

# My probability and statistics exercises

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# Contents

<b>1</b>	<b>Introduction to Probability</b>	<b>2</b>
1.1	The History of Probability . . . . .	2
1.2	Interpretations of Probability . . . . .	2
1.3	Experiments and Events . . . . .	2
1.4	Set Theory . . . . .	2
1.5	The Definition of Probability . . . . .	2
1.6	Finite Sample Spaces . . . . .	5
1.7	Counting Methods . . . . .	6
1.8	Combinatorial Methods . . . . .	7
1.9	Multinomial Coefficients . . . . .	12
1.10	The Probability of a Union of Events . . . . .	12
<b>2</b>	<b>Conditional Probability</b>	<b>13</b>
2.1	Definition of Conditional Probability . . . . .	13
2.2	Independent Events . . . . .	16
2.3	Bayes' Theorem . . . . .	18
2.4	The Gambler's Ruin Problem . . . . .	20
<b>3</b>	<b>Random Variables and Distributions</b>	<b>22</b>
3.1	Random Variables and Discrete Distributions . . . . .	22
3.2	Continuous Distributions . . . . .	22
3.3	The Cumulative Distribution Function . . . . .	23
3.4	Bivariate Distributions . . . . .	23
3.5	Marginal Distributions . . . . .	23
3.6	Conditional Distributions . . . . .	23
3.6.1	. . . . .	23
3.7	Multivariate Distributions . . . . .	23

# Chapter 1

## Introduction to Probability

### 1.1 The History of Probability

### 1.2 Interpretations of Probability

### 1.3 Experiments and Events

### 1.4 Set Theory

*Exercises in this section (or exercises similar to them) are handled in the set theory course*

### 1.5 The Definition of Probability

1	$2/5$
2	$0.7$
3a	$1/2$
3b	$1/6$
3c	$3/8$
4	$0.6$
5	$0.4$
6	$0.5$
8	$30$
11a	$1 - \pi/4$
11b	$0.75$
11c	$2/3$
11d	$0$
14a	$0.38, 0.16$
14b	$0.04$

A little notation, related to 6:

$$Pr(A) = 0.5$$

$$Pr(B) = 0.2$$

$$Pr(A \cap B) = 0.1$$

$$Pr(A \cup B) = 0.6$$

$$Pr((A \cup B) \cap (A \cap B)^c) = P(A \cup B) - P((A \cup B) \cap (A \cap B)) = P(A \cup B) - P(A \cap B) = 0.5$$

### 1.5.7

If  $Pr(A) = 0.4$  and  $Pr(B) = 0.7$ , then we follow that the maximum  $Pr(A \cap B)$  is attained if  $A \subset B$ , in which case  $Pr(A \cap B) = Pr(A) = 0.4$ . The minimum is obtained if  $A \cup B = S$ , in which case  $Pr(A \cap B) = 0.1$

### 1.5.9

The event that exactly one of the events occurs can be expressed as

$$(A \cap B^c) \cup (A^c \cap B)$$

which comes from either the definition of xor, common sense or something else, depending on your preferences. Thus we follow that

$$\begin{aligned} Pr((A \cap B^c) \cup (A^c \cap B)) &= Pr(A \cap B^c) + Pr(A^c \cap B) - Pr((A \cap B^c) \cap (A^c \cap B)) = \\ &= Pr(A \cap B^c) + Pr(A^c \cap B) - Pr((A \cap A^c) \cap (B^c \cap B)) = \\ &= Pr(A \cap B^c) + Pr(A^c \cap B) = Pr(A) - Pr(A \cap B) + Pr(B) - Pr(B \cap A) = \\ &= Pr(A) - Pr(A \cap B) + Pr(B) - Pr(A \cap B) = Pr(A) + Pr(B) - 2Pr(A \cap B) \end{aligned}$$

as desired (rules used in this derivation: association of unions,  $A \cap A^c = \emptyset$  and other trivial stuff)

### 1.5.10

$$Pr(A \cap B^c) = Pr(A) - Pr(A \cap B)$$

$$Pr(A \cap B^c) + Pr(A \cap B) = Pr(A)$$

as desired.

**1.5.12**

Suppose that  $n > m \in N$ . Then we follow that by definition

$$B_m \subseteq A_m$$

and

$$B_n \subseteq A_m^c$$

thus we follow that

$$B_m \cap B_n \subseteq A_m \cap A_m^c = \emptyset$$

thus

$$B_m \cap B_n = \emptyset$$

therefore we conclude that  $B_1, B_2, \dots$  are disjoint sets. Thus we follow that

$$Pr(\bigcup_{i=1}^n B_i) = \sum_{i=1}^n Pr(B_i)$$

For  $n = 2$  we've got that

$$B_1 \cup B_2 = A_1 \cup (A_1^c \cap A_2) = (A_1 \cup A_1^c) \cap (A_1 \cup A_2) = A_1 \cup A_2$$

and by induction we can follow that

$$\bigcup_{i=1}^n B_i = \bigcup_{i=1}^n A_i$$

thus

$$Pr(\bigcup_{i=1}^n B_i) = \sum_{i=1}^n Pr(B_i)$$

implies that

$$Pr(\bigcup_{i=1}^n A_i) = \sum_{i=1}^n Pr(B_i)$$

for  $n \in N$ . Given that  $n$  is arbitrary, we can follow that

$$Pr(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} Pr(B_i)$$

as desired.

**1.5.13**

First equation follow from induction on the result that

$$Pr(A \cup B) \leq Pr(A) + Pr(B)$$

the second equation follows from the first equation, DeMorgan laws and induction on the form

$$Pr(A \cap B) = Pr((A^c \cup B^c)^c) = 1 - Pr(A^c \cup B^c) \geq 1 - (Pr(A^c) + Pr(B^c))$$

**1.5.14**

$$Pr(A) = 0.34$$

$$Pr(B) = 0.12$$

$$Pr(O) = 0.5$$

$$Pr(AB) = 1 - 0.34 - 0.12 - 0.5 = 0.04$$

$$Pr(a - A) = 0.34 + 0.04 = 0.38$$

$$Pr(a - B) = 0.12 + 0.04 = 0.16$$

**1.6 Finite Sample Spaces**

1	1/2
2	1/2
3	2/3
4	1/7
5	4/7
6	1/4
8b	1/4

**1.6.7**

The possible genotypes are  $Aa$  and  $aa$  with probabilities  $1/2$  and  $1/2$  respectively

**1.6.8a**

The sample space of the experiment is  $\{heads, tails\} \times \{1, 2, 3, 4, 5, 6\}$ ,

## 1.7 Counting Methods

1	14
2	9000
3	120
4	24
5	5/18
6	5/324
7	0.014731
8	360 / 2401
9	1 / 20
10a	r/100
10b	r/100
10c	r/100

### 1.7.11

$$s(n) = \frac{1}{2} \log(2\pi) + (n + \frac{1}{2}) \log n - n \approx \log n!$$

$$\log n! - \log(n-m)! = \log \frac{n!}{(n-m)!}$$

$$\begin{aligned} s(n) - s(n-m) &= \frac{1}{2} \log(2\pi) + (n + \frac{1}{2}) \log n - n - (\frac{1}{2} \log(2\pi) + ((n-m) + \frac{1}{2}) \log(n-m) - (n-m)) = \\ &= (n + \frac{1}{2}) \log n - n - ((n-m) + \frac{1}{2}) \log(n-m) + (n-m) = \\ &= (n + \frac{1}{2}) \log n - ((n-m) + \frac{1}{2}) \log(n-m) - m \approx \log \frac{n!}{(n-m)!} \end{aligned}$$

$$P(n, m) = \frac{n!}{(n-m)!} = \exp(s(n) - s(n-m))$$

## 1.8 Combinatorial Methods

1	184756
2	latter
3	equal
4	1 / 10626
5	-
6	2/n
7	(n - k - 1)/C(n, k)
8	(n - k)/C(n, k)
9	(n + 1)/C(2n, n)
10	15/92 $\approx$ 0.16304
11	1/75 $\approx$ 0.01333
12	69/119 $\approx$ 0.57983
13	173/1518 $\approx$ 0.114
14	-
15	-
16a	48/175 $\approx$ 0.27429
16b	$2^{50}/C(100, 50) \approx 0$
17	$4C(13, 4)/C(52, 4) = 44/4165 \approx 0.0105$
18	$C(20, 2)^5/C(100, 10) \approx 0.0143$
19	-
20	-
21	$C(365 + 7 - 1, 7)$
22	-

### 1.8.5

*Prove that*

$$\frac{\prod_{4155 \leq i \leq 4251} i}{\prod_{2 \leq i \leq 97} i}$$

*is an integer*

$$\begin{aligned} & \frac{\prod_{4155 \leq i \leq 4251} i}{\prod_{2 \leq i \leq 97} i} = \frac{\prod_{4155 \leq i \leq 4251} i}{\prod_{1 \leq i \leq 97} i} = \\ &= \frac{\prod_{4155 \leq i \leq 4251} i}{97!} = \frac{4251!}{4154!97!} = \frac{4251!}{4154!(4251 - 4174)!} = C(4251, 4154) \end{aligned}$$

and binomial coefficients are integers (pretty sure that we can follow that by induction in some more advanced course).



**1.8.10**

There are total of  $C(24, 10)$  possible subsets of length 10 in the space of 24. We follow that there are  $C(22, 8)$  ways to pick 8 normal bulbs, which is what required to pick 2 defective bulbs. Therefore the probability is

$$\frac{C(22, 8)}{C(24, 10)} = 15/92 \approx 0.16304...$$

**1.8.12**

Using the same logic as in 1.8.10, there is a possibility  $\frac{C(33, 8)}{C(35, 10)}$  that same two guys will be in the first team, and probability of  $\frac{C(33, 23)}{C(35, 10)}$  that they'll be in the other team. Thus the total probability is the sum of two.

**1.8.14**

*Prove that for all positive integers  $n, k$  such that  $n \geq k$*

$$C(n, k) + C(n, k - 1) = C(n + 1, k)$$

$$\begin{aligned} C(n, k) + C(n, k - 1) &= \frac{n!}{(n - k)!k!} + \frac{n!}{(n - k + 1)!(k - 1)!} = \\ &= \frac{n!}{k(n - k)!(k - 1)!} + \frac{n!}{(n - k + 1)(n - k)!(k - 1)!} = \\ &= \frac{(n - k + 1)n!}{k(n - k + 1)(n - k)!(k - 1)!} + \frac{kn!}{k(n - k + 1)(n - k)!(k - 1)!} = \\ &= \frac{(n - k + 1)n! + kn!}{k(n - k + 1)(n - k)!(k - 1)!} = \frac{n!((n - k + 1) + k)}{k(n - k + 1)(n - k)!(k - 1)!} = \\ &= \frac{n!(n + 1)}{k(n - k + 1)(n - k)!(k - 1)!} = \frac{(n + 1)!}{((n + 1) - k)!k!} = C(n + 1, k) \end{aligned}$$

as desired.

**1.8.15**

*(a) Prove that*

$$\sum_{i=0}^n C(n, i) = 2^n$$

We can follow that from the fact that there are  $2^n$  subsets of any given finite set, which means that the number of subsets of different lengths sums up to  $2^n$ .

Another way to do this is to use binomial theorem:

$$(x + y)^n = \sum_{i=0}^n C(n, i)x^i y^{n-i}$$

thus if we substitute  $x$  and  $y$  for 1, we get

$$(1 + 1)^n = \sum_{i=0}^n C(n, i)1^i 1^{n-i}$$

$$2^n = \sum_{i=0}^n C(n, i)$$

(b) Prove that

$$\sum_{i=0}^n (-1)^i C(n, i) = 0$$

I'm sure that there is a neat explanation for this one as well, but using the binomial theorem once again, but now substituting 1 for  $x$  and  $-1$  for  $y$  we get

$$(1 - 1)^n = \sum_{i=0}^n C(n, i)1^i (-1)^{n-i}$$

$$\sum_{i=0}^n C(n, i)1^i (-1)^{n-i} = 0$$

we can follow through the even-odd argument that  $1^i (-1)^{n-i} = (-1)^i$ , but I'll skip it.

### 1.8.19

(rewording) Prove the formula for unordered sampling with replacement.

This thing is ought to be covered rigorously in a course for discrete maths, combinatorics or something of sorts. Currently there is a better proof at Belcastro's "Discrete mathematics with ducks".

### 1.8.20

Prove the binomial theorem 1.8.2

1.8.2 states that

$$(x + y)^n = \sum_{i=0}^n C(n, i)x^i y^{n-i}$$

Let

$$I = \{n \in \omega : (x + y)^n = \sum_{i=0}^n C(n, i)x^i y^{n-i}\}$$

We follow that

$$(x + y)^0 = C(0, 0)x^0 y^0 = 1$$

Thus  $0 \in I$ . (we can start with a base case of 1 as well for a more clear example, but I like this one more, and it suffices as well).

Now suppose that  $n \in I$ . We follow that

$$(x + y)^n = \sum_{i=0}^n C(n, i)x^i y^{n-i}$$

thus we follow that

$$(x + y)(x + y)^n = (x + y) \left[ \sum_{i=0}^n C(n, i)x^i y^{n-i} \right]$$

Left-hand side is reduced to

$$(x + y)(x + y)^n = (x + y)^{n+1}$$

Right-hand side is obviously a bit trickier, but we can follow

$$\begin{aligned} (x + y) \sum_{i=0}^n C(n, i)x^i y^{n-i} &= \\ &= x \sum_{i=0}^n C(n, i)x^i y^{n-i} + y \sum_{i=0}^n C(n, i)x^i y^{n-i} = \\ &= \sum_{i=0}^n C(n, i)x^{i+1} y^{n-i} + \sum_{i=0}^n C(n, i)x^i y^{n+1-i} = \\ &= \sum_{i=0}^n C(n, i)x^i y^{n+1-i} + \sum_{i=0}^n C(n, i)x^{i+1} y^{n-i} = \\ &= C(n, n)x^{n+1} y^0 + \sum_{i=0}^n C(n, i)x^i y^{n+1-i} + \sum_{i=0}^{n-1} C(n, i)x^{i+1} y^{n-i} = \\ &= x^{n+1} + \sum_{i=0}^n C(n, i)x^i y^{n+1-i} + \sum_{i=0}^{n-1} C(n, i)x^{i+1} y^{n-i} = \end{aligned}$$

$$\begin{aligned}
&= x^{n+1} + \sum_{i=0}^n C(n, i) x^i y^{n+1-i} + x \sum_{i=0}^{n-1} C(n, i) x^i y^{n-i} = \\
&= x^{n+1} + \sum_{i=0}^n C(n, i) x^i y^{n+1-i} + x \sum_{i=1}^n C(n, i-1) x^{i-1} y^{n-(i-1)} = \\
&= x^{n+1} + C(n, 0) x^0 y^{n+1} + \sum_{i=1}^n C(n, i) x^i y^{n+1-i} + \sum_{i=1}^n C(n, i-1) x^i y^{n+1-i} = \\
&= x^{n+1} + y^{n+1} + \sum_{i=1}^n C(n, i) x^i y^{n+1-i} + \sum_{i=1}^n C(n, i-1) x^i y^{n+1-i} = \\
&= x^{n+1} + y^{n+1} + \sum_{i=1}^n (C(n, i) + C(n, i-1)) x^i y^{n+1-i} = \\
&= x^{n+1} + y^{n+1} + \sum_{i=1}^n C(n+1, i) x^i y^{n+1-i} = x^{n+1} + C(n+1, 0) x^0 y^{n+1-0} + \sum_{i=1}^n C(n+1, i) x^i y^{n+1-i} = \\
&= x^{n+1} + \sum_{i=0}^n C(n+1, i) x^i y^{n+1-i} = x^{n+1} y^0 + \sum_{i=0}^n C(n+1, i) x^i y^{n+1-i} = \\
&= C(n+1, n+1) x^{n+1} y^{n+1-(n+1)} + \sum_{i=0}^n C(n+1, i) x^i y^{n+1-i} = \sum_{i=0}^{n+1} C(n+1, i) x^i y^{n+1-i}
\end{aligned}$$

Thus we follow

$$(x + y)^{n+1} = \sum_{i=0}^{n+1} C(n+1, i) x^i y^{n+1-i}$$

or

$$(x + y)^{n^+} = \sum_{i=0}^{n^+} C(n^+, i) x^i y^{n^+-i}$$

which means that  $n \in I \Rightarrow n^+ \in I$ , from which we conclude that  $I = \omega$ , and thus

$$(x + y)^n = \sum_{i=0}^n C(n, i) x^i y^{n-i}$$

for all  $n \in \omega$ , as desired.

### 1.8.22

Skip

## 1.9 Multinomial Coefficients

1	$(21!)/(7! * 7! * 7!)$
2	$50!/(18! * 12! * 12! * 8!)$
3	$300!/(5! * 8! * 287!)$
4	$(3!3!2!)/10! = 1/50400$
5	$M(n, (n_1, \dots, n_6))/6^n$
6	$(7!)/(2 * 6^7)$
7	$M(12, (6, 2, 4)) * M(13, (4, 6, 3))/M(25, (10, 8, 7))$
8	$M(12, (3, 3, 3, 3)) * M(40, (10, 10, 10, 10))/M(52, (13, 13, 13, 13))$
9	$4!/M(52, (13, 13, 13, 13))$
10	$(2! * 3! * 4!)/9!$

## 1.10 The Probability of a Union of Events

1	$\approx 0.11913$
2	85
3	45

### 1.10.1

$$Pr(A_1) = Pr(A_2) = Pr(A_3) = C(4, 2) * C(48, 3)/C(52, 5)$$

$$Pr(A_1 \cup A_2) = Pr(A_1 \cup A_3) = Pr(A_2 \cup A_3) = C(4, 2) * C(48, 3) * C(45, 3)/C(52, 5)^2$$

$$Pr(A_1 \cup A_2 \cup A_3) = 0$$

$$Pr(A_1 \cup A_2 \cup A_3) = 3 * C(4, 2) * C(49, 3)/C(52, 5) - 3C(4, 2) * C(49, 3) * C(46, 3)/C(52, 5)^2$$

TODO later (probably never).

## Chapter 2

# Conditional Probability

### 2.1 Definition of Conditional Probability

1	$Pr(A)/Pr(B)$
2	0
3	$Pr(A)$
4	$1/27 \approx 0.037037$
5	-
6	$2/3$
7	$1/3$
8	$0.6/0.85 \approx 0.706$
9a	$3/4$
9b	$3/5$
10	0.4485884485884486
11	-
12	-
13	$4/9$
14	0.056
15	0.47
16	$5/12$
17	-

$$Pr(A|B) = \frac{Pr(A \cap B)}{Pr(B)}$$

#### 2.1.5

$$\frac{r}{r+b} * \frac{(r+k)}{(r+k)+b} * \frac{(r+2k)}{(r+2k)+b} * \frac{b}{(r+3k)+b}$$

**2.1.6**

Let  $A$  be an event, that we've picked up a card, looked at its side and that the side is green. We can follow that

$$Pr(A) = 1/2$$

Let  $B$  be an event that we've picked up a card, and it's green on both sides. We follow that

$$Pr(B) = 1/3$$

Probability that both  $A$  and  $B$  happened are  $1/3$ . Thus we follow that

$$Pr(B|A) = \frac{Pr(A \cap B)}{Pr(A)} = \frac{1/3}{1/2} = 2/3$$

This makes me think about Monty Hall problem, as those two are (probably) closely related.

**2.1.11**

We want to prove that

$$Pr(A^c|B) = 1 - Pr(A|B)$$

we follow that by

$$Pr(A^c|B) = \frac{Pr(A^c \cap B)}{Pr(B)} = \frac{Pr(B) - Pr(A \cap B)}{Pr(B)} = 1 - \frac{Pr(A \cap B)}{Pr(B)} = 1 - Pr(A|B)$$

where

$$Pr(A^c \cap B) = Pr(B) - Pr(A \cap B)$$

is proven in Theorem 1.5.6. as desired.

**2.1.12**

$$\begin{aligned} Pr(A \cup B|D) &= \frac{Pr((A \cup B) \cap D)}{Pr(D)} = \frac{Pr((A \cap D) \cup (B \cap D))}{Pr(D)} = \\ &= \frac{Pr(A \cap D) + Pr(B \cap D) - Pr(A \cap D \cap B \cap D)}{Pr(D)} = \\ &= \frac{Pr(A \cap D) + Pr(B \cap D) - Pr(A \cap B \cap D)}{Pr(D)} = \\ &= \frac{Pr(A \cap D)}{Pr(D)} + \frac{Pr(B \cap D)}{Pr(D)} - \frac{Pr(A \cap B \cap D)}{Pr(D)} = Pr(A|D) + Pr(B|D) - Pr(A \cap B|D) \end{aligned}$$

every derivation that was done here was either justified by a theorem in section 1.5 or is a property of set operations.

**2.1.17**

We can't have

$$Pr((A|C)|B)$$

on the account that  $A|C$  is not an event, but just a funky notation introduced with the probability function. What this notation gives is just a syntactic sugar.

$$\begin{aligned} Pr(A|C) &= \frac{Pr(A \cap C)}{Pr(C)} = \frac{1}{Pr(C)} Pr(A \cap C) = \frac{1}{Pr(C)} \sum_{j=1}^n Pr(B_j) Pr(A \cap C | B_j) = \\ &= \frac{1}{Pr(C)} \sum_{j=1}^n Pr(B_j) \frac{Pr(A \cap C \cap B_j)}{Pr(B_j)} = \sum_{j=1}^n Pr(B_j) \frac{Pr(A \cap C \cap B_j)}{Pr(B_j) Pr(C)} = \\ &= \sum_{j=1}^n \frac{Pr(A \cap C \cap B_j)}{Pr(C)} = \sum_{j=1}^n \frac{Pr(B_j \cap C) Pr(A \cap C \cap B_j)}{Pr(B_j \cap C) Pr(C)} = \\ &= \sum_{j=1}^n \frac{Pr(B_j \cap C) Pr(A \cap B_j \cap C)}{Pr(C) Pr(B_j \cap C)} = \\ &= \sum_{j=1}^n \frac{Pr(B_j \cap C)}{Pr(C)} * \frac{Pr(A \cap B_j \cap C)}{Pr(B_j \cap C)} = \sum_{j=1}^n Pr(B_j|C) Pr(A|B_j \cap C) \end{aligned}$$

assuming that  $Pr(B_j \cap C), Pr(C) \neq 0$  for all  $1 \leq j \leq n$ .



## 2.2 Independent Events

1	$Pr(A^c)$
2	-
3	-
4	$1/216$
5	$1 - 10^{-6}$
6	$149/5000 = 0.0298$
7a	$23/25 = 0.92$
7b	$20/23 \approx 0.869565$
8	$1/36 \approx 0.0277778$
9	$1/7 \approx 0.142857$
10	$\frac{106}{781} \approx 0.1357234314980794$
11	$67/256 = 0.26171875$
12a	$3/4 = 0.75$
12b	$11/24 \approx 0.4583333333$
13	$0.09135172474836409$
14	$0.09561792499119552$
15	$161$

### 2.2.1

Suppose that  $A$  and  $B$  are independent events. Thus

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

and

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

thus

$$\begin{aligned}
 Pr(A^c|B^c) &= \frac{Pr(A^c \cap B^c)}{Pr(B^c)} = \frac{Pr((A \cup B)^c)}{Pr(B^c)} = \frac{1 - Pr(A \cup B)}{Pr(B^c)} = \\
 &= \frac{1 - (Pr(A) + Pr(B) - Pr(A)Pr(B))}{Pr(B^c)} = \frac{1 - Pr(A) - Pr(B) + Pr(A)Pr(B)}{Pr(B^c)} = \\
 &= \frac{1 - Pr(B) - Pr(A) + Pr(A)Pr(B)}{Pr(B^c)} = \frac{1 - Pr(B)}{Pr(B^c)} + \frac{-Pr(A) + Pr(A)Pr(B)}{Pr(B^c)} = \\
 &= 1 + \frac{Pr(A)(-1 + Pr(B))}{Pr(B^c)} = 1 - \frac{Pr(A)(1 - Pr(B))}{Pr(B^c)} = 1 - Pr(A) \frac{1 - Pr(B)}{Pr(B^c)} = \\
 &= 1 - Pr(A) = Pr(A^c)
 \end{aligned}$$

Same goes for  $Pr(B^c|A^c)$

**2.2.2**

2.2.1 implies that

$$Pr(A^c) = Pr(A^c|B^c)$$

and

$$Pr(B^c) = Pr(B^c|A^c)$$

for the nonzero cases, and if  $Pr(A) = 0$  or  $Pr(B) = 0$ , then the cases are trivial.

**2.2.3**

Suppose that  $A$  is an event and  $Pr(A) = 0$  and  $B$  is another event. We follow that

$$Pr(A \cap B) \leq Pr(A)$$

and thus

$$Pr(A \cap B) = 0$$

as desired.

**2.2.7b**

$$Pr(A|A \cup B) = \frac{Pr(A \cap (A \cup B))}{Pr(A \cup B)} = \frac{Pr(A)}{Pr(A \cup B)}$$

**2.2.9**

Assuming  $1 \leq n \leq \infty$

$$\sum (p_n)^3 = \sum (2^{-n})^3 = \sum 2^{-3n} = \sum (1/8)^n = \frac{1/8}{1 - 1/8} = 1/7$$

**2.2.10**

Let  $A$  be an event that at least 1 child in the family has blue eyes and let  $B$  be an event that at least 3 children have blue eyes. We follow that

$$Pr(B|A) = \frac{Pr(A \cap B)}{Pr(A)}$$

given that  $B \subseteq A$ , we follow that

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

We follow that

$$Pr(A) = 1 - (1 - 1/4)^5 = 781/1024$$

and

$$Pr(B) = \sum_{i \in \{3,4,5\}} C(n, i) 1/4 * C(n, n-i) (1-1/4) = \sum_{i \in \{3,4,5\}} C(n, i) (1/4)^i (3/4)^{5-i} = 53/512$$

thus

$$Pr(B|A) = \frac{Pr(B)}{Pr(A)} = \frac{106}{781} \approx 0.1357234314980794$$

### 2.2.11

If the youngest child in the family has the blue eyes, then we can't say that  $B \subseteq A$ . Given that the probabilities of children having different colored eyes are independent, we follow that we can rewrite this problem as "what's the probability of that the remaining 4 children have at least 2 blue-eyed children among them". This happens to be equal to

$$\sum_{i \in \{2,3,4\}} C(4, i) (1/4)^i (3/4)^{4-i} = 67/256 = 0.26171875$$

*Done with this section; moving on*

## 2.3 Bayes' Theorem

1	-
2	3
3	0.3
4	0.0001899658061548921
5	0.30508474576271183
6a	0.9896907216494846
6b	0.9846153846153847
7a	0, 1/10, 1/5, 3/10, 2/5
8	skip
16	-

### 2.3.1

Suppose that  $S$  can be partitioned into  $B_1, \dots, B_k$ . Suppose also that  $A$  is an event such that  $Pr(A) > 0$  and

$$Pr(B_1|A) < Pr(B_1)$$

and

$$Pr(B_i|A) \leq Pr(B_i)$$

for all  $1 < i \leq k$ . Thus we follow that

$$\sum Pr(B_i|A) < \sum Pr(B_i) = 1$$

thus

$$\begin{aligned} \sum Pr(B_i|A) &< 1 \\ \sum \frac{Pr(B_i \cap A)}{Pr(A)} &< 1 \\ \sum Pr(B_i \cap A) &< Pr(A) \end{aligned}$$

Given that  $B_i$  is a partition of  $S$ , we follow that  $B_i$ 's are disjoint (BTW if several sets are all pairwise disjoint, then all of them are disjoint), therefore we follow that  $B_j \cap A$  is disjoint from  $B_l \cap A$  for all  $1 \leq j, l \leq k$ . Thus

$$\sum Pr(B_i \cap A) = Pr(\bigcup [B_i \cap A]) = Pr(\bigcup [B_i] \cap A) = Pr(S \cap A) = Pr(A) < Pr(A)$$

which is a contradiction.

### 2.3.16

(a)

Suppose that  $D_1$  is independent of  $B$ . That is,

$$Pr(D_1) = Pr(D_1|B) = 0.01$$

Assume that for some  $n$  we've got that

$$Pr(D_n) = 0.01$$

We follow that

$$Pr(D_{n+1}|B) = 0.01$$

If  $B^c$  is true and we know that  $n$ 'th item is normal, then we can follow that

$$Pr(D_{n+1}|D_n^c \cap B^c) = 1/165$$

If  $n$ 'th item is defective, then

$$Pr(D_{n+1}|D_n \cap B^c) = 2/5$$

therefore, because  $D$  and  $D^c$  are partitioning space, we follow that

$$Pr(D_{n+1}|B^c) = Pr(D_n^c) * 1/165 + Pr(D_n) * 2/5 = 0.01$$

thus we now can follow that

$$Pr(D_{n+1}) = 0.1 * 0.7 + 0.01 * 0.3 = 0.1$$

therefore by induction we can conclude that  $Pr(D_n) = 0.01$  for all  $n \in N$

(b)

Let us assume that we've got a typo in the text, and we actually need to compute  $Pr(B|E)$ . From our initial assumptions we follow that

$$Pr(E|B) = 0.99^4 * 0.01^2 = 9.65 * 10^{-5}$$

thus we need to compute

$$Pr(B|E) = \frac{Pr(E|B) * Pr(B)}{Pr(E|B) * Pr(B) + Pr(E|B^c) * Pr(B^c)}$$

thus the only thing that we need to compute is  $Pr(E|B^c)$ . We follow that

$$\begin{aligned} Pr(E|B^c) &= \\ &= Pr(D_1^c \cap D_2^c \cap D_3 \cap D_4 \cap D_5^c \cap D_6^c | B^c) = Pr(D_1^c | B^c) Pr(D_2^c | D_1^c \cap B) Pr(D_3 | D_2^c \cap B) \dots = \\ &= 0.99 * 164/165 * 1/165 * 2/5 * 3/5 * 164/165 = 0.99 * (164/165)^2 * 1/165 * 2/5 * 3/5 = \\ &= 0.001422598347107438 \end{aligned}$$

thus we can now compute the rest and state that

$$Pr(B|E) = 0.11898006688921978 \approx 12\%$$

## 2.4 The Gambler's Ruin Problem

1	-
2	all the same
3	a
4	c
5	198
6	7
7	-

**2.4.1**

Suppose that we've got conditions from Example 2.4.2. Let  $i$  be a natural number such that  $i \leq 98$ . Probability that gambler  $A$ 's gonna win  $i$  dollars before losing  $100 - i$  is

$$a_i = \frac{(3/2)^i - 1}{(3/2)^{100} - 1}$$

we follow that  $a_i$  is an increasing function and thus we can conclude that in order to get the desired conclusion, we need to calculate the case  $i = 98$ . We follow that

$$a_{98} = \frac{(3/2)^{98} - 1}{(3/2)^{100} - 1} \approx 0.444444$$

BTW, it's not a pretty rational number.

**2.4.7**

we follow that

$$f_i = \frac{(1/3)^i - 1}{(1/3)^{i+2} - 1}$$

is the desired function. We want to show that the function is decreasing and  $a_1 < 1/4$ . Simple calculation show that  $a_1 \approx 0.14285714285714282$ . We also follow that

$$f_n - f_{n+1} = \frac{(1/3)^n - 1}{(1/3)^{n+2} - 1} - \frac{(1/3)^{n+1} - 1}{(1/3)^{n+3} - 1}$$

Maxima shows that this thing is equal to

$$-\frac{16 * 3^{n+2}}{\text{something.positive}}$$

which is good enough for me to prove that this thing is always below  $1/4$ , as desired.

*Done with this section*

## Chapter 3

# Random Variables and Distributions

### 3.1 Random Variables and Discrete Distributions

1	$6/11$
2	$1/15$
3	no
4	binomial with 10 and $1/2$
5	skip
6	0.15087890625
7	0.80589565
8	0.13295332343433508
9	$1/2$
10a	$1/120 (x + 1)(8 - x)$
10b	$1/3$
11	harmonics

### 3.2 Continous Distributions

1	$4/9$
2	$31/48, 9/16, 136/243$
3	$1/2, 13/27, 2/27$

*The rest of that damned section is just exercises in trivial calculus. Skipping all this stuff.*

### 3.3 The Cumulative Distribution Function

Exercises in this section are once again exercises in trivial calculus and graphing. Skip

### 3.4 Bivariate Distributions

1	$1/2, 1/4$
2	$0.27, \dots$

Exercises are a bunch of integrals/sums and other borderline trivial stuff. This is practically exercise in Maxima. Skip

### 3.5 Marginal Distributions

1	$f(y) = (d - c)k, f(y) = (b - a)k$
2	$f(y) = \frac{y+1}{30}, f(x) = \frac{2x+3}{15}, \text{ dependent}$

this section is also an exercise in calculus.

### 3.6 Conditional Distributions

#### 3.6.1

$$f(x|y) = \frac{3y^2}{2x^2 - 1^{3/2}}$$

for appropriate values of  $x$  and  $y$ .

The rest is calculus

### 3.7 Multivariate Distributions

1	$1/3, x_3 + 1/3x_1 + 1/3, 0.557762265046662$
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