

Research report at the 72nd IFIP WG 10.4 meeting

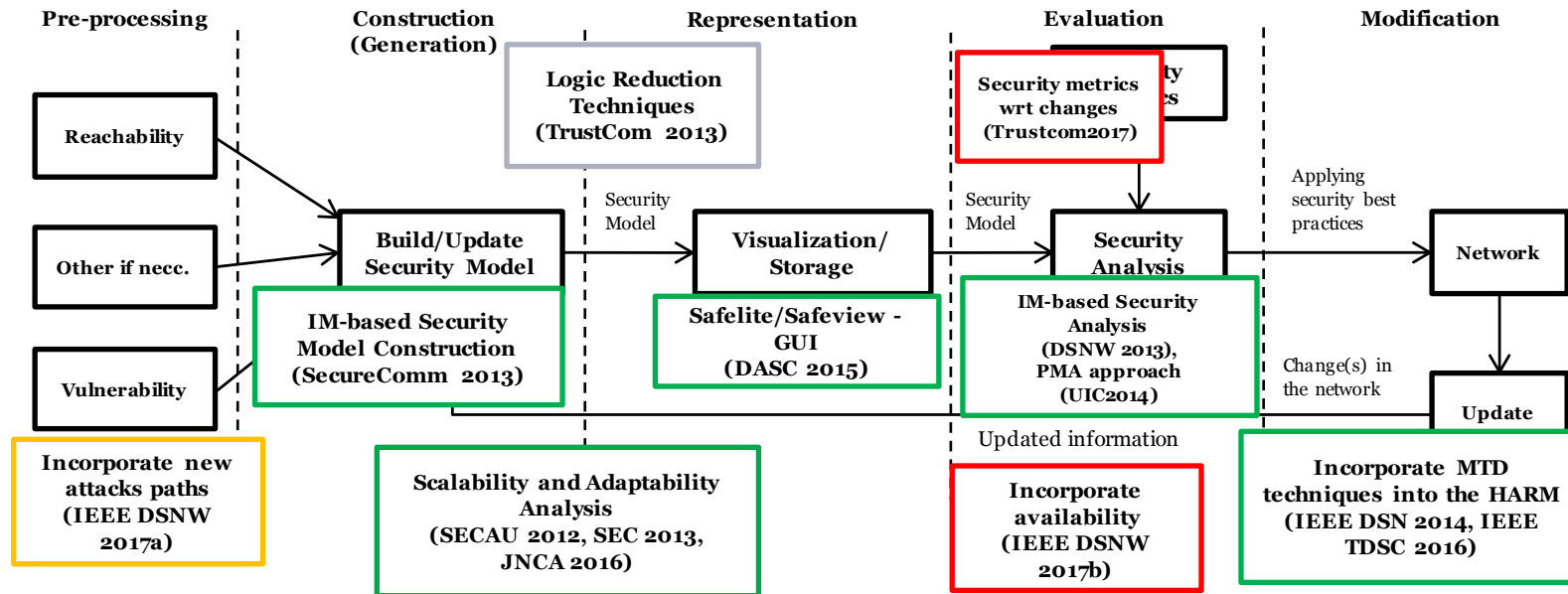
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Recent results

- Evaluating Security and Availability of Multiple Redundancy Designs, DSNW 2017
- Evaluating the Effectiveness of Security Metrics for Dynamic Networks, IEEE Trustcom 2017
- Discovering and Mitigating New Attack Paths, DSNW 2017

Graphical Security Models: our selected research contributions



SECAU 2012: "HARMs: Hierarchical Attack Representation Models for Network Security Analysis"

IFIP SEC 2013: "Performance analysis of scalable attack representation models"

IEEE TrustCom 2013: "Scalable Attack Representation Model Using Logic Reduction Techniques"

IEEE DSNW 2013: "Scalable Security Analysis in Hierarchical Attack Representation Model using Centrality Measures"

SecureComm 2013: "Scalable Security Model Generation and Analysis using k-importance Measure"

IEEE DSN 2014: "Scalable Security Models for Assessing Effectiveness of Moving Target Defenses"

IEEE DSNW 2014: "What Vulnerability Do We Need To Patch First?"

IEEE UIC 2014: "Scalable Security Analysis using Partition and Merge Approach in an Infrastructure as a Service Cloud"

IEEE DASC2015: "Towards Automated Generation and Visualization of Hierarchical Attack Representation Models"

IEEE TDSC 2016: "Assessing the Effectiveness of Moving Target Defense using Security Models"

Elsevier JNCA 2016: "Towards scalable security analysis using multi-layered security models"

IEEE DSNW2017a: "Discovering and Mitigating New Attack Paths using Graphical Security Models"

IEEE DSNW2017b: "Evaluating Security and Availability of Multiple Redundancy Designs"

IEEE Trustcom 2017: "Evaluating the Effectiveness of Security Metrics for Dynamic Networks"

Evaluating Security and Availability of Multiple Redundancy Designs when Applying Security Patches

Mengmeng Ge¹, Huy Kang Kim², Dong Seong Kim¹

¹University of Canterbury, New Zealand

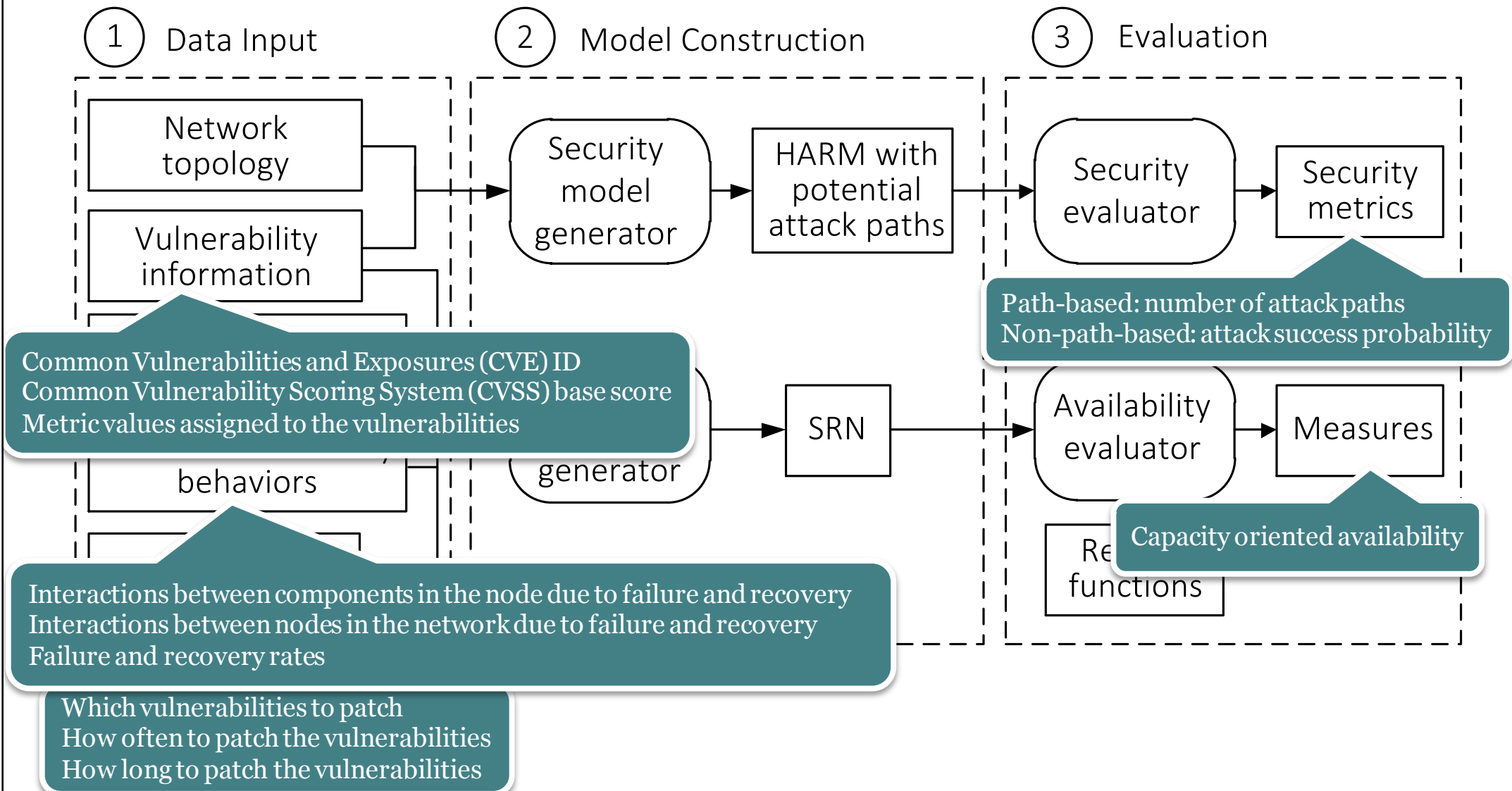
²Korea University, South Korea

Introduction

- Centralized patch management
 - Enhance security
- Some security patches require system reboot
 - Introduce downtime
- Redundant servers
 - Improve availability
 - Increase attack surface
- Balance between security and availability affected by the security patch

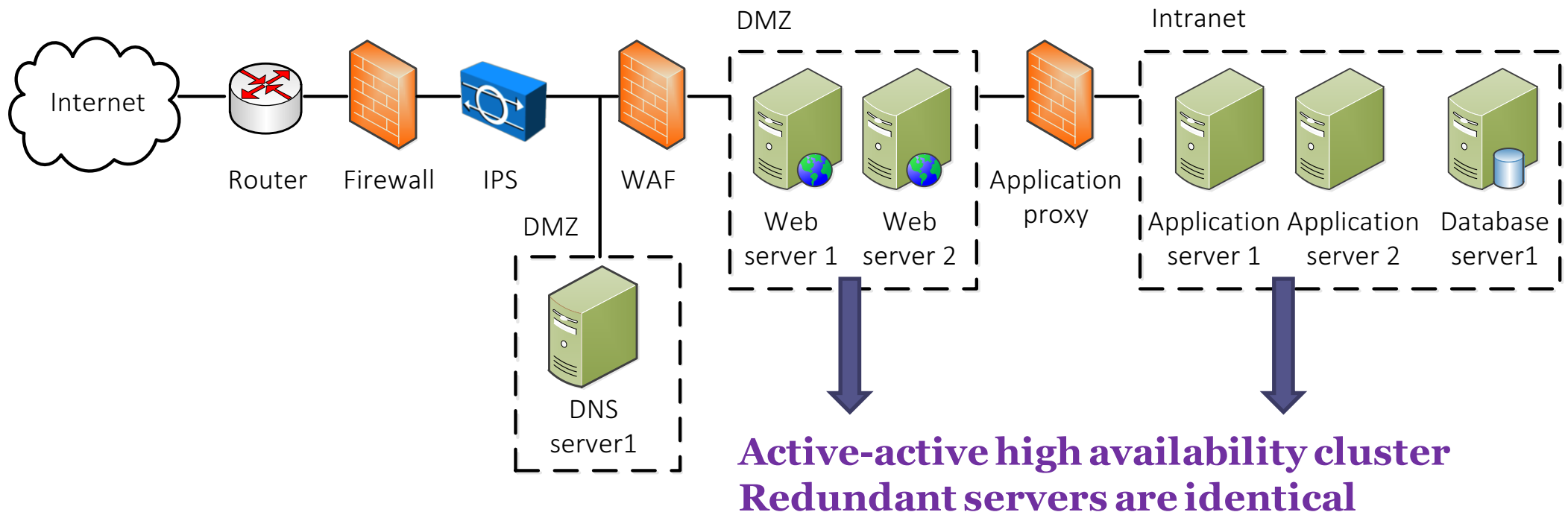
Proposed Approach

Hierarchical Attack Representation Model (HARM)
Stochastic Reward Net (SRN)



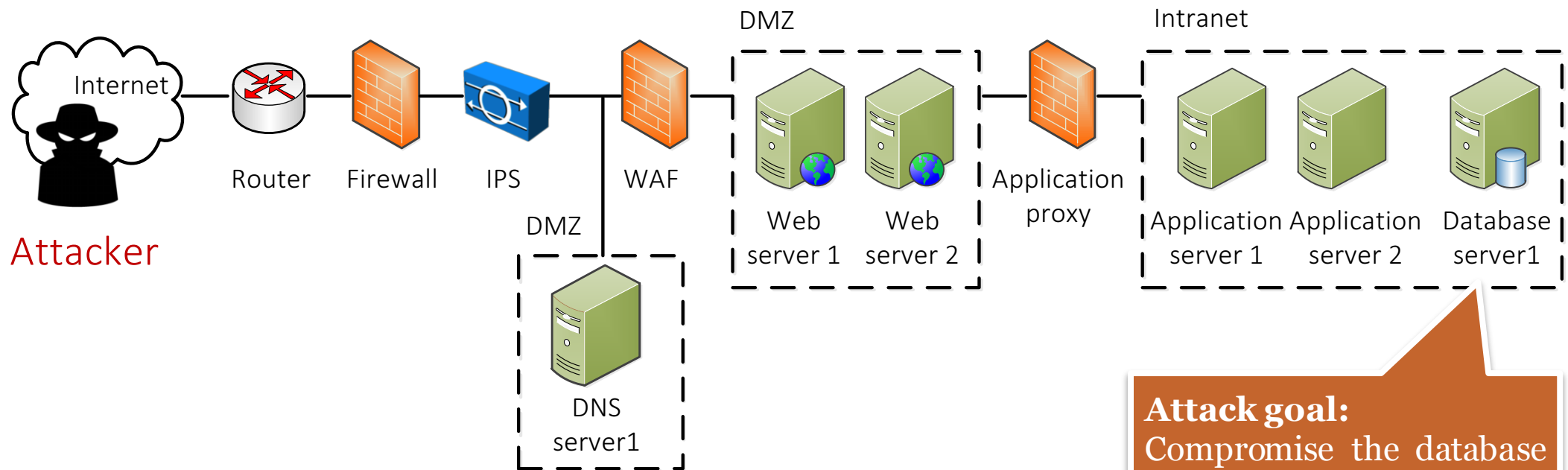
Example Enterprise Network

- 3-tier client-server architecture



An Attacker Model - Remote Attacker

- Laptop-class device with attack tools

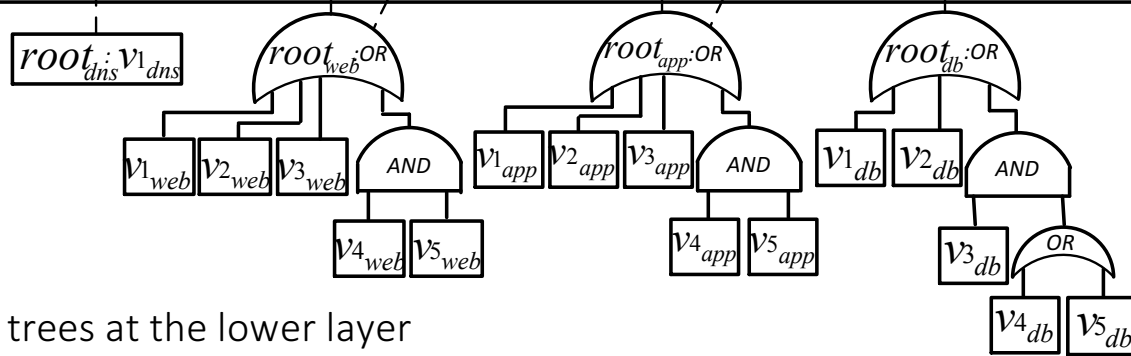
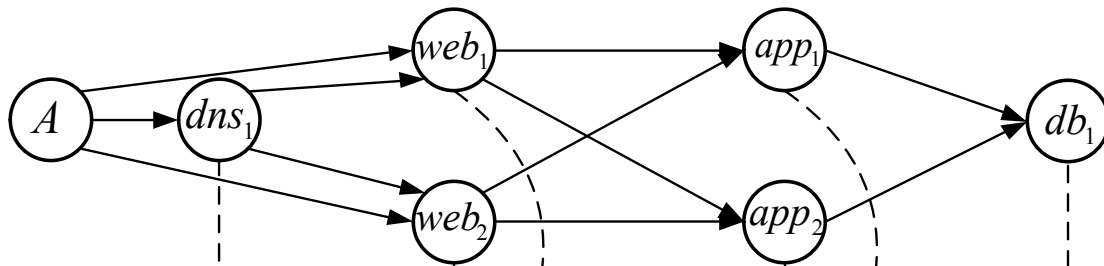


Attack goal:
Compromise the database server(s) through privilege escalation attacks

Construction of HARMs

Before patch

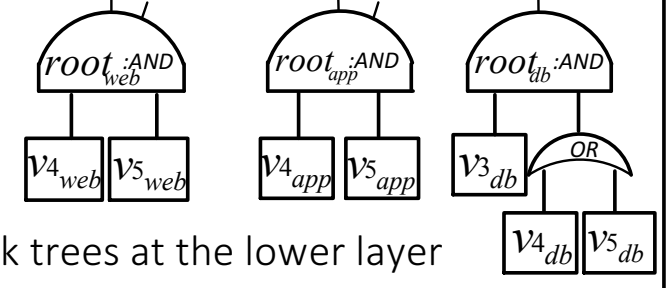
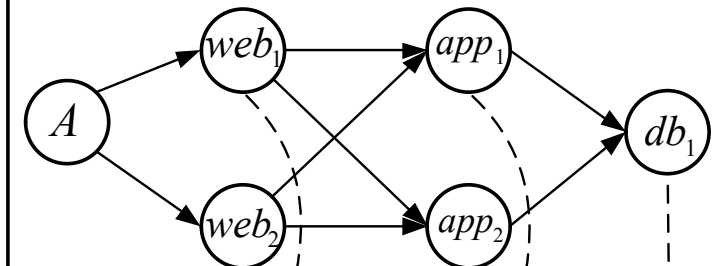
Attack graph at the upper layer



Attack trees at the lower layer

After patch

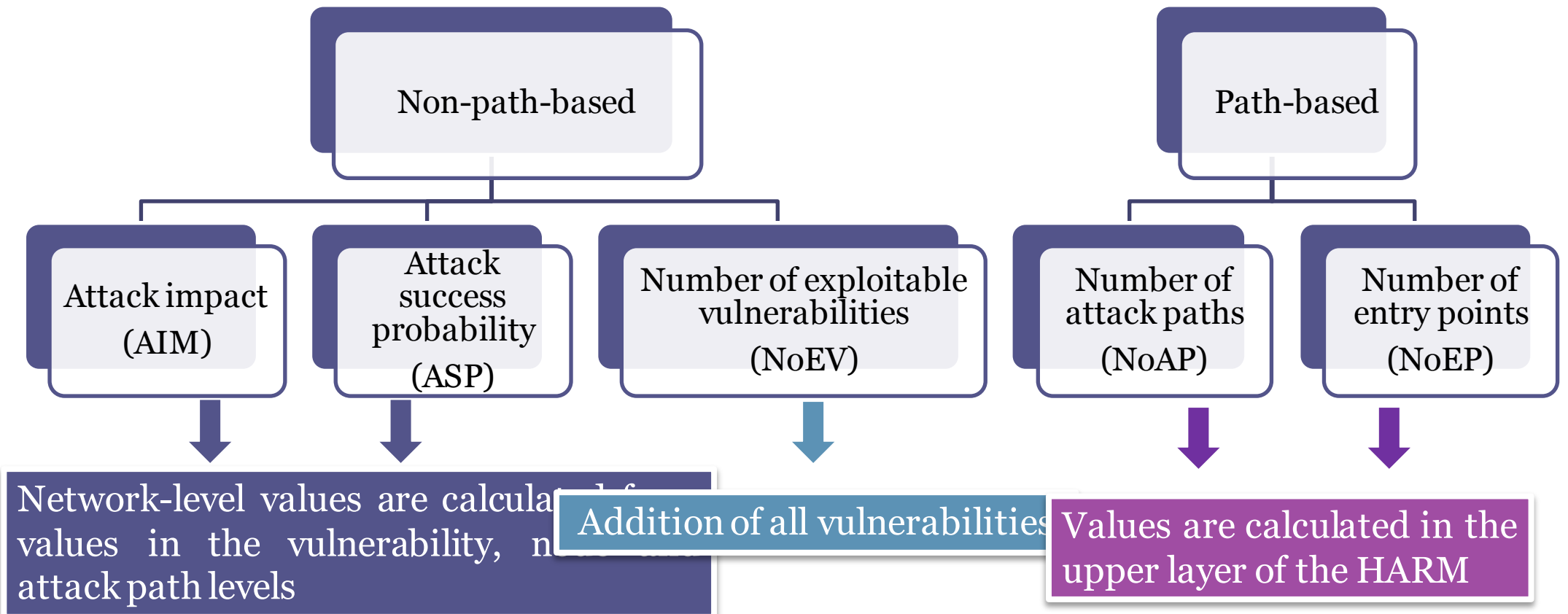
Attack graph at the upper layer



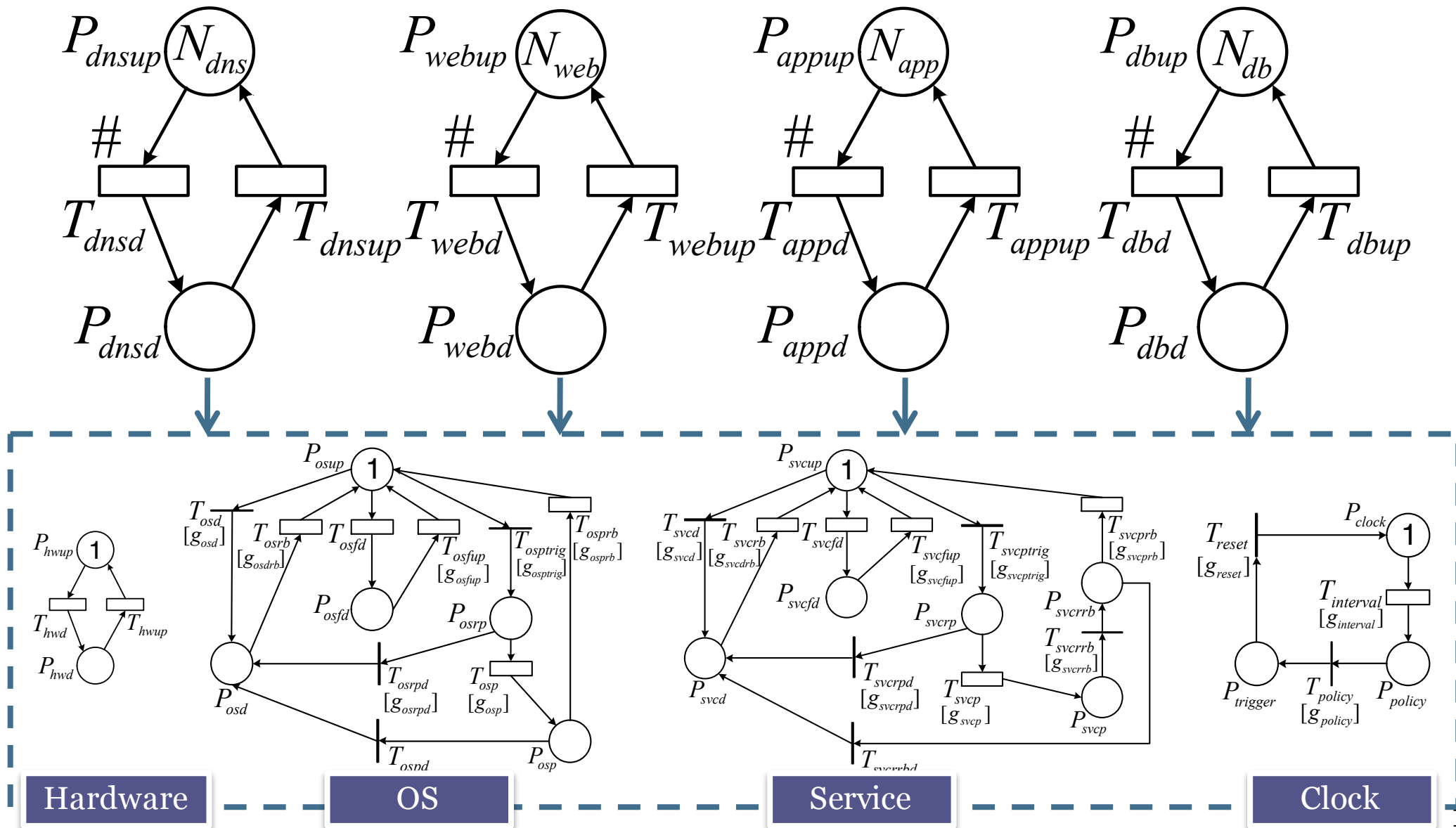
Attack trees at the lower layer

Construction of HARMs (cont.)

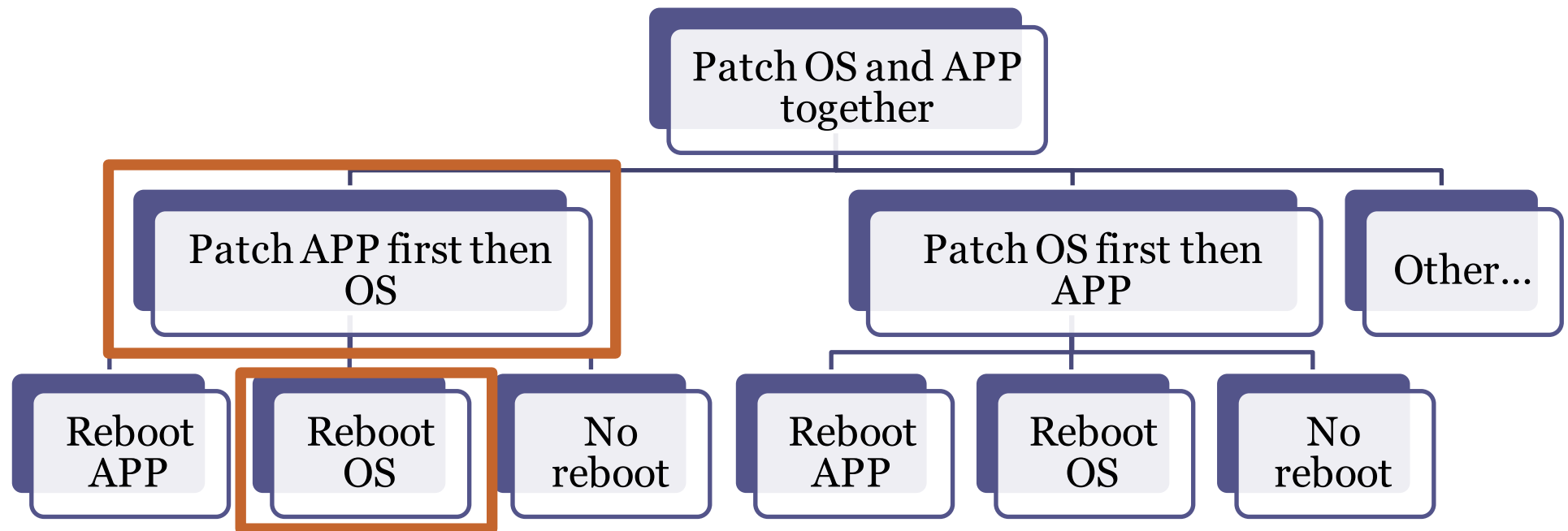
- Security metrics for the security analysis:



Construction of SRN models



Construction of SRN models (cont.)



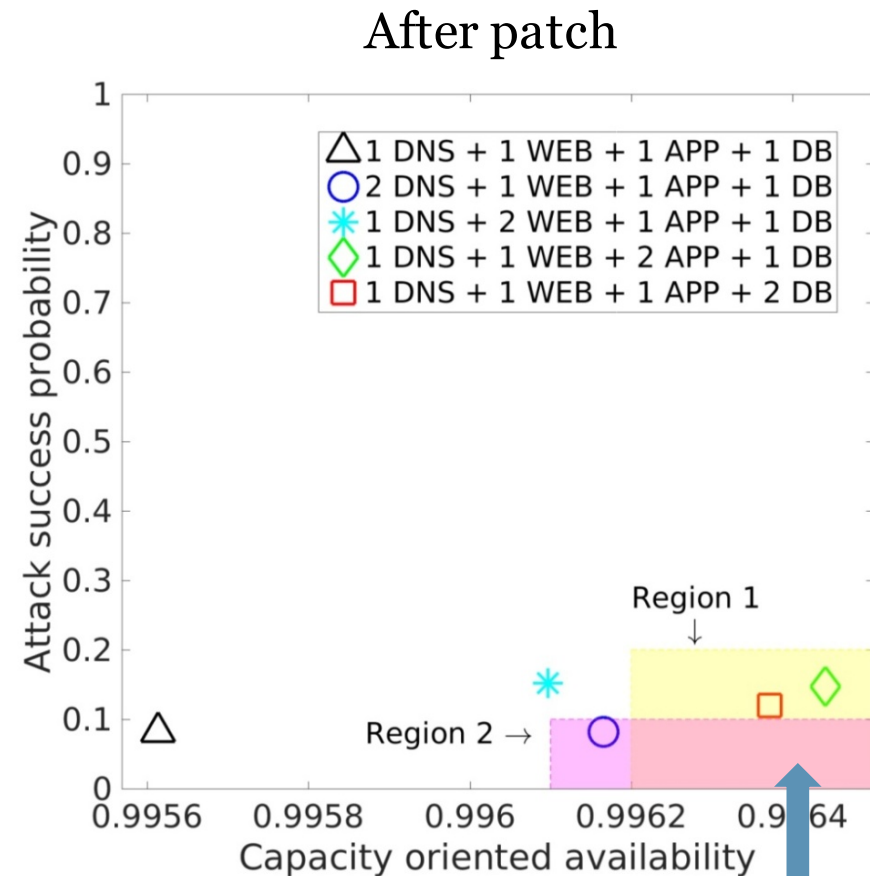
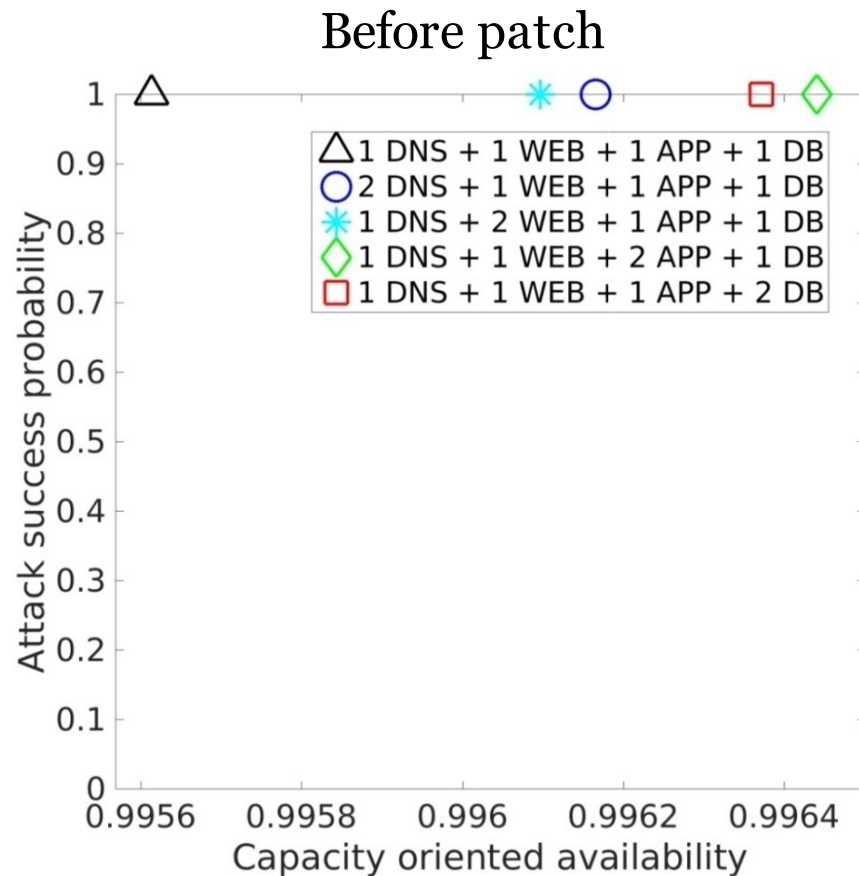
Construction of SRN models (cont.)

- Output measure of SRN sub-models for the network
 - Capacity oriented availability (COA)

Reward	Definition
<i>COA</i>	if ($\#P_{dnsup} == 1 \ \&\& \ \#P_{webup} == 2 \ \&\& \ \#P_{appup} == 2 \ \&\& \ \#P_{dbup} == 1$) 1 else if ($\#P_{dnsup} == 1 \ \&\& \ \#P_{webup} == 1 \ \&\& \ \#P_{appup} == 2 \ \&\& \ \#P_{dbup} == 1$) 0.83333 else if ($\#P_{dnsup} == 1 \ \&\& \ \#P_{webup} == 2 \ \&\& \ \#P_{appup} == 1 \ \&\& \ \#P_{dbup} == 1$) 0.83333 else if ($\#P_{dnsup} == 1 \ \&\& \ \#P_{webup} == 1 \ \&\& \ \#P_{appup} == 1 \ \&\& \ \#P_{dbup} == 1$) 0.66667 else 0

Reward rate is the number of running servers during patch divided by the total number of servers.
 0.83333 (5/6)
 0.66667 (4/6)
COA ≈ 0.99707

Numerical Analysis



$$f(ASP, COA) = \begin{cases} 1, & \text{if } ASP \leq \phi \text{ and } COA \geq \psi \\ 0, & \text{if } ASP > \phi \text{ or } COA < \psi \end{cases}$$

Limitations and Potential Extensions

- Measurement using Testbed (we have a Software Defined Cloud at UC)
- Systems
 - Large scale; heterogeneous redundancy
- SRN availability models
 - Patch schedules; reboot cases;...
- User oriented performance/performanceability
 - Queuing network (e.g., mean response time, mean waiting time, dropping probability) ...
- Other Dependability and Security Metrics
 - Opex/capex as output measure
 - Economic metrics (e.g., gain of high availability vs. cost of redundancy; loss of successful attacks vs. cost of security patch) ...

My previous work on Availability and performance:

Dong Seong Kim, Fumio Machida, Kishor S. Trivedi: Availability Modeling and Analysis of a Virtualized System. PRDC 2009: 365-371

Rahul Ghosh, Kishor S. Trivedi, Vijay K. Naik, Dong Seong Kim: End-to-End Performance Analysis for Infrastructure-as-a-Service Cloud: An Interacting Stochastic Models Approach. PRDC 2010: 125-132

Fumio Machida, Dong Seong Kim, Kishor S. Trivedi: Modeling and analysis of software rejuvenation in a server virtualized system with live VM migration. Perform. Eval. 70(3): 212-230 (2013)

Tuan Anh Nguyen, Dong Seong Kim, Jong Sou Park: Availability modeling and analysis of a data center for disaster tolerance. Future Generation Comp. Syst. 56: 27-50 (2016)

Evaluating the Effectiveness of Security Metrics for Dynamic Networks

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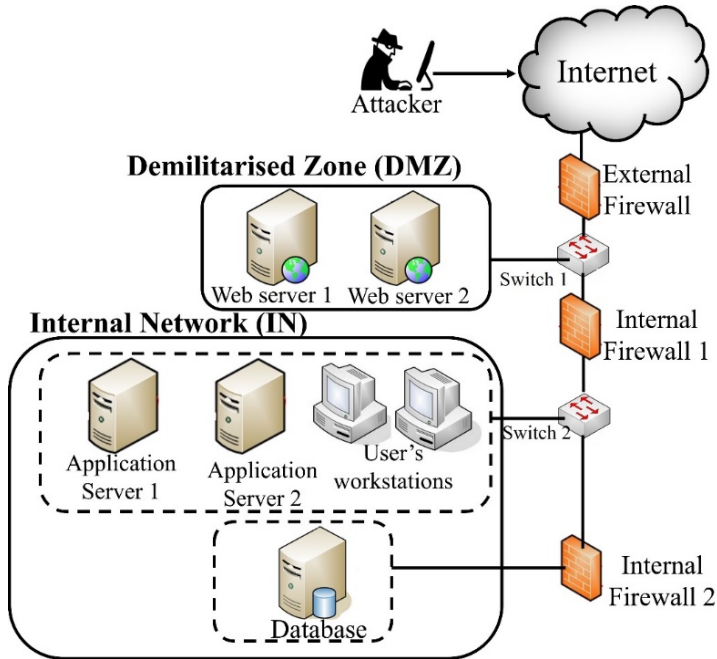
Our approach

- Previous graphical security model – Hierarchical Attack Representation Model (HARM)^{1,2}
- Need to incorporate changes in the networks
- propose to use a **Temporal**-Hierarchical Attack Representation Model (T-HARM)*
- to investigate the varying effects of security metrics when changes are observed in a dynamic network (e.g. enterprise nets, Cloud).
 - T-HARM is a layered and scalable security model that captures the temporal changes.

¹Jin B. Hong, Dong Seong Kim: Assessing the Effectiveness of Moving Target Defenses Using Security Models. *IEEE Trans. Dependable Sec. Comput.* 13(2): 163-177 (2016)

²J. B. Hong and D. S. Kim, “Towards Scalable Security Analysis using Multi-layered Security Models,” *Elsevier Journal of Network and Computer Applications*, vol. 75, pp. 156 – 168, 2016.

Example network and Attacker Model



List of vulnerabilities at t_2		
Host	Vul.	Base score
WS_1	CVE-2015-3566	4.3
	CVE-2000-1247	2.1
WS_2	CVE-2015-3566	4.3
	CVE-2000-1247	2.1
AS_1	CVE-2013-0638	10.0
	CVE-2016-0763	4.3
AS_2	CVE-2013-0900	4.3
	CVE-2015-3566	4.3
DB	CVE-2012-1675	7.5
	CVE-2016-3201	4.3
User's	CVE-2016-2834	8.8
	CVE-2016-7218	1.9

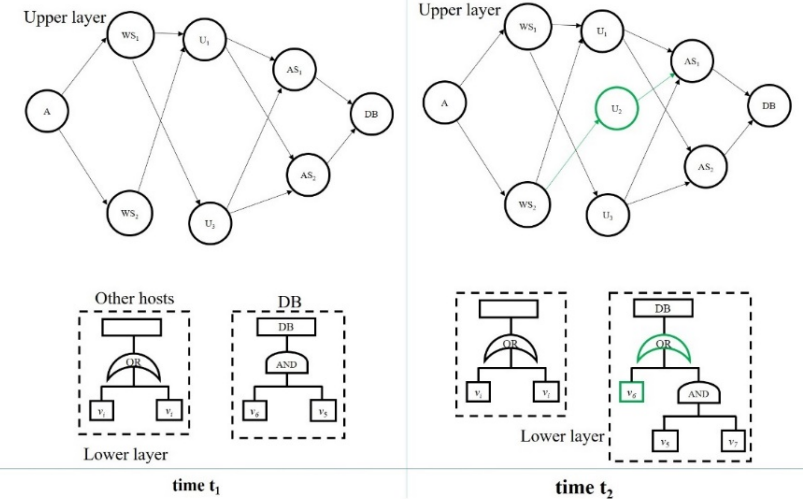


Fig: T-HARM

Table 1: OSs and Applications on hosts

Host	OS	Service
WS_1	Redhat Enterprise Linux 6	Apache http server 2.4
WS_2	Redhat Enterprise Linux 6	Apache http server 2.4
AS_1	Windows 10	WebLogic server 12.1
AS_2	Redhat Enterprise Linux 6	Apache tomcat 7.0
DB	Windows 10	Oracle database 11g
WT_s	Redhat Enterprise Linux 6	Mozilla firefox 31.1.0

Changes at t_2		
Host	Vul.	Base score
DB	CVE-2015-2465	2.1
New User	CVE-2016-2834	8.8
	CVE-2016-7218	1.9

Security metrics

- We investigate the following security metrics (ten security metrics):

Metrics	Notations
Risk on attack paths	R
Cost on attack paths	AC
Probability of attack success on paths	Pr
Return on attack paths	ROA
Standard deviation of attack path lengths [6]	SDPL
Mean of attack path lengths [14]	MAPL
Number of attack paths [12]	NAP
Mode of attack path lengths [6]	MoPL
Shortest attack path [12]	SAP
Normalised mean of attack path lengths [6]	NMPL

Security metrics (cont.)

- Based on Common Vulnerability Scoring System (CVSS) based score, we assigned values to:
 - the probability of attack success (pr),
 - attack impact (aim) and
 - attack cost (ac) to each vulnerability in the network
 - The CVSS provides standardised vulnerability score which is ranging from 0.0 to 10.0 (with 10.0 being the most severe level).
- We introduce time for each metric, we then use them for the security analysis (e.g., for Risk on attack paths (R), we use it as Risk on attack paths at time t (R_t) to compute the metric at a specific time).

Host (h)	Vul.	Base score	ac_h	aim_h	Pr_h
WS_1	CVE-2015-3566	4.3	5.7	5.0	0.43
	CVE-2000-1247	2.1	7.9	3.0	2.1
WS_2	CVE-2015-3566	4.3	5.7	5.0	0.43
	CVE-2000-1247	2.1	7.9	3.0	2.1
AS_1	CVE-2013-0638	10.0	0.1	10.0	1.0
	CVE-2016-0763	4.3	7.9	5.3	
AS_2	CVE-2013-0900	4.3	7.9	5.3	0.43
	CVE-2015-3566	4.3	7.9	5.3	0.43
DB	CVE-2012-1675	7.5	2.5	8.0	0.75
	CVE-2016-3201	4.3	7.9	5.0	0.43
User's	CVE-2016-2834	8.8	1.2	9.0	0.88
	CVE-2016-7218	1.9	8.1	2.0	0.19

Defense model

- We use the patching of vulnerabilities as the defense for our simulation.
- In particular, we adopt the prioritised set of vulnerabilities using the hybrid method* to determine important vulnerability to patch first (since it is infeasible to patch all vulnerability)
 - the Prioritized set of vulnerabilities (PSV) is defined as a set of vulnerabilities which are most important to enhance security (e.g., to minimize the system risk).

*J. B. Hong, D. S. Kim, and A. Haqiq. What Vulnerability Do We Need to Patch First? In Proceeding of the DSNW 2014.

Changes

- investigate the varying effects of security metrics when changes are observed in the network, we conduct various analysis with different types of changes via the T-HARM.
 - Emergence of new vulnerabilities
 - Patching of vulnerabilities with the emergence of vulnerabilities
 - Addition of new hosts (hosts having vulnerabilities)
 - Removal of existing hosts
 - Change of firewall rules.

Summary

- the existing security metrics response to changes in different ways when we introduced time to them.
 - We found that, depending on the types of security change the different security metrics (except the SAP) can show change in their value when there is a change in the network system and configuration.

Security metrics	Emergence of vulnerabilities	Patching and emergence of vulnerabilities	Addition of hosts	Removal of hosts	Firewall rules change
R	✓	✓	✗	✗	✓
AC	✓	✓	✗	†	†
Pr	✓	✓	✗	✗	†
ROA	✓	✓	✗	†	✓
SAP	✗	✗	✗	✗	✗
NAP	✗	✓	✓	✓	✓
MAPL	✗	✓	✓	✓	✓
NMPL	✗	✓	✓	✓	✓
SDPL	✗	✓	✓	✓	✓
MoPL	✗	✓	✓	✗	✓

Legend

- Significant change(✓)
- Small change(†)
- No change(✗)

On-going work

- Time-independent Security models
 - T-HARM takes snapshots of a dynamic network at t_i (*event, users, batch...*)
 - Issues
 - we may miss some states.
 - infeasible to cover all possible states.
- For each state (lets stay we choose sampling method (1)), we can compute the risk of the given Network (e.g., Cloud) state.
- Two ideas:
 - (a) we aggregate all the states and compute security metrics.
 - (b) we give weights to each state based on the observation and aggregate risk based on the weight of each state.
 - Make a state space model (Markov, Petri net models) to capture the state transitions and other info.

Thank you!



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