

Random Variables

A Random Variable describes a numerical property of ~~our~~^{the} outcomes in our sample space.

def Let Ω be a sample space of a probability space. A random variable is a function $X: \Omega \rightarrow \mathbb{R}$ from Ω to the real numbers.

Ex "Suppose we flip a fair coin 3 times.

Let X be the number of times that H (heads) is flipped."

Ω

• HHH	• THH
• HHT	• THT
• HTH	• TTH
• HTT	• TTT

ω	$X(\omega)$
HHH	3
HHT	2
HTH	2
HTT	1

ω	$X(\omega)$
TTH	2
THT	1
TTH	1
TTT	0

②

Remark: If X is a r.v. (random variable)

we often use " $X = a$ " for the event $\{\omega \in \Omega \mid X(\omega) = a\}$ consisting of all outcomes on which X takes the value a .

• One of the most important properties of a r.v. is the average value or expectation.

def If $X: \Omega \rightarrow \mathbb{R}$ is a random variable, then the expectation of X , denoted $E[X]$ is given by

$$E[X] = \sum_a a \cdot \Pr(X=a)$$

Remark: To treat ~~more~~ ~~cases~~ more complex cases, people

who study probability define $E[X]$ (3)

as an integral. For our purposes of studying discrete probability, this definition suffices.

Ex: Continuing with our coin example,

$$E[X] = \sum_a a \cdot \Pr(X=a)$$

$$= 0 \cdot \Pr(X=0) + 1 \cdot \Pr(X=1) + 2 \cdot \Pr(X=2) + 3 \cdot \Pr(X=3)$$

$$= 0 \cdot \frac{1}{8} + 1 \cdot \frac{3}{8} + 2 \cdot \frac{3}{8} + 3 \cdot \frac{1}{8}$$

$$= \frac{12}{8}$$

$$= \boxed{\frac{3}{2}}$$

So the expected number of heads is $\frac{3}{2} = 1.5$.

Notice that no outcome ω actually gives us

$X(\omega) = 1.5$ heads.

Linearity of Expectation

(4)

Prop Let $X: \Omega \rightarrow \mathbb{R}$ and $Y: \Omega \rightarrow \mathbb{R}$ be random variables. Define a new random variable $Z = X + Y$ by

$$Z(\omega) = X(\omega) + Y(\omega).$$

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

Proof: ~~$E[X + Y]$~~

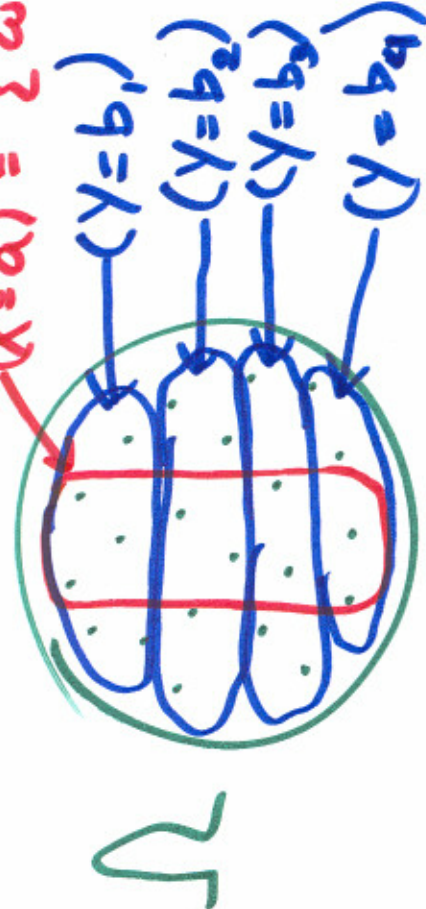
$$E[X] + E[Y] = \sum_a a \cdot \Pr(X=a) + \sum_b b \cdot \Pr(Y=b)$$

$$= \sum_a a \cdot \left(\sum_b \Pr(X=a \cap Y=b) \right) +$$

$$\sum_b b \cdot \left(\sum_a \Pr(X=a \cap Y=b) \right)$$

$$= \sum_{a,b} a \cdot \Pr(X=a \cap Y=b) + \sum_{a,b} b \cdot \Pr(X=a \cap Y=b)$$

$$X=a = \{ \omega \in \Omega \mid X(\omega) = a \}$$



$$= \sum_{a,b} (a+b) \cdot \Pr(X=a \cap Y=b)$$

(5)

Now collecting all pairs (a,b) such that $a+b=c$ yields

$$= \sum_c c \cdot \Pr\left(\sum_{\substack{a,b \\ a+b=c}} \Pr(X=a \cap Y=b)\right)$$

$$= \sum_c c \cdot \Pr(X+Y=c)$$

$$= \sum_c c \cdot \Pr(Z=c)$$

$$= E[Z]$$



Cor $E[X_1 + X_2 + \dots + X_n] = E[X_1] + E[X_2] + \dots + E[X_n]$

Linearity of Expectation can transform impossible computations into easy computations.

(Or at least difficult computations into easier computations.)

Ex: Derangements.

A total of ~~# party~~ of n people go to a concert and check their coats in the cloakroom. When they go to retrieve their coats, each person is handed one of the remaining coats at random.

Q: What is the expected number of people that receive their own coats back?

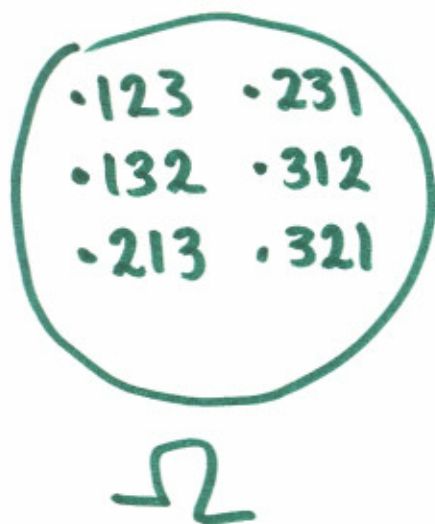
Model this problem with permutations:

$$\Omega = \{ \pi \mid \pi \text{ is a perm. of } n \text{ elts} \}$$

where the outcome $\pi \in \Omega$ represents that $\forall j$ the j th person receives $\pi(j)$'s coat. Let X be the number of people that get their own coats back; i.e.

$$X(\pi) = |\{j \mid \pi(j) = j\}|.$$

E_X : $n=3$:



π	$X(\pi)$
123	3
132	1
213	1
231	0
312	0
321	1

So $E[X] = \sum_a a \cdot \Pr(X=a)$

$$= 0 \cdot \frac{2}{6} + 1 \cdot \frac{3}{6} + ~~3~~ \cdot \frac{1}{6}$$

$$= \frac{6}{6} = \boxed{1}.$$

For general n , we let

$$A_a = \{\pi \mid |\{j \mid \pi(j) = j\}| = a\}$$

= " $(X=a)$ "

So that A_a is the event that exactly a people get their own coat back, we have by definition

$$E[X] = \sum_{a=0}^n a \cdot \Pr(X=a)$$

$$= \sum_{a=0}^n a \cdot \Pr(A_a)$$

$$= \sum_{a=0}^n a \cdot \frac{|A_a|}{|S|}$$

$$= \sum_{a=0}^n a \cdot \frac{|A_a|}{n!}$$

$$= \frac{1}{n!} \sum_{a=0}^n a \cdot |A_a|$$

$$= \dots$$

- Linearity of Expectation provides an easier way.

• For $1 \leq j \leq n$ let X_j be the r.v. given by

$$X_j(\pi) = \begin{cases} 1 & \pi(j) = j \\ 0 & \text{otherwise} \end{cases}$$

These are called indicator random variables. Note that for each $\pi \in \Omega$,

$$\begin{aligned} X(\pi) &= |\{j \mid \pi(j) = j\}| \\ &= X_1(\pi) + X_2(\pi) + \dots + X_n(\pi) \end{aligned}$$

and so $X = X_1 + X_2 + \dots + X_n$.

Therefore, by linearity of expectation,

$$E[X] = E[X_1] + E[X_2] + \dots + E[X_n]$$

What is $E[X_j]$? By defn,

$$E[X_j] = 0 \cdot \Pr(X_j = 0) + 1 \cdot \Pr(X_j = 1)$$

$$= \Pr(X_j = 1)$$

$$= \Pr(\pi(j) = j)$$

$$= \frac{|\{\pi \mid \pi(j) = j\}|}{|\Omega|}$$

$$= \frac{(n-1)!}{n!}$$

$$= \frac{1}{n}$$

Hence, $E[X] = E[X_1] + \dots + E[X_n]$

$$= \frac{1}{n} + \dots + \frac{1}{n}$$

$$= \boxed{1}$$

So the expected number of people that receive their own coats is **1** for all n , not just $n=3$.