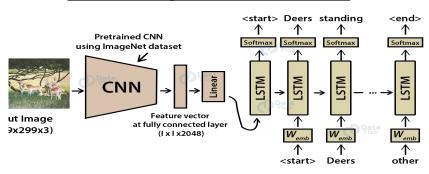
Image Caption Generator

- This project aims to construct an Image Caption Generator by integrating Convolutional Neural Networks (CNNs) and Recurrent Neural Networks (RNNs).
- Using TensorFlow and Keras.

Model - Image Caption Generator



- Step-1: CNN responsible for feature extraction, encoding the input image, specifically using Inception V3 pre trained on Images to extract image features.
- Step-2: features are passed as embeddings into an RNN built with Long Short-Term Memory (LSTM) layers.

CAPTIONS

A brown horse and a black foal on the beach

A brown horse stands near a black horse that is sitting on the ground .

A large brown horse stands over a small black colt that is kneeling on the sand .

An adult horse approaching a foal on a sandy plain .

A thin brown horse standing and a small black horse sitting on sand

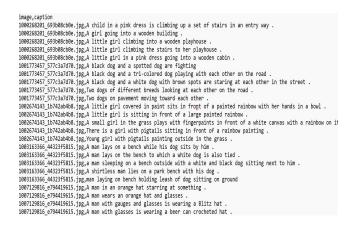


Key Difference:

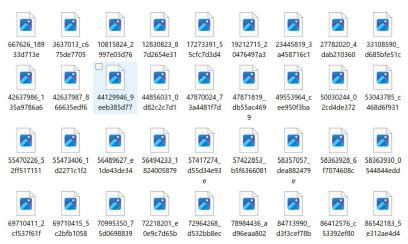
This approach differs from others by providing the image embedding as the initial input to the RNN network only once, allowing the LSTM to generate descriptive captions for the images. This combination of CNN and LSTM enables the model to learn and produce meaningful captions based on the extracted image features.

Dataset

The dataset used is Flickr 8K, consisting of 8,091 images each one paired with five different captions to provide clear descriptions.



captions



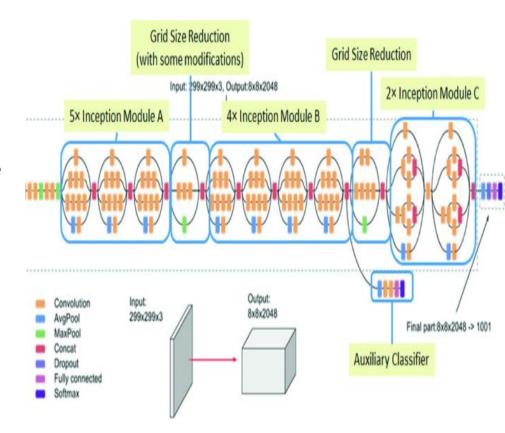
Images

ENCODER MODEL

Inception V3-

a convolutional neural network (CNN) architecture designed for image recognition tasks

- ➤ Input Layer:
- ➤ Stem:
- Inception Modules:
- Reduction Blocks:
- Auxiliary Classifiers:
- Fully Connected Layers:



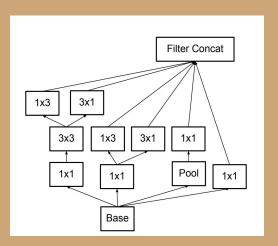
Inception Modules:

The main characteristic of InceptionV3 is the repeated use of inception modules. Each module comprises parallel convolutional operations of varying filter sizes (1x1, 3x3, 5x5), pooling, and concatenation. The purpose is to capture features at different scales and complexities.

```
# Remove the last layer of the Inception V3 model

def get_encoder():
    image_model = tf.keras.applications.InceptionV3(include_top=False, weights='imagenet')
    new_input = image_model.input
    hidden_layer = image_model.layers[-1].output

image_features_extract_model = tf.keras.Model(new_input, hidden_layer)
    return image_features_extract_model
```



get_encoder() - extracts the image feature using the InceptionV3 architecture in TensorFlow Keras.

aim - remove the last classification layer of the InceptionV3 model

Why Inception V3?

- Analysis of the models' performance showed that ResNet-50 achieved 97.5%
- ResNet has a simpler, single-scale processing unit with data pass-through connections.

- Analysis of the models' performance showed that Inception-V3 achieved 95.5%
- Inception divides processing by scale, merges the results, and repeats.

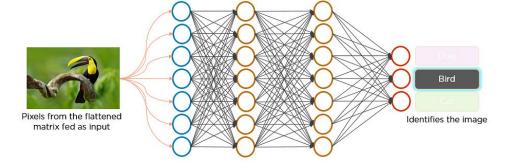
Why not Resnet-50?

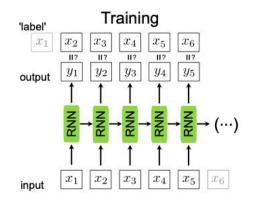
WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).



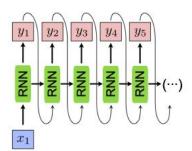
Encoder- CNN

Image detection





Generation



Decoder - RNN

Caption Generation

CNN

- Feature Extraction:
- Pre-trained Models:
- Transfer Learning:
- Feature Maps:
- Contextual Information:
- Dimensionality Reduction:

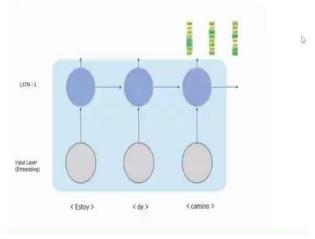
RNN

- Sequence Generation:
- Long Short-Term Memory (LSTM) or Gated Recurrent Unit (GRU):
- Word Embeddings:
- Initial State:
- Word Generation:
- Context Vector and Attention:
- Training Objective:
- Capturing Temporal Dependencies:
- Beam Search or Sampling:

Important Concept

Decoder Architecture:

Attention Mechanism: Implementing an attention mechanism in the decoder to focus on different parts of the image when generating different words. Attention can improve the model's ability to associate image features with specific words in the cap.



Estimated caption

```
def synthesize captions(captions):
    # Tokenize captions into words
    tokenized captions = [caption.split() for caption in captions]
    # Create a dictionary to hold word frequencies
    word freq = {}
    for caption in tokenized captions:
        for word in caption:
           if word not in word frea:
               word freq[word] = 1
            else:
                word freg[word] += 1
    # Select the most common words across captions
    common_words = [word for word, freq in word_freq.items() if freq >= len(captions) // 2]
    # Generate a single synthesized statement
    synthesized statement = []
    for word in common words:
        synthesized_statement.append(word)
    return ' '.ioin(synthesized statement)
```

<start> A girl runs across the sandy beach in swimsuit .a little running is on

```
['<start> A girl smiles as she runs across
the white sandy beach in her swimsuit .
<end>',
  '<start> a little girl in a flowered
bathingsuit runs through the sand at the
beach <end>',
  '<start> A little girl in a flower
swimsuit running across the beach with
waves in the background . <end>',
  '<start> There is a little girl running on
the beach . <end>',
  '<start> The young girl is running on a
sandy beach . <end>']
```

```
# Generate and print the synthesized statement
synthesized = synthesize_captions(captions)
print("Synthesized Statement:")
print(synthesized)
```

References

Dataset - https://www.kaggle.com/datasets/adityajn105/flickr8k

https://thinkautonomous.medium.com/rnns-in-computer-vision-image-captioning-597d5e1321d1

https://github.com/angeligareta/image-captioning/blob/master/notebooks/image-captioning.ipynb

THANK YOU