

Lecture 17: Machine Learning Review

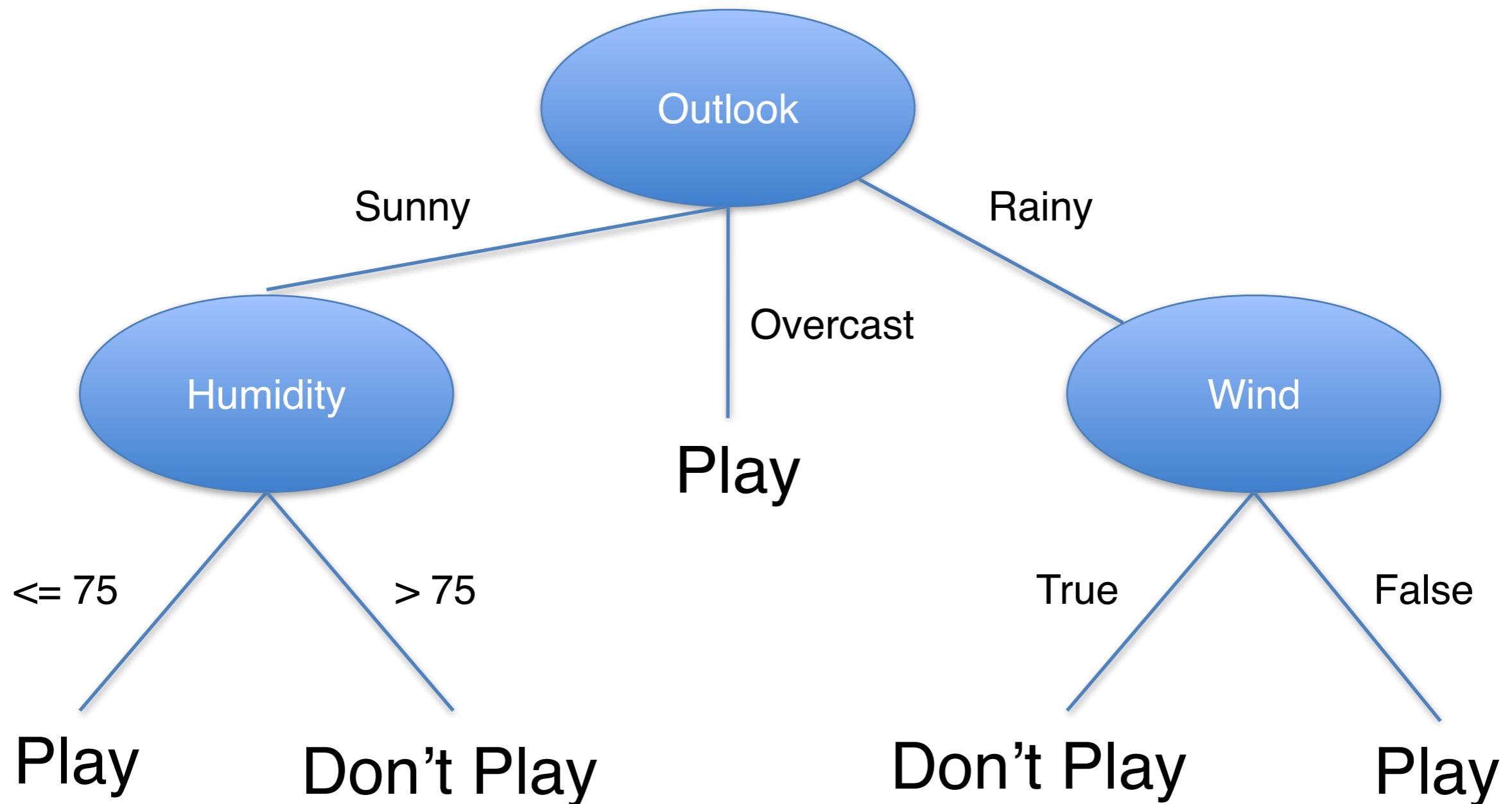
Core 109S IDWT?, Spring 2017

Michael Hay

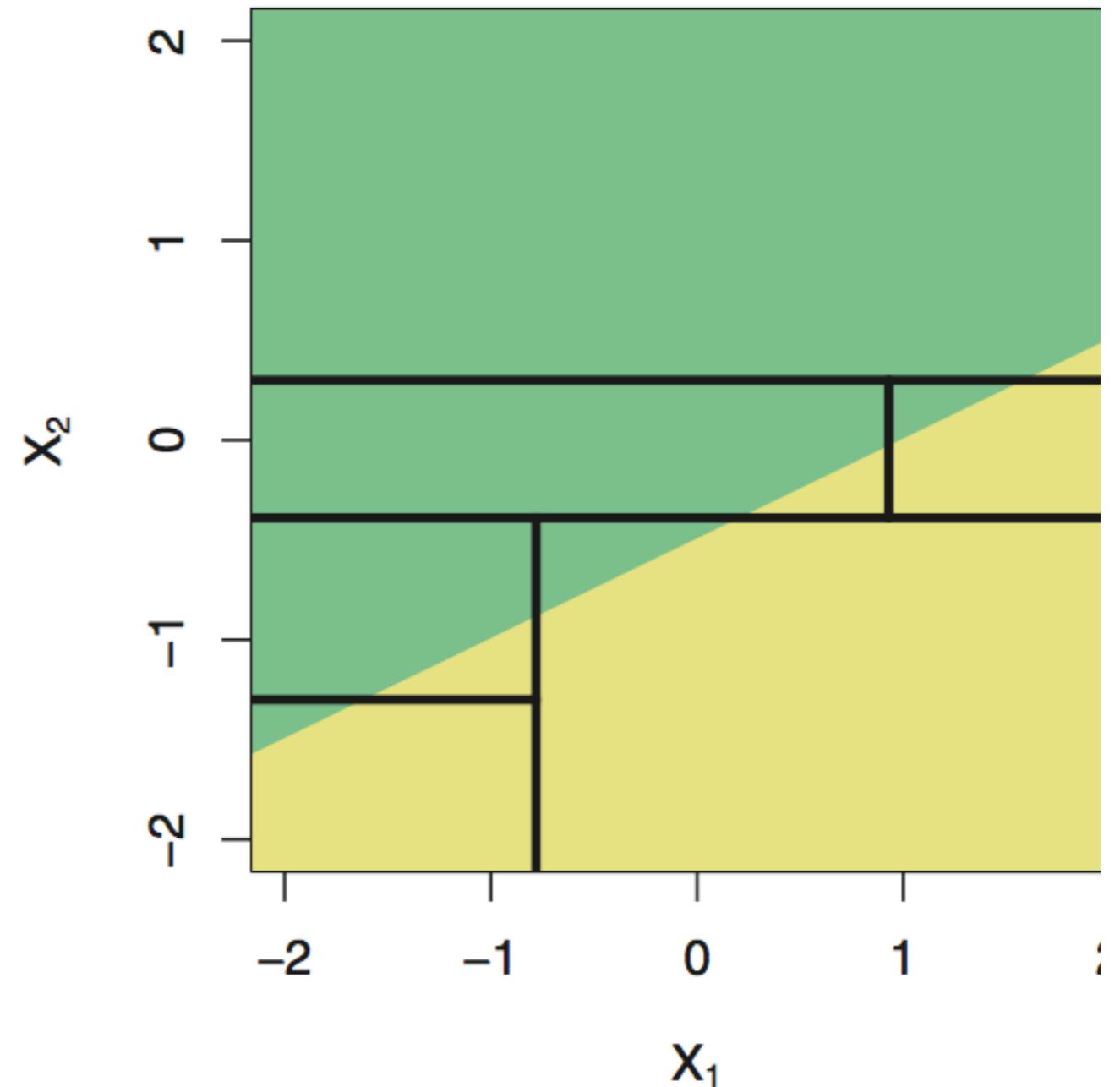
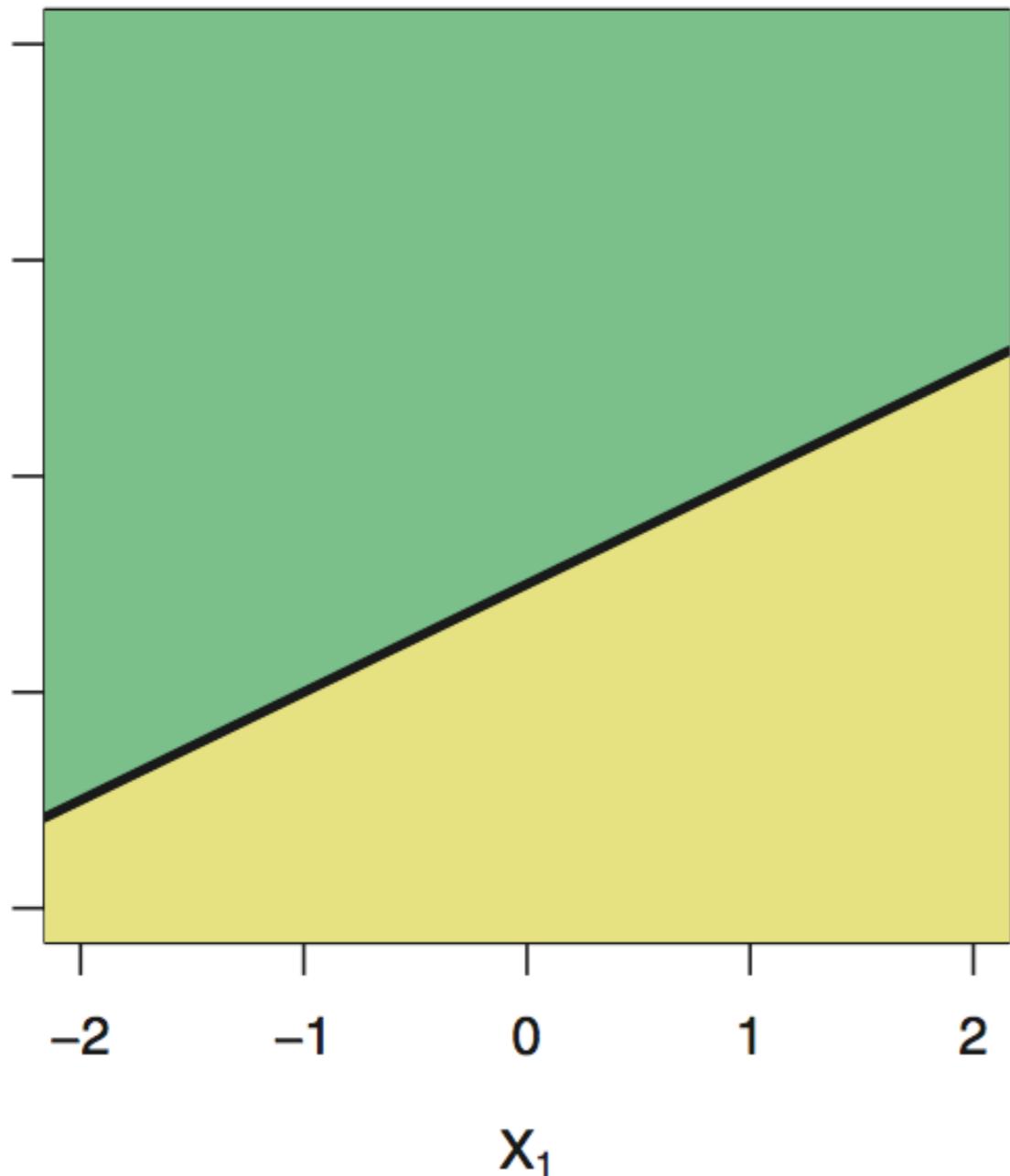
Today

- Review what we've learned so far about
 - Decision trees
 - Perceptrons (linear discriminants)
- Review is oriented through *exercises*
 - Work in groups, take your own notes, ask me questions as we go.
 - We will review answer to each question
- Wednesday: look at one more subtle but important issue with machine learning (overfitting)

“Is it a good day to play golf?”



Linear vs. trees



True boundary is linear; tree can only approximate with axis-parallel splits

Complete Q1 on
Worksheet

Example: perceptron

- Given input $x = (x_1, \dots, x_d)$ (*the d attributes of customer app.*)
- Assign a **weight** w_i for each attribute value x_i . and choose some **threshold** value

Approve credit if $\sum_{i=1}^d w_i x_i \geq \text{threshold}$

Deny credit if $\sum_{i=1}^d w_i x_i < \text{threshold}$

Writing as function $h(x)$

$$h(x) = \text{sign} \left(\left(\sum_{i=1}^d w_i x_i \right) - \text{threshold} \right)$$

sign(x) = *1 if $x \geq 0$*
 -1 if $x < 0$

$$= \text{sign} \left(\left(\sum_{i=1}^d w_i x_i \right) + w_0 \right)$$

(rename "threshold" as w_0)

$$= \text{sign} \left(\left(\sum_{i=1}^d w_i x_i \right) + w_0 x_0 \right)$$

(introduce artificial attribute $x_0=1$)

$$= \text{sign} \left(\sum_{i=0}^d w_i x_i \right)$$

Complete Q2 on
Worksheet

Algorithm 1 Algorithm for learning weights of a perceptron

```
1: procedure PERCEPTRONLEARNER(training data)
2:   Training data: collection of  $(x, y)$  pairs where  $x = (x_1, \dots, x_d)$  and  $y = +1$  or  $y = -1$ .
3:   Initialize  $w_0 = w_1 = \dots = w_d = 0$ 
4:   repeat
5:      $\triangleright$  Pick a misclassified example:
6:     Find some  $(\textcolor{blue}{x}, \textcolor{blue}{y})$  pair such that  $h(x) = \text{sign} \left( \sum_{i=0}^d \textcolor{red}{w}_i x_i \right) \neq \textcolor{blue}{y}$ 
7:      $\triangleright$  Update the weight vector:
8:     for  $i = 1$  to  $d$  do
9:       Set  $\textcolor{red}{w}_i \leftarrow \textcolor{red}{w}_i + \textcolor{blue}{y} \times \textcolor{blue}{x}_i$ 
10:      end for
11:      until There are no more misclassified points (or some other stopping criterion)
12:      Return weights  $w_0, \dots, w_d$ .
13: end procedure
```

Intuition:

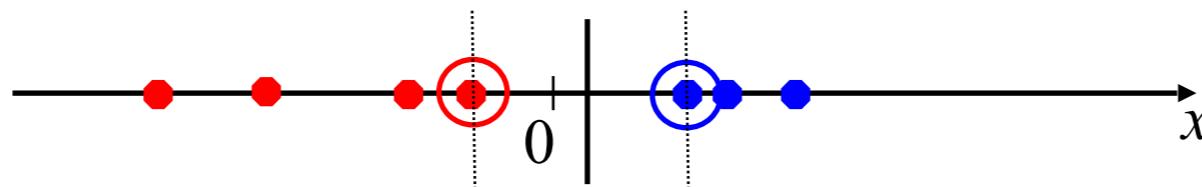
- Suppose Bob makes \$15K in salary
- Perceptron incorrectly denies Bob credit
 - $h(\text{Bob}) = -1$ but $y = 1$
- Weight update will *increase* the weight on salary, and therefore *increase* $h(\text{Bob})$

Complete Q3 on
Worksheet

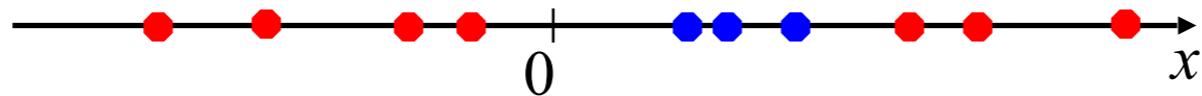
Complete Q4 on
Worksheet

Non-Linear Separators

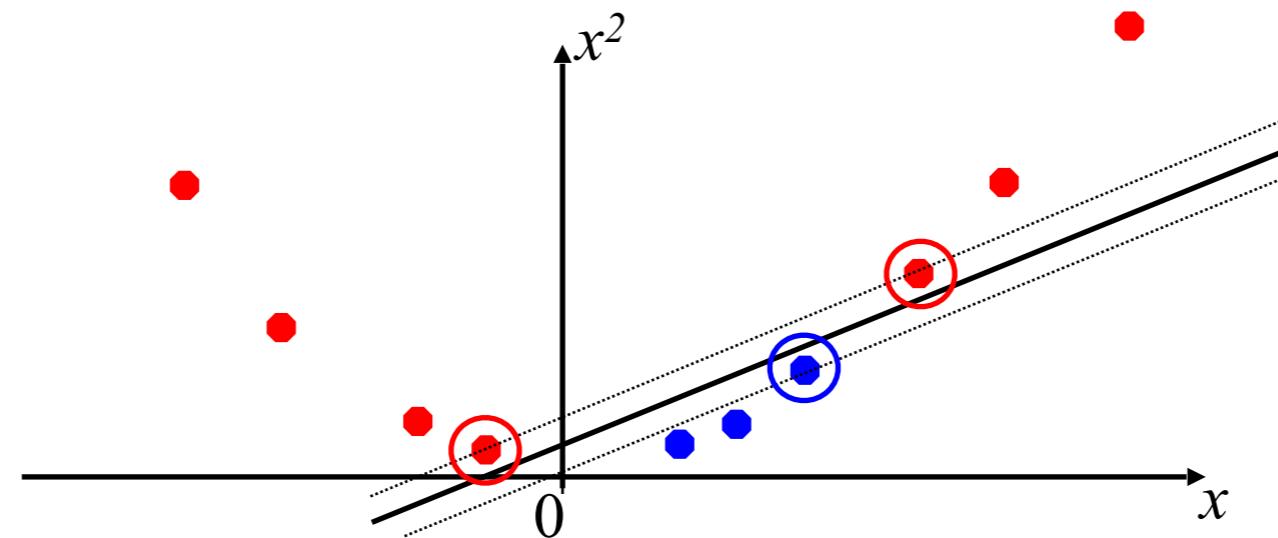
- Data that is linearly separable (with some noise) works out great:



- But what are we going to do if the dataset is just too hard?



- How about... mapping data to a higher-dimensional space:



Feature engineering

- A perceptron (aka linear discriminant) finds a line in feature space that separates the data into two groups.
- But we are free to *add* new features!
- While data may not be linearly separable in "original" feature space, it may be separable in augmented feature space.
- In general, feature engineering is a key component of ML.

Example #2: Spam Filter

- Input: email
- Predict: "spam" or "ham"
- Data: old emails, each labeled "spam" or "ham"
- **Features:** The attributes used to make the ham / spam decision
 - Words: FREE!
 - Text Patterns: \$dd, CAPS, spelling errors, ...
 - Non-text: SenderInContacts
 - ...



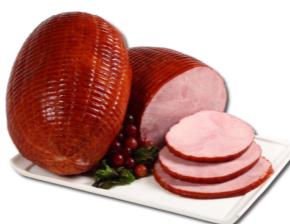
Dear Sir.

First, I must solicit your confidence in this transaction, this is by virtue of its nature as being utterly confidential and top secret. ...



TO BE REMOVED FROM FUTURE MAILINGS, SIMPLY REPLY TO THIS MESSAGE AND PUT "REMOVE" IN THE SUBJECT.

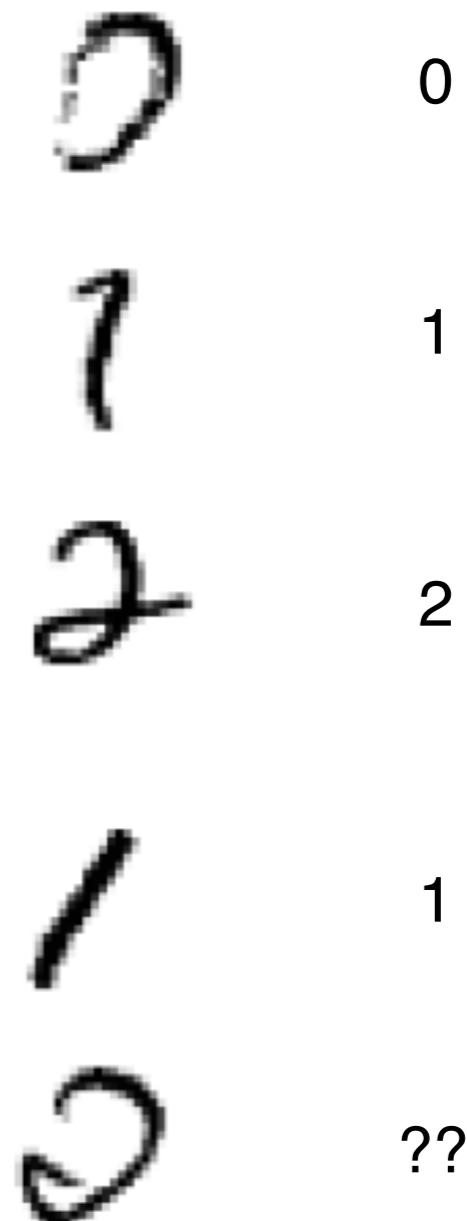
99 MILLION EMAIL ADDRESSES FOR ONLY \$99



Ok, I know this is blatantly OT but I'm beginning to go insane. Had an old Dell Dimension XPS sitting in the corner and decided to put it to use, I know it was working pre being stuck in the corner, but when I plugged it in, hit the power nothing happened.

Example #3: Digit Recognition

- Input: images / pixel grids
- Predict: which digit 0-9
- Data: images of hand-written digits, each labeled
- **Features:** The attributes used to make the digit decision
 - Pixels: (6,8)=ON
 - Shape Patterns: NumComponents, AspectRatio, NumLoops, Vertical/Horizontal Symmetry
 - ...



Complete Q5 and Q6
on Worksheet