Sports Analytics Regression

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```
library(readr)
# read data
nba_stats = read.csv('nba-stats-2013-2019.csv')
Part 1: An initial linear regression model
# split into training and test set
train = nba_stats[nba_stats$Year < 2018,]</pre>
test = nba_stats[nba_stats$Year >= 2018,]
# run regression
lm_stats_1 = lm(PTS ~ . - Team - Year - G - Playoffs - PTS, data = train)
summary(lm_stats_1)
##
## Call:
## lm(formula = PTS ~ . - Team - Year - G - Playoffs - PTS, data = train)
##
## Residuals:
##
          Min
                      1Q
                             Median
                                             3Q
## -3.562e-13 -9.761e-14 -3.014e-14
                                     4.631e-14
                                                 2.029e-12
##
## Coefficients: (2 not defined because of singularities)
##
                 Estimate Std. Error
                                         t value Pr(>|t|)
## (Intercept) -4.753e-12 1.011e-12 -4.700e+00 6.29e-06 ***
## X3P
                3.000e+00
                           8.733e-16
                                      3.435e+15
## X3PA
                1.024e-16
                           3.162e-16
                                       3.240e-01
                                                   0.7465
## X2P
                2.000e+00
                           3.671e-16
                                       5.448e+15
                                                  < 2e-16 ***
                                                   1.0000
## X2PA
                0.000e+00
                           2.431e-16
                                       0.000e+00
## FG
                       NA
                                   NA
                                              NA
                                                       NA
## FGA
                                   NA
                                                       NA
                       NA
                                              NA
                1.000e+00
## FT
                           4.327e-16
                                       2.311e+15
                                                  < 2e-16 ***
## FTA
                7.259e-16
                           3.582e-16
                                       2.027e+00
                                                   0.0447 *
                                                   0.5090
## ORB
                2.343e-16
                           3.539e-16
                                       6.620e-01
## DRB
               -1.470e-16
                           2.939e-16 -5.000e-01
                                                   0.6177
## AST
                1.725e-16
                           2.405e-16
                                      7.170e-01
                                                   0.4746
## STL
               -1.550e-16
                           4.801e-16 -3.230e-01
                                                   0.7473
## BLK
                2.267e-16
                           4.243e-16 5.340e-01
                                                   0.5941
## TOV
                2.251e-16
                           3.303e-16
                                      6.810e-01
                                                   0.4967
## PF
               -4.675e-16 2.514e-16 -1.860e+00
                                                   0.0651 .
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.857e-13 on 136 degrees of freedom
                            1, Adjusted R-squared:
## Multiple R-squared:
## F-statistic: 2.166e+31 on 13 and 136 DF, \, p-value: < 2.2e-16
```

- (a) The training set R2 is 1.
- (b) The nonzero coefficients are those on X3P, X2P, and FT.
- (c) The result of this regression is not surprising since all points must come from either 2-point FGs, 3-point FGs, or free throws. We are thus able to perfectly predict total points with these variables in the dataset.

Part 2: A better linear regression model

```
lm_stats_2 = lm(PTS ~ X3PA + X2PA + FGA + FTA + ORB + DRB + AST + STL
                + BLK + TOV + PF, data=train)
summary(lm_stats_2)
##
## Call:
  lm(formula = PTS ~ X3PA + X2PA + FGA + FTA + ORB + DRB + AST +
##
       STL + BLK + TOV + PF, data = train)
##
##
  Residuals:
##
       Min
                1Q
                    Median
                                3Q
                                        Max
   -618.88 -105.45
                      2.49
                            114.65
                                    429.92
##
## Coefficients: (1 not defined because of singularities)
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 513.82999
                          629.50509
                                       0.816 0.41576
## X3PA
                 0.69129
                            0.11843
                                       5.837 3.58e-08 ***
                            0.12259
                                             0.00371 **
## X2PA
                 0.36184
                                       2.952
## FGA
                      NA
                                          NA
                                                   NA
                                 NA
## FTA
                 0.78122
                            0.09658
                                       8.089 2.67e-13 ***
                                      -0.916
## ORB
                -0.20277
                            0.22138
                                             0.36129
## DRB
                 0.61064
                            0.18204
                                       3.354
                                             0.00103 **
## AST
                 0.91510
                            0.13105
                                       6.983 1.09e-10 ***
## STL
                 0.50506
                            0.29140
                                       1.733 0.08528 .
## BLK
                 0.07975
                            0.27270
                                       0.292
                                              0.77039
                -0.62701
## TOV
                                      -3.071
                            0.20417
                                             0.00257 **
## PF
                 0.30420
                            0.15811
                                       1.924 0.05640 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 184.4 on 139 degrees of freedom
## Multiple R-squared: 0.7944, Adjusted R-squared: 0.7796
## F-statistic: 53.69 on 10 and 139 DF, p-value: < 2.2e-16
```

- (a) X3PA, X2PA, FTA, DRB, AST, and TOV are all statistically significant at the 0.05 level.
- (b) The coefficient on X3PA implies that an increase in X3PA of 1 corresponds to an increase in PTS of 0.69129. This implies a probability of success of a 3-point attempt of 0.69129/3 = 0.23043.
- (c) FGA does not have a coefficient in the model since it is perfectly collinear with X3PA, X2PA, and FTA since all field goal attempts must be one of those types.

```
# use regression to predict test set
pred = predict(lm_stats_2, newdata=test)
# compute SSE and SST
SSE = sum((test$PTS - pred)^2)
SST = sum((mean(train$PTS) - test$PTS)^2)
pred_r2 = 1 - SSE/SST
```

pred_r2

[1] 0.9327728

(d) The test set R2 is 0.9327728.

See Part 3 in Python.