

Using loops observed in traceroute to infer the ability to spoof

Lone, Qasim; Luckie, Matthew; Korczyński, Maciej; Van Eeten, Michel

DOI

[10.1007/978-3-319-54328-4_17](https://doi.org/10.1007/978-3-319-54328-4_17)

Publication date

2017

Document Version

Accepted author manuscript

Published in

Passive and Active Measurement - 18th International Conference, PAM 2017, Proceedings

Citation (APA)

Lone, Q., Luckie, M., Korczyński, M., & Van Eeten, M. (2017). Using loops observed in traceroute to infer the ability to spoof. In S. Uhlig, J. Amann, & M. A. Kaafar (Eds.), *Passive and Active Measurement - 18th International Conference, PAM 2017, Proceedings* (pp. 229-241). (Lecture Notes in Computer Science (including subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics); Vol. 10176 LNCS). Springer. https://doi.org/10.1007/978-3-319-54328-4_17

Important note

To cite this publication, please use the final published version (if applicable).
Please check the document version above.

Copyright

Other than for strictly personal use, it is not permitted to download, forward or distribute the text or part of it, without the consent of the author(s) and/or copyright holder(s), unless the work is under an open content license such as Creative Commons.

Takedown policy

Please contact us and provide details if you believe this document breaches copyrights.
We will remove access to the work immediately and investigate your claim.

Using Loops Observed in Traceroute to Infer the Ability to Spoof

Qasim Lone¹, Matthew Luckie², Maciej Korczyński¹, and Michel van Eeten¹

¹ Delft University of Technology, the Netherlands

Q.B.Lone, Maciej.Korczynski, M.J.G.vanEeten@tudelft.nl

² University of Waikato, New Zealand

mjl@wand.net.nz

Abstract. Despite source IP address spoofing being a known vulnerability for at least 25 years, and despite many efforts to shed light on the problem, spoofing remains a popular attack method for redirection, amplification, and anonymity. To defeat these attacks requires operators to ensure their networks filter packets with spoofed source IP addresses, known as source address validation (SAV), best deployed at the edge of the network where traffic originates. In this paper, we present a new method using routing loops found in traceroute data to infer inadequate SAV at the transit provider edge, where a provider does not filter traffic that could not have come from the customer. Our method does not require a vantage point within the customer network. We present and validate an algorithm that identifies at Internet scale which loops identify a lack of ingress filtering by providers. The results show that 703 provider ASes do not implement ingress filtering on at least one of their out-degree links for 1,780 customer ASes. Most of these observations are unique compared to the existing methods of the Spoofer and Open Resolver projects. By increasing the visibility of the networks that allow spoofing, we aim to strengthen the incentives for the adoption of SAV.

1 Introduction

Despite source IP address spoofing being a known vulnerability for at least 25 years [6], and despite many efforts to shed light on the problem (e.g. [7,8,9]), spoofing remains a viable attack method for redirection, amplification, and anonymity, as evidenced in February 2014 during a 400 Gbps DDoS attack against Cloudflare [20]. That particular attack used an amplification vector in some implementations of NTP [20]; a previous attack against Spamhaus [10] in March 2013 achieved 300+ Gbps using an amplification vector in DNS. While some application-layer patches can mitigate these attacks [21], attackers continuously search for new vectors.

Defeating amplification attacks, and other threats based on IP spoofing, requires providers to filter incoming packets with spoofed source IP addresses [11] – in other words, to implement BCP 38, a Best Current Practice also known as source address validation (SAV). SAV suffers from misaligned incentives: a network that adopts SAV incurs the cost of deployment, while the security benefits

diffuse to all other networks. That being said, SAV is a widely supported norm in the community. Increasing the visibility of what networks have or have not adopted SAV reduces the incentive problem by leveraging reputation effects and the pressure of other providers and stakeholders. All of this puts a premium on our ability to measure SAV adoption.

In this paper, we report on the efficacy of a new measurement technique that is based on an idea of Jared Mauch. It allows an external observer to use traceroute to infer the absence of filtering by a provider AS at a provider-customer interconnect. This paper makes the following five contributions: (1) We show that it is feasible for providers to deploy static ingress ACLs, as their customers rarely change address space. (2) We describe a scalable algorithm for accurately inferring absence of ingress filtering from specific patterns in traceroute. (3) We validate the algorithm’s correctness using ground truth from 7 network operators. (4) We demonstrate the utility of the algorithm by analyzing Internet-scale inferences we made. (5) We build a public website showing the provider-customer edges that we inferred imply absence of filtering, combined with actionable data that operators can use to deploy filtering.

2 Background on Ingress Filtering

The canonical documents describing the use of ingress filtering methods for SAV are RFCs 2827 [11] and 3704 [5], known in the network operations and research communities as BCPs 38 and 84. BCP 38 describes the basic idea: the source address of packets should be checked at the periphery of the Internet against a set of permitted addresses. For an access network, this check could be at the point of interconnection with a single customer; for an enterprise, this could be on their edge routers to their neighbors; and for a transit provider, this could be on the provider-edge router where a customer connects. For single-homed customers, a transit provider can discard packets that have a source address outside the set of prefixes the customer announces to the transit provider, using Strict or Feasible Reverse Path Forwarding (RPF). A router using Strict RPF will drop a packet if it arrived on a different interface than the router would choose when forwarding a packet to the packet’s source address; a router using Feasible RPF will consider all paths it could use to reach the source address, not just the best path.

BCP 84 discusses challenges in deploying ingress filtering on multi-homed networks. Both Strict and Feasible RPF are not always feasible if a customer is multi-homed and does not announce all of its prefixes to each neighbor router, as it might do for traffic engineering purposes. Instead, an operator might define a set of prefixes covering source addresses in packets the router will forward, known as an Ingress Access List, or Ingress ACL. BCP 84 states that while ingress ACLs require manual maintenance if a neighbor acquires additional address space, they are “the most bulletproof solution when done properly”, and the “best fit ... when the configuration is not too dynamic, .. if the number of used prefixes is low”.

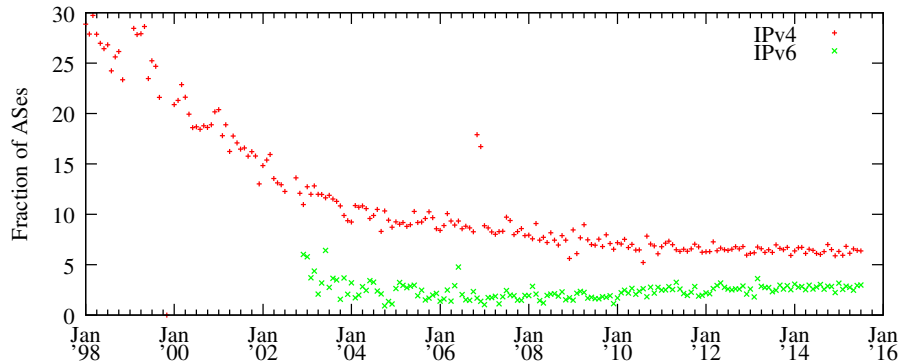


Fig. 1: Fraction of ASes whose prefix announcements changed month-to-month.

3 Related Work

Testing a network’s SAV compliance requires a measurement vantage point inside (or adjacent to) the network, because the origin network of arbitrary spoofed packets cannot be determined [5]. The approach of the Spoofer project [7] is to allow volunteers to test their network’s SAV compliance with a custom client-server system, where the client sends spoofed packets in coordination with the server, and the server infers if the client can spoof if the server receives these spoofed packets. However, the Spoofer project requires volunteer support to run to obtain a view from a given network. In May 2015, CAIDA released an updated Spoofer client that operates in the background, automatically testing attached networks once per week, and whenever the system is attached to a network it hasn’t tested in the previous week. The number of prefixes tested per month has since increased from ≈ 400 to more than ≈ 1600 in August 2016.

Jared Mauch provided the first technique to infer if a network had inadequate SAV without requiring a custom client-server system. As a product of the Open Resolver Project [2], he observed DNS resolvers embedded in home routers forwarding DNS queries from his system with IP_X to other resolvers, without rewriting the source IP address of the packet. These other resolvers returned the subsequent answer direct to IP_X , rather than to the DNS resolver in the home router as it should have.

We emphasize that these methods are complementary, and that no one technique is able to test deployment of SAV for all networks.

4 Motivation of Ingress ACLs

As described in §2, the best place to deploy filtering is at the edge. However, not all edge networks have the technical ability or motivation to filter their own traffic. A transit provider, however, is often managed by skilled network operators who may already deploy defenses to prevent their customers from announcing inappropriate routes. The provider-customer interconnect for an edge network represents the other straightforward place to deploy ingress filtering.

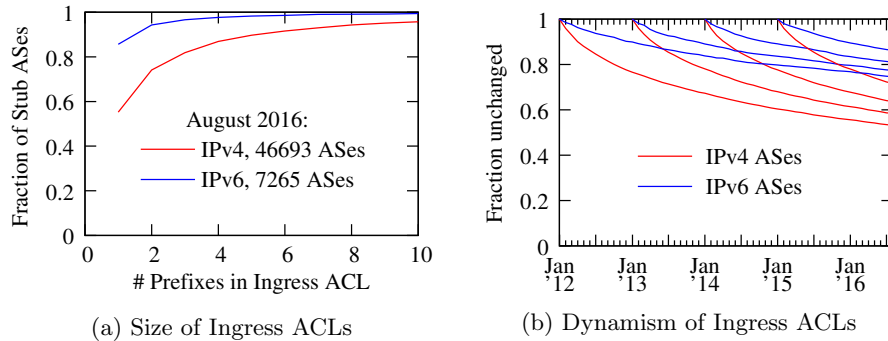


Fig. 2: Size and dynamism of ACLs required by providers to filter traffic from stub ASes.

Figure 1 examines the dynamism of address space announced by stub ASes over time. For each stub AS, we aggregated the prefixes each AS originated in BGP into the minimum prefix set, and examined month-to-month changes in the set. Perhaps a consequence of IPv4 address exhaustion, we see a trend toward stable announcement patterns. This may diminish the need for automated validation of source addresses and improve the practicality of static ingress ACLs: in May 2000, $\approx 15\%$ of stub ASes would have required deployment of a different IPv4 ingress ACL month-to-month, but in 2015, less than 5% of ASes would require the same.

As BCP 84 states that because ingress ACLs require manual maintenance they are best suited “when the configuration is not too dynamic” and “if the number of used prefixes is low”, figure 2 examines the size and dynamism of ingress ACLs required for stub ASes in August 2016. Figure 2a shows that 88.9% of stub ASes would require an IPv4 ACL of no more than 4 prefixes, and 85.6% of stub ASes would require an IPv6 ACL of a single prefix. Figure 2b shows the dynamism of ACLs over time, based on ACLs that could have been defined for all stub ASes in January 2012, 2013, 2014, and 2015. For stub ASes for these times, at least 77.4% of IPv4 ACLs would have not have to be changed over the course of one year; for those defined in January 2012, 54.4% of the inferred ACLs would not have needed to be changed even up to August 2016. Further, required IPv6 ACLs would be even less dynamic: more than 74.6% of IPv6 ACLs would not have to be changed over the course of 4.5 years until August 2016. We used public BGP data collected by Routeviews [19] and RIPE RIS [18] to indicate prefix changes in the respective ASes in figures 1 and 2

We believe the observed number of prefixes and dynamism over time imply that ingress ACLs are feasible in the modern Internet.

5 Inferring Absence of Ingress Filtering using Traceroute

The key idea of our approach is that traceroute can show absence of ingress filtering by providers of stub ASes when a traceroute path reaches the stub AS and then exits out of the stub, as the traceroute packets contain a source address

belonging to the vantage point (VP) launching the traceroute. If the provider’s router is performing SAV, it should filter the traceroute packet when it arrives from the stub AS, as it has a source address that the stub AS is not the origin for. If the provider’s router does not perform SAV, it will forward the packet, and the traceroute will show an apparent IP-level forwarding loop as subsequent packets are returned to the stub AS.

Xia *et al.* found that 50% of persistent loops were caused by missing “pull-up route” at a border router to an upstream provider [22]. However, a forwarding loop does not imply absence of SAV at the edge: a loop resulting from a transient misconfiguration or routing update can occur anywhere in the network. The key challenge in this work is inferring the provider-customer boundary in traceroute [16]. In this paper, we superimpose millions of traceroutes towards random IP addresses in /24 prefixes to build a topology graph, and use a small set of heuristics to infer provider-customer edges for stub ASes in the graph. §5.1 describes the Internet topology datasets that we used, and §5.3 describes the algorithm we used to filter the loops that imply the absence of ingress filtering by the provider – in other words, the lack of compliance with BCP 38.

5.1 Input Data

CAIDA IPv4 routed /24 topology datasets: We used CAIDA’s ongoing traceroute measurements towards every routed /24 prefix in the Internet. CAIDA’s probing of all routed /24s is especially useful here, as the goal is to find unrouted space that can result in a forwarding loop. CAIDA’s traceroute data is collected with scamper [15] using Paris traceroute which avoids spurious loops by keeping the ICMP checksum value the same for any given traceroute [4]. As of August 2016, CAIDA probes every routed /24 using 138 Vantage Points (VPs) organised into three teams; each team probes the address space independently. Each team takes roughly 1.5 days to probe every routed /24.

CAIDA IPv4 AS relationships: We used CAIDA’s ongoing BGP-based AS relationship inferences [17] to identify customer-provider interconnections in traceroute paths. The relationship files were inferred using public BGP data collected by Routeviews and RIPE RIS, using RIB files recorded on the 1-5 of each month. We also used the same BGP data to identify the origin AS announcing each prefix measured with traceroute.

CAIDA Sibling Inferences: We used CAIDA’s ongoing WHOIS-based AS-to-organization inference file [13] to identify ASes that belong to the same underlying organization (are siblings). The sibling files were inferred using textual analysis on WHOIS databases obtained from Regional Internet Registries (RIRs) at 3-month intervals. We used sibling inferences to avoid mis-classifying a loop that occurs within a single organisation using multiple ASes as one that occurs between distinct provider and customer ASes.

5.2 Construction of Topology

Our first goal is to correctly identify the provider-customer boundary towards stub ASes with high precision. Because the customer usually uses address space

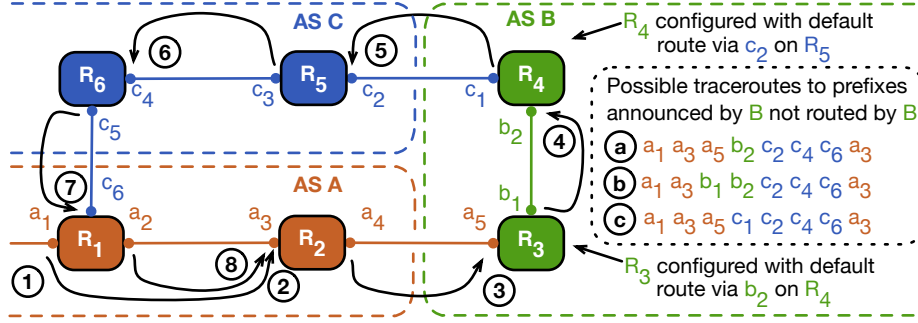


Fig. 4: A two-provider loop between ASes A and C and their customer B implying absence of filtering by C at R_5 . R_5 should discard packet 5 because it arrives with a source address outside of B, rather than forward the packet to R_6 .

assumed customer router is only observed in traceroute paths towards the stub AS, (4) that there is at least one other address originated by the provider in the traceroute path towards the stub. Criteria #3 avoids incorrectly inferring a provider-operated router as a customer edge router when a loop occurs before the stub AS (e.g. $a_1 a_3 a_2 a_3$) as a_3 is observed in traceroute paths towards both B and C. Criteria #4 avoids incorrectly inferring which router in a traceroute path is the customer edge router when the customer edge router is multihomed and the traceroute path enters via a second provider AS D (e.g., $d_2 a_4 a_5 a_4$).

Two-provider loops: Figure 4 illustrates the second case, where R_3 and R_4 are customer-edge routers belonging to AS B, with default routes configured on R_3 and R_4 . The underlying routing configuration issues are the same as a point-to-point loop, except the default route is via a different AS than the AS the traceroute entered the network. Figure 4 shows the traceroute visiting two routers operated by AS B; however, it is possible that the traceroute will never contain an IP address mapped to B, depending on how many routers in B the traceroute visits, and how the routers respond to traceroute probes. Therefore, our criteria are: (1) that the assumed customer router where the traceroute exits is only observed in paths towards the stub AS, (2) that both the ingress and egress AS in the traceroute path are inferred providers of the stub AS and not a sibling of the stub AS, (3) that there is no unresponsive traceroute hop in the traceroute path that could be where a customer router could be located, (4) that at least two consecutive IP addresses mapped to the same egress AS are observed in the loop. Criteria #2 does not require different provider ASes: if the stub AS is multi-homed to the same provider with different routers, our method will still infer an absence of filtering. Criteria #3 ensures that we do not mis-infer where the customer router is located in the path, and thus incorrectly infer the AS that has not deployed ingress filtering. Finally, criteria #4 reduces the chance that a loop inside the customer network is mis-classified as crossing into a provider network if the customer router responds with a third-party IP address.

5.4 Finding Needles in a Haystack

As discussed in §5.1, CAIDA’s Ark project uses three teams of VPs to probe a random address in every routed /24 prefix. In this section, we report on the characteristics of cycle 4947 conducted by team 3. The characteristics of data conducted by other teams and for other cycles is quantitatively very similar. In total, cycle 4947 contains 10,711,132 traceroutes, and 163,916 (1.5%) of these contain an IP-level loop. 105,685 (64.5%) of the loops were not towards a stub network.

Of the remaining 58,231 loops towards stub ASes, we inferred 31,023 (53.3%) occurred within the stub network, i.e. all IP addresses in the loop were announced in BGP by the stub, or involved the customer edge router. A further 11,352 loops (19.5%) contained an unresponsive IP address, and 1,373 (2.4%) an unrouted IP address that prevented us from inferring if the loop occurred at a provider/customer interconnect. 610 (1.0%) loops had a loop that was disqualified as occurring at a customer-provider boundary, as the loop occurred at a router observed in paths towards multiple destination ASes, and 494 (0.8%) were not recorded as the IP address observed could have been a third party address on a customer router, rather than a router operated by a provider. In total, only 2,530 (4.3%) of the loops were simple point-to-point loops, and only 93 (0.2%) of the loops were more complex two-provider loops.

5.5 Persistence of Loops

Given that we are looking for needles in haystacks, how reliably can we find them? Ideally, we would be able to consistently reproduce the relevant loops and weed out the observations that were caused by transient events. Unfortunately, there is currently no straightforward way of doing this.

To understand why, we have to take a closer look at CAIDA’s traceroute data. The data is collected via probes sent to a random IP address from each routable /24 prefix. This adds efficiency by reducing the number of probes, at the cost of potentially missing loops that occur for smaller prefixes. It also means that when such a loop is in fact discovered, the next probe might miss it again by selecting a random address outside the smaller prefix. In other words, the traceroute data itself does not tell us very much about the persistence of loops.

To better understand the impact of random address selection on the discovery of loops, we decided to reproduce as closely as possible the exact traceroute probes that identified the relevant loops. We first applied the algorithm outlined in §5.3 to the traceroute data for August 2016 and found 2,500 unique loops between 703 provider and 1,780 customer ASes. In October 2016, we reproduced the probes the exact IP addresses that generated the loops, using two different vantage points. We were able to reproduce 1,240 of the loops, between 461 provider and 1,026 customer ASes. Next, we repeated this procedure to over a year of traceroute data: August 2015–August 2016. We found 7,784 unique loops between 1,286 provider and 3,993 customer ASes. In October 2016, we were able to reproduce 1,542 unique loops between 505 provider and 1,176 customer ASes. In other words, the additional data identified 342 loops that could be reproduced.

A significant portion of all loops could not be reproduced and the longer the time lag, the higher the odds of failure. There are four main reasons for this. First, the loop might have been transient, i.e., it only occurred during routing protocol convergence [12] or temporary misconfiguration [22]. Second, it might depend on the vantage point of the probe, e.g., because of multi-homed routers. Third, the provider might have fixed the routing issue that caused the loop. Fourth, and most relevant, the provider has implemented ingress filtering.

Future work is needed to disentangle these causes. We know from our validation effort (§6) that even loops that are observed only once can correctly signal non-compliance. Some of the loops that we could not reproduce had already been validated by the provider as true positives. In the remainder of the paper, we will work with the full set of loops as identified by our algorithm.

6 Validation by Network Providers

In order to validate our results and obtain ground truth, we contacted providers in two rounds: September 2015 and September 2016. We got feedback from one hosting provider, one data center provider, one ISP, two national research and education networks, and two Tier 1 networks. Some providers were included only in one round, some in both, depending on whether we saw non-compliant links for their network at both times and on our ability to reach the right specialist in the organization. We gave all providers a formal assurance that their names would not be included in the paper.

The feedback from the providers during the first round resulted in improvements in our methodology. The final methodology was applied to both the August 2015 and August 2016 data. We then compared the final results to the feedback that we received from the providers in both rounds.

We define a result as a true positive if we identify a provider-to-customer link as non-compliant and the provider operating the AS confirms this. Note that this implies that we correctly inferred the absence of SAV as well as the boundary between provider and customer. A false positive occurs when we either incorrectly detected the boundary or the provider is actually performing SAV at boundary.

We talked to 6 providers in round 1 and 4 in round 2, where 3 providers participated in both rounds. Our methodology correctly identified the absence of ingress filtering on the provider boundary in 94 out of 98 unique links between provider and customer ASes (45 of 49 links in round 1 and 49 of 49 links in round 2).

The four false positives had different causes. Three of them occurred because of route aggregation. Providers perform route aggregation by consolidating multiple routes to its customers in a single, more general route. This practice can lead to problems with our boundary detection. Imagine this scenario: a provider is assigned a /16 prefix X by the Regional Internet Registry (RIR). It allocates a /24 subnet Y from its prefix to a customer. The customer also has its own prefix Z allocated by the RIR. Now, if the provider aggregates the /24 it allocated to the customer with other provider prefixes into a single /16 advertisement, this

Table 1: Results of validation from network providers

Provider	2015		2016	
	AS Size	Links (TP/FP)	AS Size	Links (TP/FP)
A	12800	1/0	-	-
B	8780032	0/1	-	-
C	8219392	0/1	8220672	3/0
D	90368	2/0	-	-
E	19245568	2/0	-	-
F	155795	7/0	260640	3/0
G	7830144	37/0	7778944	6/0
H	-	-	9476864	2/0
	-	-	11139328	32/0
I	-	-	1709312	0/0

would mean that prefix Y would show up as belonging to the provider AS. Our methodology would then categorize a loop between provider prefix X and customer prefix Z as signaling the absence of SAV. In truth, the loop was actually within the customer network, occurring between prefix Y and Z.

For the fourth false positive, the provider informed us that the traceroute data suggested that the loop had occurred inside their network rather than on the boundary. However, they could not reproduce it anymore and blamed it on a transient event. Note that in the second round, we found 3 loops for the same provider and they were all true positives.

One additional piece of feedback that we received was that some of the providers, while confirming the validity of our inference that they were not doing ingress filtering on their boundary, objected to the implication that they *should* be doing ingress filtering there. They saw their services as offering transit and contracted them as such, which meant no filtering on the provider’s side. In the view of these providers, the downstream customer AS should perform SAV at their last hop router. The customer ASes were business entities like ISPs, hosting providers or large enterprises.

Evaluating whether this interpretation of the RFCs is merited, falls outside the scope of this paper and is for the community to address. For this paper, the key point is that the proposed method performed accurately.

7 Results

We first summarize the results in terms of the number of networks that do not implement SAV. We then compare our method to the two alternatives: the Spoofer and the Open Resolver projects. Like those methods, our approach only observes a subset of the networks without SAV. In the absence of loops, we cannot tell anything about the presence of ingress filtering.

Using one month of traceroute data from August 2016, our approach identifies 2,500 unique loops between 703 provider ASes as lacking SAV on one or more of their out-degree links and 1,780 customer ASes. These represent approximately

1.3% and 3.2% of all advertised ASes [3], respectively. Moreover, when compared to all advertised stub ASes and their providers [17], we find 9.0% of provider ASes without ingress filtering linking with 3.8% of all customer ASes.

As discussed in §6, some of the providers argued that customer ASes should be responsible for SAV within their networks or at their borders. However, we find that about 63% of customer ASes of non-compliant providers advertise /20 or smaller prefix lengths. It is unlikely that such small entities have the resources and incentives to implement SAV in their networks. Furthermore, such small prefixes should allow the providers to implement static ACLs.

We now compare our results to the data from the Spoofer and Open Resolver projects (see §3 for details). Our method only detects the lack of ingress filtering for provider networks, which means that their customer ASes might be able to spoof. We compared those customer ASes with the Spoofer data from February to August 2016. Of 54 overlapping ASes, in 38 of them the **spoofer** probes were behind a Network Address Translation (NAT) that blocked spoofing. Of the systems not behind a NAT, 10 of the 16 stub ASes were found to allow spoofing, i.e., more than half of these ASes had not deployed SAV, suggesting the provider’s expectation for their customers to deploy filtering is not being met, and reinforcing the case for transit providers to filter their customers. This means that the connected provider ASes do not implement ingress filtering, which is consistent with our results. The probes in the 6 remaining customer ASes could not spoof, even though the provider AS did not do ingress filtering. This suggests that filtering took place in the customer AS. The overlap between both methods contains only a small sample, but it does indicate that the majority of the customer networks are not doing SAV – a finding that reinforces the point that providers should not expect their customer ASes to be willing and able do SAV, even if they are not that small.

The Open Resolver data was used 2014 by Kührer *et al.* to identify 2,692 unique ASes from within which spoofing was possible [14]. Following the same approach, we analyzed the August 2016 data from the Open Resolver project, generously provided to us by Jared Mauch, and found a total of 3,015 unique ASes that are able to spoof. We compare these to the customer ASes that our method identified as allowing spoofing – i.e., those connected to the providers which lack ingress filtering. We find only a modest overlap: 244 ASes.

In sum: these findings show that our method can add a lot of unique data points to both existing methods, and hence significantly improve visibility of networks lacking SAV. In terms of the volume of observations, it resides between Spoofer and Open Resolver. The three methods are very complementary and provide different viewpoints into the problem, contributing to improved overall visibility of SAV adoption.

8 Conclusion

In this paper we implemented and validated an algorithm which uses traceroute data to infer a lack of SAV between a stub and provider network. We were able to identify 703 providers which do not implement ingress filtering on at least one

of their out-degree links for 1,780 customer ASes. We also built a public website showing the provider-customer edges that we inferred as lacking ingress filtering: <https://spoofer.caida.org/cgi-bin/trspoofer/trspoofer.pl>. The data can be used by providers to deploy filtering, which would not only stop attackers from sending spoof packets from their network, but also block attempts to DDoS the connection between provider and customer by sending traceroutes from different locations to these loops [22].

To improve the reliability of the method, future work is needed on border detection and on untangling the different factors that prevent loops from being reproduced, so as to separate the implementation of ingress filtering from the other causes. A completely different direction for future work is to experimentally test the strength of reputation effects among providers and network operators. The networks that allow spoofing could be made public in varying ways, to see which mechanism best incentivizes providers into taking action.

For the community of network operators, the results support efforts such as the Routing Resilience Manifesto [1] and other community initiatives to improve network security. By complementing the Spoofer and Open Resolver data, our method greatly increases visibility into the adoption of SAV. Public visibility of spoofing-enabled networks is a critical step in incentivizing providers to deploy ingress filtering in their networks. It rewards those that adopt SAV and calls out those that do not. The dataset might also be useful for the national CERTs who want to push BCP 38 compliance in their countries. The problem of IP spoofing has been recognized for years, but the task to reduce its role in ongoing attacks is only becoming increasingly urgent.

References

1. Mutually Agreed Norms for Routing Security (MANRS), <https://www.routingmanifesto.org/manrs/>
2. Open Resolver Project, <http://openresolverproject.org/>
3. CIDR report. <http://www.cidr-report.org/as2.0/>, Retrieved: 2016-09-14
4. Augustin, B., Cuvellier, X., Orgogozo, B., Viger, F., Friedman, T., Latapy, M., Magnien, C., Teixeira, R.: Avoiding traceroute anomalies with Paris traceroute. In: IMC. pp. 153–158 (Oct 2006)
5. Baker, F., Savola, P.: Ingress filtering for multihomed networks. RFC 3704 (Mar 2004), IETF BCP84
6. Bellare, S.: Security problems in the TCP/IP protocol suite. CCR 19(2), 32–48 (1989)
7. Beverly, R., Bauer, S.: The spoofer project: Inferring the extent of source address filtering on the Internet. In: Proceedings of USENIX SRUTI (Jul 2005)
8. Beverly, R., Berger, A., Hyun, Y., K. Claffy: Understanding the efficacy of deployed Internet source address validation. In: IMC. pp. 356–369 (Nov 2009)
9. Beverly, R., Koga, R., K. Claffy: Initial longitudinal analysis of IP source spoofing capability on the Internet (Jul 2013), <http://www.internetsociety.org/>
10. Bright, P.: Spamhaus DDoS grows to Internet-threatening size (Mar 2013)
11. Ferguson, P., Senie, D.: Network ingress filtering: Defeating denial of service attacks which employ IP source address spoofing. RFC 2827 (May 2000), IETF BCP38

12. Francois, P., Bonaventure, O.: Avoiding transient loops during igp convergence in ip networks. In: INFOCOM. pp. 237–247 (Mar 2005)
13. Huffaker, B., Keys, K., Koga, R., kc claffy: CAIDA inferred AS to organization mapping dataset, <https://www.caida.org/data/as-organizations/>
14. Kühner, M., Hupperich, T., Rossow, C., Holz, T.: Exit from hell? reducing the impact of amplification ddos attacks. In: USENIX Security Symposium (2014)
15. Luckie, M.: Scamper: a scalable and extensible packet prober for active measurement of the Internet. In: IMC. pp. 239–245 (Nov 2010)
16. Luckie, M., Dhamdhere, A., Huffaker, B., Clark, D., k claffy: bdrmap: Inference of borders between IP networks. In: IMC (Nov 2016), <https://www.caida.org/~mjl/pubs/bdrmap.pdf>
17. Luckie, M., Huffaker, B., Dhamdhere, A., Giotsas, V., k claffy: AS relationships, customer cones, and validation. In: IMC. pp. 243–256 (Oct 2013)
18. NCC, R.: Routing information service (ris). DOI= <https://www.ripe.net/analyse/internet-measurements/routing-information-service-ris>
19. of Oregon, U.: University of oregon route views project. DOI= <http://www.route-views.org/>
20. Prince, M.: Technical details behind a 400Gbps NTP amplification DDoS attack (Feb 2014), <http://blog.cloudflare.com/>
21. Vixie, P.: Rate-limiting state: The edge of the Internet is an unruly place. ACM Queue 12(2), 1–5 (Feb 2014)
22. Xia, J., Gao, L., Fei, T.: A measurement study of persistent forwarding loops on the internet. Computer Networks 51(17), 4780–4796 (2007)