4 Things Nobody tells you about Online News a Model with Social Networks and Competition

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### Introduction

#### How good can ad-funded online news outlets get?

- Why a model?
  - Social media changed news market:
    - ★ Advertisement revenues  $\rightarrow$  profits from social media attention
    - ★ Sharing content  $\rightarrow$  consumers play incentivizing role on spreading news
  - $\Rightarrow$  "Old" market environment  $\rightarrow$  new effects?
- Why does it matter?
  - Insights into topical issues
    - ★ More than 50% of adults get news online [Pew Research Center]
    - ★ Fear of market segmentation: paywalled vs. free information
  - Need for policy recommendations
    - ★ Can we trust outlets relying on online shares?
    - ★ Should competition be encouraged?
    - ★ What kind of interventions work?

### Overview

- This model does not deal with:
  - Psychological bias agents are Bayesian
  - Partisanship agents care about the truth(\*)
  - Reputation initial viewership exogenous
- This model deals with a two-sided market:
  - Producers, paid-per-view  $\rightarrow$  ad revenue
  - Consumers, share news → networks
- Sketch of the model:
  - Underlying reality state of the world (SoW)
  - Consumers care about sharing true news; receive private information
  - Producers care about views through shares
    - $\rightarrow~$  invest if true article makes more views than false article

### Results

- High news quality can be achieved only when topic already well-known
  - Consumers believe news more easily if corresponds to:
    - ★ their private information  $\Rightarrow$  news quality bounded
    - ★ their prior  $\Rightarrow$  news articles more valuable in likely SoW
  - $\rightarrow$  Share buttons are not good enough incentives
- Competition does not necessarily help
  - More quality because followers harder to reach
  - Less quality because smaller potential readership
  - $\rightarrow$  Particularly relevant with (almost) free entry for online outlets
- Welfare created through entertainment, hardly by better decisions
   → Ad-funded news outlets are barely *news* outlets
- Flagging can help; quality certification less
  - $\rightarrow$  Timing of fact checking matters

#### Literature

- News market:
  - Two-sided news markets with producer competition.
    - ★ e.g. Allcott and Gentzkow (2017).
    - $\rightarrow$  *Introduce*: networks
  - Two-sided news markets with networks.

★ only Kranton and McAdams (2019) – KM hereafter.

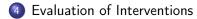
- $\rightarrow$  *Introduce*: competition (& welfare).
- Learning in Networks:
  - Dynamic learning communicating over beliefs/actions.
    - ★ e.g. see Golub and Sadler (2017) for a review.
    - ★ Hsu et al. (2019) deals with behavioral sharing casquade without competition

### Outline









# Model

#### Model

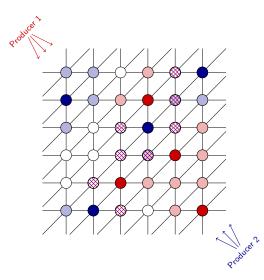
### Environment

- Binary SoW, documented through news articles & private signals
- Producers publish one article each:
  - Reach exogenous number of influencers
  - Choose the quality of the outlet x := Pr(article reports true SoW)
  - Do not directly chose the article's content!
- Consumers receive private signal + at most one article
  - Consumers are arranged on lattice of degree d
  - Influencers come across articles  $\rightarrow$  decide whether to share z
  - Followers read article if a neighbor shared
    - ★ If different articles shared, only one appears to follower (random source)

(Timing: simultaneous / equilibrium concept: NE)

#### Model

### Model

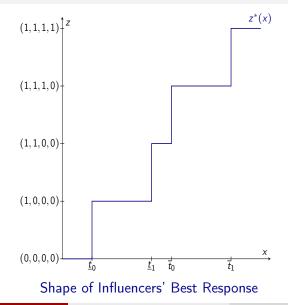


### **Objectives:** Influencers

• Payoff from sharing = 
$$\begin{cases} 1 & \text{if article true} \\ -1 & \text{otherwise} \end{cases}$$

- ⇒ share if probability that the article is true  $\geq \frac{1}{2} \rightarrow$  depends on:
  - $\star$  article content *n*
  - ★ private signal s
  - ★ outlet precision x
  - Note: strategy  $z_{n,s}(x)$  = probability of sharing article
- Share content *n* if *x* high enough given  $s \rightarrow$  thresholds.
  - Sharing is (weakly) monotone in x
  - Share anything if  $x \ge$  private signal's precision

### **Objectives:** Influencers



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### **Objectives:** Producers

- Costs: C strictly convex; c(x) marginal cost
- Revenues: expected portion of views:
  - depends on the sharing behavior of influencers  $\rightarrow$  on z
  - depends on whether my article is true  $\rightarrow$  on x
  - depends on whether others' article is true  $\rightarrow$  on  $x_{\text{others}}$

Note:  $V_{w,n}(z)$  = expected revenue from content *n* when SoW is w $\Rightarrow$  Revenues:  $w_0[xV_{0,0} + (1-x)V_{0,1}] + (1-w_0)[xV_{1,1} + (1-x)V_{1,0}]$ 

• Best-response, i.e. *incentive to invest*:

$$x^{*}(z) = c^{-1} \Big( w_{0} [V_{0,0}(z) - V_{0,1}(z)] + (1 - w_{0}) [V_{1,1}(z) - V_{1,0}(z)] \Big)$$

extra value of true article when SoW is 0

extra value of true article when SoW is 1

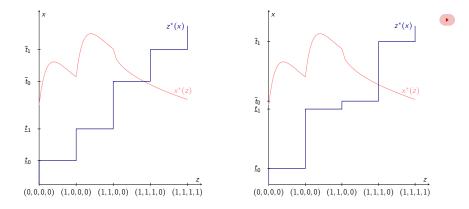
Note: 
$$V_{w,n}(z) = \frac{b}{|U|} + (1-b) \sum_{m} \Pr(m|w) \frac{P_{u|w,n}}{P_{u|w,n} + P_{-u|w,m}} \left(1 - (1 - P_{u|w,n} - P_{-u|w,m})^d\right)$$

# Equilibrium

### Without Competition

#### **Proposition 1**

Unique NE characterized by news' quality:  $x^{M} = \max\{\min\{x^{*}(1,1,0,0), \overline{t}_{0}\}, \min\{x^{*}(1,1,1,0), \overline{t}_{1}\}\}$ 



### Without Competition

#### Proposition 2

The incentive to invest is single peaked in d

Intuition: big  $d \Rightarrow$  can rely on *few* nodes to share

#### Lemma 2

The incentive to invest is increasing with the certainty about the SoW

Intuition: articles more valuable in more likely SoW

### With Competition

- Call producers a and b.
- Simplification:  $w_0 = 1/2$ 
  - $\rightarrow$  consider only subset of all possible undominated strategies:

$$z_{a|0,0} = z_{a|1,1} = z_{aT}$$
 and  $z_{a|0,1} = z_{a|1,0} = z_{aF}$ 

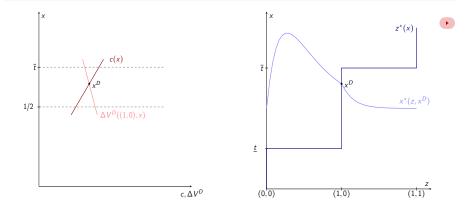
 $\Rightarrow$  two relevant thresholds  $\underline{t}, \overline{t}$ 

• Symmetric equilibria:  $z_a^* = z_b^*$  and  $x_a^* = x_b^*$ 

### With Competition

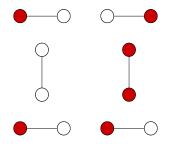
#### **Proposition 3**

# Unique symmetric NE characterized by news precision $x^{D} = \arg \min_{x \in [1/2, \gamma]} |\Delta V^{D}((1, 0); x) - c(x)|.$



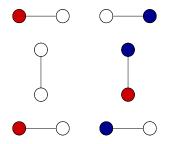
#### Theorem 1

Monopoly leads to higher incentives to invest for  $d < \bar{d}$ , while duopoly leads to more investment for  $d > \bar{d}$ 



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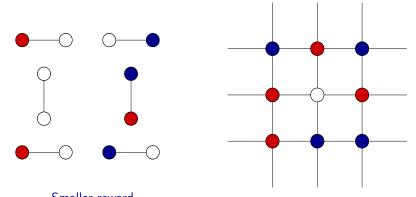
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#### Smaller reward

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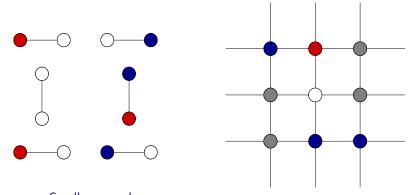
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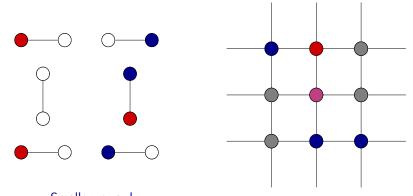
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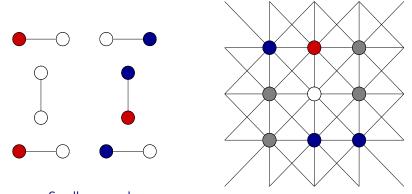
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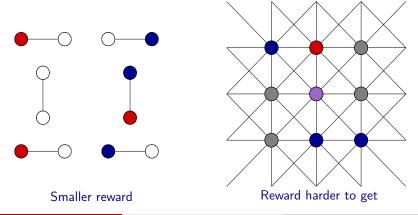
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# Welfare

### Framework

#### Proposition 5

Any equilibrium outcome on an ad-based news market is Pareto inefficient

Welfare evaluation?  $\rightarrow$  add *bet*:

• After articles spread, all consumers can take action a to match SoW

$$u_j(a_j|\omega = w) = egin{cases} 1 & ext{if } a_j = w \ -1 & ext{otherwise} \end{cases}$$

Three aspects of welfare :

- Entertainment: expected utility from sharing
- Guidance: expected utility from bet (no opting out)
- Driver: expected utility from costly bet

#### Welfare

### Welfare Analysis

Disclaimer: Literal interpretation of one signal per agent.

- Sharing strategy  $\approx$  betting strategy
- News quality increases entertainment, not necessarily guidance

#### Lemma 5 (Preliminary)

With  $w_0 = 1/2$ , news outlets are not providing guidance to influencers

• Intuition: news quality bounded by private signal

#### Theorem 2 (Preliminary)

In terms of guidance, only followers can benefit from competition

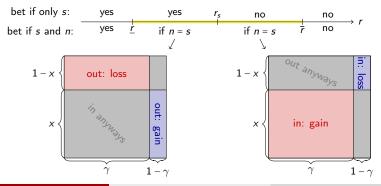
- Influencers do not take better decisions
- Cost of production doubles
- Network filters bad articles out for followers

#### Welfare

### Welfare Analysis

What about the *driver* aspects?  $\rightarrow$  price *r* to enter bet

- Receiving news article can:
  - Motivate agents to take the bet they would not have with s only
  - Discourage agents to take the bet because  $n \neq s \rightarrow$  can backfire
  - $\rightarrow$  ambiguous, depends on *r*



# Evaluation of Interventions

# Flagging

False articles are flagged (before sharing) with some probability Note: flagging is not noisy.

• Consumers care about truth  $\rightarrow$  flagged article worthless to producer

#### Remark 3

Flagging removes bound on news quality

● Intuition: flagging ≈ private signal

#### Proposition 6

With  $w_0 = 1/2$ , flagging has stronger effects in monopoly than duopoly

- Competition → strategic considerations: competitor could be flagged!
- Reward harder anyways; readership smaller with competition
  - $\rightarrow$  if all false articles flagged, monopoly outperforms duopoly on quality

# Quality Certification

Move to a sequential game

- Can help: internalize effect investment on sharing
- Depends on *total* cost function

#### Remark 4

Observable news' quality imposes the same bounds on outlets' informativeness.

Intuition: influencers *always sharing* = best producer can achieve

### Subscription-Based Revenues

- Setup:
  - Each influencer pays a subscription t(x) in order to read news.
  - Producers' unique revenue: subscriptions.
- Comparing inefficiencies:
  - ▶ No possible welfare improvement marginal benefit = marginal cost.
  - No advertisement revenue loss of surplus.
- Feasibility (preliminary):
  - The ad-based monopoly outcome is reproducible with subscriptions.
- Open question: When can the gain in news quality compensate the loss of ad-revenues?

### Conclusion

#### How good can online news outlets get?

 $\rightarrow$  not so good... without intervention

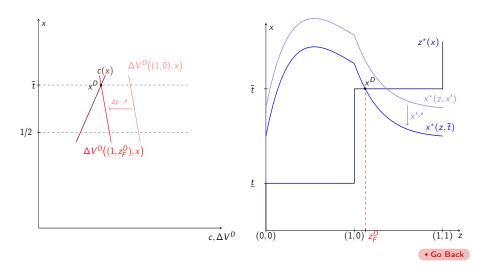
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### Bibliography

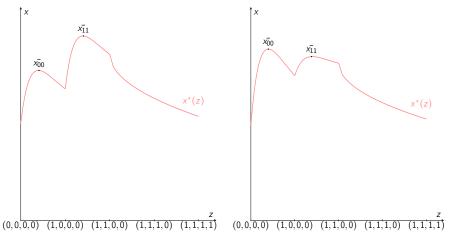
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### Equilibrium: With Competition



### Equilibrium: Monopolist Best-Response



#### Shape of Monopolist's Best Response

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# Attention-Seeking Influencers Objective

- Effect of competition between influencers, who compete for likes
  - ▶ Realistic → robustness
  - Trade-off between visibility and veracity
- Assumptions about likes:
  - Only followers (denoted f) can like posts
  - f can like i post only if f saw i's post
  - f only sees the post of **one** of his sharing neighbors **at random**
  - f likes a post iff receive a positive private signal (prior irrelevant)
- Decision rule<sup>1</sup>:

$$\mathbb{E}(\#\mathsf{likes}) \geq \tau$$

- Payoff depends on:
  - Whether news is true or false ( $\Rightarrow$  on *i*'s posterior)
  - How many other neighbors of f share ( $\Rightarrow$  on -i's sharing decision)

$${}^{1}\mathbb{E}(\# \text{likes}) = p_{\nu}(s_{i}; x_{\nu})\gamma \frac{1-b}{p_{\tau}} \left(1 - (1 - p_{\tau})^{d}\right) + (1 - p_{\nu}(s_{i}; x_{\nu}))(1 - \gamma) \frac{1-b}{p_{\tau}} \left(1 - (1 - p_{F})^{d}\right)$$

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#### Attention-Seeking Influencers Best-Response

**Disclaimer:** We focus only on symmetric strategies  $z_i = z \quad \forall i$ . We call "best-response"  $(z_T^*(x), z_F^*(x))$ , each maps  $x \to [0, 1]$  s.t.  $z^*(x, \mathbf{z}^*(\mathbf{x})) = z^*(x)$ 

#### Theorem 3

(i) For any 
$$\tau \leq \gamma \delta$$
,  $z_T^*(x; \tau) = z_F^*(x; \tau) = 1$  if and only if  $x \geq \hat{x}(\tau)$ .

(ii) For any 
$$\tau \ge (1 - \gamma)d(1 - b)$$
,  $z_{\mathcal{T}}^*(x; \tau) = z_{\mathcal{F}}^*(x; \tau) = 0$  if and only if  $x \le \mathfrak{x}(\tau)$ .

iii) For any 
$$\tau \in [\tau_1, \tau_2]$$
,  $z_T^*(x; \tau) = 1$ ,  $z_F^*(x; \tau) = 0$  if only if  $x \in [x_1(\tau), x_2(\tau)]$ .

Where:

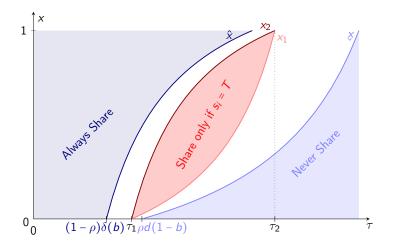
$$\delta(b) = \frac{1-b}{b} [1-(1-b)^d], \quad \tau_1(b) = \frac{1-b}{b} [1-(1-b(1-\gamma))^d], \quad \tau_2(b) = \frac{1-b}{b} [1-(1-b\gamma)^d]$$

And, given 
$$T = \frac{\frac{b\tau}{1-b} - 1 + (1-b(1-\gamma))^d}{(1-b(1-\gamma))^d - (1-b\gamma)^d}$$
,  
 $\hat{x}(\tau) = \frac{\gamma}{2\gamma - 1} \frac{\tau - (1-\gamma)\delta}{\tau}, \quad \dot{x}(\tau) = \frac{1-\gamma}{2\gamma - 1} \frac{\tau - (1-\gamma)d(1-b)}{d(1-b) - \tau}, \quad x_1(\tau) = \frac{(1-\gamma)T}{(1-\gamma)T + \gamma(1-T)}, \quad x_2(\tau) = \frac{\gamma T}{\gamma T + (1-\gamma)(1-T)}$ 

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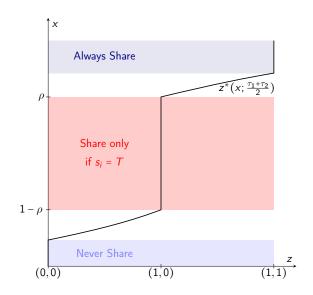
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#### Appendix



#### Pure strategy symmetric best-response

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Influencers' best-response for  $\tau = \frac{\tau_1 + \tau_2}{2}$ 

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