

# Lecture 10 – Multi-Class Classification

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October 15, 2012

# Today

- ▶ Decision-trees odds and ends
  - ▶ Pruning
  - ▶ Why entropy works
- ▶ Multi-class classification

# A Real Classification Problem

Classify handwritten digits.

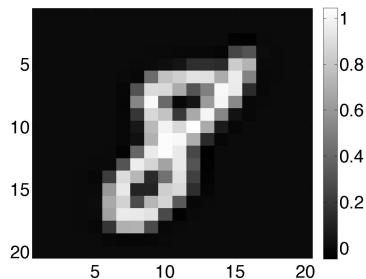


$$y \in \{0, 1, 2, 3, 4, 5, 6, 7, 8, 9\}$$

We don't know how to solve this yet

# Hand-written digit classification

Input:  $20 \times 20$  grayscale image



Unroll the image into a vector

$$\begin{bmatrix} x_1 & x_{21} & \dots & x_{381} \\ x_2 & x_{22} & \dots & x_{382} \\ & & \vdots & \\ x_{20} & x_{40} & \dots & x_{400} \end{bmatrix}$$

Feature vector  $\mathbf{x} \in \mathbb{R}^{400}$

$$\mathbf{x} = (x_1, \dots, x_{400})^T$$

# Multi-class Classification

Input:  $\mathbf{x} \in \mathbb{R}^d$  (continuous or discrete)

Labels:  $y \in \{1, \dots, K\}$

Group 1: Decision Trees

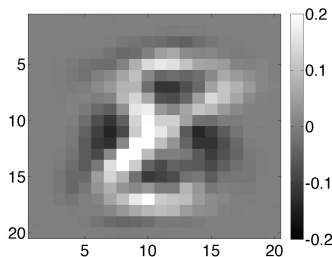
Group 2: Logistic regression

(Use one or more binary ( $y \in \{0, 1\}$ ) classifiers)

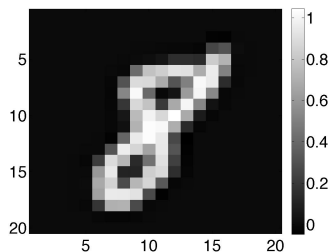
Hint: think about prediction first, then training.

# Visualization

Format weight vector as an image:



$\mathbf{w}_8$



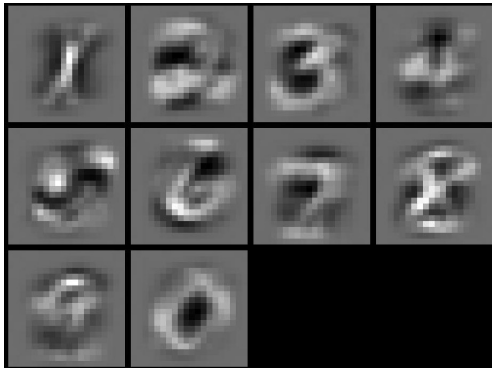
$\mathbf{x}$

Recall that

$$\text{Prediction} = \begin{cases} 1 & \mathbf{w}^T \mathbf{x} \geq 0 \\ 0 & \mathbf{w}^T \mathbf{x} < 0 \end{cases}$$

Dot product = multiply together corresponding pixels and add

## Visualization: One vs. All



## Visualization: One vs. One

