B - Book work

E - New example

T - New theory

$$f_{x}(x) = \frac{1}{2\pi}$$
 $x \in [-\pi, \pi]$

$$i)$$
 $y = g(x) = x^3$

$$x = y^{1/3}$$

$$y \in [-\pi^3, \pi^3]$$

$$g(x) = 3 x^2 = 3 y^{2/3}$$

$$f_{y}(y) = \frac{1}{|g'(x)|} f_{x}(x) = \frac{1}{3 y^{2/3}} f_{x}(x)$$

$$= \frac{1}{6\pi y^{2/3}}$$

$$y \in [-\pi^3, \pi^3]$$

ii)
$$y = g(x) = x^4$$
 $x_1 = y^{1/4}$

$$g'(x) = 4x^3$$

$$f_{\gamma}(y) = \frac{1}{|g'(x_i)|} f_{\kappa}(x_i) + \frac{1}{|g'(x_i)|} f_{\kappa}(x_i)$$

$$= \frac{1}{4y^{3/4}} \left[f_{x}(x_{l}) + f_{x}(x_{l}) \right]$$

$$= \frac{1}{4\pi y^{3/4}}$$

iii)
$$y = g(x) = \sin(x)$$

$$\chi_1 = \sin^4(y)$$
 $\chi_2 = \pi - \sin^4(y)$

$$g'(x) = \cos x$$
$$= \sqrt{1 - y^2}$$

$$f_{y}(y) = \frac{1}{|g'(x_1)|} f_{x}(x_1) + \frac{1}{|g'(x_2)|} f_{x}(x_2)$$

$$= \frac{1}{\sqrt{1-y^2}} \left(\frac{1}{2\pi} + \frac{1}{2\pi} \right)$$
$$= \frac{1}{\sqrt{1-y^2}}$$

b) i) Let
$$X' = 2 \times 1$$

Then $f_{X'}(X') = \frac{1}{2} f_{X}(X) = \frac{1}{2} e^{-X'/2}$ $X' > 0$ [2E]
$$f_{Z}(z) = \int_{0}^{z} f_{X'}(z-y) f_{Y}(y) dy$$

$$= \int_{0}^{z} \frac{1}{2} e^{-(z-y)/2} e^{-y} dy$$

$$= \frac{1}{2} e^{-z/2} \int_{0}^{z} e^{-y/2} dy$$

$$= \frac{1}{2} e^{-z/2} \cdot 2 (1 - e^{-z/2})$$

$$= e^{-z/2} - e^{-z} \qquad z > 0$$
ii) $F_{Z}(z) = P\{\min_{x \in Z} (X, y) \le z\}$

$$= 1 - P\{\sum_{x \in Z} (X, y) \le z\}$$

$$= 1 - [1 - F_{X}(z)](1 - F_{Y}(z))$$

$$= F_{X}(z) + F_{Y}(z) - F_{X}(z) F_{Y}(z)$$

$$f_{Z}(z) = f_{X}(z) + f_{Y}(z) - f_{X}(z) F_{Y}(z)$$
We have
$$f_{X}(z) = f_{Y}(z) = e^{-z} \qquad z > 0$$
Thus,
$$f_{X}(z) = 2 e^{-z} - 2 e^{z} (1 - e^{-z})$$

$$D = 1$$

$$f_{z(z)} = 2e^{-z} - 2\bar{e}^{z}(1-e^{-z})$$

$$= 2e^{-2z}$$

$$= 2e^{-2z}$$

iii)
$$Z = \max(X, Y)$$

$$\begin{cases}
F_{Z}(Z) = P\{\max(X, Y) \in Z\} \\
= P(X \in Z, Y \subseteq Z\} \\
= F_{X}(Z) F_{Y}(Z)
\end{cases}$$

$$f_{Z}(Z) = f_{X}(Z) F_{Y}(Z) + f_{X}(Z) f_{Y}(Z)$$

$$GEI$$

 $= 2 e^{-2} (1 - e^{-2})$

2. a) Let X denote the average.

The joint density
$$f(X,c) = c^n e^{-cn(\bar{X}-X_0)}$$
[3E]

has maximum if

$$\frac{\partial f(X,c)}{\partial c} = 0 \implies \hat{c} = \frac{1}{\bar{\chi} - \chi_0} \qquad [2E]$$

obviously,
$$\overline{X} = 9$$
 in this problem. So
$$\hat{C} = \frac{1}{9-5} = \frac{1}{4}$$
 [3 E]

b) From the Wiener-Hopf equation,
$$C = R^{-1}r$$

$$\sigma^{2} = r_{0} - r^{T}R^{-1}r$$
[28]

i) When
$$n=1$$
, we have
$$R=1 \qquad \qquad [2E]$$

$$r=r_1=0.643$$

Thus, $C_1 = Y_1 = 0.643$ $C_2 = 1 - Y_1^2 = 1 - 0.643^2 = 0.587$

ii) When
$$n=2$$
, we have
$$R = \begin{bmatrix} 1 & 0.643 \\ 0.643 & 1 \end{bmatrix}$$

$$\Gamma = \begin{bmatrix} -0.055 \\ 0.643 \end{bmatrix}$$

Thus,

= 0.2/4

$$C = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = R^{-1} \Gamma$$

$$= \begin{bmatrix} 1 & 0.643 \end{bmatrix}^{-1} \qquad \begin{bmatrix} -0.055 \\ 0.643 \end{bmatrix}$$

$$= \begin{bmatrix} -0.797 \\ 1.154 \end{bmatrix}$$

$$C^2 = 1 - \begin{bmatrix} -0.055 & 0.643 \end{bmatrix} \begin{bmatrix} -0.797 \\ 1.154 \end{bmatrix}$$

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3. a) i) $X(n) = \cos(n U)$ E[X(n)] = E[Cos(nu)] = 0[2 E] E[Xin)] = E[cosicnu)] = E[\(\frac{1}{2} \left(1 + \cos \(\text{knu} \right) \right) \] = \(\frac{1}{2} \) E[x(m)x(n)] = E[cos(mu) cos(nu)] [3 E] $= E \left[\frac{1}{2} \left(\cos \left((m+n)u \right) + \cos \left(m-n \right) u \right) \right]$ = 0 if $m \neq n$ Therefore, X(n) is wide-sense Stationary. it) Here, the answer is not unique. For example, one may check E[Xcm) X(n) X(r)] = E[cos(mu) coscnu) coscrul)]

 $E[\chi(m) \chi(n) \chi(r)] = E[\cos(mu)\cos(nu)\cos(nu)\cos(nu)]$ $= \frac{1}{2}E[(\cos((m+n+r)u) + \cos(m+n-r)u))\cos(mu)]$ $= \frac{1}{4}E[\cos((m+n+r)u) + \cos((m-n-r)u)]$ $= \frac{1}{4}[\delta(m+n+r) + \delta(m+n-r))$ $= \frac{1}{4}[\delta(m-n+r) + \delta(m-n-r)]$

where $\delta(n) = 1$ if n = 0 $\delta(n) = 0$ if $n \neq 0$

Consider two cases (m, n, r) = (1, 2, 3), (2, 3, 4). [27] They take different values $\frac{1}{4}$ and 0. So it doesn't satisfy the definition of strict-sense

Stationarity (which would require the same values).

i) This is the same as the time when the third patient arrives.

$$E[T_3] = \frac{3}{\lambda} = \frac{3}{0.1} = 30$$
 minutes [3E]

li) This means that the number of patients arrived in the first hour is less than three

$$P(NCt) < 3) t = 60 \text{ minutes}$$

$$= P(NCt) = 0) + P(NCt) = 1) + P(NCt) = 2)$$

$$= e^{-60/10} + (\frac{60}{10})e^{-60/10} + \frac{1}{2}(\frac{60}{10})^{2}e^{-60/10}$$

$$= 25 \cdot e^{-6}$$

$$= 0.062$$
BEI

Recall Poisson:
$$P(Nt)=K)=e^{-\lambda t}\frac{(\lambda t)^{K}}{K!}$$
 $k=0,1/2,...$

(iii) Poisson process is memoryless. So this probability is given by

$$P(N(t_{1}) \ge 2) \cdot P(N(t_{2}-t_{1}) \le 2) \qquad t_{1} = 60 \text{ minutes}$$

$$= \left[1 - P(N(t_{1}) < 2)\right] P(N(t_{2}-t_{1}) \le 2)$$

$$= \left[1 - \left[P(N(t_{1}) = 0) + P(N(t_{1}) = 1)\right]\right] P(N(t_{2}-t_{1}) \le 2)$$

$$= \left[1 - \left[e^{-6} + 6e^{-6}\right]\right] \cdot 25 \cdot e^{-6} \qquad (from ii)) \quad \vec{3} = 1$$

$$= 0.06$$

4. a) i) For any sequence of states io, ii, ...

$$P(|Y_{r+1}| = i_{r+1}| | Y_r = i_r, | Y_{r-1}| = i_{r-1}, ..., | Y_0 = i_0))$$

$$= \frac{P(|Y_{r+1}| = i_{r+1}, | Y_r = i_r, ..., | Y_0 = i_0)}{P(|Y_r = i_r, | ..., | Y_0 = i_0)}$$

$$= \frac{\prod_{s=0}^{r} P_{is,i_{s+1}}(|N_{s+1}| - N_s)}{\prod_{s=0}^{r+1} P_{is,i_{s+1}}(|N_{s+1}| - N_s)}$$

$$= P_{ir,i_{s+1}}(|N_{r+1}| - N_r)$$

$$= P_{i_r,i_{r+1}} (n_{r+1} - n_r)$$
 [27]

ii) The transition matrix is given by
$$P^{2} = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{pmatrix} \quad [3E]$$

[2 T]

b) Let
$$S_{n+1} = S_n + Z_{n+1}$$
 where $P(Z_{n+1} = 1) = P$, $P(Z_{n+1} = -1) = F$.

Thus,
$$E[Y_{htt}|Y_{n},...,Y_{o}] = E[(\frac{2}{p})^{S_{n}t}|S_{n},S_{h^{-1}},...S_{o}]$$

$$= E[(\frac{2}{p})^{S_{n}+Z_{htt}}|S_{n}] \qquad Markov$$

$$= (\frac{2}{p})^{S_{n}}[\frac{2}{p}\cdot p + (\frac{2}{p})^{T}\cdot k]$$

$$= (\frac{2}{p})^{S_{n}}$$

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i.e., fyn) is a martingale.

$$E[Y_7] = E[Y_0] = (8/p)^i$$

Since Yo = i.

We also have

ELY₇] =
$$P_i \stackrel{(a)}{=} p^0 + (1 - P_i) \stackrel{(a)}{=} p^N$$

= $P_i + (1 - P_i) \stackrel{(a)}{=} p^N$

Thus

$$P_{i} = \frac{1 - (\frac{7}{2})^{N-i}}{1 - (\frac{7}{2})^{N}} \quad \text{if } \frac{7}{3} \neq 1 \quad [2B]$$

$$P_{i} = 1 - \frac{i}{N} \quad \text{if } p = 8 = \frac{1}{2}$$

c) From the
$$T_k = 1 + p T_{k+1} + g T_{k-1}$$
, we have

$$P\left(T_{i+1}-T_i\right)=g\left(T_i-T_{i+1}\right)-1$$

[2T]

Let Min = Tin - Ti, we obtain iteration

$$M_{iH} = \frac{g}{p} M_{i} - \frac{1}{p}$$

$$= \begin{cases} (\frac{g}{p})^{i} M_{i} - \frac{1}{p - g} & p \neq f \\ M_{i} - \frac{i}{p} & p = g \end{cases}$$

$$p = g$$

This gives

$$T_{i} = \sum_{k=0}^{i-1} M_{k+1}$$

$$= \begin{cases} (M_{i} + \frac{1}{p-g}) \sum_{k=0}^{i-1} (\frac{g}{p})^{k} - \frac{i}{p-g} \\ i M_{i} - \frac{i(i-i)}{2p} \end{cases} \qquad p \neq g \qquad [2T]$$

We yet need to determine M_r from initial conditions $T_0 = 0$ $T_N = 0$

Which gives

$$M_1 + \frac{1}{p-g} = \frac{N}{p-g} \cdot \frac{1-g/p}{1-(g/p)N}$$
 $p \neq g$

$$T_{i} = \frac{N}{P-8} \cdot \frac{1-8P}{1-(8P)^{N}} \cdot \frac{1-(8P)^{i}}{1-8P} - \frac{i}{P-8}$$

$$= \int_{P-8}^{N} \cdot \frac{(1-8P)^{i}}{(1-8P)^{N}} - \frac{i}{P-8}$$

$$p \neq 8$$

=i(N-i)

$$T_{N} = 0 = NM_{1} - \frac{N(N-1)}{2p} = 0$$

$$M_{1} = \frac{N-1}{2p}$$

$$T_{i} = \frac{i(N-1)}{2p} - \frac{i(i-1)}{2p}$$

$$= \frac{i(N-i)}{2p}$$

$$P = \frac{1}{2}$$