

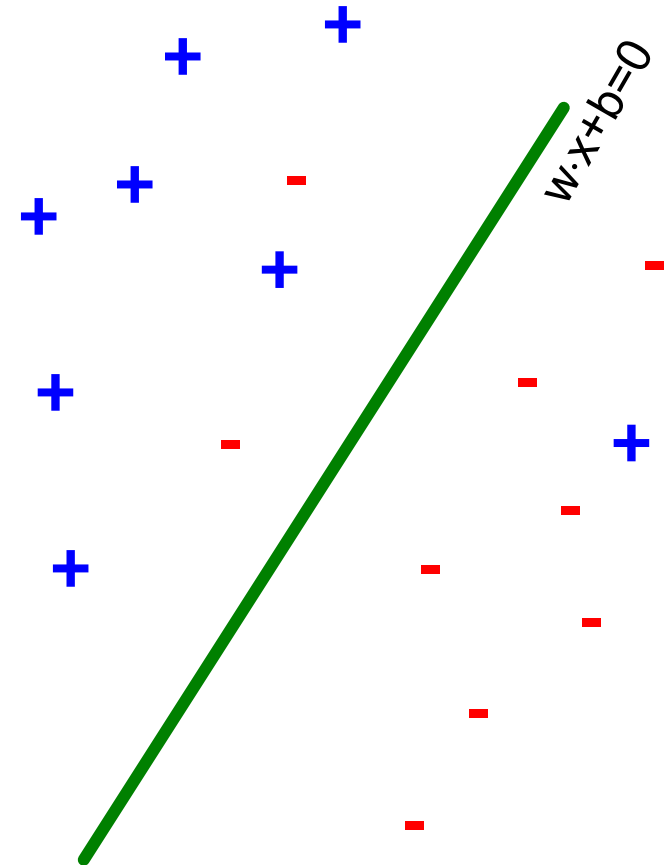
# Non linearly separable data?

- If data is **not separable** introduce **penalty**:

$$\min_w \frac{1}{2} \|w\|^2 + C \text{ (\#number of mistakes)}$$

$$s.t. \forall i, y_i (w \cdot x_i + b) \geq 1$$

- Minimize  $\|w\|^2$  plus the number of training mistakes
- Set  $C$  using cross validation
- **How to penalize mistakes?**
  - All mistakes are not equally bad!



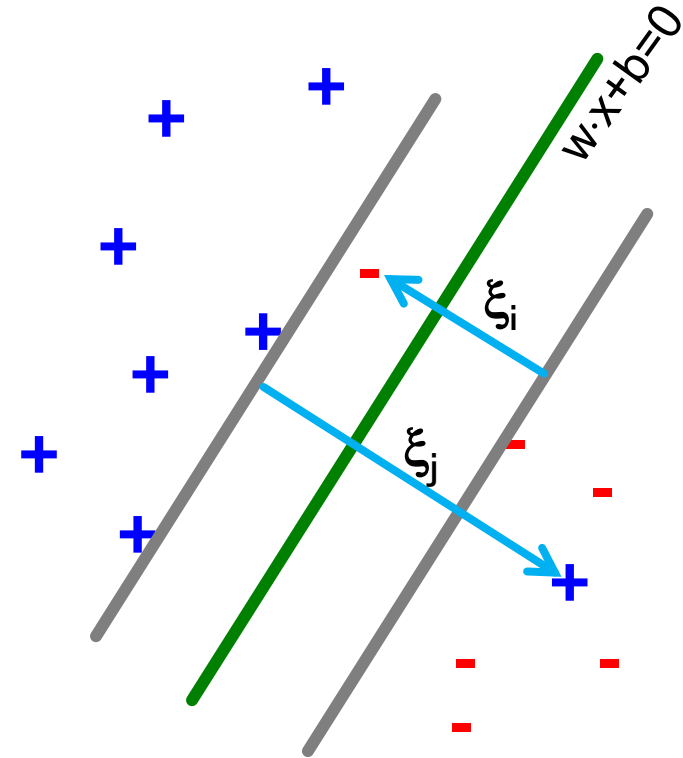
# Support Vector Machines

- Introduce **slack variables**  $\xi_i$

$$\min_{w, b, \xi_i \geq 0} \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i$$

$$s.t. \forall i, y_i (w \cdot x_i + b) \geq 1 - \xi_i$$

- If point  $\mathbf{x}_i$  is on the wrong side of the margin then get penalty  $\xi_i$



**For each datapoint:**

If margin  $\geq 1$ , don't care

If margin  $< 1$ , pay linear penalty

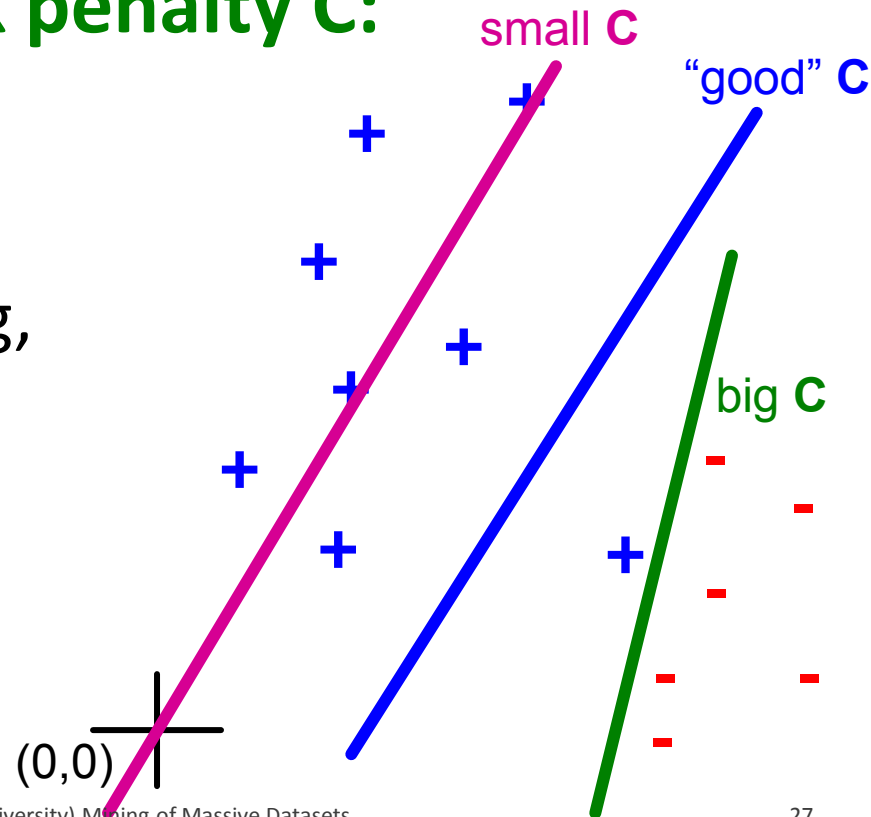
# Slack Penalty $C$

$$\min_w \frac{1}{2} \|w\|^2 + C \text{ (#number of mistakes)}$$

$$s.t. \forall i, y_i (w \cdot x_i + b) \geq 1$$

## ■ What is the role of slack penalty $C$ :

- $C=\infty$ : Only want to  $w, b$  that separate the data
- $C=0$ : Can set  $\xi_i$  to anything, then  $w=0$  (basically ignores the data)

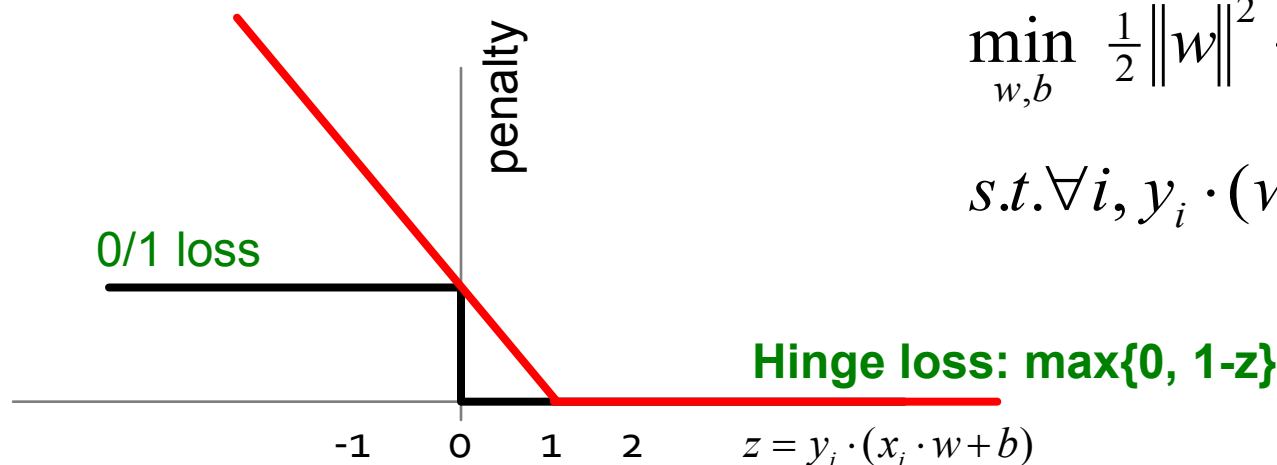


# Support Vector Machines

## ■ SVM in the “natural” form

$$\arg \min_{w,b} \underbrace{\frac{1}{2} w \cdot w}_{\text{Margin}} + \underbrace{C}_{\substack{\text{Regularization} \\ \text{parameter}}} \sum_{i=1}^n \underbrace{\max\{0, 1 - y_i(w \cdot x_i + b)\}}_{\text{Empirical loss } L \text{ (how well we fit training data)}}$$

## ■ SVM uses “Hinge Loss”:



$$\min_{w,b} \frac{1}{2} \|w\|^2 + C \sum_{i=1}^n \xi_i$$

$$s.t. \forall i, y_i \cdot (w \cdot x_i + b) \geq 1 - \xi_i$$