

System architecture and ecosystem interoperability enabling AI-driven solutions

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Background

- Main work experience
 - Assistant Professor (tenure track), Tampere University, February 2018-
 - Software Engineering
 - Industry4.0 applications focusing on software and system architectures
 - Researcher, Teaching Associate, Postdoc, Academy Postdoctoral Researcher positions since 2007 at Tampere University of Technology
 - Industrial systems, data-driven solutions and analytics based services
- Education and background
 - D.Sc. (Tech.), Tampere University of Technology, 2013
 - M.Sc. (Tech.), Tampere University of Technology, 2007

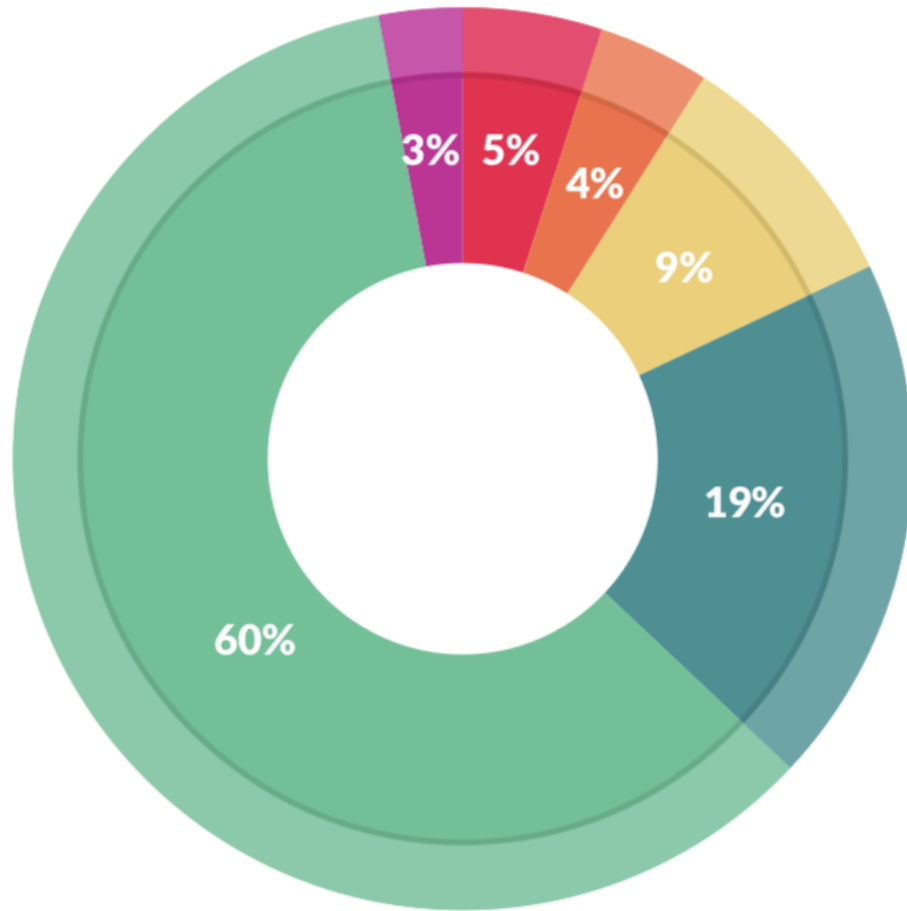
Content

- Motivation and challenge
- Architecture, interoperability
- A few examples striving for
 - Scalability and loose coupling
 - Local (distributed) reasoning
 - Flexible and continuously evolving system
- Conclusion

Typical challenges

- The data is not of sufficient quality, it is not representative
- There is not enough data
- Data is in different formats, not comparable. It is accessed differently from various systems.
- Data ownership sets limits to easily share to new use cases
- Data-driven solutions are not agile
- ...

Data is not efficiently utilized



What data scientists spend the most time doing

- Building training sets: 3%
- Cleaning and organizing data: 60%
- Collecting data sets; 19%
- Mining data for patterns: 9%
- Refining algorithms: 4%
- Other: 5%

Source: Data Science Report, CrowdFlower, 2016

Getting ready for AI (data-driven solutions)

THE DATA SCIENCE HIERARCHY OF NEEDS

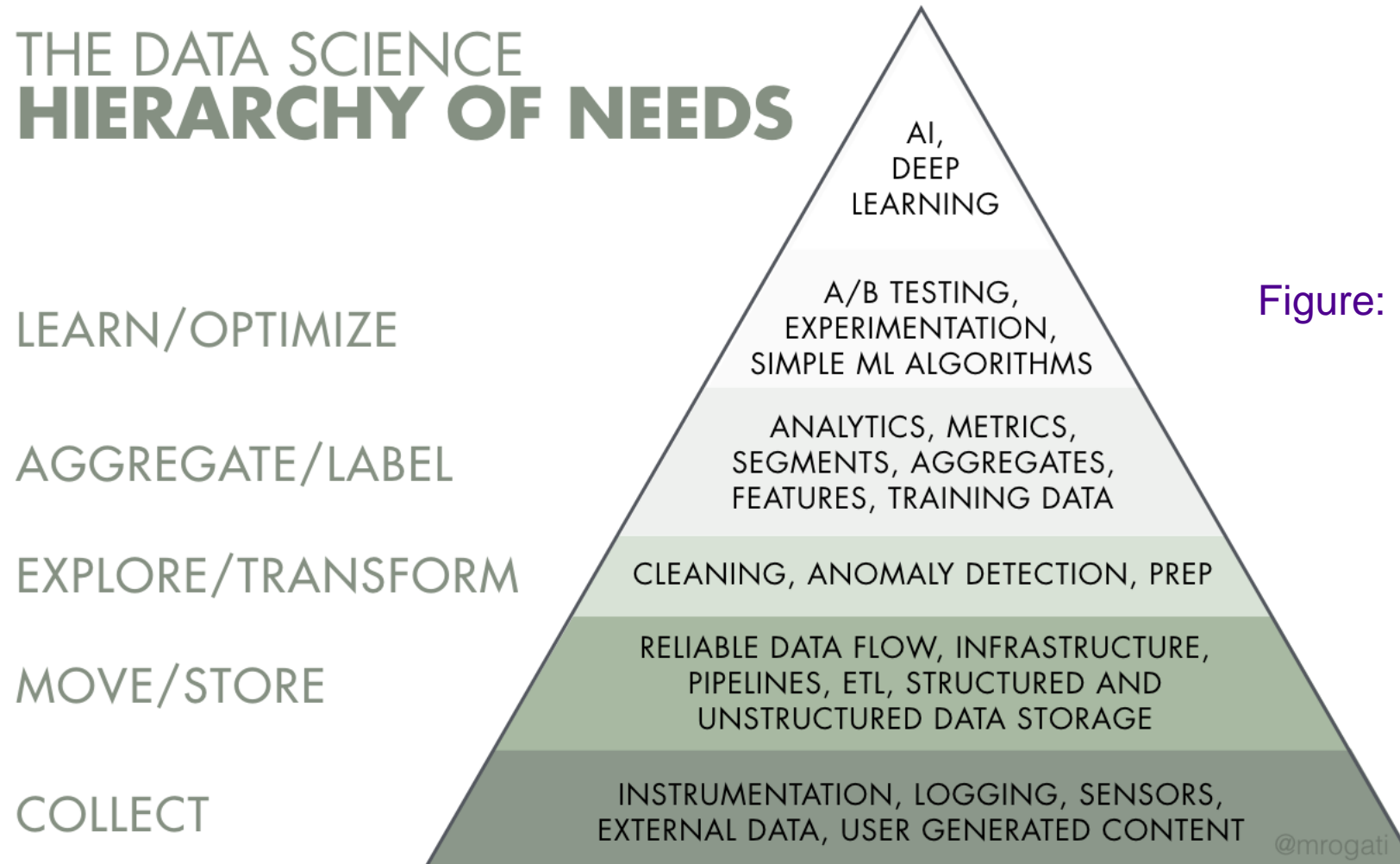


Figure: Monica Rogati

What architecture? What interoperability?

- Software and system architectures are easily overlooked when building a new system or solution in an "agile" fashion
 - Division and structure of modules, extendability
 - Interplay of components
 - Scalability, performance
 - ...
- Interoperability spans from data encoding to communication patterns between systems of systems
 - On an ecosystem level you could basically negotiate contracts on the go

Large-scale data exchange for evolving needs

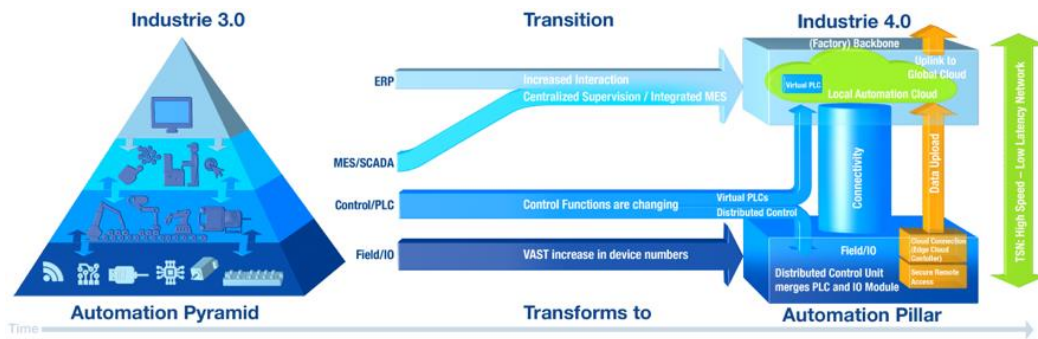
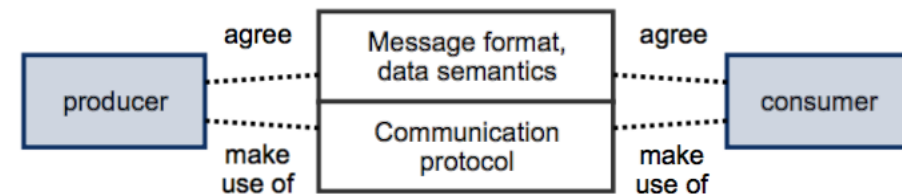
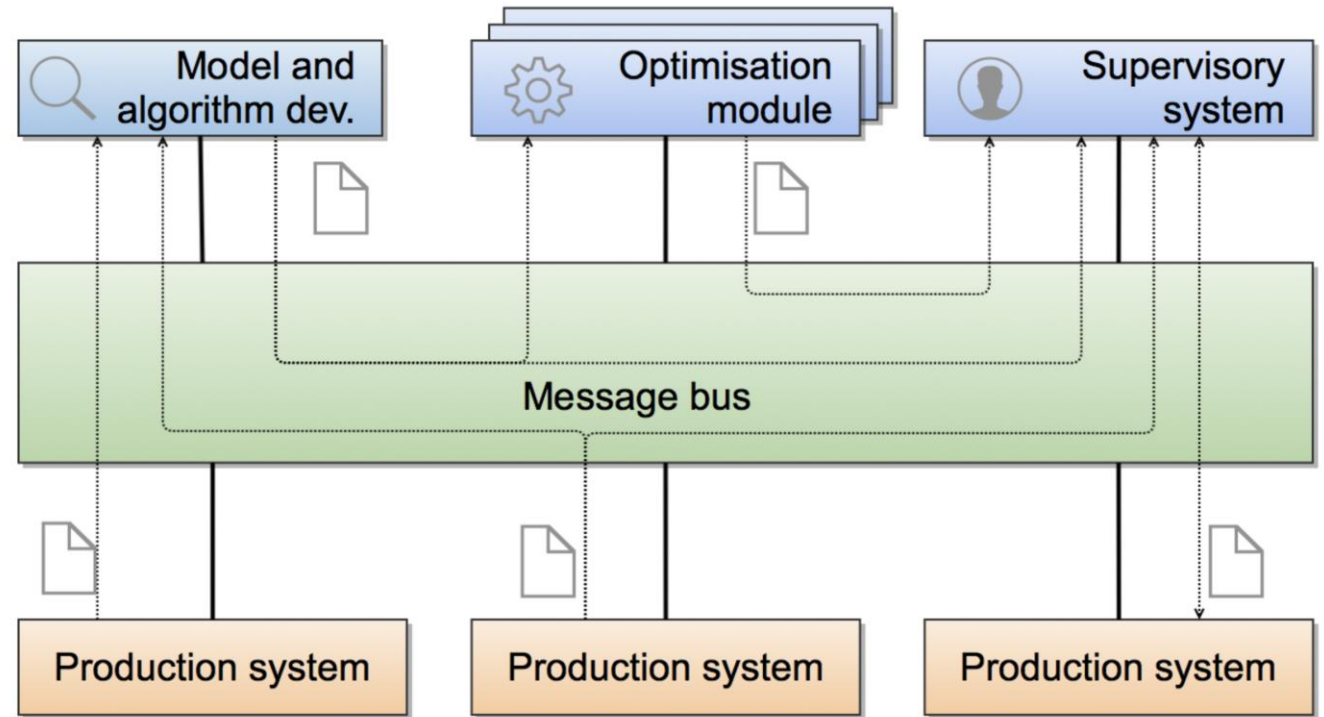
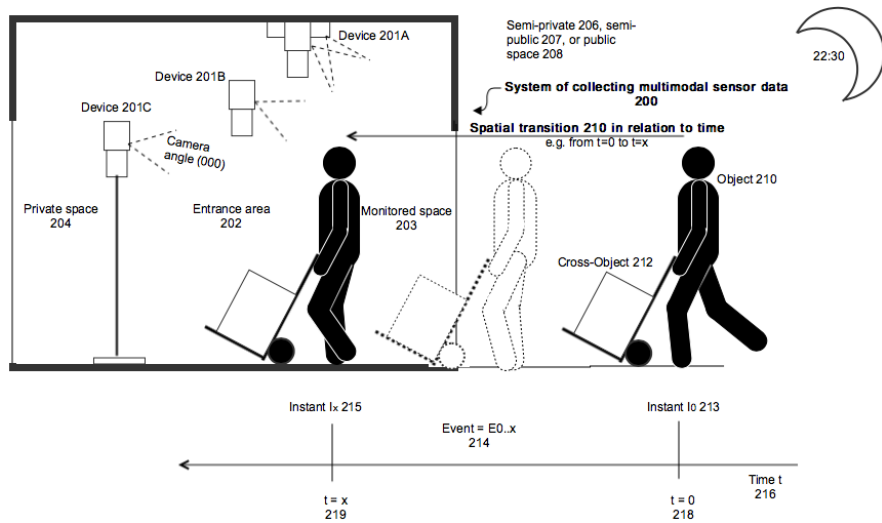
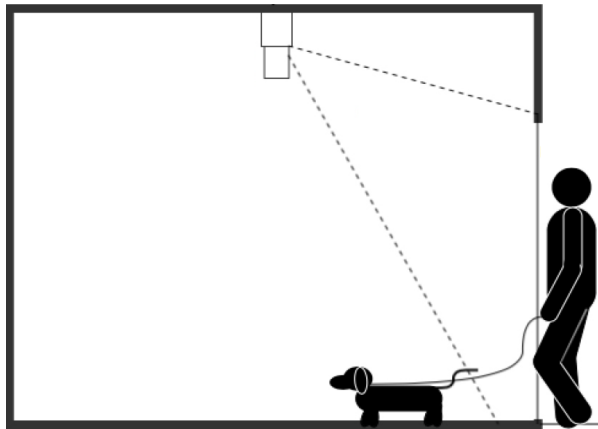


Figure: AutomationWorld

Source: Data-driven and Event-driven Integration Architecture for Plant-wide Industrial Process Monitoring and Control, Hästbacka, D., Kannisto, P. & Vilkkö, M., IECON 2018 - 44th Annual Conference of the IEEE Industrial Electronics Society

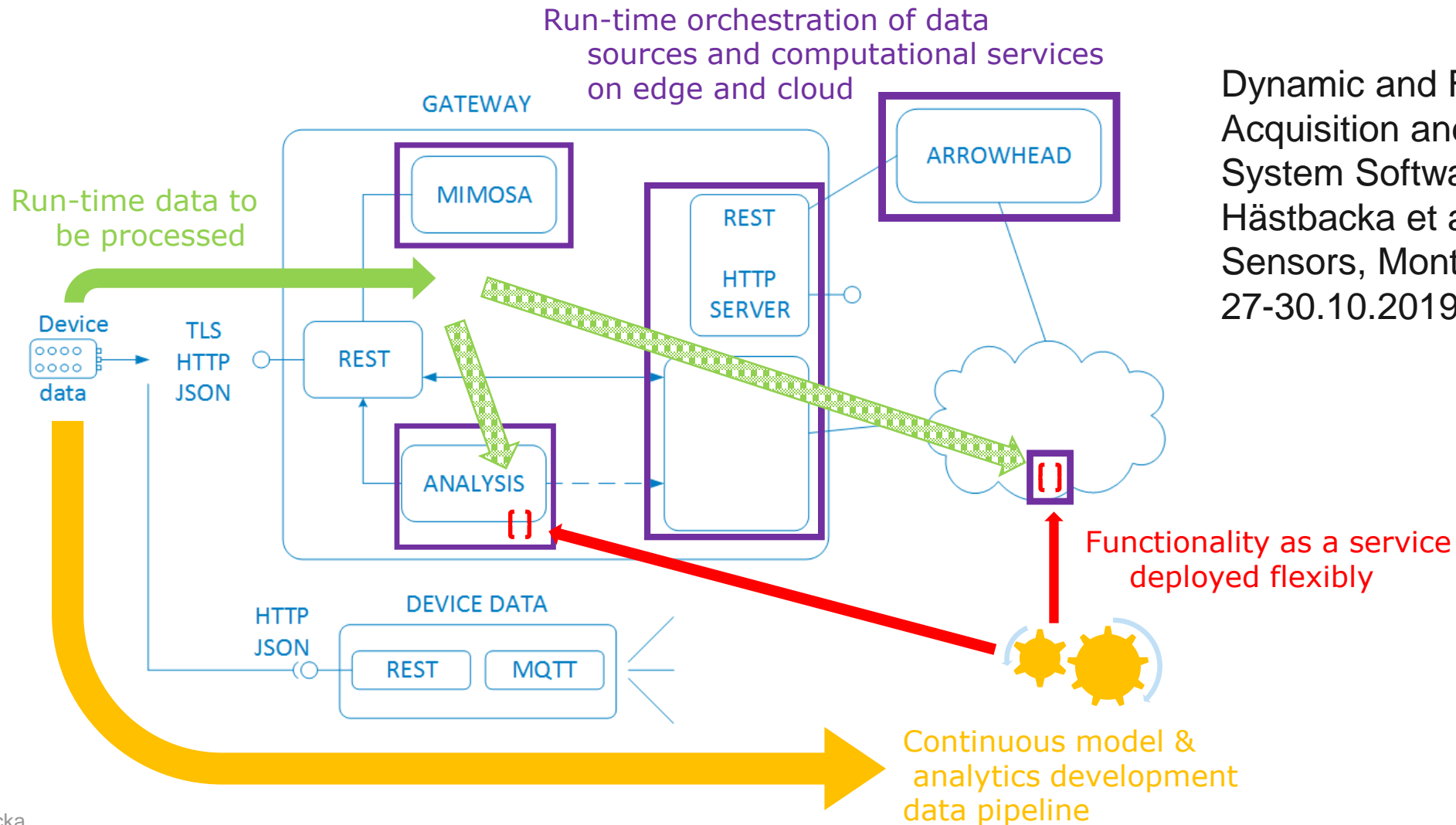


Local reasoning for control and forecasting



AIDOMUS non-intrusive sensing for understanding use of space

Building a flexible, system architecture



Dynamic and Flexible Data Acquisition and Data Analytics System Software Architecture, Hästbacka et al., IEEE Sensors, Montreal, Canada, 27-30.10.2019

Conclusion

- AI based solutions need a solid architecture
 - Purposeful, scalable and extendable (open)
 - Physical (virtual) architecture \neq logical architecture
- Ecosystem interoperability
 - Not only APIs – interoperability is needed on all levels of communication!
- Ideal future AI-driven solutions
 - Are capable of tapping into data in a plug and play manner
 - Evolve over time (i.e. a model is not only deployed and forgotten)
 - Support decentralised AI computation

Thank you!

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