

Enhancing Network Attack Detection with Distributed and In-Network Data Collection System

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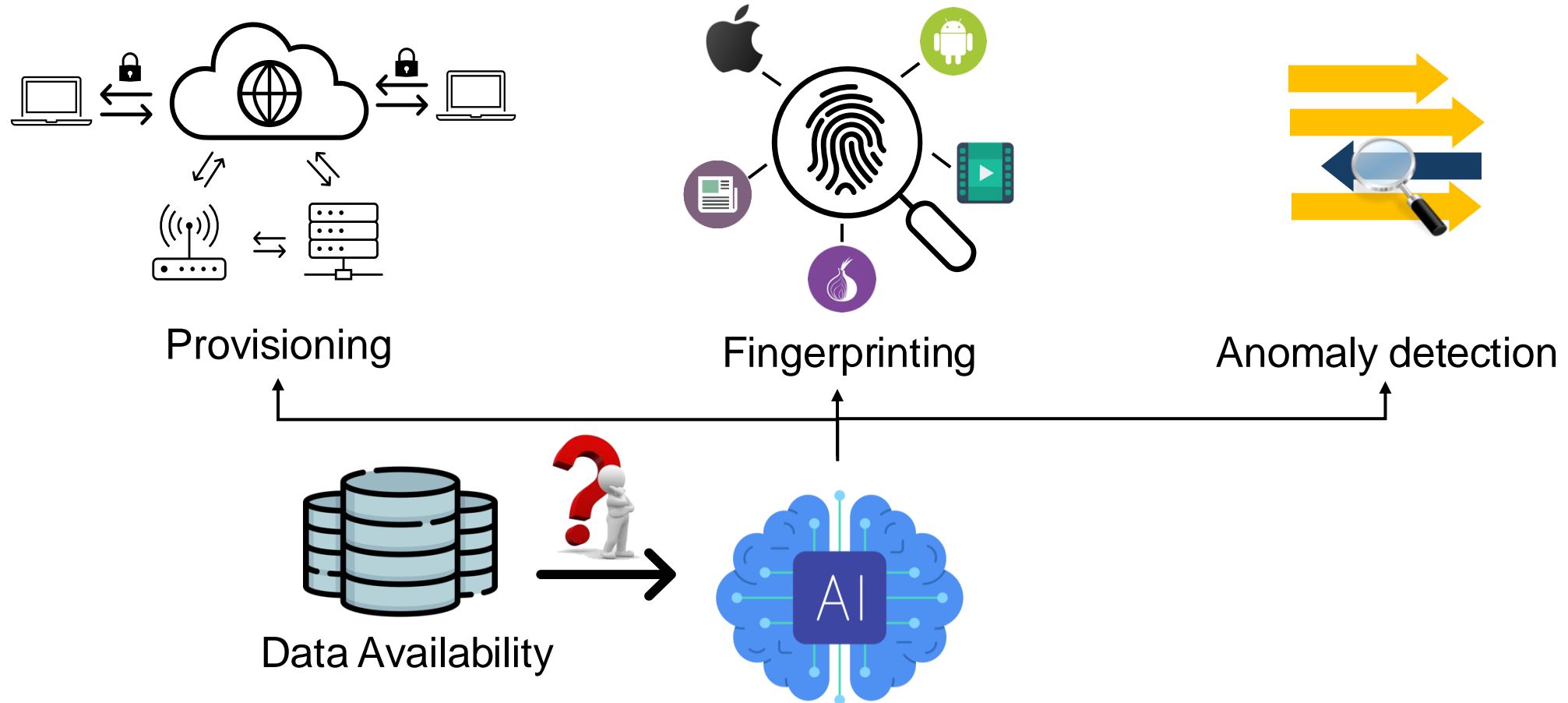
USENIX Security Symposium 2024

Outline

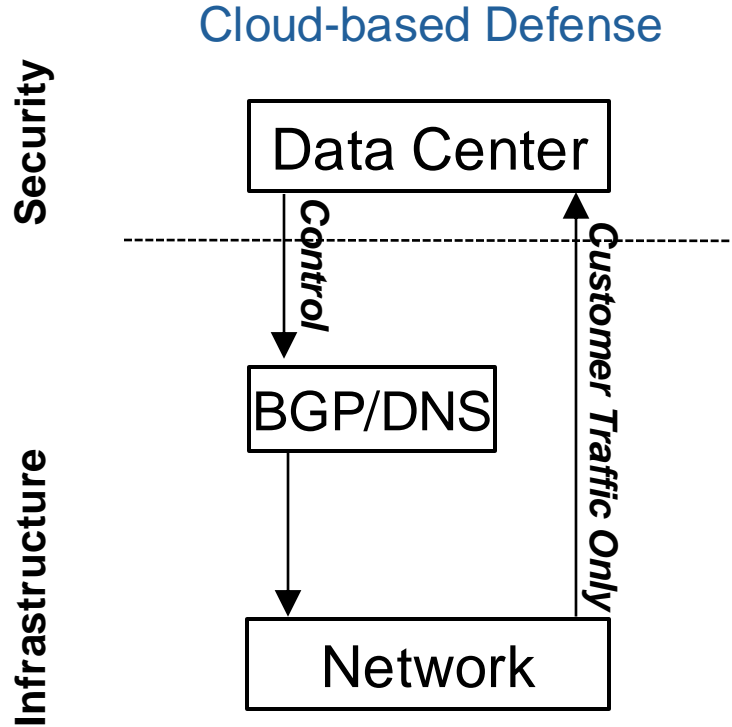


- Background - network traffic measurement
- Motivation - data availability for security
- Design goals - collaborative data collection
- Proposed system - ISDC
- Evaluation - covert channel/DDoS detection
- Conclusion

Background: Network Traffic Measurement

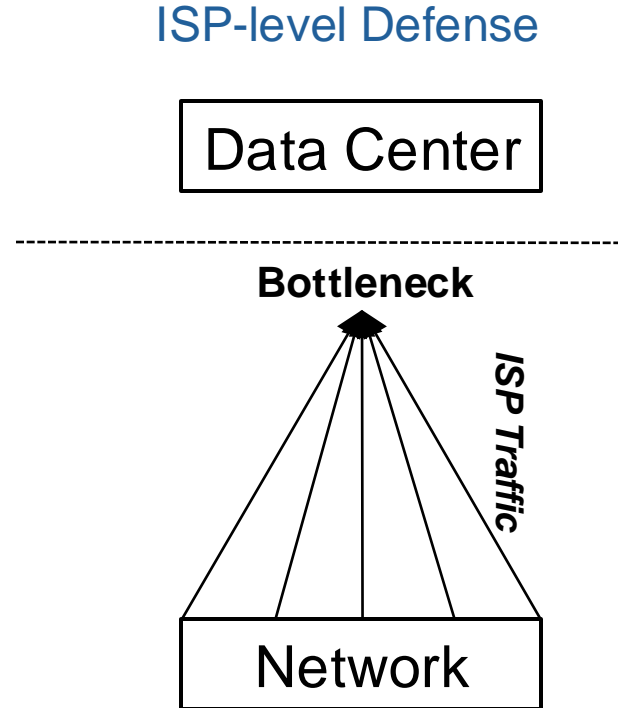


Motivation: Data Availability (DA) for Security



✗ Narrow Interest in Traffic

Weak DA

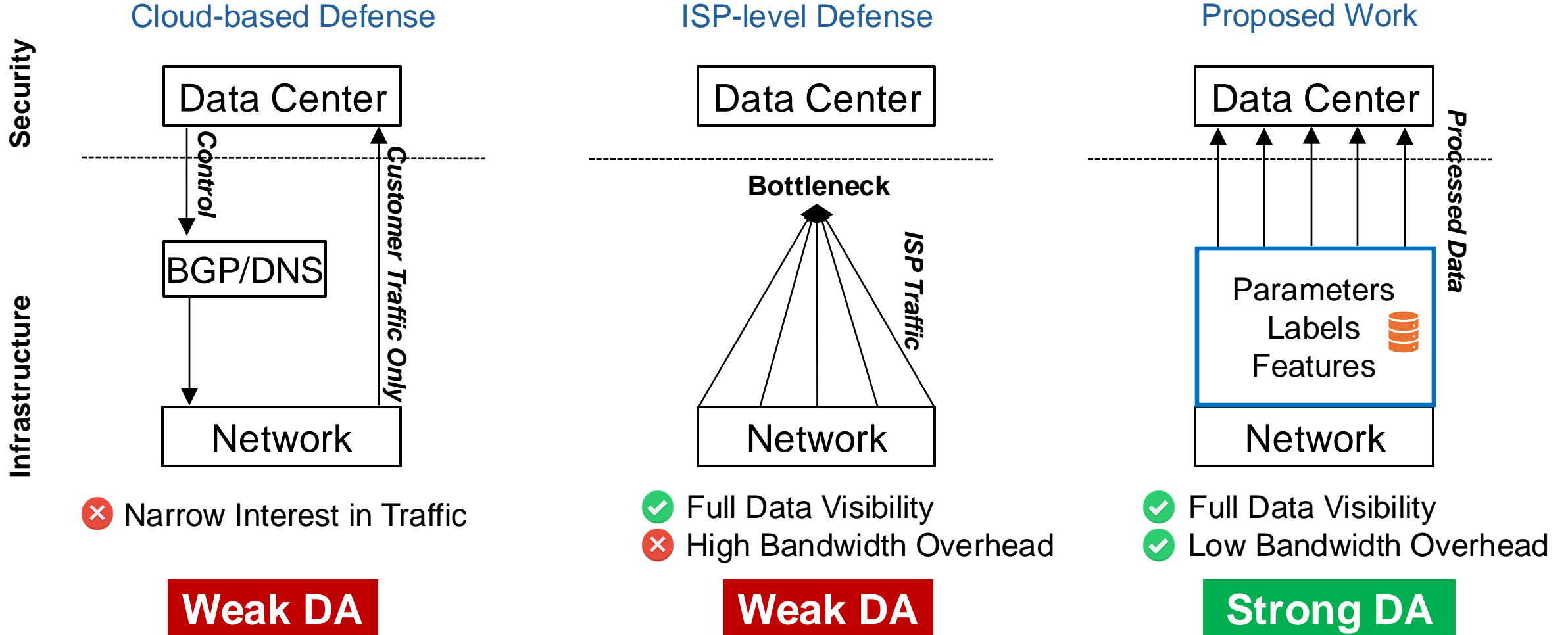


- ✓ Full Data Visibility
- ✗ High Bandwidth Overhead

Weak DA

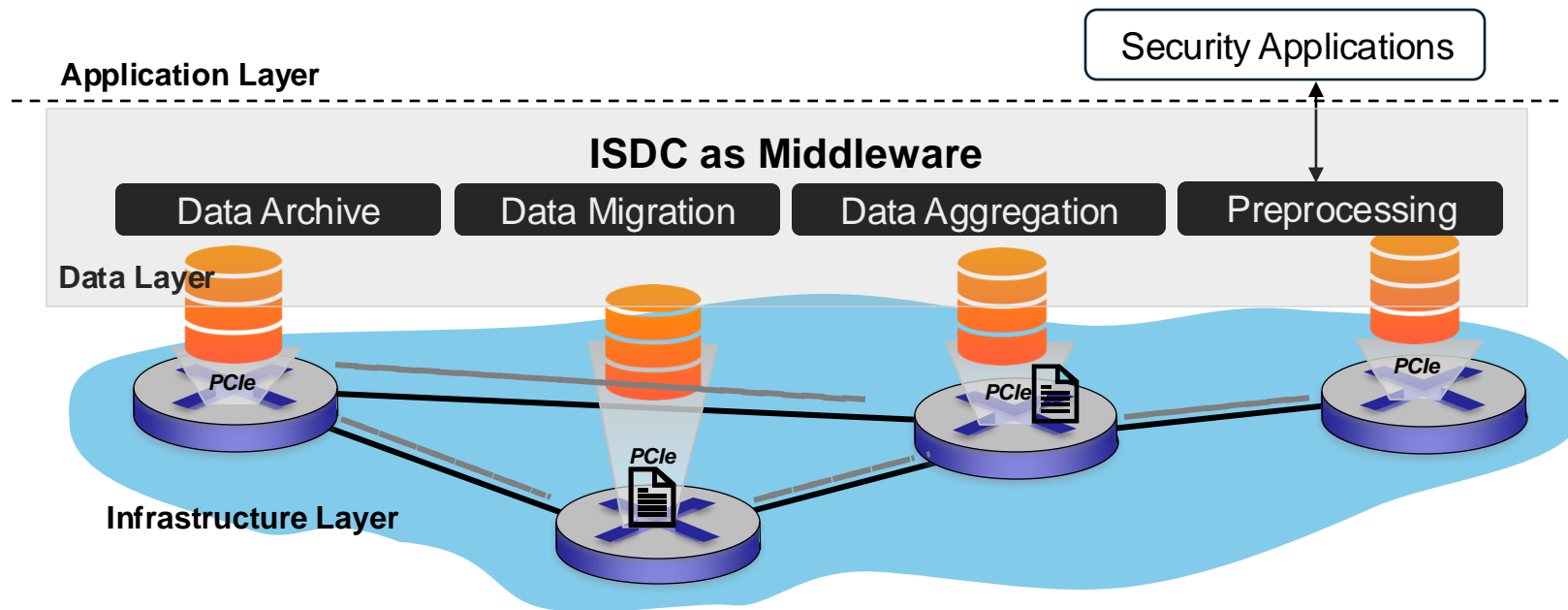


Motivation: Data Availability (DA) for Security



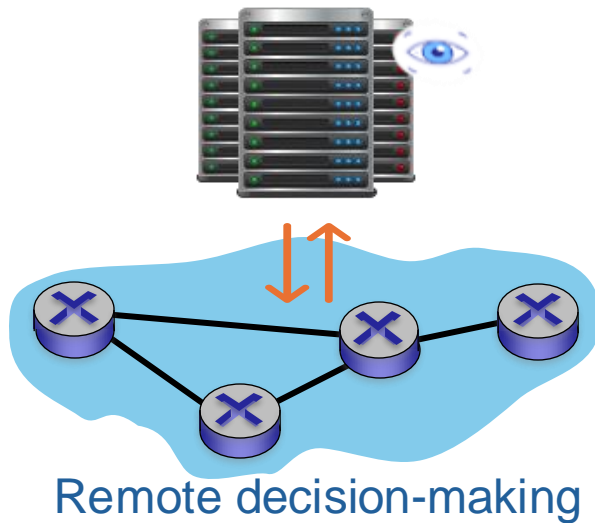
Our Goal: Collaborative Data Collection

- In-network Serverless Data Collection (ISDC)
 - Data plane collaborative network traffic measurement
 - Control plane (local switches) data aggregation/synchronization

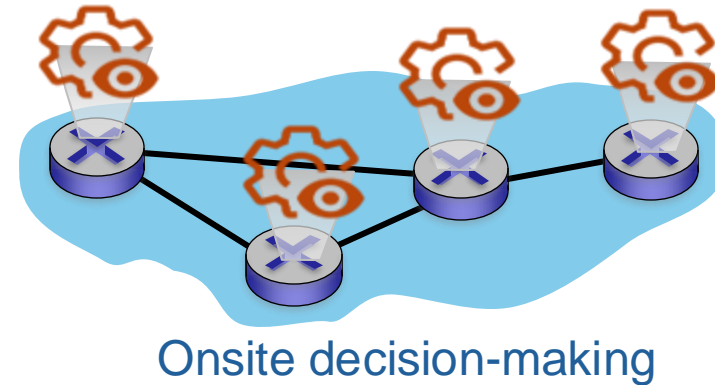


Prior Works: Resource Inefficiency during Collaboration

- Remote decision-making with global view for resource optimization ^{[1][2]}
- Onsite decision-making with local view for adaptiveness ^[3]



✗ Slow adaptation to dynamic shift



✗ Duplicated task measurement

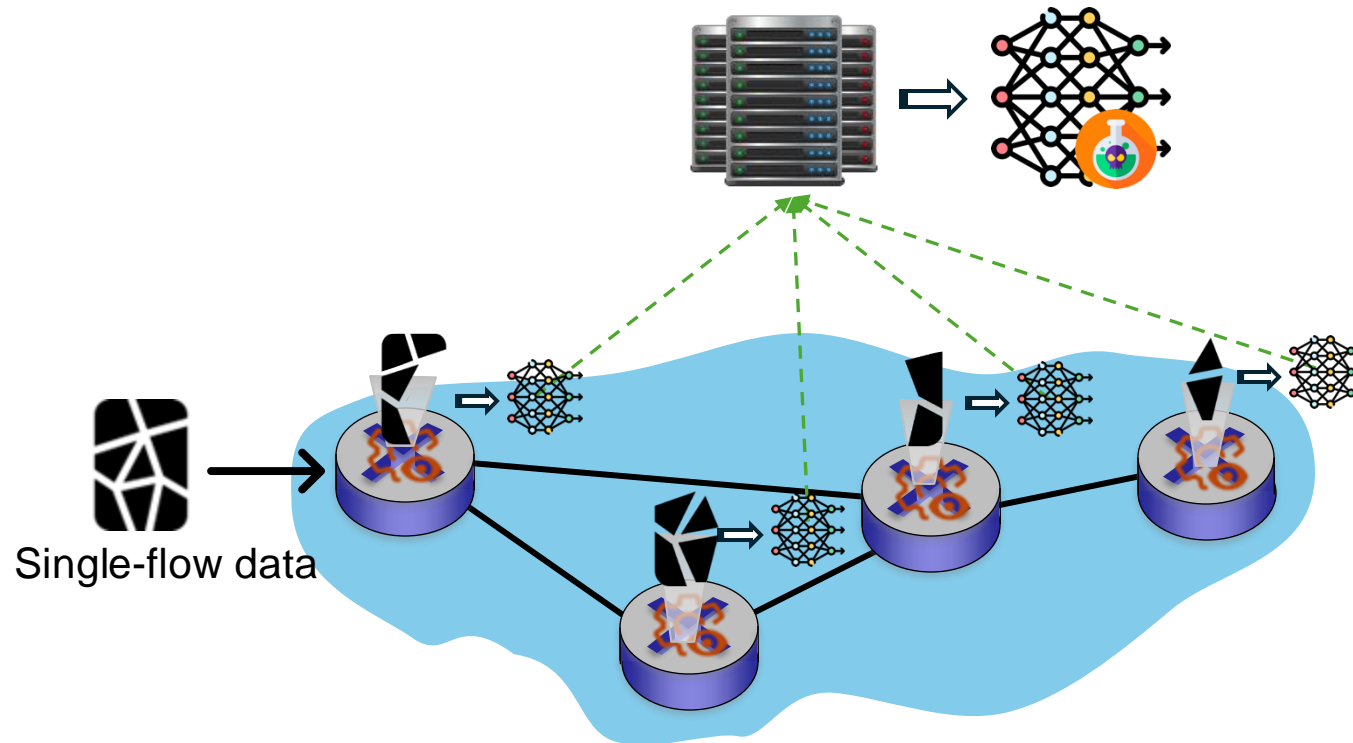
[1] Xu, Hongli, and et al. Lightweight flow distribution for collaborative traffic measurement in software defined networks. In Proc. of IEEE INFOCOM, 2019

[2] Sekar, Vyas, and et al. cSamp: A system for network-wide flow monitoring. In Proc. of USENIX NSDI, 2008

[3] Basat, Ran Ben, and et al. Cooperative network-wide flow selection. In Proc. of IEEE ICNP, 2020

New Insight: Data Fragmentation and Model Poisoning

- Local view decision-making creates fragmented data
 - Collected data is utilized as data source for distributed learning
 - Presence of fragmentation leads to model poisoning



Design Goals

Goal 1: Optimize Network Resource Usage

- Effective resource utilization according to security application demands

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Goal 2: Dynamic Task Allocation

- Efficient task coordination to maximize network-wide resources

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- Effective resource utilization according to security application demands

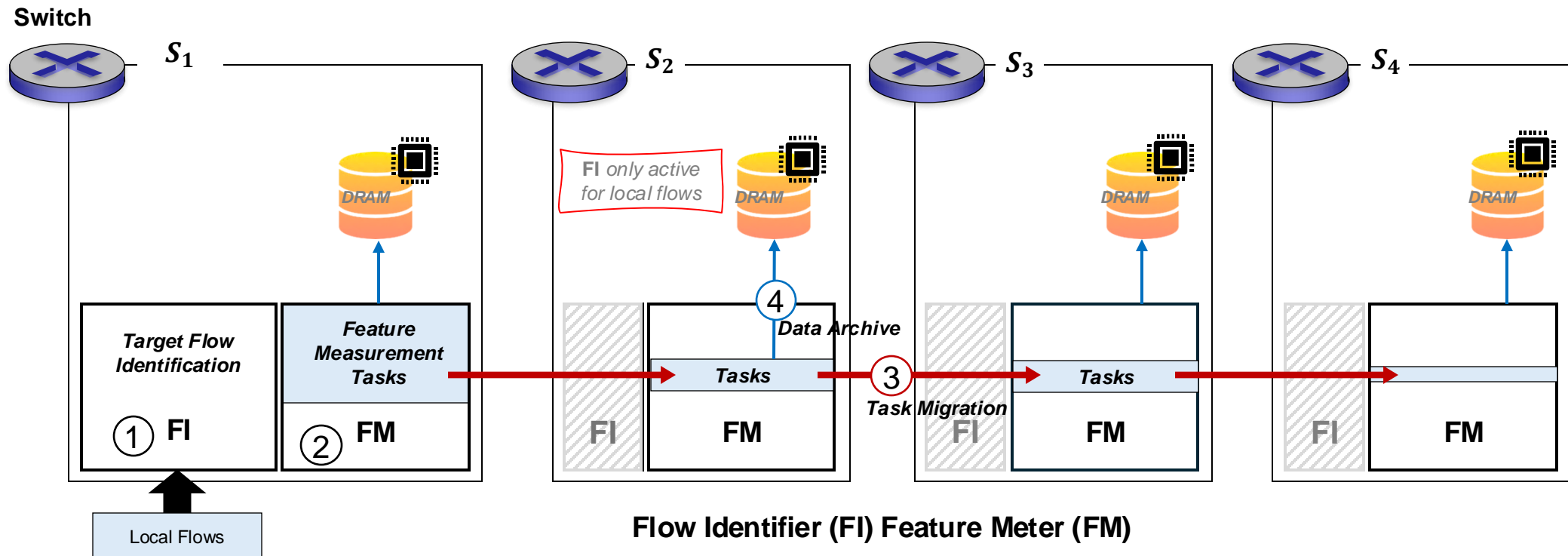
Goal 2: Dynamic Task Allocation

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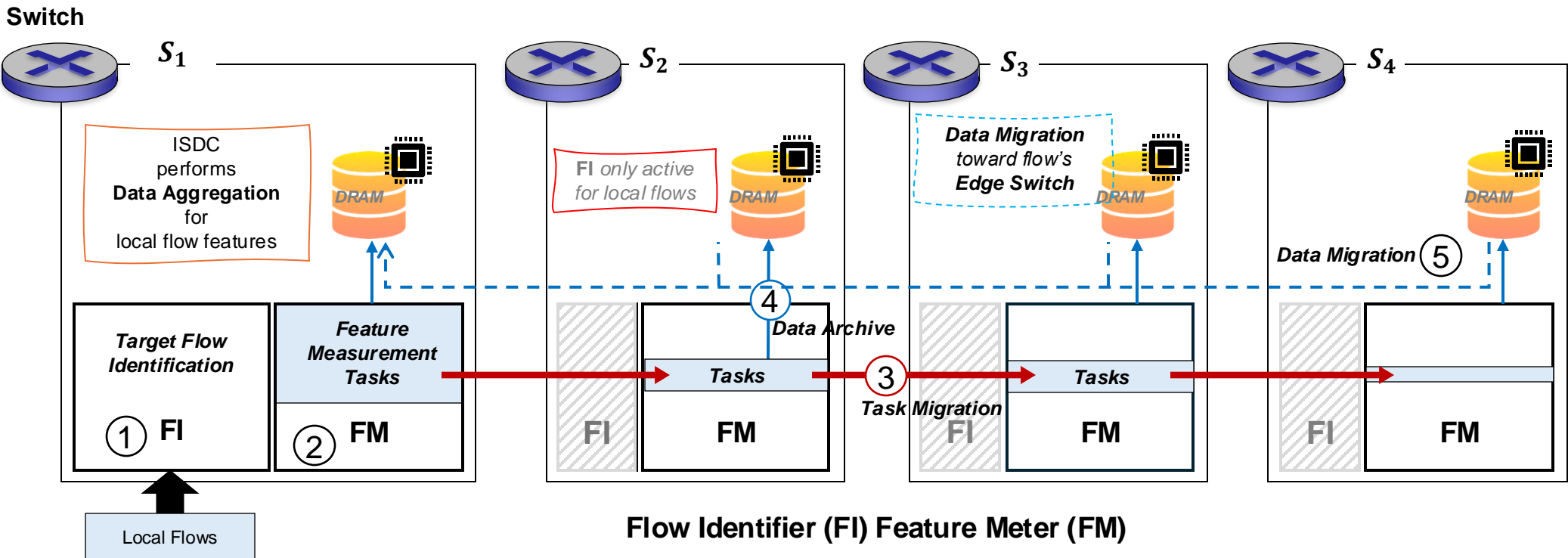
Goal 3: Reliable Data Source for Security

- Ensure data integrity to eliminate model poisoning caused by data fragmentation

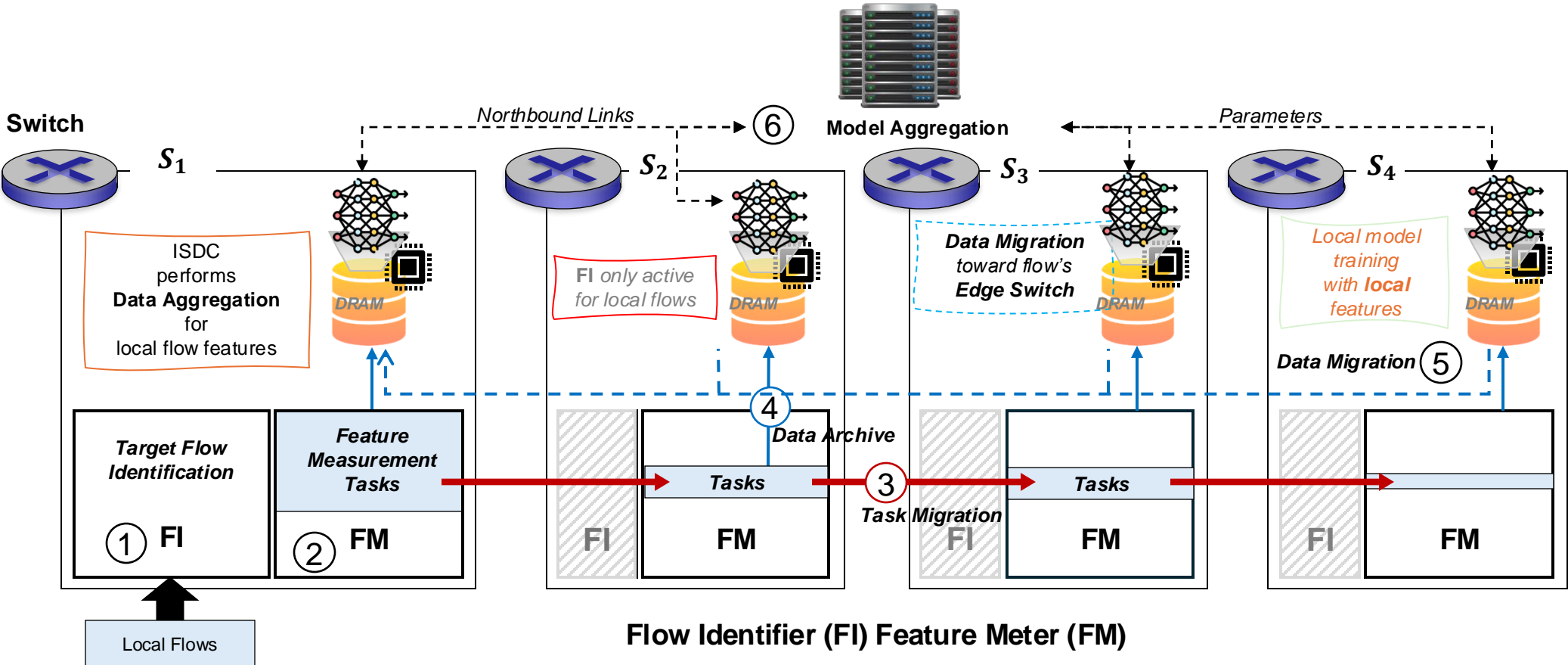
ISDC: Framework



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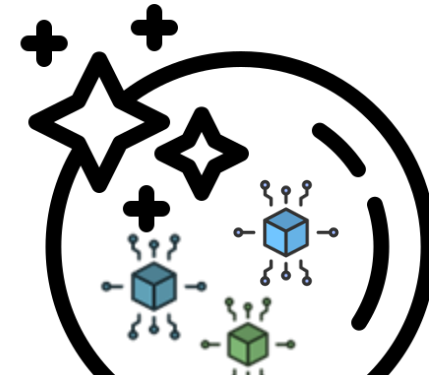


ISDC: Framework



Design 1: Task Prioritization

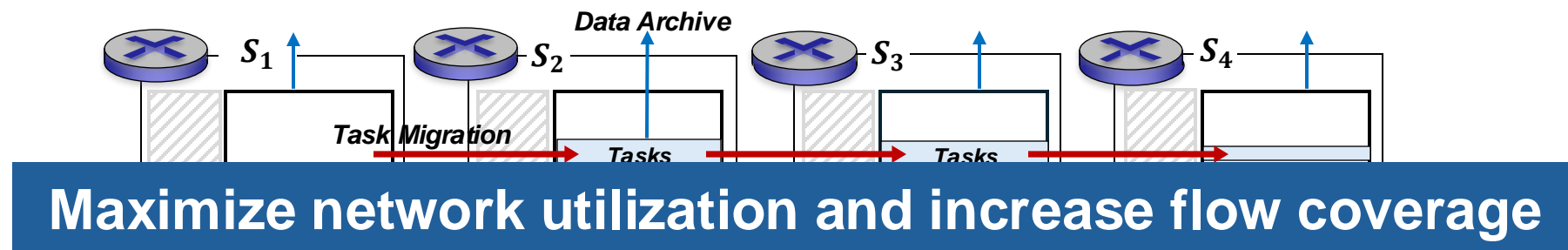
- **Challenge:** Achieving full-flow coverage is infeasible
 - With the ever-increasing traffic volume
- **Our approach:** Application-focused prioritization
 - ML/DL disfavor sparse data points created by super mice flows with one or two packets
- **Flow Identifier (FI):** Real-time large flow prediction
 - Reducing memory/computational complexity from $O(n)$ to $O(1)$



Reduced resource wastage in data collection for security application

Design 2: Dynamic Task Allocation

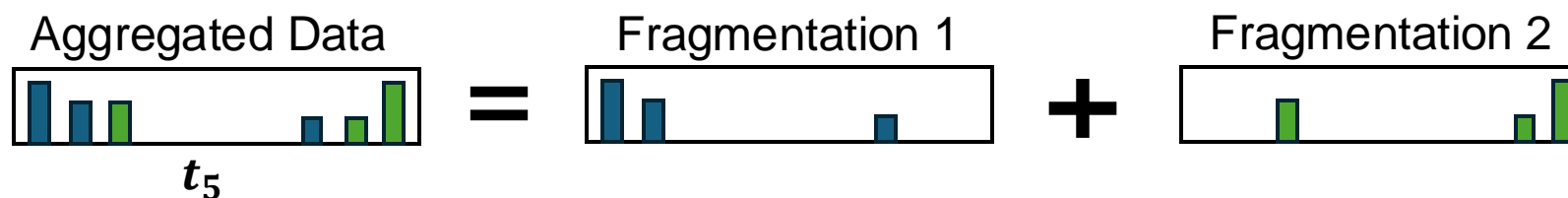
- **Challenge:** Lack of efficient collaboration
 - State-of-the-art onsite decision-making suffers from duplicated task measurement
- **Our approach:** Efficient and dynamic task collaboration
 - Having more task migration based on switch resources
 - When the task is migrated, the data is archived (task-data isolation)
- **Task migration:** A light-weight coordination protocol
 - A hybrid policy that applies two opposing strategies to maximize resource utilization and minimize task migration footprint



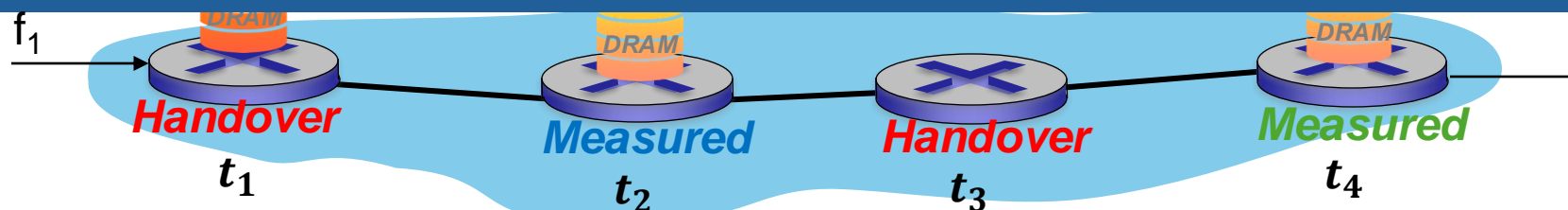
Flow Identifier (FI) Feature Meter (FM)

Design 3: Data Migration

- **Challenge:** Local view decision-making creates data fragmentation
 - Data fragmentation leads to model poisoning
- **Our approach:** In-network data aggregation
 - To enable a reliable foundation for ML/DL application
- **Data migration:** A light-weight, non-blocking protocol for data delivery/acknowledgment
 - No prior knowledge of network topology and routing path



High-quality data without fragmentation via in-network data aggregation



Evaluation: Setup

- Hardware and software implementations:

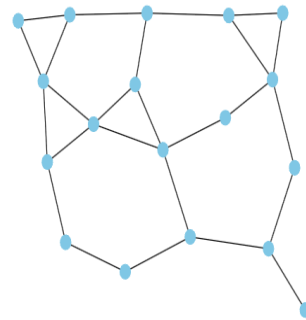
- bmv2 P4 software switch in Mininet environment
- Wedge 100BF-32X ASIC (Intel Tofino 1)

- Network topologies:

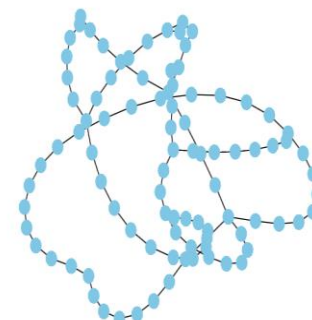
- Small: 18/25 switch/links (ASN)
- Medium: 92/96 switch/links (Vlt Wavenet)
- Large: 161/328 switch/links (Tiscali)

- Security use cases:

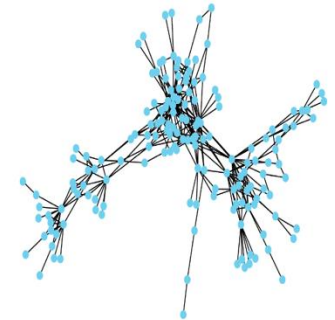
- Covert channel attack detection
- DoS/DDoS attack detection



ASN



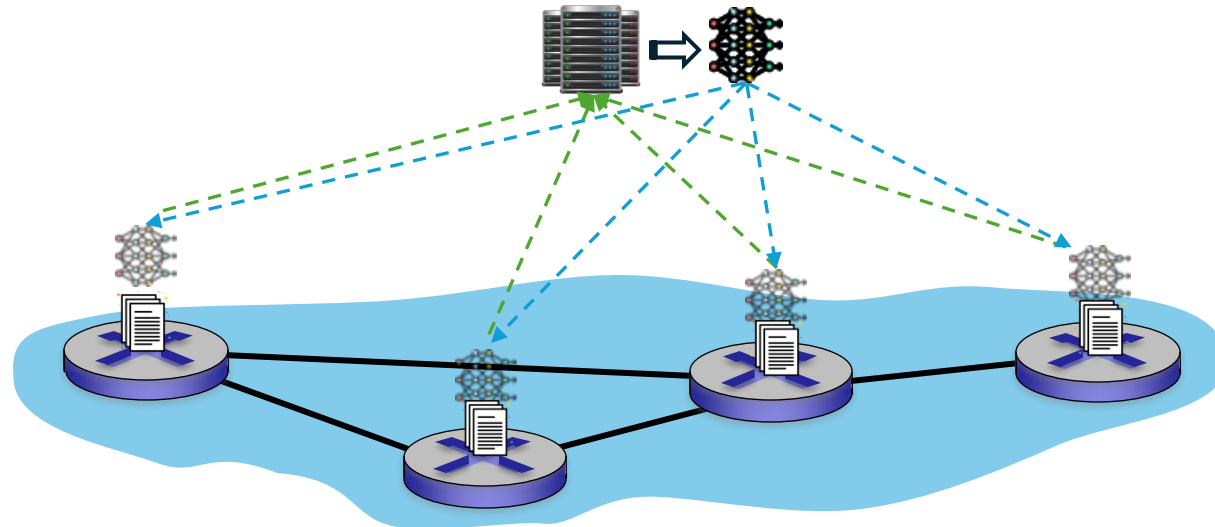
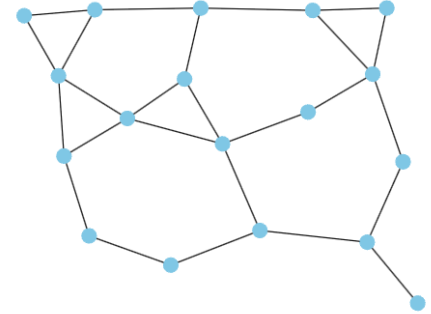
Vlt Wavenet



Tiscali (AS 3257)

Experimental Setting for Use Cases

- ASN topology with 18 switches and 25 links
- Collection of features from attack/benign traffic
 - Measured and aggregated in a distributed manner
- Distributed local data used for standard federated learning
- The global model is distributed to 18 switches for attack detection



Security Use Cases

#1: Covert Channel Detection

| Schemes | Cov. | Frag. | Mem. Waste | Avg. WMRE | F1 | | AUC | |
|-------------|--------------|-----------|--------------|-------------|--------------|--------------|--------------|--------------|
| | | | | | 1 rd. | 10 rd. | 1 rd. | 10 rd. |
| Strawman | 30.6% | 0% | 52.5% | 1.37 | 0.295 | 0.927 | 0.246 | 0.869 |
| CSAMP | 35.4% | 0% | 51.8% | 1.27 | 0.824 | 0.923 | 0.718 | 0.862 |
| NSPA | 36.9% | 0% | 51.3% | 1.25 | 0.816 | 0.927 | 0.709 | 0.868 |
| CFS | 58.1% | 53% | 62.1% | 1.67 | 0.887 | 0.942 | 0.857 | 0.894 |
| ISDC | 94.1% | 0% | 8.02% | 0.18 | 0.960 | 0.970 | 0.938 | 0.967 |

High flow coverage (Design 1, 2)

Lack of collaboration mentation (Design 3)

Remote decision-making

Onsite decision-making y (Design 1)

High feature accuracy (Design 3)

#2: DoS/DDoS Detection

| Schemes | Cov. | Frag. | Mem. Waste | Avg. WMRE | F1 | | AUC | |
|-------------|--------------|-----------|-------------|--------------|--------------|--------------|--------------|--------------|
| | | | | | 1 rd. | 10 rd. | 1 rd. | 10 rd. |
| CFS | 49.9% | 62% | 67.6% | 0.989 | 0.617 | 0.613 | 0.828 | 0.891 |
| CFS-clean | 49.9% | 0% | 67.6% | 0.868 | 0.620 | 0.756 | 0.777 | 0.892 |
| ISDC | 93.1% | 0% | 3.5% | 0.297 | 0.730 | 0.809 | 0.860 | 0.945 |

Enhanced ML performance
for security

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High flow coverage (Design 1, 2)

Zero data fragmentation (Design 3)

High memory efficiency (Design 1)

High feature accuracy (Design 3)

#2: DoS/DDoS Detection

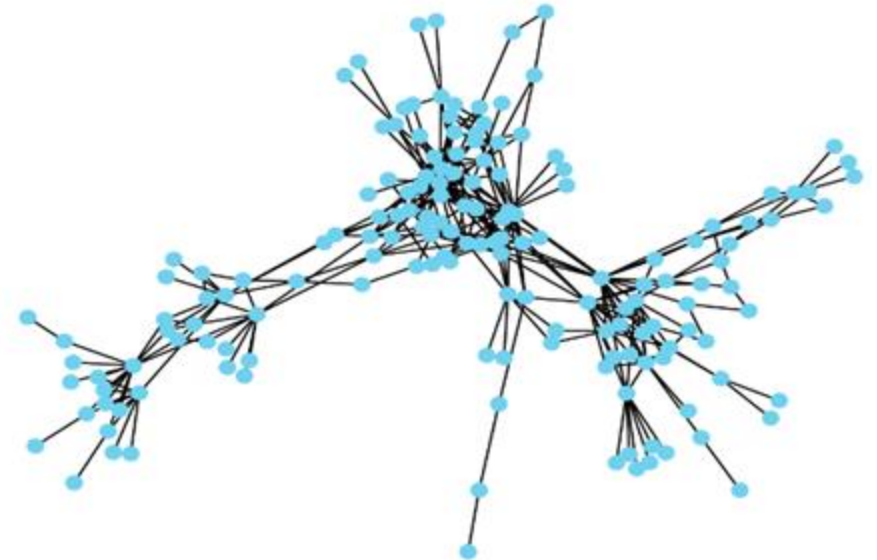
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+14% ↑

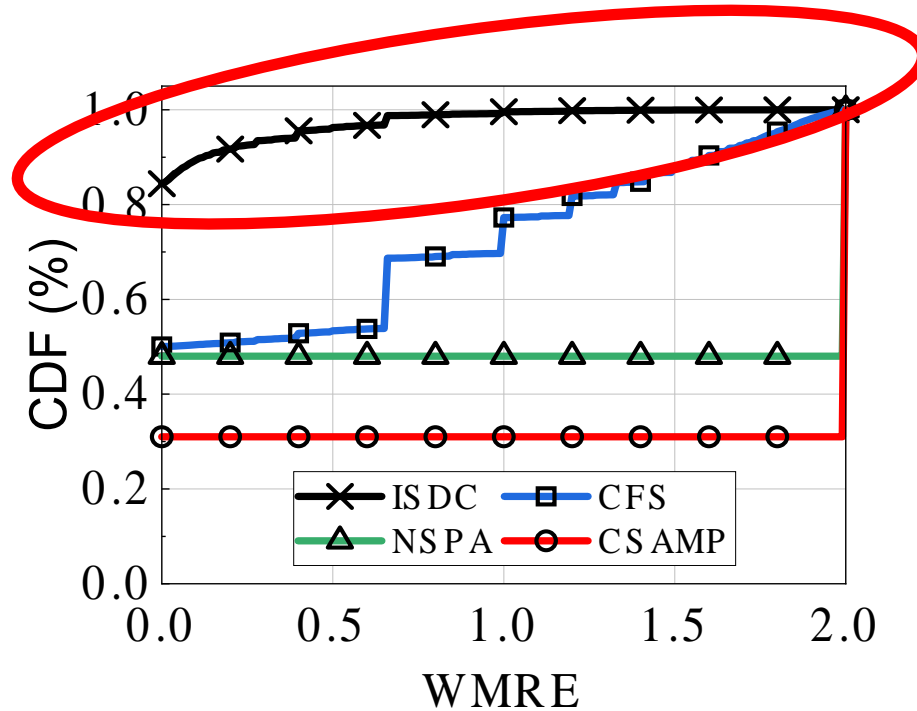
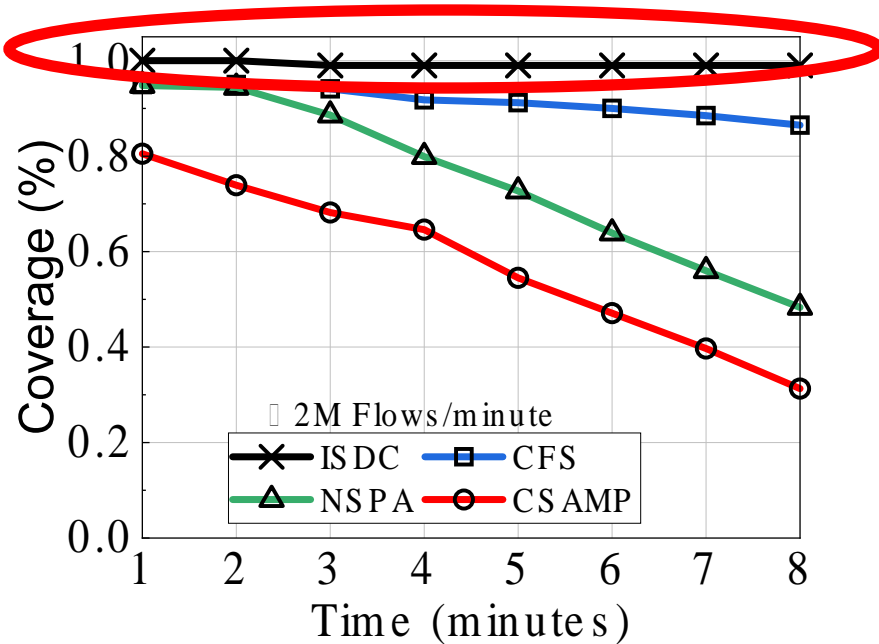
Enhanced ML performance
for security

Experimental Setting for System Evaluation

- **Large-scale** real-world topology with **161** switches and **328** links
- Used **eight**-minute CAIDA traffic, a total of **14.5/250** million flows/packets
- Metric
 - Flow coverage (%): higher is better
 - Feature quality (WMRE): smaller is better



System Performance: Data Collection



- Consistent delivery of full data coverage for top-500k flows
- Delivery of high-quality data, with **95%** of collected features have WMRE of less than **0.5**

Conclusion

- Limitation of existing collaborative framework
 - Resource wastage
 - Fragmentation caused model poisoning
- **ISDC**
 - Effective resource usage
 - Efficient resource allocation
 - Light-weight in-network data aggregation
- Achieved goal
 1. High **coverage** and **quality** data collection
 2. Enhanced **data availability** for ML/DL security application
- Source code: <https://github.com/NIDS-LAB/ISDC>

Q & A

Thank You!