



FPGA-Enhanced Real-Time Analysis of Network Traffic

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Outline

- ❑ **Motivation**
 - Common Network Traffic Measurement Tasks
 - Challenges
- ❑ **General Backgrounds**
 - Sketch-based Data Stream Algorithms
 - Flow-size/Flow-length Distribution
- ❑ **High-Speed Real-Time Processing**
 - Estimating Weibull Model Parameters
 - CAIDA 2007 DDoS
 - Witty Worm
- ❑ **Summary**
- ❑ **Q&A**

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Common Measurement Tasks

- ❑ **Heavy Hitter Detection**
- ❑ **Shannon Entropy**
 - One of the effective methods for Network Anomaly Detection
- ❑ **Flow Size/Length Distribution Estimation plays a fundamental role in**
 - Understanding network behavior
 - Optimizing performance
 - Managing resources, and
 - Detecting anomalies

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Bohan Zhao, Xiang Li, Boyu Tian, Zhiyu Mei, and Wenfei Wu. 2021. DHS: Adaptive Memory Layout Organization of Sketch Slots for Fast and Accurate Data Stream Processing. In *Proceedings of the 27th ACM SIGKDD Conference on Knowledge Discovery & Data Mining*. ACM, Virtual Event Singapore, 2285 – 2293. <https://doi.org/10.1145/3447548.3467353>

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Challenges

- ❑ **Packet arrives at a rapid rate**
- ❑ **Per-Flow Traffic Measurement is challenging**
 - Memory space
 - Processing time
- ❑ **Key space**
 - IPv4 address of 32-bit
 - High distinct number of flows
- ❑ **How to process and compute statistics on packet streams in real-time?**
- ❑ **Typical Implementation**
 - Software-based
 - Off-line approaches

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A. K. Marnerides, A. Schaeffer-Filho, and A. Mauthe, "Traffic Anomaly Diagnosis in Internet Backbone Networks," *Computer. Networks*, vol. 73, no. C, pp. 224–243, Nov. 2014.

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Sketch-based Data Stream Algorithms

- ❑ **A probabilistic data structure**
- ❑ **"Processing a long stream of data items in one pass using small memory space" [Kumar et al., IMC 03]**
- ❑ **Widely used in networking and data stream processing**
 - Capable of summarizing large amounts of data
 - Update/Query operations in a memory-efficient manner
- ❑ **Especially useful in scenarios of**
 - Infeasible to store all data due to resource constraints
 - High-speed Network Monitoring
 - Anomaly Detection, and Traffic Engineering

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Flow Size/Length

- ❑ **Flow**
 - Flow is defined by selected attributes of layer 3-4 packet header.
 - For example, by 5-tuple of:
{IP_Src, IP_Dest, TCP_SrcP, TCP_DestP, Proto}
- ❑ **Flow Size**
 - The total amount of data transferred within a flow
 - In terms of the number of **bytes**
- ❑ **Flow Length**
 - The total amount of **packets** transferred within a flow

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Security and Anomaly Detection

- ❑ **Sudden Increase in Large/Small Flows**
- ❑ **Changes in Flow Length Distribution**
 - DDoS attacks
 - Attackers generate large numbers of short flows to overload web servers or services
 - Network scanning
 - Numerous short flows could be a sign of network scanning
- ❑ **Changes in Flow Size Distribution**
 - Some malware or Worms may generate large flows
 - Attackers overwhelm network bandwidth with large amounts of data

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- ❑ **Can we do high-level coarse-grained estimation to track Changes of Flow Length and Flow Size?**
- ❑ **At wire-speed of 100Gbps and beyond?**



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Statistical Model of Flow Size/Length Distribution

- **Highly-skew, Long-tail distribution**
 - Lognormal, Pareto and Weibull are commonly used for the long-tail network traffic modeling [2] [24]
- **Due to the complexity of Internet traffic, single distribution can not be fitted accurately [26]**
- **Weibull distribution is a flexible model**
 - Can accommodate various shapes
 - Including the exponential and Rayleigh distributions as special cases
 - Tracking the change of the parameter
 - Real-time estimation of Weibull shape parameter is useful in traffic monitoring and anomaly detection [4]

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**Weibull and Log-Normal Model for
Flow Length/Size Distribution**

PARAMETER ESTIMATION

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Weibull Model

- ❑ The probability density function of a Weibull random variable
 - α : Shape
 - β : Scale
 - δ : Location

$$f(x) = \frac{\alpha}{\beta} \left(\frac{x - \delta}{\beta} \right)^{\alpha-1} e^{-((x-\delta)/\beta)^\alpha}$$

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Method of Moments

- ❑ A technique for estimating the parameters of statistical distributions
- ❑ By equating the *population moments* to the *sample moments*
- ❑ Particularly useful for distribution models of
 - Weibull
 - Log-Normal

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Estimating the First Moment m_1

- Take the **Flow Length** as an example:
- Assuming the **flow id** is the source IP address
- Packet counts of flows are updated in the counter of **$C[d][w]$**
- The total packets count of all IP flows is the total sum of each row

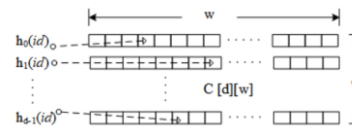
$$\sum_{id=0}^{n-1} x_{id} = \sum_{i=0}^{w-1} C[0][i]$$

- The first frequency moment **F_1**

$$F_1 = \sum_{i=0}^{w-1} C[0][i]$$

$$\frac{m_1}{\sqrt{(cm_2)}} = \frac{\Gamma(1 + \frac{1}{\alpha})}{\Gamma(1 + \frac{2}{\alpha}) - \Gamma^2[1 + \frac{1}{\alpha}]}$$

$$m_1 = \mu = \frac{1}{n} \sum_{id=0}^{n-1} x_{id}$$



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Zeroth Frequency Moment F_0

$$m_1 = \mu = \frac{1}{n} \sum_{id=0}^{n-1} x_{id}$$

- The distinct count of the observed flows: **n**
- Can be estimated by using the sketch-based algorithms

$$n \approx \hat{F}_0$$

- FM sketch [#flajolet_probabilistic_1983]
- HyperLogLog [#flajolet_hyperloglog_2007]

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Estimation of the Sample Moments

- Get a function with only the shape parameter α

$$\frac{m_1}{\sqrt{cm_2}} = \frac{\Gamma(1 + \frac{1}{\alpha})}{\Gamma(1 + \frac{2}{\alpha}) - \Gamma^2[1 + \frac{1}{\alpha}]}$$

- The first moment about the origin

$$\hat{m}_1 = \frac{\hat{F}_1}{\hat{F}_0}$$

- The second moment about the mean

$$cm_2 = \frac{\hat{F}_2}{\hat{F}_0} - \left(\frac{\hat{F}_1}{\hat{F}_0}\right)^2$$

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FPGA Functional Simulations

- Count-Min Sketch
 - cm_d = 7
 - cm_w = 65536
- FM Sketch PCSA
 - fm_n = 512
 - fm_bits = 32
- Trace: MAWI, 200403201400.dump.gz

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200403201400.dump.gz_obsT10_flow_length_slp_2024-08-18

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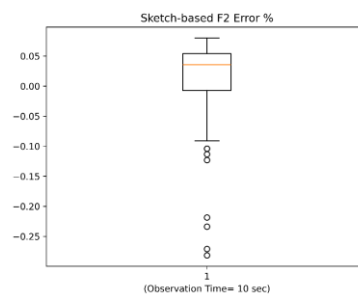
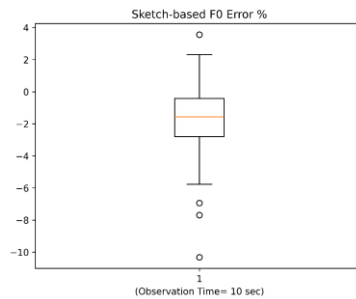


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Relative Error Estimated F_0 & F_2

□ F_0

- Median: -2%
- Q1-Q3: ~ -3%



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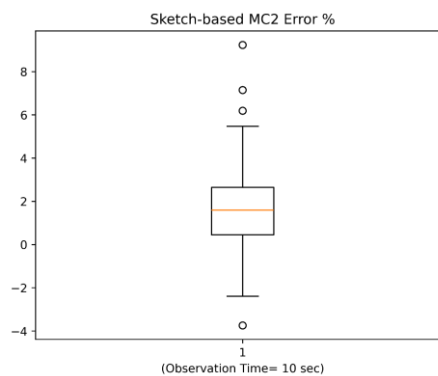


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Relative Error Second Moment about the Mean

□ MC2

- Median: 1.8%
- Q1-Q3: ~ 3%



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Synthetic DDoS Network Traffic

❑ CAIDA+MAWI

❑ CAIDA 2007 DDoS Trace (Attacking):

- The attacker sent many ICMP Echo request packets to the Victim
- Randomly created source IP addresses
- Four DDoS attacking traffic are selected from the CAIDA 2007 DDoS trace (to-victim).
 - 20070804_140436.pcap
 - 20070804_140936.pcap
 - 20070804_141436.pcap
 - 20070804_141936.pcap

❑ MAWI DITL 2019 Trace (Background)

- <https://mawi.wide.ad.jp/mawi/ditl/ditl2019/201904091800.html>

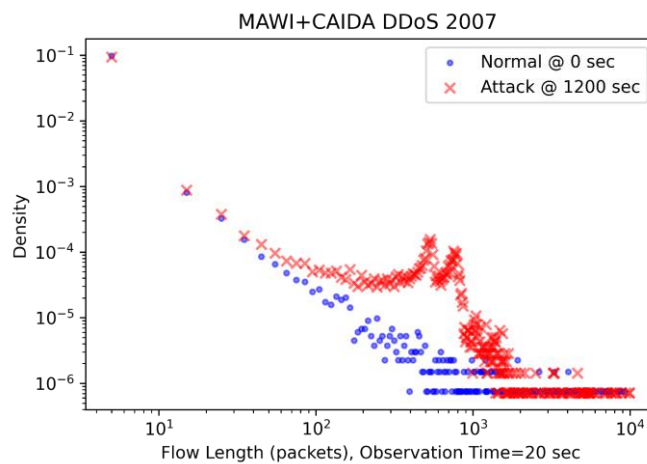
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CAIDA 2007 DDoS Flow Length Distribution



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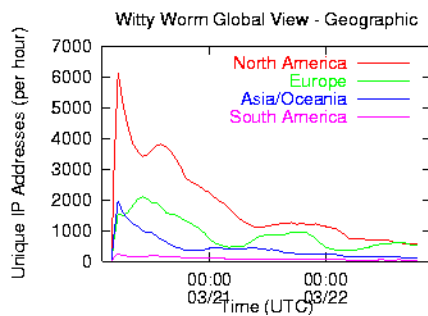
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Witty Worm Outbreak

- **March 19, 2004, Approximately at 8:45pm PST**
 - UTC time at 3:45 AM on March 20, 2004
- **Infected 12,000 machines in 45 mins**
- **Targeting several Internet Security Systems (ISS) products**
 - Buffer overflow vulnerability



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<https://www.caida.org/archive/witty/>

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Witty Worm Characteristics

- ❑ **The infecting host sends**
 - 20,000 **UDP** packets
 - Random destination IP address
 - Random size between 796 and 1307 bytes
 - from **source port 4000**

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MAWI 2004 Data Set Witty Worm

- ❑ **Outbreak begins at UTC Time 3:45 AM on March 20, 2004**
- ❑ **Captured by the daily 15-min traces**

MAWI Trace 2004	UTC Time 2004	
200403181400.dump.gz	03/18 , 05:00:01 AM	
200403191400.dump.gz	03/19 , 05:00:01 AM	
200403201400.dump.gz	03/20 , 05:00:01 AM	*****
200403211400.dump.gz	03/21 , 05:00:01 AM	****
200403221400.dump.gz	03/22 , 05:00:01 AM	**
200403231400.dump.gz	03/23 , 05:00:01 AM	*
200403241400.dump.gz	03/24 , 05:00:01 AM	*
200403251400.dump.gz	03/25 , 05:00:01 AM	
200403261400.dump.gz	03/26 , 05:00:01 AM	

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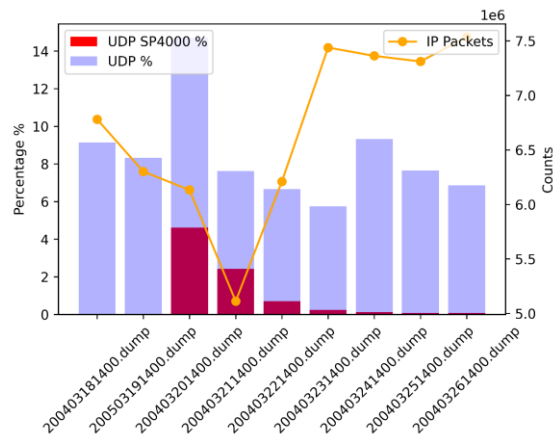
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MAWI Traces

- During the Witty Worm Outbreak
- ~5% UDP Source Port 4000 Packets



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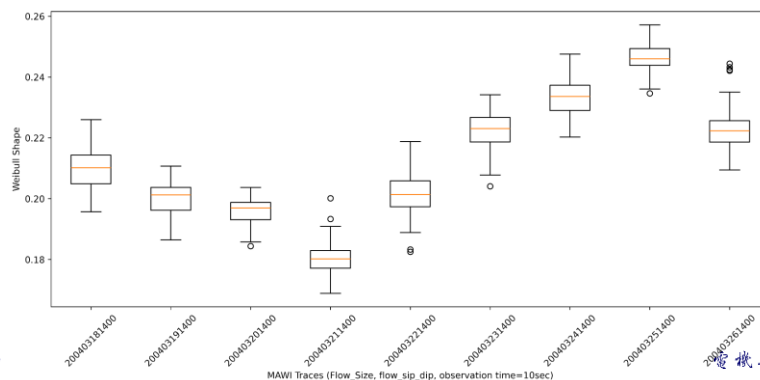
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Estimated Weibull Shape

- Shape Parameter Decreases
 - The tail of the distribution becomes heavier (longer)
 - Variability in the data increases, meaning there's less predictability in the size of any given flow



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MAWI Traces (Flow_Size, flow_sip_dip, observation time=10sec)

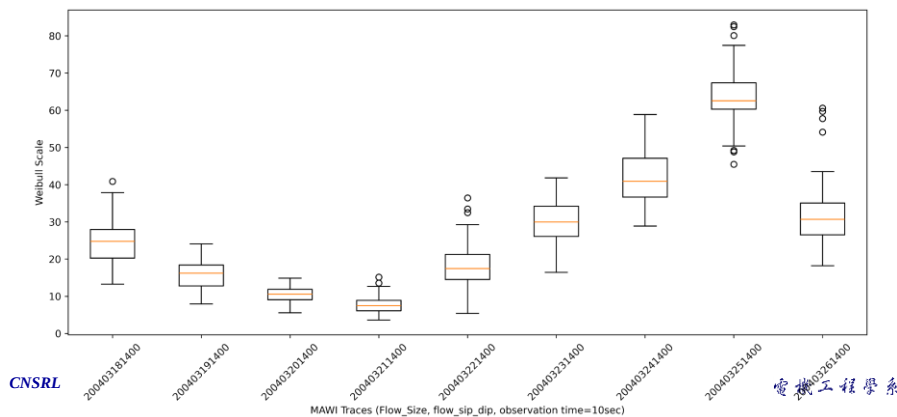


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Estimated Weibull Scale

Scale Parameter Decreases

- The distribution becomes more concentrated near zero
- The peak of the distribution shifts towards the origin



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Sketch-based FPGA Implementation

F_2 Estimation

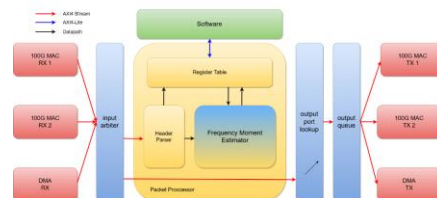
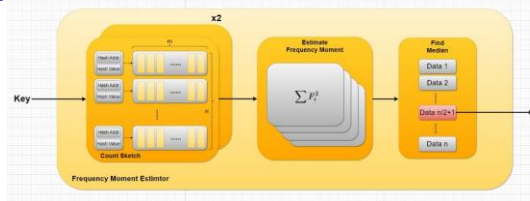
- AMS-Cormode Sketch base on Count-sketch [5]

F_0 Estimation

- FM Sketch, Probabilistic Counting with Stochastic Averaging (PCSA) [6]

NetFPGA PLUS Framework

Xilinx Alveo U200 FPGA



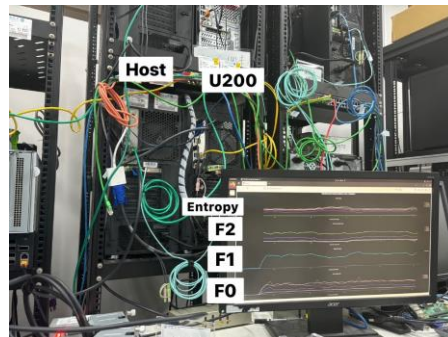
CNSRI [5] Graham Cormode and Marios Hadjieleftheriou. 2009. Finding the frequent items in streams of data. *Commun. ACM* 52, 10 (2009), 97–105. <https://doi.org/10.1145/1562764.1562789>

[6] Philippe Flajolet and G. Nigel Martin. 1985. Probabilistic counting algorithms for data base applications. *J. Comput. Syst. Sci.* 31, 2 (1985), 182–209. <http://portal.acm.org/citation.cfm?id=5215>

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Xilinx FPGA Demo

- ❑ **Replays the synthetic trace from a 2-port 100Gbps NIC**
- ❑ **Observation time of 30 seconds**



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Summary

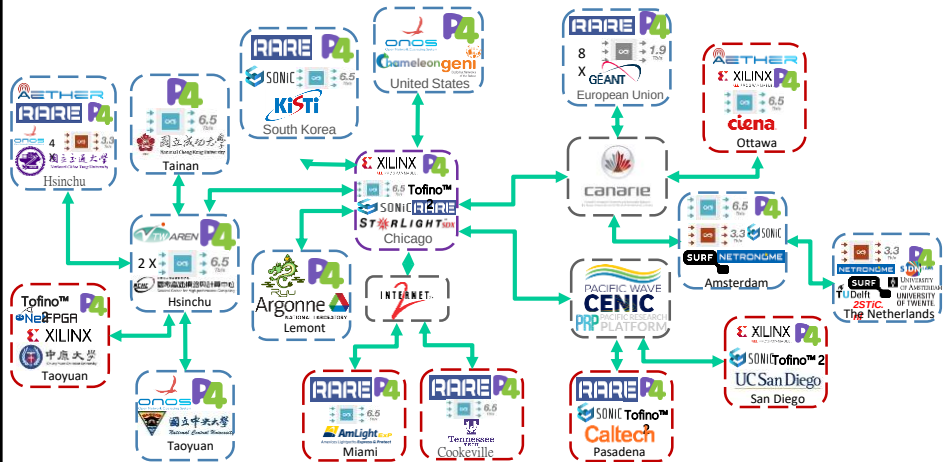
- ❑ **Real-time online processing**
 - exceeding 100Gbps line speed
- ❑ **For less than 132K bytes of memory space**
 - Count Sketch: 128K Bytes
 - FM Sketch PCSA: 4K Bytes
- ❑ **Flow-Length/ Flow-Size Distribution**
 - Weibull Parameter Estimation
 - Log-Normal Parameter Estimation
- ❑ **Sketch-based Implementation**
 - Xilinx Alveo U200 FPGA /w NetFPGA PLUS Framework
- ❑ **Suitable to implement on the high-speed network equipment**



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International P4 Experimental Networks (iP4EN)



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