

Perverse Ethics

Dongkyu Chang¹ Allen Vong²

¹CityU of HK

²UM

Motivation

Social media platforms create diverse beliefs.

Diverse beliefs create conflicts.

Platforms are urged to “ethically” internalize the conflict costs.

This paper: a cautionary tale.

- By internalizing conflict costs, platforms may aggravate conflicts.

Roadmap

1. Model and equilibrium analysis
 - Baseline version: platform is self-interested.
 - Alternative version: platform is ethical.
2. Main result: Ethicality may aggravates conflict costs.
3. Regulations
4. Literature

Model

Setup

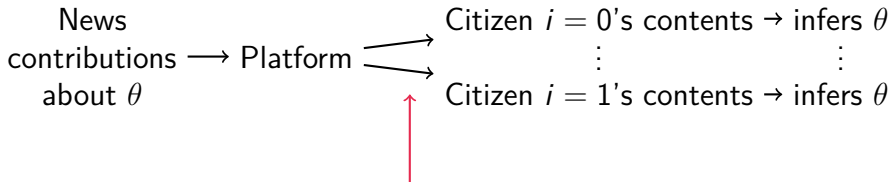
One-shot game.

Players:

- Platform.
- Rational citizens, $i \in [0, r]$ where $0 < r \leq 1$.
- Credulous citizens, $i \in (r, 1]$.

Hidden state $\theta \sim N(0, 1/p)$, where $p > 0$.

Overview



Depends on platform's upfront investment in two algorithms

One algorithm filters misinformation.

Another algorithm determines a personalized slant for each citizen.

Overview

Delta arrived in Israel. Some vaccinated people are infected.

- The state θ captures the change in Pfizer vaccine's effectiveness.

News contribution 1: (filtering: flagged on Facebook.)

“Vaccines are worthless. $> 80\%$ of the infected people are vaccinated.”

News contribution 2:

“Alarming news: half of the infected people are vaccinated.”

News contribution 3:

“Data shows that vaccine is 80% effective against infection.”

(Slanting towards negative news: omit #3 in citizen i 's content.)

Algorithms

The platform chooses a pair of algorithms (f, s) :

- Filter $f \in [0, \infty)$.
- Slant $s : [0, 1] \rightarrow \mathbf{R}$, where s_i is citizen i 's personalized slant.

The algorithms are hidden from the citizens.

But the (rational) citizens have rational expectation about (f, s)

The citizens take no actions.

Signals

Given (f, s) , each citizen i receives a private signal

$$y_i = \theta + s_i + \varepsilon_i$$

where $\varepsilon_i \sim N(0, \frac{1}{q+f})$ represents misinformation ($q > 0$: default precision).

Suppose that the citizens expect that the platform chooses (f^*, s^*) (possibly, $(f^*, s^*) \neq$ the platform's actual choice).

Each citizen i 's estimate of θ upon receiving y_i :

$$\hat{\theta}_i(y_i) = \begin{cases} \mathbf{E}^*[\theta|y_i] = \overbrace{\frac{q+f^*}{p+q+f^*}}^{\text{weight on signal}} \cdot \overbrace{(y_i - s_i^*)}^{\theta + \varepsilon_i + s_i - s_i^*} & \text{if } i \text{ is rational} \\ y_i & \text{if } i \text{ is credulous} \end{cases}$$

Platform's payoff

Revenue from rational citizens:

$$v_R(f, s; f^*, s^*) := \mathbf{E} \left[\int_0^r \overbrace{-\beta(\hat{\theta}_i(y_i) - b_i)^2 - \tau \mathbf{Var}_i^*[\theta|y_i]}^{\text{revenue from citizen } i\text{'s activities on the platform}} di \right],$$

citizen i 's estimate of θ $\in \mathbf{R}$; citizen i 's bias

Revenue from credulous citizens:

$$v_C(f, s) := \mathbf{E} \left[\int_r^1 -\beta(\hat{\theta}(y_i) - b_i)^2 di \right] = \mathbf{E} \left[\int_r^1 -\beta(y_i - b_i)^2 di \right]$$

Cost to develop the algorithms:

$$\frac{c}{2} f^2 + \frac{k}{2} \int_0^1 s_i^2 di.$$

Solution concept

Pure-strategy Bayesian Nash Equilibria, henceforth equilibria.

In equilibrium, users' expectations are correct.

(f^*, s^*) is an equilibrium if and only if

$$(f^*, s^*) \in \arg \max_{f, s} \left\{ \underbrace{v_R(f, s; f^*, s^*) + v_C(f, s)}_{=v(f, s; f^*, s^*)} - \frac{c}{2} f^2 - \frac{k}{2} \int_0^1 s_i^2 di \right\}$$

Equilibrium

Equilibrium

Proposition. *There exists an essentially unique equilibrium. In the equilibrium, the platform chooses $(f, s) = (f^S, s^S)$ where:*

1. *The filter f^S is positive and uniquely characterized by*

$$\frac{\beta r}{(p + q + f^S)^2} + \frac{\beta(1 - r)}{(q + f^S)^2} = cf^S$$

2. *For every citizen i , s_i^S is characterized by*

$$s_i^S = \begin{cases} \frac{2\beta}{k} \left(\frac{q+f^S}{p+q+f^S} \right) b_i & \text{if } i \in [0, r] \\ \frac{2\beta}{2\beta+k} b_i & \text{if } i \in (r, 1] \end{cases}$$

Proof sketch

Focus on the platform's profit from rational citizens.

Platform's incentives depend on the (rational) citizens' inferences:

Lemma. *Suppose that the rational citizens expect that the platform plays (f^*, s^*) . Then each rational citizen i 's posterior belief about the state θ , upon receiving signal y_i , is Gaussian:*

$$\theta|y_i \sim N\left(\underbrace{\frac{q + f^*}{p + q + f^*} (y_i - s_i^*)}_{=\hat{\theta}_i(y_i)}, \underbrace{\frac{1}{p + q + f^*}}_{=\text{Var}_i^*[\theta|y_i]}\right).$$

Proof sketch

Platform's (expected) revenue from the rational citizens:

$$\int_0^r \mathbf{E} \left[-\beta \left(\frac{q + f^*}{p + q + f^*} (y_i - s_i^*) - b_i \right)^2 \right] - \frac{\tau}{p + q + f^*} di$$

Thus, for each user i , platform wishes to

1. “minimize the spread” of $y_i = \theta + s_i + \epsilon_i$,
2. “pull” $y_i = \theta + s_i + \epsilon_i$ closer to bias b_i .

Proof sketch

From the platform's perspective:

$$\hat{\theta}_i(y_i) \sim N \left(\frac{q + f^*}{p + q + f^*} (s_i - s_i^*), \left(\frac{q + f^*}{p + q + f^*} \right)^2 \frac{p + q + f}{p(q + f)} \right).$$

To reduce spread, platform chooses higher f .

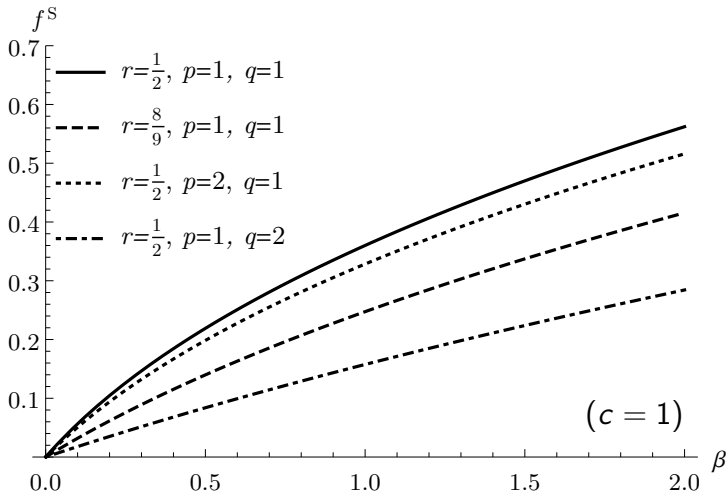
To pull $\hat{\theta}_i(y_i)$, platform chooses more extreme s_i .

In equilibrium, no incentive to reduce f or pull s_i further.

Finally, given higher p , users put less weight on signals for inference.

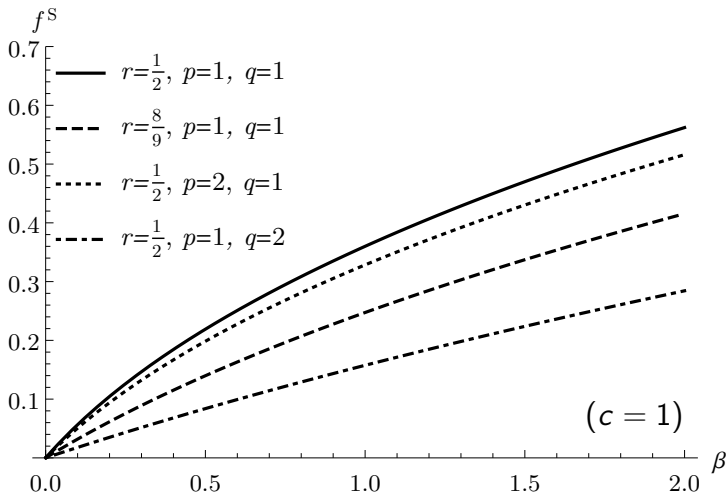
- Thus, platform has less incentives to filter.
- Platform's marginal return from slanting also decreases. □

Filter



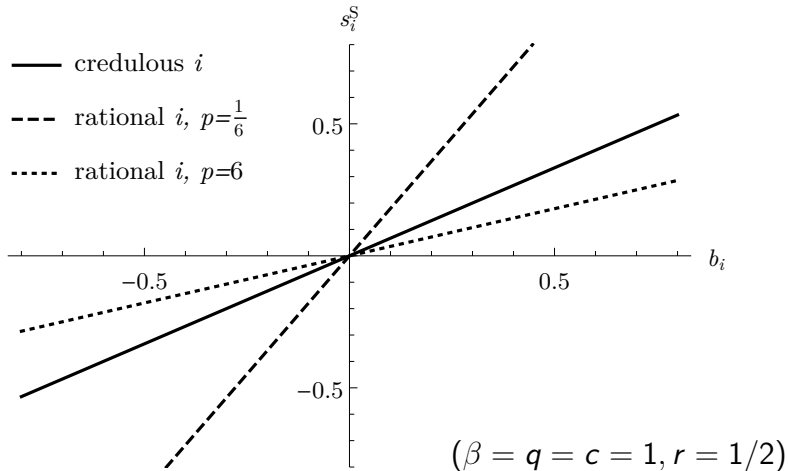
Filtering to cater to biases (β and r).

Filter



Higher p and q crowd out filtering incentives: $\frac{df^S}{dp}, \frac{df^S}{dq} \in (-1, 0) ..$

Slants



$d|s_i^S|/dp < 0$ and $\partial|s_i^S|/\partial f^S > 0$ for rational citizens.

Social Conflict and Ethics

Social Conflicts

The citizens' estimates of θ typically disagree.

A regulator wishes to minimize conflicts due to disagreements.

$$\kappa(f, s; f^*, s^*) := \mathbf{E} \left[\frac{1}{2} \int_0^1 \int_0^1 h(\hat{\theta}_j(y_j) - \hat{\theta}_i(y_i))^2 dj di \right]$$

Ethicality

An ethical platform's payoff is

$$v(f, s; f^*, s^*) - \kappa(f, s; f^*, s^*) - \frac{c}{2}f^2 - \int_0^1 \frac{k}{2}s_i^2 di.$$

The model is otherwise identical.

Equilibrium (with Ethical Platform)

Proposition. *There exists an essentially unique equilibrium. In the equilibrium, the platform chooses $(f, s) = (f^E, s^E)$ where:*

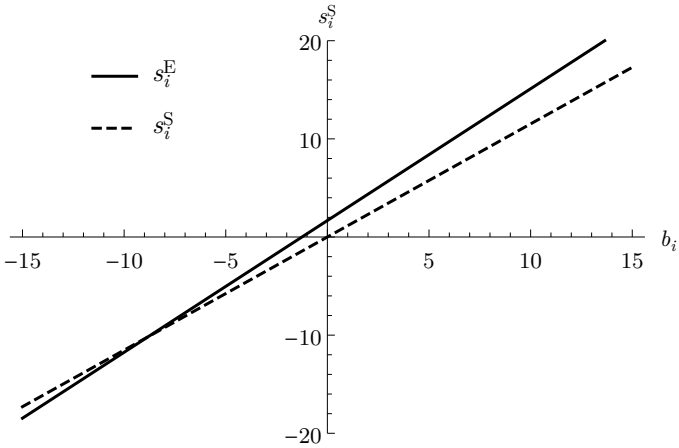
1. f^E strictly exceeds f^S and is uniquely characterized by

$$(\beta + h) \left[\frac{r}{(p + q + f^E)^2} + \frac{1 - r}{(q + f^E)^2} \right] = c'(f^E),$$

2. For almost every i , s_i^E is characterized by

$$s_i^E = \begin{cases} \frac{2\beta}{k} \left(\frac{q + f^E}{p + q + f^E} \right) \left(b_i + \frac{2h}{k + 2\beta + 2hr} \int_r^1 b_j dj \right) & \text{if } i \text{ is rational} \\ \frac{2\beta}{2\beta + k + 2h} \left(b_i + \frac{2h}{k + 2\beta + 2hr} \int_r^1 b_j dj \right) & \text{if } i \text{ is credulous} \end{cases}$$

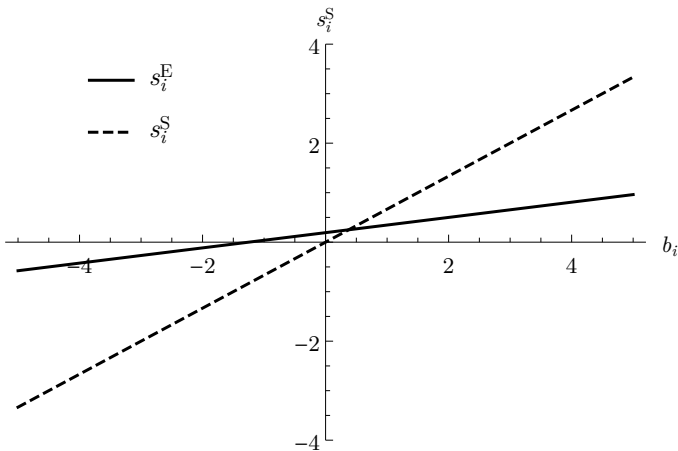
Slants



More personalized contents for rational citizens

$$(p = q = k = c = 1, r = .5, h = 5, \int_r^1 b_i di = 1)$$

Slants



Less personalized contents for credulous citizens

$$(p = q = k = c = 1, r = .5, h = 5, \int_r^1 b_i di = 1)$$

Proof sketch

Additional filtering incentives to reduce social conflict

$$\mathbf{E} [(\hat{\theta}_i(y_i) - \hat{\theta}_j(y_j))^2]$$
$$= \begin{cases} \mathbf{E} [(\mathbf{E}^*[\theta|y_j] - \mathbf{E}^*[\theta|y_i])^2] & \text{between rational } i \text{ and } j \\ \mathbf{E} [(y_j - y_i)^2] & \text{between credulous } i \text{ and } j \\ \mathbf{E} [(y_j - \mathbf{E}^*[\theta|y_i])^2] & \text{btw rational } i \text{ and credulous } j \end{cases}$$

Additional incentives to slant less for credulous citizens.

Given $f^E > f^S$, higher MR from slanting rational citizens' signals.

Perverse Ethics

Equilibrium Conflict Cost

Equilibrium conflict cost

$$\text{among rational citizens} : K_R(f, s) = \mathbf{E} \left[\frac{h}{2} \int_0^r \int_0^r (\hat{\theta}_j(y_j) - \hat{\theta}_i(y_i))^2 dj di \right]$$

$$\text{among credulous citizens} : K_C(f, s) = \mathbf{E} \left[\frac{h}{2} \int_r^1 \int_r^1 (\hat{\theta}_j(y_j) - \hat{\theta}_i(y_i))^2 dj di \right]$$

$$\text{between the two groups} : K_B(f, s) = \mathbf{E} \left[h \int_r^1 \int_1^r (\hat{\theta}_j(y_j) - \hat{\theta}_i(y_i))^2 dj di \right]$$

Proposition. *The following holds.*

1. $\exists \bar{p} > 0$ such that $K_R(f^S, s^S) < K_R(f^E, s^E)$ iff $p > \bar{p}$.
2. $K_C(f^S, s^S) > K_C(f^E, s^E)$ and $K_B(f^S, s^S) > K_B(f^E, s^E)$.

Main result

The aggregate conflict cost in equilibrium (f, s) :

$$K(f, s) := K_R(f, s) + K_B(f, s) + K_C(f, s) \quad (= \kappa(f, s; f, s))$$

Corollary (Perverse ethics).

There is $r \in (0, 1)$ such that the following holds for every $r \in [\bar{r}, 1)$:

$$\exists p' > 0 \quad \text{such that} \quad K(f^S, s^S) < K(f^E, s^E) \quad \text{whenever } p > p'.$$

“Unless the state is sufficiently uncertain, ethicality backfires.”

Proof sketch

For all rational citizens $i \in [0, r]$,

$$\hat{\theta}_i(y_i) = A(y_i - s_i^*) \quad \text{where} \quad A = \frac{q + f^*}{p + q + f^*}.$$

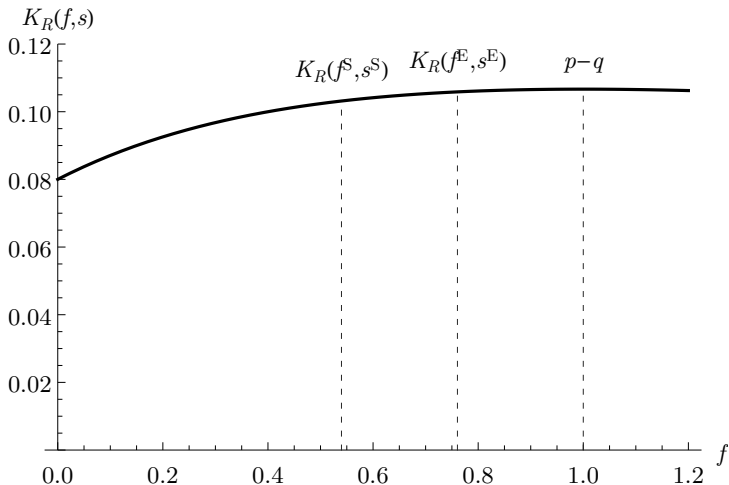
For now, suppose that the weight $A > 0$ is exogenously given.

$$K_R(f, s^*) = \frac{h}{2} \int_0^r \int_0^r A^2 \left[\frac{2}{q + f} \right] di dj$$

In equilibrium,

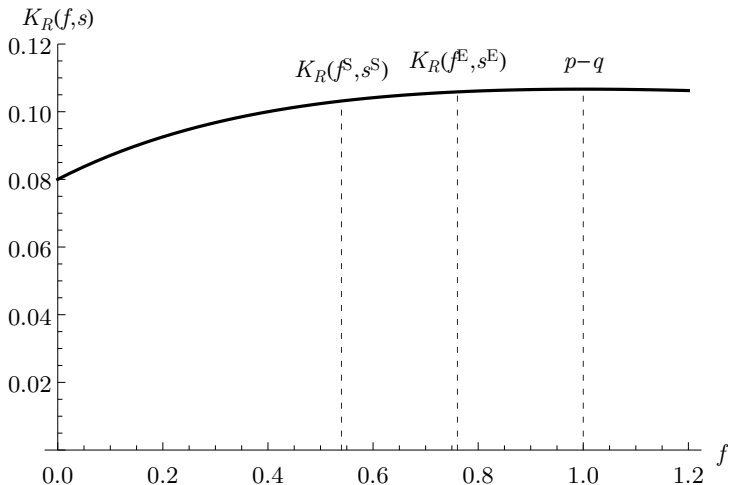
$$A = \frac{q + f^*}{p + q + f^*} = \frac{q + f}{p + q + f} \quad \text{increases in } f.$$

Proof sketch



$f \uparrow \implies \left\{ \begin{array}{l} \text{less misinformation} \\ \text{but more weight on } y_i \text{'s} \end{array} \right.$

Proof sketch



$p \uparrow \implies \left\{ \begin{array}{l} \text{smaller learning benefit by filtering} \\ \text{\& crowding out filtering incentives} \end{array} \right. \implies f^E \downarrow \quad f^S \downarrow .$

Regulations

Filtering Floor

Consider legislation that ensures $f \geq \underline{f} > 0$.

Let (f^L, s^L) denote the algorithms with the filtering floor \underline{f} . Then,

$$f^L = \begin{cases} f^S & \text{if } \underline{f} < f^S \\ \underline{f} & \text{if } \underline{f} \geq f^S. \end{cases}$$

where f^S denotes the status quo filtering level.

The legislation should be sufficiently aggressive to guarantee a success.

Proposition.

1. $K_C^L < K_C(f^S, s^S)$ and $K_B^L < K_B(f^S, s^S)$ whenever $\underline{f} > f^S$.
2. $K_R^L < K_R(f^S, s^S)$ whenever $\underline{f} > f^S \geq p - q$.
3. Suppose $f^S < p - q$. Then, there is $F > f^S$ such that

$$K_R^L < K_R(f^S, s^S) \quad \text{if and only if} \quad \underline{f} > F.$$

Arrest of Misinformation

Increase the default precision level from q to q^A .

The equilibrium filtering level increases from f^S to f^A .

$$f^A < f^S \text{ but } q^A + f^A > q + f^S.$$

Proposition.

1. $K_C^A < K_C(f^S, s^S)$ and $K_B^A < K_B(f^S, s^S)$.
2. $K_R^A < K_R(f^S, s^S)$ whenever $f^S \geq p - q$.
3. Suppose $f^S < p - q$, there is $Q > q$ such that

$$K_R^A < K_R(f^S, s^S) \text{ if and only if } q^A > Q.$$

Fairness Doctrine

Originally a regulation on radio and television news.

The platform should deliver all contrasting views on public issues.

Effectively, the platform cannot slant (i.e., $s^F = 0$).

Proposition. $K_R^F = K_R(f^S, s^S)$, $K_C^F < K_C(f^S, s^S)$, and $K_B^F < K_B(f^S, s^S)$.

Media Literacy Campaign

Consider a media literacy campaign that increases r to 1.

$\mathbf{E}[(\hat{\theta}_i(y_i) - \hat{\theta}_j(y_i))^2]$ is lower when both i and j are rational, compared to the case at least one of i and j is credulous.

Hence, the media literacy campaign decreases the conflict, *provided that the platform does not respond to the campaign.*

Proposition. *There is $\bar{\beta} \geq \underline{\beta} > 0$ such that*

- 1. $p - q < f^M < f^S$ and $K^M > K(f^S, s^S)$ whenever $\beta > \bar{\beta}$.*
- 2. $K^M < K(f^S, s^S)$ whenever $\beta < \underline{\beta}$.*

The regulator may combine media literacy campaign with a regulation on the supply side (e.g., filtering floor).

Transparency

Proposition (Transparency).

Suppose that the platform's algorithms are publicly observable. Then

$$K(f^S, s^S) \leq K(f^E, s^E).$$

With transparency,

- The citizens would infer the state from their personalized signals based on the platform's actual choice of the algorithms.
- The ethical platform correctly internalizes the conflict costs.
- MacCarthy (2020)

Literature

Literature

Ethical design in computer science

Wu (2017), Kearns & Roth (2019).

Here: cautionary tale against conventional wisdom about ethicality.

Traditional media

Mullainathan & Shleifer (2005),
Anderson, Waldfogel & Strömberg (2015), Perego & Yuksel (2021).

Here: individual signals enable conflicts; slanting given rational users.

Internet media

Peitz & Reisinger (2015), MacCarthy (2020).

Here: platforms' incentives to fight spams; implications of ethicality.

Measure for disagreement

Kartik, Lee, & Suen (2021), Zanardo (2017).