

# Generating Functions

Stochastics

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- 2 Properties
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## Definition

If  $X$  is a discrete random variable, then the *probability generating function* or simply *generating function* of  $X$  is

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Example. If  $X$  is the value of a roll with a fair 6-sided die, then

$$G(z) = \frac{1}{6}z^1 + \frac{1}{6}z^2 + \frac{1}{6}z^3 + \frac{1}{6}z^4 + \frac{1}{6}z^5 + \frac{1}{6}z^6.$$

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If  $X \sim PGEO(p)$ , then

$$G(z) = \sum_{k=0}^{\infty} p(1-p)^k z^k = p \sum_{k=0}^{\infty} ((1-p)z)^k = \frac{p}{1 - (1-p)z}.$$

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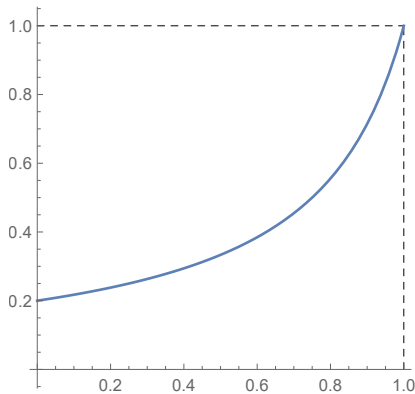
(keeping in mind that  $0^0 = 1$ ),

- the sum is convergent for  $z \in [0, 1]$ ,
- $G(z)$  is analytic on  $[0, 1)$ ,
- $G(z)$  is increasing and convex on  $[0, 1]$ .

## Example

Example. The generating function of PGEO(4/5) is

$$G(z) = \frac{1 - 4/5}{1 - 4z/5} = \frac{1}{5 - 4z}.$$



## Mean and variance

## Lemma

- $\mathbb{E}(X) = G'(1),$
- $\mathbb{D}(X) = \sqrt{G''(1) + G'(1) - (G'(1))^2}.$



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Proof.

$$\begin{aligned} G'(z) \Big|_{z=1} &= \sum_{k=0}^{\infty} (z^k)' \mathbb{P}(X = k) \Big|_{z=1} = \sum_{k=0}^{\infty} k z^{k-1} \mathbb{P}(X = k) \Big|_{z=1} = \\ &= \sum_{k=1}^{\infty} k z^{k-1} \mathbb{P}(X = k) \Big|_{z=1} = \sum_{k=1}^{\infty} k \mathbb{P}(X = k) = \mathbb{E}(X). \end{aligned}$$

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The proof for the formula for  $\mathbb{D}(X)$  is similar.

# Recovering the original distribution

The original distribution can be entirely computed from the generating function.

## Lemma

$$\mathbb{P}(X = k) = \frac{G^{(k)}(z)|_{z=0}}{k!},$$

where  $G^{(k)}$  denotes the  $k$ -th derivative of  $G$ .

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Specifically,

$$\mathbb{P}(X = 0) = G(0),$$

$$\mathbb{P}(X = 1) = G'(0),$$

$$\mathbb{P}(X = 2) = \frac{G''(0)}{2}$$

etc.

## Further properties

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$$\begin{aligned} G_{X+1}(z) &= \sum_{k=0}^{\infty} \mathbb{P}(X + 1 = k) z^k = z \sum_{k=1}^{\infty} \mathbb{P}(X = k - 1) z^{k-1} \\ &= z \sum_{k=0}^{\infty} \mathbb{P}(X = k) z^k = z G_X(z). \end{aligned}$$

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$$\begin{aligned} G_{2X}(z) &= \sum_{k=0}^{\infty} \mathbb{P}(2X = k)z^k = \sum_{k=0}^{\infty} \mathbb{P}(X = k/2)(z^2)^{k/2} = \\ &= \sum_{k=0, k \text{ even}}^{\infty} \mathbb{P}(X = k/2)(z^2)^{k/2} = \\ &= \sum_{l=0}^{\infty} \mathbb{P}(X = l)(z^2)^l = G_X(z^2). \end{aligned}$$



# Sum of two independent random variables

## Theorem

*If  $X$  and  $Y$  are independent discrete random variables with generating function  $G_X(z)$  and  $G_Y(z)$  respectively, then*

$$G_{X+Y}(z) = G_X(z)G_Y(z).$$

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Proof (sketch, through example). Let  $X$  and  $Y$  be the result of rolling two fair 6-sided dice. Let's compute  $\mathbb{P}(X + Y = 4)$ .

$$\begin{aligned}\mathbb{P}(X + Y = 4) &= \\ &= \mathbb{P}(X = 1, Y = 3) + \mathbb{P}(X = 2, Y = 2) + \mathbb{P}(X = 3, Y = 1) = \\ &= \frac{1}{6} \cdot \frac{1}{6} + \frac{1}{6} \cdot \frac{1}{6} + \frac{1}{6} \cdot \frac{1}{6} = \frac{3}{36}.\end{aligned}$$

## Sum of two independent random variables

Let's also compute the coefficient of  $z^4$  in  $G_X(z)G_Y(z)$ :

$$G_X(z)G_Y(z) = \left( \frac{1}{6}z^1 + \frac{1}{6}z^2 + \dots + \frac{1}{6}z^6 \right) \left( \frac{1}{6}z^1 + \frac{1}{6}z^2 + \dots + \frac{1}{6}z^6 \right).$$

We can get  $z^4$  by taking  $z^1$  from the first bracket and  $z^3$  from the second bracket, or  $z^2 \cdot z^2$ , or  $z^3 \cdot z^1$ . The coefficient of these terms gives

$$\frac{1}{6}z^1 \cdot \frac{1}{6}z^3 + \frac{1}{6}z^2 \cdot \frac{1}{6}z^2 + \frac{1}{6}z^3 \cdot \frac{1}{6}z^1 = \frac{3}{36}z^4.$$

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The result of the multiplication of polynomials gives exactly the probability we need.

# Deterministic numbers

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For example,

$$G_{X+1}(z) = zG_X(z)$$

follows directly from the previous lemma.

## Example - Binomial distribution

Let  $X \sim \text{BIN}(n, p)$ . Compute its generating function  $G(z)$ .

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$X$  is the number of successes from  $n$  independent trials, so by adding 1 for each success and 0 for each failure,  $X$  can be written as

$$X = Y_1 + \cdots + Y_n,$$

where  $Y_1, \dots, Y_n$  are iid Bernoulli variables with generating function  $G_Y(z) = (1 - p) + pz$ .



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$$G(z) = ((1 - p) + pz)^n.$$

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On the other hand,  $G(z)$  can be computed directly from the distribution of  $X$ :

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Due to the binomial theorem, the two formulas are equivalent.

$$G(z) = ((1-p) + pz)^n = \sum_{i=0}^n \binom{n}{i} p^i (1-p)^{n-i} z^i.$$

## Sum with a random number of terms

Let  $X_1, X_2, \dots$  be independent, identically distributed discrete random variables (iid for short) with common generating function  $G_X(z)$ , and let  $N$  be a discrete random variable with generating function  $G_N(z)$ , independent from all the  $X$ 's.

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Example. We have a blue and a red die, both are 6-sided and fair. We roll the blue die once, then roll the red die a number of times equal to the result of the blue roll. Finally we add all the results of the red rolls. What are the possible values for the sum of the red rolls?

## Sum with a random number of terms

Let  $N$  be the value of the blue die roll, and  $X_1, X_2, \dots$  be the values of the red rolls. Then  $Y = X_1 + \dots + X_N$  is the sum of the red rolls.

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If  $N = 1$ , then  $Y = X_1$ , so it has generating function

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If  $N = 2$ , then  $Y = X_1 + X_2$ , so it has generating function

$$(G_X(z))^2.$$

If  $N = 3$ , then  $Y = X_1 + X_2 + X_3$ , so it has generating function

$$(G_X(z))^3 \text{ etc.}$$

# Sum with a random number of terms

Altogether, we have

$$G_Y(z) = \mathbb{E}(z^Y) = \sum_{n=1}^6 \mathbb{E}(z^Y | N = n) \mathbb{P}(N = n) =$$
$$\frac{1}{6} G_X(z) + \frac{1}{6} (G_X(z))^2 + \cdots + \frac{1}{6} (G_X(z))^6.$$

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**Theorem**

*If  $X_1, X_2, \dots$  are iid discrete random variables with common generating function  $G_X(z)$ , and  $N$  is a discrete random variable with generating function  $G_N(z)$ , independent from all the  $X$ 's, and  $Y = X_1 + \cdots + X_N$ , then*

$$G_Y(z) = G_N(G_X(z)).$$

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First, you will have a random number of children.

Then each of your children will have a random number of children.



## Summary

$$G(z) := \sum_{k=0}^{\infty} \mathbb{P}(X = k)z^k.$$

$$G(1) = 1, \quad G(0) = \mathbb{P}(X = 0), \quad \mathbb{P}(X = k) = \frac{G^{(k)}(0)}{k!}$$

$$\mathbb{E}(X) = G'(1), \quad \mathbb{D}(X) = \sqrt{G''(1) + G'(1) - (G'(1))^2}$$

$$G_{X+Y}(z) = G_X(z)G_Y(z) \quad \text{for } X, Y \text{ independent}$$

$$G_{X_1+\dots+X_N}(z) = G_N(G_X(z)) \quad \text{for } X_1, X_2, \dots \text{ iid and } N \text{ indep.}$$

The theorem of total expectation applies for the generating function.

# Incomplete generating functions

Sometimes we work with discrete random variables which may also take infinity as a value with positive probability. In this case,

$$\sum_{k=0}^{\infty} \mathbb{P}(X = k) = 1 - \mathbb{P}(X = \infty) \leq 1.$$

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Sometimes we work with discrete random variables which may also take infinity as a value with positive probability. In this case,

$$\sum_{k=0}^{\infty} \mathbb{P}(X = k) = 1 - \mathbb{P}(X = \infty) \leq 1.$$

For such variables, the definition of the generating function

$$G(z) = \sum_{k=0}^{\infty} \mathbb{P}(X = k)z^k$$

still makes sense, but instead of  $G(1) = 1$ , we now have

$$G(1) \leq 1,$$

with

$$\mathbb{P}(X = \infty) = 1 - G(1).$$

## Problem 1

A nonnegative discrete random variable has generating function

$$G(z) = \frac{3}{8} + \frac{3}{8}z + \frac{1}{8}z^2 + \frac{1}{8}z^3.$$

Determine the distribution of  $X$  (that is, the  $\mathbb{P}(X = k)$  probabilities for  $k = 0, 1, 2, \dots$ ). Calculate its mean and variance as well.

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Solution.

$$\begin{aligned}\mathbb{P}(X = 0) &= \frac{3}{8}, & \mathbb{P}(X = 1) &= \frac{3}{8}, \\ \mathbb{P}(X = 2) &= \frac{1}{8}, & \mathbb{P}(X = 3) &= \frac{1}{8}, \\ \mathbb{P}(X = 4) &= \mathbb{P}(X = 5) = \dots = 0.\end{aligned}$$

## Problem 1

$$G'(z) = \frac{3}{8} + \frac{2}{8}z + \frac{3}{8}z^2,$$
$$G''(z) = \frac{2}{8} + \frac{6}{8}z$$

and so

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and

$$\mathbb{E}(X) = G'(1) = 1,$$
$$\text{Var}(X) = G''(1) + G'(1) - (G'(1))^2 = 1 + 1 - 1^2 = 1.$$

## Problem 2

Alice sends a letter to Bob. Postal service is not very reliable; each day, the postman will take the letter to the logistics center with probability  $1/3$  (regardless of the past). Once the letter is in the logistics center, each day it is processed with probability  $1/5$  (regardless of the past). Once it is processed, shipping it takes 1 day. (So at best, the total delivery time is 1 day.) Let  $X$  denote the total delivery time in days. Calculate the generating function and the mean of  $X$ .



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Solution. We can write  $X$  as

$$X = X_1 + X_2 + X_3,$$

where

- $X_1$  is the number of days it takes for the postman to take the letter to the logistics center,
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$X_1, X_2$  and  $X_3$  are independent. What are their distributions?

## Problem 2

Then

$$X_1 \sim \text{PGEO}(1/3), \quad G_1(z) = \frac{1}{3 - 2z}, \quad \mathbb{E}(X_1) = 2,$$

$$X_2 \sim \text{PGEO}(1/5), \quad G_2(z) = \frac{1}{5 - 4z}, \quad \mathbb{E}(X_2) = 4,$$

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(The information that “at best, the total delivery time is 1 day” implies that the geometric distributions are pessimistic.)

## Problem 2

Then

$$\begin{aligned} X_1 &\sim \text{PGEO}(1/3), & G_1(z) &= \frac{1}{3-2z}, & \mathbb{E}(X_1) &= 2, \\ X_2 &\sim \text{PGEO}(1/5), & G_2(z) &= \frac{1}{5-4z}, & \mathbb{E}(X_2) &= 4, \\ X_3 &= 1, & G_3(z) &= z, & \mathbb{E}(X_3) &= 1. \end{aligned}$$

(The information that “at best, the total delivery time is 1 day” implies that the geometric distributions are pessimistic.)

So

$$\begin{aligned} G(z) &= G_1(z)G_2(z)G_3(z) = \frac{1}{3-2z} \cdot \frac{1}{5-4z} \cdot z, \\ \mathbb{E}(X) &= G'(1) = \mathbb{E}(X_1) + \mathbb{E}(X_2) + \mathbb{E}(X_3) = 2 + 4 + 1 = 7. \end{aligned}$$

## Problem 3

An exam has two parts, A and B. Part B of the exam may be taken only by students who pass part A. Each student passes part A with probability 0.6, independent of the others. Each student who passed part A then passes part B with probability 0.5, independent of the others. 100 students take this test. Let  $X$  denote the number of students who pass part A, and  $Y$  denote the number of students who pass part B. What is the distribution of  $X$ ? Calculate  $G_X$ , the generating function of  $X$ , then derive  $G_Y$ , the generating function of  $Y$  using  $G_X$ . Can we tell the distribution of  $Y$  from  $G_Y$ ?

## Problem 3

Solution.  $Y$  can be written as

$$Y = Z_1 + \cdots + Z_X,$$

where  $Z_1, \dots, Z_X$  correspond to students who passed part A, and  $Z_i$  is 1 if student  $i$  has passed part B and 0 if not.



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$X \sim \text{BIN}(100, 0.6)$ , so

$$G_X(z) = (0.4 + 0.6z)^{100}, \quad \text{and}$$

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which is the generating function of  $\text{BIN}(100, 0.3)$ .

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Another way to see  $Y \sim \text{BIN}(100, 0.3)$  is to use total probability. For each student, the probability that he passes part B can be calculated as

$$\begin{aligned}\mathbb{P}(\text{passes B}) &= \mathbb{P}(\text{passes B}|\text{passes A})\mathbb{P}(\text{passes A}) + \\ &\quad + \mathbb{P}(\text{passes B}|\text{does not pass A})\mathbb{P}(\text{does not pass A}) = \\ &= 0.5 \cdot 0.6 + 0 \cdot 0.4 = 0.3,\end{aligned}$$

so each of the 100 students will pass part B with probability 0.3 independently.

## Problem 3

The result can be stated as a lemma in general.

## Lemma

*If  $N \sim \text{BIN}(n, p)$  and  $X_1, X_2, \dots$  are iid with Bernoulli distribution with parameter  $q$ , then*

$$Y = X_1 + \dots + X_N$$

*has distribution  $\text{BIN}(n, pq)$ .*

## Problem 4

We roll a fair six-sided die until we get two consecutive sixes. Let  $Y$  denote the number of rolls needed to get two consecutive sixes (so for the sequence 1462655661,  $Y = 9$ ). Calculate the generating function of  $Y$ , then the mean of  $Y$ . Hint: the theorem of total expectation applies for the generating function.

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Solution. Let  $X_1$  denote the number of rolls needed to get the first six.

$X_1 \sim \text{GEO}(1/6)$ , so the generating function of  $X$  is

$$G_1(z) = \sum_{k=1}^{\infty} \frac{1}{6} \left(\frac{5}{6}\right)^{k-1} z^k = \frac{\frac{1}{6}z}{1 - \frac{5}{6}z} = \frac{z}{6 - 5z}.$$

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The corresponding total expectation formula for  $G_1(z)$  is

$$G_1(z) = \frac{1}{6}z + \frac{5}{6}zG_1(z),$$

which we can solve to get

$$G_1(z) = \frac{z}{6 - 5z}$$

again.

## Problem 4

Consider  $Y$  next, the number of rolls needed to get two consecutive sixes. Since we need at least one six first,  $Y$  can be written as

$$Y = X_1 + X_2,$$

where  $X_1$  is again the number of rolls needed to get the first six, and  $X_2$  is the number of rolls needed after the first six.  $X_1$  and  $X_2$  are independent, so

$$G_Y(z) = G_1(z)G_2(z).$$

## Problem 4

We apply total expectation to  $G_Y(z)$  according to the value of *the roll following the first 6 roll*. If it is also a 6, we are finished:  $X_2 = 1$  and  $G_Y(z) = G_1(z)z$ . This has probability  $1/6$ .

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which is a linear equation for  $G_Y(z)$  with the solution

$$G_Y(z) = \frac{z^2}{5z^2 + 30z - 36}, \quad \text{and} \quad G'_Y(1) = 42.$$