

Homework 03

Christina Lee

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1.

- a. What's the chance of getting a sequential pair on two rolls of a die (eg, a 3 then a 4 counts, but a 4 then a 3 does not). (Hint: you can calculate this manually if you like, by counting up the sample space and finding the fraction of that sample space that consists of ordered pairs.)

To answer this, I listed out all sample space = 36 possible outcomes, and the total number of sequential pairs are $(1,2)(2,3)(3,4)(4,5)(5,6) = 5$ pairs. Therefore, the answer to this question is $P(seqpair) = 5/36$

(1,1)(1,2)(1,3)(1,4)(1,5)(1,6)
(2,1)(2,2)(2,3)(2,4)(2,5)(2,6)
(3,1)(3,2)(3,3)(3,4)(3,5)(3,6)
(4,1)(4,2)(4,3)(4,4)(4,5)(4,6)
(5,1)(5,2)(5,3)(5,4)(5,5)(5,6)
(6,1)(6,2)(6,3)(6,4)(6,5)(6,6)

- b. Given a dartboard with a inner circle that is $2/3$ of the total area, and a bulls-eye that is 5% of the total area (and entirely within the inner circle): if you are throwing a random dart (that is guaranteed to hit somewhere on the board, but everywhere inside is equally likely), what is the chance of hitting the bulls-eye conditional on knowing your dart is somewhere inside the inner circle?

To answer this, we know that the inner circle $P(A) = 2/3$ and $P(B) = 0.05$ and $P(A|B) = 1$. All we have to do is to put what is known into the equation which gives us the answer of 0.075% chance of hitting the bulls-eye.

$$P(B|A) = \frac{P(A|B) P(B)}{P(A)} = \frac{1 * 0.05}{\frac{2}{3}} = 0.075$$

- c. You take a test for a scary disease, and get a positive result. The disease is quite rare – 1 in 1000 in the general population. The test has a sensitivity of 95%, and a false positive rate of only 5%. What is the chance you have the disease?

To answer this, put what we know into Bayes' equation:

$$P(\text{disease}) = 0.001$$

$$P(+|\text{disease}) = 0.95$$

$$P(+|\text{No disease}) = 0.05$$

$$P(\text{No disease}) = 1 - 0.001 = 0.999$$

Therefore, given that you tested positive for this disease, there is only approximately 1.9% chance of you actually having the disease.

$$P(\text{disease}|+) = \frac{P(\text{disease}) P(+|\text{disease})}{P(\text{disease}) P(+|\text{disease}) + P(\text{No disease}) P(+|\text{No disease})}$$

$$P(\text{disease}|+) = \frac{(0.001)(0.95)}{(0.001)(0.95) + (0.999)(0.05)} = \frac{0.00095}{0.00095 + 0.04995} = \frac{0.00095}{0.0509} = 0.019$$

- d. What is the chance you have the disease if everything remains the same, but the disease is even rarer, 1 in 10,000?

To answer this, put what we know into Bayes' equation:

$$P(\text{disease}) = 0.0001$$

$$P(+|\text{disease}) = 0.95$$

$$P(+|\text{No disease}) = 0.05$$

$$P(\text{No disease}) = 1 - 0.0001 = 0.9999$$

Therefore, given that you tested positive for this disease, there is only approximately 0.19% chance of you actually having the disease.

$$P(\text{disease}|+) = \frac{P(\text{disease}) P(+|\text{disease})}{P(\text{disease}) P(+|\text{disease}) + P(\text{No disease}) P(+|\text{No disease})}$$

$$P(\text{disease}|+) = \frac{(0.0001)(0.95)}{(0.0001)(0.95) + (0.9999)(0.05)} = \frac{0.000095}{0.000095 + 0.049995} = \frac{0.000095}{0.05009} = 0.0019$$

- e. What does this tell you about the dangers of tests for rare diseases?

This tells us that the rarer the disease is –there is a less chance that you actually have the disease even though you tested positive for it. Such that, the accuracy rate for tests decreases as the disease gets rarer.

2.

- a. You have a 20-side die. Using sample, roll it 10,000 times and count the number of rolls that are 10 or less.

To answer this, I created a die with 1:20 sides and named it die1. I used sample() function to sample die1, rolled it 10,000 times, and added the replace argument. In order to count the number of rolls that are 10 or less, I placed the sample into a variable called “rollthedie” and used the sum() to count rolls with numbers that are <= 10 by adding the argument (rollthedie <= 10), which it will generate a series of TRUE = number is <= 10 and FALSE = number is > 10. The sum() will then add up number of TRUEs that appeared which essentially tells me the number of rolls that are <= 10.

```
die1 = c(1:20)

rollthedie <- sample(die1,10000,replace=TRUE)
sum(rollthedie<=10)
```

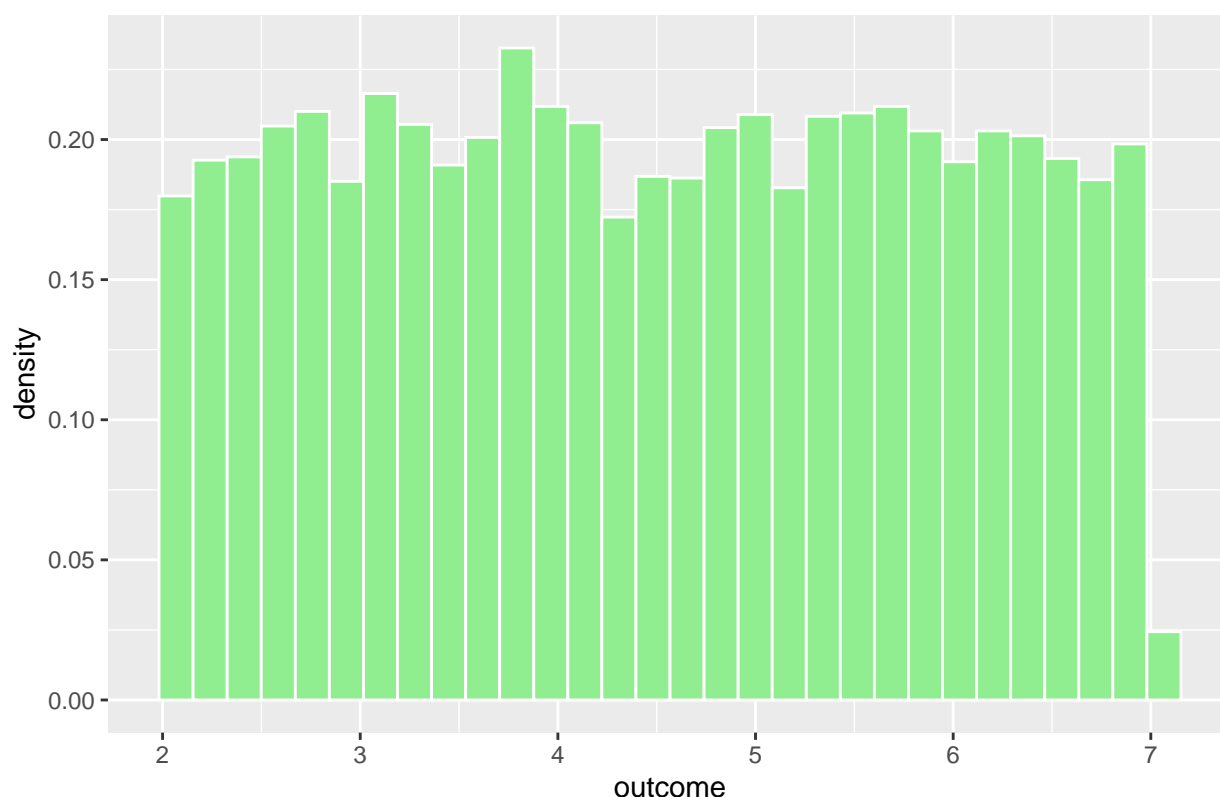
```
## [1] 5032
```

- b. Generate a histogram using ggplot of 10,000 draws from a uniform distribution between 2 and 7.

To answer this, I used the runif() function since we want a random uniform distribution and added in the first argument (generate 10000 draws), and the second and third argument (lower and upper bound, 2 and 7). I then named it “randunifs” and plotted the histogram via ggplot().

```
library(ggplot2)
randunifs <- runif(10000, 2,7)
ggplot(data=data.frame(randunifs),aes(x=randunifs)) +
  geom_histogram(aes(y=..density..),col="white",fill="lightgreen", bins=30) +
  ylab("density") + xlab("outcome") +
  ggtitle("Uniform Distribution of 10,000 Draws Between 2 and 7") +
  theme(plot.title = element_text(hjust = 0.5, face="bold"))
```

Uniform Distribution of 10,000 Draws Between 2 and 7



c. Try to write down the equation for this probability density function.

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{for } 2 \leq x \leq 7 \\ 0 & \text{for } x < 2 \text{ or } x > 7 \end{cases}$$

d. What is the probability that a draw from this distribution will be between 1.5 and 3.2?

To answer this, I used the `punif()` function since I want to know the probability from this uniform distribution. Next, the probability between 1.5 and 3.2 is basically $P(X \leq 3.2) - P(X \leq 1.5)$ so, all I have to do now is set up the arguments for each `punif()` functions (the min and max is the same as the one above since we are using the same uniform distribution as 2b). The answer is 0.24% of chance that a draw from this distribution will be between 1.5 and 3.2.

```
punif(3.2, min=2, max=7) - punif(1.5, min=2, max=7)
```

```
## [1] 0.24
```

3.

a. Using R's `cdf` for the binomial, what is the probability of getting 500 or fewer "20"s when rolling your 20-sided die 10,000 times. Looking back at 2a, how many of your rolls were actually 20s?

To answer this, since we are, again, asking for probability for the binomial, I used `pbinom()` function to calculate. I filled in the parameter for this function according to the question where 500 is q , 10000 is size, and $1/20$ is the prob, which gives the answer of 0.51% chance of getting 500 or fewer "20"s when rolling my 20-sided die 10000 times.

```
pbinom(500, 10000, 1/20)
```

```
## [1] 0.511895
```

Using the same method as 2a, I changed the argument from `<= 10` to `== 20` since I am asking for the rolls that were `= 20`.

```
die1 = c(1:20)

rollthedie <- sample(die1,10000,replace=TRUE)

sum(rollthedie == 20)

## [1] 506
```

- b. Using `rbinom`, roll a 100-sided die 100 times and report the total number of 7s you get.

To answer this, I used `rbinom()` to generate required number of random values of given probability from a given sample. To do that, I added in some arguments in the function: I only want 1 observation, 100 trial per observation, and the probability is 1/100 which is 0.01. I renamed the function and called it “hundie” for me to output the outcome and the output that I will get is the total number of 7s that I get in the 100 rolls.

```
hundie <- rbinom(1,100,0.01)
hundie

## [1] 0
```

This is another way to answer this question. Again, I used `rbinom()` to generate the required number of random values of given probability from a given sample, however, I added the following arguments in the function: this time, I want 100 observations, 1 trial per observation, and the probability is still 1/100 which is 0.01. I basically took this question and thought of it as a binary experiment. Although the probability stays the same, but instead of a dice, I thought of the question as flipping a two-sided coin, which I generated 100 flips of one coin, and the outcome can only be Heads or Tails. Heads (0) = all numbers that doesn't equal to 7 and Tails (1) = 7. The probability of getting Heads is 0.99 and 0.01 for Tails. Therefore, `ribnom()` will generate 100 numbers that are either 0 = fail or 1 = pass. I named the `rbinom()` function to “hundie2” and passed it into the `sum()` to count the number of 1s which essentially gives me the total number of 7s that was generated in the 100 rolls.

```
hundie2 <- rbinom(100,1,0.01)
sum(hundie2 == 1)

## [1] 0
```

- c. You are a klutz, and the average number of times you drop your pencil in a day is 1. Using the poisson functions in R, what's the chance of dropping your pencil two or more times in a day? (Hint: calculate the chance of dropping it one or fewer times, and then take 1 minus that.)

To answer this, I used `ppois()` function to calculate the probability of dropping pencil two or more times by first calculating the chance of dropping it one or fewer times (which is my first argument `=1`) and with a mean of 1 (my second argument). Then I took 1 minus the `ppois()` function and named it variable “pencildrop” to output the answer.

```
pencildrop <- 1-ppois(1,1)
pencildrop

## [1] 0.2642411
```

- d. Because he is lazy, your teacher has assigned grades for an exam at random, and to help hide his deception he has given the fake grades a normal distribution with a mean of 70 and a standard deviation of 10. What is the chance your exam got a score of 85 or above? What is the chance you got a score between 50 and 60?

To answer this, I used `pnorm()` function to find the probability of getting a score of 85 or above, where 85 is `x`, 70 is mean, 10 is sd, and `lower.tail = FALSE` because I am interesting in knowing the probability of value `x=85` or larger which are located on the upper tail of the distribution.

```
pnorm(85,70,10, lower.tail=FALSE)
```

```
## [1] 0.0668072
```

To answer this, I used `pnorm()` function to find the probability of getting a score between 50 and 60 by taking the probability of 60 and subtract it by probability of 50.

```
pnorm(60,70,10, lower.tail=TRUE) - pnorm(50,70,10, lower.tail=TRUE)
```

```
## [1] 0.1359051
```