Lab 3

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Hot Hands

Basketball players who make several baskets in succession are described as having a *hot hand*. Fans and players have long believed in the hot hand phenomenon, which refutes the assumption that each shot is independent of the next. However, a 1985 paper by Gilovich, Vallone, and Tversky collected evidence that contradicted this belief and showed that successive shots are independent events. This paper started a great controversy that continues to this day, as you can see by Googling *hot hand basketball*.

We do not expect to resolve this controversy today. However, in this lab we'll apply one approach to answering questions like this. The goals for this lab are to (1) think about the effects of independent and dependent events, (2) learn how to simulate shooting streaks in R, and (3) to compare a simulation to actual data in order to determine if the hot hand phenomenon appears to be real.

Getting Started

Load packages

In this lab, we will explore and visualize the data using the tidyverse suite of packages. The data can be found in the companion package for OpenIntro labs, openintro.

Let's load the packages.

```
library(tidyverse)
library(openintro)
```

Data

Your investigation will focus on the performance of one player: Kobe Bryant of the Los Angeles Lakers. His performance against the Orlando Magic in the 2009 NBA Finals earned him the title Most Valuable Player and many spectators commented on how he appeared to show a hot hand. The data file we'll use is called kobe_basket.

```
glimpse(kobe_basket)
```

This data frame contains 133 observations and 6 variables, where every row records a shot taken by Kobe Bryant. The shot variable in this dataset indicates whether the shot was a hit (H) or a miss (M).

Just looking at the string of hits and misses, it can be difficult to gauge whether or not it seems like Kobe was shooting with a hot hand. One way we can approach this is by considering the belief that hot hand shooters tend to go on shooting streaks. For this lab, we define the length of a shooting streak to be the number of consecutive baskets made until a miss occurs.

For example, in Game 1 Kobe had the following sequence of hits and misses from his nine shot attempts in the first quarter:

$$HM \mid M \mid HHM \mid M \mid M \mid M$$

You can verify this by viewing the first 9 rows of the data in the data viewer.

Within the nine shot attempts, there are six streaks, which are separated by a "|" above. Their lengths are one, zero, zero, zero, zero (in order of occurrence).

Exercise 1 What does a streak length of 1 mean, i.e. how many hits and misses are in a streak of 1? What about a streak length of 0?

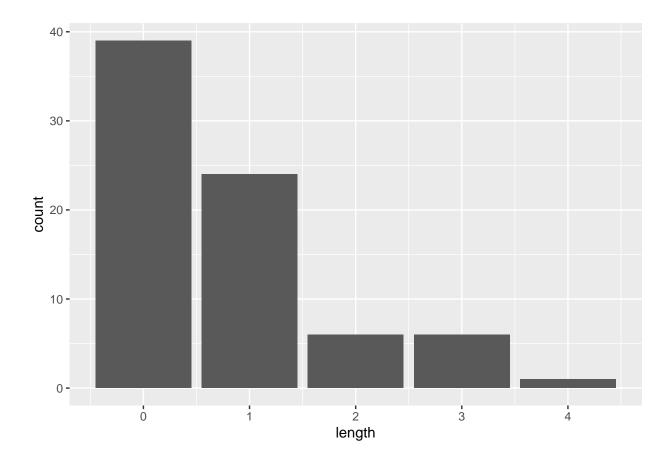
Answer 1: When there is a shot and a miss the streak length is 1. When there are zero hits and one miss the streak is 0.

Counting streak lengths manually for all 133 shots would get tedious, so we'll use the custom function calc_streak to calculate them, and store the results in a data frame called kobe_streak as the length variable.

```
kobe_streak <- calc_streak(kobe_basket$shot)</pre>
```

We can then take a look at the distribution of these streak lengths.

```
ggplot(data = kobe_streak, aes(x = length)) +
  geom_bar()
```



Exercise 2: Describe the distribution of Kobe's streak lengths from the 2009 NBA finals. What was his typical streak length? How long was his longest streak of baskets? Make sure to include the accompanying plot in your answer.

Answer 2: Kobe's streak is a right skewed distribution. As he misses the ball number of times in a row his typical streak is 0. His longest streak of baskets is 4.

Compared to What?

We've shown that Kobe had some long shooting streaks, but are they long enough to support the belief that he had a hot hand? What can we compare them to?

To answer these questions, let's return to the idea of *independence*. Two processes are independent if the outcome of one process doesn't effect the outcome of the second. If each shot that a player takes is an independent process, having made or missed your first shot will not affect the probability that you will make or miss your second shot.

A shooter with a hot hand will have shots that are *not* independent of one another. Specifically, if the shooter makes his first shot, the hot hand model says he will have a *higher* probability of making his second shot.

Let's suppose for a moment that the hot hand model is valid for Kobe. During his career, the percentage of time Kobe makes a basket (i.e. his shooting percentage) is about 45%, or in probability notation,

$$P(\text{shot } 1 = \text{H}) = 0.45$$

If he makes the first shot and has a hot hand (not independent shots), then the probability that he makes his second shot would go up to, let's say, 60%,

$$P(\text{shot } 2 = H | \text{shot } 1 = H) = 0.60$$

As a result of these increased probabilities, you'd expect Kobe to have longer streaks. Compare this to the skeptical perspective where Kobe does *not* have a hot hand, where each shot is independent of the next. If he hit his first shot, the probability that he makes the second is still 0.45.

$$P(\text{shot } 2 = H | \text{shot } 1 = H) = 0.45$$

In other words, making the first shot did nothing to effect the probability that he'd make his second shot. If Kobe's shots are independent, then he'd have the same probability of hitting every shot regardless of his past shots: 45%.

Now that we've phrased the situation in terms of independent shots, let's return to the question: how do we tell if Kobe's shooting streaks are long enough to indicate that he has a hot hand? We can compare his streak lengths to someone without a hot hand: an independent shooter.

Simulations in R

While we don't have any data from a shooter we know to have independent shots, that sort of data is very easy to simulate in R. In a simulation, you set the ground rules of a random process and then the computer uses random numbers to generate an outcome that adheres to those rules. As a simple example, you can simulate flipping a fair coin with the following.

```
set.seed(35797)
coin_outcomes <- c("heads", "tails")
sample(coin_outcomes, size = 1, replace = TRUE)</pre>
```

```
## [1] "tails"
```

The vector coin_outcomes can be thought of as a hat with two slips of paper in it: one slip says heads and the other says tails. The function sample draws one slip from the hat and tells us if it was a head or a tail.

Run the second command listed above several times. Just like when flipping a coin, sometimes you'll get a heads, sometimes you'll get a tails, but in the long run, you'd expect to get roughly equal numbers of each.

If you wanted to simulate flipping a fair coin 100 times, you could either run the function 100 times or, more simply, adjust the size argument, which governs how many samples to draw (the replace = TRUE argument indicates we put the slip of paper back in the hat before drawing again). Save the resulting vector of heads and tails in a new object called sim_fair_coin.

```
set.seed(35797)
sim_fair_coin <- sample(coin_outcomes, size = 100, replace = TRUE)</pre>
```

To view the results of this simulation, type the name of the object and then use table to count up the number of heads and tails.

```
set.seed(35797)
sim_fair_coin
```

```
## [1] "tails" "heads" "heads" "tails" "heads" "tails" "heads" "tails" "tails" "tails" "tails" "tails" "tails" "tails" "tails" "tails" "heads" "heads"
```

```
## [73] "tails" "heads" "heads" "tails" "tails" "heads" "heads" "tails"
## [82] "heads" "heads" "tails" "heads" "tails" "heads" "tails" "heads" "tails" "heads" "tails" "heads" "tails" "tails" "tails" "heads" "tails" "tails" "tails" "heads" "tails" "tails" "heads" "tails" "tails" "heads" "tails" "tails" "tails" "heads" "tails" "
```

Since there are only two elements in coin_outcomes, the probability that we "flip" a coin and it lands heads is 0.5. Say we're trying to simulate an unfair coin that we know only lands heads 20% of the time. We can adjust for this by adding an argument called prob, which provides a vector of two probability weights.

prob=c(0.2, 0.8) indicates that for the two elements in the outcomes vector, we want to select the first one, heads, with probability 0.2 and the second one, tails with probability 0.8. Another way of thinking about this is to think of the outcome space as a bag of 10 chips, where 2 chips are labeled "head" and 8 chips "tail". Therefore at each draw, the probability of drawing a chip that says "head" is 20%, and "tail" is 80%.

Exercise 3: In your simulation of flipping the unfair coin 100 times, how many flips came up heads? Include the code for sampling the unfair coin in your response. Since the markdown file will run the code, and generate a new sample each time you Knit it, you should also "set a seed" before you sample. Read more about setting a seed below.

Answer 3: In the flip of unfair coins heads came up 56 times.

```
table(sim_fair_coin)

## sim_fair_coin
## heads tails
## 56 44
```

Simulating the Independent Shooter

Simulating a basketball player who has independent shots uses the same mechanism that you used to simulate a coin flip. To simulate a single shot from an independent shooter with a shooting percentage of 50% you can type

```
shot_outcomes <- c("H", "M")
sim_basket <- sample(shot_outcomes, size = 1, replace = TRUE)</pre>
```

To make a valid comparison between Kobe and your simulated independent shooter, you need to align both their shooting percentage and the number of attempted shots.

Exercise 4: What change needs to be made to the sample function so that it reflects a shooting percentage of 45%? Make this adjustment, then run a simulation to sample 133 shots. Assign the output of this simulation to a new object called sim_basket.

```
set.seed(123)
outcomes <- c("H", "M")
sim_basket <- sample(outcomes, size = 133, replace = TRUE, prob = c(0.45, 0.55))
sim_basket</pre>
```

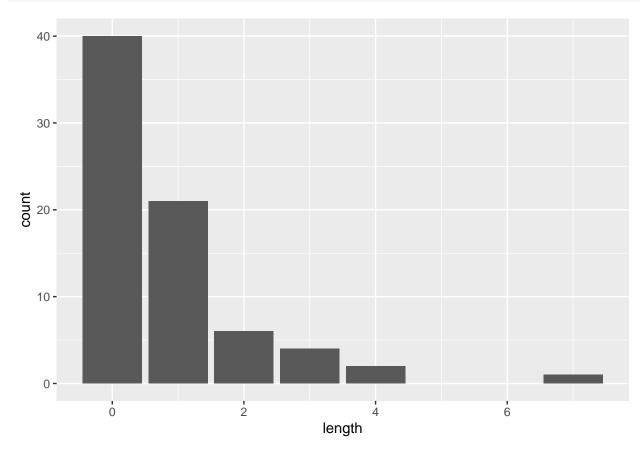
```
## sim_basket
## H M
## 60 73
```

More Practise

Comparing Kobe Bryant to the Independent Shooter

Exercise 5: Using calc_streak, compute the streak lengths of sim_basket, and save the results in a data frame called sim_streak.

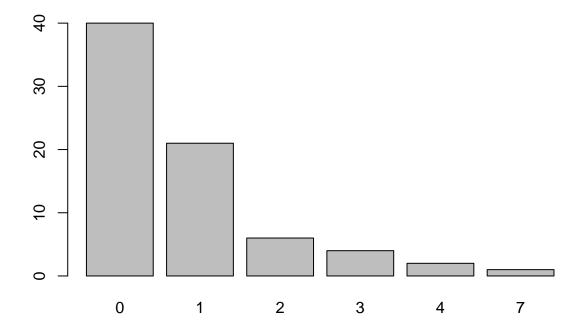
```
sim_streak <- calc_streak(sim_basket)
ggplot(data = sim_streak, aes(x = length)) +
  geom_bar()</pre>
```



Exercise 6: Describe the distribution of streak lengths. What is the typical streak length for this simulated independent shooter with a 45% shooting percentage? How long is the player's longest streak of baskets in 133 shots? Make sure to include a plot in your answer.

Answer 6: The distribution is right skewed. The typical streak length is 0. The longest streak length is 7.

```
barplot(table(sim_streak))
```

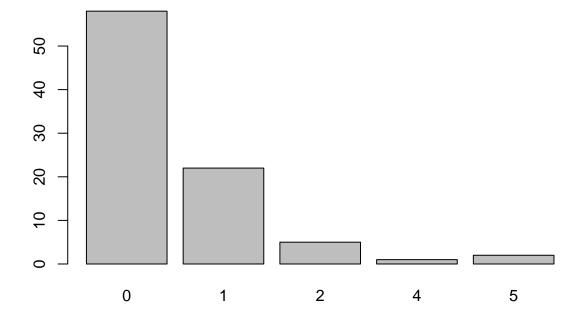


Exercise 7: If you were to run the simulation of the independent shooter a second time, how would you expect its streak distribution to compare to the distribution from the question above? Exactly the same? Somewhat similar? Totally different? Explain your reasoning.

Answer 7: The streak distribution would vary each time but will be similar to each other.

```
set.seed(1234)
sim_basket1 <- sample(outcomes, size = 133, replace = TRUE, prob = c(0.45, 0.55))
sim basket1
##
     [1] "M"
            "H" "H" "H" "H" "H" "M" "M" "H"
                                             "M"
                                                  "H"
                                                      иМи
                                                               "H"
                                                                       "H"
##
                 "M"
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                         "M" "M" "M" "M" "H" "M"
                                                      "H" "M" "M"
                     "M"
                                                                   "M" "M"
                          "H" "M" "M" "H" "H" "M" "M"
                                                      "H" "H" "H"
   [127] "M" "M" "M"
                     "M" "H" "H" "M"
table(sim_basket1)
```

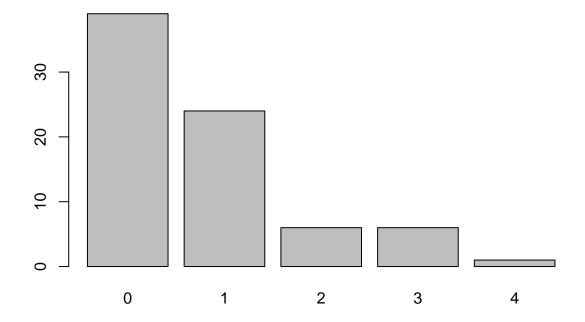
```
## sim_basket1
## H M
## 46 87
ex <- calc_streak(sim_basket1)
barplot(table(ex))</pre>
```



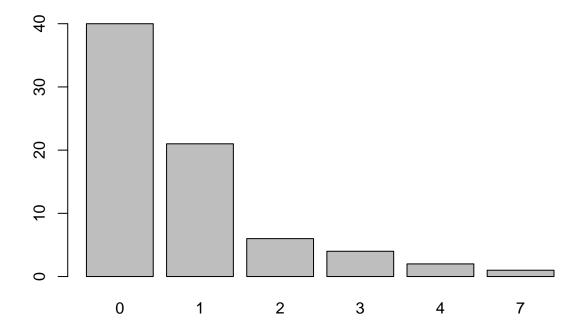
Exercise 8: How does Kobe Bryant's distribution of streak lengths compare to the distribution of streak lengths for the simulated shooter? Using this comparison, do you have evidence that the hot hand model fits Kobe's shooting patterns? Explain.

Answer 8: When we compare the 3 barplots below it is clear that hot hands cannot be concluded from this data

```
barplot(table(kobe_streak)) #kobe's graph
```



barplot(table(sim_streak)) #First simulation of independent shooter



barplot(table(ex))

#Second simulation of independent shooter

