## Sheet 0, "Introduction to Stochastic Analysis"

These exercises will be discussed in the first tutorial. You are encouraged to study them in advance, but do not have to submit any solutions.

## 1. (Conditional expectation)

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, $\mathcal{G} \subset \mathcal{F}$ a sub- $\sigma$-algebra. Let $X, Y$ absolutely integrable random variables. First, define the conditional expectation $\mathbb{E}(X \mid \mathcal{G})$. Then prove the following statements:

1. The map $X \rightarrow \mathbb{E}(X \mid \mathcal{G})$ is linear.
2. If $\mathcal{B} \subset \mathcal{G}$ is a $\sigma$-algebra, then $\mathbb{E}[\mathbb{E}(X \mid \mathcal{G}) \mid \mathcal{B}]=\mathbb{E}(X \mid \mathcal{B})$, a.s. This is called the towerproperty. In particular, this implies $\mathbb{E}[\mathbb{E}[X \mid \mathcal{G}]]=\mathbb{E}[X]$.
3. If $X \leq Y$ a.s., then $\mathbb{E}(X \mid \mathcal{G}) \leq \mathbb{E}(Y \mid \mathcal{G})$, a.s.
4. $|\mathbb{E}(X \mid \mathcal{G})| \leq \mathbb{E}(|X| \mid \mathcal{G})$ a.s.
5. Assume that there exists $n \in \mathbb{N}$ and $A_{1}, \ldots, A_{n} \subset \Omega$ such that $\left\{A_{1}, \ldots, A_{n}\right\}$ are pairwise disjoint, $\mathbb{P}\left(A_{i}\right)>0$ for all $i, \Omega=\cup_{i=1}^{n} A_{i}$ and $\mathcal{G}=\sigma\left(\left\{A_{1}, \ldots, A_{n}\right\}\right)$. Show that $\mathbb{E}(X \mid \mathcal{G})=\sum_{i=1}^{n} \mathbb{E}\left[X \mid A_{i}\right] \mathbb{1}_{A_{i}}$ a.s., where $\mathbb{E}\left[X \mid A_{i}\right]=\mathbb{E}\left[X \mathbb{1}_{A_{i}}\right] \mathbb{P}\left[A_{i}\right]^{-1}$.

## 2. (Inequalities)

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space and let $\mathcal{G} \subset \mathcal{F}$ be a sub- $\sigma$-algebra. For $X, Y$ absolutely integrable random variables, prove that:

1. Prove the conditional Markov inequality, i.e. show that, if $f: \mathbb{R}_{+} \rightarrow \mathbb{R}_{+}$is nondecreasing and such that $f(|X|)$ is integrable, then

$$
\mathbb{P}[|X| \geq \alpha \mid \mathcal{G}] \leq \frac{1}{f(\alpha)} \mathbb{E}[f(|X|) \mid \mathcal{G}] \quad \mathbb{P} \text {-a.s. }
$$

2. Let $\phi: \mathbb{R} \rightarrow \mathbb{R}$ be a convex function, $X$ and $\phi(X)$ be integrable random variables. Prove the conditional Jensen inequality

$$
\phi(\mathbb{E}[X \mid \mathcal{G}]) \leq \mathbb{E}[\phi(X) \mid \mathcal{G}] .
$$

Hint: You can use that for $x, y \in \mathbb{R}$ there exists a measurable function $c: \mathbb{R} \rightarrow \mathbb{R}$ such that

$$
\phi(x) \geq \phi(y)+c(y)(x-y) .
$$

## 3. (Independence)

Let $X, Y$ integrable random variables on $(\Omega, \mathcal{F}, \mathbb{P})$. Let $\sigma(X)$ the $\sigma$-algebra generated by $X$ (what does this mean)? Let $\mathbb{E}[X \mid Y]:=\mathbb{E}[X \mid \sigma(Y)]$ and call two $\sigma$-algebras $\mathcal{A}, \mathcal{B}$ independent if for all $A \in \mathcal{A}, B \in \mathcal{B}: \mathbb{P}(A \cap B)=\mathbb{P}(A) \cdot \mathbb{P}(B)$. From now on, we call $X$ and $Y$ independent if $\sigma(X)$ and $\sigma(Y)$ are independent. Prove:

1. If $Y$ is constant, then $X$ and $Y$ are independent.
2. If $\sigma(X)$ is independent of $\mathcal{G}$, then $\mathbb{E}[X \mid \mathcal{G}]=\mathbb{E}[X]$ a.s..
3. If $X, Y$ are independent and in $L^{2}(\mathbb{P})$, then $\mathbb{E}[X Y]=\mathbb{E}[X] \mathbb{E}[Y]$ (they are uncorrelated).

The tutorials will start in the second week of the lecture, taking place at

1. Monday 8ct
2. Monday 14 ct
3. Tuesday 14 ct

Register for the tutorials on ecampus here and send a short email containing your full name, your ecampus-ID (typically s6...) and your preferred tutorials to florian.kreten@unibonn.de. The more possibilities you send the higher the chance you get one of your preferred timeslots. Deadline for this is Thursday, 29 October, 19:00.

