

Solutions

STAC63 Final 2022

Any results established in the class or in the Exercises, appropriately referenced, can be used as part of solving these questions.

1 (a) (5 marks) Suppose you need to generate values from a distribution with density

$$f(x) = \begin{cases} cx^3 & \text{for } 0 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

where c is the normalizing constant. Determine an algorithm to do this.

Can do this by inversion or rejection
for full marks (give lots of part marks)

Inversion $\int_0^x z^3 dz = \frac{z^4}{4} \Big|_0^x = u \Rightarrow x = \sqrt[4]{4u}$
 we have that $c = 1/4$ and the cdf is given
 $F(x) = \frac{x^4}{16}$ for $0 \leq x \leq 2$ and so for $u \in [0,1]$
 we have $F^{-1}(u) = 2u^{1/4}$. Therefore,
 generate $U_1 \sim U(0,1)$ and put $X = 2U_1^{1/4}$.

OR

Rejection

Let $g(x) = \frac{1}{2}$ for $x \in [0,2]$ which is the $U(0,2)$ density. Then $f(x) = \frac{x^3}{4} \leq 4g(x)$ for all $x \in (0,2)$.
 So we can use the rejection algorithm by generating $Y \sim U(0,2)$ and putting
 generating $U_1 \sim g$ by generating $U_1 \sim U(0,1)$ and putting
 $Y = 2U_1$, and independently generating $U_2 \sim U(0,1)$
 $X = Y$ whenever $U_2 g(Y) = \frac{U_2}{2} \geq \frac{Y^3}{4}$,
 and putting $X = Y$ whenever $U_2 g(Y) = \frac{U_2}{2} > \frac{Y^3}{4}$.
 If this inequality fails then we generate
 new values of (Y, U_2) until it is satisfied.

1 (b) (5 marks) Suppose you are asked to approximate $E(\exp(\cos(X)))$ where $X \sim f$ with f as in 1(a). Show how you would construct a Monte Carlo estimate of this quantity and also how you would assess the error in your estimate.

- Generate $x_{1,2,\dots,n} \sim f$ as described in 1(a). Then estimate $E(\exp(\cos X))$ by
- ② $I_n = \frac{1}{n} \sum_{i=1}^n \exp(\cos(x_i))$. The error in the estimate is assessed by quoting the interval $I_n \pm 3S_n$ where
- ③ $S_n = \sqrt{\frac{1}{n} \sum_{i=1}^n \exp(2\cos^2 x_i) - I_n^2}$.

2 (a) (5 marks) Suppose that X_1, \dots, X_n is a sample (*i.i.d.*) from a $U(0, 1)$ distribution and let $X_{(n)} = \max(X_1, \dots, X_n)$. Determine the distribution function of $X_{(n)}$ and then prove that $X_{(n)} \xrightarrow{d} 1$.

$$\begin{aligned}
 F_{X_{(n)}}(x) &= P(\max(X_1, \dots, X_n) \leq x) \\
 &= P(X_1 \leq x, \dots, X_n \leq x) \\
 &= \prod_{i=1}^n P(X_i \leq x) \quad \text{by independence} \\
 &= \begin{cases} 1, & x \leq 1 \\ 0, & x > 1 \end{cases} \quad \text{when } 0 \leq x \leq 1
 \end{aligned}$$

$$\text{Then } \lim_{n \rightarrow \infty} F_{X_{(n)}}(x) = \begin{cases} 0, & x \leq 1 \\ 1, & x > 1 \end{cases}$$

which implies $X_{(n)} \xrightarrow{d} 1$ since the limit is the cdf of the distribution degenerate at 1 and the limit holds at every x .

2 (b) (5 marks) Suppose that X_1, \dots, X_n is a sample as in 2(a). What constants a and b guarantee that

$$Y_n = \frac{\sqrt{n}}{b} \left(\frac{1}{n} \sum_{i=1}^n X_i - a \right) \xrightarrow{d} Z \sim N(0, 1)?$$

$\mathbb{E}(X_i) = 1/2$, $\mathbb{E}(X_i^2) = \int_0^1 x^2 dx = \frac{1}{3}$ so

(1) $\mathbb{E}(X_i) = \frac{1}{2}$, $\mathbb{E}(X_i^2) = \frac{1}{3}$. Therefore

$$\text{Var}(X_i) = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}$$

(2) by the CLT $a = \frac{1}{2}$ and $b = \sqrt{\frac{1}{12}}$
guarantees that this limit holds.

2 (c) (5 marks) Suppose that Y_n is as in 2(b) and $W_n = \exp(Y_n)$. Determine an asymptotic distribution for W_n that can be used to approximate probabilities for W_n .

By the delta theorem we have that

$$\frac{\ln}{b} (\exp(Y_n) - \exp(0)) \xrightarrow{D} \left(\frac{\partial \exp(y)}{\partial y} \Big|_{y=0} \right)^{-1}$$

(3) where $z \sim N(0, 1)$ and $\frac{\partial \exp(y)}{\partial y} \Big|_{y=0} = e^0 = 1$

(1) Therefore $\frac{\ln}{b} (\exp(Y_n) - \exp(0)) \approx N(0, \frac{1}{b})$

which implies we can approximate

$$P(W_n \leq w) = P\left(\frac{\ln}{b} (W_n - \exp(0)) \leq \frac{\ln}{b} (w - e^0)\right)$$

$$\begin{aligned} &= P\left(e^{-\frac{e^0}{b}} \frac{\ln}{b} (W_n - \exp(0)) + e^0 \frac{\ln}{b} (w - e^0)\right) \\ &\approx \Phi\left(-\frac{\ln}{b} (w - e^0)\right). \end{aligned}$$

3 (a) (5 marks) Suppose the state space \mathcal{S} for a stochastic process is finite. Is it possible to have $\lim_{n \rightarrow \infty} p_{ij}^{(n)} = 0$ for all $i, j \in \mathcal{S}$. Justify your conclusion.

No it is not possible because

(2) $1 = \sum_{j \in \mathcal{S}} p_{ij}^{(n)}$ For every i and n .

So if $\lim_{n \rightarrow \infty} p_{ij}^{(n)} = 0$ for all $i, j \in \mathcal{S}$ we would

(3) have $1 = \lim_{n \rightarrow \infty} \sum_{j \in \mathcal{S}} p_{ij}^{(n)} = 0 \times$.

3 (b) (5 marks) Suppose a process $\{X_n : n \in \mathbb{N}_0\}$ has state space $\mathcal{S} = \mathbb{N}$ and $X_0 = 1, X_i = X_{i-1} + 1$ for $i \geq 1$. Is this a Markov chain and, if it is, what is the transition probability matrix P ?

When $X_0 = 1, X_1 = 1, \dots, X_n = n$ we have

that $P(X_n = i | X_0 = 1, \dots, X_{n-1} = n-1)$

$$\textcircled{3} \quad = \begin{cases} 1 & \text{when } i = n \\ 0 & \text{otherwise} \end{cases}$$

$$= P(X_n = i | X_{n-1} = n-1)$$

and so this is a Markov chain with

$$P = \begin{pmatrix} 0 & 1 & 0 & 0 & \cdots \\ 0 & 0 & 1 & 0 & \cdots \\ 0 & 0 & 0 & 1 & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

3 (c) (5 marks) Consider a Markov chain with state space $S = \{1, 2, 3\}$ and transition probability matrix

$$P = \begin{pmatrix} 1/6 & 1/3 & 1/2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

Prove that $p_{12}^{(n)} \geq 1/3$ for all $n \geq 1$ (Hint: induction) and so conclude $\sum_{n=1}^{\infty} p_{12}^{(n)} = \infty$. Does this imply $f_{12} = 1$?

$P_{12}^{(n)} = 1/3$ and assume $P_{12}^{(n-1)} > 1/3$. Then

Since $P^n = P(P_{12}^{n-1}) = P \begin{pmatrix} P_{11}^{(n-1)} & P_{12}^{(n-1)} & P_{13}^{(n-1)} \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$

(3) we have that $P_{12}^{(n)} = \frac{1}{6} P_{12}^{(n-1)} + \frac{1}{3} > \frac{1}{3}$.

The fact that $\sum_{n=1}^{\infty} P_{12}^{(n)} = \infty$ does not

(2) imply that $f_{12} = 1$ since the chain is not irreducible.

3 (d) (5 marks) Consider a Markov chain with state space $S = \{1, 2, 3\}$ and transition probability matrix

$$P = \begin{pmatrix} 1/6 & 1/3 & 1/2 \\ 1/6 & 2/3 & 1/6 \\ 2/3 & 0 & 1/3 \end{pmatrix}.$$

Determine a stationary distribution for this chain. Is the chain reversible with respect to this stationary distribution?

(3) This transition probability matrix is doubly stochastic. Therefore the uniform distribution $\pi_i = 1/3$ for $i \in \{1, 2, 3\}$ is a stationary distribution.

(2) The chain is not reversible wrt π because $\pi_1 P_{12} = \frac{1}{3} \cdot \frac{1}{3} \neq \pi_2 P_{21} = \frac{1}{3} \cdot \frac{1}{6}$.

3(e) (5 marks) For the Markov chain in 3(d) determine $\lim_{n \rightarrow \infty} p_{ij}^{(n)}$ and fully justify this.

- ② The chain is irreducible and since $p_{ii}^{(n)} > 0$ for every i the chain is aperiodic.
- ③ Therefore by the Markov Chain Convergence Theorem we have that $\lim_{n \rightarrow \infty} p_{ij}^{(n)} = \pi_j = \frac{1}{3}$.

4 (a) (5 marks) Suppose (Ω, \mathcal{A}, P) is a probability model with X and Y random variables. Using the general definition of conditional expectation prove that $E(XY | \mathcal{A}_X) = XY$ (Hint: use the uniqueness w.p. 1 of the conditional expectation).

By the definition of conditional expectation we have $\mathbb{E}(H(X,Y)) \stackrel{(1)}{=} \mathbb{E}(H(X)\mathbb{E}(Y|\mathcal{A}_X))$ for any function $H(x,y)$. But since X and Y are stats. r.v.s,

- (1) we have that $\mathbb{E}(H(X,Y)) = \mathbb{E}(H(X))\mathbb{E}(Y)$ and so we note that $\mathbb{E}(Y|\mathcal{A}_X) = Y$ satisfies equation (1). Therefore, by the uniqueness of $\mathbb{E}(Y|\mathcal{A}_X)$ we must have that $\mathbb{E}(Y|\mathcal{A}_X) = Y$.
- (2) so $\mathbb{E}(XY|\mathcal{A}_X) = X\mathbb{E}(Y|\mathcal{A}_X) = XY$.

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4 (b) (4 marks) Suppose $\{M_n : n \in \mathbb{N}_0\}$ is a martingale with respect to the stochastic process $\{X_n : n \in \mathbb{N}_0\}$, namely, M_n is a function of (X_0, X_1, \dots, X_n) , and also $\{B_n : n \in \mathbb{N}_0\}$ is a stochastic process where B_n is a function of $(X_0, X_1, \dots, X_{n-1})$ s.t. $P(|B_n| < C) = 1$ for every n . If $Y_0 = 0$ and

$$Y_n = \sum_{i=1}^n B_i(M_i - M_{i-1}),$$

then prove that $\{Y_n : n \in \mathbb{N}_0\}$ is a martingale with respect to $\{X_n : n \in \mathbb{N}_0\}$.

$$\text{We have } |Y_n| \leq \sum_{i=1}^n |B_i|(|M_i| + |M_{i-1}|)$$

$$\leq \sum_{i=1}^n C(|M_i| + |M_{i-1}|) \text{ so}$$

$$\textcircled{5} \quad \mathbb{E}|Y_n| \leq \sum_{i=1}^n C(\mathbb{E}|M_i| + \mathbb{E}|M_{i-1}|) < \infty$$

Since $\{M_n : n \in \mathbb{N}_0\}$ is a martingale.

$$\text{Now } Y_{n+1} = Y_n + B_{n+1}(M_{n+1} - M_n) \text{ so}$$

$$\mathbb{E}(Y_{n+1} | X_0, \dots, X_n) = \mathbb{E}(Y_n | X_0, \dots, X_n)$$

$$+ \mathbb{E}(B_{n+1}(M_{n+1} - M_n) | X_0, \dots, X_n)$$

$$\textcircled{6} \quad = Y_n + B_{n+1}(\mathbb{E}(M_{n+1} | X_0, \dots, X_n) - \mathbb{E}(M_n | X_0, \dots, X_n))$$

Since Y_n, B_{n+1} are functions of X_0, \dots, X_n

$$= Y_n + B_{n+1}(M_n - M_n) \text{ since } \{M_n : n \in \mathbb{N}_0\} \text{ is a martingale}$$

$$= Y_n$$

$\therefore \{Y_n : n \in \mathbb{N}_0\}$ is a martingale.

4 (c) (5 marks) Suppose for the processes defined in 4 (b) that T is a stopping time for the process $\{Y_n : n \in \mathbb{N}_0\}$. Establish that T is also a stopping time for the process $\{X_n : n \in \mathbb{N}_0\}$.

We have that $\sum_{T=n}^{\infty} \in \mathcal{A}_{Y_0, \dots, Y_n}$
but $\mathcal{A}_{Y_0, \dots, Y_n} \subseteq \mathcal{A}_{X_0, \dots, X_n}$ which
proves T is also a stopping time
for $\{X_n : n \in \mathbb{N}_0\}$.

4. (d) (5 marks) For the process $\{Y_n : n \in \mathbb{N}_0\}$ with stopping time T prove that the process $\{Y_{\min(n,T)} : n \in \mathbb{N}_0\}$ is also a martingale. (Hint: $Y_{\min(n+1,T)} = I_{\{T \leq n\}} Y_{\min(n,T)} + I_{\{T > n\}} Y_{n+1}$.)

Using the hint

$$\begin{aligned}
 & \mathbb{E}(Y_{\min(n+1,T)} | X_0, \dots, X_n) \\
 &= \mathbb{E}(I_{\{T \leq n\}} Y_{\min(n,T)} | X_0, \dots, X_n) + \mathbb{E}(I_{\{T > n\}} Y_{n+1} | X_0, \dots, X_n) \\
 &= I_{\{T \leq n\}} Y_{\min(n,T)} + I_{\{T > n\}} \mathbb{E}(Y_{n+1} | X_0, \dots, X_n) \\
 &\quad \text{since } I_{\{T \leq n\}}, Y_{\min(n,T)}, I_{\{T > n\}} = 1 - I_{\{T \leq n\}} \\
 &\quad \text{are all functions of } X_0, \dots, X_n \\
 &= I_{\{T \leq n\}} Y_{\min(n,T)} + I_{\{T > n\}} Y_n \text{ since } \{Y_n\}_{n \in \mathbb{N}_0} \\
 &\quad \text{is a martingale} \\
 &= Y_{\min(n,T)} \text{ in both cases where } T \leq n \text{ or} \\
 &\quad T > n.
 \end{aligned}$$

5. (a) (5 marks) If $\{B_t : t \geq 0\}$ is a Brownian motion, then determine $E((B_2 - B_{1/2})(B_2 - B_1)^2)$.

$$\begin{aligned}
 & E((B_2 - B_{1/2})(B_2 - B_1)^2) \\
 &= E((B_2 - B_1 + B_1 - B_{1/2})(B_2 - B_1)^2) \\
 &= E((B_2 - B_1)^3) + E((B_1 - B_{1/2})(B_2 - B_1)^2) \\
 &= E(Z^3) + E(B_1 - B_{1/2}) E((B_2 - B_1)^2) \\
 &\quad \text{since } Z = B_2 - B_1 \sim N(0, 1) \text{ independent} \\
 &\quad \text{of } B_1 - B_{1/2} \sim N(0, 1/2) \\
 &= 0 + 0 = 0
 \end{aligned}$$

Since $E(Z^3) = 0$, $E(B_1 - B_{1/2}) = 0$.

5. (b) (5 marks) If $\{B_t : t \geq 0\}$ is a Brownian motion and $M_t = \max\{B_s : 0 \leq s \leq t\}$ then determine $E(M_t)$.

We have from Prop VI.8 that

$$\textcircled{3} M_t \text{ has density } f_{M_t}(m) = \sqrt{\frac{2}{\pi t}} \exp\left\{-\frac{m^2}{2t}\right\}.$$

Then $E(M_t) = \int_{-\infty}^{\infty} m f_{M_t}(m) dm = \int_{-\infty}^{\infty} m \sqrt{\frac{2}{\pi t}} \exp\left\{-\frac{m^2}{2t}\right\} dm$

$$\textcircled{3} = \sqrt{\frac{2}{\pi t}} \left(-t \exp\left\{-\frac{m^2}{2t}\right\} \Big|_0^\infty \right)$$

$$= \sqrt{\frac{2}{\pi t}} t = \sqrt{\frac{2t}{\pi}}.$$

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6. (a) (6 marks) Suppose messages arrive according to a Poisson process with intensity λ and a message is labelled as urgent with probability $1/3$. If M_t is the total number of messages in $[0, t]$ and U_t is the number of messages labelled urgent in $[0, t]$, then compute $P(U_2 = 3)$, $P(U_2 = 3, M_5 = 4)$.

We know that $\{U_{t \wedge T}, t \geq 0\}$ is a Poisson process with intensity $\lambda/3$ so

$$\textcircled{5} \quad P(U_2 = 3) = \frac{(2\lambda/3)^3}{3!} e^{-2\lambda/3}. \quad \text{Also}$$

$M_5 - U_5 = \# \text{ of messages labelled not urgent}$
 is a Poisson process of intensity $2\lambda/3$
 and it is independent of $\{U_{t \wedge T}, t \geq 0\}$.

$$\text{Now } \{U_2 = 3, M_5 = 4\} \subset \{U_2 = 3, U_5 - U_2 = 1, M_5 - U_5 = 0\} \\ \cup \{U_2 = 3, U_5 - U_2 = 0, M_5 - U_5 = 1\}$$

$$\textcircled{5} \quad P(\cdot | \{U_2 = 3, M_5 = 4\})$$

Therefore, using independence of increments and the
 processes, and the disjointness of $\textcircled{1}$ and $\textcircled{2}$,

$$P(U_2 = 3, M_5 = 4) = P(\textcircled{1}) + P(\textcircled{2}) = \frac{(2\lambda/3)^3}{3!} e^{-2\lambda/3} \frac{(3\lambda/3)^1}{1!} e^{-3\lambda/3} e^{-5(2\lambda/3)} \\ + \left(\frac{2\lambda/3}{5!} \right)^3 e^{-2\lambda/3} e^{-3\lambda/3} \frac{(5(2\lambda/3))^1}{1!} e^{-5(2\lambda/3)}$$

6. (b) (5 marks) Suppose messages arrive from two independent sources according to a Poisson processes with intensities λ_1 and λ_2 and U_t is the number of messages labelled urgent (with probability 1/3) in $[0, t]$. Compute $P(U_2 = 3)$.

We have that the calls arrive according to a Poisson process with intensity $\lambda_1 + \lambda_2$.

⑥ Therefore $\{U_{t+2t/3}, 0\}$ is a Poisson process with intensity $(\lambda_1 + \lambda_2)/3$ which implies

$$P(U_2=3) = \frac{(z^{(\lambda_1+\lambda_2)/3})^3}{3!} e^{-z^{(\lambda_1+\lambda_2)/3}}$$