

FactoryDC: Network and Resource Planning for Emerging Applications in Future Factories

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Never Change a Running System! ...or Should we?

- IT & OT are merging
- Paradigm change in factories caused by “Build to order”
- Machine learning

Network
dimensioning?

ML
performance?

How to ac Planning! changes?

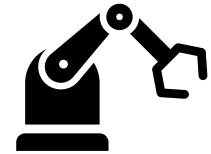
Compute
capacities?

ML resource
requirements?

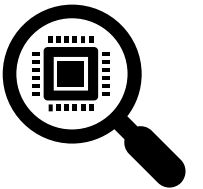
?

Proper Planning needs a Plan

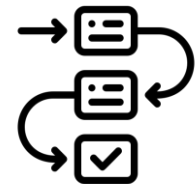
Evaluate ML-centric use cases in the factory



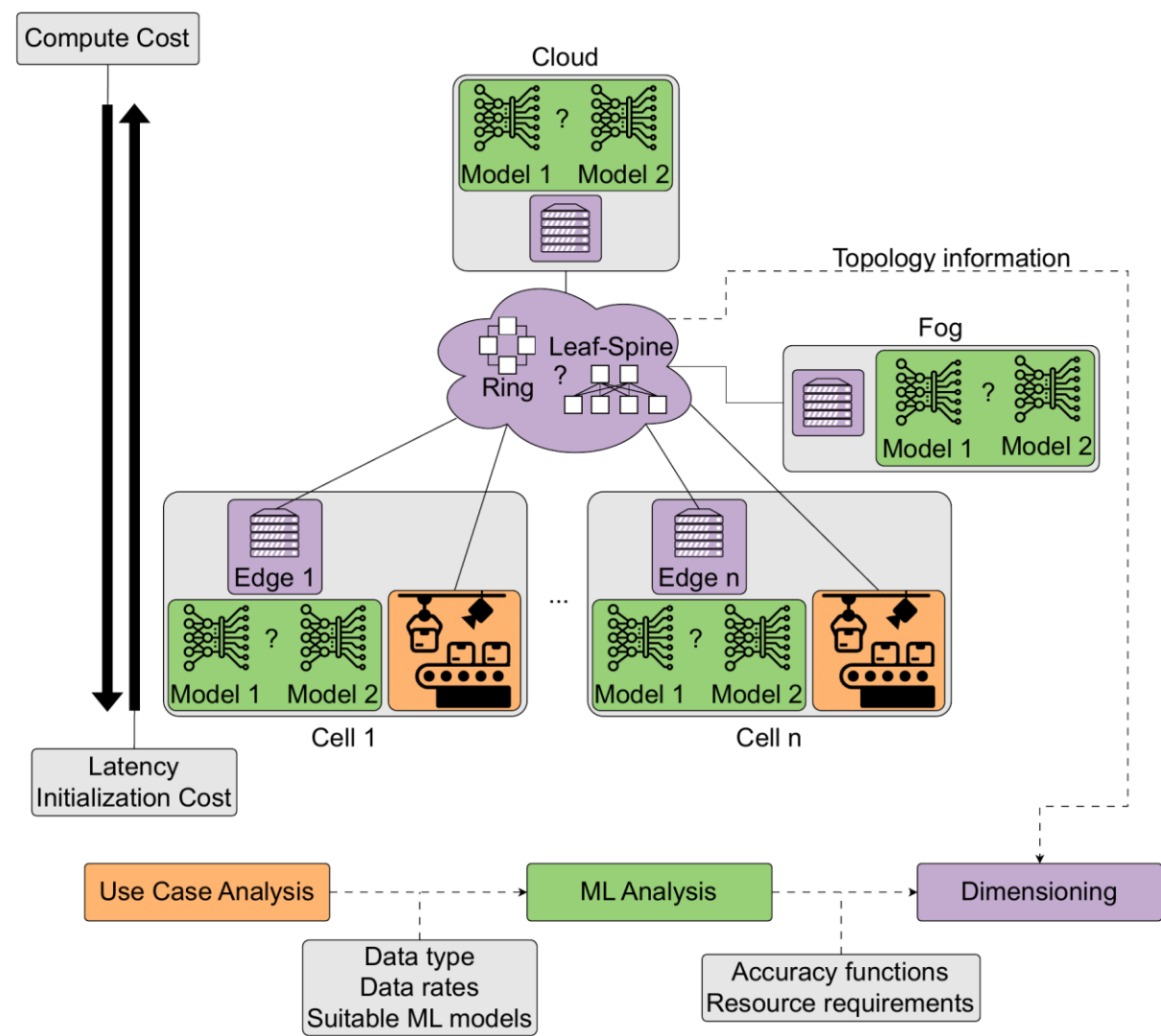
Analyse ML applications w.r.t their performance behaviour



Dimension/Plan with ML applications in mind



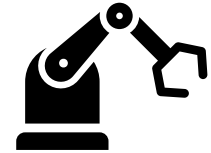
FactoryDC Pipeline – From Use Case to Dimensioning



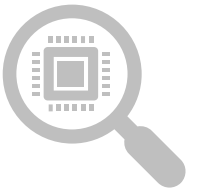
Proper Planning needs a Plan

Omitted for this presentation

Evaluate ML-centric use cases in the factory



Analyse ML applications w.r.t their performance behaviour

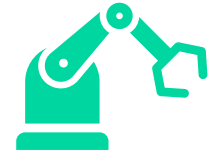


Dimension/Plan with ML applications in mind

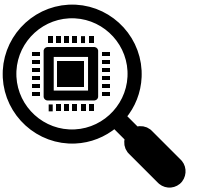


Proper Planning needs a Plan

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Dimension/Plan with ML applications in mind



Machine Learning Analysis – How does it behave?

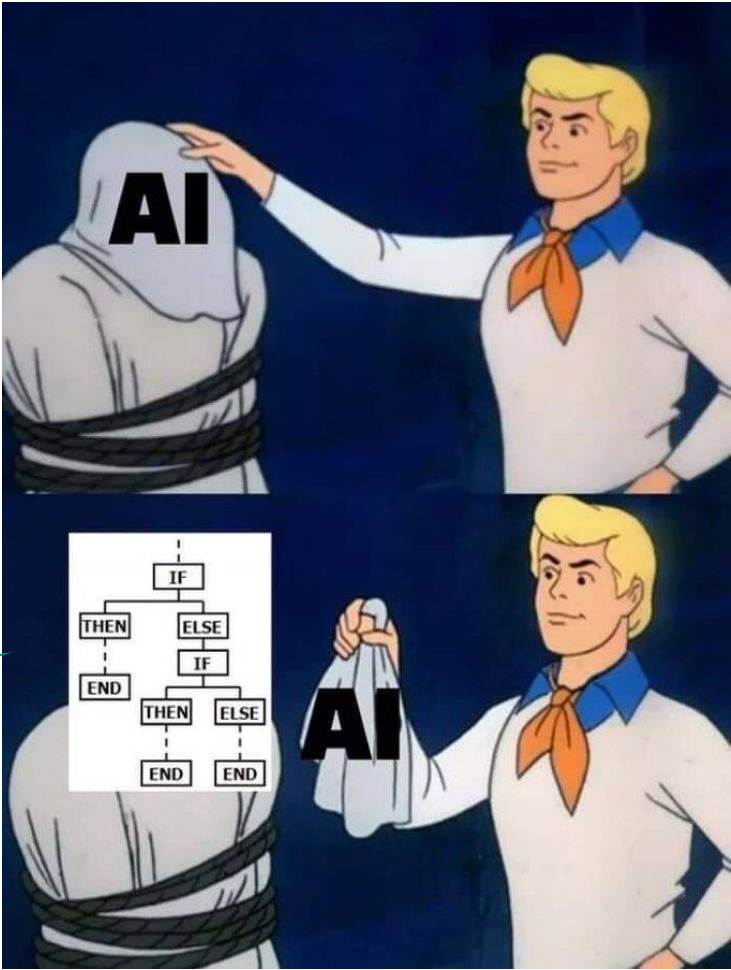
ML Influences

Data quality

Data type

ML model selected

Bitrate



ML Metrics

Accuracy



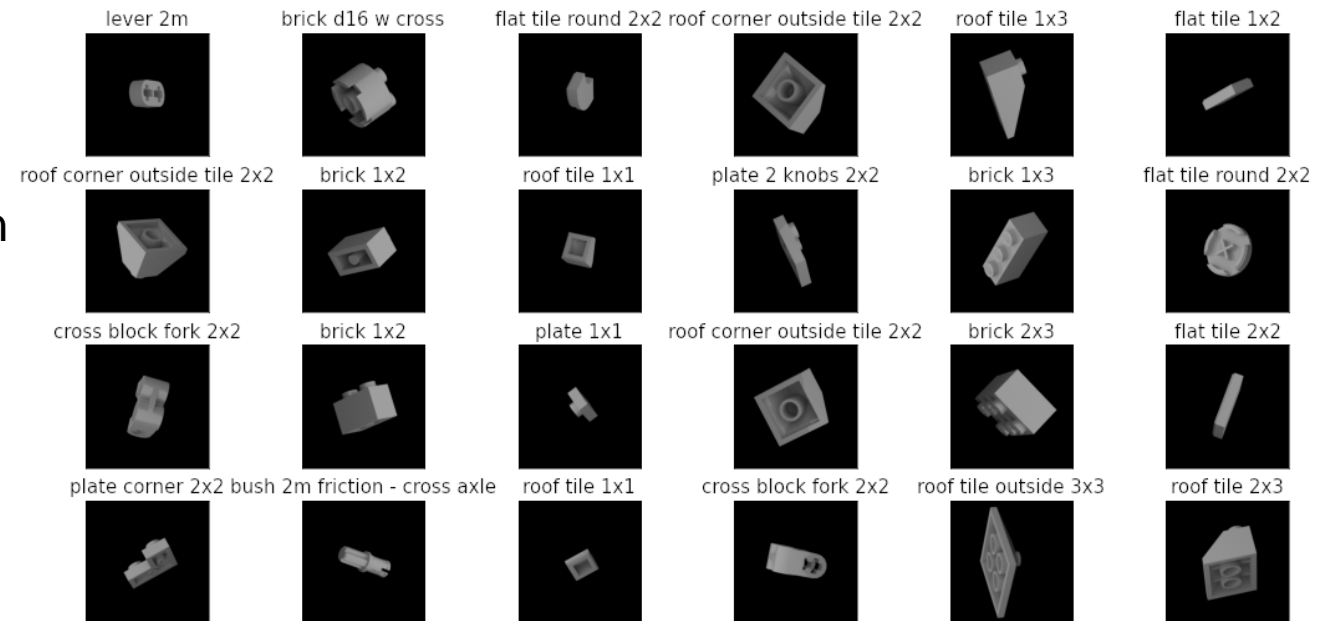
Resource requirements

➔ When does garbage in become garbage out?!

Machine Learning Analysis - Evaluation

Test Setup

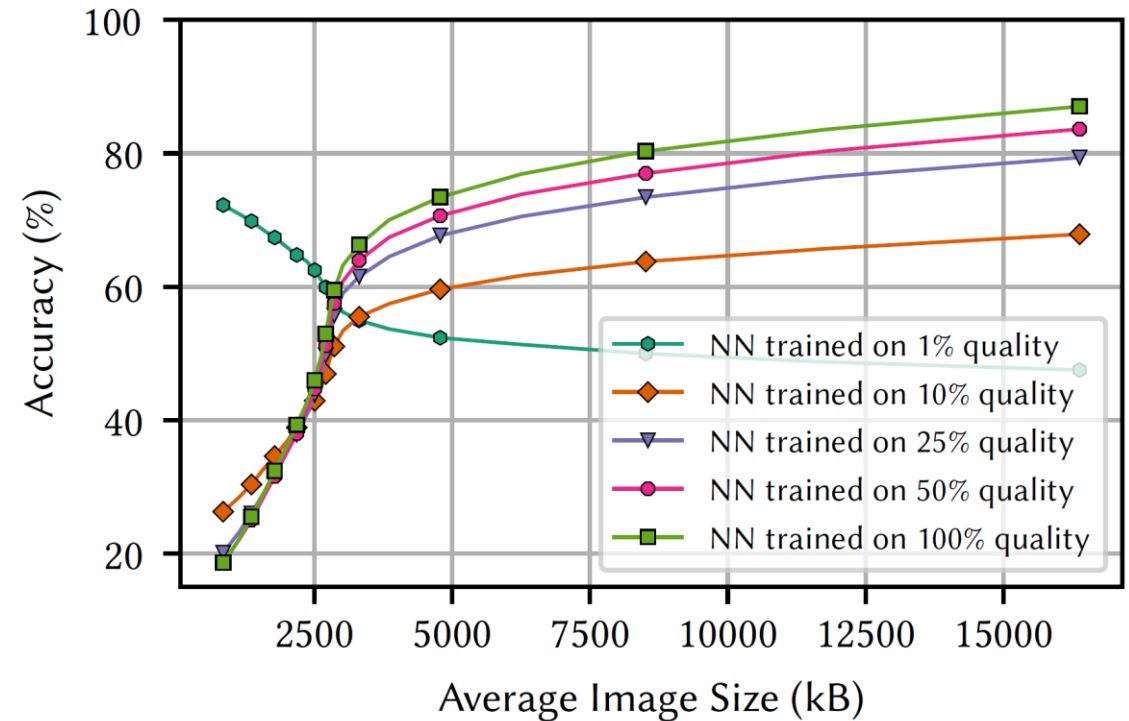
- 40000 images of lego bricks
- 50 classes of bricks
- 100/50/25/10/1% JPEG compression
- ResNet models trained to every compression stage



Machine Learning Analysis - Evaluation

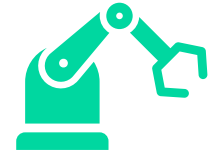
Evaluation

- Mixing bad (1%) and good (100%) images
 - Increasing mixing level from all bad to all good
 - Inference on all models for each mixing level
- Record accuracy over average image size of mix

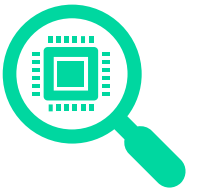


Proper Planning needs a Plan

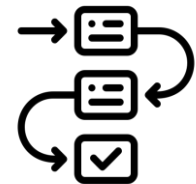
Evaluate ML-centric use cases in the factory



Analyse ML applications w.r.t their performance behaviour



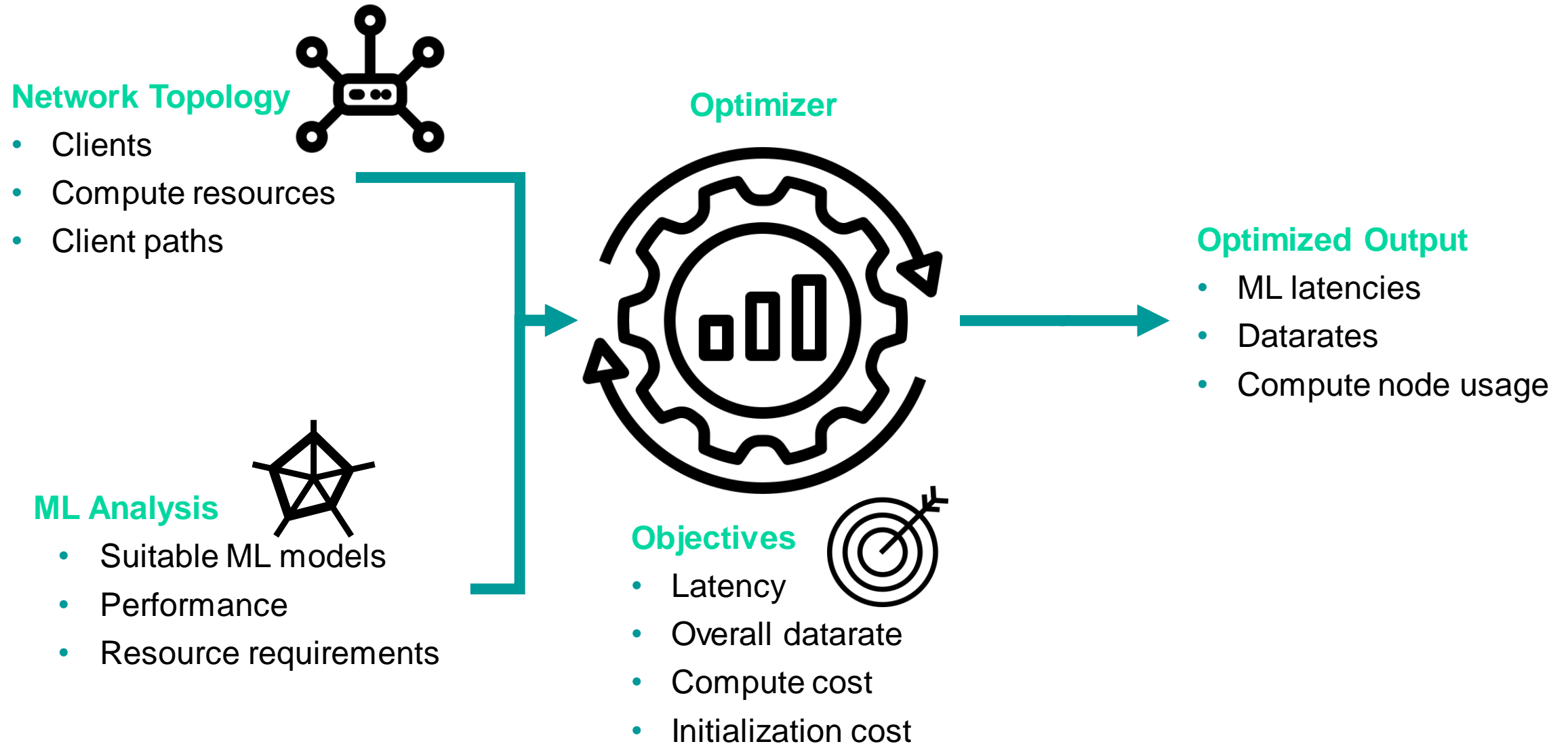
Dimension/Plan with ML applications in mind



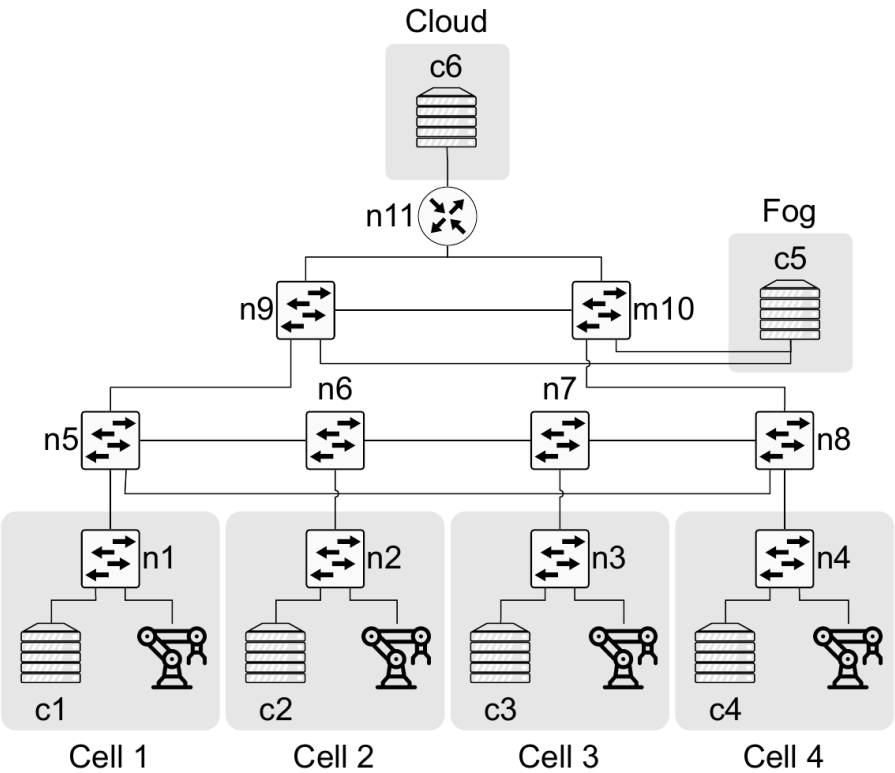
Topology Dimensioning & Planning - Introduction

- MIP to place ML applications on compute resources
 - Edge
 - FOG/Industrial datacenter
 - Cloud
- Multi-objective path-flow model between clients (Robots, AGVs, ect.) and compute nodes
 - Latency
 - Overall datarate
 - Compute cost
 - Initialization cost
- Input: ML analysis
 - Suitable ML models
 - Performance
 - Resource requirements
 - => **Weniger Bulletpoints / Grafik**

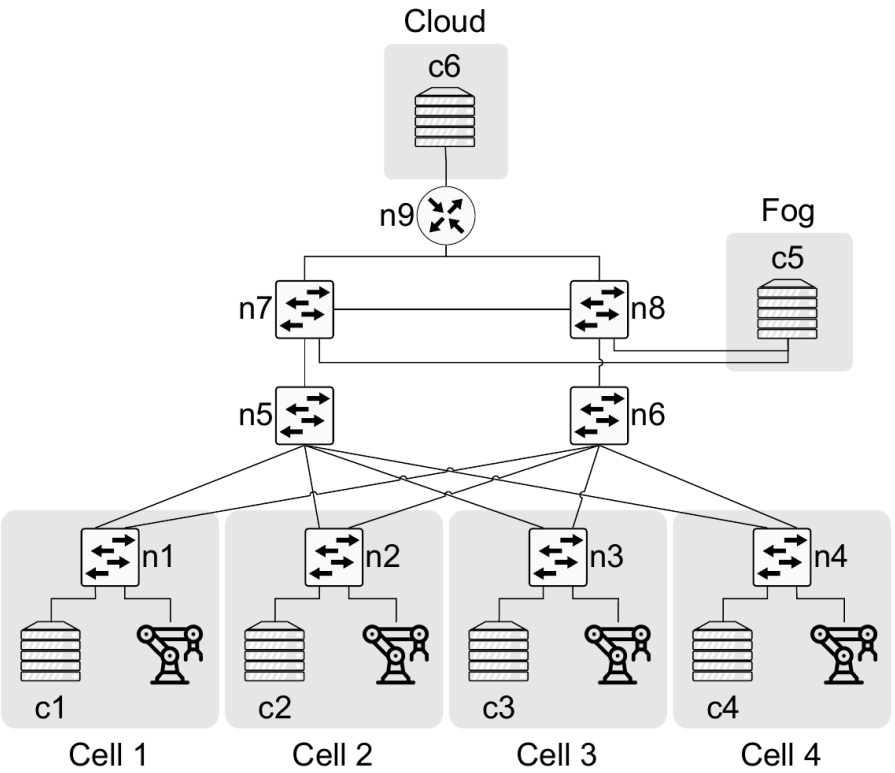
Topology Dimensioning & Planning - Introduction



Topology Dimensioning & Planning – Test Setup

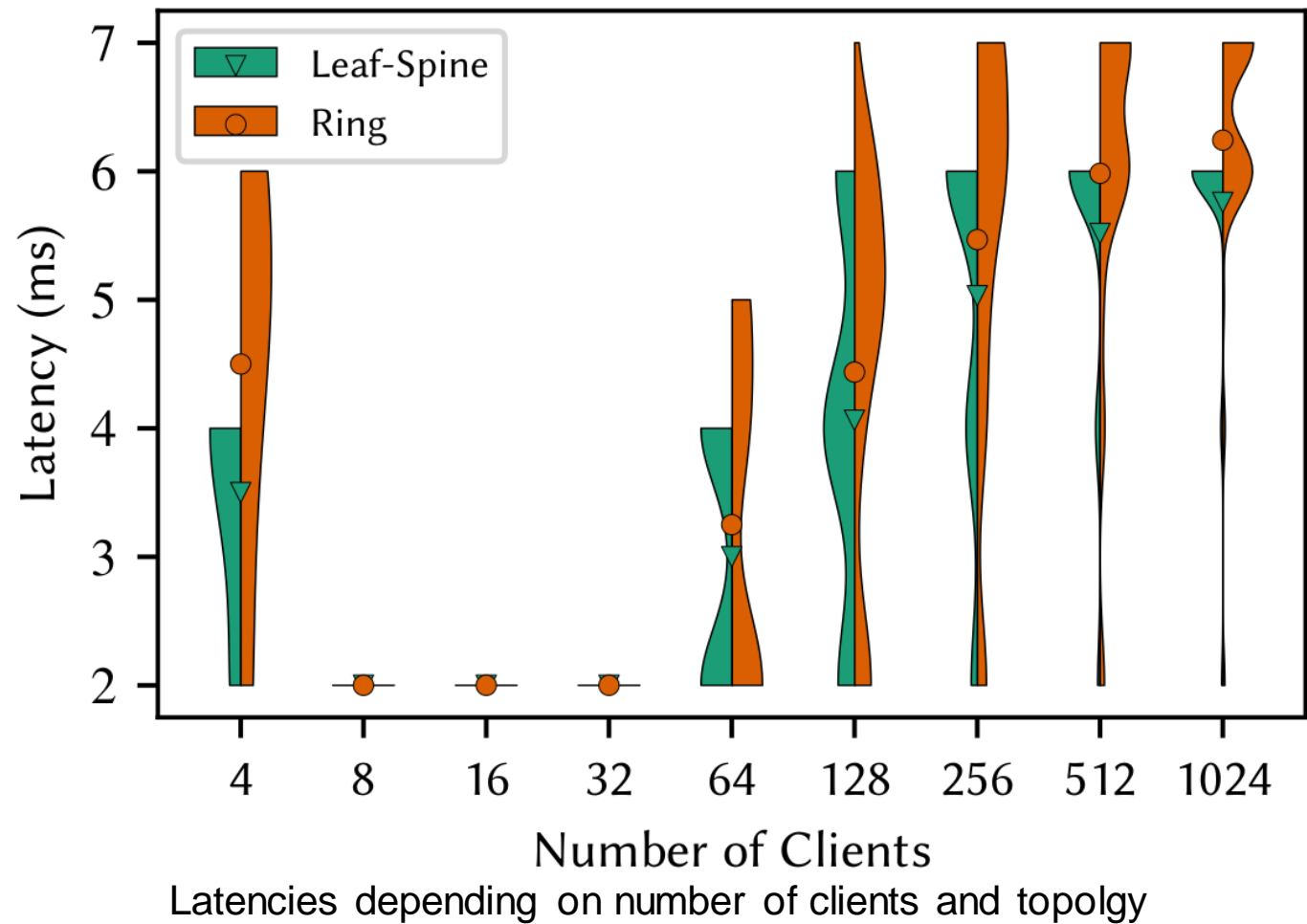


Ring Topology



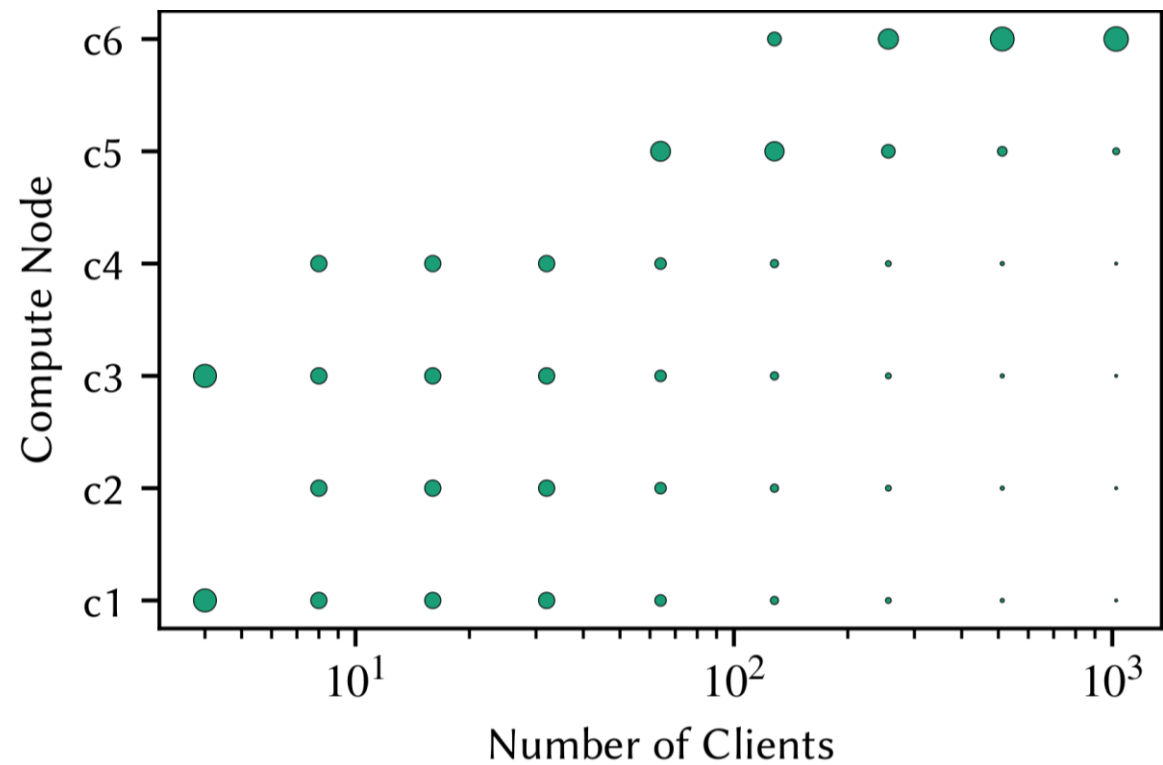
Leaf-Spine Topology

Topology Dimensioning & Planning – Evaluation

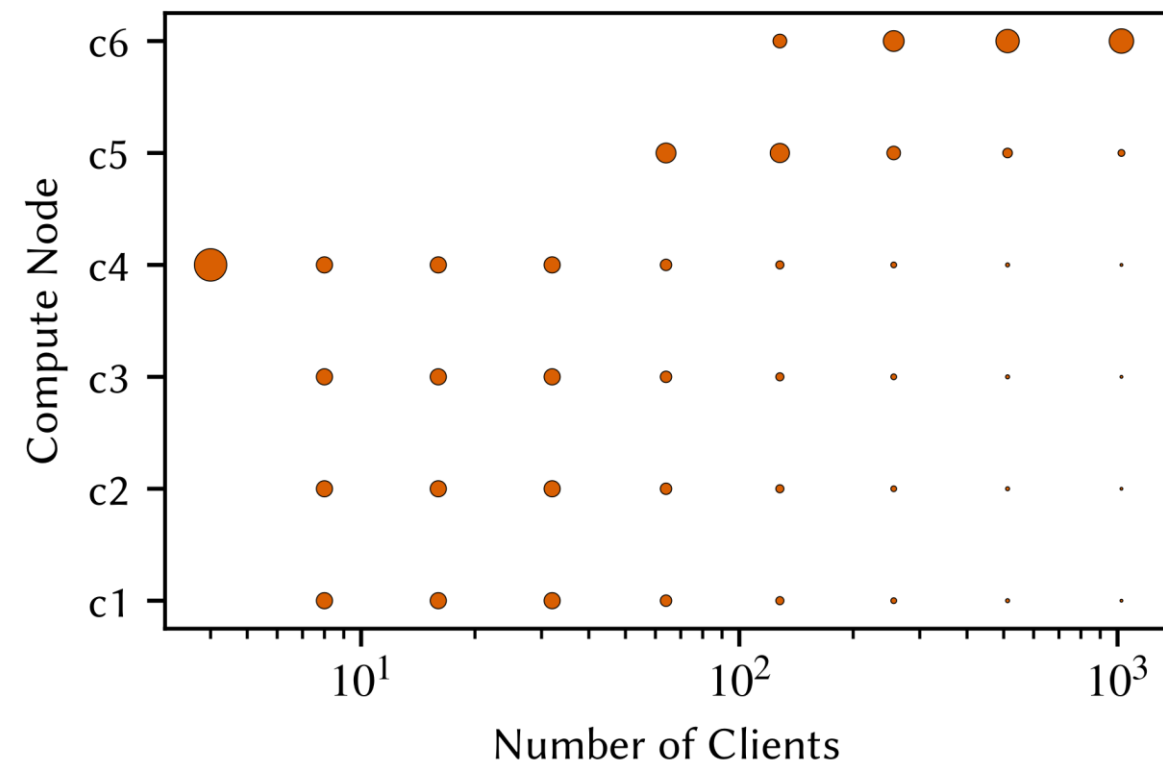


➔ Higher latencies of legacy topologies indicate inability to cope with modern requirements

Topology Dimensioning & Planning – Evaluation



Leaf-Spine compute node usage

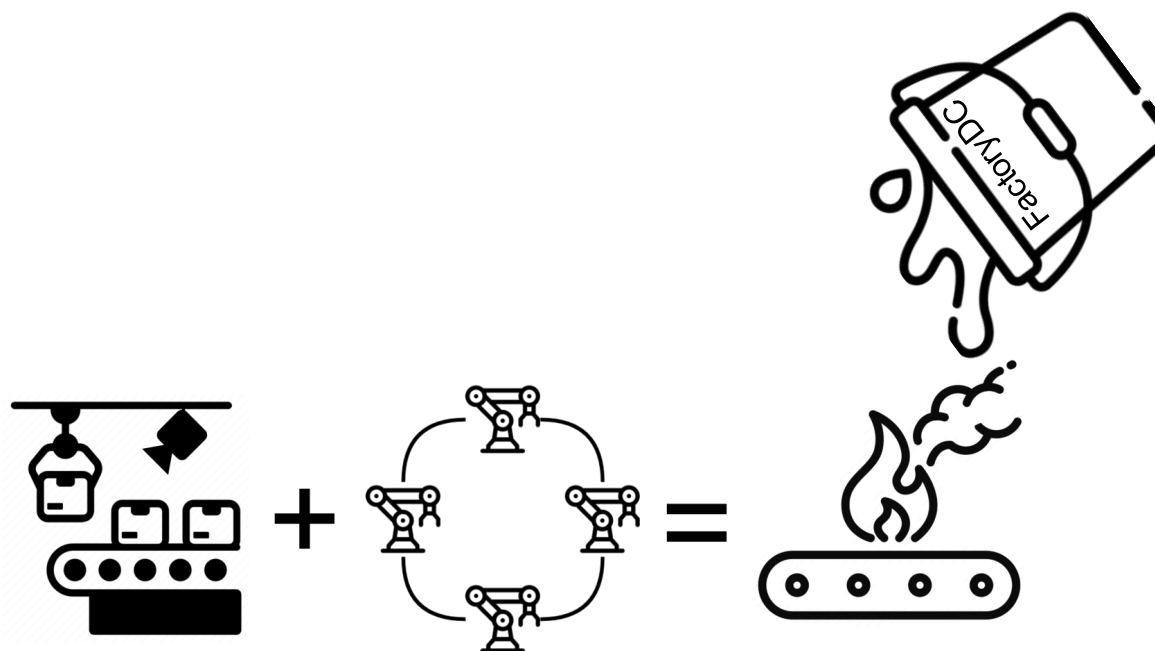


Ring compute node usage

➔ Worse compute distribution for less clients due to higher hop count

Conclusion

- ML-centric network planning essential for future factories
- New datacenter-like topologies outperform legacy networks
- Understanding ML application behaviours is key for proper planning
- Realistic dimensioning helps to save cost and improve efficiency



Future Work

- Evaluate additional network topologies
- Consider further ML applications and demands
- Increase client amount for simulation
- Introduce hard-boundaries (latency) for time critical tasks

Acknowledgements

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Thank You



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