Configuration Manual

1. System Requirements

RAM: 32GB OS: Windows and Google Colab Processor: i5 Intel GPU: 4 GB Nvidia, T4- Google Colab

2. Steps to work Here are the actions you need to take to manually run the image captioning model:

1. We gather information.

Gather any and all cricket-related visuals. Filter out incompatible graphics and download around 1000 photos.

Separate the cricket dataset into a training set of 800 photos and a test set of 200. Use an external dataset, such as Flickr8k, if you need a more substantial dataset.

2. Images are loaded and features are generated.

To use the InceptionV3 model, you'll need to load the photos and preprocess them to a specified size (for example, 299x299).

Extract 1x2048-dimensional vectors of picture features using a pretrained InceptionV3 model.

3. Textual pre-processing and image caption loading constitute this step.

Incorporate the captions into a dictionary that links the filenames of images with their descriptions.

Get rid of punctuation and capitalise just necessary words when processing text. Each caption has to have "captionstart" and "captionend" added so the model can read them.

4. Loading Data Consisting of Processed Images and Captions

Encode the sanitised captions by employing a tokenizer to convert the words into a series of integers.

Get the data ready for model fitting by splitting it into input and output pairs.

5. Model Fitting and Definition

An image feature extractor, sequence processor, and decoder make up the image captioning model, which you must define.

Train the model by giving it data in batches using a generator function.

Keep track of the several trained models over time and pick the best one according to how well it does on the development dataset.

6. Model Completion

Choose the best-performing model in an unexplored dataset (Flickr's training data for BLEU-1, cricket's test data for BLEU-2).

The completed model should be saved for later usage.

7. Assessing the Model

Put the model to the test with real data. Compute BLEU scores by comparing expected and reference captions.

8. Create Test or New Captions for Images

Bring up the tokenizer and trained model you made before.

Caption start word ("captionstart") and maximum length must be specified.

With each fresh or test picture:

>> Make use of InceptionV3 to produce picture features.

>> The trained model should be used to guess the next caption word until the "captionend" keyword is encountered.

>> To get the anticipated caption, just take off the first and last words.

Check the quality of the produced captions to see how well the model performed.

2. Dataset

First step is to find suitable dataset to build an image captioning problem. As we are more interested towards sport specific captioning, we will be choosing cricket domain related images. We have downloaded around 1000 images from internet via python code. For cricket data, as we have very small dataset, we will use 800 images as train data and 200 images as test dataset. Although some of the images from cricket data will be filtered out due to nonsupported visuals like gif files. In our final cricket captioning model, we will have total 720 images, out of which 575 will be used for model training and remaining 145 will be used for model evaluation (test data). There are many open datasets like Flickr8k dataset, Flickr30k dataset & Microsoft COCO dataset which contains 180k images. We want to build an image captioning model using normal hardware and building model using large image dataset is not feasible on normal laptops. Therefore, we will be using Flickr8k dataset for this model development. This data can be downloaded from Kaggle platform without any cost. This dataset is a benchmark collection of 8000 images from different domains with 5 captions per image. All these captions provide major content info of any image. This dataset has pre-defined 6000 images as training data, 1000 images as development data and remaining 1000 images a test data. A sample of Flickr image captioning is provided below:



Figure 1. Sample image from Flickr8k data

For above Fig 8, below are the 5 different captions provided by experts:

Caption1: a black dog is running after a white dog in the snow

Caption2: black dog chasing brown dog through snow

Caption3: two dogs chase each other across the snowy ground

Caption4: two dogs play together in the snow

Caption5: two dogs running through a low lying body of water

Cleaned descriptions of last step need to be encoded as numbers or each word should be assigned a number. This step will be helpful in upcoming sequence processing step. Keras provides a class named as Tokenizer which can learn the mapping for each word of a loaded description and assign a unique number to each word. Now we need to prepare the cleaned text data and image feature vectors as per model fitting requirements. We know that we cannot pass complete caption as target variable of image captioning model. We need to pass word-by-word. To perform that, we need to encode the captions. Each image caption will be divided into words. Model will be provided image feature vector and one word at a time and it will generate next word. Then first two words and image feature vectors will be given as inputs to generate next word. This will the process of model training. Below example (Fig 9 and Table 9) provides an easy to understand explanation of the training data preparation:



Figure 2. A sample image of cricket data

Caption Generated: A fielder is catching the ball

Table 1. Data preparation for image captioning training

<i>S.N</i> .	Image feature vector (Input1)	Word embeddings (Input2)	Output
1	Image_1	captionstart	a
2	Image_1	captionstart a	fielder
3	Image_1	captionstart a fielder	is
4	Image_1	captionstart a fielder is	catching
5	Image_1	captionstart a fielder is catching	the
6	Image_1	captionstart a fielder is catching the	ball
7	Image_1	captionstart a fielder is catching the ball	captionend

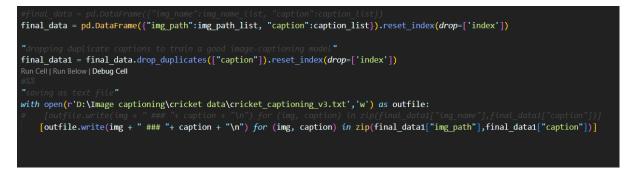
3.Code Snippets

import os
import pandas as pd
from pandas import ExcelWriter
import argparse



workbook.close()





"Finalizing the epoch with best trained model on test cricket data

import os
os.chdir(r"D:\Image captioning\coding\cricket")
import warnings
warnings.filterwarnings("ignore")

"Model evaluation on test images"

from pickle import load
from keras.preprocessing.sequence import pad_sequences
from keras.models import load_model
from nltk.translate.bleu_score import corpus_bleu
import numpy as np
from pickle import load, dump
import pandas as pd

map an integer to a word def word_for_id(integer, tokenizer): for word, index in tokenizer.word_index.items(): if index == integer: return word return None

```
# generate a description for an image
def generate_desc(model, tokenizer, photo, max_length):
    # seed the generation process
    in_text = 'captionstart'
    photo = photo.reshape(1,2048)
    # iterate over the whole length of the sequence
    for i in range(max_length):
        # integer encode input sequence
        sequence = tokenizer.texts_to_sequences([in_text])[0]
        # pad input
        sequence = pad_sequences([sequence], maxlen=max_length)
        # predict next word
        yhat = model.predict([photo,sequence], verbose=0)
        # convert probability to integer
        yhat = np.argmax(yhat)
        # map integer to word
        word = word_for_id(yhat, tokenizer)
        # stop if we cannot map the word
        if word is None:
            break
        # append as input for generating the next word
        in_text += ' ' + word
        # stop if we predict the end of the sequence
        if word == 'captionend':
            break
    return in_text
```

```
def evaluate_model(model, descriptions, photos, tokenizer, max_length):
    actual, predicted = list(), list()
    k=0
    1=0
    key list = []
    for key, desc_list in descriptions.items():
        k+=1
        print(k)
        try:
            yhat = generate_desc(model, tokenizer, photos[key], max_length)
            references = [d.split() for d in desc list]
            yhat1 = yhat.split()[1:-1]
            actual.append(references)
            predicted.append(yhat1)
            key list.append(key)
        except:
            l+=1
    print('BLEU-1: %f' % corpus bleu(actual, predicted, weights=(1.0, 0, 0, 0)))
    print('BLEU-2: %f' % corpus bleu(actual, predicted, weights=(0.5, 0.5, 0, 0)))
```

return actual, predicted, key_list

```
with open(r"D:\Image captioning\cricket data\cricket_captioning_test_v2.txt","r") as f:
    test_ids = f.read()
# prepare test image features
with open("image_features.pkl","rb") as f:
    image_extracted = load(f)
# prepare test image descriptions
with open("descriptions.pkl","rb") as f:
    descriptions = load(f)
```



```
"Checking the BLEU-performance on test data using the Finalized model in above step'

import os

os.chdir(r"D:\Image captioning\coding\cricket")

import warnings

warnings.filterwarnings("ignore")
```

"Model evaluation on test images'

from pickle import load

from keras.preprocessing.sequence import pad_sequences
from keras.models import load_model
from nltk.translate.bleu_score import corpus_bleu
import numpy as np
from pickle import load, dump
import pandas as pd

map an integer to a word

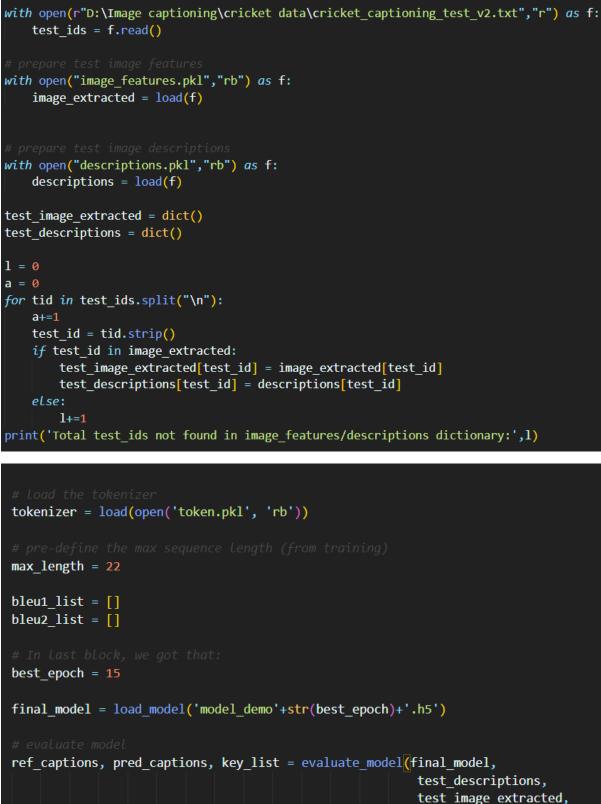
```
def word_for_id(integer, tokenizer):
    for word, index in tokenizer.word_index.items():
        if index == integer:
            return word
        return None
```

generate a description for an image

```
def generate_desc(model, tokenizer, photo, max_length):
    # seed the generation process
    in_text = 'captionstart'
    photo = photo.reshape(1,2048)

    # iterate over the whole length of the sequence
    for i in range(max_length):
        # integer encode input sequence
        sequence = tokenizer.texts_to_sequences([in_text])[0]
        # pad input
        sequence = pad_sequences([sequence], maxlen=max_length)
        # predict next word
        yhat = model.predict([photo,sequence], verbose=0)
        # convert probability to integer
        yhat = np.argmax(yhat)
        # map integer to word
        word = word_for_id(yhat, tokenizer)
        # stop if we cannot map the word
        if word is None:
            break
        # append as input for generating the next word
        in_text += ' ' + word
        # stop if we predict the end of the sequence
        if word == 'captionend':
            break
    return in_text
```

```
def evaluate_model(model, descriptions, photos, tokenizer, max_length):
   actual, predicted = list(), list()
   k=0
   1=0
   key_list = []
   for key, desc list in descriptions.items():
        k+=1
        print(k)
        try:
           yhat = generate_desc(model, tokenizer, photos[key], max_length)
            references = [d.split() for d in desc_list]
           yhat1 = yhat.split()[1:-1]
            actual.append(references)
            predicted.append(yhat1)
            key_list.append(key)
        except:
            l+=1
   print('BLEU-1: %f' % corpus_bleu(actual, predicted, weights=(1.0, 0, 0, 0)))
   print('BLEU-2: %f' % corpus_bleu(actual, predicted, weights=(0.5, 0.5, 0, 0)))
   return actual, predicted, key_list
```



tokenizer, max_length)

```
import cv2
import warnings
warnings.filterwarnings("ignore")
for i in range(min(20,len(key_list))):
    image_id = key_list[i]
    img = cv2.imread(r"{}".format(image_id))
    cv2.imshow(image_id, img)
    print(image_id)
    print('Ref. captions::')
    print([' '.join(cap1) for cap1 in ref_captions[i]])
    print('Predicted caption::')
    print([' '.join(pred_captions[i])])
    print('BLEU scoring::')
    print('BLEU-1: %f' % corpus_bleu([ref_captions[i]], [pred_captions[i]], weights=(1.0, 0, 0, 0)))
    print('BLEU-2: %f' % corpus bleu([ref captions[i]], [pred captions[i]], weights=(0.5, 0.5, 0, 0)))
    print("\n\n")
```

```
"Image captioning for new images"
import os
os.chdir(r"D:\Image captioning\coding\cricket")
import warnings
warnings.filterwarnings("ignore")
from pickle import load, dump
from numpy import argmax
from keras.preprocessing.sequence import pad_sequences
from keras.preprocessing.image import load img
```

```
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img_to_array
from keras.applications.inception_v3 import InceptionV3,preprocess_input
```

```
from keras.layers import Dense,BatchNormalization,Dropout,Embedding,RepeatVector
```

```
from keras.models import Sequential
from keras.models import Model
```

from keras.models import load model

import numpy as np

pre_trained_incept_v3 = load(open('pretrained_incep_v3.pkl', 'rb'))

```
def extract features(filename):
    TARGET_SIZE = (299, 299)
    img = load img(filename, target size=TARGET SIZE)
    img array = img to array(img)
    nimage = preprocess_input(img_array)
    nimage = np.expand dims(nimage, axis=0)
    fea vec = pre trained incept v3.predict(nimage)
    return fea vec
def word_for_id(integer, tokenizer):
    for word, index in tokenizer.word index.items():
        if index == integer:
            return word
    return None
def generate_desc(model, tokenizer, photo, max_length):
    in_text = 'captionstart'
    photo = photo.reshape(1,2048)
    for i in range(max_length):
        sequence = tokenizer.texts to sequences([in text])[0]
        sequence = pad_sequences([sequence], maxlen=max_length)
        yhat = model.predict([photo,sequence], verbose=0)
        yhat = np.argmax(yhat)
        word = word_for_id(yhat, tokenizer)
        if word is None:
            break
        in text += ' ' + word
        if word == 'captionend':
            break
    return in text
```

```
# load the tokenizer
tokenizer = load(open('token.pkl', 'rb'))
# pre-define the max sequence length (from training)
max_length = 22
# load the model
model = load_model('model_demo15.h5')
"prediction on new images"
img_folder = r'D:\Image captioning\ImageCaptioning\imgs'
imgs = os.listdir(img_folder)
for img in imgs:
    print(img)
    photo = extract_features(os.path.join(img_folder, img))
    description = generate_desc(model, tokenizer, photo, max_length)
    print(' '.join(description.split()[1:-1]))
    print('\n\n')
```

```
import os
os.chdir(r"D:\Image captioning\coding\cricket")
from keras.applications.inception_v3 import InceptionV3,preprocess_input
from keras.layers import Dense,BatchNormalization,Dropout,Embedding,RepeatVector
from keras.preprocessing.image import load_img, img_to_array
from keras.models import Sequential
from keras.models import Model
from pickle import dump, load
from keras.models import load model
import numpy as np
inception = InceptionV3(weights='imagenet')
inception.layers.pop()
for layer in inception.layers:
    layer.trainable = False
final_model = Model(input = inception.input,output = inception.layers[-1].output)
dump(final_model,open("pretrained_incep_v3.pkl","wb"))
```

```
import os
os.chdir(r"D:\Image captioning\coding\cricket")
from pickle import dump, load
from keras.applications.inception_v3 import InceptionV3,preprocess_input
from keras.layers import Dense,BatchNormalization,Dropout,Embedding,RepeatVector
from keras.preprocessing.image import load_img, img_to_array
import numpy as np
final_model =load(open('pretrained_incep_v3.pkl', 'rb'))
TARGET SIZE = (299,299) # needed to convert the image as per pre-trained inceptionv3 requirements
image_extracted = dict()
with open(r"D:\Image captioning\cricket data\cricket_captioning_v2.txt") as f:
    data = f.read()
data1 = data.split("\n")
id_list = []
for el in data1:
    id1 = el.split("###")[0].strip()
    id_list+=[id1]
id_list1 = list(set(id_list))
```

```
k=0
1=0
for image_id in id_list1:
    if k%10==0:
        print(k)
    try:
        img = load_img(r"{}".format(image_id),target_size=TARGET_SIZE)
        img_array = img_to_array(img)
        nimage = preprocess_input(img_array)
        nimage = np.expand_dims(nimage, axis=0)
        fea vec = final model.predict(nimage)
        image_extracted[image_id] = np.reshape(fea_vec, fea_vec.shape[1]) # reshape from (1, 2048) to (2048
    except Exception as e:
        1+=1
        print("Total Exception got :- \n",1)
```

```
dump(image_extracted,open("image_features.pkl","wb"))
train_image_extracted = dict()
with open(r"D:\Image captioning\cricket data\cricket_captioning_train_v2.txt","r") as f:
    train_ids = f.read()
l=0
for tid in train_ids.split("\n"):
    tid = tid.strip()
    if tid in image_extracted:
        train_image_extracted[tid] = image_extracted[tid]
    else:
        l+=1
```

```
print('Total train_ids not found in image_features dictionary:',1)
# save the file
dump(train_image_extracted,open("train_image_features.pkl","wb"))
```

```
train_image_extracted = dict()
with open(r"D:\Image captioning\cricket data\cricket_captioning_train_v2.txt","r") as f:
    train_ids = f.read()
l=0
for tid in train_ids.split("\n"):
    tid = tid.strip()
    if tid in image_extracted:
        train_image_extracted[tid] = image_extracted[tid]
    else:
        l+=1
print('Total train_ids not found in image_features dictionary:',l)
# save the file
dump(train_image_extracted,open("train_image_features.pkl","wb"))
```

```
import os
os.chdir(r"D:\Image captioning\coding\cricket")
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from pickle import dump, load
import string
descriptions = dict()
with open(r"D:\Image captioning\cricket data\cricket captioning v2.txt") as f:
    data = f.read()
for el in data.split("\n"):
    try:
        tokens = el.split("###")
        image_id , image_desc = tokens[0],tokens[1]
        image id = image id.strip()
        image desc = tokens[1].strip()
        if image id in descriptions:
            descriptions[image id].append(image desc)
```

else:

except Exception as e:

descriptions[image_id] = list()

print("Exception got :- \n",e)

descriptions[image_id].append(image_desc)



clean all the captions
descriptions = clean_descriptions(descriptions)
save the file for further use
dump(descriptions,open("descriptions.pkl","wb"))
prepare test image descriptions
with open("descriptions.pkl","rb") as f:
 descriptions = load(f)
#find pre-processed train captions from descriptions file and perform the encoding as per
train_descriptions = dict()
with open(r"D:\Image captioning\cricket data\cricket_captioning_train_v2.txt","r") as f:

```
train_file = f.read()
```

```
1=0
for tid in train file.split("\n"):
    train id = tid.strip()
    if train_id in descriptions.copy():
         train_descriptions[train_id] = descriptions.copy()[train_id]
    else:
         1+=1
print('Total train ids not found in descriptions:',1)
for key, train desc list in train descriptions.items():
    for i in range(len(train_desc_list)):
        train desc = train desc list[i]
         train_desc_list[i] = 'captionstart ' + ''.join(train_desc) + ' captionend'
dump(train descriptions,open("train descriptions.pkl","wb"))
Run Cell | Run Above | Debug Cell
import os
os.chdir(r"D:\Image captioning\coding\cricket")
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.sequence import pad_sequences
from keras.utils.np utils import to categorical
from nltk import FreqDist
corpus = ""
all_train_captions = []
with open("train_descriptions.pkl","rb") as f:
    train descriptions = load(f)
for ec in train descriptions.values():
    for el in ec:
       corpus += " "+el
       all_train_captions.append(el)
```

```
total_words = corpus.split()
```

```
#vocabulary = set(total_words)
#print("The size of vocablury is {}".format(len(vocabulary)))
```

creating frequecny distribution of words

freq_dist = FreqDist(total_words)
freq_dist.most_common(5)

convert a dictionary of clean descriptions to a list of descriptions
def to_lines(descriptions):

all_desc = list()
for key in descriptions.keys():
 [all_desc.append(d) for d in descriptions[key]]
return all_desc

calculate the length of the description with the most words

print('preprocessed words %d -> %d' % (len(word_counts), len(vocab)))

```
def max_length(descriptions):
    lines = to_lines(descriptions)
    return max(len(d.split()) for d in lines)
```

```
# determine the maximum sequence length
max_length = max_length(train_descriptions)
```

```
print('Description Length: %d' % max_length)
```

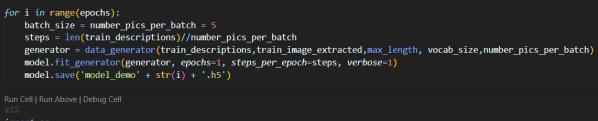
```
# Consider only words which occur at least 'word_count_threshold ' times in the corpus
word_count_threshold = 2
word_counts = {}
nsents = 0
for sent in all_train_captions:
    nsents += 1
    for w in sent.split(' '):
        word_counts[w] = word_counts.get(w, 0) + 1
vocab = [w for w in word_counts if word_counts[w] >= word_count_threshold]
```

```
ix = 1
for w in vocab:
   wordtoix[w] = ix
    ixtoword[ix] = w
   ix += 1
vocab size = len(ixtoword) + 1
print('vocab_size %d' % vocab_size)
token = Tokenizer(num words=vocab size)
token.fit_on_texts(all_train_captions)
with open("token.pkl","wb") as f:
    dump(token,f)
glove_dir = 'D:\Image captioning\glove.6B'
embeddings index = {} # empty dictionary
f = open(os.path.join(glove_dir, 'glove.6B.300d.txt'), encoding="utf-8")
k=0
for line in f:
    if k%10000==0:
        print(k)
    k+=1
    values = line.split()
    word = values[0]
    coefs = np.asarray(values[1:], dtype='float32')
    embeddings_index[word] = coefs
f.close()
print('Found %s word vectors.' % len(embeddings_index))
embedding dim = 300
embedding_matrix = np.zeros((vocab_size, embedding_dim))
for word, i in wordtoix.items():
    embedding vector = embeddings index.get(word)
    if embedding_vector is not None:
        embedding matrix[i] = embedding vector
from pickle import dump
dump(embedding matrix, open("embedding matrix.pkl","wb"))
```

```
from keras.models import Model,Input
from keras.applications.inception_v3 import InceptionV3,preprocess_input
from keras.layers import Embedding,Dense,BatchNormalization,Dropout,LSTM,add
from keras.utils import plot_model
import numpy as np
def combined_model(MAX_LENGTH,VOCAB_SIZE):
    "modeL parameters"
    # standard variables
# MAX_LENGTH = 39
# VOCAB_SIZE = 1969 # due to embedding_matrix shape
# NPIX = 299 # required image shape for pre-trained inceptionnv3 model
# TARGET_SIZE = (NPIX,NPIX,3)
EMBEDDING_SIZE = 300
#EMBEDDING_SIZE = 50
# partial caption sequence model
inputs2 = Input(shape=(MAX_LENGTH,))
se1 = Embedding(VOCAB_SIZE, EMBEDDING_SIZE, mask_zero=True)(inputs2)
se2 = Dropout(0.5)(se1)
se3 = LSTM(EMBEDDING_SIZE) (se2) # 300-dim glove vector
```

```
# image feature extractor model
inputs1 = Input(shape=(2048,)) # iceptionnv3
fe1 = Dropout(0.5)(inputs1)
fe2 = Dense(EMBEDDING_SIZE, activation='relu')(fe1)# 300-dim glove vector
#fe2 = Dense(50, activation='relu')(fe1)# 50-dim glove vector
#fe2 = Dense(50, activation='relu')(fe1)# 50-dim glove vector
#decoder1 = add([fe2, se3])
decoder2 = Dense(EMBEDDING_SIZE, activation='relu')(decoder1) # 300-dim glove vector
#decoder2 = Dense(50, activation='relu')(decoder1) # 300-dim glove vector
outputs = Dense(VOCAB_SIZE, activation='softmax')(decoder2)
# merge the two input models
# image_feature + partial caption ===> output
model = Model(inputs=[inputs1, inputs2], outputs=outputs)
# setting wight of embedded matrix that we saved earlier for words
with open("embedding_matrix.pkl", "rb") as f:
    embedding_matrix = load(f)
model.layers[2].set_weights([embedding_matrix])
model.layers[2].trainable = False
model.compile(loss='categorical_crossentropy', optimizer='adam')
return model
```

```
def data_generator(descriptions, photos, MAX_LENGTH,VOCAB_SIZE, num_photos_per_batch):
    X1, X2, y = list(), list(), list()
    n=0
    1=0
    for key, desc_list in train descriptions.items():
        try:
           n+=1
           photo = photos[key]
           for desc in desc_list:
               seq = token.texts_to_sequences([desc])
               seq = seq[0]
               for i in range(1,len(seq)):
                   in_seq , op_seq = seq[:i],seq[i]
                   in_seq = pad_sequences([in_seq],maxlen=MAX_LENGTH,padding="post")[0]
                   op_seq = to_categorical([op_seq],num_classes=VOCAB_SIZE)[0]
                   X1.append(photo)
                   X2.append(in seq)
                   y.append(op_seq)
        except:
           l+=1
        if n==num photos per batch:
           yield [[np.array(X1), np.array(X2)], np.array(y)]
           X1, X2, y = list(), list(), list()
           n=0
 with open("train image features.pkl", "rb") as f:
      train image_extracted = load(f)
 with open("train descriptions.pkl", "rb") as f:
      train descriptions = load(f)
 max length = max length
 vocab_size = vocab_size
 model = combined model(max length, vocab size)#
 epochs = 20
```



import os os.chdir(r"D:\Image captioning\coding\cricket")

```
flickr_trloss = pd.read_csv("train_loss_cricket.csv")
 import matplotlib.pyplot as plt
plt.figure(figsize=[12,6])
plt.plot(flickr_trloss['Epoch'], flickr_trloss['train_loss'], 'o-', label='train_loss')
plt.legend()
plt.Xlabel('Epoch', fontweight='bold', fontsize=12)
plt.ylabel('Train_loss', fontweight='bold', fontsize=12)
plt.title("Epoch vs train loss (circket)", fontweight='bold', fontsize=14)
```

import random

```
with open(r"D:\Image captioning\cricket data\cricket_captioning_v2.txt") as f:
    data = f.read()
data1 = data.split("\n")
id_list = []
for el in data1:
    id1 = el.split("###")[0].strip()
    id list+=[id1]
id_list1 = list(set(id_list))
random.shuffle(id_list1)
train_samples = int(0.8*len(id_list1))
train_data = id_list1[:train_samples]
test_data = id_list1[train_samples:]
with open(r"D:\Image captioning\cricket data\cricket captioning train v2.txt","w") as f:
    [f.write(x+"\n") for x in train_data]
with open(r"D:\Image captioning\cricket data\cricket_captioning_test_v2.txt","w") as f:
    [f.write(x+"\n") for x in test_data]
```

"Finalizing the epoch with best trained model on devlopment cricke import os os.chdir(r"D:\Image captioning\coding\flickr") "Model evaluation on test images" from pickle import load from keras.preprocessing.sequence import pad_sequences from keras.models import load_model from nltk.translate.bleu_score import corpus_bleu, sentence_bleu import numpy as np from pickle import load, dump import pandas as pd import warnings warnings.filterwarnings("ignore")

map an integer to a word

def word_for_id(integer, tokenizer):
 for word, index in tokenizer.word_index.items():
 if index == integer:
 return word
 return None

```
# generate a description for an image
def generate_desc(model, tokenizer, photo, max_length):
    # seed the generation process
    in_text = 'captionstart'
    photo = photo.reshape(1,2048)
    # iterate over the whole length of the sequence
    for i in range(max_length):
        # integer encode input sequence
        sequence = tokenizer.texts_to_sequences([in_text])[0]
        # pad input
        sequence = pad_sequences([sequence], maxlen=max_length)
        # predict next word
        yhat = model.predict([photo,sequence], verbose=0)
        # convert probability to integer
        yhat = np.argmax(yhat)
        # map integer to word
        word = word_for_id(yhat, tokenizer)
        # stop if we cannot map the word
        if word is None:
            break
        # append as input for generating the next word
        in_text += ' ' + word
        # stop if we predict the end of the sequence
        if word == 'captionend':
            break
        return in_text
```

```
def evaluate_model(model, descriptions, photos, tokenizer, max_length):
    actual, predicted = list(), list()
    k=0
    1=0
    key_list = []
    for key, desc_list in descriptions.items():
        k+=1
        print(k)
        try:
            yhat = generate_desc(model, tokenizer, photos[key], max_length)
            references = [d.split() for d in desc_list]
            yhat1 = yhat.split()[1:-1]
            actual.append(references)
            predicted.append(yhat1)
            key_list.append(key)
        except:
            1+=1
```

```
print('BLEU-1: %f' % corpus_bleu(actual, predicted, weights=(1.0, 0, 0, 0)))
    print('BLEU-2: %f' % corpus_bleu(actual, predicted,
                                     weights = (0.5, 0.5, 0, 0))
    return actual, predicted, key list
with open(r"D:\Image captioning\145129 343604 bundle archive\Flickr Data\
          Flickr_Data\Flickr_TextData\Flickr_8k.devImages.txt", "r") as f:
    dev_ids = f.read()
with open("image_features.pkl","rb") as f:
    image_extracted = load(f)
with open("descriptions.pkl","rb") as f:
    descriptions = load(f)
dev_image_extracted = dict()
dev descriptions = dict()
1 = 0
a = 0
for did in dev_ids.split("\n"):
    a+=1
    if a>100:
        break
    dev id = did.strip()
    if dev id in image extracted:
        dev_image_extracted[dev_id] = image_extracted[dev_id]
        dev descriptions[dev id] = descriptions[dev id]
```

else: 1+=1			
<pre>print('Total dev_ids not found in image_features/descriptions dictionary:',1)</pre>			
# save the files			
<pre>dump(dev_image_extracted,open("dev_image_features.pkl","wb"))</pre>			
<pre>dump(dev descriptions, open("dev descriptions.pkl", "wb"))</pre>			
<pre>(function) def print(*values: object,</pre>			
# lo sep: str None = " ",			
toke end: str None = "\n",			
<pre>file: SupportsWrite[str] None = None,</pre>			
<pre># pr flush: Literal[False] = False</pre>			
<pre>max_) -> None</pre>			
bleu Prints the values to a stream, or to sys.stdout by default.			
bleu sep			
string inserted between values, default a space.			
# fi end			
<i>for</i> string appended after the last value, default a newline.			
<pre>print("Epoch:%d"%i)</pre>			
<pre>model = load_model('model_demo'+str(i)+'.h5')</pre>			

```
ref captions, pred captions, key list =
    evaluate_model(model, dev_descriptions, dev_image_extracted,
                   tokenizer, max length)
    bleu1_list +=[corpus_bleu(ref_captions,
                               pred captions, weights=(1.0, 0, 0, 0)]
    bleu2 list += corpus bleu(ref captions,
                              pred_captions, weights=(0.5, 0.5, 0, 0))]
flickr_acc = pd.DataFrame({"Epoch":list(range(1,21)),
                           "BLEU-1":bleu1 list,"BLEU-2":bleu2 list})
flickr acc.to csv("epoch vs dev data bleu scoring.csv")
import matplotlib.pyplot as plt
plt.figure(figsize=[12,6])
plt.plot(flickr_acc['Epoch'], flickr_acc['BLEU-1'], 'o-', label='BLEU-1')
plt.plot(flickr_acc['Epoch'], flickr_acc['BLEU-2'], 'o-', label='BLEU-2')
plt.legend()
plt.xlabel('Epoch', fontweight='bold', fontsize=12)
plt.ylabel('BLEU score', fontweight='bold', fontsize=12)
plt.title("Epoch vs BLEU-score on development data",
          fontweight='bold', fontsize=14)
```

```
import os
os.chdir(r"D:\Image captioning\coding\flickr")
import warnings
warnings.filterwarnings("ignore")
from pickle import load
from keras.preprocessing.sequence import pad sequences
from keras.models import load model
from nltk.translate.bleu score import corpus bleu
import numpy as np
from pickle import load, dump
import pandas as pd
def word_for_id(integer, tokenizer):
   for word, index in tokenizer.word index.items():
        if index == integer:
            return word
    return None
def generate_desc(model, tokenizer, photo, max_length):
    in_text = 'captionstart'
    photo = photo.reshape(1,2048)
```

```
# iterate over the whole length of the sequence
for i in range(max_length):
    # integer encode input sequence
    sequence = tokenizer.texts_to_sequences([in_text])[0]
    # pad input
    sequence = pad_sequences([sequence], maxlen=max_length)
    # predict next word
    yhat = model.predict([photo,sequence], verbose=0)
    # convert probability to integer
    yhat = np.argmax(yhat)
    # map integer to word
    word = word_for_id(yhat, tokenizer)
    # stop if we cannot map the word
    if word is None:
        break
    # append as input for generating the next word
    in_text += ' ' + word
    # stop if we predict the end of the sequence
    if word == 'captionend':
        break
return in_text
```

```
def evaluate_model(model, descriptions, photos, tokenizer, max_length):
    actual, predicted = list(), list()
    k=0
    1=0
    key list = []
    for key, desc list in descriptions.items():
        k+=1
        print(k)
        try:
            yhat = generate_desc(model, tokenizer, photos[key], max_length)
            references = [d.split() for d in desc_list]
            yhat1 = yhat.split()[1:-1]
            actual.append(references)
            predicted.append(yhat1)
            key_list.append(key)
        except:
            l+=1
```

```
print('BLEU-1: %f' % corpus_bleu(actual,
                                     predicted, weights=(1.0, 0, 0, 0)))
    print('BLEU-2: %f' % corpus_bleu(actual,
                                      predicted, weights=(0.5, 0.5, 0, 0)))
    return actual, predicted, key_list
with open(
     r"D:\Image captioning\145129 343604 bundle archive\
        Flickr_Data\Flickr_Data\Flickr_TextData\Flickr_8k.testImages.txt","r")as f:
    test_ids = f.read()
with open("image_features.pkl","rb") as f:
    image_extracted = load(f)
with open("descriptions.pkl", "rb") as f:
    descriptions = load(f)
test_image_extracted = dict()
test_descriptions = dict()
1 = 0
a = 0
for tid in test_ids.split("\n"):
    a+=1
```

```
test_id = tid.strip()
      if test id in image extracted:
           test image extracted[test id] = image extracted[test id]
           test descriptions[test id] = descriptions[test id]
     else:
           1+=1
print('Total test ids not found in image features/descriptions dictionary:',1)
dump(test_image_extracted,open("test_image_features.pkl","wb"))
dump(test descriptions, open("test descriptions.pkl", "wb"))
tokenizer = load(open('token.pkl', 'rb'))
max length = 34
bleu1 list = []
bleu2 list = []
best_epoch = 5 #(16th epoch)
final model = load model('model demo'+str(best epoch)+'.h5')
ref_captions, pred_captions, key_list = evaluate_model(final_model, test_descriptions, test_image_extracted, tokenizer, max_length)
import os
os.chdir(r"D:\Image captioning\coding\flickr")
import cv2
import warnings
warnings.filterwarnings("ignore")
for i in range(8,min(16,len(key_list))):
    image_id = key_list[i]
    img = cv2.imread(r"D:\Image captioning\145129_343604_bundle_archive\Flickr_Data\Flickr_Data\Images\{}".format(image_id))
    cv2.imshow(image_id, img)
   print(image_id)
   print('Ref. captions::')
   print([' '.join(cap1) for cap1 in ref_captions[i]])
   print('Predicted caption::')
   print([' '.join(pred_captions[i])])
   print('BLEU scoring::')
   print('BLEU-1: %f' % corpus_bleu([ref_captions[i]], [pred_captions[i]], weights=(1.0, 0, 0, 0)))
print('BLEU-2: %f' % corpus_bleu([ref_captions[i]], [pred_captions[i]], weights=(0.5, 0.5, 0, 0)))
   print("\n\n")
```

```
import os
os.chdir(r"D:\Image captioning\coding\flickr")
from pickle import load, dump
from numpy import argmax
from keras.preprocessing.sequence import pad sequences
from keras.preprocessing.image import load img
from keras.preprocessing.image import img to array
from keras.applications.inception v3 import InceptionV3,preprocess input
from keras.layers import Dense, BatchNormalization, Dropout, Embedding, RepeatVector
from keras.models import Sequential
from keras.models import Model
from keras.models import load_model
import numpy as np
pre trained incept v3 = load(open('pretrained incep v3.pkl', 'rb'))
def extract_features(filename):
    TARGET_SIZE = (299, 299)
    img = load_img(filename,target_size=TARGET_SIZE)
    img_array = img_to_array(img)
    nimage = preprocess_input(img_array)
    nimage = np.expand dims(nimage, axis=0)
    fea vec = pre trained incept v3.predict(nimage)
    return fea_vec
```

```
def word for id(integer, tokenizer):
   for word, index in tokenizer.word index.items():
        if index == integer:
            return word
    return None
def generate_desc(model, tokenizer, photo, max_length):
   in text = 'captionstart'
   photo = photo.reshape(1,2048)
   for i in range(max length):
        sequence = tokenizer.texts_to_sequences([in_text])[0]
        sequence = pad sequences([sequence], maxlen=max length)
        yhat = model.predict([photo,sequence], verbose=0)
        yhat = np.argmax(yhat)
        word = word for id(yhat, tokenizer)
        if word is None:
            break
        in_text += ' ' + word
        if word == 'captionend':
            break
    return in_text
```

```
# load the tokenizer
tokenizer = load(open('token.pkl', 'rb'))
# pre-define the max sequence length (from training)
max_length = 34
# load the model
model = load_model('model_demo5.h5')
"prediction on new images"
img_folder = r'D:\Image captioning\ImageCaptioning\imgs'
imgs = os.listdir(img_folder)
for img in imgs:
    print(img)
    photo = extract_features(os.path.join(img_folder, img))
    description = generate_desc(model, tokenizer, photo, max_length)
    print(' '.join(description.split()[1:-1]))
    print('\n\n')
```

```
# Load the tokenizer
tokenizer = load(open('token.pkl', 'rb'))
```

```
# pre-define the max sequence length (from training)
max_length = 34
```

```
model = load model('model demo5.h5')
```

```
"prediction on new images"
```

```
img_folder = r'D:\Image captioning\ImageCaptioning\imgs'
imgs = os.listdir(img_folder)
```

```
for img in imgs:
    print(img)
    photo = extract_features(os.path.join(img_folder, img))
    description = generate_desc(model, tokenizer, photo, max_length)
    print(' '.join(description.split()[1:-1]))
    print('\n\n')
```

"Setting working directory"

import <mark>os</mark>

os.chdir(r"D:\Image captioning\coding\flickr")

Importing necessary modules

from keras.applications.inception_v3 import InceptionV3,preprocess_input
from keras.layers import Dense,BatchNormalization,Dropout,Embedding,RepeatVector
from keras.preprocessing.image import load_img, img_to_array
from keras.models import Sequential
from keras.models import Model
from pickle import dump, load
from keras.models import load_model
import numpy as np

Since we are using this as feature extractor, the last softmax layer is not useful inception = InceptionV3(weights='imagenet')

pop the last softmax layer and freezing the remaining layers (re-structure the mode inception.layers.pop()

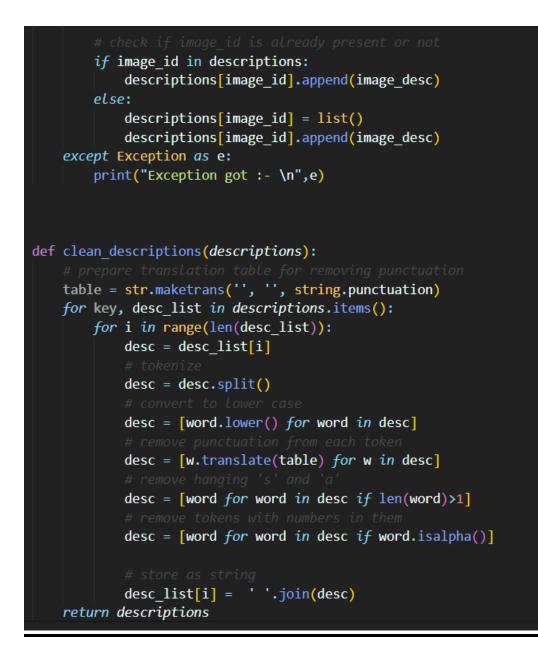
for layer in inception.layers:
 layer.trainable = False

building the final model
final_model = Model(input = inception.input,output = inception.layers[-1].output)

save the model to get image features from pre-trained inceptionnv3 model dump(final_model,open("pretrained_incep_v3.pkl","wb"))



```
img_array = img_to_array(img)
nimage = preprocess_input(img_array)
       nimage = np.expand_dims(nimage, axis=0)
fea_vec = final_model.predict(nimage)
       image_extracted[image_id] = np.reshape(fea_vec, fea_vec.shape[1]) # reshape from (1, 2048) to (2048, )
    except Exception as e:
       print("Total Exception got :- \n",1)
dump(image_extracted,open("image_features.pkl","wb"))
train_image_extracted = dict()
with open(r"D:\Image captioning\145129_343604_bundle_archive\Flickr_Data\Flickr_Data\Flickr_TextData\Flickr_8k.trainImages.txt", "r") as f:
   train_ids = f.read()
for tid in train_ids.split("\n"):
    if tid in image_extracted:
       train_image_extracted[tid] = image_extracted[tid]
print('Total train_ids not found in image_features dictionary:',1)
dump(train_image_extracted,open("train_image_features.pkl","wb"))
 import os
os.chdir(r"D:\Image captioning\coding\flickr")
import numpy as np
import pandas as pd
 import matplotlib.pyplot as plt
from pickle import dump, load
import string
PATH = r"D:\Image captioning\145129_343604_bundle_archive\Flickr_Data\Flickr_Data\Flickr_TextData"
descriptions = dict()
with open(PATH+r"\Flickr8k.token.txt") as f:
     data = f.read()
for el in data.split("\n"):
     try:
          tokens = el.split("#")
          image_id , image_desc = tokens[0],tokens[1:]
          image_id = image_id.strip()
          image_desc = " ".join(image_desc)
```



```
descriptions = clean_descriptions(descriptions)
dump(descriptions,open("descriptions.pkl","wb"))
with open("descriptions.pkl","rb") as f:
   descriptions = load(f)
train descriptions = dict()
with open(PATH+r"\Flickr 8k.trainImages.txt","r") as f:
    train file = f.read()
k=0
1=0
for tid in train_file.split("\n"):
    train id = tid.strip()
    if train id in descriptions.copy():
       train_descriptions[train_id] = descriptions.copy()[train_id]
   else:
       1+=1
print('Total train ids not found in descriptions:',1)
for key, train desc list in train descriptions.items():
    for i in range(len(train_desc_list)):
        train desc = train desc list[i]
        train desc list[i] = 'captionstart ' + ''.join(train_desc) + ' captionend'
dump(train descriptions,open("train descriptions.pkl","wb"))
Run Cell | Run Above | Debug Cell
import os
os.chdir(r"D:\Image captioning\coding\flickr")
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.sequence import pad sequences
from keras.utils.np_utils import to_categorical
```

```
# Vocablury Preparation
from nltk import FreqDist
```

```
# creating corpus
corpus = ""
```

Create a list of all the training captions

```
all_train_captions = []
with open("train_descriptions.pkl","rb") as f:
    train_descriptions = load(f)
for ec in train_descriptions.values():
    for el in ec:
        corpus += " "+el
        all_train_captions.append(el)
```

```
total_words = corpus.split()
#vocabulary = set(total_words)
#print("The size of vocablury is {}".format(len(vocabulary)))
```

creating frequecny distribution of words
freq dist = FreqDist(total words)

freq_dist.most_common(5)

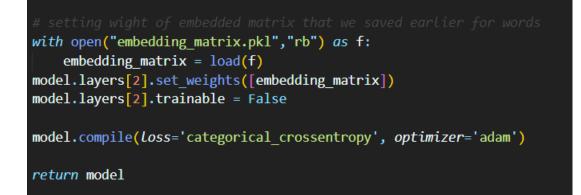
```
# convert a dictionary of clean descriptions to a list of descriptions
def to_lines(descriptions):
    all_desc = list()
    for key in descriptions.keys():
        [all_desc.append(d) for d in descriptions[key]]
    return all desc
```

```
def to_lines(descriptions):
   all_desc = list()
   for key in descriptions.keys():
        [all desc.append(d) for d in descriptions[key]]
    return all desc
def max length(descriptions):
    lines = to_lines(descriptions)
    return max(len(d.split()) for d in lines)
max_length = max_length(train_descriptions)
print('Description Length: %d' % max_length)
word count threshold = 5
word counts = \{\}
nsents = 0
for sent in all_train_captions:
   nsents += 1
   for w in sent.split(' '):
        word_counts[w] = word_counts.get(w, 0) + 1
vocab = [w for w in word_counts if word_counts[w] >= word_count_threshold]
print('preprocessed words %d -> %d' % (len(word_counts), len(vocab)))
ixtoword = {}
wordtoix = {}
```

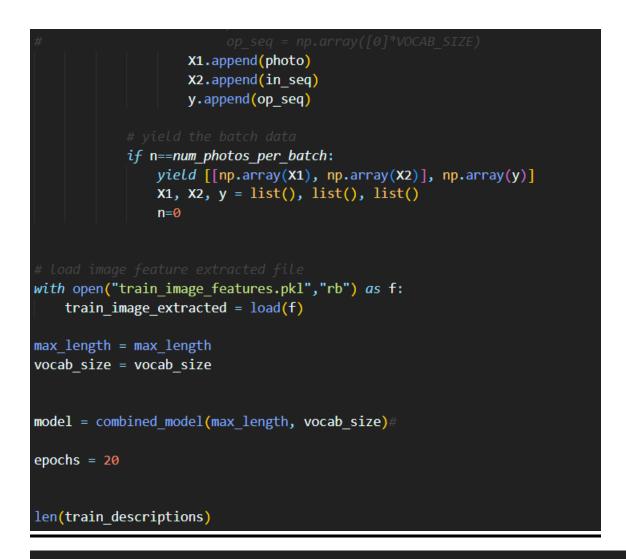
```
ix = 1
for w in vocab:
    wordtoix[w] = ix
    ixtoword[ix] = w
    ix += 1
vocab_size = len(ixtoword) + 1
token = Tokenizer(num words=vocab size)
token.fit_on_texts(all_train_captions)
with open("token.pkl","wb") as f:
    dump(token,f)
glove_dir = 'D:\Image captioning\glove.6B'
embeddings_index = {} # empty dictionary
f = open(os.path.join(glove dir, 'glove.6B.300d.txt'), encoding="utf-8")
k=0
for line in f:
    if k%1000==0:
        print(k)
    k+=1
    values = line.split()
    word = values[0]
    coefs = np.asarray(values[1:], dtype='float32')
    embeddings_index[word] = coefs
f.close()
print('Found %s word vectors.' % len(embeddings_index))
```

```
embedding dim = 300
embedding matrix = np.zeros((vocab size, embedding dim))
for word, i in wordtoix.items():
    embedding vector = embeddings index.get(word)
    if embedding vector is not None:
        embedding_matrix[i] = embedding_vector
from pickle import dump
dump(embedding matrix, open("embedding_matrix.pkl","wb"))
Run Cell | Run Above | Debug Cell
import os
os.chdir(r"D:\Image captioning\coding\flickr")
from keras.models import Model,Input
from keras.applications.inception_v3 import InceptionV3,preprocess input
from keras.layers import Embedding,Dense,BatchNormalization,Dropout,LSTM,add
from keras.utils import plot model
import numpy as np
from pickle import load, dump
```

```
from keras.models import Model,Input
from keras.applications.inception v3 import InceptionV3, preprocess input
from keras.layers import Embedding,Dense,BatchNormalization,Dropout,LSTM,add
from keras.utils import plot_model
import numpy as np
from pickle import load,dump
def combined model(MAX LENGTH, VOCAB SIZE):
    EMBEDDING SIZE = 300
    inputs2 = Input(shape=(MAX LENGTH,))
    se1 = Embedding(VOCAB SIZE, EMBEDDING SIZE, mask zero=True)(inputs2)
    se2 = Dropout(0.5)(se1)
    se3 = LSTM(EMBEDDING SIZE)(se2) # 300-dim glove vector
   inputs1 = Input(shape=(2048,)) # iceptionnv3
   fe1 = Dropout(0.5)(inputs1)
   fe2 = Dense(EMBEDDING SIZE, activation='relu')(fe1)# 300-dim glove
   decoder1 = add([fe2, se3])
   decoder2 = Dense(EMBEDDING_SIZE, activation='relu')(decoder1) # 30
   outputs = Dense(VOCAB SIZE, activation='softmax')(decoder2)
   model = Model(inputs=[inputs1, inputs2], outputs=outputs)
```



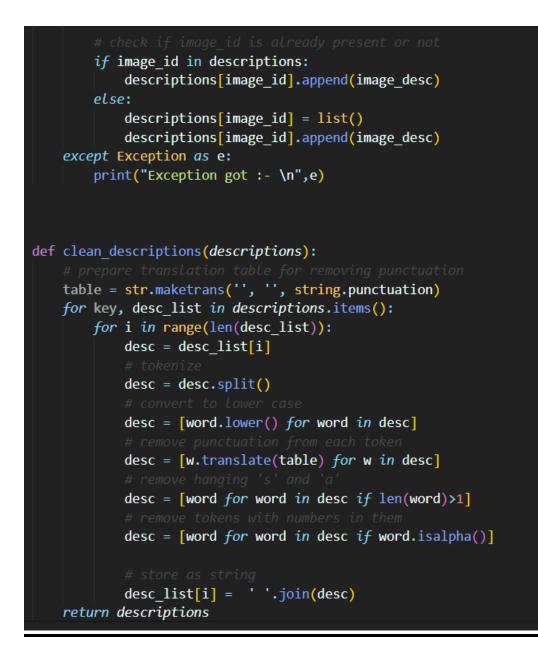




for i in range(epochs):
 batch_size = number_pics_per_batch = 5
 steps = len(train_descriptions)//number_pics_per_batch
 generator = data_generator(train_descriptions, train_image_extracted,max_length, vocab_size,number_pics_per_batch)
 model.fit_generator(generator, epochs=1, steps_per_epoch=steps, verbose=1)
 model.save('model_demo' + str(i) + '.h5')

Run Cell Run Above[Debug Cell
#%%
import os
os.chdir(r"D:\Image captioning\coding\flickr")
"MonualLy saving the training loss for each epoch and plotting here"
flickr_trloss = pd.read_csv("train_loss_flickr.csv")
import matplotlib.pyplot as plt
plt.figure(figsize=[12,6])
plt.plot(flickr_trloss['Epoch'], flickr_trloss['train_loss'], 'o-', LabeL='train_loss')
plt.legend()
plt.xlabel('Epoch', fontweight='bold', fontsize=12)
plt.title("Epoch vs train loss", fontweight='bold', fontsize=14)

```
descriptions = clean descriptions(descriptions)
dump(descriptions,open("descriptions.pkl","wb"))
with open("descriptions.pkl","rb") as f:
    descriptions = load(f)
train descriptions = dict()
with open(PATH+r"\Flickr_8k.trainImages.txt","r") as f:
    train_file = f.read()
k=0
1=0
for tid in train file.split("\n"):
    train id = tid.strip()
    if train_id in descriptions.copy():
        train descriptions[train id] = descriptions.copy()[train id]
    else:
        1+=1
print('Total train ids not found in descriptions:',1)
for key, train_desc_list in train_descriptions.items():
    for i in range(len(train desc list)):
        train desc = train desc list[i]
        train_desc_list[i] = 'captionstart ' + ''.join(train_desc) + ' captionend'
```



from numpy import array from pickle import load from keras.preprocessing.text import Tokenizer from keras.preprocessing.sequence import pad sequences from keras.utils import to_categorical from keras.utils import plot model from keras.models import Model from keras.layers import Input from keras.layers import Dense from keras.layers import LSTM from keras.layers import Bidirectional from keras.layers import Embedding from keras.layers import Dropout from keras.layers.merge import add from keras.callbacks import ModelCheckpoint from os import listdir from pickle import dump *import* string from keras.applications.inception v3 import InceptionV3 from keras.preprocessing.image import load_img from keras.preprocessing.image import img to array

from keras.applications.inception v3 import preprocess input

```
def extract_features(directory):
   model = InceptionV3()
   model.layers.pop()
   model = Model(inputs=model.inputs, outputs=model.layers[-1].output)
   print(model.summary())
   features = dict()
   for name in listdir(directory):
       filename = directory + '/' + name
       image = load_img(filename, target_size=(224, 224))
        image = img_to_array(image)
       image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
        image = preprocess_input(image)
       feature = model.predict(image, verbose=0)
        image_id = name.split('.')[0]
        features[image_id] = feature
       print('>%s' % name)
    return features
```

```
def load descriptions(doc):
    mapping = dict()
    for line in doc.split('\n'):
        tokens = line.split()
        if len(line) < 2:</pre>
            continue
        image_id, image_desc = tokens[0], tokens[1:]
        image_id = image_id.split('.')[0]
        image_desc = ' '.join(image_desc)
        if image id not in mapping:
            mapping[image_id] = list()
        mapping[image id].append(image desc)
    return mapping
def clean descriptions(descriptions):
    table = str.maketrans('', '', string.punctuation)
   for key, desc list in descriptions.items():
        for i in range(len(desc list)):
            desc = desc list[i]
            desc = desc.split()
            desc = [word.lower() for word in desc]
            desc = [w.translate(table) for w in desc]
            desc = [word for word in desc if len(word)>1]
            desc = [word for word in desc if word.isalpha()]
            desc list[i] = ' '.join(desc)
```

```
def to vocabulary(descriptions):
    all desc = set()
    for key in descriptions.keys():
        [all_desc.update(d.split()) for d in descriptions[key]]
    return all desc
def save_descriptions(descriptions, filename):
    lines = list()
    for key, desc list in descriptions.items():
        for desc in desc list:
            lines.append(key + ' ' + desc)
    data = '\n'.join(lines)
    file = open(filename, 'w')
    file.write(data)
    file.close()
filename = 'Flickr8k_text/Flickr8k.token.txt'
doc = load doc(filename)
```

```
# parse descriptions
```

```
descriptions = load_descriptions(doc)
```

```
print('Loaded: %d ' % len(descriptions))
```

clean descriptions

```
clean_descriptions(descriptions)
```

summarize vocabulary

vocabulary = to_vocabulary(descriptions)
print('Vocabulary Size: %d' % len(vocabulary))

save to file

save_descriptions(descriptions, 'descriptions.txt')

```
# load doc into memory
def load_doc(filename):
    # open the file as read only
    file = open(filename, 'r')
    # read all text
    text = file.read()
    # close the file
    file.close()
    return text
# load a pre-defined list of photo identifiers
def load_set(filename):
    doc = load_doc(filename)
    dataset = list()
    # process line by line
    for line in doc.split('\n'):
        # skip empty lines
        if len(line) < 1:
            continue
        # get the image identifier
        identifier = line.split('.')[0]
        dataset.append(identifier)
        return set(dataset)
# load clean descriptions into memory
def load_clean descriptions into memory
```

```
def load_clean_descriptions(filename, dataset):
    # load document
    doc = load_doc(filename)
    descriptions = dict()
    for line in doc.split('\n'):
        # split line by white space
        tokens = line.split()
        # split id from description
        image_id, image_desc = tokens[0], tokens[1:]
        # skip images not in the set
        if image_id in dataset:
            # create list
            if image_id not in descriptions:
                description in tokens
               description in tokens
               desc = 'startseq ' + ' '.join(image_desc) + ' endseq'
                 # store
                descriptions[image_id].append(desc)
    return descriptions
```

```
def load_photo_features(filename, dataset):
    all features = load(open(filename, 'rb'))
    features = {k: all_features[k] for k in dataset}
    return features
 filename = 'Flickr8k text/Flickr 8k.trainImages.txt'
 train = load set(filename)
 print('Dataset: %d' % len(train))
 train descriptions = load clean descriptions('descriptions.txt', train)
 print('Descriptions: train=%d' % len(train_descriptions))
 train features = load photo features('features InceptionV3.pkl', train)
 print('Photos: train=%d' % len(train features))
def to lines(descriptions):
    all desc = list()
    for key in descriptions.keys():
        [all desc.append(d) for d in descriptions[key]]
    return all desc
def create tokenizer(descriptions):
    lines = to lines(descriptions)
    tokenizer = Tokenizer()
    tokenizer.fit on texts(lines)
    return tokenizer
tokenizer = create tokenizer(train descriptions)
vocab size = len(tokenizer.word index) + 1
print('Vocabulary Size: %d' % vocab size)
```



```
def max length(descriptions):
     lines = to lines(descriptions)
     return max(len(d.split()) for d in lines)
Run Cell | Run Above | Debug Cell
def define_model(vocab_size, max length):
     inputs1 = Input(shape=(2048,))
     fe1 = Dropout(0.5)(inputs1)
     fe2 = Dense(256, activation='relu')(fe1)
     inputs2 = Input(shape=(max length,))
     se1 = Embedding(vocab size, 256, mask zero=True)(inputs2)
     se2 = Dropout(0.5)(se1)
     se3 = LSTM(256)(se2)
     decoder1 = add([fe2, se3])
     decoder2 = Dense(256, activation='relu')(decoder1)
     outputs = Dense(vocab_size, activation='softmax')(decoder2)
     model = Model(inputs=[inputs1, inputs2], outputs=outputs)
     model.compile(loss='categorical crossentropy', optimizer='adam')
     print(model.summary())
     plot model(model, to_file='model_IV3.png', show_shapes=True)
     return model
filename = 'Flickr8k text/Flickr 8k.trainImages.txt'
train = load_set(filename)
print('Dataset: %d' % len(train))
train_descriptions = load_clean_descriptions('descriptions.txt', train)
print('Descriptions: train=%d' % len(train descriptions))
train features = load photo features('features InceptionV3.pkl', train)
print('Photos: train=%d' % len(train features))
```

tokenizer = create_tokenizer(train_descriptions)
vocab_size = len(tokenizer.word_index) + 1
print('Vocabulary Size: %d' % vocab_size)
determine the maximum sequence Length
max_length = max_length(train_descriptions)
print('Description Length: %d' % max_length)
premare sequences

X1train, X2train, ytrain = create_sequences(tokenizer, max_length, train_descriptions, train_features)

filename = 'Flickr8k_text/Flickr_8k.devImages.txt'
test = load_set(filename)
print('Dataset: %d' % len(test))
descriptions

test_descriptions = load_clean_descriptions('descriptions.txt', test)
print('Descriptions: test=%d' % len(test_descriptions))

test_features = load_photo_features('features_InceptionV3.pkl', test)
print('Photos: test=%d' % len(test_features))

X1test, X2test, ytest = create_sequences(tokenizer, max_length, test_descriptions, test_features)

Run Cell | Run Above | Debug Cell #%% fit model

define the model
model = define_model(vocab_size, max_length)
define checkpoint callback
filepath = 'model_IV3-ep{epoch:03d}-loss{loss:.3f}-val_loss{val_loss:.3f}.h5'
checkpoint = ModelCheckpoint(filepath, monitor='val_loss', verbose=1, save_best_only=True, mode='min')
fit model
model.fit([X1train, X2train], ytrain, epochs=10, verbose=2, callbacks=[checkpoint], validation_data=([X1test, X2test], ytest))

from numpy import array from pickle import load from keras.preprocessing.text import Tokenizer from keras.preprocessing.sequence import pad sequences from keras.utils import to_categorical from keras.utils import plot model from keras.models import Model from keras.layers import Input from keras.layers import Dense from keras.layers import LSTM from keras.models import Sequential from keras.layers import TimeDistributed, RepeatVector, Activation # Merge from keras.layers.wrappers import Bidirectional from keras.layers import Embedding from keras.layers import Dropout from keras.layers.merge import add from keras.callbacks import ModelCheckpoint from os import listdir from pickle import dump import string from keras.applications.vgg16 import VGG16 from keras.preprocessing.image import load img from keras.preprocessing.image import img to array from keras.applications.vgg16 import preprocess_input

```
def extract_features(directory):
   model = VGG16()
   model.layers.pop()
   model = Model(inputs=model.inputs, outputs=model.layers[-1].output)
    print(model.summary())
    features = dict()
   for name in listdir(directory):
        filename = directory + '/' + name
        image = load_img(filename, target_size=(224, 224))
        image = img_to_array(image)
        image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
        image = preprocess input(image)
        feature = model.predict(image, verbose=0)
        image_id = name.split('.')[0]
        features[image id] = feature
        print('>%s' % name)
    return features
directory = 'Flicker8k Dataset'
features = extract features(directory)
print('Extracted Features: %d' % len(features))
dump(features, open('features.pkl', 'wb'))
# load doc into memor
def load_doc(filename):
```

```
# open the file as read only
file = open(filename, 'r')
# read all text
text = file.read()
# close the file
file.close()
return text
```

```
#% extract descriptions for images
def load_descriptions(doc):
    mapping = dict()
    # process lines
    for line in doc.split('\n'):
        # split line by white space
        tokens = line.split()
        if len(line) < 2:
            continue
        # take the first token as the image id, the rest as the description
        image_id, image_desc = tokens[0], tokens[1:]
        # remove filename from image id
        image_id = image_id.split('.')[0]
        # convert description tokens back to string
        image_desc = ' '.join(image_desc)
        # create the list if needed
        if image_id] = list()
        # store description
        mapping[image_id].append(image_desc)
    return mapping
</pre>
```

```
def clean descriptions(descriptions):
    table = str.maketrans('', '', string.punctuation)
   for key, desc list in descriptions.items():
        for i in range(len(desc list)):
            desc = desc list[i]
            desc = desc.split()
            desc = [word.lower() for word in desc]
            desc = [w.translate(table) for w in desc]
            desc = [word for word in desc if len(word)>1]
            desc = [word for word in desc if word.isalpha()]
            desc_list[i] = ' '.join(desc)
           the loaded descriptions into a vocabulary of words
def to_vocabulary(descriptions):
    all desc = set()
    for key in descriptions.keys():
        [all_desc.update(d.split()) for d in descriptions[key]]
    return all desc
def save descriptions(descriptions, filename):
    lines = list()
    for key, desc_list in descriptions.items():
        for desc in desc list:
            lines.append(key + ' ' + desc)
    data = '\n'.join(lines)
    file = open(filename, 'w')
    file.write(data)
    file.close()
```

```
filename = 'Flickr8k_text/Flickr8k.token.txt'
# load descriptions
doc = load_doc(filename)
# parse descriptions
descriptions = load_descriptions(doc)
print('Loaded: %d ' % len(descriptions))
# clean descriptions
clean_descriptions(descriptions)
# summarize vocabulary
vocabulary = to_vocabulary(descriptions)
print('Vocabulary Size: %d' % len(vocabulary))
# save to file
save_descriptions(descriptions, 'descriptions.txt')
```

```
# Load doc into memory
```

```
def load_doc(filename):
    # open the file as read only
    file = open(filename, 'r')
    # read all text
    text = file.read()
    # close the file
    file.close()
    return text
```

```
# load a pre-defined list of photo identifiers
def load_set(filename):
```

```
doc = load_doc(filename)
dataset = list()
# process line by line
for line in doc.split('\n'):
# skip empty lines
if len(line) < 1:
    continue
# get the image identifier
identifier = line.split('.')[0]
dataset.append(identifier)
return set(dataset)</pre>
```

```
# Load clean descriptions into memory
def load_clean_descriptions(filename, dataset):
    # Load document
    doc = load_doc(filename)
    descriptions = dict()
    for line in doc.split('\n'):
        # split line by white space
        tokens = line.split()
        # split id from description
        image_id, image_desc = tokens[0], tokens[1:]
        # skip images not in the set
        if image_id in dataset:
            # create list
            if image_id not in descriptions:
                descriptions[image_id] = list()
        # wrap description in tokens
        desc = 'startseq ' + ' '.join(image_desc) + ' endseq'
        # store
        descriptions[image_id].append(desc)
    return descriptions
```

```
def load_photo_features(filename, dataset):
    # load all features
    all_features = load(open(filename, 'rb'))
    # filter features
    features = {k: all_features[k] for k in dataset}
    return features
# load training dataset (6K)
filename = 'Flickr8k_text/Flickr_8k.trainImages.txt'
train = load_set(filename)
print('Dataset: %d' % len(train))
# descriptions
train_descriptions = load_clean_descriptions('descriptions.txt', train)
```

```
print('Descriptions: train=%d' % len(train_descriptions))
# photo features
train_features = load_photo_features('features.pkl', train)
```

```
print('Photos: train=%d' % len(train_features))
```

```
def to lines(descriptions):
    all desc = list()
    for key in descriptions.keys():
        [all desc.append(d) for d in descriptions[key]]
    return all desc
def create tokenizer(descriptions):
    lines = to lines(descriptions)
    tokenizer = Tokenizer()
    tokenizer.fit on texts(lines)
    return tokenizer
tokenizer = create tokenizer(train descriptions)
vocab size = len(tokenizer.word index) + 1
print('Vocabulary Size: %d' % vocab size)
def create_sequences(tokenizer, max_length, descriptions, photos):
    X1, X2, y = list(), list(), list()
    for key, desc_list in descriptions.items():
        for desc in desc list:
            seq = tokenizer.texts_to_sequences([desc])[0]
            for i in range(1, len(seq)):
                in_seq, out_seq = seq[:i], seq[i]
                in_seq = pad_sequences([in_seq], maxlen=max_length)[0]
                out_seq = to_categorical([out_seq], num_classes=vocab_size)[0]
                X1.append(photos[key][0])
                X2.append(in seq)
                y.append(out_seq)
    return array(X1), array(X2), array(y)
def max length(descriptions):
    lines = to lines(descriptions)
    return max(len(d.split()) for d in lines)
```

def define_model(vocab_size, max_length):

```
inputs1 = Input(shape=(4096,))
  fe1 = Dropout(0.5)(inputs1)
  fe2 = Dense(256, activation='relu')(fe1)
  fe3 = RepeatVector(max length)(fe2)
  inputs2 = Input(shape=(max length,))
  se1 = Embedding(vocab_size, 256, mask_zero=True)(inputs2)
  se2 = Dropout(0.5)(se1)
  se3 = LSTM(256, return sequences=True)(se2)
  se4 = TimeDistributed(Dense(256))(se3)
  decoder1 = add([fe3, se4])
  decoder2 = Bidirectional(LSTM(256, return_sequences=False))(decoder1)
  outputs = Dense(vocab_size, activation='softmax')(decoder2)
  model = Model(inputs=[inputs1, inputs2], outputs=outputs)
  model.compile(loss='categorical crossentropy', optimizer='rmsprop')
  print(model.summary())
  plot model(model, to file='model.png', show shapes=True)
  return model
filename = 'Flickr8k_text/Flickr_8k.trainImages.txt'
train = load set(filename)
print('Dataset: %d' % len(train))
train_descriptions = load_clean_descriptions('descriptions.txt', train)
print('Descriptions: train=%d' % len(train_descriptions))
train_features = load_photo_features('features.pkl', train)
print('Photos: train=%d' % len(train_features))
tokenizer = create_tokenizer(train_descriptions)
vocab_size = len(tokenizer.word_index) + 1
print('Vocabulary Size: %d' % vocab_size)
max length = max length(train descriptions)
print('Description Length: %d' % max_length)
X1train, X2train, ytrain = create_sequences(tokenizer, max_length, train_descriptions, train_features)
```

filename = 'Flickr&k_text/Flickr_&k.devImages.txt'
test = load_set(filename)
print('Dataset: %d' % len(test))
descriptions

test_descriptions = load_clean_descriptions('descriptions.txt', test)
print('Descriptions: test=%d' % len(test_descriptions))

test_features = load_photo_features('features.pkl', test)
print('Photos: test=%d' % len(test_features))

X1test, X2test, ytest = create_sequences(tokenizer, max_length, test_descriptions, test_features)

Run Cell | Run Above | Debug Cell #%% fit model

model = define_model(vocab_size, max_length)

define checkpoint callback
filepath = 'model_vgg_bilstm-ep{epoch:03d}-loss{loss:.3f}-val_loss{val_loss:.3f}.h5'
checkpoint = ModelCheckpoint(filepath, monitor='val_loss', verbose=1, save_best_only=True, mode='min')

model.fit([X1train, X2train], ytrain, epochs=20, verbose=2, callbacks=[checkpoint], validation_data=([X1test, X2test], ytest))

from numpy import array from pickle import load from keras.preprocessing.text import Tokenizer from keras.preprocessing.sequence import pad_sequences from keras.utils import to categorical from keras.utils import plot model from keras.models import Model from keras.layers import Input from keras.layers import Dense from keras.layers import LSTM from keras.layers import Embedding from keras.layers import Dropout from keras.layers.merge import add from keras.callbacks import ModelCheckpoint from os import listdir from pickle import dump import string from keras.applications.vgg16 import VGG16 from keras.preprocessing.image import load_img from keras.preprocessing.image import img_to_array from keras.applications.vgg16 import preprocess_input def extract features(directory): model = VGG16()model.layers.pop() model = Model(inputs=model.inputs, outputs=model.layers[-1].output) print(model.summary()) features = dict() for name in listdir(directory): filename = directory + '/' + name image = load_img(filename, target_size=(224, 224)) image = img to array(image) image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2])) image = preprocess_input(image) feature = model.predict(image, verbose=0) image_id = name.split('.')[0] features[image id] = feature print('>%s' % name) return features

```
Run Cell | Run Above | Debug Cell
directory = 'Flicker8k Dataset'
features = extract features(directory)
print('Extracted Features: %d' % len(features))
dump(features, open('features.pkl', 'wb'))
def load doc(filename):
    file = open(filename, 'r')
    text = file.read()
    file.close()
    return text
Run Cell | Run Above | Debug Cell
def load_descriptions(doc):
    mapping = dict()
    for line in doc.split('\n'):
        tokens = line.split()
        if len(line) < 2:</pre>
            continue
        image_id, image_desc = tokens[0], tokens[1:]
        image_id = image_id.split('.')[0]
        image_desc = ' '.join(image_desc)
        if image id not in mapping:
            mapping[image_id] = list()
        mapping[image_id].append(image desc)
    return mapping
```

```
def clean descriptions(descriptions):
   table = str.maketrans('', '', string.punctuation)
   for key, desc list in descriptions.items():
        for i in range(len(desc_list)):
            desc = desc_list[i]
            desc = desc.split()
            desc = [word.lower() for word in desc]
            desc = [w.translate(table) for w in desc]
            desc = [word for word in desc if len(word)>1]
            desc = [word for word in desc if word.isalpha()]
           desc_list[i] = ' '.join(desc)
def to_vocabulary(descriptions):
   all desc = set()
   for key in descriptions.keys():
        [all_desc.update(d.split()) for d in descriptions[key]]
   return all desc
```

```
# convert the loaded descriptions into a vocabulary of words
def to_vocabulary(descriptions):
    # build a list of all description strings
    all_desc = set()
    for key in descriptions.keys():
        [all_desc.update(d.split()) for d in descriptions[key]]
    return all_desc
# save descriptions to file, one per line
def save_descriptions(descriptions, filename):
    lines = list()
    for key, desc_list in descriptions.items():
        for desc in desc_list:
            lines.append(key + ' ' + desc)
        data = '\n'.join(lines)
        file = open(filename, 'w')
        file.write(data)
        file.close()
```

```
filename = 'Flickr8k_text/Flickr8k.token.txt'
# load descriptions
doc = load_doc(filename)
# parse descriptions
descriptions = load_descriptions(doc)
print('Loaded: %d ' % len(descriptions))
# clean descriptions
clean_descriptions(descriptions)
# summarize vocabulary
vocabulary = to_vocabulary(descriptions)
print('Vocabulary Size: %d' % len(vocabulary))
# save to file
save_descriptions(descriptions, 'descriptions.txt')
```

```
def load doc(filename):
    file = open(filename, 'r')
    text = file.read()
    file.close()
    return text
def load set(filename):
    doc = load_doc(filename)
    dataset = list()
    for line in doc.split('\n'):
        if len(line) < 1:
            continue
        identifier = line.split('.')[0]
        dataset.append(identifier)
    return set(dataset)
def load clean descriptions(filename, dataset):
    doc = load doc(filename)
    descriptions = dict()
    for line in doc.split('\n'):
        tokens = line.split()
        image_id, image_desc = tokens[0], tokens[1:]
        if image id in dataset:
            if image_id not in descriptions:
                descriptions[image id] = list()
            desc = 'startseq ' + ' '.join(image desc) + ' endseq'
            descriptions[image id].append(desc)
    return descriptions
```

```
def load photo features(filename, dataset):
    all_features = load(open(filename, 'rb'))
    features = {k: all_features[k] for k in dataset}
    return features
filename = 'Flickr8k text/Flickr 8k.trainImages.txt'
train = load set(filename)
print('Dataset: %d' % len(train))
train descriptions = load clean descriptions('descriptions.txt', train)
print('Descriptions: train=%d' % len(train_descriptions))
train features = load photo features('features.pkl', train)
print('Photos: train=%d' % len(train_features))
def to lines(descriptions):
   all desc = list()
   for key in descriptions.keys():
        [all desc.append(d) for d in descriptions[key]]
   return all desc
def create tokenizer(descriptions):
   lines = to lines(descriptions)
   tokenizer = Tokenizer()
   tokenizer.fit on texts(lines)
   return tokenizer
tokenizer = create tokenizer(train descriptions)
vocab size = len(tokenizer.word index) + 1
print('Vocabulary Size: %d' % vocab size)
```

```
def create sequences(tokenizer, max length, descriptions, photos):
    X1, X2, y = list(), list(), list()
    for key, desc_list in descriptions.items():
        for desc in desc_list:
            seq = tokenizer.texts_to_sequences([desc])[0]
            for i in range(1, len(seq)):
                in seq, out seq = seq[:i], seq[i]
                in seq = pad sequences([in seq], maxlen=max length)[0]
                out seq = to categorical([out seq], num classes=vocab size)[0]
                X1.append(photos[key][0])
                X2.append(in seq)
                y.append(out seq)
    return array(X1), array(X2), array(y)
def max length(descriptions):
    lines = to lines(descriptions)
    return max(len(d.split()) for d in lines)
def define_model(vocab_size, max_length):
    inputs1 = Input(shape=(4096,))
    fe1 = Dropout(0.5)(inputs1)
    fe2 = Dense(256, activation='relu')(fe1)
    inputs2 = Input(shape=(max_length,))
    se1 = Embedding(vocab size, 256, mask zero=True)(inputs2)
    se2 = Dropout(0.5)(se1)
    se3 = LSTM(256)(se2)
    decoder1 = add([fe2, se3])
    decoder2 = Dense(256, activation='relu')(decoder1)
    outputs = Dense(vocab_size, activation='softmax')(decoder2)
    model = Model(inputs=[inputs1, inputs2], outputs=outputs)
    model.compile(loss='categorical_crossentropy', optimizer='adam')
    print(model.summary())
    plot model(model, to file='model.png', show shapes=True)
    return model
```

```
filename = 'Flickr8k_text/Flickr_8k.trainImages.txt'
train = load_set(filename)
print('Dataset: %d' % len(train))
train descriptions = load clean descriptions('descriptions.txt', train)
print('Descriptions: train=%d' % len(train_descriptions))
train_features = load_photo_features('features.pkl', train)
print('Photos: train=%d' % len(train_features))
tokenizer = create tokenizer(train descriptions)
vocab_size = len(tokenizer.word_index) + 1
print('Vocabulary Size: %d' % vocab_size)
max_length = max_length(train_descriptions)
print('Description Length: %d' % max length)
X1train, X2train, ytrain = create_sequences(tokenizer, max_length, train_descriptions, train_features)
filename = 'Flickr8k_text/Flickr_8k.devImages.txt'
test = load_set(filename)
print('Dataset: %d' % len(test))
test_descriptions = load_clean_descriptions('descriptions.txt', test)
print('Descriptions: test=%d' % len(test descriptions))
test_features = load_photo_features('features.pkl', test)
print('Photos: test=%d' % len(test_features))
X1test, X2test, ytest = create_sequences(tokenizer, max_length, test_descriptions, test_features)
model = define model(vocab size, max length)
filepath = 'model-ep{epoch:03d}-loss{loss:.3f}-val_loss{val_loss:.3f}.h5'
checkpoint = ModelCheckpoint(filepath, monitor='val_loss', verbose=1, save_best_only=True, mode='min')
model.fit([X1train, X2train], ytrain, epochs=20, verbose=2,
     callbacks=[checkpoint], validation_data=([X1test, X2test], ytest))
```

```
from numpy import argmax
from pickle import load
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.sequence import pad_sequences
from keras.models import load_model
from nltk.translate.bleu_score import corpus_bleu
```

load doc into memory

```
def load_doc(filename):
    # open the file as read onl
    file = open(filename, 'r')
    # read all text
    text = file.read()
    # close the file
    file.close()
    return text
```

load a pre-defined list of photo identifiers

```
def load_set(filename):
    doc = load_doc(filename)
    dataset = list()
    # process line by line
    for line in doc.split('\n'):
        # skip empty lines
        if len(line) < 1:
            continue
        # get the image identifier
        identifier = line.split('.')[0]
        dataset.append(identifier)
    return set(dataset)
```

```
def load_clean_descriptions(filename, dataset):
    doc = load doc(filename)
    descriptions = dict()
    for line in doc.split('\n'):
        tokens = line.split()
        image_id, image_desc = tokens[0], tokens[1:]
        if image id in dataset:
            if image id not in descriptions:
                descriptions[image_id] = list()
            desc = 'startseq ' + ' '.join(image_desc) + ' endseq'
            descriptions[image_id].append(desc)
    return descriptions
def load photo features(filename, dataset):
    all features = load(open(filename, 'rb'))
    features = {k: all features[k] for k in dataset}
    return features
```

```
def load photo features(filename, dataset):
   all_features = load(open(filename, 'rb'))
    features = {k: all features[k] for k in dataset}
    return features
def to lines(descriptions):
   all_desc = list()
   for key in descriptions.keys():
        [all desc.append(d) for d in descriptions[key]]
   return all desc
def create_tokenizer(descriptions):
   lines = to lines(descriptions)
   tokenizer = Tokenizer()
   tokenizer.fit on texts(lines)
    return tokenizer
def max length(descriptions):
   lines = to lines(descriptions)
   return max(len(d.split()) for d in lines)
```

```
def word for id(integer, tokenizer):
    for word, index in tokenizer.word index.items():
        if index == integer:
            return word
    return None
def generate_desc(model, tokenizer, photo, max_length):
    in text = 'startseq'
    for i in range(max_length):
        sequence = tokenizer.texts to sequences([in text])[0]
        sequence = pad_sequences([sequence], maxlen=max_length)
        yhat = model.predict([photo,sequence], verbose=0)
        yhat = argmax(yhat)
        word = word_for_id(yhat, tokenizer)
        if word is None:
            break
        in text += ' ' + word
        if word == 'endseq':
            break
    return in_text
```

```
def evaluate_model(model, descriptions, photos, tokenizer, max_length):
    actual, predicted = list(), list()
   for key, desc_list in descriptions.items():
       if (key == '3320032226_63390d74a6'):
           yhat = generate_desc(model, tokenizer, photos[key], max_length)
           references = [d.split() for d in desc_list]
           actual.append(references)
           predicted.append(yhat.split())
    print('BLEU-1: %f' % corpus_bleu(actual, predicted, weights=(1.0, 0, 0, 0)))
    print('BLEU-2: %f' % corpus bleu(actual, predicted, weights=(0.5, 0.5, 0, 0)))
    print('BLEU-3: %f' % corpus_bleu(actual, predicted, weights=(0.3, 0.3, 0.3, 0)))
    print('BLEU-4: %f' % corpus_bleu(actual, predicted, weights=(0.25, 0.25, 0.25, 0.25)))
filename = 'Flickr8k text/Flickr 8k.trainImages.txt'
train = load set(filename)
print('Dataset: %d' % len(train))
train descriptions = load clean descriptions('descriptions.txt', train)
print('Descriptions: train=%d' % len(train_descriptions))
tokenizer = create_tokenizer(train_descriptions)
vocab size = len(tokenizer.word index) + 1
```

print('Vocabulary Size: %d' % vocab_size)

determine the maximum sequence length

max_Length = max_length(train_descriptions)
print('Description Length: %d' % max Length)

```
# Load test set
filename = 'Flickr8k_text/Flickr_8k.testImages.txt'
test = load_set(filename)
print('Dataset: %d' % len(test))
# descriptions
test_descriptions = load_clean_descriptions('descriptions.txt', test) #
print('Descriptions: test=%d' % len(test_descriptions))
# photo features
test_features = load_photo_features('features.pkl', test) # change the p
print('Photos: test=%d' % len(test_features))
# Load the model
filename = 'model-ep004-loss3.604-val_loss3.830.h5' # change the m
model = load_model(filename)
for key, desc_list in test_descriptions.items():
    if (key == '3320032226_63390d74a6'):
        # generate_description
        yhat = generate_desc(model, tokenizer, test_features[key], max_Length)
        print(yhat+'\n'+key)
```

break

```
import glob
from PIL import Image
import numpy as np
import matplotlib.pyplot as plt
import pickle
from tqdm import tqdm
import pandas as pd
from keras.preprocessing import sequence
from keras.models import Sequential
from keras.layers import LSTM, Embedding, TimeDistributed, Dense, RepeatVector, Activation, Flatten
from keras.optimizers import Adam, RMSprop
from keras.layers.wrappers import Bidirectional
from keras.applications.inception_v3 import InceptionV3
from keras.applications.inception_v3 import preprocess_input
from keras.preprocessing import image
import nltk
token = 'Flickr8k text/Flickr8k.token.txt'
captions = open(token, 'r').read().strip().split('\n')
d = \{\}
for i, row in enumerate(captions):
     row = row.split('\t')
     row[0] = row[0][:len(row[0])-2]
     if row[0] in d:
         d[row[0]].append(row[1])
     else:
         d[row[0]] = [row[1]]
print(d['1000268201_693b08cb0e.jpg'])
images = 'Flicker8k_Dataset/'
img = glob.glob(images+'*.jpg')
print(img[:5])
train_images_file = 'Flickr8k_text/Flickr_8k.trainImages.txt'
train_images = set(open(train_images_file, 'r').read().strip().split('\n'))
```

```
def split_data(l):
    temp = []
    for i in img:
        if i[len(images):] in L:
            temp.append(i)
    return temp
train img = split data(train images)
len(train_img)
val images file = 'Flickr8k text/Flickr 8k.devImages.txt'
val_images = set(open(val_images_file, 'r').read().strip().split('\n'))
val_img = split_data(val_images)
len(val img)
test images file = 'Flickr8k text/Flickr 8k.testImages.txt'
test images = set(open(test images file, 'r').read().strip().split('\n'))
test_img = split_data(test_images)
len(test_img)
Image.open(train img[0])
model = InceptionV3(weights='imagenet')
from keras.models import Model
new input = model.input
hidden layer = model.layers[-2].output
model_new = Model(new_input, hidden_layer)
def encode(image path):
    img = image.load_img(image_path, target_size=(299, 299))
   x = image.img_to_array(img)
   x = np.expand dims(x, axis=0)
    img p = preprocess input(x)
    temp enc = model new.predict(img p)
    temp enc = np.reshape(temp enc, temp enc.shape[1])
    return temp enc
```

```
encoding_train = pickle.load(open('encoded_images_inceptionV3.p', 'rb'))
encoding_train['3556792157_d09d42bef7.jpg'].shape
encoding_test = pickle.load(open('encoded_images_test_inceptionV3.p', 'rb'))
encoding_test[test_img[0][len(images):]].shape
Run Cell | Run Below | Debug Cell
train_d = {}
for i in train img:
    if i[len(images):] in d:
        train d[i] = d[i[len(images):]]
len(train d)
val_d = {}
for i in val img:
    if i[len(images):] in d:
        val_d[i] = d[i[len(images):]]
len(val d)
```

```
val_d = {}
for i in val_img:
    if i[len(images):] in d:
        val_d[i] = d[i[len(images):]]
len(val_d)
test_d = {}
for i in test_img:
    if i[len(images):] in d:
        test_d[i] = d[i[len(images):]]
len(test_d)
Run Cell | Run Above | Debug Cell
caps = []
for key, val in train_d.items():
    for i in val:
        caps.append('<start> ' + i + ' <end>')
words = [i.split() for i in caps]
unique = []
for i in words:
    unique.extend(i)
unique = list(set(unique))
with open("unique.p", "wb") as pickle_d:
     pickle.dump(unique, pickle_d)
unique = pickle.load(open('unique.p', 'rb'))
len(unique)
```

```
word2idx = {val:index for index, val in enumerate(unique)}
word2idx['<start>']
idx2word = {index:val for index, val in enumerate(unique)}
print(idx2word[5553])
max_len = 0
for c in caps:
   c = c.split()
   if len(c) > max_len:
        max_len = len(c)
print(max len)
print(len(unique), max_len)
vocab size = len(unique)
print(vocab_size)
f = open('flickr8k training dataset.txt', 'w')
f.write("image_id\tcaptions\n")
for key, val in train_d.items():
   for i in val:
        f.write(key[len(images):] + "\t" + "<start> " + i +" <end>" + "\n")
f.close()
df = pd.read_csv('flickr8k_training_dataset.txt', delimiter='\t')
print(len(df))
c = [i for i in df['captions']]
print(len(c))
imgs = [i for i in df['image_id']]
a = c[-1]
a, imgs[-1]
for i in a.split():
    print (i, "=>", word2idx[i])
samples_per_epoch = 0
for ca in caps:
    samples_per_epoch += len(ca.split())-1
print(samples per epoch)
```

```
def data_generator(batch_size = 32):
    partial_caps = []
    next_words = []
    images = []

    df = pd.read_csv('flickr8k_training_dataset.txt', delimiter='\t')
    df = df.sample(frac=1)
    iter = df.iterrows()
    c = []
    imgs = []
    for i in range(df.shape[0]):
        x = next(iter)
        c.append(x[1][1])
        imgs.append(x[1][0])
```

count = 0

```
while True:
   for j, text in enumerate(c):
       current_image = encoding_train[imgs[j]]
       for i in range(len(text.split())-1):
           count+=1
           partial = [word2idx[txt] for txt in text.split()[:i+1]]
           partial_caps.append(partial)
           n = np.zeros(vocab_size)
           n[word2idx[text.split()[i+1]]] = 1
           next_words.append(n)
            images.append(current_image)
            if count>=batch_size:
               next_words = np.asarray(next_words)
               images = np.asarray(images)
               partial_caps = sequence.pad_sequences(partial_caps, maxlen=max_len, padding='post')
               yield [[images, partial_caps], next_words]
               partial_caps = []
               next_words = []
               images = []
               count = 0
```

```
from keras.utils import plot model
from keras.models import Model
from keras.layers import Input, Dropout
from keras, layers, merge import add
embedding_size = 300
inputs1 = Input(shape=(2048,))
fe1 = Dropout(0.5)(inputs1)
fe2 = Dense(embedding size, activation='relu')(fe1)
fe3 = RepeatVector(max_len)(fe2)
inputs2 = Input(shape=(max len,))
se1 = Embedding(vocab_size, embedding_size, mask_zero=True)(inputs2)
se2 = Dropout(0.5)(se1)
se3 = LSTM(256, return sequences=True)(se2)
se4 = TimeDistributed(Dense(embedding size))(se3)
decoder1 = add([fe3, se4])
decoder2 = Bidirectional(LSTM(256, return sequences=False))(decoder1)
outputs = Dense(vocab_size, activation='softmax')(decoder2)
final_model = Model(inputs=[inputs1, inputs2], outputs=outputs)
final model.compile(loss='categorical crossentropy', optimizer='rmsprop')
print(final model.summary())
plot_model(final_model, to_file='model.png', show_shapes=True)
final_model.fit_generator(data_generator(batch_size=128), samples_per_epoch=samples_per_epoch, nb_epoch=1, verbose=2)
```

```
#%% Prediction function
```

```
def predict_captions(image):
    start_word = ["<start>"]
    while True:
        par_caps = [word2idx[i] for i in start_word]
        par_caps = sequence.pad_sequences([par_caps], maxlen=max_len, padding='post')
        e = encoding_test[image[len(images):]]
        preds = final_model.predict([np.array([e]), np.array(par_caps)])
        word_pred = idx2word[np.argmax(preds[0])]
        start_word.append(word_pred)
        if word_pred == "<end>" or len(start_word) > max_len:
            break
        return ' '.join(start_word[1:-1])
```

```
def beam_search_predictions(image, beam_index = 3):
    start = [word2idx["<start>"]]
    start_word = [[start, 0.0]]
    while len(start_word[0][0]) < max_len:</pre>
        temp = []
       for s in start_word:
            par_caps = sequence.pad_sequences([s[0]], maxlen=max_len, padding='post')
            e = encoding_test[image[len(images):]]
            preds = final_model.predict([np.array([e]), np.array(par_caps)])
            word_preds = np.argsort(preds[0])[-beam_index:]
            for w in word_preds:
                next_cap, prob = s[0][:], s[1]
                next_cap.append(w)
                prob += preds[0][w]
                temp.append([next_cap, prob])
        start_word = temp
        start_word = sorted(start_word, reverse=False, key=lambda l: l[1])
        start_word = start_word[-beam_index:]
    start word = start word[-1][0]
    intermediate_caption = [idx2word[i] for i in start_word]
    final_caption = []
```

```
for i in intermediate_caption:
    if i != '<end>':
        final_caption.append(i)
    else:
        break
    final_caption = ' '.join(final_caption[1:])
    return final_caption
Run Cell | Run Above | Debug Cell
    #255
try_image = test_img[0]
Image.open(try_image)
print ('Normal Max search:', predict_captions(try_image))
print ('Beam Search, k=3:', beam_search_predictions(try_image, beam_index=3))
print ('Beam Search, k=5:', beam_search_predictions(try_image, beam_index=3))
print ('Beam Search, k=7:', beam_search_predictions(try_image, beam_index=5))
print ('Beam Search, k=7:', beam_search_predictions(try_image, beam_index=7))
```

```
from pickle import load
from numpy import argmax
from keras.preprocessing.sequence import pad sequences
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img to array
from keras.models import Model
from keras.models import load model
from keras.applications.vgg16 import VGG16
from keras.applications.vgg16 import preprocess_input
def extract_features(filename):
   model = VGG16()
   model.layers.pop()
   model = Model(inputs=model.inputs, outputs=model.layers[-1].output)
    image = load_img(filename, target_size=(224, 224))
    image = img_to_array(image)
    image = image.reshape((1, image.shape[0], image.shape[1], image.shape[2]))
    image = preprocess_input(image)
   feature = model.predict(image, verbose=0)
   return feature
Run Cell | Run Below | Debug Cell
```

```
def word_for_id(integer, tokenizer):
    for word, index in tokenizer.word_index.items():
        if index == integer:
            return word
        return None
```

```
# generate a description for an image
def generate_desc(model, tokenizer, photo, max_length):
    # seed the generation process
    in_text = 'startseq'
    # iterate over the whole length of the sequence
    for i in range(max_length):
        # integer encode input sequence
        sequence = tokenizer.texts_to_sequences([in_text])[0]
        # pad input
        sequence = pad_sequences([sequence], maxlen=max_length)
        # predict next word
        yhat = model.predict([photo,sequence], verbose=0)
        # convert probability to integer
        yhat = argmax(yhat)
        # map integer to word
        word = word_for_id(yhat, tokenizer)
        # stop if we cannot map the word
        if word is None:
            break
        # append as input for generating the next word
        in_text += ' ' + word
        # stop if we predict the end of the sequence
        if word == 'endseq':
            break
        return in_text
```

tokenizer = load(open('tokenizer.pkl', 'rb')) $max_length = 34$ Run Cell | Run Above | Debug Cell model = load model('model vgg bilstm-ep006-loss3.949-val loss4.149.h5') Run Cell | Run Above | Debug Cell photo = extract features('test/attn.jpg') description = generate desc(model, tokenizer, photo, max length) print(description) photo = extract_features('test/woman-of-dog.jpg') description = generate desc(model, tokenizer, photo, max length) print(description) photo = extract features('test/dog-of-dog.jpg') description = generate desc(model, tokenizer, photo, max length) print(description) photo = extract_features('test/grass-attn.jpg') description = generate_desc(model, tokenizer, photo, max_length) print(description) photo = extract features('test/iist car.jpg') description = generate desc(model, tokenizer, photo, max length) print(description) photo = extract_features('test/iist_sign.jpg') description = generate_desc(model, tokenizer, photo, max_length) print(description)

```
from keras.preprocessing.text import Tokenizer
from pickle import dump
def load_doc(filename):
    file = open(filename, 'r')
    text = file.read()
    file.close()
    return text
def load_set(filename):
    doc = load_doc(filename)
    dataset = list()
    for line in doc.split('\n'):
        if len(line) < 1:</pre>
            continue
        identifier = line.split('.')[0]
        dataset.append(identifier)
    return set(dataset)
```

```
def load clean descriptions(filename, dataset):
    doc = load_doc(filename)
    descriptions = dict()
   for line in doc.split('\n'):
        tokens = line.split()
        image_id, image_desc = tokens[0], tokens[1:]
        if image_id in dataset:
            if image id not in descriptions:
                descriptions[image_id] = list()
            desc = 'startseq ' + ' '.join(image_desc) + ' endseq'
            descriptions[image id].append(desc)
    return descriptions
def to lines(descriptions):
   all_desc = list()
   for key in descriptions.keys():
        [all desc.append(d) for d in descriptions[key]]
    return all desc
def create tokenizer(descriptions):
    lines = to lines(descriptions)
    tokenizer = Tokenizer()
    tokenizer.fit on texts(lines)
    return tokenizer
filename = 'Flickr8k text/Flickr 8k.trainImages.txt'
train = load set(filename)
print('Dataset: %d' % len(train))
train descriptions = load clean descriptions('descriptions.txt', train)
print('Descriptions: train=%d' % len(train_descriptions))
tokenizer = create tokenizer(train descriptions)
dump(tokenizer, open('tokenizer.pkl', 'wb'))
```

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