

# Generating function for a sum

## Theorem 6.

Let

- $X, Y$  random variables
- $X \perp\!\!\!\perp Y$

Then

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

Proof that  $G_z(s) = G_x(s) G_y(s)$

We have  $X \perp Y$  and  $Z = X + Y$

First method :  $f_z = f_x * f_y$

$$\Rightarrow G_z(s) = G_x(s) G_y(s)$$

Second method:

$$G_z(s) = E[S^z] = E[S^{x+y}]$$

$$= E[S^x S^y] \stackrel{X \perp Y}{=} E[S^x] E[S^y]$$

$$\Rightarrow G_z(s) = G_x(s) G_y(s)$$

## Binomial random variables

Let  $X \sim \text{Bin}(n, p)$      $n \geq 1$      $p \in (0, 1)$

Then

$$P(X = k) = \binom{n}{k} p^k q^{n-k}, \quad q = 1 - p$$

Claim :

$$G_X(s) = (q - p) + ps \quad J^n$$

Main use of  $\text{Bin}(n, p)$  r.v.:

# successes in  $n$   $\perp$  trials with success  
with proba  $p$

Another way to express that:

Let  $Y_1, \dots, Y_n$  be  $n$   $\perp$  random  
variables with  $Y_j \sim B(p)$

$$\text{Set } X = \sum_{j=1}^n Y_j = Y_1 + Y_2 + \dots + Y_n.$$

Then

$$X \sim \text{Bin}(n, p)$$

Generating function

$$G_X(s) = \prod_{j=1}^n G_{Y_j}(s) = ((1-p) + ps)^n$$

# Binomial random variable (1)

Notation:

$$X \sim \text{Bin}(n, p), \text{ for } n \geq 1, p \in (0, 1)$$

State space:

$$\{0, 1, \dots, n\}$$

Pmf:

$$\mathbf{P}(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}, \quad 0 \leq k \leq n$$

Expected value, variance and generating function:

$$\mathbf{E}[X] = np, \quad \mathbf{Var}(X) = np(1 - p), \quad G_X(s) = [(1 - p) + ps]^n$$

# Binomial random variable (2)

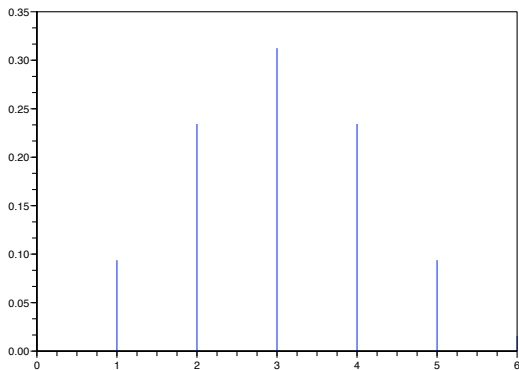
## Use 1, Number of successes in a Bernoulli trial:

- Example: Roll a dice 9 times.
- $X = \#$  of 3 obtained
- We get  $X \sim \text{Bin}(9, 1/6)$
- $\mathbf{P}(X = 2) = 0.28 = \binom{9}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^7$

## Use 2: Counting a feature in a repeated trial:

- Example: stock of 1000 pants with 10% defects
- Draw 15 times a pant at random
- $X = \#$  of pants with a defect
- We get  $X \sim \text{Bin}(15, 1/10)$

# Binomial random variable (3)



$X \in \{0, 1, \dots, 6\}$

Figure: Pmf for  $\text{Bin}(6; 0.5)$ . x-axis:  $k$ . y-axis:  $P(X = k)$

# Binomial random variable (4)

If  $n$  large  $\text{Bin}(n, p) \approx \mathcal{N}(np, np(1-p))$

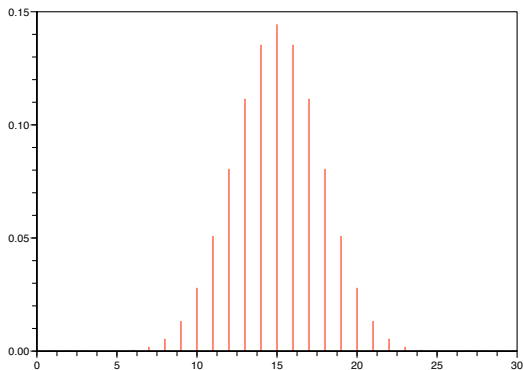


Figure: Pmf for  $\text{Bin}(30; 0.5)$ . x-axis:  $k$ . y-axis:  $\mathbf{P}(X = k)$



# Computation for $G_X$

Generating function for Bernoulli: If  $Y \sim \mathcal{B}(p)$  then

$$G_Y(s) = (1 - p) + ps$$

Decomposition of Binomial: If  $X \sim \text{Bin}(n, p)$  one can write

$$X = \sum_{i=1}^n Y_i, \quad \text{with } Y_i \text{ i.i.d, } Y_i \sim \mathcal{B}(p)$$

Computing  $G_X$ : We get

$$G_X(s) = \prod_{i=1}^n G_{Y_i}(s) = [(1 - p) + ps]^n$$

# Joint generating functions

## Definition 7.

Let

- $X_1, X_2$  random variables
- $X_1, X_2$  take values in  $\mathbb{Z}$

Then the pgf for  $(X_1, X_2)$  is

$$G_{X_1, X_2}(s_1, s_2) = \mathbf{E} [s_1^{X_1} s_2^{X_2}]$$

# Characterization of independence

## Theorem 8.

Let

- $X_1, X_2$  random variables
- $G_{X_1, X_2}$  the corresponding pgf

Then we have

$$X_1 \perp\!\!\!\perp X_2 \iff G_{X_1, X_2}(s_1, s_2) = G_{X_1}(s_1)G_{X_2}(s_2) \text{ for all } s_1, s_2$$

# Outline

- 1 Generating functions
- 2 Random walks**
- 3 Branching processes

# Definition of random walk

## Definition 9.

Let

- $X_1, \dots, X_n$  Bernoulli random variables with values  $\pm 1$ ,

$$\mathbf{P}(X_i = 1) = p, \quad \mathbf{P}(X_i = -1) = 1 - p = q$$

- The  $X_i$ 's are independent

Idea: at each step, we flip a coin to know if we are doing one step up or down

We set  $S_0 = 0$  and

$$S_n = \sum_{i=1}^n X_i$$

Then  $S$  is called **simple random walk**

# Symmetric random walk

## Definition 10.

Let

- $X_1, \dots, X_n$  Bernoulli random variables with values  $\pm 1$ ,

$$\mathbf{P}(X_i = 1) = \frac{1}{2}, \quad \mathbf{P}(X_i = -1) = \frac{1}{2}$$

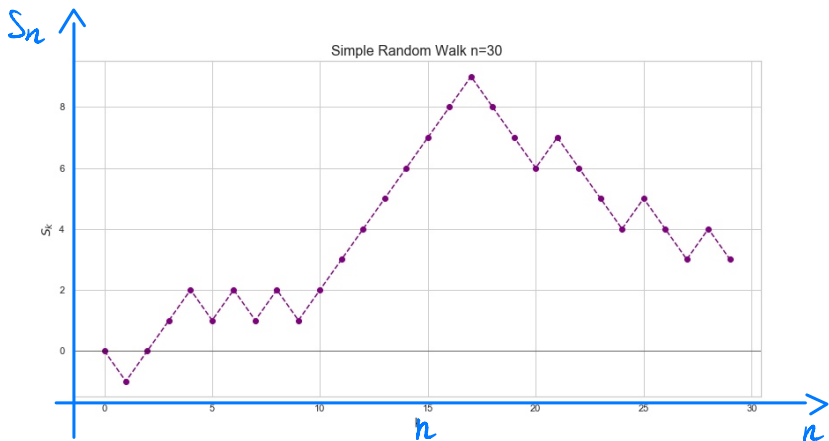
- The  $X_i$ 's are independent

We set  $S_0 = 0$  and

$$S_n = \sum_{i=1}^n X_i$$

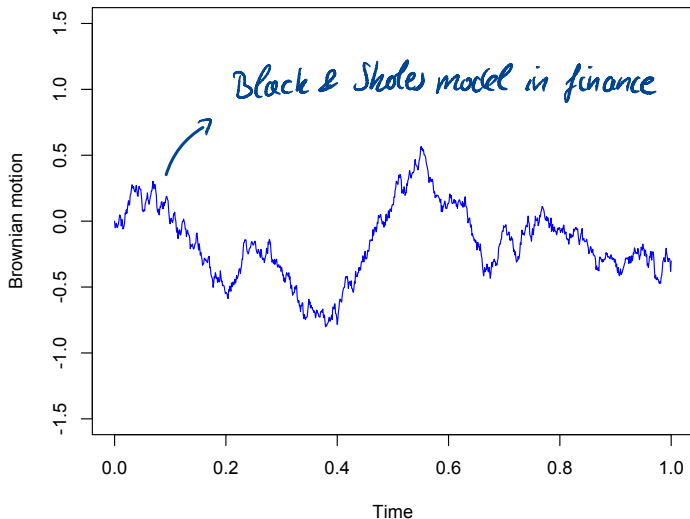
Then  $S$  is called **symmetric random walk**

# Illustration: 30 steps of a random walk



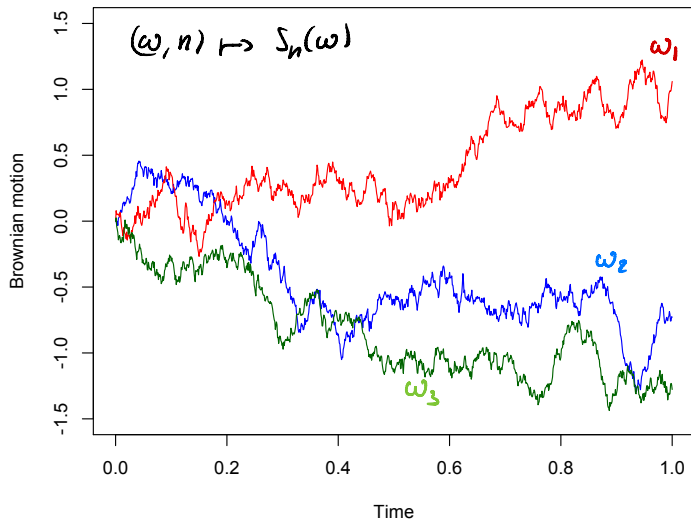
# Illustration: chaotic path (Brownian motion)

Take  $J_n$ , accelerate time, squeeze space  $\rightarrow$  Brownian motion





# Illustration: random path (Brownian motion)



# Questions about random walks

## Main questions

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

## Methodologies to answer those questions

- 1 Elementary methods based on generating functions
- 2 Later: Markov chain methods
- 3 Also useful: martingale methods

If  $X$  is a random variable, e.g.  $X \sim \text{Bin}(n, p)$

Then  $X: \Omega \longrightarrow \{0, 1, \dots, n\}$

and we should write  $X(\omega)$

Event: any  $A \subset \Omega$

Ex 1 For the Bin, one can take  $\Omega = \{0, 1\}^n$

We have  $\omega \in \Omega$ . For each  $\omega \in \Omega$ ,  
we get a value  $X(\omega)$

Ex 2 For the rw from 0 to  $n$ , one can  
take

$$\Omega = \{-1, 1\}^n$$