

Recall: random variables

- A **random variable** X on a sample space Ω is a function $X: \Omega \rightarrow \mathbb{R}$ that assigns to each sample point $\omega \in \Omega$ a real number $X(\omega)$.
- For each random variable, we should understand:
 - The set of values it can take.
 - The probabilities with which it takes on these values.
- The **distribution** of a discrete random variable X is the collection of pairs $\{(a, \Pr[X = a])\}$.

You roll two dice. Let X be the random variable that represents the sum of the numbers you roll.

What is the probability of the event $X=6$?

- A. $1/36$
- B. $1/9$
- C. $5/36$
- D. $1/6$
- E. None of the above.

You roll two dice. Let X be the random variable that represents the sum of the numbers you roll.

How many different values can X take on?

- A. 6
- B. 11
- C. 12
- D. 36
- E. None of the above.

You roll two dice. Let X be the random variable that represents the sum of the numbers you roll.

What is the distribution of X ?

- A. Uniform distribution on the set of possible values.
- B. It satisfies $\Pr[X = 2] < \Pr[X = 3] < \dots < \Pr[X = 12]$.
- C. It satisfies $\Pr[X = 2] > \Pr[X = 3] > \dots > \Pr[X = 12]$.
- D. It satisfies $\Pr[X = 2] < \Pr[X = 3] < \dots < \Pr[X = 7]$ and $\Pr[X = 7] > \Pr[X = 8] > \dots > \Pr[X = 12]$.
- E. None of the above is true.

Independent RVs: definition

- Random variables X and Y are **independent** if

$$\begin{aligned}\Pr[(X = x) \cap (Y = y)] \\ = \Pr[X = x] \cdot \Pr[Y = y]\end{aligned}$$

for all values x and y .

- Random variables X_1, X_2, \dots, X_n are **mutually independent** if for **all subsets** of $I \subseteq [n]$ and all values x_i , where $i \in I$,

$$\begin{aligned}\Pr[\cap_{i \in I} (X_i = x_i)] \\ = \prod_{i \in I} \Pr[X_i = x_i].\end{aligned}$$

You roll one die. Let X be the random variable that represents the result.

What value does X take, on average?

- A. $1/6$
- B. 3
- C. 3.5
- D. 6
- E. None of the above.

Random variables: expectation

- The **expectation** of a discrete random variable X over a sample space Ω is

$$E[X] = \sum_{\omega \in \Omega} X(\omega) \cdot \Pr[\omega].$$

- We can group together outcomes ω for which $X(\omega) = a$:

$$E[X] = \sum_a a \cdot \Pr[X = a],$$

where the sum is over all possible values a taken by X .

- The second equality is more useful for calculations.

Example: random hats

- **Example:** permutations
 - n students exchange their hats, so that everybody gets a random hat
 - R.V. X : the number of students that got their own hats.
 - E.g., if students 1,2,3 got hats 2,1,3 then $X=1$.

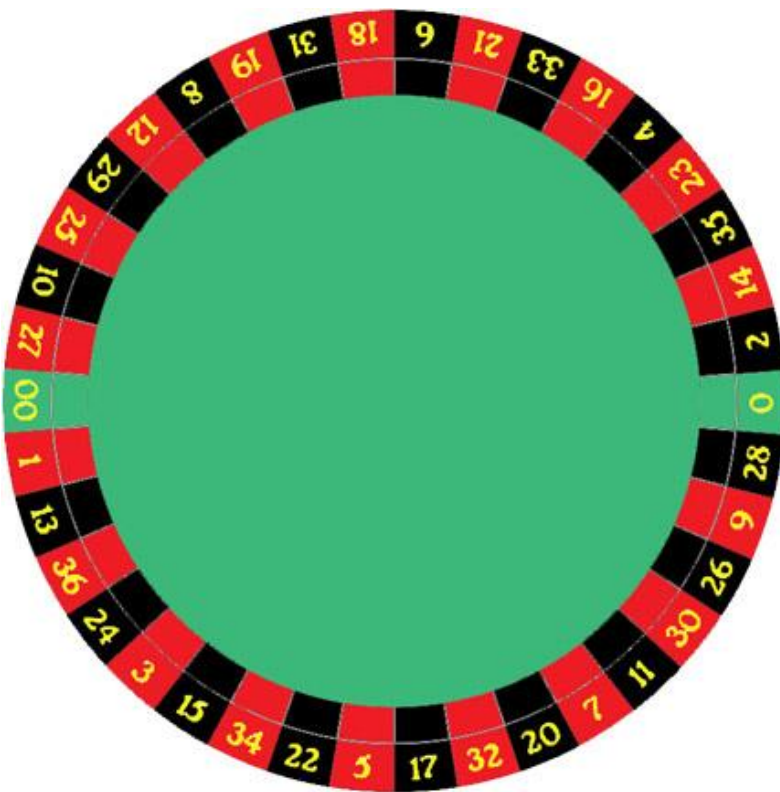
- Distribution of X :

$$\Pr[X = 0] = \frac{1}{3}, \Pr[X = 1] = \frac{1}{2}, \Pr[X = 3] = \frac{1}{6}.$$

- What's the expectation of X ?

Example: roulette

- 38 slots: 18 black, 18 red, 2 green.
- If we bet \$1 on red, we get \$2 back if red comes up. What's the expected value of our winnings?



Linearity of expectation

- **Theorem.** For any two random variables X and Y on the same probability space,

$$E[X + Y] = E[X] + E[Y].$$

Also, for all $c \in \mathbb{R}$,

$$E[cX] = c \cdot E[X].$$

Indicator random variables

- An **indicator random variable** takes on two values: 0 and 1.
- **Lemma.** For an indicator random variable X ,
$$E[X] = \Pr[X = 1].$$

You have a coin with bias $3/4$ (the bias is the probability of HEADS). Let X be the number of HEADS in 1000 tosses of your coin.

You represent X as the sum: $X = X_1 + X_2 + \cdots + X_{1000}$.

What is X_1 ?

- A. $3/4$.
- B. The number of HEADS.
- C. The probability of HEADS in toss 1.
- D. The number of heads in toss 1.
- E. None of the above.

You have a coin with bias $3/4$ (the bias is the probability of HEADS). Let X be the number of HEADS in 1000 tosses of your coin.

What is the expectation of X ?

- A. $3/4$.
- B. $4/3$.
- C. 500.
- D. 750.
- E. None of the above.