

# Lecture

# Tips for Homework 1

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IE590 DEEP LEARNING IN MACHINE VISION

# Tips for KNN

- Let  $A^{2 \times 2}$  and  $X^{3 \times 2}$  be two matrices. We want to compute the pairwise distance between each row in  $A$  to each row in  $X$ .
- Let  $D^{2 \times 3}$  be the distance matrix, where  $D(i, j)$  represents the pairwise distance between  $i^{th}$  row in  $A$  to  $j^{th}$  row in  $X$

$A$	$X$
$\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \end{bmatrix}$	$\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \end{bmatrix}$

# Tips for KNN

$A$	$X$
$\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \end{bmatrix}$	$\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \end{bmatrix}$

# Tips for KNN

$A$

$$\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \end{bmatrix}$$

$X$

$$\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \end{bmatrix}$$

$$D(A, X) = \begin{bmatrix} (a_1 - x_1)^2 + (b_1 - y_1)^2 & (a_1 - x_2)^2 + (b_1 - y_2)^2 & (a_1 - x_3)^2 + (b_1 - y_3)^2 \\ (a_2 - x_1)^2 + (b_2 - y_1)^2 & (a_2 - x_2)^2 + (b_2 - y_2)^2 & (a_2 - x_3)^2 + (b_2 - y_3)^2 \end{bmatrix}$$

# Tips for KNN

$$\begin{array}{cc} A & X \\ \begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \end{bmatrix} & \begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ x_3 & y_3 \end{bmatrix} \end{array}$$

$$D(A, X) = \begin{bmatrix} (a_1 - x_1)^2 + (b_1 - y_1)^2 & (a_1 - x_2)^2 + (b_1 - y_2)^2 & (a_1 - x_3)^2 + (b_1 - y_3)^2 \\ (a_2 - x_1)^2 + (b_2 - y_1)^2 & (a_2 - x_2)^2 + (b_2 - y_2)^2 & (a_2 - x_3)^2 + (b_2 - y_3)^2 \end{bmatrix}$$

Don't use this  
equation  
literally !!!

$$D(A, X) = \sqrt{\Sigma A^2 - 2 (A \cdot X) + \Sigma X^2}$$

<https://arxiv.org/pdf/1502.07541.pdf>

# SVM Loss

# Naïve SVM Loss

$$L_i = \sum_{j \neq y_i} \max(0, w_j^T x_i - w_{y_i}^T x_i + \Delta)$$

Example

$$s = \begin{bmatrix} 13 & -7 & 14 \\ -2 & 8 & 10 \end{bmatrix} \quad y_0 = 0, y_1 = 1$$

$$l_0 = \max(0, -7 - 13 + 1) + \max(0, 14 - 13 + 1) = 2$$

$$l_1 = \max(0, -2 - 8 + 1) + \max(0, 10 - 8 + 1) = 3$$

3 classes  
2 data points

$$Loss = 2 + 3 = 5$$

## Example

# Vectorized SVM Loss

$$s = \begin{bmatrix} 13 & -7 & 14 \\ -2 & 8 & 10 \end{bmatrix}$$

$$y_0 = 0, y_1 = 1$$

$$L_i = \sum_{j \neq y_i} \max(0, w_j^T x_i - w_{y_i}^T x_i + \Delta)$$



## Example

# Vectorized SVM Loss

$$s = \begin{bmatrix} 13 & -7 & 14 \\ -2 & 8 & 10 \end{bmatrix}$$

$$y_0 = 0, y_1 = 1$$

$$s = \begin{bmatrix} \textcircled{13} & -7 & 14 \\ -2 & \textcircled{8} & 10 \end{bmatrix}$$

$$\longrightarrow c = \begin{bmatrix} 13 \\ 8 \end{bmatrix}$$

$$L_i = \sum_{j \neq y_i} \max(0, w_j^T x_i - w_{y_i}^T x_i + \Delta)$$

## Example

# Vectorized SVM Loss

$$s = \begin{bmatrix} 13 & -7 & 14 \\ -2 & 8 & 10 \end{bmatrix}$$

$$y_0 = 0, y_1 = 1$$

$$s = \begin{bmatrix} \textcircled{13} & -7 & 14 \\ -2 & \textcircled{8} & 10 \end{bmatrix} \longrightarrow c = \begin{bmatrix} 13 \\ 8 \end{bmatrix}$$

$$L_i = \sum_{j \neq y_i} \max(0, w_j^T x_i - w_{y_i}^T x_i + \Delta)$$

$$s_1 = s - c + 1$$

$$s_1 = \begin{bmatrix} \textcircled{1} & -19 & 2 \\ -9 & \textcircled{1} & 3 \end{bmatrix}$$

$$s_2 = \begin{bmatrix} 0 & -19 & 2 \\ -9 & 0 & 3 \end{bmatrix}$$

Element wise  
 $\max(s, 0)$

$$s_2 = \begin{bmatrix} 0 & 0 & 2 \\ 0 & 0 & 3 \end{bmatrix}$$

*Loss = sum of all elements = 5*

# SoftMax Loss

## Example

# SoftMax Classifier

- Say there are 3 classes and two data points
- Let  $S$  be the score function i.e.  $S = X W$

## Example

# Vectorized SoftMax Loss

$$S = \begin{bmatrix} 13 & -7 & 14 \\ -2 & 8 & 10 \end{bmatrix}$$

$$y_0 = 0, y_1 = 1$$

Take exponential

$$L_s = -\frac{1}{M} \sum_{i=1}^M \log \frac{\exp(W_{y_i}^T f(x_i) + b_{y_i})}{\sum_{j=1}^C \exp(W_j^T f(x_i) + b_j)}$$

$$S = e^{\begin{bmatrix} 13 & -7 & 14 \\ -2 & 8 & 10 \end{bmatrix}} = \begin{bmatrix} e^{13} & e^{-7} & e^{14} \\ e^{-2} & e^8 & e^{10} \end{bmatrix}$$

$$S = 1e + 06 * \begin{bmatrix} 0.44 & 0.00 & 1.20 \\ 0.00 & 0.00 & 0.02 \end{bmatrix}$$

## Example

# Vectorized SoftMax Loss

$$S = 1e + 06 * \begin{bmatrix} 0.44 & 0.00 & 1.20 \\ 0.00 & 0.00 & 0.02 \end{bmatrix}$$

$$L_s = -\frac{1}{M} \sum_{i=1}^M \log \frac{\exp(W_{y_i}^T f(x_i) + b_{y_i})}{\sum_{j=1}^C \exp(W_j^T f(x_i) + b_j)}$$

Find max in each row

$$Z = 1e + 06 * \begin{bmatrix} 0.44 \\ 0.02 \end{bmatrix}$$

Subtract max from each row

$$S = S - Z = 1e + 06 * \begin{bmatrix} -0.76 & -1.20 & 0.00 \\ -0.02 & -0.02 & 0.00 \end{bmatrix}$$

## Example

# Vectorized SoftMax Loss

$$S = 1e + 06 * \begin{bmatrix} -0.76 & -1.20 & 0.00 \\ -0.02 & -0.02 & 0.00 \end{bmatrix}$$

$$L_s = -\frac{1}{M} \sum_{i=1}^M \log \frac{\exp(W_{y_i}^T f(x_i) + b_{y_i})}{\sum_{j=1}^C \exp(W_j^T f(x_i) + b_j)}$$

Find sum of each row

$$Q = 1e + 06 * \begin{bmatrix} -1.96 \\ -0.04 \end{bmatrix}$$

Normalize each row

$$S = \frac{S}{Z} = \begin{bmatrix} 0.39 & 0.61 & 0.00 \\ 0.54 & 0.46 & 0.00 \end{bmatrix}$$

## Example

# Vectorized SoftMax Loss

$$S = \begin{bmatrix} 0.39 & 0.61 & 0.00 \\ 0.54 & 0.46 & 0.00 \end{bmatrix}$$

$$S = \begin{bmatrix} 0.39 & 0.61 & 0.00 \\ 0.54 & 0.46 & 0.00 \end{bmatrix} \longrightarrow c = \begin{bmatrix} 0.39 \\ 0.46 \end{bmatrix}$$

$$L = -0.5 * (\log(0.39) + \log(0.46))$$

$$L = -1.44$$



# SoftMax Classifier

- No. of data points -  $n$
- Dimension of the data -  $d$
- No. of classes -  $c$
- Subscript  $i$  indicates  $i^{th}$  row
- Superscript  $j$  indicates  $j^{th}$  column
- Input data matrix -  $X^{n \times d}$
- Weight matrix -  $W^{d \times c}$
- Scores ( $S^{n \times c}$ )

# SoftMax Classifier

$$S = X W$$

- $S_i$  is  $i^{th}$  row in  $S$
- $S_i^j$  is the element in the  $i^{th}$  row and  $j^{th}$  column

$$S_i = X_i^T W$$

$$\Rightarrow S_i = X_i^T [W^1 \mid W^2 \mid \dots \mid W^c]$$