

#### Lecture 8: Random Processes, Part I

Yi, Yung (이용)

EE210: Probability and Introductory Random Processes KAIST EE

May 5, 2021

- A lot of applications in engineering systems
- Basics on Random Process
- Bernoulli Process
- Poisson Process
- Use of Bernoulli and Poisson Processes
- Coding of both processes
- Merge and Split
- Markov Chain

May 5, 2021 1 / 36

May 5, 2021 2 / 36

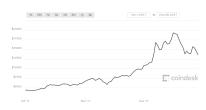
### Roadmap

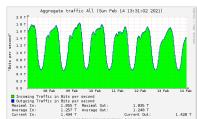
- A lot of applications in engineering systems
- Basics on Random Process
- Bernoulli Process
- Poisson Process
- Use of Bernoulli and Poisson Processes
- Coding of both processes
- Merge and Split
- Markov Chain

## KAISTEE Things that evolve in time

- Many probabilistic experiments that evolve in time
  - $\,\circ\,$  Sequence of daily prices of a stock
  - Sequence of scores in football
  - $\,^\circ\,$  Sequence of failure times of a machine
  - Sequence of traffic loads in Internet
- Random process is a mathematical model for it.







#### Random Process: A Sneak Peek



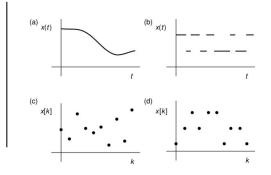
### 4 Types of Random Processes



- A random process is a sequence of random variables indexed by time.
- Time: discrete or continuous
- Notation
- $(X_t)_{t\in\mathcal{T}}$  or  $\Big(X(t)\Big)_{t\in\mathcal{T}}$ , where  $\mathcal{T}=\mathbb{R}$  (continuous) or  $\mathcal{T}=\{0,1,2,\ldots\}$  (discrete)
- For the discrete case, we also often use  $(X_n)_{n\in\mathbb{Z}_+}$ .
- We will use all of them, unless confusion arises.
- For a fixed time t,  $X_t$  is a random variable.
- The values that  $X_t$  can take: discrete or continuous

- Types of time and value

- (a) continuous time, continuous value
- (b) continuous time, discrete value
- (c) discrete time, continuous value
- (d) discrete time, discrete value



May 5, 2021 6 /

May 5, 2021 5 / 36

# Roadmap KAIST EE

**KAIST EE** 

- A lot of applications in engineering systems
- Basics on Random Process
- Bernoulli Process
- Poisson Process
- Use of Bernoulli and Poisson Processes
  - Coding of both processes
  - Merge and Split
- Markov Chain

# Bernoulli Process

- $\bullet$  At each minute, we toss a coin with probability of head 0 < p < 1.
  - Sequence of lottery wins/looses
  - $\,{}^{_{\odot}}$  Customers (each second) to a bank
  - o Clicks (at each time slot) to server
- A sequence of independent Bernoulli trials  $X_1, X_2, \ldots$
- We call index 1, 2, ... time slots (or simply slots)

### 0010100000110111001

- Discrete time, discrete value
- One of the simplest random processes
- A type of "arrival" process

May 5, 2021 7 / 36 May 5, 2021 8 /

- Question. We've already studied a sequence of Bernoulli rvs  $X_1, X_2, \dots, X_n$ . What's the difference?
- Physical difference: infinite sequence of  $X_1, X_2, \ldots$ .
  - Sample space? set of all outcomes?
  - an outcome: an infinite sequence of sample values  $x_1, x_2, \ldots, e.g., (0,1,1,0,0,1,\ldots)$
- Semantic difference: Understand i in X<sub>i</sub> as time. Also, interesting questions from the random process point of view.
  - Dependence: How  $X_1, X_2, \ldots$  are related to each other as a time series
  - Long-term behavior: What is the fraction of times that a machine is idle?
  - Other interesting questions, depending on the target random process
- Next: Key questions and answers about Bernoulli process

(Q1) # of arrivals in n slots?

- $S_n = X_1 + X_2 + \cdots + X_n$
- $S_n \sim Bin(n,p)$
- $\mathbb{E}(S_n) = np$ ,  $var(S_n) = np(1-p)$

(Q2) # of slots  $T_1$  until the first arrival?

- *T*<sub>1</sub> ∼ *Geom*(*p*)
- $\mathbb{E}(T_1) = 1/p$ ,  $var(T_1) = \frac{1-p}{p^2}$

- $T_1$  is geometric? Memoryless
- Conditioned no first arrival at some time, what's the distribution of the remaining time until the first arrival?
- Still, geometric.
- But, more than that, as we will see.
   Independence across time slots leads to many useful properties, allowing the quick solution of many problems.

May 5, 2021 9 / 36

May 5, 2021 10 / 36

### Memoryless and Fresh-start after Deterministic n



Fresh-start after Random N(1)



(Q3)  $U = X_1 + X_2 \perp \!\!\!\perp V = X_5 + X_6$ ?

- Yes
- Because  $X_i$ s are independent

(Q4) The process  $(X_n)_{n=6}^{\infty}$ ?

- $(X_1,\ldots,X_5) \perp \!\!\!\perp (X_n)_{n=6}^{\infty}$
- Fresh-start after a deterministic time n
- If you watch the on-going Bernoulli process(p) from some time n, you still see the same Bernoulli process(p).

(Q5) The process  $(X_N, X_{N+1}, X_{N+2}, ...)$ ? Fresh-start even after random N?

- Examples of *N*
- E1. Time of 3rd arrival
- **E2.** First time when 3 consecutive arrivals have been observed
- E3. Time just before 3 consecutive arrivals

- Difference of *N* from *n*
- The time when I watch the on-going Bernoulli process is random.

May 5, 2021 11 / 36

### Fresh-start after Random N (2)



### Distribution of Busy Periods



(Q5) The process  $(X_N, X_{N+1}, X_{N+2}, ...)$ ? Fresh-start even after random N?

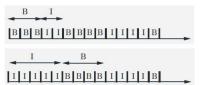
- Examples of N
- E1. Time of 3rd arrival
- **E2.** First time when 3 consecutive arrivals have been observed
- E3. Time just before 3 consecutive arrivals



- E2 | 1 | 0 | 0 | 1 | 0 | 1 | 1 | 1 | 0 | |
- **E1.** When I watch the process, N has been already determined. Yes
- **E2.** Same as **E1.** Yes
- E3. Need the future knowledge. '111' does not become random. No
- The question of N = n? can be answered just from the knowledge about  $X_1, X_2, \dots, X_n$ ? Then, Yes! (see pp. 301 for more formal description)

Regard an arrival as business of a server

• First busy period  $B_1$ : starts with the first busy slot and ends just before the first subsequent idle slot



- (Q6) Distribution of  $B_1$ ?
- N: time of the first busy slot. Fresh-start after N.
- $B_1$  is geometric with parameter (1-p)
- Question: What about the second busy period  $B_2$ ?  $B_3$ ,  $B_4$ ?

May 5, 2021 13 / 36 May 5, 2021 14 / 36

#### Time of k-th arrival



PMF of  $Y_k$ 



- Time of the first arrival  $Y_1 \sim geom(p)$
- (Q7) Time of the k-th arrival  $Y_k$ ?

- 
$$T_k = Y_k - Y_{k-1}$$
: k-th inter-arrival  $(k \ge 2, T_1 = Y_1)$   
-  $Y_k = T_1 + T_2 + \ldots + T_k$ .

- After each  $T_k$ , the fresh-start occurs.
- $\{T_i\}$  are i.i.d. and  $\sim geom(p)$
- $\mathbb{E}[Y_k] = \frac{k}{p}$ ,  $\operatorname{var}[Y_k] = \frac{k(1-p)}{p^2}$

- $Y_k = T_1 + T_2 + \ldots + T_k$ .
- $\{T_i\}$  are i.i.d. and  $\sim geom(p)$

$$\begin{split} \mathbb{P}(Y_k = t) &= \mathbb{P}\Big(X_k = 1 \text{ and } k-1 \text{ arrivals during the first } t-1 \text{ slots}\Big) \\ &= \mathbb{P}\Big(X_k = 1\Big) \cdot \mathbb{P}\Big(k-1 \text{ arrivals during the first } t-1 \text{ slots}\Big) \\ &= p \times \binom{t-c}{k-1} p^{k-1} (1-p)^{t-k} = \binom{t-c}{k-1} p^k (1-p)^{t-k}, \quad t = k, k+1, \dots \end{split}$$

- $Y_k$  is called Pascal rv with parameter (k, p).
- Pascal(1, p) = Geometric(p)



- A lot of applications in engineering systems
- Basics on Random Process
- Bernoulli Process
- Poisson Process
- Use of Bernoulli and Poisson Processes
  - Coding of both processes
  - Merge and Split
- Markov Chain

- Very useful to both continuous and discrete random processes that are "twins" and share the key properties.
  - Independence between what happens in a different time region
  - Memoryless and fresh-start property
- Choose one of discrete or continuous versions for our modeling convenience.
- Question. How do we design the continuous analog of Bernoulli process?
- Key idea: Making it as a limiting system of a sequence of Bernoulli processes
- Need a "modeling sense" to make this possible. It's a good practice for engineers!

May 5, 2021 17 / 36

May 5, 2021 18 / 36

## Key Design Idea to Develop a Continuous Twin (1)

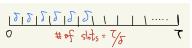


Key Design Idea to Develop a Continuous Twin (2)



- Continuous twin
  - Key point: Understand the number of arrivals over a given interval  $[0, \tau]$
  - Assume that it has some arrival rate  $\lambda$  (# of arrivals/unit time).
  - We know how to handle Bernoulli process with discrete time slots.
- Divide  $[0, \tau]$  into slots whose length  $= \delta$ . Then, n = # of slots  $= \frac{\tau}{\delta}$ .

• What's the limit as  $\delta \to 0$  (equivalently,  $n \to \infty$ )



• Now, our design idea: during one time slot of length  $\delta$ ,

 $\mathbb{P}(1 \text{ arrival}) \propto \text{arrival rate and slot length}$  $\mathbb{P}(\geq 2 \text{ arrivals}) \propto \text{something, but very small}$ for small sloth length  $\mathbb{P}(0 \text{ arrival}) = 1 - \mathbb{P}(1 \text{ arrival or } \geq 2 \text{ arrivals})$ 

$$\mathbb{P}(1 ext{ arrival}) = \lambda \delta + o(\delta)$$
 $\mathbb{P}(\geq 2 ext{ arrivals}) = 0o(\delta)$ 
 $\mathbb{P}(0 ext{ arrival}) = 1 - \lambda \delta + o(\delta)$ 

- $o(\delta)$ : some function that goes to zero faster than  $\delta$  goes to zero.
- Thus, for very small  $\delta$ ,  $o(\delta)$  becomes negligible.
- Example:  $o(\delta) = \delta^{\alpha}$ , where any  $\alpha > 1$

May 5, 2021 19 / 36 May 5, 2021

### Key Design Idea to Develop a Continuous Twin (3)



#### Poisson Process: Formalism



- 0 # of shits = 75 7
- Our interest: Prob. of k arrivals over  $[0, \tau]$
- Given "small"  $\delta$ , # of arrivals  $\sim$  Binomial(n, p), where  $n = \tau/\delta$  and  $p = \lambda\delta$
- As  $\delta \to \infty$ ,  $np = \tau/\delta \times \lambda \delta = \lambda \tau$ .
- # of arrivals over  $[0, \tau]$ ,  $\sim Poisson(\lambda \tau)$
- This is a continuous twin process of Bernous process, which we call Poisson process.

- $N_s$ : number of arrivals over the interval [0, s].
- (Independence) If s < t, the number  $N_t N_s$  of arrivals over [s, t] is independent of the times of arrivals during [0, s].
- Thus,  $N_s$  can be a random variable over any interval of length s.
- (Small interval probability) The probabilities  $\mathbb{P}(k, s)$  satisfy:

$$\mathbb{P}(0,s) = 1 - \lambda \tau + o(s)$$

$$\mathbb{P}(1,s) = \lambda s + o_1(s)$$

$$\mathbb{P}(k,s) = o_k(s) \quad \text{for } k = 2,3,\ldots,$$

where

$$\lim_{s\to 0}\frac{o(s)}{s}=0,\quad \lim_{s\to 0}\frac{o_k(s)}{s}=0$$

May 5, 2021 21 / 36

May 5, 2021 22 / 36

## Poisson Process: $\mathbb{P}(k,\tau)$ , $N_{\tau}$ , and T



Poisson Process: Example



• (Q1) Number of arrivals of any interval with length  $\tau \sim Poisson(\lambda \tau)$ , i.e.,

$$\mathbb{P}(k,\tau) = e^{-\lambda \tau} \frac{(\lambda \tau)^k}{k!}, \quad k = 0, 1, 2, \dots$$

- $\mathbb{E}[\textit{N}_{ au}] = \lambda au$  and  $\text{var}[\textit{N}_{ au}] = \lambda au$
- (Q2) Time of first arrival T

$$egin{aligned} F_T(t) &= \mathbb{P}(T \leq t) = 1 - \mathbb{P}(T > t) = 1 - P(0,t) = 1 - e^{-\lambda t} \ f_T(t) &= rac{dF_T(t)}{dt} = \lambda e^{-\lambda t}, \quad t \geq 0. \end{aligned}$$

- $T \sim expo(\lambda)$ . Thus  $\mathbb{E}[T] = 1/\lambda$  and  $var[T] = 1/\lambda^2$
- Continuous twin of geometric rv in Bernoulli process
- Memoryless

- Receive emails according to a Poisson process at rate  $\lambda=5$  messages per hour
- Mean and variance of mails received during a day
- -5\*24 = 120
- P[one new message in the next hour]

$$-\mathbb{P}(1,1) = \frac{(5\cdot1)^1 e^{-5\cdot1}}{1!} = 5e^{-5}$$

- ullet  $\mathbb{P}[\text{exactly two msgs during each of the next three hours}]$
- $-\left(\frac{5^2e^{-5}}{2!}\right)^3$

### Memoryless and Fresh-start Property



PDF of  $Y_k$ 



- Remind. Similar property for Bernoulli processes, but here no time slots.
- Fresh-start at determinsitic time: Start watching at time t, then you see the Poisson process, independent of what has happened in the past.
- Fresh-start at random time: Similarly holds. For example, when you start watching at random time  $T_1$  (time of first arrival)
- (Q3) The k-th arrival time  $Y_k$ ?
- k-th inter-arrival time  $T_k = Y_k = Y_{k-1}, k \ge 2$ , and  $T_1 = Y_1$ .
- $Y_k = T_1 + T_2 + \cdots + T_k$  is sum of i.i.d. exponential rvs.
- $\mathbb{E}[Y_k] = k/\lambda$  and  $var[Y_k] = k/\lambda^2$

• For a given  $\delta$ ,  $\delta \cdot f_{Y_k}(y)$ : prob of k-th arrival over  $[y, y + \delta]$ .

• When  $\delta$  is small, only one arrival occurs. Thus,

$$\delta \cdot f_{Y_k}(y) = \mathbb{P}\Big( ext{an arrival over } [y,y+\delta]\Big) imes \mathbb{P}\Big(k-1 ext{ arrivals before } y\Big)$$

$$pprox \lambda \delta imes \mathbb{P}(k-1,y) = \lambda \delta imes rac{\lambda^{k-1}y^{k-1}e^{-\lambda y}}{(k-1)!}$$

$$f_{Y_k}(y) = rac{\lambda^k y^{k-1}e^{-\lambda y}}{(k-1)!}, \quad y \geq 0.$$

- This is called **Erlang** rv.
- Time of first arrival: geometric / exponential
- Time of k-th arrivals: Pascal / Erlang

May 5, 2021 25 / 36

May 5, 2021 26 / 36

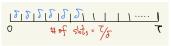
#### Poisson Process vs. Bernoulli Process



Example: Poisson Fishing (Problem 10, page 329)



-  $\mathbf{n} = \tau/\delta$ ,  $\mathbf{p} = \lambda\delta$ ,  $\mathbf{n}\mathbf{p} = \lambda\tau$ 



	POISSON	BERNOULLI
Times of Arrival	Continuous	Discrete
PMF of # of Arrivals	Poisson	Binomial
Interarrival Time CDF	Exponential	Geometric
Arrival Rate	$\lambda$ /unit time	p/per trial

- Catching fish: Poisson process  $\lambda=0.6/\text{hour}.$
- Fish for 2 hours. Within 2 hours, if he catches at least one fish, then he stops at 2 hours. Otherwise, he fishes until one fish is caught.
- (Q1)  $\mathbb{P}(\text{fishing time} > 2 \text{ hours})$ Method 1:  $\mathbb{P}(0,2)$ Method 2:  $\mathbb{P}(T_1 > 2)$
- (Q2)  $\mathbb{P}(2 < \text{fishing time} < 5)$ Method 1:  $\mathbb{P}(0,2)(1-\mathbb{P}(0,3))$ Method 2: $\mathbb{P}(2 < T_1 < 5)$
- (Q3)  $\mathbb{P}(\text{Catch at least two fish})$ Method  $1:\sum_{k=2}^{\infty} = 1 - \mathbb{P}(0,2) - \mathbb{P}(1,2)$ Method  $2:\mathbb{P}(Y_k \leq 2)$
- (Q4)  $\mathbb{E}[\text{future fi. time}|\text{already fished for 3h}]$ Fresh-start. So,  $\mathbb{E}[\exp(\lambda)] = 1/\lambda = 1/0.6$

(Q5)  $\mathbb{E}[F=\text{total fishing time}]$ 

- $2 + \mathbb{E}[F 2] = 2 + \mathbb{P}(F = 2) \cdot 0 + \\ \mathbb{P}(F > 2) \cdot \mathbb{E}[F 2|F > 2]$  $= 2 + \mathbb{P}(0, 2) \cdot \frac{1}{2}$
- (Q6)  $\mathbb{E}[\text{number of fish}] = \lambda \cdot 2 + \mathbb{P}(0,2) \cdot 1$

- A lot of applications in engineering systems
- Basics on Random Process
- Bernoulli Process
- Poisson Process
- Use of Bernoulli and Poisson Processes
  - Coding of both processes
  - Merge and Split
- Markov Chain

- Inter-arrival times facilitates coding of both processes

#### Alternative Description of the Bernoulli Process

- 1. Start with a sequence of independent geometric random variables  $T_1$ ,  $T_2$ , ..., with common parameter p, and let these stand for the interarrival times.
- 2. Record a success (or arrival) at times  $T_1$ ,  $T_1 + T_2$ ,  $T_1 + T_2 + T_3$ , etc.

#### Alternative Description of the Poisson Process

- 1. Start with a sequence of independent exponential random variables  $T_1, T_2, \ldots$ , with common parameter  $\lambda$ , and let these stand for the interarrival times.
- 2. Record an arrival at times  $T_1$ ,  $T_1 + T_2$ ,  $T_1 + T_2 + T_3$ , etc.

May 5, 2021 29 / 36

May 5, 2021 30 / 36

#### Sum of Independent Poisson rvs

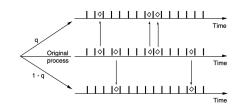


Split and Merge: Bernoulli Process



- $X \sim Poisson(\mu), Y \sim Poisson(\nu),$
- (Q1) *X* ⊥⊥ *Y*?
- (Q2) Distribution of X + Y?
  - Complex convolution, but any other easy way?
- X can be regarded as the number of arrivals of Poisson process with rate 1 over the time interval of length  $\mu$ .
- $\bullet$  Consecutive intervals of length  $\mu$  and  $\nu$
- (Q1) X ⊥⊥ Y? Yes
- (Q2) Distribution of X + Y?  $Poisson(\mu + \nu)$

- Split *Bernoulli(p)* into two processes with biased coin of head probability *q*
- Split decisions are independent of arrivals
- Split processes: also Bernoulli processes
- Bernoulli(pq) and Bernoulli(p(1 q))



- Merge Bernoulli(p) and Bernoulli(q) into one.
- Collided arrival is regarded just one arrival in the merged process
- Merged process: Bernoulli(1 - (1 - p)(1 - q))



May 5, 2021 31 / 36 May 5, 2021 32 / 36

### Split and Merge: Poisson Process



### Competing Exponential



- Split Poisson process ( $\lambda$ ) into two processes
  - Split based on the coin tossing with probability of head p
  - Poisson process  $(p\lambda)$  and Poisson process  $((1-p)\lambda)$
- Merge from Poisson process  $(\lambda_1)$  and Poisson process  $(\lambda_2)$ 
  - $\circ$  Split based on the coin tossing with probability of head p
  - Poisson process  $(\lambda_1 + \lambda_2)$
  - $\circ$  Bernoulli process of small interval  $\delta$

$$\mathbb{P}(\text{0 arrivals in the merged process}) \approx (1 - \lambda_1 \delta)(1 - \lambda_2 \delta) = 1 - (\lambda_1 + \lambda_2)\delta + o(\delta)$$

$$\mathbb{P}(\text{1 arrivals in the merged process}) \approx \lambda_1 \delta(1 - \lambda_2 \delta) + \lambda_2 \delta(1 - \lambda_1 \delta) = (\lambda_1 + \lambda_2)\delta + o(\delta)$$

- 1. Two independent light bulbs have life times  $T_a$  and  $T_b$  of exponential distributions with  $\lambda_a$  and  $\lambda_b$ .
- (Q) Distribution of  $Z = \min\{T_a, T_b\}$ ?
- $T_a$  and  $T_b$  are the first arrival times of two Poisson processes of  $\lambda_a$  and  $\lambda_b$ .
- Z is the first arrival time of merged Poisson process  $(\lambda_a + \lambda_b)$ .
- Thus,  $Z \sim exp(\lambda_a + \lambda_b)$

- 2. Three independent light bulbs have life times T of exponential distribution with  $\lambda$ .
- (Q)  $\mathbb{E}[\text{time until the last bulb burns out}]?$
- Poisson process(3 $\lambda$ )  $\xrightarrow{\text{1st burn out}}$  Poisson process(2 $\lambda$ )  $\xrightarrow{\text{2st burn out}}$  Poisson process( $\lambda$ )
- T<sub>1</sub>: time until the first burn-out, T<sub>2</sub>: time until the second burn-out, T<sub>3</sub>: time until the third burn-out
- $T_1 \sim exp(3\lambda)$ ,  $T_2 \sim exp(2\lambda)$ ,  $T_3 \sim exp(\lambda)$

$$\mathbb{E}\Big[T_1+T_2+T_3\Big]=\frac{1}{3\lambda}+\frac{1}{2\lambda}+\frac{1}{\lambda}$$

May 5, 2021 34 / 36

May 5, 2021 33 / 36

**KAISTEE** Review Questions



Questions?

1)

May 5, 2021 35 / 36 May 5, 2021 36 /