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# Henna: Hierarchical Machine Learning Inference in Programmable Switches

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[Developing the  
Science of Networks]

P4

+



Programmable  
High Throughput  
Low Latency



## Applications

Sketches

Flow Monitoring

Traffic Management

Routing and Forwarding

Service Function Chaining

Time-Sensitive Networking

In-Band Network Telemetry

Network Function Virtualization

Intrusion Detection Systems

DDoS Attack Mitigation

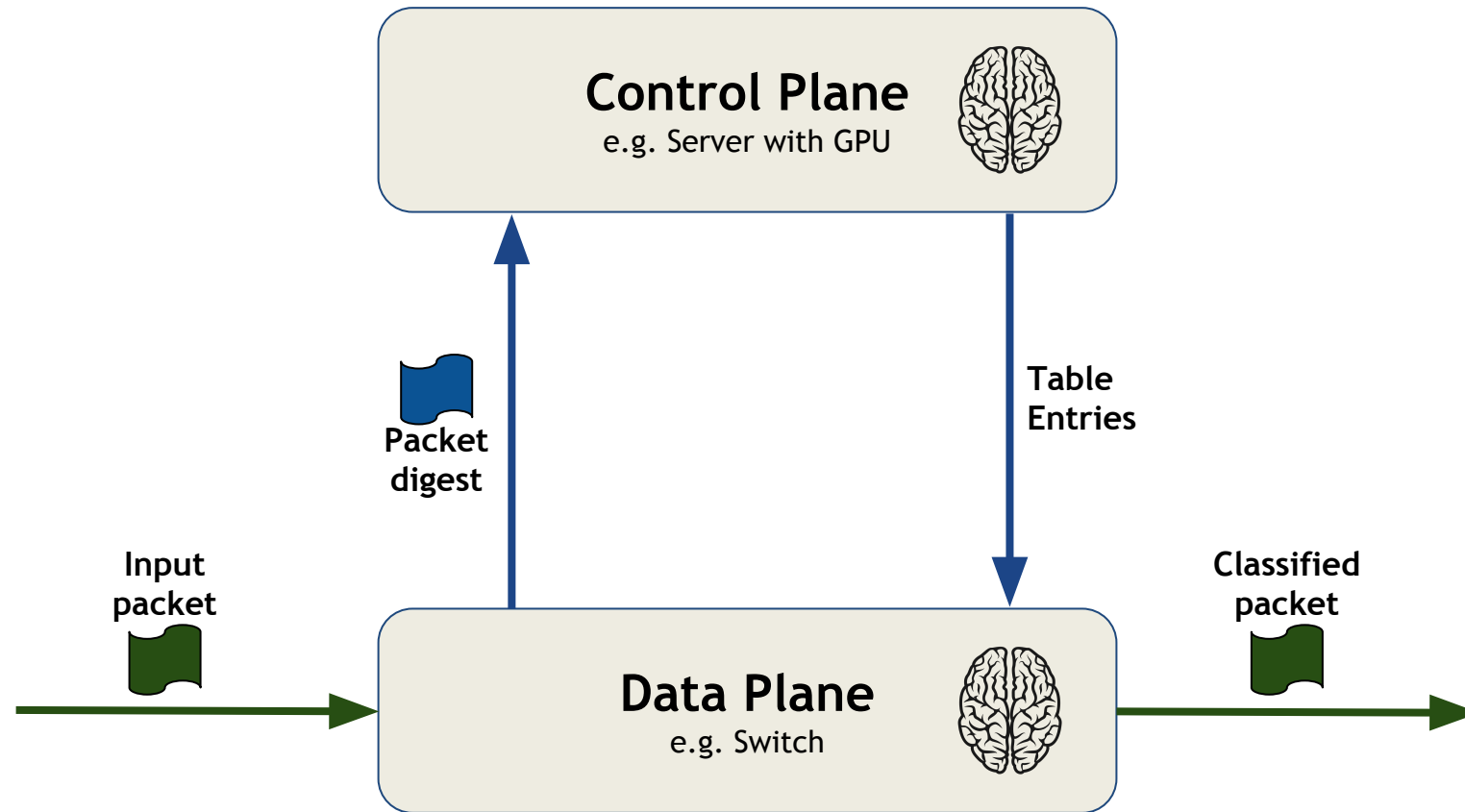
Line-rate ML Inference

etc.

# Machine Learning Inference in the Data Plane

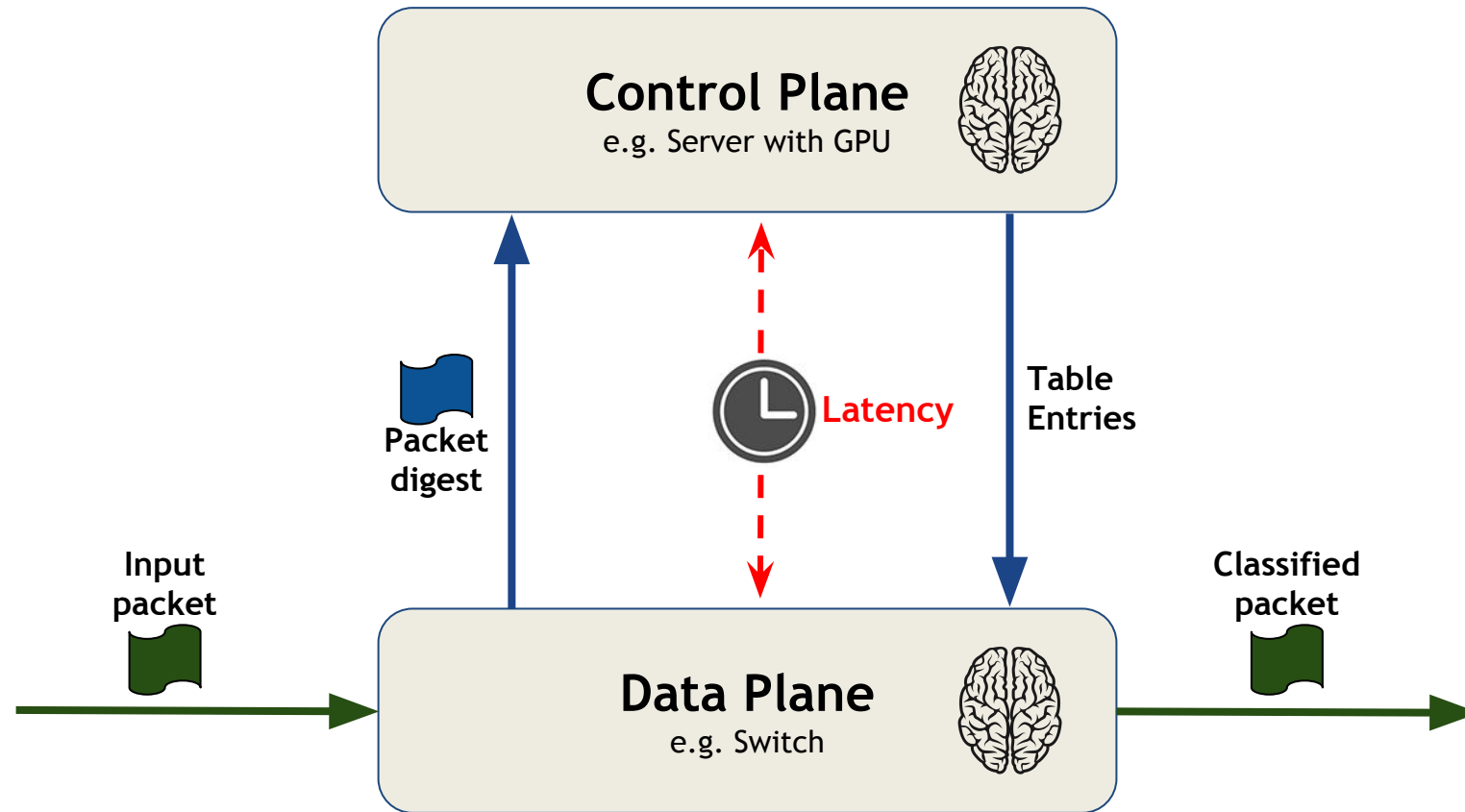


# Why Data Plane Machine Learning Inference?



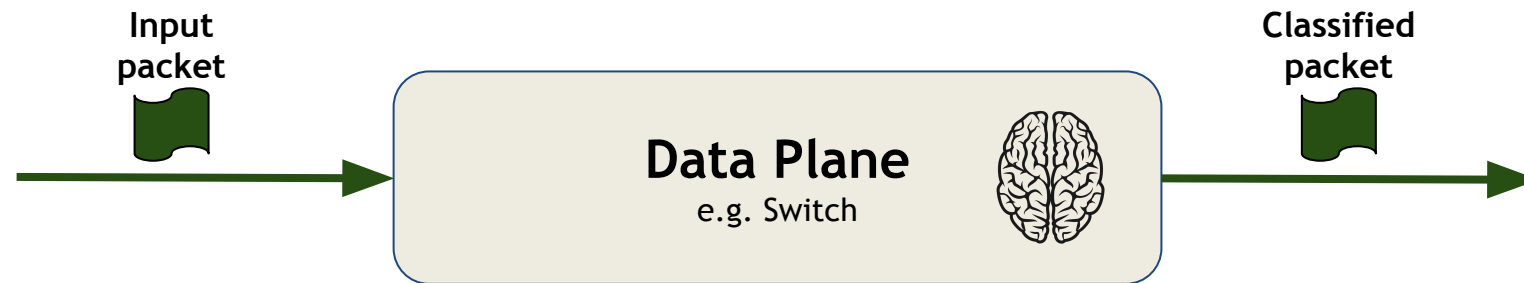
Traditional SDN

# Why Data Plane Machine Learning Inference?



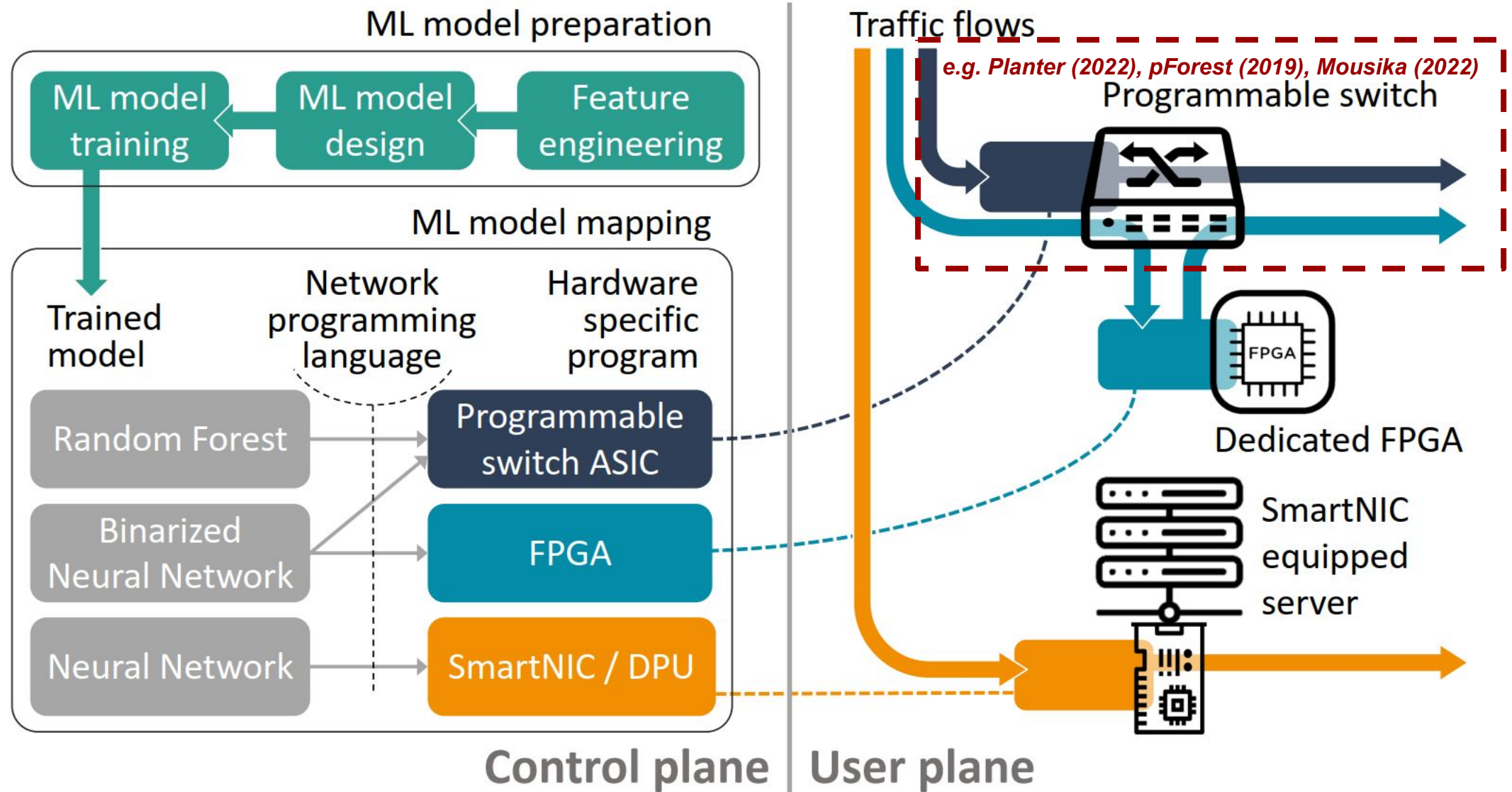
Communication with controller introduces a delay

# Why Data Plane Machine Learning Inference?



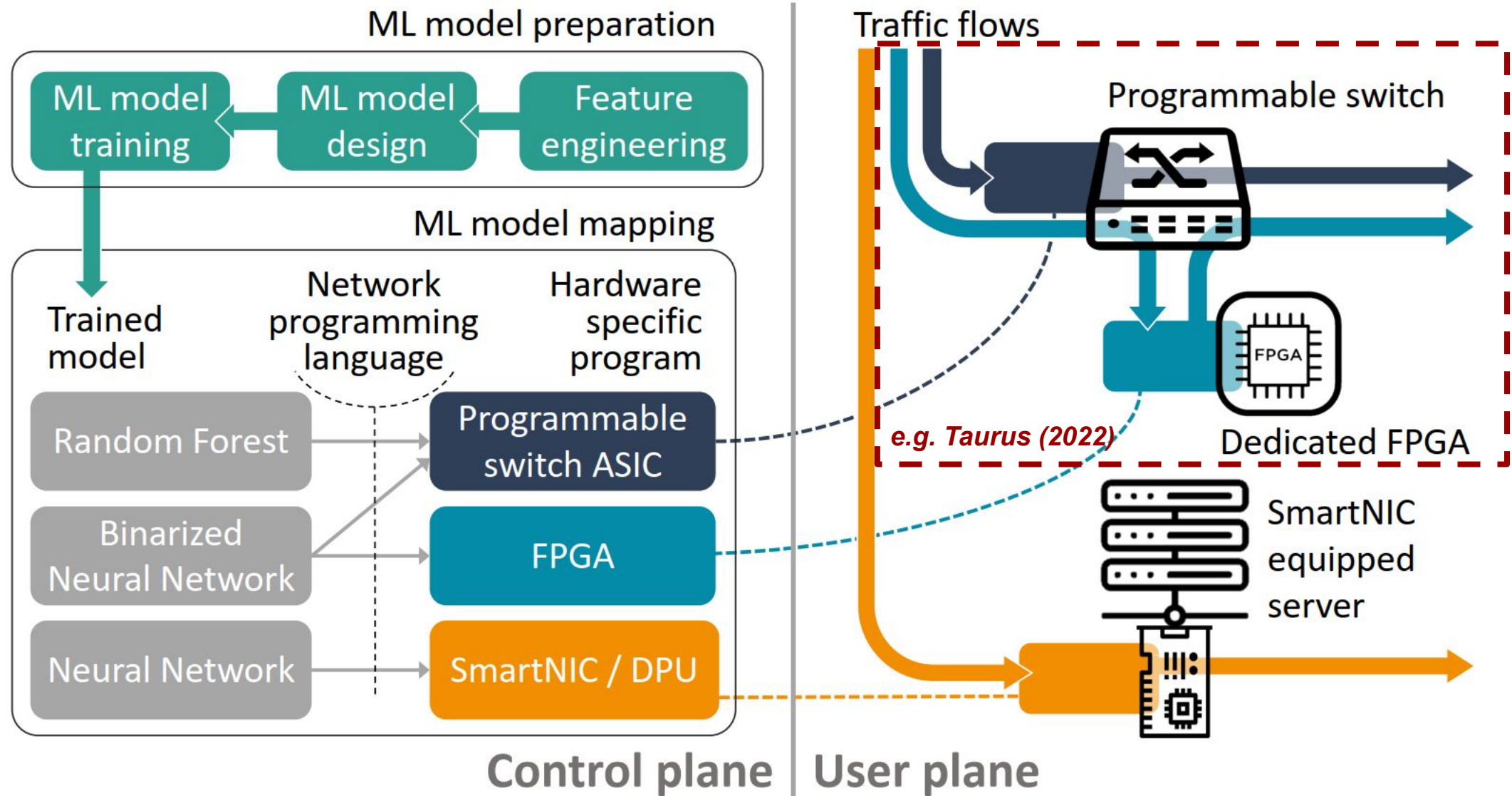
We want to eliminate the delay by bringing ML inference into the data plane

# Line-rate Machine Learning Inference



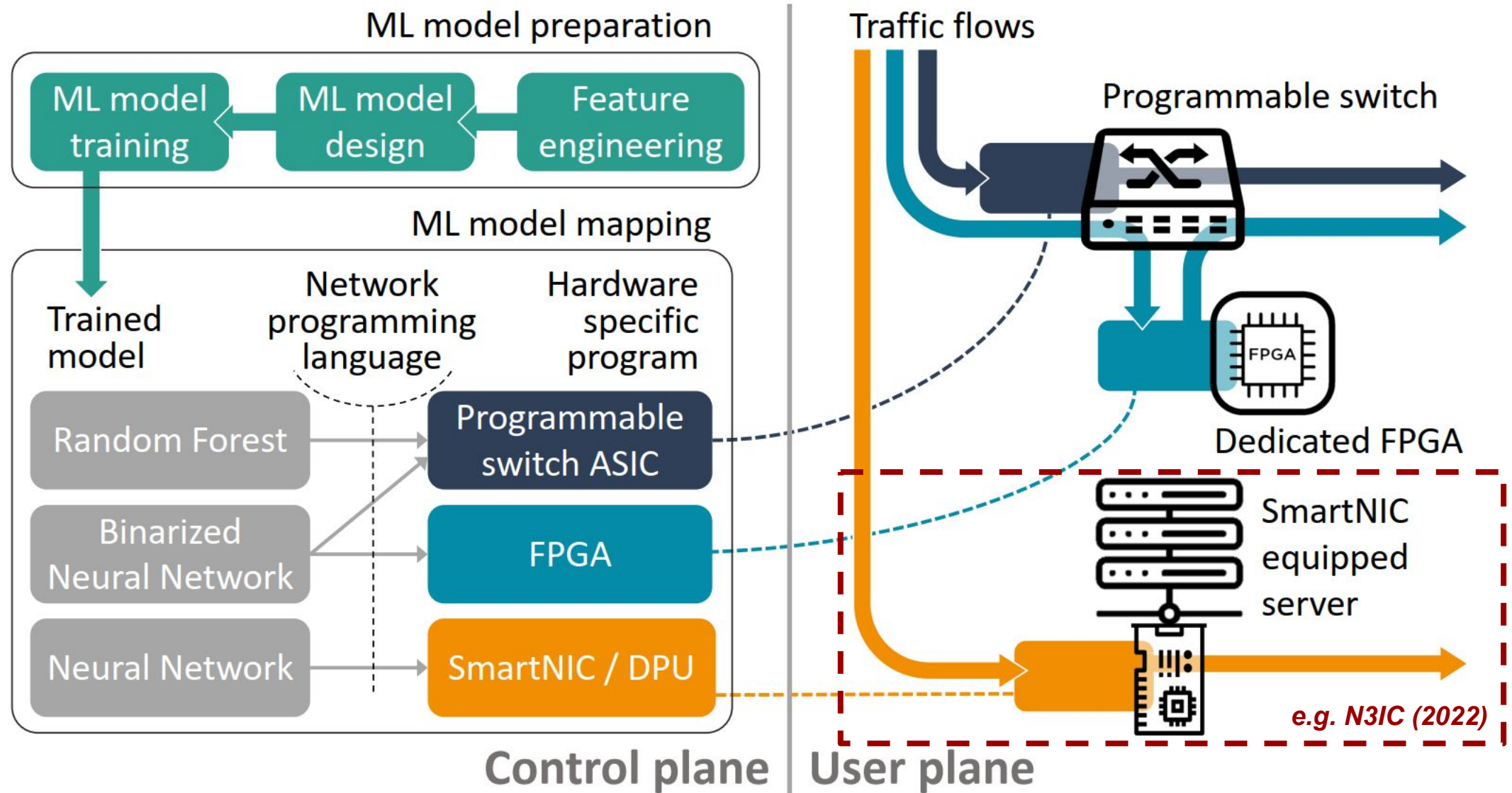


# Line-rate Machine Learning Inference





# Line-rate Machine Learning Inference



# In-Switch Machine Learning Inference with Random Forests

# State-of-the-Art

**Problem** **➡** *Train a single model for the problem and map it to the switch*

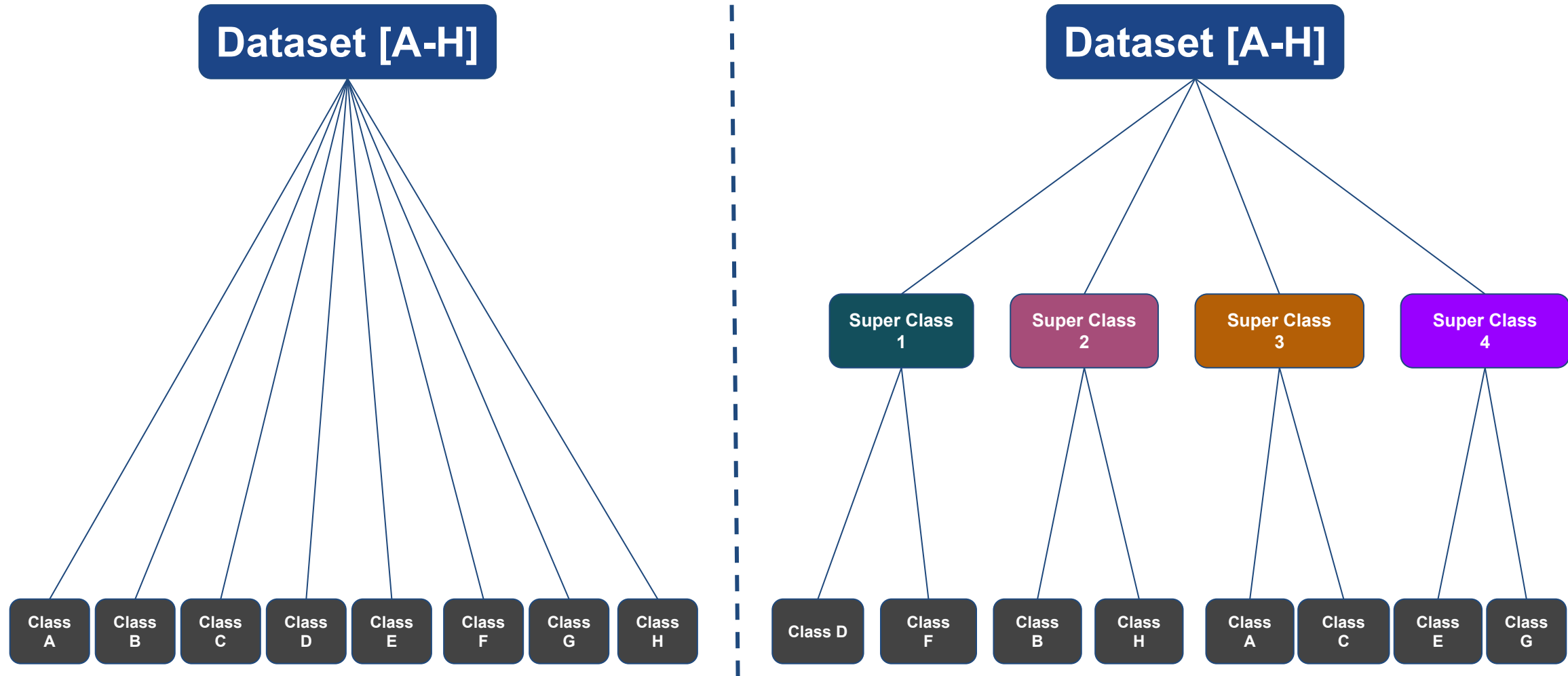
- llsy (Xiong et al, 2019)
- pForest (Busse-Grawitz et al, 2019 & 2022)
- SwitchTree (Lee et al, 2020)
- Planter (Zheng et al, 2021 & 2022)
- NERDS (Xavier et al, 2021)
- pHeavy (Zhang et al, 2021)
- Mousika (Xie et al, 2022)

For difficult tasks, a monolithic classifier often becomes too complex to fit within switch resources while attaining the desired accuracy.

If hierarchical relationships exist, the task could be split into smaller tasks that are easier to solve, improve classification accuracy and fit within switch resources.



# Hierarchical Machine Learning Inference



# Henna: In-Switch Hierarchical Machine Learning Inference

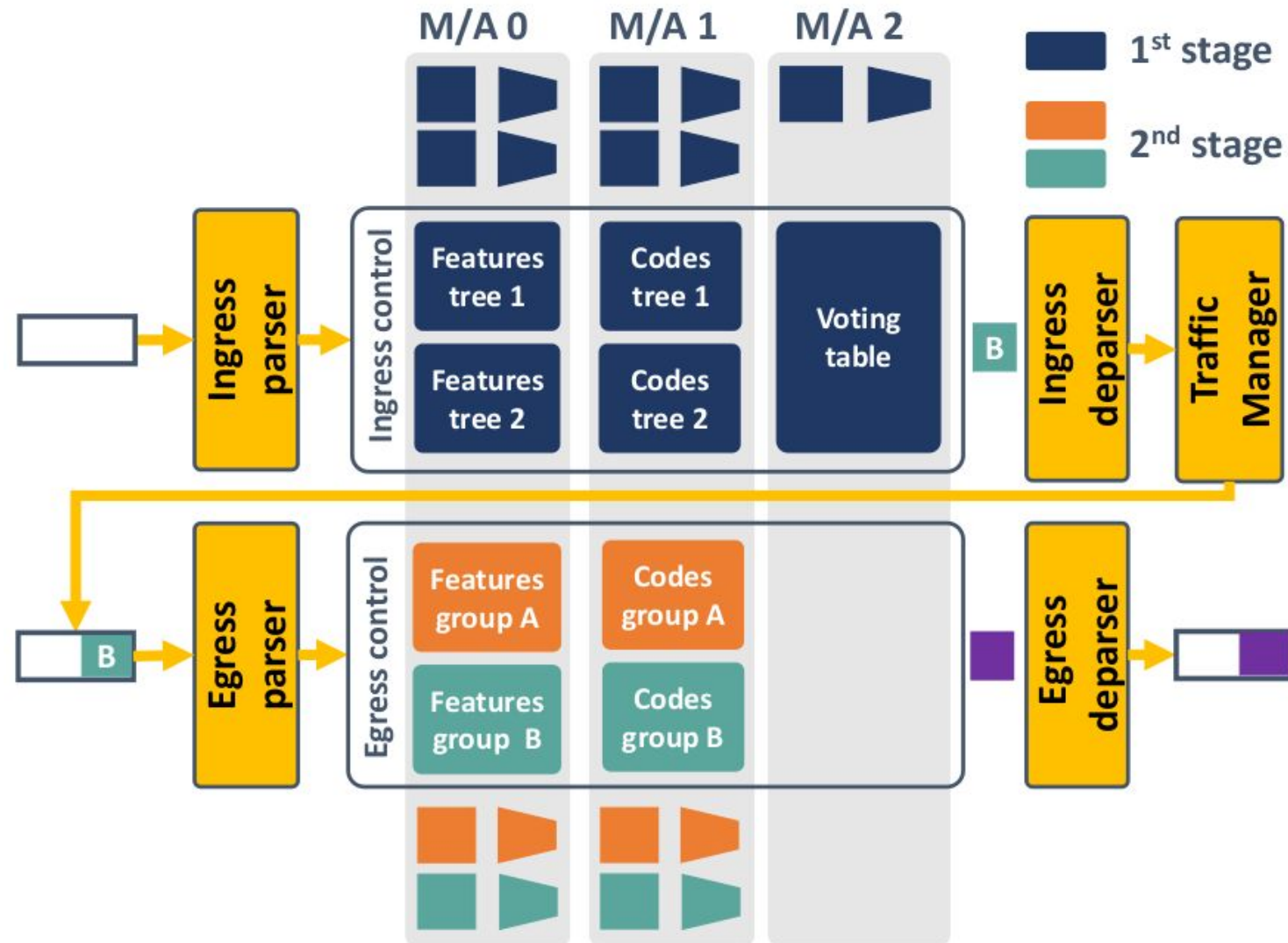
# Henna: Description

- **Implemented as a two-stage classifier**
  - *First stage - identify class groups, Second stage - identify individual classes*
- **First stage model in ingress, second stage models in egress**
  - *Allows models of both stages to share M/A stage resources in Tofino switch*
- **Models trained in Python using the Scikit-Learn library**
  - *Feature selection, grid search for hyperparameters, model validation & selection*
- **Henna uses the decision tree/random forest mapping in Planter<sup>1</sup>**
  - *Trees are mapped via feature range tables and tree code tables*

<sup>1</sup>Changgang Zheng and Noa Zilberman: Planter: seeding trees within switches. *SIGCOMM '21 Poster and Demo Sessions*. ACM, New York, NY, USA, 2021.



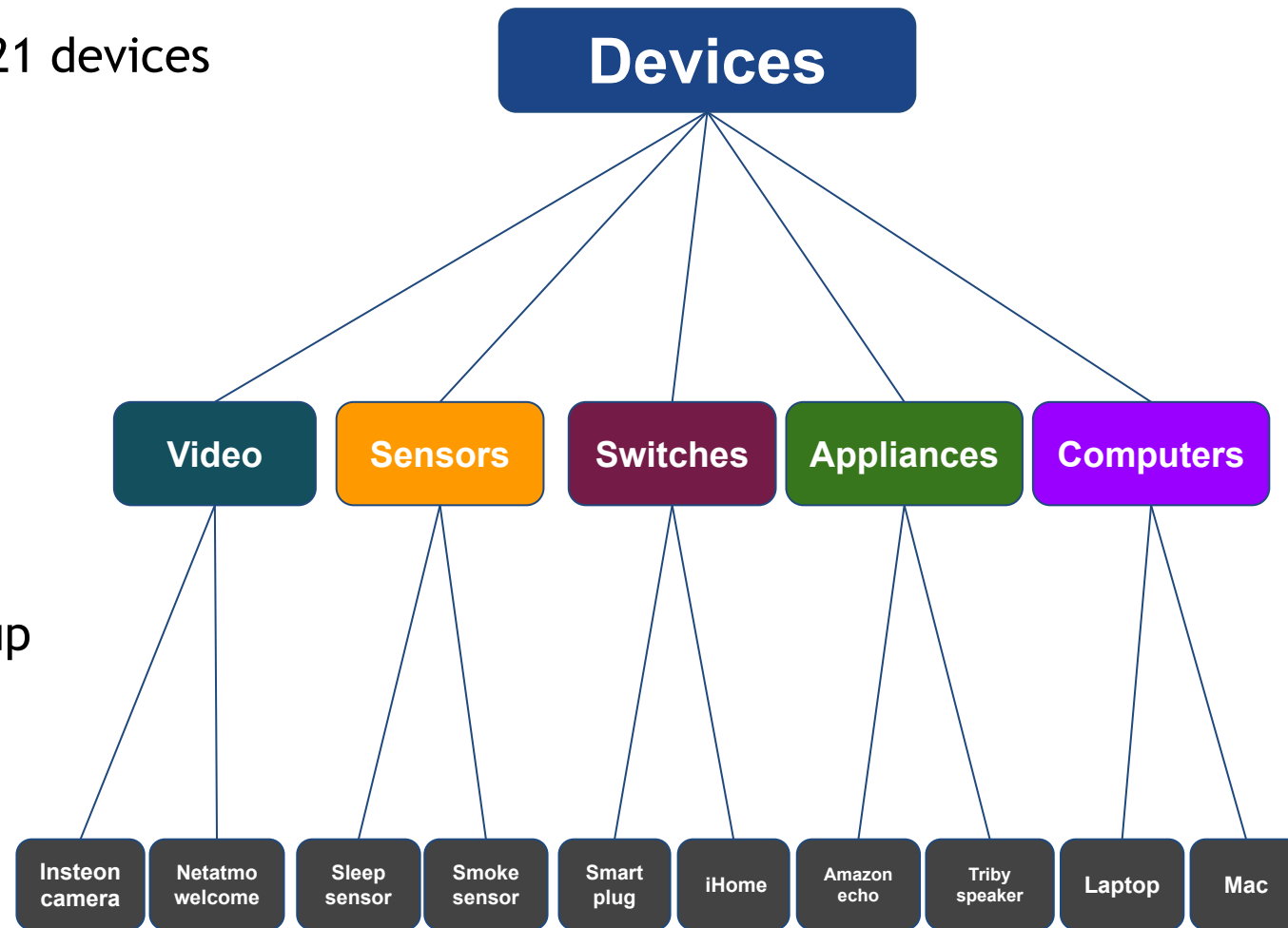
# Henna Mapping into the PISA Architecture



# Use Case and Experimental Setup

# Use Case: UNSW-IoT Traces [1]

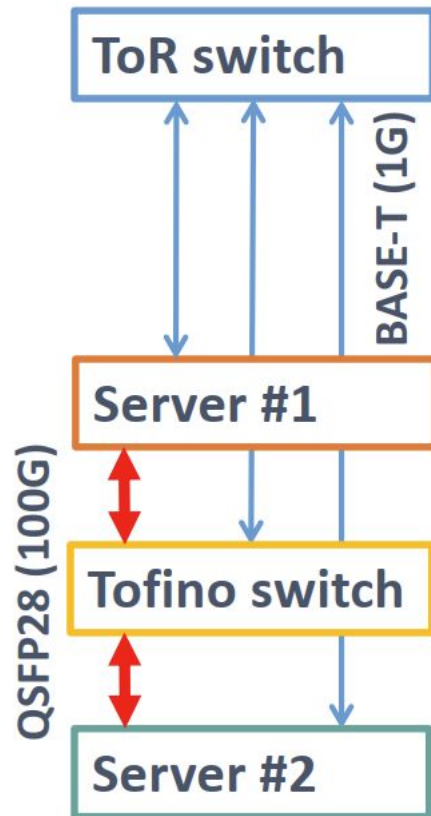
- Device Identification problem with 21 devices
- Devices clustered into 5 groups
  - Switches & Plugs
  - Sensors
  - Video Devices
  - Appliances
  - Computers
- **First stage:** identify the device group
- **Second stage:** identify the device
- **Benchmark:** single-stage classifier



[1] <https://iotanalytics.unsw.edu.au/iottraces.html>



# Experimental Setup



- Internet connectivity

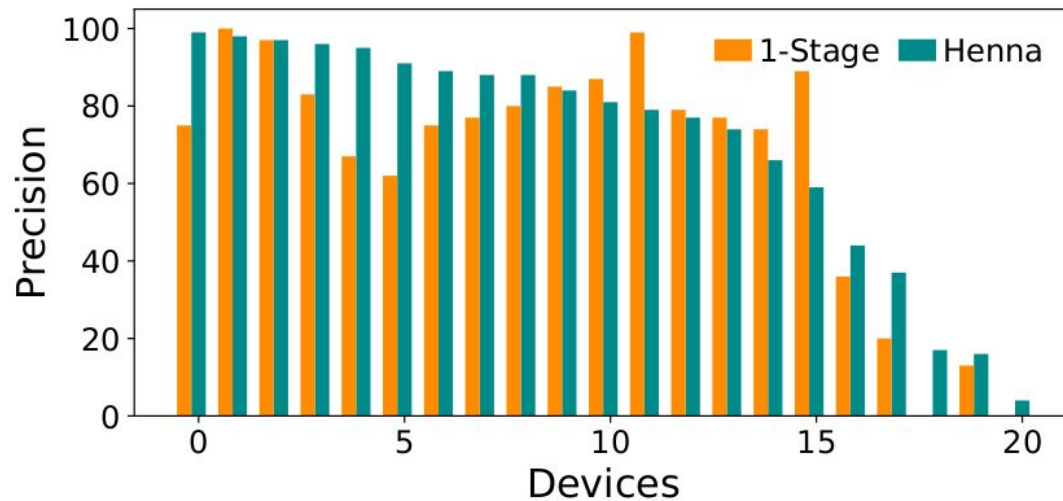
- Classification evaluation  
- End hosts

- Traffic classification

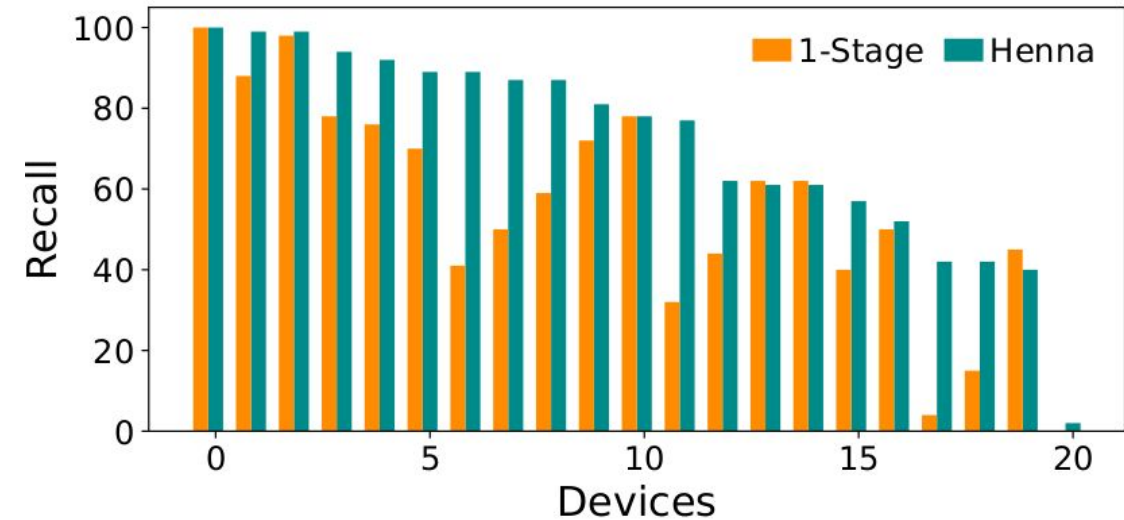
- Controller  
- Source hosts

# Results

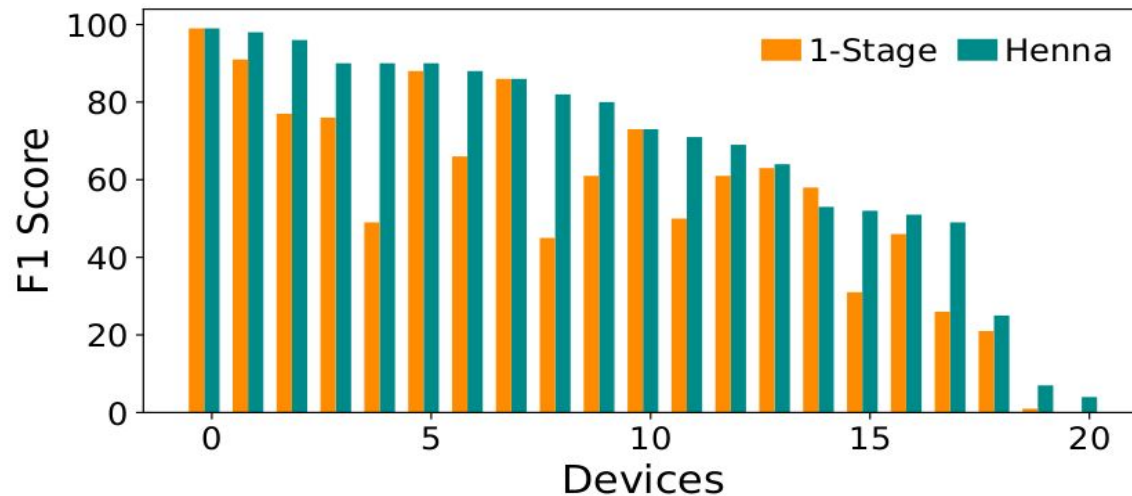
# Results - Classification Accuracy



$$\text{Precision} = \text{TP} / \text{TP} + \text{FP}$$



$$\text{Recall} = \text{TP} / \text{TP} + \text{FN}$$



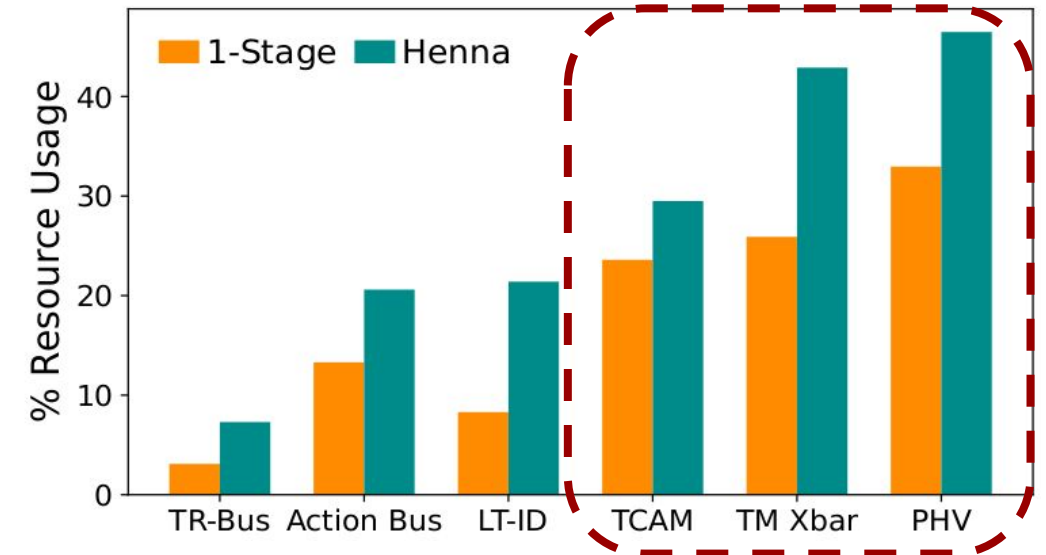
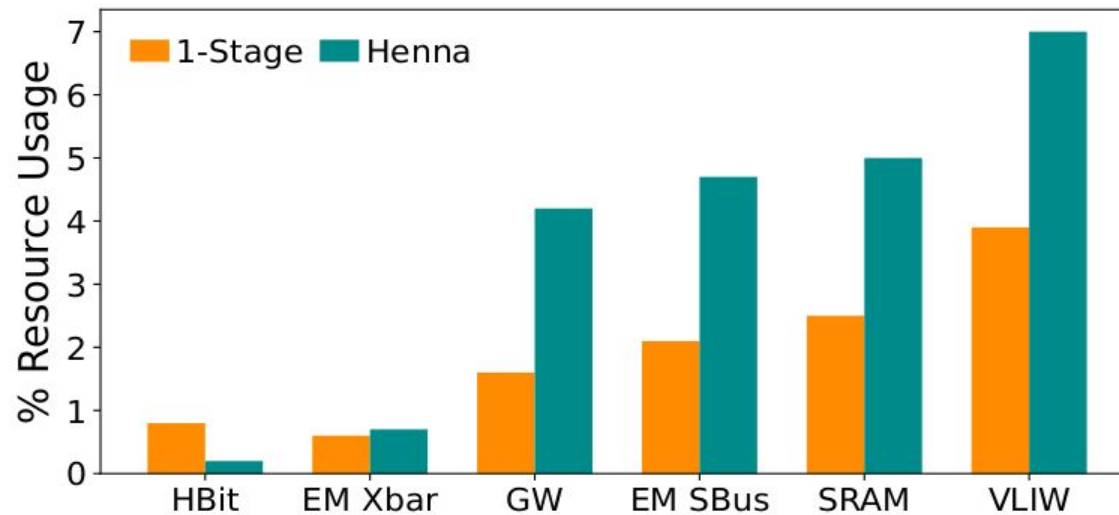
$$\text{F1 Score} = 2 \times (\text{Precision} \times \text{Recall}) / (\text{Precision} + \text{Recall})$$

**Gain in F1 score:**

- Absolute: 11.95%
- Relative: 21.52%

# Results - Resource Usage

On average Henna consumes about 8% of total switch resources



Significant use of TCAM and PHV memory (inherent to RF mapping used)



# Conclusion

# Conclusion

- We presented Henna, a two-stage decision tree-based in-switch packet classifier
- We implemented our solution in a real-world experimental platform
- Results show that Henna improves classification performance with respect to a monolithic classifier while keeping resource usage under control.
- Future work will seek to extend Henna to flow classification, reduce resource consumption and explore new use cases.

Code available: <https://github.com/nds-group/Henna>

# Thank you!



MINISTERIO  
DE CIENCIA  
E INNOVACIÓN



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