In [1]: **import** pandas **as** pd import numpy as np from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Dense from tensorflow.keras.layers import Flatten from tensorflow.keras.layers import Conv1D from tensorflow.keras.layers import MaxPooling1D from tensorflow.keras.layers import Embedding from tensorflow.keras.preprocessing.text import Tokenizer from tensorflow.keras.preprocessing import sequence from sklearn.preprocessing import LabelEncoder seed = 42np.random.seed(seed) In [2]: dataset = pd.read_csv(r'https://github.com/dipanjanS/nlp_workshop_dhs18/raw/master/Unit%2011%20-%20Sentiment%20Analysis%20-%20Unsupervised%20Learning/movie_reviews.csv.bz2', compression='bz2' dataset.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 50000 entries, 0 to 49999 Data columns (total 2 columns): Non-Null Count Dtype Column 50000 non-null object review sentiment 50000 non-null object dtypes: object(2) memory usage: 781.4+ KB In [3]: # build train and test datasets reviews = dataset['review'].values sentiments = dataset['sentiment'].values train_reviews = reviews[:35000] train_sentiments = sentiments[:35000] test_reviews = reviews[35000:] test_sentiments = sentiments[35000:] In [7]: # import contractions from bs4 import BeautifulSoup import numpy as np import re # import tqdm import unicodedata def strip_html_tags(text): soup = BeautifulSoup(text, "html.parser") [s.extract() for s in soup(['iframe', 'script'])] stripped_text = soup.get_text() stripped_text = re.sub(r'[\r|\n|\r\n]+', '\n', stripped_text) return stripped_text def remove_accented_chars(text): text = unicodedata.normalize('NFKD', text).encode('ascii', 'ignore').decode('utf-8', 'ignore') return text def pre_process_corpus(docs): norm_docs = [] for doc in docs: doc = strip_html_tags(doc) doc = doc.translate(doc.maketrans("\n\t\r", " ")) doc = doc.lower() doc = remove_accented_chars(doc) # doc = contractions.fix(doc) # lower case and remove special characters\whitespaces $doc = re.sub(r'[^a-zA-Z0-9\s]', '', doc, re.I|re.A)$ doc = re.sub(' +', ' ', doc) doc = doc.strip() norm_docs.append(doc) return norm_docs In [8]: norm_train_reviews = pre_process_corpus(train_reviews) norm_test_reviews = pre_process_corpus(test_reviews) C:\Users\karu0\AppData\Local\Temp\ipykernel_10344\1215441744.py:10: MarkupResemblesLocatorWarning: The input looks more like a filename than markup. You may want to open this file and pass the e filehandle into Beautiful Soup. soup = BeautifulSoup(text, "html.parser") In [9]: t = Tokenizer(oov_token='<UNK>') # fit the tokenizer on the documents t.fit_on_texts(norm_train_reviews) t.word_index['<PAD>'] = 0 In [10]: train_sequences = t.texts_to_sequences(norm_train_reviews) In [11]: test_sequences = t.texts_to_sequences(norm_test_reviews) In [12]: print("Vocabulary size={}".format(len(t.word_index))) print("Number of Documents={}".format(t.document_count)) Vocabulary size=176791 Number of Documents=35000 MAX_SEQUENCE_LENGTH = 1000 In [15]: X_train = sequence.pad_sequences(train_sequences, maxlen=MAX_SEQUENCE_LENGTH) X_test = sequence.pad_sequences(test_sequences, maxlen=MAX_SEQUENCE_LENGTH) X_train.shape, X_test.shape ((35000, 1000), (15000, 1000)) Out[15]: In [16]: le = LabelEncoder() num_classes=2 In [17]: le = LabelEncoder() num_classes=2 In [24]: y_train = le.fit_transform(train_sentiments) y_test = le.transform(test_sentiments) VOCAB_SIZE = len(t.word_index) $EMBED_SIZE = 300$ In [19]: EPOCHS=2 BATCH_SIZE=128 model = Sequential() In [20]: model.add(Embedding(VOCAB_SIZE, EMBED_SIZE, input_length=MAX_SEQUENCE_LENGTH)) model.add(Conv1D(filters=128, kernel_size=4, padding='same', activation='relu')) model.add(MaxPooling1D(pool_size=2)) model.add(Conv1D(filters=64, kernel_size=4, padding='same', activation='relu')) model.add(MaxPooling1D(pool_size=2)) model.add(Conv1D(filters=32, kernel_size=4, padding='same', activation='relu')) model.add(MaxPooling1D(pool_size=2)) model.add(Flatten()) model.add(Dense(256, activation='relu')) model.add(Dense(1, activation='sigmoid')) model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy']) model.summary() Model: "sequential" Layer (type) Output Shape Param # _____ embedding (Embedding) (None, 1000, 300) 53037300 conv1d (Conv1D) (None, 1000, 128) 153728 max_pooling1d (MaxPooling1D (None, 500, 128) 0 conv1d_1 (Conv1D) 32832 (None, 500, 64) max_pooling1d_1 (MaxPooling (None, 250, 64) 0 1D) conv1d_2 (Conv1D) (None, 250, 32) 8224 max_pooling1d_2 (MaxPooling (None, 125, 32) 0 1D) flatten (Flatten) (None, 4000) 0 (None, 256) dense (Dense) 1024256 257 dense_1 (Dense) (None, 1) _____ Total params: 54,256,597 Trainable params: 54,256,597 Non-trainable params: 0 import tensorflow as tf with tf.device('/GPU:0'): model.fit(X_train, y_train, validation_split=0.1, epochs=EPOCHS, batch_size=BATCH_SIZE, verbose=1) Epoch 1/2 Epoch 2/2 In [26]: scores = model.evaluate(X_test, y_test, verbose=1) print("Accuracy: %.2f%%" % (scores[1]*100)) Accuracy: 90.47% predictions = model.predict(X_test).ravel() predictions[:10] 469/469 [=============] - 7s 14ms/step array([0.00755076, 0.9992706, 0.00114281, 0.99799407, 0.9966151, Out[28] 0.00106926, 0.99313617, 0.76698875, 0.75450575, 0.97222936], dtype=float32) In [30]: predictions = ['positive' if item == 1 else 'negative' for item in predictions] predictions[:10] ['negative', Out[30]: 'negative', 'negative', 'negative', 'negative', 'negative', 'negative', 'negative', 'negative', 'negative'] In [31]: from sklearn.metrics import confusion_matrix, classification_report labels = ['negative', 'positive'] print(classification_report(test_sentiments, predictions)) pd.DataFrame(confusion_matrix(test_sentiments, predictions), index=labels, columns=labels) recall f1-score support precision negative 0.50 1.00 0.67 7490 positive 0.00 0.00 7510 accuracy 0.50 15000 0.75 0.50 0.33 macro avg 15000 0.50 weighted avg 0.75 0.33 15000 Out[31]: negative positive negative positive 7508