# Political Polarization in Online News Consumption

By Garimella et al. (2021)

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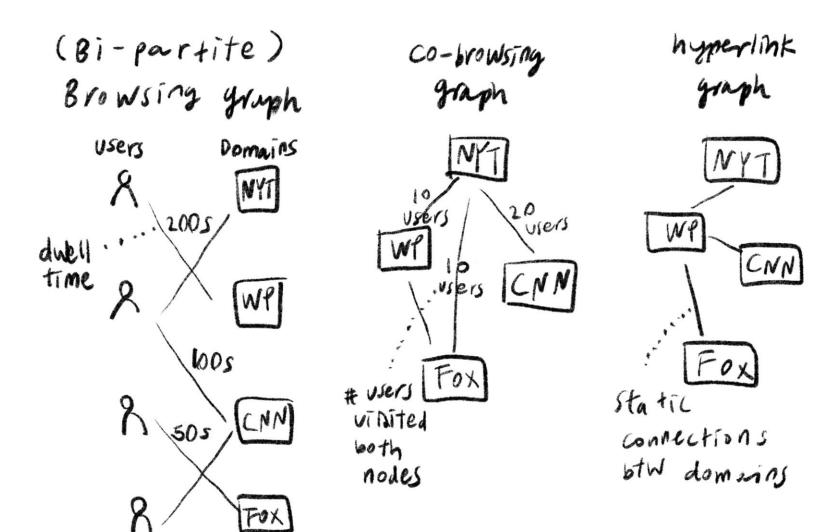
# Research Question & Dataset

- Examine the information-seeking behaviors of individual Web users to understand the relationship between mass polarization and the consumption of news content on the web
- Dataset: Large-scale dataset consisting of detailed browsing logs collected via Firefox extension

# Research Method & Participants

- Observational study: collect browsing log via Firefox extension
- Descriptive research strategy: Looking into variable of political polarization
  Political polarization pattern discovered in online news consumption
- Correlational research strategy: relationship between political polarization and online news consumption
  - The association between online news consumption and political polarization
- Measure the frequency and duration of online news consumption
- Randomly recruited from active Firefox users based in the US from 5 April 2018 for three weeks.
- Opt-in participant recruitment with informed consent

# Data analysis strategy

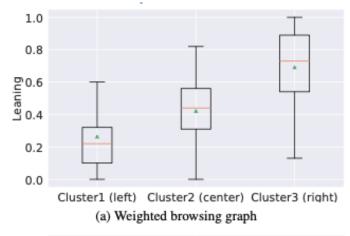


	Nodes	Edges	Avg. degree	Top PageRank
Co-browsing	1,295	176,945	273	nytimes.com, washingtonpost.com, cnn.com, theguardian.com, npr.com,
Hyperlink	1,295	323,036	498	washingtonpost.com, huffingtonpost.com, nytimes.com, businessinsider.com, theatlantic.com

Table 1: Statistics of co-browsing and hyperlink graphs.

# Result 1: Polarization pattern in news browsing

#### 1. Polarization in Dwell Times



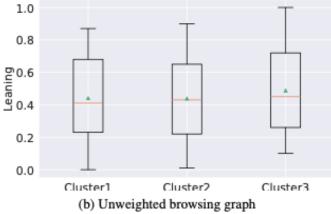


Figure 3: Boxplots of leaning scores per domain cluster, obtained by co-clustering (a) the weighted bipartite (user-by-domain) browsing graph, (b) an unweighted version of the browsing graph. Green dots are means, red lines are medians, box boundaries are quartiles.

### 2. Time Spent "On the Other Side"

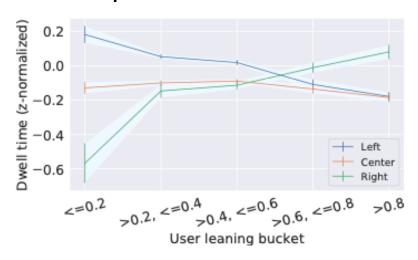


Figure 4: Average dwell time (z-score-standardized within users) spent by users of different leanings (x-axis) on domains of different leanings (three curves). Error bars show 95% confidence intervals. We see that users spend significantly more time when visiting pages on domains aligned with their own leaning.

\* leaing score:

0 = left-leaning

1 = right-leaning

# 3. Community Structure in Co-browsing Graph

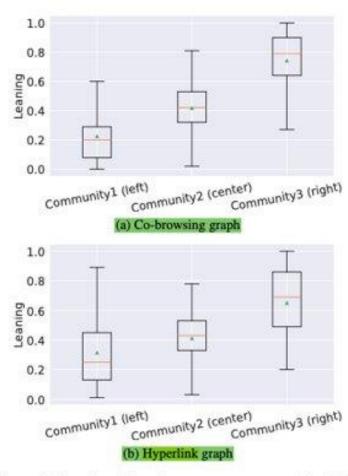


Figure 5: Boxplots of leaning scores per community of domains, obtained by running community detection on (a) the co-browsing graph, (b) the <a href="https://hyperlink.graph.">hyperlink</a> graph. Green dots are means, red lines are medians, box boundaries are quartiles.

# Result 2: Selective Exposure vs. Structure of the Web

### 1. Homophily in hyperlink and co-browsing graphs.

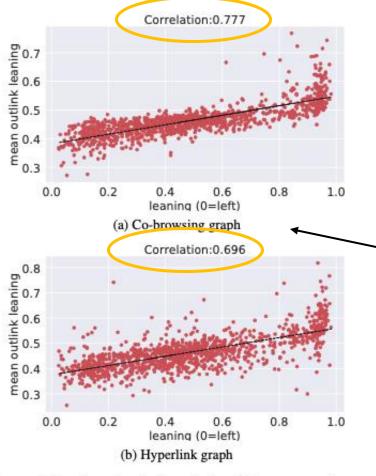
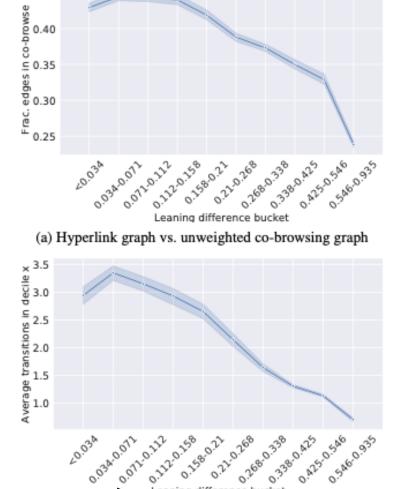


Figure 6: Leaning of each domain (x-axis) vs. average leaning of the domain's neighbors (y-axis) in (a) the co-browsing graph and (b) the hyperlink graph.

Figure 7: (a) Fraction of edges from the hyperlink graph that are also present in the co-browsing graph, when considering, for each node, only those edges that connect the node to a neighbor whose leaning difference falls into the respective decile (x-axis; deciles were computed per node over all its neighbors). (b) Analogous analysis, but with average edge weights, rather than fraction of edges, on the y-axis. The plots show that hyperlinks leading to more similar (with respect to leaning) neighbors are more likely to be chosen by users.

Pearson's correlation coefficient Significantly lower (0.696 vs. 0.777, p<0.01)



0.45

(b) Hyperlink graph vs. weighted co-browsing graph

Absolute difference with the node's neighbors in hyperlink graph

### Result 2: Selective Exposure vs. Structure of the Web

### 2. Multi-hop neighborhoods: random walks

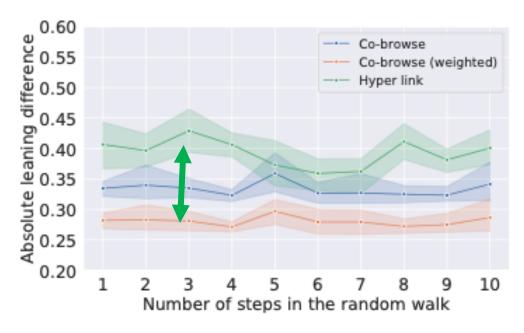


Figure 8: Leaning difference between nodes and their multihop neighborhoods in the weighted and unweighted cobrowsing graphs and in the hyperlink graph, computed by performing random walks of length 10. Error bands show 95% confidence intervals. The weighted cobrowsing graph is most homophilic, followed by the unweighted cobrowsing graph and the hyperlink graph.

# Strengths

- 1. The study created large-scale fine-grain digital observational data set to analyze online news consumption.
- 2. The study discovered three clusters in multiple sources (co-browsing graph, average dwell time on news domains, and hyperlink graph)
- 3. The study attempts to disentangle different factors in users' online news consumption by comparing hyperlink structure and co-browsing graph.
- + informed consent for participant recruitment
- ? But approved by Mozilla ethical and legal board

### Weakness

- 1. The role of algorithmic influence is absent.
  - User's own choice vs. Hyperlink?
  - Entry for news domain: social media, search engine, email newsletter, Hyperlink...?
- 2. The construct of political polarization is not clearly defined.
  - Loyal NYT reader who doesn't visit Fox news == polarized political left?
- 3. The author's interpretation of partisan audience bias score from Robertson et al. (2018) as political leaning score is misleading.
  - Missing Democrats vs. Republican context
  - Scaling from -1 to 1 to 0 to 1?
  - Matched to voter registration vs. Self-reported ideology
- + No shared any data, code, collection tool
- ? limitation of industry research with NDA?