

Deciding What to Try Next

Evaluating a Learning Algorithm

Advice for Applying Machine Learning:

Introduction

- how to choose one of the most promising avenues to spend your time pursuing.
 - number of practical suggestions,
 - advice,
 - guidelines
- on how to do that.
- And concretely what we'd focus on is the problem of, suppose you are developing a machine learning system or trying to improve the performance of a machine learning system, how do you go about deciding what are the proxy avenues to try next?

Debugging a learning algorithm:

Suppose you have implemented regularized linear regression to predict housing prices.

$$\rightarrow J(\theta) = \frac{1}{2m} \left[\sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2 + \lambda \sum_{j=1}^m \theta_j^2 \right]$$

However, when you test your hypothesis on a new set of houses, you find that it makes unacceptably large errors in its predictions. What should you try next?

Actions

- Some actions
 - Get more training examples
 - Try smaller set of features (a small set of features)
 - Try getting additional features (just the opposite)
 - Try adding polynomial features
 - Try increasing λ
 - Try decreasing λ
- People generally randomly choose one and try it, which is waste of time most of the time.

Machine learning diagnostic:

Diagnostic: A test that you can run to gain insight what is/isn't working with a learning algorithm, and gain guidance as to how best to improve its performance.

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Windows'u etkinleştirmek için Ayarlar'a gidin.

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Exercise

- Which of the following statements about diagnostics are true? Check all that apply.
 - It's hard to tell what will work to improve a learning algorithm, so the best approach is to go with gut feeling and just see what works.
 - Diagnostics can give guidance as to what might be more fruitful things to try to improve a learning algorithm.
 - Diagnostics can be time-consuming to implement and try, but they can still be a very good use of your time.
 - A diagnostic can sometimes rule out certain courses of action (changes to your learning algorithm) as being unlikely to improve its performance significantly.