



# From Single-Asset Health Monitoring to dynamic Fleet Maintenance

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# Presentation outline

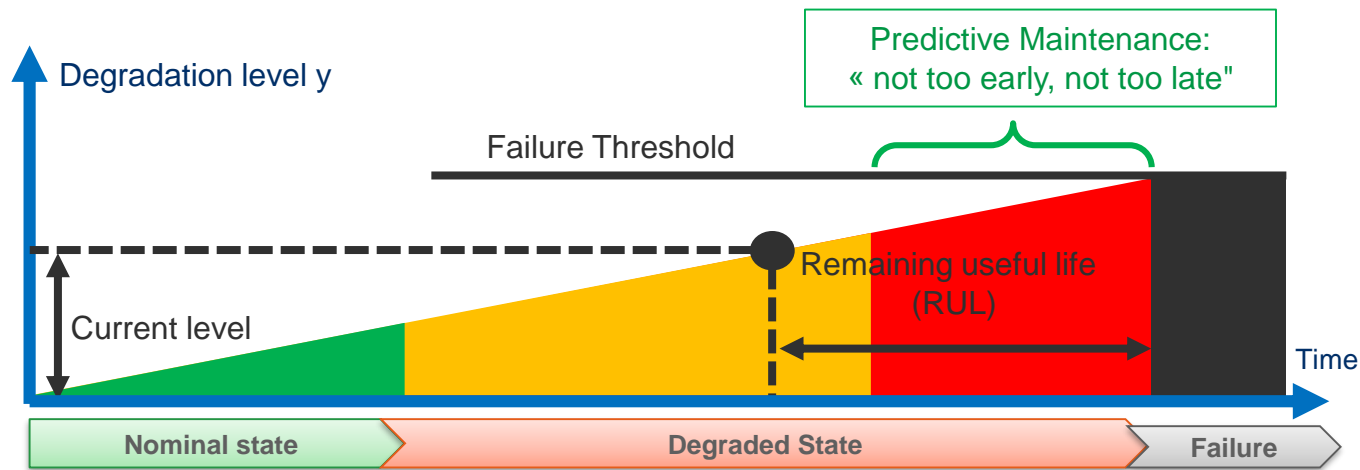
- Introduction: Asset monitoring and Fleet Challenges in Railways
- Problem Formulation: dynamic Resource allocation
- Problem Resolution 1 : a linear Programming Model
- Problem Resolution 2 : a multi-agent System
- Some preliminary Results
- Future Developments

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# Managing the Health of one individual Asset

- The search for the best maintenance policy for ONE asset can be seen as:  
How to maintain just enough to avoid failures, rather than over-maintaining



Minimize Cost subject to Availability constraint

# New Challenges when moving from individual Asset to Fleet

- An individual asset can be studied in a lab environment (test bench)
- A fleet of assets (all passenger doors on 600 trains, 1000 point machines in a country...)
  - Is comprised of numerous assets, each with its own profile, environment and context
  - Subject to operational constraints

Can be studied only in the actual field environment

# The Vision: by 20XX (xx=?), in our industry

## Real-time

- Every railway asset in the world will be remotely monitored in real-time

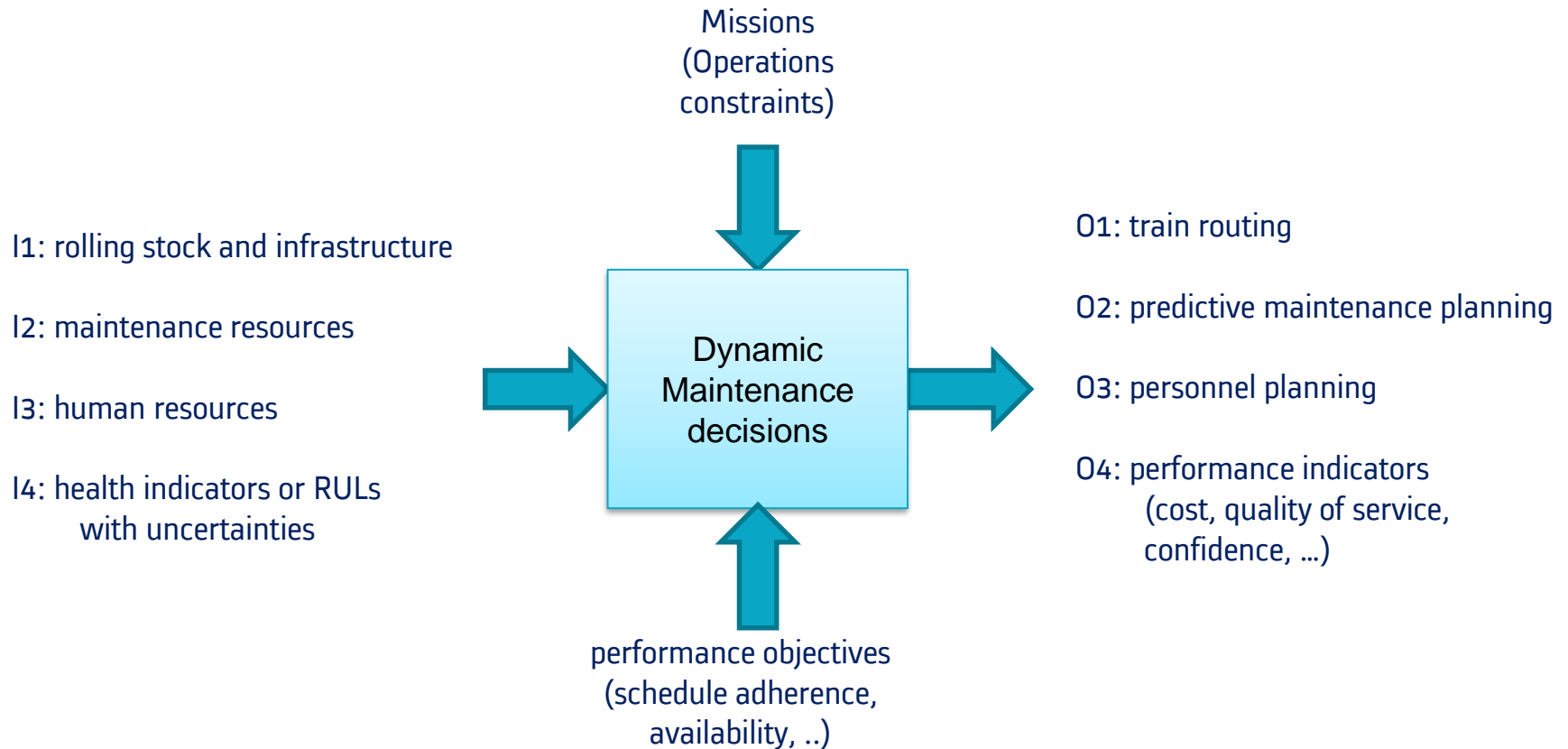
## Dynamic

- Maintenance schedules will be created dynamically based on the predicted condition of each single component

## Automatic

- Workload balancing in depots will be done automatically

# What's needed to fulfil that vision



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

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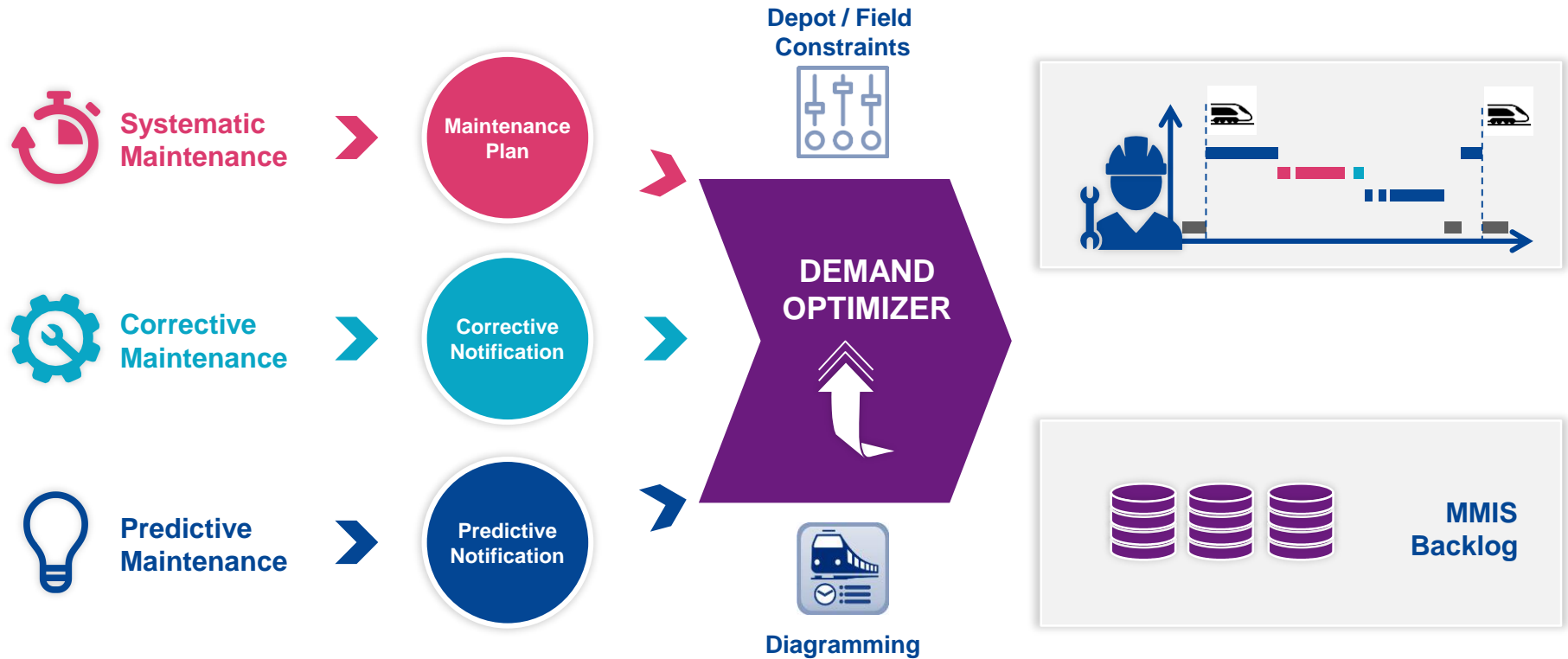
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- **Problem Formulation: dynamic Resource allocation**
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# Dynamic Resource Allocation

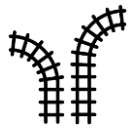


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# Resolution based on Linear Programming – Assumptions

## Network and infrastructure



track  
sections,  
turnouts, ...

## Depot



m  
tramways

## Maintenance center



- human resources  
- tools  
- spare parts

## P missions per day

(1 mission = 1 set of rides  
assigned to 1 tramway)

- each mission is associated to a  
degradation rate  $\delta$

m tramways of same type (interchangeable)

- granularity:
  - system – tramway
  - sub-system – air conditioning system
  - component – filter
- 1 RUL value for each tramway (RUL of the limiting component)
- no degradation when not in use
- the degradation of a sub-system has no impact on the degradation of other sub-systems

## Operation



P tramways  
per day

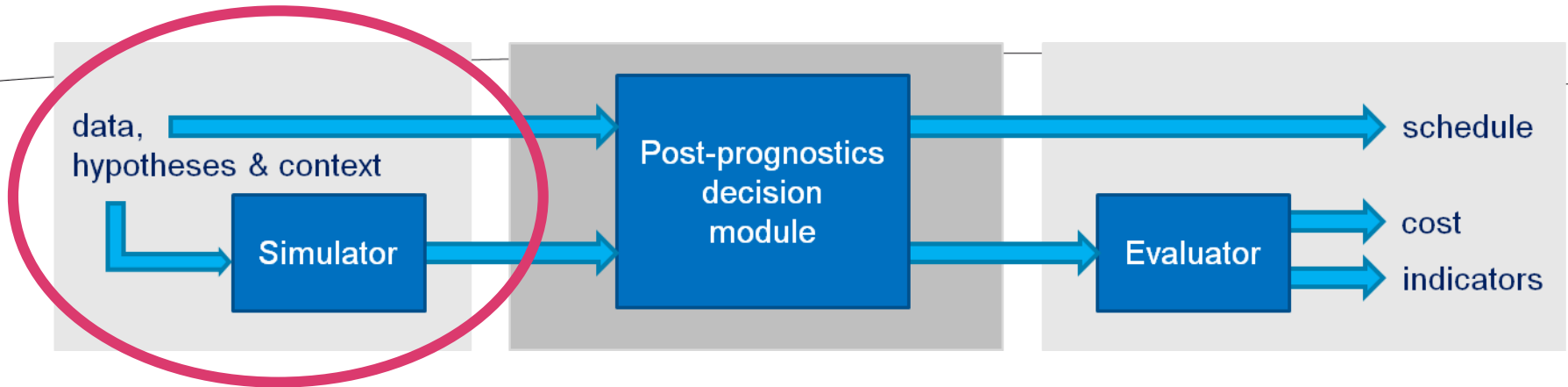
resolution based on  
Linear Programming

## Maintenance planning

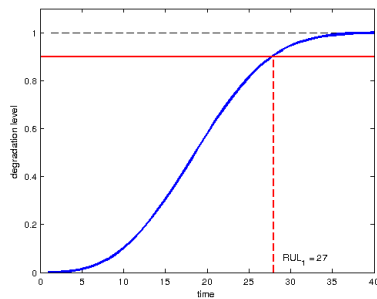


- maintenance dates
- maintenance operations (type, grouping, ...)
- choice of tramways to send to revenue service each day

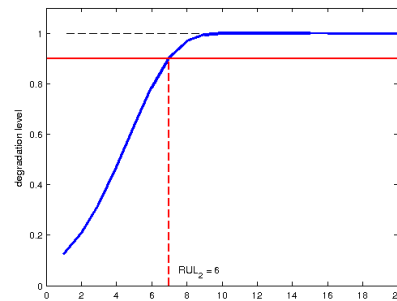
# Resolution based on Linear Programming – use case



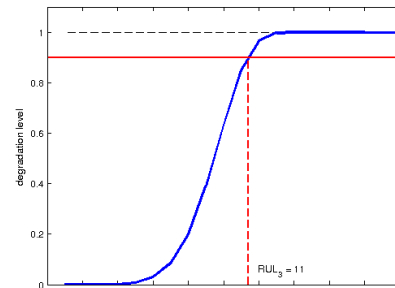
- **Assumptions:** 4 tramways, 1 sub-system per tramway for predictive maintenance
- **Simulations:** degradation for each RS unit follows a Weibull distribution  
 $[k_1 = 3, \lambda_1 = 21, \theta_1 = -1]$      $[k_2 = 8, \lambda_2 = 18, \theta_2 = -14]$      $[k_3 = 6, \lambda_3 = 9, \theta_3 = 0]$      $[k_4 = 6, \lambda_4 = 10, \theta_4 = -13]$



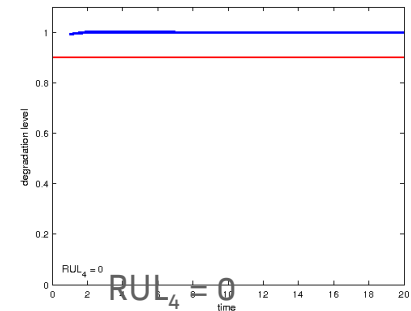
$RUL_1 = 27$



$RUL_2 = 6$

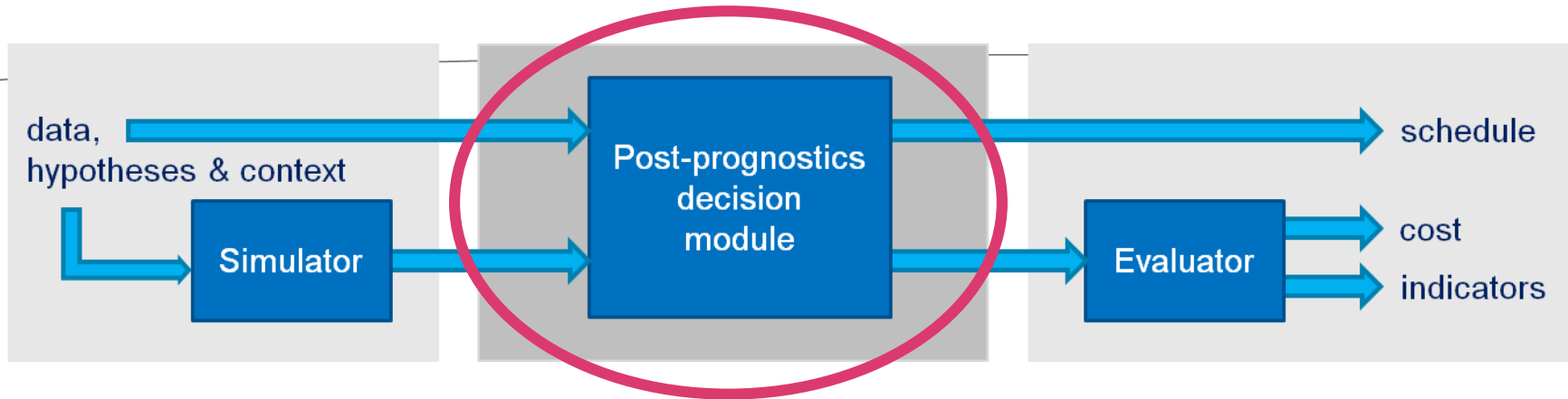


$RUL_3 = 11$



$RUL_4 = 0$

# Resolution based on Linear Programming

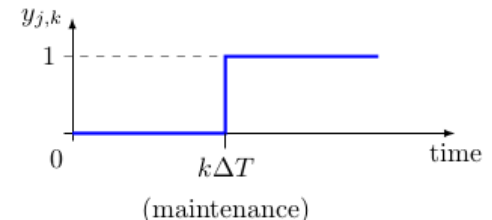
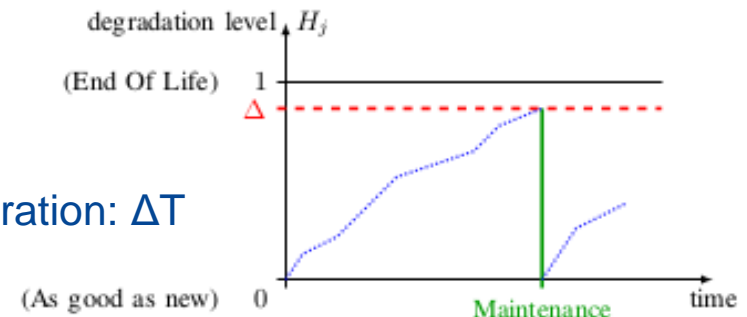


## ○ Linear Programming

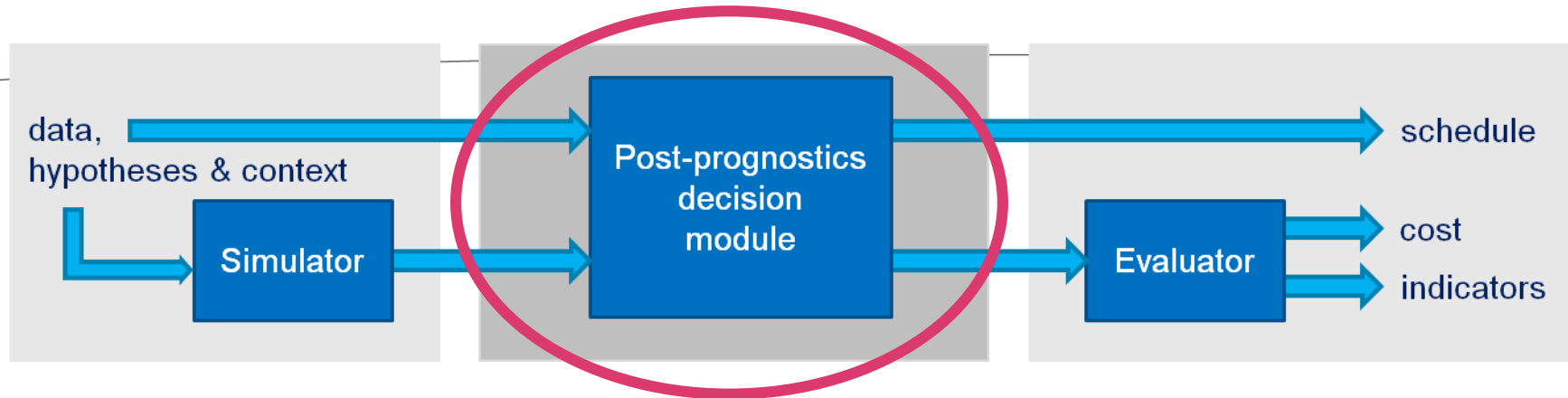
- Assumptions:
  - $m$  rolling stock units  $M_j$  ( $j=1, \dots, m$ ) –  $RUL_j$
  - $P$  missions  $T_p$  per day ( $p=1, \dots, P$ ) –  $\delta_p = \delta$
  - 1 maintenance op. for each rolling stock unit – duration:  $\Delta T$
  - maintenance capacity:  $c$  operations per  $\Delta T$
  - horizon  $H = K\Delta T$  ( $k=1, \dots, K$ )
- Decision variables:

$x_{j,k} = 1$  if the RS unit  $M_j$  performs a mission during Day  $k$

$y_{j,k} = 1$  if the RS unit  $M_j$  is maintained on Day  $k$



# Resolution based on Linear Programming : constraints



## ○ Linear Programming

- assignment of rolling stock to the P missions
- Compliance with RUL values ( modeled as deterministic)
- no assignment during maintenance
- maintenance capacity constraint
- only one maintenance op. per rolling stock unit

$$\sum_{i=1}^m x_{j,k} = P \quad \forall 1 \leq k \leq K$$

$$\sum_{k=1}^K \delta \cdot x_{j,k}(1 - y_{j,k}) \leq RUL_j \quad \forall 1 \leq j \leq m$$

$$y_{j,k} - y_{j,k-1} \leq 1 - x_{j,k} \quad \forall 1 \leq j \leq m, \forall 1 \leq k \leq K$$

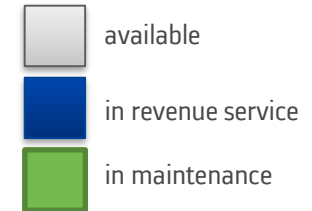
$$\sum_{i=1}^m (y_{j,k} - y_{j,k-1}) \leq c \quad \forall 1 \leq k \leq K$$

$$\sum_{k=1}^K (y_{j,k} - y_{j,k-1}) \leq 1 \quad \forall 1 \leq k \leq K$$

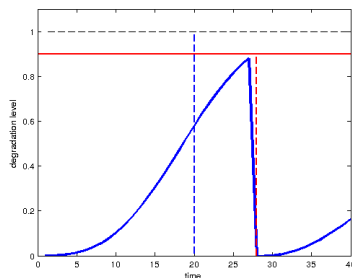
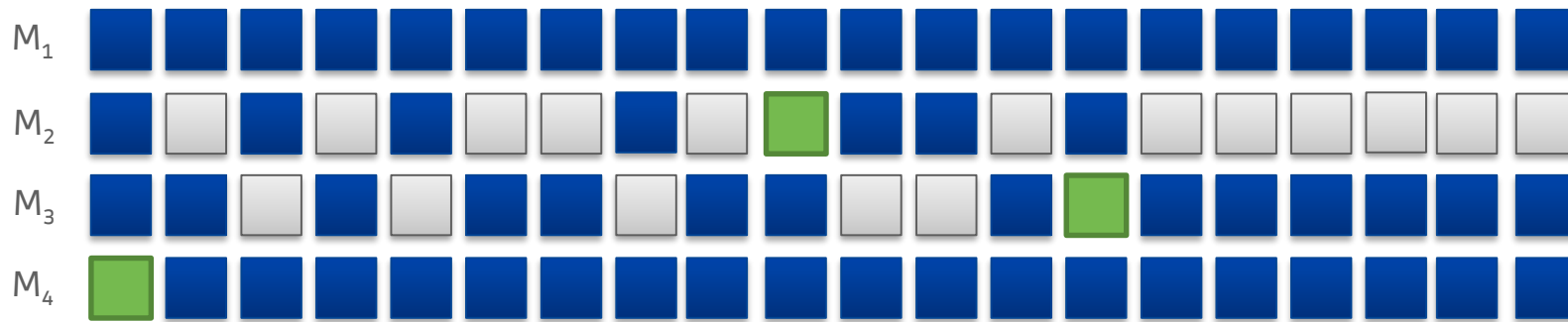
# Resolution based on Linear Programming – use case

## ○ Assumptions:

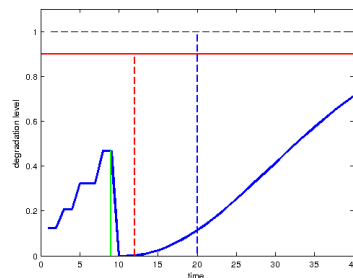
- $m = 4$  tramways,  $n = 1$  sub-system per tramway for predictive maintenance
- $P = 3$  missions per day  $k$
- $H = 20$  days ( $K = 20$ )
- $c = 1$  maintenance operation per day



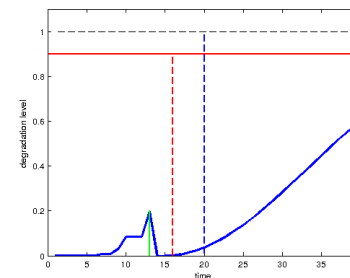
## ○ Output Schedule



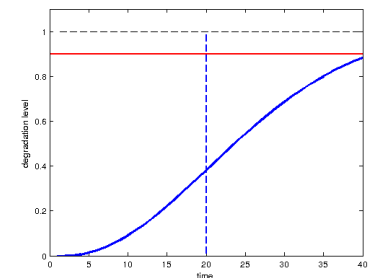
$M_1$ :  $RUL_1(0) = 27$



$M_2$ :  $RUL_2(0) = 6$   
 $RUL_2(M) = 2$



$M_3$ :  $RUL_3(0) = 11$   
 $RUL_3(M) = 3$



$M_4$ :  $RUL_4(0) = 0$   
 $RUL_4(M) = 0$



# Linear Programming (LP): various objective functions


- Objective 1: maximize degradation level before maintenance
  - Objective 2 : minimize the number of maintenance operations over the scheduling horizon.
  - Objective 3 : maximize service provided, i.e. number of missions.
- Each leads to different decisions.

LP: a sound approach but quickly breaks down when number of systems or time horizon increases.... The “curse of dimensionality”.

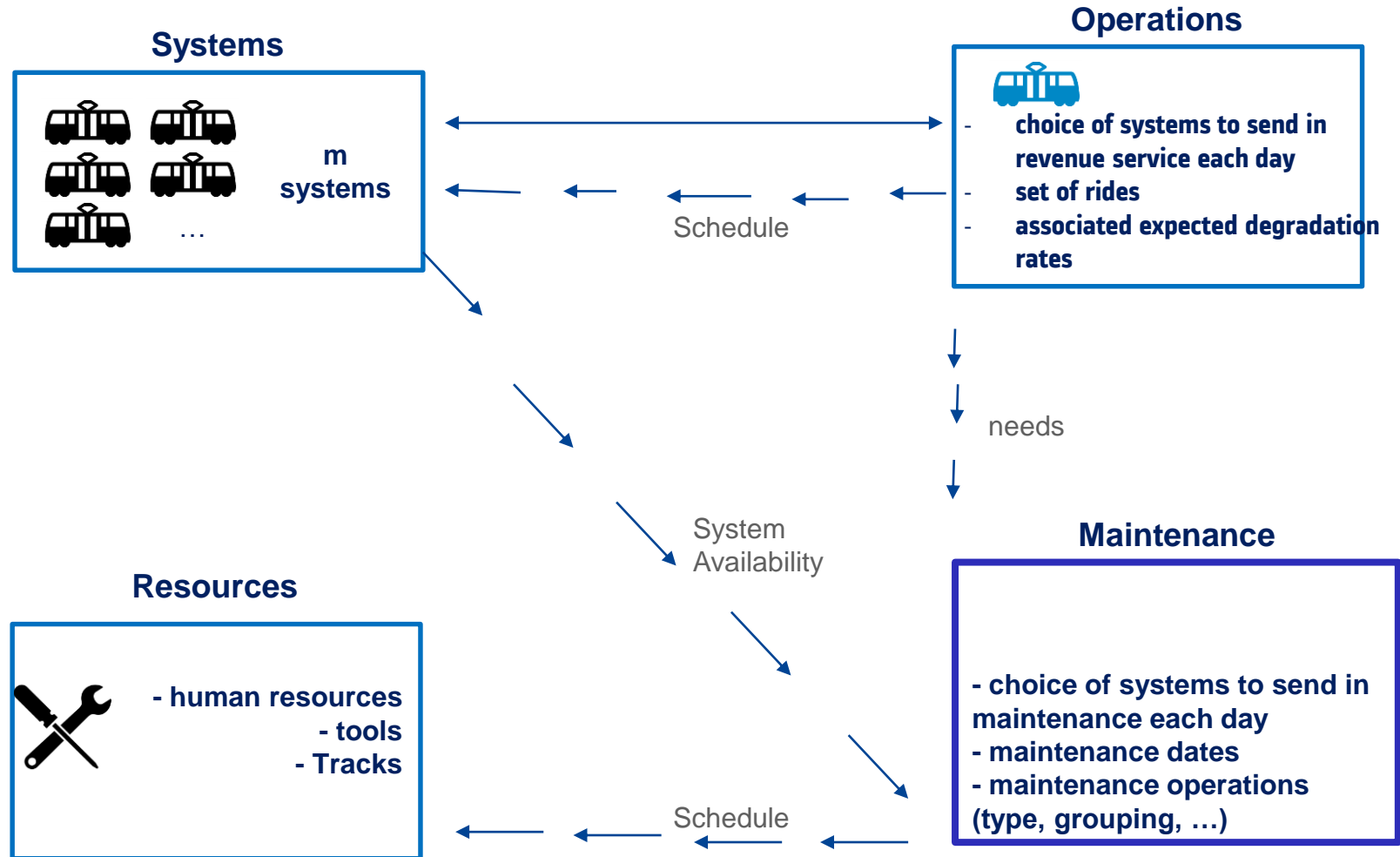
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# Multi-Agent System Approach

- Multi-Agent Systems ( MAS): approximate solutions to very complex problems by distributing them to agents, which can be seen as autonomous problem solvers (but need to coordinate)
- Well suited to the functionally and geographically distributed characteristics of a rail transport system.
- Who are the agents ?
  - A Maintenance Agent: elaborates maintenance plans
  - An Operations Agent: assigns assets to missions
  - Resource Agents: provide resources ( rolling stock, maintenance facilities)
  - System Agents: manage the state of health of each asset
- Agents' "conflicting" goals :
  - Operations: fulfil missions  Maintenance : keep assets healthy
  - They need to 'negotiate'.

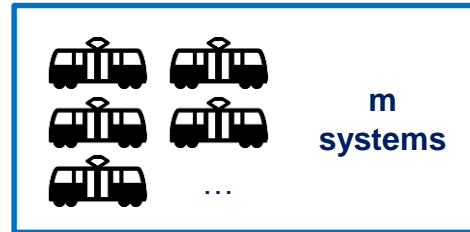
# Interactions between Agents



# Agents' Data

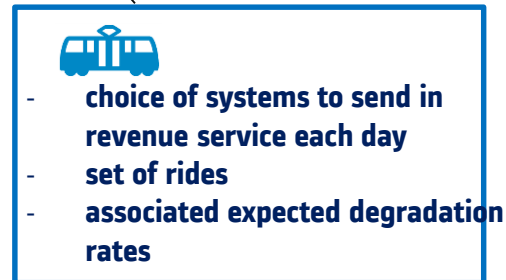
- Number of missions per day
- Length of missions
- Degradation rate of missions
- Cost of lost mission
- Planning horizon

## Systems



- Number of systems
- Sub-systems that require Predictive maintenance
- RULs of sub-systems after maintenance
- Sub-systems that require Systematic maintenance
- Periodicity of each systematic maintenance
- Cost of lost Km for each sub-system

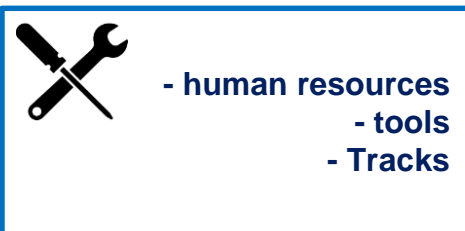
## Operations



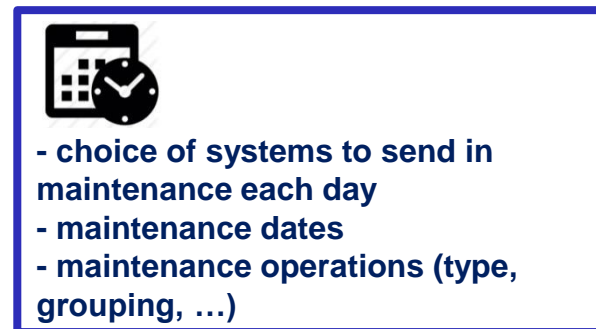
- Needed time for maint tasks
- Needed resources (Number + Time) for maint tasks
- Cost of each maint task

## Maintenance

## Maintenance



- Number of resources
- Availability of resources



# Some Preliminary Results

## ■ Linear Programming Model

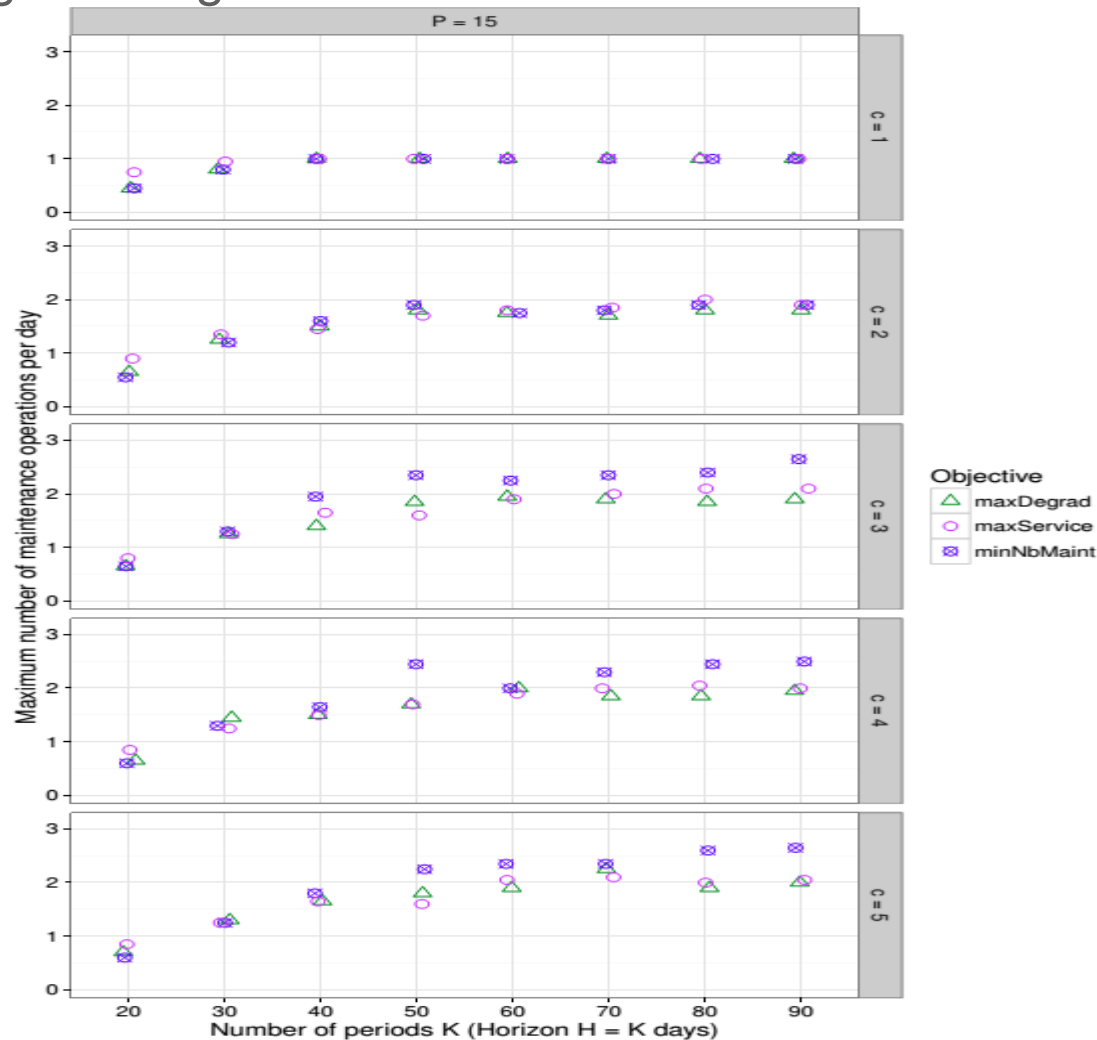


Figure 7: Maximum number of maintenance operations per day for the three objective functions and maintenance capacities  $c = 1, 2, 3, 4$  and  $5$  –  $m = 18$  systems,  $P = 15$  missions

# Evaluation Metrics ( KPIs)

KPI1: Number of lost ( non-performed) missions.

→ key for Operations

KPI2: Number of maintenance operations per day.

→ Maintenance capacity requirements

KPI3: Number of maintenance operations/ component over planning horizon → LCC.

KPI4: 'Lost km per maintenance action' : kms that still could have been performed given the state of health.

KPI5: Resource Load Ratio: % of time during which a resource is effectively utilized.

 Various policies impact those metrics differently

# Heuristic Policies

S1: assign most degraded train to easiest mission → keep best systems available

S2: assign easiest mission to most degraded system that can perform it → postpone next maintenance of each system

S3: -- in a first step, S1

-- in a second step, Operations and Maintenance agents communicate to minimize number of lost missions

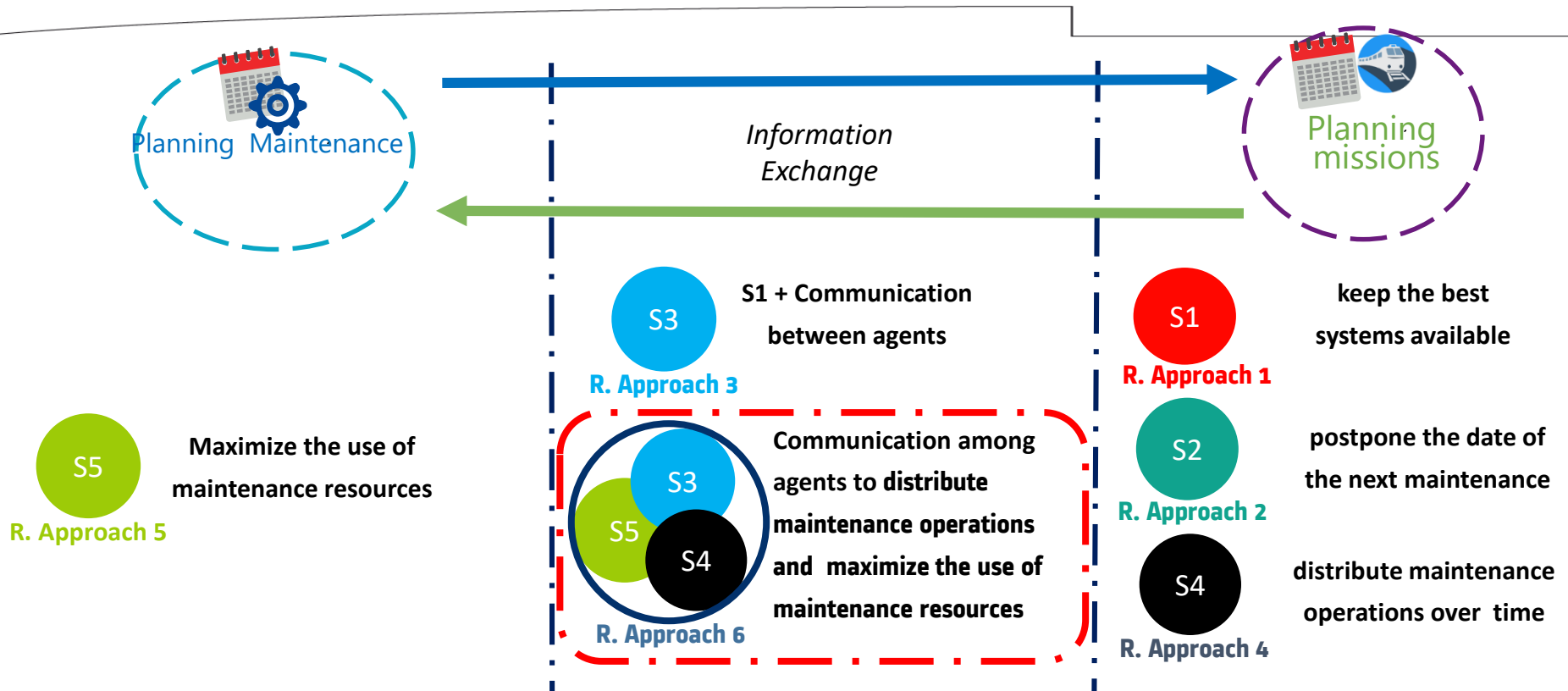
S4: Distribution maintenance operations over time

S5: choose systems to be maintained so as to maximize maintenance resource use

S6: use S4 for mission assignment and S5 for maintenance assignment



# Resolution approach based on Multi-Agent Model



Heuristic Decision Making

6 Strategies ( S1, S2,.... S6)

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# Some Future Developments

- More-general types of networks ( Main Lines– several depots)
- Model RULs as random variables
- Model more explicitly the “negotiation process” between Operations and Maintenance agents
- Encapsulate some linear-programming algorithms inside the agents
- Formalize operating costs to evaluate cost of lost missions
- Rigorous derivation of system (train)-level health attributes from subsystem ones

# Acknowledgement

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# Bibliography

- M.J. Woolridge , N.R. Jennings, “ Intelligent Agents: Theory & Practice”, The Knowledge Engineering Review, 111-152, 1995.
- B Chen, H. Cheng, “ A review of the Application of Agent Technology in Traffic and Transportation Systems”, IEEE Transactions on intelligent Transportation Systems, 11(2): 485-497, June 2010.
- “ Predictive maintenance of moving systems”, N. Herr, J-M.Nicod, C. Varnier, N. Zerhouni, P.Dersin, IEEE Prognostics & System Health Management Conference, Harbin, China, July 2017.



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