

JaylenBrownMod1_Final

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2024-07-23

```
# Clear environment
rm(list = ls())

# Load necessary Libraries
library(tidymodels)
```

```
## Warning: package 'tidymodels' was built under R version 4.3.3
```

```
## — Attaching packages ————— tidymodels 1.2.0 —
```

```
## ✓ broom      1.0.5    ✓ recipes      1.0.10
## ✓ dials      1.2.1    ✓ rsample      1.2.1
## ✓ dplyr      1.1.4    ✓ tibble       3.2.1
## ✓ ggplot2    3.5.0    ✓ tidyr        1.3.1
## ✓ infer      1.0.7    ✓ tune         1.2.0
## ✓ modeldata  1.3.0    ✓ workflows    1.1.4
## ✓ parsnip    1.2.1    ✓ workflowsets 1.1.0
## ✓ purrr      1.0.2    ✓ yardstick    1.3.1
```

```
## Warning: package 'dials' was built under R version 4.3.3
```

```
## Warning: package 'ggplot2' was built under R version 4.3.3
```

```
## Warning: package 'infer' was built under R version 4.3.3
```

```
## Warning: package 'modeldata' was built under R version 4.3.3
```

```
## Warning: package 'parsnip' was built under R version 4.3.3
```

```
## Warning: package 'recipes' was built under R version 4.3.3
```

```
## Warning: package 'rsample' was built under R version 4.3.3
```

```
## Warning: package 'tidyr' was built under R version 4.3.3
```

```
## Warning: package 'tune' was built under R version 4.3.3
```

```
## Warning: package 'workflows' was built under R version 4.3.3
```

```
## Warning: package 'workflowsets' was built under R version 4.3.3
```

```
## Warning: package 'yardstick' was built under R version 4.3.3
```

```
## — Conflicts ————— tidymodels_conflicts() —
## X purrr::discard() masks scales::discard()
## X dplyr::filter()  masks stats::filter()
## X dplyr::lag()     masks stats::lag()
## X recipes::step() masks stats::step()
## • Dig deeper into tidy modeling with R at https://www.tmr.org
```

```
library(tidytext)
library(tidyverse)
```

```
## — Attaching core tidyverse packages ————— tidyverse 2.0.0 —
## ✓ forcats 1.0.0      ✓ readr 2.1.5
## ✓ lubridate 1.9.3    ✓ stringr 1.5.1
```

```
## — Conflicts ————— tidyverse_conflicts() —
## X readr::col_factor() masks scales::col_factor()
## X purrr::discard()    masks scales::discard()
## X dplyr::filter()     masks stats::filter()
## X stringr::fixed()    masks recipes::fixed()
## X dplyr::lag()        masks stats::lag()
## X readr::spec()       masks yardstick::spec()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(plotly)
```

```
## Warning: package 'plotly' was built under R version 4.3.3
```

```
##
## Attaching package: 'plotly'
##
## The following object is masked from 'package:ggplot2':
##
##   last_plot
##
## The following object is masked from 'package:stats':
##
##   filter
##
## The following object is masked from 'package:graphics':
##
##   layout
```

```
library(scales)
library(ranger)
```

```
## Warning: package 'ranger' was built under R version 4.3.3
```

```
library(zoo)
```

```
## Warning: package 'zoo' was built under R version 4.3.3
```

```
##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric
```

```
library(ggcorrplot)
```

```
## Warning: package 'ggcorrplot' was built under R version 4.3.3
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 4.3.3
```

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

```
## Warning: package 'carData' was built under R version 4.3.3
```

```
##
## Attaching package: 'car'
##
## The following object is masked from 'package:purrr':
##
##     some
##
## The following object is masked from 'package:dplyr':
##
##     recode
```

```
library(optimx)
```

```
## Warning: package 'optimx' was built under R version 4.3.3
```

```
# Load dataset
jb <- read_csv("https://raw.githubusercontent.com/maliknyc/NBA-Prediction-Testing/main/JaylenBrownTest.csv")
```

```
## New names:
## Rows: 102 Columns: 35
## — Column specification
## _____ Delimiter: "," chr
## (6): Date, Age, Tm, ...6, Opp, ...8 dbl (28): Rk, G, GS, FG, FGA, FG%, 3P, 3PA,
## 3P%, FT, FTA, FT%, ORB, DRB, TR... time (1): MP
## 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.
## • `` -> `...6`
## • `` -> `...8`
```

```

# Preprocess the dataset
jb_cleaned <- jb %>%
  rename(TPA = `3PA`, TPP = `3P%`) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y")) %>%
  arrange(Date) %>%
  select(Date, Opp, MP, FGA, TPA, TPP, TRB, AST, STL, PTS, PRA, PR, PA, RA, SB)

# Add average points vs team
avg_points_vs_team <- jb %>%
  group_by(Opp) %>%
  summarize(avgPTS_vteam = mean(PTS)) %>%
  ungroup()

jb_cleaned <- jb_cleaned %>%
  left_join(avg_points_vs_team, by = "Opp")

# Add defensive ratings
def_ratings <- data.frame(
  Opp = c("BOS", "DEN", "OKC", "MIN", "LAC", "DAL", "NYK", "MIL", "NOP", "PHO", "CLE", "IND", "L
AL", "ORL", "PHI", "GSW", "MIA", "SAC", "HOU", "CHI", "ATL", "BRK", "UTA", "MEM", "TOR", "SAS",
"CHO", "POR", "WAS", "DET"),
  DEF_RTG = c(110.6, 112.3, 111.0, 110.4, 113.1, 114.9, 111.4, 110.2, 111.9, 113.7, 115.0, 117.
6, 114.8, 110.8, 113.0, 114.5, 111.5, 114.4, 112.8, 115.7, 115.4, 114.6, 119.6, 113.7, 115.6, 11
6.1, 116.9, 118.0, 118.6, 118.0)
)

jb_cleaned <- jb_cleaned %>%
  left_join(def_ratings, by = "Opp") %>%
  drop_na()

# Calculate days of rest and cap at 30 days
jb_cleaned <- jb_cleaned %>%
  mutate(days_rest = as.numeric(difftime(Date, lag(Date), units = 'days')),
    days_rest = ifelse(days_rest > 30, 30, days_rest)) %>%
  drop_na()

# Calculate moving averages and other features
jb_ra <- jb_cleaned %>%
  mutate(Opp = as.factor(Opp), over_21_5 = ifelse(PTS > 21.5, 1, 0),
    avgPTS_5 = rollapply(PTS, width = 5, FUN = mean, fill = NA, align = "right"),
    avgPTS_10 = rollapply(PTS, width = 10, FUN = mean, fill = NA, align = "right"),
    avgTPA_5 = rollapply(TPA, width = 5, FUN = mean, fill = NA, align = "right"),
    trendPTS = rollapply(PTS, width = 5, FUN = function(x) coef(lm(x ~ seq_along(x)))[2],
fill = NA, align = "right"),
    perc_over = rollapply(over_21_5, width = 10, FUN = mean, fill = NA, align = "right") * 1
00) %>%
  drop_na()

# Select relevant variables for the model
jb_vars <- jb_ra %>%
  mutate(intPTS5_TPA5 = avgPTS_5 * avgTPA_5,
    comp_opp_25_75 = 0.25 * DEF_RTG + 0.75 * avgPTS_vteam) %>%

```

```
select(Date, Opp, avgPTS_vteam, days_rest, avg_PTS_5, avg_TPA_5, DEF_RTG, over_21_5, comp_opp_
25_75, trend_PTS, int_PTS5_TPA5, perc_over)

# Fit logistic regression model
form.vars_cleaned1 <- "over_21_5 ~ comp_opp_25_75 + avg_PTS_5 + trend_PTS + perc_over"
glm_cleaned <- glm(as.formula(form.vars_cleaned1), jb_vars, family = binomial)

# Add predicted probabilities to jb_vars
jb_vars <- jb_vars %>%
  mutate(predicted_prob = predict(glm_cleaned, newdata = jb_vars, type = "response"))

# View the updated dataframe with the new column
jb_vars %>%
  select(Date, Opp, over_21_5, predicted_prob) %>%
  print(n = 90)
```

A tibble: 92 × 4

##	Date	Opp	over_21_5	predicted_prob
##	<date>	<fct>	<dbl>	<dbl>
## 1	2023-05-09	PHI	1	0.842
## 2	2023-05-11	PHI	0	0.518
## 3	2023-05-14	PHI	1	0.827
## 4	2023-05-17	MIA	1	0.775
## 5	2023-05-19	MIA	0	0.382
## 6	2023-05-21	MIA	0	0.0857
## 7	2023-05-23	MIA	0	0.0330
## 8	2023-05-25	MIA	0	0.111
## 9	2023-05-27	MIA	1	0.646
## 10	2023-05-29	MIA	0	0.468
## 11	2023-10-25	NYK	0	0.0190
## 12	2023-10-27	MIA	1	0.165
## 13	2023-10-30	WAS	1	0.968
## 14	2023-11-01	IND	0	0.880
## 15	2023-11-04	BRK	1	0.711
## 16	2023-11-06	MIN	1	0.754
## 17	2023-11-08	PHI	0	0.0362
## 18	2023-11-10	BRK	1	0.753
## 19	2023-11-11	TOR	1	0.951
## 20	2023-11-13	NYK	1	0.774
## 21	2023-11-17	TOR	1	0.981
## 22	2023-11-19	MEM	0	0.0204
## 23	2023-11-20	CHO	0	0.00512
## 24	2023-11-22	MIL	1	0.205
## 25	2023-11-24	ORL	0	0.325
## 26	2023-11-26	ATL	0	0.680
## 27	2023-11-28	CHI	1	0.961
## 28	2023-12-01	PHI	0	0.534
## 29	2023-12-04	IND	1	0.980
## 30	2023-12-08	NYK	0	0.194
## 31	2023-12-12	CLE	1	0.353
## 32	2023-12-14	CLE	1	0.568
## 33	2023-12-15	ORL	0	0.155
## 34	2023-12-17	ORL	1	0.830
## 35	2023-12-19	GSW	1	0.973
## 36	2023-12-20	SAC	1	0.991
## 37	2023-12-23	LAC	1	0.830
## 38	2023-12-25	LAL	0	0.158
## 39	2023-12-29	TOR	1	0.949
## 40	2023-12-31	SAS	1	0.914
## 41	2024-01-02	OKC	0	0.302
## 42	2024-01-05	UTA	0	0.0222
## 43	2024-01-06	IND	1	0.845
## 44	2024-01-08	IND	1	0.999
## 45	2024-01-10	MIN	1	1.00
## 46	2024-01-11	MIL	0	0.597
## 47	2024-01-13	HOU	1	0.638
## 48	2024-01-17	SAS	0	0.279
## 49	2024-01-19	DEN	0	0.123

## 50	2024-01-21	HOU	0	0.0455
## 51	2024-01-22	DAL	1	0.750
## 52	2024-01-25	MIA	0	0.485
## 53	2024-01-27	LAC	0	0.0231
## 54	2024-01-29	NOP	1	0.0556
## 55	2024-01-30	IND	1	0.413
## 56	2024-02-01	LAL	0	0.0139
## 57	2024-02-07	ATL	0	0.0579
## 58	2024-02-09	WAS	0	0.0568
## 59	2024-02-11	MIA	0	0.0463
## 60	2024-02-13	BRK	0	0.351
## 61	2024-02-22	CHI	0	0.295
## 62	2024-02-24	NYK	1	0.530
## 63	2024-02-27	PHI	1	0.939
## 64	2024-03-01	DAL	1	0.974
## 65	2024-03-03	GSW	1	0.960
## 66	2024-03-05	CLE	0	0.441
## 67	2024-03-07	DEN	1	0.987
## 68	2024-03-09	PHO	1	0.996
## 69	2024-03-11	POR	1	0.986
## 70	2024-03-14	PHO	1	0.999
## 71	2024-03-18	DET	1	0.998
## 72	2024-03-20	MIL	0	0.850
## 73	2024-03-22	DET	1	0.996
## 74	2024-03-25	ATL	1	0.884
## 75	2024-03-28	ATL	0	0.600
## 76	2024-03-30	NOP	0	0.182
## 77	2024-04-03	OKC	1	0.138
## 78	2024-04-07	POR	1	0.913
## 79	2024-04-09	MIL	0	0.193
## 80	2024-04-11	NYK	0	0.0518
## 81	2024-04-21	MIA	0	0.0213
## 82	2024-04-24	MIA	1	0.622
## 83	2024-04-27	MIA	1	0.832
## 84	2024-04-29	MIA	0	0.341
## 85	2024-05-01	MIA	1	0.433
## 86	2024-05-07	CLE	1	0.867
## 87	2024-05-09	CLE	0	0.758
## 88	2024-05-11	CLE	1	0.891
## 89	2024-05-13	CLE	1	0.873
## 90	2024-05-15	CLE	0	0.150
##	# i	2 more rows		


```

# Export the updated dataframe to CSV
write.csv(jb_vars %>% select(Date, Opp, over_21_5, predicted_prob), "predictions_jb.csv", row.names = FALSE)

# Plot predicted probabilities vs. comp_opp_25_75
plot_comp_opp_25_75 <- ggplot(jb_vars, aes(x = comp_opp_25_75, y = predicted_prob)) +
  geom_point(alpha = 0.5) +
  geom_smooth(method = "glm", method.args = list(family = "binomial"), se = FALSE, color = "blue") +
  labs(title = "Predicted Probability vs. comp_opp_25_75", x = "comp_opp_25_75", y = "Predicted Probability")

# Plot predicted probabilities vs. avg_PTS_5
plot_avg_PTS_5 <- ggplot(jb_vars, aes(x = avg_PTS_5, y = predicted_prob)) +
  geom_point(alpha = 0.5) +
  geom_smooth(method = "glm", method.args = list(family = "binomial"), se = FALSE, color = "blue") +
  labs(title = "Predicted Probability vs. avg_PTS_5", x = "avg_PTS_5", y = "Predicted Probability")

# Plot predicted probabilities vs. trend_PTS
plot_trend_PTS <- ggplot(jb_vars, aes(x = trend_PTS, y = predicted_prob)) +
  geom_point(alpha = 0.5) +
  geom_smooth(method = "glm", method.args = list(family = "binomial"), se = FALSE, color = "blue") +
  labs(title = "Predicted Probability vs. trend_PTS", x = "trend_PTS", y = "Predicted Probability")

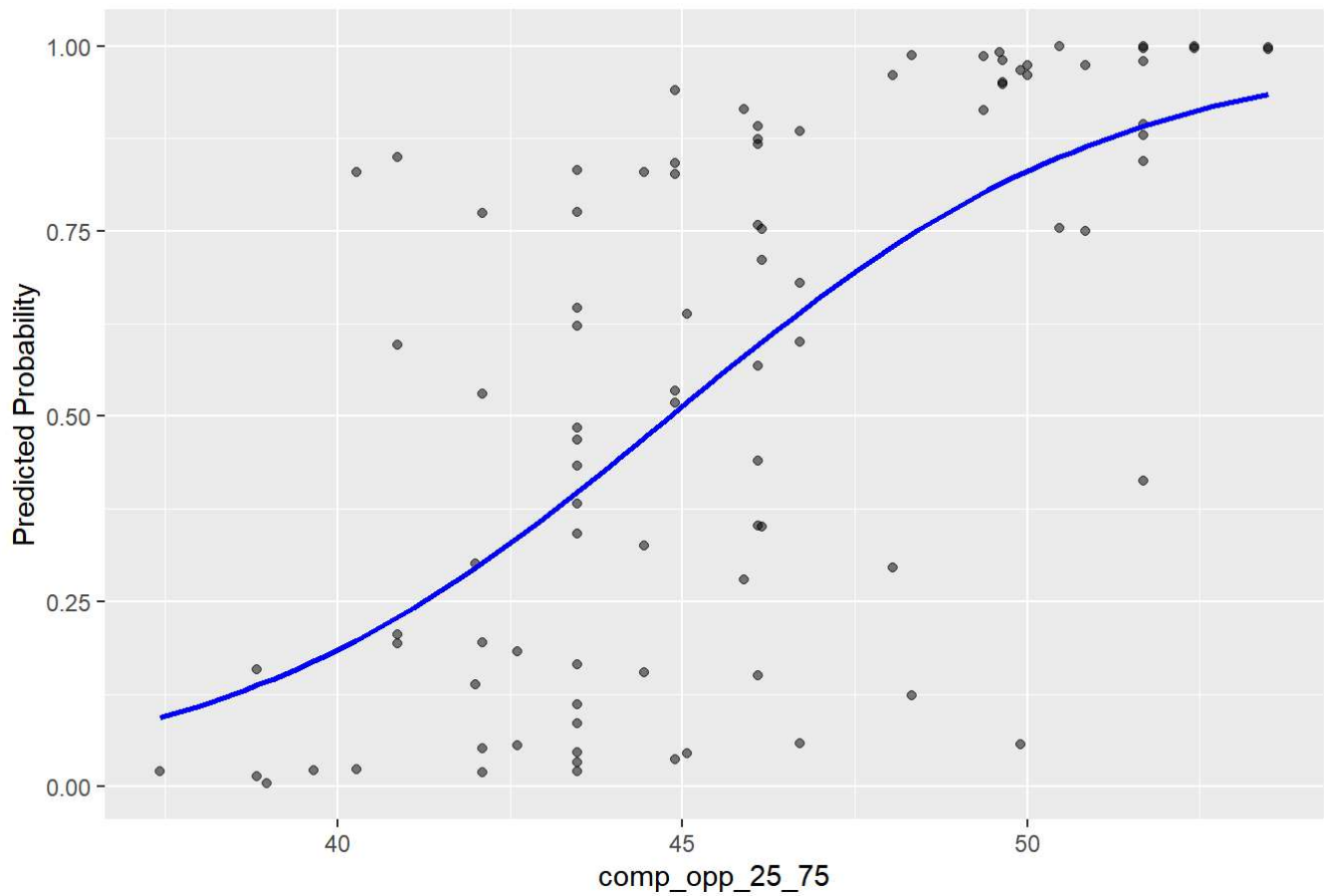
# Display the plots
print(plot_comp_opp_25_75)

```

```
## `geom_smooth()` using formula = 'y ~ x'
```

```
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

Predicted Probability vs. comp_opp_25_75

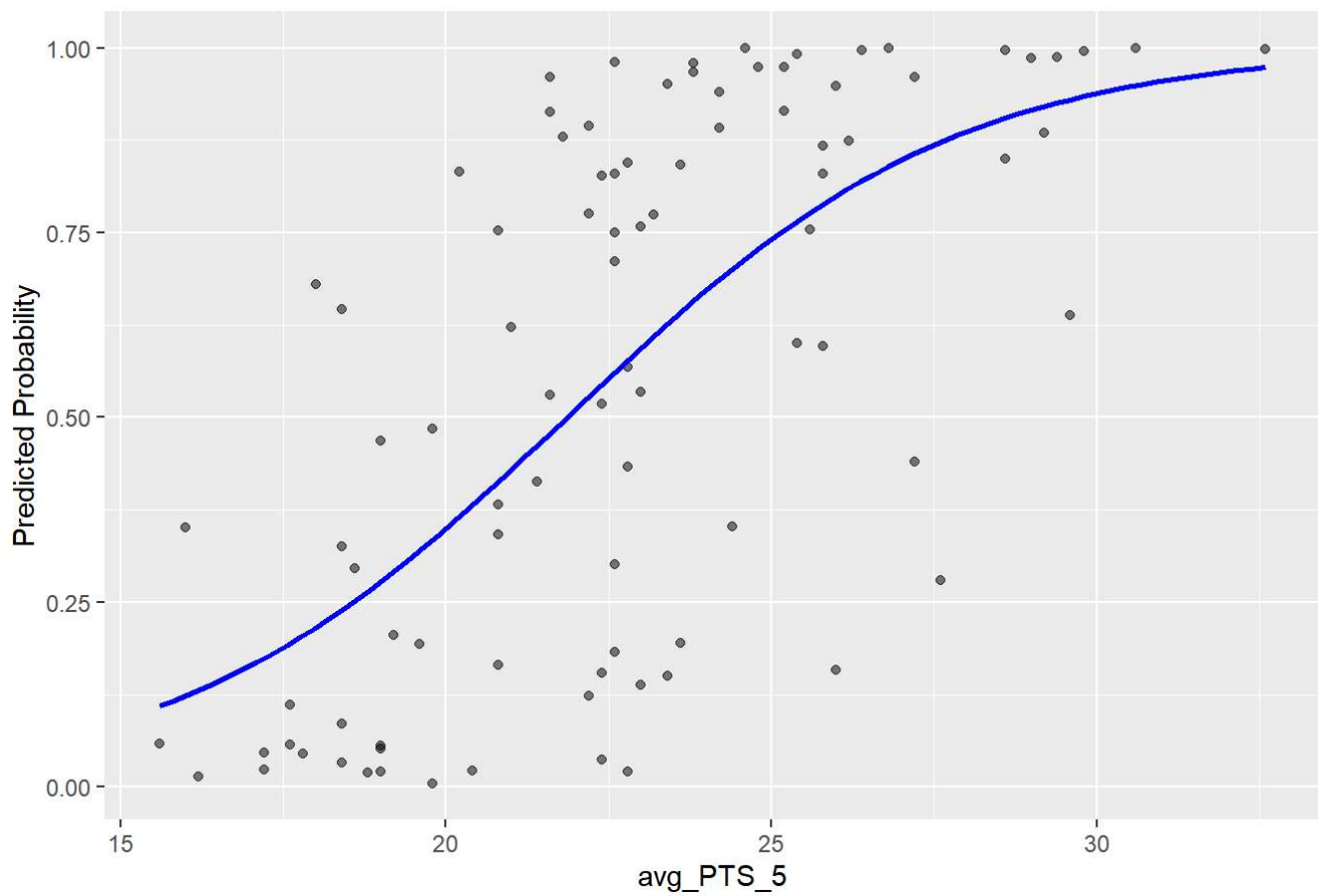


```
print(plot_avg_PTS_5)
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

```
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

Predicted Probability vs. avg_PTS_5

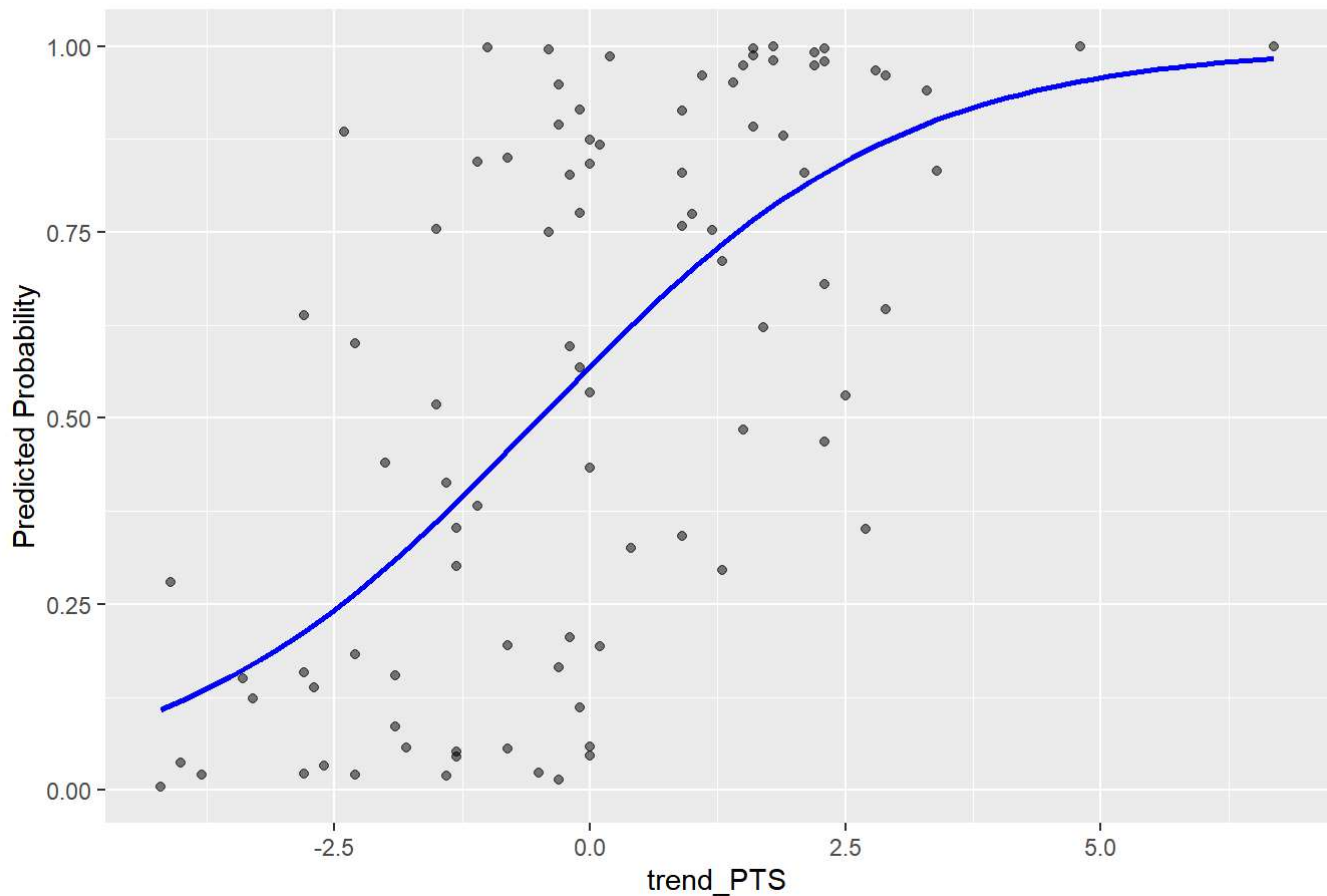


```
print(plot_trend_PTS)
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

```
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

Predicted Probability vs. trend_PTS



```
# Cross-validation
cvRes <- NULL
for(i in 1:1000) {
  inds <- sample(1:nrow(jb_vars), size = round(nrow(jb_vars)*.8), replace = F)
  train <- jb_vars %>% slice(inds)
  test <- jb_vars %>% slice(-inds)

  # Train the model on the training set
  model_clean <- glm(as.formula(form.vars_cleaned1), train, family = binomial(link = 'logit'))

  # Predict on the test set
  toEval_CV <- test %>%
    mutate(prob_over21_5 = predict(model_clean, newdata = test, type = 'response'),
           over_21_5 = factor(over_21_5, levels = c('1', '0')))

  cvRes <- cvRes %>%
    bind_rows(yardstick::roc_auc(toEval_CV, over_21_5, prob_over21_5) %>%
              mutate(cvInd = i))
}

cvRes %>%
  summarise(mean_auc = mean(.estimate))
```

```
## # A tibble: 1 × 1
##   mean_auc
##   <dbl>
## 1     0.894
```

```
# Final AUC
mean_auc <- cvRes %>%
  summarise(mean_auc = mean(.estimate)) %>%
  pull(mean_auc)

print(mean_auc) # AUC: 0.892
```

```
## [1] 0.8937853
```

```
# View updated dataset with predicted probabilities
jb_vars %>%
  select(Date, Opp, over_21_5, predicted_prob) %>%
  print(n = 90)
```

A tibble: 92 × 4

##	Date	Opp	over_21_5	predicted_prob
##	<date>	<fct>	<dbl>	<dbl>
## 1	2023-05-09	PHI	1	0.842
## 2	2023-05-11	PHI	0	0.518
## 3	2023-05-14	PHI	1	0.827
## 4	2023-05-17	MIA	1	0.775
## 5	2023-05-19	MIA	0	0.382
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## 7	2023-05-23	MIA	0	0.0330
## 8	2023-05-25	MIA	0	0.111
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## 15	2023-11-04	BRK	1	0.711
## 16	2023-11-06	MIN	1	0.754
## 17	2023-11-08	PHI	0	0.0362
## 18	2023-11-10	BRK	1	0.753
## 19	2023-11-11	TOR	1	0.951
## 20	2023-11-13	NYK	1	0.774
## 21	2023-11-17	TOR	1	0.981
## 22	2023-11-19	MEM	0	0.0204
## 23	2023-11-20	CHO	0	0.00512
## 24	2023-11-22	MIL	1	0.205
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## 31	2023-12-12	CLE	1	0.353
## 32	2023-12-14	CLE	1	0.568
## 33	2023-12-15	ORL	0	0.155
## 34	2023-12-17	ORL	1	0.830
## 35	2023-12-19	GSW	1	0.973
## 36	2023-12-20	SAC	1	0.991
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## 38	2023-12-25	LAL	0	0.158
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## 41	2024-01-02	OKC	0	0.302
## 42	2024-01-05	UTA	0	0.0222
## 43	2024-01-06	IND	1	0.845
## 44	2024-01-08	IND	1	0.999
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## 47	2024-01-13	HOU	1	0.638
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## 50	2024-01-21	HOU	0	0.0455
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## 53	2024-01-27	LAC	0	0.0231
## 54	2024-01-29	NOP	1	0.0556
## 55	2024-01-30	IND	1	0.413
## 56	2024-02-01	LAL	0	0.0139
## 57	2024-02-07	ATL	0	0.0579
## 58	2024-02-09	WAS	0	0.0568
## 59	2024-02-11	MIA	0	0.0463
## 60	2024-02-13	BRK	0	0.351
## 61	2024-02-22	CHI	0	0.295
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## 67	2024-03-07	DEN	1	0.987
## 68	2024-03-09	PHO	1	0.996
## 69	2024-03-11	POR	1	0.986
## 70	2024-03-14	PHO	1	0.999
## 71	2024-03-18	DET	1	0.998
## 72	2024-03-20	MIL	0	0.850
## 73	2024-03-22	DET	1	0.996
## 74	2024-03-25	ATL	1	0.884
## 75	2024-03-28	ATL	0	0.600
## 76	2024-03-30	NOP	0	0.182
## 77	2024-04-03	OKC	1	0.138
## 78	2024-04-07	POR	1	0.913
## 79	2024-04-09	MIL	0	0.193
## 80	2024-04-11	NYK	0	0.0518
## 81	2024-04-21	MIA	0	0.0213
## 82	2024-04-24	MIA	1	0.622
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## 87	2024-05-09	CLE	0	0.758
## 88	2024-05-11	CLE	1	0.891
## 89	2024-05-13	CLE	1	0.873
## 90	2024-05-15	CLE	0	0.150
##	# i	2 more rows		