a) Swsker à vise at $E(X_i) = e^{\mu + \frac{1}{2}\sigma^2}$. Tar whospende punkt i don moment gonererende funktionen for normal fordelingen: $M_Y(t) = e^{\mu t + \frac{1}{2}\sigma^2 t}$; sammen med def. av megf:

DEFINITION

The **moment generating function** (mgf) of a continuous random variable *X* is

$$M_X(t) = E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} f(x) dx.$$

As in the discrete case, we will say that the moment generating function exists if $M_X(t)$ is defined for an interval of numbers that includes zero in its interior, which means that it includes both positive and negative values of t.

Vi haer da:

$$E(X_i) = E(e^{Y_i}) = E(e^{Y_i}) = M_Y(1) = e^{M + \frac{1}{2}\sigma^2}$$

Vise E(X:2) er avalogt:

$$\pm (x_i^2) = \pm ((e^{x_i})^2) = \pm (e^{2x_i}) = H_Y(2)$$
 $= e^{2x_i + 1 + 2}$

b) Bruker momentmeteden med resultatet fra a):

$$E(\chi_{i}) = \frac{1}{n} \sum_{i=1}^{n} \chi_{i} = \chi = e^{\hat{n} + \frac{1}{2}\hat{\sigma}^{2}}$$

$$w\chi = \hat{n} + \frac{1}{4}\hat{\sigma}^{2}$$

$$\hat{\mu} = w\chi - \frac{1}{2}\hat{\sigma}^{2}$$

$$= [e^{2\hat{n}}] e^{2\hat{\sigma}^{2}}$$

$$\hat{\mu} = \ln x - \frac{1}{2} \left[\ln \left(\frac{1/n \sum_{i=1}^{n} x_i^2}{x^2} \right) \right]$$

$$= \ln x - \frac{1}{2} \ln |n - \frac{1}{2} \ln \sum_{i=1}^{n} x_i^2 + \frac{1}{2} \ln x^2$$

$$= 2 \ln x - \frac{1}{2} \ln n - \frac{1}{2} \ln \sum_{i=1}^{n} x_i^2$$

$$= \ln \left(x^2 / n \cdot \sum_{i=1}^{n} x^2 / n \right)$$

ALTERNATIVT!

$$Var(X_i) = E(X_i^2) - E(X_i)^2 = [e^{2\hat{\mu}+2\hat{\sigma}^2}] - [e^{\hat{\mu}+\frac{1}{2}\hat{\sigma}^2}]^2$$

$$= [e^{2\hat{\mu}+2\hat{\sigma}^2}] - [e^{2\hat{\mu}+\hat{\sigma}^2}]$$

$$= [e^{2\hat{\mu}+2\hat{\sigma}^2}] - [e^{2\hat{\mu}+\hat{\sigma}^2}]$$

$$= e^{2\hat{\mu}+\hat{\sigma}^2}[e^{\hat{\sigma}^2}-1] \qquad (I)$$

$$= e^{2\hat{\mu}+\hat{\sigma}^2} \Leftrightarrow = e^{2\hat{\mu}+\hat{\sigma}^2} \qquad (II)$$

Likeniney (I) og (II) gir:

$$Var(X_i) = [X^2](\hat{c}^2 - 1)$$

$$\hat{c}^2 = lu\left(\frac{Var(X_i)}{X^2} + 1\right)$$

$$(III)$$

Libering (*) og (III) gir:

$$\overline{X} = e^{\hat{\mu}} \cdot \exp\left(\operatorname{Im}\left(\frac{\operatorname{var}(X_i)}{\overline{X}^2} + 1\right)^2\right)$$

$$\overline{X} = e^{\hat{\mu}}$$

$$\overline{X}^2 = e^{\hat{\mu}}$$

$$\overline{X}^2 = e^{\hat{\mu}}$$

$$\overline{X}^2 = e^{\hat{\mu}}$$

$$\overline{X}^2 + 1$$

Dette or de som stour par Wilhipedia.

Oppogano 1

C) La $f(X_1, X_1; \mu, o^2)$ voire punkt-sammyulicylaten for at X_1, \dots, X_n inntreffer, denne or get ved produktet:

$$f(x_1...x_{n_1}\mu_1o^2) = \prod_{i=1}^{n} \frac{1}{[n_i \circ x_i]^2/2o^2}$$

Vi tar logaritmen:

$$[M] [f(x_1 - x_n) \mu_1 \sigma^2] = [M] [\frac{1}{12\pi} \sqrt{\frac{1}{12\pi} \sigma_{X_i}} \cdot e^{[(u_{X_i} - \mu_i)^2/2\sigma^2]}]$$

$$= \sum_{i=1}^{n} [M] [\frac{1}{12\pi} \sqrt{\frac{1}{2\pi} \sigma_{X_i}} \cdot e^{[(u_{X_i} - \mu_i)^2/2\sigma^2]}]$$

$$= \sum_{i=1}^{n} - [M] \sqrt{2\pi} \mu_{X_i} - [M] \chi_{i-\mu} \int_{0}^{1} \frac{1}{2\sigma^2}$$

$$= \sum_{i=1}^{n} - [M] \sqrt{2\pi} \chi_{i} - [M] \chi_{i-\mu} \int_{0}^{1} \frac{1}{2\sigma^2}$$

Firmor MLE ved à derivere muly parameterne u og o :

$$\frac{\partial}{\partial u} \ln \left[f(x_1...x_u; \mu, \sigma^2) \right] = \frac{\partial}{\partial u} \sum_{i=1}^{u} - \ln \int x_i - \ln u - (\ln x_i)^2 - 2\mu \ln x_i + \mu^2 \right] \frac{1}{2\sigma^2}$$

$$= \sum_{i=1}^{n} \frac{\ln x_i}{\sigma^2} - \frac{\mu}{\sigma^2}$$

$$= -\frac{n\mu}{\sigma^2} + \frac{1}{\sigma^2} \sum_{i=1}^{n} \ln x_i$$

Setter We will for a funce toppoutlet:

$$0 = -\frac{m \cdot \mu}{\sigma^2} + \frac{1}{\sigma^2} \sum_{i=1}^{n} \ln x_i = -n \cdot \mu + \sum_{i=1}^{n} \ln x_i$$

$$\hat{\mu}_{ab} = \frac{1}{n} \sum_{i=1}^{n} \ln x_i$$

Whiting, if Williagedia,

$$\frac{8}{8 \mu} \ln \left[f(x_1...x_n; \mu_1, \sigma^2) \right] = \sum_{i=1}^{n} - \ln \sqrt{2\pi} x_i - \ln \sigma - ((\ln x_i)^2 - 2\mu \ln x_i + \mu^2) \cdot \frac{1}{2\sigma^2}$$

$$= \sum_{i=1}^{n} - \frac{1}{\sigma} + ((\ln x_i)^2 - 2\mu \ln x_i + \mu^2) \cdot \frac{2}{2\sigma^2}$$

$$= -\frac{n}{\sigma} + \frac{1}{\sigma^3} \sum_{i=1}^{n} ((\ln x_i)^2 - 2\mu \ln x_i + \mu^2)$$

$$= -\frac{n}{\sigma} + \frac{1}{\sigma^3} \sum_{i=1}^{n} ((\ln x_i)^2 - 2\mu \ln x_i + \mu^2)$$

$$= -\frac{n}{\sigma} + \frac{1}{\sigma^3} \sum_{i=1}^{n} ((\ln x_i)^2 - 2\mu \ln x_i + \mu^2)$$

setter like will for a flure toppound :

$$O = -\frac{N}{\sigma} + \frac{1}{\sigma^3} \sum_{i=1}^{N} (\ln x - \mu)^2$$

$$\frac{2}{\sigma_{ads}} = \frac{1}{N} \cdot \sum_{i=1}^{N} (\ln x - \mu)^2$$
Rilly, if whippedia.

Vi har at $Y_i = \log X_i$ $\sim N(\mu_i o^2)$. MLE for normal fordulingen or gitt ved: $\hat{\mu}_{mh} = \overline{y} = \frac{1}{n} \sum_{i=1}^{n} y_i \quad \text{of} \quad \hat{\sigma}_{mh}^2 = \frac{1}{n} \sum_{i=1}^{n} (y_i - \overline{y})^2$

er rimdig a erstatte med il.

Jog setter in for you =

$$\hat{\mathcal{M}}_{indo} = \frac{1}{n} \sum_{i=1}^{n} \log x_i \qquad \text{or} \qquad \hat{\mathcal{C}}_{indo}^2 = \frac{1}{n} \sum_{i=1}^{n} \left(\log x_i - \hat{\mathcal{A}}\right)^2$$

Som et det somme som ble utedet i forrige appaave.

e) Fisher-information motivien $I(\mu, \sigma)$ for en discreation er gitt ved:

$$I(\mu, \sigma) = \begin{bmatrix} I_{11}(\mu, \sigma) & I_{12}(\mu, \sigma) \\ I_{21}(\mu, \sigma) & I_{22}(\mu, \sigma) \end{bmatrix}$$

hor $J_{11}(M_{10}) = -E(\frac{\partial^{2}}{\partial h}\log f(X; \mu, \sigma),$

$$I_{1}(\mu,\sigma) = I_{21}(\mu,\sigma) = -E\left(\frac{\partial^{2}}{\partial\mu\partial\sigma}\log f(X;\mu,\sigma)\right)$$
og
$$I_{22}(\mu,\sigma) = -E\left(\frac{\partial^{2}}{\partial\sigma^{2}}\log f(X;\mu,\sigma)\right)$$

Funco elementere:

$$I_{11}(\Lambda_{10}) = -E\left(\frac{\partial^{2}}{\partial \chi_{1}} \log f(\chi; \mu, \sigma)\right)$$

$$= -E\left(\frac{\partial^{2}}{\partial \chi_{2}} \log \frac{1}{\log \chi} \cdot e^{\int \chi_{1} (\chi - \mu)^{2}/2\sigma^{2}}\right)$$

$$= -E\left(\frac{\partial^{2}}{\partial \chi_{2}} - \log \chi_{2} - \log \chi + (-\log \chi - \mu)^{2}/2\sigma^{2}}\right)$$

$$= -E\left(\frac{\partial}{\partial \chi_{1}} - \frac{2}{2\sigma^{2}} \cdot \left[\log \chi - \lambda_{1}\right] \cdot (-1)\right)$$

$$= -E\left(\frac{\partial}{\partial \chi_{1}} \log \frac{1}{\sigma^{2}} - \frac{\lambda_{1}}{\sigma^{2}}\right) = -E\left(-\frac{1}{\sigma^{2}}\right) = \frac{1}{\sigma^{2}}$$

$$I_{12}(\Lambda_{10}) = -E\left(\frac{\partial^{2}}{\partial \mu_{30}} \log \frac{1}{f(\chi; \mu, \sigma)}\right)$$

$$= -E\left(\frac{\partial^{2}}{\partial \mu_{30}} \log \frac{1}{f(\chi; \mu, \sigma)} \cdot \left[\log \chi - \mu\right]^{2}/2\sigma^{2}\right)$$

$$= -E\left(\frac{\partial^{2}}{\partial \mu_{30}} - \log \chi_{1} - \log \chi - \mu\right] \cdot (-1)$$

$$= -E\left(\frac{\partial}{\partial \sigma} - \frac{2}{2\sigma^{2}} \cdot \left[\log \chi - \lambda_{1}\right] \cdot (-1)\right)$$

$$= -E\left(\frac{\partial}{\partial \sigma} \log \chi - \frac{\lambda_{1}}{\sigma^{2}}\right) = -E\left(-\frac{2 \ln \chi}{\sigma^{3}} + \frac{2 \mu}{\sigma^{3}}\right)$$

$$= \frac{2}{\sigma^{3}}\left[E\left(\ln \chi\right) - E\left(\lambda\right)\right] = 0$$

$$I_{11}(\Lambda_{10}) = I_{12}(\Lambda_{10}) = 0$$

$$\begin{aligned}
T_{12}(\Lambda_{10}) &= -E\left(\frac{\partial^{2}}{\partial^{2}\sigma^{2}}\log f(X;\Lambda_{10})\right) \\
&= -E\left(\frac{\partial^{2}}{\partial^{2}\sigma^{2}}\log \frac{1}{\Re x_{0}X} \cdot e^{\left[\ln X - \mu\right]^{2}/2\sigma^{2}}\right) \\
&= -E\left(\frac{\partial^{2}}{\partial^{2}\sigma^{2}} - \log \frac{1}{2\pi}X - \log \sigma + \left(-\left[\ln X - \mu\right]^{2}/2\sigma^{2}\right)\right) \\
&= -E\left(\frac{\partial}{\partial \sigma} - \frac{1}{\sigma} + \frac{2}{2\sigma^{3}}\left[\ln X - \mu\right]^{2}\right) \\
&= -E\left(\frac{1}{\sigma^{2}} - \frac{3}{\sigma^{4}} \cdot \left[\ln X - \mu\right]^{2}\right) \\
&= -\frac{1}{\sigma^{2}} + \frac{3}{\sigma^{4}} \cdot E\left(\left[\ln X - \mu\right]^{2}\right) = \frac{1}{\sigma^{2}}
\end{aligned}$$
Dethe sor the withing it...

Defe gir Fisher-informarjonsmattison:

$$I(\mu, \sigma) = \begin{bmatrix} I_{\eta}(\mu, \sigma) & I_{12}(\mu, \sigma) \\ I_{21}(\mu, \sigma) & I_{22}(\mu, \sigma) \end{bmatrix} = \begin{bmatrix} 1/2 & 0 \\ 0 & 2/\sigma \end{bmatrix}$$

No! Jeg fortsettes oppgaven med matrison fra oppgaveteksten, ikke den jeg lant. Nor autall observasjoner er til trekelig stor, er da

$$\hat{\mu}_{\text{note.}}$$
 Norm $\left(\mu_{\text{note.}}, \frac{1}{n} \cdot I^{1}(\mu, \sigma)\right)$ og $\hat{\tau}_{\text{note.}}$ Norm $\left(\sigma_{\text{note.}}, \frac{1}{n} \cdot I^{2}(\mu, \sigma)\right)$

hor I'' og I'' er diagonalelementons i den inverterte Fisher-informasjonsmakrisen. Jeg inverterer $T(u,\sigma)$, altså den appositt i appgaveteinsten:

$$I(\mu, \sigma)^{-1} = \frac{1}{\det(I(\mu, \sigma))} \cdot \begin{bmatrix} I_{22}(\mu, \sigma) & 0 \\ 0 & I_{11}(\mu, \sigma) \end{bmatrix}$$

$$T(\mu,\sigma)^{-1} = \frac{1}{\pi_{1}(\mu,\sigma) \cdot \pi_{22}(\mu,\sigma)} \cdot \begin{bmatrix} T_{22}(\mu,\sigma) & 0 \\ 0 & T_{11}(\mu,\sigma) \end{bmatrix}$$

$$= \frac{2}{\sigma \cdot 2\sigma^{4}} \begin{bmatrix} 1/2\sigma^{4} & 0 \\ 0 & 1/\sigma^{2} \end{bmatrix}$$

$$= \begin{bmatrix} \sigma^{2} & 0 \\ 0 & 2\sigma^{4} \end{bmatrix}$$

Siden $\hat{\mu}_{\text{mle.}}$ horm $(\hat{\mu}_{\text{mle.}}, \frac{1}{n} \cdot \mathbf{I}^{\text{o}}(\mu, \sigma))$ finner jeg stændardfeilen til $\hat{\mu}_{\text{mle}}$ ved: $\frac{1}{n} \cdot \mathbf{I}^{\text{o}}(\mu, \sigma) = \frac{2}{n}$

Og norten tilsvarende (jeg er litt mer usikker hex) siden ône. in Norm (om, in I'm (u, o)), finner jeg skundard feilen til ônde ved:

$$\left(\frac{1}{n} \cdot I^{22}(\lambda_1 \sigma)\right)^2 = \left(\frac{1}{2} \sigma^4/n\right)^2$$

Veldig unleer på om det steal Oppheryer i 2. Om vi hav junner informasjonen til 8 eller 82...

T (") /

Her bruker jeg at sandard feilar er variannen:

$$\begin{aligned} \text{MSE}(\hat{\mu}_{\text{min}}) &= \text{E}[\hat{\mu}_{\text{max}} - \mu)^{7} \\ &= \text{E}[\hat{\mu}_{\text{min}}^{2}] - 2\mu \text{E}[\hat{\mu}_{\text{max}}] + \mu^{2} \\ &= \text{E}[\hat{\mu}_{\text{min}}^{2}] - 2\mu^{2} + \mu^{2} \\ &= \text{E}[\hat{\mu}_{\text{min}}^{2}] - \mu^{2} = \text{Var}[\hat{\mu}_{\text{min}}] \end{aligned}$$

f) se vedlegg!

g) TODO!

a) Bruker det au forventningevordi for lovificuariog tilfolding variablel:

$$E(X_i) \simeq \int_0^{\theta} \chi f(\chi_i, \theta) d\chi = \int_0^{\theta} \chi_i \frac{1}{\theta} d\theta = \frac{t}{\theta} \frac{1}{2} \chi^2 \Big|_0^{\theta}$$
$$= \frac{1}{\theta} \Big(\frac{1}{2} \theta^2 - \frac{1}{2} \cdot 0^2 \Big) = \frac{\theta}{2}$$

Bruker définisjon au voirant:

$$Var[X_i] = \int_0^{\theta} (x - E[X_i])^2 f(x; \theta) dx = \int_0^{\theta} (x - \frac{\theta}{2})^2 \frac{1}{\theta} dx$$

$$= \int_0^{\theta} (x^2 - 2x \cdot \frac{\theta}{2} + \frac{\theta^2}{4}) \frac{1}{\theta} dy = \int_0^{\theta} (\frac{x^2}{\theta} - x + \frac{\theta}{4}) dx$$

$$= \frac{x^3}{3\theta} - \frac{x^2}{2} - \frac{\theta}{4} = \frac{\theta^2}{3\theta} - \frac{\theta^2}{2} - \frac{\theta^2}{4}$$

$$= \frac{4\theta^2 - 3\theta^2}{12} = \frac{\theta^2}{12}$$

(b) Momanmesoden gir at

$$\begin{aligned}
\mathbb{E}[X_i] &= \frac{1}{N} \sum_{i=1}^{N} \chi_i = \overline{\chi} \\
&= \overline{\chi} \\
&= \overline{\chi}
\end{aligned}$$
fra appave a)
$$\frac{\Phi}{2} = \overline{\chi} \iff \Phi_{\text{won}} = 2\overline{\chi}$$

For at estimatoren shal were forventuingment ma por det.: $E(\hat{\theta}_{non}) \stackrel{?}{=} \theta$.

$$E[\hat{\theta}_{non}] = E[\lambda \bar{X}] = \lambda E[\bar{X}] = \lambda E[\frac{1}{n} \sum_{i=1}^{n} x_i]$$

$$= \frac{\lambda}{n} \sum_{i=1}^{n} E[x_i] = \frac{\lambda}{n} \cdot [n \cdot \frac{\delta}{\lambda}] = \delta$$

2 Duan er altså forvontningsrett.

C)Standard fel ex gitt ved JE[(ômm - 0)2];

$$E[(\hat{\theta}_{\text{man}} - \theta)^{2}] = E[(2\hat{x} - \theta)^{2}] = E[4\hat{x}^{2} - 2\hat{x}\theta + \theta^{2}]$$

$$= 4E[\hat{x}^{2}] - 2\theta E[\hat{x}] + \theta^{2}$$

$$= 4E[(\hat{x}_{n} + \hat{x}_{n})^{2}] - 2\theta \frac{\theta}{2} + \theta^{2}$$

$$= \frac{4}{n^{2}} E[(\hat{x}_{n} + \hat{x}_{n})^{2}] \times \frac{4}{n^{2}} E[(n + \hat{x}_{n})^{2}]$$
Her or rog wither row on delte e

 $= 4 \pm [\chi_i^2] = \frac{40^2}{3}$ Shandardfeil: $\frac{2\theta}{\sqrt{3}}$ ext. $2\sqrt{E[x^2]}$ $= \frac{\theta^2}{12} + \frac{\theta}{4} = \frac{\theta^2}{3}$

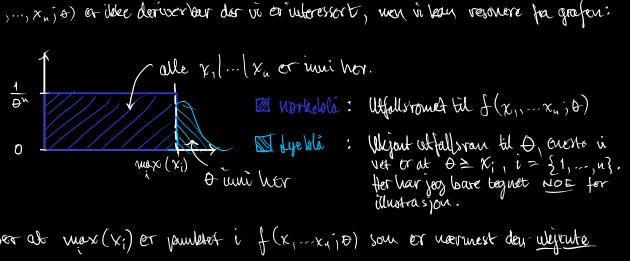
Likelihood funderjonen er gitt ved sanns ynligheten for at vi deserverer alle x-ene, alseá: d)

$$P(x_{i}|...|x_{i}) = f(x_{i};\theta) \cdot ... \cdot f(x_{i};\theta) = \prod_{i=1}^{n} f(x_{i};\theta)$$

$$= \begin{cases} \prod_{i=1}^{n} \frac{1}{\theta}, \text{ for } 0 \leq x \leq \theta \\ \prod_{i=1}^{n} 0, \text{ ellors} \end{cases}$$

$$= \begin{cases} \lambda_{i} \text{ of } 0 \leq x \leq \theta \\ 0, \text{ ellors} \end{cases}$$

f(x,,...,x,jo) or like deriver bur der vi er interessort, men in bom resonere for grafen:



Vi ser at max(x;) et pumblet i f(x,...x,j,0) som et nærmest den wijente topppunktet til punktsansynlighets funksjonen til O.

e) this of or forventual gretts har is at ELDI = 0.

$$\sharp(\hat{\theta}) = \sharp \left[\frac{n+1}{m} \widehat{\theta}_{\text{vulo}}\right] = \frac{n+1}{n} \sharp [\widehat{\theta}_{\text{vul}}] = \frac{n+1}{n} \left[\frac{n \theta}{n+1}\right] = \theta$$

Firmor standard feilen:

$$\mathbb{E}\left[\left(\tilde{\Theta} - \Theta\right)^{2}\right] = \mathbb{E}\left[\tilde{\Theta}^{2} - 2\tilde{\Theta} + \Theta^{2}\right] = \mathbb{E}\left[\frac{(n+1)^{2}}{N^{2}}\left(\hat{\Theta}_{wle}\right)^{2}\right] - 2\tilde{\Theta} + \tilde{\Theta}^{2}$$

$$= \mathbb{E}\left[\left(\frac{n+1}{N}\tilde{\Theta}_{wle}\right)^{2}\right] = \mathbb{E}\left[\frac{(n+1)^{2}}{N^{2}}\left(\hat{\Theta}_{wle}\right)^{2}\right] - 2\tilde{\Theta} + \tilde{\Theta}^{2}$$

$$= \frac{(n+1)^{2}\tilde{\Theta}^{2}}{N^{2}} + \mathbb{E}\left[\hat{\Theta}_{wle}\right] - \tilde{\Theta}^{2} = \frac{(n+1)^{2}\tilde{\Theta}^{2}}{N^{2}} + 2\tilde{\Omega}$$

$$= \frac{(n+1)^{2}\tilde{\Theta}^{2}}{N^{2} + 2\tilde{\Omega}} - \tilde{\Theta}^{2} = \frac{(n^{2} + 2\tilde{\Omega} + 1)\tilde{\Theta}^{2} - (n^{2} + 2\tilde{\Omega})\tilde{\Theta}^{2}}{N^{2} + 2\tilde{\Omega}}$$

$$= \frac{\tilde{\Theta}^{2}}{N^{2} + 2\tilde{\Omega}}$$

Ultra, Handarfeilen er $\sqrt{E[(\theta-\theta)^2]} = \frac{\theta}{\sqrt{n^2+2n}}$

f) standardfellen til
$$\hat{\theta}_{\text{non}} = 2\sqrt{E[X^2]} = \frac{2}{n} \left[E[XX] \right]$$

Standarfellen til $\hat{\theta} = \frac{1}{2} \sqrt{n^2 + 2n}$ synker med n

Bétraleter vi huordan disse endre, med n, ser vi at dan eux longikon dominarer den andre:

$$\lim_{N\to\infty} \frac{1}{\sqrt{N+2N}} \gg \lim_{N\to\infty} \frac{1}{N}$$

Jeg velger derfor , siden standardfeilen blir mindre for mange observasjoner.

g) TODO!

R-kode for debt au app gave 1:

forsikringskrav=read.table("https://www.uio.no/studier/emner/matnat/math/STK1110/data/forsikringskrav.txt")
x=forsikringskrav\$V1

```
# OPPGAVE 1B og 1C
x.mom.forvent = log(mean(x)/sqrt((var(x)/mean(x)^2)+1))
x.mom.varians = log((var(x)/mean(x)^2)+1)
c(x.mom.forvent, x.mom.varians)
x.mle.forvent = sum(log(x))/length(x)
x.mle.varians = sum((log(x)-x.mle.forvent)^2)/length(x)
c(x.mle.forvent, x.mle.varians)
# Eksempelkjøring:
#
#
#
    [1] 2.7386963 0.8902211
#
    [1] 2.7822827 0.7657692
#
# Vi ser at forventingen til MLE er litt større enn ved momentestimatoren, og
# variansen til MLE er litt mindre.
# OPPGAVE 1E
# [TODO!]
# OPPGAVE 1G
# Ikke-parametrisk bootstrap (kode tilpasset fra eksempel i forelsning) for å
# bestemme standardfeilen.
B = 1000
x.mle.forvent.star = x.mle.varians.star = rep(0,B)
for (b in 1:B)
 x.star = sample(x,length(x),replace=T)
 x.mle.forvent.star[b] = sum(log(x.star))/length(x)
  x.mle.varians.star[b] = sum((log(x.star)-x.mle.forvent.star[b])^2)/length(x)
}
c(mean(x.mle.forvent.star), mean(x.mle.varians.star)) # Sitt (saniy check)
c(sd(x.mle.forvent.star), sd(x.mle.varians.star)) # Standardfeil
# Eksempelkjøring:
#
#
#
#
    [1] 2.7819802 0.7656257
    [1] 0.01073693 0.01490560
```