



# How to Right-Size and Future-Proof a Container-First Edge AI Infrastructure

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# Edge AI: There's more to it than the model and the application

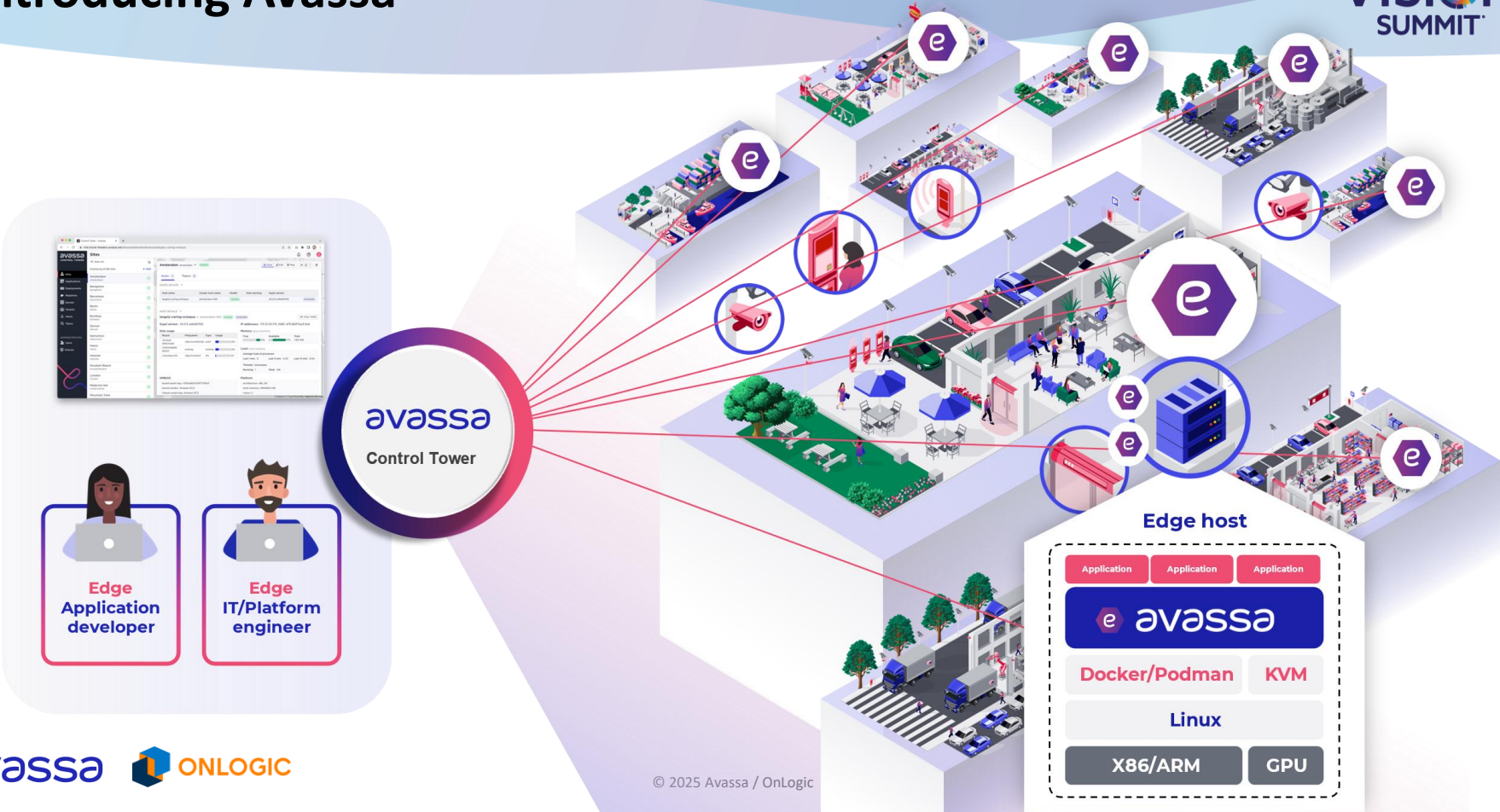
**Edge AI** starts with a model  
and an application...

...but operating AI at the edge  
in resource-constrained  
environments also requires  
**scalable, purpose-built  
infrastructure.**

# In this presentation we will...

- Learn what to consider when:
  - Selecting edge AI devices
  - Managing the lifecycle of edge AI applications
- Lay out the components of a best-of-breed approach to an edge AI stack

# Introducing Avassa



# Introducing OnLogic

## Millions of Configurations

Systems tailored for your exact needs

## Collaborative Sales Approach

No middleman when it comes to solving your challenges

## Engineered for the Edge

Rugged and reliable performance where it matters most

## Prototype to Production

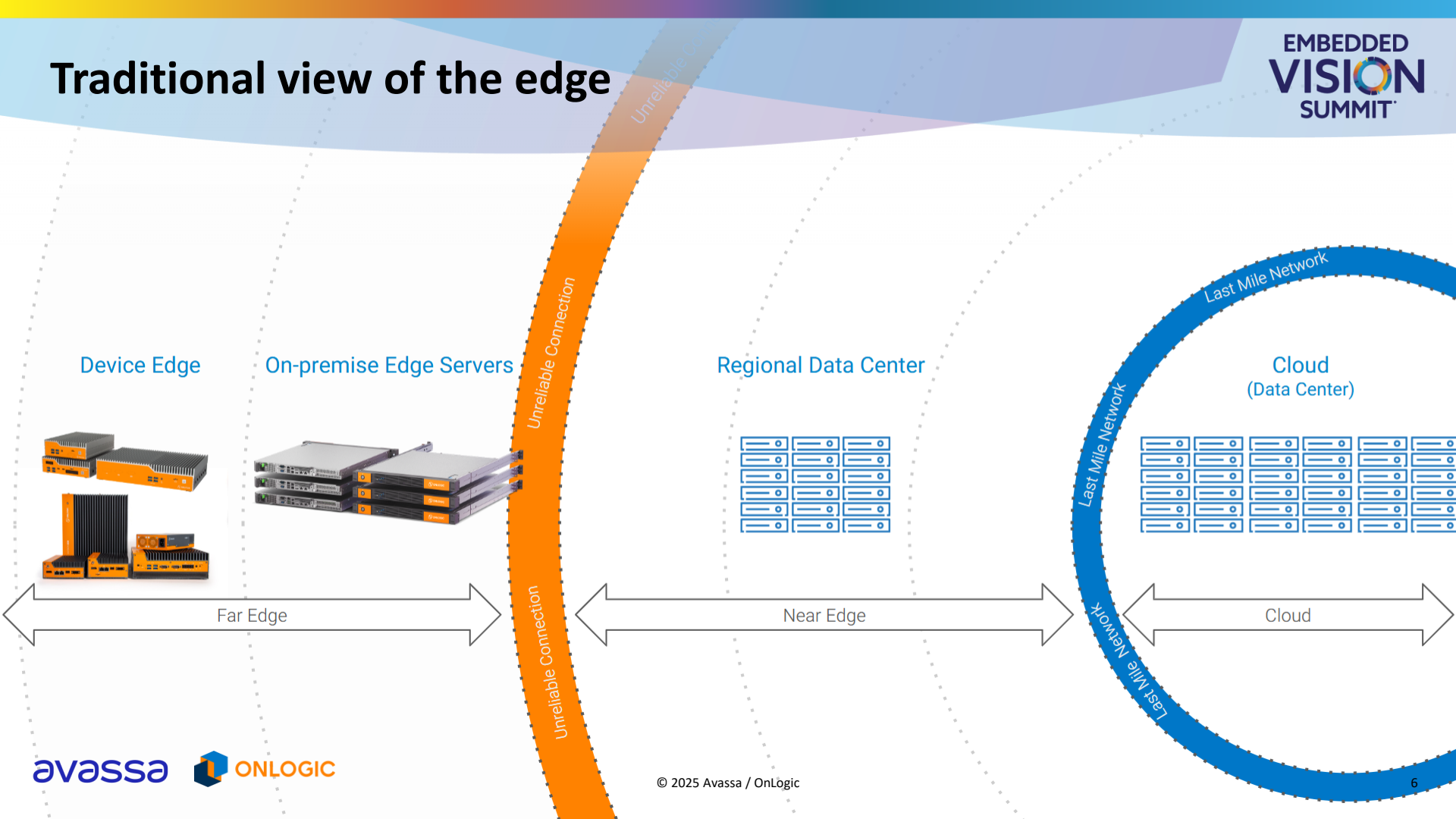
Our engineers can design, deploy, and scale any project

## Built for scale

Expansive global production capacity to meet growing demand



# Traditional view of the edge



# What do I need to consider when selecting edge devices?

# Edge hardware considerations



Temperature



Dust



Constrained  
space

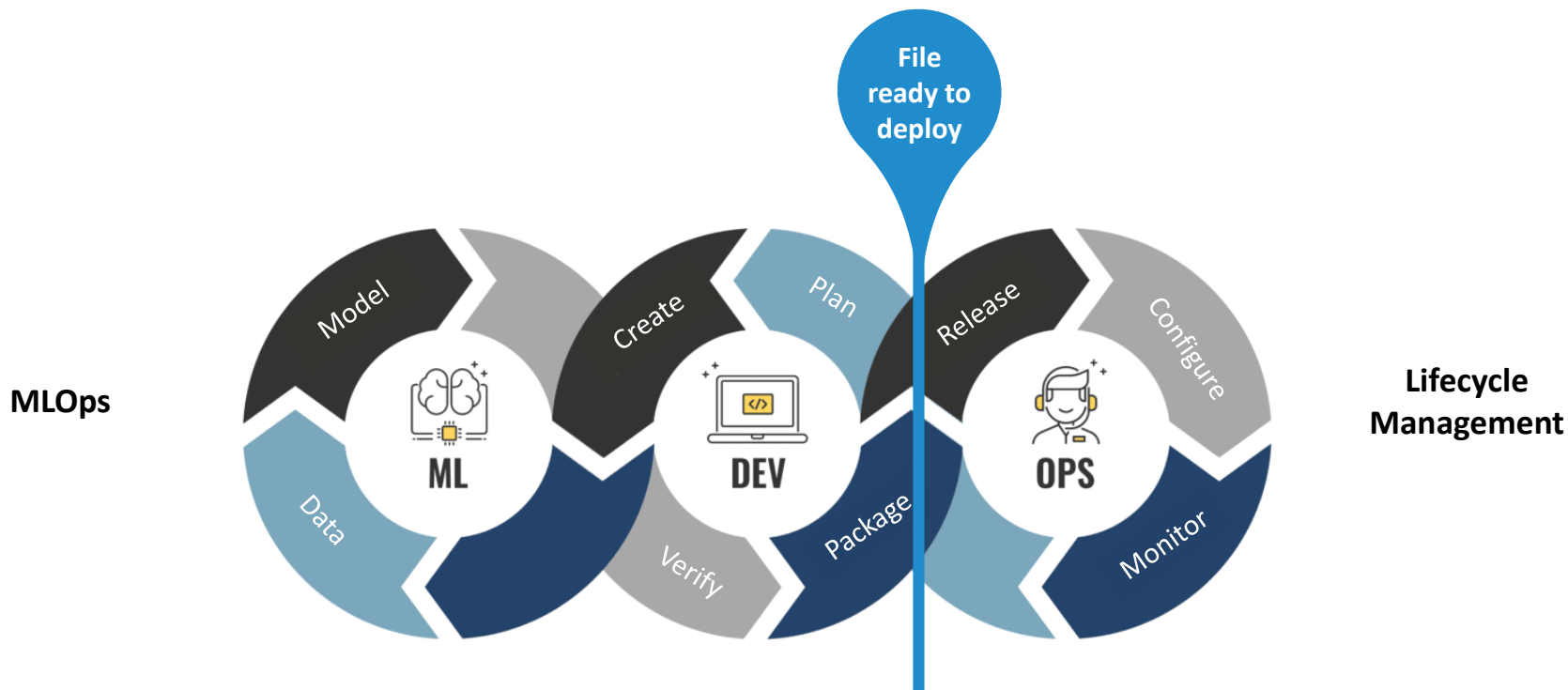


Vibration

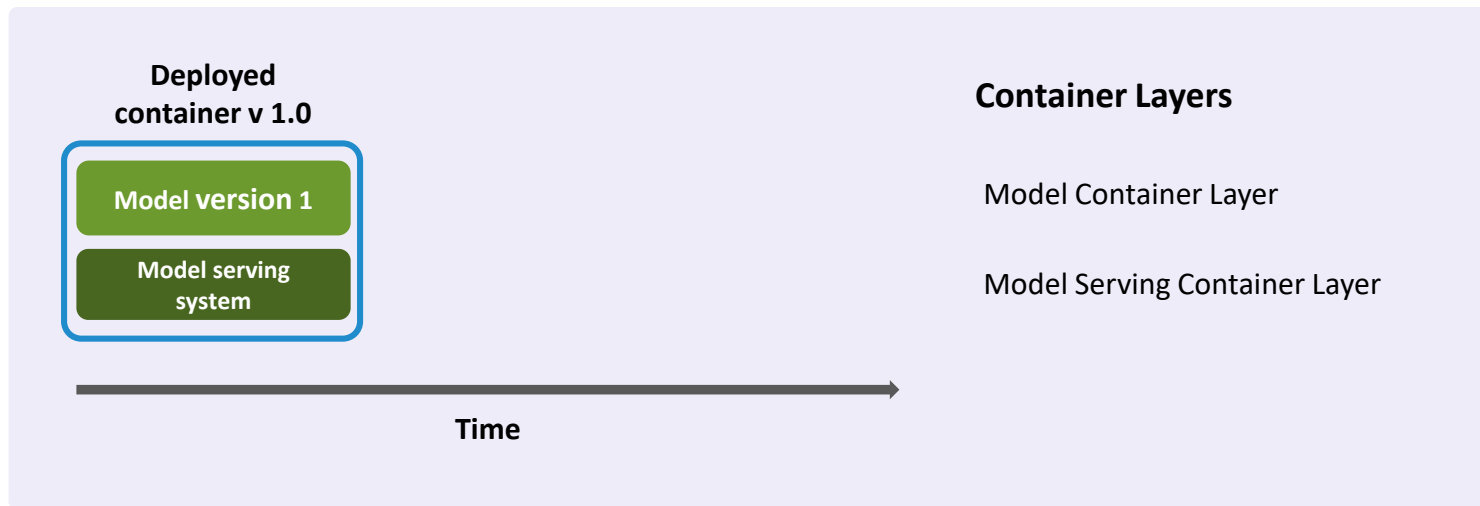
... The edge is **a complex place with different and diverse challenges.**

# What do I need to consider when managing edge AI applications?

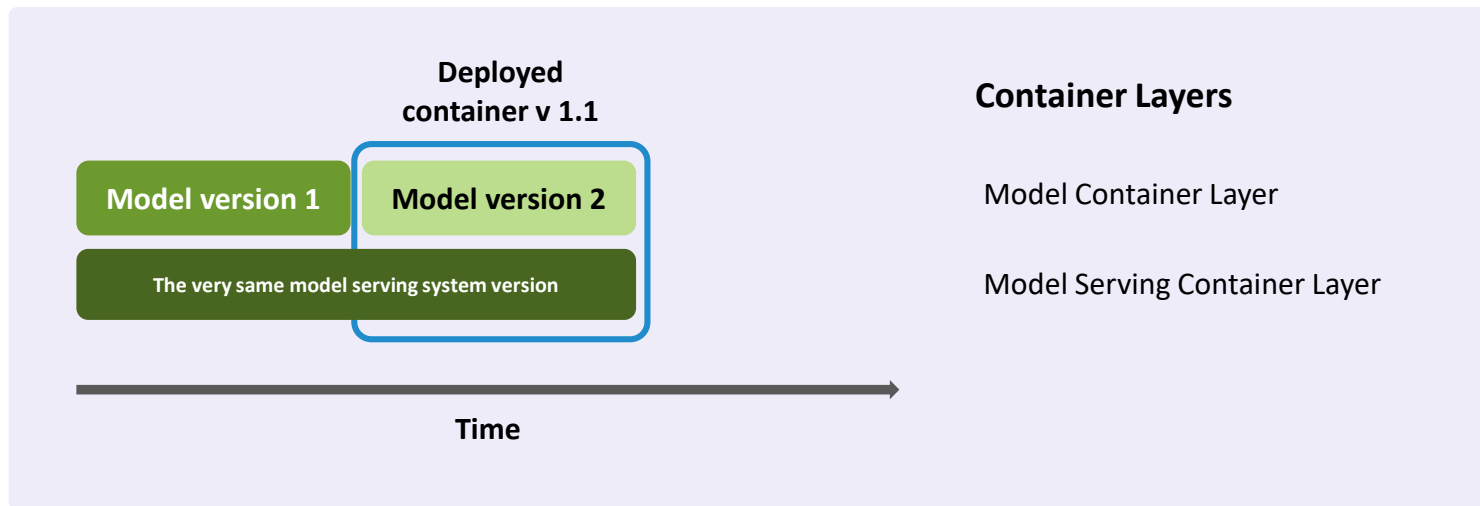
# MLOps meets lifecycle management



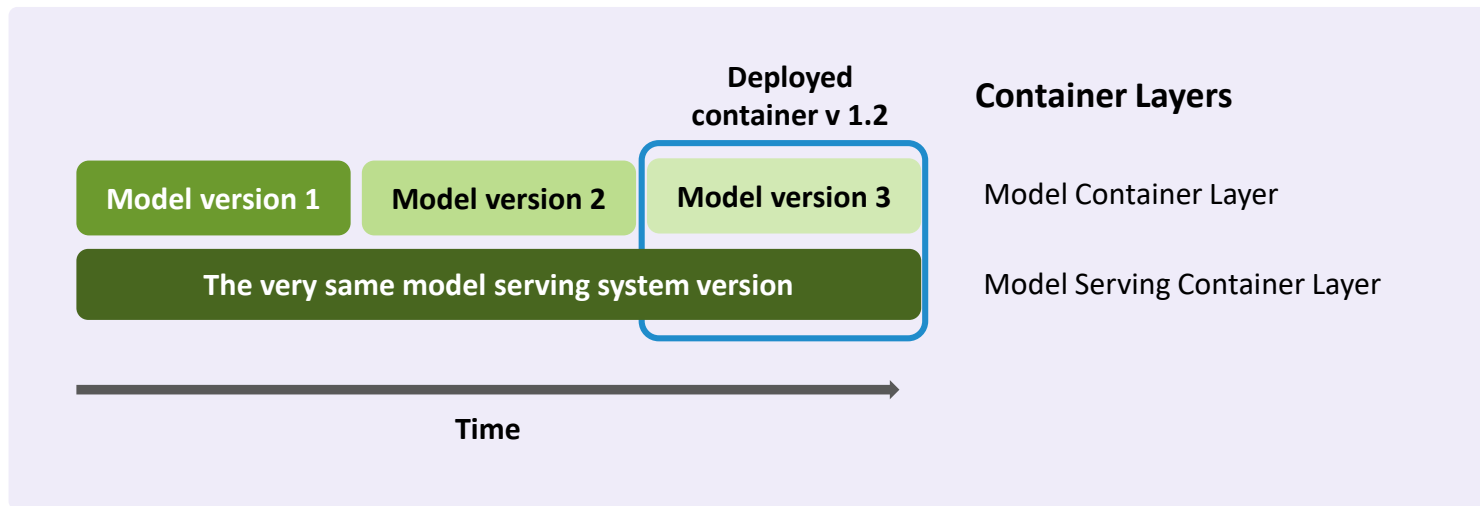
# Purpose-driven model lifecycle management



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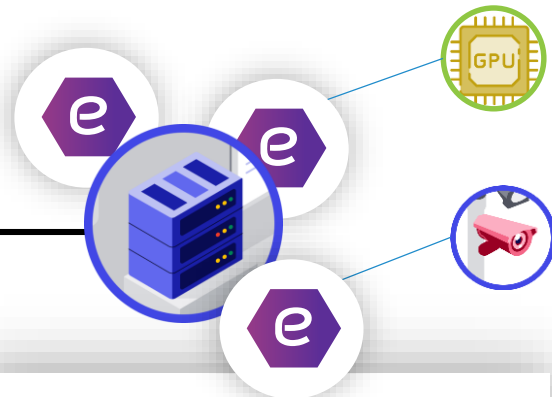
# Purpose-driven model lifecycle management



# Targeted deployments of containers, only where they are needed



Discovery and rule-based labels  
for application placement



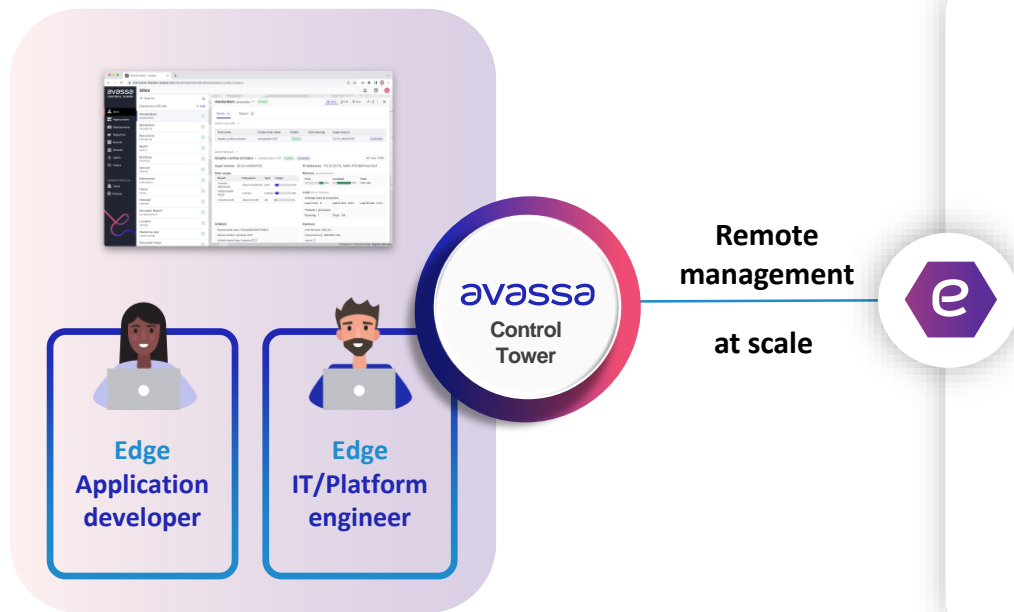
## Targeted deployments

- Declarative requirements in your application
  - Define what and where
- Automatic placement of application at the edge

## Automatic discovery

- Leaf device discovery using Linux kernel features
- Local rule-based label management
- Automatic identification and mounting of local devices
- Local placement based on GPU availability

# Lifecycle managing AI models at the edge



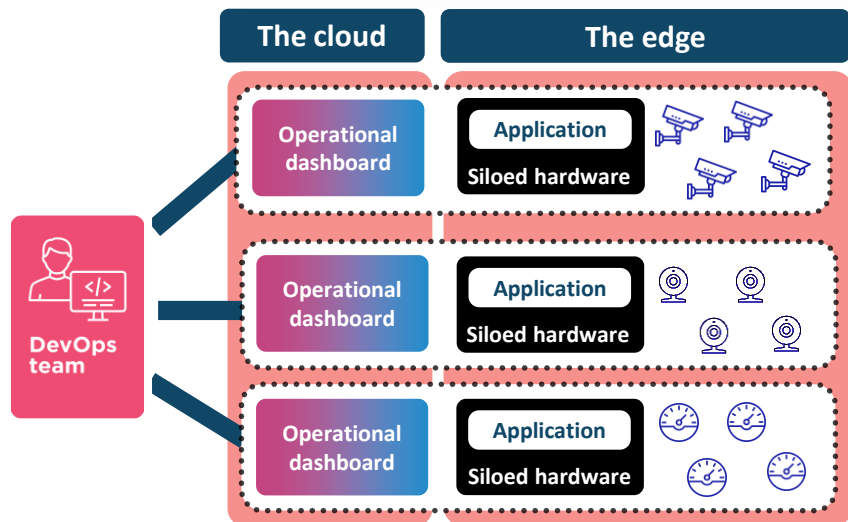
## Key considerations:

- ▶ The model changes faster **than the application**
- ▶ Hardware and GPU requirements **drive application placements**
- ▶ Allow for offline scenarios by making sure your **key application services are kept local**
- ▶ Re-utilize existing investment and **integrate with current systems**

# Why best of breed?

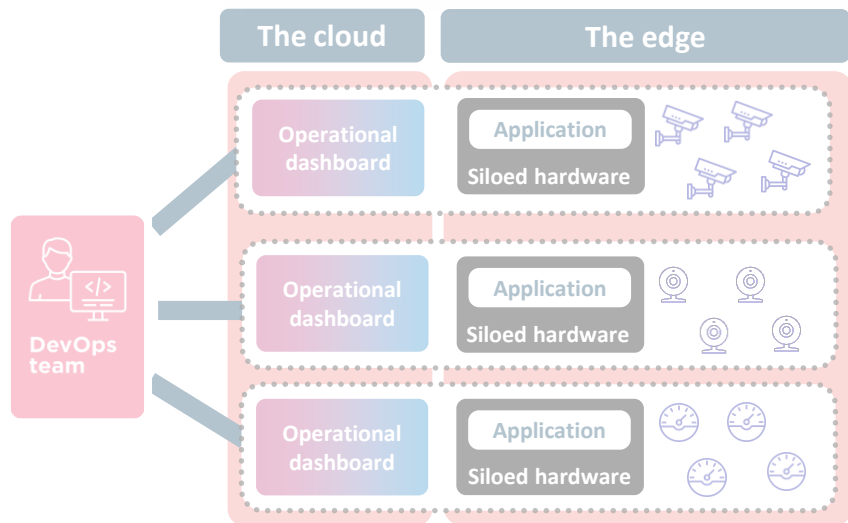
# Avoid silos and redundant stacks

Traditional silos with application-by-application infrastructure solutions

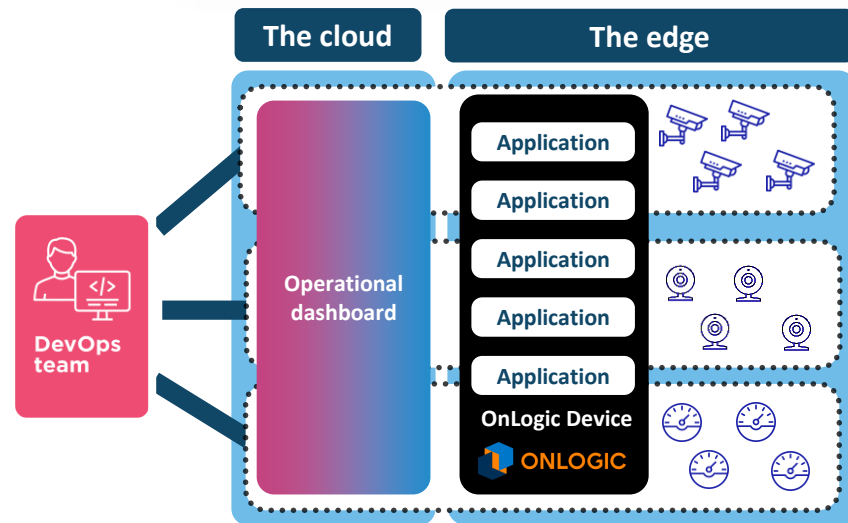


# Avoid silos and redundant stacks

Traditional silos with application-by-application infrastructure solutions



A modern unified platform with streamlined infrastructure and single-pane-of-glass overview



# Hardware that withstands its environment



## Industrial Computers

- Fanless or active cooling
- x86 and ARM architecture
- Industrial operating temp. (0-50°C)



## Rugged Computers

- Resistant to shock & vibration
- Wide input (12-48VDC)
- Wide operating temp. (-40-70°C)



## Panel PCs / HMIs

- 8.4" to 24" screen sizes
- Resistive or capacitive touch
- Up to IP69K ingress protection



## Edge Servers

- 1U to 4U sizes
- Intel and AMD options
- Highly customizable

# A modern, right-sized infrastructure stack for edge AI

## Data & Cloud Integration

App

App

...

AI

Edge Platform

Industrial PCs & HMIs

Sensors, Actuators, PLCs

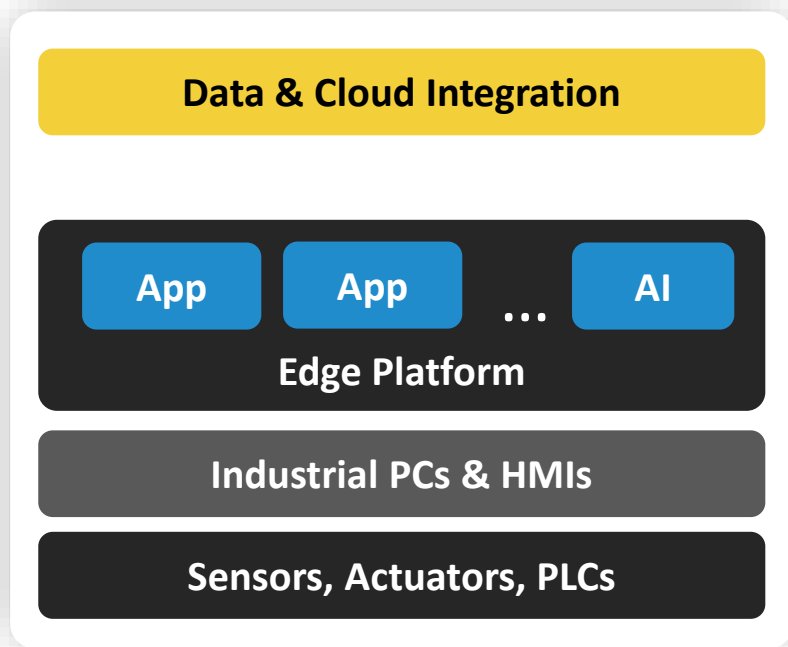
MLOps integrations to continuously deploy and monitor AI workloads through APIs

AI workloads comprise several parts (server, code, model, configuration) with separate lifecycle needs

CPUs with integrated GPUs and NPUs for efficiency, dedicated accelerators when needed.

Control logic going virtual – towards distributed, software-defined control architectures.

# A modern, right-sized infrastructure stack for edge AI



A unified stack, allowing multiple workloads to share common infrastructure

# Conclusions

- **Edge AI success** demands rugged hardware, scalable orchestration, and seamless cloud integration — not just smarter models.
- **A best-of-breed stack** right-sized infrastructure prevents silos and manual overhead, unlocking a unified, agile, and scalable edge environment ready for the challenges of tomorrow.
- **Pairing industrial-grade devices with smart application management** builds Edge AI systems that adapt and thrive over time.



# Resources

[Making the edge lovable for your development and application operations team](#)

[Optimizing Edge AI: Combining MLOps and Edge Orchestration for Success](#)

[Solution description: Avassa for Edge AI](#)

[Solution description: OnLogic for Edge AI](#)

2025 Embedded Vision Summit

Come see us at booth 521!

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