

# CS 237: Probability in Computing

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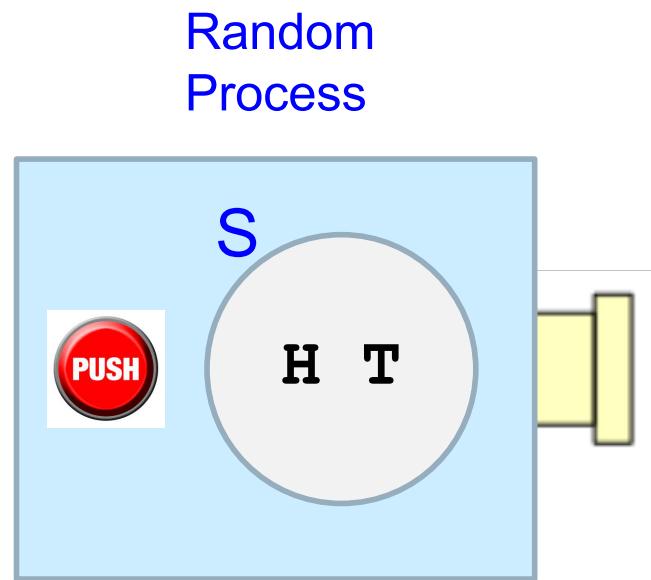
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## Lecture 7:

- Random Variables
- Discrete Random Variables
- Probability Distributions of Random Variables
- Expressions and functions of Random Variables
- Expected value of a Random Variable

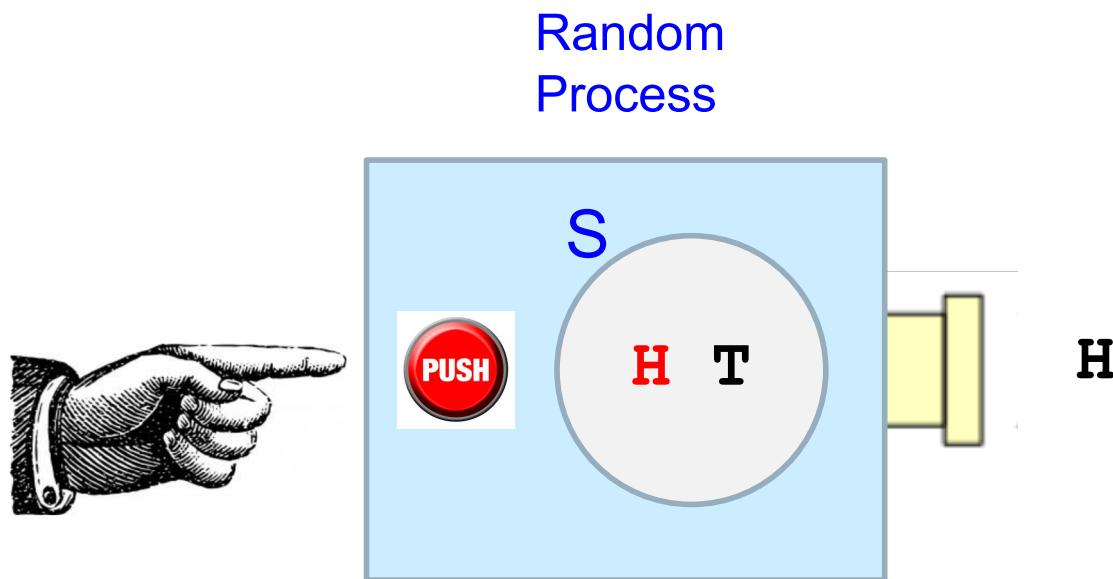
# Random Experiments and Random Variables

A Random Experiment is a process that produces uncertain outcomes from a well-defined sample space.



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# Random Variables

In order to formalize this notion, the notion of a Random Variable has been developed. A **Random Variable**  $X$  is a function from a sample space  $S$  into the reals:

$$X : S \rightarrow \mathcal{R}$$

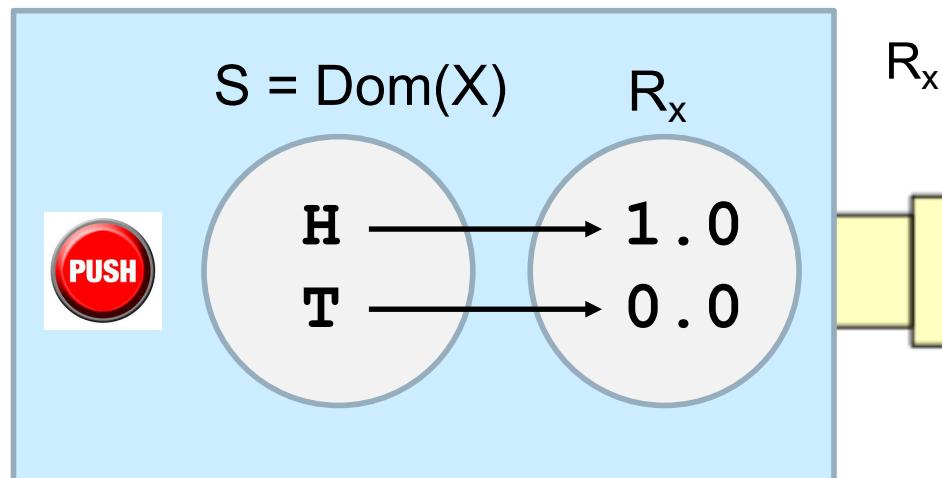
Now when an outcome is requested, the sample point is translated into a real number:

$$S = \text{Domain}(X)$$

$$X$$

$$R_x =_{\text{def}} \text{Range}(X)$$

$$R_x \subseteq \mathcal{R}$$



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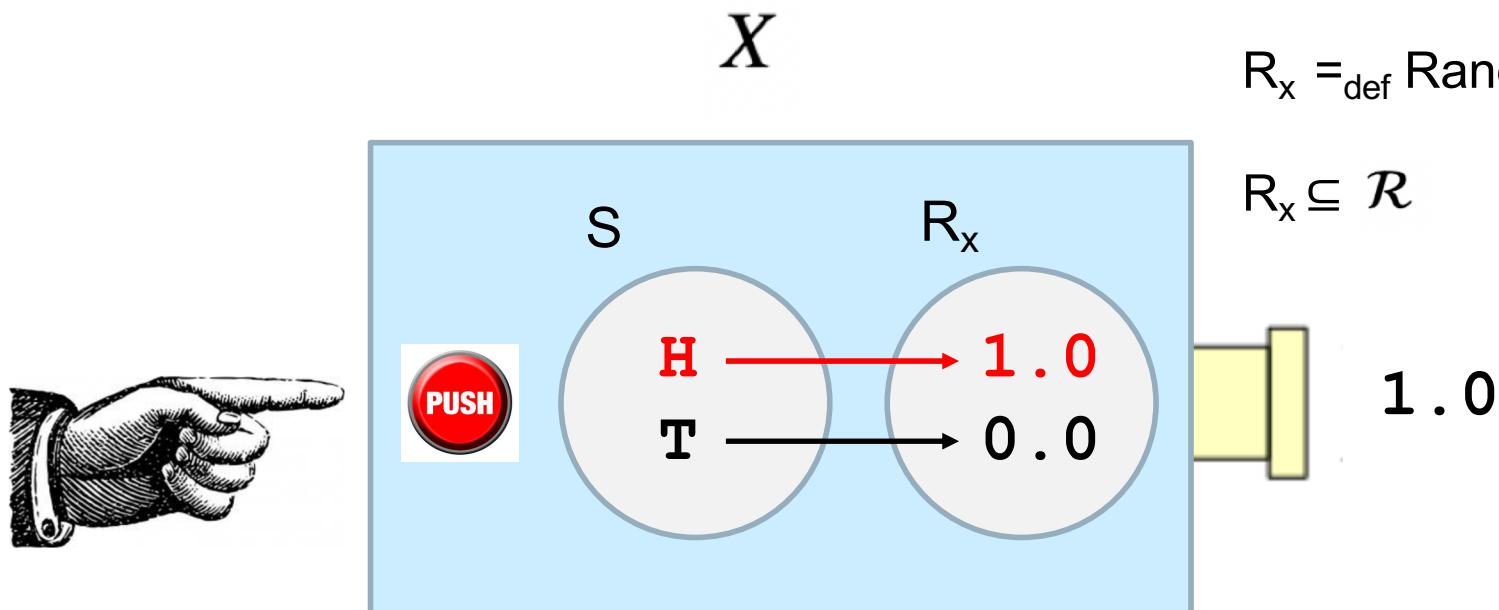
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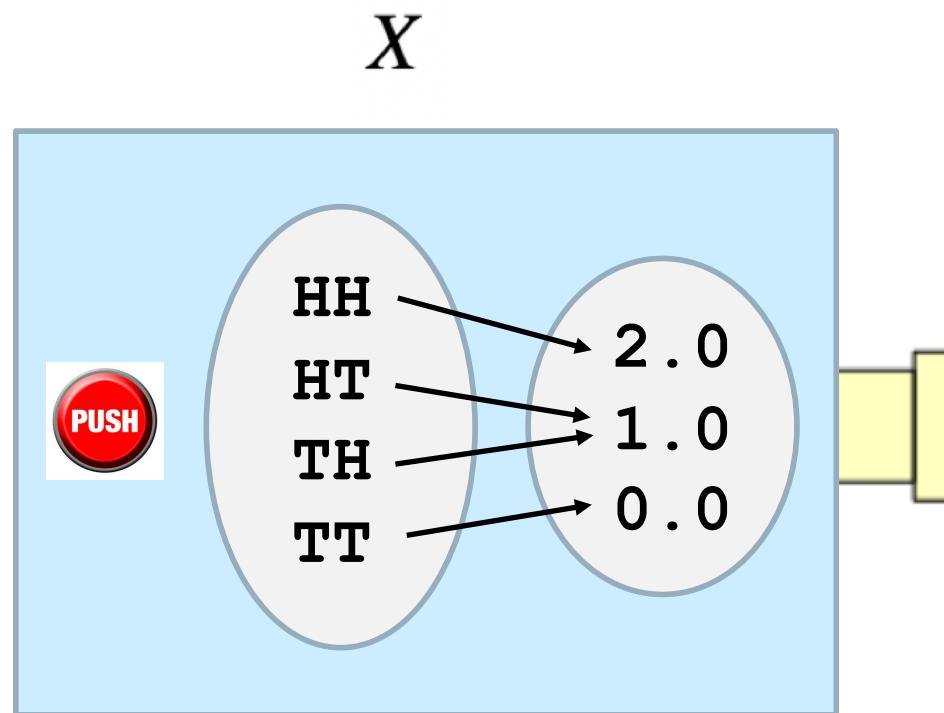
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# Random Variables

This may seem awkward, but it helps to explain the difference between random experiments whose literal outcomes are not numbers, but which are translated into numbers for clarity.

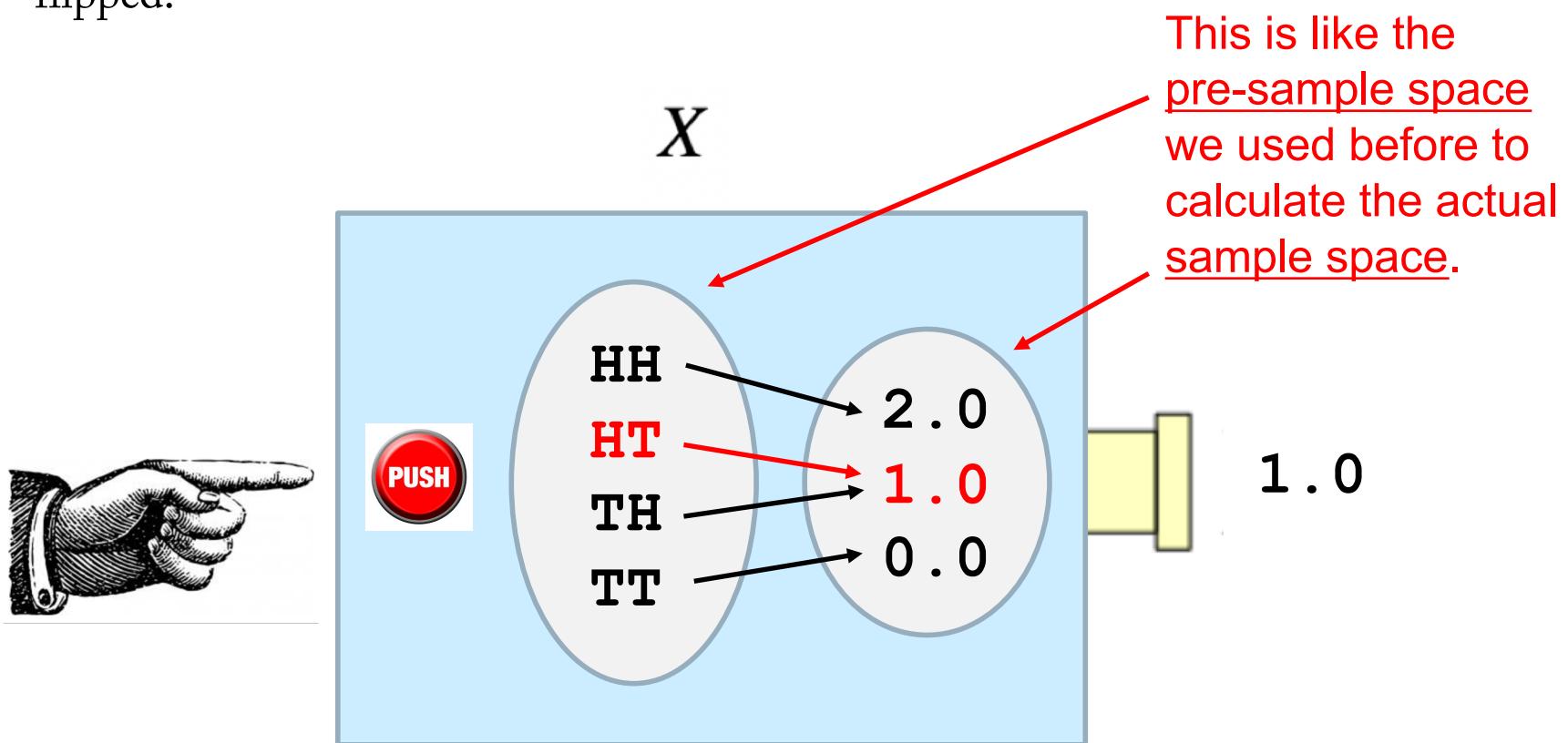
**Example:**  $X$  = “the number of heads which appear when two fair coins are flipped.”



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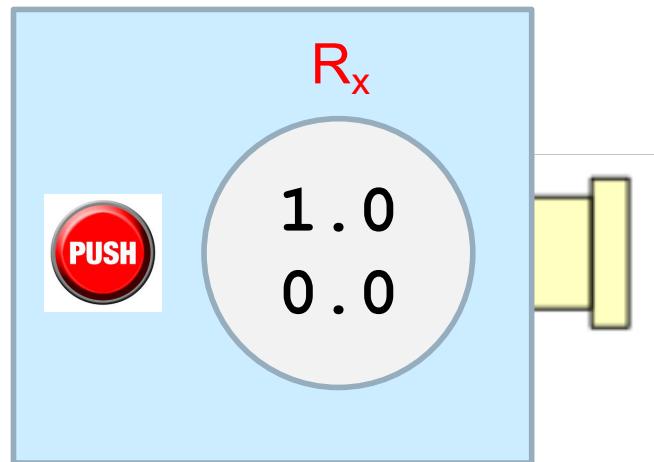
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# Random Variables

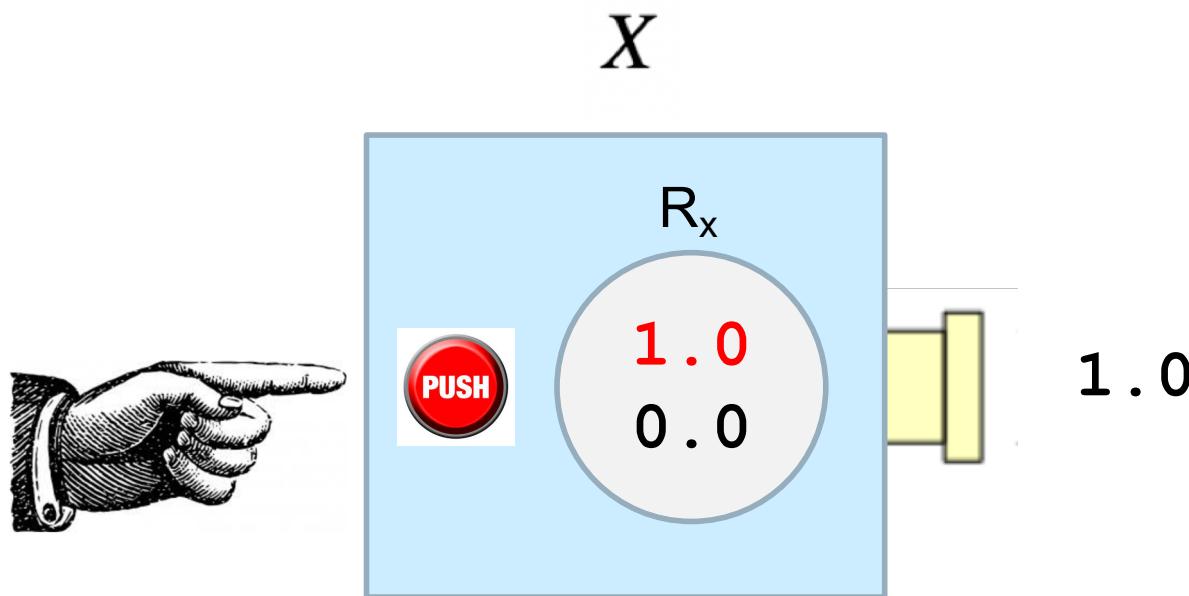
In general, in this class we will call the possible outputs  $R_x$ , since this is the symbol used in your textbook, although you could just think of it as the sample space from which the outputs are drawn.

$X$



# Random Variables

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# Discrete vs Continuous Random Variables

A random variable  $X$  is called **discrete** if  $R_x$  is **finite or countably infinite**:

Example of finite random variable:

$X$  = “the number of dots showing after rolling two dice”

$$R_x = \{ 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12 \}$$

Example of countably infinite random variable:

$Y$  = “the number of flips of a coin until a head appears”

$$R_y = \{ 1, 2, 3, \dots \}$$

A random variable is called **continuous** if  $R_x$  is **uncountable**. Example:

$Z$  = “the distance of a thrown dart from the center of a circular target of 1 meter radius”

$$R_z = [ 0.0 .. 1.0 ]$$

# Discrete vs Continuous Random Variables

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For several weeks we will only consider discrete random variables!

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$$R_z = [ 0.0 .. 1.0 ]$$

# Discrete Random Variables: Probability Distributions

The Probability Distribution of a discrete random variable  $X$  is a function  $f$  from the range of  $X$  into  $\mathcal{R}$ :

$$f_X : R_X \rightarrow [0..1]$$

such that (i)  $\forall a \in R_x \quad f_X(a) \geq 0$

(ii)  $\sum_{a \in R_x} f_X(a) = 1.0$

If there is no possibility of confusion we will write  $f$  instead of  $f_X$ .

# Discrete Random Variables: Probability Distributions

To specify a random variable precisely, you simply need to give the range  $R_X$  and the probability distribution  $f$ :

Examples:

$X$  = “The number of dots showing on a thrown die”

$$R_X = \{1, 2, 3, 4, 5, 6\}$$

$$f_X = \left\{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \right\}$$

For simplicity, we simply list the values in  $\text{Range}(f_X)$  corresponding to the listing of  $R_X$ .

$Y$  = “The number of tosses of a fair coin until a head appears”

$$R_Y = \{1, 2, 3, \dots\}$$

$$f_Y = \left\{ \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots \right\}$$

# Discrete Random Variables: Probability Distributions

How does this relate to our first definition of a probability space, events, probability function, etc., etc. ??

Probability Space

Sample Space

Random Variable X

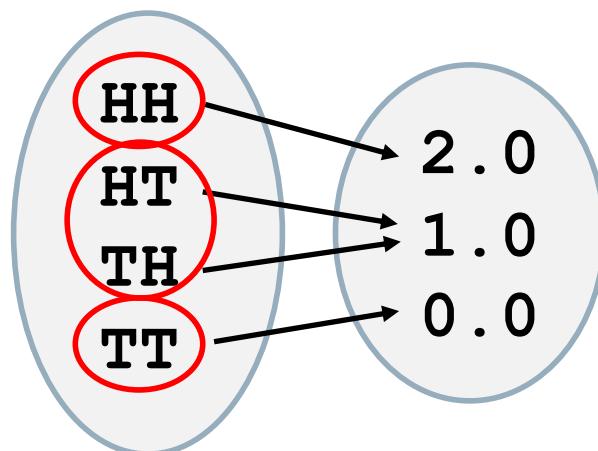
$R_X$

Event

Countable Union of Intervals on real line

Probability Function

Probability Distribution  $f_X$



# Discrete Random Variables: Probability Distributions

We will emphasize the distributions of random variables, using graphical representations (as in HW 1) to help our intuitions.

**Example:**

$$f_X : R_X \rightarrow [0..1]$$

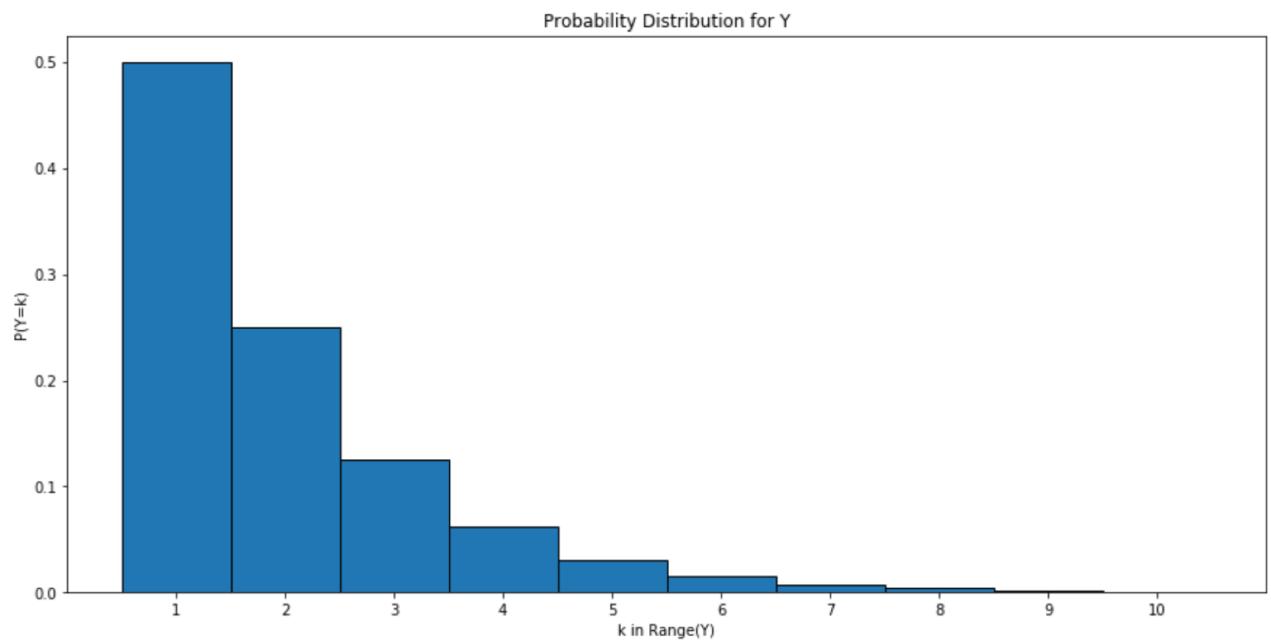
$Y$  = “The number of tosses of a fair coin until a head appears”

such that (i)  $\forall a \in R_x \quad f_X(a) \geq 0$

(ii)  $\sum_{a \in R_x} f_X(a) = 1.0$

$$R_Y = \{ 1, 2, 3, \dots \}$$

$$f_Y = \{ \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots \}$$



# Discrete Random Variables

Notation:

$$P(X = k) =_{\text{def}} f_X(k)$$

$$P(X \leq k) =_{\text{def}} \sum_{a \leq k} f_X(a)$$

$$P(k \leq X \leq m) =_{\text{def}} \sum_{k \leq a \leq m} f_X(a)$$

Example:

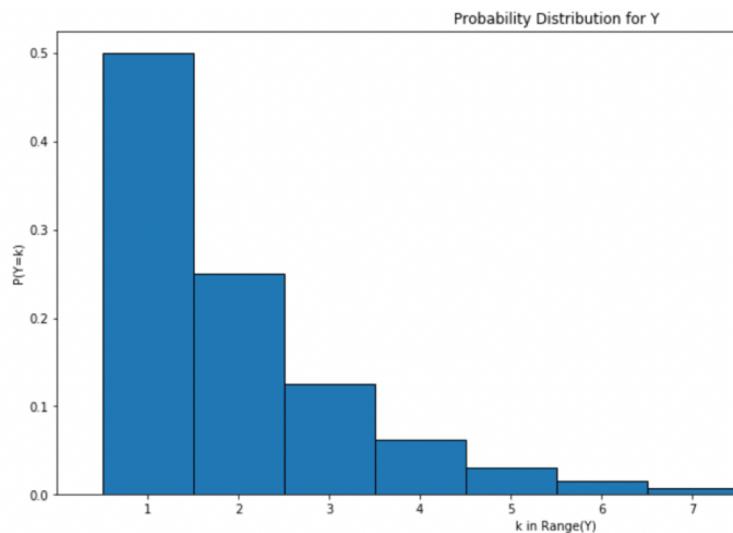
$$P(Y = 4) = \frac{1}{16}$$

$$R_Y = \{1, 2, 3, \dots\}$$

$$P(Y < 4) = \frac{7}{8}$$

$$f_Y = \left\{ \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots \right\}$$

$$P(2 \leq Y \leq 4) = \frac{7}{16}$$



# Functions of Discrete Random Variables

New random variables can be created by functions or expressions involving old random variables. But you have to be careful!

Example:  $X$  = “the number of dots on a thrown die”

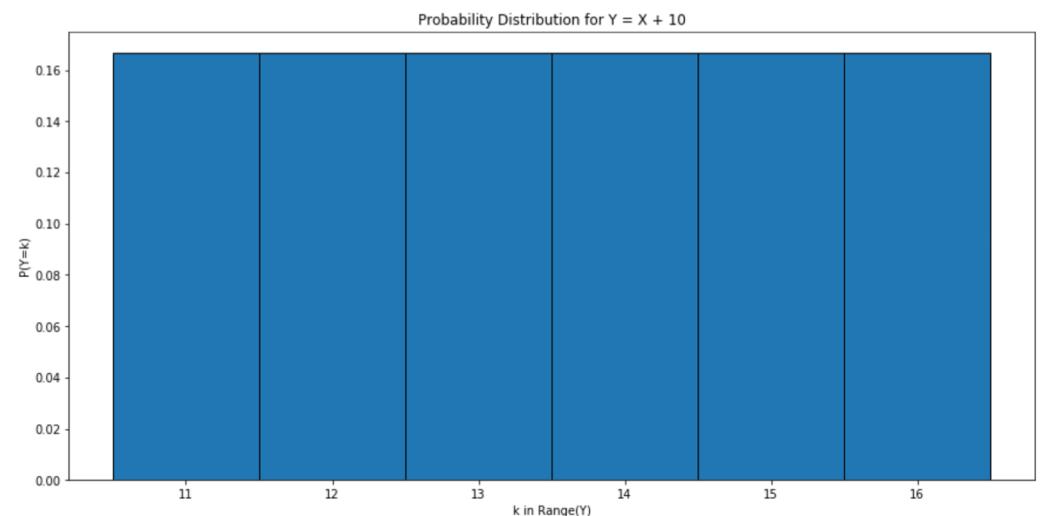
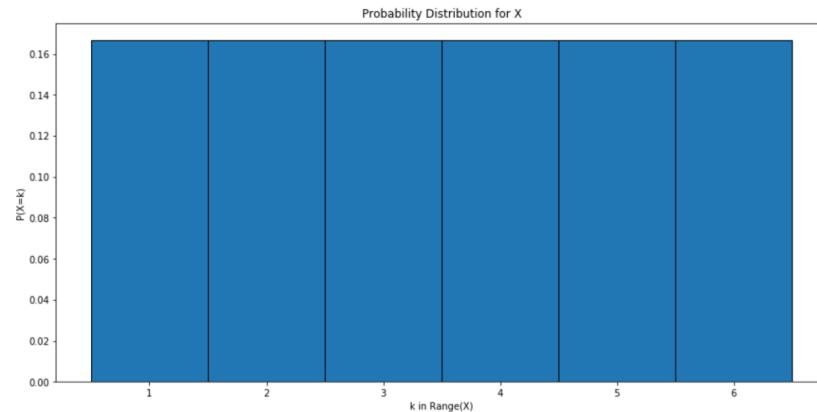
$$R_X = \{1, 2, 3, 4, 5, 6\}$$

$$f_X = \left\{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \right\}$$

Let  $Y = X + 10$

$$R_Y = \{11, 12, 13, 14, 15, 16\}$$

$$f_Y = \left\{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \right\}$$

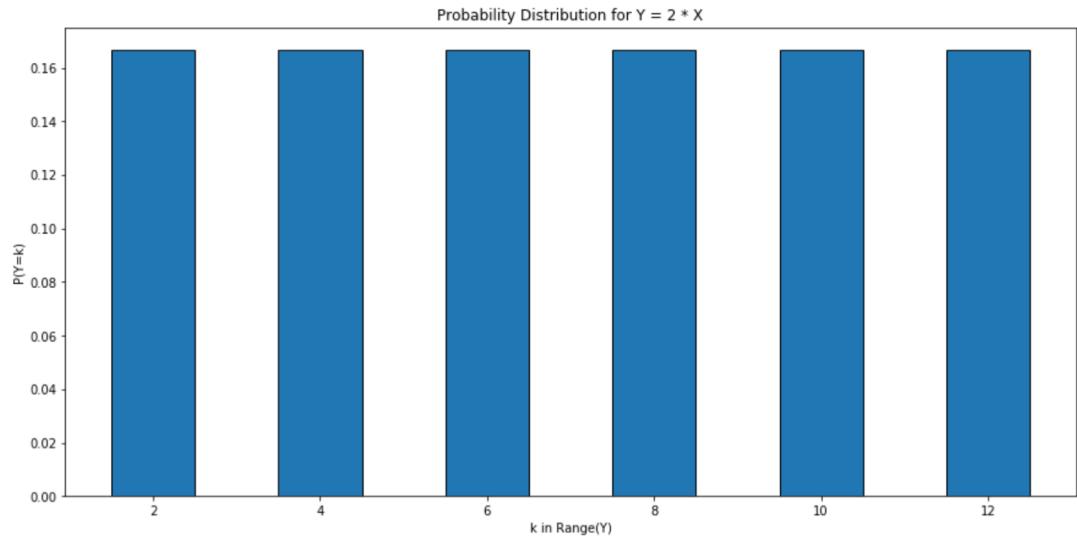


# Functions of Discrete Random Variables

Let  $Y = 2 * X$

$$R_Y = \{ 2, 4, 6, 8, 10, 12 \}$$

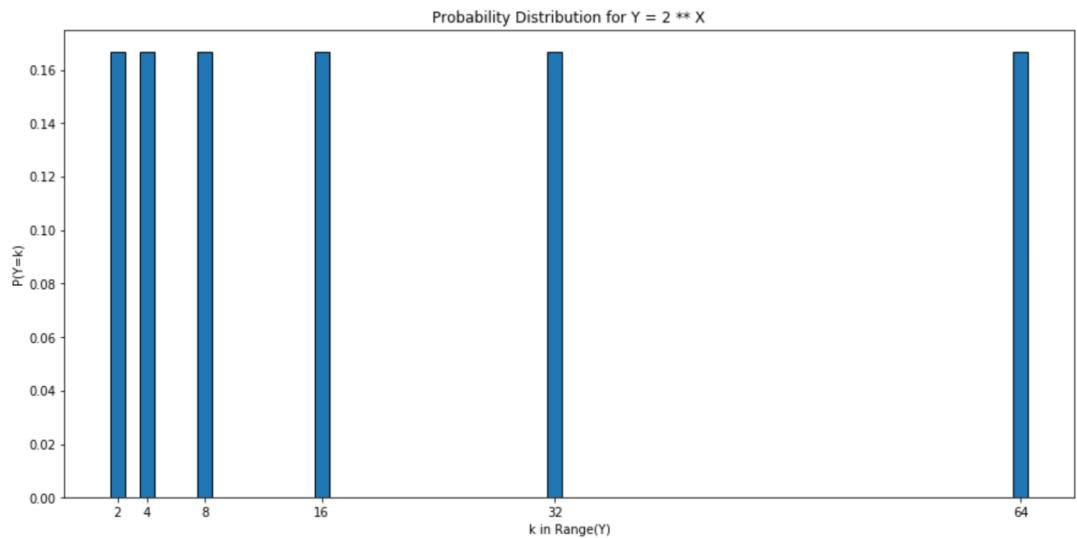
$$f_Y = \{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \}$$



Let  $Y = 2^X$

$$R_Y = \{ 2, 4, 8, 16, 32, 64 \}$$

$$f_Y = \{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \}$$



# Functions of Discrete Random Variables

Why did I say you have to be careful? Two main reasons...

One, the function of a random variable may combine outcomes...

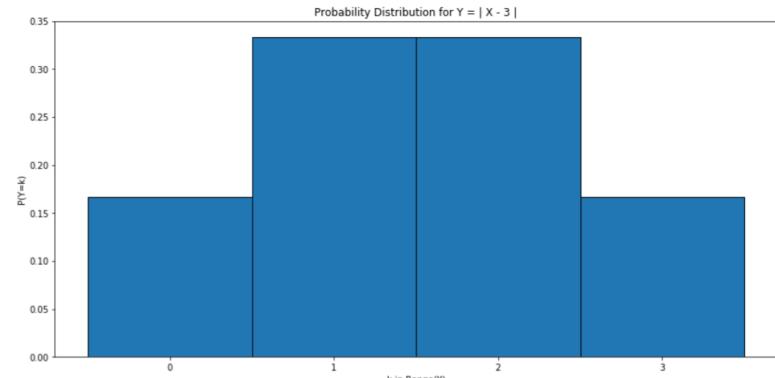
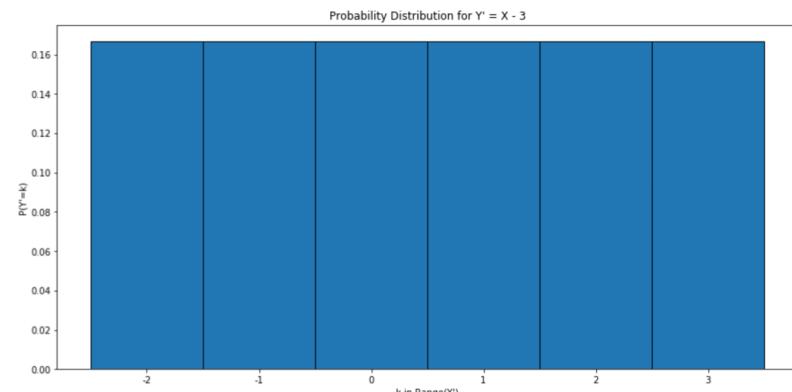
Example: Let  $Y' = X - 3$  and let  $Y = |X - 3|$

$$R_{Y'} = \{ -2, -1, 0, 1, 2, 3 \}$$

$$f_{Y'} = \{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \}$$

$$R_Y = \{ 0, 1, 2, 3 \}$$

$$f_Y = \{ \frac{1}{6}, \frac{2}{6}, \frac{2}{6}, \frac{1}{6} \}$$



$$R_X = \{1, 2, 3, 4, 5, 6\}$$

# Functions of Discrete Random Variables

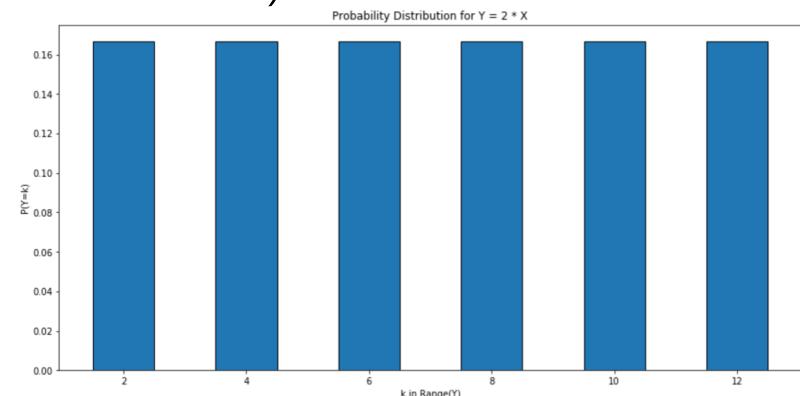
$$f_X = \left\{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \right\}$$

Two, you have to be careful when a random variable is used more than once, since each occurrence refers to a potentially different random outcome!

Let  $Y = 2 * X$  (twice the dots showing on a thrown die)

$$R_Y = \{2, 4, 6, 8, 10, 12\}$$

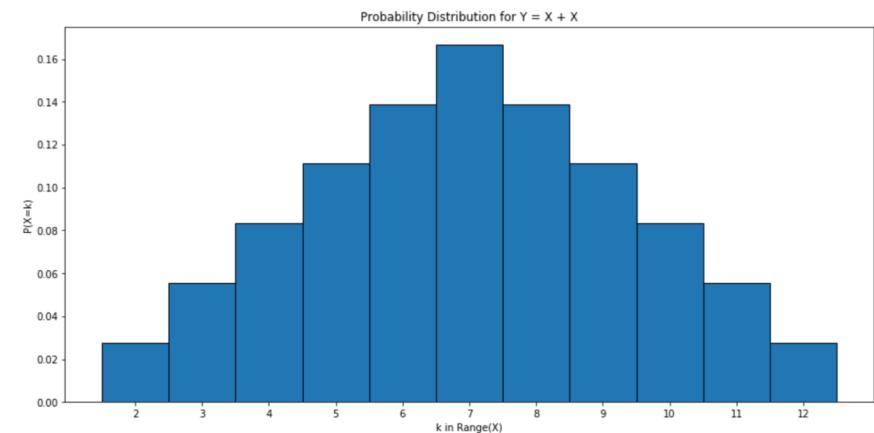
$$f_Y = \left\{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \right\}$$



Let  $Y = X + X$  (sum of the dots showing on two thrown dice)

$$R_Y = \{2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$$

$$f_Y = \left\{ \frac{1}{36}, \frac{2}{36}, \frac{3}{36}, \frac{4}{36}, \frac{5}{36}, \frac{6}{36}, \frac{5}{36}, \frac{4}{36}, \frac{3}{36}, \frac{2}{36}, \frac{1}{36} \right\}$$



# Discrete Random Variables: Expected Value

A fundamental way of characterizing a collection of real numbers is the **average** or **mean** value of the collection:

**Example:** The mean/average of  $\{ 2, 4, 6, 9 \} = 21/4 = 5.7$

The corresponding notion for a random variable  $X$  is the **Expected Value**:

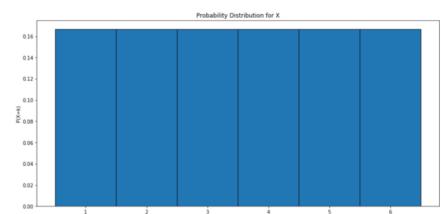
$$E(X) =_{\text{def}} \sum_{a \in R_x} a * f(a)$$

**Example:**  $X$  = “the number of dots showing on a single thrown die”

$$R_X = \{1, 2, 3, 4, 5, 6\}$$

$$E(X) = \sum_{k \in R_X} \frac{k}{6} = \frac{1 + 2 + 3 + 4 + 5 + 6}{6} = \frac{21}{6} = 3.5$$

$$f_X = \{ \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6} \}$$



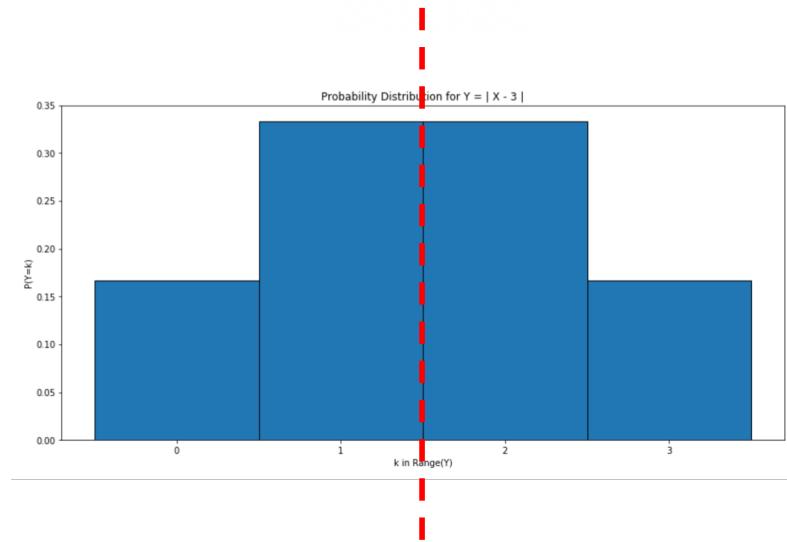
# Discrete Random Variables: Expected Value

Example:  $Y = |X - 3|$

$$R_Y = \{ 0, 1, 2, 3 \}$$

$$f_Y = \{ \frac{1}{6}, \frac{2}{6}, \frac{2}{6}, \frac{1}{6} \}$$

$E(Y)$



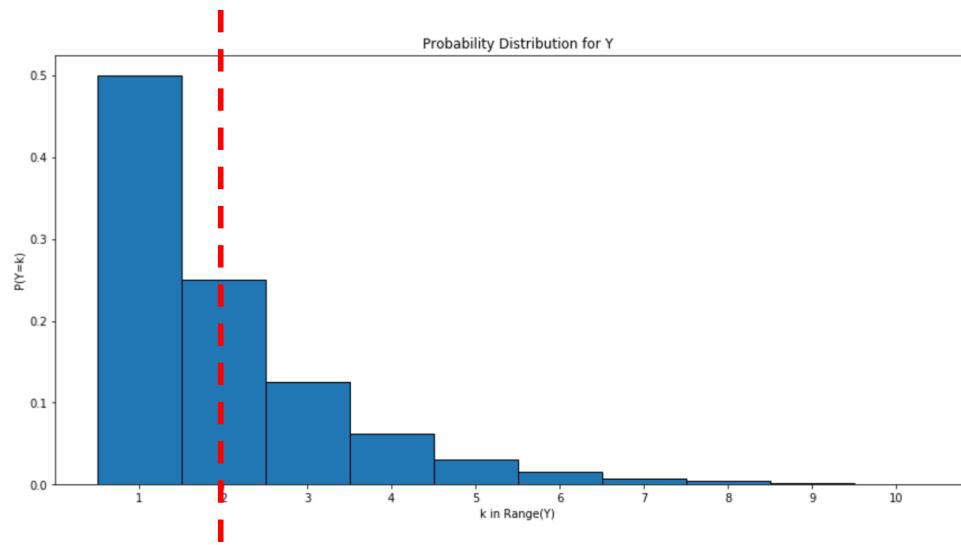
$$E(Y) = \sum_{k \in R_Y} k * f(k) = \frac{0}{6} + \frac{1}{3} + \frac{2}{3} = \frac{3}{6} = 1.5$$

# Discrete Random Variables: Expected Value

Example:  $Y$  = “tosses of a fair coin until a heads appears”

$$R_Y = \{ 1, 2, 3, \dots \}$$

$$f_Y = \{ \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots \}$$



$$E(Y) = \sum_{k \in R_Y} k * f(k) = \frac{1}{2} + \frac{2}{4} + \frac{3}{8} + \frac{4}{16} \dots = 2.0$$

# Expected Value and Fair Games

The **expected value** of a random variable is a **one-number summary of its behavior**. It describes what you can expect as the limiting behavior over many trials.

A good example of what this means occurs with games in which you win or lose money on each round or trial. Such a game can be modeled by a random variable:  $X$  = “the amount you win (+) or lose (-).” A game is **fair** if  $E(X) = 0$ .

**Example:** The rules of “Chuck-a-luck” are as follows. The player makes a bet on any number 1 through 6 and then three dice are thrown. If 1, 2, or 3 dice show the same number as the player’s choice, then he or she wins back the original bet plus 1, 2, or 3 times the original bet.

- So if you bet \$1 on **4** and the dice roll **2, 4, and 6**, you get back  $1 + 1 = 2$  for a net win of \$1.
- If you bet \$1 on **2** and the dice roll **2, 6, and 2**, you get back  $1 + 2 = 3$  for a net win of \$2.
- If you bet \$1 on **5** and no 5's show and you lose \$1.



# Expected Value and Fair Games



To analyze Chuck-a-Luck, suppose the player always bets \$1 on each round (or trial) of the game. Let

$X$  = “net win or loss for one round.”

Then  $R_X = \{ -1, 1, 2, 3 \}$

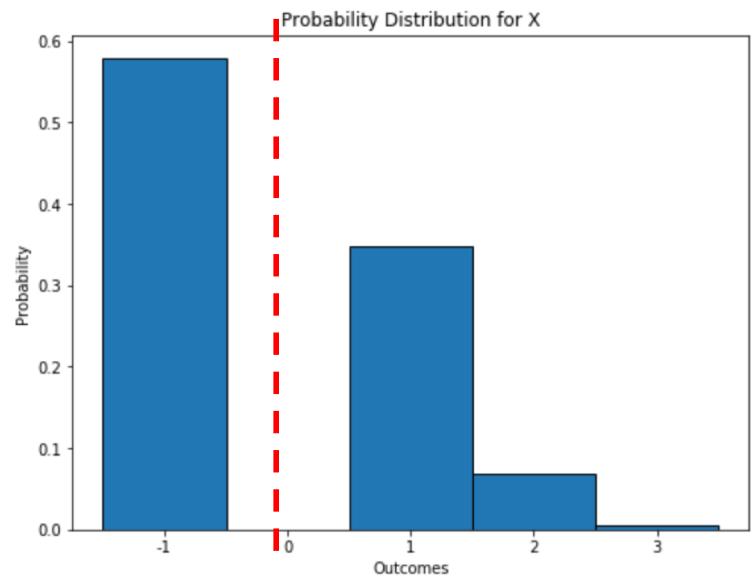
$$P(X = -1) = \binom{3}{0} * \frac{5}{6} * \frac{5}{6} * \frac{5}{6} = 0.5787$$

$$P(X = 1) = \binom{3}{1} * \frac{1}{6} * \frac{5}{6} * \frac{5}{6} = 0.3472$$

$$P(X = 2) = \binom{3}{2} * \frac{1}{6} * \frac{1}{6} * \frac{5}{6} = 0.0694$$

$$P(X = 3) = \binom{3}{3} * \frac{1}{6} * \frac{1}{6} * \frac{1}{6} = 0.0047$$

$$f_X = \{ 0.5787, 0.3472, 0.0694, 0.0047 \}$$



On average you lose about 8 cents per round, so Chuck-a-luck is **not fair**:

$$E(X) = \sum_{k \in R_X} k * f_X(k) = -0.5787 + 0.3472 + 2 * 0.0694 + 3 * 0.0047 = -0.0786$$