

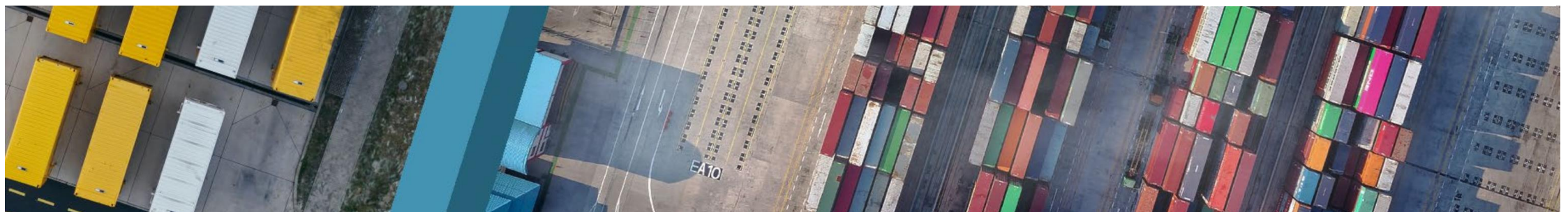


On-demand automated guided vehicles in yard logistics

[Ulrike Ritzinger](#), [Bin Hu](#), [Martin Reinthaler](#)

AIT Austrian Institute of Technology, Center for Energy

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AWARD

H2020 Project

- **All Weather Autonomous Real** logistics operations and **Demonstrations**
- Efficient and safe connected and automated heavy-duty vehicles in real logistics operations
- Potential to address key issues in commercial transportation:
 - Lack of qualified drivers in Europe
 - Number of accidents with trucks unacceptably high (due to driver failures)
 - Utilization of freight transport capacity is below 50%



AWARD has received funding from the European Union's Horizon 2020 research and innovation program under Grant Agreement No 101006817



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Scaling autonomous logistics

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H2020 Project

- Driving autonomously **on all roads at all speeds** with all sorts of **known and unknown** hazards certainly is and will be a challenge
- Developing and operating **safe autonomous transportation systems**
 - In a wide range of real-life logistic use cases in a variety of different scenarios.
 - Solution will be based on multiple sensor modalities and an embedded teleoperation system **to address 24/7 availability**
 - Vehicles will be deployed, integrated and operated in a **variety of real-life use cases** to validate their value



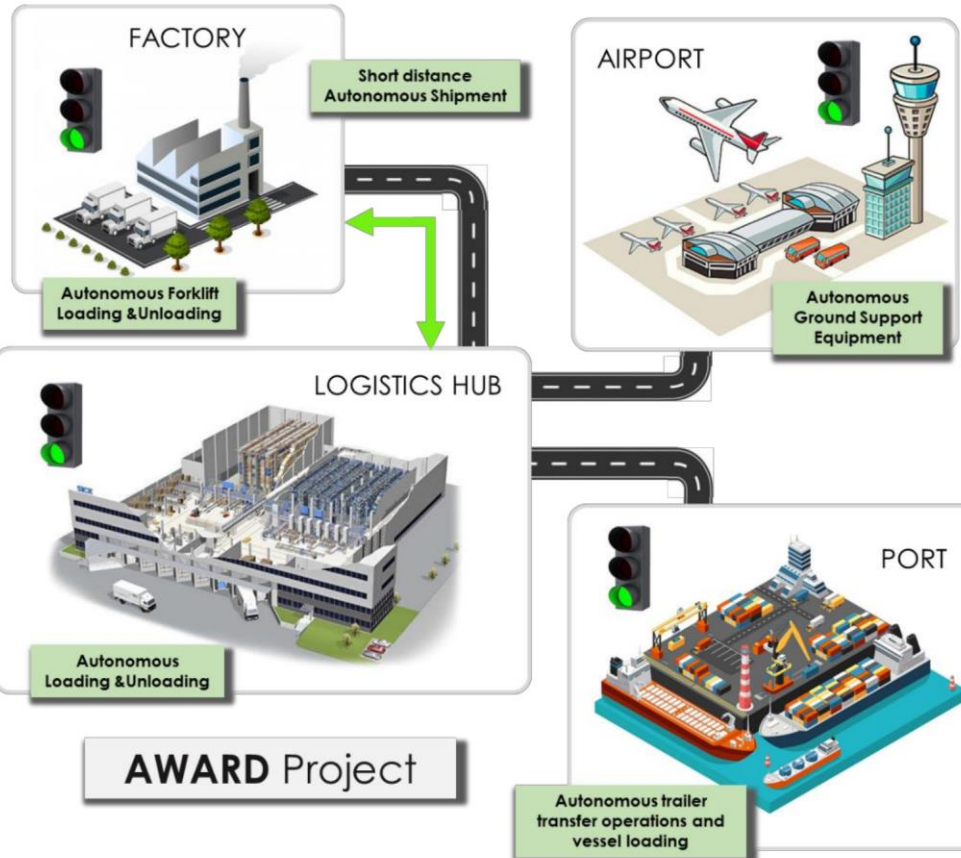
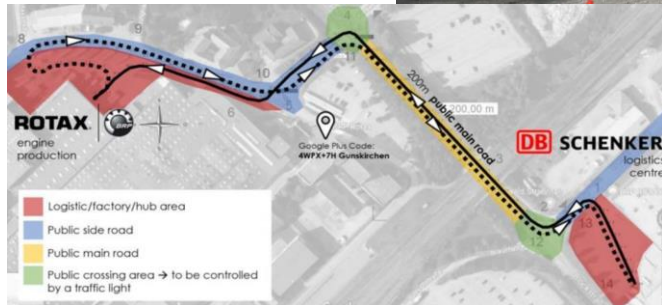
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4 Use Cases in Europe

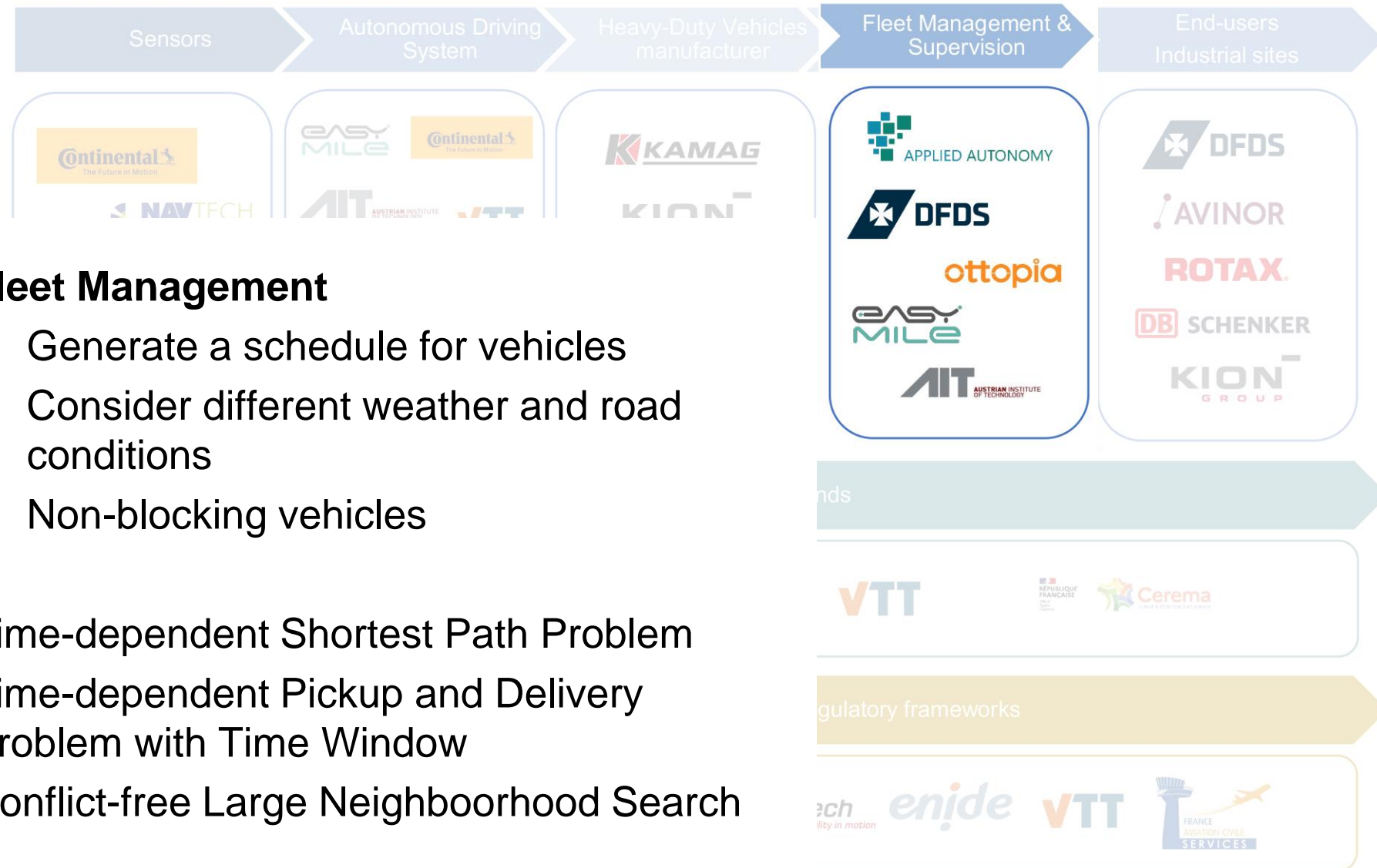


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Consortium



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Agenda

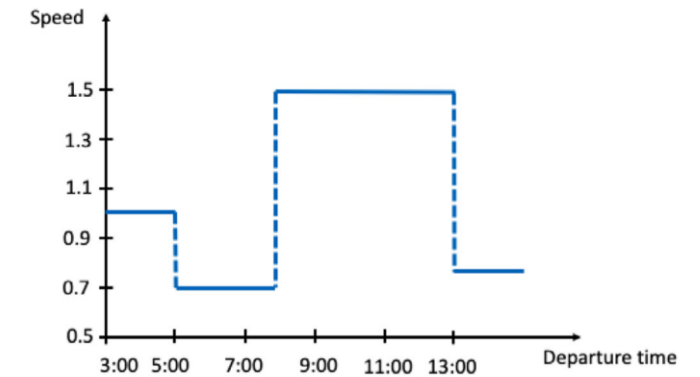


- **Fleet Management**
 - Generate a schedule for vehicles
 - Consider different weather and road conditions
 - Non-blocking vehicles
- Time-dependent Shortest Path Problem
- Time-dependent Pickup and Delivery Problem with Time Window
- Conflict-free Large Neighborhood Search

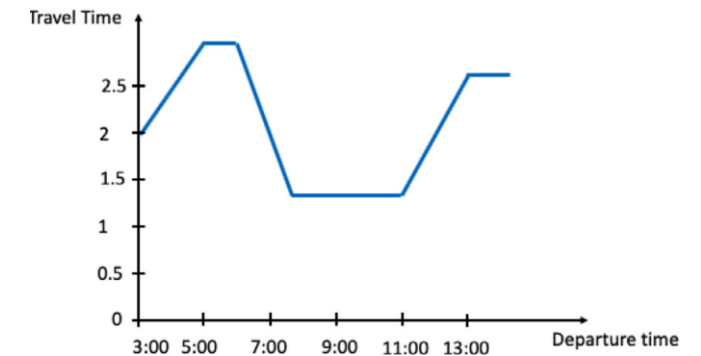
PROBLEM DESCRIPTION

Time-dependent road network

- Road network is a directed graph $G = (V, A)$
 - Set of vertices $V = \{0, \dots, n\}$
 - Set of arcs A , road segments between a pair of vertices
- Each arc $(i, j) \in A$ is associated with a
 - Distance d_{ij}
 - Time-dependent speed function $v_{ij}: t \rightarrow R^+$
 - Time-dependent cost function: $c_{ij}: t \rightarrow R^+$ (= travel time)
- Each speed function is a stepwise function from which a piecewise travel time function can be derived



(a) Speed function



(b) Travel time function
 (Gmira et al., EJOR 2021)

PROBLEM DESCRIPTION

Time-dependent Pickup and Delivery Problem with Time Windows

- Set of n requests r each with:
 - Pickup and delivery node pair $\{i, n + i\} \in V$
 - Demand for pickup q_{i+} and delivery q_{i-}
 - Service time for pickup s_{i+} and delivery s_{i-}
 - Time window $tw_r = [a_i, b_i]$ (either at pickup or delivery, soft)
- Set of m vehicles K starting at parking positions $z_k \in V$ with capacity C_k
- TDPDPTW is to generate routes for all vehicles:
 - **Feasibility:** Conflict-free, time windows, capacity constraint
 - **Objective:** Maximize the number of served requests
Minimize the total lateness and travel time of the routes

SOLUTION APPROACH I

Time-dependent Dijkstra's algorithm with new weighting function

- Time-dependent variant of Dijkstra's algorithm
 - Minimum cost path between a source and destination node at a departure time t
 - **Forward**: travel time for departure time t at source
 - **Backward**: travel time for arrival time t at destination
- Situations defined by operators
 - Time period with a start and an end time $p = [a_p, b_p]$
 - Speed defined over this interval $v_{ij}^+(p): v_{ij}(t)$ for all $t \in [a_p, b_p]$
 - Consider special case when situation is “**road closed**”
 - Either waiting until b_p or detour

SOLUTION APPROACH I

Time-dependent Dijkstra with new weighting function

- Award weighting function
 - Extension of procedure in Ichoua et al., 2003
 - From current node to successor through arc (i, j) at time t
 - Special case road closed
 - No distance covered when waiting

Algorithm 1: Travel costs to a successor of a node.

Input : arc (i, j) , arrival time t , $p = 0$

Output: costs c_{ij}

while $t \geq a_p$ **do**

$p = p + 1$

$c' = b_p - t$

$d' = c_{ij} = 0$

$d = c' \cdot v_{ij}^+(p)$ //if $v_{ij}^+(p)$ is road closed $\rightarrow d = 0$

while $d \leq d_{ij}$ **do**

$c_{ij} = c_{ij} + c'$

$p = p + 1$

$d' = d$ //if $v_{ij}^+(p)$ is road closed $\rightarrow d' = 0$

$c' = b_p - a_p$

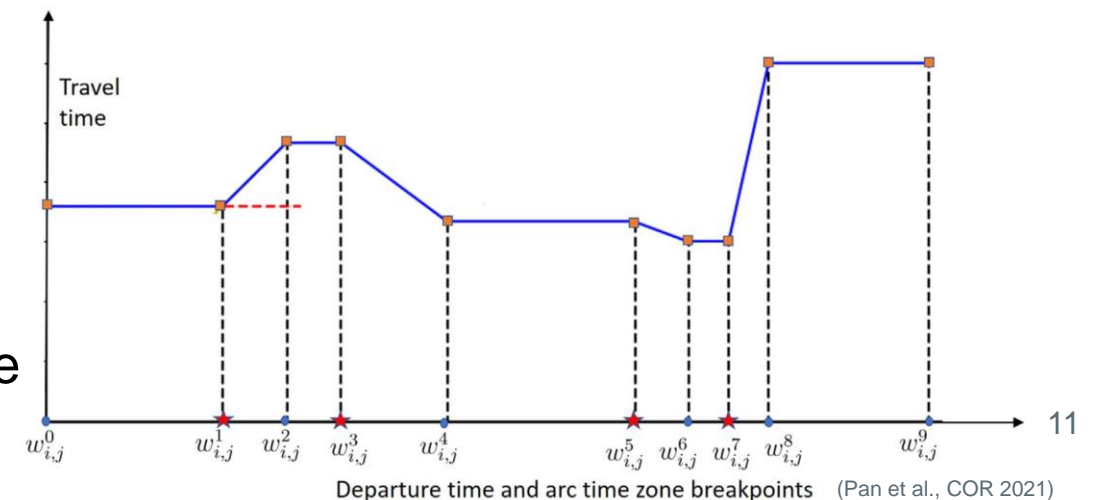
$d = d' + (c' \cdot v_{ij}^+(p))$

return $c_{ij} + ((d_{ij} - d')/v_{ij}^+(p))$

SOLUTION APPROACH II

Time-dependent distance matrix

- Efficiently store travel times between two nodes for a given time
 - Entry for every possible departure → high time resolution (ms)
- Changes in travel time depend on given situations (periods)
 - Calculate the times when the speed changes on arc (ij)
 → Set of breakpoints for arc (ij): $W_{(ij)} = \{w_{(ij)}^0, \dots, w_{(ij)}^c\}$
- Entry in matrix for not changing travel times
 - Travel cost $c_{ij}^{w_{(ij)}^0, w_{(ij)}^1}$ between i and j
 for all time points $t \in [w_{(ij)}^0, w_{(ij)}^1]$
- Travel costs of other instants stored in cache



SOLUTION APPROACH II

Compute time-dependent distance matrix

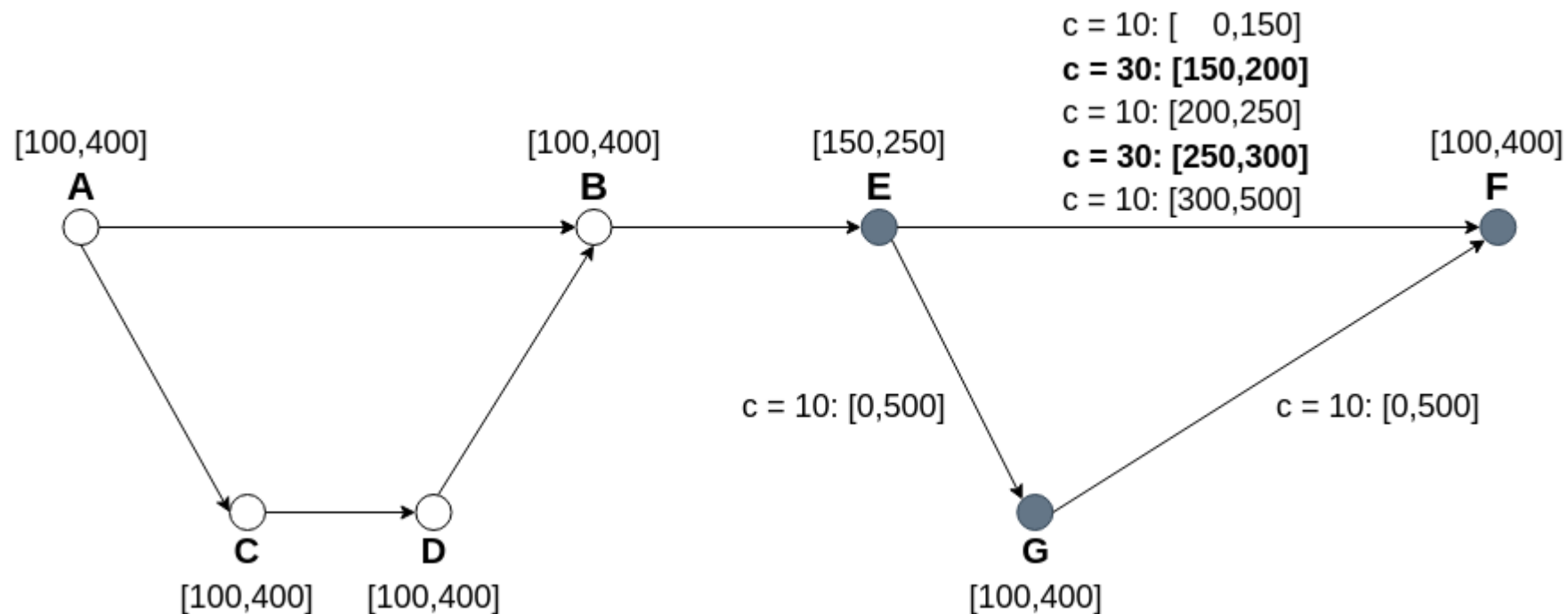
- Calculation of distance matrix entry between node x and node y
 - Determine all departure times $e \in E$ in x and their resulting paths p_{xy}^{e+} (increasing)
 - If there is no change (travel cost and path) \rightarrow entry to matrix

- Determine all departure times $e \in E$ in x :
 - Compute set of paths P with arrival in y at a_y and b_y (backward Dijkstra)
 - For all arcs (i, j) on the paths plus for all incoming arcs $(i', j), i' \neq i$
 \rightarrow compute set of breakpoints $W_{(ij)}$
 - For all breakpoints calculate departure in x and path from x to i, i' respectively
 - Departure e in x : backward Dijkstra with arrival in i at $w_{(ij)}^c$
 - Calculate path from x to i at departure e with forward Dijkstra

SOLUTION APPROACH II

Time-dependent distance matrix

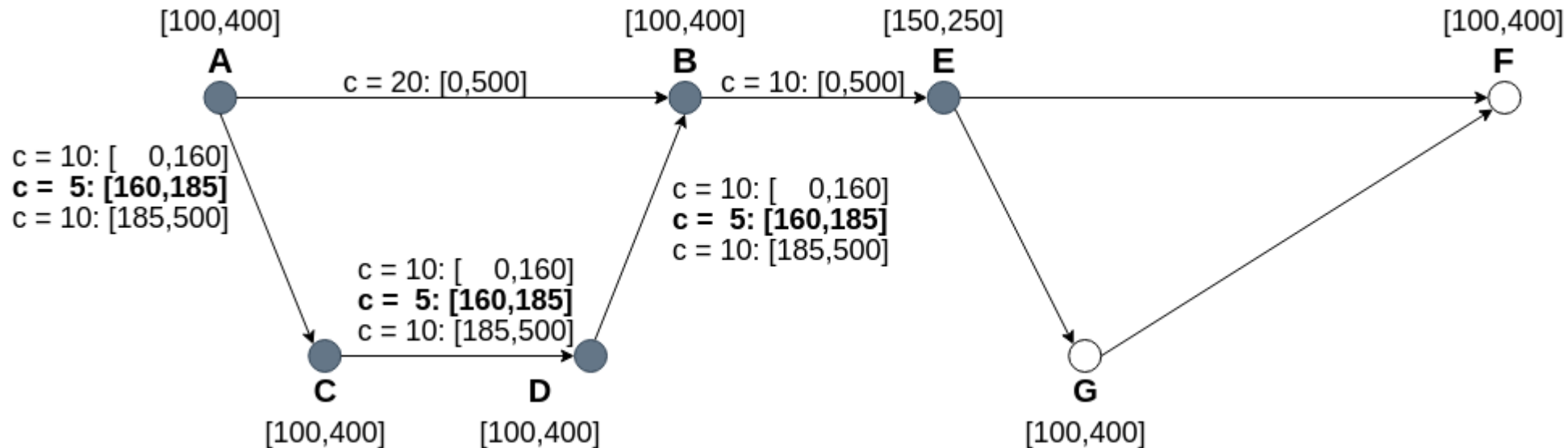
- Calculate matrix entry for **route from node E to node F**
 - Departure at 140 → path is EF with costs 10
 - Departure at 150 → path is EG.GF with costs 20 (detour)



SOLUTION APPROACH II

Time-dependent distance matrix

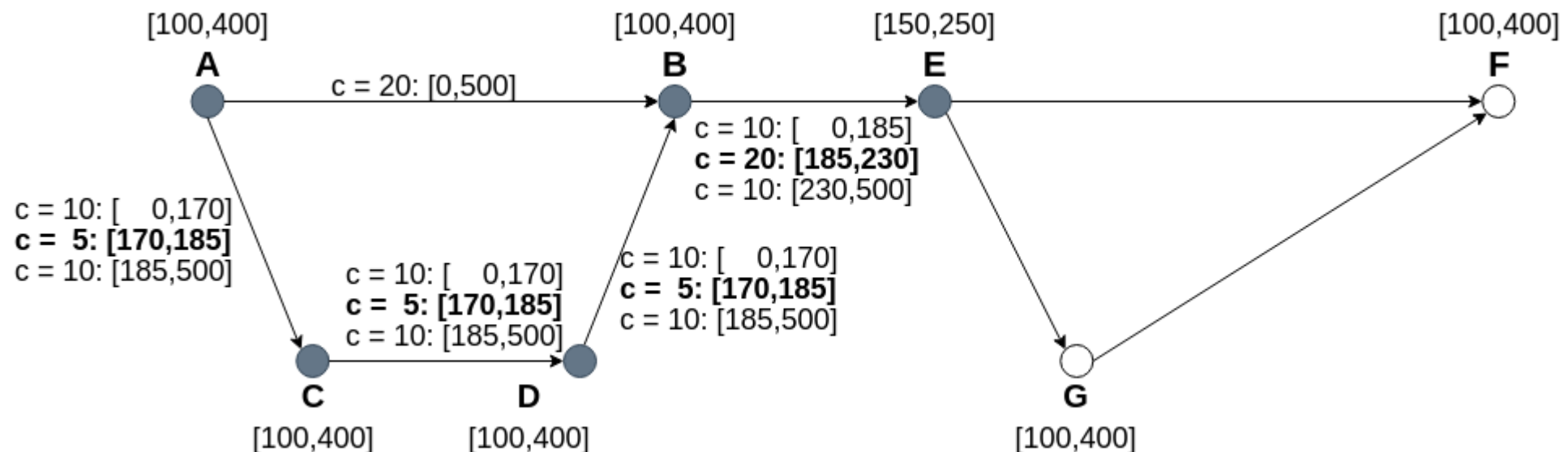
- Calculate matrix entry for **route from node A to node E**
 - Example that considering incoming arcs is necessary:
 - Departure at 120 and 220 → path is AB-BE with costs 30 (**no situations on path!**)
 - Departure at 160 → path is AC-CD-DB-BE with costs 25 (detour)



SOLUTION APPROACH II

Time-dependent distance matrix

- Calculate matrix entry for **route from node A to node E**
 - Example that considering path not only departure in A is necessary:
 - Departure at 160 → path is AB-BE with costs 35
 - Departure at 170 → path is AC-CD-DB-BE with costs 35 (**same costs put different path!**)



SOLUTION APPROACH III

TW Calculator TDPDPTW (soft)

- Improvement method to calculate time windows for the VRPTW
- Introduced by Vidal et al., 2013
- Efficiently calculate penalties according to time window constraints
 - Forward and backward
 - Concatenation

Algorithm 1 : calculateTimeWindowFeasibility(Sequence σ)

$i = first(\sigma); TW = 0; E = e_0; L = l_0; D = s_0;$

for $\forall j$ in $\sigma \setminus 0$ **do**

$\Delta = D - TW + t_{ij};$

$\Delta^{WT} = max\{0, e_j - \Delta - L\};$

$\Delta^{TW} = max\{0, E + \Delta - l_j\};$

$D = D + s_j + \Delta^{WT} + t_{i,j};$

$TW = TW + \Delta^{TW};$

$E = max\{e_j - \Delta, E\} - \Delta^{WT};$

$L = min\{l_j - \Delta, L\} + \Delta^{TW};$

$i = j;$

end for

SOLUTION APPROACH III

TW Calculator TDPDPTW (soft)

- Adapt the sequence-based TW calculation from Vidal
- Store values (duration, lateness) for earliest (E) **and** latest departure (L)
- Calculate departure times for E and L
 - When $E = L \rightarrow$ use forward extension of Vidal
 - Otherwise, if E and/or L changes:
 - Update values and check if there are changes in travel costs
 - Worst case: Update until the beginning of the sequence
- Determine best departure in depot for the node sequence

SOLUTION APPROACH IV

Conflict-free Large Neighborhood Search (LNS)

- (Adaptive) Large Neighborhood Search as in Ropke and Pisinger, 2006
 - Remove requests from solution (up to 60%) and insert them again
 - Destroy operators: Random and Worst
 - Repair operators: Greedy Repair, Regret Repair
- Only conflict-free solutions are allowed
 - Conflict: if vehicles blocks another vehicle at a stop
 - Whenever feasibility of solution is calculated:
 - Check for each vehicle route if it passes a stop where another vehicle stays
→ Solution is infeasible

PRELIMINARY RESULTS

Time-dependent PDPTW with no conflicts

- Instances on road network generated by an operator
 - Road network consists of 173 nodes and 252 edges
 - 30, 40, 50 customer requests, 5 or 10 situations, and 3 vehicles
- First results show benefit of implementing a time-deent distance matrix
 - Percentage of runtime for travel time calculation to total runtime

# requests	5 situations		10 situations	
	No matrix	Matrix	No matrix	Matrix
30	3.08	1.67	3.61	1.46
40	2.34	1.01	2.66	0.79
50	4.20	2.01	4.09	1.34

PRELIMINARY RESULTS

Time-dependent PDPTW with no conflicts

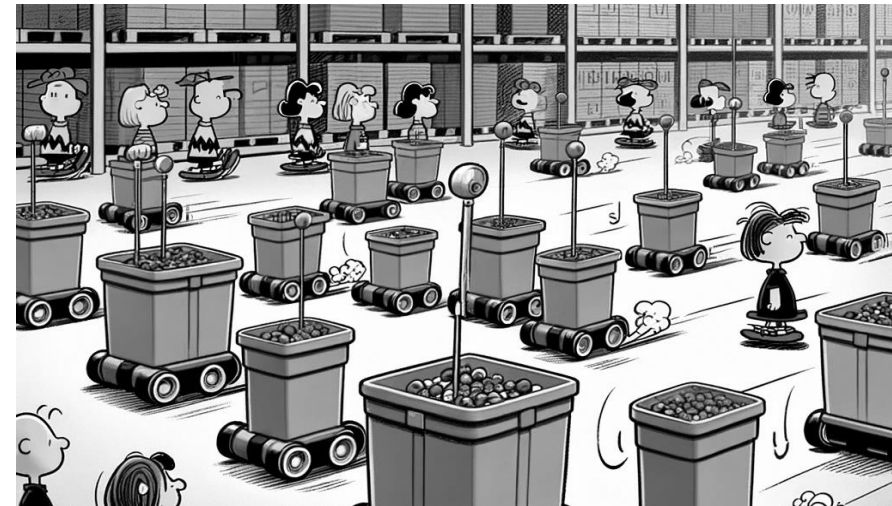
- Efficient time window calculation with time-dependent travel times
 - Implemented and tested for forward method (many possibilities)
 - Extension to backward calculation (allowing concatenation in constant time)
 - Compare to other time window calculations
- Conflict-free vehicle routes are generated for all instances
 - Dealing with larger instances (more vehicles):
 - Operator: Relocate vehicle to parking position

THANK YOU!

Ulrike Ritzinger

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ulrike.ritzinger@ait.ac.at



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