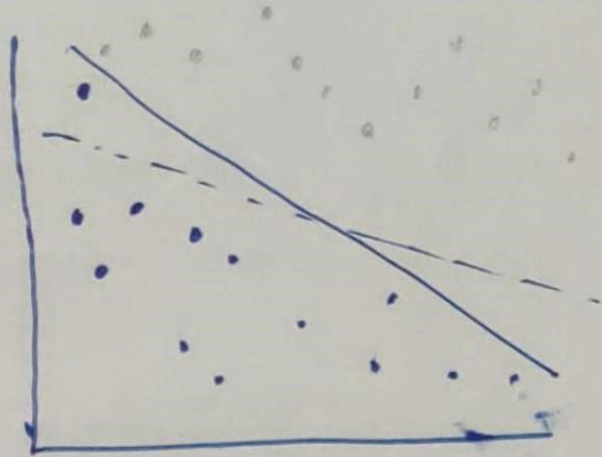
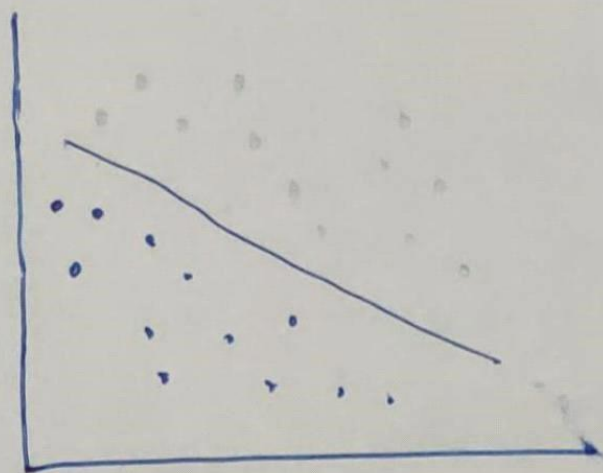


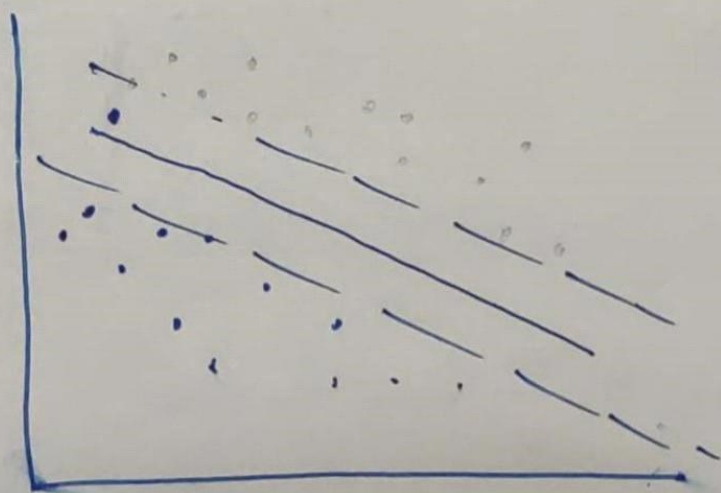
Support vector classifiers

Suppose the classes are not separable directly, then



The deviation of the maximum margin classifier will be a lot, which decreases the robustness.

So we need to choose soft-margin classifier, where we have a certain allowance of error that occurs.



Now the support vectors are given a range of tolerance.

So, now the equation changes

$$\begin{aligned} &\text{Maximize } M \\ &\beta_0, \beta_1, \beta_2, \dots, \beta_n \\ &\epsilon_1, \epsilon_2, \epsilon_3, \dots, \epsilon_n \end{aligned}$$

$$\text{Subject to } \sum_{j=1}^p \beta_j^2 = 1,$$

$$y_i (\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_p x_{ip}) \geq M(1 - \epsilon_i),$$

$$\epsilon_i \geq 0 \quad \sum_{i=1}^n \epsilon_i \leq C.$$

C is the non-negative tuning parameter. M is the width of the margin, but now slack variables are added.
($\epsilon_1, \epsilon_2, \epsilon_3, \dots$)