

## 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.

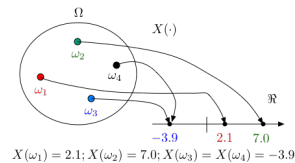
## Quick Review: Probability. Some Rules.

- ▶ **Sample Space:** Set of outcomes,  $\Omega$ .
- ▶ **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  $\Omega$  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  $\Omega$ .

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.

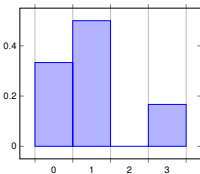
Sample Space:  $\Omega = \{123, 132, 213, 231, 312, 321\}$

How many students get back their own assignment?

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

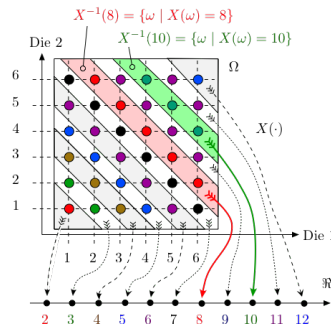
Distribution:

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



## Number of pips in two dice.

"What is the likelihood of getting  $n$  pips?"



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

## Flip three coins

Experiment: flip three coins

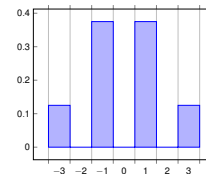
Sample Space:  $\{HHH, THH, HTH, TTH, HHT, THT, HTT, TTT\}$

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

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

Distribution:

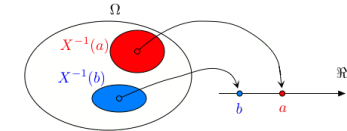
$$X = \begin{cases} -3, & \text{w. p. } 1/8 \\ -1, & \text{w. p. } 3/8 \\ 1, & \text{w. p. } 3/8 \\ 3, & \text{w. p. } 1/8 \end{cases}$$



## Distribution

The probability of  $X$  taking on a value  $a$ .

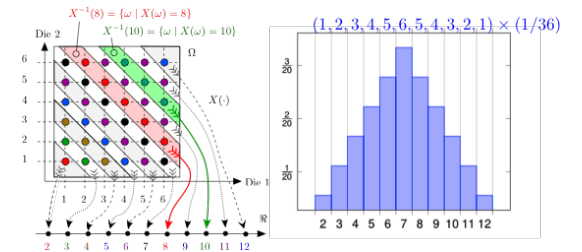
**Definition:** The **distribution** of a random variable  $X$ , is  $\{(a, Pr[X = a]) : a \in \mathcal{A}\}$ , where  $\mathcal{A}$  is the range of  $X$ .



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

## 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  $\omega$ .

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  $\omega$  if  $\omega$  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?

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

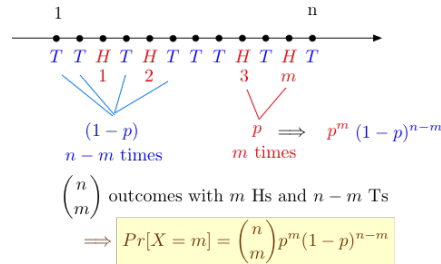
$$E[X] = \sum_i i \times Pr[X = i].$$

Uh oh? Well... It is  $pn$ .

Proof? After linearity of expectation this is easy.

Waiting is good.

## The binomial distribution.



## 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.

## 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,

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

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

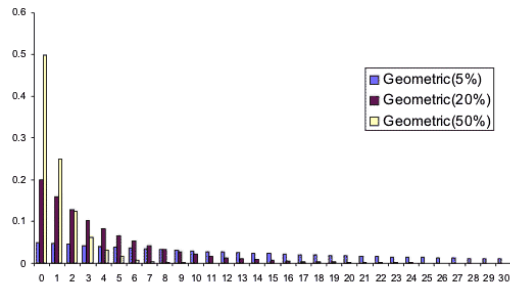
Let  $X$  be the number of flips until the first  $H$ . Then,  $X(\omega_n) = n$ .

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,

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

Hence,

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

## Poisson: Motivation and derivation.

McDonalds: How many McDonalds arrive in an hour?

Know: average is  $\lambda$ .

What is distribution?

Example:  $\Pr[2\lambda \text{ arrivals}]$ ?

Assumption: "arrivals are independent."

Derivation: cut hour into  $n$  intervals of length  $1/n$ .

$\Pr[ \text{two arrivals} ]$  is  $(\lambda/n)^2$  or small if  $n$  is large.

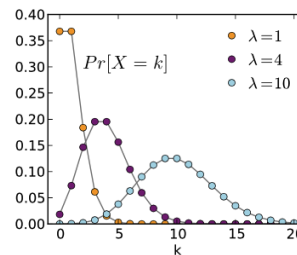
Model with binomial.

## Poisson

Experiment: flip a coin  $n$  times. The coin is such that  $\Pr[H] = \lambda/n$ .

Random Variable:  $X$  - number of heads. Thus,  $X = B(n, \lambda/n)$ .

**Poisson Distribution** is distribution of  $X$  "for large  $n$ ."



## 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:  $\{(a, \Pr[X = a]), a \in \mathcal{A}\}$ .
- ▶  $E[X] := \sum_a a \Pr[X = a]$ .
- ▶  $B(n, p), U[1 : n], G(p), P(\lambda)$ .