

# Project Proposal Template: NLP-Based Sentiment Analysis

Your Name(s)

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## 1 Introduction

[Briefly introduce the problem you are tackling.] Sentiment analysis is a crucial task in Natural Language Processing (NLP), enabling automatic classification of text into sentiment categories. This project aims to explore different machine learning models and pre-processing techniques to improve sentiment classification on social media texts.

### 1.1 Research Questions

[Clearly state your research questions. Example:]

- **RQ1:** Does the model architecture significantly affect sentiment analysis accuracy?
- **RQ2:** Do pre-processing techniques (e.g., stop-word removal, stemming) improve model performance?

## 2 Methodology

### 2.1 Technical Approach

[Describe the step-by-step approach for the project. Example steps:]

1. **Data Preprocessing:** Cleaning data by removing URLs, punctuation, and emojis.
2. **Feature Extraction:** Using TF-IDF for logistic regression and BERT embeddings for the pretrained model.
3. **Model Training:** Training both models and fine-tuning the last layer of BERT.
4. **Evaluation:** Comparing models using cross-validation.



Figure 1: [Insert an architecture diagram here. You can use tools like [draw.io](https://draw.io) to create a simple diagram.]

### 2.2 NLP Techniques

[List and explain the techniques you will use. Example:]

- **Logistic Regression:** A baseline model using TF-IDF.
- **BERT:** A pre-trained transformer model for sentiment analysis.
- **Preprocessing Techniques:** Tokenization, stop-word removal, and lemmatization.

## 3 Team Contributions

[Define who will do what. Example:]

### 3.1 Shared Responsibilities

**All Members:** Data gathering, pre-processing, train-test split creation, and final project poster preparation.

**Sample Deliverables:**

- A git repository with all materials to verify the experiments.
- A poster for the poster session.

### 3.2 Individual Responsibilities

#### 3.2.1 Student 1

**Role:** Implement and train the logistic regression model.

**Deliverables:** Trained logistic regression model.

#### 3.2.2 Student 2

**Role:** Implement and train the BERT model.

**Deliverables:** Trained BERT model.

#### 3.2.3 Student 3

**Role:** Develop the evaluation framework and calculate performance metrics.

**Deliverables:** Detailed evaluation comparing both models.

## 4 Evaluation and Dataset

### 4.1 Dataset Description

[Specify which dataset(s) you will use, consider sources such as:]

- [Hugging Face Datasets](#)
- [Kaggle Datasets](#)
- [Papers with Code Datasets](#)

[Example dataset details:] We are using the [Sentiment140\[1\]](#) dataset containing 1.4 million tweets.

- **target:** Sentiment class (0 = negative, 2 = neutral, 4 = positive).
- **id:** Unique sample identifier.
- **date:** Timestamp of the tweet.
- **user:** Username of the author.
- **text:** Tweet content.

Target	ID	Date	User	Text
0	1467811184	Mon Apr 06 22:19:57 PDT 2009	ElleCTF	My whole body feels itchy and like it's on fire.

Figure 2: A sample row from a sentiment dataset.

### 4.2 Experimental Setup

[Define how you will evaluate your models. Example metrics:]

We will evaluate using the following metrics on our train / validation / test split. For this we once evaluate BERT, then logistic regression and then a combination of different pre-processing steps.

- Accuracy
- F1 Score
- ROC-AUC Score

## References

- [1] A. Go, R. Bhayani, and L. Huang, “Twitter sentiment classification using distant supervision,” *CS224N Project Report, Stanford*, vol. 1, p. 12, 2009. Accessed: 2025-03-27.