The Werewolf Among Us: Humans vs LLMs in Multi-Agent Games

Bhavana Jonnalagadda

Riley Jones

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Introduction

- A description of the problem and its significance
- How do LLMs function in a multi-agent environment?
 - Each has limited information
 - Approximately the same ability, traits, skills
- We use LLMs to simulate whether synthetic agents can participate in complex, adversarial group dynamics
- Werewolf is a good candiate for testing multi agent systems of cooperation and secrecy
 - Need citation
 - The game tests adaptive reasoning, strategic alignment, and collective threat detection under special conditions

Related Work

Multi-Agent LLMs

- Among us game (Chi, Mao, and Tang 2024)
- Collective problem solving (Du, Rajivan, and Gonzalez 2024)
 - "analyses indicate that LLM agent groups exhibit more disagreements, complex statements, and a propensity for positive statements compared to human groups"
- Govsim (Piatti et al. 2024)
 - "In GOVSIM, a society of AI agents must collectively balance exploiting a common resource with sustaining it for future use. This environment enables the study of how ethical considerations, strategic planning, and negotiation skills impact cooperative outcomes."
- All found similar themes
 - That LLMs are capable and good at understanding the rules
 - That they can cooperate and be sneaky

LLMs and Werewolf

- Examination of improving werewolf by LLMs (Xu et al. 2024)
 - "our agents use an LLM to perform deductive reasoning and generate a diverse set of action candidates. Then an RL policy trained to optimize the decision-making ability chooses an action from the candidates to play in the game. Extensive experiments show that our agents overcome the intrinsic bias and outperform existing LLM-based agents in the Werewolf game."
- Werewolf Arena (Bailis, Friedhoff, and Chen 2024)
 - Used in this paper
- Explicitly discuss how none of the exisiting LLM+Werewolf papers examine the differences/compare from a human dataset

Methods

Data

Werewolf Among Us Human Dataset

- Human dataset description (Lai et al. 2022)
- Is specifically for a form of one-night were wolf
 - Describe key differences
- Used specifically for the text available
 - and annotations of persuasion strategy on the text

Werewolf Arena

- (Bailis, Friedhoff, and Chen 2024)
- Discuss the framework, how it works, prompts, etc
- Discuss what types of runs we did
- Discuss the data included in output
- Talk about how we had to annotate the LLM speech with persuasion strategies ourselves

Analysis

• Formatted data to match, performed various comparisons

Results

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(a) The LLM wins, by how many rounds that partiticular game had.

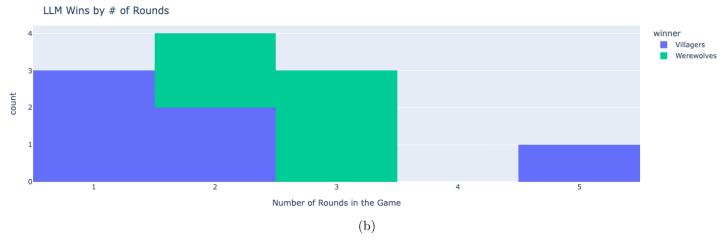


Figure 1

Source: Werewolf Among Us: Human vs LLM Analysis

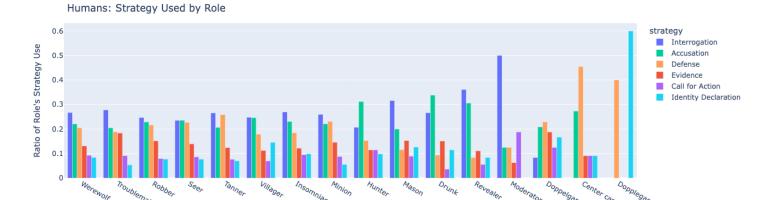


Figure 2

Role

Source: Werewolf Among Us: Human vs LLM Analysis

LLMs: Strategy Used by Role

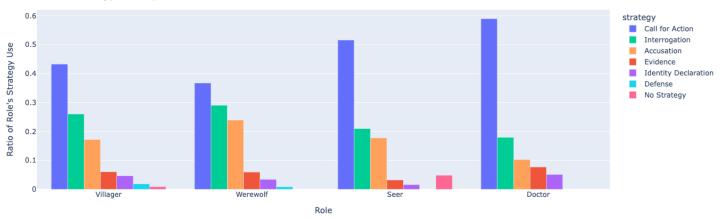


Figure 3

Source: Werewolf Among Us: Human vs LLM Analysis

Discussion and Conclusion

Interpret findings, discuss limitations, and propose future work.

Limitations

Future Work

Summary

Summarize contributions and insights from the project.

References

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Project Contributions

Bhavana Jonnalagadda:

- Paper framework (Quarto) setup
- Github repo management
- EDA on LLM dataset
- $\bullet\,$ Final comparison EDA and results analysis
- Results section
- Discussion and Conclusion section
- Abstract

Riley Jones:

- EDA on human dataset
- Werewolf Arena LLM simulation running and data aquisition
- Introduction section
- Methods section