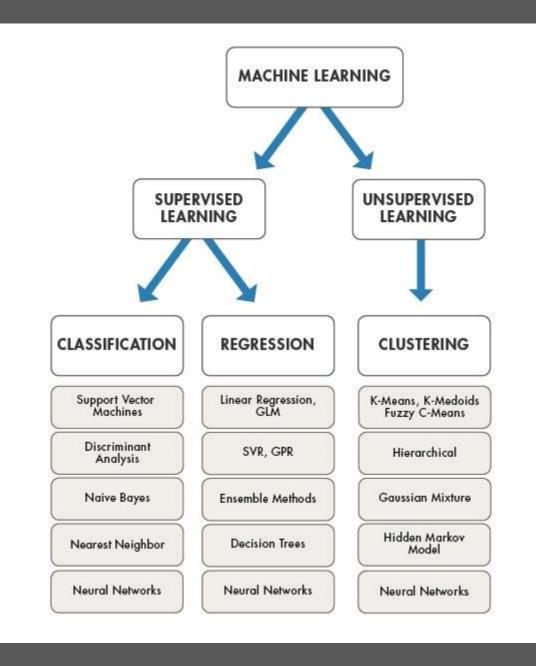
Text Analysis AMAZON FINE FOOD REVIEWS





Attribute Information

- ID
- Product id
- User id
- Profile Name
- Helpfulness Numerator
- Helpfulness Denominator
- Score 1 to 5
- Time
- summary
- Text

CONTENTS



1.ABSTRACT



2.EXISTING SYSTEM



3.PROPOSED SYSTEM



4. SCOPE OF DEVELOPMENT



5.REQUIREMENTS



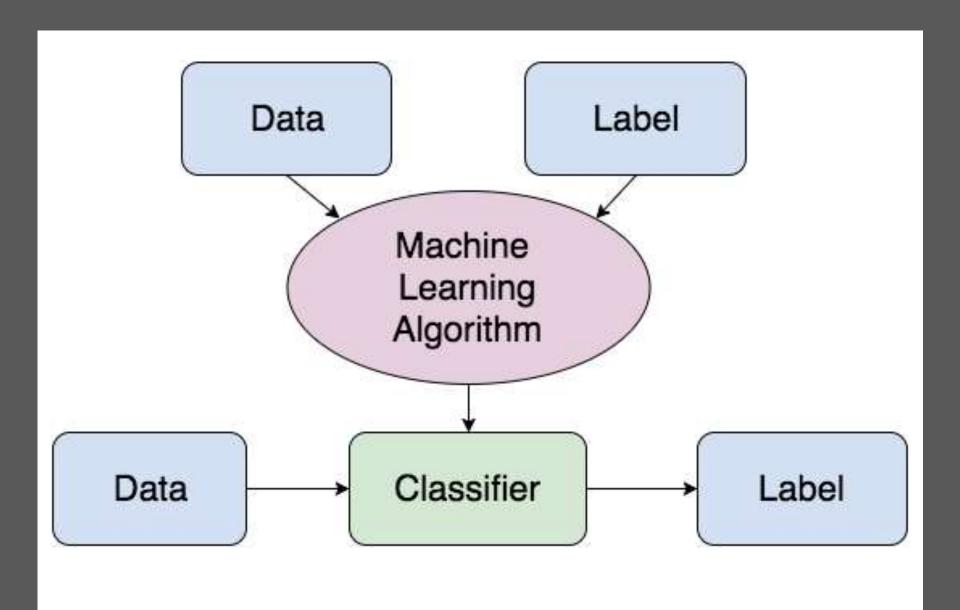
This is an Machine Learning project which can be used in Ecommerce websites. The development of the Internet changed the way people eat and responding for food. Amazon is a biggest website where users can easily purchase all kind of food they need with only a mouse-click. Here our aim is to create a classifier that classifies the reviews in to either positive or negative based on the data given such as the time at which the review was written, rating given by the customer. We can analyze the using various models like Naive Bayes, KNN, Logistic Regression etc.

Eventually we want to build an acceptable model which helps us better understand how customers rate and review the food they purchased.

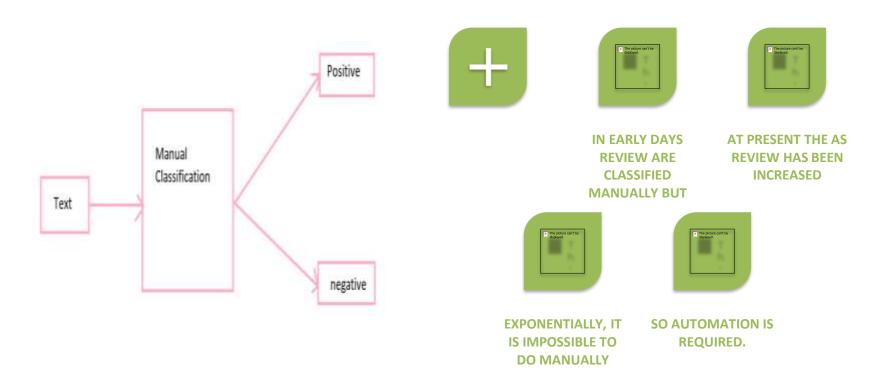
Here, we classified the reviews into 2 categories

- I) Positive: which indicates the customer is satisfied with the product
- 2) Negative: which indicates the customer is not satisfied with the product

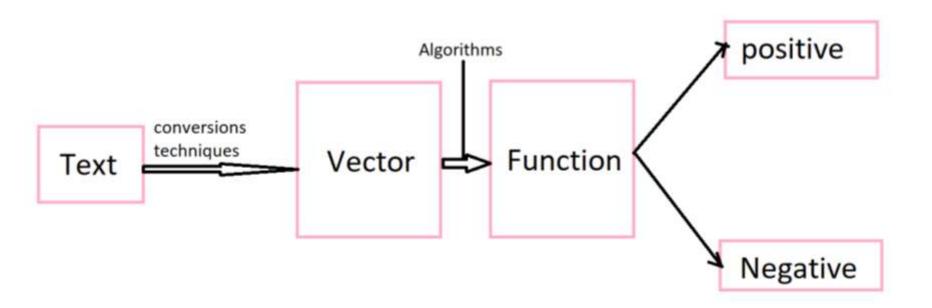
The main propaganda of this project is, it can be used in case of e-commerce websites where the review plays a vital role based on which the interest of the new customers depends.



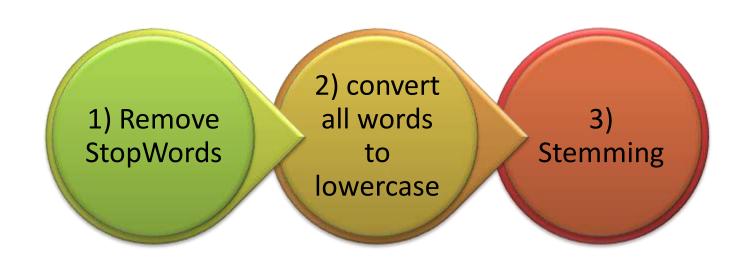
EXISTING SYSTEM



PROPOSED SYSTEM:



Text Pre_Processing



Text to Vector techniques

1. BOW(Bag Of Words)

-IDF(Term

2.TF-IDF(Term frequency Inverse Document Frequency

3

3.W2V (word to vector)

Bag Of Words

- T1: the dog is on the table
- T2: now the cats are on the table

Limitation example

- T3: This is a jntu college
- T4: This is not a jntu college

the dog is on the table



TF-IDF

$$w_{i,j} = tf_{i,j} \times \log\left(\frac{N}{df_i}\right)$$

 tf_{ij} = number of occurrences of i in j df_i = number of documents containing iN = total number of documents

Formulas



1)TF(W,R)=No of times w occur in R/Total no of word in R



0<=TF(W,R)<=1



2)IDF(W,D)= Log(N/ni)

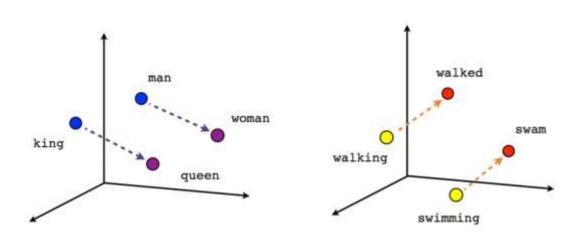


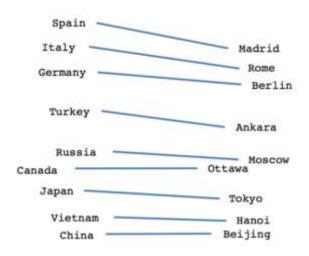
N---> Total no of Documents



ni--->No of documents contain w

W2V(Word to Vector)



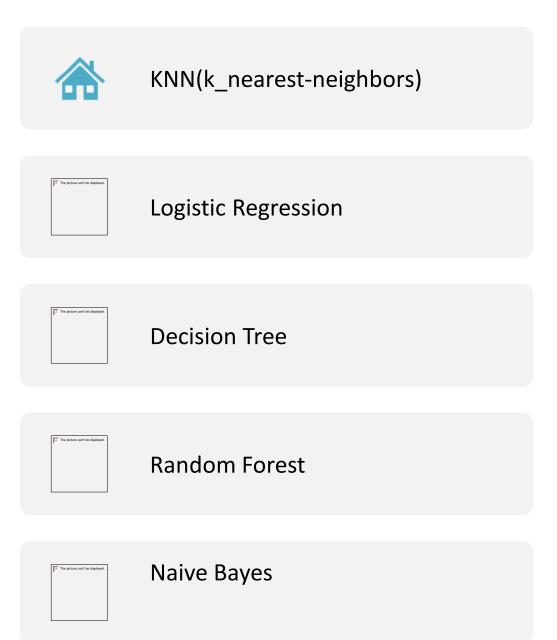


Male-Female

Verb tense

Country-Capital

Algorithms



Parameter Tuning

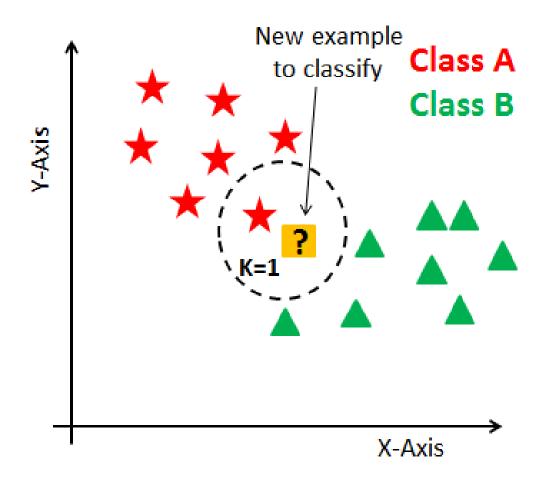
- GridSearchCV
- RandomizedSearch

fit KNN model to data after w2v

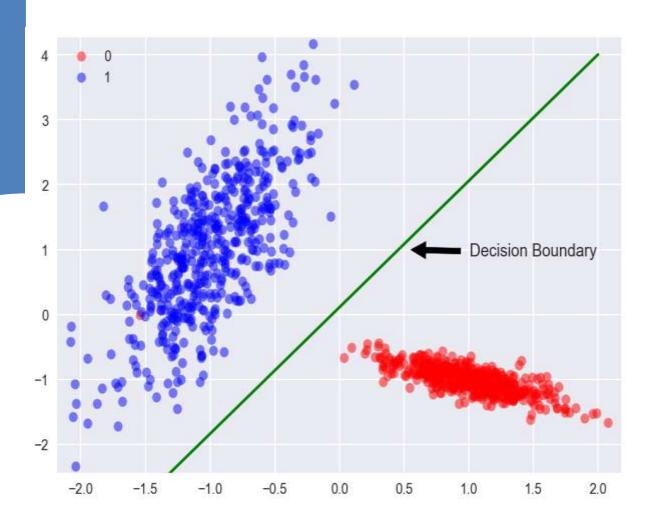
```
In [69]: from sklearn.model_selection import GridSearchCV
    from sklearn.neighbors import KNeighborsClassifier
    parameters = {'n_neighbors':[1,2,3,4,5,6,7,8]}
    neigh = KNeighborsClassifier()
    clf = GridSearchCV(neigh, parameters, cv=5, scoring="accuracy")
    clf.fit(x2_train,y2_train)
    clf.best_params_
```

Out[69]: {'n_neighbors': 7}

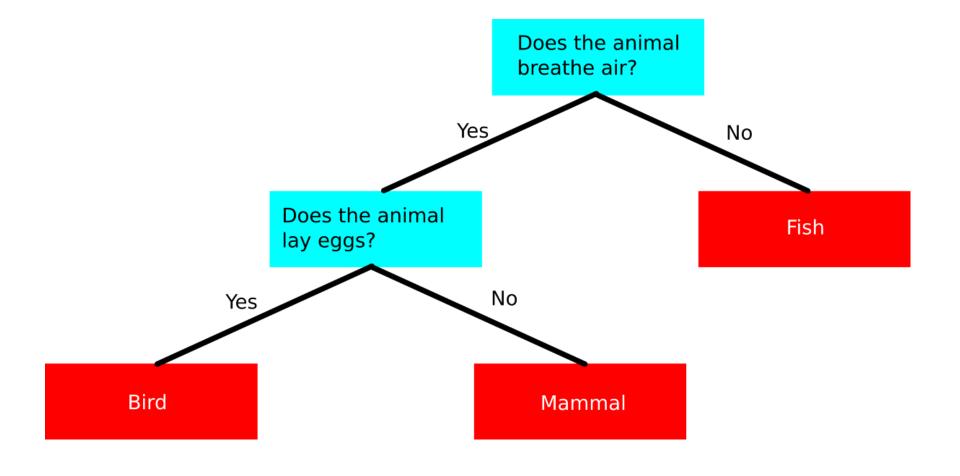
KNN



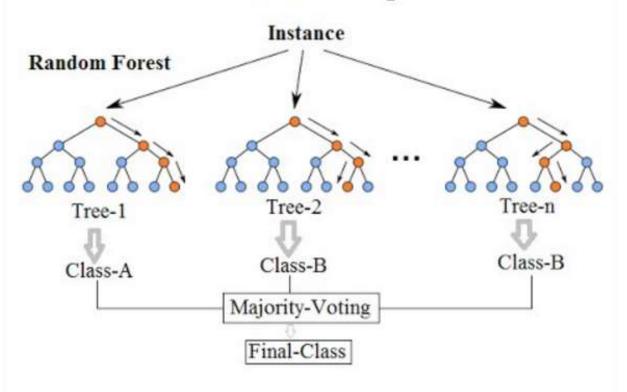
LogisticRegression



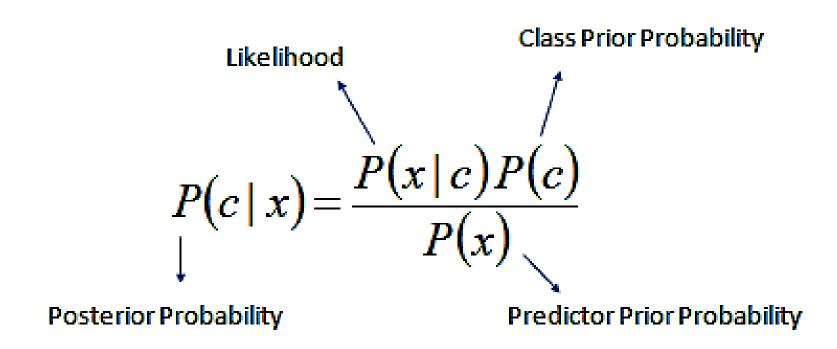
Decision Tree



Random Forest Simplified



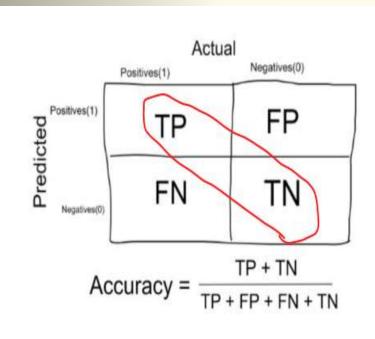
Naïve Bayes



$$P(c \mid X) = P(x_1 \mid c) \times P(x_2 \mid c) \times \cdots \times P(x_n \mid c) \times P(c)$$

Performance Metrics

- ACCURACY
- F1_SCORE



$$F_1 = 2 * \frac{precision * recall}{precision + recall}$$

Results Model Evaluation and Validation

	Algorithms	BOW	TFIDF	W2V
-	naive Bayes	80(97)	88(100)	57(50)
The prace can't be desiryed.	KNN	82(90)	90(92)	69(74)
The pricare and the displayed.	Logistic regression	90(100)	88(100)	57(85)
The pidule of solvered displayed.	Decision Tree	83(85)	83(92)	71(73)
The police can be dispersed.	Random Forest	90(94)	88(95)	70(90)

Scope of Development





We may use some advanced machine learning techniques to improve the performance, like

Deep Learning

