

# 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.

```
## Getting to know the data
dim(vikes_data)
```

```
[1] 27612    255
```

```
vikes_data<-vikes_data|>
  filter(!is.na(posteam) & posteam=="MIN") # I'm going to get only the plays where the vik
```

## 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	9	0	250	1991	0	67	9	2340
2	0	6	0	193	1870	0	34	1	1415
3	0	10	0	189	1642	0	25	1	419
4	0	289	0	57	94	726	1	0	47

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	419	547	0.434	0.566

2	2010	439	540	0.448	0.552
3	2011	440	548	0.445	0.555
4	2012	499	560	0.471	0.529
5	2013	415	590	0.413	0.587
6	2014	402	563	0.417	0.583
7	2015	435	476	0.477	0.523
8	2016	369	616	0.375	0.625
9	2017	505	589	0.462	0.538
10	2018	305	581	0.344	0.656

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     4.28     8.39
2  2010     3.98     7.54
3  2011     3.90     8.14
4  2012     3.99     8.20
5  2013     4.04     8.36
6  2014     3.79     7.70
7  2015     3.92     8.29
8  2016     3.69     7.16
9  2017     4.07     7.83
10 2018     4.22     7.83
```

Cool! Vikings were averaging a high 4.3 yards per play in 2009. The Vikings were 12-4.

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)

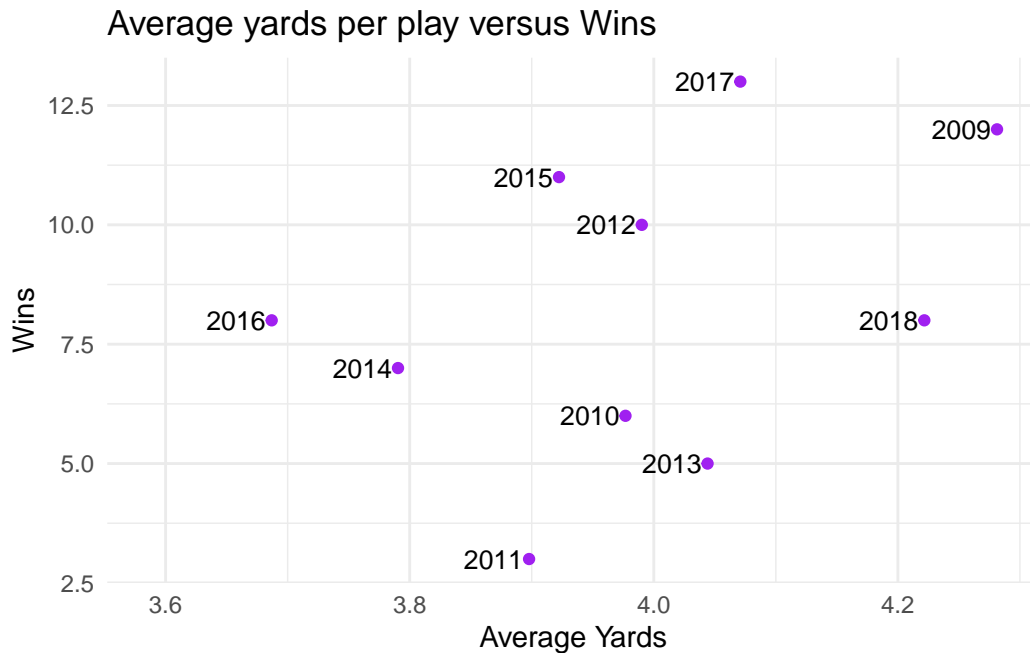
vikes_table_2$wins=wins

vikes_table_2
```

```
# A tibble: 10 x 4
  year avg_yards yards_sd wins
  <dbl>   <dbl>   <dbl> <dbl>
1  2009     4.28     8.39    12
2  2010     3.98     7.54     6
3  2011     3.90     8.14     3
4  2012     3.99     8.20    10
5  2013     4.04     8.36     5
6  2014     3.79     7.70     7
7  2015     3.92     8.29    11
8  2016     3.69     7.16     8
9  2017     4.07     7.83    13
10 2018     4.22     7.83     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 seems to be a loose positive trend between average yards per play and record.

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	4.28	8.39	4.31	6.94
2	2010	3.98	7.54	4.57	6.15
3	2011	3.90	8.14	5.27	5.25
4	2012	3.99	8.20	5.46	5.32

5	2013	4.04	8.36	5.03	5.81
6	2014	3.79	7.70	4.53	5.78
7	2015	3.92	8.29	4.78	5.96
8	2016	3.69	7.16	3.38	5.97
9	2017	4.07	7.83	4.18	6.80
10	2018	4.22	7.83	4.38	6.37

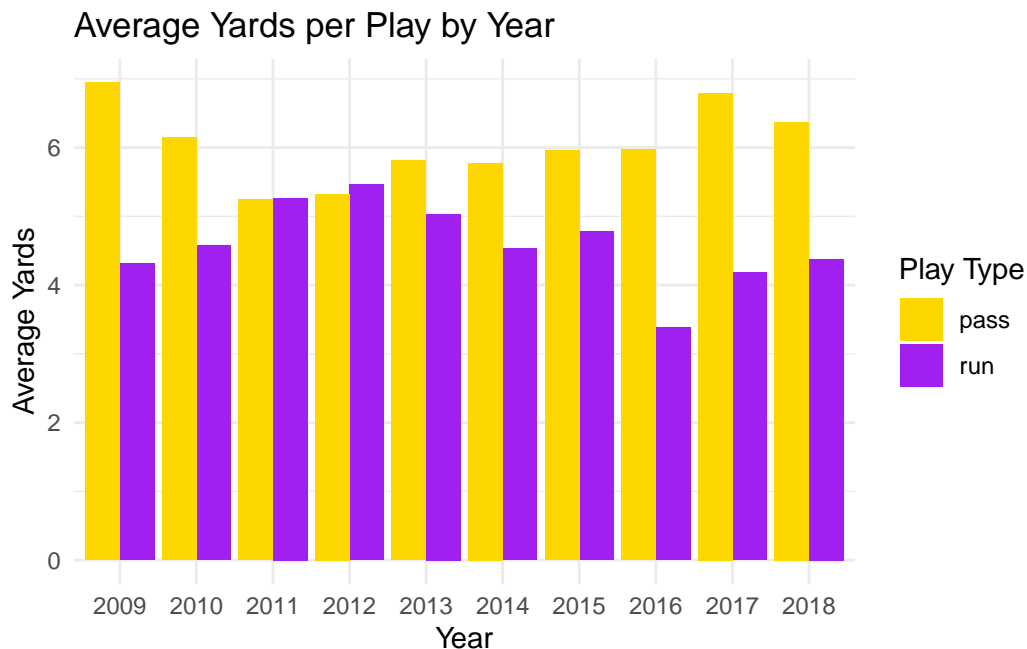
```
# I want to visualize this so I'm going to pivot longer
vikes_table_3_long<-vikes_table_3|>
  rename(
    run = avg_run_yrds,
    pass = avg_pass_yrds
  )|>
  pivot_longer(cols =c(run,pass),
               names_to = "play_type",
               values_to= "average_yards")
vikes_table_3_long
```

```
# A tibble: 20 x 5
   year avg_yards yards_sd play_type average_yards
  <dbl>   <dbl>   <dbl> <chr>         <dbl>
1  2009     4.28     8.39 run           4.31
2  2009     4.28     8.39 pass          6.94
3  2010     3.98     7.54 run           4.57
4  2010     3.98     7.54 pass          6.15
5  2011     3.90     8.14 run           5.27
6  2011     3.90     8.14 pass          5.25
7  2012     3.99     8.20 run           5.46
8  2012     3.99     8.20 pass          5.32
9  2013     4.04     8.36 run           5.03
10 2013     4.04     8.36 pass          5.81
11 2014     3.79     7.70 run           4.53
12 2014     3.79     7.70 pass          5.78
13 2015     3.92     8.29 run           4.78
14 2015     3.92     8.29 pass          5.96
15 2016     3.69     7.16 run           3.38
16 2016     3.69     7.16 pass          5.97
17 2017     4.07     7.83 run           4.18
18 2017     4.07     7.83 pass          6.80
19 2018     4.22     7.83 run           4.38
20 2018     4.22     7.83 pass          6.37
```

```

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 their complicated legacies.

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	4.28	8.39	-0.153	0.213	-0.00353	0.00506
2	2010	3.98	7.54	-0.0899	-0.112	-0.00102	-0.00123
3	2011	3.90	8.14	0.121	-0.128	0.00513	-0.00354
4	2012	3.99	8.20	0.0183	-0.0450	0.00233	-0.000512
5	2013	4.04	8.36	0.0405	-0.0944	0.00316	-0.000187
6	2014	3.79	7.70	0.0171	-0.0520	0.00239	-0.00109
7	2015	3.92	8.29	0.0219	0.0109	-0.000891	0.00203
8	2016	3.69	7.16	-0.211	0.0816	-0.00425	0.000850
9	2017	4.07	7.83	-0.0909	0.182	-0.000631	0.00547
10	2018	4.22	7.83	-0.146	0.0404	-0.00437	0.00274

# 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	4.28	8.39	-0.00353	0.00506	run	-0.153
2	2009	4.28	8.39	-0.00353	0.00506	pass	0.213
3	2010	3.98	7.54	-0.00102	-0.00123	run	-0.0899
4	2010	3.98	7.54	-0.00102	-0.00123	pass	-0.112
5	2011	3.90	8.14	0.00513	-0.00354	run	0.121
6	2011	3.90	8.14	0.00513	-0.00354	pass	-0.128
7	2012	3.99	8.20	0.00233	-0.000512	run	0.0183
8	2012	3.99	8.20	0.00233	-0.000512	pass	-0.0450
9	2013	4.04	8.36	0.00316	-0.000187	run	0.0405
10	2013	4.04	8.36	0.00316	-0.000187	pass	-0.0944
11	2014	3.79	7.70	0.00239	-0.00109	run	0.0171
12	2014	3.79	7.70	0.00239	-0.00109	pass	-0.0520
13	2015	3.92	8.29	-0.000891	0.00203	run	0.0219
14	2015	3.92	8.29	-0.000891	0.00203	pass	0.0109
15	2016	3.69	7.16	-0.00425	0.000850	run	-0.211
16	2016	3.69	7.16	-0.00425	0.000850	pass	0.0816
17	2017	4.07	7.83	-0.000631	0.00547	run	-0.0909
18	2017	4.07	7.83	-0.000631	0.00547	pass	0.182
19	2018	4.22	7.83	-0.00437	0.00274	run	-0.146
20	2018	4.22	7.83	-0.00437	0.00274	pass	0.0404

```
## 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
```

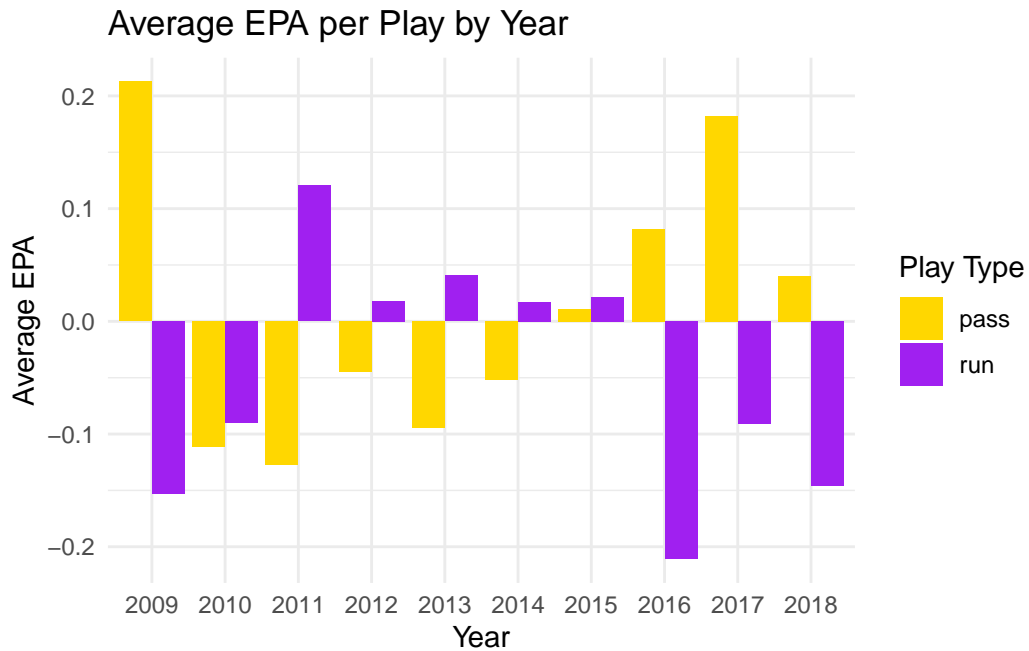
	year	avg_yards	yards_sd	avg_run_epa	avg_pass_epa	play_type	average_wpa
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<dbl>

1	2009	4.28	8.39	-0.153	0.213	run	-0.00353
2	2009	4.28	8.39	-0.153	0.213	pass	0.00506
3	2010	3.98	7.54	-0.0899	-0.112	run	-0.00102
4	2010	3.98	7.54	-0.0899	-0.112	pass	-0.00123
5	2011	3.90	8.14	0.121	-0.128	run	0.00513
6	2011	3.90	8.14	0.121	-0.128	pass	-0.00354
7	2012	3.99	8.20	0.0183	-0.0450	run	0.00233
8	2012	3.99	8.20	0.0183	-0.0450	pass	-0.000512
9	2013	4.04	8.36	0.0405	-0.0944	run	0.00316
10	2013	4.04	8.36	0.0405	-0.0944	pass	-0.000187
11	2014	3.79	7.70	0.0171	-0.0520	run	0.00239
12	2014	3.79	7.70	0.0171	-0.0520	pass	-0.00109
13	2015	3.92	8.29	0.0219	0.0109	run	-0.000891
14	2015	3.92	8.29	0.0219	0.0109	pass	0.00203
15	2016	3.69	7.16	-0.211	0.0816	run	-0.00425
16	2016	3.69	7.16	-0.211	0.0816	pass	0.000850
17	2017	4.07	7.83	-0.0909	0.182	run	-0.000631
18	2017	4.07	7.83	-0.0909	0.182	pass	0.00547
19	2018	4.22	7.83	-0.146	0.0404	run	-0.00437
20	2018	4.22	7.83	-0.146	0.0404	pass	0.00274

```

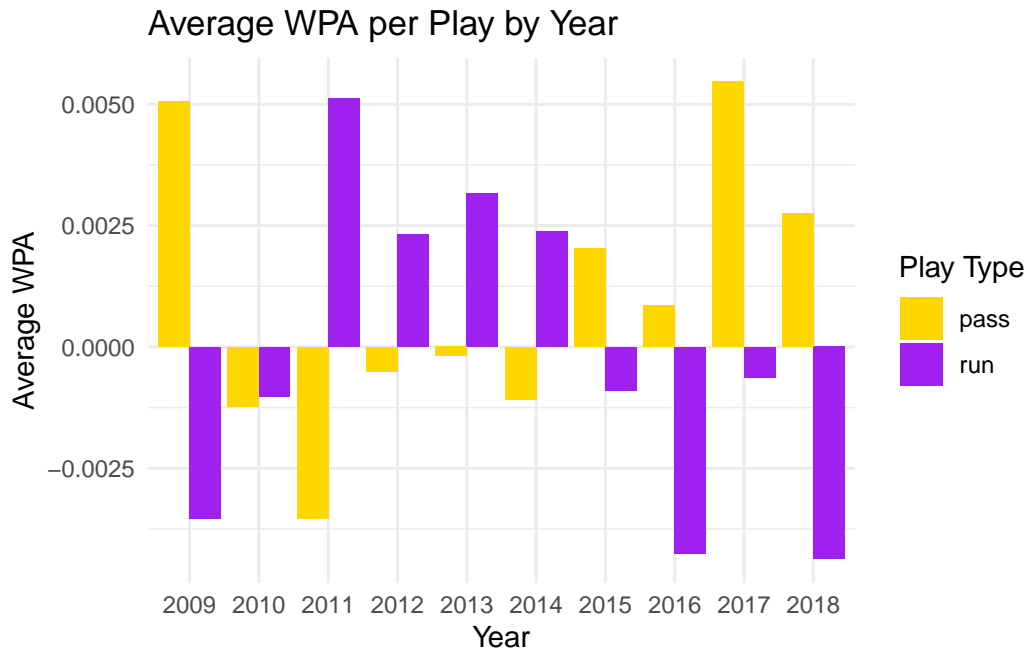
ggplot(vikes_table_4_long_ep, aes(x = factor(year), y = average_epa, fill = play_type)) +
  geom_col(position = "dodge") +
  labs(
    title = "Average EPA per Play by Year",
    x = "Year",
    y = "Average EPA",
    fill = "Play Type"
  ) +
  scale_fill_manual(
    values = c("run" = "purple",
               "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: 99 x 2
  yardline_100 avg_epa
    <dbl>     <dbl>
1         1  0.133
2         2 -0.0574
3         3  0.256
4         4  0.129
5         5  0.396
6         6  0.0666
7         7  0.367
8         8 -0.0145
```

```

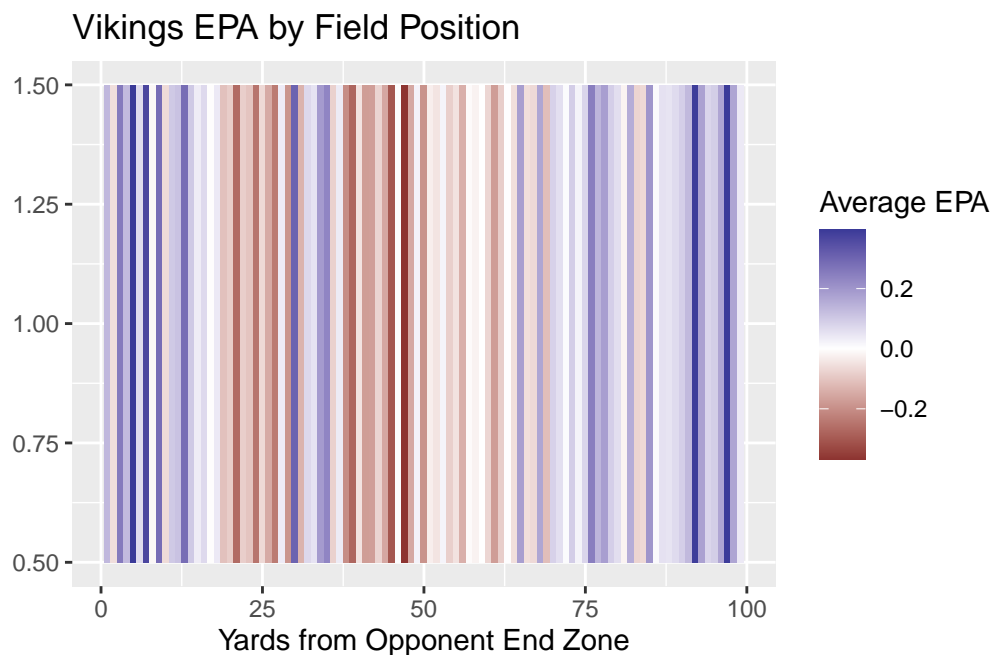
9          9  0.284
10         10 -0.0633
# i 89 more rows

```

```

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"
  )

```



I would like to facet this by play type.

```

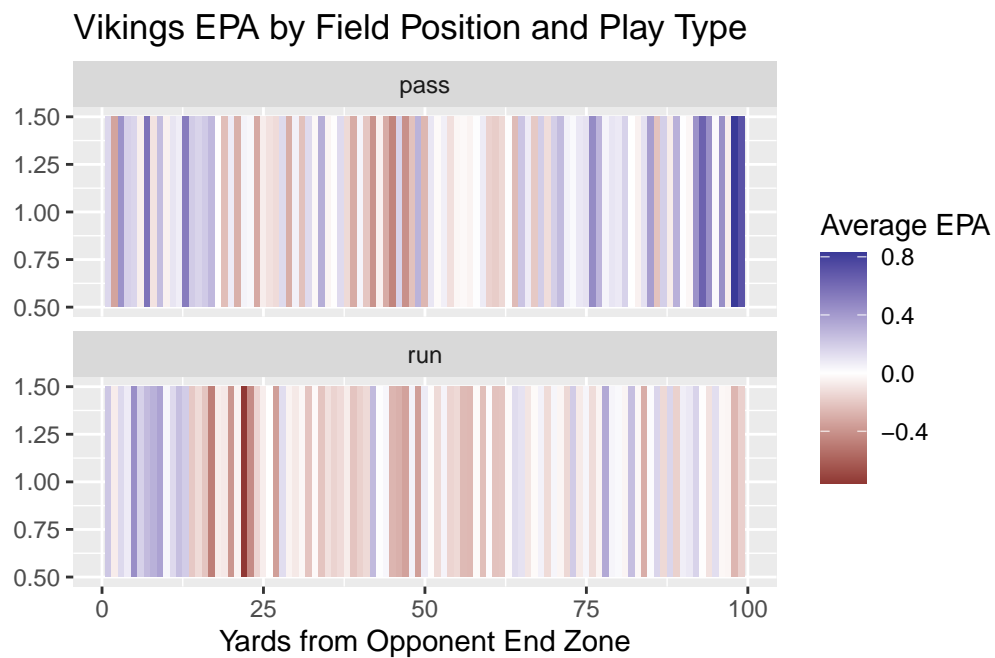
epa_by_field_type<-vikes_data|>
  filter(play_type %in% c("run","pass"))|>
  group_by(yardline_100,play_type)|>
  summarize(avg_epa = mean(epa, na.rm = TRUE), .groups = "drop")

```

```

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.