

Graph Techniques for Cybersecurity

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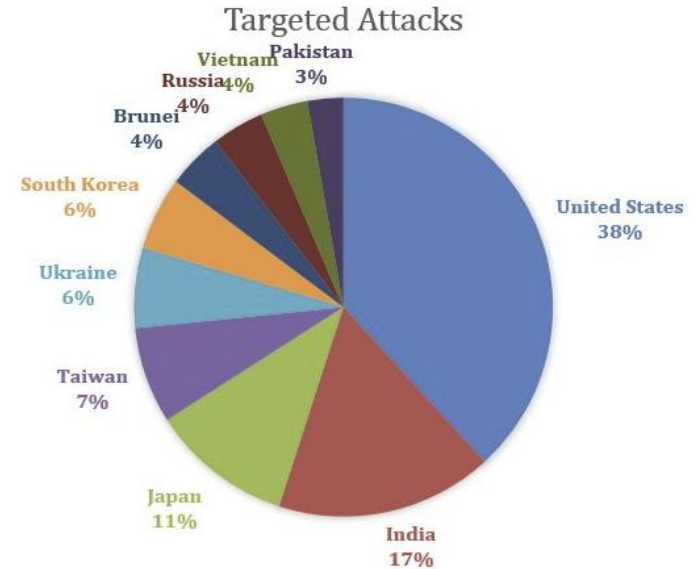


**Thank you ARCS
Foundation!**



Why Cybersecurity?

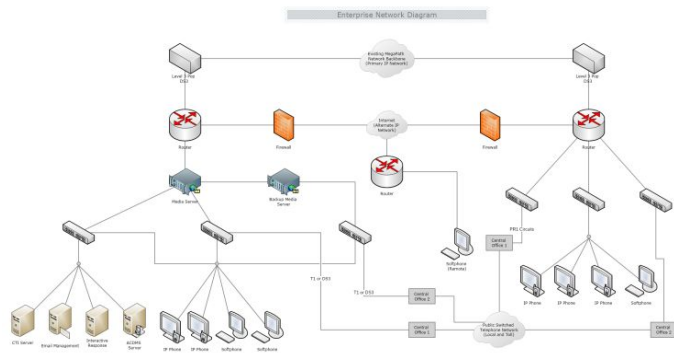
- Ransomware damage costs > **\$4 B** in 2017
- Damage related to cybercrime projected to hit **\$6 T** annually by 2021
- **20.4 B** IoT devices by 2020
- **90%** of Automobiles will be Internet Connected by 2020



Source: Symantec

Why Graphs?

- Cybersecurity relevant data highly amenable to graph representation
 - Network communication graph, program control flow graph, code property graphs

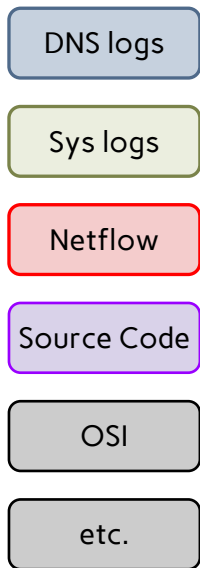


- Robust data structure capable of representing complex multimodal relationships
- Many analysis techniques ranging from traditional graph theory to more modern graph learning

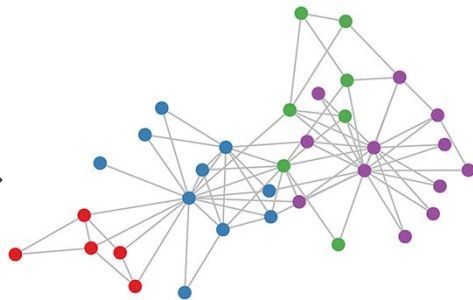


Research Vision

Cyber-Relevant Data



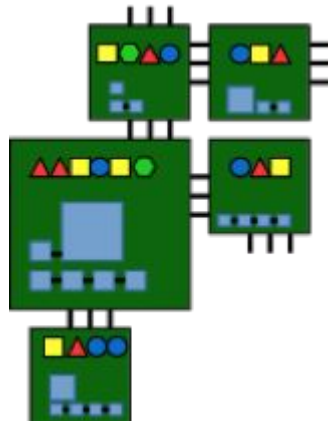
Graph Representation



Graph Algorithms



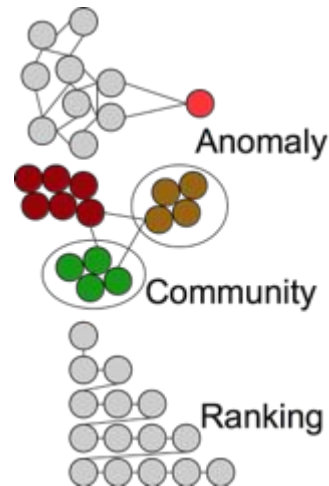
Graph System (hardware, software)



GPU, NVM,
Supercomputers



Actionable Cyber Knowledge or Insight





Anomaly Detection via Network Log Analysis



Network Security Today

- **365 day average time to detection** - monitoring and detection techniques insufficient
- Largely signature based, reactionary, requires human expert input
- Algorithmic methods often focused on singular data types

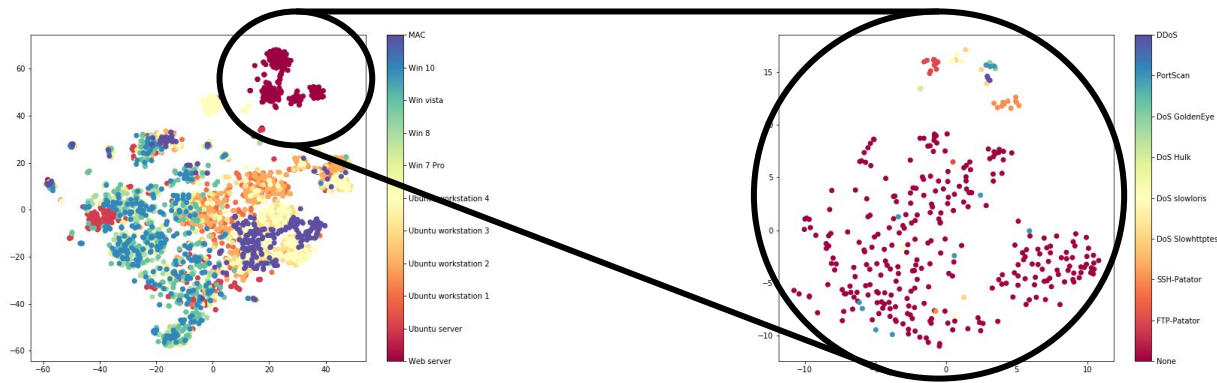


Source: Lockheed Martin

Existing techniques fail to utilize the complex relationships between disparate yet related cyber data points

Network Behavior Modeling via Unsupervised Graph Learning

- **Goal:** Utilize unsupervised graph learning techniques on streaming graphs containing many different cyber-relevant logs to learn behavior and pattern-of-life of network entities
- **Key Idea:** Identify malicious behavior as series of behavior changes





Vulnerability Detection in Software Source Code

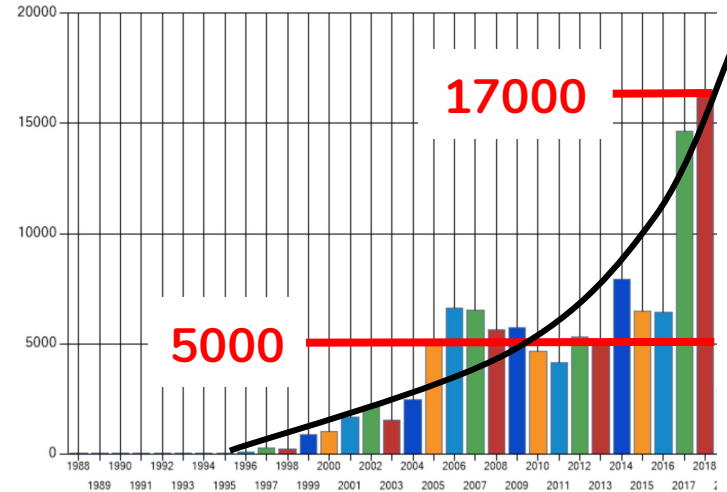


Vulnerability Detection Today

- 5000 vulnerabilities in 2013 vs 17000 in 2018
- Techniques fall into two main camps: **pattern detection** and **similarity detection**
- Patterns **manually** generated by human experts - not scalable
- Similarity detection typically based on **exact** matching - not flexible

Existing techniques are either good at finding one vulnerability in many programs, or finding many vulnerabilities in one program

Vulnerabilities by Year



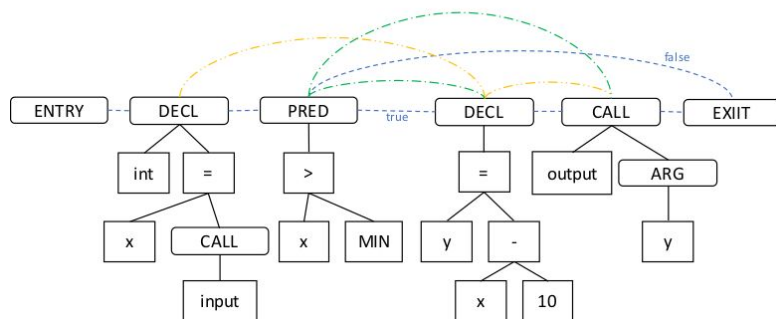
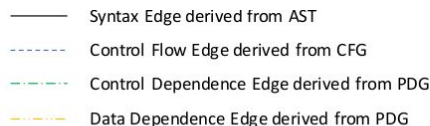
Source: NVD



vGraph: A Generalized Graph Representation for Vulnerability Detection and Discovery

- **Goal:** Be able to find many different vulnerabilities, in many different programs

System	Precision (%)	Recall (%)	F1 (%)
UDDY	85	91	88
VulPecker	90	60	70
VulDeePecker	79	83	81
vGraph	92	89	90

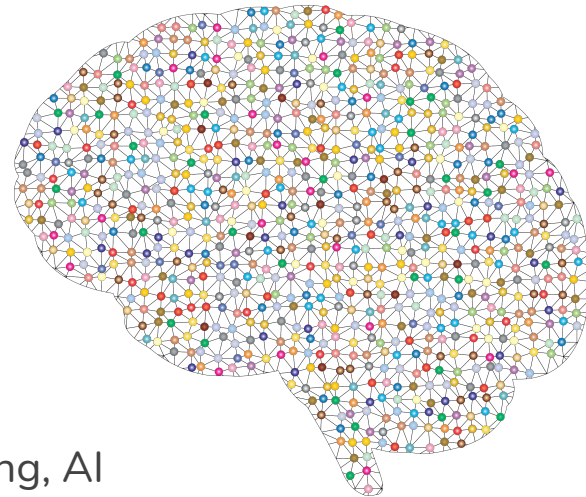


- **Key Idea:** Use a complex graph structure to represent **only the structure highly related to the vulnerability**
- Apply approximate subgraph matching techniques



Conclusion

- Graphs are powerful representations of data and have an active research community
- Recent advancements in Machine Learning, Deep Learning, AI on graphs are proving very effective
- Cybersecurity tasks and data are highly amenable to a graph representation and analysis (e.g., vulnerability detection, network behavior modeling, etc)
- Advancements in cybersecurity will be critical to the future of our increasingly digital society



Thanks for Listening!

Thank you ARCS for your support!

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