

# Probability and Random Processes

## Lecture 11

- Measurable dynamical systems
- Random processes as dynamical systems
- Stationarity
- Ergodic theory

## Measurable Dynamical System

- A probability space  $(\Omega, \mathcal{A}, P)$
- A measurable transformation  $\phi : (\Omega, \mathcal{A}) \rightarrow (\Omega, \mathcal{A})$
- The space  $(\Omega, \mathcal{A}, P, \phi)$  is called a **measurable dynamical system**

## Interpretation

- Nature selects an **initial state**  $\omega = \omega_0$
- For  $n \geq 0$ , **time acts** on  $\omega \in \Omega$  to 'move around' points in  $\Omega$ ,

$$\omega_{n+1} = \phi(\omega_n) = \{ \text{notation} \} = \phi\omega_n = \phi^{n+1}\omega_0$$

producing an **orbit**  $\{\omega_n\}$

- The orbit is **random** because  $\omega_0$  is selected at random

## Random Process as Dynamical System

- There are several ways to model a random process as a dynamical system – consider a discrete-time process  $\{X_t\}$ , for  $t \in T$  and with  $T = \mathbb{N}$
- **Approach 1:**  $\{X_t\}$  is a collection of random variables  $X_t : \Omega \rightarrow \mathbb{R}$  where for each  $t$  and  $\omega$

$$X_t(\omega) = X(\phi^t\omega)$$

for some fixed random variable  $X$

- **Approach 2:** Define  $\phi$  implicitly on  $\Omega$  by specifying a time-shift  $\phi'$  on  $T$ , i.e.,  $\phi'(t) = t + 1$  and

$$(X_0, X_1, X_2, X_3, \dots) \xrightarrow{\phi'} (X_1, X_2, X_3, X_4, \dots)$$

- The model  $X_t(\omega) = X(\phi^t\omega)$  fits with interpreting  $\{X_t\}$  as a collection of random variables
  - Note that we can also consider  $(E^T, \mathcal{E}^T, \mu, \phi)$ , i.e., the evolution of the process is described by  $\phi : E^T \rightarrow E^T$ 
    - for example,  $\phi(f(t)) = f(t+1)$
- which fits better with the time-shift in approach 2

## Continuous Time

- We will focus on systems  $(\Omega, \mathcal{A}, P, \phi)$  interpreting  $\phi$  as 'one discrete action of time' – i.e., discrete-time systems
- Continuous-time systems can be modeled using a family  $\{\phi_t\}_{t \in \Phi}$  of transformations, with  $\Phi = \mathbb{R}$  or  $\mathbb{R}^+$ , such that

$$\phi_{t+s} = \phi_t \phi_s$$

That is  $\phi_{t+s}\omega = \phi_t \phi_s \omega$

- The family  $\{\phi_t\}$  is called a **flow**
- A random waveform can be defined, e.g., as

$$X_t(\omega) = X(\phi_t\omega)$$

## Stationarity

- A dynamical system  $(\Omega, \mathcal{A}, P, \phi)$
- The system is **stationary** if

$$P(A) = P(\phi^{-1}(A)), \quad \text{for all } A \in \mathcal{A}$$

(where  $\phi^{-1}(A) = \{\omega : \phi\omega \in A\} \subset \mathcal{A}$ )

- The system is **asymptotically mean stationary (AMS)** if the limit

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} P(\phi^{-i}(A)) = \bar{P}(A)$$

exists pointwise for all  $A \in \mathcal{A}$

- $(\Omega, \mathcal{A}, P, \phi)$  AMS  $\Rightarrow \bar{P}$  is a probability measure and  $(\Omega, \mathcal{A}, \bar{P}, \phi)$  is stationary  
(of course) stationary  $\Rightarrow$  AMS and  $\bar{P} = P$

## Recurrence

- A dynamical system  $(\Omega, \mathcal{A}, P, \phi)$
- A point  $\omega \in A$  is said to be **recurrent** with respect to  $A \in \mathcal{A}$  if there is a finite  $N = N_A(\omega)$  such that  $\phi^N \omega \in A$ 
  - $\omega \in A \Rightarrow \omega$  returns to  $A$  in finite time
- $A \in \mathcal{A}$  is recurrent if  $P(A) > 0$  and  $P(\{\omega \in A : \omega \text{ is not recurrent w.r.t. } A\}) = 0$
- $(\Omega, \mathcal{A}, P, \phi)$  is recurrent if all  $A \in \mathcal{A}$  are recurrent
- **stationary  $\Rightarrow$  recurrent**

# Ergodicity

- A dynamical system  $(\Omega, \mathcal{A}, P, \phi)$
- If  $A \in \mathcal{A}$  is such that  $\phi^{-1}(A) = A$  then  $A$  is **invariant**
- Let  $\mathcal{I} = \{ \text{all invariant } A \in \mathcal{A} \}$ 
  - $\mathcal{I}$  is a  $\sigma$ -algebra (why?)
- $(\Omega, \mathcal{A}, P, \phi)$  is **ergodic** if  $P(B) = 1$  or  $0$  for all  $B \in \mathcal{I}$

## The Ergodic Theorem

- **Theorem:** If  $(\Omega, \mathcal{A}, P, \phi)$  is AMS and  $X$  is a random variable such that

$$\int X(\omega) d\bar{P} < \infty$$

then

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} X(\phi^i \omega) = \bar{E}[X|\mathcal{I}]$$

with probability one (under  $P$  and  $\bar{P}$ ), where  $\bar{E}[X|\mathcal{I}]$  is conditional expectation w.r.t.  $\bar{P}$  and where  $\mathcal{I}$  is the  $\sigma$ -algebra of invariant events

- Let

$$\bar{X} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} X(\phi^i \omega)$$

then  $\bar{X}$  is an  $\mathcal{I}$ -measurable random variable

- Note that

$$\int \bar{X} d\bar{P} = \int X d\bar{P}$$

and that if  $(\Omega, \mathcal{A}, P, \phi)$ , in addition, is ergodic then

$$\bar{X} = \int X d\bar{P}, \quad \bar{P}\text{-a.e.}$$

- In particular, if  $(\Omega, \mathcal{A}, P, \phi)$  is **stationary and ergodic**, then

$$\bar{X} = E[X] \text{ with probability one}$$

## Ergodic Decomposition

- Fix a standard measurable space  $(\Omega, \mathcal{A})$  and  $\phi : \Omega \rightarrow \Omega$
- Assume there is a  $P$  such that  $(\Omega, \mathcal{A}, P, \phi)$  is stationary
- Then there are a standard space  $(\Gamma, \mathcal{S})$ , a family  $\{P_\gamma\}_{\gamma \in \Gamma}$  and a measurable transformation  $\tau : \Omega \rightarrow \Gamma$  such that
  - ①  $\tau$  is invariant ( $\tau(\phi\omega) = \tau(\omega)$ )
  - ②  $(\Omega, \mathcal{A}, P_\gamma, \phi)$  is stationary and ergodic, for each  $\gamma \in \Gamma$
  - ③ if  $\tau$  induces  $P^*(S) = P(\tau^{-1}(S))$  on  $(\Gamma, \mathcal{S})$ , then for all  $A \in \mathcal{A}$

$$P(A) = \int P_{\tau(\omega)}(A) dP(\omega) = \int P_\gamma(A) dP^*(\gamma)$$

- ④ if  $\int f(\omega) dP(\omega) < \infty$  then

$$\int f dP = \int \left\{ \int f dP_{\tau(\omega)} \right\} dP(\omega) = \int \left\{ \int f dP_\gamma \right\} dP^*(\gamma)$$

- For any stationary  $(\Omega, \mathcal{A}, P, \phi)$  we can decompose  $P$  into a mixture of **stationary and ergodic components**  $P_\gamma$
- For  $A \in \mathcal{A}$ , the component in effect is characterized by

$$P_\gamma(A) = P(A|\tau = \gamma)$$

(regular conditional probability, given the exact outcome  $\gamma$  of the random object  $\tau : (\Omega, \mathcal{A}) \rightarrow (\Gamma, \mathcal{S})$ )

- *Interpretation:* when time starts, Nature selects which component  $P_\gamma$  will be active, with probability  $P^*$  on  $(\Gamma, \mathcal{S})$ ,  
 $\Rightarrow$  the output from a stationary system always “looks ergodic,” however we do not know beforehand which ergodic component will be active