

Intertwined Viral Marketing in Social Networks

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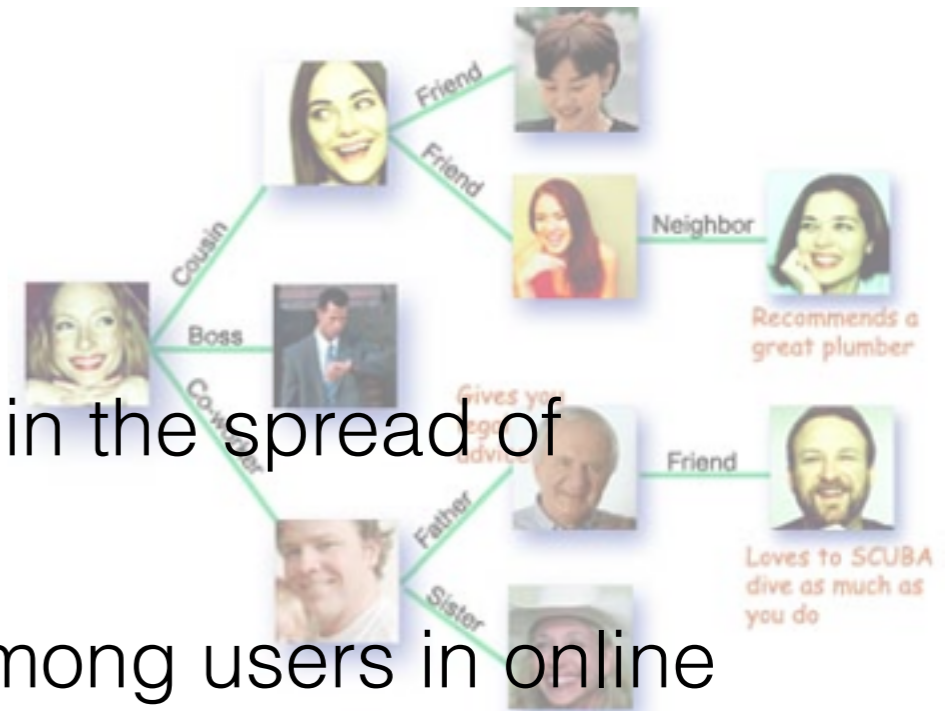
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Outline

- Background Knowledge Introduction
- Intertwined Information Diffusion Model
- Problem Formulation of TIM
 - C-TIM vs J-TIM
- Proposed Methods: TIER
 - C-TIER vs. J-TIER
- Experimental Results
- Summary

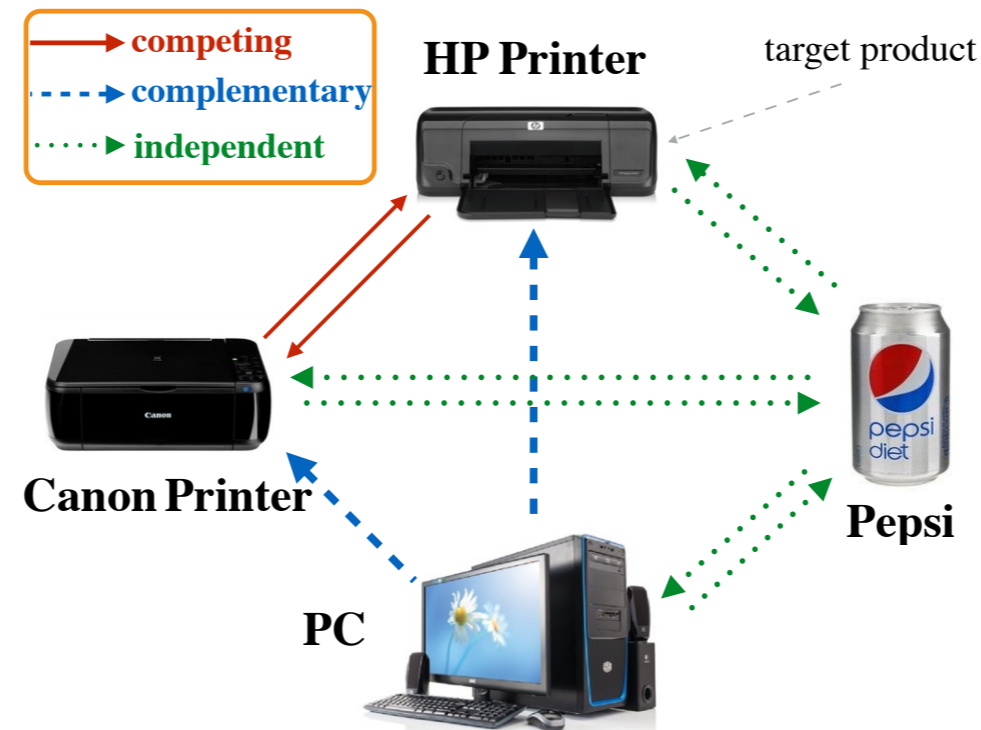
Viral Marketing Problem



- Social networks play a fundamental role in the spread of information among the users.
- To model how information propagates among users in online social networks, several information diffusion modes have been proposed:
 - IC model, LT model, SIR model, etc.
- Viral marketing problem
 - **Given:** Advertising budget, and information diffusion model in online social network
 - **Objective:** Achieve the maximum influence in the social network
 - **Problem:** Which set of users should be targeted at initially?
 - **Application:** commercial promotion, election campaign

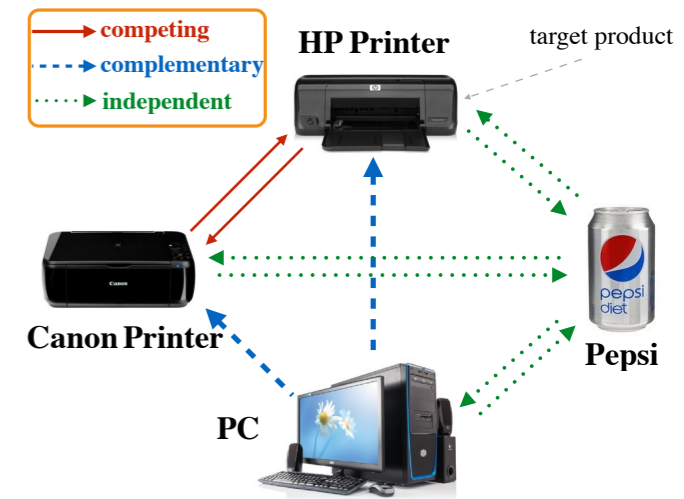
Intertwined Viral Marketing Problem

- *Observation*: Multiple products are being promoted in the social network at the same time.



- Inter**T**wined **I**nfluence **M**aximization Problem (**TIM**)
 - *Given*: target product, advertising budget, *information diffusion model in the network*, product relationships;
 - *Objective*: Achieve the maximum influence for the target product;
 - *Problem*: Identify the optimal initial seed user for the target product.

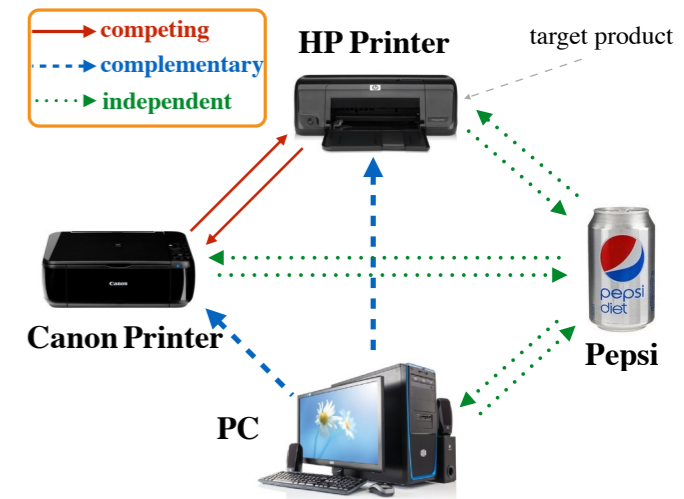
Intertwined Product Relationships



- The product relationships are intertwined:
 - *Competing*: Canon Printer \longleftrightarrow HP Printer;
 - Individuals who have Canon printer will be less likely to buy HP printer, and vice versa.
 - The purchase of Canon printer will **decrease** users' chance to buy HP printer.
 - *Complementary*: PC \longrightarrow Canon Printer, PC \longrightarrow HP Printer;
 - Individuals who have PC are more likely to buy a Canon printer or HP printer.
 - The purchase of PC will **increase** users' chance to buy printers.
 - *Independent*: PC \longleftrightarrow Diet Pepsi, Printer \longleftrightarrow Diet Pepsi.
 - The likelihood for individuals to buy PC/Printers has nothing to do with the purchase on Diet Pepsi.
 - The purchase of PC/Printers **doesn't affect** users' chance to buy Diet Pepsi.

Intertwined Product Relationships

- Intertwined Product Relation Definition



Definition (Independent, Competing and Complementary Products): Let $P(s_i^j = 1)$ (or $P(s_i^j)$ for simplicity) denote the probability that u_i is activated by product p^j and $P(s_i^j | s_i^k)$ be the conditional probability given that u_i has been activated by p^k already. For products $p^j, p^k \in \mathcal{P}$, the promotion of p^k is defined to be (1) *independent* to that of p^j if $\forall u_i \in \mathcal{V}, P(s_i^j | s_i^k) = P(s_i^j)$, (2) *competing* to that of p^j if $\forall u_i \in \mathcal{V}, P(s_i^j | s_i^k) < P(s_i^j)$, and (3) *complementary* to that of p^j if $\forall u_i \in \mathcal{V}, P(s_i^j | s_i^k) > P(s_i^j)$.

Definition (Threshold Updating Coefficient): Term $\phi_i^{l \rightarrow j} = \frac{P(s_i^j)}{P(s_i^j | s_i^l)}$ is formally defined as the “*threshold updating coefficient*” of product p^l to product p^j for user u_i , where

$$\phi_i^{l \rightarrow j} \begin{cases} < 1, & \text{if } p^l \text{ is } \textit{complementary} \text{ to } p^j, \\ = 1, & \text{if } p^l \text{ is } \textit{independent} \text{ to } p^j, \\ > 1, & \text{if } p^l \text{ is } \textit{competing} \text{ to } p^j. \end{cases}$$

Intertwined Information Diffusion Model

- Intertwined Information Diffusion Model (TLT)

- Given network structure $G = (\mathcal{V}, \mathcal{E})$, product set \mathcal{P} , users activation thresholds $\{\theta_i^j\}_{u_i \in \mathcal{V}, p^j \in \mathcal{P}}$, user influence weight $\{w_{i,k}^j\}_{(u_i, u_k) \in \mathcal{E}, p^j \in \mathcal{P}}$.
- At step 1, information propagates from the seed user sets $\{\mathcal{S}^j\}_{p^j \in \mathcal{P}}$
- At step t ($t > 1$), all active users at step $t-1$ remain active, and inactive user u_i will be activated by their neighbors($\Gamma_{out}(u_i)$) to buy product if

$$\sum_{u_l \in \Gamma_{out}(u_i)} w_{l,i}^j \geq \theta_i^j$$

- For user u_i , who has been activated by products $p^{\tau_1}, p^{\tau_2}, \dots, p^{\tau_m} \in \mathcal{P} \setminus \{p^j\}$ in a sequence, u_i 's threshold toward product p^j will be

$$(\theta_i^j)^{\tau_1} = \theta_i^j \frac{P(s_i^j)}{P(s_i^j | s_i^{\tau_1})}, (\theta_i^j)^{\tau_2} = (\theta_i^j)^{\tau_1} \frac{P(s_i^j | s_i^{\tau_1})}{P(s_i^j | s_i^{\tau_1}, s_i^{\tau_2})}, \dots, (\theta_i^j)^{\tau_m} = (\theta_i^j)^{\tau_{m-1}} \frac{P(s_i^j | s_i^{\tau_1}, \dots, s_i^{\tau_{m-1}})}{P(s_i^j | s_i^{\tau_1}, \dots, s_i^{\tau_{m-1}}, s_i^{\tau_m})},$$

- In this paper, to simplify the calculation, we assume only the most recent activation has an effect on updating current thresholds:

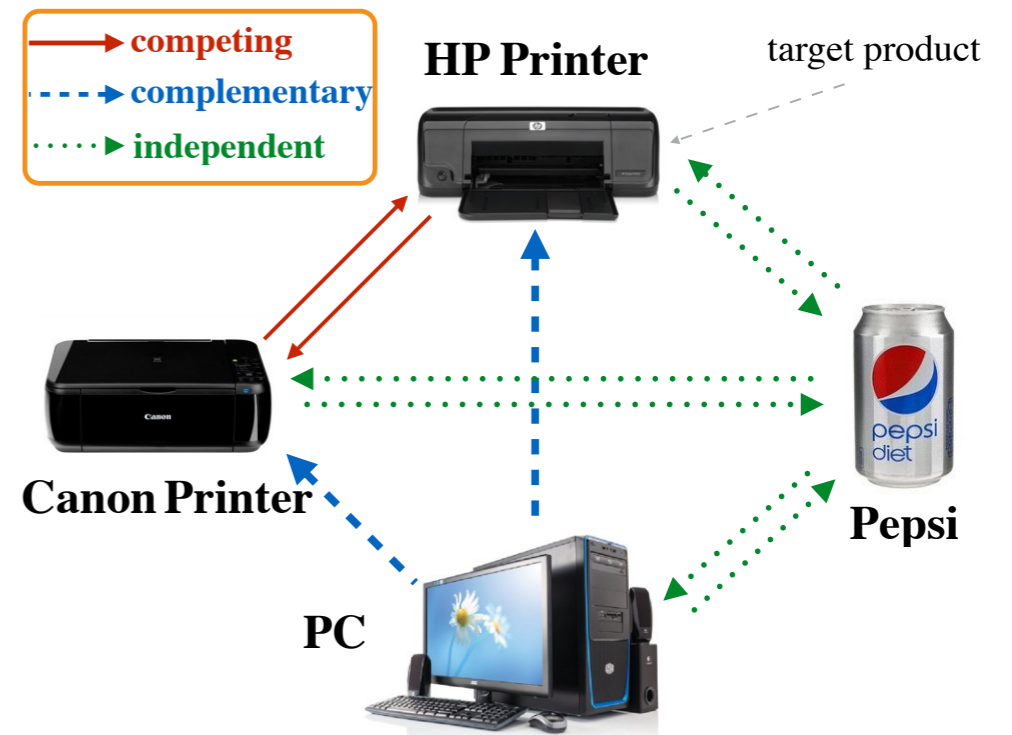
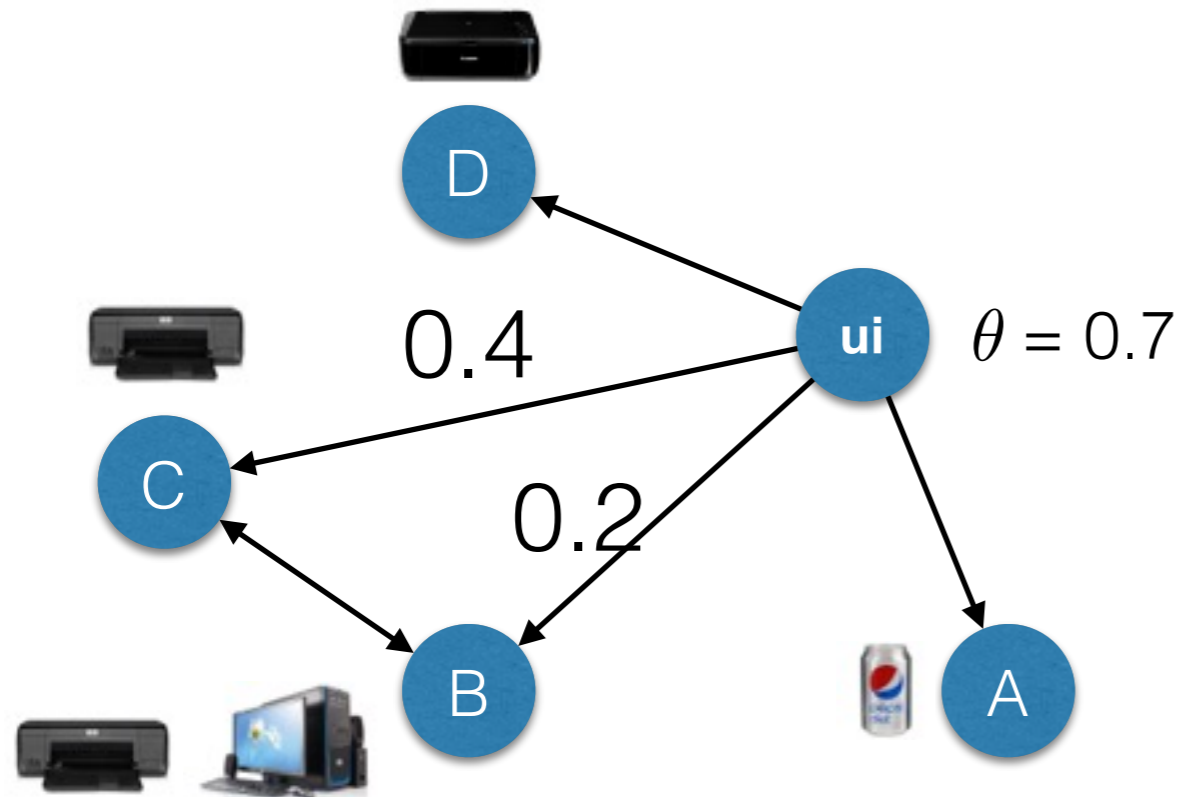
$$\frac{P(s_i^j | s_i^{\tau_1}, \dots, s_i^{\tau_{m-1}})}{P(s_i^j | s_i^{\tau_1}, \dots, s_i^{\tau_{m-1}}, s_i^{\tau_m})} \approx \frac{P(s_i^j)}{P(s_i^j | s_i^{\tau_m})} = \phi_i^{\tau_m \rightarrow j}.$$

- Therefore, we have

$$(\theta_i^j)^{\tau_m} \approx \theta_i^j \cdot \phi_i^{\tau_1 \rightarrow j} \cdot \phi_i^{\tau_2 \rightarrow j} \dots \phi_i^{\tau_m \rightarrow j}.$$

- The diffusion process stops if no further activation is possible.

Intertwined Information Diffusion Model Example



- Initially, u_i 's threshold to HP can be $\theta = 0.7$;
- u_i cannot be activated to buy HP, because

$$w_{B,u_i}^{hp} + w_{C,u_i}^{hp} < \theta_j^{hp}$$
- Assume u_i is activated by A to buy Pepsi, the new threshold will be $\theta \cdot \phi_i^{pepsi \rightarrow hp} = 0.7$;
- If u_i is activated by B to buy PC, the new threshold will be $\theta \cdot \phi_i^{pepsi \rightarrow hp} \cdot \phi_i^{pc \rightarrow hp} = 0.35$
- Therefore, user u_i will be activated by B, C to buy HP printer, since the influence is greater than the updated threshold

$$\phi_i^{pc \rightarrow canon} = 0.5$$

$$\phi_i^{pc \rightarrow hp} = 0.6$$

$$\phi_i^{canon \rightarrow hp} = 1.6$$

$$\phi_i^{hp \rightarrow canon} = 2.5$$

$$\phi_i^{\cdot \rightarrow pepsi} = \phi_i^{pepsi \rightarrow \cdot} = 1.0$$

Intertwined Viral Marketing Problem

- Two variants of the TIM problem:
 - ***Conditional TIM problem: C-TIM***
 - The other products are promoted ahead of the target product.
 - Information about other products have been propagated to users in the network already.
 - E.g., Apple to announce *iPhone 7* long after the release of iPad Pro, Samsung Galaxy S7, etc.
 - **Joint TIM problem: J-TIM**
 - The other products are being promoted simultaneously with the target product in the network.
 - Information about all the products have not be spread to users in the network yet.
 - E.g., Apple and Samsung will release the *new iPhone* and new Galaxy phone to compete for the market share.

Conditional TIM Problem

- After the spread of information about the other products, we can update the users' thresholds towards the target product.
- Based on the updated network, we can carry the promotion of the target product.
- Conditional Intertwined Influence Function Definition

Definition (Conditional Intertwined Influence Function): Let $\mathcal{S}^{-j} = (\mathcal{S}^1, \dots, \mathcal{S}^{j-1}, \mathcal{S}^{j+1}, \dots, \mathcal{S}^n)$ be the known seed user sets selected for all products in $\mathcal{P} \setminus \{p^j\}$, the *influence function* of the target product p^j given the known *seed user sets* \mathcal{S}^{-j} is defined as the *conditional intertwined influence function*: $I(\mathcal{S}^j | \mathcal{S}^{-j})$.

- C-TIM Problem Definition

C-TIM Problem: The C-TIM problem aims at selecting the optimal *marketing strategy* $\bar{\mathcal{S}}^j$ to maximize the *conditional intertwined influence function* of p^j in the network, i.e.,

$$\bar{\mathcal{S}}^j = \arg_{\mathcal{S}^j} \max I(\mathcal{S}^j | \mathcal{S}^{-j}).$$

Conditional TIM Problem Analysis and Solution

- C-TIM Problem Analysis

Theorem: The C-TIM problem is *NP-hard* based on the TLT diffusion model.

- Conditional Intertwined Influence Function Property

Theorem: For the TLT diffusion model, the *conditional influence function* is *monotone* and *submodular*.

- Solution: Conditional interTwined Influence Estimator (C-TIER)

- step-wise greedy method, which selects users who will introduce the maximum influence increase in each step

Algorithm 1 The C-TIER Algorithm

Input: input social network $G = (\mathcal{V}, \mathcal{P}, \mathcal{E})$

target product: p^j

known seed user sets of $\mathcal{P} - \{p^j\}$: \mathcal{S}^{-j}

conditional influence function of p^j : $I(\mathcal{S}^j | \mathcal{S}^{-j})$

seed user set size of p^j : k^j

Output: selected seed user set \mathcal{S}^j of size k^j

1: initialize seed user set $\mathcal{S}^j = \emptyset$

2: propagate influence of products $\mathcal{P} - \{p^j\}$ with \mathcal{S}^{-j} and update users' thresholds with intertwined threshold updating strategy

3: **while** $\mathcal{V} \setminus \mathcal{S}^j \neq \emptyset \wedge |\mathcal{S}^j| \neq k^j$ **do**

4: pick a user $u \in \mathcal{V} - \mathcal{S}^j$ according to equation $\arg \max_{u \in \mathcal{V}} I(\mathcal{S}^j \cup \{u\} | \mathcal{S}^{-j}) - I(\mathcal{S}^j | \mathcal{S}^{-j})$

5: $\mathcal{S}^j = \mathcal{S}^j \cup \{u\}$

6: **end while**

7: return \mathcal{S}^j .

Experimental Results of C-TIM Problem

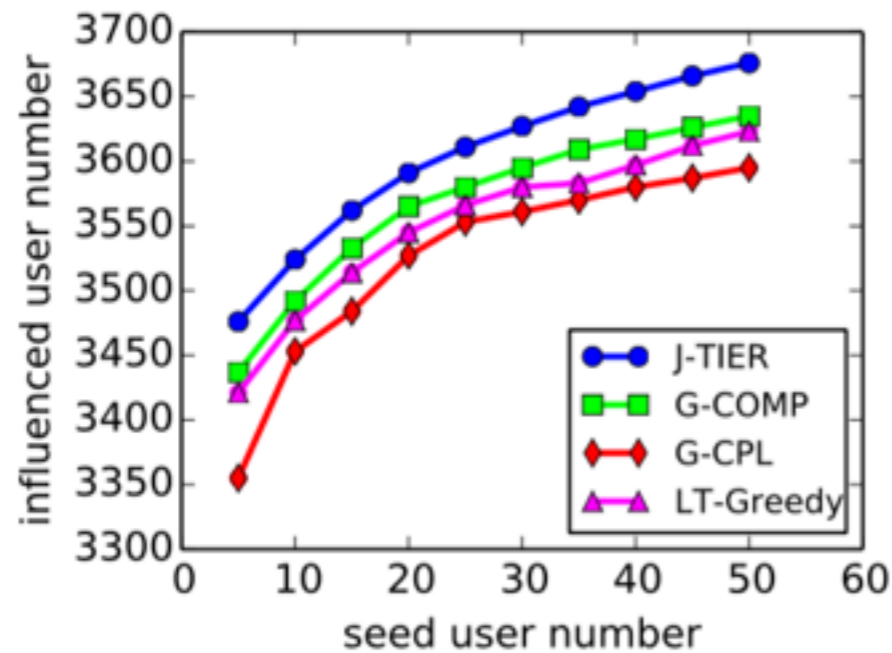
- Experimental Datasets
 - Facebook Network
 - Wikipedia Network
 - arXiv Collaboration Network
 - Epinions Network
- Comparison Methods
 - C-TIER: Step-wise greedy seed user selection method based on TLT diffusion model
 - LT-Greedy: Step-wise greedy seed user selection method based on traditional LT diffusion model without considering product relationships
 - LT-PageRank: Select nodes with the top K PageRank scores
 - LT-Degree: Select nodes with the top K degree scores
 - LT-Random: Randomly select K nodes

TABLE I
PROPERTIES OF THE DIFFERENT NETWORKS

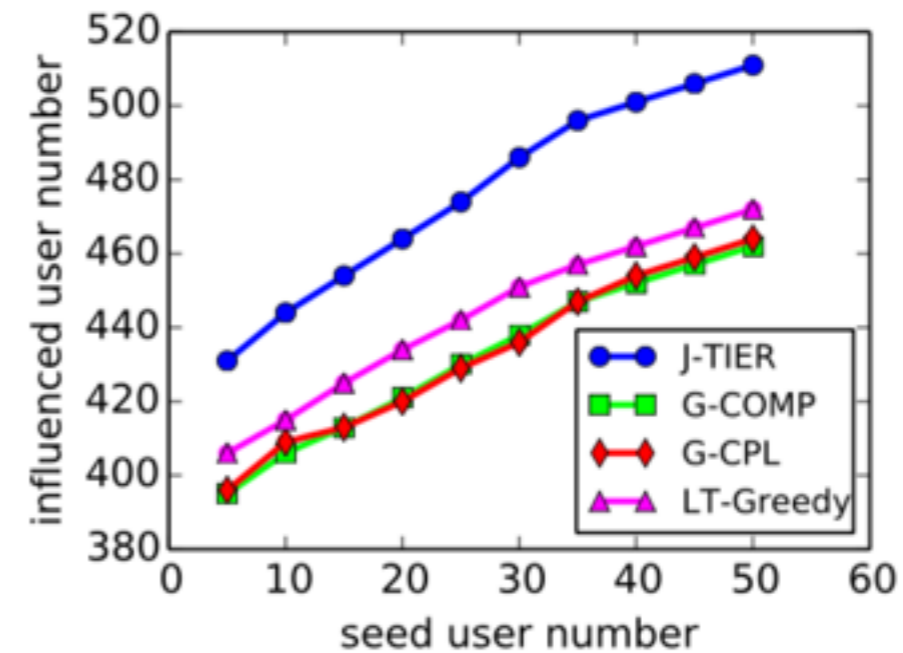
network	# nodes	# links	link type
Facebook	4,039	88,234	undirected
Wikipedia	7,115	103,689	directed
arXiv	5,242	14,496	undirected
Epinions	7,725	82,861	directed

Experimental Results of C-TIM Problem

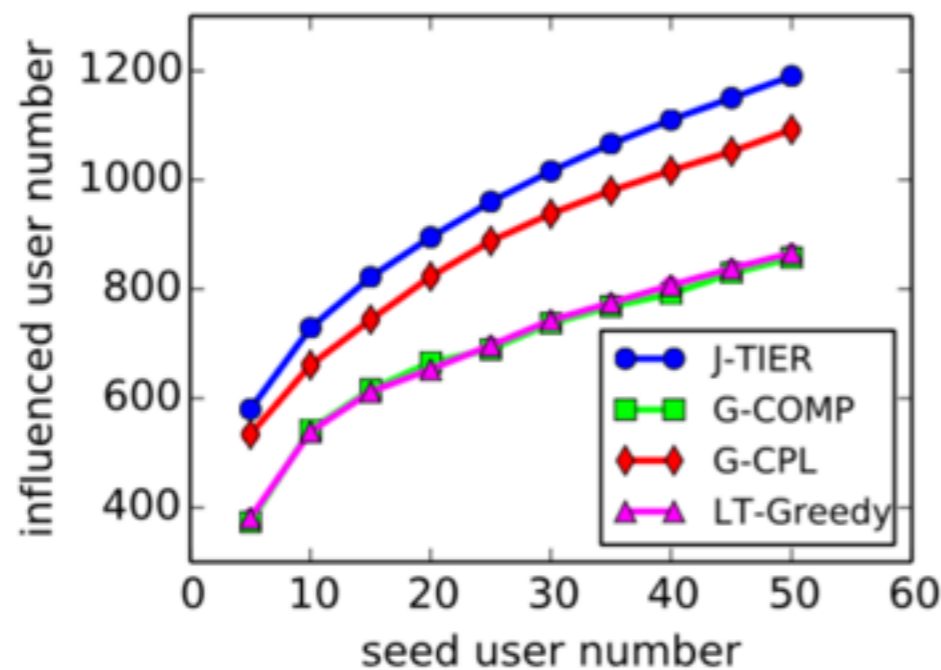
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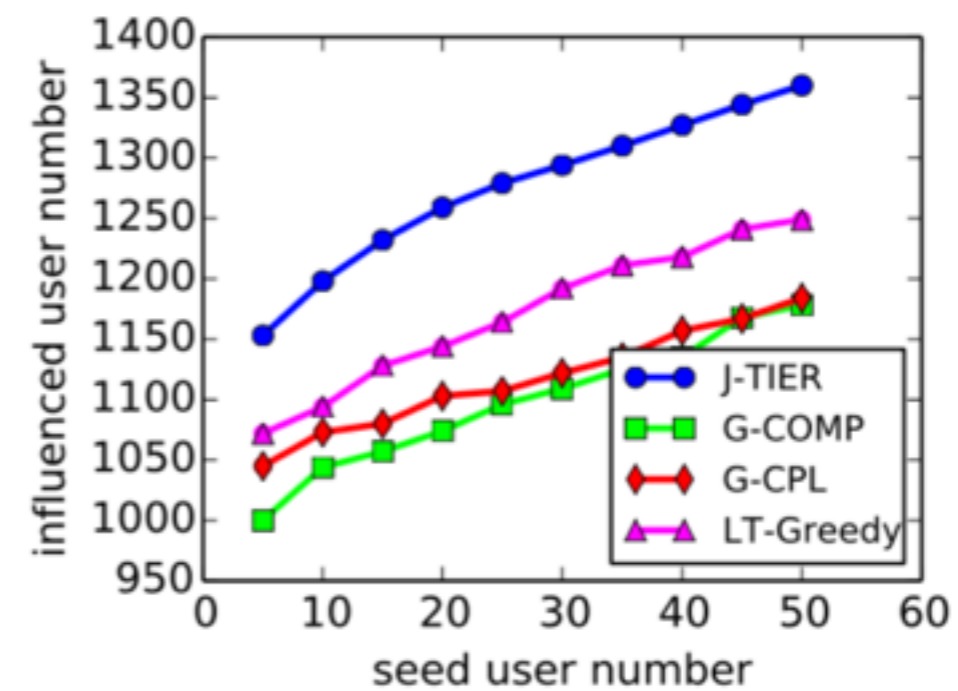
(a) Facebook



(b) Wikipedia



(c) arXiv



(d) Epinions

Joint TIM Problem

- Products with Intertwined relationships are being promoted in online social networks at the same time.
- The seed users selected by other products are unknown, and the information about other products has not been propagated yet.
- Joint Intertwined Influence Function Definition

Definition (Joint Intertwined Influence Function): When the seed user sets of products $\mathcal{P} \setminus \{p^j\}$ are unknown, i.e., \mathcal{S}^{-j} is not given, the *influence function* of product p^j together with other products in $\mathcal{P} \setminus \{p^j\}$ is defined as the *joint intertwined influence function*: $I(\mathcal{S}^j; \mathcal{S}^{-j})$.

- C-TIM Problem Definition

J-TIM Problem: J-TIM problem aims at choosing the optimal *marketing strategy* $\bar{\mathcal{S}}^j$ to maximize the *joint intertwined influence function* of p^j in the network, i.e.,

$$\bar{\mathcal{S}}^j = \arg_{\mathcal{S}^j} \max I(\mathcal{S}^j; \mathcal{S}^{-j}),$$

where set \mathcal{S}^{-j} can take any possible value.

Joint TIM Problem Analysis

- J-TIM Problem Analysis

Theorem: The J-TIM problem is *NP-hard* based on the TLT diffusion model.

- Joint Intertwined Influence Function Property

Theorem: Based on the TLT diffusion model, the *joint influence function* is *monotone* and *submodular* if all the other products are *independent* to the target product p^j .

Theorem: Based on the TLT diffusion model, the *joint influence function* is not *monotone* nor *submodular* if there exist products which are either *competing* or *complementary* to the target product p^j .

- No theoretic performance guarantee exists for the step-wise greedy seed user selection algorithm in the J-TIM problem if there exists one products either competing or complementary to the target product.

J-TIM Problem Solution: J-TIER

- Joint interTwined Influence Estimator (J-TIER)
 - In J-TIER, all the products are assumed to be “selfish” and aims at maximizing their influence gain, which leads to a “game” among products.
 - Formally, the seed users to be selected by all the products can be represented as set

$$\{\mathcal{S}_1, \mathcal{S}_2, \dots, \mathcal{S}_j, \dots, \mathcal{S}_{|\mathcal{P}|}\}$$

- J-TIER lets the products to select seed users alternatively in random order step by step. Let $(\mathcal{S})^{\tau-1}$ be the seed users selected by all the products after round $\tau - 1$
- If product p^j is randomly picked to select seed users in round τ , the selected seed user will be

$$\arg \max_{u \in \mathcal{V} - (\mathcal{S}^j)^{\tau-1}} [I((\mathcal{S}^j)^{\tau-1} \cup \{u\}; (\mathcal{S}^{-j})^{\tau-1}) - I((\mathcal{S}^j)^{\tau-1}; (\mathcal{S}^{-j})^{\tau-1})].$$

- If product p^i is randomly picked to select seed user after p^j , the selected seed user will be

$$\hat{u}^i = \arg \max_{u \in \mathcal{V} - (\mathcal{S}^i)^{\tau-1}} [I((\mathcal{S}^i)^{\tau-1} \cup \{u\}; \bar{\mathcal{S}}^{-i}) - I((\mathcal{S}^i)^{\tau-1}; \bar{\mathcal{S}}^{-i})].$$

- Such a process stops until all the products finish the seed user selection process.

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- If product p^j is randomly picked to select seed users in round \mathcal{T} , the selected seed user will be

$$\arg \max_{u \in \mathcal{V} - (\mathcal{S}^j)^{\tau-1}} [I((\mathcal{S}^j)^{\tau-1} \cup \{u\}; (\mathcal{S}^{-j})^{\tau-1}) - I((\mathcal{S}^j)^{\tau-1}; (\mathcal{S}^{-j})^{\tau-1})].$$

- If product p^i is randomly picked to select seed user after p^j , the selected seed user will be

$$\hat{u}^i = \arg \max_{u \in \mathcal{V} - (\mathcal{S}^i)^{\tau-1}} [I((\mathcal{S}^i)^{\tau-1} \cup \{u\}; \bar{\mathcal{S}}^{-i}) - I((\mathcal{S}^i)^{\tau-1}; \bar{\mathcal{S}}^{-i})].$$

- Such a process stops until all the products finish the seed user selection process.

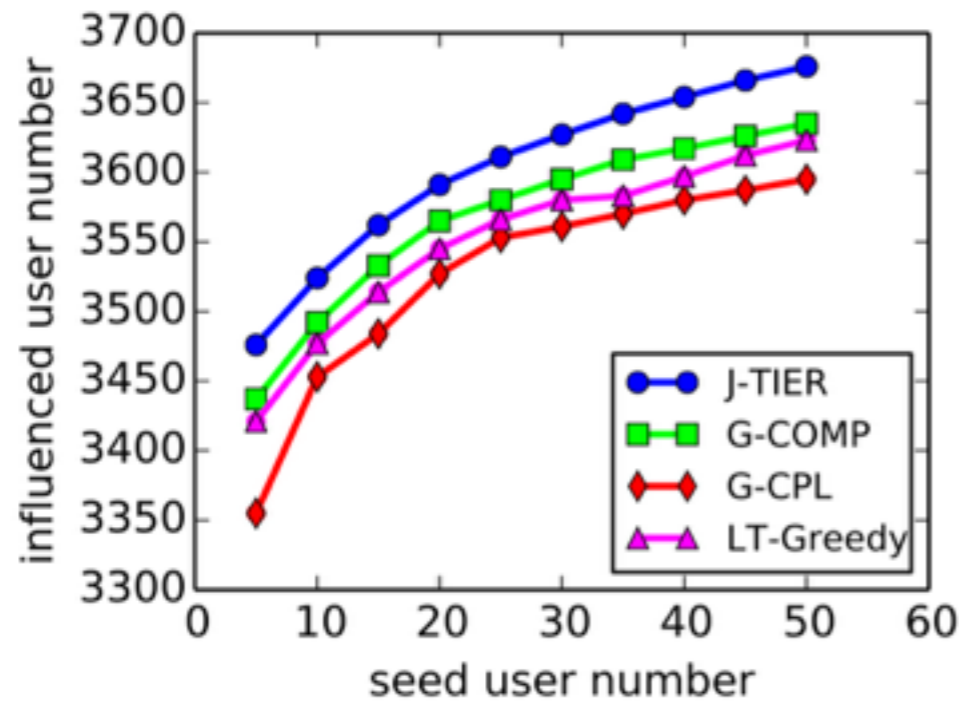
Experimental Results of J-TIM Problem

- Experimental Datasets
 - Facebook Network
 - Wikipedia Network
 - arXiv Collaboration Network
 - Epinions Network
- Comparison Methods
 - J-TIER: Iterative seed user selection method based on TLT diffusion model, which considers all products in the game.
 - G-COMP: Seed user selection considering the competing products only in the game.
 - G-CPL: Seed user selection considering the complementary products only in the game.
 - G-INDEP: Seed user selection considering the independent products only in the game.

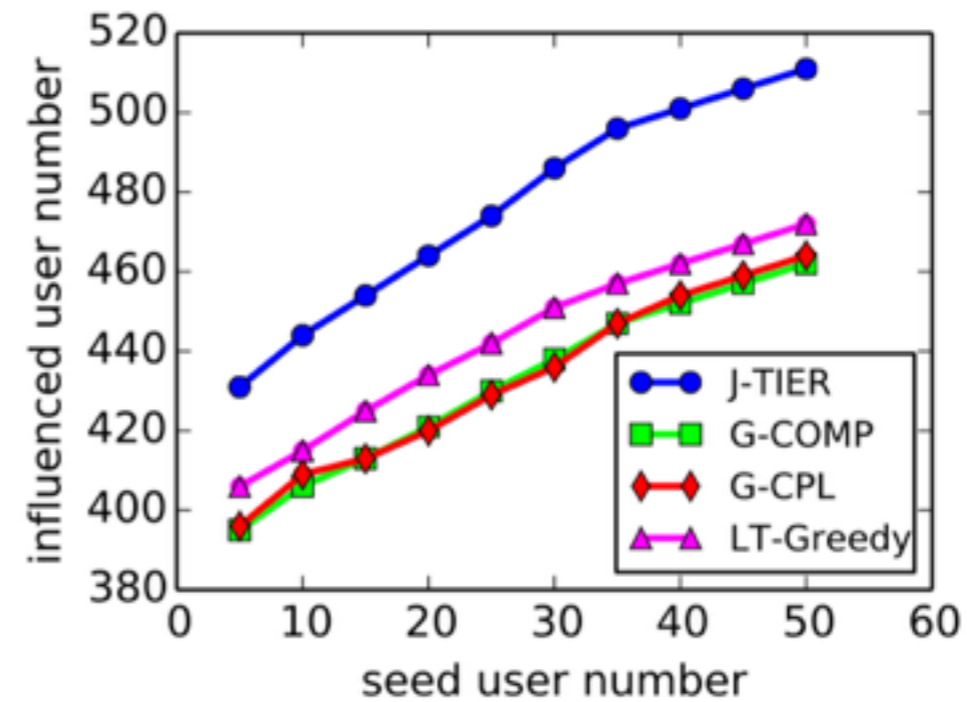
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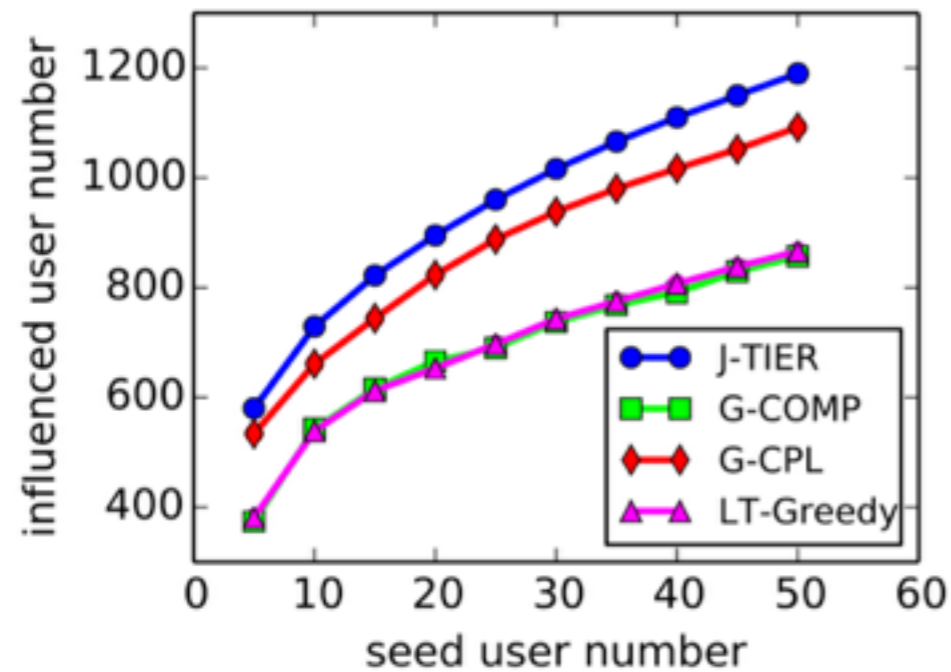
Experimental Results of J-TIM Problem



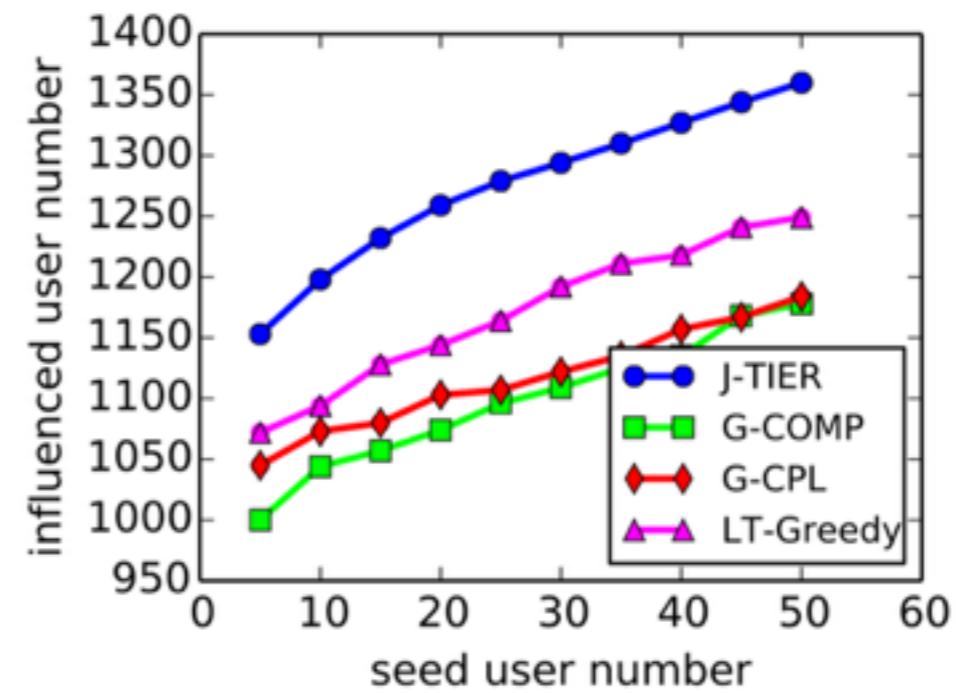
(a) Facebook



(b) Wikipedia



(c) arXiv



(d) Epinions

Summary

- **Problem Studied**

- *Intertwined viral marketing problem* in social networks with multiple products being promoted at the same time

- **Proposed Method**

- *TLT Diffusion Model*: depicts the information diffusion process in online social networks considering the intertwined relationships among the products
- *C-TIER for C-TIM problem*: step-wise greedy seed user selection, achieve $1-1/e$ approximation of the optimal result
- *J-TIER for J-TIM problem*: game based alternative seed user selection, considers the *competing*, *complementary* and *independent* products simultaneously

Intertwined Viral Marketing in Social Networks

Q&A

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