

# YouTube's recommendation algorithm creates rabbit holes, not echo chambers

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# Talk in a slide

- Why no echo chambers [on Youtube]?
- Because rabbit holes dominate.

# Definitions

- **Echo chamber:** “a bounded, enclosed media space that has the potential to both magnify the messages delivered within it and insulate them from rebuttal” *(link)*

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  - **Iterative:** one piece of content leads to the next
  - **Exhaustive:** subsequent pieces of content dig deeper into a topic
  - **Associative:** subsequent pieces of content can deviate

# Puzzle

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- **Assumption 2:** Diverse supply of information
- **Model:** Information as a consumption good (*McFadden, 2019*)
- **Expectation:** Echo chambers should prevail!
- **Puzzle:** Echo chambers are rare (*Guess et al., 2018; Barbera, 2015; Guess, 2021*)

►► Intuition

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  - ...on Youtube
  - ...in 2020

# Data and Methods

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- 600 also provided their YouTube watch histories for us

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  - Ideology (*Lai et al., 2024*)
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- For each video, we observe:
  - Ideology (*Lai et al., 2024*)
  - Video “category” from YouTube API
- **Echo Chamber:** homogeneity of ideology of videos
- **Rabbit Hole:** rate at which subsequent videos / recommendations shift to match current video

# Roadmap

- Descriptives: Watch Histories
  - **Echo chambers** don't (really) exist
  - **Rabbit holes** do exist

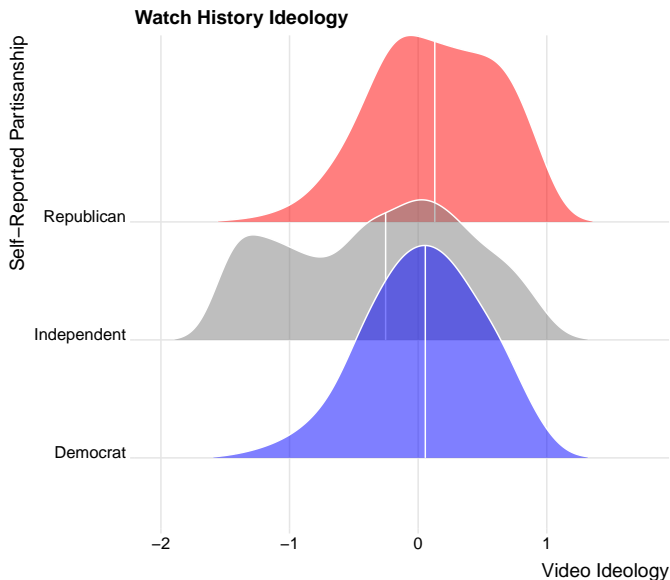
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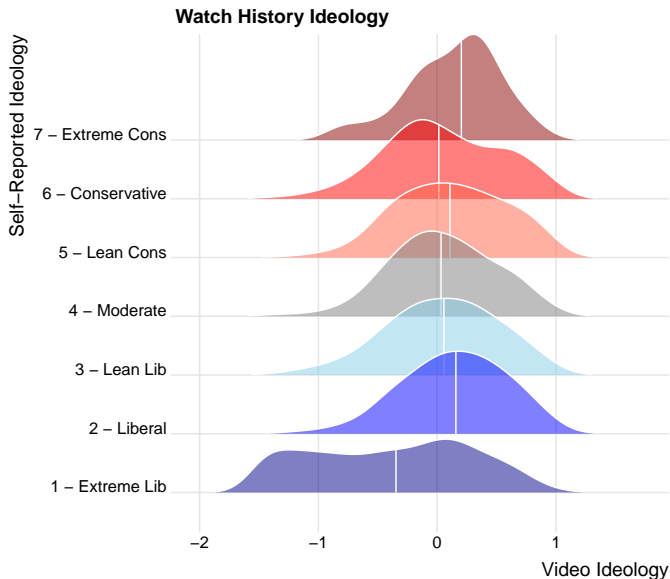
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- Inference
  - Predict recommendation video ideology as a function of current and average of watch history

# Watch Histories: Echo chambers don't (really) exist

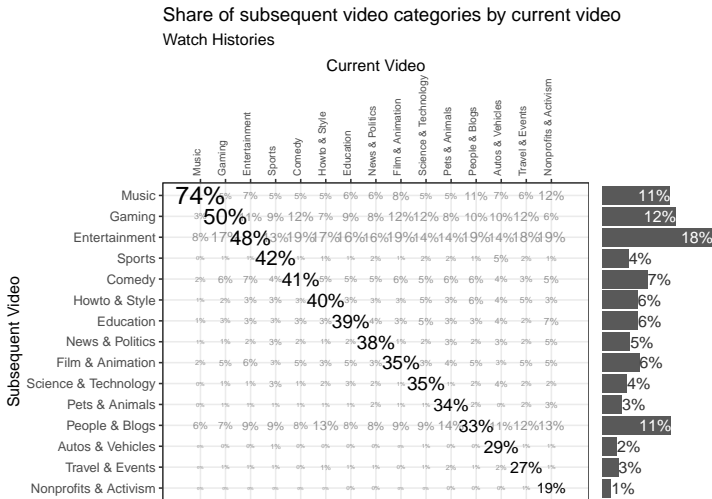




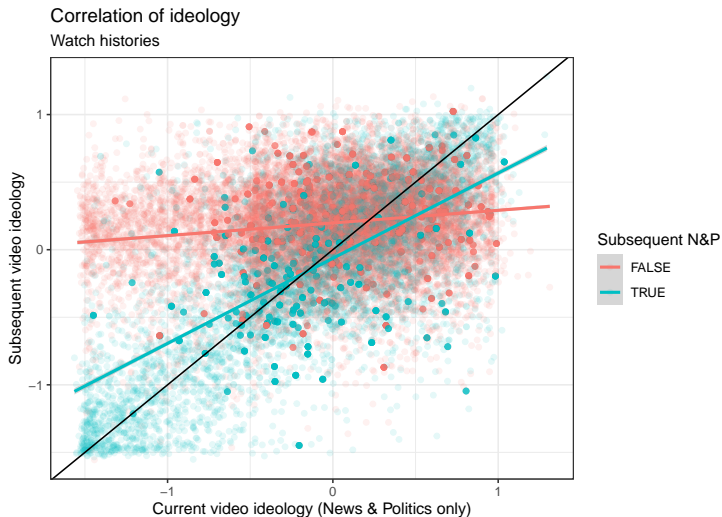
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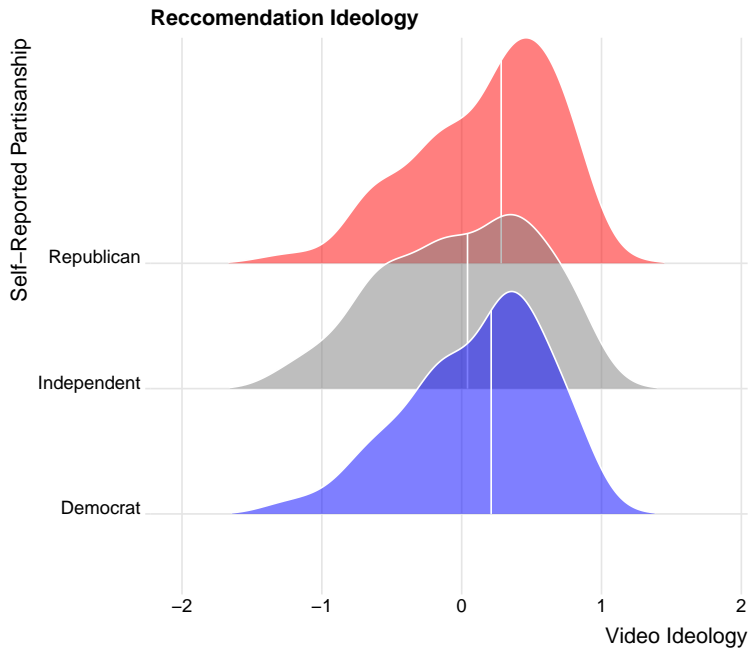
# Watch Histories: But rabbit holes might



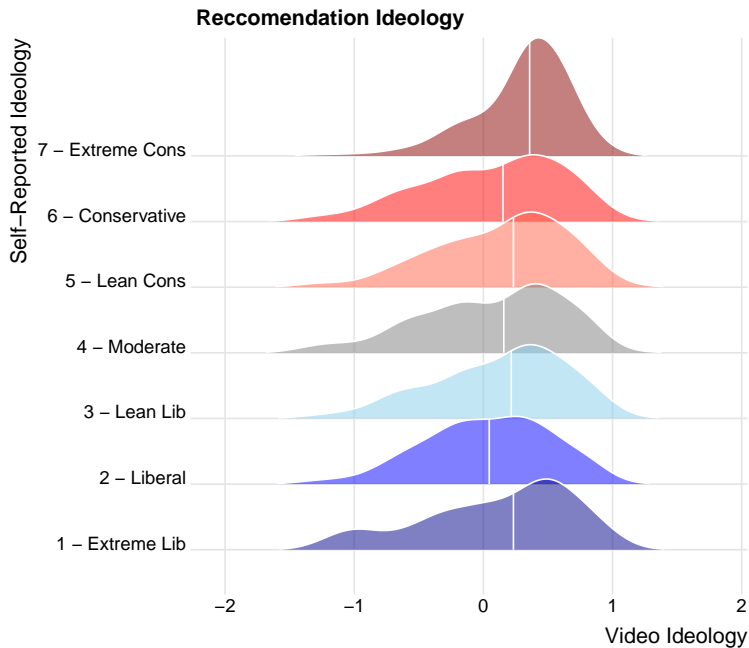
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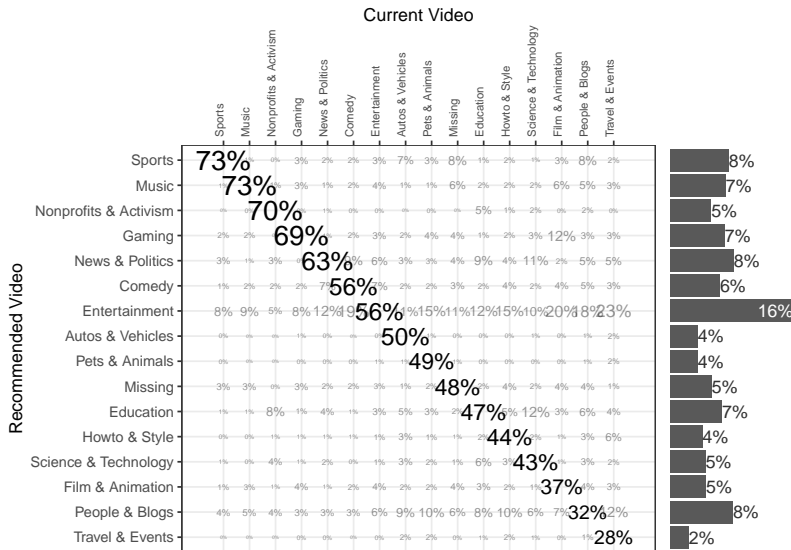
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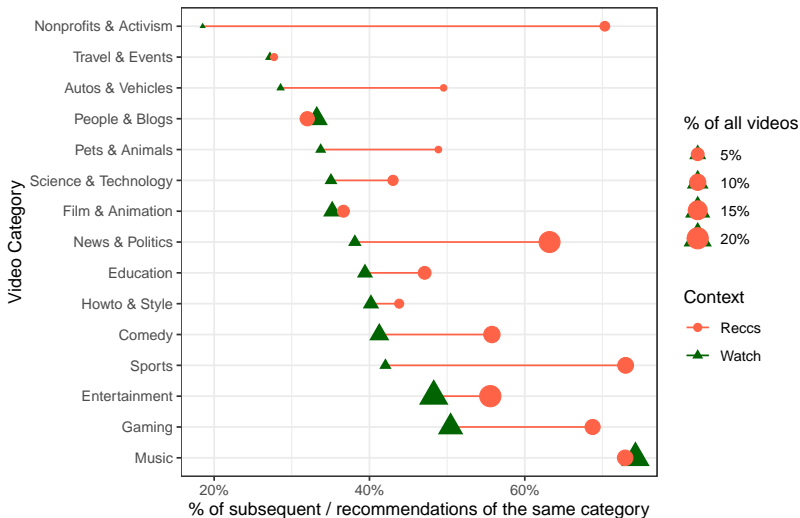


# Recommendations: But rabbit holes do

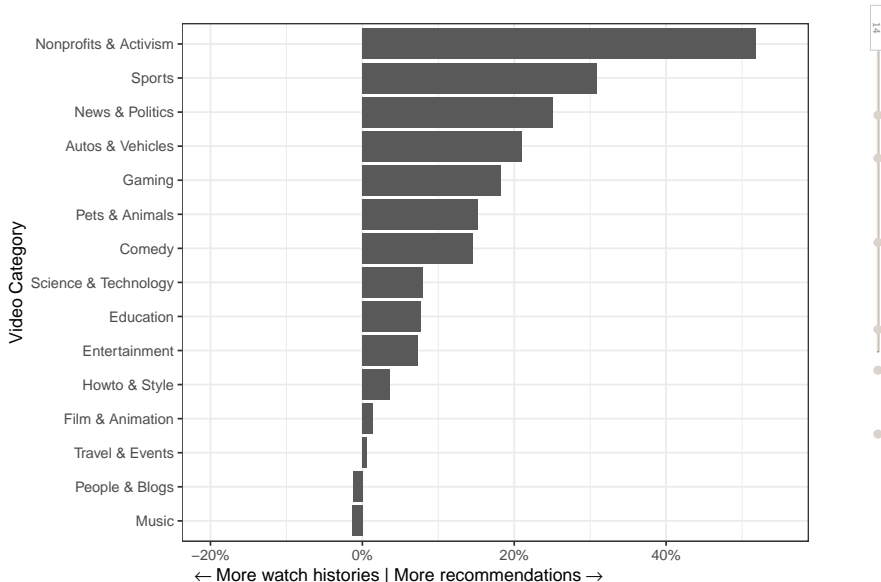


# Recommendations: Stronger rabbit holes

Echo Chamber Proportions by Context  
Watch histories versus recommendation algorithm



# Recommendations: Stronger rabbit holes





# Regressions

- What is the relative weight of a user's watch history versus the current / recently watched videos on their recommendations?

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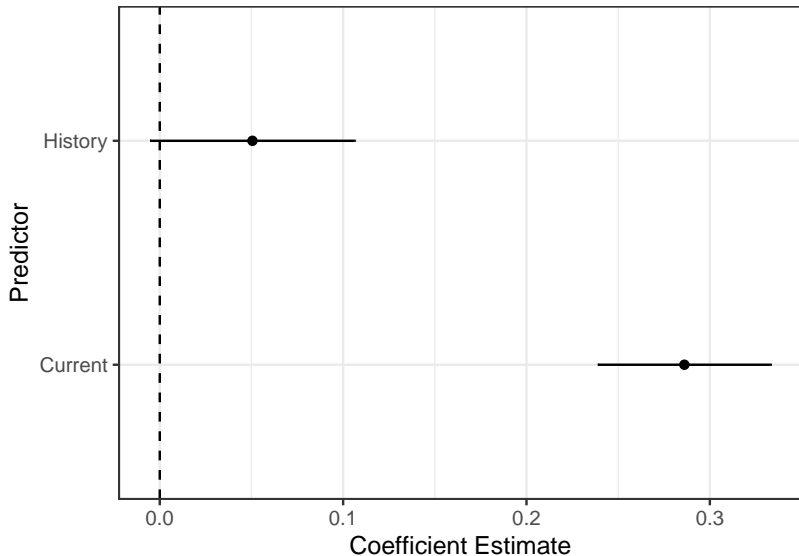
$$ideo_{irc} = \alpha_i + \rho_c + \beta_1 ideo_{itc} + \beta_2 ideo_{ic} + \varepsilon_{ic}$$

- where  $i$  indexes individuals,  $t$  indicates time (sequence),  $c$  indicates category, and  $r$  indicates recommendations
- $\alpha_i$  and  $\rho_c$  are individual and category fixed effects, respectively

# Results

## Predicting Recommendation Ideology

Watch history + Current



# Discussion

- Youtube circa 2020 suggests that users experience **rabbit holes**, not **echo chambers**
- Suggestive evidence that this is (partially) the product of the recommendation algorithm

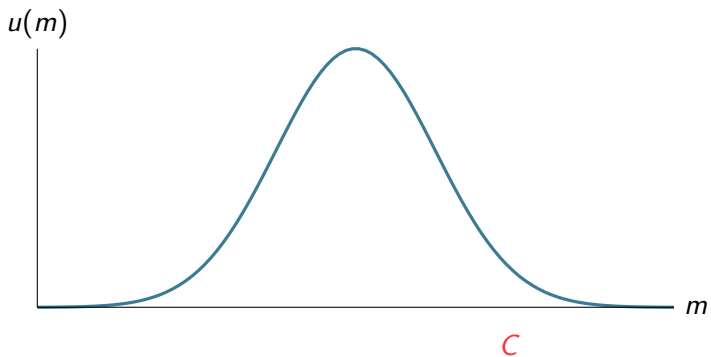
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- Youtube circa 2020 suggests that users experience **rabbit holes**, not **echo chambers**
- Suggestive evidence that this is (partially) the product of the recommendation algorithm
- Interpretation relies on how dogmatically we think the **homophily assumption** describes info consumption
  - **Dogmatist**: Youtube's algo sux
  - **Skeptic**: Can we learn about demand for info from it?
  - **Depressingly**: are we studying an algorithm instead of the humans that we actually care about?

Thank you

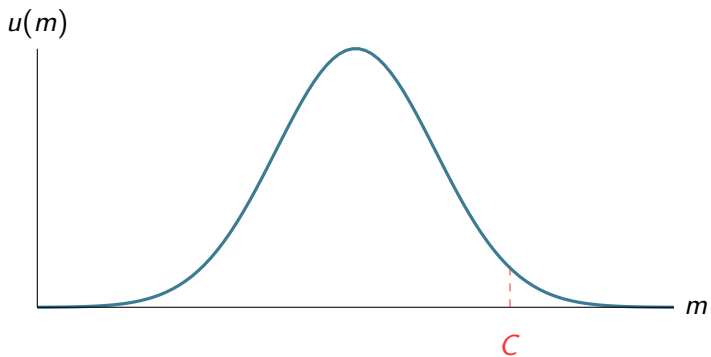
All faults are my own, Vaibhav is a star  
**[james.h.bisbee@vanderbilt.edu](mailto:james.h.bisbee@vanderbilt.edu)**

# Puzzle



» Back

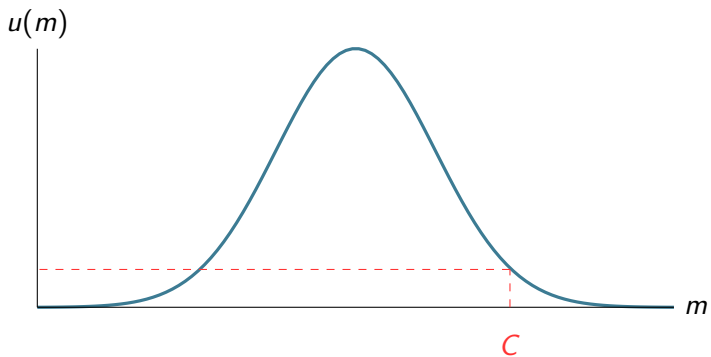
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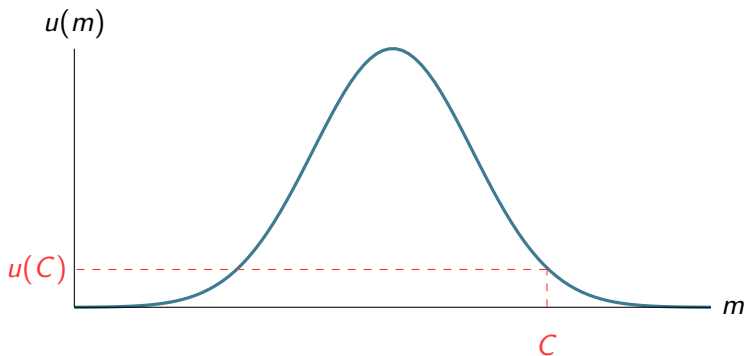


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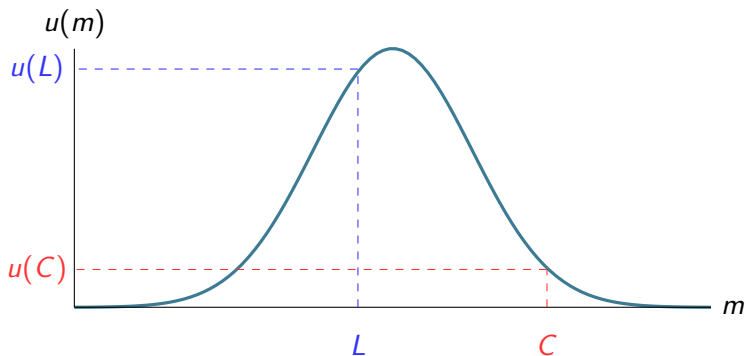
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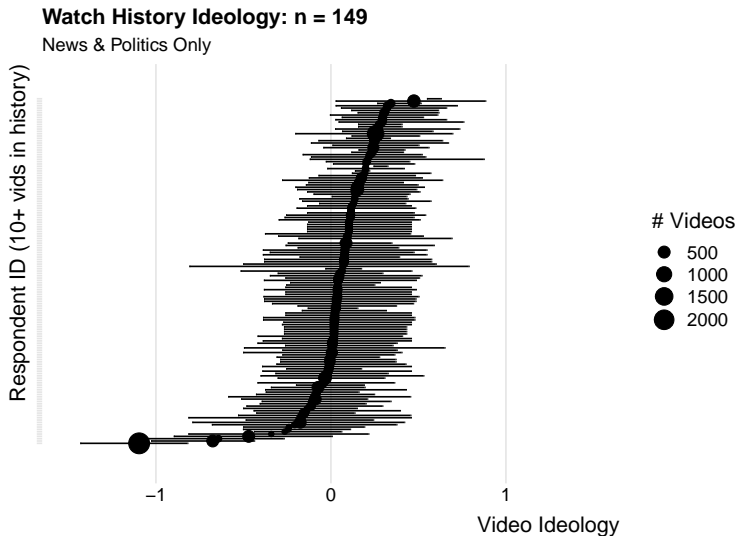
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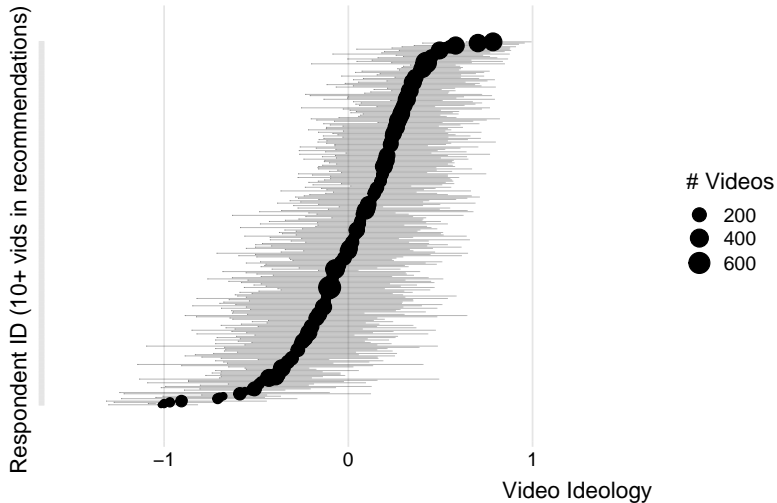
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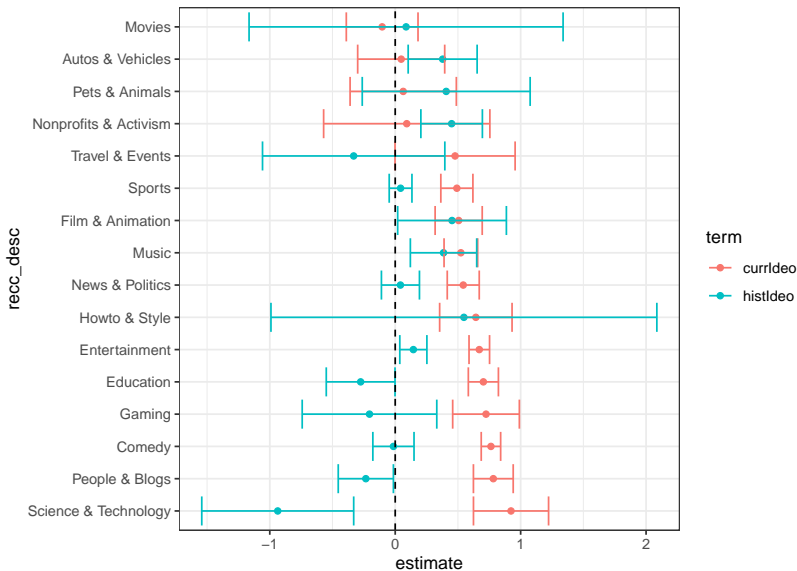
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## Recommendation Ideology: n = 298

News & Politics Only



# 3-way interaction



# Watch Histories: But rabbit holes might

