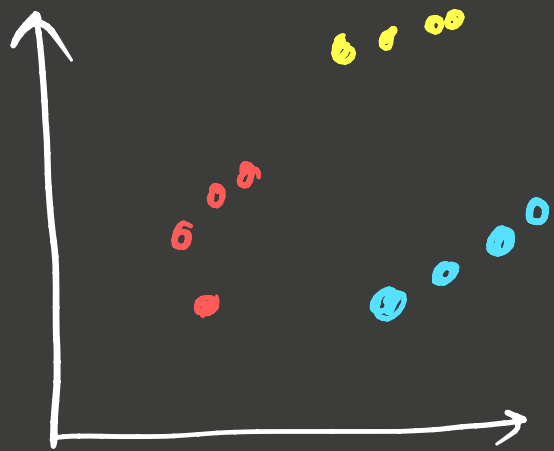
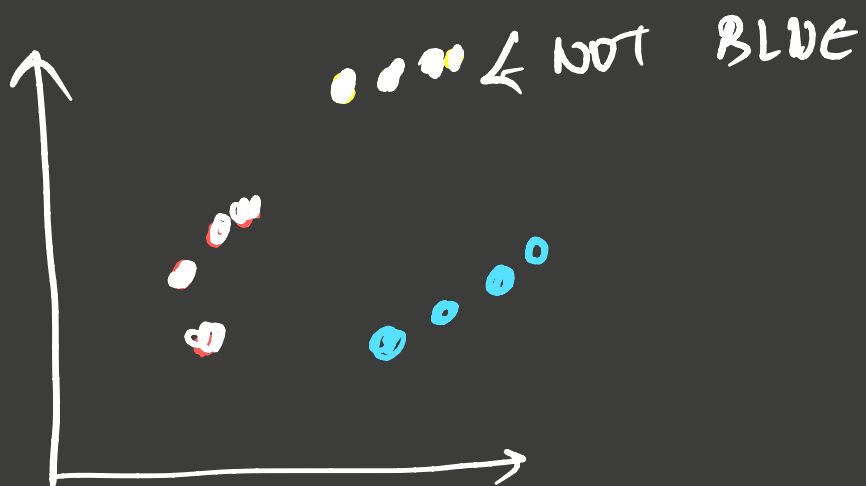


# SVM FOR MULTI-CLASS CLASSIFICATION

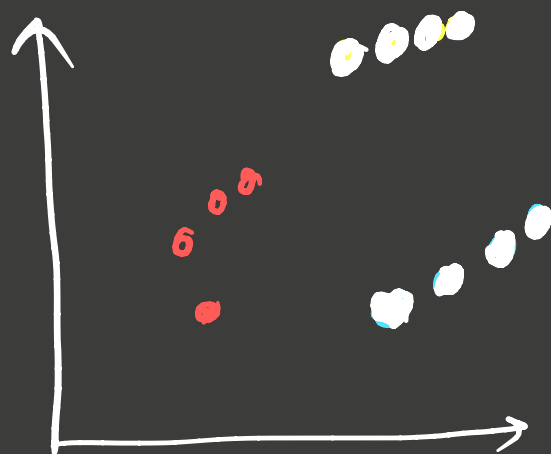
(1) ONE VS ALL



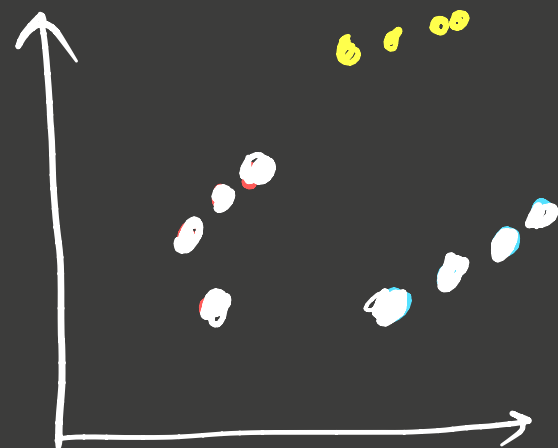
BLUE VS ALL



RED VS ALL

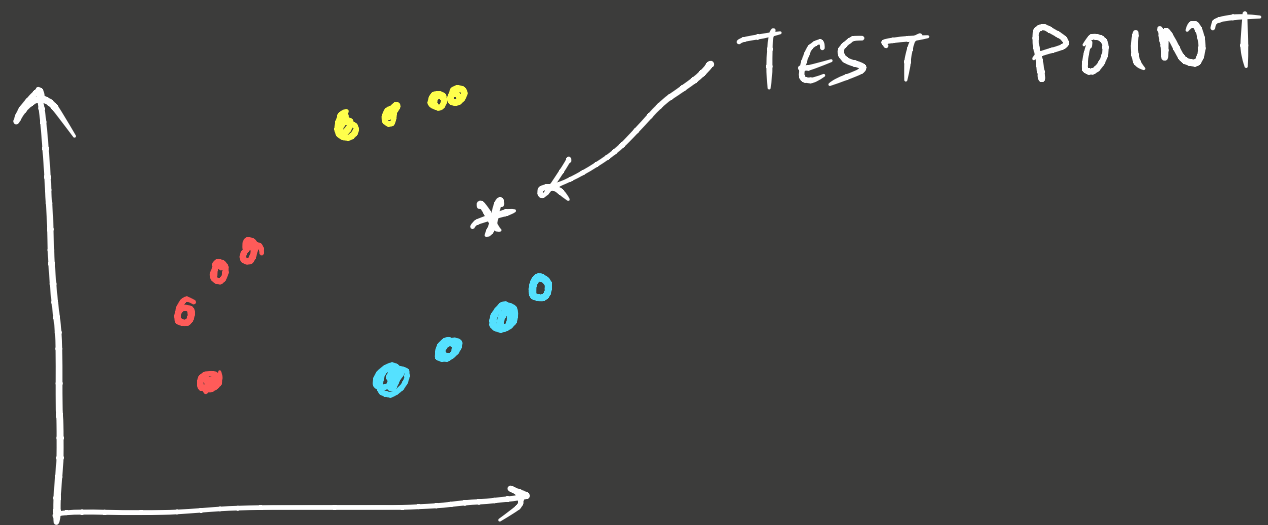


YELLOW VS ALL



# SVM FOR MULTI-CLASS CLASSIFICATION

① ONE VS ALL



BLUE VIS ALL :  $\bar{w} \cdot \bar{x}_{\text{TEST}} + b = 0.8$   
(+) (-)

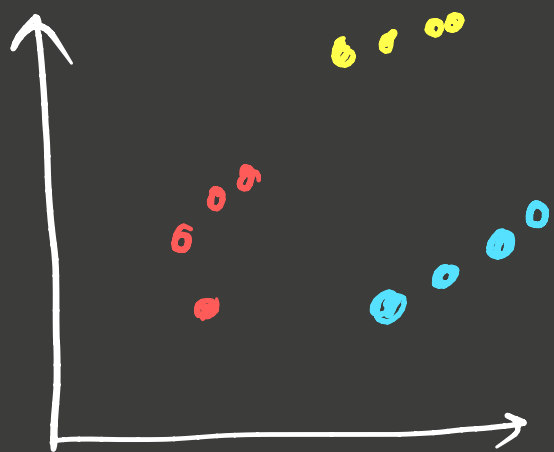
YELLOW VIS ALL :  $\bar{w} \cdot \bar{x}_{\text{TEST}} + b = 0.6$   
(+) (-)

RED VIS ALL :  $\bar{w} \cdot \bar{x}_{\text{TEST}} + b = -0.2$   
(+) (-)

} argmax = BLUE  
∴ TEST POINT IS BLUE CLASS

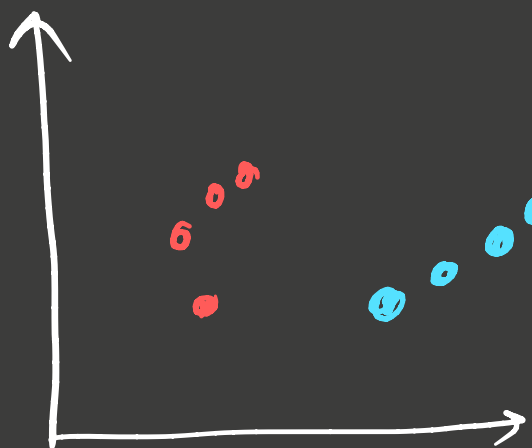
# SVM FOR MULTI-CLASS CLASSIFICATION

① ONE VS ONE

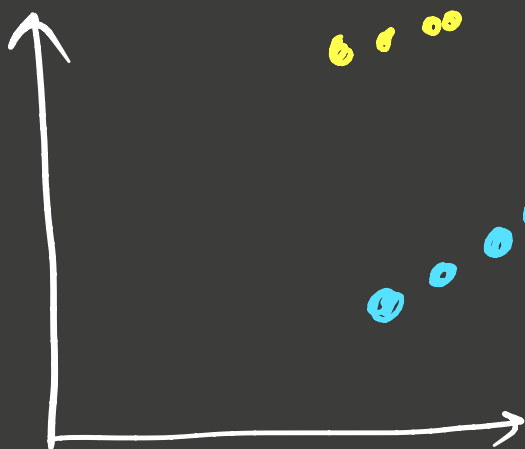


LEARN  $3C_2$  CLASSIFIERS

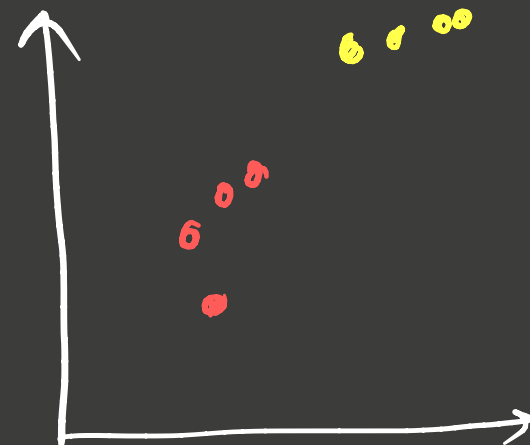
BLUE VS RED



BLUE VS YELLOW

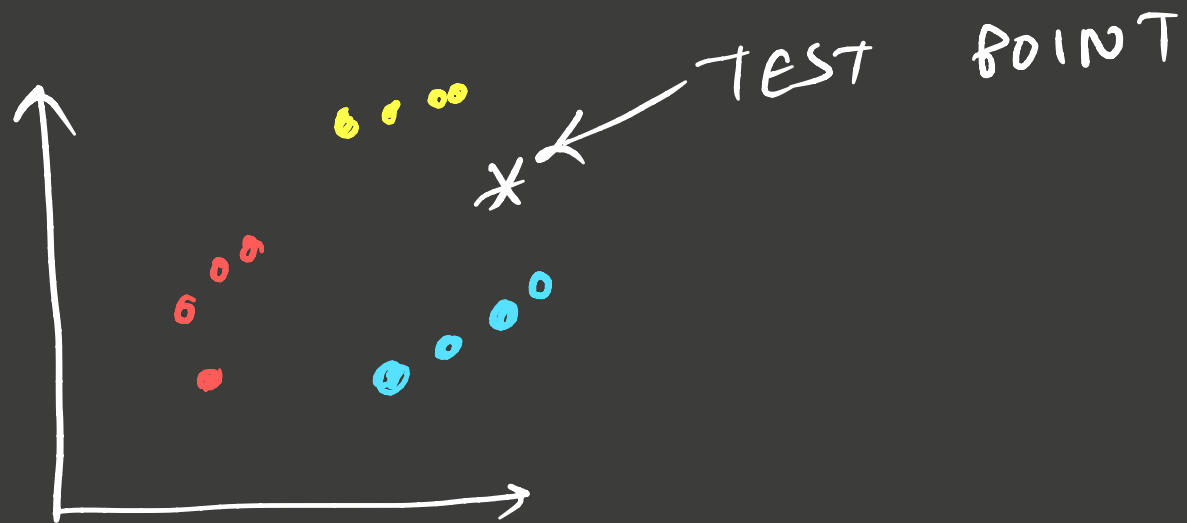


YELLOW VS RED



# SVM FOR MULTI-CLASS CLASSIFICATION

① ONE VS ONE



1) BLUE VS YELLOW  $\rightarrow$  BLUE

2) YELLOW VS RED  $\rightarrow$  RED

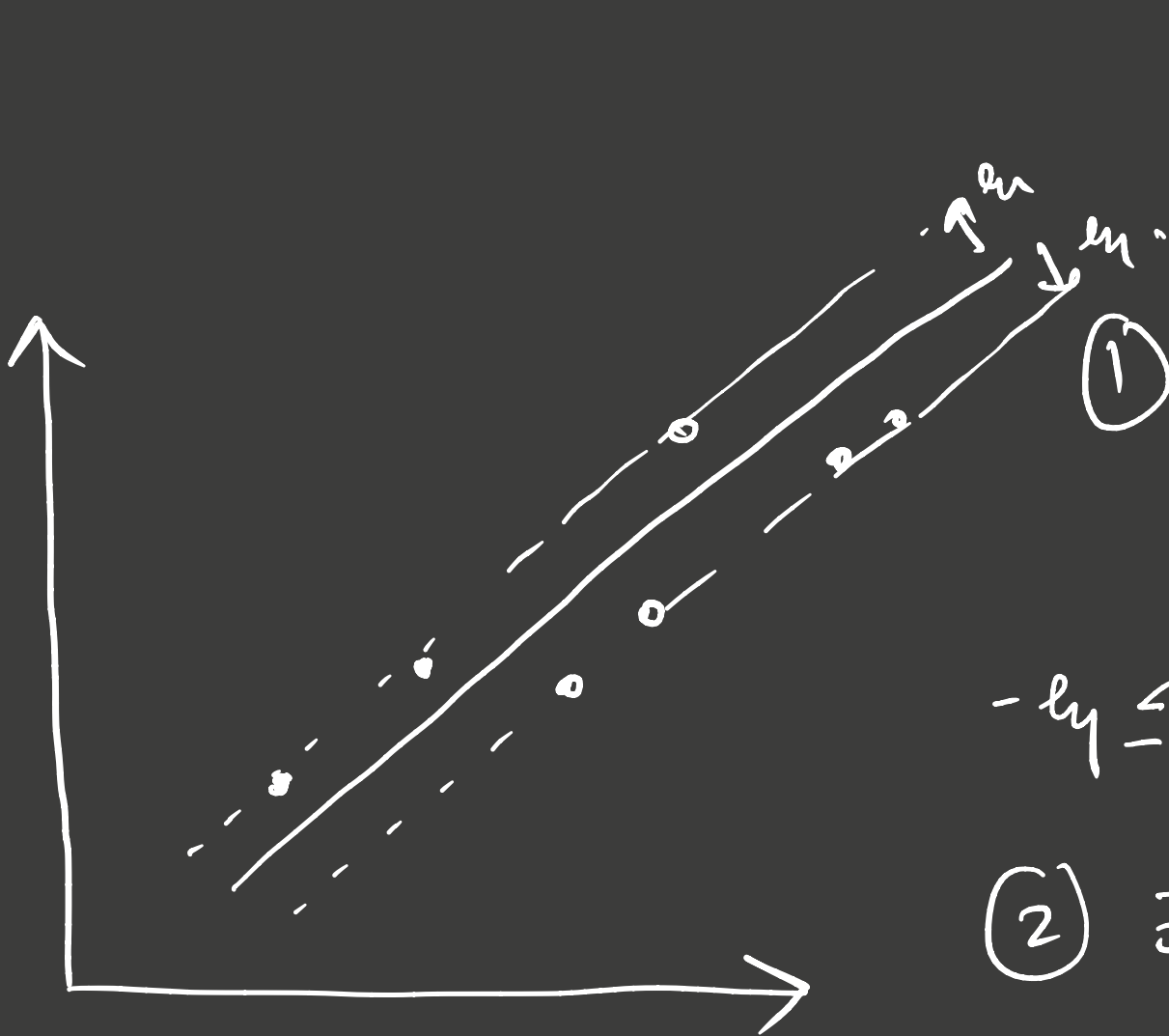
3) RED VS BLUE  $\rightarrow$  BLUE

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MAJORITY = BLUE

# SUPPORT VECTOR REGRESSION

## HARD MARGIN OR $\epsilon$ -SVR



$$\hat{y}(x) = \bar{w} \cdot \bar{x} + b$$

①  $\hat{y}_i$  has at most ' $\epsilon$ ' deviation from  $y_i$  (for train)

$$-\epsilon \leq y_i - (\bar{w} \cdot \bar{x}_i + b) \leq \epsilon$$

② Fit is as "flat" as possible

$\Rightarrow$  Regularize  $w$

i.e. Minimize  $\frac{1}{2} \|\bar{w}\|^2$