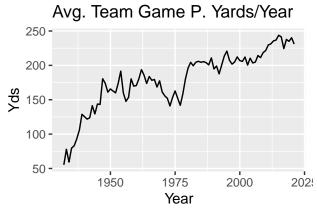
NFL_Passing_Stats_Over_the_Years

Oscar Monroy

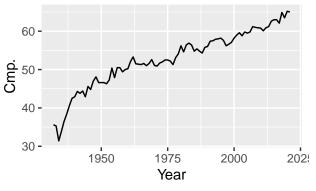
1/3/2022

```
# install.packages("ggpubr")
library(ggpubr)
## Warning: package 'ggpubr' was built under R version 3.6.3
## Loading required package: ggplot2
## Warning: package 'ggplot2' was built under R version 3.6.3
library(ggplot2)
tg <- read.csv("AvgPerTeamGame.csv")</pre>
ts <- read.csv("AvgPerTeamSeason.csv")</pre>
total <- read.csv("TotalPerSeason.csv")</pre>
sum(duplicated(tg))
## [1] O
sum(duplicated(ts))
## [1] 0
sum(duplicated(total))
## [1] 0
# summary(tg)
# summary(ts)
# summary(total)
g1 <- ggplot(tg, aes(Year, Yds)) +</pre>
  geom_line(color = "black") +
  ggtitle("Avg. Team Game P. Yards/Year")
g2 <- ggplot(tg, aes(Year, Att)) +</pre>
  geom_line(color = "black") +
  ggtitle("Avg. Team Game P. Attempts/Year")
g3 <- ggplot(tg, aes(Year, Cmp.)) +
```

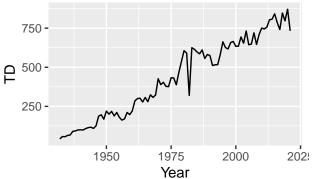


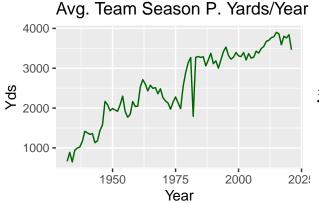


Avg. Team Game P. Comp.%/Year



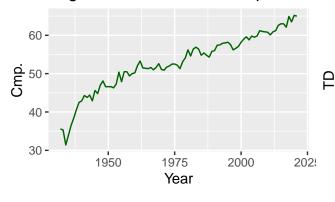
Avg. Team Game P. TD's/Year



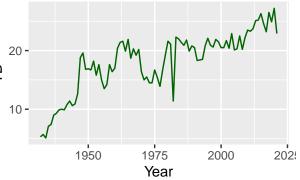


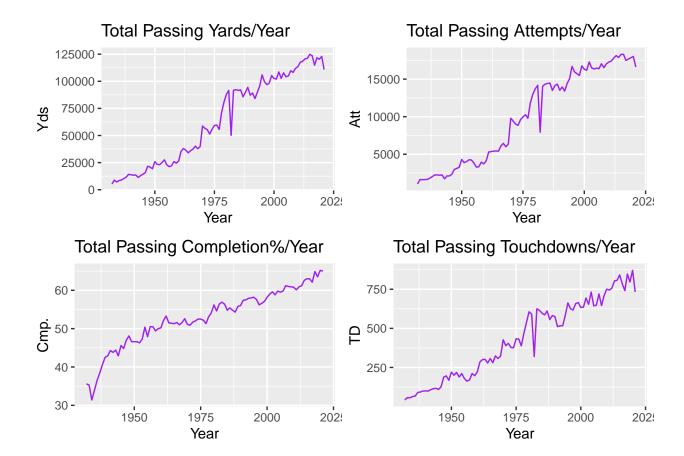
Avg. Team Season P. Attempts/Ye 5004002001950 1975 2000 2028

Avg. Team Season P. Comp.%/Yea



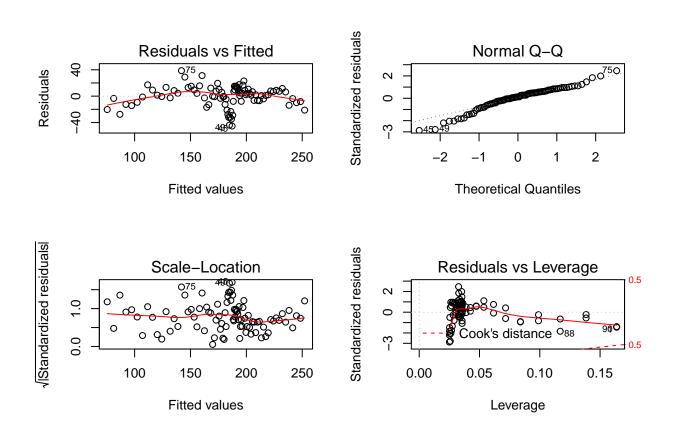
Avg. Team Season P. TD's/Year



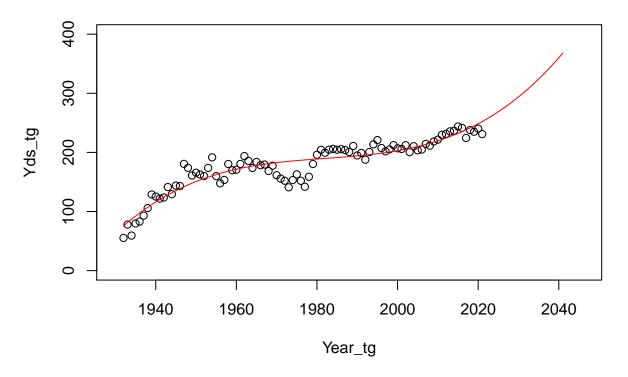


```
Yds_tg <- tg$Yds
Year_tg <- tg$Year
g_cbm <- lm(Yds_tg ~ poly(Year_tg, 3))
summary(g_cbm)</pre>
```

```
##
## Call:
## lm(formula = Yds_tg ~ poly(Year_tg, 3))
##
## Residuals:
##
      Min
                1Q
                   Median
                                3Q
                                       Max
                     2.071
                             9.563
                                    38.671
##
  -45.353
           -7.598
##
## Coefficients:
                     Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                      179.090
                                   1.682 106.467 < 2e-16 ***
## poly(Year_tg, 3)1
                     352.768
                                  15.958
                                          22.106 < 2e-16 ***
## poly(Year_tg, 3)2
                     -66.931
                                  15.958
                                         -4.194 6.63e-05 ***
## poly(Year_tg, 3)3
                       94.426
                                  15.958
                                           5.917 6.50e-08 ***
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15.96 on 86 degrees of freedom
## Multiple R-squared: 0.8629, Adjusted R-squared: 0.8581
## F-statistic: 180.4 on 3 and 86 DF, p-value: < 2.2e-16
```



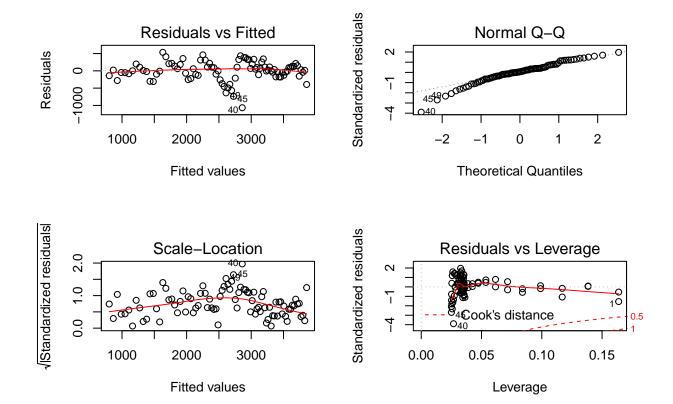
Avg. Team Season Passing Yards/Year Projections (Until 2041)



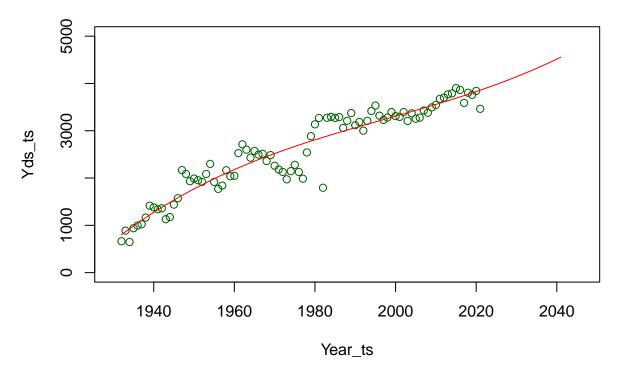
```
p_20yrs1 <- data.frame(Year = tail(years_x1, 20), Yds = tail(predictions_tg, 20))
p_20yrs1</pre>
```

```
##
       Year
                 Yds
      2022 256.0555
## 91
       2023 260.1235
## 92
## 93
       2024 264.3699
## 94
       2025 268.7990
## 95
       2026 273.4151
## 96
       2027 278.2226
## 97
       2028 283.2258
## 98
       2029 288.4291
## 99
       2030 293.8368
## 100 2031 299.4532
## 101 2032 305.2827
## 102 2033 311.3296
## 103 2034 317.5983
## 104 2035 324.0931
## 105 2036 330.8182
## 106 2037 337.7782
## 107 2038 344.9772
## 108 2039 352.4197
## 109 2040 360.1100
## 110 2041 368.0524
```

```
Yds_ts <- ts$Yds
Year_ts <- ts$Year
s_cbm <- lm(Yds_ts ~ poly(Year_ts, 3))</pre>
summary(s_cbm)
##
## Call:
## lm(formula = Yds_ts ~ poly(Year_ts, 3))
## Residuals:
                     Median
       Min
                 1Q
                                   3Q
                                           Max
## -1066.52 -122.25
                       16.45
                               145.17
                                        535.41
##
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                     2581.40
                                  29.24 88.279 < 2e-16 ***
                                 277.41 28.413 < 2e-16 ***
## poly(Year_ts, 3)1 7882.06
## poly(Year_ts, 3)2 -1102.72
                                 277.41 -3.975 0.000146 ***
## poly(Year_ts, 3)3
                     401.64
                                 277.41
                                         1.448 0.151298
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 277.4 on 86 degrees of freedom
## Multiple R-squared: 0.9056, Adjusted R-squared: 0.9023
## F-statistic: 275.1 on 3 and 86 DF, p-value: < 2.2e-16
par(mfrow=c(2,2))
plot(s_cbm) # Diagnositic plots shows that the model has no problems
```



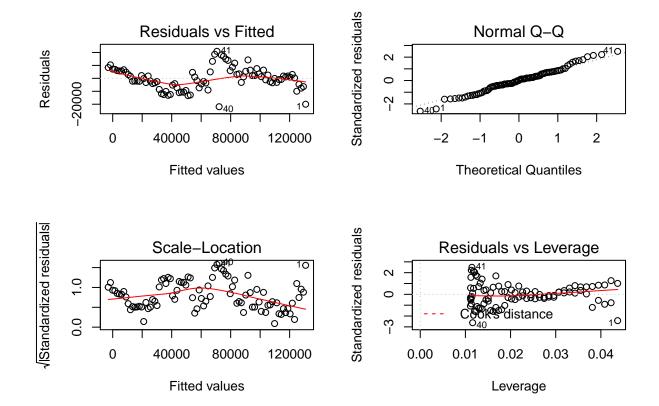
Avg. Team Game Passing Yards/Year Projections (Until 2041)



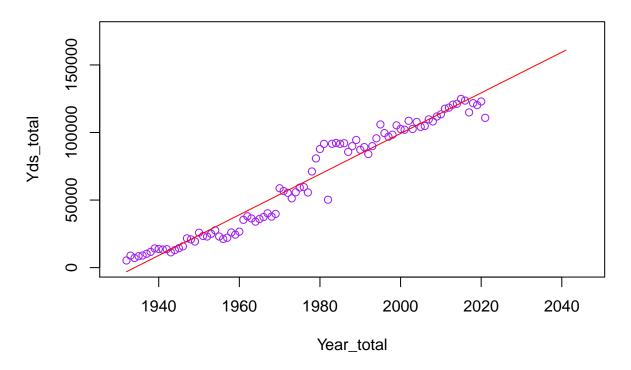
```
p_20yrs2 <- data.frame(Year = tail(years_x1, 20), Yds = tail(predictions_ts, 20))
p_20yrs2</pre>
```

```
##
       Year
                 Yds
       2022 3887.553
## 91
       2023 3917.605
## 92
## 93
       2024 3948.131
## 94
       2025 3979.148
## 95
       2026 4010.674
## 96
       2027 4042.729
## 97
       2028 4075.331
## 98
       2029 4108.498
## 99
       2030 4142.248
## 100 2031 4176.601
## 101 2032 4211.575
## 102 2033 4247.187
## 103 2034 4283.457
## 104 2035 4320.402
## 105 2036 4358.043
## 106 2037 4396.396
## 107 2038 4435.480
## 108 2039 4475.314
## 109 2040 4515.916
## 110 2041 4557.305
```

```
# Since the plot using Yds from "total" shows a linear pattern compared to the
# other data sets, we'll use linear regression instead of
# cubic regression like the previous two prediction models.
Yds_total <- total$Yds
Year_total <- total$Year
t_lm <- lm(Yds_total ~ Year_total)</pre>
summary(t_lm)
##
## Call:
## lm(formula = Yds_total ~ Year_total)
## Residuals:
##
       Min
                 1Q
                     Median
                                   3Q
## -21927.3 -4303.7 120.9 5342.5 20874.0
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.909e+06 6.756e+04 -43.06 <2e-16 ***
## Year_total 1.504e+03 3.418e+01 44.01 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 8424 on 88 degrees of freedom
## Multiple R-squared: 0.9565, Adjusted R-squared: 0.956
## F-statistic: 1937 on 1 and 88 DF, p-value: < 2.2e-16
par(mfrow=c(2,2))
plot(t_lm) # Diagnositic plots shows that the model has no problems
```



Total Passing Yards/Year Projections (Until 2041)



```
p_20yrs3 <- data.frame(Year = tail(years_x1, 20), Yds = tail(predictions_total, 20))
p_20yrs3</pre>
```

```
##
       Year
                 Yds
       2022 132354.5
## 91
       2023 133858.8
## 92
## 93
       2024 135363.1
## 94
       2025 136867.4
## 95
       2026 138371.7
## 96
       2027 139875.9
## 97
       2028 141380.2
## 98
       2029 142884.5
## 99
       2030 144388.8
## 100 2031 145893.1
## 101 2032 147397.3
## 102 2033 148901.6
## 103 2034 150405.9
## 104 2035 151910.2
## 105 2036 153414.5
## 106 2037 154918.8
## 107 2038 156423.0
## 108 2039 157927.3
## 109 2040 159431.6
## 110 2041 160935.9
```