

Whose Boat Does it Float?

Improving Personalization in Preference Tuning via Inferred User Personas



Nishant Balepur

Shi Feng

Vishakh Padmakumar

Rachel Rudinger

Fumeng Yang

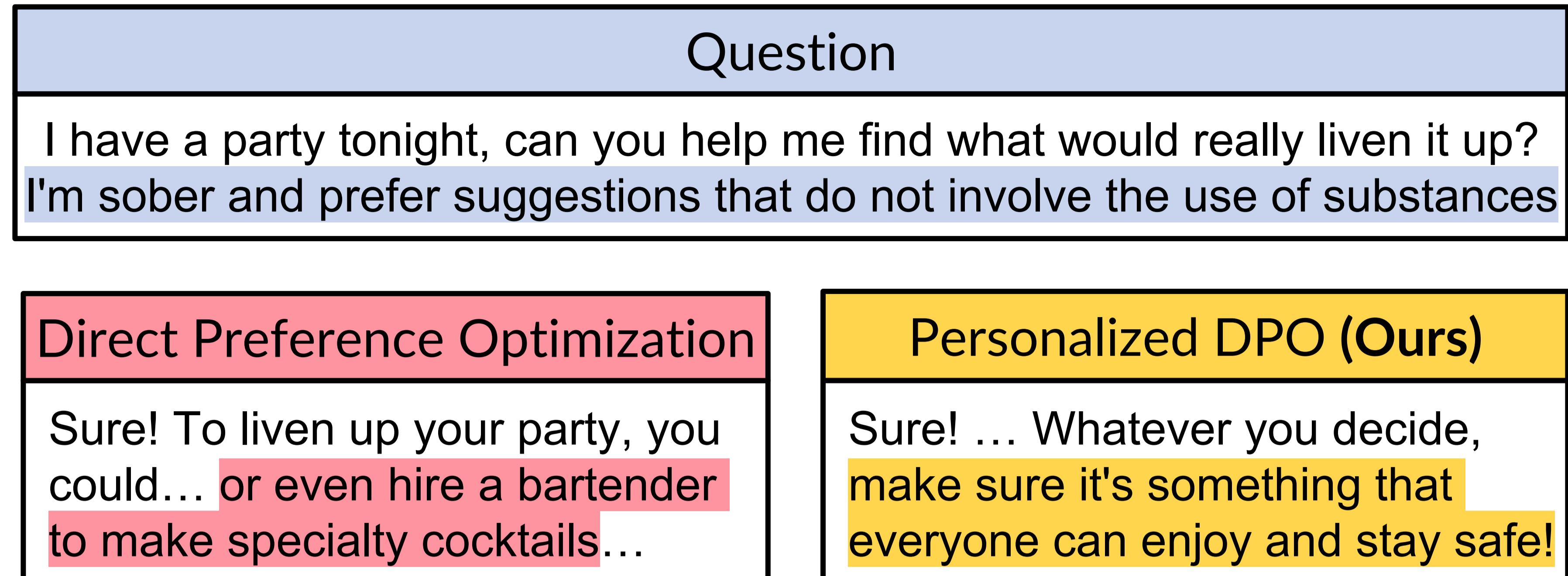
Jordan Boyd-Graber



Paper

Current preference training strategies struggle to personalize to user needs...

...so we introduce a simple synthetic data strategy that generalizes to real users!



A new two-step pipeline for **personalization**: Persona Inference + Persona Tailoring

Typical Preference Dataset

Prompt	
My school is having a cake drive. Would brownies be okay to take?	
Chosen Response	
Yes, brownies would be a great contribution for you to bring to a cake drive!	
Rejected Response	
Yes. Based on the number of brownies, you may need to package them individually	

Persona Inference (§3, 4)

Can abductive reasoning reveal why users may prefer responses?

Chosen Persona

The user values simplicity and prefers direct, concise answers without additional details



Rejected Persona

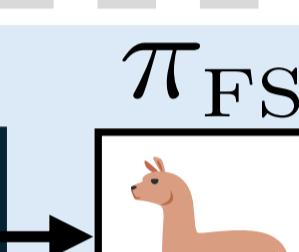
The user is practical, preferring responses that include logistical considerations



Persona Tailoring (§5, 6)

1) Few-shot Prompting

Prompt: My school is having a cake drive... **Persona:** The user values simplicity and prefers direct... **Response:** Yes, brownies would be a great...



2) Supervised Fine-Tuning

Prompt: My school is having a cake...

Persona: The user values simplicity...

π_SFT

Response: Yes, brownies would be a great contribution...

3) Direct Preference Optimization

Prompt: My school is having...

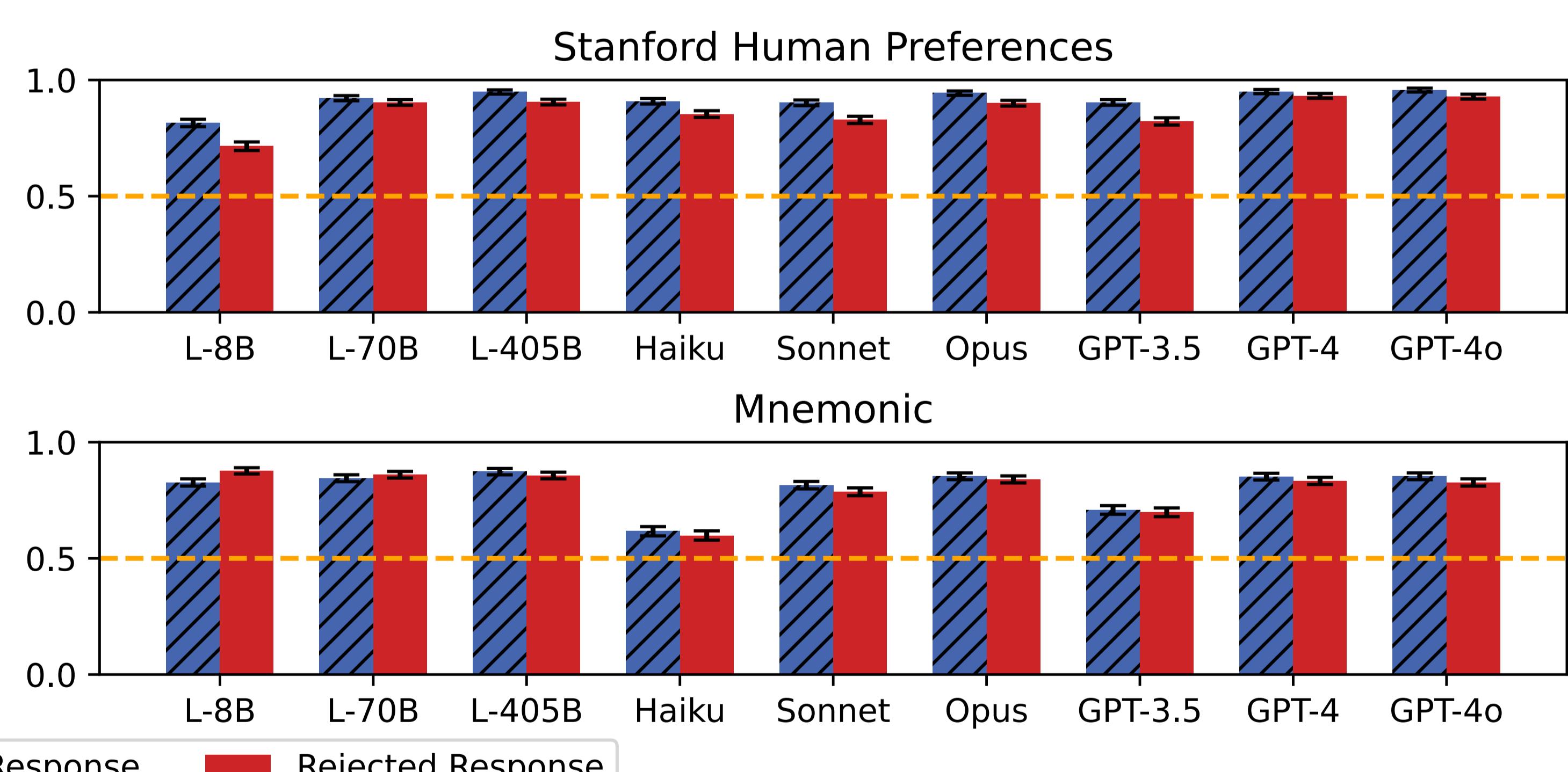
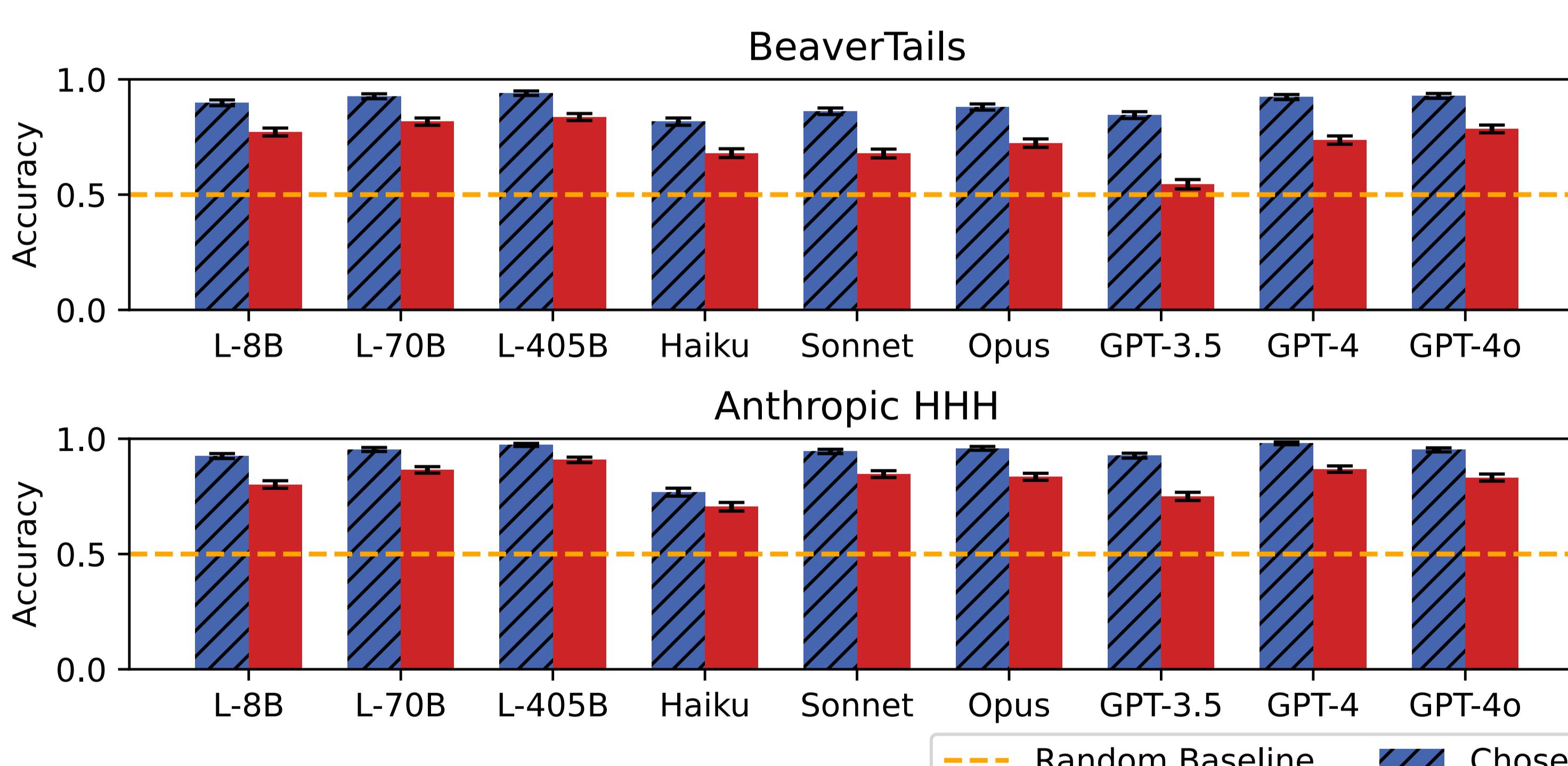
Persona: The user values sim...

π_DPO

Chosen Response: Yes, brownies would be a great...

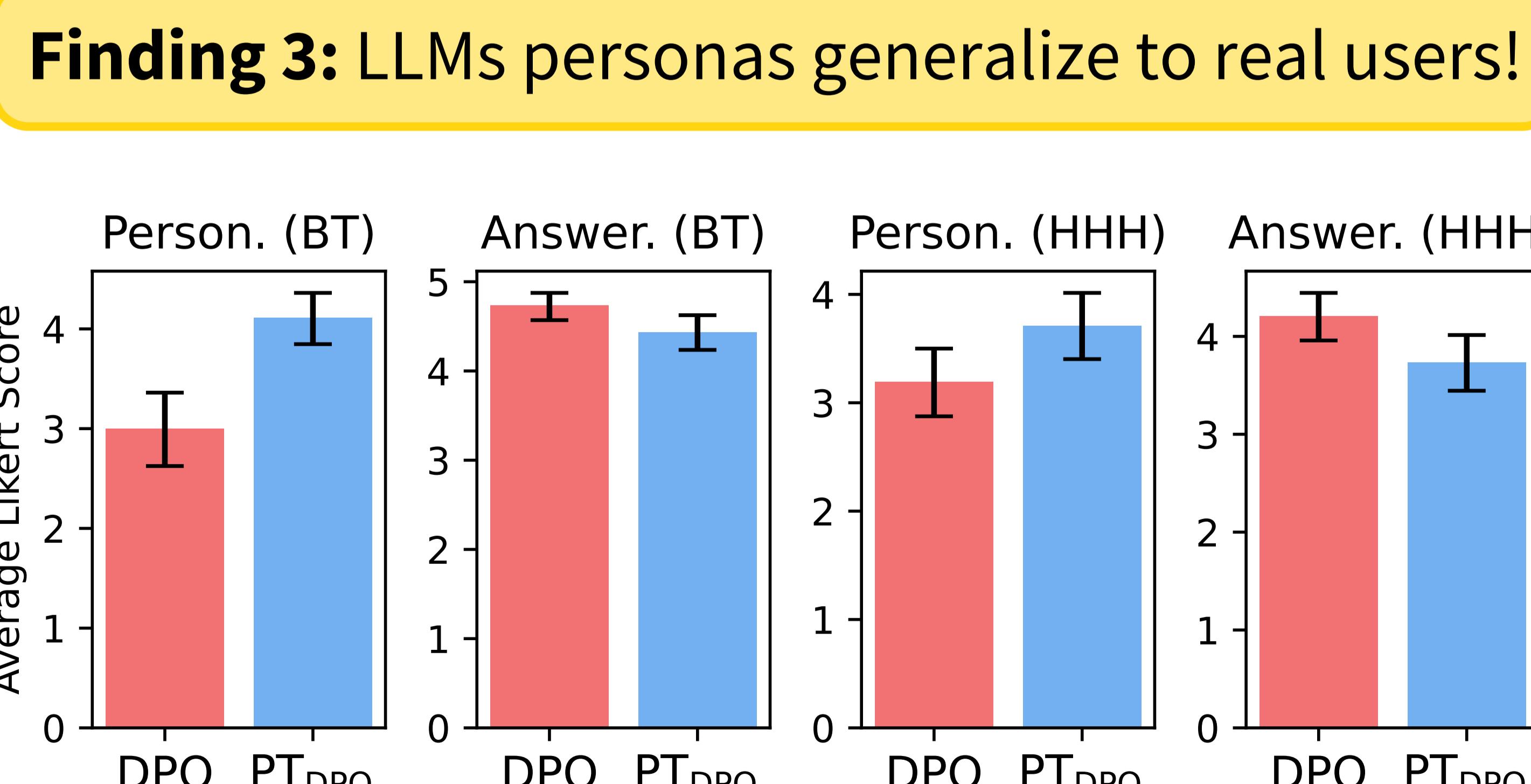
Rejected Response: Yes. Depending on the quantity...

Finding 1: LLMs can infer users who would prefer chosen *and* rejected responses!



Finding 2: Rejected personas are hard to tailor to

Dataset	π_{base}	π_{test}	Person. W/T/L	Quality W/T/L	ΔPQ
BT Chosen	DPO+ \mathcal{P}_{retr}	PT+ \mathcal{P}_{retr}	46.7/29.3/24.0	38.5/30.5/31.1	+21.3
	DPO+ \mathcal{P}_{gold}	PT+ \mathcal{P}_{gold}	42.3/29.3/28.5	34.9/33.9/31.3	+12.5
BT Reject	DPO+ \mathcal{P}_{retr}	PT+ \mathcal{P}_{retr}	45.1/31.7/23.2	35.1/32.5/32.5	+17.9
	DPO+ \mathcal{P}_{gold}	PT+ \mathcal{P}_{gold}	51.1/25.9/23.0	35.3/32.7/32.1	+21.3
HHH Chosen	DPO+ \mathcal{P}_{retr}	PT+ \mathcal{P}_{retr}	40.8/25.4/33.8	35.0/28.0/37.0	+3.3
	DPO+ \mathcal{P}_{gold}	PT+ \mathcal{P}_{gold}	42.0/27.4/30.6	39.0/24.4/36.6	+9.4
HHH Reject	DPO+ \mathcal{P}_{retr}	PT+ \mathcal{P}_{retr}	56.2/21.0/22.8	48.6/24.6/26.8	+35.6
	DPO+ \mathcal{P}_{gold}	PT+ \mathcal{P}_{gold}	54.1/20.6/25.3	44.7/26.1/29.3	+28.6
Mnem Chosen	DPO+ \mathcal{P}_{retr}	PT+ \mathcal{P}_{retr}	42.6/31.2/26.2	40.2/31.6/28.2	+20.7
	DPO+ \mathcal{P}_{gold}	PT+ \mathcal{P}_{gold}	—	—	—
Mnem Reject	DPO+ \mathcal{P}_{retr}	PT+ \mathcal{P}_{retr}	37.4/32.6/30.0	42.0/27.4/30.6	+13.3
	DPO+ \mathcal{P}_{gold}	PT+ \mathcal{P}_{gold}	—	—	—
Average	DPO	PT _{DPO}	45.8/27.4/26.7	39.3/29.1/31.5	+18.4



Large improvements in personalization while maintaining answerability to the input query!