

# STAT253/317 Winter 2014 Lecture 8

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Generating Functions

Lecture 8 - 1

## Generating Functions

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

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

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

▶  $G(s)$  is a power series converging absolutely for all  $-1 < s < 1$ .

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

▶

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

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

Lecture 8 - 2

## 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

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

Lecture 8 - 3

## 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.

Lecture 8 - 4

## 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 iterating relation (2), we can find

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

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

Lecture 8 - 5

Similarly,

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

So

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

Lecture 8 - 6

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

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

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

$$= \frac{p + pqqs^2G'_{i-1}(s)}{(1 - qsG_{i-1}(s))^2}$$

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

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

$$= \frac{1}{p} + \frac{q}{p}\left(\frac{1}{p} + \frac{q}{p}m_{i-2}\right)$$

$$= \frac{1}{p}\left[1 + \frac{q}{p} + \left(\frac{q}{p}\right)^2 + \dots + \left(\frac{q}{p}\right)^{i-1}\right] + \left(\frac{q}{p}\right)^i m_0$$

Since  $N_{01} = 1$ , which implies  $m_0 = 1$ .

$$m_i = \begin{cases} \frac{1-(q/p)^i}{p-q} + \left(\frac{q}{p}\right)^i & \text{if } p \neq 0.5 \\ 2i + 1 & \text{if } p = 0.5 \end{cases}$$

Lecture 8 - 7

## Generating Function of $N_{10}$

Let  $G(s)$  be the generating function of  $N_{10}$ . From (3), we know that

$$G(s) = \frac{1}{2}s + \frac{1}{2}E[s^{1+N_{21}+N'_{10}}] = \frac{1}{2}s + \frac{1}{2}sG(s)^2$$

which is a quadratic equation in  $G(s)$ . The two roots are

$$G(s) = \frac{1 \pm \sqrt{1-s^2}}{s}$$

Since  $G(s)$  must lie between 0 and 1 when  $0 < s < 1$ . So

$$G(s) = \frac{1 - \sqrt{1-s^2}}{s}$$

Note that

$$G'(s) = \frac{1}{\sqrt{1-s^2} + 1-s^2}, \quad E[N_{10}] = \lim_{s \rightarrow 1^-} G'(s) = \infty$$

which implies  $E[N_{00}] = 1 + E[N_{10}] = \infty$ .

$\Rightarrow$  Symmetric random walk is null recurrent.

Lecture 8 - 9

From all the above we have

$$G(s) = \frac{1 - \sum_{k=0}^{\infty} \binom{1/2}{k} (-s^2)^k}{s} = - \sum_{k=1}^{\infty} (-1)^k \binom{1/2}{k} s^{2k-1}$$

$$= \sum_{k=1}^{\infty} \frac{1}{2^{2k-1}(2k-1)} \binom{2k-1}{k} s^{2k-1}$$

So the distribution of  $N_{10}$  is

$$P(N_{10} = 2k-1) = \frac{1}{2^{2k-1}(2k-1)} \binom{2k-1}{k}$$

$$P(N_{10} = 2k) = 0$$

for  $k = 0, 1, 2, \dots$

Lecture 8 - 11

## Example: Symmetric Random Walk on $(-\infty, \infty)$

State space =  $\{\dots, -2, -1, 0, 1, 2, \dots\}$

$$P_{i,i+1} = P_{i,i-1} = 1/2, \quad \text{for all integer } i$$

- ▶ 1 classes, recurrent, null-recurrent or positive-recurrent?
- ▶ For  $i, j$ , define  $N_{ij} = \min\{m > 0 : X_m = j | X_0 = i\}$ .
- ▶ Note  $N_{00} = 1 + N_{10}$
- ▶ Given  $X_0 = 1$

$$N_{10} = \begin{cases} 1 & \text{if } X_1 = 0 \\ 1 + N_{21} + N'_{10} & \text{if } X_1 = 2 \end{cases} \quad (3)$$

Note  $N_{21}$  and  $N'_{10}$  are independent and have the same distribution as  $N_{10}$  (Why?)

Lecture 8 - 8

The power series expansion of  $G(s) = \frac{1-\sqrt{1-s^2}}{s}$  can be found via Newton's binomial formula

$$(1-s^2)^\alpha = \sum_{k=0}^{\infty} \binom{\alpha}{k} (-s^2)^k$$

where  $\binom{\alpha}{0} = 1$  and for  $k \geq 1$ ,  $\binom{\alpha}{k} = \prod_{i=0}^{k-1} (\alpha - i)/k!$ .

$$\binom{1/2}{k} = \frac{\frac{1}{2}(\frac{1}{2}-1)(\frac{1}{2}-2)(\frac{1}{2}-3)\dots(\frac{1}{2}-k+1)}{k!}$$

$$= \frac{(-1)^{k-1} 1 \cdot 1 \cdot 3 \cdot 5 \dots (2k-3)}{2^k k!}$$

$$= \frac{(-1)^{k-1} 1 \cdot 1 \cdot 3 \cdot 5 \dots (2k-3)(2k-1)}{2^k k!(2k-1)}$$

$$= \frac{(-1)^{k-1} \frac{(2k-1)!}{2^{k-1}(k-1)!}}{2^k k!(2k-1)} = \frac{(-1)^{k-1}}{2^{2k-1}(2k-1)} \binom{2k-1}{k}$$

Lecture 8 - 10

## 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

$$X_{n+1} = \sum_{i=1}^{X_n} Z_{n,i} \quad (4)$$

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 =  $\{0, 1, 2, \dots\}$ .

Lecture 8 - 12

## 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 equation

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

*Proof of (i).*

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

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):* HW today

Lecture 8 - 13

## 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  $k \geq 5$ .

Lecture 8 - 14

## Extinction Probability

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

### Proposition

The extinction probability  $\pi_0$  is a root of the equation

$$g(s) = s \quad (5)$$

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

$$\begin{aligned} \pi_0 &= P(\text{population dies out}) \\ &= \sum_{j=0}^{\infty} P(\text{population dies out} | X_1 = j) P_j \\ &= \sum_{j=0}^{\infty} \pi_0^j P_j = g(\pi_0) \end{aligned}$$

Lecture 8 - 15

## Proposition I

Let  $\mu = E[Z_{n,i}] = \sum_{j=0}^{\infty} j P_j$ . If  $\mu \leq 1$ , the extinction probability  $\pi_0$  is 1 unless  $P_1 = 1$ .

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

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

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

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  $0 \leq s < 1$

$\Rightarrow g(s) > s$

for  $0 \leq s < 1$

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

Lecture 8 - 16

## Proposition II

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.* 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  $h(s) = 0$ . 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  $[0, 1)$ .

Lecture 8 - 17