

## 6. Birth and Death Processes

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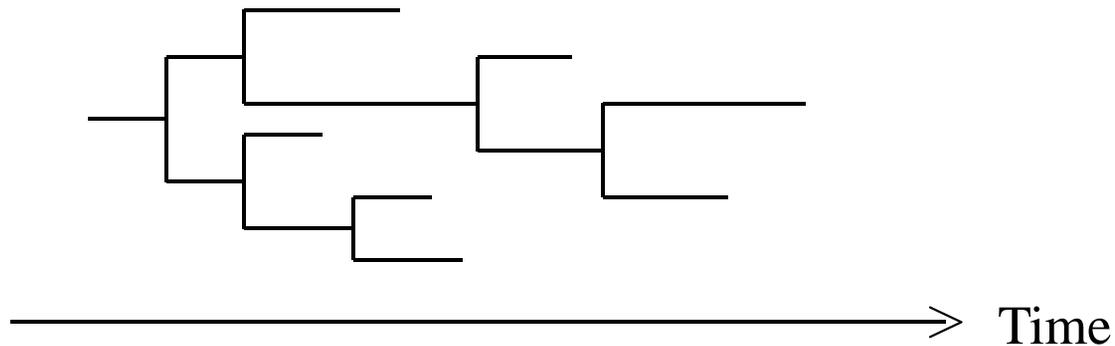
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## 6.1 Pure Birth Process (Yule-Furry Process)

Example. Consider cells which reproduce according to the following rules:

- i. A cell present at time  $t$  has probability  $\lambda h + o(h)$  of splitting in two in the interval  $(t, t + h)$
- ii. This probability is independent of age.
- iii. Events between different cells are independent



## Non-Probabilistic Analysis

$n(t)$  = no. of cells at time  $t$

$\Rightarrow \lambda n(t) \Delta t$  births occur in  $(t, t + \Delta t)$

where  $\lambda$  = birth rate per single cell.

$$n(t + \Delta t) = n(t) + n(t)\lambda\Delta t$$

$$\frac{n(t + \Delta t) - n(t)}{\Delta t} \rightarrow n'(t) = n(t)\lambda$$

or

$$\frac{n'(t)}{n(t)} = \frac{d}{dt} \log n(t) = \lambda$$

$$\log n(t) = \lambda t + c$$

$$n(t) = Ke^{\lambda t}, \quad n(0) = n_0$$

$$\boxed{n(t) = n_0 e^{\lambda t}}$$

## Probabilistic Analysis

$N(t)$  = no. of cells at time  $t$

$$P\{N(t) = n\} = P_n(t)$$

Prob. of birth in  $(t, t + h)$  if  $\{N(t) = n\} = n\lambda h + o(h)$

$$\begin{aligned} P_n(t + h) &= P_n(t)(1 - n\lambda h + o(h)) \\ &\quad + P_{n-1}(t)((n - 1)\lambda h + o(h)) \end{aligned}$$

$$P_n(t + h) - P_n(t) = -n\lambda h P_n(t) + P_{n-1}(t)(n - 1)\lambda h + o(h)$$

$$\frac{P_n(t + h) - P_n(t)}{h} = -n\lambda P_n(t) + P_{n-1}(t)(n - 1)\lambda + o(h) \text{ as } h \rightarrow 0$$

$$\boxed{P'_n(t) = -n\lambda P_n(t) + (n - 1)\lambda P_{n-1}(t)}$$

Initial condition  $P_{n_0}(0) = P\{N(0) = n_0\} = 1$

$$P'_n(t) = -n\lambda P_n(t) + (n-1)\lambda P_{n-1}(t); \quad P_{n_0}(0) = 1$$

Solution:

$$(1) \quad P_n(t) = \binom{n-1}{n-n_0} e^{-\lambda n_0 t} (1 - e^{-\lambda t})^{n-n_0} \quad n = n_0, n_0 + 1, \dots$$

Solution is negative binomial distribution; i.e. Probability of obtaining exactly  $n_0$  successes in  $n$  trials.

Suppose  $p =$  prob. of success and  $q = 1 - p =$  prob. of failure. Then in first  $(n-1)$  trials results in  $(n_0-1)$  successes and  $(n-n_0)$  failures followed by success on  $n^{\text{th}}$  trial; i.e.

$$(2) \quad \binom{n-1}{n_0-1} p^{n_0-1} q^{n-n_0} \cdot p = \binom{n-1}{n-n_0} p^{n_0} q^{n-n_0} \quad n = n_0, n_0 + 1, \dots$$

If  $p = e^{-\lambda t}$  and  $q = 1 - e^{-\lambda t}$

$\Rightarrow$  (2) is same as (1).

Yule studied this process in connection with theory of evolution; i.e. population consists of the species within a genus and creation of new element is due to mutations. Neglects probability of species dying out and size of species.

Furry used same model for radioactive transmutations.

### Notes on Negative Binomial Distribution

The geometric distribution is defined as the number of trials to achieve one success for a series of Bernoulli trials; i.e.

Geometric Distribution:  $P\{N = n\} = pq^{n-1}, n = 1, 2, \dots$

$N$  is number of trials for 1 success

$$\phi_N(s) = E(e^{-sN}) = p \sum_{n=1}^{\infty} e^{-sn} q^{n-1} = \frac{p}{q} \sum_{n=1}^{\infty} (e^{-s} q)^n.$$

But

$$\sum_{n=1}^{\infty} (e^{-s}q)^n = e^{-s}q/(1 - e^{-s}q)$$

$$\phi_N(s) = \frac{p}{q} \cdot \frac{e^{-s}q}{1 - e^{-s}q} = \frac{pe^{-s}}{1 - e^{-s}q} = \frac{pz}{1 - qz} \quad \text{if } z = e^{-s}$$

$$\phi_N(z) = pz(1 - qz)^{-1}$$

$$\phi'_N(z) = p\{(1 - qz)^{-1} + z(1 - qz)^{-2}q\}$$

$$\phi'_N(1) = p\{(1 - q)^{-1} + q(1 - q)^{-2}\} = 1 + \frac{q}{p} = \frac{p + q}{p} = \frac{1}{p}$$

$$\text{Similarly } \phi''_N(1) = \frac{2}{p^2} - \frac{2}{p} \Rightarrow V(n) = \frac{1}{p^2} - \frac{1}{p}$$

Theorem: If  $N_i$  ( $i = 1, 2, \dots, n_0$ ) are iid geometric random variables with parameter  $p$ , then  $N = N_1 + N_2 + \dots + N_{n_0}$  is a negative binomial distribution having generating function

$$\phi_N(z) = \left( \frac{pz}{1 - qz} \right)^{n_0}, \quad z = e^{-s}$$

$$\therefore E(N) = n_0/p, \quad V(N) = n_0 \left[ \frac{1}{p^2} - \frac{1}{p} \right]$$

If  $p = e^{-\lambda t}$  and  $N(t)$  is a pure birth process

$$E[N(t)] = n_0 e^{\lambda t}$$

$$V[N(t)] = n_0 [e^{2\lambda t} - e^{\lambda t}]$$

## 6.2 Generalization

In Poisson Process, the prob. of a change during  $(t, t + h)$  is independent of number of changes in  $(0, t)$ . Assume instead that if  $n$  changes occur in  $(0, t)$ , the probability of new change to  $n + 1$  in  $(t, t + h)$  is  $\lambda_n h + o(h)$ . The probability of more than one change is  $o(h)$ . Then

$$P_n(t + h) = P_n(t)(1 - \lambda_n h) + P_{n-1}(t)\lambda_{n-1}h + o(h), \quad n \neq 0$$

$$P_0(t + h) = P_0(t)(1 - \lambda_0 h) + o(h)$$

$$\Rightarrow P'_n(t) = -\lambda_n P_n(t) + \lambda_{n-1} P_{n-1}(t)$$

$$P'_0(t) = -\lambda_0 P_0(t).$$

Equations can be solved recursively with  $P_0(t) = P_0(0)e^{-\lambda_0 t}$ .

If the initial condition is  $P_{n_0}(0) = 1$ , then the resulting equations are:

$$P'_n(t) = -\lambda_n P_n(t) + \lambda_{n-1} P_{n-1}(t), \quad n > n_0$$

$$P'_{n_0}(t) = -\lambda_{n_0} P_{n_0}(t)$$

Pure birth process assumed  $\lambda_n = n\lambda$ .

### Change of Language

If  $n$  transitions take place during  $(0, t)$ , we may refer to the process as being in state  $E_n$ . Changes occur  $E_n \rightarrow E_{n+1} \rightarrow E_{n+2} \rightarrow \dots$  for the pure birth process.

### 6.3 Birth and Death Processes

Consider transitions  $E_n \rightarrow E_{n-1}$  as well as  $E_n \rightarrow E_{n+1}$  if  $n \geq 1$ .

If  $n = 0$ , we only allow  $E_0 \rightarrow E_1$ .

Assume that if the process at time  $t$  is in  $E_n$ , then during  $(t, t + h)$  the transitions  $E_n \rightarrow E_{n+1}$  have prob.  $\lambda_n h + o(h)$ ,  $E_n \rightarrow E_{n-1}$  have prob.  $\mu_n h + o(h)$  and Prob. more than 1 change occurs =  $o(h)$

$$P_n(t + h) = P_n(t)\{1 - \lambda_n h - \mu_n h\}$$

$$+ P_{n-1}(t)\{\lambda_{n-1} h\} + P_{n+1}(t)\{\mu_{n+1} h\} + o(h)$$

$$\Rightarrow \boxed{P'_n(t) = -(\lambda_n + \mu_n)P_n(t) + \lambda_{n-1}P_{n-1}(t) + \mu_{n+1}P_{n+1}(t)}$$

For  $n = 0$

$$P_0(t + h) = P_0(t)\{1 - \lambda_0 h\} + P_1(t)\mu_1 h + o(h)$$

$$\Rightarrow \boxed{P'_0(t) = -\lambda_0 P_0(t) + \mu_1 P_1(t)}$$

If the initial conditions  $P_{n_0}(0) = 1$  in which case 0 in above is replaced by  $n_0$ .

If  $\lambda_0 = 0$ , then  $E_0 \rightarrow E_1$  is impossible and  $E_0$  is an absorbing state.

If  $\lambda_0 = 0$ , then  $P'_0(t) = \mu_1 P_1(t) \geq 0$  so that  $P_0(t)$  increases monotonically.

Note:  $\lim_{t \rightarrow \infty} P_0(t) = P_0(\infty) = \text{Probability of being absorbed.}$

$$P'_0(t) = -\lambda_0 P_0(t) + \mu_1 P_1(t)$$

$$P'_n(t) = -(\lambda_n + \mu_n)P_n(t) + \lambda_{n-1}P_{n-1}(t) + \mu_{n+1}P_{n+1}(t)$$

As  $t \rightarrow \infty$ ,  $P_n(t) \rightarrow P_n$  (*limit*) hence  $P'_0(t) = 0$  for large  $t$  and  $P'_n(t) = 0$  for large  $t$ . Therefore

$$0 = -\lambda_0 P_0 + \mu_1 P_1$$

$$\Rightarrow P_1 = \frac{\lambda_0}{\mu_1} P_0$$

$$0 = -(\lambda_1 + \mu_1)P_1 + \lambda_0 P_0 + \mu_2 P_2$$

$$\Rightarrow P_2 = \frac{\lambda_0 \lambda_1}{\mu_2 \mu_1} P_0$$

$$\Rightarrow P_3 = \frac{\lambda_0 \lambda_1 \lambda_2}{\mu_1 \mu_2 \mu_3} P_0 \quad \text{etc.}$$

Note that dependence on initial conditions has disappeared.

## 6.4 Relation to Markov Chains

Define for  $t \rightarrow \infty$

$$\begin{aligned} P(E_{n+1}|E_n) &= \text{Prob. of transition } E_n \rightarrow E_{n+1} \\ &= \text{Prob. of going to } E_{n+1} \text{ conditional on being in } E_n. \end{aligned}$$

Similarly define  $P(E_{n-1}|E_n)$ .

$$P(E_{n+1}|E_n) \propto \lambda_n, \quad P(E_{n-1}|E_n) \propto \mu_n$$

$$\Rightarrow P(E_{n+1}|E_n) = \frac{\lambda_n}{\lambda_n + \mu_n}, \quad P(E_{n-1}|E_n) = \frac{\mu_n}{\lambda_n + \mu_n}$$

Same conditional probabilities hold if it is given that a transition will take place during  $(t, t + h)$  conditional on being in  $E_n$ .

## 6.5 Linear birth and death processes

$$\lambda_n = n\lambda, \quad \mu_n = n\mu$$

$$\Rightarrow P'_0(t) = \mu P_1(t)$$

$$P'_n(t) = -(\lambda + \mu)nP_n(t) + \lambda(n-1)P_{n-1}(t) + \mu(n+1)P_{n+1}(t)$$

Steady state behavior is characterized by

$$\lim_{t \rightarrow \infty} P'_0(t) = 0 \Rightarrow P_1(\infty) = 0$$

Similarly as  $t \rightarrow \infty$   $P'_n(\infty) = 0$

If  $P_0(\infty) = 1 \Rightarrow$  Probability of ultimate extinction is 1.

If  $P_0(\infty) = P_0 < 1$ , the relations  $P_1 = P_2 = P_3 \dots = 0$  imply with prob.  $1 - P_0$  the population can increase without bounds. The population must either die out or increase indefinitely.

$$P'_n(t) = -(\lambda + \mu)nP_n(t) + \lambda(n-1)P_{(n-1)}(t) + \mu(n+1)P_{n+1}(t)$$

Define Mean by  $M(t) = \sum_{n=1}^{\infty} nP_n(t)$

and consider  $M'(t) = \sum_1^{\infty} nP'_n(t)$ .

$$\begin{aligned} M'(t) &= -(\lambda + \mu) \sum_1^{\infty} n^2 P_n(t) + \lambda \sum_1^{\infty} (n-1)n P_{(n-1)}(t) \\ &\quad + \mu \sum_1^{\infty} (n+1)n P_{n+1}(t) \end{aligned}$$

Write  $(n-1)n = (n-1)^2 + (n-1)$ ,  $(n+1)n = (n+1)^2 - (n+1)$

$$\begin{aligned}
M'(t) &= -(\lambda + \mu) \sum_1^{\infty} n^2 P_n(t) \\
&\quad + \lambda \sum_1^{\infty} (n-1)^2 P_{n-1}(t) + \mu \left[ \sum_1^{\infty} (n+1)^2 P_{n+1}(t) + 1 \cdot P_1(t) \right] \\
&\quad + \lambda \sum_1^{\infty} (n-1) P_{n-1}(t) - \mu \left[ \sum_1^{\infty} (n+1) P_{n+1}(t) + P_1(t) \right] \\
&\Rightarrow M'(t) = \lambda \sum_1^{\infty} n P_n(t) - \mu \sum_1^{\infty} n P_n(t) \\
&\quad = (\lambda - \mu) M(t)
\end{aligned}$$

$$\boxed{M(t) = n_0 e^{(\lambda - \mu)t}} \quad \text{if } P_{n_0}(0) = 1$$

$$M(t) = n_0 e^{(\lambda - \mu)t}$$

$M(t) \rightarrow 0$  or  $\infty$  depending on  $\lambda < \mu$  or  $\lambda > \mu$ .

Similarly if  $M_2(t) = \sum_1^{\infty} n^2 P_n(t)$  one can show

$$M_2'(t) = 2(\lambda - \mu)M_2(t) + (\lambda + \mu)M(t)$$

and when  $\lambda > \mu$ , the variance is

$$n_0 e^{2(\lambda - \mu)t} \{1 - e^{(\mu - \lambda)t}\} \frac{\lambda + \mu}{\lambda - \mu}$$