Expectation summarizes "average" value:, a function mapping well-behaved random variables to numbers

Layman's talk : what is expectation

Expectation summarizes "average" value:
a function mapping well-behaved random variables
to numbers

▶ Distributions assign (probability values) numbers to random events associated with random variables.

Many values associated with many events...

- ► Sometimes, one would prefer to know what happens "on average". E.g., if heads for 1 bottoms for 0, we know from our daily experience, the average would be 0.5.
- ► **Expectation** formalizes the above idea and assigns a number (if exists) for one random variable.

## Example:

 $X \sim \text{Bern}(p)$  with PMF:

$$P(X = 0) = 1 - p$$
;  $P(X = 1) = p$ .

$$E(X) = 0 \times P(X = 0) + 1 \times P(X = 1) = p.$$

## Example:



 $X \sim \text{Bern}(p)$ , then Y = 2 + 2X is also an r.v. with PMF:

$$P(Y = 2) = 1 - p$$
;  $P(Y = 4) = p$ .

$$E(X) = 0 \times P(X = 0) + 1 \times P(X = 1) = p.$$

$$E(Y) = 2 \times P(Y = 2) + 4 \times P(Y = 4) = 2 + 2p.$$

For a discrete r.v. X:

$$E(X) = \sum_{x \in \text{supp}(X)} x P(X = x).$$

## For a discrete r.v. X:

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- ▶ A weighted average (with weights being the associated probability values, these weights sum up to 1) of possible values taken by X.
- ► For E(X) to be well-defined, we need the condition that this summation  $\sum_{x \in \text{supp}(X)} x P(X = x)$  is well-defined.

Expectation essentially maps a subset of the collection of all random variables to numbers. It is a function!

$$E: \{X : well-behaved r.v.'s\} \mapsto C$$

$$E(X + Y) = E(X) + E(Y)$$
 and  $E(aX) = aE(X)$  for real-valued r.v.'s  $X, Y$  and  $a \in \mathbb{R}$ .

 Expectation goes far beyond "average" by working with different transformation of random variables.

E.g., it can recover probability values via the indicator random variable  $I_{\{X=x\}}$ :

$$P(X=x)=EI_{\{X=x\}}.$$