

# HOMEWORK I

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(1) a) Since  $p_n(t) = \Pr(X(t) = n)$ , we have  $\sum_{n=0}^{\infty} p_n(t) \leq 1$ .

Let  $S_k$  be the  $k$ -th sojourn time, then  $S_k$  has Exponential ( $\lambda_k$ ) distribution.

$$(\Rightarrow) \sum_{n=0}^{\infty} p_n(t) = 1 \text{ implies } \sum_{n=0}^{\infty} \frac{1}{\lambda_n} = \infty.$$

Suppose by contradiction that  $\sum_{n=0}^{\infty} \frac{1}{\lambda_n} < \infty$ , then  $\mathbb{E}\left[\sum_{n=0}^{\infty} S_n\right] = \sum_{n=0}^{\infty} \frac{1}{\lambda_n} < \infty$

or the expected time for the population to explode is finite. Thus  $0 < \Pr(X(t) = +\infty) = 1 - \sum_{n=0}^{\infty} p_n(t)$  or  $\sum_{n=0}^{\infty} p_n(t) < 1$  (contradiction!).

$$(\Leftarrow) \sum_{n=0}^{\infty} \frac{1}{\lambda_n} = \infty \text{ implies } \sum_{n=0}^{\infty} p_n(t) = 1.$$

We have  $\mathbb{E}\left[\exp\left(-\sum_{n=0}^{\infty} S_n\right)\right] = \prod_{n=0}^{\infty} \mathbb{E}\left[e^{-S_n}\right]$  (by independence)

$$= \prod_{n=0}^{\infty} \frac{\lambda_n}{1 + \lambda_n} \leq \prod_{n=0}^{\infty} \frac{1}{e^{\lambda_n}} = \frac{1}{\exp\left(\sum_{n=0}^{\infty} \lambda_n\right)} \rightarrow 0$$

$$\text{as } \sum_{n=0}^{\infty} \frac{1}{\lambda_n} = \infty.$$

$$\text{Therefore } \Pr\left(\sum_{n=0}^{\infty} S_n = \infty\right) = \Pr\left(\exp\left(-\sum_{n=0}^{\infty} S_n\right) = 0\right) = 1$$

which implies  $\Pr(X(t) = \infty) = 0$ , so  $\sum_{n=0}^{\infty} p_n(t) = 1$ .

b) "Explosive process": choose  $\lambda_n$  s.t.  $\sum_{n=0}^{\infty} \frac{1}{\lambda_n} < \infty \Rightarrow$  any geometric series works. ( $\lambda_n = n^\alpha, \alpha > 1$ ).

② a) Use Forward Kolmogorov equation with  $\lambda_n = \lambda n + v$ ,  
 $\mu_n = \mu n$ :

$$(1) \quad P'_{i,j}(t) = [\lambda(j-1) + v] P_{i,j-1}(t) - [(\lambda + \mu)j + v] P_{i,j}(t) + \mu(j+1) P_{i,j+1}(t), \quad j \geq 1.$$

We deduce

$$\frac{d\bar{n}}{dt} = \frac{d}{dt} [E[X(t)]]$$

$$= \frac{d}{dt} \sum_{j=1}^{\infty} j P_{i,j}(t)$$

$$= \sum_{j=1}^{\infty} j \frac{d}{dt} P_{i,j}(t)$$

$$\stackrel{\text{by (1)}}{=} \sum_{j=1}^{\infty} j \left( [\lambda(j-1) + v] P_{i,j-1}(t) - [(\lambda + \mu)j + v] P_{i,j}(t) + \mu(j+1) P_{i,j+1}(t) \right)$$

$$= \sum_{j=1}^{\infty} \left[ \lambda(j-1)j P_{i,j-1}(t) - (\lambda + \mu)j^2 P_{i,j}(t) + \mu j(j+1) P_{i,j+1}(t) \right]$$

$$+ \sum_{j=1}^{\infty} vj [P_{i,j-1}(t) - P_{i,j}(t)]$$

$$= (\lambda - \mu)\bar{n} + v.$$

b) Example: The logistic model (in computational challenge 2) with immigration.

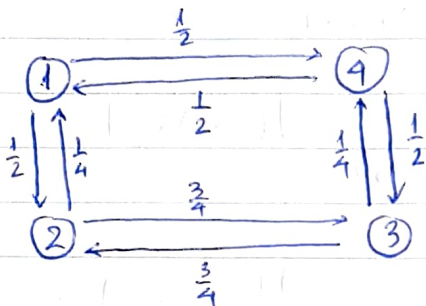
$$\lambda_n = rn \left( 1 - \frac{n}{2k} \right) + v$$

$$\mu_n = \frac{rn^2}{2k}$$

$$r = 0.015, \quad k = 2, \quad v = 0.01.$$

corresponding deterministic equation  $\frac{dn}{dt} = rn \left(1 - \frac{n}{K}\right) + v$ .  $\square$

③ a)



b) We have  $p(1 \rightarrow 2) > 0$ ,  $p(2 \rightarrow 3) > 0$ ,  $p(3 \rightarrow 4) > 0$ ,  $p(4 \rightarrow 1) > 0$ , so all the states communicate with each other, therefore the chain is irreducible.

Moreover, the chain has finite states, so it is positive recurrent.

For any state  $i \in \{1, 2, 3, 4\}$ ,  $p_{ii}^2 > 0$ ,  $p_{ii} = 0$ , so the chain is periodic with period = 2.

c) The stationary distribution  $\pi$  satisfies 
$$\begin{cases} \pi = \pi P \\ \sum_{i=1}^4 \pi_i = 1. \end{cases}$$

Solving the system yields  $\pi = \left( \frac{1}{6}, \frac{1}{3}, \frac{1}{3}, \frac{1}{6} \right)$   $\square$

④ Let  $S_n$  be the sojourn time of  $n$ -th protein, then  $S_n$  has  $\text{Poisson}(\lambda)$ .  
 $T_n$  be the time promoter that the promoter is occupied by  $n$ -th protein, then  $T_n$  are i.i.d r.v.s with mean  $\mu$ .

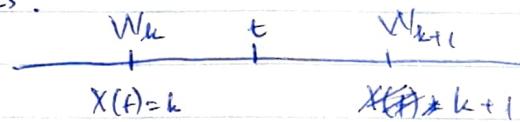
Then

$$\frac{\text{Unoccupied Time}}{\text{Total Time}} = \lim_{N \rightarrow \infty} \frac{\sum_{n=0}^N S_n}{\sum_{n=0}^N (S_n + T_n)} = \lim_{N \rightarrow \infty} \frac{\frac{1}{N} \sum_{n=0}^N S_n}{\frac{1}{N} \sum_{n=0}^N (S_n + T_n)}$$

$$= \frac{E[S_n]}{E[S_n] + E[T_n]} \quad (\text{by Law of Large Numbers})$$

$$= \frac{1}{1 + \lambda \mu} \quad \square$$

- ⑤ Let  $X(t)$  be the Poisson process modelling the sequence of action potentials.



Then, we conclude that

$$T(t) = \min_{X(t) \leq k \leq X(t)+1} |t - W_k|$$

Fixed  $s \in \mathbb{R}$ ,  $s \in \mathbb{R}$ ,

$$\Pr(T(t) > s) = \Pr\left(\min_{X(t) \leq k \leq X(t)+1} |t - W_k| > s\right)$$

$$= \Pr(|t - W_k| > s, |t - W_{k+1}| > s \mid X(t) = k)$$

$$= \Pr(t - W_k > s, W_{k+1} - t > s \mid X(t) = k)$$

$$= \Pr(W_k < t - s, W_{k+1} > t + s \mid X(t) = k)$$

$$= \Pr(X(t+s) - X(t-s) = 0, \mid X(t) = k)$$

$$= \Pr(X(t+s) - X(t) = 0 \mid X(t) = k) \Pr(X(t) - X(t-s) = 0 \mid X(t) = k)$$

$$= e^{-\lambda s} \cdot e^{-\lambda s}$$

$$= e^{-2\lambda s}$$

Therefore

$$F_{T(t)}(s) = 1 - \Pr(T(t) > s) = 1 - e^{-2\lambda s},$$

$$\text{so } f_{T(t)}(s) = 2\lambda e^{-2\lambda s}.$$

Hence,  $T(t)$  has Exponential  $(2\lambda)$  and  $E[T(t)] = \frac{1}{2\lambda}$