

A change detection primitive for the network data plane

Gonçalo Matos, Salvatore Signorello, Fernando M. V. Ramos









Traffic changes might indicate an anomaly in the network.



Traffic changes might indicate an **anomaly** in the network.

Their timely detection is of paramount importance to network operation!



Traditional change detection

Sampling via Flow-based measurements:

- Standard, e.g., Netflow and sFlow
- Sampling \rightarrow CPU, memory, bandwidth
- Low sampling rate \rightarrow Low Fidelity

Based on **fixed-function equipment** that is **difficult to manage** and **to configure**.



Networks are changing!

Software Defined Networking (SDN):

- > Separation of the network planes.
- Programmability of each plane.



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Programmable Switching Chips:

Detection of traffic changes inside the data plane.



Some solutions already enable change detection using P4 to program the data plane.

However, these are often heavy-hitter based.

But sometimes the problem is with smaller flows!

Approach	Tbps performance	Heavy-Hitter Change Detection	Generic Change Detection
Hashpipe ^[1]	Х	Х	
Elastic Sketch ^[2]	X	X	
SketchLearn ^[3]	X	X	
K-ary Sketch ^[4]		Х	Х

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Contributions

- The design of K-meleon, an online change detection system that leverages programmable switches.
- The implementation of a **prototype in P4** for Tofino switch.
- An evaluation using the software switch bmv2 that demonstrates K-meleon achieves the same level of accuracy as the k-ary algorithm.





































with Count-min











K-ary Operations

The forecasting module uses forecasting models, such as the EWMA:

$$S_{f}(t) = \begin{cases} \alpha \cdot S_{o}(t-1) + (1-\alpha) \cdot S_{f}(t-1), & t > 2\\ S_{o}(1), & t = 2 \end{cases}$$

Floating-point multiplications

The change detection module computes **estimates** on top of the error sketch:



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Programming the data plane



Protocol-Independent Switch Architecture (PISA)

Memory and budget of operations are limited!

K-ary algorithm - limitations

Performs batch-based operations at the end of each epoch, thus;

- Does not fit the **constraints of the data plane** programming model;
- Performs **complex operations** not supported in P4;

The k-meleon

Stream-based approach.

➢Performs operations incrementally with each packet.

Sends changes to the controller ad hoc.



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Evaluation





Q1. How does k-meleon perform compared to k-ary?

Q2. What is the expected performance of k-meleon?

Q3. What is the resource usage of k-meleon?

2

Experimental setup

The evaluation of k-meleon uses several packet traces from two datasets:

CSE-CIC-IDS2018^[5] \rightarrow Network attacks.

Data-center measurement (UN1)^[6] \rightarrow Microburst events.

Version	Purpose
K-ary in python 🛛 🥐	Establish a baseline
K-meleon in P4 for bmv2 🄁	Verify the correctness
K-meleon in P4 for TNA 🏼 🗛	Final target - Evaluation

Q1. K-meleon vs. k-ary

We vary the **size of the sketch**, which affects the number of complex computations performed.

Median relative difference is very low for any value of H (<0.2%).



Relative Difference (%)

Q2. Performance

Our P4 prototype for TNA compiled successfully.

Thus, it should run at line-rate, sustaining Tbps net traffic.



However, we have not yet moved

all computations to the data plane.

Q3. Resource Usage on TNA

Only 7 stages are used by our preliminary prototype in P4 for TNA.

Only ingress is used for performing computation.

Computing	Usage
Stages	58%
Meter ALU	16.7%
Hash Dist Units	13.9%
VLIWs	4.7%
Memory	Usage
SRAM	2%
TCAM	0%

Conclusions

- The design of **K-meleon**, an *online change detection system* that leverages programmable switches.
- The implementation of a prototype in P4 for Tofino Switch.
- An evaluation using the software switch bmv2 that demonstrates K-meleon achieves the same level of accuracy as the k-ary algorithm.

Discussion and future work

Detection entirely in the data plane



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Discussion and future work

Use Cases



The tuning of some parameters improves detection accuracy of our primitive for different use cases.

Discussion and future work

Reversibility





Thank You

Questions?

References

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