

# Stat 340 Group Progress Report

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## Description of Data

Baseball is an American sport played between two teams. One team plays defense and has a player (pitcher) try to throw the ball past the other team's batter who tries to hit the ball. If the batting team successfully hits the ball enough times they can score runs (points). We will use data sets from baseball-reference.com which contain team and individual statistics. This database also offers data 'splits', which show comparisons between home and away games. For our primary analysis, we will be focusing on the Houston Astros 2017 Team Batting Splits and the 2017 Individual Player Batting Splits. These data sets outline various different statistics of the Houston Astros 2017 season in which it was confirmed that they had cheated in home games. We will use the 'splits' data sets to determine a difference of statistics between home and away games. This database also contains data on the Astros for different seasons and different teams. Depending on the results of our initial analysis, we will compare the players on the roster of the 2017 Astros to previous seasons to determine a difference in performance before and after cheating. Our statistical question is whether or not the 2017 Houston Astros cheated. Two years after the alleged cheating occurred, reports from a player who was on the 2017 Astros alleged that there were cameras positioned in the Astros home field stadium, Minute Maid Park in Houston, that could see the relayed signs between the pitcher and catcher. Note: the catcher communicates to the pitcher what pitch the pitcher should throw based on a series of hand movements in between pitches. The Houston Astros camera could see these hand signals and people were in charge of watching the film and decoding the symbols to determine what pitch was coming. Once a pitch was shown, members of the Astros would relay what pitch was coming to the batter through their own series of symbols (most infamously a banging of a garbage can).

The Astros only had this technology at their Home Stadium, so they should have higher performance at home compared to on the road. Note: this sign decoding should reason that only their batting improved at home and not their pitching. We used several stats for measuring their performance.

## *Statistical Questions*

- Is there a noticeable difference between the batting splits for home and away games of the Houston Astros team during the 2017 season?
- How does this difference, if any, compare to other teams' performance in the MLB during the 2017 season?

- Is there sufficient statistical evidence to suggest the Houston Astros benefited from cheating in the 2017 season?
  - If there is, how much may the cheating have affected their season? And what would their season have looked like if they didn't cheat?

## Why We Chose This Dataset

- This has been a controversy within the baseball community regarded to be one of the biggest cheating scandals in the sport's history.
- In 2017 the Houston Astros won the world series. However, it came out in November of 2019 that they were using technology to steal signs during home games.
- To us, this data set is interesting because it contains (in detail) records of each game, each player, and almost all variables that could take place in this event. So, this dataset is very in depth and useful for testing hypotheses thoroughly regarding the incident. Like they said "Numbers don't lie"!

## Variables

Below is a list of some important variables in our dataset:

Name	Abbr.	Description
On base and slugging percentage	OPS	Measures a players On base Percentage (percentage of At bats a player has gotten on base) and a players slugging percentage (a weighted batting average)
On base and slugging percentage (Player)	tOPS	This is adjusted so that 100 is the team average, so if tOPS is less than 100, the batters did worse, and if it is higher than 100, the batters did better.
On base and slugging percentage (Team and Player)	sOPS	This is adjusted so that 100 is the league average, so if sOPS is less than 100, the batters did worse than the league average and if it is higher than 100, the batters did better.
Runs Batted In	RBI	The number of runs a player has created by hitting a player home. Weights a players higher
Hits	H	When a batter reaches base without doing so via error or fielder's choice.
Homeruns	HR	Scored when the ball is hit in such a way that the batter is able to circle the bases and reach home safely.
Walks	W	Occurs when a pitcher throws four pitches outside of the strike zone, none of which are swung at by the batter.
Batting Average	BA	Percentage of At Bats a player gets a hit.

## Loading Data

```
# will store the data for all teams in the 2017 season
season_2017 <- NA
data_empty <- TRUE

# Getting initial path data
data_path <- "../data"
data_dirs <- list.files(data_path)
data_dirs <- data_dirs[!str_detect(data_dirs, ".csv")]

# iteratively get all csv file paths and store them
for(path in data_dirs){
```

```

# gets the directory for a specific teams csv files
team_dir <- paste(data_path, path, sep="/")
csv_list <- list.files(team_dir)

# steps through csv files, loads and edits them as dataframes
for(csv in csv_list){
  # get full path, and other information from csv name
  full_path <- paste(team_dir, csv, sep="/")
  split <- ifelse(str_detect(csv, "home"), "HOME", "AWAY")
  team_name <- str_split_fixed(team_dir, pattern="data/", 2)[2]

  # Read in data
  df <- read.csv(full_path)%>%
    mutate(Name = str_split_fixed(Name, "\\\\", 2)[,1]) %>%
    mutate(split = split, team = team_name)
  df <- df[ -length(df[,1]), ]

  # binds all teams data together
  if(data_empty){
    season_2017 <- df
    data_empty <- FALSE
  }else {
    season_2017 <- rbind(season_2017, df, make.row.names=TRUE)
  }
}
}
astros_2017 <- season_2017 %>%
  filter(team=="astros", AB != 0)

# Loading Data pt 2
# btw this sucked
wins_and_losses_2017 <- read.csv("./data/win_loss_2017.csv") %>%
  mutate(split = gsub(" ", "", split, fixed = TRUE), team = gsub(" ", "", team, fixed = TRUE)) %>%
  mutate(split = toupper(split))

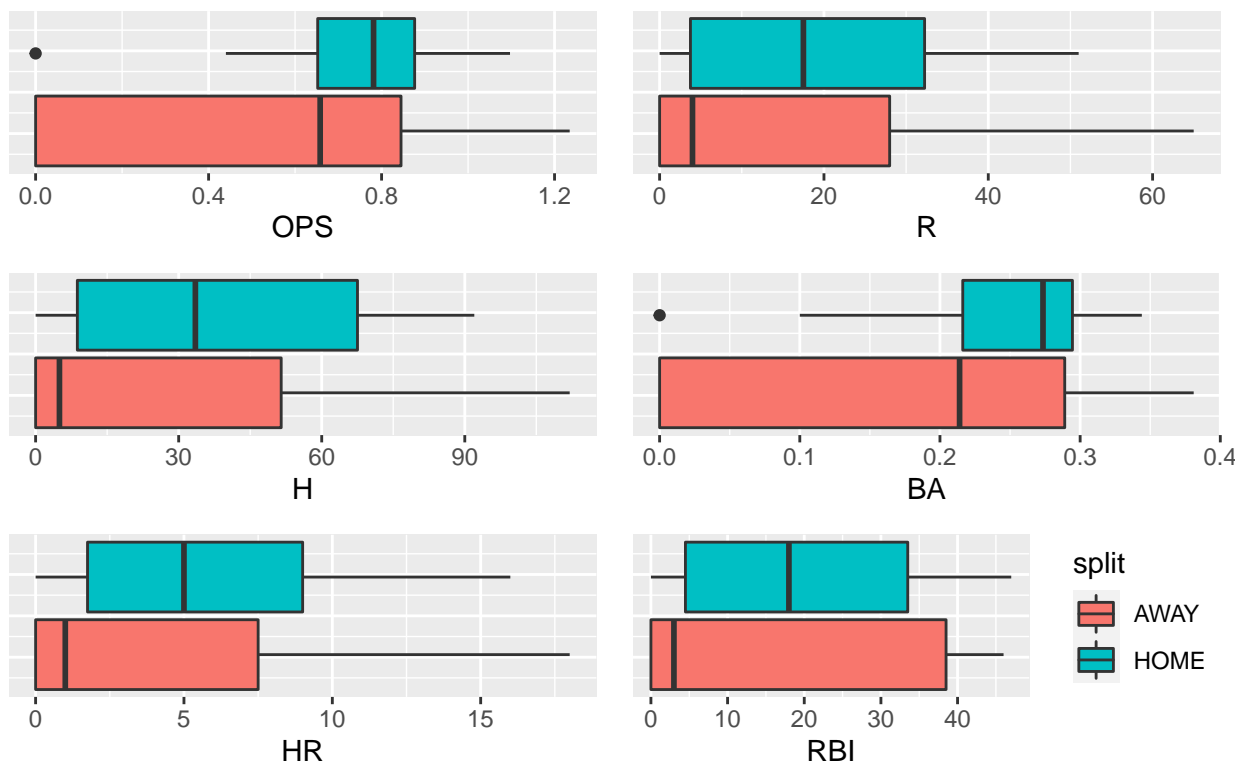
```

## Preliminary Plots

### *Comparing The Astros 2017 Home and Away Data*

to start we will look at some of the variables we take a look some comparisons of specific variables between home and away games.

## Astros 2017 data – Home vs Away



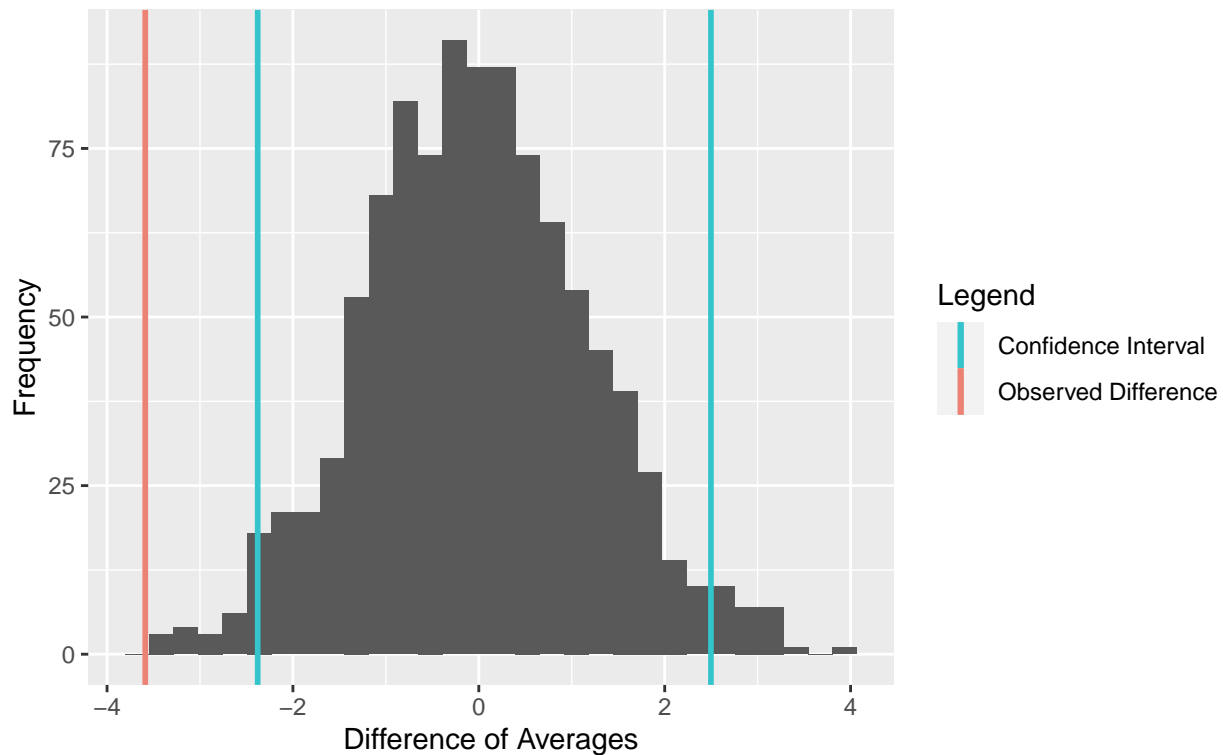
These box plots show how there is a clear difference between home and away games. The largest difference is in average Hits (H) per season that a player gets between home and away games.

### *Runs analysis - Home vs. Away*

To do this analysis, we will be doing a monte carlo simulation on the difference of average runs a player gets per season between home and away games. Runs will be modeled by a poisson random variable. Under the assumption that Runs from home and away games comes from the same distribution, we will set lambda as the mean of runs for home and away games combined.

## Difference of Average Player Runs

Home vs. Away – Astros 2017



After performing a 95% confidence interval on our data, we see that our observed difference falls outside of the confidence interval, leading us to reject our initial assumption. Thus, we can say with 95% confidence that there is statistical significance in the difference of Run means between home and away games.

However, because Away game Runs are lower than Home game Runs, this implies that the Astros cheating had a positive effect on their performance.

### *Is It Just a Home-Field Advantage*

Since our original statistical question is basically “can we prove that the Astros cheated in the 2017 season?” we want to determine if the results we found in the confidence interval above just prove that a home field advantage exists or if the Astros have a better home field advantage due to the existence of cheating.

To do this we will perform some two sample t tests to determine a difference between the average statistics between Astros at home and other teams at their home stadium. We have realized that sign stealing should have a higher effect on players batting averages because it allowed the astros players to know which kind of pitch was coming in and thus have a better chance at hitting the ball. Batting Average is calculated as  $BA = Hits/AtBats$ .

```
##  
## Welch Two Sample t-test
```

```
##
## data:  astros_home$BA and astros_away$BA
## t = 2.2676, df = 48.977, p-value = 0.0278
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  0.007848595 0.130125598
## sample estimates:
## mean of x mean of y
## 0.2466000 0.1776129
```

Above we see that the batting average of the Astros is significantly better at home than away. This again does not show conclusive evidence that they cheated, but instead adds strength to our earlier analysis that the Astros definitely have some sort of a home field advantage. Next we will compare the Astros with the rest of the league, by performing another two sample t test between the astros home games and the home games of all other teams in the league.

```
##
## Welch Two Sample t-test
##
## data:  astros_home$BA and c$BA
## t = 1.6987, df = 21.411, p-value = 0.1039
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.007287044 0.072700567
## sample estimates:
## mean of x mean of y
## 0.2466000 0.2138932
```

Here the results do not lead us anywhere conclusive as our p-value is too high to reject the hypothesis that the true mean of the distribution of the Astros batting average is the same as the true mean of the batting average of all other teams. We have also tried this in other variables in our data set as well, however the results have all been similar. There is no distinction between the astros home statistics versus other teams home statistics.

## A New Approach

«««< HEAD So far it seems that our data is too general for our statistical question. There are a myriad of different factors that play into how a team does in a season, differing game to game. We believe that to tackle this question we need to take a look at more in-depth statistics than just player hits, batting average, etc. We have found a new dataset from “Baseball Savant” that offer’s player data for the percent of pitches that are out of the strike zone that a player swings at and the percent of pitches that are in the strike zone that a player swings at.

```
test = data.frame(
  player= c("a", "b", "c", "d", "e", "f", "g", "h", "i", "j", "k", "l", "m",
            "a", "b", "c", "d", "e", "f", "g", "h", "i", "j", "k", "l", "m"),
  IZP = c(.1, .2, .3, .4, .5, .6, .7, .8, .9, .8, .14, .35, .66,
          .2, .25, .35, .45, .55, .65, .75, .85, .95, .85, .145, .355, .56),
  OZP = c(.15, .25, .35, .45, .55, .65, .76, .86, .96, .86, .146, .356, .65,
          .2, .28, .38, .48, .58, .68, .75, .85, .95, .85, .145, .355, .56),
  year = c("2017", "2017", "2017", "2017", "2017", "2017", "2017", "2017", "2017", "2017", "2017", "2017", "2017",
            "2016", "2016", "2016", "2016", "2016", "2016", "2016", "2016", "2016", "2016", "2016", "2016"),
```

```

team = c("A", "B", "C", "D", "E", "F", "G", "H", "I", "J", "K", "L", "M",
         "A", "B", "C", "D", "E", "F", "G", "H", "I", "J", "K", "L", "M")
)

IZP_diffs <- test %>%
  drop_na() %>%
  group_by(player) %>%
  mutate(change_IZP = (IZP[2] - IZP[1]), change_OZP = (OZP[2] - OZP[1])) %>%
  select(player, team, change_IZP, change_OZP) %>%
  unique() %>%
  arrange(desc(change_IZP)) %>%
  mutate(team = ifelse(team == "A", "astros", "not_astros"))
IZP_diffs$order <- 1:length(IZP_diffs$change_IZP)
IZP_diffs

```

```

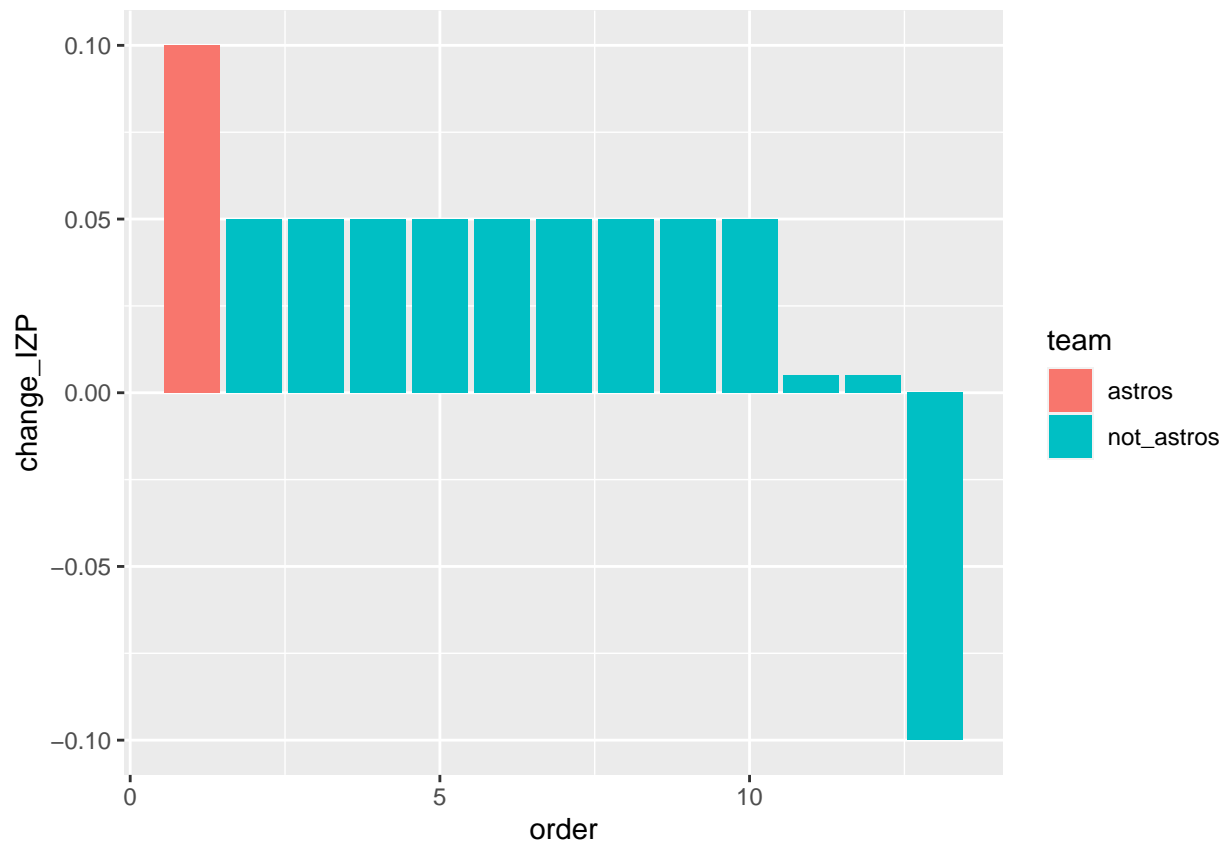
## # A tibble: 13 x 5
## # Groups:   player [13]
##   player team      change_IZP change_OZP order
##   <chr>  <chr>          <dbl>      <dbl> <int>
## 1 a      astros          0.1         0.05     1
## 2 e      not_astros      0.0500      0.0300     2
## 3 f      not_astros      0.0500      0.0300     3
## 4 g      not_astros      0.0500     -0.0100     4
## 5 b      not_astros      0.05         0.0300     5
## 6 c      not_astros      0.05         0.0300     6
## 7 d      not_astros      0.05         0.0300     7
## 8 h      not_astros      0.0500     -0.0100     8
## 9 i      not_astros      0.0500     -0.0100     9
## 10 j     not_astros      0.0500     -0.0100    10
## 11 l     not_astros      0.00500    -0.00100    11
## 12 k     not_astros      0.00500    -0.00100    12
## 13 m     not_astros     -0.1        -0.09     13

```

```

IZP_diffs %>%
  ggplot() +
  geom_col(aes(x=order, y=change_IZP, fill = team))

```



```

zones <- read.csv('./data/zone_swings.csv')
zones_clean <- zones %>%
  unite('Name', first_name:last_name, remove = TRUE) %>%
  mutate(Name = str_replace(Name, "_", " ")) %>%
  select(Name, out_zone_swing, in_zone_swing) %>%
  arrange(Name)

astros_zones_2017 <- merge(astros_2017, zones_clean, by = "Name", all = TRUE)
astros_zones_2017

```

##	Name	Rk	G	GS	PA	AB	R	H	X2B	X3B	HR	RBI	SB	CS	BB	SO
## 1	Aaron Judge	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 2	Adam Duvall	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 3	Adam Jones	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 4	Albert Pujols	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 5	Alcides Escobar	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 6	Alex Bregman	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 7	Alex Gordon	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 8	Andrelton Simmons	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 9	Andrew Benintendi	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 10	Andrew McCutchen	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 11	Anthony Rendon	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 12	Anthony Rizzo	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 13	Asdrubal Cabrera	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 14	Avisail Garcia	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA



## 15	Ben Gamel	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 16	Billy Hamilton	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 17	Brandon Crawford	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 18	Brandon Phillips	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 19	Brett Gardner	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 20	Brian Dozier	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 21	Buster Posey III	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 22	Byron Buxton	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 23	Carlos Beltran	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 24	Carlos Gonzalez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 25	Carlos Santana	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 26	Cesar Hernandez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 27	Charlie Blackmon	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 28	Chase Headley	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 29	Chris Davis	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 30	Chris Taylor	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 31	Christian Yelich	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 32	Cody Bellinger	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 33	Corey Dickerson	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 34	Corey Seager	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 35	Curtis Granderson	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 36	Daniel Murphy	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 37	Dansby Swanson	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 38	David Freese	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 39	David Peralta	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 40	Dee Strange-Gordon	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 41	Denard Span	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 42	Didi Gregorius	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 43	DJ LeMahieu	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 44	Domingo Santana	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 45	Eddie Rosario	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 46	Edwin Encarnacion	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 47	Elvis Andrus	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 48	Ender Inciarte	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 49	Eric Hosmer	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 50	Eric Thames	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 51	Eugenio Suarez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 52	Evan Longoria	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 53	Francisco Lindor	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 54	Freddie Freeman	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 55	Freddy Galvis	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 56	Gary Sanchez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 57	George Springer III	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 58	Giancarlo Stanton	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 59	Hanley Ramirez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 60	Hunter Pence	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 61	Ian Kinsler	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 62	J.T. Realmuto	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 63	Jackie Bradley Jr.	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 64	Jake Lamb	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 65	Javier Baez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 66	Jay Bruce	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 67	Jean Segura	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 68	Jed Lowrie	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA

## 69	Joe Mauer	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 70	Joe Panik	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 71	Joey Gallo	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 72	Joey Votto	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 73	Jonathan Schoop	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 74	Jordy Mercer	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 75	Jorge Polanco	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 76	Jose Abreu	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 77	Jose Altuve	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 78	Jose Bautista	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 79	Jose Peraza	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 80	Jose Ramirez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 81	Jose Reyes	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 82	Josh Bell	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 83	Josh Harrison	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 84	Josh Reddick	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 85	Justin Smoak	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 86	Justin Turner	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 87	Justin Upton	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 88	Kendrys Morales	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 89	Kevin Pillar	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 90	Khris Davis	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 91	Kole Calhoun	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 92	Kris Bryant	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 93	Kyle Seager	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 94	Logan Morrison	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 95	Lorenzo Cain	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 96	Maikel Franco	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 97	Manny Machado	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 98	Manuel Margot	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 99	Marcell Ozuna	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 100	Mark Reynolds	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 101	Mark Trumbo	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 102	Marwin Gonzalez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 103	Matt Carpenter	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 104	Matt Joyce	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 105	Max Kepler	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 106	Melky Cabrera	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 107	Miguel Cabrera	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 108	Mike Moustakas	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 109	Mike Trout	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 110	Mitch Moreland	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 111	Mookie Betts	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 112	Nelson Cruz Jr.	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 113	Nick Castellanos	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 114	Nick Markakis	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 115	Nolan Arenado	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 116	Nomar Mazara	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 117	Odubel Herrera	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 118	Orlando Arcia	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 119	Paul Goldschmidt	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 120	Robinson Cano	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 121	Roughed Odor	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 122	Ryan Zimmerman	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA

## 123	Ryon Healy	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 124	Scott Schebler	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 125	Shin-Soo Choo	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 126	Steven Souza Jr.	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 127	Tim Anderson	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 128	Tim Beckham	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 129	Todd Frazier	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 130	Tommy Joseph	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 131	Tommy Pham	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 132	Travis Shaw	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 133	Trevor Story	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 134	Trey Mancini III	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 135	Whit Merrifield	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 136	Wil Myers	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 137	Xander Bogaerts	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 138	Yadier Molina	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 139	Yangervis Solarte	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 140	Yasiel Puig	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 141	Yolmer Sanchez	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 142	Yonder Alonso	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 143	Yuli Gurriel	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 144	Zack Cozart	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 145	AJ Reed	20	2	1	6	6	0	0	0	0	0	0	0	0	0	1
## 146	Alex Bregman	7	79	75	328	290	46	84	26	3	10	41	7	3	28	53
## 147	Alex Bregman	10	76	73	298	266	42	74	13	2	9	30	10	2	27	44
## 148	Brad Peacock	11	4	3	8	7	1	2	1	0	0	2	0	0	1	3
## 149	Brian McCann	8	52	51	215	188	28	50	10	1	11	43	1	0	21	34
## 150	Brian McCann	15	45	45	184	161	19	34	2	0	7	19	0	0	17	24
## 151	Cameron Maybin	10	13	8	38	34	5	7	1	1	3	8	1	2	3	9
## 152	Cameron Maybin	19	8	6	25	25	1	4	0	0	1	5	3	1	0	7
## 153	Carlos Beltran	18	62	57	254	234	28	53	14	0	6	26	0	0	17	54
## 154	Carlos Beltran	13	67	61	255	233	32	55	15	0	8	25	0	0	16	48
## 155	Carlos Correa	4	58	58	269	236	51	71	12	1	13	44	1	1	30	61
## 156	Carlos Correa	3	51	51	212	186	31	62	13	0	11	40	1	0	23	31
## 157	Charlie Morton	24	1	1	3	3	0	0	0	0	0	0	0	0	0	0
## 158	Chris Devenski	25	3	0	1	1	0	0	0	0	0	0	0	0	0	1
## 159	Colin Moran	1	7	2	12	11	3	4	0	1	1	3	0	0	1	1
## 160	Collin McHugh	23	1	1	2	2	0	0	0	0	0	0	0	0	0	2
## 161	Dallas Keuchel	27	1	1	3	3	0	0	0	0	0	0	0	0	0	2
## 162	Derek Fisher	16	27	17	74	63	10	13	0	1	2	6	0	1	9	24
## 163	Derek Fisher	14	26	22	92	83	11	18	4	0	3	11	3	2	8	30
## 164	Evan Gattis	14	41	37	165	154	25	39	9	0	8	36	0	1	5	24
## 165	Evan Gattis	11	43	39	160	146	16	40	13	0	4	19	0	0	13	26
## 166	George Springer	5	69	66	313	277	61	81	13	0	18	43	3	2	29	60
## 167	George Springer	6	71	71	316	271	51	74	16	0	16	42	2	5	35	51
## 168	J.D. Davis	15	13	7	36	31	5	5	2	0	2	4	0	0	4	12
## 169	J.D. Davis	8	11	7	32	31	3	9	2	0	2	3	1	1	0	8
## 170	Jake Marisnick	17	58	32	139	123	26	23	5	0	6	18	3	2	12	55
## 171	Jake Marisnick	2	48	33	120	107	24	33	5	0	10	17	6	2	8	35
## 172	James Hoyt	30	3	0	1	1	0	0	0	0	0	0	0	0	0	1
## 173	Joe Musgrove	29	4	2	5	5	0	0	0	0	0	0	0	0	0	1
## 174	Jose Altuve	2	75	73	333	294	65	112	23	3	15	46	13	3	34	32
## 175	Jose Altuve	7	78	78	329	296	47	92	16	1	9	35	19	3	24	52
## 176	Josh Reddick	13	65	60	283	250	44	72	16	1	7	46	3	0	25	38

## 177	Josh Reddick	4	69	61	257	227	33	78	18	3	6	36	4	3	18	34
## 178	Juan Centeno	21	11	5	20	20	1	4	0	0	0	1	0	0	0	8
## 179	Juan Centeno	12	11	10	37	32	4	8	0	0	2	3	0	0	4	4
## 180	Lance McCullers Jr.	22	1	1	3	3	0	0	0	0	0	0	0	0	0	0
## 181	Marwin Gonzalez	3	64	58	256	221	32	72	18	0	8	43	4	2	29	38
## 182	Marwin Gonzalez	5	70	62	259	234	35	66	16	0	15	47	4	1	20	61
## 183	Max Stassi	6	9	3	19	14	4	3	1	0	1	3	0	0	4	3
## 184	Max Stassi	16	5	3	12	10	1	1	0	0	1	1	0	0	2	1
## 185	Michael Feliz	28	3	0	1	1	0	0	0	0	0	0	0	0	0	1
## 186	Mike Fiers	31	2	1	1	1	1	0	0	0	0	0	0	0	0	1
## 187	Nori Aoki	12	38	29	126	112	20	34	7	1	2	12	4	1	10	16
## 188	Nori Aoki	17	32	27	98	90	8	21	5	0	0	7	1	1	5	13
## 189	Tony Kemp	20	11	6	25	23	4	5	0	0	0	3	1	0	1	5
## 190	Tony Kemp	18	6	4	14	14	2	3	1	0	0	1	0	0	0	0
## 191	Tony Sipp	26	5	0	1	1	0	0	0	0	0	0	0	0	0	0
## 192	Tyler White	19	12	5	34	31	2	7	3	0	0	2	0	0	3	10
## 193	Tyler White	1	10	8	33	30	5	10	3	0	3	8	0	1	1	6
## 194	Yuli Gurriel	9	70	70	294	276	39	86	20	1	10	42	2	2	12	36
## 195	Yuli Gurriel	9	69	67	270	253	30	72	23	0	8	33	1	0	10	26
##	BA	OBP	SLG	OPS	TB	GDP	HBP	SH	SF	IBB	ROE	BAbip	tOPS.	sOPS.	split	
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[illegible]

[illegible]

##	143	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	<NA>
##	144	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	<NA>
##	145	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0.000	NA	-100	HOME	
##	146	0.290	0.360	0.503	0.863	146	7	6	0	4	2	3	0.320	108	134	AWAY
##	147	0.278	0.343	0.444	0.787	118	8	1	1	3	0	1	0.301	91	105	HOME
##	148	0.286	0.375	0.429	0.804	3	0	0	0	0	0	0	0.500	100	121	AWAY
##	149	0.266	0.344	0.505	0.850	95	3	3	0	3	1	1	0.267	122	130	AWAY
##	150	0.211	0.299	0.354	0.653	57	6	4	0	2	2	3	0.205	74	71	HOME
##	151	0.206	0.270	0.559	0.829	19	1	0	1	0	0	0	0.182	138	120	AWAY
##	152	0.160	0.160	0.280	0.440	7	0	0	0	0	0	2	0.176	27	12	HOME
##	153	0.226	0.280	0.363	0.643	85	4	1	0	2	1	3	0.267	94	76	AWAY
##	154	0.236	0.286	0.403	0.690	94	5	2	0	4	2	3	0.260	106	78	HOME
##	155	0.301	0.379	0.525	0.905	124	6	1	0	2	1	8	0.354	92	146	AWAY
##	156	0.333	0.406	0.581	0.986	108	6	1	0	2	4	1	0.349	110	155	HOME
##	157	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY
##	158	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY
##	159	0.364	0.417	0.818	1.235	9	0	0	0	0	0	0	0.333	100	228	AWAY
##	160	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY
##	161	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY
##	162	0.206	0.324	0.333	0.658	21	0	2	0	0	0	1	0.297	99	82	AWAY
##	163	0.217	0.293	0.373	0.667	31	1	1	0	0	1	1	0.300	100	73	HOME
##	164	0.253	0.291	0.468	0.758	72	4	4	0	2	0	1	0.250	96	104	AWAY
##	165	0.274	0.331	0.445	0.776	65	6	0	0	1	0	0	0.308	104	101	HOME
##	166	0.292	0.367	0.534	0.902	148	4	5	0	2	1	3	0.313	102	144	AWAY
##	167	0.273	0.366	0.509	0.875	138	7	6	0	2	0	5	0.282	97	127	HOME
##	168	0.161	0.278	0.419	0.697	13	3	1	0	0	0	0	0.176	86	88	AWAY
##	169	0.290	0.281	0.548	0.830	17	0	0	0	1	0	1	0.318	114	110	HOME
##	170	0.187	0.273	0.374	0.647	46	5	3	0	1	1	1	0.270	61	76	AWAY
##	171	0.308	0.373	0.636	1.008	68	0	3	2	0	0	0	0.371	145	158	HOME
##	172	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY
##	173	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY
##	174	0.381	0.449	0.633	1.081	186	6	3	1	1	2	1	0.391	125	194	AWAY
##	175	0.311	0.371	0.463	0.834	137	13	6	0	3	1	3	0.349	75	118	HOME
##	176	0.288	0.346	0.444	0.790	111	4	0	0	5	1	4	0.310	87	116	AWAY
##	177	0.344	0.381	0.529	0.910	120	5	0	1	7	0	2	0.371	114	136	HOME
##	178	0.200	0.200	0.200	0.400	4	0	0	0	0	0	0	0.333	28	11	AWAY
##	179	0.250	0.333	0.438	0.771	14	2	0	1	0	1	0	0.231	143	100	HOME
##	180	0.000	0.000	0.000	0.000	0	0	0	0	0	0	1	0.000	NA	-100	AWAY
##	181	0.326	0.416	0.516	0.932	114	7	5	1	0	4	3	0.366	108	155	AWAY
##	182	0.282	0.339	0.543	0.881	127	1	1	2	2	0	0	0.319	92	126	HOME
##	183	0.214	0.368	0.500	0.868	7	1	0	0	1	0	0	0.182	123	136	AWAY
##	184	0.100	0.250	0.400	0.650	4	1	0	0	0	0	0	0.000	65	67	HOME
##	185	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY
##	186	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY
##	187	0.304	0.357	0.438	0.795	49	5	1	0	3	1	0	0.330	118	118	AWAY
##	188	0.233	0.278	0.289	0.567	26	6	1	1	1	0	2	0.269	57	50	HOME
##	189	0.217	0.280	0.217	0.497	5	0	1	0	0	0	0	0.278	99	40	AWAY
##	190	0.214	0.214	0.286	0.500	4	0	0	0	0	0	0	0.214	101	30	HOME
##	191	0.000	0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY
##	192	0.226	0.294	0.323	0.617	10	0	0	0	0	0	0	0.333	51	70	AWAY
##	193	0.333	0.364	0.733	1.097	22	0	1	0	1	0	1	0.318	151	177	HOME
##	194	0.312	0.340	0.500	0.840	138	7	2	0	4	0	2	0.325	105	127	AWAY
##	195	0.285	0.322	0.470	0.793	119	5	5	0	2	1	5	0.290	94	104	HOME
##		team	out_zone_swing	in_zone_swing												

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## 2	<NA>	400	778
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## 133	<NA>	296	786
## 134	<NA>	376	756
## 135	<NA>	372	826
## 136	<NA>	344	813
## 137	<NA>	365	672
## 138	<NA>	310	717
## 139	<NA>	310	584
## 140	<NA>	295	683
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zones\_clean

##	Name	out_zone_swing	in_zone_swing
## 1	Aaron Judge	382	816
## 2	Adam Duvall	400	778
## 3	Adam Jones	520	783
## 4	Albert Pujols	416	750
## 5	Alcides Escobar	404	795
## 6	Alex Bregman	280	746
## 7	Alex Gordon	278	760
## 8	Andrelton Simmons	310	739
## 9	Andrew Benintendi	326	754
## 10	Andrew McCutchen	258	875
## 11	Anthony Rendon	243	813
## 12	Anthony Rizzo	390	849
## 13	Asdrubal Cabrera	270	690
## 14	Avisail Garcia	397	766
## 15	Ben Gamel	284	681
## 16	Billy Hamilton	328	830
## 17	Brandon Crawford	324	790

## 18	Brandon Phillips	439	731
## 19	Brett Gardner	328	707
## 20	Brian Dozier	383	795
## 21	Buster Posey III	245	667
## 22	Byron Buxton	312	669
## 23	Carlos Beltran	340	581
## 24	Carlos Gonzalez	330	623
## 25	Carlos Santana	264	792
## 26	Cesar Hernandez	235	715
## 27	Charlie Blackmon	375	932
## 28	Chase Headley	286	731
## 29	Chris Davis	285	597
## 30	Chris Taylor	244	786
## 31	Christian Yelich	346	867
## 32	Cody Bellinger	270	708
## 33	Corey Dickerson	581	859
## 34	Corey Seager	306	815
## 35	Curtis Granderson	238	656
## 36	Daniel Murphy	354	668
## 37	Dansby Swanson	281	690
## 38	David Freese	240	632
## 39	David Peralta	320	703
## 40	Dee Strange-Gordon	392	895
## 41	Denard Span	234	660
## 42	Didi Gregorius	409	786
## 43	DJ LeMahieu	275	827
## 44	Domingo Santana	312	828
## 45	Eddie Rosario	416	777
## 46	Edwin Encarnacion	317	914
## 47	Elvis Andrus	404	800
## 48	Ender Inciarte	458	883
## 49	Eric Hosmer	365	843
## 50	Eric Thames	303	698
## 51	Eugenio Suarez	293	824
## 52	Evan Longoria	409	812
## 53	Francisco Lindor	412	876
## 54	Freddie Freeman	302	700
## 55	Freddy Galvis	436	845
## 56	Gary Sanchez	352	607
## 57	George Springer III	283	849
## 58	Giancarlo Stanton	370	795
## 59	Hanley Ramirez	307	679
## 60	Hunter Pence	329	625
## 61	Ian Kinsler	296	743
## 62	J.T. Realmuto	299	744
## 63	Jackie Bradley Jr.	244	684
## 64	Jake Lamb	267	730
## 65	Javier Baez	428	589
## 66	Jay Bruce	381	821
## 67	Jean Segura	331	717
## 68	Jed Lowrie	285	948
## 69	Joe Mauer	244	703
## 70	Joe Panik	255	681
## 71	Joey Gallo	364	690

## 72	Joey Votto	177	952
## 73	Jonathan Schoop	445	861
## 74	Jordy Mercer	263	623
## 75	Jorge Polanco	242	665
## 76	Jose Abreu	403	812
## 77	Jose Altuve	410	688
## 78	Jose Bautista	360	836
## 79	Jose Peraza	303	638
## 80	Jose Ramirez	313	799
## 81	Jose Reyes	322	607
## 82	Josh Bell	329	761
## 83	Josh Harrison	322	713
## 84	Josh Reddick	311	584
## 85	Justin Smoak	330	754
## 86	Justin Turner	225	652
## 87	Justin Upton	315	845
## 88	Kendrys Morales	410	727
## 89	Kevin Pillar	400	728
## 90	Khris Davis	293	919
## 91	Kole Calhoun	395	806
## 92	Kris Bryant	348	830
## 93	Kyle Seager	297	776
## 94	Logan Morrison	310	752
## 95	Lorenzo Cain	327	912
## 96	Maikel Franco	343	816
## 97	Manny Machado	402	818
## 98	Manuel Margot	261	648
## 99	Marcell Ozuna	412	847
## 100	Mark Reynolds	281	796
## 101	Mark Trumbo	390	731
## 102	Marwin Gonzalez	307	589
## 103	Matt Carpenter	196	729
## 104	Matt Joyce	216	733
## 105	Max Kepler	292	630
## 106	Melky Cabrera	395	710
## 107	Miguel Cabrera	319	685
## 108	Mike Moustakas	475	770
## 109	Mike Trout	210	591
## 110	Mitch Moreland	324	692
## 111	Mookie Betts	268	704
## 112	Nelson Cruz Jr.	369	773
## 113	Nick Castellanos	430	870
## 114	Nick Markakis	320	774
## 115	Nolan Arenado	386	857
## 116	Nomar Mazara	420	705
## 117	Odubel Herrera	444	705
## 118	Orlando Arcia	346	720
## 119	Paul Goldschmidt	338	774
## 120	Robinson Cano	358	817
## 121	Rougned Odor	488	781
## 122	Ryan Zimmerman	347	626
## 123	Ryon Healy	393	677
## 124	Scott Schebler	330	646
## 125	Shin-Soo Choo	261	809

## 126	Steven Souza Jr.	276	803
## 127	Tim Anderson	422	724
## 128	Tim Beckham	346	799
## 129	Todd Frazier	286	707
## 130	Tommy Joseph	372	618
## 131	Tommy Pham	167	626
## 132	Travis Shaw	359	719
## 133	Trevor Story	296	786
## 134	Trey Mancini III	376	756
## 135	Whit Merrifield	372	826
## 136	Wil Myers	344	813
## 137	Xander Bogaerts	365	672
## 138	Yadier Molina	310	717
## 139	Yangervis Solarte	310	584
## 140	Yasiel Puig	295	683
## 141	Yolmer Sanchez	329	670
## 142	Yonder Alonso	290	679
## 143	Yuli Gurriel	316	616
## 144	Zack Cozart	229	620

```
astros_2017 %>%
  arrange(Name)
```

##	Rk	Name	G	GS	PA	AB	R	H	X2B	X3B	HR	RBI	SB	CS	BB	SO	BA
## 1	20	AJ Reed	2	1	6	6	0	0	0	0	0	0	0	0	0	1	0.000
## 2	7	Alex Bregman	79	75	328	290	46	84	26	3	10	41	7	3	28	53	0.290
## 3	10	Alex Bregman	76	73	298	266	42	74	13	2	9	30	10	2	27	44	0.278
## 4	11	Brad Peacock	4	3	8	7	1	2	1	0	0	2	0	0	1	3	0.286
## 5	8	Brian McCann	52	51	215	188	28	50	10	1	11	43	1	0	21	34	0.266
## 6	15	Brian McCann	45	45	184	161	19	34	2	0	7	19	0	0	17	24	0.211
## 7	10	Cameron Maybin	13	8	38	34	5	7	1	1	3	8	1	2	3	9	0.206
## 8	19	Cameron Maybin	8	6	25	25	1	4	0	0	1	5	3	1	0	7	0.160
## 9	18	Carlos Beltran	62	57	254	234	28	53	14	0	6	26	0	0	17	54	0.226
## 10	13	Carlos Beltran	67	61	255	233	32	55	15	0	8	25	0	0	16	48	0.236
## 11	4	Carlos Correa	58	58	269	236	51	71	12	1	13	44	1	1	30	61	0.301
## 12	3	Carlos Correa	51	51	212	186	31	62	13	0	11	40	1	0	23	31	0.333
## 13	24	Charlie Morton	1	1	3	3	0	0	0	0	0	0	0	0	0	0	0.000
## 14	25	Chris Devenski	3	0	1	1	0	0	0	0	0	0	0	0	0	1	0.000
## 15	1	Colin Moran	7	2	12	11	3	4	0	1	1	3	0	0	1	1	0.364
## 16	23	Collin McHugh	1	1	2	2	0	0	0	0	0	0	0	0	0	2	0.000
## 17	27	Dallas Keuchel	1	1	3	3	0	0	0	0	0	0	0	0	0	2	0.000
## 18	16	Derek Fisher	27	17	74	63	10	13	0	1	2	6	0	1	9	24	0.206
## 19	14	Derek Fisher	26	22	92	83	11	18	4	0	3	11	3	2	8	30	0.217
## 20	14	Evan Gattis	41	37	165	154	25	39	9	0	8	36	0	1	5	24	0.253
## 21	11	Evan Gattis	43	39	160	146	16	40	13	0	4	19	0	0	13	26	0.274
## 22	5	George Springer	69	66	313	277	61	81	13	0	18	43	3	2	29	60	0.292
## 23	6	George Springer	71	71	316	271	51	74	16	0	16	42	2	5	35	51	0.273
## 24	15	J.D. Davis	13	7	36	31	5	5	2	0	2	4	0	0	4	12	0.161
## 25	8	J.D. Davis	11	7	32	31	3	9	2	0	2	3	1	1	0	8	0.290
## 26	17	Jake Marisnick	58	32	139	123	26	23	5	0	6	18	3	2	12	55	0.187
## 27	2	Jake Marisnick	48	33	120	107	24	33	5	0	10	17	6	2	8	35	0.308
## 28	30	James Hoyt	3	0	1	1	0	0	0	0	0	0	0	0	0	1	0.000
## 29	29	Joe Musgrove	4	2	5	5	0	0	0	0	0	0	0	0	0	1	0.000
## 30	2	Jose Altuve	75	73	333	294	65	112	23	3	15	46	13	3	34	32	0.381

## 31	7	Jose Altuve	78	78	329	296	47	92	16	1	9	35	19	3	24	52	0.311
## 32	13	Josh Reddick	65	60	283	250	44	72	16	1	7	46	3	0	25	38	0.288
## 33	4	Josh Reddick	69	61	257	227	33	78	18	3	6	36	4	3	18	34	0.344
## 34	21	Juan Centeno	11	5	20	20	1	4	0	0	0	1	0	0	0	8	0.200
## 35	12	Juan Centeno	11	10	37	32	4	8	0	0	2	3	0	0	4	4	0.250
## 36	22	Lance McCullers Jr.	1	1	3	3	0	0	0	0	0	0	0	0	0	0	0.000
## 37	3	Marwin Gonzalez	64	58	256	221	32	72	18	0	8	43	4	2	29	38	0.326
## 38	5	Marwin Gonzalez	70	62	259	234	35	66	16	0	15	47	4	1	20	61	0.282
## 39	6	Max Stassi	9	3	19	14	4	3	1	0	1	3	0	0	4	3	0.214
## 40	16	Max Stassi	5	3	12	10	1	1	0	0	1	1	0	0	2	1	0.100
## 41	28	Michael Feliz	3	0	1	1	0	0	0	0	0	0	0	0	0	1	0.000
## 42	31	Mike Fiers	2	1	1	1	1	0	0	0	0	0	0	0	0	1	0.000
## 43	12	Nori Aoki	38	29	126	112	20	34	7	1	2	12	4	1	10	16	0.304
## 44	17	Nori Aoki	32	27	98	90	8	21	5	0	0	7	1	1	5	13	0.233
## 45	20	Tony Kemp	11	6	25	23	4	5	0	0	0	3	1	0	1	5	0.217
## 46	18	Tony Kemp	6	4	14	14	2	3	1	0	0	1	0	0	0	0	0.214
## 47	26	Tony Sipp	5	0	1	1	0	0	0	0	0	0	0	0	0	0	0.000
## 48	19	Tyler White	12	5	34	31	2	7	3	0	0	2	0	0	3	10	0.226
## 49	1	Tyler White	10	8	33	30	5	10	3	0	3	8	0	1	1	6	0.333
## 50	9	Yuli Gurriel	70	70	294	276	39	86	20	1	10	42	2	2	12	36	0.312
## 51	9	Yuli Gurriel	69	67	270	253	30	72	23	0	8	33	1	0	10	26	0.285
##		OBP	SLG	OPS	TB	GDP	HBP	SH	SF	IBB	ROE	BAbip	tOPS.	sOPS.	split	team	
## 1		0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	HOME	astros	
## 2		0.360	0.503	0.863	146	7	6	0	4	2	3	0.320	108	134	AWAY	astros	
## 3		0.343	0.444	0.787	118	8	1	1	3	0	1	0.301	91	105	HOME	astros	
## 4		0.375	0.429	0.804	3	0	0	0	0	0	0	0.500	100	121	AWAY	astros	
## 5		0.344	0.505	0.850	95	3	3	0	3	1	1	0.267	122	130	AWAY	astros	
## 6		0.299	0.354	0.653	57	6	4	0	2	2	3	0.205	74	71	HOME	astros	
## 7		0.270	0.559	0.829	19	1	0	1	0	0	0	0.182	138	120	AWAY	astros	
## 8		0.160	0.280	0.440	7	0	0	0	0	0	2	0.176	27	12	HOME	astros	
## 9		0.280	0.363	0.643	85	4	1	0	2	1	3	0.267	94	76	AWAY	astros	
## 10		0.286	0.403	0.690	94	5	2	0	4	2	3	0.260	106	78	HOME	astros	
## 11		0.379	0.525	0.905	124	6	1	0	2	1	8	0.354	92	146	AWAY	astros	
## 12		0.406	0.581	0.986	108	6	1	0	2	4	1	0.349	110	155	HOME	astros	
## 13		0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY	astros	
## 14		0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY	astros	
## 15		0.417	0.818	1.235	9	0	0	0	0	0	0	0.333	100	228	AWAY	astros	
## 16		0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY	astros	
## 17		0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY	astros	
## 18		0.324	0.333	0.658	21	0	2	0	0	0	1	0.297	99	82	AWAY	astros	
## 19		0.293	0.373	0.667	31	1	1	0	0	1	1	0.300	100	73	HOME	astros	
## 20		0.291	0.468	0.758	72	4	4	0	2	0	1	0.250	96	104	AWAY	astros	
## 21		0.331	0.445	0.776	65	6	0	0	1	0	0	0.308	104	101	HOME	astros	
## 22		0.367	0.534	0.902	148	4	5	0	2	1	3	0.313	102	144	AWAY	astros	
## 23		0.366	0.509	0.875	138	7	6	0	2	0	5	0.282	97	127	HOME	astros	
## 24		0.278	0.419	0.697	13	3	1	0	0	0	0	0.176	86	88	AWAY	astros	
## 25		0.281	0.548	0.830	17	0	0	0	1	0	1	0.318	114	110	HOME	astros	
## 26		0.273	0.374	0.647	46	5	3	0	1	1	1	0.270	61	76	AWAY	astros	
## 27		0.373	0.636	1.008	68	0	3	2	0	0	0	0.371	145	158	HOME	astros	
## 28		0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY	astros	
## 29		0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY	astros	
## 30		0.449	0.633	1.081	186	6	3	1	1	2	1	0.391	125	194	AWAY	astros	
## 31		0.371	0.463	0.834	137	13	6	0	3	1	3	0.349	75	118	HOME	astros	
## 32		0.346	0.444	0.790	111	4	0	0	5	1	4	0.310	87	116	AWAY	astros	

##	33	0.381	0.529	0.910	120	5	0	1	7	0	2	0.371	114	136	HOME	astros
##	34	0.200	0.200	0.400	4	0	0	0	0	0	0	0.333	28	11	AWAY	astros
##	35	0.333	0.438	0.771	14	2	0	1	0	1	0	0.231	143	100	HOME	astros
##	36	0.000	0.000	0.000	0	0	0	0	0	0	1	0.000	NA	-100	AWAY	astros
##	37	0.416	0.516	0.932	114	7	5	1	0	4	3	0.366	108	155	AWAY	astros
##	38	0.339	0.543	0.881	127	1	1	2	2	0	0	0.319	92	126	HOME	astros
##	39	0.368	0.500	0.868	7	1	0	0	1	0	0	0.182	123	136	AWAY	astros
##	40	0.250	0.400	0.650	4	1	0	0	0	0	0	0.000	65	67	HOME	astros
##	41	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY	astros
##	42	0.000	0.000	0.000	0	0	0	0	0	0	0	NA	NA	-100	AWAY	astros
##	43	0.357	0.438	0.795	49	5	1	0	3	1	0	0.330	118	118	AWAY	astros
##	44	0.278	0.289	0.567	26	6	1	1	1	0	2	0.269	57	50	HOME	astros
##	45	0.280	0.217	0.497	5	0	1	0	0	0	0	0.278	99	40	AWAY	astros
##	46	0.214	0.286	0.500	4	0	0	0	0	0	0	0.214	101	30	HOME	astros
##	47	0.000	0.000	0.000	0	0	0	0	0	0	0	0.000	NA	-100	AWAY	astros
##	48	0.294	0.323	0.617	10	0	0	0	0	0	0	0.333	51	70	AWAY	astros
##	49	0.364	0.733	1.097	22	0	1	0	1	0	1	0.318	151	177	HOME	astros
##	50	0.340	0.500	0.840	138	7	2	0	4	0	2	0.325	105	127	AWAY	astros
##	51	0.322	0.470	0.793	119	5	5	0	2	1	5	0.290	94	104	HOME	astros

## Discussion on Progress/Challenges

### Challenges:

- On our last project report, the results that we gathered from the data were not what we expected. We originally cleaned the dataset wrong and it affected the outcome, giving us reason to believe the Astros didn't cheat. After going back and changing how we read in the data, our graphs prove what we were originally expecting, and show initial evidence that they were cheating. Our original findings were very surprising because this cheating scandal wouldn't have been a scandal if it weren't for some correlation and evidence that suggest they did. This was a big challenge because after our last project report, we thought we were gonna have to go back and change our whole project. But now we are back on track and can continue the analysis of our questions.
- Another challenge we faced during this project report was trying to figure out how to use a linear model in our analysis. We originally wanted to use a regression model on all of the variables in our dataset to predict wins and then we could see if home games (the supposed cheating) was a statistically significant variable. After initially trying this, we realized this approach was not a very good/strong indication of how home games helped them cheat. Then, we thought it would be beneficial to compare individual Astro players pre-season stats, to that of the regular season, and compare home and away. This also came with its issues though because there wasn't individual player data for preseason. So we moved onto our next thought process
- From the data site Kieth gave us from the last project report, we found a perfect data set split up into the home and away games, but there were no column names for 161 variables. So we used python to assign column names from the reference sheet online.
- In doing this, we found out how hard it is to try and piece together all the datasets. This method also increases the margin of error, which is something we can compute and analyze for the final report.

### Progress:

- We came to the realization that some of our data cleaning from our last project report was skewing the results. We fixed that problem, re-fit the previous graphs, and re-analyzed them



- We spend a tremendous amount of time reading in new data and cleaning it, in order to form a regression model. After forming a linear model to predict win\_percentage based on Runs, Hits, Runs Batted In, Home Runs, and Split (Home vs, Away), we found that the model sucks, like a lot. So we will need to adapt our model to work better in the future, but we did not have enough time to finish it now.
- In experimenting with different ways to analyze home and away splits, we found that creating our own batting average would be best because we could create one for each home and away game according to the player. In doing this we also realized that teams rosters are different for home and away and this, along with players who never got up to the plate once in a whole season were skewing our results. In an attempt to resolve this, we only analyzed players who got up to bat once per game, or 162 times for all the regular season games in the pre-covid 2017 season.

## Next Steps

- We want to figure out why the new regression models we created were so bad. Before the final project is due, we will work on how to improve our model, and attend office hours to gain some insight as why they currently are the way they are. We can then plot the models to visually see how they are doing
- We also might want to look into the performance of teams throughout the league at the Astros' home stadium to see if the conditions under which teams play at that location (called park factors) have any affect on performance, indicating a possible reason for the Astros lower performance at home.
- We are also considering looking into things such as individual Astros players performance and test to see if there were any noticeable, improbable differences between years, comparing it to the change in performance for other teams' top performers.