

10.9 Test på forkholdsbal

p_1 = andelen Volvoeigarar i Noreg

p_2 = - - - - i Sverige

X_1 = talet på Volvoeigarar i utvalget i Noreg
i Sverige.

X_2 = - - - - i Sverige.

$X_1 \sim B(m_1, p_1)$, $X_2 \sim B(m_2, p_2)$

$$\hat{p}_1 = \frac{X_1}{m_1}, \quad \hat{p}_2 = \frac{X_2}{m_2}, \quad \text{Var} \hat{p}_1 = \frac{p_1(1-p_1)}{m_1}, \quad \text{Var} \hat{p}_2 = \frac{p_2(1-p_2)}{m_2}$$

$$H_0: p_1 = p_2$$

$$H_1: \begin{cases} p_1 > p_2 \\ p_1 \neq p_2 \\ p_1 < p_2 \end{cases}$$

Før stor moh m_1 og m_2 er

$$Z = \frac{\hat{p}_1 - \hat{p}_2 - (p_1 - p_2)}{\sqrt{\frac{p_1(1-p_1)}{m_1} + \frac{p_2(1-p_2)}{m_2}}} \stackrel{N(0,1)}{\sim}$$

Sidom p_1 og p_2 er ukjende, er det vanlig å erstattre p_1 og p_2 med $\hat{p} = \frac{X_1 + X_2}{m_1 + m_2}$

$$\text{og bruke at } Z^* = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p})\left(\frac{1}{m_1} + \frac{1}{m_2}\right)}} \stackrel{N(0,1)}{\sim}$$

Kap 11 Regressjonsanalyse

Føremål: Finne sammenhengar mellom ein responsvariabel og ein eller fleire forklaringsvariable. Regressjonsanalyse er eit modellingsverktøy.

11.2. Enkel linjær regresjon (linjer i koefisientane)

$$\text{Modell: } \hat{y}_i = \alpha + \beta x_i + \varepsilon_i \quad \begin{array}{l} \text{mavh.} \\ \bar{\varepsilon}_i = 0, \text{Var}[\varepsilon_i] = \sigma^2 \end{array}$$

↑ ↑
 respons regresjonsvariabel
 avh. variabel mabh. variabel
 forklaringsvariabel

x_i kan vere ein transformasjon av andre variablene.

$$\text{då } x_i = \ln(a_i), \quad x_i = \sin(\theta_i) \text{ o.s.v.}$$

Observasjoner $(y_1, x_1), (y_2, x_2), \dots, (y_m, x_m)$ skal oppfylle

$$y_1 = \alpha + \beta x_1 + \varepsilon_1$$

$$y_2 = \alpha + \beta x_2 + \varepsilon_2$$

$$\vdots$$

$$y_m = \alpha + \beta x_m + \varepsilon_m$$

Skal finne estimat, $\stackrel{\text{og } b}{\text{for }} \alpha$, $\text{og } \beta$.

Minske kvadratsumms metode

Finn da α og β som minimiserer

$$Q = \sum_{i=1}^m \varepsilon_i^2 = \sum_{i=1}^m (y_i - \alpha - \beta x_i)^2$$

$$\frac{\partial Q}{\partial \alpha} = 0 \Leftrightarrow -2 \sum_{i=1}^m (y_i - \alpha - \beta x_i) = 0$$

$$\frac{\partial Q}{\partial b} = 0 \iff -2 \sum_{i=1}^m (y_i - a - bx_i) x_i = 0$$

Dette gir dei 2 normallikningane

$$\sum_{i=1}^m a + b \sum_{i=1}^m x_i = \sum_{i=1}^m y_i \quad (1)$$

$$\sum_{i=1}^m ax_i + b \sum_{i=1}^m x_i^2 = \sum_{i=1}^m x_i y_i \quad (2)$$

Løyser vi (1) for a får vi:

$$ma + b \sum_{i=1}^m x_i = \sum_{i=1}^m y_i \Rightarrow a = \bar{y} - b \bar{x}$$

Som inntell i (2) gir vi:

$$(\bar{y} - b \bar{x}) \sum_{i=1}^m x_i + b \sum_{i=1}^m x_i^2 = \sum_{i=1}^m x_i y_i$$

$$\Rightarrow b \left(\sum_{i=1}^m x_i^2 - m \bar{x}^2 \right) = \sum_{i=1}^m x_i (y_i - \bar{y})$$

$$\Rightarrow b = \frac{\sum_{i=1}^m x_i (y_i - \bar{y})}{\sum_{i=1}^m x_i^2 - m \bar{x}^2} = \frac{\sum_{i=1}^m (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^m (x_i - \bar{x})^2} = \frac{\sum_{i=1}^m (x_i - \bar{x}) y_i}{\sum_{i=1}^m (x_i - \bar{x})^2}$$

$$\text{Hvis vi bruker } \sum_{i=1}^m \bar{x} (y_i - \bar{y}) = \bar{x} \sum_{i=1}^m (y_i - \bar{y}) = 0$$

$$\sum_{i=1}^m (x_i - \bar{x})^2 = \sum_{i=1}^m (x_i^2 - 2x_i \bar{x} + \bar{x}^2) = \sum_{i=1}^m x_i^2 - 2m \bar{x}^2 + m \bar{x}^2 = \sum_{i=1}^m x_i^2 - m \bar{x}^2$$

$$\sum_{i=1}^m \bar{y} (x_i - \bar{x}) = \bar{y} \sum_{i=1}^m (x_i - \bar{x}) = 0$$

Estimators for β and α are given by

$$\hat{\beta} = \beta = \frac{\sum_{i=1}^m (x_i - \bar{x}) y_i}{\sum_{i=1}^m (x_i - \bar{x})^2} \quad \text{og} \quad \hat{\alpha} = A = \bar{y} - \hat{\beta} \bar{x}$$

$$E[\hat{\beta}] = E\left[\frac{\sum_{i=1}^m (x_i - \bar{x}) E[y_i]}{\sum_{i=1}^m (x_i - \bar{x})^2}\right] = \frac{\sum_{i=1}^m (x_i - \bar{x})(\alpha + \beta x_i)}{\sum_{i=1}^m (x_i - \bar{x})^2} = \beta \frac{\sum_{i=1}^m (x_i - \bar{x}) x_i}{\sum_{i=1}^m (x_i - \bar{x})^2}$$

$$= \beta \frac{\sum_{i=1}^m (x_i - \bar{x})^2}{\sum_{i=1}^m (x_i - \bar{x})^2}$$

$$E[\hat{A}] = E[\bar{y}] - \hat{\beta} \bar{x} = \frac{\sum_{i=1}^m E[y_i]}{m} - \hat{\beta} \bar{x} = \frac{\sum_{i=1}^m (\alpha + \beta x_i)}{m} - \hat{\beta} \bar{x}$$

$$= \alpha + \beta \bar{x} - \hat{\beta} \bar{x} = \alpha.$$

$$\text{Var}[\hat{\beta}] \stackrel{\text{p.g.a}}{=} \frac{\sum_{i=1}^m \text{Var}[(x_i - \bar{x}) y_i]}{\left(\sum_{i=1}^m (x_i - \bar{x})^2\right)^2} = \frac{\sum_{i=1}^m (x_i - \bar{x})^2 \cdot \sigma^2}{\left(\sum_{i=1}^m (x_i - \bar{x})^2\right)^2} = \frac{\sigma^2}{\sum_{i=1}^m (x_i - \bar{x})^2}$$

$$\text{Var}[\hat{A}] \stackrel{\text{p.g.a}}{=} \text{Var}\left[\bar{y} - \hat{\beta} \bar{x}\right] = \frac{\sigma^2}{m} + \frac{\bar{x}^2 \cdot \sigma^2}{\sum_{i=1}^m (x_i - \bar{x})^2} = \sigma^2 \left[\frac{1}{m} + \frac{\bar{x}^2}{\sum_{i=1}^m (x_i - \bar{x})^2} \right]$$

Estimate for σ^2 , $s^2 = \frac{\sum_{i=1}^m (y_i - \hat{y}_i)^2}{m-2} = \frac{\sum_{i=1}^m (y_i - \hat{y}_i)^2}{m-2}$

$$s^2 = \frac{\sum_{i=1}^m (y_i - \hat{y}_i)^2}{m-2}, \quad \text{der } \hat{y}_i = A + \hat{\beta} x_i = \hat{\alpha} + \hat{\beta} x_i$$