A Quasi-Poisson Approach
Goals in Football, What contributes to them?

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Background of study

Football has become a part of the world and is one of the biggest source of entertainment in the world. It has created job and opportunities to people all over the world. It's important in this modern time cannot be overlooked.There are five major football leagues in the world; English premier league, German BundesLiga, Spanish Laliga, Italian Seria A and France League one. Football supporters of English primier league all over the world always claim that, their league is the difficult one among all the five major leagues, is this claim true?. Football fans all over the world have insufficient answer to this question. This study tries to provide answer to this question.

Problem statement

Argument: If a league has a lot of goals scored in a season it is deemed not competitive because of the quality of opponents (vice versa as well). Goal: Does the factor league determine the amount of goals combined with other variables? If not, what predictors are associated with the response Goals?

Objectives

This study is directed by the specific objectives stated below;

1. Determine if there is a significant difference in goals across the 5 major leagues.
2. Determine if there are other predictors that are associated with the goals other than the different leagues.

Methodology

This study used a data which was collected from https://fbref.com, We used the 2021-2022 season statistics for EPL, La Liga, Bundesliga, Ligue 1 and Serie A in the analysis. The data comprise of 98 observations and 18 variables. Models we considered includes Poisson regression, Quasi - Poisson Regression. Diagnostic Tests: Test for Overdispersion, Residual Vs. Fitted Plot, Normal QQ-Plot

Poisson Regression

This is type a regression used when the response variable is a COUNT variable. It models the logarithm of the expected value of occurrence of an event as a linear function of some predictor variables.

The mathematical equation for the model is given below:

$$log(E(no.Goals|X)) = \beta_0 + \beta' X$$

Alternatively, we can write it in an exponential form as:

$$E(No.Goals|X) = e^{(\beta_0 + \beta' X)}$$

The method used here to estimate the paramters is the Maximum Liklihood estimator.

Assumptions of poisson Regression

- The number of goals must be scored in a fixed time.
-

$$E(no.Gls|X) = Var(no.Gls|X)$$

Estimation of parameter

The estimation method used here is the maximum likelihood estimation method. The procedure is briefly described below: Given the model:

$$\lambda(x) = E(Y|X) = e^{\beta' x}$$

Here Y, the response follow a Poisson distribution, therefore we have

$$P(y|x, \beta') = \frac{\lambda(x)^y e^{-\lambda(x)}}{y!}$$

$$= \frac{e^{y\beta' x} e^{-e^{\beta' x}}}{y!}$$

Now suppose we have data set (X, y) with n rows, then given an set of parameters β the likelihood is given by

$$L(\beta|X, Y) = \prod_{i=1}^{n} \frac{e^{y_i \beta' x_i} e^{-e^{\beta' x_i}}}{y_i!}$$

The goal is to find β which maximize this likelihood.

Quasi-Poisson

This study used this variant of Poisson regression because, the data is under-dispersed(the conditonal variance is smaller than expected). The type of model add a parameter called the dispersion which help solve the problem of under-dispersion. This reason for this model is to produce credible inference. The dispersion parameter is given by

$$\hat{\phi} = \frac{1}{n - K} \sum_{i=1}^{n} \frac{(y_i - \hat{y}_i)^2}{\hat{y}_i}$$

where K is the number of parameters.
such that

$$Var(Y|X) = \hat{\phi}E(Y|X)$$

if $\hat{\phi} = 1$ we have Possion regression
if $\hat{\phi} > 1$ we have over-dispersion
if $\hat{\phi} < 1$ we have under-dispersion

Pearson Residuals

This is a useful measure for diagnosing the appropriateness of a Poisson regression. This type of residual is a weighted residual and when ploted with the fitted points tells us if the Poisson regression is ok and there is no issue with dispersion. This measure is computed with the formula below.

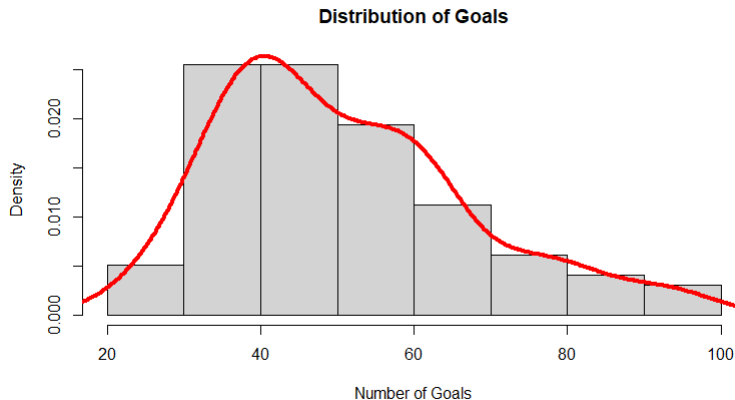$$PearsonResidual = \frac{y_i - \hat{y}_i}{\sqrt{\hat{y}_i}}$$

Results

Figure1: The plot show the goals distribution for all 5 Leagues
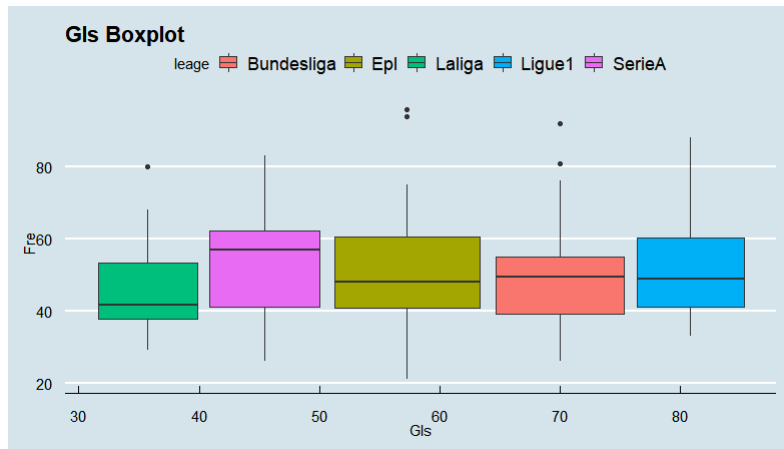
Results

Figure2: The plot show the goals distribution across the 5 Leagues

Linear Regression

Table: Naive OLS Model

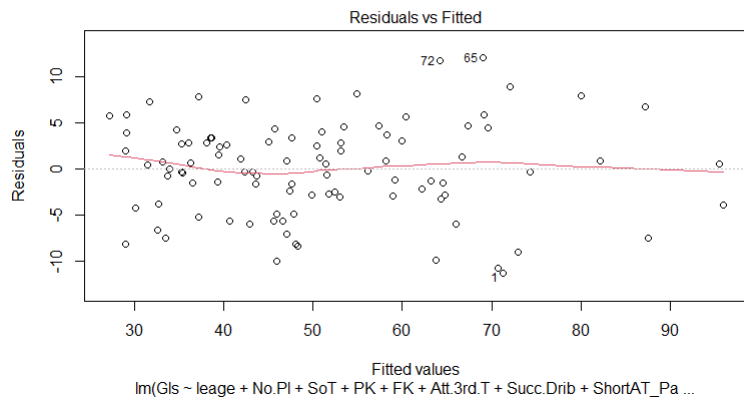|	Estimate	Std. Error	t value	Pr(>|t|)
(Intercept)	-14.943	6.097	-2.451	0.016
leageBundesliga	5.375	2.059	2.610	0.011
leageLaliga	-0.487	1.872	-0.260	0.795
leageLigue1	1.169	1.895	0.617	0.539
leageSerieA	-0.614	1.995	-0.308	0.759
No.Pl	0.133	0.166	0.802	0.425
SoT	0.296	0.034	8.672	0
PK	0.755	0.273	2.771	0.007
FK	-0.147	0.100	-1.471	0.145
Att.3rd.T	0.001	0.001	0.710	0.479
Succ.Drib	-0.022	0.012	-1.785	0.078
ShortAT_Pass	0.001	0.001	2.102	0.039
NumAtt	3.271	0.626	5.229	0.00000

Results

Figure3: The residual plots for OLS

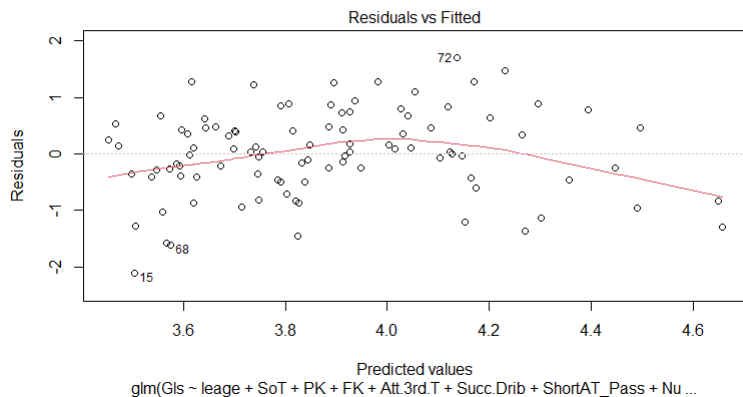# Results
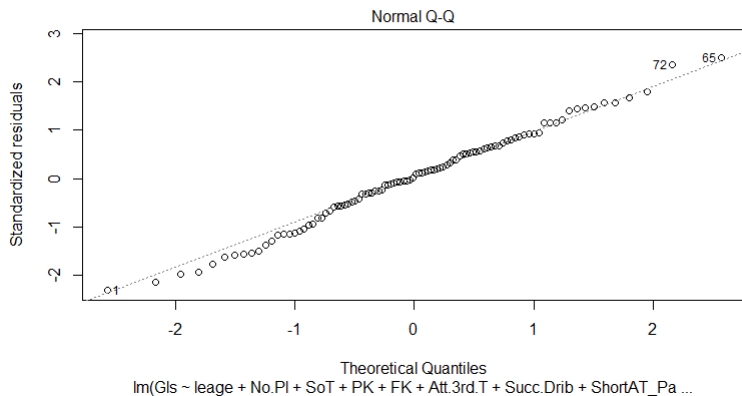
Figure4: The QQ plot for OLS

Results

Table: Poisson Regression Output

	Estimate	Std. Error	z value	$\Pr(>|z|)$
(Intercept)	2.853	0.103	27.592	0
leageBundesliga	0.093	0.053	1.770	0.077
leageLaliga	-0.007	0.048	-0.138	0.890
leageLigue1	0.036	0.048	0.742	0.458
leageSerieA	0.004	0.048	0.089	0.929
SoT	0.005	0.001	6.503	0
PK	0.016	0.007	2.244	0.025
FK	-0.003	0.003	-0.993	0.321
Att.3rd.T	-0.00001	0.00003	-0.283	0.777
Succ.Drib	-0.0005	0.0003	-1.655	0.098
ShortAT_Pass	0.00003	0.00002	1.885	0.059
NumAtt	0.073	0.015	4.719	0.00000

Dispersion parameter $= 1$

Results

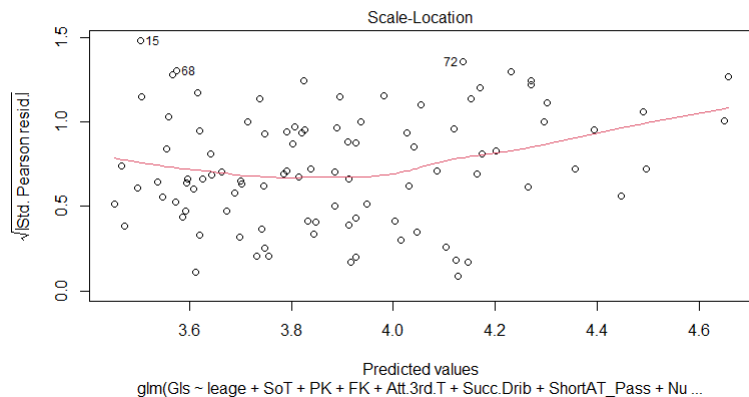Figure5: Residual Plot for Poisson Regression

Results

Figure6: Pearson Residual Plots for Poisson regression

Quasi-Poisson regression Output

Table: Quasi-Poisson regression

	Estimate	Std. Error	t value	Pr(>|t|)
(Intercept)	2.853	0.084	33.861	0
leageBundesliga	0.093	0.043	2.172	0.033
leageLaliga	-0.007	0.039	-0.169	0.866
leageLigue1	0.036	0.039	0.911	0.365
leageSerieA	0.004	0.039	0.109	0.913
SoT	0.005	0.001	7.981	0
PK	0.016	0.006	2.754	0.007
FK	-0.003	0.002	-1.219	0.226
Att.3rd.T	-0.00001	0.00002	-0.348	0.729
Succ.Drib	-0.0005	0.0002	-2.031	0.045
ShortAT_Pass	0.00003	0.00001	2.314	0.023
NumAtt	0.073	0.013	5.791	0.00000

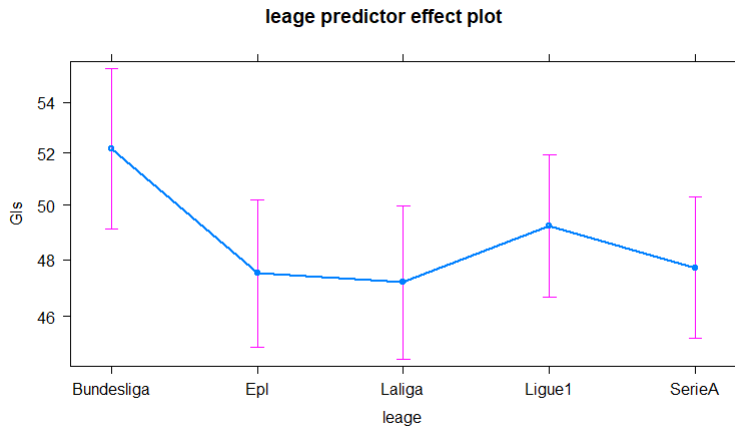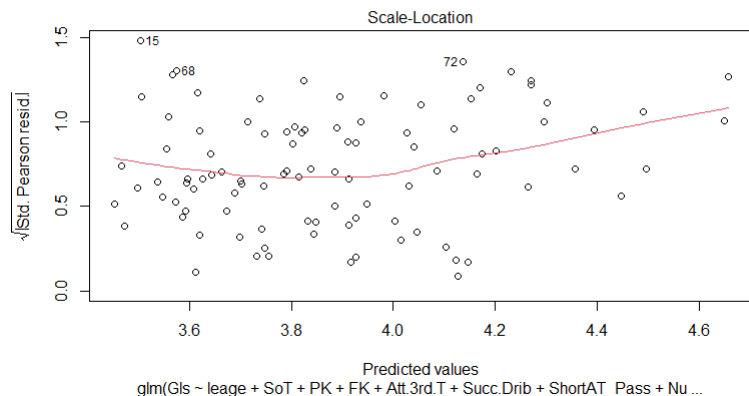Dispersion parameter= 0.6640285.

Results

Figure6: Pearson Residual Plots for Poisson regression

Results

Figure6: Pearson Residual Plots for Poisson regression

Results