

# Forensic Science and Statistics: Version 1.0.0

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# Chapter 1

## Welcome

**This is a draft version of a work in progress. It is NOT intended for circulation.**

This book introduces statistical methods that are of use in forensic science. In some cases the methods are currently used. In others the methods are described but not generally in use, although we hope that eventually they will be!

We assume that the reader has a basic understanding of mathematics, but no prior knowledge of statistics is required. We will provide a modest description of forensic methods, but as many of these are highly nuanced we do not attempt to provide a deep treatment of forensic methods here. However, when possible we will try to provide references to more information.

There are portions of this book that borrow from the treatment of statistics given in the online materials:

Online Statistics Education: A Multimedia Course of Study (<http://onlinestatbook.com/>) Project Leader: David M. Lane, Rice University

This site provides a comprehensive introduction to statistics and is worth referencing when you would like additional information. We deeply appreciate their willingness to allow the use of their work.

### 1.1 An Introductory Example

Consider the following scenario: Suppose there is a late-night break-in at a closed convenience store. No one is present to see the perpetrator, but there is low-quality CCTV footage. Unfortunately, the *only* thing that can be discerned from the video is the color of the perpetrator's hair.

Now consider two possible versions of what comes next:

**Version 1:** The video shows that the perpetrator has *brown* hair. The day after the robbery, police see a person with brown hair walking down the street. The person is arrested on suspicion of committing the break-in.

**Version 2:** The video shows that the perpetrator has *green* hair. The day after the robbery, police see a person with green hair walking down the street. The person is arrested on suspicion of committing the break-in.

For the sake of this simplified example, in both versions we assume that the color of the hair is the **only** reason the police suspect the person who is arrested.

**Question:** In which version do you think it is more likely that the police have arrested the person who committed the break-in?

Answering the question requires considering the likelihood that the police have the correct person in both versions and deciding which is greater. In most communities people with brown hair outnumber people with green hair, so let's assume that is the case here. Intuitively, it seems reasonable to expect that the likelihood of a correct arrest in Version 2 is greater than that in Version 1: Brown-haired people are relatively common, so in Version 1 the likelihood of arresting the correct brown-haired person is probably low. On the other hand, green-haired people are relatively rare, so in Version 2 the likelihood of arresting the correct person is higher.

Two important points:

- 1) Although the likelihood of the correct arrest is greater in Version 2 than in Version 1 does not mean that the likelihood in Version 2 is high. It is possible that there are more brown-haired people than green-haired people while still having a lot of green-haired people, making the correct arrest unlikely in either case.
- 2) While the above discussion is interesting, the word “likelihood” is imprecise. Going forward we will develop “probability” which will allow us to quantify the notion of likelihood, so that we will be able to specify (for instance) how much more likely a correct arrest is in Version 1 than in Version 2.

## Chapter 2

# Introduction to Probability

### 2.1 The Previous Example

The example from the previous section involves the “likelihood” of arresting the correct person based on their hair color. Here we begin the development of “probability” which allows us to quantify the notion of likelihood.

Let’s start by providing some additional context for our example: Suppose that the crime described takes place on an island of 1000 people. The number of these people with each hair color is given in the table below.

Color	Count	Table 1: Hair Color Counts
Brown	670	
Blond	200	
Red	110	
Green	20	

A **probability** is a number between 0 and 1 that provides a measure of the likelihood that something occurs, with a larger number indicating greater likelihood.

Suppose that we select a person from this population at random. To compute the probability that this person has brown hair, we take the number with brown hair and divide by the number of people:

$$P(\text{Brown hair}) = \frac{670}{1000} = 0.67$$

We use “P(...)” to denote the probability of something. For instance in this case  $P(\text{Brown hair})$  = “probability of brown hair”.

### 2.1.1 Sample Questions

Sprinkled throughout this book are Sample Questions, which also play the role of examples. In the online version, the solutions to these questions are hidden. Before looking at the solution (by hovering), we recommend that you try to do them yourself!

1. What is the probability that a randomly selected person has blond hair?

**Answer:**  $P(\text{Blond}) = \frac{200}{1000} = 0.20$ .

2. What is the probability that a randomly selected person has red hair?

**Answer:**  $P(\text{Red}) = \frac{110}{1000} = 0.11$ .

These examples are fine but do not answer the question from the previous section. In Version 1, we have a brown-haired person on the CCTV and one of the 670 brown-haired people randomly arrested. Thus the probability that the correct person is arrested is

$$P(\text{correct arrest}) = \frac{1}{670} = 0.00149$$

In Version 2 of the example we have a green-haired person on CCTV, so this time the probability that the correct person is arrested is

$$P(\text{correct arrest}) = \frac{1}{20} = 0.05$$

Dividing the two probabilities

$$\frac{0.05}{0.00149} = 33.5$$

we see that a correct arrest in Version 2 is 33.5 times as likely as in Version 1. However, even in Version 2 there is only a 0.05 probability that the police have the correct person!

### 2.1.2 Sample Questions

1. Suppose CCTV shows a red-haired person committed the robbery, and that a random red-haired person is arrested. What is the probability of a correct arrest in this instance?

**Answer:**  $P(\text{correct arrest}) = \frac{1}{110} = 0.00909$ .

2. Determine the number of times greater the likelihood of a correct arrest if the culprit has red hair vs having blond hair.

**Answer:** In the case of a blond-haired person, we have  $P(\text{correct arrest}) = \frac{1}{200} = 0.005$ . Therefore the ratio of probabilities is  $\frac{0.00909}{0.005} = 1.81818$  so a correct



arrest for a red-haired culprit is 1.81818 times as likely as for a blond-haired culprit.

## 2.2 Exercises

1. Put exercises here



## Chapter 3

# Joint Probability

### 3.1 Blood Types

Blood plays a role in many forensic science applications, with the “blood type” determined by a combination of two different blood group systems:

*Blood types are inherited and represent contributions from both parents. As of 2019, a total of 41 human blood group systems are recognized by the International Society of Blood Transfusion (ISBT). The two most important blood group systems are ABO and Rh; they determine someone’s blood type (A, B, AB, and O, with +, – or null denoting RhD status) for suitability in blood transfusion.*

Source: [https://en.wikipedia.org/wiki/Blood\\_type](https://en.wikipedia.org/wiki/Blood_type)

Suppose that we have a collection of 1000 people, with each classified based on both ABO (A, B, AB, or O) and Rh (+ or –). The number of people of each combination of ABO and Rh class (the “blood type”) is shown in the table below. (This table is based on the distribution of blood types in Canada, reported at <https://www.blood.ca/en/blood/donating-blood/whats-my-blood-type>)

	+	–	
A	360	60	Counts by ABO and Rh groups
B	76	14	
AB	25	5	
O	390	70	

Suppose that one of these people is selected at random and note that person’s groups. There are various possibilities for outcomes, such as the person selected has ABO group O and Rh group +. The likelihood that this combination occurs is the **joint probability** of the two classifications, ABO and Rh groupings.

There are 390 out of the 1000 people are O and +, so the probability of randomly selecting such a person is

$$P(\text{O and } +) = \frac{390}{1000} = 0.39$$

Here  $P(\text{O and } +)$  stands for the probability that the person selected is ABO group O **and** Rh group +. A sample of other probabilities that come from the table are

$$\begin{aligned} P(\text{AB and } -) &= \frac{5}{1000} = 0.005 \\ P(\text{B and } +) &= \frac{76}{1000} = 0.076 \\ P(\text{B and } -) &= \frac{14}{1000} = 0.014 \end{aligned}$$

We also can combine table entries to compute other probabilities. For instance, there are a total of  $76 + 14 = 90$  people with blood group B, so the probability of randomly selecting a person with blood group B is

$$P(\text{B}) = \frac{76 + 14}{1000} = \frac{90}{1000} = 0.09$$

As another example, the number of people with Rh group + is  $360 + 76 + 25 + 390 = 851$ . Thus the probability of randomly selecting a person with Rh group + is

$$P(+) = \frac{360 + 76 + 25 + 390}{1000} = \frac{851}{1000} = 0.851$$

### 3.1.1 Sample Questions

1. Find the probability that a randomly selected person has blood groups A and −.

**Answer:**  $P(\text{A and } -) = \frac{60}{1000} = 0.06$ .

2. Find the probability that a randomly selected person has ABO group AB.

**Answer:** There are  $25 + 5 = 30$  people that have ABO group AB, so that  $P(\text{AB}) = \frac{30}{1000} = 0.03$ .

3. Find the probability that a randomly selected person has ABO group A and ABO group B and Rh group +.

**Answer:** There is no person that has ABO group A *and* ABO group B, so  $P(\text{A and O and } +) = \frac{0}{1000} = 0$ .

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## 3.2 And vs Or

Above we saw that  $P(B \text{ and } +) = 0.076$ . Suppose we instead would like to know  $P(B \text{ or } +)$ ? That is, we want to know the probability that a randomly selected person is either ABO group B **or** Rh group +. Note that this includes those people who are both ABO group B **and** Rh group +.

From our table we see that the total number of people satisfying one or both conditions is  $360 + 76 + 24 + 390 + 14 = 864$  so that

$$P(B \text{ or } +) = \frac{864}{1000} = 0.864$$

Other combinations are also possible. For instance, number of people with ABO group A blood or ABO group B blood is

$$\underbrace{(360 + 60)}_{\text{ABO group A}} + \underbrace{(76 + 14)}_{\text{ABO group B}} = 510$$

Therefore we have

$$P(A \text{ or } B) = \frac{510}{1000} = 0.51$$

### 3.2.1 Sample Questions

1. Find the probability that a randomly selected person has ABO group AB or Rh group – blood.

**Answer:** The number of people who have ABO group AB **or** Rh group – is  $25 + 5 + 60 + 14 + 70 = 174$  so that  $P(AB \text{ or } -) = \frac{174}{1000} = 0.174$

2. Find the probability that a randomly selected person has ABO group O or Rh group + blood.

**Answer:** The number of people who have ABO group O **or** Rh group + is  $390 + 70 + 360 + 76 + 25 = 921$  so that  $P(O \text{ or } +) = \frac{921}{1000} = 0.921$

3. Find the probability that a randomly selected person has ABO group AB or O.

**Answer:** The number of people who have ABO group AB **or** ABO group O is  $25 + 5 + 390 + 70 = 490$  so that  $P(AB \text{ or } O) = \frac{490}{1000} = 0.49$

4. Find the probability that a randomly selected person has ABO group A or AB, and also Rh group –.

**Answer:** We have  $(A \text{ or } AB) \text{ and } (-) = (A \text{ and } -) \text{ or } (AB \text{ and } -)$ . The number of people who are  $(A \text{ and } -)$  is 60, and the number of people who are  $(AB \text{ and } -)$  is 5, so that  $P((A \text{ or } AB) \text{ and } (-)) = \frac{60+5}{1000} = 0.065$ .

### 3.3 Probability Tables

Frequently, tables provide probabilities (or percentages) instead of counts. To make the conversion from counts to probabilities, all we do is divide each table entry by the total. Thus, for our table, we divide each entry by 1000 to arrive at

	F	M
A	0.175	0.235
AB	0.016	0.024
B	0.037	0.063
O	0.202	0.248

Gender and blood type counts

Table 2 entries give the probability of selecting each combination. For instance

$$P(\text{Female and Type O}) = 0.202$$

We can add entries in this table to compute other probabilities.

For example, suppose we want to compute the probability that a randomly selected person has type O blood. Based on the table, this can happen if a person is female and has type O blood, or if a person is male and has type O blood. Because these two groups are disjoint, we can add the probabilities to arrive at

$$P(\text{Type O}) = P(\text{Female and Type O}) + P(\text{Male and Type O}) = 0.202 + 0.248 = 0.45$$

If instead (for instance), we want to compute  $P(\text{Female})$  then we add the entries in the corresponding column, giving us

$$\Pr(\text{Female}) = 0.175 + 0.016 + 0.037 + 0.202 = 0.43$$

### 3.4 Exercises

1. Put exercises here

## Chapter 4

# Conditional Probability

Now suppose that we just focus on the females in the population, which forms a subset of the population. The probability that a randomly selected female has blood type O is an example of a **conditional probability**.

There are 430 females in the population, and 202 of those have type O blood. Hence, the probability that female chosen at random has type O blood is

$$P(\text{Type O} \mid \text{Female}) = \frac{202}{430} = 0.470$$

The vertical bar in the notation is interpreted as *given that*, so that  $P(\text{Type O} \mid \text{Female})$  is read as

The probability of blood type O, given the person is female.

Conditional probabilities arise all the time when evaluating forensic evidence. Other examples of conditional probabilities:

The probability of Type AB, given that the person is male:

$$P(\text{Type AB} \mid \text{Male}) = \frac{24}{570} = 0.042$$

The probability the person is female, given Type B blood:

$$P(\text{Female} \mid \text{Type B}) = \frac{37}{100} = 0.37$$

The probability the person is male, given Type A blood:

$$P(\text{Male} \mid \text{Type A}) = \frac{235}{410} = 0.573$$

#### 4.0.1 Sample Questions

1. What is the probability of blood type AB or B given the person selected is female?

**Answer:**  $P(\text{AB or B} \mid \text{Female}) = (16 + 37)/430 = 0.123$



## Chapter 5

# Probability Rules

Earlier we computed the conditional probability

$$P(\text{Type O} \mid \text{Female}) = \frac{202}{430}$$

The numerator 202 came from the count of Females who have blood Type B, and the denominator 430 is the total number of Females. If we divide the numerator and denominator by 1000, then the quotient is not changed, so that

$$P(\text{Type O} \mid \text{Female}) = \frac{202/1000}{430/1000} = \frac{0.202}{0.43} = \frac{P(\text{Type O and Female})}{P(\text{Female})}$$

This illustrates a general property of probability: If A and B represent possible outcomes, then the probability of A given B is

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

Multiplying on both sides of the above equation by  $P(B)$  gives

$$P(A \text{ and } B) = P(A \mid B)P(B)$$

This is called the **multiplication rule**. Reversing the order of A and B above gives

$$P(B \text{ and } A) = P(B \mid A)P(A)$$

In the blood type example, it is clear that  $P(\text{Female and Type O})$  is the same as  $P(\text{Type O and Female})$ . This is true in general:  $P(A \text{ and } B) = P(B \text{ and } A)$ , it follows that

$$\Pr(A \mid B)\Pr(B) = \Pr(B \mid A)\Pr(A)$$

so that

$$\Pr(A \mid B) = \Pr(B \mid A) \frac{\Pr(A)}{\Pr(B)}$$

This formula is called **Bayes Rule**. Applied to our earlier population, we have

$$\Pr(\text{Type O} \mid \text{Female}) = \Pr(\text{Female} \mid \text{Type O}) \frac{\Pr(\text{Type O})}{\Pr(\text{Female})}$$

Two outcomes A and B are **independent** when  $\Pr(A \mid B) = \Pr(A)$ , which implies that knowing if B has occurred has no effect on the probability of A.

## Chapter 6

# Counterintuitive Applications

### 6.1 Birthday Paradox

How many people are needed in a classroom so that the probability of them sharing a birthday is at least  $\frac{1}{2}$ ? Intuitively, one could claim that 183 people in the group would imply that the probability is greater than  $\frac{1}{2}$ , since  $\frac{183}{365} = 0.501$ . However, intuition does not correctly solve this problem.

To begin, there are some assumptions that need to be made. For simplicity, we will ignore leap years and assume that all 365 birthdays have an equal probability of occurring.

We want to compute  $P(\beta)$ , the probability that at least 2 people share the same birthday. Recall from the previous chapter that  $P(\beta) = 1 - P(\beta')$ . In this case, it is much simpler to calculate the probability that no one in the group shares the same birthday with someone else.

Let's start with the simple example involving only two people. The probability that they do not share the same birthday is

$$P(\beta') = \left(\frac{365}{365}\right) \left(\frac{364}{365}\right) = 0.997$$

$$1 - P(\beta') = 0.003$$

Extending this result to a group of five people:

$$P(\beta') = \left(\frac{365}{365}\right) \left(\frac{364}{365}\right) \left(\frac{363}{365}\right) \left(\frac{362}{365}\right) \left(\frac{361}{365}\right)$$

$$= \frac{365!}{365^5(365-5)!} = 0.973$$

$$1 - P(\beta') = 0.027$$

This formula can be applied to any number of group size, so that in the general case of  $n$  people, the probability that at least two share the same birthday is

$$P(\beta) = 1 - \frac{365!}{365^n(365-n)!}$$

Finally, to solve the original question: How many people are needed in a classroom so that the probability of them sharing a birthday is at least  $\frac{1}{2}$ ? Iteratively increasing the group size from five, it is clear to see that at a group size of 23,  $P(\beta) > \frac{1}{2}$

$$P(\beta) = 1 - P(\beta') = 1 - \frac{365!}{365^{23}(365-23)!} = 0.507$$

## 6.2 Monty Hall Problem

The original problem was popularized as a letter by Craig F. Whitaker sent to Marilyn vos Savant's column in Parade magazine in 1990. You can read more about the article here: <https://web.archive.org/web/20130121183432/http://marilynvossavant.com/game-show-problem/>

Suppose you're on a game show, and you're given the choice of three doors. Behind one door is a car, behind the others, goats. You pick a door, say #1, and the host, who knows what's behind the doors, opens another door, say #3, which has a goat. He says to you, "Do you want to pick door #2?" Is it to your advantage to switch your choice of doors?

## Chapter 7

# Graphing Distributions: Qualitative Variables

Note: Portions below modeled after content from *Online Statistics Education: A Multimedia Course of Study* (<http://onlinestatbook.com/>) Project Leader: David M. Lane, Rice University

### 7.1 Introduction

When Apple Computer introduced the iMac computer in August 1998, the company wanted to learn whether the iMac was expanding Apple's market share. Was the iMac just attracting previous Macintosh owners? Or was it purchased by newcomers to the computer market and by previous Windows users who were switching over? To find out, 500 iMac customers were interviewed. Each customer was categorized as a previous Macintosh owner, a previous Windows owner, or a new computer purchaser.

This section examines graphical methods for displaying the results of the interviews. We'll learn some general lessons about how to graph data that fall into a small number of categories. A later section will consider how to graph numerical data in which each observation is represented by a number in some range. The key point about the qualitative data that occupy us in the present section is that they do not come with a pre-established ordering (the way numbers are ordered). For example, there is no natural sense in which the category of previous Windows users comes before or after the category of previous Macintosh users. This situation may be contrasted with quantitative data, such as a person's weight. People of one weight are naturally ordered with respect to people of a different weight.

## 7.2 Frequency Tables

All of the graphical methods shown in this section are derived from frequency tables. Table 1 shows a frequency table for the results of the iMac study; it shows the frequencies of the various response categories. It also shows the relative frequencies, which are the proportion of responses in each category. For example, the relative frequency for “none” is  $85/500 = 0.17$ .

Table 1. Frequency Table for the iMac Data. Previous Ownership Frequency  
Relative Frequency None 85 0.17 Windows 60 0.12 Macintosh 355 0.71 Total  
500 1.00

## 7.3 Pie Charts

The pie chart in Figure 1 shows the results of the iMac study. In a pie chart, each category is represented by a slice of the pie. The area of the slice is proportional to the percentage of responses in the category. This is simply the relative frequency multiplied by 100. Although most iMac purchasers were Macintosh owners, Apple was encouraged by the 12% of purchasers who were former Windows users, and by the 17% of purchasers who were buying a computer for the first time.

Figure 1. Pie chart of iMac purchases illustrating frequencies of previous computer ownership.

Pie charts are effective for displaying the relative frequencies of a small number of categories. They are not recommended, however, when you have a large number of categories. Pie charts can also be confusing when they are used to compare the outcomes of two different surveys or experiments. In an influential book on the use of graphs, Edward Tufte asserted, “The only worse design than a pie chart is several of them.”

Here is another important point about pie charts. If they are based on a small number of observations, it can be misleading to label the pie slices with percentages. For example, if just 5 people had been interviewed by Apple Computers, and 3 were former Windows users, it would be misleading to display a pie chart with the Windows slice showing 60%. With so few people interviewed, such a large percentage of Windows users might easily have occurred since chance can cause large errors with small samples. In this case, it is better to alert the user of the pie chart to the actual numbers involved. The slices should therefore be labeled with the actual frequencies observed (e.g., 3) instead of with percentages.

## 7.4 Bar Charts

Bar charts can also be used to represent frequencies of different categories. A bar chart of the iMac purchases is shown in Figure 2. Frequencies are shown on

the Y-axis and the type of computer previously owned is shown on the X-axis. Typically, the Y-axis shows the number of observations in each category rather than the percentage of observations as is typical in pie charts.

Figure 2. Bar chart of iMac purchases as a function of previous computer ownership.

## 7.5 Comparing Distributions

Often we need to compare the results of different surveys, or of different conditions within the same overall survey. In this case, we are comparing the “distributions” of responses between the surveys or conditions. Bar charts are often excellent for illustrating differences between two distributions. Figure 3 shows the number of people playing card games at the Yahoo website on a Sunday and on a Wednesday in the Spring of 2001. We see that there were more players overall on Wednesday compared to Sunday. The number of people playing Pinochle was nonetheless the same on these two days. In contrast, there were about twice as many people playing hearts on Wednesday as on Sunday. Facts like these emerge clearly from a well-designed bar chart.

Figure 3. A bar chart of the number of people playing different card games on Sunday and Wednesday.

The bars in Figure 3 are oriented horizontally rather than vertically. The horizontal format is useful when you have many categories because there is more room for the category labels. We’ll have more to say about bar charts when we consider numerical quantities later in the section Bar Charts.