



Smart Maintenance for Smart Mobility

Experience, Perspectives &
Challenges

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AGENDA

Smart Maintenance for Smart Mobility

- What is Smart Mobility ?
- Why does Smart Mobility require Smart Maintenance ?
- From single Asset to Complex System to Fleet
- Opportunities and Challenges

Alstom Digital Mobility

Alstom has evolved from rail equipment manufacturer to mobility solutions provider



Smart and Sustainable Mobility for tomorrow's cities

A number of innovative cities around the world have a “Smart City” program.
Example: Singapore (Land Transport Authority, LTA). Some challenges:

- Alleviating Congestion :

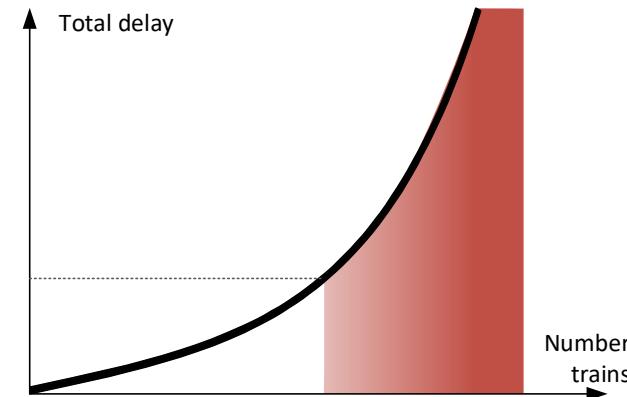
- ▶ City population growth and ecological concern → urban public transport demand expected to rise 50% by 2050
- ▶ Today, most urban rail networks are saturated at peak hours → knock-on delays

- Multi-modality :

Today, switching from one mode to another in case of incident is not easy.
Last mile usually a problem. Little offer for alternative connections

- Seamless chain between pedestrians/ cycles/busses/trains
- Inclusion of autonomous vehicles as complementary or alternative mode.

- Sustainability: optimizing life-cycle cost.



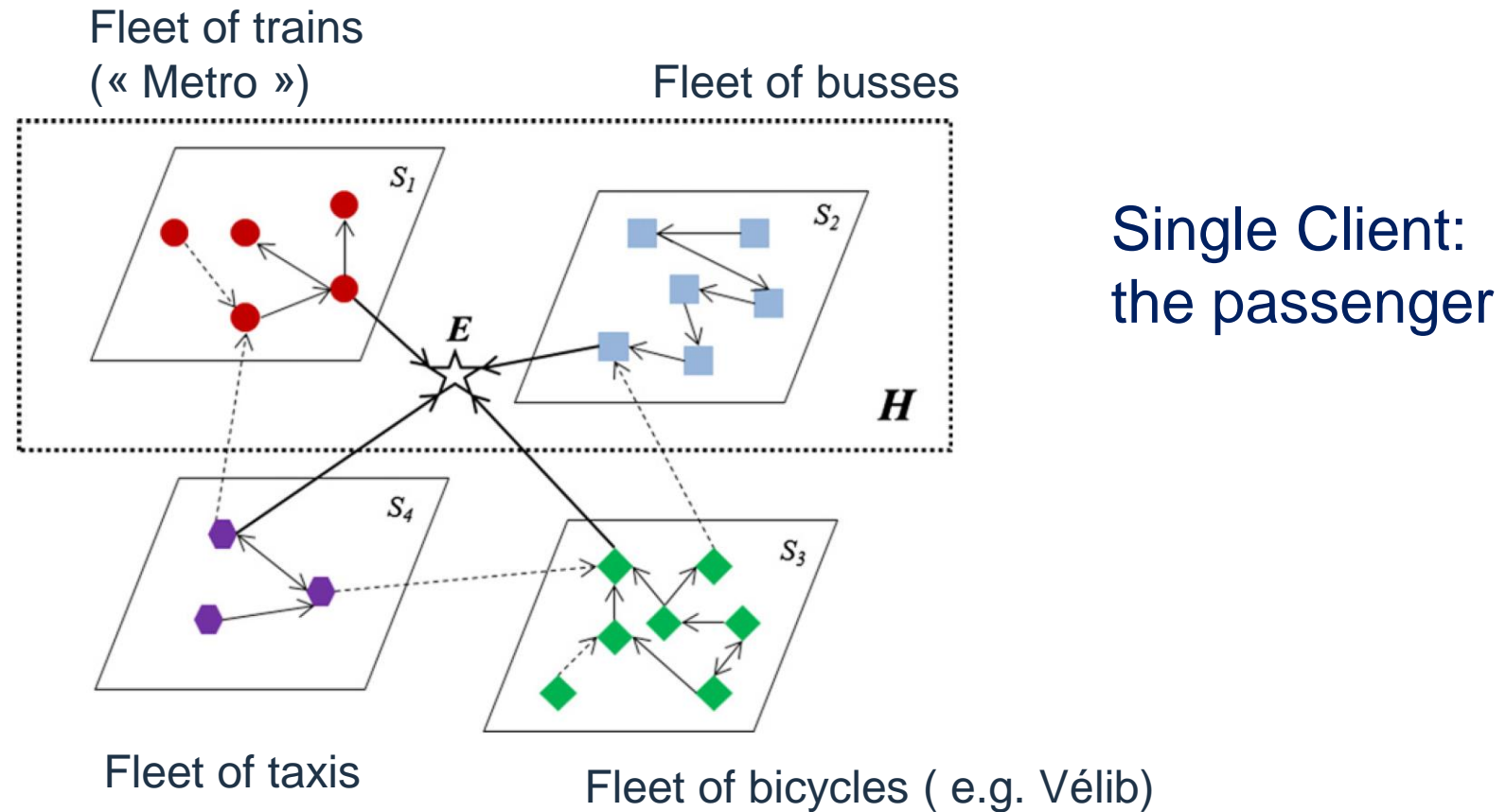
What is Smart Mobility ?

Three desirable key characteristics:

- **Flexibility:** Multiple modes of transportation allow travellers to choose which ones work best for a given situation-> resilience
- **Efficiency:** The trip gets the travellers to their destination with minimal disruption and in as little time as possible→ punctuality
- **Integration:** The full route is planned door-to-door, regardless of which modes of transportation are used-> seamless multimodal transport

Multi-modal transportation: a 'System of Systems'

Different systems are managed independently but must (or should) coordinate



Multi-modal Transportation: a 'System of Systems'

Five properties characterize a System of Systems in general:

Geographical
distribution

Heterogeneity

Operational and
managerial
independence

Evolutionary
Development

Emergent Behavior

→ Unintended consequences

Field of study since the 1980s (defense industry) . Reliability, Safety, Maintenance implications still insufficiently understood (IEEE Working Group)

Why does Smart Mobility require Smart Maintenance ?

Traditional Maintenance

- **Preventive:** systematic, i.e. scheduled at predetermined times or mileages
 - Rigid , fixed maintenance plans
- **Corrective:** post-factum, i.e. after the failure has occurred → unscheduled (randomness)
 - Maintenance teams may be overwhelmed.... or idle.
 - Resources (people, tools and parts) are dimensioned by averages or worst case (peak and failures)
 - Spare parts may be in excess... or lacking
 - Asset utilization is generally far from optimized
 - Availability may be insufficient

 Absolutely not in line with flexibility requirements of Smart Mobility
→ Need for 'smart, adaptive maintenance'

What are the requirements for Smart Maintenance ?

Ideally, Maintenance must make the Assets available for when they are needed, and achieve that goal cost-effectively

--"when they are needed": depends on variable passenger demand → need to monitor and anticipate demand

-- "make assets available": → need to monitor assets and anticipate failures
→ **condition-based / predictive maintenance**
PHM (Prognostics & Health Management)

Maintenance cannot be managed separately from Operations

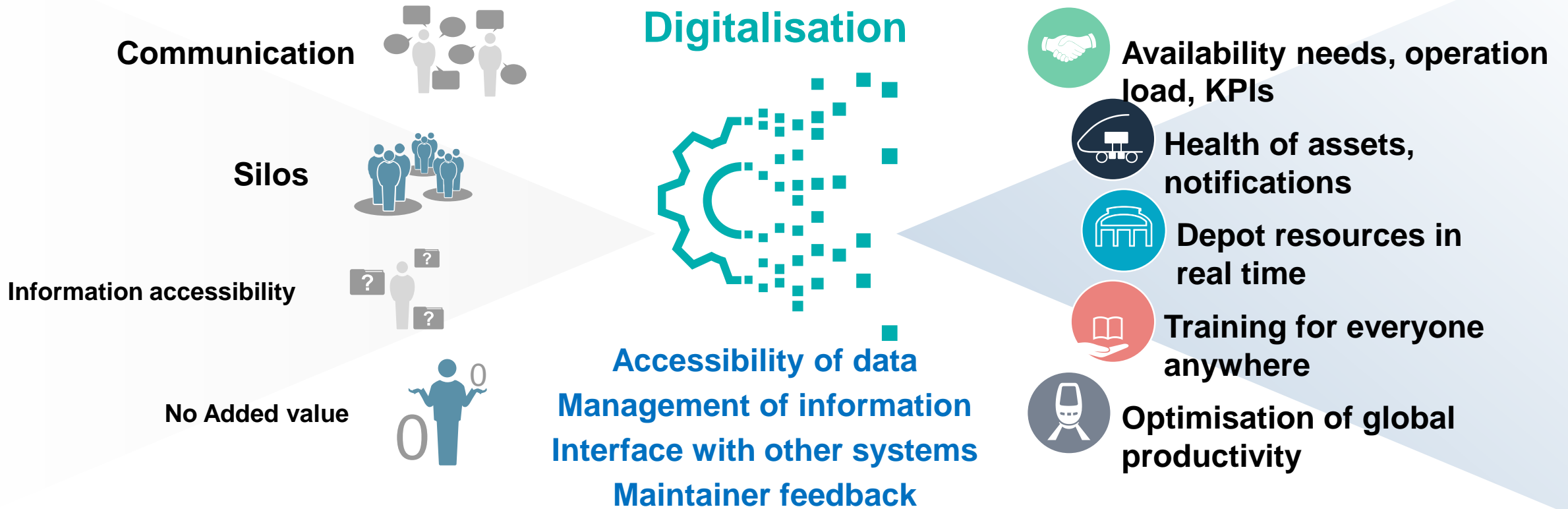


Dynamic Maintenance of a Fleet of Assets
Predictive Traffic Management

Enabler : Digitalisation is changing the world of maintenance

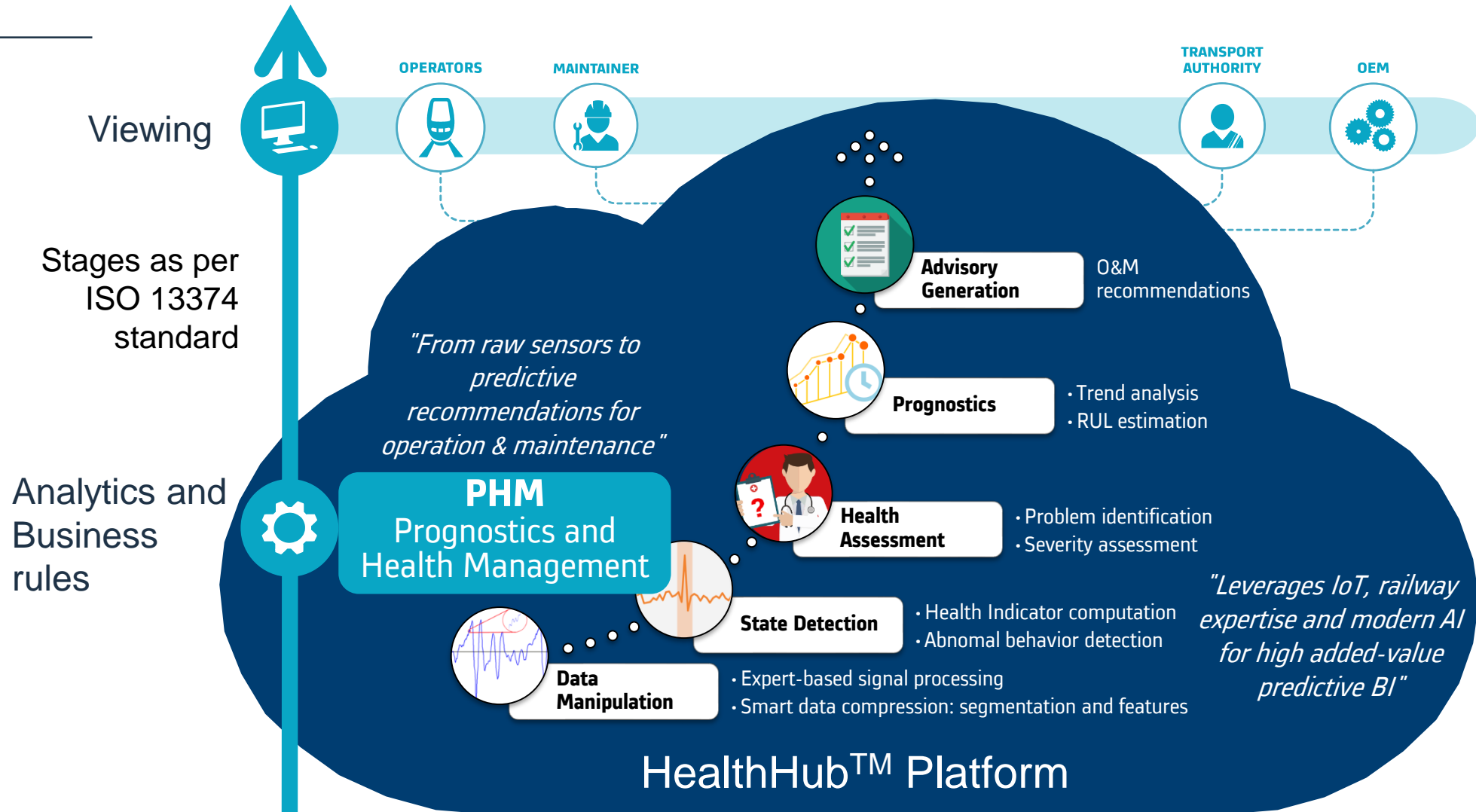
FROM

TO:



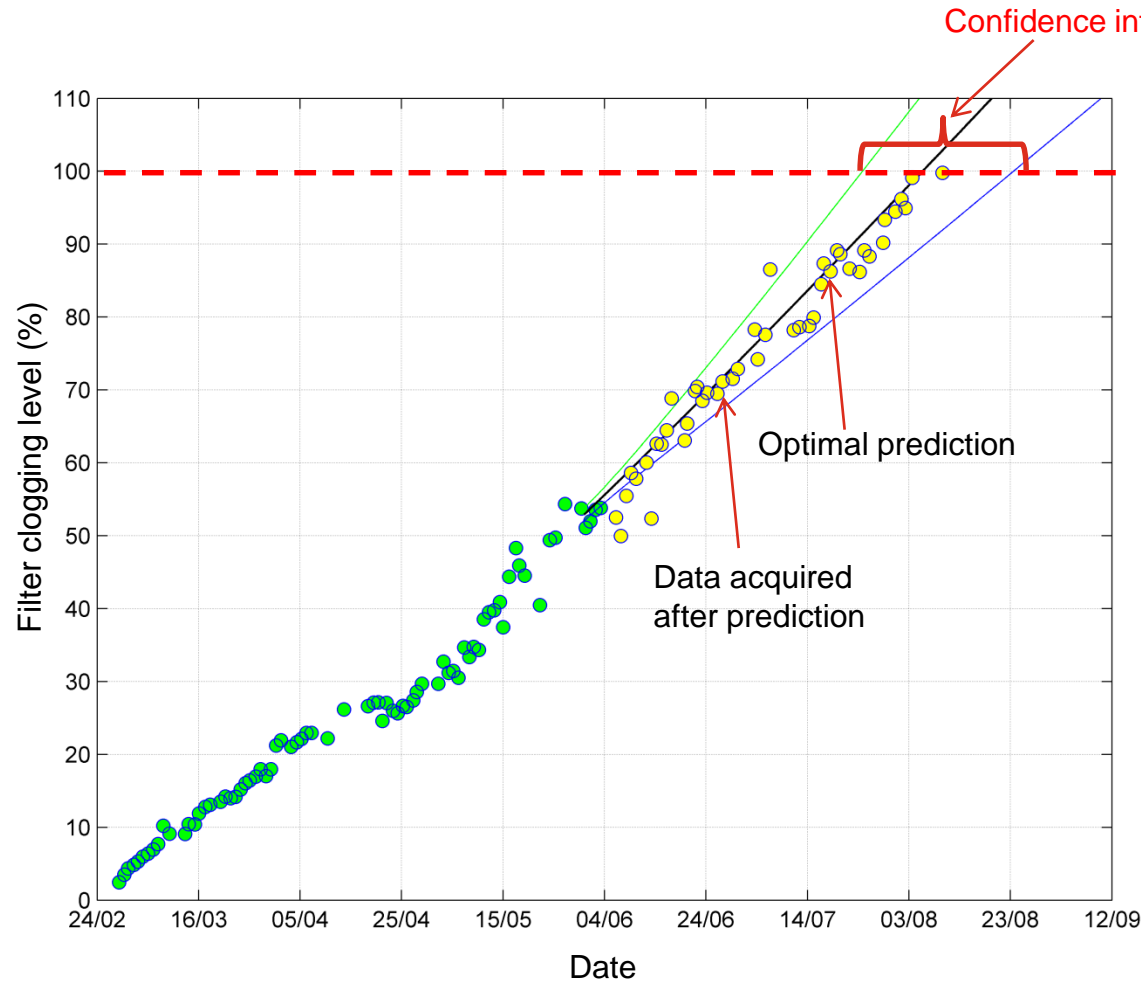
Digitalisation enables interaction between different maintenance actors and operations

Alstom's CBM and PHM (Prognostics & Health Management) Framework



RUL (Remaining Useful Life)

Example:
HVAC filter



RUL is a *random variable*

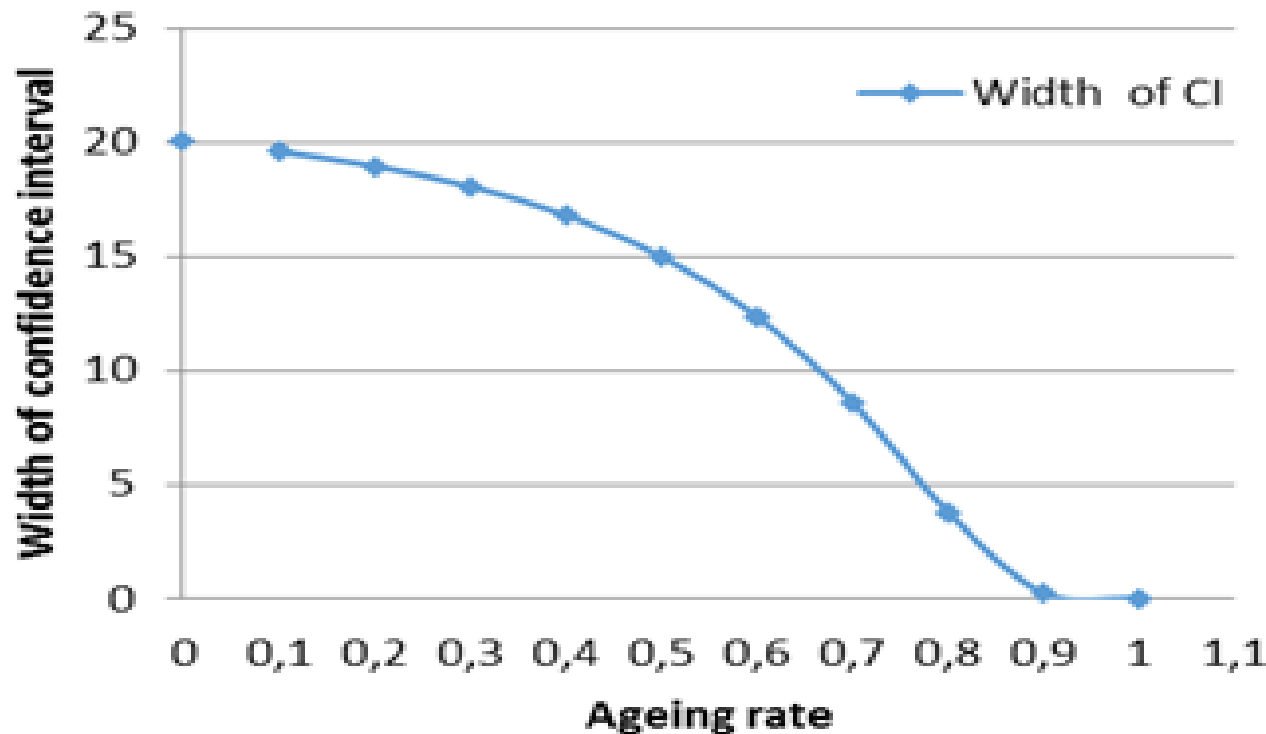
Need to deal with *uncertainty*
and *quantify it if possible*:

- model (epistemic)
- data (measurement)
- future load

Otherwise, meaningless

Confidence interval for RUL : influence of ageing rate

- ❑ The higher the “RUL Loss rate per unit of time”, or ageing rate (k), the narrower the confidence interval
- ❑ In the limit of $k \rightarrow 1$ (deterministic ageing), confidence interval shrinks to 0.
- ❑ In the opposite limit $k=0$ (exponential distribution for RUL), confidence interval width is maximal:

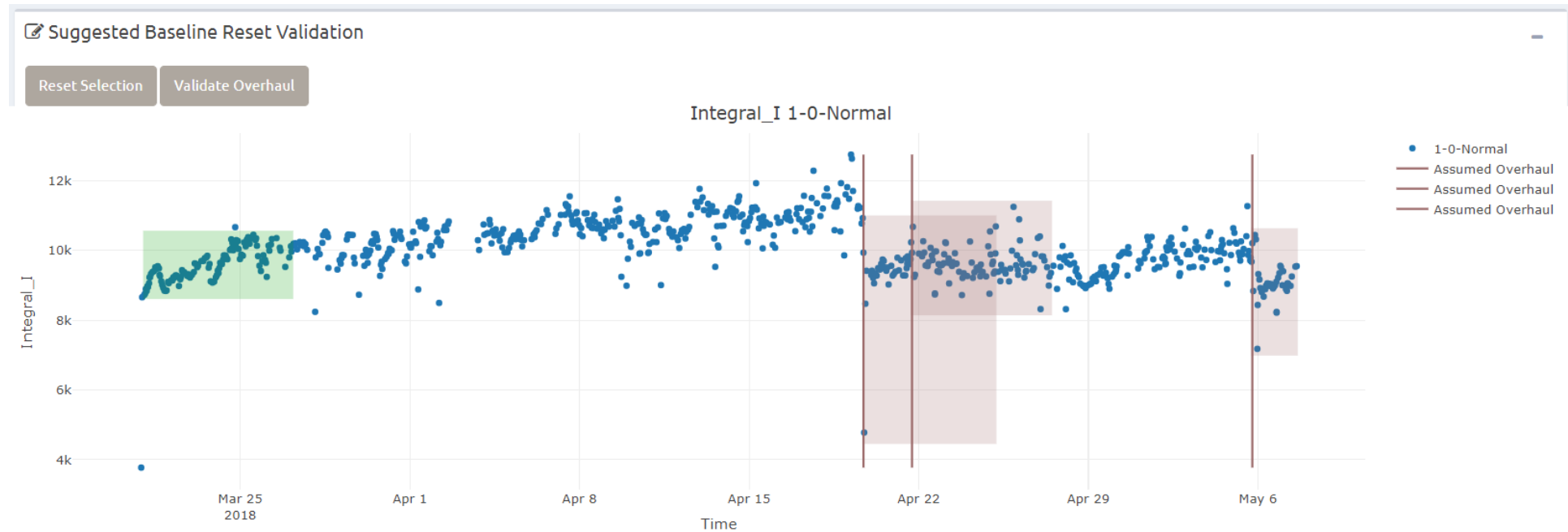


For family of distributions with
MRL linear in time

From ESREL2018 (P.Dersin, the class of lifetime distributions
with a mean residual life linear in time)

When to reset baseline ? Context evolution

When changes in context (environment but also maintenance operations) occur , algorithm has to be retrained





Key Performance Indicators

If Railway operators are to trust predictive maintenance, it is key to guarantee performance levels in terms of

- Detection (detection rate, false alarm)
 - Diagnostics (true diagnostics rate)
 - Prognostics (accuracy of RUL estimation.)
- → In spite of multiple uncertainties

HealthHub™ Fleet Support Centers



Monitoring not just individual items of equipment but complex systems

Example: Detecting, Diagnosing, Anticipating Radio Communication Issues

- **Problem:**

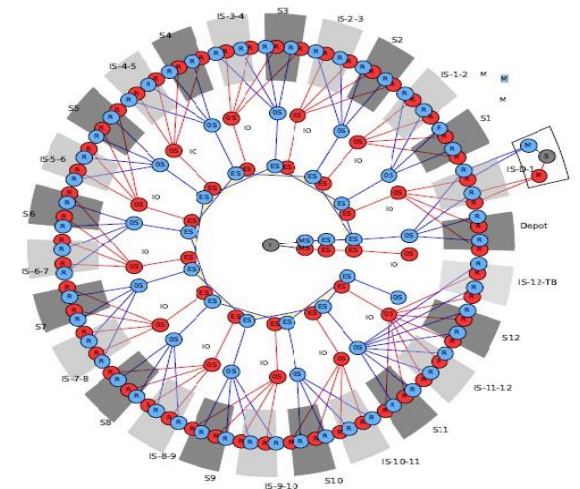
In CBTC (Communication Based Train Control) as well as ETCS (European Train Control System), data communication system is operations critical.

A number of degradations linked to radio signal quality occur, which lead to disruptions in operations, and delays:

→ Loss or delays of packets → movement authority not received on time → unwanted emergency brakes → **delays.**

- **AI/Machine learning based Solution**

→ Continuous or periodic Monitoring of radio signal strength
→ Preventive maintenance measures to avoid traffic disruptions



What is dynamic Maintenance Planning ?

From individual asset to fleet

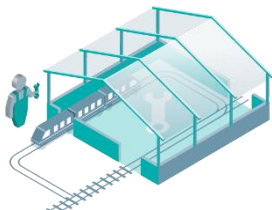
FLEET PLANNING
& MONITORING



WORKLOAD
OPTIMISATION



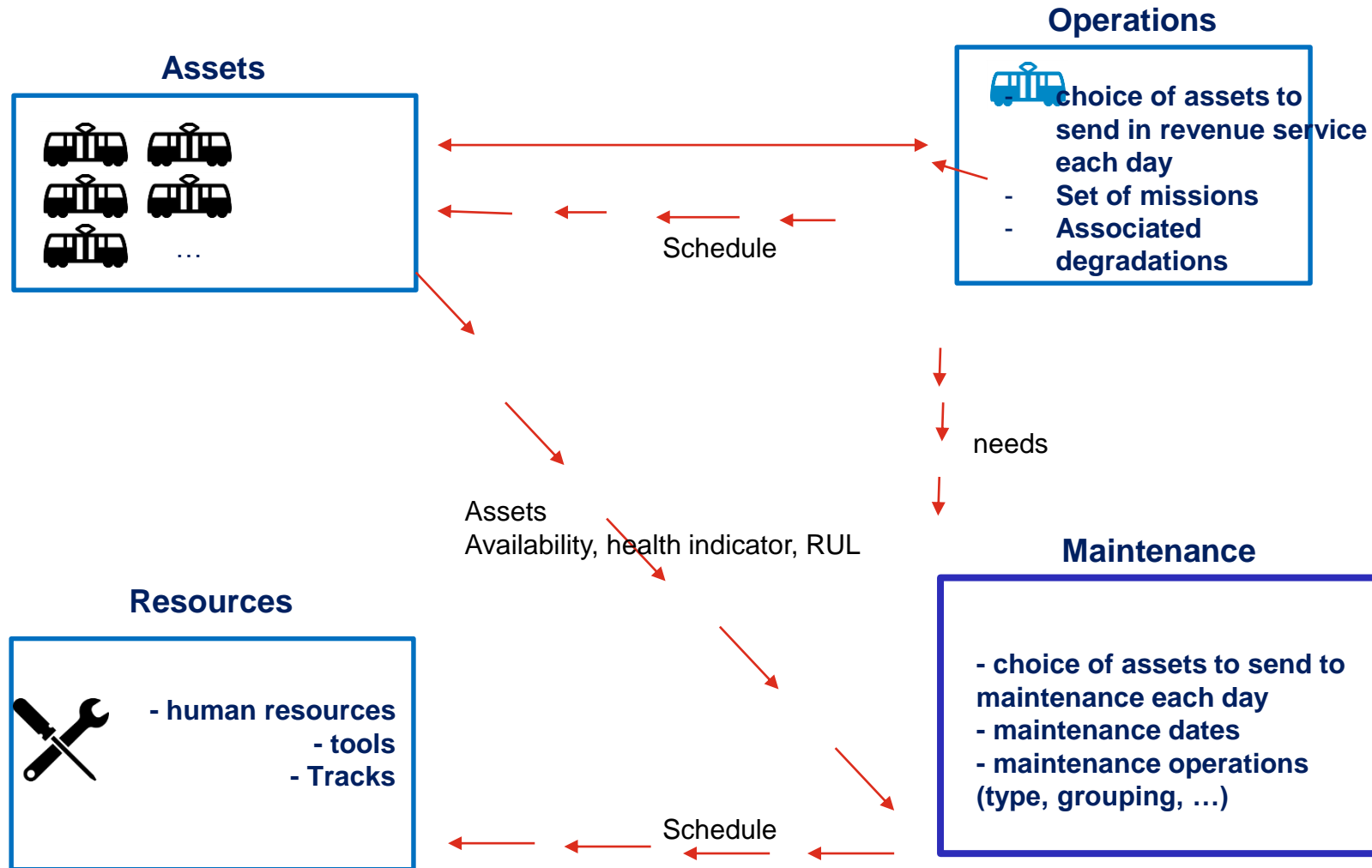
EXECUTION



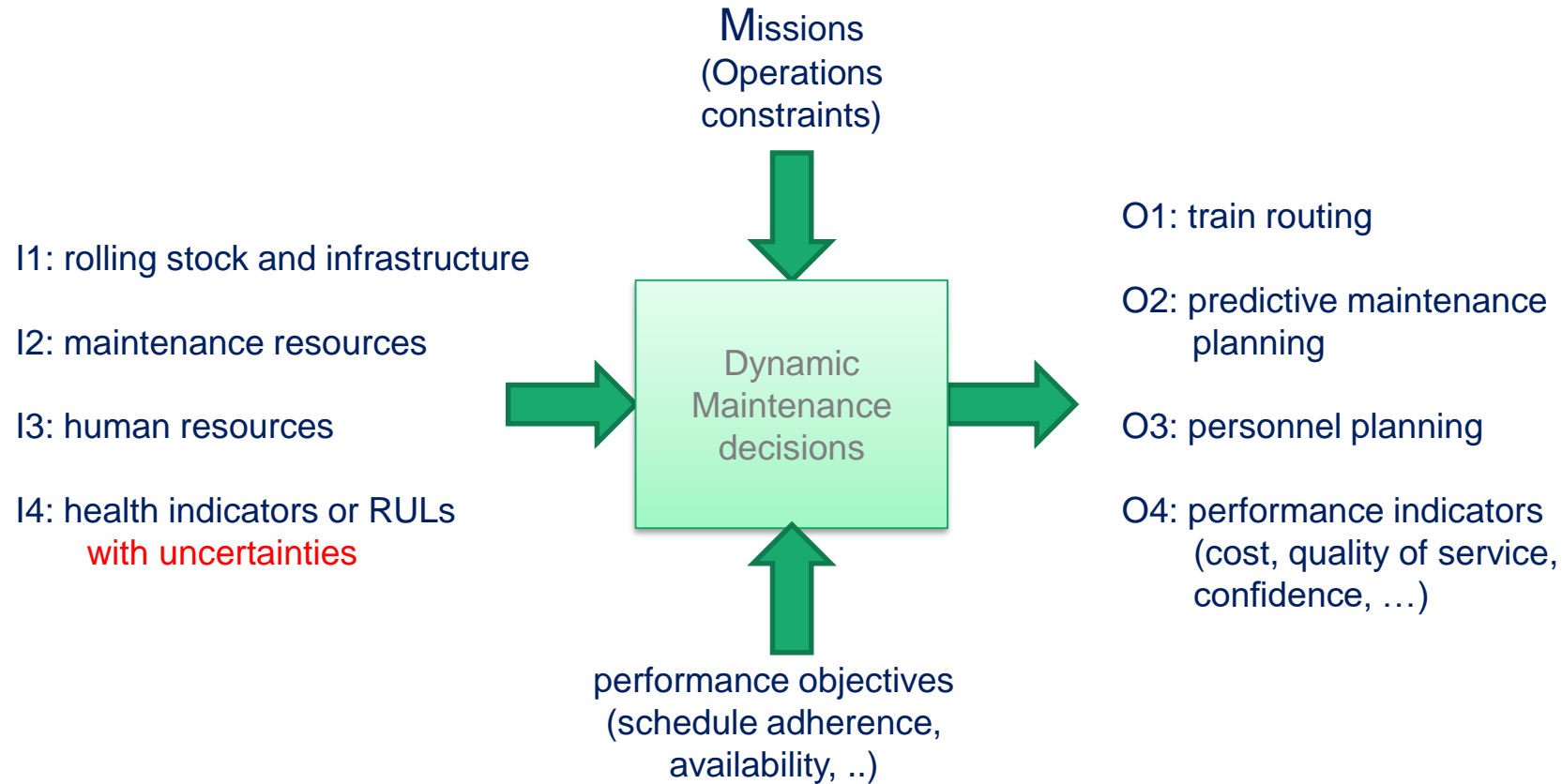
E2E Digitalisation of the operation
& maintenance execution

Fleet Level: Maintenance and Operations

Multi-agent System



Maintenance and Operations Decisions

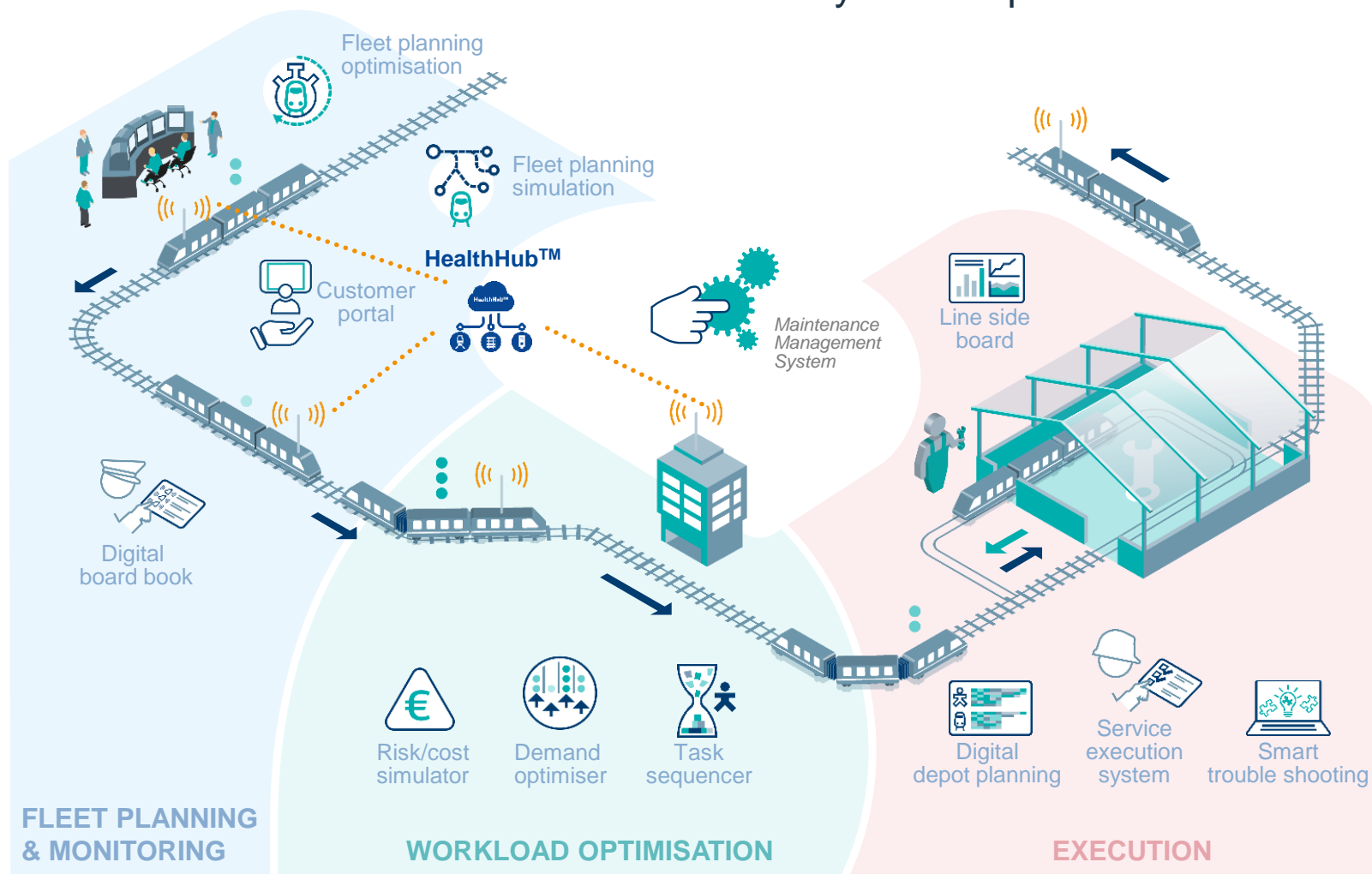


Predictive maintenance for a fleet of assets with material and human resource constraints

The Dynamic maintenance planning journey



Dynamic Optimization with Uncertainty



Enabler: Digitalisation is changing the world of Operations.

the MASTRIA® mobility platform

Multimodal coordination: all modes of transportation

Operation supervision: alternative solutions in case of incident

Predictive analytics: plan and optimize the mobility based on the demand

Facilitate MaaS ('Mobility as a Service') implementation



Public Transport



Roads



City



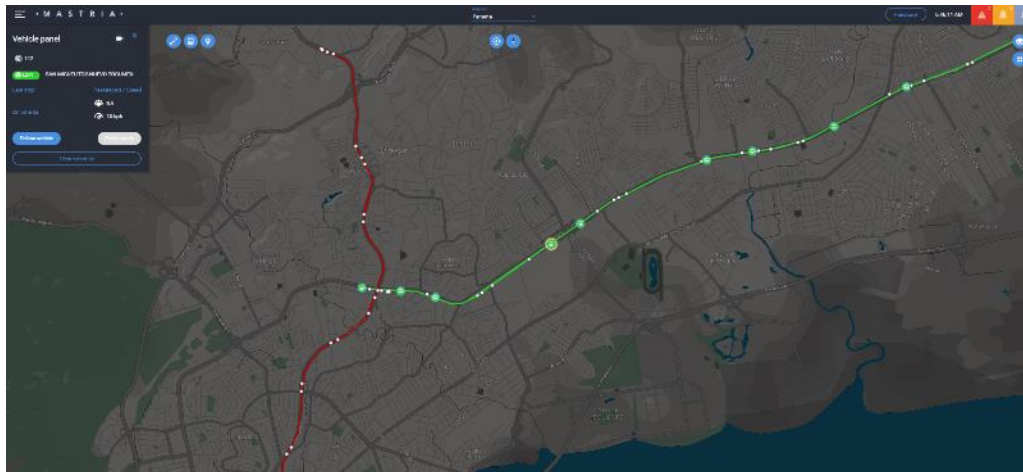
Telecom



Example of Results: network capacity optimization (Panama City)

Max "Fail to Board" from 70% to 40%

- Prediction 30 minutes in advance of "Fail to Board" alert (50%)
- Activating "direct trains" to the right stations
 - Immediate results in December 2019 – January 2020



Real-time monitoring of the trains and their load



Fail to Board at San Miguelito

3 minutes less waiting time for 30 000 passengers during peak hours
1500 hours per day -> equivalent 3M\$ GDP / year

Existing Data

Rolling stock & Signalling

- Coach, train weight & position
- Doors events & technical data

Tickets

- Validations
- People counting

Smartphone Apps

- Trip planning
- Crowd sourcing

Cameras

- Video flows
- Temperature & behavior

Telecom

- Origin-destination matrix
- Density information

« AI Powered »:

Deep reinforcement learning

Cross-data Analytics

Simulating & Planning

Analyze people flow
In trains and at stations
Anticipate demand evolution

Simulate access restriction
Measure impact on people flow

Monitoring & Predicting

Real-time people flow monitoring
In trains and stations
Predict demand evolution

React to demand spike / incidents
Activate actions at the right time

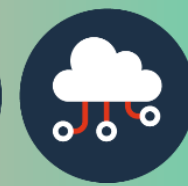
Informing Passengers



Info trains & stations

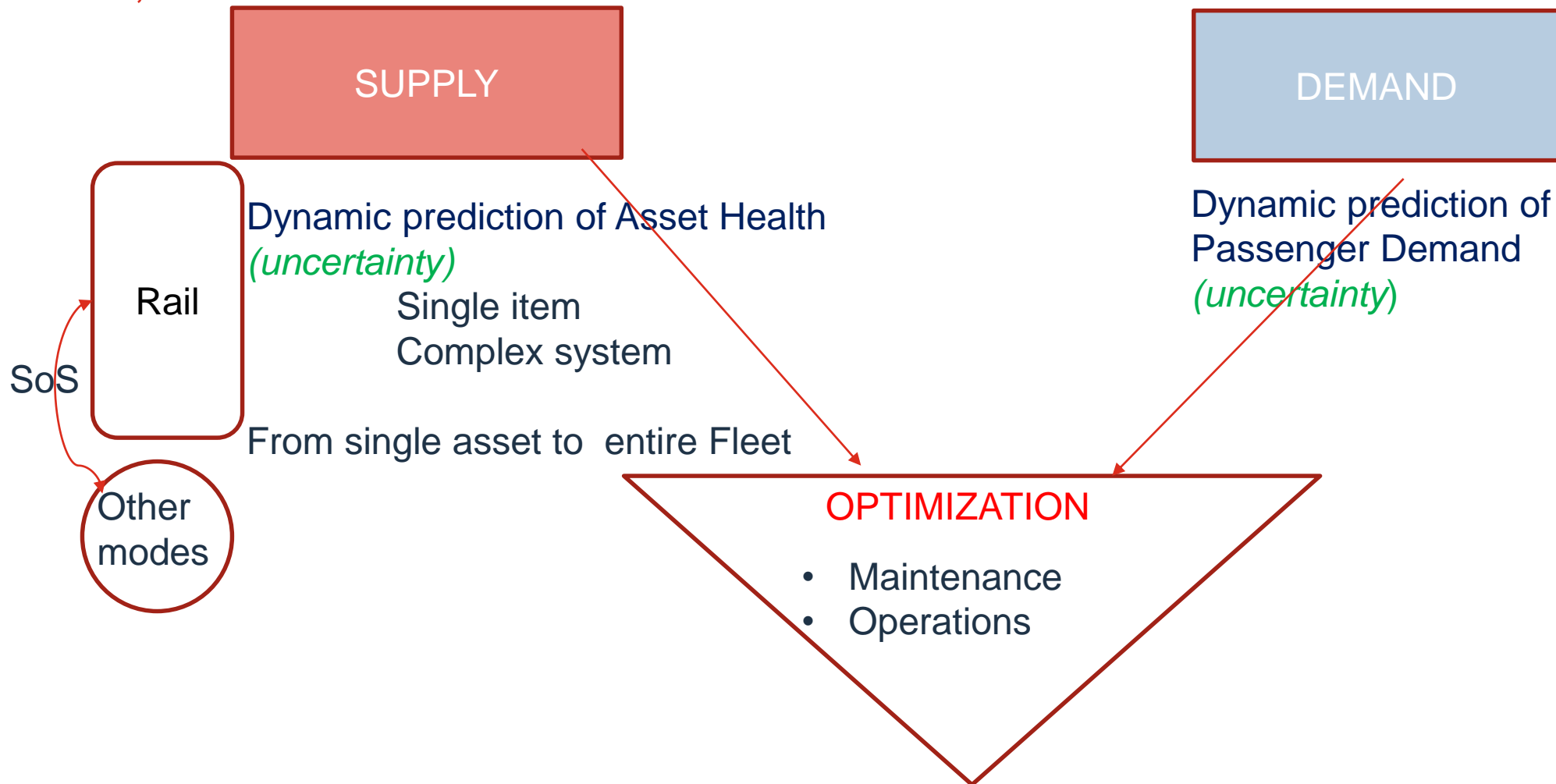


Apps



Open data

Conclusion:Global Vision



Challenges and Opportunities

Opportunities

- Leverage AI/ Machine Learning and combine with traditional optimization
 - Context-adaptive maintenance and operations policies
 - Reconfiguration algorithms learning from past experience
 - Deep Learning for PHM
- Model more explicitly the “negotiation process” between Operations and Maintenance agents→ game-theoretic approach ?

Challenges and Opportunities

Challenges

- Large –scale real time optimization of distributed systems with multiple decision makers
- Dealing with ‘Systems-of-Systems’ issues (Multimodality) : committing to performance (availability, cost) despite emergent behaviour→ Resilience
- Dealing with uncertainty, and uncertainty propagation (health indicators, RUL)
- Should decisions be
 - centralized : upload all health raw data to the cloud ?
 - or decentralized: bring intelligence (AI) down to the assets ?
- Assuring Cybersecurity and Safety

Some Bibliographical References

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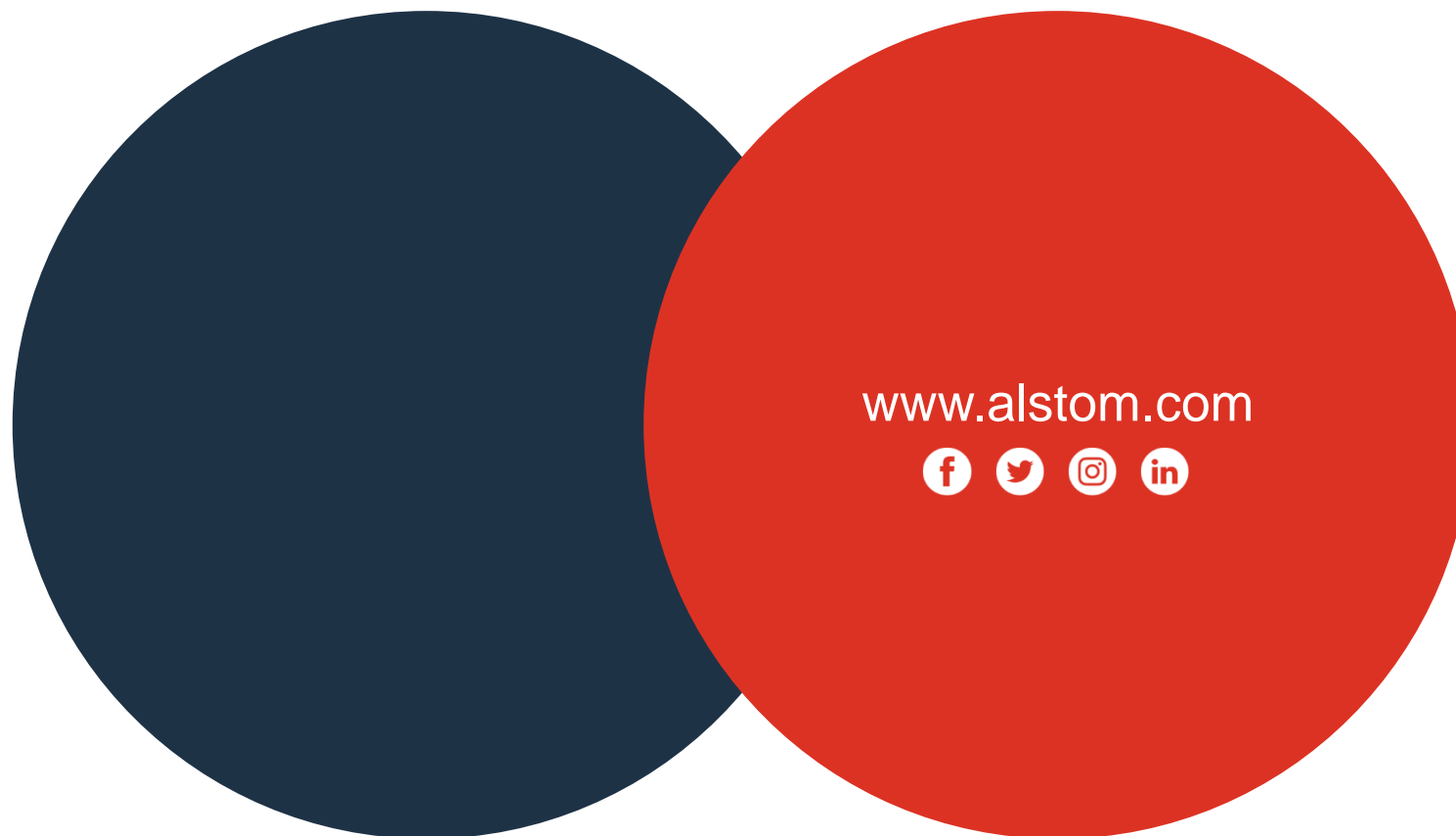


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