

# SENTIMENT ANALYSIS OF ELECTRONIC PRODUCTS IN



Lazada

FINAL PROJECT  
GROUP 2



# Meet Our Teams

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# Introduction - Background

NUMBER OF ONLINE SHOPPERS IN INDONESIA  
(in millions)



As the E-commerce markets in Indonesia is getting bigger every year, **product feedbacks** play the bigger roles in increasing the **revenue** of the sellers. This can be caused by the **change behavior** of consumer to **compare** products that they want to buy first from the **reviews** and prioritize the **services** they get from the seller. Hence, the sentiment analysis is needed to monitor and find the particular **aspects** of product from the reviews that people are expressing in positive, neutral, or negative way.

# Introduction - Goals



## Model

Build a sentiment analysis model that can interpret customer feedbacks by defining the feedback in some of the aspects.



## Insights

Provide insights of what features of product and services that needed to tailored to meet their customer needs.

# Benefits

**1**

A reference for sentiment analysis in other e-commerce or social media

**2**

Applicable for the recommendation system in e-commerce

**3**

Improve customer services of the product

**4**

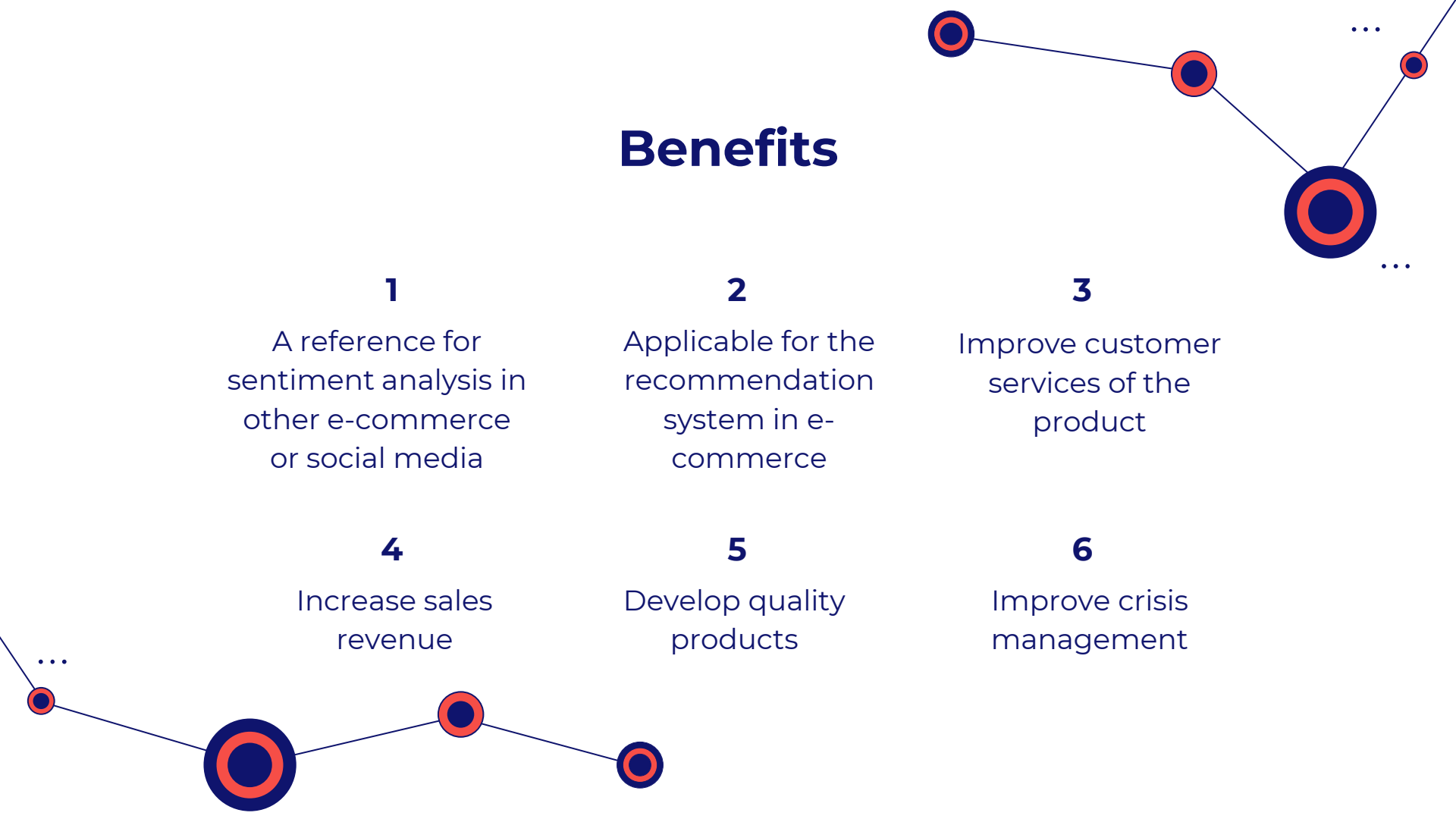
Increase sales revenue

**5**

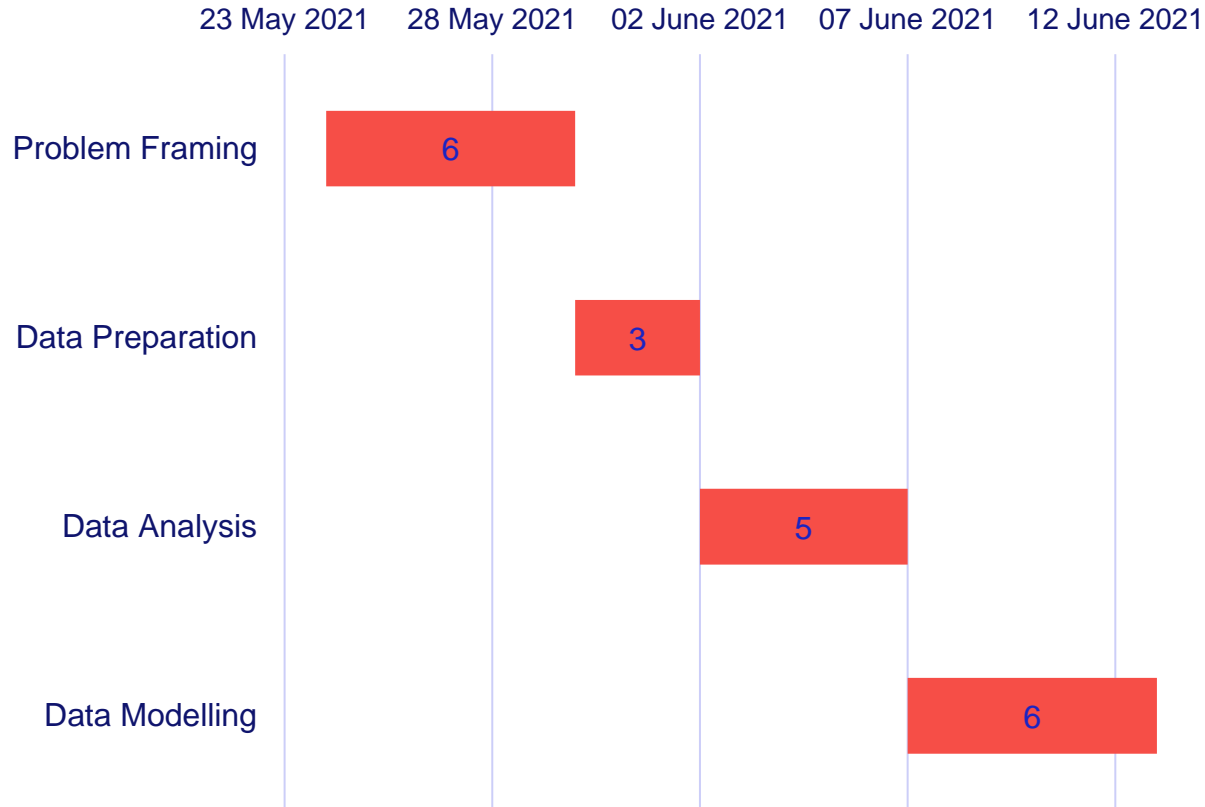
Develop quality products

**6**

Improve crisis management



# Timeline



# Tools

Jupyter Notebook



Pandas



Matplotlib



Scikit-Learn



NLTK

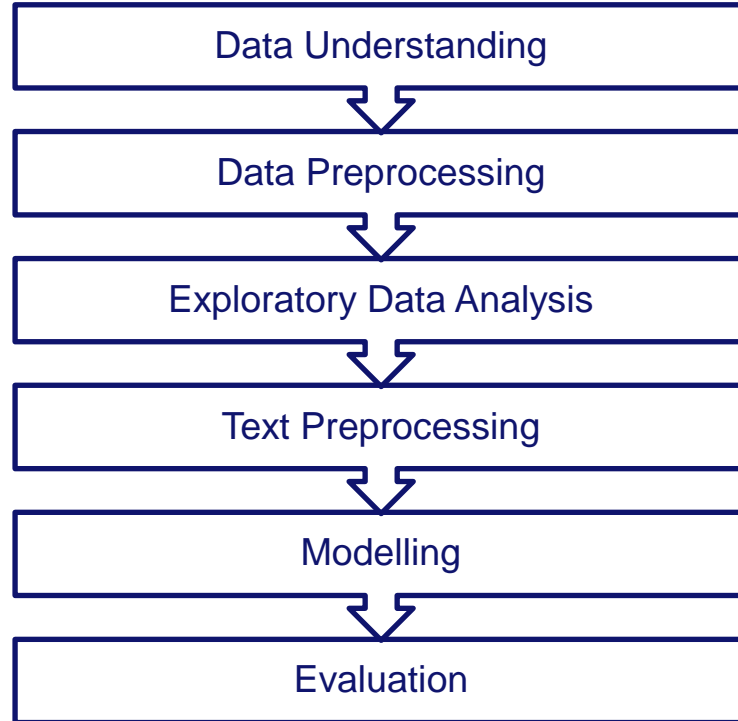


Sastrawi





# Workflow



# Workflow

## Data Understanding

- Explore and understanding variables of the dataset

## Data Preprocessing

- Preparing the dataset for analysis and modelling

## Exploratory Data Analysis

- Making the visualisation of the dataset to gain business understanding
- Investigate the shape of the dataset and number of samples

## Text Preprocessing

- Cleaning the dataset by handling the null values
- Handle the text dataset by the process of NLP

## Modelling

- Selecting the machine learning method suited with the dataset

## Evaluation

- Predict the test dataset
- Evaluating the model with performance metrics

# Data Understanding

## Lazada Indonesian Reviews

Product reviews from Lazada Indonesia based on 5 categories.

The dataset were divided into 2 items:

- 20191002-items.csv
- 20191002-reviews.csv

Source: <https://www.kaggle.com/grikomsn/lazada-indonesian-reviews?select=categories.txt>



beli-laptop



beli-smart-tv



shop-televisi-  
digital



jual-flash-  
drives



beli-harddisk-  
eksternal



# Dataset Lazada Indonesian Reviews

20191002-items.csv



Column Names	Definition	Data Type
itemid	Id of item	Int64
category	Category in e-commerce	Object
name	Title of the item	Object
brandname	Brand name of item	Object
url	Link to buy the item	Object
price	Price of item	Int64
averagerating	Average rating of item	Int64
totalreviews	Total Reviews of item	Int64
retrieveddate	Date of last calculation	Object

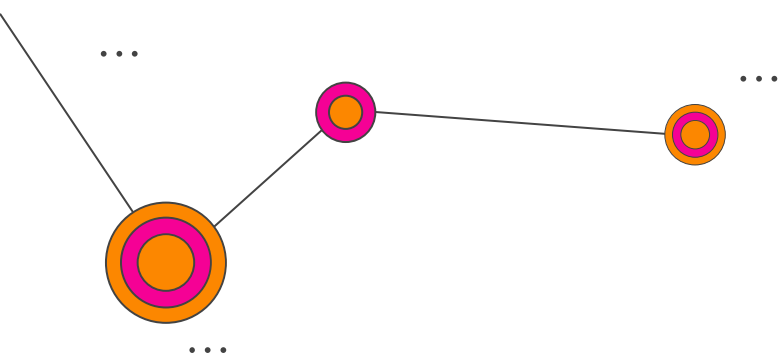


# Dataset Lazada Indonesian Reviews

**20191002-reviews.csv**



Column Names	Definition	Data Type
itemID	Id of item	int64
category	Category in e-commerce	object
name	Title of the item	object
rating	Rating by user	int64
originalRating	-	float64
reviewTitle	Title of review	object
reviewContent	Content of review	object
likeCount	Total Likes	int64
upVotes	Total upVotes	int64
downVotes	Total downvotes	int64
helpful	1 : Condition is helpful 0: Condition isnt helpful	bool
relevanceScore	Relevance score	float64
boughtDate	Date of transaction	object
clientType	Type of application	object
retrievedDate	Date of calculation	object



# Data Preprocessing

- Merge Dataset

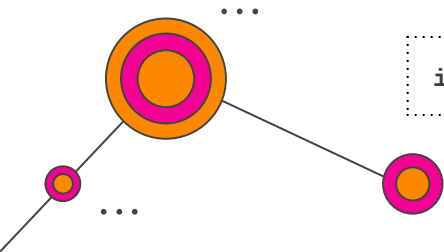
- Join two dataset with keys itemId and category resulting in 203787 rows x 22 columns. Eventhough its not related to the NLP, but it can be used for EDA

```
df =pd.merge(data, data1, how="inner",on=["itemId",'category'])
```

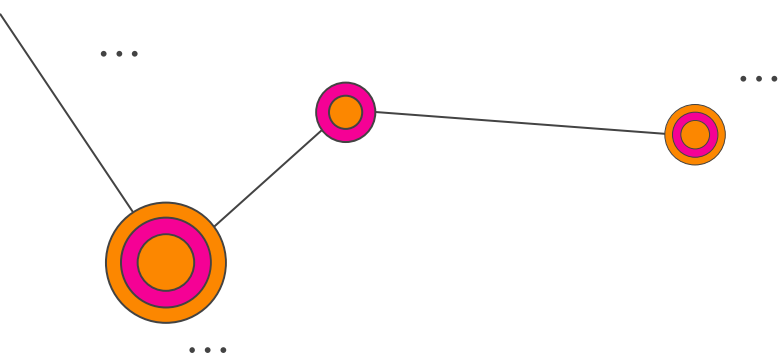
- Handling Missing Values

- Drop the missing values in reviewContent for sentiment analysis

```
df=df[df['reviewContent'].notna()]
```



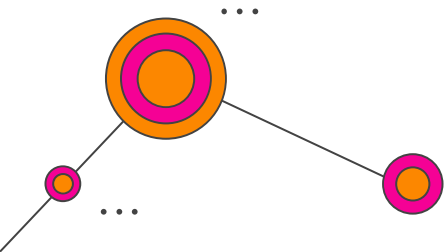
itemId	category	brandName	price	...	rating	...	reviewContent	...
--------	----------	-----------	-------	-----	--------	-----	---------------	-----



# Data Preprocessing

- Handling Outliers
  - There were some outliers in the price distribution, so we choose to investigate more about it and found the review was full of joke in range of price >Rp 40.000.000

```
df.reviewContent[df.price==df.price.max()]
```



# Reviews



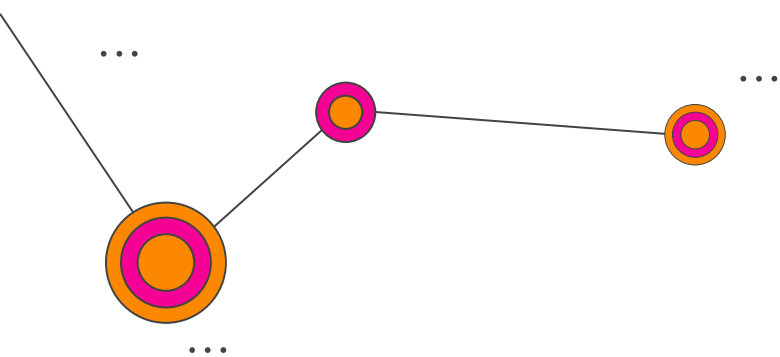
## SEBUT SAJA MAWAR

“Akhirnya kebeli juga berkat jual  
GINJAL keponakan”

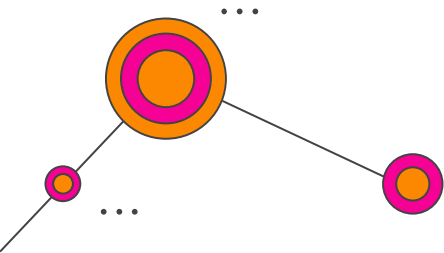
## SEBUT SAJA BUDI

“Barang nya udh dtg,pesen tadi ehh  
nyampe nya kemaren,sekarang dipake  
cuma buat alas tenis meja,thxx...”





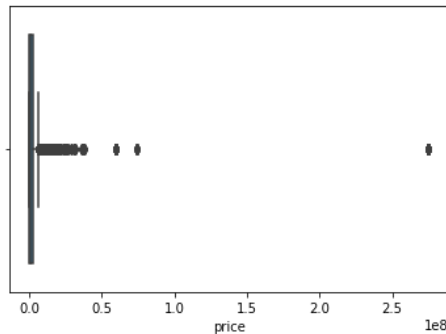
# Data Preprocessing



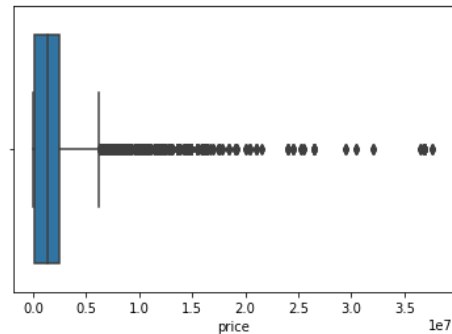
- Handling Outliers

- So we choose to remove the product that has the range of price > Rp 40.000.000

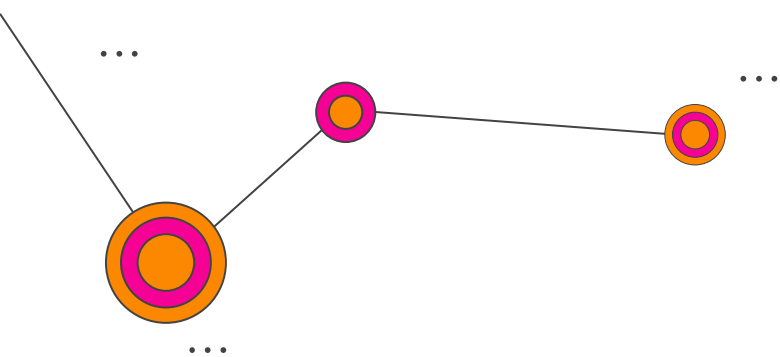
```
df = df[df.price < 40000000]  
sns.boxplot(df.price)
```



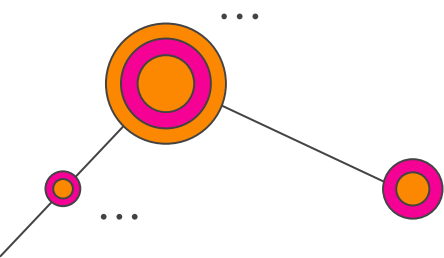
From this



To this



# Data preprocessing



- Set Samples  
We reduce the amount of dataset that will be used for modelling to shorten the computer processing. The dataset became an imbalance so we set the sample to be equal in amount that is 5000

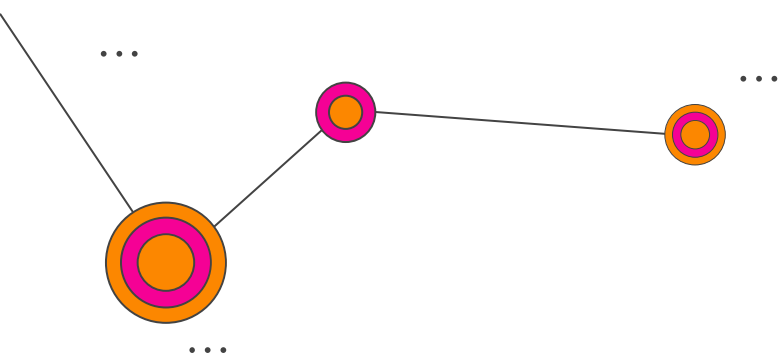
```
s1=df[df.label==1].sample(5000, replace=True)
s2=df[df.label==0].sample(5000, replace=True)

data=pd.concat([s1,s2])
```

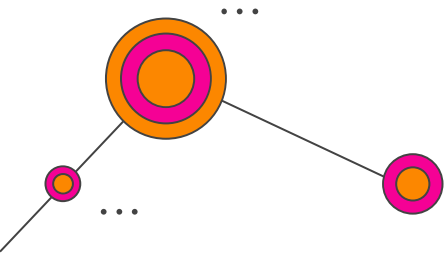
	col	row
Label 1	1	183583
Label 0	1	20204



	col	row
Label 1	1	5000
Label 0	1	5000



# Data preprocessing

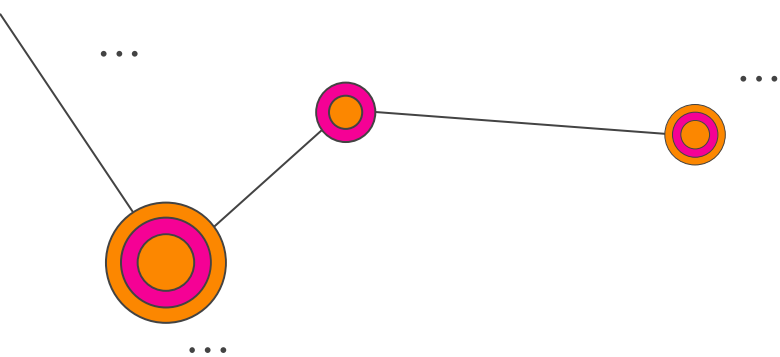


- Set label

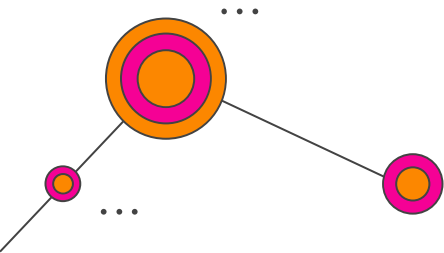
We set the satisfaction rate to be positive and negative. The positive label will be filled with 5 and 4 in rating and negative will be the rest.

```
label=[]
for index, row in df.iterrows():
    if row['rating']==5:
        label.append(1)
    elif row['rating']==4:
        label.append(1)
    else:
        label.append(0)

df['label']=label
df.head()
```



# Data preprocessing



## IMPORTANCE OF RATINGS & REVIEWS

Based on Lazada's data, products with 10 or more 4-5-star ratings have:

**23x** more traffic & **18x** more sales

Your product reviews have a big influence on your customer's purchasing decision. The more reviews your products have, the more traffic you gain.

**Remember:** For every orders you received, ask your customers for their ratings and reviews to get more traffic.

### Traffic of product with and without Reviews

No. Ratings & Reviews	Impact on Traffic
-----------------------	-------------------

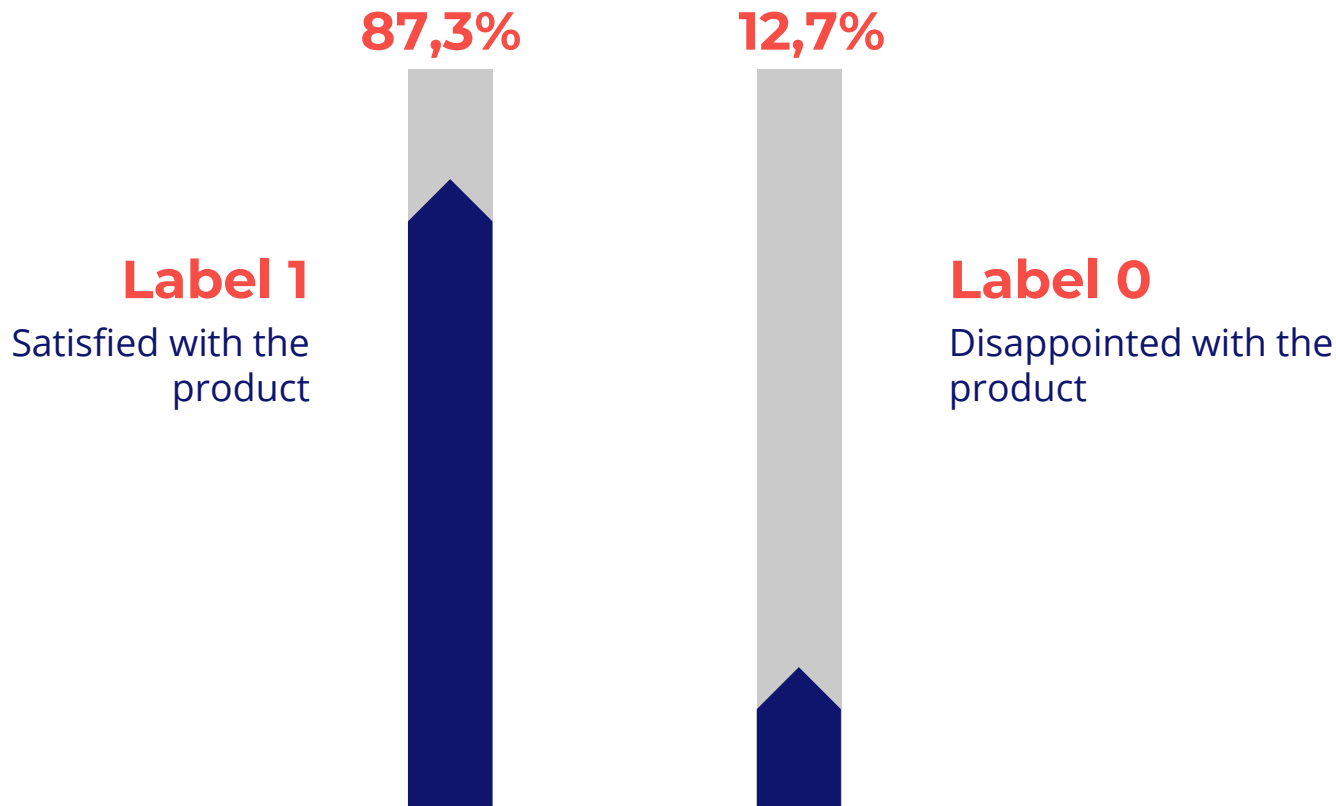
1-9	= 5x
-----	------

10-19	= 14x
-------	-------

>=20	= 29x
------	-------

- Positive Seller Rating
  - Get higher rank on Lazada's SEO
  - Chance to participated on event and campaign held by Lazada
  - Chance to join priority program
- Increase GMV of SKU
- 10 rejections = 1 bad review

# Exploratory Data Analysis

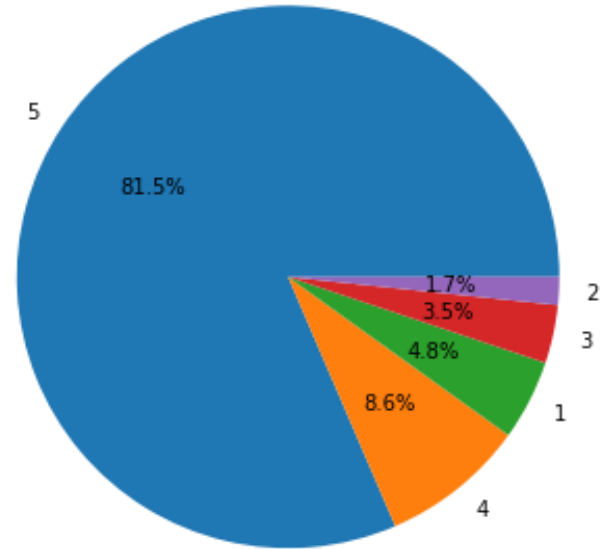


# Exploratory Data Analysis

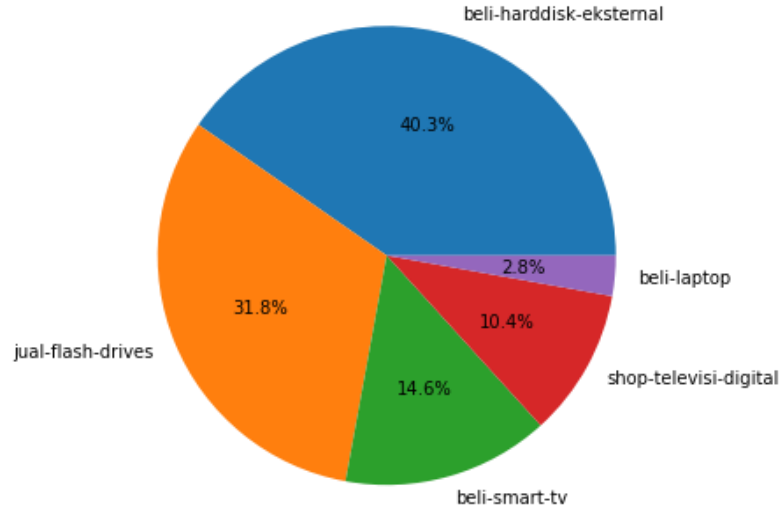
Based on Lazada's data, products with 10 or more 4-5-star ratings have: **23x** more traffic & **18x** more sales

- Positive Seller Rating
  - Get higher rank on Lazada's SEO
  - Chance to participated on event and campaign held by Lazada
  - Chance to join priority program
- Increase GMV(Gross Merchant Values) of SKU(Stock Keeping Unit)
- 10 rejections = 1 bad review

Rating pie chart



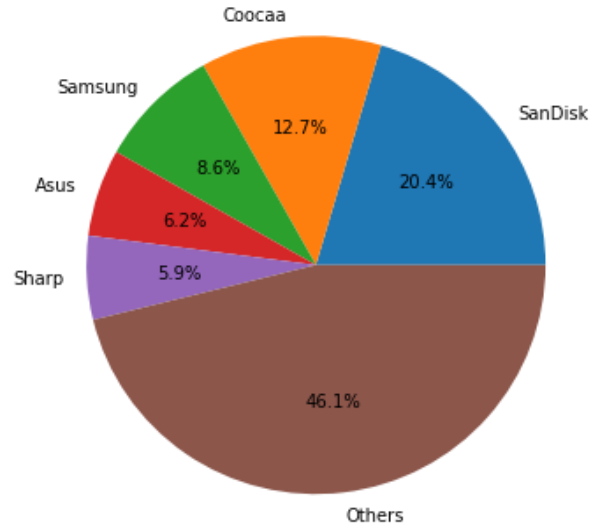
# Exploratory Data Analysis



## Category

Most of the categories exists in the dataset were 'beli-harddisk-eksternal'. But it doesn't reflect the item in the dataset because there were many seller use the unrelated categories too.

# Exploratory Data Analysis



## Brand

There were 235 brands in the dataset and the best seller brand is SanDisk



# Exploratory Data Analysis



## COOCA LED TV 40 INCH

[GRATIS ONGKIR] –  
FULL HD PANEL -SLIM

3952 units sold

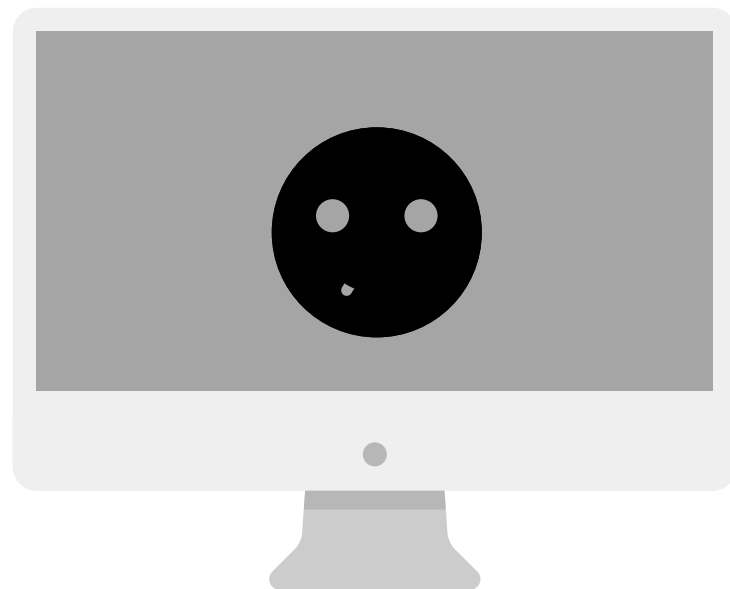
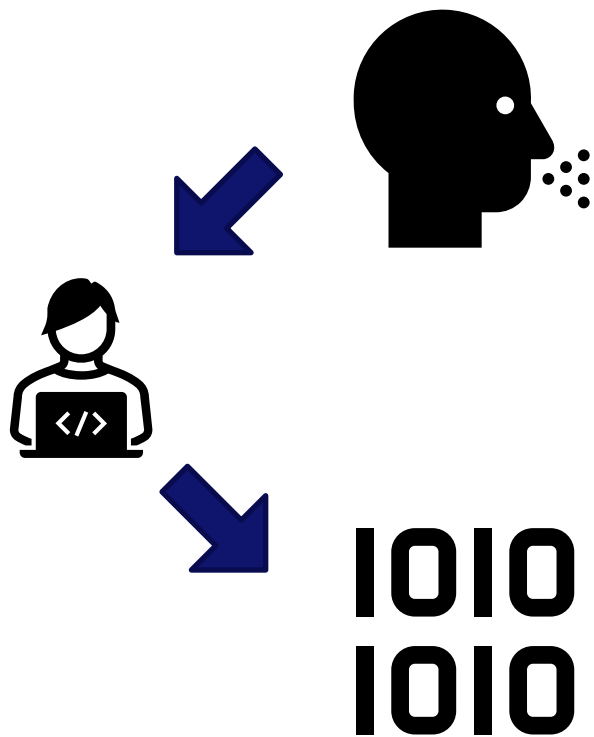


## COOCA LED TV 24 INCH

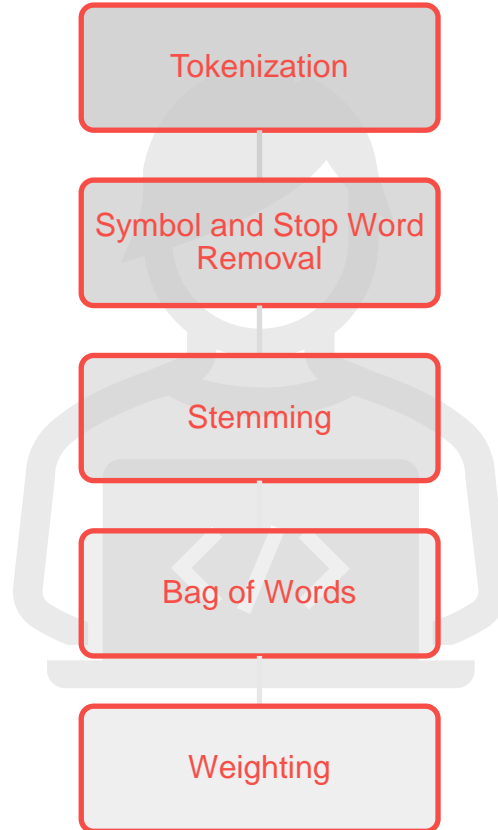
[GRATIS ONGKIR] –HD  
PANEL- SLIM

3540 units sold





# Text Preprocessing



# Text Preprocessing

- Tokenization
  - Splitting sentence into words.

```
import re
df.reviewContent=df.reviewContent.apply(lambda x: re.split(r"\s+", x))
```

Example:

['barang bagus banget',  
'barang rusak , packing buruk banget',  
'barang oke, tapi pengiriman buruk banget',  
'packing rapi , pengiriman cepat dan aman']



[['barang', 'bagus', 'banget'],  
['barang', 'rusak', '', 'packing', 'buruk', 'banget'],  
['barang', 'oke,', 'tapi', 'pengiriman', 'buruk', 'banget'],  
['packing', 'rapi', '', 'pengiriman', 'cepat', 'dan', 'aman']]

# Text Preprocessing

- Remove unimportant text
  - We removed some unwanted text that unrelated to the sentiment analysis and summarizing the text.
  - Remove punctuation and numbers

```
df.reviewContent = df.reviewContent.apply(lambda x: x.lower())  
df.reviewContent = df.reviewContent.apply(lambda x: re.sub(r'([^\a-z\s]+)', '', x))
```

- Remove Stopwords

```
from nltk.corpus import stopwords  
from string import punctuation  
stop_words=stopwords.words("indonesian")+list(punctuation)  
df.reviewContent=df.reviewContent.apply(lambda x: [w for w in x if not w in  
                                                    stop_words])  
df.reviewContent=df.reviewContent.apply(lambda x: str(' '.join(x)))
```

Example:

'Paket sudah sampai, sudah dicoba dan berfungsi dengan baik, semoga awet dan tidak ada kendala.'



Paket sampai, dicoba berfungsi baik, semoga awet kendala.'

# Text Preprocessing

- Stemming  
Reducing the affixes and suffixes to get the root of words.

```
from Sastrawi.Stemmer.StemmerFactory import StemmerFactory  
factory = StemmerFactory()  
stemmer = factory.create_stemmer()  
df.reviewContent=df.reviewContent.apply(lambda x:stemmer.stem(x))
```

Example:

'Paket sudah sampai, sudah dicoba dan berfungsi dengan baik, semoga awet dan tidak ada kendala.'

'Mereka Meniru-nirukannya.'



'paket sudah sampai sudah coba dan fungsi dengan baik  
moga awet dan tidak ada kendala'

'mereka tiru'

# Text Preprocessing

- Train Test Split  
Split the dataset to be train and test dataset

```
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(dfr.reviewContent,
                                                    dfr['label'], test_size=0.1, random_state=30)
```

- Bag of Words  
The bag-of-words model is a popular and simple feature extraction technique used when we work with text. It describes the occurrence of each word within a document.

```
from sklearn.feature_extraction.text import CountVectorizer

cv = CountVectorizer()
bog= cv.fit_transform(dfr.reviewContent)
```

	col	row
X_train	1	9000
X_test	1	1000



9132	9000
9132	1000

# Text Preprocessing

- TF-IDF

Using TF-IDF to transform sentence into vector that can be applicable for machine learning model. In information retrieval, tf-idf,  $TF \times IDF$ , or TFIDF, short for term frequency-inverse document frequency, is a numerical statistic that is intended to reflect how important a word is to a document in a collection or corpus.

```
from sklearn.feature_extraction.text import TfidfVectorizer
```

```
tf = TfidfVectorizer()  
X_train = tf.fit_transform(X_train)  
X_test = tf.transform(X_test)
```

	col	row
X_train	1	9000
X_test	1	1000



8631	9000
8631	1000



# Text Preprocessing

Examples of Bag of Words and TF-IDF

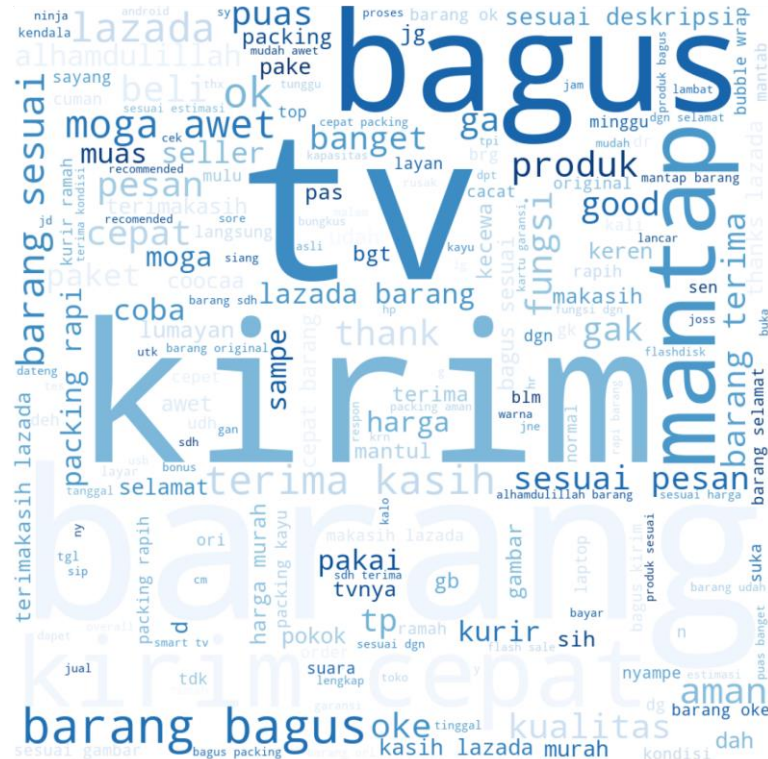
```
df =[  
    'barang bagus banget',  
    'barang rusak , packing buruk banget']
```

Bag of Words

```
['bagus', 'banget', 'barang', 'buruk', 'packing', 'rusak']  
[[1 1 1 0 0 0]  
 [0 1 1 1 1 1]]
```

TF-IDF

```
'bagus', 'banget', 'barang', 'buruk', 'packing', 'rusak']  
[[0.70490949  0.50154891  0.50154891  0.          0.          0.]  
 [0.          0.35520009  0.35520009  0.49922133  0.49922133  0.49922133]]
```

[illegible]

## Word Visualisation

- For label 1

# Text Preprocessing

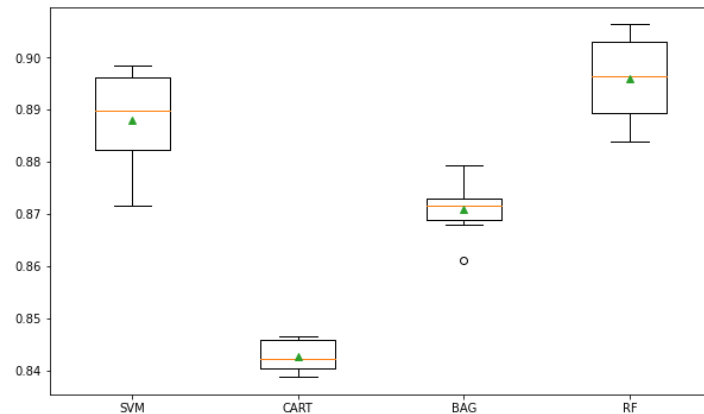


# Word Visualisation

- For label 0

# Modelling - Comparison

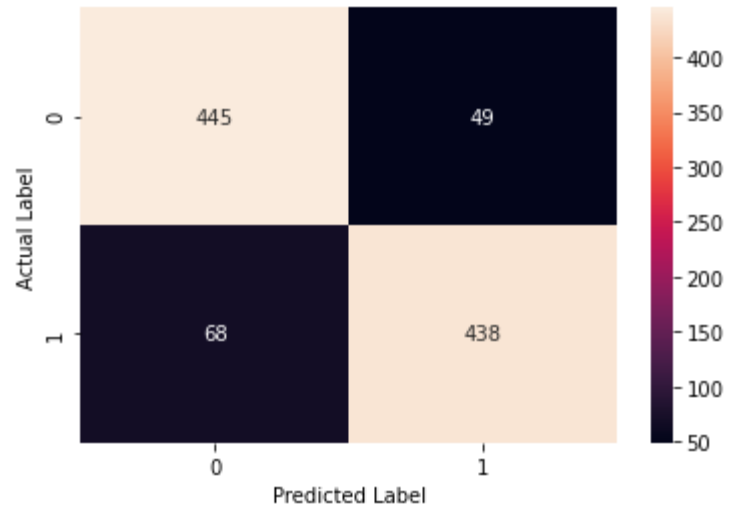
```
cv = RepeatedStratifiedKFold(n_splits=2, n_repeats=3, random_state=1)
```



- Decision Tree  
AUC : 0.843
- SVM  
AUC : 0.888
- Bagging  
AUC : 0.871
- Random Forest  
AUC : 0.896

# Modelling – Confusion Matrix

```
sns.heatmap(confusion_matrix(y_test, pred),annot=True,fmt='g')
```



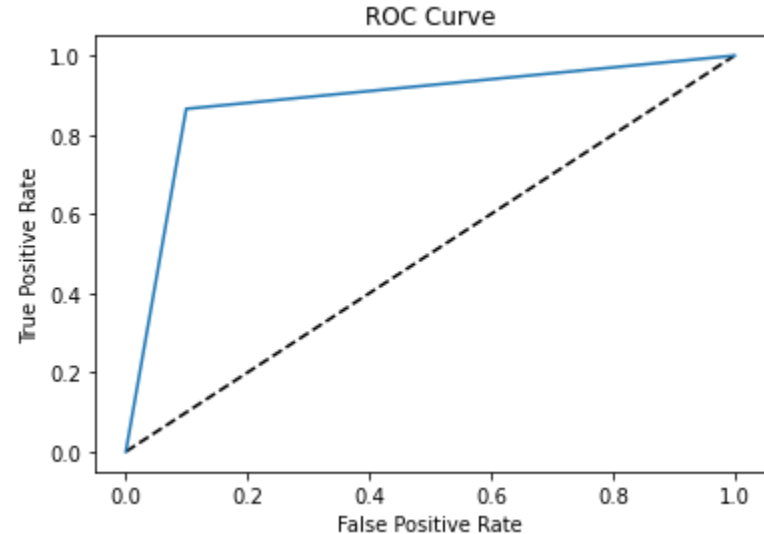
# Modelling – Classification Report

	precision	recall	f1-score	support
0	0.87	0.90	0.88	494
1	0.90	0.87	0.88	506
accuracy			0.88	1000
macro avg	0.88	0.88	0.88	1000
weighted avg	0.88	0.88	0.88	1000

# Modelling – RFC Performance

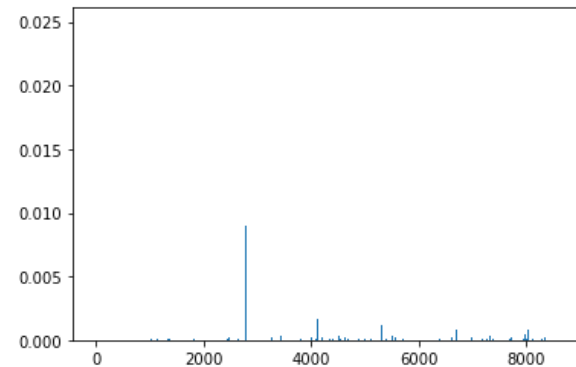
Random forest Classifier  
Algorithm with high interpretability and  
accuracy

Accuracy : 0.883  
AUC : 0.896



# Modelling – Feature Importances

RFC Feature Importances



Words	feature_ importances	Sum of Words
bagus	0.024868	1697
mantap	0.023659	494
cepat	0.018677	1130
kecewa	0.017781	823
barang	0.017632	4238
sesuai	0.014964	1345
gak	0.012157	758
kirim	0.012155	2212
lazada	0.011953	1645
awet	0.011398	484



# Conclusion

Sentiment Analysis model has been created with RFC model with the biggest AUC (0,895)

There are 3 ways that you can implement to increase your ratings and reviews:

## REMIND

Each customer only has 30 days to leave a review. **Remind** them via "Chat" or send an "invitation card" in your parcel.



## REWARD

**Reward** customer with vouchers for leaving a review.



## RESPOND

Listen and understand your customers. **Respond** to both good and bad reviews.



# Literature

- J. Wong, Natural Language Processing Workflow , 2020, (<https://towardsdatascience.com/natural-language-processing-workflow-1dddf3a48ab5>).
- B. Shetty, Natural Language Processing (NLP) for Machine Learning, 2018, (<https://towardsdatascience.com/natural-language-processing-nlp-for-machine-learning-d44498845d5b>).
- A. F. Zulfikar, Pengembangan Algoritma Stemming Bahasa Indonesia dengan Pendekatan Dictionary Base Stemming untuk Menentukan Kata Dasar dari Kata Yang Berimbuhan, Universitas Pamulang, 2017.
- <https://sellercenter.lazada.com.ph/seller/helpcenter/ratings-reviews-11737.html?spm=a2a15.helpcenter-psc-article.articles-list.4.41a71da3t3hxjY>
- <https://sellercenter.lazada.co.id/seller/helpcenter/Apa-Yang-Dimaksud-Dengan-Positif-Seller-Rating-6070.html?spm=a2a14.helpcenter-psc-search.article.1.37cb475dS3IBs5>
- <https://sellercenter.lazada.co.id/seller/helpcenter/Apa-ltu-Bisnis-Analysis-6041.html?spm=a2a14.helpcenter-psc-article.articles-list.7.5b0f4af5qYllv4>

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