

# Measuring Friendship Closeness: A Perspective of Social Identity Theory

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## Outline

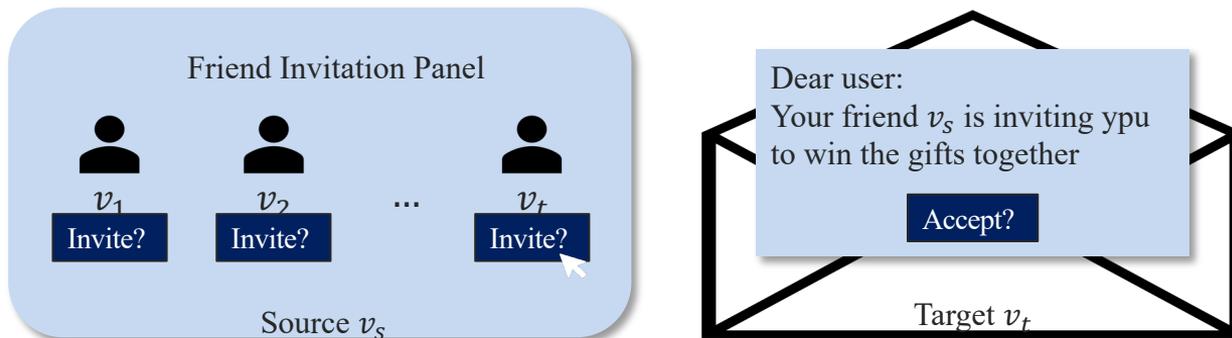
- Problem and Applications
- Existing Works
- Proposed Measures
- Experiments
- Deployments

## Problem Definition

- Given a graph  $G = (V, E)$ 
  - $v_s \in V$ : the user in the social network
  - $(v_s, v_t) \in E$ : two users are friends in the social network
- We aim to measure **TFC** (Topological Friendship Closeness) for each friend pair  $(v_s, v_t) \in E$ .

## Application Scenario

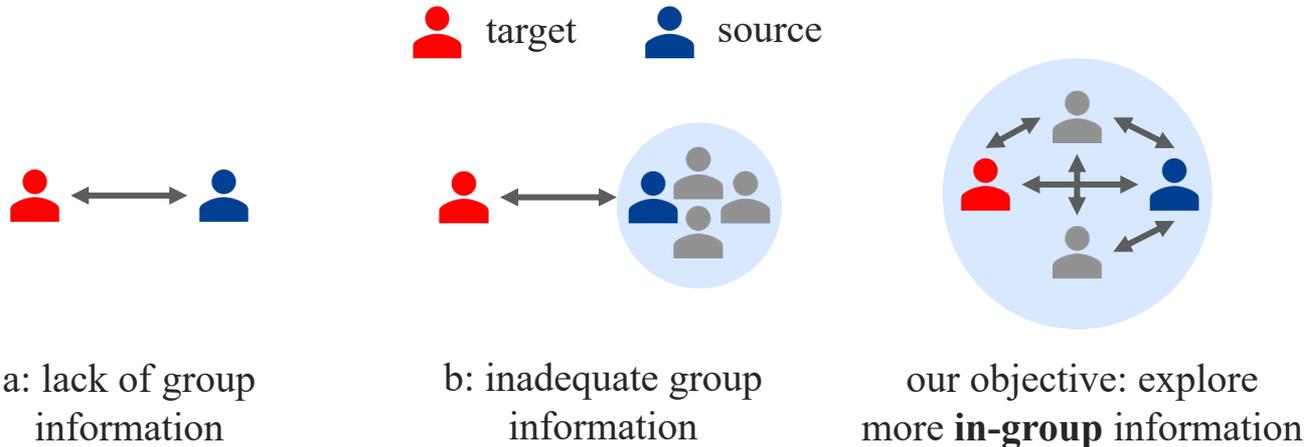
- Friendship-enhancing event in Tencent's Games
  - source  $v_s$ : the user who *sent* the invitation
  - target  $v_t$ : the user who *received* the invitation



- TFC is applied for (i) *user behavior understanding* and (ii) *target recommendation*.

# Existing Solutions

- a. Individual-level measures
  - Tie strength, # common friend
  - Personalized PageRank, similarity between node embeddings
- b. Group-level measures
  - Structural diversity, user-group tie strength



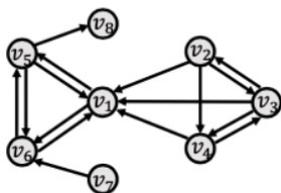
# Proposed Measures: Theory support

- Social identity theory (SIT)
  - The inclination that a target endorses behaviors of users inside the same group is affected by psychological factors.

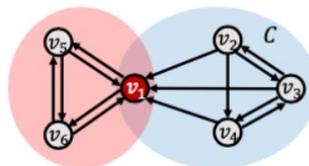
Factor	Meaning
Multi-membership	Number of groups
Inclusiveness	Number of in-group members
Solidarity	$v_t$ 's psychological bond with in-group members
Centrality	Importance of a group in $v_t$ 's cognition
Self-stereotyping	Similarity of $v_t$ and group average in $v_t$ 's cognition
In-group homogeneity	Similarity within a group
Social standing	Social standing of a group

- *Novelty: import psychological factors for TFC measures*
- *Challenge: how to reflect these factors by structural information?*

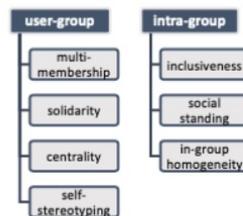
# Proposed Measures: Overview



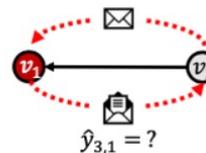
(a) Input graph



(b) Categorization



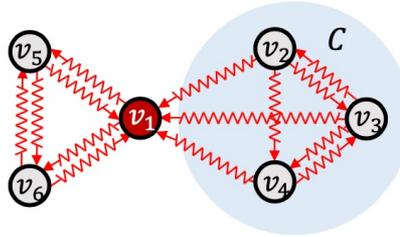
(c) Factor



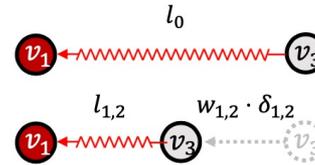
(d) Inference

- Social categorization
  - Candidate group = CC in the ego network of the target.
- **SIT-based measure definition**
  - Define quantitative measures to describe each factor.
- Inclination inference
  - Supervised learning via XGBoost.

# Proposed Measures: Definitions



(a) Local neighborhood



(b) Attractive force

- ( $C$ : the group that the given source and target are in)
- Abstract the edge as a spring
  - tie strength  $w$  as the stiffness constant.
  - similarity  $\delta$  as the displacement.
- **UGT** (solidarity, self-stereotyping)
  - The average attractive force from users in  $C$  to the target.
- **IGT** (ingroup-homogeneity)
  - The average attractive force among users in  $C$ .

# Experiments: Datasets and Setup

**Table 2: Dataset statistics ( $M = 10^6$ ,  $B = 10^9$ ).**

<b>Dataset</b>	<b><math> \mathcal{V} </math></b>	<b><math> \mathcal{E} </math></b>	<b><math> \mathcal{S} </math></b>	<b><math> \mathcal{T} </math></b>
<i>FPS</i>	<i>77.2M</i>	<i>1.1B</i>	<i>33.5M</i>	<i>43.6M</i>
<i>MOBA-A</i>	<i>111.0M</i>	<i>4.5B</i>	<i>111.0M</i>	<i>94.7M</i>
<i>MOBA-B</i>	<i>130.2M</i>	<i>6.5B</i>	<i>120.5M</i>	<i>99.7M</i>

- Individual-level competitors:
  - tie strength (**Tie**); #common friend (**COM**); Personalized PageRank (**PPR**); similarity between Node2vec embeddings (**N2V**)
- Group-level competitors:
  - Structural diversity (**#CC**); user-group tie strength (**GT**); in-group edge density (**GD**)
- UGT,IGT:  $w$ : Tie;  $\delta$ : N2V.

# Experiments: Behavior Prediction

Measure	Adoption						Invitation					
	FPS			MOBA-A			FPS			MOBA-A		
	AUC	Accuracy	F1 score									
Tie	<u>0.7154</u>	<u>0.6965</u>	0.6554	0.6017	<u>0.6021</u>	0.3958	0.6072	<u>0.5985</u>	0.4607	0.5361	0.5353	0.2200
COM	0.5488	0.5538	0.5615	0.5667	0.5576	<u>0.6219</u>	0.5456	0.5323	0.4674	0.5332	0.5281	<u>0.5565</u>
PPR	0.6565	0.6036	0.5596	0.5562	0.5388	0.4447	<u>0.6289</u>	0.5972	<u>0.5786</u>	<u>0.5846</u>	0.5589	0.5467
N2V(cos)	0.6976	0.6610	<u>0.7171</u>	0.5808	0.5626	0.5420	0.5608	0.5537	0.5683	0.5770	<u>0.5630</u>	0.5426
N2V(euc)	0.7076	0.6652	0.7091	0.5664	0.5566	0.5390	0.5679	0.5588	0.5628	0.5739	0.5585	0.5375
#CC	0.6153	0.5897	0.5378	0.5452	0.5288	0.4136	0.6091	0.5820	0.5392	0.5790	0.5551	0.5662
GT	0.6985	0.6572	0.6004	<u>0.6295</u>	0.5959	0.5777	0.5738	0.5652	0.3988	0.5397	0.5297	0.4875
GD	0.6077	0.5736	0.5269	0.6039	0.5728	0.5908	0.5811	0.5507	0.4490	0.5674	0.5508	0.4991
SIT	<u>0.7995</u>	<u>0.7206</u>	<u>0.7350</u>	<u>0.7410</u>	<u>0.6780</u>	<u>0.6638</u>	<u>0.7307</u>	<u>0.6719</u>	<u>0.6754</u>	<u>0.6550</u>	<u>0.6086</u>	<u>0.6047</u>

- Predictions for target adoption and source invitation
- **SIT outperforms all competitors** on two datasets in terms of two prediction tasks and three evaluation metrics.

# Experiments: Conversion analysis

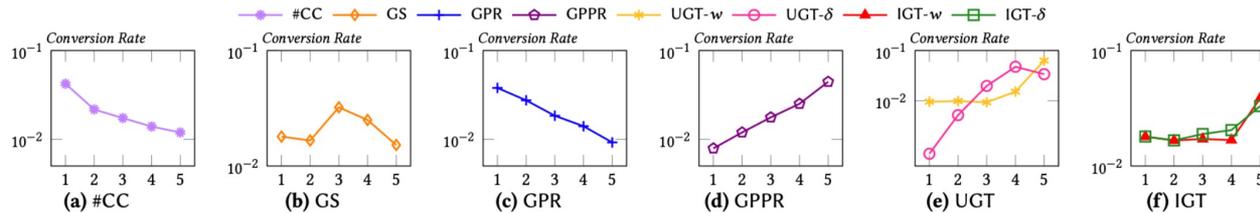


Figure 4: Conversion probability of adoption behaviors conditioned on each SIT-based measure in FPS.

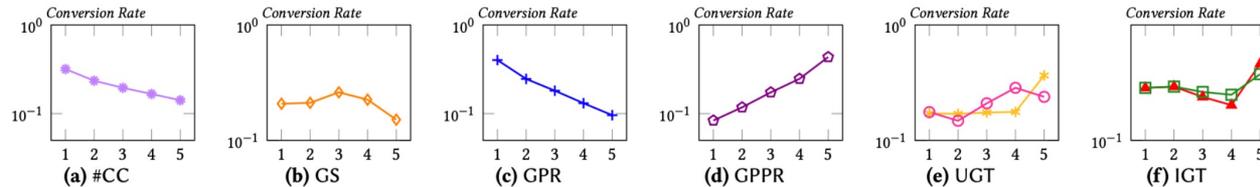


Figure 5: Conversion probability of invitation behaviors conditioned on each SIT-based measure in FPS.

- Conversion rate: fraction of inviters/adopters in given rank.
- The conversion is more sensitive to **#CC**, **GPR**, **GPPR**, **UGT**

# Deployments: Target Recommendation

**Table 7: Online performance in MOBA-A.**

Measure	Tie	COM	PPR	N2V(euc)	SIT
<b>E2E rate</b>	0.1018	0.0958	<u>0.1066</u>	0.0739	0.1431

**Table 8: Online performance in MOBA-B.**

Measure	Tie	PPR	SIT
<b>E2E rate</b>	0.1152	<u>0.1218</u>	0.1384

- E2E rate:  $\frac{\text{target friends adopting the invitations}}{\text{source users seeing the event}}$ .
- **SIT improves the best competitor by up to 34.2%.**

## Conclusion

- We propose six new TFC measures based on the social identity theory.
- The SIT-based measures are sensitive to user behavior conversion.
- The SIT-based measures can experimentally outperform the competitors.
- The SIT-based measures have been deployed to more than 10 friendship-enhancing events up till now.



**Thank You!**