



Situational Awareness Metrics from Flow and Other Data Sources

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Introduction

Need a more flexible set of metrics for
network situational awareness

- Aggregate (over large IP sets)
- Composite (multiple measures or counts)
- Can detect changes in traffic patterns
- Amenable to visualization
- Fast and scalable (simple algorithms)



Overview

Propose some new metrics for SA

Uses Flow Data

Some require additional data:

- Information on Assets

- Organizational Level Data

- Elicited Data

- Various Lists of Sites/Hosts/Domains

- DNS

- Topology



Proposed metrics

Threats - Risks - Impacts

A) Mainly flow data and results from SIMS

$N(\text{attack category}) - N(\text{method of operation}) - N(\text{system or host})$

Estimate (N): $\langle TP \rangle + \langle FN \rangle$ | exercises & pen tests

B) Flow and other data

Match attack sources with malicious domain lists

- intersection of the IP sets

Implementation levels & Compliance levels

- priority of patches or tasking orders

- criticality of hosts

Probabilities of success * Expected damage | Attack category



Flow-based Metrics: Threats

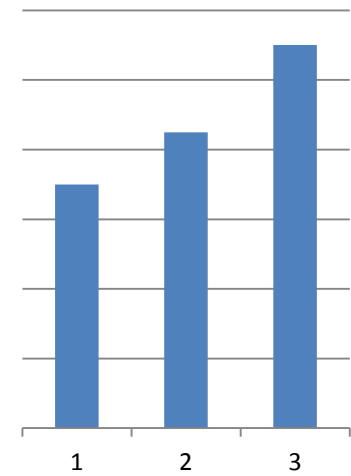
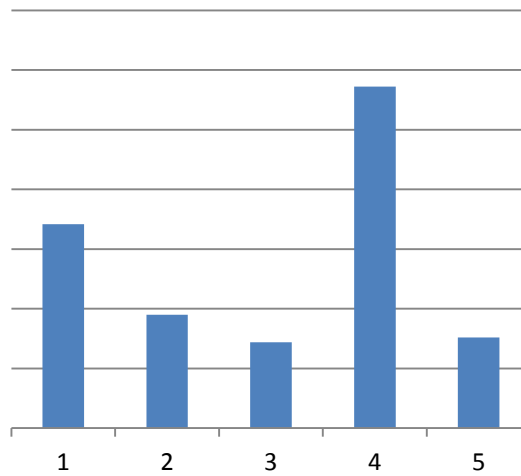
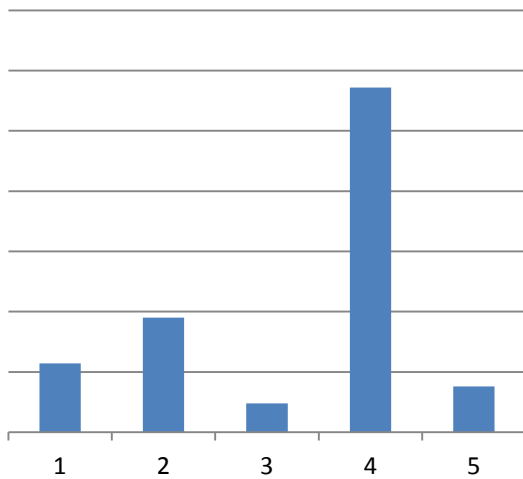
$N(i) \mid \{I\}$ = distribution of attacks by category i in set $\{I\}$
= attack scenario

$[w(i)*N(i) \mid \underline{w}]$ = seriousness-weighted attack distribution
= attack intensity

$\sum_i [w(i)*N(i)]$ = Overall seriousness > Trends
 $[\sum_i w(i) = 1]$



Illustrative Example: Numbers, Intensity, Trends



Flow-based Metrics: Risks

$N(m)$ = Distribution of attacks by “method of operation”

$[s(m)*N(m) | \underline{s}]$ = Severity-based risk scenario

$\sum_m [s(m)*N(m)]$ = Overall severity metric

$$[\sum_m s(m) = 1]$$



Flow-based Metrics - Impacts

$N(h)$ = Distribution of attacks by system/host “h”

$[v(h)*N(h) | \underline{v}]$ = Value-weighted impact of attacks

$\sum_h [v(h)*N(h)]$ = Overall value of network assets
that are being attacked by their attack rate



Other Data Needed

{Attack Categories} and {Relative Seriousness}

{Taxonomy for MOs} and {Potential Severity}

{Classification of Network Assets – value & criticality}

{True Positives | Alerts & Verification} and {False Negatives}
<- Exercises and Testing

{Lists of Malicious Domains/Ips} [Some exist]

{Status of Assets w.r.t. Compliance: Yes/No | patch or TO}

{Success rate of attacks by category | recent reports}

{Expected damage from successful attacks}



Estimation of metrics based on non-flow data

Maliciousness of Attacker Set:

$\{A\} \sim$ attacker set

$\{M\} \sim$ lists of known malicious hosts

$\{A\} \cap \{M\} =$ Degree of Attribution by Maliciousness

Risk \equiv Non-compliance (+ other factors) \gg Compliance level

$\sum_h J(h,p) * c(h) / \sum_h Y(h,p) * c(h) = I(p) =$ Implementation level of patch p

$J \supset Y$

$\sum_p u(p) * I(p) =$ Compliance level with respect to patching

Compliance by criticality & urgency

Impact:

Likelihood * Consequence

$= \pi * D =$ {probability of successful attack * expected damage}

{ $\pi(k) * D(k)$; by level of damage k}



Benefits for Situational Awareness

New metrics to supplement current measures

Additional aspects of SA

Identification of important data to be collected

Fast estimation procedures

Can track changes over time



Summary and Conclusions

Summary

Set of SA metrics: ***Threats-Risk-Impact***
Properties and interpretation of the metrics
Flow data and additional data (as identified)
Benefits from applying these metrics

Key Challenges

A processing and analysis layer between queries & reporting
Data availability
Problems with the numbers (NATs, Proxies, inconsistencies, etc.)

Future Work in Brief

Develop, validate & interpret these metrics
Collect the needed data systematically
Include the intermediate analytics capabilities



THANK YOU!

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