STAT 231: Problem Set 10B

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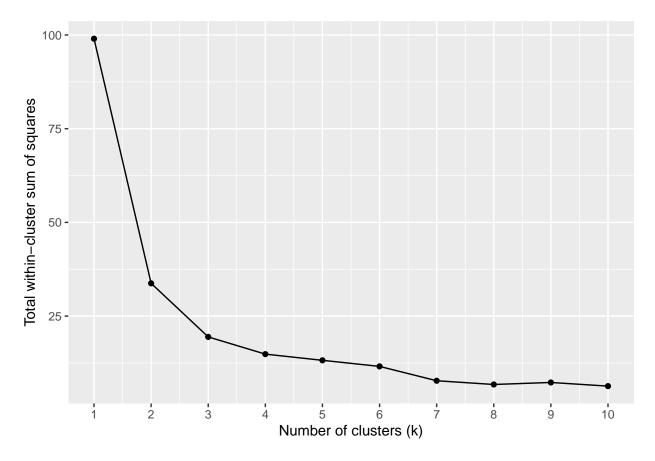
```
load("MuhammadAhsanTahir.RData")
```

```
## # A tibble: 34 x 5
                    WinPercent GoalsPerMatch PointsPerMatch GoalsConcededPerMatch
##
      Squad
##
      <chr>
                         <dbl>
                                       <dbl>
                                                      <dbl>
                                                                            <dh1>
## 1 Arsenal
                          53.7
                                       1.77
                                                      1.82
                                                                             1.13
## 2 Aston Villa
                          26.7
                                                      1.01
                                                                             1.62
                                       1.10
## 3 Bournemouth
                          29.5
                                       1.27
                                                      1.11
                                                                             1.74
## 4 Brentford
                          34.2
                                       1.26
                                                      1.21
                                                                             1.47
## 5 Brighton
                          25.3
                                                      1.1
                                                                             1.36
## 6 Burnley
                          28.2
                                       0.970
                                                      1.11
                                                                             1.40
## 7 Cardiff City
                          22.4
                                       0.868
                                                      0.842
                                                                             1.88
## 8 Chelsea
                                                      1.94
                          57.1
                                       1.82
                                                                             1.01
## 9 Crystal Palace
                          31.6
                                                                             1.41
                                       1.13
                                                      1.18
## 10 Everton
                          38.4
                                       1.36
                                                      1.42
                                                                             1.32
## # ... with 24 more rows
```

```
# set the seed for reproducibility
set.seed(1877090)

tableForClustering <- groupedTable %>%
   select(WinPercent, GoalsPerMatch, GoalsConcededPerMatch)%>%
   mutate(across(where(is.numeric), ~scale(.)[,1], .names = "{.col}_scaled")) %>%
   select(WinPercent_scaled, GoalsPerMatch_scaled, GoalsConcededPerMatch_scaled)
```

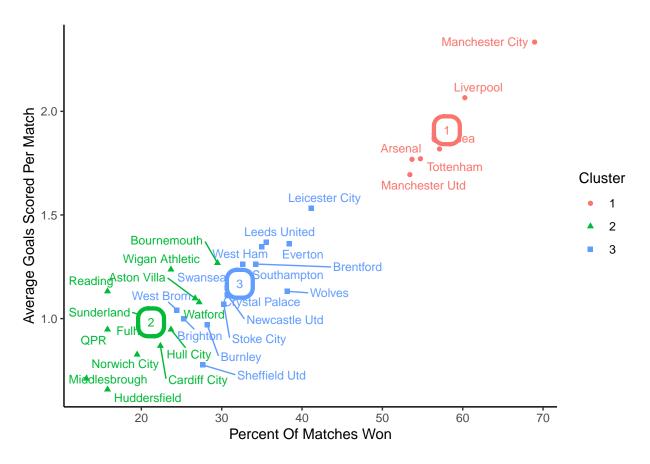
```
\# Iterate through clustering algorithm for 10 different values of k
elbow_plot <- tibble(k = 1:10) %>%
  mutate(
    # List-column of 10 kmeans objects
    # (apply `kmeans()` to each value of `k`)
   kmeans_results = purrr::map(k, ~kmeans(tableForClustering, .x)),
    # List-column of "glanced" model summaries for each kmeans object
    # (apply `glance()` to each corresponding result after running `kmeans()`)
   glanced = purrr::map(kmeans_results, glance)) %>%
  # Turn `glanced` list-column into regular tibble columns
  unnest(cols = c(glanced))
# Construct elbow plot
ggplot(elbow_plot, aes(x = k, y = tot.withinss)) +
  geom_point() +
 geom_line() +
  scale_x_continuous(breaks = 1:10) +
  labs(x = "Number of clusters (k)",
      y = "Total within-cluster sum of squares")
```



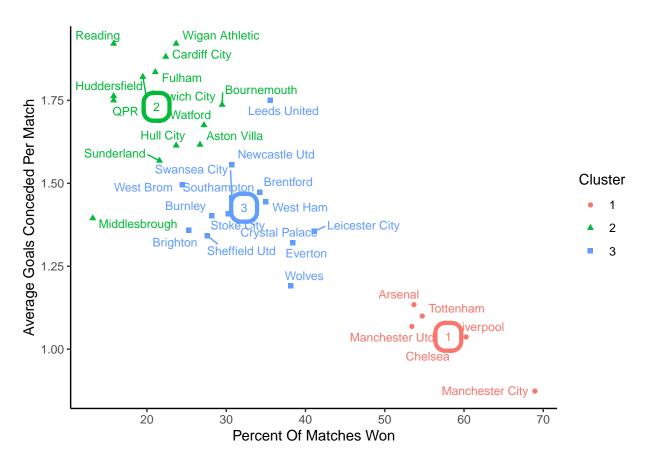
```
meanGoalsPerMatch <- mean(groupedTable$GoalsPerMatch)
meanGoalsConcededPerMatch <- mean(groupedTable$GoalsConcededPerMatch)
meanWinPercent <- mean(groupedTable$WinPercent)</pre>
```

```
sdGoalsPerMatch <- sd(groupedTable$GoalsPerMatch)</pre>
sdGoalsConcededPerMatch <- sd(groupedTable$GoalsConcededPerMatch)</pre>
sdWinPercent <- sd(groupedTable$WinPercent)</pre>
#Thus, we should use 3 clusters from looking at the elbow plot.
# Perform k-means clustering with k = 3
# set the seed for reproducibility
set.seed(1877090)
footballKmeans <- tableForClustering %>%
  kmeans(centers = 3, nstart = 20)
footballKmeansSummaries <- tidy(footballKmeans) %>%
  mutate(GoalsPerMatch = meanGoalsPerMatch + (sdGoalsPerMatch * GoalsPerMatch_scaled),
         GoalsConcededPerMatch = meanGoalsConcededPerMatch + (sdGoalsConcededPerMatch * GoalsConcededPer
         WinPercent = meanWinPercent + (sdWinPercent * WinPercent_scaled)
# Add cluster assignment as a factor to the data frame
# (argument order MUST be: kmeans object first, original data frame second)
footballWithKmeans <- augment(footballKmeans, groupedTable)</pre>
footballWithKmeans
## # A tibble: 34 x 6
                    WinPercent GoalsPerMatch PointsPerMatch GoalsConcede~1 .clus~2
##
      Squad
                                                                      <dbl> <fct>
##
      <chr>>
                          <dbl>
                                        <dbl>
                                                       <dbl>
                                        1.77
                                                       1.82
## 1 Arsenal
                           53.7
                                                                      1.13 1
## 2 Aston Villa
                           26.7
                                                                       1.62 2
                                        1.10
                                                       1.01
## 3 Bournemouth
                           29.5
                                        1.27
                                                       1.11
                                                                       1.74 2
## 4 Brentford
                           34.2
                                       1.26
                                                       1.21
                                                                       1.47 3
## 5 Brighton
                           25.3
                                       1
                                                       1.1
                                                                       1.36 3
## 6 Burnley
                           28.2
                                       0.970
                                                                       1.40 3
                                                       1.11
## 7 Cardiff City
                                                       0.842
                           22.4
                                        0.868
                                                                       1.88 2
## 8 Chelsea
                           57.1
                                       1.82
                                                       1.94
                                                                       1.01 1
## 9 Crystal Palace
                          31.6
                                       1.13
                                                       1.18
                                                                       1.41 3
## 10 Everton
                           38.4
                                        1.36
                                                       1.42
                                                                       1.32 3
## # ... with 24 more rows, and abbreviated variable names
## # 1: GoalsConcededPerMatch, 2: .cluster
# Visualize the cluster assignments and centroids
ggplot(footballWithKmeans, aes(x = WinPercent, y = GoalsPerMatch)) +
  geom_point(aes(color = .cluster, shape = .cluster)) +
  geom_text_repel(aes(label = Squad, color = .cluster),
                  size = 3, max.overlaps = 15, show.legend = FALSE) +
  # Add centroid labels to plot
  geom_label(data = footballKmeansSummaries, aes(label = cluster, color = cluster),
             size = 3,
             label.r = unit(0.5, "lines"),
             label.size = 1.5,
             label.padding = unit(0.5, "lines"),
             show.legend = FALSE) +
```

```
labs(x = "Percent Of Matches Won",
    y = "Average Goals Scored Per Match",
    color = "Cluster",
    shape = "Cluster") +
theme_classic()
```



```
# Visualize the cluster assignments and centroids
ggplot(footballWithKmeans, aes(x = WinPercent, y = GoalsConcededPerMatch)) +
  geom_point(aes(color = .cluster, shape = .cluster)) +
  geom_text_repel(aes(label = Squad, color = .cluster),
                  size = 3, max.overlaps = 15, show.legend = FALSE) +
  # Add centroid labels to plot
  geom_label(data = footballKmeansSummaries, aes(label = cluster, color = cluster),
             size = 3,
             label.r = unit(0.5, "lines"),
             label.size = 1.5,
             label.padding = unit(0.5, "lines"),
            show.legend = FALSE) +
  labs(x = "Percent Of Matches Won",
      y = "Average Goals Conceded Per Match",
       color = "Cluster",
      shape = "Cluster") +
  theme_classic()
```



```
# Visualize the cluster assignments and centroids
ggplot(footballWithKmeans, aes(x = GoalsConcededPerMatch, y = GoalsPerMatch)) +
  geom_point(aes(color = .cluster, shape = .cluster)) +
  geom_text_repel(aes(label = Squad, color = .cluster),
                  size = 3, max.overlaps = 15, show.legend = FALSE) +
  # Add centroid labels to plot
  geom_label(data = footballKmeansSummaries, aes(label = cluster, color = cluster),
             size = 3,
             label.r = unit(0.5, "lines"),
             label.size = 1.5,
             label.padding = unit(0.5, "lines"),
            show.legend = FALSE) +
  labs(x = "Average Goals Conceded Per Match",
      y = "Average Goals Scored Per Match",
      color = "Cluster",
      shape = "Cluster") +
  theme_classic()
```

