

# Learning Policies for Effective Incentive Allocation in Unknown Social Networks

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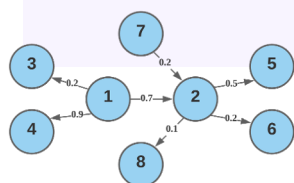


## INTRODUCTION

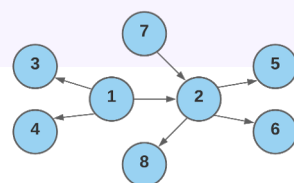
Incentivizing users to take behaviors that are profitable or beneficial to the incentive providers is a crucial problem to many fields. Such processes are computationally modeled as *incentive allocation problem*, where the goal is to incentivize users with effective incentives under a budget limitation, such that the number of users who take the behavior that the incentive provider expects is maximized.

## CHALLENGE

- To design a reasonable incentive structure and the pricing policies.
  - Overpricing the incentive would waste the budget, whereas underpricing could fail to incentivize the user.
  - Users' attributes, such as users' preferences and skill abilities can be unavailable for generating incentives.
- To engage influential users and exploit their social influence to incentivize more users.
  - Recognizing influential users in a social network is difficult in practice.
  - Influence probabilities on edges are unknown.



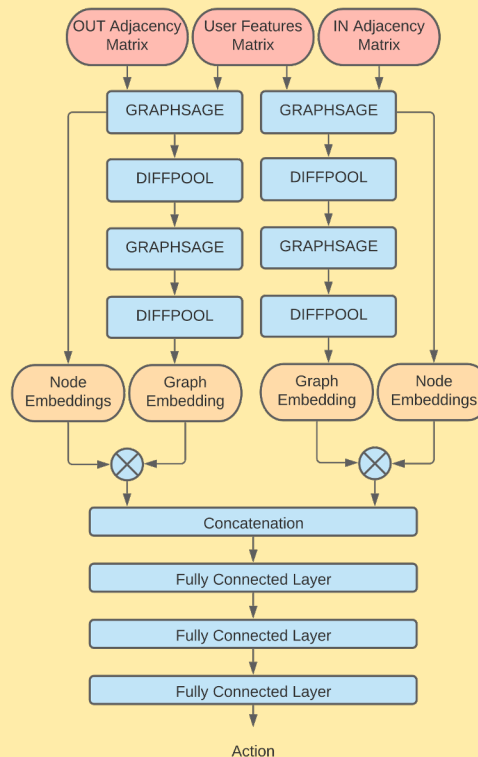
Known Social Network



Unknown Social Network

## METHODOLOGY

We propose a Reinforcement Learning-based framework, named **Geometric Actor Critic (GAC)**, to tackle the aforementioned challenges.

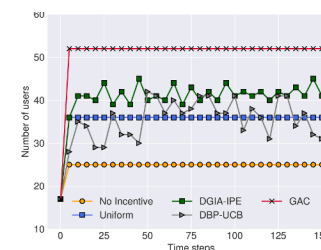


## EXPERIMENTAL SETUP

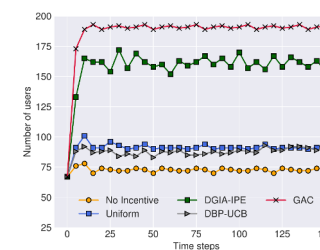
We conduct the simulation-based experiments to evaluate the performance of the proposed GAC comparing with four baseline approaches. Three real-world social network datasets are deployed.

Dataset	#Nodes	#Edges	Avg. Degree
Dolphins	62	159	5.1
Twitter	236	2478	21.0
Wiki-Vote	889	2914	6.6

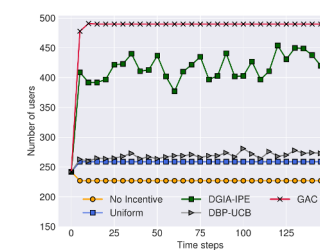
## RESULTS



Dolphins, Budget=3



Twitter, Budget=20



Wiki-Vote, Budget=40

## CONCLUSION

### Pros:

- GAC is able to solve the incentive allocation problem in unknown social networks, where only limited prior knowledge is required.
- GAC learns to represent the network from both global and local perspectives.
- The trained GAC outperforms compared baseline approaches in all three deployed datasets under a fixed budget constraint.

### Cons:

- GAC requires well-trained beforehand to obtain superior performance.
- The cost for training GAC could be exploded with the increasing size of the network, as it requires two adjacency matrices as input.