TBD - CLIP

Exploring Visual-Dialog Conversation



Rob



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Visual Question Answering

A) Supervised Loss

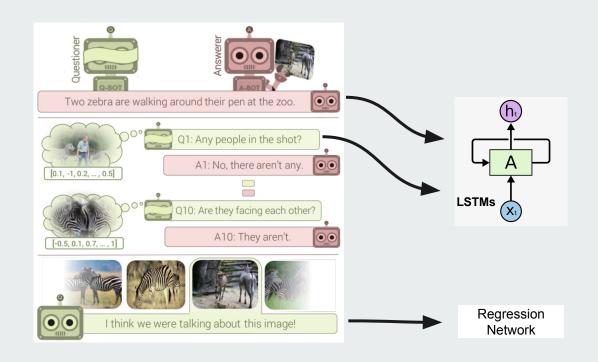
- Image Embedding prediction.
- Word level tokens.
- Diversity Loss

B) RL Fine-tuning

- Distance based rewards.
- Reinforce

Das, A. et al. "Visual Dialog." 2017 IEEE Conference on Computer Vision and Pattern Recognition (CVPR) (2017): 1080-1089.

Murahari, Vishvak S. et al. "Improving Generative Visual Dialog by Answering Diverse Questions." EMNLP/IJCNLP (2019).



Create software that can have meaningful conversations with humans about images

War of Embeddings

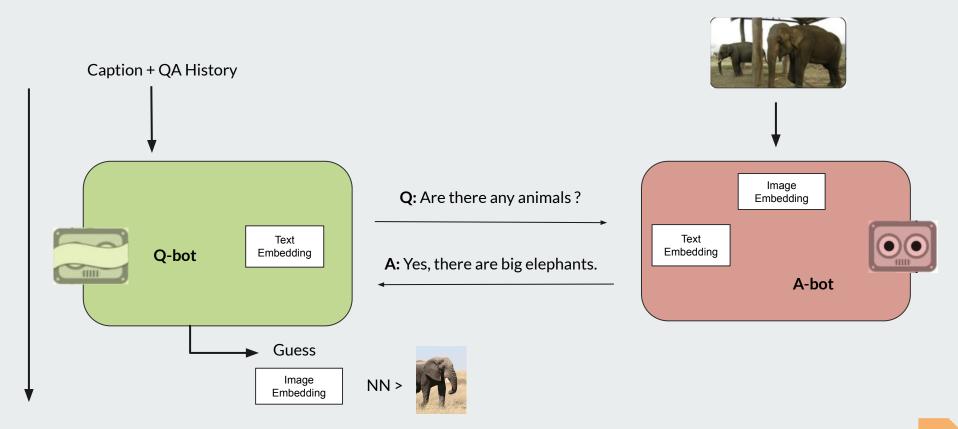
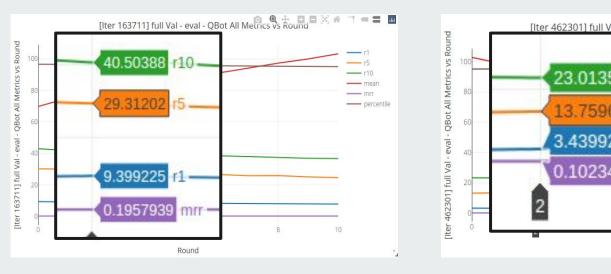
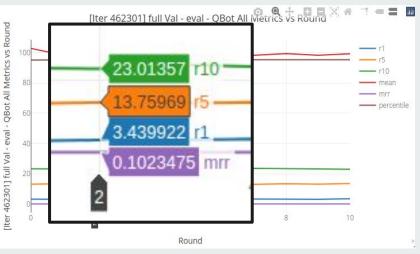


Image Embeddings | VGG -> CLIP



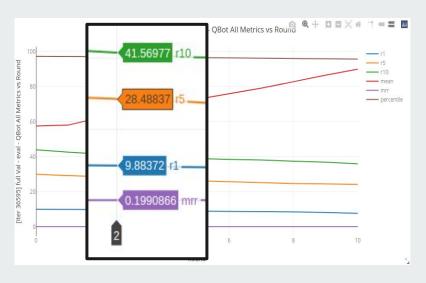
CLIP



VGG Embeddings (Murahari et al.)

OpenAl's CLIP uses a Vision Transformer (ViT) to embed images. The old way used embeddings from VGG16.

Text Embeddings | From Scratch -> Pretrained Glove



CLIP + Glove (fixed)



CLIP + (from scratch)

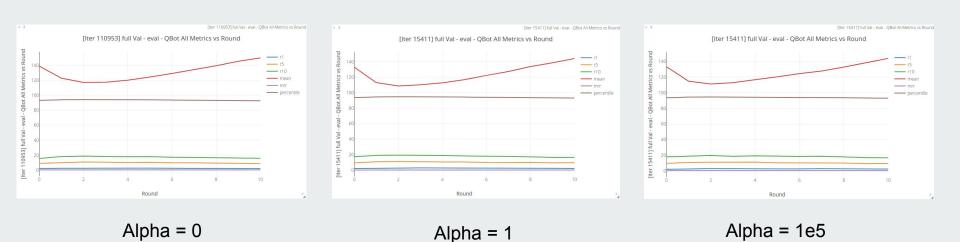
RL Fine-Tuning | Reinforce -> PPO

```
function REINFORCE Initialise \theta arbitrarily for each episode \{s_1, a_1, r_2, ..., s_{T-1}, a_{T-1}, r_T\} \sim \pi_{\theta} do for t=1 to T-1 do \theta \leftarrow \theta + \alpha \nabla_{\theta} \log \pi_{\theta}(s_t, a_t) v_t end for end for return \theta end function
```

PPO

- Make a ratio of word probabilities between the current & old policy.
- Allows for more sensible weight updates.
- Reduces training time.

RL Fine-Tuning | Contribution ?



Testing for different Coefficient for RL Fine Tuning - Negligible Change in result.

Conclusion

- Visual Dialog Conversation is still in its developmental phase.
- CLIP embeddings do better at relating language with images
 - Big compute + multi-model pre-training is good,
- RL objective does not produce better results.
 - PPO over Reinforce is still an open question.

Exploring Visual-Dialog Conversation = Lots of models, metrics and fun

Demo Time:)

Next up, Questions!







Results and Visualizations

Web Demo

Code