## Trust-based Sybil Node Detection with Robust Seed Selection and Graph Pruning on SNS

### Shuichiro Haruta, Kentaroh Toyoda, Iwao Sasase

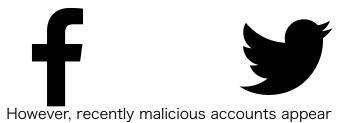
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Most of us use SNS such as Facebook or Twitter.

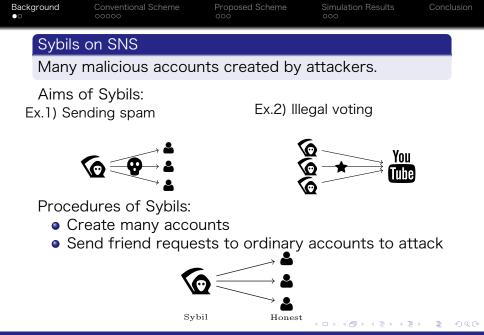


### Motivation for our research

Detect malicious account called "Sybil", effectively

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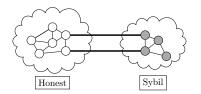


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Background ○●	Conventional Scheme	Proposed Scheme	Simulation Results	Conclusion
System	Model			

- Non-directional graph SNS (e.g., Facebook)
- We call ordinary nodes as "honest nodes"
- Sybils can make friends among Sybils
  - Sybils pretend to be honest nodes
- Sybils try to make friends with honest nodes
  - Acceptance ratio of friend request is relatively small e.g., friend request from stranger



The relationships among Sybils and honests are called "**AE (Attack Edge)**".

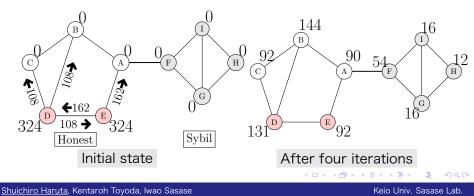
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Background oo	Conventional Scheme ●○○○○	Proposed Scheme	Simulation Results	Conclusion

## Conventional Scheme 1 : SybilRank[1]

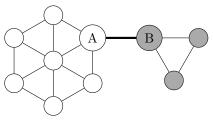
- Sybil detection scheme Cao et al. proposed
- repeatedly distribute trust values of non-Sybil nodes (honest seeds)to their neighbors
- honest nodes tend to get higher trust values
  - since trust values Sybils get are only via AE



Background oo	Conventional Scheme	Proposed Scheme 000	Simulation Results	Conclusion

### Conventional Scheme 2 : Graph pruning scheme [2]

- Zhang et al. proposed to enhance SybilRank's detection accuracy
- Prior to SybilRank, doubtful edges are pruned
- Prune the relationships between two nodes with few common friends



The number of common friends between A and B is 0. This thick edge is pruned.

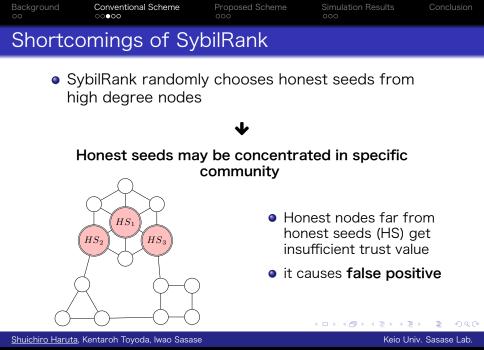
Honest



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Background oo	Conventional Scheme ○○○●○	Proposed Scheme	Simulation Results	Conclusion
Distribu	ution of high	dearee noc	les	

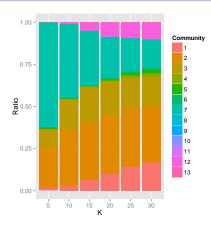


Figure: The ratio of nodes with high degree in each community.

Figure shows communities where the top K% highest degree nodes belong. (Dataset is Facebook)

Example: K = 5

Most of high degree nodes belong to 4 out of 13 communities.

### Conventional Seed Selecting …

Honest seed tends to concentrate in specific community

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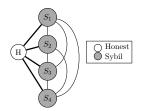
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Background	Conventional Scheme	Proposed Scheme	Simulation Results	Conclusion
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## Shortcomings of Graph Pruning Scheme

 Conventional graph pruning scheme[2] prunes AE based on the number of common friends

## Attacker can easily increase the number of common friends and in that case, pruning AE is difficult

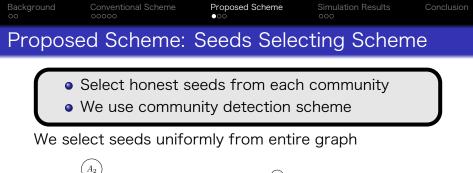


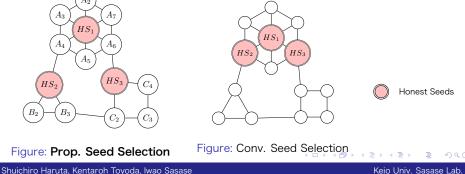
If an attacker creates relationships, the number of common friends between honest and Sybil is increased.

The common friends of H and  $S_1$  are  $S_2, S_3, S_4$ .

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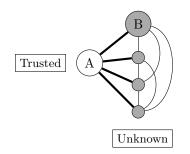
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### Proposed Scheme: Graph Pruning Based on Density of Relationships

- Consider how many trusted relationships exist
- Prune relationships between a trusted node and an untrusted node with probability



- Node A has already been regarded as trusted
- Node B is unknown and has 4 friends
- <sup>1</sup>/<sub>4</sub> of node B's friends are trusted

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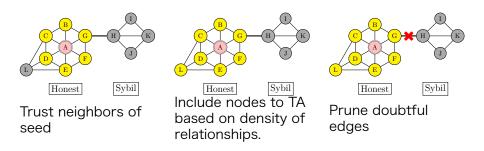
• Pruning probability between A and B is  $P_{prune}(A,B) = 1 - \frac{1}{4} = \frac{3}{4}$ 

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Area). We calculate TA with three steps.



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Background oo	Conventional Scheme	Proposed Scheme	Simulation Results	Conclusion

## Simulation Settings : Attack Scenario

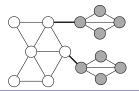
Attack scenario 1

- Conventional attack scenario
- Add totally 200 AEs from randomly chosen Sybils to Honest nodes

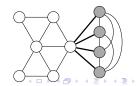
Attack scenario 2

- Add totally 200 AEs in below style
- Honest nodes that accept friend requests from strangers are limited

In the simulation, we intensively increase the common number of friends



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Simulat	ion Setting :	2		

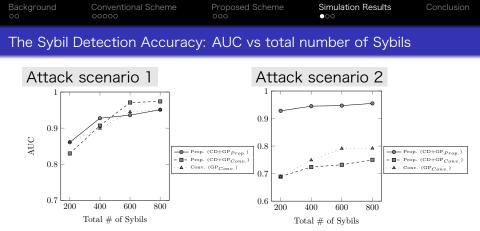
- Evaluating the AUC (Area Under Curve) of ROC (Receive Operating Characteristic) curve[3].
- AUC indicates ability of a classification algorithm and ranges 0 to 1.

parameter	value
dataset	Facebook [4]
number of nodes	4039
number of edges	88234
graph model	BA model[5]
n <sub>att</sub>	5
K	10
simulation tools	R with igraph[6]

Table: Parameter values used in the simulation.

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- In the attack scenario 1, the both proposed schemes achieve almost same accuracy with the conventional scheme
- In the attack scenario 2, the proposed scheme with community detection and our graph pruning considerably improve AUC against the conventional scheme

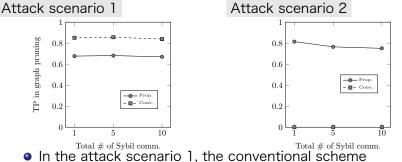
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True Positive Rate in Graph Pruning vs Total number of Sybil Comm.

We define  $\ensuremath{\text{TP}}$  (True Positive) rate as how accurately AEs are pruned



- effectively prunes AE. The proposed scheme degrades.
- In the attack scenario 2, while the conventional graph pruning scheme cannot prune AE, the proposed scheme prunes AE with high TP.

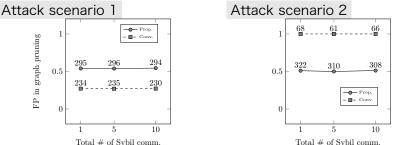
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False Positive in Rate in Graph Pruning vs Total of Sybil comm.

We define **FP** (False Positive) rate as **the ratio that the non-AE are inaccurately pruned**<sup>1</sup>.



 In the attack scenario 2, the conventional scheme cannot accurately prune AE.

• Although FP seems to be high in both scheme, it can be acceptable since the # of entire edges is about 90,000.

The numeric above a point is the average number of pruned 💿 👁

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Background oo	Conventional Scheme	Proposed Scheme	Simulation Results	Conclusion
Conclu	sion			

- We have proposed a Sybil nodes detection scheme with robust seed selection and graph pruning on SNS
- The first proposal is seed selecting scheme by detecting communities and choosing seeds from them
- The second one is a graph pruning scheme that considers trusted area
- By computer simulation, we show that our scheme achieves high detection accuracy even if attackers make a large number of common friends

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Background	Conventional Scheme	Proposed Scheme	Simulation Results	Conclusion

## Citations

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Proposed Scheme

Simulation Results

Conclusion

# Thank you for listening.

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