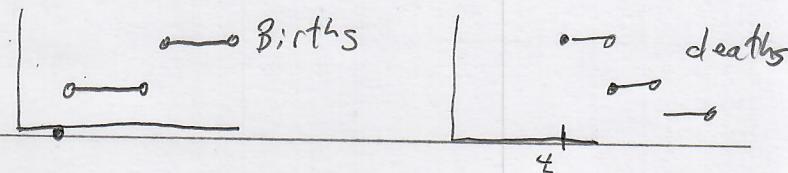


Brooke Campbell  
 OSU ST 6510 4/11/10 Sy.  
 § 6.3.81

sojourns of  $\tau$

lifetimes



We'll borrow from § 5.4 - using all the notation

except the problem has  $X(t)$  as # particles alive

@ t. - denoted  $M(t)$  in § 5.4 Our lifetimes

$y_1, y_2, \dots \sim \exp(\mu)$  so the death rates

$\mu_i = \mu + \gamma_i$ . Our birth rates are determined

by the arrival P.P. ( $\gamma$ ), and  $\gamma_i = \gamma + \delta_i$ .

Now we should show that the conditions of a BDP ( $\mu_i, \gamma_i$ ) are satisfied. Let  $Z(\tau)$  be the

P.P. ( $\gamma$ ) generating particles. We'll start by noting

$$P_{ij}(0) = P(X(0^-)=j \mid X(0)=i) = \delta_{ij} \text{ and } M_0=0$$

because  $Z(0)=0$  and the fact that there can be

no deaths until an arrival. We know the sojourn times

of  $\tau$  are exponential, and by definition the sojourns

before a death are exponential as well.

$$\text{We would like to reason } X(t) = Z(t) - Y(t)$$

Suppose  $X(t)=N$  what can happen in an interval

$t+h$ ? We can have a death  $P_{i,i-1}(h) = \mu h + o(h)$

or a birth  $P_{i,i+1}(h) = \gamma h + o(h)$  or no events

$$P_{i,i}(h) = 1 - [P_{i,i+1}(h) + P_{i,i-1}(h)]$$

Ex 6.3.83

Let  $\pi = \frac{\alpha}{\alpha + \beta}$   $\tilde{\pi} = \alpha - \beta$

$$P_{00}(t) = (1-\pi) + \pi e^{-\tilde{\pi}t} \quad P_{01} = 1 - P_{00}(t) = \pi - \pi e^{-\tilde{\pi}t}$$

$$P_{10}(t) = (1-\pi) - (1-\pi) e^{-\tilde{\pi}t} \quad P_{11}(t) = 1 - P_{10}(t) = \pi + (1-\pi) e^{-\tilde{\pi}t}$$

Using  $P(V(t)=n) = \sum_{i=0}^n q_i P_{in}(t)$  (error in book?)  
version 4

where  $q_i = P(V(0)=i)$  we see

$$q_0 = 1-\pi \quad q_1 = \pi$$

$$P(V(t)=1) = q_0 P_{01}(t) + q_1 P_{11}(t)$$

Inserting  $q$ 's and  $P$ 's from above

$$\begin{aligned} P(V(t)=1) &= (1-\pi)[\pi - \pi e^{-\tilde{\pi}t}] + \pi[\pi + (1-\pi)e^{-\tilde{\pi}t}] \\ &= \cancel{\pi} - \pi \cancel{e^{-\tilde{\pi}t}} - \cancel{\pi^2} - \pi^2 e^{-\tilde{\pi}t} + \cancel{\pi^2} + \pi \cancel{e^{-\tilde{\pi}t}} - \cancel{\pi^2} \cancel{e^{-\tilde{\pi}t}} \\ &= \pi t \end{aligned}$$

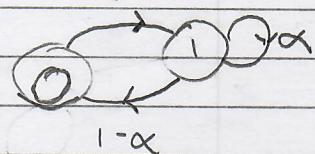
S 6.3. PI  $\xi_n$   $n=0, 1, \dots$  2 state M.C.  $\Rightarrow S = \{0, 1\}$

TPM  $P = \begin{bmatrix} 0 & 1 \\ 1-\alpha & \alpha \end{bmatrix}$  let  $N(t)$   $PP(\lambda)$

Show  $X(t) = \xi_{N(t)}$  is 2 state B.D.P.  $(\lambda_0, \mu_1)$

and find  $\lambda_0, \mu_1$  in terms of  $\lambda, \alpha$

First  $X(0) = \xi_{N(0)} = \xi_0$   $\xi_0$  could be 0 or 1



Starting at  $N(0)$  suppose  $\xi_0 = 0$ ,

the sojourn of  $N(t)$  to  $N(t)=1$  is  $\exp(\lambda t)$  and

we transition to state 1, then we sojourn in

state 1  $\exp(\lambda t)$  and either stay there with

probability  $\alpha$  or transition to 0 with probability

$1-\alpha$ . In terms of infinitesimal transition

probabilities we have

$$P(X(t+h) = 1 | X(t) = 0) = \lambda h + o(h) \quad \text{and}$$

$$P(X(t+h) = 0 | X(t) = 1) = (1-\alpha)\lambda h + o(h)$$

The internet helped me above for the second equation

I also reasoned through what happens when  $\alpha=0$

From the above  $\lambda_0 = \lambda$   $\mu_1 = (1-\alpha)\lambda$

29 § 6.3. P3

$$S = \{0, 1\}$$

30  
31  $P(X(0)=0) = 1 - \pi \quad P(X(0)=1) = \pi \quad 0 < s < t$

32  
33  $E[X(s) X(t)] = \sum_{i=0}^1 i P(X(s) X(t) = i)$

34  
35 since  $X(s) X(t) = 1 \Rightarrow X(s) = 1, X(t) = 1$

36  
37 we have

38  
39  $E[X(s) X(t)] = 0 \cdot P(X(s) X(t) = 0) + 1 P(X(s) X(t) = 1)$

40  
41  $= P(X(s) = 1, X(t) = 1)$  now condition

42  
43  $E[X(s) X(t)] = P(X(t) = 1 | X(s) = 1) P(X(s) = 1)$

44  
45  $= P(X(t-s) = 1 | X(0) = 1) P(X(s) = 1)$

46  
47 Since we're starting in the stationary  
48 distribution  $P(X(s) = 1) = \pi$ .

49  
50  $P(X(t-s) = 1 | X(0) = 1) = P_{11}(t-s) =$

51  
52  
53  $\frac{\alpha}{\alpha+\beta} + \frac{\beta}{\alpha+\beta} e^{-(\alpha+\beta)(t-s)}$  From 6.29c

54

55  
56

§ 6.3. P3

(E)

Also from 6.30 of we have

$$P_{11}(t-s) = \pi + (1-\pi) e^{-\zeta(t-s)} \quad \zeta = \alpha + \beta$$

Since  $P_{11}(t-s) = 1 - P_{10}(t-s)$  from T.P.M,

conditions on rows being probability distribution

$$\text{so } E[X(s) X(t)] = \pi - \pi P_{10}(t-s)$$

$$\text{Now } \text{Cov}(X(s), X(t)) = E[X(s) X(t)] - E[X(s)] E[X(t)]$$

$$E[X(s)] = \sum_{i=0}^1 i \cdot P(X(s)=i) = P(X(s)=1) = \pi$$

From our starting conditions, likewise

$$E[X(t)] = \pi, \text{ so}$$

$$\text{Cov}(X(s), X(t)) = \pi - \pi P_{10}(t-s) - \pi^2 - \text{Not sure}$$

why \* - if we use  $P_{11}(t-s)$  - we have

$$\text{Cov}(X(s), X(t)) = \pi P_{11}(t-s) - \pi^2 = \pi (\pi + (1-\pi) e^{-\zeta(t-s)}) - \pi^2$$

$$= \pi(1-\pi) e^{-\zeta(t-s)} = \pi(1-\pi) e^{-(\alpha+\beta)|t-s|} \quad \text{since } t > s > 0$$

1      § 6.6.82

2       $X_1(t) \quad X_2(t) \quad + \quad S = \{0, 1\}$

3  
4      i.e. generating     $A = \begin{bmatrix} 0 & 1 \\ -2 & 2 \\ 1 & -1 \end{bmatrix}$

5       $Z(t) = X_1(t) + X_2(t)$

6      Clearly  $S_Z = \{0, 1, 2\}$  by considering possibilities7  
8      for  $X_1(t)$   $X_2(t)$ , or using convolution formula9  
10     The Markov property for  $X_1(t)$   $X_2(t)$ 

11  
12      $P_{ij}^1(t) = P(X_1(t+s)=j \mid X_1(s)=i) \quad t \geq 0$

13  
14      $P_{ij}^2(t) = P(X_2(t+s)=j \mid X_2(s)=i) \quad t \geq 0$

15  
16     We need to show  $Z$  inherits this property17  
18     I googled this one because it wasn't19  
20     obvious that  $Z$  is Markov. In fact21  
22     I found lots of confusion and it's23  
24     not true in general. We will have25  
26     to use the fact that  $A$  is the same.

1       $\S 6.6.82$ 

(2)

2

3      Looking at the hint - we can just reason  
45      with the  $P_{ij}$ 's  
6

7       $P_{00}^Z = P_{00}^{X_1}(t) \cdot P_{00}^{X_2}(t)$

8       $P_{01}^Z(t) = P_{00}^{X_1}(t) \cdot P_{01}^{X_2}(t) + P_{01}^{X_1}(t) \cdot P_{00}^{X_2}(t)$

9       $P_{02}^Z = P_{01}^{X_1}(t) \cdot P_{01}^{X_2}(t)$

10      $P_{10}^Z = P_{10}^{X_1}(t) P_{00}^{X_2}(t) + P_{00}^{X_1}(t) P_{10}^{X_2}(t)$

11      $P_{11}^Z(t) = P_{11}^{X_1}(t) P_{00}^{X_2}(t) + P_{00}^{X_1}(t) P_{11}^{X_2}(t) + P_{01}^{X_1}(t) P_{10}^{X_2}(t)$

12      $P_{12}^Z(t) = P_{11}^{X_1}(t) P_{01}^{X_2}(t) + P_{01}^{X_1}(t) P_{11}^{X_2}(t)$

13      $P_{20}^Z(t) = P_{10}^{X_1}(t) P_{10}^{X_2}(t)$

14      $P_{21}^Z(t) = P_{11}^{X_1}(t) P_{10}^{X_2}(t) + P_{10}^{X_1}(t) P_{11}^{X_2}(t)$

15      $P_{22}^Z(t) = P_{11}^{X_1}(t) P_{11}^{X_2}(t)$

16     Now expressing in terms of  
17      $\lambda$ 

18      $\tau = \frac{\lambda}{\lambda + \mu} \quad \tau = \lambda + \mu \quad$  we have

Writing  $\tilde{P}^z(t) = [P_0^z(t), P_1^z(t), P_2^z(t)]$   
 the columns of  $\tilde{P}^z(t)$  are

$$\tilde{P}_0^z(t) = \begin{bmatrix} [(1-\pi) + \pi e^{-\bar{\tau}t}]^2 \\ [(1-\pi) + \pi e^{-\bar{\tau}t}] [\pi - \pi e^{-\bar{\tau}t}] \cdot 2 \\ [\pi - \pi e^{-\bar{\tau}t}]^2 \end{bmatrix}$$

$$\tilde{P}_1^z(t) = \begin{bmatrix} [(1-\pi) - (1-\pi)e^{-\bar{\tau}t}] [\pi + \pi e^{-\bar{\tau}t}] \cdot 2 \\ [\pi + (1-\pi)e^{-\bar{\tau}t}] [\pi - \pi e^{-\bar{\tau}t}] \cdot 2 + [\pi - \pi e^{-\bar{\tau}t}] [(1-\pi) - (1-\pi)e^{-\bar{\tau}t}] \\ [\pi + (1-\pi)e^{-\bar{\tau}t}] [\pi - \pi e^{-\bar{\tau}t}] \cdot 2 \end{bmatrix}$$

$$\tilde{P}_2^z(t) = \begin{bmatrix} [(1-\pi) - (1-\pi)e^{-\bar{\tau}t}]^2 \\ [\pi + (1-\pi)e^{-\bar{\tau}t}] [\pi - (1-\pi)e^{-\bar{\tau}t}] \cdot 2 \\ [\pi + (1-\pi)e^{-\bar{\tau}t}]^2 \end{bmatrix}$$

1 6.6. P1

2  $Y_n \quad n=0, 1, 2, \dots$  a M.C.  $\in TPM$   $P = \|P_{ij}\|$ 

4  $N(t) + P.P(\geq) \quad X(t) = Y_{N(t)} \quad t \geq 0$

6  $N(t) \sim Pois(t)$   $\forall t \in [0, \infty) \quad N(t) \in \{0, 1, \dots\}$ 8 so  $\forall t \quad X(t) = Y_k$  for some  $k$ 10 and  $S_x = S_y$ . Now  $P(X(t)=j | X(0)=i)$ 

12  $= P(Y_{N(t)}=j | Y_0=i) = P(Y_k=j | Y_0=i) = P_{ij}^k$

14 We also know  $P(Y_{n+1}=j | Y_0=i_0, Y_1=i_1, \dots, Y_n=i_n)$ 16  $= P(Y_{n+1}=j | Y_n=i)$ . Now let  $t_n > t_{n-1} > \dots > t_0$ 18  $P(X(t_n)=j | X(t_{n-1})=i, X(t_{n-2})=i_{n-2}, \dots, X(t_0)=i_0)$ 

20  $= P(Y_{k_n}=j | Y_{k_{n-1}}=i, \dots, Y_{k_0}=i_0) =$

22  $P(Y_{k_n}=j | Y_{k_{n-1}}=i)$  since24  $k_n \geq k_{n-1} \geq \dots \geq k_0$  from the properties of26  $N(t)$  so  $X(t)$  is M.C.

1 6.6. P1

(z)

2 like the 2-state situation we can

3 investigate A through the infinitesimal

4

5 transition probabilities

6

7

8  $P(X(t+h) = j | X(t) = i) = (\gamma h) P_{ij} + o(h)$

9

10 the prob of  $N(t)$  having an event

11

12 in  $[t, t+h]$  times the probability of a

13

14 transition from  $i$  to  $j$ . In our earlier

15

16 notation

17

18

19  $P(X(t+h) = j | X(t) = i) = P(Y_{N(t+h)} = j | Y_{N(t)} = i)$ 

20

21 and when  $h$  is small  $N(t+h) - N(t) = \{^o\}$ 

22

23 with prob

$$\frac{1 - (\gamma h) + o(h)}{\gamma h + o(h)}$$

from ch 3.

24

25

26

27

28