



# ESnet

ENERGY SCIENCES NETWORK

# ESnet6 High Touch Services

## Precision Streaming Network Telemetry

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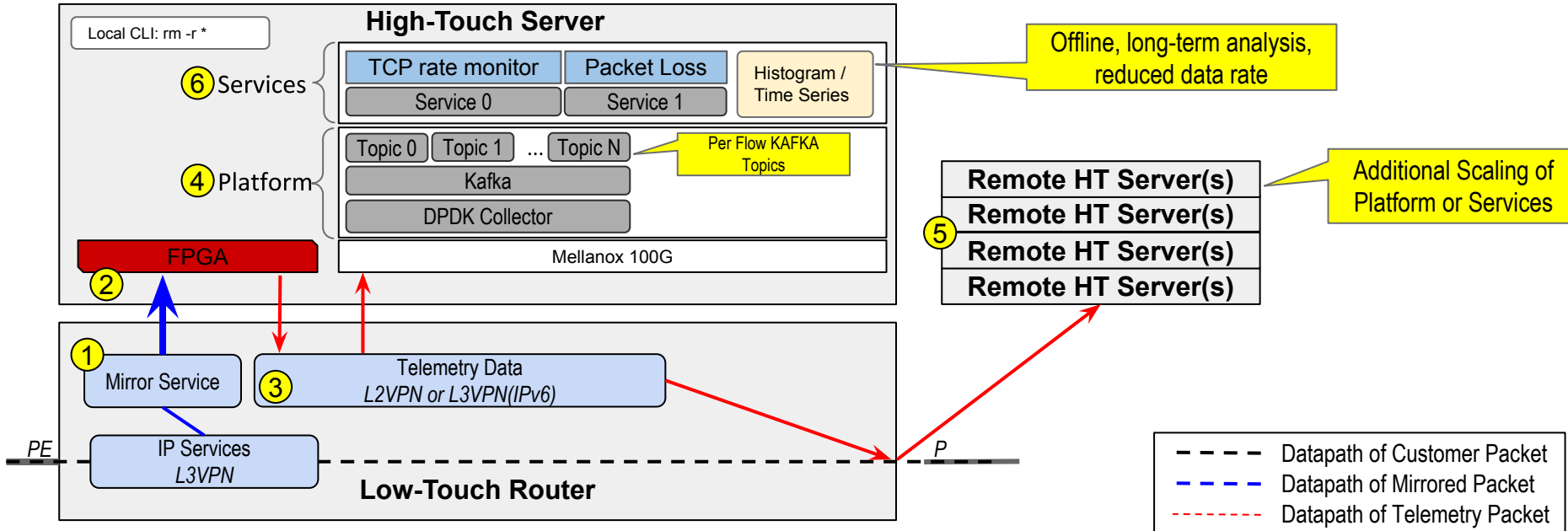


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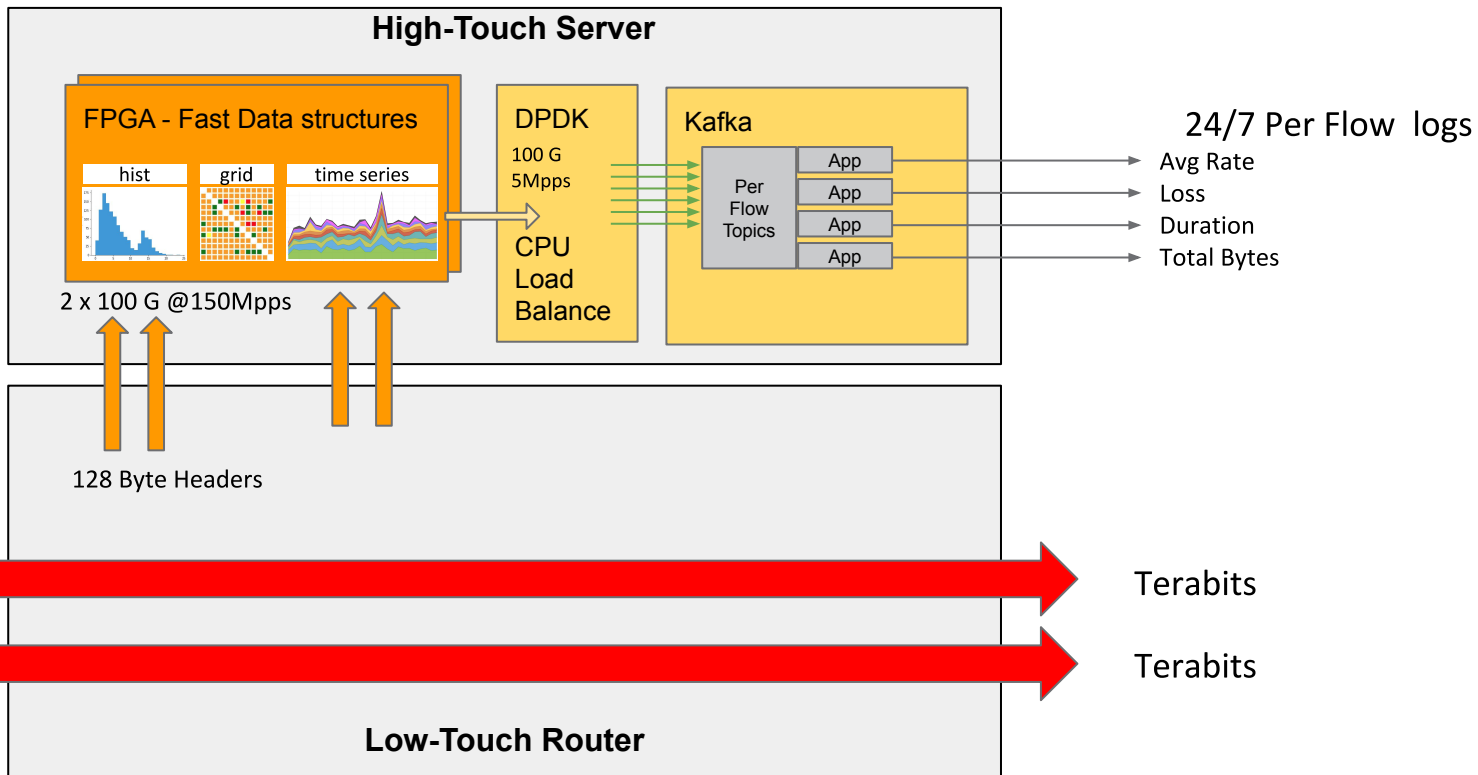


# ESnet6 High-Touch Architecture Overview

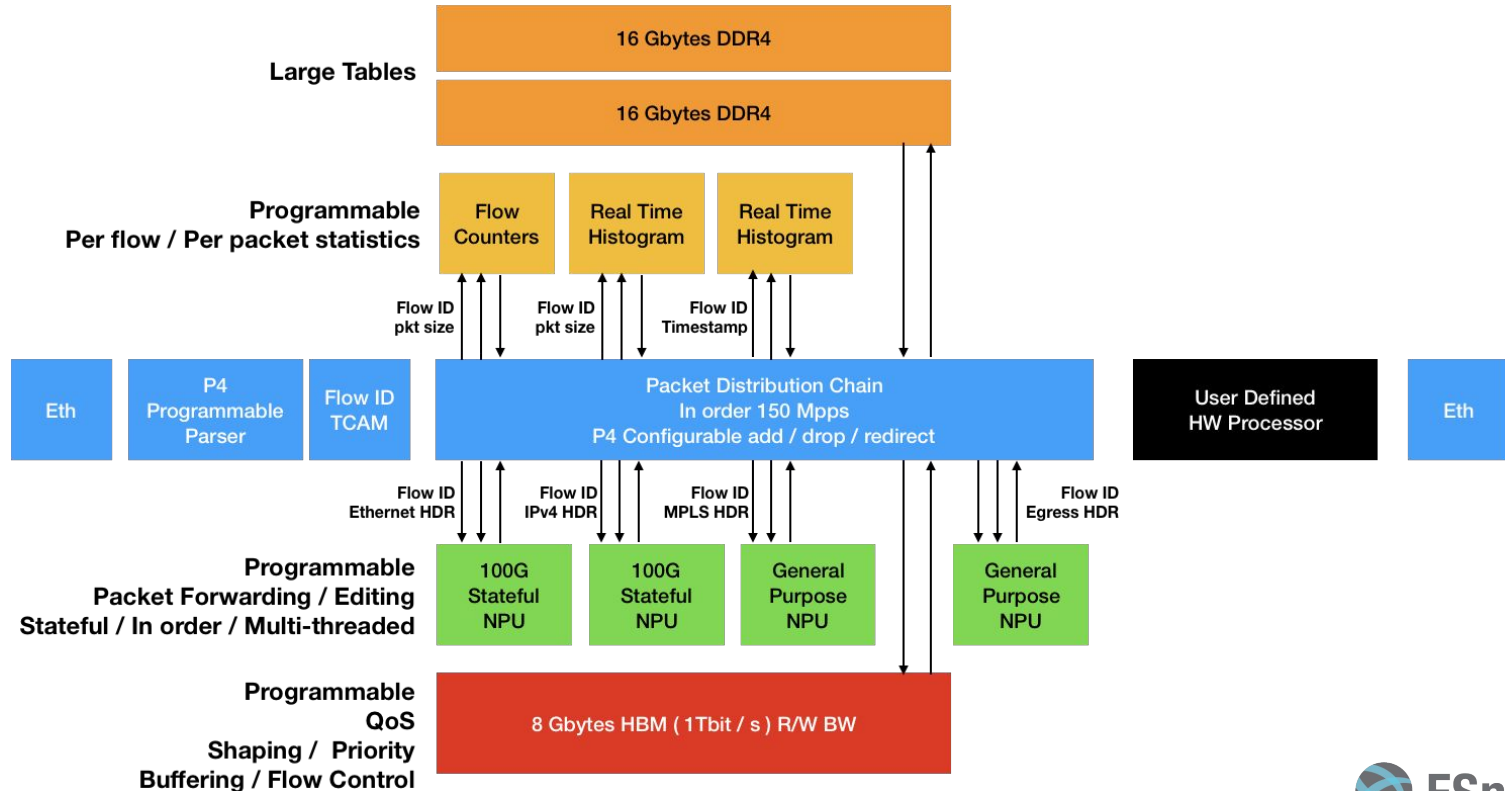


1. Mirror Service - Allows selective flows in the dataplane to be duplicated and sent to the FPGA for processing.
2. Programmable Dataplane (DP) - Appends meta-data, timestamps and repackages packet for transmission to Platform code.
3. Telemetry Data L2VPN or L3VPN(IPv6) - Connect Dataplane and Platform, possibly on different High-Touch (Local and Remote) Servers.
4. Platform - Reads telemetry packets from the network and distributes information to High Touch Services.
5. Remote Server - Hosts Platform components or Services (but not a Dataplane). Telemetry data can be directed to Remote Servers.
6. Services - Reads data from the Platform and performs real-time analysis as well as inserts selected telemetry data into database.

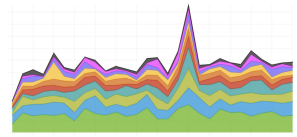
# Summary Of Rates



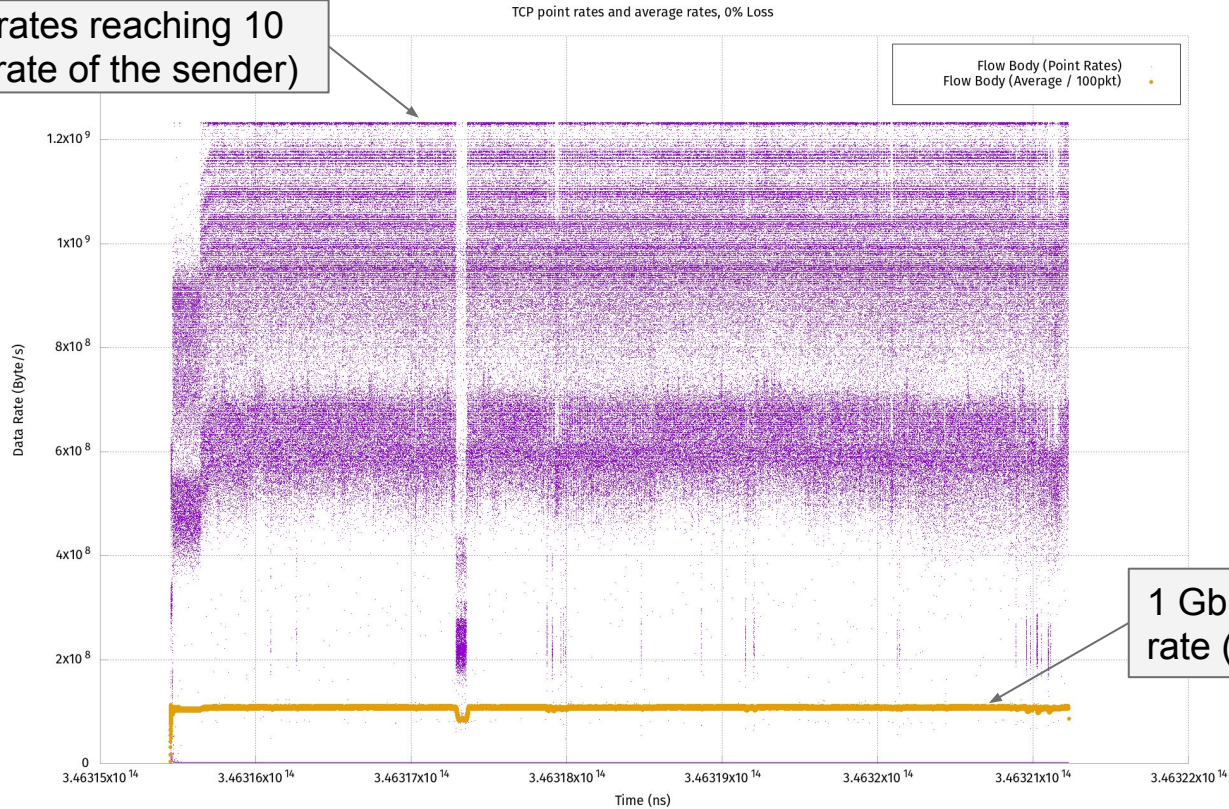
# ESnet FPGA Block Diagram - Present and Future



# 1 Gbps iPerf Flow - 600,000 Packets



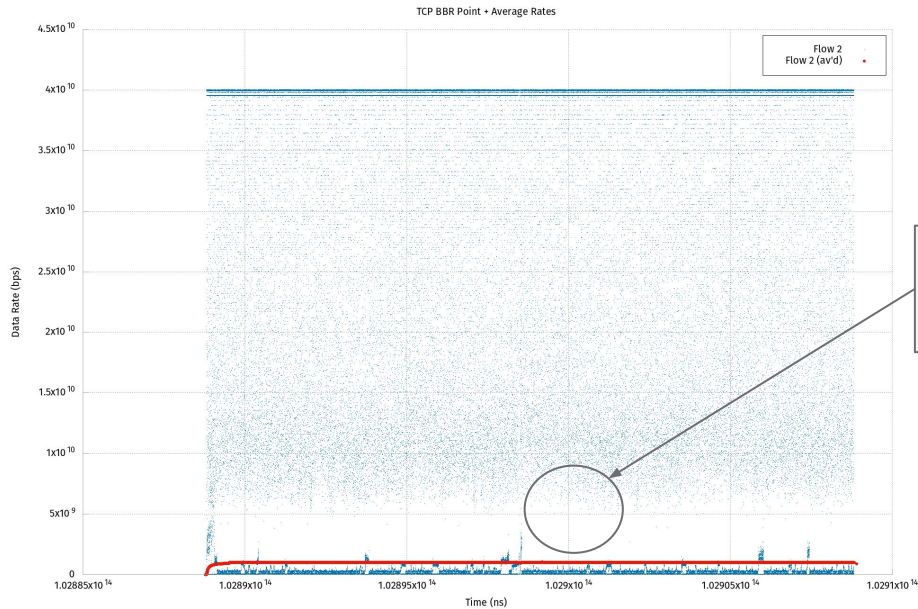
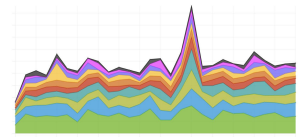
Per-packet rates reaching 10 Gbit/s (line rate of the sender)



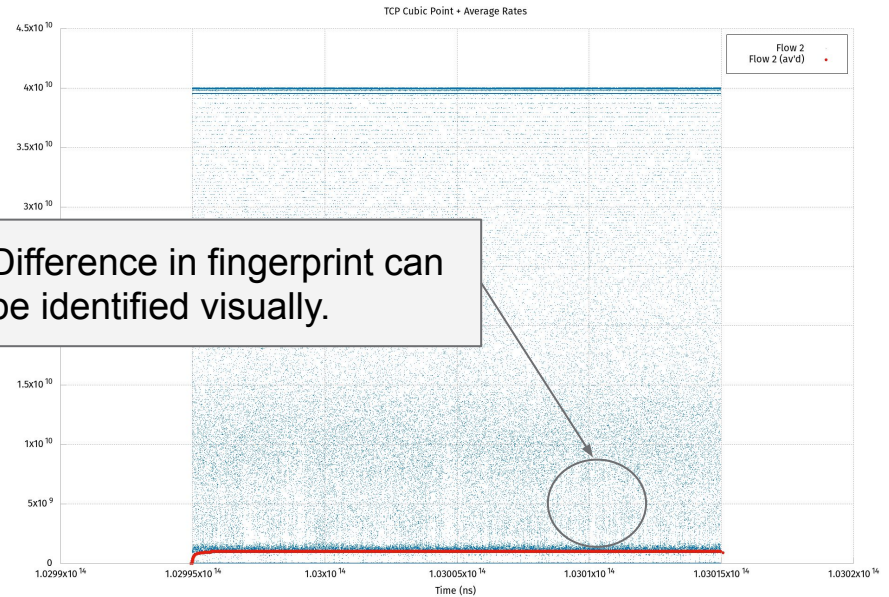
Note: Average rate is calculated using a time-weighted average of per-packet rates.



# BBR vs Cubic - Point Rates



TCP BBR

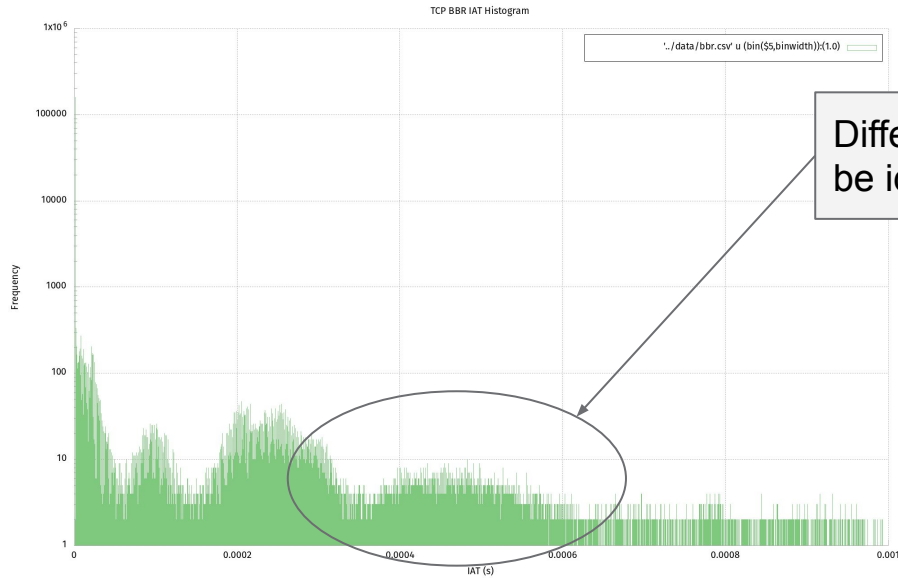
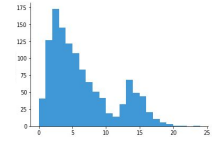


TCP Cubic

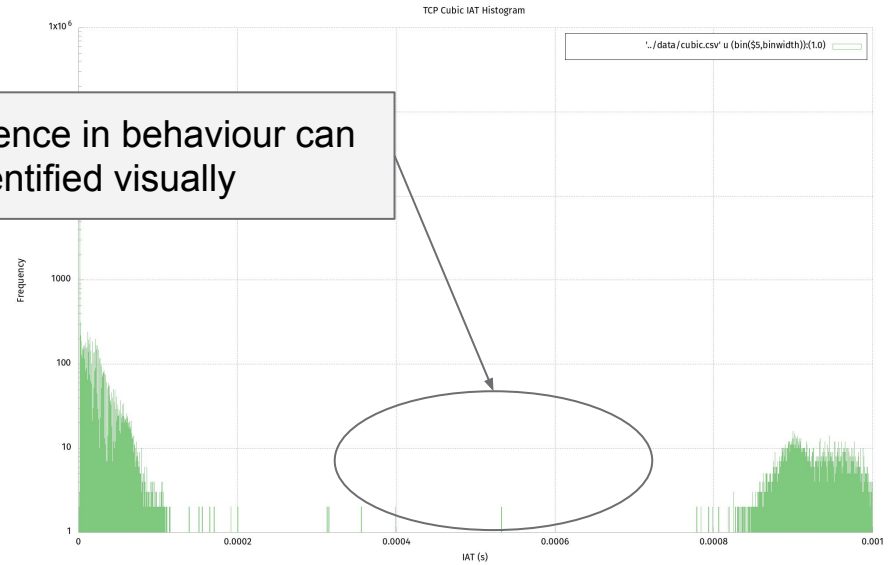
2 millions of data points shown (around 600.000 points a second generated)



# BBR vs Cubic - Inter-Arrival Time Histogram



TCP BBR (delay-based)



TCP Cubic (loss-based)

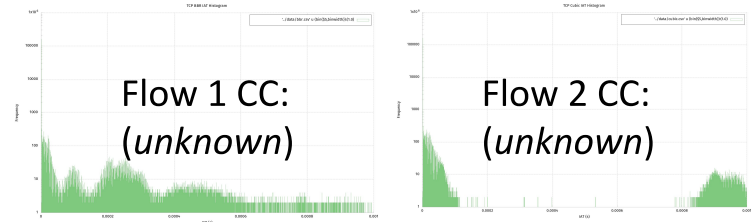
Difference in behaviour can be identified visually

BBR: inter-packet timing is more widespread than other congestion control algorithms.



# Machine Learning on Aggregated Data

- Aggregated data - such as histograms can be used to tell apart congestion control (CC) used by TCP flows
- We are using data plane histograms of inter-arrival times per flow (2000 packets per histogram)
- ML algorithms explored: Convolutional Neural Networks, k-Nearest Neighbors



Input: per-flow histograms  
of Inter-Arrival Time (IAT)

**Machine Learning**  
(trained with labeled data)

*Inference in less than 1 ms in all cases*

Flow 1 CC:  
most likely  
*TCP BBR*

Flow 2 CC:  
most likely  
*TCP RENO*

*More details, dataplane architecture, ML code in:*

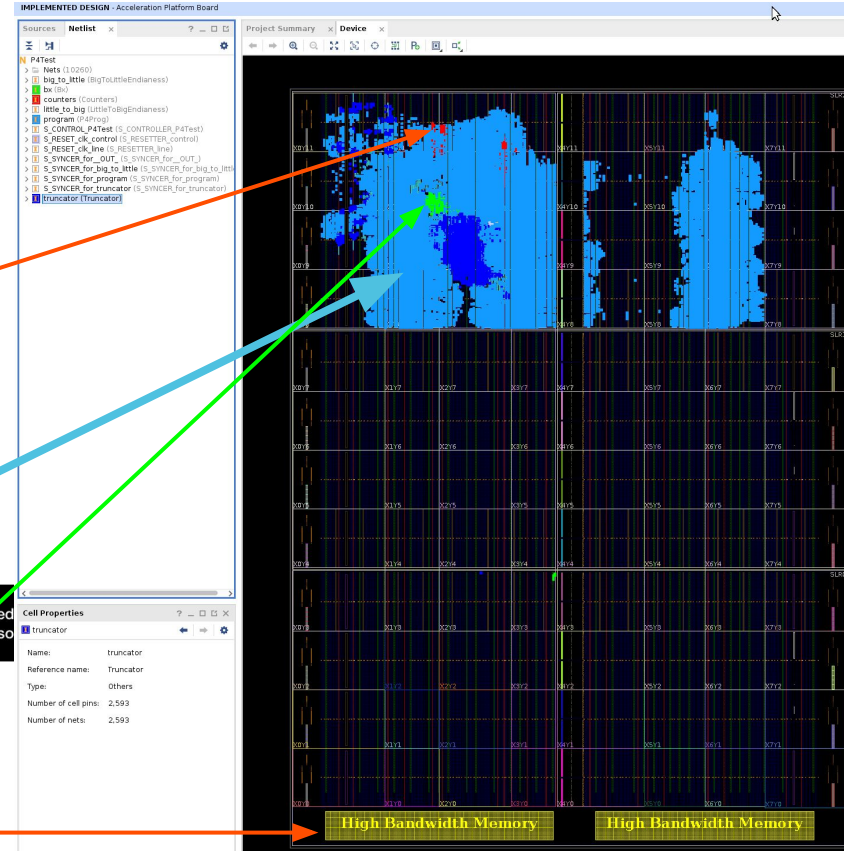
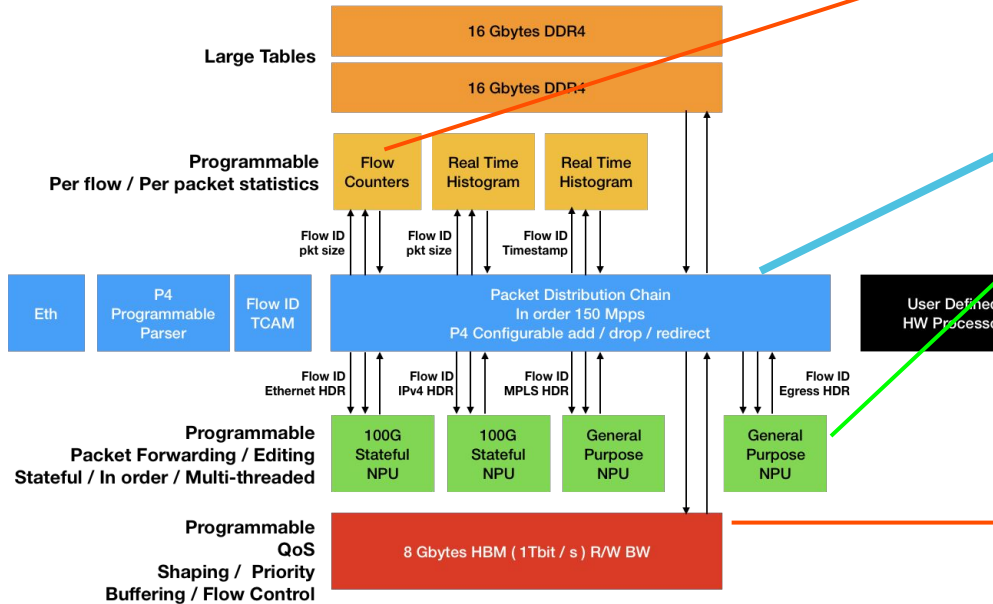
Simpson, Kyle A., Richard Cziva, and Dimitrios P. Pezaros.  
"Seiðr: Dataplane Assisted Flow Classification Using ML."  
IEEE GLOBECOM, Taipei, Taiwan (2020).



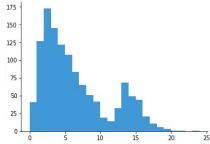


# FPGA Utilization

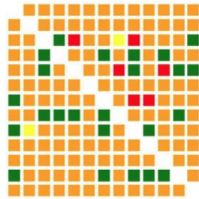
1. Plenty of room for expansion ( 10 % utilized )
2. P4 pipeline and TCAM are huge compared to NPU and counters.



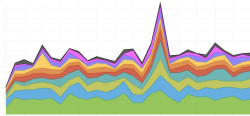
# Summary of techniques for rate reduction :



Per Flow Histograms - Counters / Sketches -- These are **polled** at Kafka friendly rates



Per Flow Single Facts - Total Bytes / (src,dst pairs ) / Duration  
Loss / Average Rate -- These are **polled** at Kafka friendly rates



Per Flow Reduced Time Series - Moving Ave Rate / 1:1 Bad Flows / Inter Arrival Time  
FFT -- These are **streamed** at Kafka friendly rates

All structures implemented in the FPGA at full line rate

Questions ?