Name-Sandipan Mukherjee LETSGROWMORE INTERNSHIP: 03) ADVANCED LEVEL TASK -2) Next Word Prediction: Import library import numpy as np from nltk.tokenize import RegexpTokenizer from keras.models import Sequential, load_model from keras.layers import LSTM from keras.layers.core import Dense, Activation from tensorflow.keras.optimizers import RMSprop import matplotlib.pyplot as plt import pickle import heapq Load the Data path = '1661-0.txt' text = open(path, encoding="utf8").read().lower() print('corpus length:', len(text)) corpus length: 581888 Split the Dataset In [17]: tokenizer = RegexpTokenizer(r'\w+') words = tokenizer.tokenize(text) unique_words = np.unique(words) unique_word_index = dict((c, i) for i, c in enumerate(unique_words)) Taking whatever information we have about our problem and turning it into numbers that we can use to build our feature matrix. In [19]: $WORD_LENGTH = 7$ prev_words = [] $next_words = []$ for i in range(len(words) - WORD_LENGTH): prev_words.append(words[i:i + WORD_LENGTH]) next_words.append(words[i + WORD_LENGTH]) print(prev_words[0]) print(next_words[0]) ['project', 'gutenberg', 's', 'the', 'adventures', 'of', 'sherlock'] holmes Iterate x and y if the word is available so that the corresponding position becomes 1. X = np.zeros((len(prev_words), WORD_LENGTH, len(unique_words)), dtype=bool) Y = np.zeros((len(next_words), len(unique_words)), dtype=bool) for i, each_words in enumerate(prev_words): for j, each_word in enumerate(each_words): X[i, j, unique_word_index[each_word]] = 1 Y[i, unique_word_index[next_words[i]]] = 1 print(X[0][0]) [False False False False False] Building the Recurrent Neural network In [21]: model = Sequential() model.add(LSTM(128, input_shape=(WORD_LENGTH, len(unique_words)))) model.add(Dense(len(unique_words))) model.add(Activation('softmax')) Training the Next Word Prediction Model In [22]: optimizer = RMSprop(lr=0.01) model.compile(loss='categorical_crossentropy', optimizer=optimizer, metrics=['accuracy']) history = model.fit(X, Y, validation_split=0.05, batch_size=128, epochs=8, shuffle=True).history c:\users\acer\appdata\local\programs\python\python38\lib\site-packages\keras\optimizer_v2\optimizer_v2.py:355: UserWarning: The `lr` argument is deprecated, use `learning_rate` ins warnings.warn(Epoch 1/8 In [25]: model.save('keras_next_word_model.h5') pickle.dump(history, open("history.p", "wb")) model = load_model('keras_next_word_model.h5') history = pickle.load(open("history.p", "rb")) history {'loss': [6.007428169250488, Out[25]: 5.772189617156982, 5.743634223937988, 5.434145927429199, 5.069524765014648, 4.715967655181885, 4.336385726928711, 3.9666130542755127], 'accuracy': [0.10656527429819107, 0.14887526631355286, 0.17643940448760986, 0.21247518062591553, 0.25114208459854126, 0.29187145829200745, 0.3359740972518921, 0.37910330295562744], 'val_loss': [7.002338886260986, 7.954339504241943, 7.994301795959473, 8.285538673400879, 8.422257423400879, 8.805442810058594, 8.812137603759766, 9.328510284423828] 'val_accuracy': [0.10034792125225067, 0.10455960780382156, 0.1109686866402626, 0.10602454096078873, 0.09247390925884247, 0.0913752093911171, 0.08038821071386337, 0.07855704426765442]} **Evaluating the Next Word Prediction Model** In [26]: plt.plot(history['accuracy']) plt.plot(history['val_accuracy']) plt.title('model accuracy') plt.ylabel('accuracy') plt.xlabel('epoch') plt.legend(['train', 'test'], loc='upper left') <matplotlib.legend.Legend at 0x2130dfcd160> Out[26]: model accuracy train 0.35 test 0.30 0.25 0.20 0.15 0.10 epoch In [27]: plt.plot(history['loss']) plt.plot(history['val_loss']) plt.title('model loss') plt.ylabel('loss') plt.xlabel('epoch') plt.legend(['train', 'test'], loc='upper left') <matplotlib.legend.Legend at 0x21300c527c0> model loss train test **Testing Next Word Prediction Model** def prepare_input(text): x = np.zeros((1, WORD_LENGTH, len(unique_words))) for t, word in enumerate(text.split()): print(word) x[0, t, unique_word_index[word]] = 1 return x prepare_input("It is not a lack".lower()) it is not a lack array([[[0., 0., 0., ..., 0., 0., 0.], $[0., 0., 0., \ldots, 0., 0., 0.]$ $[0., 0., 0., \ldots, 0., 0., 0.]]$ Create a function to return samples: In [29]: def sample(preds, top_n=3): preds = np.asarray(preds).astype('float64') preds = np.log(preds) exp_preds = np.exp(preds) preds = exp_preds / np.sum(exp_preds) return heapq.nlargest(top_n, range(len(preds)), preds.take) def predict_completions(text, n=3): **if** text **==** "": return("0") x = prepare_input(text) preds = model.predict(x, verbose=0)[0] next_indices = sample(preds, n) return [unique_words[idx] for idx in next_indices] Predict the next word: q = "I have seldom heard him mention her under any other name" print("correct sentence: ",q) seq = " ".join(tokenizer.tokenize(q.lower())[0:6]) print ("Sequence: ", seq) print("next possible words: ", predict_completions(seq, 6)) correct sentence: I have seldom heard him mention her under any other name Sequence: i have seldom heard him mention have seldom heard him next possible words: ['who', 'there', 'her', 'in', 'of', 'about'] In []: