# **Berliner Digitaler Bahnbetrieb (BerDiBa)**

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**Cooperation:** Siemens Mobility GmbH TU Berlin (MDT) IBB, EU & State of Berlin Funding:



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# Project

Motivation:

- Ensuring reliable, cost-efficient, and secure public transport
- Reduction of the emission of greenhouse gases and pollutants
- Increase quality of life in cities and metropolitan areas
- The Berlin Mobility Act (Berliner Mobilitätsgesetz) from 2018

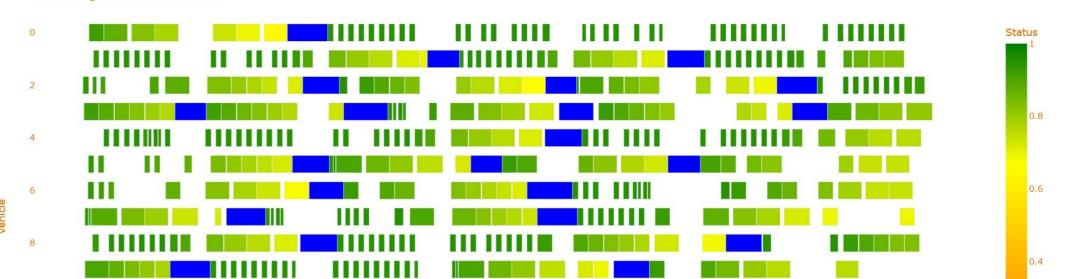
Project goal:

- Establishment of digital and holistic twins
- Development and efficient implementation of railway transportation
- Autonomous driving and ultimately fully automated operation

#### Our task:

# **RSRP with Predictive Maintenance**

- Given: Railway network, timetable, vehicles with individual health states, estimations of degradation caused by trips
- Goals: Rolling stock rotations operating all trips, schedule maintenance based on the predicted health states



- Assign vehicles to train services, automatically schedule the necessary maintenance operations and optimize the utilization of trains
- Incorporate predictive maintenance into rolling stock rotation planning (RSRP)

Example for vehicle schedules showing their health state in green to red and maintenance activities in blue.

# Graph Model

- Arcs represent the trips, deadheads, waiting, and maintenance
- Nodes correspond to the arrival or departure at a location
- Costs of trip arcs depend on their failure probabilities
- Finite set of values  $\mathscr{D}$  containing the parameters of the nodes, i.e., each node exists for each value of  $\mathscr{D}$ 
  - $\rightarrow$  The graph has different layers depending on  $\mathscr{D}$
  - $\rightarrow$  Arcs between layers are altering the parameters

#### Assumptions:

- PDFs of health states belong to a parametric family
- Degradation functions of trips are monotonic increasing and Lipschitz
- Failure probability is monotonic and Lipschitz w.r.t. the parameters

#### Solution:

- Graph approximates the original problem
- Granularity of  $\mathscr{D}$  determines the accuracy of the approximation
- Solution for the approximate problem is given by a cost-minimal set of paths covering exactly one arc corresponding to each trip

# **Solution Approach**

A solution to the approximate problem can be determined by solving the following integer program (IP) formulation:

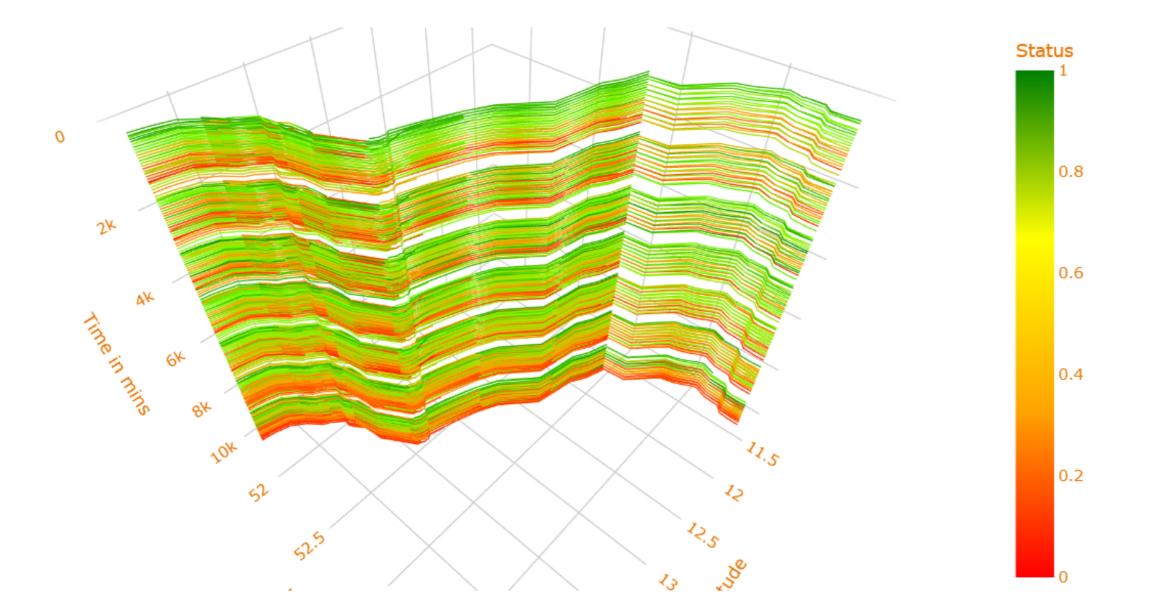
$$\begin{array}{ll} \min \sum_{a \in A} c_a x_a \\ \text{s.t.} \sum_{a \in A(t)} x_a = 1 & \forall t \in T \\ \sum_{a \in \delta^+(v)} x_a = \sum_{a \in \delta^-(v)} x_v & \forall v \in V \\ & x_a \in \mathbb{Z}_{\geq 0} & \forall a \in A \end{array}$$

- Refining  $\mathcal{D}$  gives rise to closer approximations of the original problem
- IP-based heuristic: Refine  $\mathcal{D}$  randomly
- Iterative refinement approach:

Train Usage with Health Statu

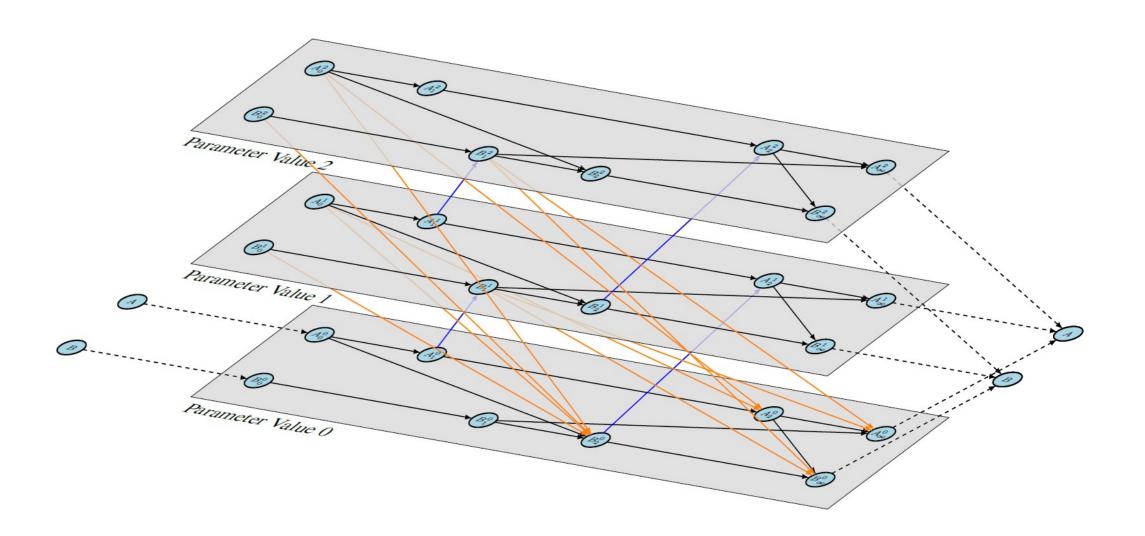
- 1. Choose  $\mathscr{D}$  randomly or evenly spaced
- 2. Determine the corresponding graph and solve the IP

Health Status



Example for a time-expanded rotation plan showing the health states of the vehicles.

- 3. Refine node-wise around parameter values of nodes contained in the resulting paths
- 4. Go to 2 and repeat



Example for a parameter-expanded event-graph. Here, deadheads and waiting arcs are black, maintenance arcs are orange, and trip arcs are blue.

### Funding



