'day of the week'; ▼

**Correct!****Feature selection**

Including "IQ" and "GPA" in ▼

**Correct!****Deep learning**

a type of machine learning ▼

**Question 10****0.2 / 0.2 pts**

If you are picking among many different model variants you generally split your data into three different groups. Match the group to a property of that group.

**Correct!****Training set**

data used to create the model ▼

**Correct!****Validation set**

like a test set but it is used to tune the model ▼

**Correct!****Test set**

the data set used to evaluate the model ▼

**Correct!**

▼

**Question 11****0.2 / 0.2 pts**

The difference between regression and classification is that regression predicts nominal variables/labels, while classification predicts continuous variables.

☐ True☒ False**Correct!****Question 12****0.2 / 0.2 pts**

Regression models can only use continuous variables as features and classification models can only use categorical/nominal variables as features.

☐ True☒ False**Correct!****Question 13****0.2 / 0.2 pts**

Match the method of normalization with the result

**$$\text{new\_score} = (\text{old\_score} - \text{min}) / (\text{max} - \text{min})$$**

range between 0 and 1 ▼

**$$\text{new\_score} = (\text{old\_score} - \text{mean}) / \text{standard\_deviation}$$**

0 mean and unit variance ▼

**Correct!****Correct!**

**Question 14****0.2 / 0.2 pts**

Let's assume you expect your friend carry a double-headed coin in their pocket 0.1% of the time. Match the following probabilities. Hint: you don't need to calculate them explicitly (use your intuitions), but you can if you want to.

**Correct!****P(10 heads|a trick coin)**

1 ▼

**Correct!****p(a fair coin)**

0.999 ▼

**Correct!****p(10 heads in a row | a fair coin)** $1/2^{10} = \sim 0.001$  ▼**Correct!****p(a trick coin | 10 heads in a row)** $\sim 0.2$  ▼**Correct!****p(10 heads in a row)** $\sim 0.002$  ▼**Question 15****0.2 / 0.2 pts**

Match the terminology with the appropriate conditional probability

**Correct!****p(data | model)**

likelihood ▼

**Correct!****p(model | data)**

posterior (the most sought) ▼

**Correct!****p(model)**

prior ▼



**Correct!****p(data)**

normalization term ▼

**Question 16****0.2 / 0.2 pts**

Which are true about Naive Bayes classifiers?

**Correct!**

The most common functional form used in naive bayes is gaussian (hence "Gaussian Naive Bayes or GNB) since all that is needed is the mean and variance of the data for each class to calculate the likelihood.

**Correct!**

Although Bayes rule is a mathematical truth, the assumption of independent features and gaussian probability functions are just approximations that, though efficient, can lead to misclassifications

**Correct!**

Naive Bayes classification is fast since each feature can be treated independently and the functional form for the probabilities are often simple to calculate.

**Correct!**

The "Naive" part is because we assume the likelihoods of each feature are independent (so the overall probability is found by simply multiplying them)

**Question 17****0.2 / 0.2 pts**

When using the k-Nearest Neighbors classifier, the k represents the number of nearest neighbors from the training set to find, and the class is the most represented class among those neighbors.

**Correct!**☒ True☐ False**Question 18****0.2 / 0.2 pts**

Normalization is not important in k-NN classification because the features with the larger range should always have a larger influence on the k-NN distance metric than other features.

☐ True**Correct!**☒ False**Question 19****0.1 / 0.2 pts**

Which equations are linear equations if the variables are x's and constants are a's?

**Correct Answer**☐  $y = a_0 x_0 + x_1 + a_2 x_2$ ☐  $y = a_0 x_0 + a_1 x_1 + a_2 (x_2)^2$ **Correct!**☒  $y = a_0 + a_1 x_1 + a_2 x_2$ ☐  $y = a_0 x_0 x_1 + a_2 x_2$ **Question 20****0.2 / 0.2 pts**

Match the goal for a linear regression model with the appropriate introduced feature

Correct!

have an intercept (e.g. the "b" in " $y = a x + b$ ")

Introduce a new feature w ▾

Correct!

Create a quadratic (2nd degree polynomial fit) from a linear regression model

add a feature that is the square of the original feature ▾

Correct!

Use a regression model for binary classification

change the target variable ▾

### Question 21

0.2 / 0.2 pts

Match the regularization strategy with the expected outcome

Correct!

No regularization

Irrelevant features will not be penalized ▾

Correct!

Ridge regression

Irrelevant feature coefficients are shrunk towards zero ▾

Correct!

Lasso regression

Irrelevant feature coefficients are set to zero ▾

### Question 22

0.2 / 0.2 pts

Which technique is more aptly considered a classifier?

☐ Ridge regression

**Correct!**

- ☐ Regularized regression
- ☐ Lasso regression
- ☒ Logistic regression

**Question 23****0.2 / 0.2 pts**

What is the purpose of regularization in linear regression?

**Correct!**

To decrease the coefficient values for irrelevant terms in the regression model

**Correct!**

To improve prediction accuracy on a future test set better than ordinary linear regression

**Correct!**

To diminish the contribution of irrelevant features to the resulting model, effectively performing automated feature selection during learning



To lasso the ridge with an elastic net

**Question 24****0.2 / 0.2 pts**

Match the term with the definition using the following values in a classification task

TP: True positives

FP: False positives - classified a positive, but truly negative

TN: True negatives

FN: False negatives - classified as negative, but truly positive

**Correct!**

	<b>Sensitivity (Recall for the positive case)</b>	TP / (TP + FN) ▼
<b>Correct!</b>	<b>Precision (for the positive case)</b>	TP / (TP + FP) ▼
<b>Correct!</b>	<b>Accuracy</b>	(TP + TN) / (TP + TN + FP + FN) ▼
<b>Correct!</b>	<b>Specificity (Recall for the negative case)</b>	TN / (TN + FP) ▼
<b>Correct!</b>	<b>F1 Score</b>	The harmonic mean of pre- and recall ▼

**Question 25****0.2 / 0.2 pts**

Match the cross-validation technique with the description

<b>Correct!</b>	<b>K-fold cross-validation</b>	Train your model on K-1 groups and validate on the remaining 1 group ▼
<b>Correct!</b>	<b>Leave one out cross-validation</b>	Same as K-fold cross-validation with K=1 ▼
<b>Correct!</b>	<b>Subject-wise cross-validation</b>	When you use the data from each subject once as training and once as testing ▼
<b>Correct!</b>	<b>Nested cross-validation</b>	A cross-validation technique that separates the data into training and testing sets twice ▼

**Question 26****0.2 / 0.2 pts**

Match the regression error metric with the appropriate statement

Correct!

**Root Mean Square  
Logarithmic Error**

An error of 110 instead of ▼

Correct!

**Mean Square Error**

An error of 110 instead of ▼

Correct!

**Mean Absolute Error**

An error of 110 instead of ▼

### Question 27

0.2 / 0.2 pts

When you want to know how well a product will work on a new person without any individual-specific training, it is better to use K-fold cross-validation than subject-wise cross-validation, because K-fold cross-validation may have an individual's data in both the training and test sets, which is what you want in that case.

☐ True

Correct!

☒ False

### Question 28

0.2 / 0.2 pts

When you use cross-validation to select the right hyperparameter, you need a separate set of data to properly measure the accuracy of the model with that hyperparameter (due to potential overfitting). Unfortunately, many scientists don't do this (although they should)

Correct!

☒ True

☐ False

### Question 29

0.2 / 0.2 pts

Match the term with the definition using the following values in a classification task

TP: True positives

FP: False positives - classified a positive, but truly negative

TN: True negatives

FN: False negatives - classified as negative, but truly positive

Correct!

**Sensitivity (Recall for the positive case)**

TP / (TP + FN) ▼

Correct!

**Precision (for the positive case)**

TP / (TP + FP) ▼

Correct!

**Accuracy**

(TP + TN) / (TP + TN + FP + FN) ▼

Correct!

**Specificity (Recall for the negative case)**

TN / (TN + FP) ▼

Correct!

**F1 Score**

The harmonic mean of precision and recall ▼

### Question 30

0.2 / 0.2 pts

Match the cross-validation technique with the description

Correct!

**K-fold cross-validation**

Train your model on K-1 groups and test on the remaining group ▼

**Correct!****Leave one out cross-validation**

Same as K-fold cross-valid ▼

**Correct!****Subject-wise cross-validation**

When you use the data frc ▼

**Correct!****Nested cross-validation**

A cross-validation techniq ▼

**Question 31****0.2 / 0.2 pts**

Match the regression error metric with the appropriate statement

**Correct!****Root Mean Square  
Logarithmic Error**

An error of 110 instead of ▼

**Correct!****Mean Square Error**

An error of 110 instead of ▼

**Correct!****Mean Absolute Error**

An error of 110 instead of ▼

**Question 32****0.2 / 0.2 pts**

When you want to know how well a product will work on a new person without any individual-specific training, it is better to use K-fold cross-validation than subject-wise cross-validation, because K-fold cross-validation may have an individual's data in both the training and test sets, which is what you want in that case.

☐ True**Correct!**☒ False



**Question 33****0.2 / 0.2 pts**

When you use cross-validation to select the right hyperparameter, you need a separate set of data to properly measure the accuracy of the model with that hyperparameter (due to potential overfitting). Unfortunately, many scientists don't do this (although they should)

**Correct!**☒ True☐ False**Question 34****0.2 / 0.2 pts**

Match the concepts to the descriptions for support vector machines

**Correct!****Support Vectors**

The samples that are near ▼

**Correct!****Margin**

The distance between clas ▼

**Correct!****Kernel**

The functions used to proj ▼

**Correct!****Slack Variable**

A parameter to allow sampl ▼

**Question 35****0.2 / 0.2 pts**

Check all of the features of the Random Forest classifier

**Correct!**☒ Averages the result of a set of decision trees to arrive at a final prediction**Correct!**☒ Is an ensemble method using decision trees☐ Relies on a "random walk" strategy**Correct!**☒ Creates trees using a subset of features**Question 36****0.2 / 0.2 pts**

Of the following classifiers, select the three that are more likely to work well on a complex learning problem

**Correct!**☒ Multi-layer neural network☐ Perceptron**Correct!**☒ SVM with radial basis function kernel☐ SVM with no kernel**Correct!**☒ Random Forest☐ Decision tree**Question 37****0.2 / 0.2 pts**

Match the following classifiers with the types of boundaries between classes they can construct in the feature space

**Correct!****Decision trees**

Vertical and Horizontal bc ▾

**Correct!****Perceptron**

A single separating line ▼

**Correct!****SVM with kernel functions,  
multilayer neural networks**

Arbitrarily complex separa ▼

**Question 38****0.2 / 0.2 pts**

The perceptron is simple, one-layer neural networks which is guaranteed to find a classification if the data is linearly separable. However, it fails on problems as simple as training for the XOR (exclusive-or) function.

**Correct!**☒ True☐ False**Question 39****0.2 / 0.2 pts**

Backpropagation is the ability of deep learning neural networks, like google's Deep Dream, to propagate signals from higher level neurons down to low-level neurons - allowing us in some cases to see the "hallucinated" images.

**Correct!**☐ True☒ False**Question 40****0.2 / 0.2 pts**

A random forest classifier uses boosting on a set of decision trees to increase performance

**Correct!**☐ True☒ False**Question 41****0.2 / 0.2 pts**

Boosting combines the predictions of all the learners with equal weight.

☐ True☒ False**Correct!****Question 42****0.2 / 0.2 pts**

Bootstrap estimation is a statistical technique which relies on resampling (often with replacement) from a distribution of data to arrive at a distribution of parameter values for the data (e.g. means, variances...) rather than simply one number.

☒ True☐ False**Correct!****Question 43****0.2 / 0.2 pts**

In bagging, both the features (commonly the columns in a data set) and samples/observations (commonly the rows in a data set) may be sampled. And this can be done with or without replacement.

**Correct!**☒ True☐ False**Question 44****0.2 / 0.2 pts**

If you are teams #2 and #3 in a competition and you want to merge your teams to beat #1, you will more likely be using bagging rather than boosting to win.

☐ True**Correct!**☒ False**Question 45****0.2 / 0.2 pts**

Bagging is applied to learners that are more likely to overfit, while boosting is often applied to weak learners (to avoid overgeneralization). e.g. Bagging decision trees may have no limits on size, whereas boosted decision trees may be limited by depth.

**Correct!**☒ True☐ False**Question 46****0.2 / 0.2 pts**

The key characteristic distinguishing ensemble methods from general model selection strategies is that you retain the candidate models in ensemble methods and use all of them to arrive at a prediction.

**Correct!**☒ True☐ False**Question 47****0 / 0.2 pts**

Here is an analogy:

"Rose" is to "Flower" as "Porsche" is to "Automobile", because the first word is a type of the second word.

"North" is to "South" as "Black" is to "White" because second word is the opposite of the first word.

and so on...

The following is analogy can be said for four important concepts in machine learning. Fill in the blank.

Classification is to regression in supervised learning as \_\_\_\_\_ is to dimensionality reduction in unsupervised learning.

Or more succinctly

Classification is to regression as \_\_\_\_\_ is to dimensionality reduction

☐ Reinforcement learning☒ PCA☐ Factor Analysis☐ Clustering**You Answered****Correct Answer****Question 48****0 / 0.2 pts**

Select all the model hyperparameters where a larger value leads to overfitting/higher model complexity rather than overgeneralization/simpler models.

**Correct!**

☒ the degree of the polynomial in polynomial regression

☐ the k in k nearest neighbors

**You Answered**

☒ the slack variable in support vector machine (larger means more slack or acceptance of errors))

**Correct Answer**

☐ the maximum depth parameter for decision trees

### Question 49

0.2 / 0.2 pts

Which of the following is just an ensemble method applied to a simpler classifier?

**Correct!**

☒ Random Forest

☐ Support Vector Machines

☐ K Nearest Neighbors

☐ Regularized Logistic Regression

### Question 50

0.2 / 0.2 pts

Asking a thousand people hundreds of questions about their personalities, you can use which technique to find numbers which may approximate the "Big 5" personality characteristics.

**Correct!**  
**You Answered**

☐ Clustering

☐ Classification

☒ Dimensionality reduction

☒ Regression

Quiz Score: **9.5** out of 10