## STAT 305: Chapter 5

Part I

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## Chapter 5: Probability

Mathematically Describing Randomness

and

Discrete Random Variables

### **Probability Theory**

# What is Probability Theory?

In a mathematics, the field of **Probability** is the axioms, theories, concepts, terminology, and discoveries that are used to allow for random variation in a strict, rigorous, and mathematically (or logically) sound way. Using probability theories, we can use our existing knowledge of mathematics to deal with elements in a system that behave in chaotic ways.

#### History

In the long history of mathematics, Probability is a fairly young branch. Initial attempts to make random chance events the subject of mathematical study go back to the 17th century, but the strongest mathematical foundations were largely laid in the 20th century, with much credit going to Andrey Kolmogorov (1903-1987).

## **Probability**

# What is Probability Theory?

#### History, Cont

In it's foundations, Probability is relies on much of **measure theory**, a branch of mathematics concerned with measurement.

In Kolmogorov's application, a probability is a way of measuring the likelihood of a given outcome. By grafting the earlier probability concepts onto the elements of measure theory, Kolmogorov created an axiomatic basis for probability on which others could base their work with certainty.

## Probability Basics

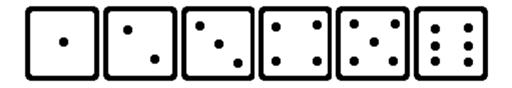
What is Probability Theory?

An Example: Throwing Dice

#### **Basics**

To introduce the core concepts, we can start with some examples that go back further than the field itself.

Consider a 6-sided die.



- a) In terms of the number of dots facing up, what are the possible outcomes from a single roll of the die?
- b) What would it mean for the die to be a "fair" die?

#### Basic Terminology I

# What is Probability Theory?

We call the process of tossing the die and observing the number of dots facing up a **random experiment** - meaning, we are rigorous about how we do the process but we still expect the end result to be change.

#### **Basics**

A few more key terms:

**Sample space**: The set of all possible outcomes from a random experiment.

**Event**: A subset of the sample space.

**Fair**: A system is fair if all the outcomes in the sample space are equally likely to occur.

#### An Example: Throwing Dice

What is Probability Theory?

In our example we have the following sample space:

S = {"one dot showing up", "two dots showing up", ..., "six dots facing up"}

#### **Basics**

That's a little cumbersome to write out, but if we all agree we are talking about dots facing up (instead of, say, cows in the field), then it's OK to write this as:

$$S = \{1, 2, 3, 4, 5, 6\}$$

All of our existing set notation and rules work on sample spaces: union, intersection, compliment, subset, etc.

#### Review of Working with Sets

# What is Probability Theory?

#### **Basics**

We have the following key terms and set operations:

- **element**: The term used for a member of a set.
- **universe**: The universe is the set of all elements (in probability, the sample space is our universe)
- union: We define  $A \cup B$  as the set of all elements in either A or B.
- **intersection**: We define  $A \cap B$  as the set of all elements in A and B.
- **compliment**: We define  $A^c$  as the set of all elements in the universe that are **not** in A.
- **subset**: We say that *A* is a subset of *B* if every element in *A* is in *B*.
- **empty set**: the empty set is the set with no elements. We write this as  $\{\}$  or  $\emptyset$ .

#### Review of Working with Sets

# What is Probability Theory?

Suppose that U is the set of all letters. Suppose  $A=\{a,b,c\}, B=\{b,c,e\}$  and  $A_1=\{x,y,z\}.$ 

Find the following:

#### **Basics**

- 1.  $A^c$
- $2. A \cup B$
- $3. A \cap B$
- $A \cdot A \cap A_1$
- 5.  $A^c \cup B$
- 6.  $A^c \cap B$

#### An Example: Throwing Dice

# What is Probability Theory?

Any subset of our sample space is an event. Using the sample space

$$S = \{1, 2, 3, 4, 5, 6\}$$

#### **Basics**

- If we take the subset  $E_1=\{1,3,5\}$ , then  $E_1$  is the event that an odd number of dots are facing up.
- If we are interested in whether or not the number of dots facing up is less that 3, we could write  $E_2=\{1,2\}$  as an event.
- If the only outcome we care about is if the roll results in 6 dots facing up, we would be interested in the event  $E_3 = \{6\}$ .
- We can still do set math with events: for example,  $E_1 \cap E_2 = \{1\}$

Note: For a samples space S, S is also an event and so is  $\emptyset$ .

### An Example: Throwing Dice

What is Probability Theory?

Basics

Since our sample space has 6 possible outcomes,

$$S = \{1, 2, 3, 4, 5, 6\}$$

then for a die roll to be "fair" each of the 6 should have the same chance of occuring. We could think of think of this by saying

• For each possible result, there is a 1 in 6 chance that the next toss will be that result.

or

• If we were able to continue tossing this die infinitly, then each outcome will be seen on 1/6 of the tosses.

#### Basic Terminology II: Probability and Its Axioms

# What is Probability Theory?

**Probability**: A special measurement used to describe the likelihood of a specific event. A probability of 0 means the event will not occur. A probability of 1 means the event will occur.

#### **Basics**

In order to valid for a sample space, S, the probability must follow these rules:

- 1. For any event A, P(A) > 0
- 2. P(S) = 1
- 3. If  $A \cap B = \emptyset$  then  $P(A \cup B) = P(A) + P(B)$ .

We can translate these rules into words:

- 1. All probabilities are at least 0 (they can be 0).
- 2. The probability that something occurs is 1.
- 3. The probability of any event is the sum of the probabilities of its parts.

#### Basic Terminology II: Probability and Its Axioms

What is Probability Theory?

By combining these three rules we get the many other rules, such as the following:

#### **Basics**

• 
$$P(A^C) = 1 - P(A)$$

- For any events A and B,  $P(A \cup B) \leq P(A) + P(B)$
- $P(\emptyset) = 0$
- ullet For disjoint events  $A_1,A_2,...,A_k,\ P(A_1\cup A_2\cup\ldots\cup A_k)=P(A_1)+P(A_2)+\ldots+P(A_k)$

.

#### Basic Terminology II: Probability and Its Axioms

What is Probability Theory?

**Basics** 

#### Example

Suppose that A and B are two events of a sample space S. Using the probability axioms, it can be shown that

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

## Working with Sample Spaces

From Sample Spaces to Random Variables

#### Example: Red die, blue die

### Axioms Example

A fair red die and fair blue die are tossed at the same time. The number of dots facing up on each die are recorded. We can ask a lot of questions related to the outcome of the toss:

# Sample Space Example

- 1. What is the probability that the red die shows 4 dots facing up?
- 2. What is the probability that the blue die and the red die both have 4 dots facing up?
- 3. What is the probability that one of the die has 4 dots facing up?
- 4. What is the probability that neither of the die has 4 dots facing up?
- 5. What is the most likely total number of dots facing up?
- 6. Assuming the red die shows a 6, what is the probability that the blue die shows a 4?
- 7. What is the probability that the total number of dots facing up is 5?

### Example: Red die, blue die (cont)

#### Axioms Example

#### **Creating a Random Variable**

## Sample Space Example

With the sample space in hand, we might find many of our questions about the total are easy to handle by defining a new variable:

Let T be the total number of dots facing up on both die after performing the random experiment.

So for instance, if the outcome we observe is (1,3) (red is 1, blue is 3) then T=4. In this way, the value T takes varies based on the outcome of a random experiment. Further, the probability of the outcomes of the experiment determine the probability of the value that T takes. We call such a variable a **Random Variable**.

#### Example: Red die, blue die (cont)

#### **Creating a Random Variable**

**def: Random Variable**: A variable which takes numeric values based on the outcome of a random experiment. We use capital letters for the variables and lower case letters when we need to generically refer to values it may take after the outcome of the random experiment is observed.

Since T takes values based on the outcome of our experiment and our outcomes have probabilities, then the value of T inherits the probability. We use a capital letter for the variable and a lower case letter for the specific value it takes after a random experiment.

t	2	3	4	5	6	7	8	9	10	11	12
#Outcomes	1	2	3	4	5	6	5	4	3	2	1
P(T=t)	1/36	2/36	3/36	4/36	5/36	6/36	5/36	4/36	3/36	2/36	1/36

### Example: Red die, blue die (cont)

### Axioms Example

### Sample Space Example

#### **Using Random Variables**

Because a random variable inherits its probability from the sample space, it inherits the probability rules from the sample space too. For instance, we can write things like:

$$P(T < 4) = 1 - P(T \ge 4) = 1 - P(T > 5)$$

or

$$P(2 < T < 6) = P(T = 3) + P(T = 4) + P(T = 5)$$

Again, this all goes back to our *probability axioms* but now that T is a numeric random variable we don't have to go back and think about the events/sets involved. The rules stay the same.

### Example: Red die, blue die (cont)

### Axioms Example

#### **More Random Variables?**

Suppose that we are interested in other aspects of the random event. We could define other random variables:

# Sample Space Example

- $X_R$  as the number of dots facing up on the red die,
- $X_B$  as the number of dots facing up on the blue die,
- $Y=X_R-X_B$ , and
- $Z = \max$

$$X_R, X_B$$

.

These are all random variables, too, though they may not be useful unless we are very concerned with the outcomes they represent. Notice that a single random event will result in all of these random variables taking a value at the same time.

### Example: Red die, blue die (cont)

### Axioms Example

#### **More Random Variables?**

Notice that a single random event will result in all of these random variables taking a value at the same time.

# Sample Space Example

So if we role a (3, 2) then we get:

• 
$$T = 3 + 2 = 5$$

• 
$$X_R = 3$$

• 
$$X_B = 2$$

• 
$$Y = 1$$

• 
$$Z = 3$$
.

In other words, it is valid to ask questions involving multiple random variables, such as:

• What is 
$$P(T \le 9, Z = 6)$$
?

• What is 
$$P(T \le 9, Z = 6, Y = 1)$$
?

#### **Example: Deck of Cards**

#### Axioms Example

Goal 1: Find the probability that if you are dealt two cards that you will have a pair

# Sample Space Example

Goal 2: Find the probability that if you are dealt five cards you will have a full house (3 of one rank, 2 of another rank)

#### Deck of Cards

#### **Problems**

- Sample space very large, can we avoid writing it out?
- How do we account for the order the cards are dealt (ignore ordering vs use ordering)?

### Summary

### Axioms Example

#### **Understanding Random Experiments**

## Sample Space Example

- We talked about how random experiments result in outcomes
- That the set of all outcomes is called a **sample space**
- That we can group outcomes together into **events**
- That the likelihood of the outcomes can be measured using probability

#### Summary

#### **Understanding Probability**

- We talked about the rules that a probability must follow to be a valid way of measuring likelihood.
- the simple set of rules (called the **probability axioms**) can be used to show many more complicated rules that must also be true.

### Summary (cont)

#### Axioms Example

# Sample Space Example

#### Summary

#### **Random Variables**

- We gave a definition of random variables.
- We can create random variables to make answering questions about our outcomes easier.
- We can create multiple random variables for the same experiment
- The random variables we create can be used together to answer certain questions.

How Does Partial Information Effect Probabilities?

and

The Monty Hall Problem

## **Conditional Probability**

#### What is it?

#### What is Conditional Probability?

Most of what we've discussed up until this point has assumed that we have a random experiment and the outcome we observe is revealed to us all at once.

But what if we could know *partial information* about the outcome?

For instance, if we roll a pair of dice (one red, one blue) and you are told the total, could you better guess what number was on the red die?

In terms of probability, that information changes the conditions in which a specific outcome's likelihood is being measured - it means that we are now dealing with something called a **conditional probability**.

#### How Does Conditional Probability Work?

What is it?

How it works

#### Here's how it works:

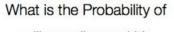
- By learning some detail about the actual outcome, we know that there are only some outcomes that have that detail and some that do not have that detail
- Since we have learned that the actual observed outcome does have that detail, then the outcomes that did not have that detail could not have been the outcome that occured
- and it also means that the only possible outcomes that could have occured are a subset of the sample space
- In other words, we know that some event has occured!

### How Does Conditional Probability Work?

What is it?

Conditional Probability

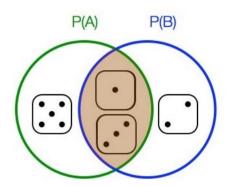
#### How it works



rolling a dice and it's value is less than 4

$$P(B \mid A) = \frac{P(A \cap B)}{P(A)}$$

knowing that the value is an odd number



#### How Does Conditional Probability Work?

What is it?

**Example**: Suppose I deal you two cards. You are interested in predicting if the second card I dealt you is Red.

#### How it works

- 1. What is the probability that the second card is red?
- 2. What is the probability that the first card is red?
- 3. What is the probability that the second card is red *given* the first card is red?
- 4. What is the probability that the second card is red *given* the first card is black?
- 5. What is the probability that the second card is red *given* the second card is red?
- 6. What is the probability that the second card is red *given* the second card is black?

#### **Conditional Probability Notation**

#### **Event Notation**

What is it?

Since the information can be described in terms of events, we use the following notation:

How it works

• P(A|B): probability the outcome is in event A given that the outcome is in event B

#### **Notation**

Since we know that the event is in B, then for it to be in A we must have in  $A \cap B$ . We can actually get values for conditional probabilities if we know the original probabilities:

$$P(A|B) = rac{P(A \cap B)}{P(B)}$$

Which also means that we can write

$$P(A|B) \cdot P(B) = P(A \cap B)$$

### **Conditional Probability Notation**

What is it?

How it works

**Notation** 

#### **Random Variable Notation**

Since random variables are just sets of outcomes too, the we can use a similar notation when dealing with random variables:

• P(X = x | Y = y): the probability the random variable X takes the specific values x given that the random variable Y took the value y.

And similarly we can find the conditional probability using:

$$P(X=x|Y=y)=rac{P(X=x,Y=y)}{P(Y=y)}$$

Which also means that we can write

$$P(X=x|Y=y)\cdot P(Y=y) = P(X=x,Y=y)$$

#### Independence

A related concept is **independence**:

What is it?

**Independent Events** 

How it works

Events A and B are said to be **independent** if knowing event B has occured does not impact the probability that A will occur. In other words,

**Notation** 

$$P(A|B) = P(A)$$

#### Independence

#### **Independent Random Variables**

Random variables X and Y are said to be **independent** if knowing the values taken by Y has not impact on the probabilities associated with values taken by X. In other words, for any value of x and y,

$$P(X = x | Y = y) = P(X = x)$$

#### **Bayes Theorem**

What is it?

Wildly important and useful way of connecting conditional probabilities

#### How it works

**Bayes Theorem** 

For events A and B,

**Notation** 

 $P(B|A) = rac{P(A|B) \cdot P(B)}{P(A)}$ 

#### Independence

For Random Variables X and Y,

Bayes' Theorem

$$P(Y=y|X=x)=rac{P(X=x|Y=y)\cdot P(Y=y)}{P(X=x)}$$

#### Example

I flip a coin. If I flip heads, I roll a six-sided die. If I flip tails, I roll a 10 sided die. I tell you the number on the die. You tell me the flip of the coin.

## Wrap Up Example

Who killed the king?

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem

Example



King Joffrey dies during his wedding feast. There are multiple scenarios of his cause of death, and poison is suspected as a manner of death. The poison might have been given by any of his enemies

The question is that who killed the king?

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem



### Example

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem

Example



#### Suppose that

- There is 70% chance that he was poisoned.
- Apart form that, he could have died of internal bleeding by 15% chance.
- He could been poisoned and also had internal bleeding by 7% chance.

What is the probability that he died of poisoning OR internal bleeding?

The maester in charge of investigating Joffrey?s death figured out that the poisoned used was made by either *Tyrion* 

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem



#### Example

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem

Example

or lord Baelish.



What is it?

How it works

**Notation** 

Independence

Bayes' Theorem

Example



Tyrion was accused to poison the king by 60%.

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem





Lord Baelish was accused to poison the king by 45%.

### The quesion is that given that he was poisoned

who was more likely to poison the king?

What is it?

How it works

**Notation** 

Independence

Bayes' Theorem

Example

Tyrion:

$$P(T|P) = \frac{P(T \cap P)}{P(P)} = \frac{60\%}{70\%} = 0.85$$

Lord Baelish:

$$P(LB|P) = rac{P(LB \cap P)}{P(P)} = rac{45\%}{70\%} = 0.64$$

