

Generator:  $(P_h)_{ij} \triangleq \mathbb{P}(Y_n=j | Y_0=i)$  for time-homogeneous CTMC  $\{Y_t\}$ , is called the semi-group operator

$$\text{check: } \begin{cases} P_{h_1} P_{h_2} = P_{h_1+h_2} = P_{h_2} P_{h_1} \\ P_0 = I \end{cases}$$

$$\text{generator } G \triangleq \lim_{h \rightarrow 0} \frac{P_h - I}{h} \quad \left. \frac{d}{dt} P_t \right|_{t=0}$$

$$\text{Calculation: } G_{ij} = \begin{cases} -q_i & \text{if } i=j \\ q_i p_{ij} & \text{else} \end{cases}$$

provides "rate" interpretation of CTMC!

$$\begin{cases} P_{ii}(h) = 1 + \underbrace{G_{ii}}_{\text{neg flow rate out of } i} h + o(h) \\ \forall i \neq j, P_{ij}(h) = \underbrace{G_{ij}}_{\text{flow rate } i \rightarrow j} h + o(h) \end{cases} \quad (h \rightarrow 0)$$

$$\text{so } G_{ii} + \sum_{j \neq i} G_{ij} = 0 \quad \text{for } \forall i$$

(row sums up to 0)

Recover  $P_t$  from  $G$ ?

$$\begin{cases} P_t' = G P_t = P_t G \quad (\text{forward/backward eqn}) \\ P_0 = I \end{cases}$$

$G$  and  $P$  has 1-to-1 correspondence, uniquely characterizes CTMC.

exponential analogue:  $\left\{ \begin{array}{l} y' = ay \Rightarrow y = c \cdot e^{at} \\ P_t' = G P_t \Rightarrow P_t = e^{tG} \end{array} \right.$

What happens if MC has cts state?

Then matrix dimension becomes uncountable, instead of a seq of matrix  $P_t$ , we have a seq of function  $p(t, x)$ , forward/backward eqn becomes PDE!

↓  
space variable

eg:  $\lambda, \mu > 0$ , MC on  $\{1, 2\}$ ,  $G = \begin{pmatrix} -\mu & \mu \\ \lambda & -\lambda \end{pmatrix}$

(a): Forward eqn, solve for  $P_{ij}(t)$

Def:  $P_t = \begin{pmatrix} P_{11}(t) & P_{12}(t) \\ P_{21}(t) & P_{22}(t) \end{pmatrix}$ ,  $P_t' = G P_t$  writes

$$\begin{cases} P_{11}'(t) = -\mu P_{11}(t) + \lambda P_{12}(t) \\ P_{12}'(t) = \mu P_{11}(t) - \lambda P_{12}(t) \\ P_{21}'(t) = -\mu P_{21}(t) + \lambda P_{22}(t) \\ P_{22}'(t) = \mu P_{21}(t) - \lambda P_{22}(t) \end{cases} \quad \text{with } P_{ij}(0) = \delta_{ij}$$

Since  $P_{11}(t) + P_{12}(t) = 1$ ,  $\begin{cases} P_{11}'(t) = -(\lambda + \mu) P_{11}(t) + \lambda \\ P_{11}(0) = 1 \end{cases}$

$$\Downarrow \\ P_{11}(t) = \frac{\lambda}{\lambda + \mu} + \frac{\mu}{\lambda + \mu} e^{-(\lambda + \mu)t}$$

$$\Downarrow \\ P_{12}(t) = \frac{\mu}{\lambda + \mu} - \frac{\mu}{\lambda + \mu} e^{-(\lambda + \mu)t}$$

Similarly,  $P_{21}(t) = \frac{\lambda}{\lambda + \mu} - \frac{\lambda}{\lambda + \mu} e^{-(\lambda + \mu)t}$

$$P_{22}(t) = \frac{\mu}{\lambda + \mu} + \frac{\lambda}{\lambda + \mu} e^{-(\lambda + \mu)t}$$

(b): Calculate  $G^n$  and  $\sum_{n=0}^{\infty} \frac{t^n}{n!} G^n$ , compare to part (a).

pf:

Def of  $e^{tG}$  for operator  $G$

$G$  has eigenval 0 and  $-(\lambda+\mu)$ , with eigvec

$$\begin{pmatrix} 1 \\ 1 \end{pmatrix} \text{ and } \begin{pmatrix} \mu \\ -\lambda \end{pmatrix}$$

$$\text{So } G = P D P^{-1}, \quad P = \begin{pmatrix} 1 & \mu \\ 1 & -\lambda \end{pmatrix}, \quad D = \begin{pmatrix} 0 & 0 \\ 0 & -(\lambda+\mu) \end{pmatrix}$$

$$P^{-1} = \begin{pmatrix} \frac{\lambda}{\lambda+\mu} & \frac{\mu}{\lambda+\mu} \\ \frac{1}{\lambda+\mu} & -\frac{1}{\lambda+\mu} \end{pmatrix}$$

$$G^n = P D^n P^{-1} = [-(\lambda+\mu)]^n \cdot \begin{pmatrix} \frac{\mu}{\lambda+\mu} & -\frac{\mu}{\lambda+\mu} \\ -\frac{\lambda}{\lambda+\mu} & \frac{\lambda}{\lambda+\mu} \end{pmatrix} \quad (n \geq 1)$$

$$e^{tG} = \sum_{n=1}^{\infty} \frac{[-(\lambda+\mu)t]^n}{n!} \begin{pmatrix} \frac{\mu}{\lambda+\mu} & -\frac{\mu}{\lambda+\mu} \\ -\frac{\lambda}{\lambda+\mu} & \frac{\lambda}{\lambda+\mu} \end{pmatrix} + I$$

$$= \begin{pmatrix} \frac{\mu}{\lambda+\mu} & -\frac{\mu}{\lambda+\mu} \\ -\frac{\lambda}{\lambda+\mu} & \frac{\lambda}{\lambda+\mu} \end{pmatrix} \cdot e^{-(\lambda+\mu)t} + I -$$

$$\begin{pmatrix} \frac{\mu}{\lambda+\mu} & -\frac{\mu}{\lambda+\mu} \\ -\frac{\lambda}{\lambda+\mu} & \frac{\lambda}{\lambda+\mu} \end{pmatrix}$$

$$= \begin{pmatrix} \frac{\mu}{\lambda+\mu} & -\frac{\mu}{\lambda+\mu} \\ -\frac{\lambda}{\lambda+\mu} & \frac{\lambda}{\lambda+\mu} \end{pmatrix} \cdot e^{-(\lambda+\mu)t} + \begin{pmatrix} \frac{\lambda}{\lambda+\mu} & \frac{\mu}{\lambda+\mu} \\ \frac{\lambda}{\lambda+\mu} & \frac{\mu}{\lambda+\mu} \end{pmatrix}$$

$$= P_t \quad \checkmark$$



(c): Find stat dist  $\pi$ . Check  $P_{ij}(t) \rightarrow \pi_j (t \rightarrow \infty)$

Pf: By def,  $\forall t, \pi P_t = \pi, \pi(P_t - I) = 0,$

so divide by  $t$  and set  $t \rightarrow 0$  to get

$$\underline{\pi G = 0}$$

$$\Downarrow$$
$$\pi = \begin{pmatrix} \frac{\lambda}{\lambda + \mu} & \frac{\mu}{\lambda + \mu} \end{pmatrix}$$

and  $P_{ij}(t) \rightarrow \pi_j (t \rightarrow \infty) \checkmark$

Consistent with the ergodic thm.

(d): Calculate  $IP(X_t=2 | X_0=1, X_{3t}=1)$

Pf:

$$= \frac{IP(X_{3t}=1 | X_0=1, X_t=2) \cdot IP(X_t=2 | X_0=1)}{IP(X_{3t}=1 | X_0=1)}$$

Markov

$$= \frac{P_{2,1}(2t) \cdot P_{1,2}(t)}{P_{1,1}(3t)}$$

use (a)

$$= \frac{\left[ \frac{\lambda}{\lambda + \mu} - \frac{\lambda}{\lambda + \mu} e^{-(\lambda + \mu)2t} \right] \left[ \frac{\mu}{\lambda + \mu} - \frac{\mu}{\lambda + \mu} e^{-(\lambda + \mu)t} \right]}{\frac{\lambda}{\lambda + \mu} + \frac{\mu}{\lambda + \mu} e^{-(\lambda + \mu)3t}}$$

## Applications of CTMC:

e.g.  $\langle N_t \rangle$  is Poisson process (intensity  $\lambda$ ),  $N_0 = 0$ ,  $\phi(t, z) \triangleq \mathbb{E} z^{N_t}$  is g.f. of  $N_t$ , find an integral equation for  $\phi(t, z)$  and check that  $\phi(t, z) = e^{\lambda t(z-1)}$  is the solution.

Def: Poisson process is CTMC, pure birth process with birth rate  $\lambda_i = \lambda$  at any state  $i$ .

By **first step decomposition**, let  $S_1$  be the first time a state transition happens.

$$\phi(t, z) = \mathbb{E} [\mathbb{E}(z^{N_t} | S_1)]$$

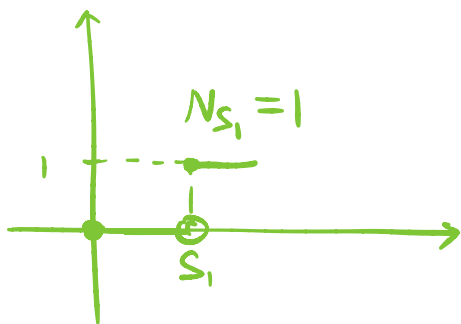
$$\text{with } \mathbb{E}(z^{N_t} | S_1 = s)$$

$$= \mathbb{E}(z^{N_s} \cdot z^{N_t - N_s} | S_1 = s)$$

$$= z \cdot \mathbb{E}(z^{N_t - N_s} | S_1 = s)$$

Markov property,  $t-s$  time left,  
a single particle at time  $S_1$

$$= z \cdot \mathbb{E} z^{N_{t-s}} = z \cdot \phi(t-s, z) \quad \text{if } s < t$$



$$S_0: \mathbb{E}(z^{N_t} | S_1) = z \cdot \phi(t - S_1, z) \text{ if } S_1 < t$$

When taking expectation w.r.t.  $S_1$ , notice that  $S_1$  may not be well-defined since  $S_1 \sim \mathcal{E}(\lambda)$  and it's possible that no transition has happened up to time  $t$ .

$$\begin{aligned} S_0 \quad \underline{\phi(t, z)} &= \mathbb{E}(z^{N_t} \cdot \underbrace{I_{\{S_1 > t\}}}_{\substack{\text{under this} \\ \text{event, } N_t = 0}}) + \mathbb{E}(z^{N_t} \cdot I_{\{S_1 < t\}}) \\ &= \mathbb{P}(S_1 > t) + z \cdot \mathbb{E}[\phi(t - S_1, z) \cdot I_{\{S_1 < t\}}] \\ &= \underline{e^{-\lambda t} + z \cdot \int_0^t \phi(t - s, z) \cdot \lambda e^{-\lambda s} ds} \end{aligned}$$

provides an integral eqn for  $\phi$ .

Check:  $\phi(t, z) = e^{\lambda t(z-1)}$  is solution.

$$\begin{aligned} \text{RHS} &= e^{-\lambda t} + z \cdot e^{\lambda(z-1)t} \cdot \int_0^t \lambda \cdot e^{-\lambda s z} ds \\ &= e^{-\lambda t} + z \cdot e^{\lambda(z-1)t} \cdot \frac{1}{z} (1 - e^{-\lambda t z}) \\ &= e^{\lambda(z-1)t} = \phi(t, z) \quad \checkmark \end{aligned}$$

RMK: Why last step decomposition does not work?

Let  $T$  be last transition time before  $t$ , condition on  $T$  brings difficulty since  $N_T$  is unknown except  $N_T + 1 = N_t$  while in first time decamp  $\sim N_{S_1} = N_0 + 1 = 1$ .  
↓                      ↓  
both random

e.g:  $(X_t)$  is cts-time BDC on  $\mathbb{N}$  with birth rate  $\lambda_n$  death rate  $\mu_n$ , assume  $X_0 = i$ , let  $T_{i+1}$  be first hitting time to  $i+1$ . Derive a recursive formula calculating

$$m_i \triangleq \mathbb{E}_i T_{i+1}.$$

pf: Still use **first step decomposition**, let  $S_1$  be the first transition time,  $S_1 \sim \mathcal{E}(\lambda_i + \mu_i)$   
↑  
since  $X_0 = i$

$$m_i = \mathbb{E}_i T_{i+1} = \mathbb{E}_i [\mathbb{E}_i(T_{i+1} | S_1)]$$

where  $\mathbb{E}_i(T_{i+1} | S_1 = s)$   $X_{S_1}$  could take value  $i-1$  or  $i+1$  w.p.  $\frac{\mu_i}{\lambda_i + \mu_i}$  or  $\frac{\lambda_i}{\lambda_i + \mu_i}$

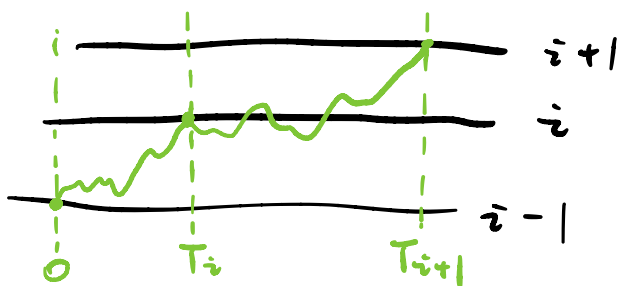
$$= \mathbb{P}_i(N_{S_1} = i+1 | S_1 = s) \cdot \mathbb{E}_i(T_{i+1} | S_1 = s, N_{S_1} = i+1) + \mathbb{P}_i(N_{S_1} = i-1 | S_1 = s) \cdot \mathbb{E}_i(T_{i+1} | S_1 = s, N_{S_1} = i-1)$$

$$= \frac{\lambda_i}{\lambda_i + \mu_i} \cdot s + \frac{\mu_i}{\lambda_i + \mu_i} \cdot (s + \mathbb{E}_{i-1} T_{i+1})$$

*Markov property*

$$= s + \frac{\mu_i}{\lambda_i + \mu_i} \mathbb{E}_{i-1} T_{i+1} \implies \text{can't represent using } \{m_k\} \text{ need further simplification!}$$

Structure of BDC implies that if hits  $\bar{z}+1$  starting from  $\bar{z}-1$ , one must first hit  $\bar{z}$ .



$$IE_{\bar{z}-1} T_{\bar{z}+1} = IE_{\bar{z}-1} T_{\bar{z}} + IE_{\bar{z}-1} (T_{\bar{z}+1} - T_{\bar{z}})$$

$$= IE_{\bar{z}-1} T_{\bar{z}} + IE_{\bar{z}-1} \left[ IE_{\bar{z}-1} (T_{\bar{z}+1} - T_{\bar{z}} | T_{\bar{z}}) \right]$$

$$= IE_{\bar{z}-1} T_{\bar{z}} + IE_{\bar{z}-1} \left[ \underbrace{IE_{\bar{z}-1} (T_{\bar{z}+1} | T_{\bar{z}})}_{\text{Strong Markov property}} - T_{\bar{z}} \right]$$

$$= IE_{\bar{z}-1} T_{\bar{z}} + IE_{\bar{z}} T_{\bar{z}+1}$$

Strong Markov property

$$= \underbrace{T_{\bar{z}}}_{\text{already waited}} + \underbrace{IE_{\bar{z}} T_{\bar{z}+1}}_{\text{to wait in the future}}$$

$$S_0: IE_{\bar{z}} (T_{\bar{z}+1} | S_1 = s) = s + \frac{\mu_{\bar{z}}}{\lambda_{\bar{z}} + \mu_{\bar{z}}} (m_{\bar{z}-1} + m_{\bar{z}})$$

$$m_{\bar{z}} = IE S_1 + \frac{\mu_{\bar{z}}}{\lambda_{\bar{z}} + \mu_{\bar{z}}} (m_{\bar{z}-1} + m_{\bar{z}})$$

⇓

$$\begin{cases} m_{\bar{z}} = \frac{1}{\lambda_{\bar{z}}} + \frac{\mu_{\bar{z}}}{\lambda_{\bar{z}}} m_{\bar{z}-1} \\ m_0 = \frac{1}{\lambda_0} \end{cases}$$

eg:  $\begin{cases} \text{sunny} \rightarrow \text{wildfire occurs with rate } 0.5 \text{ per day} \\ \text{cloudy} \rightarrow \text{-----} \text{-----} 0.1 \text{ -----} \end{cases}$  (Poisson)

weather: two-state CTMC, sunny lasting on average 2 days, cloudy lasts on average 1 day. Wildfire occurs at time  $T_1, T_2, \dots$ , as arrival times of a cts-time counting process and weather indep of wildfire occurring.

$F_t \triangleq$  # of fires up to time  $t$ .

(1): Compute  $\lim_{t \rightarrow \infty} \frac{EF_t}{t}$

Qf: Modelling:  $S_t$  is weather at time  $t$ , state space  $\{s, c\}$ , only need to specify holding rates  $q_s$  and  $q_c$ , clearly,

$$\frac{1}{q_s} = 2, \quad \frac{1}{q_c} = 1, \quad \text{so } q_s = \frac{1}{2}, \quad q_c = 1.$$

$\{S_t\}$  irreducible, by ergodic thm,  $S_t \xrightarrow{d} \pi$  ( $t \rightarrow \infty$ ) with  $\pi$  as stat dist.

Write out generator matrix  $G = \begin{matrix} & \begin{matrix} s & c \end{matrix} \\ \begin{matrix} s \\ c \end{matrix} & \begin{pmatrix} -\frac{1}{2} & \frac{1}{2} \\ 1 & -1 \end{pmatrix} \end{matrix}$

so  $\pi G = 0 \quad \Rightarrow \quad \pi = \left(\frac{2}{3}, \frac{1}{3}\right)$

When  $t \rightarrow \infty$ ,  $\frac{2}{3}$  days are sunny,  $\frac{1}{3}$  days are cloudy.

So: as  $t \rightarrow \infty$ ,  $IEF_t \sim \frac{2}{3} \cdot \underline{(0.5t)} + \frac{1}{3} \cdot \underline{(0.1t)}$

PP with  $\lambda = 0.5$

PP with

$F_t \sim \mathcal{O}(0.5t)$

$\lambda = 0.1$

$IEF_t = 0.5t$

$$\lim_{t \rightarrow \infty} \frac{IEF_t}{t} = \frac{2}{3} \cdot 0.5 + \frac{1}{3} \cdot 0.1 = \frac{11}{30}.$$

(2): Explain why  $X_t = (S_t, F_t)$  is Markov and write down its generator matrix.

Pf: Given  $X_t = (s, n)$ , next transition can only be  $\rightarrow (c, n)$ , depends on which happens first.  
 $\rightarrow (s, n+1)$

Weather  $s$  has holding time  $\sim E(\tau_s)$ , fire number  $n$  has holding time  $\sim E(0.5)$  when weather is sunny, so holding time of  $(s, n)$  is minimum of both,

$$\sim E(\tau_s + 0.5)$$

it's still exponentially dist, and different holding times are independent, so it's still CTMC.

Intuitively,  $\{F_t\}$  is Poisson process with  $S_t$ -dependent intensity, forming as tuple provides Markov prop.



In addition,  $\begin{cases} q_{(s,n)} = q_s + 0.5 = 1 \\ q_{(c,n)} = q_c + 0.1 = 1.1 \end{cases}$  are holding rates of  $\{X_t\}$ .

Still need to figure out the transition prob of the underlying discrete-time MC:

$$\begin{cases} P_{(s,n), (c,n)} = \frac{q_s}{q_{(s,n)}} = \frac{1}{2} \\ P_{(s,n), (s,n+1)} = 1 - \frac{1}{2} = \frac{1}{2} \\ P_{(c,n), (s,n)} = \frac{q_c}{q_{(c,n)}} = \frac{1.0}{1.1} \\ P_{(c,n), (c,n+1)} = 1 - \frac{1.0}{1.1} = \frac{1}{11} \end{cases}$$

By def of generator,

$$G^X = \begin{matrix} (s,0) \\ (s,1) \\ \vdots \\ (c,0) \\ (c,1) \\ \vdots \end{matrix} \begin{pmatrix} (s,0) & (s,1) & (s,2) & \dots & (c,0) & (c,1) & (c,2) & \dots \\ -1 & \frac{1}{2} & 0 & \dots & \frac{1}{2} & 0 & 0 & \dots \\ 0 & -1 & \frac{1}{2} & \dots & 0 & \frac{1}{2} & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots \\ 1 & 0 & 0 & \dots & -1.1 & 0.1 & 0 & \dots \\ 0 & 1 & 0 & \dots & 0 & -1.1 & 0.1 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

(3): If it's sunny now, find expected time until next fire  $E_{(s,0)} T_1$ .

Df: First step decomposition, condition on first state transition time  $S_1$  of  $\{X_t\}$ .

$$E_{(s,0)} T_1 = E_{(s,0)} [E_{(s,0)}(T_1 | S_1)]$$

$$\begin{aligned} \text{where } E_{(s,0)}(T_1 | S_1 = s) &= \underbrace{\frac{1}{2}}_{P(s,0,(s,1))} \cdot s + \underbrace{\frac{1}{2}}_{P(s,0,(c,0))} \cdot (s + E_{(c,0)} T_1) \\ &= s + \frac{1}{2} E_{(c,0)} T_1 \end{aligned}$$

So:  $E_{(s,0)} T_1 = 1 + \frac{1}{2} E_{(c,0)} T_1$   $S_1 \sim \mathcal{E}(\lambda_{(s,0)}) = \mathcal{E}(1)$   
if  $X_0 = (s, 0)$

On the other hand,

$$E_{(c,0)} T_1 = E_{(c,0)} [E_{(c,0)}(T_1 | S_1)]$$

$$\begin{aligned} \text{where } E_{(c,0)}(T_1 | S_1 = s) &= \underbrace{\frac{1}{11}}_{P(c,0,(c,1))} \cdot s + \underbrace{\frac{10}{11}}_{P(c,0,(s,0))} (s + E_{(s,0)} T_1) \\ &= s + \frac{10}{11} E_{(s,0)} T_1 \end{aligned}$$

So:  $E_{(c,0)} T_1 = \frac{1}{1.1} + \frac{10}{11} E_{(s,0)} T_1$   $S_1 \sim \mathcal{E}(\lambda_{(c,0)}) = \mathcal{E}(1.1)$   
if  $X_0 = (c, 0)$

$$\Rightarrow \begin{cases} E_{(s,0)} T_1 = \boxed{\frac{8}{3}} \rightarrow \text{answer} \\ E_{(c,0)} T_1 = \frac{10}{3} \end{cases}$$

In long term, avg rate of fire  $\frac{11}{30}$  so expected waiting time until next fire  $\approx \frac{30}{11}$ , notice that

$$\underbrace{\frac{8}{3}}_{\text{sunny}} < \underbrace{\frac{30}{11}}_{\text{long-term avg}} < \underbrace{\frac{10}{3}}_{\text{cloudy}}.$$