

Branching Processes

Stat 528
4-6-17

Each individual produces j offspring with probability p_j , independent of all other individuals. ①

Assume $0 \leq p_j < 1$ and $\sum_{j=0}^{\infty} p_j = 1$

$X_0 = \#$ of individuals initially present
(generation 0)

$X_n = \#$ of individuals in the n^{th} generation

The state space is $\{0, 1, 2, \dots\}$ ②

$P_{00} = 1$ So 0 is an absorbing state

If $P_0 > 0$, then

$$P_{i0} = P_0^i > 0$$

[by independence]

So all other states must be transient.

If the $\#$ of transient states is finite, then
the total $\#$ of visits to all transient

State is also finite, so eventually the process would have to visit state 0.

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\therefore Either the # of transient states is infinite or the population dies out.

\therefore In the limit as $n \rightarrow \infty$, the population either increases without bound or dies out.

Let μ and σ^2 be the mean and variance of the # of offspring per individual.

That is, let $Z_i = \# \text{ offspring for } i^{\text{th}} \text{ individual}$

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$$\mu = E[Z_i] = \sum_{j=0}^{\infty} j P_j = \sum_{j=1}^{\infty} j P_j$$

$$\begin{aligned} \sigma^2 = \text{Var}[Z_i] &= E[Z_i^2] - \mu^2 \\ &= \sum_{j=0}^{\infty} j^2 P_j - \mu^2 = \sum_{j=1}^{\infty} j^2 P_j - \mu^2 \end{aligned}$$

Goal: find $E[X_n]$ and $\text{Var}[X_n]$

Note $X_n = \sum_{i=1}^{X_{n-1}} Z_i$. Now, condition on X_{n-1} (5)

$$\begin{aligned} E[X_n] &= E[E[X_n | X_{n-1}]] \\ &= E\left[E\left[\sum_{i=1}^{X_{n-1}} Z_i \mid X_{n-1}\right]\right] \\ &= E[\mu \cdot X_{n-1}] = \mu E[X_{n-1}] \end{aligned}$$

Recall: this is the Law of Iterated Expectations

$$\text{So } E[X_1] = \mu E[X_0] = \mu X_0$$

$$E[X_2] = \mu E[X_1] = \mu \cdot \mu X_0 = \mu^2 X_0 \quad (6)$$

$$\vdots$$
$$E[X_n] = \mu^n X_0$$

$$V[X_n] = \underbrace{E[V[X_n | X_{n-1}]]}_{(1)} + \underbrace{V[E[X_n | X_{n-1}]]}_{(2)}$$

$$\textcircled{2}: V[\mu X_{n-1}] = \mu^2 V[X_{n-1}]$$

$$\textcircled{1}: E\left[V\left[\sum_{i=1}^{X_{n-1}} z_i \mid X_{n-1}\right]\right] \quad \textcircled{7}$$

$$= E\left[\sum_{i=1}^{X_{n-1}} V[z_i]\right] = E[\sigma^2 X_{n-1}]$$

$$= \sigma^2 E[X_{n-1}] = \sigma^2 \mu^{n-1} X_0$$

$$\therefore V[X_n] = \sigma^2 \mu^{n-1} X_0 + \mu^2 V[X_{n-1}]$$

$$= \sigma^2 \mu^{n-1} X_0 + \mu^2 [\sigma^2 \mu^{n-2} X_0 + \mu^2 V[X_{n-2}]]$$

$$= \sigma^2 (\mu^{n-1} + \mu^n) X_0 + \mu^4 V[X_{n-2}]$$

$$= \sigma^2 (\mu^{n-1} + \mu^n) X_0 + \mu^4 [\sigma^2 \mu^{n-3} X_0 + \mu^2 V[X_{n-3}]] \quad \textcircled{8}$$

$$= \sigma^2 (\mu^{n-1} + \mu^n + \mu^{n+1}) X_0 + \mu^6 V[X_{n-3}]$$

...

$$= \sigma^2 (\underbrace{\mu^{n-1} + \mu^n + \mu^{n+1} + \dots + \mu^{2n-2}}_{n \text{ terms}}) X_0 + \mu^{2n} \underbrace{V[X_0]}_0$$

$$\therefore V[X_n] = \sigma^2 (\mu^{n-1}) (1 + \mu + \dots + \mu^{n-1}) X_0$$

$$= \sigma^2 \mu^{n-1} \frac{1 - \mu^n}{1 - \mu} X_0 \quad \text{if } \mu \neq 1$$

$$= n\sigma^2 X_0$$

$$\text{if } \mu = 1$$

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Let π_0 = prob that the population dies out,
given that $X_0 = 1$

$$\text{That is, } \pi_0 = \lim_{n \rightarrow \infty} P[X_n = 0 | X_0 = 1]$$

Case 1: $\mu < 1$

$$E[X_n] = \mu^n \quad (\text{since } X_0 = 1)$$

$$\text{So } \mu^n = E[X_n] = \sum_{j=0}^{\infty} j P[X_n = j]$$

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$$= \sum_{j=1}^{\infty} j P[X_n = j]$$

$$\geq \sum_{j=1}^{\infty} 1 \cdot P[X_n = j] = P[X_n \geq 1]$$

$$= 1 - P[X_n = 0]$$

$$\mu^n \geq 1 - P[X_n = 0]$$

$$\text{So } \lim_{n \rightarrow \infty} \mu^n \geq 1 - \lim_{n \rightarrow \infty} P[X_n = 0]$$

$$\therefore \lim_{n \rightarrow \infty} P[X_n = 0] = 1$$

$$\therefore \pi_0 = 1$$

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Case 2: $\mu = 1$ It can be shown that, again, $\pi_0 = 1$ Case 3: $\mu > 1$

$$\pi_0 = P[\text{population dies out}]$$

$$= \sum_{j=0}^{\infty} P[\text{population dies out} \mid X_1 = j] P_j$$

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

↑ by independence

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Example 1: $P_0 = .5, P_1 = .25, P_2 = .25, X_0 = 1$

$$\begin{aligned} \mu &= \sum_j j P_j = 0(.5) + 1(.25) + 2(.25) \\ &= .75 \end{aligned} \quad \therefore \pi_0 = 1$$

$$E[X_n] = \mu^n X_0 = (.75)^n$$

$$\begin{aligned} E[Z_i^2] &= \sum_j j^2 P_j = 0^2(.5) + 1^2(.25) + 2^2(.25) \\ &= 1.25 \end{aligned}$$

$$\sigma^2 = 1.25 - \mu^2 = \frac{11}{16} = .6875$$

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$$\begin{aligned}
 V[X_n] &= \sigma^2 \mu^{n-1} \frac{1-\mu^n}{1-\mu} X_0 \\
 &= \frac{11}{16} (.75)^{n-1} \frac{1-.75^n}{.25}
 \end{aligned}$$

Example 2 : $p_0 = .25$, $p_1 = .25$, $p_2 = .5$

$$\mu = 0(.25) + 1(.25) + 2(.5) = 1.25$$

$$\sigma^2 = \frac{11}{16} = .6875$$

HW #1 follows this page, due Thur 4/13

Stat 568
HW1
Due 4/13/17

63. For the Markov chain with states 1, 2, 3, 4 whose transition probability matrix P is as specified below find f_{i3} and s_{i3} for $i = 1, 2, 3$.

$$P = \begin{bmatrix} 0.4 & 0.2 & 0.1 & 0.3 \\ 0.1 & 0.5 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.2 & 0.1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

64. Consider a branching process having $\mu < 1$. Show that if $X_0 = 1$, then the expected number of individuals that ever exist in this population is given by $1/(1 - \mu)$. What if $X_0 = n$?
65. In a branching process having $X_0 = 1$ and $\mu > 1$, prove that π_0 is the *smallest* positive number satisfying Equation (4.20).

Hint: Let π be any solution of $\pi = \sum_{j=0}^{\infty} \pi^j P_j$. Show by mathematical induction that $\pi \geq P\{X_n = 0\}$ for all n , and let $n \rightarrow \infty$. In using the induction argue that

$$P\{X_n = 0\} = \sum_{j=0}^{\infty} (P\{X_{n-1} = 0\})^j P_j$$

66. For a branching process, calculate π_0 when
- (a) $P_0 = \frac{1}{4}, P_2 = \frac{3}{4}$.
 - (b) $P_0 = \frac{1}{4}, P_1 = \frac{1}{2}, P_2 = \frac{1}{4}$.
 - (c) $P_0 = \frac{1}{6}, P_1 = \frac{1}{2}, P_3 = \frac{1}{3}$.