

## Linear Regression:

$$y = mx + c$$

slope    intercept

$$y = ax + bz + c$$

$n^{\text{th}}$  dimension  
line eq<sup>n</sup>  $(n-1)$  dimensions.

$$y = mx + c$$

$$y = \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3 + \dots + \theta_{n-1} x_{n-1} + c$$

$$y = \theta_0 x_0 + \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_{n-1} x_{n-1} \quad (x_0 = 1) \quad \text{let } (\theta_0 = c)$$

$$y = \sum_{i=0}^{n-1} \theta_i x_i$$

hypothesis func. of  
linear regression

( $\theta_0$ : intercept  
 $x_0: 1$ )

loss used in this case:

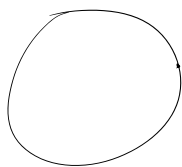
$$L_2 \text{ loss: } J = \min \frac{1}{2} \sum_{i=0}^n (h(x_i) - y_i)^2$$

99

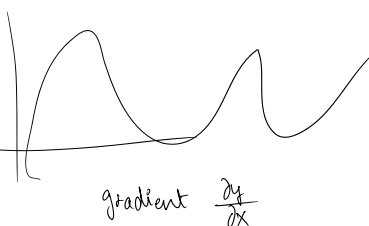
$$y = mx + c$$

$$y = m(2) + c$$

## Gradient Descent:



$$\frac{dy}{dx} = \text{slope}$$

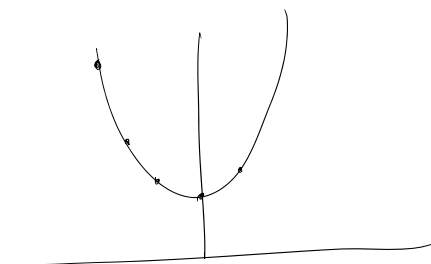


$$\text{gradient } \frac{\partial y}{\partial x}$$

$$y = ax + bz + c \quad \frac{\partial y}{\partial x}, \frac{\partial y}{\partial z}$$

$$\frac{\partial J}{\partial \theta_i}$$

$$\theta_j = \theta_j - \alpha \frac{\partial J}{\partial \theta_j}$$



$$y = 4x + 2$$

$$y = 2x + 1$$

$\alpha$ : learning rate ( $\alpha: 0.001, 0.003$ )  
Hyper parameter

$$\frac{\partial J}{\partial \theta_j}$$

$$J = \min \frac{1}{2} \sum_{i=0}^n (h(x_i) - y_i)^2$$

$$h(x_i) = \sum_{i=0}^{n-1} \theta_i x_i$$

$$\frac{\partial J}{\partial \theta_j} = \frac{1}{n} \sum (h(x_i) - y_i) \cdot \frac{\partial}{\partial \theta_j} (h(x_i) - y_i)$$

(100%) Dataset?  $m \times n$

(70-80%) → Training set

(20-10%) → Testing set

(10-10%) → Validation

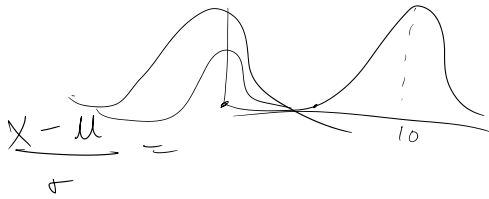


$m \times 3$

$$y = ax + bz + c$$

set

$M_L$



$$\partial \theta_j \quad \neq$$

$$\partial \theta_j$$

$$(h(x_i) - y_i) \cdot \frac{\partial}{\partial \theta_j} (\sum_i \theta_i x_i - y_i)$$

$$\boxed{\frac{\partial J}{\partial \theta_j} = (h(x_i) - y_i) (x_j)}$$

