

Chapter 2 - Probability

CUNY MSDA - IS606 - Homework 2

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Homework

OpenIntro Statistics

Practice: 2.5, 2.7, 2.19, 2.29, 2.43

Graded: 2.6, 2.8, 2.20, 2.30, 2.38, 2.44

2.6 Dice rolls.

(a) getting a sum of 1?

Let's build a probability table

```
# Dice 1 Sample space
dice1 <- c(1,2,3,4,5,6)
Pdice1 <- c(1/6,1/6,1/6,1/6,1/6,1/6)
# Dice 2 sample space
dice2 <- c(1,2,3,4,5,6)
Pdice2 <- c(1/6,1/6,1/6,1/6,1/6,1/6)
# Possible outcomes sums
dsums <- c(2,3,4,5,6,7,8,9,10,11,12)
# Let's calculate the probabilities for each sum
P <- c(1/36, 2/36, 3/36, 4/36, 5/36, 6/36, 5/36, 4/36, 3/36, 2/36, 1/36)
# Let's build the table
dicef <- data.frame(dsums, P)
names(dicef) <- c("Sums", "Probability")
dicef
```

##	Sums	Probability
## 1	2	0.02777778
## 2	3	0.05555556
## 3	4	0.08333333
## 4	5	0.11111111
## 5	6	0.13888889
## 6	7	0.16666667
## 7	8	0.13888889
## 8	9	0.11111111
## 9	10	0.08333333
## 10	11	0.05555556
## 11	12	0.02777778

Based on the possibilities, there's no possible way to obtain a sum of 1, since none of the faces on the dice has a value of zero.

Hence: $P(X+Y=0) = 0$.

(b) getting a sum of 5?

For this, we have different possibilities:

Let's say X represent the first die and Y the second die.

The possible outcomes will be as follows:

$$\text{Outcome 1} = P(X=1) * P(Y=4) = 1/6 * 1/6 = 1/36$$

$$\text{Outcome 2} = P(X=2) * P(Y=3) = 1/6 * 1/6 = 1/36$$

$$\text{Outcome 3} = P(X=3) * P(Y=2) = 1/6 * 1/6 = 1/36$$

$$\text{Outcome 4} = P(X=4) * P(Y=1) = 1/6 * 1/6 = 1/36$$

There are 4 possible ways to obtain a sum of 5.

$$P(X+Y = 5) = \text{Outcome 1} + \text{Outcome 2} + \text{Outcome 3} + \text{Outcome 4}$$

$$P(X+Y = 5) = 1/36 + 1/36 + 1/36 + 1/36$$

$$P(X+Y = 5) = 4/36$$

(c) getting a sum of 12?

For this, we have different possibilities:

Let's say X represent the first die and Y the second die.

The possible outcomes will be as follows:

$$\text{Outcome 1} = P(X=6) * P(Y=6) = 1/6 * 1/6 = 1/36$$

$$P(X+Y = 12) = 1/36$$

2.8 Poverty and language.

Let's define as follows:

A: Americans living below the poverty line.

F: Speak a language other than English (foreign language) at home.

$$P(A) = 14.6\% = 0.146$$

$$P(F) = 20.7\% = 0.207$$

$$P(A \text{ and } F) = 4.2\% = 0.042$$

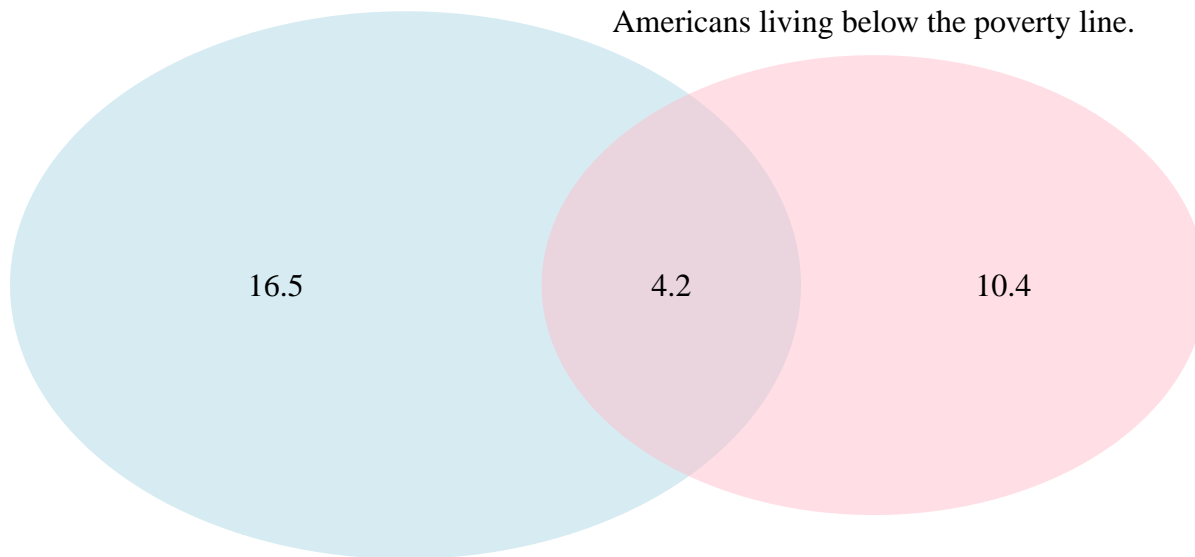
(a) Are living below the poverty line and speaking a foreign language at home disjoint?

No, they are not disjoint since both are happening mutually.

(b) Draw a Venn diagram summarizing the variables and their associated probabilities.

Speak a language other than English (foreign language) at home.

Americans living below the poverty line.



```
## (polygon[GRID.polygon.1], polygon[GRID.polygon.2], polygon[GRID.polygon.3], polygon[GRID.polygon.4],
```

(c) What percent of Americans live below the poverty line and only speak English at home?

$$P(A \text{ and Speak English at Home}) = P(A) - P(A \text{ and F})$$

$$P(A \text{ and Speak English at Home}) = 0.146 - 0.042$$

$$P(A \text{ and Speak English at Home}) = 0.104$$

Answer: The percent of Americans live below the poverty line and only speak English at home is 10.4%

(d) What percent of Americans live below the poverty line or speak a foreign language at home?

$$P(A \text{ or F}) = P(A) + P(F) - P(A \text{ and F})$$

$$P(A \text{ or F}) = 0.146 + 0.207 - 0.042$$

$$P(A \text{ or F}) = 0.311$$

Answer: The percent of Americans live below the poverty line or speak a foreign language at home is 31.1%

(e) What percent of Americans live above the poverty line and only speak English at home?

AC: Complement of A

FC: Complement of C

$$P(AC \text{ and FC}) = 1 - (P(A) + P(F) - P(A \text{ and F}))$$

$$P(\text{AC and FC}) = 1 - 0.311$$

$$P(\text{AC and FC}) = 0.689$$

Answer: The percent of Americans live above the poverty line and only speak English at home is 68.9%

(f) Is the event that someone lives below the poverty line independent of the event that the person speaks a foreign language at home?

Let's build our independence condition:

$$P(A \text{ and } F) = P(A) * P(F)$$

$$0.042 = 0.146 * 0.207$$

$$\text{Since } 0.042 \neq 0.030$$

We conclude that these events are not independent, since the independency multiplication rule is not satisfied.

2.20 Assortative mating.

(a) What is the probability that a randomly chosen male respondent or his partner has blue eyes?

By applying the general addition rule:

$$P(\text{M_Blue or F_Blue}) = P(\text{M_Blue}) + P(\text{F_Blue}) - P(\text{M_Blue and F_Blue})$$

$$P(\text{M_Blue or F_Blue}) = 108/204 + 114/204 - 78/204$$

$$P(\text{M_Blue or F_Blue}) = 0.7059$$

Answer: The probability that a randomly chosen male respondent or his partner has blue eyes is 70.59%.

(b) What is the probability that a randomly chosen male respondent with blue eyes has a partner with blue eyes?

$$P(\text{F_Blue} \mid \text{M_Blue}) = P(\text{F_Blue and M_Blue}) / P(\text{M_Blue})$$

$$P(\text{F_Blue} \mid \text{M_Blue}) = (78/204) / (114 / 204)$$

$$P(\text{F_Blue} \mid \text{M_Blue}) = 0.6842$$

Answer: The probability that a randomly chosen male respondent with blue eyes has a partner with blue eyes is 68.42%

(c) What is the probability that a randomly chosen male respondent with brown eyes has a partner with blue eyes? What about the probability of a randomly chosen male respondent with green eyes having a partner with blue eyes?

$$P(\text{F_Blue} \mid \text{M_Brown}) = P(\text{F_Blue and M_Brown}) / P(\text{M_Brown})$$

$$P(\text{F_Blue} \mid \text{M_Brown}) = (23/204) / (54/204)$$

$$P(\text{F_Blue} \mid \text{M_Brown}) = 0.4259$$

Answer: The probability that a randomly chosen male respondent with brown eyes has a partner with blue eye is 42.59%

$$P(\text{F_Blue} \mid \text{M_Green}) = P(\text{F_Blue and M_Green}) / P(\text{M_Green})$$

$$P(\text{F_Blue} \mid \text{M_Green}) = (13/204) / (36/204)$$

$$P(F_Blue \mid M_Green) = 0.3611$$

Answer: The probability of a randomly chosen male respondent with green eyes having a partner with blue eyes is 36.11%

(d) Does it appear that the eye colors of male respondents and their partners are independent? Explain your reasoning.

Let's build our independence condition:

$$P(M_Blue \text{ and } F_Blue) = P(M_Blue) * P(F_Blue)$$

$$(78/204) = (114/204) * (108/204)$$

Since $0.3824 \neq 0.2958$

We conclude that these events are not independent, since the independence multiplication rule is not satisfied.

2.30 Books on a bookshelf.

(a) Find the probability of drawing a hardcover book first then a paperback fiction book second when drawing without replacement.

$$P(\text{First H and Second being P_fiction}) = P(\text{First H}) * P(\text{Second being P_fiction})$$

$$P(\text{First H and Second being P_fiction}) = (28/95) * (59/94)$$

$$P(\text{First H and Second being P_fiction}) = 0.185$$

Answer: The probability of drawing a hardcover book first then a paperback fiction book second when drawing without replacement is 18.5%

(b) Determine the probability of drawing a fiction book first and then a hardcover book second, when drawing without replacement.

$$P(\text{Fiction and second being H}) = P(\text{Fiction}) * P(\text{second being H})$$

$$P(\text{Fiction and second being H}) = (72/95) * (28/94)$$

$$P(\text{Fiction and second being H}) = 0.2258$$

Answer: The probability of drawing a fiction book first and then a hardcover book second, when drawing without replacement is 22.58%

(c) Calculate the probability of the scenario in part (b), except this time complete the calculations under the scenario where the first book is placed back on the bookcase before randomly drawing the second book.

$$P(\text{Fiction and second being H}) = P(\text{Fiction}) * P(\text{second being H})$$

$$P(\text{Fiction and second being H}) = (72/95) * (28/95)$$

$$P(\text{Fiction and second being H}) = 0.2234$$

Answer: The probability of drawing a fiction book first and then a hardcover book second, when drawing with replacement is 22.34%

(d) The final answers to parts (b) and (c) are very similar. Explain why this is the case.

Answer: In this case the answers are very similar, this is because when the possible events are considerable large, the outcome will not be affected by much when there's no replacement in random drawings.

2.38 Baggage fees.

(a) Build a probability model, compute the average revenue per passenger, and compute the corresponding standard deviation.

```
# Number of bags
bags <- c(0, 1, 2)
# Fees charges for 0 pieces of luggage in dollars
Luggage_0 <- 0
# Fees charges for 1st luggage in dollars
Luggage_1 <- 25
# Fees charges for 2nd luggage in dollars
Luggage_2 <- Luggage_1 + 35

# Baggage fees table
baggage_fees <- c(Luggage_0, Luggage_1, Luggage_2)

# Percentage of passengers that check baggage in decimal form
baggage_percent_per_pax <- c(0.54, 0.34, 0.12)

# Find Expected value for each  $x_i$ 
E_revenue <- baggage_fees * baggage_percent_per_pax

# Find the overall Expected value AKA  $\mu$ 
Ex <- sum(E_revenue)

# Expected Revenue per passenger
Ex

## [1] 15.7

# Create  $\mu$  column
mu <- c(Ex, Ex, Ex)

# Create data frame
baggage <- data.frame(bags, baggage_fees, baggage_percent_per_pax)

# Find The variance  $_i$  of  $x_i$  and  $\mu$ 
baggage_variance <- baggage_fees - Ex

# Calculate the Variance2 and  $P(X=x_i)$ 
baggage_EVariance <- baggage_variance2 * baggage_percent_per_pax

# Create visual representation of the table
baggage <- cbind(baggage, E_revenue, mu, baggage_variance, baggage_variance2, baggage_EVariance)

# Name columns for the baggage data frame
names(baggage) <- c("bags", "x_i", "P(X=x_i)", "E(X_i)", "mu", "Variance", "Variance2", "Variance2*P(X=x_i)")

# View Table
```

```

baggage

##   bags x_i P(X=x_i) E(X_i)   mu Variance Variance^2 Variance^2*P(X=x_i)
## 1    0   0    0.54   0.0 15.7    -15.7    246.49          133.1046
## 2    1  25    0.34   8.5 15.7     9.3     86.49          29.4066
## 3    2  60    0.12   7.2 15.7    44.3    1962.49         235.4988

# Find the overall value for the Variance^2
Variance2 <- sum(baggage_EVariance)

# Print the overall Variance
# Variance2

# Find the standard deviation by calculating the square root of the variance
sd <- Variance2^(1/2)

# Print the standard deviation
sd

## [1] 19.95019

```

(b) About how much revenue should the airline expect for a flight of 120 passengers? With what standard deviation? Note any assumptions you make and if you think they are justified.

```

# Number of bags
bags <- c(0, 1, 2)
# Fees charges for 0 pieces of luggage in dollars
Luggage_0 <- 0
# Fees charges for 1st luggage in dollars
Luggage_1 <- 25
# Fees charges for 2nd luggage in dollars
Luggage_2 <- Luggage_1 + 35

# Baggage fees table
baggage_fees <- c(Luggage_0, Luggage_1, Luggage_2)

# Percentage of passengers that check baggage in decimal form
baggage_percent_per_pax <- c(0.54, 0.34, 0.12)

# Number of passengers
pax <- 120

# Find Expected value for each x_i
E_revenue <- baggage_fees * baggage_percent_per_pax * pax

# Find the overall Expected value AKA mu
Ex <- sum(E_revenue)

# Expected revenue
Ex

## [1] 1884

```

```
# Create mu column
mu <- c(Ex, Ex, Ex)

# Create data frame
baggage <- data.frame(bags, baggage_fees, baggage_percent_per_pax)

# Find The variance_i of x_i and mu
baggage_variance <- baggage_fees - Ex

# Calculate the Variance^2 and P(X=x_i)
baggage_EVariance <- baggage_variance^2 * baggage_percent_per_pax

# Create visual representation of the table
baggage <- cbind(baggage, E_revenue, mu, baggage_variance, baggage_variance^2, baggage_EVariance)

# Name columns for the baggage data frame
names(baggage) <- c("bags", "x_i", "P(X=x_i)", "E(X_i)", "mu", "Variance", "Variance^2", "Variance^2*P(X=x_i)")

# View Table
baggage

##   bags x_i P(X=x_i) E(X_i)   mu Variance Variance^2 Variance^2*P(X=x_i)
## 1    0  0    0.54     0 1884   -1884   3549456         1916706.2
## 2    1 25    0.34   1020 1884   -1859   3455881         1174999.5
## 3    2 60    0.12    864 1884   -1824   3326976         399237.1

# Find the overall value for the Variance^2
Variance2 <- sum(baggage_EVariance)

# Print the overall Variance
#Variance2

# Find the standard deviation by calculating the square root of the variance
sd <- Variance2^(1/2)

# Print the standard deviation
sd

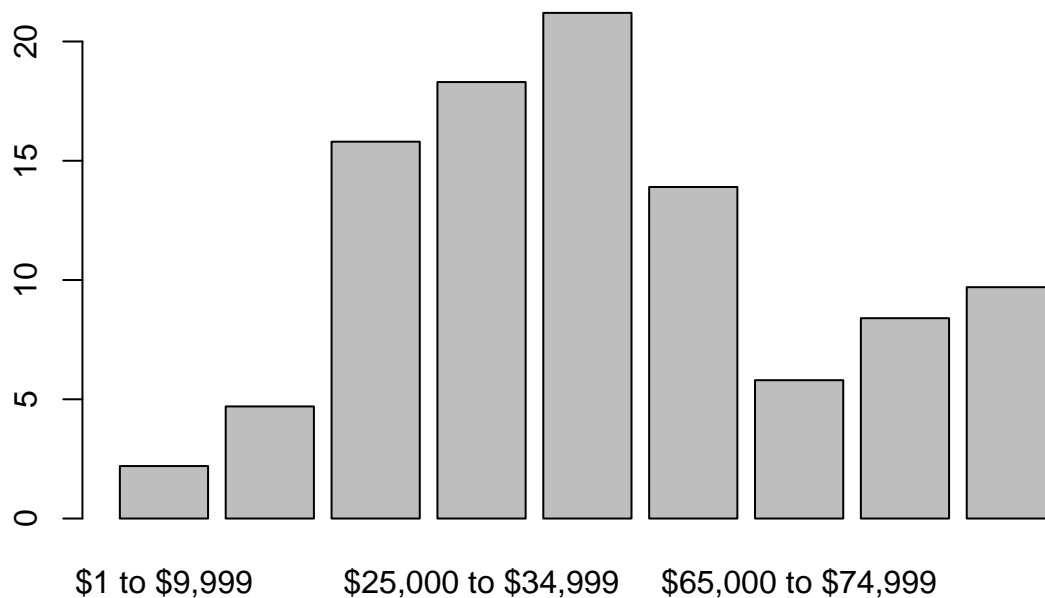
## [1] 1868.407
```

```
##          income total
```



```
## 1      $1 to $9,999  2.2
## 2 $10,000 to $14,999  4.7
## 3 $15,000 to $24,999 15.8
## 4 $25,000 to $34,999 18.3
## 5 $35,000 to $49,999 21.2
## 6 $50,000 to $64,999 13.9
## 7 $65,000 to $74,999  5.8
## 8 $75,000 to $99,999  8.4
## 9  $100,000 or more  9.7
```

```
barplot(dist$total, names.arg=income)
```



(b) What is the probability that a randomly chosen US resident makes less than \$50,000 per year?

$P(\text{Resident} < \$50,000) = P(\$1 \text{ to } \$9,999) + P(\$10,000 \text{ to } \$14,999) + P(\$15,000 \text{ to } \$24,999) + P(\$25,000 \text{ to } \$34,999) + P(\$35,000 \text{ to } \$49,999)$

$P(\text{Resident} < \$50,000) = 0.022 + 0.047 + 0.158 + 0.183 + 0.212$

$P(\text{Resident} < \$50,000) = 0.622$

```
sum(dist[1:5,2])
```

```
## [1] 62.2
```

(c) What is the probability that a randomly chosen US resident makes less than \$50,000 per year and is female? Note any assumptions you make.

This sample is comprised of 59% males and 41% females.

Assumption:

Since we don't know the relationship between the probability of an income of less than \$50,000 and being female.

I will be assuming that they are independent events then $P(A \text{ and } B) = P(A) \times P(B)$.

$$P(\text{Resident} < \$50000 \text{ and } F) = P(\text{Resident} < \$50000) * P(F)$$

$$P(\text{Resident} < \$50000 \text{ and } F) = 0.622 * 0.41$$

$$P(\text{Resident} < \$50000 \text{ and } F) = 0.2550$$

Answer: The probability that a randomly chosen US resident makes less than \$50,000 per year and is female is 25.50%.

(d) The same data source indicates that 71.8% of females make less than \$50,000 per year. Use this value to determine whether or not the assumption you made in part (c) is valid.

$$P(\text{Resident} < \$50000 \text{ and } F) = P(\text{Resident} < \$50000) * P(F)$$

$$0.718 = 0.622 * 0.41$$

$$\text{Since } 0.718 \neq 0.2550$$

We conclude that these events are not independent, since the independency multiplication rule is not satisfied.