Project1

Information Exposure Maximization

Evaluator & Heuristic Search

Outline

- A brief review of information exposure maximization
- An estimation method for balanced information exposure
- A heuristic algorithm for information exposure maximization
- Summary

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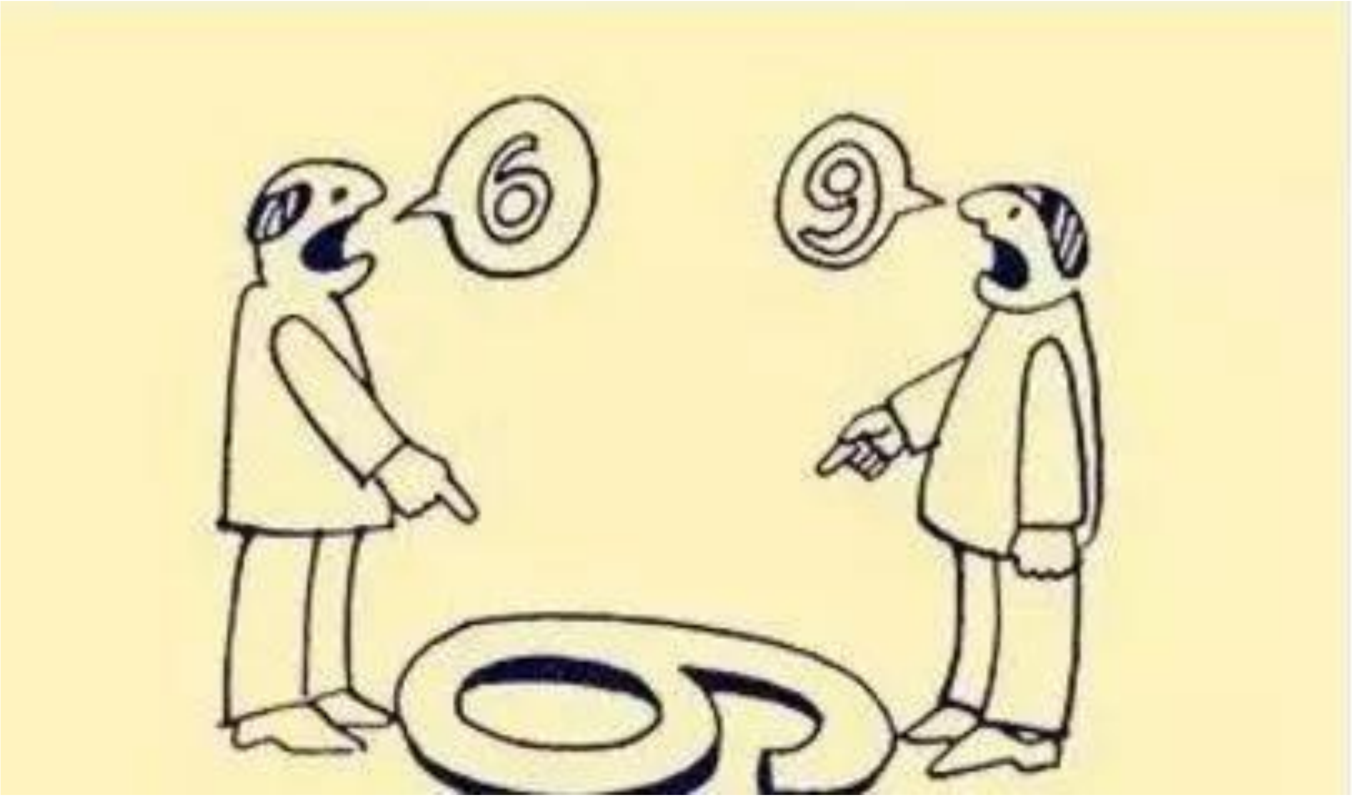
Brief review of IEM

- The Information Exposure Maximization (IEM) problem is proposed to solve the echo chamber effect on social media.



Brief review of IEM

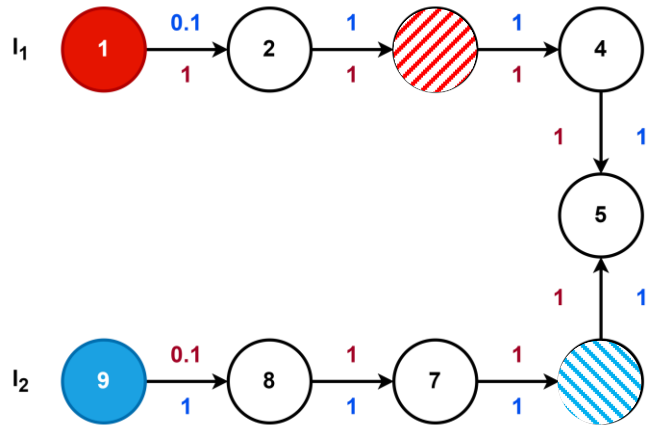
Given a social network $G = (V, E)$, two initial seed sets I_1 and I_2, and a budget k.

The IEM is **to find two balanced seed sets S_1 and S_2**, where $|S_1| + |S_2| \leq k$, and **maximize the balanced information exposure**, i.e.,

$$\max \Phi(S_1, S_2) = \max \mathbb{E}\left[\left|V \setminus \left(r_1(I_1 \cup S_1) \bigtriangleup r_2(I_2 \cup S_2)\right)\right|\right]$$

$$\text{s.t.} \quad |S_1| + |S_2| \leq k$$

$$S_1, S_2 \subseteq V$$



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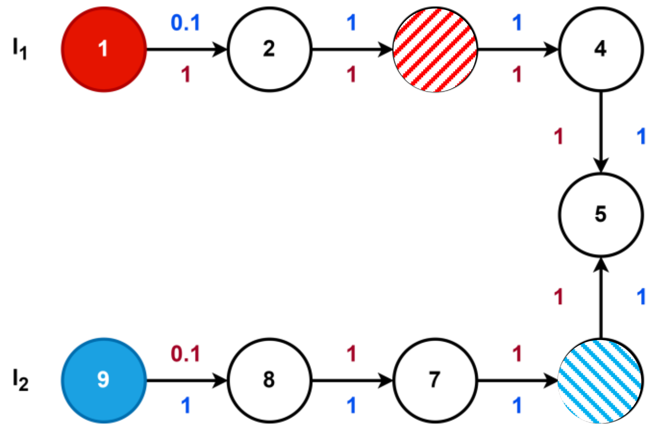
Both are random variables determined by the stochastic process of the diffusion model and their diffusion probabilities



Brief review of IEM

- Finding an optimal solution of IEM is NP-hard.
- Computing the balanced information exposure for a given solution is NP-hard.

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Balanced information exposure estimation

Monte Carlo simulation

- A computational algorithm that uses **repeated random sampling** to obtain the likelihood of a range of results of occurring

Balanced information exposure estimation

Monte Carlo simulation

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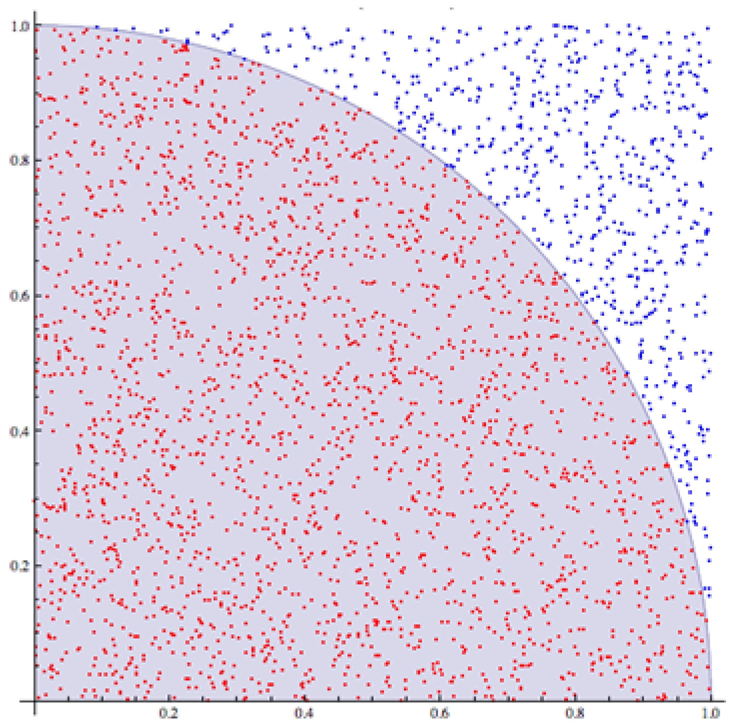
Example2:Estimate π

As shown in the figure, if n=3000 points are randomly generated in the square region of 1*1, 2375 points are obtained in the quarter circle with radius 1, then the value of π can be evaluated as?



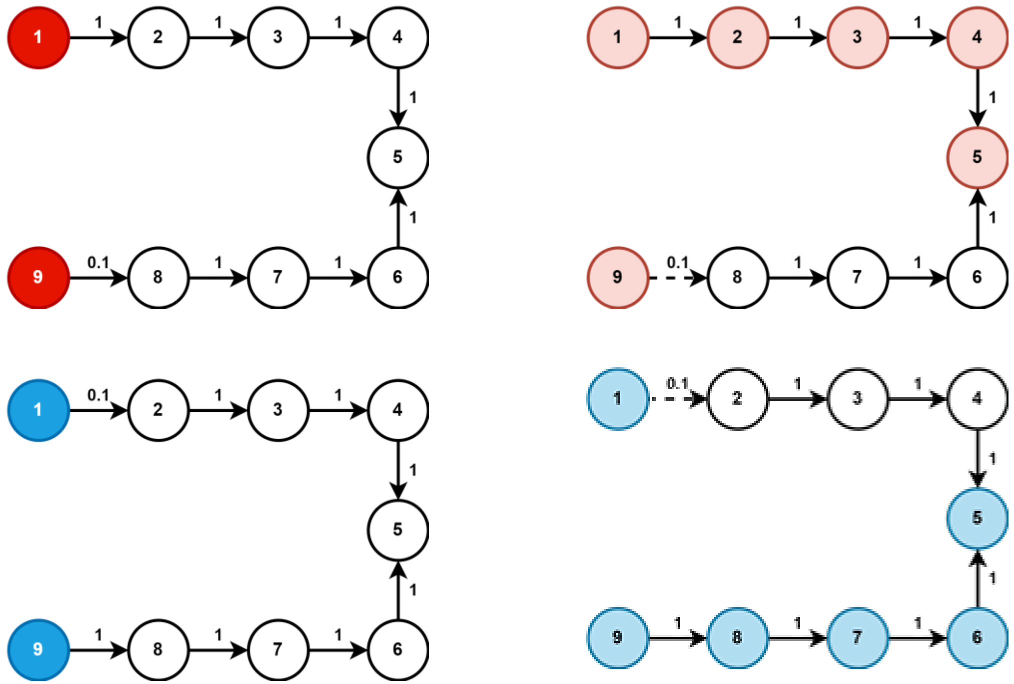
Balanced information exposure estimation

Monte Carlo simulation

- A computational algorithm that uses **repeated random sampling** to obtain the likelihood of a range of results of occurring

Estimate balanced information exposure:

$$\Phi_{g\sim G}(S_1, S_2) = \left| V \setminus \left(r_1(I_1 \cup S_1) \triangle r_2(I_2 \cup S_2) \right) \right|_g = |\{1, 2, 5, 8, 9\}| = 5$$



Balanced information exposure estimation

Monte Carlo simulation

- A computational algorithm that uses **repeated random sampling** to obtain the likelihood of a range of results of occurring

Estimate balanced information exposure:

$$\max \Phi(S_1, S_2) = \max \mathbb{E}\left[\left|V \setminus \left(r_1(I_1 \cup S_1) \triangle r_2(I_2 \cup S_2)\right)\right|\right]$$

$$\boxed{\widehat{\Phi}(S_1, S_2) = \frac{\sum_{i=1}^{N} \Phi_{g_i}(S_1, S_2)}{N}}$$

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Heuristic algorithm for IEM

Greedy best-first search

- Main idea: expand the node with the largest $h(v)$ value

$h(v)$ = increment to the balanced information exposure

Algorithm: Greedy best-first search

$S_1 \leftarrow S_2 \leftarrow \emptyset$;

while $|S_1| + |S_2| \leq k$ do

$v_1^* \leftarrow \arg\max_v \big(\Phi(S_1 \cup \{v\}, S_2) - \Phi(S_1, S_2)\big)$;

$v_2^* \leftarrow \arg\max_v \big(\Phi(S_1, S_2 \cup \{v\}) - \Phi(S_1, S_2)\big)$;

add the better option between $< v_1^*, \emptyset >$ and $< \emptyset, v_2^* >$ to $< S_1, S_2 >$ while respecting the budget.

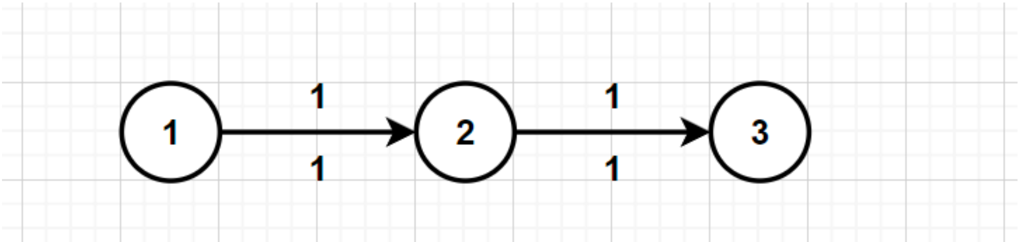
Heuristic algorithm for IEM

Greedy best-first search

Input:

$I_1 = \{1\}, I_2 = \{3\}, k = 2$

Network $G = (V, E)$



Iteration 1:

$I_1 = \{1\}, I_2 = \{3\}, S_1 = S_2 = \{\}$

Step1:Find v_1^*:

v =1, U1={1}, U2={3}, E1={1,2,3},E2={3}, Φ =1

v = 2, U1={1,2}, U2={3}, E1={1,2,3}, E2={3}, Φ =1

v = 3, U1={1,3}, U2={3}, E1={1,2,3}, E2={3}, Φ =1

$v_1^* = 1$

Step2:Find v_2^*:

v =1, U1={1}, U2={1,3}, E1={1,2,3},E2={1,2,3}, Φ =3

v = 2, U1={1}, U2={2,3}, E1={1,2,3}, E2={2,3}, Φ =2

v = 3, U1={1}, U2={3}, E1={1,2,3}, E2={3}, Φ =1

$v_2^* = 1$

Step3:add the better option between $< v_1^*, \emptyset >$ and $< \emptyset, v_2^* >$ to $< S_1, S_2 >$

<{},{}> → <{},{1}>

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Summary

- Information exposure maximization is computationally complex
- **Monte Carlo simulations** for balanced information exposure estimation
- **Greedy best-first search** to find balanced seed sets
- **Improvements in solution quality or computing efficiency are encouraged**