

# Lösungen zur 9. Übung Stochastik für Informatiker

## Aufg 35

$$\begin{aligned} a) E[ax + bY] &= \sum_{x,y} (ax + by) P(X=x, Y=y) \\ &= \sum_{x,y} ax P(X=x, Y=y) + \sum_{x,y} by P(X=x, Y=y) \\ &= a \sum_x x P(X=x) + b \sum_y y P(Y=y) \\ &= a E[X] + b E[Y] \end{aligned}$$

$$b) X \leq Y \Rightarrow Y - X \geq 0$$

$$\Rightarrow E[Y - X] = \left\{ \begin{array}{l} \sum_{x,y} (y-x) P(X=x, Y=y) \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (y-x) f(x,y) dy dx \end{array} \right\} \geq 0$$

$$\stackrel{\text{Lin.}}{\Rightarrow} E[Y] - E[X] \geq 0 \Rightarrow E[X] \leq E[Y]$$

## Aufg 34 b)

$$\text{Setze } T: \begin{cases} (0, \infty)^2 \rightarrow (0, \infty)^2 \\ (x_1, x_2) \mapsto \left(\frac{x_1}{x_2}, x_2\right) \end{cases}$$

$$\Rightarrow T^{-1}(a, b) = (ab, b)$$

wie a)  $\Rightarrow$  T ist bijektiv auf  $(0, \infty)^2$

$$\det \left( \frac{\partial T_i}{\partial x_j} \right) = \begin{pmatrix} \frac{1}{x_2} & -\frac{x_1}{x_2^2} \\ 0 & 1 \end{pmatrix} = \frac{1}{x_2} \neq 0$$

$\Rightarrow \left(\frac{x_1}{x_2}, x_2\right)$  abs. stetig mit Dichte

$$f_{\left(\frac{x_1}{x_2}, x_2\right)}(z, t) = t \cdot f_{(x_1, x_2)}(zt, t) \cdot \mathbb{1}_{(0, \infty)^2}(z, t)$$

$$\stackrel{\text{st.u.}}{\Rightarrow} f_{\frac{x_1}{x_2}}(z) = \int_0^{\infty} t f_{x_1}(zt) f_{x_2}(t) dt \cdot \mathbb{1}_{(0, \infty)}(z)$$

## Aufg 36

$$a) \bar{E}[X] = \sum_{i=0}^{\infty} i \frac{\lambda^i}{i!} e^{-\lambda} = e^{-\lambda} \sum_{i=1}^{\infty} \frac{\lambda^i}{(i-1)!} = \lambda e^{-\lambda} \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} \\ = \lambda e^{-\lambda} e^{\lambda} = \lambda$$

$$\text{Var}[X] = E[X^2] - (E[X])^2$$

$$E[X^2] = \sum_{i=0}^{\infty} i^2 \frac{\lambda^i}{i!} e^{-\lambda} = \lambda e^{-\lambda} \sum_{i=1}^{\infty} \frac{i \lambda^{i-1}}{(i-1)!} = \lambda e^{-\lambda} \sum_{i=1}^{\infty} \frac{d}{d\lambda} \lambda^i \\ = \lambda e^{-\lambda} \frac{d}{d\lambda} \sum_{i=1}^{\infty} \frac{\lambda^i}{(i-1)!} = \lambda e^{-\lambda} \frac{d}{d\lambda} \lambda e^{\lambda} = \lambda e^{-\lambda} (e^{\lambda} + \lambda e^{\lambda}) \\ = \lambda(1+\lambda) = \lambda + \lambda^2$$

$$\text{Var}[X] = \lambda + \lambda^2 - \lambda^2 = \lambda$$

$$b) \text{Bin}(n, p) = \underbrace{\text{Geo}(p) * \dots * \text{Geo}(p)}_n$$

$$Y_i \sim \text{Geo}(p) \quad i=1, \dots, n \quad \text{st. u.}$$

$$X = \sum_{i=1}^n Y_i$$

$$E[X] = \sum_{i=1}^n E[Y_i] = n E[Y_1] = \frac{n(1-p)}{p}$$

$$E[Y_1] = E[Y_1(Y_1-1) + Y_1] = E[Y_1(Y_1-1)] + E[Y_1]$$

$$= \sum_{k=0}^{\infty} k(k-1) q^k p + \frac{q}{p}$$

$$= p q^2 \sum_{k=2}^{\infty} k(k-1) q^{k-2} + \frac{q}{p}$$

$$= p q^2 \sum_{k=2}^{\infty} \frac{\partial^2}{\partial q^2} q^k + \frac{q}{p}$$

$$= p q^2 \frac{\partial^2}{\partial q^2} \sum_{k=2}^{\infty} q^k + \frac{q}{p}$$

$$= p q^2 \frac{\partial^2}{\partial q^2} \frac{q}{1-q} + \frac{q}{p}$$

$$= pq^2 \frac{d}{dq} \frac{2q(1-q) + q^2}{(1-q)^2} + \frac{q}{p}$$

$$= pq^2 \frac{d}{dq} \frac{2q - q^2}{(1-q)^2} + \frac{q}{p}$$

$$= pq^2 \frac{(2-2q)(1-q)^2 + 2(1-q)(2q-q^2)}{(1-q)^4} + \frac{q}{p}$$

$$= pq^2 \frac{(2-2q)(1-q) + 2(2q-q^2)}{(1-q)^3} + \frac{q}{p}$$

$$= pq^2 \frac{2}{p^3} + \frac{q}{p} = \frac{2q^2}{p^2} + \frac{q}{p}$$

$$\text{Var}[Y_1] = E[Y_1^2] - (E[Y_1])^2 = \frac{2q^2 + pq - q^2}{p^2} = \frac{q(q+p)}{p^2} = \frac{1-p}{p^2}$$

$$\Rightarrow \text{Var}[X] = \sum_{i=1}^n \text{Var}[Y_i] = n \text{Var}[Y_1] = \frac{n(1-p)}{p^2}$$

$$c) E[X] = \int_0^{\infty} x \frac{R^k}{\Gamma(k)} x^{k-1} e^{-Rx} dx$$

$$= \frac{R^k}{\Gamma(k)} \int_0^{\infty} x^k e^{-Rx} dx$$

$$= \frac{R^k}{\Gamma(k)} \frac{\Gamma(k+1)}{R^{k+1}} \underbrace{\int_0^{\infty} \frac{R^{k+1}}{\Gamma(k+1)} x^k e^{-Rx} dx}_{=1}$$

$$= \frac{k}{R}$$

$$E[X^2] = \int_0^{\infty} x^2 \frac{R^k}{\Gamma(k)} x^{k-1} e^{-Rx} dx$$

$$= \frac{R^k}{\Gamma(k)} \frac{\Gamma(k+2)}{R^{k+2}} \int_0^{\infty} \frac{R^{k+2}}{\Gamma(k+2)} x^{k+1} e^{-Rx} dx = \frac{(k+1)k}{R^2}$$

$$\Rightarrow \text{Var}[X] = \frac{(k+1)k}{R^2} - \frac{k^2}{R^2} = \frac{k}{R^2}$$

$$d) E(X^+) = \int_0^{\infty} \frac{x}{\pi(1+x^2)} dx = \lim_{t \rightarrow \infty} \int_0^t \frac{x}{\pi(1+x^2)} dx = \lim_{t \rightarrow \infty} \frac{1}{2\pi} \ln(1+t^2) = \infty \Rightarrow E[X] \neq$$

$$\Rightarrow \text{Var}[X] \neq$$

# Zusatzaufgabe zum 9. Kapitel

Seien  $X_1, \dots, X_n$  iid  $\sim \text{Poi}(\lambda)$   
Bestimmen Sie für eine Stichprobe  $x_1, \dots, x_n$  den  
Maximum-Likelihood-Schätzer  $\lambda^*$  von  $\lambda$ .

Lösung:  $L(\lambda | x_1, \dots, x_n) = \prod_{i=1}^n e^{-\lambda} \frac{\lambda^{x_i}}{x_i!} = e^{-n\lambda} \frac{\lambda^{\sum_{i=1}^n x_i}}{\prod_{i=1}^n x_i!}$

$$\log L(\lambda | x_1, \dots, x_n) = -n\lambda + \left(\sum_{i=1}^n x_i\right) \log \lambda - \underbrace{\sum_{i=1}^n \log x_i!}_{\text{unabh. von } \lambda}$$

$\Rightarrow$  maximiere  $g(\lambda) = -n\lambda + \left(\sum_{i=1}^n x_i\right) \log \lambda$

notwendiges Kriterium für ein Maximum:

$$g'(\lambda) = -n + \frac{\sum_{i=1}^n x_i}{\lambda} \stackrel{!}{=} 0$$

$\Rightarrow \lambda^* = \frac{\sum_{i=1}^n x_i}{n} = \bar{x}$

hinreichendes Kriterium:

$$g''(\lambda) = \sum_{i=1}^n x_i \left(-\frac{1}{\lambda^2}\right) < 0$$

gilt, falls  $\sum_{i=1}^n x_i > 0$

Falls  $\sum_{i=1}^n x_i = 0$ :  $\log L(\lambda | x_1=0, \dots, x_n=0) = -n\lambda$

$\Rightarrow$  Maximum bei  $\lambda^* = 0$  ( $= \frac{\sum_{i=1}^n x_i}{n}$ , da  $\sum_{i=1}^n x_i = 0$ )

$\Rightarrow$  Für alle  $x_1, \dots, x_n$  ist  $\lambda^* = \frac{\sum_{i=1}^n x_i}{n}$  ML-Schätzer  
für  $\lambda$ .

$\sum_{i=1}^n x_i = 0, x_i \geq 0$   
 $\Rightarrow x_i = 0 \forall i$