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# ----- 1) IMPORTS & DATA CLEANING -----
import pandas as pd
import string, re, nltk
from nltk.corpus import stopwords
# Download NLTK stop-word list (one-time per environment)
nltk.download('stopwords')
# Load the CSV that contains the raw text and sentiment labels
df = pd.read csv("legal sentiment dataset.csv") # Columns: Text, Sentiment
# Create a reusable cleaning function
stop words = set(stopwords.words('english'))
def clean text(text: str) -> str:
   text = text.lower()
                                                          # a) lowercase
   text = re.sub(r"\d+", "", text)
                                                          # b) remove numbers
   text = text.translate(str.maketrans("", "", string.punctuation)) # c) strip punctuation
   text = " ".join(w for w in text.split() if w not in stop words) # d) remove stop-words
   return text
# Apply cleaning and store in new column
df["clean text"] = df["Text"].apply(clean text)
# Show a small sample
print(df.head(3))
\rightarrow
                                                    Text Sentiment \
       [Case 3708] The appeal is hereby dismissed for... Negative
       [Case 6994] The plaintiff's motion for summary... Positive
       [Case 2793] The legal claim is barred by the s... Negative
                                            clean text
    0
               case appeal hereby dismissed lack merit
      case plaintiffs motion summary judgment granted
           case legal claim barred statute limitations
    [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk data] Package stopwords is already up-to-date!
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# ----- 2) TRAIN-TEST SPLIT & TF-IDF ------
from sklearn.model selection import train test split
from sklearn.feature extraction.text import TfidfVectorizer
# Separate features & labels
X = df["clean text"]
y = df["Sentiment"]
# 80 % training / 20 % test
X train, X test, y train, y test = train test split(
   X, y, test_size=0.20, random_state=42, stratify=y
# TF-IDF converts text to numeric vectors
vectorizer = TfidfVectorizer(max features=5000) # top 5 000 terms
X_train_vec = vectorizer.fit_transform(X_train) # fit + transform (train)
X test vec = vectorizer.transform(X test)
                                                # transform only (test)
print("Training matrix shape:", X_train_vec.shape)
print("Example feature names:", vectorizer.get feature names out()[:10])
    Training matrix shape: (80, 72)
    Example feature names: ['acquitted' 'agreement' 'amicable' 'appeal' 'barred' 'beyond' 'breach'
     'case' 'caused' 'charges']
# ----- 3) MODEL, METRICS, PREDICTION, SAVE -----
from sklearn.linear model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report
import pickle
# a) Train a binary classifier
model = LogisticRegression(max_iter=1000) # more iterations for convergence
model.fit(X train vec, y train)
# b) Evaluate on the held-out test set
y_pred = model.predict(X_test_vec)
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print("Accuracy on test set:", accuracy_score(y_test, y_pred))
print("\nDetailed classification report:\n", classification_report(y_test, y_pred))

# c) Helper for predictions on new sentences
def predict_sentiment(raw_text: str) -> str:
    cleaned = clean_text(raw_text)
    vec = vectorizer.transform([cleaned])
    return model.predict(vec)[0]

print("\nSample inference:",
    predict_sentiment("The injunction is granted to prevent further harm."))

# d) Persist model + vectorizer for later reuse
with open("legal_sentiment_model.pkl", "wb") as f:
    pickle.dump((model, vectorizer), f)
print("\nModel saved as legal sentiment model.pkl")
```

Accuracy on test set: 1.0

Detailed classification report:

	precision	recall	f1-score	support
Negative	1.00	1.00	1.00	9
Positive	1.00	1.00	1.00	11
accuracy			1.00	20
macro avg	1.00	1.00	1.00	20
weighted avg	1.00	1.00	1.00	20

Sample inference: Positive

Model saved as legal_sentiment_model.pkl

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