```
import pandas as pd
import numpy as np
from datasets import load dataset
# Load IMDB dataset
dataset = load dataset("imdb")
# Convert training data to Pandas DataFrame
train_df = pd.DataFrame(dataset["train"])
/usr/local/lib/python3.12/dist-packages/huggingface hub/utils/
auth.py:94: UserWarning:
The secret `HF TOKEN` does not exist in your Colab secrets.
To authenticate with the Hugging Face Hub, create a token in your
settings tab (https://huggingface.co/settings/tokens), set it as
secret in your Google Colab and restart your session.
You will be able to reuse this secret in all of your notebooks.
Please note that authentication is recommended but still optional to
access public models or datasets.
 warnings.warn(
{"model id":"d7e9fa593d5d46fb903651a2c6d9a407","version major":2,"vers
ion minor":0}
{"model id": "59a240a7dfec4d2394118d24a1a126ea", "version major": 2, "vers
ion minor":0}
{"model id": "3c4fe2d038fe47ad9b9bf30d50bf586a", "version major": 2, "vers
ion minor":0}
{"model id":"fc793fccd9db49f093cd82cb6a3b8831","version major":2,"vers
ion minor":0}
{"model id": "76af4ed7005a4752a1bb310880815714", "version_major": 2, "vers
ion minor":0}
{"model id": "5adeb59aacec4eb8ade07d9758c20d7f", "version major": 2, "vers
ion minor":0}
{"model id":"d9142465ea1b46e6bfe64c7d7780415a","version major":2,"vers
ion minor":0}
# First 5 rows
print(train df.head())
# Info about dataset
print("\n--- INFO ---")
print(train df.info())
# Check missing values
print("\n--- NULL VALUES ---")
```

```
print(train df.isnull().sum())
# Check balance of labels
print("\n--- LABEL COUNTS ---")
print(train df["label"].value counts())
# Add text length column
train df["text length"] = train df["text"].apply(len)
print("\n--- TEXT LENGTH STATS ---")
print(train df["text length"].describe())
                                                 text label
   I rented I AM CURIOUS-YELLOW from my video sto...
  "I Am Curious: Yellow" is a risible and preten...
                                                           0
1
  If only to avoid making this type of film in t...
                                                           0
  This film was probably inspired by Godard's Ma...
                                                           0
4 Oh, brother...after hearing about this ridicul...
                                                           0
--- INFO ---
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 25000 entries, 0 to 24999
Data columns (total 2 columns):
     Column Non-Null Count Dtype
#
- - -
0
     text
             25000 non-null
                             object
1
    label
             25000 non-null
                            int64
dtypes: int64(1), object(1)
memory usage: 390.8+ KB
None
--- NULL VALUES ---
text
         0
label
dtype: int64
--- LABEL COUNTS ---
label
     12500
0
1
     12500
Name: count, dtype: int64
--- TEXT LENGTH STATS ---
count
         25000.00000
          1325.06964
mean
std
          1003.13367
min
            52.00000
           702.00000
25%
50%
           979,00000
75%
          1614.00000
```

```
13704.00000
max
Name: text length, dtype: float64
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
# Download necessary resources
nltk.download("stopwords")
nltk.download("punkt")
nltk.download("punkt tab")
nltk.download("wordnet")
# Load stopwords & lemmatizer
stop words = set(stopwords.words("english"))
# Keep negations for sentiment analysis
for keep in ["not", "no", "nor", "n't"]:
    if keep in stop words:
        stop words.remove(keep)
lemmatizer = WordNetLemmatizer()
[nltk data] Downloading package stopwords to /root/nltk data...
[nltk data] Unzipping corpora/stopwords.zip.
[nltk data] Downloading package punkt to /root/nltk data...
[nltk_data] Unzipping tokenizers/punkt.zip.
[nltk data] Downloading package punkt tab to /root/nltk data...
[nltk_data] Unzipping tokenizers/punkt_tab.zip.
[nltk data] Downloading package wordnet to /root/nltk_data...
def clean text(text):
    # 1. Lowercase
    text = text.lower()
    # 2. Remove HTML tags
    text = re.sub(r"<.*?>", "", text)
    # 3. Remove punctuation & numbers
    text = re.sub(r"[^a-z\s]", "", text)
    # 4. Tokenize
    words = nltk.word tokenize(text)
    # 5. Remove stopwords & lemmatize
    words = [lemmatizer.lemmatize(w) for w in words if w not in
stop words]
    return " ".join(words)
# Apply cleaning
```

```
train_df["clean_text"] = train_df["text"].apply(clean_text)
# Show before & after for first review
print("Before:\n", train_df["text"].iloc[0])
print("\nAfter:\n", train_df["clean_text"].iloc[0])
```

## Before:

I rented I AM CURIOUS-YELLOW from my video store because of all the controversy that surrounded it when it was first released in 1967. I also heard that at first it was seized by U.S. customs if it ever tried to enter this country, therefore being a fan of films considered "controversial" I really had to see this for myself.<br /><br />The plot is centered around a young Swedish drama student named Lena who wants to learn everything she can about life. In particular she wants to focus her attentions to making some sort of documentary on what the average Swede thought about certain political issues such as the Vietnam War and race issues in the United States. In between asking politicians and ordinary denizens of Stockholm about their opinions on politics, she has sex with her drama teacher, classmates, and married men.<br/>
/><br />What kills me about I AM CURIOUS-YELLOW is that 40 years ago, this was considered pornographic. Really, the sex and nudity scenes are few and far between, even then it's not shot like some cheaply made porno. While my countrymen mind find it shocking, in reality sex and nudity are a major staple in Swedish cinema. Even Ingmar Bergman, arguably their answer to good old boy John Ford, had sex scenes in his films.<br />spr />I do commend the filmmakers for the fact that any sex shown in the film is shown for artistic purposes rather than just to shock people and make money to be shown in pornographic theaters in America. I AM CURIOUS-YELLOW is a good film for anyone wanting to study the meat and potatoes (no pun intended) of Swedish cinema. But really, this film doesn't have much of a plot.

## After:

rented curiousyellow video store controversy surrounded first released also heard first seized u custom ever tried enter country therefore fan film considered controversial really see myselfthe plot centered around young swedish drama student named lena want learn everything life particular want focus attention making sort documentary average swede thought certain political issue vietnam war race issue united state asking politician ordinary denizen stockholm opinion politics sex drama teacher classmate married menwhat kill curiousyellow year ago considered pornographic really sex nudity scene far even not shot like cheaply made porno countryman mind find shocking reality sex nudity major staple swedish cinema even ingmar bergman arguably answer good old boy john ford sex scene filmsi commend filmmaker fact sex shown film shown artistic purpose rather shock people make money shown pornographic theater america curiousyellow good film anyone wanting study meat potato no pun intended swedish cinema really film doesn't much plot

```
from sklearn.model selection import train test split
from sklearn.feature extraction.text import TfidfVectorizer
# 1. Train-test split (80% train, 20% test)
X_train, X_test, y_train, y_test = train_test_split(
    train_df["clean_text"], train_df["label"],
    test size=0.2,
    random state=42
print("Train size:", len(X_train))
print("Test size:", len(X_test))
# 2. TF-IDF Vectorization
vectorizer = TfidfVectorizer(max features=5000) # limit vocab to top
5000 words
X train vec = vectorizer.fit transform(X train)
X test vec = vectorizer.transform(X test)
print("Vectorized shape (train):", X_train_vec.shape)
print("Vectorized shape (test):", X_test_vec.shape)
Train size: 20000
Test size: 5000
Vectorized shape (train): (20000, 5000)
Vectorized shape (test): (5000, 5000)
from sklearn.linear model import LogisticRegression
from sklearn.metrics import accuracy score, classification report
# 1. Initialize model
model = LogisticRegression(max iter=1000)
# 2. Train the model on training data
model.fit(X train vec, y train)
# 3. Make predictions on test data
y pred = model.predict(X test vec)
# 4. Evaluate
print("Accuracy:", accuracy_score(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test,
y pred))
Accuracy: 0.8802
Classification Report:
               precision recall f1-score support
                   0.89
                             0.87
                                       0.88
                                                 2515
           1
                   0.87
                             0.89
                                       0.88
                                                 2485
```

```
0.88
                                                 5000
    accuracy
                             0.88
                                       0.88
                                                 5000
   macro avq
                   0.88
                   0.88
                             0.88
                                       0.88
                                                 5000
weighted avg
# Test with your own custom reviews
sample reviews = [
    "I hated this movie, waste of time!", # Positive
    # "The film was too long and very boring.",
                                                          # Negative
    # "An average movie, not too bad but not great either." # Neutral-
like
1
# Clean the reviews (using the same function as before)
sample cleaned = [clean text(r) for r in sample reviews]
# Vectorize with the SAME vectorizer used earlier
sample vec = vectorizer.transform(sample cleaned)
# Predict
predictions = model.predict(sample vec)
# Show results
for review, pred in zip(sample reviews, predictions):
    sentiment = "Positive ☺" if pred == 1 else "Negative ☺"
    print(f"Review: {review}\nPredicted Sentiment: {sentiment}\n")
Review: I hated this movie, waste of time!
Predicted Sentiment: Negative 😂
def predict sentiment(review):
    # Step 1: Clean the input review
    cleaned = clean text(review)
    # Step 2: Convert it into vector form using the same TF-IDF
vectorizer
    vec = vectorizer.transform([cleaned])
    # Step 3: Predict using the trained model
    pred = model.predict(vec)[0]
    # Step 4: Return the result
    if pred == 1:
        return "Positive ©"
    else:
        return "Negative 😂"
\# \sqcap Test the function
print(predict sentiment("I loved the acting, such a brilliant
```

```
movie!"))
print(predict sentiment("Worst movie ever, I wasted my time."))
print(predict sentiment("hey this movie was great"))
Positive ©
Negative 😩
Positive ©
import gradio as gr
def predict sentiment(review):
    # Clean the text
    clean = clean text(review)
    # Transform text into features
    features = vectorizer.transform([clean])
    # Predict class (0 = Negative, 1 = Positive)
    prediction = model.predict(features)[0]
    # Predict probability
    prob = model.predict_proba(features)[0] # gives [prob_negative,
prob positive]
    # Select probability of predicted class
    confidence = prob[prediction] * 100
    # Final readable output
    sentiment = "Positive ⊕" if prediction == 1 else "Negative ⊕"
    result = f"{sentiment} (Confidence: {confidence:.2f}%)"
    return result
# Create UI
ui = gr.Interface(
    fn=predict sentiment,
    inputs=gr.Textbox(lines=5, placeholder="Type a movie review
here..."),
    outputs="text",
    title="☐ Movie Review Sentiment Analyzer",
    description="Enter a movie review and see if it's Positive or
Negative with confidence score!"
ui.launch()
It looks like you are running Gradio on a hosted Jupyter notebook,
which requires `share=True`. Automatically setting `share=True` (you
can turn this off by setting `share=False` in `launch()` explicitly).
Colab notebook detected. To show errors in colab notebook, set
```

debug=True in launch()
\* Running on public URL: https://becb919aa002439a8a.gradio.live

This share link expires in 1 week. For free permanent hosting and GPU upgrades, run `gradio deploy` from the terminal in the working directory to deploy to Hugging Face Spaces (https://huggingface.co/spaces)

<IPython.core.display.HTML object>