

Final Project Data Analysis and Linear Model

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Data Analysis:

The dataset contains basketball NBA statistics for players that have a higher EFF value. The dataset contains information such as their in-game stats, points, rebounds, assists, blocks, field goal percentage, etc. There are two categorical variables, one of which is a response/target variable for our model, `CHANGED_FRANCHISE` and the other one is `SEASON_ID`.

Trends:

Basic Statistical Overview:

```
df <- read.csv('./train.csv')%>% mutate(CHANGED_FRANCHISE = factor(CHANGED_FRANCHISE))
head(df, 10)
```

##	X	PLAYER_ID	RANK	PLAYER	TEAM_ID	TEAM	GP	MIN	FGM	FGA	FG_PCT				
## 1	0	764	1	David Robinson	1610612759	SAN	82	3019	711	1378	0.516				
## 2	1	893	2	Michael Jordan	1610612741	CHI	82	3090	916	1850	0.495				
## 3	2	252	3	Karl Malone	1610612762	UTH	82	3113	789	1520	0.519				
## 4	3	165	4	Hakeem Olajuwon	1610612745	HOU	72	2797	768	1494	0.514				
## 5	4	255	5	Grant Hill	1610612765	DET	80	3260	564	1221	0.462				
## 6	5	787	6	Charles Barkley	1610612756	PHX	71	2632	580	1160	0.500				
## 7	6	358	7	Anfernee Hardaway	1610612753	ORL	82	3015	623	1215	0.513				
## 8	7	431	8	Shawn Kemp	1610612760	SEA	79	2631	526	937	0.561				
## 9	8	913	9	Larry Johnson	1610612766	CHH	81	3274	583	1225	0.476				
## 10	9	304	10	John Stockton	1610612762	UTH	82	2915	440	818	0.538				
##	FG3M	FG3A	FG3_PCT	FTM	FTA	FT_PCT	OREB	DREB	REB	AST	STL	BLK	TOV	PF	PTS
## 1	3	9	0.333	626	823	0.761	319	681	1000	247	111	271	190	262	2051
## 2	111	260	0.427	548	657	0.834	148	395	543	352	180	42	197	195	2491
## 3	16	40	0.400	512	708	0.723	175	629	804	345	138	56	199	245	2106
## 4	3	14	0.214	397	548	0.724	176	608	784	257	113	207	247	242	1936
## 5	5	26	0.192	485	646	0.751	127	656	783	548	100	48	263	242	1618
## 6	49	175	0.280	440	566	0.777	243	578	821	262	114	56	218	208	1649
## 7	89	283	0.314	445	580	0.767	129	225	354	582	166	41	229	160	1780
## 8	5	12	0.417	493	664	0.742	276	628	904	173	93	127	315	299	1550
## 9	67	183	0.366	427	564	0.757	249	434	683	355	55	43	182	173	1660
## 10	95	225	0.422	234	282	0.830	54	172	226	916	140	15	246	207	1209
##	EFF	AST_TOV	STL_TOV	CHANGED_FRANCHISE	SEASON_ID										
## 1	2626	1.30	0.58	False	1995-96										
## 2	2368	1.79	0.91	False	1995-96										
## 3	2323	1.73	0.69	False	1995-96										
## 4	2173	1.04	0.46	False	1995-96										
## 5	2016	2.08	0.38	False	1995-96										
## 6	1978	1.20	0.52	False	1995-96										
## 7	1967	2.54	0.72	False	1995-96										
## 8	1950	0.55	0.29	False	1995-96										
## 9	1835	1.95	0.30	False	1995-96										
## 10	1834	3.72	0.57	False	1995-96										

```
print(summary(df))
```

```
##          X          PLAYER_ID          RANK          PLAYER
## Min.   :    0   Min.   :    2   Min.   :    1   Length:13829
## 1st Qu.: 3457   1st Qu.:  1594   1st Qu.:119   Class :character
## Median : 6914   Median : 101129   Median :239   Mode  :character
## Mean   : 6914   Mean   : 378186   Mean   :241
## 3rd Qu.:10371   3rd Qu.: 203473   3rd Qu.:358
## Max.   :13828   Max.   :1642013   Max.   :603
## TEAM_ID          TEAM          GP          MIN
## Min.   :1.611e+09   Length:13829   Min.   :    1.00   Min.   :    1
## 1st Qu.:1.611e+09   Class :character   1st Qu.:31.00   1st Qu.: 372
## Median :1.611e+09   Mode  :character   Median :57.00   Median :1102
## Mean   :1.611e+09          Mean   :51.12   Mean   :1198
## 3rd Qu.:1.611e+09          3rd Qu.:73.00   3rd Qu.:1905
## Max.   :1.611e+09          Max.   :85.00   Max.   :3485
## FGM          FGA          FG_PCT          FG3M
## Min.   :    0.0   Min.   :    0.0   Min.   :0.0000   Min.   :    0.00
## 1st Qu.: 43.0   1st Qu.: 101.0   1st Qu.:0.3990   1st Qu.:    0.00
## Median :143.0   Median : 320.0   Median :0.4390   Median :   13.00
## Mean   :187.2   Mean   : 411.1   Mean   :0.4367   Mean   :   37.97
## 3rd Qu.:286.0   3rd Qu.: 628.0   3rd Qu.:0.4830   3rd Qu.:   62.00
## Max.   :978.0   Max.   :2173.0   Max.   :1.0000   Max.   :402.00
## FG3A          FG3_PCT          FTM          FTA
## Min.   :    0.0   Min.   :0.0000   Min.   :    0.00   Min.   :    0.0
## 1st Qu.:    3.0   1st Qu.:0.0000   1st Qu.: 16.00   1st Qu.: 24.0
## Median :   43.0   Median :0.3090   Median : 55.00   Median : 76.0
## Mean   : 106.2   Mean   :0.2487   Mean   : 90.29   Mean   :119.2
## 3rd Qu.: 176.0   3rd Qu.:0.3680   3rd Qu.:127.00   3rd Qu.:170.0
## Max.   :1028.0   Max.   :1.0000   Max.   :756.00   Max.   :972.0
## FT_PCT          OREB          DREB          REB
## Min.   :0.0000   Min.   :    0.00   Min.   :    0.0   Min.   :    0
## 1st Qu.:0.6470   1st Qu.: 12.00   1st Qu.: 42.0   1st Qu.: 57
## Median :0.7500   Median : 35.00   Median :123.0   Median : 163
## Mean   :0.6999   Mean   : 55.66   Mean   :155.4   Mean   : 211
## 3rd Qu.:0.8180   3rd Qu.: 77.00   3rd Qu.:226.0   3rd Qu.: 305
## Max.   :1.0000   Max.   :443.00   Max.   :894.0   Max.   :1247
## AST          STL          BLK          TOV
## Min.   :    0.0   Min.   :    0   Min.   :    0.00   Min.   :    0.00
## 1st Qu.: 19.0   1st Qu.: 10   1st Qu.:    4.00   1st Qu.: 18.00
## Median : 66.0   Median : 30   Median : 13.00   Median : 53.00
## Mean   :111.3   Mean   : 38   Mean   : 24.35   Mean   : 69.01
## 3rd Qu.:152.0   3rd Qu.: 57   3rd Qu.: 31.00   3rd Qu.:103.00
## Max.   :935.0   Max.   :231   Max.   :332.00   Max.   :464.00
## PF          PTS          EFF          AST_TOV
## Min.   :    0.0   Min.   :    0.0   Min.   : -8.0   Min.   :    0.000
## 1st Qu.: 39.0   1st Qu.: 113.0   1st Qu.: 135.0   1st Qu.: 0.820
## Median :100.0   Median : 376.0   Median : 453.0   Median : 1.330
## Mean   :104.3   Mean   : 502.6   Mean   : 565.5   Mean   : 1.476
## 3rd Qu.:159.0   3rd Qu.: 768.0   3rd Qu.: 869.0   3rd Qu.: 2.000
## Max.   :371.0   Max.   :2832.0   Max.   :3039.0   Max.   :21.000
## STL_TOV          CHANGED_FRANCHISE          SEASON_ID
## Min.   :0.0000   False:9539   Length:13829
## 1st Qu.:0.3700   True :4290   Class :character
## Median :0.5300          Mode  :character
## Mean   :0.6088
## 3rd Qu.:0.7500
## Max.   :7.0000
```

It can be observed from the code above, that there are two categorical variables: CHANGED_FRANCHISE and SEASON_ID

Correlation Analysis:

```
cor_matrix <- df %>%
  select(where(is.numeric)) %>%
  cor(use = "complete.obs")
print(cor_matrix)
```

```
##          X          PLAYER_ID          RANK          TEAM_ID          GP
## X          1.00000000000 0.69758068 0.188089646 0.0008072137 -0.127965017
## PLAYER_ID 0.6975806789 1.000000000 0.239181730 0.0215986827 -0.181465233
## RANK      0.1880896464 0.23918173 1.000000000 0.0045755525 -0.851062299
## TEAM_ID   0.0008072137 0.02159868 0.004575553 1.0000000000 0.006424455
## GP        -0.1279650173 -0.18146523 -0.851062299 0.0064244546 1.000000000
## MIN       -0.1262424516 -0.17695208 -0.939386069 -0.0015658663 0.860512759
## FGM       -0.0539410040 -0.11554195 -0.891443223 -0.0049534671 0.725021118
## FGA       -0.0684890272 -0.12612815 -0.879182688 -0.0046799720 0.731005091
## FG_PCT    0.0505689123 0.01092808 -0.367647969 0.0001148484 0.305932995
## FG3M      0.1717371509 0.06023054 -0.518874879 -0.0258053187 0.483023061
```

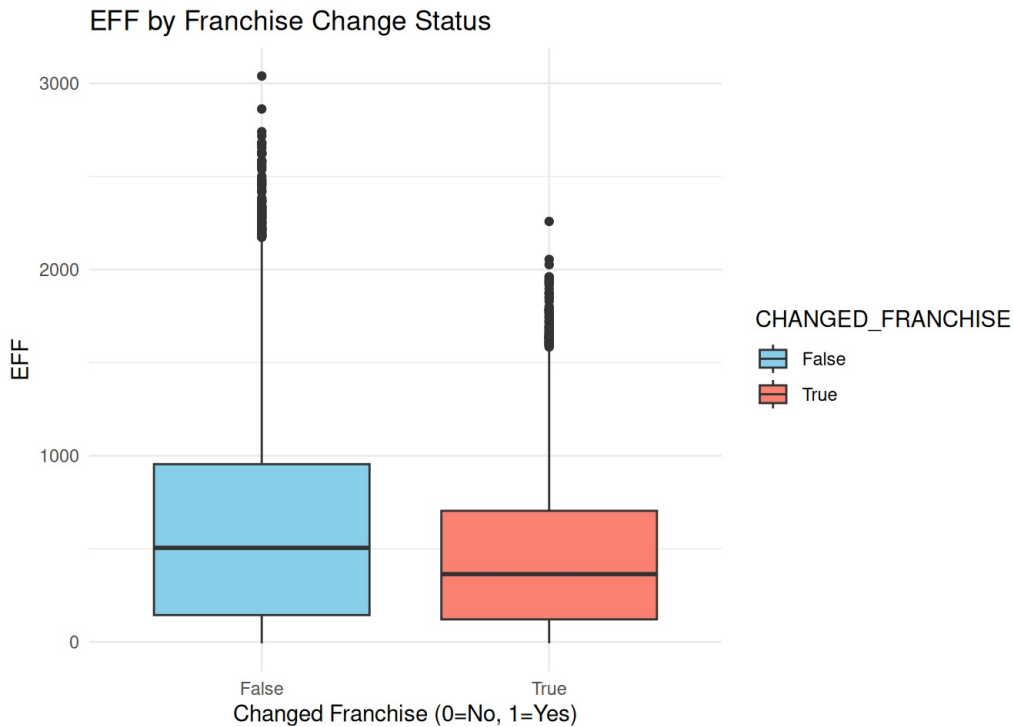
## FG3A	0.1783718551	0.06735812	-0.531988171	-0.0256319845	0.498144872
## FG3_PCT	0.1678313503	0.11388656	-0.209747391	0.0031076195	0.240353001
## FTM	-0.1186178116	-0.15276703	-0.783217597	0.0031229577	0.595410070
## FTA	-0.1358204296	-0.16256860	-0.796019205	0.0045738777	0.609977864
## FT_PCT	0.0074393483	-0.05256895	-0.393150309	0.0047170228	0.396849677
## OREB	-0.1553248198	-0.15187326	-0.683920147	-0.0010021762	0.582553869
## DREB	-0.0469579716	-0.12134592	-0.848612461	-0.0088076964	0.705640652
## REB	-0.0821521636	-0.13514393	-0.828171865	-0.0067105810	0.692928971
## AST	-0.0432926079	-0.09827119	-0.688746979	-0.0047070247	0.560885680
## STL	-0.1257618042	-0.15142420	-0.804528172	-0.0016970160	0.717438097
## BLK	-0.0803930385	-0.09781998	-0.568029861	0.0049296410	0.468990682
## TOV	-0.1441344572	-0.17442657	-0.853739053	-0.0046089680	0.708803020
## PF	-0.1978707507	-0.20652752	-0.868152599	-0.0008470234	0.857025992
## PTS	-0.0470588046	-0.11173605	-0.881312483	-0.0057282847	0.714221741
## EFF	-0.0402785069	-0.11369899	-0.925318112	-0.0072267092	0.750343608
## AST_TOV	0.1598303283	0.10045472	-0.155599597	-0.0011475117	0.165300616
## STL_TOV	0.0859808763	0.08254657	0.015601172	0.0080966682	0.060231191
##	MIN	FGM	FGA	FG_PCT	FG3M
## X	-0.126242452	-0.053941004	-0.068489027	0.0505689123	0.171737151
## PLAYER_ID	-0.176952077	-0.115541952	-0.126128145	0.0109280829	0.060230536
## RANK	-0.939386069	-0.891443223	-0.879182688	-0.3676479686	-0.518874879
## TEAM_ID	-0.001565866	-0.004953467	-0.004679972	0.0001148484	-0.025805319
## GP	0.860512759	0.725021118	0.731005091	0.3059329947	0.483023061
## MIN	1.000000000	0.922759354	0.928241051	0.2619212620	0.600240538
## FGM	0.922759354	1.000000000	0.988779093	0.2699827082	0.599261029
## FGA	0.928241051	0.988779093	1.000000000	0.2052189500	0.657262296
## FG_PCT	0.261921262	0.269982708	0.205218950	1.0000000000	-0.001889547
## FG3M	0.600240538	0.599261029	0.657262296	-0.0018895471	1.000000000
## FG3A	0.617178100	0.615853495	0.679630690	-0.0156888286	0.991505798
## FG3_PCT	0.262199215	0.253334333	0.290558782	0.0080153964	0.501796708
## FTM	0.803485176	0.891036455	0.882012097	0.2187249815	0.450741405
## FTA	0.809292355	0.889169721	0.871354138	0.2494614152	0.395463983
## FT_PCT	0.362515557	0.336525489	0.359294122	0.1631717565	0.354374099
## OREB	0.623451868	0.578558183	0.514681499	0.3905539224	-0.026553784
## DREB	0.825575909	0.803725350	0.762075394	0.3454793971	0.317306555
## REB	0.792733795	0.762973328	0.712954685	0.3718265110	0.222381606
## AST	0.736393909	0.723587381	0.749041082	0.1032383305	0.561862173
## STL	0.859038891	0.789423512	0.809356490	0.1750122064	0.554529677
## BLK	0.511563730	0.485658516	0.427424225	0.3318046704	0.007902937
## TOV	0.885247948	0.901467292	0.907195699	0.2156481422	0.509633432
## PF	0.872434179	0.765794698	0.752559111	0.3284779732	0.365642013
## PTS	0.917497434	0.993230238	0.989289990	0.2456726999	0.646111954
## EFF	0.929454217	0.956197643	0.927348452	0.3295870244	0.525293243
## AST_TOV	0.191082496	0.138082348	0.167186121	-0.0566261003	0.290810327
## STL_TOV	0.002043001	-0.075229466	-0.064331401	0.0086239699	0.065332789
##	FG3A	FG3_PCT	FTM	FTA	FT_PCT
## X	0.17837186	0.167831350	-0.118617812	-0.135820430	0.007439348
## PLAYER_ID	0.06735812	0.113886558	-0.152767027	-0.162568599	-0.052568952
## RANK	-0.53198817	-0.209747391	-0.783217597	-0.796019205	-0.393150309
## TEAM_ID	-0.02563198	0.003107619	0.003122958	0.004573878	0.004717023
## GP	0.49814487	0.240353001	0.595410070	0.609977864	0.396849677
## MIN	0.61717810	0.262199215	0.803485176	0.809292355	0.362515557
## FGM	0.61585350	0.253334333	0.891036455	0.889169721	0.336525489
## FGA	0.67963069	0.290558782	0.882012097	0.871354138	0.359294122
## FG_PCT	-0.01568883	0.008015396	0.218724982	0.249461415	0.163171757
## FG3M	0.99150580	0.501796708	0.450741405	0.395463983	0.354374099
## FG3A	1.00000000	0.494462570	0.473366659	0.419747759	0.359809840
## FG3_PCT	0.49446257	1.000000000	0.170371192	0.127050848	0.331884430
## FTM	0.47336666	0.170371192	1.000000000	0.987178184	0.307386073
## FTA	0.41974776	0.127050848	0.987178184	1.000000000	0.260056969
## FT_PCT	0.35980984	0.331884430	0.307386073	0.260056969	1.000000000
## OREB	-0.01817250	-0.156603805	0.524809210	0.597194474	0.075350690
## DREB	0.33125217	0.064091149	0.713887753	0.756147860	0.217162033
## REB	0.23510765	-0.001846324	0.681067423	0.734160356	0.181114092
## AST	0.58802015	0.303921305	0.672048334	0.647442988	0.309236852
## STL	0.58107882	0.277805680	0.706390421	0.705370275	0.311077115
## BLK	0.01367898	-0.128494008	0.442706604	0.508509829	0.047366992
## TOV	0.53802630	0.214324558	0.867384229	0.872096527	0.312314610
## PF	0.37921808	0.101792692	0.670362709	0.702517917	0.285673567
## PTS	0.66232579	0.277239057	0.921348021	0.911189845	0.352396411
## EFF	0.53848776	0.197593676	0.877270338	0.886745226	0.313339217
## AST_TOV	0.29949463	0.345670529	0.080102557	0.048552018	0.265486079
## STL_TOV	0.06623744	0.140395599	-0.116499164	-0.120730281	0.101045393
##	OREB	DREB	REB	AST	STL
## X	-0.155324820	-0.046957972	-0.082152164	-0.043292608	-0.125761804
## PLAYER_ID	-0.151873260	-0.121345921	-0.135143932	-0.098271193	-0.151424200
## RANK	-0.683920147	-0.848612461	-0.828171865	-0.688746979	-0.804528172
## TEAM_ID	-0.001002176	-0.008807696	-0.006710581	-0.004707025	-0.001697016
## GP	0.582553869	0.705640652	0.692928971	0.560885680	0.717438097

## MIN	0.623451868	0.825575909	0.792733795	0.736393909	0.859038891
## FGM	0.578558183	0.803725350	0.762973328	0.723587381	0.789423512
## FGA	0.514681499	0.762075394	0.712954685	0.749041082	0.809356490
## FG_PCT	0.390553922	0.345479397	0.371826511	0.103238331	0.175012206
## FG3M	-0.026553784	0.317306555	0.222381606	0.561862173	0.554529677
## FG3A	-0.018172498	0.331252174	0.235107651	0.588020146	0.581078823
## FG3_PCT	-0.156603805	0.064091149	-0.001846324	0.303921305	0.277805680
## FTM	0.524809210	0.713887753	0.681067423	0.672048334	0.706390421
## FTA	0.597194474	0.756147860	0.734160356	0.647442988	0.705370275
## FT_PCT	0.075350690	0.217162033	0.181114092	0.309236852	0.311077115
## OREB	1.000000000	0.837270141	0.917662243	0.206039272	0.440744270
## DREB	0.837270141	1.000000000	0.985604169	0.484378476	0.645257012
## REB	0.917662243	0.985604169	1.000000000	0.415714262	0.605199099
## AST	0.206039272	0.484378476	0.415714262	1.000000000	0.777677146
## STL	0.440744270	0.645257012	0.605199099	0.777677146	1.000000000
## BLK	0.753435042	0.723554590	0.758784417	0.145495244	0.345128376
## TOV	0.530761481	0.744082454	0.704850896	0.837485519	0.815423224
## PF	0.742686765	0.813925928	0.821135417	0.518242220	0.718796069
## PTS	0.536373181	0.780038568	0.732715980	0.738546655	0.793533740
## EFF	0.697376053	0.903039586	0.871885608	0.739809715	0.807619912
## AST_TOV	-0.174557811	-0.001908639	-0.055360949	0.456518981	0.272933602
## STL_TOV	-0.061375992	-0.056532965	-0.060061158	-0.043124695	0.179048249
##	BLK	TOV	PF	PTS	EFF
## X	-0.080393038	-0.144134457	-0.1978707507	-0.047058805	-0.040278507
## PLAYER_ID	-0.097819981	-0.174426567	-0.2065275244	-0.111736046	-0.113698992
## RANK	-0.568029861	-0.853739053	-0.8681525994	-0.881312483	-0.925318112
## TEAM_ID	0.004929641	-0.004608968	-0.0008470234	-0.005728285	-0.007226709
## GP	0.468990682	0.708803020	0.8570259917	0.714221741	0.750343608
## MIN	0.511563730	0.885247948	0.8724341791	0.917497434	0.929454217
## FGM	0.485658516	0.901467292	0.7657946981	0.993230238	0.956197643
## FGA	0.427424225	0.907195699	0.7525591108	0.989289990	0.927348452
## FG_PCT	0.331804670	0.215648142	0.3284779732	0.245672700	0.329587024
## FG3M	0.007902937	0.509633432	0.3656420127	0.646111954	0.525293243
## FG3A	0.013678980	0.538026302	0.3792180812	0.662325790	0.538487764
## FG3_PCT	-0.128494008	0.214324558	0.1017926918	0.277239057	0.197593676
## FTM	0.442706604	0.867384229	0.6703627085	0.921348021	0.877270338
## FTA	0.508509829	0.872096527	0.7025179167	0.911189845	0.886745226
## FT_PCT	0.047366992	0.312314610	0.2856735670	0.352396411	0.313339217
## OREB	0.753435042	0.530761481	0.7426867649	0.536373181	0.697376053
## DREB	0.723554590	0.744082454	0.8139259283	0.780038568	0.903039586
## REB	0.758784417	0.704850896	0.8211354166	0.732715980	0.871885608
## AST	0.145495244	0.837485519	0.5182422200	0.738546655	0.739809715
## STL	0.345128376	0.815423224	0.7187960692	0.793533740	0.807619912
## BLK	1.000000000	0.435903416	0.6248471940	0.453984718	0.602875331
## TOV	0.435903416	1.000000000	0.7694513881	0.906173450	0.893995836
## PF	0.624847194	0.769451388	1.0000000000	0.747899868	0.812549620
## PTS	0.453984718	0.906173450	0.7478998682	1.000000000	0.950128653
## EFF	0.602875331	0.893995836	0.8125496198	0.950128653	1.000000000
## AST_TOV	-0.161081192	0.152325629	0.0160242690	0.150173222	0.156083892
## STL_TOV	-0.056113583	-0.149420392	-0.0259122088	-0.073677927	-0.047509393
##	AST_TOV	STL_TOV			
## X	0.159830328	0.085980876			
## PLAYER_ID	0.100454722	0.082546572			
## RANK	-0.155599597	0.015601172			
## TEAM_ID	-0.001147512	0.008096668			
## GP	0.165300616	0.060231191			
## MIN	0.191082496	0.002043001			
## FGM	0.138082348	-0.075229466			
## FGA	0.167186121	-0.064331401			
## FG_PCT	-0.056626100	0.008623970			
## FG3M	0.290810327	0.065332789			
## FG3A	0.299494634	0.066237441			
## FG3_PCT	0.345670529	0.140395599			
## FTM	0.080102557	-0.116499164			
## FTA	0.048552018	-0.120730281			
## FT_PCT	0.265486079	0.101045393			
## OREB	-0.174557811	-0.061375992			
## DREB	-0.001908639	-0.056532965			
## REB	-0.055360949	-0.060061158			
## AST	0.456518981	-0.043124695			
## STL	0.272933602	0.179048249			
## BLK	-0.161081192	-0.056113583			
## TOV	0.152325629	-0.149420392			
## PF	0.016024269	-0.025912209			
## PTS	0.150173222	-0.073677927			
## EFF	0.156083892	-0.047509393			
## AST_TOV	1.000000000	0.416675373			
## STL_TOV	0.416675373	1.000000000			

Looking at the large blob above, there seems to be a very high correlation (often times ≈ 0.9) among variables that represent similar or related stats such as `c('FGM', 'FG_PCT', 'FG3M')` that represent different 'Field Goal' metrics.

Visualization:

```
box_plot <- ggplot(df, aes(x = CHANGED_FRANCHISE, y = EFF, fill = CHANGED_FRANCHISE)) +  
  geom_boxplot() +  
  scale_fill_manual(values = c("skyblue", "salmon")) +  
  labs(title = "EFF by Franchise Change Status",  
       x = "Changed Franchise (0=No, 1=Yes)",  
       y = "EFF") +  
  theme_minimal()  
print(box_plot)
```



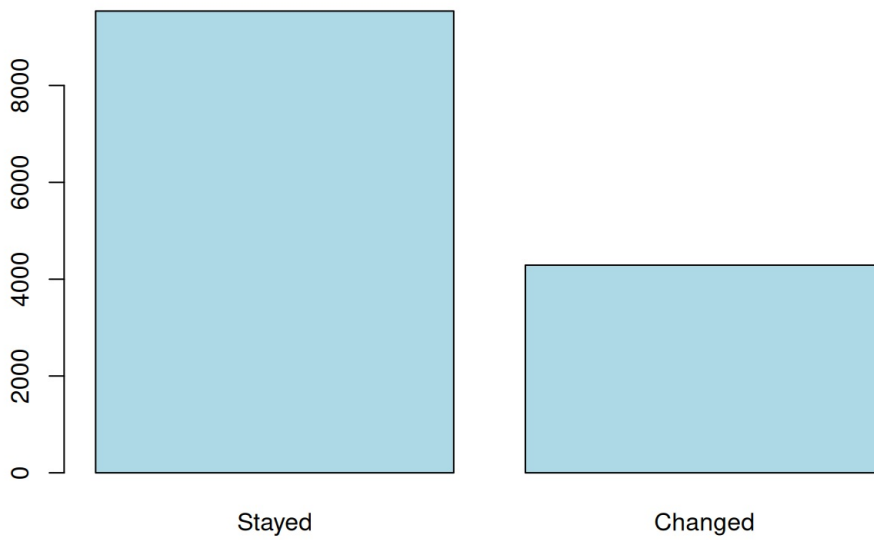
There are statistically significant outliers as the EFF increases for both Categories (CHANGED_FRANCHISE : True or False)

```
table(df$CHANGED_FRANCHISE)
```

```
##  
## False  True  
##  9539  4290
```

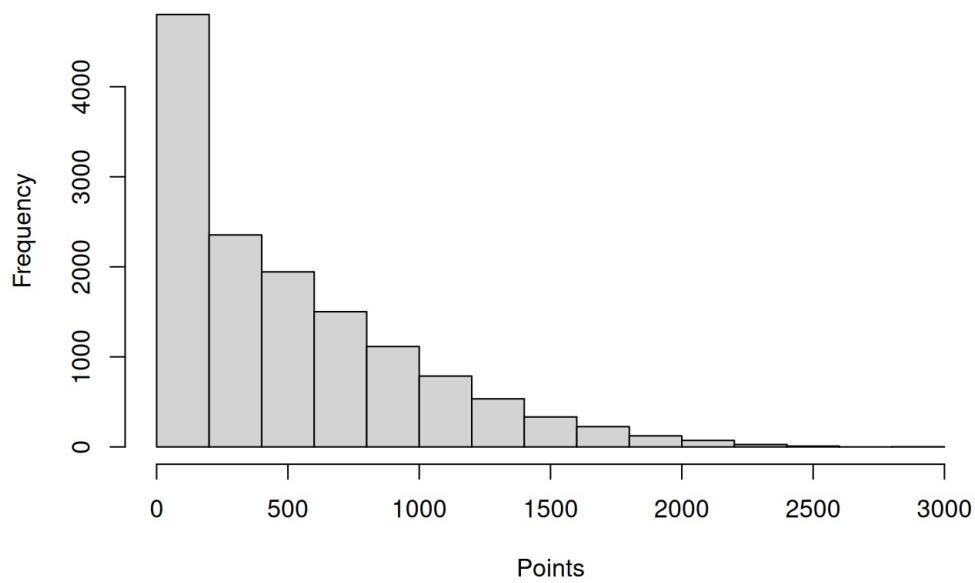
```
barplot(table(df$CHANGED_FRANCHISE),  
       names.arg = c("Stayed", "Changed"), col = "lightblue",  
       main = "Distribution of Franchise Changes")
```

Distribution of Franchise Changes



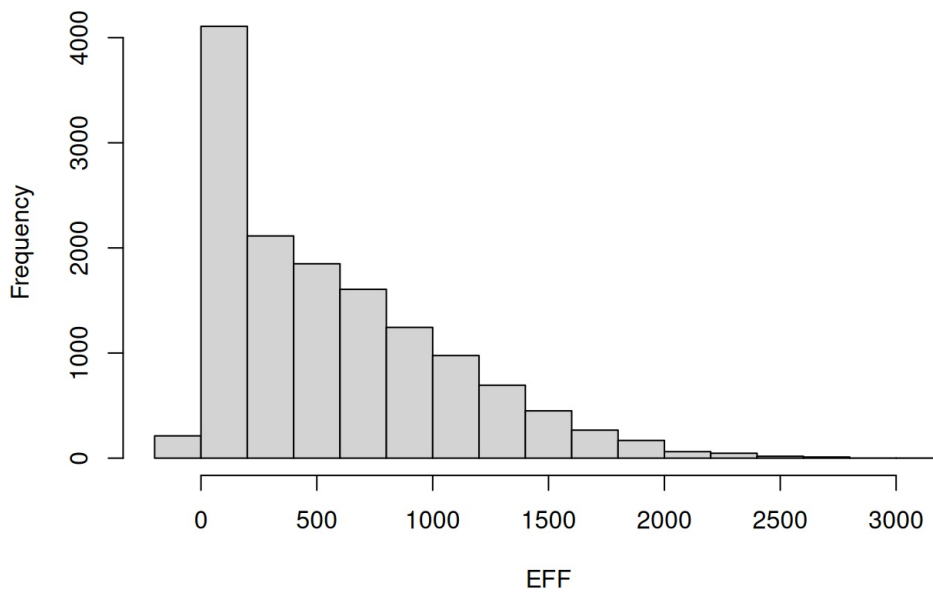
```
hist(df$PTS, main = "Distribution of Points per Game", xlab = "Points")
```

Distribution of Points per Game



```
hist(df$EFF, main = "Player Efficiency Distribution", xlab = "EFF")
```

Player Efficiency Distribution



Understanding Outliers:

```
numeric_cols <- c("EFF")
z_scores <- df %>%
  select(all_of(numeric_cols)) %>%
  scale() %>%
  abs()

outliers <- z_scores > 3
df_outliers <- df %>%
  filter(rowSums(outliers) > 0) %>%
  select(PLAYER, all_of(numeric_cols))

print(df_outliers)
```

```
##          PLAYER  EFF
## 1    David Robinson 2626
## 2    Michael Jordan 2368
## 3      Karl Malone 2323
## 4   Hakeem Olajuwon 2173
## 5      Karl Malone 2478
## 6    Michael Jordan 2215
## 7      Grant Hill 2130
## 8      Karl Malone 2370
## 9      Tim Duncan 2170
## 10   Shaquille O'Neal 2672
## 11     Kevin Garnett 2330
## 12      Karl Malone 2223
## 13      Gary Payton 2159
## 14     Chris Webber 2093
## 15   Shaquille O'Neal 2293
## 16     Kevin Garnett 2251
## 17      Tim Duncan 2125
## 18      Tim Duncan 2558
## 19     Kevin Garnett 2279
## 20     Kevin Garnett 2630
## 21      Tim Duncan 2425
## 22      Kobe Bryant 2298
## 23     Dirk Nowitzki 2218
## 24     Tracy McGrady 2160
## 25     Kevin Garnett 2717
## 26     Kevin Garnett 2621
## 27     LeBron James 2259
## 28     Dirk Nowitzki 2194
## 29   Amar'e Stoudemire 2141
## 30     Shawn Marion 2073
## 31     Shawn Marion 2337
## 32     LeBron James 2323
## 33     Kevin Garnett 2303
## 34     Elton Brand 2253
## 35     Kobe Bryant 2226
```

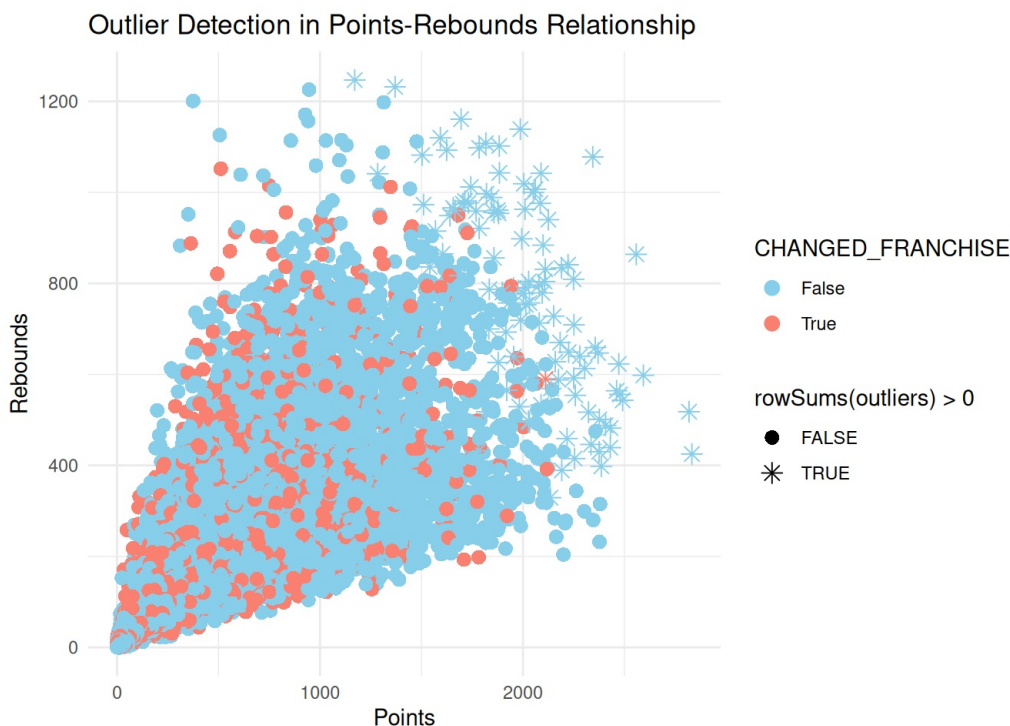
## 36	Dirk Nowitzki	2218
## 37	Kevin Garnett	2217
## 38	Kobe Bryant	2129
## 39	Dirk Nowitzki	2098
## 40	LeBron James	2275
## 41	Amar'e Stoudemire	2247
## 42	Chris Paul	2231
## 43	Dwight Howard	2194
## 44	Kobe Bryant	2181
## 45	LeBron James	2501
## 46	Chris Paul	2376
## 47	Dwyane Wade	2314
## 48	Dwight Howard	2133
## 49	LeBron James	2464
## 50	Kevin Durant	2293
## 51	David Lee	2186
## 52	Dwight Howard	2101
## 53	LeBron James	2258
## 54	Dwight Howard	2208
## 55	Blake Griffin	2102
## 56	Pau Gasol	2083
## 57	Kevin Durant	2462
## 58	LeBron James	2446
## 59	Kevin Durant	2572
## 60	Kevin Love	2328
## 61	LeBron James	2255
## 62	Blake Griffin	2082
## 63	James Harden	2202
## 64	Chris Paul	2125
## 65	Stephen Curry	2073
## 66	Stephen Curry	2424
## 67	Russell Westbrook	2283
## 68	James Harden	2281
## 69	Kevin Durant	2149
## 70	LeBron James	2092
## 71	Russell Westbrook	2740
## 72	James Harden	2623
## 73	Karl-Anthony Towns	2485
## 74	Anthony Davis	2336
## 75	LeBron James	2291
## 76	Giannis Antetokounmpo	2270
## 77	LeBron James	2681
## 78	Anthony Davis	2476
## 79	Karl-Anthony Towns	2384
## 80	Russell Westbrook	2359
## 81	Giannis Antetokounmpo	2306
## 82	James Harden	2170
## 83	Andre Drummond	2125
## 84	James Harden	2581
## 85	Giannis Antetokounmpo	2538
## 86	Karl-Anthony Towns	2338
## 87	Nikola Jokić	2311
## 88	Nikola Vučević	2242
## 89	Rudy Gobert	2183
## 90	Kevin Durant	2178
## 91	Andre Drummond	2153
## 92	Russell Westbrook	2118
## 93	Paul George	2110
## 94	James Harden	2220
## 95	Giannis Antetokounmpo	2180
## 96	Nikola Jokić	2585
## 97	Nikola Jokić	2862
## 98	Giannis Antetokounmpo	2343
## 99	Joel Embiid	2304
## 100	Karl-Anthony Towns	2088
## 101	Nikola Jokić	2622
## 102	Domantas Sabonis	2456
## 103	Joel Embiid	2369
## 104	Luka Dončić	2214
## 105	Jayson Tatum	2209
## 106	Shai Gilgeous-Alexander	2073
## 107	Nikola Jokić	3039
## 108	Domantas Sabonis	2679
## 109	Giannis Antetokounmpo	2655
## 110	Luka Dončić	2580
## 111	Anthony Davis	2548
## 112	Shai Gilgeous-Alexander	2416
## 113	LeBron James	2126
## 114	Kevin Durant	2075

Looking at the list of outliers based on EFF values, it can be said that the list boils down to some of the more popular household names in basketball.

```
pts_iqr <- df %>%
  summarise(
    Q1 = quantile(PTS, 0.25, na.rm = TRUE),
    Q3 = quantile(PTS, 0.75, na.rm = TRUE),
    IQR = Q3 - Q1
  )

pts_outliers <- df %>%
  filter(PTS < (pts_iqr$Q1 - 1.5 * pts_iqr$IQR) |
         PTS > (pts_iqr$Q3 + 1.5 * pts_iqr$IQR))

outlier_plot <- ggplot(df, aes(x = PTS, y = REB)) +
  geom_point(aes(color = CHANGED_FRANCHISE, shape = rowSums(outliers) > 0), size = 3) +
  scale_color_manual(values = c("skyblue", "salmon")) +
  scale_shape_manual(values = c(16, 8)) +
  labs(title = "Outlier Detection in Points-Rebounds Relationship",
       x = "Points",
       y = "Rebounds") +
  theme_minimal()
print(outlier_plot)
```



Model Creation:

Logistic Regression Model:

The model aims to predict CHANGED_FRANCHISE, a categorical variable of boolean type based on all of the other relevant parameters.

We decided to use a logistic regression because our response variable is binomial, a player changes franchise or does not change franchise.

```
# Selecting desired columns.
cols_list <- c('RANK', 'TEAM_ID', 'GP', 'MIN', 'FGM', 'FG_PCT', 'FG3M', 'FG3A', 'FG3_PCT', 'FTM', 'FTA', 'FT_PCT',
              'OREB', 'DREB', 'REB', 'AST', 'STL', 'BLK', 'TOV', 'PF', 'PTS', 'EFF', 'AST_TOV', 'STL_TOV', 'CHANGED_FRANCHISE',
              'SEASON_ID')
df <- df[, cols_list]
#head(train_df, 10)

# Convert to factor.
df$CHANGED_FRANCHISE <- as.factor(df$CHANGED_FRANCHISE)
df$SEASON_ID <- as.factor(df$SEASON_ID)

#Logistic regression model
logit_mod <- glm(CHANGED_FRANCHISE ~ ., family=binomial(link='logit'), data = df)
```

```
summary(logit_mod)
```

```
##
## Call:
## glm(formula = CHANGED_FRANCHISE ~ ., family = binomial(link = "logit"),
##      data = df)
##
## Coefficients: (2 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  9.040e+06  3.534e+06   2.558 0.010521 *
## RANK         -4.685e-03  7.051e-04 -6.645 3.04e-11 ***
## TEAM_ID      -5.612e-03  2.194e-03 -2.558 0.010521 *
## GP           6.780e-04  1.996e-03   0.340 0.734053
## MIN          3.643e-05  1.229e-04   0.296 0.767000
## FGM          -1.261e-03  9.497e-04 -1.328 0.184273
## FG_PCT       -1.762e-01  2.360e-01 -0.747 0.455131
## FG3M         8.186e-03  4.025e-03   2.034 0.041980 *
## FG3A        -3.087e-03  1.504e-03 -2.053 0.040073 *
## FG3_PCT      2.397e-02  1.349e-01   0.178 0.858962
## FTM          8.610e-04  2.246e-03   0.383 0.701448
## FTA         -8.039e-04  1.661e-03 -0.484 0.628459
## FT_PCT       1.225e-01  1.160e-01   1.055 0.291288
## OREB         -8.225e-04  1.222e-03 -0.673 0.500836
## DREB         3.064e-03  8.854e-04   3.460 0.000539 ***
## REB          NA         NA         NA         NA
## AST          9.928e-05  9.098e-04   0.109 0.913103
## STL         -1.036e-03  1.694e-03 -0.612 0.540649
## BLK         -4.722e-03  1.400e-03 -3.372 0.000745 ***
## TOV          2.853e-03  1.707e-03   1.671 0.094683 .
## PF          -7.020e-04  8.128e-04 -0.864 0.387774
## PTS          NA         NA         NA         NA
## EFF         -2.424e-03  7.161e-04 -3.384 0.000713 ***
## AST_TOV      5.576e-02  2.823e-02   1.975 0.048223 *
## STL_TOV     -6.334e-03  5.777e-02 -0.110 0.912684
## SEASON_ID1996-97 1.593e+01  1.122e+02   0.142 0.887104
## SEASON_ID1997-98 1.575e+01  1.122e+02   0.140 0.888314
## SEASON_ID1998-99 1.523e+01  1.122e+02   0.136 0.892002
## SEASON_ID1999-00 1.559e+01  1.122e+02   0.139 0.889497
## SEASON_ID2000-01 1.570e+01  1.122e+02   0.140 0.888664
## SEASON_ID2001-02 1.565e+01  1.122e+02   0.140 0.889008
## SEASON_ID2002-03 1.553e+01  1.122e+02   0.138 0.889853
## SEASON_ID2003-04 1.600e+01  1.122e+02   0.143 0.886544
## SEASON_ID2004-05 1.620e+01  1.122e+02   0.144 0.885170
## SEASON_ID2005-06 1.565e+01  1.122e+02   0.139 0.889067
## SEASON_ID2006-07 1.537e+01  1.122e+02   0.137 0.891015
## SEASON_ID2007-08 1.562e+01  1.122e+02   0.139 0.889263
## SEASON_ID2008-09 1.579e+01  1.122e+02   0.141 0.888038
## SEASON_ID2009-10 1.581e+01  1.122e+02   0.141 0.887891
## SEASON_ID2010-11 1.621e+01  1.122e+02   0.144 0.885125
## SEASON_ID2011-12 1.540e+01  1.122e+02   0.137 0.890792
## SEASON_ID2012-13 1.598e+01  1.122e+02   0.142 0.886707
## SEASON_ID2013-14 1.598e+01  1.122e+02   0.142 0.886714
## SEASON_ID2014-15 1.587e+01  1.122e+02   0.142 0.887471
## SEASON_ID2015-16 1.578e+01  1.122e+02   0.141 0.888146
## SEASON_ID2016-17 1.589e+01  1.122e+02   0.142 0.887371
## SEASON_ID2017-18 1.578e+01  1.122e+02   0.141 0.888114
## SEASON_ID2018-19 1.598e+01  1.122e+02   0.142 0.886690
## SEASON_ID2019-20 1.592e+01  1.122e+02   0.142 0.887129
## SEASON_ID2020-21 1.591e+01  1.122e+02   0.142 0.887213
## SEASON_ID2021-22 1.600e+01  1.122e+02   0.143 0.886567
## SEASON_ID2022-23 1.591e+01  1.122e+02   0.142 0.887192
## SEASON_ID2023-24 1.597e+01  1.122e+02   0.142 0.886758
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 17128  on 13828  degrees of freedom
## Residual deviance: 16225  on 13778  degrees of freedom
## AIC: 16327
##
## Number of Fisher Scoring iterations: 15
```

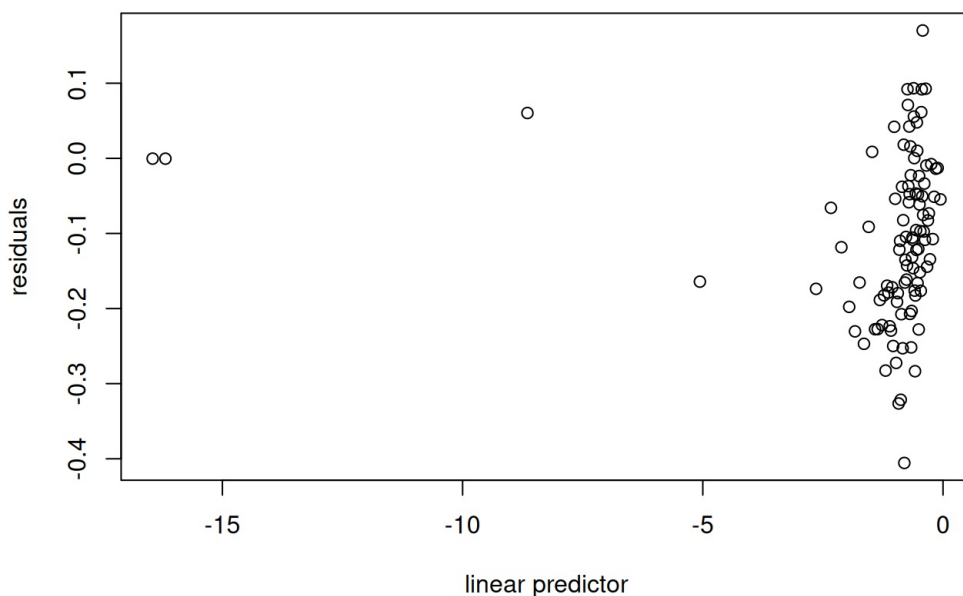
Diagnostics For Logistic Regression:

```
df <- mutate(df,
              residuals = residuals(logit_mod),
              linpred = predict(logit_mod))

gdf <- group_by(df,
                bin = cut(linpred, breaks = unique(quantile(linpred, (1:100)/101))))

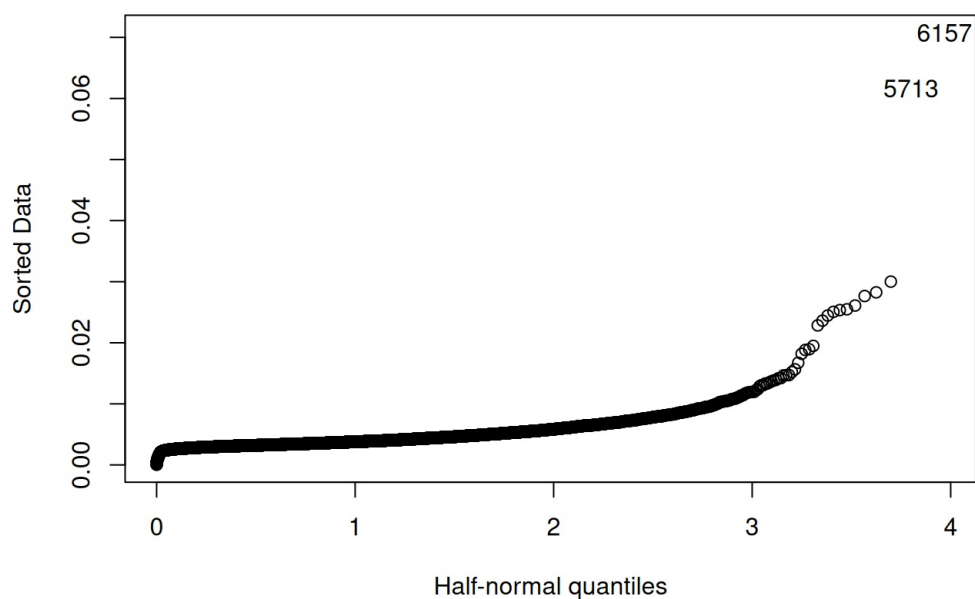
diagdf <- summarise(gdf,
                    residuals = mean(residuals),
                    linpred = mean(linpred))

plot(residuals ~ linpred, diagdf, xlab = "linear predictor")
```



The values seem to cluster up close to zero, and this breaks our constant variance assumption. Because there is not a random scatter in our fitted versus residual plot, we can say that this model does not meet our assumption of linearity.

```
#Half norm plot
library(faraway)
halfnorm(hatvalues(logit_mod))
```



From the half norm plot, we can see that there are only two outliers. The plot seems to look similar to a sigmoid function, which is expected since the model is logistic regression.

Model Comparison:

```
#Smaller model from omitting non significant predictors
columns <- c('RANK', 'TEAM_ID', 'FG3M', 'FG3A', 'DREB', 'BLK', 'EFF', 'AST_TOV', 'CHANGED_FRANCHISE')
new_train <- df[, columns]

new_train$CHANGED_FRANCHISE <- as.factor(new_train$CHANGED_FRANCHISE)

logit_mod_reduced <- glm(CHANGED_FRANCHISE ~ ., family=binomial(link='logit'), data = new_train)
#Summary of new log mod
summary(logit_mod_reduced)
```

```
##
## Call:
## glm(formula = CHANGED_FRANCHISE ~ ., family = binomial(link = "logit"),
##      data = new_train)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  8.938e+06  3.484e+06   2.566   0.0103 *
## RANK         -2.638e-03  3.886e-04  -6.788 1.14e-11 ***
## TEAM_ID      -5.550e-03  2.163e-03  -2.566   0.0103 *
## FG3M          4.091e-03  2.997e-03   1.365   0.1722
## FG3A         -9.430e-04  1.150e-03  -0.820   0.4121
## DREB          3.328e-03  4.265e-04   7.802 6.08e-15 ***
## BLK          -4.883e-03  1.046e-03  -4.668 3.05e-06 ***
## EFF          -2.203e-03  1.629e-04 -13.524 < 2e-16 ***
## AST_TOV       1.033e-01  2.029e-02   5.093 3.52e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 17128  on 13828  degrees of freedom
## Residual deviance: 16681  on 13820  degrees of freedom
## AIC: 16699
##
## Number of Fisher Scoring iterations: 4
```

```
#Compare the reduced vs full models
anova(logit_mod_reduced, logit_mod, test = "Chisq")
```

```
## Analysis of Deviance Table
##
## Model 1: CHANGED_FRANCHISE ~ RANK + TEAM_ID + FG3M + FG3A + DREB + BLK +
##      EFF + AST_TOV
## Model 2: CHANGED_FRANCHISE ~ RANK + TEAM_ID + GP + MIN + FGM + FG_PCT +
##      FG3M + FG3A + FG3_PCT + FTM + FTA + FT_PCT + OREB + DREB +
##      REB + AST + STL + BLK + TOV + PF + PTS + EFF + AST_TOV +
##      STL_TOV + SEASON_ID
##      Resid. Df Resid. Dev Df Deviance  Pr(>Chi)
## 1      13820      16681
## 2      13778      16225 42    456.43 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

The p-value is 2.2e-16 which is extremely small. This means that the additional predictors in Model 2(the full model) significantly improve the fit of the model and help explain the response variable(changed franchise).

Hypothesis Test

Null Hypothesis(H_0): The reduced model(Model 1) fits just as well as the full model(Model 2) and the additional predictors in the full model do not improve the model.

Alternative Hypothesis(H_1): The full model(Model 2) is better fit than the reduced model(Model 1). $\alpha = 0.05$

$D = \text{Deviance}_{\{\text{Model1}\}} - \text{Deviance}_{\{\text{Model2}\}} = 16681 - 16225 = 456.43$

$D = 456.43$

$df = 42$

$p\text{-value} = 2.2e-16$

Conclusion:

Since the p-value is extremely small, we reject the null hypothesis. This indicates that the full model offers a significantly improved fit over the reduced model, and the additional predictors included in the full model are meaningful in explaining the response variable.

Confidence Interval:

```
ci <- confint.default(logit_mod) # default 95% CI based on standard error.  
print(ci)
```

```
##              2.5 %      97.5 %  
## (Intercept)  2.113917e+06 1.596509e+07  
## RANK         -6.066968e-03 -3.303133e-03  
## TEAM_ID     -9.912439e-03 -1.312501e-03  
## GP          -3.233260e-03  4.589204e-03  
## MIN         -2.045272e-04  2.773803e-04  
## FGM         -3.122370e-03  6.004597e-04  
## FG_PCT      -6.387294e-01  2.862465e-01  
## FG3M        2.968644e-04  1.607413e-02  
## FG3A       -6.033614e-03 -1.398778e-04  
## FG3_PCT     -2.404015e-01  2.883386e-01  
## FTM         -3.540946e-03  5.262984e-03  
## FTA         -4.060248e-03  2.452347e-03  
## FT_PCT      -1.049855e-01  3.499189e-01  
## OREB        -3.217301e-03  1.572263e-03  
## DREB        1.328452e-03  4.799117e-03  
## REB         NA           NA  
## AST        -1.683820e-03  1.882375e-03  
## STL        -4.355693e-03  2.283248e-03  
## BLK        -7.466725e-03 -1.977690e-03  
## TOV        -4.929675e-04  6.198874e-03  
## PF         -2.294969e-03  8.910467e-04  
## PTS         NA           NA  
## EFF        -3.827187e-03 -1.020099e-03  
## AST_TOV     4.357724e-04  1.110850e-01  
## STL_TOV    -1.195542e-01  1.068855e-01  
## SEASON_ID1996-97 -2.039316e+02  2.357823e+02  
## SEASON_ID1997-98 -2.041034e+02  2.356105e+02  
## SEASON_ID1998-99 -2.046269e+02  2.350870e+02  
## SEASON_ID1999-00 -2.042714e+02  2.354425e+02  
## SEASON_ID2000-01 -2.041531e+02  2.355608e+02  
## SEASON_ID2001-02 -2.042020e+02  2.355119e+02  
## SEASON_ID2002-03 -2.043220e+02  2.353919e+02  
## SEASON_ID2003-04 -2.038521e+02  2.358617e+02  
## SEASON_ID2004-05 -2.036569e+02  2.360570e+02  
## SEASON_ID2005-06 -2.042104e+02  2.355035e+02  
## SEASON_ID2006-07 -2.044869e+02  2.352270e+02  
## SEASON_ID2007-08 -2.042382e+02  2.354757e+02  
## SEASON_ID2008-09 -2.040643e+02  2.356496e+02  
## SEASON_ID2009-10 -2.040434e+02  2.356705e+02  
## SEASON_ID2010-11 -2.036505e+02  2.360634e+02  
## SEASON_ID2011-12 -2.044552e+02  2.352587e+02  
## SEASON_ID2012-13 -2.038752e+02  2.358387e+02  
## SEASON_ID2013-14 -2.038763e+02  2.358376e+02  
## SEASON_ID2014-15 -2.039838e+02  2.357301e+02  
## SEASON_ID2015-16 -2.040797e+02  2.356342e+02  
## SEASON_ID2016-17 -2.039696e+02  2.357443e+02  
## SEASON_ID2017-18 -2.040750e+02  2.356388e+02  
## SEASON_ID2018-19 -2.038728e+02  2.358410e+02  
## SEASON_ID2019-20 -2.039352e+02  2.357787e+02  
## SEASON_ID2020-21 -2.039471e+02  2.357668e+02  
## SEASON_ID2021-22 -2.038553e+02  2.358586e+02  
## SEASON_ID2022-23 -2.039442e+02  2.357697e+02  
## SEASON_ID2023-24 -2.038826e+02  2.358313e+02
```

A coefficient is statistically significant if its confidence interval doesn't include 0, and if it does then it is not statistically significant.

This happens because including 0 means that there is a plausible chance that the true effect is zero i.e., no effect.

List of Significant Parameters:

```
has_zero <- ci[, 1] <0 & ci[, 2] >0  
significant_params <- rownames(ci)[!has_zero]  
significant_params
```

```
## [1] "(Intercept)" "RANK"      "TEAM_ID"    "FG3M"      "FG3A"  
## [6] "DREB"        NA          "BLK"        NA           "EFF"  
## [11] "AST_TOV"
```

List of Insignificant Parameters:

```
insignificant_params <- rownames(ci)[has_zero]
insignificant_params
```

```
## [1] "GP" "MIN" "FGM" "FG_PCT"
## [5] "FG3_PCT" "FTM" "FTA" "FT_PCT"
## [9] "OREB" NA "AST" "STL"
## [13] "TOV" "PF" NA "STL_TOV"
## [17] "SEASON_ID1996-97" "SEASON_ID1997-98" "SEASON_ID1998-99" "SEASON_ID1999-00"
## [21] "SEASON_ID2000-01" "SEASON_ID2001-02" "SEASON_ID2002-03" "SEASON_ID2003-04"
## [25] "SEASON_ID2004-05" "SEASON_ID2005-06" "SEASON_ID2006-07" "SEASON_ID2007-08"
## [29] "SEASON_ID2008-09" "SEASON_ID2009-10" "SEASON_ID2010-11" "SEASON_ID2011-12"
## [33] "SEASON_ID2012-13" "SEASON_ID2013-14" "SEASON_ID2014-15" "SEASON_ID2015-16"
## [37] "SEASON_ID2016-17" "SEASON_ID2017-18" "SEASON_ID2018-19" "SEASON_ID2019-20"
## [41] "SEASON_ID2020-21" "SEASON_ID2021-22" "SEASON_ID2022-23" "SEASON_ID2023-24"
```