

1. Branching processes.
Models for dynamics of population

$X_0 = 1$

X_n : size of population time n

Z_i : offspring distribution

$\mu = \mathbb{E}(Z)$, $\sigma^2 = \text{Var}(Z)$

$P(Z=j) = P_j$

Have shown: If $P_0 > 0$ either $X_n \rightarrow 0$ or $X_n \rightarrow \infty$

Let $\pi_0 = \lim_{n \rightarrow \infty} P(X_n = 0 | X_0 = 1)$ i.e. the probability that the population dies out

π_0 ?

$0 < \mu < 1$: $\mu^n = E(X_n) = \sum_{j=0}^n j P(X_n = j | X_0 = 1)$
 $\geq \sum_{j=1}^n P(X_n = j | X_0 = 1)$
 $P(X_n \geq 1 | X_0 = 1)$

Since $\mu^n \rightarrow 0$, then
 $\lim_{n \rightarrow \infty} P(X_n = 0 | X_0 = 1) = 1 - \lim_{n \rightarrow \infty} P(X_n \geq 1 | X_0 = 1) = 1$

and $\lim_{n \rightarrow \infty} P(X_n = 0 | X_0 = 1) \rightarrow \pi_0$

feb. 11-11:14

$\mu > 1$. $\pi_0 = P(\text{population dies out} | X_0 = 1)$

$$= \sum_{j=0}^{\infty} P(\text{population dies out} | X_1 = j) \cdot P(X_1 = j | X_0 = 1)$$

$$= \sum_{j=0}^{\infty} \pi_0^j P_j$$

Then π_0 is the smallest value that satisfies $\pi_0 = \sum_{j=0}^{\infty} \pi_0^j P_j$

$\mu = 1$: Then one can show $\lim_{n \rightarrow \infty} P(X_n = 0 | X_0 = 1) \rightarrow \pi_0$

feb. 11-11:26

Examples: $P_0 = 1/2, P_1 = 1/4, P_2 = 1/4$

Then $\mu = 0 \cdot 1/2 + 1 \cdot 1/4 + 2 \cdot 1/4 = 3/4 < 1$

Example $P_0 = 1/4, P_1 = 1/4, P_2 = 1/2$

$\mu = 1 \cdot 1/4 + 2 \cdot 1/2 = 5/4 > 1$

π_0 satisfies
 $\pi_0 = \frac{1}{4} + \frac{1}{4} \pi_0 + \frac{1}{2} \pi_0^2$

so $2\pi_0^2 - 3\pi_0 + 1 = 0$ with solution
 $\frac{3 \pm \sqrt{9-8}}{2 \cdot 2} = \left\{ \frac{1}{2}, \frac{1}{2} \right\}$

so $\pi_0 = 1/2$

Example if $X_0 = n$ then
 $\lim_{n \rightarrow \infty} P(\text{population dies out}) = \pi_0^n$

feb. 11-11:33

4.8 Time-reversible Markov chains

Suppose that X_n is ergodic (irreducible, aperiodic, positive recurrent)

Let P be transition matrix of X_n

We know that $\exists \pi = \pi P$ $\pi_j > 0 \forall j$

Let $Y_n = X_{n-N}$, $n = 0, 1, \dots, N$ and consider Y_0, Y_1, \dots, Y_N which is X_n run backward.

Is Y_n a Markov chain?

If yes, what is the transition matrix?

$$P(Y_n = j | Y_{n-1} = i) = P(X_{n-N} = j | X_{n-N-1} = i)$$

$$= \frac{P(X_{n-N} = i | X_{n-N-1} = j) P(X_{n-N-1} = i)}{P(X_{n-N} = i)}$$

$$= P_{ji} \frac{\pi_j}{\pi_i} = Q_{ij}$$

Now $P(Y_0 = i_0, \dots, Y_N = i_N) = P(X_0 = i_N, \dots, X_N = i_0)$

$$= \pi_{i_0} P_{i_0 i_1} \dots P_{i_{N-1} i_N}$$

$$= \pi_{i_0} \left(\frac{\pi_{i_1}}{\pi_{i_0}} Q_{i_0 i_1} \right) \dots \left(\frac{\pi_{i_N}}{\pi_{i_{N-1}}} Q_{i_{N-1} i_N} \right) \pi_{i_N} Q_{i_{N-1} i_N}$$

feb. 11-11:33

so $P(Y_n = i_n | Y_{n-1} = i_{n-1}, \dots, Y_0 = i_0) = P(Y_n = i_n | Y_{n-1} = i_{n-1})$

which implies that Y_n is a Markov chain with transition probability $Q = \{Q_{ij}\}$

Defn: If $P_{ij} = \frac{\pi_j}{\pi_i} P_{ji}$ the Markov chain is time reversible.

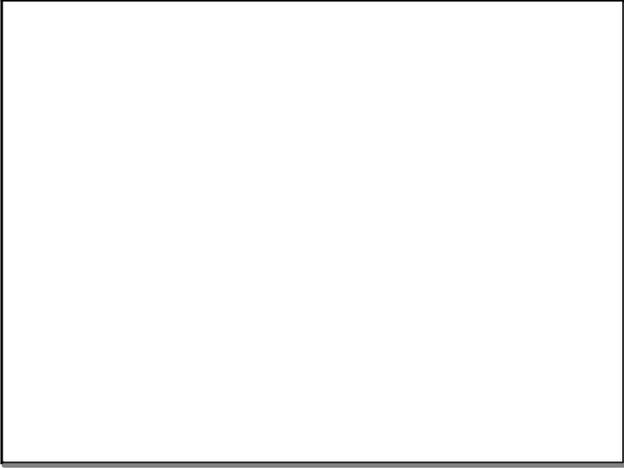
An alternative expression is

$\pi_i P_{ij} = \pi_j P_{ji}$ all i, j

feb. 11-11:33

- 1 -

feb. 11-11:43



feb. 11-11:54