



The Impact of Social Nudges on User-Generated Content for Social Network Platforms

Renyu (Philip) Zhang

(Joint works with Zhiyu Zeng, Hengchen Dai, Dennis J. Zhang, Heng Zhang, Max Shen)





- Introduction and Contributions
- Experiments and Empirics
- Social Network Model and Global Effect





Nudge: Intervention without Money or Restriction



- Nudge: Intervention that intends to change individual behaviors without altering financial incentives or imposing restrictions (Thaler and Sunstein 2009).
- Nudge is a central concept in behavioral science and very widely applied in practice:







•

Key Research (and Business) Questions



- Literature: Peer recognition NOT accompanied by financial incentives may backfire.
- How does the underlying social network impact social nudges?
- How to quantify the global effect of social nudges on the entire social network, considering nudge diffusion?

Combined Analytics Framework: Field Experiment (A/B Testing) + Social Network Modeling





Related Literature



- Content production on UGC platforms: Peer recognition NOT accompanied by financial incentives may backfire.
 - Burtch et al. (2018), Huang et al. (2019), Restivo and Van de Rijt (2014), Gallus et al. (2020), etc.
- Peer effects on social networks: People's behaviors are largely influenced by those of their neighbors on the network.
 - Jackson (2010), Ballester et al. (2006), Candogan et al. (2012), Bimpkis et al. (2016), Zhou and Chen (2016), Gelper et al. (2021), etc.
- Information-based interventions: Offering the right information to the right people improves the customer engagement and service capacity.
 - Buell and Norton (2011), Buell et al. (2017), Parker et al. (2016), Cui et al. (2019), Xu et al. (2021), etc.
- Platform operations: Optimizing operational strategies for online platforms.
 - Banerjee et al. (2016), Cachon et al. (2017), Bai et al. (2019), Kabra et al. (2020), etc.



Highlight of Main Contributions



- Social nudge: A Novel Intervention to Boost UGC Production
 - Direct effect: Receiving social nudges boosts the production of the provider by +13.21%.
 - Low-cost, information-based and flexible approach leveraging the embedded social network to address a central platform operations problem: Increasing UGC production.

• Impact of the Social Network on Social Nudges

- Indirect effect (diffusion): Nudge receivers send +15.57% more social nudges to providers.
- Both direct and indirect effects of social nudges are stronger with a stronger tie and last for a few days.
- Global Effect of Social Nudges under Interferences
 - A social network model captures the direct and indirect effects of social nudges.
 - Bonacich Centrality for Edges helps quantify the global effect of social nudges (~+1% of total production).







- Introduction and Contributions
- Experiments and Empirics
- Social Network Model and Global Effect



Field Setting



- A Chinese online short-video sharing social network platform (referred to as Platform O hereafter).
- 300 million of daily-active users (DAU), half-billion MAU, 20 million USD advertising revenue per day.
- Each user can have two roles: Content provider and content viewer.
- Each user can follow other users: An embedded social network on Platform O.

Typical User-Interfaces of a Short-Video Sharing Platform











- Main experiment:
 - Sept. 12, 2018 ~ Sept. 14, 2018
 - Sample size: 993,676 providers (496,976 treatment and 496,700 control)
 - They had never received any social nudges before the experiment.
 - Most (94%) received only one social nudge during this experiment.
- Replication experiment (for robustness check):
 - Sept 14, 2018 ~ Sept. 20, 2018.
 - 655,001 providers non-overlapping with the main experiment.
- Data from both experiments pass the randomization checks, i.e., treatment and control providers are comparable.
- Outcome variables:
 - Number of videos uploaded.
 - Number of social nudges sent to other providers by the recipient.

Question: How do social nudges impact UGC production on a social network platform?



Social Nudges Boost UGC Production



- We focus on the day a provider was sent the first social nudge in the experiment (the first reception day).
- Model specifications (a two-way tie means the nudge receiver also follows the sender):

Number of Videos Uploaded_i = $\beta_0 + \beta_1 Treatment_i + \epsilon_i$

 $Number \ of \ Videos \ Uploaded_i = \beta_0 + \beta_1 Treatment_i + \beta_2 Two_Way_Tie_i + \alpha_3 Treatment_i * Two_Way_Tie_i + \epsilon_i$

• Direct effect on production:

Social Nudge

Treatment Effect	Average	2-way Tie	1-way Tie
Relative Effective Size	+13.21%***	+17.40%***	+9.35%***

Note: All p-values are smaller than 0.001.

- Social nudges boost UGC production.
- The direct effect of social nudges is stronger with a stronger tie.
- (i) Video views, likes, and comments 1; (ii) video quality (like-rate and comment-rate) ; (iii) more effective than platform-initiated nudges; (iv) production boost not driven by the additional likes and comments.
- Literature: Peer recognition without financial incentives may backfire.



Social Nudges Diffuse



- How does the underlying social network impact the effect of social nudges?
- Model specifications:

Social Nudge

Number of Social Nudges $Sent_i = \beta_0 + \beta_1 Treatment_i + \epsilon_i$

 $Number \ of \ Social \ Nudges \ Sent_i = \beta_0 + \beta_1 Treatment_i + \beta_2 Two_Way_Tie_i + \alpha_3 Treatment_i * Two_Way_Tie_i + \epsilon_i$

• Indirect effect of social nudges:

Treatment Effect	Average	2-way Tie	1-way Tie
Relative Effective Size	+15.57%***	+30.02%***	+2.87%*

Note: ***: p-value<0.001; *: p-value<0.05.

- Social nudges diffuse on the social network platform.
 - The social nudge diffusion is stronger with a stronger tie.





- How to evaluate the global effect of social nudges on the entire social network?
 - The global effect is NOT directly measurable by user-side randomized experiments due to the interference from nudge diffusion.





- Introduction and Contributions
- Experiments and Empirics
- Social Network Model and Global Effect



Social Network Model



- A social network G = (V, E), where V is the set of users and E is the set of following relationships.
- Discrete time period: $t=1,2,3,\cdots$ (t=1 is the day when the social nudge function was launched).
- μ_e : the number of social nudges e_o sends to e_d without receiving any nudge (called the organic nudges).
- p_e : the production boost of e_d after receiving a social nudge sent by e_o , i.e., the direct effect of social nudge.
- $d_{e\ell}$: the increase in the number of nudges sent on ℓ after e_d receives a nudge sent by e_o ($e_d = \ell_o$), i.e., the indirect effect.



• The production boost of user i in period t (direct effect):

$$x_i(t) = \sum_{s=1}^{t} \alpha_p^{t-s} \sum_{\ell_d=i} p_\ell y_\ell(s) + \epsilon_i^x(t)$$

• The number of social nudges on edge e in period t (indirect effect):

$$y_e(t) = \mu_e + \sum_{s=1}^t \alpha_d^{t-s} \sum_{\ell_d = e_o} d_{\ell e} y_\ell(s) + \epsilon_e^y(t)$$

Direct and indirect effects decay to 0 exponentially as $p_e \alpha_p^{\Delta t}$ and $d_{e\ell} \alpha_d^{\Delta t}$.

$$\mathbf{D} := (d_{\ell e} : (\ell, e) \in E^2) \quad \mu := (\mu_e : e \in E) \quad \eta := (p_e/(1 - \alpha_p) : e \in E)$$





Bonacich Centrality for Edges and the Equilibrium



• Equilibrium nudge vector $\mathbf{y}^* := \lim_{t\uparrow+\infty} \mathbb{E}[\mathbf{y}(t)]$ and production boost $\mathbf{x}^* := \lim_{t\uparrow+\infty} \sum_{i\in V} \mathbb{E}[x_i(t)]$.

Theorem. The equilibrium nudge vector is $\mathbf{y}^* = \mathcal{BE}(\mathbf{D}, \mu)$ and the equilibrium production boost is $\mathbf{x}^* = \eta^T \mathbf{y}^*$.

• Bonacich Centrality for Edges (BCE) (I is the identity matrix, assuming the limit exists):

$$\mathcal{BE}(\mathbf{D},\mu) = \left(\mathbf{I} - \frac{1}{1-\alpha_d}\mathbf{D}\right)^{-1} \mu = \left(\mathbf{I} + \sum_{k=1}^{+\infty} \frac{1}{(1-\alpha_d)^k}\mathbf{D}^k\right) \mu$$

Direct Effect Indirect Effect

- The nudge vector of the kth-order diffusion: $\frac{1}{(1-\alpha_d)^k} \mathbf{D}^k \mu$; the production boost from kth-order diffusion: $\frac{1}{(1-\alpha_d)^k} \eta^T \mathbf{D}^k \mu$
- Core challenge: Dimension of the diffusion matrix **D** for Platform O is 10³².



Approximation for the Global Effect

- Dimension of the diffusion matrix **D** for Platform O: 10^{32} . ٠
 - First approximation: Consider the first-order diffusion only and ignore the higher-order diffusions (~ -0.72% gap for Platform O).

$$\widetilde{\mathbf{y}}(1) = \widetilde{\mathcal{BE}}(\mathbf{D}, \mu, 1) = \left(\mathbf{I} + \frac{1}{1 - \alpha_d}\mathbf{D}\right)\mu, \ \widetilde{\mathbf{x}}(1) = \eta^T \widetilde{\mathbf{y}}(1)$$

First-order Diffusion

Second approximation: Randomly down-sample a subset of providers \tilde{V} , compute $\tilde{x}(1, \tilde{V}) := \sum_{i \in \tilde{V}} \left(\sum_{e \in E, e_d = i} \left(\frac{\mu_e p_e}{1 - \alpha_p} + \sum_{\ell \in E, e_d = i} \frac{\mu_e d_{e\ell} p_\ell}{(1 - \alpha_p)(1 - \alpha_d)} + \sum_{\ell \in E, e_d = i} \frac{\mu_e d_{e\ell} p_\ell}{(1 - \alpha_p)(1 - \alpha_d)} \right)$ ٠



Direct Effect

First-order Diffusion



Proposition. $\kappa \cdot \tilde{x}(1, \tilde{V})$ is an unbiased estimate of $\tilde{\mathbf{x}}(1)$, where $\kappa := |V|/|\tilde{V}|$ is the scaling factor.



Parameter Estimation with Experimental Data



- To quantify the global effect of social nudges, it suffices to estimate $(\mathbf{D}, \mathbf{p}, \mu, \alpha_p, \alpha_d)$ with data from the experiments.
- μ_e (the likelihood of sending organic nudges): Sampling edges in E whose origin was in the control group and cross-validate a logistic regression model.
- Jointly estimate (p_e, α_p) by minimize the mean-squared-error between the true over-time direct effect (from the experiment) and the one predicted by our social network model $p_e \alpha_p^t$ for t = 0,1,2.
- Jointly estimate $(d_{e\ell}, \alpha_d)$ by minimize the mean-squared-error between the true over-time indirect effect (from the experiment) and the one predicted by our social network model $d_{e\ell}\alpha_d^t$ for t = 0,1.
- The estimates can be replicated by the data from the second experiment (relative gap <10% for each parameter), suggesting the robustness and accuracy of our approach.

The estimated parameter values are the inputs of the social network model to evaluate the global effect of social nudges, $\kappa \cdot \tilde{x}(1, \tilde{V})$.



Estimation Results of the Global Effect



- Baseline estimation: Global Effect = +48.65 videos per day
 - Number of providers who received any social nudges per day multiplies the production boost of receiving social nudges (estimated from the experiment).

Estimation with Our Social Network Model				
Global Effect = $+140.67$ videos per day ($\sim +1\%$ of the total UGC production of Platform O)				
Accumulated Direct Effect = +130.08 videos per day	Accumulated Indirect Effect (diffusion) = $+10.59$ videos per day			
167% higher than the baseline estimation	Amounting to 22% of the baseline estimation and 8.14% of the direct effect			

Note: The absolute global effect is scaled to protect sensitive data.

- The estimation of global effect is robust and accurate, which can be replicated by the second experiment.
 - The ratio between the indirect effect and the direct effect is 8.38%.

Insight: Our social network model helps de-bias the substantial underestimates of the natural baseline estimation approach.

と 海 短 約 大 学 NYU SHANGHAI	Takeaways	UHIK SINESS SCHOOL
• Deve lever •	elop the social nudge as a novel low-cost, information-based, and flexible intervention that rages the embedded social network to increase content production on a large-scale UGC platforn Social nudges boost UGC production and diffuse on the social network.	n.
• Socia interf	I network model: Evaluate the global effect of social nudges on a social network with the ferences from nudge diffusion.	
• The operation	combined framework of A/B testing and social network model has the potential to study other ations problems for large-scale social network platforms.	

23





Thank You!

Questions?

renyu.zhang@nyu.edu and philipzhang@cuhk.edu.hk

https://rphilipzhang.github.io/rphilipzhang/index.html