

Predicting Average

Salary From NBA Stats

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NBA CONTRACTS HISTORY



Data Set

B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z	AA	AB
nba_contracts_history																										
CONTRACT_START	CONTRACT_END	Avg_Salary	Age	GP	W	L	Min	PTS	Fgm	Fga	Fg%	3Pm	3P%	FTm	FTA	FT%	OREB	DREB	REB	AST	TOV	STL	BLK	PF	+/-	
2019	2020	2564753.0	32.0	69.0	27.0	42.0	2091.0	840.0	279.0	698.0	40.0	150.0	403.0	37.2	132.0	163.0	81.0	32.0	138.0	170.0	160.0	91.0	54.0	17.0	160.0	-90.0
2015	2017	21165675.0	27.0	72.0	34.0	38.0	2100.0	1236.0	506.0	987.0	51.3	1.0	10.0	10.0	223.0	274.0	81.4	214.0	321.0	535.0	50.0	104.0	43.0	126.0	206.0	-104.0
2011	2014	10759763.5	22.0	80.0	31.0	49.0	2047.0	566.0	234.0	341.0	68.6	0.0	1.0	0.0	98.0	217.0	45.2	207.0	368.0	575.0	42.0	101.0	41.0	142.0	255.0	-100.0
2015	2018	8143323.5	25.0	82.0	39.0	43.0	2581.0	1258.0	512.0	1100.0	46.5	57.0	179.0	31.8	177.0	232.0	76.3	110.0	396.0	506.0	192.0	170.0	101.0	39.0	245.0	105.0
2018	2019	13410739.0	32.0	81.0	35.0	46.0	2463.0	1347.0	506.0	911.0	55.5	1.0	7.0	14.3	334.0	582.0	57.4	255.0	757.0	1012.0	105.0	208.0	48.0	131.0	249.0	89.0
2015	2016	7055398.0	22.0	76.0	46.0	30.0	1563.0	530.0	203.0	496.0	40.9	39.0	131.0	29.8	85.0	126.0	67.5	27.0	125.0	152.0	155.0	65.0	48.0	14.0	146.0	-128.0
2016	2017	6135000.0	28.0	76.0	18.0	58.0	1615.0	586.0	218.0	561.0	38.9	96.0	268.0	35.8	54.0	63.0	85.7	20.0	154.0	174.0	83.0	45.0	48.0	5.0	82.0	-322.0
2019	2020	4767000.0	29.0	65.0	31.0	34.0	1371.0	611.0	230.0	476.0	48.3	71.0	176.0	40.3	80.0	101.0	79.2	104.0	305.0	409.0	50.0	87.0	45.0	34.0	193.0	-214.0
2015	2018	4051312.5	25.0	51.0	15.0	36.0	824.0	294.0	119.0	242.0	49.2	12.0	43.0	27.9	44.0	57.0	77.2	46.0	153.0	199.0	59.0	55.0	31.0	39.0	110.0	-123.0
2014	2017	8000000.0	28.0	77.0	41.0	36.0	2723.0	1107.0	389.0	853.0	45.6	180.0	442.0	40.7	149.0	193.0	77.2	99.0	376.0	475.0	191.0	132.0	126.0	20.0	179.0	196.0
2013	2016	11000000.0	25.0	78.0	45.0	33.0	2983.0	1786.0	626.0	1388.0	45.1	272.0	600.0	45.3	262.0	291.0	90.0	59.0	255.0	314.0	539.0	240.0	126.0	12.0	198.0	133.0
2017	2019	7000000.0	25.0	74.0	29.0	45.0	1494.0	585.0	207.0	545.0	38.0	120.0	308.0	39.0	51.0	64.0	79.7	27.0	129.0	156.0	93.0	46.0	46.0	6.0	96.0	-60.0
2016	2019	12500000.0	25.0	59.0	31.0	28.0	866.0	246.0	93.0	208.0	44.7	24.0	74.0	32.4	36.0	42.0	85.7	39.0	127.0	166.0	57.0	33.0	37.0	11.0	68.0	46.0
2015	2017	1224935.0	21.0	35.0	2.0	33.0	261.0	73.0	26.0	67.0	38.8	5.0	19.0	26.3	16.0	24.0	66.7	15.0	24.0	39.0	11.0	11.0	6.0	1.0	15.0	27.0
2017	2019	5708251.0	36.0	67.0	33.0	34.0	1753.0	678.0	239.0	515.0	46.4	162.0	359.0	45.1	38.0	42.0	90.5	7.0	179.0	186.0	110.0	69.0	34.0	21.0	107.0	-169.0
2011	2015	4250000.0	25.0	82.0	40.0	42.0	2140.0	867.0	300.0	629.0	47.7	105.0	253.0	41.5	162.0	218.0	74.3	111.0	207.0	318.0	108.0	77.0	87.0	20.0	139.0	-31.0
2018	2019	1682025.0	25.0	82.0	49.0	33.0	1488.0	441.0	165.0	390.0	42.3	74.0	210.0	35.2	37.0	44.0	84.1	36.0	125.0	161.0	91.0	43.0	23.0	22.0	120.0	141.0
2017	2019	31633333.33333300	31.0	60.0	36.0	24.0	2244.0	1344.0	426.0	918.0	46.4	193.0	468.0	41.2	299.0	365.0	81.9	48.0	238.0	286.0	417.0	173.0	88.0	19.0	170.0	356.0
2015	2018	10375000.0	23.0	27.0	9.0	18.0	899.0	374.0	121.0	300.0	40.3	26.0	68.0	38.2	106.0	129.0	82.2	19.0	95.0	114.0	82.0	52.0	17.0	5.0	64.0	-129.0
2017	2018	5250000.0	33.0	62.0	32.0	30.0	1596.0	444.0	174.0	395.0	44.1	41.0	120.0	34.2	55.0	75.0	73.3	54.0	216.0	270.0	107.0	58.0	96.0	31.0	97.0	-65.0
2015	2019	17000450.0	29.0	78.0	41.0	37.0	2640.0	1275.0	502.0	1002.0	50.1	90.0	259.0	34.7	181.0	234.0	77.4	81.0	193.0	274.0	350.0	173.0	78.0	15.0	195.0	20.0
2016	2020	23217391.5	27.0	70.0	41.0	29.0	2448.0	1046.0	372.0	873.0	42.6	139.0	399.0	34.8	163.0	192.0	84.9	55.0	373.0	428.0	403.0	205.0	65.0	43.0	114.0	211.0
2019	2020	2174310.0	26.0	76.0	45.0	31.0	2028.0	618.0	232.0	572.0	40.6	104.0	327.0	31.8	50.0	95.0	52.6	25.0	137.0	162.0	167.0	68.0	47.0	23.0	207.0	104.0

Inflation Index

Purpose

Since the salaries were from a range of years, we decided to adjust for inflation and standardize the values.

Process

From a table of CPI-U values, multiply the base year CPI to the nominal salary, divide by the nominal salary year's CPI, and multiply by 100% to get the adjusted salary.

Inflation Code

```
# Function to adjust salary for inflation
adjust_for_inflation <- function(year, salary) {
  # Extract CPI values for the year
  yearly_cpi <- cpi_data$CPI[format(cpi_data$date, "%Y") == as.character(year)]

  # If CPI for that year is not found, return NA
  if (length(yearly_cpi) == 0) {
    return(NA)
  }

  # Use the average CPI for the year
  avg_cpi_year <- mean(yearly_cpi)

  # Calculate the adjusted salary
  adjusted_salary <- (salary * cpi_2023) / avg_cpi_year
  return(adjusted_salary)
}

# Apply the function to adjust each salary
nba$AdjustedSalary <- mapply(adjust_for_inflation, nba$CONTRACT_END, nba$SALARY, SIMPLIFY = FALSE)

# Convert the list to a vector
nba$AdjustedSalary <- unlist(nba$AdjustedSalary)

# Check the output
print(nba$AdjustedSalary)
```



Stepwise Regression

```
> # Summary of the selected model  
> summary(stepwise_model)
```

Call:
lm(formula = target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +
 FTA + DREB + PF, data = data.frame(target, predictors))

Residuals:

Min	1Q	Median	3Q	Max
-12702205	-3410513	-22293	3096859	10520624

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	8666745	1608968	5.387	3.32e-07 ***
GP	-253372	48355	-5.240	6.43e-07 ***
W	168299	45652	3.687	0.000334 ***
MIN	3373	2040	1.653	0.100691
FGM	62338	21114	2.952	0.003752 **
FGA	-22228	10309	-2.156	0.032941 *
ThreePM	-84708	57142	-1.482	0.140693
ThreePA	48161	23892	2.016	0.045914 *
FTA	13205	4684	2.819	0.005579 **
DREB	16891	5824	2.900	0.004393 **
PF	-35172	13849	-2.540	0.012295 *

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 4892000 on 128 degrees of freedom
Multiple R-squared: 0.762, Adjusted R-squared: 0.7434
F-statistic: 40.98 on 10 and 128 DF, p-value: < 2.2e-16



Hypothetical Models

1st Order (Reduced) Model

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \dots + \beta_{10} X_{10} + \varepsilon$$

2nd Order (Complete) Model

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_{10} X_{10} + \beta_{11} X_1^2 + \beta_{12} X_2^2 + \dots + \beta_{20} X_{10}^2 + \text{Interaction Terms} + \epsilon$$

NESTED MODELS

Reduced

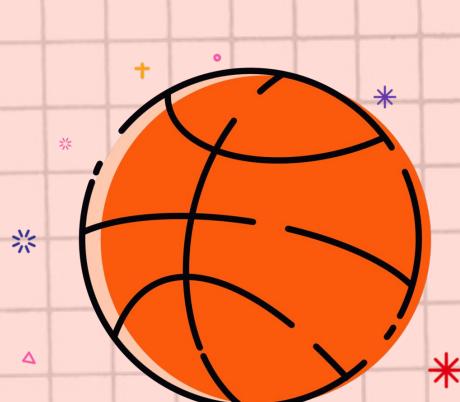
```
> first_order_model
```

Call:

```
lm(formula = target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +
  FTA + DREB + PF, data = data.frame(target, predictors))
```

Coefficients:

	GP	W	MIN	FGM	FGA	ThreePM
8666745	-253372	168299	3373	62338	-22229	-84708
ThreePA	FTA	DREB	PF			
48161	13205	16891	-35172			



```
> second_order_model
```

Call:

```
lm(formula = target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +
  FTA + DREB + PF + I(GP^2) + I(W^2) + I(MIN^2) + I(FGM^2) +
  I(FTA^2) + I(ThreePM^2) + I(ThreePA^2) + I(DREB^2) +
  I(PF^2) + GP * W + GP * MIN + GP * FGM + GP * FGA + GP *
  ThreePM + GP * ThreePA + GP * FTA + GP * DREB + GP * PF +
  W * MIN + W * FGM + W * FGA + W * ThreePM + W * ThreePA +
  W * FTA + W * DREB + W * PF + MIN * FGM + MIN * FGA + MIN *
  ThreePM + MIN * ThreePA + MIN * FTA + MIN * DREB + MIN *
  PF + FGM * FGA + FGM * ThreePM + FGM * ThreePA + FGA +
  FGM * DREB + FGM * PF + FGA * ThreePM + FGA * ThreePA + FGA *
  FTA + FGA * DREB + FGA * PF + ThreePM * ThreePA + ThreePM *
  FTA + ThreePM * DREB + ThreePM * PF + ThreePA * FTA + ThreePA *
  DREB + ThreePA * PF + FTA * DREB + FTA * PF + DREB * PF,
  data = data.frame(target, predictors))
```

Coefficients:

(Intercept)	GP	W	MIN	FGM	FGA
5.875e+06	-4.350e+05	2.744e+05	1.174e+04	4.606e+05	-1.300e+05
ThreePM	ThreePA	FTA	DREB	PF	I(GP^2)
-3.624e+05	1.176e+05	-1.496e+03	-3.842e+04	-1.102e+04	4.696e+03
I(W^2)	I(MIN^2)	I(FGM^2)	I(FTA^2)	I(ThreePM^2)	I(ThreePA^2)
5.718e+03	2.080e+00	1.256e+03	1.370e+02	-1.795e+04	-2.578e+03
I(DREB^2)	I(PF^2)	GP:W	GP:MIN	GP:FGM	
2.163e+01	-1.218e+02	2.162e+01	-1.282e+04	-2.349e+02	-8.802e+03
GP:FGA	FGM:FGA	GP:ThreePM	GP:ThreePA	GP:DREB	
3.076e+03	-8.562e+02	9.686e+03	-3.418e+03	8.380e+02	2.525e+03
GP:PF	W:MIN	W:FGM	W:DREB	W:ThreePM	W:ThreePA
1.209e+03	2.456e+02	-3.518e+03	1.418e+03	2.905e+03	-2.055e+03
W:FTA	W:DREB	W:PF	MIN:FGM	MIN:DREB	MIN:ThreePM
-9.012e+01	-9.960e+02	3.253e+02	8.052e+01	-4.990e+01	8.116e+01
MIN:ThreePA	MIN:FTA	MIN:DREB	MIN:PF	FGM:ThreePM	FGM:ThreePA
-4.703e+01	1.084e+01	2.642e+01	-8.220e+01	4.433e+03	-1.327e+03
FGM:FTA	FGM:DREB	FGM:PF	FGA:ThreePM	FGA:ThreePA	FGA:FTA
-8.825e+02	7.842e+00	2.655e+02	-2.124e+03	6.948e+02	2.820e+02
FGA:DREB	FGA:PF	ThreePM:ThreePA	ThreePM:FTA	ThreePM:DREB	ThreePM:PF
-8.906e+01	-9.760e+01	1.370e+04	-3.029e+02	-1.441e+03	-1.630e+03
ThreePA:FTA	ThreePA:DREB	ThreePA:PF	FTA:DREB	FTA:PF	DREB:PF
-2.948e+01	6.221e+02	9.111e+02	1.795e+02	1.418e+02	-3.686e+02
GP:FGM:FGA					
1.054e+00					

Complete Model Summary Statistics

```
> summary(second_order_model)
```

Call:

```
lm(formula = target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +
    FTA + DREB + PF + I(GP^2) + I(W^2) + I(MIN^2) + I(FGM^2) +
    I(FGA^2) + I(ThreePM^2) + I(ThreePA^2) + I(FTA^2) + I(DREB^2) +
    I(PF^2) + GP * W + GP * MIN + GP * FGM * GP * FGA + GP *
    ThreePM + GP * ThreePA + GP * FTA + GP * DREB + GP * PF +
    W * MIN + W * FGM + W * FGA + W * ThreePM + W * ThreePA +
    W * FTA + W * DREB + W * PF + MIN * FGM + MIN * FGA + MIN *
    ThreePM + MIN * ThreePA + MIN * FTA + MIN * DREB + MIN *
    PF + FGM * FGA + FGM * ThreePM + FGM * ThreePA + FGM * FTA +
    FGM * DREB + FGM * PF + FGA * ThreePM + FGA * ThreePA + FGA *
    FTA + FGA * DREB + FGA * PF + ThreePM * ThreePA + ThreePM *
    FTA + ThreePM * DREB + ThreePM * PF + ThreePA * FTA + ThreePA *
    DREB + ThreePA * PF + FTA * DREB + FTA * PF + DREB * PF,
    data = data.frame(target, predictors))
```

Residuals:

Min	1Q	Median	3Q	Max
-9149718	-1947195	45188	1993341	11768921



Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	5.875e+06	2.823e+06	2.081	0.0410 *
GP	-4.359e+05	2.877e+05	-1.515	0.1341
W	2.744e+05	3.188e+05	0.861	0.3922
MIN	1.174e+04	2.488e+04	0.472	0.6384
FGM	4.606e+05	2.675e+05	1.722	0.0894 .
FGA	-1.300e+05	1.255e+05	-1.036	0.3035
ThreePM	-3.624e+05	6.107e+05	-0.593	0.5548
ThreePA	1.176e+05	2.659e+05	0.442	0.6596
FTA	-1.496e+03	6.473e+04	-0.023	0.9816
DREB	-3.842e+04	6.473e+04	-0.594	0.5547
PF	-1.102e+05	1.369e+05	-0.805	0.4234
I(GP^2)	4.699e+03	5.892e+03	0.798	0.4278
I(W^2)	5.718e+03	5.109e+03	1.119	0.2668
I(MIN^2)	2.080e+00	1.165e+01	0.179	0.8588
I(FGM^2)	1.256e+03	1.021e+03	1.230	0.2226
I(FGA^2)	1.370e+02	2.635e+02	0.520	0.6047
I(ThreePM^2)	-1.795e+04	8.817e+03	-2.036	0.0455 *
I(ThreePA^2)	-2.578e+03	1.536e+03	-1.678	0.0977 .
I(FTA^2)	2.163e+01	4.409e+01	0.491	0.6252
I(DREB^2)	-1.218e+02	6.776e+01	-1.797	0.0765 .
I(PF^2)	2.162e+01	5.201e+02	0.042	0.9670

GP:FGM:FGA 1.054e+00 7.503e-01 1.404 0.1646

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4703000 on 72 degrees of freedom

Multiple R-squared: 0.8763, Adjusted R-squared: 0.7628

F-statistic: 7.725 on 66 and 72 DF, p-value: 8.996e-16

>

GP:#	-1.282e+04	8.561e+03	-1.497	0.1387
GP:MIN	-2.349e+02	4.978e+02	-0.472	0.6385
GP:FGM	-8.802e+03	4.663e+03	-1.887	0.0631 .
GP:FGA	3.076e+03	2.425e+03	1.268	0.2988
FGM:FGA	-8.562e+02	1.018e+03	-0.841	0.4933
GP:ThreePM	9.686e+03	1.241e+04	0.781	0.4375
GP:ThreePA	-3.418e+03	5.036e+03	-0.679	0.4995
GP:FTA	8.380e+02	1.407e+03	0.596	0.5533
GP:DREB	2.525e+03	1.491e+03	1.802	0.0757 .
GP:PF	1.209e+03	2.964e+03	0.498	0.6845
W:MIN	2.456e+02	3.288e+02	0.747	0.4576
W:FGM	-3.518e+03	3.982e+03	-0.883	0.3799
W:FGA	1.418e+03	1.934e+03	0.733	0.4659
W:ThreePM	2.905e+03	9.637e+03	0.301	0.7640
W:ThreePA	-2.055e+03	3.948e+03	-0.520	0.6044
W:FTA	-9.812e+01	9.533e+02	-0.095	0.9249
W:DREB	-9.960e+02	1.047e+03	-0.951	0.3448
W:PF	3.253e+03	2.120e+03	1.535	0.1293
MIN:FGM	8.052e+01	1.769e+02	0.455	0.6504
MIN:FGA	-4.990e+01	9.345e+01	-0.534	0.5950
MIN:ThreePM	8.116e+01	3.839e+02	0.211	0.8332
MIN:ThreePA	-4.703e+01	1.666e+02	-0.282	0.7785
MIN:FTA	1.084e+01	3.796e+01	0.286	0.7760
MIN:DREB	2.642e+01	5.006e+01	0.528	0.5994
MIN:PF	-8.220e+01	1.077e+02	-0.764	0.4477
FGM:ThreePM	4.433e+03	3.951e+03	1.122	0.2656
FGM:ThreePA	-1.327e+03	1.741e+03	-0.762	0.4483
FGM:FTA	-8.825e+02	5.305e+02	-1.664	0.1005
FGM:DREB	7.841e+00	3.314e+02	0.024	0.9812
FGM:PF	2.655e+02	1.296e+03	0.205	0.8383
FGA:ThreePM	-2.124e+03	2.038e+03	-1.042	0.3010
FGA:ThreePA	6.948e+02	9.069e+02	0.766	0.4461
FGA:FTA	2.820e+02	2.457e+02	1.148	0.2549
FGA:DREB	-8.906e+01	1.727e+02	-0.516	0.6077
FGA:PF	-9.760e+01	6.026e+02	-0.162	0.8718
ThreePM:ThreePA	1.370e+02	7.278e+03	1.882	0.0639 .
ThreePM:FTA	-3.029e+02	1.447e+03	-0.209	0.8348
ThreePM:DREB	-1.441e+03	1.299e+03	-1.110	0.2707
ThreePM:PF	-1.630e+03	2.814e+03	-0.579	0.5641
ThreePA:FTA	-2.948e+01	5.599e+02	-0.053	0.9582
ThreePA:DREB	6.221e+02	5.235e+02	1.188	0.2386
ThreePA:PF	9.111e+02	1.160e+03	0.786	0.4346
FTA:DREB	1.795e+02	1.367e+02	1.313	0.1934
FTA:PF	1.418e+02	2.084e+02	0.680	0.4984
DREB:PF	-3.686e+02	3.592e+02	-1.026	0.3082

ANOVA TEST

TAKEAWAYS

The p-value = 0.2447. Therefore, we decided to go with the more parsimonious model and chose the REDUCED MODEL.

```
> anova(first_order_model, second_order_model)
```

Analysis of Variance Table

Model 1: target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA + FTA + DREB + PF

Model 2: target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA + FTA + DREB + PF + I(GP^2) + I(W^2) + I(MIN^2) + I(FGM^2) + I(FGA^2) + I(ThreePM^2) + I(ThreePA^2) + I(FTA^2) + I(DREB^2) + I(PF^2) + GP * W + GP * MIN + GP * FGM * GP * FGA + GP * ThreePM + GP * ThreePA + GP * FTA + GP * DREB + GP * PF + W * MIN + W * FGM + W * FGA + W * ThreePM + W * ThreePA + W * FTA + W * DREB + W * PF + MIN * FGM + MIN * FGA + MIN * ThreePM + MIN * ThreePA + MIN * FTA + MIN * DREB + MIN * PF + FGM * FGA + FGM * ThreePM + FGM * ThreePA + FGM * FTA + FGM * DREB + FGM * PF + FGA * ThreePM + FGA * ThreePA + FGA * FTA + FGA * DREB + FGA * PF + ThreePM * ThreePA + ThreePM * FTA + ThreePM * DREB + ThreePM * PF + ThreePA * FTA + ThreePA * DREB + ThreePA * PF + FTA * DREB + FTA * PF + DREB * PF

Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
--------	-----	----	-----------	---	--------

1	128	3.0626e+15			
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2	72	1.5923e+15	56	1.4704e+15	1.1873 0.2447
---	----	------------	----	------------	---------------

Reduced Model Summary Statistics

```
> first_order_model <- lm(target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA + FTA + DREB + PF , data = data.frame(target, predictors))  
> summary(first_order_model)
```

Call:

```
lm(formula = target ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +  
    FTA + DREB + PF, data = data.frame(target, predictors))
```

Residuals:

	Min	1Q	Median	3Q	Max
-12702205	-3410513	-22293	3096859	10520624	

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	8666745	1608968	5.387	3.32e-07 ***
GP	-253372	48355	-5.240	6.43e-07 ***
W	168299	45652	3.687	0.000334 ***
MIN	3373	2040	1.653	0.100691
FGM	62338	21114	2.952	0.003752 **
FGA	-22228	10309	-2.156	0.032941 *
ThreePM	-84708	57142	-1.482	0.140693
ThreePA	48161	23892	2.016	0.045914 *
FTA	13205	4684	2.819	0.005579 **
DREB	16891	5824	2.900	0.004393 **
PF	-35172	13849	-2.540	0.012295 *

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 4892000 on 128 degrees of freedom

Multiple R-squared: 0.762, Adjusted R-squared: 0.7434

F-statistic: 49.08 on 10 and 128 DF, p-value: < 2.2e-16

$$\begin{aligned} y = & -253372x_1 + 168299x_2 + 3373x_3 + 62338x_4 - 22228x_5 - 84708x_6 + 48161x_7 \\ & + 13205x_8 + 16891x_9 - 35172x_{10}. \end{aligned}$$



Determining Homoscedasticity



```
t1 <- target[1:70] # First 70 elements
p1<-predictors[1:70,]
t2 <- target[71:length(target)] # Elements from 71 to the end
p2<-predictors[71:nrow(predictors),]

first_order_model_1 <- lm(t1 ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA + FTA + DREB + PF , data = data.frame(t1, p1))

first_order_model_2 <- lm(t2 ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA + FTA + DREB + PF , data = data.frame(t2, p2))

summary(first_order_model_1)
summary(first_order_model_2)

var.test(first_order_model_1, first_order_model_2)
```

> summary(first_order_model_1)

Call:

```
lm(formula = t1 ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +
    FTA + DREB + PF, data = data.frame(t1, p1))
```

Residuals:

Min	1Q	Median	3Q	Max
-7667234	-3360339	-364052	2425137	11516259

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	10300497.4	2096925.1	4.912	7.49e-06 ***
GP	-255960.8	63963.3	-4.002	0.000178 ***
W	155530.5	59085.9	2.632	0.010808 *
MIN	173.3	2791.7	0.062	0.950721
FGM	40403.6	26292.7	1.537	0.129716
FGA	-8598.0	14073.9	-0.611	0.543602
ThreePM	-46227.5	71811.4	-0.644	0.522242
ThreePA	28895.0	30299.5	0.954	0.344153
FTA	17064.7	5608.7	3.043	0.003499 **
DREB	25017.9	7831.9	3.194	0.002250 **
PF	-26955.5	16834.8	-1.601	0.114679

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 4609000 on 59 degrees of freedom
Multiple R-squared: 0.8144, Adjusted R-squared: 0.7829
F-statistic: 25.89 on 10 and 59 DF, p-value: < 2.2e-16

> summary(first_order_model_2)

Call:

```
lm(formula = t2 ~ GP + W + MIN + FGM + FGA + ThreePM + ThreePA +
    FTA + DREB + PF, data = data.frame(t2, p2))
```

Residuals:

Min	1Q	Median	3Q	Max
-12025836	-3146869	-1205	2415391	10918239

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	6824008	2532444	2.695	0.00920 **
GP	-233498	79220	-2.947	0.00461 **
W	157004	81783	1.920	0.05981 .
MIN	5696	3266	1.744	0.08644 .
FGM	84465	39746	2.125	0.03785 *
FGA	-34274	17691	-1.937	0.05757 .
ThreePM	-86007	101350	-0.849	0.39958
ThreePA	53591	42207	1.270	0.20926
FTA	11519	8955	1.286	0.20345
DREB	9442	9310	1.014	0.31470
PF	-44220	26792	-1.651	0.10424

Signif. codes: 0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 5234000 on 58 degrees of freedom
Multiple R-squared: 0.7401, Adjusted R-squared: 0.6953
F-statistic: 16.51 on 10 and 58 DF, p-value: 1.381e-13

Variance Test

```
> var.test(first_order_model_1, first_order_model_2)
```

F test to compare two variances

```
data: first_order_model_1 and first_order_model_2  
F = 0.77553, num df = 59, denom df = 58, p-value = 0.3332  
alternative hypothesis: true ratio of variances is not equal to 1  
95 percent confidence interval:  
 0.4619306 1.3005260  
sample estimates:  
ratio of variances  
 0.775531
```

Predicted Values

	Names	Contract_Start	Contract_End	TargetTest	PredictedValues	APE	31	Ricky Rubio	2015	2018	16847316	8606826	48.912775
1	DeAndre Jordan	2011	2014	13983783	7686193	45.034950	32	LeBron James	2015	2016	34569618	33027018	4.462301
2	Mike Conley	2011	2015	10418932	14923507	43.234510	33	Draymond Green	2015	2019	19736585	21135401	7.087424
3	Thaddeus Young	2011	2014	10812412	13330709	23.290798	34	Kent Bazemore	2016	2019	21060381	15115484	28.227870
4	Marc Gasol	2011	2014	18683500	13592977	27.246691	35	Austin Rivers	2016	2018	14695076	8511023	42.082487
5	Ian Mahinmi	2012	2015	5192387	4781268	7.917733	36	Spencer Dinwiddie	2016	2017	1412699	7336946	419.356636
6	Danilo Gallinari	2012	2015	14422107	14950364	3.662829	37	Jeremy Lamb	2016	2018	8576816	10771612	25.589874
7	Stephen Curry	2013	2016	14101166	21750738	54.247791	38	Spencer Dinwiddie	2016	2018	1595724	7336946	359.787798
8	Zaza Pachulia	2013	2015	6750103	6394612	5.266455	39	Kevin Durant	2016	2017	32346241	39041068	20.697388
9	Chris Paul	2013	2017	26947234	21378245	20.666273	40	Bismack Biyombo	2016	2019	20458655	9844122	51.882852
10	James Johnson	2013	2014	1954269	1676129	14.232429	41	Hassan Whiteside	2016	2019	29309893	22857551	22.014210
11	Kyle Korver	2013	2016	7691545	11885260	54.523700	42	Seth Curry	2016	2017	3719377	53535696	43.456684
12	Trevor Ariza	2014	2017	10041499	19173745	90.945049	43	Jared Dudley	2016	2018	12252594	7172904	41.458074
13	Paul George	2014	2018	24667093	28854086	16.974004	44	Eric Gordon	2016	2019	15911656	11040776	30.612027
14	Quincy Pondexter	2014	2017	4393156	7817527	77.947869	45	Kyle Korver	2017	2019	6869596	9990250	45.427027
15	Garrett Temple	2014	2015	1351115	-1689618	225.053615	46	Kyle Lowry	2017	2019	38069145	22293527	41.439380
16	Patrick Patterson	2014	2016	7755642	8507149	9.689811	47	Thabo Sefolosha	2017	2018	6432612	9326248	44.983848
17	Thabo Sefolosha	2014	2016	5127697	11062352	115.737248	48	JaMychal Green	2017	2018	10169653	9620485	5.400065
18	Udonis Haslem	2014	2015	3626194	5679599	56.626982	49	Quinn Cook	2017	2018	1062780	7885551	641.974312
19	DeMarcus Cousins	2014	2017	19533973	22559785	15.489998	50	Paul Millsap	2017	2019	36243912	22205538	38.733053
20	Derrick Favors	2014	2017	15062248	12404715	17.643669	51	Jeff Teague	2017	2019	22865556	17980755	21.363142
21	Brook Lopez	2015	2017	26566887	15024077	43.448108	52	Bojan Bogdanovic	2017	2018	12865223	12504884	2.800882
22	Alec Burks	2015	2018	12712066	9376533	26.239107	53	Quinn Cook	2017	2018	1062780	7885551	641.974312
23	Kawhi Leonard	2015	2019	26179919	22087094	15.633453	54	Patrick McCaw	2018	2019	2987026	6550429	119.295995
24	Tobias Harris	2015	2018	19604150	17945336	8.461542	55	Jahlil Okafor	2018	2019	1967336	5451630	177.107165
25	Robin Lopez	2015	2018	16545749	11390974	31.154682	56	Nerlens Noel	2018	2019	2190458	5131594	134.270421
26	Jae Crowder	2015	2019	8424152	9142763	8.530365	57	Edmond Sumner	2018	2019	1474103	8441947	472.683751
27	Tyson Chandler	2015	2018	15928372	20846036	30.873614	58	Joe Harris	2018	2019	9627603	9376069	2.612631
28	Ed Davis	2015	2017	8367916	7017060	16.143277	59	Rajon Rondo	2019	2020	3048942	9140298	199.785939
29	Tristan Thompson	2015	2019	19736376	13103284	33.608458	60	Robin Lopez	2019	2020	5666941	6387387	12.713132
30	Khris Middleton	2015	2019	21085280	13743450	34.819691							

Paired Difference Test

```
> t.test(target_test, predicted_values, paired = TRUE, alternative="two.side")
```

Paired t-test Null Hypothesis: Mean Difference = 0

data: target_test and predicted_values

t = 1.297, df = 59, p-value = 0.1997

alternative hypothesis: true mean difference is not equal to 0

95 percent confidence interval:

-641209.1 3004080.0

sample estimates:

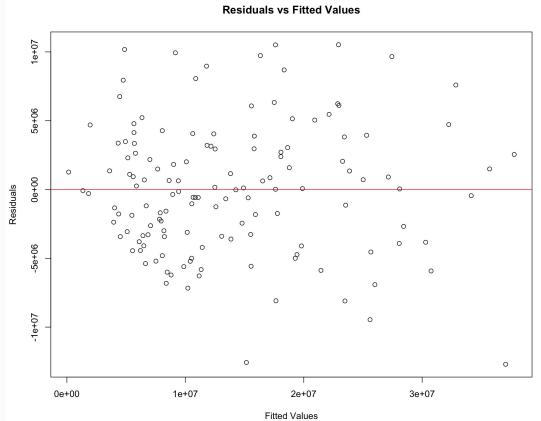
mean difference

1181435

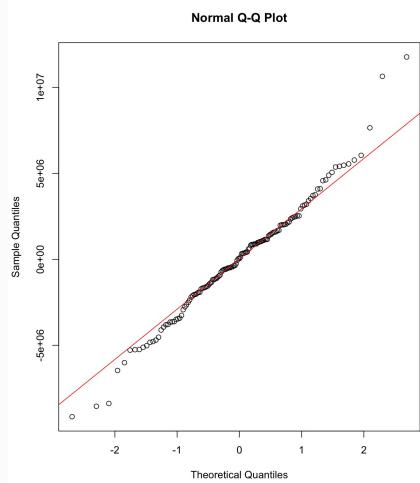
p-value = 0.1997 > 0.05, therefore we cannot reject
the null and we can conclude that there is no
significant mean difference between the values



Residuals vs. Fitted Values Plot



Q-Q Plot



Cook's Distance

```
> cooks.distance(first_order_model)
      159      179      14     195      170      50     118      43     197      193
1.868320e-03 3.669328e-04 2.561298e-05 4.249411e-02 7.589519e-03 2.675489e-04 5.608942e-03 1.099776e-03 1.200256e-04 7.808866e-05
      153       90       91     187      185      92     137      99      72      26
8.575132e-03 5.883973e-04 1.045973e-03 2.350675e-03 5.515502e-04 3.262444e-03 9.631915e-04 2.060704e-03 1.657917e-02 1.359437e-02
       7     196     183     164      78      81     192     103     117      76
4.277578e-03 1.967439e-02 5.642879e-05 5.852480e-03 7.845100e-03 1.068339e-03 1.366539e-04 2.035266e-05 2.431557e-02 1.443466e-02
     143      32     109     198     177      74      23     155      53     135
5.325977e-03 1.842739e-02 1.694131e-03 4.478592e-02 3.004701e-02 8.527088e-04 2.157899e-02 1.332008e-04 6.002085e-03 9.716409e-03
     161      162      34      69     181     178      63     141      97     186
6.235383e-03 2.808030e-03 2.824190e-03 2.948240e-07 6.892402e-04 1.523627e-04 3.386747e-05 8.685631e-04 6.834917e-03 2.497188e-07
       38      21      41     188      60      16     116      94       6      86
3.312861e-04 3.437004e-03 9.027465e-04 4.860320e-03 2.250259e-03 9.046606e-03 1.105943e-03 1.877718e-06 2.166347e-03 1.456605e-04
     140      39     190     194     157       4      13     156     127      52
2.562961e-02 1.016807e-02 5.578498e-03 4.083219e-03 3.956273e-02 8.615835e-03 1.264084e-02 9.147899e-03 1.921873e-03 5.873497e-02
     22      89     168     110      25      87      35      40     112      30
1.435179e-02 3.674920e-03 4.016948e-05 2.560506e-02 3.691003e-03 2.516495e-06 3.023460e-03 8.577337e-05 3.627973e-03 4.524658e-05
      12      31     126     120      64     182     191      93      96      71
2.654897e-03 1.861877e-03 3.611353e-02 1.579717e-04 8.217710e-06 1.009803e-02 2.097235e-03 1.516433e-03 2.678629e-03 5.696928e-04
       67     163      79      85      37       8      51     176     136      98
1.172575e-03 1.710566e-02 8.651326e-03 7.058579e-05 1.210946e-06 1.706134e-03 9.733462e-03 7.404100e-02 3.044846e-03 3.305972e-02
     102     139     154      84      46      17      62      95      54     123
1.195307e-02 3.389185e-02 2.322475e-03 1.695987e-03 2.844726e-04 1.687517e-03 1.947909e-02 7.430167e-03 8.032072e-04 3.201655e-04
     107      24     124     166     128     108     180     131     100      27
4.744336e-03 1.791954e-03 4.451680e-05 1.050819e-03 3.738842e-03 1.393323e-04 1.261507e-02 9.566555e-04 4.365091e-02 4.281317e-03
       42       5      70     144      88      82     148      55      36     138
6.569628e-04 6.226283e-02 1.494948e-03 1.478967e-02 2.858400e-03 6.264795e-04 4.957033e-03 5.372092e-03 2.365261e-02 4.111030e-03
     150       9      48      56      77     134       1     122     169
2.208728e-02 2.620138e-04 1.989401e-02 1.987042e-02 4.961994e-06 1.873799e-06 1.219195e-02 1.875847e-03 9.651578e-03
```

