Report On

Image Caption Generator

Submitted in partial fulfillment of the requirements of the Course project in Semester VII of Fourth Year Artificial Intelligence & Data Science Engineering

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(A.Y. 2023-24)

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CERTIFICATE

This is to certify that the project entitled "Image Caption Generator" is a bonafide work of "Devashree Pawar (Roll No. 21), Naina Roy (Roll No. 24), Khushi Upadhyay (Roll No. 30)" submitted to the University of Mumbai in partial fulfillment of the requirement for the Course project in semester VII of Fourth Year Artificial Intelligence and Data Science engineering.

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Abstract

An image caption generator is an artificial intelligence (AI) system that uses computer vision and natural language processing (NLP) techniques to generate captions or descriptions for images. The system typically analyses the visual content of an image and generates a corresponding textual description that captures the key objects, attributes, and relationships in the scene. The development of image caption generators has been driven by the growing demand for automated image understanding and annotation in various applications, such as image retrieval, multimedia indexing, and assistive technologies for the visually impaired. The underlying technology relies on deep learning models, such as convolutional neural networks (CNNs) for image analysis and recurrent neural networks (RNNs) for language generation. The image caption generation process typically involves several steps, including image preprocessing, feature extraction, sequence modeling, and decoding. The system first pre-processes the input image to normalize its size and color, and then extracts high-level features using a pretrained CNN. These features are then fed into an RNN to generate a sequence of words that forms the caption. The RNN can be trained using various techniques, such as maximum likelihood estimation or reinforcement learning, to optimize the caption quality and coherence.

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1.1 Problem Statement:

The rapid growth of digital images and videos has resulted in a vast amount of visual data that needs to be analyzed and understood. However, analyzing and interpreting visual data is a complex and challenging task that requires human-level intelligence and expertise.

Therefore, there is a need for automated solutions that can extract meaningful information from visual data and facilitate their understanding and utilization.

2.1 BLOCK DIAGRAM

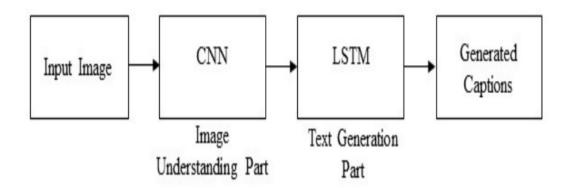


Fig 1: Block Diagram of basic image caption generation

The above figure shows a basic block diagram of our project image caption generator. The project typically consists of two parts: encoder and decoder

The encoder component uses a pre-trained CNN model to extract features from the input image. The CNN model is typically trained on a large dataset such as ImageNet and is able to extract high-level features from the image such as objects, shapes, and textures. The output of the CNN model is a fixed-length vector of feature values, which captures the visual content of the image.

The decoder component uses an LSTM model to generate the caption for the input image. The LSTM model takes in the feature vector generated by the encoder as its initial input, and then generates the caption word by word, based on the previous work in the sequence. The LSTM model learns the relationships between the words in the caption and generates a sequence of words that are coherent and relevant to the input image.

The CNN-LSTM model is a widely used approach for image caption generation as it can effectively learn the visual features of the image and can give more natural-sounding captions.

Module Description:

The process design for the image caption generator with GUI project can be divided into several stages:

- 1. Planning and Requirements Gathering: This stage involves understanding the project requirements, identifying the user needs, and planning the overall architecture of the system.
- 2. Data Collection and Preprocessing: This stage involves collecting and preprocessing the data needed to train the model. This can include finding and downloading image datasets, annotating the images with captions, and preprocessing the images to ensure they are of consistent quality. The dataset used by us is Flicker30k data. It is of size 9GB and has around 16,000 images and nearly 5 captions for each image. The preprocessing step also involves cleaning the caption data and preprocessing all the captions and tokenizing the caption dataset.
- 3. Model Training: This stage involves training the model using the pre-processed data. The model will typically consist of a combination of the Inception V3 model, an architecture for convolutional neural networks (CNNs) for image processing, and LSTM, a technique of recurrent neural networks (RNNs) for caption generation. The model can be trained using Keras, a deep-learning library that simplifies the process of building and training neural networks.
- 4. Model Evaluation: This stage involves evaluating the performance of the trained model using various metrics such as accuracy, perplexity, and BLEU score. This is important to ensure that the model is generating accurate captions.
- 5. GUI Development: This stage involves designing and developing the graphical user interface (GUI) that will allow users to interact with the image caption generator. The GUI should be intuitive, user-friendly, and responsive, allowing users to upload an image, and adjust input image settings, and caption settings. Our GUI is simple and is built by using the Tkinter library provided by Python. In the GUI, we can upload an image by browsing through our laptop and then click on generate caption to generate the caption. The caption generated is through two searching algorithms namely Beam search and greedy search which are a part of Artificial Intelligence.

- 6. System Integration: This stage involves integrating the various components of the system, such as the pre-processing module, feature extraction module, caption generation module, and GUI. This involves programming, testing, and optimizing the system to ensure that it works seamlessly.
- 7. Documentation: This stage involves creating user guides, technical documents, and other documentation to facilitate the use and development of the system. This documentation should be comprehensive and accessible to all stakeholders, including developers, users, and other interested parties.

2.2 DETAILS OF HARDWARE AND SOFTWARE

***** Hardware:

- > A personal computer
- ➤ Windows 10/11
- > CPU 8 GB or more
- > GPU 2 GB or more

Software:

- > VS code
- > Jupyter Notebook
- > Python

3.1 Code:

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D

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                🌞 gui,py 4, M + Code + Markdown | ▶ Run All 🗮 Clear All Outputs | 🗏 Outline …
         X ■ image-captionin...

Import numpy as np # linear algebra

import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
            > images

≣ caption-img.pkl
            image-captioning.ipy...

inception_v3_weights...

▷ ∨
                                                               from keras.preprocessing.sequence import pad_sequences
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.text import Tokenizer
from keras.preprocessing.text import Model
from keras.layers import Flatten, Dense, LSTM, Dropout, Embedding, Activation
from keras.layers import concatenate, BatchNormalization, Input
from keras.layers.merge import add
from keras.layers.merge import add

    ixtoword.pkl

    new-model.h5

    README.md

                                                                from keras.utils import to_categorical
from keras.applications.inception_v3 import InceptionV3, preprocess_input

    wordtoix.pkl

                                                                from keras.utils import plot_model
                                                                import string import time
> OUTLINE
```

Fig 2. Importing packages

```
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O
                                                                                                                                                                                                                                                                     def clean_description(desc):
                                               for key, des_list in desc.items():
    for i in range(len(des_list)):
                                                                                                                                                                                                                                                                                                                                                                  caption = des_list[i]
caption = [ch for ch in caption if ch not in string.punctuation]
                                                                                                                                                                                                                                                                                                                                                                    caption = ''.join(caption)
caption = caption.split(' ')
caption = [word.lower() for word in caption if len(word)>1 and word.isalpha()]
caption = ' '.join(caption)

≡ image-caption-30k-3...
                                                   image-captioning.ipy...

inception_v3_weights...

inception_v3_weights...
                                                                                                                                                                                                                                                                                                                                                                          des_list[i] = caption
                                                   ≣ ixtoword.pkl

    LICENSE.md

                                                   E new-model-1.h5

E new-model.h5
                                                                                                                                                                                                                                                                           clean_description(descriptions)
descriptions['1000268201_693b08cb0e']
                                           # READMEND

(a) READMEND

(b) READMEND

(c) READMEND

(c)
                                                                                                                                                                                                                                                                               def to_vocab(desc):
    words = set()
                                       > OUTLINE
```

Fig 3. Preprocessing data

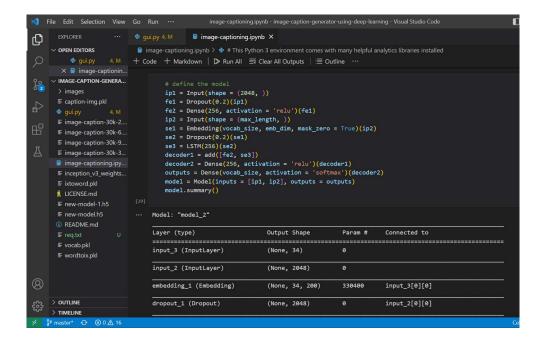


Fig 4. Define model

```
image-captioning.ipynb - image-caption-generator-using-deep-learning - Visual Studio Code
               ··· 🕏 gui.py 4, M 📳 image-captioning.ipynb 🗙
Ð
   V OPEN EDITORS
     ♣ gui.py 4, MX ■ image-captionin...
                              model.fit([X1, X2], y, epochs = 1, batch_size = 256)
    if(i%2 == 0):
    model.save_weights("image-caption-weights" + str(i) + ".h5")
                        Output exceeds the \underline{\text{size limit}}. Open the full output data \underline{\text{in a text editor}} Epoch 1/1

    image-caption-30k-9...
    image-caption-30k-9...

≡ image-caption-30k-3...
                        Epoch 1/1
292328/292328 [=========] - 55s 187us/step - loss: 3.1549
                        Epoch 1/1
292328/292328 [=========] - 54s 186us/step - loss: 2.9185

    ixtoword.pkl

                        ≣ new-model.h5
                        292328/292328 [==============] - 54s 184us/step - loss: 2.6496

    README.md

                        Epoch 1/1
292328/292328 [=========] - 54s 183us/step - loss: 2.5559

■ wordtoix.pkl

                         292328/292328 [==================] - 53s 183us/step - loss: 2.4086

→ ⊗ 0 ▲ 16
```

Fig 5. Training model

```
··· 🕏 gui.py 4, M 🔋 image-captioning.ipynb 🗙
D
       208384/292328 [========>.....] - ETA: 15s - loss: 1.8688
                               def greedy_search(pic):

≡ image-caption-30k-2....
                                   start = 'startseq'
for i in range(max_length):

≡ image-caption-30k-3...
      image-captioning.ipy...
      ≡ ixtoword.pkl
      € LICENSE.md
                                  break
final = start.split()
      ≣ new-model.h5
                                  final = final[1:-1]
final = ' '.join(final)
return final

    README md

                               pic = list(encoding_test.keys())[250]
                               img = encoding_test[pic].reshape(1, 2048)
x = plt.imread(images + pic)
plt.imshow(x)
    > OUTLINE
    > TIMELINE
```

Fig 6. Training model with search algorithm

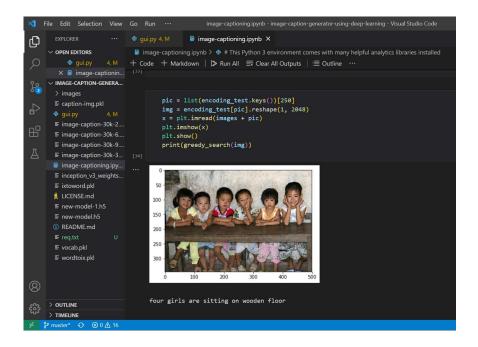


Fig 7. Testing the model

```
| File Edit Selection View Go Run Terminal Help guipy-image-caption-generator-using-deep-learning-Visual Studio Code
| Community | Prince | Prince
```

Fig 8. GUI creation

3.2 Results:

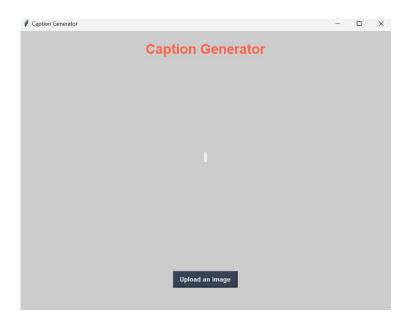


Fig 9. Caption generators first look

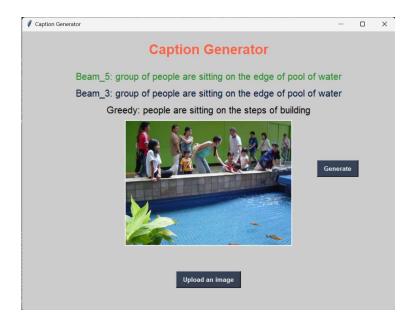


Fig 10. Output 1.

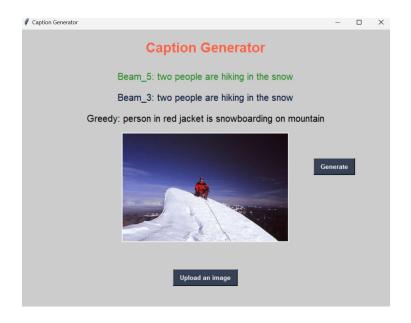


Fig 11. Output 2.

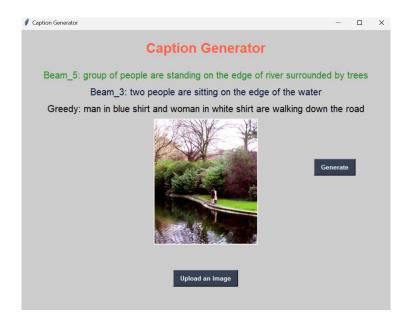


Fig 12. Output 3.

3.3 CONCLUSION

In conclusion, the image caption generator made with Keras, Inception V3 model, and LSTM is an advanced tool that uses deep learning techniques to generate accurate and natural-sounding captions for images. The GUI allows users to easily upload images and view the generated captions. In this project, a deep learning method for image caption generation using neural networks is presented, the proposed method was applied to a Flickr 30k dataset.

The encoder-decoder architecture is a strong method for creating picture captions, to sum up. The model can learn meaningful representations of images and produce natural language captions thanks to the usage of a convolutional neural network (CNN) as the encoder and an LSTM network as the decoder.

The model is effective in producing captions that are relevant and diverse, as evidenced by the evaluation of the generated captions using metrics like BLEU, METEOR, ROUGE, and CIDEr. Further enhancing the quality of the captions is the use of beam search and other methods for caption generation.

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