18.445 HOMEWORK 4 SOLUTIONS

Exercise 1. Let X, Y be two random variables on $(\Omega, \mathcal{F}, \mathbb{P})$. Let $\mathcal{A} \subset \mathcal{F}$ be a sub- σ -algebra. The random variables X and Y are said to be independent conditionally on \mathcal{A} is for every non-negative measurable functions f, g, we have

$$\mathbb{E}[f(X)g(Y) \mid \mathcal{A}] = \mathbb{E}[f(X) \mid \mathcal{A}] \times \mathbb{E}[g(Y) \mid \mathcal{A}] \quad a.s.$$

Show that X, Y are independent conditionally on \mathcal{A} is and only if for every non-negative \mathcal{A} -measurable random variable Z, and every non-negative measurable functions f, g, we have

$$\mathbb{E}[f(X)g(Y)Z] = \mathbb{E}[f(X)Z\mathbb{E}[g(Y) \mid \mathcal{A}]].$$

Proof. If X and Y are independent conditionally on \mathcal{A} and Z is \mathcal{A} -measurable, then

$$\mathbb{E}[f(X)g(Y)Z] = \mathbb{E}\left[\mathbb{E}[f(X)g(Y)Z \mid \mathcal{A}]\right]$$
$$= \mathbb{E}\left[\mathbb{E}[f(X)g(Y) \mid \mathcal{A}]Z\right]$$
$$= \mathbb{E}\left[\mathbb{E}[f(X) \mid \mathcal{A}]\mathbb{E}[g(Y) \mid \mathcal{A}]Z\right]$$
$$= \mathbb{E}\left[\mathbb{E}\left[f(X)\mathbb{E}[g(Y) \mid \mathcal{A}]Z \mid \mathcal{A}\right]\right]$$
$$= \mathbb{E}\left[f(X)Z\mathbb{E}[g(Y) \mid \mathcal{A}]Z \mid \mathcal{A}\right]$$

Conversely, if this equality holds for every nonnegative \mathcal{A} -measurable Z, then in particular, for every $A \in \mathcal{A}$,

$$\mathbb{E}[f(X)g(Y)\mathbb{1}_A] = \mathbb{E}\Big[f(X)\mathbb{E}[g(Y) \mid \mathcal{A}]\mathbb{1}_A\Big].$$

It follows from the definition of conditional expectation that

$$\mathbb{E}[f(X)g(Y) \mid \mathcal{A}] = \mathbb{E}\Big[f(X)\mathbb{E}[g(Y) \mid \mathcal{A}] \mid \mathcal{A}\Big] = \mathbb{E}[f(Y) \mid \mathcal{A}]\mathbb{E}[g(Y) \mid \mathcal{A}],$$

so X and Y are independent conditionally on \mathcal{A} .

Exercise 2. Let $X = (X_n)_{n \ge 0}$ be a martingale.

(1). Suppose that T is a stopping time, show that X^T is also a martingale. In particular, $\mathbb{E}[X_{T \wedge n}] = \mathbb{E}[X_0]$. *Proof.* Since X is a martingale, first we have

$$\mathbb{E}[|X_n^T|] \le \mathbb{E}[\max_{i \le n} |X_i|] \le \sum_{i=1}^n \mathbb{E}[|X_i|] < \infty.$$

Moreover, for every $n \ge m$,

$$\mathbb{E}[X_n^T | \mathcal{F}_{n-1}] = \mathbb{E}[X_{n-1}^T + (X_n - X_{n-1}) \mathbb{1}_{T > n-1} | \mathcal{F}_{n-1}]$$

= $\mathbb{E}[X_{n-1}^T] + \mathbb{1}_{T > n-1} \mathbb{E}[X_n - X_{n-1} | \mathcal{F}_{n-1}]$
= $\mathbb{E}[X_{n-1}^T].$

We conclude that X^T is a martingale.

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(2). Suppose that $S \leq T$ are bounded stopping times, show that $\mathbb{E}[X_T | \mathcal{F}_S] = X_S, a.s.$ In particular, $\mathbb{E}[X_T] = \mathbb{E}[X_S].$

Proof. Suppose S and T are bounded by a constant $N \in \mathbb{N}$. For $A \in \mathcal{F}_S$,

$$\mathbb{E}[X_N \mathbb{1}_A] = \sum_{i=1}^N \mathbb{E}[X_N \mathbb{1}_A \mathbb{1}_{S=i}]$$

$$= \sum_{i=1}^N \mathbb{E}\Big[\mathbb{E}[X_N \mid \mathcal{F}_S] \mathbb{1}_A \mathbb{1}_{S=i}\Big]$$

$$= \sum_{i=1}^N \mathbb{E}\Big[\mathbb{E}[X_N \mid \mathcal{F}_i] \mathbb{1}_A \mathbb{1}_{S=i}\Big]$$

$$= \sum_{i=1}^N \mathbb{E}\Big[X_i \mathbb{1}_A \mathbb{1}_{S=i}\Big]$$

$$= \mathbb{E}[X_S \mathbb{1}_A],$$

so $\mathbb{E}[X_N | \mathcal{F}_S] = X_S$. Similarly, $\mathbb{E}[X_N | \mathcal{F}_T] = X_T$. We conclude that $\mathbb{E}[X_T | \mathcal{F}_S] = \mathbb{E}\left[\mathbb{E}[X_N | \mathcal{F}_T] | \mathcal{F}_S\right] = \mathbb{E}[X_N | \mathcal{F}_S] = X_S$.

(3). Suppose that there exists an integrable random variable Y such that $|X_n| \leq Y$ for all n, and T is a stopping time which is finite a.s., show that $\mathbb{E}[X_T] = \mathbb{E}[X_0]$.

Proof. Since $|X_n| \leq Y$ for all n and T is finite a.s., $|X_{n \wedge T}| \leq Y$. Then the dominated convergence theorem implies that

$$\lim_{n \to \infty} \mathbb{E}[X_{n \wedge T}] = \mathbb{E}[\lim_{n \to \infty} X_{n \wedge T}] = \mathbb{E}[X_T].$$

As $n \wedge T$ is a bounded stopping time, Part (2) implies that $\mathbb{E}[X_{n \wedge T}] = \mathbb{E}[X_0]$. Hence we conclude that $\mathbb{E}[X_T] = \mathbb{E}[X_0]$.

(4). Suppose that X has bounded increments, i.e. $\exists M > 0$ such that $|X_{n+1} - X_n| \leq M$ for all n, and T is a stopping time with $\mathbb{E}[T] < \infty$, show that $\mathbb{E}[X_T] = \mathbb{E}[X_0]$.

Proof. We can write $\mathbb{E}[X_T] = \mathbb{E}[X_0] + \mathbb{E}[\sum_{i=1}^T (X_i - X_{i-1})]$, so it suffices to show that the last term is zero. Note that

$$\mathbb{E}[|\sum_{i=1}^{T} (X_i - X_{i-1})|] \le \mathbb{E}[\sum_{i=1}^{T} |X_i - X_{i-1}|] \le M\mathbb{E}[T] < \infty.$$

Then the dominated convergence theorem implies that

$$\mathbb{E}[\sum_{i=1}^{T} (X_i - X_{i-1})] = \mathbb{E}[\sum_{i=1}^{\infty} (X_i - X_{i-1}) \mathbb{1}_{T \ge i}]$$

= $\sum_{i=1}^{\infty} \mathbb{E}[(X_i - X_{i-1}) \mathbb{1}_{T \ge i}]$
= $\sum_{i=1}^{\infty} \mathbb{E}[X_i - X_{i-1}] \mathbb{P}[T \ge i]$
= 0,

where we used that $X_i - X_{i-1}$ is independent of $\{T \ge i\} = \{T < i-1\}$ as T is a stopping time of the martingale X.

Exercise 3. Let $X = (X_n)_{n \ge 0}$ be Gambler's ruin with state space $\Omega = \{0, 1, 2, ..., N\}$: $X_0 = k$, $\mathbb{P}[X_{n+1} = X_n + 1 \mid X_n] = \mathbb{P}[X_{n+1} = X_n - 1 \mid X_n] = 1/2$, $\tau = \min\{n : X_n = 0 \text{ or } N\}$. (1). Show that $Y = (Y_n := X_n^2 - n)_{n \ge 0}$ is a martingale.

Proof. By the definition of X,

$$\mathbb{E}[Y_n \mid \mathcal{F}_{n-1}] = \mathbb{E}[X_n^2 - n \mid \mathcal{F}_{n-1}]$$

= $\mathbb{E}[(X_n - X_{n-1})^2 + 2(X_n - X_{n-1})X_{n-1} + X_{n-1}^2 - n \mid \mathcal{F}_{n-1}]$
= $\mathbb{E}[(X_n - X_{n-1})^2 \mid X_{n-1}] + 2\mathbb{E}[X_n - X_{n-1} \mid X_{n-1}]X_{n-1} + X_{n-1}^2 - n$
= $1 + 0 + X_{n-1}^2 - n = Y_{n-1},$

so Y is a martingale.

(2). Show that Y has bounded increments.

Proof. It is clear that

$$\begin{aligned} |Y_n - Y_{n-1}| &= |X_n^2 - X_{n-1}^2 - 1| \\ &\leq |X_n + X_{n-1}| |X_n - X_{n-1}| + 1 \\ &\leq |X_{n-1}| + 1 + |X_{n-1}| + 1 \\ &\leq 2N + 2, \end{aligned}$$

so Y has bounded increments.

(3). Show that $\mathbb{E}[\tau] < \infty$.

Proof. First, let α be the probability that the chain increases for N consecutive steps, i.e.

$$\alpha = \mathbb{P}[X_{i+1} - X_i = 1, X_{i+2} - X_{i+1} = 1, \dots, X_{i+N} - X_{i+N-1} = 1]$$

which is positive and does not depend on *i*. If $\tau > mN$, then the chain never increases N times consecutively in the first mN steps. In particular,

$$\{\tau > mN\} \subset \bigcap_{i=0}^{m-1} \{X_{iN+1} - X_{iN} = 1, X_{iN+2} - X_{iN+1} = 1, \dots, X_{iN+N} - X_{iN+N-1} = 1\}^c.$$

Since the events on the right-hand side are independent and each have probability $1 - \alpha < 1$,

$$\mathbb{P}[\tau > mN] \le (1-\alpha)^m$$

For $mN \leq l < (m+1)N$, $\mathbb{P}[\tau > l] \leq \mathbb{P}[\tau > mN]$, so

$$\mathbb{E}[\tau] = \sum_{l=0}^{\infty} \mathbb{P}[\tau > l] \le \sum_{m=0}^{\infty} N \mathbb{P}[\tau > mN] \le N \sum_{m=0}^{\infty} (1-\alpha)^m < \infty.$$

(4). Show that $\mathbb{E}[\tau] = k(N-k)$.

Proof. Since $\mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n] = 0$ and $|X_{n+1} - X_n| = 1$, X is a martingale with bounded increments. We also showed that Y is a martingale with bounded increments. As $\mathbb{E}[\tau] < \infty$, Exercise 2 Part (4) implies that

$$k = \mathbb{E}[X_0] = \mathbb{E}[X_\tau] = \mathbb{P}[X_\tau = 0] \cdot 0 + \mathbb{P}[X_\tau = N] \cdot N \tag{1}$$

and
$$k^2 = \mathbb{E}[Y_0] = \mathbb{E}[Y_\tau] = \mathbb{E}[X_\tau^2] - \mathbb{E}[\tau].$$
 (2)

Then (1) gives, $\mathbb{P}[X_{\tau} = N] = k/N$. Hence it follows from (2) that

$$\mathbb{E}[\tau] = \mathbb{E}[X_{\tau}^2] - k^2 = \mathbb{P}[X_{\tau} = 0] \cdot 0 + \mathbb{P}[X_{\tau} = N] \cdot N^2 - k^2 = kN - k^2 = k(N - k).$$

Exercise 4. Let $X = (X_n)_{n \ge 0}$ be the simple random walk on \mathbb{Z} .

(1). Show that $(Y_n := X_n^3 - 3nX_n)_{n \ge 0}$ is a martingale.

Proof. We have

$$\begin{split} & \mathbb{E}[Y_n - Y_{n-1} \mid \mathcal{F}_{n-1}] \\ &= \mathbb{E}[X_n^3 - 3nX_n - X_{n-1}^3 + 3(n-1)X_{n-1} \mid \mathcal{F}_{n-1}] \\ &= \mathbb{E}[(X_n - X_{n-1})^3 + 3(X_n - X_{n-1})^2 X_{n-1} + 3(X_n - X_{n-1})X_{n-1}^2 - 3n(X_n - X_{n-1}) - 3X_{n-1} \mid \mathcal{F}_{n-1}] \\ &= \mathbb{E}[(X_n - X_{n-1})^3] + 3\mathbb{E}[(X_n - X_{n-1})^2]X_{n-1} + 3\mathbb{E}[X_n - X_{n-1}]X_{n-1}^2 - 3n\mathbb{E}[X_n - X_{n-1}] - 3X_{n-1}] \\ &= 0 + 3X_{n-1} + 0 - 0 - 3X_{n-1} \\ &= 0, \end{split}$$

so Y is a martingale.

(2). Let τ be the first time that the walker hits either 0 or N. Show that, for $0 \le k \le N$, we have

$$\mathbb{E}_k[\tau \,|\, X_\tau = N] = \frac{N^2 - k^2}{3}$$

Proof. Since $0 \le X_n^{\tau} \le N$, the martingale Y^{τ} is bounded and thus has bounded increments. The stopping time τ is the same as in Exercise 3, so the same argument implies that

$$k^3 = \mathbb{E}[Y_0] = \mathbb{E}[Y_\tau] = \mathbb{E}[X_\tau^3] - 3\mathbb{E}[\tau X_\tau].$$

We compute that $\mathbb{E}[X_{\tau}^3] = \mathbb{P}[X_{\tau} = 0] \cdot 0 + \mathbb{P}[X_{\tau} = N] \cdot N^3 = kN^2$. Hence

$$\frac{kN^2 - k^3}{3} = \mathbb{E}[\tau X_{\tau}] = \mathbb{P}[X_{\tau} = 0] \cdot 0 + \mathbb{P}[X_{\tau} = N] \cdot \mathbb{E}[\tau N \mid X_{\tau} = N] = k\mathbb{E}[\tau \mid X_{\tau} = N].$$

We conclude that

$$\mathbb{E}[\tau \mid X_{\tau} = N] = \frac{N^2 - k^2}{3}.$$

Exercise 5. Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space with filtration $(\mathcal{F}_n)_{n\geq 0}$.

(1). For any $m, m' \ge n$ and $A \in \mathcal{F}_n$, show that $T = m \mathbf{1}_A + m' \mathbf{1}_{A^c}$ is a stopping time.

Proof. Assume without loss of generality that $m \leq m'$ (since we can flip the roles of A and A^c). If l < m, then $\{T \leq l\} = \emptyset \in \mathcal{F}_l$. If $m \leq l < m'$, then $\{T \leq l\} = A \in \mathcal{F}_n \subset \mathcal{F}_l$ as $n \leq m \leq l$. If $l \geq m'$, then $\{T \leq l\} = \Omega \in \mathcal{F}_l$. Hence T is a stopping time.

(2). Show that an adapted process $(X_n)_{n\geq 0}$ is a martingale if and only if it is integrable, and for every bounded stopping time T, we have $\mathbb{E}[X_T] = \mathbb{E}[X_0]$.

Proof. The "only if" part was proved in Exercise 2 Part (2) with $S \equiv 0$.

Conversely, suppose for every bounded stopping time T, we have $\mathbb{E}[X_T] = E[X_0]$. In particular, $\mathbb{E}[X_m] = \mathbb{E}[X_0]$ for every $m \in \mathbb{N}$. Moreover, for $n \leq m$ and $A \in \mathcal{F}_n$, Part (1) implies that $T = n\mathbb{1}_A + m\mathbb{1}_{A^c}$ is a bounded stopping time. Thus

$$\mathbb{E}[X_m] = \mathbb{E}[X_0] = \mathbb{E}[X_T] = \mathbb{E}[X_n \mathbb{1}_A + X_m \mathbb{1}_{A^c}],$$

so $\mathbb{E}[X_m \mathbb{1}_A] = \mathbb{E}[X_n \mathbb{1}_A]$. By definition, this means $\mathbb{E}[X_m | \mathcal{F}_n] = X_n$, so X is a martingale.

Exercise 6. Let $X = (X_n)_{n \ge 0}$ be a martingale in L^2 .

(1). Show that its increments $(X_{n+1} - X_n)_{n \ge 0}$ are pairwise orthogonal, i.e. for all $n \ne m$, we have $\mathbb{E}[(X_{n+1} - X_n)(X_{m+1} - X_m)] = 0.$

Proof. First, note that for any $n \leq m$,

$$\mathbb{E}[X_n X_m] = \mathbb{E}\Big[\mathbb{E}[X_n X_m \mid \mathcal{F}_n]\Big] = \mathbb{E}\Big[X_n \mathbb{E}[X_m \mid \mathcal{F}_n]\Big] = \mathbb{E}[X_n^2].$$

Now assume without loss of generality that n < m. Then

$$\mathbb{E}[(X_{n+1} - X_n)(X_{m+1} - X_m)] = \mathbb{E}[X_{n+1}X_{m+1}] - \mathbb{E}[X_nX_{m+1}] - \mathbb{E}[X_{n+1}X_m] + \mathbb{E}[X_nX_m]$$
$$= \mathbb{E}[X_{n+1}^2] - \mathbb{E}[X_n^2] - \mathbb{E}[X_{n+1}^2] + \mathbb{E}[X_n^2] = 0.$$

(2). Show that X is bounded in L^2 if and only if

$$\sum_{n\geq 0} \mathbb{E}[(X_{n+1} - X_n)^2] < \infty.$$

Proof. Note that

$$\mathbb{E}[X_0(X_{n+1} - X_n)] = \mathbb{E}[X_0^2] - \mathbb{E}[X_0^2] = 0$$

by the computation in Part (1). Thus for any m, we have

$$\mathbb{E}[X_m^2] = \mathbb{E}\Big[\Big(X_0 + \sum_{n=0}^{m-1} (X_{n+1} - X_n)\Big)^2\Big] = \mathbb{E}[X_0^2] + \sum_{n=0}^{m-1} \mathbb{E}[(X_{n+1} - X_n)^2]$$

where the cross terms disappear by Part (1). Therefore,

$$\sup_{m \ge 0} \mathbb{E}[X_m^2] = \mathbb{E}[X_0^2] + \sum_{n \ge 0} \mathbb{E}[(X_{n+1} - X_n)^2].$$
(3)

If X is bounded in L^2 , i.e. the left-hand side in (3) is bounded, then the sum on the right-hand side is bounded. Conversely, if the sum is bounded, since X_0 is in L^2 , the left-hand side is also bounded.

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