

# AirBnB



Group 10

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# Introduction



## Company Overview

- \$3.4 billion in revenue in 2020
- 193 million bookings in Airbnb in 2020
- Over 7 million listings in Airbnb, run by 4 million hosts

## Objective

Predict if a property will receive a good or bad rating based on sentiment analysis and average rating score.



# Value Proposition of Project



**Small Airbnb Managers:**

Determine most important attributes

Compare property to competition



**Large Airbnb Managers:**

Clarify user rating drivers

Guide purchasing decisions

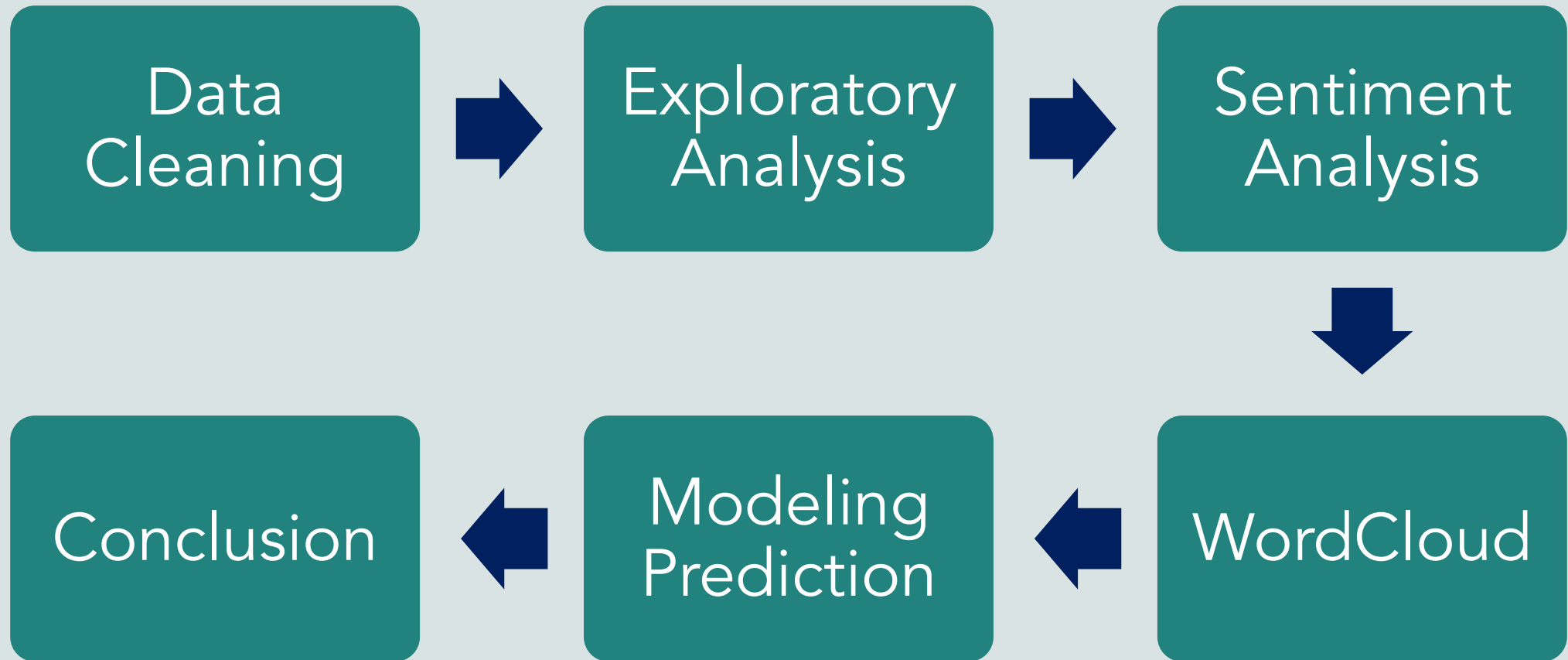


**Airbnb Corporate:**

Vet aspiring Airbnb hosts

Improve recommendation algorithm

# Process Overview



# Data Cleaning |





# Original Dataset

- Two datasets - 'Listings' and 'Reviews'
  - Combined into one
  - Reviews - 876,200 entries
  - Listings - 37,713 properties
- 43+ columns/attributes in combined dataset
- Reviews in multiple languages
- "Corrupted" reviews containing non-character symbols like <sup>TM</sup>, ®, & etc



# Dataset Pre-processing

---

1

Remove all non strings, digits and symbols

2

Remove punctuation, white spaces and converting upper case to lower case

3

Remove Stopwords using nltk stopwords

4

Remove null values

# Cleaned Dataset



- 15,722 rows and two columns in 'Reviews'
- 12,876 properties in 'Listings'
- 42 columns in 'Listings'



10 variables included from 2 different datasets, namely Listings and Reviews, such as:

Cleanliness  
Subjectivity  
Polarity  
Value  
Rent



# Exploratory Analysis

```
object to mirror_ob  
mirror_mod.mirror_object = mirror_ob  
operation == "MIRROR_X":  
    mirror_mod.use_x = True  
    mirror_mod.use_y = False  
    mirror_mod.use_z = False  
operation == "MIRROR_Y":  
    mirror_mod.use_x = False  
    mirror_mod.use_y = True  
    mirror_mod.use_z = False  
operation == "MIRROR_Z":  
    mirror_mod.use_x = False  
    mirror_mod.use_y = False  
    mirror_mod.use_z = True
```

```
selection at the end -add back the deselected  
ob.select= 1  
ob.select=1  
context.scene.objects.active = modifier_ob  
selected" + str(modifier_ob)) # modifier ob  
mirror_ob.select = 0  
context.selected_objects[0]  
context.objects[one.name].select = 1
```

```
print("please select exactly two objects, %s" % len(context.selected_objects))
```

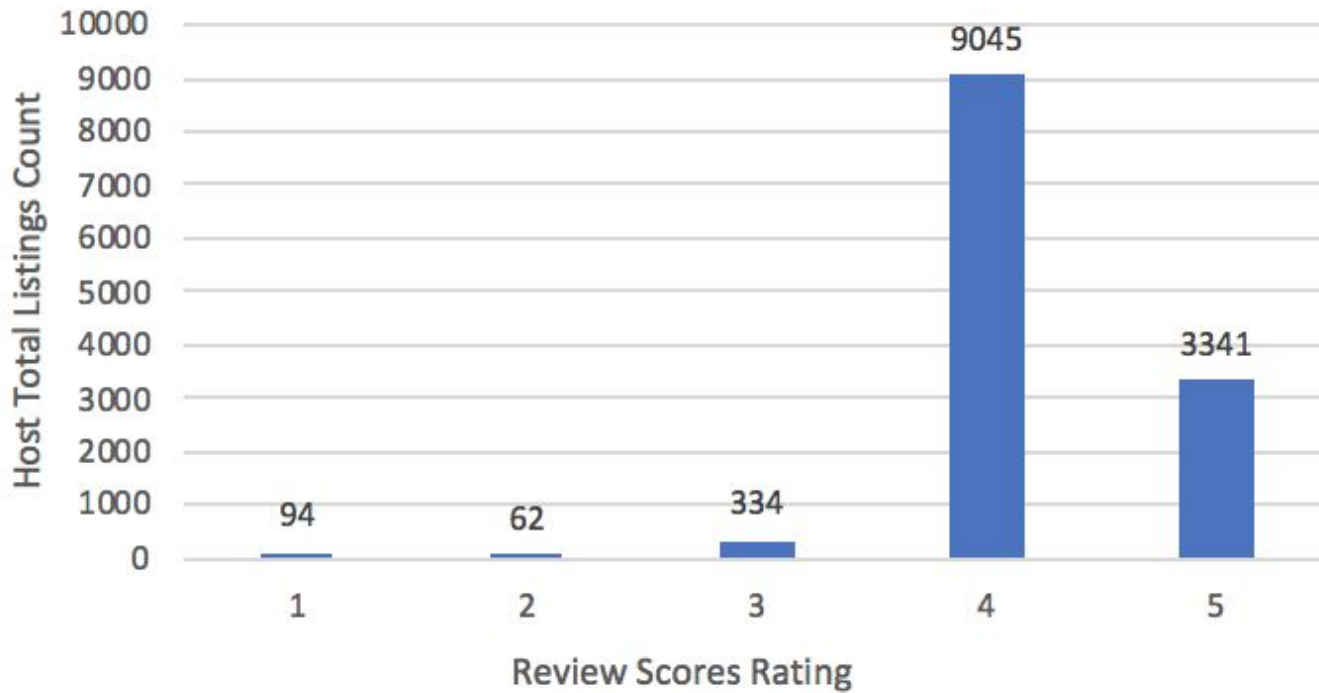
OPERATOR CLASSES -----

```
class MirrorOperator(bpy.types.Operator):  
    """Mirror to the selected object"""  
    bl_idname = "mirror_mirror_x"  
    bl_label = "Mirror X"
```

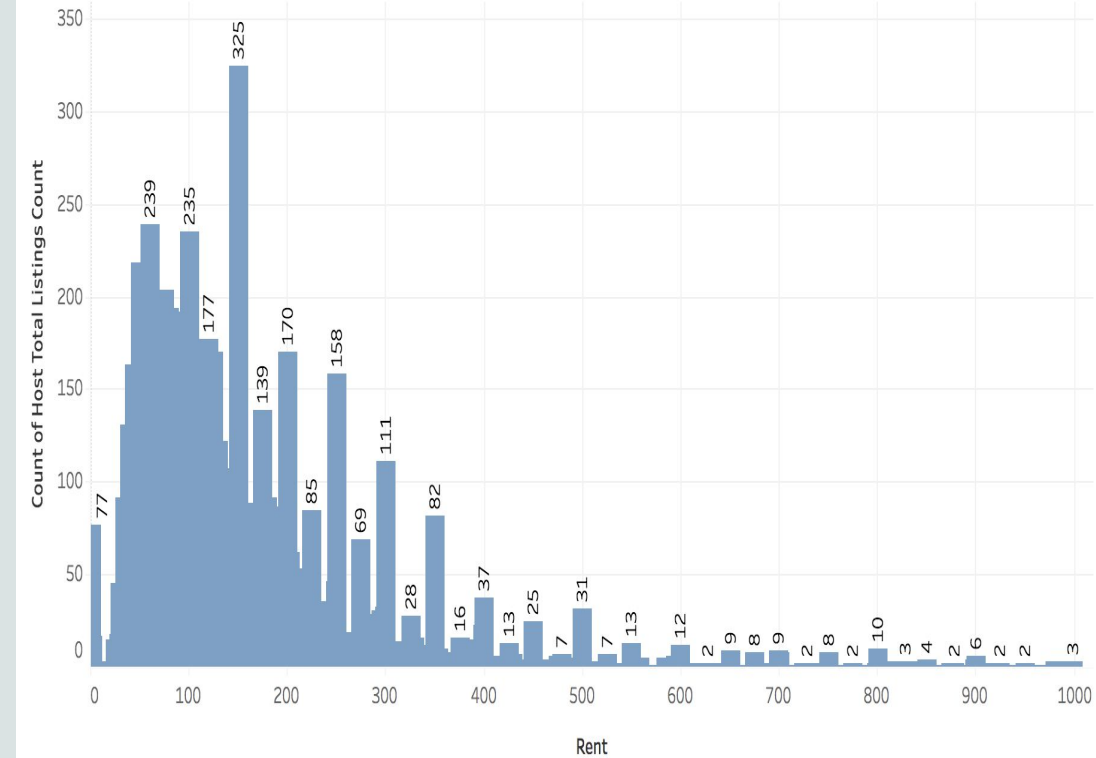
```
def execute(self, context):  
    if context.selected_objects[0] is not None
```

# Demographic Visualizations

## Review Rating Distribution for New York Hosts

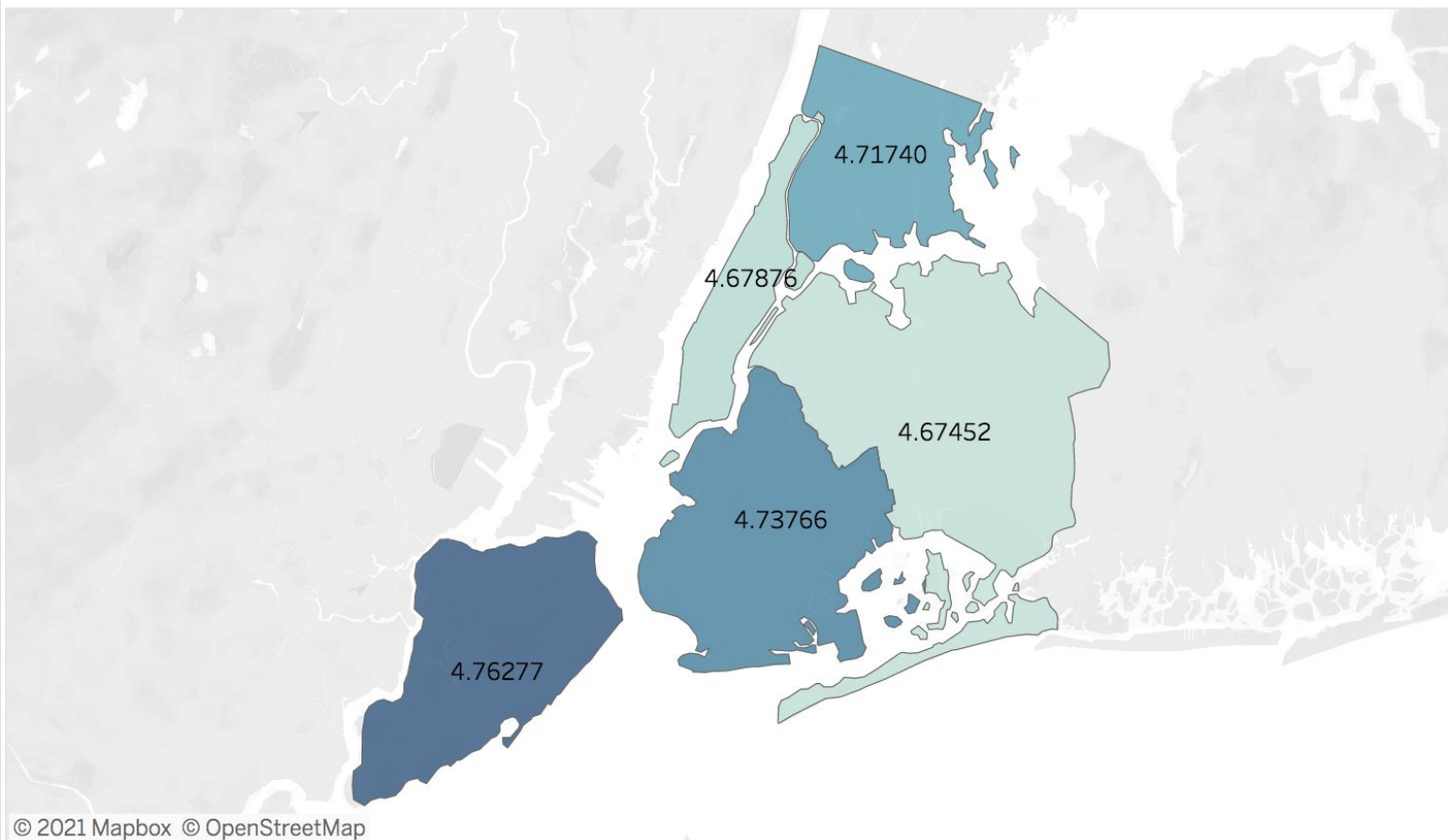


## Rent Distribution for New York Hosts



# Demographic Visualizations

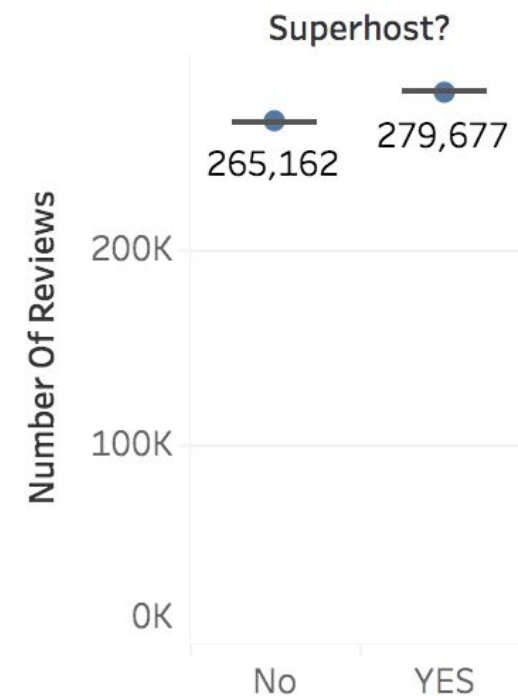
Average Rating Score for New York Region



Avg. Review Scores Rating



# of Reviews Comparison for Hosts





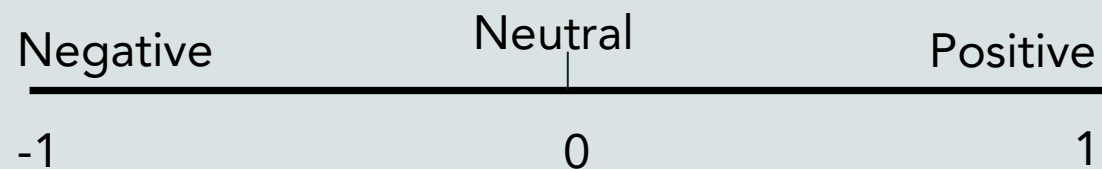
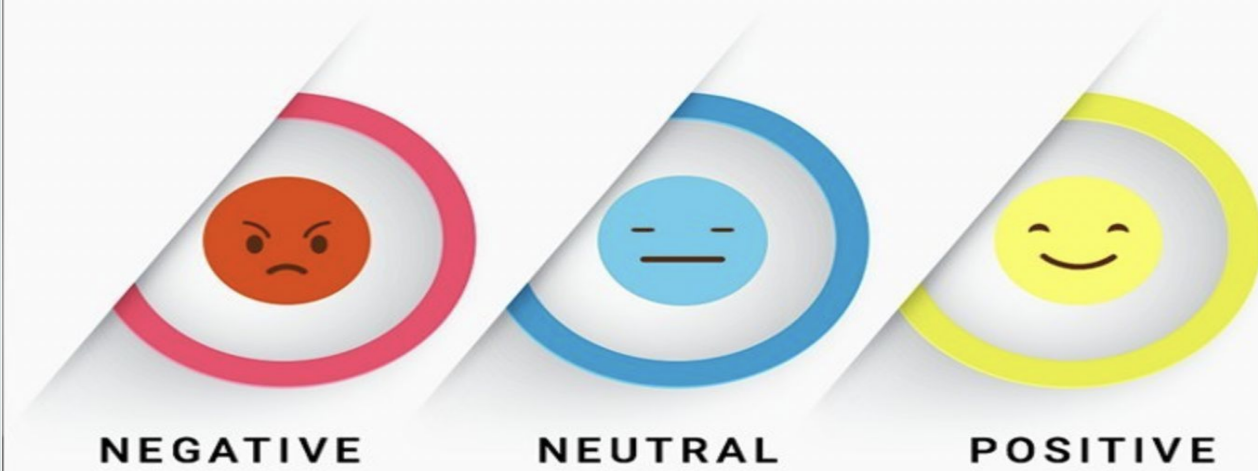
# Sentiment Analysis

Textblob

Polarity

Subjectivity

## SENTIMENT ANALYSIS

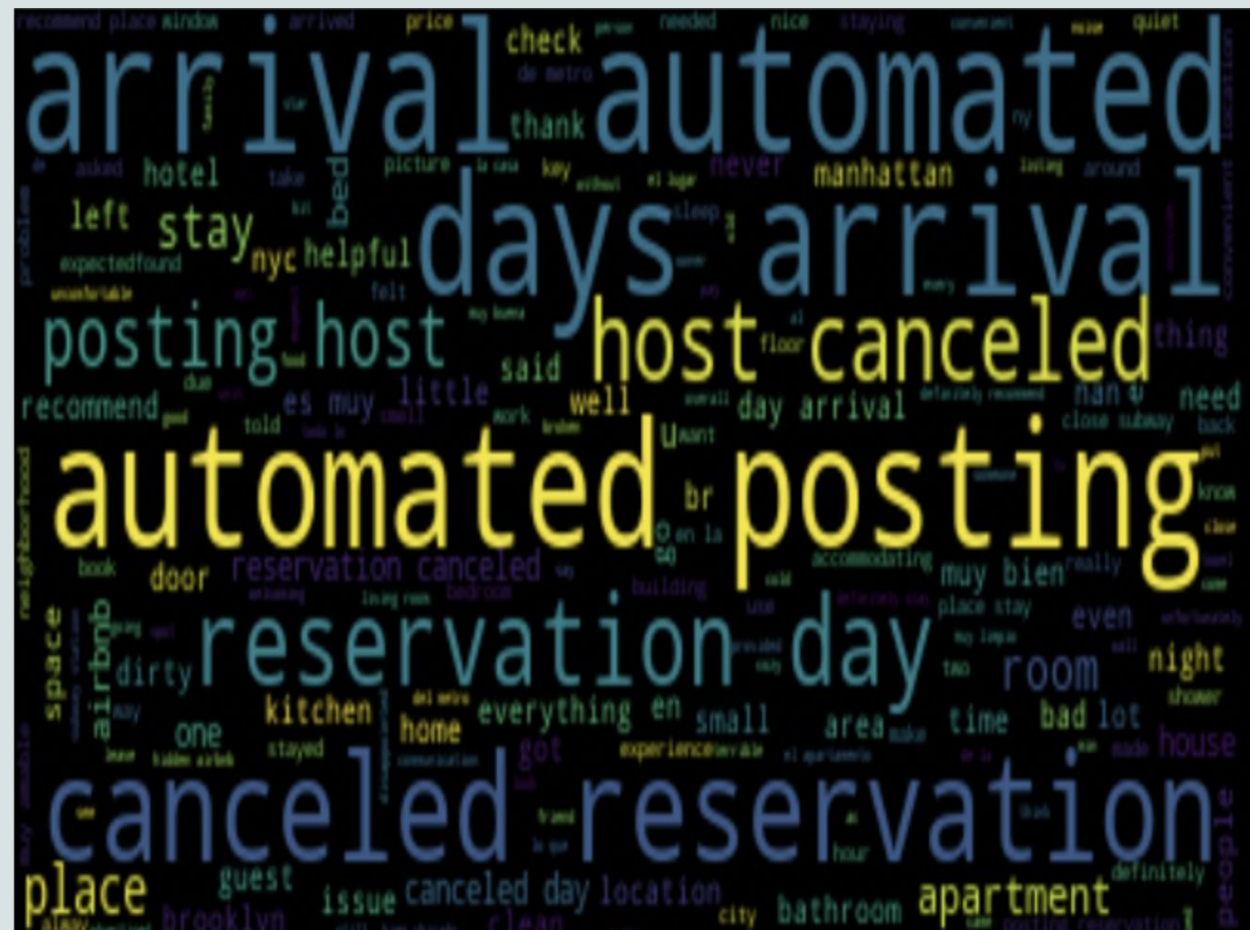


# WordCloud

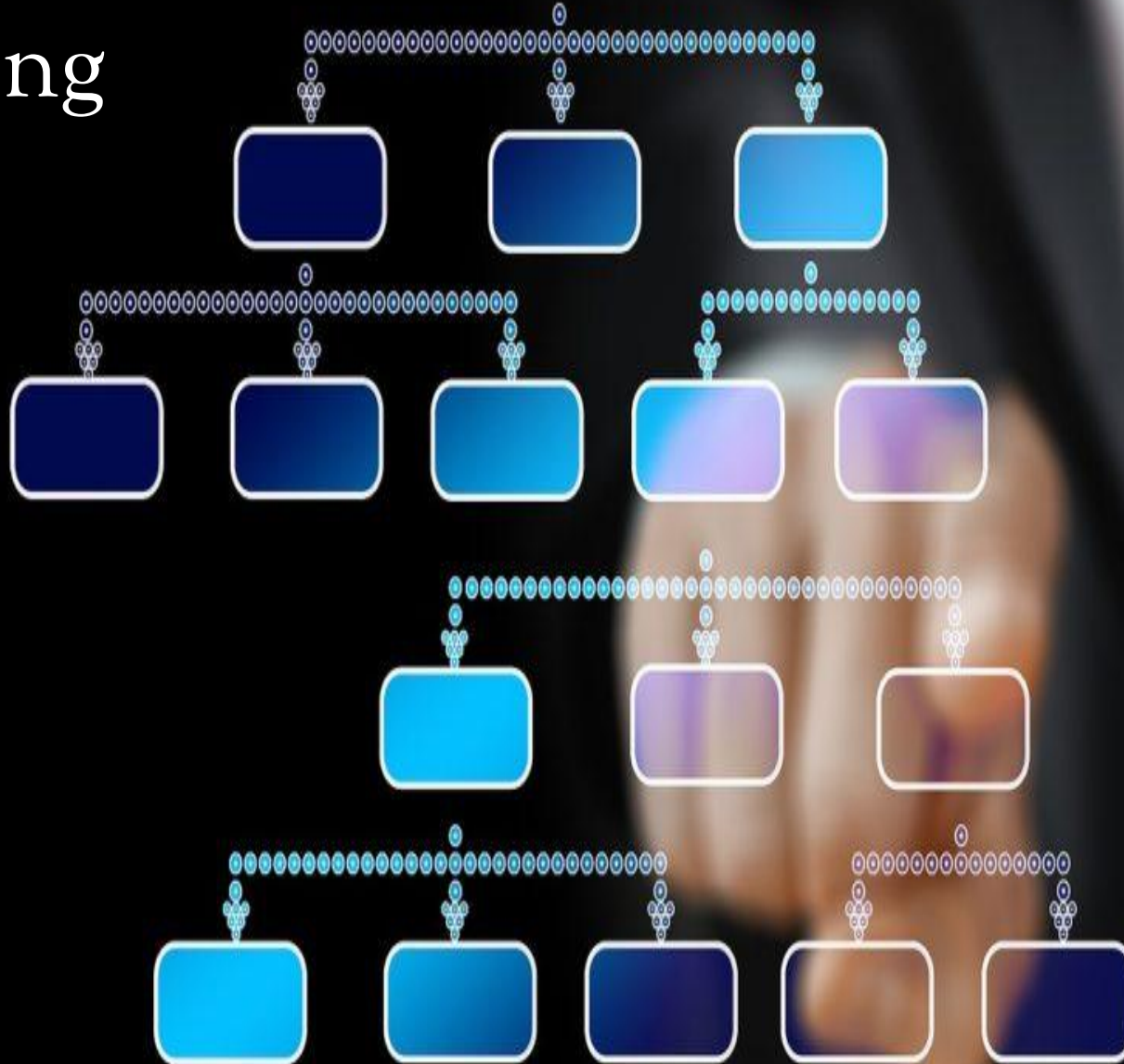
# Positive



# Negative



# Data Modelling

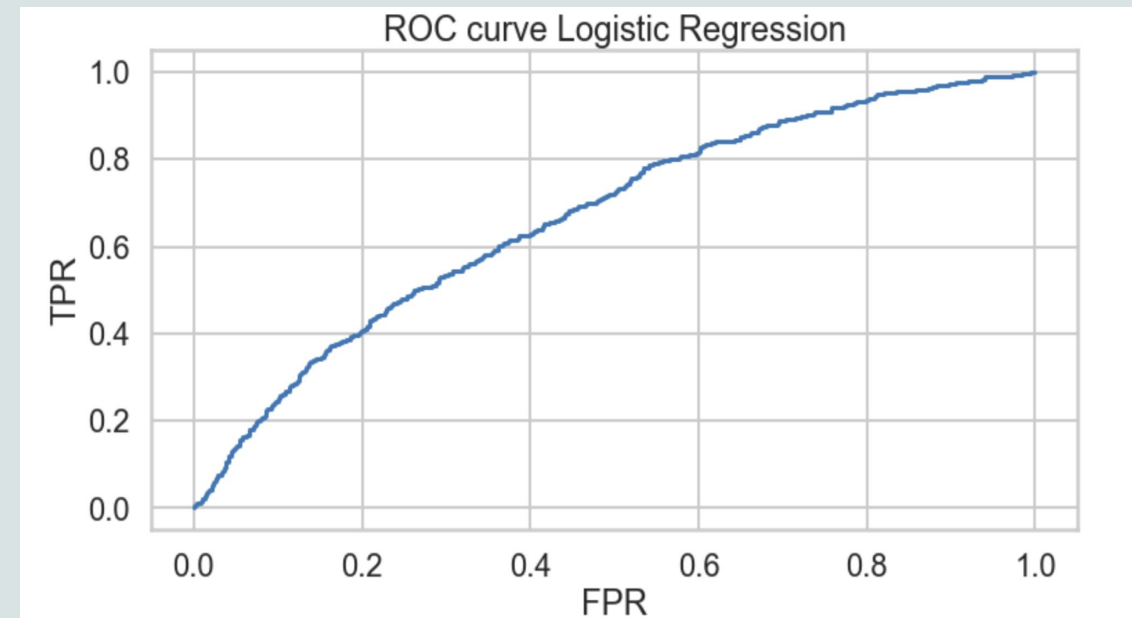




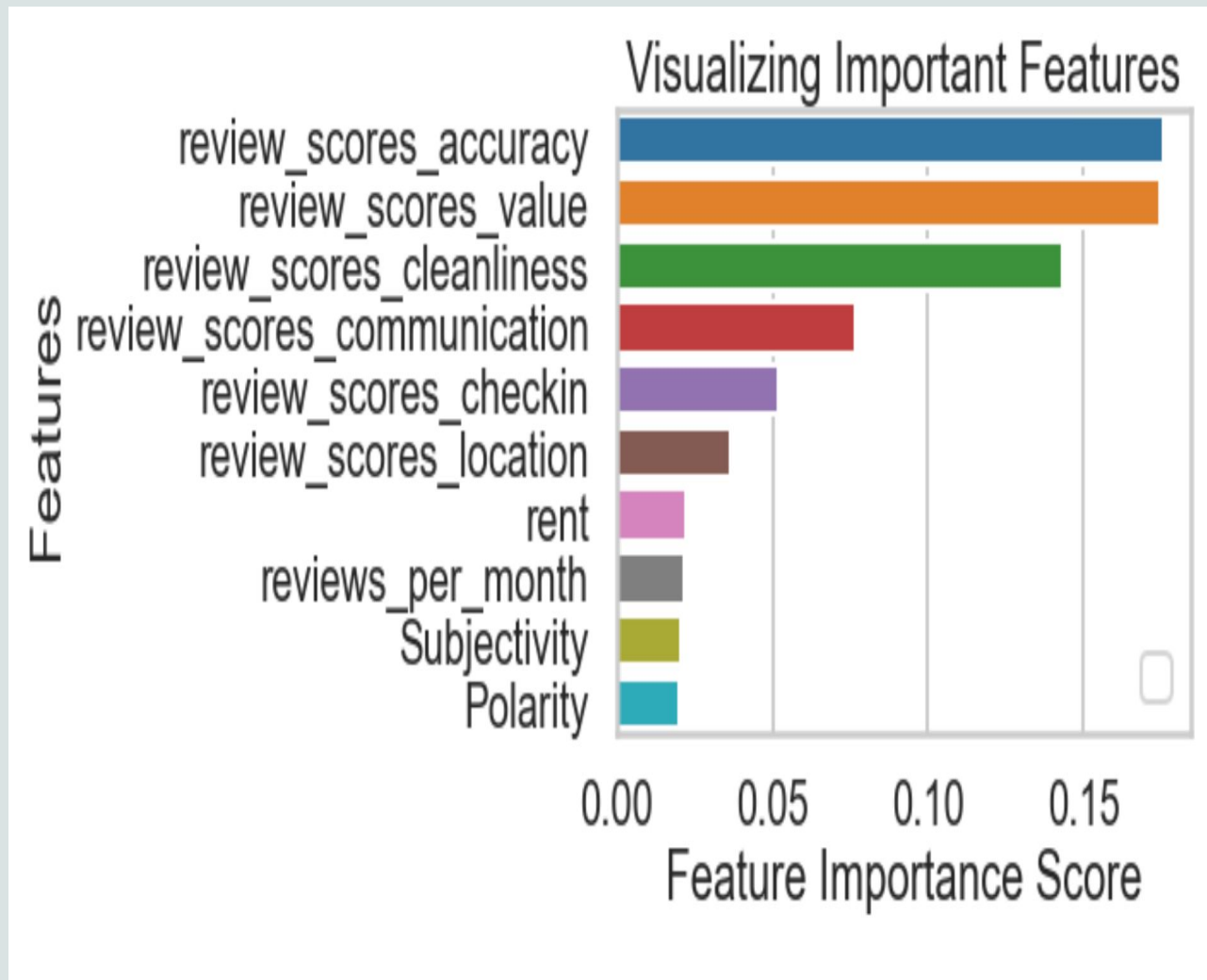
# Logistic Regression

- Without processing the data
- Benchmark model

	precision	recall	f1-score	support
0	0.69	0.94	0.80	1121
1	0.58	0.16	0.25	553
accuracy			0.68	1674
macro avg	0.64	0.55	0.52	1674
weighted avg	0.66	0.68	0.62	1674



# Feature Selection

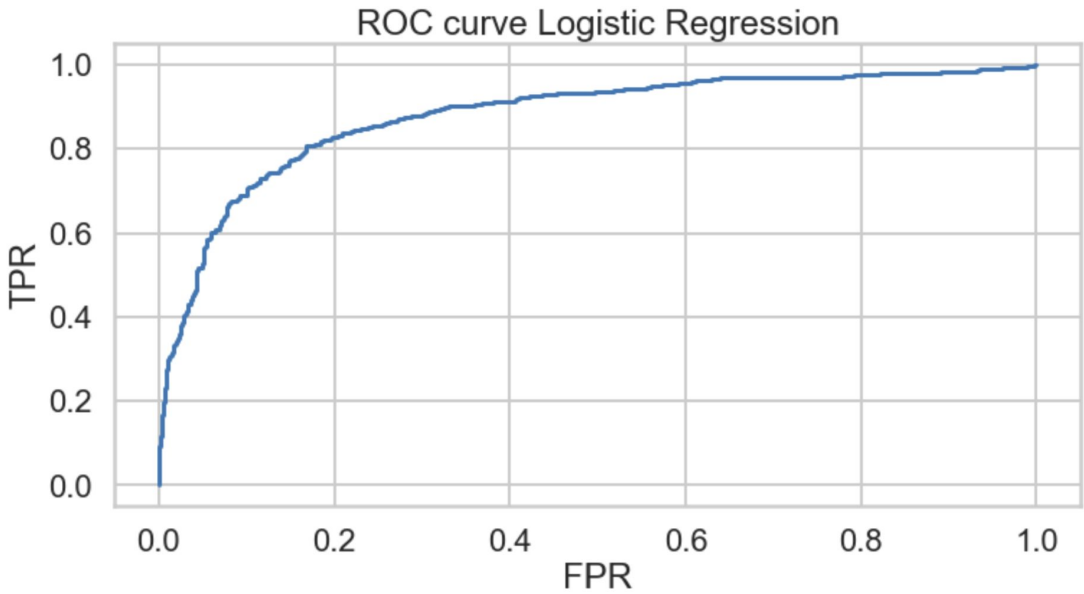


# Logistic Regression with processed data

Optimization terminated successfully.  
Current function value: 0.292887  
Iterations 8

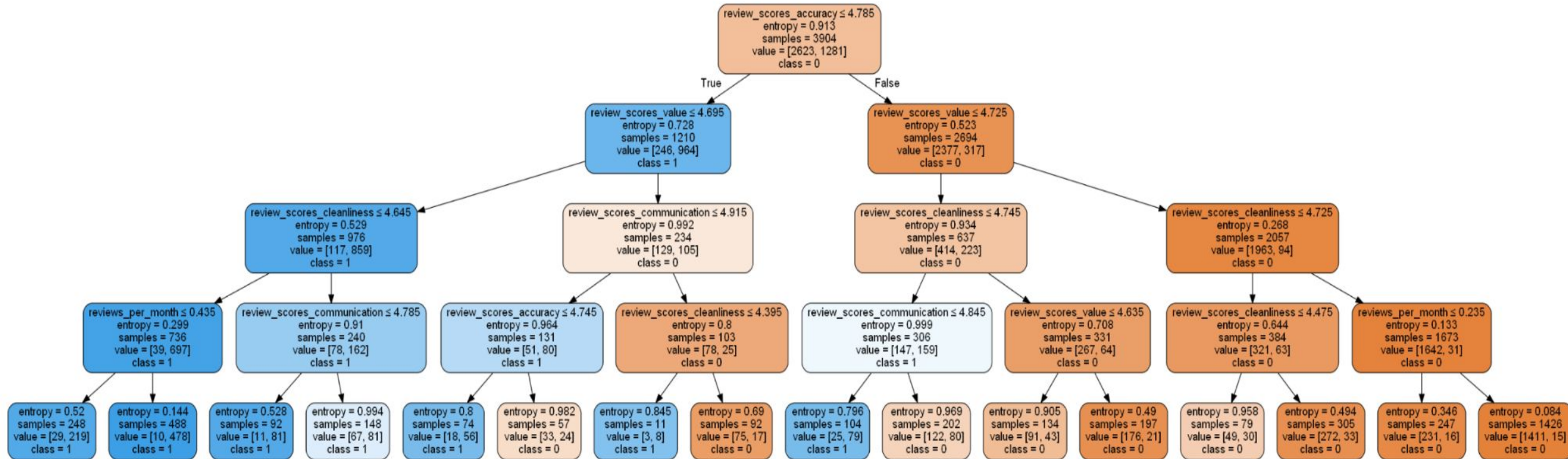
Logit Regression Results						
Dep. Variable:	Ratings	No. Observations:	5578			
Model:	Logit	Df Residuals:	5568			
Method:	MLE	Df Model:	9			
Date:	Mon, 29 Nov 2021	Pseudo R-squ.:	0.5375			
Time:	18:11:46	Log-Likelihood:	-1633.7			
converged:	True	LL-Null:	-3532.7			
Covariance Type:	nonrobust	LLR p-value:	0.000			
	coef	std err	z	P> z	[0.025	0.975]
Intercept	97.9725	3.001	32.648	0.000	92.091	103.854
review_scores_checkin	-1.8481	0.384	-4.819	0.000	-2.600	-1.096
reviews_per_month	0.0454	0.016	2.781	0.005	0.013	0.077
review_scores_accuracy	-5.9947	0.414	-14.463	0.000	-6.807	-5.182
review_scores_cleanliness	-3.9142	0.221	-17.718	0.000	-4.347	-3.481
review_scores_communication	-3.8983	0.415	-9.403	0.000	-4.711	-3.086
review_scores_location	-1.6668	0.203	-8.225	0.000	-2.064	-1.270
review_scores_value	-3.4218	0.290	-11.815	0.000	-3.989	-2.854
Subjectivity	0.8386	1.064	0.788	0.431	-1.247	2.925
Polarity	-0.5390	0.940	-0.574	0.566	-2.381	1.302

	precision	recall	f1-score	support
0	0.82	0.95	0.88	1121
1	0.84	0.57	0.68	553
accuracy			0.82	1674
macro avg	0.83	0.76	0.78	1674
weighted avg	0.82	0.82	0.81	1674

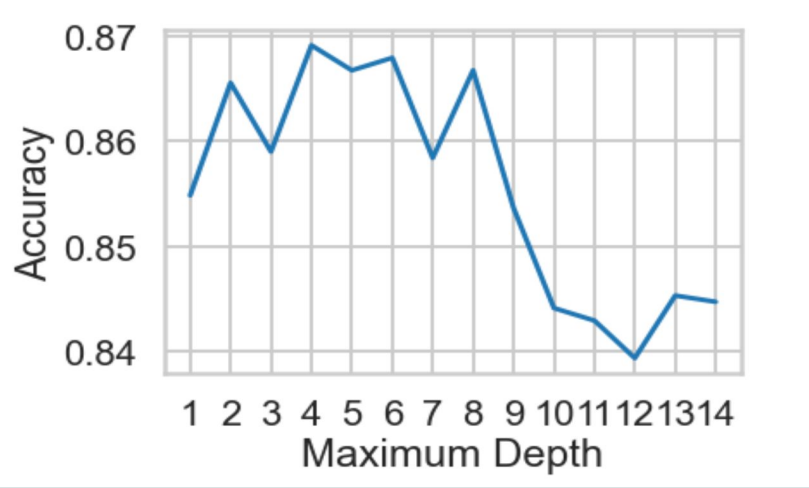




# Decision Tree Classifier



# Decision Tree Classifier

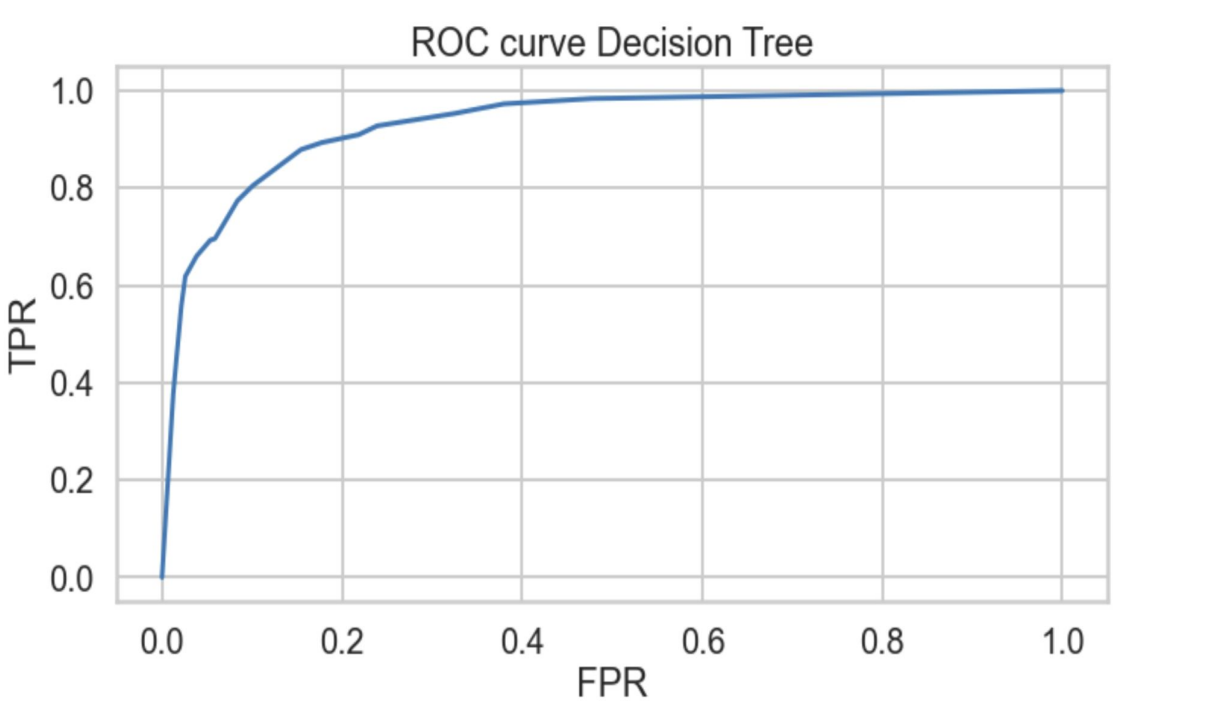


Accuracy: 0.8691756272401434

Precision: 0.8199233716475096

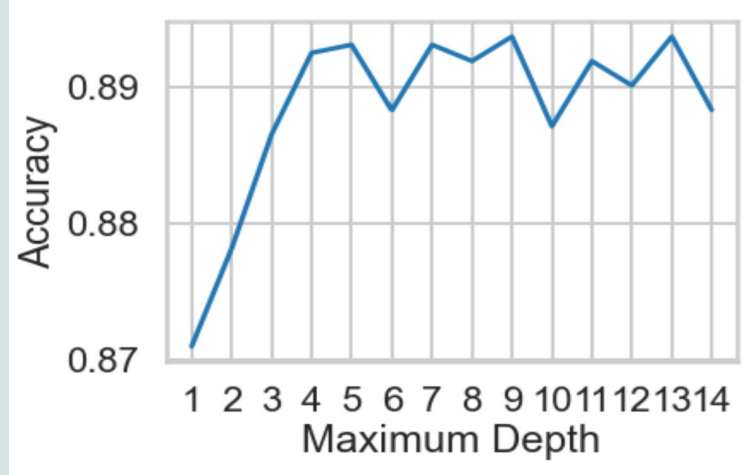
Recall: 0.7739602169981917

	precision	recall	f1-score	support
0	0.89	0.92	0.90	1121
1	0.82	0.77	0.80	553
accuracy			0.87	1674
macro avg	0.86	0.85	0.85	1674
weighted avg	0.87	0.87	0.87	1674



AUC - 0.9316823489747754

# Random Forest

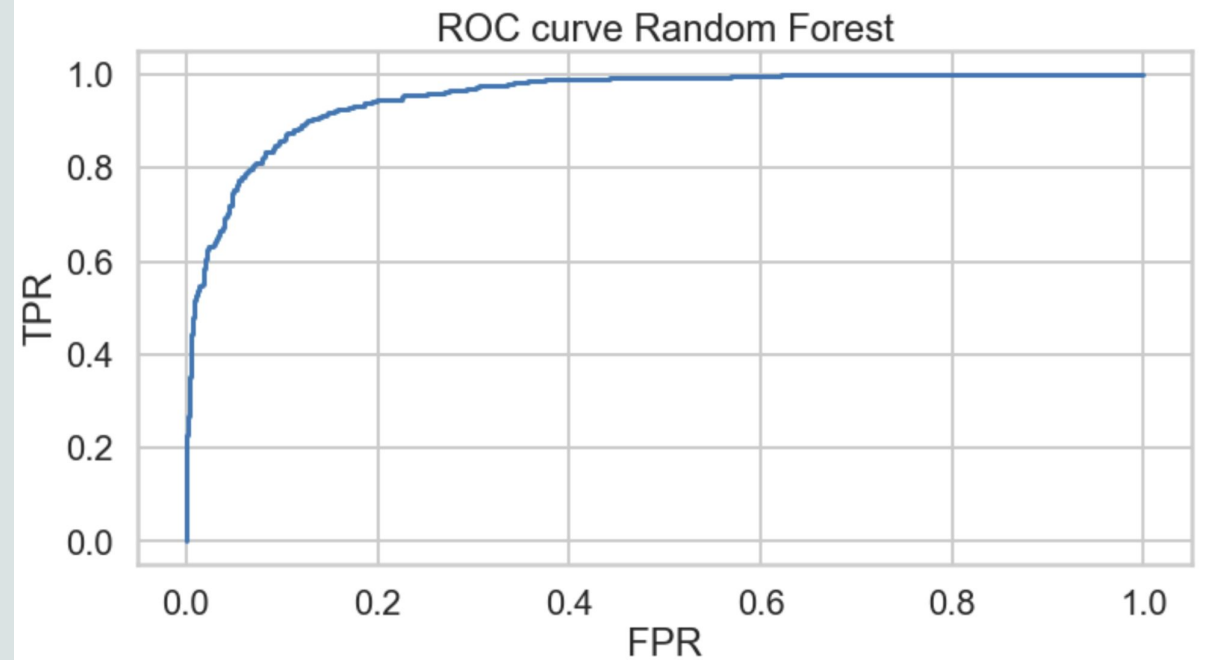


Accuracy: 0.8918757467144564

Precision: 0.8381818181818181

Recall: 0.833634719710669

	precision	recall	f1-score	support
0	0.92	0.92	0.92	1121
1	0.84	0.83	0.84	553
accuracy			0.89	1674
macro avg	0.88	0.88	0.88	1674
weighted avg	0.89	0.89	0.89	1674



AUC - 0.9536483345243607



# Conclusion

	Precision	Recall	F1-Score	AUC
Logistic Regression(BenchMark)	0.58	0.16	0.25	0.669504
Logistic Regression	0.84	0.57	0.68	0.880348
Decision Tree	0.82	0.77	0.8	0.931682
Random Forest	0.84	0.83	0.84	0.953648

## KEY TAKEAWAYS

- Changing hyperparameters can change the accuracy of your model
- Avoid overfitting by limiting number of features/attributes used in prediction models
- We must make a decision regarding what attributes to use in model
  - Sentiment analysis did not rate highly in significance but was integral to what we set out to do

[illegible]