# Visual Dialog

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## Visual Dialog

- Requires an Al agent to hold a meaningful dialog with humans about visual content.
- Input:
  - Image
  - Dialog history
  - Question
- Output:
  - Answer to the question



## VQA vs Visual Dialog



#### **VQA**

Q: How many people on wheelchairs?

A: Two

Q: How many wheelchairs?

A: One

#### Captioning

Two people are in a wheelchair and one is holding a racket.

#### **Visual Dialog**

Q: How many people are on wheelchairs?

A: Two

Q: What are their genders?

A: One male and one female

Q: Which one is holding a

racket?

A: The woman



#### **Visual Dialog**

Q: What is the gender of the one in the white shirt?

A: She is a woman

Q: What is she doing?

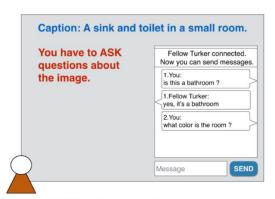
A: Playing a Wii game

Q: Is that a man to her right

A: No, it's a woman

### VisDial Dataset

- Contains ~123k images and 10 question-answer pairs for each image
- Images are from COCO dataset
- Question-answer pairs are collected on AMT with human dialog





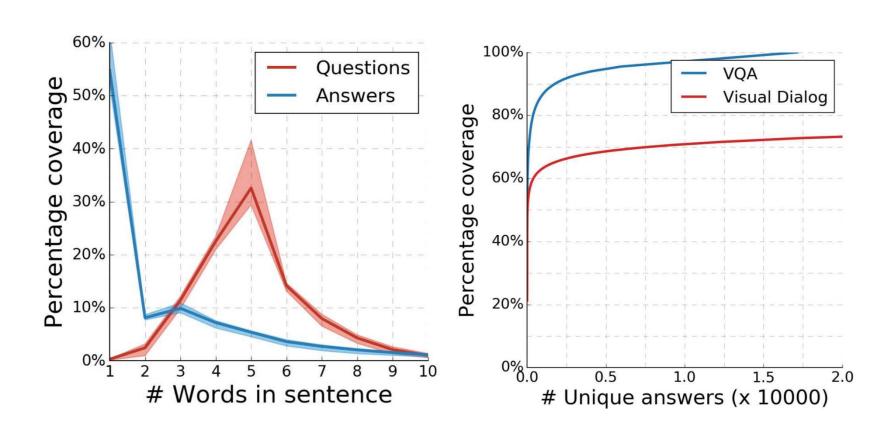


(a) What the 'questioner' sees.

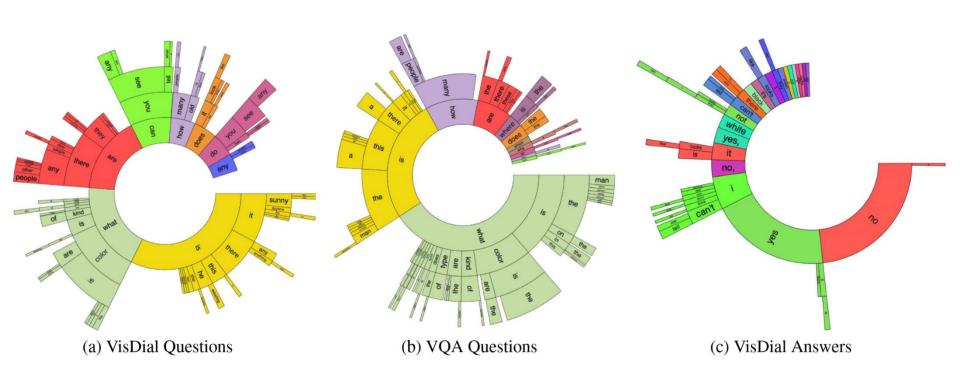
(b) What the 'answerer' sees.

(c) Example dialog from our VisDial dataset.

### VisDial Dataset



### VisDial Dataset



### **Evaluation**

- Given N = 100 candidate answers, return a sorting of them
- Candidate answers:
  - The human response
  - Answers to 50 most similar questions
  - 30 most popular answers from the dataset
  - 19 random answers
- Retrieval metrics:

MRR, recall@k, average rank of the human response

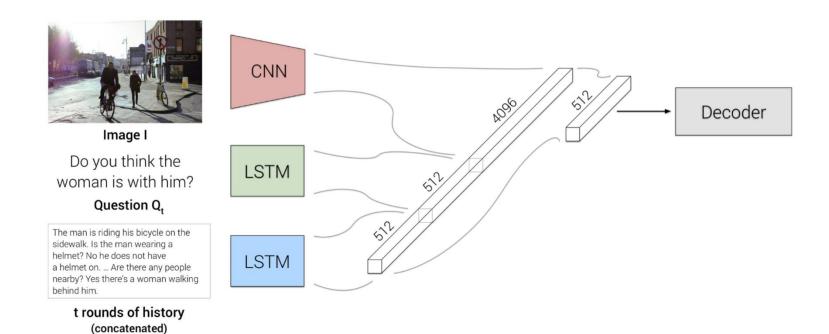
### Models

- Following the encoder-decoder framework
- 2 kinds of decoder
  - Generative Decoder
  - Discriminative Decoder
- 3 kinds of encoder
  - Late Fusion Encoder
  - Hierarchical Recurrent Encoder
  - Memory Network Encoder

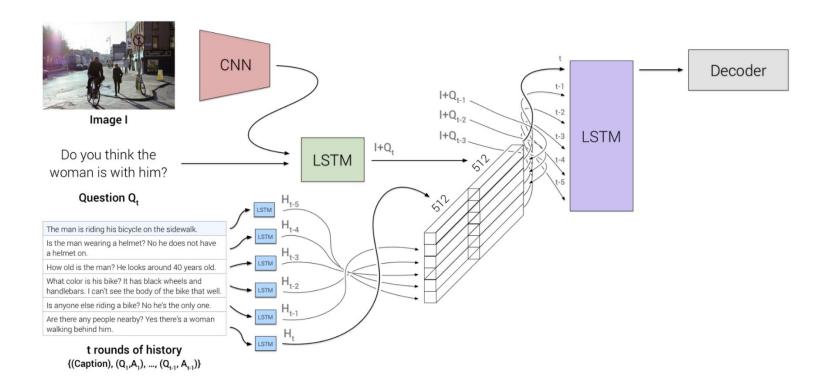
### Decoders

- Generative Decoder
  - LSTM decoder
  - Maximize the log-likelihood of the ground truth answer
  - Use the model's log-likelihood scores for ranking
- Discriminative Decoder
  - Compute similarity between the input encoding and LSTM encoding for candidate answers
  - Maximize softmax score of the ground truth answer
  - Use the similarities for ranking

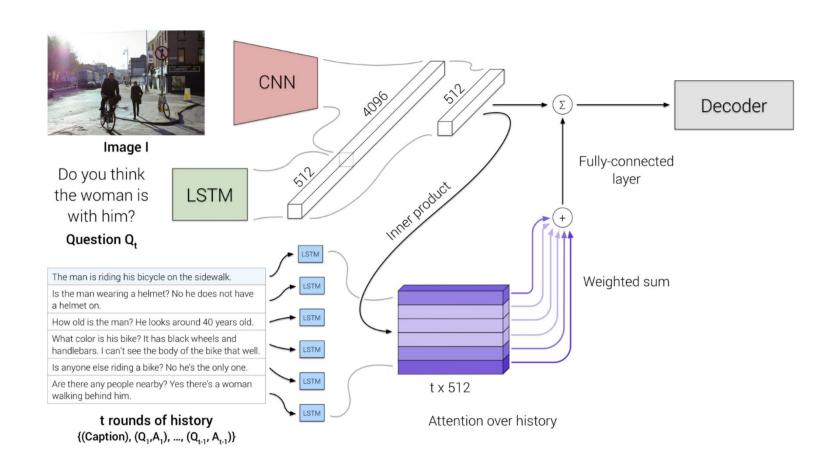
## Late Fusion (LF) Encoder



## Hierarchical Recurrent Encoder (HRE)



## Memory Network (MN) Encoder



### Experiments

Dataset: VisDial v0.9

#### Baseline

• NN-Q:

Find k nearest neighbor questions for a test question, and score answers by their mean similarity with these k answers

NN-QI:

Find K nearest neighbor questions for a test question, then find a subset of size k based on image feature similarity. Score answers by their mean similarity with these k answers

### VQA models

- SAN and HieCoAtt
- Feed VQA outputs to their discriminative decoder, and train end-toend on VisDial

## Results

	Model	MRR	R@1	R@5	R@10	Mean
Baseline	Answer prior	0.3735	23.55	48.52	53.23	26.50
	NN-Q	0.4570	35.93	54.07	60.26	18.93
	NN-QI	0.4274	33.13	50.83	58.69	19.62
Generative	LF-Q-G	0.5048	39.78	60.58	66.33	17.89
	LF-QH-G	0.5055	39.73	60.86	66.68	17.78
	LF-QI-G	0.5204	42.04	61.65	67.66	16.84
	LF-QIH-G	0.5199	41.83	61.78	67.59	17.07
	HRE-QH-G	$0.5\overline{1}0\overline{2}$	$-40.\overline{15}$	61.59	67.36	17.47
	HRE-QIH-G	0.5237	42.29	62.18	67.92	17.07
	HREA-QIH-G	0.5242	42.28	62.33	68.17	16.79
	$\overline{MN}$ - $\overline{QH}$ - $\overline{G}$	$0.5\overline{1}1\overline{5}$	$-40.\overline{42}$	$\bar{6}1.\bar{5}\bar{7}$	$-67.\overline{44}^{-}$	17.74
	MN-QIH-G	0.5259	42.29	62.85	68.88	17.06
Discriminative	LF-Q-D	0.5508	41.24	70.45	79.83	7.08
	LF-QH-D	0.5578	41.75	71.45	80.94	6.74
	LF-QI-D	0.5759	43.33	74.27	83.68	5.87
	LF-QIH-D	0.5807	43.82	74.68	84.07	5.78
	HRE-QH-D	0.5695	$-4\overline{2}.\overline{7}0^{-}$	73.25	82.97	6.11
	HRE-QIH-D	0.5846	44.67	74.50	84.22	5.72
	HREA-QIH-D	0.5868	44.82	74.81	84.36	5.66
	$\bar{M}\bar{N}$ - $\bar{Q}\bar{H}$ - $\bar{D}$	0.5849	$-44.\overline{0}3$	75.26	84.49	5.68
	MN-QIH-D	0.5965	45.55	76.22	85.37	5.46
VQA	SAN1-QI-D	0.5764	43.44	74.26	83.72	5.88
	HieCoAtt-QI-D	0.5788	43.51	74.49	83.96	5.84