VQA-E: Explaining, Elaborating, and Enhancing Your Answers for Visual Questions

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Introduction

Goal: Generate Textual Justifications for Predicted Answers

- Be Accessible to visually impaired people.
- Provide beneficial feedbacks that enable the questioners to extend the conversation for effective communication.

Q:What is the hotel's name?
A:Butternut.

From where can we tell the name of the hotel?

Explaining....

E:A view of a red brick building which has a sign that says 'BUTTERNUT' on the side.

Q:Is the elephant in the boat?
A:No.

It is not in the boat. Then where is it?

Elaborating....

E:An elephant is walking down the hill near a boat in the water.

Q:Is the dog wearing anything?
A:Yes..

'Anything' is too vague to tell what the dog is wearing.

Enhancing....

E:A dog in a madonna shirt is sitting next to feet in high heels.

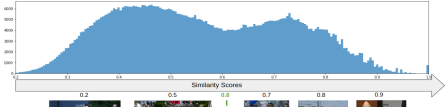
Contributions:

- Constructed a new dataset with textual justifications for the answers, which is automatically derived from the VQA v2 dataset by intelligently exploiting the available captions.
- Proposed a novel multi-task learning framework which can generate a sentence to explain the predicted answer.
- Outperformed the state-of-the-art methods by a clear margin on the VQA v2 dataset.

VQA-E Dataset

- Explanation Synthesis

- Dataset Examples

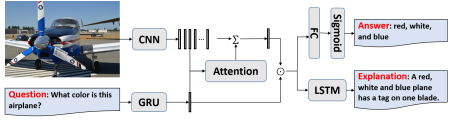


- Dataset Analysis

Dataset	Split	#Images	#Q&A	#E	#Unique Q	#Unique A	#Unique E
VQA-E	Train	72,680	181,298	181,298	77,418	9,491	115,560
	Val	35,645	88,488	88,488	42,055	6,247	56,916
	Total	108,325	269,786	269,786	108,872	12,450	171,659
VQA-v2	Train	82,783	443,757	0	151,693	22,531	0
	Val	40,504	214,354	0	81,436	14,008	0
	Total	123,287	658,111	0	215,076	29,332	0

Download: https://github.com/liqing-ustc/VQA-E

Multi-task Model



$$L = L_{\text{vqa}} + L_{\text{vqe}}$$

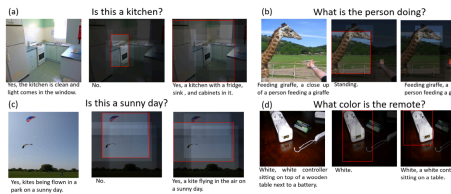
Experiment Results

Explanation Generation

Model	Image features	B-1	B-2	B-3	B-4	M	C	R
Q-E	-	26.80	10.90	4.20	1.80	7.98	13.42	24.90
I-E	Global	32.50	17.20	9.30	5.20	12.38	48.58	29.79
Q&I-E	Global	34.70	19.30	11.00	6.50	14.07	61.55	31.87
	Grid	36.30	21.10	12.50	7.60	15.30	73.70	34.00
	Bottom-up	38.00	22.60	13.80	8.60	16.57	84.07	34.92
Q-AE	Global	35.10	19.70	11.30	6.70	14.40	64.62	32.39
	Grid	38.30	22.90	14.00	8.80	16.85	87.04	35.16
	Bottom-up	**39.30**	**23.90**	**14.80**	**9.40**	**17.37**	**93.08**	**36.33**

Answer Prediction

Model	Image features	All	Yes/No	Number	Other
Q-A	Global	57.26	77.19	39.73	46.74
	Grid	59.25	76.31	39.99	51.38
	Bottom-up	61.78	78.63	41.30	52.54
Q-AE	Global	57.92	78.01	40.46	47.25
	Grid	60.57	78.35	39.36	52.66
	Bottom-up	**63.51**	**80.85**	**43.02**	**54.16**
QI-AE(random)	Bottom-up	58.74	78.75	40.79	48.26
QI-AE(relevant)	Bottom-up	62.18	79.02	41.07	53.26



Insight: the additional supervision from explanations helps the model better localize and understand the important image regions.