

# Final Project

Navya

2024-03-02

```
data = read.csv("C:/Users/navya/Documents/DATA557-Project/Factors-Influencing-Online-News-Popularity/data/OnlineNewsPopularity.csv")  
head(data)
```

```

##                                     url timedelta
## 1   http://mashable.com/2013/01/07/amazon-instant-video-browser/      731
## 2   http://mashable.com/2013/01/07/ap-samsung-sponsored-tweets/      731
## 3   http://mashable.com/2013/01/07/apple-40-billion-app-downloads/    731
## 4   http://mashable.com/2013/01/07/astronaut-notre-dame-bcs/          731
## 5   http://mashable.com/2013/01/07/att-u-verse-apps/                  731
## 6   http://mashable.com/2013/01/07/beewi-smart-toys/                  731
##   n_tokens_title n_tokens_content n_unique_tokens n_non_stop_words
## 1           12           219           0.6635945           1
## 2           9           255           0.6047431           1
## 3           9           211           0.5751295           1
## 4           9           531           0.5037879           1
## 5          13          1072           0.4156456           1
## 6          10           370           0.5598886           1
##   n_non_stop_unique_tokens num_hrefs num_self_hrefs num_imgs num_videos
## 1           0.8153846           4           2           1           0
## 2           0.7919463           3           1           1           0
## 3           0.6638655           3           1           1           0
## 4           0.6656347           9           0           1           0
## 5           0.5408895          19          19          20           0
## 6           0.6981982           2           2           0           0
##   average_token_length num_keywords data_channel_is_lifestyle
## 1           4.680365           5           0
## 2           4.913725           4           0
## 3           4.393365           6           0
## 4           4.404896           7           0
## 5           4.682836           7           0
## 6           4.359459           9           0
##   data_channel_is_entertainment data_channel_is_bus data_channel_is_socmed
## 1           1           0           0
## 2           0           1           0
## 3           0           1           0
## 4           1           0           0
## 5           0           0           0
## 6           0           0           0
##   data_channel_is_tech data_channel_is_world kw_min_min kw_max_min kw_avg_min
## 1           0           0           0           0           0
## 2           0           0           0           0           0
## 3           0           0           0           0           0
## 4           0           0           0           0           0
## 5           1           0           0           0           0
## 6           1           0           0           0           0
##   kw_min_max kw_max_max kw_avg_max kw_min_avg kw_max_avg kw_avg_avg
## 1           0           0           0           0           0           0
## 2           0           0           0           0           0           0
## 3           0           0           0           0           0           0
## 4           0           0           0           0           0           0
## 5           0           0           0           0           0           0
## 6           0           0           0           0           0           0
##   self_reference_min_shares self_reference_max_shares
## 1           496           496
## 2           0           0

```

```

## 3          918          918
## 4          0          0
## 5          545         16000
## 6          8500         8500
## self_reference_avg_sharess weekday_is_monday weekday_is_tuesday
## 1          496.000          1          0
## 2          0.000          1          0
## 3          918.000          1          0
## 4          0.000          1          0
## 5          3151.158          1          0
## 6          8500.000          1          0
## weekday_is_wednesday weekday_is_thursday weekday_is_friday
## 1          0          0          0
## 2          0          0          0
## 3          0          0          0
## 4          0          0          0
## 5          0          0          0
## 6          0          0          0
## weekday_is_saturday weekday_is_sunday is_weekend LDA_00 LDA_01
## 1          0          0          0 0.50033120 0.37827893
## 2          0          0          0 0.79975569 0.05004668
## 3          0          0          0 0.21779229 0.03333446
## 4          0          0          0 0.02857322 0.41929964
## 5          0          0          0 0.02863281 0.02879355
## 6          0          0          0 0.02224528 0.30671758
## LDA_02 LDA_03 LDA_04 global_subjectivity
## 1 0.04000468 0.04126265 0.04012254 0.5216171
## 2 0.05009625 0.05010067 0.05000071 0.3412458
## 3 0.033335142 0.03333354 0.68218829 0.7022222
## 4 0.49465083 0.02890472 0.02857160 0.4298497
## 5 0.02857518 0.02857168 0.88542678 0.5135021
## 6 0.02223128 0.02222429 0.62658158 0.4374086
## global_sentiment_polarity global_rate_positive_words
## 1          0.09256198          0.04566210
## 2          0.14894781          0.04313725
## 3          0.32333333          0.05687204
## 4          0.10070467          0.04143126
## 5          0.28100348          0.07462687
## 6          0.07118419          0.02972973
## global_rate_negative_words rate_positive_words rate_negative_words
## 1          0.013698630          0.7692308          0.2307692
## 2          0.015686275          0.7333333          0.2666667
## 3          0.009478673          0.8571429          0.1428571
## 4          0.020715631          0.6666667          0.3333333
## 5          0.012126866          0.8602151          0.1397849
## 6          0.027027027          0.5238095          0.4761905
## avg_positive_polarity min_positive_polarity max_positive_polarity
## 1          0.3786364          0.10000000          0.7
## 2          0.2869146          0.03333333          0.7
## 3          0.4958333          0.10000000          1.0
## 4          0.3859652          0.13636364          0.8
## 5          0.4111274          0.03333333          1.0

```

```
## 6          0.3506100          0.13636364          0.6
##  avg_negative_polarity min_negative_polarity max_negative_polarity
## 1          -0.3500000          -0.600          -0.2000000
## 2          -0.1187500          -0.125          -0.1000000
## 3          -0.4666667          -0.800          -0.1333333
## 4          -0.3696970          -0.600          -0.1666667
## 5          -0.2201923          -0.500          -0.0500000
## 6          -0.1950000          -0.400          -0.1000000
##  title_subjectivity title_sentiment_polarity abs_title_subjectivity
## 1          0.5000000          -0.1875000          0.0000000
## 2          0.0000000          0.0000000          0.5000000
## 3          0.0000000          0.0000000          0.5000000
## 4          0.0000000          0.0000000          0.5000000
## 5          0.4545455          0.1363636          0.04545455
## 6          0.6428571          0.2142857          0.14285714
##  abs_title_sentiment_polarity shares
## 1          0.1875000          593
## 2          0.0000000          711
## 3          0.0000000          1500
## 4          0.0000000          1200
## 5          0.1363636          505
## 6          0.2142857          855
```

```
data$n_tokens_content = scale(data$n_tokens_content)
data$n_tokens_title = scale(data$n_tokens_title)
```

```
data$is_popular = data$shares>30000
# View the modified dataset
head(data)
```

```

##                                     url timedelta
## 1  http://mashable.com/2013/01/07/amazon-instant-video-browser/      731
## 2  http://mashable.com/2013/01/07/ap-samsung-sponsored-tweets/      731
## 3  http://mashable.com/2013/01/07/apple-40-billion-app-downloads/    731
## 4  http://mashable.com/2013/01/07/astronaut-notre-dame-bcs/         731
## 5  http://mashable.com/2013/01/07/att-u-verse-apps/                 731
## 6  http://mashable.com/2013/01/07/beewi-smart-toys/                  731
##  n_tokens_title n_tokens_content n_unique_tokens n_non_stop_words
## 1      0.7574377      -0.69520168      0.6635945      1
## 2     -0.6616483     -0.61878600      0.6047431      1
## 3     -0.6616483     -0.71218294      0.5751295      1
## 4     -0.6616483     -0.03293246      0.5037879      1
## 5      1.2304663      1.11542538      0.4156456      1
## 6     -0.1886196     -0.37468036      0.5598886      1
##  n_non_stop_unique_tokens num_hrefs num_self_hrefs num_imgs num_videos
## 1      0.8153846          4          2          1          0
## 2      0.7919463          3          1          1          0
## 3      0.6638655          3          1          1          0
## 4      0.6656347          9          0          1          0
## 5      0.5408895         19         19         20          0
## 6      0.6981982          2          2          0          0
##  average_token_length num_keywords data_channel_is_lifestyle
## 1      4.680365          5          0
## 2      4.913725          4          0
## 3      4.393365          6          0
## 4      4.404896          7          0
## 5      4.682836          7          0
## 6      4.359459          9          0
##  data_channel_is_entertainment data_channel_is_bus data_channel_is_socmed
## 1          1          0          0
## 2          0          1          0
## 3          0          1          0
## 4          1          0          0
## 5          0          0          0
## 6          0          0          0
##  data_channel_is_tech data_channel_is_world kw_min_min kw_max_min kw_avg_min
## 1          0          0          0          0          0
## 2          0          0          0          0          0
## 3          0          0          0          0          0
## 4          0          0          0          0          0
## 5          1          0          0          0          0
## 6          1          0          0          0          0
##  kw_min_max kw_max_max kw_avg_max kw_min_avg kw_max_avg kw_avg_avg
## 1          0          0          0          0          0          0
## 2          0          0          0          0          0          0
## 3          0          0          0          0          0          0
## 4          0          0          0          0          0          0
## 5          0          0          0          0          0          0
## 6          0          0          0          0          0          0
##  self_reference_min_shares self_reference_max_shares
## 1          496          496
## 2          0          0

```

```

## 3          918          918
## 4          0          0
## 5          545         16000
## 6          8500         8500
## self_reference_avg_sharess weekday_is_monday weekday_is_tuesday
## 1          496.000          1          0
## 2          0.000          1          0
## 3          918.000          1          0
## 4          0.000          1          0
## 5          3151.158          1          0
## 6          8500.000          1          0
## weekday_is_wednesday weekday_is_thursday weekday_is_friday
## 1          0          0          0
## 2          0          0          0
## 3          0          0          0
## 4          0          0          0
## 5          0          0          0
## 6          0          0          0
## weekday_is_saturday weekday_is_sunday is_weekend LDA_00 LDA_01
## 1          0          0          0 0.50033120 0.37827893
## 2          0          0          0 0.79975569 0.05004668
## 3          0          0          0 0.21779229 0.03333446
## 4          0          0          0 0.02857322 0.41929964
## 5          0          0          0 0.02863281 0.02879355
## 6          0          0          0 0.02224528 0.30671758
## LDA_02 LDA_03 LDA_04 global_subjectivity
## 1 0.04000468 0.04126265 0.04012254 0.5216171
## 2 0.05009625 0.05010067 0.05000071 0.3412458
## 3 0.033335142 0.03333354 0.68218829 0.7022222
## 4 0.49465083 0.02890472 0.02857160 0.4298497
## 5 0.02857518 0.02857168 0.88542678 0.5135021
## 6 0.02223128 0.02222429 0.62658158 0.4374086
## global_sentiment_polarity global_rate_positive_words
## 1          0.09256198          0.04566210
## 2          0.14894781          0.04313725
## 3          0.32333333          0.05687204
## 4          0.10070467          0.04143126
## 5          0.28100348          0.07462687
## 6          0.07118419          0.02972973
## global_rate_negative_words rate_positive_words rate_negative_words
## 1          0.013698630          0.7692308          0.2307692
## 2          0.015686275          0.7333333          0.2666667
## 3          0.009478673          0.8571429          0.1428571
## 4          0.020715631          0.6666667          0.3333333
## 5          0.012126866          0.8602151          0.1397849
## 6          0.027027027          0.5238095          0.4761905
## avg_positive_polarity min_positive_polarity max_positive_polarity
## 1          0.3786364          0.10000000          0.7
## 2          0.2869146          0.03333333          0.7
## 3          0.4958333          0.10000000          1.0
## 4          0.3859652          0.13636364          0.8
## 5          0.4111274          0.03333333          1.0

```

```
## 6          0.3506100          0.13636364          0.6
##  avg_negative_polarity min_negative_polarity max_negative_polarity
## 1          -0.3500000          -0.600          -0.2000000
## 2          -0.1187500          -0.125          -0.1000000
## 3          -0.4666667          -0.800          -0.1333333
## 4          -0.3696970          -0.600          -0.1666667
## 5          -0.2201923          -0.500          -0.0500000
## 6          -0.1950000          -0.400          -0.1000000
##  title_subjectivity title_sentiment_polarity abs_title_subjectivity
## 1          0.5000000          -0.1875000          0.0000000
## 2          0.0000000          0.0000000          0.5000000
## 3          0.0000000          0.0000000          0.5000000
## 4          0.0000000          0.0000000          0.5000000
## 5          0.4545455          0.1363636          0.04545455
## 6          0.6428571          0.2142857          0.14285714
##  abs_title_sentiment_polarity shares is_popular
## 1          0.1875000      593      FALSE
## 2          0.0000000      711      FALSE
## 3          0.0000000     1500      FALSE
## 4          0.0000000     1200      FALSE
## 5          0.1363636      505      FALSE
## 6          0.2142857      855      FALSE
```

```
# Logistic Model with only title length
```

```
model_title = glm(is_popular ~ n_tokens_title, family=binomial, data = data)
summary(model_title)
```

```
##
## Call:
## glm(formula = is_popular ~ n_tokens_title, family = binomial,
##      data = data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -4.52774    0.04913  -92.159  < 2e-16 ***
## n_tokens_title  0.15846    0.04795   3.305 0.000951 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 4736.9  on 39643  degrees of freedom
## Residual deviance: 4726.0  on 39642  degrees of freedom
## AIC: 4730
##
## Number of Fisher Scoring iterations: 7
```

```
# Logistic Model with only content length
```

```
model_content = glm(is_popular ~ n_tokens_content, family=binomial, data = data)
summary(model_content)
```

```
##
## Call:
## glm(formula = is_popular ~ n_tokens_content, family = binomial,
##      data = data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -4.51931    0.04875 -92.708  <2e-16 ***
## n_tokens_content -0.09358    0.05456  -1.715   0.0863 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 4736.9  on 39643  degrees of freedom
## Residual deviance: 4733.7  on 39642  degrees of freedom
## AIC: 4737.7
##
## Number of Fisher Scoring iterations: 7
```

```
# Logistic Model with title Length and content Length combined
model_title_content_combined = glm(is_popular ~ n_tokens_content + n_tokens_title, family=binomial, data = data)
summary(model_title_content_combined)
```

```
##
## Call:
## glm(formula = is_popular ~ n_tokens_content + n_tokens_title,
##      family = binomial, data = data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -4.53198    0.04935 -91.841  < 2e-16 ***
## n_tokens_content -0.09657    0.05445  -1.773  0.076157 .
## n_tokens_title    0.16043    0.04805   3.339  0.000841 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 4736.9  on 39643  degrees of freedom
## Residual deviance: 4722.6  on 39641  degrees of freedom
## AIC: 4728.6
##
## Number of Fisher Scoring iterations: 7
```

```
# Confidence Interval for Combined Model
(confint(model_title_content_combined, level = 0.95))
```

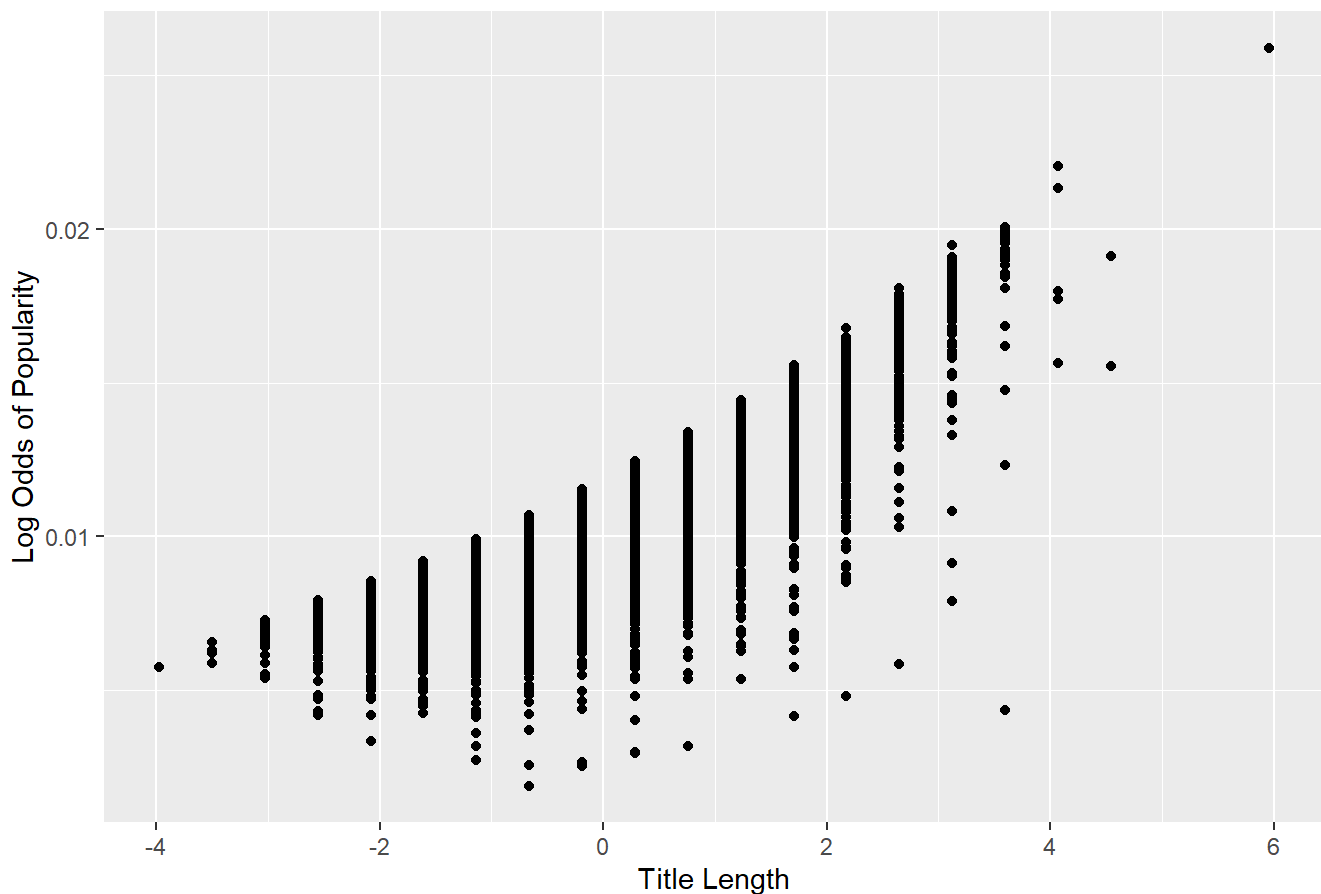


```
## Waiting for profiling to be done...
```

```
##           2.5 %       97.5 %
## (Intercept) -4.63030261 -4.436811090
## n_tokens_content -0.20736289  0.005849317
## n_tokens_title  0.06604559  0.254393222
```

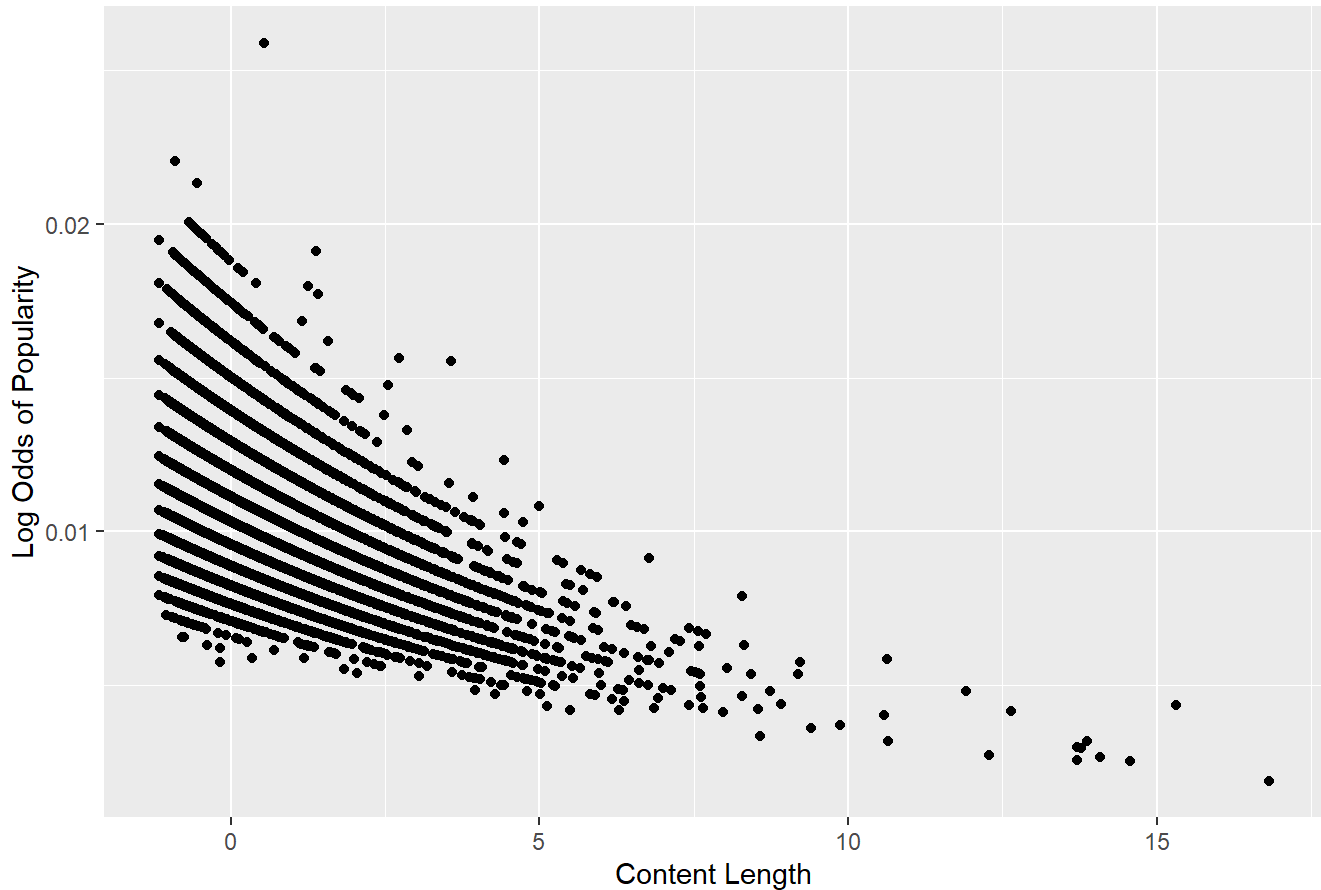
```
# Testing logistic regression assumption - Linearity between log odds of probability and title length
library(ggplot2)
ggplot(data, aes(x = n_tokens_title, y = predict(model_title_content_combined, type = "response"))) +
  geom_point() +
  labs(title = "Logit of is_popular vs. title length", x = "Title Length", y = "Log Odds of Popularity")
```

Logit of is\_popular vs. title length



```
# Testing Logistic regression assumption - Linearity between log odds of probability and content length
library(ggplot2)
ggplot(data, aes(x = n_tokens_content, y = predict(model_title_content_combined, type = "response"))) +
  geom_point() +
  labs(title = "Logit of is_popular vs. content length", x = "Content Length", y = "Log Odds of Popularity")
```

Logit of is\_popular vs. content length



```
# Testing Logistic Regression Assumption of multicollinearity
(correlation <- cor(data[, "n_tokens_title"], data[, "n_tokens_content"]))
```

```
##           [,1]
## [1,] 0.01815965
```

```
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```

par(cex = 0.75)

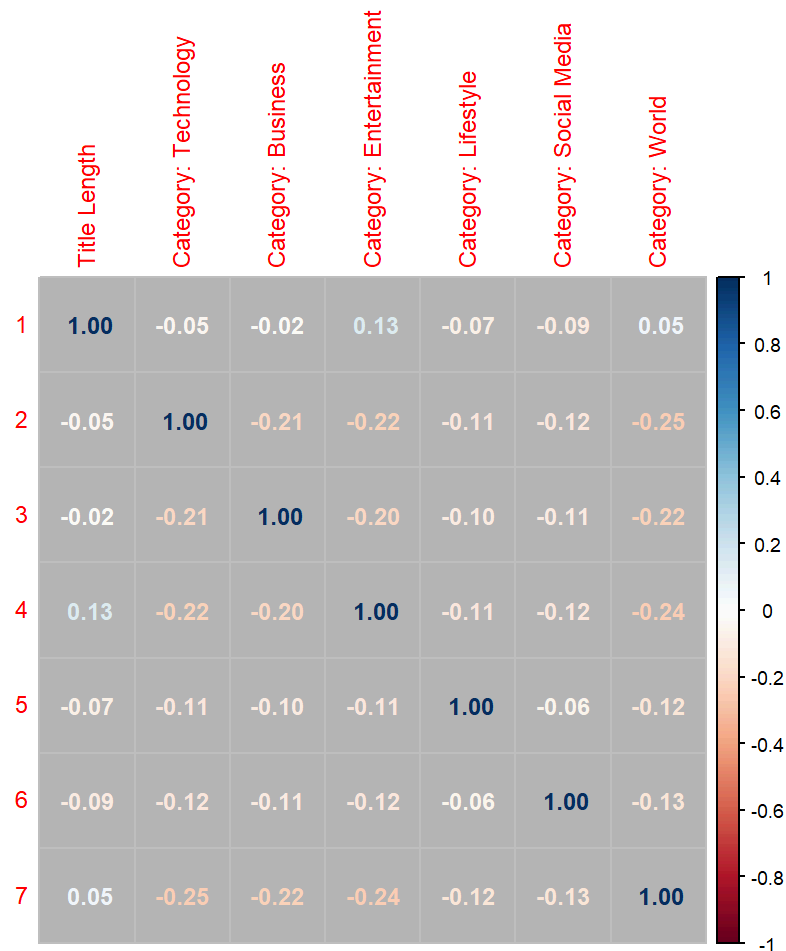
# Assuming house_data is a data frame
data_list <- list(data$n_tokens_title, data$data_channel_is_tech, data$data_channel_is_bus, data
$data_channel_is_entertainment, data$data_channel_is_lifestyle, data$data_channel_is_socmed, dat
a$data_channel_is_world)

# Convert the list of variables into a matrix and calculate correlations
correlation_values <- cor(do.call(cbind, data_list))

# Add column names to the correlation matrix
colnames(correlation_values) <- c("Title Length", "Category: Technology", "Category: Business",
"Category: Entertainment", "Category: Lifestyle", "Category: Social Media", "Category: World")

# Plot the correlation matrix
corrplot(correlation_values, method = "number", bg = "#B4B4B8", number.digits = 2)

```



```

# Logistic regression model of title length, content length and their combine effect
model_interaction = glm(is_popular ~ n_tokens_content * n_tokens_title, family=binomial, data =
data)
summary(model_interaction)

```

```
##
## Call:
## glm(formula = is_popular ~ n_tokens_content * n_tokens_title,
##      family = binomial, data = data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      -4.53154    0.04932 -91.871 < 2e-16 ***
## n_tokens_content  -0.09029    0.05502  -1.641 0.100765
## n_tokens_title      0.15864    0.04816   3.294 0.000987 ***
## n_tokens_content:n_tokens_title -0.03330    0.05144  -0.647 0.517319
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 4736.9  on 39643  degrees of freedom
## Residual deviance: 4722.2  on 39640  degrees of freedom
## AIC: 4730.2
##
## Number of Fisher Scoring iterations: 7
```

$$\text{logit}(\text{is\_popular}) = \beta_0 + \beta_1[\text{n\_tokens\_title}] + \beta_2[\text{Category}] + \beta_3[\text{n\_tokens\_title} * \text{Category}]$$

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
data <- data %>%
  mutate(Category = case_when(
    data_channel_is_lifestyle == 1 ~ "Lifestyle",
    data_channel_is_entertainment == 1 ~ "Entertainment",
    data_channel_is_bus == 1 ~ "Business",
    data_channel_is_socmed == 1 ~ "Social Media",
    data_channel_is_tech == 1 ~ "Technology",
    data_channel_is_world == 1 ~ "World",
    TRUE ~ "Other"
  ))
head(data$Category)
```

```
## [1] "Entertainment" "Business"      "Business"      "Entertainment"
## [5] "Technology"     "Technology"
```

```
lm_model_total_before = glm(is_popular ~ n_tokens_title + Category, family=binomial, data = data)
summary(lm_model_total_before)
```

```
##
## Call:
## glm(formula = is_popular ~ n_tokens_title + Category, family = binomial,
##      data = data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -4.86833     0.14504  -33.566 < 2e-16 ***
## n_tokens_title     0.17048     0.04957   3.439 0.000583 ***
## CategoryEntertainment  0.32479     0.18457   1.760 0.078460 .
## CategoryLifestyle     0.31292     0.26326   1.189 0.234576
## CategoryOther         1.39312     0.16303   8.545 < 2e-16 ***
## CategorySocial Media  0.06328     0.27793   0.228 0.819891
## CategoryTechnology    -0.50040     0.22485  -2.225 0.026048 *
## CategoryWorld        -0.41173     0.20952  -1.965 0.049399 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 4736.9  on 39643  degrees of freedom
## Residual deviance: 4507.8  on 39636  degrees of freedom
## AIC: 4523.8
##
## Number of Fisher Scoring iterations: 8
```

```
lm_model_total = glm(is_popular ~ n_tokens_title * Category, data= data)
summary(lm_model_total)
```

```
##
## Call:
## glm(formula = is_popular ~ n_tokens_title * Category, data = data)
##
## Coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.0078959   0.0013051    6.050 1.46e-09 ***
## n_tokens_title  0.0040882   0.0012754    3.205  0.00135 **
## CategoryEntertainment  0.0031840   0.0018262    1.744  0.08125 .
## CategoryLifestyle    0.0017713   0.0027061    0.655  0.51275
## CategoryOther        0.0224274   0.0018546   12.093 < 2e-16 ***
## CategorySocial Media -0.0001018   0.0026240   -0.039  0.96906
## CategoryTechnology   -0.0033340   0.0017787   -1.874  0.06088 .
## CategoryWorld        -0.0029736   0.0017251   -1.724  0.08476 .
## n_tokens_title:CategoryEntertainment -0.0036863   0.0017810   -2.070  0.03848 *
## n_tokens_title:CategoryLifestyle    -0.0052152   0.0027993   -1.863  0.06247 .
## n_tokens_title:CategoryOther        -0.0004642   0.0018897   -0.246  0.80596
## n_tokens_title:CategorySocial Media -0.0039626   0.0025010   -1.584  0.11311
## n_tokens_title:CategoryTechnology   -0.0047672   0.0017542   -2.718  0.00658 **
## n_tokens_title:CategoryWorld        -0.0009313   0.0017102   -0.545  0.58606
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.01062735)
##
##      Null deviance: 424.36  on 39643  degrees of freedom
## Residual deviance: 421.16  on 39630  degrees of freedom
## AIC: -67634
##
## Number of Fisher Scoring iterations: 2
```

```
(confint(lm_model_total, level = 0.95))
```

```
## Waiting for profiling to be done...
```

##	2.5 %	97.5 %
## (Intercept)	0.0053380233	0.0104537328
## n_tokens_title	0.0015883820	0.0065879509
## CategoryEntertainment	-0.0003952369	0.0067632337
## CategoryLifestyle	-0.0035325971	0.0070752846
## CategoryOther	0.0187923474	0.0260623596
## CategorySocial Media	-0.0052447988	0.0050412401
## CategoryTechnology	-0.0068202057	0.0001521602
## CategoryWorld	-0.0063547438	0.0004074641
## n_tokens_title:CategoryEntertainment	-0.0071769643	-0.0001955612
## n_tokens_title:CategoryLifestyle	-0.0107018071	0.0002714072
## n_tokens_title:CategoryOther	-0.0041680299	0.0032396270
## n_tokens_title:CategorySocial Media	-0.0088644877	0.0009393052
## n_tokens_title:CategoryTechnology	-0.0082054028	-0.0013289836
## n_tokens_title:CategoryWorld	-0.0042831509	0.0024205788

```
(confint(lm_model_total_before, level = 0.95))
```

```
## Waiting for profiling to be done...
```

##	2.5 %	97.5 %
## (Intercept)	-5.16647579	-4.5966811800
## n_tokens_title	0.07309991	0.2674126496
## CategoryEntertainment	-0.03275729	0.6926683532
## CategoryLifestyle	-0.22303378	0.8147687380
## CategoryOther	1.08263274	1.7231711663
## CategorySocial Media	-0.50710501	0.5893672240
## CategoryTechnology	-0.94837258	-0.0634185836
## CategoryWorld	-0.82500679	-0.0008748532

```
(summary(data$n_tokens_title))
```

```
##      V1
##  Min.   :-3.9728
## 1st Qu.: -0.6616
## Median :-0.1886
## Mean   : 0.0000
## 3rd Qu.: 0.7574
## Max.   : 5.9608
```

```
(summary(data$n_tokens_content))
```

```
##          V1
##  Min.    :-1.1601
## 1st Qu.  :-0.6379
## Median   :-0.2919
## Mean     : 0.0000
## 3rd Qu.  : 0.3598
## Max.     :16.8273
```