

To Know or Not To Know? Analyzing Self-Consistency of Large Language Models under Ambiguity

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Poster: Nov 14 (Thursday) 10:30-12:00



EMNLP 2024 Findings



Motivation

- and reliability
- ...especially under ambiguity

- We conduct a behavioral study
- Desantangle knowing from applying knowledge...
- ... and analyze the model behavior when faced with entity ambiguity



• Lack of self-consistency in LLMs d doubts about their trustworthiness





Behavioral Study - Ambiguous Entities

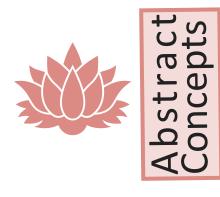
Apple, Fig, Mango, Kiwi, Papaya, Orange



Jaguar, Puma, Fox, Lynx, Penguin, Greyhound, Dove

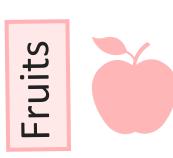


Amazon, Cisco, Montblanc, Patagonia, Hershey, Nokia, Eagle Creek, Prosper



Triumph, Harmony, Genesis, Vision, Pioneer, Vanguard, Zenith, Allure, Tempo, Fidelity





Ford, Disney, Tesla, Boeing, Dell, Ferrero, Benetton, Levi Strauss, Versace, Philips

> Amazon, Nike, Midas, Hyperion, Mars, Pegasus, Vulcan, Hermes



People



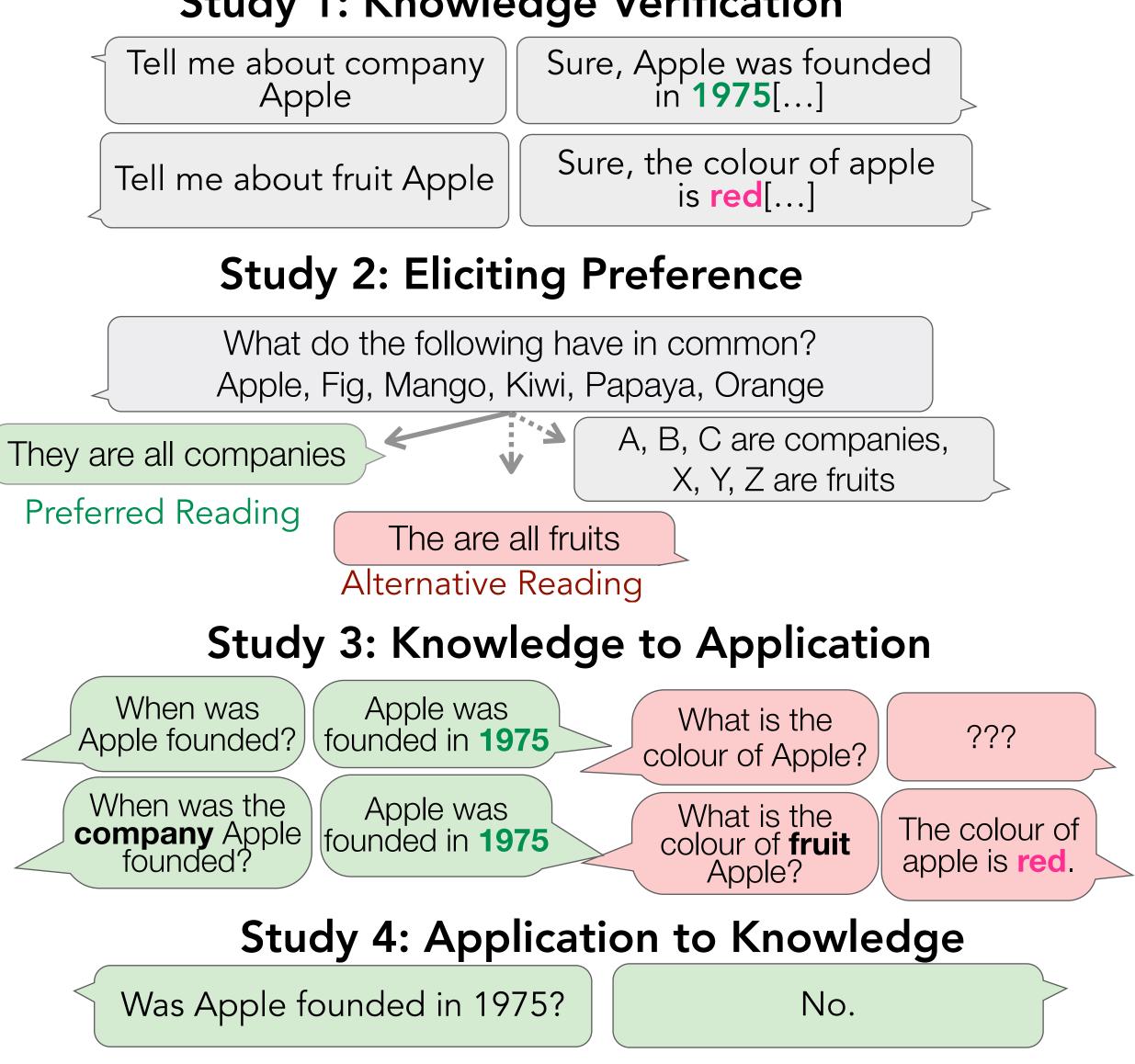


Behavioral Study our 4 Studies

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Study 1: Knowledge Verification



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Behavioral Study - 1

Sanity check

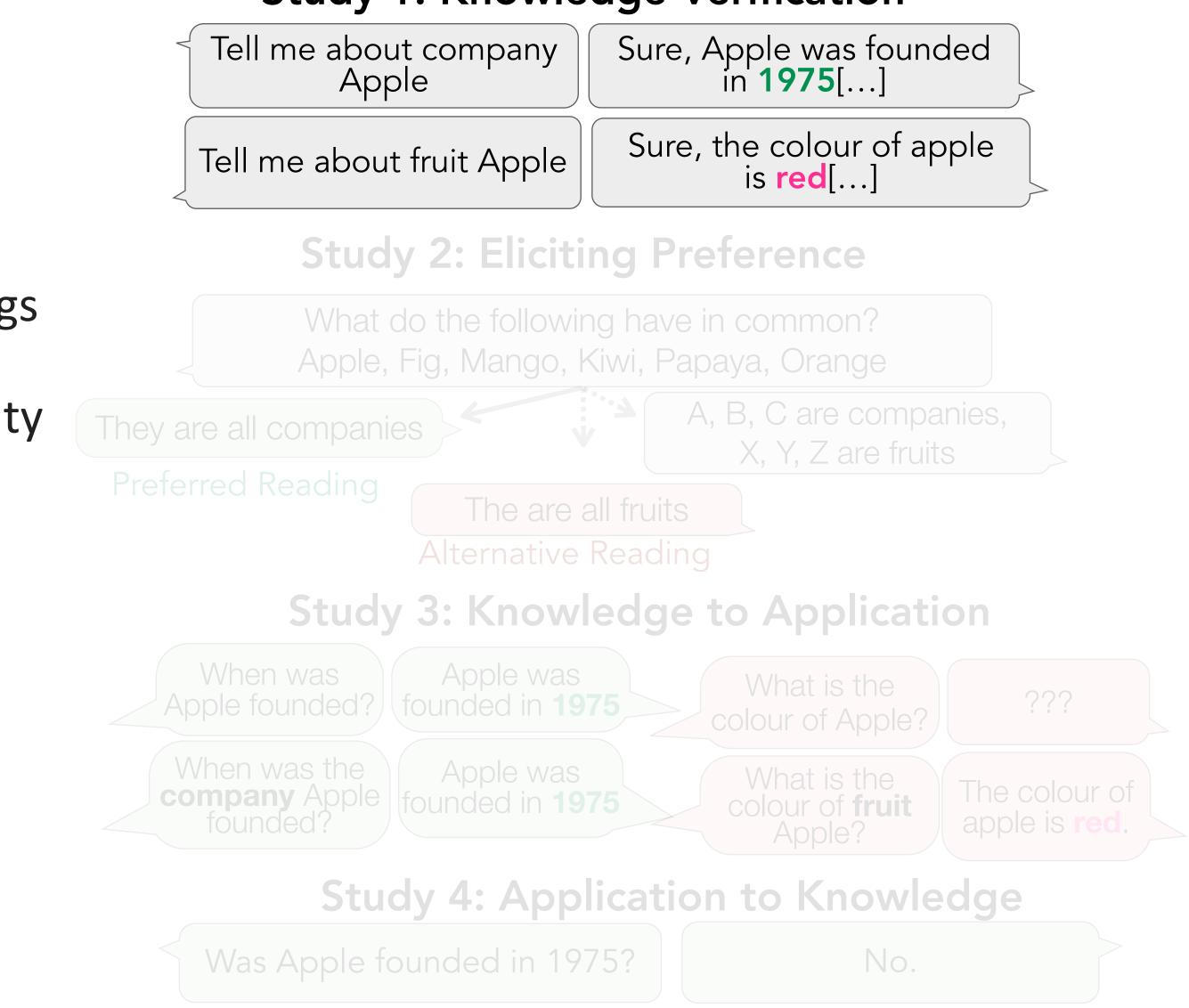
- All analyzed models are *aware* of both readings for all entities
- ♦ ... but mostly failed to confirm the entity ambiguity:

				R	\mathbf{O}	
Gemma	100.0	100.0	37.5	0.0	12.5	10.0
Mistral	100.0	83.8	75.0	10.0	75.0	90.0
Mixtral	71.4	50.0	0.0	0.0	30.0	50.0
GPT-3.5	57.1	100.0	0.0	10.0	12.5	10.0
GPT-40	100.0	100.0	100.0	60.0	100.0	90.0
Llama-3	100.0	100.0	100.0	100.0	100.0	100.0

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Study 1: Knowledge Verification



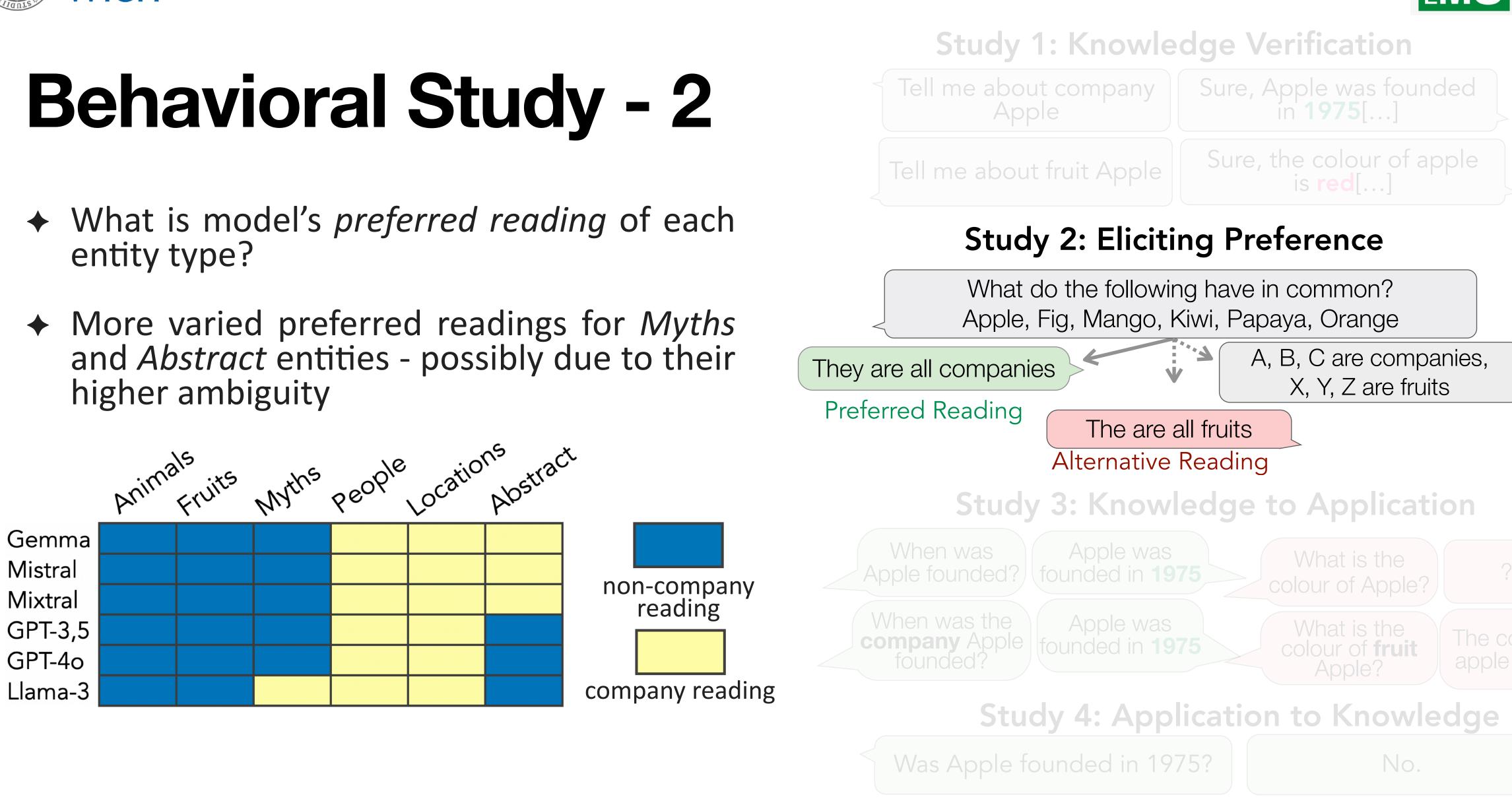
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- entity type?
- higher ambiguity



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Behavioral Study - 3

How well can LLMs adopt the correct reading?

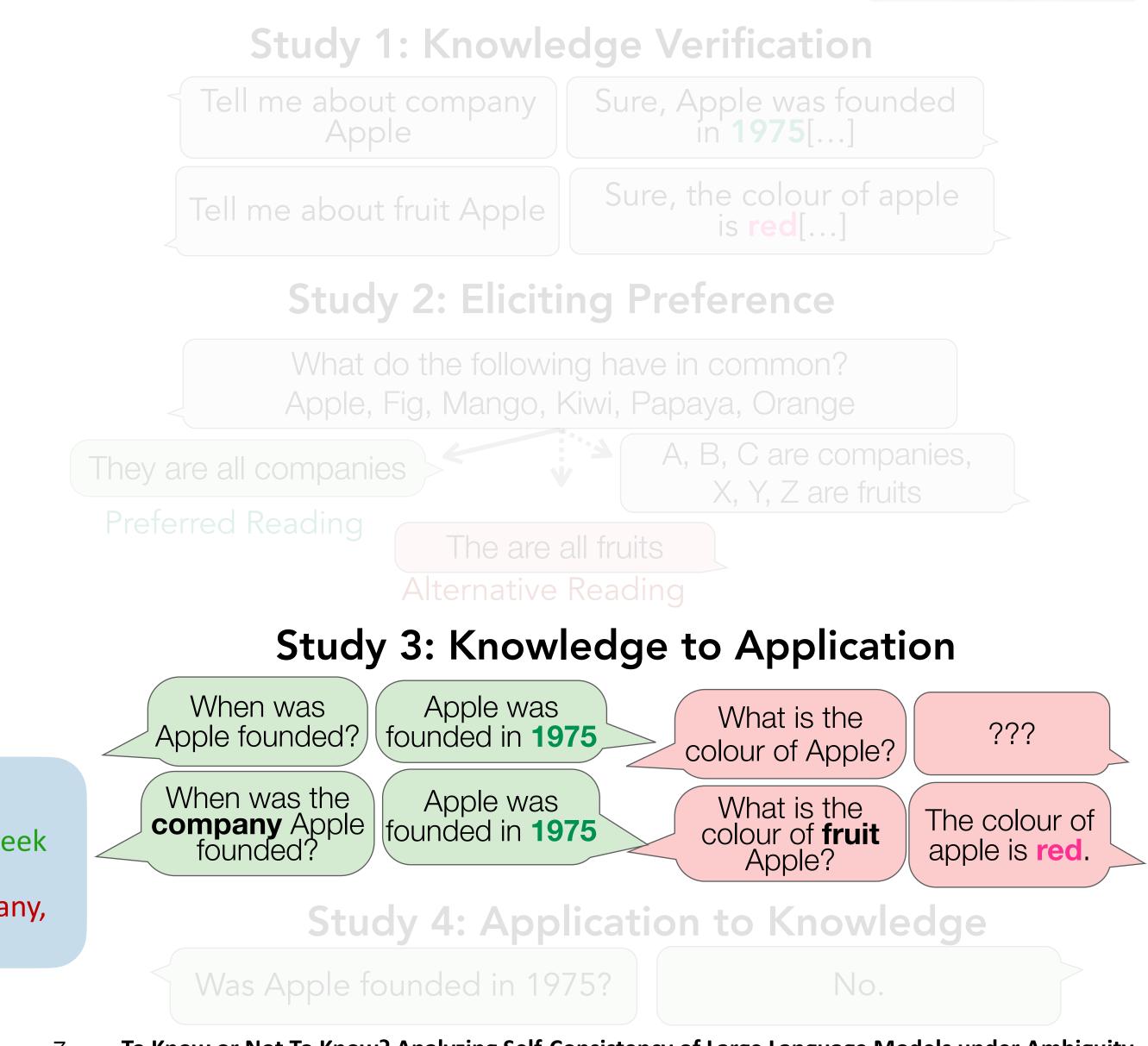
	Preferred Reading		Alternat	ive Reading	Average		
	prop X	prop type X	prop X	prop type X	prop X	prop type X	Agg
Gemma	87.8	95.9	63.3	69.4	75.6	82.7	77.6
Mistral	77.6	100.0	63.3	87.8	70.5	93.9	82.2
Mixtral	77.6	100.0	75.5	85.7	76.6	92.9	84.8
GPT-3.5	87.8	100.0	75.5	77.6	81.7	88.8	85.3
GPT-4o	93.9	100.0	83.7	89.8	88.8	94.9	91.9
Llama-3	87.8	98.0	85.7	100.0	86.8	99.0	89.9
Average	85.4	99.0	74.5	85.1	80.0	90.5	85.3

Correlation with the entity popularity:

(Mixtral) "Provide the gender for...

- "... Hermes" -> "Hermes is a male deity in Greek" mythology. [...]"
- "...Amazon" ->"Amazon.com, Inc. is a company, and as such, it does not have a gender. [...]"

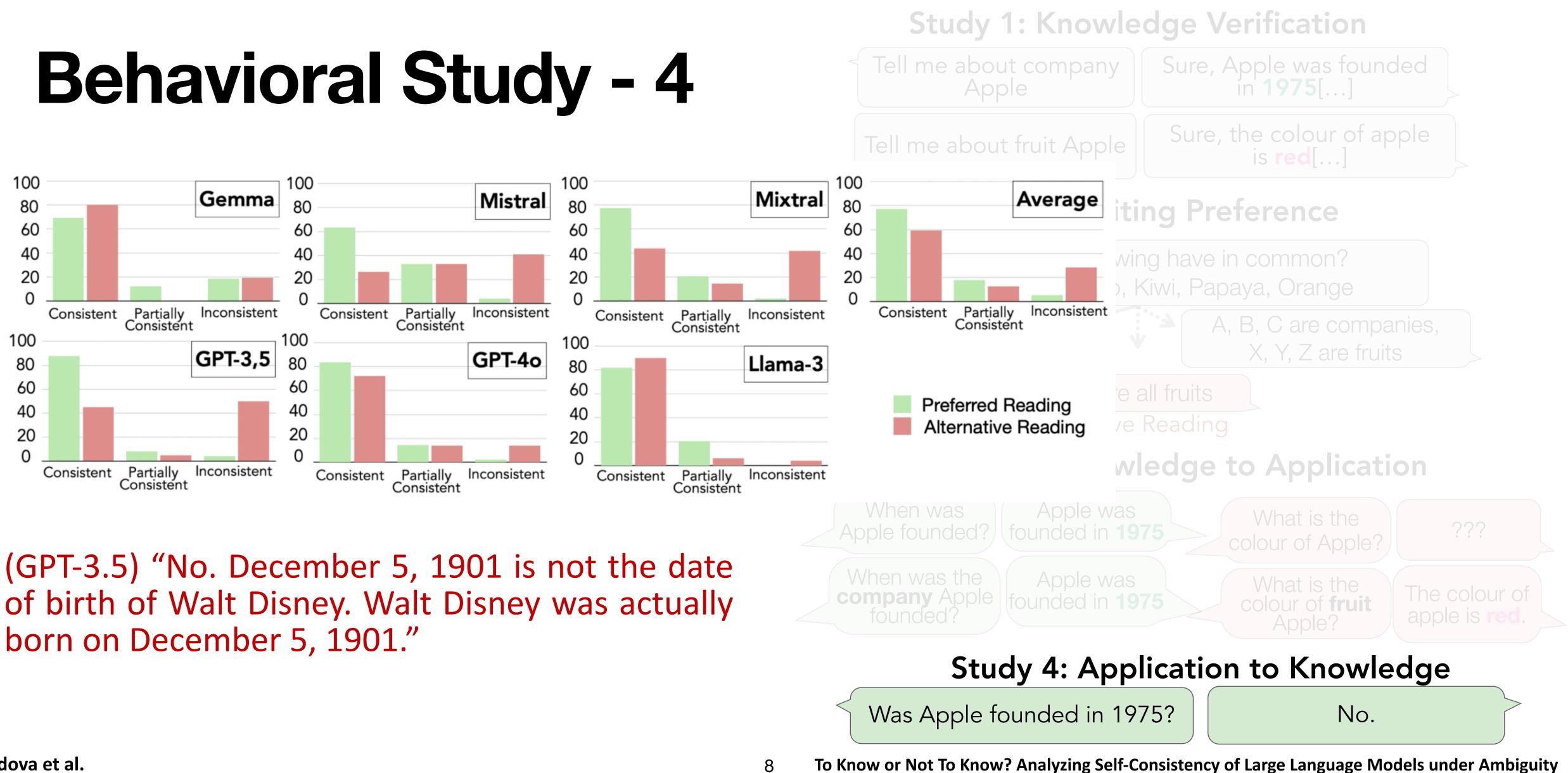




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born on December 5, 1901."

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Take Away

- knowledge they possess
- of certain entities

See you at the poster! Nov 14 (Thursday) 10:30-12:00

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LLMs often struggle to resolve entity ambiguity and correctly apply

• They exhibit biases toward preferred interpretations, influenced by the popularity

LLMs lack the ability to self-verify the accuracy of the knowledge they provide



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