

$$\textcircled{1} S_X(f) = (1 - |f|/2) \mathcal{I}_{[-2,2]}(f)$$

$$\begin{aligned} S_Y(f) &= |H(f)|^2 S_X(f) \\ &= \frac{1}{1 - |f|/2} (1 - |f|/2) \mathcal{I}_{[-1,1]}(f) \\ &= \mathcal{I}_{[-1,1]}(f) \end{aligned}$$

$$\therefore R_Y(\tau) = 2 \frac{\sin(2\pi\tau)}{2\pi\tau} = \sqrt{2} \sqrt{R_X(f)}$$

$$\begin{aligned} \textcircled{2} (a) E[|X_n - X|^2] &= E[X_n^2] - 2E[X_n X] + E[X^2] \\ &= 1 - 2(1 - 1/(2n^2)) + 1 \\ &= \frac{1}{n^2} \rightarrow 0. \end{aligned}$$

$$\therefore X_n \xrightarrow{L^2} X.$$

$$\begin{aligned} (b) \sum_{n=1}^{\infty} \mathcal{P}(|X_n - X| \geq \varepsilon) &\leq \sum_{n=1}^{\infty} \frac{E[|X_n - X|^2]}{\varepsilon^2}, \text{ by Markov's ineq.} \\ &= \frac{1}{\varepsilon^2} \sum_{n=1}^{\infty} \frac{1}{n^2} < \infty. \end{aligned}$$

$$\therefore X_n \rightarrow X \text{ a.s.}$$

$$\begin{aligned} \textcircled{3} E[X_{t_1} Z_{t_2}] &= E[X_{t_1} \text{sgn}(Y_{t_2})] \\ &= E[E[X_{t_1} \text{sgn}(Y_{t_2}) | Y_{t_2}]] \\ &= E[E[X_{t_1} | Y_{t_2}] \text{sgn}(Y_{t_2})] \\ &= E[A Y_{t_2} \text{sgn}(Y_{t_2})], \end{aligned}$$

$$\begin{aligned} A C_{Y_{t_2}} &= C_{X_{t_1} Y_{t_2}} \\ A &= \frac{R_{XY}(t_1 - t_2)}{R_Y(0)} \end{aligned}$$

Next,

$$\begin{aligned} E[Y_{t_2} \text{sgn}(Y_{t_2})] &= \int y \text{sgn}(y) f_Y(y) dy \\ &= \int |y| f(y) dy \end{aligned}$$

$$\begin{aligned}
&= 2 \int_0^{\infty} y \frac{e^{-(y/\sigma)^2/2}}{\sqrt{2\pi} \sigma} dy \quad \sigma^2 = R_y(0) \\
&= \frac{2\sigma}{\sqrt{2\pi}} \cdot \left( -e^{-(y/\sigma)^2/2} \right) \Big|_0^{\infty} \\
&= \frac{2\sigma}{\sqrt{2\pi}} = \frac{2R_y(0)^{1/2}}{\sqrt{2\pi}}.
\end{aligned}$$

So,

$$R_{xz}(t_1, t_2) = \sqrt{\frac{2}{\pi R_y(0)}} R_{xy}(t_1 - t_2).$$

(4) As we know,  $X_n \xrightarrow{L^2} X \Rightarrow E[X_n^2] \rightarrow E[X^2]$  and  $E[X_n] \rightarrow E[X]$ .

$$\therefore \frac{p_n}{1-p_n} = E[X_n] \rightarrow E[X] \quad \boxed{1}$$

and

$$\frac{p_n}{(1-p_n)^2} + \left[ \frac{p_n}{1-p_n} \right]^2 = E[X_n^2] \rightarrow E[X^2].$$

$$\text{By } \boxed{1}, \left( \frac{p_n}{1-p_n} \right)^2 \rightarrow E[X]^2 \quad \boxed{2}$$

$$\text{+ So } \frac{p_n}{(1-p_n)^2} \rightarrow \text{var}(X). \quad \boxed{3}$$

$$\text{Then } \boxed{2} + \boxed{3} \Rightarrow \boxed{p_n = \frac{\left( \frac{p_n}{1-p_n} \right)^2}{\frac{p_n}{(1-p_n)^2}} \rightarrow \frac{E[X]^2}{\text{var}(X)}}.$$

We now have

$$G_{X_n}(z) = \frac{1-p_n}{1-zp_n} \rightarrow \frac{1-p}{1-zp}, \quad p := \frac{E[X]^2}{\text{var}(X)}.$$

$$\therefore X \sim \text{geo}_0(p).$$

⑤ Let's identify  $\{t: M_t \geq n\} = \{t: N_{v(t)} \geq n\} = \{t: T_n \leq v(t)\}$ ,  
 where  $T_n$  is the  $n$ th occurrence time of  $\{N_t, t \geq 0\}$ . So,

$$\begin{aligned} \{t: M_t \geq n\} &= \{t: v^{-1}(T_n) \leq t\} \\ &= [v^{-1}(T_n), \infty). \end{aligned}$$

$$\therefore S_n = \min [v^{-1}(T_n), \infty) = v^{-1}(T_n).$$

Next,

$$\begin{aligned} F_{S_n}(s) &= \mathcal{P}(S_n \leq s) = \mathcal{P}(v^{-1}(T_n) \leq s) = \mathcal{P}(T_n \leq v(s)) \\ &= F_{T_n}(v(s)) \end{aligned}$$

$\therefore$

$$f_{S_n}(s) = f_{T_n}(v(s))v'(s).$$

Since  $T_n \sim \text{Erlang}(n, 1)$ ,

$$f_{S_n}(s) = \frac{v(s)^{n-1} e^{-v(s)}}{(n-1)!} v'(s)$$