

# Jess, Anna and Seth Project

12/6/19

## project r markdown document

```
source('packages.R')
```

```
## Loading required package: optimx
## Loading required package: parallel
## Loading required package: minqa
## Loading required package: lme4
## Loading required package: Matrix
## Loading required package: segmented
## Loading required package: dplyr
##
## 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
## Loading required package: ggplot2
## Loading required package: ggcorrplot
```

```
source('styleguide.R')
source('helpers.R')
source('cleaner.R')
source('models.R')
```

```
## [1] "Dim Check Successful"
## [1] "Dim Check Successful"
## refitting model(s) with ML (instead of REML)
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00304692 (tol = 0.002, component 1)
## Loading required namespace: dfoptim
## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control$checkConv, :
## Model failed to converge with max|grad| = 0.00938467 (tol = 0.002, component 1)
## Loading required namespace: dfoptim
## refitting model(s) with ML (instead of REML)
```

## Get the Data

```
# Read in Clean DF
df.clean <- add_time("complete_data_clean.csv")
df.clean <- add_coach_change(df.clean)
df.tourney <- add_time("tourney_data_clean.csv")
df.tourney <- add_coach_change(df.tourney)

# Check dimensions - len(unique schools) * len(unique years) must equal # of rows
dim_checker(df.clean)

## [1] "Dim Check Successful"

dim_checker(df.tourney)

## [1] "Dim Check Successful"

get_newprop = cbind(df.tourney$School, get_prop_df(df.tourney))

# gets the minimum slope
sorted_alpha_school = df.tourney[order(df.tourney$School),]
sorted_alpha_prop = get_newprop[order(get_newprop$`df.tourney$School`),]
sorted_alpha_school$winner = (sorted_alpha_school$W.L. >= 0.5) * 1

total_wins <- aggregate(sorted_alpha_school$winner, by=list(sorted_alpha_school$School), FUN=sum)[2]
total_3s_made <- aggregate(sorted_alpha_prop$X3P, by=list(sorted_alpha_prop$`df.tourney$School`), FUN=sum)
total_3s_attempted <- aggregate(sorted_alpha_prop$X3PA, by=list(sorted_alpha_prop$`df.tourney$School`), FUN=sum)
total_sos <- aggregate(sorted_alpha_school$SOS, by=list(sorted_alpha_school$School), FUN=sum)[2]

idx_min <- which.min(coef(lmer3d)$School[,2])
school_list <- sort(unique(df.tourney$School)) # alphabetically sort the list
school_list[idx_min]

## [1] Butler
## 232 Levels: Air Force Akron Alabama Alabama A&M ... Yale

coef(lmer3d)$School[idx_min, 2]

## [1] -0.005638882

new_df <- data.frame(school = school_list, slope = coef(lmer3d)$School[,2], wins = total_wins, threes_made = total_3s_made, threes_attempted = total_3s_attempted, sos = total_sos)
new_df = new_df[order(new_df$slope),]
```

Find the Schools Most Prone and Least Prone by taking the Absolute value (slopes closer to 0 will be least impacted by the change) and we define most affected as a larger absolute value

```
# check for with absolute value
abs_df <- new_df
abs_df$slope = abs(abs_df$slope)
abs_df = abs_df[order(abs_df$slope),]
abs_df

##              school      slope  x      x.1      x.2
```

## 62	George Mason	2.790870e-05	12	5.721800	12.11
## 192	Temple	4.965195e-05	14	7.185309	78.46
## 193	Tennessee	5.843993e-05	14	6.813296	131.42
## 223	Western Kentucky	9.733913e-05	14	6.897856	-12.55
## 59	Florida State	1.097211e-04	15	6.462685	125.03
## 127	Nebraska	1.219053e-04	10	6.310638	114.91
## 139	North Carolina-Wilmington	1.276611e-04	7	6.744223	-13.77
## 6	Alabama-Birmingham	1.326660e-04	13	6.522186	26.18
## 135	North Carolina A&T	1.548420e-04	4	6.284508	-127.78
## 29	Central Connecticut State	1.712815e-04	4	6.017776	-109.53
## 161	Princeton	1.775000e-04	11	7.689089	-47.77
## 153	Oregon	3.229018e-04	14	7.516776	111.73
## 27	Cal State Fullerton	3.261316e-04	8	7.021127	-38.99
## 151	Old Dominion	3.668240e-04	15	5.791018	-6.92
## 148	Ohio State	3.708165e-04	15	6.761041	126.37
## 183	Southern Illinois	3.806682e-04	10	5.606899	38.59
## 231	Xavier	4.581981e-04	16	6.679811	99.84
## 77	Indiana	5.206040e-04	12	7.008277	129.17
## 167	Robert Morris	5.720029e-04	11	7.052139	-92.65
## 155	Pacific	5.836199e-04	9	6.494608	5.00
## 205	UC-Irvine	5.994744e-04	10	6.647123	-22.18
## 168	Saint Joseph's	6.363121e-04	11	6.897087	57.38
## 30	Central Florida	6.419951e-04	11	6.862061	18.65
## 186	St. Bonaventure	6.996270e-04	9	6.064039	22.12
## 145	Notre Dame	7.377932e-04	14	8.085865	118.65
## 65	Georgia	7.423079e-04	10	5.953399	119.52
## 96	Long Island University	7.861274e-04	9	7.053123	-118.59
## 67	Georgia Tech	7.952019e-04	7	5.608172	123.91
## 21	Bradley	8.013806e-04	8	6.167268	43.95
## 64	Georgetown	8.138741e-04	13	6.675569	124.15
## 14	Austin Peay	8.166123e-04	10	6.229611	-54.87
## 48	Delaware State	8.344733e-04	6	6.385202	-114.52
## 178	South Carolina	8.441718e-04	9	6.655275	110.81
## 38	College of Charleston	8.900277e-04	13	7.226049	-45.47
## 91	Lamar	9.010590e-04	9	6.142502	-83.81
## 215	Virginia	9.427137e-04	13	6.483533	128.79
## 90	Lafayette	9.736049e-04	5	8.156764	-73.30
## 39	Colorado	9.935891e-04	10	6.365386	105.60
## 162	Providence	9.961643e-04	10	6.544888	122.92
## 130	New Mexico	1.066940e-03	13	7.314909	50.60
## 124	Morgan State	1.134026e-03	5	5.015951	-133.87
## 58	Florida A&M	1.139391e-03	1	5.988670	-128.30
## 83	Jacksonville State	1.165320e-03	6	6.407183	-85.46
## 177	South Alabama	1.190902e-03	8	6.773379	-39.59
## 47	Delaware	1.193842e-03	5	6.198627	-24.96
## 28	Cal State Northridge	1.253521e-03	3	5.511947	-29.70
## 197	Texas Christian	1.258382e-03	6	6.459154	85.72
## 166	Richmond	1.278710e-03	11	7.023925	45.12
## 80	Iowa	1.280903e-03	11	6.650355	120.12
## 147	Ohio	1.307250e-03	11	7.244677	-3.65
## 171	Saint Peter's	1.319750e-03	6	6.359713	-58.27
## 132	Niagara	1.330204e-03	8	7.162730	-50.53
## 69	Green Bay	1.352620e-03	13	6.269094	-4.07
## 22	Brigham Young	1.371272e-03	15	7.156491	59.49

## 211	Valparaiso	1.389770e-03	11	6.976101	-4.98
## 101	Loyola (MD)	1.402737e-03	6	5.840579	-58.56
## 74	Houston	1.406796e-03	13	7.688357	32.48
## 78	Indiana State	1.456388e-03	5	6.684037	48.49
## 11	Arkansas	1.469489e-03	13	6.539985	97.80
## 70	Hampton	1.491714e-03	9	5.823553	-137.64
## 61	Gardner-Webb	1.493835e-03	9	7.225362	-75.49
## 60	Fresno State	1.515991e-03	8	7.215728	28.25
## 88	Kentucky	1.524159e-03	16	6.154633	137.21
## 129	Nevada-Las Vegas	1.529315e-03	15	6.885973	64.19
## 33	Cincinnati	1.530182e-03	14	6.602163	100.89
## 181	Southern	1.557363e-03	6	5.211291	-161.81
## 143	Northwestern	1.562525e-03	6	7.670868	115.15
## 108	Memphis	1.625274e-03	16	6.569505	68.71
## 15	Baylor	1.632899e-03	12	7.124214	126.00
## 97	Louisiana	1.653640e-03	10	6.676045	-32.86
## 159	Portland State	1.696784e-03	9	6.892471	-68.40
## 7	Albany (NY)	1.697746e-03	11	6.131447	-89.52
## 152	Oral Roberts	1.718486e-03	12	6.156634	-22.52
## 156	Penn State	1.722025e-03	6	6.484239	119.87
## 224	Western Michigan	1.727121e-03	11	6.394350	-6.74
## 75	Illinois	1.729081e-03	12	6.995798	133.43
## 31	Charlotte	1.771142e-03	7	7.041463	25.91
## 54	Eastern Kentucky	1.800401e-03	8	8.211955	-74.73
## 119	Mississippi Valley State	1.856378e-03	4	5.889948	-141.20
## 36	Coastal Carolina	1.875451e-03	10	6.724475	-86.19
## 16	Belmont	1.877112e-03	15	9.586143	-52.17
## 118	Mississippi State	1.881707e-03	12	6.823209	87.91
## 206	UC-Santa Barbara	1.895579e-03	13	6.725307	-20.67
## 122	Montana	1.902487e-03	14	6.873324	-53.46
## 4	Alabama A&M	1.910670e-03	2	5.397392	-177.09
## 87	Kent State	1.915808e-03	16	6.676689	-3.86
## 116	Minnesota	1.986769e-03	12	6.003933	126.07
## 52	Duke	2.011376e-03	16	7.887467	155.52
## 107	Massachusetts	2.019466e-03	8	6.887404	44.09
## 213	Vermont	2.026286e-03	15	6.061072	-74.68
## 26	Cal Poly	2.070102e-03	4	7.171318	-24.21
## 149	Oklahoma	2.086706e-03	12	6.866157	137.27
## 56	Fairleigh Dickinson	2.123456e-03	5	6.329318	-93.07
## 225	Wichita State	2.204181e-03	15	7.019215	60.38
## 5	Alabama State	2.213276e-03	7	5.898918	-168.43
## 184	Southern Methodist	2.215674e-03	7	6.250370	38.12
## 49	DePaul	2.273036e-03	4	6.047734	112.75
## 170	Saint Mary's (CA)	2.280819e-03	16	7.471634	33.82
## 66	Georgia State	2.326856e-03	8	6.626472	-34.88
## 111	Miami (OH)	2.333651e-03	6	6.361490	18.82
## 32	Chattanooga	2.363351e-03	10	7.350013	-55.39
## 182	Southern California	2.379242e-03	9	6.079078	113.76
## 227	Wisconsin	2.387344e-03	15	6.979702	141.29
## 228	Wofford	2.412120e-03	10	7.551247	-44.25
## 179	South Florida	2.439797e-03	3	5.047055	78.34
## 103	Marquette	2.440084e-03	15	7.204356	119.20
## 99	Louisville	2.466948e-03	16	7.496409	130.88
## 196	Texas A&M-Corpus Christi	2.470178e-03	10	5.321892	-73.10

## 195	Texas A&M	2.494641e-03	13	6.044516	104.01
## 212	Vanderbilt	2.523020e-03	12	7.993621	118.38
## 114	Middle Tennessee	2.538578e-03	14	6.000197	-18.79
## 165	Rhode Island	2.543603e-03	12	5.964717	44.10
## 20	Boston University	2.607326e-03	10	7.259560	-69.89
## 23	Bucknell	2.621217e-03	12	6.741496	-58.56
## 121	Monmouth	2.626323e-03	6	6.447372	-69.77
## 115	Milwaukee	2.714735e-03	9	7.419218	1.10
## 109	Mercer	2.722499e-03	8	7.094293	-60.00
## 134	North Carolina	2.729929e-03	16	6.271221	157.06
## 10	Arizona State	2.758011e-03	9	7.383960	104.62
## 207	UCLA	2.813067e-03	13	6.447906	121.42
## 17	Binghamton	2.816150e-03	2	6.259588	-85.43
## 216	Virginia Commonwealth	2.835503e-03	16	7.245661	29.08
## 57	Florida	2.847409e-03	15	7.588434	126.60
## 144	Northwestern State	2.879030e-03	8	6.211679	-69.89
## 51	Drake	2.883595e-03	6	7.336464	40.55
## 169	Saint Louis	2.885301e-03	11	5.734167	51.13
## 141	Northeastern	2.911363e-03	10	7.076050	-12.96
## 46	Dayton	2.965604e-03	14	6.703542	55.20
## 218	Wake Forest	2.997841e-03	8	6.216456	130.39
## 137	North Carolina-Asheville	3.012658e-03	8	6.429142	-65.19
## 154	Oregon State	3.019166e-03	8	5.747199	89.39
## 123	Morehead State	3.023707e-03	9	6.133421	-55.72
## 172	Sam Houston State	3.024168e-03	14	7.054085	-79.98
## 208	University of California	3.026062e-03	11	5.676435	114.80
## 34	Clemson	3.053811e-03	14	6.581742	119.01
## 98	Louisiana State	3.118372e-03	12	6.086522	102.33
## 201	Texas-El Paso	3.145531e-03	11	5.901999	9.68
## 176	Siena	3.168709e-03	9	5.891237	-36.43
## 8	American	3.217724e-03	10	6.462683	-79.77
## 131	New Mexico State	3.225442e-03	13	6.253432	-10.87
## 120	Missouri	3.231862e-03	10	7.055662	115.65
## 63	George Washington	3.258254e-03	10	5.648510	34.79
## 117	Mississippi	3.289159e-03	12	7.074049	96.19
## 136	North Carolina State	3.318133e-03	12	6.563238	124.84
## 194	Texas	3.351889e-03	14	6.603725	140.57
## 92	Lehigh	3.378968e-03	12	6.770819	-86.33
## 175	Seton Hall	3.430851e-03	12	6.559728	113.11
## 45	Davidson	3.441282e-03	16	8.925587	-4.21
## 157	Pennsylvania	3.444370e-03	7	6.806416	-30.58
## 9	Arizona	3.455211e-03	16	6.456764	127.18
## 13	Auburn	3.470796e-03	6	7.512625	97.08
## 100	Loyola (IL)	3.601446e-03	7	5.874828	0.23
## 43	Cornell	3.645366e-03	4	7.596776	-46.09
## 140	North Texas	3.717216e-03	11	6.054695	-45.62
## 203	Troy	3.816750e-03	4	8.779607	-52.43
## 174	San Diego State	3.828083e-03	14	6.031051	67.69
## 89	La Salle	3.909470e-03	6	6.832841	36.17
## 126	Murray State	3.922218e-03	15	7.033542	-63.31
## 142	Northern Iowa	4.059802e-03	14	7.250434	60.82
## 71	Harvard	4.069929e-03	10	6.375169	-40.46
## 110	Miami (FL)	4.106440e-03	13	7.058740	117.45
## 3	Alabama	4.111946e-03	15	6.144707	117.73

## 84	James Madison	4.124645e-03	5	6.370749	-27.90
## 18	Boise State	4.139649e-03	11	7.599539	28.63
## 222	West Virginia	4.257394e-03	14	7.405296	135.42
## 94	Lipscomb	4.273298e-03	10	7.756064	-70.72
## 44	Creighton	4.305770e-03	15	8.211294	81.65
## 85	Kansas	4.321536e-03	16	6.806275	155.83
## 82	Jackson State	4.322197e-03	5	5.397949	-156.81
## 160	Prairie View	4.403416e-03	3	5.471237	-148.17
## 95	Little Rock	4.423617e-03	7	5.804571	-41.89
## 188	Stanford	4.620269e-03	12	5.993873	114.23
## 35	Cleveland State	4.682915e-03	6	6.536270	1.94
## 204	Tulsa	4.724622e-03	12	6.203486	45.07
## 1	Air Force	4.791179e-03	7	7.551358	28.56
## 133	Norfolk State	4.849900e-03	8	5.633081	-135.23
## 138	North Carolina-Greensboro	4.873969e-03	6	6.754625	-49.34
## 202	Texas-San Antonio	4.927869e-03	8	6.895634	-51.29
## 219	Washington	4.992617e-03	14	6.433361	109.75
## 221	Weber State	5.013173e-03	13	7.284673	-57.12
## 209	Utah	5.060943e-03	10	6.760732	88.26
## 232	Yale	5.079997e-03	10	5.948941	-44.96
## 163	Purdue	5.099678e-03	12	6.741528	130.76
## 146	Oakland	5.215012e-03	9	7.514128	-16.78
## 164	Radford	5.256009e-03	8	6.228929	-82.25
## 113	Michigan State	5.265836e-03	16	6.583259	145.03
## 189	Stephen F. Austin	5.313053e-03	14	5.719587	-101.35
## 112	Michigan	5.328781e-03	13	7.552010	141.21
## 12	Arkansas-Pine Bluff	5.352610e-03	2	5.302537	-141.64
## 180	Southeastern Louisiana	5.354288e-03	10	6.071915	-95.57
## 104	Marshall	5.420877e-03	8	7.229566	17.83
## 68	Gonzaga	5.447725e-03	16	6.716946	63.89
## 86	Kansas State	5.456449e-03	15	6.048941	126.17
## 158	Pittsburgh	5.519558e-03	13	6.027521	120.84
## 220	Washington State	5.589081e-03	6	6.726650	96.44
## 93	Liberty	5.599490e-03	7	6.709843	-85.98
## 25	Butler	5.638882e-03	13	7.592211	77.56
## 150	Oklahoma State	5.681481e-03	13	6.902494	134.27
## 190	Stony Brook	5.696688e-03	9	5.964597	-86.47
## 191	Syracuse	5.703737e-03	16	6.402456	132.36
## 24	Buffalo	5.785106e-03	13	6.801858	-2.92
## 187	St. John's (NY)	5.786628e-03	7	5.467189	122.86
## 229	Wright State	5.835519e-03	13	6.273270	-13.54
## 76	Illinois-Chicago	5.889671e-03	8	6.128775	-7.25
## 53	East Tennessee State	5.897911e-03	14	6.681445	-52.78
## 40	Colorado State	5.923076e-03	10	6.260372	55.15
## 210	Utah State	5.931023e-03	15	6.799605	9.68
## 73	Holy Cross	6.048317e-03	7	6.021283	-71.05
## 55	Eastern Washington	6.074115e-03	7	7.771243	-46.53
## 50	Detroit Mercy	6.136221e-03	7	6.198857	0.05
## 198	Texas Southern	6.179689e-03	8	5.725888	-140.53
## 102	Manhattan	6.204649e-03	7	5.983108	-47.39
## 37	Colgate	6.251096e-03	4	7.063172	-78.87
## 125	Mount St. Mary's	6.280049e-03	7	6.767301	-90.44
## 72	Hawaii	6.412211e-03	12	5.974273	-12.30
## 173	San Diego	6.443620e-03	8	6.695405	26.83

```
## 81          Iowa State 6.528966e-03 11 7.404039 128.89
## 200        Texas-Arlington 6.685139e-03 11 6.889081 -51.57
## 2          Akron 6.838891e-03 14 8.093001 -7.63
## 226        Winthrop 6.847741e-03 12 7.143976 -79.14
## 214        Villanova 6.990611e-03 15 7.944232 135.88
## 199        Texas Tech 7.033891e-03 9 5.538718 124.81
## 19        Boston College 7.400558e-03 7 6.588496 117.58
## 41        Connecticut 7.528914e-03 13 5.894859 114.57
## 128        Nevada 7.869737e-03 12 6.260792 38.89
## 217        Virginia Tech 7.964500e-03 11 6.269554 102.13
## 106 Maryland-Baltimore County 7.983819e-03 4 6.211620 -85.24
## 185        Southern Mississippi 8.523459e-03 8 5.687583 15.37
## 42        Coppin State 8.868988e-03 2 6.734071 -99.70
## 79          Iona 8.913573e-03 11 7.580094 -36.21
## 105        Maryland 8.971031e-03 16 6.165858 124.85
## 230        Wyoming 9.297598e-03 9 6.490703 43.16
```

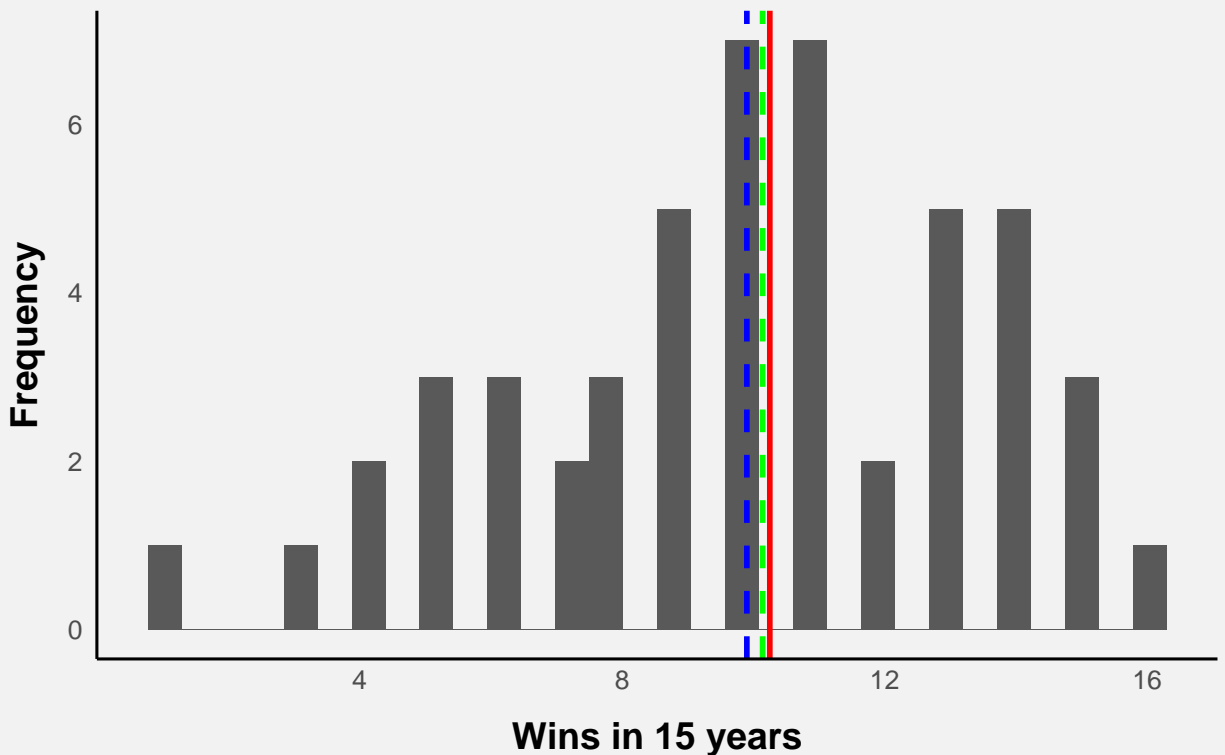
```
close_zero_least <- abs_df[1:50,]
close_zero_most <- abs_df[181:230,]
```

```
# histograms - teams least prone
```

```
ggplot(close_zero_least, aes(x = x)) +
  geom_histogram() +
  geom_vline(aes(xintercept=mean(total_wins$x)),
    color="red", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_least$x)),
    color="blue", linetype="dashed", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_most$x)),
    color="green", linetype="dashed", size=1) +
  theme_hodp() +
  labs(title="Teams Least Prone to 3PA Rate Change") +
  xlab("Wins in 15 years") +
  ylab("Frequency")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

# Teams Least Prone to 3PA Rate Change



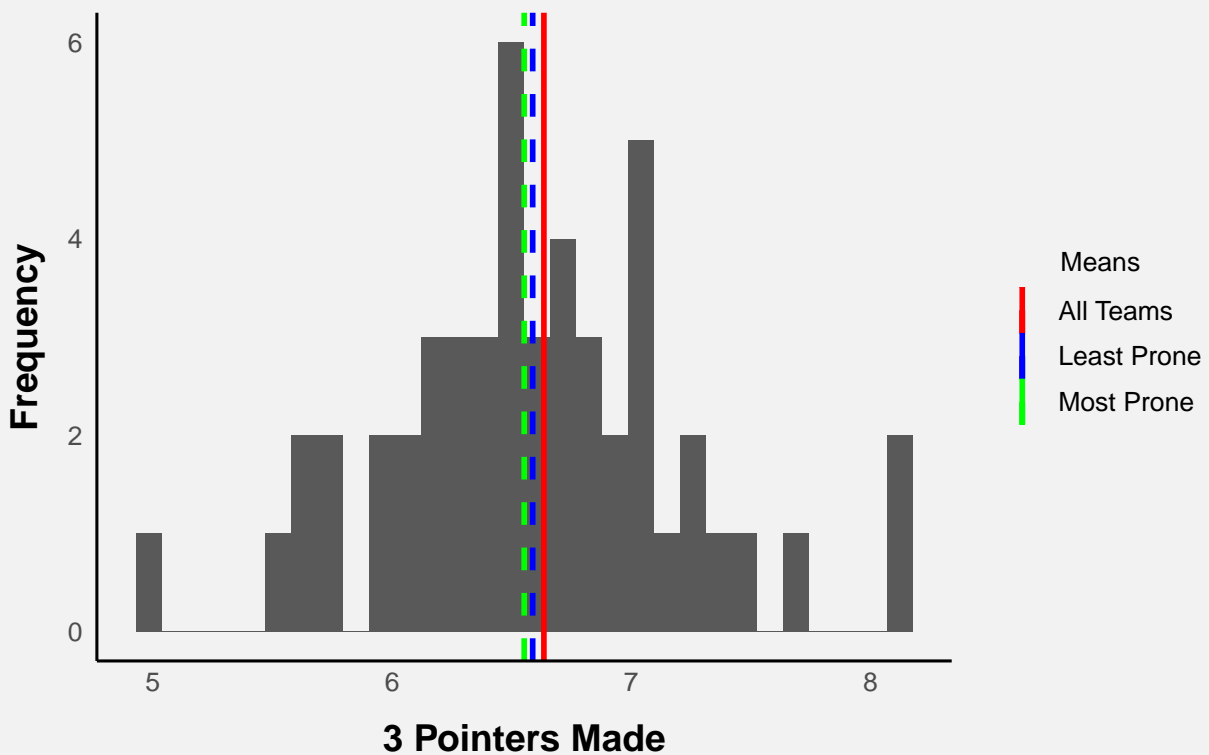
```
cols <- c("All Teams"="red", "Most Prone"="green", "Least Prone"="blue")
```

```
ggplot(close_zero_least, aes(x = x.1)) +
  geom_histogram() +
  geom_vline(aes(xintercept=mean(total_3s_made$x), color="All Teams"), size=1) +
  geom_vline(aes(xintercept=mean(close_zero_least$x.1), color="Least Prone"), linetype="dashed", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_most$x.1), color="Most Prone"), linetype="dashed", size=1) +
  theme_hodp() +
  labs(title="Teams Least Prone to 3PA Rate Change") +
  theme(legend.position = "right") +
  scale_color_manual("Means", values = cols) +
  xlab("3 Pointers Made") +
  ylab("Frequency")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



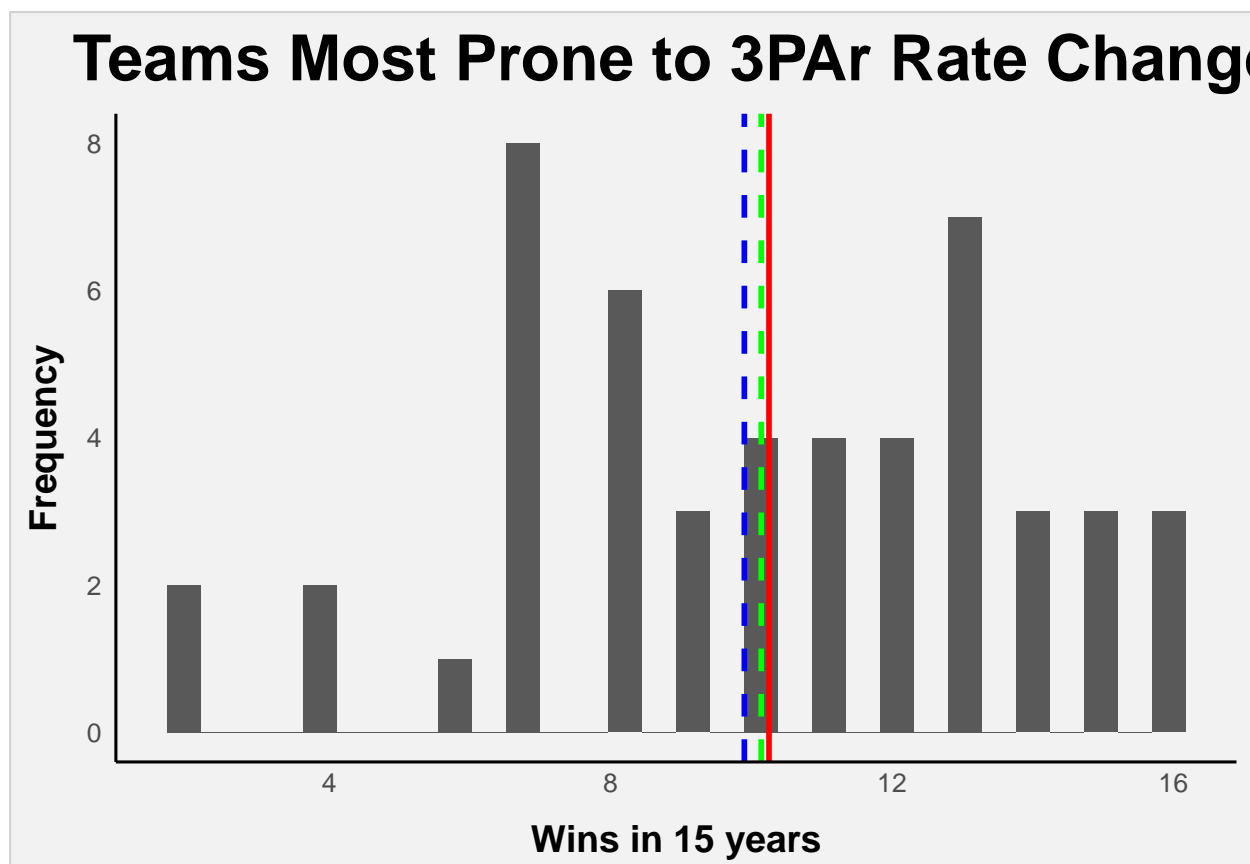
## Teams Least Prone to 3PAr Rate Change



```
# histogram teams most prone
```

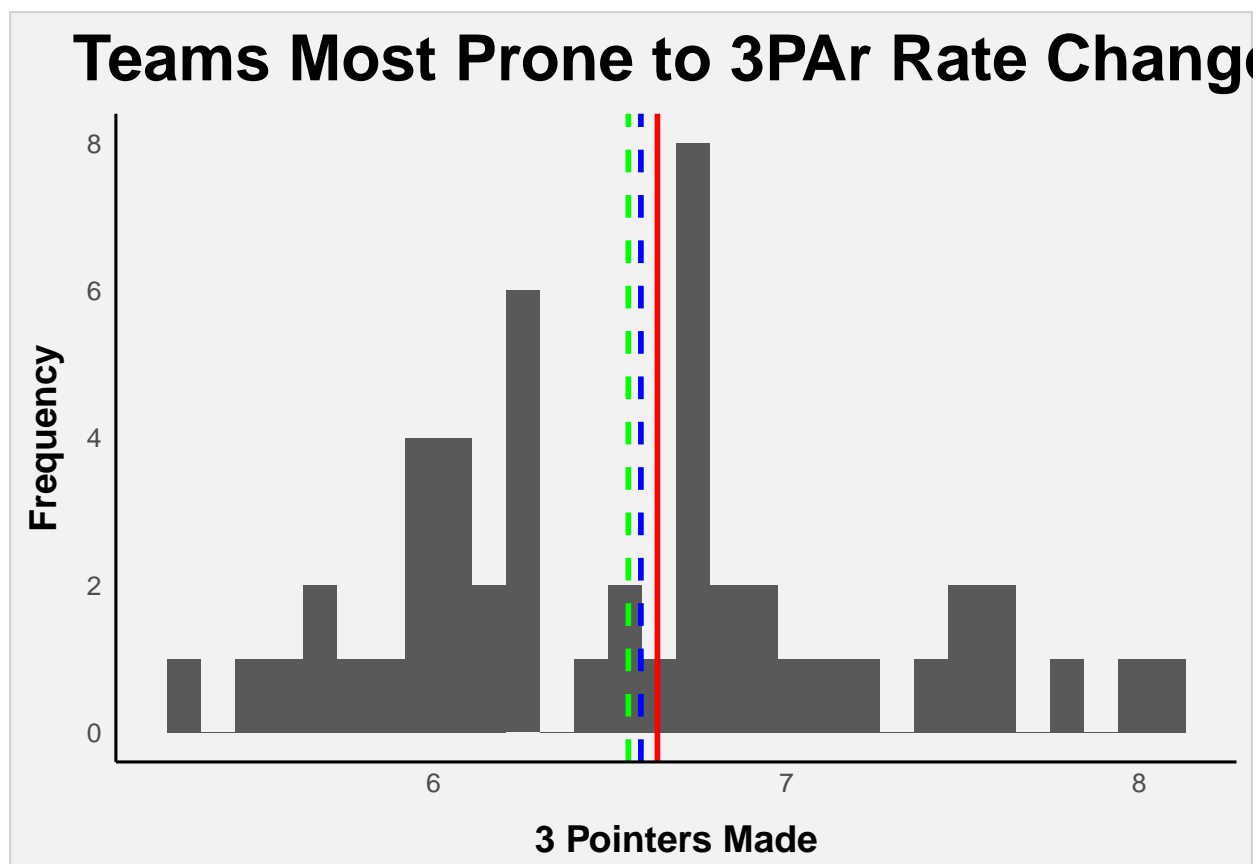
```
ggplot(close_zero_most, aes(x = x)) +
  geom_histogram() +
  geom_vline(aes(xintercept=mean(total_wins$x)),
    color="red", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_most$x)),
    color="green", linetype="dashed", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_least$x)),
    color="blue", linetype="dashed", size=1) +
  theme_hodp() +
  labs(title="Teams Most Prone to 3PAr Rate Change") +
  xlab("Wins in 15 years") +
  ylab("Frequency")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
ggplot(close_zero_most, aes(x = x.1)) +
  geom_histogram() +
  geom_vline(aes(xintercept=mean(total_3s_made$x)),
    color="red", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_most$x.1)),
    color="green", linetype="dashed", size=1) +
  geom_vline(aes(xintercept=mean(close_zero_least$x.1)),
    color="blue", linetype="dashed", size=1) +
  theme_hodp() +
  labs(title="Teams Most Prone to 3PAr Rate Change") +
  theme(legend.position = c(0.8, 0.2)) +
  xlab("3 Pointers Made") +
  ylab("Frequency")
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



## Comparing the Means with formal hypothesis testing

```
# between the most and least affected teams - games
t.test(close_zero_least$x, close_zero_most$x)
```

```
##
## Welch Two Sample t-test
##
## data: close_zero_least$x and close_zero_most$x
## t = -0.33755, df = 97.98, p-value = 0.7364
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.650956 1.170956
## sample estimates:
## mean of x mean of y
## 9.90 10.14
```

```
# between the most and least affected - 3 pointers made per game
t.test(close_zero_least$x.1, close_zero_most$x.1)
```

```
##
## Welch Two Sample t-test
##
## data: close_zero_least$x.1 and close_zero_most$x.1
## t = 0.27396, df = 97.588, p-value = 0.7847
```

```
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.2216405 0.2926328
## sample estimates:
## mean of x mean of y
## 6.587455 6.551959
```

## Conducting a similar method on the piecewise linear model

```
# more tests on the piece slopes
piece.slopes = coef(lmer9a)$School
piece.slopes$time = abs(piece.slopes$time)
piece.slopes = piece.slopes[order(piece.slopes$time),]

era.zeroslopes = piece.slopes[, 2]
era.oneslopes = era.zeroslopes + piece.slopes[, 5]
era.twoslopes = era.zeroslopes + piece.slopes[, 6]

piece_df <- data.frame(school = school_list, slope0 = era.zeroslopes, slope1 = era.oneslopes, slope2 = era.twoslopes)
piece_df0 = piece_df[order(piece_df$slope0),]
piece_df1 = piece_df[order(piece_df$slope1),]
piece_df2 = piece_df[order(piece_df$slope2),]
```

## Comparing the Means for piecewise with formal hypothesis testing

```
piece_df0_least <- piece_df0[1:50, ]
piece_df0_most <- piece_df0[(nrow(piece_df0) - 50):nrow(piece_df0), ]
piece_df1_least <- piece_df1[1:50, ]
piece_df1_most <- piece_df1[(nrow(piece_df1) - 50):nrow(piece_df1), ]
piece_df2_least <- piece_df2[1:50, ]
piece_df2_most <- piece_df2[(nrow(piece_df2) - 50):nrow(piece_df2), ]
```

```
# t tests for wins
mean(total_wins$x)
```

```
## [1] 10.25
```

```
t.test(piece_df0_least$x, piece_df0_most$x)
```

```
##
```

```
## Welch Two Sample t-test
```

```
##
```

```
## data: piece_df0_least$x and piece_df0_most$x
```

```
## t = -2.7837, df = 86.811, p-value = 0.006595
```

```
## alternative hypothesis: true difference in means is not equal to 0
```

```
## 95 percent confidence interval:
```

```
## -3.4838802 -0.5812178
```

```
## sample estimates:
```

```
## mean of x mean of y
```

```
## 9.34000 11.37255
```

```

t.test(piece_df1_least$x, piece_df1_most$x)

##
## Welch Two Sample t-test
##
## data: piece_df1_least$x and piece_df1_most$x
## t = -2.7837, df = 86.811, p-value = 0.006595
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -3.4838802 -0.5812178
## sample estimates:
## mean of x mean of y
## 9.34000 11.37255

t.test(piece_df2_least$x, piece_df2_most$x)

##
## Welch Two Sample t-test
##
## data: piece_df2_least$x and piece_df2_most$x
## t = -2.7837, df = 86.811, p-value = 0.006595
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -3.4838802 -0.5812178
## sample estimates:
## mean of x mean of y
## 9.34000 11.37255

# t tests for 3 pointers made, p value 0.05
mean(total_3s_made$x)

## [1] 6.6343

t.test(piece_df0_least$x.1, piece_df0_most$x.1)

##
## Welch Two Sample t-test
##
## data: piece_df0_least$x.1 and piece_df0_most$x.1
## t = 1.7417, df = 95.697, p-value = 0.08477
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.03686786 0.56456499
## sample estimates:
## mean of x mean of y
## 6.794718 6.530870

t.test(piece_df1_least$x.1, piece_df1_most$x.1)

##
## Welch Two Sample t-test
##
## data: piece_df1_least$x.1 and piece_df1_most$x.1
## t = 1.7417, df = 95.697, p-value = 0.08477
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.03686786 0.56456499

```

```
## sample estimates:
## mean of x mean of y
## 6.794718 6.530870

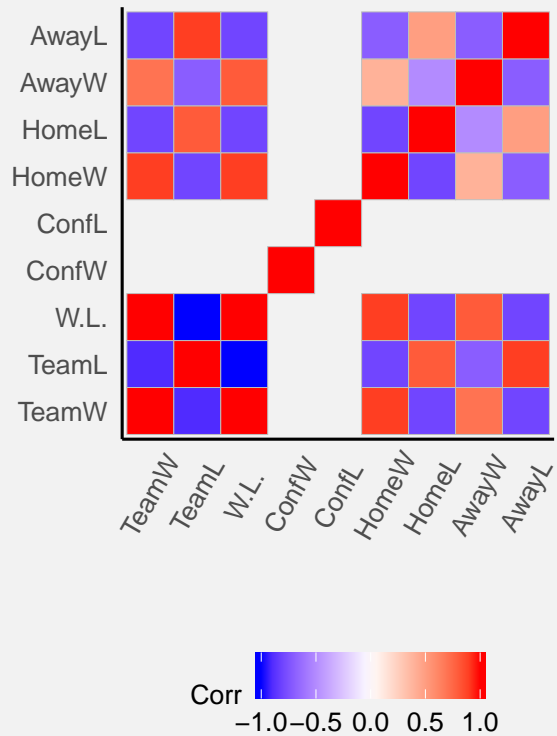
t.test(piece_df2_least$x.1, piece_df2_most$x.1)

##
## Welch Two Sample t-test
##
## data: piece_df2_least$x.1 and piece_df2_most$x.1
## t = 1.7417, df = 95.697, p-value = 0.08477
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.03686786 0.56456499
## sample estimates:
## mean of x mean of y
## 6.794718 6.530870
```

## Comparisons between Schools

```
#eda correlation
data <- read.csv('data/full_data_raw.csv')
wl <- data %>% select(TeamW, TeamL, W.L., ConfW, ConfL, HomeW, HomeL, AwayW, AwayL)
cor <- round(cor(wl), 1)
p <- ggcorrplot(cor) +
  labs(title='Corr Plot for W-L Vars') +
  xlab('') + ylab('') +
  theme_hodp() +
  theme(axis.text.x=element_text(angle=60))
p
```

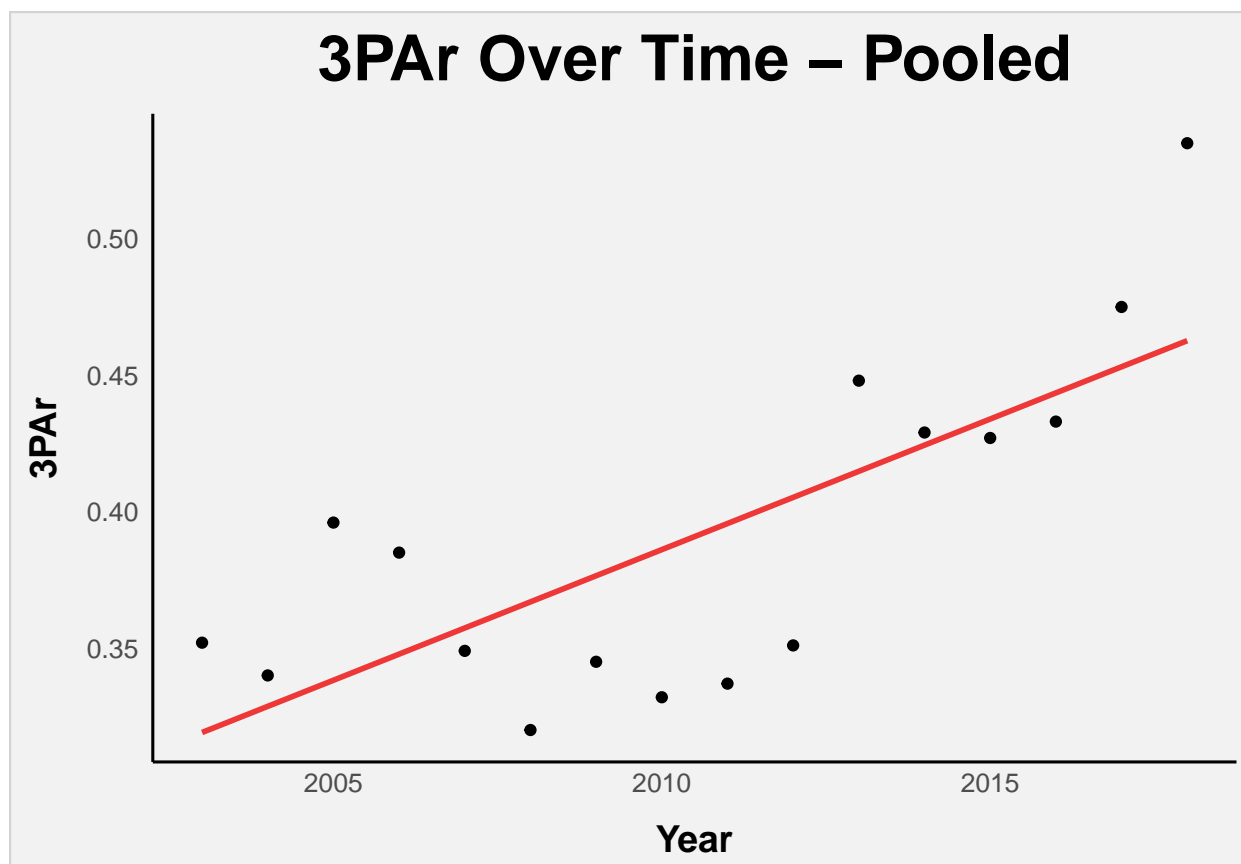
# Corr Plot for W-L Vars



```
#look at effects for individual teams
uva = df.tourney[df.tourney$School == 'Virginia',]
nova = df.tourney[df.tourney$School == 'Villanova',]
data_teams = rbind(uva, nova)
#interaction between school and time and time and era
data_teams$era <- as.factor((data_teams$year > 2006) + (data_teams$year > 2012))

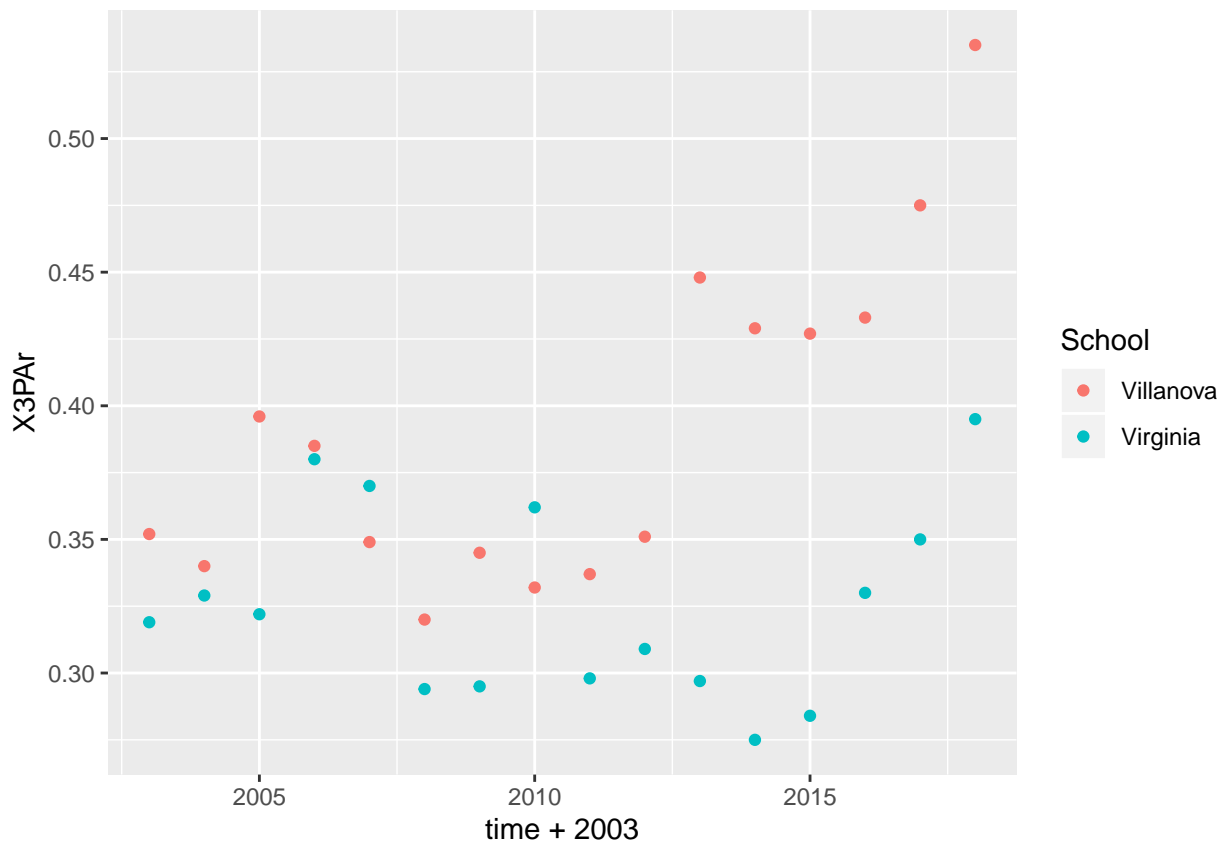
#plot villanova and overall linear fit
p <- ggplot(nova, aes(x = time + 2003, y = X3PAr)) +
  geom_point() +
  stat_smooth(method = "lm", col = '#EE3838', se = F) +
  labs(title="3PAr Over Time - Pooled") +
  xlab("Year") +
  ylab("3PAr") +
  #ylim(c(0,0.6)) +
  theme_hodp()
```

p



```
#plot villanova and uva by color  
p <- ggplot(data_teams, aes(x = time + 2003, y = X3PAr, color = School)) +  
  geom_point(aes(color = School))  
p
```





```
#interaction between time and school
lmer9 <- lmer(X3PAR ~ time*era + (1 + time|School), data=data_teams) # fails to converge
lm2 <- lm(X3PAR ~ time*School, data_teams)
summary(lm2)
```

```
##
## Call:
## lm(formula = X3PAR ~ time * School, data = data_teams)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.058660 -0.030860 -0.005394  0.032978  0.072346
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.319096   0.019212  16.610 4.97e-16 ***
## time           0.009571   0.002182   4.386 0.000148 ***
## SchoolVirginia  0.006853   0.027169   0.252 0.802705
## time:SchoolVirginia -0.009622  0.003086  -3.118 0.004189 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.04024 on 28 degrees of freedom
## Multiple R-squared:  0.5901, Adjusted R-squared:  0.5462
## F-statistic: 13.44 on 3 and 28 DF, p-value: 1.287e-05
```

```

names(df.clean)

## [1] "School"      "G"           "TeamW"       "TeamL"       "W.L."
## [6] "SRS"         "SOS"         "Tm."         "Opp."        "FTr"
## [11] "X3PAr"       "TS."         "TRB."        "AST."        "BLK."
## [16] "eFG."        "TOV."        "FT.FGA"      "FG"          "FGA"
## [21] "FG."         "X3P"         "X3PA"        "X3P."        "FT"
## [26] "FTA"         "FT."         "ORB"         "TRB"         "AST"
## [31] "STL"         "BLK"         "TOV"         "PF"          "year"
## [36] "time"        "same.coach"

lm3 <- lm(X3PAr ~ time*(School + era), data_teams)
lmer10 <- lmer(X3PAr ~ time*era + (1 + time|School), data=data_teams) # fails to converge

# Compare mixed model to lm
summary(lm3)

##
## Call:
## lm(formula = X3PAr ~ time * (School + era), data = data_teams)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.04196 -0.01550 -0.00673  0.02005  0.06335
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.324624   0.021217  15.300 7.05e-14 ***
## time           0.021361   0.009922   2.153 0.041598 *
## SchoolVirginia  0.006853   0.021031   0.326 0.747367
## era1           0.018645   0.039896   0.467 0.644467
## era2          -0.177502   0.068934  -2.575 0.016618 *
## time:SchoolVirginia -0.009622  0.002389  -4.028 0.000491 ***
## time:era1      -0.019093   0.011169  -1.709 0.100269
## time:era2       0.002593   0.011169   0.232 0.818392
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.03115 on 24 degrees of freedom
## Multiple R-squared:  0.7895, Adjusted R-squared:  0.7281
## F-statistic: 12.86 on 7 and 24 DF,  p-value: 9.297e-07

summary(lmer10)

## Linear mixed model fit by REML ['lmerMod']
## Formula: X3PAr ~ time * era + (1 + time | School)
## Data: data_teams
##
## REML criterion at convergence: -87.2
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.4080 -0.5117 -0.2005  0.6737  2.0573
##
## Random effects:
## Groups   Name      Variance Std.Dev. Corr

```

```
## School (Intercept) 2.303e-05 0.004799
## time 4.539e-05 0.006737 -1.00
## Residual 9.315e-04 0.030520
## Number of obs: 32, groups: School, 2
##
## Fixed effects:
## Estimate Std. Error t value
## (Intercept) 0.328050 0.018372 17.856
## time 0.016550 0.010763 1.538
## era1 0.018645 0.039090 0.477
## era2 -0.177502 0.067543 -2.628
## time:era1 -0.019093 0.010944 -1.745
## time:era2 0.002593 0.010944 0.237
##
## Correlation of Fixed Effects:
## (Intr) time era1 era2 tim:r1
## time -0.788
## era1 -0.454 0.332
## era2 -0.263 0.192 0.123
## time:era1 0.695 -0.791 -0.731 -0.189
## time:era2 0.695 -0.791 -0.327 -0.639 0.778
```

```
AIC(lm3)
```

```
## [1] -122.4083
```

```
AIC(lmer10)
```

```
## [1] -67.22047
```

```
# lm outperforms mixed model
```

```
intercept.0 <- summary(lm3)$coef[1,1]
slope.0 <- summary(lm3)$coef[2,1]
intercept.virginia <- summary(lm3)$coef[3,1]
intercept.1 <- summary(lm3)$coef[4,1]
intercept.2 <- summary(lm3)$coef[5,1]
slope.virginia <- summary(lm3)$coef[6,1]
slope.1 <- summary(lm3)$coef[7,1]
slope.2 <- summary(lm3)$coef[8,1]
```

```
#plot lm3
```

```
year <- 2003:2017
```

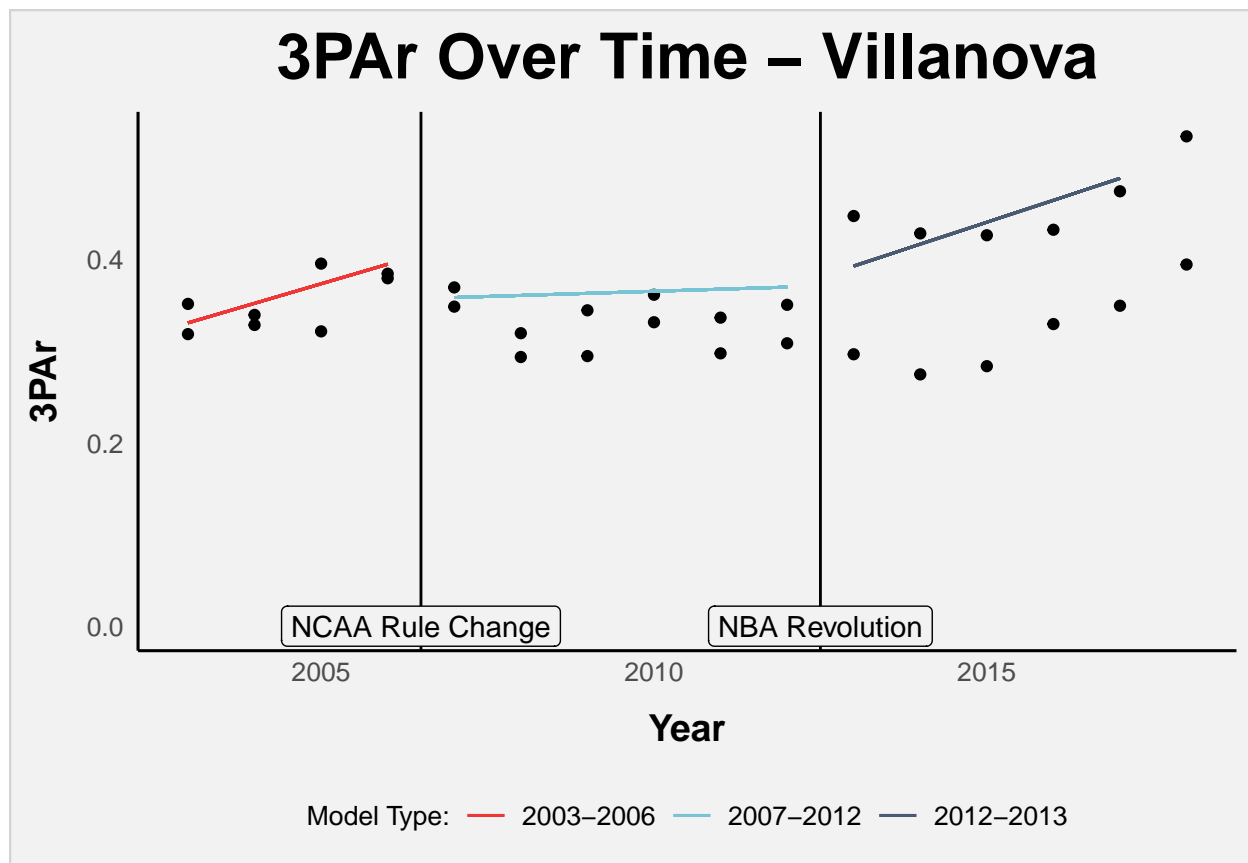
```
fnnova <- function(year, era1, era2) {
  return(intercept.0 + (year - 2003) * slope.0 +
         intercept.1*era1 + slope.1 * (year - 2003) * era1 +
         intercept.2*era2 + slope.2 * (year - 2003) * era2)
}
```

```
fnuva <- function(year, era1, era2) {
  return(intercept.0 + intercept.virginia + (year - 2003 + slope.virginia) * slope.0 +
         intercept.1*era1 + slope.1 * (year - 2003) * era1 +
         intercept.2*era2 + slope.2 * (year - 2003) * era2)
}
```

```

#plot interaction between school and time and time and era for just uva
p <- ggplot(data_teams, aes(x = time + 2003, y = X3PAr)) +
  geom_point() +
  geom_segment(aes(x = 2003, y = fnuva(2003,0,0), xend = 2006, yend = fnuva(2006,0,0),
    colour = '#EE3838'),
    data = data_teams) +
  geom_segment(aes(x = 2007, y = fnuva(2007,1,0), xend = 2012, yend = fnuva(2012,1,0),
    colour = '#78C4D4'),
    data = data_teams) +
  geom_segment(aes(x = 2013, y = fnuva(2013,0,1), xend = 2017, yend = fnuva(2017,0,1),
    colour = '#4B5973'),
    data = data_teams) +
  geom_vline(xintercept = 2012.5) +
  geom_vline(xintercept = 2006.5) +
  scale_colour_identity(name="Model Type:",
    breaks = c('#EE3838', '#78C4D4', '#4B5973'),
    labels = c("2003-2006", "2007-2012", "2012-2013"),
    guide = "legend") +
  annotate(geom="label", x = 2012.5, y = 0, label = "NBA Revolution", fill = "#F2F2F2", color = "black")
  annotate(geom="label", x = 2006.5, y = 0, label = "NCAA Rule Change", fill = "#F2F2F2", color = "black")
  labs(title="3PAr Over Time - Villanova") +
  xlab("Year") +
  ylab("3PAr") +
  theme_hodp()
p

```



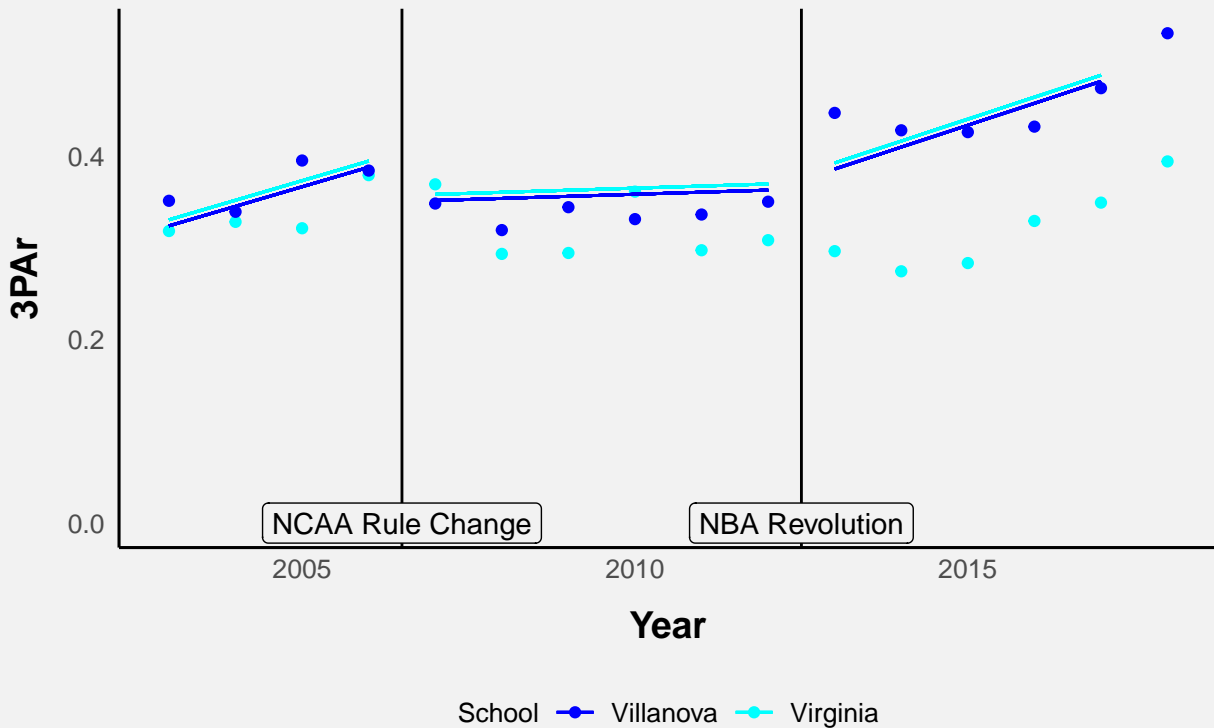
```

#plot interaction between school and time and time and era for both schools
p <- ggplot(data_teams, aes(x = time + 2003, y = X3PAr, color = School)) +
  geom_point(aes(color = School)) +
  scale_color_manual(values = c("#0000FF", "#00FFFF", "#0000FF", "#00FFFF"))+
  #geom_point(color = c('#EE3838', '#78C4D4')) +
  geom_segment(aes(x = 2003, y = fnnova(2003,0,0), xend = 2006, yend = fnnova(2006,0,0),
    colour = 'Villanova'),
    data = data_teams) +
  geom_segment(aes(x = 2007, y = fnnova(2007,1,0), xend = 2012, yend = fnnova(2012,1,0),
    colour = 'Villanova'),
    data = data_teams) +
  geom_segment(aes(x = 2013, y = fnnova(2013,0,1), xend = 2017, yend = fnnova(2017,0,1),
    colour = 'Villanova'),
    data = data_teams) +

  geom_segment(aes(x = 2003, y = fnuva(2003,0,0), xend = 2006, yend = fnuva(2006,0,0),
    colour = 'Virginia'),
    data = data_teams) +
  geom_segment(aes(x = 2007, y = fnuva(2007,1,0), xend = 2012, yend = fnuva(2012,1,0),
    colour = 'Virginia'),
    data = data_teams) +
  geom_segment(aes(x = 2013, y = fnuva(2013,0,1), xend = 2017, yend = fnuva(2017,0,1),
    colour = 'Virginia'),
    data = data_teams) +
  geom_vline(xintercept = 2012.5) +
  geom_vline(xintercept = 2006.5) +
  annotate(geom="label", x = 2012.5, y = 0, label = "NBA Revolution", fill = "#F2F2F2", color = "black")
  annotate(geom="label", x = 2006.5, y = 0, label = "NCAA Rule Change", fill = "#F2F2F2", color = "black")
  labs(title="3PAr Over Time By School") +
  xlab("Year") +
  ylab("3PAr") +
  theme_hodp()
p

```

# 3PAR Over Time By School



```
#interaction between school and time and time and era and school and era
data_teams$era <- as.factor((data_teams$year > 2006) + (data_teams$year > 2012))
lm4 <- lm(X3PAR ~ time*(School + era) +School*era , data_teams)
summary(lm4)
```

```
##
## Call:
## lm(formula = X3PAR ~ time * (School + era) + School * era, data = data_teams)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-0.034627	-0.019078	-0.000164	0.017521	0.042160

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.3427594	0.0184098	18.618	5.92e-15 ***
time	0.0169937	0.0087054	1.952	0.0638 .
SchoolVirginia	-0.0294187	0.0203126	-1.448	0.1616
era1	0.0098848	0.0381567	0.259	0.7980
era2	-0.1297586	0.0667713	-1.943	0.0649 .
time:SchoolVirginia	-0.0008875	0.0058036	-0.153	0.8799
time:era1	-0.0190929	0.0093065	-2.052	0.0523 .
time:era2	0.0025929	0.0093065	0.279	0.7831
SchoolVirginia:era1	0.0175208	0.0374620	0.468	0.6446
SchoolVirginia:era2	-0.0954875	0.0680944	-1.402	0.1748

```
## ---
```

```

## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02595 on 22 degrees of freedom
## Multiple R-squared:  0.866, Adjusted R-squared:  0.8112
## F-statistic: 15.8 on 9 and 22 DF,  p-value: 1.14e-07

intercept.0 <- summary(lm4)$coef[1,1]
slope.0 <- summary(lm4)$coef[2,1]
intercept.virginia <- summary(lm4)$coef[3,1]
intercept.1 <- summary(lm4)$coef[4,1]
intercept.2 <- summary(lm4)$coef[5,1]
slope.virginia <- summary(lm4)$coef[6,1]
slope.1 <- summary(lm4)$coef[7,1]
slope.2 <- summary(lm4)$coef[8,1]
intercept.1.virginia <- summary(lm4)$coef[9,1]
intercept.2.virginia <- summary(lm4)$coef[10,1]

year <- 2003:2017
fnnova <- function(year, era1, era2) {
  return(intercept.0 + (year - 2003) * slope.0 +
    intercept.1*era1 + slope.1 * (year - 2003) * era1 +
    intercept.2*era2 + slope.2 * (year - 2003) * era2)
}
fnuva <- function(year, era1, era2) {
  return(intercept.0 + intercept.virginia + (year - 2003 + slope.virginia) * slope.0 +
    (intercept.1 + intercept.1.virginia) * era1 + slope.1 * (year - 2003) * era1 +
    (intercept.2 + intercept.2.virginia) * era2 + slope.2 * (year - 2003) * era2)
}

p <- ggplot(data_teams, aes(x = time + 2003, y = X3PAr, color = School)) +
  geom_point(aes(color = School)) +
  scale_color_manual(values = c("#EE3838", "#0000FF", "#EE3838", "#0000FF"))+
  #geom_point(color = c('#EE3838', '#78C4D4')) +
  geom_segment(aes(x = 2003, y = fnnova(2003,0,0), xend = 2006, yend = fnnova(2006,0,0),
    colour = 'Villanova'),
    data = data_teams) +
  geom_segment(aes(x = 2007, y = fnnova(2007,1,0), xend = 2012, yend = fnnova(2012,1,0),
    colour = 'Villanova'),
    data = data_teams) +
  geom_segment(aes(x = 2013, y = fnnova(2013,0,1), xend = 2017, yend = fnnova(2017,0,1),
    colour = 'Villanova'),
    data = data_teams) +

  geom_segment(aes(x = 2003, y = fnuva(2003,0,0), xend = 2006, yend = fnuva(2006,0,0),
    colour = 'Virginia'),
    data = data_teams) +
  geom_segment(aes(x = 2007, y = fnuva(2007,1,0), xend = 2012, yend = fnuva(2012,1,0),
    colour = 'Virginia'),
    data = data_teams) +
  geom_segment(aes(x = 2013, y = fnuva(2013,0,1), xend = 2017, yend = fnuva(2017,0,1),
    colour = 'Virginia'),
    data = data_teams) +
  geom_vline(xintercept = 2012.5) +

```

```
geom_vline(xintercept = 2006.5) +
  annotate(geom="label", x = 2012.5, y = 0, label = "NBA Revolution", fill = "#F2F2F2", color = "black")
  annotate(geom="label", x = 2006.5, y = 0, label = "NCAA Rule Change", fill = "#F2F2F2", color = "black")
labs(title="3PAr Over Time By School") +
  xlab("Year") +
  ylab("3PAr") +
  theme_hodp()
```

p

