

Support Vector Machine

- well used classifier
- Binary classifier
- Large-margin classifier

Margin b/w +ve & -ve classes is larger than as compared to other binary classifiers

e.g. Misclassification

decision boundary for some binary classifiers

+ + +
+ + +
+ + +
+

Misclassification

+ + + ⊕
+ + + ⊕
+ + + ⊕

← →

maximum margin for SVM

from border pts.

Support Vector

Why SVM should perform better?

data ⇒ sample is not proper representation of whole data. if new point comes which at border b'coz first D.B it is misclassified

SVM by choosing Maximum Margin tries to reduce this Misclassification because of improper D.B.

Development of SVM from Logistic Regression

LR \Rightarrow ~~$h(x) = \sigma(\theta^T x)$~~
 $z = \theta^T x$

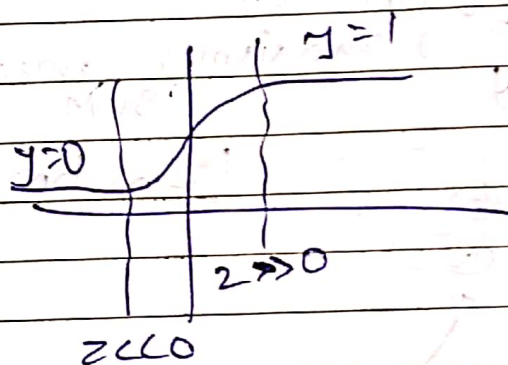
$$y = 1 \quad h(x) > 0.5 \quad z > 0$$

$$y = 0 \quad h(x) < 0.5 \quad z < 0$$

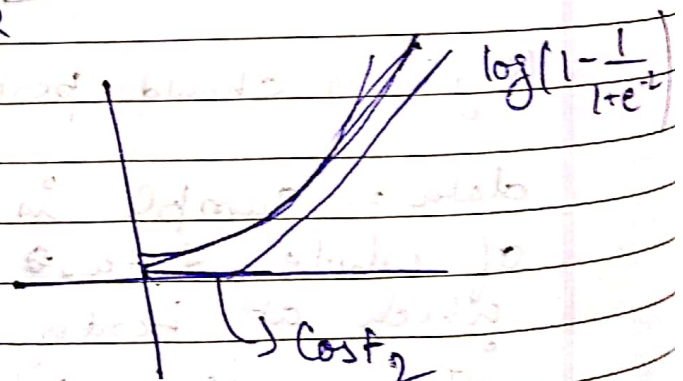
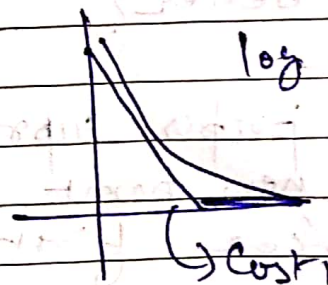
SVM

$$y = 1 \quad h(x) \approx 1.0 \quad \theta^T x \gg 0$$

$$y = 0 \quad h(x) \approx 0.0 \quad \theta^T x \ll 0$$



Cost function for LR



piece wise linearization for SVM

Cost

SVM

$$\min \sum_{i=1}^m y^i \text{cost}_1(\theta^T x_i) + (1-y^i) \text{cost}_0(\theta^T x_i)$$

A

$$+ \frac{\lambda}{2} \sum_{j=0}^n \theta_j^2$$

B

Regularization

Regularization is done so that there is no overfitting on training data of Machine Learning Parameters.

$$\min ((u-5)^2 + 1) \times 10 \rightarrow u=5$$

$$\min 10 (u-5)^2 + 10 \rightarrow u=5$$

$$A + \lambda B$$

Minimize
Remain same
when multiplied

$$\min A + \lambda B \equiv \frac{1}{\lambda} A + B$$

$$\equiv C A + B$$