

Statistikk, Innlevering 4, Vegard G. Jord

Oppg. 1a)

$$\bar{X} \sim N(\mu, \sigma^2)$$

$$P(X > \mu + \sigma) = P\left(\frac{\bar{X} - \mu}{\sigma} > 1\right)$$

$$Z := \frac{\bar{X} - \mu}{\sigma} \sim N(0, 1)$$

$$P\left(\frac{\bar{X} - \mu}{\sigma} > 1\right) = 1 - P(Z \leq 1) = 1 - 0,8413$$

$$P(\bar{X} > 10,29) = 0,1587$$

$$\bar{Y} := \frac{X_1 + X_2}{2} \rightarrow \sigma_Y = \frac{\sigma_X}{\sqrt{2}}, \mu_Y = \mu_X$$

$$P(|\bar{Y} - \mu_X| > \sigma_X) = P(\bar{Y} - \mu_X < -\sigma_X) + P(\bar{Y} - \mu_X > \sigma_X)$$

$$= P(\bar{Y} < \mu_X - \sigma_X) + P(\bar{Y} > \mu_X + \sigma_X)$$

$$Z := \frac{\bar{Y} - \mu_X}{\sigma_Y} = \frac{\bar{Y} - \mu_X}{\frac{\sigma_X}{\sqrt{2}}} \sim N(0, 1)$$

$$P(|\bar{Y} - \mu_X| > \sigma_X) = P\left(Z < \frac{-\sigma_X}{\frac{\sigma_X}{\sqrt{2}}}\right) + P(Z > \sqrt{2})$$

$$= 2 \cdot P(Z < -\sqrt{2})$$

$$= 0,1486$$

Oppg 16) $\hat{\mu}_A = \bar{x}_1 \rightarrow E(\hat{\mu}_A) = E(\bar{x}_1) = 10g$

$E(\hat{\mu}_B) = 10g$

$Var(\hat{\mu}_A) = Var(\bar{x}_1) = 0,2^2 g^2$

$Var(\hat{\mu}_B) = 0,2^2 g^2$

$\Sigma_1 := \bar{x}_1 + \bar{x}_2 \quad \Sigma_2 = \bar{x}_1 - \bar{x}_2$

$\tilde{\mu}_A := \frac{\Sigma_1 + \Sigma_2}{2} \quad \tilde{\mu}_B := \frac{\Sigma_1 - \Sigma_2}{2}$

$E(\tilde{\mu}_A) = \frac{1}{2} E(\Sigma_1 + \Sigma_2) = \frac{1}{2} E(2\bar{x}_1) = 10g$

$E(\tilde{\mu}_B) = \frac{1}{2} E(\Sigma_1 - \Sigma_2) = E(\bar{x}_2) = 10g$

$Var(\tilde{\mu}_A) = Var\left(\frac{\Sigma_1 + \Sigma_2}{2}\right) = \frac{1}{4} Var(\Sigma_1 + \Sigma_2)$
 $= \frac{1}{4} (Var(\Sigma_1) + Var(\Sigma_2) + Cov(\Sigma_1, \Sigma_2))$

$Var(\Sigma_1) = Var(\bar{x}_1) + Var(\bar{x}_2) = 2\sigma_x^2$

$Var(\Sigma_2) = Var(\bar{x}_1) + Var(\bar{x}_2) = 2\sigma_x^2$

$Cov(\Sigma_1, \Sigma_2) = E(\Sigma_1 \Sigma_2) - E(\Sigma_1) E(\Sigma_2)$
 $= E((\bar{x}_1 + \bar{x}_2)(\bar{x}_1 - \bar{x}_2)) \stackrel{=0}{=}$
 $= E(\bar{x}_1^2) - E(\bar{x}_2^2)$
 $= 0$

$\rightarrow Var(\tilde{\mu}_A) = \sigma_x^2$ dette er feil, man vet ikke hvorfor " $Var[\sum a_i x_i] = \sum a_i^2 Var[x_i]$

Gitt at Fesit er riktig og

$\text{Var}(\tilde{\mu}_n) = \frac{1}{2} \sigma^2$ Vile jeg ønsket
 $\tilde{\mu}_n$ fordi mindre varians og den
er forventningsrett.

Oppg. 2a)

$$P(Z > 10) = 1 - P(Z \leq 10) \\ = 1 - F(10, \lambda)$$

$$P(Z > 10) = 0,61$$

$$P(Z > 20 | Z > 10) = \frac{P(Z > 20)}{P(Z > 10)} = \frac{e^{-1}}{e^{-0.5}} = 0,61$$

$$\begin{aligned} Z(0) \quad Z < 1 \rightarrow M=0 \quad P(M=0) &= F(1, \lambda) \\ 1 < Z < 2 \rightarrow M=1 \quad P(M=1) &= P(1 < Z < 2) = F(2, \lambda) - F(1, \lambda) \\ m < Z < (m+1) \rightarrow M=m \quad P(M=m) &= P(m < Z < (m+1)) \end{aligned}$$

$$\begin{aligned} &= F(m+1, \lambda) - F(m) \\ &= 1 - e^{-\lambda(m+1)} - (1 - e^{-\lambda m}) \\ &= -e^{-\lambda} e^{-\lambda m} + e^{-\lambda m} \end{aligned}$$

$$P(M=m) = e^{-\lambda m} (1 - e^{-\lambda}) \quad \square$$

$$ZC) L(m_1, m_2, \dots, m_n; \lambda) = \prod_{i=1}^n (1 - e^{-\lambda}) e^{-\lambda m_i} = (1 - e^{-\lambda})^n \exp\left(-\lambda \sum_{i=1}^n m_i\right)$$

$$l(\vec{m}; \lambda) = \ln(L(\vec{m}; \lambda)) \\ = n \ln(1 - e^{-\lambda}) - \lambda \sum_{i=1}^n m_i$$

$$\frac{dl}{d\lambda} = 0 \rightarrow \frac{ne^{-\lambda}}{1 - e^{-\lambda}} - \sum_{i=1}^n m_i = 0$$

$$ne^{-\lambda} = (1 - e^{-\lambda}) \sum m_i$$

$$e^{-\lambda} (n + \sum m_i) = \sum m_i$$

$$\lambda = \ln\left(\frac{n + \sum m_i}{\sum m_i}\right)$$

$$\frac{d^2 l}{d\lambda^2} = n \left(\frac{-e^{-\lambda}}{1 - e^{-\lambda}} + e^{-\lambda} (e^{-\lambda} (1 - e^{-\lambda})^2) \right) \\ = \frac{-ne^{-\lambda}}{(1 - e^{-\lambda})^2} < 0 \rightarrow \lambda = \ln\left(\frac{n}{\sum m_i} + 1\right) \\ \text{er maks}$$

oppg 3) kumulativ sannsynlighet for at
base komponent i der er

$$F_{x_i}(x) \prod_{j \neq i} (1 - F_{x_j}(x))$$

$\uparrow$ komp. i der $\nwarrow$ ingen andre der

Sannsynlighet for at en vilkårlig komp. der

$$P(\text{ended}) = \sum_i F_{x_i}(x) \prod_{j \neq i} (1 - F_{x_j}(x)) \quad \text{Binomisk}$$

$$P(\text{ended}) = n F_x(x) (1 - F_x(x))^{n-1} \quad \checkmark \text{ fordelt}$$

$$= n (1 - e^{-(\lambda x)^\alpha}) e^{(1-n)(\lambda x)^\alpha}$$

$$= n (e^{(1-n)(\lambda x)^\alpha} - e^{-n(\lambda x)^\alpha})$$

Oppg. 4)

$$f(x) = \lambda e^{-\lambda x}, x > 0 \quad \Sigma = X_1 + X_2$$

$$\text{Shal vise: } f(y) = \lambda^2 y e^{-\lambda y}, y > 0$$

Finne kumulativ sannsynlighet for Σ

$$P(\Sigma < y) = P(X_1 + X_2 < y) = P(X_2 < y - X_1)$$

$$= \int_0^y \int_0^{y-x_1} f(x_1) f(x_2) dx_2 dx_1$$

$$= \int_0^y \lambda^2 e^{-\lambda x_1} \frac{1}{\lambda} [e^{-\lambda x_2}]_0^{y-x_1} dx_1$$

$$= -\lambda \int_0^y e^{-\lambda x_1} (e^{-\lambda(y-x_1)} - 1) dx_1$$

$$= -\lambda \int_0^y e^{-\lambda y} - e^{-\lambda x_1} dx_1$$

$$P(\Sigma < y) = -\lambda y e^{-\lambda y} - e^{-\lambda y} + 1$$

$$f(y) = \frac{dP(\Sigma < y)}{dy} = \lambda^2 y e^{-\lambda y} - \lambda e^{-\lambda y} + \lambda e^{-\lambda y}$$

$$f(y) = \lambda^2 y e^{-\lambda y} \quad \square$$
