

Operational scheduling in automotive industry

Scheduling Seminar

5-Nov-24

www.artelys.com



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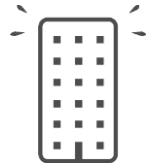
1. Artelys in a nutshell
2. Industrial context
3. Optimization problem
4. Resolution & Results



Artelys in a nutshell

Artelys in a nutshell

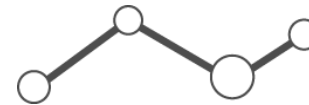
Artelys is an independent company, created in 2000, specialized in **optimization, decision-support, modeling.**



2000
CREATION
Arnaud Renaud



15% annual
sustained growth



90 EXPERTS
MSc and PhD



35% of our activity
is dedicated to **R&D**



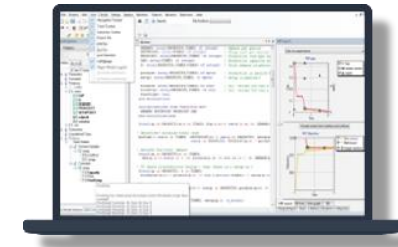
SOFTWARE EDITION

Custom software,
off-the-shelf software,
Numerical solvers



SERVICES & CONSULTING

Optimisation,
Data Science
and business expertise



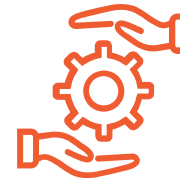
Artelys in a nutshell



Strong skills in Operational Research, Modeling and Decision Support



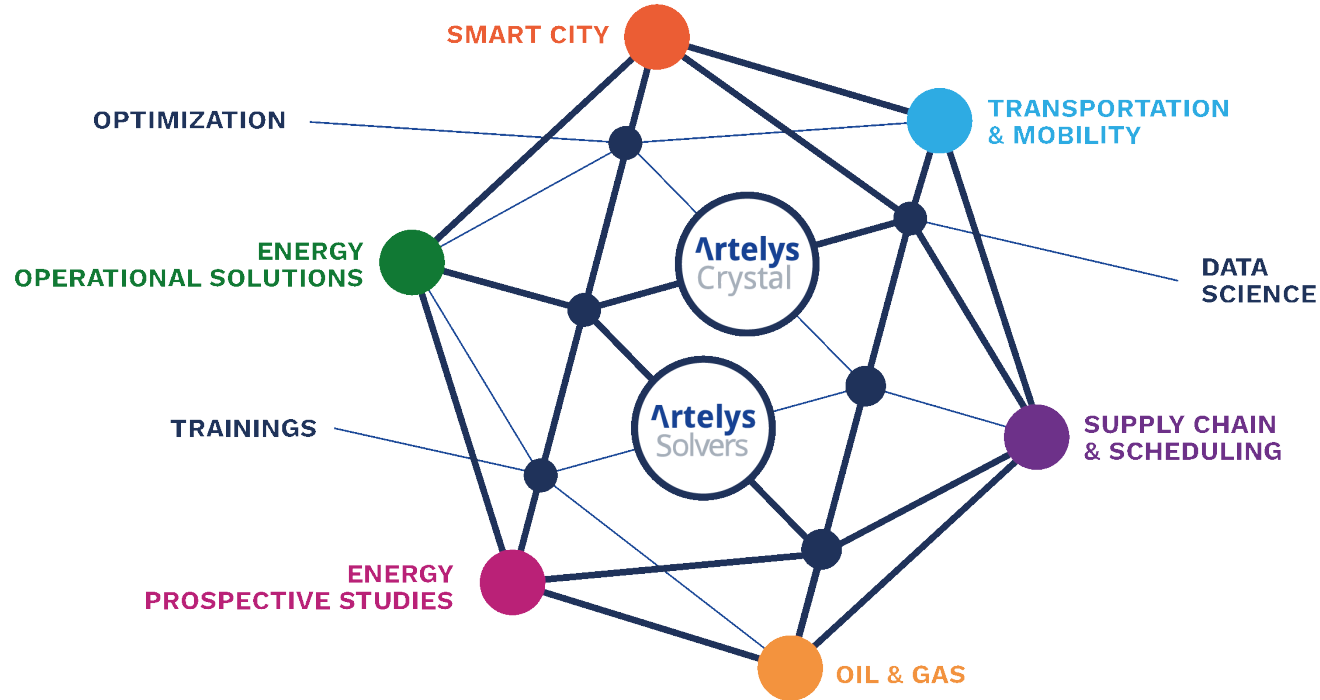
Recognized know-how in scientific consulting and project management



A business experience of energy, transport and logistics issues



Mastery of methods and tools for numerical optimization and statistical analysis



Numerical optimization tools

4 Artelys Knitro

- | Industry leading solver for very large, difficult quadratic and nonlinear optimization problems (QP, NLP, MINLP)

↳ <https://github.com/Artelys/knitro-modeling-examples>



4 FICO Xpress Optimization Suite

- | High performance linear, quadratic and mixed integer programming solver (LP, MIP, QP)



4 Artelys Kalis

- | Object-oriented environment to model and solve problems with **constraint programming**



4 AMPL

- | Comprehensive **modeling language** for Mathematical Programming



Example scheduling projects at Artelys

┌ Maintenance planning for public transit operators (subway, trams)

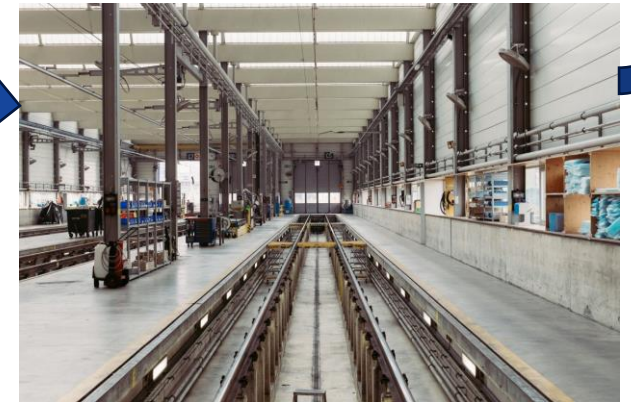
| When to apply maintenance tasks on each train ?

- ↳ Satisfy maintenance shop capacities
- ↳ Satisfy min/max travelled distance between each maintenance
- ↳ Satisfy precedence/series constraints for maintenance tasks
- ↳ Satisfy unavailability of trains and maintenance shops

| Different horizons:

- ↳ Strategic planning (next years)
- ↳ Tactical planning (current year)
- ↳ Operational planning (next 2 weeks)

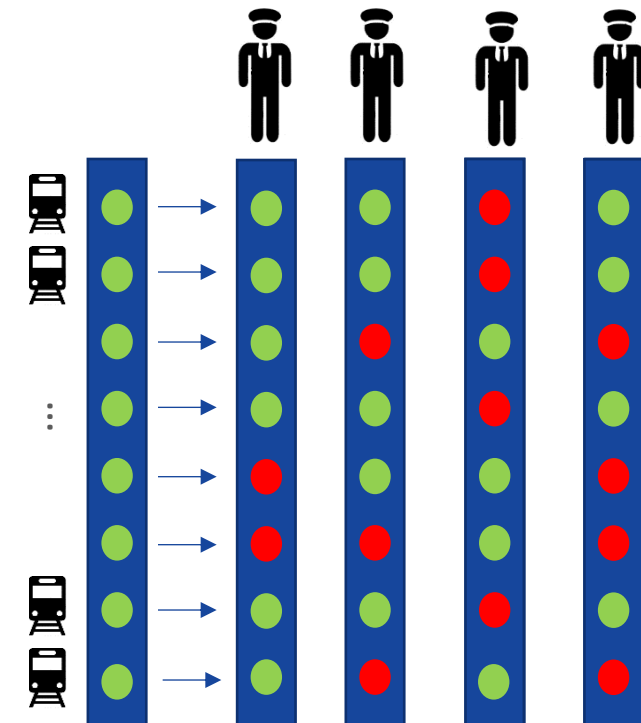
| CP + MILP

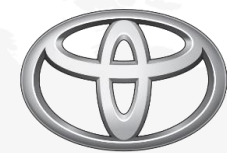


Example scheduling projects at Artelys

4 Shift planning for on-board railway crews

- | Which operator on which train ?
- | Pairing: Generate back-and-forth trips for crews
- | Planning: generate feasible shifts for each operator (sequence of trips)
- | Satisfy work rules and skill requirements
 - ↳ min/max trips
 - ↳ balance for different types of trips
 - ↳ number of overnight stays
- | Maximize train coverage
- | Operational planning for each month
- | CP + MILP





TOYOTA

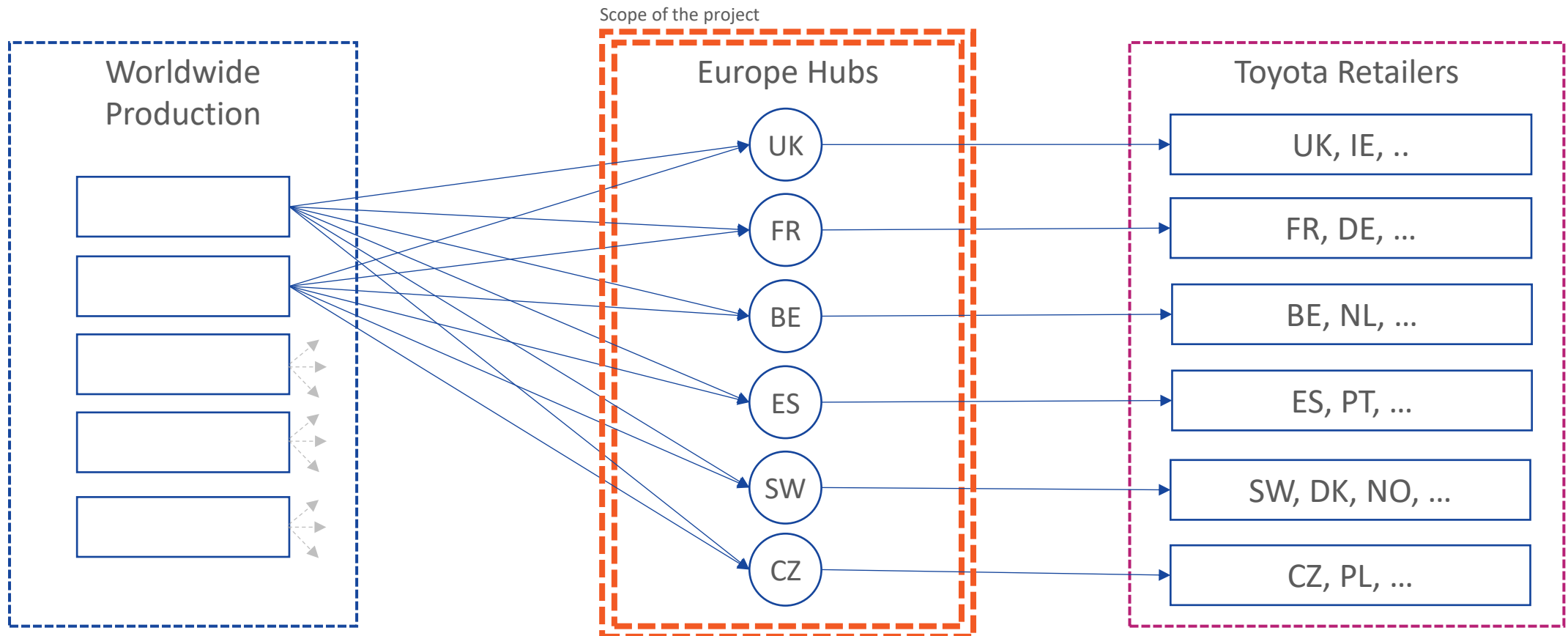
Operational scheduling in automotive industry

Context

Context

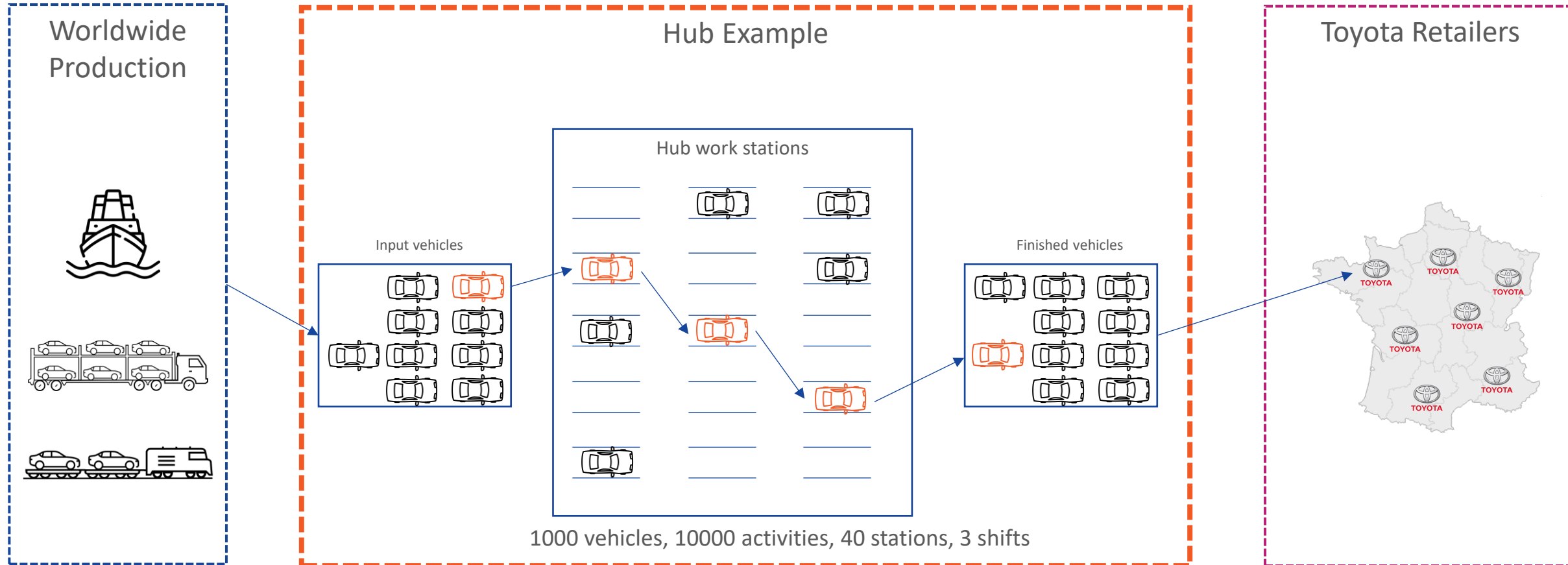
Toyota Motors Europe (TME) operations:

- Hubs receive produced vehicles and finalize them with Post-Production Options (PPO) before final retailers delivery



Context

One hub example



Context

▣ One hub example





Optimization problem

Hard constraints

▴ Disjunctive resources

- | A station can execute at most one task at a time
- | At most one task executed at a time on a vehicle

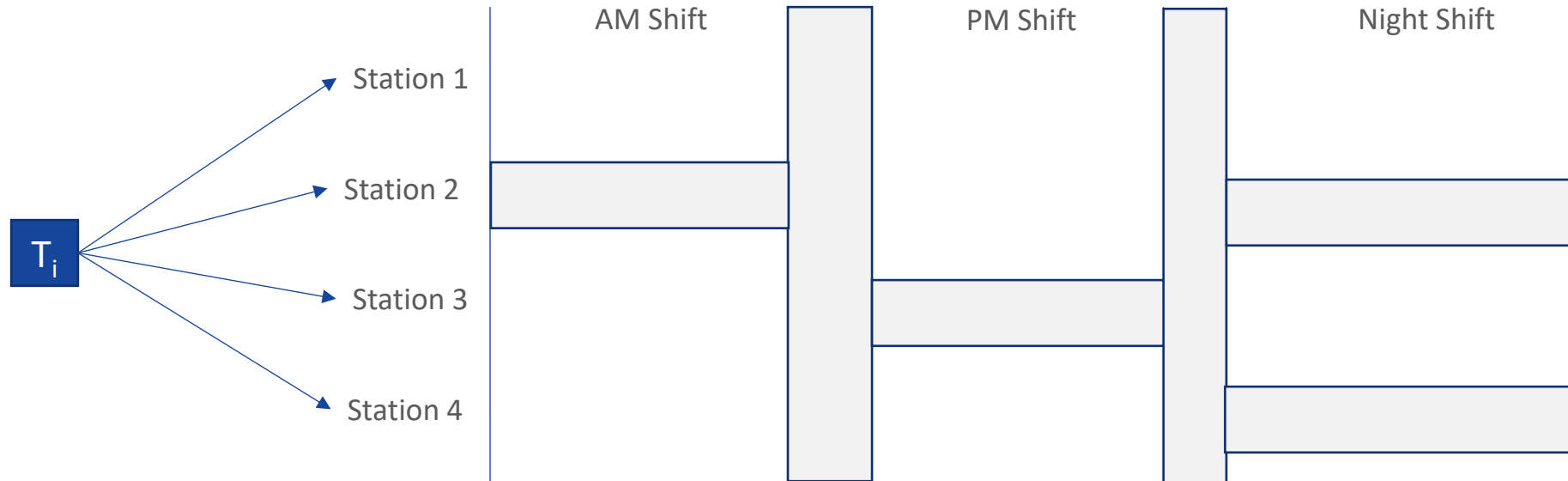
Hard constraints

▣ Disjunctive resources

- | A station can execute at most one task at a time
- | At most one task executed at a time on a vehicle

▣ Tasks can be assigned to several stations

▣ Tasks can be performed only during specific shifts



Hard constraints

▴ Disjunctive resources

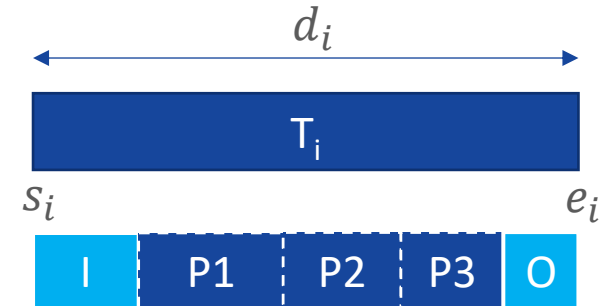
- | A station can execute at most one task at a time
- | At most one task executed at a time on a vehicle

▴ Tasks can be assigned to several stations

▴ Tasks can be performed only during specific shifts

▴ Task duration is variable:

- | Nominal duration: PPOs of a vehicle are **aggregated** into tasks based on business rules
- | Operator's skills / Number of operators
- | **In/out times** for parking trips varying for stations/shifts
- | Stations have **hard-breaks** and **soft-breaks** and work during a **shift**

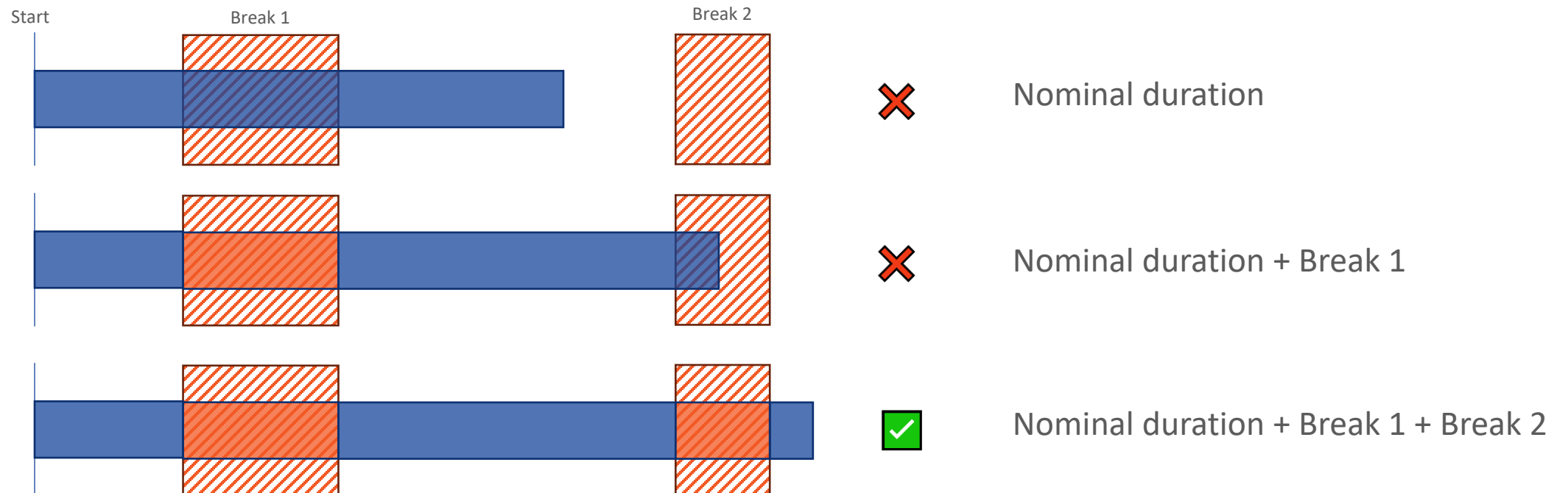


Hard constraints

Task duration is variable:

Soft-breaks

- ↳ Breaks are specific to each station/shift (cannot pre-process them)
- ↳ Cannot be dealt with hard breaks
- ↳ Duration of task depends fully on the start time of the task (must not branch on duration variables)

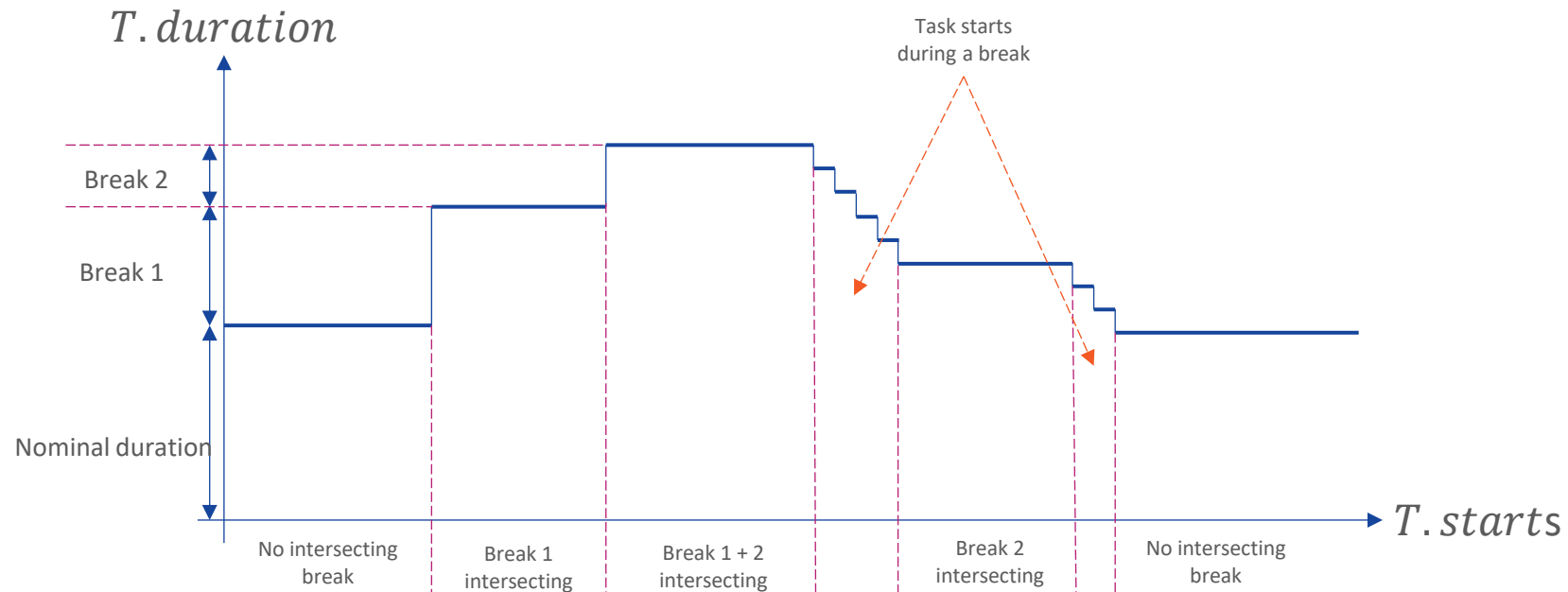


Hard constraints

Task duration is variable:

Soft-breaks

- ↳ Breaks are specific to each station/shift (cannot pre-process them)
- ↳ Cannot be dealt with hard breaks
- ↳ Duration of task depends fully on the start time of the task (must not branch on duration variables)
 - “Start based duration” global constraint



Hard constraints

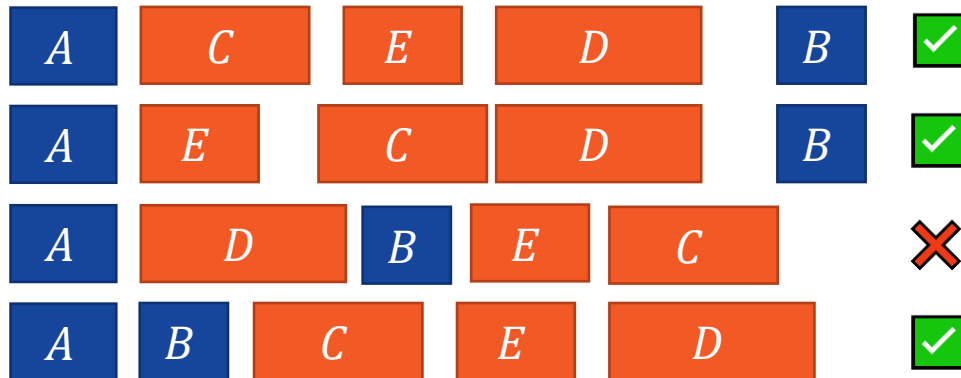
▴ Precedence and sequence constraints:

| Precedence constraints:

↳ Task A must be done **before** Task B

| Sequence constraints:

↳ Task C, D, E must be done without any other task **in-between them**

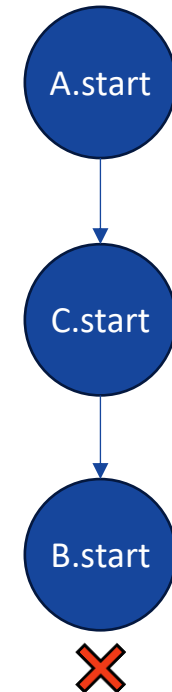


| High-level rule-based definition:

↳ $*-A-*-[C-E-D]-*-B-*$

↳ User does not generate the precedence matrix by himself

Impact on tree search

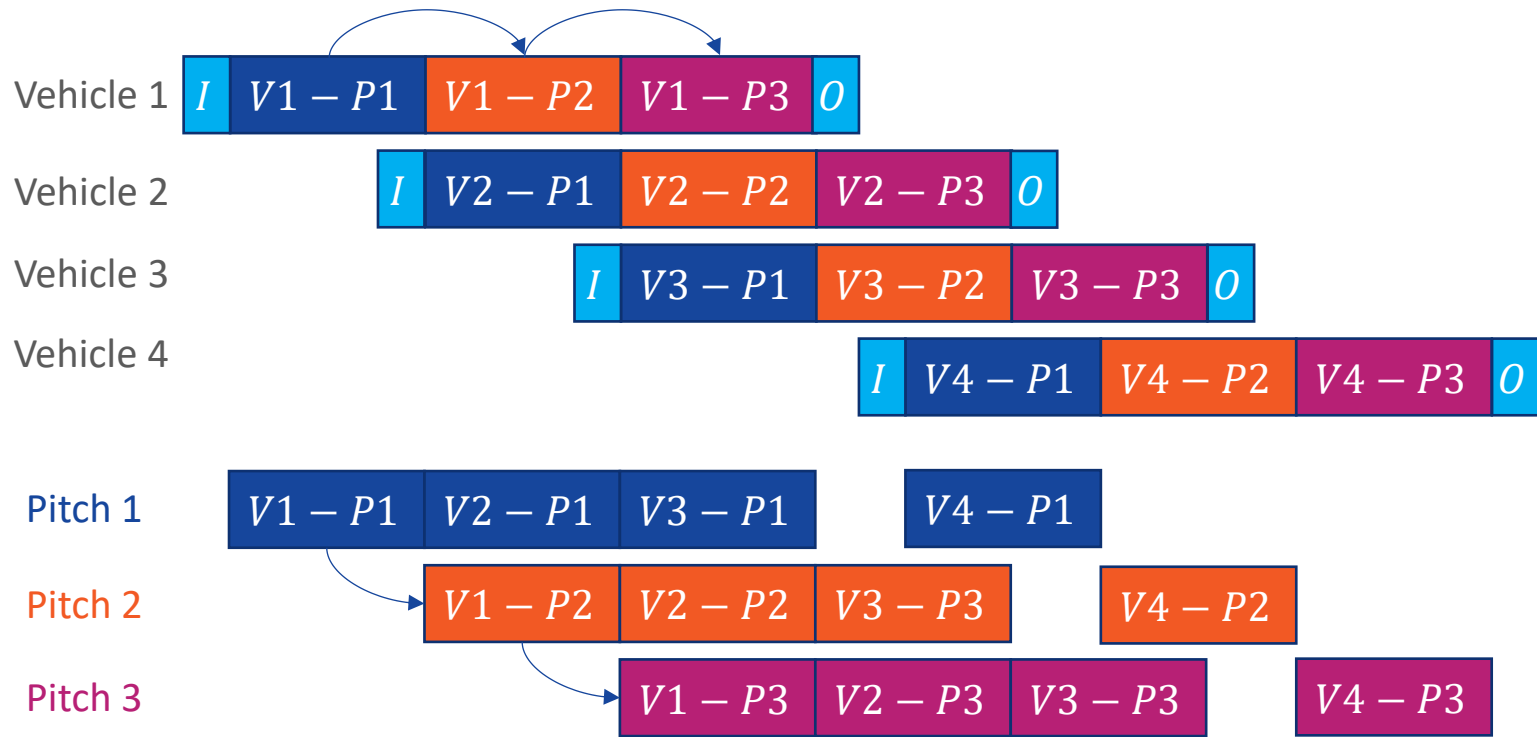


“No good” choice

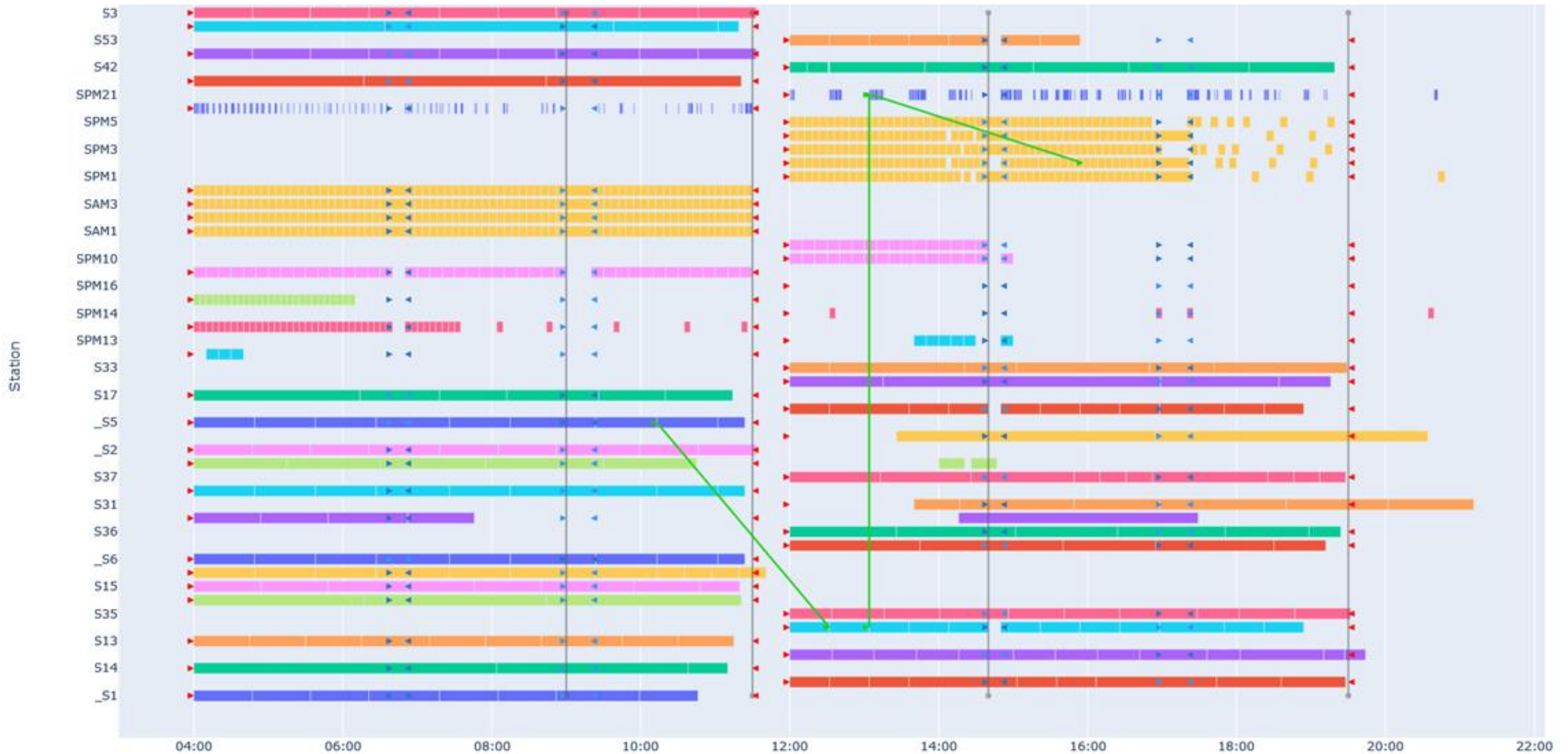
Hard constraints

4 Conveyors

- | The vehicle must pass through n pitches
- | 1 pitch takes a fixed number of seconds (e.g. 100 seconds)
- | This is not a resource of capacity n
- | Once in the conveyor, the vehicle must go through all pitches
- | I/O times only applied to vehicle resources
- | Breaks applied on all pitches



Example visualization



Soft-constraints

▴ Multi-objective

- | Maximize the number of tasks planned within shifts
- | Maximize the number of vehicles ending before their cutoff time
- | Minimize vehicles time span (minutes)
- | Minimize shifts durations (minutes)
- | Minimize successive complex tasks on stations
- | Minimize successive different family tasks

Deadline constraints

Workshop efficiency

Stations efficiency

Soft-constraints

Multi-objective

- | Maximize the number of tasks planned within shifts | 0.5%
- | Maximize the number of vehicles ending before their cutoff date | 10%
- | Minimize successive complex tasks on stations | 20%
- | Minimize successive different family tasks | 20%
- | Minimize shifts durations (minutes) | 20%
- | Minimize vehicles span times (minutes) | 20%

Goal-programming method:

- | Ranking + Equivalency tolerance
- | Two solutions comparison:
 - ↳ Solution 1 is better

Objective	Sense	Solution 1	Solution 2	Gap
NB_PPOS_PLANNED_IN_SHIFTS	MAX	980	979	-0.1%
NB_VEHICLE_BEFORE_CUTOFF	MAX	430	434	+9%
NB_SUCESSIVE_COMPLEX	MIN	10	10	0%
NB_MODEL_PPOS_CHANGES	MIN	80	150	+87%
TOTAL_SHIFTS_DURATION	MIN	960	1050	
TOTAL_VEHICLE_SPAN_TIME	MIN	50000	40000	

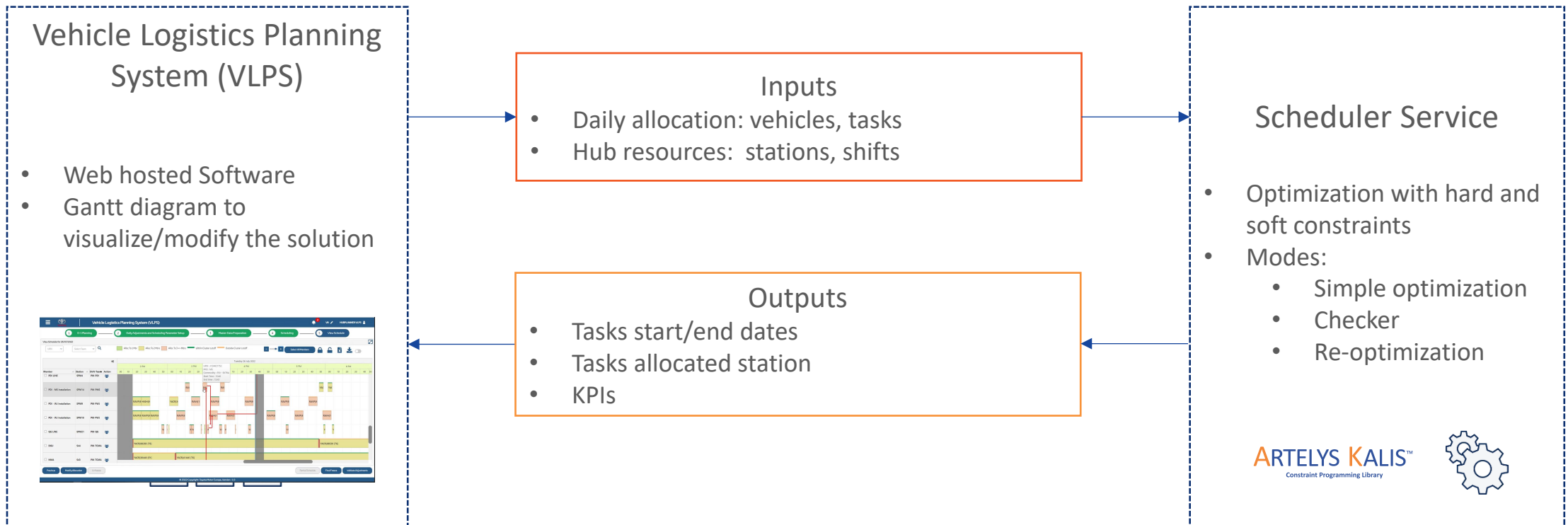


Resolution & Results

Resolution

4 VLPS Scheduler Service

- | **Optimize a daily scheduling plan** taking into account operational constraints
- | Fully integrated as a **stateless microservice** with TME's **Vehicle Logistics Planning System**



Resolution

4 Optimization technologies

- | Artelys-Kalis
- | Modelling with FICO Xpress-Mosel
 - ↳ Easy to update the model for a specification update
- | Short running times
 - ↳ 10000+ activities
 - ↳ 1s timestep
 - ↳ 5 minutes time limit

4 Remote computing

- | Multiple users can call the service simultaneously
- | Horizontal scaling

ARTELYS KALIS™
Constraint Programming Library



On the use of constraint programming

⚠ MILP would **fail**

- | Too many binary variables
- | Bad overall relaxation
- | Good to get bounds on sub-models

⚠ CP

- | Interval variables are essential
- | Customized search is still required to achieve operational performance
- | Expertise on the CP solver is still important
 - ↳ What global constraint to use ? There is no common API for CP Solvers
 - ↳ What propagation level is enough ?
 - ↳ Where is the time spent at each node ?

Results

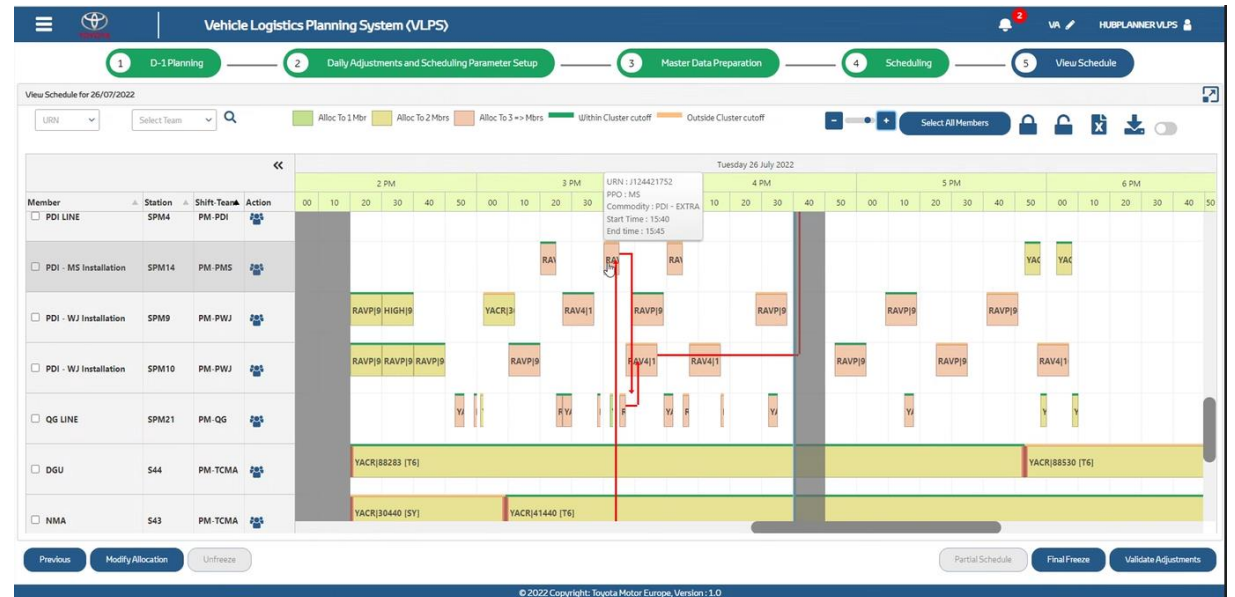
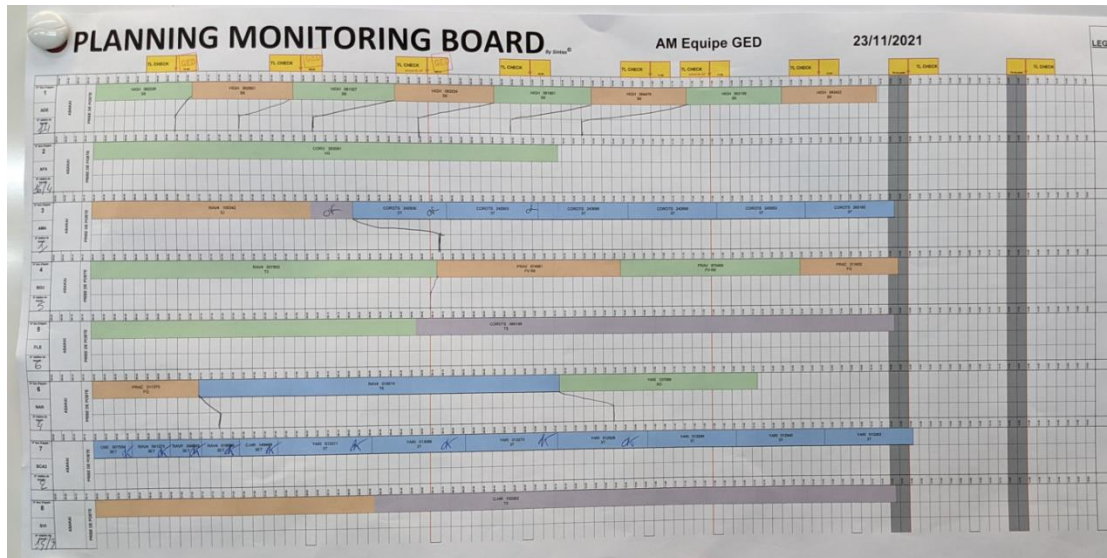
➤ Producing daily production plans on European hubs

➤ Before:

- | Excel sheets
- | Gantt charts printed on paper
- | 3h
- | Reduced scope

➤ After:

- | Web-based application
- | 100% digital, allowing management in real-time
- | 20 minutes
- | Improved planning quality





Q&A
