## Final\_Project\_Image\_Captioning\_pc3019

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## 0.0.1 Image Captioning

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Another objective of this project was to perform image captioning of the book cover images. The image captioning model was trained using the Flickr8K dataset. It is a dataset of 8,000 images where each image is mapped to 5 captions which describe the content of the image. The image embedding was implemented using Inception v3 which is a model pretrained on ImageNet. It accepts an input shape of 299x299x3 and creates an embedding vector of dimensions 256x1. The captions from the dataset are preprocessed before passing to the RNNs. Text preprocessing includes converting to lowercase, removing special characters, and more. "startseq" is added to the beginning of each caption. Similarly, "endseq" is added to the end of each caption. The word embedding technique used is GloVe. GloVe is an unsupervised learning algorithm to obtain the word vector representations. The GloVe embedding representation used in this project has 6 billion tokens and 200 features. To obtain the predicted image caption, the image is passed to the model with the input string "startseq". Then the model predicts the next word and this word is appended to the input string. This repeats until "endseq" is reached or the maximum sentence length is reached.

```
[81]: from time import time
      import os
      import glob
      from tqdm import tqdm
      import numpy as np
      import tensorflow.keras as keras
      import matplotlib.pyplot as plt
      from tensorflow.keras.preprocessing import image
      import string
      import random
      from tensorflow.keras.utils import to categorical
      from tensorflow.keras.layers import add
      from tensorflow.keras.models import Model
      from pickle import dump, load
      from tensorflow.keras.preprocessing.sequence import pad_sequences
      from tensorflow.keras.applications.inception_v3 import preprocess_input
      from tensorflow.keras.applications.inception_v3 import InceptionV3
      from tensorflow.keras.layers import Input, Dense, LSTM, Dropout, Embedding
      from tensorflow.keras.utils import plot_model
```

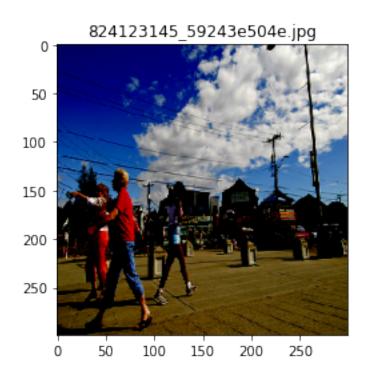
```
[82]: os.listdir('../input/flickr8k')
```

```
[82]: ['captions.txt', 'Images']
 []:
[83]: my_images = glob.glob('../input/flickr8k/Images/' +'*.jpg')
[84]: # Defining utility function which is used to load captions.txt
      def loadFileUtil(file_name):
          f = open(file_name,'r')
          content = f.read()
          f.close()
          return content
      captions = loadFileUtil('../input/flickr8k/captions.txt')
      captions = captions.split('\n')
[85]: captions.pop(0)
[85]: 'image, caption'
[86]: captions.pop(-1)
[86]: ''
[87]: # Formatting the captions
      my_images = []
      for x in range(len(captions)):
          curr_img = captions[x].split(',')
          curr_img = curr_img[0]
          my_images.append(curr_img)
      my_images= set(my_images)
 []:
[88]: # Populating a dictionary where the image name is the key and captions are the
       \rightarrow value
      imgNameDict = {}
      for curr_img in my_images:
          imgNameDict[curr_img] = []
      for curr_image_caption in captions:
          curr_image_caption = curr_image_caption.split(',')
          curr_img,mycap = curr_image_caption[0],curr_image_caption[-1]
          imgNameDict[curr_img].append(mycap)
 []:
```

```
[]:
 []:
[89]: # Creating a translation table for punctuation removal
      mytable = str.maketrans('', '', string.punctuation)
[90]: for curr_img, captionList in imgNameDict.items():
          for x,curr_cap in enumerate(captionList):
              curr_cap = curr_cap.split()
              curr_cap = [y.lower() for y in curr_cap]
              curr_cap = [y.translate(mytable) for y in curr_cap]
              curr_cap = [y for y in curr_cap if len(y) >1]
              curr_cap = [y for y in curr_cap if y.isalpha()]
              captionList[x] = ' '.join(curr_cap)
 []:
[91]: # Defining function to save the caption data
      def saveCaptionData(caps,file_name):
          lines = []
          for img,capList in caps.items():
              for curr_cap in capList:
                  lines.append(img + ',' + curr_cap)
              content = '\n'.join(lines)
              f = open(file_name,'w')
              f.write(content)
              f.close()
[92]: # Saving the cleaned captions
      saveCaptionData(imgNameDict,'./cleanedCaptionData.txt')
[93]: # Defining function to load the caption data
      def loadCaptionData(file_name,image_path):
          myfile = loadFileUtil(file_name)
          myImageDict = {}
          my_images = glob.glob(image_path +'*.jpg')
          for idx,curr_image in enumerate(my_images):
              curr_image = curr_image.split('/')[-1]
              my_images[idx] = curr_image
```

```
myImageDict[curr_image] = []
          for line in myfile.split('\n'):
              tokens = line.split(',')
              my_image= tokens[0]
              curr_cap=tokens[1:]
              if my_image in my_images:
                  curr_cap = 'startseq ' + ' '.join(curr_cap) + ' endseq'
                  myImageDict[my_image].append(curr_cap)
          return myImageDict
      # Loading the cleaned caption data
      imgNameDict = loadCaptionData('./cleanedCaptionData.txt','../input/flickr8k/

¬Images/')
[94]: # Defining utility function to preprocess the image
      def preprocessingUtil(myPath):
          curr_img = image.load_img(myPath,target_size= (299,299))
          temp= image.img_to_array(curr_img)
          temp = np.expand_dims(temp,axis =0)
          temp = preprocess_input(temp)
          return temp
[95]: random_image = random.choice(list(my_images))
      temp2 = preprocessingUtil('../input/flickr8k/Images/' + random_image)
      for cap in imgNameDict[random_image]:
          print(cap)
      plt.imshow(temp2[0])
      plt.title(random_image)
     startseq people walk outside on wooden walkway endseq
     startseq three woman walk on the sidewalk endseq
     startseq two elderly women are walking past younger woman on public path endseq
     startseq walkers on concrete boardwalk under blue sky endseq
     startseq women walking beneath blue sky and powerlines endseq
[95]: Text(0.5, 1.0, '824123145 59243e504e.jpg')
```



```
[96]: # Loading the InceptionV3 model
     mymodel = InceptionV3(weights = 'imagenet')
[]:
[97]: # Removing the last layer
     featExtractor = Model(mymodel.input,mymodel.layers[-2].output)
     featExtractor.summary()
    Model: "model_2"
    Layer (type)
                               Output Shape
                                               Param #
                                                          Connected to
    ______
    =============
    input_4 (InputLayer)
                              [(None, 299, 299, 3) 0
    conv2d_94 (Conv2D)
                              (None, 149, 149, 32) 864
                                                          input_4[0][0]
    batch_normalization_94 (BatchNo (None, 149, 149, 32) 96 conv2d_94[0][0]
    activation_94 (Activation) (None, 149, 149, 32) 0
```

batch_normalization_94[0][0]				
conv2d_95 (Conv2D) activation_94[0][0]		147, 147, 32)	9216	
batch_normalization_95 (BatchNo	(None,	147, 147, 32)	96 	conv2d_95[0][0]
activation_95 (Activation) batch_normalization_95[0][0]		147, 147, 32)		
conv2d_96 (Conv2D) activation_95[0][0]		147, 147, 64)		
batch_normalization_96 (BatchNo	(None,		192	
activation_96 (Activation) batch_normalization_96[0][0]	(None,	147, 147, 64)	0	
max_pooling2d_4 (MaxPooling2D) activation_96[0][0]	(None,	73, 73, 64)	0	
conv2d_97 (Conv2D) max_pooling2d_4[0][0]	(None,	73, 73, 80)	5120	
batch_normalization_97 (BatchNo				
activation_97 (Activation) batch_normalization_97[0][0]	(None,	73, 73, 80)	0	
conv2d_98 (Conv2D) activation_97[0][0]	(None,	71, 71, 192)	138240	
batch_normalization_98 (BatchNo	(None,	71, 71, 192)	576	conv2d_98[0][0]
activation_98 (Activation)		71, 71, 192)		

batch_normalization_98[0][0]					
max_pooling2d_5 (MaxPooling2D) activation_98[0][0]	(None,				
conv2d_102 (Conv2D) max_pooling2d_5[0][0]	(None,				12288
batch_normalization_102 (BatchN conv2d_102[0][0]	(None,	35,	35,	64)	192
activation_102 (Activation) batch_normalization_102[0][0]	(None,	35,	35,	64)	0
conv2d_100 (Conv2D) max_pooling2d_5[0][0]	(None,				
conv2d_103 (Conv2D) activation_102[0][0]	(None,				
batch_normalization_100 (BatchN conv2d_100[0][0]	(None,	35,			144
batch_normalization_103 (BatchN conv2d_103[0][0]			35,	96)	288
activation_100 (Activation) batch_normalization_100[0][0]	(None,	35,	35,	48)	0
activation_103 (Activation) batch_normalization_103[0][0]	(None,				0
average_pooling2d_9 (AveragePoomax_pooling2d_5[0][0]	(None,	35,	35,	192)	0
conv2d_99 (Conv2D)	(None,				

max_pooling2d_5[0][0]						
conv2d_101 (Conv2D) activation_100[0][0]	(None,	35,	35,	64)	76800	
conv2d_104 (Conv2D) activation_103[0][0]	(None,	35,	35,	96)	82944	
conv2d_105 (Conv2D) average_pooling2d_9[0][0]	(None,				6144	
batch_normalization_99 (BatchNo	(None,	35,	35,	64)	192	conv2d_99[0][0]
batch_normalization_101 (BatchN conv2d_101[0][0]						
batch_normalization_104 (BatchN conv2d_104[0][0]					288	
batch_normalization_105 (BatchN conv2d_105[0][0]	(None,	35,	35,	32)	96	
activation_99 (Activation) batch_normalization_99[0][0]	(None,	35,	35,	64)	0	
activation_101 (Activation) batch_normalization_101[0][0]	(None,					
activation_104 (Activation) batch_normalization_104[0][0]	(None,					
activation_105 (Activation) batch_normalization_105[0][0]	(None,					
mixed0 (Concatenate) activation_99[0][0]	(None,					

activation_101[0][0] activation_104[0][0] activation_105[0][0]						
conv2d_109 (Conv2D)	(None,					mixed0[0][0]
batch_normalization_109 (BatchN conv2d_109[0][0]					192	
activation_109 (Activation) batch_normalization_109[0][0]	(None,				0	
 conv2d_107 (Conv2D)	(None,	35,	35,	48)	12288	mixed0[0][0]
 conv2d_110 (Conv2D) activation_109[0][0]	(None,					
batch_normalization_107 (BatchN conv2d_107[0][0]				48)	144	
batch_normalization_110 (BatchN conv2d_110[0][0]					288	
activation_107 (Activation) batch_normalization_107[0][0]	(None,				0	
activation_110 (Activation) batch_normalization_110[0][0]	(None,	35,	35,	96)	0	
average_pooling2d_10 (AveragePo						mixed0[0][0]
conv2d_106 (Conv2D)	(None,	35,	35,	64)	16384	mixed0[0][0]
conv2d_108 (Conv2D) activation_107[0][0]	(None,	35,	35,	64)		

conv2d_111 (Conv2D) activation_110[0][0]	(None,	35,	35,	96)	82944
conv2d_112 (Conv2D) average_pooling2d_10[0][0]	(None,	35,	35,	64)	16384
batch_normalization_106 (BatchN (conv2d_106[0][0]	(None,	35,	35,	64)	192
batch_normalization_108 (BatchN (conv2d_108[0][0]	(None,	35,	35,	64)	192
batch_normalization_111 (BatchN (conv2d_111[0][0]					
batch_normalization_112 (BatchN (conv2d_112[0][0]	(None,	35,	35,	64)	192
activation_106 (Activation) batch_normalization_106[0][0]					
activation_108 (Activation) batch_normalization_108[0][0]	(None,	35,	35,	64)	0
batch_normalization_111[0][0]	(None,				0
	(None,				0
mixed1 (Concatenate) activation_106[0][0] activation_108[0][0] activation_111[0][0] activation_112[0][0]	(None,	35,	35,	288)	0

conv2d_116 (Conv2D)	(None,	35,	35,	64)	18432	mixed1[0][0]
batch_normalization_116 (BatchN conv2d_116[0][0]	(None,	35,	35,	64)	192	
activation_116 (Activation) batch_normalization_116[0][0]	(None,	35,	35,	64)	0	
conv2d_114 (Conv2D)						mixed1[0][0]
conv2d_117 (Conv2D) activation_116[0][0]	(None,					
batch_normalization_114 (BatchN conv2d_114[0][0]					144	
batch_normalization_117 (BatchN conv2d_117[0][0]					288	
activation_114 (Activation) batch_normalization_114[0][0]	(None,	35,	35,	48)	0	
activation_117 (Activation) batch_normalization_117[0][0]	(None,	35,	35,	96)	0	
average_pooling2d_11 (AveragePo						mixed1[0][0]
 conv2d_113 (Conv2D)	(None,	35,	35,	64)	18432	mixed1[0][0]
conv2d_115 (Conv2D) activation_114[0][0]	(None,	35,	35,	64)	76800	
conv2d_118 (Conv2D) activation_117[0][0]	(None,	35,	35,	96)	82944	
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conv2d_119 (Conv2D) average_pooling2d_11[0][0]	(None,	35,			18432	
batch_normalization_113 (BatchN conv2d_113[0][0]						
batch_normalization_115 (BatchN conv2d_115[0][0]			35,	64)	192	
batch_normalization_118 (BatchN conv2d_118[0][0]	(None,	35,	35,	96)	288	
batch_normalization_119 (BatchN conv2d_119[0][0]					192	
activation_113 (Activation) batch_normalization_113[0][0]	(None,					
activation_115 (Activation) batch_normalization_115[0][0]	(None,	35,	35,	64)	0	
activation_118 (Activation) batch_normalization_118[0][0]	(None,	35,	35,	96)	0	
activation_119 (Activation) batch_normalization_119[0][0]	(None,					
mixed2 (Concatenate) activation_113[0][0] activation_115[0][0] activation_118[0][0] activation_119[0][0]	(None,	35,	35,	288)	0	
conv2d_121 (Conv2D)	(None,	35,	35,	64)	18432	mixed2[0][0]
batch_normalization_121 (BatchN conv2d_121[0][0]	(None,	35,	35,	64)	192	

activation_121 (Activation) batch_normalization_121[0][0]	(None,	35,	35,	64)	0	
 conv2d_122 (Conv2D) activation_121[0][0]	(None,	35,	35,	96)	55296	
batch_normalization_122 (BatchN conv2d_122[0][0]				96)	288	
activation_122 (Activation) batch_normalization_122[0][0]	(None,			96)	0	
conv2d_120 (Conv2D)						mixed2[0][0]
conv2d_123 (Conv2D) activation_122[0][0]	(None,	17,				
batch_normalization_120 (BatchN conv2d_120[0][0]						
batch_normalization_123 (BatchN conv2d_123[0][0]					288	
activation_120 (Activation) batch_normalization_120[0][0]	(None,					
activation_123 (Activation) batch_normalization_123[0][0]	(None,	17,	17,	96)	0	
	(None,	17,	17,	288)	0	mixed2[0][0]
mixed3 (Concatenate) activation_120[0][0] activation_123[0][0] max_pooling2d_6[0][0]	(None,	17,	17,	768)	0	

conv2d_128 (Conv2D)	(None,	17,	17,	128)	98304	mixed3[0][0]
batch_normalization_128 (BatchN conv2d_128[0][0]						
	(None,					
conv2d_129 (Conv2D) activation_128[0][0]					114688	
batch_normalization_129 (BatchN conv2d_129[0][0]	(None,	17,	17,	128)	384	
activation_129 (Activation) batch_normalization_129[0][0]	(None,	17,	17,	128)	0	
conv2d_125 (Conv2D)					98304	mixed3[0][0]
conv2d_125 (Conv2D)	(None,	17,	17,	128)	114688	
conv2d_125 (Conv2D)	(None,	17,	17,	128)	114688	
conv2d_125 (Conv2D)	(None,	17, 17, 17,	17, 17, 17,	128)  128)  128)	114688  384 	
conv2d_125 (Conv2D)	(None,  (None,  (None,	17, 17, 17, 17,	17, 17, 17, 17,	128) 128) 128) 128)	114688 	
conv2d_125 (Conv2D)	(None,  (None,  (None,	17, 17, 17, 17,	17, 17, 17, 17,	128) 128) 128) 128)	114688 	

conv2d_126 (Conv2D) activation_125[0][0]	(None,	17,	17,	128)	114688	
conv2d_131 (Conv2D) activation_130[0][0]	(None,	17,	17,	128)	114688	
batch_normalization_126 (BatchN conv2d_126[0][0]	(None,	17,	17,	128)	384	
batch_normalization_131 (BatchN conv2d_131[0][0]	(None,	17,	17,	128)	384	
activation_126 (Activation) batch_normalization_126[0][0]	(None,	17,				
activation_131 (Activation) batch_normalization_131[0][0]	(None,					
average_pooling2d_12 (AveragePo	(None,	17,	17,	768) 	0	mixed3[0][0]
conv2d_124 (Conv2D)						mixed3[0][0]
conv2d_127 (Conv2D) activation_126[0][0]	(None,	17,	17,	192)	172032	
 conv2d_132 (Conv2D) activation_131[0][0]					172032	
conv2d_133 (Conv2D) average_pooling2d_12[0][0]	(None,	17,	17,	192)	147456	
batch_normalization_124 (BatchN conv2d_124[0][0]	(None,	17,	17,			
batch_normalization_127 (BatchN conv2d_127[0][0]	(None,	17,	17,	192)	576	

batch_normalization_132 (BatchN conv2d_132[0][0]						
batch_normalization_133 (BatchN conv2d_133[0][0]						
activation_124 (Activation) batch_normalization_124[0][0]	(None,	17,	17,	192)	0	
activation_127 (Activation) batch_normalization_127[0][0]	(None,	17,	17,	192)	0	
activation_132 (Activation) batch_normalization_132[0][0]	(None,	17,	17,	192)	0	
activation_133 (Activation) batch_normalization_133[0][0]	(None,	17,	17,	192)	0	
mixed4 (Concatenate) activation_124[0][0] activation_127[0][0] activation_132[0][0] activation_133[0][0]	(None,	17,	17,	768)	0	
conv2d_138 (Conv2D)						mixed4[0][0]
batch_normalization_138 (BatchN conv2d_138[0][0]	(None,	17,	17,	160)	480	
activation_138 (Activation) batch_normalization_138[0][0]	(None,	17,	17,	160)	0	
conv2d_139 (Conv2D) activation_138[0][0]	(None,	17,	17,	160)	179200	
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batch_normalization_139 (BatchN conv2d_139[0][0]						
activation_139 (Activation) batch_normalization_139[0][0]	(None,					
conv2d_135 (Conv2D)	(None,	17,	17,	160)	122880	mixed4[0][0]
conv2d_140 (Conv2D) activation_139[0][0]	(None,					
batch_normalization_135 (BatchN conv2d_135[0][0]	(None,	17,	17,	160)	480	
batch_normalization_140 (BatchN conv2d_140[0][0]	(None,	17,	17,	160)	480	
activation_135 (Activation) batch_normalization_135[0][0]	(None,					
activation_140 (Activation) batch_normalization_140[0][0]	(None,					
conv2d_136 (Conv2D) activation_135[0][0]			17,	160)	179200	
conv2d_141 (Conv2D) activation_140[0][0]	(None,	17,			179200	
batch_normalization_136 (BatchN conv2d_136[0][0]	(None,	17,	17,	160)	480	
batch_normalization_141 (BatchN conv2d_141[0][0]	(None,	17,	17,	160)	480	
activation_136 (Activation)	(None,					

batch_normalization_136[0][0]						
activation_141 (Activation) batch_normalization_141[0][0]	(None,	17,	17,	160)	0	
average_pooling2d_13 (AveragePo					0	
conv2d_134 (Conv2D)	(None,	17,	17,	192)	147456	mixed4[0][0]
conv2d_137 (Conv2D) activation_136[0][0]					215040	
conv2d_142 (Conv2D) activation_141[0][0]					215040	
conv2d_143 (Conv2D) average_pooling2d_13[0][0]	(None,	17,	17,	192)	147456	
batch_normalization_134 (BatchN conv2d_134[0][0]					576	
batch_normalization_137 (BatchN conv2d_137[0][0]					576	
batch_normalization_142 (BatchN conv2d_142[0][0]						
batch_normalization_143 (BatchN conv2d_143[0][0]	(None,	17,	17,	192)	576	
activation_134 (Activation) batch_normalization_134[0][0]	(None,	17,	17,	192)	0	
activation_137 (Activation) batch_normalization_137[0][0]	(None,					

activation_142 (Activation) batch_normalization_142[0][0]	(None,	17,	17,	192)	0	
activation_143 (Activation) batch_normalization_143[0][0]	(None,	17,	17,	192)	0	
mixed5 (Concatenate) activation_134[0][0] activation_137[0][0] activation_142[0][0] activation_143[0][0]	(None,	17,	17,	768)	0	
conv2d_148 (Conv2D)					122880	mixed5[0][0]
batch_normalization_148 (BatchN conv2d_148[0][0]	(None,	17,	17,	160)	480	
activation_148 (Activation) batch_normalization_148[0][0]	(None,					
conv2d_149 (Conv2D) activation_148[0][0]					179200	
batch_normalization_149 (BatchN conv2d_149[0][0]	(None,	17,	17,	160)	480	
activation_149 (Activation) batch_normalization_149[0][0]	(None,					
conv2d_145 (Conv2D)	(None,	17,	17,	160)	122880	mixed5[0][0]
conv2d_150 (Conv2D) activation_149[0][0]					179200	
batch_normalization_145 (BatchN conv2d_145[0][0]						<b></b>

batch_normalization_150 (BatchN conv2d_150[0][0]	(None,	17,	17,	160)	480	
activation_145 (Activation) batch_normalization_145[0][0]	(None,	17,	17,	160)	0	
activation_150 (Activation) batch_normalization_150[0][0]	(None,	17,	17,	160)	0	
conv2d_146 (Conv2D) activation_145[0][0]	(None,	17,	17,	160)	179200	
conv2d_151 (Conv2D) activation_150[0][0]	(None,	17,	17,	160)	179200	
batch_normalization_146 (BatchN conv2d_146[0][0]	(None,	17,	17,	160)	480	
batch_normalization_151 (BatchN conv2d_151[0][0]						
activation_146 (Activation) batch_normalization_146[0][0]	(None,					
activation_151 (Activation) batch_normalization_151[0][0]						
average_pooling2d_14 (AveragePo	(None,	17,	17,	768)	0	mixed5[0][0]
conv2d_144 (Conv2D)						mixed5[0][0]
conv2d_147 (Conv2D) activation_146[0][0]	(None,	17,	17,	192)	215040	

conv2d_152 (Conv2D) activation_151[0][0]	(None,	17,	17,	192)	215040	
conv2d_153 (Conv2D) average_pooling2d_14[0][0]	(None,				147456	
batch_normalization_144 (BatchN conv2d_144[0][0]						
batch_normalization_147 (BatchN conv2d_147[0][0]	(None,	17,	17,	192)	576	
batch_normalization_152 (BatchN conv2d_152[0][0]						
batch_normalization_153 (BatchN conv2d_153[0][0]	(None,	17,	17,	192)	576	
activation_144 (Activation) batch_normalization_144[0][0]	(None,					
activation_147 (Activation) batch_normalization_147[0][0]	(None,	17,	17,	192)	0	
activation_152 (Activation) batch_normalization_152[0][0]	(None,	,	,	,	0	
activation_153 (Activation) batch_normalization_153[0][0]	(None,	17,	•		0	
mixed6 (Concatenate) activation_144[0][0] activation_147[0][0] activation_152[0][0] activation_153[0][0]	(None,	17,				
conv2d_158 (Conv2D)	(None,	17,	17,	192)	147456	mixed6[0][0]

batch_normalization_158 (BatchN conv2d_158[0][0]			17,	192)	576	
activation_158 (Activation) batch_normalization_158[0][0]	(None,	17,	17,	192)	0	
conv2d_159 (Conv2D) activation_158[0][0]	(None,	17,	17,	192)	258048	
batch_normalization_159 (BatchN conv2d_159[0][0]	(None,	17,	17,	192)	576	
activation_159 (Activation) batch_normalization_159[0][0]	(None,	-	-		0	
conv2d_155 (Conv2D)				192)		mixed6[0][0]
conv2d_160 (Conv2D) activation_159[0][0]		-	17,	192)	258048	
batch_normalization_155 (BatchN conv2d_155[0][0]			17,	192)	576	
batch_normalization_160 (BatchN conv2d_160[0][0]						
activation_155 (Activation) batch_normalization_155[0][0]	(None,	17,	17,	192)	0	
activation_160 (Activation) batch_normalization_160[0][0]	(None,		17,	192)		
conv2d_156 (Conv2D) activation_155[0][0]	(None,	17,			258048	
	<b></b>				<b>_</b>	

conv2d_161 (Conv2D) activation_160[0][0]	(None,	17,	17,	192)	258048	
batch_normalization_156 (BatchN conv2d_156[0][0]	(None,	17,	17,	192)	576	
batch_normalization_161 (BatchN conv2d_161[0][0]	(None,	17,	17,	192)	576	
activation_156 (Activation) batch_normalization_156[0][0]	(None,	17,	17,	192)	0	
activation_161 (Activation) batch_normalization_161[0][0]	(None,	17,	17,	192)	0	
average_pooling2d_15 (AveragePo	(None,	17,			0	
 conv2d_154 (Conv2D)						mixed6[0][0]
conv2d_157 (Conv2D) activation_156[0][0]					258048	
conv2d_162 (Conv2D) activation_161[0][0]	(None,	17,	17,	192)	258048	
conv2d_163 (Conv2D) average_pooling2d_15[0][0]					147456	
batch_normalization_154 (BatchN conv2d_154[0][0]						
batch_normalization_157 (BatchN conv2d_157[0][0]						
batch_normalization_162 (BatchN					576	<b>_</b>

conv2d_162[0][0]						
batch_normalization_163 (BatchN conv2d_163[0][0]	(None,	17,	17,	192)	576	
activation_154 (Activation) batch_normalization_154[0][0]	(None,	17,	17,	192)	0	
activation_157 (Activation) batch_normalization_157[0][0]	(None,					
activation_162 (Activation) batch_normalization_162[0][0]	(None,					
activation_163 (Activation) batch_normalization_163[0][0]	(None,					
mixed7 (Concatenate) activation_154[0][0] activation_157[0][0] activation_162[0][0] activation_163[0][0]	(None,	17,	17,	768)	0	
 conv2d_166 (Conv2D)	(None,	17,	17,	192)	147456	mixed7[0][0]
batch_normalization_166 (BatchN conv2d_166[0][0]	(None,	17,			576	
activation_166 (Activation) batch_normalization_166[0][0]	(None,					
 conv2d_167 (Conv2D) activation_166[0][0]	(None,	17,	17,	192)	258048	
batch_normalization_167 (BatchN conv2d_167[0][0]	(None,	17,	17,	192)	576	

activation_167 (Activation) batch_normalization_167[0][0]	(None,	17, 17, 192)	0	
conv2d_164 (Conv2D)	(None,	17, 17, 192)	147456	mixed7[0][0]
conv2d_168 (Conv2D) activation_167[0][0]	(None,	17, 17, 192)	258048	
batch_normalization_164 (BatchN conv2d_164[0][0]	(None,	17, 17, 192)	576	
batch_normalization_168 (BatchN conv2d_168[0][0]				
activation_164 (Activation) batch_normalization_164[0][0]		17, 17, 192)	0	
	(None,	17, 17, 192)		
conv2d_165 (Conv2D) activation_164[0][0]	(None,	8, 8, 320)	552960	
conv2d_169 (Conv2D) activation_168[0][0]	(None,	8, 8, 192)	331776	
batch_normalization_165 (BatchN conv2d_165[0][0]			960	
batch_normalization_169 (BatchN conv2d_169[0][0]		8, 8, 192)	576	
activation_165 (Activation) batch_normalization_165[0][0]	(None,	8, 8, 320)	0	

activation_169 (Activation) batch_normalization_169[0][0]	(None,				0	
max_pooling2d_7 (MaxPooling2D)		8,	8,	768)	0	mixed7[0][0]
mixed8 (Concatenate) activation_165[0][0] activation_169[0][0] max_pooling2d_7[0][0]	(None,				0	
conv2d_174 (Conv2D)	(None,	8,	8,	448)	573440	mixed8[0][0]
batch_normalization_174 (BatchN conv2d_174[0][0]						
activation_174 (Activation) batch_normalization_174[0][0]	(None,	8,	8,	448)	0	
conv2d_171 (Conv2D)	(None,	8,	8,			mixed8[0][0]
conv2d_175 (Conv2D) activation_174[0][0]				384)	1548288	
batch_normalization_171 (BatchN conv2d_171[0][0]				384)	1152	
batch_normalization_175 (BatchN conv2d_175[0][0]	(None,	8,	8,	384)	1152	
activation_171 (Activation) batch_normalization_171[0][0]				384)	0	
activation_175 (Activation) batch_normalization_175[0][0]		8,	8,	384)	0	
conv2d_172 (Conv2D)	(None,	8,	8,	384)	442368	

activation_171[0][0]			
conv2d_173 (Conv2D) activation_171[0][0]	(None, 8, 8, 384)	442368	
conv2d_176 (Conv2D) activation_175[0][0]	(None, 8, 8, 384)	442368	
conv2d_177 (Conv2D) activation_175[0][0]	(None, 8, 8, 384)		
average_pooling2d_16 (AveragePo	(None, 8, 8, 1280)	0	mixed8[0][0]
_	(None, 8, 8, 320)	409600	
batch_normalization_172 (BatchN conv2d_172[0][0]	(None, 8, 8, 384)	1152	
batch_normalization_173 (BatchN conv2d_173[0][0]	(None, 8, 8, 384)	1152	
batch_normalization_176 (BatchN conv2d_176[0][0]	(None, 8, 8, 384)	1152	
batch_normalization_177 (BatchN conv2d_177[0][0]		1152	
conv2d_178 (Conv2D) average_pooling2d_16[0][0]	(None, 8, 8, 192)		
batch_normalization_170 (BatchN conv2d_170[0][0]	(None, 8, 8, 320)	960	
	(None, 8, 8, 384)		

activation_173 (Activation) batch_normalization_173[0][0]	(None,	8,	8,	384)	0	
activation_176 (Activation) batch_normalization_176[0][0]	(None,	8,	8,	384)	0	
activation_177 (Activation) batch_normalization_177[0][0]	(None,	8,	8,	384)	0	
batch_normalization_178 (BatchN conv2d_178[0][0]	(None,	8,	8,	192)	576	
activation_170 (Activation) batch_normalization_170[0][0]	(None,	8,	8,	320)	0	
mixed9_0 (Concatenate) activation_172[0][0] activation_173[0][0]	(None,	8,	8,	768)	0	
concatenate_2 (Concatenate) activation_176[0][0] activation_177[0][0]	(None,	8,	8,	768)	0	
activation_178 (Activation) batch_normalization_178[0][0]	(None,	8,	8,	192)	0	
mixed9 (Concatenate) activation_170[0][0]  concatenate_2[0][0] activation_178[0][0]				2048)	0	mixed9_0[0][0]
 conv2d_183 (Conv2D)						mixed9[0][0]
batch_normalization_183 (BatchN conv2d_183[0][0]	(None,	8,	8,	448)	1344	

activation_183 (Activation) batch_normalization_183[0][0]	(None,	8,	8,	448)	0	
conv2d_180 (Conv2D)	(None,	8,	8,	384)	786432	mixed9[0][0]
 conv2d_184 (Conv2D) activation_183[0][0]	(None,	8,	8,	384)	1548288	
batch_normalization_180 (BatchN conv2d_180[0][0]	(None,	8,	8,	384)	1152	
batch_normalization_184 (BatchN conv2d_184[0][0]	(None,	8,	8,	384)	1152	
activation_180 (Activation) batch_normalization_180[0][0]	(None,	8,	8,	384)	0	
activation_184 (Activation) batch_normalization_184[0][0]	(None,	8,	8,	384)	0	
conv2d_181 (Conv2D) activation_180[0][0]	(None,	8,	8,	384)	442368	
conv2d_182 (Conv2D) activation_180[0][0]	(None,	8,	8,	384)	442368	
conv2d_185 (Conv2D) activation_184[0][0]				384)		
conv2d_186 (Conv2D) activation_184[0][0]	(None,	8,	8,	384)	442368	
average_pooling2d_17 (AveragePo	(None,	8,	8,	2048)	0	mixed9[0][0]
conv2d_179 (Conv2D)					655360	

batch_normalization_181 (BatchN conv2d_181[0][0]	(None, 8, 8, 384)	1152
batch_normalization_182 (BatchN conv2d_182[0][0]	(None, 8, 8, 384)	1152
batch_normalization_185 (BatchN conv2d_185[0][0]		1152
batch_normalization_186 (BatchN conv2d_186[0][0]	(None, 8, 8, 384)	1152
conv2d_187 (Conv2D) average_pooling2d_17[0][0]	(None, 8, 8, 192)	393216
batch_normalization_179 (BatchN conv2d_179[0][0]	(None, 8, 8, 320)	960
activation_181 (Activation) batch_normalization_181[0][0]	(None, 8, 8, 384)	0
activation_182 (Activation) batch_normalization_182[0][0]	(None, 8, 8, 384)	0
activation_185 (Activation) batch_normalization_185[0][0]	(None, 8, 8, 384)	0
activation_186 (Activation) batch_normalization_186[0][0]	(None, 8, 8, 384)	0
batch_normalization_187 (BatchN conv2d_187[0][0]		576
activation_179 (Activation) batch_normalization_179[0][0]	(None, 8, 8, 320)	0

```
mixed9_1 (Concatenate)
                             (None, 8, 8, 768) 0
     activation_181[0][0]
     activation 182[0][0]
     concatenate_3 (Concatenate) (None, 8, 8, 768)
     activation_185[0][0]
     activation_186[0][0]
     activation_187 (Activation) (None, 8, 8, 192)
     batch_normalization_187[0][0]
     mixed10 (Concatenate)
                                  (None, 8, 8, 2048) 0
     activation_179[0][0]
                                                                  mixed9_1[0][0]
     concatenate 3[0][0]
     activation_187[0][0]
                                                                mixed10[0][0]
     avg_pool (GlobalAveragePooling2 (None, 2048)
     ============
     Total params: 21,802,784
     Trainable params: 21,768,352
     Non-trainable params: 34,432
     ______
[98]: # Performing image embedding with featExtractor
      def encodeData(curr_image,featExtractor):
          curr_image = preprocessingUtil(curr_image)
          featVector = featExtractor.predict(curr_image)
          featVector = np.ravel(featVector)
          return featVector
[108]: | # Dividing the data into training, validation, and testing data
      AllData = list(my_images)
      trainingData = AllData[:6000]
      validationData = AllData[6000:7000]
      TestingData = AllData[7000:]
      print(f'Number of training images {len(trainingData)}')
      print(f'Number of testing images {len(TestingData)}')
      print(f'Number of validation images {len(validationData)}')
```

```
Number of training images 6000
      Number of testing images 1091
      Number of validation images 1000
[109]: trainingEnc = {}
       validationEnc = {}
       testingEnc = {}
[110]: # Encoding the training data
       for x in range(1,len(trainingData)):
           myloc = '../input/flickr8k/Images/' + trainingData[x]
           trainingEnc[trainingData[x]] = encodeData(myloc,featExtractor)
[111]: # Storing the encoded training images
       with open('./training_images_enc.pkl','wb') as pickEnc:
           dump(trainingEnc,pickEnc)
[112]: # Similarly we encode and store the validation and testing data
       for curr_img in tqdm(validationData):
           validationEnc[curr_img] = encodeData('../input/flickr8k/Images/' +_
        ⇔curr_img,featExtractor)
      100%|
                | 1000/1000 [01:07<00:00, 14.88it/s]
[113]: with open('./validation_images_enc.pkl','wb') as pickEnc:
           dump(validationEnc,pickEnc)
[114]: for curr_img in tqdm(TestingData):
           testingEnc[curr_img] = encodeData('../input/flickr8k/Images/' +__
        ⇔curr_img,featExtractor)
      100%1
                | 1091/1091 [01:12<00:00, 15.00it/s]
[115]: with open('./testing_images_enc.pkl','wb') as pickEnc:
           dump(testingEnc,pickEnc)
[116]: |cp ./testing_images_enc.pkl /content/drive/MyDrive
      cp: cannot create regular file '/content/drive/MyDrive': No such file or
      directory
[117]: # Loading the features
       featTrain = load(open('./training_images_enc.pkl','rb'))
       featValid = load(open('./validation_images_enc.pkl','rb'))
       featTest = load(open('./testing_images_enc.pkl','rb'))
       print(f'The number of training features {len(featTrain)}')
       print(f'The number of testing features {len(featTest)}')
```

```
print(f'The number of validation features {len(featValid)}')
      The number of training features 5999
      The number of testing features 1091
      The number of validation features 1000
[118]: # Creating separate lists for storing the training, testing, and validation
        \hookrightarrow captions
       allCaptionsTrain = []
       allCaptionsVal = []
       allCaptionsTest = []
       for x in range(1,len(trainingData)):
           curr_img=trainingData[x]
           captions = imgNameDict[curr_img]
           for mycap in captions:
               allCaptionsTrain.append(mycap)
       print(f'Number of all training captions {len(allCaptionsTrain)}')
      Number of all training captions 29995
[119]: for curr_img in validationData:
           captions = imgNameDict[curr_img]
           for mycap in captions:
               allCaptionsVal.append(mycap)
       print(f'Number of all validation captions {len(allCaptionsVal)}')
      Number of all validation captions 5000
[120]: for img in TestingData:
           captions = imgNameDict[img]
           for mycap in captions:
               allCaptionsTest.append(mycap)
       print(f'Number of all test captions {len(allCaptionsTest)}')
      Number of all test captions 5455
[121]: wc = {}
       for mysentence in allCaptionsTrain:
           for curr_word in mysentence.split(' '):
               wc[curr_word] = wc.get(curr_word,0) + 1
       myvocab = [curr_word for curr_word in wc if wc[curr_word] >= 10]
[122]: id_word = {}
       word id = \{\}
       idx = 1
       for curr_word in myvocab:
           word_id[curr_word] = idx
           id_word[idx] = curr_word
```

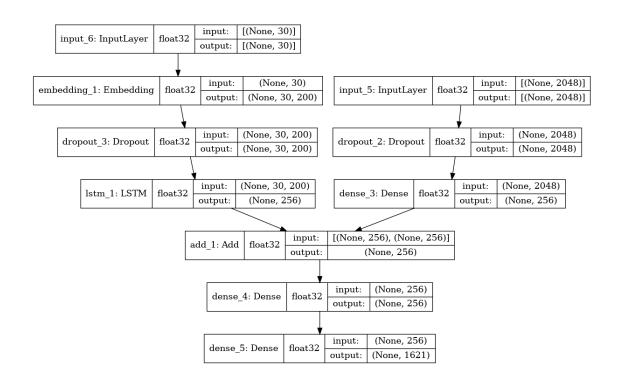
```
idx +=1
[123]: with open('./word_id.pkl','wb') as word_id1:
           dump(word_id,word_id1)
[124]: with open('./id_word.pkl','wb') as id_word1:
           dump(id_word,id_word1)
[125]: vocabSize = len(id_word) +1
       vocabSize
[125]: 1621
[126]: trainDict = {}
       for x in range(1,len(trainingData)):
           my_image=trainingData[x]
           trainDict[my_image] = imgNameDict[my_image]
[127]: # Defining function to find maximum length of captions
       def maxFun(imageCaptions):
           lines = to_lines(imageCaptions)
           return max(len(line.split()) for line in lines)
       # Defining function which accepts dictionary of clean captions as input
       # Returns captions list
       def to lines(imageCaptions):
           allCaptions = []
           for x in imageCaptions.keys():
               [allCaptions.append(mycap) for mycap in imageCaptions[x]]
           return allCaptions
       maxLength = maxFun(trainDict)
       print(f'The captions maximum length is {maxLength}')
      The captions maximum length is 30
[128]: # Loading GloVe
       embeddingsIndex = {}
       myfile = open('../input/glove6b/glove.6B.200d.txt',encoding = 'utf-8')
       for x in tqdm(myfile):
           vals = x.split()
           curr_word = vals[0]
           coefficients = np.asarray(vals[1:],dtype = 'float32')
           embeddingsIndex[curr_word] = coefficients
       myfile.close()
       print('Number of word vectors is %s' % len(embeddingsIndex))
```

```
[129]: embedDim = 200
[130]: # Creating the embedding matrix
      embedMat = np.zeros((vocabSize,embedDim))
      print(f'The embedding matrix dimensions are {embedMat.shape}')
     The embedding matrix dimensions are (1621, 200)
[131]: for curr_word,idx in word_id.items():
          embedVec = embeddingsIndex.get(curr_word)
          if embedVec is not None:
              embedMat[idx] = embedVec
 []:
 []:
[132]: # Building the model
      inp1 = Input(shape = (2048,))
      fe1 = Dropout(0.5)(inp1)
      fe2 = Dense(256,activation = 'relu')(fe1)
      inp2 = Input(shape = (maxLength,))
      se1 = Embedding(vocabSize,embedDim,mask_zero = True)(inp2)
      se2 = Dropout(0.5)(se1)
      se3 = LSTM(256)(se2)
      dec1 = add([fe2,se3])
      dec2 = Dense(256,activation = 'relu')(dec1)
      out = Dense(vocabSize,activation = 'softmax')(dec2)
[133]: mymodel2 = Model(inputs = [inp1,inp2],outputs = out)
      mymodel2.summary()
     Model: "model 3"
     Layer (type)
                                    Output Shape
                                                       Param #
                                                                  Connected to
      ______
     =============
     input_6 (InputLayer)
                                    [(None, 30)]
     input_5 (InputLayer)
                                   [(None, 2048)]
```

400000it [00:18, 21609.09it/s]

Number of word vectors is 400000

	embedding_1 (Embedding)	(None,	30, 200)	324200	input_6[0][0]		
	dropout_2 (Dropout)	(None,	2048)	0	input_5[0][0]		
	dropout_3 (Dropout) embedding_1[0][0]		30, 200)				
	dense_3 (Dense)	(None,	256)	524544	dropout_2[0][0]		
	lstm_1 (LSTM)	(None,	256)	467968	dropout_3[0][0]		
	add_1 (Add)		256)		dense_3[0][0] lstm_1[0][0]		
	dense_4 (Dense)		256)		_		
	dense_5 (Dense)				=		
,	Total params: 1,799,101 Trainable params: 1,799,101 Non-trainable params: 0						
34]:	<pre># Displaying the model layers plot_model(mymodel2,</pre>	_					



```
[135]: print(f'Embedding Matrix Shape {embedMat.shape}')
      embedWeight = mymodel2.layers[2].get_weights()
      print(f'Embedding Weight Shape {embedWeight[0].shape}')
      mymodel2.layers[2].set_weights([embedMat])
      mymodel2.layers[2].trainable = False
      Embedding Matrix Shape (1621, 200)
      Embedding Weight Shape (1621, 200)
[136]: mymodel2.summary()
      Model: "model_3"
      Layer (type)
                                     Output Shape
                                                          Param #
                                                                      Connected to
      _____
      input_6 (InputLayer)
                                     [(None, 30)]
                                     [(None, 2048)]
      input_5 (InputLayer)
                                                                      input_6[0][0]
                                    (None, 30, 200)
                                                          324200
      embedding_1 (Embedding)
```

```
(None, 2048) 0
     dropout_2 (Dropout)
                                                          input_5[0][0]
     dropout_3 (Dropout)
                               (None, 30, 200) 0
     embedding_1[0][0]
     dense_3 (Dense)
                              (None, 256) 524544 dropout_2[0][0]
     lstm_1 (LSTM)
                               (None, 256) 467968 dropout_3[0][0]
                               (None, 256) 0
     add_1 (Add)
                                                           dense_3[0][0]
                                                           lstm_1[0][0]
     dense_4 (Dense)
                               (None, 256)
                                           65792 add_1[0][0]
     ______
     dense 5 (Dense)
                               (None, 1621) 416597 dense_4[0][0]
     _____
     Total params: 1,799,101
     Trainable params: 1,474,901
     Non-trainable params: 324,200
[137]: | mymodel2.compile(optimizer = 'adam',loss = 'categorical_crossentropy')
[138]: epochs = 10
     steps = len(trainDict)//6
     mymodel2.optimizer.lr = 1e-4
 []:
[139]: def data_generator(trainDict,featTrain,word_id,maxLength,pics_p_batch):
         V = []
         n = 0
         X1 = []
         X2=[]
         while True:
            for curr_image,captionsList in trainDict.items():
               myfeatVect = featTrain[curr_image]
```

```
for mycap in captionsList:
                     encoded_cap = [word_id[word] for word in mycap.split(' ') if_
       →word in word_id]
                     for idx in range(1,len(encoded cap)):
                        in_seq = encoded_cap[:idx]
                        out seq = encoded cap[idx]
                        in_seq = pad_sequences([in_seq],maxlen = maxLength)[0]
                        out_seq = to_categorical([out_seq],num_classes =__
       yocabSize)[0]
                        y.append(out_seq)
                        X1.append(myfeatVect)
                        X2.append(in seq)
                 if n == pics_p_batch:
                    X1,X2,y = np.array(X1),np.array(X2),np.array(y)
                    yield [X1,X2],y
                    y = []
                    n=0
                    X1 = []
                    X2 = []
     '3448855727_f16dea7b03.jpg' in featTrain
[140]:
[140]: True
[141]: for curr_epoch in range(epochs):
          genvar = data_generator(trainDict, featTrain, word_id, maxLength, 6)
         mymodel2.fit_generator(genvar, epochs=1, steps_per_epoch=steps, verbose=1)
     /opt/conda/lib/python3.7/site-
     packages/tensorflow/python/keras/engine/training.py:1844: UserWarning:
     `Model.fit_generator` is deprecated and will be removed in a future version.
     Please use `Model.fit`, which supports generators.
       warnings.warn('`Model.fit_generator` is deprecated and '
     999/999 [========== ] - 113s 109ms/step - loss: 5.5713
     999/999 [========== ] - 111s 111ms/step - loss: 4.4140
     999/999 [========= ] - 117s 117ms/step - loss: 4.0736
     999/999 [============ ] - 108s 108ms/step - loss: 3.8797
     999/999 [=========== ] - 110s 110ms/step - loss: 3.7496
     999/999 [=========== ] - 117s 118ms/step - loss: 3.6473
     999/999 [=========== ] - 108s 108ms/step - loss: 3.5669
     999/999 [=========== ] - 107s 107ms/step - loss: 3.4978
     999/999 [========== ] - 117s 117ms/step - loss: 3.4407
     999/999 [=========== ] - 115s 115ms/step - loss: 3.3863
```

```
[142]: # Saving the model
       mymodel2.save('model_flickr8k')
       # Saving the model weights
       mymodel2.save_weights('./model_flickr8k.h5')
  []:
[143]: |zip -r model_flickr8k.zip model_flickr8k
      updating: model_flickr8k/ (stored 0%)
      updating: model_flickr8k/saved_model.pb (deflated 90%)
      updating: model_flickr8k/variables/ (stored 0%)
      updating: model_flickr8k/variables/variables.data-00000-of-00001 (deflated 7%)
      updating: model_flickr8k/variables/variables.index (deflated 64%)
      updating: model_flickr8k/assets/ (stored 0%)
[144]: |#To obtain the predicted image caption, the image is passed to the model with
        ⇔the input string "startseg".
       # Then the model predicts the next word and this word is appended to the input_{\sqcup}
        ⇔string. This repeats until
       # "endseq" is reached or the maximum sentence length is reached.
       def greedy_search(featVect, verbose = 0):
           in_text ='startseq'
           for i in range(maxLength):
               sequence = [word_id[x] for x in in_text.split() if x in word_id]
               sequence = pad_sequences([sequence],maxlen = maxLength)
               yhat = mymodel2.predict([featVect,sequence],verbose = verbose) #__
        \hookrightarrow [(1,2048),(1,31)]
               yhat = np.argmax(yhat)
               curr_word = id_word[yhat]
               in_text += ' ' + curr_word
               if curr_word == 'endseq':
                   break
           res = in_text.split()
           res = res[1:-1]
           res = ' '.join(res)
           return res
       img = random.choice(TestingData)
       featVect = featTest[img].reshape((1,2048))
       print(f'feature_vector: {featVect.shape}')
       final = greedy_search(featVect,1)
       final
```

feature\_vector: (1, 2048)

```
1/1 [======= ] - 1s 1s/step
     1/1 [======] - Os 21ms/step
     1/1 [=======] - 0s 20ms/step
     1/1 [=======] - 0s 20ms/step
                                ==] - Os 21ms/step
                                 =] - 0s 22ms/step
                                 =] - 0s 20ms/step
     1/1 [======] - 0s 22ms/step
                                ==] - Os 23ms/step
[144]: 'woman wearing red shirt and glasses and glasses'
[145]: img = random.choice(TestingData)
     feature_vector = featTest[img].reshape((1,2048))
     x = plt.imread('../input/flickr8k/Images/' + img)
     plt.imshow(x)
     plt.title(greedy_search(feature_vector,0))
     plt.show()
```



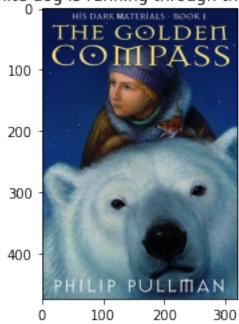


```
[146]: # Performing image captioning for a single sample image from the dataset
    img = 'fantasy20.jpg'
    testingEnc[img] = encodeData('../input/sampleimage/' + img,featExtractor)

[147]: feature_vector = testingEnc[img].reshape((1,2048))
    x = plt.imread('../input/sampleimage/fantasy20.jpg')
```

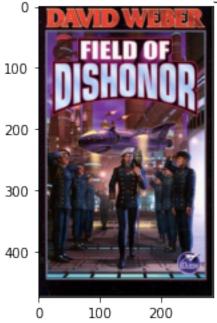
```
plt.imshow(x)
plt.title(greedy_search(feature_vector,0))
plt.show()
```

## white dog is running through the snow

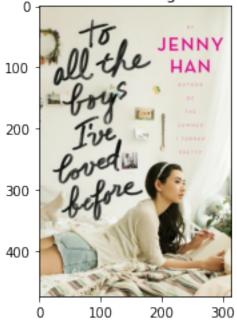


```
[148]: # Performing image captioning for several sample images from the dataset
for myfile in os.listdir('../input/sampleimage/'):
    img = myfile
    encoded_data = encodeData('../input/sampleimage/' + img,featExtractor)
    feature_vector = encoded_data.reshape((1,2048))
    x = plt.imread('../input/sampleimage/'+img)
    plt.imshow(x)
    plt.title(greedy_search(feature_vector,0))
    plt.show()
```

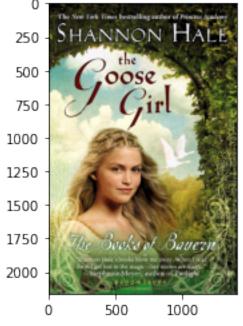
man in red shirt and red shirt is standing in front of crowd



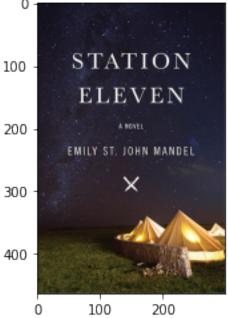
two women are sitting on the floor



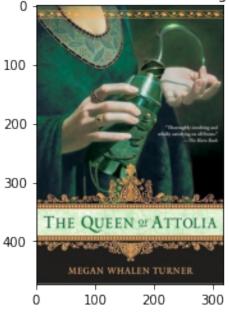
woman in red shirt and white shirt is sitting on the sidewalk



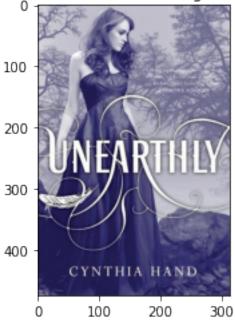
man in red jacket and white hat is sitting on the beach



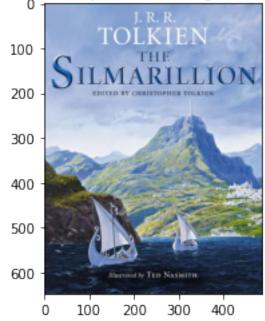
woman in black shirt and black shirt is standing in front of the camera



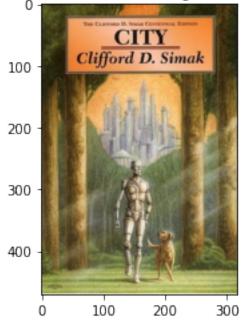
woman in black dress is sitting on the beach



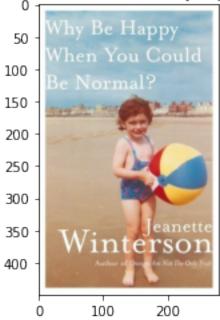
man in white jacket is sitting on the beach



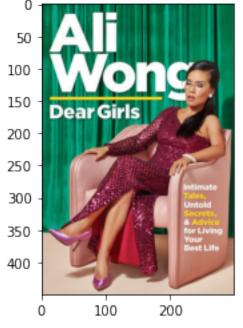
man and woman are sitting on the ground



woman in black and white shirt is jumping into the air



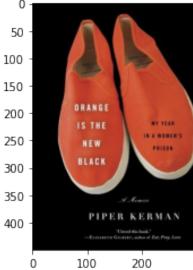
woman in black dress and white shirt is sitting on the street



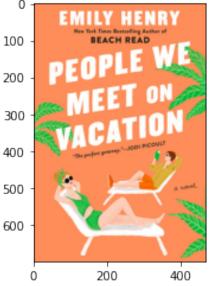
man in red shirt is standing in front of the camera



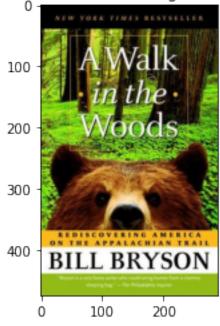
woman wearing red shirt and black shirt and white shirt is sitting on the ground



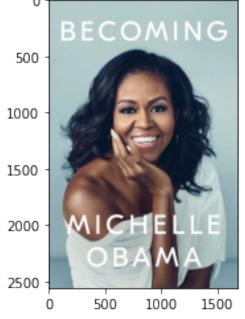
woman wearing red shirt and white shirt and white shirt is holding her head



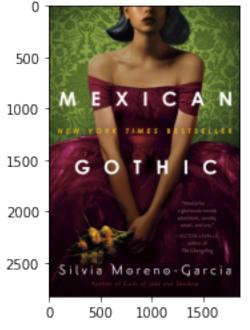
man in red shirt is sitting on the grass



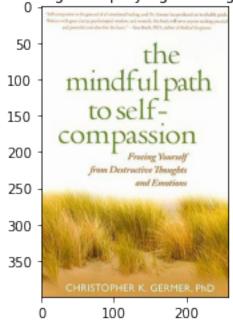
woman wearing sunglasses and sunglasses is sitting on the camera



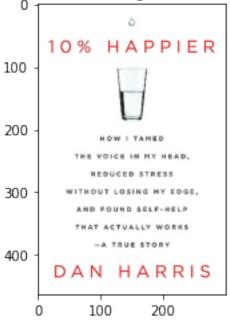
woman in black dress and black hat and woman in black dress

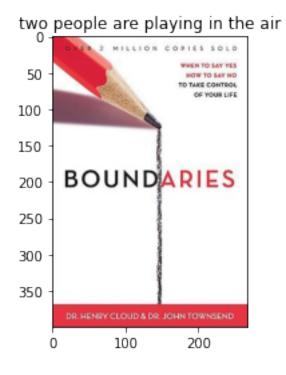




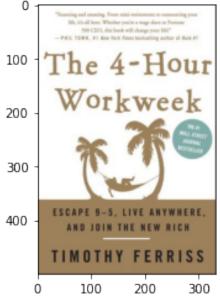


## man wearing black shirt and sunglasses is sitting on the camera

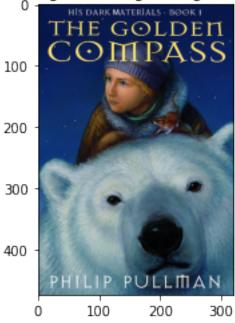




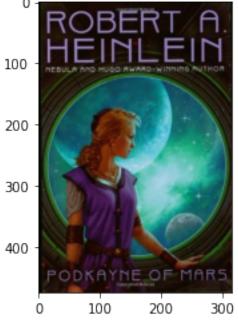
man in black shirt and black shirt and white shirt is sitting on the ground



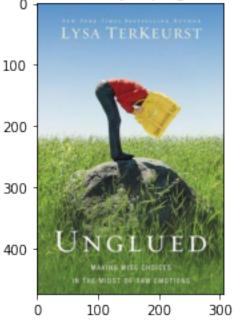
white dog is running through the snow



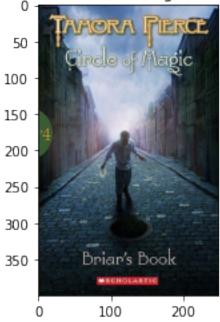
man in red shirt and red shirt is sitting on the street



boy in red shirt is jumping over the air



man in red shirt is standing on the sidewalk



woman wearing red shirt and sunglasses and white shirt and whi

