

# MATH180C: Introduction to Stochastic Processes II

[www.math.ucsd.edu/~ynemish/teaching/180c](http://www.math.ucsd.edu/~ynemish/teaching/180c)

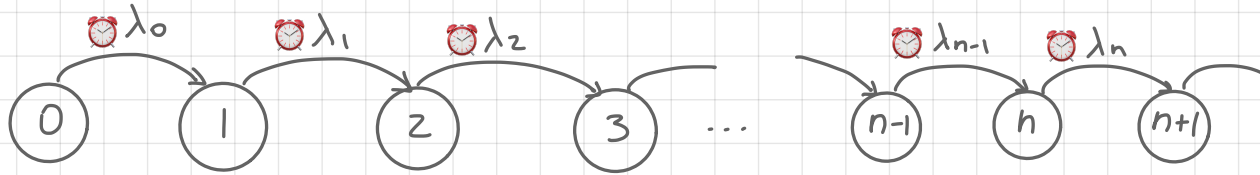
Today: Pure death processes.  
Birth and death processes  
> Q&A: October 9  
Next: PK 6.5

Week 1:

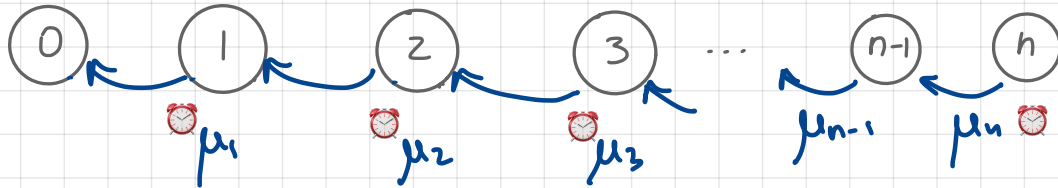
- homework 1 (due Friday October 9)
- join Piazza

# Pure death processes

Pure birth process



What if the chain moves in the opposite direction?



Pure death process:

- exponential sojourn times with rates  $\mu_i$
- only negative jumps of magnitude 1 allowed

## Pure death processes

### Infinitesimal description:

Pure death process  $(X_t)_{t \geq 0}$  of rates  $(\mu_k)_{k=1}^N$  is a continuous time MC taking values in  $\{0, 1, 2, \dots, N-1, N\}$  (state 0 is absorbing) with stationary infinitesimal transition probability functions

$$(a) \quad P_{k, k-1}(h) = \mu_k h + o(h) \quad k=1, \dots, N$$

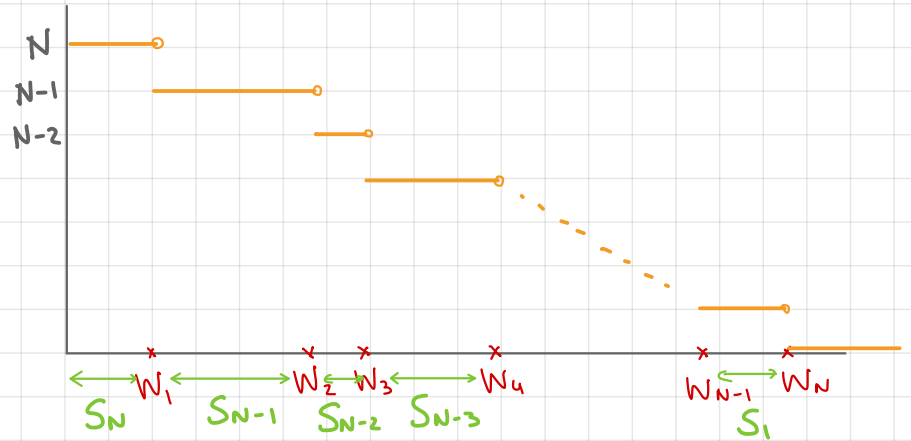
$$(b) \quad P_{kk}(h) = 1 - \mu_k h + o(h), \quad k=1, \dots, N$$

$$(c) \quad P_{kj}(h) = 0 \quad \text{for } j > k.$$

State 0 is absorbing ( $\mu_0 = 0$ )

## Pure death process

$$S_k \sim \text{Exp}(\mu_k)$$



## Sojourn time / jump description:

Pure death process of rates  $(\mu_k)_{k=1}^N$  is a nonincreasing right-continuous process taking values in  $\{0, 1, \dots, N\}$

- with sojourn times  $S_1, S_2, S_3, \dots, S_N$  being independent exponential r.v.s of rates  $\mu_1, \mu_2, \dots, \mu_N$  and
- jumps  $X_{W_{i+1}} - X_{W_i} = -1$  of magnitude 1

## Differential equations for pure birth processes

Define  $P_n(t) = P(X_t = n \mid X_0 = N)$  distribution of  $X_t$   
 $\uparrow$  starting in state  $N$

(a), (b), (c) implies (check)

$$\begin{cases} P_n'(t) = -\mu_n P_n(t) + \mu_{n+1} P_{n+1}(t) & \text{for } n=0 \dots N-1 \\ P_N'(t) = -\mu_N P_N(t) & \end{cases} \quad (\text{note that } \mu_0 = 0)$$

Initial conditions:  $P_N(0)=1$ ,  $P_n(0)=0$  for  $n=0 \dots N-1$

Solve recursively:  $P_N(t) = e^{-\mu_N t} \rightarrow P_{N-1}(t) \rightarrow \dots \rightarrow P_0(t)$

General solution (assume  $\mu_i \neq \mu_j$ )

$$P_n(t) = \mu_{n+1} \cdots \mu_N (A_{nn} e^{-\mu_n t} + \cdots + A_{N,n} e^{-\mu_N t}), \quad A_{kn} = \prod_{\substack{\ell=n \\ \ell \neq k}}^N \frac{1}{\mu_\ell - \mu_k}$$

## Linear death process

Similar to Yule process:

death rate is proportional to the size of the population

$$\mu_k = k\alpha \text{ (linear dependence on } k\text{)}$$

Compute  $P_n(t)$ : •  $\mu_{n+1} \cdots \mu_N = \alpha^{N-n} \frac{N!}{n!}$

$$\bullet A_{kn} = \prod_{\substack{l=n \\ l \neq k}}^N \frac{1}{\mu_l - \mu_k} = \frac{1}{\alpha^{N-n} (-1)^{n-k} (k-n)! (N-k)!}$$

$$\left\{ \mu_l - \mu_k = \alpha(l-k) \right.$$

$$\bullet P_n(t) = \alpha^{N-n} \frac{N!}{n!} \cdot \frac{1}{\alpha^{N-n}} \sum_{k=n}^N \frac{1}{(-1)^{n-k} (k-n)! (N-k)!} \cdot e^{-k\alpha t} \quad \left\{ \begin{array}{l} j = k-n \\ k = j+n \end{array} \right.$$

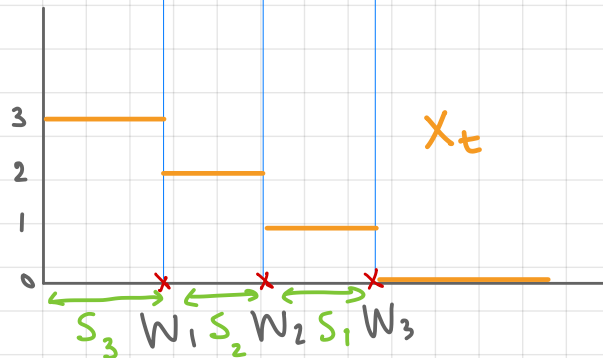
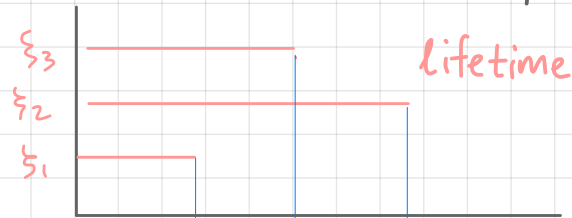
$$= \frac{N!}{n!} \sum_{j=0}^{N-n} \frac{(-1)^j e^{-(j+n)\alpha t}}{j! (N-n-j)!}$$

$$= \frac{N!}{n!} e^{-n\alpha t} \sum_{j=0}^{N-n} \frac{1}{j! (N-n-j)!} (-e^{-\alpha t})^j = \frac{N!}{n! (N-n)!} e^{-n\alpha t} (1 - e^{-\alpha t})^{N-n}$$

$$\nearrow X_t \sim \text{Bin}(N, e^{-\alpha t})$$

## Interpretation of $X_t \sim \text{Bin}(n, e^{-\alpha t})$

Consider the following process: Let  $\xi_i, i=1 \dots N$ , be i.i.d. r.v.s,  $\xi_i \sim \text{Exp}(\alpha)$ . Denote by  $X_t$  the number of  $\xi_i$ 's that are bigger than  $t$  ( $\xi_i$  is the lifetime of an individual,  $X_t$  = size of the population at  $t$ ).  $X_0 = N$ .



Then:  $S_k \sim \text{Exp}(\alpha k)$ , independent

$\hookrightarrow (X_t)_{t \geq 0}$  is a pure death process

Probability that an individual

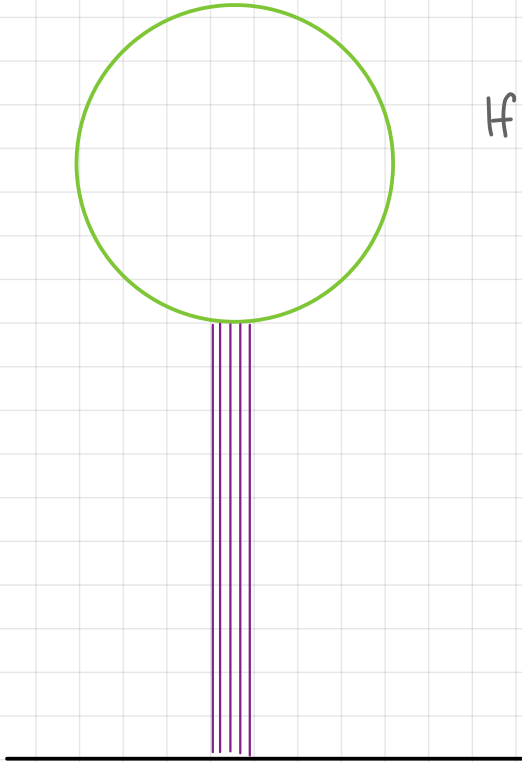
survives to time  $t$  is  $e^{-\alpha t}$

Probability that exactly  $n$

individuals survive to time  $t$  is

$$\binom{N}{n} e^{-\alpha n t} (1 - e^{-\alpha t})^{N-n} = P(X_t = n)$$

## Example . Cable



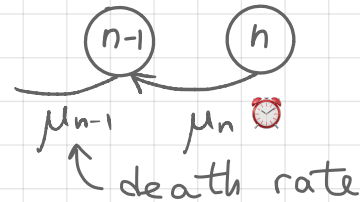
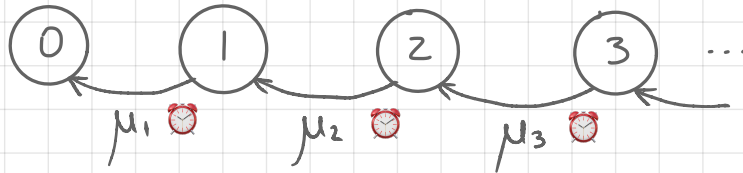
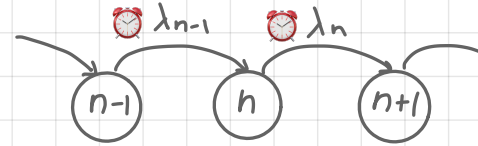
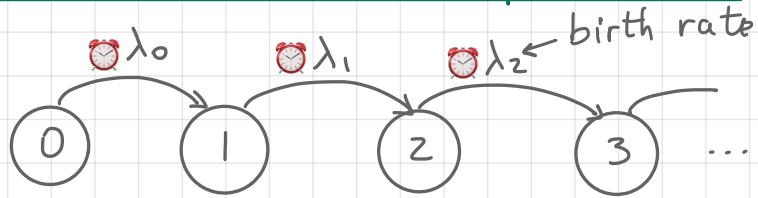
$X_t$  = number of fibers in the cable

If a fiber fails, then this increases the load on the remaining fibers, which results in a shorter lifetime.

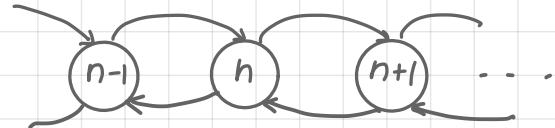
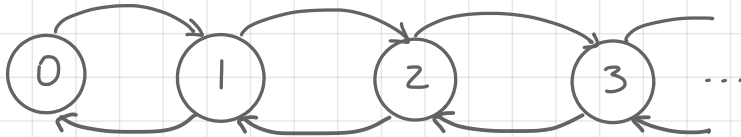
↳ pure death process



# Birth and death processes



Combine both



Birth and death processes

## Infinitesimal definition

Def. Let  $(X_t)_{t \geq 0}$  be a continuous time MC,  $X_t \in \{0, 1, 2, \dots\}$  with stationary transition probabilities. Then  $(X_t)_{t \geq 0}$  is called a birth and death process with birth rates  $(\lambda_k)$  and death rates  $(\mu_k)$  if

$$1. P_{i, i+1}(h) = \lambda_i h + o(h)$$

$$2. P_{i, i-1}(h) = \mu_i h + o(h)$$

$$3. P_{i, i}(h) = 1 - (\lambda_i + \mu_i)h + o(h)$$

$$4. P_{ij}(0) = \delta_{ij} \quad (P(X_0=j | X_0=i) = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases})$$

$$5. \mu_0 = 0, \lambda_0 > 0, \lambda_i, \mu_i > 0$$

## Example: Linear growth with immigration

Dynamics of a certain population is described by the following principles:

during any small period of time of length  $h$

- each individual gives birth to one new member with probability  $\beta h + o(h)$  independently of other members;
- each individual dies with probability  $\alpha h + o(h)$  independently of other members;
- one external member joins the population with probability  $a h + o(h)$

Can be modeled as a Markov process

## Example: Linear growth with immigration

Let  $(X_t)_{t \geq 0}$  denote the size of the population.

Using a similar argument as for the Yule/pure death models:

•  $P_{n,n+1}(h) = n\beta h + ah + o(h)$

← pure birth growth

↑ immigration growth

•  $P_{n,n-1}(h) = n\alpha h + o(h)$

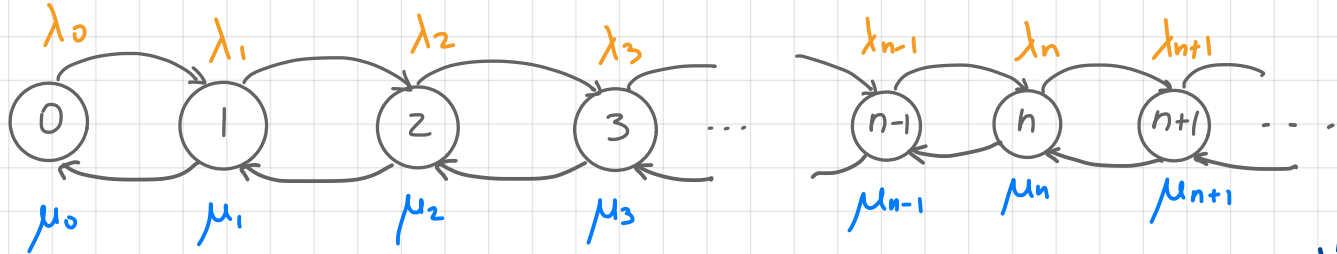
•  $P_{n,n}(h) = 1 - (n\beta h + ah + n\alpha h) + o(h)$

↳ birth and death process with

$$\lambda_n = n\beta + a$$

$$\mu_n = n\alpha$$

# Alternative (jump and hold) characterization



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

$\swarrow \searrow$   
 $\text{Exp}(z) \quad \text{Exp}(\mu)$

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  $\frac{\mu_i}{\lambda_i + \mu_i}$