



# Conversational Linguistic Features That Inform Social Network Learning

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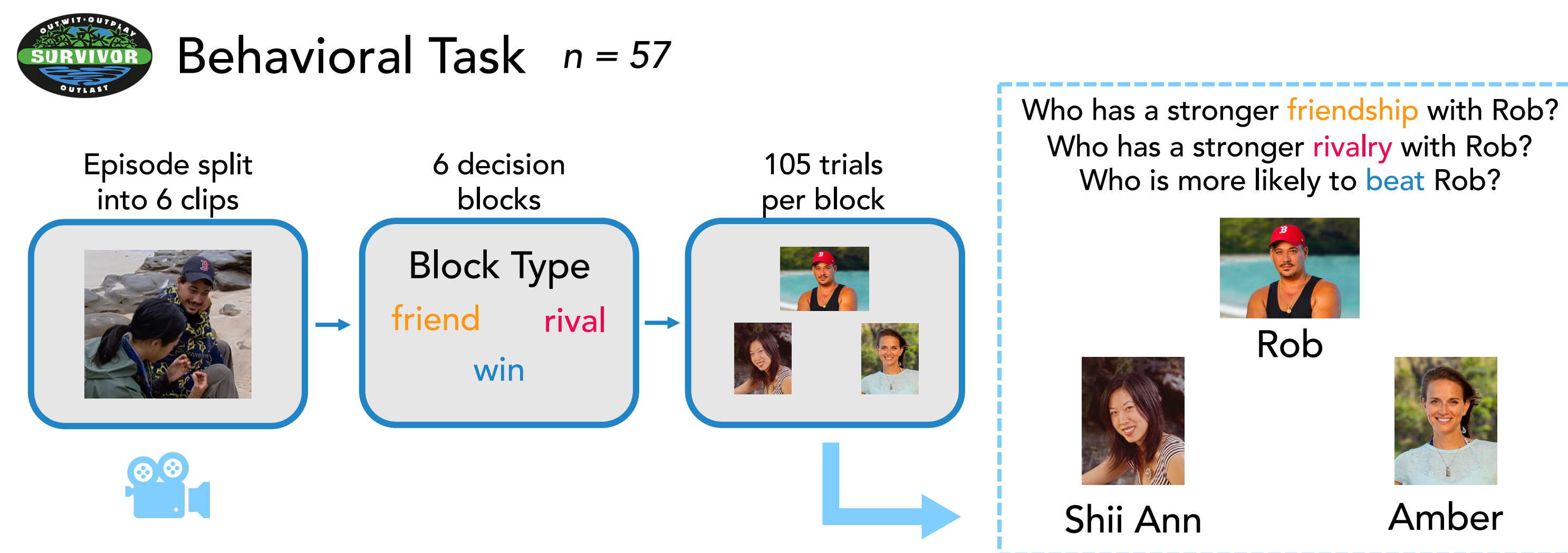
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## Introduction

- We know that social networks exist<sup>1</sup>, but we don't fully understand how people learn about these networks. People use social features like club membership and shared hobbies<sup>2</sup>, as well as traits<sup>3</sup>, to make inferences about a social network, but it is still unclear how communication between members contributes to passive learning about these social networks.
- Dialogue contains a wealth of information that can reveal the social structure between speakers. We can leverage tools like **natural language processing**<sup>4,5,6,7</sup> to quantitatively represent this information and show how people differentially use linguistic elements to make inferences and decisions about social networks.

## Method



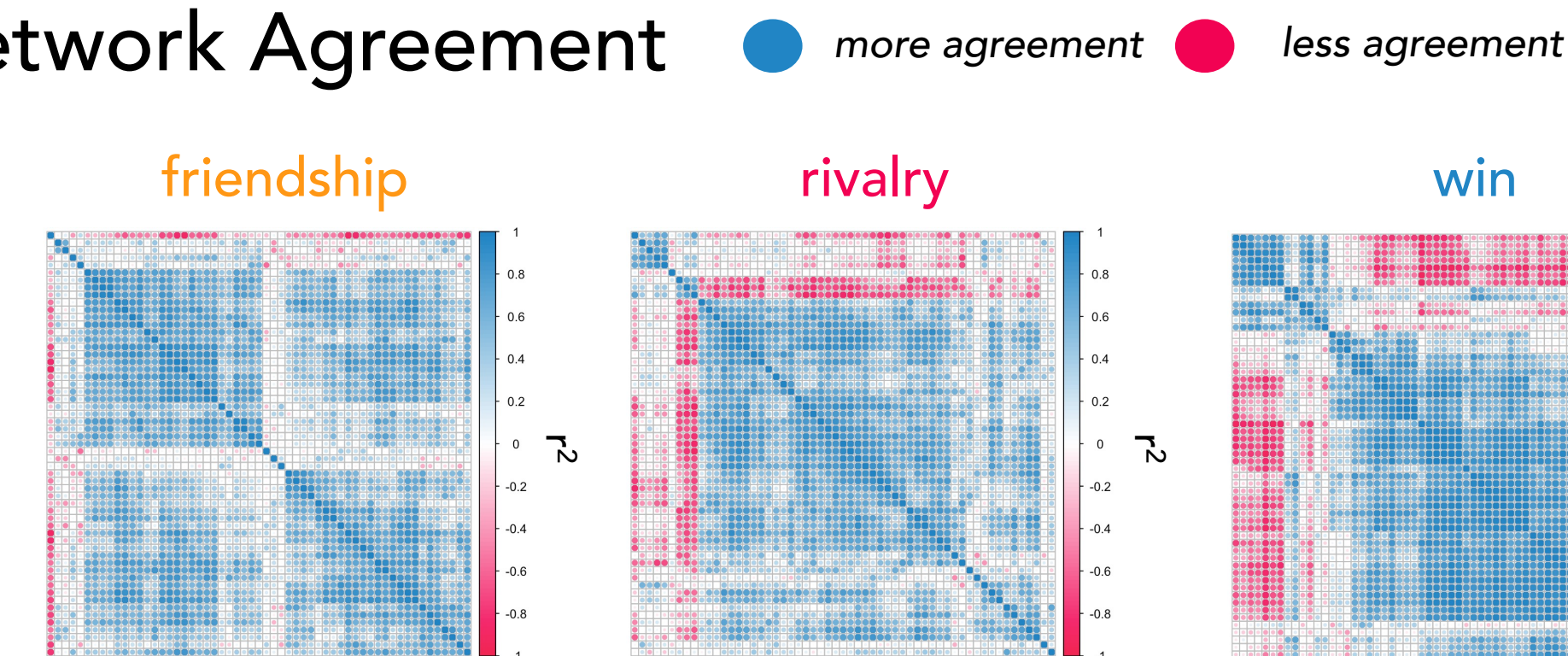
## Natural Language Processing on Transcribed Dialogue

	Speaker	Recipient	Dialogue	Similarity	Sentiment	Confidence
<b>similarity</b>	Shii Ann	Rob	If you make it to the final two, I'd be shocked.	0.1607	-0.1508	50
	Rob	Shii Ann	I know.	0.1607	0	1
<b>sentiment</b>	Shii Ann	Rob	With your strategy, current strategy.	0.1607	0	99
	Shii Ann	Rob	And I hope you do well, but, you know.	0.1607	0.4333	98.62
<b>confidence</b>	Shii Ann	Rob	I'm wishing myself luck in the next immunity challenge.	0.1607	0.4333	1.38

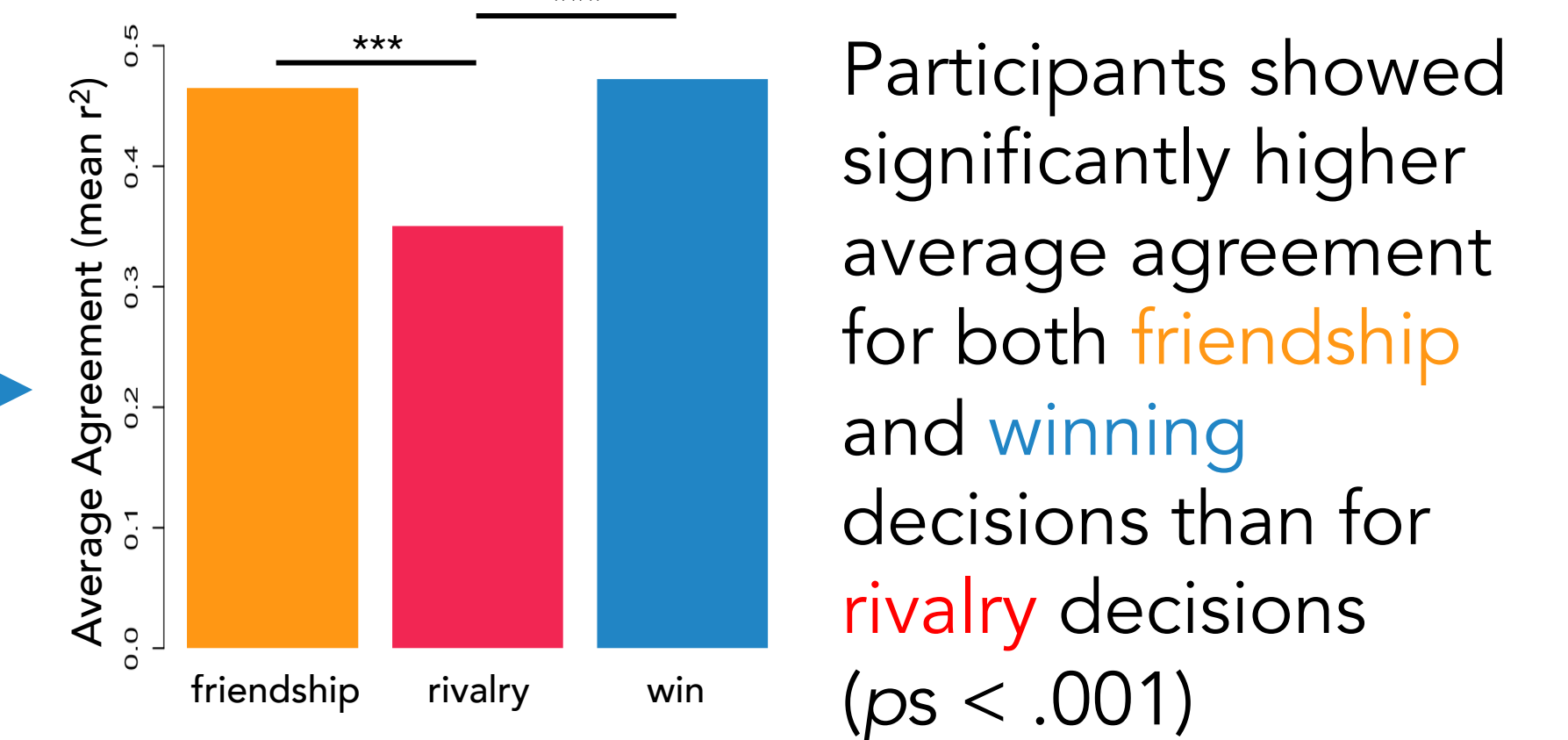
Similarity scores calculated using the Universal Sentence Encoder<sup>6</sup>, sentiment scores calculated using SentimentR package<sup>5</sup>, and confidence scores calculated using LIWC software<sup>7</sup>.

## Results

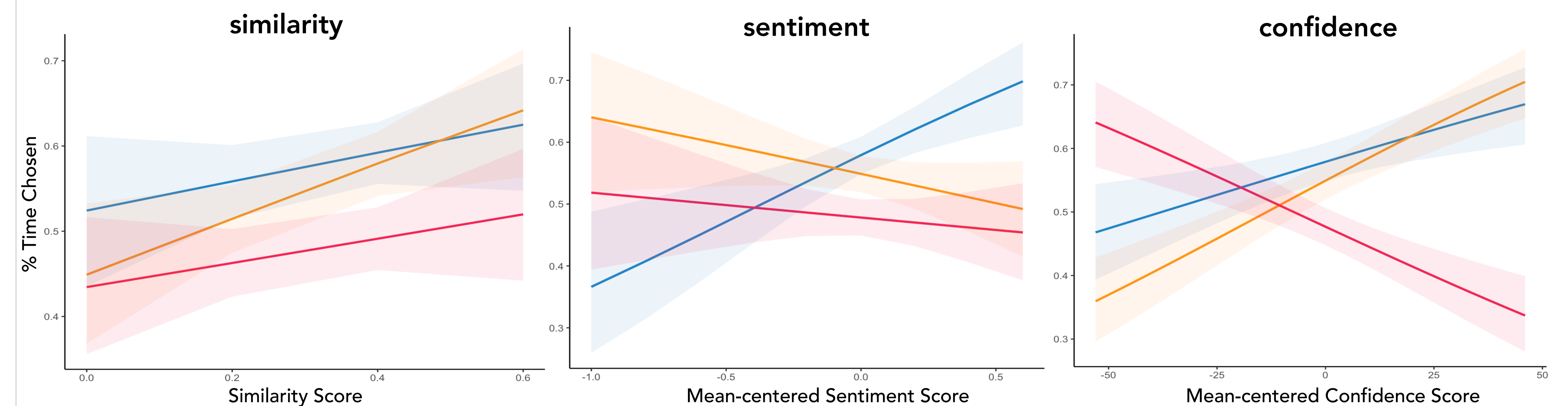
### Network Agreement



Participants showed high levels of network agreement for all relationship choices



### Linguistic Features Inform Social Network Decisions



As dialogue **similarity increases** between two contestants, participants are more likely to decide those contestants are **friends** and less likely to decide they are **rivals**.

As **sentiment becomes more positive**, participants are less likely to decide that contestants are **friends** and less likely to decide they are **rivals**.

As **confidence increases**, participants are more likely to decide that contestants are **friends** and less likely to decide they are **rivals**.

## Summary

- Overall, participants showed high agreement about friendships, rivalries, and likelihood of winning within the social network
- Participants were more likely to infer **friendship** with greater similarity, more negative sentiment, and higher confidence
- Meanwhile, participants were more likely to infer **rivalry** with reduced similarity, more negative sentiment, and lower confidence

## Future Directions

- Repeat the behavioral task while participants undergo fMRI to investigate the neural basis of social network learning. We expect to see interactions between social areas (dorsomedial PFC, precuneus) and memory regions (hippocampus, amygdala, perirhinal cortex) during social network learning.
- Repeat natural language processing analyses on dialogue from all episodes in the season that preceded this current episode to assess if language predicts social relationship changes over time.

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**References** – 1) McPherson et al. (2001). Birds of a Feather: Homophily in Social Networks. *Annual Review of Sociology*, 27, 415-444. 2) Son, J.Y., Bhandari, A., & FeldmanHall, O. (2021). Cognitive maps of social features enable flexible inference in social networks. *PNAS*, 118(39), 1-11. 3) Alt et al. (2021). The Face of Social Networks: Naïve Observers' Accurate Assessment of Others' Social Network Positions From Faces. *Social Personality and Personality Science*, 13(1), 118-126. 4) Jackson et al. (2021). From Text to Thought: How Analyzing Language Can Advance Psychological Science. *Perspectives on Psychological Science*, 17(3), 805-826. 5) Rinker, T. (2021) Package 'sentimentr': Calculate Text Polarity Sentiment. CRAN. 6) Cer et al. (2018) Universal Sentence Encoder for English. *Proceedings of the 2018 Conference on Empirical Methods in Natural Language Processing: System Demonstrations*, 169-174. 7) Pennebaker et al. (2015) The development and psychometric properties of LIWC2015. Austin, TX: University of Texas at Austin.