Markus Reiß
Stochastik II / Stochastic Processes I
Winter 2023/24
Humboldt-Universität zu Berlin


## Exercises: sheet 1

1. Flies and wasps land on your dinner plate in the manner of independent Poisson processes with respective intensities $\mu$ and $\lambda$. Show that the arrival of flying beasts forms a Poisson process of intensity $\lambda+\mu$ (superposition). The probability that an arriving fly is a blow-fly is $p$. Does the arrival of blow-flies also form a Poisson process? (thinning)
2. Let $\left(N_{t}, t \geqslant 0\right)$ be a Poisson process of intensity $\lambda>0$ and let $\left(Y_{k}\right)_{k \geqslant 1}$ be a sequence of i.i.d. random variables, independent of $N$. Then $X_{t}:=\sum_{k=1}^{N_{t}} Y_{k}$, $t \geqslant 0$, is called compound Poisson process $\left(X_{t}:=0\right.$ if $\left.N_{t}=0\right)$.
(a) Show that $\left(X_{t}, t \geqslant 0\right)$ has independent and stationary increments.
(b) Determine the expectation of $X_{t}$ in the case $Y_{k} \in L^{1}$.
(c) Introduce the Lévy measure $\nu(B):=\lambda P\left(Y_{1} \in B\right), B \in \mathfrak{B}_{\mathbb{R}}$. Show that $X_{t}$ has characteristic function

$$
\varphi_{t}(u)=\mathbb{E}\left[e^{i u X_{t}}\right]=\exp \left(t \int_{\mathbb{R}}\left(e^{i u x}-1\right) \nu(d x)\right), \quad u \in \mathbb{R}
$$

(d) Find a sequence of compound Poisson processes $\left(X_{t}^{(n)}, t \geqslant 0\right)$ with Lévy measures $\nu_{n}$ such that $X_{t}^{(n)} \xrightarrow{d} N(0,1)$ as $n \rightarrow \infty$ for some fixed $t>0$. Describe heuristically how the sample paths evolve.
$\left(\mathrm{e}^{*}\right)$ Characterize all sequences $\left(\nu_{n}\right)_{n \geqslant 1}$ with $X_{1}^{(n)} \xrightarrow{d} N(0,1)$ in (d).
3. The number of busses that arrive until time $t$ at a bus stop follows a Poisson process with intensity $\lambda>0$ (in our model). Adam and Berta arrive together at time $t_{0}>0$ at the bus stop and discuss how long they have to wait in the mean for the next bus.
Adam: Since the waiting times are $\operatorname{Exp}(\lambda)$-distributed and the exponential distribution is memoryless, the mean is $\lambda^{-1}$.
Berta: The time between the arrival of two busses is $\operatorname{Exp}(\lambda)$-distributed and has mean $\lambda^{-1}$. Since on average the same time elapses before our arrival and after our arrival, we obtain the mean waiting time $\frac{1}{2} \lambda^{-1}$ (at least assuming that at least one bus had arrived before time $t_{0}$ ).
What is the correct answer to this waiting time paradoxon?
4. Let the processes $\left(X_{t}, t \geqslant 0\right)$ and $\left(Y_{t}, t \geqslant 0\right)$ be versions of each other and each have right-continuous sample paths. Prove that $\left(X_{t}, t \geqslant 0\right)$ and $\left(Y_{t}, t \geqslant 0\right)$ are indistinguishable.

Submit the solutions before the lecture on Thursday, 26 October 2023.

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## Exercises: sheet 2

1. Show for a given Markov chain that the set $M$ of invariant initial distributions $\mu$ is convex. Find examples where (a) $M$ consists of one element only, (b) $M$ has infinitely many elements and (c) $M$ is empty.
2. Let $C([0, \infty))$ be equipped with the topology of uniform convergence on compacts using the metric $d(f, g):=\sum_{k \geqslant 1} 2^{-k}\left(\sup _{t \in[0, k]}|f(t)-g(t)| \wedge 1\right)$. Prove:
(a) $(C([0, \infty)), d)$ is Polish.
(b) The Borel $\sigma$-algebra is the smallest $\sigma$-algebra such that all coordinate projections $\pi_{t}: C([0, \infty)) \rightarrow \mathbb{R}, t \geqslant 0$, are measurable.
(c) For any continuous stochastic process $\left(X_{t}, t \geqslant 0\right)$ on $(\Omega, \mathscr{F}, \mathbb{P})$ the mapping $\bar{X}: \Omega \rightarrow C([0, \infty))$ with $\bar{X}(\omega)_{t}:=X_{t}(\omega)$ is Borel-measurable.
(d) The law of $\bar{X}$ is uniquely determined by the finite-dimensional distributions of $X$.
3. Prove the regularity lemma: Let $\mathbb{P}$ be a probability measure on the Borel $\sigma$ algebra $\mathfrak{B}$ of any metric space. Then

$$
\mathcal{D}:=\left\{B \in \mathfrak{B} \mid \mathbb{P}(B)=\sup _{K \subseteq B \text { compact }} \mathbb{P}(K)=\inf _{O \supseteq B \text { open }} \mathbb{P}(O)\right\}
$$

is closed under set differences and countable unions ( $\mathcal{D}$ is a $\sigma$-ring).
Conclude for a Polish space, using the lecture results, that $\mathcal{D}$ is a $\sigma$-algebra and $\mathcal{D}=\mathfrak{B}$.
4. Abstract construction of discrete-time Markov chains: Let $(S, \mathcal{P}(S))$ be a countable state space and let an initial counting density $\mu^{(0)}$ (i.e. $\mu_{i}^{(0)} \geqslant 0$, $\sum_{i \in S} \mu_{i}^{(0)}=1$ ) as well as transition probabilities $p_{i j}$ (i.e. $p_{i j} \geqslant 0$ and
$\sum_{j \in S} p_{i j}=1$ ) be given. $\left.\sum_{j \in S} p_{i j}=1\right)$ be given.
(a) Show that $(S, \mathcal{P}(S))$ becomes a Polish space when equipped with the discrete metric $d(i, j)=\mathbf{1}(i \neq j), i, j \in S$.
(b) For $A \subseteq S^{n+1}$ define

$$
\mu_{n}(A):=\sum_{i_{0} \in S} \cdots \sum_{i_{n} \in S} \mathbf{1}_{A}\left(i_{0}, \ldots, i_{n}\right) \mu_{i_{0}}^{(0)} p_{i_{0} i_{1}} \cdots p_{i_{n-1} i_{n}} .
$$

Show the one-step consistency condition

$$
\mu_{n+1}\left(\pi_{\{0, \ldots, n+1\} \rightarrow\{0, \ldots, n\}}^{-1}(A)\right)=\mu_{n}(A), \quad A \subseteq S^{n+1}
$$

(c) Conclude that $\mu_{\left\{t_{1}, \ldots, t_{n}\right\}}(B):=\mu_{t_{n}}\left(\pi_{\left\{0, \ldots, t_{n}\right\} \rightarrow\left\{t_{1}, \ldots, t_{n}\right\}}^{-1}(B)\right)$ for $n \geqslant 1,0 \leqslant$ $t_{1} \leqslant \cdots \leqslant t_{n}$ and $B \subseteq S^{n}$ defines a projective family and that a Markov chain $\left(X_{n}, n \geqslant 0\right)$ with $\mathbb{P}\left(X_{0}=j\right)=\mu_{j}^{(0)}, \mathbb{P}\left(X_{n+1}=j \mid X_{n}=i\right)=p_{i j}$ for all $n \geqslant 0$ and $i, j \in S$ with $\mathbb{P}\left(X_{n}=i\right)>0$ exists.

Submit the solutions before the lecture on Thursday, 2 November 2023.

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## Exercises: sheet 3

1. A process $\left(B_{t}, t \geqslant 0\right)$ is called Brownian motion if
(a) $B_{t} \sim N(0, t), t \geqslant 0$, holds;
(b) the increments are stationary and independent: for $0 \leqslant t_{0}<t_{1}<\cdots<t_{m}$ we have

$$
\left(B_{t_{1}}-B_{t_{0}}, \ldots, B_{t_{m}}-B_{t_{m-1}}\right) \sim N\left(0, \operatorname{diag}\left(t_{1}-t_{0}, \ldots, t_{m}-t_{m-1}\right)\right)
$$

(c) $B$ has continuous sample paths.

Prove that a process $\left(B_{t}, t \geqslant 0\right)$ with properties (a) and (b) exists by showing that these properties are satisfied by a Gaussian process with mean zero and covariance function $c(t, s)=t \wedge s, t, s \geqslant 0$. For the existence of such a Gaussian process the representation $c(t, s)=\int_{0}^{\infty} \mathbf{1}_{[0, t]}(u) \mathbf{1}_{[0, s]}(u) d u$ might be useful to derive positive-semidefiniteness.
2. (Proof of $\left.C([0,1]) \notin \mathfrak{B}_{\mathbb{R}}^{\otimes[0,1]}\right)$ We say that $A \subseteq \mathbb{R}^{[0,1]}:=\{f:[0,1] \rightarrow \mathbb{R}\}$ only depends on countably many coordinates if there is a countable index set $T(A) \subseteq[0,1]$ with

$$
\forall f \in A, g \in \mathbb{R}^{[0,1]}:\left.f\right|_{T(A)}=\left.g\right|_{T(A)} \Rightarrow g \in A
$$

Let $\mathcal{A}:=\left\{A \subseteq \mathbb{R}^{[0,1]} \mid A\right.$ only depends on countably many coordinates $\}$.
(a) Show that $\left\{f \in \mathbb{R}^{[0,1]} \mid f\left(t_{0}\right) \in B\right\}$ for any $t_{0} \in[0,1], B \in \mathfrak{B}_{\mathbb{R}}$ lies in $\mathcal{A}$.
(b) Verify that $\mathcal{A}$ is a $\sigma$-algebra and deduce that $\mathfrak{B}_{\mathbb{R}}^{\otimes[0,1]} \subseteq \mathcal{A}$.
(c) Prove $C([0,1])=\left\{f \in \mathbb{R}^{[0,1]} \mid f\right.$ is continuous $\} \notin \mathfrak{B}_{\mathbb{R}}^{\otimes[0,1]}$.
3. Let $(X, Y)$ be a two-dimensional random vector with Lebesgue density $f^{X, Y}$.
(a) For $x \in \mathbb{R}$ with $f^{X}(x)>0$ (recall $\left.f^{X}(x)=\int f^{X, Y}(x, \eta) d \eta\right)$ consider the conditional density

$$
f^{Y \mid X=x}(y):=\frac{f^{X, Y}(x, y)}{f^{X}(x)} .
$$

Which regularity condition on $f^{X, Y}$ ensures for any Borel set $B$

$$
\lim _{h \downarrow 0} \mathbb{P}(Y \in B \mid X \in[x, x+h])=\int_{B} f^{Y \mid X=x}(y) d y \quad ?
$$

(b) Show that for $Y \in L^{2}$ (without any condition on $f^{X, Y}$ ) the function

$$
\varphi_{Y}(x):= \begin{cases}\int y f^{Y \mid X=x}(y) d y, & \text { if } f^{X}(x)>0 \\ 0, & \text { otherwise }\end{cases}
$$

minimizes the $L^{2}$-distance $\mathbb{E}\left[(Y-\varphi(X))^{2}\right]$ over all measurable functions $\varphi$. We write $\mathbb{E}[Y \mid X=x]:=\varphi_{Y}(x)$ and $\mathbb{E}[Y \mid X]:=\varphi_{Y}(X)$.
4. In the situation of problem 3 prove the following properties directly from the definition:
(a) $\mathbb{E}[\mathbb{E}[Y \mid X]]=\mathbb{E}[Y]$;
(b) if $X$ and $Y$ are independent, then $\mathbb{E}[Y \mid X]=\mathbb{E}[Y]$ holds a.s.;
(c) if $Y \geqslant 0$ a.s., then $\mathbb{E}[Y \mid X] \geqslant 0$ a.s.;
(d) for all $\alpha, \beta \in \mathbb{R}, \alpha \neq 0$ we have $\mathbb{E}[\alpha Y+\beta \mid X]=\alpha \mathbb{E}[Y \mid X]+\beta$ a.s.;
(e) if $g: \mathbb{R} \rightarrow \mathbb{R}$ is such that $(x, y) \mapsto(x, y g(x))$ is a diffeomorphism and $Y g(X) \in L^{1}$, then $\mathbb{E}[Y g(X) \mid X]=\mathbb{E}[Y \mid X] g(X)$ a.s.

Submit the solutions before the lecture on Thursday, 9 November 2023.

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## Exercises: sheet 4

1. Let $\Omega=\bigcup_{n \in \mathbb{N}} B_{n}, B_{m} \cap B_{n}=\varnothing$ for $m \neq n$, be a measurable, countable partition for given $(\Omega, \mathscr{F}, \mathbb{P})$ and put $\mathcal{B}:=\sigma\left(B_{n}, n \in \mathbb{N}\right)$. Show:
(a) Any $\mathcal{B}$-measurable random variable $X$ can be written as $X=$ $\sum_{n \in \mathbb{N}} \alpha_{n} \mathbf{1}_{B_{n}}$ with suitable $\alpha_{n} \in \mathbb{R}$. For $Y \in L^{1}$ we have

$$
\mathbb{E}[Y \mid \mathcal{B}]=\sum_{n: \mathbb{P}\left(B_{n}\right)>0}\left(\frac{1}{\mathbb{P}\left(B_{n}\right)} \int_{B_{n}} Y d \mathbb{P}\right) \mathbf{1}_{B_{n}} \quad \mathbb{P} \text {-a.s. }
$$

(b) Specify $\Omega=[0,1)$ with Borel $\sigma$-algebra and $\mathbb{P}=U([0,1))$, the uniform distribution. For $Y(\omega):=\omega, \omega \in[0,1)$, determine

$$
\mathbb{E}[Y \mid \sigma([(k-1) / n, k / n), k=1, \ldots, n)]
$$

For $n=1,3,5,10$ plot the conditional expectations and $Y$ itself as functions on $\Omega$.
2. Let $(X, Y)$ be a two-dimensional $N(\mu, \Sigma)$-random vector.
(a) For which $\alpha \in \mathbb{R}$ are $X$ and $Y-\alpha X$ uncorrelated?
(b) Conclude that $X$ and $Y-(\alpha X+\beta)$ are independent for these values $\alpha$ and for arbitrary $\beta \in \mathbb{R}$ such that $\mathbb{E}[Y \mid X]=\alpha X+\beta$ with suitable $\beta \in \mathbb{R}$.

Remark: In the Gaussian case the conditional expectation is linear!
3. Let $\mathscr{G}$ be a sub- $\sigma$-algebra of $\mathscr{F}$. Prove:
(a) $Y_{n} \in \mathcal{M}^{+}(\Omega, \mathscr{F}) \Rightarrow \mathbb{E}\left[\liminf _{n \rightarrow \infty} Y_{n} \mid \mathscr{G}\right] \leqslant \liminf _{n \rightarrow \infty} \mathbb{E}\left[Y_{n} \mid \mathscr{G}\right]$ a.s. (Fatou's Lemma);
(b) $Y_{n} \in \mathcal{M}(\Omega, \mathscr{F}), Y_{n} \rightarrow Y,\left|Y_{n}\right| \leqslant Z$ with $Z \in L^{1}(\Omega, \mathscr{F}, \mathbb{P})$ implies $\mathbb{E}\left[Y_{n} \mid \mathscr{G}\right] \rightarrow \mathbb{E}[Y \mid \mathscr{G}]$ a.s. as $n \rightarrow \infty$ (dominated convergence).

Hint: Use the monotone convergence theorem for conditional expectations, recalling the arguments for the Lebesgue integral / expectation.
4. For $Y \in L^{2}$ define the conditional variance of $Y$ given $X$ by

$$
\operatorname{Var}(Y \mid X):=\mathbb{E}\left[(Y-\mathbb{E}[Y \mid X])^{2} \mid X\right]
$$

(a) Why is $\operatorname{Var}(Y \mid X)$ well defined?
(b) Show $\operatorname{Var}(Y)=\operatorname{Var}(\mathbb{E}[Y \mid X])+\mathbb{E}[\operatorname{Var}(Y \mid X)]$.
(c) Use (b) to prove for independent random variables $\left(Z_{k}\right)_{k \geqslant 1}$ and $N$ in $L^{2}$ with $\left(Z_{k}\right)$ identically distributed and $N \mathbb{N}$-valued:

$$
\operatorname{Var}\left(\sum_{k=1}^{N} Z_{k}\right)=\mathbb{E}\left[Z_{1}\right]^{2} \operatorname{Var}(N)+\mathbb{E}[N] \operatorname{Var}\left(Z_{1}\right)
$$

What is the variance of the compound Poisson process $\left(X_{t}, t \geqslant 0\right)$ from Exercise 1.2 at time $t$ (assuming $Y_{k} \in L^{2}$ )?

Submit the solutions before the lecture on Thursday, 16 November 2023.

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## Exercises: sheet 5

1. Let $\left(X_{n}\right)_{n \geqslant 0}$ be an $\left(\mathscr{F}_{n}\right)$-adapted family of random variables in $L^{1}$. Show that $\left(X_{n}\right)_{n \geqslant 0}$ is a martingale if and only if for all bounded $\left(\mathscr{F}_{n}\right)$-stopping times $\tau$ the identity $\mathbb{E}\left[X_{\tau}\right]=\mathbb{E}\left[X_{0}\right]$ holds.
Hint for ' if ': Deduce first $\mathbb{E}\left[X_{n}\right]=\mathbb{E}\left[X_{0}\right]$ and consider then $\tau=n \mathbf{1}_{G^{C}}+(n+$ 1) $\boldsymbol{1}_{G}$ for suitable events $G$.
2. Let $\left(\mathcal{F}_{n}^{X}\right)_{n \geqslant 0}$ be the natural filtration of a process $\left(X_{n}\right)_{n \geqslant 0}$ and consider a finite stopping time $\tau$ with respect to $\left(\mathcal{F}_{n}^{X}\right)$.
(a) Prove $\mathcal{F}_{\tau}=\sigma\left(\tau, X_{\tau \wedge n}, n \geqslant 0\right)$.

Hint: for ' $\subseteq$ ' write $A \in \mathcal{F}_{\tau}$ as $A=\bigcup_{n} A \cap\{\tau=n\}$.
$\left(\mathrm{b}^{*}\right)$ Show that even $\mathcal{F}_{\tau}=\sigma\left(X_{\tau \wedge n}, n \geqslant 0\right)$ holds.
3. Let $\left(S_{n}\right)_{n \geqslant 0}$ be the symmetric simple random walk, that is $S_{0}=0, S_{n}=$ $\sum_{i=1}^{n} X_{i}, n \geqslant 1$, with independent $X_{i}$ and $\mathbb{P}\left(X_{i}=+1\right)=\mathbb{P}\left(X_{i}=-1\right)=1 / 2$.
(a) Argue that $\left(\left|S_{n}\right|\right)_{n \geqslant 0}$ is a submartingale with respect to the natural filtration $\left(\mathscr{F}_{n}^{S}\right)_{n \geqslant 0}$ of $\left(S_{n}\right)$ (and then also to the natural filtration $\left(\mathscr{F}_{n}^{|S|}\right)_{n \geqslant 0}$ of $\left.\left(\left|S_{n}\right|\right)\right)$.
(b) Verify that $A_{n}=\sum_{i=0}^{n-1} \mathbf{1}\left(\left|S_{i}\right|=0\right), n \geqslant 1$, yields the compensator of $\left(\left|S_{n}\right|\right)_{n \geqslant 0} . A_{n}$ is called local time of the random walk at zero.
(c) Show $\mathbb{P}\left(S_{2 j}=0\right)=\binom{2 j}{j} 2^{-2 j}$ and conclude

$$
\mathbb{E}\left[\left|S_{n}\right|\right]=\sum_{i=0}^{n-1} \mathbb{P}\left(S_{i}=0\right)=\sum_{j=0}^{\lfloor(n-1) / 2\rfloor}\binom{2 j}{j} 2^{-2 j}
$$

4. Generating function of a random walk's first passage time:

Let $\left(S_{n}\right)_{n \geqslant 0}$ be a simple random walk with $S_{0}=0, S_{n}=\sum_{i=1}^{n} X_{i}, n \geqslant 1$, where the $X_{i}$ are independent and $\mathbb{P}\left(X_{i}=+1\right)=p, \mathbb{P}\left(X_{i}=-1\right)=q=1-p$, $p \in(0,1)$. Prove:
(a) With $M(\lambda)=p e^{\lambda}+q e^{-\lambda}, \lambda \in \mathbb{R}$, the process

$$
Y_{n}^{(\lambda)}:=e^{\lambda S_{n}} M(\lambda)^{-n}, \quad n \geqslant 0,
$$

is a martingale with respect to $\left(\mathscr{F}_{n}^{S}\right)$.
(b) For $M(\lambda) \geqslant 1, a, b \in \mathbb{Z}$ with $a<0<b$ and the stopping time $\tau:=$ $\inf \left\{n \geqslant 0 \mid S_{n} \in\{a, b\}\right\}$ we have

$$
e^{a \lambda} \mathbb{E}\left[M(\lambda)^{-\tau} \mathbf{1}_{\left\{S_{\tau}=a\right\}}\right]+e^{b \lambda} \mathbb{E}\left[M(\lambda)^{-\tau} \mathbf{1}_{\left\{S_{\tau}=b\right\}}\right]=1 .
$$

(c) This implies for all $s \in(0,1]$ (solve $s=M(\lambda)^{-1}$ )

$$
\begin{aligned}
& \mathbb{E}\left[s^{\tau} \mathbf{1}_{\left\{S_{\tau}=a\right\}}\right]=\frac{\nu_{+}(s)^{b}-\nu_{-}(s)^{b}}{\nu_{+}(s)^{b} \nu_{-}(s)^{a}-\nu_{+}(s)^{a} \nu_{-}(s)^{b}}, \\
& \mathbb{E}\left[s^{\tau} \mathbf{1}_{\left\{S_{\tau}=b\right\}}\right]=\frac{\nu_{-}(s)^{a}-\nu_{+}(s)^{a}}{\nu_{+}(s)^{b} \nu_{-}(s)^{a}-\nu_{+}(s)^{a} \nu_{-}(s)^{b}}
\end{aligned}
$$

with $\nu_{ \pm}(s)=\left(1 \pm \sqrt{1-4 p q s^{2}}\right) /(2 p s)$ and continuous extension of the quotient in the case $\nu_{+}(s)=\nu_{-}(s)$.
(d) Now let $a \downarrow-\infty$ and infer that the generating function of the first passage time $\tau_{b}:=\inf \left\{n \geqslant 0 \mid S_{n}=b\right\}$ is given by

$$
\varphi_{\tau_{b}}(s):=\mathbb{E}\left[s^{\tau_{b}} \boldsymbol{1}_{\left\{\tau_{b}<\infty\right\}}\right]=\nu_{+}(s)^{-b}=\left(\frac{1-\sqrt{1-4 p q s^{2}}}{2 q s}\right)^{b}, \quad s \in(0,1] .
$$

In particular, we have $\mathbb{P}\left(\tau_{b}<\infty\right)=\varphi_{\tau_{b}}(1)=\min (1, p / q)^{b}$.

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## Exercises: sheet 6

1. Prove that a family $\left(X_{i}\right)_{i \in I}$ of real-valued random variables is uniformly integrable if and only if $\sup _{i \in I}\left\|X_{i}\right\|_{L^{1}}<\infty$ holds as well as

$$
\forall \varepsilon>0 \exists \delta>0 \forall A \in \mathscr{F}: \mathbb{P}(A)<\delta \Rightarrow \sup _{i \in I} \mathbb{E}\left[\left|X_{i}\right| \mathbf{1}_{A}\right]<\varepsilon
$$

2. Show for an $L^{p}$-bounded martingale $\left(M_{n}\right)$ (i.e. $\sup _{n} \mathbb{E}\left[\left|M_{n}\right|^{p}\right]<\infty$ ) with $p \in$ $(1, \infty)$ :
(a) $\left(M_{n}\right)$ converges a.s. and in $L^{1}$ to some $M_{\infty} \in L^{1}$.
(b) Use $\left|M_{\infty}\right| \leqslant \sup _{n \geqslant 0}\left|M_{n}\right|$ and Doob's inequality to infer $M_{\infty} \in L^{p}$.
(c) Prove with dominated convergence that $\left(M_{n}\right)$ converges to $M_{\infty}$ in $L^{p}$.
3. Give a martingale proof of Kolmogorov's 0-1 law:
(a) Let $\left(\mathscr{F}_{n}\right)$ be a filtration and $\mathscr{F}_{\infty}=\sigma\left(\mathscr{F}_{n}, n \geqslant 0\right)$. Then for $A \in \mathscr{F}_{\infty}$ we have $\lim _{n \rightarrow \infty} \mathbb{E}\left[\mathbf{1}_{A} \mid \mathscr{F}_{n}\right]=\mathbf{1}_{A}$ a.s.
(b) For a sequence $\left(X_{n}\right)_{n \geqslant 1}$ of independent random variables consider the natural filtration $\left(\mathscr{F}_{n}^{X}\right)$ and the terminal $\sigma$-algebra $\mathcal{T}:=\bigcap_{n \geqslant 1} \sigma\left(X_{k}, k \geqslant\right.$ $n$ ). Then for $A \in \mathcal{T}$ deduce $\mathbb{P}(A)=\mathbb{E}\left[\mathbf{1}_{A} \mid \mathscr{F}_{n}^{X}\right] \rightarrow \mathbf{1}_{A}$ a.s. for $n \rightarrow \infty$, implying $\mathbb{P}(A) \in\{0,1\}$.
4. A monkey types at random the 26 capital letters of the Latin alphabet. Let $\tau$ be the first time by which the monkey has completed the sequence ABRACADABRA. Prove that $\tau$ is almost surely finite and satisfies

$$
\mathbb{E}[\tau]=26^{11}+26^{4}+26
$$

Give an example of an 11 -letter word where $\mathbb{E}[\tau]=26^{11}$.
Hint: You may look at a fair game with gamblers $G_{n}$ arriving before times $n=1,2, \ldots$ Then $G_{n}$ bets 1 Euro on 'A' for letter $n$; if she wins, she puts 26 Euro on 'B' for letter $n+1$, otherwise she stops. If she wins again, she puts $26^{2}$ Euro on 'R', otherwise she stops etc. What is the balance of the game maker at time $\tau$ ?

Submit the solutions before the lecture on Thursday, 30 November 2023.

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## Exercises: sheet 7

1. Let $Z_{n}(x)=(3 / 2)^{n} \sum_{k \in\{0,2\}^{n}} \mathbf{1}_{I(k, n)}(x), x \in[0,1)$, with intervals $I(k, n):=$ $\left[\sum_{i=1}^{n} k_{i} 3^{-i}, \sum_{i=1}^{n} k_{i} 3^{-i}+3^{-n}\right)$. Show:
(a) $\left(Z_{n}\right)_{n \geqslant 0}$ with $Z_{0}=1$ forms a martingale on $\left([0,1), \mathfrak{B}_{[0,1]}, \lambda,\left(\mathscr{F}_{n}\right)\right)$ with Lebesgue measure $\lambda$ on $[0,1)$ and $\mathscr{F}_{n}:=\sigma\left(I(k, n), k \in\{0,1,2\}^{n}\right)$.
(b) $\left(Z_{n}\right)$ converges $\lambda$-a.s., but not in $L^{1}\left([0,1], \mathfrak{B}_{[0,1]}, \lambda\right)$.
(c) Interpret $Z_{n}$ as the density of a probability measure $\mathbb{P}_{n}$ with respect to $\lambda$. Then $\left(\mathbb{P}_{n}\right)$ converges weakly to some probability measure $\mathbb{P}_{\infty}\left(\mathbb{P}_{\infty}\right.$ is called Cantor measure). Identify a Borel set $C \subseteq[0,1]$ with $\mathbb{P}_{\infty}(C)=1$, $\lambda(C)=0$ so that $\mathbb{P}_{\infty} \perp \lambda$.
Hint: Show that the distribution functions converge.
2. Let $\left(X_{k}\right)_{k \geqslant 1}$ be a sequence of i.i.d. $\{-1,+1\}$-valued random variables. Under the probability measure $\mathbb{P}_{0}$ (the null hypothesis $H_{0}$ ) we have $\mathbb{P}_{0}\left(X_{k}=+1\right)=p_{0}$ with $p_{0} \in(0,1)$, while under $\mathbb{P}_{1}$ (the alternative $\left.H_{1}\right)$ we have $\mathbb{P}_{1}\left(X_{k}=+1\right)=p_{1}$ with $p_{1} \in(0,1), p_{1} \neq p_{0}$.
(a) Explain why the likelihood quotient $L_{n}=\frac{d\left(\otimes_{i=1}^{n} \mathbb{P}_{1}^{X_{i}}\right)}{d\left(\otimes_{i=1}^{n} \mathbb{P}_{0}^{X_{i}}\right)}$ after $n$ observations $X_{1}, \ldots, X_{n}$ is given by

$$
L_{n}=\frac{p_{1}^{\left(n+S_{n}\right) / 2}\left(1-p_{1}\right)^{\left(n-S_{n}\right) / 2}}{p_{0}^{\left(n+S_{n}\right) / 2}\left(1-p_{0}\right)^{\left(n-S_{n}\right) / 2}} \text { with } S_{n}=\sum_{k=1}^{n} X_{k}
$$

(b) Show that the likelihood process $\left(L_{n}\right)_{n \geqslant 0}$ (put $L_{0}:=1$ ) forms a nonnegative martingale under the hypothesis $H_{0}$ (i.e. under $\mathbb{P}_{0}$ ) with respect to its natural filtration.
(c) A sequential likelihood-quotient test, based on $0<A<B$ and the stopping time

$$
\tau_{A, B}:=\inf \left\{n \geqslant 1 \mid L_{n} \geqslant B \text { or } L_{n} \leqslant A\right\}
$$

rejects $H_{0}$ if $L_{\tau_{A, B}} \geqslant B$, and accepts $H_{0}$ if $L_{\tau_{A, B}} \leqslant A$. Determine the probability for errors of the first and second kind (i.e., $\mathbb{P}_{0}\left(L_{\tau_{A, B}} \geqslant B\right)$ and $\left.\mathbb{P}_{1}\left(L_{\tau_{A, B}} \leqslant A\right)\right)$ in the case $p_{0}=0.4, p_{1}=0.6, A=(2 / 3)^{5}, B=(3 / 2)^{5}$. Calculate $\mathbb{E}\left[\tau_{A, B}\right]$.
$\left(d^{*}\right)$ Compare the error probabilities of this sequential test with those of the test which after $n=\left\lfloor\mathbb{E}\left[\tau_{A, B}\right]\right\rfloor$ observations rejects $H_{0}$ if $L_{n} \geqslant 1$ and accepts $H_{0}$ if $L_{n}<1$.
3. Prove in detail for probability measures $\mathbb{Q} \ll \mathbb{P}, Z=\frac{d \mathbb{Q}}{d \mathbb{P}}$ and $Y \in L^{1}(\mathbb{Q})$ that $Y Z$ is in $L^{1}(\mathbb{P})$ and that the identity

$$
\mathbb{E}_{\mathbb{Q}}[Y]=\mathbb{E}_{\mathbb{P}}[Y Z] \text {, i.e. } \int Y d \mathbb{Q}=\int Y \frac{d \mathbb{Q}}{d \mathbb{P}} d \mathbb{P}
$$

holds. Give an example where $Y$ is in $L^{1}(\mathbb{Q})$, but not in $L^{1}(\mathbb{P})$.
4. Suppose $\mu_{0}, \mu_{1}, \mu_{2}$ are measures on $(\Omega, \mathscr{F})$ so that $\mu_{2}$ has a $\mu_{1}$-density $f_{2,1}$ and $\mu_{1}$ has a $\mu_{0}$-density $f_{1,0}$ (i.e., $\mu_{1}(A)=\int_{A} f_{1,0} d \mu_{0}$ etc.). Show:
(a) $\mu_{0}$ and $\mu_{1}$ are equivalent if and only if $f_{1,0}>0$ holds $\mu_{0}$-a.e. In that case $f_{0,1}:=f_{1,0}^{-1}$ is $\mu_{0}$-a.e. and $\mu_{1}$-a.e. the $\mu_{1}$-density of $\mu_{0}$.
Short-hand notation: $\frac{d \mu_{0}}{d \mu_{1}}=\left(\frac{d \mu_{1}}{d \mu_{0}}\right)^{-1}$.
(b) We have $\mu_{2} \ll \mu_{0}$ and $f_{2,0}:=f_{2,1} f_{1,0}$ is $\mu_{0}$-a.e. the $\mu_{0}$-density of $\mu_{2}$. Short-hand notation: $\frac{d \mu_{2}}{d \mu_{0}}=\frac{d \mu_{2}}{d \mu_{1}} \frac{d \mu_{1}}{d \mu_{0}}$.

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## Exercises: sheet 8

1. The recursion $X_{n}=a X_{n-1}+\varepsilon_{n}$ for $n \geqslant 1$ with $a \in \mathbb{R}$ and independent random variables $\varepsilon_{n} \sim N\left(0, \sigma^{2}\right), X_{0} \sim N\left(\mu_{0}, \sigma_{0}^{2}\right)$ defines a so called autoregressive process of order one.
(a) Show that $\left(X_{n}, n \geqslant 0\right)$ forms a Gaussian process.
(b) For which values of $a$ do $\mu_{0} \in \mathbb{R}, \sigma_{0}>0$ exist such that $\left(X_{n}, n \geqslant 0\right)$ is stationary?
$\left(c^{*}\right)$ (optional) Simulate several trajectories for $a \in\{-1,-0.5,0,1,2\}$ and different $\mu_{0}, \sigma_{0}$. Explain what you see.
2. Let $\left(X_{n}\right)_{n \geqslant 0}$ be a time-homogeneous Markov chain with initial distribution $\mu$. Show that the following are equivalent:
(a) $\left(X_{n}\right)$ is a stationary process;
(b) $\mu$ is an invariant initial distribution, i.e. $\mathbb{P}_{\mu}\left(X_{1} \in B\right)=\mu(B)$ for all $B \subseteq S$.

Consider the one-step transition matrix of a Markov chain on $S=\{1,2,3\}$

$$
P(1)=\left(\begin{array}{ccc}
p_{11} & p_{12} & 0 \\
0 & p_{22} & p_{23} \\
0 & p_{32} & p_{33}
\end{array}\right)
$$

with each $p_{i j}>0$. Visualise this by a graph with directed edges along positive transition probabilities. Then determine an invariant initial distribution $\mu$.
3. Let $\mathscr{I}_{T}$ be the $\sigma$-algebra of invariant events for the measure-preserving map $T$ on $(\Omega, \mathscr{F}, \mathbb{P})$. Show:
(a) A random variable $Y$ is $\mathscr{I}_{T}$-measurable if and only if $Y \circ T=Y$ holds P-a.s.
(b) $T$ is ergodic if and only if all bounded random variables $Y$ with $Y \circ T=Y$ $\mathbb{P}$-a.s. are constant $\mathbb{P}$-a.s.
(c) For all invariant events $A$ there is a strictly invariant event $B$ (i.e., $T^{-1}(B)=B$ holds) such that $\mathbb{P}(A \Delta B):=\mathbb{P}(A \backslash B \cup B \backslash A)=0$.
4. Read Ryan Tibshirani's slides on Google's PageRank algorithm (lecture 3 under http://www.stat.cmu.edu/~ryantibs/datamining) and explain briefly the main ideas.

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## Exercises: sheet 9

1. Extend Birkhoff's ergodic theorem to an $L^{p}$-ergodic theorem:

For measure-preserving $T$ and $X \in L^{p}, p \geqslant 1$, consider $A_{n}:=\frac{1}{n} \sum_{i=0}^{n-1} X \circ T^{i}$. Then $\left(\left|A_{n}\right|^{p}\right)_{n \geqslant 1}$ is uniformly integrable and $A_{n} \rightarrow \mathbb{E}\left[X \mid \mathscr{I}_{T}\right]$ holds in $L^{p}$.
2. Show that a measure-preserving map $T$ on $(\Omega, \mathscr{F}, \mathbb{P})$ is ergodic if and only if for all $A, B \in \mathscr{F}$

$$
\lim _{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \mathbb{P}\left(A \cap T^{-k} B\right)=\mathbb{P}(A) \mathbb{P}(B)
$$

Hint: For one direction apply an ergodic theorem to $\mathbf{1}_{B}$.
(*optional) Extension: If even $\lim _{n \rightarrow \infty} \mathbb{P}\left(A \cap T^{-n} B\right)=\mathbb{P}(A) \mathbb{P}(B)$ holds, then $T$ is called mixing. Show that $T$ mixing implies $T$ ergodic, but not conversely (e.g., consider rotation by an irrational angle).
3. Gelfand's Problem: Does the decimal representation of $2^{n}$ ever start with the initial digit 7? Study this as follows:
(a) Determine the relative frequencies of the initial digits of $\left(2^{n}\right)_{1 \leqslant n \leqslant 30}$.
(b) Let $A \sim U([0,1])$. Prove that the relative frequency of the initial digit $k$ in $\left(10^{A} 2^{n}\right)_{1 \leqslant n \leqslant m}$ converges as $m \rightarrow \infty$ a.s. to $\log _{10}(k+1)-\log _{10}(k)$.
Hint: consider $X_{n}=A+n \log _{10}(2) \bmod 1$ and argue via ergodicity.
(c) Prove that the convergence in (b) even holds everywhere. In particular, the relative frequency of the initial digit 7 in the powers of 2 converges to $\log _{10}(8 / 7) \approx 0,058$.
Hint: Show for trigonometric polynomials $p(a)=\sum_{|m| \leqslant M} c_{m} e^{2 \pi i m a}$ that $\frac{1}{n} \sum_{k=0}^{n-1} p(a+k \eta) \rightarrow \int_{0}^{1} p(x) d x$ holds for all $\eta \in \mathbb{R} \backslash \mathbb{Q}, a \in[0,1]$ (calculate explicitly for monomials!) and approximate (you may use Weierstraß's Theorem: trigonometric polynomials are dense in $\left.\left(C([0,1]),\|\bullet\|_{\infty}\right)\right)$.

Suggested reading: Benford's law and fraud detection for election results, tax declarations and corona statistics, e.g.https://en.wikipedia.org/wiki/ Benford\%27s_law.
4. Consider the set $\mathscr{I}$ of all invariant initial distributions of a recurrent Markov chain on a state space $S$. Prove:
(a) $\mathscr{I}$ is convex.
(b) If $\pi \in \mathscr{I}$ is even ergodic (that is $\mathbb{P}_{\pi}$ is ergodic), then there is a connected component $[x]$ with $\pi([x])=1$ and $\pi(\{y\})>0$ for all $y \in[x]$.
(c) If $\pi, \pi^{\prime} \in \mathscr{I}$ are both ergodic, then $\pi=\pi^{\prime}$ or $\pi \perp \pi^{\prime}$ follows.
(d) Suppose $\mathscr{I} \neq \varnothing$. By decomposing $S=\bigcup_{n}\left[x_{n}\right]$ into pairwise disjoint components $\left[x_{n}\right]$, there are ergodic $\pi_{n} \in \mathscr{I}$ with $\pi_{n}\left(\left[x_{n}\right]\right)=1$ so that we can write any $\pi \in \mathscr{I}$ as convex combination $\pi=\sum_{n} \alpha_{n} \pi_{n}$ with $\alpha_{n} \geqslant 0$, $\sum_{n} \alpha_{n}=1$. In particular, for an irreducible chain $\mathscr{I}$ contains at most one element, which is then ergodic.
Here, the union and the sum extend over finitely or countably many $n$.

Submit the solutions before the lecture on Thursday, 18 January 2024.

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## Exercises: sheet 10

1. Let the initial distribution $\pi$ of a Markov chain with one-step transition probabilities $p_{x y}(1)$ satisfy $\pi(\{x\}) p_{x y}(1)=\pi(\{y\}) p_{y x}(1)$ for all states $x, y$ (then $\pi$ is called reversible). Show:
(a) $\pi$ is an invariant initial distribution.
(b) $\mathbb{P}_{\pi}\left(X_{0}=x_{0}, \ldots, X_{n}=x_{n}\right)=\mathbb{P}_{\pi}\left(X_{n}=x_{0}, \ldots, X_{0}=x_{n}\right)$ holds for all $x_{0}, \ldots, x_{n} \in S$ (use induction over $n \geqslant 1$ ). Explain in your words what this reversibility of the Markov chain means.
(c) The transition operator $P$ is $L^{2}(\pi)$-self-adjoint in the sense $\langle P f, g\rangle_{\pi}=$ $\langle f, P g\rangle_{\pi}$ for all $f, g \in L^{2}(\pi)$.
2. For random variables $X, Y$ on $(\Omega, \mathscr{F}, \mathbb{P})$ with values in a Polish space $(S, d)$ with Borel $\sigma$-algebra define $d_{0}(X, Y):=\mathbb{E}[d(X, Y) \wedge 1]$. Show:
(a) $\omega \mapsto d(X(\omega), Y(\omega))$ is measurable and $d_{0}$ defines a metric on the space $L^{0}(\Omega ; S)$ of all $S$-valued random variables on $(\Omega, \mathscr{F}, \mathbb{P})$, when $\mathbb{P}$-a.s. equal random variables are identified.
(b) $d_{0}\left(X_{n}, X\right) \rightarrow 0 \Longleftrightarrow X_{n} \xrightarrow{\mathbb{P}} X$ (stochastic convergence).
(c) $X_{n} \xrightarrow{\mathbb{P}} X$ implies $X_{n} \xrightarrow{d} X$ (convergence in distribution).
(d) $X_{n} \xrightarrow{d} c$ for some constant $c \in S$ implies $X_{n} \xrightarrow{\mathbb{P}} c$.
3. Let $\left(X_{k}\right)_{k \geqslant 1}$ be an i.i.d. sequence of random variables in $L^{2}$ with $\mu=\mathbb{E}\left[X_{k}\right]$. Introduce the sample mean $\bar{X}_{n}:=\frac{1}{n} \sum_{k=1}^{n} X_{k}$ and the sample variance $\bar{\sigma}_{n}^{2}:=$ $\frac{1}{n-1} \sum_{k=1}^{n}\left(X_{k}-\bar{X}_{n}\right)^{2}$. Use a CLT and Slutsky's Lemma to prove for $n \rightarrow \infty$

$$
\frac{\sqrt{n}\left(\bar{X}_{n}-\mu\right)}{\bar{\sigma}_{n}} \xrightarrow{d} N(0,1) .
$$

Determine approximately a real number $c>0$ such that

$$
\lim _{n \rightarrow \infty} \mathbb{P}\left(\mu \in\left[\bar{X}-c \frac{\bar{\sigma}_{n}}{\sqrt{n}}, \bar{X}+c \frac{\bar{\sigma}_{n}}{\sqrt{n}}\right]\right)=0.95
$$

4. Let $\alpha \in(0,1)$. Choose $X_{0} \in[0,1]$ and perform the following independent iterations for $n \in \mathbb{N}$ : given $X_{n-1} \in[0,1]$, go with probability $1 / 2$ left, setting $X_{n}=\alpha X_{n-1}$, and with probability $1 / 2$ right, setting $X_{n}=(1-\alpha)+\alpha X_{n-1}$.
(a) Write $X_{n}=\alpha X_{n-1}+(1-\alpha) Z_{n}, n \in \mathbb{N}$, with suitable i.i.d. random variables $\left(Z_{n}\right)$. Interpret $\left(X_{n}, n \geqslant 0\right)$ as a Markov process on $\left([0,1], \mathfrak{B}_{[0,1]}\right)$.
(b) For $\alpha=1 / 2$ and $\alpha=1 / 3$ determine an invariant initial distribution $\mu$ such that $\left(X_{n}, n \geqslant 0\right)$ becomes stationary with $X_{0} \sim \mu$.
Hint: Represent $x \in[0,1]$ in a dyadic or triadic expansion.
(c) Show that, whatever the initial distribution of $X_{0}$ is, we have $X_{n} \xrightarrow{d} \mu$ in (b). Conclude that with $X_{0} \sim \mu$ the process $\left(X_{n}, n \geqslant 0\right)$ is ergodic.
(d*) (Optional, but beautiful!) Consider the triangle $\Delta$ spanned by the corner points $(0,0),(1,0),(0,1)$ in $\mathbb{R}^{2}$. Perform iterations, where for given $X_{n-1} \in \Delta$ with probability $1 / 3$ one of the corners is selected and $X_{n}$ is obtained as the middle point between that corner and $X_{n-1}$. Expand $x \in \Delta$ as $x=\sum_{i} b_{i} 2^{-i}$ with certain $b_{i} \in\{0,1\}^{2}$ and describe the unique invariant initial distribution $\mu$. Plot the support set of $\mu$ approximately by simulating $\left(X_{n}\right)$. Try to understand and explore further!
Application: a treasure is hidden in the triangle spanned by three pyramids. A treasure hunter starts digging somewhere in the triangle and then moves half way to one of the pyramids at random to dig next etc. Does he asymptotically dig in a dense subset of the triangle and thus find the treasure eventually?

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## Exercises: sheet 11

1. Consider a distribution $\pi$ on $S$ with $\pi(\{x\})>0$ for all $x \in S$ and an irreducible Markov chain on $S$ with transition probabilities $q_{x y}(1), x, y \in S$, satisfying $q_{x y}(1)>0$ if and only if $q_{y x}(1)>0$ for $x, y \in S$. Prove that the Markov chain with transition probabilities

$$
p_{x y}(1):= \begin{cases}\min \left(q_{x y}(1), \frac{\pi(\{y\})}{\pi(\{x\})} q_{y x}(1)\right), & \text { if } x \neq y \\ 1-\sum_{z \neq x} p_{x z}(1), & \text { if } x=y\end{cases}
$$

is reversible with respect to $\pi$ and irreducible. If the transition matrix $Q(1)$ is aperiodic or if $\pi$ is not reversible with respect to $Q(1)$, deduce that the transition matrix $P(1)$ is aperiodic.
2. Read Example 18.16 (Ising model) in the book by Klenke. Write down the Boltzmann distribution $\pi$ on $S=\{-1,+1\}^{\Lambda}$ and explain briefly the quantities appearing. Prove in detail that the proposal Markov chain there satisfies the properties in Problem 1. Then derive rigorously that the Markov chain $X_{n}=$ $F_{n}\left(X_{n-1}\right)$ has invariant distribution $\pi$. Is $\left(X_{n}\right)$ aperiodic?
3. For probability measures $\mathbb{P}$ and $\mathbb{Q}$ on a measurable space $(\Omega, \mathcal{F})$ their total variation distance is given by $\|\mathbb{P}-\mathbb{Q}\|_{T V}=\sup _{A \in \mathcal{F}}|\mathbb{P}(A)-\mathbb{Q}(A)|$. Prove that convergence in total variation implies weak convergence on metric spaces.
Decide whether for $n \rightarrow \infty$ the probabilities $\mathbb{P}_{n}$ with the following Lebesgue densities $f_{n}$ on $\mathbb{R}$ converge in total variation, weakly or not at all:

$$
f_{n}(x)=n e^{-n x} \mathbf{1}_{[0, \infty)}(x), \quad f_{n}(x)=\frac{n+1}{n} x^{1 / n} \mathbf{1}_{[0,1]}(x), \quad f_{n}(x)=\frac{1}{n} \mathbf{1}_{[0, n]}(x)
$$

4. Prove: Every relatively (weakly) compact family $\left(\mathbb{P}_{i}\right)_{i \in I}$ of probability measures on a Polish space $\left(S, \mathfrak{B}_{S}\right)$ is uniformly tight. Proceed as follows (compare the proof of Ulam's Theorem):
(a) For $k \geqslant 1$ consider open balls $\left(A_{k, m}\right)_{m \geqslant 1}$ of radius $1 / k$ that cover $S$. If $\lim _{M \rightarrow \infty} \inf _{i} \mathbb{P}_{i}\left(\bigcup_{m=1}^{M} A_{k, m}\right)<1$ were true, then by assumption and by the Portmanteau Theorem we would have $\lim _{M \rightarrow \infty} \mathbb{Q}\left(\bigcup_{m=1}^{M} A_{k, m}\right)<1$ for some limiting probability measure $\mathbb{Q}$, which is contradictory.
(b) Conclude that for any $\varepsilon>0, k \geqslant 1$ there are indices $M_{k, \varepsilon} \geqslant 1$ such that $\inf _{i} \mathbb{P}_{i}(K)>1-\varepsilon$ holds with $K:=\bigcap_{k \geqslant 1} \bigcup_{m=1}^{M_{k, \varepsilon}} A_{k, m}$. Moreover, $K$ is relatively compact in $S$, which suffices.

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## Exercises: sheet 12

1. We say that a family of real-valued random variables $\left(X_{i}\right)_{i \in I}$ is stochastically bounded, notation $X_{i}=O_{\mathbb{P}}(1)$, if $\lim _{R \rightarrow \infty} \sup _{i \in I} \mathbb{P}\left(\left|X_{i}\right|>R\right)=0$.
(a) Show $X_{i}=O_{\mathbb{P}}(1)$ if and only if the laws $\left(\mathbb{P}^{X_{i}}\right)_{i \in I}$ are uniformly tight.
(b) Prove that any $L^{p}$-bounded family of random variables is stochastically bounded, hence has uniformly tight laws.
(c) If $X_{n} \xrightarrow{\mathbb{P}} 0$ holds, then we write $X_{n}=o_{\mathbb{P}}(1)$. Check the symbolic rules $O_{\mathbb{P}}(1)+O_{\mathbb{P}}(1)=O_{\mathbb{P}}(1)$ and $O_{\mathbb{P}}(1) o_{\mathbb{P}}(1)=o_{\mathbb{P}}(1)$.
2. For probability measures $\mathbb{P}, \mathbb{Q}$ on a metric space $(S, d)$ with Borel $\sigma$-algebra define the Bounded-Lipschitz metric

$$
d_{B L}(\mathbb{P}, \mathbb{Q}):=\sup \left\{\left|\int_{S} f d \mathbb{P}-\int_{S} f d \mathbb{Q}\right| \mid f \in B L_{1}(S)\right\}
$$

with $B L_{1}(S)=\left\{f: S \rightarrow \mathbb{R}\left|\|f\|_{\infty} \leqslant 1, \forall x, y \in S:|f(x)-f(y)| \leqslant d(x, y)\right\}\right.$. Prove that $d_{B L}$ is indeed a metric and that $d_{B L}\left(\mathbb{P}_{n}, \mathbb{P}\right) \rightarrow 0 \Rightarrow \mathbb{P}_{n} \xrightarrow{w} \mathbb{P}$.
For $S=[0, T]$ use the Arzelà-Ascoli Theorem to prove

$$
d_{B L}\left(\mathbb{P}_{n}, \mathbb{P}\right) \rightarrow 0 \Longleftrightarrow \mathbb{P}_{n} \xrightarrow{w} \mathbb{P}
$$

Remark: This holds in fact on any Polish space $(S, d)$.
3. Let $\left(B_{t}, t \geqslant 0\right)$ be a Brownian motion. Verify that the following processes are also Brownian motions:
(a) $\left(-B_{t}, t \geqslant 0\right)$;
(b) $\left(a^{-1 / 2} B_{a t}, t \geqslant 0\right)$ for any $a>0$ ('time change');
(c) $\left(X_{t}, t \geqslant 0\right)$ with $X_{t}=t B_{1 / t}$ for $t>0$ and $X_{0}=0$ ('time inversion').
*4. (Optional) We want to show that a Brownian motion $B$ is a.s. not $1 / 2$-Hölder continuous at zero and a.s. hits zero again immediately after start in zero.
(a) Let $A_{s}:=\left\{\exists t \in(0, s]: B_{t} \geqslant K \sqrt{t}\right\}, s>0$, for some $K>0$. Use invariance of $B$ under time changes to prove $\mathbb{P}\left(A_{s}\right)=\mathbb{P}\left(A_{1}\right)$ for all $s>0$.
(b) By letting $s \downarrow 0$ deduce

$$
\mathbb{P}\left(\inf \left\{t>0 \mid B_{t} \geqslant K \sqrt{t}\right\}=0\right) \geqslant \mathbb{P}\left(B_{1} \geqslant K\right)>0 .
$$

(c) Apply Blumenthal's 0-1 law (follows from Kolmogorov's 0-1 law, e.g. Thm. 21.15 in Klenke) to deduce that $\inf \left\{t>0 \mid B_{t} \geqslant K \sqrt{t}\right\}=0$ almost surely.
(d) This implies that with probability one there is for any $\varepsilon>0$ a sequence $\left(t_{K}\right)_{K \geqslant 1} \subseteq(0, \varepsilon)$ with $B_{t_{K}} \geqslant K \sqrt{t_{K}}$ for all $K \in \mathbb{N}$, that is $\lim \sup _{t \rightarrow 0} t^{-1 / 2} B_{t}=\infty$ a.s.
(e) By Problem 3(a) we obtain further $\lim \inf _{t \rightarrow 0} t^{-1 / 2} B_{t}=-\infty$ a.s. so that $\inf \left\{t>0 \mid B_{t}=0\right\}=0$ a.s. By Problem 3(c) we get with probability one $\limsup _{t \rightarrow \infty} t^{-1 / 2} B_{t}=\infty, \liminf _{t \rightarrow \infty} t^{-1 / 2} B_{t}=-\infty, \sup \left\{t>0 \mid B_{t}=0\right\}=\infty$.

Submit the solutions before the lecture on Thursday, 8 February 2024.


[^0]:    Submit the solutions before the lecture on Thursday, 7 December 2023.

