Applied Machine Learning with Big Data "EE 6973"



Topic: Linear Regression

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Logistics

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• Office hours: Tuesday &Thursday 10:15:11:00 (30 minutes after our class)

Course Web: https://github.com/ml6973/Course

Mailing list: TBD

Course Social Network: TBD

Outline

Prerequisites:

Calculus

Python

Linear Algebra

Probabilitiy

3 Types of Learning

Supervised Unsupervised Reinforcement

- Learning from labeled data
- ➤ E.g., Spam classification
 - Classification
 - Regression
 - Ranking

- Discover structure in unlabeled data
- E.g., Document clustering
 - Clustering
 - Hidden Markov Models

- Learning by "doing" with delayed reward
- ➤ E.g., Chess computer

Supervised Learning

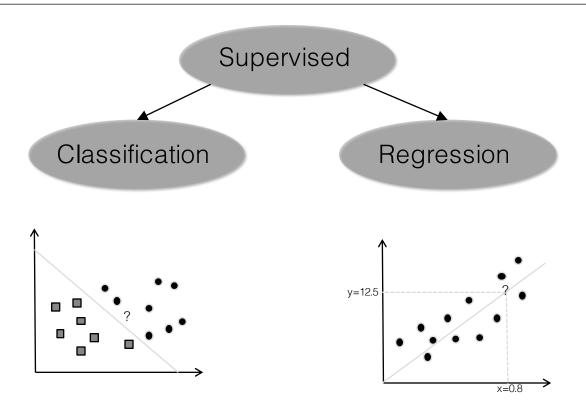
Given examples of a function (X, F(X))

Predict function F(X) for new examples X

Discrete F(X): Classification

Continuous F(X): Regression

F(X) = Probability(X): Probability estimation



Regression and Classification Examples

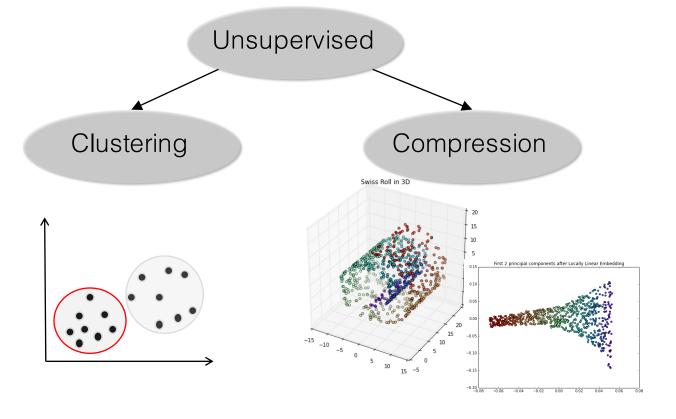
Stock prediction

- * Predict the price of a stock (y)
 - * Depends on x =
 - Recent history of stock price
 - News events
 - Related commodities

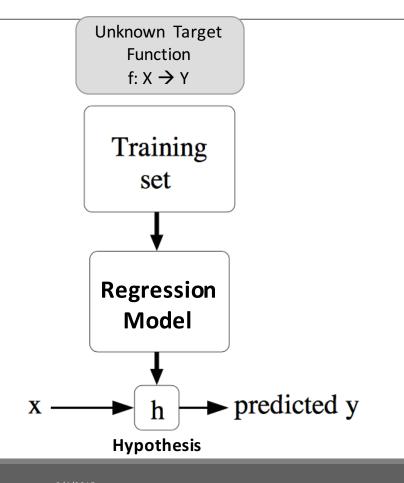
Spam or Not spam emails

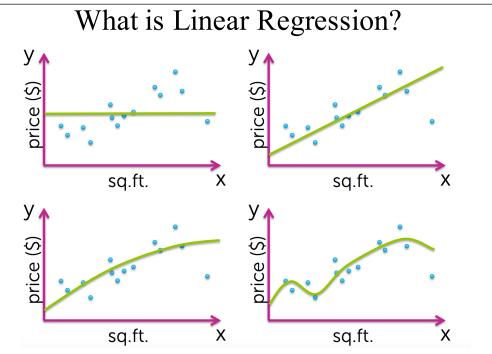
Music or Tweeter Sentiment Analysis

Unsupervised Learning



Linear Regression Learning Model

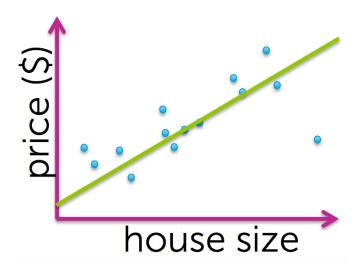




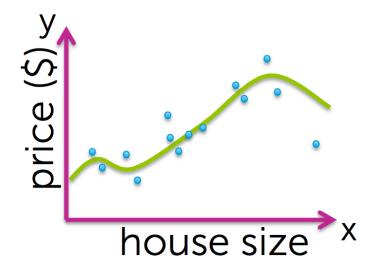
Simple Regression vs. Multiple Regression

What makes it simple?

1 input and just fit a line to data

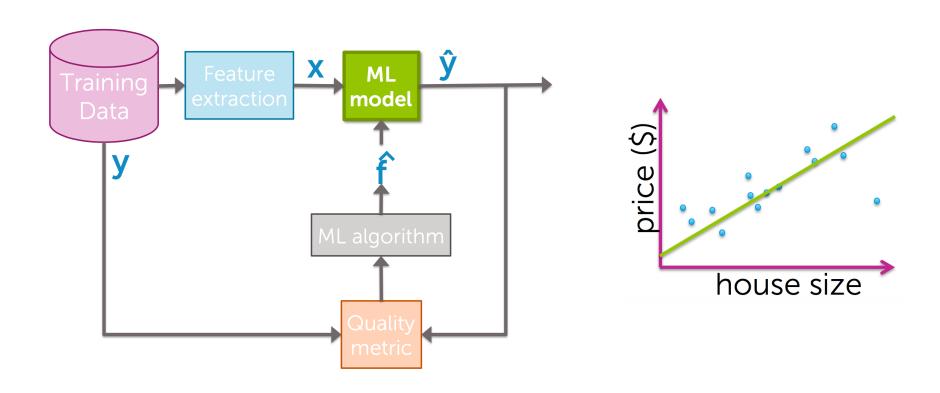


Fit more complex relationships than just a line



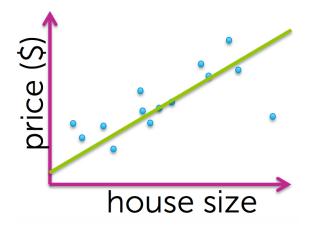
2/4/2015 PHD THESIS PROPOSAL

Regression Model



Predicting house prices

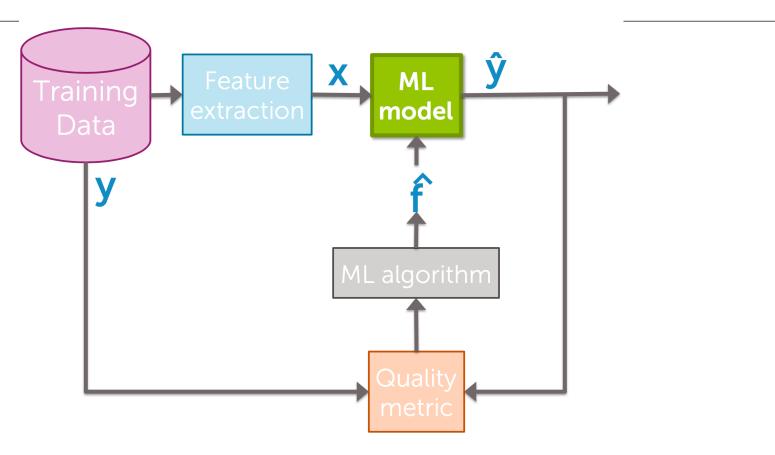
How much is my house worth?



Look at recent sales in my neighborhood

• How much did they sell for?

Regression Model



2/4/2015

Regression Algorithm

We will need a starting value for the slope and intercept, a step_size and a tolerance initial_intercept = 0, nitial_slope = 0, step_size = 0.05, tolerance = 0.01

The algorithm

In each step of the gradient descent we will do the following:

- 1. Compute the predicted values given the current slope and intercept
- 2. Compute the prediction errors (prediction Y)
- 3. Update the intercept:

compute the derivative: sum(errors)

compute the adjustment as step size times the derivative

decrease the intercept by the adjustment

4. Update the slope:

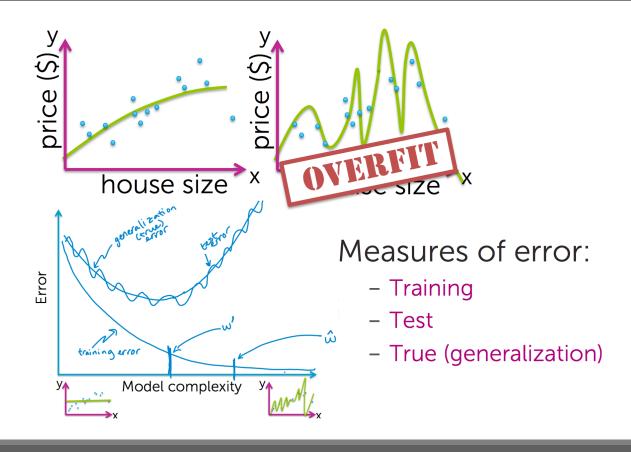
compute the derivative: sum(errors*input)

compute the adjustment as step_size times the derivative

decrease the slope by the adjustment

- 5. Compute the magnitude of the gradient
- 6. Check for convergence

Assessing Performance



Real world examples

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X = \# hours of exercise in a week \rightarrow Y = body mass index
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Y = # hours of studying in a week \rightarrow Y = grade for the course

Χ

Υ

Objective/Error/Cost Function

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Thank you

Question #1

What is an appropriate application of linear regression?

- 1) Predicting a person's height, given the amounts of certain substances in their diet
- 2) Predicting whether or not a person likes cake, given the amounts of certain substances in their diet

Question #2

What would lead to a negative r-squared?

- 1) We predict the opposite of the target value
- 2) We predict the negative of the target value
- 3) We predict worse than the mean of the target values

Assignment #1 - closed form example

Computing regression parameters The data

Consider the following 5 point synthetic data set:

X: 0,1,2,3,4

Y: 1,3,7,13,21

We want the line that "best fits" this data set as measured by residual sum of squares

In summary, we have:

slope = 5, intercept = -1

Finally we can add the line to our plot from above