

Understanding network traffic through Intraflow data

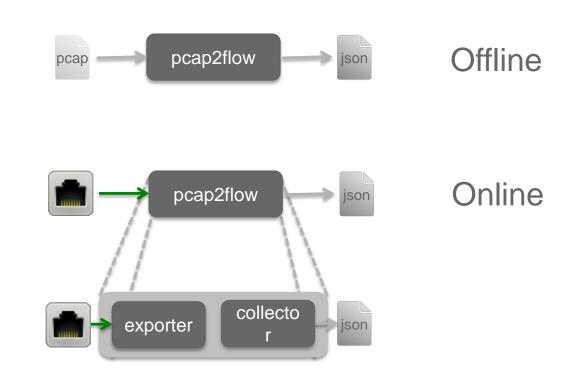
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FloCon 2016

Exploring threat data features at scale



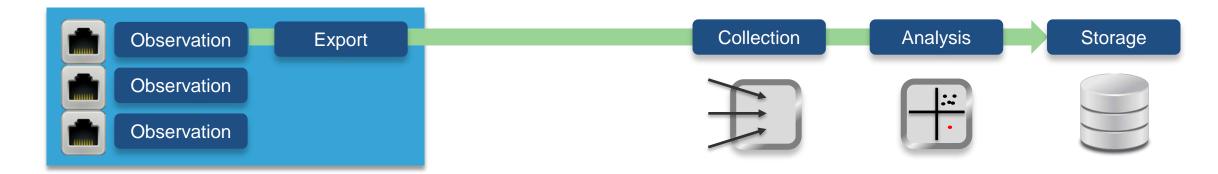


https://github.com/davidmcgrew/joy



Flow Monitoring

srcIP, dstIP, srcPort, dstPort, prot, startTime, stopTime, numBytes, numPackets





"I need to understand traffic even when it is encrypted"

Known threats and malware
Evasive applications and tunnels
TLS, SSH, and other encrypted
traffic (on **any** port)

"I need to understand all the traffic in my network, not just traffic that passes through a security appliance"

Monitoring internal traffic

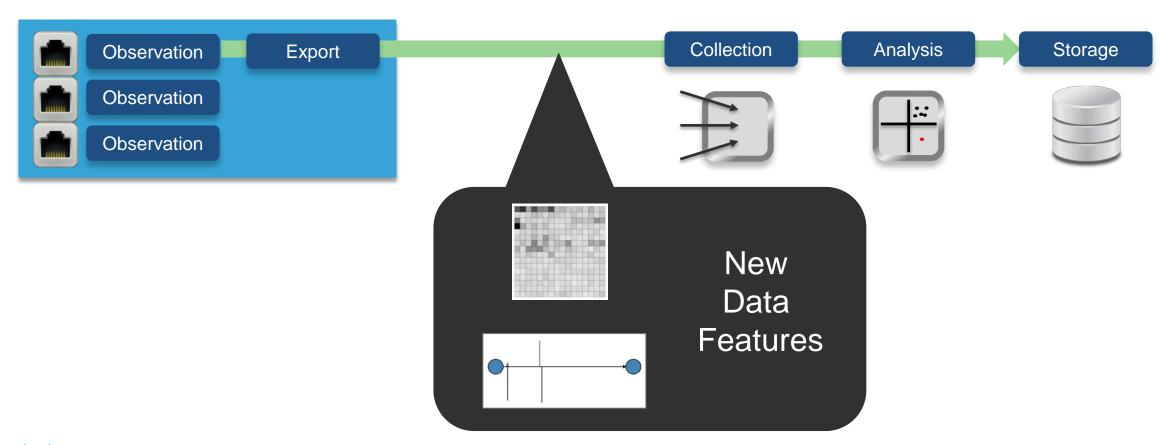
Forensics

Crypto usage audit



Flow Monitoring

srcIP, dstIP, srcPort, dstPort, prot, startTime, stopTime, numBytes, numPackets





Intraflow data

Information about events or data inside of flows that can be conveniently collected, stored, and analyzed within a flow monitoring framework

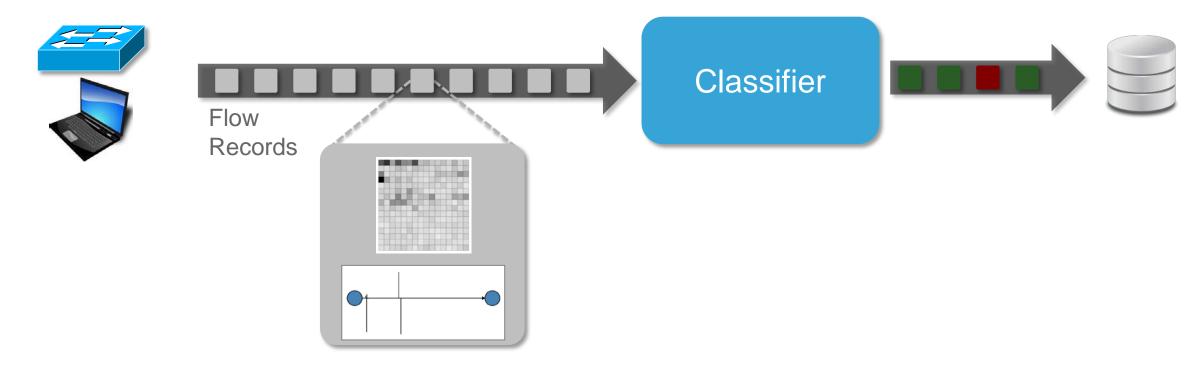


Intraflow data

- Economical observation
 - Unidirectional
 - Minimal computation
 - Small snaplen
- Application/protocol independence
- Compactness
 - Observation
 - Transmission and storage
- Composability

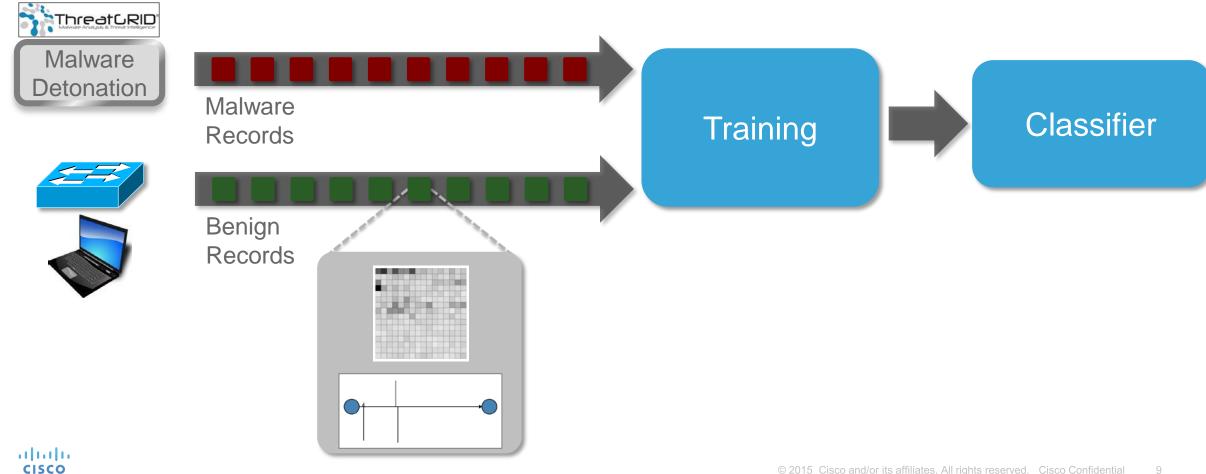


Architecture





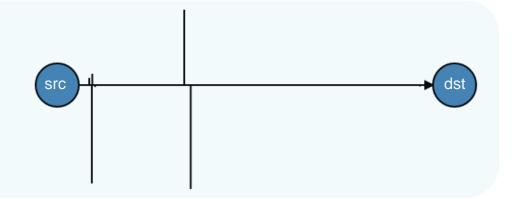
Training architecture



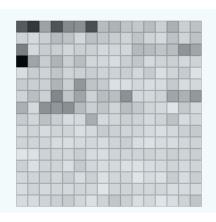
New Telemetry Data Features

Enhanced Telemetry Data Types

 SPLT – Sequence of Packet Lengths and Arrival Times



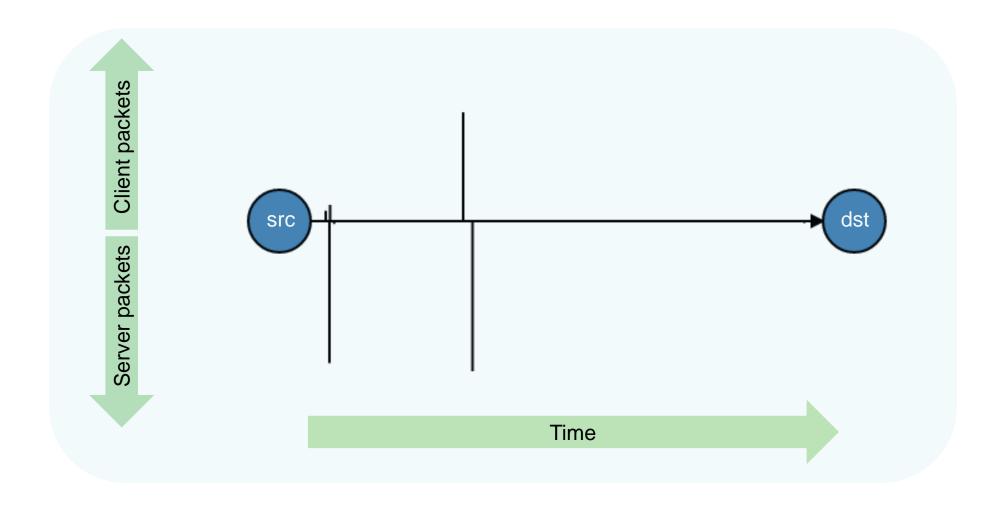
- Byte Distribution
 - Relative frequency for each byte in a flow
- Byte Entropy



Initial Data Packet



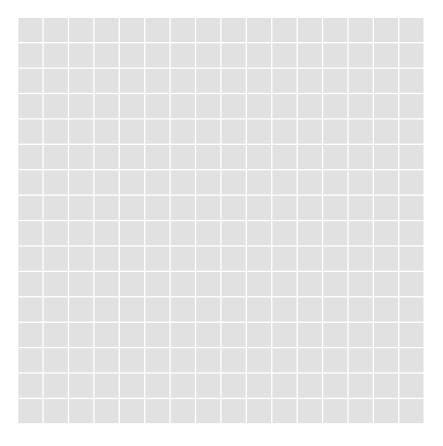
Sequence of Packet Lengths and Times





H T T P / 1 . 1 2 0 0 0 K

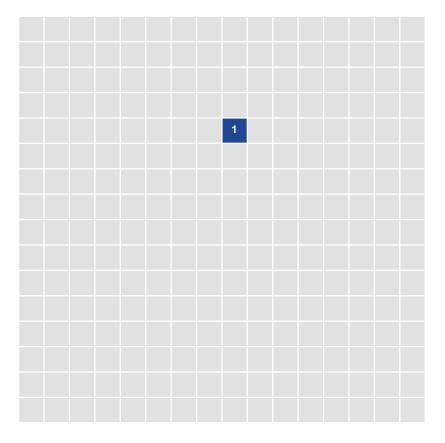
48 54 54 50 2f 31 2e 31 20 32 30 30 20 4f 4b





H T T P / 1 . 1 2 0 0 O K

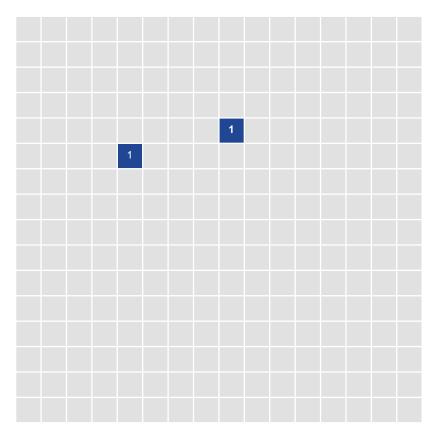
48 54 54 50 2f 31 2e 31 20 32 30 30 20 4f 4b





H T T P / 1 . 1 2 0 0 0 K

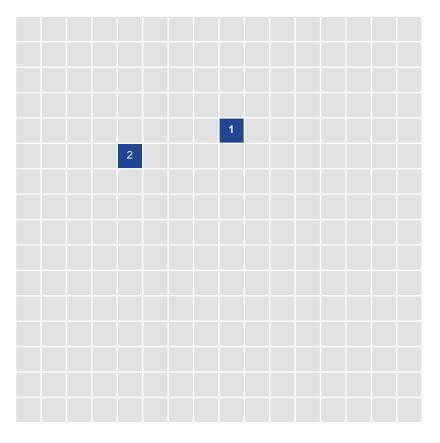
48 54 54 50 2f 31 2e 31 20 32 30 30 20 4f 4b





H T T P / 1 . 1 2 0 0 0 K

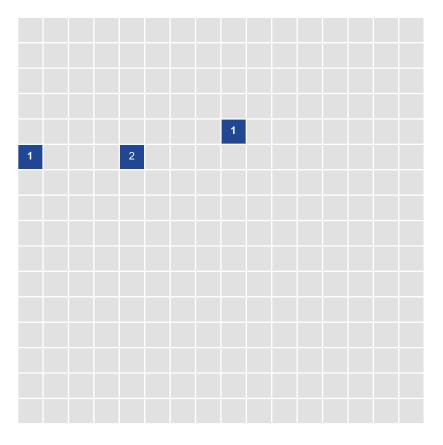
48 54 54 50 2f 31 2e 31 20 32 30 30 20 4f 4b





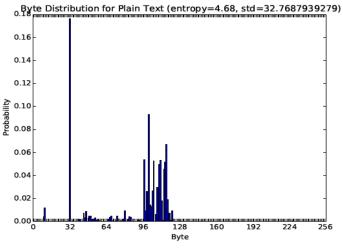
H T T P / 1 . 1 2 0 0 O K

48 54 54 50 2f 31 2e 31 20 32 30 30 20 4f 4b





Byte Distribution for different encodings



128

160

192

0.06

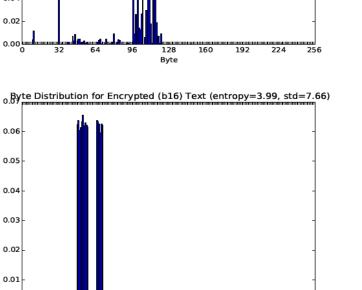
0.05

و 0.03

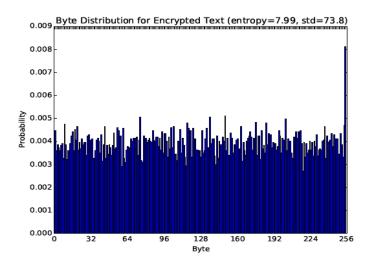
0.02

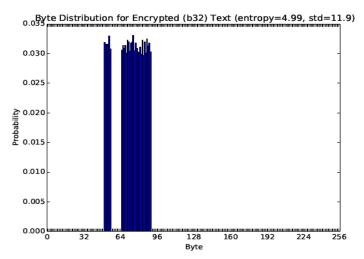
0.01

32



224







JSON flow data

Conventional flow data

Intraflow data

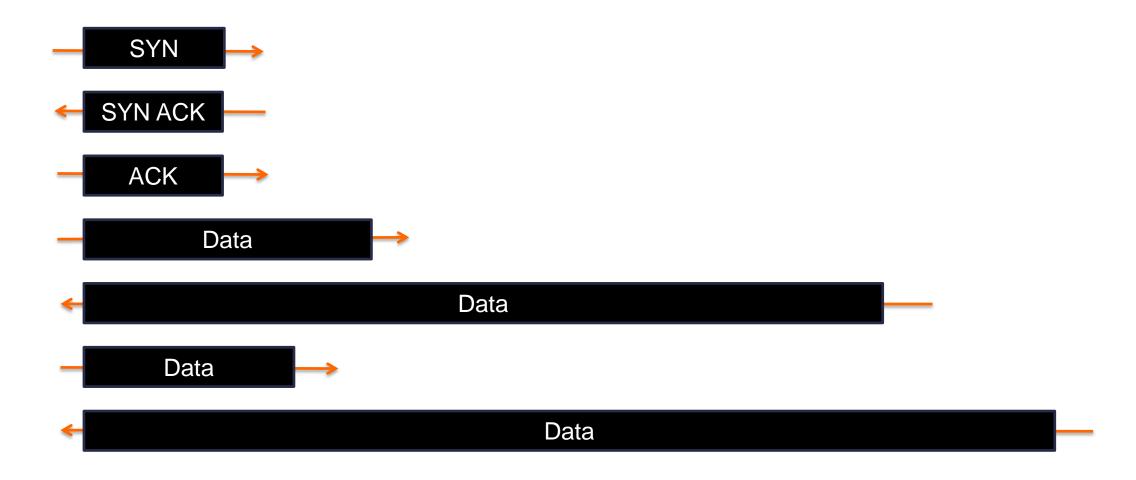
Extracted parameters

Easy to use with data analytics and machine learning tools

```
"da": "98.196.11.87",
           1460, "dir": "<", "ipt": 27 },
     "b": 1460, "dir": "<", "ipt": 395 },
     "b": 1460, "dir": "<", "ipt": 34 ]
"be": 7,960630
```

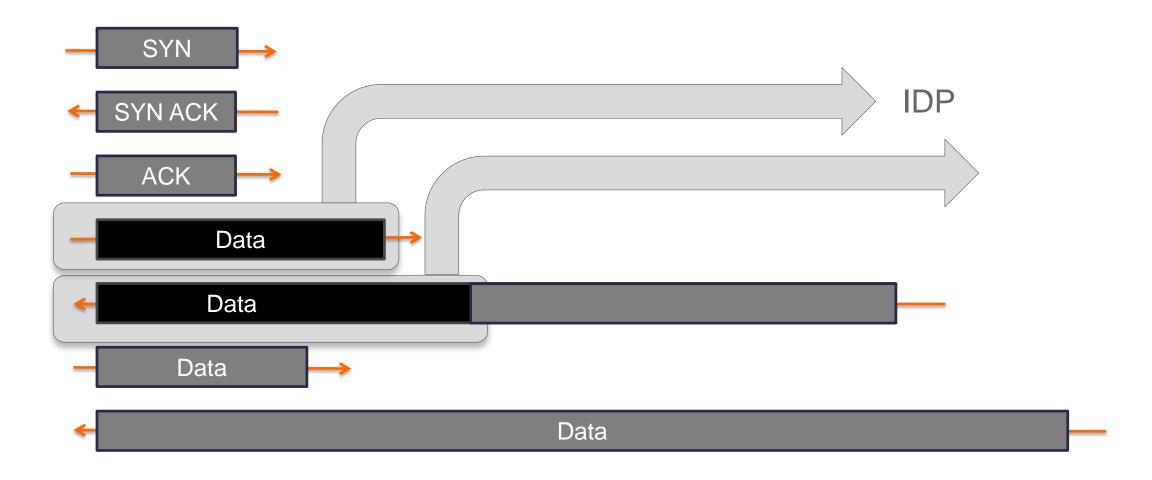
Options Buffers Tools Javascript Help

Initial Data Packet





Initial Data Packet

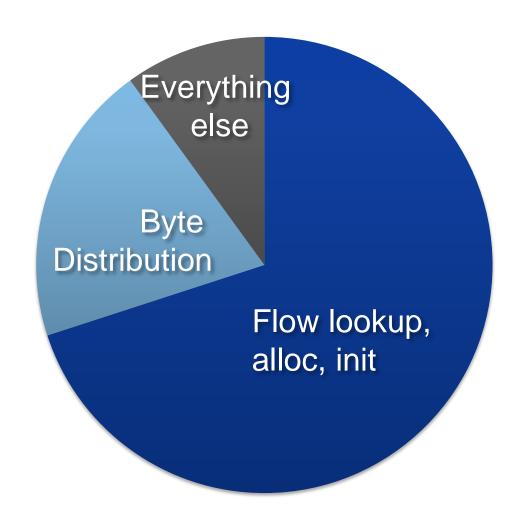




Experimental results

Performance

- CPU: Intel Xeon @ 2.70GHz
 - 17% utilization at 1.0e7 bytes/sec,
 1.2e4 packets/sec
 - Approx 870 Mbits/sec at full utilization
- RAM: 8Gbyte
 - 2.7% utilization (216 Mbyte)





Detecting malware with SPLT and Byte Distribution

Analytics User Interface

LAUI 0.2a Home Contact Local Analytics -

Number of flows classified: 244

P(Malware)	P(TLS)	Source Address	Dest.Address	Source Port	Dest. Port 🔷	Inbound Packets	Outbound Packets
1.0	0.99	172.16.45.20	104.237.132.39	1034	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1035	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1036	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1044	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1045	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1064	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1065	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1070	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1071	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1072	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1073	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1074	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1076	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1077	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1080	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1115	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1116	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1117	443	16	11
1.0	0.99	172.16.45.20	104.237.132.39	1118	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1119	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1120	443	15	11



LAUI 0.2a Home Contact Local Analytics -

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	1						
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1.0	0.99	172.16.45.20	104.237.132.39	1036	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1044	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1045	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1064	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1065	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1070	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1071	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1072	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1073	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1074	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1076	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1077	443	15	11
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1.0	0.99	172.16.45.20	104.237.132.39	1116	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1117	443	16	11
1.0	0.99	172.16.45.20	104.237.132.39	1118	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1119	443	15	11
1.0	0.99	172.16.45.20	104.237.132.39	1120	443	15	11

Classifying flows as malicious/benign

- L1-logistic regression
- SPLT + 5-tuple

- L1-logistic regression
- SPLT + 5-tuple + BD



Classifying flows as malicious/benign

- L1-logistic regression
- SPLT + 5-tuple
 - 160 non-zero parameters
 - 0.01 FDR: 51.11%
 - Total Accuracy: 98.44%

- L1-logistic regression
- SPLT + 5-tuple + BD
 - 128 non-zero parameters
 - 0.01 FDR: 98.92%
 - Total Accuracy: 99.81%



Intraflow data

- Economical observation
 - Unidirectional
 - Minimal computation
 - Small snaplen
- Application/protocol independence
- Compactness
 - Observation
 - Transmission and storage

Composability

SPLT 10 packets

Yes

Yes

Yes

Yes

10 bytes

10 bytes

Yes

Byte Distribution

Yes

Yes

No

Yes

256 bytes

256 bytes

Yes

Intraflow data

- Economical observation
 - Unidirectional
 - Minimal computation
 - Small snaplen
- Application/protocol independence
- Compactness
 - Observation
 - Transmission and storage

Composability

SPLT 10 packets

Yes

Yes

Yes

Yes

10 bytes

10 bytes

Yes

Byte Distribution

Yes

Yes

No

Yes

16 bytes

256 bytes

256 bytes

Yes

Conclusions

- Intraflow data is feasible to implement, enables useful inferences
- SPLT is valuable and relatively cheap
- Byte Distribution is valuable but more costly
- Training classifiers is key
 - Data fusion





Thank You

Joy applications

