

## STAT253/317 Lecture 8 Generating Functions

For a non-negative-integer-valued random variable  $T$ , the generating function of  $T$  is the expected value of  $s^T$  as a function of  $s$

$$G(s) = E[s^T] = \sum_{k=0}^{\infty} s^k P(T = k),$$

in which  $s^T$  is defined as 0 if  $T = \infty$ .

Since  $0 \leq P(T = k) \leq 1$ , the generating function is always defined for  $-1 \leq s \leq 1$

## Examples of Generating Functions

- ▶ If  $T$  has a geometric distribution:  $P(T = k) = p(1 - p)^k$ ,  $k = 0, 1, 2, \dots$ , the generating function of  $T$  is

$$G(s) = \sum_{k=0}^{\infty} s^k P(T = k) = \sum_{k=0}^{\infty} s^k p(1-p)^k = \frac{p}{1 - (1-p)s}$$

- ▶ If  $T$  has a Binomial distribution  $P(T = k) = \binom{n}{k} p^k (1-p)^{n-k}$ ,  $k = 0, 1, 2, \dots, n$ , the generating function of  $T$  is

$$\begin{aligned} G(s) &= \sum_{k=0}^{\infty} s^k P(T = k) = \sum_{k=0}^{\infty} s^k \binom{n}{k} p^k (1-p)^{n-k} \\ &= (ps + (1-p))^n \end{aligned}$$

# Properties of Generating Function

$$G(s) = E[s^T] = \sum_{k=0}^{\infty} s^k P(T = k)$$

- ▶  $G(s)$  is a power series converging absolutely for all  $-1 \leq s \leq 1$ .  
since  $0 \leq P(T = k) \leq 1$  and  $\sum_k P(T = k) \leq 1$ .

- ▶  $G(1) = P(T < \infty) \begin{cases} = 1 & \text{if } T \text{ is finite w/ prob. } 1 \\ < 1 & \text{otherwise} \end{cases}$

- ▶  $P(T = k) = \frac{G^{(k)}(0)}{k!}$

Knowing  $G(s) \Leftrightarrow$  Knowing  $P(T = k)$  for all  $k = 0, 1, 2, \dots$

## More Properties of Generating Functions

$$G(s) = E[s^T] = \sum_{k=0}^{\infty} s^k P(T = k)$$

- ▶  $E[T] = \lim_{s \rightarrow 1^-} G'(s)$  if it exists because

$$G'(s) = \frac{d}{ds} E[s^T] = E[Ts^{T-1}] = \sum_{k=1}^{\infty} s^{k-1} k P(T = k).$$

- ▶  $E[T(T-1)] = \lim_{s \rightarrow 1^-} G''(s)$  if it exists because

$$G''(s) = E[T(T-1)s^{T-2}] = \sum_{k=2}^{\infty} s^{k-2} k(k-1) P(T = k)$$

- ▶ If  $T$  and  $U$  are **independent** non-negative-integer-valued random variables, with generating function  $G_T(s)$  and  $G_U(s)$  respectively, then the generating function of  $T + U$  is

$$G_{T+U}(s) = E[s^{T+U}] = E[s^T]E[s^U] = G_T(s)G_U(s)$$

### 4.5.3 Random Walk w/ Reflective Boundary at 0

- ▶ State Space =  $\{0, 1, 2, \dots\}$
- ▶  $P_{01} = 1, P_{i,i+1} = p, P_{i,i-1} = 1 - p = q$ , for  $i = 1, 2, 3 \dots$
- ▶ Only one class, irreducible
- ▶ For  $i < j$ , define

$$N_{ij} = \min\{m > 0 : X_m = j | X_0 = i\}$$

= time to reach state  $j$  starting in state  $i$

- ▶ Observe that  $N_{0n} = N_{01} + N_{12} + \dots + N_{n-1,n}$   
By the Markov property,  $N_{01}, N_{12}, \dots, N_{n-1,n}$  are indep.
- ▶ Given  $X_0 = i$

$$N_{i,i+1} = \begin{cases} 1 & \text{if } X_1 = i + 1 \\ 1 + N_{i-1,i}^* + N_{i,i+1}^* & \text{if } X_1 = i - 1 \end{cases} \quad (1)$$

where  $N_{i-1,i}^* \sim N_{i-1,i}$ ,  $N_{i,i+1}^* \sim N_{i,i+1}$ , and  $N_{i-1,i}^*, N_{i,i+1}^*$  are indep.

## Generating Function of $N_{i,i+1}$

Let  $G_i(s)$  be the generating function of  $N_{i,i+1}$ . From (1), and by the independence of  $N_{i-1,i}^*$  and  $N_{i,i+1}^*$ , we get that

$$G_i(s) = ps + qE[s^{1+N_{i-1,i}^*+N_{i,i+1}^*}] = ps + qsG_{i-1}(s)G_i(s)$$

So

$$G_i(s) = \frac{ps}{1 - qsG_{i-1}(s)} \quad (2)$$

Since  $N_{01}$  is always 1, we have  $G_0(s) = s$ . Using the iterative relation (2), we can find

$$G_1(s) = \frac{ps}{1 - qsG_0(s)} = \frac{ps}{1 - qs^2} = ps \sum_{k=0}^{\infty} (qs^2)^k = \sum_{k=0}^{\infty} pq^k s^{2k+1}$$

$$\text{So } P(N_{12} = n) = \begin{cases} pq^k & \text{if } n = 2k + 1 \text{ for } k = 0, 1, 2, \dots \\ 0 & \text{if } n \text{ is even} \end{cases}$$

Similarly,

$$\begin{aligned} G_2(s) &= \frac{ps}{1 - qsG_1(s)} = \frac{ps(1 - qs^2)}{1 - q(1 + p)s^2} \\ &= \frac{ps}{1 - q(1 + p)s^2} - \frac{pqs^3}{1 - q(1 + p)s^2} \\ &= ps \sum_{k=0}^{\infty} (q(1 + p)s^2)^k - pqs^3 \sum_{k=0}^{\infty} (q(1 + p)s^2)^k \\ &= \sum_{k=0}^{\infty} pq^k(1 + p)^k s^{2k+1} - \sum_{k=0}^{\infty} pq^{k+1}(1 + p)^k s^{2k+3} \\ &= ps + \sum_{k=1}^{\infty} pq^k[(1 + p)^k - (1 + p)^{k-1}]s^{2k+1} \\ &= ps + \sum_{k=1}^{\infty} p^2 q^k(1 + p)^{k-1} s^{2k+1} \end{aligned}$$

So

$$P(N_{23} = n) = \begin{cases} p & \text{if } n = 1 \\ p^2 q^k(1 + p)^{k-1} & \text{if } n = 2k + 1 \text{ for } k = 1, 2, \dots \\ 0 & \text{if } n \text{ is even} \end{cases}$$

## Mean of $N_{i,i+1}$

Recall that  $G'_i(1) = E(N_{i,i+1})$ . Let  $m_i = E(N_{i,i+1}) = G'_i(1)$ .

$$\begin{aligned} G'_i(s) &= \frac{p(1 - qsG_{i-1}(s)) + ps(qG_{i-1}(s) + qsG'_{i-1}(s))}{(1 - qsG_{i-1}(s))^2} \\ &= \frac{p + pqs^2 G'_{i-1}(s)}{(1 - qsG_{i-1}(s))^2} \end{aligned}$$

Since  $N_{i,i+1} < \infty$ ,  $G_i(1) = 1$  for all  $i = 0, 1, \dots, n-1$ . We have

$$m_i = G'_i(1) = \frac{p + pqG'_{i-1}(1)}{(1 - q)^2} = \frac{1 + qG'_{i-1}(1)}{p} = \frac{1}{p} + \frac{q}{p}m_{i-1}$$

We get the same iterative equation as in Lecture 7.



## 4.7 Branching Processes Revisit

Recall a Branching Process is a population of individuals in which

- ▶ all individuals have the same lifespan, and
- ▶ each individual will produce a random number of offsprings at the end of its life

Let  $X_n$  = size of the  $n$ th generation,  $n = 0, 1, 2, \dots$ . Let  $Z_{n,i} = \#$  of offsprings produced by the  $i$ th individuals in the  $n$ th generation. Then

$$X_{n+1} = \sum_{i=1}^{X_n} Z_{n,i} \quad (3)$$

Suppose  $Z_{n,i}$ 's are i.i.d with probability mass function

$$P(Z_{n,i} = j) = P_j, \quad j \geq 0.$$

We suppose the non-trivial case that  $P_j < 1$  for all  $j \geq 0$ .  $\{X_n\}$  is a Markov chain with state space  $= \{0, 1, 2, \dots\}$ .

## Generating Functions of the Branching Processes

Let  $g(s) = E[s^{Z_{n,i}}] = \sum_{k=0}^{\infty} P_k s^k$  be the generating function of  $Z_{n,i}$ , and  $G_n(s)$  be the generating function of  $X_n$ ,  $n = 0, 1, 2, \dots$ . Then  $\{G_n(s)\}$  satisfies the following two iterative equations.

$$(i) \quad G_{n+1}(s) = G_n(g(s)) \quad \text{for } n = 0, 1, 2, \dots$$

$$(ii) \quad G_{n+1}(s) = g(G_n(s)) \quad \text{if } X_0 = 1, \text{ for } n = 0, 1, 2, \dots$$

*Proof of (i).*

$$\begin{aligned} E[s^{X_{n+1}} | X_n] &= E \left[ s^{\sum_{i=1}^{X_n} Z_{n,i}} \right] = E \left[ \prod_{i=1}^{X_n} s^{Z_{n,i}} \right] \\ &= \prod_{i=1}^{X_n} E[s^{Z_{n,i}}] \quad \text{by indep. of } Z_{n,i}'\text{'s} \\ &= \prod_{i=1}^{X_n} g(s) \quad \text{as } g(s) = E[s^{Z_{n,i}}] \\ &= g(s)^{X_n} \end{aligned}$$

From which, we have

$$G_{n+1}(s) = E[s^{X_{n+1}}] = E[E[s^{X_{n+1}} | X_n]] = E[g(s)^{X_n}] = G_n(g(s))$$

since  $G_n(s) = E[s^{X_n}]$ .

## Proof of (ii) $G_{n+1}(s) = g(G_n(s))$ if $X_0 = 1$

Suppose there are  $k$  individuals in the first generation ( $X_1 = k$ ). Let  $Y_i$  be the number offspring of the  $i$ th individual in the first generation in the  $(n+1)$ st generation. Obviously,

$$X_{n+1} = Y_1 + \dots + Y_k.$$

Observe  $Y_1, \dots, Y_k$ 's are indep and each has the same distn. as  $X_n$  since they are all the size of the  $n$ th generation of a single ancestor. Thus, by ndep. of  $Y_i$ 's

$$E[s^{X_{n+1}} | X_1 = k] = E[s^{Y_1 + \dots + Y_k}] = E\left[\prod_{i=1}^k s^{Y_i}\right] = \prod_{i=1}^k E[s^{Y_i}]$$

Since  $Y_i$ 's have the same dist'n as  $X_n$  and  $G_n(s) = E[s^{X_n}]$ , we have

$$E[s^{X_{n+1}} | X_1 = k] = \prod_{i=1}^k G_n(s) = (G_n(s))^k$$

Since  $X_0 = 1$ ,  $X_1 = Z_{1,1}$ , and hence  $P(X_1 = k) = P_k$ .

$$G_{n+1}(s) = E[s^{X_{n+1}}] = \sum_{k=0}^{\infty} E[s^{X_{n+1}} | X_1 = k] P_k = \sum_{k=0}^{\infty} (G_n(s))^k P_k = g(G_n(s)),$$

where the last equality comes from that  $g(s) = \sum_{k=0}^{\infty} P_k s^k$ .

## Example

Suppose  $X_0 = 1$ , and  $(P_0, P_1, P_2) = (1/4, 1/2, 1/4)$ . Find the distribution of  $X_2$ .

*Sol.*

$$g(s) = \frac{1}{4}s^0 + \frac{1}{2}s^1 + \frac{1}{4}s^2 = (1+s)^2/4.$$

Since  $X_0 = 1$ ,  $G_0(s) = E[s^{X_0}] = E[s^1] = s$ . From (i) we have

$$G_1(s) = G_0(g(s)) = g(s) = (1+s)^2/4$$

$$\begin{aligned} G_2(s) &= G_1(g(s)) = \frac{1}{4}\left(1 + \frac{1}{4}(1+s)^2\right)^2 = \frac{1}{64}(5 + 2s + s^2)^2 \\ &= \frac{1}{64}(25 + 20s + 14s^2 + 4s^3 + s^4) = \sum_{k=0}^{\infty} P(X_2 = k)s^k \end{aligned}$$

The coefficient of  $s^k$  in the polynomial of  $G_2(s)$  is the chance that  $X_2 = k$ .

$k$	0	1	2	3	4
$P(X_2 = k)$	$\frac{25}{64}$	$\frac{20}{64}$	$\frac{14}{64}$	$\frac{4}{64}$	$\frac{1}{64}$

and  $P(X_2 = k) = 0$  for  $k \geq 5$ .

# Extinction Probability of a Branching Process

$$\begin{aligned}\text{Let } \pi_0 &= \lim_{n \rightarrow \infty} P(X_n = 0 | X_0 = 1) \\ &= P(\text{the population will eventually die out} | X_0 = 1)\end{aligned}$$

As  $G_n(s) = E[s^{X_n}] = \sum_{k=0}^{\infty} P(X_n = k)s^k$ , plugging in  $s = 0$ , we get

$$G_n(0) = P(X_n = 0) = P(\text{extinct by the } n\text{th generation}).$$

Recall that if  $X_0 = 1$ ,  $G_1(s) = g(s)$ , and  $G_{n+1}(s) = g(G_n(s))$ . We can compute  $G_n(0)$  iteratively as follows

$$\begin{aligned}G_1(0) &= g(0) \\ G_{n+1}(0) &= g(G_n(0)), \quad n = 1, 2, 3, \dots\end{aligned}$$

Finally, we can get the extinction probability by taking the limit

$$\pi_0 = \lim_{n \rightarrow \infty} G_n(0).$$

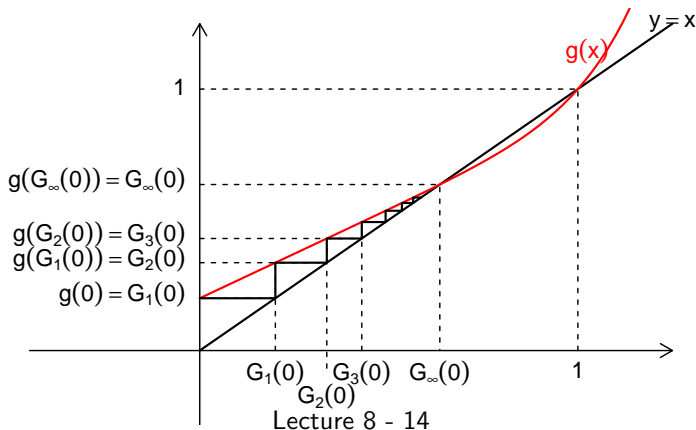
## Extinction Probability of a Branching Process

If  $X_0 = 1$ , the extinction probability  $\pi_0$  is a **smallest root** of the equation

$$g(s) = s \quad (4)$$

in the range  $0 \leq s \leq 1$ , where  $g(s) = \sum_{k=0}^{\infty} P_k s^k$  is the generating function of  $Z_{n,i}$ .

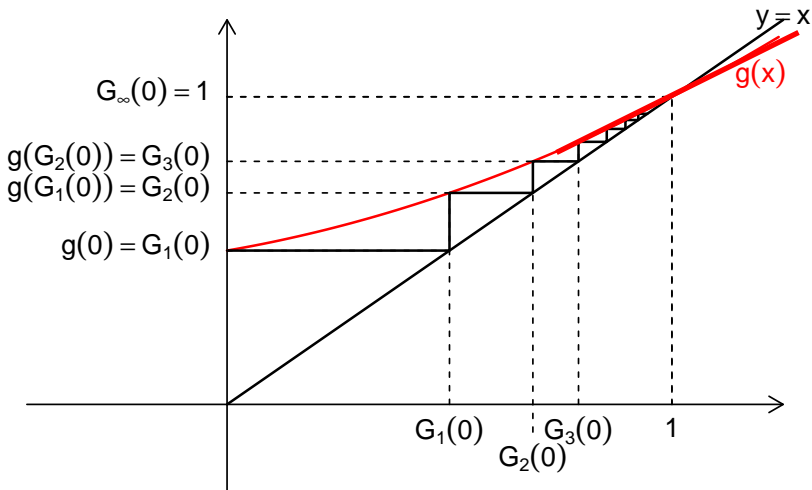
*Proof.*



## A Branching Process Will Become Extinct If $\mu \leq 1$

Let  $\mu = E[Z_{n,i}] = \sum_{j=0}^{\infty} jP_j$ . If  $\mu \leq 1$ , the extinction probability  $\pi_0$  is 1 unless  $P_1 = 1$ . mu = g'(1) = slope of the curve g(x) at x = 1

*Proof.*



## Formal Proof

Let  $h(s) = g(s) - s$ . Since  $g(1) = 1$ ,  $g'(1) = \mu$ ,

$$h(1) = g(1) - 1 = 0,$$

$$h'(s) = \left( \sum_{j=1}^{\infty} jP_j s^{j-1} \right) - 1 \leq \left( \sum_{j=1}^{\infty} jP_j \right) - 1 = \mu - 1 \quad \text{for } 0 \leq s < 1$$

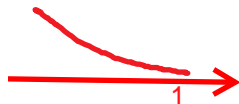
Thus  $\mu \leq 1 \Rightarrow h'(s) \leq 0$  for  $0 \leq s < 1$

$\Rightarrow h(s)$  is non-increasing in  $[0, 1)$

$\Rightarrow h(s) > h(1) = 0$  for  $0 \leq s < 1$

$\Rightarrow g(s) > s$

$\Rightarrow$  There is no root in  $[0, 1)$ .



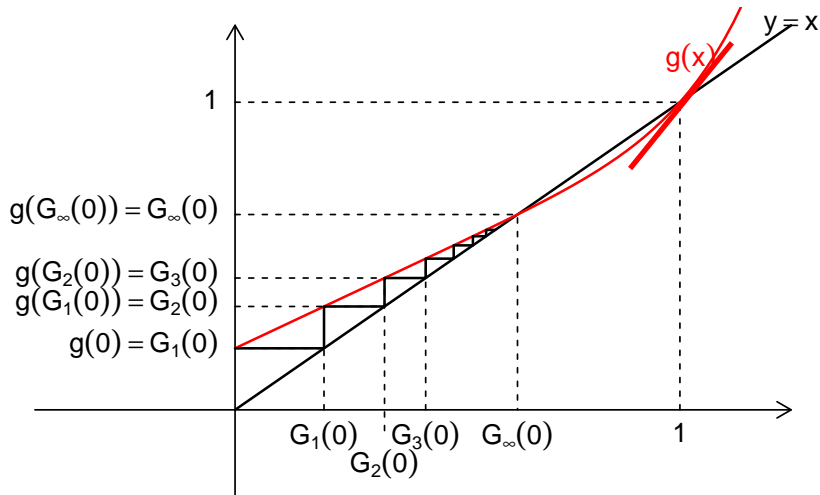
for  $0 \leq s < 1$



## Extinction Probability When $\mu > 1$

If  $\mu > 1$ , there is a unique root of the equation  $g(s) = s$  in the domain  $[0, 1)$ , and that is the extinction probability.

*Proof.*



## Formal Proof

Let  $h(s) = g(s) - s$ . Observe that

$$h(0) = g(0) = P_0 > 0$$

$$h'(0) = g'(0) - 1 = P_1 - 1 < 0$$

$$\text{Then } \mu > 1 \Rightarrow h'(1) = \mu - 1 > 0$$

$\Rightarrow h(s)$  is increasing near 1

$\Rightarrow h(1 - \delta) < h(1) = 0$  for  $\delta > 0$  small enough

Since  $h(s)$  is continuous in  $[0, 1)$ , there must be a root to  $h(s) = s$ . The root is unique since

$$h''(s) = g''(s) = \sum_{j=2}^{\infty} j(j-1)P_j s^{j-2} \geq 0 \quad \text{for } 0 \leq s < 1$$

$h(s)$  is convex in  $[0, 1)$ .

$g(s)$  is also convex.

