

# Solution to homework 2

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## 1 Problem 1

We would like to minimize this function:

$$L(m, b) = \sum_{i=1}^n (mx_i + b - y_i)^2. \quad (1.1)$$

Compute the first-order derivative:

$$\frac{\partial L}{\partial m} = 2 \sum_{i=1}^n (mx_i + b - y_i)x_i, \quad \frac{\partial L}{\partial b} = 2 \sum_{i=1}^n (mx_i + b - y_i). \quad (1.2)$$

Compute the second-order derivative:

$$\frac{\partial^2 L}{\partial m^2} = 2 \sum_{i=1}^n x_i^2, \quad \frac{\partial^2 L}{\partial b^2} = 2 \sum_{i=1}^n 1, \quad \frac{\partial^2 L}{\partial m \partial b} = 2 \sum_{i=1}^n x_i. \quad (1.3)$$

Then the Hessian matrix is:

$$H = \begin{pmatrix} \frac{\partial^2 L}{\partial m^2} & \frac{\partial^2 L}{\partial m \partial b} \\ \frac{\partial^2 L}{\partial m \partial b} & \frac{\partial^2 L}{\partial b^2} \end{pmatrix} = \begin{pmatrix} 2 \sum_{i=1}^n x_i^2 & 2 \sum_{i=1}^n x_i \\ 2 \sum_{i=1}^n x_i & 2 \sum_{i=1}^n 1 \end{pmatrix} \quad (1.4)$$

We introduce the matrix  $X \in \mathbb{R}^{n \times 2}$ :

$$X = \begin{pmatrix} x_1 & 1 \\ x_2 & 1 \\ \dots & \dots \\ x_n & 1 \end{pmatrix} \quad (1.5)$$

Then

$$X^T X = \begin{pmatrix} x_1 & x_2 & \dots & x_n \\ 1 & 1 & \dots & 1 \end{pmatrix} \begin{pmatrix} x_1 & 1 \\ x_2 & 1 \\ \dots & \dots \\ x_n & 1 \end{pmatrix} = \begin{pmatrix} \sum_{i=1}^n x_i^2 & \sum_{i=1}^n x_i \\ \sum_{i=1}^n x_i & \sum_{i=1}^n 1 \end{pmatrix} \quad (1.6)$$

Therefore, we have

$$H = 2X^T X \quad (1.7)$$

For any vector  $z \in \mathbb{R}^2$ , we compute

$$z^T H z = 2z^T X^T X z = 2(Xz)^T Xz = 2\|Xz\|^2 \geq 0. \quad (1.8)$$

Moreover, it is easy to see that  $Xz = 0$  indicates  $z = 0$ , under the assumption that  $x_i$  are not identical. This proves that  $H$  is positive-definite.

## 2 Problem 2

We would like to minimize the function

$$L(m_1, m_2, b) = \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i)^2 \quad (2.1)$$

1. Compute the derivatives:

$$\frac{\partial L}{\partial m_1} = 2 \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i) x_i^{(1)} \quad (2.2)$$

$$\frac{\partial L}{\partial m_2} = 2 \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i) x_i^{(2)} \quad (2.3)$$

$$\frac{\partial L}{\partial b} = 2 \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i) \quad (2.4)$$

2. Setting

$$\frac{\partial L}{\partial m_1} = 0, \quad \frac{\partial L}{\partial m_2} = 0, \quad \frac{\partial L}{\partial b} = 0. \quad (2.5)$$

We have

$$\begin{aligned} \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i) x_i^{(1)} &= 0, \\ \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i) x_i^{(2)} &= 0, \\ \sum_{i=1}^n (m_1 x_i^{(1)} + m_2 x_i^{(2)} + b - y_i) &= 0. \end{aligned} \quad (2.6)$$

We can rewrite it as a linear system of  $(m_1, m_2, b)$ :

$$A \begin{pmatrix} m_1 \\ m_2 \\ b \end{pmatrix} = \begin{pmatrix} \sum_{i=1}^n x_i^{(1)} y_i \\ \sum_{i=1}^n x_i^{(2)} y_i \\ \sum_{i=1}^n y_i \end{pmatrix} \quad (2.7)$$

where

$$A = \begin{pmatrix} \sum_{i=1}^n (x_i^{(1)})^2 & \sum_{i=1}^n x_i^{(1)} x_i^{(2)} & \sum_{i=1}^n x_i^{(1)} \\ \sum_{i=1}^n x_i^{(1)} x_i^{(2)} & \sum_{i=1}^n (x_i^{(2)})^2 & \sum_{i=1}^n x_i^{(2)} \\ \sum_{i=1}^n x_i^{(1)} & \sum_{i=1}^n x_i^{(2)} & \sum_{i=1}^n 1 \end{pmatrix} \quad (2.8)$$

We introduce the matrix  $X \in \mathbb{R}^{n \times 3}$ :

$$X = \begin{pmatrix} x_1^{(1)} & x_1^{(2)} & 1 \\ x_2^{(1)} & x_2^{(2)} & 1 \\ \dots & \dots & \dots \\ x_n^{(1)} & x_n^{(2)} & 1 \end{pmatrix} \quad (2.9)$$

and

$$Y = (y_1, y_2, \dots, y_n)^T. \quad (2.10)$$

Then it is easy to verify that  $A = X^T X$  and the linear system can be rewritten as

$$X^T X \begin{pmatrix} m_1 \\ m_2 \\ b \end{pmatrix} = X^T Y \quad (2.11)$$

Therefore, we can solve out

$$\begin{pmatrix} m_1 \\ m_2 \\ b \end{pmatrix} = (X^T X)^{-1} X^T Y. \quad (2.12)$$

3. By computing second order derivatives, we can prove that the Hessian matrix can be rewritten as:

$$H = 2X^T X \quad (2.13)$$

Therefore, for any vector  $z \in \mathbb{R}^3$ , we compute

$$z^T H z = 2z^T X^T X z = 2(Xz)^T Xz = 2\|Xz\|^2 \geq 0. \quad (2.14)$$

Moreover, it is easy to see that  $Xz = 0$  indicates  $z = 0$ , under the assumption that  $(x_i^{(1)}, x_i^{(2)})$  are not identical. This proves that  $H$  is positive-definite.