

Support vector machines.



model-based learning.



$$x_i = \{x_1, \dots, x_n\}$$

$$w = \{w_1, \dots, w_n\}$$

$$\sum_i w_i x_i + b = \text{SCORE}$$

Geometric approach

Illustration:

2 features
2 classes

binary classific.

positive class

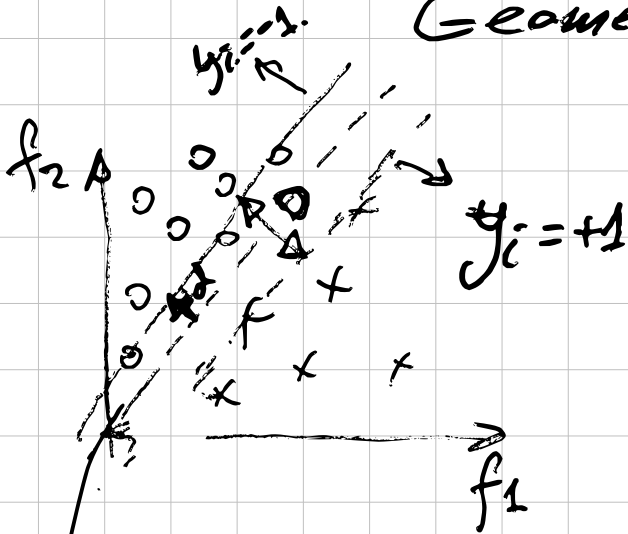
negative class

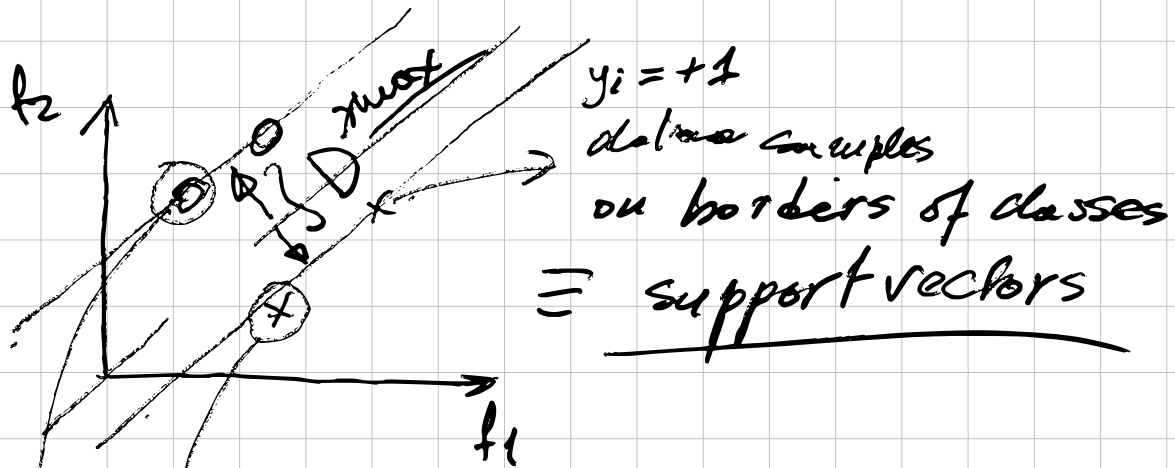
$y_i = +1$

$y_i = -1$

model
hyperplane in general
d is the distance from model to line delimiting a class

Data sample
Data point \equiv VECTOR of features





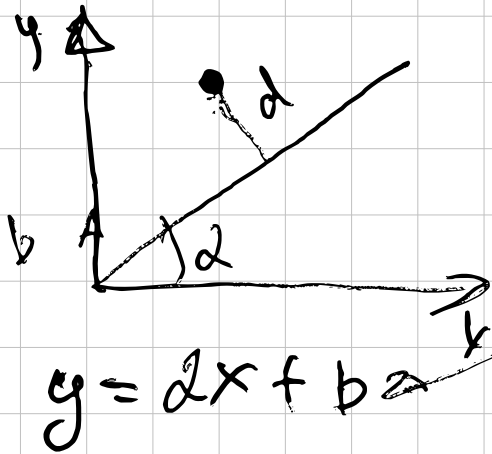
$$\left. \begin{aligned}
 w x_i + b &\geq +1 & \text{if } y_i = +1 \\
 w x_i + b &\leq -1 & \text{if } y_i = -1
 \end{aligned} \right\} \quad b)$$

$$y_i \cdot (w x_i + b) \geq 1$$

We want to max $D = 2 \times d$
where d is dist. from model
to support hyper plane

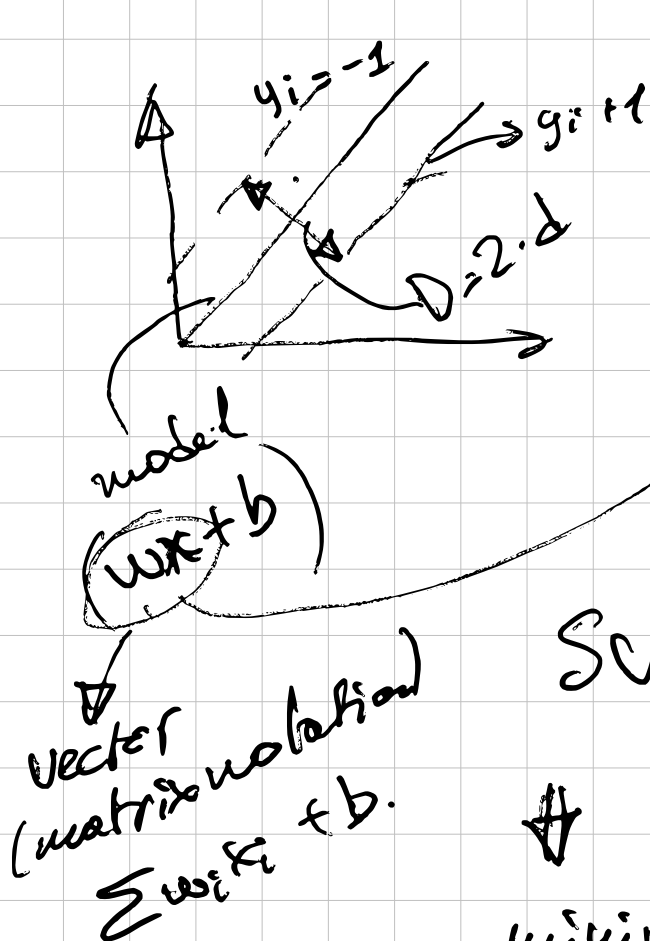
In 2D case hyper plane is line

from basic geometry



$$Ax + By + C = 0$$

$$d = \frac{|Ax + By + C|}{\sqrt{A^2 + B^2}}$$



$$d = \frac{|Ax + By + C|}{\sqrt{A^2 + B^2}}$$

$$d = \frac{|w x + b|}{\sqrt{\sum w_i^2}} = \frac{1}{\|w\|}$$

SVM: Word to maximize

$$\# \quad f() = 2 \cdot d = \frac{2}{\|w\|}$$

$$\text{minimize: } \frac{1}{f()} = \frac{1}{2} \|w\|$$

To make $f()$ convex: $\frac{1}{2} \|w\|^2$

Primary optimization formulation



$$f^* = \frac{1}{2} \|w\|^2$$

constraint:

$$y_i (w x_i + b) = 1$$

$$g: y_i (w x_i + b) - 1 = 0$$

Solve optimization problem using

Lagrangean method

$$\Rightarrow L = f^*(x) - 2 g(x)$$

$$+ \sum \lambda_i y_i$$

$$\text{w.o. w dot prod.} = \frac{1}{2} \|w\|^2 - \sum \lambda_i [y_i (x_i w + b) - 1]$$

$$\left[\begin{array}{l} \frac{\partial L}{\partial w} = 0 \\ \frac{\partial L}{\partial b} = 0 \end{array} \right. : \begin{array}{l} w - \sum d_i y_i x_i = 0 \\ -\sum d_i y_i = 0 \end{array}$$

\swarrow
 $w_i = \sum d_i y_i x_i$

Formulate secondary optimization problem # of samples

$$L^* = \sum \underline{d_i} - \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \underline{d_i y_i d_j y_j} (x_i \cdot x_j)$$

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When dual problem is solved

few $\alpha_i > 0$

$$\cancel{f} = \underline{w}x + b$$

$$\textcircled{w} = \sum \alpha_i y_i x_i$$

$$\textcircled{b} = y_i - w x_i$$

Multiclass
classif.

ϕ

one vs. one

ϕ vs
all

$$\frac{n(n-1)}{2} \text{ SVs}$$

