**CREATE A CHATBOT IN PYTHON**

**COLLEGE NAME: ARULMIGU MEENAKSHI AMMAN COLLEGE OF ENGINEERING**

**COLLEGE CODE:5103**

**TEAM MEMBER**

**REG NO: 510321104013**

**Name: THAMIZHENIYAN.K**

**Phase-4: Development Part 2**

**Project Title: Create a chatbot in python**



**Basic introduction for development part 2:**

**Here are some tips for feature engineering for a chatbot in Python:**

* **Identify the relevant features:**The first step is to identify the features that are relevant to the chatbot's task. For example, if you are building a customer service chatbot, the relevant features might include the customer's name, account number, and the issue they are having.
* **Preprocess the data:** Once you have identified the relevant features, you need to preprocess the data. This involves cleaning the data, removing any noise or errors, and converting the data to a format that is compatible with your machine learning model.
* **Create new features:**You can also create new features by combining existing features or transforming existing features in new ways. For example, you could create a new feature that represents the customer's sentiment or the urgency of their issue.
* **Select the most important features:** Not all features are created equal. Some features will be more important for the chatbot's task than others. You can use feature selection techniques to identify the most important features and remove the less important features.

**Here are some examples of specific feature engineering techniques that can be used for a chatbot in Python:**

* **Tokenization:** Tokenization is the process of breaking down a text string into smaller units, such as words or tokens. This can be useful for tasks such as natural language processing (NLP) and machine learning.
* **Lemmatization:**Lemmatization is the process of reducing a word to its root form. This can be useful for tasks such as NLP and machine learning, as it can help to improve the accuracy of the model.
* **Part-of-speech tagging:** Part-of-speech tagging is the process of assigning a part-of-speech tag to each word in a sentence. This can be useful for tasks such as NLP and machine learning, as it can help to improve the accuracy of the model.
* **Named entity recognition (NER):** NER is the process of identifying and classifying named entities, such as people, places, and organizations, in a text. This can be useful for tasks such as NLP and machine learning, as it can help to extract important information from the text.
* **Feature hashing:** Feature hashing is a technique for reducing the dimensionality of a dataset. This can be useful for tasks such as machine learning, as it can help to improve the performance of the model.

Once you have engineered your features, you can use them to train a machine learning model. The machine learning model will learn to predict the chatbot's response based on the features.

**Here are some Python libraries that can be used for feature engineering for a chatbot:**

* **NLTK: NLTK is a Python library for NLP:** It provides a variety of tools for tokenization, lemmatization, part-of-speech tagging, NER, and other NLP tasks.
* **SpaCy:**SpaCy is another Python library for NLP. It provides a variety of tools for tokenization, lemmatization, part-of-speech tagging, NER, and other NLP tasks.
* **Scikit-learn:** Scikit-learn is a Python library for machine learning. It provides a variety of tools for feature selection, machine learning model training, and model evaluation.

**To train a model for a chatbot in Python, we can follow these steps:**

1. **Choose a machine learning algorithm.** There are many different machine learning algorithms that can be used for chatbots. Some popular choices include:
   * Support vector machines (SVMs)
   * Random forests
   * Gradient boosted trees
   * Neural networks
2. **Prepare the training data:** The training data should consist of pairs of inputs and outputs. The inputs are the user's messages, and the outputs are the chatbot's responses. You can collect training data from a variety of sources, such as customer service logs, chat transcripts, or social media conversations.
3. **Preprocess the data:** Once you have collected the training data, you need to preprocess it. This involves cleaning the data, removing any noise or errors, and converting the data to a format that is compatible with your machine learning algorithm.
4. **Train the model:**Once the data is preprocessed, you can train the machine learning model. This involves feeding the training data to the model and allowing the model to learn the relationship between the inputs and outputs.
5. **Evaluate the model.** Once the model is trained, you need to evaluate its performance on a held-out test set. This will give you an idea of how well the model will generalize to new data.
6. Deploy the model. Once you are satisfied with the model's performance, you can deploy it to production. This means making the model available so that it can be used to generate responses to user messages.

**Here are some Python libraries that can be used for model training for a chatbot:**

* **Scikit-learn:** Scikit-learn is a Python library for machine learning. It provides a variety of tools for machine learning model training, evaluation, and deployment.
* **TensorFlow:**TensorFlow is a Python library for deep learning. It can be used to train neural network models for chatbots.
* **PyTorch:**PyTorch is another Python library for deep learning. It can be used to train neural network models for chatbots.

Once you have trained a model for your chatbot, you can use it to generate responses to user messages. You can do this by integrating the model with a chatbot platform, such as Dialogflow or Rasa.

**Here are some tips for training a model for a chatbot in Python:**

* Use a large and diverse training dataset. The larger and more diverse the training dataset, the better the model will be able to generalize to new data.
* Use a variety of feature engineering techniques. Feature engineering can help to improve the accuracy and performance of the model.
* Use a regularization technique. Regularization can help to prevent the model from overfitting the training data.
* Evaluate the model on a held-out test set. This will give you an idea of how well the model will generalize to new data.
* Monitor the model's performance in production. This will help you to identify any problems with the model and to make necessary adjustments.

**To evaluate a chatbot in Python, you can use a variety of metrics, such as:**

* **Accuracy:**Accuracy is the percentage of user messages that the chatbot responds to correctly.
* **Precision:** Precision is the percentage of chatbot responses that are relevant to the user's message.
* **Recall:** Recall is the percentage of relevant chatbot responses that are actually generated by the chatbot.
* **F1 score:** The F1 score is a harmonic mean of precision and recall. It is a good overall measure of the chatbot's performance.
* **User satisfaction:** User satisfaction is a subjective measure of how happy users are with the chatbot's performance. You can collect user feedback through surveys or interviews.

In addition to these metrics, you can also evaluate the chatbot's performance on specific tasks, such as customer service or booking appointments.

**Here are some tips for evaluating a chatbot in Python:**

* Use a held-out test set. The test set should be different from the training set. This will help you to evaluate how well the chatbot generalizes to new data.
* Use a variety of evaluation metrics. Different metrics measure different aspects of the chatbot's performance. By using a variety of metrics, you can get a more complete picture of the chatbot's strengths and weaknesses.
* Get feedback from users. User feedback is essential for evaluating the chatbot's performance. Ask users for their feedback on the chatbot's accuracy, relevance, and overall satisfaction.

**Here are some Python libraries that can be used for evaluation for a chatbot:**

* **Scikit-learn:** Scikit-learn provides a variety of tools for machine learning model evaluation, including metrics such as accuracy, precision, recall, and F1 score.
* **NLTK:**NLTK provides a variety of tools for NLP evaluation, including metrics such as word error rate and BLEU score.
* **SpaCy:** SpaCy provides a variety of tools for NLP evaluation, including metrics such as word error rate and BLEU score.

**INTRODUCTION:**

A chatbot is a computer program that can simulate human conversation. Chatbots can be used for a wide range of purposes, from answering customer service inquiries to providing recommendations for products and services.

Python is a popular programming language for creating chatbots. It is relatively easy to learn and has a large community of users and developers. There are also a number of Python libraries that make it easy to build chatbots.

**To create a chatbot in Python, we will need to:**

* **Install the necessary dependencies:** This includes the Python programming language and a chatbot library such as Chatter Bot or Rasa.
* **Create a chatbot object:** This object will represent your chatbot and will be responsible for handling incoming messages and generating responses.
* **Train your chatbot:** This can be done by providing it with a dataset of example conversations.
* **Deploy your chatbot:** Once your chatbot is trained, you can deploy it to a production environment so that users can interact with it.

**GIVEN DATA SET:**

**DATA SET LINK**:( <https://www.kaggle.com/datasets/grafstor/simple-dialogs-for-chatbot>.)

hi, how are you doing? i'm fine. how about yourself?

i'm fine. how about yourself? i'm pretty good. thanks for asking.

i'm pretty good. thanks for asking. no problem. so how have you been?

no problem. so how have you been? i've been great. what about you?

i've been great. what about you? i've been good. i'm in school right now.

i've been good. i'm in school right now. what school do you go to?

what school do you go to? i go to pcc.

i go to pcc. do you like it there?

do you like it there? it's okay. it's a really big campus.

it's okay. it's a really big campus. good luck with school.

good luck with school.

Here is how to apply design thinking to chatbot design:

**1. Empathize:**

Start by understanding your users and their needs. What are their goals? What challenges do they face? What kind of experience do they want to have with your chatbot?

You can gather this information through user research methods such as interviews, surveys, and usability testing.

**2. Define:**

Once you understand your users, you can define the problem you want to solve with your chatbot. What specific task or task do you want it to be able to perform?

Be as specific as possible, and avoid defining the problem in terms of the solution. For example, instead of saying "I want to create a chatbot that helps people book flights," you could say "I want to create a chatbot that helps people book flights on our airline, without having to go through our website or call our customer service center."

**3. Ideate:**

Once you have a clear definition of the problem, you can start generating ideas for solutions. This is where design thinking gets creative!

Brainstorm as many ideas as you can, no matter how crazy they seem. The goal is to come up with as many different ways to solve the problem as possible.

**4. Prototype:**

Once you have a list of ideas, it's time to start prototyping. This means building a working model of your chatbot so you can test it with users.

Prototypes don't have to be perfect. They can be as simple as a wireframe or a mockup. The important thing is that they are functional and allow you to test the core user experience.

**5. Test:**

Once you have a prototype, you can start testing it with users. This is the most important step in the design thinking process, as it allows you to get feedback on your chatbot and make improvements.

Pay attention to how users interact with your chatbot. What are they doing easily? What are they struggling with? What features do they like? What features do they not like?

Use this feedback to iterate on your prototype and make it better.

Repeat the process

Design thinking is an iterative process, so be prepared to repeat the steps above as needed. The more you test your chatbot with users, the better it will be

# Import Librarie:

In [1]:

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 tensorflow.keras.layers import LSTM,Dense,Embedding,Dropout,LayerNormalization

In [2]:

df=pd.read\_csv('/kaggle/input/simple-dialogs-for-chatbot/dialogs.txt',sep='**\t**',names=['question','answer'])

print(f'Dataframe size: **{**len(df)**}**')

df.head()

Dataframe size: 3725

Out[2]:

|  | Question | Answer |
| --- | --- | --- |
| 0 | hi, how are you doing? | i'm fine. how about yourself? |
| 1 | i'm fine. how about yourself? | i'm pretty good. thanks for asking. |
| 2 | i'm pretty good. thanks for asking. | no problem. so how have you been? |
| 3 | no problem. so how have you been? | i've been great. what about you? |
| 4 | i've been great. what about you? | i've been good. i'm in school right now. |

# Data Preprocessing:

## Data Visualization:

In [3]:

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))

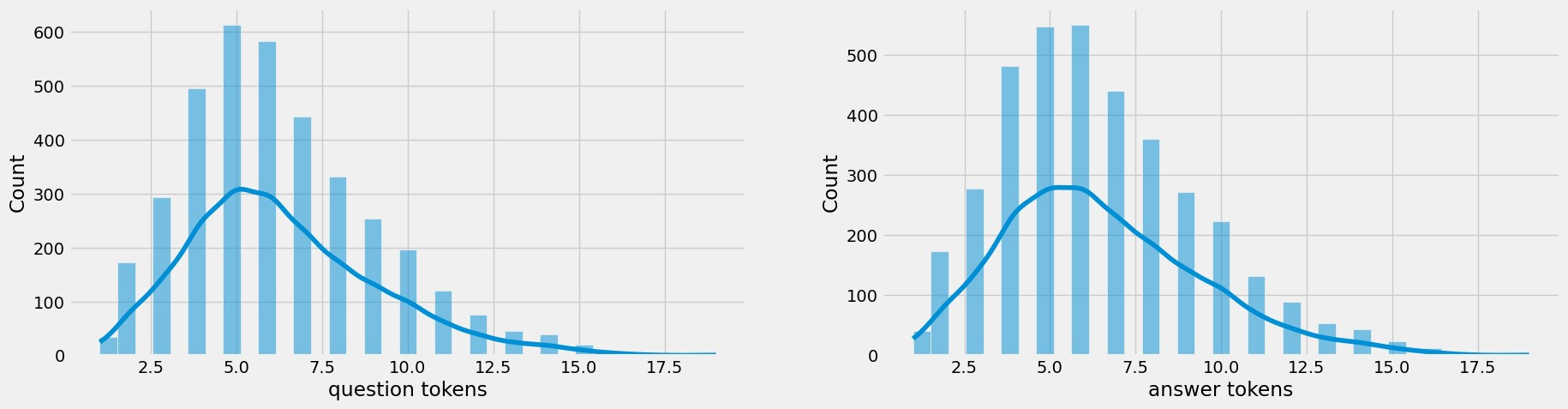
sns.set\_palette('Set2')

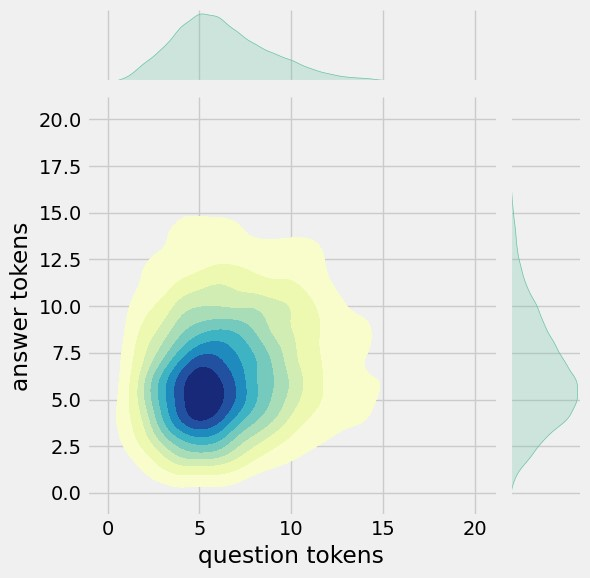
sns.histplot(x=df['question tokens'],data=df,kde=True,ax=ax[0])

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()





## Text Cleaning[:](https://www.kaggle.com/code/sudhakar26/my-chatbot#Text-Cleaning)

In [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)

Out[4]:

|  | question | answer | encoder\_inputs | decoder\_targets | decoder\_inputs |
| --- | --- | --- | --- | --- | --- |
| 0 | hi, how are you doing? | i'm fine. how about yourself? | hi , how are you doing ? | i ' m fine . how about yourself ? <end> | <start> i ' m fine . how about yourself ? <end> |
| 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> | <start> i ' m pretty good . thanks for asking... |
| 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> | <start> no problem . so how have you been ? ... |
| 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> | <start> i ' ve been great . what about you ? ... |
| 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... |
| 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> | <start> what school do you go to ? <end> |
| 6 | what school do you go to? | i go to pcc. | what school do you go to ? | i go to pcc . <end> | <start> i go to pcc . <end> |
| 7 | i go to pcc. | do you like it there? | i go to pcc . | do you like it there ? <end> | <start> do you like it there ? <end> |
| 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... |
| 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> | <start> good luck with school . <end> |

In [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()))

plt.style.use('fivethirtyeight')

fig,ax=plt.subplots(nrows=1,ncols=3,figsize=(20,5))

sns.set\_palette('Set2')

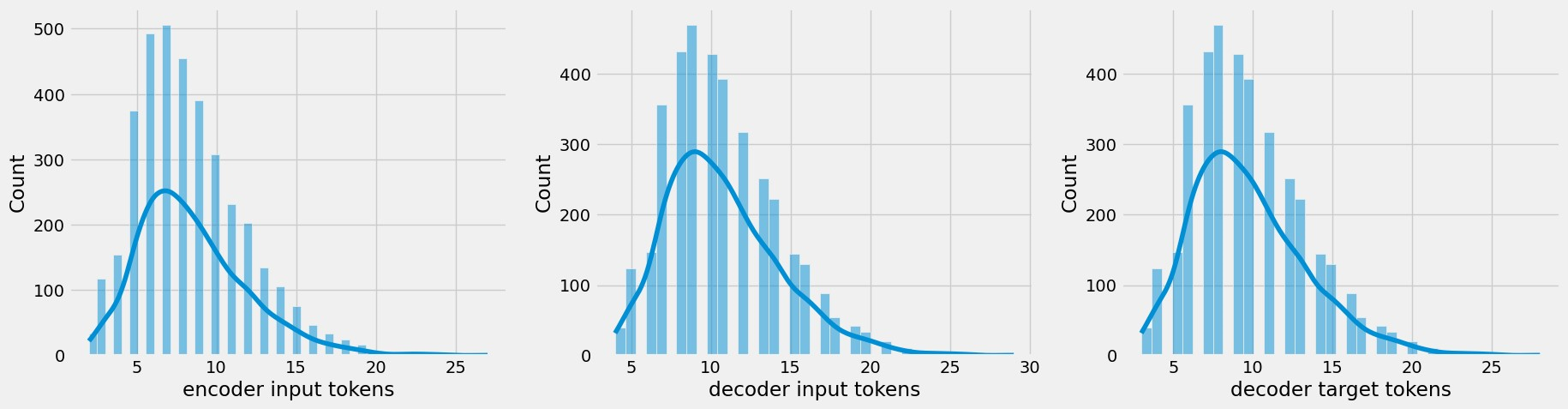
sns.histplot(x=df['encoder input tokens'],data=df,kde=True,ax=ax[0])

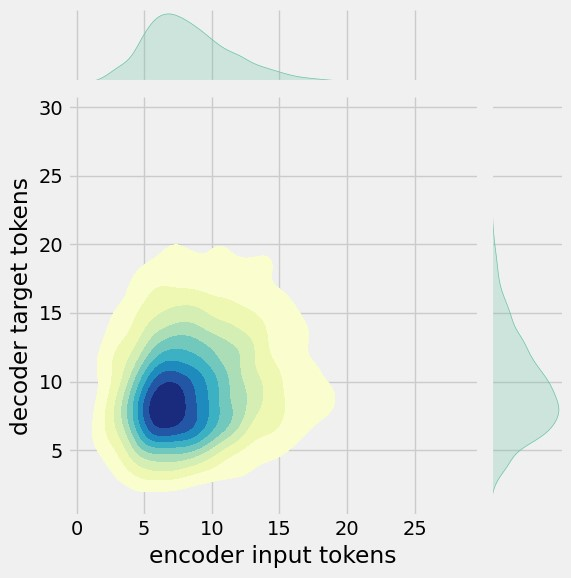
sns.histplot(x=df['decoder input tokens'],data=df,kde=True,ax=ax[1])

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,

"lstm\_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)

After preprocessing: for example , if your birth date is january 1 2 , 1 9 8 7 , write 0 1 / 1 2 / 8 7 .

Max encoder input length: 27

Max decoder input length: 29

Max decoder target length: 28

Out[6]:

|  | encoder\_inputs | decoder\_targets | decoder\_inputs |
| --- | --- | --- | --- |
| 0 | hi , how are you doing ? | i ' m fine . how about yourself ? <end> | <start> i ' m fine . how about yourself ? <end> |
| 1 | i ' m fine . how about yourself ? | i ' m pretty good . thanks for asking . <end> | <start> i ' m pretty good . thanks for asking... |
| 2 | i ' m pretty good . thanks for asking . | no problem . so how have you been ? <end> | <start> no problem . so how have you been ? ... |
| 3 | no problem . so how have you been ? | i ' ve been great . what about you ? <end> | <start> i ' ve been great . what about you ? ... |
| 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... |
| 5 | i ' ve been good . i ' m in school right now . | what school do you go to ? <end> | <start> what school do you go to ? <end> |
| 6 | what school do you go to ? | i go to pcc . <end> | <start> i go to pcc . <end> |
| 7 | i go to pcc . | do you like it there ? <end> | <start> do you like it there ? <end> |
| 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... |
| 9 | it ' s okay . it ' s a really big campus . | good luck with school . <end> | <start> good luck with school . <end> |

**Tokenization:**

In [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']

In [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)

In [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 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] ...

In [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)

**Build Models:**

**Build Encoder:**

In [ ]:

In [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\_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(lstm\_cells,embedding\_dim,vocab\_size,name='encoder')

encoder.call(\_[0])

Out[11]:

(<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

In [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(lstm\_cells,embedding\_dim,vocab\_size,name='decoder')

decoder(\_[1][:1],encoder(\_[0][:1]))

Out[12]:

<tf.Tensor: shape=(1, 30, 2443), dtype=float32, numpy=

array([[[3.4059247e-04, 5.7348556e-05, 2.1294907e-05, ...,

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[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

In [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

In [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])

Out[14]:

<tf.Tensor: shape=(149, 30, 2443), dtype=float32, numpy=

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[1.46621116e-03, 8.02504110e-06, 5.40619949e-05, ...,

1.91874733e-05, 9.72440175e-05, 7.64339056e-05],

[9.69291723e-05, 2.74417835e-05, 1.37613132e-03, ...,

3.60095728e-05, 1.55378671e-04, 1.83973272e-04],

...,

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[1.90027885e-03, 6.92659756e-04, 1.43461803e-04, ...,

1.95525470e-04, 1.71066222e-05, 1.02524005e-04],

[1.90027885e-03, 6.92659756e-04, 1.43461803e-04, ...,

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[2.30146630e-04, 3.91469621e-06, 2.72463716e-04, ...,

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...,

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[[1.19036995e-03, 8.10516722e-05, 2.42324077e-05, ...,

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...,

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[3.52622545e-03, 1.26781175e-03, 1.02695449e-04, ...,

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...,

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2.15980137e-04, 3.02832137e-04, 1.77760507e-04],

...,

[2.41102395e-03, 1.29279669e-03, 9.11735406e-05, ...,

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[2.41102395e-03, 1.29279669e-03, 9.11735406e-05, ...,

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3.03280340e-05, 2.54765386e-04, 2.82170397e-04],

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1.86366087e-04, 7.02239413e-05, 2.54370330e-04],

...,

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6.66230699e-05, 5.76460850e-04, 1.52416309e-04],

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2.55555002e-04, 2.28998848e-04, 4.37303359e-04],

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...,

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3.07609705e-04, 6.09844255e-06, 8.61325825e-05],

[5.20280097e-03, 3.23211338e-04, 2.47709468e-05, ...,

3.07609705e-04, 6.09844255e-06, 8.61325825e-05]]], dtype=float32)>

**Train Model**

In [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)

]

)

Epoch 1/100

23/23 [==============================] - ETA: 0s - loss: 1.6590 - accuracy: 0.2180

Epoch 1: val\_loss improved from inf to 1.21875, saving model to ckpt

23/23 [==============================] - 68s 3s/step - loss: 1.6515 - accuracy: 0.2198 - val\_loss: 1.2187 - val\_accuracy: 0.3072

Epoch 2/100

23/23 [==============================] - ETA: 0s - loss: 1.2327 - accuracy: 0.3087

Epoch 2: val\_loss improved from 1.21875 to 1.10877, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 1.2287 - accuracy: 0.3092 - val\_loss: 1.1088 - val\_accuracy: 0.3415

Epoch 3/100

23/23 [==============================] - ETA: 0s - loss: 1.1008 - accuracy: 0.3368

Epoch 3: val\_loss did not improve from 1.10877

23/23 [==============================] - 22s 973ms/step - loss: 1.0984 - accuracy: 0.3370 - val\_loss: 1.1161 - val\_accuracy: 0.3315

Epoch 4/100

23/23 [==============================] - ETA: 0s - loss: 1.0209 - accuracy: 0.3536

Epoch 4: val\_loss improved from 1.10877 to 0.95189, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 1.0186 - accuracy: 0.3540 - val\_loss: 0.9519 - val\_accuracy: 0.3718

Epoch 5/100

23/23 [==============================] - ETA: 0s - loss: 0.9622 - accuracy: 0.3673

Epoch 5: val\_loss did not improve from 0.95189

23/23 [==============================] - 23s 979ms/step - loss: 0.9672 - accuracy: 0.3670 - val\_loss: 0.9642 - val\_accuracy: 0.3666

Epoch 6/100

23/23 [==============================] - ETA: 0s - loss: 0.9159 - accuracy: 0.3801

Epoch 6: val\_loss improved from 0.95189 to 0.94015, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.9182 - accuracy: 0.3796 - val\_loss: 0.9401 - val\_accuracy: 0.3598

Epoch 7/100

23/23 [==============================] - ETA: 0s - loss: 0.8737 - accuracy: 0.3908

Epoch 7: val\_loss improved from 0.94015 to 0.83293, saving model to ckpt

23/23 [==============================] - 52s 2s/step - loss: 0.8746 - accuracy: 0.3900 - val\_loss: 0.8329 - val\_accuracy: 0.4180

Epoch 8/100

23/23 [==============================] - ETA: 0s - loss: 0.8389 - accuracy: 0.4013

Epoch 8: val\_loss improved from 0.83293 to 0.77748, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.8395 - accuracy: 0.4013 - val\_loss: 0.7775 - val\_accuracy: 0.4305

Epoch 9/100

23/23 [==============================] - ETA: 0s - loss: 0.8148 - accuracy: 0.4094

Epoch 9: val\_loss did not improve from 0.77748

23/23 [==============================] - 23s 983ms/step - loss: 0.8187 - accuracy: 0.4084 - val\_loss: 0.8608 - val\_accuracy: 0.3830

Epoch 10/100

23/23 [==============================] - ETA: 0s - loss: 0.7889 - accuracy: 0.4200

Epoch 10: val\_loss improved from 0.77748 to 0.73131, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.7923 - accuracy: 0.4188 - val\_loss: 0.7313 - val\_accuracy: 0.4515

Epoch 11/100

23/23 [==============================] - ETA: 0s - loss: 0.7624 - accuracy: 0.4284

Epoch 11: val\_loss did not improve from 0.73131

23/23 [==============================] - 22s 965ms/step - loss: 0.7615 - accuracy: 0.4282 - val\_loss: 0.8036 - val\_accuracy: 0.4472

Epoch 12/100

23/23 [==============================] - ETA: 0s - loss: 0.7433 - accuracy: 0.4361

Epoch 12: val\_loss did not improve from 0.73131

23/23 [==============================] - 23s 984ms/step - loss: 0.7452 - accuracy: 0.4354 - val\_loss: 0.7384 - val\_accuracy: 0.4623

Epoch 13/100

23/23 [==============================] - ETA: 0s - loss: 0.7246 - accuracy: 0.4493

Epoch 13: val\_loss did not improve from 0.73131

23/23 [==============================] - 23s 988ms/step - loss: 0.7281 - accuracy: 0.4488 - val\_loss: 0.8017 - val\_accuracy: 0.4449

Epoch 14/100

23/23 [==============================] - ETA: 0s - loss: 0.7080 - accuracy: 0.4513

Epoch 14: val\_loss did not improve from 0.73131

23/23 [==============================] - 23s 995ms/step - loss: 0.7080 - accuracy: 0.4509 - val\_loss: 0.7568 - val\_accuracy: 0.4259

Epoch 15/100

23/23 [==============================] - ETA: 0s - loss: 0.6853 - accuracy: 0.4620

Epoch 15: val\_loss did not improve from 0.73131

23/23 [==============================] - 22s 974ms/step - loss: 0.6826 - accuracy: 0.4616 - val\_loss: 0.7376 - val\_accuracy: 0.4502

Epoch 16/100

23/23 [==============================] - ETA: 0s - loss: 0.6731 - accuracy: 0.4673

Epoch 16: val\_loss did not improve from 0.73131

23/23 [==============================] - 23s 983ms/step - loss: 0.6733 - accuracy: 0.4672 - val\_loss: 0.7646 - val\_accuracy: 0.4538

Epoch 17/100

23/23 [==============================] - ETA: 0s - loss: 0.6576 - accuracy: 0.4732

Epoch 17: val\_loss improved from 0.73131 to 0.66131, saving model to ckpt

23/23 [==============================] - 52s 2s/step - loss: 0.6539 - accuracy: 0.4738 - val\_loss: 0.6613 - val\_accuracy: 0.4714

Epoch 18/100

23/23 [==============================] - ETA: 0s - loss: 0.6468 - accuracy: 0.4807

Epoch 18: val\_loss improved from 0.66131 to 0.65303, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.6458 - accuracy: 0.4805 - val\_loss: 0.6530 - val\_accuracy: 0.4993

Epoch 19/100

23/23 [==============================] - ETA: 0s - loss: 0.6353 - accuracy: 0.4881

Epoch 19: val\_loss did not improve from 0.65303

23/23 [==============================] - 23s 994ms/step - loss: 0.6357 - accuracy: 0.4876 - val\_loss: 0.7331 - val\_accuracy: 0.4677

Epoch 20/100

23/23 [==============================] - ETA: 0s - loss: 0.6194 - accuracy: 0.4968

Epoch 20: val\_loss improved from 0.65303 to 0.55054, saving model to ckpt

23/23 [==============================] - 54s 2s/step - loss: 0.6188 - accuracy: 0.4967 - val\_loss: 0.5505 - val\_accuracy: 0.5221

Epoch 21/100

23/23 [==============================] - ETA: 0s - loss: 0.6160 - accuracy: 0.4978

Epoch 21: val\_loss did not improve from 0.55054

23/23 [==============================] - 23s 987ms/step - loss: 0.6182 - accuracy: 0.4965 - val\_loss: 0.6790 - val\_accuracy: 0.4979

Epoch 22/100

23/23 [==============================] - ETA: 0s - loss: 0.6011 - accuracy: 0.5052

Epoch 22: val\_loss did not improve from 0.55054

23/23 [==============================] - 23s 996ms/step - loss: 0.6011 - accuracy: 0.5051 - val\_loss: 0.6221 - val\_accuracy: 0.5277

Epoch 23/100

23/23 [==============================] - ETA: 0s - loss: 0.5950 - accuracy: 0.5079

Epoch 23: val\_loss did not improve from 0.55054

23/23 [==============================] - 23s 987ms/step - loss: 0.5934 - accuracy: 0.5081 - val\_loss: 0.6142 - val\_accuracy: 0.5198

Epoch 24/100

23/23 [==============================] - ETA: 0s - loss: 0.5810 - accuracy: 0.5160

Epoch 24: val\_loss did not improve from 0.55054

23/23 [==============================] - 22s 971ms/step - loss: 0.5803 - accuracy: 0.5170 - val\_loss: 0.5759 - val\_accuracy: 0.5137

Epoch 25/100

23/23 [==============================] - ETA: 0s - loss: 0.5716 - accuracy: 0.5227

Epoch 25: val\_loss did not improve from 0.55054

23/23 [==============================] - 23s 986ms/step - loss: 0.5733 - accuracy: 0.5229 - val\_loss: 0.6344 - val\_accuracy: 0.5169

Epoch 26/100

23/23 [==============================] - ETA: 0s - loss: 0.5676 - accuracy: 0.5225

Epoch 26: val\_loss did not improve from 0.55054

23/23 [==============================] - 22s 963ms/step - loss: 0.5708 - accuracy: 0.5210 - val\_loss: 0.6254 - val\_accuracy: 0.4882

Epoch 27/100

23/23 [==============================] - ETA: 0s - loss: 0.5616 - accuracy: 0.5291

Epoch 27: val\_loss did not improve from 0.55054

23/23 [==============================] - 23s 988ms/step - loss: 0.5624 - accuracy: 0.5280 - val\_loss: 0.6774 - val\_accuracy: 0.5379

Epoch 28/100

23/23 [==============================] - ETA: 0s - loss: 0.5531 - accuracy: 0.5318

Epoch 28: val\_loss did not improve from 0.55054

23/23 [==============================] - 22s 949ms/step - loss: 0.5543 - accuracy: 0.5310 - val\_loss: 0.7284 - val\_accuracy: 0.5302

Epoch 29/100

23/23 [==============================] - ETA: 0s - loss: 0.5398 - accuracy: 0.5389

Epoch 29: val\_loss did not improve from 0.55054

23/23 [==============================] - 23s 1s/step - loss: 0.5391 - accuracy: 0.5398 - val\_loss: 0.7385 - val\_accuracy: 0.5193

Epoch 30/100

23/23 [==============================] - ETA: 0s - loss: 0.5375 - accuracy: 0.5416

Epoch 30: val\_loss improved from 0.55054 to 0.50346, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.5384 - accuracy: 0.5417 - val\_loss: 0.5035 - val\_accuracy: 0.5411

Epoch 31/100

23/23 [==============================] - ETA: 0s - loss: 0.5270 - accuracy: 0.5481

Epoch 31: val\_loss did not improve from 0.50346

23/23 [==============================] - 22s 958ms/step - loss: 0.5262 - accuracy: 0.5477 - val\_loss: 0.5805 - val\_accuracy: 0.5457

Epoch 32/100

23/23 [==============================] - ETA: 0s - loss: 0.5304 - accuracy: 0.5447

Epoch 32: val\_loss did not improve from 0.50346

23/23 [==============================] - 22s 963ms/step - loss: 0.5329 - accuracy: 0.5435 - val\_loss: 0.5374 - val\_accuracy: 0.5725

Epoch 33/100

23/23 [==============================] - ETA: 0s - loss: 0.5196 - accuracy: 0.5520

Epoch 33: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 975ms/step - loss: 0.5211 - accuracy: 0.5518 - val\_loss: 0.6217 - val\_accuracy: 0.5066

Epoch 34/100

23/23 [==============================] - ETA: 0s - loss: 0.5129 - accuracy: 0.5558

Epoch 34: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 1000ms/step - loss: 0.5129 - accuracy: 0.5556 - val\_loss: 0.6070 - val\_accuracy: 0.5653

Epoch 35/100

23/23 [==============================] - ETA: 0s - loss: 0.5059 - accuracy: 0.5620

Epoch 35: val\_loss did not improve from 0.50346

23/23 [==============================] - 22s 966ms/step - loss: 0.5081 - accuracy: 0.5614 - val\_loss: 0.6153 - val\_accuracy: 0.5452

Epoch 36/100

23/23 [==============================] - ETA: 0s - loss: 0.5037 - accuracy: 0.5619

Epoch 36: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 980ms/step - loss: 0.5063 - accuracy: 0.5617 - val\_loss: 0.5328 - val\_accuracy: 0.5873

Epoch 37/100

23/23 [==============================] - ETA: 0s - loss: 0.4977 - accuracy: 0.5682

Epoch 37: val\_loss did not improve from 0.50346

23/23 [==============================] - 22s 969ms/step - loss: 0.4980 - accuracy: 0.5682 - val\_loss: 0.5976 - val\_accuracy: 0.5693

Epoch 38/100

23/23 [==============================] - ETA: 0s - loss: 0.4939 - accuracy: 0.5704

Epoch 38: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 993ms/step - loss: 0.4953 - accuracy: 0.5687 - val\_loss: 0.5937 - val\_accuracy: 0.5236

Epoch 39/100

23/23 [==============================] - ETA: 0s - loss: 0.4860 - accuracy: 0.5758

Epoch 39: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 986ms/step - loss: 0.4868 - accuracy: 0.5746 - val\_loss: 0.6155 - val\_accuracy: 0.5457

Epoch 40/100

23/23 [==============================] - ETA: 0s - loss: 0.4809 - accuracy: 0.5778

Epoch 40: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 1s/step - loss: 0.4821 - accuracy: 0.5760 - val\_loss: 0.5046 - val\_accuracy: 0.5662

Epoch 41/100

23/23 [==============================] - ETA: 0s - loss: 0.4781 - accuracy: 0.5817

Epoch 41: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 990ms/step - loss: 0.4782 - accuracy: 0.5821 - val\_loss: 0.5256 - val\_accuracy: 0.5907

Epoch 42/100

23/23 [==============================] - ETA: 0s - loss: 0.4713 - accuracy: 0.5836

Epoch 42: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 982ms/step - loss: 0.4729 - accuracy: 0.5824 - val\_loss: 0.6387 - val\_accuracy: 0.5456

Epoch 43/100

23/23 [==============================] - ETA: 0s - loss: 0.4641 - accuracy: 0.5904

Epoch 43: val\_loss did not improve from 0.50346

23/23 [==============================] - 23s 1s/step - loss: 0.4627 - accuracy: 0.5908 - val\_loss: 0.5668 - val\_accuracy: 0.5741

Epoch 44/100

23/23 [==============================] - ETA: 0s - loss: 0.4608 - accuracy: 0.5921

Epoch 44: val\_loss improved from 0.50346 to 0.49920, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.4618 - accuracy: 0.5920 - val\_loss: 0.4992 - val\_accuracy: 0.5768

Epoch 45/100

23/23 [==============================] - ETA: 0s - loss: 0.4592 - accuracy: 0.5902

Epoch 45: val\_loss did not improve from 0.49920

23/23 [==============================] - 22s 970ms/step - loss: 0.4599 - accuracy: 0.5887 - val\_loss: 0.5423 - val\_accuracy: 0.5854

Epoch 46/100

23/23 [==============================] - ETA: 0s - loss: 0.4535 - accuracy: 0.5978

Epoch 46: val\_loss improved from 0.49920 to 0.48429, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.4552 - accuracy: 0.5966 - val\_loss: 0.4843 - val\_accuracy: 0.6049

Epoch 47/100

23/23 [==============================] - ETA: 0s - loss: 0.4528 - accuracy: 0.5987

Epoch 47: val\_loss improved from 0.48429 to 0.47868, saving model to ckpt

23/23 [==============================] - 54s 2s/step - loss: 0.4537 - accuracy: 0.5990 - val\_loss: 0.4787 - val\_accuracy: 0.5906

Epoch 48/100

23/23 [==============================] - ETA: 0s - loss: 0.4441 - accuracy: 0.6016

Epoch 48: val\_loss did not improve from 0.47868

23/23 [==============================] - 23s 982ms/step - loss: 0.4439 - accuracy: 0.6025 - val\_loss: 0.5746 - val\_accuracy: 0.5542

Epoch 49/100

23/23 [==============================] - ETA: 0s - loss: 0.4436 - accuracy: 0.6041

Epoch 49: val\_loss did not improve from 0.47868

23/23 [==============================] - 22s 951ms/step - loss: 0.4432 - accuracy: 0.6045 - val\_loss: 0.5058 - val\_accuracy: 0.5753

Epoch 50/100

23/23 [==============================] - ETA: 0s - loss: 0.4435 - accuracy: 0.6033

Epoch 50: val\_loss did not improve from 0.47868

23/23 [==============================] - 22s 949ms/step - loss: 0.4441 - accuracy: 0.6043 - val\_loss: 0.6037 - val\_accuracy: 0.5473

Epoch 51/100

23/23 [==============================] - ETA: 0s - loss: 0.4382 - accuracy: 0.6069

Epoch 51: val\_loss did not improve from 0.47868

23/23 [==============================] - 22s 957ms/step - loss: 0.4383 - accuracy: 0.6067 - val\_loss: 0.5206 - val\_accuracy: 0.6154

Epoch 52/100

23/23 [==============================] - ETA: 0s - loss: 0.4293 - accuracy: 0.6125

Epoch 52: val\_loss did not improve from 0.47868

23/23 [==============================] - 23s 971ms/step - loss: 0.4284 - accuracy: 0.6123 - val\_loss: 0.4997 - val\_accuracy: 0.5840

Epoch 53/100

23/23 [==============================] - ETA: 0s - loss: 0.4309 - accuracy: 0.6109

Epoch 53: val\_loss improved from 0.47868 to 0.42987, saving model to ckpt

23/23 [==============================] - 52s 2s/step - loss: 0.4317 - accuracy: 0.6094 - val\_loss: 0.4299 - val\_accuracy: 0.6062

Epoch 54/100

23/23 [==============================] - ETA: 0s - loss: 0.4292 - accuracy: 0.6120

Epoch 54: val\_loss did not improve from 0.42987

23/23 [==============================] - 22s 980ms/step - loss: 0.4309 - accuracy: 0.6115 - val\_loss: 0.6996 - val\_accuracy: 0.5592

Epoch 55/100

23/23 [==============================] - ETA: 0s - loss: 0.4225 - accuracy: 0.6115

Epoch 55: val\_loss did not improve from 0.42987

23/23 [==============================] - 22s 976ms/step - loss: 0.4224 - accuracy: 0.6102 - val\_loss: 0.5500 - val\_accuracy: 0.5769

Epoch 56/100

23/23 [==============================] - ETA: 0s - loss: 0.4220 - accuracy: 0.6180

Epoch 56: val\_loss did not improve from 0.42987

23/23 [==============================] - 23s 995ms/step - loss: 0.4236 - accuracy: 0.6169 - val\_loss: 0.5689 - val\_accuracy: 0.5817

Epoch 57/100

23/23 [==============================] - ETA: 0s - loss: 0.4173 - accuracy: 0.6210

Epoch 57: val\_loss did not improve from 0.42987

23/23 [==============================] - 22s 976ms/step - loss: 0.4161 - accuracy: 0.6217 - val\_loss: 0.4614 - val\_accuracy: 0.6048

Epoch 58/100

23/23 [==============================] - ETA: 0s - loss: 0.4183 - accuracy: 0.6198

Epoch 58: val\_loss did not improve from 0.42987

23/23 [==============================] - 23s 1s/step - loss: 0.4183 - accuracy: 0.6201 - val\_loss: 0.4372 - val\_accuracy: 0.6067

Epoch 59/100

23/23 [==============================] - ETA: 0s - loss: 0.4120 - accuracy: 0.6251

Epoch 59: val\_loss did not improve from 0.42987

23/23 [==============================] - 23s 994ms/step - loss: 0.4136 - accuracy: 0.6237 - val\_loss: 0.6183 - val\_accuracy: 0.5948

Epoch 60/100

23/23 [==============================] - ETA: 0s - loss: 0.4090 - accuracy: 0.6239

Epoch 60: val\_loss did not improve from 0.42987

23/23 [==============================] - 23s 980ms/step - loss: 0.4101 - accuracy: 0.6225 - val\_loss: 0.5042 - val\_accuracy: 0.6161

Epoch 61/100

23/23 [==============================] - ETA: 0s - loss: 0.4051 - accuracy: 0.6314

Epoch 61: val\_loss did not improve from 0.42987

23/23 [==============================] - 23s 1s/step - loss: 0.4077 - accuracy: 0.6296 - val\_loss: 0.5100 - val\_accuracy: 0.6128

Epoch 62/100

23/23 [==============================] - ETA: 0s - loss: 0.4016 - accuracy: 0.6326

Epoch 62: val\_loss did not improve from 0.42987

23/23 [==============================] - 24s 1s/step - loss: 0.4029 - accuracy: 0.6322 - val\_loss: 0.5295 - val\_accuracy: 0.6005

Epoch 63/100

23/23 [==============================] - ETA: 0s - loss: 0.4049 - accuracy: 0.6323

Epoch 63: val\_loss did not improve from 0.42987

23/23 [==============================] - 23s 981ms/step - loss: 0.4069 - accuracy: 0.6316 - val\_loss: 0.5103 - val\_accuracy: 0.6088

Epoch 64/100

23/23 [==============================] - ETA: 0s - loss: 0.3951 - accuracy: 0.6335

Epoch 64: val\_loss did not improve from 0.42987

23/23 [==============================] - 22s 981ms/step - loss: 0.3943 - accuracy: 0.6341 - val\_loss: 0.5366 - val\_accuracy: 0.5869

Epoch 65/100

23/23 [==============================] - ETA: 0s - loss: 0.3967 - accuracy: 0.6344

Epoch 65: val\_loss improved from 0.42987 to 0.40702, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.3972 - accuracy: 0.6352 - val\_loss: 0.4070 - val\_accuracy: 0.6452

Epoch 66/100

23/23 [==============================] - ETA: 0s - loss: 0.3942 - accuracy: 0.6351

Epoch 66: val\_loss did not improve from 0.40702

23/23 [==============================] - 22s 961ms/step - loss: 0.3954 - accuracy: 0.6337 - val\_loss: 0.4963 - val\_accuracy: 0.6039

Epoch 67/100

23/23 [==============================] - ETA: 0s - loss: 0.3884 - accuracy: 0.6409

Epoch 67: val\_loss did not improve from 0.40702

23/23 [==============================] - 22s 951ms/step - loss: 0.3879 - accuracy: 0.6424 - val\_loss: 0.4651 - val\_accuracy: 0.6276

Epoch 68/100

23/23 [==============================] - ETA: 0s - loss: 0.3876 - accuracy: 0.6398

Epoch 68: val\_loss improved from 0.40702 to 0.38016, saving model to ckpt

23/23 [==============================] - 52s 2s/step - loss: 0.3870 - accuracy: 0.6388 - val\_loss: 0.3802 - val\_accuracy: 0.6614

Epoch 69/100

23/23 [==============================] - ETA: 0s - loss: 0.3897 - accuracy: 0.6394

Epoch 69: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 961ms/step - loss: 0.3895 - accuracy: 0.6395 - val\_loss: 0.4046 - val\_accuracy: 0.6587

Epoch 70/100

23/23 [==============================] - ETA: 0s - loss: 0.3855 - accuracy: 0.6433

Epoch 70: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 967ms/step - loss: 0.3870 - accuracy: 0.6432 - val\_loss: 0.4162 - val\_accuracy: 0.6475

Epoch 71/100

23/23 [==============================] - ETA: 0s - loss: 0.3828 - accuracy: 0.6422

Epoch 71: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 986ms/step - loss: 0.3828 - accuracy: 0.6423 - val\_loss: 0.4099 - val\_accuracy: 0.6612

Epoch 72/100

23/23 [==============================] - ETA: 0s - loss: 0.3825 - accuracy: 0.6460

Epoch 72: val\_loss did not improve from 0.38016

23/23 [==============================] - 24s 1s/step - loss: 0.3831 - accuracy: 0.6449 - val\_loss: 0.5160 - val\_accuracy: 0.6117

Epoch 73/100

23/23 [==============================] - ETA: 0s - loss: 0.3795 - accuracy: 0.6451

Epoch 73: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 1s/step - loss: 0.3797 - accuracy: 0.6448 - val\_loss: 0.4963 - val\_accuracy: 0.6231

Epoch 74/100

23/23 [==============================] - ETA: 0s - loss: 0.3769 - accuracy: 0.6479

Epoch 74: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 975ms/step - loss: 0.3783 - accuracy: 0.6459 - val\_loss: 0.4888 - val\_accuracy: 0.6084

Epoch 75/100

23/23 [==============================] - ETA: 0s - loss: 0.3719 - accuracy: 0.6541

Epoch 75: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 971ms/step - loss: 0.3724 - accuracy: 0.6538 - val\_loss: 0.5175 - val\_accuracy: 0.6032

Epoch 76/100

23/23 [==============================] - ETA: 0s - loss: 0.3697 - accuracy: 0.6555

Epoch 76: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 1s/step - loss: 0.3687 - accuracy: 0.6548 - val\_loss: 0.4598 - val\_accuracy: 0.6059

Epoch 77/100

23/23 [==============================] - ETA: 0s - loss: 0.3702 - accuracy: 0.6552

Epoch 77: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 954ms/step - loss: 0.3713 - accuracy: 0.6540 - val\_loss: 0.5650 - val\_accuracy: 0.5824

Epoch 78/100

23/23 [==============================] - ETA: 0s - loss: 0.3685 - accuracy: 0.6548

Epoch 78: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 982ms/step - loss: 0.3675 - accuracy: 0.6557 - val\_loss: 0.4115 - val\_accuracy: 0.6292

Epoch 79/100

23/23 [==============================] - ETA: 0s - loss: 0.3659 - accuracy: 0.6584

Epoch 79: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 970ms/step - loss: 0.3662 - accuracy: 0.6577 - val\_loss: 0.3868 - val\_accuracy: 0.6516

Epoch 80/100

23/23 [==============================] - ETA: 0s - loss: 0.3626 - accuracy: 0.6628

Epoch 80: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 994ms/step - loss: 0.3627 - accuracy: 0.6638 - val\_loss: 0.4733 - val\_accuracy: 0.6388

Epoch 81/100

23/23 [==============================] - ETA: 0s - loss: 0.3623 - accuracy: 0.6578

Epoch 81: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 970ms/step - loss: 0.3621 - accuracy: 0.6577 - val\_loss: 0.5189 - val\_accuracy: 0.5979

Epoch 82/100

23/23 [==============================] - ETA: 0s - loss: 0.3603 - accuracy: 0.6612

Epoch 82: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 982ms/step - loss: 0.3600 - accuracy: 0.6614 - val\_loss: 0.4210 - val\_accuracy: 0.6280

Epoch 83/100

23/23 [==============================] - ETA: 0s - loss: 0.3608 - accuracy: 0.6604

Epoch 83: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 1s/step - loss: 0.3627 - accuracy: 0.6592 - val\_loss: 0.5621 - val\_accuracy: 0.6082

Epoch 84/100

23/23 [==============================] - ETA: 0s - loss: 0.3605 - accuracy: 0.6640

Epoch 84: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 998ms/step - loss: 0.3628 - accuracy: 0.6634 - val\_loss: 0.4241 - val\_accuracy: 0.6462

Epoch 85/100

23/23 [==============================] - ETA: 0s - loss: 0.3498 - accuracy: 0.6713

Epoch 85: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 976ms/step - loss: 0.3484 - accuracy: 0.6713 - val\_loss: 0.4425 - val\_accuracy: 0.6489

Epoch 86/100

23/23 [==============================] - ETA: 0s - loss: 0.3537 - accuracy: 0.6663

Epoch 86: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 1s/step - loss: 0.3543 - accuracy: 0.6656 - val\_loss: 0.4006 - val\_accuracy: 0.6716

Epoch 87/100

23/23 [==============================] - ETA: 0s - loss: 0.3503 - accuracy: 0.6698

Epoch 87: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 987ms/step - loss: 0.3493 - accuracy: 0.6697 - val\_loss: 0.4375 - val\_accuracy: 0.6527

Epoch 88/100

23/23 [==============================] - ETA: 0s - loss: 0.3497 - accuracy: 0.6714

Epoch 88: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 986ms/step - loss: 0.3495 - accuracy: 0.6710 - val\_loss: 0.5339 - val\_accuracy: 0.6160

Epoch 89/100

23/23 [==============================] - ETA: 0s - loss: 0.3500 - accuracy: 0.6671

Epoch 89: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 970ms/step - loss: 0.3501 - accuracy: 0.6666 - val\_loss: 0.4148 - val\_accuracy: 0.6438

Epoch 90/100

23/23 [==============================] - ETA: 0s - loss: 0.3494 - accuracy: 0.6661

Epoch 90: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 995ms/step - loss: 0.3529 - accuracy: 0.6647 - val\_loss: 0.4992 - val\_accuracy: 0.6324

Epoch 91/100

23/23 [==============================] - ETA: 0s - loss: 0.3479 - accuracy: 0.6718

Epoch 91: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 986ms/step - loss: 0.3482 - accuracy: 0.6715 - val\_loss: 0.6037 - val\_accuracy: 0.6195

Epoch 92/100

23/23 [==============================] - ETA: 0s - loss: 0.3436 - accuracy: 0.6767

Epoch 92: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 964ms/step - loss: 0.3452 - accuracy: 0.6764 - val\_loss: 0.4368 - val\_accuracy: 0.6462

Epoch 93/100

23/23 [==============================] - ETA: 0s - loss: 0.3377 - accuracy: 0.6793

Epoch 93: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 984ms/step - loss: 0.3372 - accuracy: 0.6795 - val\_loss: 0.5267 - val\_accuracy: 0.6275

Epoch 94/100

23/23 [==============================] - ETA: 0s - loss: 0.3433 - accuracy: 0.6743

Epoch 94: val\_loss did not improve from 0.38016

23/23 [==============================] - 22s 964ms/step - loss: 0.3453 - accuracy: 0.6736 - val\_loss: 0.4532 - val\_accuracy: 0.6314

Epoch 95/100

23/23 [==============================] - ETA: 0s - loss: 0.3409 - accuracy: 0.6780

Epoch 95: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 987ms/step - loss: 0.3407 - accuracy: 0.6775 - val\_loss: 0.4901 - val\_accuracy: 0.6680

Epoch 96/100

23/23 [==============================] - ETA: 0s - loss: 0.3378 - accuracy: 0.6791

Epoch 96: val\_loss did not improve from 0.38016

23/23 [==============================] - 23s 991ms/step - loss: 0.3388 - accuracy: 0.6793 - val\_loss: 0.5620 - val\_accuracy: 0.6063

Epoch 97/100

23/23 [==============================] - ETA: 0s - loss: 0.3389 - accuracy: 0.6763

Epoch 97: val\_loss improved from 0.38016 to 0.33265, saving model to ckpt

23/23 [==============================] - 53s 2s/step - loss: 0.3402 - accuracy: 0.6765 - val\_loss: 0.3327 - val\_accuracy: 0.6854

Epoch 98/100

23/23 [==============================] - ETA: 0s - loss: 0.3408 - accuracy: 0.6768

Epoch 98: val\_loss did not improve from 0.33265

23/23 [==============================] - 22s 974ms/step - loss: 0.3407 - accuracy: 0.6766 - val\_loss: 0.4046 - val\_accuracy: 0.6695

Epoch 99/100

23/23 [==============================] - ETA: 0s - loss: 0.3388 - accuracy: 0.6795

Epoch 99: val\_loss did not improve from 0.33265

23/23 [==============================] - 23s 985ms/step - loss: 0.3394 - accuracy: 0.6791 - val\_loss: 0.4475 - val\_accuracy: 0.6622

Epoch 100/100

23/23 [==============================] - ETA: 0s - loss: 0.3358 - accuracy: 0.6787

Epoch 100: val\_loss did not improve from 0.33265

23/23 [==============================] - 22s 968ms/step - loss: 0.3385 - accuracy: 0.6773 - val\_loss: 0.3742 - val\_accuracy: 0.6796

**Visualize Metrics:**

In [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()

**Save Model:**

In [17]:

model.load\_weights('ckpt')

model.save('models',save\_format='tf')

In [18]:

for idx,i **in** enumerate(model.layers):

print('Encoder layers:' if idx==0 else 'Decoder layers: ')

for j **in** i.layers:

print(j)

print('---------------------')

Encoder layers:

<keras.layers.core.embedding.Embedding object at 0x782084b9d190>

<keras.layers.normalization.layer\_normalization.LayerNormalization object at 0x7820e56f1b90>

<keras.layers.rnn.lstm.LSTM object at 0x7820841bd650>

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Decoder layers:

<keras.layers.core.embedding.Embedding object at 0x78207c258590>

<keras.layers.normalization.layer\_normalization.LayerNormalization object at 0x78207c78bd10>

<keras.layers.rnn.lstm.LSTM object at 0x78207c258a10>

<keras.layers.core.dense.Dense object at 0x78207c2636d0>

---------------------

Create Inference Model

In [19]:

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],name='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,decoder\_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"

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Layer (type) Output Shape Param #

=================================================================

input\_1 (InputLayer) [(None, None)] 0

encoder\_embedding (Embeddin (None, None, 256) 625408

g)

layer\_normalization (LayerN (None, None, 256) 512

ormalization)

encoder\_lstm (LSTM) [(None, None, 256), 525312

(None, 256),

(None, 256)]

=================================================================

Total params: 1,151,232

Trainable params: 1,151,232

Non-trainable params: 0

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Model: "chatbot\_decoder"

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Layer (type) Output Shape Param # Connected to

==================================================================================================

input\_4 (InputLayer) [(None, None)] 0 []

decoder\_embedding (Embedding) (None, None, 256) 625408 ['input\_4[0][0]']

layer\_normalization (LayerNorm (None, None, 256) 512 ['decoder\_embedding[0][0]']

alization)

input\_2 (InputLayer) [(None, 256)] 0 []

input\_3 (InputLayer) [(None, 256)] 0 []

decoder\_lstm (LSTM) [(None, None, 256), 525312 ['layer\_normalization[1][0]',

(None, 256), 'input\_2[0][0]',

(None, 256)] 'input\_3[0][0]']

decoder\_dense (Dense) (None, None, 2443) 627851 ['decoder\_lstm[0][0]']

==================================================================================================

Total params: 1,779,083

Trainable params: 1,779,083

Non-trainable params: 0

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In [20]:

tf.keras.utils.plot\_model(chatbot.encoder,to\_file='encoder.png',show\_shapes=True,show\_layer\_activations=True)

Out[20]:

In [21]:

tf.keras.utils.plot\_model(chatbot.decoder,to\_file='decoder.png',show\_shapes=True,show\_layer\_activations=True)

Out[21]:

**Time to Chat:**

def print\_conversation(texts):

for text **in** texts:

print(f'You: **{**text**}**')

print(f'Bot: **{**chatbot(text)**}**')

print('=====================)

print\_conversation([‘hi,how are you doing?’,

‘how about your self?’,

‘what about you?’

‘what school do you go to?’

‘do you like it there?’

‘how’s it going?’

‘are you enjoying it there?’

‘how are you doing today?’])

**OUTPUT:**

**You:hi,how are you doing?**

**Bot:I am fine.**

**=============================================**

**You:how about your self?**

**Bot:I am pretty good.**

**==============================================**

**You:what about you?**

**Bot:I’ve been good.i’m in school right now.**

**===============================================**

**You:what school do you go to?**

**Bot:I go to pcc.**

**================================================**

**You:how’s it going?**

**Bot:I am doing well.**

**=================================================**

**You:do you like it there?**

**Bot:it’s ok.it’s a really big campus.**

**==================================================**

**You:are you enjoing it there?**

**Bot:it’s not bad.there are lot of people there.**

**===================================================**

**You:how are doing today?**

**Bot:I am doing great.**

**====================================================**

**CONCLUSION:**

In conclusion, building a chatbot in Python is a rewarding and educational experience that can expose you to a variety of useful skills, including natural language processing, machine learning, and software engineering. Chatbots can be used for a wide range of purposes, from customer service to education to entertainment.