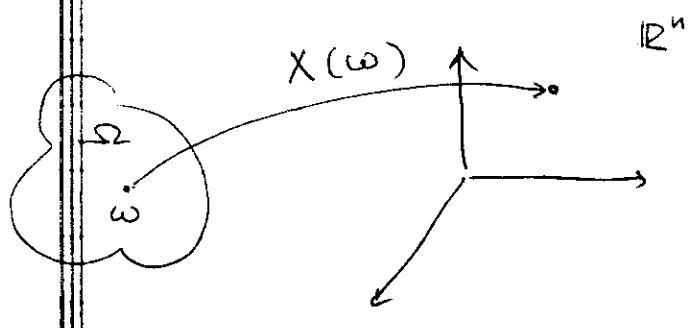


Summary : To specify a random experiment  
 $(\Omega, \mathcal{F}, \mathcal{P})$  ~ probability space.  
 A RV  $X(\omega)$  is a function that maps  
 elements of  $\Omega$  onto the real line



A random vector  
 $\underline{X}(\omega)$  is a mapping of  
 elements of  $\Omega$  onto  $\mathbb{R}^n$



Events :  
 $A = \{ \omega : X(\omega) \leq \alpha \} \in \mathcal{F}$   
 for all  $\alpha \in \mathbb{R}^n$

Complete Characterization:

$$P_x(\alpha) = \Pr[\omega : X(\omega) \leq \alpha]$$

for all  $\alpha$

Vectors: Product type events.

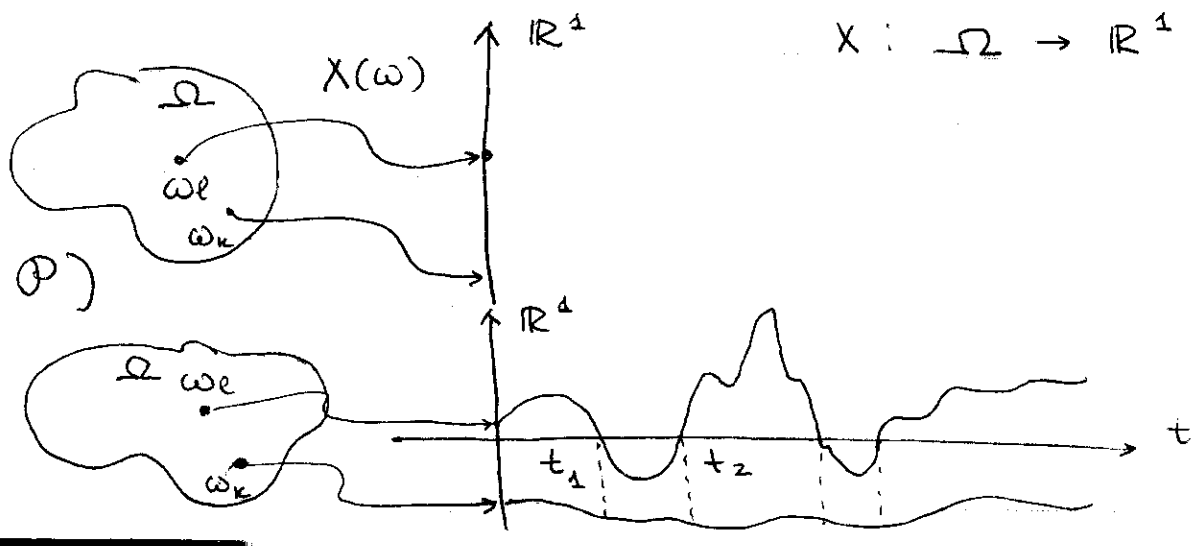
Partial Characterizations:

$$m_\alpha, \sigma_\alpha^2, \text{ moments } E[X^k], k = 0, 1, 2, \dots$$

Random Processes

R.V.

$(\Omega, \mathcal{F}, \mathcal{P})$



$$X : \Omega \rightarrow \mathbb{R}^1$$

(2)

$$X: \underbrace{\Omega \times T}_{\text{domain}} \rightarrow \underbrace{S}_{\substack{\text{range} \\ \text{space}}} \quad \begin{matrix} \mathbb{R}^1 \\ \mathbb{R}^2 \\ \vdots \end{matrix}$$

ex (index set)

$$T = \{t_1, t_2, t_3, \dots, t_n, \dots\} \quad \begin{matrix} \swarrow \text{discrete-time process} \\ \text{countable} \end{matrix}$$

$$T = \{t, t \in \mathbb{R}^1\} \quad \begin{matrix} \text{random} \\ \text{signal} \end{matrix}$$

A Random Process  $X(t)$  is specified iff a rule is given or implied for determining the joint c.d.f.  $P_{X(t)}$  for any finite set of observation instances  $(t_1, \dots, t_n)$ .

$$T = \{t : t \in \mathbb{R}^2\} \quad \text{2D-space}$$

$$T = \{t : t \in \mathbb{R}^2 \times \mathbb{R}^1\} \quad \text{picture change}$$

Terminology:

$$X(t, \omega)$$

for  $t$  fixed

for  $\omega$  fixed

$t$  and  $\omega$  vary

random variable

sample function

ensemble

Sample process at

$t_1, t_2, \dots, t_n$ , then

$$\underbrace{X(t_1, \omega), X(t_2, \omega), \dots, X(t_n, \omega)}$$

- joint RVs

Characterization:  $X(t, \omega)$  is completely characterized if the joint c.d.f.

$$\begin{aligned} & \Pr [ X(t_1) \leq x_1, X(t_2) \leq x_2, \dots, X(t_n) \leq x_n ] \\ & = P_{X(t_1), X(t_2), \dots, X(t_n)}(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n) \end{aligned}$$

③ is known for all finite joint sets  $(t_1, t_2, \dots, t_n) \in T$  and all  $x_1, x_2, \dots, x_n \in S$ .

### Specifications of R.P.s

(1) Rule of construction for the edf

$$P_{X(t)}(\underline{x})$$

(2) Parameterization

$$X(\omega, t) = \cos(2\pi ft + \theta(\omega))$$

difficult, but possible to find  $P_{X(t)}(\underline{x})$  via  $f_{\theta}(\theta)$ .

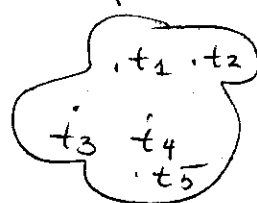
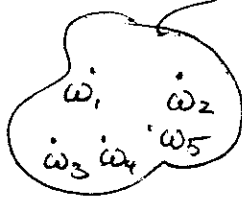
$$RV \sim f_{\theta}(\theta)$$

(3) transformation of a completely characterized R.P.

usually possible for G.R.P.

### Random Process:

$$X: \Omega \times T \rightarrow \mathbb{R}^n$$



$$X(\omega, t)$$

ex.

("rule")

Markov Process

discrete  $S$

continuous

$$X: \Omega \times T \rightarrow S$$

state space

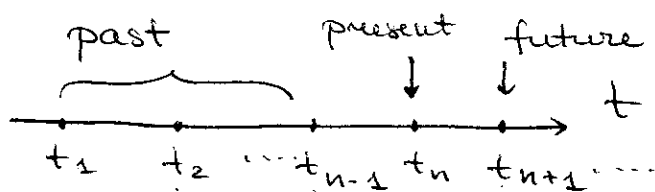
parameter space (index set)

discrete

$T$   
(param. space)

continuous

	discrete	continuous
discrete	discrete Markov chain	discrete parameter Markov process
continuous	continuous parameter Markov chain	continuous parameter Markov process



4.

Defn.  $X(t)$  is a Markov process if

$$\Pr [ X(t_n) \leq \alpha_n \mid X(t_{n-1}) \leq \alpha_{n-1}, \dots, X(t_1) \leq \alpha_1 ]$$

$$= \Pr [ X(t_n) \leq \alpha_n \mid X(t_{n-1}) \leq \alpha_{n-1} ] .$$

Then the joint distribution:

$$\Pr [ X(t_1) \leq \alpha_1, X(t_2) \leq \alpha_2, \dots, X(t_n) \leq \alpha_n ]$$

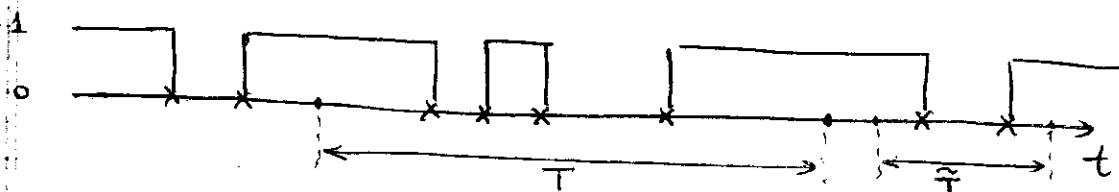
$$= \Pr [ X(t_n) \leq \alpha_n \mid X(t_{n-1}) \leq \alpha_{n-1} ]$$

$$\times \Pr [ X(t_{n-1}) \leq \alpha_{n-1} \mid X(t_{n-2}) \leq \alpha_{n-2} ]$$

$$\times \dots \times \Pr [ X(t_2) \leq \alpha_2 \mid X(t_1) \leq \alpha_1 ] \cdot \Pr [ X(t_1) \leq \alpha_1 ] .$$

- Specify:
- Type of Markov process marginal cdf
  - Transition probabilities
  - marginal cdf.

ex. 2: ("rule") Random Telegraph Wave



$X(t)$   
takes '0' or '1'

Rules:

1.  $\Pr [ X(t) = 0 ] = \Pr [ X(t) = 1 ] = \frac{1}{2}$

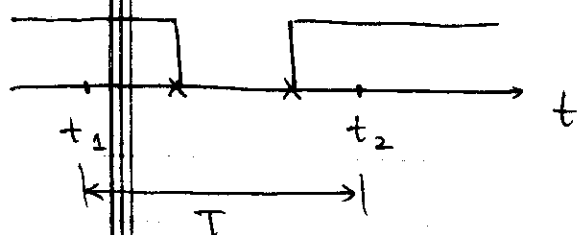
2. Changes  $0 \rightarrow 1$  or  $1 \rightarrow 0$  occur at each event.

3.  $\Pr [ k \text{ events in the interval } T ]$   
 $= \frac{(\lambda T)^k}{k!} \cdot e^{-\lambda T}$

( $\lambda \sim$  average intensity of events)

4. the number of events in disjoint intervals are statistically independent.

5.



Find:  $\Pr [X(t_2) = 1 | X(t_1) = 1; t_1, t_2]$

$= \Pr [\text{even \# events in } [t_1, t_2]]$

$= \sum_{k \text{ even}} \frac{(\lambda T)^k}{k!} \cdot e^{-\lambda T}$

use this trick

$= \sum_{k=0}^{+\infty} \frac{1}{2} [1 + (-1)^k] \frac{(\lambda T)^k}{k!} \cdot e^{-\lambda T}$

$= \frac{e^{-\lambda(t_2-t_1)}}{2} \cdot \left( \underbrace{\sum_{k=0}^{+\infty} \frac{(\lambda T)^k}{k!}}_{e^{\lambda T}} + \underbrace{\sum_{k=0}^{+\infty} \frac{(-\lambda T)^k}{k!}}_{e^{-\lambda T}} \right)$

$= \frac{1}{2} [1 + e^{-2\lambda(t_2-t_1)}]$

We can also find:  $\Pr [X(t_2) = 0 | X(t_1) = 1]$

$= \frac{1}{2} [1 - e^{-2\lambda(t_2-t_1)}]$

$\Pr [X(t_2) = 1 | X(t_1) = 0]$

$\Pr [X(t_2) = 0 | X(t_1) = 0]$

Consider:  $\Pr [X(t_3) = 1, X(t_2) = 1, X(t_1) = 1]$

$= \Pr [X(t_3) = 1 | X(t_2) = 1, X(t_1) = 1]$

$\times \Pr [X(t_2) = 1 | X(t_1) = 1] \cdot \underbrace{\Pr [X(t_1) = 1]}_{1/2}$

$\Pr [X(t_3) = 1 | X(t_2) = 1]$

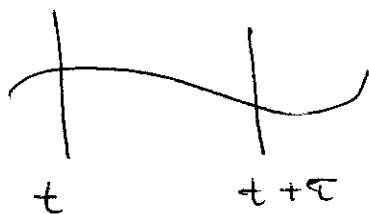
$= \left(\frac{1}{2}\right)^3 \cdot [1 + e^{-2\lambda(t_3-t_2)}][1 + e^{-2\lambda(t_2-t_1)}]$

Defn.: A RP  $X(t)$  is a strict sense stationary of order 1 if

$P_{X(t+\tau)}(\alpha; t+\tau) = P_{X(t)}(\alpha; t)$

6.

A sample function



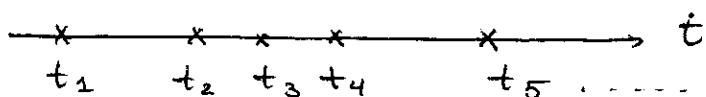
A RP is SSS of order 2 if

$$P_{X(t_1+\tau), X(t_2+\tau)}(\alpha_1, \alpha_2; t_1+\tau, t_2+\tau) = P_{X(t_1), X(t_2)}(\alpha_1, \alpha_2; t_1, t_2)$$

A RP is SSS if it is SSS for all orders.

ex.

Random Telegraph Wave



$$\Pr [ X(t_n) = 1, X(t_{n-1}) = 1, \dots, X(t_2) = 1 ] = \frac{1}{2^n} \cdot \prod_{i=2}^n [ 1 + e^{-2\lambda(t_i - t_{i-1})} ]$$

Consider a shift by  $\tau$ :

$$\frac{1}{2^n} \prod_{i=2}^n [ 1 + e^{-2\lambda(t_i + \tau - t_{i-1} - \tau)} ]$$

$\Rightarrow$  The RP is SSS.

