

## Towards an Efficient Detection of Pivoting Activity

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Part I

# Introduction

# Motivation & Goals

## Motivation

- Pivoting is a powerful technique, when used by adversaries
  - If a network is accessible only through a VPN server, the adversary may exploit the VPN server to get access inside
- Pivoting detection capabilities are limited:
  - Many phenomena may look like pivoting and cause false positive detections

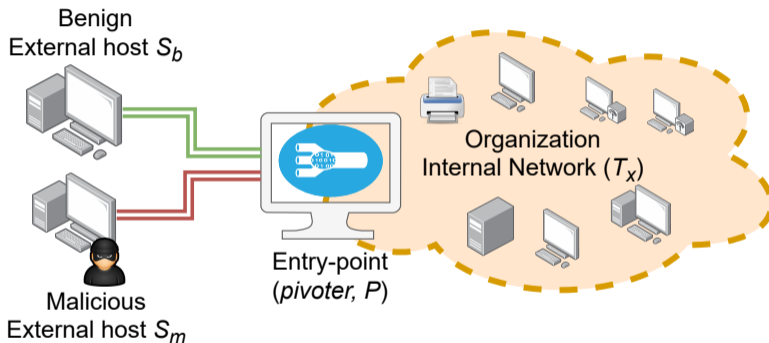
## Research questions

1. Which phenomena are similar to pivoting?
2. Which NetFlow features are intrinsic to pivoting?
3. How to automatically reduce false positives?

Part II

## Pivoting

# Pivoting Scenario



From now on,  $S$  stands for Source,  $P$  for Pivot, and  $T$  for Target.

## Existing Pivoting Detection Methods

- G. Apruzzese, F. Pierazzi, M. Colajanni, and M. Marchetti, “Detection and threat prioritization of pivoting attacks in large networks,” *IEEE Transactions on Emerging Topics in Computing*, vol. 8, no. 2, pp. 404–415, 2020.
- Temporal graph-analytics to detect pivoting by analyzing NetFlow data
  - $S$  and  $T$  must go through  $P$
  - $P$  forwards the communications to  $T$  after receiving the communications from  $S$
  - Pairs of NetFlows linking  $S$  with  $P$ , and  $P$  with  $T$ , are separated by at most  $\epsilon$  time units
- Evaluated only in internal network with no background noise
- Would generate large number of false positives even with very small values of  $\epsilon$

## Pivoting Detection Methodology

Splitting the detection in two phases:

- Pivoting candidate detection using the existing algorithm
- Pivoting selection over detection candidates

Pivoting candidate detection - consider network flows  $f_1$  and  $f_2$ :

- destination IP of  $f_1$  is source IP of  $f_2$
- destination IP of  $f_2$  is not source IP of  $f_1$
- protocol and destination ports are the same in  $f_1$  and  $f_2$
- $\epsilon = 1s$ :  $f_2$  appears after  $f_1$  but no later than 1 second after  $f_1$

Pivoting selection

- Manual and automated, as shown later

## Discovery of Phenomena Similar to Pivoting

Pivoting is expected if:

- the protocol that allow remote access, such as SSH, RDP, and Telnet
- $S$  is anywhere in the Internet or  $T$  is in the internal network
- The set of  $S$ ,  $P$ , and  $T$  is rather unique
- The duration of flows and numbers of transferred bytes/packets/flows are similar

False positive detection is expected if:

- The protocol is DNS, SMTP, or other benign protocol (in the context of pivoting)
- All actors are from the same network (e.g., internal) or are well-known and the activity may be explained otherwise
- There are many candidates with the same  $P$  and  $T$  but different  $S$



## Automatically Identifying Intrinsic Features

The following features will be observed:

- $S, P, T$  Count – number of candidates that share the same  $S, P$ , or  $T$
- $S, P, T$  Location – location of the given IP address, i.e., 'Private' for ranges like 10.0.0.0/8, 'Internal' for the IP range of the organization, 'External' for IP addresses in the Internet
- Duration, In/Out Packets, In/Out Bytes, Flows ratios – the values of the first biflow (source to pivot) divided by the second biflow (from pivot to target)

Principal Component Analysis will be used to select the most meaningful features and their combinations

## Part III

# Experiment Results

# Experiment Setup

## Measurements

- Masaryk University campus network (public /16 IPv4 and /48 IPv6, private networks)
- NetFlow probe near server segment, including a VPN server and Wi-Fi access points
- Data from a full working day

## Content

- Internet <-> Public IPs, Public IPs <-> Private IPs
- Only a few firewall rules, most of the traffic is enabled

## Pivoting Candidate Detection

Candidates	Dst Port	Service	Comment
69,393	TCP 443	HTTPS	Mostly benign, further inspection required
6,102	TCP 53	DNS	Benign, legitimate servers
5,362	TCP 25	SMTP	Benign, legitimate servers
3,456	TCP 51413	BitTorrent	Benign, personal computers
2,251	UDP 6881	BitTorrent	Benign, personal computers
133	TCP 22	SSH	Suspicious candidates

- No candidates found on ports 3389 (RDP) and 23 (Telnet)
- Numerous other candidates were found in BitTorrent, VoIP, and online gaming traffic

# Manual Investigation of Pivoting Candidates

## SSH Candidates

- 7 distinct pivots, 3 benign and 4 suspicious
- Benign pivots initiated connections only to one other host, e.g.,
  - The machine accepted many connections throughout the day from many hosts
  - The single outgoing connection was a git checkout/commit or data back-up
  - No similarity between the flows -> random false positive
- Suspicious pivots – 13 candidates
  - Either Internet->Public->Internet or Public->Internet->Public

## HTTP(S) candidates

- Mostly common and benign behavior of the protocol
- Left for future work, deeper analysis of HTTP headers or TLS is recommended

## Investigation of Pivoting Candidates

	HTTP(S)	70,312
<b>False positives</b>	other services - DNS, SMTP, NTP	12,714
	p2p - BitTorrent, VoIP, online gaming	7,615
	SSH	120
<b>True pivoting</b>	SSH	<b>13</b>
<b>Total</b>		<b>90,774</b>

- Huge imbalance, very low number of suspicious candidates complicates the analysis
- Strong motivation for the reduction of false positives and future work

# Principal Component Analysis

## Setup

- Manually selected SSH candidates labeled as *Suspicious* or *Benign*
- Kaiser criterion – selecting only principal components with eigenvalue  $> 1$
- Requirement on at least 90% of the variance explained for future analysis

## Findings

- Strong association with 'Internal'  $S$  and  $T$  and 'External'  $P$  or vice versa
- Strong association with 'count' features, namely  $P$  and  $T$  count
- Rare associations with 'ratio' features

## Part IV

# Conclusion



# Conclusion

## Conclusion

- Accurate detection of pivoting activities is of paramount importance
- Existing approaches work only under specific circumstances
- The rate of false positives is overwhelming due to similar phenomena (e.g., p2p)
- Location of the actors is the most significant trait of pivoting candidates

## Future Work

- Proposal of an efficient detector based on PCA-refined methods
- Further analysis of malicious and benign candidates

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