# Player Style Clustering

### Group E

12/9/2021

```
knitr::opts_chunk$set(message = FALSE, warning = FALSE)
```

I am following the cluster analysis steps from this **ATP cluster analysis**. The goal is to compare the results from this WTA cluster analysis to the results in the ATP cluster analysis to explore how playing styles might differ.

```
library(naniar)
# from https://cran.r-project.org/web/packages/naniar/vignettes/replace-with-na.html
tennis_results <- tennis_results %>%
  replace_with_na(replace = list(w_SvGms = 0, l_SvGms = 0, minutes = 0))
# selecting matches from 2011 onward
# creating the statistics that they mentioned in the table
playerstyle_WTA_cluster <- tennis_results %>%
  filter(year >= 2011, tour == "WTA", !is.na(w_1stIn), !is.na(w_svpt), !is.na(l_1stIn), !is.na(l_svpt),
  mutate(w_1stsvpct = w_1stIn/w_svpt,
         l_1stsvpct = l_1stIn/l_svpt,
         w_svpctWon = (w_1stWon+w_2ndWon)/w_svpt,
         l_svpctWon = (l_1stWon+l_2ndWon)/l_svpt,
         w_1stsvWon = w_1stWon/w_1stIn,
         l_1stsvWon = l_1stWon/l_1stIn,
         w_2ndsvWon = w_2ndWon/(w_svpt-w_1stIn),
         1_2ndsvWon = 1_2ndWon/(1_svpt-l_1stIn),
         w_acepct = w_ace/w_svpt,
         l_acepct = l_ace/l_svpt,
         w_dfpct = w_df/w_svpt,
         l_dfpct = l_df/l_svpt,
         w_ptspersvgame = w_svpt/w_SvGms,
         1_ptspersvgame = 1_svpt/l_SvGms,
         w_bpSavepct = w_bpSaved/w_bpFaced,
         1_bpSavepct = 1_bpSaved/1_bpFaced,
         w_bppersvgame = w_bpFaced/w_SvGms,
         l_bppersvgame = l_bpFaced/l_SvGms,
         w_pctptWon = (w_1stWon+w_2ndWon+(l_svpt-l_1stWon-l_2ndWon))/(w_svpt+l_svpt),
         l_pctptWon = (l_1stWon+l_2ndWon+(w_svpt-w_1stWon-w_2ndWon))/(w_svpt+l_svpt),
         w_1stretWon = (l_1stIn-l_1stWon)/l_1stIn,
         1_1stretWon = (w_1stIn-w_1stWon)/w_1stIn,
         w_2ndretWon = (l_svpt-l_1stIn-l_2ndWon)/(l_svpt-l_1stIn),
         1_2ndretWon = (w_svpt-w_1stIn-w_2ndWon)/(w_svpt-w_1stIn),
         w_retpctWon = 1-(l_1stWon+l_2ndWon)/l_svpt,
         l_retpctWon = 1-(w_1stWon+w_2ndWon)/w_svpt,
```

```
w_ptsperretgame = l_svpt/l_SvGms,
         l_ptsperretgame = w_svpt/w_SvGms,
         w_bpConvpct = 1-l_bpSaved/l_bpFaced,
         l_bpConvpct = 1-w_bpSaved/w_bpFaced,
         w_bpperretgame = l_bpFaced/l_SvGms,
         l_bpperretgame = w_bpFaced/w_SvGms,
         w_retace = l_ace/l_svpt,
         l_retace = w_ace/w_svpt,
         w_retdf = l_df/l_svpt,
         l_retdf = w_df/w_svpt,
         ptspermin = (w_svpt+l_svpt)/minutes) %>%
  select(1:30, 49:89)
# creating two rows for winner/loser
cluster_w <- playerstyle_WTA_cluster %>%
  select(tour, tourney_id, tourney_name, surface, tourney_date, year, month, winner_id, winner_name, wi
 mutate(result = 1)
colnames(cluster_w) <- c("tour", "tourney_id", "tourney_name", "surface", "tourney_date", "year", "mont
cluster_l <- playerstyle_WTA_cluster %>%
  select(tour, tourney_id, tourney_name, surface, tourney_date, year, month, loser_id, loser_name, lose
  mutate(result = 0)
colnames(cluster_1) <- c("tour", "tourney_id", "tourney_name", "surface", "tourney_date", "year", "mont."</pre>
final_cluster_data <- rbind(cluster_w,cluster_l)</pre>
# calculating stats by player
WTA_player_stats <- final_cluster_data %>%
  group_by(id, name) %>%
  summarize(height = mean(height),
            age = max(age),
            win_perc = mean(result),
            perc_points_won = mean(pctptWon, na.rm = TRUE),
            "1st_serv_perc" = mean(`1stsvpct`, na.rm = TRUE),
            "1st_win" = mean(`1stsvWon`, na.rm = TRUE),
            ace_perc = mean(acepct, na.rm = TRUE),
            df_perc = mean(dfpct, na.rm = TRUE),
            "2nd_win" = mean(`2ndsvWon`, na.rm = TRUE),
            svc_perc_win = mean(svpctWon, na.rm = TRUE),
            points_per_svc_game = mean(ptspersvgame, na.rm = TRUE),
            break_point_save_perc = mean(bpSavepct, na.rm = TRUE),
            bp_per_game = mean(bppersvgame, na.rm = TRUE),
            return_1st_win = mean(`1stretWon`, na.rm = TRUE),
            return_ace_perc = mean(retace, na.rm = TRUE),
            return_df_perc = mean(retdf, na.rm = TRUE),
            return_2nd_win = mean(`2ndretWon`, na.rm = TRUE),
            return_perc_win = mean(retpctWon, na.rm = TRUE),
            points_per_return_game = mean(ptsperretgame, na.rm = TRUE),
            bp_convert_perc = mean(bpConvpct, na.rm = TRUE),
            return_bp_per_game = mean(bpperretgame, na.rm = TRUE),
            points_per_minute = mean(ptspermin, na.rm = TRUE))
```

```
surface_stats <- final_cluster_data %>%
  group_by(id, name, surface) %>%
  summarize(count = n()) %>%
  mutate(freq = count / sum(count)) %>%
  pivot_wider(id_cols = c(id, name), names_from = surface, values_from = freq) %%
  summarize(clay_perc = Clay, grass_perc = Grass, hard_perc = Hard) %>%
  select(-2)
final_WTA <- cbind(surface_stats, WTA_player_stats) %>%
  select(-1) %>%
  rename(id = id...5) %>%
  select(4:7, 1:3, 8:27)
library(broom)
# cluster analysis
final_WTA_km <- final_WTA %>%
  drop_na() %>%
  select(height:points_per_minute) %>%
  mutate(across(height:points_per_minute, scale))
set.seed(13)
final_WTA_kclusts <-</pre>
  tibble(k = 1:9) \%
  mutate(final_WTA_kclust = map(k, ~kmeans(final_WTA_km, .x)),
    glanced = map(final_WTA_kclust, glance),
    tidied = map(final_WTA_kclust, tidy),
    augmented = map(final_WTA_kclust, augment, final_WTA_km)
 )
clusters <-
  final_WTA_kclusts %>%
  unnest(cols = c(tidied))
assignments <-
  final_WTA_kclusts %>%
  unnest(cols = c(augmented))
clusterings <-</pre>
  final_WTA_kclusts %>%
  unnest(cols = c(glanced))
clusterings %>%
  ggplot(aes(x = k, y = tot.withinss)) +
  geom_line() +
  geom_point() + ylab("") +
  ggtitle("Total Within Sum of Squares")
```

### **Total Within Sum of Squares**

set.seed(47)

<fct>

## 1 between ~

## 2 top play~

## 3 strong s~

## 4 tier 2

##

## #

WTA\_clustered <- final\_WTA\_km %>%

<dbl> <dbl>

170. 26.5

177. 26.6

172. 30.8

177. 30.4

<dbl>

0.430

0.237

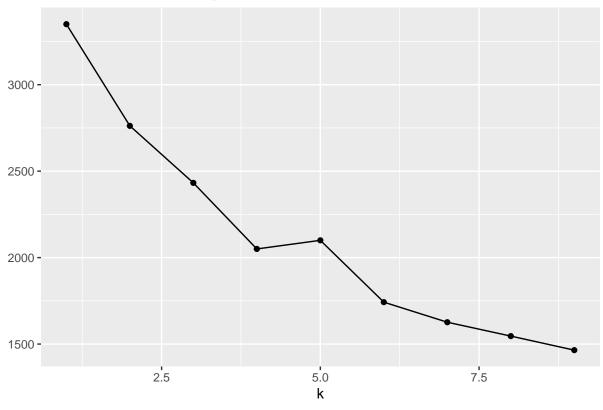
0.213

0.289

ace\_perc <dbl>, df\_perc <dbl>, 2nd\_win <dbl>, svc\_perc\_win <dbl>,

return\_1st\_win <dbl>, return\_ace\_perc <dbl>, return\_df\_perc <dbl>,

## # ... with 18 more variables: 1st\_serv\_perc <dbl>, 1st\_win <dbl>,



```
kmeans(centers = 4)

WTA_clusters <- cbind(WTA_clustered$cluster, final_WTA %>% drop_na())

WTA_clusters %>%
    rename(cluster = `WTA_clustered$cluster`) %>%
    group_by(cluster) %>%
    select(cluster, height:points_per_minute) %>%
    summarise_each(funs(mean(., na.rm = TRUE))) %>%
    mutate(cluster = as.factor(cluster)) %>%
    mutate(cluster = recode(cluster, "1" = "between tours", "2" = "top players", "3" = "strong servers (A"

## # A tibble: 4 x 26

## cluster height age clay_perc grass_perc hard_perc win_perc perc_points_won
```

<dbl>

0.479

0.657

0.684

0.559

<dbl>

0.450

0.633

0.500

0.330

<dbl>

0.490

0.522

0.502

0.471

<dbl>

0.0916

0.106

0.103

0.152

points\_per\_svc\_game <dbl>, break\_point\_save\_perc <dbl>, bp\_per\_game <dbl>,

```
## # return_2nd_win <dbl>, return_perc_win <dbl>, points_per_return_game <dbl>,
## # bp_convert_perc <dbl>, return_bp_per_game <dbl>, points_per_minute <dbl>
```

 ${\it\# from\ https://stackoverflow.com/questions/21807987/calculate-the-mean-for-each-column-of-a-matrix-in-relative}$ 

#### ATP Replication

```
library(naniar)
# from https://cran.r-project.org/web/packages/naniar/vignettes/replace-with-na.html
tennis_results <- tennis_results %>%
  replace_with_na(replace = list(w_SvGms = 0, l_SvGms = 0, minutes = 0))
# selecting matches from 2011 onward
# creating the statistics that they mentioned in the table
playerstyle_ATP_cluster <- tennis_results %>%
  filter(year >= 2011, tour == "ATP", !is.na(w_1stIn), !is.na(w_svpt), !is.na(l_1stIn), !is.na(l_svpt),
  mutate(w_1stsvpct = w_1stIn/w_svpt,
         1_1stsvpct = l_1stIn/l_svpt,
         w_svpctWon = (w_1stWon+w_2ndWon)/w_svpt,
         l_svpctWon = (l_1stWon+l_2ndWon)/l_svpt,
         w_1stsvWon = w_1stWon/w_1stIn,
         1 1stsvWon = 1 1stWon/l 1stIn,
         w_2ndsvWon = w_2ndWon/(w_svpt-w_1stIn),
         1_2ndsvWon = 1_2ndWon/(1_svpt-l_1stIn),
         w_acepct = w_ace/w_svpt,
         l_acepct = l_ace/l_svpt,
         w_dfpct = w_df/w_svpt,
         l_dfpct = l_df/l_svpt,
         w_ptspersvgame = w_svpt/w_SvGms,
         l_ptspersvgame = l_svpt/l_SvGms,
         w_bpSavepct = w_bpSaved/w_bpFaced,
         1_bpSavepct = 1_bpSaved/1_bpFaced,
         w_bppersvgame = w_bpFaced/w_SvGms,
         l_bppersvgame = l_bpFaced/l_SvGms,
         w_pctptWon = (w_1stWon+w_2ndWon+(l_svpt-l_1stWon-l_2ndWon))/(w_svpt+l_svpt),
         l_pctptWon = (l_1stWon+l_2ndWon+(w_svpt-w_1stWon-w_2ndWon))/(w_svpt+l_svpt),
         w_1stretWon = (l_1stIn-l_1stWon)/l_1stIn,
         l_1stretWon = (w_1stIn-w_1stWon)/w_1stIn,
         w_2ndretWon = (l_svpt-l_1stIn-l_2ndWon)/(l_svpt-l_1stIn),
         1_2ndretWon = (w_svpt-w_1stIn-w_2ndWon)/(w_svpt-w_1stIn),
         w_retpctWon = 1-(l_1stWon+l_2ndWon)/l_svpt,
         l_retpctWon = 1-(w_1stWon+w_2ndWon)/w_svpt,
         w_ptsperretgame = l_svpt/l_SvGms,
         1_ptsperretgame = w_svpt/w_SvGms,
         w_bpConvpct = 1-l_bpSaved/l_bpFaced,
         l_bpConvpct = 1-w_bpSaved/w_bpFaced,
         w_bpperretgame = l_bpFaced/l_SvGms,
         l_bpperretgame = w_bpFaced/w_SvGms,
         w_retace = l_ace/l_svpt,
         l_retace = w_ace/w_svpt,
         w retdf = 1 df/l svpt,
         l_retdf = w_df/w_svpt,
```

```
ptspermin = (w_svpt+l_svpt)/minutes) %>%
    select(1:30, 49:89)
# creating two rows for winner/loser
cluster_w_m <- playerstyle_ATP_cluster %>%
    select(tour, tourney_id, tourney_name, surface, tourney_date, year, month, winner_id, winner_name, win
    mutate(result = 1)
colnames(cluster_w_m) <- c("tour", "tourney_id", "tourney_name", "surface", "tourney_date", "year", "modern tourney_date", "year", "modern tourney_date", "tourney_date", "tou
cluster_l_m <- playerstyle_ATP_cluster %>%
    select(tour, tourney_id, tourney_name, surface, tourney_date, year, month, loser_id, loser_name, lose
    mutate(result = 0)
colnames(cluster_l_m) <- c("tour", "tourney_id", "tourney_name", "surface", "tourney_date", "year", "mod</pre>
final_cluster_data_m <- rbind(cluster_w_m,cluster_l_m)</pre>
# calculating stats by player
ATP_player_stats <- final_cluster_data_m %>%
    group_by(id, name) %>%
    summarize(height = mean(height), age = max(age), win_perc = mean(result), perc_points_won = mean(pctp
surface_stats_m <- final_cluster_data_m %>%
    group_by(id, name, surface) %>%
    summarize(count = n()) %>%
    mutate(freq = count / sum(count)) %>%
    pivot_wider(id_cols = c(id, name), names_from = surface, values_from = freq) %>%
    summarize(clay_perc = Clay, grass_perc = Grass, hard_perc = Hard) %%
    select(-2)
final_ATP <- cbind(surface_stats_m, ATP_player_stats) %>%
    select(-1) %>%
    rename(id = id...5) %>%
    select(4:7, 1:3, 8:27)
# cluster analysis
final_ATP_km <- final_ATP %>%
    drop_na() %>%
    select(height:points_per_minute) %>%
    mutate(across(height:points_per_minute, scale))
set.seed(7)
final_ATP_kclusts <-</pre>
    tibble(k = 1:9) \%
    mutate(final_ATP_kclust = map(k, ~kmeans(final_ATP_km, .x)),
        glanced = map(final_ATP_kclust, glance),
        tidied = map(final_ATP_kclust, tidy),
        augmented = map(final_ATP_kclust, augment, final_ATP_km)
    )
```

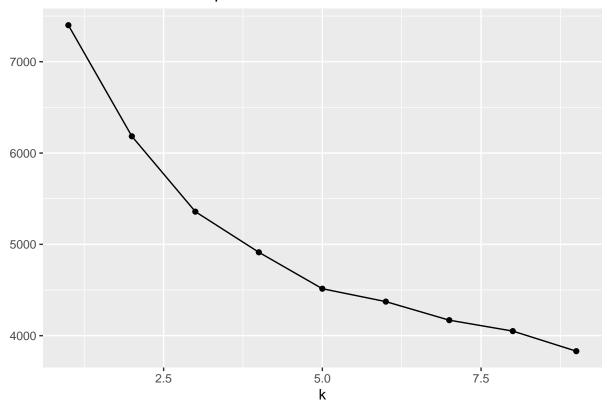
```
clusters_m <-
  final_ATP_kclusts %>%
  unnest(cols = c(tidied))

assignments_m <-
  final_ATP_kclusts %>%
  unnest(cols = c(augmented))

clusterings_m <-
  final_ATP_kclusts %>%
  unnest(cols = c(glanced))
```

```
clusterings_m %>%
  ggplot(aes(x = k, y = tot.withinss)) +
  geom_line() +
  geom_point() + ylab("") +
  ggtitle("Total Within Sum of Squares")
```

## Total Within Sum of Squares



```
set.seed(47)
ATP_clustered <- final_ATP_km %>%
   kmeans(centers = 4) # use 4 to compare to WTA

ATP_clusters <- cbind(ATP_clustered$cluster, final_ATP %>% drop_na())
```

```
ATP_clusters %>%
    rename(cluster = `ATP_clustered$cluster`) %>%
    group_by(cluster) %>%
    select(cluster, height:points_per_minute) %>%
    summarise_each(funs(mean(., na.rm = TRUE))) %>%
    mutate(cluster = as.factor(cluster)) %>%
     mutate(cluster = recode(cluster, "1" = "tier 2", "2" = "top players", "3" = "between tours", "4" = "
## # A tibble: 4 x 26
        cluster height
##
                                                  age clay_perc grass_perc hard_perc win_perc perc_points_won
                                <dbl> <dbl>
##
          <fct>
                                                                    <dbl>
                                                                                            <dbl>
                                                                                                                 <dbl>
                                                                                                                                     <dbl>
                                                                                                                                     0.284
## 1 tier 2
                                   184. 31.9
                                                                    0.261
                                                                                          0.177
                                                                                                                 0.559
                                                                                                                                                                       0.465
## 2 top play~
                                   186. 30.4
                                                                    0.291
                                                                                          0.0895
                                                                                                                 0.614
                                                                                                                                    0.610
                                                                                                                                                                       0.517
                                    184. 31.2
## 3 between ~
                                                                    0.420
                                                                                          0.0963
                                                                                                                 0.482
                                                                                                                                     0.402
                                                                                                                                                                       0.487
                                    192. 30.3
                                                                    0.215
                                                                                          0.130
                                                                                                                 0.651
                                                                                                                                    0.465
                                                                                                                                                                       0.496
## 4 strong s~
## # ... with 18 more variables: 1st serv perc <dbl>, 1st win <dbl>,
             ace_perc <dbl>, df_perc <dbl>, 2nd_win <dbl>, svc_perc_win <dbl>,
              points_per_svc_game <dbl>, break_point_save_perc <dbl>, bp_per_game <dbl>,
## #
             return_1st_win <dbl>, return_ace_perc <dbl>, return_df_perc <dbl>,
## #
             return_2nd_win <dbl>, return_perc_win <dbl>, points_per_return_game <dbl>,
              bp_convert_perc <dbl>, return_bp_per_game <dbl>, points_per_minute <dbl>
## #
\# from https://stackoverflow.com/questions/21807987/calculate-the-mean-for-each-column-of-a-matrix-in-relative for the sum of th
# combining the two datasets
ATP_cluster_results <- ATP_clusters %>%
    rename(cluster = `ATP_clustered$cluster`) %>%
    group_by(cluster) %>%
    select(cluster, height:points_per_minute) %>%
    summarise_each(funs(mean(., na.rm = TRUE))) %>%
    mutate(tour = "ATP",
                   cluster = as.factor(cluster)) %>%
    mutate(cluster = recode(cluster, "1" = "tier 2", "2" = "top players",
                                                         "3" = "between tours",
                                                         "4" = "strong servers (ATP)/\nstrong returners (WTA)")) %>%
    select(27, 1:26)
```