



A Hybrid P4/NFV Architecture for Cloud Gaming Traffic Detection with Unsupervised ML

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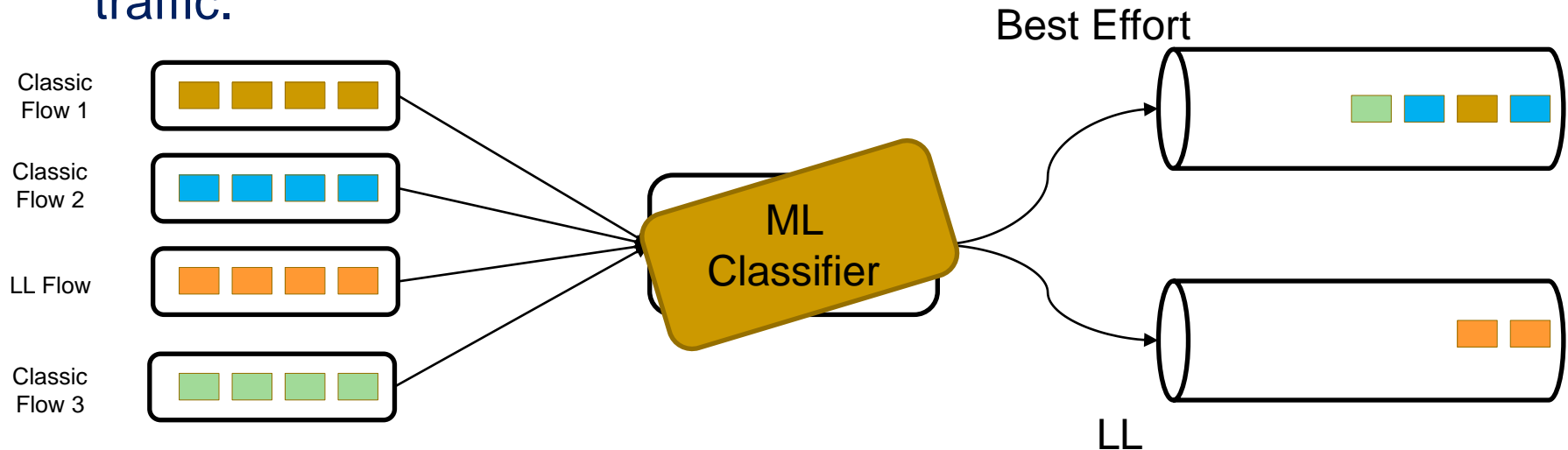


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1. Context & Motivation

- Increasing pressure of Cloud Gaming (CG) traffic on current network infrastructures.
- Stringent network requirements, especially in terms of **jitter and latency**.
- Recent network technologies such as **L4S** (Low Latency Low Loss Scalable throughput) can reduce the latency of low-latency (LL) traffic.



2. Previous works

- **Cloud Gaming traffic** from 4 main CG platforms available in Europe (Stadia (STD), GeForceNow (GFN), Xbox Cloud (XC) and PlayStationNow (PSN)):
 - Traffic on **normal conditions**
 - Traffic with **network constraints**
- **Non-Cloud Gaming Traffic:** (Video conferencing (VC); Video streaming (VS); Live video streaming (LV); Facebook navigation (FN))
- **New datasets:** (new games and new CG platforms (Moonlight, Steam))

- Mean packet size
- Average Inter Arrival Times (IAT)
- Total number of packets
- Standard deviation of packet size
- Standard deviation of IATs
- Total size of application data

- Decision Tree (DT) model (supervised ML) in a set of VNF to detect CG traffic [Graff et al.].

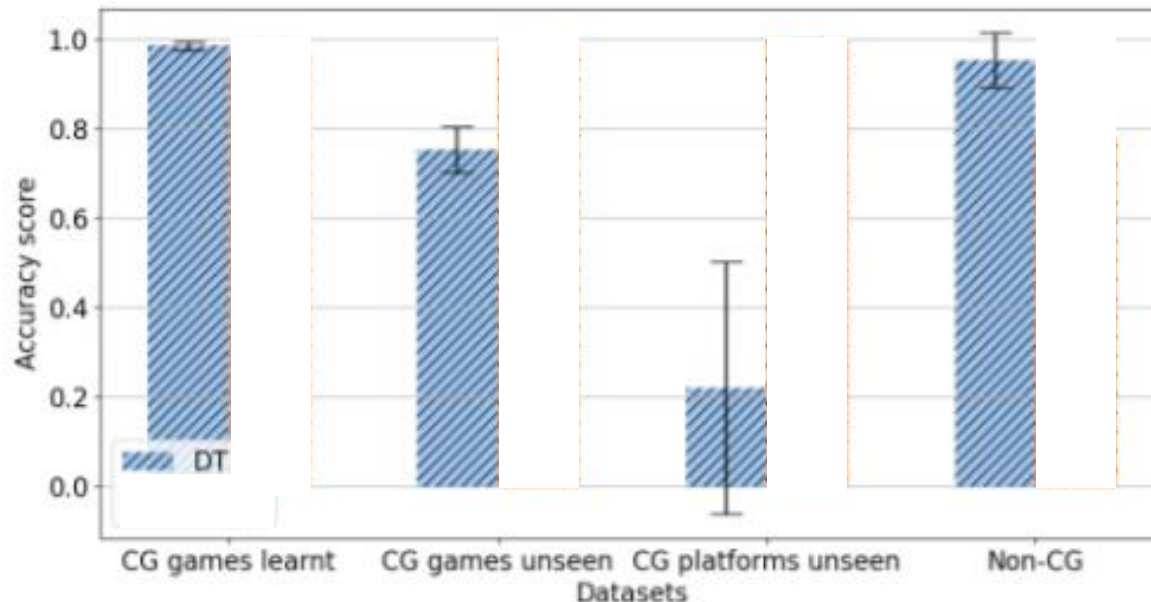
Performance degradations with new CG traffic (« unseen » CG traffic or new CG platforms).

12 network traffic features (computed for uplink and downlink traffic) for a window traffic of 33ms

3. ML-based CG Traffic Detection

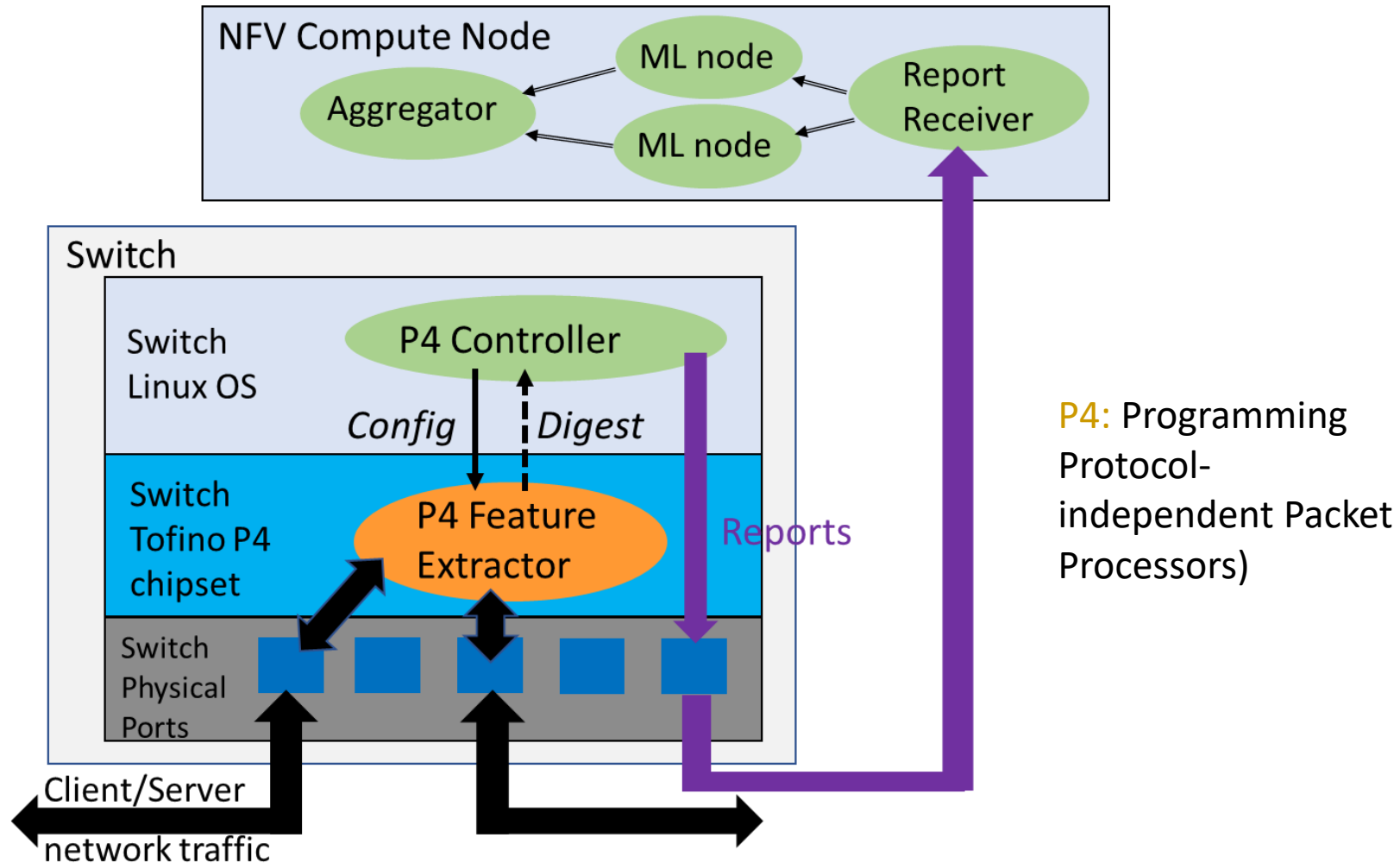


- ML models:
 - **DT** (Decision Tree): CG traffic on normal conditions and NCG traffic (with labels)
 - **USAD** (UnSupervised Anomaly Detection): CG traffic on normal conditions only (without labels) [Audibert et al.]:
- ML models Performance:



4. Hybrid P4/NFV Architecture

- 1 hardware P4 switch, 1 NFV compute node



5. Evaluation



■ Accuracy & F1-score for all traffic

Traffic	Type	Performance with P4		Performance with application		Difference (%)	
		Accuracy	F1	Accuracy	F1	Accuracy	F1
Normal CG	STD	0.965	0.983	0.984	0.992	-1.90	-0.97
	GFN	0.990	0.995	0.979	0.989	1.16	0.58
	XC	0.903	0.942	0.966	0.981	-6.35	-3.85
	PSN	0.981	0.990	0.966	0.983	1.46	0.76
	Overall	0.958 (± 0.054)	0.979 (± 0.03)	0.973 (± 0.021)	0.986 (± 0.011)	-1.44	-0.74
CG with network constraints	STD	0.999	0.999	1.000	1.000	-0.01	-0.01
	XC	0.954	0.977	0.992	0.996	3.33	1.55
	PSN	0.995	0.998	0.961	0.983	-3.73	-1.91
	Overall	0.983 (± 0.035)	0.993 (± 0.019)	0.984 (± 0.019)	0.992 (± 0.010)	-0.14	0.05
Non CG	VC	0.938	0.971	0.867	0.941	7.18	3.00
	LV	0.980	0.990	0.978	0.989	0.17	0.10
	VS	0.993	0.996	0.991	0.996	0.14	0.01
	FB	0.988	0.994	0.989	0.995	-0.10	-0.04
	Overall	0.959 (± 0.056)	0.983 (± 0.031)	0.918 (± 0.114)	0.970 (± 0.066)	4.13	1.30
New games CG learned platforms	GFN	0.973	0.986	0.995	0.997	-2.15	-1.10
	XC	0.969	0.984	0.979	0.989	-0.97	-0.50
	PSN	0.980	0.990	0.999	1.000	-1.97	-1.00
	Overall	0.974 (± 0.010)	0.987 (± 0.005)	0.991 (± 0.011)	0.996 (± 0.006)	-1.70	-0.84
New CG platforms	MoonLight	0.999	0.999	1.00	1.00	-0.01	-0.01
	Steam	0.999	0.999	1.00	1.00	-0.01	-0.01
	Overall	0.999 (± 0.00)	0.999 (± 0.00)	1.00 (± 0.00)	1.00 (± 0.00)	-0.01	-0.01

6. P4 Limitations

- Limited computation capabilities due to:
 - Limited number of **operations**
 - Variable **multiplication/division limited** => Standard deviation can not be computed

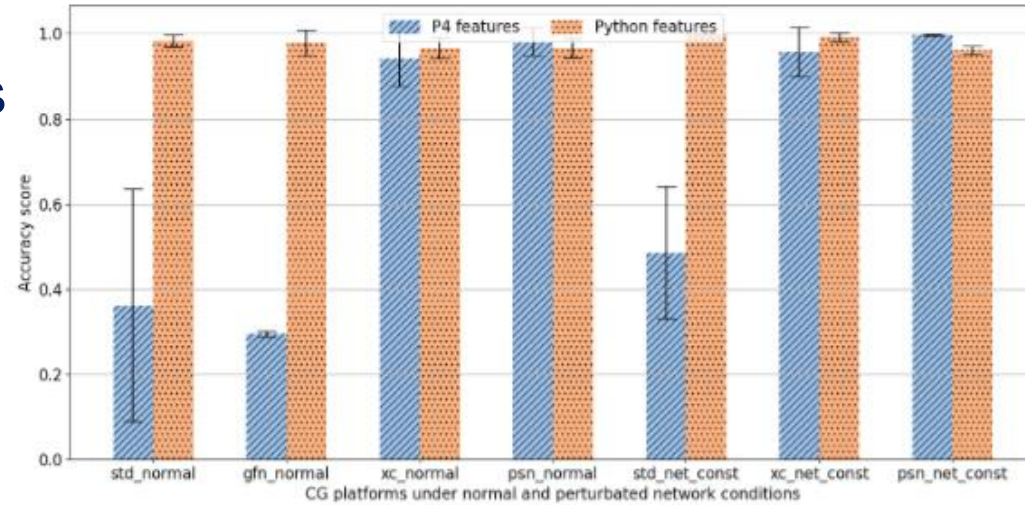
- Not a **packet buffering/copying capability** in P4
 - => Impossible to keep in memory packet features value to compute standard deviation

- Processing trigger upon packet arrival
 - => Reports are not sent exactly each 33ms because there is no timer
 - If sporadic traffic, empty reports need to be send to acknowledge the lack of traffic

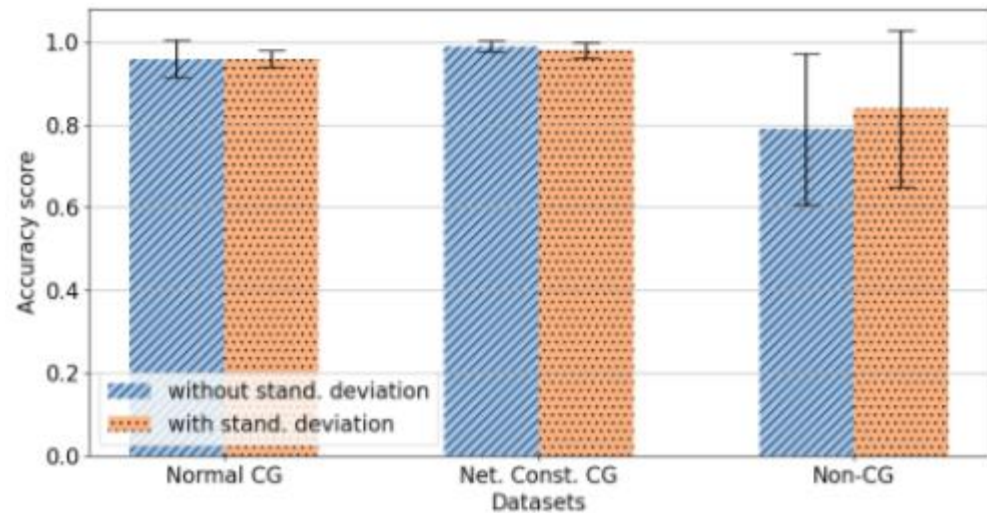
6. P4 Limitations



- Standard deviation approximation with previous mean value.



- Remove standard deviation from ML features



7. Conclusion

- CG traffic detection implementation on P4 hardware module demonstrates **excellent performance**.
- Lessons learned from P4 hardware implementation:
 - P4 not suitable for handling complex computational tasks.
 - Time event-based programs are not ideal for P4.
 - Code optimization is required with P4
- Trade-off between **high-speed line-rate packet processing** and ease of programming.
- Hybrid P4/NFV architecture is a promising approach to efficiently **split processing tasks**.
- Evaluation of the solution with operational traffic from network operators will be considered in future work.

Thank you

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<https://www.mosaico-project.org/>

https://github.com/mosaico-anr/P4_NFV_CG_Detector

8. References



- [Graff et al.]: Philippe Graff, Xavier Marchal, Thibault Cholez, Bertrand Mathieu, Olivier Festor. Efficient Identification of Cloud Gaming Traffic at the Edge. NOMS 2023 - 36th IEEE/IFIP Network Operations and Management Symposium, May 2023, Miami, United States. pp.10.
- [Audibert et al.]: Julien Audibert, Pietro Michiardi, Frédéric Guyard, Sébastien Marti, and Maria A. Zuluaga. 2020. USAD: UnSupervised Anomaly Detection on Multivariate Time Series. In Proceedings of the 26th ACM SIGKDD International Conference on Knowledge Discovery & Data Mining (KDD '20). Association for Computing Machinery, New York, NY, USA, 3395–3404.

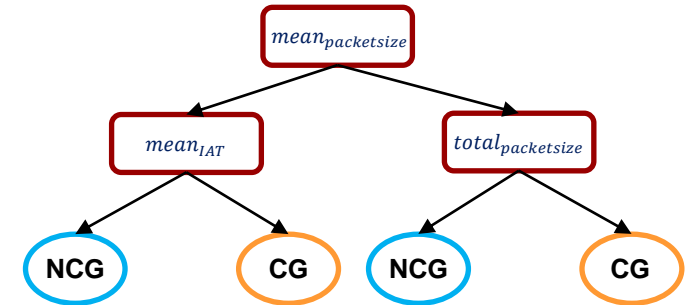
Appendix



A-1. ML Models



■ Decision Tree (DT):



■ UnSupervised Anomaly Detection (USAD) [Audibert et al.]:

