

Final Project - Predicting March Madness

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```
library(tidyverse)
library(tidymodels)
library(Stat2Data)
library(caret)
library(leaps)
library(MASS)

cbb <- read_csv("data/cbb.csv")

#Remove non-postseason teams and R68 losers
cbb <- cbb[!is.na(cbb$POSTSEASON),]
cbb <- filter(cbb, !grepl("R68", POSTSEASON))

#Cleaning up variable names, variables, etc
cbb$POSTSEASON <- str_trim(cbb$POSTSEASON, side = c("both"))
cbb <- rename(cbb, march_madness = POSTSEASON)

cbb <- cbb |>
  mutate(march_madness = case_when(
    march_madness == "Sweet Sixteen" ~ "S16",
    march_madness == "Elite Eight" ~ "E8",
    march_madness == "Final Four" ~ "F4",
    march_madness == "Finals" ~ "2ND",
    march_madness == "CHAMPS" ~ "Champions",
    TRUE ~ march_madness
  ))

cbb
```

```
# A tibble: 896 x 19
```

```
TEAM      march~1 Confe~2 ADJOE ADJDE `EFG%` `EFGD%` TOR TORD ORB DRB
```

	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	Gonzaga	S16	WCC	120.	89.9	58.7	43.1	15.7	16.3	29.1
2	Gonzaga	2ND	WCC	126.	90.9	61	47.7	16.1	19.3	30.1
3	Gonzaga	E8	WCC	124.	91	59	44.2	14.9	19	31.5
4	Villanova	Champi~	BE	129.	94.4	59.5	48.5	15	18.2	29.6
5	Gonzaga	2ND	WCC	118.	87.2	56.6	41.1	16.2	17.1	30
6	Villanova	Champi~	BE	124	92	56.1	46.7	16.3	20.6	28.2
7	Kentucky	F4	SEC	122.	86.1	51.5	39.6	16.3	21.3	39.5
8	Louisville	S16	Amer	119.	88.2	53.5	43.9	15.3	25	37.1
9	Louisville	Champi~	BE	117	86	50.6	44.8	18.3	27	38.2
10	Kentucky	Champi~	SEC	124.	90.4	53.8	42	17.1	17.6	37.8

... with 886 more rows, 8 more variables: FTR <dbl>, FTRD <dbl>, `2P%` <dbl>,
`2P%D` <dbl>, `3P%` <dbl>, `3P%D` <dbl>, `ADJ T.` <dbl>, ...19 <dbl>, and
abbreviated variable names 1: march_madness, 2: Conference

```
#separated each row by round for determining differences
cbb <- mutate(cbb, round_64 = if_else(march_madness == "R64", FALSE, TRUE))
round_64 <- cbb

cbb <- mutate(cbb, round_32 =
  case_when(march_madness == "R32" ~ FALSE,
    march_madness %in% c("S16", "E8", "F4", "2ND", "Champions") ~ TRUE,
    TRUE ~ NA))
round_32 <- cbb[!is.na(cbb$round_32),]

cbb <- mutate(cbb, sweet_sixteen =
  case_when(march_madness == "S16" ~ FALSE,
    march_madness %in% c("E8", "F4", "2ND", "Champions") ~ TRUE,
    TRUE ~ NA))
sweet_sixteen <- cbb[!is.na(cbb$sweet_sixteen),]

cbb <- mutate(cbb, elite_eight =
  case_when(march_madness == "E8" ~ FALSE,
    march_madness %in% c("F4", "2ND", "Champions") ~ TRUE,
    TRUE ~ NA))
elite_eight <- cbb[!is.na(cbb$elite_eight),]

cbb <- mutate(cbb, final_four =
  case_when(march_madness == "F4" ~ FALSE,
    march_madness %in% c("2ND", "Champions") ~ TRUE,
    TRUE ~ NA))
final_four <- cbb[!is.na(cbb$final_four),]
```

```

cbb <- mutate(cbb, champ_game =
  case_when(march_madness == "2ND" ~ FALSE,
    march_madness %in% c("Champions") ~ TRUE,
    TRUE ~ NA))
champ_game <- cbb[!is.na(cbb$champ_game),]

round_64_model <- glm(round_64 ~ ADJOE + ADJDE + `EFG%` +
  `EFGD%` + TOR + TORD + ORB +
  DRB + FTR + FTRD + `2P%` +
  `2P%D` + `3P%` + `ADJ T.` +
  ADJOE*ADJDE + `EFG%`*`EFGD%` +
  TOR*TORD + `2P%`*`2P%D`,
  data = round_64,
  family = "binomial")

round_64_min <- glm(round_64 ~ 1,
  data = round_64,
  family = "binomial")

stepAIC(round_64_min,
  scope = list(lower = round_64_min, upper = round_64_model),
  data = round_64, direction = "both")

```

Start: AIC=1244.12

round_64 ~ 1

	Df	Deviance	AIC
+ ADJOE	1	1031.8	1035.8
+ ADJDE	1	1040.5	1044.5
+ `EFGD%`	1	1197.5	1201.5
+ TOR	1	1207.0	1211.0
+ ORB	1	1210.9	1214.9
+ `2P%D`	1	1211.7	1215.7
+ FTRD	1	1224.4	1228.4
+ `EFG%`	1	1227.7	1231.7
+ `2P%`	1	1228.1	1232.1
+ `3P%`	1	1237.0	1241.0
+ FTR	1	1237.7	1241.7
+ TORD	1	1239.3	1243.3
<none>		1242.1	1244.1
+ `ADJ T.`	1	1241.7	1245.7

+ DRB 1 1242.0 1246.0

Step: AIC=1035.84

round_64 ~ ADJOE

	Df	Deviance	AIC
+ ADJDE	1	871.82	877.82
+ `EFGD%`	1	972.14	978.14
+ `2P%D`	1	999.46	1005.46
+ TORD	1	1002.69	1008.69
+ `EFG%`	1	1003.77	1009.77
+ `3P%`	1	1007.29	1013.29
+ `2P%`	1	1021.48	1027.48
+ ORB	1	1023.33	1029.33
+ `ADJ T.`	1	1028.65	1034.65
<none>		1031.84	1035.84
+ FTRD	1	1030.72	1036.72
+ TOR	1	1030.98	1036.98
+ FTR	1	1031.05	1037.05
+ DRB	1	1031.42	1037.42
- ADJOE	1	1242.12	1244.12

Step: AIC=877.82

round_64 ~ ADJOE + ADJDE

	Df	Deviance	AIC
+ ADJOE:ADJDE	1	863.63	871.63
+ `2P%D`	1	867.06	875.06
+ TOR	1	868.60	876.60
+ DRB	1	869.41	877.41
+ `EFG%`	1	869.60	877.60
<none>		871.82	877.82
+ TORD	1	869.99	877.99
+ `EFGD%`	1	870.22	878.22
+ `2P%`	1	870.32	878.32
+ `3P%`	1	870.87	878.87
+ FTRD	1	871.67	879.67
+ `ADJ T.`	1	871.71	879.71
+ FTR	1	871.81	879.81
+ ORB	1	871.82	879.82
- ADJDE	1	1031.84	1035.84
- ADJOE	1	1040.51	1044.51

Step: AIC=871.63

round_64 ~ ADJOE + ADJDE + ADJOE:ADJDE

	Df	Deviance	AIC
+ `2P%D`	1	857.43	867.43
+ TOR	1	860.62	870.62
+ DRB	1	861.05	871.05
+ `EFGD%`	1	861.37	871.37
+ `EFG%`	1	861.44	871.44
<none>		863.63	871.63
+ TORD	1	861.93	871.93
+ `2P%`	1	862.12	872.12
+ `3P%`	1	862.65	872.65
+ FTRD	1	863.49	873.49
+ FTR	1	863.50	873.50
+ `ADJ T.`	1	863.51	873.51
+ ORB	1	863.59	873.59
- ADJOE:ADJDE	1	871.82	877.82

Step: AIC=867.43

round_64 ~ ADJOE + ADJDE + `2P%D` + ADJOE:ADJDE

	Df	Deviance	AIC
+ DRB	1	853.77	865.77
+ `EFG%`	1	855.18	867.18
<none>		857.43	867.43
+ `2P%`	1	855.57	867.57
+ TOR	1	855.88	867.88
+ `EFGD%`	1	856.73	868.73
+ `3P%`	1	856.82	868.82
+ TORD	1	857.01	869.01
+ ORB	1	857.27	869.27
+ FTR	1	857.32	869.32
+ FTRD	1	857.40	869.40
+ `ADJ T.`	1	857.40	869.40
- `2P%D`	1	863.63	871.63
- ADJOE:ADJDE	1	867.06	875.06

Step: AIC=865.77

round_64 ~ ADJOE + ADJDE + `2P%D` + DRB + ADJOE:ADJDE

	Df	Deviance	AIC
+ TOR	1	850.64	864.64

<none>		853.77	865.77
+ `EFG%`	1	852.87	866.87
+ `2P%`	1	853.10	867.10
+ FTRD	1	853.25	867.25
+ `EFGD%`	1	853.25	867.25
- DRB	1	857.43	867.43
+ TORD	1	853.44	867.44
+ `3P%`	1	853.46	867.46
+ `ADJ T.`	1	853.68	867.68
+ ORB	1	853.75	867.75
+ FTR	1	853.77	867.77
- `2P%D`	1	861.05	871.05
- ADJOE:ADJDE	1	863.82	873.82

Step: AIC=864.64

round_64 ~ ADJOE + ADJDE + `2P%D` + DRB + TOR + ADJOE:ADJDE

	Df	Deviance	AIC
<none>		850.64	864.64
- TOR	1	853.77	865.77
+ ORB	1	850.09	866.09
+ `EFGD%`	1	850.26	866.26
+ TORD	1	850.27	866.27
+ `EFG%`	1	850.33	866.33
+ FTRD	1	850.35	866.35
+ FTR	1	850.44	866.44
+ `2P%`	1	850.46	866.46
+ `3P%`	1	850.48	866.48
+ `ADJ T.`	1	850.61	866.61
- DRB	1	855.88	867.88
- `2P%D`	1	856.01	868.01
- ADJOE:ADJDE	1	860.21	872.21

Call: glm(formula = round_64 ~ ADJOE + ADJDE + `2P%D` + DRB + TOR +
ADJOE:ADJDE, family = "binomial", data = round_64)

Coefficients:

(Intercept)	ADJOE	ADJDE	`2P%D`	DRB	TOR
-135.80213	1.42933	1.14692	0.08999	0.06315	-0.09017
ADJOE:ADJDE					
-0.01290					

Degrees of Freedom: 895 Total (i.e. Null); 889 Residual
Null Deviance: 1242
Residual Deviance: 850.6 AIC: 864.6

```
round_32_model <- glm(round_32 ~ ADJOE + ADJDE + `EFG%` +
  `EFGD%` + TOR + TORD + ORB +
  DRB + FTR + FTRD + `2P%` +
  `2P%D` + `3P%` + `ADJ T.` ,
  data = round_32,
  family = "binomial")

round_32_min <- glm(round_32 ~ 1,
  data = round_32,
  family = "binomial")

stepAIC(round_32_min,
  scope = list(lower = round_32_min,
    upper = round_32_model),
  data = round_32, direction = "both")
```

Start: AIC=623.06
round_32 ~ 1

	Df	Deviance	AIC
+ ADJOE	1	563.40	567.40
+ ADJDE	1	590.88	594.88
+ `EFG%`	1	605.82	609.82
+ `2P%`	1	607.28	611.28
+ FTRD	1	612.35	616.35
+ `EFGD%`	1	613.05	617.05
+ `2P%D`	1	614.75	618.75
+ `3P%`	1	615.23	619.23
+ TOR	1	616.56	620.56
+ ORB	1	616.60	620.60
<none>		621.06	623.06
+ `ADJ T.`	1	620.38	624.38
+ DRB	1	620.57	624.57
+ FTR	1	620.73	624.73
+ TORD	1	620.83	624.83

Step: AIC=567.4
round_32 ~ ADJOE

	Df	Deviance	AIC
+ ADJDE	1	519.16	525.16
+ `EFGD%`	1	550.12	556.12
+ `2P%D`	1	555.62	561.62
+ `3P%`	1	559.58	565.58
+ TORD	1	559.67	565.67
+ TOR	1	561.19	567.19
<none>		563.40	567.40
+ FTRD	1	561.82	567.82
+ ORB	1	562.15	568.15
+ `EFG%`	1	562.35	568.35
+ FTR	1	563.12	569.12
+ `ADJ T.`	1	563.37	569.37
+ DRB	1	563.38	569.38
+ `2P%`	1	563.38	569.38
- ADJOE	1	621.06	623.06

Step: AIC=525.16
round_32 ~ ADJOE + ADJDE

	Df	Deviance	AIC
<none>		519.16	525.16
+ `2P%D`	1	517.35	525.35
+ `EFGD%`	1	518.11	526.11
+ FTRD	1	518.20	526.20
+ DRB	1	518.31	526.31
+ FTR	1	518.42	526.42
+ ORB	1	518.74	526.74
+ `ADJ T.`	1	518.74	526.74
+ `3P%`	1	518.76	526.76
+ `2P%`	1	519.03	527.03
+ TOR	1	519.09	527.09
+ TORD	1	519.11	527.11
+ `EFG%`	1	519.15	527.15
- ADJDE	1	563.40	567.40
- ADJOE	1	590.88	594.88

Call: glm(formula = round_32 ~ ADJOE + ADJDE, family = "binomial",


```
data = round_32)
```

Coefficients:

(Intercept)	ADJOE	ADJDE
-2.7505	0.1842	-0.1949

Degrees of Freedom: 447 Total (i.e. Null); 445 Residual

Null Deviance: 621.1

Residual Deviance: 519.2 AIC: 525.2

```
sweet_sixteen_model <- glm(sweet_sixteen ~ ADJOE + ADJDE +
  `EFG%` + `EFGD%` + TOR + TORD +
  ORB + DRB + FTR + FTRD + `2P%` +
  `2P%D` + `3P%` + `ADJ T.` ,
  data = sweet_sixteen,
  family = "binomial")

sweet_sixteen_min <- glm(sweet_sixteen ~ 1,
  data = sweet_sixteen,
  family = "binomial")

stepAIC(sweet_sixteen_min,
  scope = list(lower = sweet_sixteen_min,
    upper = sweet_sixteen_model),
  data = sweet_sixteen, direction = "both")
```

Start: AIC=312.53

sweet_sixteen ~ 1

	Df	Deviance	AIC
+ ADJOE	1	296.66	300.66
+ ADJDE	1	302.34	306.34
+ ORB	1	305.41	309.41
+ `EFGD%`	1	306.80	310.80
+ `2P%`	1	307.64	311.64
+ `2P%D`	1	308.51	312.51
<none>		310.53	312.53
+ `EFG%`	1	309.10	313.10
+ TOR	1	309.43	313.43
+ FTR	1	309.68	313.68
+ FTRD	1	310.33	314.33

+ `3P%`	1	310.44	314.44
+ TORD	1	310.50	314.50
+ DRB	1	310.52	314.52
+ `ADJ T.`	1	310.52	314.52

Step: AIC=300.66
sweet_sixteen ~ ADJOE

	Df	Deviance	AIC
+ ADJDE	1	280.66	286.66
+ `3P%`	1	288.30	294.30
+ `EFGD%`	1	290.17	296.17
+ ORB	1	292.83	298.83
+ `EFG%`	1	293.43	299.43
+ `2P%D`	1	293.78	299.78
+ TORD	1	294.61	300.61
<none>		296.66	300.66
+ FTR	1	295.15	301.15
+ FTRD	1	295.76	301.76
+ TOR	1	295.93	301.93
+ DRB	1	296.42	302.42
+ `2P%`	1	296.53	302.53
+ `ADJ T.`	1	296.55	302.55
- ADJOE	1	310.53	312.53

Step: AIC=286.66
sweet_sixteen ~ ADJOE + ADJDE

	Df	Deviance	AIC
+ `3P%`	1	276.63	284.63
+ `EFG%`	1	278.34	286.34
<none>		280.66	286.66
+ DRB	1	279.25	287.25
+ FTRD	1	279.33	287.33
+ `2P%D`	1	279.58	287.58
+ ORB	1	279.89	287.89
+ FTR	1	280.08	288.08
+ `2P%`	1	280.10	288.10
+ `ADJ T.`	1	280.47	288.47
+ `EFGD%`	1	280.53	288.53
+ TOR	1	280.62	288.62
+ TORD	1	280.65	288.65
- ADJDE	1	296.66	300.66

- ADJOE 1 302.34 306.34

Step: AIC=284.63

sweet_sixteen ~ ADJOE + ADJDE + `3P%`

	Df	Deviance	AIC
<none>		276.63	284.63
+ DRB	1	275.41	285.41
+ FTRD	1	275.48	285.48
+ `2P%D`	1	275.99	285.99
+ `2P%`	1	276.11	286.11
+ `ADJ T.`	1	276.17	286.17
+ FTR	1	276.24	286.24
+ `EFG%`	1	276.36	286.36
+ TOR	1	276.40	286.40
+ ORB	1	276.53	286.53
+ TORD	1	276.61	286.61
+ `EFGD%`	1	276.62	286.62
- `3P%`	1	280.66	286.66
- ADJDE	1	288.30	294.30
- ADJOE	1	301.82	307.82

Call: glm(formula = sweet_sixteen ~ ADJOE + ADJDE + `3P%`, family = "binomial",
 data = sweet_sixteen)

Coefficients:

(Intercept)	ADJOE	ADJDE	`3P%`
-2.9987	0.1815	-0.1382	-0.1447

Degrees of Freedom: 223 Total (i.e. Null); 220 Residual

Null Deviance: 310.5

Residual Deviance: 276.6 AIC: 284.6

```
elite_eight_model <- glm(elite_eight ~ ADJOE + ADJDE +
  `EFG%` + `EFGD%` + TOR + TORD +
  ORB + DRB + FTR + FTRD + `2P%` +
  `2P%D` + `3P%` + `ADJ T.` ,
  data = elite_eight,
  family = "binomial")

elite_eight_min <- glm(elite_eight ~ 1,
```

```

        data = elite_eight,
        family = "binomial")

stepAIC(elite_eight_min,
        scope = list(lower = elite_eight_min,
                      upper = elite_eight_model),
        data = elite_eight, direction = "both")

```

Start: AIC=157.26

elite_eight ~ 1

	Df	Deviance	AIC
+ ADJDE	1	151.74	155.74
+ ADJOE	1	152.77	156.77
+ FTRD	1	153.07	157.07
<none>		155.26	157.26
+ `2P%`	1	153.99	157.99
+ TOR	1	154.41	158.41
+ `EFGD%`	1	154.44	158.44
+ DRB	1	154.59	158.59
+ `ADJ T.`	1	154.69	158.69
+ `3P%`	1	154.73	158.73
+ ORB	1	154.80	158.80
+ FTR	1	154.93	158.93
+ `2P%D`	1	155.03	159.03
+ `EFG%`	1	155.11	159.11
+ TORD	1	155.26	159.26

Step: AIC=155.73

elite_eight ~ ADJDE

	Df	Deviance	AIC
+ ADJOE	1	147.48	153.48
+ FTRD	1	149.09	155.09
+ TOR	1	149.47	155.47
<none>		151.74	155.74
+ `3P%`	1	149.91	155.91
+ `ADJ T.`	1	150.53	156.53
+ `2P%`	1	150.57	156.57
+ TORD	1	150.78	156.78
+ FTR	1	150.87	156.87
+ `2P%D`	1	150.94	156.94

-	ADJDE	1	155.26	157.26
+	DRB	1	151.32	157.32
+	`EFGD%`	1	151.36	157.36
+	ORB	1	151.71	157.71
+	`EFG%`	1	151.73	157.73

Step: AIC=153.48
 elite_eight ~ ADJDE + ADJOE

	Df	Deviance	AIC
+	`2P%`	1	139.57 147.57
+	`EFG%`	1	143.28 151.28
<none>			147.48 153.48
+	`ADJ T.`	1	145.90 153.90
+	`2P%D`	1	146.00 154.00
+	FTRD	1	146.75 154.75
+	`EFGD%`	1	146.85 154.85
+	FTR	1	147.01 155.01
+	TOR	1	147.18 155.18
+	TORD	1	147.26 155.26
+	`3P%`	1	147.36 155.36
+	DRB	1	147.36 155.36
+	ORB	1	147.46 155.46
-	ADJOE	1	151.74 155.74
-	ADJDE	1	152.77 156.77

Step: AIC=147.57
 elite_eight ~ ADJDE + ADJOE + `2P%`

	Df	Deviance	AIC
+	FTRD	1	136.54 146.54
+	ORB	1	137.46 147.46
+	`2P%D`	1	137.53 147.53
<none>			139.57 147.57
+	DRB	1	138.26 148.26
+	FTR	1	138.31 148.31
+	`EFGD%`	1	138.40 148.40
+	TORD	1	139.09 149.09
+	`ADJ T.`	1	139.29 149.29
+	TOR	1	139.39 149.39
+	`EFG%`	1	139.40 149.40
+	`3P%`	1	139.56 149.56
-	ADJDE	1	146.76 152.76

```
- `2P%`      1    147.48 153.48
- ADJOE      1    150.57 156.57
```

Step: AIC=146.54

```
elite_eight ~ ADJDE + ADJOE + `2P%` + FTRD
```

	Df	Deviance	AIC
+ `2P%D`	1	134.45	146.45
<none>		136.54	146.54
+ ORB	1	134.74	146.74
+ `EFGD%`	1	135.40	147.40
+ DRB	1	135.54	147.54
- FTRD	1	139.57	147.57
+ FTR	1	136.02	148.02
+ `ADJ T.`	1	136.13	148.13
+ `EFG%`	1	136.23	148.23
+ `3P%`	1	136.47	148.47
+ TORD	1	136.49	148.49
+ TOR	1	136.54	148.54
- ADJDE	1	143.77	151.77
- ADJOE	1	144.81	152.81
- `2P%`	1	146.75	154.75

Step: AIC=146.45

```
elite_eight ~ ADJDE + ADJOE + `2P%` + FTRD + `2P%D`
```

	Df	Deviance	AIC
<none>		134.45	146.45
- `2P%D`	1	136.54	146.54
+ `ADJ T.`	1	133.41	147.41
+ ORB	1	133.41	147.41
- FTRD	1	137.53	147.53
+ DRB	1	133.94	147.94
+ `EFG%`	1	134.15	148.15
+ FTR	1	134.29	148.29
+ TORD	1	134.30	148.30
+ `3P%`	1	134.37	148.37
+ `EFGD%`	1	134.39	148.39
+ TOR	1	134.41	148.41
- ADJDE	1	143.38	153.38
- ADJOE	1	143.83	153.83
- `2P%`	1	145.29	155.29

```
Call: glm(formula = elite_eight ~ ADJDE + ADJOE + `2P%` + FTRD + `2P%D`,
  family = "binomial", data = elite_eight)
```

Coefficients:

(Intercept)	ADJDE	ADJOE	`2P%`	FTRD	`2P%D`
12.22212	-0.23456	0.16972	-0.28213	-0.07466	0.14523

Degrees of Freedom: 111 Total (i.e. Null); 106 Residual

Null Deviance: 155.3

Residual Deviance: 134.5 AIC: 146.5

```
final_four_model <- glm(final_four ~ ADJOE + ADJDE + `EFG%` +
  `EFGD%` + TOR + TORD + ORB + DRB +
  FTR + FTRD + `2P%` + `2P%D` +
  `3P%` + `ADJ T.` ,
  data = final_four,
  family = "binomial")
```

```
final_four_min <- glm(final_four ~ 1,
  data = final_four,
  family = "binomial")
```

```
stepAIC(final_four_min,
  scope = list(lower = final_four_min,
    upper = final_four_model),
  data = final_four, direction = "both")
```

Start: AIC=79.63

final_four ~ 1

	Df	Deviance	AIC
+ ADJOE	1	72.316	76.316
+ `2P%`	1	72.766	76.766
+ TOR	1	73.669	77.669
+ `EFG%`	1	74.721	78.721
<none>		77.632	79.632
+ DRB	1	75.667	79.667
+ ADJDE	1	76.231	80.231
+ `ADJ T.`	1	76.675	80.675
+ `2P%D`	1	76.940	80.940

+ FTRD	1	77.267	81.267
+ `3P%`	1	77.385	81.385
+ TORD	1	77.426	81.426
+ `EFGD%`	1	77.476	81.476
+ FTR	1	77.605	81.605
+ ORB	1	77.624	81.624

Step: AIC=76.32
final_four ~ ADJOE

	Df	Deviance	AIC
+ ADJDE	1	69.269	75.269
<none>		72.316	76.316
+ `2P%D`	1	71.084	77.084
+ `2P%`	1	71.131	77.131
+ DRB	1	71.182	77.182
+ FTR	1	71.655	77.655
+ `EFGD%`	1	71.718	77.718
+ `3P%`	1	71.775	77.775
+ TOR	1	71.794	77.794
+ `ADJ T.`	1	71.981	77.981
+ FTRD	1	72.026	78.026
+ `EFG%`	1	72.240	78.240
+ ORB	1	72.277	78.277
+ TORD	1	72.293	78.293
- ADJOE	1	77.632	79.632

Step: AIC=75.27
final_four ~ ADJOE + ADJDE

	Df	Deviance	AIC
<none>		69.269	75.269
+ TOR	1	67.499	75.499
+ DRB	1	68.132	76.132
- ADJDE	1	72.316	76.316
+ `2P%`	1	68.541	76.541
+ TORD	1	68.724	76.724
+ ORB	1	68.730	76.730
+ `EFGD%`	1	68.944	76.944
+ FTRD	1	69.061	77.061
+ `EFG%`	1	69.149	77.149
+ `3P%`	1	69.158	77.158
+ `ADJ T.`	1	69.163	77.163

+ FTR	1	69.182	77.182
+ `2P%D`	1	69.268	77.268
- ADJOE	1	76.231	80.231

```
Call: glm(formula = final_four ~ ADJOE + ADJDE, family = "binomial",
  data = final_four)
```

Coefficients:

(Intercept)	ADJOE	ADJDE
-4.2298	0.1535	-0.1517

Degrees of Freedom: 55 Total (i.e. Null); 53 Residual

Null Deviance: 77.63

Residual Deviance: 69.27 AIC: 75.27

```
champ_game_model <- glm(champ_game ~ ADJOE + ADJDE +
  `EFG%` + `EFGD%` + TOR + TORD +
  ORB + DRB + FTR + FTRD + `2P%` +
  `2P%D` + `3P%` + `ADJ T.` ,
  data = champ_game,
  family = "binomial")
```

Warning: glm.fit: algorithm did not converge

Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

```
champ_game_min <- glm(champ_game ~ 1,
  data = champ_game,
  family = "binomial")

stepAIC(champ_game_min,
  scope = list(lower = champ_game_min,
    upper = champ_game_model),
  data = champ_game, direction = "both")
```

Start: AIC=40.82

champ_game ~ 1

	Df	Deviance	AIC
+ DRB	1	31.186	35.186
+ `3P%`	1	32.187	36.187
+ ADJOE	1	34.854	38.854
<none>		38.816	40.816
+ TORD	1	37.295	41.295
+ ORB	1	37.445	41.445
+ FTR	1	37.755	41.755
+ FTRD	1	38.393	42.393
+ `EFGD%`	1	38.443	42.443
+ `EFG%`	1	38.460	42.460
+ ADJDE	1	38.641	42.641
+ `2P%`	1	38.670	42.670
+ `2P%D`	1	38.718	42.718
+ TOR	1	38.801	42.801
+ `ADJ T.`	1	38.813	42.813

Step: AIC=35.19

champ_game ~ DRB

	Df	Deviance	AIC
+ `3P%`	1	21.981	27.981
+ ADJOE	1	24.532	30.532
+ `EFG%`	1	27.351	33.351
+ FTR	1	27.531	33.531
+ FTRD	1	28.161	34.161
<none>		31.186	35.186
+ TOR	1	30.319	36.319
+ `2P%`	1	30.333	36.333
+ `2P%D`	1	30.530	36.530
+ `EFGD%`	1	31.028	37.028
+ `ADJ T.`	1	31.124	37.124
+ ADJDE	1	31.152	37.152
+ ORB	1	31.178	37.178
+ TORD	1	31.186	37.186
- DRB	1	38.816	40.816

Step: AIC=27.98

champ_game ~ DRB + `3P%`

	Df	Deviance	AIC
+ ADJOE	1	19.881	27.881
<none>		21.981	27.981

+ ORB	1	20.541	28.541
+ `EFGD%`	1	21.270	29.270
+ FTR	1	21.305	29.305
+ `2P%D`	1	21.469	29.469
+ FTRD	1	21.788	29.788
+ `ADJ T.`	1	21.816	29.816
+ `EFG%`	1	21.905	29.905
+ `2P%`	1	21.921	29.921
+ ADJDE	1	21.928	29.928
+ TORD	1	21.962	29.962
+ TOR	1	21.964	29.964
- `3P%`	1	31.186	35.186
- DRB	1	32.187	36.187

Step: AIC=27.88

champ_game ~ DRB + `3P%` + ADJOE

	Df	Deviance	AIC
<none>		19.881	27.881
- ADJOE	1	21.981	27.981
+ `EFG%`	1	18.062	28.062
+ `2P%`	1	18.301	28.301
+ TOR	1	18.372	28.372
+ ORB	1	18.697	28.697
+ FTRD	1	19.537	29.537
+ ADJDE	1	19.556	29.556
+ `2P%D`	1	19.698	29.699
+ FTR	1	19.733	29.733
+ `ADJ T.`	1	19.756	29.756
+ TORD	1	19.834	29.834
+ `EFGD%`	1	19.850	29.850
- `3P%`	1	24.532	30.532
- DRB	1	31.482	37.482

Call: glm(formula = champ_game ~ DRB + `3P%` + ADJOE, family = "binomial",
data = champ_game)

Coefficients:

(Intercept)	DRB	`3P%`	ADJOE
-63.5938	0.6698	0.5753	0.1957

Degrees of Freedom: 27 Total (i.e. Null); 24 Residual
Null Deviance: 38.82
Residual Deviance: 19.88 AIC: 27.88