#### STUDY 1: IMAGE CAPTIONING USING ATTENTION MODEL

#### MODEL: ATTENTION MECHANISM + ResNet50 (Convolutional Network)

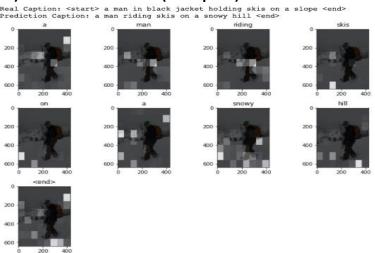
#### I. MODEL DESCRIPTION:

Attention Mechanism			Number of Training Images Epochs		Total Datapoints	Training Batch Size	
Bahdanau Attention	7,000	6,000	30	5	30,000	64	

# **II. PREDICTION RESULTS:**

In this study, we have analyzed the predicted results for 50 different images from the MS-COCO dataset and 3 other random images from the internet. We classify these image caption predictions as "Good", "Fair" and "Bad". Note that, the classifications into "Good", "Fair" and "Bad" are based only on how close the predicted caption is to the real caption - based on human judgement. In the next section, we will analyze NLP metrics such as BLEU, GLEU and WER scores. We calculate the mean of these metrics from the 53 outcomes to make inferences about the model and the convolutional layer used. A few examples of "Good", "Fair" and "Bad" are:

### 1) GOOD PREDICTION (example 1)



INDUVIDUAL N GRAM					
1 GRAM	0.111111				
2 GRAM	1.000000				
3 GRAM	1.000000				
4 GRAM	1.000000				

CUMMULATIVE N GRAM					
BLEU 1	0.111111				
BLEU 2	0.333333				
BLEU 3	0.484284				
BLEU 4	0.577350				

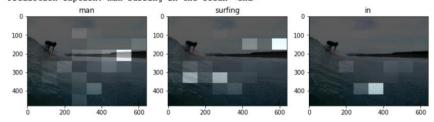
GLEU S	CORE
Sentenc Frequ	
1 to 4 grams	0.2647
1 to 2 grams	0.4215

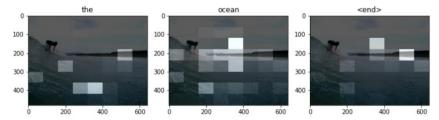
WER I	nat	rix	(9)	k10	):				
.0]]	1.	2.	3.	4.	5.	6.	7.	8.	9.]
[1.	0.	1.	2.	3.	4.	5.	6.	7.	8.]
[2.	1.	1.	2.	3.	4.	5.	6.	7.	8.]
[3.	2.	2.	2.	3.	4.	4.	5.	6.	7.]
[4.	3.	3.	3.	3.	4.	5.	4.	5.	6.]
[5.	4.	4.	4.	4.	4.	5.	5.	4.	5.]
[6.	5.	5.	5.	5.	5.	5.	6.	5.	5.]
[7.	6.	6.	6.	6.	6.	6.	6.	6.	6.]
[8.	7.	7.	7.	7.	7.	7.	7.	7.	7.]]
7									

[Next Page]

# 2) GOOD PREDICTION (example 2)

Real Caption: <start> the surfer is riding on a wave in the ocean <end> Prediction Caption: man surfing in the ocean <end>





#### BLEU SCORE, GLEU SCORE and WER (Word Error Rate) metric comparison between sentences:

INDUVIDUAL N GRAM					
1 GRAM	0.111111				
2 GRAM	1.000000				
3 GRAM	1.000000				
4 GRAM	1.000000				

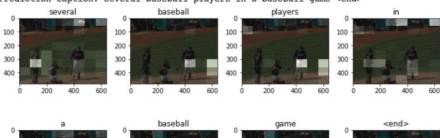
CUMMUL	ATIVE N GRAM			
BLEU 1	0.111111			
BLEU 2	0.333333			
BLEU 3	0.484284			
BLEU 4	0.577350			

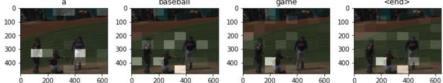
GLEU S	CORE
Sentenc	
Frequ	ency
1 to 4 grams	0.1176
1 to 2 grams	0.2105

```
WER matrix (9x10):
[[0 1 2 3 4 5 6 7 8 9]
[1 1 2 3 4 5 6 7 8 9]
[2 2 2 2 3 4 5 6 7 8]
[3 3 3 3 3 4 5 6 7 8]
[4 4 4 4 4 3 4 5 6 7 8]
[5 5 5 5 5 4 3 4 5 6 7]
[6 6 6 6 6 5 4 4 5 6 7]
[7 7 7 7 7 6 5 5 5 5 6 7]
[8 8 8 8 8 7 6 6 6 6 6 7]]
```

#### 3) GOOD PREDICTION (example 3)

Real Caption: <start> group of baseball players preparing next move in a game <end> Prediction Caption: several baseball players in a baseball game <end>





INDUVIDU	AL N GRAM
1 GRAM	0.125000
2 GRAM	1.000000
3 GRAM	1.000000
4 GRAM	1.000000

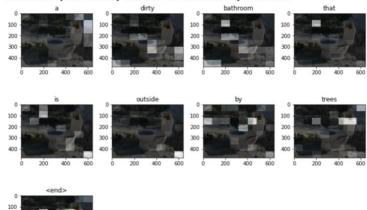
CUMMUL	ATIVE N GRAM
BLEU 1	0.125000
BLEU 2	0.353553
BLEU 3	0.503478
BLEU 4	0.594604

CORE
e Level
ency
0.2058
0.3684

WER matrix			(8x10):						
[[0.	1.	2.	3.	4.	5.	6.	7.	8.	9.]
[1.	1.	1.	2.	3.	4.	5.	6.	7.	8.]
[2.	2.	2.	1.	2.	3.	4.	5.	6.	7.]
[3.	3.	3.	2.	2.	3.	4.	4.	5.	6.]
[4.	4.	4.	3.	3.	3.	4.	5.	4.	5.]
[5.	5.	4.	4.	4.	4.	4.	5.	5.	5.]
[6.	6.	5.	5.	5.	5.	5.	5.	6.	5.]
[7.	7.	6.	6.	6.	6.	6.	6.	6.	6.]]
6									

#### 4) FAIR PREDICTION (example 1)

Prediction Caption: a dirty bathroom that is outside by trees <end>



# 

#### BLEU SCORE, GLEU SCORE and WER (Word Error Rate) metric comparison between sentences:

INDUVIDUAL N GRAM						
1 GRAM	0.111111					
2 GRAM	1.000000					
3 GRAM	1.000000					
4 GRAM	1.000000					

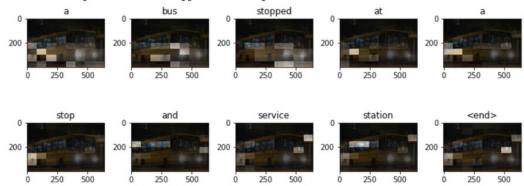
CUMMULATIVE N GRAM						
BLEU 1	0.111111					
BLEU 2	0.333333 0.484284					
BLEU 3						
BLEU 4	0.577350					

GLEU SC	CORE			
Sentence Level Frequency				
1 to 4 grams	0.0789			
1 to 2 grams	0.1428			

WER matrix (9x11): [[0. 1. 2. 3. 4. 5. 6. 7. 8. 9.10] [1. 1. 2. 3. 4. 5. 6. 7. 8. 9.10] [2. 2. 2. 3. 4. 5. 6. 7. 8. 9.10] [3. 3. 3. 3. 4. 5. 6. 7. 8. 9.10] [4. 4. 3. 4. 4. 5. 6. 7. 8. 9.10] [5. 5. 4. 3. 4. 5. 6. 7. 8. 9.10]	
[ 1. 1. 2. 3. 4. 5. 6. 7. 8. 9. 10 [ 2. 2. 2. 3. 4. 5. 6. 7. 8. 9. 10 [ 3. 3. 3. 3. 4. 5. 6. 7. 8. 9. 10 [ 4. 4. 3. 4. 4. 5. 6. 7. 8. 9. 10 [ 5. 5. 4. 3. 4. 5. 6. 7. 8. 9. 10	
[ 2. 2. 2. 3. 4. 5. 6. 7. 8. 9. 10 [ 3. 3. 3. 3. 4. 5. 6. 7. 8. 9. 10 [ 4. 4. 3. 4. 4. 5. 6. 7. 8. 9. 10 [ 5. 5. 4. 3. 4. 5. 6. 7. 8. 9. 10	.]
[ 3. 3. 3. 3. 4. 5. 6. 7. 8. 9. 10 [ 4. 4. 3. 4. 4. 5. 6. 7. 8. 9. 10 [ 5. 5. 4. 3. 4. 5. 6. 7. 8. 9. 10	.]
[ 4. 4. 3. 4. 4. 5. 6. 7. 8. 9. 10 [ 5. 5. 4. 3. 4. 5. 6. 7. 8. 9. 10	.]
[ 5. 5. 4. 3. 4. 5. 6. 7. 8. 9. 10	. ]
	. j
	.]
[ 6. 6. 5. 4. 4. 5. 6. 7. 8. 9. 10	.]
[ 7. 7. 6. 5. 5. 5. 6. 7. 8. 9. 10	.]
[8. 8. 7. 6. 6. 6. 6. 7. 8. 9. 10	.]]
10	

# 5) FAIR PREDICTION (example 2)

Real Caption: <start> a commuter bus parked at a stop at night time <end> Prediction Caption: a bus stopped at a stop and service station <end>



INDUVIDL	INDUVIDUAL N GRAM							
1 GRAM	0.100000							
2 GRAM	1.000000							
3 GRAM	1.000000							
4 GRAM	1.000000							

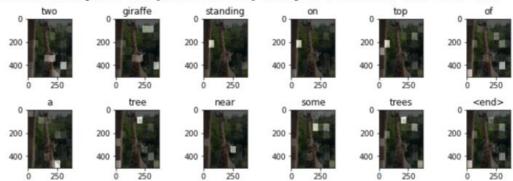
CUMMU	CUMMULATIVE N GRAM					
BLEU 1	0.100000					
BLEU 2	0.316288					
BLEU 3	0.467735					
BLEU 4	0.562341					

GLEU SCORE					
Sentence Freque					
1 to 4 grams	0.2352				
1 to 2 grams	0.3684				

WER I	mati	rix	(1	0x1	0):				
.0]]	1.	2.	3.	4.	5.	6.	7.	8.	9.]
[1.	1.	1.	2.	3.	4.	5.	6.	7.	8.]
[2.	2.	2.	2.	3.	4.	5.	6.	7.	8.]
[3.	3.	3.	3.	2.	3.	4.	5.	6.	7.]
[4.	4.	4.	4.	3.	2.	3.	4.	5.	6.]
[5.	5.	5.	5.	4.	3.	2.	3.	4.	5.]
[6.	6.	6.	6.	5.	4.	3.	3.	4.	5.]
[7.	7.	7.	7.	6.	5.	4.	4.	4.	5.]
[8.	8.	8.	8.	7.	6.	5.	5.	5.	5.]
[9.	9.	9.	9.	8.	7.	6.	6.	6.	6.]]

# 6) FAIR PREDICTION (example 3)

Real Caption: <start> two giraffes in an wooden and cable fence <end> Prediction Caption: two giraffe standing on top of a tree near some trees <end>



#### BLEU SCORE, GLEU SCORE and WER (Word Error Rate) metric comparison between sentences:

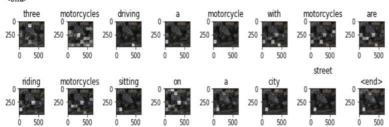
INDUVIDUAL N GRAM					
0.083333					
1.000000					
1.000000					
1.000000					

CUMMULATIVE N GRAM						
BLEU 1	0.083333					
BLEU 2	0.288675 0.440423					
BLEU 3						
BLEU 4	0.537285					

CLELLCCORE		WER m	WER matrix (12x8):						
GLEU SCORE			1.	2.	3.	4.	5.	6.	7.]
		[ 1.	1.	2.	3.	4.	5.	6.	7.]
		[ 2.	2.	2.	3.	4.	5.	6.	7.]
Sentenc	e Level	[ 3.	3.	3.	3.	4.	5.	6.	7.]
		[ 4.	4.	4.	4.	4.	5.	6.	7.]
Frequ	ency	[ 5.	5.	5.	5.	5.	5.	6.	7.]
20 (5) 32		[ 6.	6.				6.	6.	7.]
1 to 4 grams	0.0238	[ 7.	7.	7.	7.	7.	7.	7.	7.]
	9	[ 8.	8.	8.	8.	8.	8.	8.	8.]
1 to 2 grams	0.0434	[ 9.	9.	9.	9.	9.	9.	9.	9.]
1 to 2 grams	0.0434	[10.	10.	10.	10.	10.	10.	10.	10.]
	1	111.	11.	11.	11.	11.	11.	11	11.1

# 7) BAD PREDICTION (example 1)

Real Caption: <start> there are three police motor cycles parked together <end>
Prediction Caption: three motorcycles driving a motorcycle with motorcycles are riding motorcycles sitting on a city street <end>

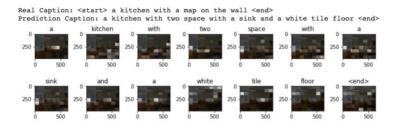


INDUVIDU	INDUVIDUAL N GRAM							
1 GRAM	0.062500							
2 GRAM	1.000000							
3 GRAM	1.000000							
4 GRAM	1.000000							

CUMMULATIVE N GRAM				
BLEU 1	0.062500			
BLEU 2	0.250000			
BLEU 3	0.400535			
BLEU 4	0.500000			

GLEU SCORE			
Sentence Level Frequency			
1 to 4 grams	0.0344		
1 to 2 grams	0.0645		

# 8) BAD PREDICTION (example 2)

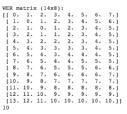


#### BLEU SCORE, GLEU SCORE and WER (Word Error Rate) metric comparison between sentences:

INDUVIDUAL N GRAM					
1 GRAM	0.071429				
2 GRAM	1.000000				
3 GRAM	1.000000				
4 GRAM	1.000000				

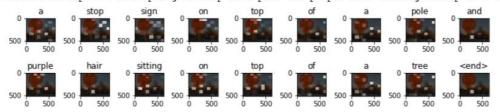
CUMMULATIVE N GRAM					
BLEU 1	0.071429				
BLEU 2	0.267261				
BLEU 3	0.418579				
BLEU 4	0.516973				

GLEU SCORE			
Sentence Level Frequency			
1 to 4 grams	0.1600		
1 to 2 grams	0.2592		



#### 9) BAD PREDICTION (example 3)

Real Caption: <start> a close of image of a dirty stop sign <end>
Prediction Caption: a stop sign on top of a pole and purple hair sitting on top of a tree <end>



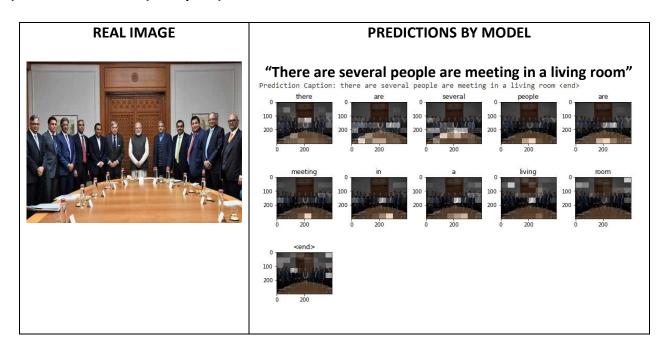
INDUVIDUAL N GRAM				
0.055556				
1.000000				
1.000000				
1.000000				

CUMMULATIVE N GRAM				
BLEU 1	0.055556			
BLEU 2	0.235702			
BLEU 3	0.385265			
BLEU 4	0.485492			

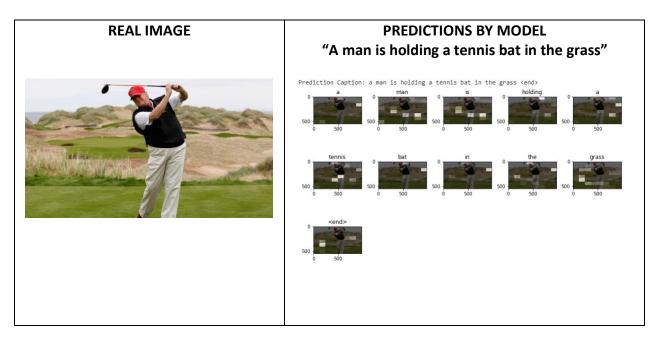
GLEU SCORE			
Sentence Level Frequency			
1 to 4 grams	0.1212		
1 to 2 grams	0.2285		

	M NAW	attix	TOXA)					
	[[ 0.	1. 2	2. 3.	4.	5.	6.	7.	8.]
	[ 1.	1. 2	2. 3.	4.	5.	6.	6.	7.]
	[ 2.	2. 2	2. 3.	4.	5.	6.	7.	6.)
	[ 3.	3. 3	3. 3.	4.	5.	6.	7.	7.]
ī	[ 4.	4. 4	4.	4.	5.	6.	7.	8.]
	[ 5.	5. 4	1. 5.	4.	5.	6.	7.	8.]
	[ 6.	6. :	5. 5.	5.	4.	5.	6.	7.1
	[ 7.	7. 6	5. 6.	6.	5.	5.	6.	7.1
	[8.	8. 7	7. 7.	7.	6.	6.	6.	7.]
	[ 9.	9. 8	8. 8.	8.	7.	7.	7.	7.1
	[10.	10. 9	). 9.	9.	8.	8.	8.	8.)
	[11.	11. 10	). 10.	10.	9.	9.	9.	9.1
	[12.	12. 11	1. 11.	11.	10.	10.	10.	10.]
	[13.	13. 12	2. 12.	12.	11.	11.	11.	11.]
	[14.	14. 13	3. 13.	12.	12.	12.	12.	12.1
	[15.	15. 14	. 14.	13.	12.	13.	13.	13.]
	[16.	16. 15	5. 15.	14.	13.	13.	14.	14.]
	[17.	17. 16	5. 16.	15.	14.	14.	14.	15.]]

# 10) RANDOM IMAGES (example 1)



# 11) RANDOM IMAGES (example 2)



# 12) RANDOM IMAGES (example 2)

This is an image of myself (on the right) and my teammate!!

# The model predicted the following for my picture: **REAL IMAGE:** PREDICTION 1: "A pair of people posing on" PREDICTION 2: "A picture of <unk> <unk> at a train"

