

Vikings_EDA

Exploring Vikings NFL Data

```
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.4      v readr      2.1.5
v forcats    1.0.0      v stringr    1.5.1
v ggplot2    4.0.0      v tibble     3.2.1
v lubridate  1.9.4      v tidyr      1.3.1
v purrr      1.0.2
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
vikes_data <- read_csv("final_vikings_data.csv")
```

```
Rows: 27612 Columns: 255
```

```
-- Column specification -----
```

```
Delimiter: ","
```

```
chr   (74): home_team, away_team, posteam, posteam_type, defteam, side_of_fi...
```

```
dbl   (147): play_id, game_id, yardline_100, quarter_seconds_remaining, half_...
```

```
lgl   (32): lateral_receiver_player_id, lateral_receiver_player_name, latera...
```

```
date   (1): game_date
```

```
time   (1): time
```

```
i Use `spec()` to retrieve the full column specification for this data.
```

```
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

Contingency Tables and Numeric Summaries

First, visualize vikings play type per down.

```
table(vikes_data$down, vikes_data$play_type)
```

	extra_point	field_goal	kickoff	no_play	pass	punt	qb_kneel	qb_spike	run
1	0	17	0	516	4169	0	133	35	4522
2	0	17	0	400	3800	0	58	1	2771
3	0	16	0	371	3304	0	34	2	818
4	0	562	0	118	194	1488	1	0	102

Seems like the Vikings are more likely to pass over run on later downs.

Let's look at counts and ratios of pass and run plays per year.

```
library(lubridate)
vikes_table_1 <-vikes_data|>
  mutate(year = year(game_date)) |>
  group_by(year) |>
  summarize(run_count=sum(play_type=="run",na.rm=TRUE ), pass_count = sum(play_type=="pass
  mutate(run_ratio = run_count/(run_count+pass_count),pass_ratio = pass_count/(run_count+p

vikes_table_1
```

A tibble: 10 x 5

	year	run_count	pass_count	run_ratio	pass_ratio
	<dbl>	<int>	<int>	<dbl>	<dbl>
1	2009	758	1098	0.408	0.592
2	2010	832	1088	0.433	0.567
3	2011	872	1142	0.433	0.567
4	2012	942	1246	0.431	0.569
5	2013	843	1275	0.398	0.602
6	2014	837	1136	0.424	0.576
7	2015	811	1028	0.441	0.559
8	2016	758	1228	0.382	0.618
9	2017	897	1198	0.428	0.572
10	2018	678	1057	0.391	0.609

I want to create a table that shows average yards per play by year.

```
vikes_table_2 <- vikes_data |>
  mutate(year = year(game_date)) |>
  group_by(year) |>
  summarize(
    avg_yards = mean(yards_gained, na.rm = TRUE),
    yards_sd = sd(yards_gained, na.rm = TRUE)
  )
```

```
vikes_table_2
```

```
# A tibble: 10 x 3
  year avg_yards yards_sd
  <dbl>   <dbl>   <dbl>
1  2009     3.94     8.08
2  2010     3.77     7.52
3  2011     3.94     7.95
4  2012     3.79     7.64
5  2013     4.11     8.31
6  2014     3.79     7.47
7  2015     3.86     8.06
8  2016     3.60     7.23
9  2017     3.66     7.36
10 2018     3.87     7.79
```

Cool! Vikings were averaging a high 4.1 yards per play in 2013. Surprisingly, the Vikings were 5-10-1 in spite of this.

Let's add the Vikings wins to this table to look at how yards per play relates to games won.

```
# Adding in a vector with wins is easier than trying to extract this information from a pl

wins<-c(12,6,3,10,5,7,11,8,13,8)

vikes_table_2$wins=wins

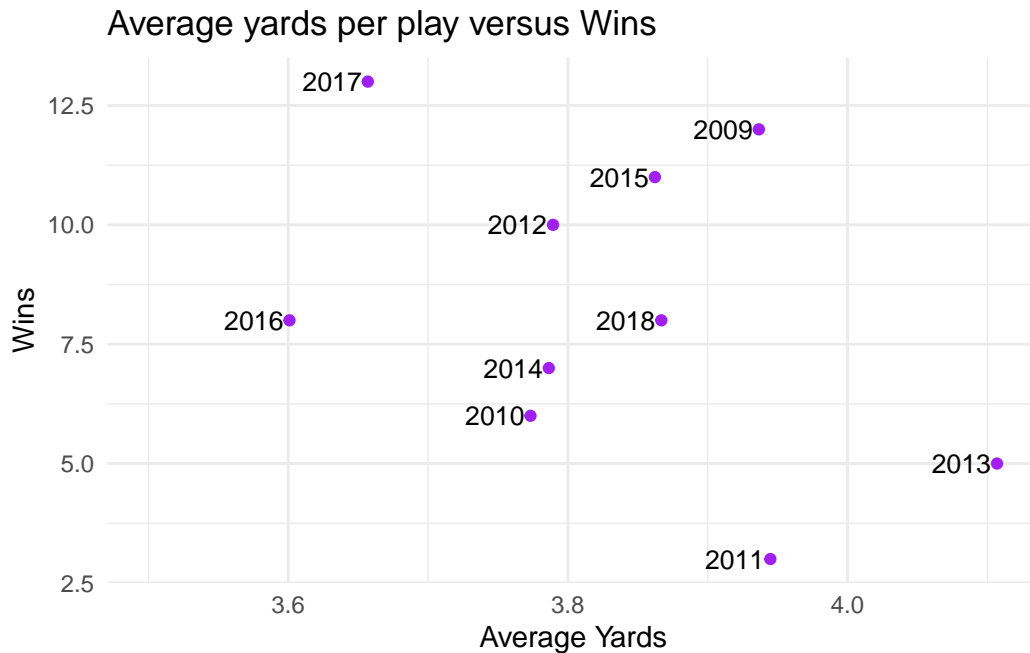
vikes_table_2
```

```
# A tibble: 10 x 4
  year avg_yards yards_sd wins
  <dbl>   <dbl>   <dbl> <dbl>
1  2009     3.94     8.08    12
```

2	2010	3.77	7.52	6
3	2011	3.94	7.95	3
4	2012	3.79	7.64	10
5	2013	4.11	8.31	5
6	2014	3.79	7.47	7
7	2015	3.86	8.06	11
8	2016	3.60	7.23	8
9	2017	3.66	7.36	13
10	2018	3.87	7.79	8

Visualization

```
library(ggplot2)
ggplot(vikes_table_2, aes(x=avg_yards, y = wins))+
  geom_point(color="purple")+
  geom_text(aes(label=year), hjust = 1.1, size = 3.5)+
  labs(
    title = "Average yards per play versus Wins",
    x = "Average Yards",
    y = "Wins"
  )+
  theme_minimal()+
  expand_limits(x = min(vikes_table_2$avg_yards) - 0.1)
```



There is no obvious relationship between these two variables.

Let's next look at yards/run plays and yards/pass plays by year.

```
vikes_table_3 <- vikes_data |>
  mutate(year = year(game_date)) |>
  group_by(year) |>
  summarize(
    avg_yards = mean(yards_gained, na.rm = TRUE),
    yards_sd = sd(yards_gained, na.rm = TRUE),
    avg_run_yrds = mean(ifelse(play_type=="run", yards_gained, NA), na.rm=TRUE),
    avg_pass_yrds = mean(ifelse(play_type=="pass", yards_gained, NA), na.rm=TRUE)
  )
vikes_table_3
```

A tibble: 10 x 5

	year	avg_yards	yards_sd	avg_run_yrds	avg_pass_yrds
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	2009	3.94	8.08	4.18	6.51
2	2010	3.77	7.52	4.31	6.06
3	2011	3.94	7.95	4.65	6.15
4	2012	3.79	7.64	4.81	5.63

5	2013	4.11	8.31	4.60	6.29
6	2014	3.79	7.47	4.52	6.00
7	2015	3.86	8.06	4.64	6.15
8	2016	3.60	7.23	3.77	5.81
9	2017	3.66	7.36	4.06	6.03
10	2018	3.87	7.79	4.29	6.19

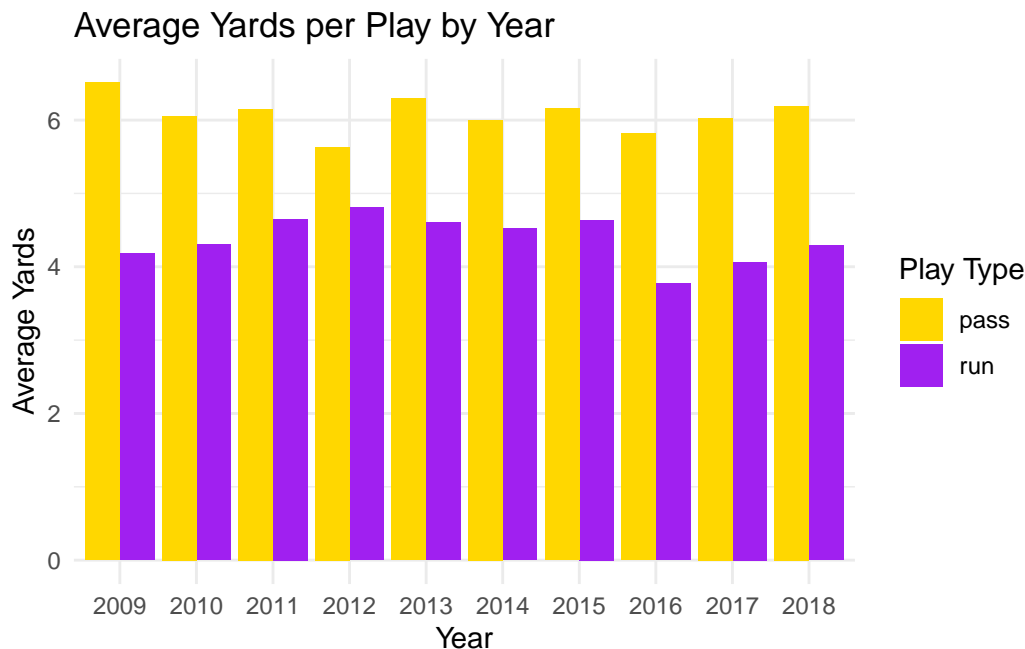
```
# I want to visualize this so I'm going to pivot longer
vikes_table_3_long<-vikes_table_3|>
  rename(
    run = avg_run_yrds,
    pass = avg_pass_yrds
  )|>
  pivot_longer(cols =c(run,pass),
               names_to = "play_type",
               values_to= "average_yards")
vikes_table_3_long
```

```
# A tibble: 20 x 5
   year avg_yards yards_sd play_type average_yards
  <dbl>   <dbl>   <dbl> <chr>         <dbl>
1  2009     3.94     8.08 run           4.18
2  2009     3.94     8.08 pass          6.51
3  2010     3.77     7.52 run           4.31
4  2010     3.77     7.52 pass          6.06
5  2011     3.94     7.95 run           4.65
6  2011     3.94     7.95 pass          6.15
7  2012     3.79     7.64 run           4.81
8  2012     3.79     7.64 pass          5.63
9  2013     4.11     8.31 run           4.60
10 2013     4.11     8.31 pass          6.29
11 2014     3.79     7.47 run           4.52
12 2014     3.79     7.47 pass          6.00
13 2015     3.86     8.06 run           4.64
14 2015     3.86     8.06 pass          6.15
15 2016     3.60     7.23 run           3.77
16 2016     3.60     7.23 pass          5.81
17 2017     3.66     7.36 run           4.06
18 2017     3.66     7.36 pass          6.03
19 2018     3.87     7.79 run           4.29
20 2018     3.87     7.79 pass          6.19
```

```

ggplot(vikes_table_3_long, aes(x = factor(year), y = average_yards, fill = play_type)) +
  geom_col(position = "dodge") +
  labs(
    title = "Average Yards per Play by Year",
    x = "Year",
    y = "Average Yards",
    fill = "Play Type"
  ) +
  scale_fill_manual(
    values = c("run" = "purple",
               "pass" = "gold")
  ) +
  theme_minimal()

```



- Vikings fans will fondly remember 2012 as Adrian Peterson rushing for 2000+ yards in 2012 after tearing his ACL and winning MVP. SKOL. This year the Vikings had the highest rush yards/attempt of any year.
- Vikings fans will also remember 2009 as the year Brett Favre threw for 4200 yards and took the Vikings to the NFC championship game, only to lose to the Saints after an infamous “too many men on the field” penalty, and “bountygate”—a system that incentivized Saints defensive players to try to knock opposing players out of the game. This was the year with the highest pass yards/attempt.

- Let us also not forget that Favre and Peterson have both endured their fair share of scandals, and I don't want to reminisce on their glory days without noting this.

Lets look at some other things:

```
vikes_table_4 <- vikes_data |>
  mutate(year = year(game_date)) |>
  group_by(year) |>
  summarize(
    avg_yards = mean(yards_gained, na.rm = TRUE),
    yards_sd = sd(yards_gained, na.rm = TRUE),
    avg_run_epa = mean(ifelse(play_type=="run", epa, NA), na.rm=TRUE),
    avg_pass_epa = mean(ifelse(play_type=="pass", epa, NA), na.rm=TRUE),
    avg_run_wpa = mean(ifelse(play_type=="run", wpa, NA), na.rm=TRUE),
    avg_pass_wpa = mean(ifelse(play_type=="pass", wpa, NA), na.rm=TRUE)
  )
vikes_table_4
```

A tibble: 10 x 7

	year	avg_yards	yards_sd	avg_run_epa	avg_pass_epa	avg_run_wpa	avg_pass_wpa
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	2009	3.94	8.08	-0.168	0.118	-0.00372	0.00282
2	2010	3.77	7.52	-0.0978	-0.0806	-0.000535	-0.00118
3	2011	3.94	7.95	0.0108	0.0373	0.00229	0.00244
4	2012	3.79	7.64	-0.0451	0.0145	0.000343	0.00138
5	2013	4.11	8.31	0.00710	0.0360	0.000662	0.00244
6	2014	3.79	7.47	0.00778	0.00918	0.00195	0.00178
7	2015	3.86	8.06	-0.0299	0.0298	-0.00137	0.00150
8	2016	3.60	7.23	-0.152	0.0166	-0.00288	-0.000410
9	2017	3.66	7.36	-0.112	0.0315	-0.00160	0.00127
10	2018	3.87	7.79	-0.0822	-0.0276	-0.00199	0.00158

```
# I want to visualize this so I'm going to pivot longer for EPA
vikes_table_4_long_ep<-vikes_table_4|>
  rename(
    run = avg_run_epa,
    pass = avg_pass_epa
  )|>
  pivot_longer(cols =c(run,pass),
               names_to = "play_type",
               values_to= "average_epa")
vikes_table_4_long_ep
```



```
# A tibble: 20 x 7
```

	year	avg_yards	yards_sd	avg_run_wpa	avg_pass_wpa	play_type	average_epa
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<dbl>
1	2009	3.94	8.08	-0.00372	0.00282	run	-0.168
2	2009	3.94	8.08	-0.00372	0.00282	pass	0.118
3	2010	3.77	7.52	-0.000535	-0.00118	run	-0.0978
4	2010	3.77	7.52	-0.000535	-0.00118	pass	-0.0806
5	2011	3.94	7.95	0.00229	0.00244	run	0.0108
6	2011	3.94	7.95	0.00229	0.00244	pass	0.0373
7	2012	3.79	7.64	0.000343	0.00138	run	-0.0451
8	2012	3.79	7.64	0.000343	0.00138	pass	0.0145
9	2013	4.11	8.31	0.000662	0.00244	run	0.00710
10	2013	4.11	8.31	0.000662	0.00244	pass	0.0360
11	2014	3.79	7.47	0.00195	0.00178	run	0.00778
12	2014	3.79	7.47	0.00195	0.00178	pass	0.00918
13	2015	3.86	8.06	-0.00137	0.00150	run	-0.0299
14	2015	3.86	8.06	-0.00137	0.00150	pass	0.0298
15	2016	3.60	7.23	-0.00288	-0.000410	run	-0.152
16	2016	3.60	7.23	-0.00288	-0.000410	pass	0.0166
17	2017	3.66	7.36	-0.00160	0.00127	run	-0.112
18	2017	3.66	7.36	-0.00160	0.00127	pass	0.0315
19	2018	3.87	7.79	-0.00199	0.00158	run	-0.0822
20	2018	3.87	7.79	-0.00199	0.00158	pass	-0.0276

```
## And pivot longer for WPA
vikes_table_4_long_wp<-vikes_table_4|>
  rename(
    run = avg_run_wpa,
    pass = avg_pass_wpa
  )|>
  pivot_longer(cols =c(run,pass),
               names_to = "play_type",
               values_to= "average_wpa")
vikes_table_4_long_wp
```

```
# A tibble: 20 x 7
```

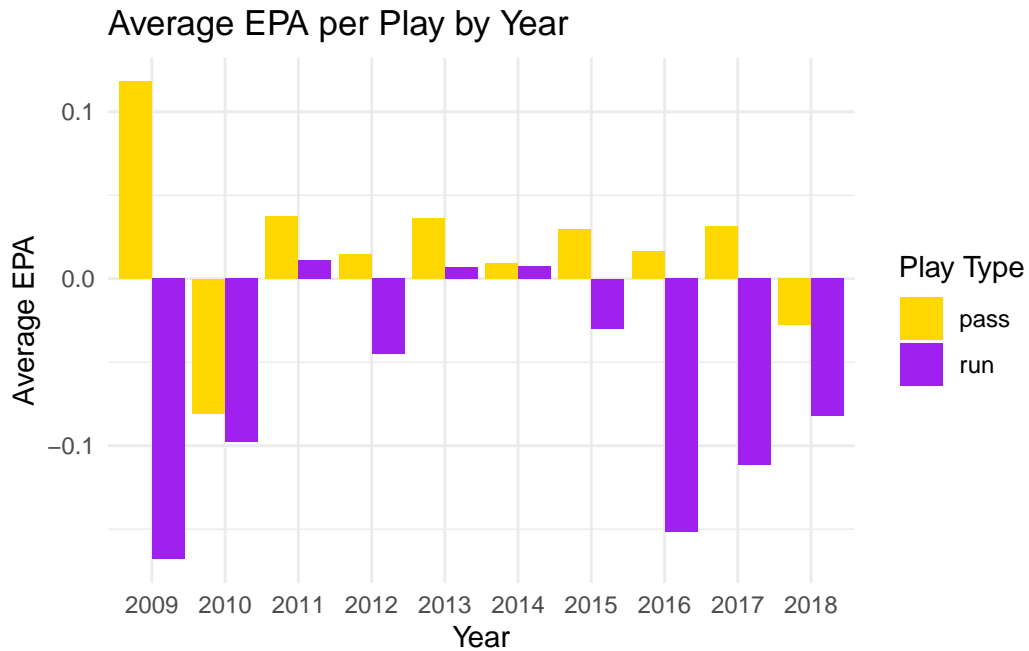
	year	avg_yards	yards_sd	avg_run_epa	avg_pass_epa	play_type	average_wpa
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<dbl>
1	2009	3.94	8.08	-0.168	0.118	run	-0.00372
2	2009	3.94	8.08	-0.168	0.118	pass	0.00282
3	2010	3.77	7.52	-0.0978	-0.0806	run	-0.000535

4	2010	3.77	7.52	-0.0978	-0.0806	pass	-0.00118
5	2011	3.94	7.95	0.0108	0.0373	run	0.00229
6	2011	3.94	7.95	0.0108	0.0373	pass	0.00244
7	2012	3.79	7.64	-0.0451	0.0145	run	0.000343
8	2012	3.79	7.64	-0.0451	0.0145	pass	0.00138
9	2013	4.11	8.31	0.00710	0.0360	run	0.000662
10	2013	4.11	8.31	0.00710	0.0360	pass	0.00244
11	2014	3.79	7.47	0.00778	0.00918	run	0.00195
12	2014	3.79	7.47	0.00778	0.00918	pass	0.00178
13	2015	3.86	8.06	-0.0299	0.0298	run	-0.00137
14	2015	3.86	8.06	-0.0299	0.0298	pass	0.00150
15	2016	3.60	7.23	-0.152	0.0166	run	-0.00288
16	2016	3.60	7.23	-0.152	0.0166	pass	-0.000410
17	2017	3.66	7.36	-0.112	0.0315	run	-0.00160
18	2017	3.66	7.36	-0.112	0.0315	pass	0.00127
19	2018	3.87	7.79	-0.0822	-0.0276	run	-0.00199
20	2018	3.87	7.79	-0.0822	-0.0276	pass	0.00158

```

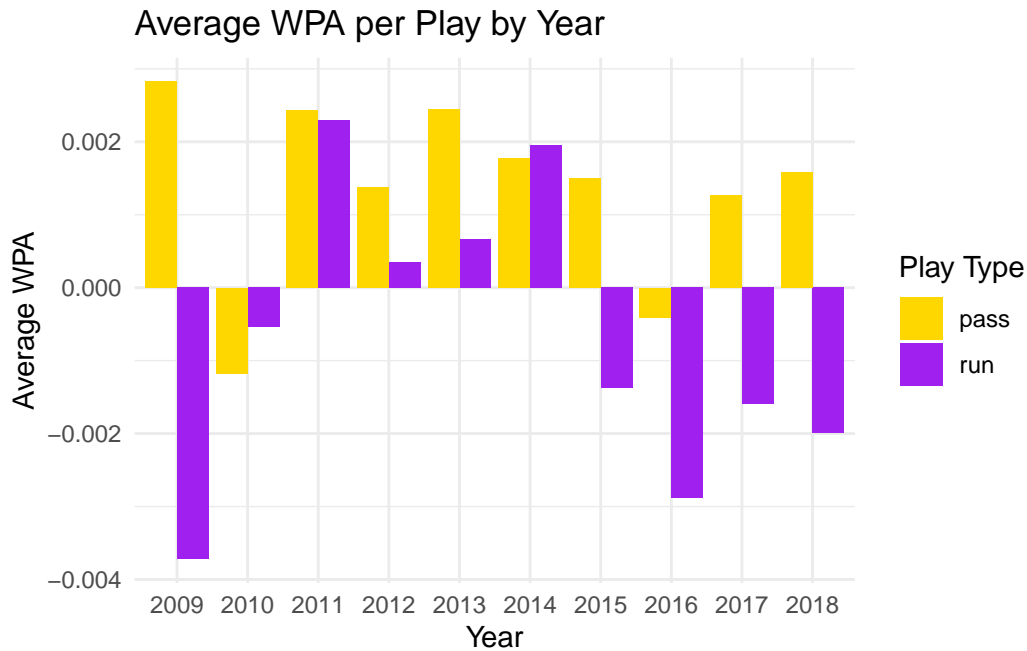
ggplot(vikes_table_4_long_ep, aes(x = factor(year), y = average_epa, fill = play_type)) +
  geom_col(position = "dodge") +
  labs(
    title = "Average EPA per Play by Year",
    x = "Year",
    y = "Average EPA",
    fill = "Play Type"
  ) +
  scale_fill_manual(
    values = c("run" = "purple",
              "pass" = "gold")
  ) +
  theme_minimal()

```



- Similar trends are visible here. Note that the 2010 Vikings were 6-10.

```
ggplot(vikes_table_4_long_wp, aes(x = factor(year), y = average_wpa, fill = play_type)) +
  geom_col(position = "dodge") +
  labs(
    title = "Average WPA per Play by Year",
    x = "Year",
    y = "Average WPA",
    fill = "Play Type"
  ) +
  scale_fill_manual(
    values = c("run" = "purple",
              "pass" = "gold")
  ) +
  theme_minimal()
```



- This is also a fascinating breakdown. Note that in 2009, running was not advantageous at all, although Adrian Peterson did run for 1300+ yards that year.

Let's try to visualize some other things. I want to try do a heatmap relating EPA and field position.

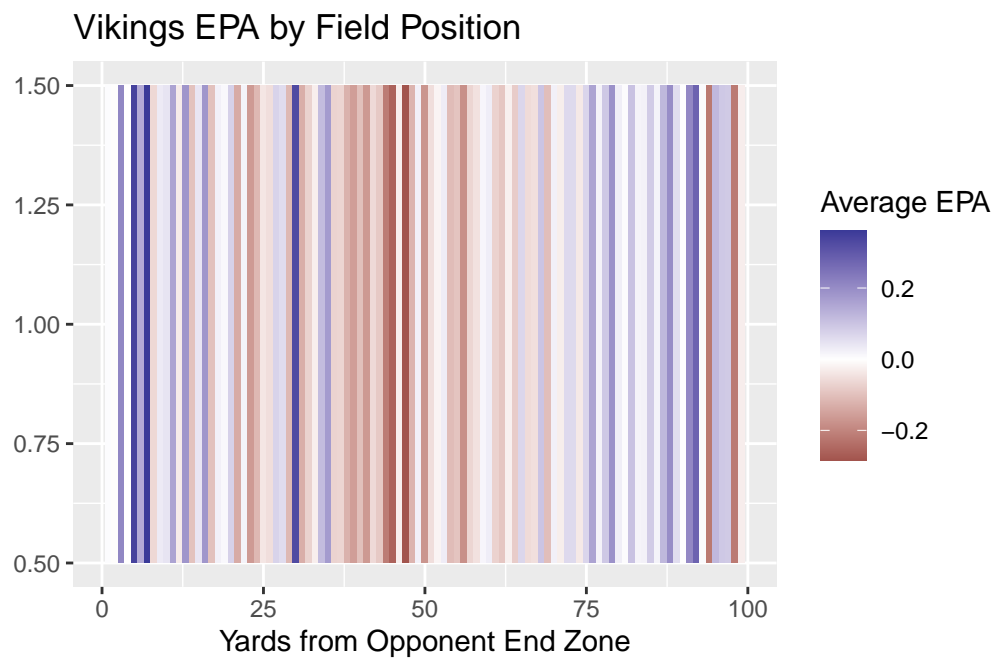
```
epa_by_field<-vikes_data|>
  group_by(yardline_100)|>
  summarize(avg_epa = mean(epa, na.rm=TRUE))
epa_by_field
```

```
# A tibble: 100 x 2
  yardline_100 avg_epa
    <dbl>      <dbl>
1         1  0.00506
2         2 -0.00389
3         3  0.211
4         4 -0.00103
5         5  0.344
6         6  0.146
7         7  0.362
8         8 -0.0618
9         9  0.0346
```

```
10          10  0.0460
# i 90 more rows
```

```
ggplot(epa_by_field, aes(x=yardline_100, y=1, fill = avg_epa))+
  geom_tile()+
  scale_fill_gradient2()+
  labs(
    title = "Vikings EPA by Field Position",
    x = "Yards from Opponent End Zone",
    y = "",
    fill = "Average EPA"
  )
```

Warning: Removed 1 row containing missing values or values outside the scale range (`geom_tile()`).



I would like to facet this by play type.

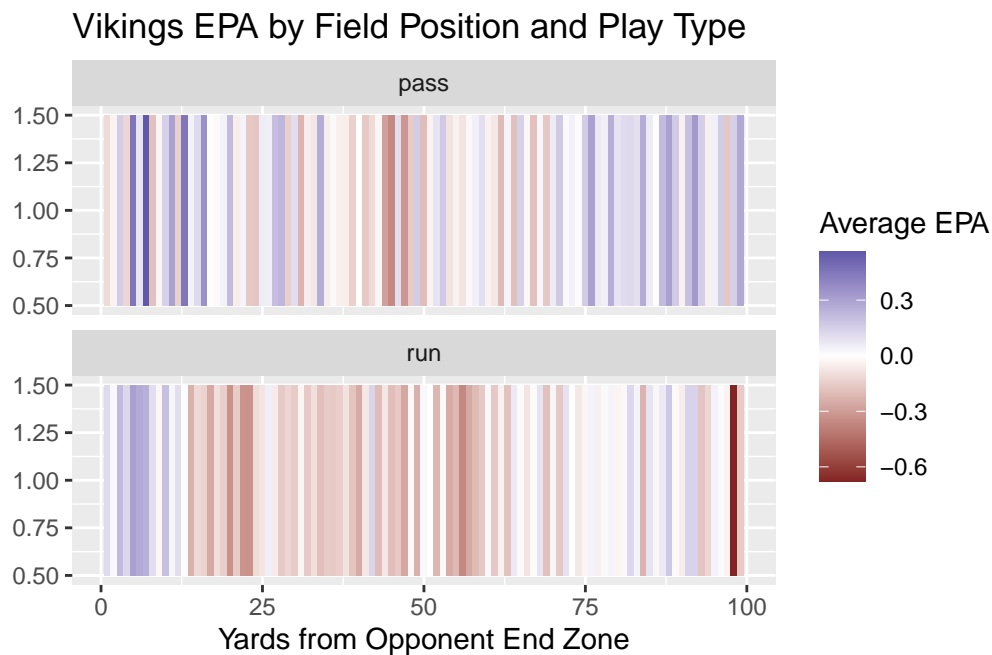
```
epa_by_field_type<-vikes_data|>
  filter(play_type %in% c("run","pass"))|>
```

```

group_by(yardline_100,play_type)|>
summarize(avg_epa = mean(epa, na.rm = TRUE), .groups = "drop")

ggplot(epa_by_field_type, aes(x=yardline_100, y=1, fill=avg_epa))+
  geom_tile()+
  scale_fill_gradient2()+
  facet_wrap(~play_type, ncol = 1) +
  labs(
    title = "Vikings EPA by Field Position and Play Type",
    x = "Yards from Opponent End Zone",
    y = "",
    fill = "Average EPA"
  )

```



Seems like passing generally has a higher EPA from just about anywhere in the field.