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import pandas as pd
import tensorflow as tf
import joblib
import numpy as np
from sklearn.feature extraction.text import TfidfVectorizer
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad sequences
from app.globals import train set, test set, train labels, test labels
from sklearn.model selection import train test split
from imblearn.under sampling import RandomUnderSampler
def under_sampling(df, target_column):
    X = df.drop(target column, axis=1)
    y = df[target column]
    rus = RandomUnderSampler(sampling strategy='not minority',
random state=101)
    X \text{ res}, y \text{ res} = \text{rus.fit resample}(X, y)
    df_undersampled = pd.DataFrame(X_res, columns=X.columns)
    df undersampled[target column] = y res
    return df undersampled
# Load and process the dataset
# Define the list of CSV file paths
file paths = [
    'data/imdb cleaned 1.csv',
    'data/imdb cleaned 2.csv',
    'data/imdb cleaned 3.csv',
    'data/imdb cleaned 4 5 6.csv',
    'data/imdb cleaned 7.csv',
    'data/imdb cleaned 8.csv',
    'data/imdb cleaned 9.csv',
# Load all CSV files into a single DataFrame
dataframes = [pd.read csv(file, usecols=['review', 'sentiment']) for
file in file paths]
# Concatenate all dataframes
imdb df = pd.concat(dataframes, ignore_index=True)
# Drop any rows with missing values
imdb df.dropna(inplace=True)
imdb df = under sampling(imdb_df, 'sentiment')
review df = imdb df['review'] # DataFrame that contains review text
(feature column)
sentiment df = imdb df['sentiment'] # DataFrame that contains
sentiment label (target column)
# Split the dataset into training and testing sets
train set, test set, train labels, test labels =
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train test split(review df, sentiment df, test size=0.2,
random state=101)
def preprocess text(text):
    return text.lower()
def feature extraction(X train, X test):
    # Replace NaN values with an empty string
    X train = ["" if pd.isna(text) else text for text in X train]
    X test = ["" if pd.isna(text) else text for text in X test]
    vectorizer = TfidfVectorizer()
    X train vectors = vectorizer.fit transform(X train)
    X test vectors = vectorizer.transform(X test)
    return X train vectors, X test vectors
def preprocess data for deep learning(data):
    train_set, test_set, train_labels, test_labels = data
    train set = ["" if pd.isna(text) else str(text) for text in
train set]
    test set = ["" if pd.isna(text) else str(text) for text in
test set]
    vocab size = \max(len(set(text.split()))) for text in train set +
test set)
    all texts = train set + test set
    sequence lengths = [len(text.split()) for text in all texts]
    \max length = \min(\max(\text{sequence lengths}), 500)
    oov tok = <00V>
    tokenizer = Tokenizer(num words=vocab size, oov token=oov tok)
    tokenizer.fit_on_texts(train_set)
    train sequences = tokenizer.texts to sequences(train set)
    test sequences = tokenizer.texts_to_sequences(test_set)
    X train = pad sequences(train sequences, maxlen=max length)
    X test = pad sequences(test sequences, maxlen=max length)
    return X train, X test, train labels, test labels, vocab size,
max length
def pred user sentence(pred sentences, model, tokenizer=None,
type='ml'):
    labels = ['Negative', 'Neutral', 'Positive']
    processed sentences = [preprocess text(sentence) for sentence in
pred sentences]
    if type == 'pretrained':
        encoded input = tokenizer(processed sentences, max length=512,
padding=True, truncation=True, return tensors='pt')
        dummy labels = [0] * len(processed sentences)
        pred dataset = IMDbDataset(encoded input, dummy labels)
        predictions = model.predict(pred dataset)
        y pred = np.argmax(predictions.predictions, axis=-1)
        result labels = [labels[idx] for idx in y pred]
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elif type == 'ml':
        _, transformed_sentences = feature_extraction(train set,
processed sentences)
        label indices = model.predict(transformed sentences)
        result labels = [labels[idx] for idx in label indices]
    elif type == 'dl':
_, processed_sentences_sequences, _, _, _, _ = preprocess_data_for_deep_learning(data=(train_set,
processed sentences, None, None))
        predictions = model.predict(processed sentences sequences)
        result labels = [labels[np.argmax(pred)] for pred in
predictions]
    return result labels
def load and predict(pred sentence, model name, tokenizer=None):
    type = 'ml' if model_name in ['lr', 'svm', 'rf'] else 'dl'
    print(f"Loading model: {model name}")
    if type == 'dl':
        filename = f'model/{model name.lower()}.keras'
        model = tf.keras.models.load model(filename)
    else:
        filename = f'model/{model name.lower()}.sav'
        model = joblib.load(filename)
    return pred user sentence([pred sentence], model=model,
tokenizer=tokenizer, type=type)
text = 'cincai'
model name = 'svm'
sentiment =load and predict(text, model name)
print(sentiment)
Loading model: svm
c:\Users\Asus\anaconda3\Lib\site-packages\sklearn\base.py:318:
UserWarning: Trying to unpickle estimator SVC from version 1.5.0 when
using version 1.2.2. This might lead to breaking code or invalid
results. Use at your own risk. For more info please refer to:
https://scikit-learn.org/stable/model persistence.html#security-
maintainability-limitations
  warnings.warn(
ValueError
                                            Traceback (most recent call
last)
Cell In[8], line 4
      1 text = 'cincai'
      2 model name = 'svm'
----> 4 sentiment =load and predict(text, model name)
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5 print(sentiment)
Cell In[7], line 60, in load and predict(pred sentence, model name,
tokenizer)
     58
            filename = f'model/{model name.lower()}.sav'
     59
            model = joblib.load(filename)
---> 60 return pred_user_sentence([pred_sentence], model=model,
tokenizer=tokenizer, type=type)
Cell In[7], line 43, in pred user sentence(pred sentences, model,
tokenizer, type)
     41 elif type == 'ml':
            _, transformed_sentences = feature extraction(train set,
processed sentences)
            label indices = model.predict(transformed sentences)
---> 43
            result labels = [labels[idx] for idx in label indices]
     45 elif type == 'dl':
File c:\Users\Asus\anaconda3\Lib\site-packages\sklearn\svm\
base.py:820, in BaseSVC.predict(self, X)
            y = np.argmax(self.decision function(X), axis=1)
    818
    819 else:
            y = super().predict(X)
--> 820
    821 return self.classes .take(np.asarray(y, dtype=np.intp))
File c:\Users\Asus\anaconda3\Lib\site-packages\sklearn\svm\
base.py:433, in BaseLibSVM.predict(self, X)
    417 def predict(self, X):
            """Perform regression on samples in X.
    418
    419
    420
            For an one-class model, +1 (inlier) or -1 (outlier) is
returned.
   (\ldots)
    431
                The predicted values.
            .....
    432
            X = self._validate_for predict(X)
--> 433
    434
            predict = self. sparse predict if self. sparse else
self. dense predict
    435
            return predict(X)
File c:\Users\Asus\anaconda3\Lib\site-packages\sklearn\svm\
base.py:613, in BaseLibSVM. validate for predict(self, X)
    610 check is fitted(self)
    612 if not callable(self.kernel):
--> 613
            X = self. validate data(
    614
                Χ,
    615
                accept sparse="csr",
    616
                dtype=np.float64,
    617
                order="C",
    618
                accept large sparse=False,
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619
                reset=False,
    620
            )
    622 if self._sparse and not sp.isspmatrix(X):
        X = sp.csr matrix(X)
File c:\Users\Asus\anaconda3\Lib\site-packages\sklearn\base.py:588, in
BaseEstimator._validate_data(self, X, y, reset, validate_separately,
**check params)
    585
            out = X, y
    587 if not no val X and check params.get("ensure 2d", True):
--> 588
            self. check n features(X, reset=reset)
    590 return out
File c:\Users\Asus\anaconda3\Lib\site-packages\sklearn\base.py:389, in
BaseEstimator. check n features(self, X, reset)
    386
            return
    388 if n features != self.n features in :
--> 389
            raise ValueError(
                f"X has {n_features} features, but
    390
\{\text{self.}\_\text{class}\_.\_\text{name}\ \} "
    391
                f"is expecting {self.n features in } features as
input."
    392 )
ValueError: X has 10363 features, but SVC is expecting 10201 features
as input.
```