

Traffic Monitoring and Diagnosis with Multivariate Statistical Network Monitoring: A Case Study

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Motivation - I

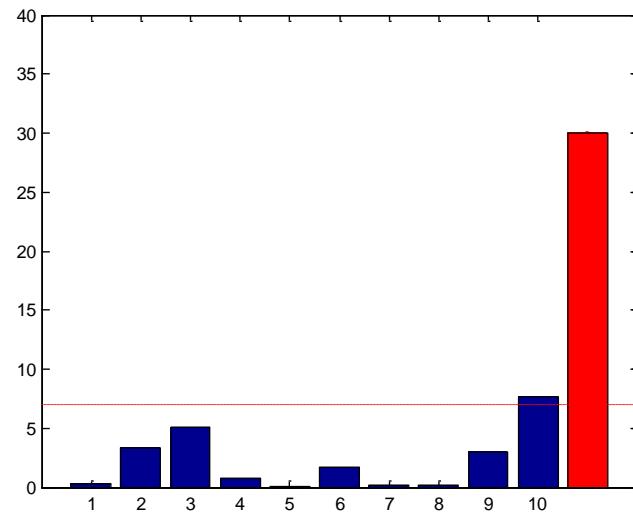
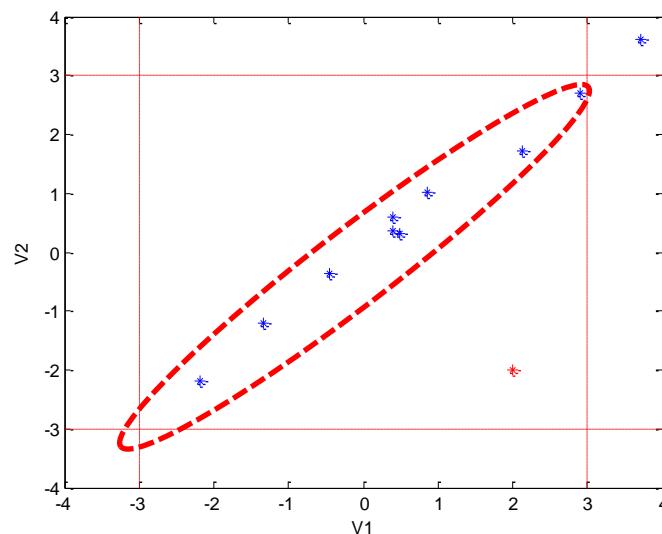
- ✓ Anomaly Detection from Traffic Data
 - ✓ Accuracy & Variate Sources
- ✓ Software for CyberSec:
 - ✓ Pivoting (**Specific**)
 - ✓ **High False Anomalies** (Correlation)
- ✓ CyberSec Research → ML
 - ✓ Data Fusion (**General**)
 - ✓ **High Detection** but **Semantic Gap**
- ✓ Multivariate Statistical Network Monitoring
 - ✓ Data Fusion (**General**)
 - ✓ **High Detection & Diagnosis**



Multivariate approach

In a data set with many measured variables, the interesting information is contained in a (much lower) number of **latent variables**

Multivariate Statistical Control (PCA)



Multivariate Network Security Monitoring (MSNM)

COMPUTERS & SECURITY 59 (2016) 118–137



Available online at www.sciencedirect.com
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Computers & Security



PCA-based multivariate statistical network monitoring for anomaly detection

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2014 IEEE INFOCOM Workshops: 2014 IEEE INFOCOM Workshop on Security and Privacy in Big Data

Tackling the Big Data 4 Vs for Anomaly Detection

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Perspective

Received: 12 February 2016, Revised: 21 April 2016, Accepted: 25 April 2016, Published online in Wiley Online Library (wileyonlinelibrary.com) DOI: 10.1002/cem.2806

CHEMOMETRICS

Networkmetrics: multivariate big data analysis in the context of the internet

José Camacho^{a*}, Roberto Magán-Carrión^a, Pedro García-Teodoro^a
 and James J. Treinen^b

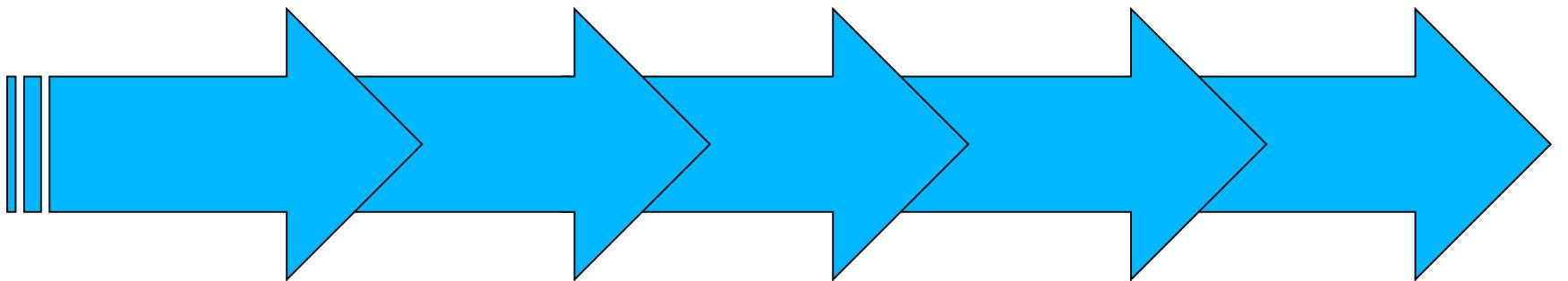
Hierarchical PCA-Based Multivariate Statistical Network Monitoring for Anomaly Detection

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In all areas of knowledge. In chemistry and related disciplines, the chemometric effort to understand and solve problems mainly from a multivariate and exploratory point of view, of broader applicability, even in areas of knowledge far from chemistry. In fact, the net of devices that allow an interconnected world where all types of data communication services can be provided. Problems in the Internet or in general in from chemometric problems. Building on this parallelism, we review four classes of problems, on, anomaly detection, optimization, and classification. We present an illustrative multivariate perspective may lead to significant improvements from state-of-the-art methods. In other words, we call the approach of treating these problems from that multivariate perspective. Networkmetric problems have their own specificities, mainly, their typical Big Data structured data. We argue that multivariate analysis is, indeed, useful to tackle these problems. We work; networkmetrics; Big Data

2016 IEEE International Workshop on Information Forensics and Security (WIFS)

- ✓ MSNM: 5 steps from the hay to the needle



(1) Parsing

(2) Fusion

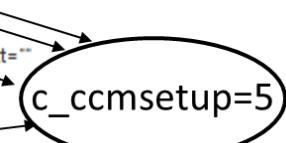
(3) Detection

(4) Diagnosis

(5) De-parsing

(1) Parsing: Feature-as-a-counter

```
<![LOG[      SCCM.CONTOSO.COM]LOG!> <time="21:36:59.151+000" date="03-30-2010" component="ccmsetup" context="" type="1" thread="4304" file="ccmsetup.cpp:4542">
<![LOG[Updated security on object C:\Windows\ccmsetup\]LOG!> <time="21:36:59.167+000" date="03-30-2010" component="ccmsetup" context="" type="0" thread="4304" file="ccmsetup.cpp:8849">
<![LOG[Sending Fallback Status Point message, STATEID='100']LOG!> <time="21:36:59.183+000" date="03-30-2010" component="ccmsetup" context="" type="1" thread="4304" file="ccmsetup.cpp:9326">
<![LOG[State message with TopicType 800 and TopicId (9EBF02F2-54F8-4E7E-8CC1-6982AC49CD98) has been sent to the FSP]LOG!>
<![LOG[Running as user "SYSTEM"]LOG!> <time="21:36:59.370+000" date="03-30-2010" component="ccmsetup" context="" type="1" thread="2928" file="ccmsetup.cpp:2690">
<![LOG[Detected 16747 MB free disk space on system drive.]LOG!> <time="21:36:59.370+000" date="03-30-2010" component="ccmsetup" context="" type="1" thread="2928" file="ccmsetup.cpp:463">
```

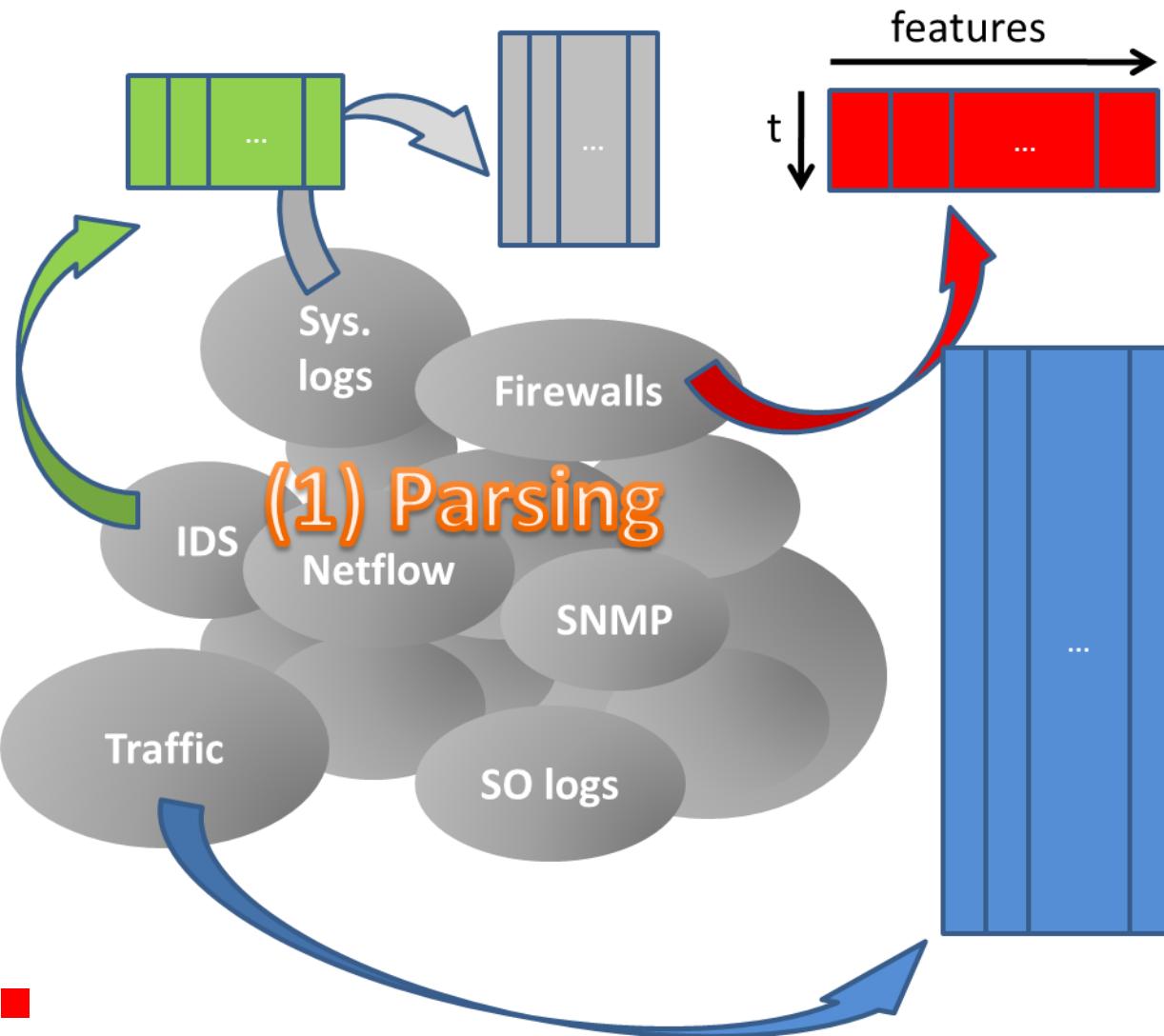


Time	FSPStateMessage	ccmstup	thread_4304
T=20s	1	5	4
T=40s	2	3	3
T=60s	1	3	3
T=80s	1	1	4

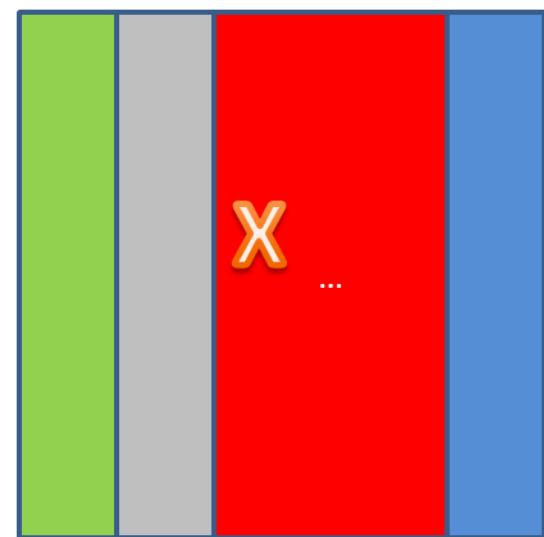
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Tackling the Big Data 4 Vs for Anomaly Detection

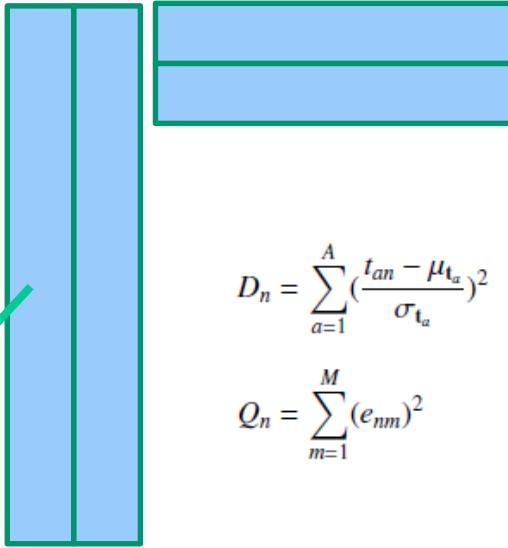
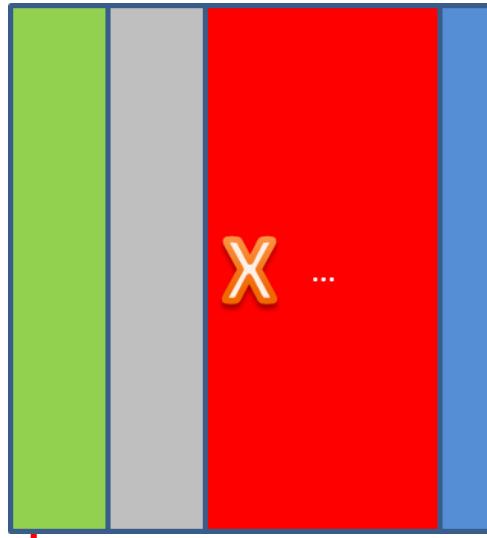
José Camacho, Gabriel Maciá-Fernández, Jesús Díaz-Verdejo and Pedro García-Teodoro
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(2) Fusion



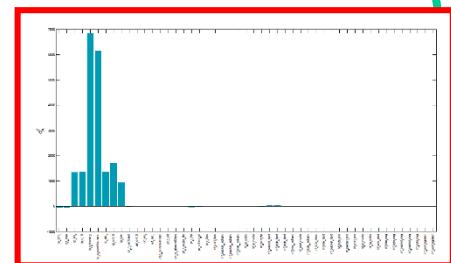
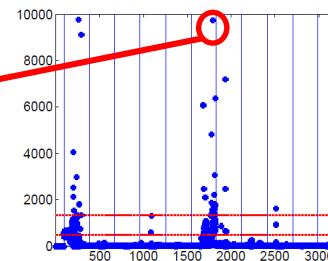
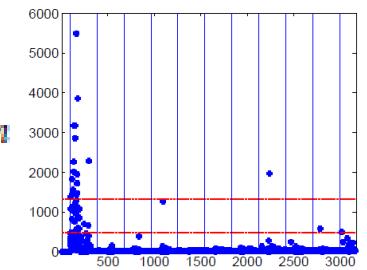
Combine any sources: from low level sensors (e.g. netflow) to high level info
(e.g. correlation rules at SIEM)



$$D_n = \sum_{a=1}^A \left(\frac{t_{an} - \mu_{t_a}}{\sigma_{t_a}} \right)^2$$

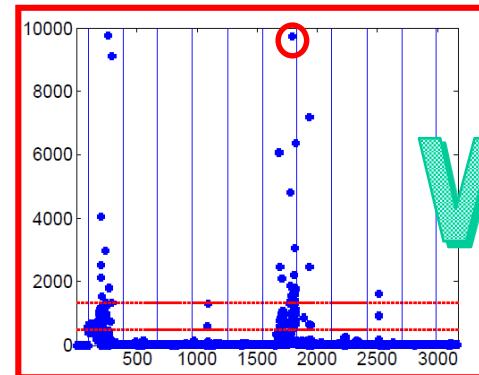
$$Q_n = \sum_{m=1}^M (e_{nm})^2$$

(3) Detection

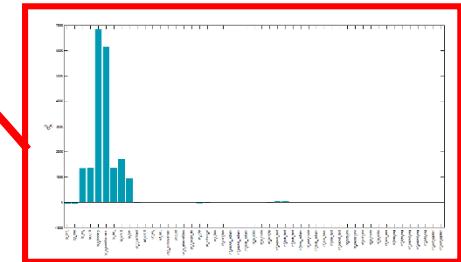
 D_n  Q_n 

(4) Diagnosis

MSNM creates a normality model for detection (which answers when an anomaly takes place) and diagnosis (in which source/s of info it is detected)



When?



Where?

```
IDS-03292012-1hr.txt | [**] [1:2103003:7] GPL NETBIOS SMB-DS Session Setup NTMLSSP unicode asnl overflow attempt [**]  
[Classification: Generic Protocol Command Decode] [Priority: 3]  
03/29-14:48:31.019982 172.23.0.216:1251 -> 172.23.0.10:445  
TCP TTL:128 TOS:0x0 ID:1696 Iplen:20 DgmLen:1500 DF  
***A*** Seq: 0xA9B345B0 Ack: 0x4522D27D Win: 0xFF3A TcpLen: 20  
[Xref => http://www.microsoft.com/technet/security/bulletin/MS04-007.mspx][Xref =>  
http://cgi.nessus.org/plugins/dump.php3?id=12065][Xref =>  
http://cgi.nessus.org/plugins/dump.php3?id=12052][Xref =>  
http://cve.mitre.org/cgi-bin/cvename.cgi?name=2003-0818][Xref =>  
http://www.securityfocus.com/bid/9635][Xref => http://www.securityfocus.com/bid/9633]  
  
[**] [1:2102466:9] GPL NETBIOS SMB-DS IPC$ unicode share access [**]  
[Classification: Generic Protocol Command Decode] [Priority: 3]  
03/29-14:48:31.024896 172.23.0.216:1251 -> 172.23.0.10:445  
TCP TTL:128 TOS:0x0 ID:1698 Iplen:20 DgmLen:138 DF  
***AP*** Seq: 0xA9B35020 Ack: 0x4522D402 Win: 0xFDB5 TcpLen: 20  
  
[**] [1:2103003:7] GPL NETBIOS SMB-DS Session Setup NTMLSSP unicode asnl overflow attempt [**]  
[Classification: Generic Protocol Command Decode] [Priority: 3]  
03/29-14:48:32.421373 172.23.0.211:1308 -> 172.23.0.10:445  
TCP TTL:128 TOS:0x0 ID:1843 Iplen:20 DgmLen:1500 DF  
***A*** Seq: 0xA1B4DB42 Ack: 0x5D2556D8 Win: 0xFF3A TcpLen: 20  
[Xref => http://www.microsoft.com/technet/security/bulletin/MS04-007.mspx][Xref =>  
http://cgi.nessus.org/plugins/dump.php3?id=12065][Xref =>  
http://cgi.nessus.org/plugins/dump.php3?id=12052][Xref =>  
http://cve.mitre.org/cgi-bin/cvename.cgi?name=2003-0818][Xref =>  
http://www.securityfocus.com/bid/9635][Xref => http://www.securityfocus.com/bid/9633]
```

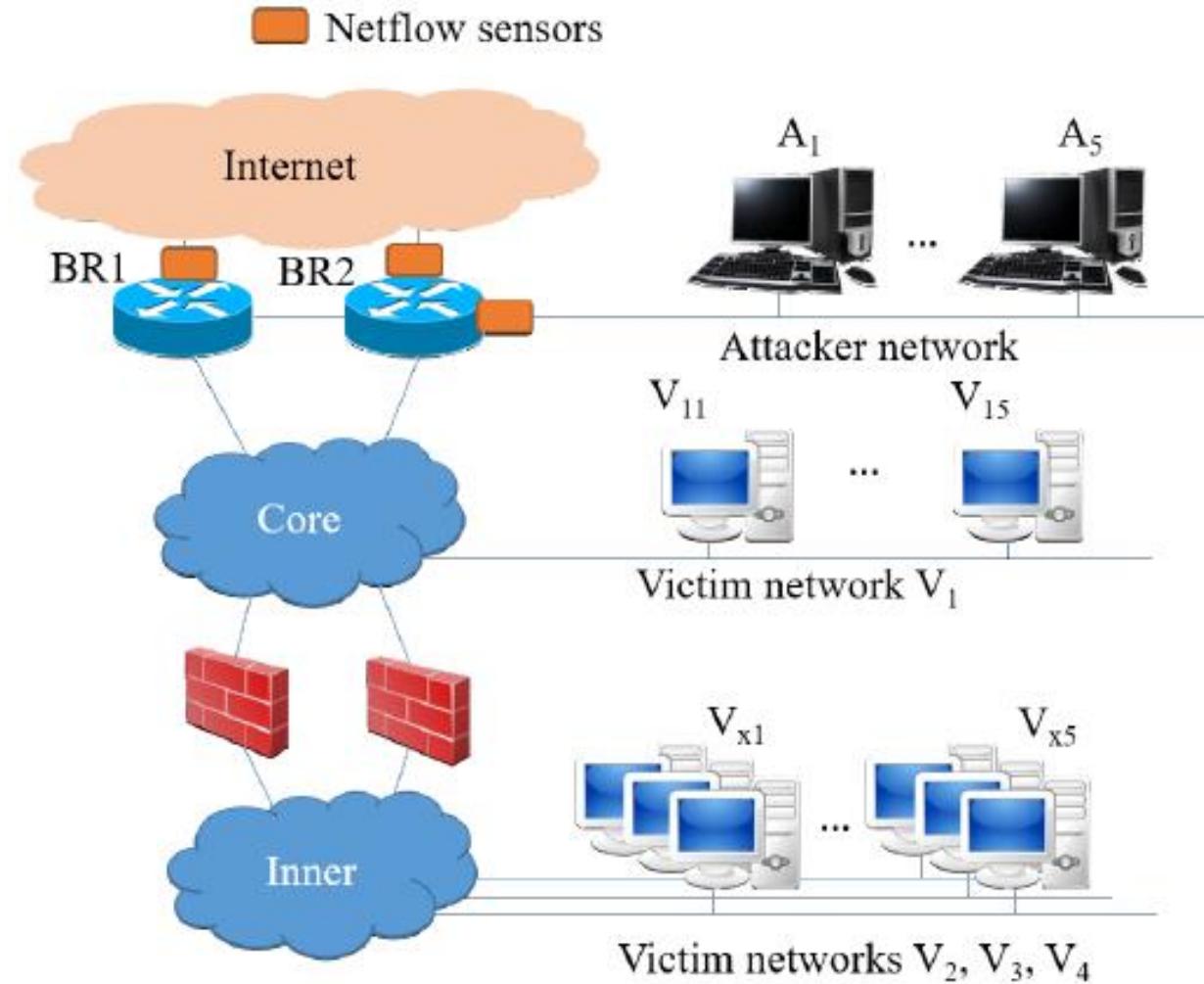
(5) De-parsing

Logs with detailed info of the anomaly are manually identified from the information provided by MSNM

Case Study - I

Network Engineering and Security Group
University of Granada
José Camacho, Ph.D.





- ✓ Synthetic attacks: DoS, scan, exfiltration (3 types)

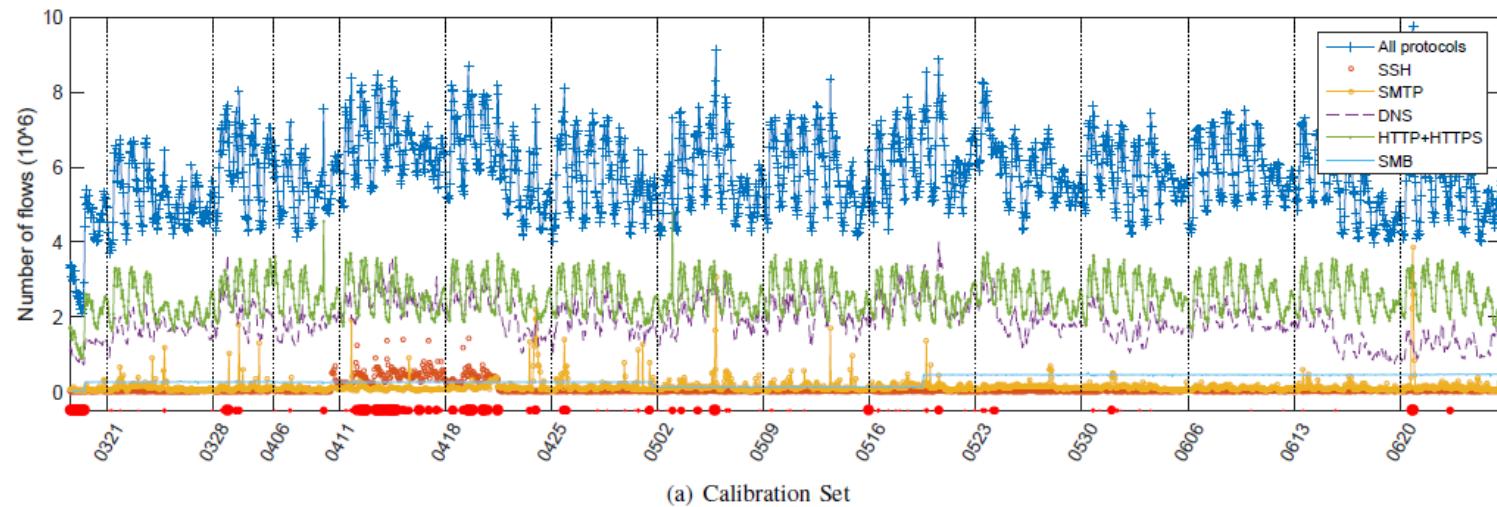
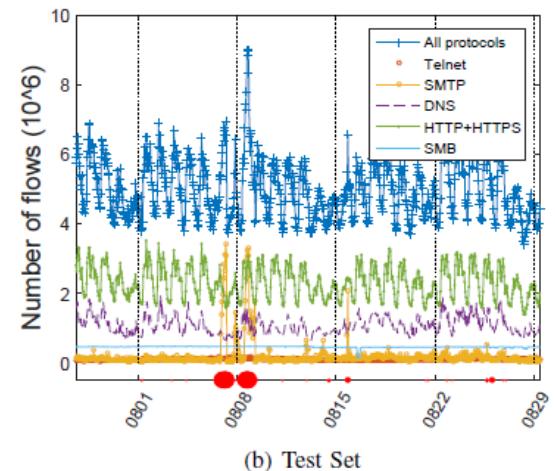


TABLE IV
FEATURES OF THE CALIBRATION AND THE TEST SETS.

Feature	Calibration	Test
Capture start	10:52h 03/18/2016	13:43h 07/27/2016
Capture end	18:27h 06/26/2016	09:27h 08/29/2016
Attacks start	N/A	00:00h 07/28/2016
Attacks end	N/A	12:00h 08/09/2016
Number of files	17	6
Size (compressed)	181GB	55GB
# Connections	\approx 13,000M	\approx 3,900M



✓ Features (138, no Fusion)

Variable	#features → values
Source IP	2 → <i>public, private</i>
Destination IP	2 → <i>public, private</i>
Source port	50 → <i>specific services, Other</i>
Destination port	50 → <i>specific services, Other</i>
Protocol	5 → <i>TCP, UDP, ICMP, IGMP, Other</i>
Flags	6 → <i>A, S, F, R, P, U</i>
ToS	3 → <i>0, 192, Other</i>
# Packets in	5 → <i>very low, low, medium, high, very high</i>
# Packets out	5 → <i>very low, low, medium, high, very high</i>
# Bytes in	5 → <i>very low, low, medium, high, very high</i>
# Bytes out	5 → <i>very low, low, medium, high, very high</i>

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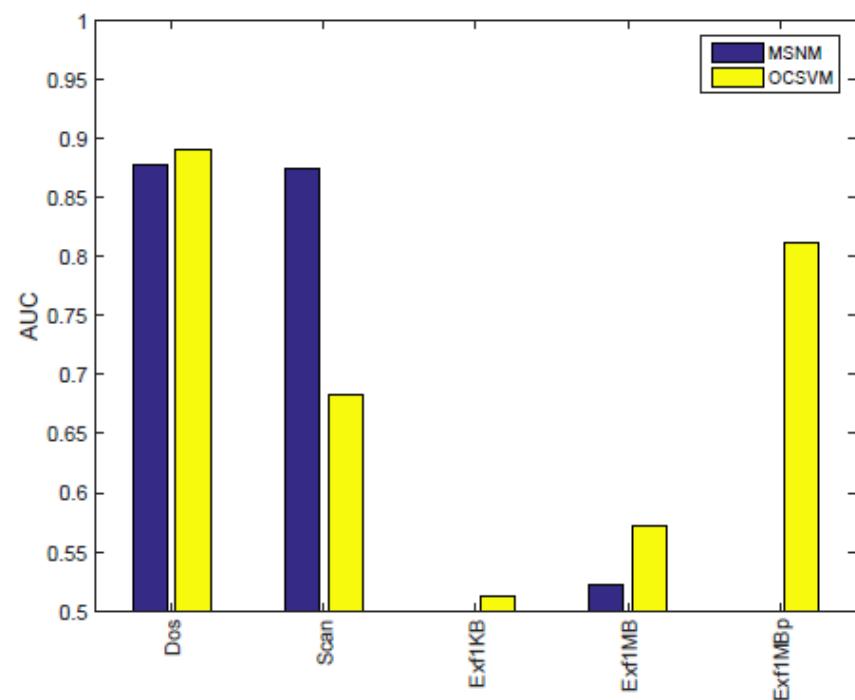
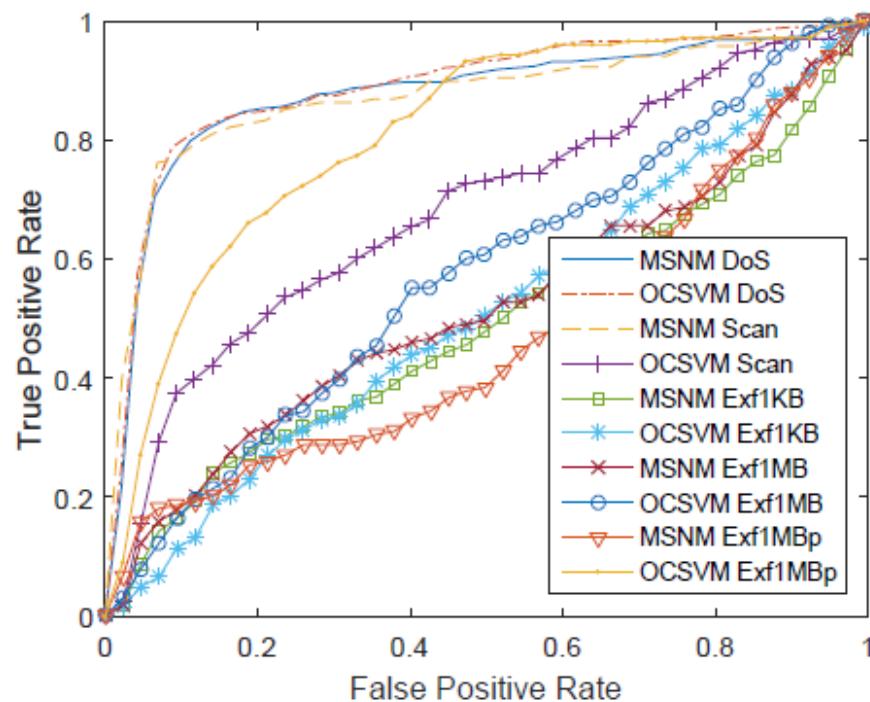
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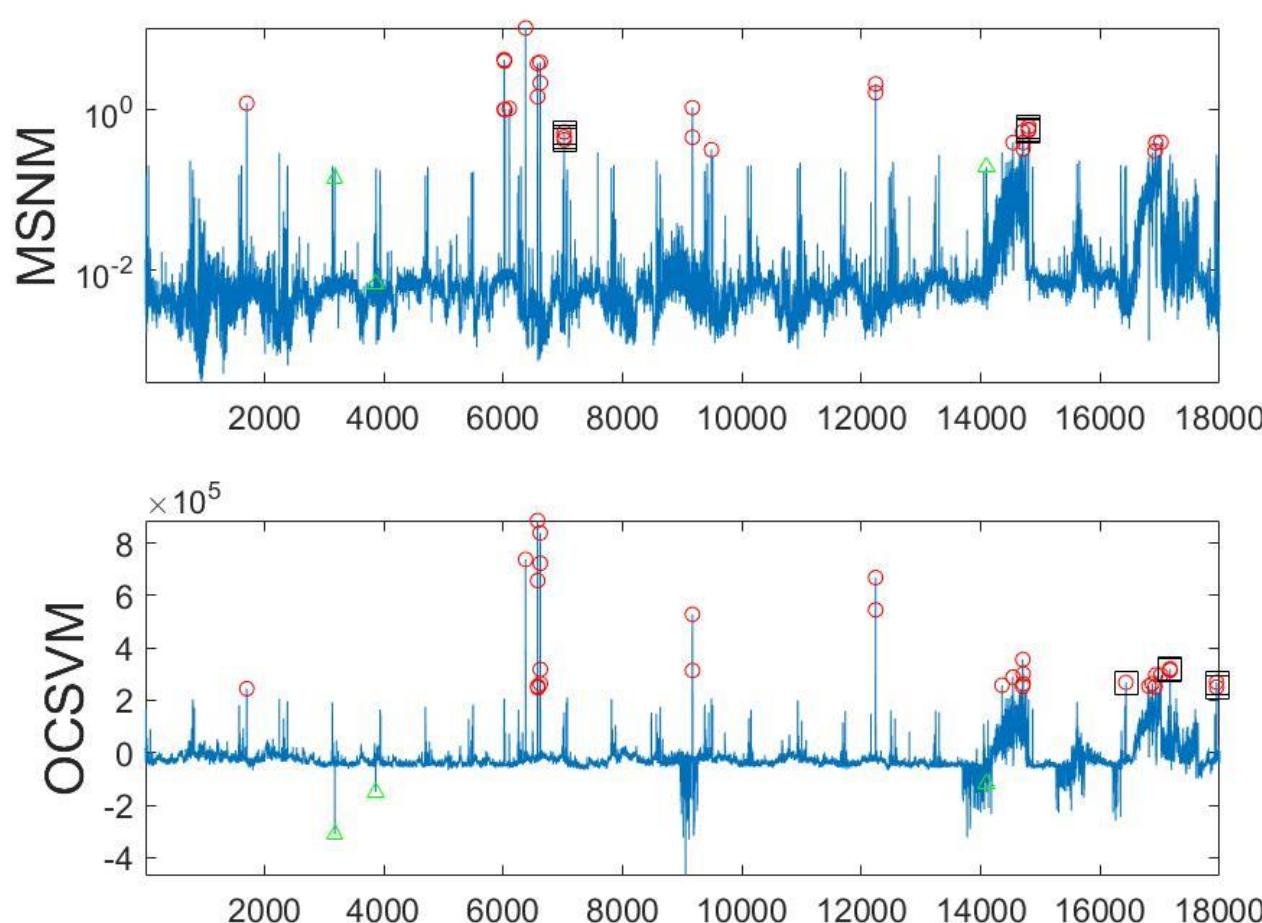
✓ One-class SVM

- ✓ B. Scholkopf, A. J. Smola, R. C. Williamson, and P. L. Bartlett, “New Support Vector Algorithms,” Neural computation, vol. 12, no. 5, pp.1207–1245, 2000.
- ✓ B. Scholkopf, J. C. Platt, J. Shawe-Taylor, A. J. Smola, and R. C. Williamson, “Estimating the Support of a High-Dimensional Distribution,” Neural Computation, vol. 13, no. 7, pp. 1443–1471, 2001.

✓ Synthetic attacks

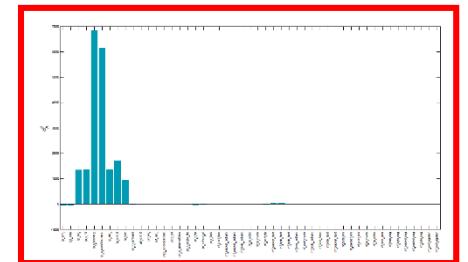


✓ Real attacks: Detection scores



- 20 Top detections
- Top & Synth.
- △ Least agreement

✓ Real attacks: Diagnosis



Time interval 20160806t2039-20160807t0559

'dport_smtp' 'in_nbytes_low' 'tcpflags_SYN' 'protocol_tcp'
'tcpflags_PSH' 'in_npackets_low' '--protocol_udp' '--dport_dns'
'tcpflags_RST' 'srctos_zero'



Many SMTP short
connections

Advantage over ML

Diagnosis is useful, but too
complicate!

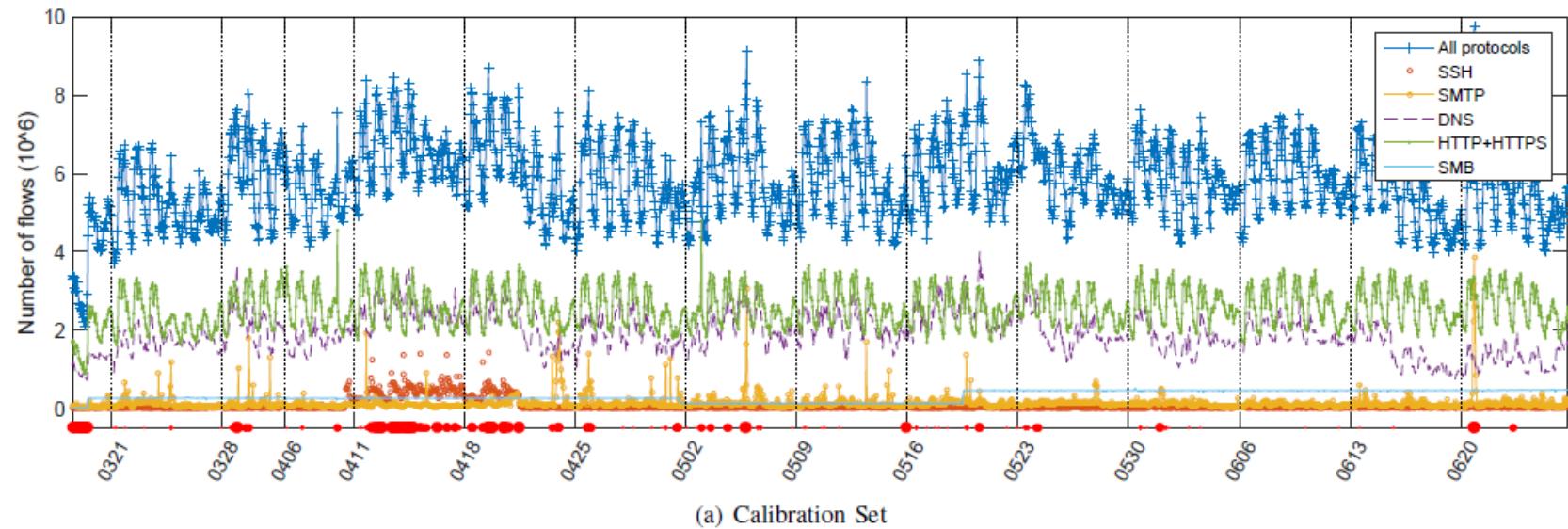
✓ Real attacks: De-parsing example

```
Time interval 20160806t2039-20160807t0559
'dport_smtp' 'in_nbytes_low' 'tcpflags_SYN' 'protocol_tcp'
'tcpflags_PSH' 'in_npackets_low' '--protocol_udp' '--dport_dns'
'tcpflags_RST' 'srctos_zero'
```

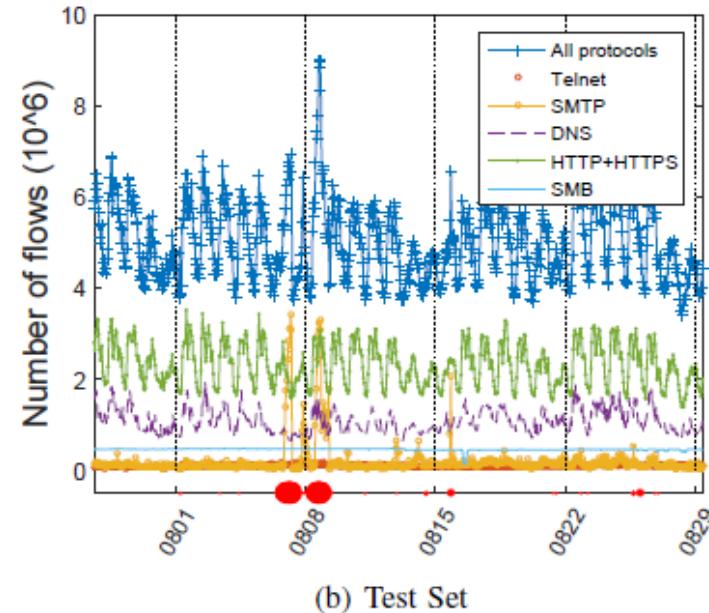
```
MSNM rocks :-) $ nfdump -R $inputs -t 2016/08/06:20:39:00-2016/08/07:06:00:00
'dst port = 25 or (bytes > 150 and bytes < 1001) ... ' -w $output
```

SPAM campaing				
Date first seen	Duration	Proto	Dst Port	Flows(%)
2016-08-06 20:39:01.448	33657.436	GRE	0	16537(0.1)
2016-08-06 20:40:21.164	33579.712	TCP	25	11.5 M(38.2)
2016-08-06 20:39:02.756	33658.236	TCP	443	1.4 M(4.6)
2016-08-06 20:39:08.644	33652.352	TCP	80	1.5 M(4.9)
2016-08-06 21:00:20.876	27042.808	TCP	20	109(0.0)

Normal				
Date first seen	Duration	Proto	Dst Port	Flows(%)
2016-08-20 11:07:01.452	895.896	GRE	0	378(0.1)
2016-08-20 11:07:27.128	869.912	TCP	80	32569(9.5)
2016-08-20 11:07:27.116	870.680	TCP	443	26252(7.6)
2016-08-20 11:11:27.008	583.996	TCP	26242	4(0.0)
2016-08-20 11:07:01.000	888.752	TCP	22	898(0.3)



(a) Calibration Set



(b) Test Set

- ✓ MSNM ~ OCSVM in detection.
- ✓ Good detection performance when including real attacks.
- ✓ MSNM has diagnosis support → reduces the time from detection to response.
- ✓ Unlike other methods, MSNM takes advantage of a large number of features → improves Diagnosis
- ✓ Future work:
 - ✓ Other sources (Host), MSNM-SIEM, Privacy & Scalability, ...

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This work is partly supported by the Spanish Ministry of
Economy and Competitiveness and FEDER funds through
project TIN2014-60346-R

