

## Branching process example

$$S = \mathbb{N}$$

Dynamics: 
$$X_{n+1} = \sum_{k=1}^{X_n} z_k^{(n+1)} = \sum_{k=1}^{\infty} z_k \mathbb{1}_{(k \leq X_n)}$$

Thus

$$X_{n+1} = \varphi(X_n, \underline{z}^{(n+1)})$$

where  $\underline{z}^{(n+1)} = \{ z_k^{(n+1)} ; k \geq 1 \}$ ,  $z_k^{(n+1)} \in \mathbb{N}$

$\varphi(x, \underline{z}) = \sum_{k=1}^{\infty} z_k \mathbb{1}_{(k \leq x)}$

*sequence* (under  $\underline{z}$ )

*generations* (over  $z_k^{(n+1)}$ )

*offspring # k* (under  $z_k^{(n+1)}$ )

The sequences are i.i.d

$\Rightarrow$

$X$  is a Markov chain

$$\varphi(i, z) = \sum_{k=1}^i z_k = \sum_{k=1}^{\infty} z_k \mathbb{1}_{(k \leq i)}$$

Transition

$$\begin{aligned} P_{ij} &= P(\varphi(i, z^{(i)}) = j) \\ &= P\left(\sum_{k=1}^i z_k^{(i)} = j\right) \end{aligned}$$

We have seen that those quantities could be expressed in terms of generating functions. If  $G \equiv$  pgf for  $z^{(1)}$ , then

$$\text{pgf}\left(\sum_{k=1}^i z_k^{(i)}\right) = G^i$$

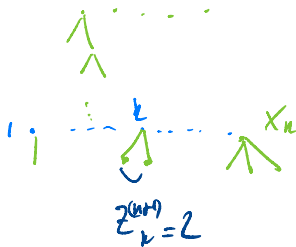
Thus

$$P\left(\sum_{k=1}^i z_k^{(i)} = j\right) = \frac{1}{j!} \left\{ \begin{array}{l} \text{coefficient of } s^j \text{ for} \\ \text{the function } [G(s)]^i \end{array} \right\}$$

# Branching process case (1)

State space:

$$S = \mathbb{N}$$



Markov property: We have seen

- $X_{n+1} = \sum_{k=1}^{X_n} Z_k^{(n+1)} = \varphi(X_n, \mathbf{Z}^{(n+1)})$
- $\mathbf{Z}^{(n)} = \{Z_k^{(n)}; k \geq 1\}$  is a sequence
- $\varphi(x, \mathbf{z}) = \sum_{k=1}^x z_k$
- $\{\mathbf{Z}^{(n)}; n \geq 1\}$  i.i.d family  
     $\hookrightarrow$  with  $(Z_k^{(n)})_{k \geq 1}$  i.i.d with common pgf  $G$

Thus

$X$  is a Markov chain

## Branching process case (2)

Transition probability: We have

$$\begin{aligned} p_{ij} &= \mathbf{P} \left( \sum_{k=1}^i Z_k^{(1)} = j \right) \\ &= \frac{1}{j!} \times \text{Coefficient of } s^j \text{ in } (G(s))^i \end{aligned}$$

$n$ -step transition: We obtain

$$p_{ij}(n) = \frac{1}{j!} \times \text{Coefficient of } s^j \text{ in } \widehat{(G_n(s))^i} = (G^{o_n}(s))^i$$

# Outline

- 1 Markov processes
- 2 Classification of states
- 3 Classification of chains
- 4 Stationary distributions and the limit theorem
  - Stationary distributions
  - Limit theorems
- 5 Reversibility
- 6 Chains with finitely many states
- 7 Branching processes revisited

# Questions about Markov chains

## Main questions

- 1 Does the MC  $X_n$  go to  $\infty$  when  $n \rightarrow \infty$ ?
- 2 Does it return to state  $i$  after  $n = 0$ ?
- 3 How often does it return to  $i$ ?
- 4 What is the range of  $X_n(\omega)$ ?

## Methodologies to answer those questions

- 1 We have seen: pgf's for random walks and branching
- 2 **Now: Markov chain methods**

# Persistent and transient states

## Definition 11.

Let

- $X$  Markov chain
- $i$  state in  $S$

Then

- ①  $i$  is called **persistent** or **recurrent** if

*we will always  
return to  $i$*

$$\mathbf{P}(X_n = i \text{ for some } n \geq 1 | X_0 = i) = 1$$

- ②  $i$  is called **transient** if

*we are not sure to  
return to  $i$*

$$\mathbf{P}(X_n = i \text{ for some } n \geq 1 | X_0 = i) < 1$$

# First passage time probabilities

## Definition 12.

Let

- $X$  Markov chain and  $i, j$  states in  $S$

Then we define

- 1 Probability that  
↪ 1st visit to  $j$  starting from  $i$  takes place at step  $n$ :

$$f_{ij}(n) = \mathbf{P}(X_1 \neq j, \dots, X_{n-1} \neq j, X_n = j | X_0 = i)$$

- 2 Probability that  $X$  ever visits  $j$  starting from  $i$ :

$$f_{ij} = \sum_{n=1}^{\infty} f_{ij}(n)$$



Recall that  $f_{ij}(n) = P(X_1 \neq j, \dots, X_{n-1} \neq j, X_n = j \mid X_0 = i)$

Set

$$T_j = \inf \{ k \geq 1 ; X_k = j \}$$

Then (i)  $f_{ij}(n) = P(T_j = n \mid X_0 = i)$

$$(ii) f_{ij} = \sum_{n=1}^{\infty} f_{ij}(n) =$$

$$= \sum_{n=1}^{\infty} P(T_j = n \mid X_0 = i)$$

$$= P(T_j < \infty \mid X_0 = i)$$

If  $f_{ii} = 1$ , then  $P(T_i < \infty \mid X_0 = i) = 1$

$\Rightarrow i$  persistent

## Alternative definition for $f_{ij}(n)$

First visit to  $j$ : We set  $T_j = \infty$  if there is no visit to  $j$ , and

$$T_j = \inf \{n \geq 1; X_n = j\}$$

Expression for  $f_{ij}(n)$ : We have

$$\begin{aligned} f_{ij}(n) &= \mathbf{P}(X_1 \neq j, \dots, X_{n-1} \neq j, X_n = j | X_0 = i) \\ &= \mathbf{P}(T_j = n | X_0 = i) \end{aligned}$$

# Some pgf's

Pgf's  $P$  and  $F$ : We set

*generating function*

$$P_{ij}(s) = \sum_{n=0}^{\infty} p_{ij}(n) s^n,$$

*pgf for  $T_j$  conditioned on  $(X_0=i)$*

$$F_{ij}(s) = \sum_{n=0}^{\infty} \underbrace{f_{ij}(n)}_{P(T_j=n | X_0=i)} s^n$$

Remarks:

$$p_{ij}(0) = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{otherwise} \end{cases}$$

- 1 Conventions above:  $p_{ij}(0) = \delta_{ij}$  and  $f_{ij}(0) = 0$
- 2  $i$  persistent iff  $f_{ii} = 1$
- 3 For  $|s| < 1$ , the series  $P_{ij}(s)$  and  $F_{ij}(s)$  are convergent
- 4  $P_{ij}(1)$  and  $F_{ij}(1)$  are defined through Abel's theorem
- 5  $f_{ij} = F_{ij}(1)$

# Relation between $F$ and $P$

## Theorem 13.

Let  $X_n$  be a Markov chain with transition  $p$ . Then

- 1  $P_{ii}$  and  $F_{ii}$  satisfy

$$P_{ii}(s) = 1 + F_{ii}(s)P_{ii}(s)$$

- 2 For  $i \neq j$ , the function  $P_{ij}$  verifies

$$P_{ij}(s) = F_{ij}(s)P_{jj}(s)$$