Probability

Solutions to Exam B, Fall 2015

Solution Q1:

(A) $S_m - S_n = X_{n+1} + \dots + X_m$. For any t, s we have that $\{S_m - S_n \leq t\} \in \sigma(X_{n+1}, \dots, X_m)$ and $\{S_n \leq s\} \in \sigma(X_1, \dots, X_n)$. These σ -algebras are independent, so $\{S_m - S_n \leq t\}$ and $\{S_n \leq s\}$ are independent events. So for all t, s,

$$\mathbb{P}[S_m - S_n \le t , S_n \le s] = \mathbb{P}[S_m - S_n \le t] \cdot \mathbb{P}[S_n \le s].$$

This implies that $S_m - S_n$ and S_n are independent.

(B) The random variable $S_n \cdot \mathbf{1}_{A_n}$ is a function of X_1, \ldots, X_n . The random variable $S_m - S_n$ is a function of X_{n+1}, \ldots, X_m . Since (X_1, \ldots, X_n) is independent of (X_{n+1}, \ldots, X_m) , we get that also $S_m - S_n$ is independent of $S_n \cdot \mathbf{1}_{A_n}$. So these random variables are also uncorrelated and

$$\mathbb{E}[(S_m - S_n) \cdot S_n \cdot \mathbf{1}_{A_n}] = \mathbb{E}[S_m - S_n] \cdot \mathbb{E}[S_n \cdot \mathbf{1}_{A_n}].$$

Since $\mathbb{E}[S_m - S_n] = \mathbb{E}[X_{n+1}] + \dots + \mathbb{E}[X_m] = 0$ we have that $\mathbb{E}[(S_m - S_n) \cdot S_n \cdot \mathbf{1}_{A_n}] = 0.$

(C) Write

$$\mathbb{E}[S_m^2 \cdot \mathbf{1}_{A_n}] = \mathbb{E}[(S_m - S_n + S_n)^2 \cdot \mathbf{1}_{A_n}]$$

= $\mathbb{E}[(S_m - S_n)^2 \cdot \mathbf{1}_{A_n}] + \mathbb{E}[S_n^2 \cdot \mathbf{1}_{A_n}] + 2 \mathbb{E}[(S_m - S_n) \cdot S_n \cdot \mathbf{1}_{A_n}]$
= $\mathbb{E}[(S_m - S_n)^2 \cdot \mathbf{1}_{A_n}] + \mathbb{E}[S_n^2 \cdot \mathbf{1}_{A_n}]$
 $\geq \mathbb{E}[S_n^2 \cdot \mathbf{1}_{A_n}].$

(D) We will show that

$$\{M_n \ge a\} = \biguplus_{k=1}^n A_k.$$

So $\mathbf{1}_{\{M_n \ge a\}}(\omega) = 1$ if and only if there is exactly one $1 \le k \le n$ such that $\mathbf{1}_{A_k}(\omega) = 1$. This implies that

$$\mathbf{1}_{\{M_n\geq a\}}=\sum_{k=1}^n\mathbf{1}_{A_k}.$$

If $M_n \ge a$, then, there exists $j \le n$ such that $|S_j| \ge a$. Thus, there exists a minimal such j; that is, there exists k for which $|S_j| < a$ if j < k and $|S_k| \ge a$. So we have shown that $\{M_n \ge a\} \subset \bigcup_{k=1}^n A_k$.

On the other hand, if A_k occurs then $|S_k| \ge a$, so also $M_n \ge a$. That is, $A_k \subset \{M_n \ge a\}$ for all $k \le n$, which implies that $\bigcup_{k=1}^n A_k \subset \{M_n \ge a\}$.

The two inclusions prove that

$$\{M_n \ge a\} = \bigcup_{k=1}^n A_k.$$

So we are left with showing that the union is disjoint.

If k > j we have

$$A_k \cap A_j \subset \{|S_j| < a\} \cap \{|S_j| \ge a\} = \emptyset.$$

So $(A_k)_k$ are pairwise disjoint.

(E) Since $S_n^2 \ge S_n^2 \mathbf{1}_{\{M_n \ge a\}}$, by linearity,

$$\mathbb{E}[S_n^2] \ge \mathbb{E}[S_n^2 \mathbf{1}_{\{M_n \ge a\}}] = \sum_{k=1}^n \mathbb{E}[S_n^2 \mathbf{1}_{A_k}] \ge \sum_{k=1}^n \mathbb{E}[S_k^2 \mathbf{1}_{A_k}],$$

where the last inequality is from (C).

Markov's inequality gives that for the non-negative random variable $S_n^2 \mathbf{1}_{A_n}$,

$$\mathbb{P}[S_n^2 \mathbf{1}_{A_n} \ge a^2] \le \mathbb{E}[S_n^2 \mathbf{1}_{A_n}] \cdot \frac{1}{a^2}.$$

Also, if $M_n \ge a$ then there exists $k \le n$ such that $\mathbf{1}_{A_k} = 1$ and $S_k^2 \ge a^2$. So,

$$\mathbb{P}[M_n \ge a] \le \mathbb{P}[\bigcup_{k=1}^n \left\{ S_k^2 \mathbf{1}_{A_k} \ge a^2 \right\}]$$

$$\le \sum_{k=1}^n \mathbb{P}[S_k^2 \mathbf{1}_{A_k} \ge a^2] \le \frac{1}{a^2} \cdot \sum_{k=1}^n \mathbb{E}[S_k^2 \mathbf{1}_{A_k}]$$

$$\le \frac{1}{a^2} \cdot \mathbb{E}[S_n^2].$$

Finally, since $\mathbb{E}[S_n] = \sum_{k=1}^n \mathbb{E}[X_k] = 0$ we have that $\mathbb{E}[S_n^2] = \operatorname{Var}[S_n]$ and so

$$\mathbb{P}[M_n \ge a] \le \frac{1}{a^2} \cdot \operatorname{Var}[S_n].$$

Solution Q2:

(A) Since X is absolutely continuous,

$$\mathbb{E}[(X^{+})^{2}] = \int_{-\infty}^{\infty} (t^{+})^{2} f_{X}(t) dt = \int_{0}^{\infty} t^{2} f_{X}(t) dt$$
$$= \int_{0}^{\infty} \int_{0}^{t} 2s ds f_{X}(t) dt = \int_{0}^{\infty} \int_{0}^{t} 2s f_{X}(t) ds dt$$
$$= \int_{0}^{\infty} \int_{s}^{\infty} 2s f_{X}(t) dt ds = \int_{0}^{\infty} 2s \mathbb{P}[X > s] ds.$$

(B) We have

$$\mathbb{E}[X] = \sum_{k=0}^{\infty} k \mathbb{P}[X=k] = \sum_{k=1}^{\infty} \sum_{m=0}^{k-1} \mathbb{P}[X=k]$$
$$= \sum_{m=0}^{\infty} \sum_{k=m+1}^{\infty} \mathbb{P}[X=k] = \sum_{m=0}^{\infty} \mathbb{P}[X \ge m+1]$$
$$= \sum_{m=0}^{\infty} \mathbb{P}[X > m].$$

Solution Q3:

(A) Note that for any k we have that $n_k, n_{k+1} \ge n_k$ so

$$\mathbb{P}[|Y_{k+1} - Y_k| > 2^{-k}] = \mathbb{P}[|X_{n_{k+1}} - X_{n_k}| > 2^{-k}] < 2^{-k}.$$

For any n we have

$$\mathbb{P}\left[\bigcup_{k \ge n} \left\{ |Y_k - Y_{k+1}| > 2^{-k} \right\} \right] \le \sum_{k \ge n} \mathbb{P}\left[|Y_{k+1} - Y_k| > 2^{-k}\right] \le \sum_{k=n}^{\infty} 2^{-k} = 2^{-n+1}.$$

For any *n* we have that $F \subset \bigcup_{k \ge n} \{ |Y_k - Y_{k+1}| > 2^{-k} \}$, so $\mathbb{P}[F] \le 2^{-n+1}$ for all *n*. Thus, $\mathbb{P}[F] = 0$.

(B) We know that $(Y_k)_k$ converges if and only if $(Y_k)_k$ is a Cauchy sequence. So we need to show that $(Y_k)_k$ is a Cauchy sequence a.s.

Now, if $\omega \notin F$ (for F as in (A)), then

$$\omega \in F^c = \bigcup_{n} \bigcap_{k \ge n} \{ |Y_k - Y_{k+1}| \le 2^{-k} \}.$$

That is, if $\omega \notin F$ then there exists n such that for all $k \ge n$ we have $|Y_{k+1}(\omega) - Y_k(\omega)| \le 2^{-k}$. In this case, for any $k \ge n$ and $m \ge 0$ we have that

$$|Y_{k+m}(\omega) - Y_k(\omega)| \le \sum_{j=0}^{m-1} |Y_{k+j+1}(\omega) - Y_{k+j}(\omega)| \le \sum_{j=0}^{\infty} |Y_{k+i+1}(\omega) - Y_{k+i}(\omega)| \le \sum_{j=0}^{\infty} 2^{-k-j} = 2^{-k+1}$$

That is, for any $\varepsilon > 0$ there exists n such that for all $k, m \ge n$ we have $|Y_k(\omega) - Y_m(\omega)| \le \varepsilon$. So for any $\omega \notin F$ we have that $(Y_k(\omega))_k$ forms a Cauchy sequence. Thus,

 $\mathbb{P}[(Y_k)_k \text{ is a Cauchy sequence }] \geq \mathbb{P}[F^c] = 1.$

(C) $A = \{Z = W\} = \{Z - W = 0\} = (Z - W)^{-1}(\{0\})$. Since Z, W are random variables, so is Z - W, and so $A = (Z - W)^{-1}(\{0\})$ is an event.

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Now, for any ω we have that the limit $\lim_k X_k(\omega)$ exists if and only if $Z(\omega) = W(\omega)$. If the limit exists, then it is equal to $Z(\omega) = W(\omega)$. Thus,

$$Y(\omega) = \left\{ \begin{array}{ll} Z(\omega) & \text{if } Z(\omega) = W(\omega) \\ 0 & \text{if } Z(\omega) \neq W(\omega) \end{array} \right\} = Z(\omega) \mathbf{1}_{\{Z=W\}}(\omega).$$

That is, $Y = Z \cdot \mathbf{1}_A$. Because A is an event, $\mathbf{1}_A$ is a random variable, and thus so is $Y = Z \cdot \mathbf{1}_A$ as a product of two random variables.