

Today

Random Variables.

Questions about outcomes ...

Experiment: roll two dice.

Sample Space: $\{(1,1), (1,2), \dots, (6,6)\} = \{1, \dots, 6\}^2$

How many pips?

Experiment: flip 100 coins.

Sample Space: $\{HHH \dots H, THH \dots H, \dots, TTT \dots T\}$

How many heads in 100 coin tosses?

Experiment: choose a random student in cs70.

Sample Space: $\{Adam, Jin, Bing, \dots, Angeline\}$

What midterm score?

Experiment: hand back assignments to 3 students at random.

Sample Space: $\{123, 132, 213, 231, 312, 321\}$

How many students get back their own assignment?

In each scenario, each outcome gives a number.

The number is a (known) function of the outcome.

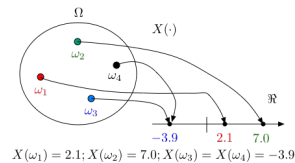
Quick Review: Probability. Some Rules.

- ▶ **Sample Space:** Set of outcomes, Ω .
- ▶ **Probability:** $Pr[\omega]$ for all $\omega \in \Omega$.
 - ▶ $0 \leq Pr[\omega] \leq 1$.
 - ▶ $\sum_{\omega \in \Omega} Pr[\omega] = 1$.
- ▶ **Event:** $A \subseteq \Omega$. $Pr[A] = \sum_{\omega \in A} Pr[\omega]$
 - ▶ Inclusion/Exclusion: $Pr[A \cup B] = Pr[A] + Pr[B] - Pr[A \cap B]$.
 - ▶ Simple Total Probability: $Pr[B] = Pr[A \cap B] + Pr[\bar{A} \cap B]$.
 - ▶ Complement: $Pr[\bar{A}] = 1 - Pr[A]$.
 - ▶ Union Bound. Total Probability.
- ▶ **Conditional Probability:** $Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]}$
- ▶ **Bayes' Rule:** $Pr[A_m|B] = p_m q_m / (p_1 q_1 + \dots + p_M q_M)$.
- ▶ **Product Rule:**
 $Pr[A_1 \cap \dots \cap A_n] = Pr[A_1] Pr[A_2|A_1] \dots Pr[A_n|A_1 \cap \dots \cap A_{n-1}]$.
- ▶ **Total Probability/Product:** $Pr[B] = Pr[B|A] Pr[A] + Pr[B|\bar{A}] Pr[\bar{A}]$.

Random Variables.

A **random variable**, X , for an experiment with sample space Ω is a **function** $X: \Omega \rightarrow \mathfrak{R}$.

Thus, $X(\cdot)$ assigns a real number $X(\omega)$ to each $\omega \in \Omega$.



Function $X(\cdot)$ defined on outcomes Ω .

Function $X(\cdot)$ is **not random, not a variable!**

What varies at random (among experiments)? **The outcome!**

Note: Random variable induces partition:

$$A_y = \{\omega \in \Omega : X(\omega) = y\} = X^{-1}(y)$$

Random Variables

Random Variables

1. Random Variables.
2. Expectation
3. Distributions.

Example 1 of Random Variable

Experiment: roll two dice.

Sample Space: $\{(1,1), (1,2), \dots, (6,6)\} = \{1, \dots, 6\}^2$

Random Variable X : number of pips.

$$X(1,1) = 2$$

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

\vdots

$$X(6,6) = 12,$$

$$X(a,b) = a + b, (a,b) \in \Omega.$$

Example 2 of Random Variable

Experiment: flip three coins

Sample Space: $\{HHH, THH, HTH, TTH, HHT, THT, HTT, TTT\}$

Winnings: if win 1 on heads, lose 1 on tails: X

$X(HHH) = 3$ $X(THH) = 1$ $X(HTH) = 1$ $X(TTH) = -1$
 $X(HHT) = 1$ $X(THT) = -1$ $X(HTT) = -1$ $X(TTT) = -3$

Handing back assignments

Experiment: hand back assignments to 3 students at random.

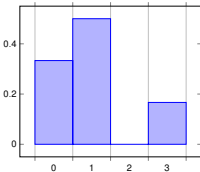
Sample Space: $\Omega = \{123, 132, 213, 231, 312, 321\}$

How many students get back their own assignment?

Random Variable: values of $X(\omega) : \{3, 1, 1, 0, 0, 1\}$

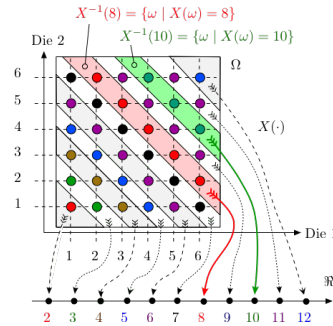
Distribution:

$$X = \begin{cases} 0, & \text{w.p. } 1/3 \\ 1, & \text{w.p. } 1/2 \\ 3, & \text{w.p. } 1/6 \end{cases}$$



Number of pips in two dice.

"What is the likelihood of getting n pips?"

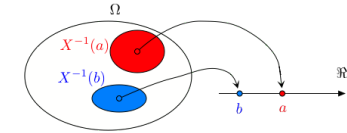


$$Pr[X = 10] = 3/36 = Pr[X^{-1}(10)]; Pr[X = 8] = 5/36 = Pr[X^{-1}(8)].$$

Distribution

The probability of X taking on a value a .

Definition: The **distribution** of a random variable X , is $\{(a, Pr[X = a]) : a \in \mathcal{A}\}$, where \mathcal{A} is the range of X .



$$Pr[X = a] := Pr[X^{-1}(a)] \text{ where } X^{-1}(a) := \{\omega \mid X(\omega) = a\}.$$

Flip three coins

Experiment: flip three coins

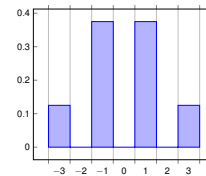
Sample Space: $\{HHH, THH, HTH, TTH, HHT, THT, HTT, TTT\}$

Winnings: if win 1 on heads, lose 1 on tails: X

Random Variable: $\{3, 1, 1, -1, -1, -1, -3\}$

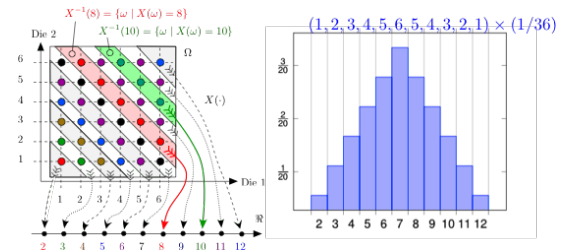
Distribution:

$$X = \begin{cases} -3, & \text{w. p. } 1/8 \\ -1, & \text{w. p. } 3/8 \\ 1, & \text{w. p. } 3/8 \\ 3, & \text{w. p. } 1/8 \end{cases}$$



Number of pips.

Experiment: roll two dice.



Expectation.

How did people do on the midterm?

Distribution.

Summary of distribution?

Average!



An Example

Flip a fair coin three times.

$\Omega = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}$.

$X =$ number of H 's: $\{3, 2, 2, 2, 1, 1, 1, 0\}$.

Thus,

$$\sum_{\omega} X(\omega) Pr[\omega] = \{3+2+2+2+1+1+1+0\} \times \frac{1}{8}.$$

Also,

$$\sum_a a \times Pr[X = a] = 3 \times \frac{1}{8} + 2 \times \frac{3}{8} + 1 \times \frac{3}{8} + 0 \times \frac{1}{8}.$$

What's the answer? Uh... $\frac{3}{2}$

Expectation - Definition

Definition: The **expected value** of a random variable X is

$$E[X] = \sum_a a \times Pr[X = a].$$

The expected value is also called the mean.

According to our intuition, we expect that if we repeat an experiment a large number N of times and if X_1, \dots, X_N are the successive values of the random variable, then

$$\frac{X_1 + \dots + X_N}{N} \approx E[X].$$

That is indeed the case, in the same way that the fraction of times that $X = x$ approaches $Pr[X = x]$.

This (nontrivial) result is called the [Law of Large Numbers](#).

The subjectivist(bayesian) interpretation of $E[X]$ is less obvious.

Expectation and Average.

There are n students in the class;

$X(m) =$ score of student m , for $m = 1, 2, \dots, n$.

"Average score" of the n students: add scores and divide by n :

$$\text{Average} = \frac{X(1) + X(2) + \dots + X(n)}{n}.$$

Experiment: choose a student uniformly at random.

Uniform sample space: $\Omega = \{1, 2, \dots, n\}$, $Pr[\omega] = 1/n$, for all ω .

Random Variable: midterm score: $X(\omega)$.

Expectation:

$$E(X) = \sum_{\omega} X(\omega) Pr[\omega] = \sum_{\omega} X(\omega) \frac{1}{n}.$$

Hence,

$$\text{Average} = E(X).$$

This holds for a [uniform](#) probability space.

Expectation: A Useful Fact

Theorem:

$$E[X] = \sum_{\omega} X(\omega) \times Pr[\omega].$$

Proof:

$$\begin{aligned} E[X] &= \sum_a a \times Pr[X = a] \\ &= \sum_a a \times \sum_{\omega: X(\omega)=a} Pr[\omega] \\ &= \sum_a \sum_{\omega: X(\omega)=a} X(\omega) Pr[\omega] \\ &= \sum_{\omega} X(\omega) Pr[\omega] \end{aligned}$$

Distributive property of multiplication over addition. □

Named Distributions.

Some distributions come up over and over again.

...like "choose" or "stars and bars"....

Let's cover some.

The binomial distribution.

Flip n coins with heads probability p .

Random variable: number of heads.

Binomial Distribution: $Pr[X = i]$, for each i .

How many sample points in event " $X = i$ "?

i heads out of n coin flips $\implies \binom{n}{i}$

What is the probability of ω if ω has i heads?

Probability of heads in any position is p .

Probability of tails in any position is $(1-p)$.

So, we get

$$Pr[\omega] = p^i (1-p)^{n-i}.$$

Probability of " $X = i$ " is sum of $Pr[\omega]$, $\omega \in "X = i"$.

$$Pr[X = i] = \binom{n}{i} p^i (1-p)^{n-i}, i = 0, 1, \dots, n: B(n, p) \text{ distribution}$$

Expectation of Binomial Distribution

Parameter p and n . What is expectation?

$$Pr[X = i] = \binom{n}{i} p^i (1-p)^{n-i}, i = 0, 1, \dots, n: B(n, p) \text{ distribution}$$

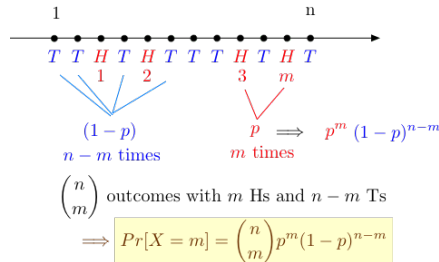
$$E[X] = \sum_i i \times Pr[X = i].$$

Uh oh? Well... It is pn .

Proof? After linearity of expectation this is easy.

Waiting is good.

The binomial distribution.



Uniform Distribution

Roll a six-sided balanced die. Let X be the number of pips (dots).

Then X is equally likely to take any of the values $\{1, 2, \dots, 6\}$. We say that X is *uniformly distributed* in $\{1, 2, \dots, 6\}$.

More generally, we say that X is uniformly distributed in $\{1, 2, \dots, n\}$ if

$Pr[X = m] = 1/n$ for $m = 1, 2, \dots, n$.

In that case,

$$E[X] = \sum_{m=1}^n m Pr[X = m] = \sum_{m=1}^n m \times \frac{1}{n} = \frac{1}{n} \frac{n(n+1)}{2} = \frac{n+1}{2}.$$

Error channel and...

A packet is corrupted with probability p .

Send $n+2k$ packets.

Probability of at most k corruptions.

$$\sum_{i \leq k} \binom{n+2k}{i} p^i (1-p)^{n+2k-i}.$$

Also distribution in polling, experiments, etc.

Geometric Distribution

Let's flip a coin with $Pr[H] = p$ until we get H .



For instance:

$\omega_1 = H$, or
 $\omega_2 = T H$, or
 $\omega_3 = T T H$, or
 $\omega_n = T T T T \dots T H$.

Note that $\Omega = \{\omega_n, n = 1, 2, \dots\}$.

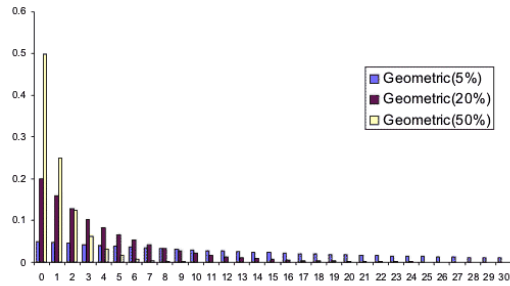
Let X be the number of flips until the first H . Then, $X(\omega_n) = n$.

Also,

$$Pr[X = n] = (1-p)^{n-1} p, n \geq 1.$$

Geometric Distribution

$$\Pr[X = n] = (1 - p)^{n-1} p, n \geq 1.$$



Geometric Distribution

$$\Pr[X = n] = (1 - p)^{n-1} p, n \geq 1.$$

Note that

$$\sum_{n=1}^{\infty} \Pr[X_n] = \sum_{n=1}^{\infty} (1 - p)^{n-1} p = p \sum_{n=1}^{\infty} (1 - p)^{n-1} = p \sum_{n=0}^{\infty} (1 - p)^n.$$

Now, if $|a| < 1$, then $S := \sum_{n=0}^{\infty} a^n = \frac{1}{1-a}$. Indeed,

$$\begin{aligned} S &= 1 + a + a^2 + a^3 + \dots \\ aS &= a + a^2 + a^3 + a^4 + \dots \\ (1-a)S &= 1 + a - a + a^2 - a^2 + \dots = 1. \end{aligned}$$

Hence,

$$\sum_{n=1}^{\infty} \Pr[X_n] = p \frac{1}{1 - (1 - p)} = 1.$$

Geometric Distribution: Expectation

$$X =_D G(p), \text{ i.e., } \Pr[X = n] = (1 - p)^{n-1} p, n \geq 1.$$

One has

$$E[X] = \sum_{n=1}^{\infty} n \Pr[X = n] = \sum_{n=1}^{\infty} n (1 - p)^{n-1} p.$$

Thus,

$$\begin{aligned} E[X] &= p + 2(1-p)p + 3(1-p)^2 p + 4(1-p)^3 p + \dots \\ (1-p)E[X] &= (1-p)p + 2(1-p)^2 p + 3(1-p)^3 p + \dots \\ pE[X] &= p + (1-p)p + (1-p)^2 p + (1-p)^3 p + \dots \\ &\quad \text{by subtracting the previous two identities} \\ &= \sum_{n=1}^{\infty} \Pr[X = n] = 1. \end{aligned}$$

Hence,

$$E[X] = \frac{1}{p}.$$

Summary

Random Variables

- ▶ A random variable X is a function $X : \Omega \rightarrow \mathfrak{R}$.
- ▶ $\Pr[X = a] := \Pr[X^{-1}(a)] = \Pr[\{\omega \mid X(\omega) = a\}]$.
- ▶ $\Pr[X \in A] := \Pr[X^{-1}(A)]$.
- ▶ The distribution of X is the list of possible values and their probability: $\{(a, \Pr[X = a]), a \in \mathcal{A}\}$.
- ▶ $E[X] := \sum_a a \Pr[X = a]$.
- ▶ Expectation is Linear.
- ▶ $B(n, p), U[1 : n], G(p), P(\lambda)$.