





NPTEL ONLINE CERTIFICATION COURSES

Course Name: Deep Learning

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Topic

Lecture 14: Linear Machine and Multiclass SVM

CONCEPTS COVERED

Concepts Covered:

- ☐ Linear Classifier
- ☐ Support Vector Machine
- ☐ Linear Machine
- Multiclass Support Vector Machine







$$f: R^{D} \to R^{K}$$
$$f(X_{i}, W, b) = WX_{i} + b = s$$

$$\begin{bmatrix} W_{11} & W_{12} & W_{13} & \dots & W_{1D} \\ W_{21} & W_{22} & W_{23} & \dots & W_{2D} \\ \dots & \dots & \dots & \dots & \dots \\ W_{K1} & W_{K2} & W_{K3} & \dots & W_{KD} \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ \dots \\ X_D \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ \dots \\ b_K \end{bmatrix} = \begin{bmatrix} s_1 \\ s_2 \\ \dots \\ s_K \end{bmatrix}$$





$$\begin{bmatrix} 0.2 & 0.6 & -1.0 & 0.8 \\ 1.5 & 0.9 & 3.1 & 0.1 \\ 0.5 & 1.1 & 0.7 & 0.0 \\ 2.1 & 0.3 & 0.2 & 0.5 \end{bmatrix} \begin{bmatrix} 45 \\ 110 \\ 21 \\ 16 \end{bmatrix} + \begin{bmatrix} 1.1 \\ 5.3 \\ -2.1 \\ 0.6 \end{bmatrix} \Rightarrow \begin{bmatrix} 67.9 \\ 238.5 \\ 156.1 \\ 0.6 \end{bmatrix} \xrightarrow{Cat score}$$

 $f(X_i, W, b)$



Interpretation



$$\begin{bmatrix} 0.2 & 0.6 & -1.0 & 0.8 \\ 1.5 & 0.9 & 3.1 & 0.1 \\ 0.5 & 1.1 & 0.7 & 0.0 \\ 2.1 & 0.3 & 0.2 & 0.5 \end{bmatrix} \begin{bmatrix} 45 \\ 110 \\ 21 \\ 16 \end{bmatrix} + \begin{bmatrix} 1.1 \\ 5.3 \\ -2.1 \\ 0.6 \end{bmatrix} \Rightarrow \begin{bmatrix} 67.9 \\ 238.5 \\ 156.1 \\ 140.3 \end{bmatrix}$$
 Cat score Dog score

 $f(X_i, W, b)$









Bias Trick



$$\begin{bmatrix} 0.2 & 0.6 & -1.0 & 0.8 & 1.1 \\ 1.5 & 0.9 & 3.1 & 0.1 & 5.3 \\ 0.5 & 1.1 & 0.7 & 0.0 & -2.1 \\ 2.1 & 0.3 & 0.2 & 0.5 & 0.1 \end{bmatrix} \begin{bmatrix} 45 \\ 110 \\ 21 \\ 16 \\ 1 \end{bmatrix} \Rightarrow \begin{bmatrix} 67.9 \\ 238.5 \\ 156.1 \\ 140.3 \end{bmatrix} Cat score \\ Dog score \\ Car score \\ f(X_i, W)$$



Multiclass SVM

$$\left. \begin{array}{l} s_{j} = f(X_{i}, W)_{j} \\ = WX_{i} \end{array} \right\} \longrightarrow Score \ for \ j^{th} \ Class \ of \ i^{th} \ Vector \ (X_{i}, y_{i}) \\ \\ s_{y_{i}} = f(X_{i}, W)_{y_{i}} \longrightarrow should \ be \ max \ imum \\ \\ s_{y_{i}} - s_{j} \ge \Delta \\ \\ L_{i} = \sum_{i} \max(0, s_{j} - s_{y_{i}} + \Delta) \end{array}$$



Loss Function: An Example

For some (X_i, y_i) where $y_i = 2$

$$s = (10 \ 30 \ -20 \ 25)^t$$
 $\Delta = 10$

$$L_i = \sum_{j \neq y_i} \max(0, s_j - s_{y_i} + \Delta)$$

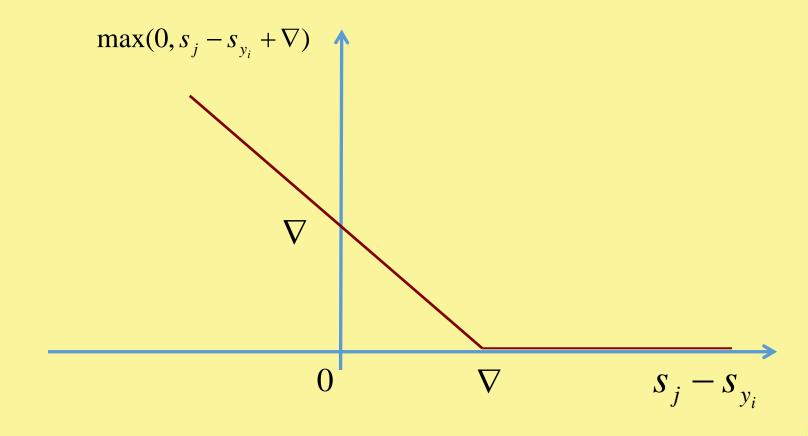
$$= \max(0,10-30+10) + \max(0,-20-30+10) + \max(0,25-30+10)$$

$$=0+0+15$$

$$=15$$



Hinge Loss





Regularization

$$S_{j} - S_{y_{i}} = W_{j}^{t} X_{i} - W_{y_{i}}^{t} X_{i}$$

Scaling W by $\lambda: W \leftarrow \lambda W$

$$S_{j} - S_{y_{i}} \leftarrow \lambda (S_{j} - S_{y_{i}})$$



Regularization

Include a regularization term R(W)

$$R(W) = \lambda \sum_{k} \sum_{l} W_{kl}^{2}$$

$$L = \frac{1}{N} \sum_{i} L_{i} + \lambda R(W)$$

$$L = \frac{1}{N} \sum_{i} \sum_{j \neq y_i} [\max(0, f(X_i, W)_j - f(X_i, W)_{y_i} + \nabla) + \lambda \sum_{k} \sum_{l} W_{kl}^2]$$









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Thank you