

# Machine Learning II — Visual Question Answer

#### **Group 7**

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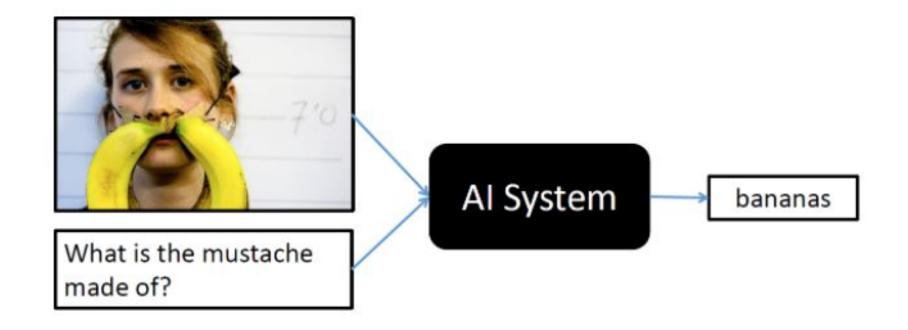
### **Outline**



- Introduction
- Data Process
- VAQ Model
- Results
- Conclusion

# Introduction





### View Dataset - MSCOCO



#### Image Set

204,721 Image quantities

Input: 4096 dimensions feature

#### **Question Set**

1,105,904 questions

Input: 384 dimensions feature

#### **Answer Set**

Top 1000 answers from

11,059,040 ground truth answer

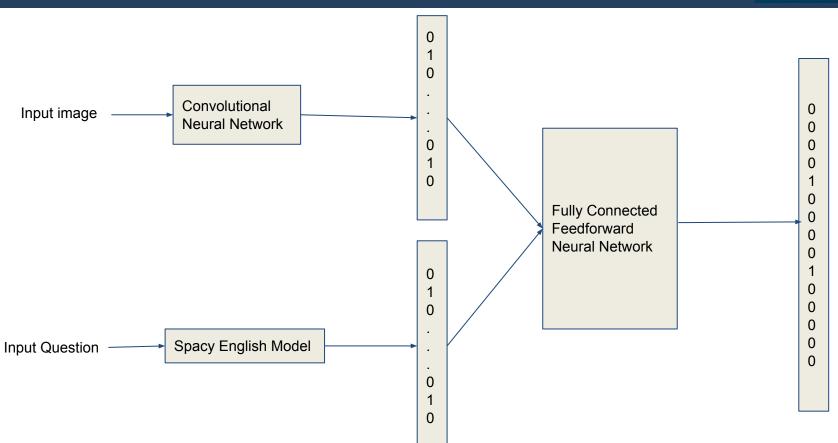


ice cream



```
sipeng wang@ml-class-ubuntul6:~/final/data/preprocessed$ cat guestions val2014.t
xt | head -10
Where is he looking?
What are the people in the background doing?
What is he on top of?
What website copyrighted the picture?
Is this a creamy soup?
Is this rice noodle soup?
What is to the right of the soup?
What is the man doing in the street?
How many photo's can you see?
What does the truck on the left sell?
sipeng wang@ml-class-ubuntu16:~/final/data/preprocessed$ cat answers val2014 mod
al.txt |head -10
down
watching
picnic table
foodiebakercom
yes
chopsticks
walking
```





# Data Pre-process



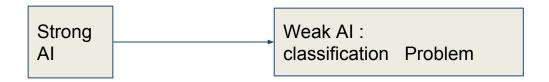
Extract data

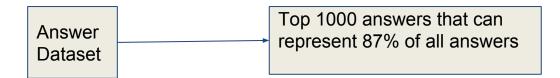
Generate Answer

Categorize answer data



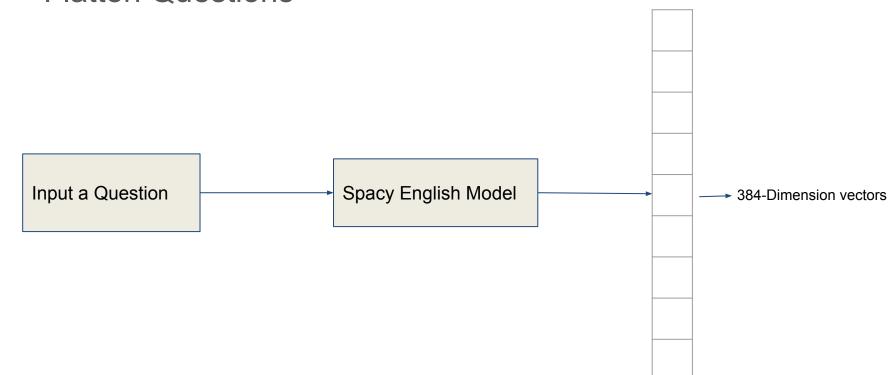
Generate Answers





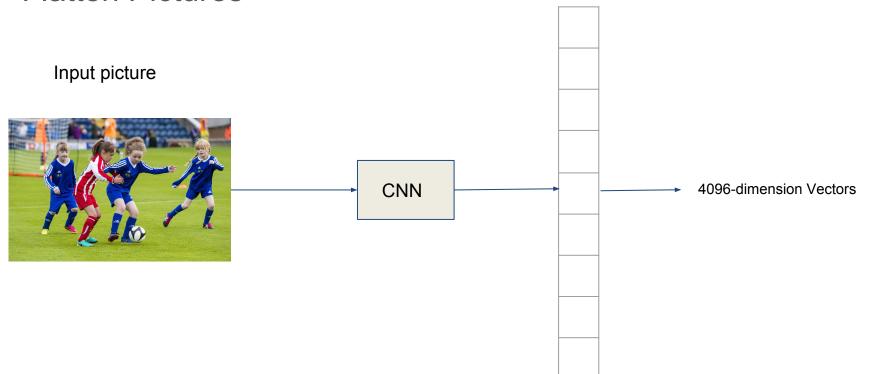


Flatten Questions





Flatten Pictures



#### **VGG 16**

```
model.add(ZeroPadding2D((1,1)))
                                                             model.add(Convolution2D(512, 3, 3, activation='relu'))
def VGG 16 (weights path=None):
                                                             model.add(ZeroPadding2D((1,1)))
   model = Sequential()
                                                            model.add(Convolution2D(512, 3, 3, activation='relu'))
   model.add(ZeroPadding2D((1,1),input shape=(3,224,224)))
                                                            model.add(ZeroPadding2D((1,1)))
   model.add(Convolution2D(64, 3, 3, activation='relu'))
                                                            model.add(Convolution2D(512, 3, 3, activation='relu'))
   model.add(ZeroPadding2D((1,1)))
                                                            model.add(MaxPooling2D((2,2), strides=(2,2)))
   model.add(Convolution2D(64, 3, 3, activation='relu'))
   model.add(MaxPooling2D((2,2), strides=(2,2)))
                                                            model.add(ZeroPadding2D((1,1)))
                                                             model.add(Convolution2D(512, 3, 3, activation='relu'))
   model.add(ZeroPadding2D((1,1)))
                                                            model.add(ZeroPadding2D((1,1)))
   model.add(Convolution2D(128, 3, 3, activation='relu'))
                                                            model.add(Convolution2D(512, 3, 3, activation='relu'))
   model.add(ZeroPadding2D((1,1)))
                                                            model.add(ZeroPadding2D((1,1)))
   model.add(Convolution2D(128, 3, 3, activation='relu'))
                                                            model.add(Convolution2D(512, 3, 3, activation='relu'))
   model.add(MaxPooling2D((2,2), strides=(2,2)))
                                                             model.add(MaxPooling2D((2,2), strides=(2,2)))
   model.add(ZeroPadding2D((1,1)))
                                                            model.add(Flatten())
   model.add(Convolution2D(256, 3, 3, activation='relu'))
                                                            model.add(Dense(4096, activation='relu'))
   model.add(ZeroPadding2D((1,1)))
                                                            model.add(Dropout(0.5))
   model.add(Convolution2D(256, 3, 3, activation='relu'))
                                                             model.add(Dense(4096, activation='relu'))
   model.add(ZeroPadding2D((1,1)))
                                                            model.add(Dropout(0.5))
   model.add(Convolution2D(256, 3, 3, activation='relu'))
                                                            model.add(Dense(1000, activation='softmax'))
   model.add(MaxPooling2D((2,2), strides=(2,2)))
                                                             if weights path:
                                                                 model.load weights (weights path)
                                                             return model
```

input: (None, 224, 224, 3)		
input_1: InputLayer output: (None, 224, 224, 3)	block5 conv1: Conv2D input: (None, 14, 14, 512)	dense_3: Dense input: (None, 4608) output: (None, 1024)
block1_conv1: Conv2D input: (None, 224, 224, 3)	output: (None, 14, 14, 512)	
output: (None, 224, 224, 64)		
input: (None, 224, 224, 64)	input: (None, 14, 14, 512)	activation_1: Activation input: (None, 1024) output: (None, 1024)
block1_conv2: Conv2D output: (None, 224, 224, 64)	block5_conv2: Conv2D output: (None, 14, 14, 512)	(1010, 1021)
•	(1016, 14, 14, 11a)	
block1_pool: MaxPooling2D input: (None, 224, 224, 64) output: (None, 112, 112, 64)		dropout_2: Dropout input: (None, 1024) output: (None, 1024)
•	block5 conv3: Conv2D input: (None, 14, 14, 512)	(1000, 1021)
block2_conv1: Conv2D   input: (None, 112, 112, 64)   output: (None, 112, 112, 128)	output: (None, 14, 14, 512)	<u> </u>
(1000, 112, 112, 120)		dense_4: Dense input: (None, 1024) output: (None, 1024)
block2_conv2: Conv2D input: (None, 112, 112, 128)	input: (None, 14, 14, 512)	outain (cont, 1021)
output: (None, 112, 112, 128)	block5_pool: MaxPooling2D input: (None, 14, 14, 512) output: (None, 7, 7, 512)	
block2_pool: MaxPooling2D input: (None, 112, 112, 128)	output. (Noie, 1, 1, 312)	activation_2: Activation input: (None, 1024) output: (None, 1024)
output: (None, 56, 56, 128)	•	Carpan (Com, 1021)
input: (None, 56, 56, 128)	flatten_1: Flatten input: (None, 7, 7, 512)	
block3_conv1: Conv2D output: (None, 56, 56, 256)	output: (None, 25088)	dropout_3: Dropout input: (None, 1024) output: (None, 1024)
block3_conv2: Conv2D input: (None, 56, 56, 256) output: (None, 56, 56, 256)		
	dense_1: Dense input: (None, 25088) output: (None, 4096)	dense_5: Dense input: (None, 1024) output: (None, 1024)
block3_conv3: Conv2D input: (None, 56, 56, 256) output: (None, 56, 56, 256)	output: (None, 4096)	
	<b>↓</b>	input: (None, 1024)
block3_pool: MaxPooling2D input: (None, 56, 56, 256)	dropout_1: Dropout input: (None, 4096)   lstm_1_input: InputLayer input: (None, 30, 384)	activation_3: Activation input: (None, 1024) output: (None, 1024)
output: (None, 28, 28, 256)	output: (None, 4096) Istin_1_input_ayer output: (None, 30, 384)	
block4_conv1: Conv2D input: (None, 28, 28, 256)		input: (None, 1024)
output: (None, 28, 28, 512)	input: (None, 4096) input: (None, 30, 384)	dropout_4: Dropout output: (None, 1024)
input: (None, 28, 28, 512)	dense_2: Dense	
block4_conv2: Conv2D output: (None, 28, 28, 512)	output. (Note, 4090)	input: (None, 1024)
input: (None, 28, 28, 512)		dense_6: Dense output: (None, 1000)
block4_conv3: Conv2D output: (None, 28, 28, 512)	merge_1: Merge input: [(None, 512), (None, 4096)]	
	output: (None, 4608)	input: (None, 1000)
block4_pool: MaxPooling2D input: (None, 28, 28, 512) output: (None, 14, 14, 512)		activation_4: Activation output: (None, 1000)
	<b>*</b>	

# Neuraltalk2



a man is playing tennis on a tennis court



a train is traveling down the tracks at a train station



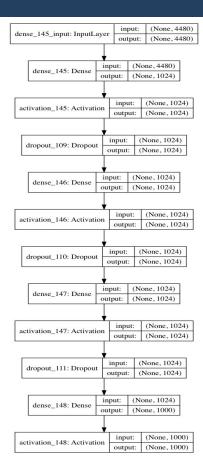
a cake with a slice cut out of it



a bench sitting on a patch of grass next to a sidewalk

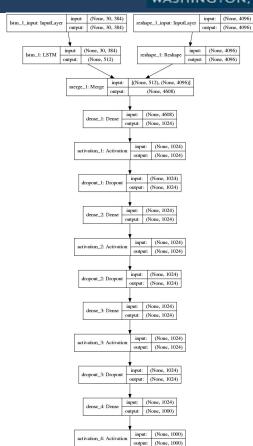
### Model





Multi-layer-Perceptron(MLP)

Long-Short-Term-Memory(LSTM)



# **Training**



#### MLP

Epochs number	1	5	10	 20
Time(min)	36.5	182.5	420	 1680
Loss(Ir = 0.01)	5.25	3.89	3.08	 3.07

#### **LSTM**

Epochs Number	1	5	10	 20
Time(min)	63	315	630	 2520
Loss(lr = 0.01)	3.52	3.01	2.90	 3.02

## Accuracy

THE GEORGE WASHINGTON UNIVERSITY WASHINGTON, DC

Our MLP:24.0%

Our LSTM:36.5%

2017 challenge 1st: 69%

2017 challenge 27th: 37.3%



what color is that train?

# Loaded /Users/zack wang/a UserWarning)

42.28 % blue

20.94 % red

07.67 % orange

07.64 % silver

07.37 % yellow

### **Problems & Future work**



 The VGG Model training implement was limited by the computational power.

 The loss of both MLP/LSTM model went smooth at around 3 after 10 epochs.