

# Exposing Insecure Configurations of Network Session and Permission Graphs

Final talk for the Master's Thesis by

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## Introduction

### **Problem:**

Lateral Movement in Windows networks

### **Goal:**

Prevention of Identity Snowball Attacks

### **Contribution:**

Solution based on graph-theoretic metrics

# Agenda

Problem and Motivation

Problem Analysis

Exposing Insecure Configurations

Design and Implementation

Evaluation

Related Work

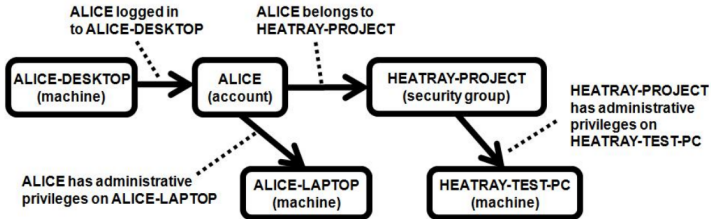


Figure 1: Identity Snowball Attack [1]



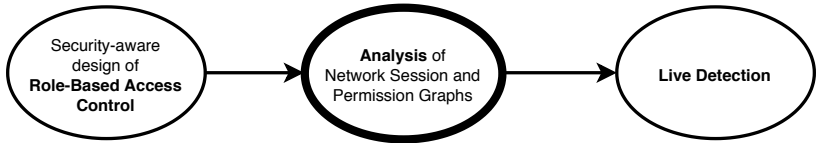


Figure 3: Possible points in time

⇒ Analysis before the actual attack

# Problem Analysis

## Problem Statement

### Analysis of Network Session and Permission Graphs

⇒ We want to ...

- ... **define undesired graph configurations**, deduced from real-world issues.
- ... **detect undesired configurations** within the graph.
- ... be able to **scale** the solution even **for large graphs**.
- ... apply the solutions for network session and permission graphs, not only Active Directory.

## Users, Groups, Computers

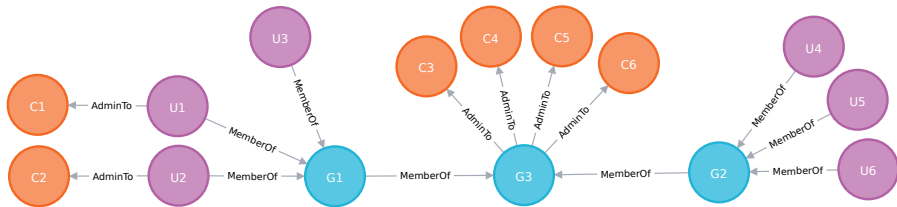


Figure 4: Minimalist sample graph



# Exposing Insecure Configurations

## Degree Centrality

"Direct neighbors of the attacker"

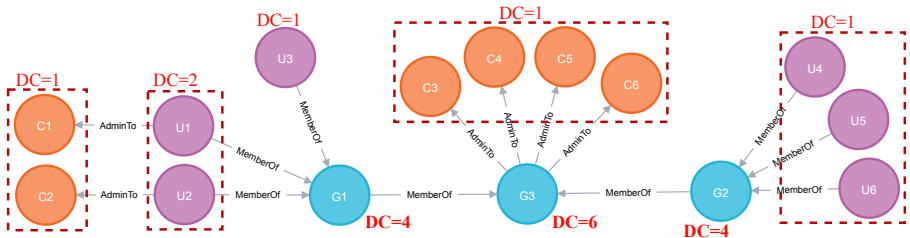


Figure 5: Sample graph with Degree Centrality (DC)

# Exposing Insecure Configurations

## Betweenness Centrality

"Nodes on the attacker's path"

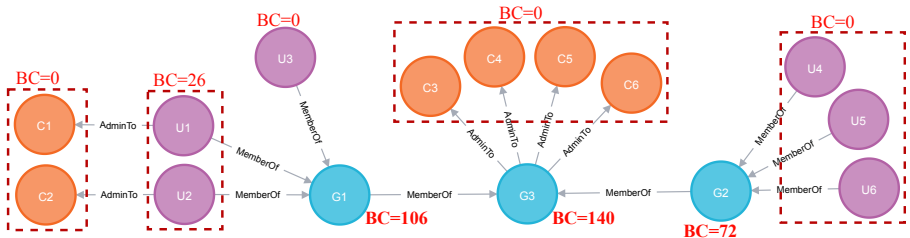


Figure 6: Sample graph with Betweenness Centrality (BC)

# Exposing Insecure Configurations

## Closeness Centrality

"Closeness to the attacker"

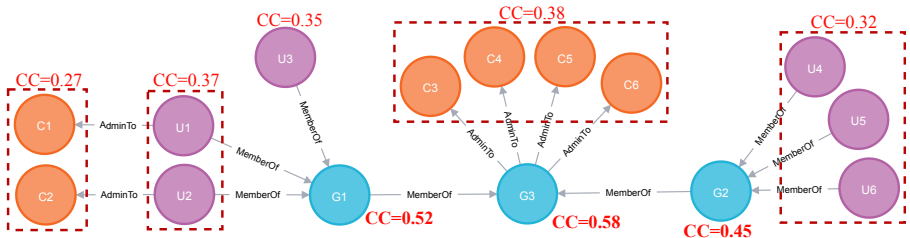


Figure 7: Sample graph with Close Centrality (CC)

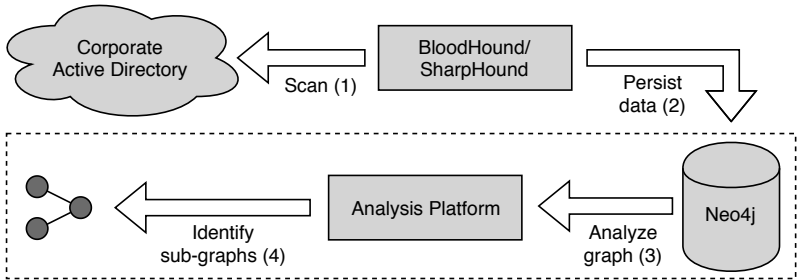


Figure 8: Technical perspective of the investigation process

- Simulated environment
  1. Randomly generate graphs
  2. Inject undesired configurations
  3. Find undesired configurations
  
- Real-world environment
  1. Collect network session and permission graph
  2. Find undesired configurations
  3. Discussion about meaningfulness of findings

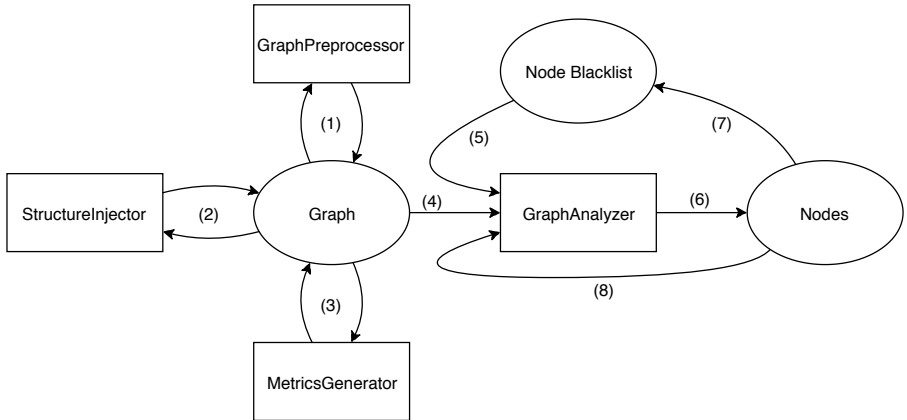


Figure 9: Analysis platform

## Injection of Undesired Configuration

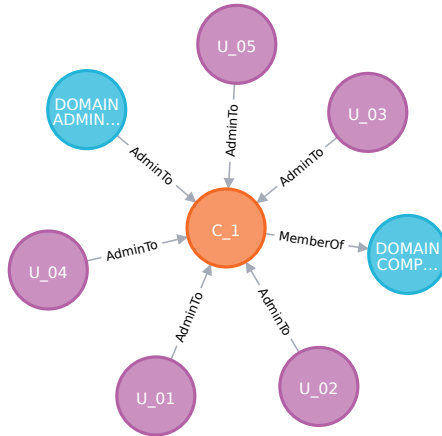


Figure 10: Undesired configuration: One computer with five admin users

## Node with Betweenness Centrality $BC = 7660$

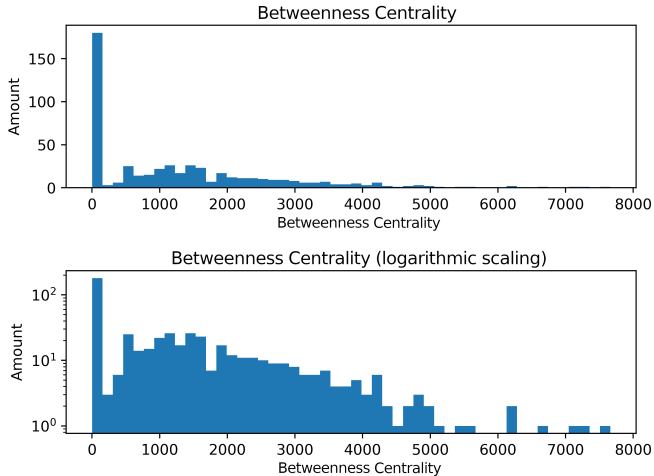


Figure 11: Betweenness Centrality of a graph (injected one computer with ten admin users)



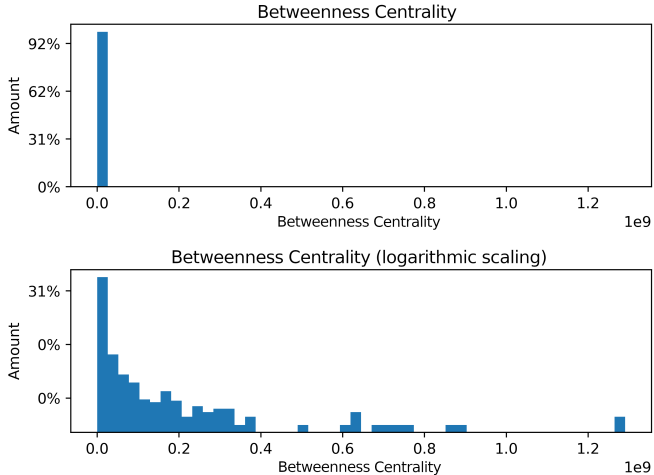


Figure 12: Betweenness Centrality of a real AD network

# Evaluation

## Results

- Found common AD groups  
e.g. DOMAIN USERS, DOMAIN ADMINS  
⇒ Intentionally blacklisted
- Found undesired sub-structures (intentionally injected)  
e.g. users with many administrative permissions
- Found odd configurations  
e.g. users in many groups
- PageRank may be a comprehensive metric

## Related Work

### Comparison to Other Approaches

- **Automated Analysis:** Heat-ray by Microsoft [1]  
⇒ Limited usage of graph metrics (only one variation of Betweenness Centrality)
- **Manual Analysis:** BloodHound AD [3]  
⇒ Manual investigation per path, user, group or machine is necessary

- **Problems** with existing solutions to prevent Identity Snowball Attacks
- **Analysis** of Network Session and Permission Graphs
- **Novel centralized approach** for detection with centrality metrics
- **Evaluation** in a real-world and a simulated environment

- [1] John Dunagan, Alice X. Zheng, and Daniel R. Simon.  
Heat-ray: Combating identity snowball attacks using machine learning, combinatorial optimization and attack graphs.  
In *Proceedings of the ACM SIGOPS 22Nd Symposium on Operating Systems Principles, SOSP '09*, pages 305–320. ACM.
- [2] Didier Stevens.  
Windows Credential Guard & Mimikatz.  
<https://blog.nviso.be/2018/01/09/windows-credential-guard-mimikatz/>.
- [3] Andrew Robbins, Rohan Vazarkar, and Will Schroeder.  
BloodHound: Six Degrees of Domain Admin.  
<https://github.com/BloodHoundAD/BloodHound>.

# Additional Ideas (Backup Slide)

## Page Rank

Importance/Influence of a node

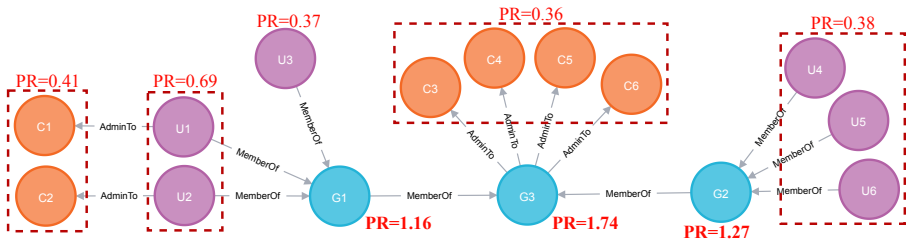


Figure 13: Example graph with Page Rank (PR)