Concept Quiz Over Week 4 Material

Due Oct 29 at 11:59pmPoints 1Questions 10Available Oct 25 at 12am - Oct 29 at 11:59pmTime Limit None

Score for this survey: **1** out of 1 Submitted Oct 28 at 9:12pm This attempt took 60 minutes.

| | Question 1 | | | |
|-------------|--|--|--|--|
| | A hard-margin linear SVM only has a solution when the data is linearly separable. | | | |
| | ○ True | | | |
| ou Answered | False | | | |
| | True. The constrained optimization only considers solutions where all datapoints are correctly classified. | | | |

Question 2 Finding the optimal weight vector for a hard-margin SVM requires solving an unconstrained optimization problem. True ou Answered False

False. The optimization includes constraints for the data to be correctly classified.

Question 3

The SVM dual formulation shows us that the optimal weight vector always depends on all data points.

True

ou Answered

False

False. While the weight vector is a weighted combination of all training points, points not within or on the margin have zero weight.

Question 4

Classifying new points for an SVM requires computing dot products between support vectors and the new point.

ou Answered

True

False

True. Substituting the definition of the optimal weight vector into $m{w}^T m{x} + m{b}$ shows this fact.

ou Answered

| | Question 5 |
|-------------|---|
| | Both the SVM primal and dual formulations can be solved using |
| | Gradient Descent |
| ou Answered | Quadratic Program Solvers |
| | Matrix Inverse |
| | Quadratic Program Solvers |
| | Question 6 |
| | Which of the following algorithms can handle multiclass classification? |
| ou Answered | k Nearest Neighbors |
| | Standard Logistic Regression |
| ou Answered | Multinomial Logistic Regression |
| | |

k Nearest Neighbors and Naive Bayes can handle multiple classes in their standard definitions. However, for logistic regression we had to go through a new derivation in order to handle multiclass problems.

Question 7

Naive Bayes

Tree classifiers, one-vs-all classifiers, and all-vs-all classifiers are schemes to let binary classification models work for multiple classes.

ou Answered

| _ | | |
|---|----|---|
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| | | |

False

True! We can use any binary classifier in constructing one of these tree / one-vs-all / all-vs-all classifiers to make multiclass predictions.

Question 8

What would the softmax function output for the input vector [5, -2, 10]^T?

Your Answer:

[0.006698, 0.000006, 0.993296]

$$Softmax \begin{pmatrix} \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_d \end{bmatrix} \end{pmatrix} = \begin{bmatrix} e^{z_1} \\ e^{z_2} \\ \vdots \\ e^{z_d} \end{bmatrix} \frac{1}{\sum_{i=1}^d e^{z_i}}$$

Implementing this in numpy with some pretty printing options:

```
import numpy as np
np.set_printoptions(formatter={'float': lambda x: "{0:0.3f}".format(x)})

a = np.array([5, -2, 10])
z = np.exp(a)
z = z/np.sum(z)
print(z)
[0.007 0.000 0.993]
```

Question 9

When deriving logistic regression originally, we set up our task of learning the optimal line w as maximizing the likelihood of our data under a conditional Bernoulli likelihood model. When deriving mulitclass logistic regression, we use the same formulation.

ou Answered

True

False

False. A Bernoulli likelihood only considers the y label to be 0 or 1. We used a Categorical likelihood which allowed the y label to be from a discrete set corresponding to our classes.

Question 10

What is a confusion matrix and what can you learn about a classifier by looking at one?

Your Answer:

confusion matrix is a table used in machine learning and statistics to evaluate the performance of a classification model. It provides a comprehensive summary of the model's predictions and how they compare to the actual outcomes, a confusion matrix is particularly useful for understanding the quality of a classifier's predictions in a misclassification problem.

Confusion matrices keep track of a classifier's predictions relative to the true value of its inputs. For instance, the i, j'th entry in a confusion matrix counts the number of times an instance of class i is labeled class j by the classifier. Looking at a confusion matrix can show you which classes a classifier commonly mixes up.

Survey Score: 1 out of 1