

# CSE 427 : Machine Learning

Quiz  $\rightarrow 15\%$   
Assign  $\rightarrow 15\%$   
Mid  $\rightarrow 20\%$   
Final  $\rightarrow 50\%$



Experience Task Performance

Training Data	Testing Data
70%	30%
80%	20%

i) Supervised Learning

ii) Unsupervised Learning

iii) Reinforcement Learning

Nobel Prize Name  $\Rightarrow$   
Mathematics = Fields Medal  
Comp. Sci. = ACM AM Turing Award

Dr. Holly Kreiger

Youtube Channel  $\Rightarrow$  Number file

# আমান্নাং কাছে ১L ও ৩L এর বটল আছে। আমান্নাকে ৭L পানি মেপে দিতে হবে।

$\Rightarrow$  ১/১L জের

২/১L থেকে ৩L এ ঢালব (১L এ ২L থাকবে)

৩/ ৩L খালি বসাব।

৪/ ১L এর সর্বোচ্চ যে ২L আছে সেটা ৩L এ ঢালব।

৫/ ১L জের।

৬/ ১L থেকে ৩L এর বোতল পূর্ণাঙ্গ জের।

(৩L এর সর্বোচ্চ আছে ২L ছিল, এখন তাতে ১L জেরল)

৭/  $\therefore$  ১L এ এখন ৭L আছে।

Ceftron

# Supervised Learning

Regression  
(value দ্বারা)

Classification  
(class দ্বারা)

1000 mail এর মধ্যে 850  $\rightarrow$  non-spam } Classification  
150  $\rightarrow$  spam

Weather Data  $\rightarrow$  আগামীকাল Temp. কত হবে?  $\rightarrow$  Regression  
 $\rightarrow$  আগামীকাল বৃষ্টি হবে কি হবেনা?  $\rightarrow$  Classification.

1000 image data  $\rightarrow$  Male/Female  $\rightarrow$  Classification  
 $\rightarrow$  Age  $\rightarrow$  Regression.

## Single Variable Linear Regression

১টি feature দ্বারা

① Dataset

(Feature) $x$	$y$ (Output)
0	4
1	7
2	7
3	8

## ② Hypothesis function:

$$h_{\theta}(x) = \theta_0 + \theta_1 * x$$

$$y = mx + c$$

## ③ Parameter Initialization:

$\theta_0, \theta_1 \rightarrow$  parameter

যদি,  $\theta_0 = 2, \theta_1 = 2$

(Randomly initialize করে  
Dataset range এর ভিতর করে)

$$h_{\theta}^{(1)}(x) = \theta_0 + \theta_1 * x^{(1)}$$
$$= 2 + 2 * 0 = 2$$

$$h_{\theta}^{(2)}(x) = \theta_0 + \theta_1 * x^{(2)}$$
$$= 2 + 2 * 1 = 4$$

$$h_{\theta}^{(3)}(x) = \theta_0 + \theta_1 * x^{(3)}$$
$$= 2 + 2 * 2 = 6$$

$$h_{\theta}^{(4)}(x) = \theta_0 + \theta_1 * x^{(4)}$$
$$= 2 + 2 * 3 = 8$$

feature  $n$  অংখ্যক হলে  
parameter  $n+1$  অংখ্যক হবে।

④ cost function: (ফুলের পরিমাণ/error)  
করা কর

$$\begin{aligned}
 J(\theta) &= \frac{1}{2m} \sum_{i=1}^m \left( h_{\theta}^{(i)}(x) - y^{(i)} \right)^2 \\
 &= \frac{1}{2 \times 4} \left\{ (2-1)^2 + (4-7)^2 + (6-7)^2 + (8-8)^2 \right\} \\
 &= \frac{1}{8} (1 + 9 + 1 + 0) \\
 &= \frac{1}{8} \times 11 \\
 &= 1.375 \text{ error/cost.}
 \end{aligned}$$

Next Step A  $\theta_0, \theta_1$  এমনভাবে change করা যাবে error কমাবে।

⑤ Gradient Descent:

repeat until convergence

$$\left\{ \begin{array}{l} \text{আমরা} \\ \theta_j = \theta_j - \left[ \frac{\alpha}{m} \frac{d}{d\theta_j} \{ J(\theta) \} \right] \end{array} \right.$$

Learning rate.

# of data entry

$$\theta_0 := \theta_0 - \frac{\alpha}{m} \sum_{i=1}^m \left( h_{\theta}^{(i)} - y^{(i)} \right)$$

$$\theta_1 := \theta_1 - \frac{\alpha}{m} \sum_{i=1}^m \left( h_{\theta}^{(i)} - y^{(i)} \right) \cdot x^{(i)}$$

Let,  $\alpha = 0.01$  (ছোট অংখ্যক বসাব)

$$\begin{aligned}\theta_0 &:= 2 - \frac{0.01}{4} \{ (2-4) + (4-7) + (6-7) + (8-8) \} \\ &= 2 - \frac{0.01}{4} \{ -2 - 3 - 1 + 0 \} \\ &= 2.015\end{aligned}$$

$$\begin{aligned}\theta_1 &:= 2 - \frac{0.01}{4} \{ (2-4) \cdot 0 + (4-7) \cdot 1 + (6-7) \cdot 2 + (8-8) \cdot 3 \} \\ &= 2 - \frac{0.01}{4} \{ 0 - 3 - 2 + 0 \} \\ &= 2.0125\end{aligned}$$

এবার নতুন  $\theta_0, \theta_1$  দিয়ে Hypothesis function বের করব।  
cost function বের করব। আবার  $\theta_0, \theta_1$  update করব।

এভাবে,  $h_\theta(x) = \theta_0 + \theta_1 * x$   
 $\hookrightarrow x$  এর যেকোনো value-র জন্য  
 $\rightarrow y$  এর prediction পাব।

Plot:  $x$  এর সাপেক্ষে  $h$   
 $x$  এর সাপেক্ষে  $y$ .



## Multi variable Linear Regression

(i) Dataset:

$x_1$	$x_2$	$y$
1	0.5	2
1	1.5	3
2	1	4
3	1	4

(ii)

$\theta_0, \theta_1, \theta_2$  Let,  $\theta_0 = 1, \theta_1 = 0.5, \theta_2 = 1$

(iii) Hypothesis function:

$$h_{\theta}(x) = \theta_0 + \theta_1 * x_1 + \theta_2 * x_2$$

$$\begin{aligned} h_{\theta}^{(1)}(x) &= \theta_0 + \theta_1 * x_1^{(1)} + \theta_2 * x_2^{(1)} \\ &= 1 + 0.5 * 1 + 1 * 0.5 = 2 \end{aligned}$$

$$\begin{aligned} h_{\theta}^{(2)}(x) &= \theta_0 + \theta_1 * x_1^{(2)} + \theta_2 * x_2^{(2)} \\ &= 1 + 0.5 * 1 + 1 * (1.5) \\ &= 3 \end{aligned}$$

$$\begin{aligned} h_{\theta}^{(3)}(x) &= \theta_0 + \theta_1 * x_1^{(3)} + \theta_2 * x_2^{(3)} \\ &= 1 + 0.5 * 2 + 1 * 1 \\ &= 3 \end{aligned}$$

$$\begin{aligned} h_{\theta}^{(4)}(x) &= \theta_0 + \theta_1 * x_1^{(4)} + \theta_2 * x_2^{(4)} \\ &= 1 + 0.5 * 3 + 1 * 1 \\ &= 3.5 \end{aligned}$$

④ Cost Function:

$$J(\theta) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}^{(i)} - y^{(i)})^2$$

$$= \frac{1}{2 \times 4} \{ (2-2)^2 + (3-3)^2 + (3-4)^2 + (3.5-4)^2 \}$$

$$= \frac{1}{8} (0+0+1+0.25) = \frac{1.25}{8} = 0.15625$$

⑤ Gradient Descent:

Repeat until convergence:

$$\theta_{j \in \{0,1,2\}} := \theta_j - \frac{\alpha}{m} \frac{\partial}{\partial \theta_j} \{J(\theta)\}$$

$$\theta_0 := \theta_0 - \frac{\alpha}{m} \sum_{i=1}^m (h_{\theta}^{(i)}(x) - y^{(i)})$$

$$\theta_1 := \theta_1 - \frac{\alpha}{m} \sum_{i=1}^m (h_{\theta}^{(i)}(x) - y^{(i)}) \cdot x_1^{(i)}$$

$$\theta_2 := \theta_2 - \frac{\alpha}{m} \sum_{i=1}^m (h_{\theta}^{(i)}(x) - y^{(i)}) \cdot x_2^{(i)}$$

$$\therefore \theta_0 := 1 - \frac{0.01}{4} \{ (2-2) + (3-3) + (3-4) + (3.5-4) \}$$

$$= 1 - \frac{0.01}{4} \{ 0 + 0 + (-1) + (-0.5) \}$$

$$= 1.00375$$

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$$\begin{aligned}
 \theta_1 &:= 0.5 - \frac{0.01}{4} \{ (2-2) \cdot 1 + (3-3) \cdot 1 + (3-4) \cdot 2 + (3.5-4) \cdot 3 \} \\
 &= 0.5 - \frac{0.01}{4} \{ 0 + 0 + (-2) + (-1.5) \} \\
 &= 0.50125
 \end{aligned}$$

$$\begin{aligned}
 \theta_2 &:= 1 - \frac{0.01}{4} \{ (2-2) (0.5) + (3-3) (1.5) + (3-4) \cdot 1 + (3.5-4) \cdot 1 \} \\
 &= 1 - \frac{0.01}{4} \{ 0 + 0 + (-1) + (-0.5) \} \\
 &= 1.00375
 \end{aligned}$$



## Dataset Vectorization:

$x_1, x_2, x_3, \dots, x_{100}$  → feature set/vector

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{100} \end{bmatrix} \rightarrow \text{feature vector}$$

$$x^{(1)} = \begin{bmatrix} x_1^{(1)} \\ x_2^{(1)} \\ x_3^{(1)} \\ \vdots \\ x_{100}^{(1)} \end{bmatrix}$$

$$x^{(3)} = \begin{bmatrix} x_1^{(3)} \\ x_2^{(3)} \\ x_3^{(3)} \\ \vdots \\ x_{100}^{(3)} \end{bmatrix}$$

যে matrix এর row n-ডাইমেনশন, column ১টি, তাকে Matrix কে বলব vector.  $n \times 1$   
↓ ↓  
row colm.

$$\theta = \begin{bmatrix} \theta_0 \\ \theta_1 \\ \vdots \\ \theta_{100} \end{bmatrix}$$

$$h_{\theta}(x) = \theta^T \cdot x$$

$$= [\theta_0, \theta_1, \theta_2, \dots, \theta_{100}]$$

$$\begin{bmatrix} x_0=1 \\ x_1 \\ x_2 \\ \vdots \\ x_{100} \end{bmatrix} \quad 101 \times 1$$

$$= \theta_0 \cdot x_0 + \theta_1 \cdot x_1 + \theta_2 \cdot x_2 + \dots + \theta_{100} \cdot x_{100}$$

$$= \theta_0 + \theta_1 \cdot x_1 + \theta_2 \cdot x_2 + \dots + \theta_{100} \cdot x_{100}$$

\*for solving large number of Dataset it can be done in dot product which is vectorization.

### Normalization:

$x_1$	$x_2$	$x_3$
0.2	105	15280
0.7	270	17570
0.125	380	19390
0.42	170	2010

Normalization কখন করব?

⇒ যখন Database এর feature এ বিভিন্ন range এর value থাকবে।

### Mean Normalization:

$$\begin{aligned}x_1^{(1)} &= \frac{x_i - \mu_i}{S_i} \\&= \frac{0.2 - 0.36125}{0.575} \\&= -0.2804\end{aligned}$$

$$\begin{aligned}x_2^{(1)} &= \frac{105 - 231.25}{275} \\&= -0.4590\end{aligned}$$

$$\mu_i(x_1) = \frac{0.2 + 0.7 + 0.125 + 0.42}{4}$$

$$= 0.36125$$

$$\begin{aligned}S_i(x_1) &= \max(x_i) - \min(x_i) \\&= 0.7 - 0.125 \\&= 0.575\end{aligned}$$

$$\begin{aligned}\mu_i(x_2) &= \frac{105 + 270 + 380 + 170}{4} \\&= 231.25\end{aligned}$$

$$\begin{aligned}S_i(x_2) &= 380 - 105 \\&= 275\end{aligned}$$

$[-1, 1]$  এর মধ্যে থাকবে।

$$\begin{aligned} x_3 &= \frac{15280 - 13562.5}{17380} \\ &= 0.0988 \end{aligned}$$

$$\begin{aligned} \mu_i(x_3) &= \frac{15280 + 17570 + 19390 + 2010}{4} \\ &= 13562.5 \end{aligned}$$

$$\begin{aligned} S_i(x_3) &= 19390 - 2010 \\ &= 17380 \end{aligned}$$

এভাবে প্রত্যেক  $x_1, x_2, x_3$  এর জন্য update value  
বের করা লাগবে।

Normalization এর জন্য  $\alpha$  not needed,  $\alpha$  এর  
value normalization এর উপর dependent.