

Machine Learning System Design

最新提交作业的评分

80%

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:

1/1 分

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

For reference:

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

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



Correct

The classifier correctly predicted the true positives and the true negatives = $85 + 10$, so the accuracy is $95/1000 = 0.095$

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.

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Which are the two?

☐ We train a learning algorithm with a small number of parameters (that is thus unlikely to overfit).

☒ We train a model that does not use regularization.



This should not be selected

Even with a very large dataset, some regularization is still likely to help the algorithm's performance, so you should use cross-validation to select the appropriate regularization parameter.

☐ The features x contain sufficient information to predict y accurately. (For example, one way to verify this is if a human expert on the domain can confidently predict y when given only x).

☒ We train a learning algorithm with a large number of parameters (that is able to learn/represent fairly complex functions).



Correct

You should use a "low bias" algorithm with many parameters, as it will be able to make use of the large dataset provided. If the model has too few parameters, it will underfit the large training set.

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

1/1 分

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

☐ The classifier is likely to now have lower precision

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

4. Suppose you are working on a spam classifier, where spam

1/1 分

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.

- ☒ 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.

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

- ☒ 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 with are positive, which is 1%.

- ☒ 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%.

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

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- ☒ 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.

- ☐ If your model is underfitting the training set, then obtaining more data is likely to help.
- ☒ 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.

- ☐ After training a logistic regression classifier, you **must** use 0.5 as your threshold for predicting whether an example is positive or negative.
- ☐ 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.