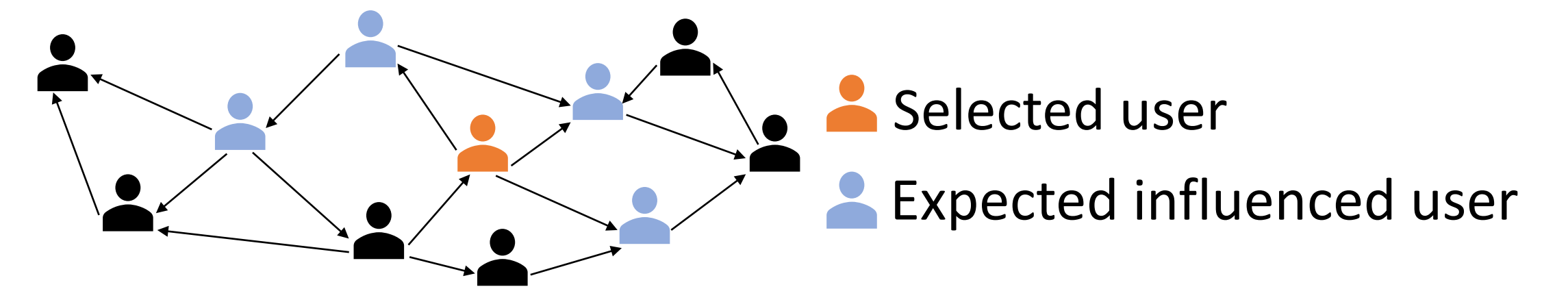


## Introduction

**Influence Maximization (IM):** find a set of users  $S$  (👤) that maximizes their expected influence (👤 + 👤) in a social network

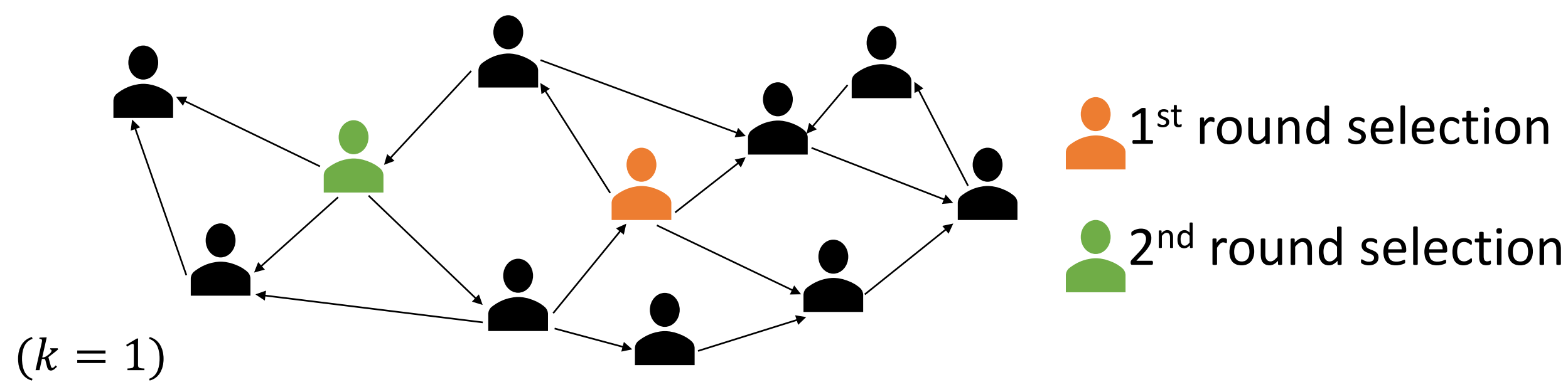


### IM with a Matroid Constraint:

$S$  must satisfy a matroid constraint  $M = (U, I): S \subseteq U$  and  $S \in I$ , where  $I$  represents certain feasible solutions. A matroid  $M$  allows constraints across **multiple sets of users** or **different objects**.

#### Example #1: IM in Multiple Rounds (MRIM)

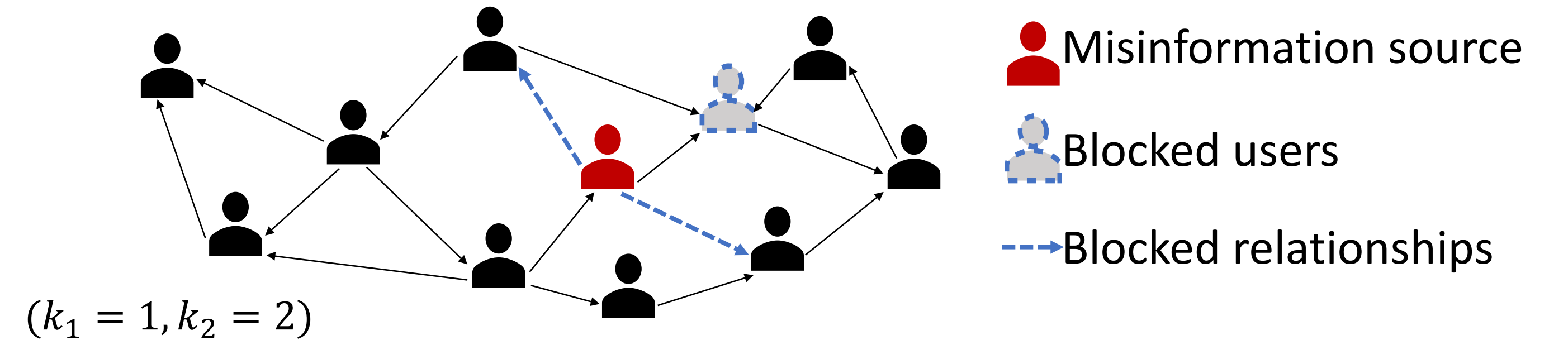
**Constraint:** select  $\leq k$  users each round



**Objective:** maximize expected influence from selected promoters (👤 and 👤)

#### Example #2: Adversarial attacks on IM (AdvIM)

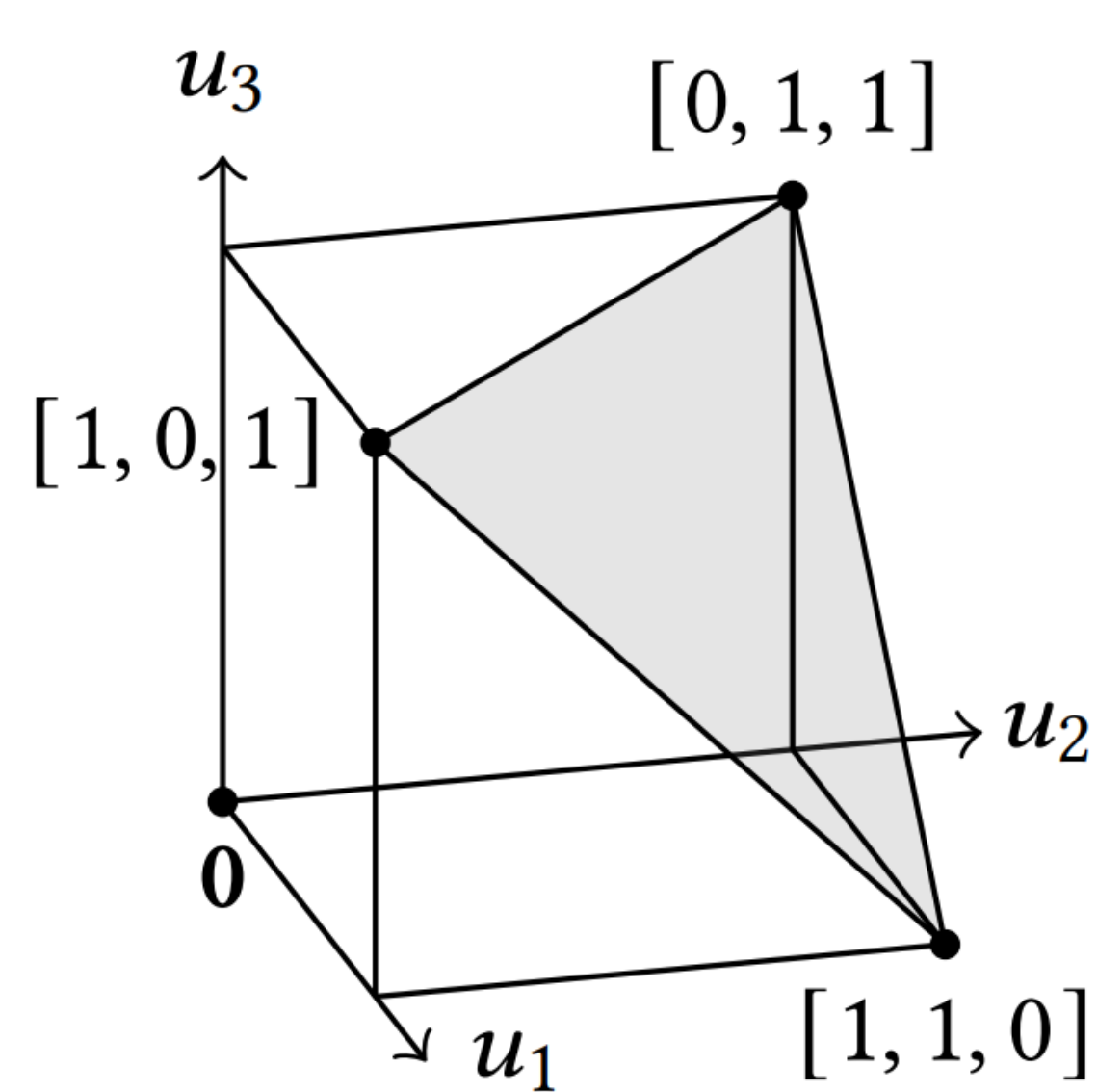
**Constraint:** block  $\leq k_1$  users and  $\leq k_2$  relationships



**Objective:** minimize expected influence from misinformation source (👤)

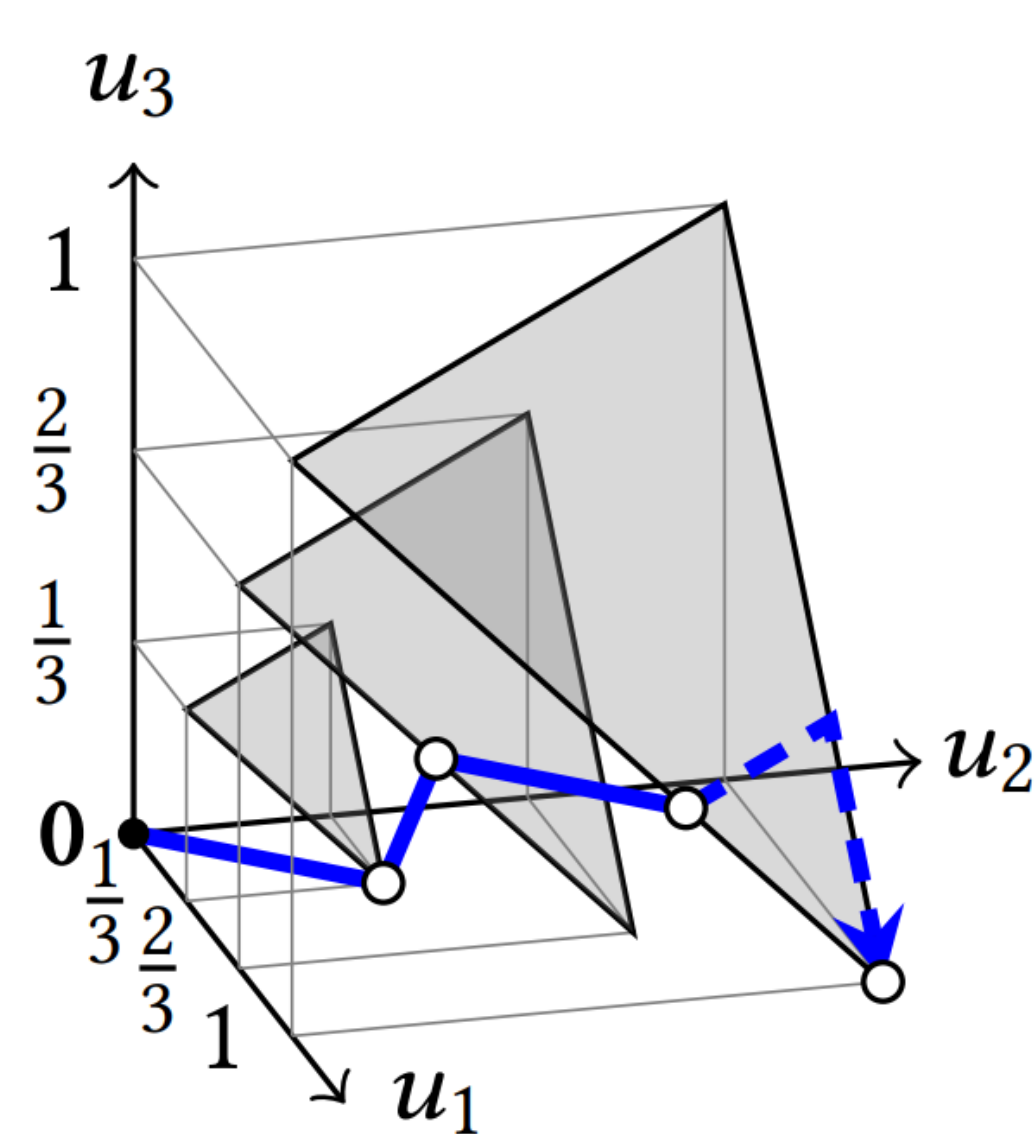
## Proposed Solution: Boosting Approximation from $1/2-\epsilon$ to $(1-1/e-\epsilon)$

A hypergeometric view of a matroid with  $U = \{u_1, u_2, u_3\}$  and  $I = \{S \subseteq U: |S| \leq 2\}$



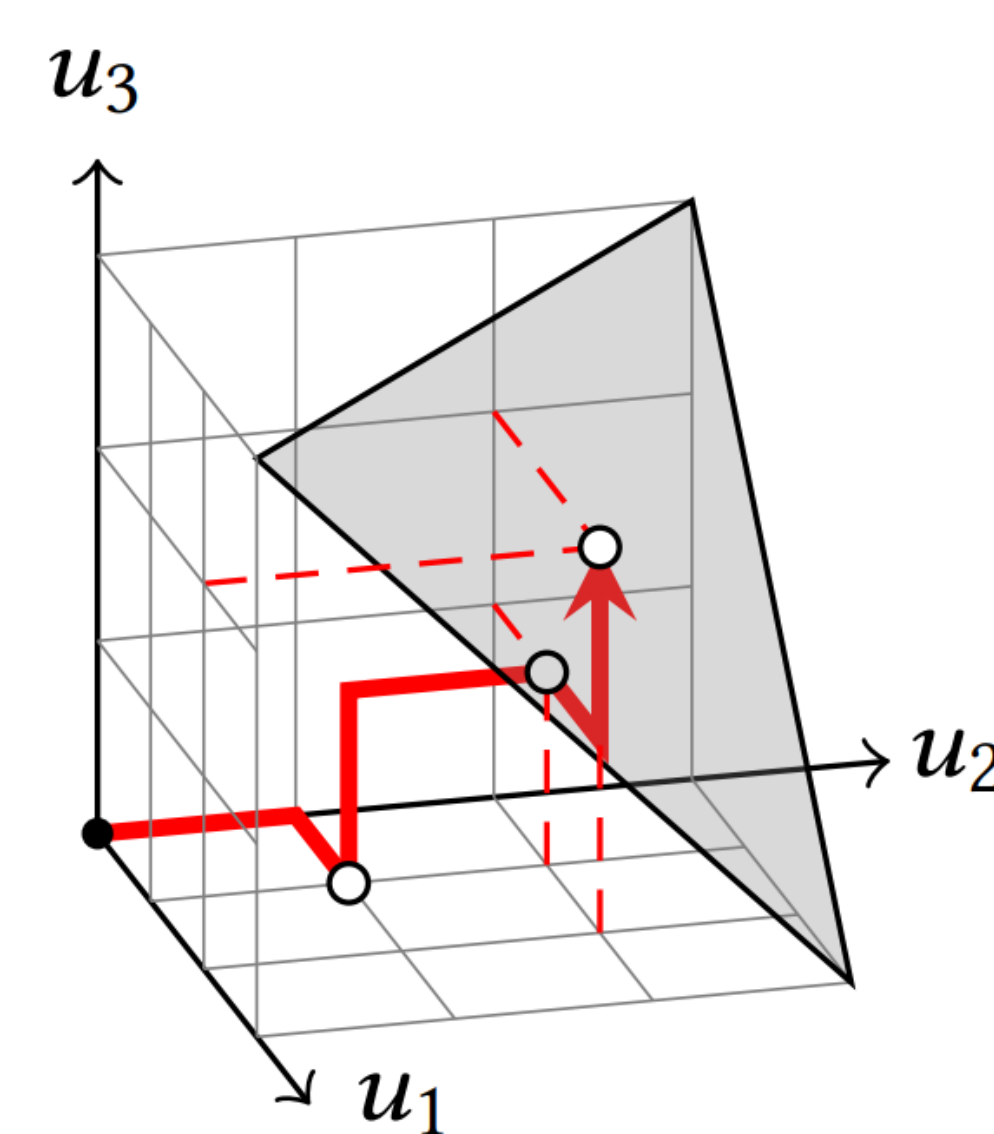
### Overview

- polyhedron  $\rightarrow$  solution space
- shaded face  $\rightarrow$  partial solutions
- labeled dots  $\rightarrow$  final solutions



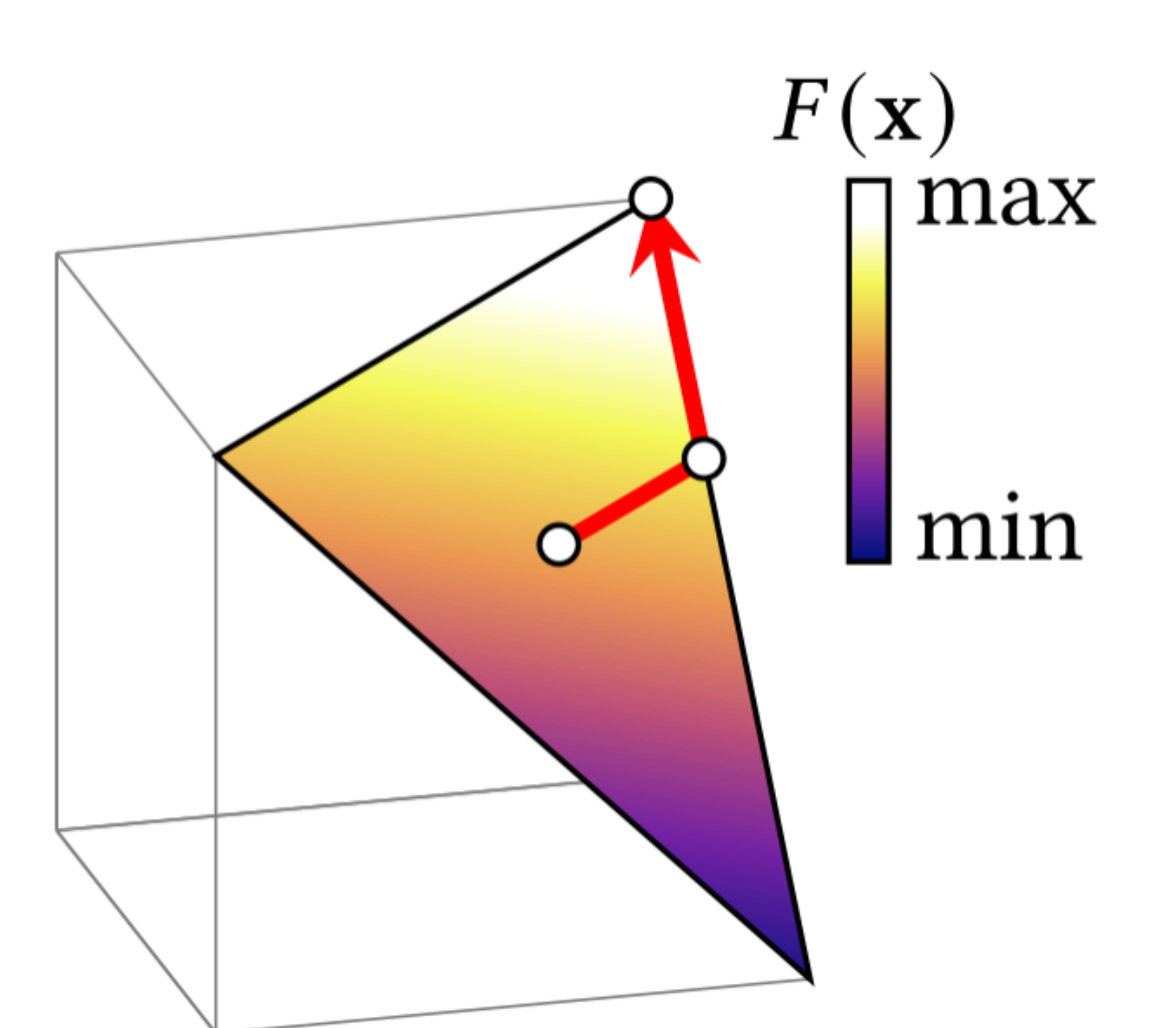
### Previous Solver

- coarse-grained hill-climbing
- using sampling to estimate partial solutions
- $O(n^7 \log n)$  running time



### Proposed Searching

- fine-grained search
- calculate partial solutions efficiently & **deterministically**
- $O_\epsilon(n \cdot \text{poly}(\log n))$  time



### Proposed Rounding

- round partial solutions to final solutions deterministically
- no loss** in solution quality

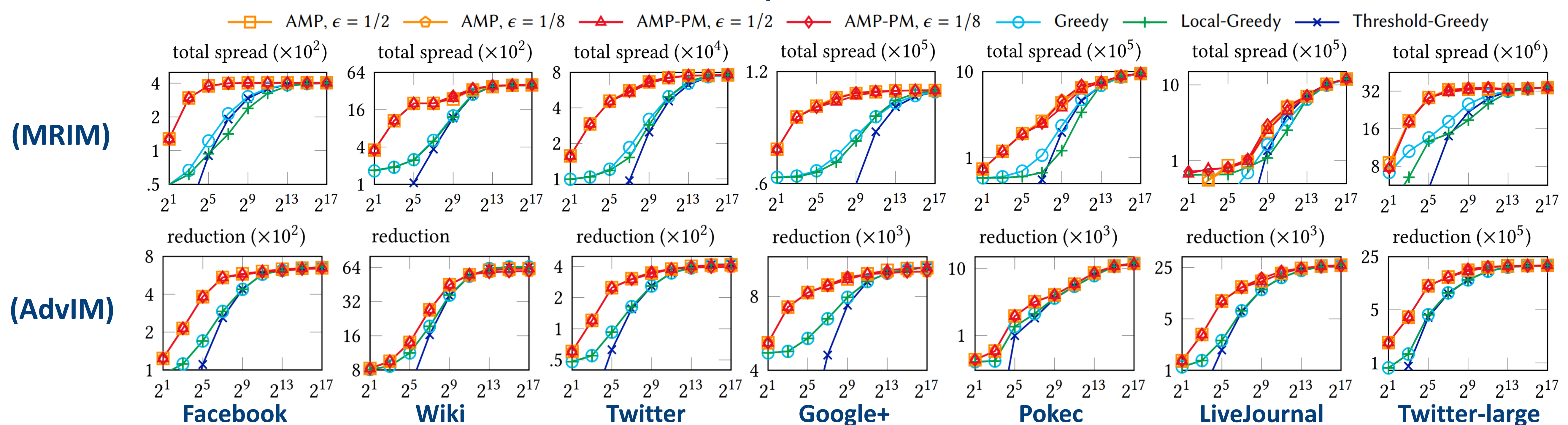
**Scalable implementations:** follow the framework of OPIM-C and redesign the constants with rigorous analysis

**Final Algorithm (AMP and RAMP):**  $(1 - 1/e - \epsilon)$  – approximation + scalable running time

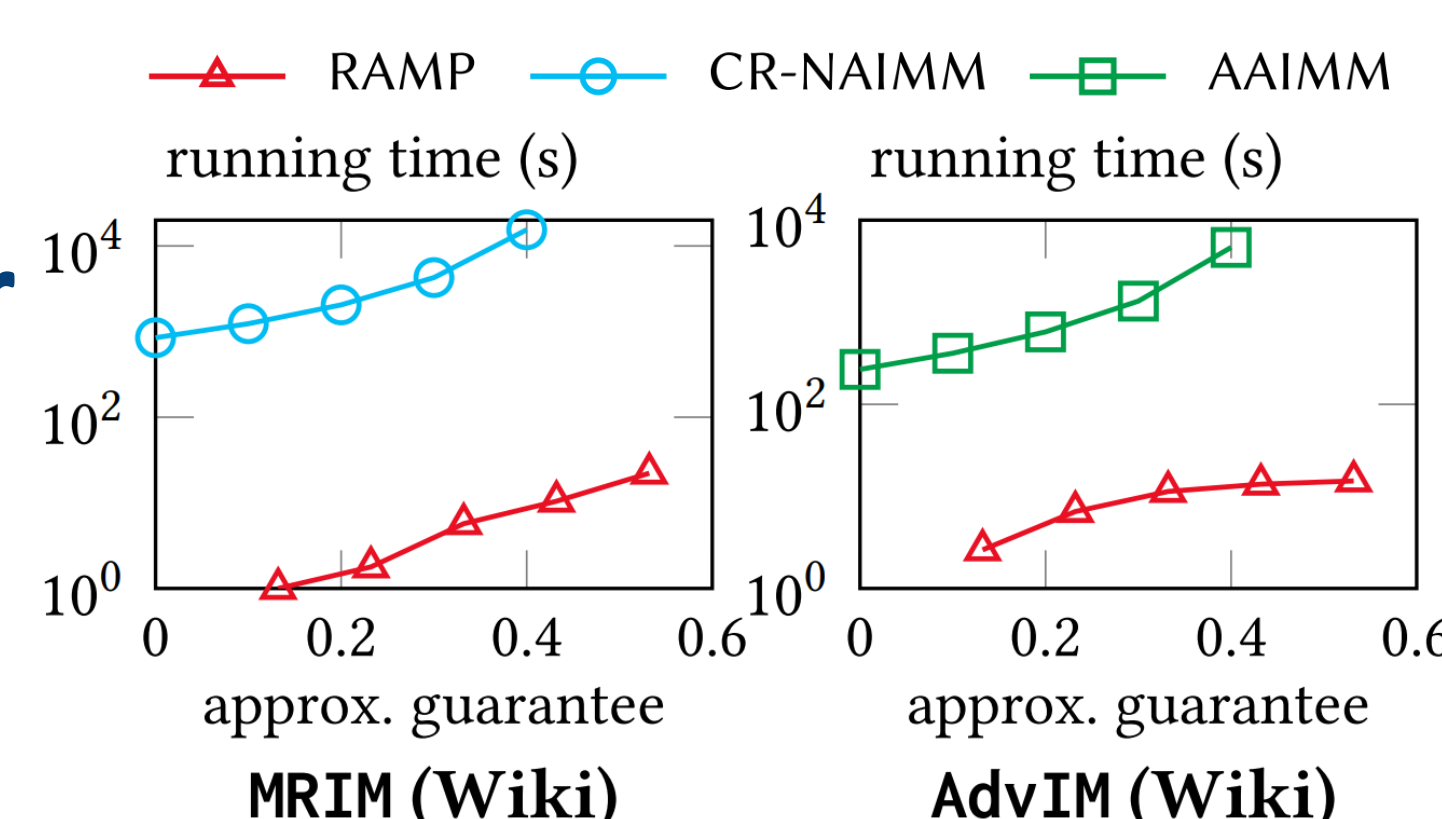
## Experiments

- AMP and its variants outperform all other solvers on 7 public datasets in terms of solution quality**

X-axis: computation resource



- RAMP outperforms other scalable algorithms in terms of running time**



- RAMP outperforms other solutions in actual spread of Tencent e-games**

Algorithm	Spread
Most-Click	7.34M
Greedy	7.36M
RAMP	<b>7.50M</b>