

Branching Process

(Ross 4.1)

Population Modeling :

$\frac{X_0}{\text{initial P.P.}}$

$\frac{X_1}{\text{Gen. 1}}$

$\frac{X_2}{\text{Gen. 2}}$

...

Each individual has

$$\phi(\text{reproduce } j \text{ offspring}) = \phi_j$$

If $X_k = 0$, then no one left. ~~then~~

$$\text{So } X_{k+1} = 0.$$

$\Rightarrow P_{00} = 1$ state 0 is recurrent.

If $P_0 > 0$, then

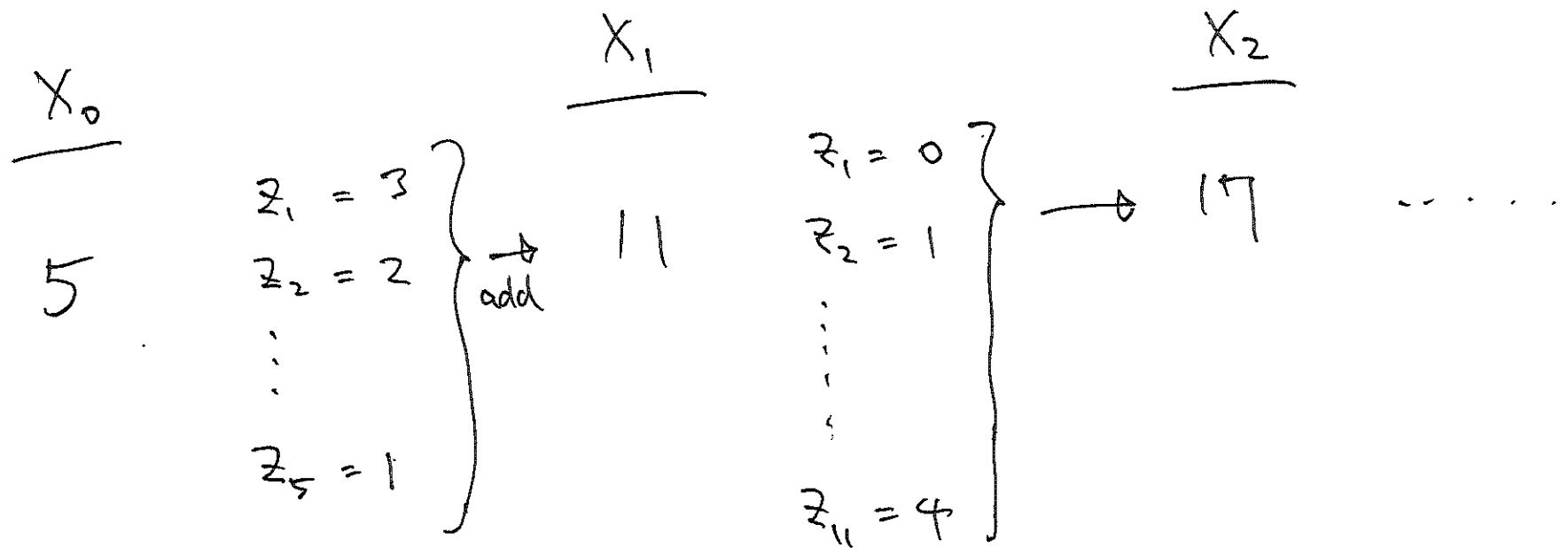
$$\begin{aligned} P_{i0} &= P(\text{all of } i \text{ people have 0 offspring}) \\ &= P_0^i > 0. \end{aligned}$$

Since this is same as

$$P(\text{go from } i \text{ to } 0 \text{ and never come back}) > 0.$$

All state i are transient.

\Rightarrow This process either die out or diverge.



$$P(z_i = j) = P_j \quad \text{i.i.d.}$$

(same for all individual.)

(independent ")

(same for each generation)

$P_0 > 0$. it is possible to have no offspring .

$$Z_k \sim \overset{\text{distribution}}{\text{?}} \Rightarrow P_0, P_1, P_2, \dots$$

$$X_n = \sum_{k=1}^{X_{n-1}} Z_k$$

$$\mu = E(Z_k)$$

$$\sigma^2 = v(Z_k)$$

What are $E(X_n)$ and $v(X_n)$?

$$E(X_n) = E\left(\sum_{k=1}^{X_{n-1}} Z_k\right)$$

$$= E\left(E\left(\sum_{k=1}^{X_{n-1}} Z_k \mid X_{n-1} = i\right)\right)$$

$$= E\left(\sum_{k=1}^i E(Z_k)\right)$$

$$= E(i\mu)$$

$$= \mu E(\overset{i}{\cancel{X_{n-1}}}) = \mu E(X_{n-1}).$$

$$\text{If } X_0 = a,$$

$$E(X_0) = a$$

$$E(X_1) = a\mu$$

$$E(X_2) = a\mu^2$$

\vdots

$$E(X_n) = a\mu^n$$

$$E(X_n) = \mu E(X_{n-1})$$

$$\text{Var}(X_n) = E[\text{Var}(X_n | X_{n-1})] + \text{Var}[E(X_n | X_{n-1})]$$

$$\begin{cases} E(X_n | X_{n-1}) = E\left(\sum_{k=1}^{X_{n-1}} \overset{\text{not random}}{\downarrow} Z_k\right) = X_{n-1} \mu \\ \text{Var}(X_n | X_{n-1}) = \text{Var}\left(\sum_{k=1}^{X_{n-1}} \overset{\text{not random}}{\downarrow} Z_k\right) = X_{n-1} \text{Var}(Z_k) = X_{n-1} \sigma^2 \end{cases}$$

$$\begin{cases} E(X_{n-1} \sigma^2) = \sigma^2 E(X_{n-1}) \\ V(X_{n-1} \mu) = \mu^2 V(X_{n-1}) \end{cases}$$

$$\boxed{\text{Var}(X_n) = \sigma^2 a \mu^{n-1} + \mu^2 V(X_{n-1})}$$

$$X_0 = a$$

$$\text{Var}(X_0) = 0$$

$$V(X_n) = a\sigma^2\mu^{n-1} + \mu^2 V(X_{n-1})$$

$$\text{Var}(X_1) = a\sigma^2\mu^0$$

$$\text{Var}(X_2) = a\sigma^2\mu + a\sigma^2\mu^2 = a\sigma^2(\mu + \mu^2)$$

$$\text{Var}(X_3) = a\sigma^2\mu^2 + a\sigma^2(\mu + \mu^2)\mu^2$$

$$= a\sigma^2\mu^2(1 + \mu + \mu^2)$$

$$\text{Var}(X_n) = \begin{cases} a\sigma^2\mu^{n-1} \left(\frac{1-\mu^n}{1-\mu} \right) & \mu \neq 1 \\ n\sigma^2 a & \mu = 1 \end{cases}$$

Prob. of Die out ~~Alt~~

$$E(X_n) = a\mu^n \quad \text{if } \mu < 1, \quad \text{then}$$

$$\lim_{n \rightarrow \infty} E(X_n) = 0.$$

Then observe that

$$E(X_n) = \sum_{j=0}^{\infty} j P(X_n = j)$$

$$\geq \sum_{j=1}^{\infty} 1 \cdot P(X_n = j) = P(X_n \geq 1).$$

Therefore, if $\mu < 1$,

$$\lim_{n \rightarrow \infty} P(X_n \geq 1) = 0,$$

i.e., population will die out.

(if $\mu = 1$, this is still true.)

If $\mu > 1$, (assume $X_0 = 1$),

$$P(\text{Die out})$$

$$= \sum_{j=0}^{\infty} \underbrace{P(\text{Die out} \mid X_1 = j)}_{\downarrow P(\text{Die out})^j} P_j$$

$$P(\text{P.O.}) = \sum_{j=0}^{\infty} P(\text{D.O.})^j P_j$$

$P(\text{P.O.})$ = smallest positive number satisfying above eqn.