

MATH180C: Introduction to Stochastic Processes II

<https://mathweb.ucsd.edu/~ynemish/teaching/180c>

Today: Hitting probabilities.
Absorption times. General CTMC.
Matrix exponentials
Next: PK 6.5, 6.6, Durrett 4.1

Week 2:

- HW1 due Friday, April 14 on Gradescope
- Important: Midterm 1 will take place on Friday, April 28

Stopping times

Def (Informal). Let $(X_t)_{t \geq 0}$ be a stochastic process and let $T \geq 0$ be a random variable. We call T a **stopping time** if the event

$$\{T \leq t\}$$

can be determined from the knowledge of the process up to time t (i.e., from $\{X_s : 0 \leq s \leq t\}$)

Examples: Let $(X_t)_{t \geq 0}$ be right-continuous

1. $\min\{t \geq 0 : X_t = i\}$ is a stopping time
2. W_k is a stopping time
3. $\sup\{t \geq 0 : X_t = i\}$ is not a stopping time

Strong Markov property

Theorem (no proof)

Let $(X_t)_{t \geq 0}$ be a MC, let T be a stopping time of $(X_t)_{t \geq 0}$. Then, conditional on $T < \infty$ and $X_T = i$,

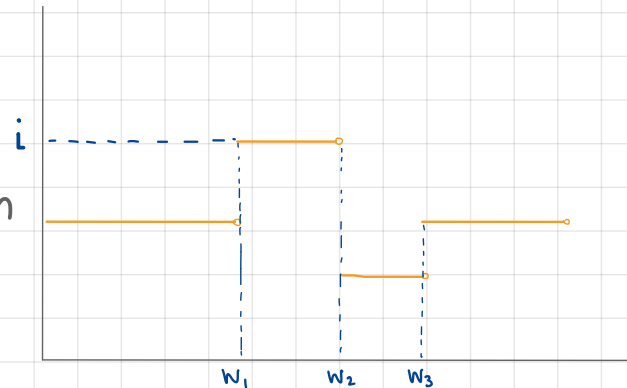
$$(X_{T+t})_{t \geq 0}$$

(i) is independent of $\{X_s, 0 \leq s \leq T\}$

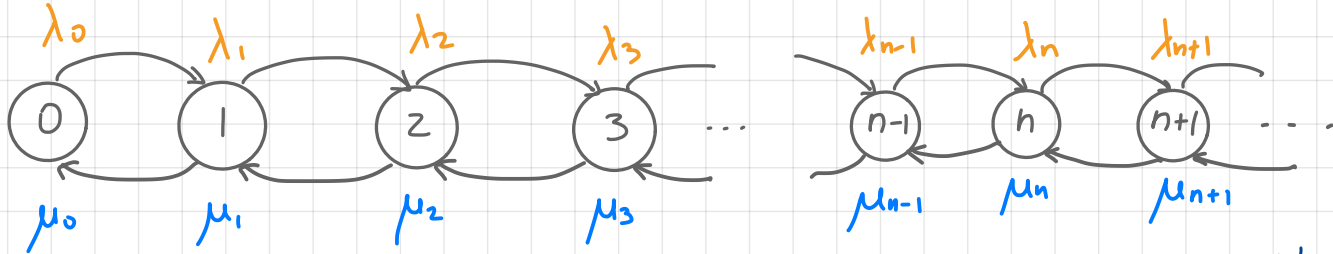
(ii) has the same distribution as $(X_t)_{t \geq 0}$ starting from i .

Example

$(X_{W_i+t})_{t \geq 0}$ has the same distribution as $(X_t)_{t \geq 0}$ conditioned on $X_0 = i$ and is indep. of what happened before



Alternative (jump and hold) characterization



$$\lambda = \mu = 1 \quad \lambda' = \mu' = 2$$

Sojourn times S_k are independent, $\frac{\lambda}{\lambda + \mu} = \frac{\lambda'}{\lambda' + \mu'} = \frac{1}{2}$

$$\text{Exp}(2) \quad \text{Exp}(4)$$

Each transition has two parts

- wait in state i for time $\sim \text{Exp}(\lambda_i + \mu_i)$
- then choose where to go:

go $\rightarrow (i+1)$ with probability $\frac{\lambda_i}{\lambda_i + \mu_i}$

go $\leftarrow (i-1)$ with probability $\frac{\mu_i}{\lambda_i + \mu_i}$

Alternative (jump and hold) characterization

"Proof" cont.

$$G_i(t) = P(S_k > t \mid X_{W_k} = i)$$

$$G_i'(t) = -(\lambda_i + \mu_i) G_i(t), \quad G_i(0) = 1$$

$$\hookrightarrow G_i(t) = e^{-(\lambda_i + \mu_i)t} = P(S_k > t \mid X_{W_k} = i)$$

✓ $\hookrightarrow S_k \sim \text{Exp}(\lambda_i + \mu_i)$ (given that the process sojourns in i)

Suppose the process waits $\text{Exp}(\lambda_i + \mu_i)$, then

jumps to $i+1$ with probability $\lambda_i / (\lambda_i + \mu_i)$

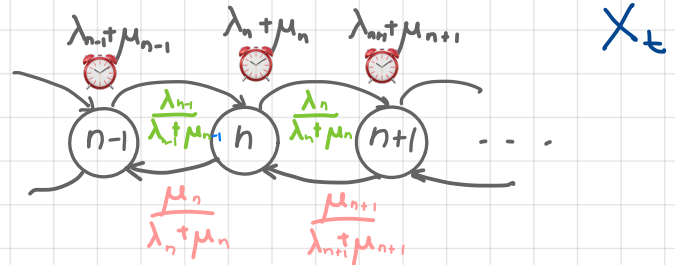
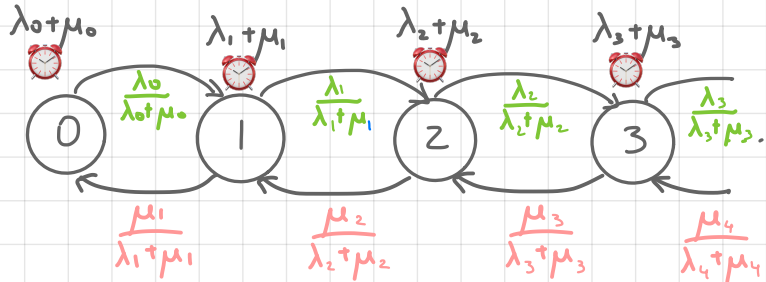
to $i-1$ with probability $\mu_i / (\lambda_i + \mu_i)$

$$P_{i,i+1}(h) = P(S_{\circ} \leq h, \text{jump to } i+1 \mid X_{\circ} = i) + o(h)$$

$$= (1 - e^{-(\lambda_i + \mu_i)h}) \frac{\lambda_i}{\lambda_i + \mu_i} + o(h) = (1 - (1 - (\lambda_i + \mu_i)h)^{+o(h)}) \frac{\lambda_i}{\lambda_i + \mu_i} + o(h) = \lambda_i h + o(h) \quad \checkmark$$

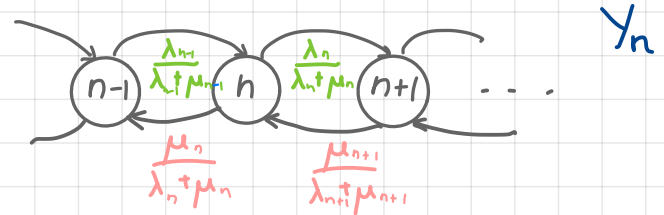
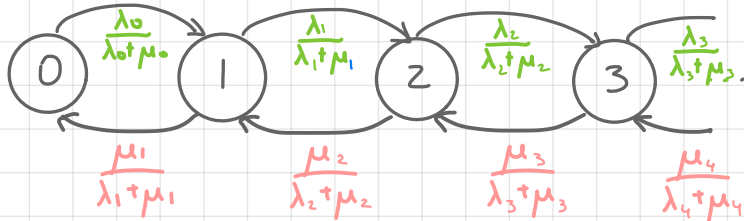
$$P_{i,i-1}(h) = (\lambda_i + \mu_i)h \cdot \frac{\mu_i}{\lambda_i + \mu_i} + o(h) = \mu_i h + o(h)$$

Related discrete time MC.



X_t

Def. Let $(X_t)_{t \geq 0}$ be a continuous time MC, let $W_n, n \geq 0$, be the corresponding waiting (arrival, jump) times. Then we call $(Y_n)_{n \geq 0}$ defined by $Y_n = X_{W_n}$ the jump chain of $(X_t)_{t \geq 0}$.



Y_n

random walk

Absorption probabilities for B&D processes

Let $(X_t)_{t \geq 0}$ be a birth and death process, and assume that the state 0 is absorbing, $\lambda_0 = 0$. Then

$$\begin{aligned} P((X_t)_{t \geq 0} \text{ gets absorbed in } 0 \mid X_0 = i) \\ = P((Y_n)_{n \geq 0} \text{ gets absorbed in } 0 \mid Y_0 = i) \end{aligned}$$

↳ use the first step analysis to compute the absorption probabilities for $(Y_n)_{n \geq 0}$ (and for $(X_t)_{t \geq 0}$)

Denote $u_i = P((Y_n) \text{ is absorbed in } 0 \mid Y_0 = i)$

Then $u_0 = 1$, $u_n = \frac{\mu_n}{\lambda_n + \mu_n} \cdot u_{n-1} + \frac{\lambda_n}{\lambda_n + \mu_n} \cdot u_{n+1}$

Absorption probabilities for B&D processes

$$u_0 = 1, \quad u_n = \frac{\mu_n}{\lambda_n + \mu_n} u_{n-1} + \frac{\lambda_n}{\lambda_n + \mu_n} u_{n+1}$$

Rewrite $(\lambda_n + \mu_n) u_n = \mu_n u_{n-1} + \lambda_n u_{n+1}$

$$\lambda_n (u_{n+1} - u_n) = \mu_n (u_n - u_{n-1})$$

$$u_{n+1} - u_n = \frac{\mu_n}{\lambda_n} (u_n - u_{n-1})$$

$$= \underbrace{\frac{\mu_n}{\lambda_n} \cdot \frac{\mu_{n-1}}{\lambda_{n-1}} \cdots \frac{\mu_1}{\lambda_1}}_{\rho_n} (u_1 - u_0)$$

$$(*) \quad u_{n+1} - u_n = \rho_n (u_1 - 1)$$

Note that $\sum_{k=1}^{n-1} (u_{k+1} - u_k) = u_n - u_1 = (u_1 - 1) \sum_{k=1}^{n-1} \rho_k \quad (**)$

If $\sum_{n=1}^{\infty} \rho_n = \infty$, then $u_1 = 1$ and from (*) $u_n = 1 \quad \forall n \geq 0$.

Absorption probabilities for B&D processes

Let $\sum_{k=1}^{\infty} p_k < \infty$. We are looking for the **minimal** solution

that satisfies $u_n \in [0, 1] \forall n$. We rewrite (***) as

$$u_n = u_1 + (u_1 - 1) \sum_{k=1}^{n-1} p_k = 1 + (u_1 - 1) \left(1 + \sum_{k=1}^{n-1} p_k \right)$$

Choose smallest $u_1 \in [0, 1]$ for which $1 + (u_1 - 1) \left(1 + \sum_{k=1}^{n-1} p_k \right) \geq 0 \forall n$

$$u_1 \geq 1 - \frac{1}{1 + \sum_{k=1}^{n-1} p_k} \geq 1 - \frac{1}{1 + \sum_{k=1}^{\infty} p_k} \quad , \quad u_1 = \frac{\sum_{k=1}^{\infty} p_k}{1 + \sum_{k=1}^{\infty} p_k}$$

$$u_n = \frac{\sum_{k=n}^{\infty} p_k}{1 + \sum_{k=1}^{\infty} p_k}$$