Final Project

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```
library(dplyr)
Attaching package: 'dplyr'
The following objects are masked from 'package:stats':
    filter, lag
The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
  library(tidyverse)
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v forcats 1.0.0 v readr
                                  2.1.4
v ggplot2 3.4.3 v stringr
v lubridate 1.9.2 v tibble
                                  1.5.0
                                  3.2.1
           1.0.2
                     v tidyr
                                  1.3.0
v purrr
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag() masks stats::lag()
i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become
  library(tidymodels)
```

```
-- Attaching packages -----
                                      ----- tidymodels 1.1.1 --
v broom
              1.0.5
                        v rsample
                                      1.2.0
v dials
              1.2.0
                                      1.1.2
                        v tune
              1.0.5
                                      1.1.3
v infer
                        v workflows
v modeldata
              1.2.0
                        v workflowsets 1.0.1
v parsnip
              1.1.1
                        v yardstick
                                      1.2.0
v recipes
              1.0.8
-- Conflicts ----- tidymodels conflicts() --
x scales::discard() masks purrr::discard()
x dplyr::filter()
                   masks stats::filter()
x recipes::fixed() masks stringr::fixed()
x dplyr::lag()
                   masks stats::lag()
x yardstick::spec() masks readr::spec()
x recipes::step()
                   masks stats::step()
* Dig deeper into tidy modeling with R at https://www.tmwr.org
  library(ggplot2)
  library(Stat2Data)
  tennis <- read.csv("data/wta_matches_qual_itf_2023.csv")</pre>
```

Introduction and Data

Research Question

Does the court type impact the duration of a women's single tennis match, considering the winner's height, age, and playing hand?

The data we chose was the Women's Tennis Association data set from Awesomedata's GitHub repository (https://github.com/JeffSackmann/tennis_wta/blob/master/wta_matches_qual_itf_2023.csv) The data was created in 2023 and was collected from the International Tennis Federation. The data contains 34,323 observations and 49 variables. The variables of interest in our research include surface, winner and loser height, winner and loser age, and winner and loser hand. These variables will help us answer the question of the court types impact on the duration of tennis matches. There are some NA variables corresponding to the players height. Since this is our variable of interest, we will be dropping all NA values corresponding to height. This will leave us with 1,256 observations. Additionally, since grand slams are played on either clay, grass, or hard courts, we will be dropping the matches that were played on carpet. This will then leave us with 1,239 observations. The motivation behind this project is to help Women's Tennis players better prepare for the length of their match based on the surface they will be playing on. With the Olympics coming up, this data will help the tennis player better prepare for a match.

```
tennis <- tennis %>%
  filter(!is.na(winner_ht) & !is.na(loser_ht) & surface != "Carpet")

Variables of Interest
Surface: Surface the match was played on (clay, grass, or hard)
```

Winner_ht: Height of the winner in centimeters (cm)

Winner_age: Age of the winner

loser_hand: Dominant playing hand of the loser (right, left, undecided)

Winner_hand: Dominant playing hand of the winner (right, left, undecided)

loser_ht: Height of the loser in centimeters (cm)

loser_age: Age of the loser

Minutes: Duration of the tennis match in minutes

```
winner_hand_counts <- tennis |>
  count(winner_hand)

loser_hand_counts <- tennis |>
  count(loser_hand)

print(winner_hand_counts)
```

```
winner_hand n
1 L 116
2 R 1110
3 U 13
```

```
print(loser_hand_counts)
```

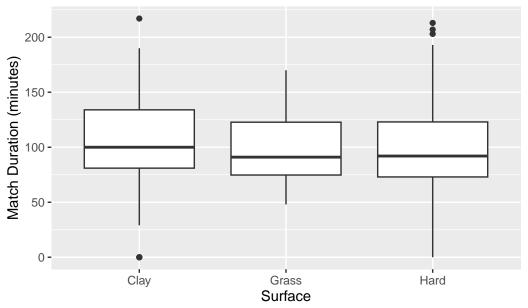
```
loser_hand n
1 L 107
2 R 1107
3 U 25
```

Methodology

```
ggplot(data = tennis, aes(x = surface, y = minutes)) +
  geom_boxplot() +
  labs(title = "Match Duration by Court Surface",
        x = "Surface",
        y = "Match Duration (minutes)")
```

Warning: Removed 620 rows containing non-finite values (`stat_boxplot()`).

Match Duration by Court Surface



```
library(MASS)
```

Attaching package: 'MASS'

The following object is masked from 'package:dplyr':

select

```
tennis_binary <- tennis |>
    mutate(minutes = case_when(minutes >= 101.3 ~ 1,
                              minutes <= 101.3 \sim 0),
           minutes = as.factor(minutes))
  winner_mins <- glm(minutes ~ surface + winner_age +</pre>
                     as.numeric(winner ht) + winner hand, data = tennis binary,
                    family = "binomial")
  summary(winner_mins)
Call:
glm(formula = minutes ~ surface + winner_age + as.numeric(winner_ht) +
    winner_hand, family = "binomial", data = tennis_binary)
Coefficients:
                       Estimate Std. Error z value Pr(>|z|)
                      -3.051243 2.321364 -1.314 0.1887
(Intercept)
surfaceGrass
                      -0.246180 0.296445 -0.830 0.4063
surfaceHard
                      -0.198074 0.179377 -1.104 0.2695
                      winner_age
as.numeric(winner_ht) 0.008849 0.012878 0.687 0.4920
winner_handR
                      0.121194 0.291980 0.415
                                                    0.6781
                   -12.810120 535.411309 -0.024 0.9809
winner handU
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 849.49 on 618 degrees of freedom
Residual deviance: 840.17 on 612 degrees of freedom
  (620 observations deleted due to missingness)
AIC: 854.17
Number of Fisher Scoring iterations: 12
  loser_mins <- glm(minutes ~ surface + loser_age +</pre>
                   as.numeric(loser_ht) + loser_hand, data = tennis_binary,
                   famil = "binomial")
  summary(loser_mins)
```

```
Call:
glm(formula = minutes ~ surface + loser_age + as.numeric(loser_ht) +
    loser_hand, family = "binomial", data = tennis_binary)
Coefficients:
                     Estimate Std. Error z value Pr(>|z|)
                    -6.106356
(Intercept)
                                2.286333 -2.671 0.00757 **
surfaceGrass
                    -0.211784 0.298861 -0.709 0.47855
surfaceHard
                    -0.201920 0.180160 -1.121 0.26238
                     0.009084
                                         0.464 0.64240
loser_age
                                0.019564
                                           2.405 0.01618 *
as.numeric(loser_ht) 0.030503
                                0.012685
                     0.537760
loser_handR
                                0.308438
                                           1.743 0.08125 .
                                0.873335 -0.498 0.61881
loser_handU
                    -0.434522
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 849.49 on 618 degrees of freedom
Residual deviance: 837.20 on 612 degrees of freedom
  (620 observations deleted due to missingness)
AIC: 851.2
Number of Fisher Scoring iterations: 4
  exp(coef(loser_mins))
         (Intercept)
                            surfaceGrass
                                                  surfaceHard
         0.002228658
                             0.809139613
                                                  0.817160408
           loser_age as.numeric(loser_ht)
                                                  loser_handR
         1.009125753
                             1.030972719
                                                  1.712166551
         loser_handU
         0.647574123
  exp(coef(winner_mins))
```

1.048390e+00 winner_handU 2.732973e-06 1.008888e+00

1.128843e+00