# NFL Prediction: Analysis of Passing EPA based on first nine weeks\*

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This paper focuses on the later Passing EPA in 2023 NFL based on the first weeks league data.

## 1 Introduction

This paper explores the total expected points added on passing for quarterback position players. The quarterback position is a role that passes to a receiver from an offensive lineman and passes the ball to a running back, which means the quarterback position plays an essential role in winning a football game. In addition, a team without good quarterback players has less chance to win the football games in a league.

Therefore, the total expected points added on passing for quarterback players is one of the most important factors in evaluating each quarterback player. This study focuses on the correlation between "passing\_epa" and other related factors in the football game.

The data from the first nine weeks of the 2023 NFL season were used to predict the later 2023 season's total expected points added on passing.

In the following section, Section 2 explains the data used and the cleaning process, and Section 3 analyzes the built model for this study.

### 2 Data

The raw data used was retrieved from NFLverse(Sebastian Carl 2023). The data was cleaned by R(R Core Team 2023) and Tidyverse(citeTidyverse?), and analyzed by r. The raw data from Sebastian Carl (2023) contains all data from 1999. The cleaned data contains only player data

<sup>\*</sup>https://github.com/kakaomonk/nfl-prediction

Table 1: Cleaned NFL data

# A tibble: 646 x 11							
	player_id	season	week	completions	attempts	passing_yards	passing_tds
	<chr></chr>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1	00-0023459	2023	1	0	1	0	0
2	00-0026158	2023	13	23	44	254	2
3	00-0026158	2023	14	26	45	311	3
4	00-0026158	2023	15	28	44	374	2
5	00-0026158	2023	16	27	42	368	3
6	00-0026158	2023	17	19	29	309	3
7	00-0026498	2023	1	24	38	334	0
8	00-0026498	2023	2	34	55	307	1
9	00-0026498	2023	3	18	33	269	1
10	00-0026498	2023	4	27	40	319	1
	"						

<sup>#</sup> i 636 more rows

in the first 9 weeks of 2023. The variables in interest are 'passing\_epa', 'player\_id', 'season', 'week', 'completions', 'attempts', 'passing\_yards', 'passing\_tds', 'interceptions', 'sacks', and 'sack\_yards'.

## 3 Model

#### 3.1 Model set-up

Define  $y_i$  as the number of seconds that the plane remained aloft. Then  $\beta_i$  is the wing width and  $\gamma_i$  is the wing length, both measured in millimeters.

$$y = B_0 + B_1 X_{completions} + B_2 X_{attempts} + B_3 X_{passing_y ards} \tag{1}$$

$$+B_4X_{passing_tds} + B_5X_{interceptions} + B_6X_{sacks} + B_7X_{sack_yard}$$
 (2)

The model was ran in R(R Core Team 2023) by using rstanarm (Goodrich et al. 2022).

<sup>#</sup> i 4 more variables: interceptions <dbl>, sacks <dbl>, sack\_yards <dbl>,

<sup>#</sup> passing\_epa <dbl>

Table 2: Explanatory models of flight time based on wing width and wing length

```
parsnip model object
```

#### Call:

```
stats::lm(formula = passing_epa ~ completions + attempts + passing_yards +
passing_tds + interceptions + sacks + sack_yards, data = data)
```

#### Coefficients:

(Intercept)	completions	attempts	passing_yards	passing_tds
-0.93981	0.41778	-0.75037	0.09604	1.68490
interceptions	sacks	sack_yards		
-3.81529	-1.33799	-0.09272		

## 3.1.1 Model justification

The model considers 6 predictor variables (completions, attempts, passing\_yards, passing\_tds, interceptions, sacks, sack\_yard) to predict passing\_epa.

## 4 Results

The results from the model is in Table 2.

As a result, the paper could predict that the passing distance (passing\_yards (+0.09604)), the number of touchdowns (passing\_tds (1.68490)), number of sacks (sacks (-1.33799)), and yards of lost sack (sack\_yards (-0.09272)) influence to the total expected points added on passing. In addition, the number of touchdowns (passing\_tds (1.68490)) and number of sacks (sacks (-1.33799)) influence to the total expected points added on passing significantly.

# References

- Goodrich, Ben, Jonah Gabry, Imad Ali, and Sam Brilleman. 2022. "Rstanarm: Bayesian Applied Regression Modeling via Stan." https://mc-stan.org/rstanarm/.
- R Core Team. 2023. R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for Statistical Computing. https://www.R-project.org/.
- Sebastian Carl, Lee Sharpe, Ben Baldwin. 2023. Nflverse: Easily Install and Load the 'Nflverse'. https://CRAN.R-project.org/package=nflverse.