

Quiz

Machine Learning System Design

1. You are working on a spam classification system using regularized logistic regression. "Spam" is a positive class ($y = 1$) and "not spam" is the negative class ($y = 0$). You have trained your classifier and there are $m = 1000$ examples in the cross-validation set. The chart of predicted class vs. actual class is:

0 / 1 point

	Actual Class: 1	Actual Class: 0
Predicted Class: 1	85	890
Predicted Class: 0	15	10

For reference:

- Accuracy = (true positives + true negatives) / (total examples)
- Precision = (true positives) / (true positives + false positives)
- Recall = (true positives) / (true positives + false negatives)
- F_1 score = $(2 * \text{precision} * \text{recall}) / (\text{precision} + \text{recall})$

What is the classifier's accuracy (as a value from 0 to 1)?

Enter your answer in the box below. If necessary, provide at least two values after the decimal point.

0.095



Incorrect

The answer you gave is not a number.

2. Suppose a massive dataset is available for training a learning algorithm. Training on a lot of data is likely to give good performance when two of the following conditions hold true.

1 / 1 point

Which are the two?

- ☒ A human expert on the application domain
can confidently predict y when given only the features x
(or more generally, if we have some way to be confident
that x contains sufficient information to predict y
accurately).

✓ Correct

It is important that the features contain sufficient information, as otherwise no amount of data can solve a learning problem in which the features do not contain enough information to make an accurate prediction.

- ☐ The classes are not too skewed.
- ☒ Our learning algorithm is able to
represent fairly complex functions (for example, if we
train a neural network or other model with a large
number of parameters).

✓ Correct

You should use a complex, "low bias" algorithm, as it will be able to make use of the large dataset provided. If the model is too simple, it will underfit the large training set.

- ☐ When we are willing to include high
order polynomial features of x (such as $x_1^2, x_2^2,$
 $x_1 x_2$, etc.).

3. Suppose you have trained a logistic regression classifier which is outputting $h_\theta(x)$.

1 / 1 point

Currently, you predict 1 if $h_\theta(x) \geq \text{threshold}$, and predict 0 if $h_\theta(x) < \text{threshold}$, where currently the threshold is set to 0.5.

Suppose you **increase** the threshold to 0.7. Which of the following are true? Check all that apply.

- ☒ The classifier is likely to now have lower recall.

✓ **Correct**

Increasing the threshold means more $y = 0$ predictions. This will increase the decrease of true positives and increase the number of false negatives, so recall will decrease.

- ☐ The classifier is likely to have unchanged precision and recall, but higher accuracy.
- ☐ The classifier is likely to now have lower precision.
- ☐ The classifier is likely to have unchanged precision and recall, but lower accuracy.

4. Suppose you are working on a spam classifier, where spam emails are positive examples ($y = 1$) and non-spam emails are negative examples ($y = 0$). You have a training set of emails in which 99% of the emails are non-spam and the other 1% is spam. Which of the following statements are true? Check all that apply.

1 / 1 point

- ☒ If you always predict spam (output $y = 1$), your classifier will have a recall of 100% and precision of 1%.

✓ **Correct**

Since every prediction is $y = 1$, there are no false negatives, so recall is 100%. Furthermore, the precision will be the fraction of examples which are positive, which is 1%.

- ☐ If you always predict spam (output $y = 1$), your classifier will have a recall of 0% and precision of 99%.

- ☒ If you always predict non-spam (output $y = 0$), your classifier will have a recall of 0%.

✓ **Correct**

Since every prediction is $y = 0$, there will be no true positives, so recall is 0%.

- ☒ If you always predict non-spam (output $y = 0$), your classifier will have an accuracy of 99%.

✓ **Correct**

Since 99% of the examples are $y = 0$, always predicting 0 gives an accuracy of 99%. Note, however, that this is not a good spam system, as you will never catch any spam.

5. Which of the following statements are true? Check all that apply.

1 / 1 point

- ☒ On skewed datasets (e.g., when there are more positive examples than negative examples), accuracy is not a good measure of performance and you should instead use F_1 score based on the precision and recall.

✓ **Correct**

You can always achieve high accuracy on skewed datasets by predicting the most the same output (the most common one) for every input. Thus the F_1 score is a better way to measure performance.

- ☒ Using a **very large** training set makes it unlikely for model to overfit the training data.

✓ **Correct**

A sufficiently large training set will not be overfit, as the model cannot overfit some of the examples without doing poorly on the others.

- ☐ It is a good idea to spend a lot of time collecting a **large** amount of data before building your first version of a learning algorithm.
- ☐ After training a logistic regression classifier, you **must** use 0.5 as your threshold for predicting whether an example is positive or negative.
- ☐ If your model is underfitting the training set, then obtaining more data is likely to help.