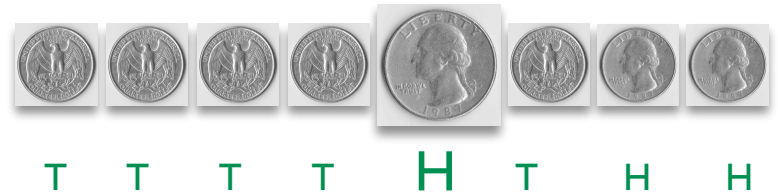

random variables



let $X_1 =$ index of

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

$$\Pr(X=k) = \Pr(\{\omega \mid X(\omega)=k\})$$

random variables

A *random variable* is a numeric function of the outcome of an experiment, not the outcome itself.

Ex.

Let H be the *number* of Heads when 20 coins are tossed

Let T be the *total* of 2 dice rolls

Let X be the *number* of coin tosses needed to see 1st head

Note: even if the underlying experiment has “equally likely outcomes,” the associated random variable *may not*

Outcome	$X = \#H$	$P(X)$
TT	0	$P(X=0) = 1/4$
TH	1	} $P(X=1) = 1/2$
HT	1	
HH	2	$P(X=2) = 1/4$

$\Omega =$ collecting all subsets of size 3.

$$\Pr(\{5, 9, 13\}) = \frac{1}{\binom{20}{3}}$$

numbered balls

20 balls numbered 1, 2, ..., 20

Draw a subset of size 3 at random.

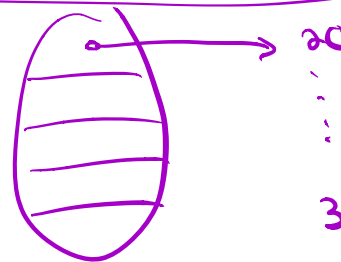
$$|\Omega| = \binom{20}{3}$$
$$\Pr(\omega) =$$

Let $X =$ the maximum of the numbers on those 3 balls

What is $P(X \geq 17)$

$$\Pr(X=20) = \frac{\binom{19}{2}}{\binom{20}{3}}$$

$$\Pr(X=18) = \frac{\binom{17}{2}}{\binom{20}{3}}$$



20 balls numbered 1, 2, ..., 20

Draw a subset of size 3 at random.

Let X = the maximum of the numbers on those 3 balls

What is $P(X \geq 17)$

$$P(X = 20) = \binom{19}{2} / \binom{20}{3} = \frac{3}{20} = 0.150$$

$$P(X = 19) = \binom{18}{2} / \binom{20}{3} = \frac{18 \cdot 17 / 2!}{20 \cdot 19 \cdot 18 / 3!} \approx 0.134$$

\vdots

$$\sum_{i=17}^{20} P(X = i) \approx 0.508$$

$$P(X \geq 17) = 1 - P(X < 17) = 1 - \binom{16}{3} / \binom{20}{3} \approx 0.508$$

Flip a (biased) coin (probability p of Heads) repeatedly until 1st head observed

What is the sample space?

$$\Omega = \{ \underset{x(\omega)}{H}, \underset{1}{TH}, \underset{2}{TTH}, \dots \}$$

How many flips? Let X be that number.

$$P(X=1) = p$$

$$P(X=2) = (1-p) \cdot p$$

$$P(X=3) = (1-p)^2 p.$$

...

$$P(X=i) = (1-p)^{i-1} p.$$

Flip a (biased) coin repeatedly until 1st head observed

How many flips? Let X be that number.

$$P(X=1) = P(H) = p$$

$$P(X=2) = P(TH) = (1-p)p$$

$$P(X=3) = P(TTH) = (1-p)^2p$$

...

$$P(X=i) = P(T^{i-1}H) = (1-p)^{i-1}p$$

$$\sum_{i \geq 0} x^i = \frac{1}{1-x},$$

when $|x| < 1$

memorize me!

A function whose domain is the sample space.

probability mass functions

A *discrete* random variable is one taking on a *countable* number of possible values.

Ex:

$X = \text{sum of 3 dice, } 3 \leq X \leq 18, X \in \mathbb{N}$

$Y = \text{position of 1}^{\text{st}} \text{ head in seq of coin flips, } 1 \leq Y, Y \in \mathbb{N}$

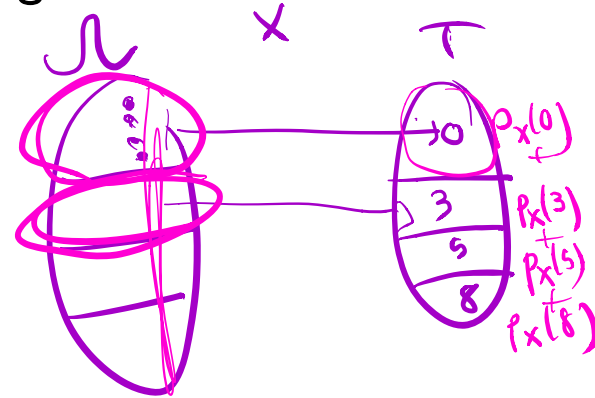
$Z = \text{largest prime factor of } (1+Y), Z \in \{2, 3, 5, 7, 11, \dots\}$

probability mass functions

A *discrete* random variable is one taking on a *countable* number of possible values.

Ex:

$$X = \text{sum of 3 dice}, \quad 3 \leq X \leq 18, \quad X \in \mathcal{N}$$



Definition: If X is a discrete random variable taking on values from a countable set $T \subseteq \mathcal{R}$, then

$$p_X(a) = \begin{cases} P(X = a) & \text{for } a \in T \\ 0 & \text{otherwise} \end{cases}$$

is called the *probability mass function*. Note: $\sum_{a \in T} p_X(a) = 1$

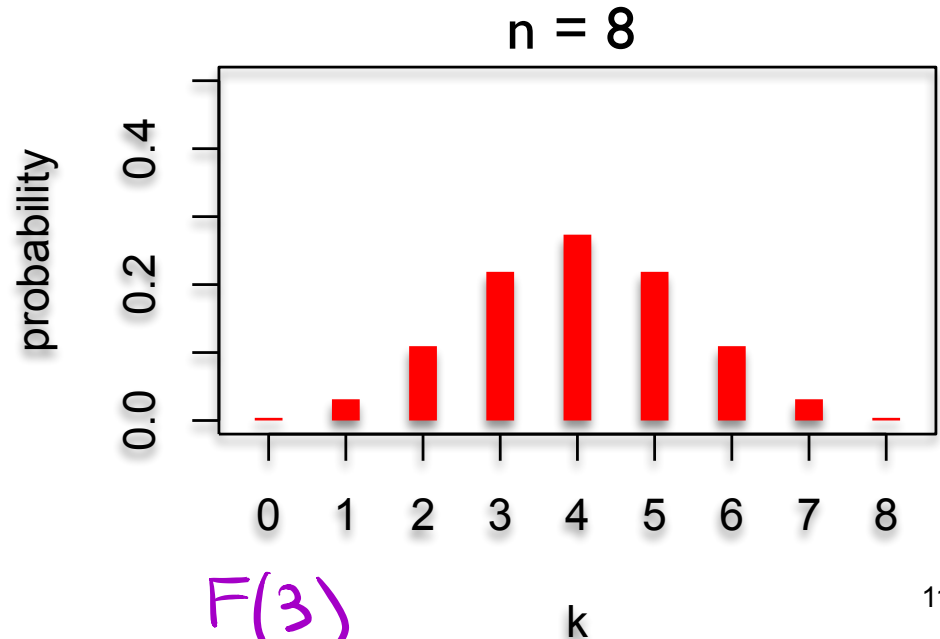
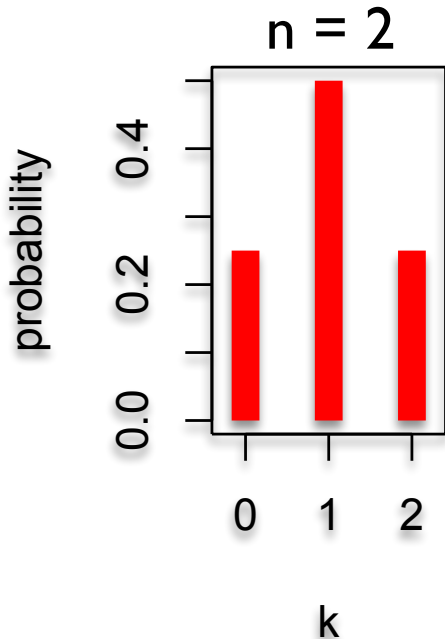
Let X be the number of heads in n independent coin tosses, each with probability p of heads.

$$p_X(k) = Pr(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$$

Let X be the number of heads observed in n coin flips

$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}, \text{ where } p = P(H)$$

Probability mass function ($p = 1/2$):



$$= \Pr(X \leq 3)$$

cumulative distribution function

The *cumulative distribution function* for a random variable X is the function $F: \mathcal{R} \rightarrow [0, 1]$ defined by

$$F(a) = P[X \leq a]$$

1 2 3

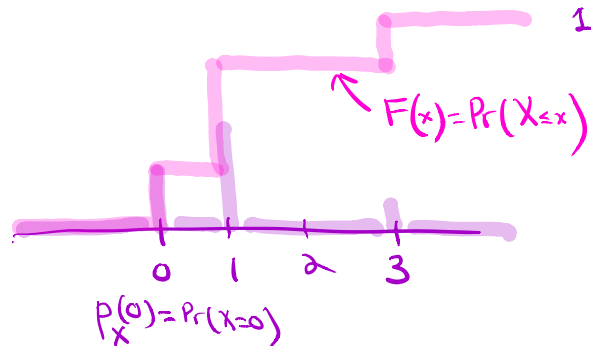
Ex: 3 students; homework returned according to random permutation.

X is number of homeworks returned to their correct homework.

What is probability mass function?

Cumulative distribution function?

Prob	outcome	X
1/6	123	3
1/6	132	1
1/6	213	1
1/6	231	0
1/6	312	0
1/6	321	1



cumulative distribution function

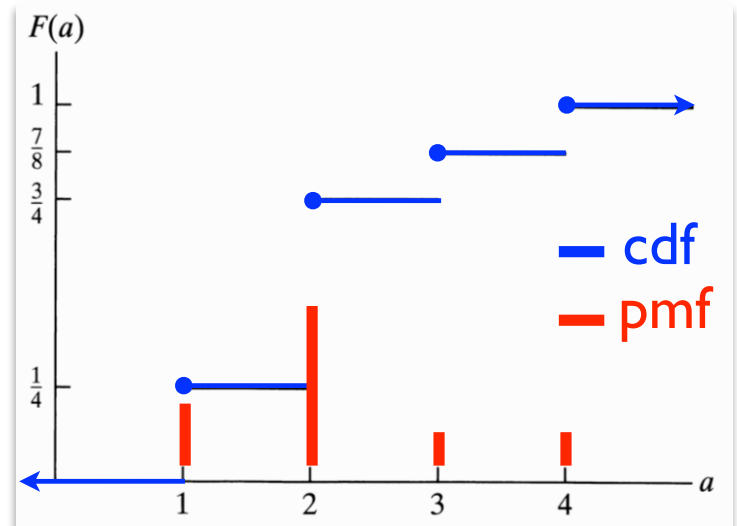
The *cumulative distribution function* for a random variable X is the function $F: \mathcal{R} \rightarrow [0,1]$ defined by

$$F(a) = P[X \leq a]$$

Ex: if X has **probability mass function** given by:

$$p(1) = \frac{1}{4} \quad p(2) = \frac{1}{2} \quad p(3) = \frac{1}{8} \quad p(4) = \frac{1}{8}$$

$$F(a) = \begin{cases} 0 & a < 1 \\ \frac{1}{4} & 1 \leq a < 2 \\ \frac{3}{4} & 2 \leq a < 3 \\ \frac{7}{8} & 3 \leq a < 4 \\ 1 & 4 \leq a \end{cases}$$



NB: for discrete random variables, be careful about “ \leq ” vs “ $<$ ”

expectation

For a discrete r.v. X with p.m.f. $p(\cdot)$, the *expectation of X* , aka *expected value* or *mean*, is

$$E[X] = \sum_x xp(x)$$

average of random values, *weighted*
by their respective probabilities

For the equally-likely outcomes case, this is just the average of the possible random values of X

For *unequally-likely* outcomes, it is again the average of the possible random values of X , **weighted by their respective probabilities**

Ex 1: Let X = value seen rolling a fair die $p(1), p(2), \dots, p(6) = 1/6$

$$E(X) = \sum_{i=1}^6 i \underbrace{\Pr(X=i)}_{p(i)} = 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + \dots + 6 \cdot \frac{1}{6} = 3.5$$

For a discrete r.v. X with p.m.f. $p(\bullet)$, the *expectation of X* , aka *expected value* or *mean*, is

$$E[X] = \sum_x xp(x)$$

average of random values, weighted by their respective probabilities

For the equally-likely outcomes case, this is just the average of the possible random values of X

For *unequally-likely* outcomes, it is again the average of the possible random values of X , **weighted by their respective probabilities**

Ex 1: Let X = value seen rolling a fair die $p(1), p(2), \dots, p(6) = 1/6$

$$E[X] = \sum_{i=1}^6 ip(i) = \frac{1}{6}(1 + 2 + \dots + 6) = \frac{21}{6} = 3.5$$

Ex 2: Coin flip; $X = +1$ if H (win \$1), -1 if T (lose \$1)

$$E(X) = 1 \cdot \underbrace{\text{Pr}(X=1)}_{\frac{1}{2}} + (-1) \cdot \underbrace{\text{Pr}(X=-1)}_{\frac{1}{2}} = 0$$

For a discrete r.v. X with p.m.f. $p(\cdot)$, the *expectation of X* , aka *expected value* or *mean*, is

$$E[X] = \sum_x xp(x)$$

average of random values, weighted
by their respective probabilities

For the equally-likely outcomes case, this is just the average of the possible random values of X

For *unequally-likely* outcomes, it is again the average of the possible random values of X , **weighted by their respective probabilities**

Ex 1: Let X = value seen rolling a fair die $p(1), p(2), \dots, p(6) = 1/6$

$$E[X] = \sum_{i=1}^6 ip(i) = \frac{1}{6}(1 + 2 + \dots + 6) = \frac{21}{6} = 3.5$$

Ex 2: Coin flip; $X = +1$ if H (win \$1), -1 if T (lose \$1)

$$E[X] = (+1) \cdot p(+1) + (-1) \cdot p(-1) = 1 \cdot (1/2) + (-1) \cdot (1/2) = 0$$

For a discrete r.v. X with p.m.f. $p(\bullet)$, the *expectation of X* , aka *expected value* or *mean*, is

$$E[X] = \sum_x xp(x)$$

average of random values, weighted
by their respective probabilities

Another view: A 2-person gambling game. If X is how much you win playing the game once, how much would you expect to win, on average, per game, when repeatedly playing?

For a discrete r.v. X with p.m.f. $p(\cdot)$, the *expectation of X* , aka *expected value* or *mean*, is

$$E[X] = \sum_x xp(x)$$

average of random values, weighted
by their respective probabilities

Another view: A 2-person gambling game. If X is how much you win playing the game once, how much would you expect to win, on average, per game, when repeatedly playing?

Ex 1: Let X = value seen rolling a fair die $p(1), p(2), \dots, p(6) = 1/6$
If you win X dollars for that roll, how much do you expect to win?

$$E[X] = \sum_{i=1}^6 ip(i) = \frac{1}{6}(1 + 2 + \dots + 6) = \frac{21}{6} = 3.5$$

Ex 2: Coin flip; $X = +1$ if H (win \$1), -1 if T (lose \$1)

$$E[X] = (+1) \cdot p(+1) + (-1) \cdot p(-1) = 1 \cdot (1/2) + (-1) \cdot (1/2) = 0$$

“a fair game”: in repeated play you expect to win as much as you lose. Long term net gain/loss = 0.

Let X be the number of flips up to & including 1st head observed in repeated flips of a biased coin (with probability p of coming up heads). If I pay you \$1 per flip, how much money would you expect to make?

X : # flips till see a H's.

$$E(X) = \sum_{i=1}^{\infty} i \Pr(X=i) = \sum_{i=1}^{\infty} i (1-p)^{i-1} p = p \sum_{i=1}^{\infty} i \underbrace{(1-p)^{i-1}}_x = p \frac{1}{(1-(1-p))^2} = \frac{1}{p}$$

$$\sum_{i=1}^{\infty} i x^{i-1} = \frac{1}{(1-x)^2}$$

$$\sum_{i \geq 0} x^i = \frac{1}{1-x},$$

when $|x| < 1$

memorize me!

Let X be the number of flips up to & including 1st head observed in repeated flips of a biased coin. If I pay you \$1 per flip, how much money would you expect to make?

$$P(H) = p; \quad P(T) = 1 - p = q$$

$$p(i) = pq^{i-1} \quad \leftarrow \text{PMF}$$

$$E[X] = \sum_{i \geq 1} ip(i) = \sum_{i \geq 1} ipq^{i-1} = p \sum_{i \geq 1} iq^{i-1} \quad (*)$$

A calculus trick:

$$\sum_{i \geq 1} iy^{i-1} = \sum_{i \geq 1} \frac{d}{dy} y^i = \sum_{i \geq 0} \frac{d}{dy} y^i = \frac{d}{dy} \sum_{i \geq 0} y^i = \frac{d}{dy} \frac{1}{1-y} = \frac{1}{(1-y)^2}$$

So (*) becomes:

$$E[X] = p \sum_{i \geq 1} iq^{i-1} = \frac{p}{(1-q)^2} = \frac{p}{p^2} = \frac{1}{p}$$

E.g.:

$p=1/2$; on average head every 2nd flip
 $p=1/10$; on average, head every 10th flip.

How much would you pay to play?

Let X be the number of heads observed in n repeated flips of a biased coin. If I pay you \$1 per head, how much money would you expect to make?

$$E(X) = \sum_{i=0}^n i \Pr(X=i)$$

$$= \sum_{i=0}^n i \binom{n}{i} p^i (1-p)^{n-i}$$

E.g.:

	$n=100$
$p=1/2$	50
$p=1/10$	10

Let X be the number of heads observed in n repeated flips of a biased coin. If I pay you \$1 per head, how much money would you expect to make?

E.g.:

$p=1/2$; on average,
 $n/2$ heads

$p=1/10$; on average,
 $n/10$ heads

$$\begin{aligned}
 E[X] &= \sum_{i=0}^n i \binom{n}{i} p^i (1-p)^{n-i} \\
 &= \sum_{i=1}^n i \binom{n}{i} p^i (1-p)^{n-i} \\
 &= \sum_{i=1}^n n \binom{n-1}{i-1} p^i (1-p)^{n-i} \\
 &= np \sum_{i=1}^n \binom{n-1}{i-1} p^{i-1} (1-p)^{n-i} \\
 &= np \sum_{j=0}^{n-1} \binom{n-1}{j} p^j (1-p)^{n-1-j} \\
 &= np(p + (1-p))^{n-1} = np
 \end{aligned}$$

For a discrete r.v. X with p.m.f. $p(\bullet)$, the *expectation of X* , aka *expected value* or *mean*, is

$$E[X] = \sum_x xp(x) = \sum_{\substack{x \\ \text{values} \\ \text{takes}}} x \Pr(X=x)$$

Another view:

$$E[X] = \sum_{s \in S} X(s) \cdot p(s)$$

$$E(X) = \sum_{\omega \in \Omega} X(\omega) \Pr(\omega)$$

