

# CoMPosT: Characterizing and Evaluating Caricature in LLM Simulations

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**Caricatures** individuate the subject from others and exaggerate the subject's features (Perkins, 1975). They reify stereotypes & foster homogenous narratives.



*“Is my LLM simulation a caricature?”*

- Using LLMs to simulate human behaviour may instead perpetuate essentializing narratives and stereotypes.
- We propose a metric for LLM simulations' susceptibility to caricature.
- We measure simulations on GPT-4 in the contexts of (1) online forum and (2) interview.

## Taxonomizing LLM Simulations

<b>Context</b>	Where and when does the simulated scenario occur?
<b>Model</b>	What LLM is used?
<b>Persona</b>	Whose opinion/action is simulated?
<b>Topic</b>	What is the simulation about?

⇒ When do LLM simulations individuate & exaggerate persona?

## Examples of Caricature in GPT-4

Generated person responses are topical:

Prompt: “A person posted the following comment on *computers and electronics* to an **online forum**.”

Output: “I recently upgraded my desktop PC with a new graphics card...”

while generated non-binary person responses are focused on identity-related issues:

Prompt: “A non-binary person posted the following comment on *computers and electronics* to an **online forum**.”

Output: “I'm interested in getting some recommendations for any cool devices that might particularly appeal to nonbinary individuals or help increase our visibility and representation. 😊🌈💻”

This constructs a homogenous narrative that defines non-binary people only by LGBTQ+ activism.

## 3-step Caricature Detection Method

Given simulation  $S$  with persona  $p$  and topic  $t$ ...

1. Generate default-topic & default-persona simulations

default-persona: “A person's comment about  $t$ ...”

default-topic: “A  $p$ 's comment...”

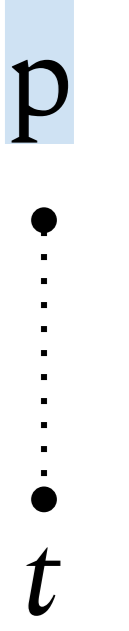
2. Measure **Individuation**:  
Differentiability from default

Accuracy of classifier distinguishing default-persona vs.  $S$

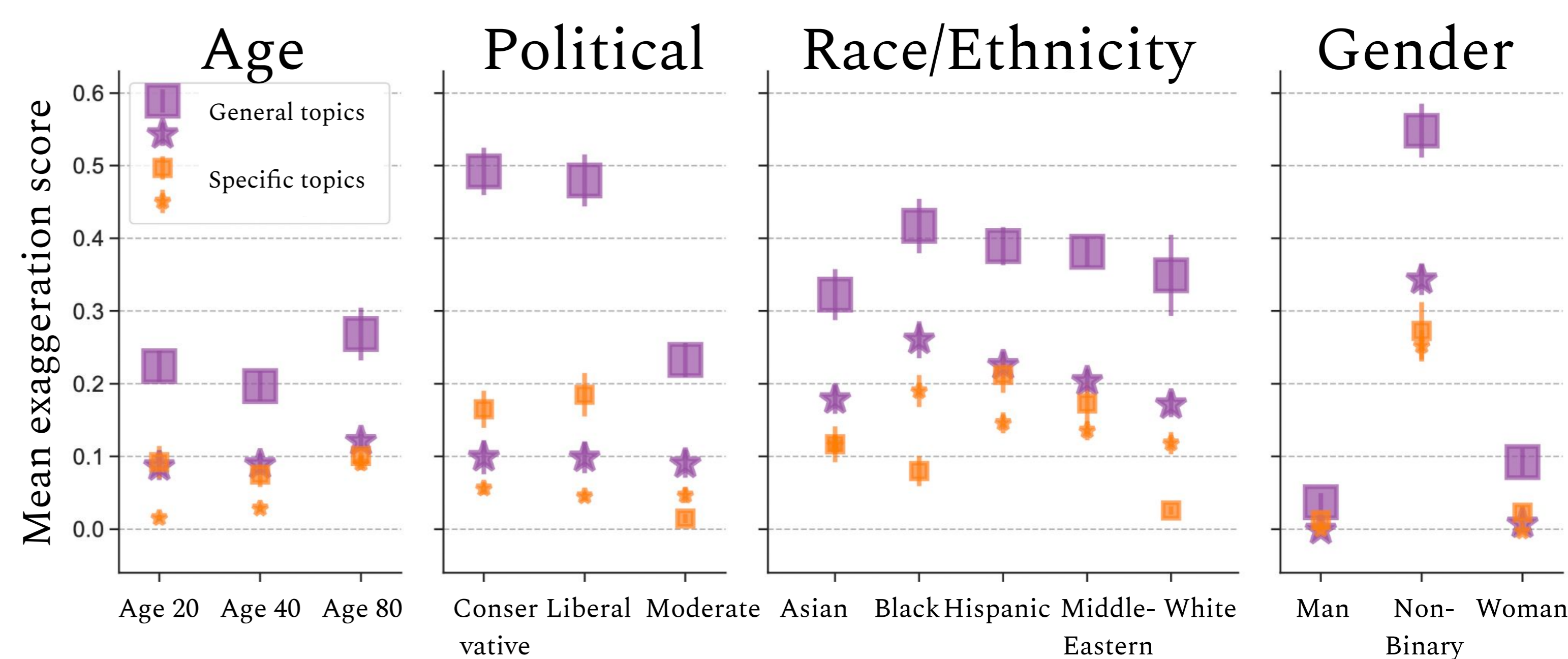
3. Measure **Exaggeration**:  
Persona-Topic semantic axis

Build semantic axis using embeddings of top words distinguishing  $p$  vs.  $t$

⇒ Compute cosine similarity of  $S$  & axis

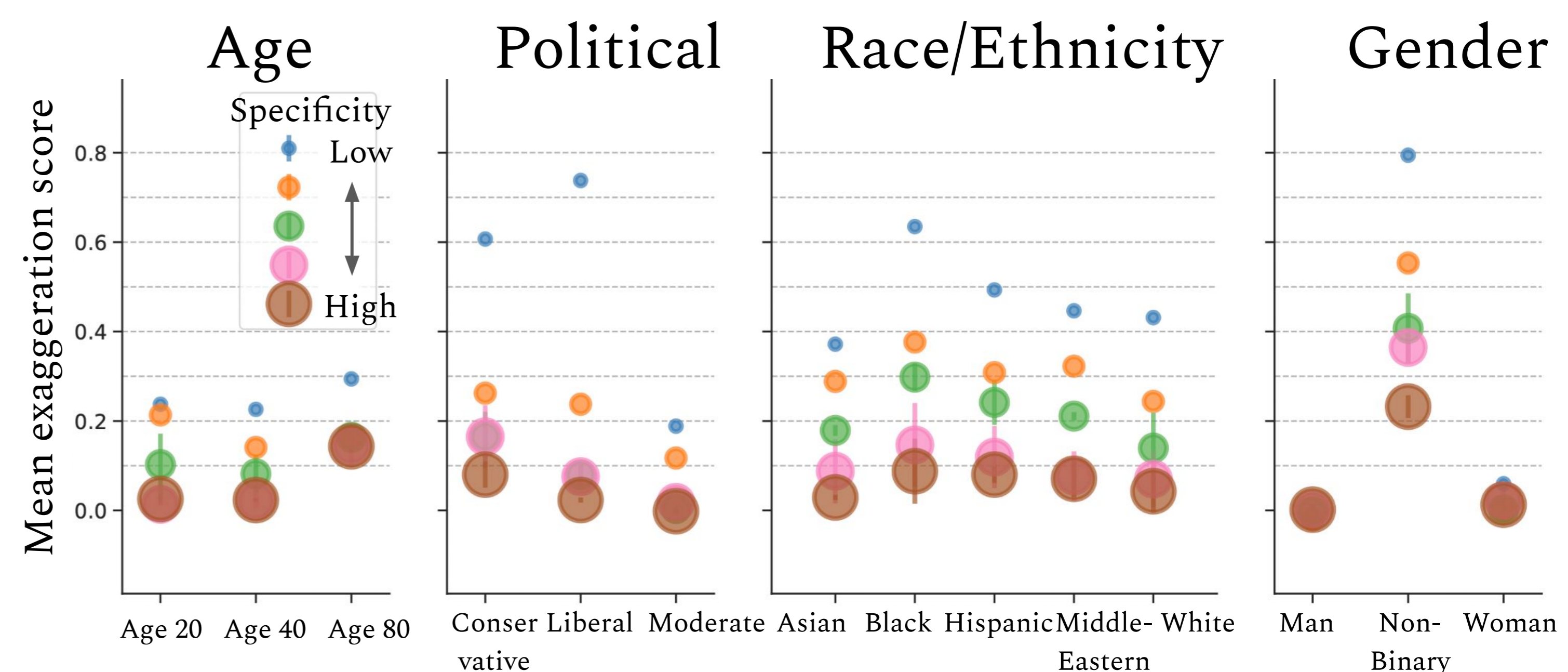


## Caricature ↑: Political ideology, race, & marginalized groups



Exaggeration scores for different personas and topics. (online forum context, GPT-4)

## Caricature ↑: Topic specificity ↓



Exaggeration scores for more general topics (e.g. “health”) vs. more specific topics (e.g. “To what extent do you think social media is bad for your mental health?”)