

# INTRODUCTION TO DATA MINING

INSTRUCTED BY ASMA SANAM

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## CONTENTS

INTRODUCTION .....	3
CONDUCTING DATA ANALYSIS .....	4
Step 1-BUSINESS UNDERSTANDING .....	4
Step 2- DATA UNDERSTANDING .....	5
Step 3- DATA PREPARATION.....	7
Step-4 MODELLING .....	8
Step 5- EVALUATION .....	15
CROSS VALIDATING A MODEL A: .....	16
Step-6 DEPLOYMENT .....	16
REFERENCES .....	18

## **INTRODUCTION**

### **OVER VIEW:**

Mashable is a global, multi-platform media and entertainment company. Powered by its own proprietary technology, Mashable is the go-to source for tech, digital culture and entertainment content for its dedicated and influential audience around the globe.

### **PROBLEM IN HAND:**

The organization ([www.mashable.com](http://www.mashable.com)) wants to evaluate the popularity of an article and the structure of it. The study is going to help in understanding the key factors that help in making an article a hit.

### **DATA SET INFORMATION:**

This dataset summarizes a heterogeneous set of features about articles published by Mashable in a period of two years.

### **OBJECTIVES OF THE CURRENT STUDY**

- The goal is to predict the number of shares in social networks (popularity).
- To determine the specific predators associated to gauge the popularity a news.

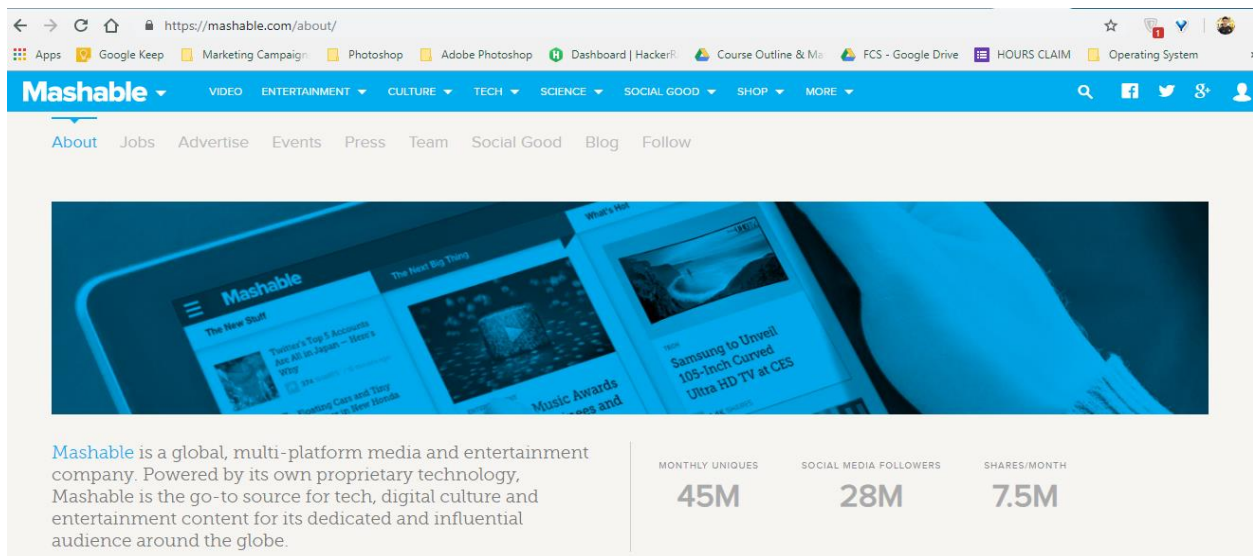
### **OUTLINE OF THE STUDY**

We were assigned a task to develop a statistical report. The topic we have chosen is “Online News Popularity”. This dataset has been picked from an online website named, UCI-Machine Learning Repository. There were all together 61 factors available that were claimed to affect the number of shares of an article.

## CONDUCTING DATA ANALYSIS

In order to systematically conduct data mining analysis, **Cross-Industry Standard Process** (CRISP) is used. CRISP is an industry standard process consisting of a sequence of steps that are usually involved in a data mining study.

### Step 1-BUSINESS UNDERSTANDING



### HOW DOES MASHABLE MAKE MONEY?

Mashable is monetized by offering a variety of advertising formats to its 45M unique monthly readers.

### HOW WILL THIS STUDY BOOST MASHABLE'S BUSINESS?

The study on the respective dataset will figure out the factors that make an article go viral on the internet. Being viral means a solid amount of views, comments, likes, and shares. Keeping in mind the business structure of the organization which earns money via advertisements, hence maximum views on an article means greater advertisement display duration which then results in generation of revenue for the firm. This study will state what factors, for example the use of positive or negative words, images or view content, strong title, timestamp of publication, linked content etc., have the optimum involvement in making an article reach to maximum audience.

## **Step 2- DATA UNDERSTANDING**

We are using R-statistical software to interpret the dataset.

### **ATTRIBUTE INFORMATION:**

Number of Attributes: 61 (58 predictive attributes, 2 non-predictive, 1 goal field)

Attribute Information:

0. url: URL of the article (non-predictive)
1. timedelta: Days between the article publication and the dataset acquisition (non-predictive)
2. n\_tokens\_title: Number of words in the title
3. n\_tokens\_content: Number of words in the content
4. n\_unique\_tokens: Rate of unique words in the content
5. n\_non\_stop\_words: Rate of non-stop words in the content
6. n\_non\_stop\_unique\_tokens: Rate of unique non-stop words in the content
7. num\_hrefs: Number of links
8. num\_self\_hrefs: Number of links to other articles published by Mashable
9. num\_imgs: Number of images
10. num\_videos: Number of videos
11. average\_token\_length: Average length of the words in the content
12. num\_keywords: Number of keywords in the metadata
13. data\_channel\_is\_lifestyle: Is data channel 'Lifestyle'?
14. data\_channel\_is\_entertainment: Is data channel 'Entertainment'?
15. data\_channel\_is\_bus: Is data channel 'Business'?
16. data\_channel\_is\_socmed: Is data channel 'Social Media'?
17. data\_channel\_is\_tech: Is data channel 'Tech'?
18. data\_channel\_is\_world: Is data channel 'World'?
19. kw\_min\_min: Worst keyword (min. shares)
20. kw\_max\_min: Worst keyword (max. shares)
21. kw\_avg\_min: Worst keyword (avg. shares)
22. kw\_min\_max: Best keyword (min. shares)
23. kw\_max\_max: Best keyword (max. shares)
24. kw\_avg\_max: Best keyword (avg. shares)
25. kw\_min\_avg: Avg. keyword (min. shares)
26. kw\_max\_avg: Avg. keyword (max. shares)
27. kw\_avg\_avg: Avg. keyword (avg. shares)
28. self\_reference\_min\_shares: Min. shares of referenced articles in Mashable
29. self\_reference\_max\_shares: Max. shares of referenced articles in Mashable
30. self\_reference\_avg\_shares: Avg. shares of referenced articles in Mashable
31. weekday\_is\_monday: Was the article published on a Monday?
32. weekday\_is\_tuesday: Was the article published on a Tuesday?
33. weekday\_is\_wednesday: Was the article published on a Wednesday?
34. weekday\_is\_thursday: Was the article published on a Thursday?
35. weekday\_is\_friday: Was the article published on a Friday?
36. weekday\_is\_saturday: Was the article published on a Saturday?
37. weekday\_is\_sunday: Was the article published on a Sunday?
38. is\_weekend: Was the article published on the weekend?
39. LDA\_00: Closeness to LDA topic 0
40. LDA\_01: Closeness to LDA topic 1
41. LDA\_02: Closeness to LDA topic 2

- 42. LDA\_03: Closeness to LDA topic 3
- 43. LDA\_04: Closeness to LDA topic 4
- 44. global\_subjectivity: Text subjectivity
- 45. global\_sentiment\_polarity: Text sentiment polarity
- 46. global\_rate\_positive\_words: Rate of positive words in the content
- 47. global\_rate\_negative\_words: Rate of negative words in the content
- 48. rate\_positive\_words: Rate of positive words among non-neutral tokens
- 49. rate\_negative\_words: Rate of negative words among non-neutral tokens
- 50. avg\_positive\_polarity: Avg. polarity of positive words
- 51. min\_positive\_polarity: Min. polarity of positive words
- 52. max\_positive\_polarity: Max. polarity of positive words
- 53. avg\_negative\_polarity: Avg. polarity of negative words
- 54. min\_negative\_polarity: Min. polarity of negative words
- 55. max\_negative\_polarity: Max. polarity of negative words
- 56. title\_subjectivity: Title subjectivity
- 57. title\_sentiment\_polarity: Title polarity
- 58. abs\_title\_subjectivity: Absolute subjectivity level
- 59. abs\_title\_sentiment\_polarity: Absolute polarity level
- 60. shares: Number of shares (target)

### Step 3- DATA PREPARATION

We did few sample t-tests to check the significance that each factor holds and finalized 35 factors that we think are more impactful.

Coefficients:				
	Estimate	Std. Error	t value	Pr(> t )
n_tokens_title	7.182e+01	2.797e+01	2.568	0.010233 *
n_tokens_content	5.364e-01	1.912e-01	2.806	0.005017 **
n_unique_tokens	2.045e+03	8.488e+02	2.409	0.016004 *
num_hrefs	2.968e+01	6.492e+00	4.572	4.84e-06 ***
num_self_hrefs	-6.353e+01	1.730e+01	-3.672	0.000241 ***
num_imgs	1.426e+01	8.251e+00	1.728	0.083916 .
average_token_length	-5.092e+02	1.348e+02	-3.776	0.000160 ***
num_keywords	1.139e+02	3.266e+01	3.488	0.000488 ***
data_channel_is_lifestyle	-1.104e+03	3.886e+02	-2.841	0.004499 **
data_channel_is_entertainment	-1.456e+03	2.431e+02	-5.989	2.13e-09 ***
data_channel_is_bus	-1.128e+03	3.784e+02	-2.981	0.002873 **
data_channel_is_socmed	-7.576e+02	3.645e+02	-2.078	0.037674 *
data_channel_is_tech	-7.935e+02	3.668e+02	-2.163	0.030533 *
data_channel_is_world	-7.546e+02	3.688e+02	-2.046	0.040742 *
kw_avg_min	2.926e-01	1.130e-01	2.591	0.009583 **
kw_max_avg	-1.145e-01	1.954e-02	-5.860	4.66e-09 ***
kw_avg_avg	1.004e+00	9.467e-02	10.602	< 2e-16 ***
self_reference_avg_sharess	2.021e-02	2.442e-03	8.278	< 2e-16 ***
weekday_is_monday	2.786e+02	2.626e+02	1.061	0.288591
weekday_is_tuesday	-2.543e+02	2.587e+02	-0.983	0.325652
weekday_is_wednesday	-1.010e+02	2.587e+02	-0.390	0.286278
weekday_is_thursday	-2.682e+02	2.593e+02	-1.034	0.301024
weekday_is_friday	-2.342e+02	2.686e+02	-0.872	0.0383315 .
weekday_is_saturday	3.727e+02	3.206e+02	1.163	0.0244999 .
LDA_00	1.429e+06	5.949e+05	2.403	0.016285 *
LDA_01	1.429e+06	5.949e+05	2.401	0.016337 *
LDA_02	1.428e+06	5.949e+05	2.400	0.016401 *
LDA_03	1.429e+06	5.949e+05	2.402	0.016292 *
LDA_04	1.429e+06	5.949e+05	2.402	0.016326 *
global_subjectivity	2.649e+03	8.001e+02	3.311	0.000929 ***
global_rate_positive_words	-7.757e+03	4.051e+03	-1.915	0.055530 .

global_rate_negative_words	-3.750e+03	5.942e+03	-0.631	0.0527926.
avg_positive_polarity	-1.350e+03	7.567e+02	-1.784	0.074477 .
avg_negative_polarity	-1.647e+03	5.402e+02	-3.048	0.002306 **
title_sentiment_polarity	3.134e+02	2.247e+02	1.395	0.163039
---				
signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1				

## Step-4 MODELLING

**TYPE OF REGRESSION MODEL:** Observational (values of x's(attributes) are uncontrolled

### Modeling a response:

a) **Multiple Regression:** A Model Relating E(y) for Qualitative Independent Variables

$E(y) = \beta_0 + \beta_1 n\_tokens\_title + \beta_2 n\_tokens\_content + \beta_3 n\_unique\_tokens + \beta_4 num\_hrefs + \beta_5 num\_self\_hrefs + \beta_6 num\_imgs + \beta_7 average\_token\_length + \beta_8 num\_keywords + \beta_9 data\_channel\_is\_lifestyle + \beta_{10} data\_channel\_is\_entertainment + \beta_{11} data\_channel\_is\_bus + \beta_{12} data\_channel\_is\_socmed + \beta_{13} data\_channel\_is\_tech + \beta_{14} data\_channel\_is\_world + \beta_{15} kw\_avg\_min + \beta_{16} kw\_max\_avg + \beta_{17} kw\_avg\_avg + \beta_{18} self\_reference\_avg\_shares + \beta_{19} weekday\_is\_monday + \beta_{20} weekday\_is\_tuesday + \beta_{21} weekday\_is\_wednesday + \beta_{22} weekday\_is\_thursday + \beta_{23} weekday\_is\_friday + \beta_{24} weekday\_is\_saturday + \beta_{25} LDA\_00 + \beta_{26} LDA\_01 + \beta_{27} LDA\_02 + \beta_{28} LDA\_03 + \beta_{29} LDA\_04 + \beta_{30} global\_subjectivity + \beta_{31} global\_rate\_positive\_words + \beta_{32} global\_rate\_negative\_words + \beta_{33} avg\_positive\_polarity + \beta_{34} avg\_negative\_polarity + \beta_{35} title\_sentiment\_polarity + \varepsilon$

Where:

y=number of shares of an article (shares ranging from 1 to 843300)

### Dummy variables

data_channel_is_lifestyle	data_channel_is_entertainment	data_channel_is_bus	data_channel_is_socmed	data_channel_is_tech	data_channel_is_world	weekday_is_monday
1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No	1-If yes 0-If not
Base case: No	Base case: No	Base case: No	Base case: No	Base case: No	Base case: No	Base Case: No



weekday_is_tuesday	weekday_is_wednesday	weekday_is_Thursday	weekday_is_friday	weekday_is_Saturday
1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No	1-If Yes 0-If No
Base case: Good	Base case: No	Base case: No	Base case: No	Base case: No

## MINITAB FOR MULTIPLE REGRESSION

Coefficients:				
	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-1.429e+06	5.950e+05	-2.401	0.016339 *
n_tokens_title	7.182e+01	2.797e+01	2.568	0.010233 *
n_tokens_content	5.364e-01	1.912e-01	2.806	0.005017 **
n_unique_tokens	2.045e+03	8.488e+02	2.409	0.016004 *
num_hrefs	2.968e+01	6.492e+00	4.572	4.84e-06 ***
num_self_hrefs	-6.353e+01	1.730e+01	-3.672	0.000241 ***
num_imgs	1.426e+01	8.251e+00	1.728	0.083916 .
average_token_length	-5.092e+02	1.348e+02	-3.776	0.000160 ***
num_keywords	1.139e+02	3.266e+01	3.488	0.000488 ***
data_channel_is_lifestyleyes	-1.104e+03	3.886e+02	-2.841	0.004499 **
data_channel_is_entertainmentyes	1.456e+03	2.431e+02	-5.989	2.13e-09 ***
data_channel_is_busyes	1.128e+03	3.784e+02	-2.981	0.002873 **
data_channel_is_socmedyes	7.576e+02	3.645e+02	-2.078	0.037674 *
data_channel_is_techyes	7.935e+02	3.668e+02	-2.163	0.030533 *
data_channel_is_worldyes	-7.546e+02	3.688e+02	-2.046	0.040742 *
kw_avg_min	2.926e-01	1.130e-01	2.591	0.009583 **
kw_max_avg	-1.145e-01	1.954e-02	-5.860	4.66e-09 ***
kw_avg_avg	1.004e+00	9.467e-02	10.602	< 2e-16 ***
self_reference_avg_shares	2.021e-02	2.442e-03	8.278	< 2e-16 ***
weekday_is_mondayYes	2.786e+02	2.626e+02	1.061	0.288591
weekday_is_tuesdayYes	-2.543e+02	2.587e+02	-0.983	0.325652
weekday_is_wednesdayYes	-1.010e+02	2.587e+02	-0.390	0.696278
weekday_is_thursdayYes	-2.682e+02	2.593e+02	-1.034	0.301024
weekday_is_fridayYes	-2.342e+02	2.686e+02	-0.872	0.383315
weekday_is_saturdayYes	3.727e+02	3.206e+02	1.163	0.244999

LDA_00	1.429e+06	5.949e+05	2.403	0.016285	*
LDA_01	1.429e+06	5.949e+05	2.401	0.016337	*
LDA_02	1.428e+06	5.949e+05	2.400	0.016401	*
LDA_03	1.429e+06	5.949e+05	2.402	0.016292	*
LDA_04	1.429e+06	5.949e+05	2.402	0.016326	*
global_subjectivity	2.649e+03	8.001e+02	3.311	0.000929	***
global_rate_positive_words	-7.757e+03	4.051e+03	-1.915	0.055530	.
global_rate_negative_words	-3.750e+03	5.942e+03	-0.631	0.527926	
avg_positive_polarity	-1.350e+03	7.567e+02	-1.784	0.074477	.
avg_negative_polarity	-1.647e+03	5.402e+02	-3.048	0.002306	**
title_sentiment_polarity	3.134e+02	2.247e+02	1.395	0.163039	
---					
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1					

## **COEFFICIENT INTERPRETATIONS**

Coefficient of `n_tokens_title`: Keeping all other variables fixed, when `n_tokens_title` is increased by 1 word, the share is increased by  $7.182e+01$  on average.

Coefficient of `n_tokens_content`: Keeping all other variables fixed, when `n_tokens_content` is increased by 1 word, the share is increased by  $5.364e-01$  on average

Coefficient of `n_unique_tokens`: Keeping all other variables fixed, when `n_unique_content` is increased by 1 unique word, the share is increased by  $2.045e+03$  on average

Coefficient of `num_hrefs`: Keeping all other variables fixed, when `num_hrefs` is increased by 1 link, the share is increased by  $2.968e+01$  on average

Coeffocient of `num_self_hrefs`: Keeping all other variables fixed, when `num_self_hrefs` is increased by 1 link, the share is decreased by  $6.353e+01$  on average

Coefficient of `num_imgs`: Keeping all other variables fixed, when `num_images` is increased by 1 image, the share is increased by  $1.426e+01$  on average

Coefficient of `average_token_length`: Keeping all other variables fixed, when `average_token_length` is increased by 1 word, the share is decreased by  $5.092e+02$  on average.

Coefficient of `num_keywords`: Keeping all other variables fixed, when `num_keywords` increased by 1 word, the share is increased  $1.139e+02$  on average.

Coefficient of `data_channel_is_lifestyle`: Keeping all other variables fixed, if an article is about lifestyle the shares will be  $1.104e+03$  less than the shares of an article which is not about lifestyle.

Coefficient of `data_channel_is_entertainment`: Keeping all other variables fixed, if an article is about entertainment the shares will be  $1.456e+03$  more than the shares of an article which is not about entertainment

Coefficient of `data_channel_is_bus`: Keeping all other variables fixed, if an article is about business the shares will be  $1.128e+03$  more than the shares of an article which is not about business

Coefficient of `data_channel_is_socmed`: Keeping all other variables fixed, if an article is about social media the shares will be  $7.576e+02$  more than the shares of an article which is not about social media

Coefficient of `data_channel_is_tech`: Keeping all other variables fixed, if an article is about technology the shares will be  $7.935e+02$  more than the shares of an article which is not about technology

Coefficient of data\_channel\_is\_world: Keeping all other variables fixed, if an article is about world affairs the shares will be  $-7.546 \times 10^2$  less than the shares of an article which is not about world affairs

### MODEL SUMMARY

Residual standard error: 11510 on 39608 degrees of freedom	
Multiple R-squared: 0.6627	Adjusted R-squared: 0.6041

The value  $R^2 = .5527$  is highlighted on the printout. This implies that using the independent variables, the model explains 55.27% of the total sample variation in shares,  $y$ . Thus,  $R^2$  is a sample statistic that tells how well the model fits the data and thereby represents a measure of the usefulness of the entire model.

### ANOVA

F-statistic: 24.6 on 35 and 39608 DF, p-value: $< 2.2 \times 10^{-16}$
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### TESTING THE UTILITY OF A MODEL: THE ANALYSIS OF VARIANCE F-TEST

**Null hypothesis:**  $\beta_1 = \beta_2 = \beta_3 = \dots = \beta_{35} = 0$

**Alternate hypotheses:** At least one of the coefficients is non-zero

The test statistic used to test this hypothesis is an F statistic, the statistical software calculates the F statistic):

$$\text{Test statistic: } F = (SS_{yy} - SSE) / k / SSE / [n - (k + 1)] = 24.6$$

**Conclusion:** Since  $p\text{-value} < \text{level of significance} = 0.05$  we will reject null hypothesis and conclude that at least one of coefficient is non-zero. (conclusion based on p-value given in the table)

## MODEL 2:

### b) Interaction Model: An interaction model with qualitative predictors

$$E(y) = \beta_0 + \beta_1 n\_tokens\_title + \beta_2 n\_tokens\_content + \beta_3 n\_unique\_tokens + \beta_4 num\_hrefs + \beta_5 num\_self\_hrefs + \beta_6 num\_imgs + \beta_7 average\_token\_length + \beta_8 num\_keywords + \beta_9 data\_channel\_is\_lifestyle + \beta_{10} data\_channel\_is\_entertainment + \beta_{11} data\_channel\_is\_bus + \beta_{12} data\_channel\_is\_socmed + \beta_{13} data\_channel\_is\_tech + \beta_{14} data\_channel\_is\_world + \beta_{15} kw\_avg\_min + \beta_{16} kw\_max\_avg + \beta_{17} kw\_avg\_avg + \beta_{18} self\_reference\_avg\_shares + \beta_{19} weekday\_is\_monday + \beta_{20} weekday\_is\_tuesday + \beta_{21} weekday\_is\_wednesday + \beta_{22} weekday\_is\_thursday + \beta_{23} weekday\_is\_friday + \beta_{24} weekday\_is\_saturday + \beta_{25} LDA\_00 + \beta_{26} LDA\_01 + \beta_{27} LDA\_02 + \beta_{28} LDA\_03 + \beta_{29} LDA\_04 + \beta_{30} global\_subjectivity + \beta_{31} global\_rate\_positive\_words + \beta_{32} global\_rate\_negative\_words + \beta_{33} avg\_positive\_polarity + \beta_{34} avg\_negative\_polarity + \beta_{35} num\_hrefs * num\_self\_hrefs + \beta_{36} n\_tokens\_content * n\_unique\_tokens + \epsilon$$

Coefficients:			
	Estimate	Std. Error	t value Pr(> t )
(Intercept)	5.146e+06	1.639e+06	3.139 0.001695 **
n_tokens_title	7.174e+01	2.797e+01	2.565 0.010319 *
n_tokens_content	2.503e+00	4.973e-01	5.033 4.84e-07 ***
n_unique_tokens	1.996e+03	8.500e+02	2.348 0.018884 *
num_hrefs	3.312e+01	7.358e+00	4.501 6.77e-06 ***
num_self_hrefs	-7.287e+01	2.555e+01	-2.852 0.004344 **
num_imgs	8.351e+00	8.372e+00	0.997 0.318543
average_token_length	-4.309e+02	1.366e+02	-3.155 0.001607 **
num_keywords	1.107e+02	3.271e+01	3.382 0.000719 ***
data_channel_is_lifestyle	-1.054e+03	3.888e+02	-2.711 0.006702 **
data_channel_is_entertainment	-1.350e+03	2.443e+02	-5.528 3.27e-08 ***
data_channel_is_bus	-1.111e+03	3.784e+02	-2.937 0.003317 **
data_channel_is_socmed	-7.849e+02	3.645e+02	-2.154 0.031284 *
data_channel_is_tech	-7.697e+02	3.669e+02	-2.098 0.035929 *
data_channel_is_world	-6.786e+02	3.691e+02	-1.838 0.066014 .
kw_avg_min	2.887e-01	1.129e-01	2.556 0.010595 *
kw_max_avg	-1.136e-01	1.954e-02	-5.813 6.17e-09 ***
kw_avg_avg	9.986e-01	9.467e-02	10.548 < 2e-16 ***

self_reference_avg_sharess	2.022e-02	2.441e-03	8.285	<2e-16	***
weekday_is_monday	2.921e+02	2.625e+02	1.113	0.265894	
weekday_is_tuesday	-2.459e+02	2.587e+02	-0.951	0.341809	
weekday_is_wednesday	-8.789e+01	2.587e+02	-0.340	0.734001	
weekday_is_thursday	-2.616e+02	2.592e+02	-1.009	0.312956	
weekday_is_friday	-2.282e+02	2.686e+02	-0.849	0.395624	
weekday_is_saturday	3.889e+02	3.205e+02	1.213	0.225047	
LDA_00	-5.145e+06	1.639e+06	-3.139	0.001697	**
LDA_01	-5.146e+06	1.639e+06	-3.139	0.001694	**
LDA_02	-5.147e+06	1.639e+06	-3.140	0.001691	**
LDA_03	-5.145e+06	1.639e+06	-3.139	0.001696	**
LDA_04	-5.146e+06	1.639e+06	-3.139	0.001695	**
global_subjectivity	2.608e+03	7.999e+02	3.260	0.001116	**
global_rate_positive_words	-7.131e+03	4.053e+03	-1.759	0.078501	.
global_rate_negative_words	-1.678e+03	5.963e+03	-0.281	0.778424	
avg_positive_polarity	-1.266e+03	7.569e+02	-1.673	0.094426	.
avg_negative_polarity	-1.828e+03	5.417e+02	-3.374	0.000742	***
title_sentiment_polarity	3.159e+02	2.247e+02	1.406	0.159685	
num_hrefs:num_self_hrefs	4.030e-01	5.617e-01	0.717	0.0473090	*
n_tokens_content:n_unique_tokens	-5.945e+00	1.388e+00	-4.282	1.86e-05	***

## ANALYSIS OF VARIANCE

**Residual standard error: 11510 on 39606 degrees of freedom**

**Multiple R-squared: 0.6674, Adjusted R-squared: 0.6011**

**F-statistic: 23.8 on 37 and 39606 DF, p-value: < 2.2e-16**

## TESTING THE OVERALL UTILITY OF MODEL USING THE GLOBAL F-TEST AT $\alpha = .05$

The global F-test is used to test the null hypothesis

**Null hypothesis:**  $\beta_1 = \beta_2 = \beta_3 = \dots = \beta_{37} = 0$

**Alternate hypotheses:** At least one of the coefficients is non-zero

The test statistic and p-value of the test (highlighted on the MINITAB printout) are  $F = 23.8$  and  $p = 2.2e-16$ , respectively. Since  $\alpha = .05$  exceeds the p-value, there is sufficient evidence to conclude that the model fit is a statistically useful predictor of shares,  $y$ . Reject null hypothesis.

## Step 5- EVALUATION

After observing both the models:

When we looked at the adjusted R-squared value for both the regression models, they are relatively close, but since the interaction test is insignificant we may declare model A to be more suitable to predict the shares/ popularity of an article

## CONCLUSION

### **Selected Model- Model A:**

$$E(y) = \beta_0 + \beta_1 n\_tokens\_title + \beta_2 n\_tokens\_content + \beta_3 n\_unique\_tokens + \beta_4 num\_hrefs + \beta_5 num\_self\_hrefs + \beta_6 num\_imgs + \beta_7 average\_token\_length + \beta_8 num\_keywords + \beta_9 data\_channel\_is\_lifestyle + \beta_{10} data\_channel\_is\_entertainment + \beta_{11} data\_channel\_is\_bus + \beta_{12} data\_channel\_is\_socmed + \beta_{13} data\_channel\_is\_tech + \beta_{14} data\_channel\_is\_world + \beta_{15} kw\_avg\_min + \beta_{16} kw\_max\_avg + \beta_{17} kw\_avg\_avg + \beta_{18} self\_reference\_avg\_shares + \beta_{19} weekday\_is\_monday + \beta_{20} weekday\_is\_tuesday + \beta_{21} weekday\_is\_wednesday + \beta_{22} weekday\_is\_thursday + \beta_{23} weekday\_is\_friday + \beta_{24} weekday\_is\_saturday + \beta_{25} LDA\_00 + \beta_{26} LDA\_01 + \beta_{27} LDA\_02 + \beta_{28} LDA\_03 + \beta_{29} LDA\_04 + \beta_{30} global\_subjectivity + \beta_{31} global\_rate\_positive\_words + \beta_{32} global\_rate\_negative\_words + \beta_{33} avg\_positive\_polarity + \beta_{34} avg\_negative\_polarity + \beta_{35} title\_sentiment\_polarity + \epsilon$$

## CROSS VALIDATING A MODEL A:

To generate training and testing data for cross validating online news popularity model:  
70% of the data is trained using model A. The rest 30% is referred as test data. Next, the model is build using test data to predict shares.

**Correlation table of actual and predicted for cross validating sample**

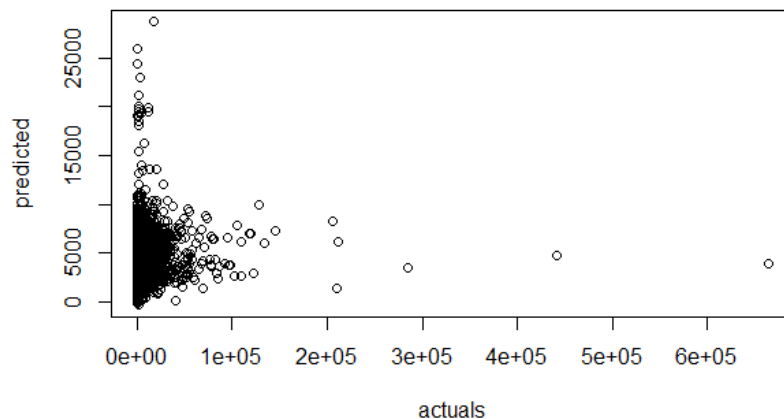
	actuals	predicted
actuals	1.0000000	0.6580839
predicted	0.6580839	1.0000000

We can say that the model selected i.e. **MODEL A**, is 65.8% good in giving accurate results.

## Step-6 DEPOLYMENT

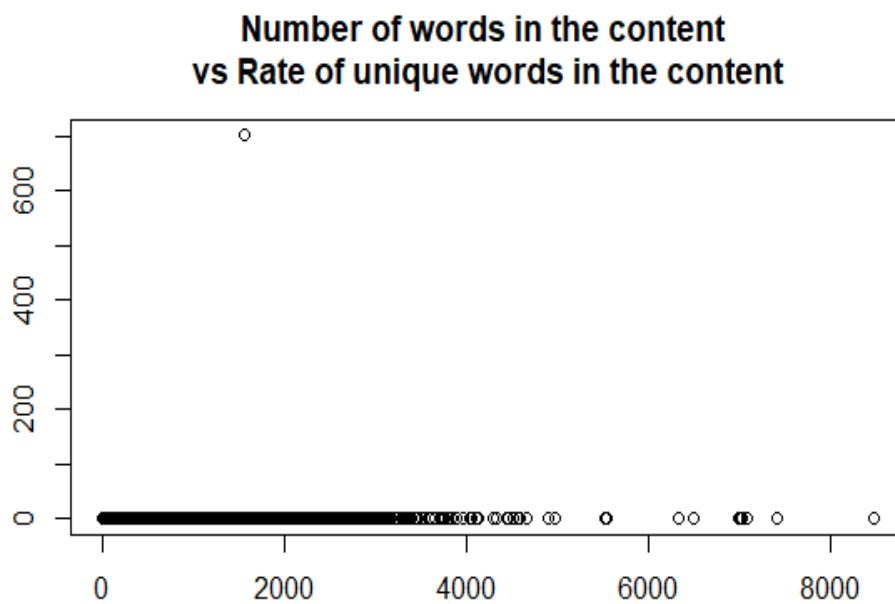
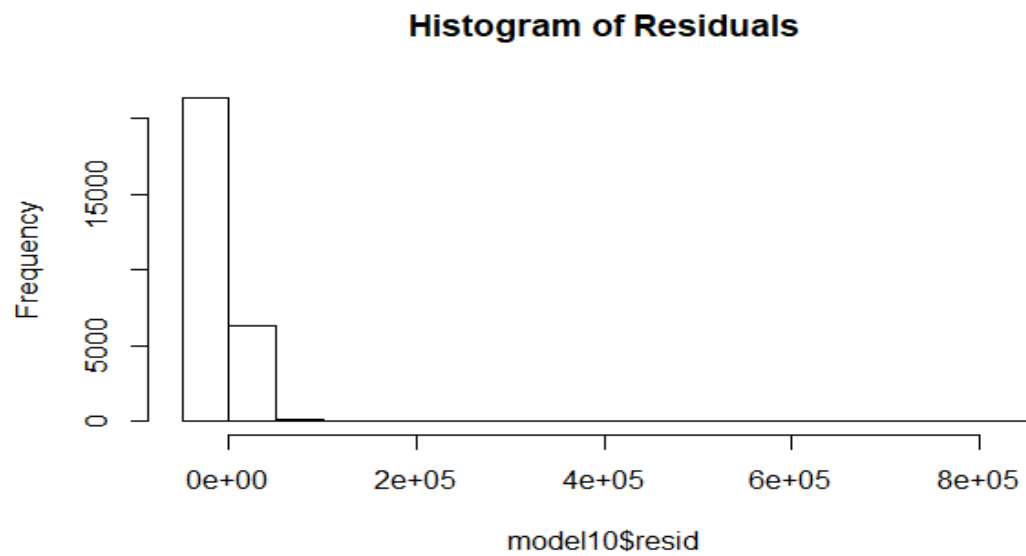
The model is finally ready for deployment

## DATA REPRESENTATION USING MODEL A



Scatter plot for predicted shares vs actual shares





## REFERENCES

1. Pedro Vinagre (pedro.vinagre.sousa@ptgmail.com) - ALGORITMI Research Centre, Universidade do Minho, Portugal
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