

# Performance\_popularity\_salary

2023-01-13

```
library(readr)
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
library(magrittr)
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 4.1.3
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
library(leaps)
library(MASS)
```

```
##
```

```
## Attaching package: 'MASS'
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      select
```

```
library(car)
```

```
## Loading required package: carData
```

```
## Warning: package 'carData' was built under R version 4.1.3
```

```
##
```

```
## Attaching package: 'car'
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      recode
```

```
library(readr)
read.csv('/R Files/PPS/Performance.csv', header = TRUE)
```

##	Players	MPG	PPG	PER	AllStar	Instafollowers	Instagram
## 1	Marcus Smart	32.3	12.1	13.67	0	768000	1
## 2	Jaylen Brown	33.6	23.6	18.98	1	2500000	1
## 3	Jayson Tatum	35.9	26.9	21.87	3	5200000	1
## 4	Robert Williams III	29.6	10.0	22.10	0	237000	1
## 5	Derrick White	27.4	11.0	13.01	0	158000	1
## 6	Al Horford	29.1	10.2	16.71	5	665000	1
## 7	Trae Young	34.9	28.4	25.48	2	4800000	1
## 8	John Collins	30.8	16.2	18.75	0	209000	1
## 9	Bogdan Bogdanovic	29.3	15.1	15.49	0	620000	1
## 10	De'Andre Hunter	29.8	13.4	10.66	0	94300	1
## 11	Kevin Huerter	29.6	12.1	11.91	0	118000	1
## 12	Clint Capela	27.6	11.1	21.43	0	403000	1
## 13	Kyrie Irving	37.7	27.4	21.49	7	17300000	1
## 14	Kevin Durant	37.2	29.9	25.69	12	13000000	1
## 15	Goran Dragic	25.5	7.3	8.63	1	482000	1
## 16	Joe Harris	30.2	11.3	9.64	0	125000	1
## 17	Seth Curry	29.9	14.9	14.61	0	2100000	1
## 18	Patty Mills	29.0	11.4	10.43	0	588000	1
## 19	Miles Bridges	35.5	20.2	17.97	0	701000	1
## 20	Terry Rozier	33.7	19.3	17.62	0	552000	1
## 21	LaMelo Ball	32.3	20.1	19.76	1	9300000	1
## 22	Gordon Hayward	31.9	15.9	15.11	1	1200000	1
## 23	PJ Washington	27.2	10.3	13.60	0	827000	1
## 24	Kelly Oubre	26.3	15.0	15.01	0	1700000	1
## 25	Demar Derozan	36.1	27.9	23.13	5	3200000	1
## 26	Zach Lavine	34.7	24.4	20.08	2	2500000	1
## 27	Lonzo Ball	34.6	13.0	14.49	0	14900000	1
## 28	Nikola Vucevic	33.1	17.6	18.27	2	0	0
## 29	Alex Caruso	28.0	7.4	11.75	0	1600000	1
## 30	Coby White	27.5	12.7	12.60	0	427000	1
## 31	Darius Garland	35.7	21.7	19.04	1	436000	1
## 32	Evan Mobley	33.8	15.0	16.15	0	218000	1
## 33	Jarrett Allen	32.3	16.1	23.08	1	254000	1
## 34	Lauri Markkanen	30.8	14.8	14.76	0	280000	1
## 35	Caris LeVert	29.8	13.6	12.45	0	253000	1
## 36	Isaac Okoro	29.6	8.8	10.39	0	172000	1
## 37	Luka Doncic	35.4	28.4	25.13	3	7300000	1
## 38	Dorian Finney-Smith	33.1	11.0	12.56	0	94100	1
## 39	Jalen Brunson	31.9	16.3	17.14	0	190000	1
## 40	Tim Hardaway Jr.	29.6	14.2	12.79	0	492000	1
## 41	Spencer Dinwiddie	28.3	15.8	19.03	0	300000	1
## 42	Reggie Bullock	28.0	8.6	9.25	0	191000	1
## 43	Nikola Jokic	33.5	27.1	32.94	4	0	0
## 44	Will Barton	32.1	14.7	13.42	0	0	0
## 45	Aaron Gordon	31.7	15.0	15.38	0	1100000	1
## 46	Monte Morris	29.9	12.6	14.75	0	140000	1
## 47	Jeff Green	24.7	10.3	12.51	0	0	0
## 48	Austin Rivers	22.1	6.0	7.24	0	418000	1
## 49	Saddiq Bey	33.0	16.1	14.08	0	57000	1

## 50	Cade Cunningham	32.6	17.4	13.12	0	576000	1
## 51	Jerami Grant	31.9	19.2	16.12	0	186000	1
## 52	Marvin Bagley	27.2	14.6	17.76	0	1020000	1
## 53	Isaiah Stewart	25.6	8.3	14.05	0	119000	1
## 54	Killian Hayes	25.0	6.9	9.35	0	221000	1
## 55	Steph Curry	34.5	25.5	21.47	8	46300000	1
## 56	Klay Thompson	29.4	20.4	16.81	5	13500000	1
## 57	Jordan Poole	30.0	18.5	16.20	0	1500000	1
## 58	Andrew Wiggins	31.9	17.2	15.04	1	2100000	1
## 59	Draymond Green	28.9	7.5	14.30	4	4000000	1
## 60	Otto Porter	22.1	8.2	15.94	0	0	0
## 61	Jalen Green	31.9	17.3	12.56	0	2200000	1
## 62	Kevin Porter	31.3	15.6	13.56	0	546000	1
## 63	Jae'Sean Tate	26.4	11.8	14.36	0	0	0
## 64	Christian Wood	30.8	17.9	19.18	0	200000	1
## 65	Eric Gordon	29.3	13.4	12.17	0	257000	1
## 66	Dennis Schroder	26.9	10.9	11.49	0	2100000	1
## 67	Malcom Brogdon	33.5	19.1	18.10	0	153000	1
## 68	Buddy Hield*	35.6	18.2	14.49	0	530000	1
## 69	Myles Turner	29.4	12.9	17.45	0	248000	1
## 70	Chris Duarte	28.0	13.1	12.31	0	338000	1
## 71	Jalen Smith	24.7	13.4	20.29	0	49900	1
## 72	Tyrese Haliburton	36.1	17.5	20.49	0	211000	1
## 73	Paul George	34.7	24.3	18.67	7	9700000	1
## 74	Reggie Jackson	31.2	16.8	12.19	0	189000	1
## 75	Marcus Morris Sr.	29.0	15.4	13.74	0	334000	1
## 76	Terance Mann	28.6	10.8	13.40	0	147000	1
## 77	Luke Kennard	27.4	11.9	13.31	0	208000	1
## 78	Eric Bledsoe	25.2	9.9	12.58	0	520000	1
## 79	LeBron James	37.2	30.3	26.25	18	133000000	1
## 80	Anthony Davis	35.1	23.2	23.97	8	7100000	1
## 81	Russell Westbrook	34.3	18.5	15.01	9	20600000	1
## 82	Malik Monk	28.1	13.8	14.49	0	554000	1
## 83	Carmelo Anothony	26.0	13.3	14.80	10	7900000	1
## 84	Talen Horton-Tucker	25.2	10.0	11.07	0	383000	1
## 85	Ja Morant	33.1	27.4	24.53	1	7700000	1
## 86	Desmond Bane	29.8	18.2	17.62	0	161000	1
## 87	Dillon Brooks	27.7	18.4	15.55	0	140000	1
## 88	Jaren Jackson	27.3	16.3	17.09	0	372000	1
## 89	Steven Adams	26.3	6.9	17.60	0	518000	1
## 90	De'Anthony Melton	22.7	10.8	15.23	0	69900	1
## 91	Jimmy Butler	33.9	21.4	23.65	6	7500000	1
## 92	Kyle Lowry	33.9	13.4	15.02	6	1500000	1
## 93	Tyler Herro	32.6	20.7	16.30	0	2300000	1
## 94	Bam Adebayo	32.6	19.1	21.91	1	844000	1
## 95	PJ Tucker	27.9	7.6	11.35	0	1000000	1
## 96	Duncan Robinson	25.9	10.9	10.65	0	294000	1
## 97	Jrue Holiday	32.9	18.3	19.84	1	530000	1
## 98	Giannis Antetokounmpo	32.9	29.9	32.12	6	13500000	1
## 99	Khris Middleton	32.4	20.1	18.19	3	403000	1
## 100	Bobby Portis	28.2	14.6	17.79	0	277000	1
## 101	Grayson Allen	27.3	11.1	12.77	0	383000	1
## 102	Pat Connaughton	26.0	9.9	13.45	0	187000	1
## 103	Anthony Edwards	34.3	21.3	16.55	0	1200000	1

## 104	Karl-Anthony Towns	33.4	24.6	24.21	3	3800000	1
## 105	D'Angelo Russell	32.0	18.1	16.36	1	3600000	1
## 106	Jaden McDaniels	25.8	9.2	10.31	0	77900	1
## 107	Patrick Beverley	25.4	9.2	14.80	0	603000	1
## 108	Jarred Vanderbilt	25.4	6.9	15.35	0	114000	1
## 109	Brandon Ingram	34.0	22.7	18.78	1	2200000	1
## 110	CJ McCollum	33.8	24.3	22.01	0	1300000	1
## 111	Josh Hart	33.5	13.4	15.19	0	663000	1
## 112	Jonas Valanciunas	30.3	17.8	21.30	0	263000	1
## 113	Herbert Jones	29.9	9.5	12.35	0	70900	1
## 114	Devonte' Graham	28.4	11.9	12.10	0	257000	1
## 115	Julius Randle	35.3	20.1	15.80	1	907000	1
## 116	RJ Barrett	34.5	20.0	13.72	0	1100000	1
## 117	Evan Fournier	29.5	14.1	12.34	0	301000	1
## 118	Alec Burks	28.6	11.7	13.72	0	8087	1
## 119	Mitchell Robinson	25.7	8.5	20.78	0	144000	1
## 120	Kemba Walker	25.6	11.6	13.96	4	1300000	1
## 121	Shai Gilgeous-Alexander	34.7	24.5	20.99	0	1100000	1
## 122	Luguentz Dort	32.6	17.2	12.31	0	176000	1
## 123	Josh Giddey	31.5	12.5	13.34	0	409000	1
## 124	Darius Bazley	27.9	10.8	12.62	0	112000	1
## 125	Aaron Wiggins	24.2	8.3	10.29	0	28600	1
## 126	Vit Krejci	23.0	6.2	8.52	0	20400	1
## 127	Cole Anthony	31.7	16.3	13.55	0	725000	1
## 128	Franz Wagner	30.7	15.2	14.75	0	102000	1
## 129	Wendell Carter	29.9	15.0	18.41	0	157000	1
## 130	Gary Harris	28.4	11.1	10.89	0	157000	1
## 131	Jalen Suggs	27.2	11.8	8.66	0	542000	1
## 132	Mo Bamba	25.7	10.6	16.50	0	712000	1
## 133	James Harden	37.7	21.0	21.84	10	11600000	1
## 134	Tyrese Maxey	35.3	17.5	16.27	0	472000	1
## 135	Tobias Harris	34.8	17.2	16.00	0	528000	1
## 136	Joel Embiid	33.8	30.6	31.24	5	6000000	1
## 137	Matisse Thybulle	25.5	5.7	11.22	0	400000	1
## 138	Georges Niang	22.8	9.2	10.70	0	100000	1
## 139	Devin Booker	34.5	26.8	21.38	3	5400000	1
## 140	Mikal Bridges	34.8	14.2	14.49	0	228000	1
## 141	Chris Paul	32.9	14.7	20.86	12	11500000	1
## 142	Deandre Ayton	29.5	17.2	21.99	0	504000	1
## 143	Jae Crowder	28.1	9.4	11.33	0	518000	1
## 144	Cameron Johnson	26.1	12.5	15.26	0	220000	1
## 145	Damian Lillard	36.4	24.0	18.53	6	9600000	1
## 146	Norman Powell	33.3	18.7	15.32	0	321000	1
## 147	Drew Eubanks	29.5	14.5	19.92	0	20100	1
## 148	Anfernee Simons	29.5	17.3	15.34	0	212000	1
## 149	Jusuf Nurkic	28.2	15.0	20.10	0	420000	1
## 150	Justise Winslow	26.8	10.7	14.13	0	544000	1
## 151	DeAaron Fox	35.3	23.2	17.49	0	1200000	1
## 152	Domantas Sabonis	33.6	18.9	21.01	2	261000	1
## 153	Harrison Barnes	33.6	16.4	15.75	0	995000	1
## 154	Davion Mitchell	27.7	11.5	10.84	0	191000	1
## 155	Donte Divincenzo	26.6	10.3	13.02	0	106000	1
## 156	Justin Holiday	25.6	8.3	7.93	0	271000	1
## 157	Dejounte Murray	34.8	21.1	22.32	1	1100000	1

## 158	Keldon Johnson	31.9	17.0	15.26	0	145000	1
## 159	Jakob Poeltl	29.0	13.5	20.85	0	72800	1
## 160	Devin Vassell	27.3	12.3	13.39	0	55900	1
## 161	Josh Richardson	24.4	11.4	14.06	0	273000	1
## 162	Doug McDermott	24.0	11.3	11.47	0	255000	1
## 163	Pascal Siakam	37.9	22.8	20.31	1	677000	1
## 164	Fred VanVleet	37.9	20.3	17.38	1	716000	1
## 165	OG Anunoby	36.0	17.1	14.84	0	214000	1
## 166	Scottie Barnes	35.4	15.3	16.37	0	575000	1
## 167	Gary Trent Jr	35.0	18.3	14.71	0	281000	1
## 168	Precious Achiuwa	23.6	9.1	12.69	0	178000	1
## 169	Donovan Mitchell	33.8	25.9	21.64	3	3500000	1
## 170	Rudy Gobert	32.1	15.6	24.76	3	826000	1
## 171	Royce O'Neale	31.2	7.4	10.31	0	134000	1
## 172	Bojan Bogdanovic	30.9	18.1	15.71	0	119000	1
## 173	Mike Conley	28.6	13.7	17.28	1	570000	1
## 174	Jordan Clarkson	27.1	16.0	14.96	0	1700000	1
## 175	Bradley Beal	36.0	23.2	17.58	3	1300000	1
## 176	Kyle Kuzma	33.4	17.1	15.21	0	4900000	1
## 177	Kentavious Caldwell-Pope	30.2	13.2	12.60	0	462000	1
## 178	Deni Avdija	24.2	8.4	11.62	0	215000	1
## 179	Corey Kispert	23.4	8.2	10.96	0	71400	1
## 180	Kristaps Porzingis	28.2	22.1	27.79	1	1500000	1

##	Twitterfollowers	Tiktok	Tiktokmentions	Salary
## 1	447900	0	9.340e+07	14339285
## 2	627400	1	1.921e+08	26758928
## 3	1000000	0	6.939e+08	28103550
## 4	90700	0	1.420e+07	3661976
## 5	34800	0	4.700e+06	15178571
## 6	372900	0	4.610e+07	27000000
## 7	1000000	0	9.048e+08	8326471
## 8	34700	1	3.170e+07	23000000
## 9	293900	0	7.900e+06	18000000
## 10	14900	0	8.292e+05	7775400
## 11	44700	0	6.084e+05	4253357
## 12	95200	0	7.600e+06	18603448
## 13	4500000	1	1.800e+09	35328700
## 14	20500000	0	1.900e+09	42018900
## 15	0	0	1.576e+05	5771889
## 16	0	0	7.100e+06	17357143
## 17	383700	0	6.520e+07	8207518
## 18	422500	1	3.070e+07	5890000
## 19	208000	0	1.840e+08	5421493
## 20	186800	0	2.640e+07	17905263
## 21	1000000	1	2.500e+09	8231760
## 22	655400	0	3.170e+07	29925000
## 23	104100	0	3.880e+07	4215120
## 24	137800	1	3.782e+08	12000000
## 25	773000	0	1.963e+08	26000000
## 26	599400	0	3.149e+08	19500000
## 27	1300000	0	7.335e+08	18604651
## 28	172700	0	5.000e+06	24000000
## 29	383300	0	1.668e+08	8650651
## 30	68000	0	5.800e+06	5837760

## 31	80200	1	3.800e+07	7040880
## 32	46600	1	1.990e+07	8075160
## 33	75400	0	2.010e+07	20000000
## 34	108100	0	7.600e+06	15690909
## 35	54100	0	4.600e+06	17500000
## 36	36500	1	5.200e+06	6720720
## 37	1700000	1	1.000e+09	10174391
## 38	20900	0	2.700e+06	4000000
## 39	80600	0	3.020e+07	1802057
## 40	138100	0	5.000e+06	21308816
## 41	122600	0	1.090e+07	17142857
## 42	111700	0	2.000e+06	9536000
## 43	0	0	2.738e+08	31579390
## 44	46400	0	1.400e+06	15625000
## 45	208300	1	8.780e+07	16409091
## 46	23500	0	1.000e+06	8449074
## 47	154000	0	7.800e+06	4500000
## 48	288800	0	2.910e+07	2401537
## 49	9713	0	2.000e+06	2824320
## 50	102900	0	7.240e+07	10050120
## 51	53400	0	4.600e+06	20002500
## 52	96500	0	4.400e+06	11312114
## 53	19800	0	2.030e+07	3277080
## 54	16700	0	1.120e+07	5572680
## 55	17000000	1	2.600e+09	45780966
## 56	1800000	0	8.820e+08	37980720
## 57	1860	0	6.351e+08	2161440
## 58	590900	0	1.285e+08	31579390
## 59	2000000	0	6.100e+08	24026712
## 60	2624	0	1.670e+07	2389641
## 61	193800	1	3.556e+08	8992200
## 62	70800	0	2.660e+07	1782621
## 63	18400	0	1.100e+03	1517981
## 64	62800	0	1.900e+07	13666667
## 65	231900	0	5.400e+06	18218818
## 66	0	1	2.660e+07	5890000
## 67	35600	0	4.100e+06	21700000
## 68	148800	0	2.870e+07	22768890
## 69	126900	0	8.900e+06	18000000
## 70	19500	1	1.360e+07	3749400
## 71	13600	0	1.100e+06	4458000
## 72	85300	0	2.470e+07	4023600
## 73	2400000	0	4.812e+08	39344900
## 74	121100	0	2.670e+07	10384500
## 75	158800	0	1.580e+07	15627907
## 76	27300	0	8.700e+06	1782621
## 77	92500	0	3.200e+06	13347727
## 78	233600	0	3.700e+06	18125000
## 79	52300000	0	5.800e+09	41180544
## 80	2200000	0	5.408e+08	35361360
## 81	6900000	0	9.743e+08	44211146
## 82	165800	0	2.540e+07	1789256
## 83	9200000	1	2.157e+08	2641691
## 84	80300	0	1.890e+07	9500000

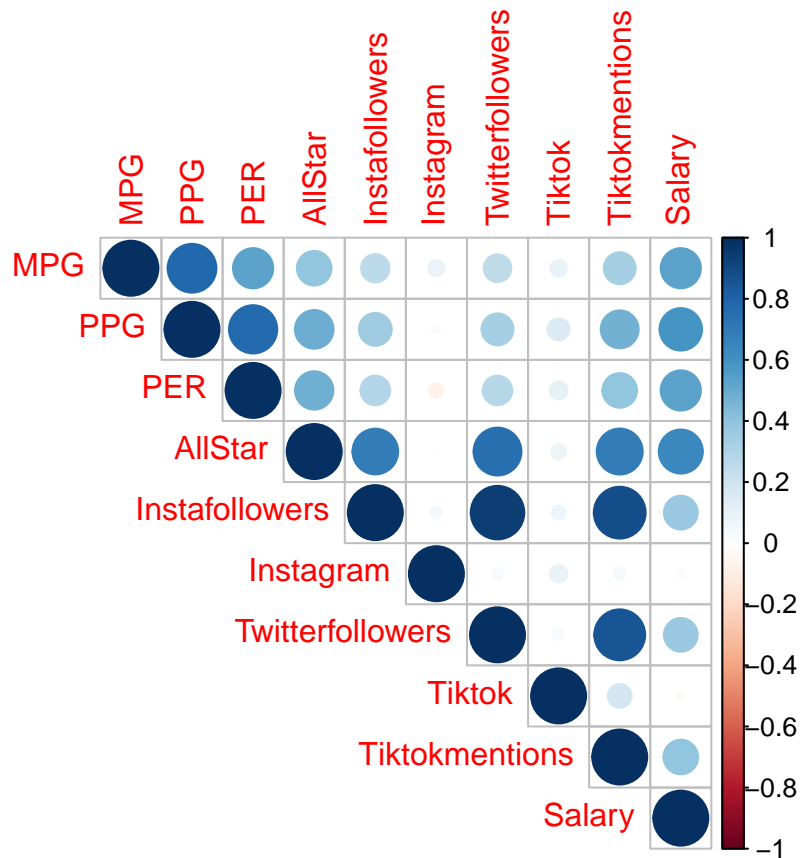
## 85	2400000	1	2.100e+09	9603360
## 86	57900	0	9.500e+06	2033160
## 87	30600	0	1.970e+07	12200000
## 88	218300	0	1.940e+07	9180560
## 89	216800	0	1.379e+08	17073171
## 90	24900	0	7.908e+05	8805976
## 91	871600	1	4.472e+08	36016200
## 92	923100	0	1.027e+08	26894128
## 93	355100	0	8.360e+07	4004280
## 94	226500	0	8.100e+07	28103550
## 95	0	0	5.650e+07	7000000
## 96	49300	0	3.400e+07	15650000
## 97	196000	0	5.080e+07	32413333
## 98	2600000	1	8.511e+08	39344970
## 99	115600	0	2.910e+07	35500000
## 100	172300	0	2.720e+07	4347600
## 101	257100	0	1.870e+07	4054695
## 102	87600	0	2.600e+06	5333334
## 103	183100	1	2.371e+08	10245480
## 104	719200	1	9.850e+07	31650600
## 105	660900	0	6.000e+07	30013500
## 106	12600	0	3.500e+06	2063280
## 107	395100	0	1.102e+08	14320988
## 108	63000	0	8.702e+06	4050000
## 109	340200	0	5.010e+07	29467800
## 110	625100	0	2.860e+07	30864198
## 111	550100	1	7.440e+07	12000000
## 112	176200	0	5.500e+06	14000000
## 113	17400	0	2.469e+05	1700000
## 114	33800	0	9.600e+06	11000000
## 115	473000	0	4.380e+07	21780000
## 116	218700	0	4.680e+07	8623920
## 117	360200	0	1.020e+07	17000000
## 118	53800	0	4.801e+05	9536000
## 119	61700	0	5.700e+06	1802057
## 120	458200	0	1.070e+08	8729020
## 121	131100	1	3.130e+07	5495532
## 122	35600	0	2.900e+06	1782621
## 123	92000	1	9.500e+07	5988000
## 124	23600	0	1.120e+07	2513040
## 125	6801	0	1.914e+05	31579390
## 126	3045	0	1.676e+05	925258
## 127	87000	1	3.010e+07	3449400
## 128	11900	0	5.300e+06	5007840
## 129	53400	1	1.100e+06	6920027
## 130	64400	0	6.300e+06	20482143
## 131	45600	0	4.930e+07	6592920
## 132	43700	0	2.763e+08	7568743
## 133	7700000	1	1.400e+09	44310840
## 134	133200	1	3.230e+07	2602920
## 135	270200	0	1.440e+07	35995950
## 136	2100000	0	3.536e+08	31579390
## 137	97200	1	1.410e+07	2840160
## 138	91700	1	8.919e+05	3300000

## 139	1200000	1	1.100e+09	31650600
## 140	63100	0	1.740e+07	5557725
## 141	8500000	0	4.664e+08	30000000
## 142	162200	1	4.140e+07	12632950
## 143	213800	0	1.480e+07	9720900
## 144	0	0	1.200e+06	4437000
## 145	3100000	0	3.450e+08	39344900
## 146	78200	1	3.400e+06	15517241
## 147	5477	0	1.650e+05	2239820
## 148	38600	0	7.000e+06	3938818
## 149	145500	0	1.440e+07	12000000
## 150	119200	0	2.253e+05	3902439
## 151	291900	0	3.580e+07	28103550
## 152	49300	0	8.000e+06	19800000
## 153	590100	0	4.100e+06	20284091
## 154	22200	1	9.900e+06	4603320
## 155	16700	0	4.100e+06	4675830
## 156	30200	1	1.900e+06	6006420
## 157	225400	0	7.930e+07	15428880
## 158	0	0	1.100e+07	2145720
## 159	1774	0	1.300e+06	8750000
## 160	17000	0	8.646e+05	4235160
## 161	0	0	1.700e+06	11615328
## 162	0	0	8.536e+05	13750000
## 163	158900	0	3.330e+07	33003936
## 164	196700	0	2.870e+07	19675926
## 165	0	0	9.300e+06	16071429
## 166	61000	1	6.840e+07	7280400
## 167	79500	0	7.000e+06	16000000
## 168	32600	0	3.300e+06	2711280
## 169	1300000	0	1.680e+08	28103550
## 170	606100	0	1.349e+08	35344828
## 171	52300	0	8.446e+05	8800000
## 172	26100	0	2.400e+06	18700000
## 173	537700	0	6.400e+06	21000000
## 174	508600	1	3.062e+08	12420000
## 175	687200	0	1.012e+08	33724200
## 176	1200000	0	1.223e+08	13000000
## 177	109600	0	3.000e+06	13038862
## 178	0	1	1.790e+07	4692840
## 179	11900	1	1.100e+06	3383640
## 180	370500	1	2.480e+07	31650600

```
data <- read.csv('/R Files/PPS/Performance.csv')
data1 = subset(data, select = c(2,3,4,5,6,7,8,9,10,11))
```

```
correlation = cor(data1)
corrplot(correlation, type = "upper")
```





```
mod_performance = lm(Salary ~ MPG + PPG + PER + AllStar, data = data1)
summary(mod_performance)
```

```
##
## Call:
## lm(formula = Salary ~ MPG + PPG + PER + AllStar, data = data1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -25675628 -4917316  -480113   5849051  26113040
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -12603112   6483277  -1.944   0.0535 .
## MPG          530580     258594    2.052   0.0417 *
## PPG          327201     227929    1.436   0.1529
## PER          244281     209238    1.167   0.2446
## AllStar      1915821     262416    7.301 9.65e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7997000 on 175 degrees of freedom
## Multiple R-squared:  0.5294, Adjusted R-squared:  0.5187
## F-statistic: 49.22 on 4 and 175 DF, p-value: < 2.2e-16
```

```
vif(mod_performance)
```

```
##      MPG      PPG      PER AllStar  
## 2.718814 4.831639 2.643038 1.377024
```

```
step(mod_performance, direction="both")
```

```
## Start: AIC=5726.99  
## Salary ~ MPG + PPG + PER + AllStar  
##  
##      Df Sum of Sq      RSS      AIC  
## - PER      1 8.7175e+13 1.1280e+16 5726.4  
## <none>                1.1193e+16 5727.0  
## - PPG      1 1.3180e+14 1.1324e+16 5727.1  
## - MPG      1 2.6925e+14 1.1462e+16 5729.3  
## - AllStar  1 3.4090e+15 1.4602e+16 5772.9  
##
```

```
## Step: AIC=5726.39  
## Salary ~ MPG + PPG + AllStar  
##  
##      Df Sum of Sq      RSS      AIC  
## <none>                1.1280e+16 5726.4  
## + PER      1 8.7175e+13 1.1193e+16 5727.0  
## - MPG      1 2.2300e+14 1.1503e+16 5727.9  
## - PPG      1 4.9636e+14 1.1776e+16 5732.1  
## - AllStar  1 3.7704e+15 1.5050e+16 5776.3
```

```
##  
## Call:  
## lm(formula = Salary ~ MPG + PPG + AllStar, data = data1)  
##  
## Coefficients:  
## (Intercept)      MPG      PPG      AllStar  
## -9623320      474443      494254      1975815
```

```
mod_performance1 = lm(Salary ~ MPG + PPG + AllStar, data = data1)  
summary(mod_performance1)
```

```
##  
## Call:  
## lm(formula = Salary ~ MPG + PPG + AllStar, data = data1)  
##  
## Residuals:  
##      Min      1Q   Median      3Q      Max  
## -26402228 -5059677  -240585  5872822 25618885  
##  
## Coefficients:  
##      Estimate Std. Error t value Pr(>|t|)  
## (Intercept) -9623320    5965885  -1.613  0.10852  
## MPG          474443     254346   1.865  0.06380 .  
## PPG          494254     177601   2.783  0.00598 **
```

```
## AllStar      1975815      257601    7.670 1.12e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8006000 on 176 degrees of freedom
## Multiple R-squared:  0.5257, Adjusted R-squared:  0.5177
## F-statistic: 65.04 on 3 and 176 DF,  p-value: < 2.2e-16
```

```
vif(mod_performance1)
```

```
##      MPG      PPG AllStar
## 2.624810 2.927472 1.324219
```

```
mod_popularity = lm(Salary ~ Instafollowers + Instagram + Twitterfollowers + Tiktok + Tiktokmentions, data = data1)
summary(mod_popularity)
```

```
##
## Call:
## lm(formula = Salary ~ Instafollowers + Instagram + Twitterfollowers +
##      Tiktok + Tiktokmentions, data = data1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -22961668 -8508036 -2445819  6326560 22471864
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.292e+07  4.349e+06   2.970  0.0034 **
## Instafollowers  5.902e-03  2.576e-01   0.023  0.9817
## Instagram     9.577e+05  4.447e+06   0.215  0.8297
## Twitterfollowers 2.097e-01  5.574e-01   0.376  0.7073
## Tiktok       -2.625e+06  2.017e+06  -1.301  0.1948
## Tiktokmentions  6.641e-03  3.100e-03   2.142  0.0335 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10650000 on 174 degrees of freedom
## Multiple R-squared:  0.1707, Adjusted R-squared:  0.1468
## F-statistic: 7.162 on 5 and 174 DF,  p-value: 4.052e-06
```

```
vif(mod_popularity)
```

```
## Instafollowers      Instagram Twitterfollowers      Tiktok
##      12.361980      1.011873      9.904102      1.116178
## Tiktokmentions
##      5.274117
```

```
step(mod_popularity, direction = "both")
```

```
## Start: AIC=5830.98
## Salary ~ Instafollowers + Instagram + Twitterfollowers + Tiktok +
```

```

##      Tiktokmentions
##
##           Df  Sum of Sq      RSS      AIC
## - Instafollowers  1 5.9490e+10 1.9725e+16 5829.0
## - Instagram      1 5.2572e+12 1.9730e+16 5829.0
## - Twitterfollowers 1 1.6040e+13 1.9741e+16 5829.1
## - Tiktok         1 1.9201e+14 1.9917e+16 5830.7
## <none>                                1.9725e+16 5831.0
## - Tiktokmentions  1 5.2034e+14 2.0245e+16 5833.7
##
## Step:  AIC=5828.99
## Salary ~ Instagram + Twitterfollowers + Tiktok + Tiktokmentions
##
##           Df  Sum of Sq      RSS      AIC
## - Instagram      1 5.2890e+12 1.9730e+16 5827.0
## - Twitterfollowers 1 4.3311e+13 1.9768e+16 5827.4
## - Tiktok         1 1.9336e+14 1.9918e+16 5828.7
## <none>                                1.9725e+16 5829.0
## + Instafollowers  1 5.9490e+10 1.9725e+16 5831.0
## - Tiktokmentions  1 6.6626e+14 2.0391e+16 5833.0
##
## Step:  AIC=5827.03
## Salary ~ Twitterfollowers + Tiktok + Tiktokmentions
##
##           Df  Sum of Sq      RSS      AIC
## - Twitterfollowers 1 4.3941e+13 1.9774e+16 5825.4
## - Tiktok         1 1.8906e+14 1.9919e+16 5826.8
## <none>                                1.9730e+16 5827.0
## + Instagram      1 5.2890e+12 1.9725e+16 5829.0
## + Instafollowers  1 9.1261e+10 1.9730e+16 5829.0
## - Tiktokmentions  1 6.6592e+14 2.0396e+16 5831.0
##
## Step:  AIC=5825.43
## Salary ~ Tiktok + Tiktokmentions
##
##           Df  Sum of Sq      RSS      AIC
## <none>                                1.9774e+16 5825.4
## - Tiktok         1 2.5045e+14 2.0024e+16 5825.7
## + Twitterfollowers 1 4.3941e+13 1.9730e+16 5827.0
## + Instafollowers  1 2.8114e+13 1.9746e+16 5827.2
## + Instagram      1 5.9185e+12 1.9768e+16 5827.4
## - Tiktokmentions  1 3.9939e+15 2.3768e+16 5856.5
##
##
## Call:
## lm(formula = Salary ~ Tiktok + Tiktokmentions, data = data1)
##
## Coefficients:
##      (Intercept)          Tiktok  Tiktokmentions
##      1.383e+07      -2.889e+06       8.156e-03

mod_popularity1 = lm(Salary ~ Tiktok + Tiktokmentions, data = data1)
summary(mod_popularity1)

```

```
##
## Call:
## lm(formula = Salary ~ Tiktok + Tiktokmentions, data = data1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -23096348 -8534542 -2386863  6469970 22705206
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.383e+07  9.138e+05  15.129 < 2e-16 ***
## Tiktok       -2.889e+06  1.929e+06  -1.497  0.136
## Tiktokmentions 8.156e-03  1.364e-03   5.979 1.21e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10570000 on 177 degrees of freedom
## Multiple R-squared:  0.1686, Adjusted R-squared:  0.1592
## F-statistic: 17.95 on 2 and 177 DF, p-value: 8e-08
```

```
vif(mod_popularity1)
```

```
##           Tiktok Tiktokmentions
##           1.036635           1.036635
```

```
mod_combined = lm(Salary ~ MPG + PPG + AllStar + Instafollowers + Instagram + Twitterfollowers + Tiktok +
summary(mod_combined)
```

```
##
## Call:
## lm(formula = Salary ~ MPG + PPG + AllStar + Instafollowers +
##      Instagram + Twitterfollowers + Tiktok + Tiktokmentions, data = data1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -23856410 -4867400  -204601  4965092 25115109
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -8.616e+06  6.242e+06  -1.380  0.16934
## MPG           3.754e+05  2.477e+05   1.516  0.13148
## PPG           5.644e+05  1.805e+05   3.127  0.00208 **
## AllStar       2.927e+06  3.663e+05   7.991  1.9e-13 ***
## Instafollowers  4.384e-01  1.907e-01   2.299  0.02274 *
## Instagram      1.308e+06  3.237e+06   0.404  0.68661
## Twitterfollowers -1.351e+00  4.555e-01  -2.967  0.00344 **
## Tiktok         -3.137e+06  1.459e+06  -2.150  0.03297 *
## Tiktokmentions  -2.881e-03  2.406e-03  -1.197  0.23287
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7689000 on 171 degrees of freedom
## Multiple R-squared:  0.5749, Adjusted R-squared:  0.555
## F-statistic: 28.91 on 8 and 171 DF, p-value: < 2.2e-16
```

```
step(mod_combined,direction = "both")
```

```
## Start: AIC=5716.7
## Salary ~ MPG + PPG + AllStar + Instafollowers + Instagram + Twitterfollowers +
## Tiktok + Tiktokmentions
##
##           Df Sum of Sq      RSS      AIC
## - Instagram      1 9.6575e+12 1.0120e+16 5714.9
## - Tiktokmentions  1 8.4754e+13 1.0196e+16 5716.2
## <none>                1.0111e+16 5716.7
## - MPG            1 1.3581e+14 1.0247e+16 5717.1
## - Tiktok         1 2.7328e+14 1.0384e+16 5719.5
## - Instafollowers  1 3.1239e+14 1.0423e+16 5720.2
## - Twitterfollowers 1 5.2046e+14 1.0631e+16 5723.7
## - PPG            1 5.7805e+14 1.0689e+16 5724.7
## - AllStar        1 3.7755e+15 1.3886e+16 5771.8
##
## Step: AIC=5714.87
## Salary ~ MPG + PPG + AllStar + Instafollowers + Twitterfollowers +
## Tiktok + Tiktokmentions
##
##           Df Sum of Sq      RSS      AIC
## - Tiktokmentions  1 8.4695e+13 1.0205e+16 5714.4
## <none>                1.0120e+16 5714.9
## - MPG            1 1.4673e+14 1.0267e+16 5715.5
## + Instagram      1 9.6575e+12 1.0111e+16 5716.7
## - Tiktok         1 2.6574e+14 1.0386e+16 5717.5
## - Instafollowers  1 3.1401e+14 1.0434e+16 5718.4
## - Twitterfollowers 1 5.1913e+14 1.0640e+16 5721.9
## - PPG            1 5.6973e+14 1.0690e+16 5722.7
## - AllStar        1 3.7664e+15 1.3887e+16 5769.8
##
## Step: AIC=5714.37
## Salary ~ MPG + PPG + AllStar + Instafollowers + Twitterfollowers +
## Tiktok
##
##           Df Sum of Sq      RSS      AIC
## <none>                1.0205e+16 5714.4
## + Tiktokmentions  1 8.4695e+13 1.0120e+16 5714.9
## - MPG            1 1.6781e+14 1.0373e+16 5715.3
## + Instagram      1 9.5989e+12 1.0196e+16 5716.2
## - Instafollowers  1 2.3069e+14 1.0436e+16 5716.4
## - Tiktok         1 3.7395e+14 1.0579e+16 5718.8
## - PPG            1 4.9444e+14 1.0700e+16 5720.9
## - Twitterfollowers 1 5.5763e+14 1.0763e+16 5721.9
## - AllStar        1 3.6937e+15 1.3899e+16 5768.0
##
## Call:
## lm(formula = Salary ~ MPG + PPG + AllStar + Instafollowers +
##     Twitterfollowers + Tiktok, data = data1)
##
## Coefficients:
```

```
##      (Intercept)          MPG          PPG          AllStar
##      -7.625e+06      4.127e+05      5.021e+05      2.880e+06
##   Instafollowers  Twitterfollowers          Tiktok
##      3.344e-01      -1.394e+00      -3.527e+06
```

```
mod_combined1 = lm(formula = Salary ~ MPG + PPG + AllStar + Instafollowers + Twitterfollowers + Tiktok,
summary(mod_combined1)
```

```
##
## Call:
## lm(formula = Salary ~ MPG + PPG + AllStar + Instafollowers +
##     Twitterfollowers + Tiktok, data = data1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -22230797 -4859230  -188134   5207357  25048218
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -7.625e+06  5.744e+06  -1.327  0.18614
## MPG           4.127e+05  2.447e+05   1.687  0.09347 .
## PPG           5.021e+05  1.734e+05   2.895  0.00428 **
## AllStar       2.880e+06  3.640e+05   7.913 2.88e-13 ***
## Instafollowers  3.344e-01  1.691e-01   1.978  0.04957 *
## Twitterfollowers -1.394e+00  4.534e-01  -3.075  0.00245 **
## Tiktok        -3.527e+06  1.401e+06  -2.518  0.01272 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7680000 on 173 degrees of freedom
## Multiple R-squared:  0.5709, Adjusted R-squared:  0.556
## F-statistic: 38.37 on 6 and 173 DF, p-value: < 2.2e-16
```

```
vif(mod_combined1)
```

```
##      MPG          PPG          AllStar  Instafollowers
##      2.639735      3.033342      2.871877      10.232132
##  Twitterfollowers          Tiktok
##      12.593616      1.035028
```

```
mod_combined2 = lm(formula = Salary ~ MPG + PPG + AllStar + Twitterfollowers + Tiktok, data = data1)
summary(mod_combined2)
```

```
##
## Call:
## lm(formula = Salary ~ MPG + PPG + AllStar + Twitterfollowers +
##     Tiktok, data = data1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -25979536 -4754957  -266579   5630033  25135863
##
```

```
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -7.943e+06  5.790e+06  -1.372  0.17188
## MPG         4.075e+05  2.467e+05   1.652  0.10040
## PPG         5.456e+05  1.735e+05   3.145  0.00195 **
## AllStar     2.744e+06  3.604e+05   7.614 1.62e-12 ***
## Twitterfollowers -5.859e-01  1.980e-01  -2.959  0.00352 **
## Tiktok      -3.338e+06  1.409e+06  -2.368  0.01896 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7744000 on 174 degrees of freedom
## Multiple R-squared:  0.5612, Adjusted R-squared:  0.5486
## F-statistic: 44.51 on 5 and 174 DF,  p-value: < 2.2e-16
```

```
vif(mod_combined2)
```

```
##              MPG              PPG              AllStar Twitterfollowers
##          2.639428          2.984654          2.769434          2.362108
##          Tiktok
##          1.030192
```

```
anova(mod_performance1, mod_combined2)
```

```
## Analysis of Variance Table
##
## Model 1: Salary ~ MPG + PPG + AllStar
## Model 2: Salary ~ MPG + PPG + AllStar + Twitterfollowers + Tiktok
##   Res.Df      RSS Df Sum of Sq    F Pr(>F)
## 1     176 1.1280e+16
## 2     174 1.0436e+16  2 8.4398e+14 7.036 0.001152 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
par(mfrow=c(2,3))
plot(formula = Salary ~ MPG + PPG + AllStar + Twitterfollowers + Tiktok, data = data1)
plot(mod_combined2)
```



