

Ex 2.2

a) $X \sim \delta_0 \Leftrightarrow \varphi_X(t) = 1 \quad \forall t \in \mathbb{R}$

$$\varphi_X(t) \stackrel{\text{def}}{=} \mathbb{E}(\exp(itx)) = \sum_{\substack{\text{discrete} \\ k \in \mathbb{Z}}} e^{itk} f_X(k)$$

$x \mapsto \{0, 1\}$

$$f_X(x) = \begin{cases} 1 & \text{if } x=0 \\ 0 & \text{else} \end{cases} \Rightarrow \mathbb{E}(\exp(itx)) = \exp(it \cdot 0) \cdot 1 + 0 = 1$$

b) $U(0, \frac{1}{n}) \xrightarrow{d} \delta_0$ Note "weakly converges" \Leftrightarrow "converges in distribution"

using the hint:

$$\varphi_{U(0, \frac{1}{n})}(t) = e^{-\frac{1}{n} t} \xrightarrow[n \rightarrow \infty]{} 1 = \varphi_{\delta_0}(t) \quad \forall t \in \mathbb{R}$$

c) $X_n \sim N(0, n) \quad X_n \xrightarrow{d} X?$

using hint

$$\varphi_{X_n}(t) = e^{-n \frac{t^2}{2}} \xrightarrow[n \rightarrow \infty]{} 0 \quad \forall t \neq 0$$

if $t=0 \quad \varphi_{N(0, n)}(0) = 1$

d) $X \sim U[a, b] \quad a < b$

$$\varphi_X(t) \stackrel{\text{def}}{=} \mathbb{E}(\exp(itx)) = \int \exp(itx) f_X(x) dx$$

$$= \int_a^b \exp(itx) \frac{1}{b-a} dx$$

$$= \frac{1}{it} \exp(itx) \frac{1}{b-a} \Big|_a^b$$

$$= \frac{1}{it(b-a)} (\exp(itb) - \exp(ita))$$

e) $U[0, \frac{1}{n}] \xrightarrow{d} \delta_0$

$$\lim_{n \rightarrow \infty} \frac{\exp(it \frac{1}{n}) - \exp(it \cdot 0)}{it (\frac{1}{n} - 0)} = \lim_{n \rightarrow \infty} \frac{-\frac{1}{n^2} it \exp(it \frac{1}{n})}{-it \frac{1}{n^2}} = \lim_{n \rightarrow \infty} \exp(it \frac{1}{n}) = 1 \quad \forall t \in \mathbb{R}$$

f) $X_n \sim U[0, n]$

$$\varphi_{X_n}(t) = \frac{\exp(itn) - \exp(it \cdot 0)}{itn} = \frac{\cos(tn) + i \sin(tn) - 1}{itn}$$

$$= \frac{i}{tn} + \frac{\cos(tn) - 1}{tn} + \frac{\sin(tn)}{tn}$$

$$= i \left(\frac{1}{t} (1 - \cos(tn)) \right) + \frac{\sin(tn)}{tn}$$

$\downarrow 0$

$$= i \left(\underbrace{\frac{1}{tn} (1 - \cos(tn))}_{\text{Imaginary}} \right) + \underbrace{\frac{\sin(tn)}{tn}}_{\text{Real}}$$

$$= 0 \quad \forall t \neq 0$$

$$\text{if } t = 0 \quad \lim_{\substack{n \rightarrow \infty \\ t \rightarrow 0}} \frac{e^{itn} - 1}{itn} = \dots = 1 \quad \text{not cont. around } 0$$

\Rightarrow not a valid

r.v.

Ex 23 $X \sim \text{Bin}(n, p)$

$$\phi_X(t) = E(\exp(itX)) = \sum_{k=0}^{\infty} e^{itk} p^k (1-p)^{n-k} \binom{n}{k}$$

$$= \sum_{k=0}^{\infty} (e^{itp})^k (1-p)^{n-k} \binom{n}{k}$$

$$= \left(e^{itp} + (1-p) \right)^n \quad \forall t \in \mathbb{R}$$

$$E(X) = \frac{\partial \phi_X(t)}{\partial t} \Big|_{t=0} = \frac{1}{i} n (pe^{it} + (1-p))^{n-1} p e^{it} \Big|_{t=0} = np$$

$$E(X^2) = \frac{\partial^2 \phi_X(t)}{\partial t^2} \Big|_{t=0} = \dots = n(n-1)p^2 + np$$

$$\text{Var}(X) = E(X^2) - E(X)^2 = \cancel{n^2 p^2} - np^2 + np - \cancel{n^2 p^2} = np(1-p)$$

b) same wie im Blatt.

Ex 20:

$$\text{s.t.} \quad \lim_{n \rightarrow \infty} P(|X_n - 1| > \varepsilon) = 0 \quad \forall \varepsilon > 0$$

$$\Rightarrow \exists n_0 \geq \varepsilon - 1 : P(X_n - 1 > \varepsilon) = P(X_n = n) \quad \forall n \geq n_0$$

$$\lim_{n \rightarrow \infty} P(|X_n - 1| > \varepsilon) = \lim_{n \rightarrow \infty} \frac{1}{n^\alpha} = 0 \quad \forall \alpha > 0$$

b)

same wie in der Lösung

c)

$$X_n \xrightarrow{d} 1 \Leftrightarrow E(f(X_n)) - E(f(1)) \rightarrow 0 \quad \forall f \in C_b$$

So let $f \in C_b$

$$\Rightarrow E(f(X_n)) = f(1) \left(1 - \frac{1}{n^\alpha}\right) + f(n) \frac{1}{n^\alpha}$$

$$\Rightarrow \lim_{n \rightarrow \infty} E(f(X_n)) - E(f(1)) = f(1) \left(1 - \frac{1}{n^\alpha}\right) + f(n) \frac{1}{n^\alpha} - f(1)$$

$$= \frac{1}{n^\alpha} (f(n) - f(1)) \rightarrow 0$$

$$\Rightarrow \quad n \rightarrow \infty \quad \mathbb{E}(f(X_n)) - \mathbb{E}(f(X)) = f(1)(1 - n^{-\alpha}) + f(n) n^{-\alpha} - f(1)$$

$$= \underbrace{\frac{1}{n^\alpha}}_{\nearrow 0} \left(\underbrace{f(n)}_{\substack{\nearrow 0 \\ f \text{ bounded}}} - f(1) \right) \xrightarrow{n \rightarrow \infty} 0$$

Ex. 2.1

$$\text{s.t. } \hat{\sigma}_n^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x}_n)^2 \xrightarrow{P} \sigma^2$$

$$\hat{\sigma}_n^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \mu) - (\bar{x}_n - \mu)^2 = \dots = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2 - 2\bar{x}_n + 2\mu\bar{x}_n - 2\mu^2 + \bar{x}_n - 2\mu\bar{x}_n + \mu^2$$

$$= \underbrace{\frac{1}{n} \sum_{i=1}^n (x_i - \mu)}_{\sigma^2} - \underbrace{(\bar{x}_n - \mu)^2}_{\xrightarrow{P} 0} \quad \text{LLN: } \bar{x}_n \xrightarrow{P} \mu \Leftrightarrow \bar{x}_n - \mu \xrightarrow{P} 0 = \text{const.}$$

$$\Rightarrow \bar{x}_n - \mu \xrightarrow{P} 0 \Rightarrow (\bar{x}_n - \mu)^2 \xrightarrow{P} 0$$

CLT: $y = x^2 - \text{const} \Rightarrow (\bar{x}_n - \mu)^2 \xrightarrow{P} 0$

$$\begin{pmatrix} \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2 \\ (\bar{x}_n - \mu)^2 \end{pmatrix} \xrightarrow{P} \begin{pmatrix} \sigma^2 \\ 0 \end{pmatrix}$$

$$\text{CLT } f\left(\frac{x}{\sqrt{y}}\right) = x - y - \text{const}$$