

Generator:  $(P_h)_{ij} \triangleq \text{IP}(Y_h=j \mid Y_0=i)$  for time-homo

CTMC  $\{Y_t\}$ , is called the semi-group operator

check:  $\begin{cases} P_h P_{h_2} = P_{h+h_2} = P_{h_2} P_h \\ P_0 = I \end{cases}$

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

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

provides "rate" interpretation of CTMC!

neg flow rate out of  $i$

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

so  $G_{ii} + \sum_{j \neq i} G_{ij} = 0$  for  $\forall i$   
 (row sums up to 0)

Recover  $P_t$  from  $G$ ?

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

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

exponential analogue:

$$\left\{ \begin{array}{l} y' = ay \Rightarrow y = C \cdot e^{at} \\ P'_t = GP_t \Rightarrow P_t = e^{tG} \end{array} \right.$$

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

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

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

Def:  $P_t = \begin{pmatrix} P_{11}(t) & P_{12}(t) \\ P_{21}(t) & P_{22}(t) \end{pmatrix}, \quad P'_t = GP_t \text{ unites}$

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

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

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

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

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).

Def:

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

$G$  has eigenvalues  $0$  and  $-(\lambda + \mu)$ , with eigenvectors  
 $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$  and  $\begin{pmatrix} \mu \\ -\lambda \end{pmatrix}$

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

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

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

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

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

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

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

$= P_t$

(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

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



$$\pi = \left( \frac{\lambda}{\lambda+\mu} \quad \frac{\mu}{\lambda+\mu} \right)$$

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

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)

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

## Applications of CTMC:

e.g:  $\langle N_t \rangle$  is Poisson process,  $N_0 = 0$ ,  $\phi(t, z) \triangleq \mathbb{E} z^{N_t}$  (Intensity  $\lambda$ )  
 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.

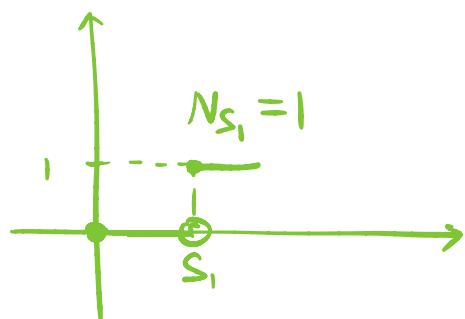
$$\phi(t, z) = \mathbb{E} [\mathbb{E}(z^{N_t} | S_1)]$$

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

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

$$= z \cdot \underbrace{\mathbb{E}(z^{N_t - N_s} | S_1 = s)}_{\text{Markov property, } t-s \text{ time left, a single particle at time } S_1}$$

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



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

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  $t$ .

$$S_0 \quad \underline{\phi(t, z)} = \mathbb{E}\left(z^{N_t} \cdot I_{\{S_1 \geq t\}}\right) + \mathbb{E}\left(z^{N_t} \cdot I_{\{S_1 < t\}}\right)$$

under this  
event,  $N_t = 0$

$$= \mathbb{P}(S_1 > t) + z \cdot \mathbb{E}[\phi(t - S_1, z) \cdot I_{\{S_1 < t\}}]$$

$$= e^{-\lambda t} + z \cdot \underline{\int_0^t \phi(t-s, z) \cdot \lambda e^{-\lambda s} ds}$$

provides an integral eqn for  $\phi$ .

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

$$\text{RHS} = e^{-\lambda t} + z \cdot e^{\lambda(z-1)t} \cdot \int_0^t \lambda \cdot e^{-\lambda s z} ds$$

$$= e^{-\lambda t} + z \cdot e^{\lambda(z-1)t} \cdot \frac{1}{z} (1 - e^{-\lambda t z})$$

$$= e^{\lambda(z-1)t} = \phi(t, z) \quad \checkmark$$

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 decompos~  $N_0 = N_0 + 1 = 1$ . 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  $i+1$ . Define a recursive formula calculating  $m_i \triangleq \mathbb{E}_i T_{i+1}$ .

Def: Still use **first step decomposition**, let  $S_i$  be the first transition time,  $S_i \sim \mathcal{E}(\lambda_i + \mu_i)$

$\uparrow$   
since  $X_0 = i$

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

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

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

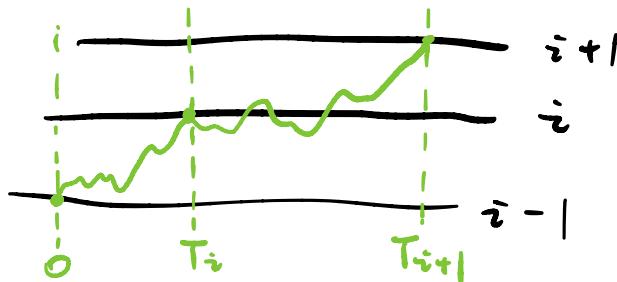
$$\mathbb{P}_i(N_{S_i} = i-1 | S_i = s) \cdot \mathbb{E}_i(T_{i+1} | S_i = s, N_{S_i} = i-1)$$

*Markov property*

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

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

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



$$\begin{aligned}
 \mathbb{E}_{\bar{i}-1} T_{\bar{i}+1} &= \mathbb{E}_{\bar{i}-1} T_i + \mathbb{E}_{\bar{i}-1} (T_{\bar{i}+1} - T_i) \\
 &= \mathbb{E}_{\bar{i}-1} T_i + \mathbb{E}_{\bar{i}-1} [\mathbb{E}_{\bar{i}-1} (T_{\bar{i}+1} - T_i | T_i)] \\
 &= \mathbb{E}_{\bar{i}-1} T_i + \mathbb{E}_{\bar{i}-1} [\mathbb{E}_{\bar{i}-1} (T_{\bar{i}+1} | T_i) - T_i] \\
 &= \mathbb{E}_{\bar{i}-1} T_i + \mathbb{E}_i T_{\bar{i}+1} \quad \text{Strong Markov property} \\
 &= \underbrace{T_i}_{\text{already waited}} + \underbrace{\mathbb{E}_i T_{\bar{i}+1}}_{\text{to wait in the future}}
 \end{aligned}$$

$$S_0: \mathbb{E}_i (T_{\bar{i}+1} | S_1 = s) = s + \frac{\mu_i}{\lambda_i + \mu_i} (m_{\bar{i}-1} + m_i)$$

$$m_i = \mathbb{E} S_1 + \frac{\mu_i}{\lambda_i + \mu_i} (m_{\bar{i}-1} + m_i)$$

↓

$$\begin{cases} m_i = \frac{1}{\lambda_i} + \frac{\mu_i}{\lambda_i} m_{\bar{i}-1} \\ m_0 = \frac{1}{\lambda_0} \end{cases}$$

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

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.

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

$$(1): \text{Compute } \lim_{t \rightarrow \infty} \frac{\mathbb{E} F_t}{t}$$

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.

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

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

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

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

PP with  $\lambda = 0.5$       PP with  $\lambda = 0.1$

$$F_t \sim \mathcal{D}(0.5t) \quad \text{IEF}_t = 0.5t$$

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

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

Def: Given  $X_t = (S, n)$ , next transition can only be  $\xrightarrow{(c, n)}$ , depends on which happens first.  
 $\xrightarrow{(S, n+1)}$

Weather  $s$  has holding time  $\sim \mathcal{E}(\gamma_s)$ , fire number  $n$  has holding time  $\sim \mathcal{E}(0.5)$  when weather is sunny, so holding time of  $(S, n)$  is minimum of both,  
 $\sim \mathcal{E}(\gamma_s + 0.5)$

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 a 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:

$$\left\{ \begin{array}{l} P_{(s,n), (s,n)} = \frac{q_s}{q_{(s,n)}} = \frac{1}{2} \\ P_{(s,n), (s,n+1)} = 1 - \frac{1}{2} = \frac{1}{2} \\ P_{(c,n), (s,n)} = \frac{q_c}{q_{(c,n)}} = \frac{10}{11} \\ P_{(c,n), (c,n+1)} = 1 - \frac{10}{11} = \frac{1}{11} \end{array} \right.$$

By def of generator,

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

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

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

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

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

$$\text{So: } \underline{E_{(S,0)} T_1 = 1 + \frac{1}{2} E_{(C,0)} T_1} \quad S_1 \sim \mathcal{E}(q_{(S,0)}) = \mathcal{E}(1) \quad \text{if } X_0 = (S,0)$$

On the other hand,

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

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

$$\text{So: } \underline{E_{(C,0)} T_1 = \frac{1}{11} + \frac{10}{11} E_{(S,0)} T_1} \quad S_1 \sim \mathcal{E}(q_{(C,0)}) = \mathcal{E}(1.1) \quad \text{if } X_0 = (C,0)$$

$$\Rightarrow \left\{ \begin{array}{l} E_{(S,0)} T_1 = \frac{8}{3} \rightarrow \text{answer} \\ E_{(C,0)} T_1 = \frac{10}{3} \end{array} \right.$$

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

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