# Segmentation of text with emojis



**Presented by**: Zakharov Artem, Shatalov Andrey and Keseli Timur, БПАД212

Project supervisor: Titova Natalia Nikolaevna

## **RACI Matrix**

TASK	Zakharov Artem	Shatalov Andrew 🤩	Keseli Timur
Data Preprocessing	R	R	R
Data Analysis	R	R	R
KNN	1	I	R
BERTweet	1	R	I
MLP	I	R	I
Siamese	R	I	I
Presentation	R	R	R

## Project aim

Combine text and emoji analysis in text segmentation. Our dataset requires to classify tweets based on text and emoji into 4 classes:

0: sad 😿

1: happy 🤣

2: angry 👪

3: love 🥰

#### Dataset

This dataset contains 3085 Twitter tweets labelled with 0-3 values, where 0= sad, 1 = happy, 2 = angry and 3 = love

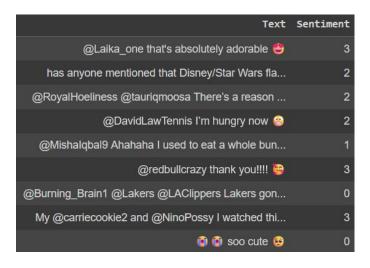


Figure 1. Data Samples.

## Data preprocessing

- 1. Cleaning sentences
- 2. Deleting stop words
- 3. Lemmatization
- 4. Extracting emojis
- 5. Vectorisation of words / emojis
- 6. Concatenation of vectors (dimension = 600)

## Data preprocessing

Text	cleaned_text	lemmatized_text	emojis	label
I'm already starting and it's all upwards and	im already starting and its all upwards and on	im already starting upwards onwards	[😬]	2
The Chinese style. 😞	the chinese style	chinese style	[2]	0
Just WT 🐶 ? 🤢 M feeds her magpies on the bedroom	just wt m feeds her magpies on the bedroom win	wt feed magpie bedroom window sill thought hac	[ <b>ড</b> , 😉, 🥜]	2
if i unfollow or unfriend you, dont take it pe	if i unfollow or unfriend you dont take it per	unfollow unfriend dont take personally ayoko I	[😚]	3
@CallMeAgent00 Thanks man 😔 I've entered 27272	thanks man ive entered giveaways in my time of	thanks man ive entered giveaway time living	[@]	0
if I'm not like this next Christmas it's over	if im not like this next christmas its over ca	im like next christmas cause ima pissed	[3]	2
"Turkey's president has warned that he would e	turkeys president has warned that he would evi	turkey president warned would evict u force tw	0	1
it doesn't feel like Christmas 🥴	it doesnt feel like christmas	doesnt feel like christmas	[@]	0
we were literally all in tears 🔞	we were literally all in tears	literally tear	[0]	0
my boyfriend got me the best gifts ever!!!!!!	my boyfriend got me the best gifts ever first	boyfriend got best gift ever first one got boo	[😉]	0

Figure 2. Data before and after Preprocessing.

### Data preprocessing, Word2Vec

#### Word2Vec Pre-trained model:

- Google News: pre-trained vectors trained on part of Google News dataset (about 100 billion words).
- The model contains 300-dimensional vectors for 3 million words and phrases.

#### **Emoji2Vec Pre-trained model:**

- Proved to have better accuracy with Google News Word2Vec on analysing emoji texts and encodes 300 dimensional vectors.
- Contains description of 1661 emoji symbols.

### Data analysis

t-SNE Visualization of Most Frequent Words and Emojis

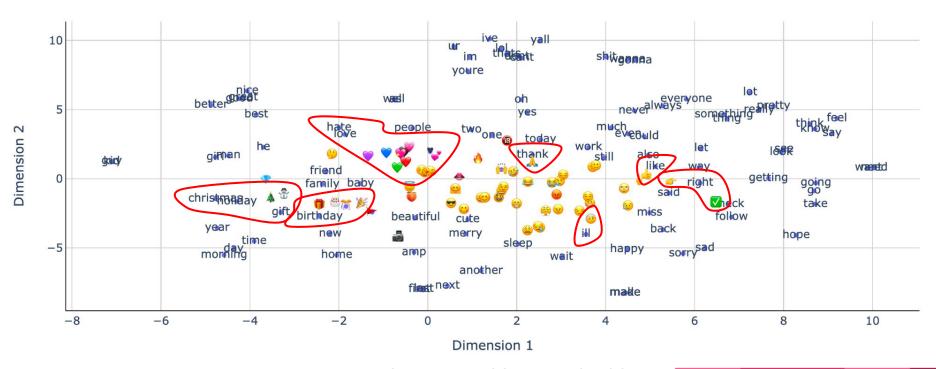


Figure 3. Stochastic neighbour embedding.

## Data analysis

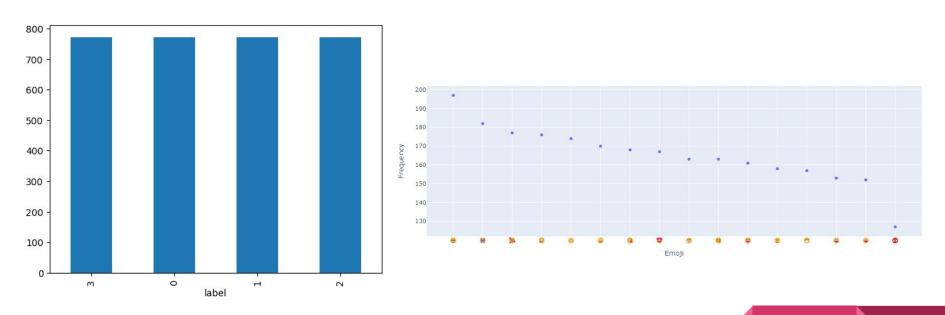


Figure 4 & 5. Class and Emoji Distribution.

### Models and metrics

- 1. KNN
- 2. Random Forests on BERT Embeddings
- 3. SIAMESE
- 4. SimpleNN

- 1. Accuracy
- 2. Precision
- 3. Recall
- 4. F1 Score

### **KNN**

Implements K-Nearest Neighbors for classification.

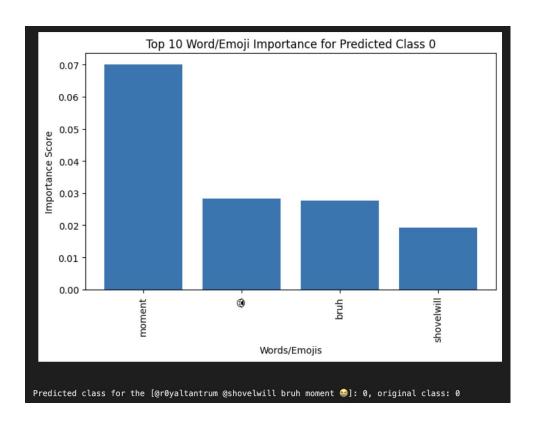
- Uses 4 neighbors with Euclidean distance as the metric.
- Trained on normalized input data to classify sentiment.

### Siamese

Siamese Neural Network with triplet loss for classification.

- Normalizes input data to improve model performance and convergence.
- Input data is reshaped to add a time dimension for convolutional processing.
- Consists of two shared convolutional layers for feature extraction.
- Employs a fully connected dense layer for generating embeddings.
- Utilizes cosine similarity on sNN embeddings for classification.

### Word Importance for SiameseNN



- Compute mean embeddings
- Calculate the gradient of the similarity score with respect to the input features (words and emojis).
- The absolute values of these gradients indicate how much each feature (word/emoji) contributes to the similarity score.

### BERTweet + Random forest

- Performs data augmentation on text data using synonym replacement.
- Extracts emojis from text using regular expressions.
- Combines text and emoji embeddings using BERTweet.
- Trains a Random Forest model to classify sentiment.

0

### **MLP**

Utilizes a Multi-Layer Perceptron (MLP) with multiple dense layers.

- Four hidden layers with ReLU activation functions.
- Dropout layers for regularization to prevent overfitting.
- Output layer uses softmax activation for multi-class classification.
- Trained using the Adam optimizer.

## Results

	Accuracy	Precision	Recall	F1 Score
KNN	0.7277	0.7286	0.7277	0.7280
BERTweet + RF	0.7615	0.7632	0.7615	0.7611
MLP	0.7763	0.7818	0.7763	0.7771
Siamese	0.7812	0.8086	0.7812	0.7855

### References

- 1. Dataset, Kaggle, <a href="https://www.kaggle.com/datasets/juyana054/sentiment-data-16-emoji">https://www.kaggle.com/datasets/juyana054/sentiment-data-16-emoji</a>
- Code, Google Colaboratory,
- 3. Emoji2Vec, GitHub, <a href="https://github.com/uclnlp/emoji2vec">https://github.com/uclnlp/emoji2vec</a>
- 4. Google News vector model, GitHub, <a href="https://github.com/mmihaltz/word2vec-GoogleNews">https://github.com/mmihaltz/word2vec-GoogleNews</a>