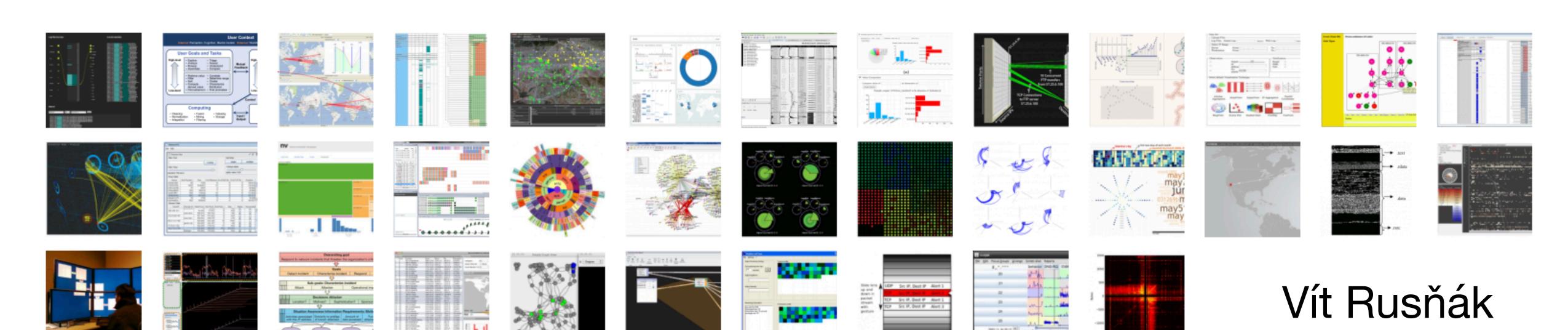


Visualizations for Cybersecurity

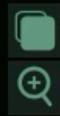
PA214 — Visualization II



STATISTICS DATA SOURCES BUZZ WIDGET











Talk Overview

- Taxonomy of Cybersecurity Visualizations
 - Data sources, data types, and tasks
- Application Scenarios
 - Monitoring, SIEM Dashboards, Analysis (threat hunting, forensics, malware)
- Research Trends in Cybersecurity Visualization

Typical Users

Cybersecurity operations (L1)



- monitoring, countermeasures
- CSIRT, Incident handlers



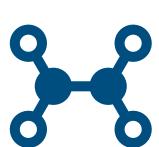
Cybersecurity Analysts (L2-L3)

network traffic anomalies, malware analysts



Management (both IT and non-IT background)

Chief information security officer (CISO), policy makers



Cybersecurity Researchers

simulations, application of ML/Al

Data Sources

Applications

Network Services

Users

Operating System

Intrusion Detection Systems

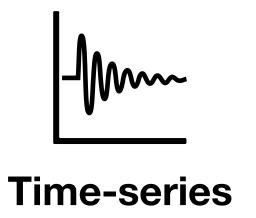
Firewalls

Passive Network Analysis

Traffic Flows

Packet Captures





Data Types

- Network traffic
- Log files and events
- User behavior
- Alerts and incidents
- Malware analysis data

Goals and Tasks

Monitoring Network traffic, log files, events, alerts

Anomaly detection > Alerts, events

Incident investigation > User behavior data, network traffic, log files, events

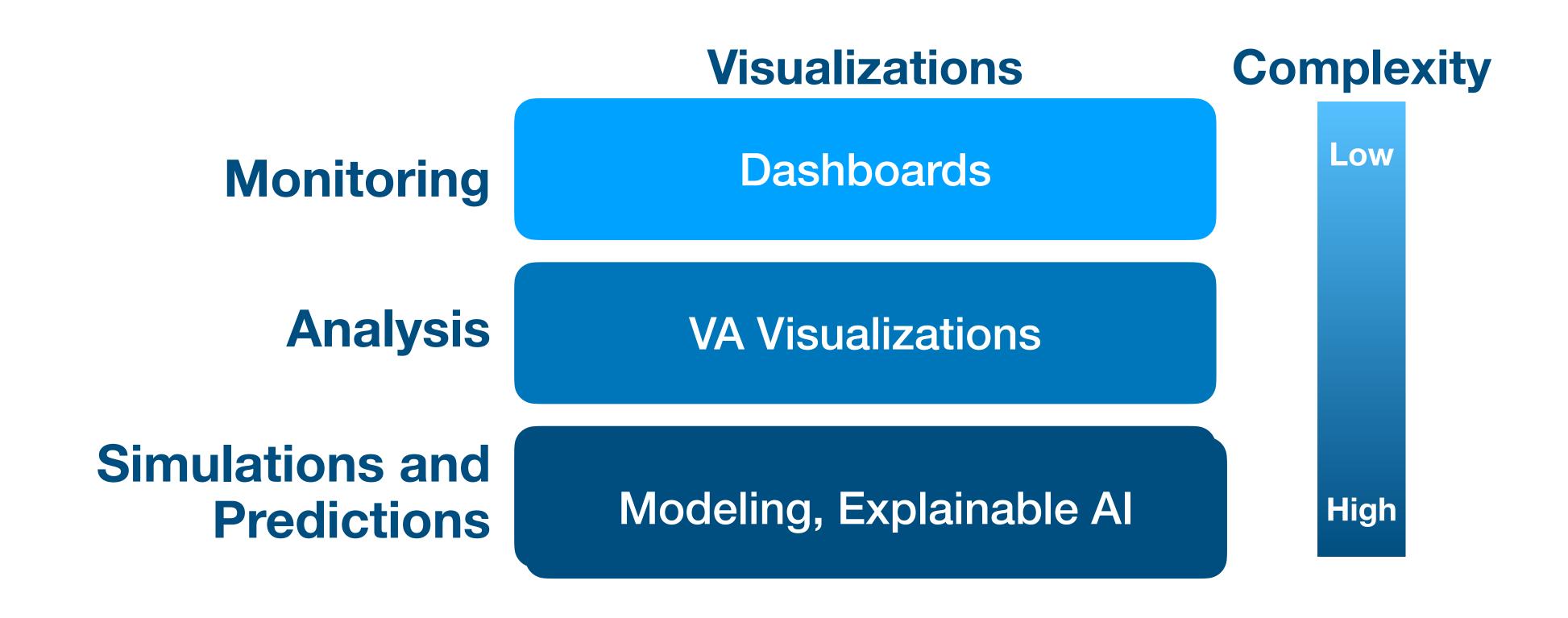
Situational awareness Network traffic, log files, user behavior data

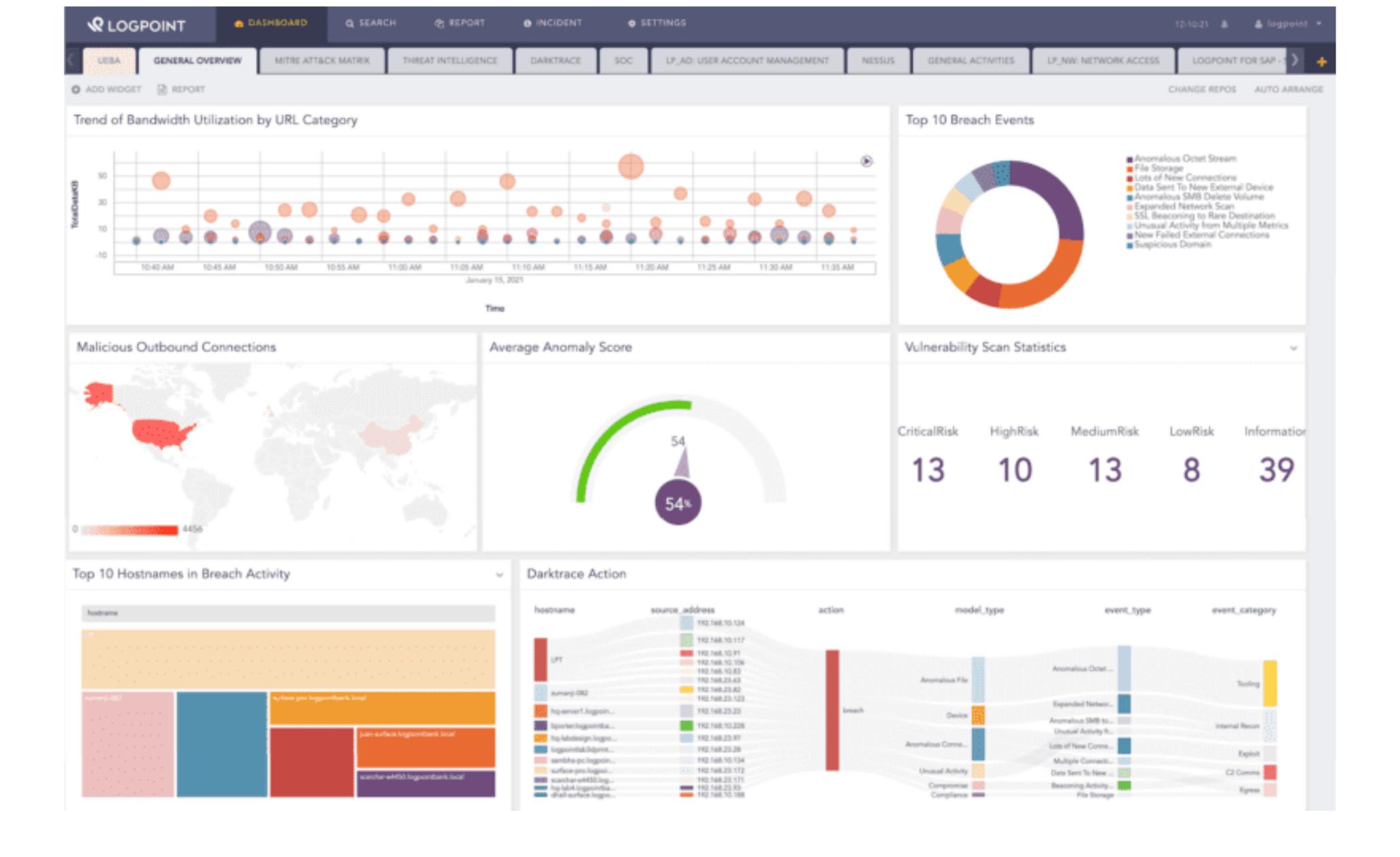
Atrribution and forensics Network traffic, alerts, events, user behavior data, malware analysis data

Visualizations

- Time-series (e.g., stacked graphs, horizon graphs)
- Graph-based (e.g., node-link diagrams)
- Spatial/geo (e.g., attack maps)
- Matrix-based (e.g., heatmaps, adjacency matrices)
- Hybrid/multiview systems

Complexity of Visualizations





Monitoring

Characteristics

- Dashboards are prevalent
 - Typically easy to read, decode and understand, multiple views (panels)
- Goal(s): situational awareness, trends, outliers and anomalies (e.g., peaks)
- **Typical visualizations:** tables, line/area charts, sparklines (microvisualizations), basic 2D charts (bar charts, heatmaps), basic geovisualizations (choropleth, links)
- Shortcuts and click-throughs allowing drill-down in analytical tools

Dashboards

"A dashboard is a visual display of the most important information needed to achieve one or more objectives that has been consolidated in a single computer screen so it can be monitored at a glance."

Stephen Few, Information Dashboard Design

Provide

- a snapshot of the current state (number of detected events, blocked IP addresses, ...)
- comparison to target measures (KPIs, warnings, trends)

Types

- Operational (monitoring, single source of information)
- Tactical (planning, communication)
- Strategic (management, decision)

Dashboards

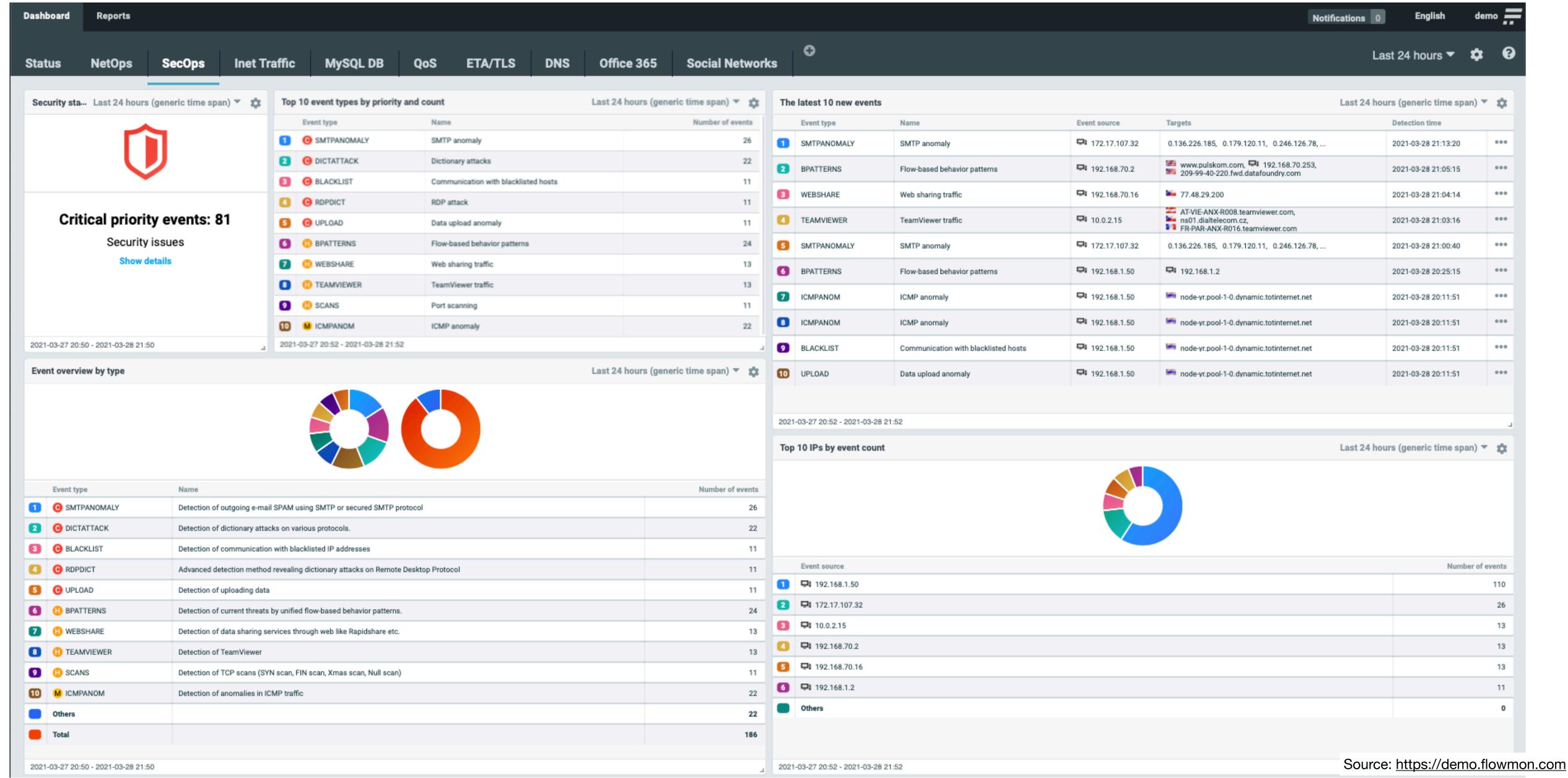
Design principles

- Contextual awareness
- Prioritization and alerting
- Configurability vs. cognitive overload

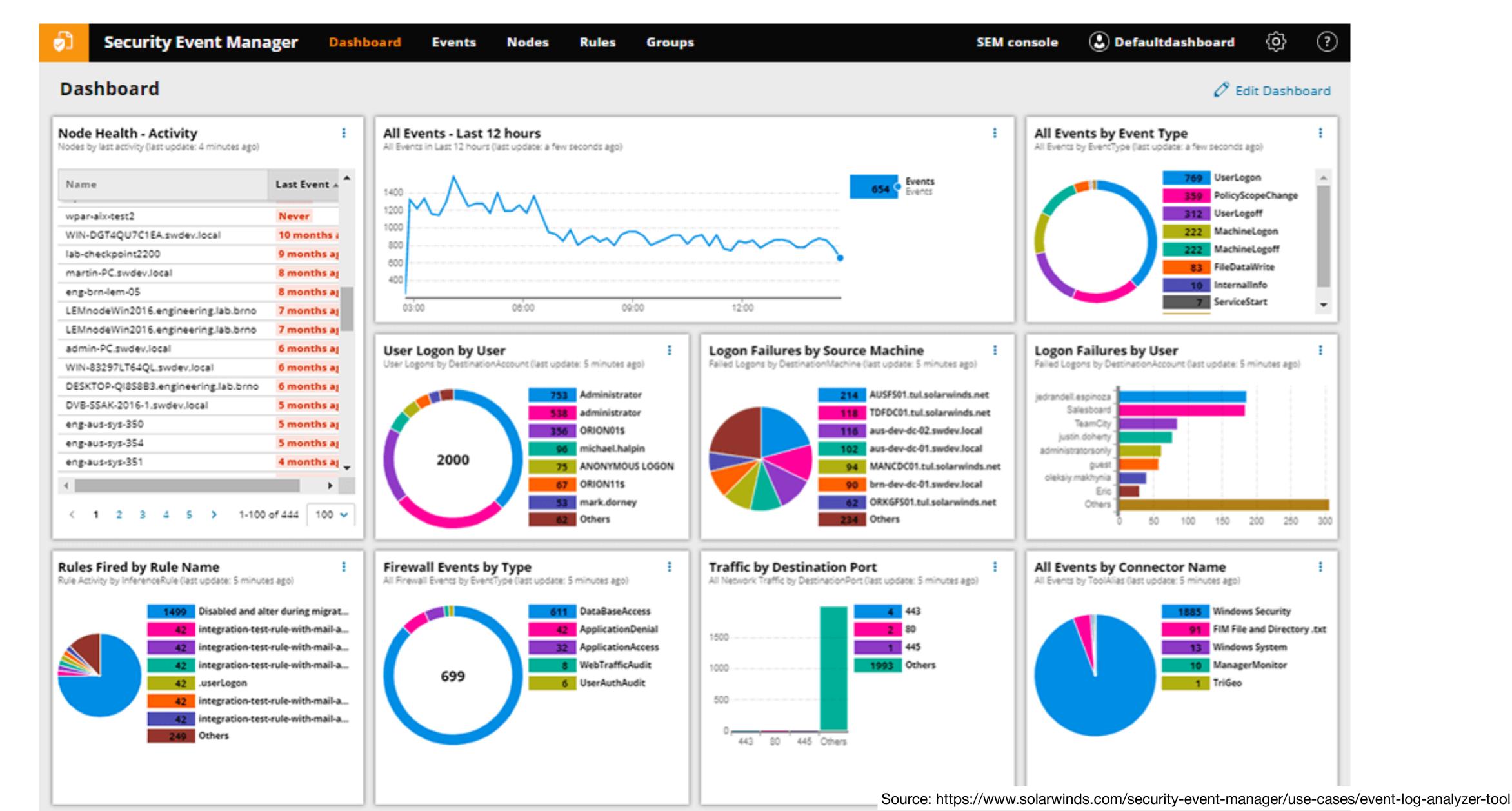
Common pitfalls

- Over-reliance on static views
- Poor encoding choices
- Lack of drill-down capabilities

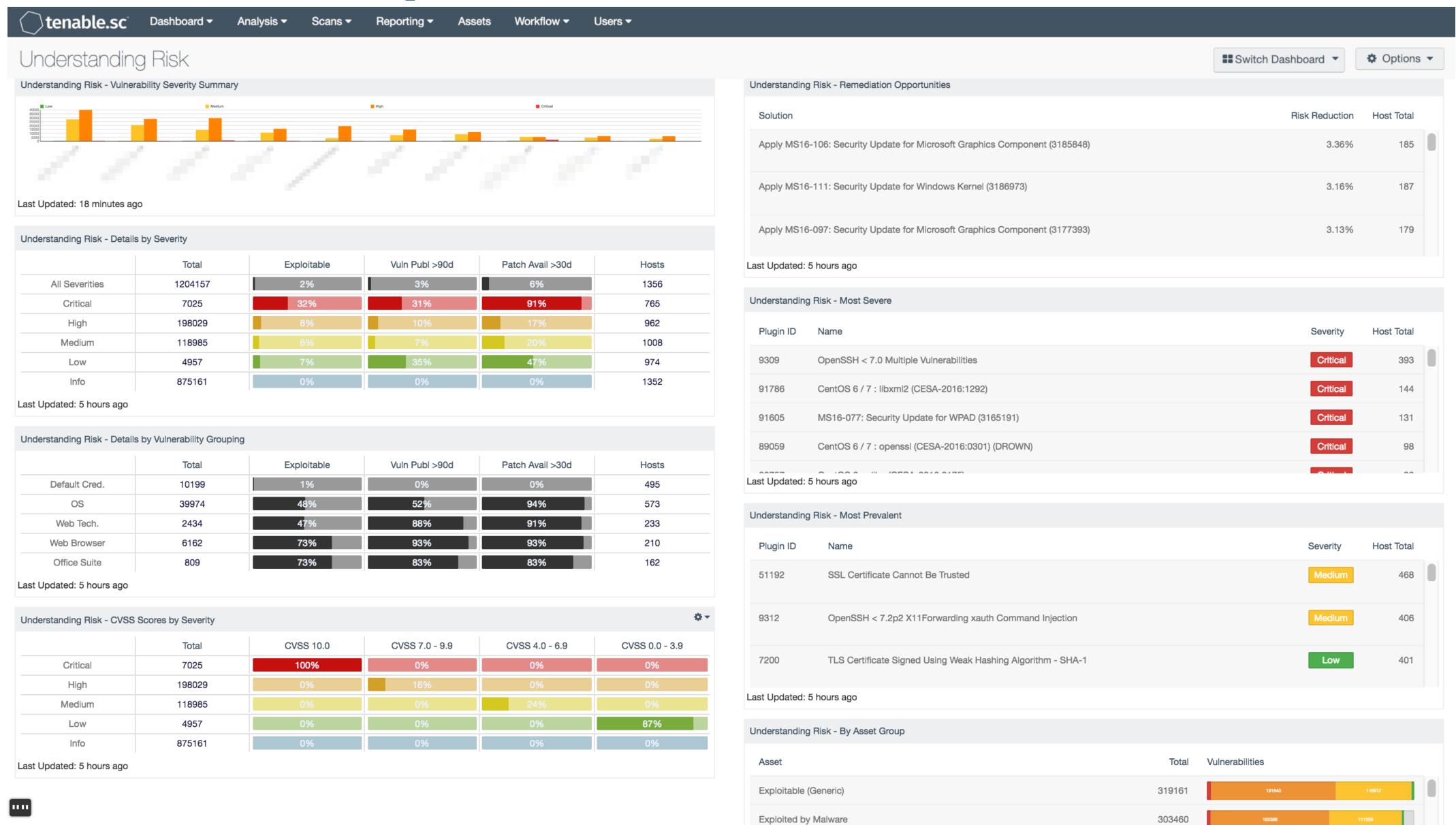
Examples: Commercial Tools



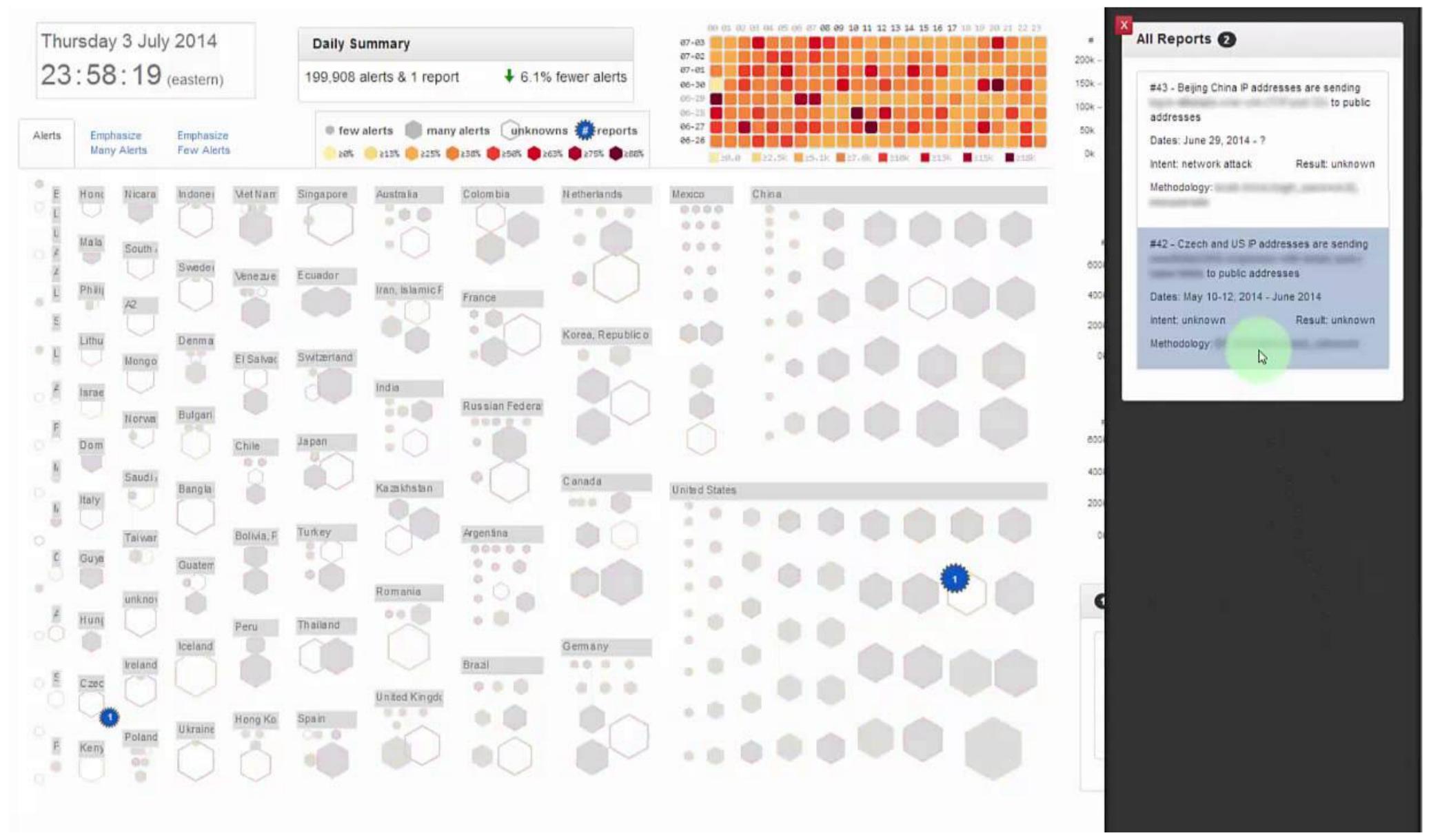
Examples: Commercial Tools

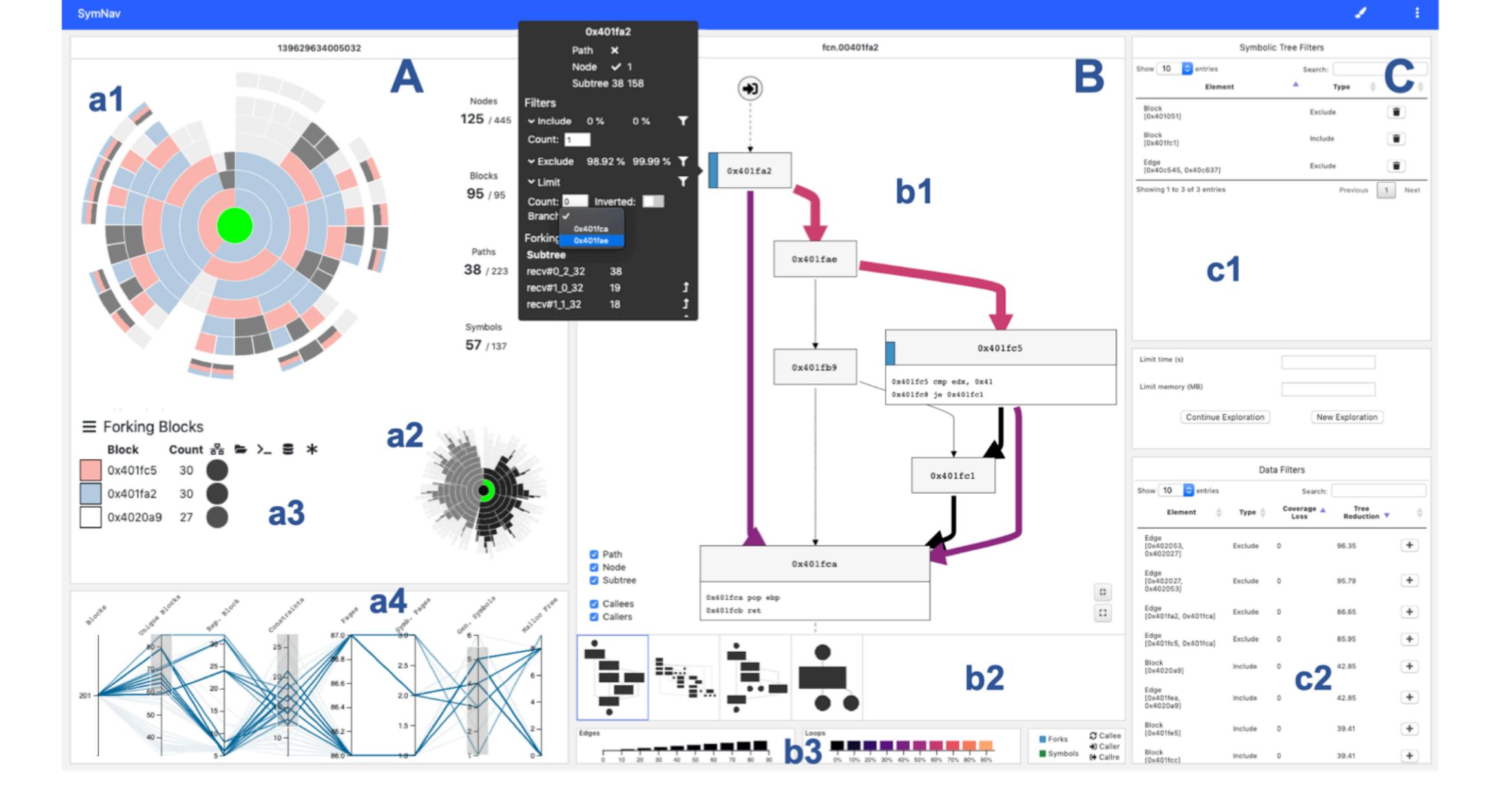


Examples: Commercial Tools



Examples: Research



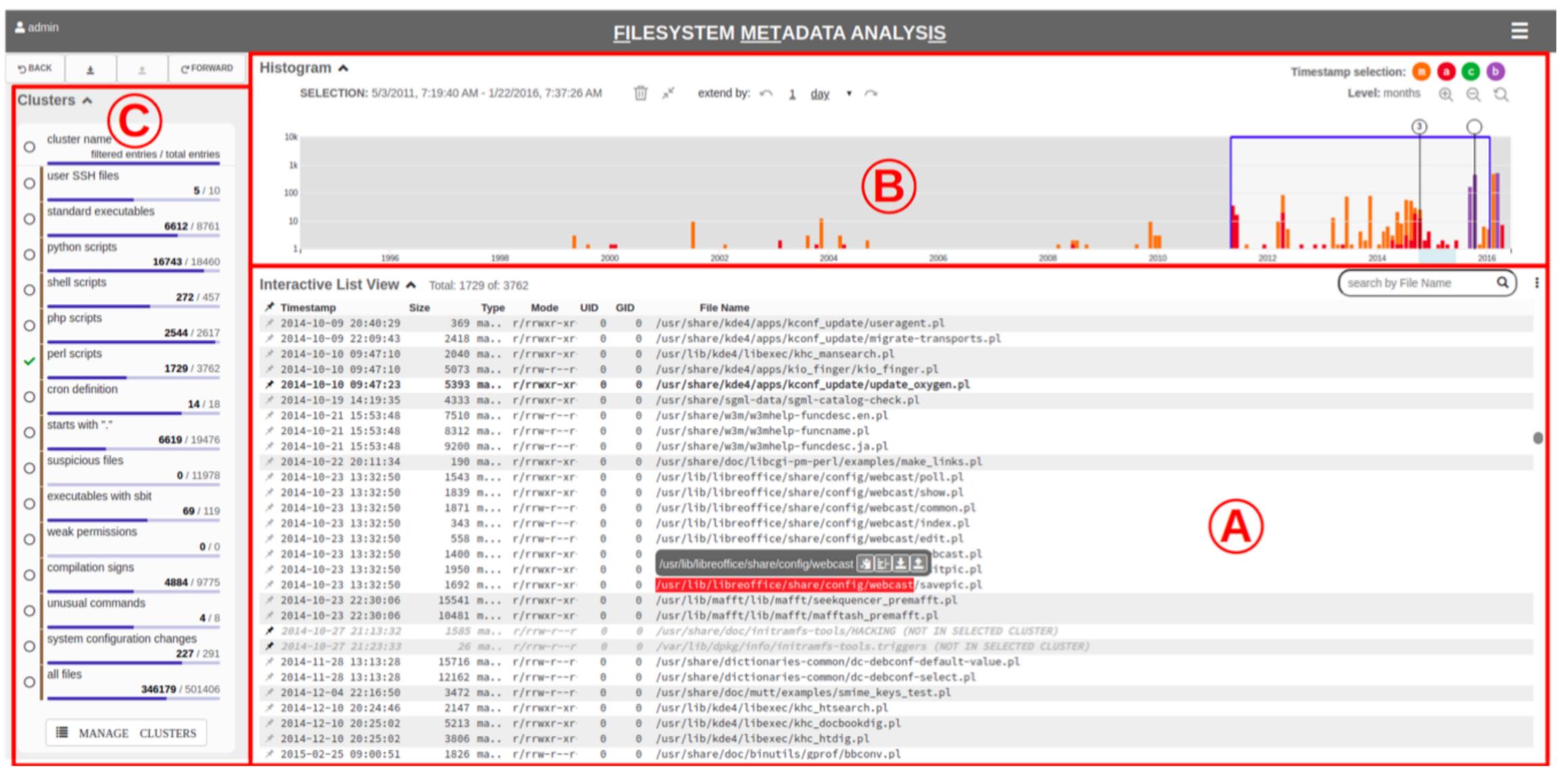


Analysis

Characteristics

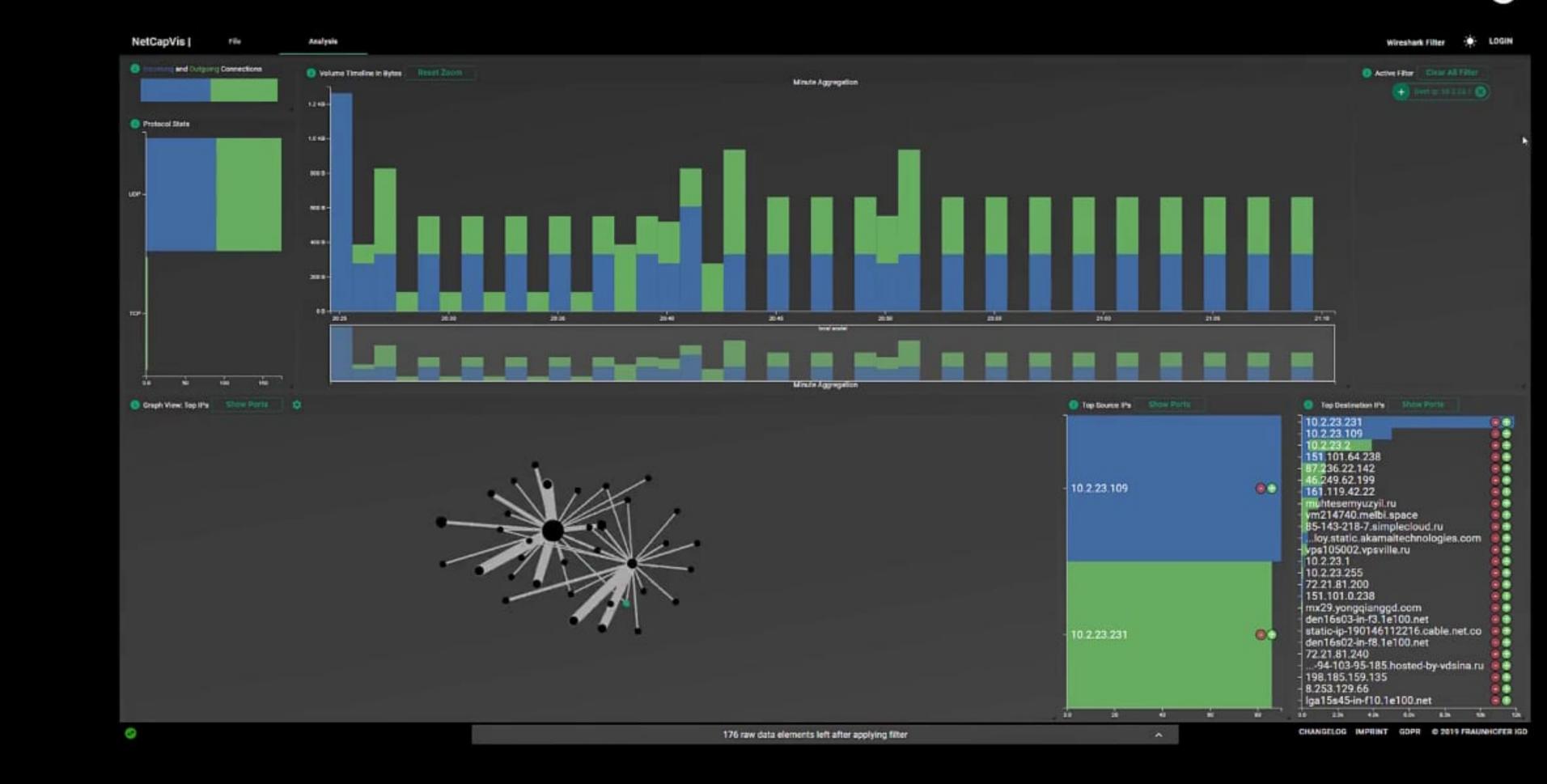
- Drill-down Visual Analytics Tools
 - Usually designed for particular use-case (e.g., malware vs. network analysis)
- Goal(s): Reduce "time-to-insight", automate repetitive tasks, help to identify anomalies in data
- Typical visualizations: linked views, basic but also novel visualization types
- Extend command line interface, use of APIs
 - Supported in existing systems (e.g., Splunk, Flowmon ADS) vs. custom-made tools
- Computational notebooks (e.g., Jupyter) are also in this category

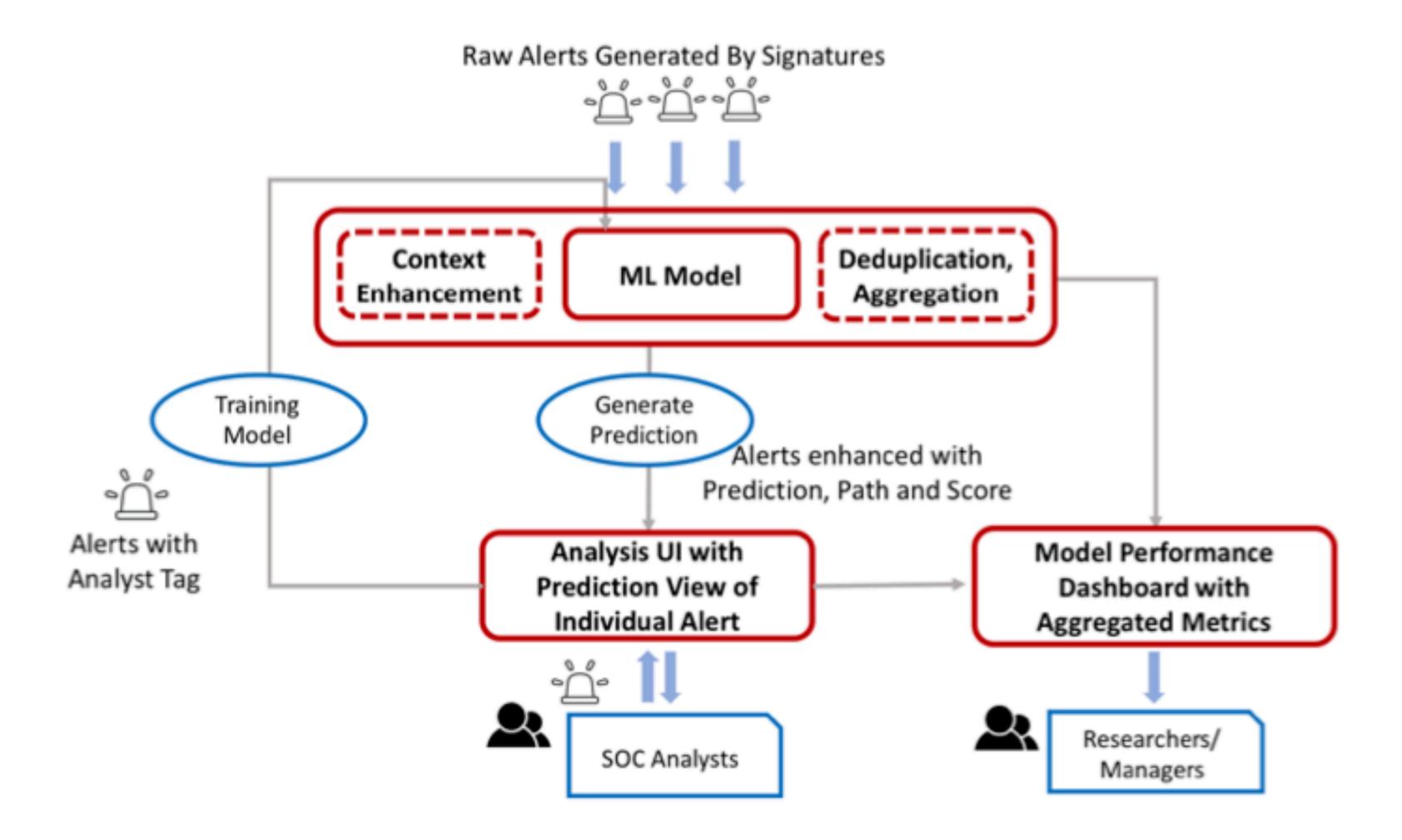
Example: File System Analysis



Example: Traffic Analysis

Web-based Visual Interactive Analyis





Predictions and Simulations

Characteristics

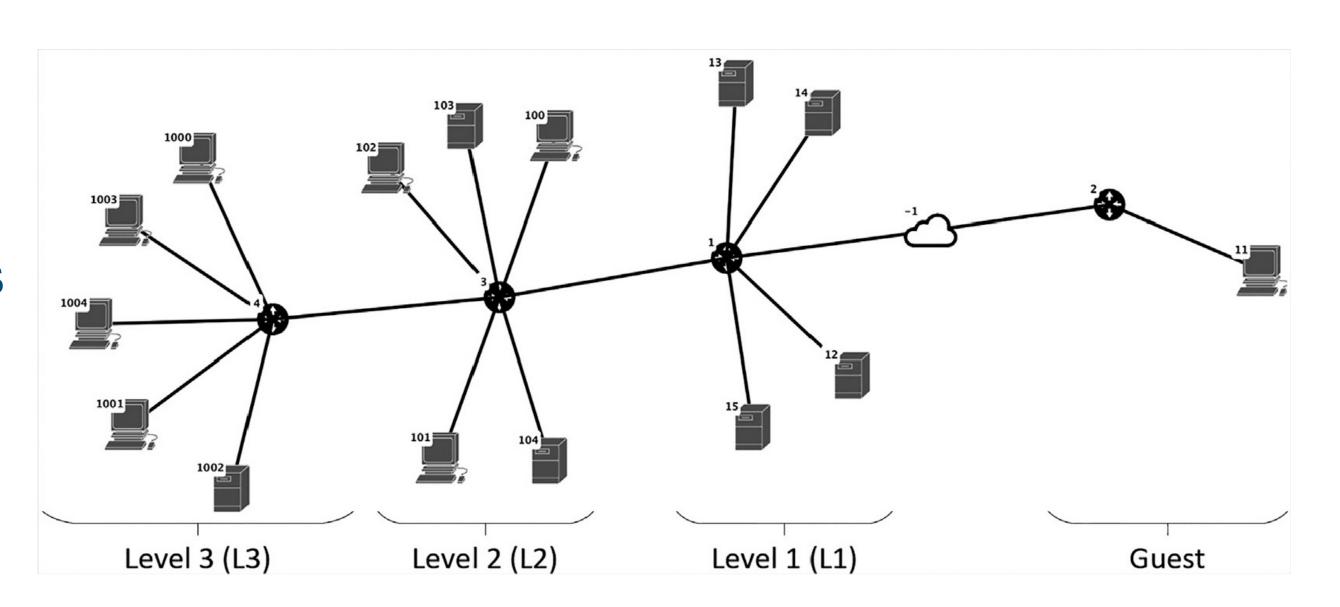
- Visual support for understanding **ML/AI** techniques, visualizations for explainability (**XAI** = **eXplainable AI**)
- Goal(s): understanding ML/Al techniques, behavior explanation, trust building
- Typical visualizations: clustering visualizations (for dimensionality reduction methods), linked views, basic visualizations
- Rise on popularity correlates with growing application of ML/AI in cybersecurity.
- Explainability approaches are transferable between different domains

Al in Cybersecurity

- Application of AI in cybersecurity is substantially difficult comparing to domains such as image recognition
- Three main areas:
 - Insights Generation: analyze the data to discover hidden patterns which can be used by decision-makers in order to react to anomalies.
 - Recommendations: the model discovers patterns in the data and provides recommendations on what should be best to do to a security specialist.
 - Autonomous mitigation: the model discovers patterns and tries to automatically solve problems without needing user input (e.g., approvals).

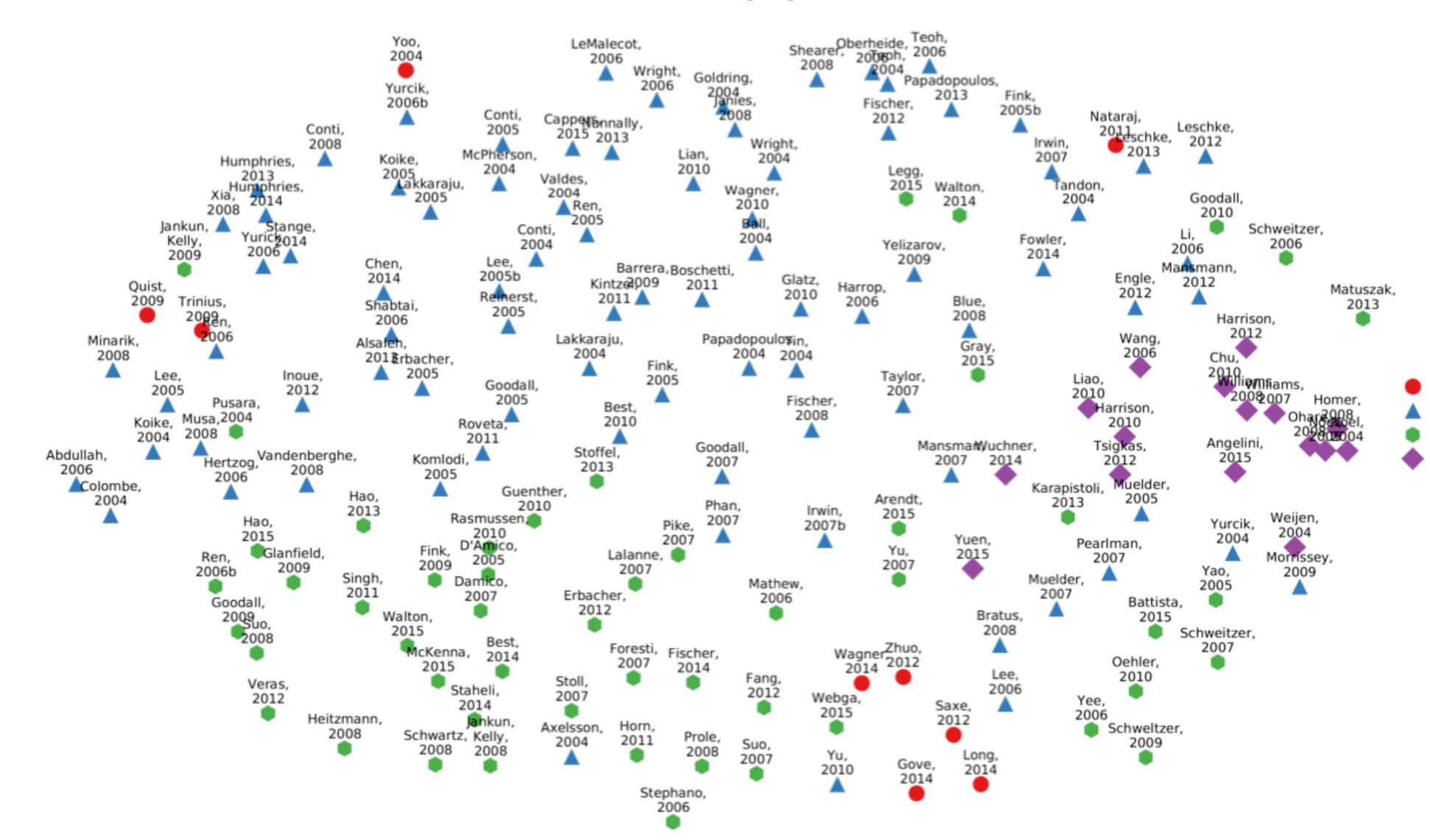
Simulations

- Largely unexplored
- Areas:
 - Attack surface and attack vectors
 - Scenario modelling tool
 - Autonomous agents (attackres) behavior
 - Comparison and explanation of their decisions



CyberSecVis Research

VizSec papers 2004—2015

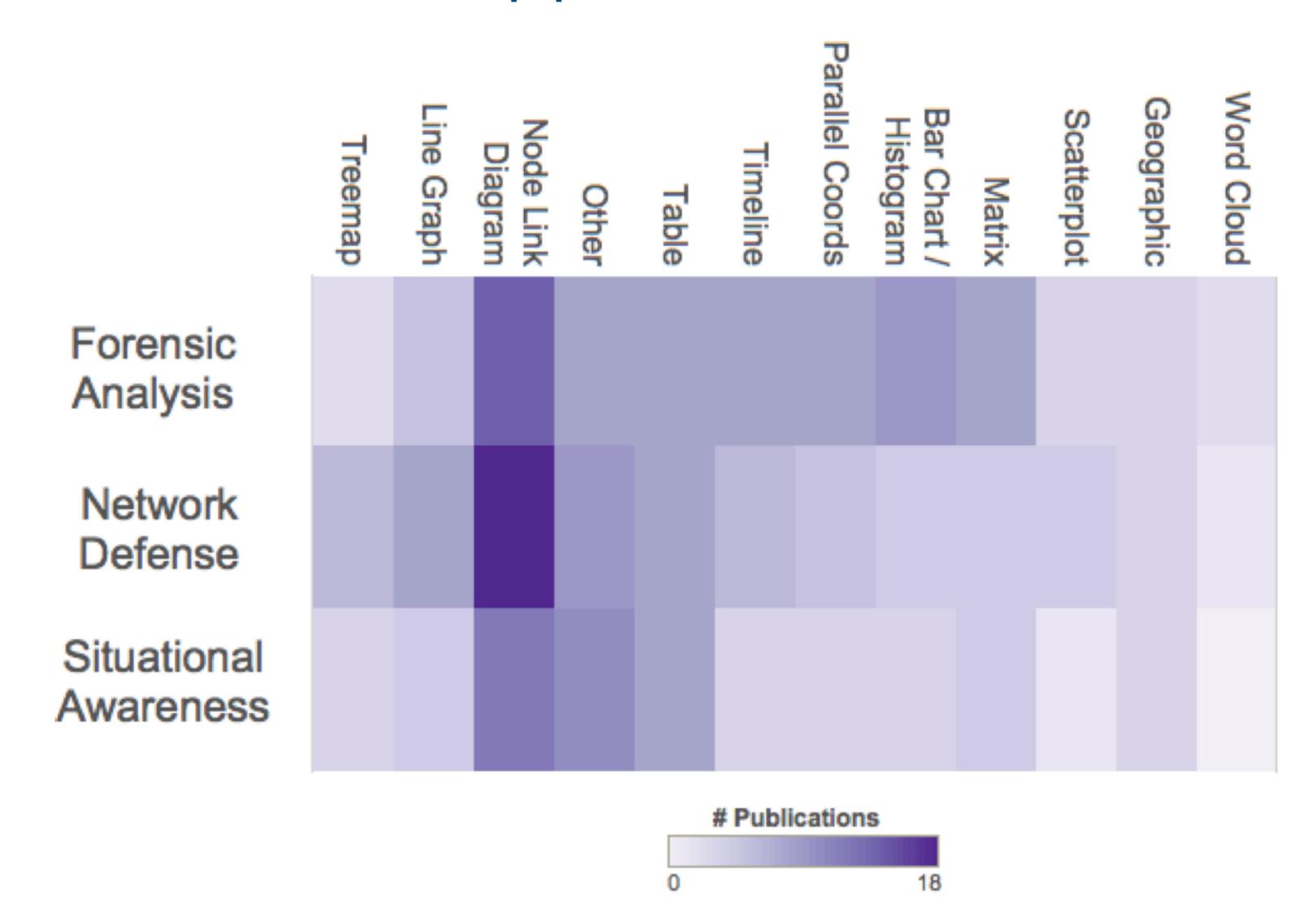


Top 5 Words in Cluster

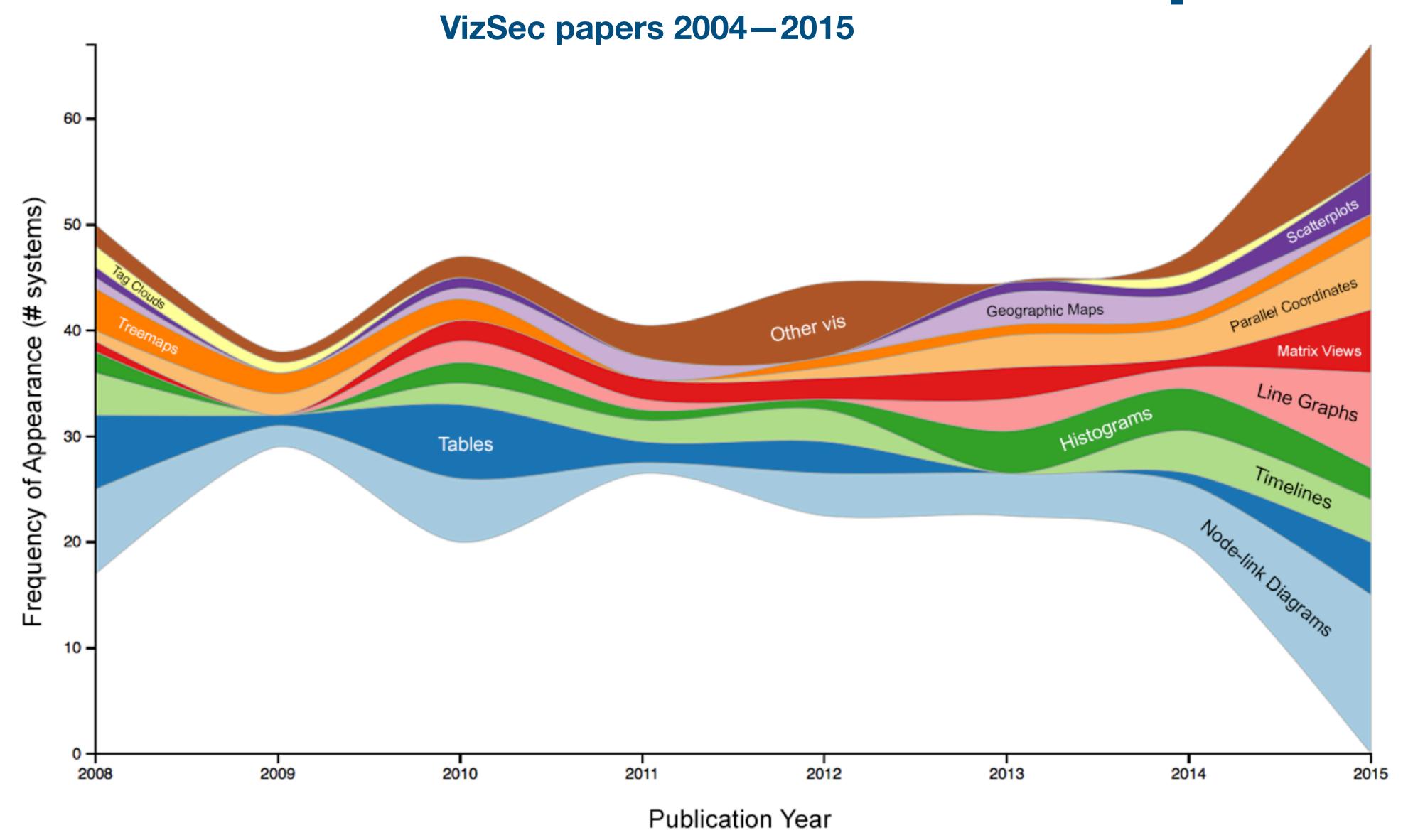
malware, sample, execution, imaging, virus ip, ported, hosts, traffic, packet analysts, task, models, cyber, alerts attacks, graph, node, vulnerabilities, exploit

Utilization of Visualizations

VizSec papers 2004—2015

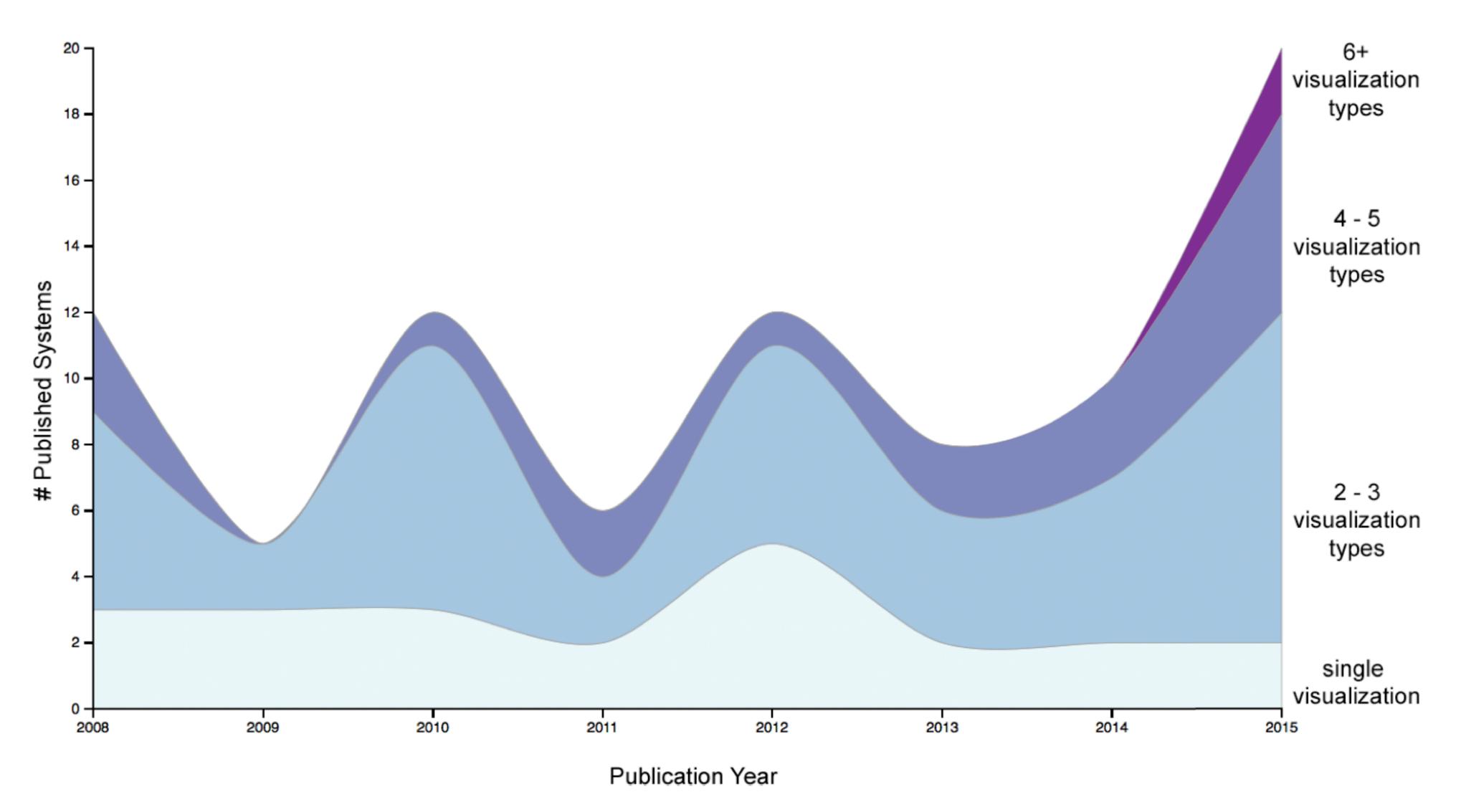


Utilization of Visual Metaphors



Interface Complexity

VizSec papers 2004-2015



Take-aways

- What makes cybersecurity data challenging?
 - volume, velocity, heterogenity, context
- Why visualization matters?
 - cognitive amplification, anomaly detection, decision making
- The commercial tools use only common charts and visualizations ...
 - ... → lot of space for improvements

Resources

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- [5] M. Angelini et al., "SymNav: Visually Assisting Symbolic Execution," 2019 IEEE Symposium on Visualization for Cyber Security (VizSec), Vancouver, BC, Canada, 2019, pp. 1-11, doi: 10.1109/VizSec48167.2019.9161524.
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- [10] B. C. M. Cappers and J. J. van Wijk, "SNAPS: Semantic network traffic analysis through projection and selection," 2015 IEEE Symposium on Visualization for Cyber Security (VizSec), Chicago, IL, USA, 2015, pp. 1-8, doi: 10.1109/VIZSEC.2015.7312768.
- [11] Moskal S, Yang SJ, Kuhl ME. Cyber threat assessment via attack scenario simulation using an integrated adversary and network modeling approach. The Journal of Defense Modeling and Simulation. 2018;15(1):13-29. doi:10.1177/1548512917725408

Other

- IEEE Symposium on Visualization for Cyber Security https://vizsec.dbvis.de
- Shixia Liu, Xiting Wang, Mengchen Liu, Jun Zhu, Towards better analysis of machine learning models: A visual analytics perspective, Visual Informatics, Volume 1, Issue 1, 2017, Pages 48-56, ISSN 2468-502X