#### Hindi Sentiment Analysis

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# **Problem Statement**

Sentiment analysis, also known as opinion mining, is a natural language processing (NLP) task that involves determining the sentiment or emotion expressed in a piece of text. This task is crucial for understanding public opinion, customer feedback, and social media discussions. The focus of this problem statement is to perform sentiment analysis on Hindi text data, which is a critical task in understanding sentiment within the Hindi-speaking community.

## **Technologies Used**

- 1. **Python:** The code is written in the Python programming language, which is widely used in data analysis, machine learning, and NLP tasks.
- 2. **Pandas:** The pandas library is used for data manipulation and analysis. It provides data structures like DataFrames, making it easier to work with structured data.
- 3. **Scikit-Learn (sklearn):** Scikit-Learn is a popular machine learning library in Python. In your code, you use several modules from Scikit-Learn, including:
- 4. **train\_test\_split:** This module is used to split the dataset into training and testing sets.
- 5. **TfidfVectorizer:** It is used to convert the text data into TF-IDF (Term Frequency-Inverse Document Frequency) features, which are essential for training a text classification model.
- 6. **SVC** (Support Vector Classification): This is the machine learning algorithm used for classification. In your code, you use an SVM classifier with a linear kernel.
- 7. accuracy score: It is used to calculate the accuracy of the model.
- 8. **classification\_report:** This module provides a comprehensive report on various classification metrics.
- 9. **Joblib:** The joblib library is used for model serialization and deserialization. It is used to save and load machine learning models and other objects efficiently.
- 10. **CSV Data:** Your code reads data from a CSV file ('emotions.csv') using Pandas. CSV (Comma-Separated Values) is a common format for storing structured data.
- 11. **Text Preprocessing:** Although not explicitly mentioned in your code, text preprocessing is an essential step in NLP. Preprocessing may include tasks like text cleaning, tokenization, and handling stopwords.
- 12. **Machine Learning Model:** Your code uses a Support Vector Machine (SVM) model for text classification. SVMs are a popular choice for binary and multi-class classification tasks, including sentiment analysis.
- 13. **TF-IDF:** The TF-IDF vectorization technique is used to convert text data into numerical features, which are then used as input for the SVM model. This technique is essential for transforming text data into a format that machine learning models can work with.
- 14. **Model Serialization:** The code saves the trained model and TF-IDF vectorizer to files using joblib for later use. This is a common practice in machine learning to avoid retraining the model every time.

# **Input Dataset**

Link for the dataset:

https://www.kaggle.com/datasets/chiragmvarma/hindi-sentiment-analysis/data

### **Source code:**

```
import joblib
import pandas as pd
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report

data = pd.read_csv('C:/Users/umang/Downloads/NLP+project -
Copy/NLP+project/emotions.csv')

X = data['Sentences']
y = data['Label']

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

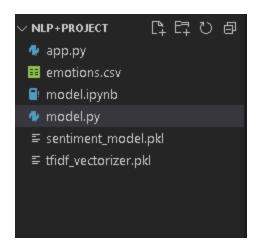
tfidf_vectorizer = TfidfVectorizer(max_features=5000)
X_train_tfidf = tfidf_vectorizer.fit_transform(X_train)
X_test_tfidf = tfidf_vectorizer.transform(X_test)
```

```
model = SVC(kernel='linear', C=1.0, probability=True)
model.fit(X train tfidf, y train)
y pred = model.predict(X test tfidf)
accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy:.2f}")
report = classification report(y test, y pred)
print("Classification Report:\n", report)
joblib.dump(model, 'sentiment_model.pkl')
joblib.dump(tfidf vectorizer, 'tfidf vectorizer.pkl')
loaded model = joblib.load('sentiment model.pkl')
loaded vectorizer = joblib.load('tfidf vectorizer.pkl')
new text = ["This is amazing!", "Worst experience ever."]
new text tfidf = loaded vectorizer.transform(new text)
predictions = loaded model.predict(new text tfidf)
print("Predictions:", predictions)
print("Model Accuracy:", accuracy)
print("Classification Report:\n", report)
```

```
import streamlit as st
     import joblib
     import pandas as pd
     model = joblib.load('sentiment model.pkl')
     tfidf_vectorizer = joblib.load('tfidf_vectorizer.pkl')
     st.title('Hindi Sentiment Analysis Web App')
     user input = st.text area('Enter your text in Hindi')
     def analyze sentiment(text):
          text tfidf = tfidf vectorizer.transform([text])
         prediction = model.predict(text_tfidf)
         return prediction[0]
     if st.button('Analyze Sentiment'):
         if user input:
              sentiment = analyze sentiment(user input)
             st.write('Sentiment:', sentiment)
         else:
              st.warning('Please enter some text for analysis.')
     st.sidebar.markdown("Disclaimer: This is a simplified example using a
small dataset and may not accurately predict sentiment for all inputs.")
```

# **Screenshots**

### **Structure of the project:**



## App.py file:

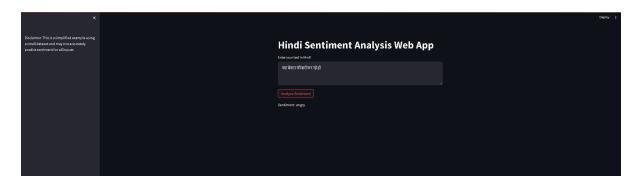
#### **Emotions.csv file:**



#### Model.py file:

### Output:

### Using streamlit.



## **Conclusion**

Hindi sentiment analysis using text vectorization, SVM, model serialization, and rigorous evaluation is a powerful approach to gain insights from Hindi text data. By following these steps, one can develop a robust sentiment analysis model capable of classifying sentiment in Hindi text, opening doors to a wide range of applications in understanding and responding to sentiment within the Hindi-speaking community. This approach combines NLP and machine learning to provide valuable insights and actionable intelligence from textual data.