



# The Automated Design of Network Graph Algorithms with Applications in Cybersecurity

Aaron Scott Pope, Ph.D.

A-4: Advanced Research in Cyber Systems

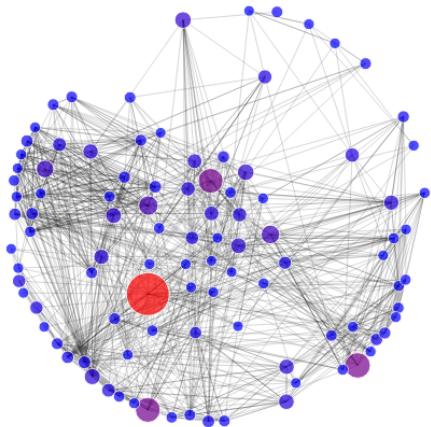
[apope@lanl.gov](mailto:apope@lanl.gov)

LA-UR-24-23429



Managed by Triad National Security, LLC, for the U.S. Department of Energy's NNSA.

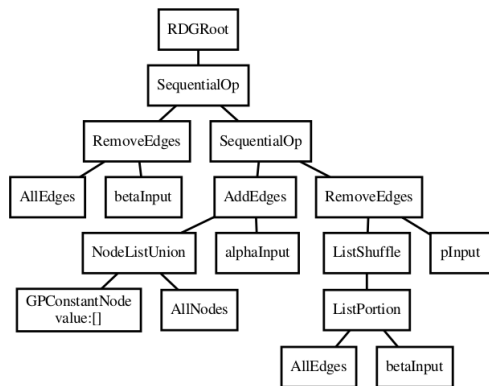
# Automated Design of Network Algorithms



Use automated heuristic search techniques to improve off-the-shelf algorithm performance for specific applications.

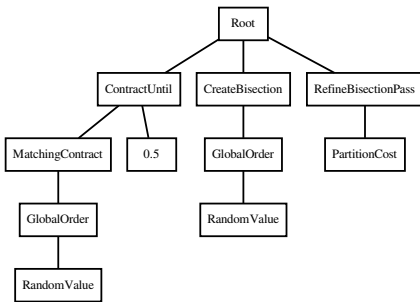
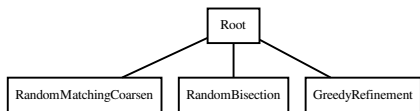
- Complex network applications typically rely on approximation heuristics for efficiency
- These heuristics can be tailored to leverage problem characteristics for an application to improve accuracy, speed, etc.
- Doing this manually can be expensive and time-consuming
- The optimization can be automated using bio-inspired search techniques

# Automated Heuristic Optimization



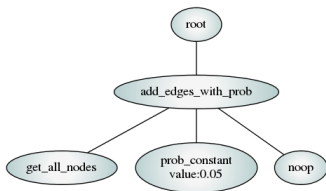
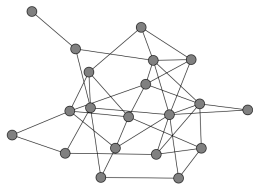
- Extract functionality from related algorithms to build a set of “algorithmic primitives”
- Construct entire algorithms from primitives (e.g., parse tree)
- Measure algorithm quality based on the application
- Use heuristic search algorithm (e.g., genetic programming) to optimize algorithm structure

# Heuristic Search Scalability for Real-world Applications



- Granularity level of primitive operations has a huge impact on scalability
- Automated primitive granularity control can help address scaling issues for heuristic searches on complex real-world problems

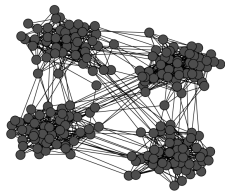
## Application: Data-Driven Network Model Generation



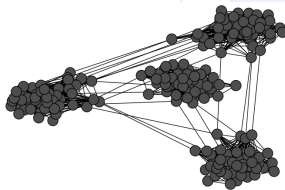
- Automate the design of algorithms for generating random networks with characteristics of interest
  - Investigate network properties
  - Make predictions
  - Generate synthetic data
- Can be trained on a single or multi-objective definition of graph quality:
  - Similarity to sample networks
  - Graph or application specific metrics

# Static Modeling: Reproducing Random Community Graphs

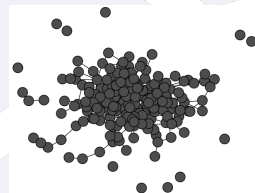
## Actual Graph



## Granular Model Generation

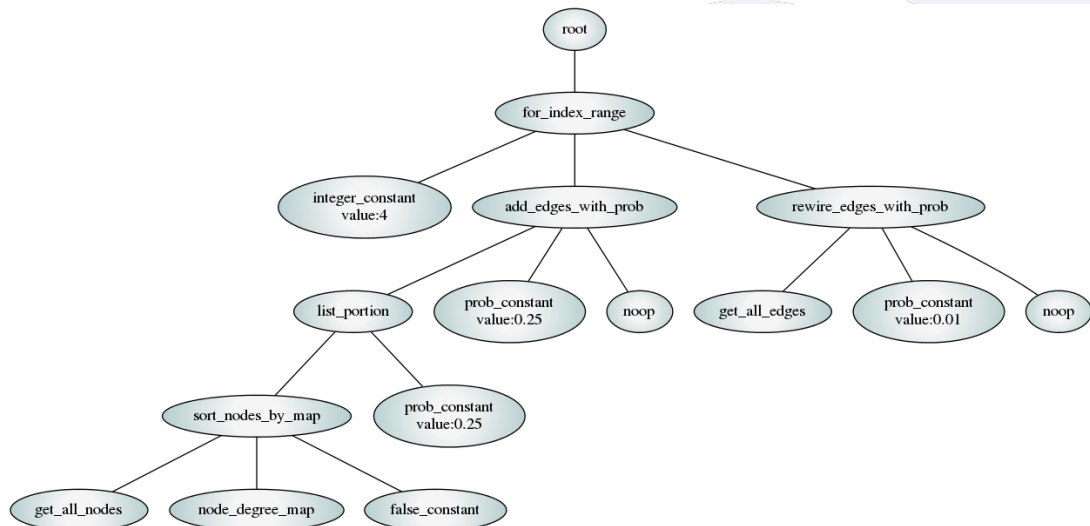


## Naive Model Fitting



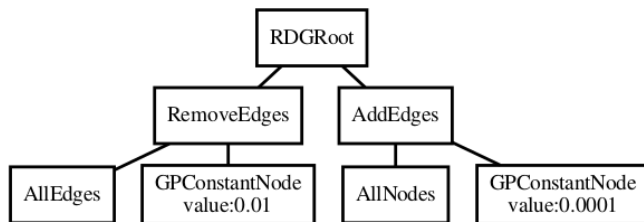
Similarity Metric	Granular		Comparison	Naive	
	Mean	$\sigma$		Mean	$\sigma$
Degree	0.436	0.075	<	0.458	0.055
Betweenness	0.209	0.105	<	0.320	0.126
PageRank	0.127	0.029	<	0.150	0.036

# Static Modeling: Random Community Network Generator



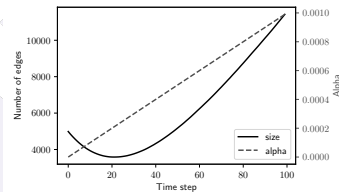
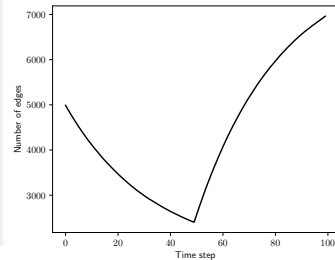
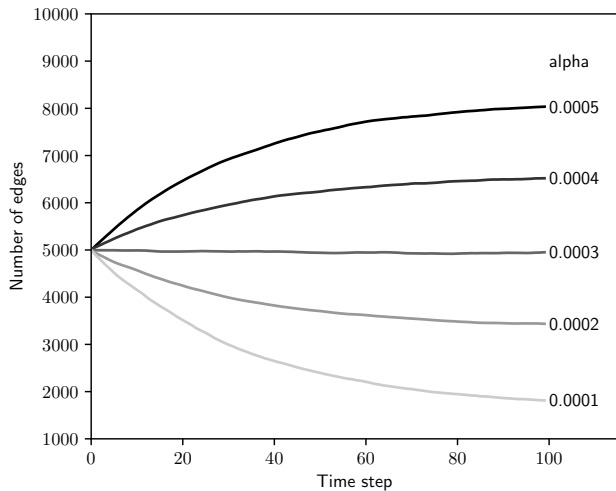
## Application: Data-Driven Dynamic Network Modeling

- Extends model generation to dynamic networks
- Generate algorithm that “updates” the network at each time step
- Learn to mimic target network behavior

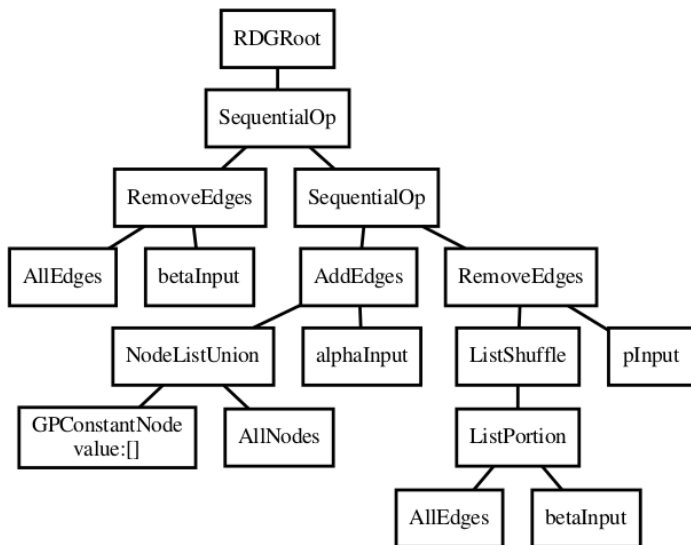




# Dynamic Modeling: Dynamic Erdős-Rényi Model



## Dynamic Modeling: Example Generated Algorithm

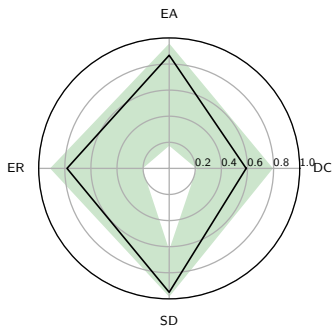


# Dynamic Modeling: Real-World Enterprise Network Behavior

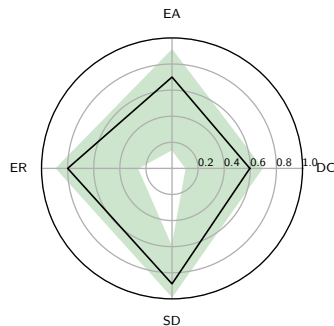
Model activity of Los Alamos National Laboratory (LANL) enterprise computer network

- User-computer authentication events
- NetFlow communication sessions between pairs of computers

## Authentication

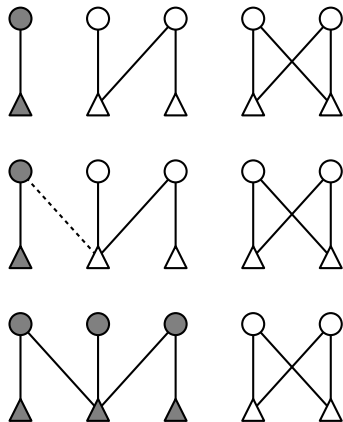


## NetFlow



## Application: Automated Network Security Metric Design

### Attack Simulation



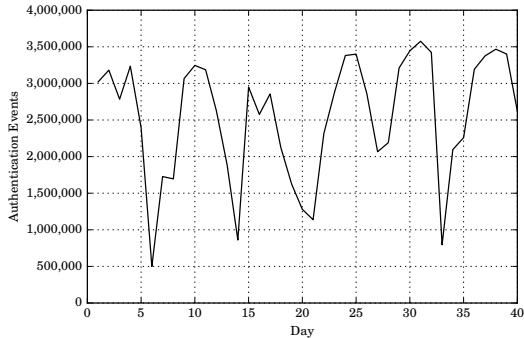
- Automated the design of network security metrics for large networks
- Trained on real or simulated event data
- Simulated attacks using real LANL network data

### LANL Authentication Dataset Details

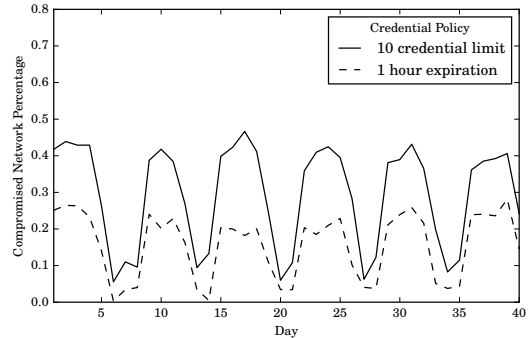
Unique Users	10,044
Unique Computers	15,779
Unique (User, Computer) Pairs	124,020
Total Authentication Events	101,918,344
Average Daily Authentication Events	2,547,959

# Application: Automated Network Security Metric Design

## Daily Authentication Events

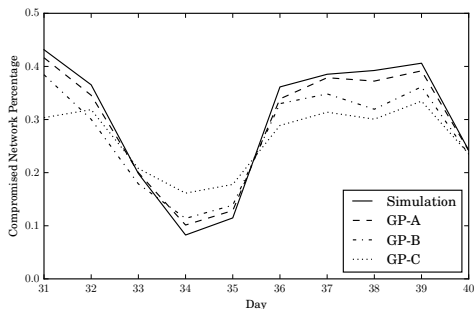


## Simulation Results

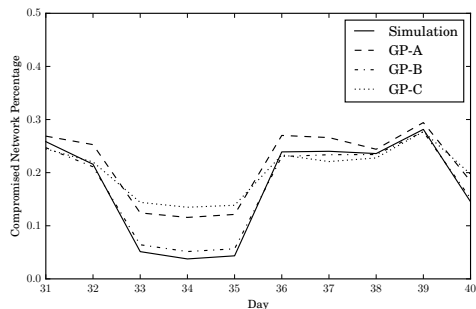


# Application: Automated Network Security Metric Design

## 10 credential limit policy



## 1 hour expiration policy



## Application: Tailored Anomaly Detection Heuristics

- Automated the design of novel link prediction heuristics for anomaly detection
- Link prediction: predict the existence of a relationship or rank relationships by likelihood
- Relies on historical or contextual information
- Predictive performance can be optimized by tailoring for an application

## Tailored Link Prediction Heuristics: Experiment

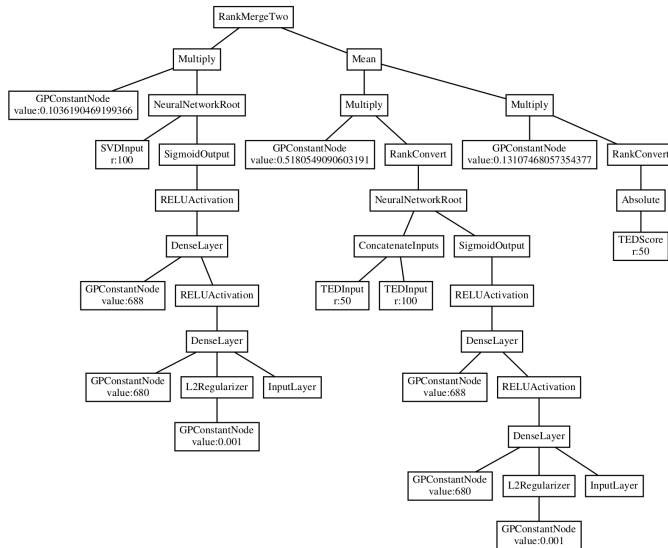
- Data from the network at Los Alamos National Laboratory
  - User-Process (UP), Computer-Process (CP), NetFlow (NF)
- Differentiate legitimate activity from anomalies
  - Positive “new” links
  - Randomly generated negative links
- Use heuristic to calculate scores for a set of input links
- Fitness: area under ROC curve (AUC)
- $AUC \in [0, 1]$ , maximized when positive and negative samples are clearly differentiated by scores



## Results

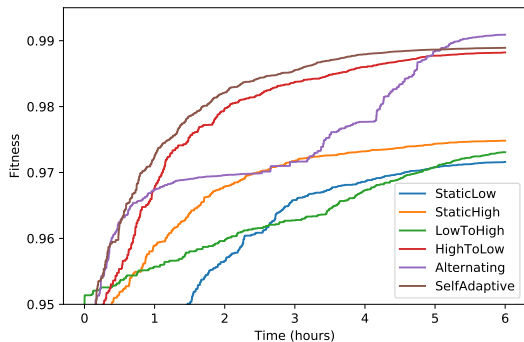
Method	Application		
	UP	CP	NF
NP	0.76963	0.74226	0.52967
TSVD	0.94186	0.90334	0.92936
TED	0.97478	0.97697	0.92390
NN	0.98725	0.98661	<b>0.98836</b>
GP-UP	<b>0.99066</b>	<b>0.98718</b>	0.98051
GP-CP	<b>0.98897</b>	<b>0.98996</b>	<b>0.99090</b>
GP-NF	<b>0.98867</b>	<b>0.98874</b>	<b>0.99241</b>

# Tailored Link Prediction: Generated Heuristic

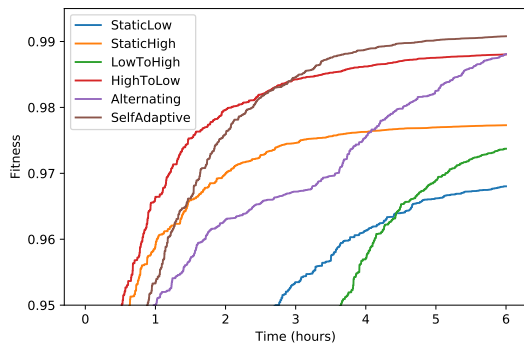


# Tailored Link Prediction: Dynamic Granularity Control

## User-Process

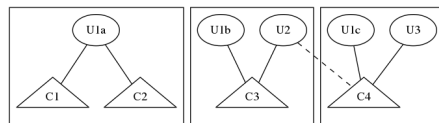
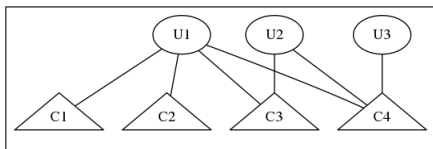


## Computer-Process



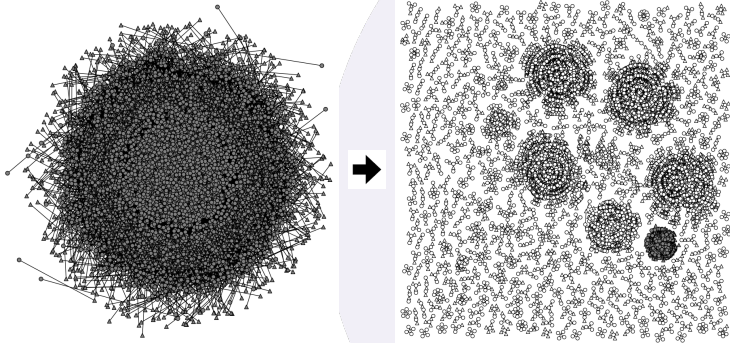
## Application: Network Segmentation Algorithms

- Automate the segmentation of a network to limit adversarial traversal using stolen credentials
- Reduce the size of connected components within the network by:
  - Revoking a user's access to a computer to remove a path
  - Split a user into multiple accounts (different credentials)
- Minimize changes to reduce impact on user productivity



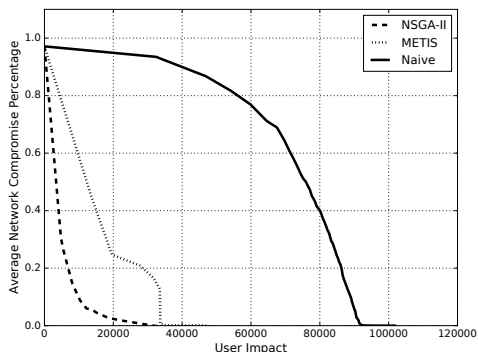
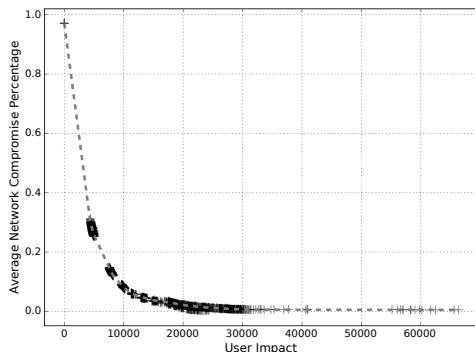
# Application: Network Segmentation Algorithms

Segmenting LANL network bipartite authentication graph (BAG)



# Application: Network Segmentation Algorithms

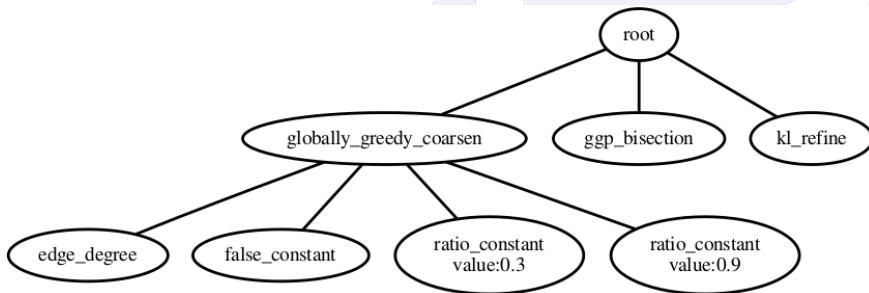
## BAG Partitioning Results



- 1-2 orders of magnitude lower user impact compared to traditional graph partitioning
- Significant reduction in network vulnerability to intrusion

## Application: Design of Network Segmentation Algorithms

Leverage heuristic search to automate the design and optimization of multi-level graph partitioning algorithms that are tailored to specific applications

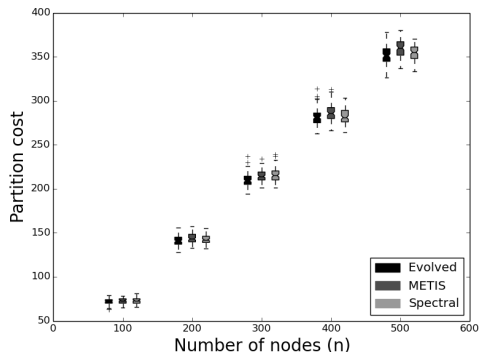


# Application: Design of Network Segmentation Algorithms

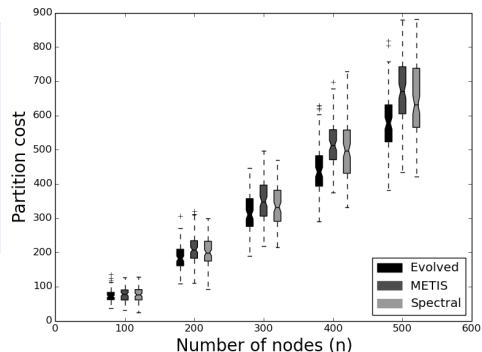
Target graph classes:

- Random graph models (Erdős-Rényi and Barabási-Albert)
- Los Alamos National Laboratory (LANL) authentication graphs

## Barabási-Albert

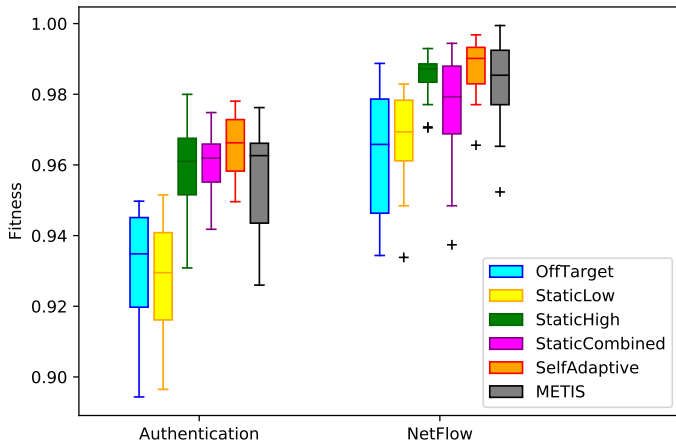


## LANL network



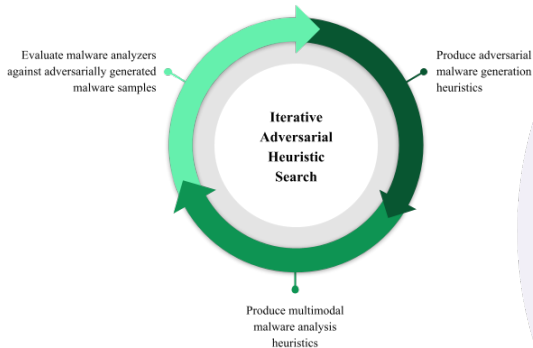


# Network Segmentation: Dynamic Granularity Control



Authentication	
OffTarget	0.93089
StaticLow	0.92903
StaticHigh	0.95844
StaticCombined	0.96001
SelfAdaptive	<b>0.96510</b>
METIS	0.95539
NetFlow	
OffTarget	0.96292
StaticLow	0.96839
StaticHigh	0.98556
StaticCombined	0.97700
SelfAdaptive	<b>0.98787</b>
METIS	0.98374

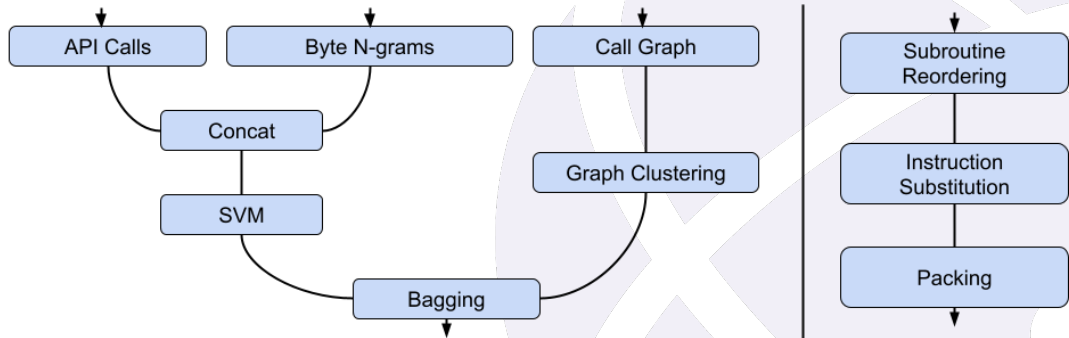
# Proposed Work: Automated Algorithm Design for Adversarial Malware Analysis



Design and optimize novel algorithms for detecting and classifying malicious software

- Machine-learning based malware analyzers can be easy to defeat with simple obfuscation methods
- Automate the design of both malware analyzers and adversarial malware generators
- Use competitive co-evolution to train robust malware classifiers

# Proposed Work: Automated Algorithm Design for Adversarial Malware Analysis



## Summary

Bio-inspired heuristic search techniques can be used to automate the design and optimization of application-tailored algorithms. Demonstrated on:

- Complex network modeling, both static and dynamic
- Network segmentation
- Anomaly detection using link prediction
- Novel network security metrics
- Co-evolving attacker and defender strategies
- Proposed: Adversarial malware analysis

# Questions?

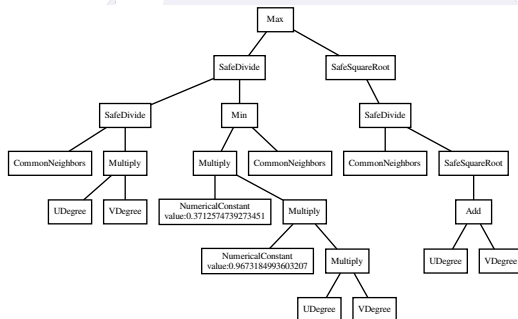
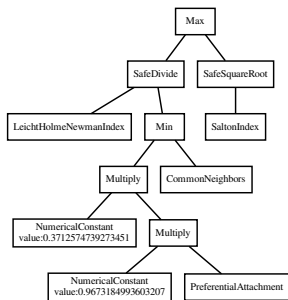
## Dynamic Primitive Granularity Control: Motivation

- Conventionally, primitive operation set is decided a priori
- Proper construction of set is crucial to heuristic search
- Functionality can be implemented at different levels of abstraction or granularity
- Complex, high-level operations:
  - Leverage more domain knowledge
  - Improve early results
  - Limit search flexibility to fine-tune
- Basic, low-level operations:
  - Allow greater algorithmic expressiveness
  - Dramatically increase search space
  - Requires “reinventing the wheel”

## Dynamic Primitive Granularity Control: Approach

- Implement operations at multiple granularity levels
- Construct high-level “macro” primitives from basic operations
- Granularity level can be set dynamically throughout search
- Controls operations available to variation mechanics
- Macro primitives can be decomposed into basic components

# Dynamic Primitive Granularity Control: Example



# Dynamic Primitive Granularity Control

## Dynamic Granularity Control Schemes:

**StaticLow:** low throughout evolution

**StaticHigh:** high throughout evolution

**LowToHigh:** low initially, change to high at midpoint

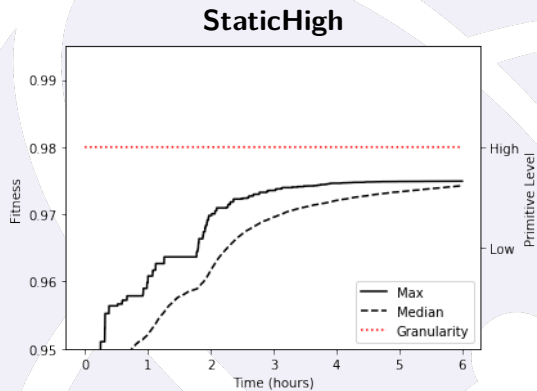
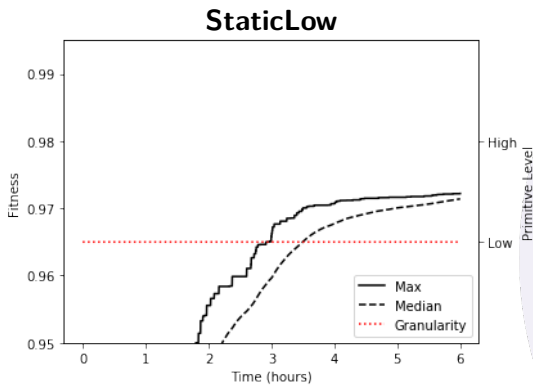
**HighToLow:** high initially, change to low at midpoint

**Alternating:** random initially, alternate on convergence

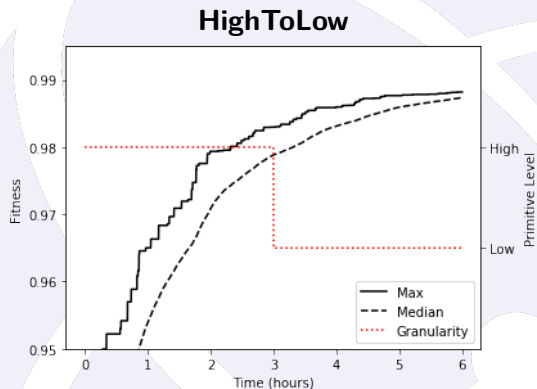
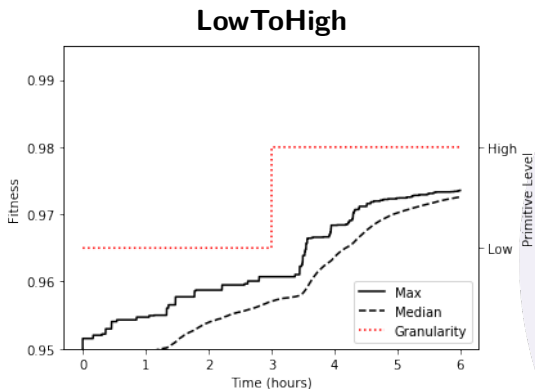
**SelfAdaptive:** self-adaptive granularity level



# Tailored Link Prediction: Dynamic Granularity Control

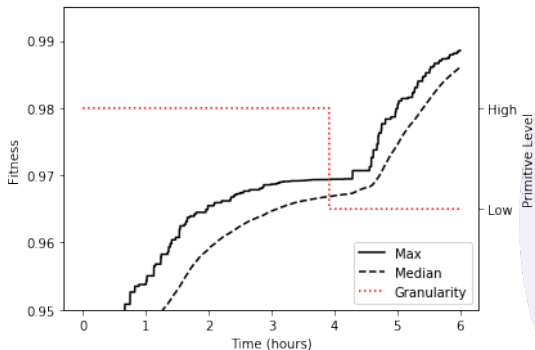


# Tailored Link Prediction: Dynamic Granularity Control

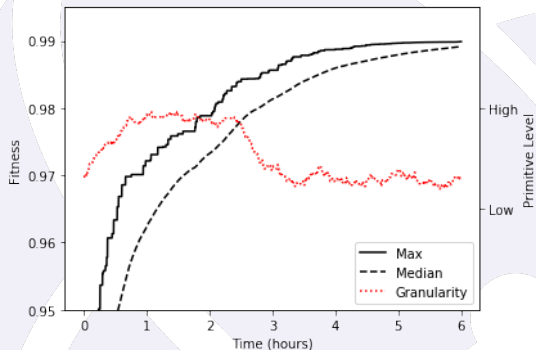


# Tailored Link Prediction: Dynamic Granularity Control

## Alternating



## SelfAdaptive



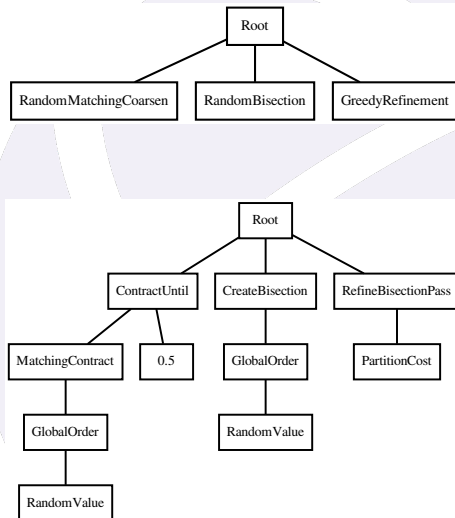
## Tailored Link Prediction: Dynamic Granularity Control

Link Prediction Accuracy

Method	Application		
	UP	CP	NF
Ensemble	0.98757	0.98734	0.9884
Best-UP	——	0.97995	0.98133
Best-CP	0.98277	——	0.97816
Best-NF	0.98518	0.98098	——
StaticLow	0.97269	0.97005	0.9296
StaticHigh	0.975	0.97748	0.94082
LowToHigh	0.97428	0.97625	0.95065
HighToLow	0.98863	0.98835	0.9895
Alternating	<b>0.9911</b>	<b>0.99019</b>	0.98343
SelfAdaptive	0.98906	<b>0.99106</b>	<b>0.99285</b>

# Self-Adaptive Granularity Control for Network Segmentation

- Evolution of MLP heuristics can be improved using dynamic primitive granularity control
- Leverage self-adaptive control scheme
- Target real-world networks for improving security through segmentation



## Self-Adaptive Granularity Control for Network Segmentation

<b>Authentication</b>	
Unique users	9,924
Unique computers	14,822
Unique user-computer pairs	106,693

<b>NetFlow</b>	
Unique devices	60,185
Unique communication pairs	1,136,854

- Segmenting **Authentication** graphs revokes user-computer access to limit traversal of insider or intruder with stolen credentials
- Segmenting **NetFlow** graphs identifies low-cost plans for separating network domains or placing intrusion detection monitors

# Self-Adaptive Granularity Control for Network Segmentation

## Example Heuristic Evolved for NetFlow Application

