

# Application oriented vehicle problems in public bus transportation

Gábor Galambos

University of Szeged, Hungary

*Joint work with*

*Viktor Árgilán, János Balogh, József Békési,  
Balázs Dávid, Miklós Krész, Attila Tóth*

Workshop on Traffic Optimization  
Heidelberg

8.october 2015.





# Szeged

## The city of Szeged





# Szeged

## The University of Szeged

---





# Szeged

## Public transportation in Szeged



- Motivation
- R&D Project



# Overview

---

- Scheduling problems in public transportation
- Vehicle scheduling
- Vehicle assignment
- „Driver-friendly” vehicle schedules
- Vehicle rescheduling



# Scheduling problems in public transportation

---

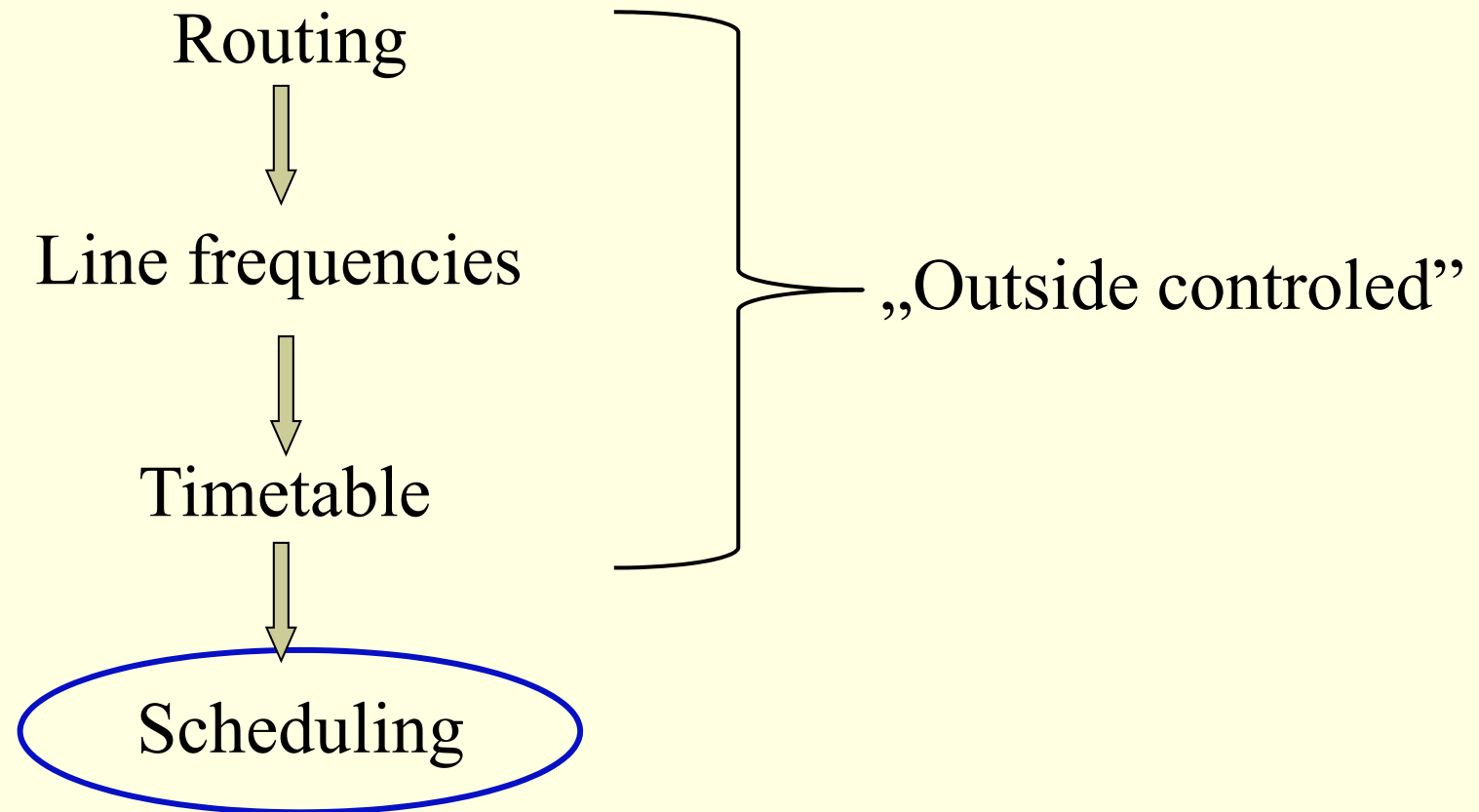
For public transportation services certain number of stations, and previously determined – bus or other vehicles – lines are given. Each line connects a pair of stations.

The lines are fixed in timetables which provides the departure and arrival time of the trips for each line, and – sometimes – further services for each day are also fixed.

One of the most important subject of a public transport company is to **minimize its operational costs**.



# Scheduling problems in public transportation (Operations planning)





# Scheduling problems in public transportation

This complex problem is divided into subtasks and they are to be solved as separated optimization problems

Vehicle scheduling



Driver scheduling



Driver rostering





# Vehicle Scheduling Problem





# Vehicle scheduling problem

## Input:

- Timetabled trips
- Deadhead trips
- Depots
  - Bus types (solo, double, etc.)
  - Locations

## Costs:

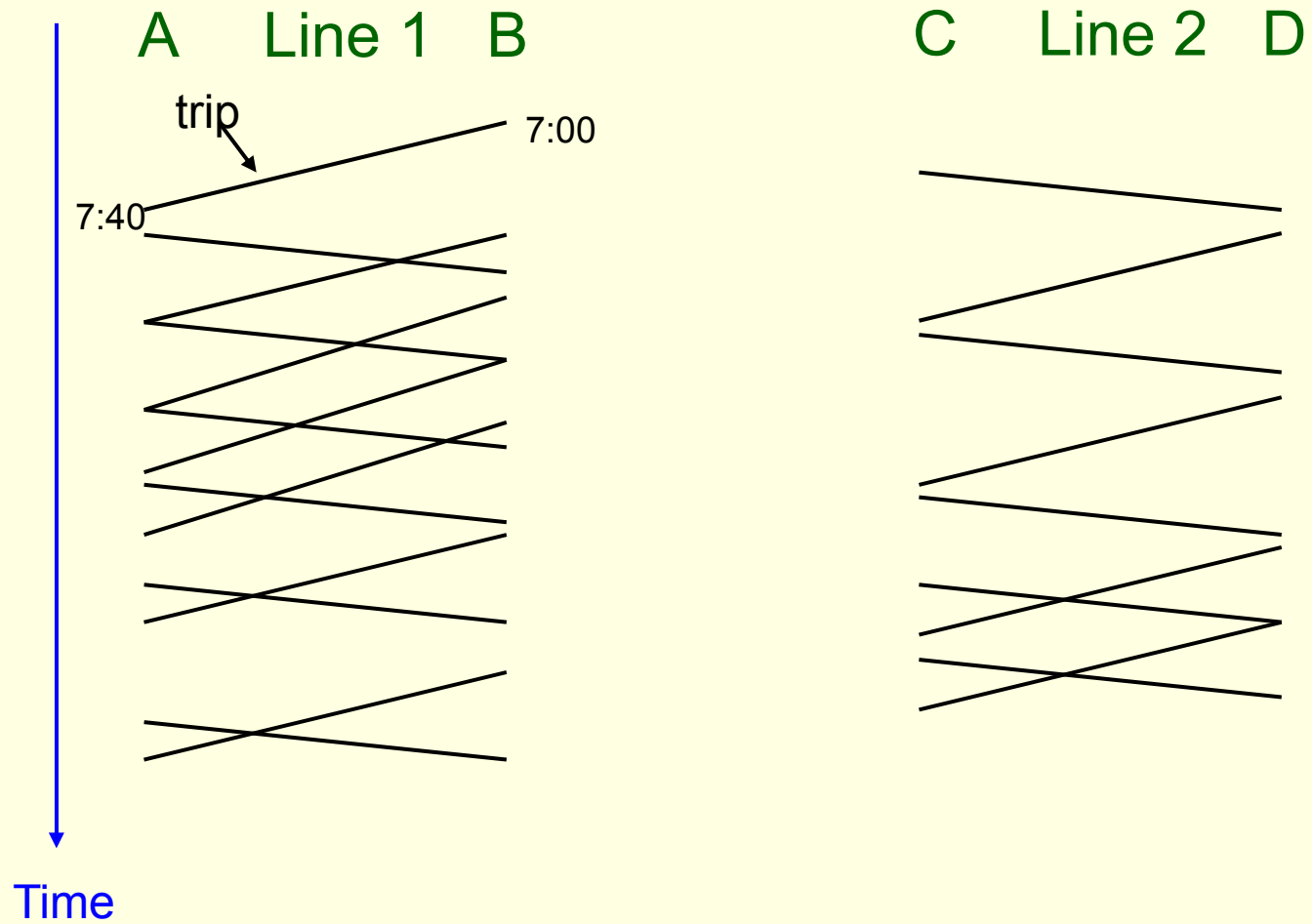
- Daily cost of a vehicle (maintenance)
- Cost of the covered distance (transportation costs, deadhead costs)

## Aim:

- Execute each trip exactly once (with regards to depot compatibility and capacities), minimizing the cost

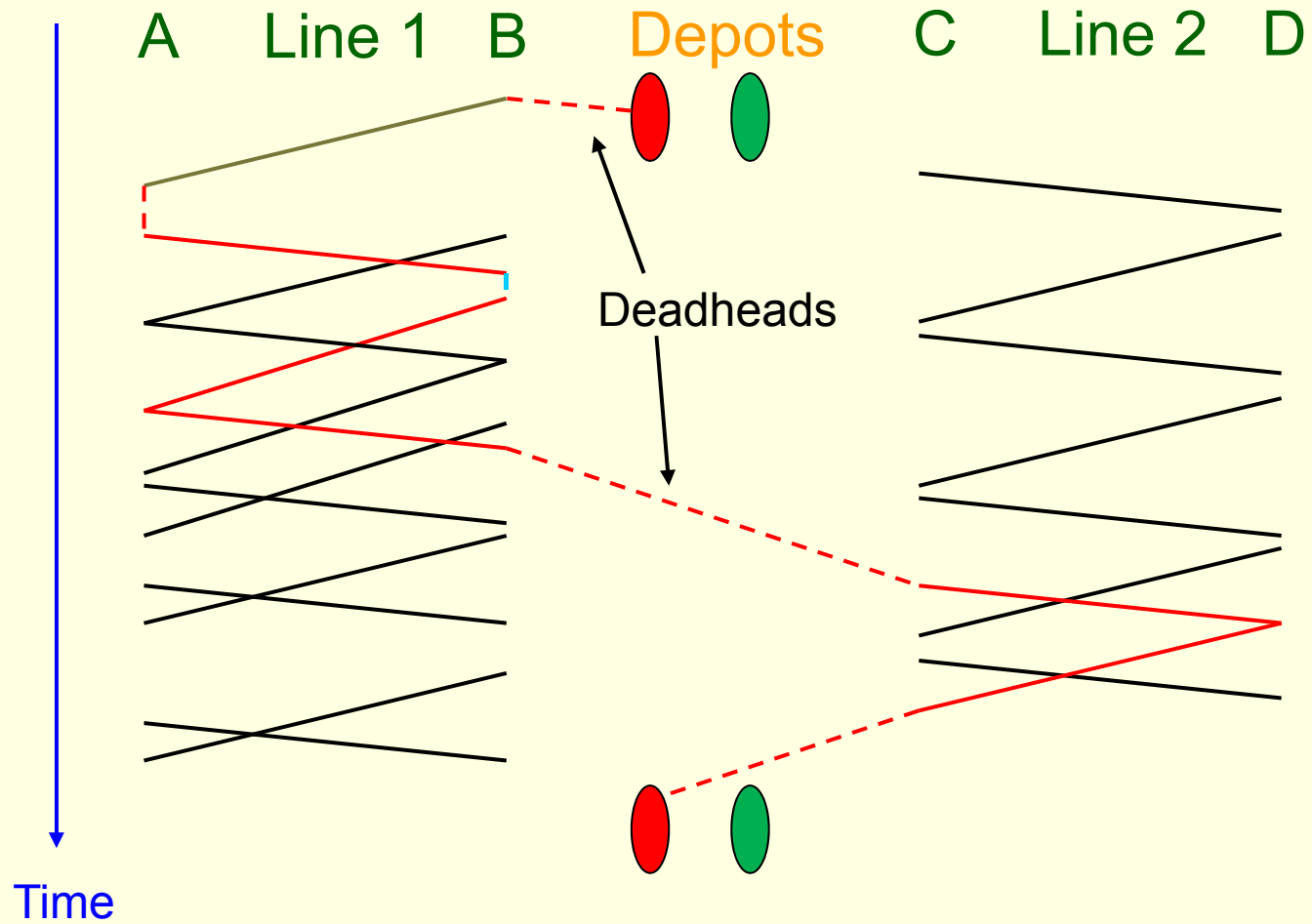


# Vehicle scheduling problem



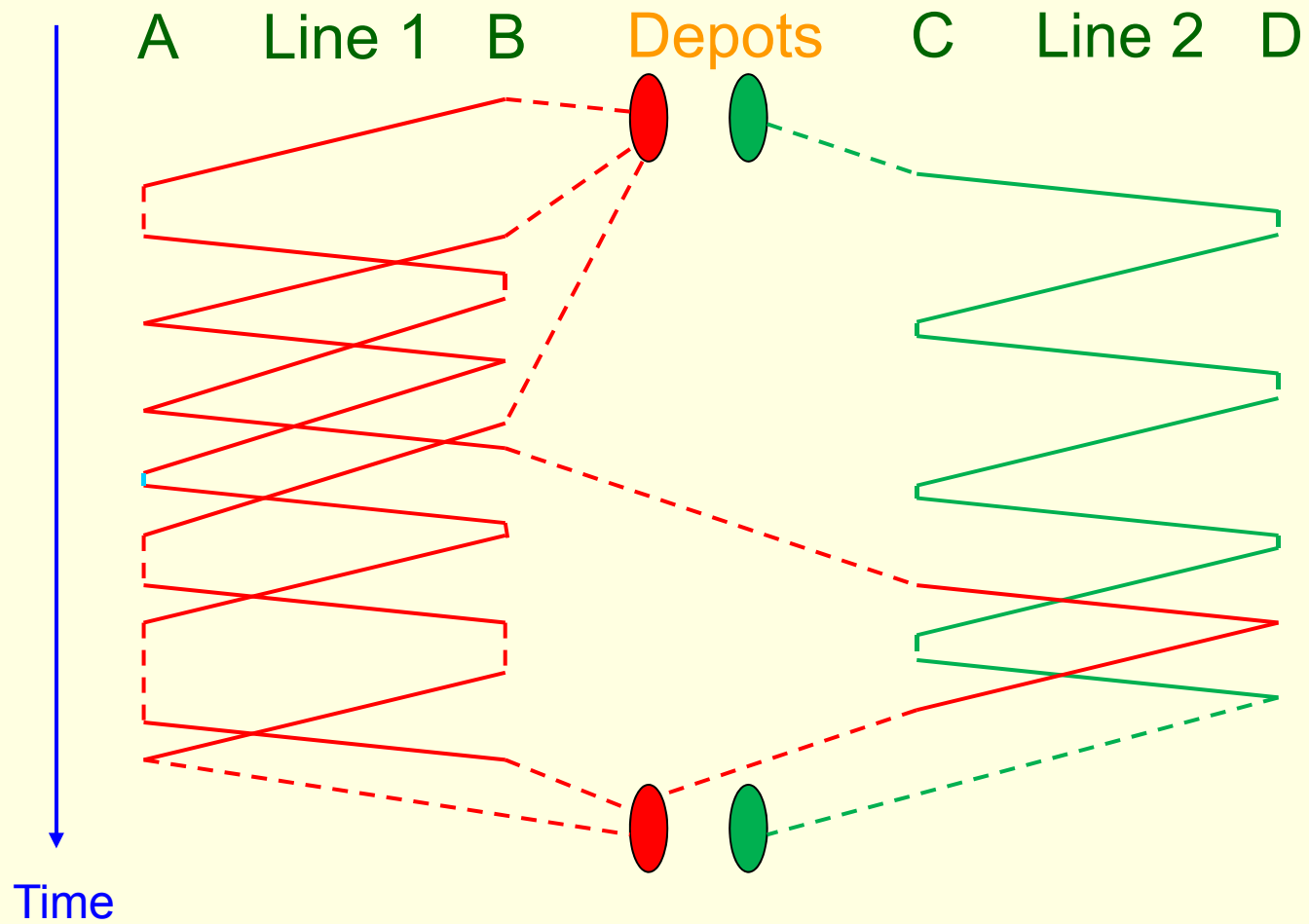


# Vehicle scheduling problem





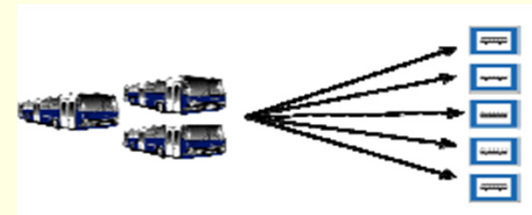
# Vehicle scheduling problem



# Vehicle scheduling problem

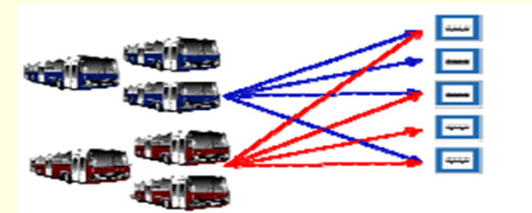
## ■ Single-Depot Vehicle Scheduling Problem(SDVSP)

- Solvable in polynomial time  
(Matching problem,  
Minimum cost network flow)



## ■ Multiple-Depot Vehicle Scheduling Problem(MDVSP)

- NP-hard (Bertossi et al., 1987)



Solution by multi-commodity network flow minimization

- Connection based network model (Löbel, 1997)
- Time space network model (Kliewer et al., 2006)



# Connection based network

$D$  = set of depots

$U$  = set of trips

$V$  = set of vehicles

$D_u \subseteq D$  set of those depots which can serve the trip  $u \in U$

$U_d \subseteq U$  set of those trips which can be served by the depot  $d \in D$

$at(d)$  = arrival time to the depot  $d$

$dt(d)$  = departure time from the depot  $d$

$at(u)$  = arrival time of the trip  $u$

$dt(u)$  = departure time of the trip  $u$

Trips  $u_i$  and  $u_j$  are compatible if  $at(u_i) \leq dt(u_j)$  and  $dt(u_j) - at(u_j) \geq T_{ij}$



# Connection based network

(define a network)

Let  $N$  be the set of vertices of the network. Then

$$N = \{ dt(u) \mid u \in U \} \cup \{ at(u) \mid u \in U \} \cup \{ dt(d) \mid d \in D \} \\ \cup \{ at(d) \mid d \in D \}$$

If  $E_d$  is the set of timetabled trips belonging to the depot  $d$ , then

$$E_d = \{ (dt(u), at(u)) \mid u \in U_d \}, \quad \forall d \in D \\ B_d = \{ (at(u), dt(v)) \mid u, v \in U_d \text{ are compatible} \}, \forall d \in D$$

where  $B_d$  is the set of the depot deadhead trips.





# Connection based network

(define a network)

Let  $R_d$  be the set of ingoing and outgoing trips of depot  $d$ , then

$$R_d = \{ (dt(d), dt(u)), (at(u), at(d)) \mid u \in U_d \}, \quad \forall d \in D$$

Let  $K_d$  is the set of depot-circle arcs of the depot  $d$ , then

$$K_d = \{(at(d), dt(d))\}, \quad \forall d \in D$$

Then the set of arcs belonging to the depot  $d$  in the network is

$$A_d = E_d \cup B_d \cup R_d \cup K_d$$

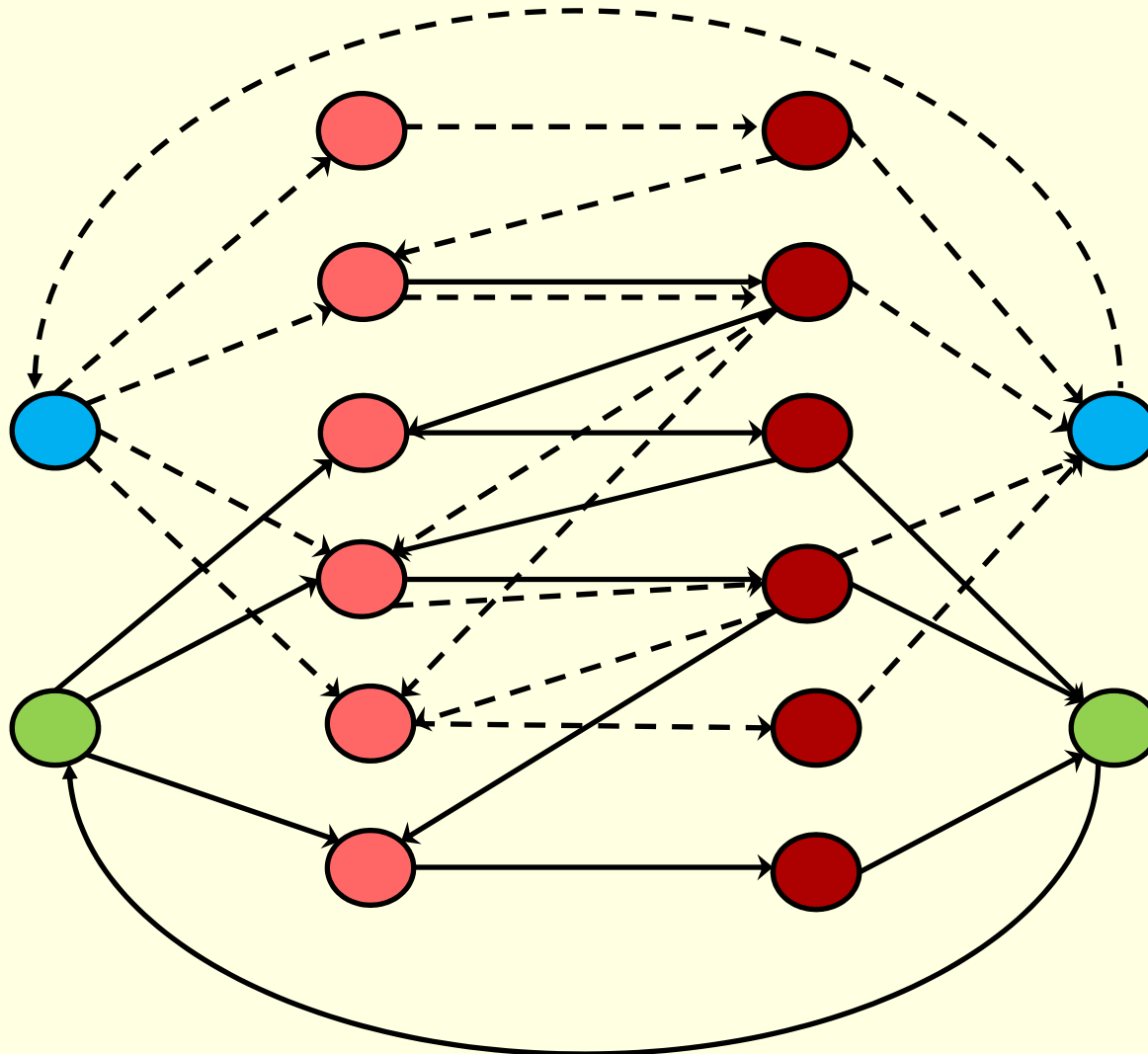
and

$$E = \bigcup_{d \in D} A_d$$



# Connection based network

(define a network)





# Connection based network

(define MDVSP on the network)

So, we have a directed graph  $G = (N, E)$ . We can define the MDVSP on this graph as follows:

Let  $X$  be an integer vector with dimension of  $|E|$ . Its component is belonging to the arc  $e \in E$  is  $x_e^d$  if  $e \in A_d$ .

$X$  represent a multicommodity flow.

Considering a schedule  $S$ . Then  $x_e^d$  we can state

$$x_e^d = \begin{cases} 1, & \text{if } x_e^d \in S \\ 0, & \text{otherwise} \end{cases}$$



# Connection based network

(define MDVSP on the network)

1. No common edges for two trips:

$$\sum_{d \in Du, e=(dt(u),at(u)) \in Ed} x_e^d = 1 \quad \forall u \in U.$$

2. Vehicles must return to the depots:

$$\sum_{e \in n^+} x_e^d - \sum_{e \in n^-} x_e^d = 0, \quad \forall u \in U, \forall n \in N$$

3. If we have capacity constraint for the depots:

$$x_{at(d),dt(d)}^d \leq k_d$$



# Connection based network

(define MDVSP on the network)

---

The objective function is

$$\min \sum_{e \in E} c_e x_e$$

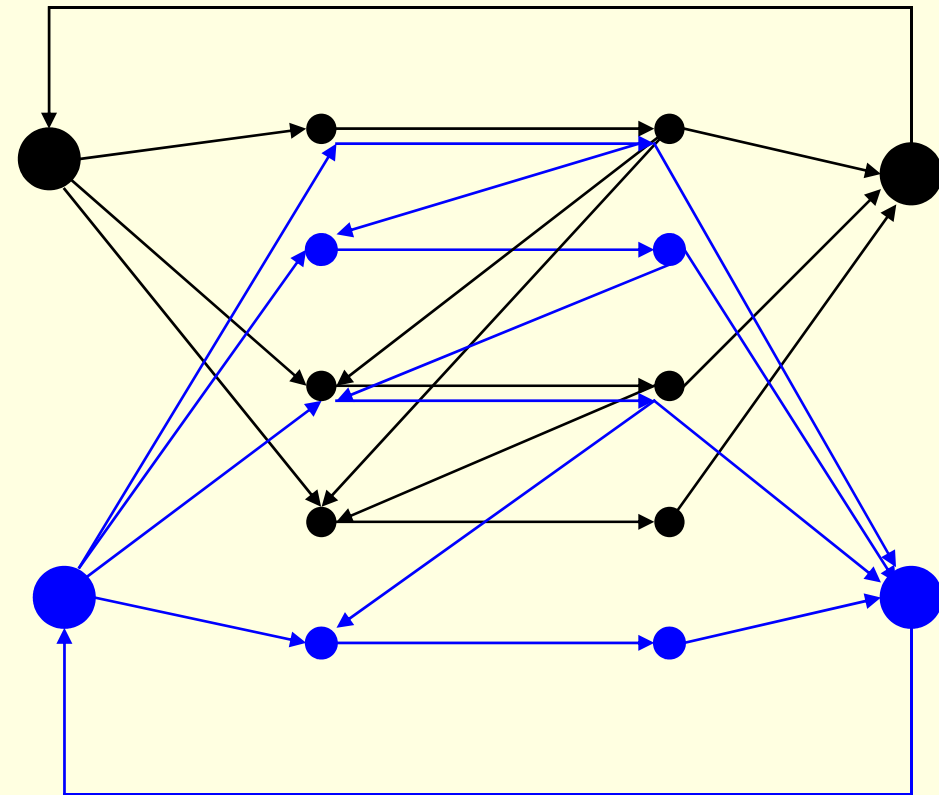
where  $c_e$  is the cost of the edge  $e$ .



# Connection based network

(define MDVSP on the network)

- Large size
  - Due to high number of possible deadheads  
(~5.000.000 variables\*)

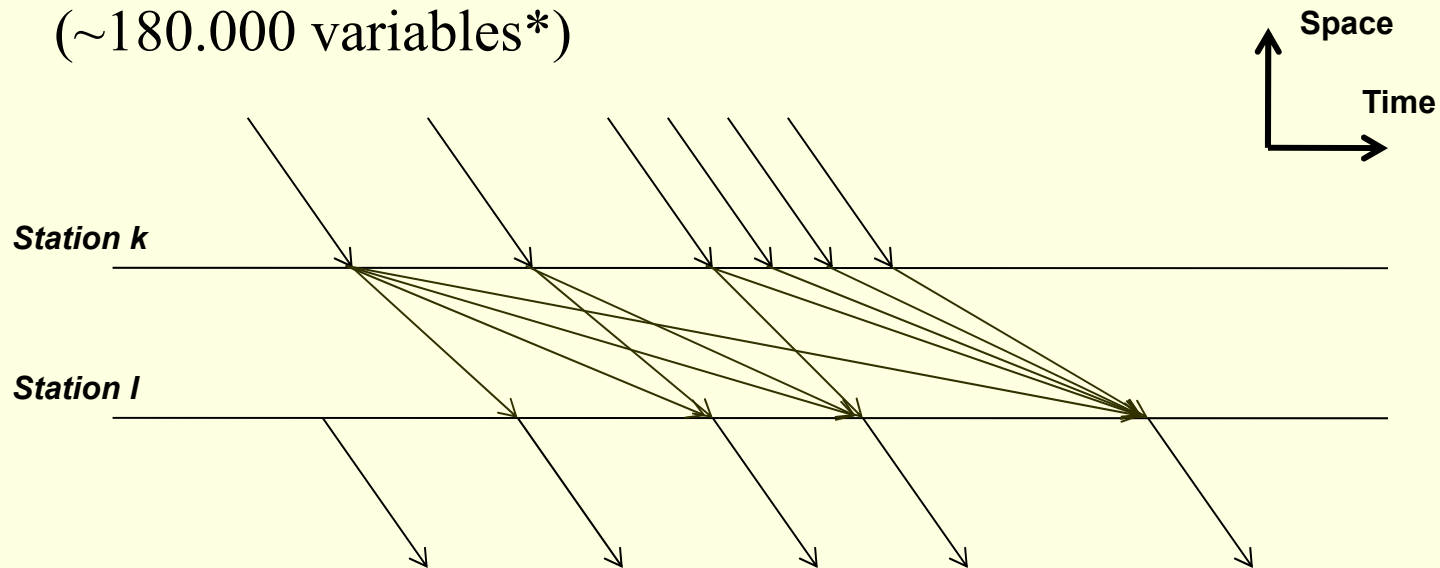


\*Tisza Volán test case



# Time space network

- Introduces timelines
- Nodes are points of time
- Aggregates possible connections into waiting edges
  - Significant decrease in model size  
(~180.000 variables\*)

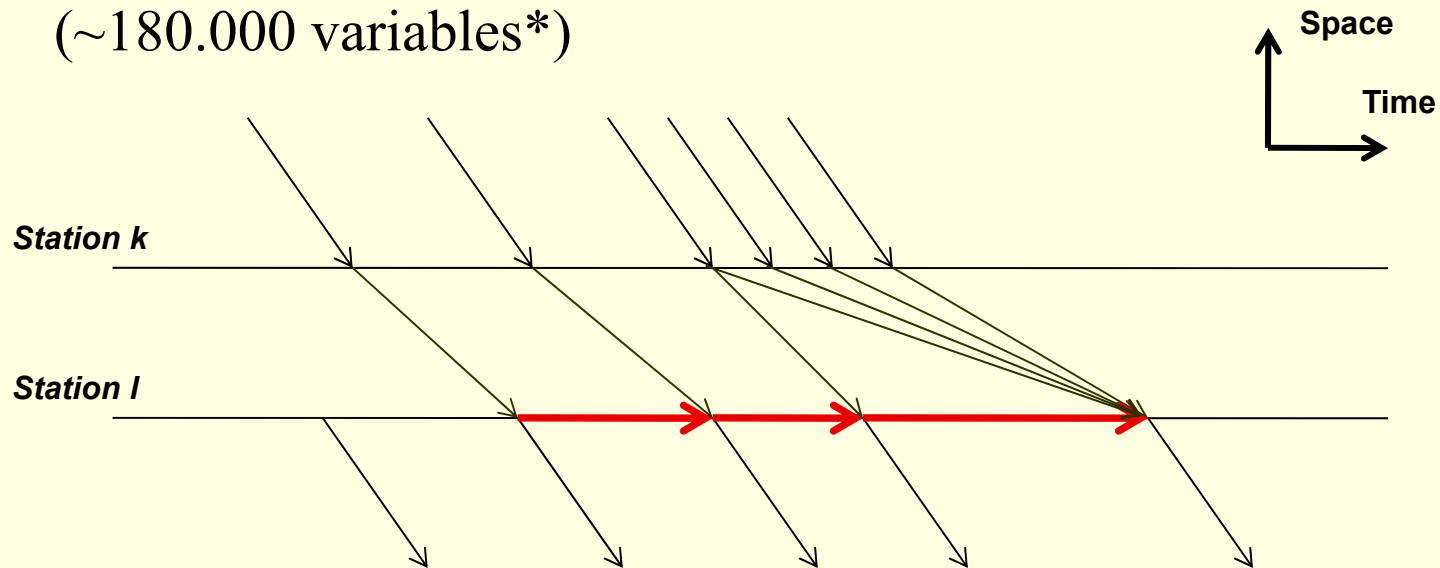


\*Tisza Volán test case



# Time space network

- Introduces timelines
- Nodes are points of time
- Aggregates possible connections into waiting edges
  - Significant decrease in model size  
(~180.000 variables\*)



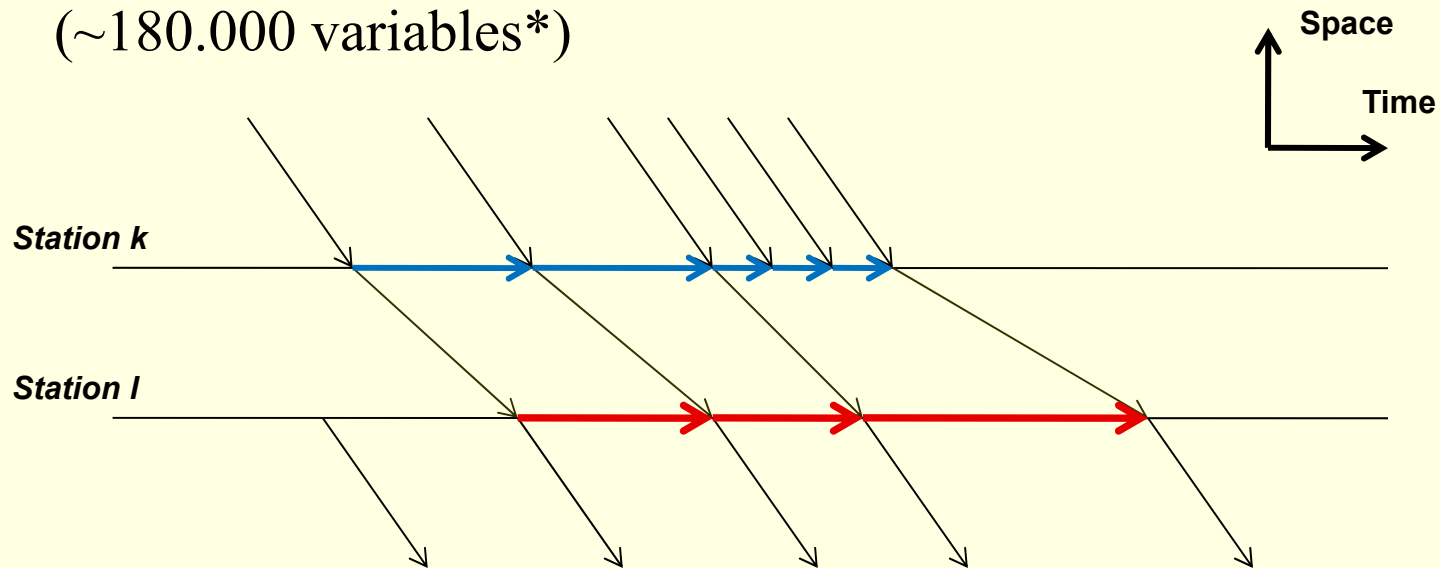
\*Tisza Volán test case





# Time space network

- Introduces timelines
- Nodes are points of time
- Aggregates possible connections into waiting edges
  - Significant decrease in model size  
(~180.000 variables\*)



\*Tisza Volán test case



# MDVSP Solution Methods

- Mathematical methods
  - Exact solution of the IP
  - Column generation
  - Lagrangian relaxation
- Combinatorial
  - Tabu search
- Mixed methods
  - Rounding heuristic
  - Variable fixing (Gintner et al., 2005)



# Application oriented (greedy) variable fixing (Dávid & Krész, 2013)

---

- Based on the idea of Gintner et al.
- An SDVSP relaxation is solved for the problem
- Consecutive trips of the result are fixed together, if they share some common property:
  - Same number of depots (at least one compatible)
  - Belongs to the same bus line (and have a compatible depot)



## Case study: Szeged urban bus transportation

---

- Fourth largest city of Hungary
- Population: 168.273
- Bus company: Tisza Volán, Urban Transport Division, [www.tiszavolan.hu](http://www.tiszavolan.hu)
- 40 lines, 120 buses
- 4 depots: combinations of
  - Conventional vs articulated (gas fuel)
  - Low floor vs normal



# Test results

	Time ratio(%)*	„bad” working pieces**	Cost difference max.(%)
Rounding	50,53	3,5	0,1029
Variable-fixing	14,44	13,75	0,2694
Greedy-chains	7,97	4,75	1,0219

•Comparing to „first feasible exact solution”

\*\*not „driver-friendly”

## ■ Efficiency

- running time: variable fixing and greedy-chains
- cost: all heuristics perform well

## ■ Integrity with driver rostering

- a rounding and greedy-chains

➔ The greedy-chains heuristics satisfies all aspects

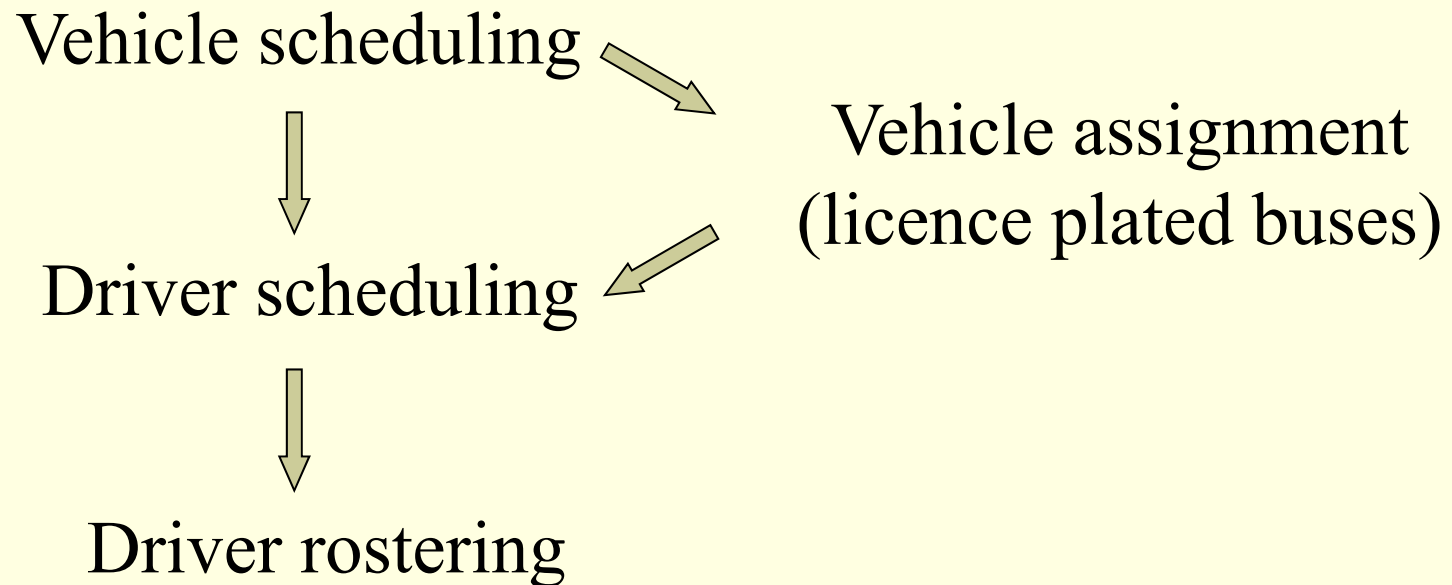


# Vehicle Assignment Problem





# Scheduling problems in public transportation





# Vehicle assignment problem

- Problem of the VSP:
  - Vehicles in the depots are considered uniform (low floor)
  - Hard to integrate vehicle specific tasks to the model
    - refueling, parking, etc.
- Aim: Solving vehicle scheduling with assigning physical buses
- Satisfy vehicle-specific requirements
  - Refueling
  - Parking
  - Maintenance
- Classical MDVSP models do not support these kind of requirements





# Vehicle assignment problem

## Solution methods

- Sequential approach
  - Transform an initial vehicle schedule with regards to the vehicle-specific tasks
- Integrated model
  - Build a model for the problem that takes these tasks into consideration
  - It is a 3D assignment model:
    - - lines
    - - „physical” vehicles
    - - vehicle-specific events



# Vehicle assignment problem

## Sequential approach (Árgilán et al, 2013)

---

### Input:

- Set of vehicle schedules
  - Solving an MDVSP model
- Set of vehicles
- Refueling stations with parameters
  - Fuel types, capacity of fuel pumps, opening times
- Fuel pumps:
  - Service times with fixed length
  - Service time may vary depending on fuel type



# Vehicle assignment problem

## Sequential approach

- Variable:

$$X_{ijkt} = \begin{cases} 1 & \text{if schedule } i \text{ is refueled with vehicle } j \text{ at station } k \text{ in time } t \\ 0 & \text{otherwise} \end{cases}$$

- $X_{ijkt}$  exists (not forbidden), if

- The depot  $i$  corresponds to the depot of vehicle  $j$
- The fuel type  $k$  corresponds to the fuel type of vehicle  $j$
- Schedule  $i$  is idle in time period  $t$
- The running distances allow the refueling at time  $t$
- Other conditions can be added



# Vehicle assignment problem

## Sequential approach

$$\min \sum_{i,j,k,t} c_{ijkl} x_{ijkl}$$

s.t.

$$\sum_{j,k,t} x_{ijkl} = 1 \text{ for all } i, (1 \leq i \leq n)$$

$$\sum_{i,k,t} x_{ijkl} \leq 1 \text{ for all } j, (1 \leq j \leq m)$$

$$\sum_{i,j} x_{ijkl} \leq d_{kt} \text{ for all } k, t (1 \leq k \leq p), t (1 \leq t \leq T_d(k))$$



# Vehicle assignment problem

## Sequential approach

---

- Solve the multi-dimensional assignment problem above
- Problem: „dense” schedules
  - Change for a different bus
  - Remove events
  - New buses



# Vehicle assignment problem

## Sequential approach – Test cases

Problem	Trips	Depot	Sched. phase	Assign. phase	B	SB	Decr.
<i>Szeged#1</i>	2724	4	1179	14	107	96	8,28%
<i>Szeged#2</i>	2690	4	872	8	107	96	8,28%
<i>Szeged#3</i>	1981	4	431	5	65	54	14,92%
<i>Szeged#4</i>	1768	4	250	1	54	44	16,52%



# Vehicle assignment problem

## Integrated approach

(Dávid et al, 2014)

### Input:

- Set of trips ( $T$ )
- Set of different vehicle types ( $V$ )
  - common structural parameters:
  - depot, fuel type, capacity, etc.
- Refueling possibilities ( $R$ )
  - All the legal time periods, where a vehicle can be refueled
  - capacity  $k_r$  for every  $r \in R$



# Vehicle assignment problem

## Integrated approach

- $S_v$ : Set of legal schedules of vehicle  $v$

$$x_s = \begin{cases} 1 & \text{if schedule } s \text{ is carried out} \\ 0 & \text{otherwise} \end{cases}$$

- Variable to connect tasks and schedules

$$a_{is} = \begin{cases} 1 & \text{if task } a \text{ is carried out in schedule } s \\ 0 & \text{otherwise} \end{cases}$$

- (tasks can be trips, refuelings, or others)





# Vehicle assignment problem

## Integrated approach

$$\sum_{v=1}^n \sum_{s \in S_v} c_s x_s \rightarrow \min$$

s.t.

$$\sum_{v=1}^n \sum_{s \in S_v} a_{ts} x_s = 1 \quad \forall t \in T$$

$$\sum_{v=1}^n \sum_{s \in S_v} a_{rs} x_s \leq k_r \quad \forall r \in R$$

$$\sum_{s \in S_v} x_s \leq q_v \quad \forall v \in V$$

$$x_s \in \{0, 1\} \quad \forall s \in S_v$$



# Vehicle assignment problem

## Integrated approach

---

- Solve the problem using column generation
  - Master problem
  - Pricing problem
    - Resource constrained shortest path
    - Check refueling capacities



# Vehicle assignment problem

## Integrated approach – Test cases

<b>Problem</b>	<b>Trips</b>	<b>Depots</b>	<b>Integrated step</b>	<b>CB</b>	<b>IB</b>	<b>Decr.</b>
<i>Szeged#1</i>	2724	4	4264	107	95	13,34%
<i>Szeged#2</i>	2690	4	3542	107	96	14,22%
<i>Szeged#3</i>	1981	4	2745	65	54	16,04%
<i>Szeged#4</i>	1768	4	2687	54	44	19,63%



# „Driver-friendly” Vehicle Scheduling





# „Driver-friendly” vehicle scheduling (Árgilán et al., 2011)


---

- Schedules given by vehicle scheduling/assignment
- The problem of „dense” schedules
  - Driver rules have to be considered
- Sequential heuristic
  - Uses the results of the previous phases



# „Driver-friendly” vehicle schedules Classification

Class A:  length  $\leq M$ , needs 1 driver

Class B:   $M < \text{length} \leq 1,5 * M$ , needs 2 drivers  
(1,5 drivers actually)

Class C:   $1,5 * M < \text{length} \leq 2M$ ,  
needs 2 drivers



# „Driver-friendly” vehicle schedules Transformation

---

- Cut-and-join for trips of class B
- Split trips of class C
- Different number of break depending on length
- Insertion of breaks
  - No transformation needed
  - Trips have to be removed
    - Can be moved to another schedule
    - Have to be moved to the „free-list”



# „Driver-friendly” vehicle schedules

## Iteration steps

---

- New input for vehicle scheduling
  - „Free-list”
- Classification and Transformation on the new schedules
- Do while the „free-list” is empty





# „Driver-friendly” vehicle scheduling

