

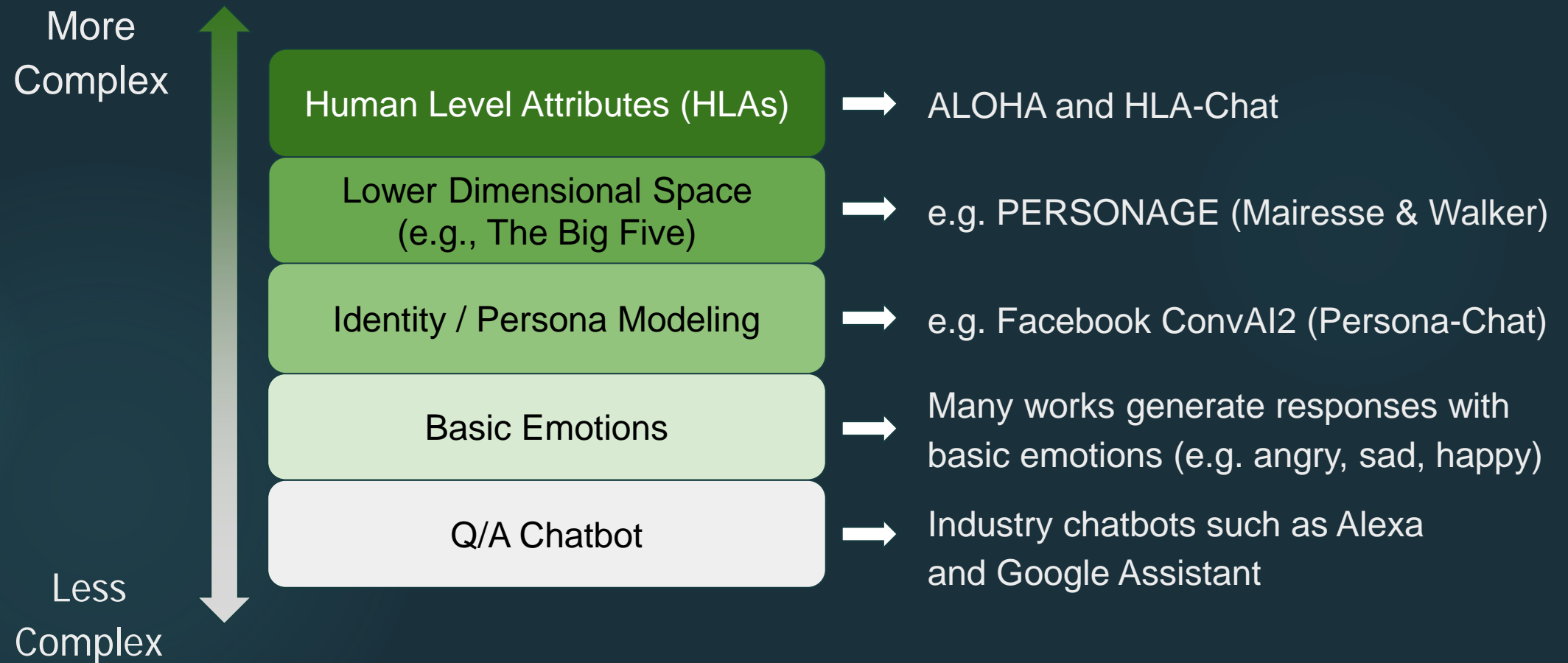
# ALOHA: Artificial Learning of Human Attributes for Dialogue Agents

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# Overall Goal

- ▶ Giving chatbots and virtual assistants the ability to imitate and express human emotions/personality
- ▶ How?
  - ❖ **Human-Level Attributes (HLA)** - Based on tropes: aspects of fictional characters representative of their identity
  - ❖ **HLA-Chat**: Dataset of characters with their HLAs + dialogue
  - ❖ **Artificial Learning of Human Attributes (ALOHA)**: System to retrieve character specific language models

# Related Work



# Human Level Attributes (HLAs)

Broad attribute: Friendly



[HLA: Childhood Friends]

...



[HLA: Vitriolic Best Buds]

Broad attribute: Trustworthy



[HLA: The Reliable One]

...



[HLA: Only Friend]

Broad attribute: Helpful



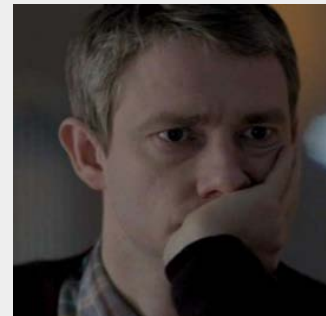
[HLA: The Caretaker]

...



[HLA: We Help the Helpless]

Broad attribute: Curious



[HLA: The Watson]

...

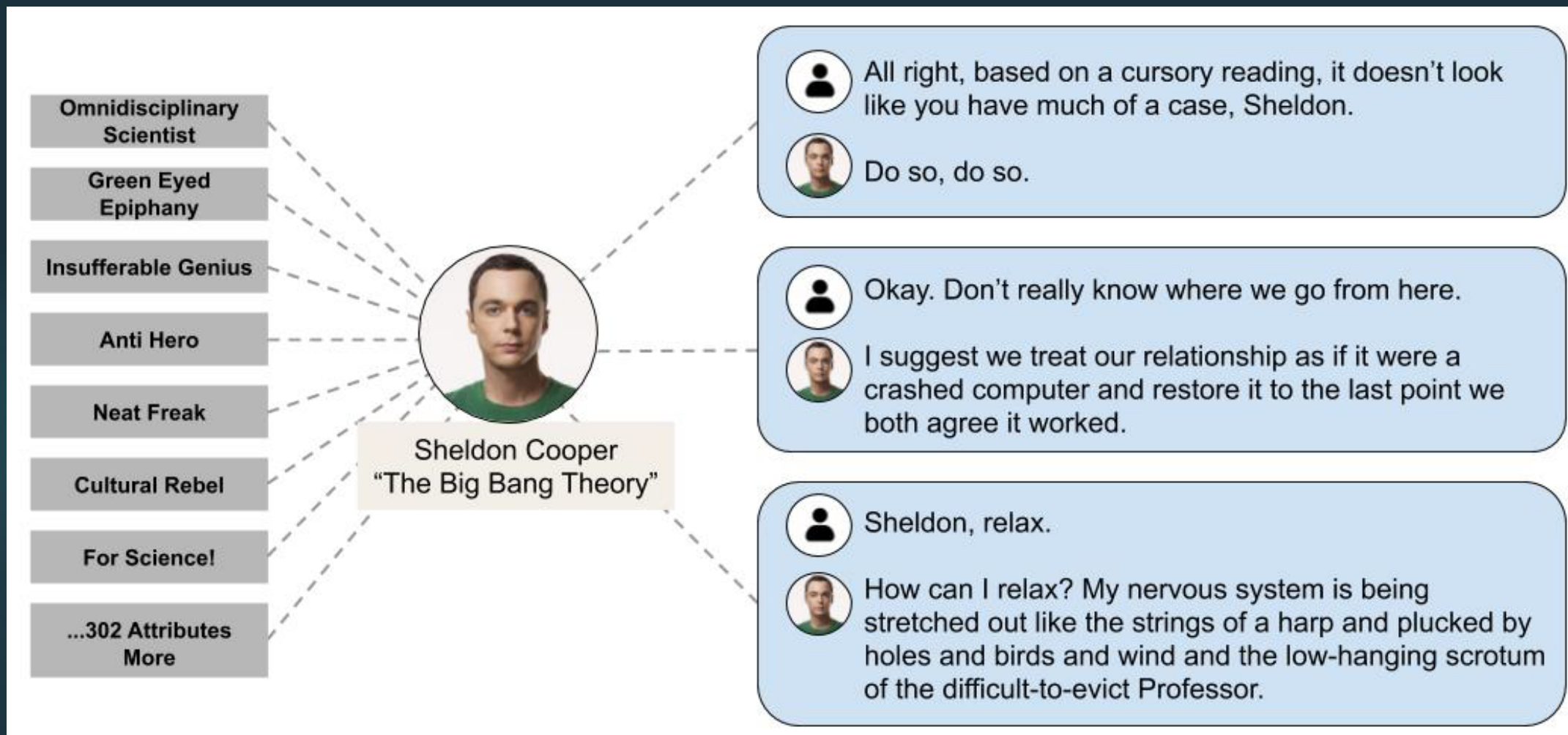


[HLA: Cute Bookworm]

# Our Dataset: HLA-Chat

- ▶ Present a dataset, **HLA-Chat**, with:
  - ❖ CHARACTERS
  - ❖ ATTRIBUTES (HLAs) of characters
  - ❖ Character DIALOGUES

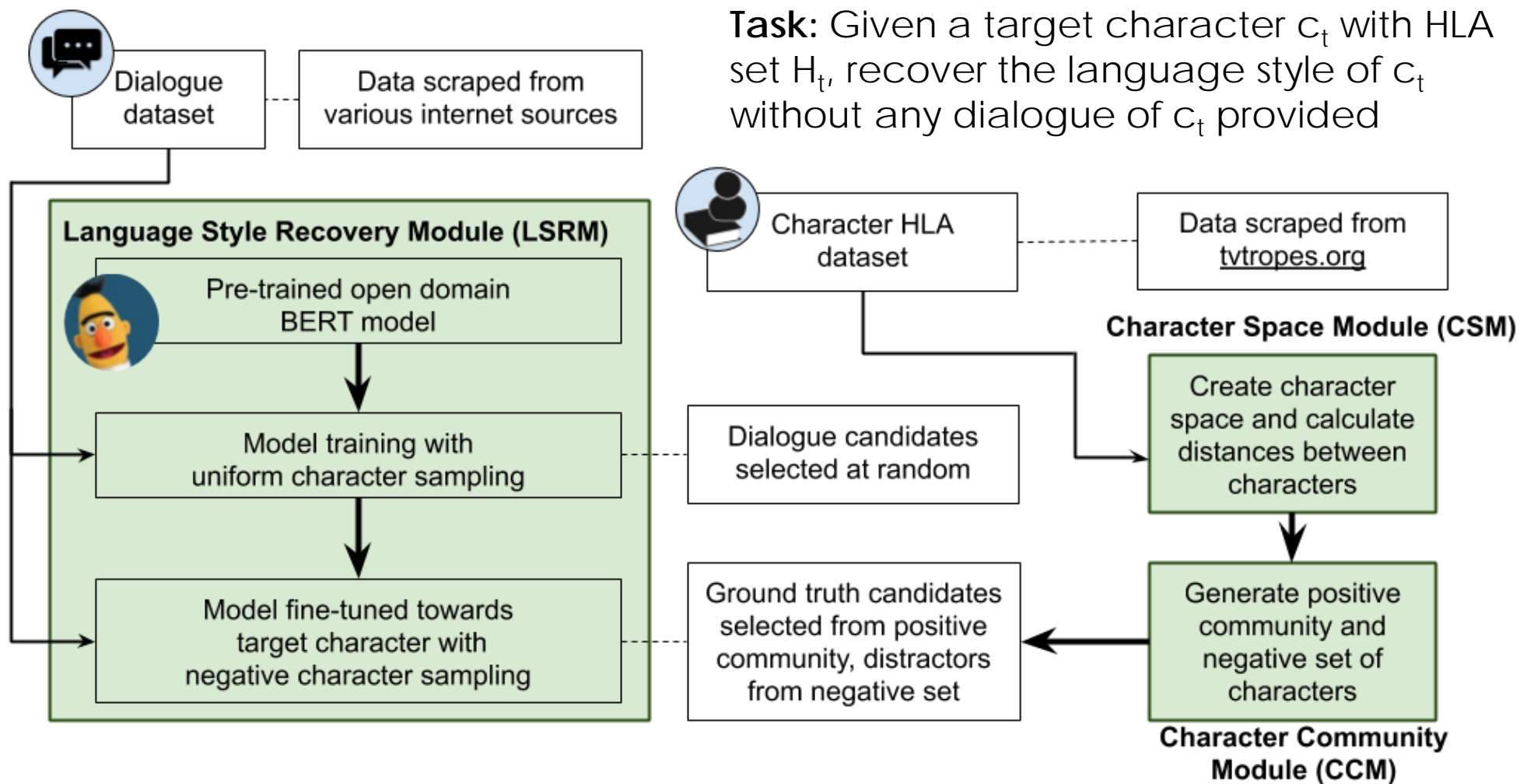
# Our Dataset: HLA-Chat



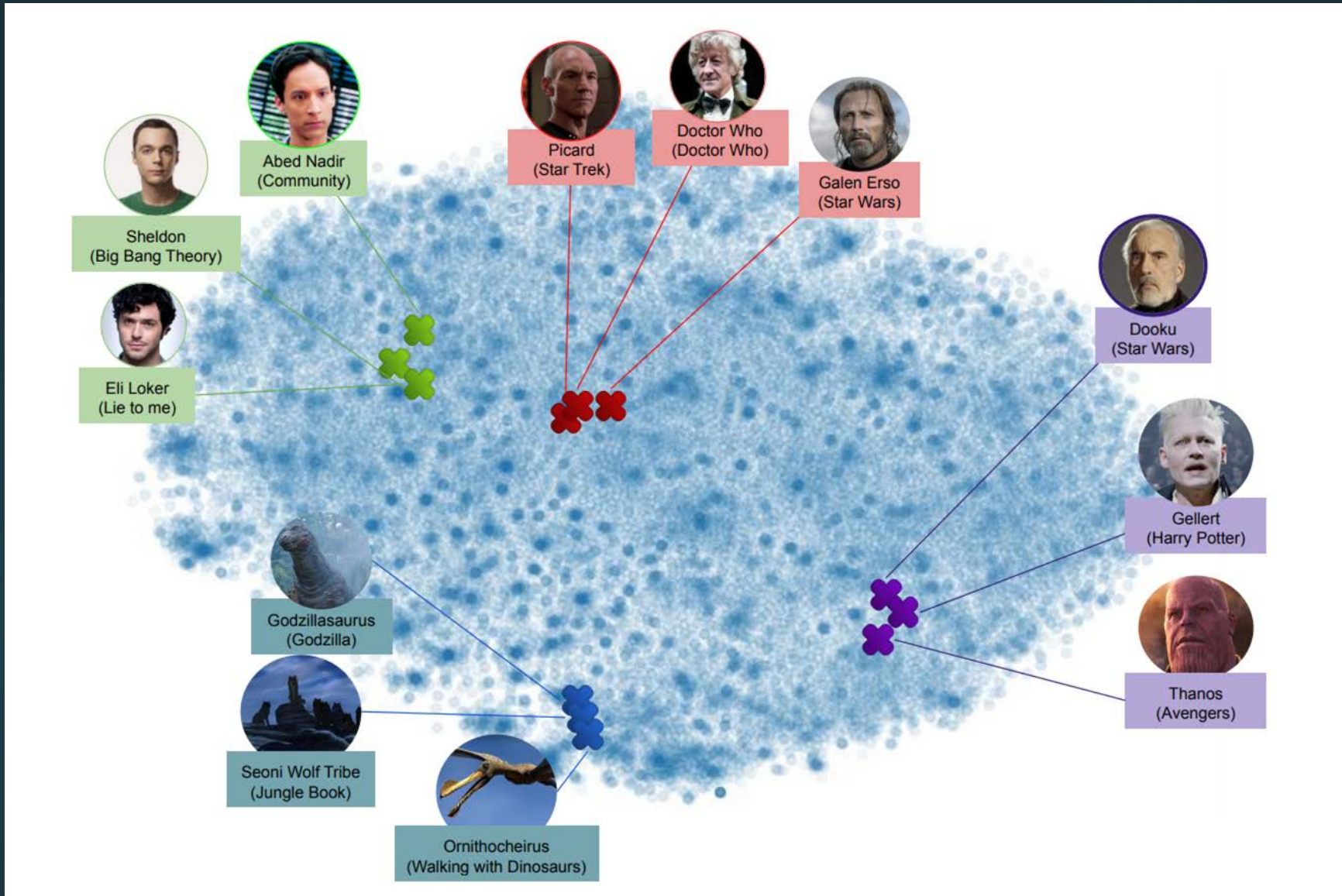


# Our System: ALOHA

7

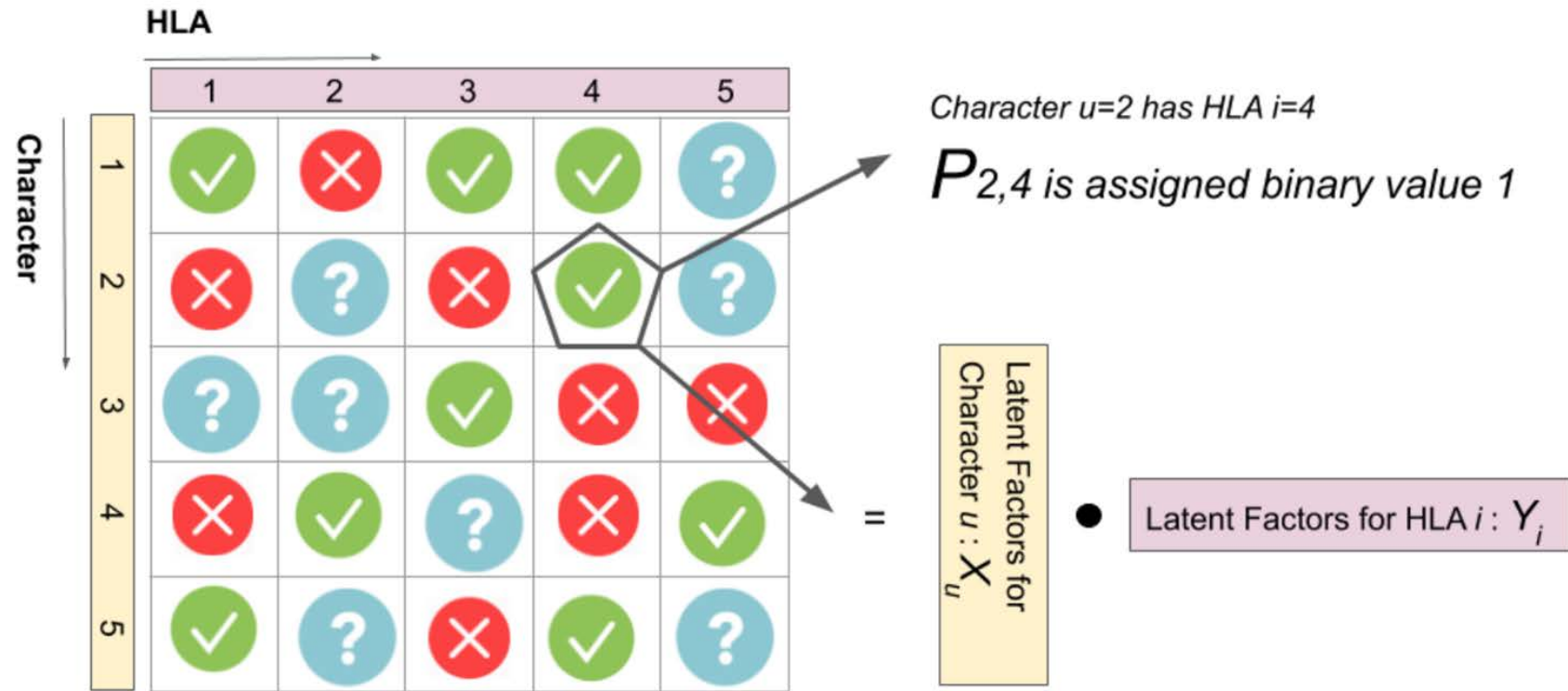


# Character Space Module (CSM)





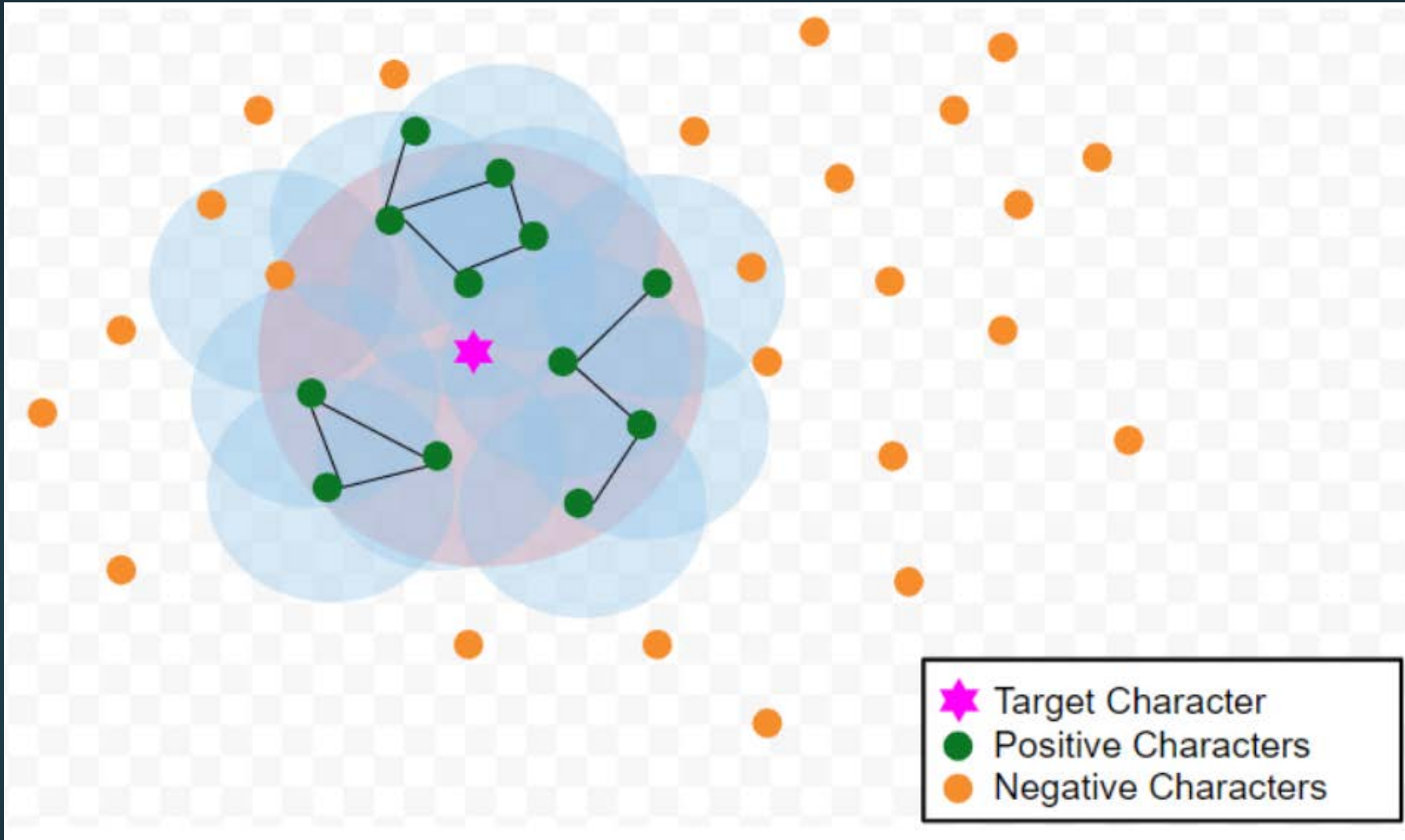
# CSM: Learning HLAs



$$loss = \sum_u \sum_i (\alpha P_{u,i} - X_u^T Y_i)^2 + \lambda (\|X_u\|^2 + \|Y_i\|^2)$$

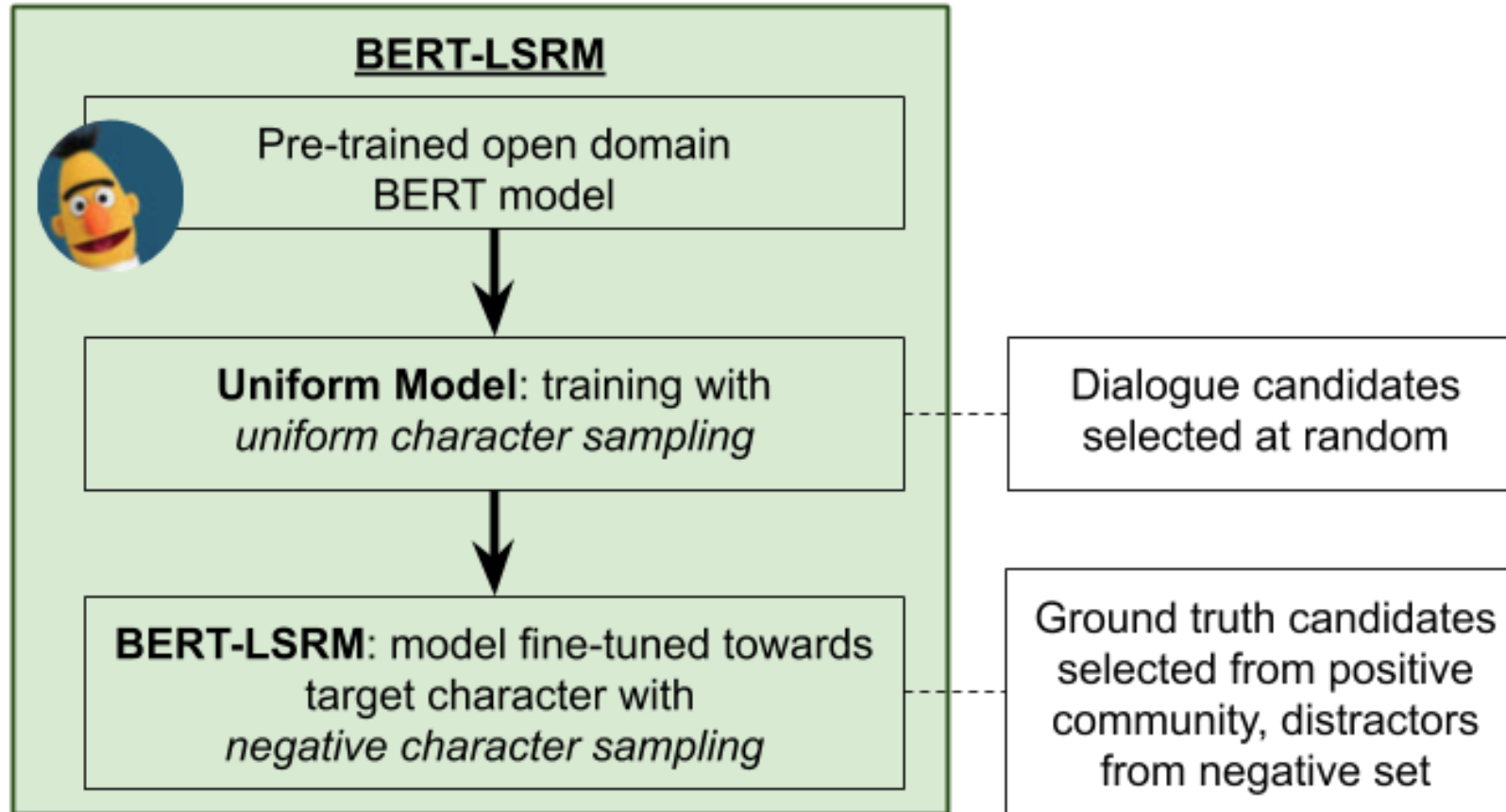
# Character Community Module (CCM)

10



# Language Style Retrieval Module (LSRM)

11



# Evaluation: Five Target Characters



# Evaluation Format

How would Sheldon Cooper respond to this:

All right, based on a cursory reading, it doesn't look like you have much of a case, Sheldon.



Yeah. But I think the thing that makes me maddest is that he's right.

- Not exactly a Manhattan hot spot.
- Nothing.
- It's Valentine's Day. [Name], of course.
- Well, the female body is a... work of art. The male body is utilitarian, it's for gettin' around, like a jeep.
- Yeah. About a quart of wine.



Do so, do so.

- No, I gotta talk to her today. If something's over, it's just got to be over. [Name], she's...she's not an affair.
- I cut off all my fingers? Ted, you're a great guy. I know it, you know it, she knows it. I would bet you a gazillion dollars--no, I'm even more confident. I would bet you a floppity jillion dollars that she's not calling to break up with you.
- I know that I want [Name] dead, which puts me squarely on team you. Besides, if you two ever need me to swap places with [Name] again, the less people know that I'm here, the better. Think about it, [Name]. Come on. Be smart.
- So, they deal with terrorism. Among other things. You know, Washington has had a wary eye on the heavens for decades, [Name], watching for any threat from beyond the stars. I've heard they call that program [Name].
- I'm sorry.
- Go. Now.
- Will you go to the prom with me?
- All this time I thought the reason I survived that Porsche accident was because of you, [Name]. Now I'm not so sure. See, I've always tried to explain everything by looking outside myself. But maybe the truth lies inside my own physiology. Maybe I am a freak.
- Oh it's your cat!
- A sore point with me and [Name].
- Very big blocks of time.
- [Name] [Name], hey, look, I'm sorry. I'm late and I'm trying to get to class.
- French fry convention?

Human Participant Selection:



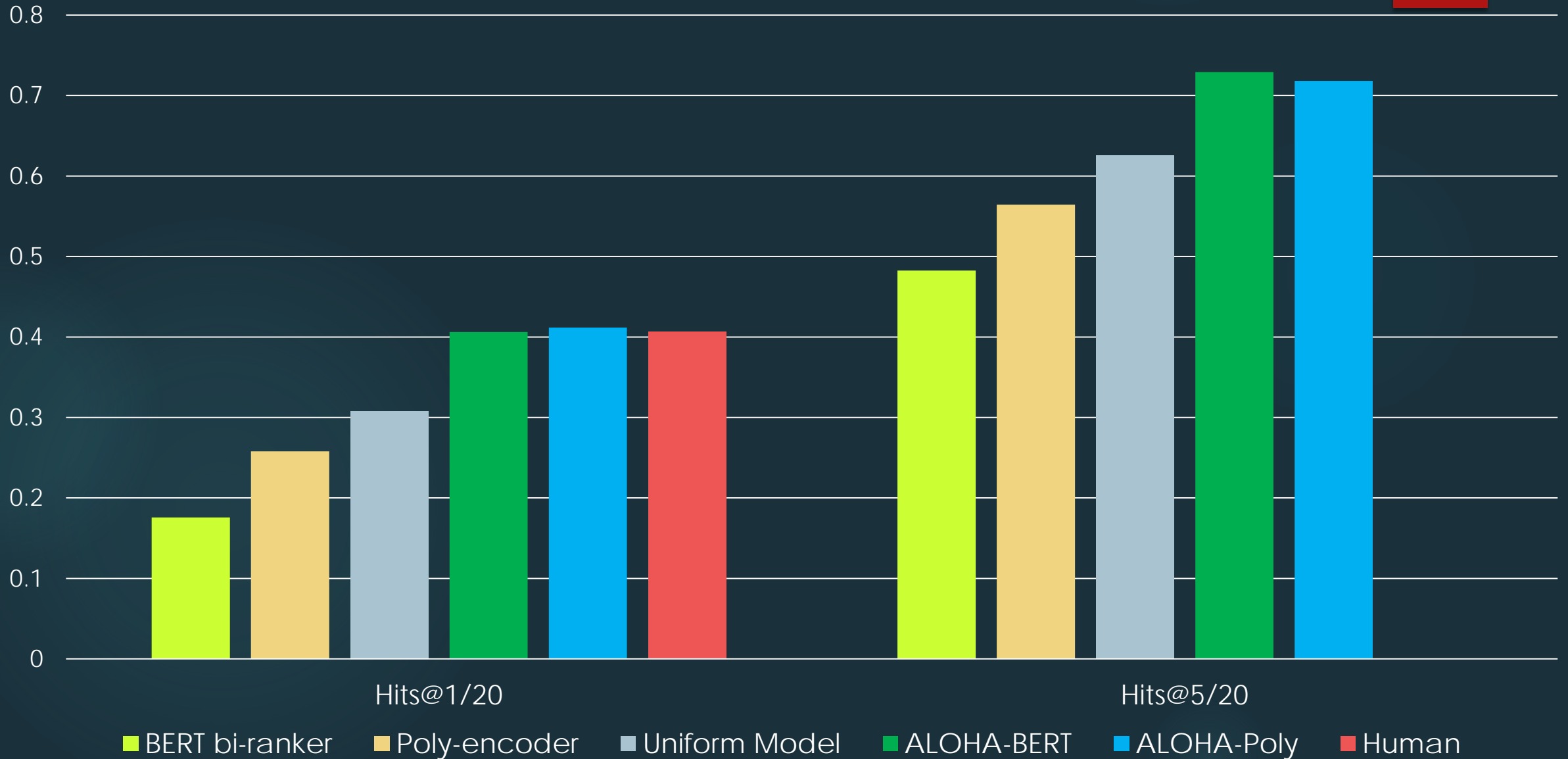
Ground Truth Response:





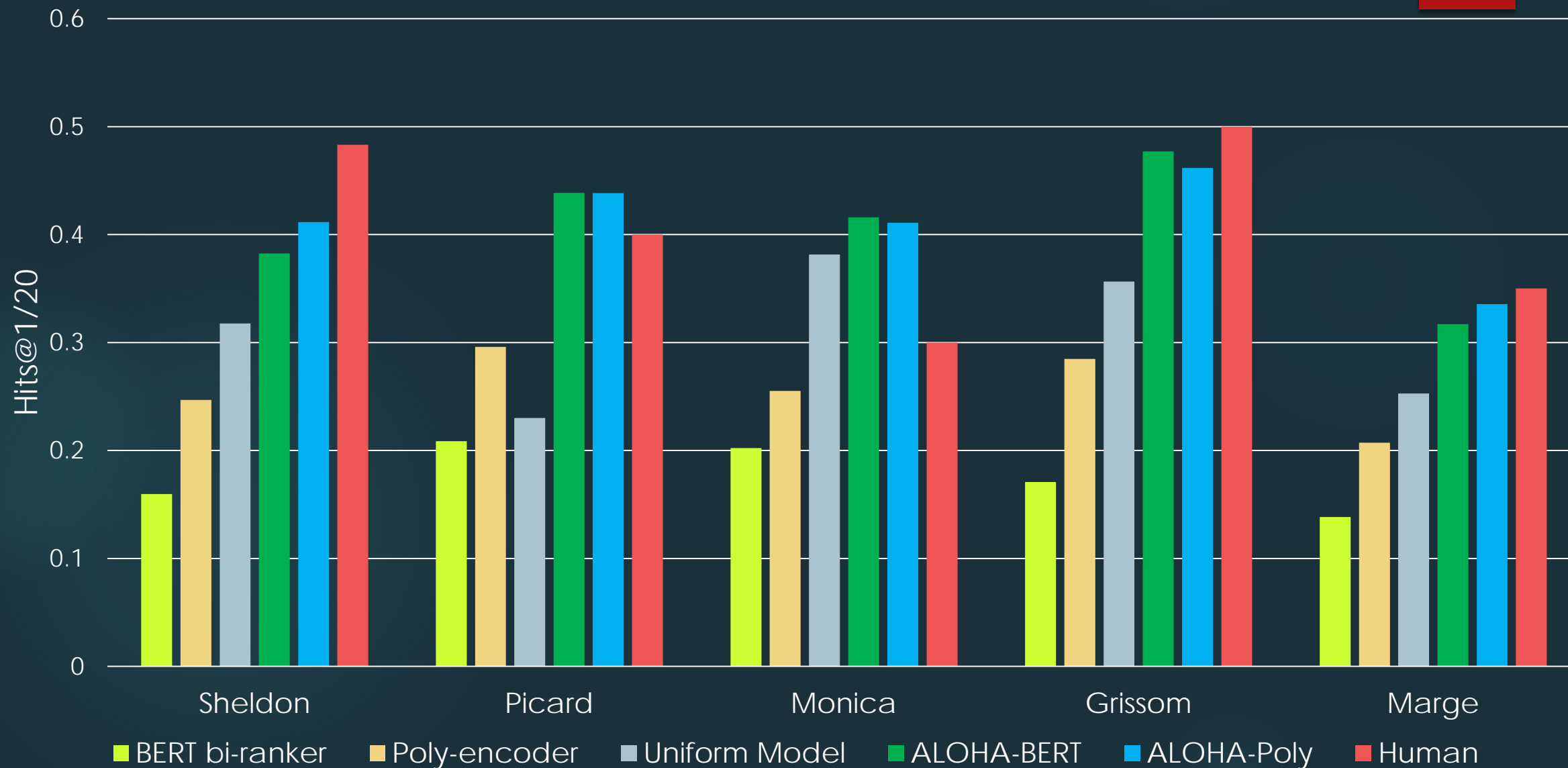
# Average Evaluation Results (Hits@1/20 and 5/20)

14



# Average Hits@1/20 by Evaluation Character

15



# Future Work

- ▶ Model conversations with a multi-turn response approach
- ▶ Model the dialogue counterpart
- ▶ HLA-aligned generative models
- ▶ Determining HLAs from text
- ▶ Larger and more extensive human evaluation

# Semantic Text Exchange (STE)

- ▶ Task: correct the semantic content of text (e.g. personalized dialogue agent responses) using the original text as a template
- ▶ Goal: correct semantics while preserving sentiment and fluency

- ▶ Example:

User input:	<i>What is the weather like today?</i>
Original output text (negative emotion):	<i>It is <b>sunny</b> outside. I know, it sucks! But you should <b>wear sunscreen</b> even if it's <b>sticky</b>.</i>
Replacement entity:	<i>Rainy</i>
Desired output text (negative emotion):	<i>It is <b>rainy</b> outside. I know, it sucks! But you should <b>bring an umbrella</b> even if it's <b>cumbersome</b>.</i>

- ▶ SMERTI: Similarity Masking, Entity Replacement, Text Infilling

# Summary

18

- ▶ **Major Challenge** - Giving chatbots and virtual assistants the ability to imitate human emotions and personality
- ▶ **Human-Level Attributes (HLAs)** - Train dialogue agents with tropes: aspects of fictional characters' identity and profile
  - ▶ HLAs accurately model human impressions of characters
- ▶ **HLA-Chat** - Dataset of characters with their HLAs and dialogue
- ▶ **Artificial Learning of Human Attributes (ALOHA)** - Three-stage system to retrieve character/personality specific language styles
  - ▶ ALOHA effectively uses HLAs to retrieve character specific language styles
  - ▶ ALOHA shown to be stable and robust



# Thanks for Listening!

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<https://github.com/newpro/aloha-chatbot>



<https://arxiv.org/abs/1910.08293>

19

**Input:** *What do you think of my AAI paper?*

**Sheldon (Reviewer 1):** *Very good. Author of one of the greatest horror stories ever written: Turn of the Screw. And I'm looking for one.*

**Picard (Reviewer 2):** *Class M in all respects. Quite Earth-like.*

**Marge (Reviewer 3):** *I think it's awesome. I love it.*

**Input:** *Let us go to a party!*

**Sheldon:** *No chance, hot pants! I'd rather die and kill all of you than party for one more millisecond!*

**Picard:** *Whoa, whoa, whoa! I enjoy partying as much as the next fellow, but this is Oktoberfest. Good day, sir!*

**Marge:** *I think you cannot have a party. Your mother left me in charge.*

# Appendices

# Mathematical Analysis

- ▶ Pearson correlation values:
  - ❖ Human vs. Uniform: 0.047
  - ❖ Human vs. ALOHA: 0.4149 and 0.5468
  
- ▶ Paired t-test p-value results (of Hits@1/20 scores):
  - ❖ ALOHA-BERT vs. BERT Bi-ranker: 0.0004
  - ❖ ALOHA-Poly vs. Poly-encoder: <0.0001
  - ❖ ALOHA-BERT vs. Uniform: 0.0329
  - ❖ ALOHA-Poly vs. Uniform: 0.0234