

Evaluating a Learning Algorithm

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- Video:** Deciding What to Try Next
5 min
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- Video:** Evaluating a Hypothesis
7 min
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- Reading:** Evaluating a Hypothesis
4 min
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- Video:** Model Selection and Train/Validation/Test Sets
12 min
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- Reading:** Model Selection and Train/Validation/Test Sets
3 min

Bias vs. Variance

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- Video:** Diagnosing Bias vs. Variance
7 min
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- Reading:** Diagnosing Bias vs. Variance
3 min
-
- Video:** Regularization and Bias/Variance
11 min
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- Reading:** Regularization and Bias/Variance
3 min
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- Video:** Learning Curves
11 min
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- Reading:** Learning Curves
3 min
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- Video:** Deciding What to Do Next Revisited
6 min
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- Reading:** Deciding What to do Next Revisited
3 min

Review

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- Reading:** Lecture Slides
10 min
-
- Quiz:** Advice for Applying Machine Learning
5 questions
-
- Programming Assignment:** Regularized Linear Regression and Bias/Variance
3h

Building a Spam Classifier

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- Video:** Prioritizing What to Work On
9 min
-
- Reading:** Prioritizing What to Work On
3 min



Model Selection and Train/Validation/Test Sets

Just because a learning algorithm fits a training set well, that does not mean it is a good hypothesis. It could over fit and as a result your predictions on the test set would be poor. The error of your hypothesis as measured on the data set with which you trained the parameters will be lower than the error on any other data set.

Given many models with different polynomial degrees, we can use a systematic approach to identify the 'best' function. In order to choose the model of your hypothesis, you can test each degree of polynomial and look at the error result.

One way to break down our dataset into the three sets is:

- Training set: 60%
- Cross validation set: 20%
- Test set: 20%

We can now calculate three separate error values for the three different sets using the following method:

1. Optimize the parameters in Θ using the training set for each polynomial degree.
2. Find the polynomial degree d with the least error using the cross validation set.
3. Estimate the generalization error using the test set with $J_{test}(\Theta^{(d)})$, (d = theta from polynomial with lower error);

This way, the degree of the polynomial d has not been trained using the test set.

✓ Complete

Go to next item