

4 Least Squares

P7

example

$$\begin{aligned}\|\vec{x}\omega - \vec{y}\|_2^2 &= (\vec{x}\omega - \vec{y})^T (\vec{x}\omega - \vec{y}) \\ &= \omega^2 \vec{x}^T \vec{x} - 2\omega \vec{x}^T \vec{y} + \vec{y}^T \vec{y}\end{aligned}$$

$$\begin{aligned}\text{Derivative } \frac{d}{d\omega}() &= 2\omega \vec{x}^T \vec{x} - 2\vec{x}^T \vec{y} \stackrel{!}{=} 0 \\ \rightarrow \omega^* &= \underline{\underline{\frac{\vec{x}^T \vec{y}}{\vec{x}^T \vec{x}}}}\end{aligned}$$

P8

example

$$\begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} \omega = \begin{bmatrix} 0 \\ 1.5 \\ 2.2 \end{bmatrix}$$

$$\frac{d}{d\omega} [(\omega - 1.5)^2 + (2\omega - 2.2)^2] = 0$$

$$\omega^* = \underline{\underline{1.18}}$$

thus:

$$y = 1.18 \cdot 10 = \underline{\underline{11.8}}$$

p 11

example

$$X = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 4 \\ 1 & 3 & 9 \end{bmatrix} \quad y = \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix}$$

$$\underline{\underline{\bar{w}^*}} = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix}$$

$$y = [1 \ 5 \ 25] \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix} = \underline{\underline{9}}$$

p 16

$$X = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 4 \\ 1 & 2 & 16 \end{bmatrix} \quad y = \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix}$$

$$\bar{w}^* = \begin{bmatrix} 6.33 \\ -6.0 \\ 0.66 \end{bmatrix}$$

$$y = [1 \ 10 \ 100] \begin{bmatrix} 6.33 \\ -6.0 \\ 0.66 \end{bmatrix} = \underline{\underline{13}}$$

p 17

$$y = w_1 \exp(w_2 x + w_3 x^2) \quad | \quad \ln \circ$$

$$\ln y = \ln(w_1) + \ln \exp(w_2 x + w_3 x^2)$$

$$\ln y = \ln w_1 + w_2 x + w_3 x^2$$

$$\Rightarrow \hat{y} = \hat{w}_1 + w_2 x + w_3 x^2$$

p 18

$$X = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & \frac{1}{2} \end{bmatrix}$$

\Rightarrow if we can compute the eigenvalues they are the same as the ~~of~~ singular vals

$$\lambda_1 = 0.5, \lambda_2 = 1, \lambda_3 = 2$$

$$K = \frac{1}{0.5} = 2 \quad \text{"well conditioned"}$$

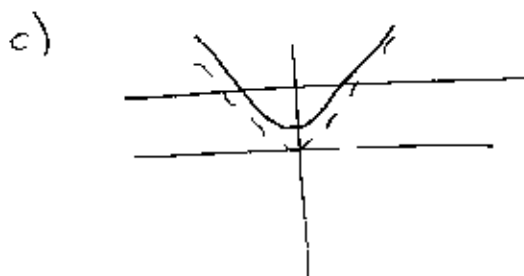
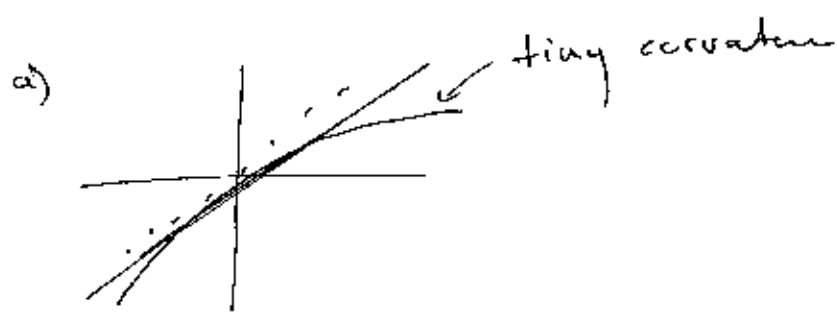
p 20

$$\begin{aligned} \|\mathbf{X}\vec{\omega} - \vec{y}\|_2^2 &= (\mathbf{X}\vec{\omega} - \vec{y})^T (\mathbf{X}\vec{\omega} - \vec{y}) \\ &= \vec{y}^T \vec{y} - \vec{y}^T \mathbf{X} \vec{\omega} - \vec{\omega}^T \mathbf{X}^T \vec{y} + \vec{\omega}^T \mathbf{X}^T \mathbf{X} \vec{\omega} \end{aligned}$$

$$\text{Derivative } \frac{d}{d\vec{\omega}} = -2\mathbf{X}^T \vec{y} + 2\mathbf{X}^T \mathbf{X} \vec{\omega} = 0$$

$$\Rightarrow \omega^* = \underline{(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \vec{y}}$$

p 21



p22

$$\omega_L^* = \argmin \sum_{i \neq j} (\cancel{x_i} + \omega_L - \cancel{x_i} - \hat{\omega}_L)^2 + (\cancel{x_j} + \omega_L - \cancel{x_j} - \hat{\omega}_L - \varepsilon_j)^2$$

$$\begin{aligned} \frac{d}{d\omega_L} (c.) &= 2 \sum_{i \neq j} (\omega_L - \hat{\omega}_L) + 2(\omega_L - \hat{\omega}_L - \varepsilon_j) \stackrel{!}{=} 0 \\ &= (N-1)(\omega_L - \hat{\omega}_L) + \omega_L - \hat{\omega}_L - \varepsilon_j = 0 \\ \rightarrow \omega_L &= \frac{\hat{\omega}_L N + \varepsilon_j}{N} \end{aligned}$$

p23

$$X = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \end{bmatrix} \leadsto \text{rank} = 2 \Rightarrow \text{not invertible}$$

decomposition (from the tips)

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

hence:

$$X^{-1} \approx \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \underline{\underline{\begin{bmatrix} 1 & 0 \\ 0 & 2 \\ 0 & 0 \end{bmatrix}}}$$

then:

$$X \hat{\omega} = \hat{y} \Rightarrow \hat{\omega}^* = X^{-1} \hat{y} = \underline{\underline{\begin{bmatrix} 2 \\ 10 \\ 0 \end{bmatrix}}}$$