

Unit 4 Discrete Random Variables

Probability Mass Functions and Expectations

A random variable is loosely speaking, a numerical quantity whose value is determined by the outcome of a probabilistic experiment e.g. the weight of a randomly selected person

Discrete: takes values in finite or countable set

Random variable examples:

Bernoulli

Uniform

Binomial

Geometric

Expected value of a random variable aka Expectation (mean)
weighted average of the values of the random variable
weighted on their probabilities

Random variables mathematically: A function from the sample space omega to the real numbers

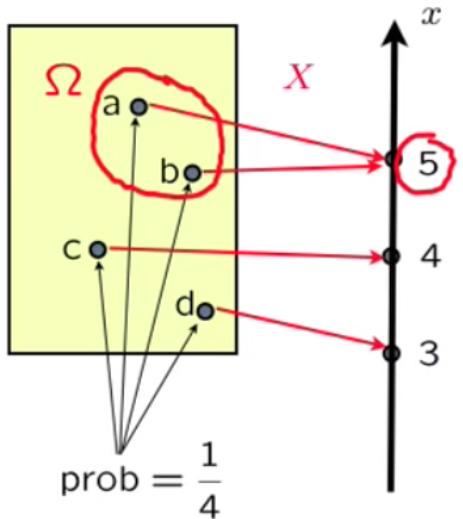
Notation: random variable X numerical value x

A function of one or more random variables is also a random variable

Probability Mass Function (PMF) aka probability law or probability distribution of X

Probability space of getting different values

$$P(5) = 1/2 \text{ a and b}$$



Notation:

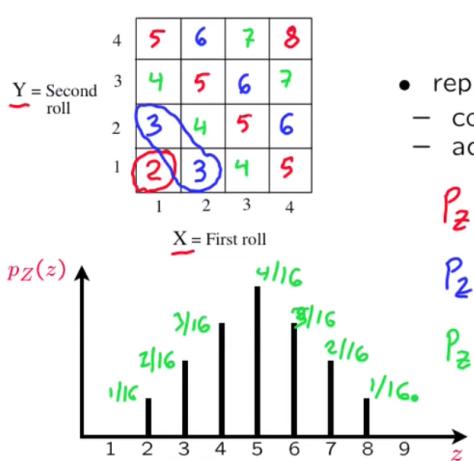
$$p_X(x) = \mathbf{P}(X = x) = \mathbf{P}(\{\omega \in \Omega \text{ s.t. } X(\omega) = x\})$$

- Properties:** $p_X(x) \geq 0$

$$\sum_x p_X(x) = 1$$

PMF calculation

- Two rolls of a tetrahedral die
- Let every possible outcome have probability 1/16



$$Z = X + Y \quad \text{Find } p_Z(z) \quad \text{for all } z$$

- repeat for all z :
 - collect all possible outcomes for which Z is equal to z
 - add their probabilities

$$P_Z(2) = P(Z=2) = 1/16$$

$$P_Z(3) = P(Z=3) = 2/16$$

$$P_Z(4) = P(Z=4) = 3/16$$

⋮

Exercise: Random variables versus numbers

1/2 points (graded)

Let X be a random variable that takes integer values, with PMF $p_X(x)$. Let Y be another integer-valued random variable and let y be a number.

a) Is $p_X(y)$ a random variable or a number?

Random variable ✗ Answer: Number

b) Is $p_X(Y)$ a random variable or a number?

Random variable ✓ Answer: Random variable

Solution:

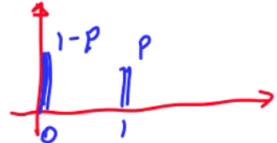
a) Recall that $p_X(\cdot)$ is a function that maps real numbers to real numbers. So, when we give it a numerical argument, y , we obtain a number.

b) In this case, we are dealing with a function, the function being $p_X(\cdot)$, of a random variable Y . And a function of a random variable is a random variable. Intuitively, the "random" value of $p_X(Y)$ is generated as follows: we observe the realized value y of the random variable Y , and then look up the numerical value $p_X(y)$.

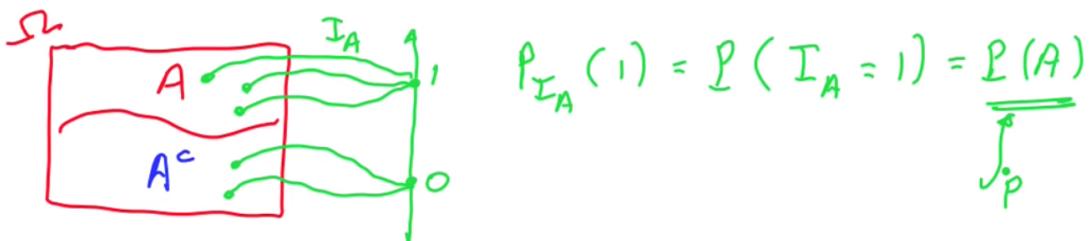
The simplest random variable: Bernoulli with parameter $p \in [0, 1]$

$$X = \begin{cases} 1, & \text{w.p. } p \\ 0, & \text{w.p. } 1 - p \end{cases}$$

$$\begin{aligned} p_x(0) &= 1 - p \\ p_x(1) &= p \end{aligned}$$



- Models a trial that results in success/failure, Heads/Tails, etc.
- Indicator r.v. of an event A : $I_A = 1$ iff A occurs



Exercise: Indicator variables

2/2 points (graded)

Let A and B be two events (subsets of the same sample space Ω), with nonempty intersection. Let I_A and I_B be the associated indicator random variables.

For each of the two cases below, select one statement that is true.

a) $I_A + I_B$:

is not the indicator random variable of any event ✓

Answer: is not the indicator random variable of any event

b) $I_A \cdot I_B$:

is the indicator variable of the event $A \cap B$ ✓

Answer: is the indicator variable of the event $A \cap B$

(*Bug warning:* In some browsers, the mathematical content in each choice in the dropdown menu may appear duplicated, e.g. $A \cup B$ may show up twice as $A \cup BA \cup B$.)

Solution:

a) If the outcome of the experiment lies in the intersection of the events A and B , then $I_A + I_B$ takes the value of 2. But indicator random variables can take only the values 0 or 1. Therefore, $I_A + I_B$ is not an indicator random variable.

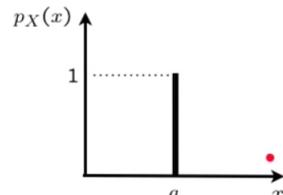
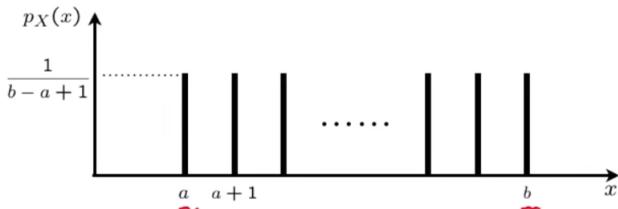
b) Note that $I_A \cdot I_B$ can take only the values 0 or 1. It is equal to 1 if and only if $I_A = 1$ (i.e., event A occurs) and $I_B = 1$ (i.e., event B occurs). Thus, $I_A \cdot I_B$ takes the value of 1 if and only if both A and B occur, and so it is the indicator random variable of the event $A \cap B$.

Discrete uniform random variable; parameters a, b

- **Parameters:** integers a, b ; $a \leq b$
- **Experiment:** Pick one of $a, a+1, \dots, b$ at random; all equally likely
- **Sample space:** $\{a, a+1, \dots, b\}$ $b-a+1$ possible values
- **Random variable X :** $X(\omega) = \omega$ $11:52:26 \quad \{0, 1, \dots, 59\}$
- **Model of:** complete ignorance

Special case: $a = b$

constant/deterministic r.v.



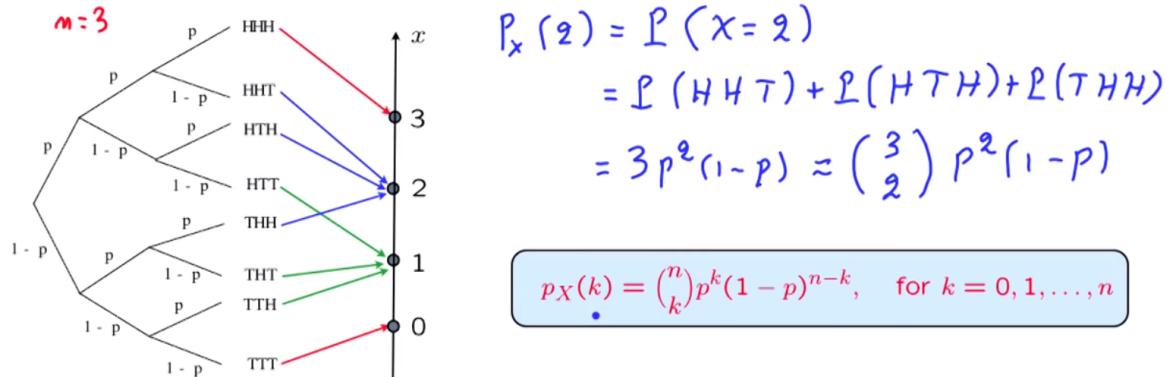
e.g. imagine looking at a clock at time 11:52:26 but only looking at the seconds
the probability of getting a number 0 to 59 is equally likely

Special case where $a = b$ is still a random variable in the mathematical sense, it

just so happens that it can only be one value in this case

Binomial random variable: parameters: positive integer n ; $p \in [0, 1]$

- **Experiment:** n independent tosses of a coin with $P(\text{Heads}) = p$
- **Sample space:** Set of sequences of H and T, of length n
- **Random variable X :** number of Heads observed
- **Model of:** number of successes in a given number of independent trials



Same formula as earlier but with slightly different notation

Exercise: The binomial PMF

2/2 points (graded)

You roll a fair six-sided die (all 6 of the possible results of a die roll are equally likely) 5 times, independently. Let X be the number of times that the roll results in 2 or 3. Find the numerical values of the following.

a) $p_X(2.5) =$ 0 ✓ Answer: 0

b) $p_X(1) =$ 0.3292 ✓ Answer: 0.32922

Solution:

a) A value of 2.5 is not possible for X since the number of rolls must be an integer, and therefore $p_X(2.5) = 0$.

b) For each die roll, there is a probability $2/6 = 1/3$ of obtaining a 2 or a 3. Hence, the random variable X is binomial with parameters $n = 5$ and $p = 1/3$, so that $p_X(1) = \binom{5}{1} \cdot (1/3) \cdot (2/3)^4 \approx 0.32922$.

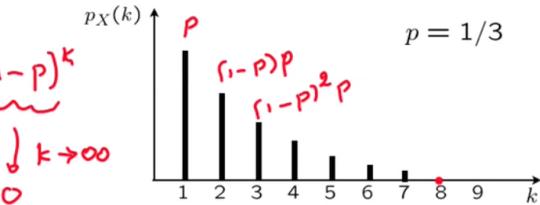
Geometric random variable; parameter p : $0 < p \leq 1$

- **Experiment:** infinitely many independent tosses of a coin; $P(\text{Heads}) = p$
- **Sample space:** Set of infinite sequences of H and T $\overbrace{\text{TTTTHHT...}}^{\text{X=5}}$
- **Random variable X :** number of tosses until the first Heads $X = 5$

- **Model of:** waiting times; number of trials until a success

$$p_X(k) = P(X=k) = P(\underbrace{\text{T...T}}_{k-1} \text{H}) = (1-p)^{k-1} p \quad k=1, 2, 3, \dots$$

$$\begin{aligned} P(\text{no Heads ever}) &\leq P(\underbrace{\text{T...T}}_k) = (1-p)^k \\ \underbrace{\text{TTT...}}_{\text{"X=\infty"}} &= 0 \end{aligned}$$



$P(\text{Tails forever})$ becomes 0 as k goes to infinity

Exercise: Geometric random variables

0/2 points (graded)

Let X be a geometric random variable with parameter p . Find the probability that $X \geq 10$. Express your answer in terms of p using [standard notation](#) (click on the "STANDARD NOTATION" button below.)

$$P(X \geq 10) = \frac{p}{1 - ((1-p)^9)}$$

✖ Answer: $(1-p)^9$

$$\frac{p}{1 - ((1-p)^9)}$$

[STANDARD NOTATION](#)

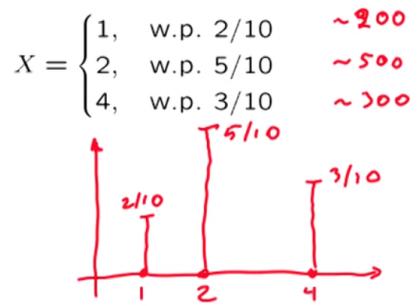
Solution:

We can calculate the desired probability by adding the probabilities of the events $\{X = 10\}$, $\{X = 11\}$, $\{X = 12\}$, etc., and using the formula for the sum of a geometric series. However, we can get the answer in an easier way, using the interpretation of geometric random variables as the number of trials until the first success. The event $\{X \geq 10\}$ is the event that the first 9 trials resulted in failure, and therefore its probability is $(1 - p)^9$.

Expectation/mean of a random variable

- **Motivation:** Play a game 1000 times.
Random gain at each play described by:
- "Average" gain:

$$\begin{aligned} & \frac{1 \cdot 200 + 2 \cdot 500 + 4 \cdot 300}{1000} \\ &= 1 \cdot \frac{2}{10} + 2 \cdot \frac{5}{10} + 4 \cdot \frac{3}{10} \end{aligned}$$



- **Definition:** $E[X] = \sum_x x p_X(x)$

- **Interpretation:** Average in large number of independent repetitions of the experiment

- **Caution:** If we have an infinite sum, it needs to be well-defined.
We assume $\sum_x |x| p_X(x) < \infty$

Expectation of a Bernoulli r.v.

Video position. Press space to toggle playback

$$X = \begin{cases} 1, & \text{w.p. } p \\ 0, & \text{w.p. } 1 - p \end{cases} \quad E[X] = 1 \cdot p + 0 \cdot (1 - p) = p$$

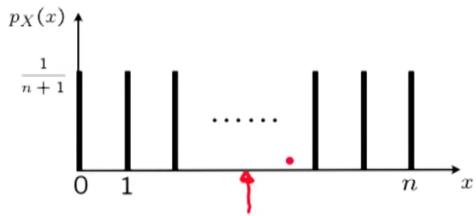
If X is the indicator of an event A , $X = I_A$:

$$X = 1 \text{ iff } A \text{ occurs} \quad p = P(A)$$

$$E[I_A] = P(A)$$

Expectation of a uniform r.v.

- Uniform on $0, 1, \dots, n$



• Definition: $E[X] = \sum_x x p_X(x)$

$$\begin{aligned} E[X] &= 0 \cdot \frac{1}{n+1} + 1 \cdot \frac{1}{n+1} + \dots + n \cdot \frac{1}{n+1} \\ &= \frac{1}{n+1} (0 + 1 + \dots + n) = \frac{1}{n+1} \cdot \frac{n(n+1)}{2} = \frac{n}{2} \end{aligned}$$

Elementary properties of expectations

- If $X \geq 0$, then $E[X] \geq 0$
for all ω : $X(\omega) \geq 0$

• Definition: $E[X] = \sum_x x p_X(x)$

$\geq 0 \quad \geq 0 \quad \geq 0$

- If $a \leq X \leq b$, then $a \leq E[X] \leq b$

for all ω : $a \leq X(\omega) \leq b$

$$\begin{aligned} E[X] &= \sum_x x p_X(x) \geq \sum_x a p_X(x) \\ &= a \sum_x p_X(x) = a \cdot 1 = a \end{aligned}$$

- If c is a constant, $E[c] = c$

$\xrightarrow{\text{c}}$

$$E[c] = c \cdot p(c) = c$$

Exercise: Random variables with bounded range

3/3 points (graded)

Suppose a random variable X can take any value in the interval $[-1, 2]$ and a random variable Y can take any value in the interval $[-2, 3]$.

a) The random variable $X - Y$ can take any value in an interval $[a, b]$. Find the values of a and b :

$$a = \boxed{-4} \quad \checkmark$$

$$b = \boxed{4} \quad \checkmark$$

b) Can the expected value of $X + Y$ be equal to 6?

$$\text{No} \quad \checkmark$$

The expected value rule, for calculating $E[g(X)]$

- Let X be a r.v. and let $Y = g(X)$
- Averaging over y : $E[Y] = \sum_y y p_Y(y)$
 $3 \cdot (0.1+0.2) + 4 \cdot (0.3+0.4)$
- Averaging over x : $3 \cdot 0.1 + 3 \cdot 0.2 + 4 \cdot 0.3 + 4 \cdot 0.5$

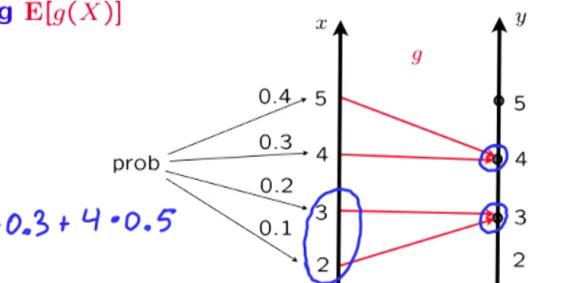
$$E[Y] = E[g(X)] = \sum_x g(x) p_X(x)$$

Proof:

$$\sum_y \sum_{x: g(x)=y} g(x) p_x(x)$$

$$= \sum_y \sum_{x: g(x)=y} p_x(x) = \sum_y \sum_{x: g(x)=y} p_x(x)$$

$$= \sum_y y p_Y(y) = E[Y]$$



- $E[X^2] = \sum_x x^2 p_x(x)$
- Caution: In general, $E[g(X)] \neq g(E[X])$

$$E[X^2] \neq (E[X])^2$$

Linearity of expectation: $E[aX + b] = aE[X] + b$

$X = \text{Salary}$ $E[X] = \text{average salary}$

$Y = \text{new salary} = 2X + 100$ $E[Y] = E[2X + 100] = 2E[X] + 100$

- Intuitive

- **Derivation**, based on the expected value rule:

$$E[Y] = \sum_x g(x) p_x(x)$$

$$= \sum_x (ax + b) p_x(x) = a \sum_x x p_x(x) + b \sum_x p_x(x)$$

$$E[g(x)] = g(E[x]) = aE[x] + b$$

exceptional g

$$g(x) = ax + b$$

$$Y = g(x)$$

The blue equality is only true for linear functions

Exercise: Linearity of expectations

3/3 points (graded)

The random variable X is known to satisfy $E[X] = 2$ and $E[X^2] = 7$. Find the expected value of $8 - X$ and of $(X - 3)(X + 3)$.

a) $E[8 - X] =$ ✓

b) $E[(X - 3)(X + 3)] =$ ✓