

Early Intrusion Projection and Impact Assessment for Cyber Situational Awareness



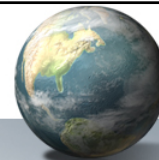
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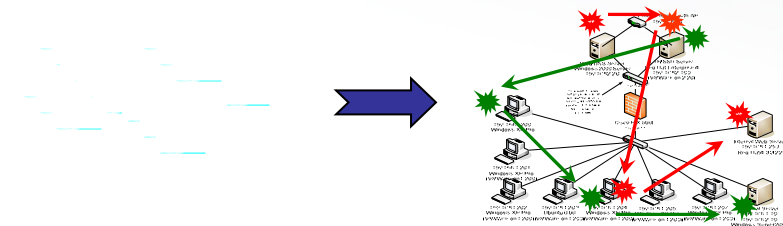
² CMIF, SUNY Buffalo

Assessing Cyber Attacks



Q: What to do with overwhelming intrusion alerts?

- Alert aggregator/correlator forms attack tracks
- Estimate impact of observed attack tracks
 - Rule/scenario based [Porras & Fong '02], [Valeur, et al. '04]
- Project attack actions
 - Matching attack plans [Qin, Lee'04], [Wang, et al. '07]

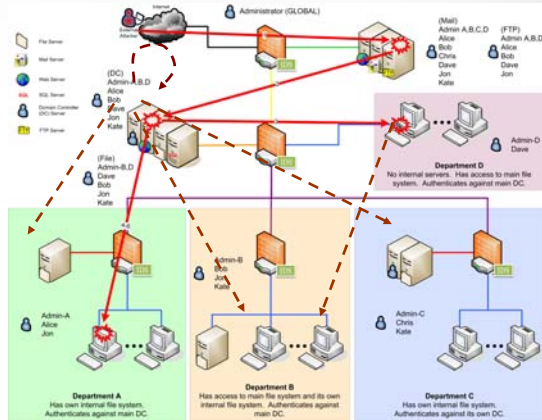


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Current Impact vs. Future Threat

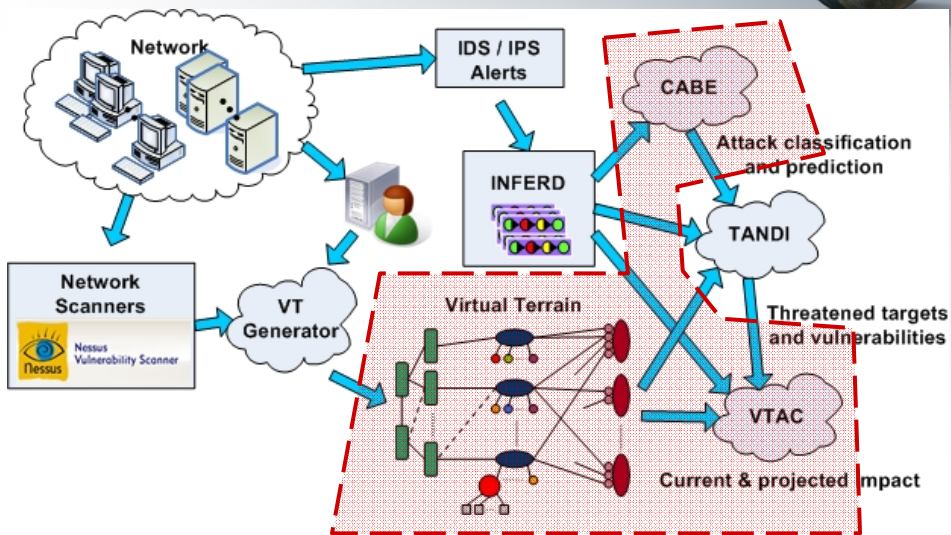


- Current: potential damage caused by observed attack tracks
- Future: anticipated moves (actions, targets, & time?) of attackers and the associated impact



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Overall System



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Cyber Context Needed

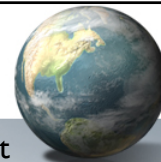


- Network information
 - Firewall / permission rules (analyst)
 - Users, accounts, and privileges (analyst)
 - Host, service, and user criticalities (analyst)
 - Local and remote services (scanner)
 - Mapping from services to vulnerabilities (scanner + databases)
 - Physical and logical subnet connectivity (scanner + analyst)
 - Exposure damage scores (common scoring system)

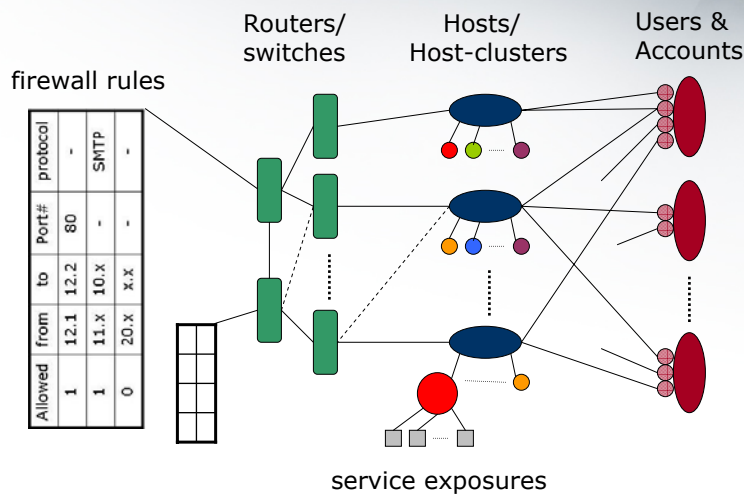
- Attack data
 - Network and host IDS alerts (correlation ground truth)
 - IDS locations (synchronized?)
 - Attack action (not alerts) ground truth?
 - Damage caused by attack actions (services and/or accounts compromised)

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Virtual Terrain



- A graph-based model for projection & impact assessment

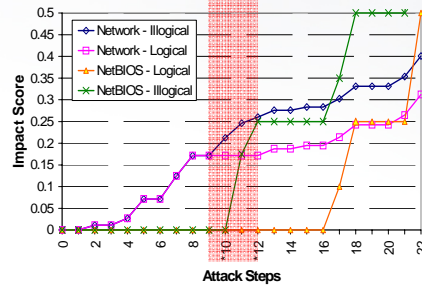


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VTAC for Impact Assessment



- VTAC (Virtual Terrain assisted impact Assessment for Cyber attacks)
 - Graph-based VT reduce the complexity of defining attack prerequisite-consequence rules
 - Identify illogical steps, potentially due to
 - error in VT model, missing alerts, zero-day attacks, coordinated attacks
 - Ranks network entities for each attack track
 - Ranks attack tracks w.r.t. different impact scores

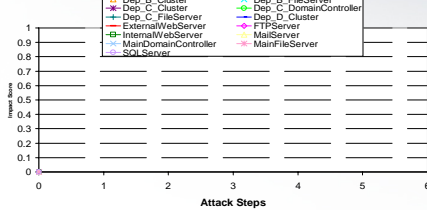


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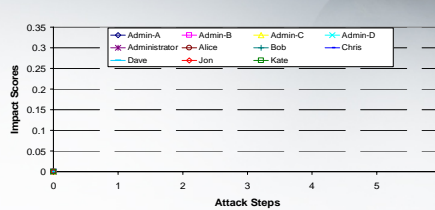
Step 0

Step	Description	Source IP	Dest IP	Port/Prot	Alert Signature
1	Compromise FTP Server	129.21.168.101	192.168.1.4	21/tcp	WEB-MISC /home/ftp access
2	Step to Internal Web Server	192.168.1.4	192.168.3.3	80/tcp	WEB-HTTP .asa HTTP header buffer overflow attempt
3	Attack Department D Cluster	192.168.3.3	192.168.4.100	23/tcp	TELNET bsd telnet exploit response
4	Ping Dep A Cluster	192.168.3.3	192.168.11.103	4566/tcp	ICMP-PING Microsoft Windows
5	Scan Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	SCAN SSH Version map attempt
6	Attack Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	EXPLOIT ssh CRC32 overflow

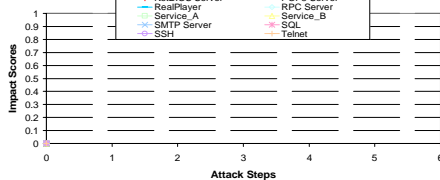
Hosts



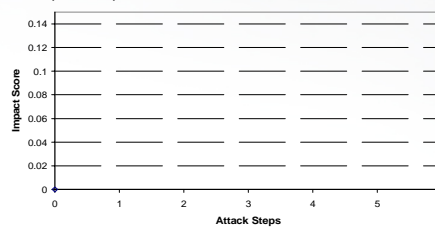
Users



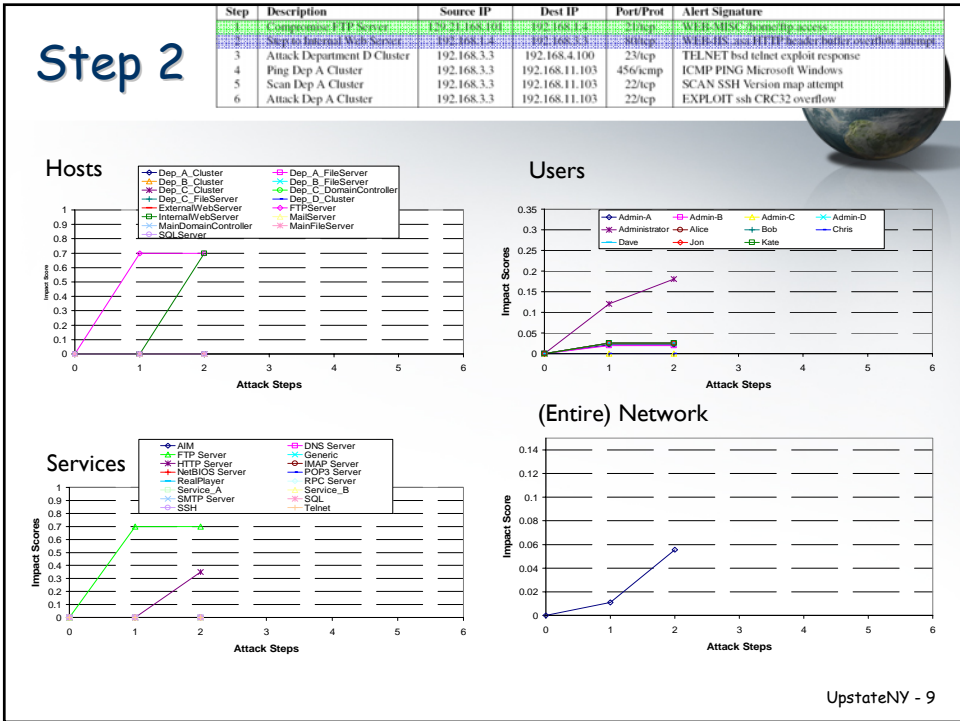
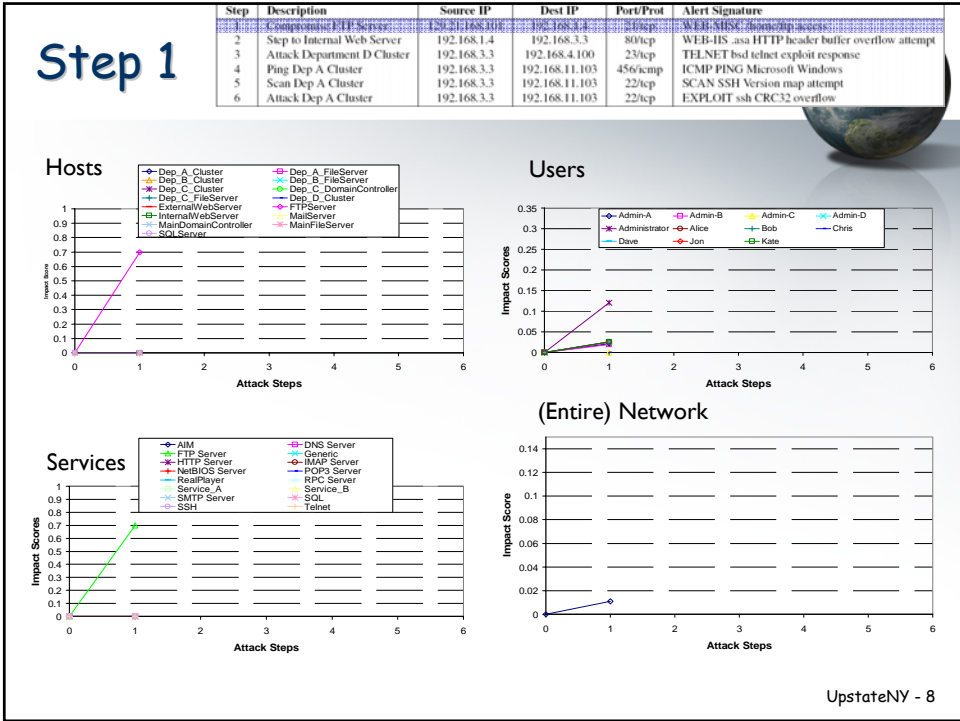
Services



(Entire) Network



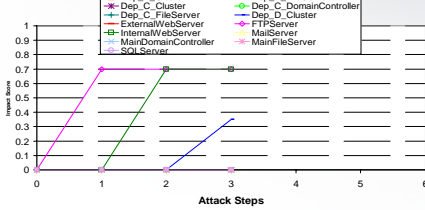
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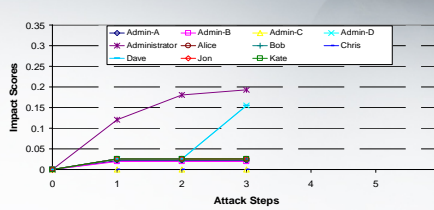
Step 3

Step	Description	Source IP	Dest IP	Port/Prot	Alert Signature
1	Compromise FTP Server	129.21.168.101	192.168.3.4	21/tcp	WEB-MISC:homeftp-access
2	Step to Internal Web Server	192.168.3.4	192.168.3.3	80/tcp	WEB-IDS:asa HTTP header buffer overflow attempt
3	Attack Department D Cluster	192.168.3.3	192.168.4.100	23/tcp	TELNET:bad telnet exploit response
4	Ping Dep A Cluster	192.168.3.3	192.168.11.103	456/icmp	ICMP-PING:Microsoft Windows
5	Scan Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	SCAN:SSH Version map attempt
6	Attack Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	EXPLOIT:ssh CRC32 overflow

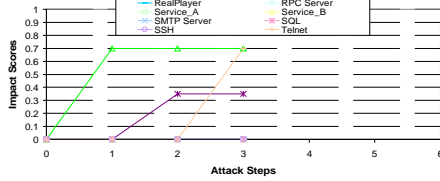
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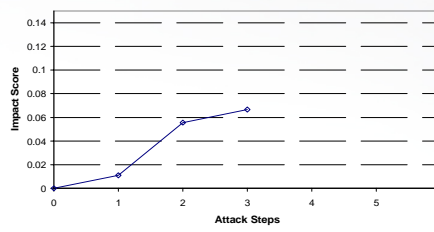
Users



Services



(Entire) Network

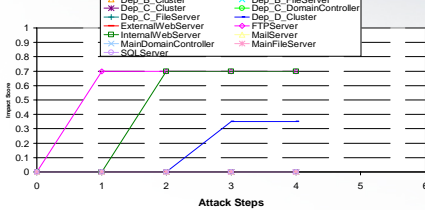


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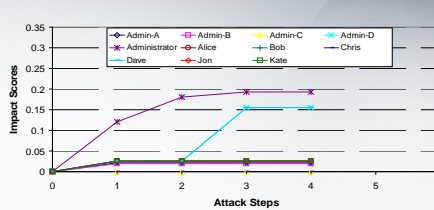
Step 4

Step	Description	Source IP	Dest IP	Port/Prot	Alert Signature
1	Compromise FTP Server	129.21.168.101	192.168.3.4	21/tcp	WEB-MISC:homeftp-access
2	Step to Internal Web Server	192.168.3.4	192.168.3.3	80/tcp	WEB-IDS:asa HTTP header buffer overflow attempt
3	Attack Department D Cluster	192.168.3.3	192.168.4.100	23/tcp	TELNET:bad telnet exploit response
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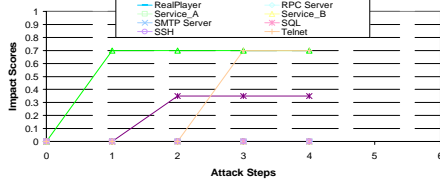
Hosts



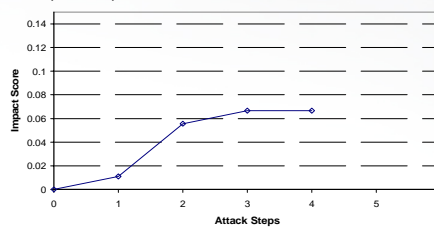
Users



Services



(Entire) Network

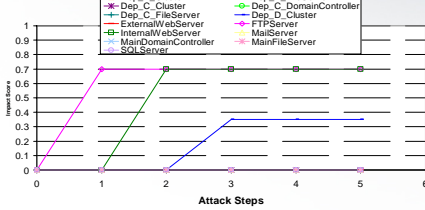


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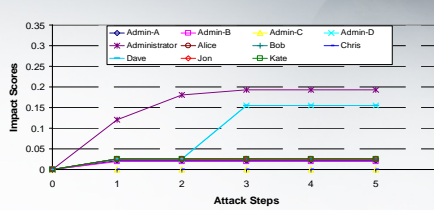
Step 5

Step	Description	Source IP	Dest IP	Port/Prot	Alert Signature
1	Compromise FTP Server	129.21.168.101	192.168.1.4	21/tcp	WEB-MISC:homeftp:access
2	Step to Internal Web Server	192.168.1.4	192.168.3.3	80/tcp	WEB-HTTP:asa:HTTP:header:buffer:overflow:attempt
3	Attack Department D Cluster	192.168.3.3	192.168.4.100	23/tcp	TELNET:bsd:telnet:exploit:response
4	Ping Dep A Cluster	192.168.3.3	192.168.3.1103	4566/tcp	ICMP-PING:Microsoft:Windows
5	Scan Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	SCAN-SSH:Microsoft:Windows
6	Attack Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	EXPLOIT:ssh:CRCC32:overflow

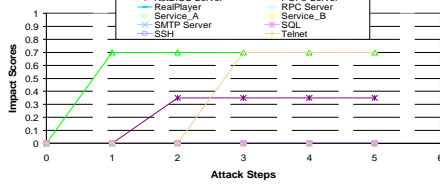
Hosts



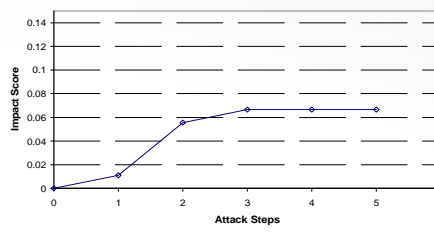
Users



Services



(Entire) Network

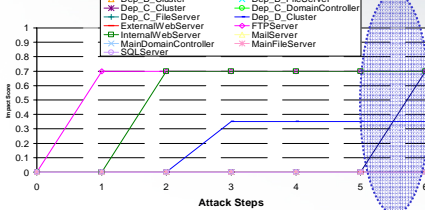


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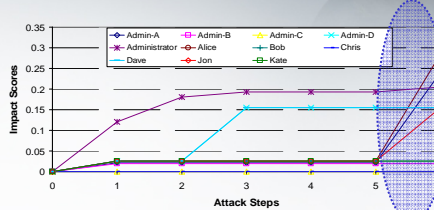
Projection

Step	Description	Source IP	Dest IP	Port/Prot	Alert Signature
1	Compromise FTP Server	129.21.168.101	192.168.1.4	21/tcp	WEB-MISC:homeftp:access
2	Step to Internal Web Server	192.168.1.4	192.168.3.3	80/tcp	WEB-HTTP:asa:HTTP:header:buffer:overflow:attempt
3	Attack Department D Cluster	192.168.3.3	192.168.4.100	23/tcp	TELNET:bsd:telnet:exploit:response
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6	Attack Dep A Cluster	192.168.3.3	192.168.11.103	22/tcp	EXPLOIT:ssh:CRCC32:overflow

Hosts



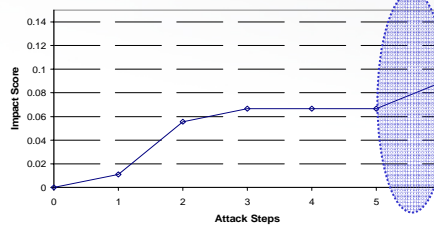
Users



Services

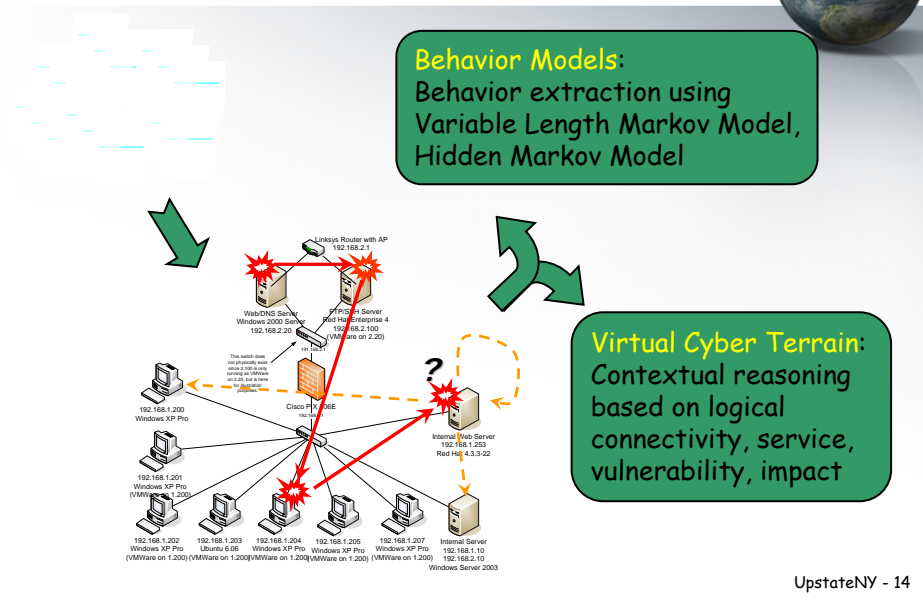


(Entire) Network



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Projection: Terrain vs. Behavior



Behavior analysis - how?



- Expert developed behavior model
 - E.g., guidance template, Bayesian Network
 - Diverse SME opinions (knowledge elicitation?)
 - Costly to maintain and update
 - **Attack tracks** → time-stamp ordered sequences of symbols
 - Context-based model
 - Adaptive Bayesian Network [Qin, Lee'04], Data Mining [Li et al.'07]
 - 0th, 1st, 2nd, 3rd order Markov Model
 - **Variable-length Markov Model (VLMM)**
 - Universal Predictor [Jacquet et al '02]
 - Q: What should be the context?
 - State-based model
 - Hidden Markov Model (feasible?)
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Alert Prediction Example



- Alerts Generated by Attack Actions
 - Predicted Alerts
- | | |
|--|---|
| <ol style="list-style-type: none"> 1. K (http_inspect) Oversize Request-URI Directory 2. F (http_inspect) Bare Byte Unicode Encoding 3. A ICMP PING NMAP 4. H ICMP L3retriever Ping 5. J WEB-MISC Invalid HTTP Version String 6. J WEB-MISC Invalid HTTP Version String 7. A ICMP PING NMAP 8. H ICMP L3retriever Ping 9. H ICMP L3retriever Ping 10. I NETBIOS SMB-DS IPC\$ unicode share access 11. A ICMP PING NMAP 12. H ICMP L3retriever Ping | <ol style="list-style-type: none"> 1. (no prediction) 2. J WEB-MISC Invalid HTTP Version String 3. J WEB-MISC Invalid HTTP Version String 4. H ICMP L3retriever Ping 5. A ICMP PING NMAP 6. J WEB-MISC Invalid HTTP Version String 7. J WEB-MISC Invalid HTTP Version String 8. H ICMP L3retriever Ping 9. F (http_inspect) Bare Byte Unicode Encoding 10. I NETBIOS SMB-DS IPC\$ unicode share access 11. A ICMP PING NMAP 12. H ICMP L3retriever Ping |
|--|---|

Bottom Line: Is there a reliable pattern (in what context) to extract for prediction?

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Translating Alerts



- <Alert>
 - <Description>ICMP PING NMAP</Description>
 - <Dest_IP>100.20.0.0</Dest_IP>
 - <Category>Recon_Scanning</Category>
- </Alert>
- <Alert>
 - <Description>SCAN SOCKS Proxy attempt</Description>
 - <Dest_IP>100.10.0.1</Dest_IP>
 - <Category>Recon_Scanning</Category>
- </Alert>
- <Alert>
 - <Description>WEB-IIS nsiislog.dll access</Description>
 - <Dest_IP>100.20.0.0</Dest_IP>
 - <Category>Intrusion_Root</Category>
- </Alert>

Category & target IP (Ω_t): AaB

Description (Ω_d): ABC

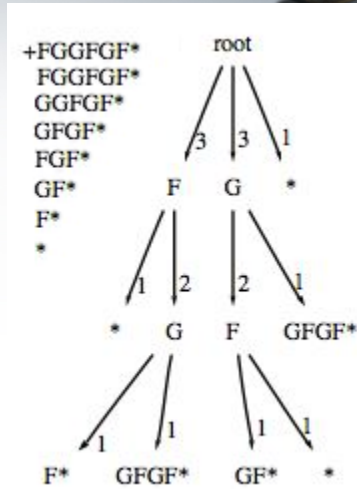
Category (Ω_c): AAB

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Suffix Tree and Prediction



- +FGGFGF*
 - +: start of attack track
 - F: WEB-IIS nsiislog.dll access
 - G: WEB-MISC Invalid HTTP Version String
 - *: end of attack track
- What follows +GF?
 - -1th order: $P=1/3$
 - 0th order: $P\{G\}=P\{F\}=3/7$, $P\{*\}=1/7$
 - 1st order: $P\{G|F\} = 2/3$, $P\{*|F\} = 1/3$
 - 2nd order: $P\{G|GF\} = 1/2$, $P\{*|GF\} = 1/2$
 - VLMM – blending the estimates

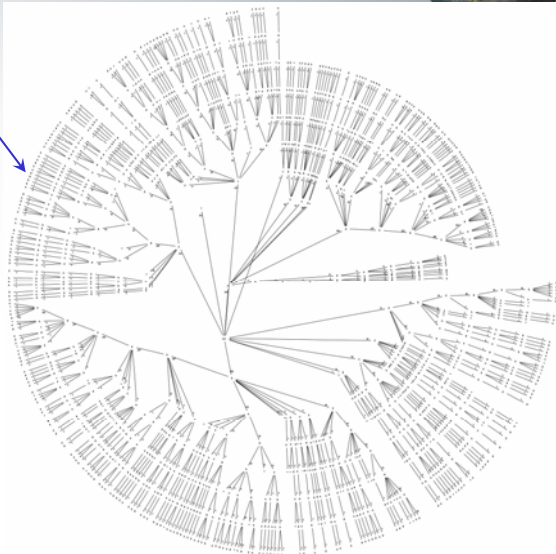


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Suffix tree from historical data



- Historical attack sequences builds suffix tree
- Suffix tree embeds patterns exhibited in finite-contexts
- Each unfolding attack sequence matches part of suffix tree for prediction



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Experiment Setup



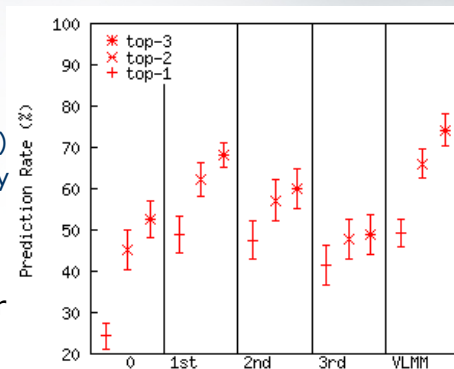
- Ground truth data generated via scripted attacks on a VMWare network
- A total of 1,113 attack sequences composed of 4,723 alerts after $\Delta t=1$ filtering [Valuer'04]
- 10 independent runs with random 50-50 splits of training vs. test data
- Alphabet choices:
 - Specific attack method (Ω_d)
 - Category of attack method (Ω_c)
 - Category + target IP (Ω_t)
- Top- k prediction rate ($k=1, 2, 3$):
 - % of correct prediction falls in the top- k choices

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0 to 3rd Order and VLMM (Ω_d)



- Dominance of 1st order prediction
- VLMM combines n-order and offers better predictions
- Top 3 actions:
 - ICMP PING NMAP (43%), WEB-MISC Invalid HTTP Version String (22.4%), (http inspect) BARE BYTE UNICODE ENCODING (9.0%)
 - ICMP PING NMAP followed by ICMP PING NMAP 87.7% of the time
- Predicts better for repeating actions? Blending with longer context helps for predicting transitions?



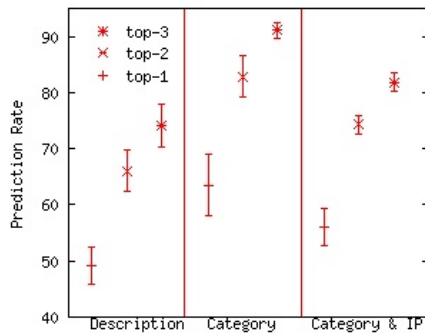
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Prediction rate for transitions

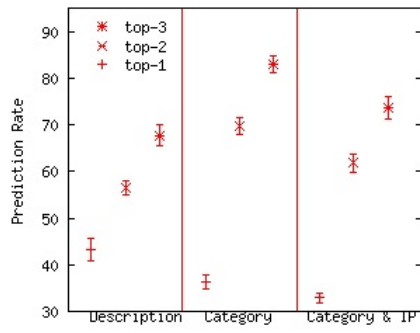


- Predicting transitions will be better off by training with data sets with no repetition
- Predicting attack category is easier and more reasonable than predicting specific attack method

Trained with no repetition

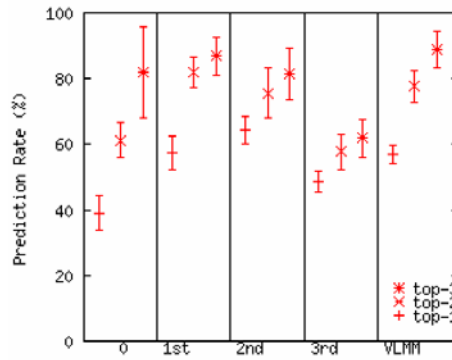
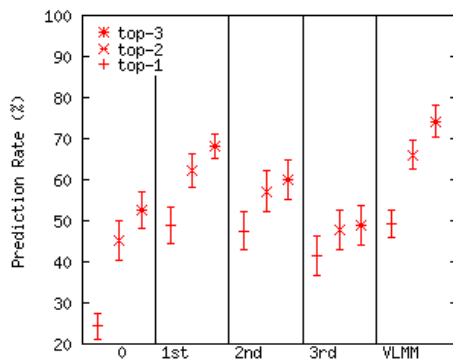


Trained with repetition



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Ω_c better than Ω_d



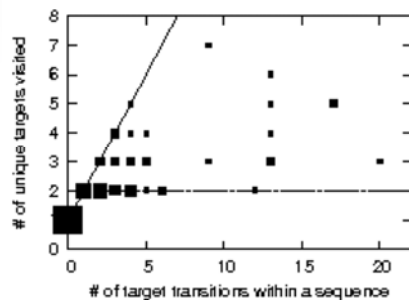
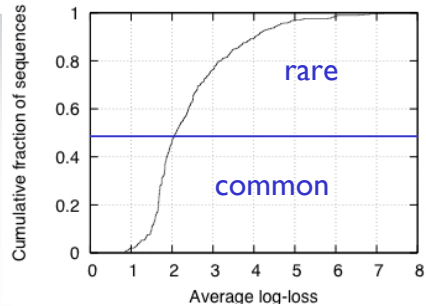
- Coarser granularity yield better prediction rate – as expected
- Network independent prediction – analyst may only want to know prediction at a coarser granularity

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Classification?



- Can we categorize cyber attack types (with no ground truth)?
- Average Log-loss:
 - Rarity of attack sequence
 - Threshold=2.0 (Ω_c , no repetition)
 - 0.83 vs. 0.69 prediction rates
- # target trans vs. # targets visited:
 - Agility of attack
 - Most targets suffered 2 scans
 - Most popular targets: 1,735 and 814 out of a total of 4,723
 - Are more agile attacks harder to predict?



Conclusion



- **Proactive** impact assessment and projection of cyber attacks!!
- Graph-based VT defines dynamic relationships between network entities
 - Automatic update is not an easy task
- VTAC determines attack's impact to network elements
 - How to validate its performance?
- Behavior-based attack prediction
 - A new theoretical and real-world problem
 - Diverse, changing, and noisy behavior