

Capacity Constrained Influence Maximization in Social Networks

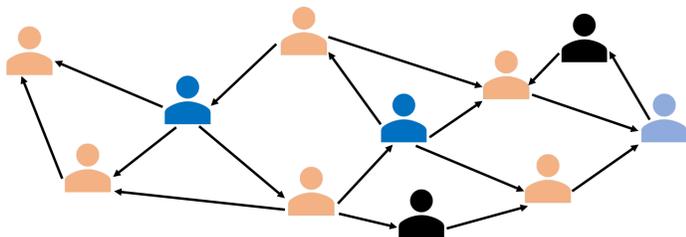
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Influence Maximization

In viral marketing, merchants

- pay k influencers
- hope word-of-mouth promotes the product
- create a cascade of influenced individuals



viral marketing on the social network

Influence Maximization (IM)

For a fixed k , how to pick k influencers to maximize the eventual influence spread $\#(\text{blue}) + \#(\text{orange})$

Limitations

Limitation #1 of conventional IM

- Conventional IM
 - ignores user's **capacity** for spending on the promotion
- User's capacity
 - is **crucial** as it determines the adoption of the product
 - is **limited** online (e.g., while playing e-games)

Limitation #2 of conventional IM

- Conventional IM
 - assumes influencers **unconditionally** be initial adopters
- Observation from real-world scenarios
 - influencers tend to be **friends of the initial adopters**



source: <https://zootopia.fandom.com/wiki>

New Problem: Capacity Constraint Influence Maximization

Input: d initial adopters and a capacity constant k

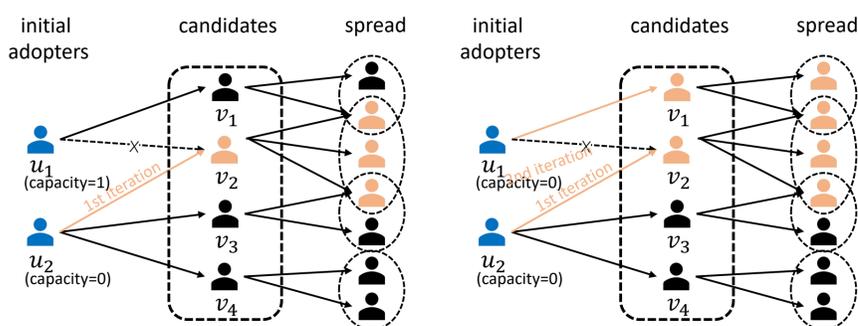
Output: k influential friends (**seeds**) for each of d initial adopters

Objective: maximize the spread of the set of all selected seeds

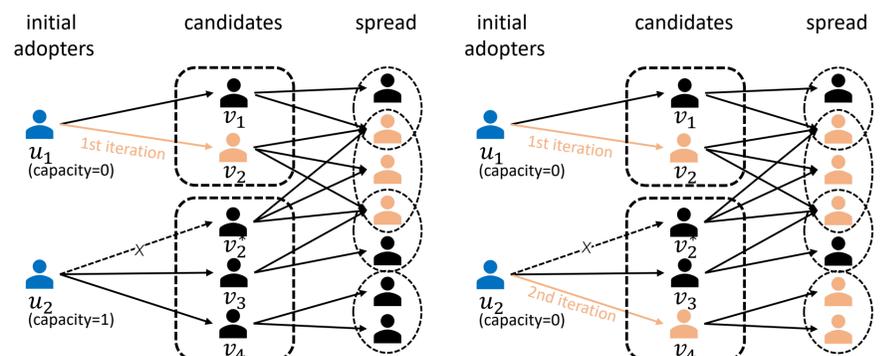
CIM is NP-hard

Greedy Algorithms and Scalable Implementations

MG-Greedy



RR-Greedy



Scalable implementations: follow the framework of OPIM-C and redesign the constants with rigorous analysis

Instances: first-cut versions (MG-OPIM, RR-OPIM) and final version (RR-OPIM+)

Experiments

CIM solutions

Local competitors:

- Degree, PageRank, IMM, OPIM-C

Greedy solutions:

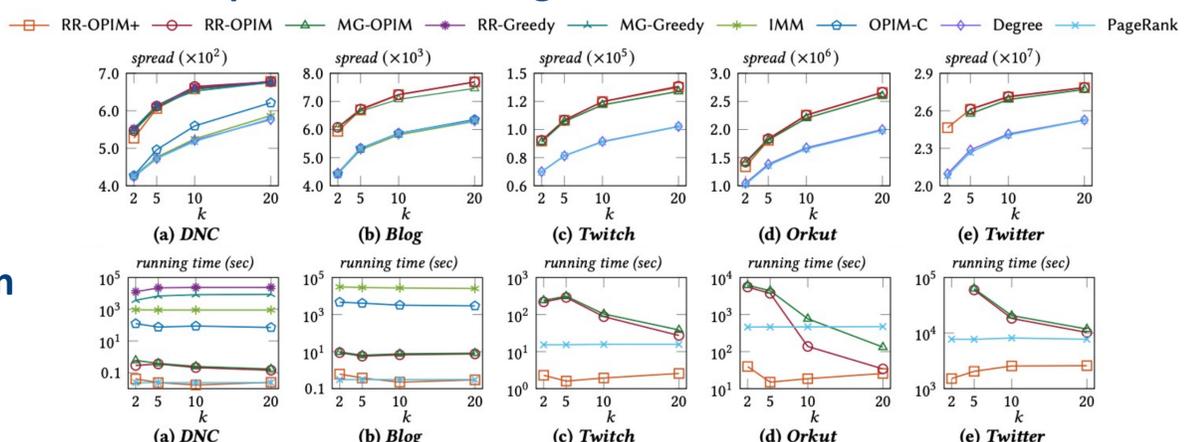
- MG-Greedy, RR-Greedy

Scalable implementations:

- MG-OPIM, RR-OPIM, RR-OPIM+

RR-OPIM+ outperforms all other solutions in terms of actual spread of Tencent e-games

RR-OPIM+ outperforms all other solutions on 5 public datasets in terms of spread and running time.



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Offline evaluation					
Solution	RR-OPIM+	MG-OPIM	RR-OPIM	Degree	PageRank
Spread	1,632	1,625	1,609	1,488	1,471

Online deployment		
Solution	Treatment	Control
Spread	60.69K	58.28K