

Roll 2 dice. Define random variables X, Y

$\hookrightarrow X$: sum of two dice.

$$X[(1,2)] = 3$$

$$X[(5,5)] = 10$$

$\hookrightarrow Y$: parity of sum.

$$Y[(1,2)] = 1$$

odd

$$Y[(5,5)] = 0$$

even

Think of these as functions, mapping sample space to a number.

e.g., $\left. \begin{array}{l} \{(1,1), (1,2), (1,3), (1,4), (1,5), (1,6), \\ (2,1), (2,2), \dots \\ \vdots \\ (6,1), (6,2), \dots \dots (6,6)\} \end{array} \right\} \begin{array}{l} X \rightarrow 2 \dots 12 \\ Y \rightarrow 0, 1 \end{array}$

Example usage: $P(X < 3) = \frac{1}{36}$ $P(Y = 1) = \frac{1}{2}$ $P(X = 13) = 0$

Expected value = weighted average

$$E[X] = \sum y \cdot P(X=y)$$

*
↳ * over all possible values, y , that X could be.

example: roll 2 dice. $X = |\text{difference between the two}|$

possible values of $X \rightarrow 0 \quad 1 \quad 2 \quad 3 \quad 4 \quad 5$

outcomes per value $\rightarrow 6 \quad 5 \cdot 2 \quad 4 \cdot 2 \quad 3 \cdot 2 \quad 2 \cdot 2 \quad 1 \cdot 2$

(for probability, divide by 36)

$$E[X] = \frac{0 + 10 + 16 + 18 + 16 + 10}{36} \sim 1.944$$

LINEARITY OF EXPECTATION $E[X + Y] = E[X] + E[Y]$

Generally, for $c_i \in \mathbb{R}$

$$E[c_1 X_1 + c_2 X_2 + \dots + c_n X_n] = c_1 E[X_1] + c_2 E[X_2] + \dots + c_n E[X_n]$$

$$E[\sum X_i] = \sum E[X_i]$$

Does NOT
assume independence \longrightarrow

$$P(X=a \ \& \ Y=b) = P(X=a) \cdot P(Y=b)$$

for all $a, b \dots$

2 dice, A, B. $X = \text{result of A.}$ $Y = \text{result of B.}$ $Z = X + Y$

$$E[Z] = E[X + Y] = E[X] + E[Y] = 2 \cdot 3.5 = 7$$

1000 dice, expected value of sum = $1000 \cdot 3.5$

EXPECTATION : PROPERTIES

$$E[X+Y] = E[X] + E[Y] \quad \rightarrow \text{always true}$$

$$\text{but } E[X \cdot Y] = E[X] \cdot E[Y] \quad \rightarrow \text{NOT always true}$$

If X & Y are independent, then $E[X \cdot Y] = E[X] \cdot E[Y]$

However, $E[X \cdot Y] = E[X] \cdot E[Y]$ does NOT imply
 X & Y are independent.

INDICATOR RANDOM VARIABLES

(IRV)

(taking value 0 or 1)

We already saw this: Y : parity of rolling one die.

Another example: flip a coin 10 times.

X = #times we see pattern HT

$$E[X] = ?$$

INDICATOR RANDOM VARIABLES

flip a coin 10 times.

$X = \#$ times we see pattern HT

HT could appear at flips 1&2, or 2&3, ..., or 9 & 10

Define r.v. $X_i = \begin{cases} 1 & \text{if flips } i \text{ \& } i+1 \text{ produce HT} \\ 0 & \text{otherwise} \end{cases}$

Notice X_1 & X_2 are not independent. $P(X_i=1) = \frac{1}{4}$
 $P(X_1 \wedge X_2) = 0$

$$X = X_1 + X_2 + \dots + X_9$$

$$\begin{aligned} E[X] &= E[X_1 + X_2 + \dots + X_9] \\ &= E[X_1] + E[X_2] + \dots + E[X_9] \end{aligned}$$

linearity of expectation

$$\begin{aligned} E[X_i] &= 0 \cdot P(X_i=0) + 1 \cdot P(X_i=1) = \frac{1}{4} \\ &= 9 \cdot \frac{1}{4} \end{aligned}$$

INDICATOR RANDOM VARIABLES

The hat-check problem (a.k.a. coat-check)

- ◆ n people at a party leave their hats with an attendant
- ◆ The attendant gives hats back randomly.

How many people do we expect to get their own hats back?

The hat-check problem

$$X = \# \text{ people who get their own hat back} = \sum_{k=1}^n X_k$$

$$X_k = \begin{cases} 1 & \text{if person } k \text{ gets their own hat back} \\ 0 & \text{otherwise} \end{cases}$$

$$E[X] = E\left[\sum_{k=1}^n X_k\right]$$

$$= \sum_{k=1}^n E[X_k]$$

$$= \sum_{k=1}^n \frac{1}{n} = \mathbf{1}$$

linearity of expectation

$$E[X_k] = \frac{1}{n} \quad \begin{array}{l} \text{(random)} \\ P(X_k = 1) \end{array}$$

The hiring problem: you need one assistant.

◆ n candidates, interviewed in random order.

◆ No 2 equally skilled.

◆ any time you interview someone better than all previous, you hire the new person & fire the current assistant.

How many people do you expect to hire?

The hiring problem

$$X = \# \text{ people you will hire} = \sum_{k=1}^n X_k$$

$$X_k = \begin{cases} 1 & \text{if you hire candidate } k \\ 0 & \text{otherwise} \end{cases}$$

$$E[X] = E\left[\sum_{k=1}^n X_k\right]$$

$$= \sum_{k=1}^n E[X_k]$$

$$= \sum_{k=1}^n \frac{1}{k} \leq 1 + \ln n$$

linearity of expectation

$$E[X_k] = \frac{1}{k}$$

person k is hired
iff better than
all $k-1$ previous

The birthday problem

How many people do we need in a room so that we expect to have (at least) one birthday match?

The birthday problem

X = # birthday matches among n people

For what n do we get $E[X] \geq 1$?

↳ Set up $E[X]$ as function of n

The birthday problem

X = # birthday matches among n people

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}$$

all $\binom{n}{2}$ pairs

$$X_{ij} = \begin{cases} 1 & \text{if persons } i \text{ \& } j \text{ match} \\ 0 & \text{otherwise} \end{cases}$$

$$E[X_{ij}] = \frac{1}{365}$$

$$E[X] = E\left[\sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}\right] = \sum_{i=1}^{n-1} \sum_{j=i+1}^n E[X_{ij}]$$

linearity of expectation
we said we want $E[X]=1$

$$= \sum_{i=1}^{n-1} \sum_{j=i+1}^n \frac{1}{365} = \frac{n \cdot (n-1)}{2} \cdot \frac{1}{365} = 1 \Rightarrow n \approx 28$$

□

$P(\geq 2 \text{ people in a group of } k \text{ have same birthday})$

(new problem)
Not about
expected value or IRV

$= 1 - P(\text{nobody has the same birthday})$

$$= 1 - \frac{365}{365} \cdot \frac{364}{365} \cdot \frac{363}{365} \cdot \dots \cdot \frac{365 - (k-1)}{365} = 1 - \frac{365!}{(365-k)! 365^k}$$

Diagram annotations: Green brackets above the fractions label them as "person #1", "person #2", "person #3", and "person #k".

$P(\geq 2 \text{ people in a group of } k \text{ have same birthday})$

$$k=2 \rightarrow P \sim 0.27\% \left(\frac{1}{365}\right)$$

$$k=4 \rightarrow P \sim 1.64\%$$

$$k=23 \rightarrow P \sim 50.73\%$$

$$k=30 \rightarrow P \sim 70.6\%$$

$$k=70 \rightarrow P \sim 99.9\%$$

$$k \sim 116 \rightarrow P \sim 1 - \frac{1}{10^9}$$

$$k=300 \rightarrow P \sim 1 - \frac{1}{10^{80}}$$

$(10^{80} \sim \# \text{ atoms in universe})$

$$(k > 365 \rightarrow P=1)$$