

# Multimodal Coreference Resolution

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# TOSHIBA

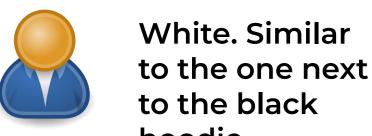
## Introduction

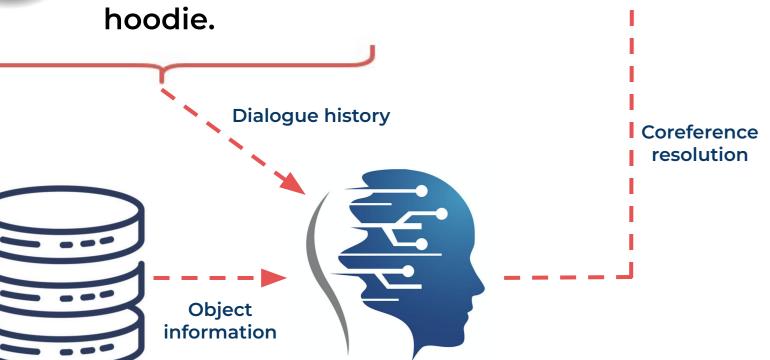


Hi, do you have short sleeve t-shirts?

Yes, we do.
What color do
you prefer?







- Applications:
  - E-commerce virtual assistant: answer customer inquiries about objects.
  - In-site interpreter: improved translations using both scene and textual context.
  - Boost other natural language tasks, like question answering or generation.
- SIMMC2 dataset published by Facebook Research is used for investigation.
  - It contains dialogues, object descriptions and scene images.
- The 10th Dialog System Technology Challenge (DSTC10) partially focused on the multimodal coreference resolution task.
  - SIMMC2 dataset was used for the competition.
  - Best performing systems are studied as a enhanced baseline.

# **SOTA MMCR Systems**

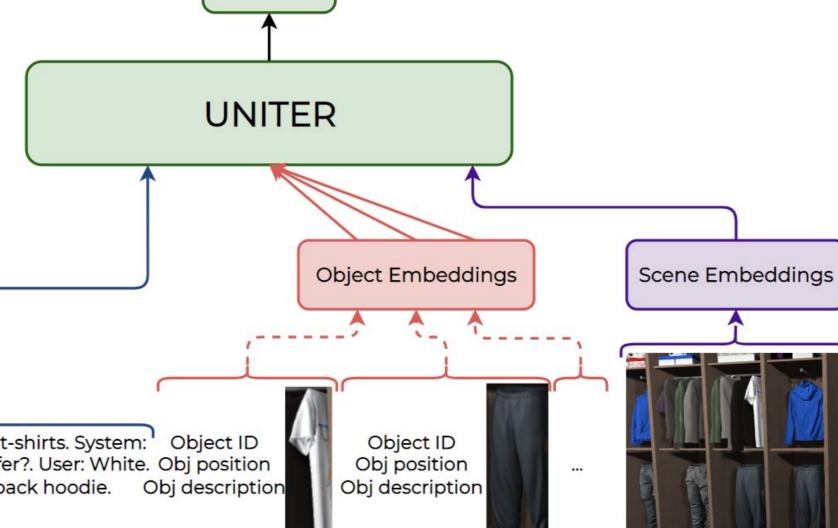
Coreference

Head

### **UNITER-based model**

- Strengths:
  - Visual features do not need to be heavily preprocessed
  - Easily adaptable to different domains
- Weaknesses:
- Computationally expensive:
   25-30 hours to train,
  - ~ 1 s. for response.

User: Hi, do you have short sleeve t-shirts. System: Object ID Yes, we do. What color do you prefer?. User: White. Obj position Similar to the one next to the back hoodie. Obj description



# BART Encoder BART Encoder BART encoder embedder Dialogue history Object 7 attributes Object 2 attributes Object position

<SYS> Yes, we do. What color do you prefer?. <USR><SOO><7><fashion\_82>[(1.2, 3.5, 2.7)]<21><fashion\_4>[(1.6, 3.5, 2.7)] ... <EOO> White. Similar to the one next to the back hoodie.

- Strengths:
  - Winner of DSTC10 on this task.
- Computationally cheaper:

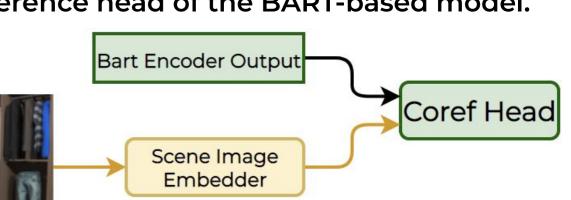
   around 5 hours to train,
   seconds for response.
- Weaknesses:
  - Bad at handling objects not seen in training.
- Scene images need to be described in natural language to be used.

# **Proposed improvements**

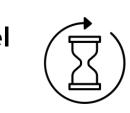
 Include object descriptions in the input of the BART-based model.



• Provide image embeddings to improve the coreference head of the BART-based model.



 Suppress object IDs in UNITER-based model to make it scene-independent.



## Results

Model	Object F1-Score
GPT-2 Baseline (Facebook Research)	36.6%
UNITER-based (New York Uni. Shanghai)	67.4.%
BART-based (KAIST & Samsung Research)	74.3%
BART using object descriptions (Ours)	76.1%

Multimodal Coreference Resolution performance on devtest split

## References

[1] Satwik Kottur et.al. SIMMC 2.0: A Task-oriented Dialog Dataset for Immersive Multimodal Conversations. *Association for Computational Linguistics*. 2021.

[2] Yichen Huang et. al. UNITER-Based Situated Coreference Resolution with Rich Multimodal Input. *Computing Research Repository*. 2021.

[3] Haeju Lee et. al. Tackling Situated Multi-Modal Task-Oriented Dialogs with a Single Transformer Model. *Association for Computational Linguistics*. 2021.

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