



Carnegie Mellon
Software Engineering Institute

CERT
Situational
Awareness

Analysis of the US-CERT DAC

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Outline

- Data
- Graphical Displays
- Detecting Trends
- Anomaly Detection
- Roadmap



Data

- **Snort**
 - Signature-based alerts
 - Pre-processor alerts
- **Origin**
 - Multiple networks of varying size
- **Volume**
 - ~30-50 million alerts per month
- **Ancillary Information**
 - Country code
 - Netblock



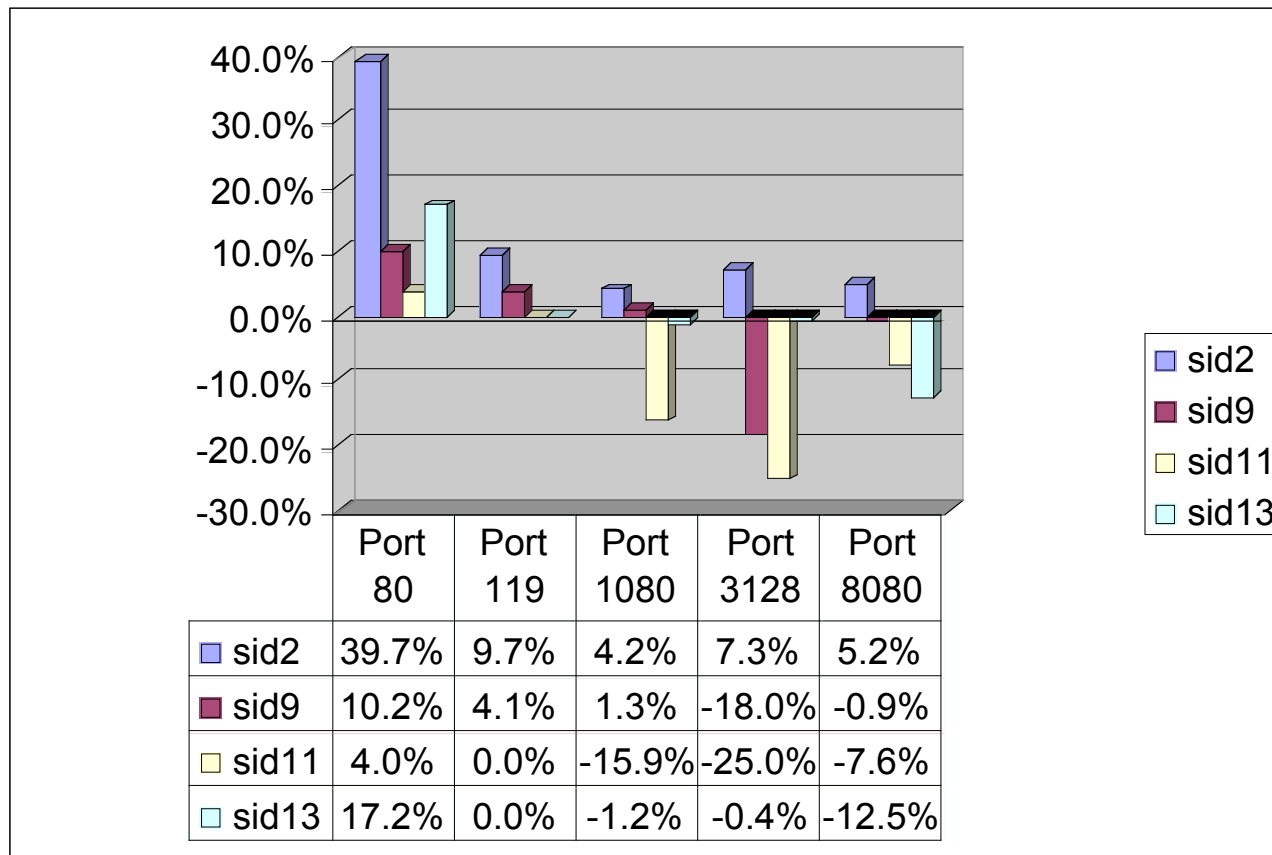
IDS Data: challenges

- No new attacks
 - Only matches known signatures
- Lack of context
 - Don't know what we are not seeing
- Non-standardized signature rule sets
 - No administrative control
- Missing Data
 - Uncertainty: Sensor failure vs. no intrusion attempts



TCP Destination Port Changes

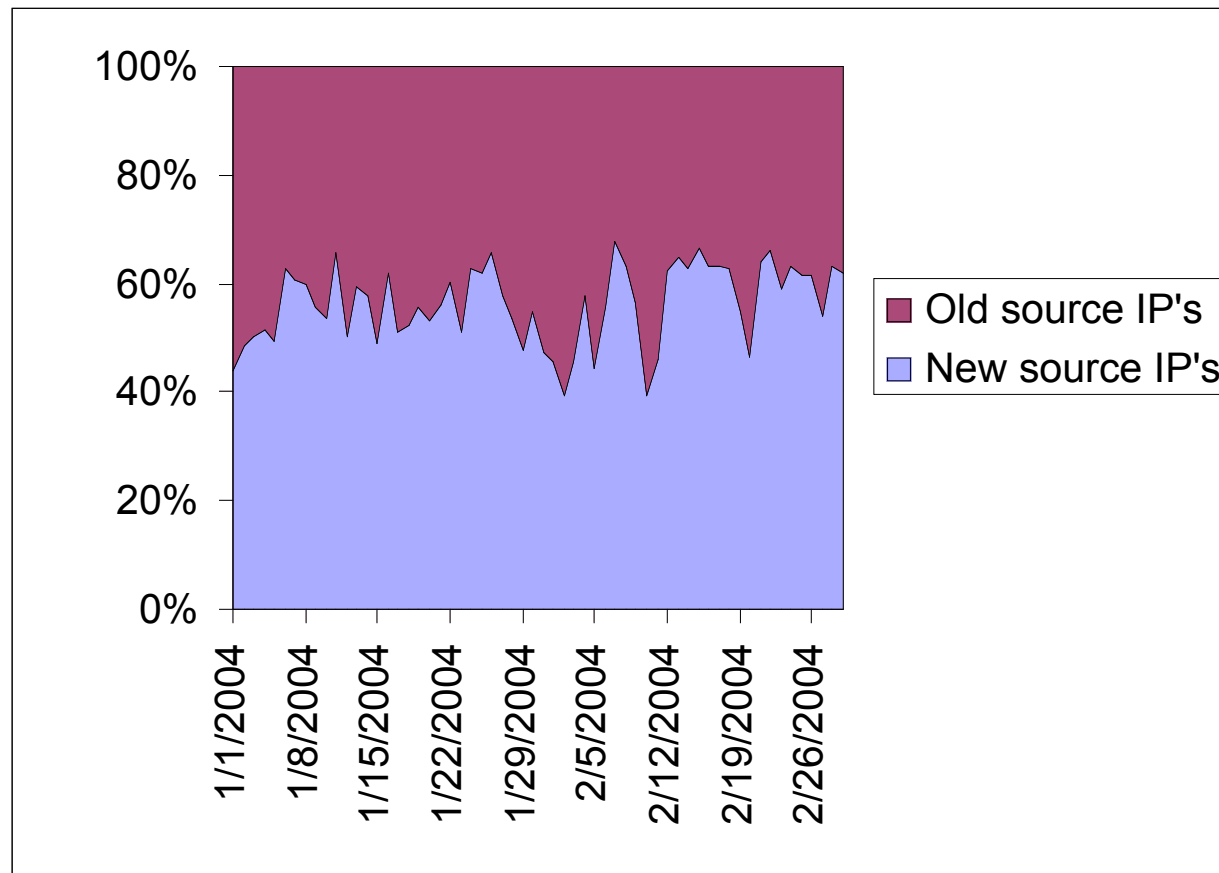
Comparison of port activity across organizations shows monthly trends.





Share of New Source IP Addresses

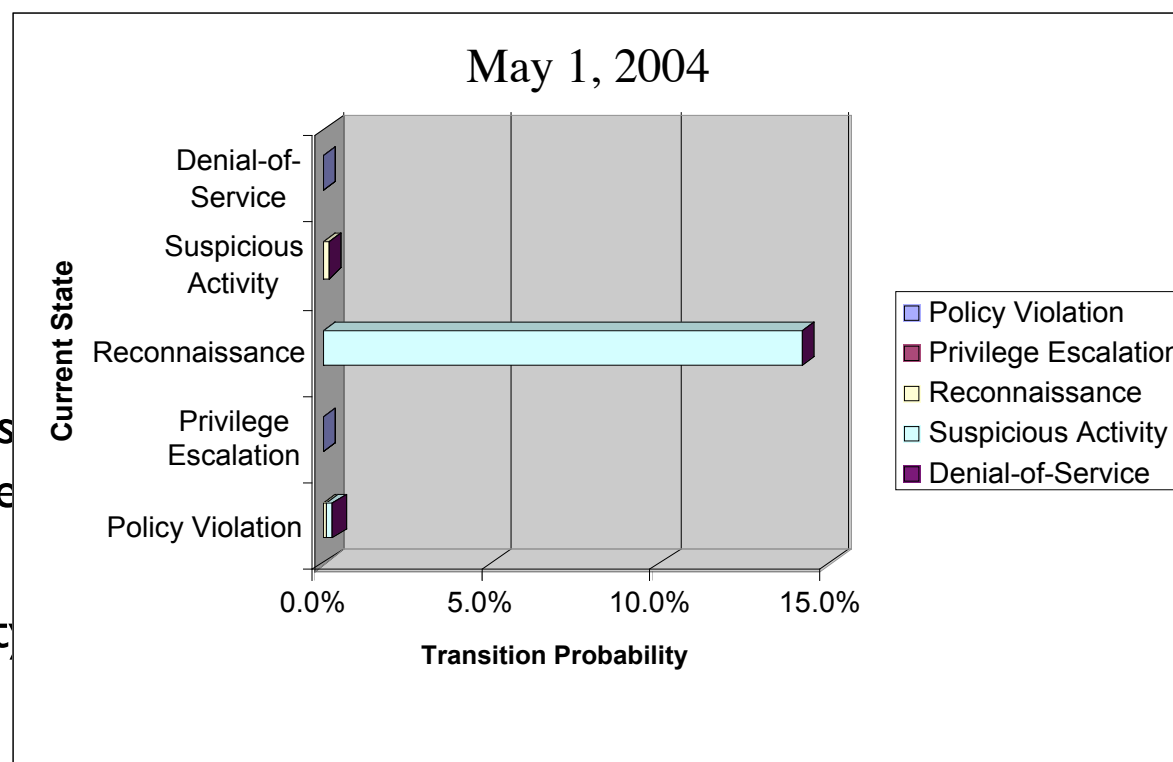
Share of new daily source IP addresses stays fairly consistent.



Signature Class Transition

Transition probabilities highlight sequential patterns in data.

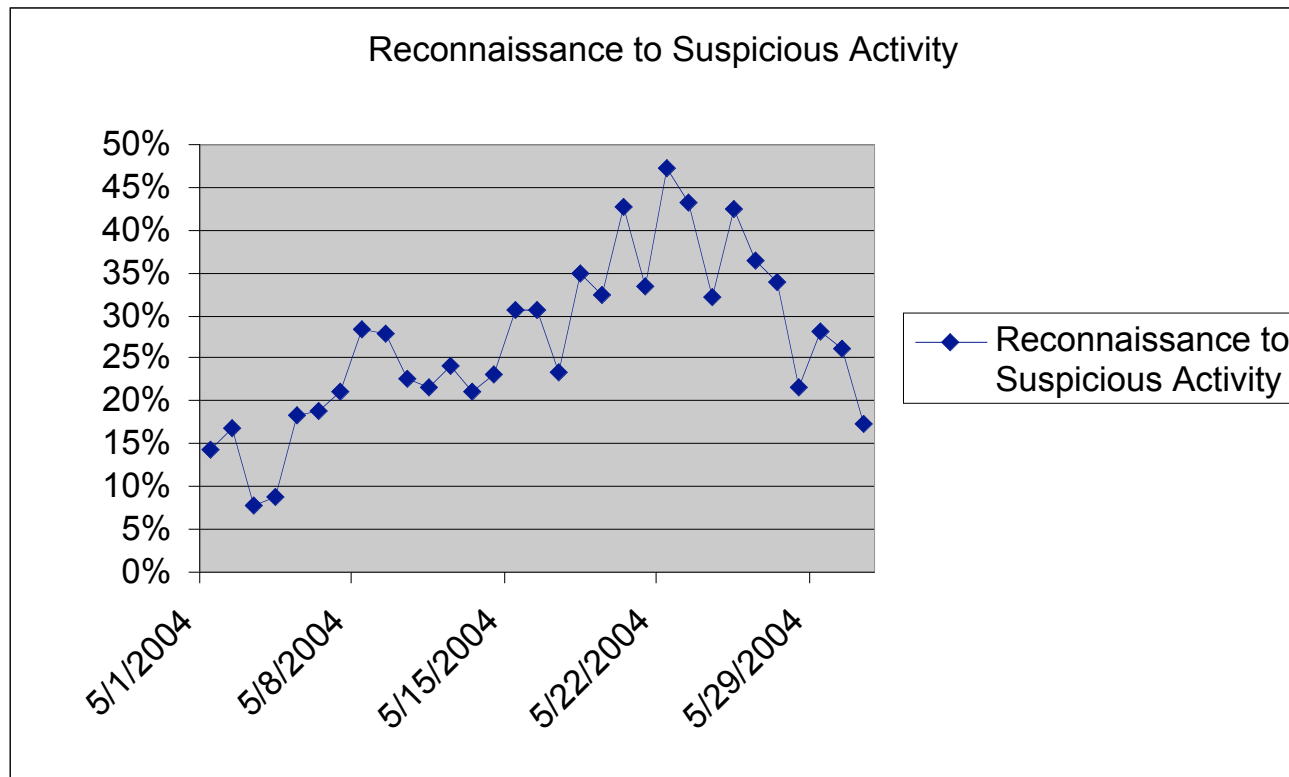
- **Current State**
 - Source IP records alert on Destination IP
- **Transition probability**
 - Percent chance for next class of alert recorded
- **Most source/dest combos involve only one signature class**
- **Small transition probability for**
 - Privilege Escalation





Daily Transition Probabilities

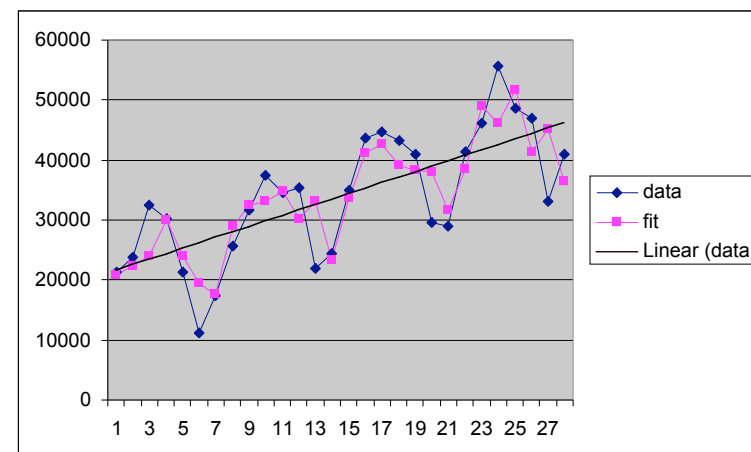
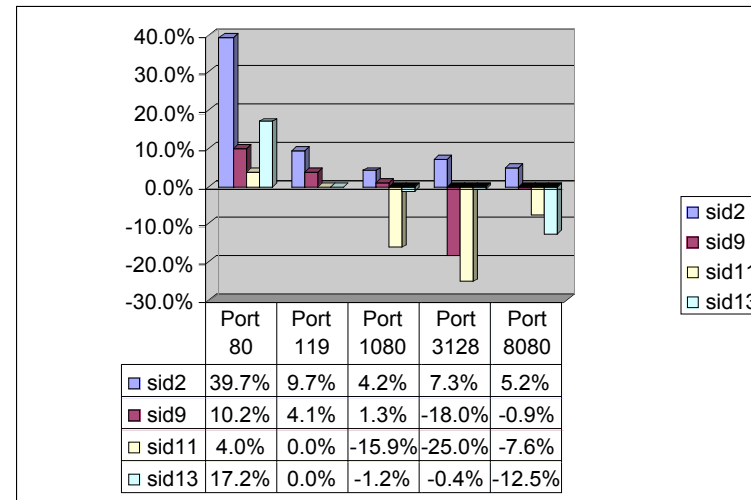
Transition probabilities can be monitored over time to identify consistent sequences.





Trend Detection

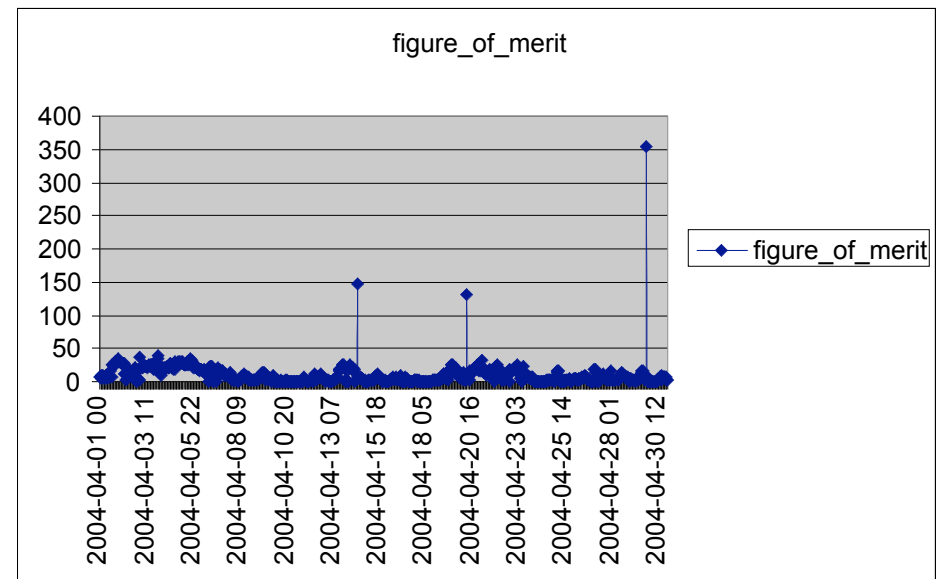
- Current month vs. previous month
 - Across organizations
 - % changes
- Time Series
 - Fit trend line
 - Arbitrary time period
 - Seasonal Components
 - Regression with ARMA errors





Anomaly Detection

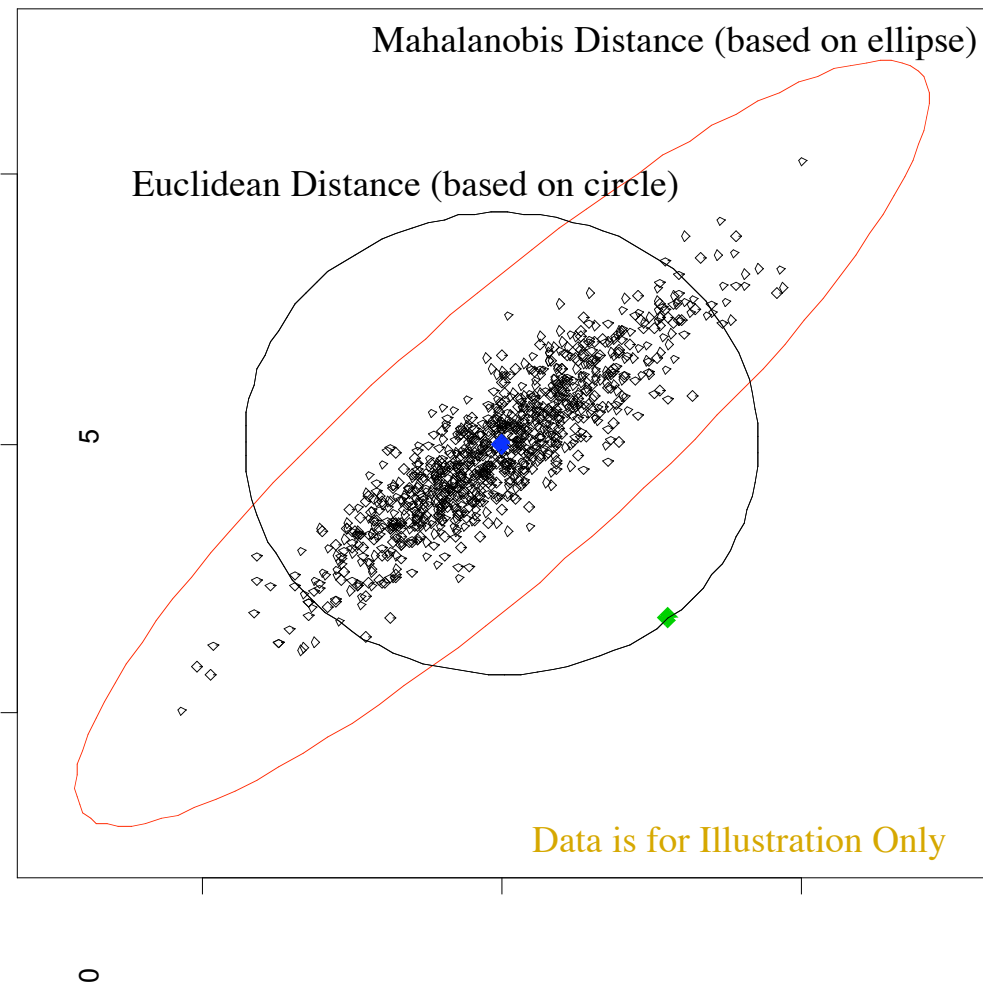
- Goal: Identify data points which deviate from overall pattern of data
- Our current implementation (Figure of Merit)
 - Evaluate hours
 - Record # alerts, # source IP addresses, # destination IP addresses, # signatures
- For each hour, we want measure of how deviant it was.





Mahalanobis distance: 2D case

- Compute distance metric between each hour and the **average** hour
- When measuring **Euclidean** (**Mahalanobis**) Distance, all points along **circle** (**ellipse**) are same distance from the center
 - Points on larger circle/ellipse are greater distance from center
- Shape of the ellipse
 - Function of correlation between variables
- Generalizes to n dimensions (**Ellipsoid**)





Analysis Roadmap

- Incorporate flow data
- Automating trend detection
 - Time series analysis
- Clustering
 - Group sources by similar activity patterns
 - Temporal correlation
 - Targeting similarities
 - Signature usage
 - Look for evidence of possible coordination