

Visual Question Answering

Presented By :

Urvi Parekh

Abhinav Rahate

Manavi Agrawal

WHAT SPORT IS IT?

SOCCER

99%

FRISBEE

1%



Contents



1. What is VQA?
2. Applications
3. Workflow
4. Data Pre-Processed
5. Model
6. Deployment
7. Challenges
8. Research Methodology
9. Limitation



What is VQA?



Given an **image**, can our
machine answer the
corresponding questions in
natural language?



What is VQA?



What is VQA?



In order to do this, our model would need to understand several things - let's break them down into sub-tasks:

1. Identifying the various objects in the image (the train, traffic signals, tracks, pavement, person, etc)
2. Processing the text of the question itself, which can be processed as a 'sequence' of words
3. Mapping the appropriate sections of the image (in this case - the train) to the input text question.
4. Generating natural language text in the form of an answer.



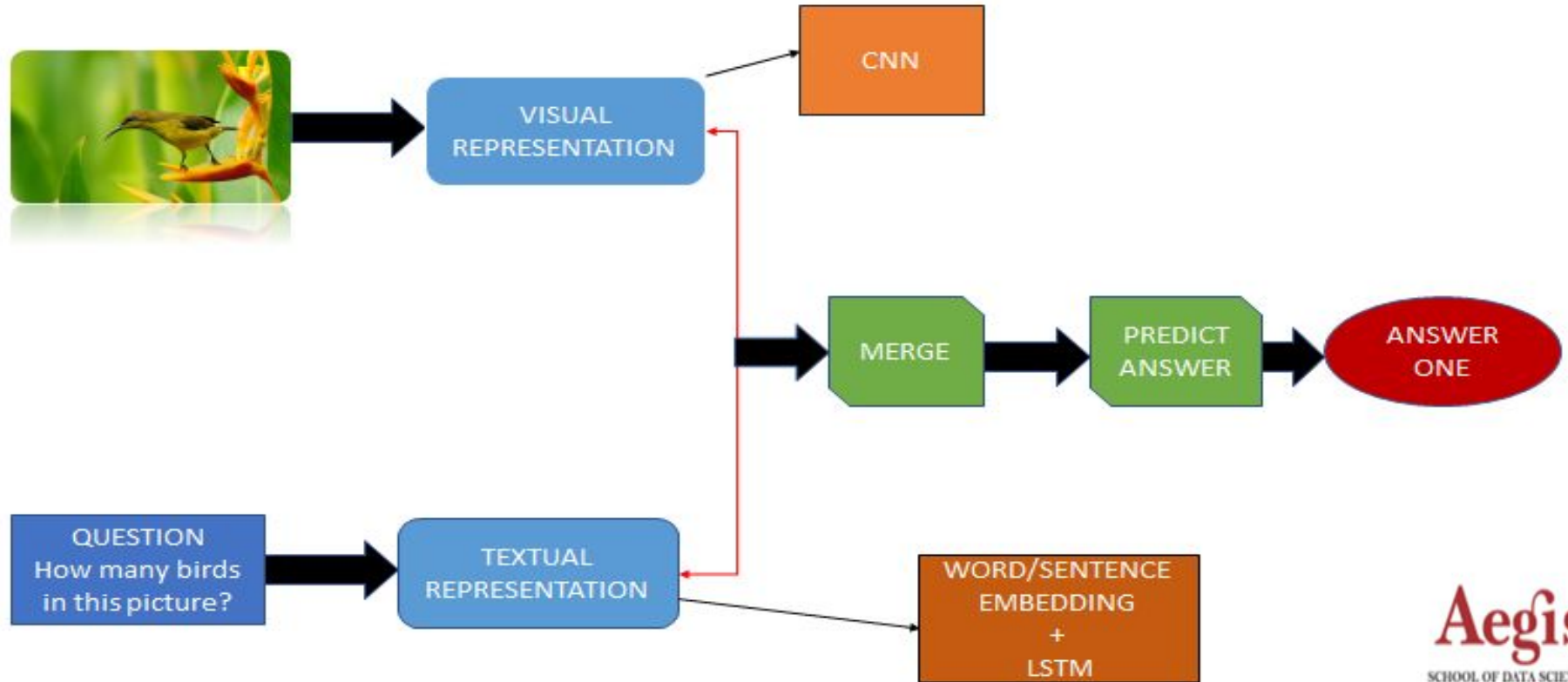
Applications



1. Educating a child playing a game on a touch screen.
2. Providing information to a spectator at an art gallery, or interacting with a robot.
3. Shopping apps like amazon and Flipkart could make their product search just so much better , like you could specify the particular type of design and it would be there.



Workflow



Workflow



1. Pre-Processed the questions
2. Extracted Image Features
3. Calculated Glove Weights
4. Building Image Model
5. Building Language Model
6. Combining The Model
7. Testing the Model
8. Deployment



Data Pre-Processing



1. Pre-processed questions.(Using nltk)
2. Tokenization and Padding questions.
3. Calculating the glove weights.
4. Reshaping the image into $224 * 224 * 3$
5. Extracting Image features.(using pre-trained model VGG16)
6. Converting the answer label into integer form.(using LabelEncoder)



Data Pre-Processing

	ans	img_id	ques_id	question	image_name	image_feature
0	5218	25	25000	front giraffes	COCO_train2014_000000000025.jpg	[1.38433e-05, 5.34205e-05, 0.0001280025, 0.000...
1	1887	25	25001	giraffes common	COCO_train2014_000000000025.jpg	[1.38433e-05, 5.34205e-05, 0.0001280025, 0.000...
10	2993	25	25010	ground next giraffe right	COCO_train2014_000000000025.jpg	[1.38433e-05, 5.34205e-05, 0.0001280025, 0.000...
100	3363	149	149001	sky clear	COCO_train2014_000000000149.jpg	[3.93909e-05, 1.6657e-05, 9.6804e-06, 3.1817e-...
1000	4219	1522	1522001	kind place	COCO_train2014_000000001522.jpg	[9.9e-09, 3.589e-07, 9.6e-09, 7e-10, 5.8e-09, ...

```
array([[0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 69., 81.],
       [0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 81., 905.],
       [0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        53., 122., 96., 36.],
       [0., 0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 0., 0., 0., 0., 0., 0., 0., 0.,
        0., 0., 48., 249.]])
```

Question features 

Model



Steps followed:

1. Image Model

Layer (type)	Output Shape	Param #
=====		
dense_3 (Dense)	(None, 2048)	2050048

activation_3 (Activation)	(None, 2048)	0
=====		
Total params: 2,050,048		
Trainable params: 2,050,048		
Non-trainable params: 0		

None		



Model



2. Language Model

1. The language model is build using LSTM, which is recurrent neural network.
2. Problem of using RNN:
 - a. Vanishing Gradient Descent
 - i. As more layers using certain activation functions are added to neural networks, the gradients of the loss function approaches zero, making the network hard to train.



Model



2. Language Model

Layer (type)	Output Shape	Param #
=====		
embedding_1 (Embedding)	(None, 26, 300)	2083800

lstm_1 (LSTM)	(None, 26, 64)	93440

lstm_2 (LSTM)	(None, 64)	33024

dense_2 (Dense)	(None, 2048)	133120

activation_2 (Activation)	(None, 2048)	0
=====		
Total params: 2,343,384		
Trainable params: 259,584		
Non-trainable params: 2,083,800		

None		



Model

3. Combine Model

Layer (type)	Output Shape	Param #	Connected to
dense_1 (Dense)	(None, 2048)	2050048	dense_input_1[0][0]
activation_1 (Activation)	(None, 2048)	0	dense_1[0][0]
embedding_1 (Embedding)	(None, 26, 300)	2083800	embedding_input_1[0][0]
lstm_1 (LSTM)	(None, 26, 64)	93440	embedding_1[0][0]
lstm_2 (LSTM)	(None, 64)	33024	lstm_1[0][0]
dense_2 (Dense)	(None, 2048)	133120	lstm_2[0][0]
activation_2 (Activation)	(None, 2048)	0	dense_2[0][0]
dense_3 (Dense)	(None, 1024)	2098176	merge_1[0][0]
dense_4 (Dense)	(None, 1000)	1025000	dense_3[0][0]
dense_5 (Dense)	(None, 5666)	5671666	dense_4[0][0]
Total params: 13,188,274			
Trainable params: 11,104,474			
Non-trainable params: 2,083,800			

Model Accuracy

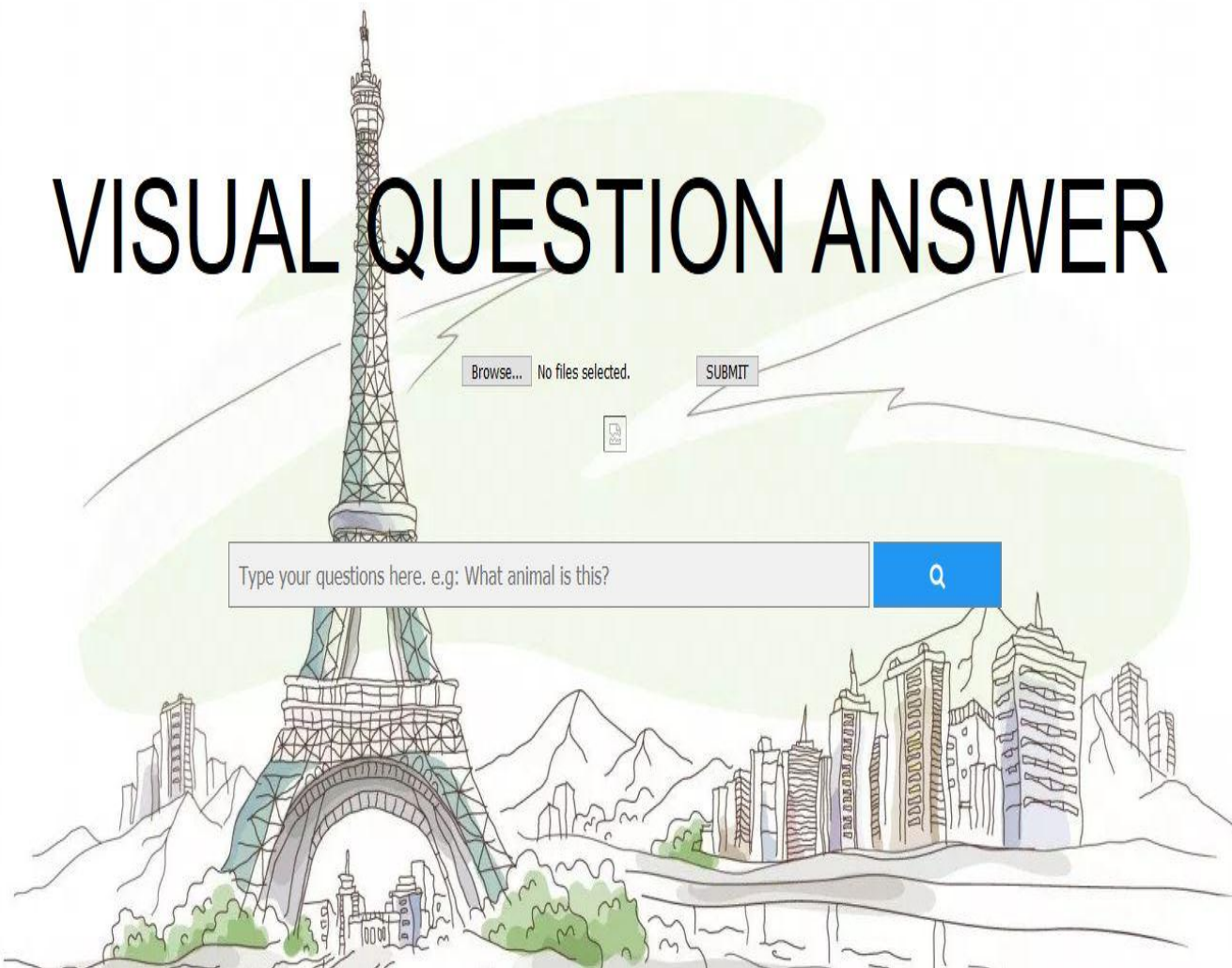


With 10,000 images trained we are able to achieve **40% training accuracy** and **40% testing accuracy**



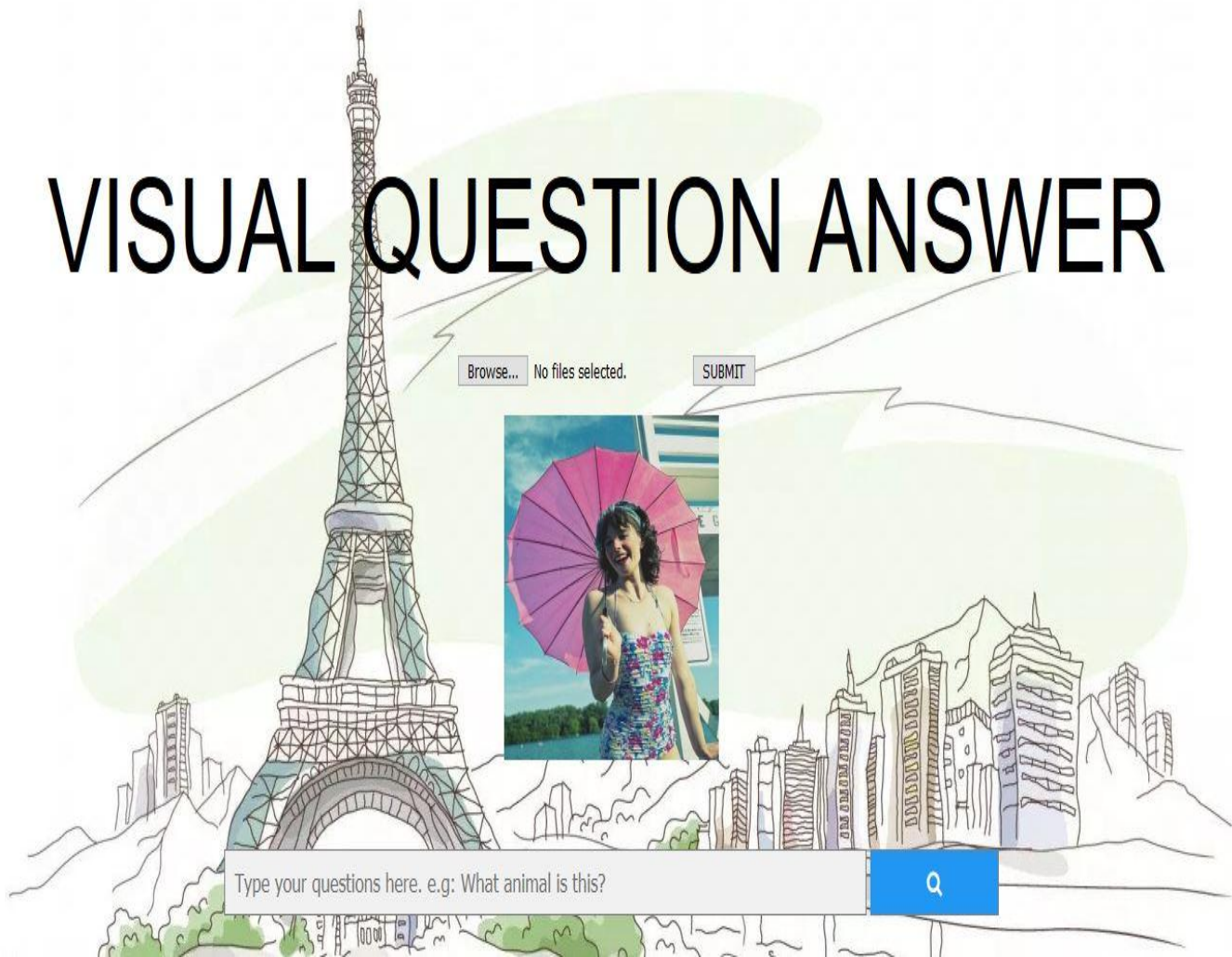
Deployment

VISUAL QUESTION ANSWER



Deployment

VISUAL QUESTION ANSWER



VISUAL QUESTION ANSWER

Deployment



Challenges



1. To handle data of different type(questions features and image features)
2. Every row had array of 30 as qus features and array of 1000 as image features. Because this we had a lot of trouble while passing it to embedding layer and dese layer respectively as it was throwing shape errors.
3. To merge two layers “Merge “ was removed in newer version of keras.
4. To combine models both model had to have same shape.
5. Issue we are still facing is processing power because of which we could only use 1000 images





Research Methodology

1. There has been lot of research done on VQA from last couple of years .
2. There are research papers available with different methodology for achieving the good accuracy in answering the question pertaining to the image.
3. We had taken reference from those research and methodology and build a model/product that answer more accurately .

Research Methodology



Reference Research Papers

<https://arxiv.org/pdf/1708.02711v1.pdf>

<https://arxiv.org/pdf/1705.06676v1.pdf>

<https://arxiv.org/abs/1606.00061>

<https://www.coursehero.com/file/36229255/150500468pdf/>

http://openaccess.thecvf.com/content_cvpr_2017/papers/Goyal_Making_the_v_CVPR_2017_paper.pdf

<https://arxiv.org/abs/1612.00837>

http://openaccess.thecvf.com/content_cvpr_2016/papers/Shih_Where_to_Look_CVPR_2016_paper.pdf





Limitation

1. Due to limitation of the processing power we went ahead with 10k images and respective questions for them.
2. As this is very complicated problem and needs huge amount of data to process and make the vocabulary accordingly we were not able to reach acceptable accuracy.



THANK YOU