

Classifying Internet One-way Traffic

Eduard Glatz,
Xenofontas Dimitropoulos

ETH Zurich

May 15, 2012

Overview

- ▶ Classification scheme for dissecting one-way traffic that relies solely on flow-level data
- ▶ Observation on one-way traffic based on a massive dataset of 457 billion flows
- ▶ Show how one-way flows are useful for service availability monitoring

Preliminaries

- ▶ Study **incoming** one-way traffic at the network level: connections that do not receive a reply.
- ▶ Example causes of one-way traffic:
 - ▶ Failures & Policies
 - ▶ Attacks
 - ▶ Special application behavior

Preliminaries

- ▶ Study **incoming** one-way traffic at the network level: connections that do not receive a reply.
- ▶ Example causes of one-way traffic:
 - ▶ Failures & Policies
 - ▶ Attacks
 - ▶ Special application behavior
- ▶ Sampling and asymmetric routing can result in artificial one-way traffic
- ▶ One-way traffic can be measured in edge networks

Classification Scheme

- ▶ Associate each one-way flow with a number of **signs**
- ▶ Introduce 18 signs exploiting in 4 cases techniques from the literature
- ▶ Classify flows based on their signs
- ▶ Classes:
 - ▶ Unreachable services
 - ▶ P2P applications
 - ▶ Scanning
 - ▶ Backscatter
 - ▶ Suspected Benign
 - ▶ Bogon

Signs: Host pair behavior

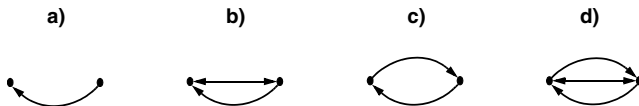


Figure: Mixture of incoming one- and two-way flows exchanged between a host pair. Hosts are represented by nodes and the presence of inflow/outflow/biflows by arrows.

Signs: Host pair behavior

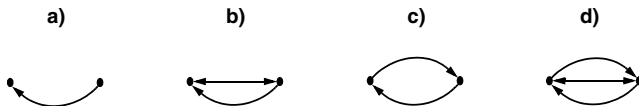


Figure: Mixture of incoming one- and two-way flows exchanged between a host pair. Hosts are represented by nodes and the presence of inflow/outflow/biflows by arrows.

- ▶ End-hosts-communicating: One-way flow between productive host pair
- ▶ Limited dialog: One-way flows between unproductive host pair

Signs: Local host behavior

- ▶ Unused local address: Unpopulated local IP address
- ▶ Service unreachable: Unanswered request to local service
- ▶ Peer-to-peer¹: Flow towards local P2P host

¹W. John and S. Tafvelin. Heuristics to classify internet backbone traffic based on connection patterns. International Conference on Information Networking (ICOIN), 2008

Signs: Remote host behavior

- ▶ Service sole reply: no biflow on $\text{srcIP} \wedge \text{dstPort} \geq 1024 \wedge \text{srcPort} < 1024$
- ▶ Remote scanner ¹²: TRW algorithm (suspected scanner)
- ▶ Remote scanner ²³: Host classification (suspected scanner)
- ▶ Remote non-scanner: TRW algorithm (suspected regular host)

²J. Jung, V. Paxson, A. Berger, and H. Balakrishnan. Fast portscan detection using sequential hypothesis testing. In Proceedings of the IEEE Symposium on Security and Privacy, 2004

³M. Allman, V. Paxson, and J. Terrell. A brief history of scanning. In Proceedings of the 7th ACM SIGCOMM IMC, 2007

Signs: Flow feature

- ▶ Artifact: UDP/TCP flow with both port numbers=0
- ▶ Single packet: Flow contains one packet only
- ▶ Large flow: Flow carries ≥ 10 packets or ≥ 10240 bytes
- ▶ Bogon: Source IP belongs to bogon space
- ▶ Protocol: IP protocol type of flow

Classification Rules

Final classifier includes 17 classification rules

Class Name	Rule #	Flow Membership Rules
Malicious Scanning	1	$\{ \overline{TRWscan}, \overline{HCscan}, \overline{PotOk} \} \Rightarrow Scanner$
	2	$\{ \overline{HCscan}, \overline{TRWscan}, \overline{TRWnom}, \overline{PotOk} \} \Rightarrow Scanner$
	3	$\{ \overline{TRWscan}, \overline{HCscan}, \overline{PotOk} \} \Rightarrow Scanner$
	4	$\{ \overline{TRWnom}, \overline{HCscan} \} \Rightarrow Scanner$
	5	$\{ \overline{GreyIP}, \overline{Onepkt}, \overline{TRWscan}, \overline{HCscan}, \overline{Backsc}, \overline{ICMP}, \overline{UDP}, \overline{bogon} \} \Rightarrow Scanner$
	6	$\{ \overline{GreyIP}, \overline{TRWscan}, \overline{HCscan}, \overline{Onepkt}, \overline{ICMP}, \overline{Backsc}, \overline{bogon} \} \Rightarrow Scanner$
	7	$\{ \overline{Onepkt}, \overline{GreyIP}, \overline{ICMP}, \overline{TRWscan}, \overline{HCscan}, \overline{TRWnom}, \overline{bogon}, \overline{P2P}, \overline{Unreach}, \overline{PotOk}, \overline{Backsc}, \overline{Large} \} \Rightarrow Scanner$
	8	$\{ \overline{GreyIP}, \overline{Onepkt}, \overline{TRWscan}, \overline{HCscan}, \overline{Backsc}, \overline{ICMP}, \overline{TCP}, \overline{bogon} \} \Rightarrow Scanner$
	9	$\{ \overline{ICMP}, \overline{TRWscan}, \overline{TRWnom}, \overline{HCscan}, \overline{InOut}, \overline{bogon}, \overline{PotOk} \} \Rightarrow Scanner$
Backscatter	10	$\{ \overline{Backsc}, \overline{TRWscan}, \overline{HCscan}, \overline{P2P}, \overline{InOut}, \overline{PotOk} \} \Rightarrow Backscatter$
Service Unreachable	11	$\{ \overline{Unreach}, \overline{TRWscan}, \overline{HCscan}, \overline{bogon}, \overline{P2P} \} \Rightarrow Unreachable$
Benign P2P Scanning	12	$\{ \overline{P2P}, \overline{TRWscan}, \overline{HCscan}, \overline{bogon} \} \Rightarrow P2P$
Suspected Benign	13	$\{ \overline{PotOk}, \overline{Unreach}, \overline{P2P}, \overline{TRWnom}, \overline{bogon} \} \Rightarrow Benign$
	14	$\{ \overline{Large}, \overline{GreyIP}, \overline{TRWscan}, \overline{HCscan}, \overline{P2P}, \overline{Unreach}, \overline{PotOk}, \overline{ICMP}, \overline{Backsc}, \overline{bogon}, \overline{TRWnom} \} \Rightarrow Benign$
	15	$\{ \overline{TRWnom}, \overline{GreyIP}, \overline{HCscan}, \overline{P2P}, \overline{Unreach}, \overline{bogon}, \overline{Backsc} \} \Rightarrow Benign$
	16	$\{ \overline{ICMP}, \overline{InOut}, \overline{TRWscan}, \overline{HCscan}, \overline{TRWnom}, \overline{bogon}, \overline{PotOk} \} \Rightarrow Benign$
Bogon	17	$\{ \overline{bogon}, \overline{TRWscan}, \overline{HCscan}, \overline{Backsc} \} \Rightarrow Bogon$

Data-Sets

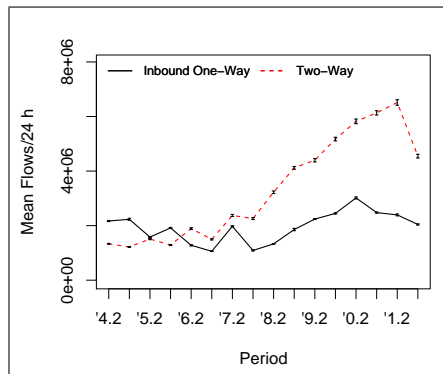
- ▶ Use data from the Swiss academic backbone network (SWITCH)
- ▶ Analyze the first 400 hours of each Feb and Aug between 2004 and 2011
- ▶ The studied traffic data correspond to:
 - ▶ 457 billion flows
 - ▶ 7.41 petabytes
 - ▶ cover 9% of the total number of flows

Data Sanitization

- ▶ Double-counting elimination reduces total traffic volume by 32.3%
- ▶ Defragmentation reduces the number of flows by a fraction ranging between 20.6% and 39.6% for different years
- ▶ Bi-flow Pairing:
 - ▶ For TCP and UDP based on standard 5-tuple
 - ▶ For other protocols based on 3-tuple

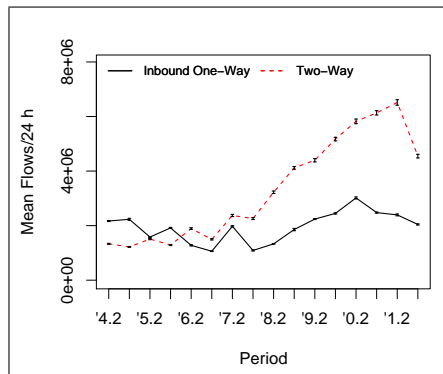
Evolution of One- and Two-way Traffic

- ▶ One-way flows are a large fraction of all flows:
 - ▶ In 2004, 2 out of every 3 flows were one-way
 - ▶ From 2007 to 2010, 1 out of every 3 flows were one-way



Evolution of One- and Two-way Traffic

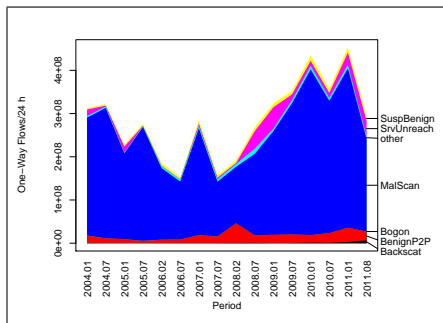
- ▶ One-way flows are a large fraction of all flows:
 - ▶ In 2004, 2 out of every 3 flows were one-way
 - ▶ From 2007 to 2010, 1 out of every 3 flows were one-way
- ▶ The number of one-way flows in 2011 is almost equal to 2004
- ▶ The fraction of one-way flows has declined



Composition of One-way Traffic

Class	% of flows	% of pkts	pkts/flow
Scanning	83.5%	62.6%	1.6
P2P applications	6.7%	13.0%	6.8
Unreach services	4.8%	10.1%	4.1
Suspected Benign	2.6%	9.1%	12.1
Other	2.2%	4.7%	4.6
Backscatter	0.3%	0.5%	3.3

- ▶ The top sources of one-way traffic are scanning, P2P protocols, and unreachable services



Service Availability Monitoring

- ▶ One-way flows are very useful for service availability monitoring
- ▶ Traditional service availability monitoring is based on active probing

Service Availability Monitoring

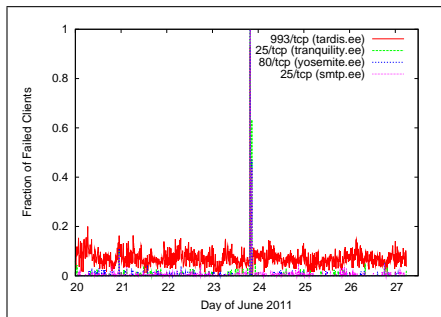
- ▶ One-way flows are very useful for service availability monitoring
- ▶ Traditional service availability monitoring is based on active probing
- ▶ Advantages of flow-based approach:
 - ▶ Provides a tangible assessment of the impact of disruptions
 - ▶ Discovers running services without requiring manual configuration
 - ▶ Exploits passive measurements

Outages and Misconfigurations in ETH Zurich

- ▶ Examine a week of NetFlow data from the EE Department of ETH Zurich
- ▶ Found 32 main services (> 99% availability) and 11 transient services

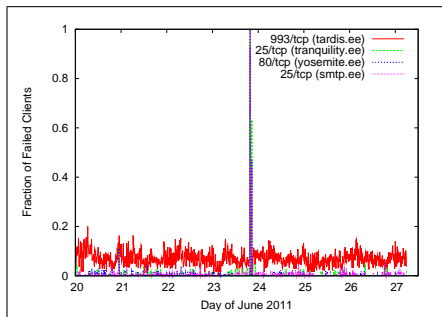
Outages and Misconfigurations in ETH Zurich

- ▶ Examine a week of NetFlow data from the EE Department of ETH Zurich
- ▶ Found 32 main services (> 99% availability) and 11 transient services
- ▶ Identified a coinciding global outage



Outages and Misconfigurations in ETH Zurich

- ▶ Examine a week of NetFlow data from the EE Department of ETH Zurich
- ▶ Found 32 main services (> 99% availability) and 11 transient services
- ▶ Identified a coinciding global outage
- ▶ During the identified interval **287,583 unique IP addresses failed** to access target services!



Conclusions

- ▶ Classification scheme for one-way traffic that relies on 18 signs derived from flow data
- ▶ Observations based on a very large data-set:
 - ▶ One-way flows are a large fraction of all flows
 - ▶ In terms of flows, the share of one-way traffic has declined since 2004
 - ▶ The top sources of one-way traffic are scanning, P2P protocols, and unreachable services
- ▶ One-way traffic is very useful for assessing the impact of failures

Questions?

Contact: fontas@gmail.com

E. Glatz and X. Dimitropoulos. Classifying Internet One-way Traffic. TIK-Report 336, ETH Zurich, May 2012

Validation

- ▶ Collect packet traces from a small campus network
- ▶ Exploit additional information:
 - ▶ Extended host profiles
 - ▶ ICMP types and codes
 - ▶ TCP flags (Check protocol state machine)
 - ▶ DPI-based application identification⁴
 - ▶ Precise timestamps

⁴H. Kim, K. Claffy, M. Fomenkov, D. Barman, M. Faloutsos, and K. Lee. Internet traffic classification demystified: myths, caveats, and the best practices. ACM CoNEXT, 2008

Validation

- ▶ Collect packet traces from a small campus network
- ▶ Exploit additional information:
 - ▶ Extended host profiles
 - ▶ ICMP types and codes
 - ▶ TCP flags (Check protocol state machine)
 - ▶ DPI-based application identification⁴
 - ▶ Precise timestamps

Class Name	Recall [%]	Precision [%]
Malicious Scanning	99.9	99.8
Service Unreachable	99.6	96.1
Benign P2P Scanning	95.3	95.5
Backscatter	62.4	88.4
Suspected Benign	85.1	75.0
Bogon	40.4	100.0

⁴H. Kim, K. Claffy, M. Fomenkov, D. Barman, M. Faloutsos, and K. Lee. Internet traffic classification demystified: myths, caveats, and the best practices. ACM CoNEXT, 2008

Outages and Misconfigurations in ETH Zurich

- ▶ Found server that was not reachable during the studied week in total by 2.2 million unique clients!
- ▶ What was this server? Hint: Switzerland is famous for chocolate, banking, swiss army knives, and watches

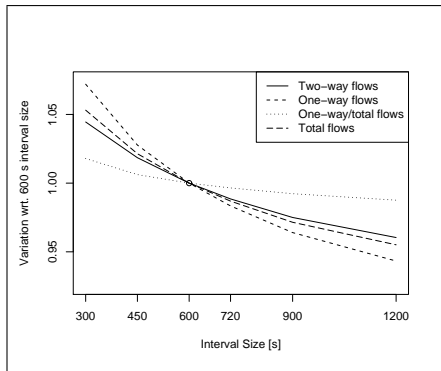
Outages and Misconfigurations in ETH Zurich

- ▶ Found server that was not reachable during the studied week in total by 2.2 million unique clients!
- ▶ What was this server? Hint: Switzerland is famous for chocolate, banking, swiss army knives, and watches
- ▶ Popular NTP server `swisstime.ee.ethz.ch` preconfigured in NTP clients and used in NTP “hello world” examples
- ▶ It was not reachable to 12.9% of its clients cause by invalid CRC checksums and a filtering policy

Impact of the Interval Size

Doubling the interval size:

- ▶ decreases absolute count metrics by 3-5%.
- ▶ decreases relative volume metrics by 1.2% and does not
- ▶ decrease further with an increasing interval size.



Signs

Sign Type	Sign Name	Detection Criterion/Algorithm
Host pair behavior	End-hosts-communicating Limited dialog	One-way flow between productive host pair One-way flows between unproductive host pair
Remote host behavior	Service sole reply Remote scanner 1 Remote scanner 2 Remote non-scanner	no biflow on $\text{srcIP} \wedge \text{dstPort} \geq 1024 \wedge \text{srcPort} < 1024$ TRW algorithm (suspected scanner) Host classification (suspected scanner) TRW algorithm (suspected regular host)
Local host behavior	Unused local address Service unreachable Peer-to-peer	Unpopulated local IP address Unanswered request to local service Flow towards local P2P host
Flow feature	Artifact Single packet Large flow Bogon Protocol	UDP/TCP flow with both port numbers=0 Flow contains one packet only Flow carries ≥ 10 packets or ≥ 10240 bytes Source IP belongs to bogon space IP protocol type of flow