

Løsningsforslag, STK4040, uke 44

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Oppgave 1

Vi har at

$$\text{MSE}(\hat{y}) = \mathbb{E}[(\hat{y} - Y)^2] \quad (1)$$

$$= \mathbb{E}[\hat{y}^2] - 2\mathbb{E}[\hat{y}Y] + \mathbb{E}[Y^2] \quad (2)$$

$$= \mathbb{E}[\hat{y}^2] - 2\mathbb{E}\hat{y}\mathbb{E}Y + \mathbb{E}[Y^2], \quad (3)$$

hvor siste overgang gjelder fordi Y og \hat{y} er uavhengige, gitt \mathbf{x} . Videre er

$$\text{Bias}^2(\hat{y}) = (\mathbb{E}[\hat{y} - Y])^2 = (\mathbb{E}\hat{y} - \mathbb{E}Y)^2 \quad (4)$$

$$= (\mathbb{E}\hat{y})^2 - 2\mathbb{E}\hat{y}\mathbb{E}Y + (\mathbb{E}Y)^2 \quad (5)$$

og

$$\text{Var}(\hat{y} - Y) = \text{Var}(\hat{y}) - 2\text{Cov}(\hat{y}, Y) + \text{Var}(Y) \quad (6)$$

$$= \text{Var}(\hat{y}) + \text{Var}(Y) \quad (7)$$

$$= \mathbb{E}[(\hat{y} - \mathbb{E}\hat{y})^2] + \mathbb{E}[(Y - \mathbb{E}Y)^2] \quad (8)$$

$$= \mathbb{E}[\hat{y}^2] - (\mathbb{E}\hat{y})^2 + \mathbb{E}[Y^2] - (\mathbb{E}Y)^2. \quad (9)$$

Vi ser da at $\text{MSE}(\hat{y}) = \text{Bias}^2(\hat{y}) + \text{Var}(\hat{y} - Y)$.

Oppgave 2

Det er en trykkfeil i oppgaveteksten: residualene e_i er lik $y_i - \hat{y}_i$, ikke $\hat{y}_i - y_i$.

a)

For å beregne $e_{(i)}$, trenger vi $\mathbf{b}_{(i)}$. Vi har at $\mathbf{b}_{(i)} = (\mathbf{X}_{(i)}^t \mathbf{X}_{(i)})^{-1} \mathbf{X}_{(i)}^t \mathbf{y}_{(i)}$. Første del av dette uttrykket (inversen) kjenner vi fra oppgaveteksten. Siste del er

$$\mathbf{X}_{(i)}^t \mathbf{y}_{(i)} = (\mathbf{x}_1, \dots, \mathbf{x}_{i-1}, \mathbf{x}_{i+1}, \dots, \mathbf{x}_n) \begin{pmatrix} y_1 \\ \vdots \\ y_{i-1} \\ y_{i+1} \\ \vdots \\ y_n \end{pmatrix} \quad (10)$$

$$= \mathbf{x}_1 y_1 + \dots + \mathbf{x}_{i-1} y_{i-1} + \mathbf{x}_{i+1} y_{i+1} + \dots, \mathbf{x}_n y_n \quad (11)$$

$$= \sum_{j \neq i} \mathbf{x}_j y_j = \sum_{j=1}^n \mathbf{x}_j y_j - \mathbf{x}_i y_i \quad (12)$$

$$= \mathbf{X}^t \mathbf{y} - \mathbf{x}_i y_i, \quad (13)$$

så vi får

$$\mathbf{b}_{(i)} = \left((\mathbf{X}^t \mathbf{X})^{-1} + \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i \mathbf{x}_i^t (\mathbf{X}^t \mathbf{X})^{-1}}{1 - h_i} \right) (\mathbf{X}^t \mathbf{y} - \mathbf{x}_i y_i) \quad (14)$$

$$= (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{y} - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i y_i \quad (15)$$

$$+ \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i \mathbf{x}_i^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{y}}{1 - h_i} - \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i \mathbf{x}_i^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i y_i}{1 - h_i}. \quad (16)$$

Nå er $(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{X}^t \mathbf{y} = \mathbf{b}$ og $\mathbf{x}_i^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i = h_i$, så

$$\mathbf{b}_{(i)} = \mathbf{b} - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i y_i + \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i \mathbf{x}_i^t \mathbf{b}}{1 - h_i} - \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i h_i y_i}{1 - h_i} \quad (17)$$

$$= \mathbf{b} - \frac{(1 - h_i)(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i y_i - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i \mathbf{x}_i^t \mathbf{b} + h_i (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i y_i}{1 - h_i} \quad (18)$$

$$= \mathbf{b} - \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i y_i - (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i \mathbf{x}_i^t \mathbf{b}}{1 - h_i} \quad (19)$$

$$= \mathbf{b} - \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i (y_i - \mathbf{x}_i^t \mathbf{b})}{1 - h_i} \quad (20)$$

$$= \mathbf{b} - \frac{(\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i}{1 - h_i} e_i. \quad (21)$$

$$(22)$$

Dermed blir

$$e_{(i)} = y_i - \mathbf{x}_i^t \mathbf{b}_{(i)} \quad (23)$$

$$= y_i - \mathbf{x}_i^t \mathbf{b} + \frac{\mathbf{x}_i^t (\mathbf{X}^t \mathbf{X})^{-1} \mathbf{x}_i}{1 - h_i} e_i \quad (24)$$

$$= e_i + \frac{h_i}{1 - h_i} e_i \quad (25)$$

$$= \frac{1}{1 - h_i} e_i, \quad (26)$$

hvilket skulle bevises.

b)

Denne oppgaven var kanskje litt uklart formulert. Poenget var å vise at $h_i = 1/n + (\mathbf{x}_i - \bar{\mathbf{x}})^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} (\mathbf{x}_i - \bar{\mathbf{x}})$, hvor $\bar{\mathbf{x}}$ er kolonnesnittene til \mathbf{X}_1 . $g_i = h_i - 1/n$ kan dermed tolkes som leverage'ne til en regresjon med sentrerte variabler.

Siden vi har separert ut konstantleddet fra \mathbf{x} , så er

$$h_i = (1 - \mathbf{x}_i) (\mathbf{X}^t \mathbf{X})^{-1} \begin{pmatrix} 1 \\ \mathbf{x}_i \end{pmatrix}. \quad (27)$$

Vi har at

$$\mathbf{X}^t \mathbf{X} = \begin{pmatrix} \mathbf{1}^t \\ \mathbf{X}_1^t \end{pmatrix} (\mathbf{1} \quad \mathbf{X}_1) = \begin{pmatrix} n & \mathbf{1}^t \mathbf{X}_1 \\ \mathbf{X}_1^t \mathbf{1} & \mathbf{X}_1^t \mathbf{X}_1 \end{pmatrix} = \begin{pmatrix} n & n \bar{\mathbf{x}}^t \\ n \bar{\mathbf{x}} & \mathbf{X}_1^t \mathbf{X}_1 \end{pmatrix}. \quad (28)$$

Hed hjelp av A.2.4(VII) får vi da at

$$(\mathbf{X}^t \mathbf{X})^{-1} = \begin{pmatrix} \frac{1}{n} & \mathbf{0}^t \\ \mathbf{0} & 0 \end{pmatrix} + \begin{pmatrix} -\frac{1}{n} n \bar{\mathbf{x}}^t \\ \mathbf{I} \end{pmatrix} (\mathbf{X}_1^t \mathbf{X}_1 - n \bar{\mathbf{x}} \bar{\mathbf{x}}^t)^{-1} \begin{pmatrix} -\frac{1}{n} n \bar{\mathbf{x}} & \mathbf{I} \end{pmatrix}. \quad (29)$$

Det er lett å vise at $\mathbf{X}_1^t \mathbf{X}_1 - n \bar{\mathbf{x}} \bar{\mathbf{x}}^t = \mathbf{X}_c^t \mathbf{X}_c$, så

$$(\mathbf{X}^t \mathbf{X})^{-1} = \begin{pmatrix} \frac{1}{n} & \mathbf{0}^t \\ \mathbf{0} & 0 \end{pmatrix} + \begin{pmatrix} -\bar{\mathbf{x}}^t \\ \mathbf{I} \end{pmatrix} (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \begin{pmatrix} -\bar{\mathbf{x}} & \mathbf{I} \end{pmatrix} \quad (30)$$

$$= \begin{pmatrix} \frac{1}{n} & \mathbf{0}^t \\ \mathbf{0} & 0 \end{pmatrix} + \begin{pmatrix} -\bar{\mathbf{x}}^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \\ (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \end{pmatrix} \begin{pmatrix} -\bar{\mathbf{x}} & \mathbf{I} \end{pmatrix} \quad (31)$$

$$= \begin{pmatrix} \frac{1}{n} + \bar{\mathbf{x}}^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \bar{\mathbf{x}} & -\bar{\mathbf{x}}^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \\ -(\mathbf{X}_c^t \mathbf{X}_c)^{-1} \bar{\mathbf{x}} & (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \end{pmatrix}. \quad (32)$$

Dette gir

$$h_i = \begin{pmatrix} 1 & \mathbf{x}_i \end{pmatrix} (\mathbf{X}^t \mathbf{X})^{-1} \begin{pmatrix} 1 \\ \mathbf{x}_i \end{pmatrix} \quad (33)$$

$$= \frac{1}{n} + \bar{\mathbf{x}}^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \bar{\mathbf{x}} - \mathbf{x}_i^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \bar{\mathbf{x}} - \bar{\mathbf{x}}^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \mathbf{x}_i + \mathbf{x}_i^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \mathbf{x}_i \quad (34)$$

$$= \frac{1}{n} - (\mathbf{x}_i - \bar{\mathbf{x}})^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \bar{\mathbf{x}} + (\mathbf{x}_i - \bar{\mathbf{x}})^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} \mathbf{x}_i \quad (35)$$

$$= \frac{1}{n} + (\mathbf{x}_i - \bar{\mathbf{x}})^t (\mathbf{X}_c^t \mathbf{X}_c)^{-1} (\mathbf{x}_i - \bar{\mathbf{x}}). \quad (36)$$