



# Computational Linguistics

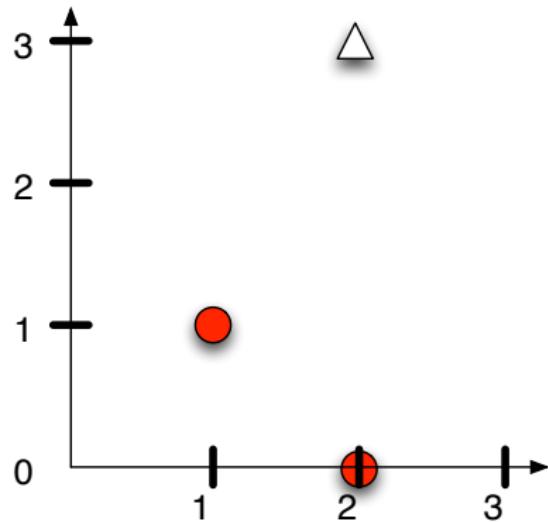
Computational Linguistics: Jordan Boyd-Graber  
University of Maryland  
**CLASSIFICATION EXAMPLES**

Slides adapted from Tom Mitchell, Eric Xing, and Lauren Hannah

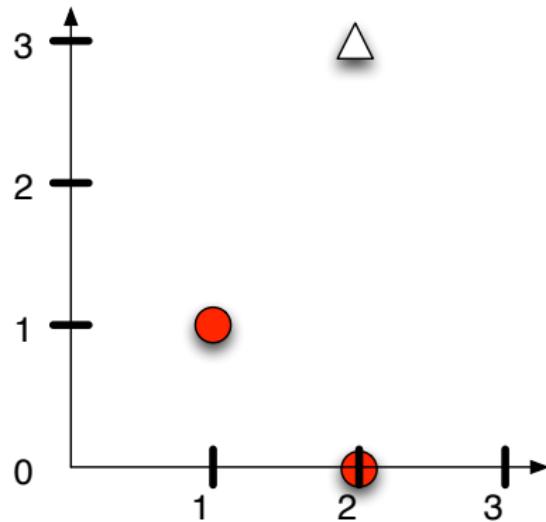
## Content Questions

## Administrative Questions

## Find the maximum margin hyperplane



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Which are the support vectors?

## Walkthrough example: building an SVM over the data shown

Working geometrically:

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- Remember that prediction has to be  $\pm 1$  for support vectors

$$w_1 + w_2 + b = -1 \quad (1)$$

$$\frac{3}{2}w_1 + 2w_2 + b = 0 \quad (2)$$

$$2w_1 + 3w_2 + b = +1 \quad (3)$$

## Walkthrough example: building an SVM over the data shown

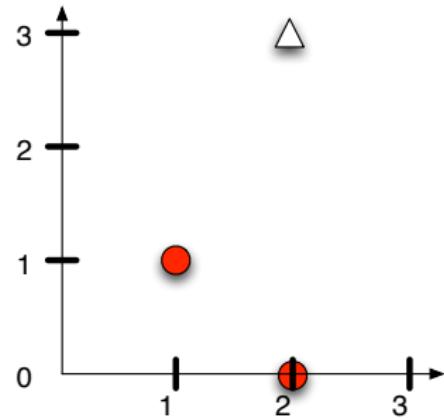
Working geometrically:

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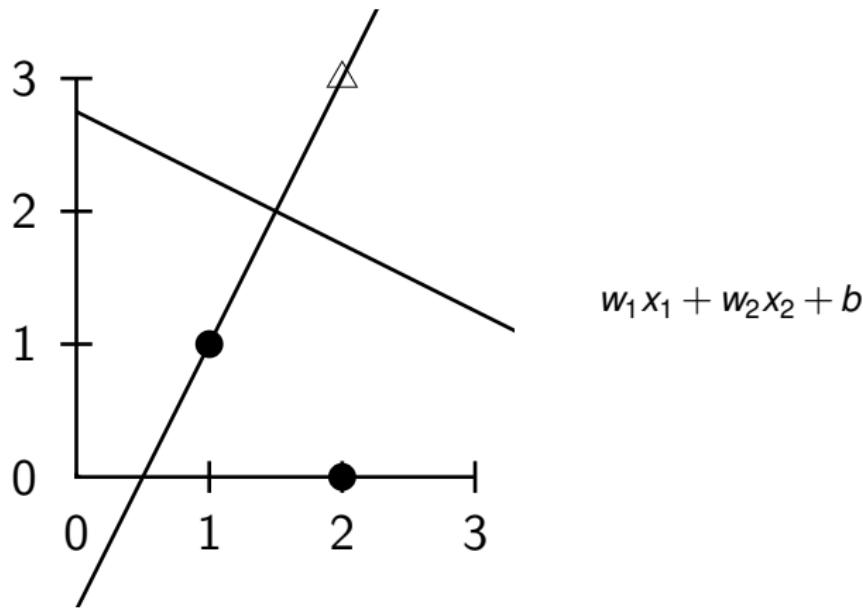
$$2w_1 + 3w_2 + b = +1 \quad (3)$$



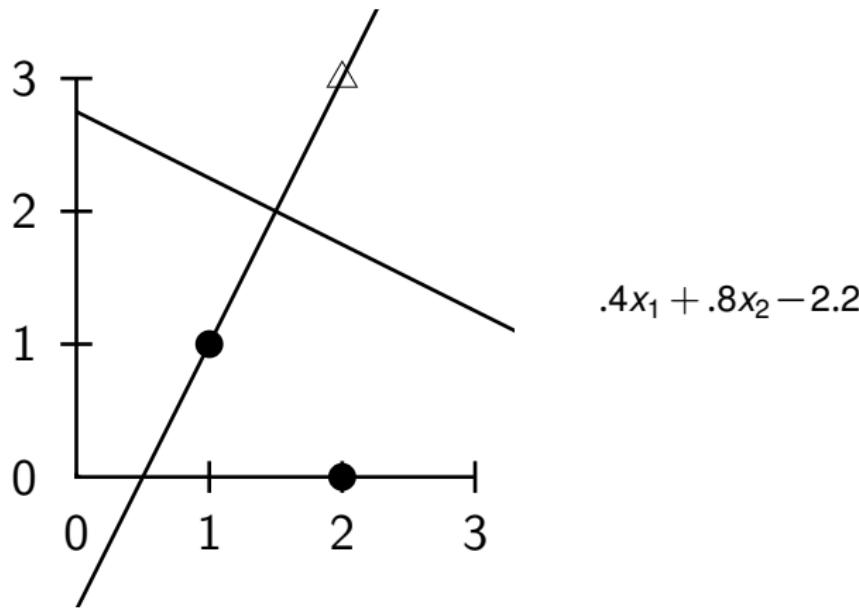
The SVM decision boundary is:

$$0 = \frac{2}{5}x + \frac{4}{5}y - \frac{11}{5}$$

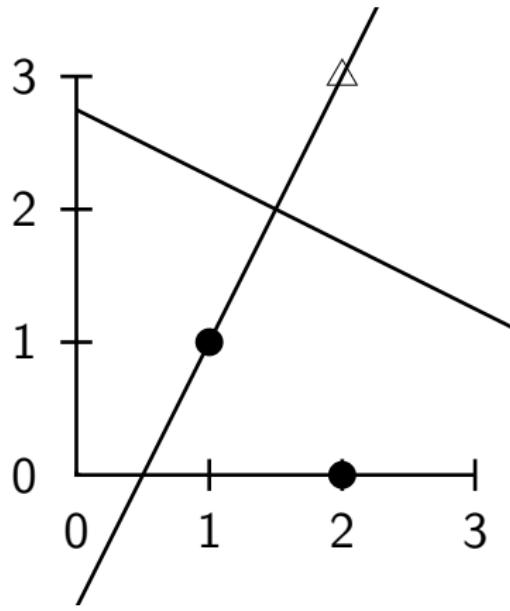
## Cannonical Form



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$$.4x_1 + .8x_2 - 2.2$$

- $.4 \cdot 1 + .8 \cdot 1 - 2.2 = -1$
- $.4 \cdot \frac{3}{2} + .8 \cdot 2 = 0$
- $.4 \cdot 2 + .8 \cdot 3 - 2.2 = +1$

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$$\frac{1}{\|w\|} = \frac{1}{\sqrt{\left(\frac{2}{5}\right)^2 + \left(\frac{4}{5}\right)^2}} = \frac{1}{\sqrt{\frac{20}{25}}} = \frac{5}{\sqrt{5}\sqrt{4}} = \frac{\sqrt{5}}{2} \quad (5)$$

## Reminder: Logistic Regression

$$P(Y = 0|X) = \frac{1}{1 + \exp[\beta_0 + \sum_i \beta_i X_i]} \quad (6)$$

$$P(Y = 1|X) = \frac{\exp[\beta_0 + \sum_i \beta_i X_i]}{1 + \exp[\beta_0 + \sum_i \beta_i X_i]} \quad (7)$$

- Discriminative prediction:  $p(y|x)$
- Classification uses: ad placement, spam detection
- What we didn't talk about is how to learn  $\beta$  from data

## Logistic Regression: Objective Function

$$\mathcal{L} \equiv \ln p(Y|X, \beta) = \sum_j \ln p(y^{(j)} | x^{(j)}, \beta) \quad (8)$$

$$= \sum_j y^{(j)} \left( \beta_0 + \sum_i \beta_i x_i^{(j)} \right) - \ln \left[ 1 + \exp \left( \beta_0 + \sum_i \beta_i x_i^{(j)} \right) \right] \quad (9)$$

## Algorithm

1. Initialize a vector  $B$  to be all zeros
2. For  $t = 1, \dots, T$ 
  - For each example  $\vec{x}_i, y_i$  and feature  $j$ :
    - Compute  $\pi_i \equiv \Pr(y_i = 1 | \vec{x}_i)$
    - Set  $\beta[j] = \beta[j]' + \lambda(y_i - \pi_i)x_i$
3. Output the parameters  $\beta_1, \dots, \beta_d$ .

## Example Documents

$$\beta[j] = \beta[j] + \lambda(y_i - \pi_i)x_i$$

$$\vec{\beta} = \langle \beta_{bias} = 0, \beta_A = 0, \beta_B = 0, \beta_C = 0, \beta_D = 0 \rangle$$

$y_1 = 1$

A A A A B B B C

(Assume step size  $\lambda = 1.0$ .)

$y_2 = 0$

B C C C D D D D

You first see the positive example. First, compute  $\pi_1$

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$\pi_1 = 0.5$  What's the update for  $\beta_{bias}$ ?

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$y_2 = 0$

B C C C D D D D

(Assume step size  $\lambda = 1.0$ .)

What's the update for  $\beta_D$ ?

$$\beta_D = \beta'_D + \lambda \cdot (y_2 - \pi_2) \cdot x_{2,D} = 0.0 + 1.0 \cdot (0.0 - 0.97) \cdot 4.0 = -3.88$$