

# ML-Based Modeling and Virtualization of Reconfigurable Multi-Accelerator Systems

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## **Motivation**





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### **Traditional Scenario**



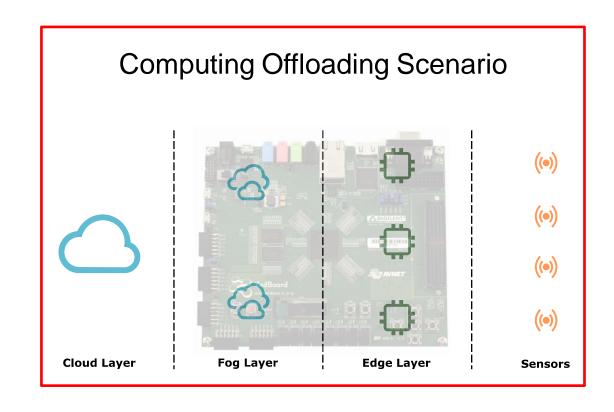




## **Motivation**

### **Traditional Scenario**









## Goals

 Real-Time Modeling and Management of the Reconfigurable multi-accelerator systems

2. Virtualization Support to the reconfigurable multi-accelerator systems





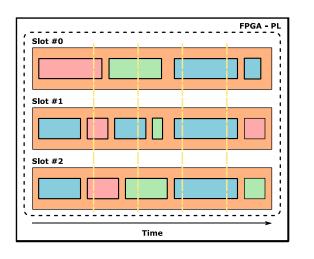
## **Proposed Solution**

Machine learning based modeling for reconfigurable multi-accelerator systems

Run-time power consumption and performance monitorization system

Computing offloading workload manager







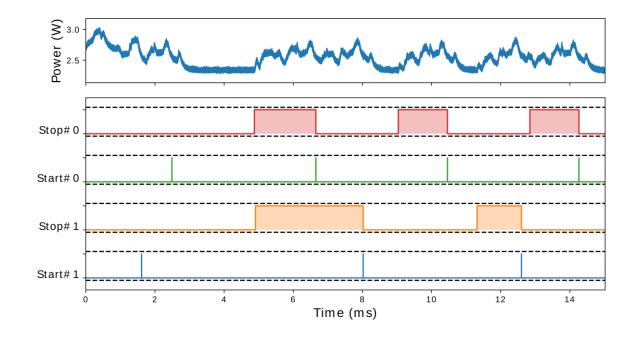


# **Monitoring Infrastructure**

- Non-intrusive
- Synchronized power and performance traces

## Components:

- Measurement board
- Monitoring IP
- Software components





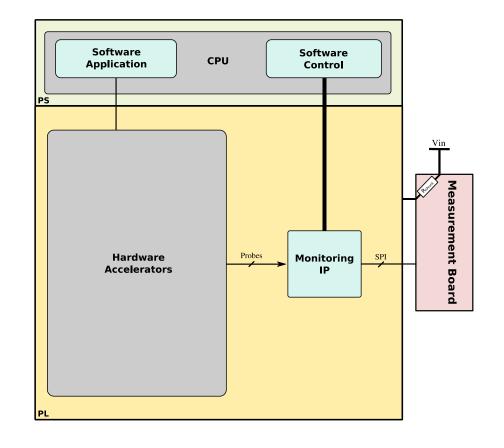


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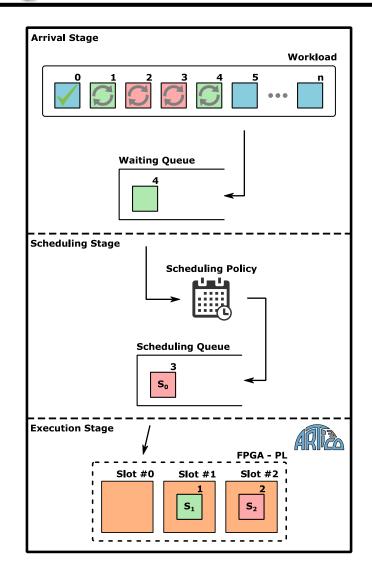






## **Workload Manager**

- Generate configurable computing offloading workloads
- Execute hardware acceleration requests
- Monitor and track every request throughout the entire process

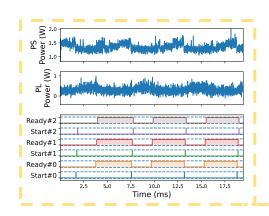


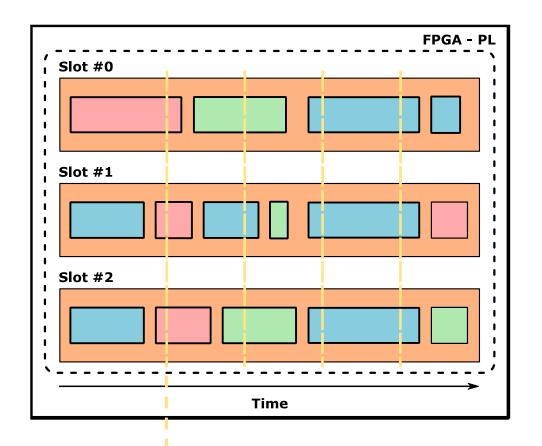




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# **Test Setup**







## **Modeling Approach**

- Machine Learning-based models
  - Power consumption
  - Performance
- Machine learning algorithms:
  - Support Vector Regression (SVR)
  - Regression Tree Ensemble (RTE)





## **Experimental Setup**

## MachSuite [1]

## Training and validation datasets:

- Edge computing workload with combinations of 11 different MachSuite kernels, generated with the Workload Manager
- PS power consumption, PL power consumption and performance traces obtained with the Monitoring Infrastructure integrated in the Workload Manager

[1] B. Reagen, R. Adolf, Y. S. Shao, G. Wei, and D. Brooks, "MachSuite: Benchmarks for accelerator design and customized architectures"





# **Experimental Results – Analysis**

#### **Kernel Interaction Impact on Execution Time**







Model	RMSE	MAPE
PS Power Consumption (SVR)	0.0447	1.329
PL Power Consumption (SVR)	0.0072	1.486
Execution Time (RTE)	0.3549	6.335

RMSE: Root Mean Squared Error

MAPE: Mean Absolute Percentage Error







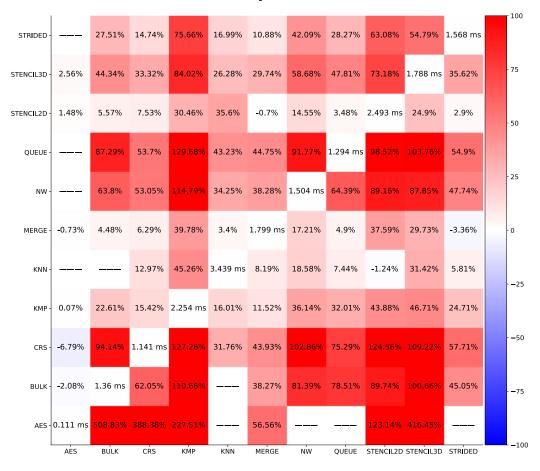
11-kernel combinations | Up to 8 accelerators per kernel | Around 5000 test observations

Observations obtained under a Linux-based OS





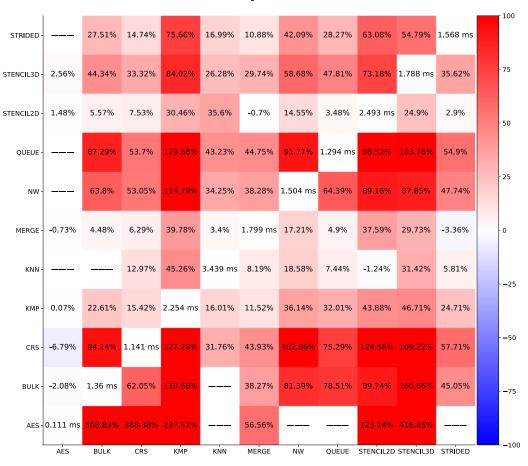
#### **Kernel Interaction Impact on Execution Time**



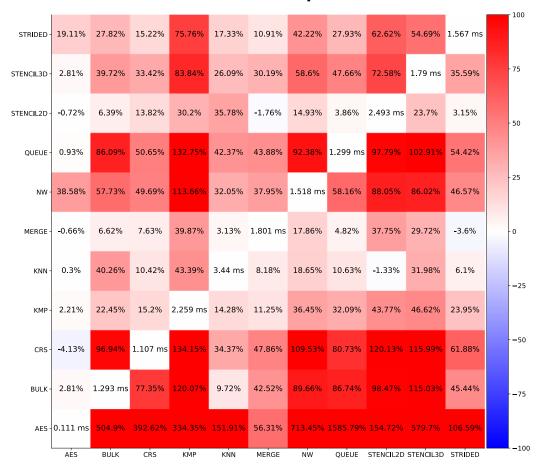




#### **Kernel Interaction Impact on Execution Time**



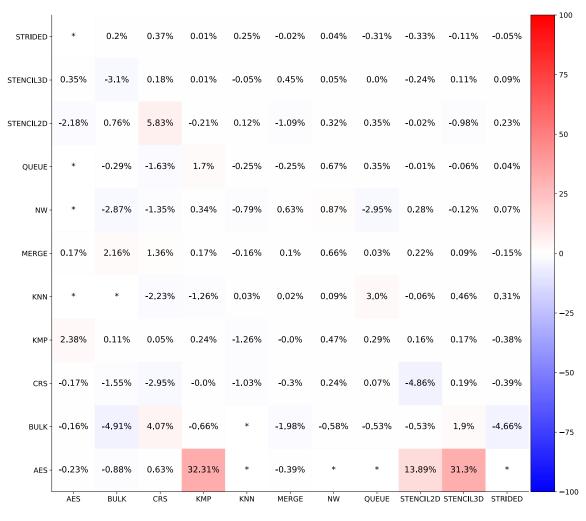
#### **Predicted Kernel Interaction Impact on Execution Time**







#### Relative Error when Predicting Kernel Interaction Impact on Execution Time







## **Conclusions**

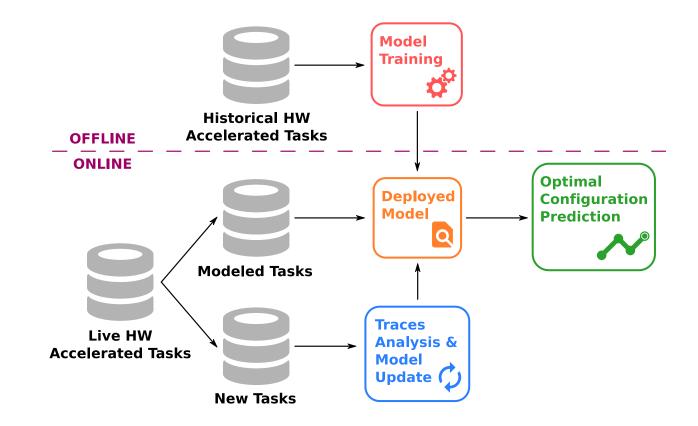
- Dynamic workload management infrastructure for effective computing offloading workloads generation, offloading and monitoring in FPGA-based systems.
- ML-based models are very accurate when predicting power consumption and performance in reconfigurable multi-accelerator systems as well as modeling the interaction between kernels.





## **Future Work**

- Online modeling
- Run-time self-adaptation

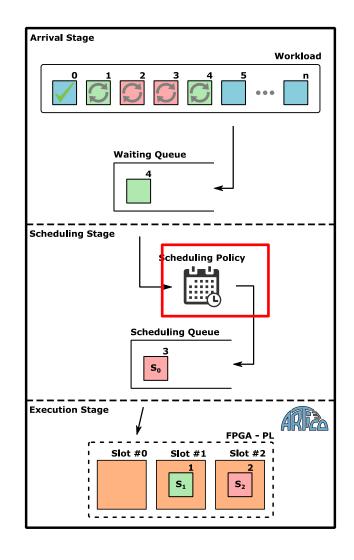






## **Future Work**

- Smart scheduling approach
- Advanced resource management







# **THANK YOU FOR YOUR ATTENTION**



