



# Data Science Intern at Data Glacier

**Project:** Hate Speech Detection using Transformers (Deep Learning)

**Week 8:** Deliverables

**Name:** Amir Ali

**University:** Warsaw University of Technology

**Email:** [amirali.researcher@gmail.com](mailto:amirali.researcher@gmail.com)

**Country:** Poland

**Specialization:** Data Science

**Batch Code:** LISUM08

**Date:** 25 May 2022

**Submitted to:** Data Glacier

## Table of Contents:

1. Project Plan .....	3
2. Problem Statement .....	3
3. Data Collection .....	3
4. Data Preprocessing .....	4
4.1.Text Cleaning .....	4
4.1.1. Lower Case .....	4
4.1.2. Remove Punctuation .....	4
4.1.3. Remove URL.....	4
4.1.4. Remove @tags .....	4
4.1.5. Remove Special Characters .....	4

## 1. Project Plan

Weeks	Date	plan
Weeks 07	May 18, 2022	Problem Statement, Data Collection, Data Report
Weeks 08	May 25, 2022	Data Preprocessing (Text Cleaning)
Weeks 09	June 1, 2022	Data Preprocessing (Preprocessing Operation + Feature Extraction)
Weeks 10	June 8, 2022	Building the Model
Weeks 11	June 14, 2022	Model Result Evaluation
Weeks 12	June 21, 2022	Flask Development + Heroku
Weeks 13	June 30, 2022	Final Submission (Report + Code + Presentation)

## 2. Problem Statement

The term hate speech is understood as any type of verbal, written or behavioural communication that attacks or uses derogatory or discriminatory language against a person or group based on what they are, in other words, based on their religion, ethnicity, nationality, race, color, ancestry, sex or another identity factor. In this problem, we will take you through a hate speech detection model with Machine Learning and Python.

Hate Speech Detection is a task of sentiment classification. So, for training, a model that can classify hate speech from a certain piece of text can be achieved by training it on a data that is used to classify sentiments. So, for the task of hate speech detection model, we will use the Twitter tweets to identify tweets containing Hate speech.

## 3. Data Collection

The Data is about Twitter hate Speech taken from Kaggle [1] which contains the 3 number of features and 31962 number of observations. Dataset using Twitter data, it was used to research hate-speech detection. The text is classified as: hate-speech, offensive language, and neither. Due to the nature of the study, it is important to note that this dataset contains text that can be considered racist, sexist, homophobic, or offensive.

Table 1: Data Information

<b>Total number of observations</b>	31962
<b>Total number of files</b>	1
<b>Total number of features</b>	3
<b>Base format of the file</b>	csv
<b>Size of the data</b>	2.95 MB

## **4. Data Preprocessing**

In part, we explain the data preprocessing approach that we apply in the text data.

### **4.1 Text Cleaning**

First, we clean our text because it was so messy data.

#### **4.1.1 Lowercase**

Converting a word to lower case (NLP -> nlp). Words like Racism and racism mean the same but when not converted to the lower case those two are represented as two different words in the vector space model (resulting in more dimensions). Therefore, we convert all text word into lower case letter.

#### **4.1.2 Remove Punctuation**

It is important to remove the Punctuation because is not important. Therefore, we remove that Punctuation in order to do that we use regular expression.

#### **4.1.3 Remove URLs**

In this part, we remove URLs because we are working on hate speech application which detect the hate and free speech and to get the output, we need to give only text not URLs therefore, we remove the URLs because we need only clean text input.

#### **4.1.4 Remove @tags**

In this part, we remove @tags which basically used when we mentioned someone So, it's doesn't concern to our application therefore, we remove @tags by using regular expressions.

#### **4.1.5 Remove Special Characters**

Remove Special Characters is essentially the following set of symbols [!''#\$%&'()\*+,-./:;<=>?@[^\_`{|}~] which basically don't have meaning. Therefore, we remove that kind of symbols because we don't need that. In order to remove we use python isalnum method.

## Reference

- [1] [https://www.kaggle.com/datasets/vkrahul/twitter-hate-speech?select=train\\_E6oV3lV.csv](https://www.kaggle.com/datasets/vkrahul/twitter-hate-speech?select=train_E6oV3lV.csv)