Ask Your Neurons: A Neural-based Approach to Answering Questions about Images

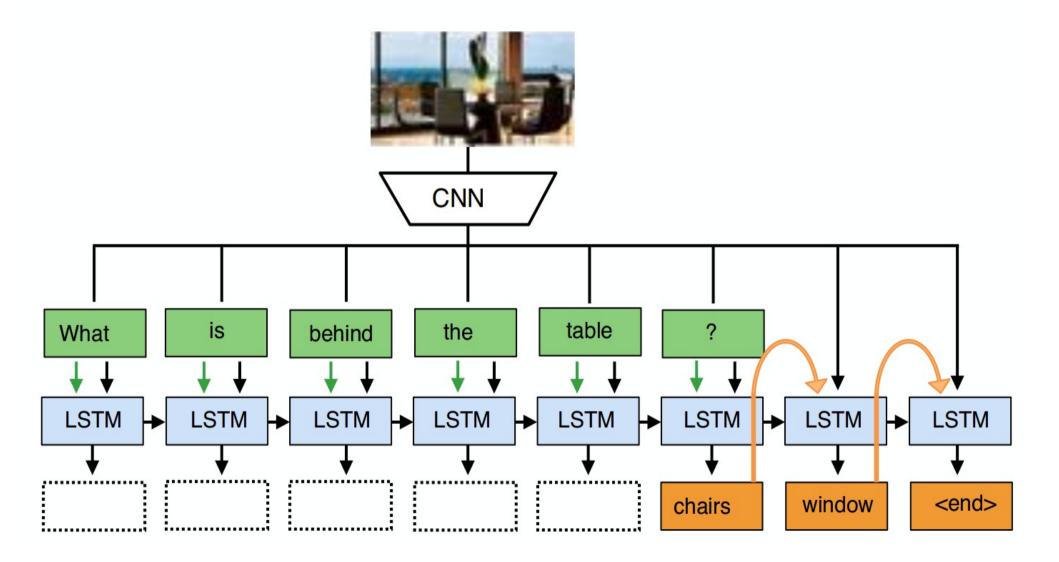
Authors Malinowski et al.

Experiment by Huihuang Zheng

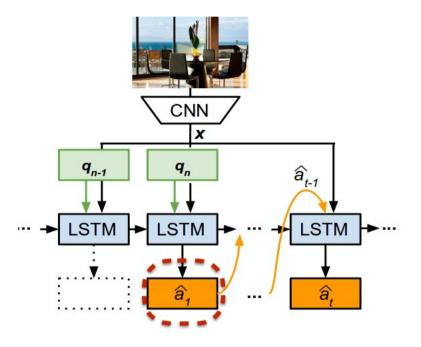
Outline

- Method in the paper
- Successful & fail examples
- Question without image
- Ask questions not relating to image
- Image with Gaussian Filter Smoothing
- Conclusion

Neural Approach to Answer



Formula:



- Predicting answer sequence
 - Recursive formulation

 $\begin{aligned} \hat{\boldsymbol{a}}_t &= \operatorname*{arg\,max}_{\boldsymbol{a} \in \mathcal{V}} p(\boldsymbol{a} | \boldsymbol{x}, \boldsymbol{q}, \hat{A}_{t-1}; \boldsymbol{\theta}), \, \boldsymbol{x} \text{-} \text{ image representation} \\ \boldsymbol{q} &= \left[\boldsymbol{q}_1, \dots, \boldsymbol{q}_{n-1}, \llbracket ? \rrbracket \right], \, \boldsymbol{q}_j \text{-} \text{ question word index} \\ \mathcal{V} \text{-} \text{ vocabulary,} \quad \hat{A}_{t-1} = \{ \hat{\boldsymbol{a}}_1, \dots, \hat{\boldsymbol{a}}_{t-1} \} \text{-} \text{ previous answer words} \end{aligned}$

Slide Credit: Malinowski, Rohrbach and Fritz

Experiment 1: Success & Fail

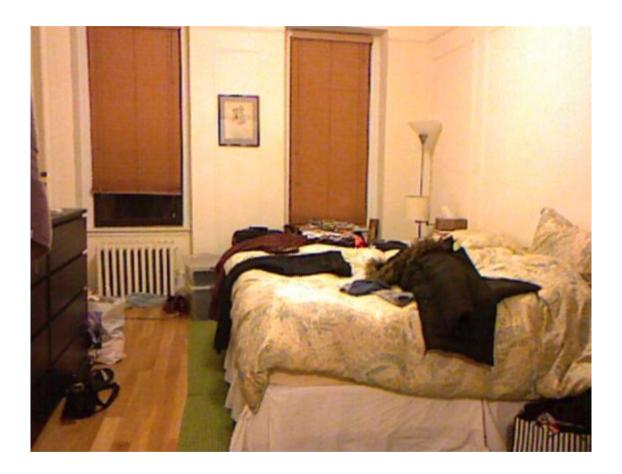
- Experiment setting:
 - Hardware
 - CPU: Intel® Core™ i7-4720HQ CPU @ 2.60GHz × 8
 - GPU: Geforce GTX 960M in laptop
 - Memory: 16GB
 - Tool:
 - Caffe-recurrent (caffe branch with RNN)
 - Python script
 - From original paper example and I modified to do other things

Experiment 1: Success & Fail

- Experiment setting:
 - Dataset:
 - DAQUER test set (dataset in paper)
 - Other dataset:
 - VQA (selected indoor images)*
 - MIT indoor scene
 - Images from Google image search

What Question: Correct

- what is on the bed ?
- clothes



How Many Question: Correct

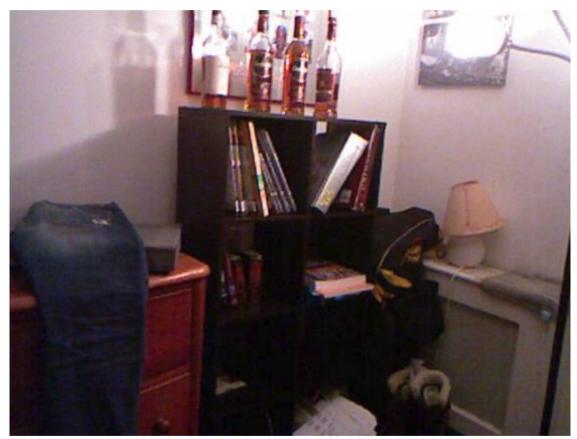
 how many doors are there?

• 1



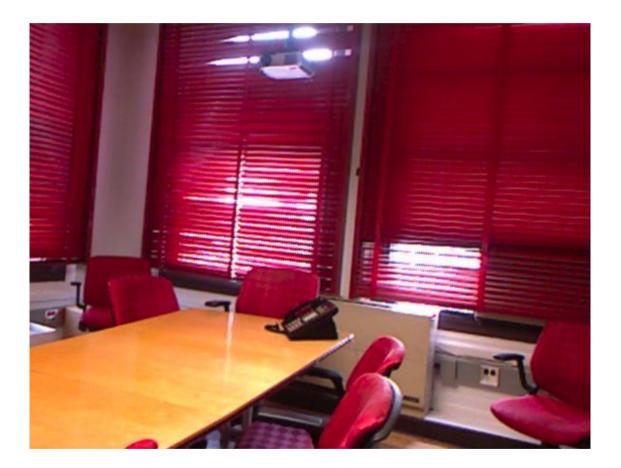
What Question: error

- what are the things on the shelf
 ? (answer: books)
- Book, photo, toy



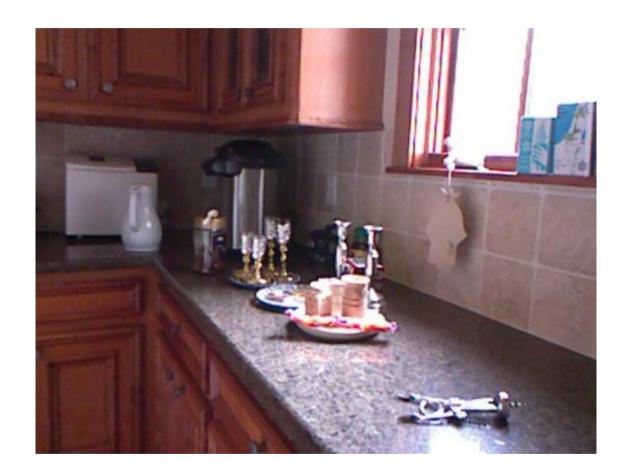
What Question: error

- what is the colour of the screen ?
 (answer: red)
- green



How Many Question: error

- how many glass cups are there? (answer 4)
- 1



Summary 1

- The NLP for questions
 - Answer object for what question
 - Answer number for how many question
 - Among first 1000 questions, only 2 (Q574, Q779) answer number fro what question. 1 (Q947) answer color for object question.
- Combining image feature and NLP is not easy.
 - GoogleNet, extract great image feature.

NLP error: Answer number for color

 what color are the bookshelves in this picture in the image84 ? (answer: brown)

• 1



NLP error: Answer Number for What Question

- what objects are on the cabinet below the medal rack? (answer: paper, photo, cup)
- 1



NLP error: Answer object for color

- what color are the drawer knobs ? (answer: white)
- pillow

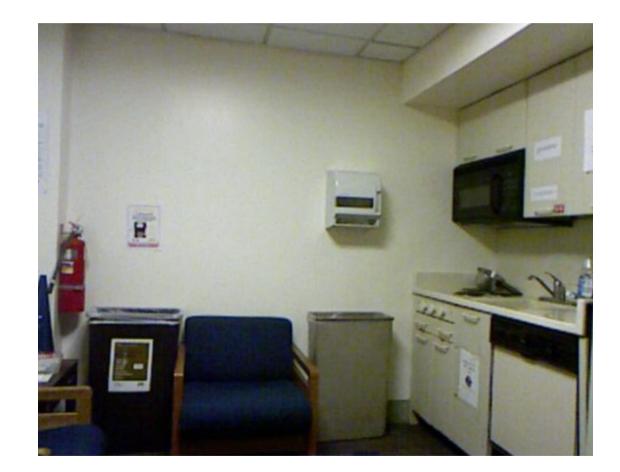


Experiment 2: Question without image

- Experiment setting:
 - Input black image, DAQUAR questions
 - Compare answers to original answers
- Result:
 - Only 1457 / 5673 (25.68%) answers are different
 - Suggest that LSTM doesn't learn a lot from images in training.

Black vs Origin

- Q0: what is on the left side of the white oven on the floor and on right side of the blue armchair ?
 - (garbage bin)
- Black image: pillow
- True image: pillow



Black vs Origin

- Q1: what is the object placed in front of the filling shelves ?
 - (table)
- Black image: projector_screen
- True image: paper



Black vs Origin

- Q2: how many garbage bins are there ? (2)
- A2:
- Black image: 1
- True image: 1



Experiment 3: Ask questions not relating to image

- Experiment setting:
 - Shuffle DAQUAR test images
 - Compare answers to original answers
- Result:
 - Only 699 / 5673 (12.32%) answers are different

Black vs Shuffle vs Origin

- Q0: what is on the left side of the white oven on the floor and on right side of the blue armchair ?
 - (garbage bin)
- Black: pillow
- Wrong Image: pillow
- True Image: pillow





Black vs Shuffle vs Origin

- Q1: what is the object placed in front of the filling shelves ?
 - (table)
- Black image: projector_screen
- Wrong image: projector_screen
- True image: paper





Black vs Shuffle vs Origin

- Q2: how many garbage bins are there ? (2)
- Black image: 1
- Wrong image: 1
- True image: 1





Experiment 3

- Experiment setting:
 - Perform transformation in images to see whether the answer is invariant under some translation
 - Gaussian Smooth DAQUAR test images
 - For 3 color channels, Gaussian filter with size 5.
 - Move down the image top edge by 80 pixels. (origin 425 * 500 pixels)
- Result:
 - 785 / 5673 (13.84% vs 12.32% of random shuffle) answers are different in Gaussian Filter
 - Not invariant to Gaussian Filter
 - 337 / 5673 (5.9 %) answers are different in moving down
 - Somehow invariant to moving, shifting

Wrong Image but Correct Answer!

- Question: what is on the left side of the toilet? (tissue_roll)
- Black & Gaussian: glass
- Wrong & true image: tissue_roll







Wrong Image but Correct Answer!

- what is the white object on top of the brown table? (printer)
- Black & Gaussian: vase
- True image: book
- Wrong image: printer



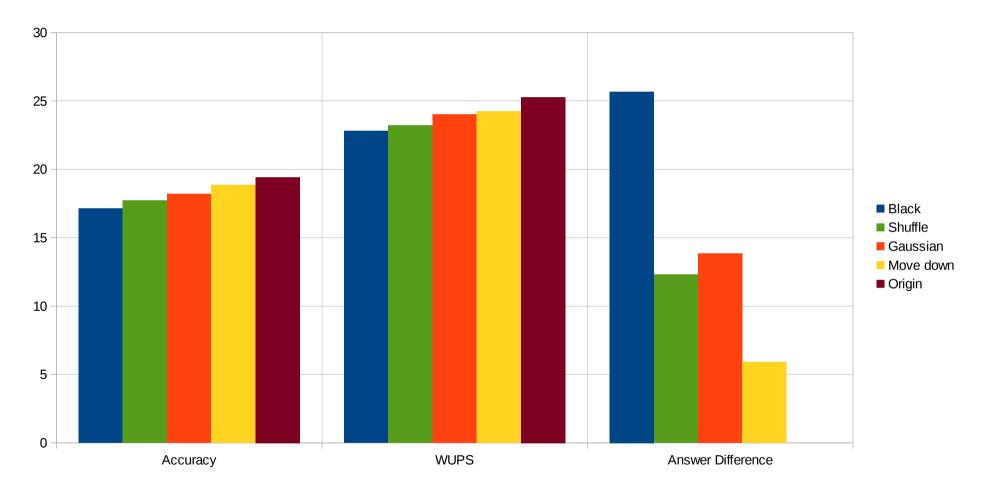




Quantity Summary:

	Accuracy (%)	WUPS (%)	Answer Diff (%)
Black	17.15	22.8	25.68
Random Shuffle	17.73	23.2	12.32
Gaussian Filtered	18.18	24.01	13.83
Movedown	18.86	24.23	5.9
Origin	19.42	25.18	0

Experiment Accuracy (%)



Accuracy is similar!

Summary 2

- LSTM doesn't learn a lot from image features of CNN in training, which causes bad performance.
- That illustrates why vision + language doesn't improve much from language only

Methods	Accuracy	WUPS @0.9
Baseline: Symbolic (NIPS'14)	7.86%	11.86%
Language Only (Our)	17.15%	22.80%
Vision + Language (Our)	19.43%	25.28%
Human performance (NIPS'14)	50.20%	50.82%

Result Table from: Malinowski, Rohrbach and Fritz

Conclusion

- In the experiments
 - CNN + LSTM learns great in NLP question
 - Answer numbers for how many question
 - Answer object for what question
 - Answer colors for color question
 - LSTM doesn't learn a lot from CNN feature
 - The result using correct or wrong input images doesn't influence much in result.
 - Still a long way to go.
- Thank you : -)

Reference

- Ask Your Neurons: A Neural-based Approach to Answering Questions about Images.
 - M. Malinowski, M. Rohrbach, and M. Fritz. arXiv 2015. (ICCV'15, Oral, to appear).
- Their ICCV 2015 (Oral) slide