

Ekse.  $X$  er  $\left\{ \begin{array}{l} \text{binomisk} \\ \text{poisson} \\ \text{normal} \\ \text{eksponential} \end{array} \right.$  Kva med  $\sum X_i$

Def. 7.2

Momentgenererende funksjon er definert ved

$$M_X(t) = E[e^{tx}] = \begin{cases} \sum_x e^{tx} f(x), & X \text{ diskret} \\ \int_{-\infty}^{\infty} e^{tx} f(x) dx, & X \text{ kontinuert} \end{cases}$$

$$\text{Vi har } \frac{d^n M_X(t)}{dt^n} = \begin{cases} \sum_x x^n e^{tx} f(x), & X \text{ diskret} \\ \int_{-\infty}^{\infty} x^n e^{tx} f(x) dx, & X \text{ kontinuert} \end{cases}$$

$$\text{slik at } E[X^n] = \left. \frac{d^n M_X(t)}{dt^n} \right|_{t=0} = \mu_n'$$

Eksempel

$X \sim$  Bernoulli fordelt  $\therefore$ 

$x$	$0$	$1$
$P(X=x)$	$1-p$	$p$

$$M_X(t) = E[e^{tx}] = e^{t \cdot 0} (1-p) + e^t p = 1-p + pe^t$$

$$\frac{dM_X(t)}{dt} = pe^t \Rightarrow E[X] = \left. \frac{dM_X(t)}{dt} \right|_{t=0} = p$$

$$E[X^2] = \left. \frac{d^2 M_X(t)}{dt^2} \right|_{t=0} = pe^t \Big|_{t=0} = p \Rightarrow \text{Var}[X] = p - p^2 = p(1-p)$$

$X \sim$  geometrisk fordelt med sannsyn  $p$   $\therefore P(X=x) = (1-p)^{x-1} p, x=1,2,3$

$$M_X(t) = \sum_{x=1}^{\infty} e^{tx} (1-p)^{x-1} p = pe^t \sum_{x=1}^{\infty} e^{t(x-1)} (1-p)^{x-1}$$

$$= p e^t \sum_{x=1}^{\infty} (e^t(1-p))^{x-1} = \frac{p e^t}{1 - e^t(1-p)}, \quad e^t(1-p) < 1 \Leftrightarrow t < -\ln(1-p)$$

$$M_x'(t) = \frac{p e^t}{(1 - e^t(1-p))^2} \Rightarrow M_x'(0) = \frac{1}{p} = E[X]$$

$$M_x''(t) = \frac{p e^t (1 + e^t(1-p))}{(1 - e^t(1-p))^3} \Rightarrow M_x''(0) = \frac{2-p}{p^2}$$

$$\Rightarrow \text{Var}[X] = \frac{2-p}{p^2} - \frac{1}{p^2} = \frac{1-p}{p^2}$$

$X \sim$  exponentialfordelt med parameter  $\frac{1}{\beta}$

$$f_X(x) = \begin{cases} \frac{1}{\beta} e^{-\frac{x}{\beta}}, & x > 0 \\ 0, & \text{elles} \end{cases}$$

$$M_x(t) = \int_0^{\infty} e^{tx} \frac{1}{\beta} e^{-\frac{x}{\beta}} dx = \frac{1}{\beta} \int_0^{\infty} e^{(t - \frac{1}{\beta})x} dx = \frac{1}{\beta} \left[ \frac{1}{t - \frac{1}{\beta}} e^{(t - \frac{1}{\beta})x} \right]_0^{\infty}$$

$$= \frac{1}{1 - \beta t}, \quad t < \frac{1}{\beta}$$

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

$$\Rightarrow M_x(t) = \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{\sigma} \int_{-\infty}^{\infty} e^{tx} e^{-\frac{1}{2} \left( \frac{x-\mu}{\sigma} \right)^2} dx$$

$$\text{Vi har: } tx - \frac{(x-\mu)^2}{2\sigma^2} = -\frac{(x^2 - 2x(\mu + \sigma^2 t) + \mu^2)}{2\sigma^2}$$

$$= -\frac{((x - (\mu + \sigma^2 t))^2 - \mu^2 - 2\mu\sigma^2 t + \sigma^4 t^2 + \mu^2)}{2\sigma^2}$$

$$= -\frac{(x - (\mu + \sigma^2 t))^2}{2\sigma^2} + \mu t + \frac{\sigma^2 t^2}{2}$$

Dette giver  $M_X(t) = e^{\mu t + \frac{\sigma^2 t^2}{2}}$

$$= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{-\frac{(x-(\mu+\sigma^2 t))^2}{2\sigma^2}} dx$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} dz = e^{\mu t + \frac{\sigma^2 t^2}{2}}$$

Egenskaper til momentgenererende funktioner.

7.8.  $M_{X+a}(t) = E[e^{t(X+a)}] = e^{at} M_X(t)$

7.9.  $M_{aX}(t) = E[e^{t(aX)}] = M_X(at)$

7.10.  $X_1, X_2, \dots, X_m$  uafh. La  $Y = \sum_{i=1}^m X_i$

$$M_Y(t) = E[e^{t \sum_{i=1}^m X_i}] = \int \dots \int e^{t \sum_{i=1}^m x_i} f(x_1, x_2, \dots, x_m) dx_1 \dots dx_m$$

$$= \int e^{t x_1} f_{X_1}(x_1) dx_1 \dots \int e^{t x_m} f_{X_m}(x_m) dx_m = M_{X_1}(t) \dots M_{X_m}(t)$$

Teorem 7.7

Dersom to stokastiske variable  $X$  og  $Y$  har samme momentgenererende funktion, har dei og samme fordeling.

Eks.  $X_1 \sim N(\mu_1, \sigma_1^2)$ ,  $X_2 \sim N(\mu_2, \sigma_2^2)$  og uafh.

La  $Y = a_1 X_1 + a_2 X_2$

$$M_Y(t) = M_{X_1}(a_1 t) \cdot M_{X_2}(a_2 t) = e^{\mu_1 a_1 t + \frac{\sigma_1^2 a_1^2 t^2}{2}} \cdot e^{\mu_2 a_2 t + \frac{\sigma_2^2 a_2^2 t^2}{2}}$$

$$= e^{(a_1 \mu_1 + a_2 \mu_2) t + \frac{(a_1^2 \sigma_1^2 + a_2^2 \sigma_2^2) t^2}{2}}$$

$\therefore Y \sim N(a_1 \mu_1 + a_2 \mu_2, a_1^2 \sigma_1^2 + a_2^2 \sigma_2^2)$

Generelt 7.11:  $X_i \sim N(\mu_i, \sigma_i^2)$  og uafh  $\Rightarrow \sum_{i=1}^n a_i X_i \sim N(\sum_{i=1}^n a_i \mu_i, \sum_{i=1}^n a_i^2 \sigma_i^2)$

## Resultat

La  $X \sim B(m, p)$ ,  $X = \sum_{j=1}^m I_j$  der  $I_j$  er Bernoulli fordelt.

$$\Rightarrow M_X(t) = (1 - p + pe^t)^m$$

La  $X_1 \sim B(m_1, p)$  og  $X_2 \sim B(m_2, p)$  og uafh.

Da er  $Y = X_1 + X_2 \sim B(m_1 + m_2, p)$

Bevís  $M_Y(t) = M_{X_1}(t) \cdot M_{X_2}(t) = (1 - p + pe^t)^{m_1 + m_2}$

Resultat.  $X_1 \sim P(\lambda_1 \delta_1)$ ,  $X_2 \sim P(\lambda_2 \delta_2)$  og uafh.

Da er  $Y = X_1 + X_2 \sim P(\lambda_1 \delta_1 + \lambda_2 \delta_2)$

Bevís: La  $X \sim P(\lambda \delta)$

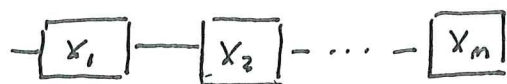
$$\Rightarrow M_X(t) = \sum_{x=0}^{\infty} e^{tx} \frac{(\lambda \delta)^x e^{-\lambda \delta}}{x!} = e^{-\lambda \delta} \sum_{x=0}^{\infty} \frac{(e^{\delta} \lambda \delta)^x}{x!} = \frac{e^{-\lambda \delta} e^{\lambda \delta} e^{\lambda \delta (e^{\delta} - 1)}}{e} = \frac{\lambda \delta (e^{\delta} - 1)}{e}$$

$$M_Y(t) = M_{X_1}(t) \cdot M_{X_2}(t) = \frac{\lambda_1 \delta_1 (e^{\delta_1} - 1)}{e} \cdot \frac{\lambda_2 \delta_2 (e^{\delta_2} - 1)}{e} = \frac{(\lambda_1 \delta_1 + \lambda_2 \delta_2)(e^{\delta} - 1)}{e}$$

## Ordningvariable

La oss sjå på eit serie-system sammansett av komponentar med uavhengige og identisk fordelte levetider.

$$X_1, \dots, X_m. \quad P(X_i \leq x) = F_X(x), \quad i = 1, 2, \dots, m$$



eks. juletrelys

La  $U$  vere levetida til systemet.  $U = \min(X_1, X_2, \dots, X_m)$

$$\begin{aligned} P(U \leq u) &= 1 - P(U > u) = 1 - P(X_1 > u, X_2 > u, \dots, X_m > u) \\ &= 1 - P(X_1 > u) \cdot P(X_2 > u) \dots P(X_m > u) = 1 - (1 - F_X(u))^m \end{aligned}$$

$$f_U(u) = m(1 - F_X(u))^{m-1} f_X(u)$$

eks.

$$F_X(x) = \begin{cases} 1 - e^{-\lambda x}, & x > 0, \\ 0, & \text{elles} \end{cases} \quad i = 1, 2, \dots, m$$

$\therefore$  eksponentialfordelt med parameter  $\beta = \frac{1}{\lambda}$

La  $U = \min(X_1, X_2, \dots, X_m)$

$$\Rightarrow F_U(u) = \begin{cases} 1 - (1 - 1 + e^{-\lambda u})^m, & u > 0 \\ 0, & u \leq 0 \end{cases} = \begin{cases} 1 - e^{-m\lambda u}, & u > 0 \\ 0, & u \leq 0 \end{cases}$$

$\therefore$  eksponentialfordelt med parameter  $\beta = \frac{1}{m\lambda}$

$$\text{og } E[U] = \frac{1}{m\lambda} = \frac{E[X_i]}{m}$$