

Probability in Computing



Reminders

• HW 3 due on Thursday

LECTURE 6 Last time

- Spinners and Continuous Probability
- Geometric Probability
- Anomalies with Continuous Probability

Today

- Five kinds of sample spaces
- Random Variables



Discrete Sample Spaces are finite or countably infinite

- We assign a probability to each outcome
- The probability of an event is the sum of the probabilities of its members: $A = \{a_1, \dots, a_n\} \rightarrow \Pr = \Pr(a_1) + \dots + \Pr(a_n)$
- Can be uniform or non-uniform
- **1. Finite Uniform Sample Spaces:** Example $Pr(A) = \frac{|A|}{|\Omega|}$
 - Experiment: Roll a die, count the number of dots showing
 - Sample space: $\Omega = \{ 1, 2, 3, 4, 5, 6 \}$
 - Probability Function: $P(1) = \frac{1}{6}, \quad P(2) = \frac{1}{6}, \quad P(3) = \frac{1}{6}$ $P(4) = \frac{1}{6}, \quad P(5) = \frac{1}{6}, \quad P(6) = \frac{1}{6}$



2. Finite Non-Uniform Sample Spaces: Example

- Experiment: Throw two dice, count the number of dots showing
- Sample space: $\Omega = \{ 2, 3, 4, ..., 12 \}$
- Probability Function:





	1	2	3	4	5	6
1	2	3	4	5	6	7
2	3	4	5	6	7	8
3	4	5	6	7	8	9
4	5	6	7	8	9	10
5	6	7	8	9	10	11
6	7	8	9	10	11	12

CS Summary: 5 Types of Sample Spaces

3. Countably Infinite Non-Uniform Sample Spaces: Example 1

- Experiment: Flip a coin until a head, count how many flips.
- Sample space:
- Probability Function

 $\Omega = \{ 1, 2, 3, 4, 5, \dots \}$ notion $\omega: 1 2 3 4 \dots k \dots$ $Pr(\omega): 1/2 1/4 1/8 1/16 \dots 1/2^k \dots$



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CS Summary: 5 Types of Sample Spaces

3. Countably Infinite Non-Uniform Sample Spaces: Example 2

- Experiment: Roll a die until a 6 appears, count how many rolls.
- Sample space: $\Omega = \{ 1, 2, 3, 4, \}$
- Probability Function:





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Continuous sample spaces are not countable

- Individual outcomes have probability zero
- We assign probabilities to the extent (length, area, volume, etc.) taken up by the event of interest

4. Uncountably Infinite Uniform Sample Spaces: Example

- Experiment: Uniformly pick an interval from 0 to 1.
- Sample space: $\Omega = [0 .. 1]$
- Event: $\omega \in [a..b]$
- Probability Function: Pr([a..b]) = b a



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5. Uncountably Infinite Non-Uniform Sample Spaces: Example

- Experiment: Throw a dart at a circular target of radius 1.0, what is the probability of hitting the annulus below?
- Sample space: $\Omega = [0 \dots 1]$
- Event: $\omega \in [a..b]$
- Probability Function:



CS Top Hat question (Join Code: 413437)

Experiment: Pick a BU student uniformly at random and measure their height in meters, rounded to 2 decimal places.

The Sample Space is:

- A. Finite and uniform
- B. Finite and not uniform
- C. Countably infinite
- D. Uncountably infinite and uniform
- E. Uncountably infinite and not uniform



Random Experiments and RandomVariables

A Random Experiment is a repeatable procedure that produces uncertain outcomes from a well-defined sample space.

Example: Flip a coin!





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Definition. A Random Variable X is a function from a sample space Ω into the reals:

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

It *interprets* the random experiment as a real number. Now when an outcome is requested, the sample point is translated into a real number:

X = "Flip a coin and count the number of heads showing"





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RandomVariables

A random variable will not in general simply rename outcomes with real numbers, it will compute some useful information about the outcome.

V = "Flip 3 coins: return 1 if there are more heads showing than tails, 0 if more tails than heads "

V simulates a



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CS Range of random variables

- Since a random variable is a function, we can talk about its range.
- The range of a random variable X, shown by Range(X) or R_X , is the set of possible values for \underline{X} .
- I toss a coin five times. Suppose that I'm interested in the number of heads. Let X be a random variable whose value is the number of heads.

Range(X)=
$$R_X$$
={0,1,2,3,4,5}

CS Top Hat question (Join Code: 033357)

Consider the following experiment:

"I toss a coin until the first heads appears. Let Y be the total number of coin tosses."

What is the range of the random variable Y?:

A.
$$R_Y = \{H, T\}$$

B. $R_Y = \{0, 1, 2, 3, 4, 5 \dots\}$
C. $R_Y = \{\sum \frac{1}{2^n} : n \in N\}$
D. $R_Y = \{\frac{1}{2}, \frac{1}{3}, \frac{1}{4}, \frac{1}{5}, \frac{1}{6}, \dots\}$
 $_{2/7/25}$ $R_Y = \{\frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \frac{1}{32}, \frac{1}{64}, \dots\}$

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Joint Random Variables are combinations of 2 or more RVs, giving *multiple interpretations* of the same experiment.

Example (from LLM):

"Flip 3 coins and

(C) return the number of heads showing;

(M) return 1 if all heads or all tails, 0 otherwise"



CS Top Hat question (Join Code: 033357)

Consider the following experiment:

"Flip 3 coins and

- (T) return the number of tails showing;
- (V) return 1 if there are more heads showing than tails, 0 if more tails than heads

Suppose the random experiment chooses HHT from Ω ; then the result is:

A. T = 1 V = 0

- **B.** T = 2 V = 0
- **C.** T = 2 V = 1
- $_{2/7/2}$ **D**. None of the above



The term "Random Variable" is a misnomer: It's really a function!

Two things to remember about RVs as functions:

1. A random variable X is characterized by its range:

Range(X) = {
$$r \mid \exists \omega \in \Omega. X(\omega) = r$$
 }

X is finite/countably infinite/continuous/uniform/non-uniform if it behaves that way on its range.

Example: Choose a student from this class:

X = "Their height rounded to the nearest mm;"(Finite/non-uniform)Y = "Their age;"(Continuous/non-uniform)Z = "How many tattoos they have"(Finite/non-uniform)



Second thing to remember about random variables as functions:

2. Random variables behave just like Python functions which return random values.

X = "Flip a coin and count the number of heads showing"





"Flip 3 coins and

- (C) return the number of heads showing;
- (M) return 1 if all heads or all tails, 0 otherwise"

```
In [3]:
              def CM():
            1
                omega = (X(), X(), X())
               print("outcome = ", omega)
            3
                C = sum(omega)
            4
                   M = (1 \text{ if } C == 3 \text{ or } C == 0 \text{ else } 0)
            5
                   return (C,M)
            6
            7
            8
              CM()
          outcome = (1, 0, 0)
 Out[3]: (1, 0)
In [22]:
              CM()
            1
          outcome = (1, 1, 1)
Out[22]: (3, 1)
```



A very typical notion for probabilities involving a random variable X is:

Pr(< some Boolean expression involving X >)

Examples for discrete RVs:

$$Pr(X = k) = P_X(k)$$

$$Pr(X \neq k) = 1.0 - P_X(k)$$

$$Pr(X \leq k) = \sum_{a \leq k} P_X(a)$$

$$Pr(j < X \leq k) = \sum_{j < a \leq k} P_X(a)$$

A useful notation in the discrete case from the alternate textbook:

 $P_X(k)$ = the probability that X returns k

Example: X = "Count the number of flips until the coin shows heads"

Discrete Random Variables: Notation



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For the random variable

X = "Count the number of flips until the coin shows heads"

What is Pr(X > k) for k > 0?



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CS 237

Probability Mass Function (PMF)

Let A = {X = x_k} be the set of outcomes s ∈ S for which the corresponding value of X is equal to x_k. In particular,

$$A = \{s \in S \mid X(s) = x_k\}$$

- The probabilities of events {X = x_k} are formally shown by the Probability Mass Function (PMF) of X, denoted as P_X.
- PMF is a probability measure that gives us probabilities of the possible values for a random variable.



- I toss a fair coin twice, and let X be defined as the number of heads I observe. Find R_X and P_X
- Sample space: $S = \{HH, HT, TH, TT\}$ $R_X = \{0, 1, 2\}$
- $P_X(k) = P(X = k)$ for k = 0,1,2
- $P_X(0) = P(X = 0) = P(TT) = \frac{1}{4}$
- $P_X(1) = P(X = 1) = P(\{HT, TH\}) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$
- $P_X(2) = P(X = 2) = P(HH) = \frac{1}{4}$



- Homework 3 due on Thursday, February 9th
 - Analytical problems
 - Programming assignment
- Ask your questions on Piazza!
- This lecture is covered on the second text book, chapters:
 - P 3.1.1-3.1.3,3.1.6,
 - P 3.2.1, 4.0-4.1



