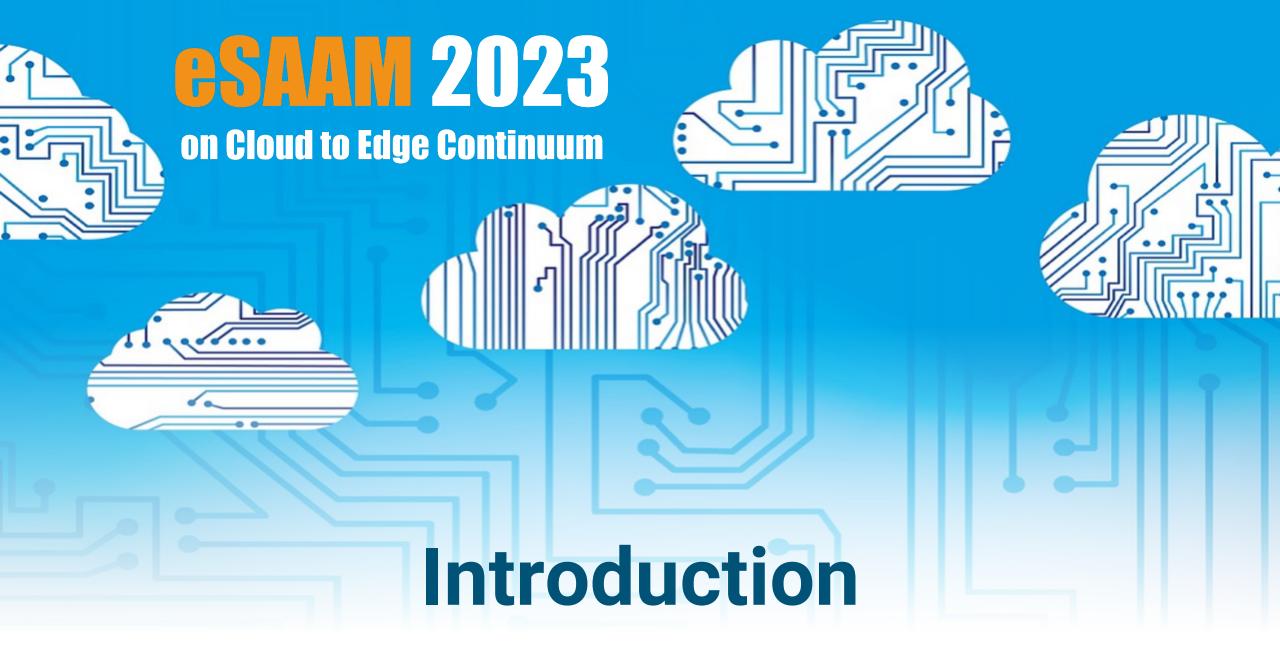


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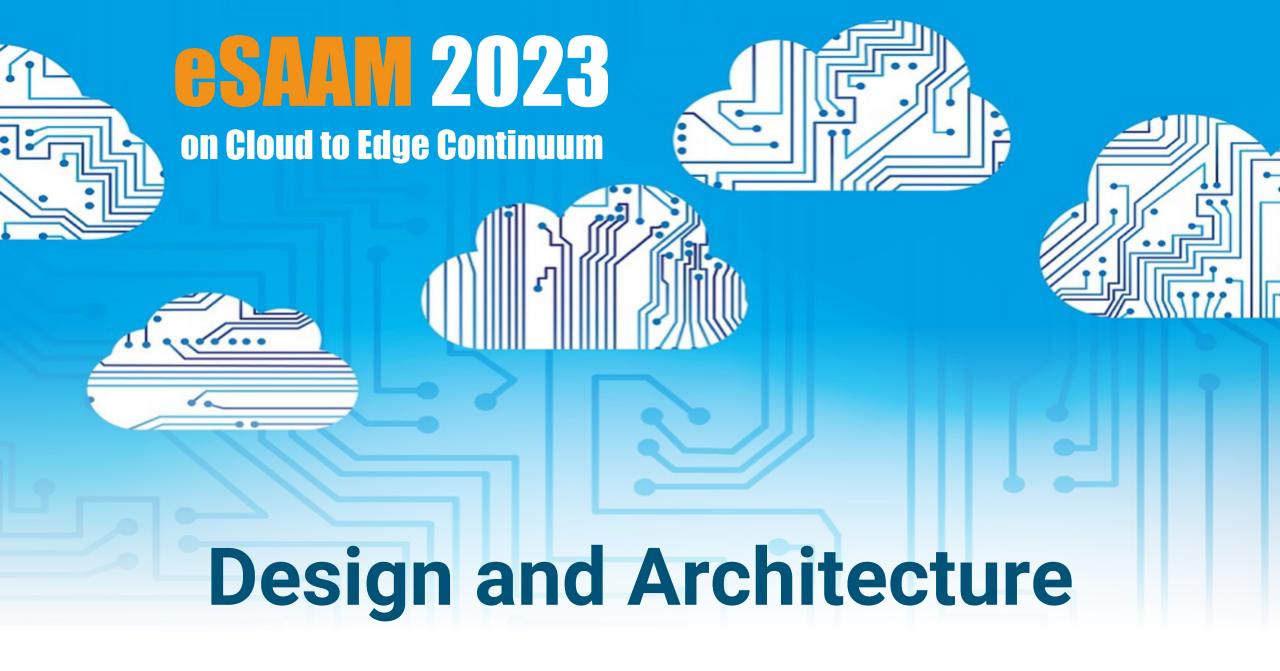
Oct. 17, 2023

Project Aim



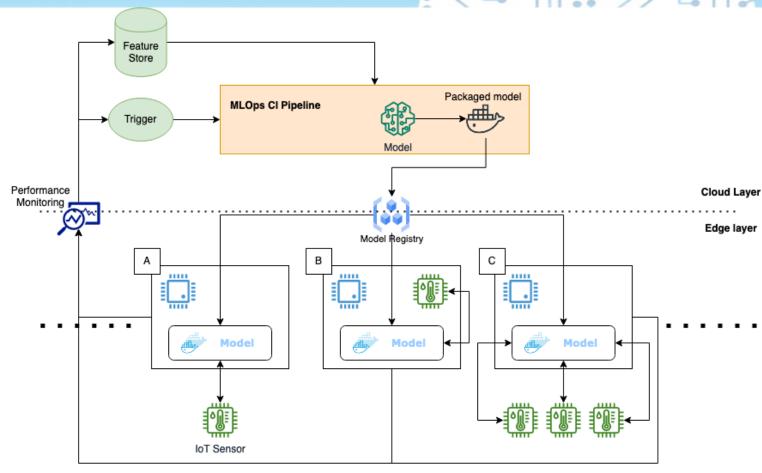


The aim of the project is to research, design, and develop a platform that enables seamless integration of MLOps practices with edge devices.



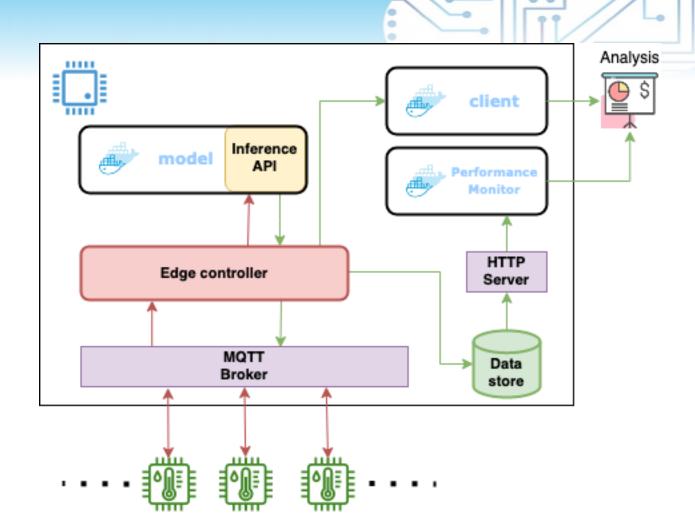






Edge Architecture





Implementation



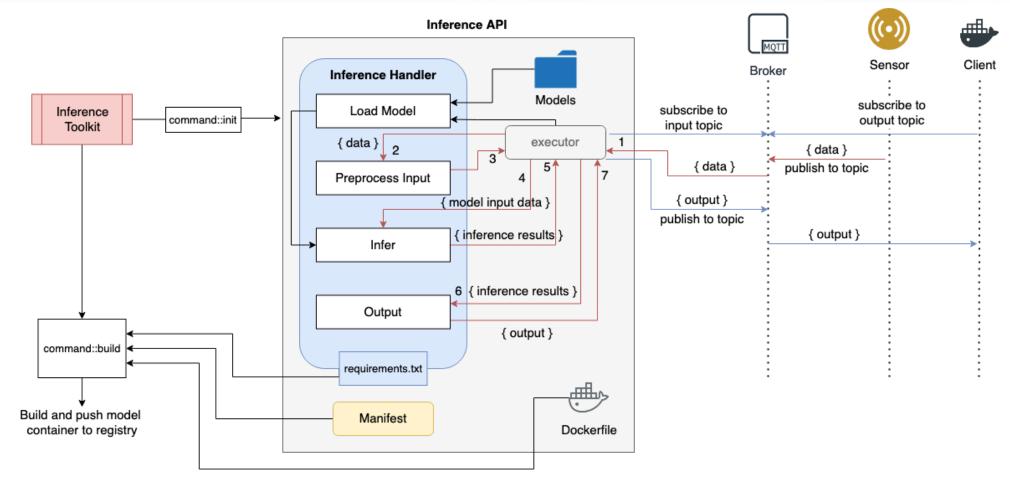


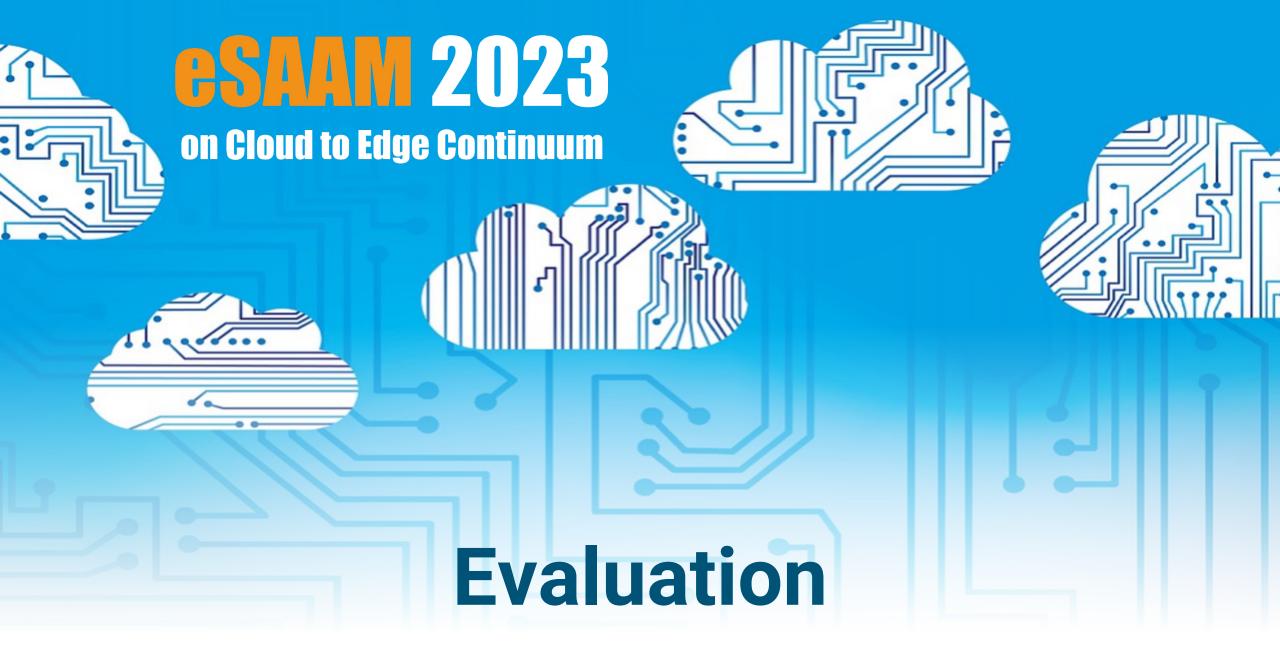
- The solution consists of two command-line tools.
 - Inference Toolkit packaging models with inference API into a Docker container (used in the CI pipeline)
 - **Edge Manager** setting up and managing infrastructure components of an edge device (used in the edge device)

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Implementation - Inference Toolkit/API





Oct. 17, 2023



Finding a dataset

- Predictive maintenance is a very valid use case for ML applications at the edge, as anomalies must be identified as quickly as possible.
- The operating environment must consist of hardware resources capable of running ML inference.
- Running inference on the cloud is not a viable option.



Dataset

- Open dataset aimed to detect component failures in an air pressure system (APS) of trucks.
- 76,000 samples, each containing 171 attributes.
- Positive class component failures for a specific component of the APS.
- Negative class component failures not related to the APS.

11/19



Presents an ideal case to develop a validation scenario:

- Air pressure system failures must be identified immediately as failure to do so may result in severe damage.
- Computer systems in modern trucks come equipped with hardware capable of running ML inference.
- Trucks operating on the road may not always have internet access.



Prediction Model

- Training set with 60,000 examples:
 - 59,000 belonging to the negative class.
 - 1,000 belonging to the positive class.
- Test set with 16,000 examples:
 - 15,625 belong to a negative class.
 - 375 to a positive class.

MLOps Pipeline

AWS Sagemaker Pipelines

Results





Inference Load Testing

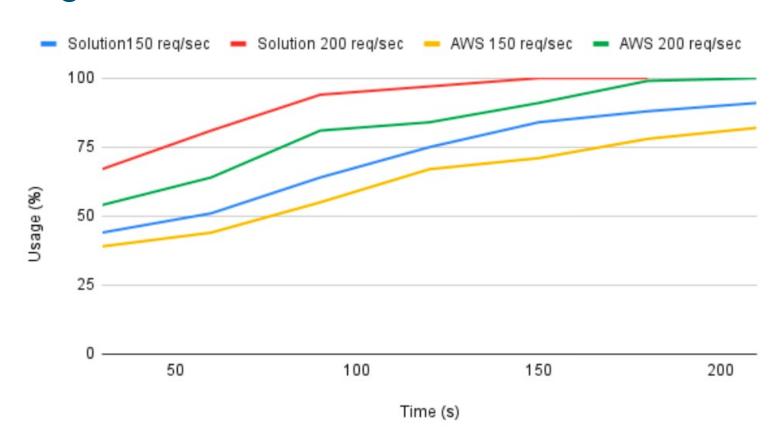


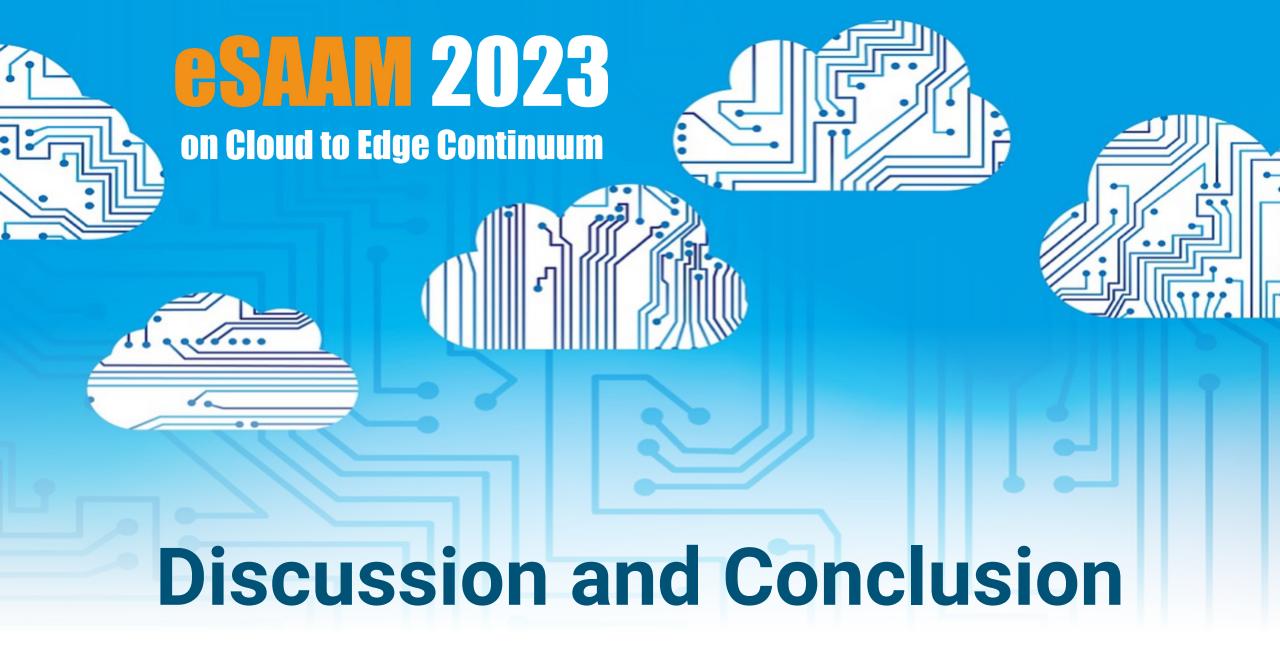
Results





CPU Usage





Discussion





- A similar trend could be observed in LinkEdge and AWS Greengrass, although Greengrass performed slightly better in comparison.
 - Can be attributed to the fact that the Greengrass application does not run in a containerised environment.
- Although using Greengrass provides slightly better performance, it comes at the cost of setting up OS and application-specific libraries on each device individually.
- Running inference in containerised environments has the benefit of having the flexibility to work with device-specific models more efficiently.
- A single-edge device running on LinkEdge can handle requests of up to 150 per second, which is sufficient in most cases that require ML inference at the edge

Conclusion





- The main objective addressed is how can a platform that enables the integration of MLOps practices with edge devices to be developed using state-of-the-art tools and methods.
- LinkEdge, is evaluated to have a performance that matches that of existing tools and services.
- LinkEdge offers end users an open-source tool to set up Edge-MLOps infrastructure with flexibility without relying on third-party software.



on Cloud to Edge Continuum







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