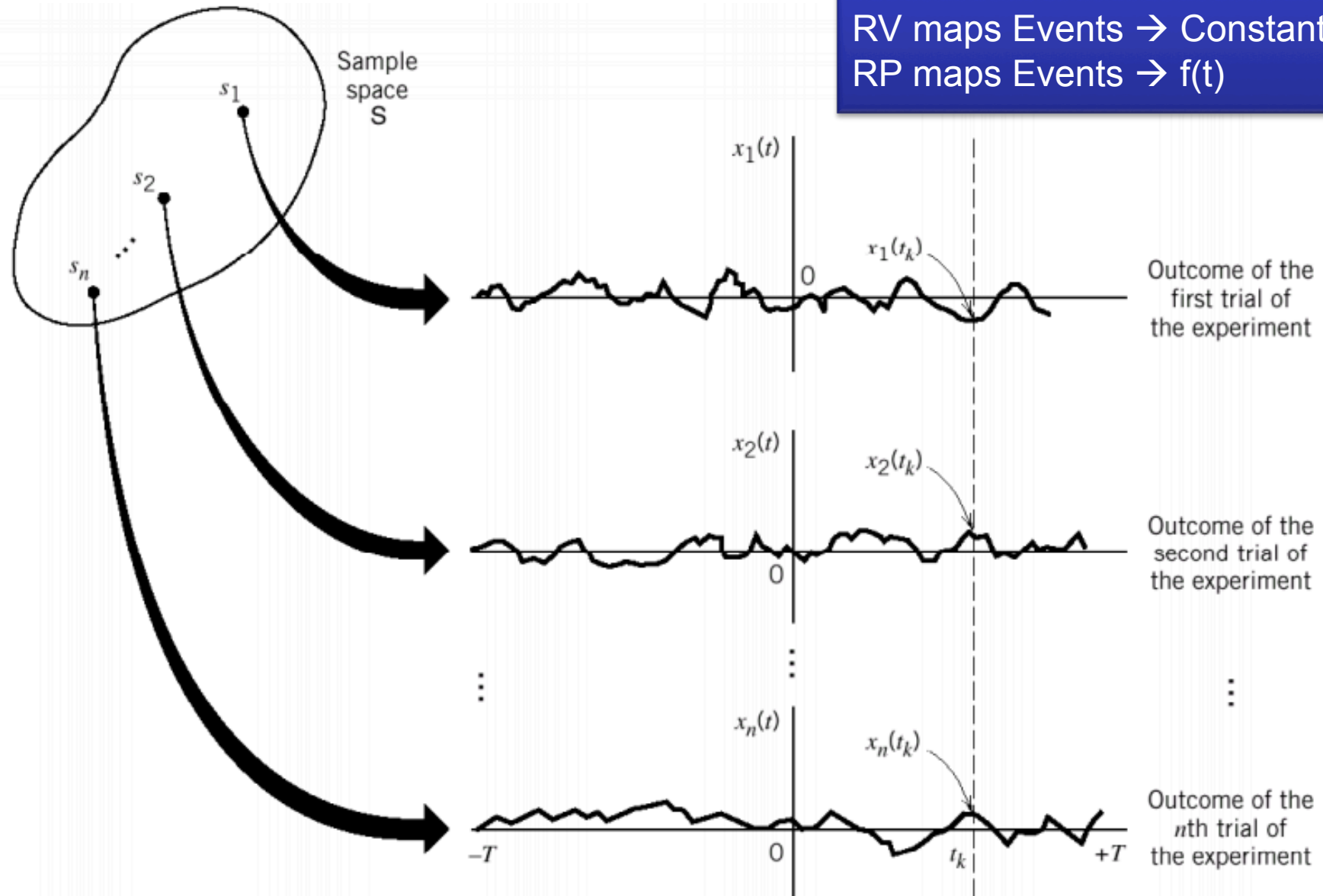


Random Process

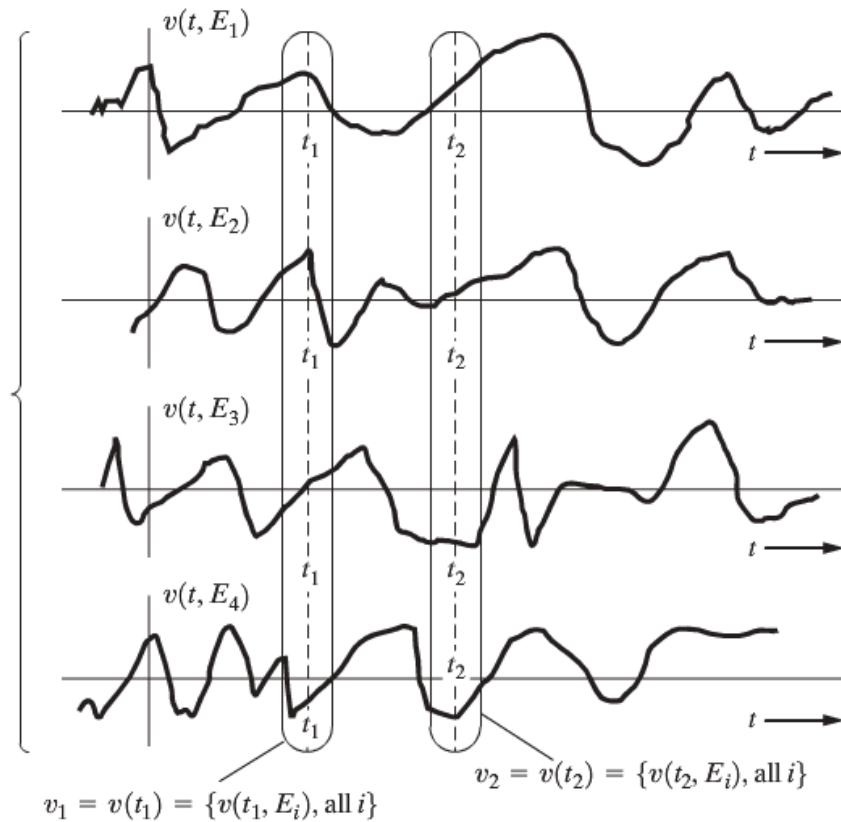
- A random process is a time-varying function that assigns the outcome of a random experiment to each time instant: $X(t)$.
- For a fixed (sample path): a random process is a time varying function, e.g., a signal.
 - For fixed t : a random process is a **random variable**.
- If one scans all possible outcomes of the underlying random experiment, we shall get an **ensemble** of signals.
- Random Process can be **continuous or discrete**
- Real random process also called **stochastic process**
 - Example: Noise source (Noise can often be modeled as a Gaussian random process).

An Ensemble of Signals

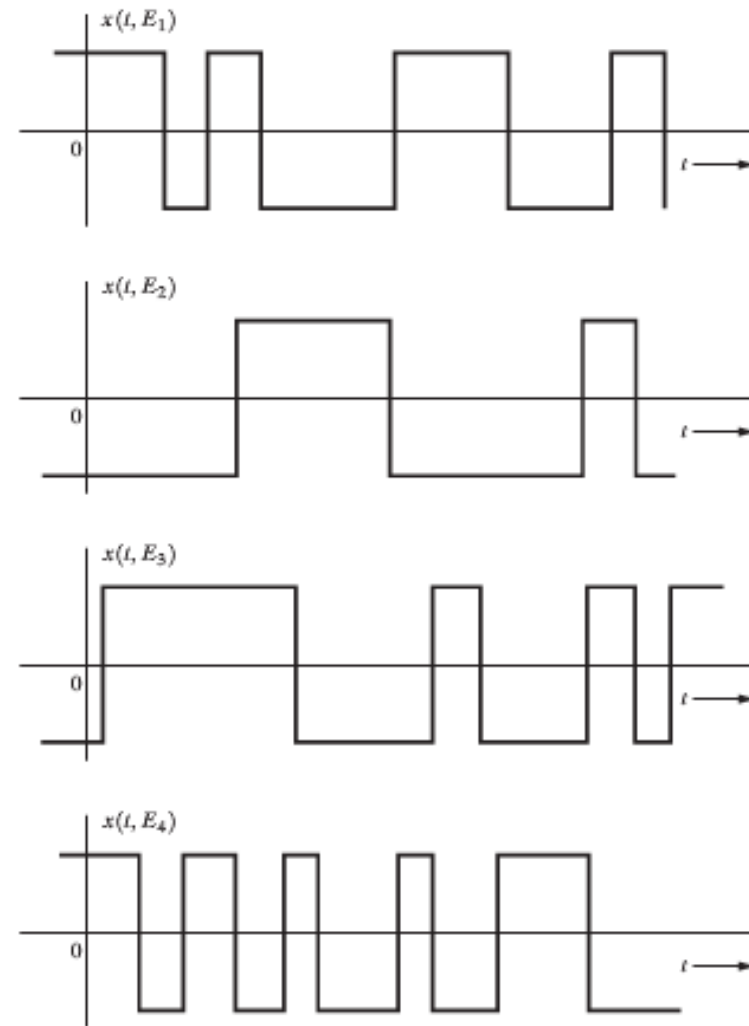
Remember:
RV maps Events \rightarrow Constants
RP maps Events \rightarrow $f(t)$



RP: Discrete and Continuous



The set of all possible sample functions $\{v(t, E_i)\}$ is called the ensemble and defines the random process $v(t)$ that describes the noise source.



Sample functions of a binary random process.

RP Characterization

- Random variables x_1, x_2, \dots, x_n represent amplitudes of sample functions at t_1, t_2, \dots, t_n .
 - A random process can, therefore, be viewed as a collection of an infinite number of random variables:

joint PDF $f_X(x_1, x_2, \dots, x_n, t_1, t_2, \dots, t_n)$:

RP Characterization – First Order

- CDF

$$F_{\mathbf{x}}(x, t) = P\{\mathbf{x}(t) \leq x\}$$

- PDF

$$f_{\mathbf{x}}(x, t) = \frac{dF_{\mathbf{x}}(x, t)}{dx}$$

- Mean

$$m_{\mathbf{x}}(t) = \overline{\mathbf{x}(t)} = E\{\mathbf{x}(t)\} = \int_{-\infty}^{+\infty} x f_{\mathbf{x}}(x, t) dx$$

- Mean-Square

$$\overline{\mathbf{x}^2(t)} = E\{\mathbf{x}^2(t)\} = \int_{-\infty}^{+\infty} x^2 f_{\mathbf{x}}(x, t) dx$$

Statistics of a Random Process

- For fixed t : the random process becomes a random variable, with mean

$$\mu_X(t) = E[X(t)] = \int_{-\infty}^{\infty} x f_X(x; t) dx$$

- In general, the mean is a function of t .

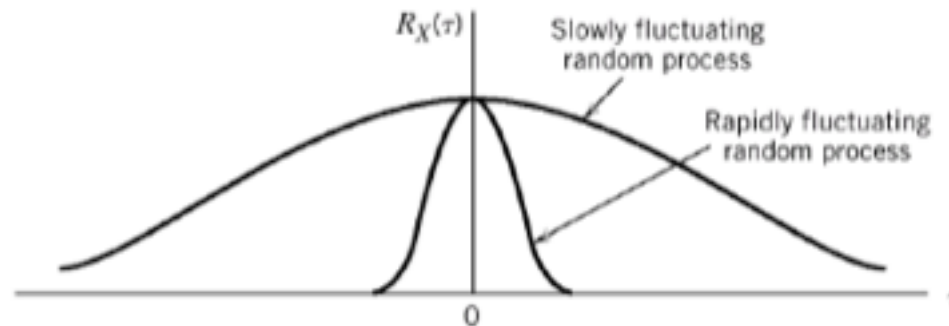
- Autocorrelation function

$$R_X(t_1, t_2) = E[X(t_1)X(t_2)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_X(x, y; t_1, t_2) dx dy$$

- In general, the autocorrelation function is a two-variable function.
- It measures the correlation between two samples.

RP Characterization – Second Order

- The first order does not provide sufficient information as to how **rapidly** the RP is changing as a function of time → We use second order estimation



RP Characterization – Second Order

- The first order does not provide sufficient information as to how **rapidly** the RP is changing as a function of time → We use second order estimation

$$F_{\mathbf{x}}(x_1, x_2, t_1, t_2) = P\{x(t_1) \leq x_1, x(t_2) \leq x_2\}$$

- CDF

$$f_{\mathbf{x}}(x_1, x_2, t_1, t_2) = \frac{\partial^2 F_{\mathbf{x}}(x_1, x_2, t_1, t_2)}{\partial x_1 \partial x_2}$$

- PDF

- Auto-correlation

(statistical average of the product of RVs)

$$R_{\mathbf{x}}(t_1, t_2) = E\{x(t_1)x(t_2)\}$$

- Cross-Correlation

(measure of correlation between sample function amplitudes of processes $x(t)$ and $y(t)$ at time instants t_1 and t_2 , respectively)

$$R_{\mathbf{xy}}(t_1, t_2) = E\{x(t_1)y(t_2)\}$$