

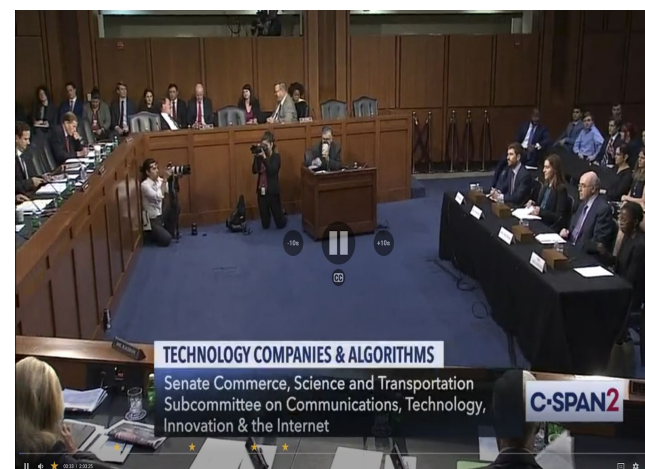
Background and Objectives

HodIPal

- HodIPal is building a civic engagement platform that empowers communities with collective trust. HodIPal's topic-specific reputation scores enable people to filter content collaboratively and explore diverse perspectives.

Background

- People generally obtain information on social networks that algorithmically filter content
- Highly polarizing content or “clickbait” spreads faster through social networks and is more likely to be shared [1]
- Simulations and empirical studies have been conducted to study dynamics of information propagation in large social networks [2]



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Objectives

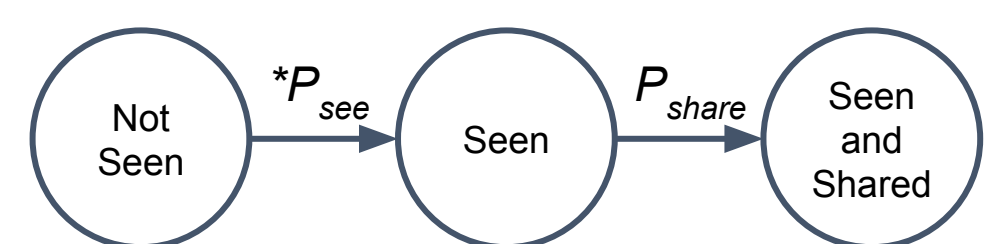
- Simulate the impact of HodIPal towards accelerating the spread of credible information over non-credible information
- Test how much HodIPal can improve the digital health metrics of shared attention and variety of perspective using its Lens feature and credibility score
- Gain user feedback on platform design and improve UX

Materials and Methods

Simulation Setup

- Information diffusion in HodIPal and in a generic social platform were simulated
- Model system consists of a 10k-node social graph with 1-way follow relationships
- Scenario: non-credible information injected into 5 strongly-connected nodes
- Simulation code developed in Python using Linux-based tools and environments
- Based on Soil agent-based social network simulation framework
- Assumptions:
 - Power law distributions of node degrees in directed social graph [3, 4]
 - Power law probability of seeing information if shared [5, 6]
 - Exponential probability of sharing information if seen [7]
 - Reduction in P_{see} of non-credible information by 2x using HodIPal trust system

Multi-Agent Hidden Markov Model



Transition

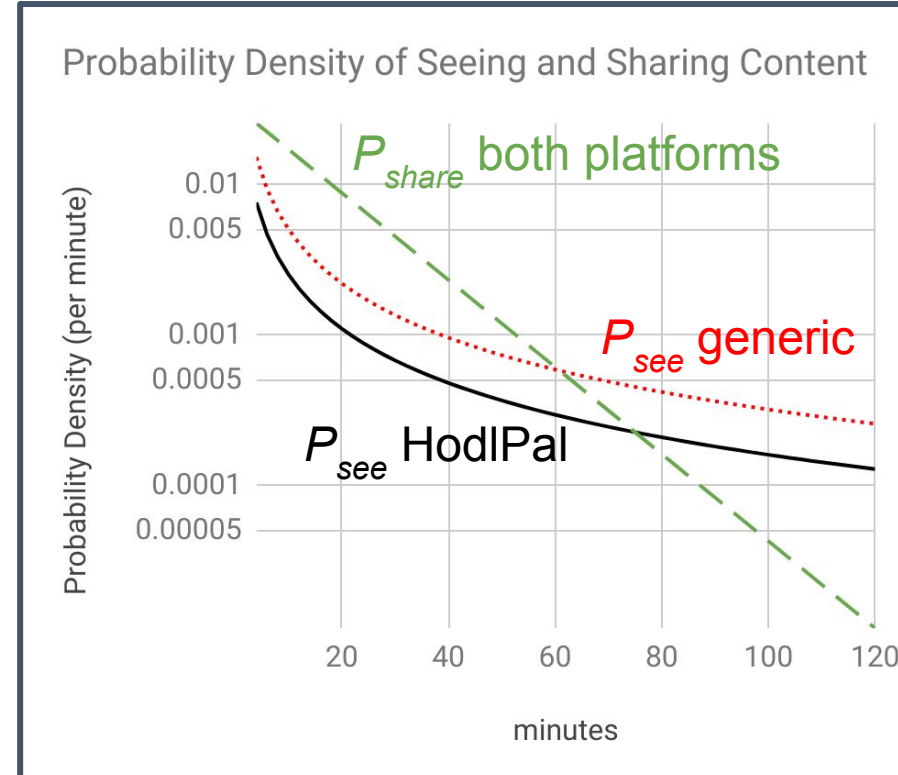
Probabilities

$$P_{see}(a_j, t) = C (t - t_j)^{-a}$$

$$P_{share}(a_j, t) = B \exp\left(\frac{t - t_j}{\tau}\right)$$

$$*P_{see}(a_j, t) = \frac{1}{N_j} \sum_{a_k \in \{A_j\}} P_{see}(a_k, t)$$

a_j - agent in network
 N_j - # of neighbors of a_j
 t_j - time a_j transitioned
 $\{A_j\}$ - agents sharing at a_j
 a, B, C, τ - constants

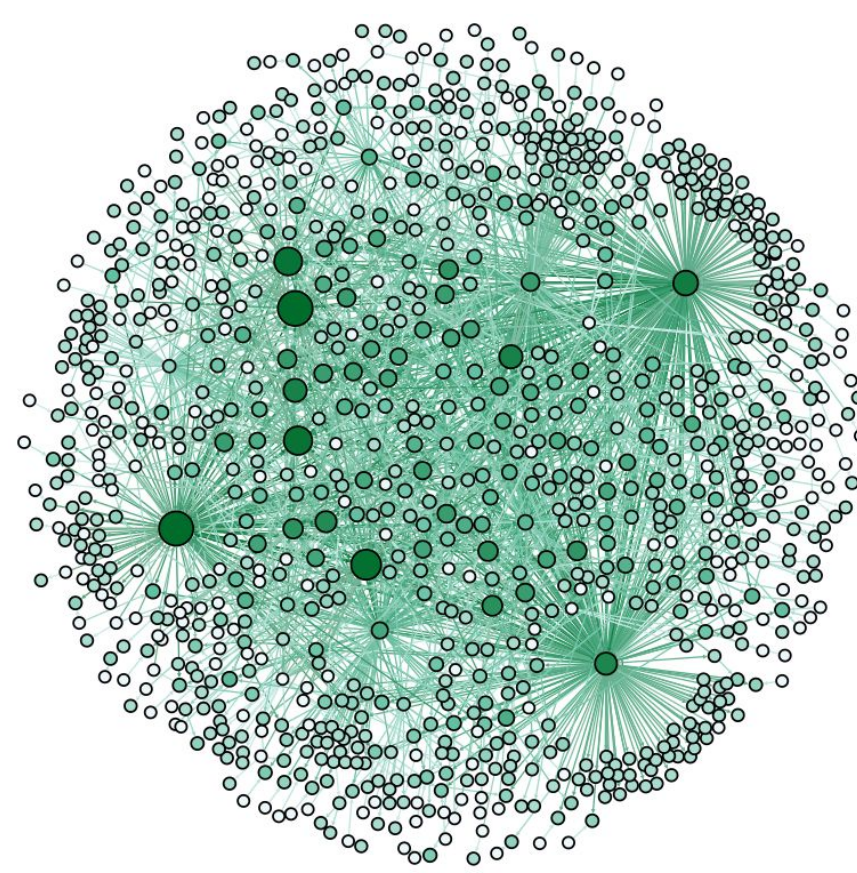


1:1 Interviews with Potential HodIPal Users

- Metrics for digital health were researched [8] and incorporated into design
- Proto.io was used to develop a prototype of the HodIPal UI on a mobile device
- UserInterviews.com was used to survey and screen 88 interviewees as potential power users of the HodIPal platform in as applied towards political discourse
- Screener Questions were designed to pick power users for civic engagement
- Interview Questions designed to test effectiveness of platform in addressing the metrics, including: variety of perspective, receptivity, awareness/self-awareness
- Six potential users were interviewed for 60 minutes each

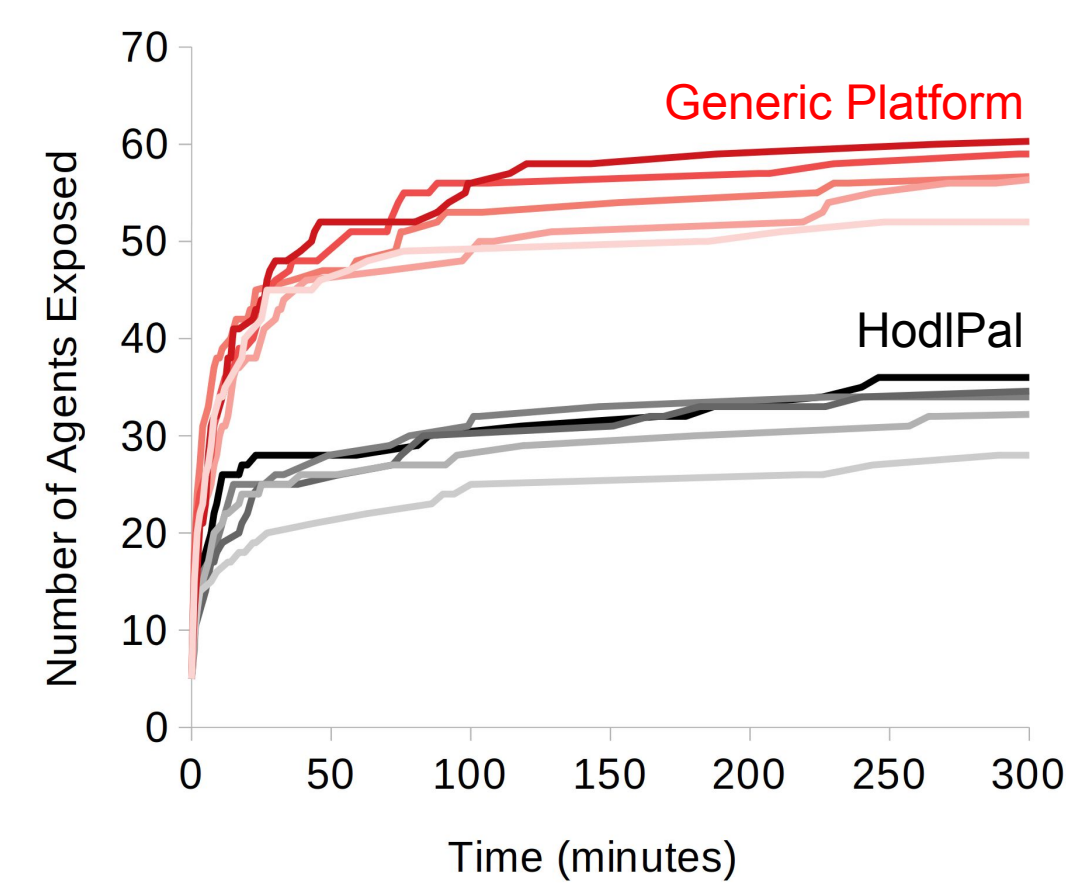
Results and Discussion

1k-Node Social Graph Subset



- Larger node sizes and darker shades = more followers
- Strongly connected “influencer” nodes are clearly visible

Simulated Spread of Information



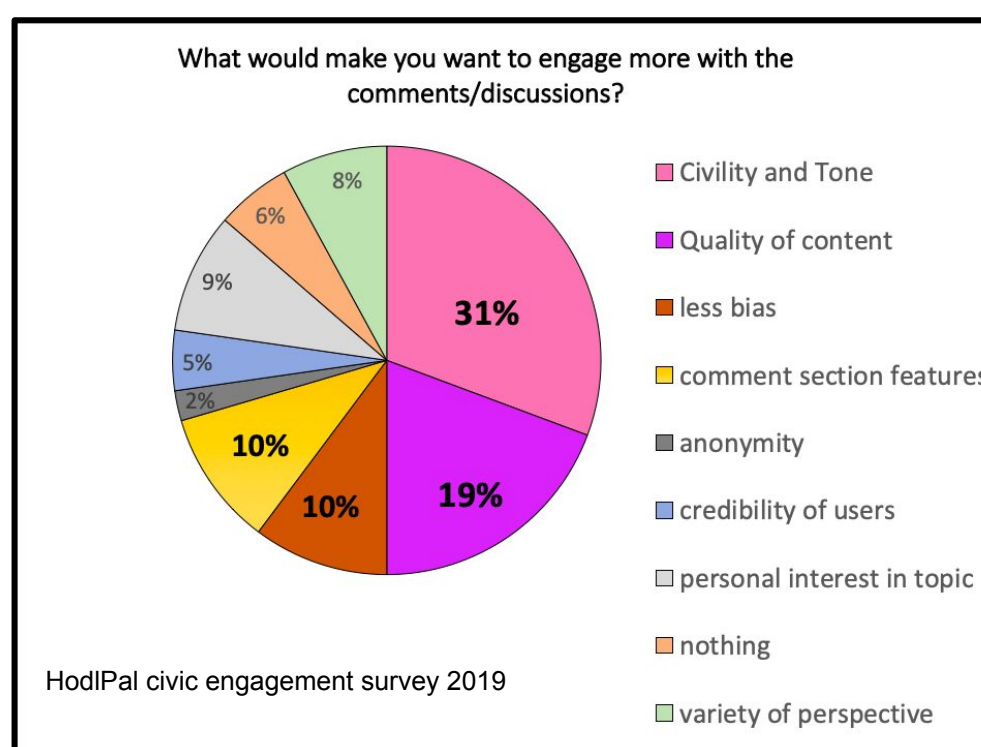
- Stochastic response repeated 10x
- HodIPal suppresses the spread of assumed non-credible information (lower probability of being seen)

Survey Data

- 47% of survey respondents are not satisfied with social media as an avenue for political news and are looking for a new platform
- 58% of survey respondents read, react or respond to comments online
- 70% of respondents found that political content they see online is at least somewhat biased

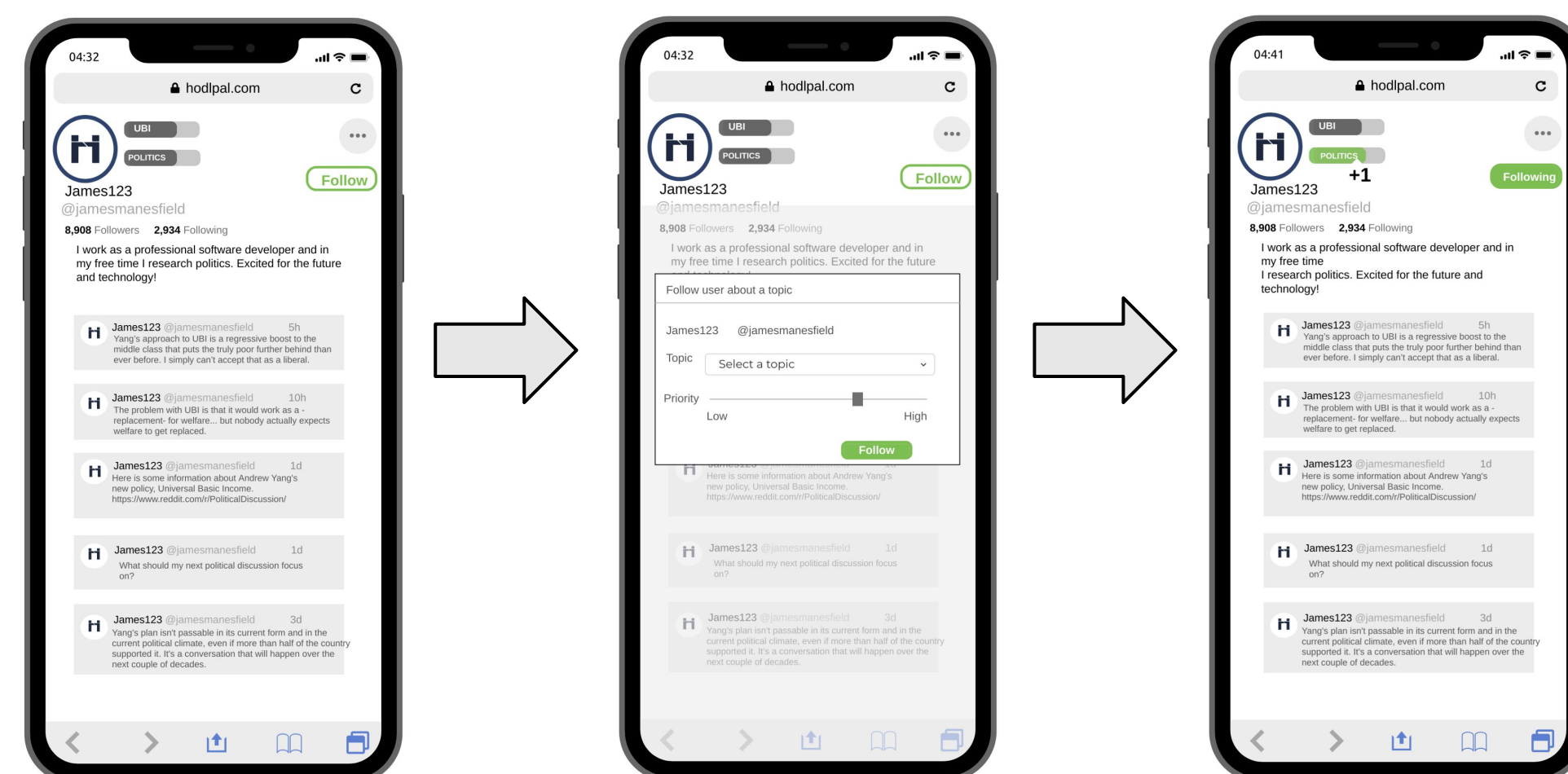
Survey Findings

Users described experiences in existing platforms. Bias and issues with content in the comment sections gave users problems when engaging with political news. Users either avoid reading comments, commenting their own thoughts or they use multiple sources to compare views on topics.

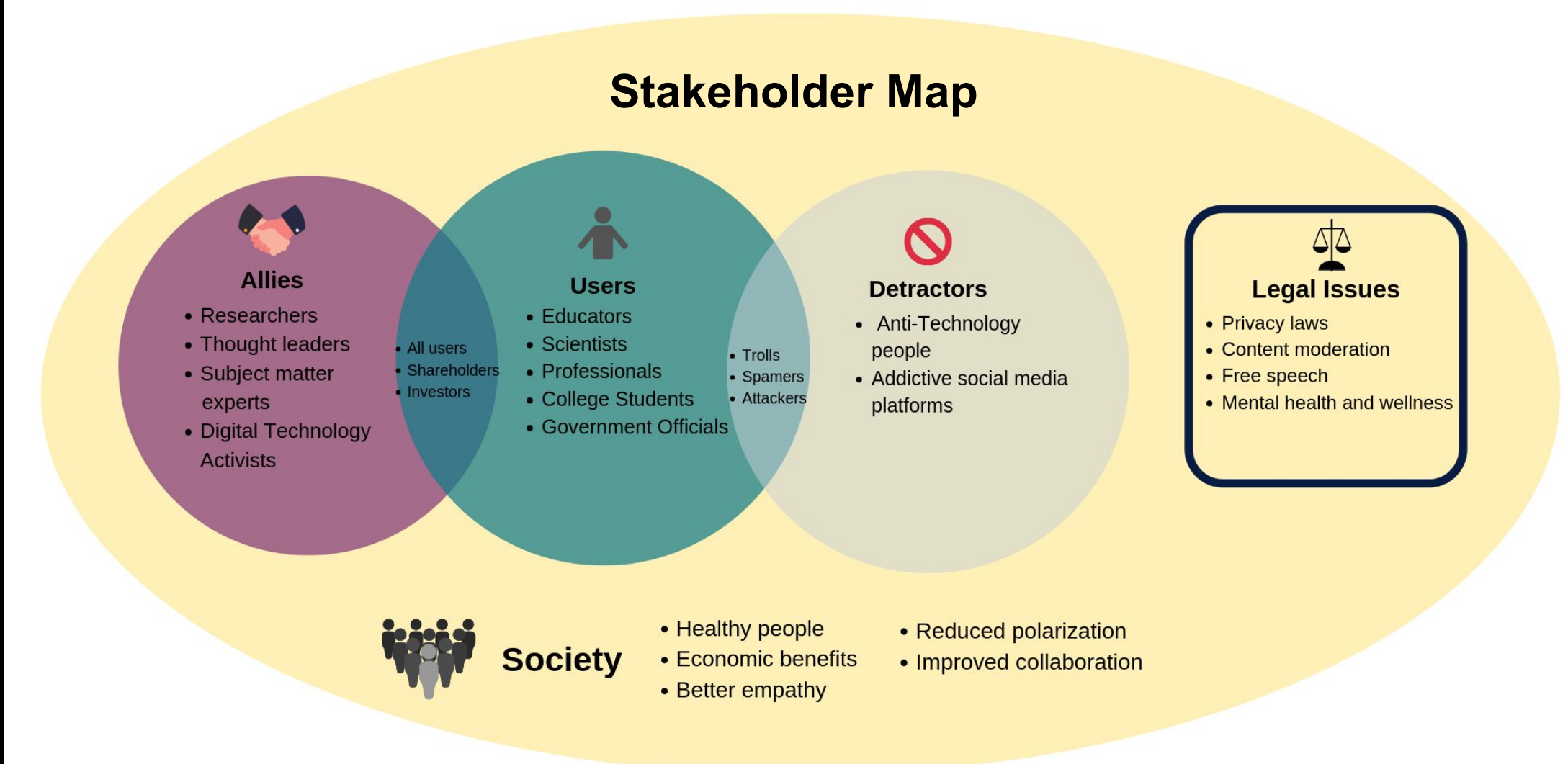


Prototype and Interview

Participants were led through the process of following a user on HodIPal by choosing a topic, setting a priority for that topic and seeing the effect on credibility score.



Social Implications



Through this research, HodIPal aims to show that it can improve the quality of conversations online. The platform can reduce toxicity, promote the spread of accurate information, and help people circumvent misinformation. Promoting these practices enables people to mitigate the influence of filter bubbles, consequently boosting the potential for collective intelligence to arise on the platform.

Conclusions

- Stochastic agent-based hidden markov simulation framework was developed
- Empirical data from literature provides a basis for probabilistic simulations
- Explicit separation of agent “see” vs. “share” events enable independent modeling of platform filtering processes vs. decisions made by users, highlighting intentionality
- Based on the survey we concluded that the quality of the news content and its comments have a significant impact on how a user engages with political news
- Interviewees responded positively to HodIPal's Credibility Score and Lens features, designed to provide diverse perspectives while engaging in meaningful conversations

Future Work

- Add more layers of complexity to the simulation such as edge weights and chance of node creation/deletion, as well as improve existing model methods and parameters
- Gain up-to-date user feedback as additional features are altered or created
- Continue work to release a working test version of the platform

Acknowledgements

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