

Baseball Data Scraping

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Part 1. Scrape baseball-reference.com with rvest

```
library(rvest)
library(dplyr)
# starting page
teampage <- read_html("http://www.baseball-reference.com/teams/")
teams <- teampage %>%
  html_nodes(".left a") %>%
  html_text()

session <- html_session("http://www.baseball-reference.com/teams/")

# create a table called baseball that contains all of the teams' franchise histories

baseball <- data.frame()
for(i in seq_along(teams)) {
  team_name <- teams[i]
  team_history <- session %>%
    follow_link(team_name) %>%
    read_html() %>%
    html_node("#franchise_years") %>%
    html_table()
  team_history$GB <- as.character(team_history$GB)
  team_history$current_name <- team_history$Tm[1]
  if(i==1) {
    baseball <- team_history
  } else {
    baseball <- full_join(baseball, team_history)
  }
}

# at the end, be sure to print out the dimensions of your baseball table
dim(baseball)

## [1] 2744 22

# also print the head of the table
head(baseball)
```

```
##      Year              Tm      Lg   G   W   L Ties   W-L%   pythW-L%   Finish
## 1 2021 Arizona Diamondbacks NL West 162 52 110    0 0.321    0.377 5th of 5
## 2 2020 Arizona Diamondbacks NL West  60 25  35    0 0.417    0.458 5th of 5
## 3 2019 Arizona Diamondbacks NL West 162 85  77    0 0.525    0.541 2nd of 5
## 4 2018 Arizona Diamondbacks NL West 162 82  80    0 0.506    0.533 3rd of 5
## 5 2017 Arizona Diamondbacks NL West 162 93  69    0 0.574    0.594 2nd of 5
## 6 2016 Arizona Diamondbacks NL West 162 69  93    0 0.426    0.424 4th of 5
##      GB      Playoffs   R  RA Attendance BatAge PAge #Bat #P
## 1 55.0                679 893   1,043,010   28.9 28.5   64 41
## 2 18.0                269 295                29.1 27.7   45 26
## 3 21.0                813 743   2,135,510   28.7 28.6   45 27
## 4  9.5                693 644   2,242,695   29.2 29.6   49 30
## 5 11.0 Lost NLDS (3-0) 812 659   2,134,375   28.3 28.7   45 23
## 6 22.0                752 890   2,036,216   26.7 26.4   50 29
##      Top Player      Managers      current_name
## 1      E.Escobar (2.3) T.Lovullo (52-110) Arizona Diamondbacks
## 2      Z.Gallen (2.5) T.Lovullo (25-35) Arizona Diamondbacks
## 3      K.Marte (6.9) T.Lovullo (85-77) Arizona Diamondbacks
## 4 P.Goldschmidt (5.5) T.Lovullo (82-80) Arizona Diamondbacks
## 5 P.Goldschmidt (6.3) T.Lovullo (93-69) Arizona Diamondbacks
## 6      J.Segura (6.4)   C.Hale (69-93) Arizona Diamondbacks
```

Some light text clean up

```
## [1] "Lengths (21, 20) differ (comparison on first 20 components)"
## [2] "13 element mismatches"

## [1] TRUE
```

Part 2. dplyr to summarize the baseball data

```
# Printing a summary table of our scraped data
```

```
baseball_summary <- baseball %>%
  filter(Year %in% 2001:2020) %>%
  group_by(current_name) %>%
  summarise("Wins" = sum(W), "Losses" = sum(L), "Runs" = sum(R), "Runs Allowed" = sum(RA), "Win Pct" = sum(W-L)/sum(W+L))
  arrange(desc('Win Pct'))
print(baseball_summary, n=30)
```

```
## # A tibble: 30 x 6
## # Groups:   current_name [30]
##   current_name      Wins Losses  Runs 'Runs Allowed' 'Win Pct'
##   <chr>          <int>  <int> <int>         <int>     <dbl>
## 1 New York Yankees    1832   1303 16187         13838     0.584
## 2 St. Louis Cardinals  1747   1388 14767         13081     0.557
## 3 Los Angeles Dodgers  1738   1400 14042         12468     0.554
## 4 Boston Red Sox      1731   1406 16249         14303     0.552
## 5 Atlanta Braves      1675   1460 14319         13274     0.534
## 6 Oakland Athletics   1674   1463 14469         13296     0.534
## 7 Los Angeles Angels  1666   1472 14604         13838     0.531
```

## 8	Cleveland Indians	1616	1520	14772	14113	0.515
## 9	San Francisco Giants	1608	1527	13471	13315	0.513
## 10	Philadelphia Phillies	1600	1537	14351	14174	0.510
## 11	Minnesota Twins	1595	1544	14582	14524	0.508
## 12	Chicago Cubs	1593	1543	14056	13528	0.508
## 13	Houston Astros	1578	1559	14103	13851	0.503
## 14	Texas Rangers	1570	1569	15523	15630	0.500
## 15	Washington Nationals	1549	1587	13734	13828	0.494
## 16	Toronto Blue Jays	1548	1589	14771	14576	0.493
## 17	New York Mets	1540	1596	13602	13752	0.491
## 18	Chicago White Sox	1540	1598	14243	14607	0.491
## 19	Arizona Diamondbacks	1538	1600	14127	14366	0.490
## 20	Seattle Mariners	1531	1607	13603	14089	0.488
## 21	Tampa Bay Rays	1525	1612	13874	14361	0.486
## 22	Milwaukee Brewers	1521	1617	13872	14445	0.485
## 23	Cincinnati Reds	1472	1666	13853	14927	0.469
## 24	Miami Marlins	1470	1665	13341	14393	0.469
## 25	San Diego Padres	1469	1670	12954	14060	0.468
## 26	Colorado Rockies	1465	1675	15371	16064	0.467
## 27	Detroit Tigers	1455	1678	14165	15273	0.464
## 28	Pittsburgh Pirates	1423	1710	13124	14644	0.454
## 29	Baltimore Orioles	1404	1732	14023	15632	0.448
## 30	Kansas City Royals	1379	1759	13622	15524	0.439

3. Regular expressions to extract values in the Managers Column

```
# Using regular expressions to extract first and last names

managers_data <- str_match_all(baseball$Managers, "([A-Z]\\.[^\\(\\)]+) \\((\\d+)-(\\d+)")

names <- character(0)
wins <- numeric(0)
losses <- numeric(0)

# Extracting the data we want from the matrices into vector form
for(i in seq_along(managers_data)){
  for(j in seq_along(1:nrow(managers_data[[i]]))){
    names <- append(names, managers_data[[i]][j,2])
    wins <- append(wins, as.numeric(managers_data[[i]][j,3]))
    losses <- append(losses, as.numeric(managers_data[[i]][j,4]))
  }
}

# Using the vectors to create a tibble, and then using dplyr to get the desired result
managers <- tibble(
  Name = names,
  Wins = wins,
  Losses = losses
)

managers %>%
  mutate(Games = Wins + Losses) %>%
  group_by(Name) %>%
```

```
summarise(Games = sum(Games), Wins = sum(Wins), Losses = sum(Losses), Win_Pct = sum(Wins)/sum(Games))
arrange(desc(Games))
```

```
## 'summarise()' ungrouping output (override with '.groups' argument)
```

```
## # A tibble: 592 x 5
##   Name      Games Wins Losses Win_Pct
##   <chr>    <dbl> <dbl> <dbl> <dbl>
## 1 C.Mack    7679  3731  3948  0.486
## 2 T.La Russa 5255  2821  2434  0.537
## 3 B.Cox     4505  2504  2001  0.556
## 4 D.Baker   4500  2406  2094  0.535
## 5 B.Harris  4377  2158  2219  0.493
## 6 J.McGraw  4373  2583  1790  0.591
## 7 J.Torre   4323  2326  1997  0.538
## 8 B.Bochy   4032  2003  2029  0.497
## 9 S.Anderson 4028  2194  1834  0.545
## 10 G.Mauch  3939  1902  2037  0.483
## # ... with 582 more rows
```