

SMS Spam or Ham project

NLP – Clustering



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# INTRODUCTION

We are a communications company that cares about fulfilling our customers' desires and listening to their requirements.

We received many complaints about the large number of unclear messages they had, which caused a lack of distinction between ham and spam messages.





## Solution

To solve this problem due to the desire of our customers, we used a set of data and artificial intelligence algorithms.



To o ls:

1

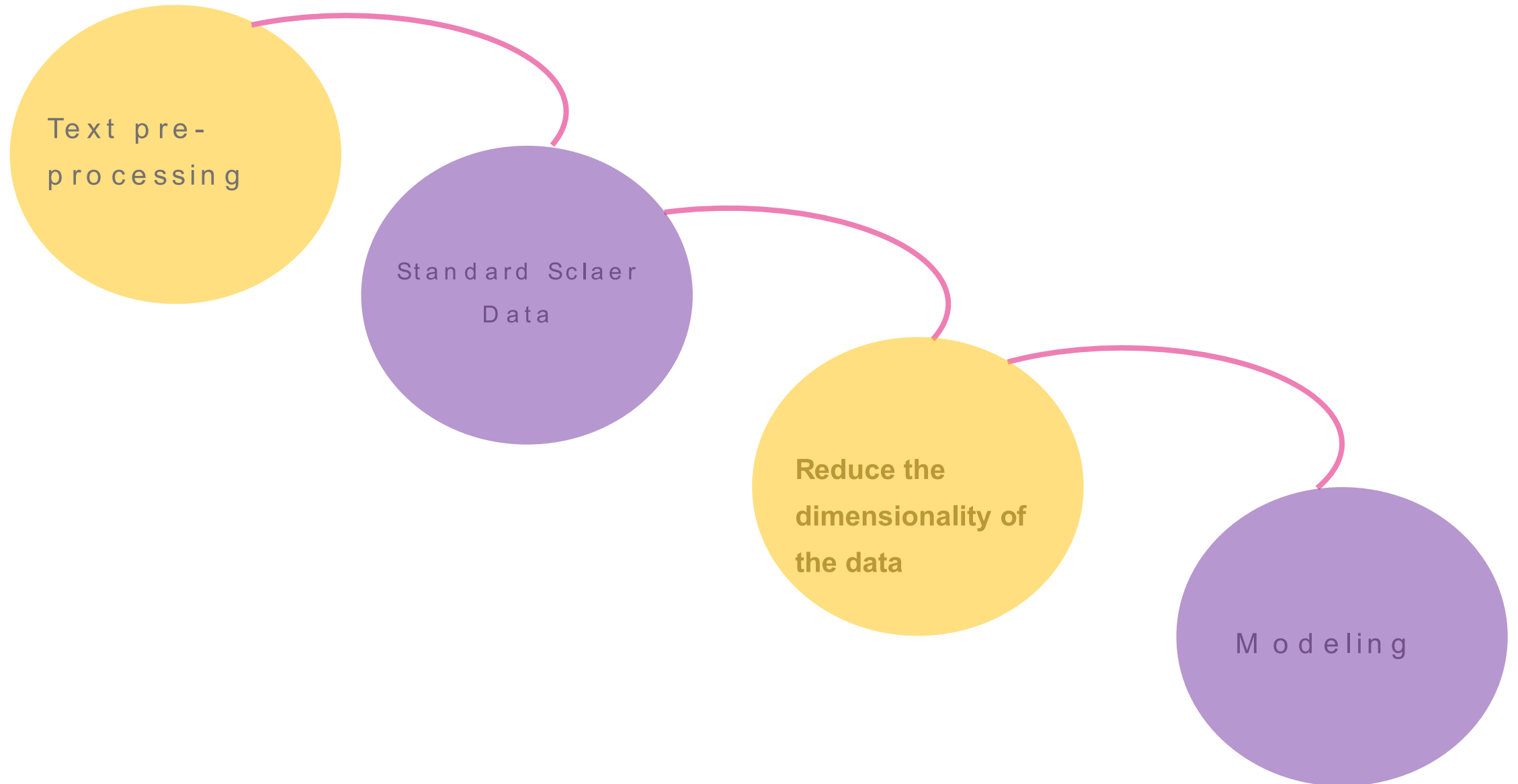
**Python  
Libraries**

2

**Jupyter  
Notebook**



Steps:



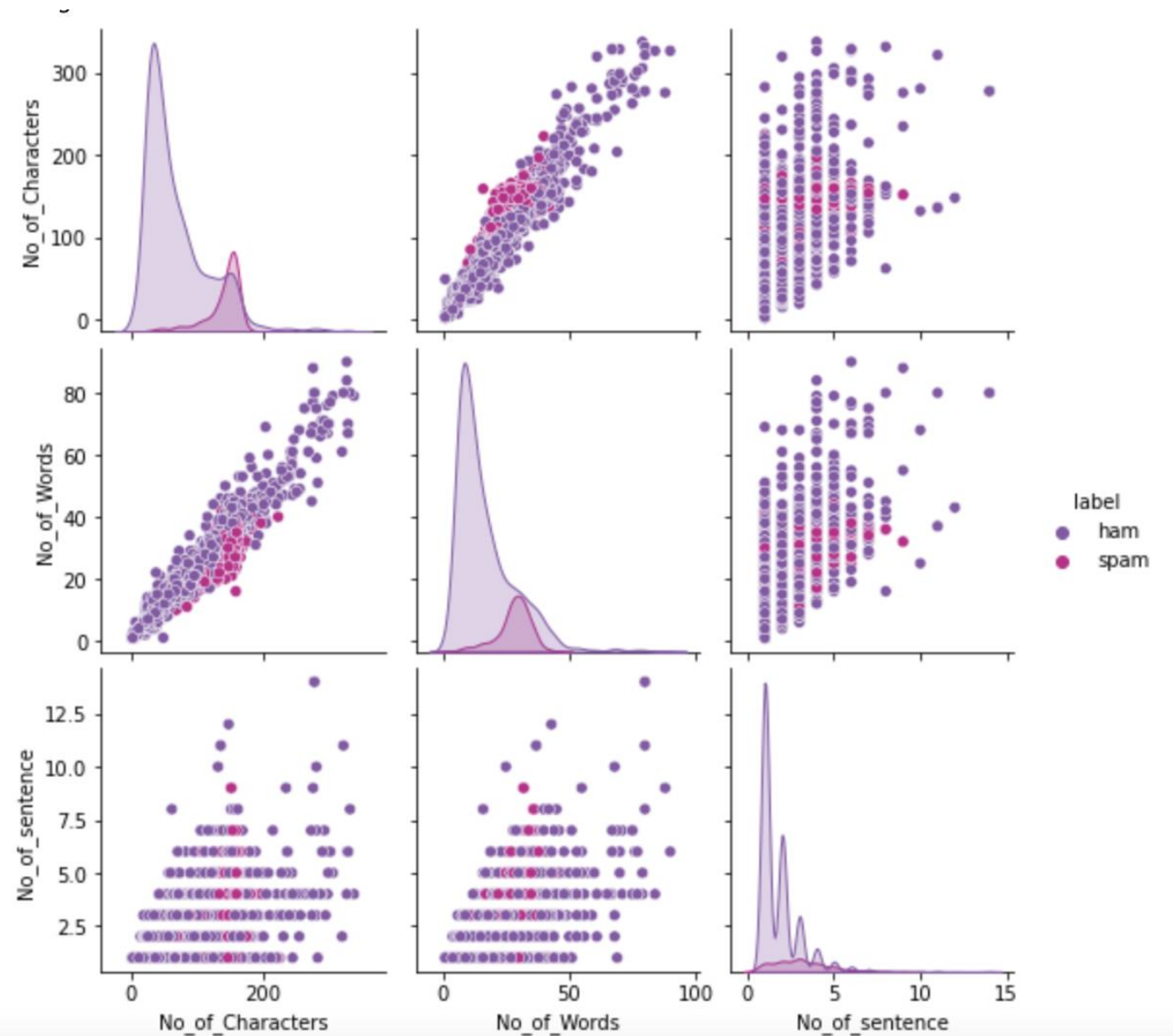
# Feature Engineering :

Number of characters  
in the text message

Number of words in  
the text message

Number of sentences  
in the text message

# Feature Engineering :



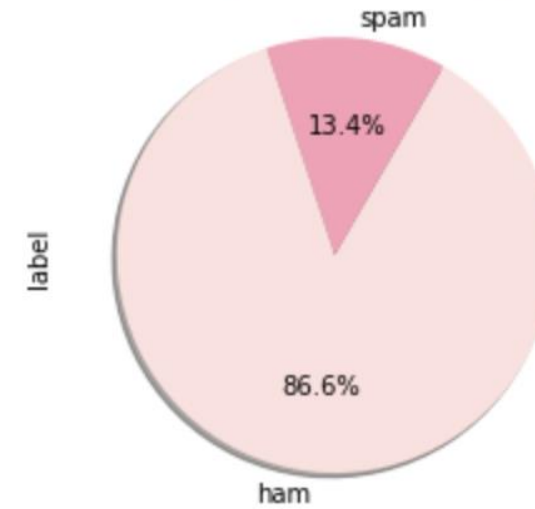


# Data Exploration

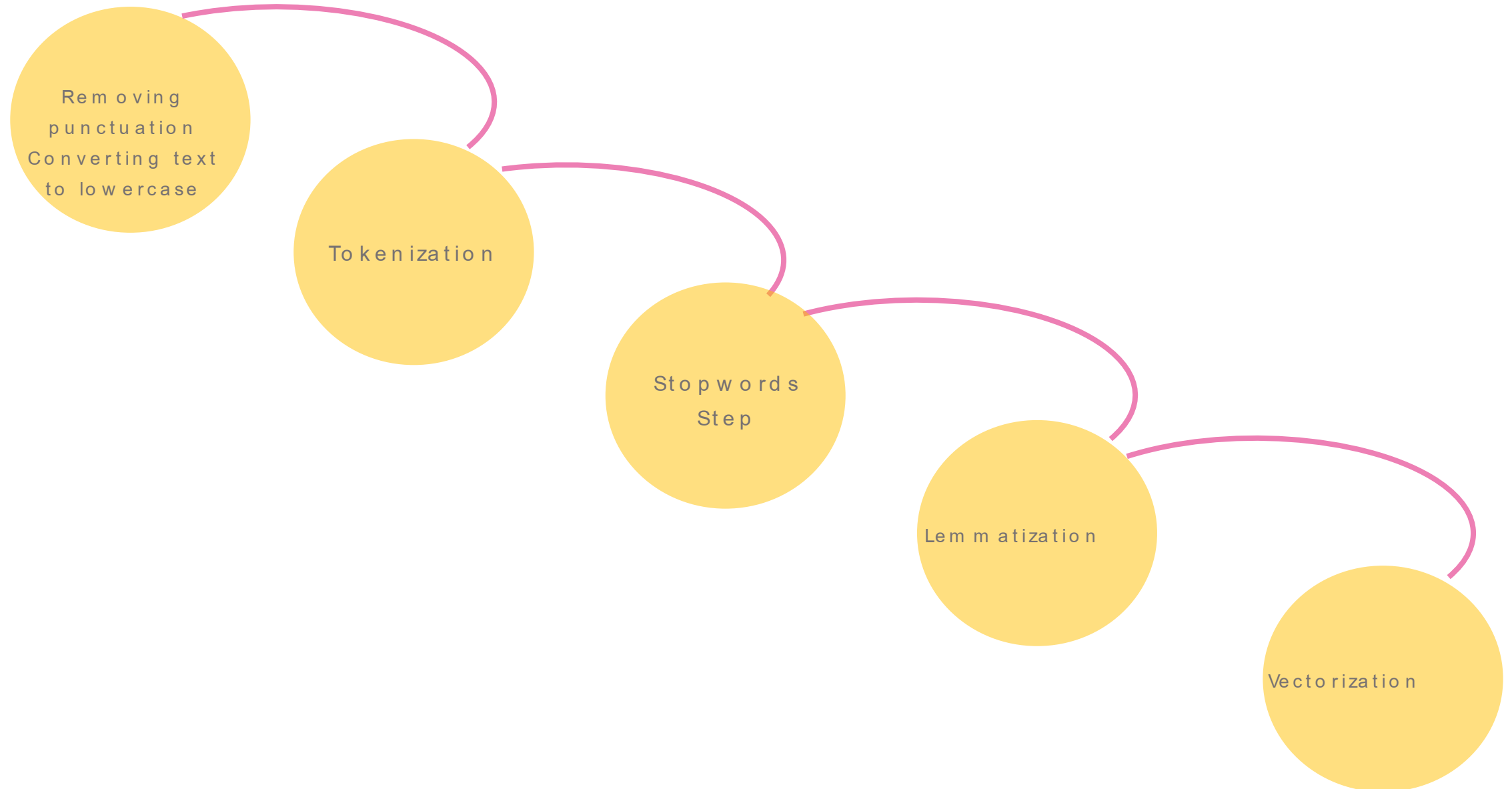
Count Plot of Classes



plot of Count percentage of Classes



# NLP Steps



Removing punctuation Converting text to lowercase

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The First 7 Texts after cleaning:

go until jurong point crazy available only in bugis n great world la e buffet cine there got amore wat  
ok lar joking wif u oni

free entry in a wkly comp to win fa cup final tkts st may text fa to to receive entry question std txt rate t c s ap  
ply over s

u dun say so early hor u c already then say

nah i don t think he goes to usf he lives around here though

freemsg hey there darling it s been week s now and no word back i d like some fun you up for it still tb ok xxx std  
chgs to send to rcv

even my brother is not like to speak with me they treat me like aids patent

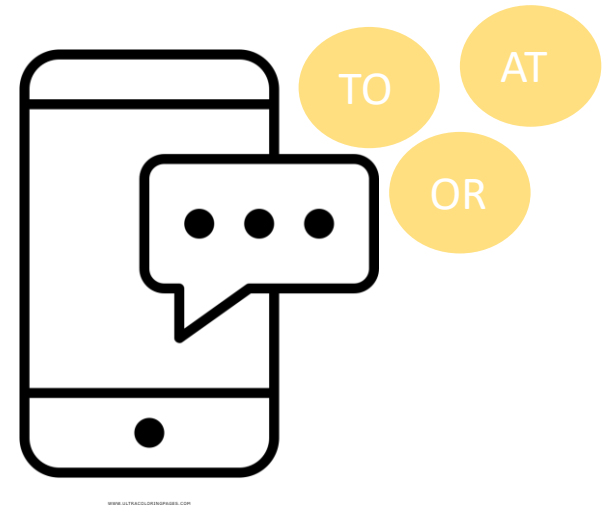


## To k e n i z a t i o n

The First 7 Texts after Tokenizing:

```
['go', 'until', 'jurong', 'point', 'crazy', 'available', 'only', 'in', 'bugis', 'n', 'great', 'world', 'la', 'e', 'b  
uffet', 'cine', 'there', 'got', 'amore', 'wat']  
['ok', 'lar', 'joking', 'wif', 'u', 'oni']  
['free', 'entry', 'in', 'a', 'wkly', 'comp', 'to', 'win', 'fa', 'cup', 'final', 'tkts', 'st', 'may', 'text', 'fa',  
'to', 'to', 'receive', 'entry', 'question', 'std', 'txt', 'rate', 't', 'c', 's', 'apply', 'over', 's']  
['u', 'dun', 'say', 'so', 'early', 'hor', 'u', 'c', 'already', 'then', 'say']  
['nah', 'i', 'don', 't', 'think', 'he', 'goes', 'to', 'usf', 'he', 'lives', 'around', 'here', 'though']  
['freemsg', 'hey', 'there', 'darling', 'it', 's', 'been', 'week', 's', 'now', 'and', 'no', 'word', 'back', 'i', 'd',  
'like', 'some', 'fun', 'you', 'up', 'for', 'it', 'still', 'tb', 'ok', 'xxx', 'std', 'chgs', 'to', 'send', 'to', 'rc  
v']  
['even', 'my', 'brother', 'is', 'not', 'like', 'to', 'speak', 'with', 'me', 'they', 'treat', 'me', 'like', 'aids',  
'patent']
```

# Stop words Step



The First 7 Texts after removing the stopwords:

['go', 'jurong', 'point', 'crazy', 'available', 'bugis', 'n', 'great', 'world', 'la', 'e', 'buffet', 'cine', 'got', 'amore', 'wat']

['ok', 'lar', 'joking', 'wif', 'u', 'oni']

['free', 'entry', 'wkly', 'comp', 'win', 'fa', 'cup', 'final', 'tkts', 'st', 'may', 'text', 'fa', 'receive', 'entry', 'question', 'std', 'txt', 'rate', 'c', 'apply']

['u', 'dun', 'say', 'early', 'hor', 'u', 'c', 'already', 'say']

['nah', 'think', 'goes', 'usf', 'lives', 'around', 'though']

['freemsg', 'hey', 'darling', 'week', 'word', 'back', 'like', 'fun', 'still', 'tb', 'ok', 'xxx', 'std', 'chgs', 'send', 'rcv']


['even', 'brother', 'like', 'speak', 'treat', 'like', 'aids', 'patent']

# Lemmatization

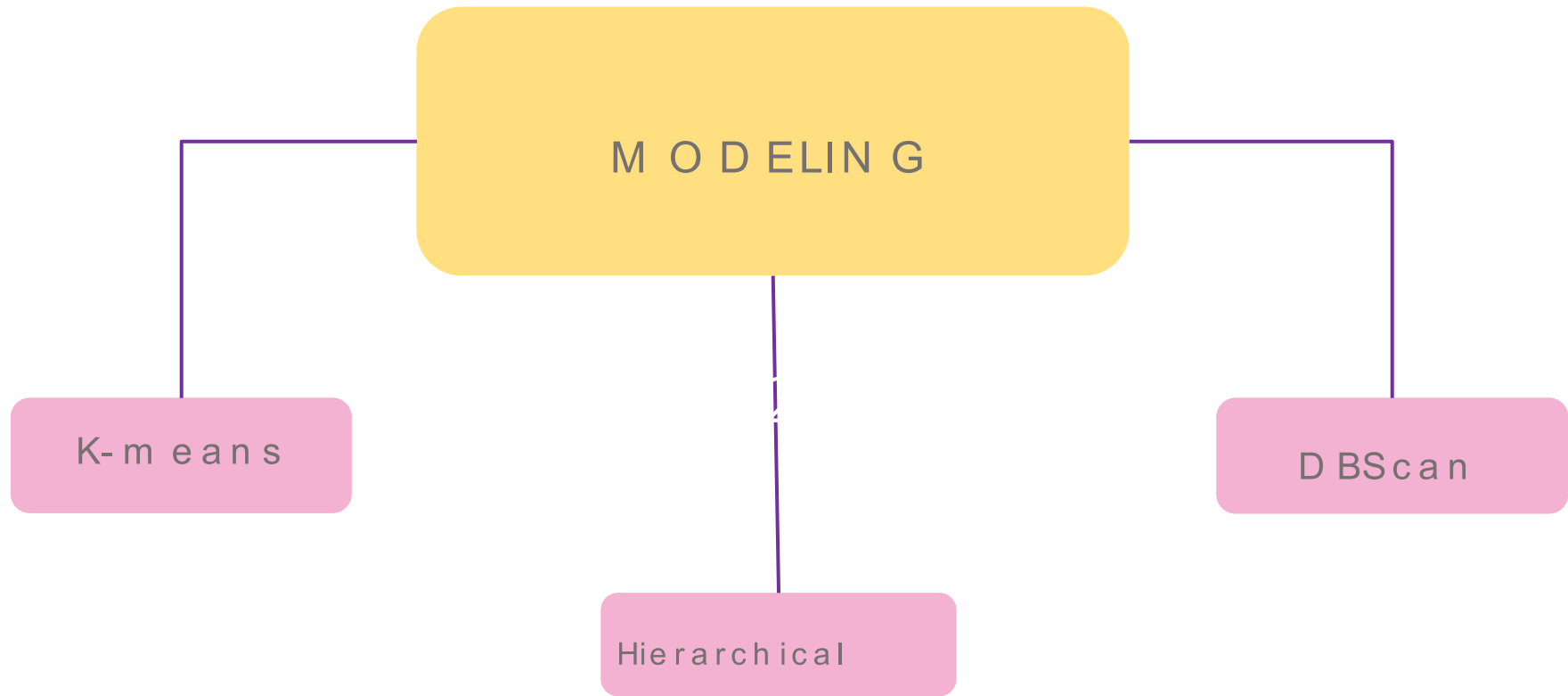
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The First 7 Texts after lemmatization:

```
['go', 'jurong', 'point', 'crazy', 'available', 'bugis', 'n', 'great', 'world', 'la', 'e', 'buffet', 'cine', 'get',  
'amore', 'wat']  
['ok', 'lar', 'joke', 'wif', 'u', 'oni']  
['free', 'entry', 'wkly', 'comp', 'win', 'fa', 'cup', 'final', 'tkts', 'st', 'may', 'text', 'fa', 'receive', 'entr  
y', 'question', 'std', 'txt', 'rate', 'c', 'apply']  
['u', 'dun', 'say', 'early', 'hor', 'u', 'c', 'already', 'say']  
['nah', 'think', 'go', 'usf', 'live', 'around', 'though']  
['freemsg', 'hey', 'darling', 'week', 'word', 'back', 'like', 'fun', 'still', 'tb', 'ok', 'xxx', 'std', 'chgs', 'sen  
d', 'rcv']  
['even', 'brother', 'like', 'speak', 'treat', 'like', 'aid', 'patent']
```



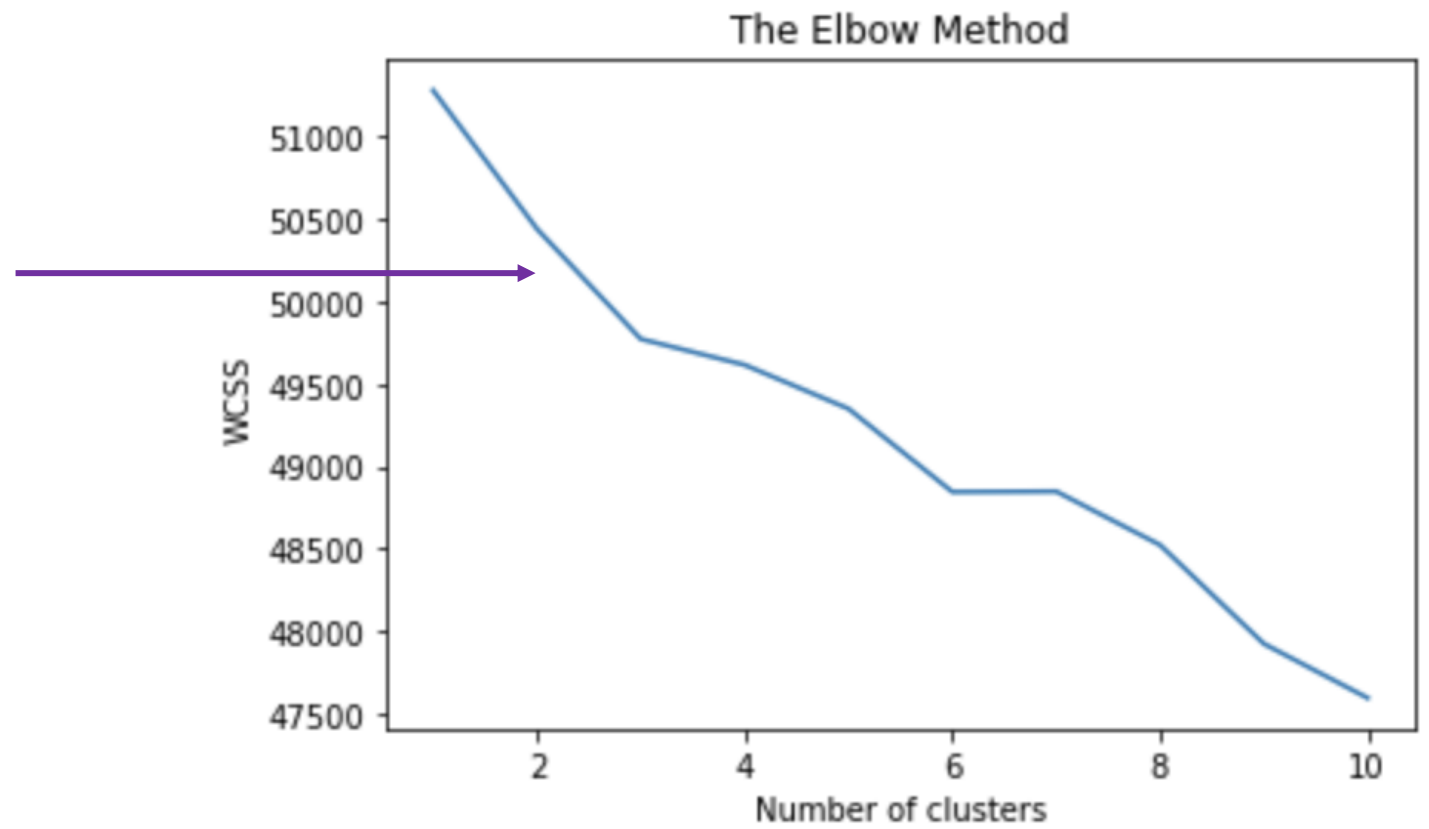




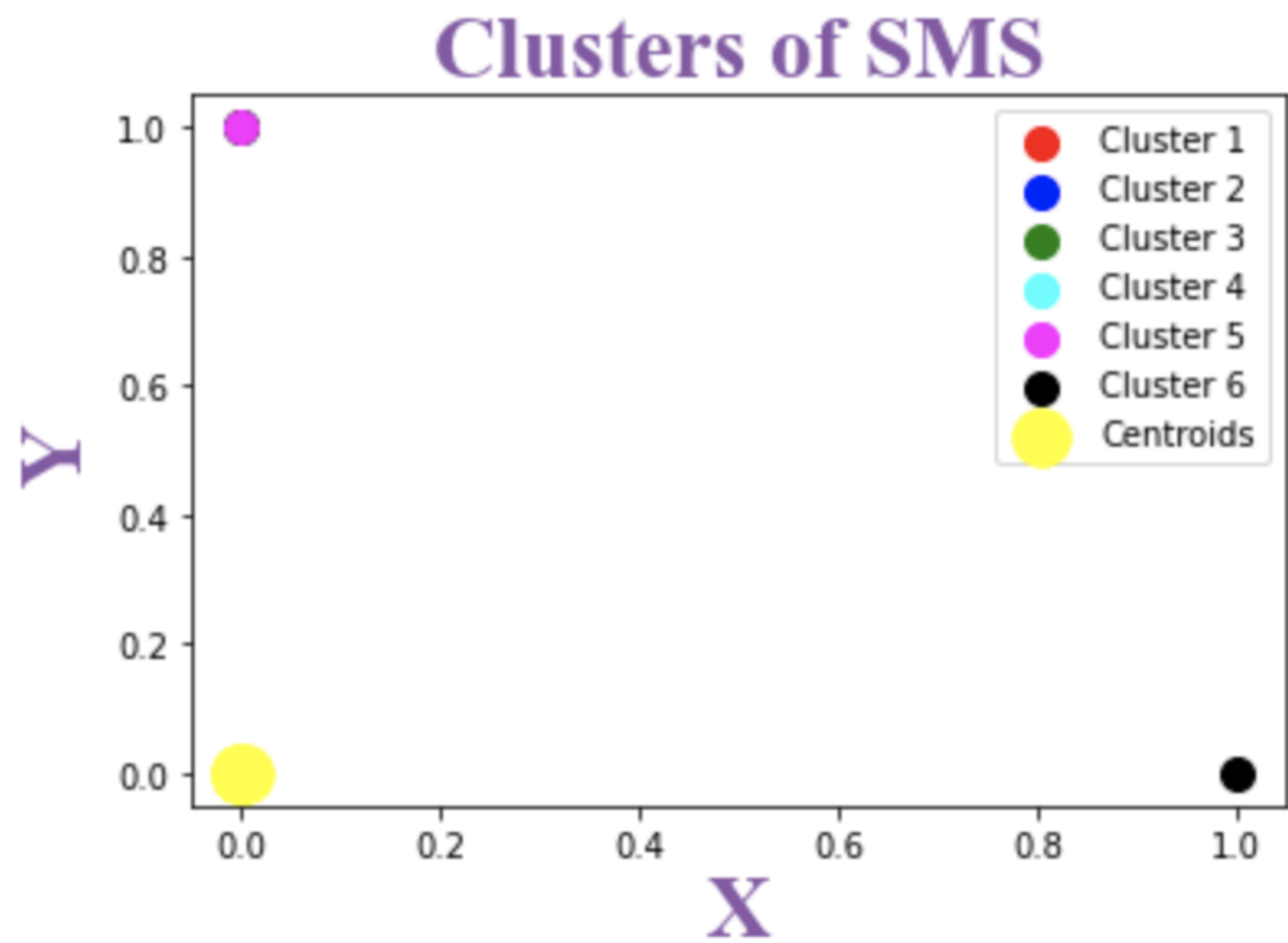


# Elbow

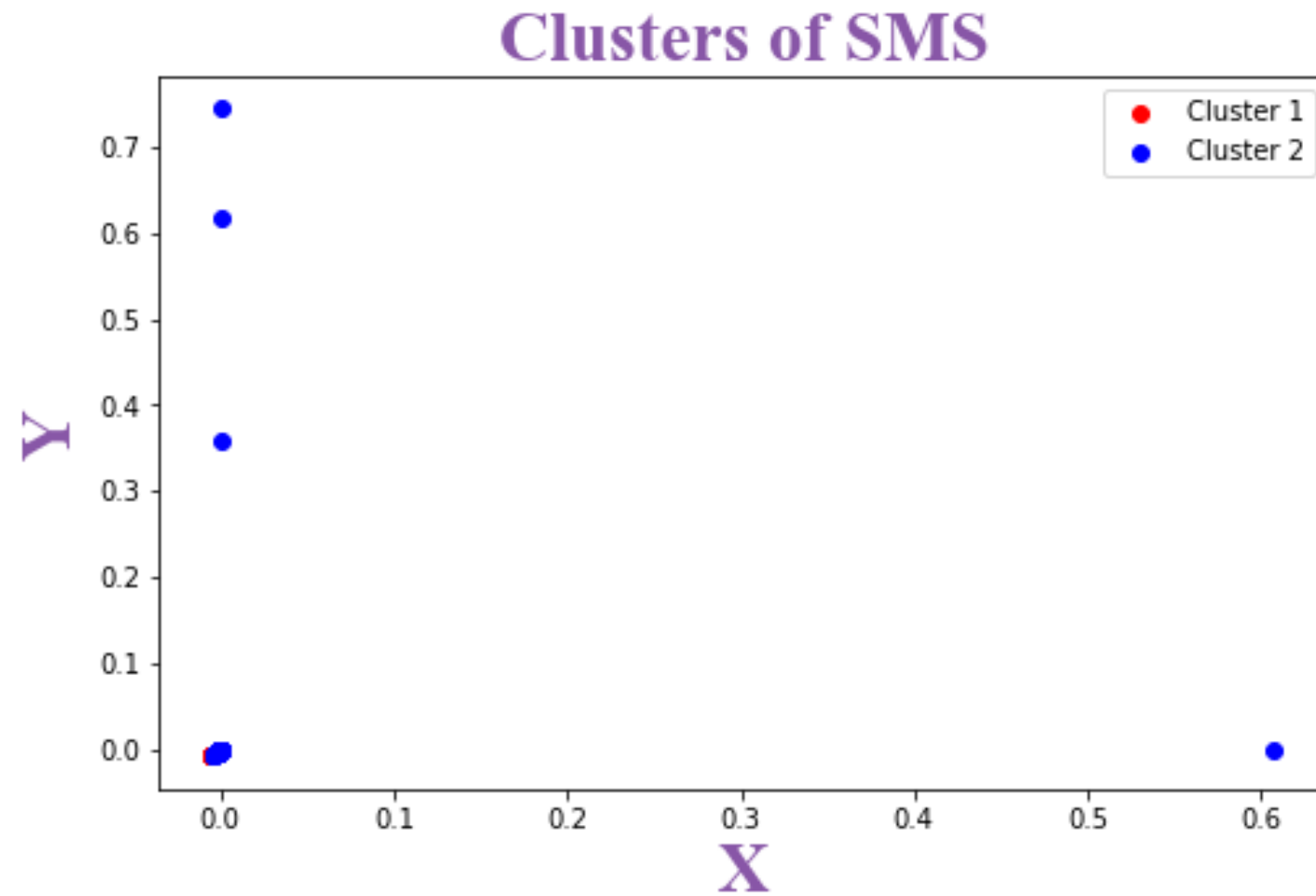
We use the Elbow Function To find the number of Cluster



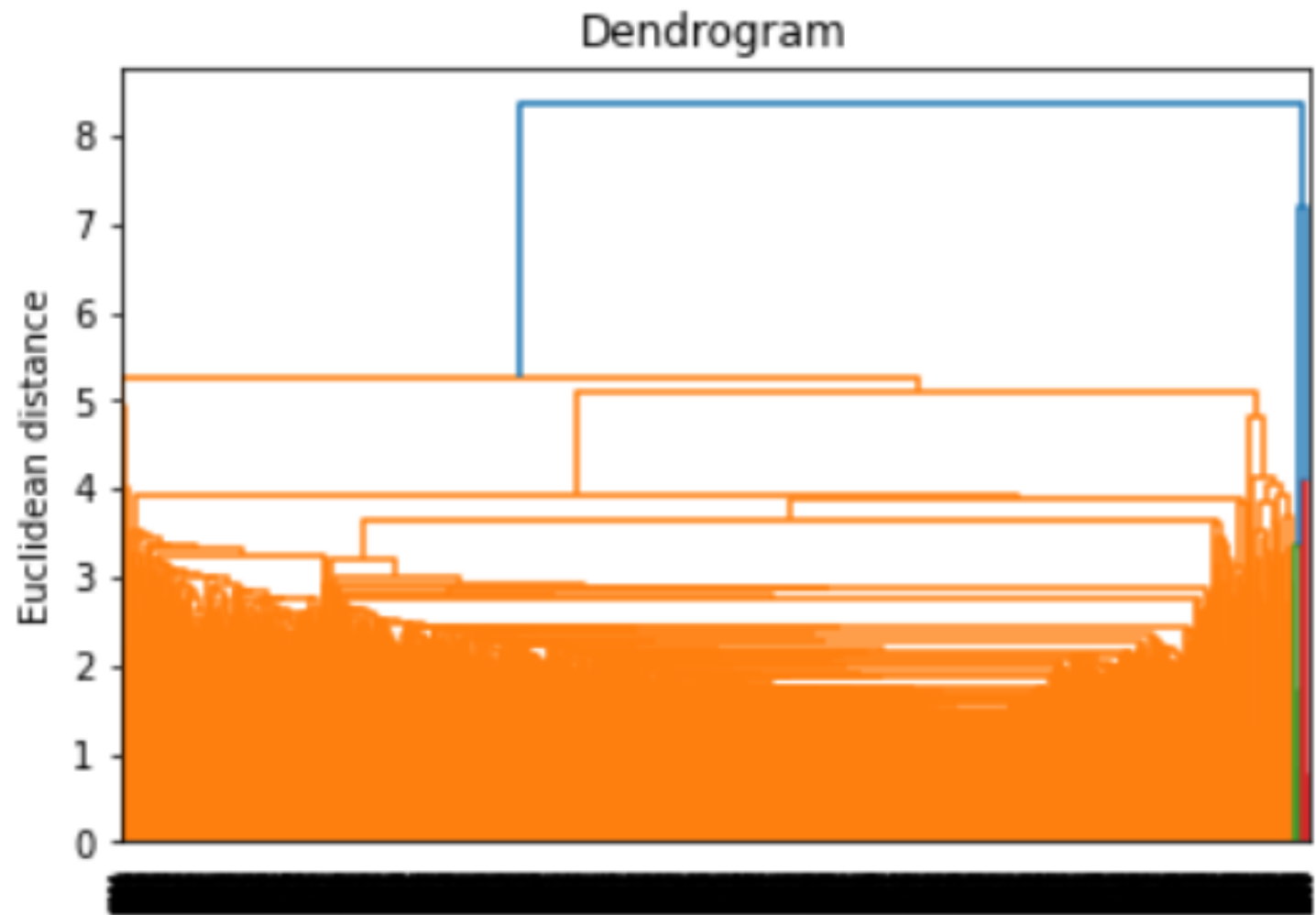
# K-means Model



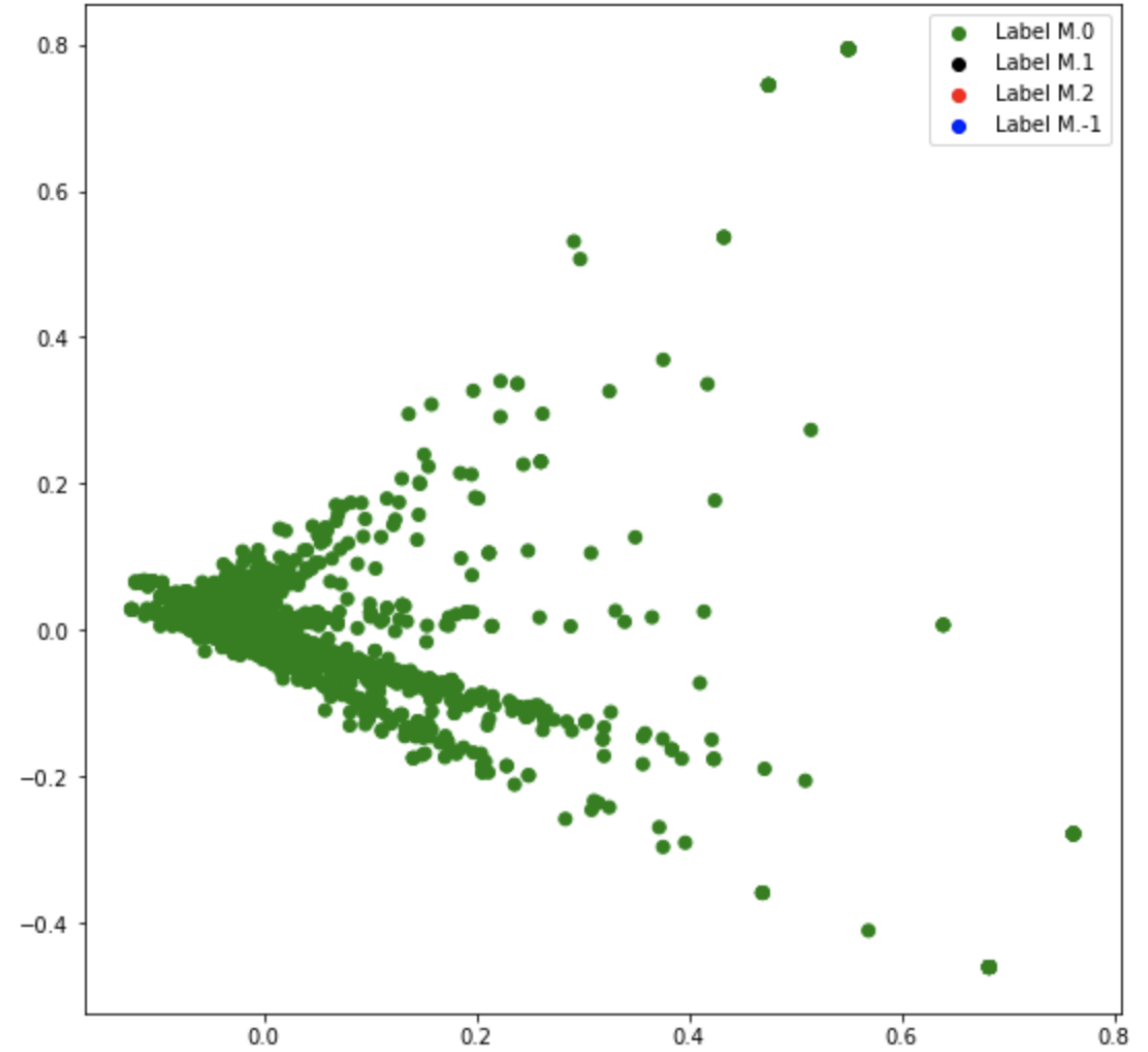
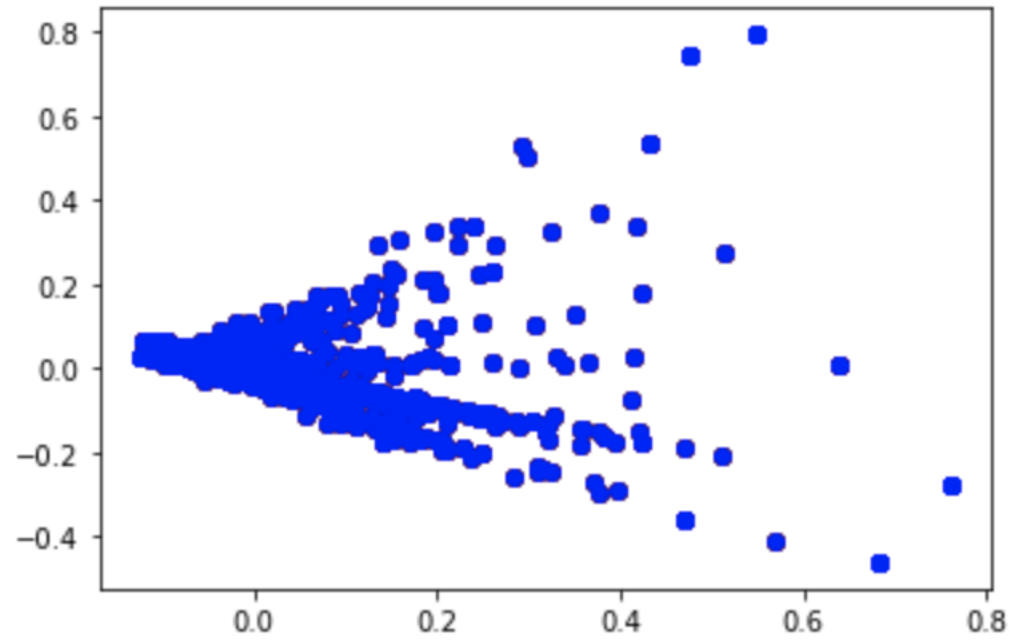
# Hierarchical Model

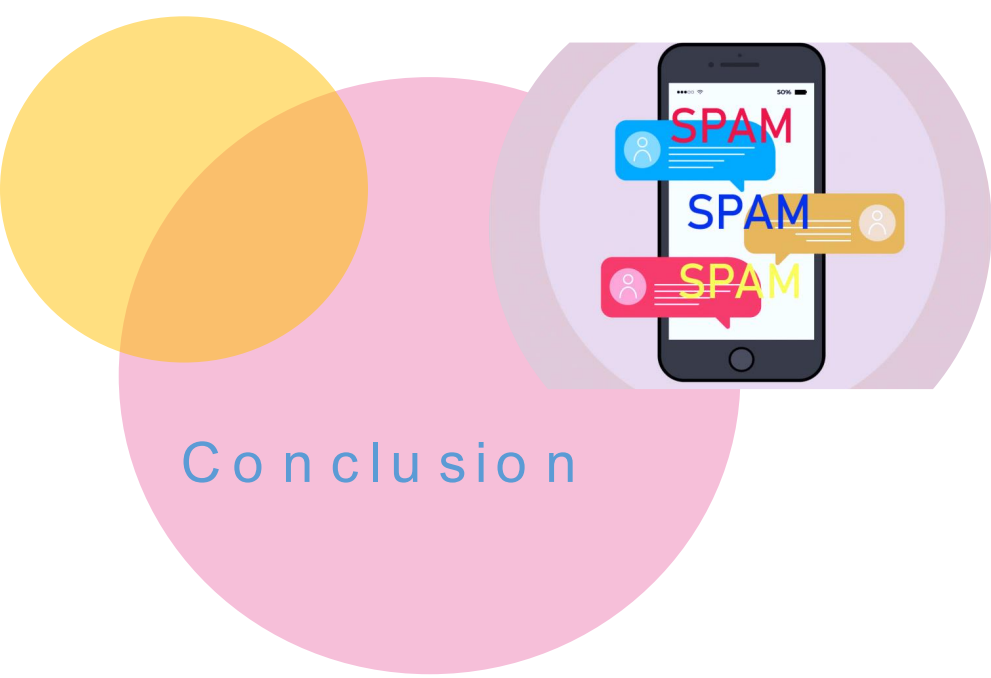


# Hierarchical Model



# DBScan Model

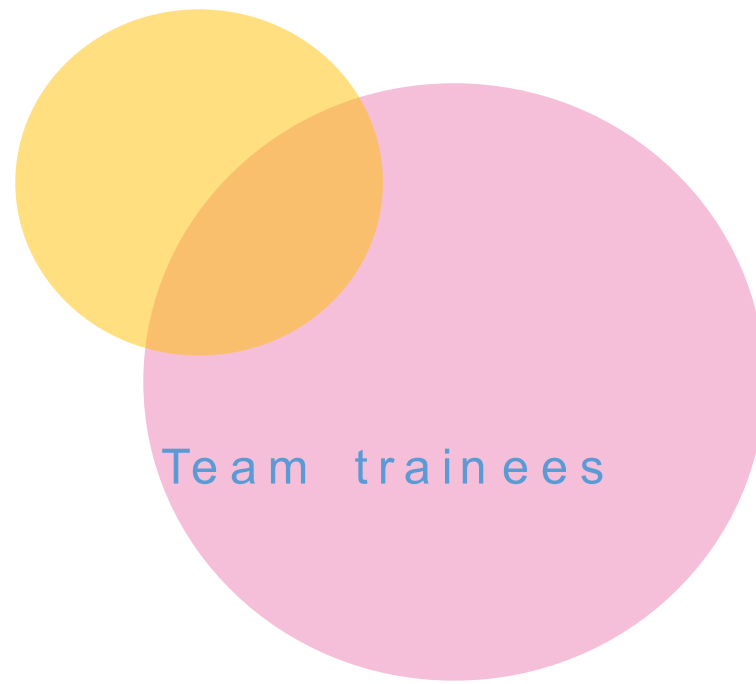




## Conclusion

After analyzing the text and cleaning by NLP Steps, we found out that the text consists of two types (spam, ham SMS) according to the cluster models. so we will help our client to select the ham messages and ignore spam messages by ML tools.





## Team trainees

- N a d a Alq a b b a n i
- Sh a h a d Alm u b k i
- N a d a Alh a m a d
- Sarah Alam e e r
- Hala Alm ulh im

### **Instructor:**

M o h a m m e d B a d d a r