# Predicting NFL Win Totals

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## Executive Summary:

This code builds and evaluates a linear regression model to predict NFL team wins (W) using historical data. It involves splitting the data into training and testing sets, preprocessing with a recipe, fitting the model, making predictions, and then assessing the model's performance using RMSE and R-squared. The residual plots help to visualize how well the model's predictions align with actual outcomes.

```
#getting websites for win total lines for each season
page2023 <- read_html("https://www.cbssports.com/nfl/news/2023-nfl-win-totals-awards-odds-lines-picks-f
page2022 <- read_html("https://www.cbssports.com/nfl/news/2022-nfl-win-totals-futures-odds-predictions-
page2021 <- read_html("https://www.cbssports.com/nfl/news/2021-nfl-win-totals-odds-predictions-best-bet

#creating variables for the win total tables
team2023 <- page2023 %>%
    html_nodes("td:nth-child(1)") %>%
    html_text()

line2023 <- page2023 %>%
    html_nodes("td:nth-child(2)") %>%
    html_text() %>%
    as.numeric()
```

## Warning in page2023 %>% html\_nodes("td:nth-child(2)") %>% html\_text() %>% : NAs
## introduced by coercion

```
team2022 <- page2022 %>%
  html_nodes("td:nth-child(1)") %>%
  html_text()

line2022 <- page2022 %>%
  html_nodes("td:nth-child(2)") %>%
  html_text() %>%
  as.numeric()

team2021 <- page2021 %>%
  html_nodes("td:nth-child(1)") %>%
  html_text()
```

```
html_nodes("td:nth-child(2)") %>%
html_text() %>%
as.numeric()
```

```
#creating tables for win totals
lines2023 <- tibble(</pre>
 team = team2023,
 line = line2023
lines2023 <- lines2023 %>%
  filter(team != "Team")
lines2022 <- tibble(</pre>
 team = team2022,
 line = line2022
lines2022 <- lines2022 %>%
 filter(team != "Team")
lines2021 <- tibble(</pre>
 team = team2021.
 line = line2021
lines2021 <- lines2021 %>%
filter(team != "Team")
```

```
#changing team names in win total line tables to make sense
lines2023$team[lines2023$team == "Buffalo Bills "] <- "Buffalo Bills"</pre>
lines2023$team[lines2023$team == "New England Patriots "] <- "New England Patriots"
lines2023$team[lines2023$team == "New York Jets "] <- "New York Jets"</pre>
lines2023$team[lines2023$team == "Miami Dolphins "] <- "Miami Dolphins"
lines2023$team[lines2023$team == "Pittsburgh Steelers"] <- "Pittsburgh Steelers"
lines2023$team[lines2023$team == "Baltimore Ravens"] <- "Baltimore Ravens"
lines2023$team[lines2023$team == "Cleveland Browns"] <- "Cleveland Browns"
lines2023$team[lines2023$team == "Cincinnati Bengals "] <- "Cincinnati Bengals"
lines2023$team[lines2023$team == "Indianapolis Colts "] <- "Indianapolis Colts"
lines2023$team[lines2023$team == "Tennessee Titans"] <- "Tennessee Titans"</pre>
lines2023$team[lines2023$team == "Jacksonville Jaguars"] <- "Jacksonville Jaguars"</pre>
lines2023$team[lines2023$team == "Houston Texans"] <- "Houston Texans"</pre>
lines2023$team[lines2023$team == "Kansas City Chiefs"] <- "Kansas City Chiefs"</pre>
lines2023$team[lines2023$team == "Los Angeles Chargers"] <- "Los Angeles Chargers"</pre>
lines2023$team[lines2023$team == "Denver Broncos"] <- "Denver Broncos"</pre>
lines2023$team[lines2023$team == "Las Vegas Raiders "] <- "Las Vegas Raiders"
lines2023$team[lines2023$team == "Philadelphia Eagles"] <- "Philadelphia Eagles"
lines2023$team[lines2023$team == "New York Giants"] <- "New York Giants"</pre>
lines2023$team[lines2023$team == "Dallas Cowboys"] <- "Dallas Cowboys"</pre>
lines2023$team[lines2023$team == "Washington Commanders"] <- "Washington Commanders"
lines2023$team[lines2023$team == "Green Bay Packers"] <- "Green Bay Packers"</pre>
lines2023$team[lines2023$team == "Chicago Bears"] <- "Chicago Bears"</pre>
lines2023$team[lines2023$team == "Detroit Lions "] <- "Detroit Lions"</pre>
```

```
lines2023$team[lines2023$team == "Minnesota Vikings"] <- "Minnesota Vikings"
lines2023$team[lines2023$team == "Tampa Bay Buccaneers"] <- "Tamps Bay Buccaneers"
lines2023$team[lines2023$team == "Atlanta Falcons"] <- "Atlanta Falcons"
lines2023$team[lines2023$team == "New Orleans Saints "] <- "New Orleans Saints"
lines2023$team[lines2023$team == "Carolina Panthers"] <- "Carolina Panthers"
lines2023$team[lines2023$team == "San Francisco 49ers"] <- "San Francisco 49ers"
lines2023$team[lines2023$team == "Seattle Seahawks"] <- "Seattle Seahawks"
lines2023$team[lines2023$team == "Arizona Cardinals"] <- "Arizona Cardinals"
lines2023$team[lines2023$team == "Los Angeles Rams"] <- "Los Angeles Rams"
lines2022$team[lines2022$team == "Bills"] <- "Buffalo Bills"</pre>
lines2022$team[lines2022$team == "Patriots"] <- "New England Patriots"</pre>
lines2022$team[lines2022$team == "Jets"] <- "New York Jets"</pre>
lines2022$team[lines2022$team == "Dolphins"] <- "Miami Dolphins"</pre>
lines2022$team[lines2022$team == "Steelers"] <- "Pittsburgh Steelers"</pre>
lines2022$team[lines2022$team == "Ravens"] <- "Baltimore Ravens"</pre>
lines2022$team[lines2022$team == "Browns"] <- "Cleveland Browns"</pre>
lines2022$team[lines2022$team == "Bengals"] <- "Cincinnati Bengals"</pre>
lines2022$team[lines2022$team == "Colts"] <- "Indianapolis Colts"</pre>
lines2022$team[lines2022$team == "Titans"] <- "Tennessee Titans"</pre>
lines2022$team[lines2022$team == "Jaguars"] <- "Jacksonville Jaguars"</pre>
lines2022$team[lines2022$team == "Texans"] <- "Houston Texans"</pre>
lines2022$team[lines2022$team == "Chiefs"] <- "Kansas City Chiefs"</pre>
lines2022$team[lines2022$team == "Chargers"] <- "Los Angeles Chargers"</pre>
lines2022$team[lines2022$team == "Broncos"] <- "Denver Broncos"</pre>
lines2022$team[lines2022$team == "Raiders"] <- "Las Vegas Raiders"</pre>
lines2022$team[lines2022$team == "Eagles"] <- "Philadelphia Eagles"
lines2022$team[lines2022$team == "Giants"] <- "New York Giants"</pre>
lines2022$team[lines2022$team == "Cowboys"] <- "Dallas Cowboys"</pre>
lines2022$team[lines2022$team == "Commanders"] <- "Washington Commanders"
lines2022$team[lines2022$team == "Packers"] <- "Green Bay Packers"</pre>
lines2022$team[lines2022$team == "Bears"] <- "Chicago Bears"</pre>
lines2022$team[lines2022$team == "Lions"] <- "Detroit Lions"</pre>
lines2022$team[lines2022$team == "Vikings"] <- "Minnesota Vikings"</pre>
lines2022$team[lines2022$team == "Buccaneers"] <- "Tamps Bay Buccaneers"</pre>
lines2022$team[lines2022$team == "Falcons"] <- "Atlanta Falcons"</pre>
lines2022$team[lines2022$team == "Saints"] <- "New Orleans Saints"</pre>
lines2022$team[lines2022$team == "Panthers"] <- "Carolina Panthers"</pre>
lines2022$team[lines2022$team == "49ers"] <- "San Francisco 49ers"</pre>
lines2022$team[lines2022$team == "Seahawks"] <- "Seattle Seahawks"</pre>
lines2022$team[lines2022$team == "Cardinals"] <- "Arizona Cardinals"</pre>
lines2022$team[lines2022$team == "Rams"] <- "Los Angeles Rams"</pre>
lines2021$team[lines2021$team == "Bills"] <- "Buffalo Bills"</pre>
lines2021$team[lines2021$team == "Patriots"] <- "New England Patriots"</pre>
lines2021$team[lines2021$team == "Jets"] <- "New York Jets"</pre>
lines2021$team[lines2021$team == "Dolphins"] <- "Miami Dolphins"</pre>
lines2021$team[lines2021$team == "Steelers"] <- "Pittsburgh Steelers"</pre>
lines2021$team[lines2021$team == "Ravens"] <- "Baltimore Ravens"</pre>
lines2021$team[lines2021$team == "Browns"] <- "Cleveland Browns"</pre>
lines2021$team[lines2021$team == "Bengals"] <- "Cincinnati Bengals"</pre>
lines2021$team[lines2021$team == "Colts"] <- "Indianapolis Colts"</pre>
lines2021$team[lines2021$team == "Titans"] <- "Tennessee Titans"</pre>
```

```
lines2021$team[lines2021$team == "Jaguars"] <- "Jacksonville Jaguars"</pre>
lines2021$team[lines2021$team == "Texans"] <- "Houston Texans"</pre>
lines2021$team[lines2021$team == "Chiefs"] <- "Kansas City Chiefs"</pre>
lines2021$team[lines2021$team == "Chargers"] <- "Los Angeles Chargers"</pre>
lines2021$team[lines2021$team == "Broncos"] <- "Denver Broncos"</pre>
lines2021$team[lines2021$team == "Raiders"] <- "Las Vegas Raiders"</pre>
lines2021$team[lines2021$team == "Eagles"] <- "Philadelphia Eagles"</pre>
lines2021$team[lines2021$team == "Giants"] <- "New York Giants"</pre>
lines2021$team[lines2021$team == "Cowboys"] <- "Dallas Cowboys"</pre>
lines2021$team[lines2021$team == "Washington"] <- "Washington Commanders"
lines2021$team[lines2021$team == "Packers"] <- "Green Bay Packers"</pre>
lines2021$team[lines2021$team == "Bears"] <- "Chicago Bears"</pre>
lines2021$team[lines2021$team == "Lions"] <- "Detroit Lions"</pre>
lines2021$team[lines2021$team == "Vikings"] <- "Minnesota Vikings"</pre>
lines2021$team[lines2021$team == "Buccaneers"] <- "Tamps Bay Buccaneers"</pre>
lines2021$team[lines2021$team == "Falcons"] <- "Atlanta Falcons"</pre>
lines2021$team[lines2021$team == "Saints"] <- "New Orleans Saints"</pre>
lines2021$team[lines2021$team == "Panthers"] <- "Carolina Panthers"</pre>
lines2021$team[lines2021$team == "49ers"] <- "San Francisco 49ers"</pre>
lines2021$team[lines2021$team == "Seahawks"] <- "Seattle Seahawks"</pre>
lines2021$team[lines2021$team == "Cardinals"] <- "Arizona Cardinals"</pre>
lines2021$team[lines2021$team == "Rams"] <- "Los Angeles Rams"</pre>
```

# #importing offense and defense stats and the records for each year defense23 <- read.csv("/Users/lindaungerbock/Downloads/X2023defense.csv",skip = 1)</pre> offense23 <- read.csv("/Users/lindaungerbock/Downloads/X2023offense.csv", skip = 1) defense22 <- read.csv("/Users/lindaungerbock/Downloads/X2022defense.csv",skip = 1)</pre> offense22 <- read.csv("/Users/lindaungerbock/Downloads/X2022offense.csv", skip = 1) defense21 <- read.csv("/Users/lindaungerbock/Downloads/X2021defense.csv",skip = 1)</pre> offense21 <- read.csv("/Users/lindaungerbock/Downloads/X2021offense.csv", skip = 1) AFCrecords23 <- read.csv("/Users/lindaungerbock/Downloads/X2023AFCRecords.csv") %>% select("Tm", "W", "L") NFCrecords23 <- read.csv("/Users/lindaungerbock/Downloads/X2023NFCRecords.csv") %>% select("Tm", "W", "L") AFCrecords23\$Tm[AFCrecords23\$Tm == "Buffalo Bills\*"] <- "Buffalo Bills" AFCrecords23\$Tm[AFCrecords23\$Tm == "Miami Dolphins+"] <- "Miami Dolphins" AFCrecords23\$Tm[AFCrecords23\$Tm == "Cleveland Browns+"] <- "Cleveland Browns" AFCrecords23\$Tm[AFCrecords23\$Tm == "Baltimore Ravens\*"] <- "Baltimore Ravens" AFCrecords23\$Tm[AFCrecords23\$Tm == "Pittsburgh Steelers+"] <- "Pittsburgh Steelers" AFCrecords23\$Tm[AFCrecords23\$Tm == "Kansas City Chiefs\*"] <- "Kansas City Chiefs" AFCrecords23\$Tm[AFCrecords23\$Tm == "Houston Texans\*"] <- "Houston Texans" NFCrecords23\$Tm[NFCrecords23\$Tm == "Philadelphia Eagles+"] <- "Philadelphia Eagles" NFCrecords23\$Tm[NFCrecords23\$Tm == "Dallas Cowboys\*"] <- "Dallas Cowboys" NFCrecords23\$Tm[NFCrecords23\$Tm == "Detroit Lions\*"] <- "Detroit Lions" NFCrecords23\$Tm[NFCrecords23\$Tm == "Green Bay Packers+"] <- "Green Bay Packers" NFCrecords23\$Tm [NFCrecords23\$Tm == "Tampa Bay Buccaneers\*"] <- "Tampa Bay Buccaneers" NFCrecords23\$Tm[NFCrecords23\$Tm == "San Francisco 49ers\*"] <- "San Francisco 49ers" NFCrecords23\$Tm[NFCrecords23\$Tm == "Los Angeles Rams+"] <- "Los Angeles Rams"

```
records23 <- AFCrecords23 %>%
  rbind(NFCrecords23) %>%
  rename(team = Tm)
AFCrecords22 <- read.csv("/Users/lindaungerbock/Downloads/X2022AFCRecords.csv") %>%
  select("Tm", "W", "L")
NFCrecords22 <- read.csv("/Users/lindaungerbock/Downloads/X2022NFCRecords.csv") %>%
  select("Tm", "W", "L")
AFCrecords22$Tm[AFCrecords22$Tm == "Buffalo Bills*"] <- "Buffalo Bills"
AFCrecords22$Tm[AFCrecords22$Tm == "Miami Dolphins+"] <- "Miami Dolphins"
AFCrecords22$Tm [AFCrecords22$Tm == "Cincinnati Bengals*"] <- "Cincinnati Bengals"
AFCrecords22$Tm[AFCrecords22$Tm == "Baltimore Ravens+"] <- "Baltimore Ravens"
AFCrecords22$Tm [AFCrecords22$Tm == "Jacksonville Jaguars*"] <- "Jacksonville Jaguars"
AFCrecords22$Tm [AFCrecords22$Tm == "Kansas City Chiefs*"] <- "Kansas City Chiefs"
AFCrecords22$Tm[AFCrecords22$Tm == "Los Angeles Chargers+"] <- "Los Angeles Chargers"
NFCrecords22$Tm [NFCrecords22$Tm == "Philadelphia Eagles*"] <- "Philadelphia Eagles"
NFCrecords22$Tm[NFCrecords22$Tm == "Dallas Cowboys+"] <- "Dallas Cowboys"
NFCrecords22$Tm[NFCrecords22$Tm == "New York Giants+"] <- "New York Giants"
NFCrecords22$Tm [NFCrecords22$Tm == "Minnesota Vikings*"] <- "Minnesota Vikings"
NFCrecords22$Tm[NFCrecords22$Tm == "Tampa Bay Buccaneers*"] <- "Tampa Bay Buccaneers"
NFCrecords22$Tm[NFCrecords22$Tm == "San Francisco 49ers*"] <- "San Francisco 49ers"
NFCrecords22$Tm [NFCrecords22$Tm == "Seattle Seahawks+"] <- "Seattle Seahawks"
records22 <- AFCrecords22 %>%
 rbind(NFCrecords22) %>%
 rename(team = Tm)
AFCrecords21 <- read.csv("/Users/lindaungerbock/Downloads/X2021AFCRecords.csv") %>%
  select("Tm", "W", "L")
NFCrecords21 <- read.csv("/Users/lindaungerbock/Downloads/X2021NFCRecords.csv") %>%
  select("Tm", "W", "L")
AFCrecords21$Tm[AFCrecords21$Tm == "Buffalo Bills*"] <- "Buffalo Bills"
AFCrecords21$Tm[AFCrecords21$Tm == "New England Patriots+"] <- "New England Patriots"
AFCrecords21$Tm[AFCrecords21$Tm == "Cincinnati Bengals*"] <- "Cincinnati Bengals"
AFCrecords21$Tm[AFCrecords21$Tm == "Pittsburgh Steelers+"] <- "Pittsburgh Steelers"
AFCrecords21$Tm[AFCrecords21$Tm == "Tennessee Titans*"] <- "Tennessee Titans"
AFCrecords21$Tm[AFCrecords21$Tm == "Kansas City Chiefs*"] <- "Kansas City Chiefs"
AFCrecords21$Tm[AFCrecords21$Tm == "Los Angeles Chargers+"] <- "Los Angeles Chargers"
AFCrecords21$Tm[AFCrecords21$Tm == "Las Vegas Raiders+"] <- "Las Vegas Raiders"
NFCrecords21$Tm[NFCrecords21$Tm == "Dallas Cowboys*"] <- "Dallas Cowboys"
NFCrecords21$Tm [NFCrecords21$Tm == "Philadelphia Eagles+"] <- "Philadelphia Eagles"
NFCrecords21$Tm[NFCrecords21$Tm == "Green Bay Packers*"] <- "Green Bay Packers"
NFCrecords21$Tm[NFCrecords21$Tm == "Tampa Bay Buccaneers*"] <- "Tampa Bay Buccaneers"
NFCrecords21$Tm[NFCrecords21$Tm == "Los Angeles Rams*"] <- "Los Angeles Rams"
NFCrecords21$Tm[NFCrecords21$Tm == "San Francisco 49ers+"] <- "San Francisco 49ers"
NFCrecords21$Tm [NFCrecords21$Tm == "Arizona Cardinals+"] <- "Arizona Cardinals"
NFCrecords21$Tm [NFCrecords21$Tm == "Washington Football Team"] <- "Washington Commanders"
records21 <- AFCrecords21 %>%
 rbind(NFCrecords21) %>%
```

### rename(team = Tm) str(offense23) ## 'data.frame': 35 obs. of 28 variables: ## \$ Rk : int 1 2 3 4 5 6 7 8 9 10 ... "Dallas Cowboys" "Miami Dolphins" "San Francisco 49ers" "Baltimore Ravens" ... ## \$ Tm : chr ## \$ G : int 17 17 17 17 17 17 17 17 17 17 ... ## \$ PF : num 509 496 491 483 461 451 433 404 402 396 ... ## \$ Yds : num 6317 6822 6773 6296 6712 ... : num 1122 1053 1024 1076 1137 ... ## \$ Ply : num 5.6 6.5 6.6 5.9 5.9 5.7 5.4 5.6 5.1 5.2 ... ## \$ Y.P ## \$ TO : num 16 25 18 19 23 28 28 18 18 22 ... ## \$ FL 6 10 6 12 11 10 12 5 7 12 ... : num ## \$ X1stD : num 385 360 383 360 375 381 377 351 337 324 ... : num 428 393 336 328 408 385 369 361 406 355 ... ## \$ Cmp ## \$ Att : num 614 566 491 494 606 579 563 583 606 574 ... ## \$ Yds.1 : num 4397 4514 4384 3635 4401 ... ## \$ TD : num 36 30 33 27 30 29 24 26 28 18 ... ## \$ Int : num 10 15 12 7 12 18 16 13 11 10 ... ## \$ NY.A : num 6.7 7.6 8.4 6.8 6.9 6.9 6.4 6.6 6.2 6 ... ## \$ X1stD.1: num 229 223 207 180 228 199 197 206 199 178 ... \$ Att.1 : num 468 456 499 541 500 512 510 477 480 479 ... ## ## \$ Yds.2 : num 1920 2308 2389 2661 2311 ... ## \$ TD.1 : num 14 27 27 26 27 22 22 18 13 19 ... ## \$ Y.A : num 4.1 5.1 4.8 4.9 4.6 4.3 4.3 4.3 3.6 4.3 ... ## \$ X1stD.2: num 113 113 147 145 124 158 149 110 112 113 ... ## \$ Pen : num 115 97 101 102 97 106 95 89 96 95 ... ## \$ Yds.3 : num 964 767 933 955 843 883 785 720 846 685 ... ## \$ X1stPy : num 43 24 29 35 23 24 31 35 26 33 ... ## \$ Sc. : num 50.3 43.5 45.3 43.1 40.6 41.4 42.9 41.1 36.6 35.7 ... ## \$ TO. : num 8.9 13.4 10.1 9.6 11.8 14.9 15.3 9.2 8.2 9.2 ... ## \$ EXP : num 193 129 269 139 191 ... #pulling only the team stats we want to use in the model key offense23 <- offense23 %>% select("Tm", "PF", "Yds", "Ply", "Y.P", "TO", "X1stD", "Sc.") %>% rename(team = Tm, pf = PF, oyds = Yds, o\_plays = Ply, o\_yds\_play = `Y.P`, turnovers = TO, o\_fd = `X1stD`, o\_score\_pct = `Sc.`) key\_defense23 <- defense23 %>% select("Tm", "PA", "Yds", "Ply", "Y.P", "TO", "X1stD", "Sc.") %>% rename(team = Tm, pa = PA, dyds = Yds, d\_plays = Ply, d\_yds\_play = `Y.P`, takeaways = TO, d\_fd = `X1stD`, d\_score\_pct = `Sc.`) key\_offense22 <- offense22 %>% select("Tm", "PF", "Yds", "Ply", "Y.P", "TO", "X1stD", "Sc.") %>% rename(team = Tm, pf = PF, oyds = Yds, o\_plays = Ply, o\_yds\_play = `Y.P`, turnovers = TO, o\_fd = `X1stD`, o\_score\_pct = `Sc.`)

key\_defense22 <- defense22 %>%

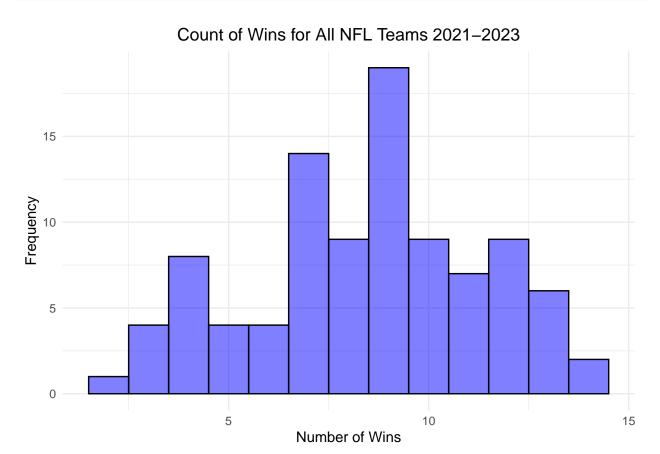
```
select("Tm", "PA", "Yds", "Ply", "Y.P", "TO", "X1stD", "Sc.") %>%
  rename(team = Tm, pa = PA, dyds = Yds, d_plays = Ply, d_yds_play = `Y.P`,
         takeaways = TO, d_fd = `X1stD`, d_score_pct = `Sc.`)
key_offense21 <- offense21 %>%
  select("Tm", "PF", "Yds", "Ply", "Y.P", "TO", "X1stD", "Sc.") %>%
  rename(team = Tm, pf = PF, oyds = Yds, o_plays = Ply, o_yds_play = `Y.P`,
         turnovers = TO, o fd = `X1stD`, o score pct = `Sc.`)
key_offense21$team[key_offense21$team == "Washington Football Team"] <- "Washington Commanders"
key_defense21 <- defense21 %>%
  select("Tm", "PA", "Yds", "Ply", "Y.P", "TO", "X1stD", "Sc.") %>%
  rename(team = Tm, pa = PA, dyds = Yds, d_plays = Ply, d_yds_play = `Y.P`,
         takeaways = TO, d_fd = `X1stD`, d_score_pct = `Sc.`)
key_defense21$team[key_defense21$team == "Washington Football Team"] <- "Washington Commanders"
#combining offense and defense stats for each year and then combining all three years into one table
stats_2023 <- left_join(records23, key_offense23, by = "team") %>%
 left join(key defense23, by = "team") %>%
 left_join(lines2023, by = "team") %>%
 mutate(year = 2023) %>%
 na.omit(stats_2023)
stats_2022 <- left_join(records22, key_offense22, by = "team") %>%
  left_join(key_defense22, by = "team") %>%
  left_join(lines2022, by = "team") %>%
  mutate(year = 2022) %>%
  na.omit(stats_2023)
stats_2021 <- left_join(records21, key_offense21, by = "team") %>%
  left_join(key_defense21, by = "team") %>%
  left_join(lines2021, by = "team") %>%
  mutate(year = 2021) %>%
  na.omit(stats_2023)
stats <- rbind(stats 2023, stats 2022, stats 2021) %>%
  mutate(ppg = pf/17, pag = pa/17)
records23$W <- as.numeric(as.character(records23$W))</pre>
records22$W <- as.numeric(as.character(records22$W))</pre>
records21$W <- as.numeric(as.character(records21$W))</pre>
all_records <- records23 %>%
 rbind(records22) %>%
 rbind(records21)
saving data to use in python
# Save the all_records data frame to a CSV file
write.csv(all_records, "all_records.csv", row.names = FALSE)
```

write.csv(all\_records, "all\_records.csv", row.names = FALSE)
write.csv(stats, "stats\_combined.csv", row.names = FALSE)

Before fitting the model, we analyze the distribution of wins (W) in the 2021-2023 NFL seasons. The histogram below shows the count of wins across 96 data points, revealing a roughly normal distribution with

nine wins being the most common. This aligns with the fact that nine wins is approximately half of the 17-game season. Given this normal distribution, no data transformation is needed, and a linear regression model is suitable

```
ggplot(all_records, aes(x = W)) +
  geom_histogram(binwidth = 1, fill = "blue", color = "black", alpha = 0.5) +
  labs(title = "Count of Wins for All NFL Teams 2021-2023", x = "Number of Wins", y = "Frequency") +
  theme_minimal() +
  theme(plot.title = element_text(hjust = 0.5))
```



Predict future win totals based on historical data using linear regression

A linear regression model was initially fitted to all variables, with the data split into 75% training and 25% testing sets. The training set produced an intercept of 42.57—far exceeding the season's maximum wins of 17—and many predictors were not significant at common alpha levels, suggesting overfitting. Additionally, several offensive and defensive predictors behaved counterintuitively, further indicating model issues. Despite this, the model showed a low RMSE of about 1 and a relatively high R-squared of 0.85 on the training data, suggesting good initial performance.

```
#Split the data into training (75%) and testing (25%)
set.seed(1122)
stats_split <- initial_split(stats)
stats_train <- training(stats_split)
stats_test <- testing(stats_split)</pre>
```

Splitting the data into a training set (75%) and a testing set (25%)

```
#Set to linear model
stats_mod <- linear_reg() %>%
set_engine("lm")
```

Set up a linear regression model using the lm engine and the model will predict the W variable (wins).

```
#Create a recipe
stats_rec <- recipe(W ~ ., data = stats_train) %>%
    step_rm(L, year, pf, pa) %>%
    update_role(team, new_role = "id")
```

Creating a recipe. A recipe defines the data processing steps for the model:  $W \sim .:$  This specifies that you're predicting the W variable using all other variables in the dataset (.). step\_rm(L, year, pf, pa): Removes unnecessary variables (L, year, pf, and pa) from the recipe that aren't needed for the prediction. update\_role(team, new\_role = "id"): Sets team as an identifier, meaning it's not included as a predictor but just for reference.

```
#Build a workflow for fitting the model
stats_wflow <- workflow() %>%
  add_model(stats_mod) %>%
  add_recipe(stats_rec)
```

```
#Fit the model to training data
stats_fit <- stats_wflow %>%
  fit(data = stats_train)

tidy(stats_fit)
```

```
## # A tibble: 16 x 5
##
      term
                  estimate std.error statistic p.value
##
      <chr>
                     <dbl>
                               <dbl>
                                         <dbl>
                                                 <dbl>
##
   1 (Intercept) 42.6
                            39.8
                                         1.07 0.290
##
  2 oyds
                   0.00198
                             0.00480
                                         0.412 0.682
## 3 o_plays
                  -0.0138
                             0.0263
                                        -0.526 0.601
## 4 o_yds_play -2.50
                             5.08
                                        -0.491 0.625
## 5 turnovers
                  -0.131
                             0.0444
                                        -2.96 0.00457
## 6 o_fd
                   0.0294
                             0.0178
                                         1.65 0.106
## 7 o_score_pct 0.111
                             0.0809
                                         1.37 0.177
## 8 dyds
                   0.00676
                             0.00519
                                         1.30 0.199
## 9 d_plays
                  -0.0161
                             0.0278
                                        -0.582 0.563
                  -7.58
                                        -1.36 0.179
## 10 d_yds_play
                             5.56
## 11 takeaways
                   0.0441
                             0.0426
                                         1.04 0.305
## 12 d_fd
                  -0.0411
                             0.0163
                                        -2.52 0.0147
## 13 d_score_pct 0.0386
                             0.0952
                                         0.405 0.687
## 14 line
                                         2.38 0.0212
                   0.246
                             0.103
## 15 ppg
                   0.0958
                             0.164
                                         0.584 0.562
## 16 pag
                  -0.0926
                             0.169
                                        -0.549 0.585
```

```
## # A tibble: 16 x 5
##
      term
                 estimate std.error statistic p.value
                    <dbl>
##
      <chr>
                              <dbl>
                                        <dbl>
                                        1.07 0.290
##
  1 (Intercept) 42.6
                           39.8
##
   2 oyds
                  0.00198 0.00480
                                        0.412 0.682
                 -0.0138
                            0.0263
                                       -0.526 0.601
##
  3 o_plays
  4 o_yds_play -2.50
                            5.08
                                       -0.491 0.625
## 5 turnovers
                 -0.131
                            0.0444
                                        -2.96 0.00457
## 6 o_fd
                  0.0294
                            0.0178
                                        1.65 0.106
## 7 o_score_pct 0.111
                            0.0809
                                        1.37 0.177
## 8 dyds
                  0.00676
                            0.00519
                                        1.30 0.199
                                        -0.582 0.563
## 9 d_plays
                 -0.0161
                            0.0278
## 10 d_yds_play -7.58
                            5.56
                                       -1.36 0.179
## 11 takeaways
                  0.0441
                                        1.04 0.305
                            0.0426
## 12 d_fd
                 -0.0411
                            0.0163
                                        -2.52 0.0147
## 13 d_score_pct 0.0386
                            0.0952
                                        0.405 0.687
## 14 line
                                        2.38 0.0212
                  0.246
                            0.103
## 15 ppg
                  0.0958
                            0.164
                                        0.584 0.562
## 16 pag
                 -0.0926
                                        -0.549 0.585
                            0.169
rmse(stats_train_pred,
     truth = W,
     estimate = .pred)
## # A tibble: 1 x 3
     .metric .estimator .estimate
##
     <chr>>
            <chr>>
                            <dbl>
## 1 rmse
            standard
                             1.01
rsq(stats_train_pred,
     truth = W,
     estimate = .pred)
## # A tibble: 1 x 3
##
     .metric .estimator .estimate
##
     <chr>
            <chr>
                            <dbl>
                            0.853
## 1 rsq
            standard
```

The model, though likely overfit, was used to predict the testing dataset. The RMSE increased slightly to 1.88, and R-squared dropped to 0.73. The residual plot shows most predictions were within two wins of the actual values, indicating a reasonably accurate estimate.

# Residual Plot 2 2 -2 3 6 9 Predicted Values

```
rmse(stats_test_pred,
     truth = W,
     estimate = .pred)
## # A tibble: 1 x 3
     .metric .estimator .estimate
             <chr>
                            <dbl>
##
     <chr>
                             1.88
## 1 rmse
             standard
rsq(stats_test_pred,
     truth = W,
     estimate = .pred)
## # A tibble: 1 x 3
##
     .metric .estimator .estimate
     <chr> <chr>
                            <dbl>
             standard
                            0.730
## 1 rsq
```

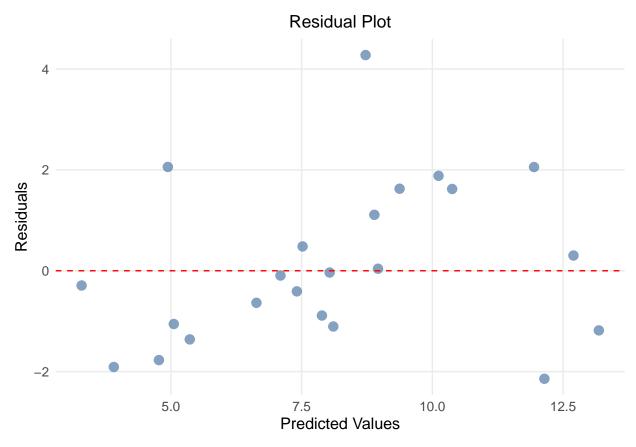
To address overfitting, many predictors were removed to improve model quality. While the initial model suggested turnovers, first downs allowed, and win total line as significant predictors, points per game and points allowed per game were intuitively selected along with win total line. This refinement yielded a well-fitting model with meaningful coefficients and small p-values. The intercept, now 4.9, falls within a realistic range of 0 to 16. Points per game positively correlates with wins, while points allowed per game negatively correlates, aligning with expectations. Coefficients and intercept details are shown below

```
stats_rec2 <- recipe(W ~ ppg + pag + line, data = stats_train)</pre>
stats_wflow2 <- workflow() %>%
  add_model(stats_mod) %>%
  add_recipe(stats_rec2)
stats_fit2 <- stats_wflow2 %>%
  fit(data = stats_train)
tidy(stats_fit2)
## # A tibble: 4 x 5
##
     term
                 estimate std.error statistic p.value
     <chr>>
                    <dbl>
                               <dbl>
                                         <dbl>
                                                  <dbl>
                    4.92
                              2.08
                                          2.37 2.09e- 2
## 1 (Intercept)
## 2 ppg
                    0.433
                              0.0409
                                         10.6 8.34e-16
## 3 pag
                   -0.336
                              0.0605
                                         -5.56 5.54e- 7
## 4 line
                    0.159
                              0.100
                                          1.59 1.18e- 1
stats_train_pred2 <- predict(stats_fit2, new_data = stats_train) %>%
  bind_cols(stats_train %>% select(W, team, year, line))%>%
  mutate(residual = W - .pred)
glance(stats_fit2)
## # A tibble: 1 x 12
##
     r.squared adj.r.squared sigma statistic p.value
                                                           df logLik
                                                                       AIC
                                                                             BIC
##
         <dbl>
                       <dbl> <dbl>
                                        <dbl>
                                                 <dbl> <dbl>
                                                               <dbl> <dbl> <dbl>
## 1
                       0.783 1.24
                                         82.7 3.81e-22
                                                            3 -111.
                                                                      231.
## # i 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
rmse(stats_train_pred2,
     truth = W,
     estimate = .pred)
## # A tibble: 1 x 3
##
     .metric .estimator .estimate
     <chr>>
             <chr>>
                             <dbl>
## 1 rmse
             standard
                              1.20
rsq(stats_train_pred2,
     truth = W,
     estimate = .pred)
```

## # A tibble: 1 x 3

```
##
     .metric .estimator .estimate
##
     <chr>>
             <chr>>
                             <dbl>
## 1 rsq
             standard
                             0.792
## # A tibble: 4 x 5
##
     term
                  estimate std.error statistic p.value
##
     <chr>
                               <dbl>
                                          <dbl>
                                                   <dbl>
                     <dbl>
## 1 (Intercept)
                     4.92
                              2.08
                                           2.37 2.09e- 2
                                          10.6 8.34e-16
## 2 ppg
                     0.433
                              0.0409
                                          -5.56 5.54e- 7
## 3 pag
                    -0.336
                              0.0605
## 4 line
                     0.159
                              0.100
                                           1.59 1.18e- 1
```

As expected, removing predictors slightly increased the RMSE to 1.2 and decreased the R-squared to 0.79, but these values remain strong, with predictions typically within one win of actual totals. On the testing dataset, performance was consistent, with an RMSE of 1.5 and an improved R-squared of 0.81. The residual plot below highlights the differences between predicted and actual wins in the training set, showing most predictions within two wins of the true values.



```
rmse(stats_test_pred2,
    truth = W,
     estimate = .pred)
## # A tibble: 1 x 3
     .metric .estimator .estimate
##
   <chr> <chr>
                           <dbl>
            standard
                           1.55
## 1 rmse
rsq(stats_test_pred2,
    truth = W,
    estimate = .pred)
## # A tibble: 1 x 3
## .metric .estimator .estimate
## <chr> <chr>
                           <dbl>
```

0.814

## 1 rsq standard