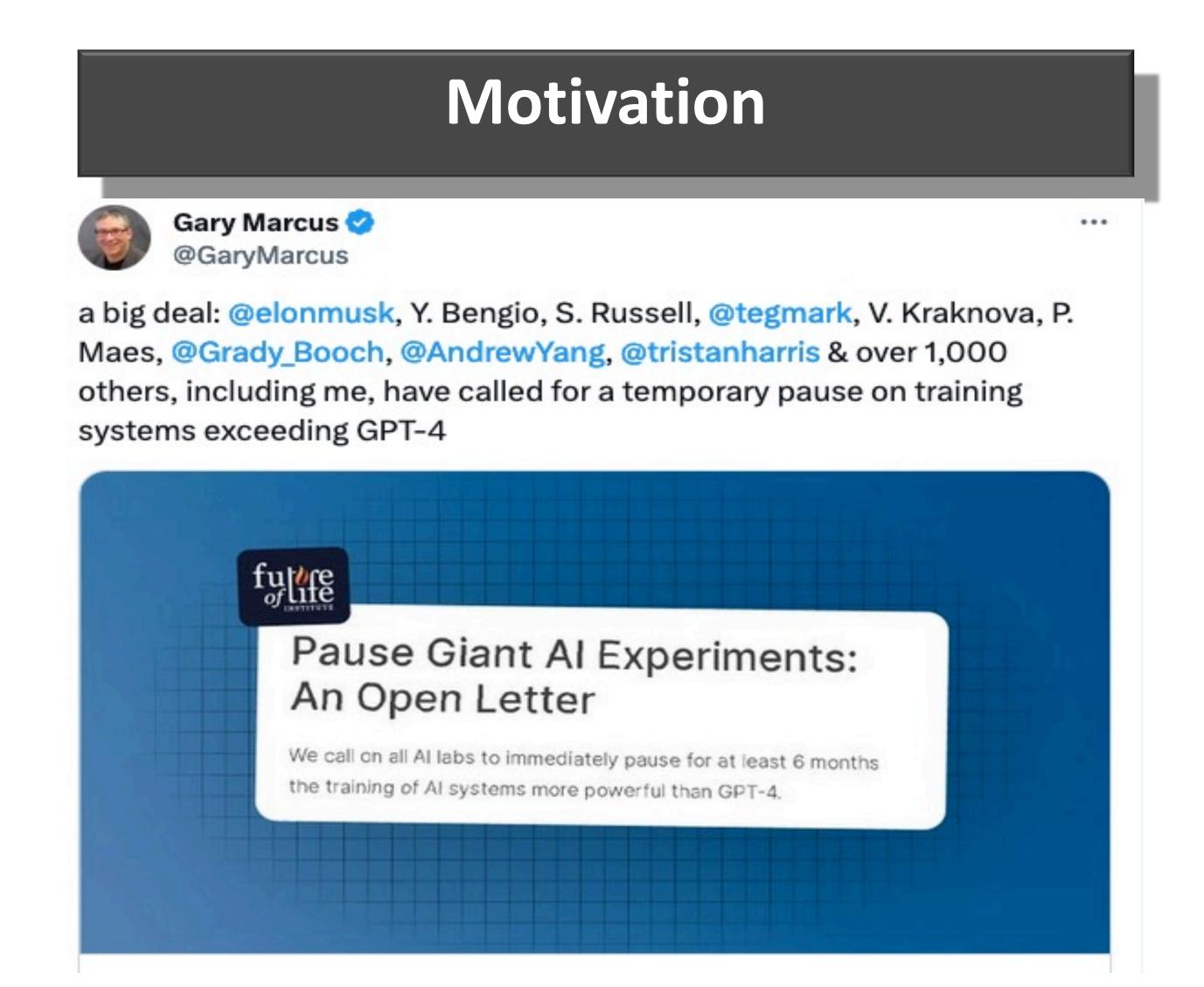
## Excitements and Concerns in the Post-ChatGPT Era: Deciphering Public Perception of AI through Social Media Analysis

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As AI continues to advance and integrate into society, it's essential to gain insights into how public perceive it. We aims to study public perception towards AI, particularly for generative AI.

- □ RQ1: What specific topics characterize the discussion of AI on Reddit? How do topics vary across subreddits?
- □ RQ2: What is the prevailing sentiment/opinions surrounding the most discussed topics, and do these sentiments differ among subreddits?

## Methods

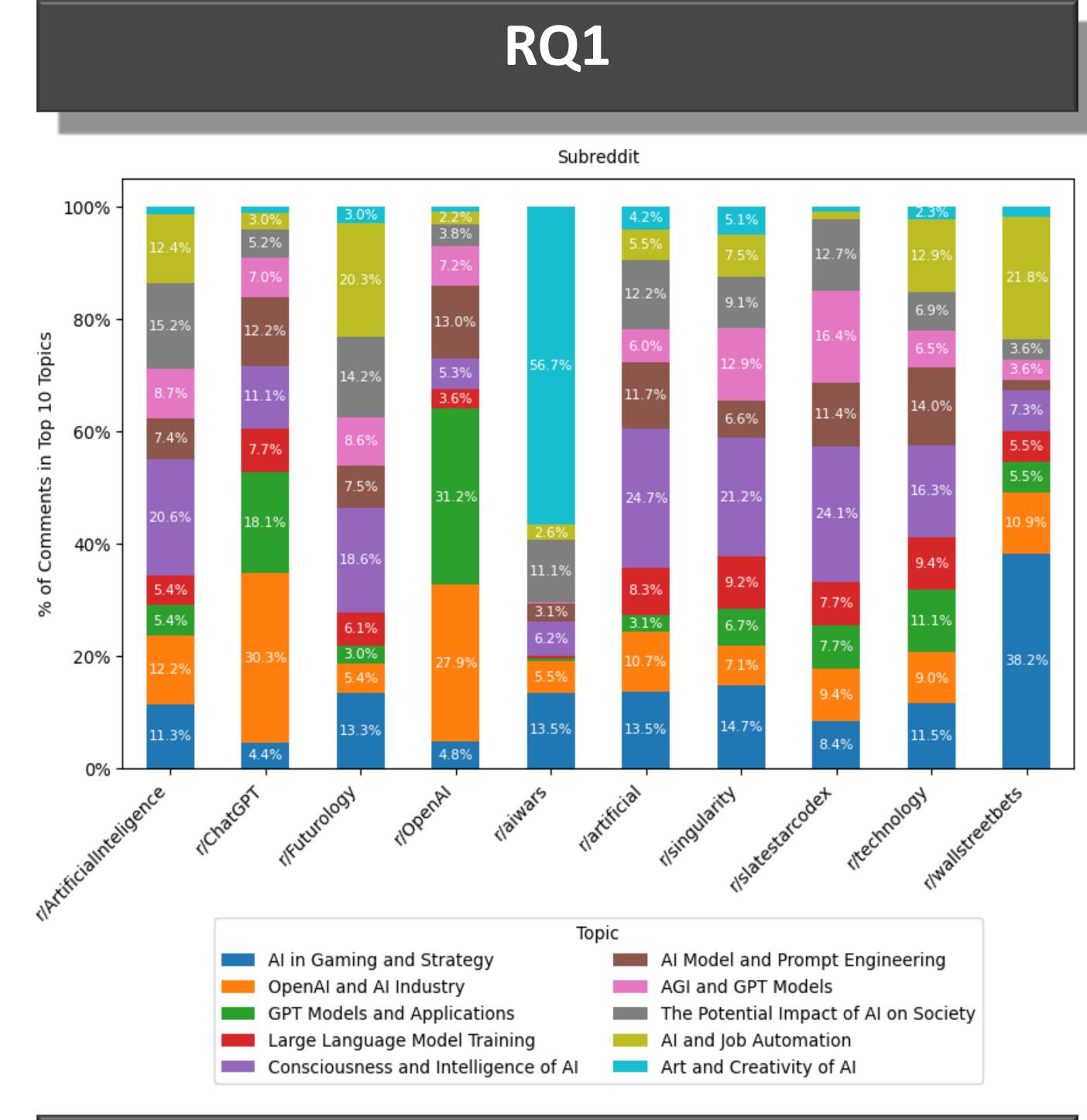
**Data Collection**: We employ Reddit PRAW API with curated keywords to extract relevant subreddits. Then we retrieve the latest Reddit comments..

**Topic Modeling:** BERTopic [1] is used. For model architecture, we select all-MiniLM-L6-v2 as embedding component and DBSCAN as clustering component.

**Sentiment classification:** We leverage GPT-3.5-turbo [2] in zero-shot prompting, assigning labels (Positive, Negative, Neutral) to each Reddit Comment.

**Regression:** To explore whether perceptions of AI differ between tech-centric and non-tech groups, we run LIWC [3] and then regress emotion and tone on if comment belongs to tech group.

$$LIWC_{Attr} = \alpha_0 + \alpha_1 \cdot Tech + \epsilon$$

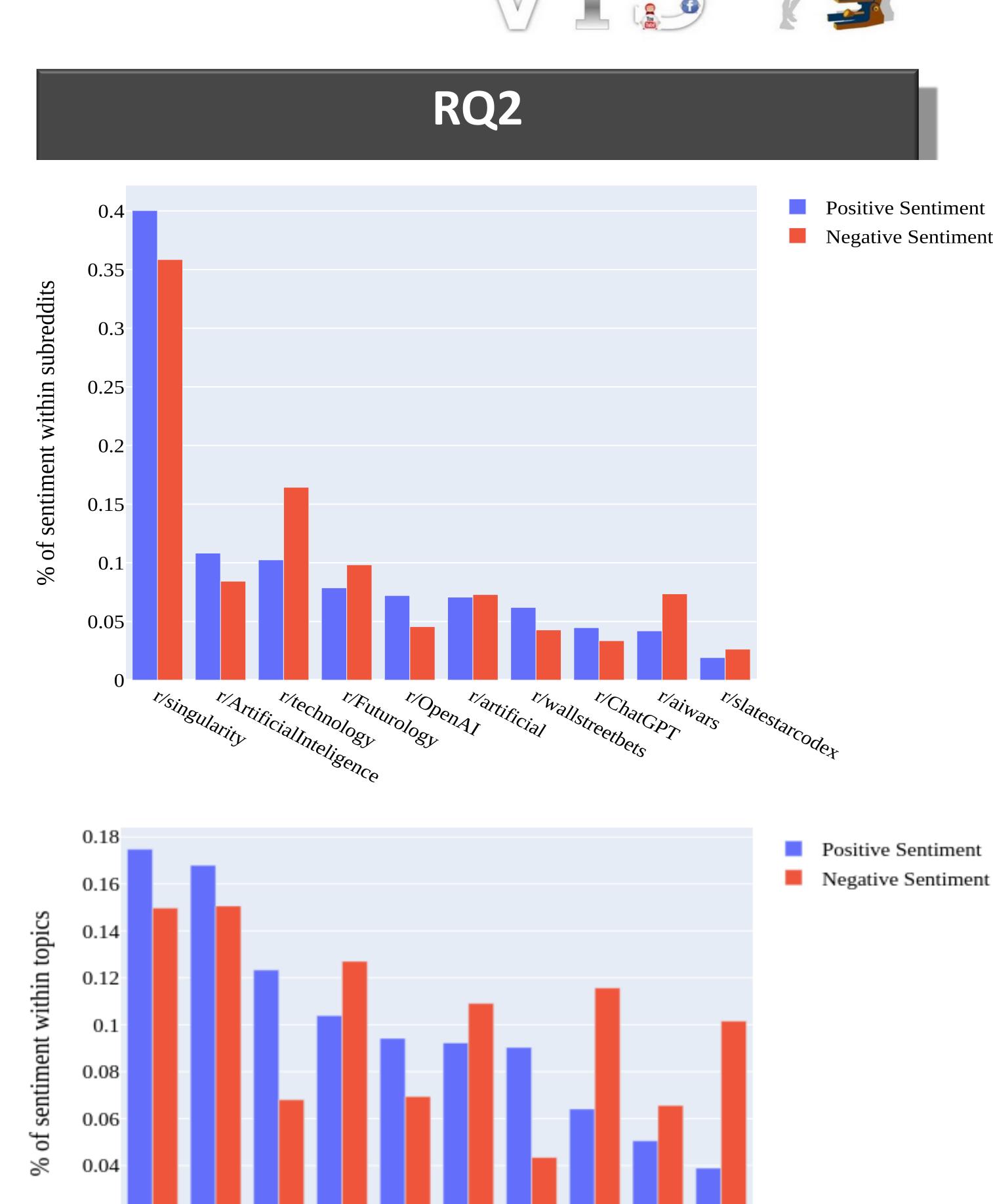


## References

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	(1)	(2)	(3)	(4)	(5)	(6)
	Positive tone	Negative tone	Positive emotion	Negative emotion	Prosocial	Conflict
Tech	0.173**	-0.0160	0.0562*	0.102***	0.0632*	0.0153
	(0.0630)	(0.0519)	(0.0282)	(0.0297)	(0.0287)	(0.0205)
Reference (	Group: Non-tech subredo	lits				
N	18764	18764	18764	18764	18764	18764

Note: This table presents the estimation coefficients of the regressions. Standard errors of each coefficient are in parentheses. The p-values indicating significance at the 90%, 95%, and 99% confidence levels have been adjusted using the Bonferroni correction.

\* p < 0.1, \*\* p < 0.05, \*\*\* p < 0.01