

# **Probability and Random Processes (15B11MA301)**

## **Lecture-32**



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## Semi-Random telegraph signal process

**Semi-Random Telegraph Signal Process:** A random process  $X(t)$  defined as

$$X(t) = (-1)^{N(t)}$$

Where  $N(t)$  represents the number of occurrences of a specified event in  $(0, t)$ .  $N(t)$  follows Poisson process.

i.e.,

$$P[N(t) = r] = \frac{e^{-\lambda t} (\lambda t)^r}{r!}, r = 0, 1, 2, \dots$$

$$P[X(t)=1] = P[N(t) \text{ is even}]$$

$$= P[N(t) = 0] + P[N(t) = 2] + P[N(t) = 4] + \dots\dots\dots$$

$$= e^{-\lambda t} \left[ 1 + \frac{(\lambda t)^2}{2!} + \frac{(\lambda t)^4}{4!} + \dots\dots\dots \right]$$

$$= e^{-\lambda t} \text{Cosh } \lambda t$$

$$P[X(t)=-1] = P[N(t) \text{ is odd}]$$

$$= P[N(t) = 1] + P[N(t) = 3] + \dots\dots\dots$$

$$= e^{-\lambda t} \left[ \frac{\lambda t}{1!} + \frac{(\lambda t)^3}{3!} + \dots\dots\dots \right]$$

$$= e^{-\lambda t} \sinh \lambda t$$

Then means  $E[X(t)] = 1 \times P[X(t)=1] - 1 \times P[X(t)= -1]$

$$= e^{-\lambda t} \text{Cosh} \lambda t - e^{-\lambda t} \sinh \lambda t$$

$$= e^{-\lambda t} [\text{Cosh} \lambda t - \text{Sinh} \lambda t]$$

$$= e^{-\lambda t} . e^{-\lambda t}$$

$$E[X(t)] = e^{-2\lambda t}$$

Autocorrelation:

$$R(t_1, t_2) = E[X(t_1).X(t_2)]$$

$$X(t_1).X(t_2) \text{ may be } \begin{matrix} +1 \\ -1 \end{matrix} \left[ \begin{array}{l} X(t_1) = 1, X(t_2) = 1 \\ X(t_1) = -1, X(t_2) = -1 \\ X(t_1) = 1, X(t_2) = -1 \\ X(t_1) = -1, X(t_2) = 1 \end{array} \right]$$

$$\begin{aligned}
 R(t_1, t_2) &= 1 \times P[X(t_1).X(t_2) = 1] - 1 \times P[X(t_1).X(t_2) = -1] \\
 &= P[X(t_1) = 1, X(t_2) = 1] + P[X(t_1) = -1, X(t_2) = -1] \\
 &\quad - P[X(t_1) = -1, X(t_2) = 1] - P[X(t_1) = 1, X(t_2) = -1] \dots \dots (1)
 \end{aligned}$$

$$\begin{aligned}
 P[X(t_1) = 1, X(t_2) = 1] &= P[X(t_1) = 1 | X(t_2) = 1] \times P[X(t_2) = 1] \\
 &\quad (t_2 \leq t_1)
 \end{aligned}$$

$$\begin{aligned}
 &= P[\text{an even no. of occurrences of event in } (t_1 - \\
 &\quad t_2)] \times P[X(t_2) = 1]
 \end{aligned}$$

$$= e^{-\lambda \tau} \text{Cosh} \lambda \tau \times e^{-\lambda t_2} \text{Cosh} \lambda t_2 \quad \text{where } \tau = t_1 - t_2$$

$$P[X(t_1) = 1, X(t_2) = 1] = e^{-\lambda \tau} \text{Cosh} \lambda \tau . e^{-\lambda t_2} \text{Cosh} \lambda t_2$$

$$P[X(t_1) = 1, X(t_2) = 1] = e^{-\lambda\tau} \text{Cosh}\lambda\tau \cdot e^{-\lambda t_2} \text{Cosh}\lambda t_2$$

$$P[X(t_1) = -1, X(t_2) = -1] = P[X(t_1) = -1 | X(t_2) = -1] \times P[X(t_2) = -1]$$

$$= P[\text{an even no. of occurrences of event in } (t_1 - t_2)] \cdot P[X(t_2) = 1]$$

$$= e^{-\lambda\tau} \text{Cosh}\lambda\tau \times e^{-\lambda t_2} \text{Sinh}\lambda t_2$$

$$P[X(t_1) = -1, X(t_2) = -1] = e^{-\lambda\tau} \text{Cosh}\lambda\tau \cdot e^{-\lambda t_2} \text{Sinh}\lambda t_2$$

Similarly,

$$P[X(t_1) = 1, X(t_2) = -1] = e^{-\lambda\tau} \text{Sinh}\lambda\tau \cdot e^{-\lambda t_2} \text{Sinh}\lambda t_2$$

and

$$P[X(t_1) = -1, X(t_2) = 1] = e^{-\lambda\tau} \text{Sinh}\lambda\tau \cdot e^{-\lambda t_2} \text{Cosh}\lambda t_2$$



Then by equation 1,

$$\begin{aligned} R(t_1, t_2) &= e^{-\lambda\tau} \text{Cosh}\lambda\tau \times e^{-\lambda t_2} \text{Sinh}\lambda t_2 \\ &+ e^{-\lambda\tau} \text{Cosh}\lambda\tau . e^{-\lambda t_2} \text{Sinh}\lambda t_2 - e^{-\lambda\tau} \text{Sinh}\lambda\tau . e^{-\lambda t_2} \text{Cosh}\lambda t_2 \\ &- e^{-\lambda\tau} \text{Sinh}\lambda\tau . e^{-\lambda t_2} \text{Sinh}\lambda t_2 \end{aligned}$$

$$\begin{aligned} R(t_1, t_2) &= e^{-\lambda\tau} \text{Cosh}\lambda\tau \times e^{-\lambda t_2} . e^{-\lambda t_2} - e^{-\lambda\tau} \text{Sinh}\lambda\tau e^{-\lambda t_2} e^{\lambda t_2} \\ &= e^{-\lambda\tau} [\text{Cosh}\lambda\tau - \sinh \lambda\tau] = e^{-\lambda\tau} . e^{-\lambda\tau} = e^{-2\lambda\tau} \end{aligned}$$

$$R(t_1, t_2) = e^{-2\lambda(t_1 - t_2)} = e^{-2\lambda\tau}$$

In semi random telegraph signal process,

Mean  $\neq$  constant &  $R(t_1, t_2) = f(t_1 - t_2)$

$= \{X(t)\}$  is evolutionary process

Imp Formulas:

$$\left\{ \begin{array}{l} \sinh x = \frac{e^x - e^{-x}}{2} \\ \cosh x = \frac{e^x + e^{-x}}{2} \\ \frac{\sinh x + \cosh x}{2} = e^x \\ \frac{\cosh x - \sinh x}{2} = e^{-x} \end{array} \right.$$

## Random telegraph Signal process:

A random process  $\{Y(t)\}$  defined as:

$$Y(t) = \alpha X(t)$$

Where  $\alpha$  is a random variable independent of  $X(t)$  defined as:

$$\alpha = \{1, -1\} \text{ with } P[\alpha = 1] = \frac{1}{2} \text{ and } P[\alpha = -1] = \frac{1}{2}.$$

and  $X(t)$  is semi random telegraph signal process.

**MEAN:**

$$E[Y(t)] = E[\alpha.X(t)]$$

$$= E[\alpha]. E[X(t)]$$

$$\left\{ \begin{aligned} E[\alpha] &= 1 \times P[\alpha = 1] - 1 \times P[\alpha = -1] \\ &= 1 \times \frac{1}{2} - 1 \times \frac{1}{2} = \frac{1}{2} - \frac{1}{2} = 0 \end{aligned} \right\}$$

$$E[Y(t)] = E[\alpha]. E[X(t)]$$

$$= 0 \times e^{-2\lambda t} = 0$$

$$E[Y(t)] = 0$$

$$E[Y(t_1).Y(t_2)] = R_y(t_1, t_2)$$

$$= R_y(t_1, t_2) = E[\alpha X(t_1).\alpha X(t_2)]$$

$$= E[\alpha^2 X(t_1).X(t_2)]$$

$$= E[\alpha^2].E[X(t_1).X(t_2)]$$

$$= E[\alpha^2].R_x(t_1, t_2)$$

$$= 1.e^{-2\lambda\tau}$$

$$= e^{-2\lambda\tau}$$

$$R_y(t_1, t_2) = e^{-2\lambda(t_1-t_2)} = e^{-2\lambda\tau}$$

$$\left\{ \begin{array}{l} E[\alpha^2] = (1)^2 \times \frac{1}{2} + (-1)^2 \times \frac{1}{2} \\ \qquad \qquad \qquad = \frac{1}{2} + \frac{1}{2} = 1 \end{array} \right\}$$

In random telegraph signal process

$$\text{Mean} = E\{Y(t)\} = 0$$

$$\& R(t_1, t_2) = e^{-2\lambda\tau}$$

That implies  $\{Y(t)\}$  is WSS process.

## References

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