

Separating Hyperplanes (support vector machine)

LDA / QDA / NB $p(x)$ and $p(y|x)$

Logistic Regression

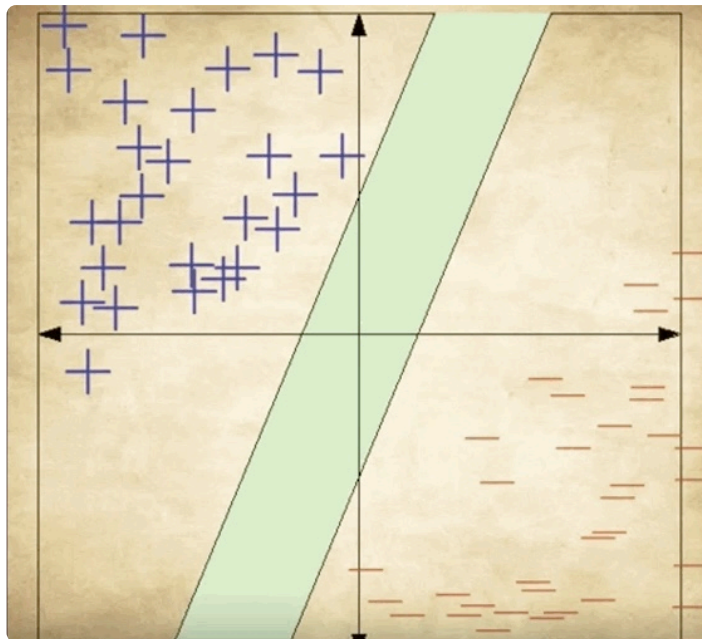
Support Vector Machine $w^T x + b \geq 0$

directly estimate $\theta = \begin{bmatrix} w \\ b \end{bmatrix}$

Vapnik's Principle:

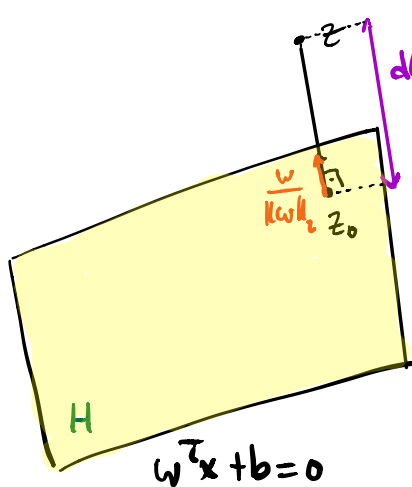
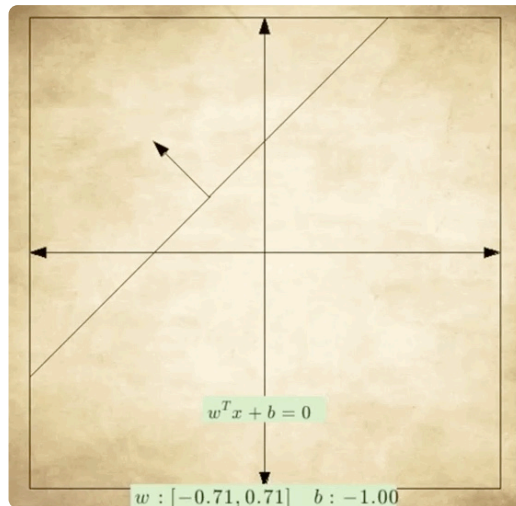
"When solving a problem of interest, do not solve a more general problem as an intermediate step."

"Don't solve a harder problem than you have to"



Hyperplane:

$$H = \{x : w^T x + b = 0\}$$



$$d(z, H) = |r|$$

$$z = z_0 + \frac{w}{\|w\|_2} \cdot r$$

↓

$$w^T z + b = \underbrace{w^T z_0 + b}_0 + \|w\|_2 \cdot r$$

$$w^T z + b = \|w\|_2 \cdot r$$

distance btw z and H : $d(z, H) = |r| = \frac{|w^T z + b|}{\|w\|_2}$

Let $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$ be training data $y_i \in \{-1, 1\}$

Data is linearly separable if $\exists w, \exists b$

$$y_i (w^T x_i + b) \geq 0 \quad \forall i = 1, \dots, n$$

$H = \{x: w^T x + b = 0\}$ is a separating hyperplane

Maximum Margin Hyperplane

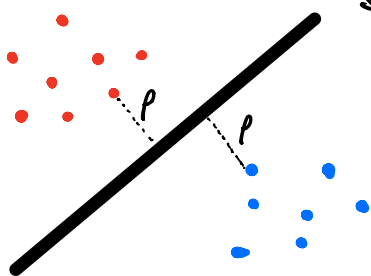
Margin ρ of a sep. hyperplane is

$$\rho(w, b) = \min_{i=1, \dots, n} d(x_i, H) = \min_{i=1, \dots, n} \frac{|w^T x_i + b|}{\|w\|_2}$$

Max. margin separating hyperplane is the solution of

$$\max_{w, b} \rho(w, b)$$

$$\text{s.t. } y_i(w^T x_i + b) \geq 0 \quad \forall i$$



not unique

$(\alpha w, \alpha b)$ gives the same hyperplane

Scale w and b by $\frac{1}{\min_{i=1, \dots, n} |w^T x_i + b|}$.

$$\text{Now } \rho(w, b) = \frac{1}{\|w\|_2}$$

$$\max_{w, b} \frac{1}{\|w\|_2}$$

$$= \max_{w, b} \frac{1}{\|w\|_2}$$

$$y_i(w^T x_i + b) \geq 1 \quad \forall i$$

$$y_i(w^T x_i + b) \geq 1 \quad \forall i$$

$y_i(w^T x_i + b) = 1$ for some i \leftarrow this constraint is automatically satisfied (we can drop it)

$$\min_{w, b} \frac{1}{2} \|w\|_2^2$$

$$y_i (w^T x_i + b) \geq 1 \quad \forall i$$

Hard-Margin
Support Vector Machine

Constrained Optimization Problem (Quadratic optimization with linear constraints).

$$\min \frac{1}{2} \|w\|_2^2 + C \cdot \frac{1}{n} \sum_{i=1}^n s_i$$

$$y_i (w^T x_i + b) \geq 1 - s_i \quad \forall i$$

$$s_i \geq 0 \quad \forall i \quad (\text{slack variables})$$

Soft-Margin
Support Vector Machine

C is a tuning parameter