### Jordan V. Lebron

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```
# Import libraries and data here
library(dplyr)
library(readr)
library(tidyverse)
library(rvest)
library(ggplot2)
library(tidyr)
library(esquisse)
# Downloaded Lebron Data from BasketBall Reference
# Link: https://www.basketball-reference.com/players/j/jamesle01.html#per_game
LebronPerGame <- read_csv('./LebronNew/LebronPerGameNew.csv')</pre>
LebronTotals <- read_csv('./LebronNew/LebronTotalStatsNew.csv')
LebronAdvanced <- read_csv('./LebronNew/LebronAdvancedNew.csv')
LebronPer100Poss <- read_csv('./LebronNew/LebronPer100Poss.csv')</pre>
LebronAllStarGames <- read_csv('./LebronNew/LebronAllStarGame.csv')
# Downloaded Jordan Data from BasketBall Reference
# Link: https://www.basketball-reference.com/players/j/jordami01.html
JordanPerGame <- read_csv('./JordanNew/JordanPerGameNew.csv')</pre>
JordanTotals <- read_csv('./JordanNew/JordanTotalStatsNew.csv')</pre>
JordanAdvanced <- read_csv('./JordanNew/JordanAdvancedNew.csv')</pre>
JordanPer100Poss <- read_csv('./JordanNew/JordanPer100Poss.csv')</pre>
JordanAllStarGames <- read_csv('./JordanNew/JordanAllStarGame.csv')</pre>
```

## Guiding Question:

• Who is the Greatest Basketball Player of All Time in NBA History, Lebron or Jordan?

## Become acquainted with your data sources:

### Where did you find them?

• Data retrieved from BasketBall Reference.

### Who collected/maintains them?

- Sean Lahman donated much of the initial data.
- Chip Hart and Kevin Cohen provided input on the site's design and data collection, respectively.
- Tom Timmerman, Dick Pfander, Todd Spehr, Matt Shuh, Justin Kubatko, Sean Burrill, Tariq Jabbar, Mike Lynch, Michael Hamel, Mark Montieth, and Robert Bradley have all contributed significantly to the collection of NBA and ABA data.
- Matthew Maurer, an NBA Draft historian, has supplied information on the NBA Draft.

### When & Why were they originally collected?

Basketball Reference was launched in April of 2004 by Justin Kubatko. The primary goal was to create a comprehensive database that tracks basketball information. This encompasses player statistics, team performances, game results, and much more, making it an invaluable resource for fans, analysts, and researchers interested in the sport's history and current events.

### What does a case represent in each data source, and how many total cases are available?

- For player statistics, each row or case represents a season's worth of performance data for that player.
- Game logs would treat each game as a case.
- Draft data entries represent individual players selected in the NBA draft.
- There are thousands of cases covering decades of basketball history across the NBA and ABA.

### What are some of the variables that you plan to use?

- Points per game (PointsPerGame)
- Assists per game (AssistsPerGame)
- Rebounds per game (ReboundsPerGame)
- Total Points (TotalPoints)
- Total Assists (TotalAssists)
- Total Rebounds (TotalRebounds)
- Player efficiency rating (PlayerEfficiencyRating)
- Career achievements (Career Achievements)
- All-star game appearances (AllStarGameAppearances)
- Game scores (GameScores)

## Explore intuition related to the research question

## 1. Comparing Lebron and Jordans total and per-game stats

### Bargraph showcasing a comparison of per game-stats

```
LebronStatsPerGame <- LebronTotals %>%
summarise(Player = "Lebron James", AssistsPerGame = sum(AST, na.rm = TRUE) / sum(G, na.rm = TRUE), ReboundsP

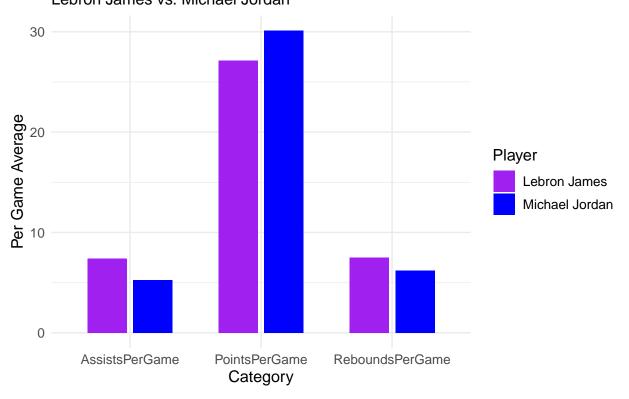
JordanStatsPerGame <- JordanTotals %>%
summarise(Player = "Michael Jordan", AssistsPerGame = sum(AST, na.rm = TRUE) / sum(G, na.rm = TRUE), Rebound

StatsComparison <- bind_rows(LebronStatsPerGame, JordanStatsPerGame)

StatsComparisonLong <- StatsComparison %>%
pivot_longer(cols = c(AssistsPerGame, ReboundsPerGame, PointsPerGame), names_to = "Statistic", values_to = "

ggplot(StatsComparisonLong, aes(x = Statistic, y = Value, fill = Player)) +
geom_bar(stat = "identity", position = position_dodge(width = 0.7), width = 0.6) +
theme_minimal() +
labs(title = "Comparison of Per Game Statistics", subtitle = "Lebron James vs. Michael Jordan", x = "Categor scale_fill_manual(values = c("Lebron James" = "purple", "Michael Jordan" = "blue")) +
theme(text = element_text(size = 12))
```

## Comparison of Per Game Statistics Lebron James vs. Michael Jordan



#### print(StatsComparison)

```
## # A tibble: 2 x 4
##
     Player
                     AssistsPerGame ReboundsPerGame PointsPerGame
##
     <chr>>
                               <dbl>
                                                <dbl>
                                                               <dbl>
## 1 Lebron James
                                7.38
                                                 7.50
                                                                27.1
                                5.25
                                                 6.22
## 2 Michael Jordan
                                                                30.1
```

LeBron has averaged 27.1 points, 7.4 assists, and 7.5 rebounds per game over the course of his 21 year career. Jordan has averaged 30.1 points, 5.3 assists, and 6.2 rebounds over the course of his 15 year career. Even though Jordan has averaged a higher amount of points, LeBron was never seen as a scorer. LeBron was known for his playmaking and getting his team involved (We can see that with the 7.4 assists), while Jordan was predominantly known as a prolific scorer.

### Bargraph showcasing a comparison of total stats

```
LebronTotalsStats <- LebronTotals %>%
summarise(Player = "Lebron James", TotalRebounds = sum(TRB, na.rm = TRUE), TotalAssists = sum(AST, na.rm = T

JordanTotalsStats <- JordanTotals %>%
summarise(Player = "Michael Jordan", TotalRebounds = sum(TRB, na.rm = TRUE), TotalAssists = sum(AST, na.rm =

StatsComparison <- bind_rows(LebronTotalsStats, JordanTotalsStats)

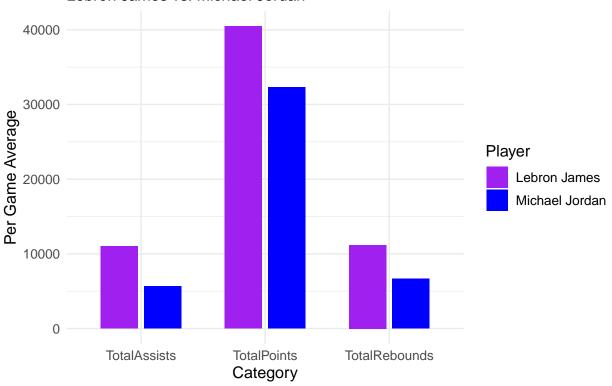
StatsComparisonLong <- StatsComparison %>%
pivot_longer(cols = c(TotalRebounds, TotalAssists, TotalPoints), names_to = "Statistic", values_to = "Value"

ggplot(StatsComparisonLong, aes(x = Statistic, y = Value, fill = Player)) +
geom_bar(stat = "identity", position = position_dodge(width = 0.7), width = 0.6) +
theme_minimal() +
```

```
labs(title = "Comparison of Total Game Statistics", subtitle = "Lebron James vs. Michael Jordan", x = "Categ
scale_fill_manual(values = c("Lebron James" = "purple", "Michael Jordan" = "blue")) +
theme(text = element_text(size = 12))
```

# Comparison of Total Game Statistics





#### print(StatsComparison)

```
## # A tibble: 2 x 4
##
     Player
                     TotalRebounds TotalAssists TotalPoints
                                           <dbl>
##
     <chr>
                             <dbl>
                                                        <dbl>
## 1 Lebron James
                                           11009
                                                        40474
                             11185
## 2 Michael Jordan
                              6672
                                            5633
                                                        32292
```

LeBron over the course of his 21 years has accumulated 40,474 points, 11,009 assists, and 11,185 total rebounds. Jordan in his 15 years in the league has totaled 32,292 points, 5,633 assists, and 6,672 total rebounds. LeBron has just overtaken Kareem as the all time leading scorer and sits in 4th place for total all time assists while Jordan barely breaks the top five for points and has nowhere near the rebounds and assists to be a leader in the stat. Even though LeBron has played 6 more seasons than Jordan, he is currently putting up 27 points per game for his entire career including this current year at 39 years old. Jordan had to retire at 39 because he was such a liability as a player.

## 2. Using Advanced and Per100Poss stats for comparison

Dotplot showcasing VORP (Value over replacement player)

```
LebronLongevity <- LebronAdvanced %>%

mutate(PlayerName = "Lebron James", Games = G, TrueShootingPercentage = as.numeric("TS%"), WinShare = WS, Vaselect(PlayerName, Season, Games, TrueShootingPercentage, WinShare, ValueOverReplacementPlayer)
```

```
JordanLongevity <- JordanAdvanced %>%
  mutate(PlayerName = "Michael Jordan", Games = G, TrueShootingPercentage = as.numeric("TS%"), WinShare
  select(PlayerName, Season, Games, TrueShootingPercentage, WinShare, ValueOverReplacementPlayer)

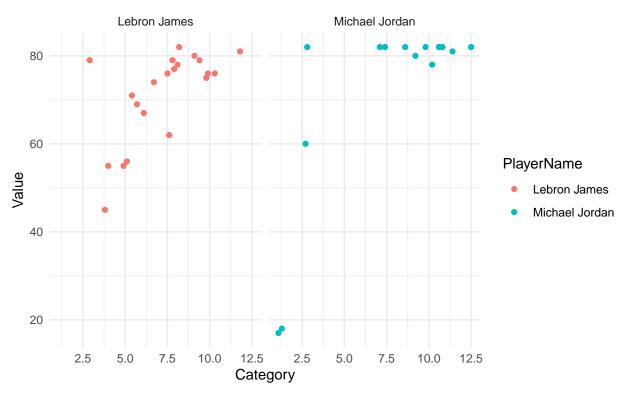
CombinedData <- bind_rows(LebronLongevity, JordanLongevity)

CombinedDataLonger <- CombinedData %>%
  pivot_longer(cols = c(Games), names_to = "Statistic", values_to = "Value")

CombinedDataLonger %>%
  ggplot(aes(x = ValueOverReplacementPlayer, y = Value, color = PlayerName)) +
  geom_point(shape = "circle", size = 1.5) +
  scale_color_hue(direction = 1) +
  theme_minimal() +
  facet_wrap(vars(PlayerName)) +
  labs(title = "Effeciency using (VORP)", subtitle = "Lebron James vs. Michael Jordan", x = "Category", y = "V
```

### Effeciency using (VORP)

#### Lebron James vs. Michael Jordan



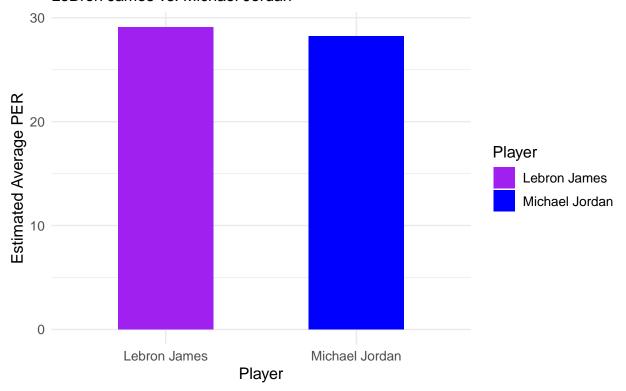
This bar plot shows an advanced stat, Value over replacement player. This is showing us per 100 possessions each players had accounted for that benefited their team. We can see that Jordans ceiling is higher and has had more instances in his career where his VORP is higher than LeBrons. We can also see that LeBron is way more consistent with his lowest points on the bar plot accounting for his rookie seasons in the NBA when he was 18 coming out of high school. This plot leans both ways with what bias you hold towards either player, but overall we think this shows how valuable LeBron really is to his team.

### Bargraph showcasing Player Effeciency Ratings (PER)

```
PER <- function(x){
  x %>%
  mutate(uPER = (PTS + TRB + AST + STL + BLK - ((FGA - FG) + (FTA - FT) + TOV)) / G) %>%
```

```
summarise(AveragePER = mean(uPER, na.rm = TRUE))
}
LebronPER <- PER(LebronTotals) %>%
  mutate(Player = "Lebron James")
JordanPER <- PER(JordanTotals) %>%
  mutate(Player = "Michael Jordan")
PERComparison <- bind_rows(LebronPER, JordanPER)</pre>
print(PERComparison)
## # A tibble: 2 x 2
##
     AveragePER Player
##
          <dbl> <chr>
           29.1 Lebron James
## 1
## 2
           28.2 Michael Jordan
ggplot(PERComparison, aes(x = Player, y = AveragePER, fill = Player)) +
  geom_bar(stat = "identity", position = position_dodge(), width = 0.5) +
  theme_minimal() +
  labs(title = "Comparison of Estimated Player Efficiency Rating (PER)", subtitle = "LeBron James vs. Michael
  scale_fill_manual(values = c("Lebron James" = "purple", "Michael Jordan" = "blue")) +
  theme(text = element_text(size = 12))
```

## Comparison of Estimated Player Efficiency Rating (PER) LeBron James vs. Michael Jordan



This graph compares LeBron and Jordans efficiency ratings over the course of their careers when comparing their turnover percentages to their scoring ability. The graph shows that Lebrons per 100 possesions stats makes LeBron a more efficient player.

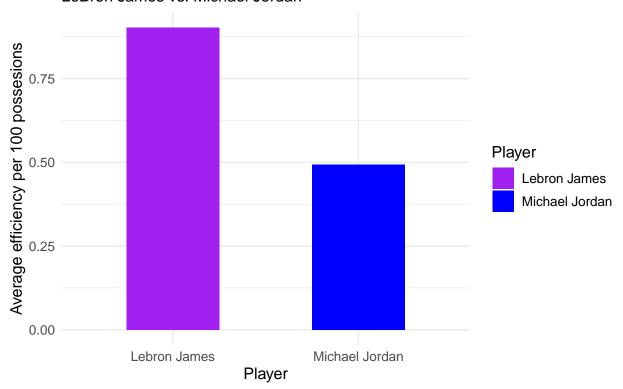
## 3. Showcasing and comparing LeBron and Jordans Legacys

OldBron <- LebronPer100Poss %>%

Barpgraph that shows LeBron and Jordans Effeciency ratings above age 30

```
filter(Season > "2014-15") %>%
  mutate(PER = (PTS + TRB + AST + STL + BLK - ((FGA - FG) + (FTA - FT) + TOV)) / G) %>%
  mutate(Player = "Lebron James")
OldJordan <- JordanPer100Poss%>%
  filter(Season > "1994-95") %>%
  mutate(PER = (PTS + TRB + AST + STL + BLK - ((FGA - FG) + (FTA - FT) + TOV)) / G) %>%
  mutate(Player = "Michael Jordan")
OldComp <- bind_rows(OldBron, OldJordan)</pre>
print(OldComp)
## # A tibble: 14 x 34
                                                                FGA 'FG%'
                                          G
                                               GS
                                                     MP
                                                           FG
##
      Season
                Age Tm
                          Lg
                                Pos
##
      <chr>
              <dbl> <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
                 31 CLE
##
   1 2015-16
                          NBA
                                SF
                                         76
                                               76
                                                   2709
                                                         14
                                                                26.9 0.52
                                                                            1.7
   2 2016-17
                 32 CLE
                          NBA
                                SF
                                         74
                                               74
                                                   2794
                                                         13.1
                                                               24
                                                                    0.548
                                                                            2.2
##
##
   3 2017-18
                 33 CLE
                          NBA
                                PF
                                         82
                                               82
                                                   3026
                                                         13.9 25.6 0.542
                                                                            2.4
                                         55
                                                   1937
                                                         13.4 26.3 0.51
##
   4 2018-19
                 34 LAL
                          \mathtt{NBA}
                                SF
                                               55
                                                                            2.7
##
   5 2019-20
                 35 LAL
                          NBA
                                PG
                                         67
                                               67
                                                   2316
                                                         13.2 26.8 0.493
## 6 2020-21
                 36 LAL
                          NBA
                                PG
                                         45
                                               45 1504 13.7 26.6 0.513
                                                                            3.4
   7 2021-22
                 37 LAL
                          NBA
                                C
                                         56
                                               56 2084 14.7 28.1 0.524
                                                                            3.7
##
##
   8 2022-23
                 38 LAL
                          NBA
                                PF
                                         55
                                               54 1954 14.8 29.6 0.5
                                                                            2.9
## 9 2023-24
                 39 LAL
                          NBA
                                PF
                                         71
                                               71
                                                  2504 13
                                                                24.1 0.54
                                                                            2.8
## 10 1995-96
                          NBA
                                SG
                                         82
                                               82 3090 15.6 31.5 0.495
                                                                            1.9
                 32 CHI
## 11 1996-97
                                SG
                                         82
                                               82 3106 15.8 32.5 0.486
                 33 CHI
                          NBA
                                                                            1.9
## 12 1997-98
                 34 CHI
                          NBA
                                SG
                                         82
                                               82
                                                   3181 14.9 32.1 0.465
                                                                            0.5
## 13 2001-02
                 38 WAS
                                SF
                                         60
                                               53
                                                   2093 14.3 34.4 0.416
                          NBA
                                                                            0.3
## 14 2002-03
                 39 WAS
                          NBA
                                SF
                                         82
                                               67 3031 12.2 27.4 0.445
                                                                            0.3
## # i 22 more variables: '3PA' <dbl>, '3P%' <dbl>, '2P' <dbl>, '2PA' <dbl>,
       '2P%' <dbl>, FT <dbl>, FTA <dbl>, 'FT%' <dbl>, ORB <dbl>, DRB <dbl>,
## #
       TRB <dbl>, AST <dbl>, STL <dbl>, BLK <dbl>, TOV <dbl>, PF <dbl>, PTS <dbl>,
## #
      ...30 <lgl>, ORtg <dbl>, DRtg <dbl>, PER <dbl>, Player <chr>
## #
ggplot(OldComp, aes(x = Player, y = PER, fill = Player)) +
  geom_bar(stat = "identity", position = position_dodge(), width = 0.5) +
  theme minimal() +
  labs(title = "Efficiency per 100 possesions at age 30 plus", subtitle = "LeBron James vs. Michael Jordan", y
  scale_fill_manual(values = c("Lebron James" = "purple", "Michael Jordan" = "blue")) +
  theme(text = element_text(size = 12))
```

## Efficiency per 100 possesions at age 30 plus LeBron James vs. Michael Jordan



This bar graph compares Michael Jordan and LeBrons efficiency's per 100 possessions at age 30 plus. We can clearly see the longevity LeBron holds. When comparing points, assists, rebounds, field goal percentages, and turnover rate, we see that LeBron dominates. At age 39 LeBron is still considered a top player in the NBA while Jordan was a dwindling old man expiring in his puny body.

#### Comparing total all star appearences

```
LebronAllStarAppearances <- nrow(LebronAllStarGames)

JordanAllStarAppearances <- nrow(JordanAllStarGames)

print(paste("Lebron James All-Star Game Appearances:", LebronAllStarAppearances))

## [1] "Lebron James All-Star Game Appearances: 21"

print(paste("Michael Jordan All-Star Game Appearances:", JordanAllStarAppearances))

## [1] "Michael Jordan All-Star Game Appearances: 15"

AllStarData <- data.frame(Player = c("Lebron James", "Michael Jordan"), Appearances = c(LebronAllStarAppearance

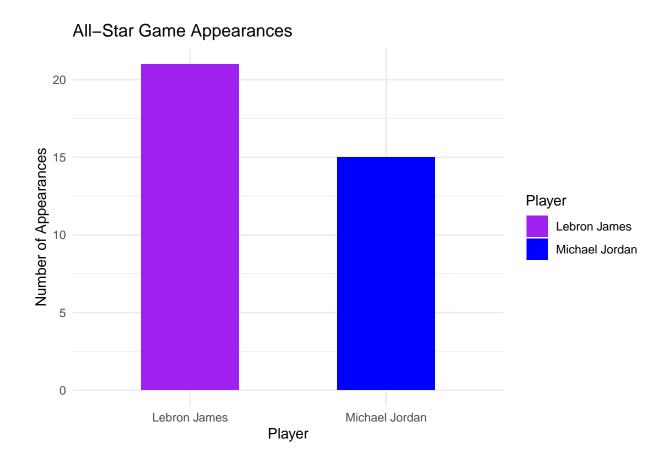
ggplot(AllStarData, aes(x = Player, y = Appearances, fill = Player)) +

geom_bar(stat = "identity", position = position_dodge(), width = 0.5) +

theme_minimal() +

labs(title = "All-Star Game Appearances", y = "Number of Appearances", fill = "Player") +

scale_fill_manual(values = c("Lebron James" = "purple", "Michael Jordan" = "blue"))
```



<sup>&</sup>quot;' When it comes to All-Star game appearences Lebron has been slected to one every season he has played in the NBA and sits at the most all-star selections ever. Jordans 15, although impressive, are expected when you are considered one of the greatest of all time. Lebrons Longevity to be able to make it to an all star game 21 years, even at his current age of 39 is unheard of.