## SONATA: Query-Driven Network Telemetry

# Arpit Gupta Princeton University

Rob Harrison, Ankita Pawar, Rüdiger Birkner,

Marco Canini, Nick Feamster, Jennifer Rexford, Walter Willinger

## **Conventional Network Telemetry**



## **Conventional Network Telemetry**



## Collection is not driven by Analysis

## **Problems with Status Quo**

## • Expressibility

- Configure collection & analysis stages separately
- Static (and often coarse) data collection
- Brittle analysis setup---specific to collection tools

## **Problems with Status Quo**

- Expressibility
  - Configure collection & analysis stages separately
  - Static (and often coarse) data collection
  - Brittle analysis setup---specific to collection tools
- Scalability

As Traffic Volume or # Monitoring Queries increases

Hard to answer queries in real-time

Hard to **express** & **scale** queries for network telemetry tasks!

# **SONATA: Query-Driven Telemetry**

Uniform Programming Abstraction

Express queries as dataflow operations over pkt. tuples

- Query-Driven Data Reduction
   Execute subset of dataflow operations in data plane
- Coordinated Data Collection & Analysis
   Select query plans that make best use of available
   resources

# **Uniform Programming Abstraction**

Extensible Packet-tuple Abstraction

Queries operate over all packet tuples, at every location in the network

- Expressive Dataflow Operators
  - Most telemetry applications require
    - collecting aggregate statistics over subset of traffic
    - joining results of one analysis with the other
  - Easy to express them as declarative queries composed of dataflow operators

# **Example Query**

## **Detecting DNS Reflection Attack**

Detect hosts for which # of unique source IPs sending DNS response messages exceeds threshold (Th)

victimIPs = pktStream(W)
 .filter(p => p.srcPort == 53)
 .map(p => (p.dstIP, p.srcIP))
 .distinct()
 .map((dstIP, srcIP) => (dstIP, 1))
Express queries without worrying about
where and how they get executed

# **Changing Status Quo**

## • Expressibility

- Express dataflow queries over packet tuples
- Not tied to low-level (3<sup>rd</sup> party/platform-specific) APIs
- Trivial to add new queries and change collection tools

# **Query Execution**

Process all (or subset of) captured packet tuples using state-of-the-art **Stream Processor** 



## Expressible but not Scalable!

## **PISA Targets for Data Reduction**

Programmable parsing

Allow new query-specific header fields for parsing

- State in packets & registers
   Support simple stateful computations
- Customizable hash functions
   Support hash functions over flexible set of fields
- Flexible match/action table pipelines Support match/action tables with prog. actions

# **Compiling Dataflow Operators**

• Map, Filter & Sample

Apply sequence of match-action tables

## Distinct & Reduce

- Compute index, & read value from hash tables
- Apply function (e.g., bit\_or for distinct) & then update the hash table
- Use sketches, e.g. reduce(sum)  $\rightarrow$  CM Sketches

## Limitations

- Complex transformations, e.g. log, regex, etc.

# **Compiling Dataflow Queries**

- Compiling a Single Query
  - Generate & update query-specific metadata fields
  - Apply operator's match-action tables in sequence
  - Clone packet if *report bit* set
- Compiling Multiple Queries
  - Generate & update metadata fields for all queries
  - Apply operators for all queries in sequence
  - Clone packet if *report bit* is set for at least one query

# **Coordinated Data Coll. & Analysis**

## Query Partitioning

- Execute subset of dataflow operators in data plane
- Reduce packet tuples at the cost of additional state in the data plane

## Iterative Refinement

- Iteratively zoom-in on traffic of interests
- Reduce state at the cost of additional detection delay

## How to select the best query plan?

# **Query Planning**

- Reflection Attack Query
- Partitioning Plans

Plan 1: Data Plane only Plan 2: Stream Processor only

## Refinement Plans

- Refinement key: dstIP
- Refinement levels: {/8, /32}

```
pktStream(W)
.filter(p => p.srcPort == 53)
.map(p => (p.dstIP, p.srcIP))
.distinct()
.map((dstIP, srcIP)=>(dstIP,1))
.reduceByKey(sum)
.filter((dstIP,count)=>count>Th)
.map((dstIP, count) => dstIP)
```

# **Query Planning**

- Reflection Attack Query
- Partitioning Plans
   Plan 1: Data Plane only
   Plan 2: Stream Processor only
- Refinement Plans
  - Refinement key: dstIP
  - Refinement levels: {/8, /32}





## Selects plan with smallest weighted cost

## Implementation



Data Plane Target

## **Evaluation**

## • Workload

Large-IXP network: 2 hours long IPFIX trace, 3 Tbps peak traffic, packet sampling rate = 1/10K

### • Queries

DDoS-UDP, SSpreader, PortScan, Reflection Attack

### Comparisons

Stream-Only, Part-OF, Part-PISA, Fixed-Refinement

# **Benefits of Query Planning**



- B<sub>max</sub>: Max. state data plane can support
- N<sub>max</sub>: Max. pkt. tuples stream processor can process
- Each color represents a unique query plan

# SONATA makes best use of available resources

## **Scaling Query Executions**



#### Number of pkt tuples processed by Stream Processor

Executing stateful operations in data plane reduces workload on Stream Proc.

## **Scaling Query Executions**



#### State (KB) required by data plane targets

# Iterative refinement reduces state required by the data plane targets

# **Changing Status Quo**

- Expressibility
  - Express Dataflow queries over packet tuples
  - Not worry about how and where the query is executed
  - Adding new queries and collection tools is trivial
- Scalability
  - Answers hundreds of queries in real-time for traffic volume as high as few Tb/s

## Expressible & Scalable!

- · แม่และ ที่เป็นสรรสน by เมื่อ รแลสมม ที่เป็นสรรบเ
- state in the data plane

# Summary

- SONATA makes it easier to express and scale network monitoring queries using
  - Programmable Data Plane
  - Scalable Stream Processor
- Running Code
  - Github: <u>github.com/Sonata-Princeton/SONATA-DEV</u>
  - Run test queries or express new ones
- SONATA@arxiv: <u>arxiv.org/abs/1705.01049</u>