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DEPARTMENT OF ELECTRONICS AND COMMUNICATION ENGINEERING

NAAN MUDHALVAN-IBM(AI) PROJECT

IBM AL 101 ARTIFICIAL INTELLIGENCE-GROUP 1(TEAM 5)

PROJECT TITLE:

CREATE A CHATBOT USING PYTHON

SUBMITTED BY:

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Phase 4: DEVELOPMENT PART 2:

1.What is a Chatbot?

A chatbot is an AI-based software designed to interact with humans in their natural languages. These chatbots are usually converse via auditory or textual methods, and they can effortlessly mimic human languages to communicate with human beings in a human-like manner. A chatbot is arguably one of the best applications of natural language processing.

2. How to Make a Chatbot in Python?

- In the past few years, chatbots in Python have become wildly popular in the tech and business sectors. These intelligent bots are so adept at imitating natural human languages and conversing with humans, that companies across various industrial sectors are adopting them. From e-commerce firms to healthcare institutions, everyone seems to be leveraging this nifty tool to drive business benefits.
- To build a chatbot in Python, import all the necessary packages and initialize the variables you want to use in chatbot project. Also, when working with text data, we need to perform data preprocessing on your dataset before designing an ML model.
- This is where tokenizing helps with text data it helps fragment the large text dataset into smaller, readable chunks (like words). Once that is done, you can also go for lemmatization that transforms a word into its lemma form. Then it creates a pickle file to store the python objects that are used for predicting the responses of the bot.
- Another vital part of the chatbot development process is creating the training and testing datasets.

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I. Import Libraries:

This code snippet imports TensorFlow, NumPy, Pandas, Matplotlib, Seaborn, and various components from TensorFlow's Keras module. It also imports the re and string modules for regular expressions and string manipulation. The code prepares your environment for working with deep learning and natural language processing.

Input 1-2:

import tensorflow as tf

import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

from tensorflow.keras.layers import TextVectorization

import re, string

 $from\ tensor flow. keras. layers\ import\ LSTM, Dense, Embedding, Dropout, Layer Normalization$

II. Data Preprocessing:

Data Visualization:

This code calculates the number of tokens (words) in the 'question' and 'answer' columns of a Pandas DataFrame and then visualizes the token distribution using Matplotlib and Seaborn. The resulting plots are displayed in a single figure with two subplots for token distributions and a joint distribution between 'question' and 'answer' tokens.

Input 3:

cmap='YIGnBu')

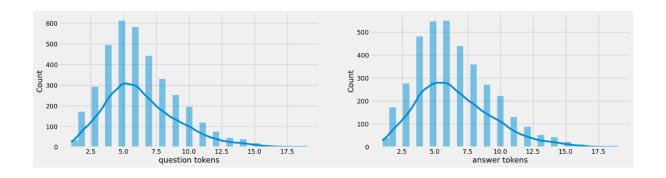
plt.show()

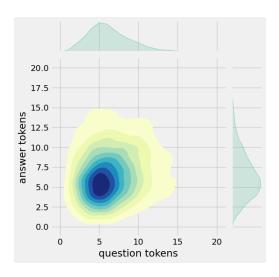
```
df['question tokens'] = df['question'].apply(lambda x: len(x.split()))

df['answer tokens'] = df['answer'].apply(lambda x: len(x.split()))

import matplotlib.pyplot as plt
import seaborn as sns

plt.style.use('fivethirtyeight')
fig, ax = plt.subplots(nrows=1, ncols=2, figsize=(20, 5))
sns.set_palette('Set2')
sns.histplot(x=df['question tokens'], data=df, kde=True, ax=ax[0])
sns.histplot(x=df['answer tokens'], data=df, kde=True, ax=ax[1])
sns.jointplot(x='question tokens', y='answer tokens', data=df, kind='kde', fill=True,
```





> Text Cleaning:

This code defines a clean_text function to clean the text and then applies this function to the 'question' and 'answer' columns in the DataFrame. It also modifies the DataFrame by creating 'encoder_inputs', 'decoder_targets', and 'decoder_inputs' columns.

Input 4:

```
def clean_text(text):
    text = re.sub('-', ' ', text.lower())
    text = re.sub('[.]', ' . ', text)
    text = re.sub('[1]', ' 1 ', text)
    text = re.sub('[2]', ' 2 ', text)
    text = re.sub('[3]', ' 3 ', text)
    text = re.sub('[4]', ' 4 ', text)
    text = re.sub('[5]', ' 5 ', text)
    text = re.sub('[6]', ' 6 ', text)
    text = re.sub('[7]', ' 7 ', text)
```

```
text = re.sub('[8]', '8', text)
  text = re.sub('[9]', '9', text)
  text = re.sub('[0]', '0', text)
  text = re.sub(',', ' , ', text)
  text = re.sub('?', ' ? ', text)
  text = re.sub('!', '!', text)
  text = re.sub('$', '$', text)
  text = re.sub('&', ' & ', text)
  text = re.sub('/', ' / ', text)
  text = re.sub(':', ':', text)
  text = re.sub(';', '; ', text)
  text = re.sub('*', ' * ', text)
  text = re.sub("", " ' ", text)
  text = re.sub("", ' " ', text)
  text = re.sub('\t', ' ', text)
  return text
df.drop(columns=['answer tokens', 'question tokens'], axis=1, inplace=True)
df['encoder_inputs'] = df['question'].apply(clean_text)
df['decoder_targets'] = df['answer'].apply(clean_text) + ' <end>'
df['decoder inputs'] = '<start> ' + df['answer'].apply(clean text) + ' <end>'
```

df.head(10)

0	hi, how are you doing?	i'm fine. how about yourself?	hi , how are you doing ?	i 'm fine . how about yourself ? <end></end>	<pre><start> i ' m fine . how about yourself ? <end></end></start></pre>

1	i'm fine. how about yourself?	i'm pretty good. thanks for asking.	i ' m fine . how about yourself ?	i 'm pretty good . thanks for asking . <end></end>	<start> i ' m pretty good . thanks for asking</start>
2 i'm pretty good. thanks for asking.		no problem. so how have you been?	i ' m pretty good . thanks for asking .	no problem . so how have you been ? <end></end>	<start> no problem . so how have you been ?</start>
3	no problem. so how have you been?	i've been great. what about you?	no problem . so how have you been ?	i ' ve been great . what about you ? <end></end>	<start> i ' ve been great . what about you ?</start>
4	i've been great. what about you?	i've been good. i'm in school right now.	i ' ve been great . what about you ?	i've been good .i'm in school right now	<start> i ' ve been good . i ' m in school ri</start>
5	i've been good. i'm in school right now.	what school do you go to?	i ' ve been good . i ' m in school right now .	what school do you go to ? <end></end>	<start> what school do you go to ? <end></end></start>
6	what school do you go to?	i go to pcc.	what school do you go to ?	i go to pcc . <end></end>	<start> i go to pcc . <end></end></start>
7	i go to pcc.	do you like it there?	i go to pcc.	do you like it there ? <pre><end></end></pre>	<start> do you like it there ? <end></end></start>
8	do you like it there?	it's okay. it's a really big campus.	do you like it there ?	it's okay . it's a really big campus . <	<start> it ' s okay . it ' s a really big cam</start>
9	it's okay. it's a really big campus.	good luck with school.	it's okay . it's a really big campus .	good luck with school . <end></end>	<start> good luck with school . <end></end></start>

Input 5:

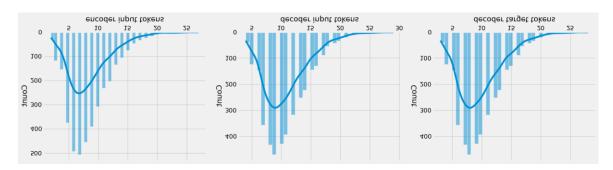
```
df['encoder input tokens'] = df['encoder_inputs'].apply(lambda x: len(x.split()))
df['decoder input tokens'] = df['decoder_inputs'].apply(lambda x: len(x.split()))
df['decoder target tokens'] = df['decoder_targets'].apply(lambda x: len(x.split()))
```

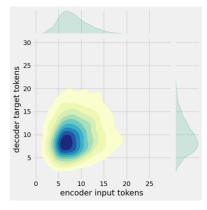
import matplotlib.pyplot as plt import seaborn as sns

```
plt.style.use('fivethirtyeight')
fig, ax = plt.subplots(nrows=1, ncols=3, figsize=(20, 5))
```

```
sns.set_palette('Set2')
sns.histplot(x=df['encoder input tokens'], data=df, kde=True, ax=ax[0])
sns.histplot(x=df['decoder input tokens'], data=df, kde=True, ax=ax[1])
sns.histplot(x=df['decoder target tokens'], data=df, kde=True, ax=ax[2])
sns.jointplot(x='encoder input tokens', y='decoder target tokens', data=df, kind='kde', fill=True, cmap='YlGnBu')
plt.show()
```

This code calculates the token counts for 'encoder_inputs', 'decoder_inputs', and 'decoder_targets' columns in the DataFrame and then visualizes the token distribution using Matplotlib and Seaborn. The resulting plots are displayed in a single figure with three subplots for the token counts and a joint distribution between 'encoder input tokens' and 'decoder target tokens'.





Input 6:

```
print(f"After preprocessing: {' '.join(df[df['encoder input tokens'].max()==df['encoder input
tokens']]['encoder inputs'].values.tolist())}")
print(f"Max encoder input length: {df['encoder input tokens'].max()}")
print(f"Max decoder input length: {df['decoder input tokens'].max()}")
print(f"Max decoder target length: {df['decoder target tokens'].max()}")
df.drop(columns=['question','answer','encoder input tokens','decoder input
tokens','decoder target tokens'],axis=1,inplace=True)
params={
  "vocab_size":2500,
  "max sequence length":30,
  "learning rate":0.008,
  "batch size":149,
  "Istm cells":256,
  "embedding dim":256,
  "buffer size":10000
learning rate=params['learning rate']
batch_size=params['batch_size']
embedding dim=params['embedding dim']
lstm cells=params['lstm_cells']
vocab size=params['vocab size']
buffer_size=params['buffer_size']
max sequence length=params['max sequence length']
df.head(10)
```

Output:

encoder_inputs	decoder_targets	decoder_inputs	
0	hi , how are you doing ?	i ' m fine . how about yourself ? <end></end>	<start> i ' m fine . how about yourself ? <end></end></start>
1	i ' m fine . how about yourself ?	i ' m pretty good . thanks for asking . <end></end>	<start> i ' m pretty good . thanks for asking</start>
2	i ' m pretty good . thanks for asking .	no problem . so how have you been ? <end></end>	<start> no problem . so how have you been ?</start>
3	no problem . so how have you been ?	i ' ve been great . what about you ? <end></end>	<start> i ' ve been great . what about you ?</start>
4	i ' ve been great . what about you ?	i ' ve been good . i ' m in school right now	<start> i ' ve been good . i ' m in school ri</start>
5	i ' ve been good . i ' m in school right now .	what school do you go to ? <end></end>	<start> what school do you go to ? <end></end></start>
6	what school do you go to?	i go to pcc . <end></end>	<start> i go to pcc . <end></end></start>
7	i go to pcc.	do you like it there ? <end></end>	<start> do you like it there ? <end></end></start>
8	do you like it there ?	it's okay . it's a really big campus . <	<start> it 's okay . it 's a really big cam</start>
9	it's okay . it's a really big campus .	good luck with school . <end></end>	<start> good luck with school . <end></end></start>

> Tokenization:

This code snippet involves data preprocessing, including text vectorization using TensorFlow's TextVectorization layer, conversion between sequences and IDs, and the creation of training and validation datasets using TensorFlow's Dataset API. It also prints various details about the data, such as batch sizes and shapes.

Input 7:

```
vectorize_layer=TextVectorization(
   max_tokens=vocab_size,
```

```
standardize=None,
  output_mode='int',
  output_sequence_length=max_sequence_length
)
vectorize_layer.adapt(df['encoder_inputs']+' '+df['decoder_targets']+' <start> <end>')
vocab_size=len(vectorize_layer.get_vocabulary())
print(f'Vocab size: {len(vectorize_layer.get_vocabulary())}')
print(f'{vectorize_layer.get_vocabulary()[:12]}')
Vocab size: 2443
[", '[UNK]', '<end>', '.', '<start>', """, 'i', '?', 'you', ',', 'the', 'to']
Input 8:
def sequences2ids(sequence):
  return vectorize_layer(sequence)
def ids2sequences(ids):
  decode="
  if type(ids)==int:
    ids=[ids]
  for id in ids:
    decode+=vectorize_layer.get_vocabulary()[id]+''
  return decode
x=sequences2ids(df['encoder_inputs'])
yd=sequences2ids(df['decoder_inputs'])
```

```
y=sequences2ids(df['decoder_targets'])
print(f'Question sentence: hi , how are you ?')
print(f'Question to tokens: {sequences2ids("hi , how are you ?")[:10]}')
print(f'Encoder input shape: {x.shape}')
print(f'Decoder input shape: {yd.shape}')
print(f'Decoder target shape: {y.shape}')
Question sentence: hi , how are you?
Question to tokens: [1971 9 45 24 8 7 0 0 0 0]
Encoder input shape: (3725, 30)
Decoder input shape: (3725, 30)
Decoder target shape: (3725, 30)
Input 9:
print(f'Encoder input: {x[0][:12]} ...')
print(f'Decoder input: {yd[0][:12]}...') # shifted by one time step of the target as input to
decoder is the output of the previous timestep
print(f'Decoder target: {y[0][:12]} ...')
Encoder input: [1971 9 45 24 8 194 7 0 0 0 0] ...
Decoder input: [ 4 6 5 38 646 3 45 41 563 7 2 0] ...
Decoder target: [ 6 5 38 646 3 45 41 563 7 2 0 0] ...
Input 10:
data=tf.data.Dataset.from_tensor_slices((x,yd,y))
```

```
data=data.shuffle(buffer size)
train data=data.take(int(.9*len(data)))
train data=train data.cache()
train data=train data.shuffle(buffer size)
train data=train data.batch(batch size)
train data=train data.prefetch(tf.data.AUTOTUNE)
train data iterator=train data.as numpy iterator()
val data=data.skip(int(.9*len(data))).take(int(.1*len(data)))
val data=val data.batch(batch size)
val data=val data.prefetch(tf.data.AUTOTUNE)
=train data iterator.next()
print(f'Number of train batches: {len(train_data)}')
print(f'Number of training data: {len(train_data)*batch_size}')
print(f'Number of validation batches: {len(val data)}')
print(f'Number of validation data: {len(val_data)*batch_size}')
print(f'Encoder Input shape (with batches): {_[0].shape}')
print(f'Decoder Input shape (with batches): { [1].shape}')
print(f'Target Output shape (with batches): { [2].shape}')
Number of train batches: 23
Number of training data: 3427
Number of validation batches: 3
Number of validation data: 447
```

Encoder Input shape (with batches): (149, 30)

Decoder Input shape (with batches): (149, 30)

Target Output shape (with batches): (149, 30)

III. Build Models:

Build Encoder:

This code defines classes for the encoder and decoder in a sequence-to-sequence model. The encoder processes input sequences, and the decoder generates output sequences. The provided code includes details about the layers, embeddings, and initializations used in both the encoder and decoder components. It also demonstrates the usage of these components by making a forward pass with example data.

Input 11:

```
class Encoder(tf.keras.models.Model):
    def __init__(self,units,embedding_dim,vocab_size,*args,**kwargs) -> None:
        super().__init__(*args,**kwargs)
        self.units=units
        self.vocab_size=vocab_size
        self.embedding_dim=embedding_dim
        self.embedding=Embedding(
            vocab_size,
            embedding_dim,
            name='encoder_embedding',
            mask_zero=True,
            embeddings_initializer=tf.keras.initializers.GlorotNormal()
        )
        self.normalize=LayerNormalization()
```

```
self.lstm=LSTM(
      units,
      dropout=.4,
      return state=True,
      return sequences=True,
      name='encoder Istm',
      kernel initializer=tf.keras.initializers.GlorotNormal()
    )
  def call(self,encoder_inputs):
    self.inputs=encoder inputs
    x=self.embedding(encoder_inputs)
    x=self.normalize(x)
    x=Dropout(.4)(x)
    encoder_outputs,encoder_state_h,encoder_state_c=self.lstm(x)
    self.outputs=[encoder_state_h,encoder_state_c]
    return encoder_state_h,encoder_state_c
encoder=Encoder(lstm_cells,embedding_dim,vocab_size,name='encoder')
encoder.call(_[0])
OUTPUT:
(<tf.Tensor: shape=(149, 256), dtype=float32, numpy=
array([[ 0.16966951, -0.10419625, -0.12700348, ..., -0.12251794,
    0.10568858, 0.14841646],
   [0.08443093, 0.08849293, -0.09065959, ..., -0.00959182,
```

```
0.10152507, -0.12077457],
   [ 0.03628462, -0.02653611, -0.11506603, ..., -0.14669597,
    0.10292757, 0.13625325],
   [-0.14210635, -0.12942064, -0.03288083, ..., 0.0568463,
    -0.02598592, -0.22455114],
   [0.20819993, 0.01196991, -0.09635217, ..., -0.18782297,
    0.10233591, 0.20114912],
   [0.1164271, -0.07769038, -0.06414707, ..., -0.06539135,
    -0.05518465, 0.25142196]], dtype=float32)>,
<tf.Tensor: shape=(149, 256), dtype=float32, numpy=
array([[ 0.34589 , -0.30134732, -0.43572 , ..., -0.3102559,
    0.34630865, 0.2613009],
   [0.14154069, 0.17045322, -0.17749965, ..., -0.02712595,
    0.17292541, -0.2922624],
   [0.07106856, -0.0739173, -0.3641197, ..., -0.3794833,
    0.36470377, 0.23766585],
   [-0.2582597, -0.25323495, -0.06649272, ..., 0.16527973,
   -0.04292646, -0.58768904],
   [0.43155715, 0.03135502, -0.33463806, ..., -0.47625306,
    0.33486888, 0.35035062],
   [0.23173636, -0.20141824, -0.22034441, ..., -0.16035017,
    -0.17478186, 0.48899865]], dtype=float32)>)
```

Build Encoder## Build Decoder

Input 12:

class Decoder(tf.keras.models.Model):

```
def __init__(self,units,embedding_dim,vocab_size,*args,**kwargs) -> None:
 super(). init (*args,**kwargs)
  self.units=units
  self.embedding dim=embedding dim
  self.vocab size=vocab size
  self.embedding=Embedding(
    vocab size,
    embedding dim,
    name='decoder_embedding',
    mask_zero=True,
    embeddings initializer=tf.keras.initializers.HeNormal()
  self.normalize=LayerNormalization()
  self.lstm=LSTM(
    units,
    dropout=.4,
    return_state=True,
    return_sequences=True,
    name='decoder_lstm',
    kernel_initializer=tf.keras.initializers.HeNormal()
  self.fc=Dense(
    vocab_size,
    activation='softmax',
    name='decoder_dense',
    kernel_initializer=tf.keras.initializers.HeNormal()
```

```
)
  def call(self,decoder inputs,encoder states):
    x=self.embedding(decoder inputs)
    x=self.normalize(x)
    x=Dropout(.4)(x)
    x,decoder state h,decoder state c=self.lstm(x,initial state=encoder states)
    x=self.normalize(x)
    x=Dropout(.4)(x)
    return self.fc(x)
decoder=Decoder(Istm cells,embedding dim,vocab size,name='decoder')
decoder( [1][:1],encoder( [0][:1]))
OUTPUT:
<tf.Tensor: shape=(1, 30, 2443), dtype=float32, numpy=
array([[[3.4059247e-04, 5.7348556e-05, 2.1294907e-05, ...,
    7.2067953e-05, 1.5453645e-03, 2.3599296e-04],
    [1.4662130e-03, 8.0250365e-06, 5.4062020e-05, ...,
    1.9187471e-05, 9.7244098e-05, 7.6433855e-05],
    [9.6929165e-05, 2.7441782e-05, 1.3761305e-03, ...,
    3.6009602e-05, 1.5537882e-04, 1.8397317e-04],
    ...,
    [1.9002777e-03, 6.9266016e-04, 1.4346189e-04, ...,
    1.9552530e-04, 1.7106640e-05, 1.0252406e-04],
    [1.9002777e-03, 6.9266016e-04, 1.4346189e-04, ...,
```

```
1.9552530e-04, 1.7106640e-05, 1.0252406e-04],

[1.9002777e-03, 6.9266016e-04, 1.4346189e-04, ...,

1.9552530e-04, 1.7106640e-05, 1.0252406e-04]]], dtype=float32)>
```

Build Training Model:

This code defines a ChatBotTrainer class for training and testing a chatbot model. It includes custom loss and accuracy functions, training and testing steps, and the compilation of the model. The code then performs a forward pass with the model using example data.

INPUT-13

```
class ChatBotTrainer(tf.keras.models.Model):
  def init (self,encoder,decoder,*args,**kwargs):
    super(). init (*args,**kwargs)
    self.encoder=encoder
    self.decoder=decoder
  def loss_fn(self,y_true,y_pred):
    loss=self.loss(y_true,y_pred)
    mask=tf.math.logical not(tf.math.equal(y true,0))
    mask=tf.cast(mask,dtype=loss.dtype)
    loss*=mask
    return tf.reduce mean(loss)
  def accuracy_fn(self,y_true,y_pred):
    pred_values = tf.cast(tf.argmax(y_pred, axis=-1), dtype='int64')
    correct = tf.cast(tf.equal(y true, pred values), dtype='float64')
    mask = tf.cast(tf.greater(y_true, 0), dtype='float64')
```

```
n_correct = tf.keras.backend.sum(mask * correct)
  n total = tf.keras.backend.sum(mask)
  return n correct / n total
def call(self,inputs):
  encoder inputs, decoder inputs=inputs
  encoder states=self.encoder(encoder inputs)
  return self.decoder(decoder inputs,encoder states)
def train_step(self,batch):
  encoder inputs, decoder inputs, y=batch
  with tf.GradientTape() as tape:
    encoder states=self.encoder(encoder inputs,training=True)
    y pred=self.decoder(decoder inputs,encoder states,training=True)
    loss=self.loss_fn(y,y_pred)
    acc=self.accuracy_fn(y,y_pred)
  variables=self.encoder.trainable_variables+self.decoder.trainable_variables
  grads=tape.gradient(loss,variables)
  self.optimizer.apply_gradients(zip(grads,variables))
  metrics={'loss':loss,'accuracy':acc}
  return metrics
def test_step(self,batch):
  encoder_inputs,decoder_inputs,y=batch
  encoder_states=self.encoder(encoder_inputs,training=True)
```

```
y_pred=self.decoder(decoder_inputs,encoder_states,training=True)
loss=self.loss_fn(y,y_pred)
acc=self.accuracy_fn(y,y_pred)
metrics={'loss':loss,'accuracy':acc}
return metrics
INPUT-14
model=ChatBotTrainer(encoder,decoder,name='chatbot_trainer')
model.compile(
loss=tf.keras.losses.SparseCategoricalCrossentropy(),
optimizer=tf.keras.optimizers.Adam(learning_rate=learning_rate),
weighted_metrics=['loss','accuracy']
)
model(_[:2])
```

> Train Model:

In this code, the model.fit function is used to train the model for 100 epochs with training data (train_data) and validation data (val_data). Two callbacks are specified: the TensorBoard callback for monitoring the training process and the ModelCheckpoint callback to save the best model during training. The training history is stored in the history variable.

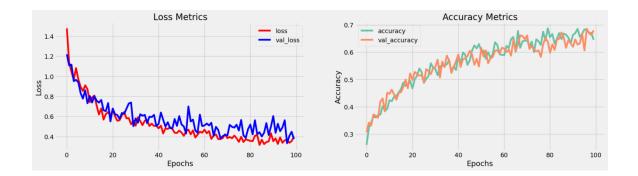
```
Input- 15
history=model.fit(
train_data,
epochs=100,
validation_data=val_data,
```

```
callbacks=[
    tf.keras.callbacks.TensorBoard(log_dir='logs'),
    tf.keras.callbacks.ModelCheckpoint('ckpt',verbose=1,save_best_only=True)
]
```

IV.Visualize Metrics:

This code creates a figure with two subplots to visualize training and validation loss and accuracy metrics over training epochs. It uses Matplotlib for plotting and shows the resulting figure.

```
Input-16
fig,ax=plt.subplots(nrows=1,ncols=2,figsize=(20,5))
ax[0].plot(history.history['loss'],label='loss',c='red')
ax[0].plot(history.history['val_loss'],label='val_loss',c = 'blue')
ax[0].set_xlabel('Epochs')
ax[1].set_xlabel('Epochs')
ax[0].set_ylabel('Loss')
ax[1].set_ylabel('Accuracy')
ax[0].set_title('Loss Metrics')
ax[1].set_title('Accuracy Metrics')
ax[1].plot(history.history['accuracy'],label='accuracy')
ax[1].plot(history.history['val_accuracy'],label='val_accuracy')
ax[0].legend()
ax[1].legend()
plt.show()
```



V.Save Model:

```
model.load_weights('ckpt')
model.save('models',save_format='tf')
for idx,i in enumerate(model.layers):
    print('Encoder layers:' if idx==0 else 'Decoder layers: ')
    for j in i.layers:
        print(j)
    print('------')
```

VI.Create Inference Model:

```
class ChatBot(tf.keras.models.Model):
    def __init__(self,base_encoder,base_decoder,*args,**kwargs):
        super().__init__(*args,**kwargs)
        self.encoder,self.decoder=self.build inference model(base encoder,base decoder)
```

```
def build_inference_model(self,base_encoder,base_decoder):
      encoder_inputs=tf.keras.Input(shape=(None,))
      x=base encoder.layers[0](encoder inputs)
      x=base encoder.layers[1](x)
      x,encoder state h,encoder state c=base encoder.layers[2](x)
encoder=tf.keras.models.Model(inputs=encoder_inputs,outputs=[encoder_state_h,encoder_state_c],nam
e='chatbot encoder')
      decoder input state h=tf.keras.Input(shape=(lstm cells,))
      decoder_input_state_c=tf.keras.Input(shape=(lstm_cells,))
      decoder inputs=tf.keras.Input(shape=(None,))
      x=base decoder.layers[0](decoder inputs)
      x=base encoder.layers[1](x)
x,decoder_state_h,decoder_state_c=base_decoder.layers[2](x,initial_state=[decoder_input_state_h,decod
er_input_state_c])
      decoder_outputs=base_decoder.layers[-1](x)
      decoder=tf.keras.models.Model(
        inputs=[decoder_inputs,[decoder_input_state_h,decoder_input_state_c]],
        outputs=[decoder_outputs,[decoder_state_h,decoder_state_c]],name='chatbot_decoder'
      )
      return encoder, decoder
    def summary(self):
      self.encoder.summary()
      self.decoder.summary()
    def softmax(self,z):
      return np.exp(z)/sum(np.exp(z))
```

```
def sample(self,conditional_probability,temperature=0.5):
  conditional_probability = np.asarray(conditional_probability).astype("float64")
  conditional probability = np.log(conditional probability) / temperature
  reweighted conditional probability = self.softmax(conditional probability)
  probas = np.random.multinomial(1, reweighted conditional probability, 1)
  return np.argmax(probas)
def preprocess(self,text):
  text=clean text(text)
  seq=np.zeros((1,max_sequence_length),dtype=np.int32)
  for i,word in enumerate(text.split()):
    seq[:,i]=sequences2ids(word).numpy()[0]
  return seq
def postprocess(self,text):
  text=re.sub(' - ','-',text.lower())
  text=re.sub(' [.] ','. ',text)
  text=re.sub(' [1] ','1',text)
  text=re.sub(' [2] ','2',text)
  text=re.sub(' [3] ','3',text)
  text=re.sub(' [4] ','4',text)
  text=re.sub(' [5] ','5',text)
  text=re.sub(' [6] ','6',text)
  text=re.sub(' [7] ','7',text)
  text=re.sub(' [8] ','8',text)
  text=re.sub(' [9] ','9',text)
  text=re.sub(' [0] ','0',text)
```

```
text=re.sub(' [,] ',', ',text)
  text=re.sub(' [?] ','? ',text)
  text=re.sub(' [!] ','! ',text)
  text=re.sub(' [$] ','$ ',text)
  text=re.sub(' [&] ','& ',text)
  text=re.sub(' [/] ','/ ',text)
  text=re.sub(' [:] ',': ',text)
  text=re.sub(' [;] ','; ',text)
  text=re.sub(' [*] ','* ',text)
  text=re.sub(' [\'] ','\'',text)
  text=re.sub(' [\"] ','\"',text)
  return text
def call(self,text,config=None):
  input seq=self.preprocess(text)
  states=self.encoder(input seq,training=False)
  target seq=np.zeros((1,1))
  target_seq[:,:]=sequences2ids(['<start>']).numpy()[0][0]
  stop condition=False
  decoded=[]
  while not stop condition:
    decoder_outputs,new_states=self.decoder([target_seq,states],training=False)
      index=tf.argmax(decoder outputs[:,-1,:],axis=-1).numpy().item()
    index=self.sample(decoder_outputs[0,0,:]).item()
    word=ids2sequences([index])
    if word=='<end> ' or len(decoded)>=max sequence length:
       stop condition=True
    else:
```

#

```
decoded.append(index)
  target_seq=np.zeros((1,1))
  target_seq[:,:]=index
  states=new_states
return self.postprocess(ids2sequences(decoded))
```

chatbot=ChatBot(model.encoder,model.decoder,name='chatbot')
chatbot.summary()

Model: "chatbot_encoder"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, None)]	0
<pre>encoder_embedding (Embeddin g)</pre>	(None, None, 256)	625408
<pre>layer_normalization (LayerN ormalization)</pre>	(None, None, 256)	512
encoder_lstm (LSTM)	[(None, None, 256), (None, 256), (None, 256)]	525312

Total params: 1,151,232 Trainable params: 1,151,232 Non-trainable params: 0

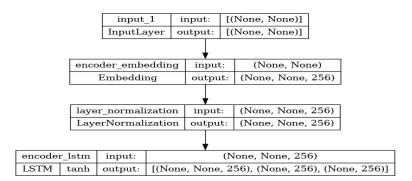
Model: "chatbot_decoder"

Layer (type)	Output Shape	Param #	Connected to
======= input_4 (InputLayer)	[(None, None)]	0	[]
decoder_embedding (Embedding)	(None, None, 256)	625408	['input_4[0][0]']
<pre>layer_normalization (LayerNorm [0]'] alization)</pre>	n (None, None, 256)	512	['decoder_embedding[0]
input_2 (InputLayer)	[(None, 256)]	0	[]

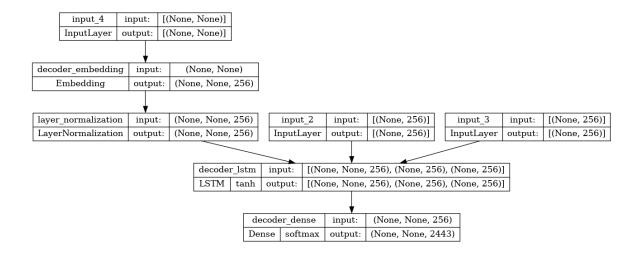
```
input 3 (InputLayer)
                        [(None, 256)]
                                                 decoder_lstm (LSTM)
                        [(None, None, 256), 525312
                                                 ['layer_normalization[1]
[0]',
                         (None, 256),
                                                  'input_2[0][0]',
                                                  'input_3[0][0]']
                         (None, 256)]
decoder_dense (Dense)
                        (None, None, 2443)
                                                 ['decoder_lstm[0][0]']
                                        627851
______
=======
```

Total params: 1,779,083 Trainable params: 1,779,083 Non-trainable params: 0

tf.keras.utils.plot_model(chatbot.encoder,to_file='encoder.png',show_shapes=True,show_layer_activation s=True)



tf.keras.utils.plot_model(chatbot.decoder,to_file='decoder.png',show_shapes=True,show_layer_activation s=True)



VII.Time to Chat

```
def print conversation(texts):
  for text in texts:
    print(f'You: {text}')
    print(f'Bot: {chatbot(text)}')
    print('=======')
print conversation([
  'hi',
  'do yo know me?',
  'what is your name?',
  'you are bot?',
  'hi, how are you doing?',
  "i'm pretty good. thanks for asking.",
  "Don't ever be in a hurry",
  "I'm gonna put some dirt in your eye ",
  "'You're trash ",
  "I've read all your research on nano-technology ",
  "You want forgiveness? Get religion",
  "While you're using the bathroom, i'll order some food.",
```

```
"Wow! that's terrible.",
  "We'll be here forever.",
  "I need something that's reliable.",
 "A speeding car ran a red light, killing the girl.",
 "'Tomorrow we'll have rice and fish for lunch."',
 "I like this restaurant because they give you free bread."])
OUTPUT:
You: hi
Bot: i have to go to the bathroom.
_____
You: do yo know me?
Bot: yes, it's too close to the other.
_____
You: what is your name?
Bot: i have to walk the house.
You: you are bot?
Bot: no, i have. all my life.
_____
You: hi, how are you doing?
Bot: i'm going to be a teacher.
You: i'm pretty good. thanks for asking.
```

Bot: no problem. i'll have to give you the english assignments from my mind. _____ You: Don't ever be in a hurry Bot: it's not a great. _____ You: I'm gonna put some dirt in your eye Bot: that's a good idea. _____ You: You're trash Bot: the tv news is reporting a bank robbery. _____ You: I've read all your research on nano-technology Bot: it's the weather. i've gone around the world. _____ You: You want forgiveness? Get religion Bot: no, i'll be my. _____ You: While you're using the bathroom, i'll order some food. Bot: don't order for me. i've been a cheater. You: Wow! that's terrible. Bot: never park your car under the house. _____ You: We'll be here forever. Bot: we'll be there in half an hour.

You: I need something that's reliable.

```
Bot: you need a car with low mileage.
```

You: A speeding car ran a red light, killing the girl.

Bot: what happened?

You: Tomorrow we'll have rice and fish for lunch.

Bot: i'll make a sandwich.

You: I like this restaurant because they give you free bread.

Bot: well, i think that's a good idea.

FULL SOURCE CODE FOR THIS PROJECT:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
import re,string
```

```
data = {
  'question': [
```

```
"hi, how are you doing?",
    "i'm fine. how about yourself?",
    "i'm pretty good. thanks for asking.",
    "no problem. so how have you been?",
    "i've been great. what about you?"
  ],
  'answer': [
    "i'm fine. how about yourself?",
    "i'm pretty good. thanks for asking.",
    "no problem. so how have you been?",
    "i've been great. what about you?",
    "i've been good. i'm in school right now."
  1
}
df = pd.DataFrame(data)
print(df)
df['question tokens']=df['question'].apply(lambda x:len(x.split()))
df['answer tokens']=df['answer'].apply(lambda x:len(x.split()))
plt.style.use('fivethirtyeight')
fig,ax=plt.subplots(nrows=1,ncols=2,figsize=(20,5))
sns.set_palette('Set2')
sns.histplot(x=df['question tokens'],data=df,kde=True,ax=ax[0])
sns.histplot(x=df['answer tokens'],data=df,kde=True,ax=ax[1])
sns.jointplot(x='question tokens',y='answer tokens',data=df,kind='kde',fill=True,cmap='YlGnBu')
plt.show()
def clean text(text):
  text=re.sub('-',' ',text.lower())
```

```
text=re.sub('[.]',' . ',text)
  text=re.sub('[1]',' 1 ',text)
  text=re.sub('[2]',' 2 ',text)
  text=re.sub('[3]',' 3 ',text)
  text=re.sub('[4]',' 4 ',text)
  text=re.sub('[5]',' 5 ',text)
  text=re.sub('[6]',' 6 ',text)
  text=re.sub('[7]',' 7 ',text)
  text=re.sub('[8]',' 8 ',text)
  text=re.sub('[9]',' 9 ',text)
  text=re.sub('[0]',' 0 ',text)
  text=re.sub('[,]',',',text)
  text=re.sub('[?]',' ? ',text)
  text=re.sub('[!]',' ! ',text)
  text=re.sub('[$]',' $ ',text)
  text=re.sub('[&]',' & ',text)
  text=re.sub('[/]',' / ',text)
  text=re.sub('[:]',':',text)
  text=re.sub('[;]',';',text)
  text=re.sub('[*]',' * ',text)
  text=re.sub('[\']',' \' ',text)
  text=re.sub('[\"]',' \" ',text)
  text=re.sub('\t',' ',text)
  return text
df.drop(columns=['answer tokens','question tokens'],axis=1,inplace=True)
df['encoder_inputs']=df['question'].apply(clean_text)
df['decoder_targets']=df['answer'].apply(clean_text)+' <end>'
```

```
df['decoder_inputs']='<start> '+df['answer'].apply(clean_text)+' <end>'
df.head(10)
df['encoder input tokens']=df['encoder inputs'].apply(lambda x:len(x.split()))
df['decoder input tokens']=df['decoder inputs'].apply(lambda x:len(x.split()))
df['decoder target tokens']=df['decoder targets'].apply(lambda x:len(x.split()))
plt.style.use('fivethirtyeight')
fig,ax=plt.subplots(nrows=1,ncols=3,figsize=(20,5))
sns.set palette('Set2')
sns.histplot(x=df['encoder input tokens'],data=df,kde=True,ax=ax[0])
sns.histplot(x=df['decoder input tokens'],data=df,kde=True,ax=ax[1])
sns.histplot(x=df['decoder target tokens'],data=df,kde=True,ax=ax[2])
sns.jointplot(x='encoder input tokens',y='decoder target
tokens',data=df,kind='kde',fill=True,cmap='YlGnBu')
plt.show()
print(f"After preprocessing: { ' '.join(df[df['encoder input tokens'].max()==df['encoder input
tokens']]['encoder inputs'].values.tolist())}")
print(f"Max encoder input length: {df['encoder input tokens'].max()}")
print(f"Max decoder input length: {df['decoder input tokens'].max()}")
print(f"Max decoder target length: {df['decoder target tokens'].max()}")
df.drop(columns=['question','answer','encoder input tokens','decoder input tokens','decoder target
tokens'],axis=1,inplace=True)
params={
  "vocab size":2500,
  "max sequence length":30,
  "learning rate":0.008,
  "batch size":149,
  "Istm cells":256,
  "embedding_dim":256,
```

```
"buffer_size":10000
}
learning_rate=params['learning_rate']
batch size=params['batch size']
embedding dim=params['embedding dim']
lstm cells=params['lstm cells']
vocab size=params['vocab size']
max sequence length=params['max sequence length']
df.head(10)
vectorizelayer=TextVectorization(
max tokens=vocab size,
standardize=None,
output_mode='int',
output_sequence_length=max_sequence_length
)
vectorize layer.adapt(df['encoder inputs']+' '+df['decoder targets']+' <start> <end>')
vocab size=len(vectorize layer.get vocabulary())
print(f'Vocab size: {len(vectorize_layer.get_vocabulary())}')
print(f'{vectorize layer.get vocabulary()[:12]}')
def sequences2ids(sequence):
  return vectorize layer(sequence)
def ids2sequences(ids):
  decode="
  if type(ids)==int:
    ids=[ids]
  for id in ids:
    decode+=vectorize_layer.get_vocabulary()[id]+''
```

return decode

```
x=sequences2ids(df['encoder_inputs'])
yd=sequences2ids(df['decoder inputs'])
y=sequences2ids(df['decoder targets'])
print(f'Question sentence: hi , how are you ?')
print(f'Question to tokens: {sequences2ids("hi , how are you ?")[:10]}')
print(f'Encoder input shape: {x.shape}')
print(f'Decoder input shape: {yd.shape}')
print(f'Decoder target shape: {y.shape}')
print(f'Encoder input: {x[0][:12]} ...')
print(f'Decoder input: {yd[0][:12]} ...') # shifted by one time step of the target as input to decoder is the
output of the previous timestep
print(f'Decoder target: {y[0][:12]} ...')
data=tf.data.Dataset.from_tensor_slices((x,yd,y))
data=data.shuffle(buffer size)
class Encoder(tf.keras.models.Model):
  def __init__(self,units,embedding_dim,vocab_size,*args,**kwargs) -> None:
    super().__init__(*args,**kwargs)
    self.units=units
    self.vocab_size=vocab_size
    self.embedding_dim=embedding_dim
    self.embedding=Embedding(
      vocab size,
      embedding dim,
      name='encoder_embedding',
      mask zero=True,
      embeddings_initializer=tf.keras.initializers.GlorotNormal()
```

```
)
    self.normalize=LayerNormalization()
    self.lstm=LSTM(
      units,
      dropout=.4,
      return state=True,
      return sequences=True,
      name='encoder_lstm',
      kernel initializer=tf.keras.initializers.GlorotNormal()
    )
  def call(self,encoder_inputs):
    self.inputs=encoder_inputs
    x=self.embedding(encoder_inputs)
    x=self.normalize(x)
    x=Dropout(.4)(x)
    encoder outputs, encoder state h, encoder state c=self.lstm(x)
    self.outputs=[encoder_state_h,encoder_state_c]
    return encoder state h,encoder state c
encoder=Encoder(Istm cells,embedding dim,vocab size,name='encoder')
encoder.call(_[0])
train_data=data.take(int(.9*len(data)))
train_data=train_data.cache()
train_data=train_data.shuffle(buffer_size)
train data=train data.batch(batch size)
train_data=train_data.prefetch(tf.data.AUTOTUNE)
```

```
train_data_iterator=train_data.as_numpy_iterator()
val_data=data.skip(int(.9*len(data))).take(int(.1*len(data)))
val data=val data.batch(batch size)
val data=val data.prefetch(tf.data.AUTOTUNE)
_=train_data_iterator.next()
print(f'Number of train batches: {len(train data)}')
print(f'Number of training data: {len(train data)*batch size}')
print(f'Number of validation batches: {len(val data)}')
print(f'Number of validation data: {len(val data)*batch size}')
print(f'Encoder Input shape (with batches): {_[0].shape}')
print(f'Decoder Input shape (with batches): {_[1].shape}')
print(f'Target Output shape (with batches): { [2].shape}')
class Decoder(tf.keras.models.Model):
  def init (self,units,embedding dim,vocab size,*args,**kwargs) -> None:
    super(). init (*args,**kwargs)
    self.units=units
    self.embedding dim=embedding dim
    self.vocab size=vocab size
    self.embedding=Embedding(
      vocab_size,
      embedding dim,
      name='decoder_embedding',
      mask zero=True,
      embeddings initializer=tf.keras.initializers.HeNormal()
    self.normalize=LayerNormalization()
```

```
self.lstm=LSTM(
      units,
      dropout=.4,
      return_state=True,
      return_sequences=True,
      name='decoder Istm',
      kernel initializer=tf.keras.initializers.HeNormal()
    )
    self.fc=Dense(
      vocab_size,
      activation='softmax',
      name='decoder_dense',
      kernel_initializer=tf.keras.initializers.HeNormal()
    )
 def call(self,decoder_inputs,encoder_states):
   x=self.embedding(decoder inputs)
    x=self.normalize(x)
    x=Dropout(.4)(x)
    x,decoder state h,decoder state c=self.lstm(x,initial state=encoder states)
   x=self.normalize(x)
    x=Dropout(.4)(x)
    return self.fc(x)
decoder=Decoder(lstm_cells,embedding_dim,vocab_size,name='decoder')
decoder(_[1][:1],encoder(_[0][:1]))
class ChatBotTrainer(tf.keras.models.Model):
 def __init__(self,encoder,decoder,*args,**kwargs):
```

```
super().__init__(*args,**kwargs)
 self.encoder=encoder
 self.decoder=decoder
def loss_fn(self,y_true,y_pred):
 loss=self.loss(y true,y pred)
 mask=tf.math.logical not(tf.math.equal(y true,0))
 mask=tf.cast(mask,dtype=loss.dtype)
 loss*=mask
  return tf.reduce mean(loss)
def accuracy_fn(self,y_true,y_pred):
  pred_values = tf.cast(tf.argmax(y_pred, axis=-1), dtype='int64')
 correct = tf.cast(tf.equal(y_true, pred_values), dtype='float64')
 mask = tf.cast(tf.greater(y_true, 0), dtype='float64')
  n correct = tf.keras.backend.sum(mask * correct)
  n total = tf.keras.backend.sum(mask)
  return n_correct / n_total
def call(self,inputs):
 encoder inputs,decoder inputs=inputs
 encoder_states=self.encoder(encoder_inputs)
  return self.decoder(decoder inputs,encoder states)
def train step(self,batch):
 encoder inputs, decoder inputs, y=batch
 with tf.GradientTape() as tape:
    encoder_states=self.encoder(encoder_inputs,training=True)
```

```
y_pred=self.decoder(decoder_inputs,encoder_states,training=True)
      loss=self.loss_fn(y,y_pred)
      acc=self.accuracy_fn(y,y_pred)
    variables=self.encoder.trainable_variables+self.decoder.trainable_variables
    grads=tape.gradient(loss,variables)
    self.optimizer.apply gradients(zip(grads,variables))
    metrics={'loss':loss,'accuracy':acc}
    return metrics
  def test_step(self,batch):
    encoder_inputs,decoder_inputs,y=batch
    encoder_states=self.encoder(encoder_inputs,training=True)
    y_pred=self.decoder(decoder_inputs,encoder_states,training=True)
    loss=self.loss_fn(y,y_pred)
    acc=self.accuracy_fn(y,y_pred)
    metrics={'loss':loss,'accuracy':acc}
    return metrics
  model=ChatBotTrainer(encoder,decoder,name='chatbot trainer')
model.compile(
  loss=tf.keras.losses.SparseCategoricalCrossentropy(),
  optimizer=tf.keras.optimizers.Adam(learning_rate=learning_rate),
  weighted metrics=['loss','accuracy']
model(_[:2])
history=model.fit(
  train data,
  epochs=100,
```

)

```
validation_data=val_data,
  callbacks=[
    tf.keras.callbacks.TensorBoard(log_dir='logs'),
    tf.keras.callbacks.ModelCheckpoint('ckpt',verbose=1,save best only=True)
  ]
)
fig,ax=plt.subplots(nrows=1,ncols=2,figsize=(20,5))
ax[0].plot(history.history['loss'],label='loss',c='red')
ax[0].plot(history.history['val loss'],label='val loss',c = 'blue')
ax[0].set xlabel('Epochs')
ax[1].set xlabel('Epochs')
ax[0].set_ylabel('Loss')
ax[1].set_ylabel('Accuracy')
ax[0].set title('Loss Metrics')
ax[1].set title('Accuracy Metrics')
ax[1].plot(history.history['accuracy'],label='accuracy')
ax[1].plot(history.history['val accuracy'],label='val accuracy')
ax[0].legend()
ax[1].legend()
plt.show()
model.load weights('ckpt')
model.save('models',save_format='tf')
for idx,i in enumerate(model.layers):
  print('Encoder layers:' if idx==0 else 'Decoder layers: ')
  for j in i.layers:
    print(j)
  print('----')
class ChatBot(tf.keras.models.Model):
```

```
def init (self,base encoder,base decoder,*args,**kwargs):
    super().__init__(*args,**kwargs)
    self.encoder,self.decoder=self.build inference model(base encoder,base decoder)
  def build inference model(self,base encoder,base decoder):
    encoder inputs=tf.keras.Input(shape=(None,))
    x=base_encoder.layers[0](encoder_inputs)
    x=base encoder.layers[1](x)
    x,encoder state h,encoder state c=base encoder.layers[2](x)
encoder=tf.keras.models.Model(inputs=encoder inputs,outputs=[encoder state h,encoder state c],nam
e='chatbot encoder')
    decoder input state h=tf.keras.Input(shape=(lstm cells,))
    decoder_input_state_c=tf.keras.Input(shape=(lstm_cells,))
    decoder inputs=tf.keras.Input(shape=(None,))
    x=base decoder.layers[0](decoder inputs)
    x=base encoder.layers[1](x)
x,decoder state h,decoder state c=base decoder.layers[2](x,initial state=[decoder input state h,decod
er_input_state_c])
    decoder_outputs=base_decoder.layers[-1](x)
    decoder=tf.keras.models.Model(
      inputs=[decoder_inputs,[decoder_input_state_h,decoder_input_state_c]],
      outputs=[decoder_outputs,[decoder_state_h,decoder_state_c]],name='chatbot_decoder'
    return encoder, decoder
  def summary(self):
    self.encoder.summary()
```

```
self.decoder.summary()
def softmax(self,z):
  return np.exp(z)/sum(np.exp(z))
def sample(self,conditional probability,temperature=0.5):
  conditional probability = np.asarray(conditional probability).astype("float64")
  conditional probability = np.log(conditional probability) / temperature
  reweighted conditional probability = self.softmax(conditional probability)
  probas = np.random.multinomial(1, reweighted conditional probability, 1)
  return np.argmax(probas)
def preprocess(self,text):
  text=clean text(text)
  seq=np.zeros((1,max sequence length),dtype=np.int32)
  for i,word in enumerate(text.split()):
    seq[:,i]=sequences2ids(word).numpy()[0]
  return seq
def postprocess(self,text):
  text=re.sub(' - ','-',text.lower())
  text=re.sub(' [.] ','. ',text)
  text=re.sub(' [1] ','1',text)
  text=re.sub(' [2] ','2',text)
  text=re.sub(' [3] ','3',text)
  text=re.sub(' [4] ','4',text)
  text=re.sub(' [5] ','5',text)
  text=re.sub(' [6] ','6',text)
```

```
text=re.sub(' [7] ','7',text)
  text=re.sub(' [8] ','8',text)
  text=re.sub(' [9] ','9',text)
  text=re.sub(' [0] ','0',text)
  text=re.sub(' [,] ',', ',text)
  text=re.sub(' [?] ','? ',text)
  text=re.sub(' [!] ','! ',text)
  text=re.sub(' [$] ','$ ',text)
  text=re.sub(' [&] ','& ',text)
  text=re.sub(' [/] ','/ ',text)
  text=re.sub(' [:] ',': ',text)
  text=re.sub(' [;] ','; ',text)
  text=re.sub('[*]','*',text)
  text=re.sub(' [\'] ','\'',text)
  text=re.sub(' [\"] ','\"',text)
  return text
def call(self,text,config=None):
  input seq=self.preprocess(text)
  states=self.encoder(input seq,training=False)
  target_seq=np.zeros((1,1))
  target_seq[:,:]=sequences2ids(['<start>']).numpy()[0][0]
  stop condition=False
  decoded=[]
  while not stop condition:
    decoder_outputs,new_states=self.decoder([target_seq,states],training=False)
      index=tf.argmax(decoder outputs[:,-1,:],axis=-1).numpy().item()
    index=self.sample(decoder_outputs[0,0,:]).item()
```

#

```
word=ids2sequences([index])
      if word=='<end> ' or len(decoded)>=max_sequence_length:
        stop condition=True
      else:
        decoded.append(index)
        target seq=np.zeros((1,1))
        target seq[:,:]=index
        states=new states
    return self.postprocess(ids2sequences(decoded))
chatbot=ChatBot(model.encoder,model.decoder,name='chatbot')
chatbot.summary()
tf.keras.utils.plot model(chatbot.encoder,to file='encoder.png',show shapes=True,show layer activation
s=True)
tf.keras.utils.plot model(chatbot.decoder,to file='decoder.png',show shapes=True,show layer activation
s=True)
def print conversation(texts):
  for text in texts:
    print(f'You: {text}')
    print(f'Bot: {chatbot(text)}')
    print('=======')
print_conversation([
  'hi',
  'do yo know me?',
  'what is your name?',
  'you are bot?',
  'hi, how are you doing?',
  "i'm pretty good. thanks for asking.",
  "Don't ever be in a hurry",
```

```
""I'm gonna put some dirt in your eye "",
""You're trash "",
""I've read all your research on nano-technology "",
""You want forgiveness? Get religion",
""While you're using the bathroom, i'll order some food."",
""Wow! that's terrible."',
""We'll be here forever."',
""I need something that's reliable."',
""A speeding car ran a red light, killing the girl."',
""Tomorrow we'll have rice and fish for lunch."',
```

""I like this restaurant because they give you free bread.""])