# Probability and Stochastic Processes I I - Lecture 6d

## Michael Evans University of Toronto https://utstat.utoronto.ca/mikevans/stac62/staC632024.html

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### VI.4 Queueing Theory

- consider a queue of customers arriving at a server

- suppose customers arrive at the queue according to interrival times  $Y_1, Y_2, \ldots \overset{i.i.d.}{\sim}$  exponential $(\lambda)$  and the service times of these customers are  $S_1, S_2, \ldots \overset{i.i.d.}{\sim}$  exponential $(\mu)$ ,

- let  $T_n = \sum_{i=1}^n Y_i$  = arrival time of *n*-th customer and

 $Q_t = \#$  of customers in the queue at time t

- with these assumptions  $\{Q_t:t\geq 0\}$  is called a M/M/1 queue

- clearly, because of the memoryless property of the exponential distribution, for any  $0 \le t_1 \le \cdots \le t_n \le t$ , then

$$P(Q_t = j | Q_{t_i} = j_i \text{ for } i = 1, ..., n) = P(Q_t = j | Q_{t_n} = j_n)$$

and this depends on time only through  $t - t_n$ , so  $\{Q_t : t \ge 0\}$  is a time homogeneous Markov process

- note that the number of arrivals is a Poisson process of intensity  $\lambda$  .

**Proposition VI.28** A M/M/1 queue with interarrival times exponential( $\lambda$ ) and service times exponential( $\mu$ ) has generator matrix

$$G = \begin{pmatrix} -\lambda & \lambda & 0 & \dots \\ \mu & -\lambda - \mu & \lambda & \\ 0 & \mu & -\lambda - \mu & \ddots \\ \vdots & & \ddots & \ddots \end{pmatrix}$$

Proof: See text and use Lemma VI.14.

**Proposition VI.29** If  $\lambda < \mu$ , the stationary distribution  $\pi$  of a M/M/1 queue with interarrival times exponential( $\lambda$ ) and service times exponential( $\mu$ ) is geometric( $1 - \lambda/\mu$ ).

Proof: The distribution  $\pi$  satisfies  $\pi G = 0$ . So  $\pi_0(-\lambda) + \pi_1(\mu) = 0$ implying  $\pi_1 = (\lambda/\mu)\pi_0$ . Also  $\pi_0(\lambda) + \pi_1(-\lambda-\mu) + \pi_2(\mu) = 0$  or  $\pi_0(\lambda) + \pi_0(\lambda/\mu)(-\lambda-\mu) + \pi_2(\mu) = \pi_0(-\lambda^2/\mu) + \pi_2(\mu) = 0$  so  $\pi_2 = (\lambda/\mu)^2\pi_0$  and in general  $\pi_k = (\lambda/\mu)^k\pi_0$ . since  $1 = \sum_{k=0}^{\infty} \pi_k$  this implies that  $\pi_0 = (\sum_{k=1}^{\infty} (\lambda/\mu)^k)^{-1} = 1 - \lambda/\mu$  which completes the proof. **Proposition VI.30** If  $\lambda < \mu$ , then for a M/M/1 queue with interarrival times exponential( $\lambda$ ) and service times exponential( $\mu$ )

$$\lim_{t\to\infty} P(Q_t=k) = (\lambda/\mu)^k (1-\lambda/\mu).$$

Proof: Since the transition  $i \rightarrow i + 1$  always has positive probability the process is irreducible. So by Prop. VI.26 the result follows.

- what happens when  $\lambda \ge \mu$ ? it can be shown that  $\lim_{k\to\infty}\lim_{t\to\infty} P(Q_t\ge k)\to 1$ 

- when  $\lambda < \mu$  and t is large there are k individuals in the queue with probability  $(\lambda/\mu)^k (1 - \lambda/\mu)$  and so a newly arrived individual will have to wait  $W = S_1 + \cdots + S_k \sim \text{gamma}(k, \mu)$  for service so (intuitively) when t is large, and  $G(\cdot; k, \mu)$  is the gamma $(i, \mu)$  cdf,

$$P(W > w) = \sum_{k=0}^{\infty} (\lambda/\mu)^k (1 - \lambda/\mu) (1 - G(w; k, \mu) < \infty)$$

- a G/G/1 queue assumes interarrival times  $Y_n$  are *i.i.d.*, the service times  $S_n$  are *i.i.d.* independent of the interarrvial times but these distributions are not assumed to be exponential

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#### VI.5 Renewal Theory

- consider r.v.'s  $Y_1$ ,  $Y_2$ ,  $Y_3$ , ... which are mutually stat. ind. and  $Y_2$ ,  $Y_3$ , ... are *i.i.d*.

- consider a machine that has been functioning properly but after  $Y_1$  units of time from when it is being monitored, it needs maintenance and then needs maintenance after  $Y_2$  units of time, then  $Y_3$  units of time, etc. and denote the maintenance times times by  $T_n = Y_1 + \cdots + Y_n$  with  $T_0 = 0$ 

**Definition VI.10** The process  $\{N_t : t \in T\}$ , where  $N_t = \#\{n : T_n \leq t\} = \#$  of renewals up to time *t*, is called a *renewal* process.

- if  $Y_1, Y_2, Y_3, \ldots$  are *i.i.d.*, then there is *zero delay* and and when the common distribution is exponential( $\lambda$ ), then this is a Poisson process

- fact,  $\{N_t : n \in \mathbb{N}_0\}$  is a Markov process iff it is a Poisson process (recall memoryless property of exponential distribution)

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#### **Example VI.10** Renewal process from a Markov chain

- let  $\{X_n: n \in \mathbb{N}_0\}$  be an irreducible, recurrent Markov chain and assume  $X_0 = j$ 

- put 
$$T_1(i) = \min\{k : X_k = i\}$$
 and for  $n \ge 2$  put  
 $T_n(i) = \min\{k > T_{n-1}(i) : X_k = i\} =$ time of *n*-th visit to state *i*

- put 
$$Y_n = T_n - T_{n-1}$$
 so

$$P(Y_1 = y | X_0 = j) = P(X_y = i, X_m \neq i, 1 \le m \le y - 1 | X_0 = j)$$

and using MP, for  $n \ge 2$ 

$$P(Y_n = y | X_{n-1} = i)$$
  
=  $P(X_y = i, X_m \neq i, 1 \le m \le y - 1 | X_0 = i)$  by TH  
=  $P(Y_1 = y | X_0 = i)$ 

- note for z > y

$$P(Y_1 = y, Y_2 = z \mid X_0 = j)$$
  
=  $P(Y_2 = z \mid Y_1 = y, X_0 = j)P(Y_1 = y \mid X_0 = j)$  and

$$P(Y_{2} = z \mid Y_{1} = y, X_{0} = j)$$

$$= \begin{array}{l} P(X_{y+z} = i, X_{m} \neq i, y+1 \leq m \leq z+y-1 \mid X_{y} = i, X_{m} \neq i, 1 \leq m \leq y-1, X_{0} = j) \\ = P(X_{y+z} = i, X_{m} \neq i, y+1 \leq m \leq z+y-1 \mid X_{y} = i) \text{ by MP} \\ = P(X_{z} = i, X_{m} \neq i, 1 \leq m \leq z-1 \mid X_{0} = i) \text{ by TH} \\ = P(Y_{1} = z \mid X_{0} = i) \end{array}$$

so in general  $Y_1, Y_2, Y_3, \ldots$  are mutually stat. ind. and  $Y_2, Y_3, \ldots$  are *i.i.d.*  $P(Y_i = z \mid X_0 = i) = P(Y_1 = z \mid X_0 = i)$ 

- consider  $N_t/t$  = renewal rate

**Proposition VI.31** (Elementary Renewal Theorem) For a renewal process  $\{N_t : t \in T\}$  with  $E(Y_n) = \mu < \infty$  for  $n \ge 2$ , then as  $t \to \infty$ 

(i) 
$$\frac{N_t}{t} \xrightarrow{wp1} \frac{1}{\mu}$$
, (ii)  $E\left(\frac{N_t}{t}\right) \to \frac{1}{\mu}$ .

Proof: (i) By the SLLN  $\frac{1}{n}T_n = \frac{1}{n}\sum_{i=1}^n Y_i \xrightarrow{wp_1} \mu$  which implies  $\lim_{t\to\infty} N_t = \lim_{t\to\infty} \#\{n: T_n \leq t\} = \infty$  wp1 (otherwise  $T_n = \infty$  infinitely often which contradicts  $\mu < \infty$ ). Therefore,

$$\lim_{t\to\infty}\frac{T_{N_t}}{N_t}=\lim_{n\to\infty}\frac{T_n}{n}=\mu \text{ wp1}$$

and since

(ii) fact. 🔳

$$\frac{T_{N_t}}{N_t} \le \frac{t}{N_t} \le \frac{T_{N_t+1}}{N_t}$$

this implies  $t/N_t \xrightarrow{wp1} \mu$ .

**Example VI.10** Renewal process from a Markov chain (continued)

- recall  $T_n(i) = \min\{k > T_n(i) : X_k = i\} =$  time of *n*-th visit to state *i* 

- when the chain is positive recurrent then  $\mu = m_i$  mean recurrence time  $< \infty$  and recall the unique stationary distribution is given by  $\pi_i = 1/m_i$ 

**Proposition VI.32** (Blackwell Renewal Theorem) For a renewal process with  $\mu < \infty$  and s.t.  $Y_2$  is non-arithmetic (there isn't  $\delta > 0$  s.t.  $P(Y_2 \in \{i\delta : i \in \mathbb{Z}\}) = 1$ ) then  $\lim_{t\to\infty} E(N_{t+h} - N_t) = h/\mu$  for any h > 0.

Proof: fact

**Corollary VI.33** If h > 0 is s.t.  $P(Y_2 < h) = 0$ , then  $\lim_{t\to\infty} P(\exists n \text{ s.t. } t < T_n \le t + h) = h/\mu$ .

Proof: We have

$$P(\exists n \text{ s.t. } t < T_n \le t+h) = P(N_{t+h} - N_t \ge 1) \\ = P(N_{t+h} - N_t = 1) = E(N_{t+h} - N_t)$$

whence the result follows from the proposition.  $\blacksquare$ 

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- suppose at renewal *i* there is a reward (or cost)  $R_i$  where  $R_1, R_2, \ldots$  are *i.i.d.* and define the *renewal reward process* by

$$R_t = \sum_{i=1}^{N_t} R_i$$
 = total reward up to time  $t$ 

**Proposition VI.34** (Renewal Reward Theorem) For a renewal process with  $\mu < \infty$ , then  $\frac{R_t}{t} \xrightarrow{wp1} \frac{E(R_1)}{\mu}$ . Proof: We have

$$rac{\mathsf{R}_t}{t} = rac{1}{t}\sum_{i=1}^{N_t} \mathsf{R}_i = rac{\mathsf{N}_t}{t}\left(rac{1}{\mathsf{N}_t}\sum_{i=1}^{N_t} \mathsf{R}_i
ight)$$

and by the SLLN

$$\lim_{t\to\infty}\frac{1}{N_t}\sum_{i=1}^{N_t}R_i=\lim_{n\to\infty}\frac{1}{n}\sum_{i=1}^nR_i=E(R_1).$$

Exercise VI.14 Text 4.5.4

Exercise VI.15 Text 4.5.6

Exercise VI.16 Text 4.6.9

Exercise VI.17 Text 4.6.10

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