

# Classification problems

## Examples:

- ❖ Text categorization
- ❖ Image recognition

Ref.: L. Bottou, F. E. Curtis, and J. Nocedal, "Optimization methods for large-scale machine learning," SIAM Review, vol. 60, no. 2, pp. 223–311, 2018

# A general setup

**$K$  categories**

Input data:  $\{(z_i, c_i)\}, i = 1, 2, \dots, n$

$z_i \in R^D$  - vector of values

$c_i \in \{1, 2, \dots, K\}$  - label

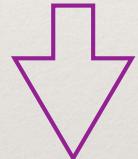
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## **Two classes: $j$ and NOT $j$**

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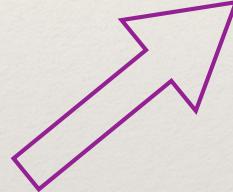
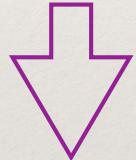
$c_i \in \{1, 2, \dots, K\}$  - label

## Feature Space

Input data:  $\{(x_i, y_i)\}, i = 1, 2, \dots, n$

$x_i = \phi(z_i) \in R^d$  - vector of values

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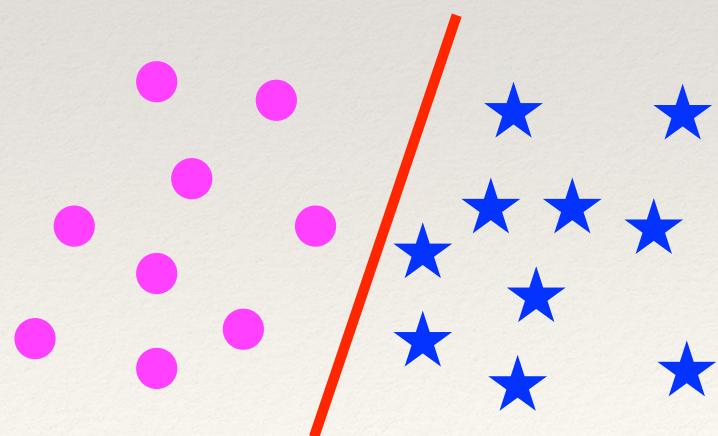
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## **Hope: separable by a hyperplane**



# Text categorization

D. D. Lewis, Y. Yang, T. G. Rose, and F. Li, “RCV1: A new benchmark collection for text categorization research,” *Journal of Machine Learning Research*, vol. 5, pp. 361–397, 2004

Reuters Corpus Volume 1 (RCV1):  
manually categorized archive of news stories

- ❖ Over 800,000 stories
- ❖ Most stories < 1000 words (less than two A4 pages)
- ❖ Feature space is defined by a vocabulary of 47,152 words

*Apart from the terrible memories this stirs up for me personally  
(coding stories through the night etc.),*

*I can't find fault with your account.*

– Reuters editor commenting on a draft of section 2 of this paper.

# Statement of problem

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Find a “good” hyperplane     $w^\top x - b = 0$     such that

$$\begin{cases} w^\top x_i - b > 0, & y_i = 1, \\ w^\top x_i - b < 0, & y_i = -1. \end{cases}$$

*Prediction function:*     $h(x, w, b) := w^\top x - b$

Evaluate  $\text{sign}(h(x_i, w, b))$  to attribute  $x_i$  to category 1 or -1

# Support-vector machines

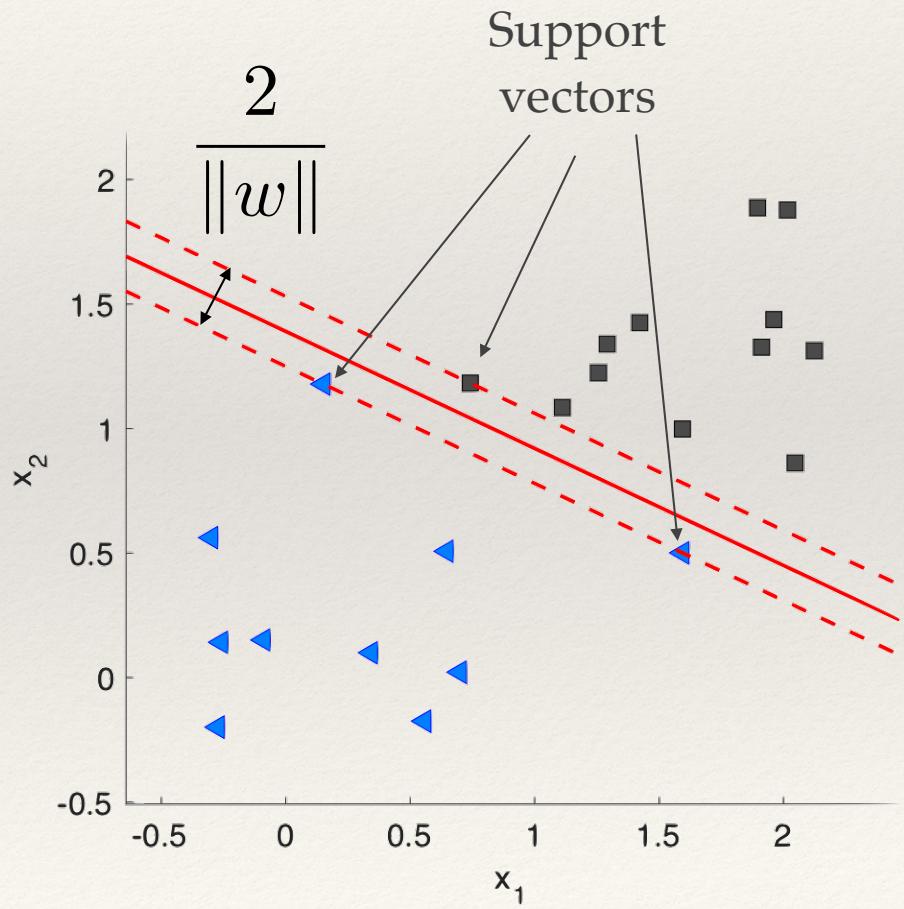
C. Cortes and V. Vapnik, “Support-vector networks,” Machine Learning, vol. 20, pp. 273–297, 1995

$$\begin{cases} w^\top x_i - b \geq 1, & y_i = 1, \\ w^\top x_i - b \leq -1, & y_i = -1. \end{cases}$$

$$\frac{1}{2} \|w\|^2 \rightarrow \min$$

subject to

$$y_i(w^\top x_i - b) \geq 1$$



# A smooth loss function

*Prediction function:*  $h(x, w, b) := w^\top x - b$

*Log-loss function:*  $l(h, y) := \log(1 + \exp[-yh(x_i, w, b)])$

*Minimization problem:*

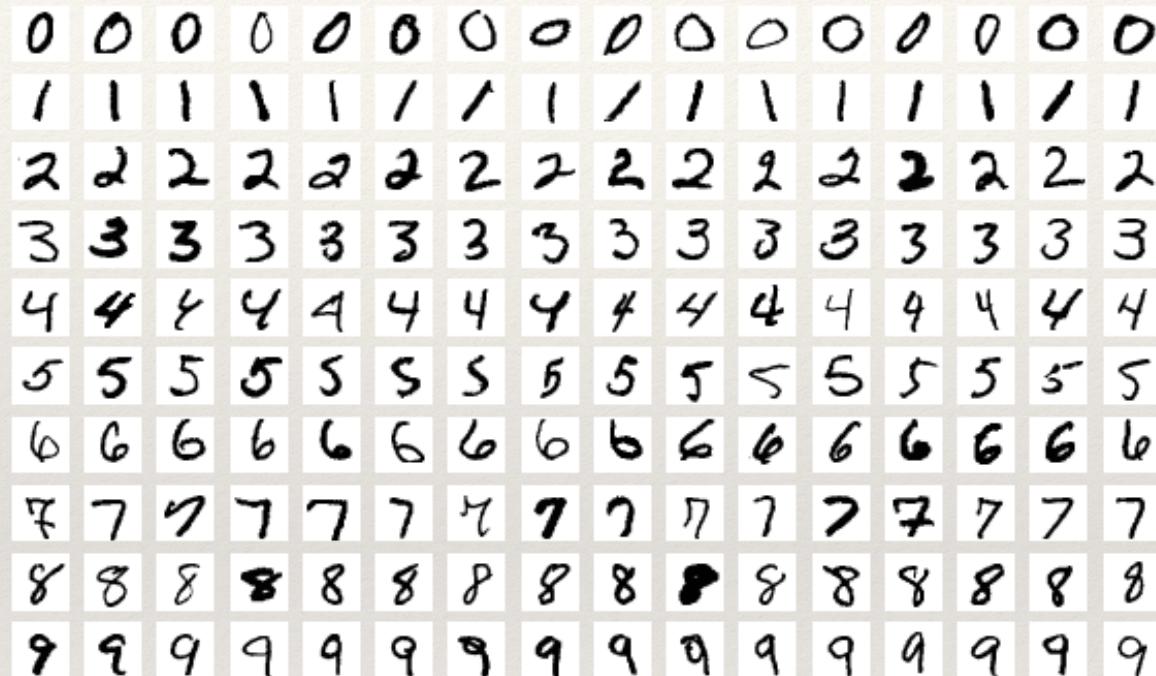
$$\min_{(w,b) \in \mathbb{R}^d \times \mathbb{R}} \frac{1}{n} \sum_{i=1}^n l[h(x_i, w, b), y_i] + \frac{\lambda}{2} \|w\|^2.$$

*positive parameter*

*Tikhonov  
regularization*

# Image recognition

Sample digit images from MNIST database of handwritten characters



Example from Bottou-Curtis-Nocedal



# Deep neural networks (DNNs)

Input:  $\{(x_i^{(0)}, y_i)\}, \quad i = 1, \dots, n$

Map to a feature space by a composition of functions:

$$x_i^{(j)} = s\left(W_j x_i^{(j-1)} + b_j\right) \in \mathbb{R}^{d_j}, \quad j = 1, \dots, J$$

$W_j$  = matrix,  $b_j$  = vector,

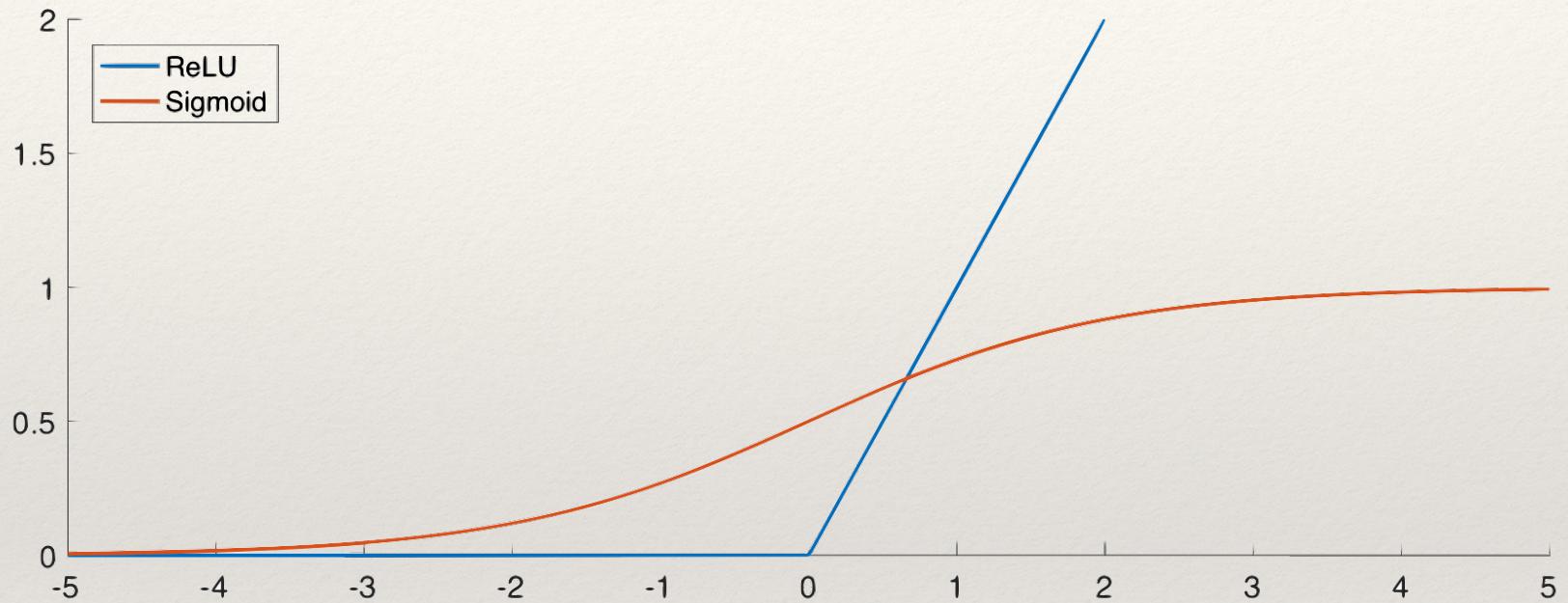
$s(\dots)$  = a nonlinear activation function

$J$  = the number of layers

Example with three layers, i.e.,  $J = 3$ :

$$x^{(3)} = s\left(W_3 s\left(W_2 s\left(W_1 x^{(0)} + b_1\right) + b_2\right) + b_3\right)$$

# Popular activation functions



Rectified Linear Unit (ReLU):

$$s(x) = \max\{0, x\}$$

Sigmoid:

$$s(x) = \frac{1}{1 + e^{-x}}$$

# Optimization problem for DNNs

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Vector of parameters:

$$\mathbf{w} := \{(W_1, b_1), (W_2, b_2), \dots, (W_J, b_J)\}$$

Minimization problem:

$$\min_{\mathbf{w} \in \mathbb{R}^d} \frac{1}{n} \sum_{i=1}^n l [h(x_i, \mathbf{w}), y_i] ,$$

$h(x_i, \mathbf{w})$       is a prediction function

$l(h(\cdot), y)$       is a loss function

# Examples of DNNs for image recognition

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- ❖ AlexNet (8 layers) (2012)
- ❖ ResNet (up to 152 layers) (residual NN, matlab) (2016)

Thanks to Avi Schwarzschild, AMSC PhD student