Image Caption Generation with LSTM and GRU

Group 8:

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Introduction to Image Captioning

Image captioning is the process of generating textual descriptions for images.

Applications include accessibility for visually impaired users, content recommendation systems, and visual search enhancement.

- Aim: Generate captions for images using deep learning architectures (CNN + RNN).
- Approach: Combine CNNs for extracting image features with LSTM and GRU for generating descriptive captions.
- Evaluation: Assess caption quality using BLEU scores and semantic distance with LLMs.

Flickr_8K Dataset Overview

- Approximately 8,000 images available.
- Approximately 40,000 captions.
- Approximately 5 captions per image
- Diverse range of subjects.
- Captions include varying lengths.



the white dog is playing in a green field with a yellow toy .

a white dog is trying to catch a ball in midair over a grassy field .

a dog leaps to catch a ball in a field .

a black and white dog jumps up towards a yellow toy .

a black and white dog jumping in the air to get a toy .



two people are at the edge of a lake, facing the water and the city skyline.

a young boy waves his hand at the duck in the water surrounded by a green park.

a little boy at a lake watching a duck .

a large lake with a lone duck swimming in it with several people around the edge of it.

a child and a woman are at waters edge in a big city .



couple with a baby sit outdoors next to their stroller.

a man and woman care for an infant along the side of a body of water.

a couple with their newborn baby sitting under a tree facing a lake .

a couple sit on the grass with a baby and stroller .

a couple and an infant , being held by the male , sitting next to a pond with a near by stroller .



this is a black dog splashing in the water .

the black dog runs through the water .

a dog splashes in the water

a black lab with tags frolicks in the water .

a black dog running in the surf .



two men are ice fishing .

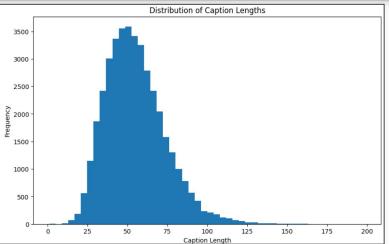
a person standing on a frozen lake .

a person in the snow drilling a hole in the ice .

a man is drilling through the frozen ice of a pond .

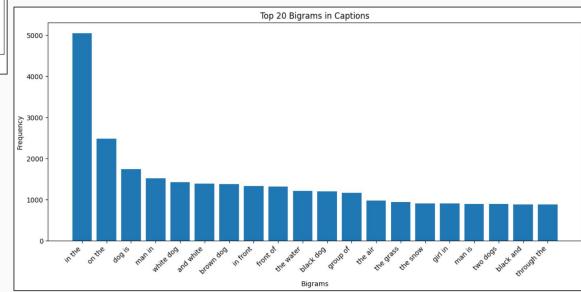
a man drilling a hole in the ice .

Caption Length Distribution & N-gram Analysis



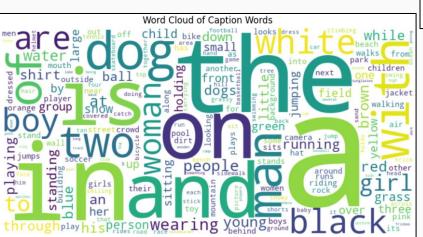
- Most captions are 40-70 characters
- Normal distribution with peak at ~50
- Few captions exceed 100 characters

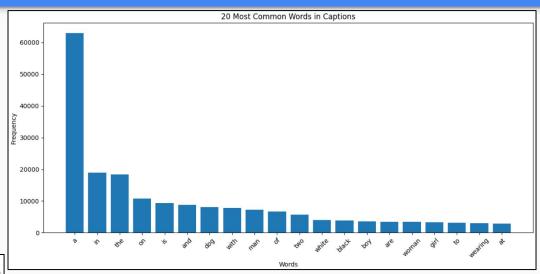
- "in the" is most frequent bigram
- Common bigrams describe objects
- Top 20 bigrams show clear patterns



Most Common Words and Word Cloud

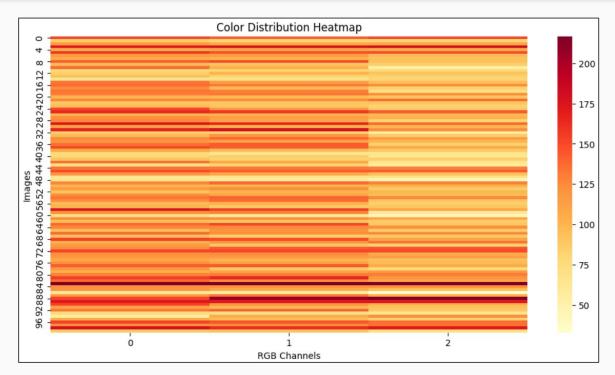
- Larger words are more frequent
- Common verbs: "is", "are", "playing"





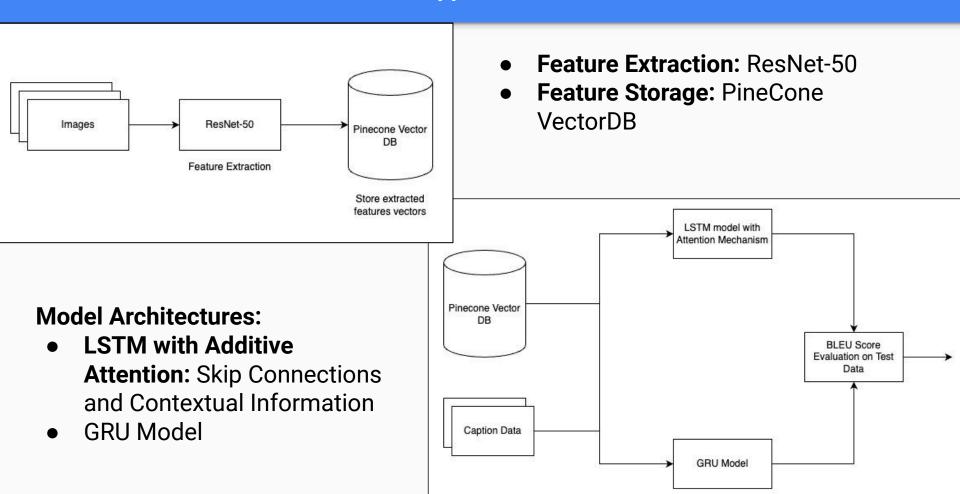
- "a" is overwhelmingly common
- Prepositions and articles dominate
- Frequent nouns: "dog", "cat", "man"

Color Distribution Heatmap



- RGB channels visualized for 10 images
- Red channel has highest intensity
- Lowest density near 0 (light yellow)

Technical Approach/Workflow



LSTM Model Implementation

- Input Layers: Two inputs one for sequential data and one for non-sequential data
- Embedding & Dropout: The sequential input is passed through an embedding layer and a dropout layer is applied to both inputs.
- LSTM and Attention: The embedded input is processed by an LSTM layer, followed by an attention mechanism. The output is concatenated with the LSTM output.
- Feature Fusion: Another LSTM processes the concatenated output. A dense layer processes the non-sequential input, and the two are combined using element-wise addition.
- Output Layers: The combined result is passed through two dense layers, with the final output size being 6329.
- Total parameters: 5.15 million.

Layer (type)	Output Shape	Param #	Connected to
input_5 (InputLayer)	[(None, 40)]	0	[]
input_4 (InputLayer)	[(None, 2048)]	0	[]
embedding_1 (Embedding)	(None, 40, 256)	1620224	['input_5[0][0]']
dropout_2 (Dropout)	(None, 2048)	0	['input_4[0][0]']
dropout_3 (Dropout)	(None, 40, 256)	0	['embedding_1[0][0]']
dense_3 (Dense)	(None, 256)	524544	['dropout_2[0][0]']
lstm_1 (LSTM)	(None, 40, 256)	525312	['dropout_3[0][0]']
additive_attention (Additi veAttention)	(None, 40, 256)	256	['lstm_1[0][0]', 'dense_3[0][0]']
concatenate (Concatenate)	(None, 40, 512)	0	['additive_attention[0][0]', 'lstm_1[0][0]']
lstm_2 (LSTM)	(None, 256)	787456	['concatenate[0][0]']
add_1 (Add)	(None, 256)	0	['dense_3[0][0]', 'lstm_2[0][0]']
dense_4 (Dense)	(None, 256)	65792	['add_1[0][0]']
dense_5 (Dense)	(None, 6329)	1626553	['dense_4[0][0]']

Total params: 5150137 (19.65 MB)
Trainable params: 5150137 (19.65 MB)
Non-trainable params: 0 (0.00 Byte)

Results and Evaluation (BLEU scores)



Referance Captions:

A man be wear a Sooner red football shirt and helmet .

A Oklahoma Sooner football player wear his jersey number 28 .

A Sooner football player weas the number 28 and black armband .

Guy in red and white football uniform

The American footballer be wear a red and white strip .

Predicted Caption:

A football player in a red helmet .

bleu score: 0.8091067115702212

Mean BLEU Score on Test Data

100%| 1000/1000 [13:34<00:00, 1.23it/s]

Bleu score on Greedy search Score: 0.4776410409015063

Semantic Similarity Computation with Large Language Model (LLM)

1. Reference Captions:

- Dog be in the snow in front of a fence.
- · Dog play on the snow.
- · Two brown dog playful fight in the snow.
- · Two brown dog wrestle in the snow.
- · Two dog play in the snow.

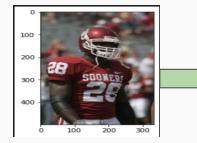
Predicted Caption:

· A doberman be run through the grass.

Analysis: The reference captions focus on dogs playing in or interacting with snow, whereas the predicted caption talks about a doberman running on grass. The context (dog activity) is somewhat similar (running and playing), but the environments (snow vs. grass) and dog breeds differ significantly.

Semantic Similarity Score: 0.4





5. Reference Captions:

- · A man be wear a Sooner red football shirt and helmet.
- A Oklahoma Sooner football player wear his jersey number 28.
- · A Sooner football player weas the number 28 and black armband.
- · Guy in red and white football uniform.
- The American footballer be wear a red and white strip.

Predicted Caption:

· A football player in a red helmet.

Analysis: Both the reference captions and the predicted caption describe a football player wearing a red helmet. While the predicted caption is less detailed, it is still fairly close to the reference descriptions.

Semantic Similarity Score: 0.8

GRU Architecture

- Feature Extraction: Image features reduced to (None, 256) via a dense layer. Text sequences converted to 128-dimensional embeddings, processed by a GRU to output (None, 256).
- **Fusion Layer:** Concatenates features into a vector of size (None, 512).
- Output Layer: Final dense layer produces output for caption generation.

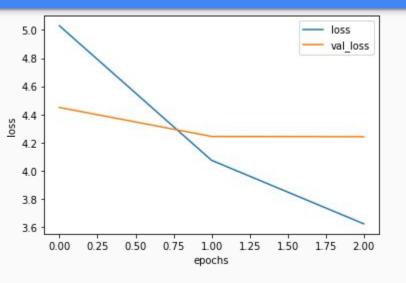
Layer (type)	Output Shape	Param #	Connected to
input_layer_2 (InputLayer)	(None, 50)	0	-
input_layer_1 (InputLayer)	(None, 100352)	0	-
embedding (Embedding)	(None, 50, 128)	1,280,000	input_layer_2[0]
ImageFeature (Dense)	(None, 256)	25,690,368	input_layer_1[0]
CaptionFeature (GRU)	(None, 256)	296,448	embedding[0][0]
concatenate (Concatenate)	(None, 512)	0	ImageFeature[0][CaptionFeature[0
dense (Dense)	(None, 10000)	5,130,000	concatenate[0][0]

Total params: 32,396,816 (123.58 MB)

Trainable params: 32,396,816 (123.58 MB)

Non-trainable params: 0 (0.00 B)

GRU Results



1. Training Loss (Blue Line):

- The training loss decreases steadily over the course of the 2 epochs, indicating that the model is learning and improving on the training data.
- This consistent decline suggests that the model is fitting well to the training set.

2. Validation Loss (Orange Line):

- The validation loss decreases initially but then plateaus after around 1 epoch.
- This plateau suggests that while the model is improving on the training data, it is not significantly improving on unseen validation data after a certain point.

The Mean BLEU-1 Score for the Test Set is 0.138 The Mean BLEU-2 Score for the Test Set is 0.050 The Mean BLEU-3 Score for the Test Set is 0.029 The Mean BLEU-4 Score for the Test Set is 0.022

Analysis:

- Potential Overfitting
- Can Include dropout and normalization layers to avoid overfitting

LSTM	GRU
0.47	0.23

Comparative Analysis of BLEU Scores

Analysis Results:

- 1. LSTM achieves higher accuracy due to the Attention Mechanism.
- 2. GRUs are computationally more efficient, making them suitable for applications where speed is critical.
- 3. GRU's scores can be increased by increasing the complexity of the architecture.
- 4. LSTM's scores can be increased by training for more epochs and more data.

Future Directions

- Train for more epochs to improve model convergence.
- Increase dataset size for better generalization.
- Incorporate attention with GRU to enhance sequence learning.
- Experiment with transformer architecture for improved performance on long sequences.

Thank You