

HW 2.1

Part I. Analytic Assignment

HW 2.1

AMLY90-02

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- 1) Compute the gradient vector for a plane in 3D space.

$$z = f(x, y) = ax + by + c$$

$$\begin{aligned} f_x &= a \\ f_y &= b \\ f_z &= -1 \end{aligned} \Rightarrow \nabla f(x, y) = \begin{bmatrix} a \\ b \\ -1 \end{bmatrix}$$

- 2) Compute the gradient vector for a hyperplane.

$$z = f(x) = f(x_1, x_2, \dots, x_N) = \sum_{i=1}^N a_i(x_i - b_i) + c = a_1x_1 + a_2x_2 + \dots + a_Nx_N + c$$

$$\begin{aligned} f_x &= a_1 + a_2 + a_3 + \dots + a_N \\ f_z &= -1 \end{aligned} \Rightarrow \nabla f(x) = \begin{bmatrix} a_1 + a_2 + a_3 + \dots + a_N \\ -1 \end{bmatrix}$$

- 3) Compute the partial derivative of the paraboloid function.

$$z = f(x, y) = A(x - x_0)^2 + B(y - y_0)^2 + c$$

$$\begin{aligned} f_x(x, y) &= \frac{\partial f(x, y)}{\partial x} = 2Ax - 2Ax_0 \\ f_y(x, y) &= \frac{\partial f(x, y)}{\partial y} = 2By - 2By_0 \end{aligned}$$

- 4) Given the following matrices and vectors. Compute the following quantities and specify the shape of the output, if an operation is not defined then just say "not defined"

$$\textcircled{1} \quad x^T = (3 \ 1 \ 4) \\ [1 \times 3]$$

$$\textcircled{2} \quad y^T = \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} \\ [3 \times 1]$$

$$\textcircled{3} \quad B^T = \begin{bmatrix} 3 & 5 & 1 \\ 5 & 2 & 4 \end{bmatrix} \\ [2 \times 3]$$

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④

$$X \cdot X = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix}$$

$$= 9 + 1 + 16 = 26$$

$$[1 \times 1]$$

⑤

$$X \cdot Y^T = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}$$

$$= 6 + 1 + 4 = 11$$

$$[1 \times 1]$$

~~$$X \cdot X = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix}$$~~

⑦

$$Y \cdot X = [2 \ 1] \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix}$$

$$= 6 + 1 + 4 = 11$$

$$[1 \times 1]$$

⑥

$$X \cdot Y = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} [2 \ 1]$$

$$= \begin{bmatrix} 6 & 15 & 3 \\ 2 & 5 & 1 \\ 8 & 20 & 4 \end{bmatrix}$$

$$[3 \times 3]$$

⑧

$$A \cdot X = \begin{bmatrix} 4 & 5 & 2 \\ 3 & 1 & 5 \\ 6 & 4 & 3 \end{bmatrix} \cdot \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix}$$

$$= \begin{bmatrix} 4 \times 3 + 5 \times 1 + 2 \times 4 \\ 3 \times 3 + 1 \times 1 + 5 \times 4 \\ 6 \times 3 + 4 \times 1 + 3 \times 4 \end{bmatrix} = \begin{bmatrix} 35 \\ 30 \\ 34 \end{bmatrix}$$

$$[3 \times 1]$$

⑨

$$A \cdot B = \begin{bmatrix} 4 & 5 & 2 \\ 3 & 1 & 5 \\ 6 & 4 & 3 \end{bmatrix} \cdot \begin{bmatrix} 3 & 5 \\ 5 & 2 \\ 1 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} 4 \times 3 + 5 \times 5 + 2 \times 1 & 4 \times 5 + 5 \times 2 \\ 3 \times 3 + 1 \times 5 + 5 \times 1 & 3 \times 5 + 1 \times 2 \\ 6 \times 3 + 4 \times 5 + 3 \times 1 & 6 \times 5 + 4 \times 2 \end{bmatrix}$$

$$= \begin{bmatrix} 39 & 38 \\ 19 & 37 \\ 41 & 40 \end{bmatrix}$$

$$[3 \times 2]$$

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$$B_{\text{reshape}}(1,6) = (3 \ 5 \ 5 \ 2 \ 1 \ 4)$$

$$[1 \times 6]$$

5) Linear Least Squares (LLS) = Single-variable

$$L_{\text{cp}} = L(m, b) = \sum_{i=1}^N (\hat{y}_i - M(\hat{x}_i, m, b))^2$$

$$= (y_1 - (mx_1 + b))^2 + (y_2 - (mx_2 + b))^2 + \dots + (y_N - (mx_N + b))^2$$

$$= y_1^2 - 2y_1(mx_1 + b) + (mx_1 + b)^2 + y_2^2 - 2y_2(mx_2 + b) + (mx_2 + b)^2 + \dots$$

$$+ y_N^2 - 2y_N(mx_N + b) + (mx_N + b)^2$$

$$\cancel{y_1^2 - 2y_1m}$$

$$= y_1^2 - 2x_1y_1m - 2y_1b + x_1^2m^2 + 2x_1mb + b^2 + \dots + y_N^2 - 2x_Ny_Nm - 2y_Nb + x_N^2m^2 + 2x_Nmb + b^2$$

$$= (y_1^2 + y_2^2 + \dots + y_N^2) - 2m(x_1y_1 + x_2y_2 + \dots + x_Ny_N) - 2b(y_1 + y_2 + \dots + y_N)$$

$$+ (x_1^2 + x_2^2 + \dots + x_N^2)m^2 + 2(x_1 + x_2 + \dots + x_N)mb + Nb^2$$

$$= N \cdot \bar{y}^2 - 2mN \cdot \bar{xy} - 2bN \cdot \bar{y} + m^2 \cdot N \cdot \bar{x}^2 + 2mb \cdot N \cdot \bar{x} + Nb^2$$

$$\frac{dL}{dm} = -2N \bar{xy} + 2N \bar{x}^2 m + 2bN \bar{x} = 0$$

$$\Rightarrow \bar{x}^2 m + b \bar{x} - \bar{xy} = 0$$

$$\frac{dL}{db} = -2N \bar{y} + 2mN \bar{x} + 2Nb = 0$$

$$\Rightarrow m \bar{x} - \bar{y} + b = 0$$

$$\begin{cases} \bar{x}^2 m + b \bar{x} - \bar{xy} = 0 \\ m \bar{x} - \bar{y} + b = 0 \end{cases}$$

$$\Rightarrow \begin{cases} \bar{x}^2 m + (\bar{y} - m \bar{x}) \bar{x} - \bar{xy} = 0 \\ \bar{x}^2 m + \bar{x} \bar{y} - m \bar{x}^2 - \bar{xy} = 0 \end{cases}$$

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$$\begin{cases} m = \frac{\bar{x}\bar{y} - \overline{xy}}{(\bar{x})^2 - \overline{x^2}} \\ b = \bar{y} - m\bar{x} = \bar{y} - \frac{\bar{x}\bar{y} - \overline{xy}}{(\bar{x})^2 - \overline{x^2}} \bar{x} \end{cases}$$

$$\begin{aligned} \text{Since } \text{var}(X) &= \frac{1}{N} \sum_{i=1}^n (x_i - \bar{x})^2 = \cancel{\frac{1}{N} \sum_{i=1}^n x_i^2} \cdot \frac{1}{N} (x_1^2 - 2x_1\bar{x} + (\bar{x})^2 + \dots + x_n^2 - 2x_n\bar{x} + (\bar{x})^2) \\ &= \frac{1}{N} (N \cdot \overline{x^2} - 2N\bar{x} \cdot \bar{x} + N(\bar{x})^2) \\ &= \overline{x^2} - (\bar{x})^2 \end{aligned}$$

$$\begin{aligned} \text{cov}(X, Y) &= \frac{1}{N} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \cancel{\frac{1}{N} \sum_{i=1}^n x_i y_i} \cdot \frac{1}{N} (x_1 y_1 - x_1 \bar{y} - \bar{x} y_1 + \bar{x} \bar{y} + \dots + x_n y_n - x_n \bar{y} - \bar{x} y_n + \bar{x} \bar{y}) \\ &= \frac{1}{N} (N \cdot \overline{xy} - N\bar{x}\bar{y} - N\bar{y}\bar{x} + N\bar{x}\bar{y}) \\ &= \frac{1}{N} (N \cdot \overline{xy} - N\bar{x}\bar{y}) \\ &= \overline{xy} - \bar{x}\bar{y} \end{aligned}$$

$$m = \frac{-\text{cov}(X, Y)}{-\text{var}(X)} = \frac{\text{cov}(X, Y)}{\text{var}(X)}$$

$$b = \bar{y} - \frac{\text{cov}(X, Y)}{\text{var}(X)} \bar{x}$$