

Exercise 1

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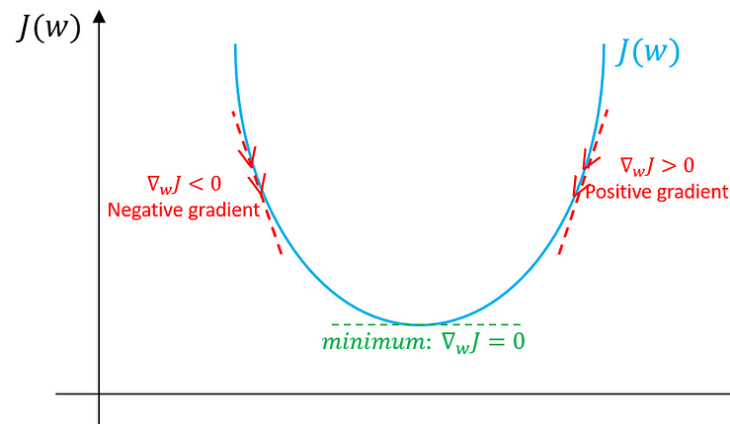
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Understanding Linear Regression

- • Models relationship between input x and output y .
- • Types:
 - - Simple Linear Regression: $f(x) = wx$
 - - Linear Regression with Bias: $f(x) = w_1x + w_0$
- • Goal: Find the best parameters w or $[w_0, w_1]$ that minimize the error.

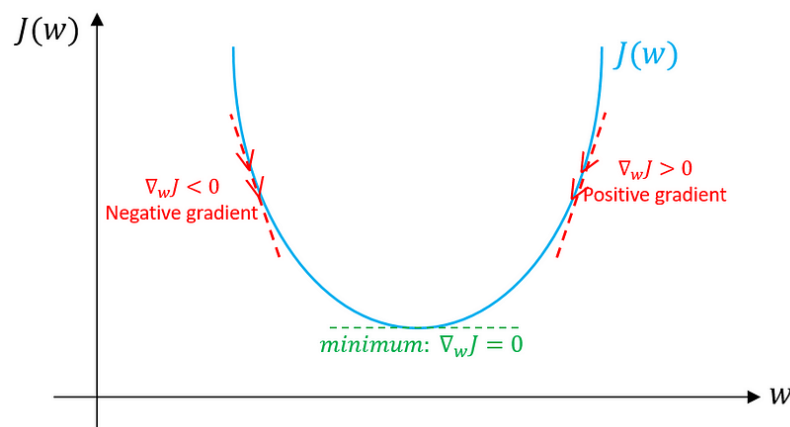
The Meaning of Derivative(w)

- In linear regression, Derivative represents the rate of change of a function at a given point, also known as the gradient (slope).
- In mathematics, the extremum of a function occurs at a critical point(臨界點), where the derivative is equal to zero.
- Therefore, we find the extremum by setting the rate of change of the function to zero.



The Meaning of Derivative(導數)

- 在線性回歸中，導數代表函數在某一點的變化率，也被稱為梯度（斜率）。
- 在**數學**中，函數的極值(大or小)發生在**臨界點**，即導數等於零的地方。
- 因此，我們透過設函數的變化率為 0 來求極值。



(a) Solving for w in $f(x) = wx$

- Objective: Minimize the error

$$J(w) = \sum (wx_i - y_i)^2$$

- Find the first derivative of the function with respect to w (對函數 $J(w)$ 做 w 偏微分)

$$\frac{\partial J(w)}{\partial w} = \frac{\partial J(w)}{\partial (wx_i - y_i)} \cdot \frac{\partial (wx_i - y_i)}{\partial w} = \frac{1}{n} \sum_{i=1}^n 2(wx_i - y_i) \cdot x_i = \frac{2}{n} \sum_{i=1}^n x_i \cdot (wx_i - y_i)$$

Chain Rule

$$\frac{2}{n} \sum_{i=1}^n x_i \cdot (wx_i - y_i) = 0$$

$$\sum_{i=1}^n x_i y_i = \sum_{i=1}^n wx_i^2$$

$$\sum_{i=1}^n \frac{x_i y_i}{x_i^2} = w$$

- Solution:

$$w = \sum \frac{x_i y_i}{x_i^2}$$

$$w = 1.3$$

(b) Solving for w_0, w_1 using Normal Equation

- Model: $f(x_i) = y_i = w_1 x_i + w_0$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_i \end{bmatrix}_{i \times 1} = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ \vdots & \vdots \\ 1 & x_i \end{bmatrix}_{i \times 2} \begin{bmatrix} w_0 \\ w_1 \end{bmatrix}_{2 \times 1}$$

- Apply Normal Equation: $w = (X^T X)^{-1} X^T Y$
- Matrix Formulation:

$$X = \begin{bmatrix} 1 & -3 \\ 1 & -1 \\ 1 & 0 \\ 1 & 2 \\ 1 & 4 \end{bmatrix}, Y = \begin{bmatrix} -4 \\ 1 \\ 1 \\ 2 \\ 6 \end{bmatrix}$$

- Solution:

$$w = \frac{1}{146} \begin{bmatrix} 30 & -2 \\ -2 & 5 \end{bmatrix} \begin{bmatrix} 6 \\ 39 \end{bmatrix} = \frac{1}{146} \begin{bmatrix} 102 \\ 183 \end{bmatrix}$$

(c) Predictions for $x = 1$

- Using $f(x) = 1.3x$:Substitute $x = 1$, compute y .
- Using $f(x) = 1.253x + 0.698$:Substitute $x = 1$, compute y .
- Solution:
 - (a) $f(1) = 1.3$
 - (b) $f(1) = 1.253 + 0.698 = 1.951$

(d) Gradient Descent Explanation

- $J(w) = \frac{1}{n} \sum_{i=1}^n (wx_i - y_i)^2$
- Update Rule:

$$W^{(t+1)} = w^t - \alpha \frac{\partial J}{\partial w}$$

$$\frac{\partial J(w)}{\partial w} = \frac{2}{n} \sum_{i=1}^n x_i \cdot (wx_i - y_i)$$

- Hyperparameters:
 - Learning Rate: $\alpha = 0.1$
 - Initial Value: $w = 0$

(d) Gradient Descent Iterations

- Iteration 1: Compute new w .
- Iteration 2: Compute next w .
- Show step-by-step calculations.

$$w_1 = w_0 - \frac{0.1}{5} \cdot \left[\sum_{i=1}^n x_i \cdot (w_0 x_i - y_i) \right]$$

$$w_1 = w_0 - \frac{0.1}{5} \cdot [(-3) \cdot 4 + (-1) \cdot (-1) + 0 \cdot (-1) + 2 \cdot (-2) + 4 \cdot (-6)] = -\frac{0.1}{5} \cdot -39 \\ = 0.78$$

$$w_2 = w_1 - \frac{0.1}{5} \cdot \left[\sum_{i=1}^n x_i \cdot (0.78x_i - y_i) \right]$$

$$w_2 = w_1 - \frac{0.1}{5} \cdot [-3 \cdot 1.66 + 1.78 + 2.044 + 4.288] = 0.78 - \frac{0.1}{5} \cdot (15.6) = 1.092$$

Problem 2

(a)

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Linear regression equations: $y_1 = w_0x_1 + w_1x_2, y_2 = w'_0x_1 + w'_1x_2$

In matrix format: $f(x) = Y = Xw$

$$\begin{bmatrix} y_1 & y_2 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} w_0 & w'_0 \\ w_1 & w'_1 \end{bmatrix}$$
$$Y = \begin{bmatrix} 7 & 10 \\ 4 & 5 \\ 2 & 9 \\ 6 & 6 \end{bmatrix}_{4 \times 2} \quad X = \begin{bmatrix} 3 & 5 \\ 2 & 2 \\ 1 & 2 \\ 3 & 3 \end{bmatrix}_{4 \times 2} \quad W = \begin{bmatrix} w_0 & w'_0 \\ w_1 & w'_1 \end{bmatrix}$$

To estimate W , use normal equation: $W = (X^T X)^{-1} X^T Y = \begin{bmatrix} \frac{18}{11} & -1 \\ \frac{25}{26} & 3 \end{bmatrix}$

$$y_1 = \frac{18}{11}x_1 + \frac{25}{26}x_2$$
$$y_2 = -1x_1 + 3x_2$$

(b)

- Substitute $x_1 = 7, x_2 = 4$, into
 $y_1 = w_0x_1 + w_1x_2, y_2 = w'_0x_1 + w'_1x_2$
- *To get the predicted outputs*

$$[y_1 \quad y_2] = [7 \quad 4] \begin{bmatrix} \frac{18}{11} & -1 \\ 25 & 3 \\ \frac{66}{33} & 5 \end{bmatrix} = \begin{bmatrix} 428 & 5 \\ 33 & 5 \end{bmatrix}$$