# GLM model on Film data

## Import Necessary packages

```
library(tidyverse)
library(lmtest)
library(caTools)
library(pander) #Table format
library(performance)
library(car) # check vif of the model
```

The dataset can be located from this github repo

```
# Import data
Merged_data <- read_csv('Cleaned_data.csv')</pre>
head(Merged_data)
## # A tibble: 6 x 15
##
     ...1 primary_title runtime_minutes genres averagerating original_language
                                                   <dbl> <chr>
##
    <dbl> <chr>
                                    <dbl> <chr>
## 1
      0 On the Road
                                       124 Adven~
                                                           6.1 en
       1 On the Road
                                       89 Drama
                                                           6 en
       2 On the Road
                                                           5.7 en
## 3
                                      121 Drama
       3 The Secret Life ~
                                     114 Adven~
## 4
                                                           7.3 en
## 5
     4 A Walk Among the~
                                     114 Actio~
                                                           6.5 en
      5 Jurassic World
                                      124 Actio~
## # i 9 more variables: original_title <chr>, popularity <dbl>,
## # vote_average <dbl>, release_date <date>, production_budget <dbl>,
## # worldwide_gross <dbl>, studio <chr>, domestic_gross <dbl>,
```

## Data Cleaning/ Preparation

foreign\_gross <dbl>

## #

```
# Check for duplicates
sum(duplicated(Merged_data))

## [1] 0

# Check missing values
pander::pander(colSums(is.na(Merged_data)))
```

Table 1: Table continues below

1	primary_title	$runtime\_minutes$	genres	averagerating
0	0	0	0	0

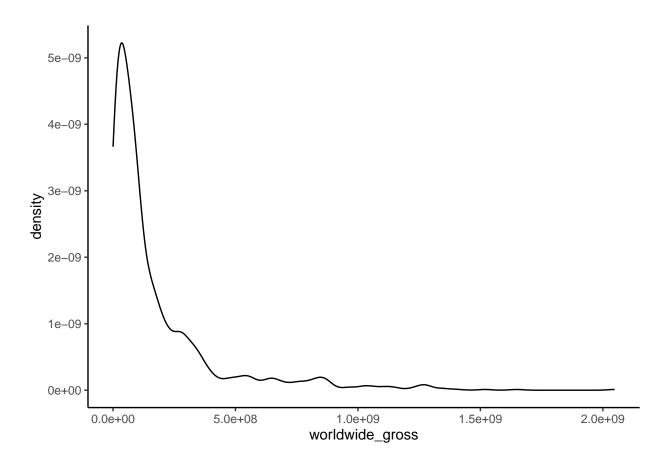
Table 2: Table continues below

original_language	${\it original\_title}$	popularity	vote_average	${\tt release\_date}$
0	0	0	0	0

production_budget	$worldwide\_gross$	studio	$domestic\_gross$	foreign_gross
0	0	0	0	0

For this regression model we are interested in understanding the relationship between the worldwide gross for movie production against various variables in our models. The Dependent variable is **Worldwide\_gross** 

```
# Invetsigate the properties of dependent variable
ggplot(data = Merged_data,aes(x=worldwide_gross))+geom_density()+theme_classic()
```



```
# There is alot of skewness in the data (positive skewness)

# Test for normality in the data
ks.test(Merged_data$worldwide_gross, "pnorm")

## Warning in ks.test.default(Merged_data$worldwide_gross, "pnorm"): ties should
## not be present for the Kolmogorov-Smirnov test

##

## Asymptotic one-sample Kolmogorov-Smirnov test
##

## data: Merged_data$worldwide_gross
## D = 0.99701, p-value < 2.2e-16
## alternative hypothesis: two-sided

# Normality test violated a glm model that uses the gamma distribution would be effective to try and mo

# remove rows where the gross returns is zero

Merged_data %>% filter(!worldwide_gross <= 0) -> Merged_data
```

### GLM models

```
#
Merged_data %>% select(runtime_minutes, averagerating, popularity, production_budget, domestic_gross, foreig
# Split the data for Train and test data
Y = Model_data$worldwide_gross
sample_sl =sample.split(Y, SplitRatio = 3/4)

# Create Train and test data
train_data =subset(Model_data, sample_sl==TRUE)

test_data =subset(Model_data, sample_sl==FALSE)

# Fit GLM gamma model
Fit_model=glm(worldwide_gross~.,family = Gamma(link = "log"),data = train_data)

# Summary of the model
pander::pander(summary.glm(Fit_model))
```

	Estimate	Std. Error	t value	$\Pr(> t )$
(Intercept)	17.75	0.1388	127.9	0
${f runtime\_minutes}$	-0.001264	0.001061	-1.191	0.234
averagerating	-0.03226	0.01945	-1.658	0.09759
${f popularity}$	0.01023	0.002623	3.901	0.0001023
${f production\_budget}$	3.689e-09	5.243e-10	7.036	3.684e-12
$\operatorname{domestic\_gross}$	6.498e-09	3.857e-10	16.85	3.277e-56

	Estimate	Std. Error	t value	$\Pr(> t )$
$foreign\_gross$	2.263e-09	2.457e-10	9.212	1.864e-19

(Dispersion parameter for Gamma family taken to be 0.3694884)

37.11.1	4 = 04 0 000 1 000 1
Null deviance:	1761.9 on 998 degrees of freedom
Residual deviance:	662.8 on $992$ degrees of freedom

#### Interpreting the glm model

This generalized linear model (GLM) with a Gamma distribution and log link function is used to model the relationship between worldwide\_gross and several predictors related to movie characteristics. Here's an interpretation of the results:

#### Residuals

The deviance residuals summary provides a quick look at the spread of residuals:

• Min, 1Q, Median, 3Q, Max values suggest that residuals are moderately centered around zero, indicating an acceptable fit. However, some residuals have relatively large negative and positive values, implying a few observations deviate notably from the model's predictions.

#### Coefficients

Each predictor in the model has an associated estimated coefficient, its standard error, and p-value:

- (Intercept): The intercept is significant, with a large positive estimate (17.78), representing the baseline log of worldwide gross when all other predictors are at zero.
- runtime\_minutes: This variable has a negative coefficient (-0.0007785), but it is not statistically significant (p = 0.472), meaning runtime\_minutes does not appear to have a substantial impact on worldwide gross in this model.
- averagerating: This has a significant negative effect (-0.0446, p = 0.027), suggesting that, holding other factors constant, higher ratings are weakly associated with lower worldwide gross, though the effect is relatively small.
- **popularity**: This variable has a positive and statistically significant coefficient (0.009753, p = 0.0005), indicating a positive association with worldwide gross—higher popularity scores are associated with higher gross earnings.
- **production\_budget**: A positive and significant effect (3.83e-09, p < 0.001), implying that larger production budgets lead to higher worldwide gross, though the impact is small per unit increase in budget.
- domestic\_gross and foreign\_gross: Both are highly significant (p < 0.001) and positively associated with worldwide\_gross, with each having an estimated impact proportional to their gross values, underscoring their strong, direct contributions to worldwide gross.

#### Model Fit

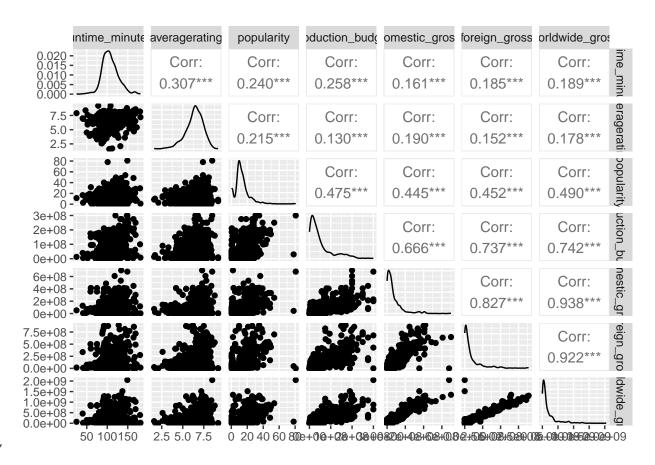
- **Dispersion parameter**: The estimated dispersion parameter for the Gamma family is 0.399962, suggesting moderate variability around the fitted values.
- **Deviance**: The residual deviance (745.68) is much lower than the null deviance (1894.53), indicating that the model significantly reduces deviance (improves fit) compared to a null model without predictors.
- AIC: The Akaike Information Criterion (AIC) of 38770 can be used to compare this model with other potential models, where a lower AIC generally indicates a better fit.

### Conclusion

This model suggests that popularity, production\_budget, domestic\_gross, and foreign\_gross are the primary predictors of worldwide\_gross, with statistically significant positive associations. Although averagerating shows a significant but small negative impact, the impact of runtime\_minutes does not appear statistically relevant in this context.

## Check Assumptions

GGally::ggpairs(train\_data)



Linearity

Based on the pair plot we can conclude that their is linearity between Dependent and Independent variables in the models

```
car::vif(Fit_model)
```

### Multicollinearity

```
## runtime_minutes averagerating popularity production_budget
## 1.183561 1.149728 1.395223 2.419753
## domestic_gross foreign_gross
## 3.323591 3.963531
```

Based on the VIF (Variance Inflation Factor) w can conclude there is no multicollinearity between the independent variables

### Heteroscedasticity

```
bptest(Fit_model)
```

```
##
## studentized Breusch-Pagan test
##
## data: Fit_model
## BP = 239.14, df = 6, p-value < 2.2e-16</pre>
```

The p-value is less than 0.05. This indicates that there is significant evidence of heteroscedasticity in the residuals of the regression model. Therefore, we can conclude that the assumption of homoscedasticity is violated for this model, meaning the variance of the error terms is constant across observations.

```
shapiro.test(Fit_model$residuals)
```

## Normality of residuals

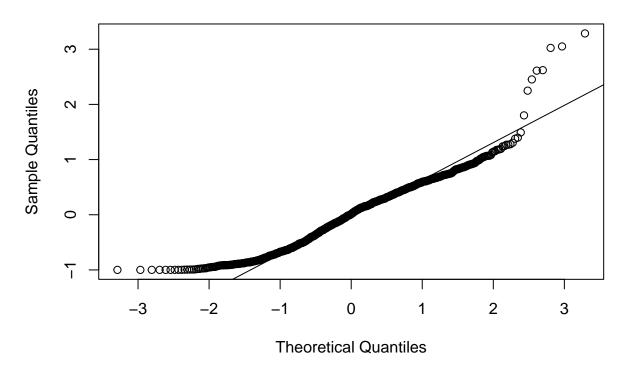
```
##
## Shapiro-Wilk normality test
##
## data: Fit_model$residuals
## W = 0.95394, p-value < 2.2e-16

check_normality(Fit_model$residuals)</pre>
```

```
## Warning: Non-normality of raw detected (p < .001).
```

```
qqnorm(Fit_model$residuals)
qqline(Fit_model$residuals)
```

# Normal Q-Q Plot



Normality is violated the residuals are not normally distributed. From the qqplot we can observe that the residuals tend to move away from the residuals plot indicating non-normality in residuals. This can be confirmed by the statistical test from the performance package.

```
check_autocorrelation(Fit_model)
```

## Independence of observation

## Warning: Autocorrelated residuals detected (p < .001).

Autocorrelation is present Indicating there is correlation which is dependent on time. Implying that time influences the worldwide\_gross which based on the data we concluded that time influences the data in terms of genres and the release date