

modeling

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4/20/2022

```
fifa_model = read_csv2("fifa_model.csv")

## Using ',' as decimal and '.' as grouping mark. Use read_delim() for more control.
## Warning: Missing column names filled in: 'X1' [1]
## Parsed with column specification:
## cols(
##   X1 = col_double(),
##   sofifa_id = col_double(),
##   short_name = col_character(),
##   age = col_double(),
##   league_name = col_character(),
##   overall = col_double(),
##   value_eur = col_double(),
##   preferred_foot = col_character(),
##   team_position = col_character(),
##   pace = col_double(),
##   shooting = col_double(),
##   passing = col_double(),
##   dribbling = col_double(),
##   defending = col_double(),
##   physic = col_double(),
##   BMI = col_double(),
##   log_value_eur = col_double()
## )

fit.all =
lm(log_value_eur ~ age + league_name + overall + preferred_foot + team_position + pace + shooting + pas

null = lm(log_value_eur ~ 1, data = fifa_model)
full = lm(log_value_eur ~ age + league_name + overall + preferred_foot + team_position + pace + shooting

#stepAIC(object = null, scope = list(upper = full),
#         direction = "forward", k = 2)
#
#stepAIC(object = null, scope = list(upper = full),
#         direction = "forward", k = log(2756))
#
#stepAIC(object = full, scope = list(upper = full),
#         direction = "backward", k = log(2756))
#
#stepAIC(object = null, scope = list(upper = full),
#         direction = "both", k = log(2756))
```

Final model

```
#Final MODEL FOR stepwise

#fit.AIC.fw = lm(formula = log_value_eur ~ overall + age + shooting + team_position +
# league_name + physic + defending + dribbling, data = fifa_model)

fit.BIC.fw = lm(formula = log_value_eur ~ overall + age + shooting + team_position,
  data = fifa_model)

fit.BIC.bw = lm(formula = log_value_eur ~ age + overall + team_position +
  shooting + defending + physic, data = fifa_model)

fit.BIC.sw = lm(formula = log_value_eur ~ overall + age + shooting + team_position,
  data = fifa_model)

# Stepwise is same as forward
```

```
require(broom)
```

```
## Loading required package: broom
```

```
## Warning: package 'broom' was built under R version 3.6.2
```

```
Model_sum<-bind_rows(glance(null),
  glance(full),
  glance(fit.BIC.fw),
  glance(fit.BIC.bw),
  glance(fit.BIC.sw))%>%

  round(.,3)
Model_sum$names<-c("Intercept",
  "Full",
  "BIC Forward",
  "BIC Backward",
  "BIC Stepwise")

names(Model_sum)
```

```
## [1] "r.squared"      "adj.r.squared" "sigma"          "statistic"
## [5] "p.value"        "df"             "logLik"         "AIC"
## [9] "BIC"            "deviance"       "df.residual"    "nobs"
## [13] "names"
```

```
Model_sum %>%
  dplyr::select(names, r.squared, adj.r.squared,AIC, BIC, sigma,df, df.residual)
```

```
## # A tibble: 5 x 8
##   names      r.squared adj.r.squared   AIC   BIC sigma   df df.residual
##   <chr>      <dbl>      <dbl> <dbl> <dbl> <dbl> <dbl>      <dbl>
## 1 Intercept      0          0  9736. 9748. 1.42    NA       2755
## 2 Full          0.98        0.98 -1023. -917. 0.2     16       2739
## 3 BIC Forward    0.98        0.98 -995. -954. 0.202    5       2750
## 4 BIC Backward   0.98        0.98 -1007. -953. 0.201    7       2748
## 5 BIC Stepwise   0.98        0.98 -995. -954. 0.202    5       2750
```

CV

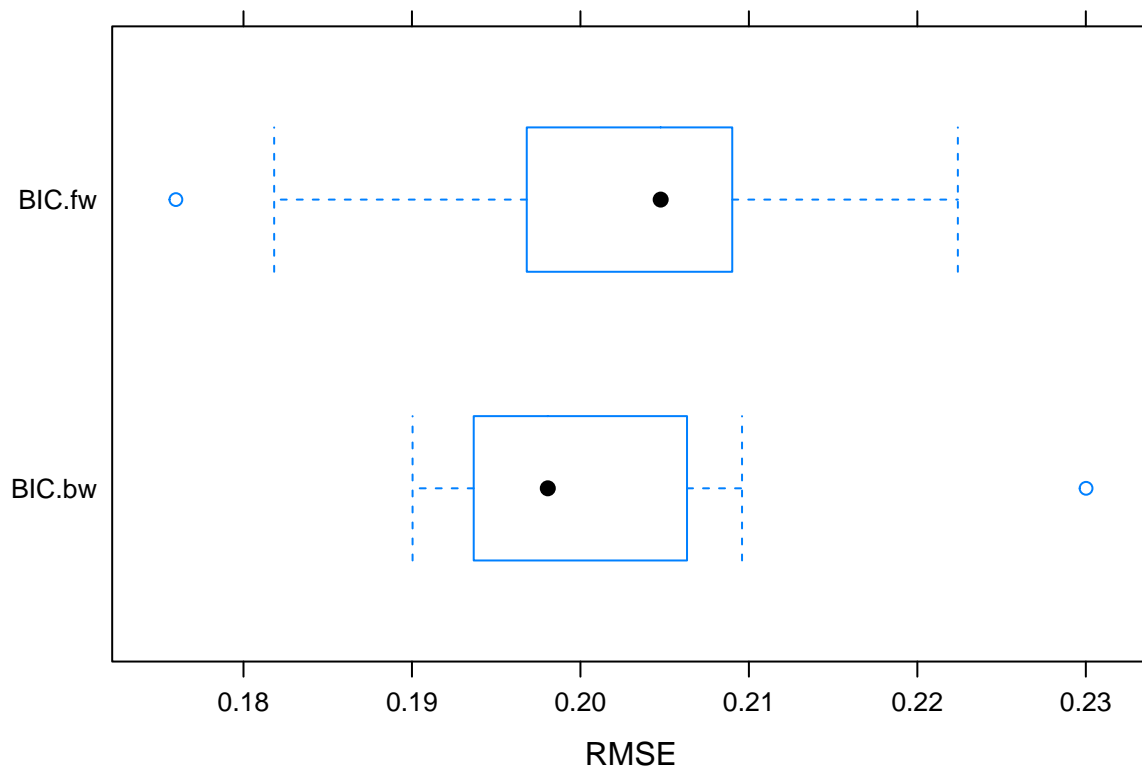
```
set.seed(6950)
tran.control = trainControl(method = "cv", number = 10)

fit.BIC.fw.cv = train(
  log_value_eur ~ overall + age + shooting + team_position,
  fifa_model, method = "lm",
  trControl = tran.control)

fit.BIC.bw.cv = train(
  log_value_eur ~ age + overall + team_position + shooting + defending + physic,
  fifa_model, method = "lm",
  trControl = tran.control)

resamp <- resamples(list(BIC.fw = fit.BIC.fw.cv,
                        BIC.bw = fit.BIC.bw.cv))
#summary(resamp)

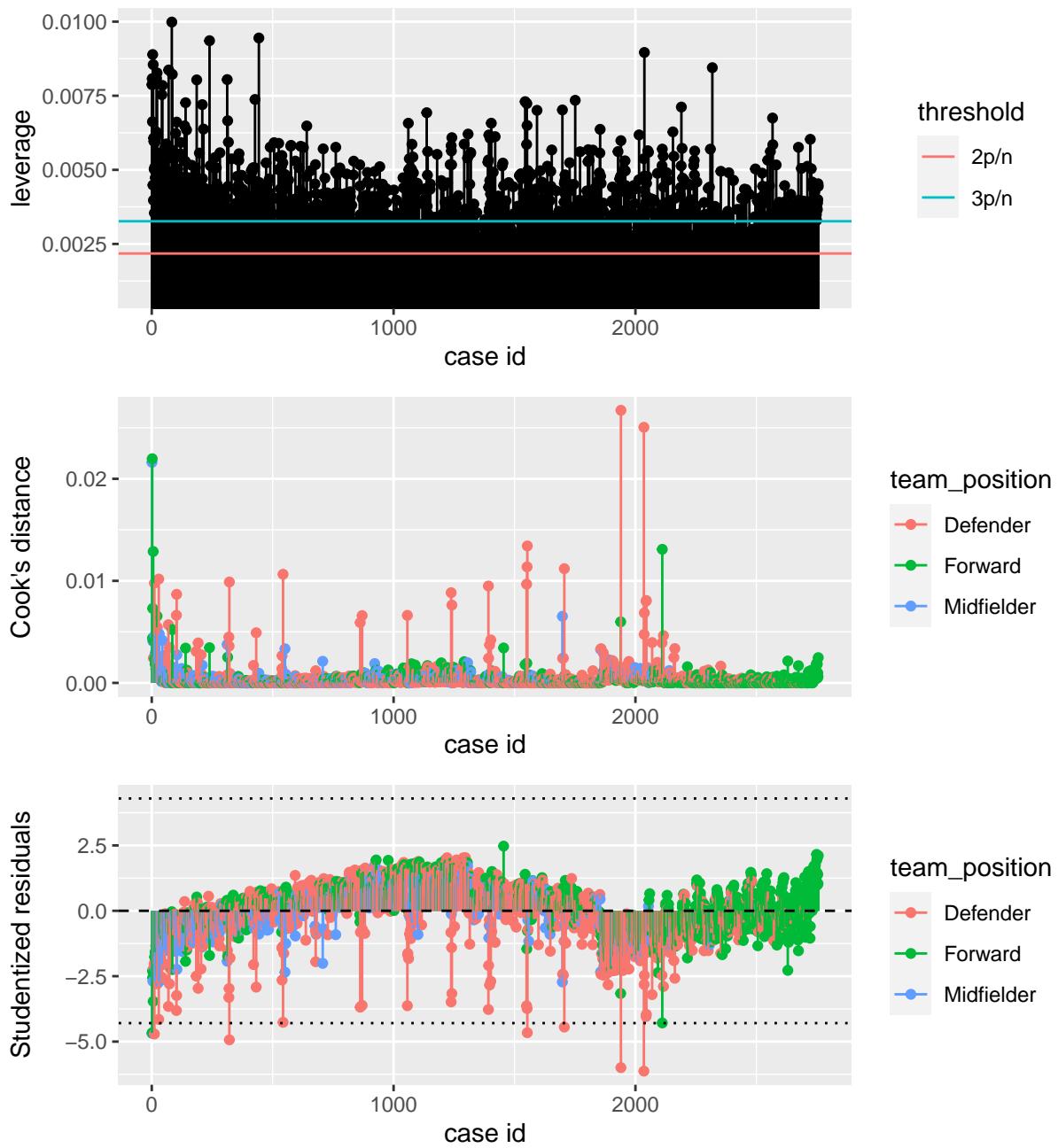
bwplot(resamp, metric = "RMSE")
```

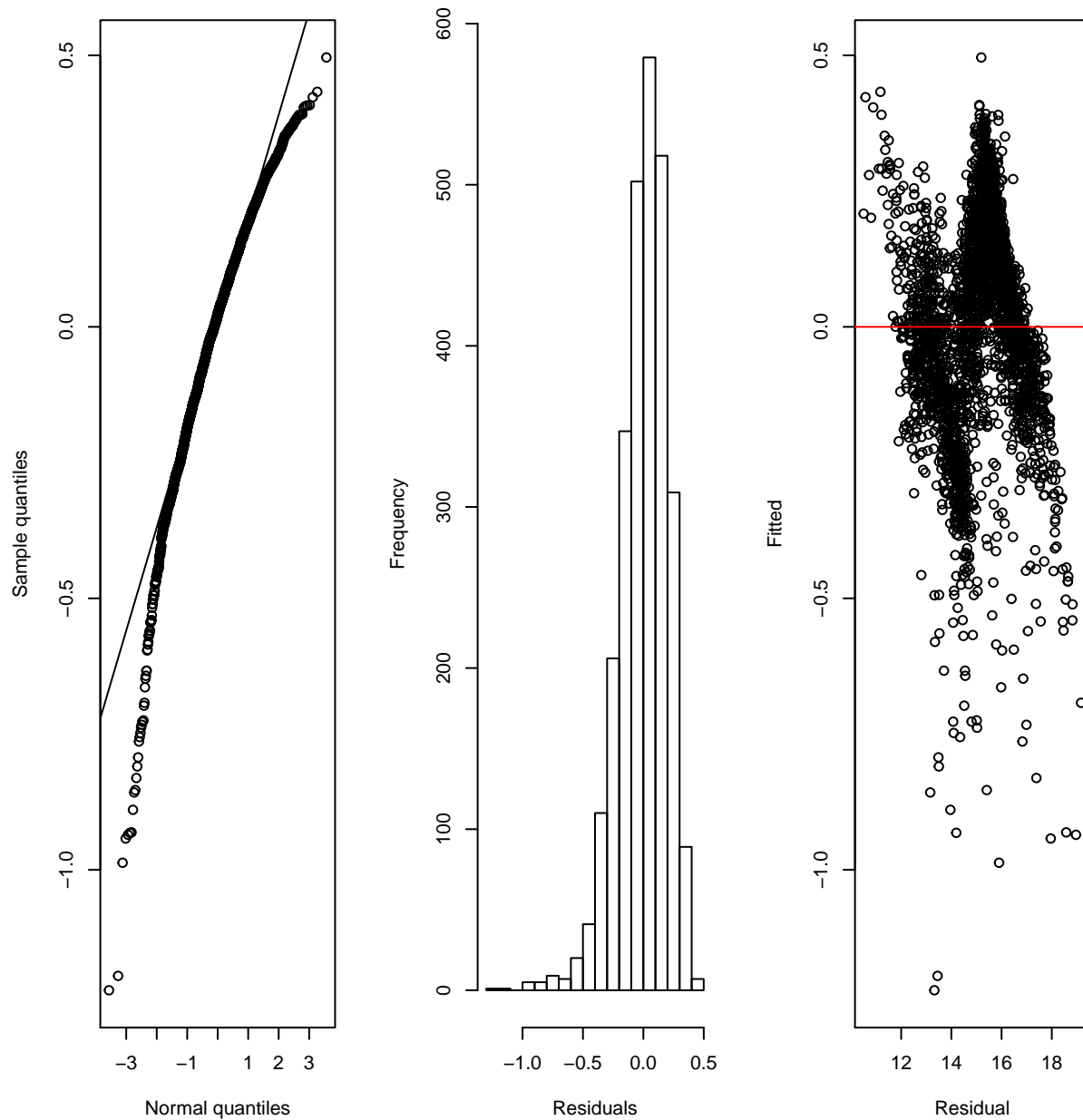


```
# select backward : fit.BIC.bw
# param 6

n.param = dim(fifa_model)[1] - fit.BIC.bw$df.residual
```

Model diagnosis





All age >30, and majority of them are defender. Therefore, we want to investigate whether there are interaction effects between team position and age.

```
fit.BIC.quart<-lm(formula = log_value_eur ~ poly(age,2) + league_name + overall + team_position +
  shooting + dribbling + defending + physic, data = fifa_model)
summary(fit.BIC.quart)
```

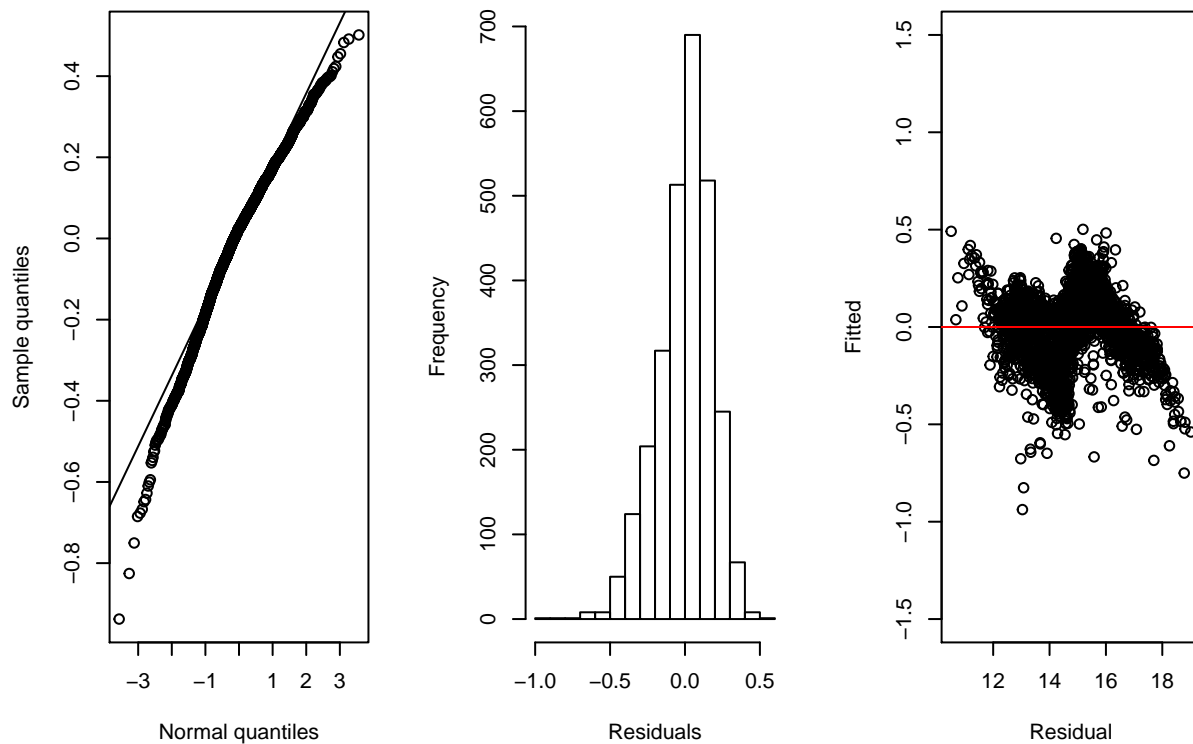
```
##
## Call:
## lm(formula = log_value_eur ~ poly(age, 2) + league_name + overall +
##   team_position + shooting + dribbling + defending + physic,
##   data = fifa_model)
##
## Residuals:
```

##	Min	1Q	Median	3Q	Max
----	-----	----	--------	----	-----

```
## -0.93781 -0.10756 0.02452 0.12707 0.50152
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.587e-01  5.238e-02   3.030 0.002468 **
## poly(age, 2)1     -1.566e+01  2.471e-01 -63.368 < 2e-16 ***
## poly(age, 2)2     -4.598e+00  2.005e-01 -22.931 < 2e-16 ***
## league_nameFrench Ligue 1    -6.097e-02  1.116e-02  -5.463 5.10e-08 ***
## league_nameGerman 1. Bundesliga -1.464e-02  1.131e-02  -1.295 0.195562
## league_nameItalian Serie A   -3.722e-02  1.092e-02  -3.408 0.000665 ***
## league_nameSpain Primera Division -3.296e-02  1.087e-02  -3.033 0.002442 **
## overall          2.052e-01  1.067e-03 192.245 < 2e-16 ***
## team_positionForward    -2.538e-02  8.974e-03  -2.828 0.004713 **
## team_positionMidfielder    2.585e-02  1.096e-02   2.358 0.018444 *
## shooting           2.323e-03  4.899e-04   4.743 2.21e-06 ***
## dribbling          4.015e-04  7.345e-04   0.547 0.584657
## defending           -9.294e-04  3.351e-04  -2.774 0.005581 **
## physic             2.369e-04  5.743e-04   0.412 0.680075
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1834 on 2742 degrees of freedom
## Multiple R-squared:  0.9833, Adjusted R-squared:  0.9832
## F-statistic: 1.239e+04 on 13 and 2742 DF,  p-value: < 2.2e-16

#fit.BIC.interact<- lm(formula = log_value_eur ~ poly(age,2) + league_name + overall * team_position +
#      shooting + dribbling + defending + physic, data = fifa_model)
#summary(fit.BIC.interact)
#
#summary(lm(data=fifa_mode))

model_diag_quart = augment(fit.BIC.quart, fifa_model)
par(mfrow=c(1,3))
qqnorm(fit.BIC.quart$residuals,
       xlab="Normal quantiles", ylab="Sample quantiles", main="")
qqline(fit.BIC.quart$residuals)
hist(fit.BIC.quart$residuals, xlab="Residuals", main="")
plot(y=model_diag_quart %>% pull(.resid) , x=model_diag_quart %>% pull(.fitted),ylim = c(-1.5,1.5),xlab=
abline(h=0,col="red")
```



```
# Appendix XX
model_diag %>%
  dplyr::select(short_name, log_value_eur, age, league_name, overall, team_position, shooting,
    dribbling, defending, physic, .ti) %>%
  filter(abs(.ti) > abs(qt((0.05/(2*dim(fifa_model)[1])), df = dim(fifa_model)[1]-9 - 1))) %>%
  knitr::kable()
```

short_name	log_value_eur	age	league_name	overall	team_position	shooting	dribbling	defending	physic	.ti
L. Messi	18.02764	33	Spain Primera Division	93	Midfielder	92	95	38	65	- 4.685423
Cristiano Ronaldo	17.64415	35	Italian Serie A	92	Forward	93	89	35	77	- 4.663520
Sergio Ramos	17.01418	34	Spain Primera Division	89	Defender	70	73	88	85	- 4.709715
B. Ivanović	14.91412	36	English Premier League	80	Defender	61	61	81	83	- 4.933934
Pedro López	13.26213	36	Spain Primera Division	72	Defender	60	69	73	65	- 4.660271
F. Peluso	13.07107	36	Italian Serie A	71	Defender	49	59	74	68	- 4.445786
N. Spolli	12.25486	37	Italian Serie A	69	Defender	39	42	71	71	- 5.996256
C. Terzi	12.10071	36	Italian Serie A	68	Defender	41	50	70	72	- 6.129129