

Controlling Fake Reviews

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Introduction

- ▶ Rating systems play key roles in platform markets:
 - ▶ Hollenbeck et al (2019): ratings vs advertisement in hotel industry
 - ▶ Reimers and Waldfogel (2020): ratings vs professional reviews for books
- ▶ At the same time, the incentive to make fake reviews is growing.
- ▶ Effort to reduce fake reviews:
 - ▶ Amazon is strictly prohibiting incentivized reviews since 2016.
 - ▶ In 2019, FTC filed the first case challenging fake paid reviews:
Cure Encapsulations:
 - selling a weight loss pill (\$12.8 million in sales on Amazon)
 - paid AmazonVerifiedReview.com for fake reviews

Question: (How) Should a platform reduce fake reviews?

- ▶ Are fake reviews harmful?
 - ▶ Rational buyers might not be fooled by the fake reviews.
 - ▶ Costly fake reviews might work as a signal of good quality.
 - It might pay off only for high quality sellers through future sales.
(Nelson; 70,74)
- ▶ Instruments of the platform:
 1. intensity of censorship on fake reviews
 2. weights on previous reviews and new reviews,

Overview

- ▶ The number of fake reviews is increasing in quality, decreasing in reputation.
- ▶ The stringent censorship reduces
 - ▶ the number of fake reviews in expectation,
 - ▶ the effects of fake reviews.
- ▶ For rational consumers:
 - ▶ a rating with fake reviews can be more informative than one without fake reviews
 - ▶ transition speed of the rating should be slower than the optimal level without fake reviews.
- ▶ For credulous consumers:
 - ▶ the stringent censorship reduces bias for the credulous consumers as long as positive number of fake reviews are observed.

Literature

Design of Rating Systems

- ▶ **[certification]** Lizzeri (1999), Harbaugh and Rasmusen (2018), DeMarzo, Kremer, Skrzypacz (2019), Hopenhayn and Saeedi (2019), Hui et al (2018), Zapechelnyuk (2020)
- ▶ **[scoring][one-shot]** Ball (2019), **[dynamic]** Vellodi (2019): entry/exit, directed search; Horner and Lambert (2018), Bonatti and Cisternas (2020), **this paper: signal jamming**

	platform controls..		eqm action	credulous consumer
	weights	censorship		
HL2018	Y	(Y)	constant	N
BC2020	Y	N	Markov	N
This paper	Y	Y	Markov	Y

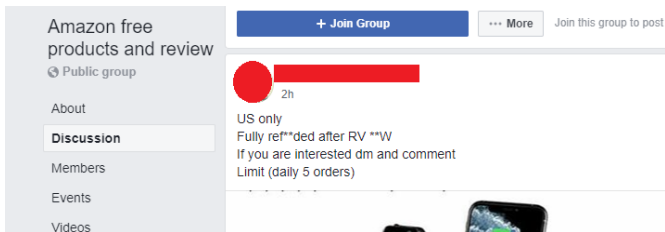
Promotion and Signaling (Q: The higher quality, the more promotion?)

- ▶ Nelson (1970, 1974), Kihlstrom and Riordan (1984), Milgrom and Roberts (1986), Horstmann and Moorthy (2003), Mayzlin (2006), Dellarocas (2006): **One shot promotion**
- ▶ Horstmann and MacDonald (1994), Saraiva (2020) [numerical/empirical], and **this paper: Repeated promotions; true quality and reputation** play different/interactive roles

Motivating example

Fake reviews with “verified purchase” on Amazon

1. Fake reviewers are **refunded** by the seller. Refunds are done outside of the platform.



- cf) official review programs on Amazon
 - *Early Reviewer Program*: Amazon offers \$1-3 for a review of a previously purchased product
 - *Vine Voice*: Reviews for free not-yet-released products (invitation only)
2. The platform takes a **transaction fee** from each transaction
 3. The platform can detect a part of fake reviews.

Model (1/3)

- ▶ Time: $t \in [0, \infty)$
- ▶ Players: a long lived seller, many short lived buyers
 - ▶ (a platform can control parameters before the game starts)
- ▶ Action at time t
 - ▶ Seller:
 - (sell one unit of the product: **fixed**)
 - choose the amount of the fake reviews: $F_t \in \mathbb{R}$
 - ▶ Buyers:
 - buy the product, or not
 - form the equilibrium price: $p_t = E[\theta_t | Y_t] \equiv M_t$ [▶ Details](#)
- ▶ State:
 - ▶ θ_t : seller's type (quality of the product) at t
 - ▶ Y_t : seller's rating at t
- ▶ Information:
 - ▶ Seller at time t : the whole history so far $= (\theta_s, Y_s, F_s, p_s)_{s \in [0, t]}$
 - ▶ Buyers at time t : current rating $= Y_t$

Model (2/3)

- State transition:
 - quality θ_t follows

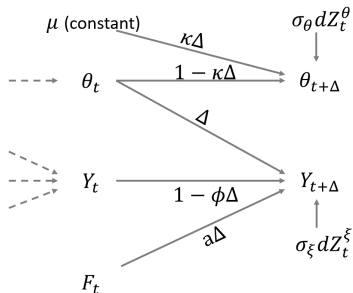
$$d\theta_t = -\kappa(\theta_t - \mu)dt + \sigma_\theta dZ_t^\theta$$

exogenous for players (seller/buyers) and for the platform

- rating Y_t follows

$$dY_t = -\phi Y_t dt + aF_t dt + \theta_t dt + \sigma_\xi dZ_t^\xi$$

$a > 0$: effectiveness of fake reviews. (low a = stringent censorship)



Model (3/3)

- Seller's instantaneous payoff:

$$\begin{aligned}\pi_t &= \underbrace{(1 - \tau) p_t (1 + F_t)}_{\text{revenue}} - \underbrace{p_t \cdot F_t}_{\text{costs of refund}} - \underbrace{\frac{c}{2} F_t^2}_{\text{additional costs}} \\ &= (1 - \tau) p_t - \tau p_t \cdot F_t - \frac{c}{2} F_t^2\end{aligned}$$

- τ : transaction fee imposed by the platform.
- The market determines $p_t = E[\theta_t | Y_t] \equiv M_t$

$$\pi_t = (1 - \tau) M_t - \tau M_t \cdot F_t - \frac{c}{2} F_t^2$$

- $\tau = 0$: a. la. Holmstrom (1999), a special case of Horner and Lambert (2018)

Definition of Equilibrium

Stationary Linear Markov equilibrium

Definition

A linear Markov strategy $F = (F_t)_{t \geq 0}$ s.t. $F_t = \alpha \theta_t + \beta Y_t + \gamma$ is a stationary linear Markov equilibrium if

1. $F = \arg \max_{(\tilde{F}_t)_{t \geq 0}} E_0 \left[\int_0^\infty e^{-tr} \left((1 - \tau) M_t - \tau M_t \cdot \tilde{F}_t - \frac{c}{2} \tilde{F}_t^2 \right) \right]$
2. $M_t = E^F [\theta_t | Y_t]$
3. $(\theta_t, Y_t)_{t \geq 0}$ induced by F is stationary Gaussian

Stationarity of Equilibrium

- ▶ Transition of (θ_t, Y_t) (in discrete analogue):

$$\theta_{t+dt} = \theta_t (1 - \kappa dt) + \mu \kappa dt + \sigma_\theta dZ_t^\theta$$

$$Y_{t+dt} = Y_t (1 - (\phi - a\beta) dt) + \theta_t (1 + a\alpha) dt + a\gamma dt + \sigma_\xi dZ_t^\xi$$

- ▶ (θ_t, Y_t) is stationary Gaussian if $\phi - a\beta > 0$
- ▶ When (θ_t, Y_t) is stationary Gaussian, then

$$M_t \equiv E[\theta_t | Y_t] = \underbrace{E[\theta_t]}_{\equiv \mu} + \underbrace{\frac{\text{Cov}(\theta_t, Y_t)}{\text{Var}(Y_t)}}_{\equiv \lambda} [Y_t - \underbrace{E[Y_t]}_{\equiv \bar{Y}}]$$

Characterize Equilibrium

- ▶ HJB equation:

$$\begin{aligned} rV(\theta, M) = & \sup_{F \in \mathbb{R}} (1 - \tau) M \cdot q - \tau M \cdot F - \frac{c}{2} F^2 \\ & - \kappa(\theta - \mu) V_\theta + \{-\phi Y_t + a\lambda F + \theta\} V_M \\ & + \frac{\sigma_\theta^2}{2} V_{\theta\theta} + \frac{\sigma_\xi^2}{2} V_{YY} \\ \text{s.t. } & M = \mu + \lambda[Y - \bar{Y}] \end{aligned}$$

- ▶ Note: θ appears in the transition of states
- ▶ The equilibrium is characterized by guess-and-verify of
 - ▶ $F = \alpha\theta + \beta Y + \gamma$ (linear strategy)
 - ▶ $V = v_0 + v_1\theta + v_2 Y + v_3\theta^2 + v_4 Y^2 + v_5 Y\theta$ (quadratic value function)
 - ▶ $\phi - a\beta > 0$ (stationarity)

Theorem (Existence and uniqueness)

There **exists** a stationary linear Markov equilibrium. In this equilibrium, $\alpha > 0$, $\beta < 0$, $\lambda > 0$. The equilibrium is **unique** and continuously differentiable in parameters if a loose condition in parameters holds.

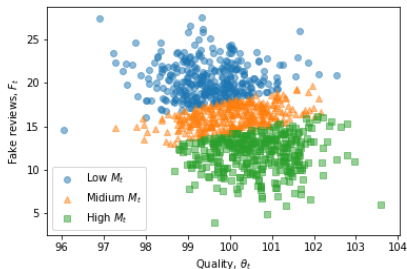
- ▶ Reminder: $F_t = \alpha\theta_t + \beta Y_t + \gamma$
- ▶ Uniqueness holds if $\phi > \kappa$ (rating evolves faster than underlying quality).
- ▶ $\beta < 0$:
 - ▶ Intuition: Higher rating, higher price, higher marginal cost of fake reviews, less fake reviews [▶ Details](#)
- ▶ $\alpha > 0$:
 - ▶ Note: $\alpha > 0$ is **not from** the incentive to increase **revenue**
 - Higher θ_t , higher p_{t+dt} even without fake reviews
 - $\frac{dp_{t+dt}}{dF_t}$ does not depend on θ_t
 - ▶ Higher quality, more cost-saving in the future
 - Once its rating is boosted, the future self will reduce the fake reviews.
 - This effect remains for a long time, given high θ_t . [▶ Details](#)

Consistency to data:

- ▶ $\beta < 0$ is consistent with Luca and Zervas (2016)
 - ▶ More manipulation after a drop of a rating

Implication to empirical literature:

- ▶ Hard to capture positive relationship b/w promotion level and quality.
 1. The rating should **not** be used as a proxy for quality
 2. Even with true quality index, researcher needs to control reputation.



Assume that the platform can change a and ϕ

► Recall:

- Rating: $Y_{t+dt} = Y_t \times (1 - \phi dt) + d\xi_t$ (higher ϕ , a higher weight on today's review & faster transition)
- New review: $d\xi_t = aF_t dt + \theta_t dt + \sigma_\xi dZ_t^\xi$ (smaller a , more stringent filtering)
- [In the paper, I am working on comparative statics about τ and σ_ξ]

Proposition

$E[F_t]$ is increasing in a . $E[F_t] \geq 0$ for sufficiently large a .

Proposition (The effects of fake reviews)

$a \cdot \alpha$, $a \cdot \beta$, $a \cdot \gamma$ goes to zero as $a \rightarrow 0$.

- ▶ Reminder: $aF_t = a\alpha\theta_t + a\beta Y_t + a\gamma$ = the effect of fake reviews
- ▶ Stringent censorship can **reduce** the **expected amount** and the **effects** of fake reviews.
- ▶ Note: $(\alpha, \beta, \gamma) \rightarrow 0$ even when $E[F_t] \rightarrow 0$ or $(a\alpha, a\beta, a\gamma) \rightarrow 0$

Q: **Should** the platform reduce fake reviews?

Criteria: $\rho^2 = \frac{\text{Cov}(\theta_t, Y_t)^2}{\text{Var}(\theta_t)\text{Var}(Y_t)}$

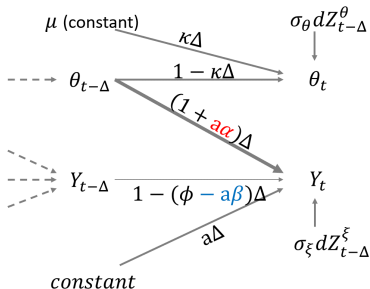
- ▶ Motivation:
- ▶ Regulators often want to make rating systems informative.
- ▶ For the platform, if the rating system is not informative, the sellers and buyers might move out to other platforms.
 - ▶ Maximization of ρ^2 is equivalent to minimizing $\text{Var}(\theta|Y)$

$$\text{Var}(\theta|Y) = \underbrace{\text{Var}(\theta)}_{\text{exogenous}} (1 - \rho^2)$$

- ▶ Note: M_t is an unbiased estimate of θ_t ($E[E[\theta_t|Y_t]] = E[\theta_t]$)

Q: Should the platform reduce fake reviews?

Criteria: $\rho^2 = \frac{(\phi - a\beta)}{(\kappa + \phi - a\beta)} \frac{(a\alpha + 1)^2}{((a\alpha + 1)^2 + \kappa(\sigma_\xi/\sigma_\theta)^2(\kappa + \phi - a\beta))}$ (given any α, β, δ)



► Impacts of the fake reviews:

1. $a \cdot \alpha > 0$ enhances the positive relationship between the true quality θ_t and the rating Y_t .
2. $-a\beta > 0$ increases the transition speed of the rating Y_t .

Proposition (Informativeness of fake reviews)

*The **equilibrium** strategy is **more informative** than **no-fake** strategy under a set of parameters such that*

1. *a is sufficiently large, or*
2. *(i) a is sufficiently small and*
(ii) $\phi^2 < \frac{\sigma_\theta^2}{\sigma_\xi^2} + \kappa^2$

Intuition:

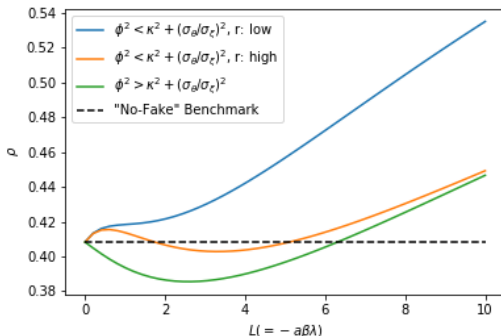
1. Higher a , higher $a \cdot \alpha$, more informative.
 2. $a\beta < 0$ makes the reputation transition speed $\phi - a\beta$ faster.
 - ▶ This is good if ϕ was too small, and
 - ▶ bad if ϕ was already sufficiently high.
- ▶ First effect dominates for large a , and second effect dominate for small a .

Sketch of the proof:

- $\rho^2 = \frac{(\phi+L)}{(\kappa+\phi+L)} \frac{(A(L)+1)^2}{((A(L)+1)^2 + \kappa(\sigma_\xi/\sigma_\theta)^2(\kappa+\phi+L))} = \rho^2(L, \phi)$ (given eqm α, β, δ)
 - L (eqm effect on the transition speed) is positive and increasing in a .

1. $\lim_{L \rightarrow \infty} \rho^2 = 1$

2. $\frac{\partial \rho^2}{\partial L}|_{L=0} > 0$ iff $\phi^2 < \frac{\sigma_\theta^2}{\sigma_\xi^2} + \kappa^2$



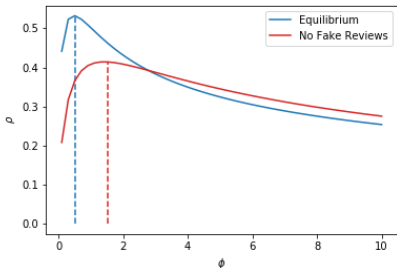
Q: How does the optimal ϕ (transition speed of the rating) change due to the fake reviews?

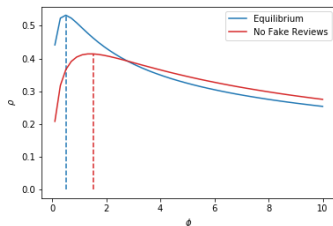
- ▶ Comparison with the optimal ϕ without fake reviews, ϕ^0
- ▶ $\phi^0 = \arg \max_{\phi} \rho^2(0, \phi)$
- ▶ $\phi^* = \arg \max_{\phi} \rho^2(L(\phi), \phi)$

Proposition

$\rho^2(L(\phi), \phi)$ is decreasing in ϕ at $\phi = \phi^0$.

Furthermore, $\rho^2(0, \phi^0) < \rho^2(L(\phi^*), \phi^*)$ for sufficiently small r .





Intuition

- ▶ $\rho^2(L(\phi), \phi)$ is decreasing in ϕ at $\phi = \phi^0$.
 - ▶ w.o./ fake reviews: the platform can control the transition speed ϕ
 - ▶ w/ fake reviews: effective transition speed is $\phi + L$
 - ▶ \rightarrow the platform should adjust ϕ downward.
- ▶ $\rho^2(0, \phi^0) < \rho^2(L(\phi^L), \phi^L)$ for sufficiently small r
 - ▶ small $r \Rightarrow$ high weight on the future \Rightarrow high $\alpha > 0 \Rightarrow$ informative (ie, $\rho^2(0, \phi^0) < \rho^2(L(\phi^0), \phi^0)$ given small r)
 - ▶ By definition of ϕ^* , $\rho^2(L(\phi^0), \phi^0) < \rho^2(L(\phi^*), \phi^*)$

Credulous Consumers

- ▶ Credulous consumers believe that
 - ▶ they face a stationary Gaussian distribution of (θ_t, Y_t)
 - ▶ there is no fake reviews by the seller. (ie, assume $\alpha = \beta = \gamma = 0$)
- ▶ Reputation:
 - ▶ rational consumers: $M_t = E^F[\theta_t | Y_t] = \mu + \lambda(\alpha, \beta)[Y_t - \bar{Y}(\alpha, \beta, \gamma)]$
 - ▶ credulous consumers: $\tilde{M}_t = \tilde{E}[\theta_t | Y_t] = \mu + \lambda(0, 0)[Y_t - \bar{Y}(0, 0, 0)]$
 - belief based an wrong joint distribution of (θ_t, Y_t)
- ▶ Seller's payoff:
 - ▶ $\pi_t = (1 - \tau)p_t - \tau p_t \cdot F_t - \frac{\epsilon}{2} F_t^2$
 - ▶ $p_t = \eta M_t + (1 - \eta) \tilde{M}_t$ where $\eta \in [0, 1]$
 - ▶ Interpretation:
 - η captures the rationality of all consumers
 - η is the ratio of rational consumers in the market.

▶ Details

Theorem

Existence and uniqueness given the same condition as the baseline model

Proposition

Existence of the credulous consumers decreases $E[F_t]$.

► Intuition:

- Credulous consumers are less sensitive to the rating than rational consumers.
 - Rational consumers regard the rating informative because of $a\alpha > 0$.
- Less marginal benefit with credulous consumers.
- Less fake reviews with credulous consumers.

Criteria for the credulous consumers:

$$\text{Bias} = E \left[\tilde{E} [\theta_t | Y_t] - \theta_t \right]$$

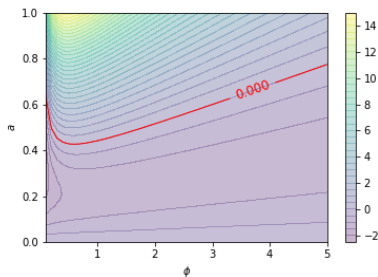
Lemma

$\text{Bias} \geq 0$ iff $E[F_t] \geq 0$.

Suppose there are only credulous consumers in the market.

Proposition

Stringent censorship policy reduces Bias as long as $E[F_t] \geq 0$.



Summary

Positive Analysis:

- ▶ The number of fake reviews is increasing in quality, decreasing in reputation.
- ▶ The stringent censorship
 - ▶ reduces fake reviews in expectation, but
 - ▶ reduces the effects of fake reviews.

Normative Analysis:

- ▶ For rational consumers:
 - ▶ a rating with fake reviews can be more informative than without fake reviews
 - ▶ Transition speed of the rating should be slower than the optimal level without fake reviews.
- ▶ For credulous consumers:
 - ▶ As long as $E[F_t] \geq 0$, the more stringent censorship, the less bias for the credulous consumers.

Intuition of the Equilibrium Strategy

► Back to Theorem

- Reminder: $V = v_0 + v_1\theta + v_2Y + v_3\theta^2 + v_4Y^2 + v_5\theta Y$
- FOC: $F_t = -\frac{\tau}{c}M_t + \frac{a}{c}\{v_2 + 2Y_tv_4 + \theta v_5\}$
- $\beta < 0$
 - $\beta = -\frac{\tau}{c}\lambda + 2\frac{a}{c}v_4 = -\frac{\tau}{c}\lambda + \frac{a}{c}\frac{-\beta\lambda\tau}{(-a\beta+r+2\phi)}$
 - $\beta < 0$ since today's cost saving incentive dominates.
- $v_4 = \frac{-\beta\lambda\tau}{2(-a\beta\lambda+r+2\phi)} > 0$
 - Higher reputation, less promotion, less costly.
- $\alpha > 0$
 - $\alpha = \frac{a}{c}v_5$
 - $v_5 = \frac{1}{\kappa+r+\phi}\left\{ \underbrace{2(a\alpha + bq)}_{\text{feedback to future } Y_t} v_4 - \alpha\lambda\tau \right\}$
 - **Driving Force:** Higher quality, higher reputation in the future, cost reduction in the future.
 - **Counteracting effect:** Higher quality, more promotion today/in the near future (if $\alpha > 0$)

Microfoundation of the price: $p_t = M_t$

► Back to Model

- (Reminder: $M_t \equiv E[\theta_t | Y_t]$)
- Suppose there is a mass (2) of buyers.
- Consumer $i \in [0, 2]$ feels

$$u_i = \begin{cases} \theta + \epsilon_i - p & \text{if the consumer purchase the product} \\ 0 & \text{otherwise} \end{cases}$$

- $\epsilon_i \sim i.i.d. F(\cdot)$ where $F(\cdot)$ is symmetric around zero
- Given Y , rational consumer purchases iff $M + \epsilon_i - p \geq 0$
- Market clearing

$$\begin{aligned} 1 &= 2q \cdot (1 - F(p - M)) \\ \Leftrightarrow p &= M \end{aligned}$$

Mixture of the rational/credulous consumers

► Back to Model

- $M = E[\theta|Y]$: rational consumer's belief (on the seller's quality)
- $\tilde{M} = \tilde{E}[\theta|Y]$: credulous consumer's belief (on the seller's quality)

Rationale:

- 2η rational consumers and $2(1 - \eta)$ credulous consumers in mkt
- Consumer $i \in [0, 2]$ feels

$$u_i = \begin{cases} \theta + \epsilon_i - p & \text{if the consumer purchase the product} \\ 0 & \text{otherwise} \end{cases}$$

- $\epsilon_i \sim U(-C, C)$: iid over the consumer types.
- Rational consumer purchases iff $M_t + \epsilon_i - p \geq 0$
- Credulous consumer purchases iff $\tilde{M}_t + \epsilon_i - p \geq 0$
- Market clearing

$$\begin{aligned} 1 &= 2\eta \cdot (1 - F(p - M)) + 2(1 - \eta) \cdot (1 - F(p - \tilde{M})) \\ \Leftrightarrow p &= \eta M + (1 - \eta) \tilde{M} \end{aligned}$$