

Stochastic processes

①.

So far, $(X_i)_{i=1}^n$ a collection of random variables,

indexed by $i=1, \dots, n$.

"Can think of as discrete time,

Def: A stochastic process is a family of random variables $X(t)$,

$$\bar{X} = \{ X(t) : t \in T \}$$

on a probability space (Ω, \mathcal{F}, P)

where T is the index set of the process. //

Often $T = \mathbb{R}$ or $T = [0, \infty)$, but also $T = \mathbb{R}^2$
or $T = \mathbb{N}$.

3 ways of thinking about $\underline{\bar{X}}$:

(2)

- For each fixed $t \in T$, $X(t)$ is a random variable

$$\Omega \rightarrow \mathbb{R}.$$

- $\underline{\bar{X}}$ is a (measurable) function

$$\underline{\bar{X}}: T \times \Omega \rightarrow \mathbb{R}$$

$$(t, \omega) \mapsto X(t, \omega).$$

- For each fixed $\omega \in \Omega$, $t \mapsto X(t, \omega)$

is function of t called the sample path. //

Examples: • Brownian motion / Wiener process. \Rightarrow continuous in time.

• Poisson process \Rightarrow discrete increments.

• Let $f: [0, \infty) \rightarrow \mathbb{R}$ be deterministic, and Y a random variable, then

$$X(t, \omega) := f(t) Y(\omega), \quad T = [0, \infty)$$

is a stochastic process (if f is Borel measurable).

• Let $\varphi \in U(0, 2\pi)$ and set

(3.)

$$X(t) := A \cdot \sin(\omega \cdot t + \varphi), \quad t \in \mathbb{R} = T.$$

A : constant, amplitude "random wave"

ω : constant angular frequency

φ : random phase "Phase shift keying".

What about the distribution of \underline{X} ?

(4.)

Issue is that T is typically uncountable, e.g. $T = [0, \infty)$

Way out is to look at finite-dimensional distributions:

For each $n \in \mathbb{N}$ and times $t_1, t_2, \dots, t_n \in \overline{T}$, we set

$$F_{t_1, t_2, \dots, t_n}(x_1, x_2, \dots, x_n) := P(X(t_1) \leq x_1, X(t_2) \leq x_2, \dots, X(t_n) \leq x_n)$$

for $x_1, x_2, \dots, x_n \in \mathbb{R}$.

$$\mathcal{F} = \{F_{t_1, t_2, \dots, t_n}\}_{(t_1, t_2, \dots, t_n) \in T^n, n \in \mathbb{N}}$$

Collection of all finite-dimensional distributions.

Notations: • Mean function of a stochastic process

(5.)

$$\mu_{\bar{X}}(t) := \mathbb{E}[X(t)] \quad t \in T$$

• Variance function:

$$\text{Var}_{\bar{X}}(t) := \mathbb{E}[(X(t))^2] - (\mu_{\bar{X}}(t))^2, \quad t \in T$$

• Auto correlation function: $R_{\bar{X}}(t, s)$, $t, s \in T$

$$R_{\bar{X}}(t, s) := \mathbb{E}[X(t) \cdot X(s)] \quad \begin{matrix} \text{deterministic} \\ \text{depend on} \end{matrix}$$

• Auto Covariance function: $\text{Cov}_{\bar{X}}(t, s)$, $t, s \in T$ two "times"

$$\text{Cov}_{\bar{X}}(t, s) := R_{\bar{X}}(t, s) - \mu_{\bar{X}}(t) \cdot \mu_{\bar{X}}(s)$$

• A stochastic process is called weakly stationary if

- 1.) $\mu_{\bar{X}}(t) = \text{constant} = \mu$

- 2.) $R_{\bar{X}}(t, s)$ is a function of $|t-s|$ only, i.e.

$$R_{\bar{X}}(t, s) = h(|t-s|) \text{ for some function } h. \quad //$$

Remark: To compute the autocorrelation function,

it suffices to know $F_{t_1, t_2}(x_1, x_2)$

⑥

Lemma: $R_{\bar{X}}(t, s)$ is the autocorrelation function of

a stochastic process $\bar{X} = \{X(t) : t \in T\}$

if and only if

1.) Symmetry: $R_X(t, s) = R_{\bar{X}}(s, t) \quad \forall t, s \in T.$

2.) Non-negative definiteness

$$\textcircled{*} \sum_{i,j=1}^n \underbrace{x_i R_{\bar{X}}(t_i, t_j) x_j}_{\text{"n by n matrix with}} \geq 0 \quad \begin{array}{l} \text{for all } x_1, x_2, \dots, x_n \in \mathbb{R} \\ \text{for all } t_1, t_2, \dots, t_n \in T \end{array}$$

and all $n \in \mathbb{N}.$

" $\vec{x}' R_{\bar{X}} \vec{x} \geq 0$ "

→ C. f. non-negative definiteness
for covariance matrices of
random vectors.

Proof: Only necessity of (*)

7.

$$\sum_{i=1}^n \sum_{j=1}^n x_i x_j R_{\bar{X}}(t_i, t_j)$$

definition of $R_{\bar{X}}$

$$= \mathbb{E} \left[\sum_{i=1}^n \sum_{j=1}^n x_i X(t_i) \cdot x_j X(t_j) \right]$$

$$= \mathbb{E} \left[\underbrace{\left(\sum_{i=1}^n x_i \cdot X(t_i) \right)^2}_{\geq 0} \right] \geq 0$$

QED

Exercises: • Show that

$$|R_{\bar{X}}(t,s)| \leq \sqrt{R_X(t,t)} \cdot \sqrt{R_X(s,s)}$$

Hint: $|\mathbb{E}[XY]| \leq \sqrt{\mathbb{E}[X^2]} \cdot \sqrt{\mathbb{E}[Y^2]}$ Cauchy-Schwarz.

- $X(t) = A \sin(\omega t + \varphi)$, $\varphi \in U(0, 2\pi)$.
Find $M_{\bar{X}}(t)$ and $R_{\bar{X}}(t_1, t_2)$.

Hint: $\sin(\alpha) \sin(\beta) = \frac{\cos(\alpha - \beta) - \cos(\alpha + \beta)}{2}$

Answers: $M_{\bar{X}}(t) = \mathbb{E}[A \sin(\omega t + \varphi)] = \int \frac{1}{2\pi} A \sin(\omega t + \varphi) d\varphi = 0$

$R_{\bar{X}}(t_1, t_2) = \frac{A^2}{2} \cos(\omega(t_2 - t_1)) = R_X(s, t)$ $X(t)$ is weakly stationary!

Suppose we are given a family of "finite-dimensional" distribution functions F_{t_1, t_2, \dots, t_n} , for $n \in \mathbb{N}$, all $t_1, \dots, t_n \in T$,

$$F = \{F_{t_1, \dots, t_n}\}_{(t_1, \dots, t_n) \in T^n, n \in \mathbb{N}}$$

Is there a stochastic process \bar{X} with F as its family of finite dimensional distributions?

Thm (Kolmogorov consistency)

Assume that for any $n < m$ and any times $(t_i)_{i=1}^m$ it holds that

$$\lim_{x_{n+1}, x_{n+2}, \dots, x_m \rightarrow +\infty} F_{t_1, t_2, \dots, t_n, t_{n+1}, \dots, t_m}(x_1, x_2, \dots, x_n, x_{n+1}, \dots, x_m) = F_{t_1, t_2, \dots, t_n}(x_1, x_2, \dots, x_n), \quad \forall x_1, \dots, x_n \in \mathbb{R}$$

then there exists a probability space (Ω, \mathcal{F}, P) and a stochastic process \bar{X} on that probability space, having F as finite-dimensional distributions.