In-Network Probabilistic Monitoring Primitives under the Influence of Adversarial Network Inputs

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Fast control loop systems



Monitoring and Debugging

FlowRadar [NSDI'16] | LossRadar [CoNEXT'16] | Dapper [SOSR'17] | ConQuest [CoNEXT'19] | SpiderMon [SOSR'20] | Continuous In-network RTT Monitoring [SIGCOMM'22]

Load Balancing and Caching

HULA [SOSR'16] | SilkRoad [SIGCOMM'17] | NetCache [SOSP'17]

Routing

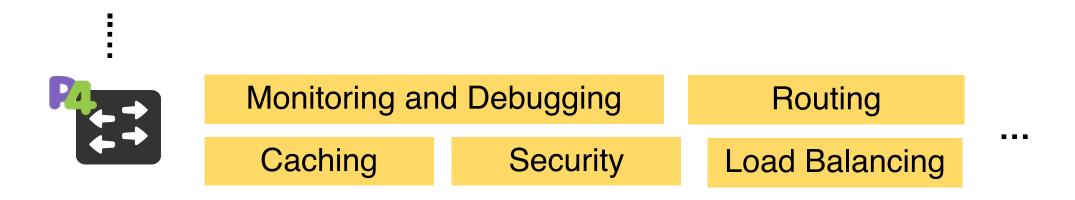
Blink [NSDI'19] | Contra [NSDI'20] | RouteScout [SOSR'21]

Security

Nethof [ICNP'19] | Poise [USENIX Security'20] | NetWarden [USENIX Security'20] | Jaqen [USENIX Security'21]

In-network monitoring primitives

Q: Is a flow new or old? FlowRadar [NSDI'16]
Q: How many times has a key been accessed? NetCache [SOSP'17]

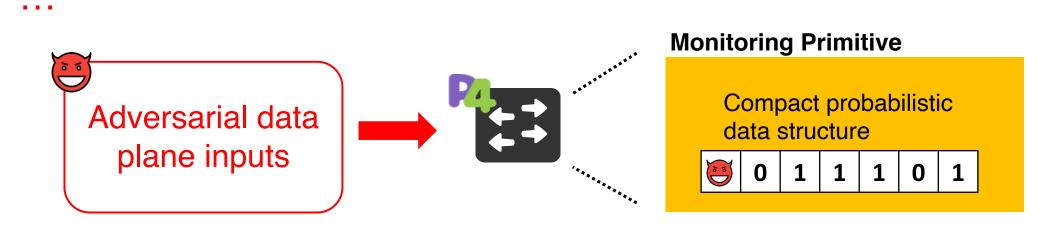


Compact probabilistic data structures!

Bloom filters (BFs), Count Minimum Sketches (CMS), Inverted Bloom Lookup Tables (IBLTs)....

Impact of attacks on probabilistic monitoring primitives

Corruption of network statistics, Denial of service, Performance degradation



Pollute the bloom filters and sketches through targeted attacks!

In this work....

Question

What are the negative impacts of polluting compact probabilistic data structures that drive in-network monitoring primitives?

Key contribution

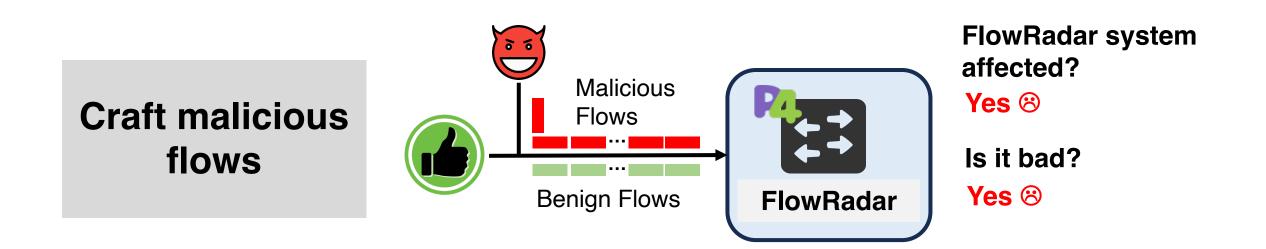
Empirically gauge the impact of bloom filter pollution attacks on *FlowRadar*¹

¹Li, Y., Miao, R., Kim, C., & Yu, M. (2016). Flowradar: A better netflow for data centers. In 13th {USENIX} Symposium on Networked Systems Design and Implementation ({NSDI} 16) (pp. 311-324).

Our approach

Define threat model

Adversarial privileges? Adversarial objectives?



How bad are the effects?

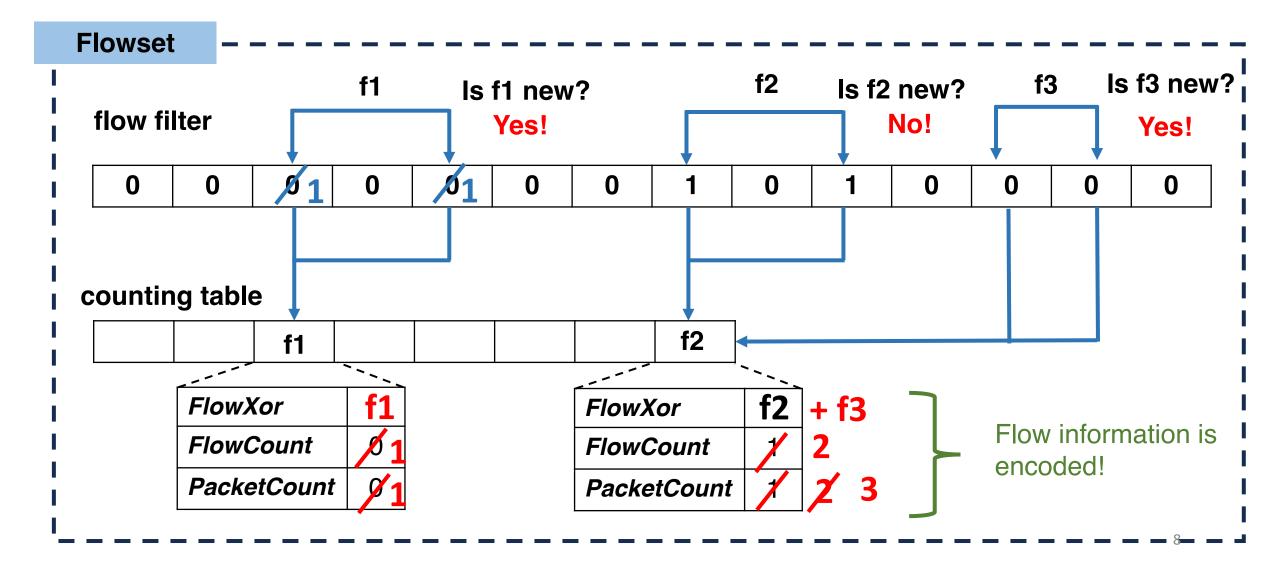
Threat model

5 5

Bloom filter size Number of hash functions Type of hash function **Chosen Insertion Adversary (CIA)** Query only Adversary (QOA) 10 **Malicious flow Malicious flow** 0 0 0 1 0 1 0 1 Ø 0 **Ø**1 0 Statistics behind the BF **False positive!** polluted affected! **Benign new flow**

Bloom filters in FlowRadar

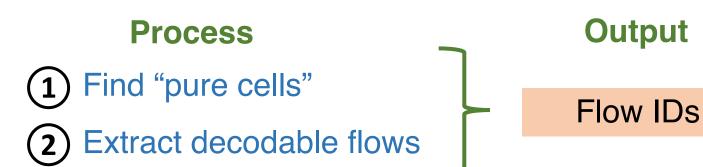
Network monitoring system that maintains flows and their counters



FlowRadar decode operations

Single Decode (SD) Extract decodable flows

Input Flowset -



Counter Decode (CD) Extract packet counts

InputProcessOutputFlow IDs1 Represent as linear equationsPacket counts!2 Solve using approximationsPacket counts!

Crafting Malicious flows

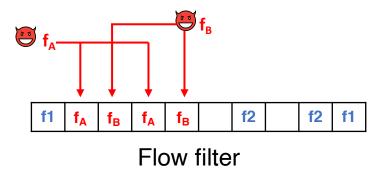
(1) Generate flow IDs (2) Interlace generated flows among benign flows

Chosen Insertion Adversar	ry (CIA)	Query or

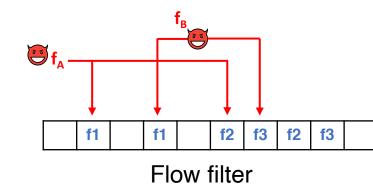
Query only Adversary (QOA)

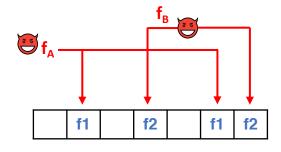
Subset

Flow IDs do not collide amongst themselves in the flowset's flowfilter



Flow IDs map to already set locations in the flowset's flowfilter Flow IDs are identical to benign flows





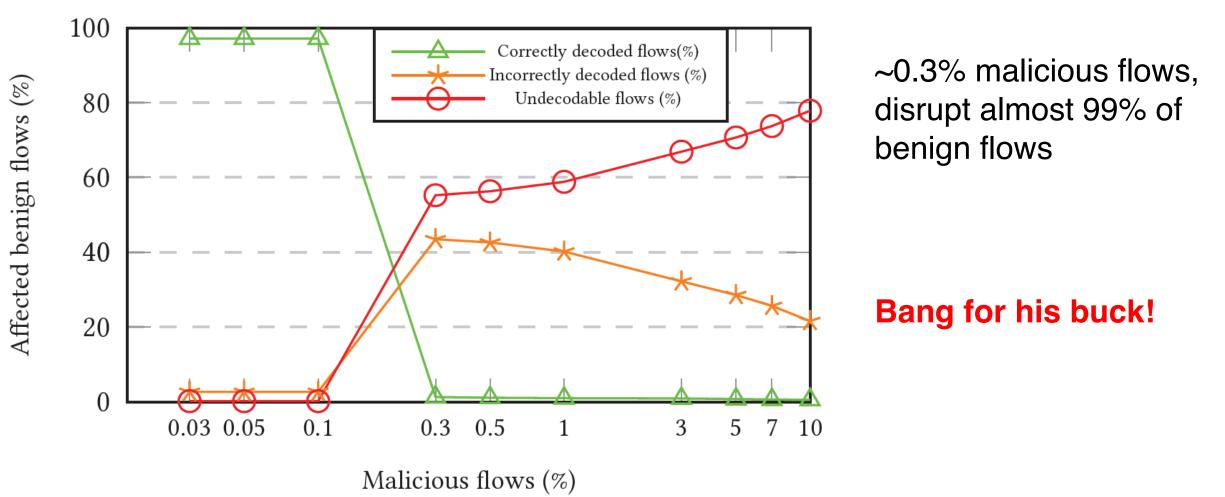
Flow filter

Results: CIA & QoA

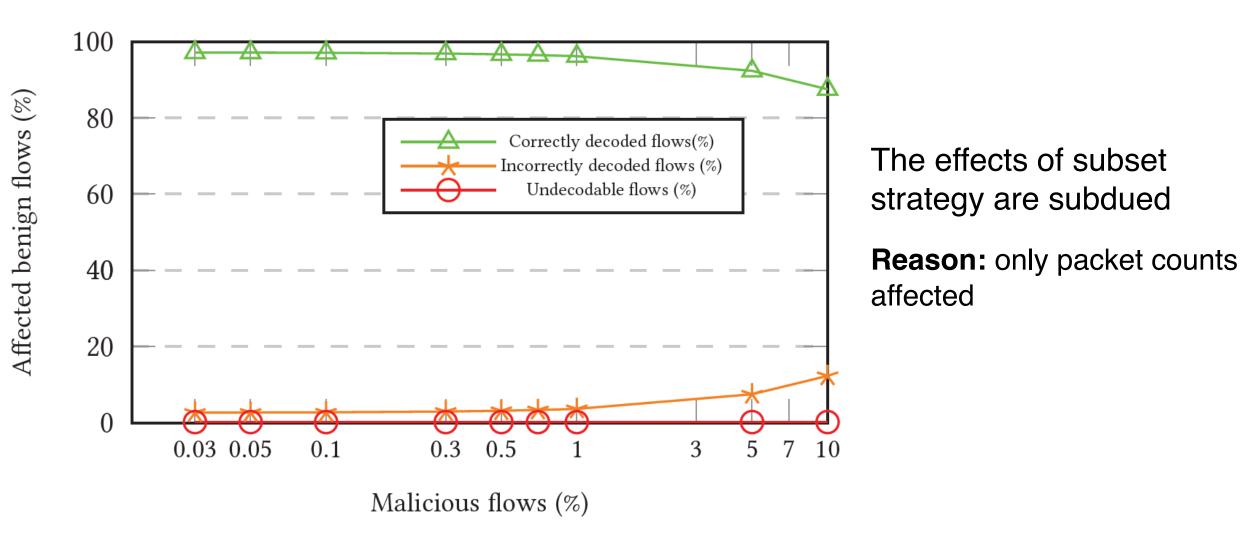
Correctly decoded flows

Incorrectly decoded flows

Undecodable flows



Results: Subset



Summary and Future work

- We empirically gauged the impact of polluting attack on FlowRadar
- An adversary can corrupt the traffic statistics with only a few crafted malicious flows

• Future work:

- Extend our analysis to other fast control loop systems (e.g., RouteScout)
- Develop detection and defence mechanisms

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Thank You

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