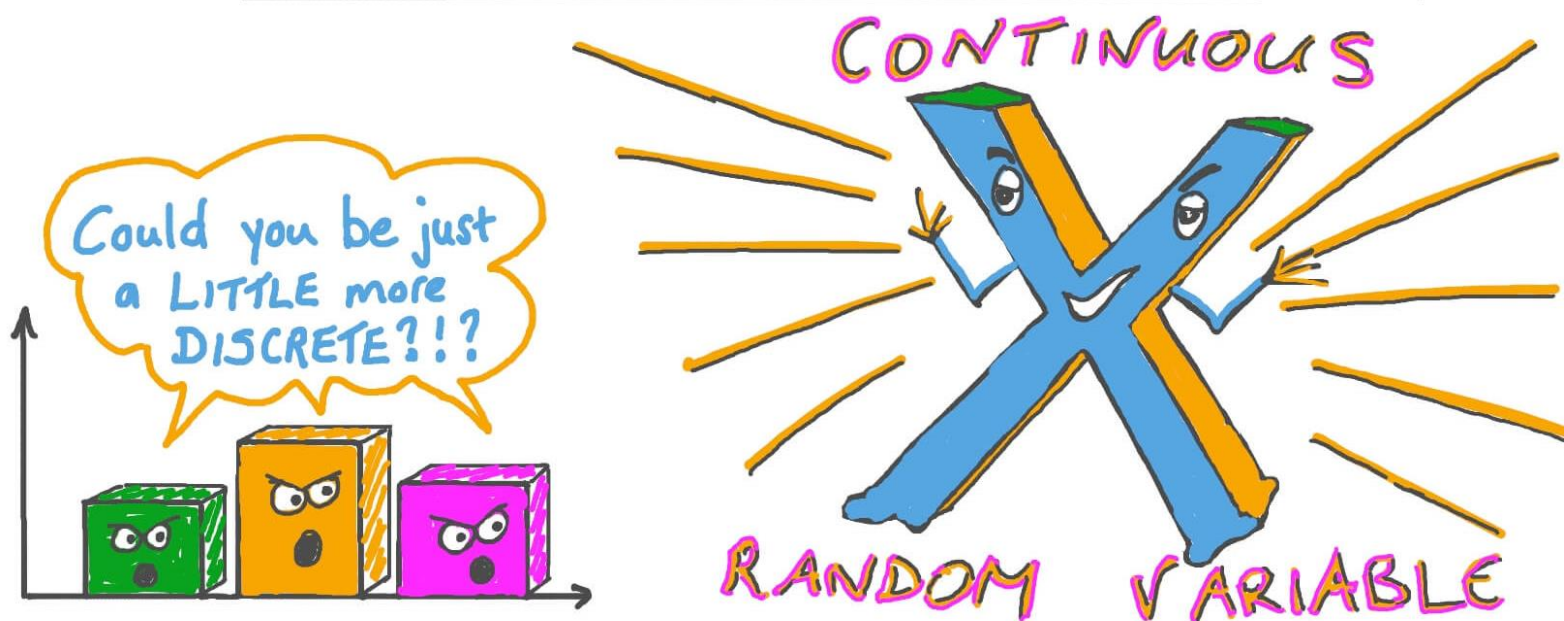


CONTINUOUS RANDOM VARIABLES



LECTURE 6

Last time

- Continuous Probability Spaces
- Anomalies with Continuous Probability

Today

- Random variables
- Probability Mass Function

GET READY FOR VOTING!

- ▶ Students from A1 section:
 1. Enter this URL: <https://app.tophat.com/e/037447>
 2. Login as needed using your Top Hat account info.
 3. If asked about enrolling, click *Enroll*.
- ▶ Students from A2 section:
 1. Enter this URL: <https://app.tophat.com/e/606363>
 2. Login as needed using your Top Hat account info.
 3. If asked about enrolling, click *Enroll*.

TOP HAT QUESTION

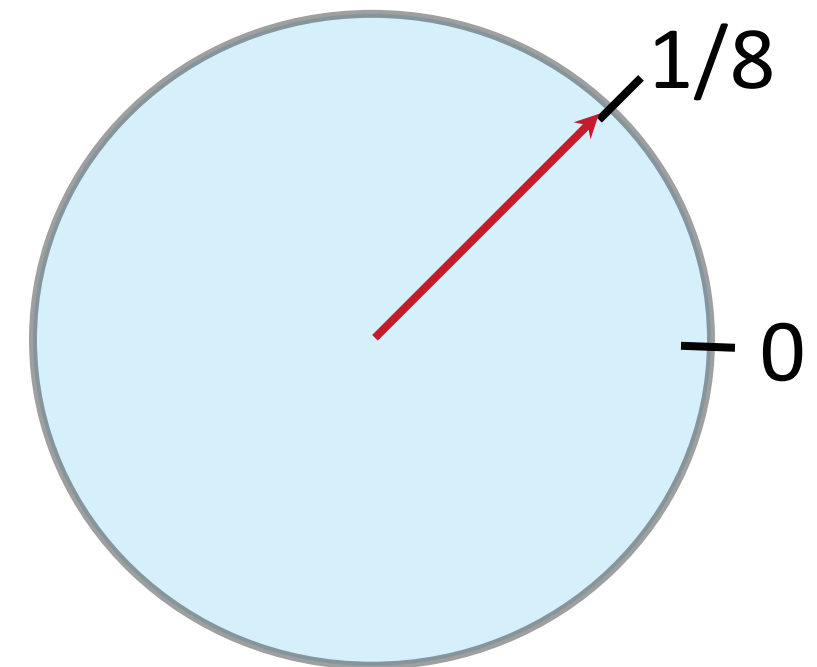
- What is the probability that the dial lands on $1/8$?

(A) $1/8 = 0.125$

(B) 0

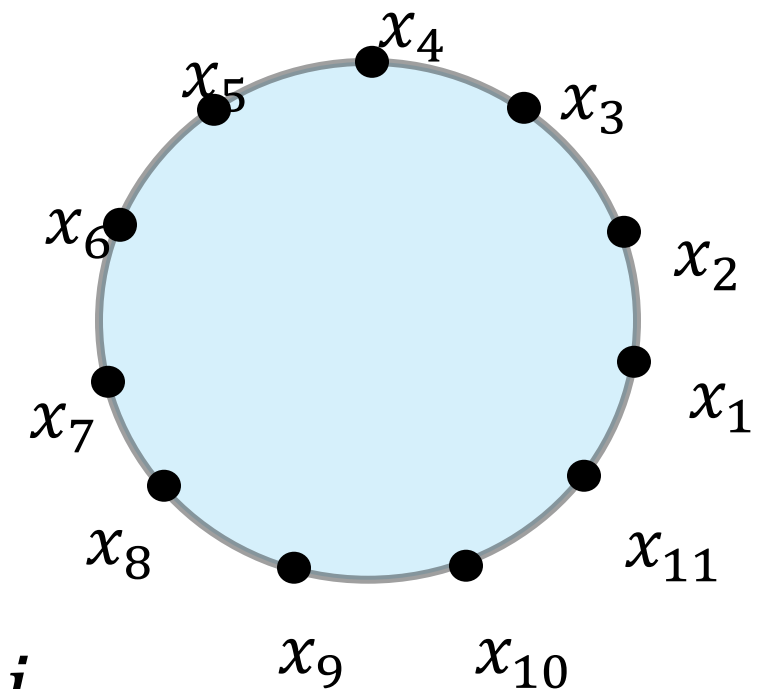
(C) 0.01

(D) none of the above



SPINNERS

- ▶ What is the probability that the dial lands on $1/8$?
- ▶ Suppose we suspect that it is, say, 0.1
- ▶ Pick 11 distinct points on the circle



by symmetry: $Pr(\text{dial lands on } x_i) = 0.1$ for all i

by additivity: $Pr(\text{dial lands on one of } x_1, \dots, x_{11}) = 11 \cdot 0.1 = 1.1$

contradiction!

SPINNERS

- ▶ Let p be the probability of an outcome, e.g., $1/8$
- ▶ We can show that $p = 0$ by contradiction
- ▶ If $p \neq 0$: pick distinct points x_1, \dots, x_k , where $k > 1/p$

by symmetry: $Pr(\text{dial lands on } x_i) = p \ \forall i \in \{1, 2, \dots, k\}$

by additivity: $Pr(\text{dial lands on one of } x_1, \dots, x_k) = k \cdot p > 1$

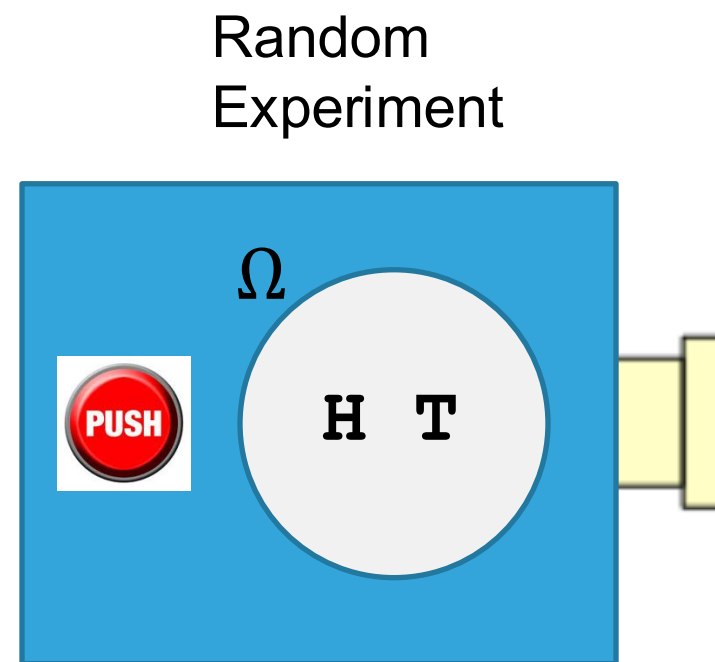
contradiction!

BEYOND EVENTS

- ▶ In many settings, we would like to know **more than the probabilities** of events
 - ▶ How many days will a system run before it fails?
 - ▶ How many people support candidate A?
 - ▶ How much money will I make if I invest in this stock?
 - ▶ How many heads in n coin tosses?
 - ▶ How many die rolls until we see a 6?
 - ▶ How much time will it take me to mine a Bitcoin on my laptop?

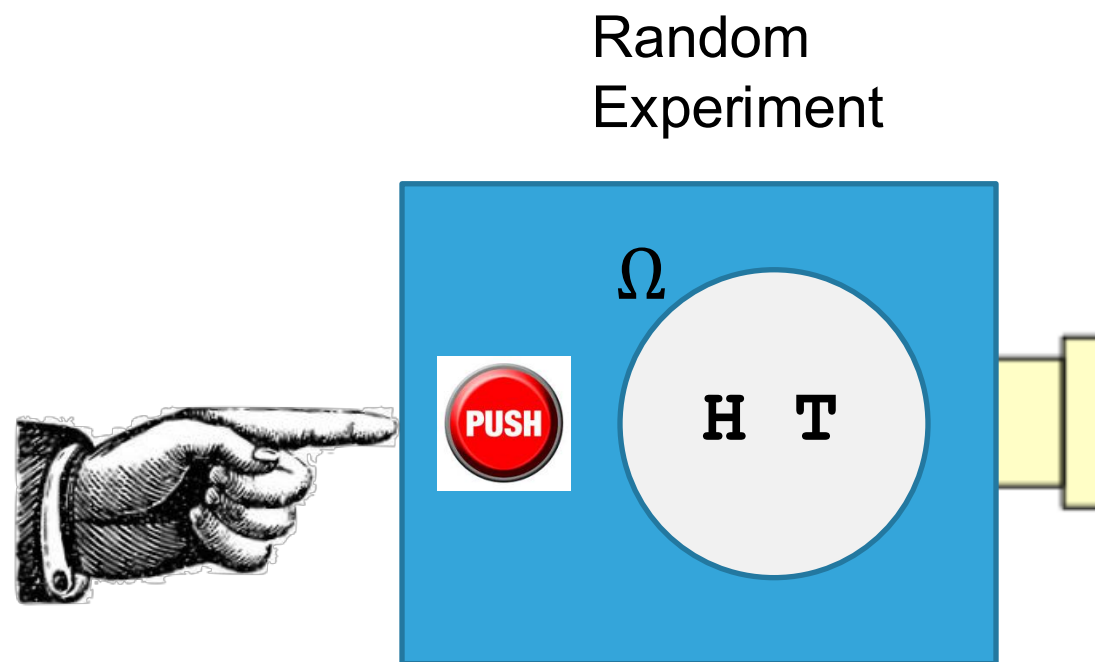
RANDOM EXPERIMENT

- ▶ A **Random Experiment** is a repeatable procedure that produces uncertain outcomes from a well-defined sample space.
- ▶ **Example: Flip a coin!**



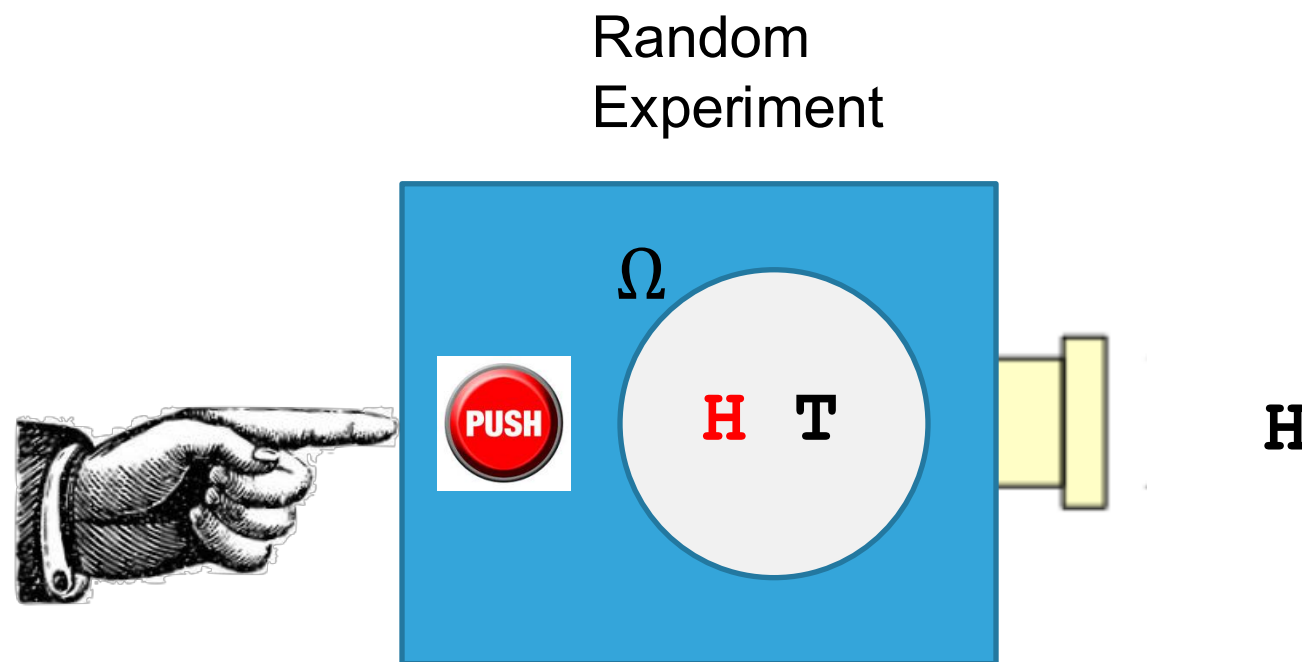
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RANDOM EXPERIMENT

- ▶ A **Random Experiment** is a repeatable procedure that produces uncertain outcomes from a well-defined sample space.
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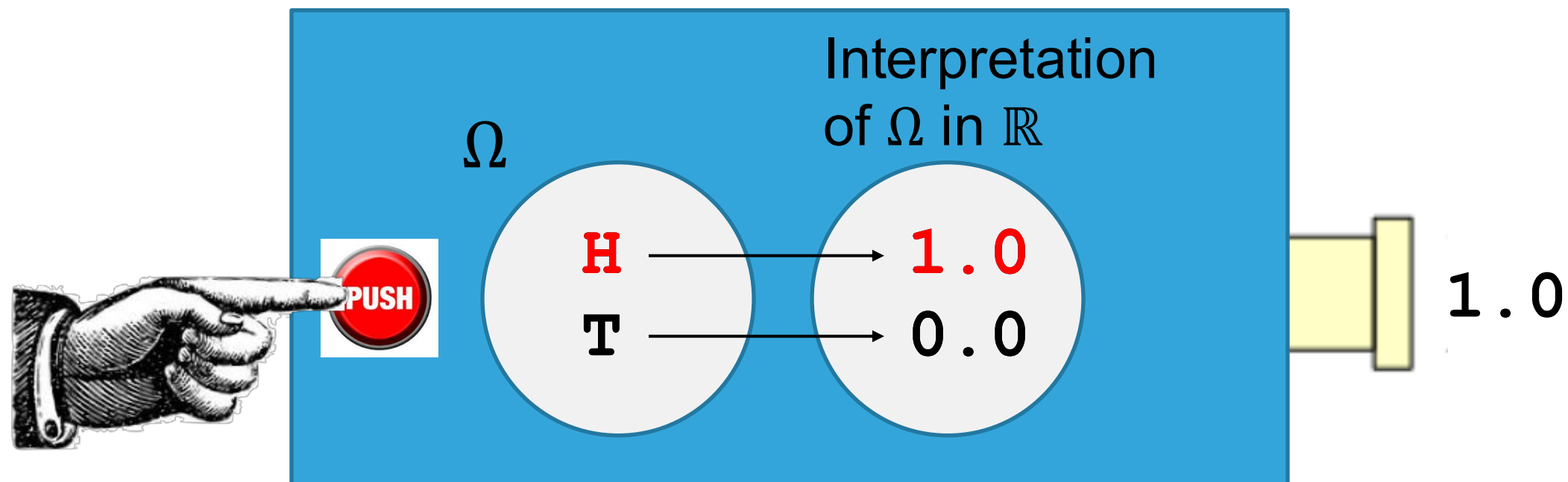


A RANDOM VARIABLE X IS A FUNCTION

- ▶ It maps a sample space Ω into \mathbb{R} , *i.e.*, *interprets* the random experiment as a real number.
- ▶ Now when an outcome is requested, the sample point is translated into a real number:

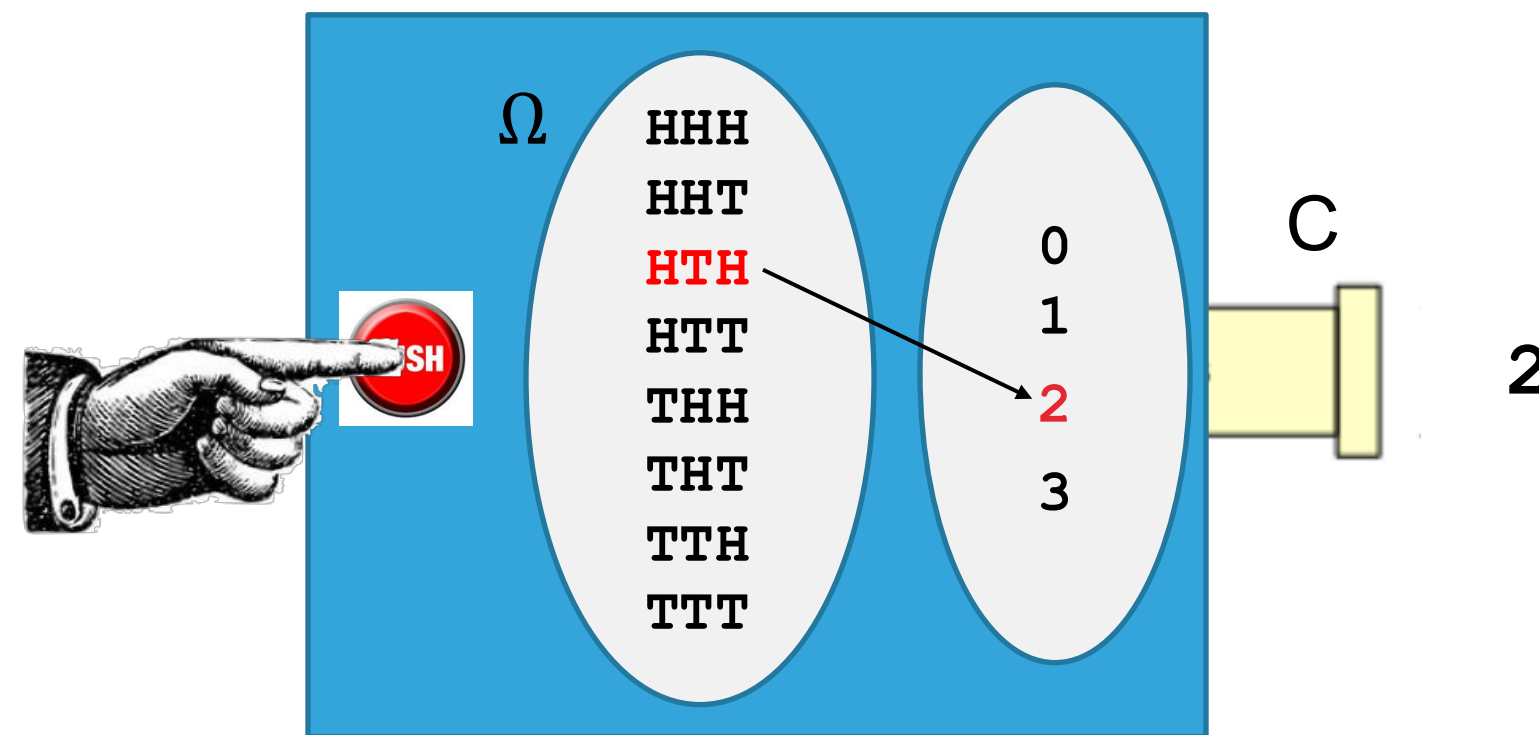
$$X : \Omega \longrightarrow \mathbb{R}$$

- ▶ X = “Flip a coin and count the number of heads showing”

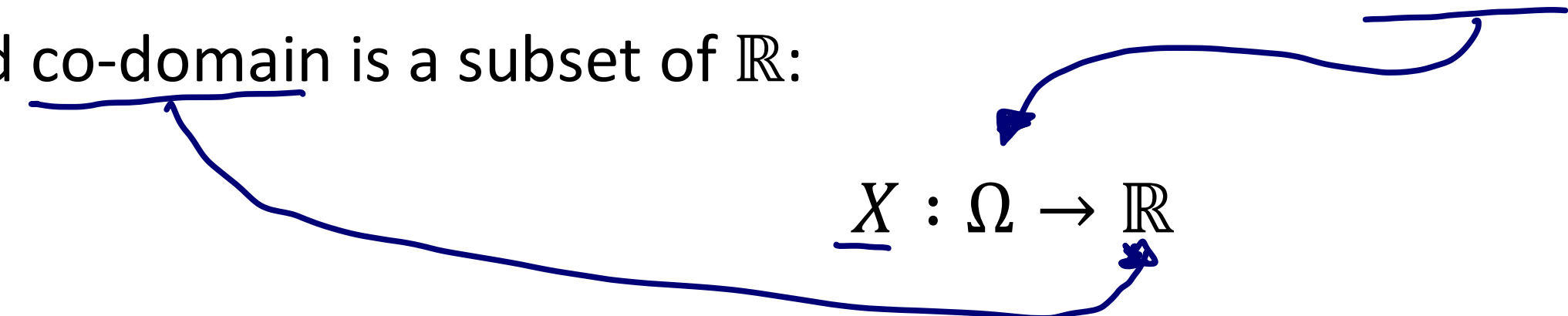
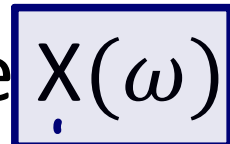


RANDOM VARIABLES

- ▶ A random variable will not in general simply rename outcomes with real numbers, it will **compute some useful information about the outcome**.
- ▶ $C = \text{"Flip 3 coins and return the number of heads showing"}$



RANDOM VARIABLES: DEFINITION

- ▶ **Definition:** A random variable X is a function whose domain is the sample space and co-domain is a subset of \mathbb{R} :

$$\underline{X} : \Omega \rightarrow \mathbb{R}$$
- ▶ X maps each outcome $\omega \in \Omega$ to a value $X(\omega) \in \mathbb{R}$

- ▶ Once an outcome is requested it is translated into a real number
- ▶ The name “random variable” is a misnomer, random variables are actually functions

RANDOM VARIABLES: DEFINITION

- ▶ **Definition:** A random variable X is a function from the sample space Ω to numbers.
- ▶ For a given value x_k , the event

$$\{X = x_k\} = \{ \omega \in \Omega : X(\omega) = x_k \}$$

- ▶ is the subset of outcomes in the sample space that make X equal to x_k .

$$C: \Omega \rightarrow \{0, 1, 2, 3\}$$

RANDOM VARIABLES - EXAMPLE

- ▶ **Experiment:** toss a fair coin 3 times
- ▶ Let C be the number of heads that appear
- ▶ How should we think about C ?

$$\Omega = \{HHH, HHT, HTH, THH, TTH, THT, HTT, TTT\}$$

$$C(HHH) = 3, \quad C(HHT) = C(HTH) = C(THH) = 2$$

$$C(TTH) = C(THT) = C(HTT) = 1$$

$$C(TTT) = 0$$

$$\{C = 2\} = \{\omega \in \Omega \mid C(\omega) = 2\} = \{HHT, HTH, THH\}$$

RANDOM VARIABLES - EXAMPLE

- ▶ **Experiment:** toss a fair coin 3 times
- ▶ Let C be the number of heads that appear
- ▶ How should we think about C ?
- ▶ C maps each outcome to a value

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RANDOM VARIABLES - EXAMPLE

- ▶ **Experiment:** toss a fair coin 3 times
- ▶ Let C be the number of heads that appear
- ▶ How should we think about C ?
- ▶ C maps each outcome to a value

$$C(HHH) = 3 \quad C(HHT) = C(HTH) = C(THH) = 2$$

$$C(TTT) = 0 \quad C(TTH) = C(THT) = C(HTT) = 1$$

- ▶ C maps each outcome to a value, i.e., C is a **function** from the sample space to values
 $C: \Omega \rightarrow \{0,1,2,3\}$

TOP HAT QUESTION

Consider the following experiment:

“I toss a coin until the first heads appears. Let Y be the total number of coin tosses.”

- ▶ What is the range of the random variable Y ?

~~A.~~ $R_Y = \{H, T\}$

B. $R_Y = \{1, 2, 3, 4, 5, \dots\}$

~~C.~~ $R_Y = \left\{ \sum \frac{1}{2^n} : n \in N \right\}$

~~D.~~ $R_Y = \left\{ \frac{1}{2}, \frac{1}{3}, \frac{1}{4}, \frac{1}{5}, \frac{1}{6}, \dots \right\}$

~~E.~~ $R_Y = \left\{ \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \frac{1}{32}, \frac{1}{64}, \dots \right\}$

Random variables and Events

$C = \#$ of heads after tossing a fair coin 3 times

RANDOM VARIABLES: NOTATION

- A very typical notion for probabilities involving a random variable X is:

$\Pr(\text{ < some Boolean expression involving } X >)$

$$\Pr(C \neq 2) = 1 - \Pr(C = 2) = \Pr(\{C = 0\} \cup \{C = 1\} \cup \{C = 3\}) = \Pr(C = 0) + \Pr(C = 1) + \Pr(C = 3)$$

$$\Pr(C = 1) = \Pr(\{C = 1\})$$

$$\Pr(X = k) = \Pr_X(k) = \Pr(\{\omega \in \Omega : X(\omega) = k\})$$

$$\Pr(X \neq k) = 1 - \Pr_X(k)$$

$$\Pr(X \leq k) = \sum_{j \leq a \leq k} \Pr_X(a)$$

- Here, $\Pr_X(k)$ means the probability that X returns k

$$\begin{aligned} \Pr(C \leq 1) &= 1 - \Pr(C > 1) \\ &= \Pr(C = 0) + \Pr(C = 1) \end{aligned}$$

$$\begin{aligned} \Pr(\{C = 1\}) &= \Pr(\{\omega \in \Omega : C(\omega) = 1\}) \\ &= \Pr(\{TTH, THT, HTT\}) \end{aligned}$$

RANDOM VARIABLES AND EVENTS

- ▶ The set of outcomes that map to specific values of an r.v. also defines an event.
- ▶ C = number of heads in three tosses

$$\underbrace{TTT}_{C=0}, \underbrace{HTT, THT, TTH}_{C=1}, \underbrace{HHT, HTH, THH}_{C=2}, \underbrace{HHH}_{C=3}$$

$$C(HHT) = C(HTH) = C(THH) = 2$$

$$Pr(C = 2) = Pr(\{HHT, HTH, THH\}) = 3/8$$

$$Pr(HHT) + Pr(HTH) + Pr(THH) = 3/8$$

$$Pr(C \leq 1) = Pr(\{TTT, HTT, THT, TTH\}) = 1/2$$

Examples of Random Variables

RANDOM VARIABLES

- ▶ Discrete random variables = discrete range
 - ▶ The possible values (outputs) are listable/countable
 - ▶ Number of H in 3 tosses, sum of two rolls of a die, number of tosses until H, ...
- ▶ Continuous random variables = continuous range
 - ▶ The possible values (outputs) form an interval of \mathbb{R}
 - ▶ Spinners, darts, ...

RANDOM VARIABLES: EXAMPLES

- ▶ **Experiment:** toss a fair coin 3 times

$$M = \begin{cases} 1 & \text{if all three tosses match (all H or all T)} \\ 0 & \text{otherwise} \end{cases}$$

$$M: \Omega \rightarrow \{0, 1\}$$

$$M(HHH) = M(TTT) = 1$$

$$M(HHT) = M(HTH) = M(HTT) = 0$$

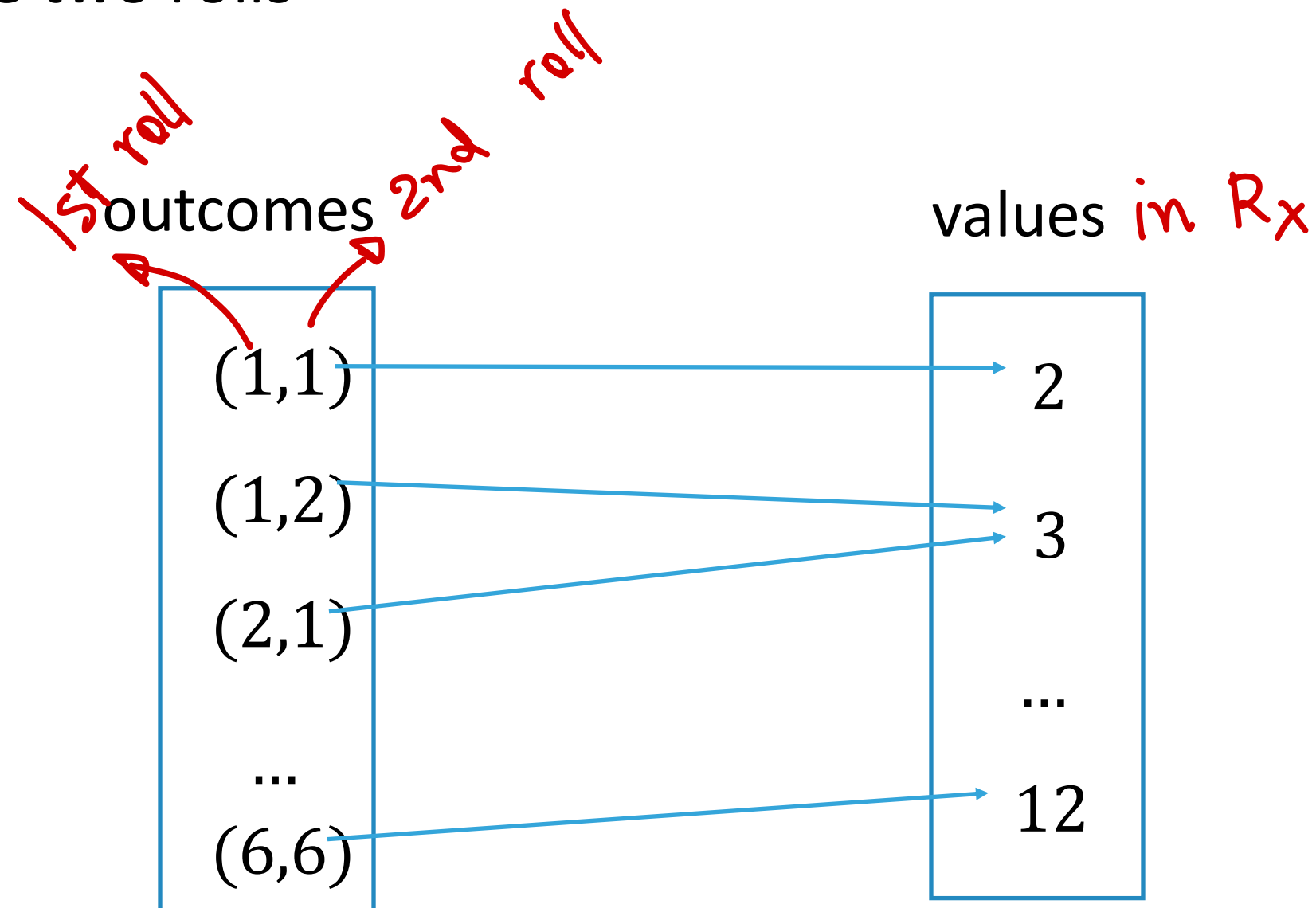
$$M(THH) = M(THT) = M(TTH) = 0$$

- ▶ An R.V. whose codomain is $\{0, 1\}$ is an indicator variable.
- ▶ Indicator variables are another way to define *events*.

→ This R.V. is super important!

RANDOM VARIABLES: EXAMPLES

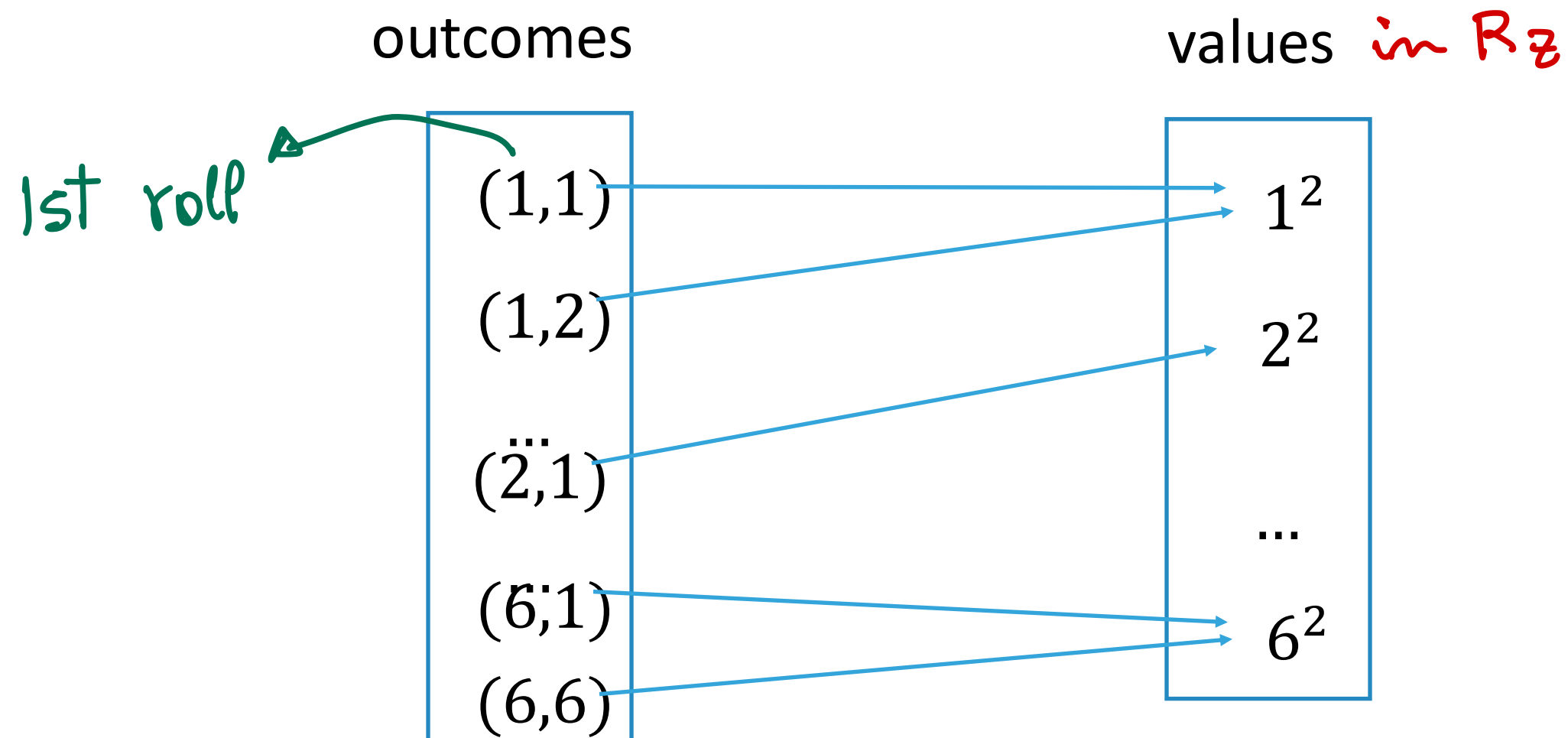
- ▶ **Experiment:** roll a fair die twice
- ▶ X = sum of the two rolls



$$X: \Omega \rightarrow \{2, 3, \dots, 12\}$$

RANDOM VARIABLES: EXAMPLES

- ▶ **Experiment:** roll a fair die twice
 - ▶ $Z = \text{first roll squared} \Rightarrow \text{1st-roll to the power of 2}$
- $R_Z = \{1^2, 2^2, 3^2, 4^2, 5^2, 6^2\}$



RANDOM VARIABLES: EXAMPLES

- ▶ **Experiment:** sample a value uniformly from $[-1, 1]$

$$X: [-1, 1] \rightarrow \mathbb{R}, X(r) = r \quad \forall r \in [-1, 1]$$

domain (inputs)

$[-1, 1]$

values (outputs)

$[-1, 1]$

RANDOM VARIABLES: EXAMPLES

- ▶ **Experiment:** sample a value uniformly from $[-1, 1]$

$$Y: [-1, 1] \rightarrow \mathbb{R}, Y(r) = r^2 \quad \forall r \in [-1, 1]$$

domain (inputs)

$[-1, 1]$

values (outputs)

$[0, 1]$

RANDOM VARIABLES: EXAMPLES

- **Experiment:** sample a value x uniformly from $[-1, 1]$

$$Z: [-1, 1] \rightarrow \mathbb{R}, Z(r) = \begin{cases} 1 & \text{if } r > 0 \text{ , w.p. } 0.5 \\ 0 & \text{if } r = 0 \text{ , w.p. } 0 \\ -1 & \text{if } r < 0 \text{ , w.p. } 0.5 \end{cases}$$

domain (inputs)

$[-1, 1]$

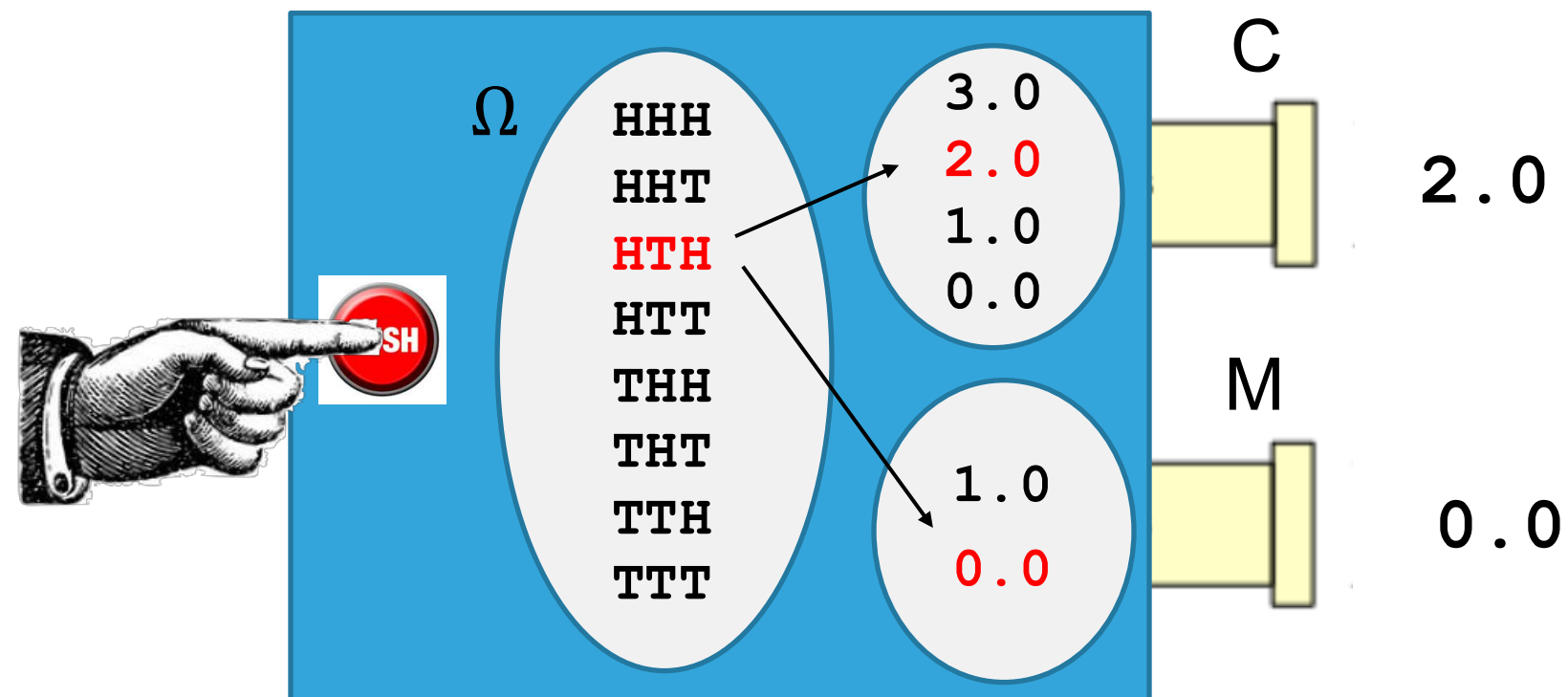
values (outputs)

$\{1, 0, -1\}$

discrete Range

JOINT RANDOM VARIABLES

- ▶ **Joint Random Variables** are combinations of 2 or more RVs, giving *multiple interpretations* of the same experiment.
- ▶ Example (from LLM):
- ▶ “Flip 3 coins and
(C) return the number of heads showing;
(M) return 1 if all heads or all tails, 0 otherwise”



TOP HAT QUESTION

Consider the following experiment:

“Flip 3 coins and

(T) return the number of tails showing;

(V) return 1 if there are more heads showing than tails, 0 if more tails than heads

► Suppose the random experiment chooses **HHT** from Ω then the result is:

A. $T = 1$ $V = 0$

B. $T = 2$ $V = 0$

C. $T = 2$ $V = 1$

D. None of the above

$$T(HHT) = 1$$

$$V(HHT) = 1$$

PROBABILITY MASS FUNCTION (PMF)

$$x_k \in \mathcal{R}_X$$

- ▶ For a given value x_k , the event

$$\{X = x_k\} = \{\omega \in \Omega : X(\omega) = x_k\}$$

- ▶ is the subset of outcomes in the sample space that make X equal to x_k .
- ▶ The **Probability Mass Function (PMF)** of X , denoted by P_X , assigns probabilities to these values:

$$P_X(x_k) = P(X = x_k) = \Pr(\{\omega \in \Omega : X(\omega) = x_k\})$$

- ▶ In other words, the PMF gives the probability that the random variable takes each of its possible values.

PROBABILITY MASS FUNCTION (PMF)


- ▶ A discrete random variable X takes values in a set $R_X = \{x_1, x_2, \dots\}$.
- ▶ The probability that X equals a particular value $x \in R_X$ is given by the **PMF**:

$$P_X(x) = P(X = x).$$

- ▶ The **PMF** assigns a probability to each possible value in the range of X .
- ▶ These probabilities must satisfy:

$$P_X(x) \geq 0 \text{ for all } x, \text{ and } \sum_x P_X(x) = 1.$$

PMF: EXAMPLE

- ▶ Let X be the number of heads obtained when a fair coin is tossed twice.
Find R_X and P_X 

PMF: EXAMPLE

- ▶ Let X be the number of heads obtained when a fair coin is tossed twice.

Find R_X and P_X *P.M.F.*

Range

- ▶ Sample space: $S = \{HH, HT, TH, TT\}$

$$R_X = \{0, 1, 2\}$$

- ▶ $P_X(k) = P(X = k)$ for $k = 0, 1, 2$

- ▶ $P_X(0) = P(X = 0) = P(TT) = \frac{1}{4}$

$$\frac{1}{4} + \frac{1}{2} + \frac{1}{4} = 1$$

- ▶ $P_X(1) = P(X = 1) = P(\{HT, TH\}) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$

- ▶ $P_X(2) = P(X = 2) = P(HH) = \frac{1}{4}$

PROBABILITY MASS FUNCTION (PMF)

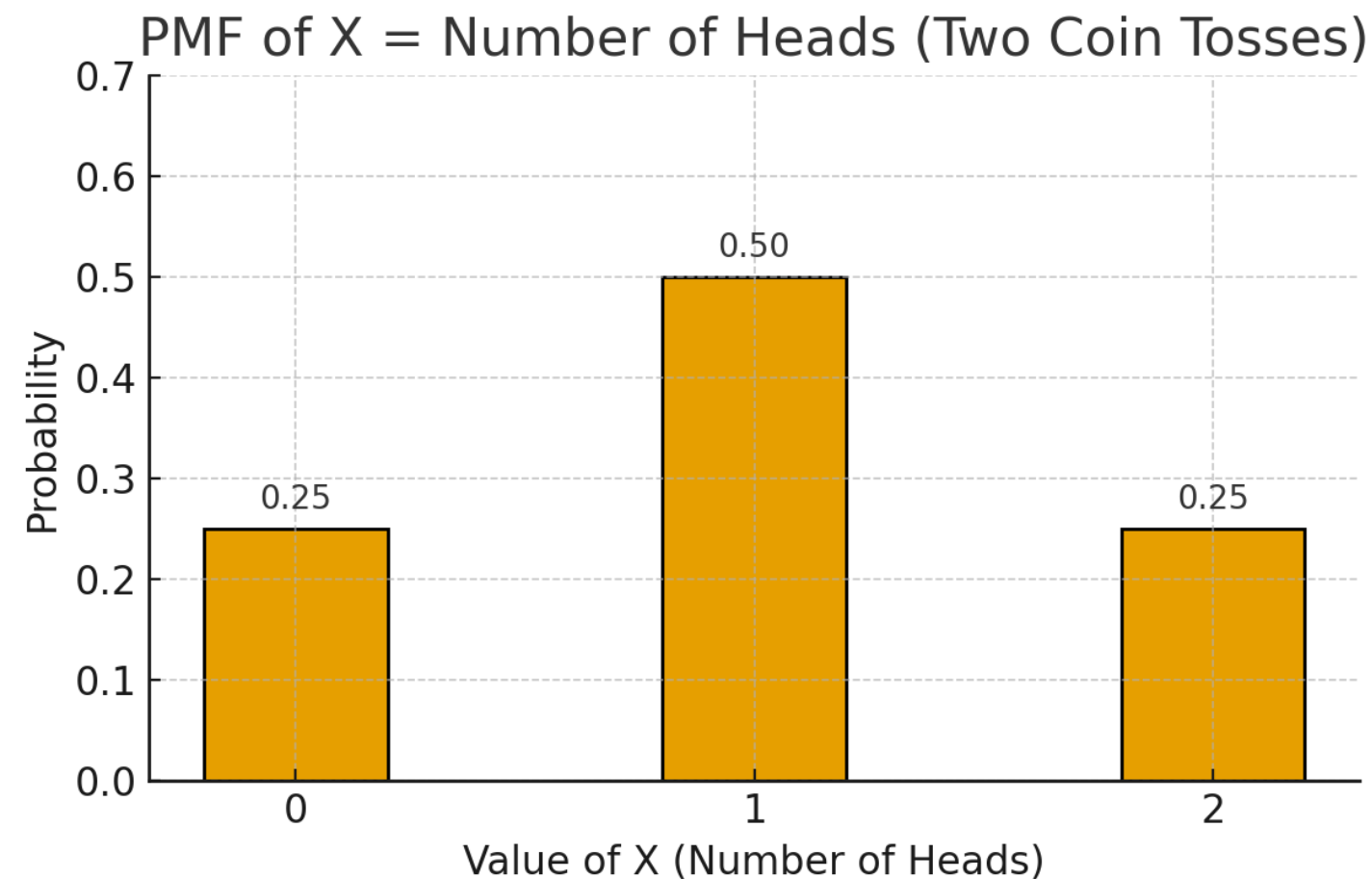
- ▶ Example: Toss two fair coins. Let X = number of heads.

PROBABILITY MASS FUNCTION (PMF)

- ▶ Example: Toss two fair coins. Let X = number of heads.

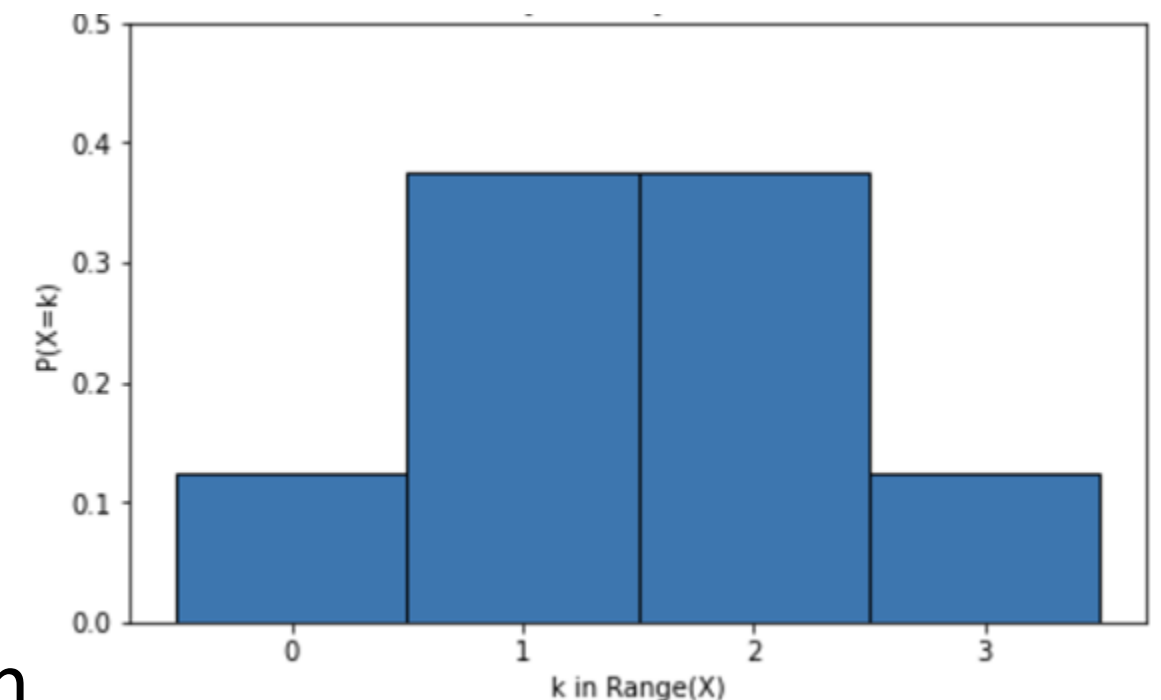
$$P_X(0) = \frac{1}{4}, P_X(1) = \frac{1}{2}, P_X(2) = \frac{1}{4}.$$

- ▶ The PMF shows how probability is distributed across the values of X .



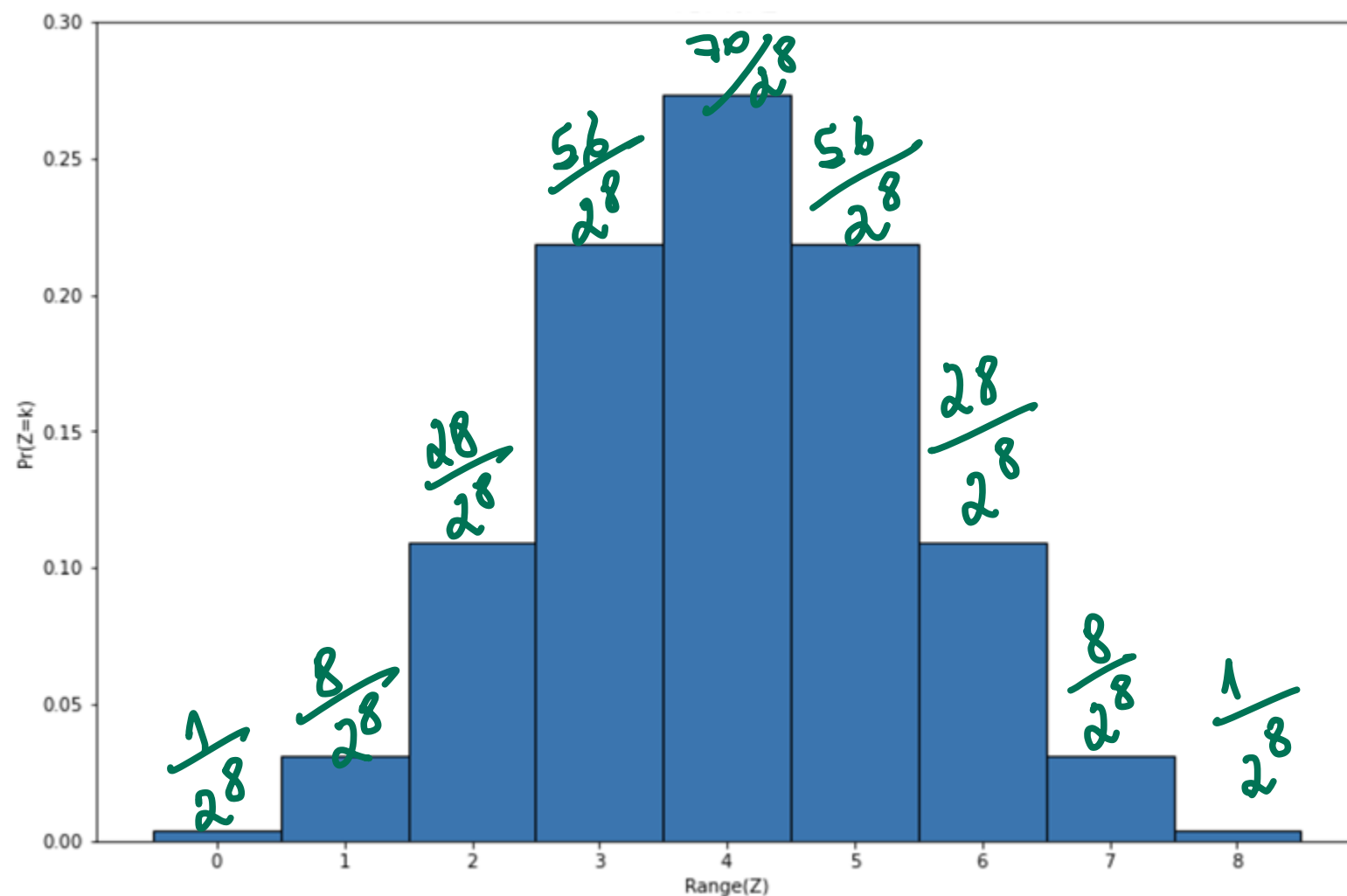
PROBABILITY MASS FUNCTION

- ▶ For discrete random variables, the **Probability Mass Function** (PMF) is also called the **probability distribution**. Thus, when asked to find the **probability distribution** of a discrete random variable X , we can do this by finding its PMF.
- ▶ $C =$ "Number of heads after three coin-flips"
- ▶ $\text{Range}(C) = \{0, 1, 2, 3\}$
- ▶ Probability Mass Function P_C
 $P_C(C=0) = P_C(C=3) = \frac{1}{8}$
 $P_C(C=1) = P_C(C=2) = \frac{3}{8}$
- ▶ A PMF contains all the information you need to know and is often illustrated as a histogram.



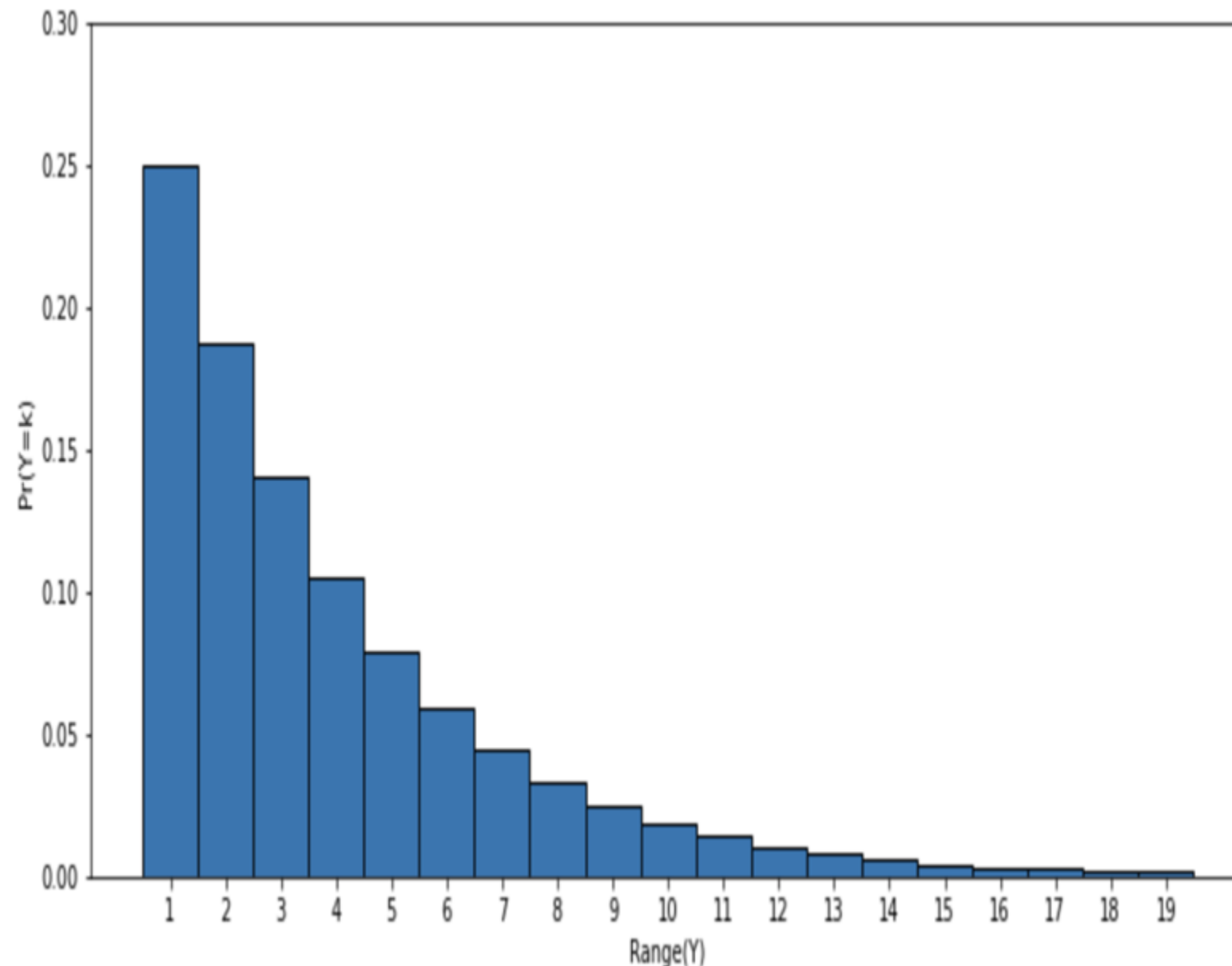
PROBABILITY MASS FUNCTION

- ▶ $Z = \text{"Toss 8 fair coins and count the number of heads"}$
- ▶ $\text{Range}(Z) = \{0, 1, 2, 3, 4, 5, 6, 7, 8\}$



PROBABILITY MASS FUNCTION

- Y = “Flip a 4-sided die and count the number of flips until it lands a 4.”



$$R_Y = \{1, 2, \dots\}$$

$$Pr(Y=1) = \frac{1}{4}$$

$$Pr(Y=2) = \frac{3}{4} \cdot \frac{1}{4} = \frac{3}{16}$$

$$Pr(Y=3) = \frac{3}{4} \cdot \frac{3}{4} \cdot \frac{1}{4}$$