

# A Markovian approach for dynamic transit assignment: Results in subnetworks of Santiago

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25 de Octubre, 2023, Valparaíso, Chile

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# The MDTrA Approach

## Overview

- It deals with time-dependent demand and supply and uncertainty in passengers' route-choices
- Adapts the *Markovian Traffic Equilibrium* concept (Baillon and Cominetti, 2008) to a dynamic and stochastic context (Addison and Heydecker, 1996, 1998; Heydecker and Addison, 2005; Han, 2003; Heydecker and Addison, 1997)
- Integrates a *reasonability* concept
- It is arc-based, rather than route-based
- The route-choice results from recursive arc-choices

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Overview

The MDTrA model

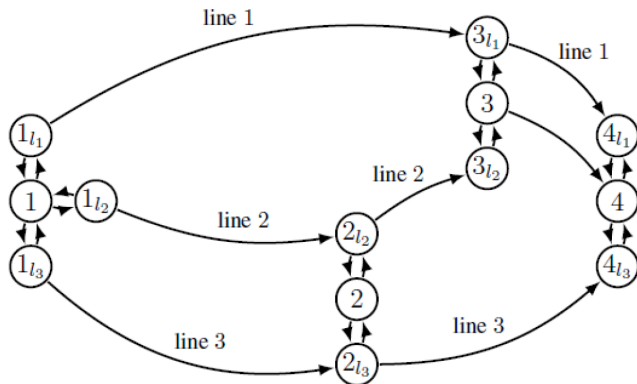
The MDTrA algorithm

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Summary

## Underlying digraph

A representation of the options



## Travel is focused on destinations

### Definition

Given a destination node  $d$ , an arc  $(i, j)$  will be a *reasonable arc towards  $d$*  if the minimum cost of going from  $j$  to  $d$  is not greater than the minimum cost of going from  $i$  to  $d$ .

The set of all reasonable arcs towards  $d$  is denoted  $R_d$

## Travel is focused on destinations

### Definition

Given a destination node  $d$ , an arc  $(i, j)$  will be a *reasonable arc towards  $d$*  if the minimum cost of going from  $j$  to  $d$  is not greater than the minimum cost of going from  $i$  to  $d$ .

The set of all reasonable arcs towards  $d$  is denoted  $R_d$

### Assumption

Motorists only travel through reasonable arcs to their destinations



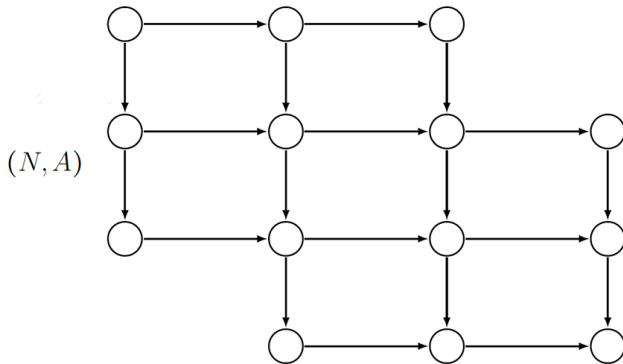
## The MDTrA model

### Overview

- It is composed by *demand and supply profiles*, a *time and cost functions* and an *arc-choice model*
- The demand and supply profiles are exogenous
- The time and cost functions classify arcs by type
- The arc choice model is a dynamic version of the one associated with the MTE

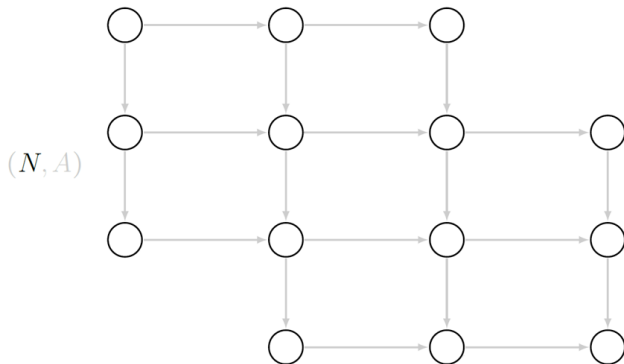
## Transport network

### The digraph



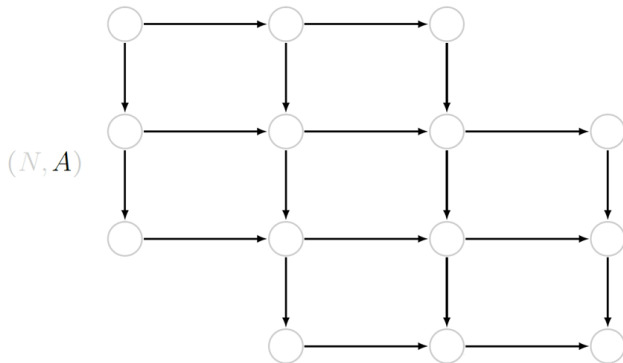
## Transport network

### The set of nodes



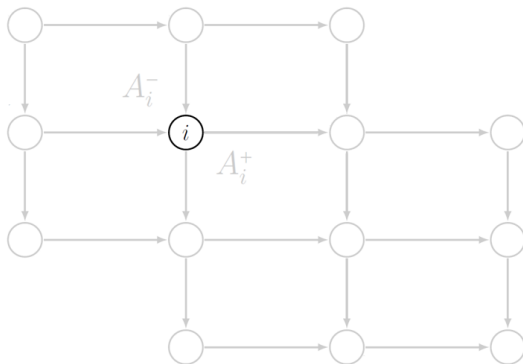
## Transport network

### The set of arcs



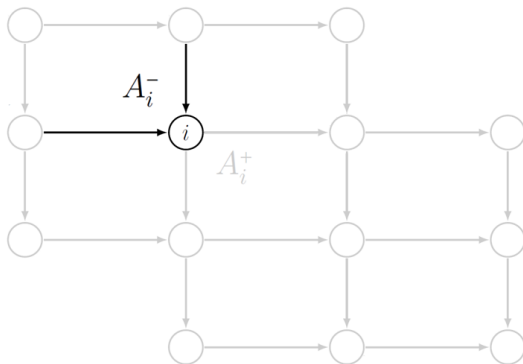
## Sets associated with a node

$$i \in N$$



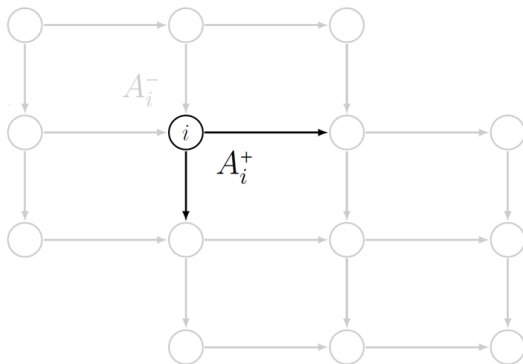
## Sets associated with a node

Set of incoming arcs to a node,  $A_i^-$



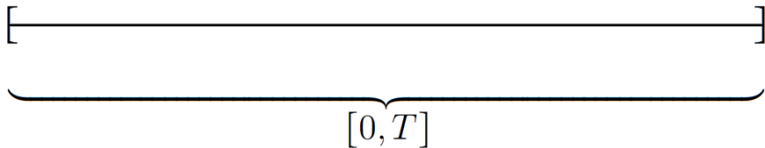
## Sets associated with a node

Set of outgoing arcs from a node,  $A_i^+$



## Analyzed time period

The time interval





## The first structure

# The demand and supply profiles

## The demand and supply profiles

### The demand profile

For each  $(o, d)$  of the transport network, the demand rate function from the origin  $o$  to the destination  $d$ , denoted as  $\mathcal{D}_{(o,d)}(\cdot)$ , is exogenous information

### The supply profile

The supply rate function is represented through the frequency function of each service at each stop, and is exogenous information

## The second structure

# The time and cost functions

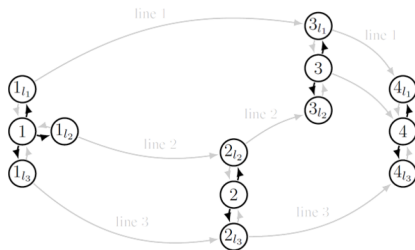
## The time and cost functions

### Overview

- Represents the interactions within each arc  $a = (i, j)$
- Relates the inflow and outflow rates going to each destination
- Locate the flows temporarily
- Respects FIFO rule

## The time and cost functions

$$\forall d \in D, \forall a \in A, \forall t \in [0, T]$$

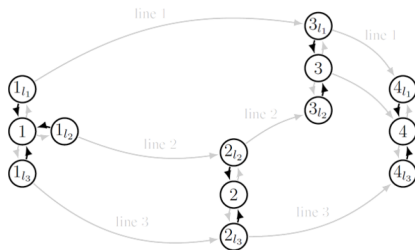


$$T_a(t) = b_a + \frac{1}{\phi_a(b_a+t)}$$

$$C_{ad}(t) = \alpha b_a + \frac{\beta}{\phi_a(b_a+t)}$$

## The time and cost functions

$$\forall d \in D, \forall a \in A, \forall t \in [0, T]$$

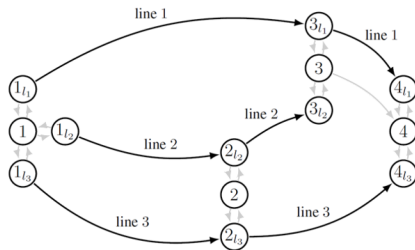


$$T_a(t) = l_a$$

$$C_{ad}(t) = \begin{cases} \delta l_a, & \text{if } a \in A_a \text{ and } j = d \\ \delta l_a + p_a, & \text{if } a \in A_a \text{ and } j \neq d \end{cases}$$

## The time and cost functions

$$\forall d \in D, \forall a \in A, \forall t \in [0, T]$$

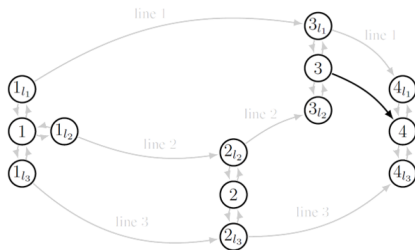


$$T_a(t) = V_a$$

$$C_{ad}^*(t) = \gamma V_a + H_a \sum_{d' \in D} E_a^{d'}(t)$$

## The time and cost functions

$$\forall d \in D, \forall a \in A, \forall t \in [0, T]$$



$$T_a(t) = w_a$$

$$C_{ad}(t) = \rho w_a$$



## The third structure

# The arc-choice model

## The arc-choice model

### Overview

- Extends to a dynamic assignment the one associated with the MTE considering a logit model
- First, computes recursively the expected minimum costs to each destination
- Then, aggregates flow rates arriving to each node
- Finally, assigns the inflow rates to the outgoing arcs of each node

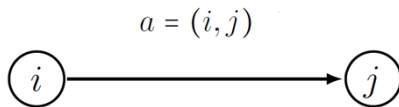
## The arc-choice model

### Logit model specifications

The *dispersion parameter*  $\theta$  is considered known

## The arc-choice model: Expected minimum costs

$$\forall d \in D, \forall a = (i, j) \in A, \forall t \in [0, T]$$

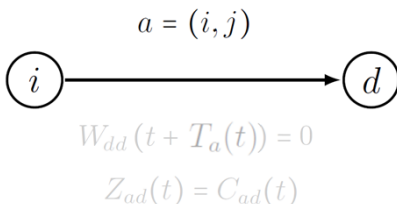


$$W_{jd}(t + T_a(t)) = 0$$

$$Z_{ad}(t) = C_{ad}(t)$$

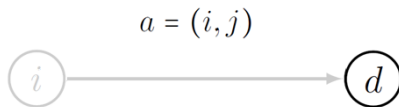
## The arc-choice model: Expected minimum costs

If  $j = d$



## The arc-choice model: Expected minimum costs

Expected minimum cost from  $j$  to  $d$  at  $t + C_a(t)$ ,  $W_{jd}(t + C_a(t))$

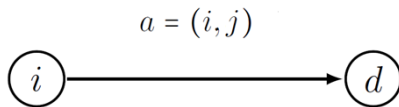


$$W_{dd}(t + T_a(t)) = 0$$

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## The arc-choice model: Expected minimum costs

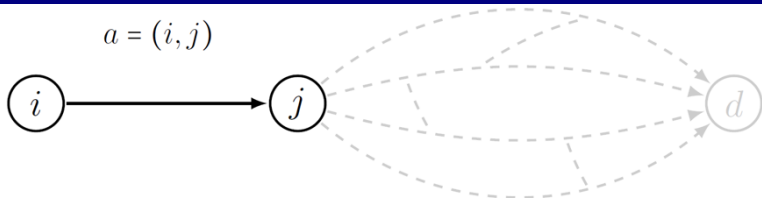
Expected minimum cost of using arc  $a$  to go to  $d$  at  $t$ ,  $Z_{ad}(t)$



$$W_{dd}(t + T_a(t)) = 0$$

$$Z_{ad}(t) = C_{ad}(t)$$

## The arc-choice model: Expected minimum costs

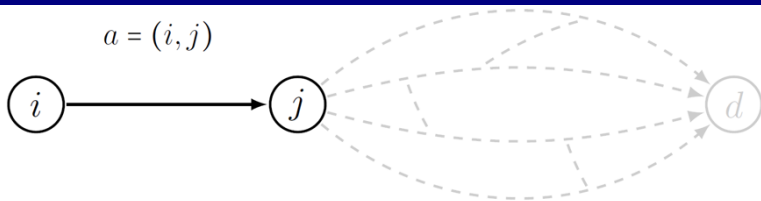
If  $j \neq d$ 

$$Z_{ad}(t) = C_{ad}(t) - \frac{1}{\theta} \ln \left( \sum_{b \in A_j^+} \exp(-\theta Z_{bd}(t + T_b(t))) \right)$$



## The arc-choice model: Expected minimum costs

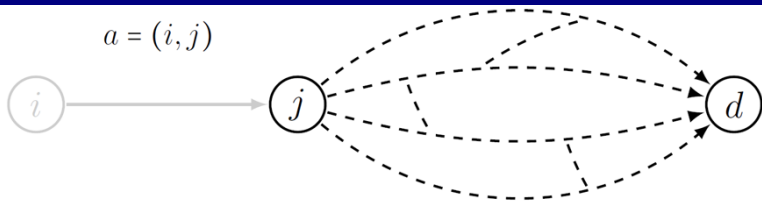
$Z_{ad}(t)$  components: Cost of the arc  $a$  at  $t$



$$Z_{ad}(t) = C_{ad}(t) - \frac{1}{\theta} \ln \left( \sum_{b \in A_j^+} \exp(-\theta Z_{bd}(t + C_b(t))) \right)$$

## The arc-choice model: Expected minimum costs

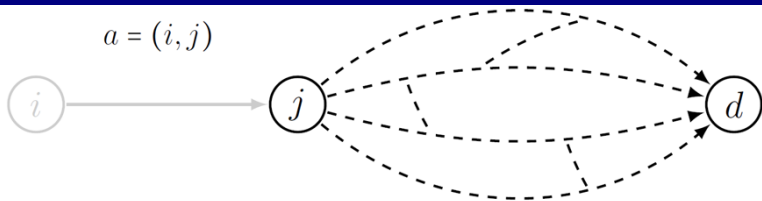
$Z_{ad}(t)$  components: Expected minimum cost from  $j$  to  $d$



$$Z_{ad}(t) = C_{ad}(t) - \frac{1}{\theta} \ln \left( \sum_{b \in A_j^+} \exp(-\theta Z_{bd}(t + T_b(t))) \right)$$

## The arc-choice model: Expected minimum costs

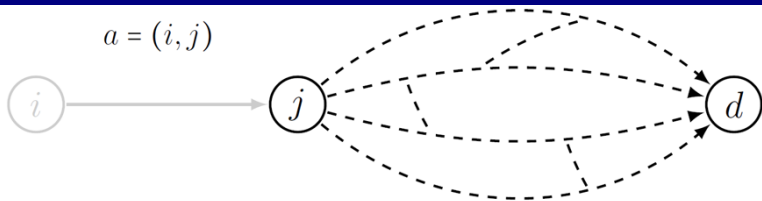
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## The arc-choice model: Expected minimum costs

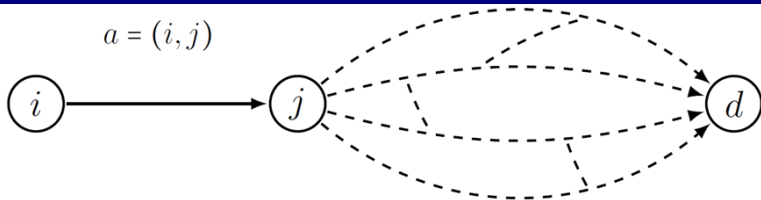
$Z_{ad}(t)$  components: Expected minimum cost from  $j$  to  $d$



$$Z_{ad}(t) = C_{ad}(t) + W_{jd}(t + T_a(t))$$

## The arc-choice model: Expected minimum costs

Expected minimum cost of using arc  $a$  to go to  $d$  at  $t$



$$Z_{ad}(t) = C_{ad}(t) + W_{jd}(t + T_a(t))$$

## The arc-choice model: Expected minimum costs

$$\forall d \in D, \forall a = (i, j) \in A, \forall t \in [0, T]$$

If  $j = d$ , then

$$W_{jd}(t + T_a(t)) = 0 \quad (1)$$

$$Z_{ad}(t) = C_a(t) \quad (2)$$

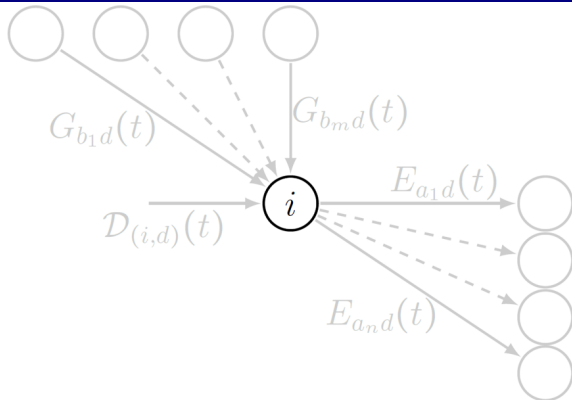
otherwise, if  $j \neq d$

$$W_{jd}(t + T_a(t)) = -\frac{1}{\theta} \ln \left( \sum_{b \in A_j^+} \exp(-\theta (Z_{bd}(t + C_a(t)))) \right) \quad (3)$$

$$Z_{ad}(t) = C_{ad}(t) + W_{jd}(t + C_a(t)) \quad (4)$$

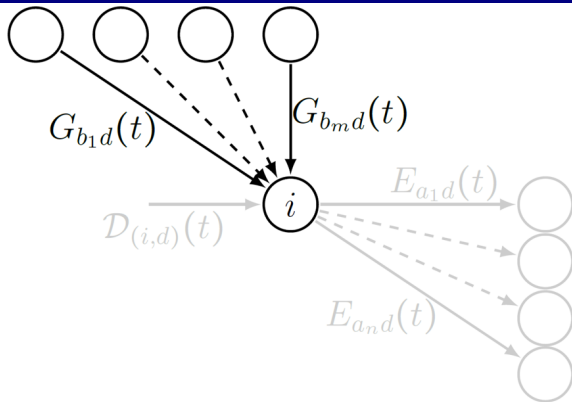
## The arc-choice model: Assignment

Given a destination node  $d \in D, \forall i \in N$



## The arc-choice model: Assignment

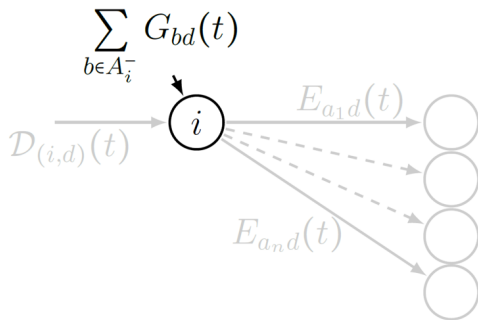
$$\forall d \in D, \forall t \in [0, T]$$





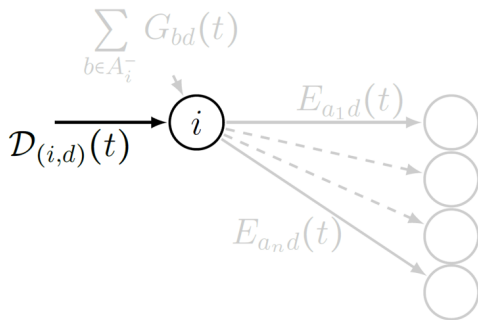
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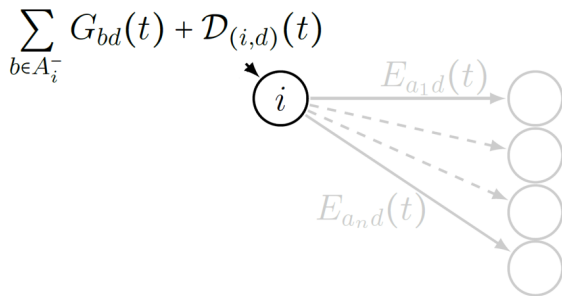
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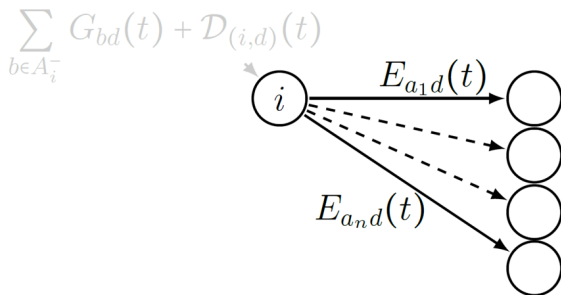
## The arc-choice model: Assignment

$$\forall d \in D, \forall t \in [0, T]$$



## The arc-choice model: Assignment

$$\forall d \in D, \forall a \in A_i^+, \forall t \in [0, T]$$



## The arc-choice model: Assignment

$$\forall d \in D, \forall i \in N, \forall a \in A_i^+, \forall t \in [0, T]$$

$$E_{ad}(t) = \frac{\exp(-\theta Z_{ad}(t))}{\sum_{b \in A_i^+ \cap R_d} \exp(-\theta Z_{bd}(t))} \left( \sum_{b \in A_i^-} G_{bd}(t) + \mathcal{D}_{(i,d)}(t) \right) \quad (5)$$

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The MDTrA model

**The MDTrA algorithm**

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Summary

# The MDTrA algorithm

## Overview

- Works over a discretization of the analyzed time period
- The processes at each time increment are based on a reversed *Dial's Algorithm*

## Overview

### The inputs

- $(N, A)$ , digraph associated to the transport network
- $O \subseteq N$ , set of origins
- $D \subseteq N$ , set of destinations
- $OD \subseteq O \times D$ , set of  $O - D$  pairs
- $\mathcal{D}(\cdot)$ , time-dependent demand rate
- $T$ , length of the period of time to analyze
- $\theta$ , dispersion parameter



## Overview

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- $\mathcal{D}(\cdot)$ , time-dependent demand rate
- $T$ , length of the period of time to analyze
- $\theta$ , dispersion parameter
- $\Delta t$ , timestep size

## Overview

### The outputs

The discretization will result in  $K$  time intervals. Then,  $\forall d \in D$ ,  
 $\forall a \in A$ ,  $\forall k = 1, \dots, K$

- $E_{ad}^k$ , inflow rate of arc  $a$  going to  $d$  at  $k$
- $G_{ad}^k$ , outflow rate of arc  $a$  going to  $d$  at  $k$
- $C_{ad}^k$ , cost of arc  $a$  going to  $d$  at  $k$
- $Z_{ad}^k$ , expected minimum cost from arc  $a$  to  $d$  at  $k$
- $W_{id}^k$ , expected minimum cost from node  $i$  to  $d$  at  $k$

## How it works

### Technical settings

- STEP 0: Initialization

### From $k = 1$ to $k = K$ or until meeting stop criteria

- STEP 1: **Backward**: Computes expected minimum costs
- STEP 2: Computes assignment factors (logit model coefficients)
- STEP 3: **Forward**: Assigns inflow rates to each destination
- STEP 4: Updates costs of arcs
- STEP 5: Checks if stop criteria are meet

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Overview

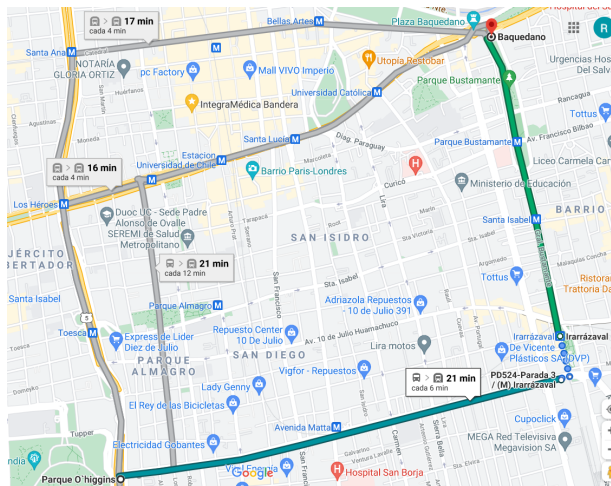
The MDTrA model

The MDTrA algorithm

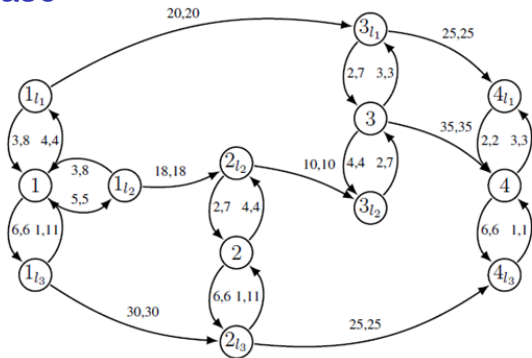
**Computational Implementation**

Summary

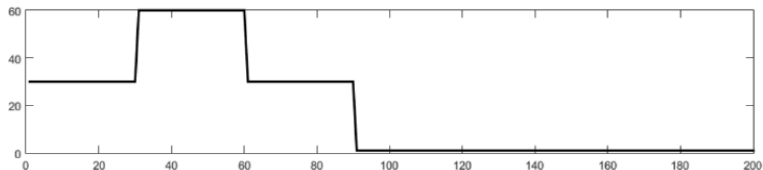
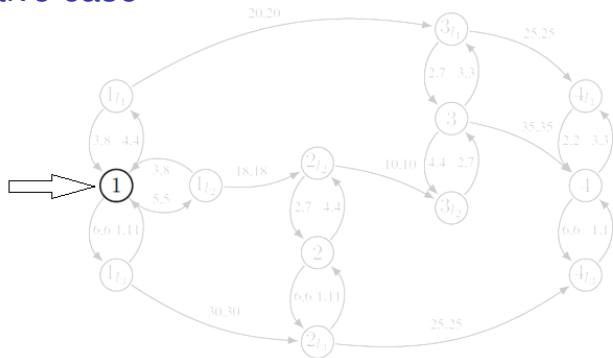
## Illustrative case



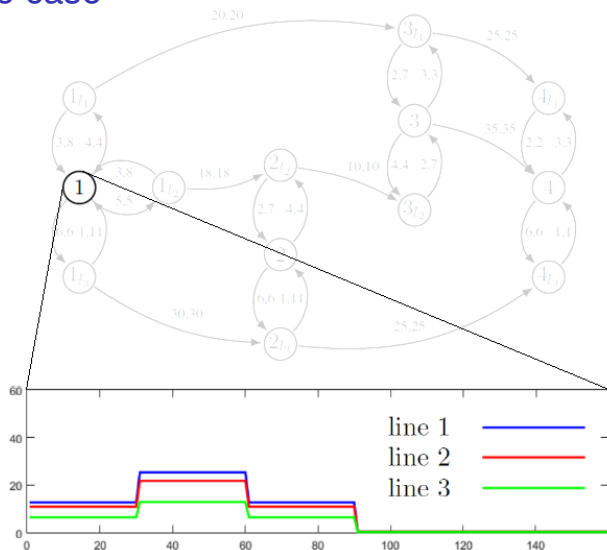
## Illustrative case



## Illustrative case

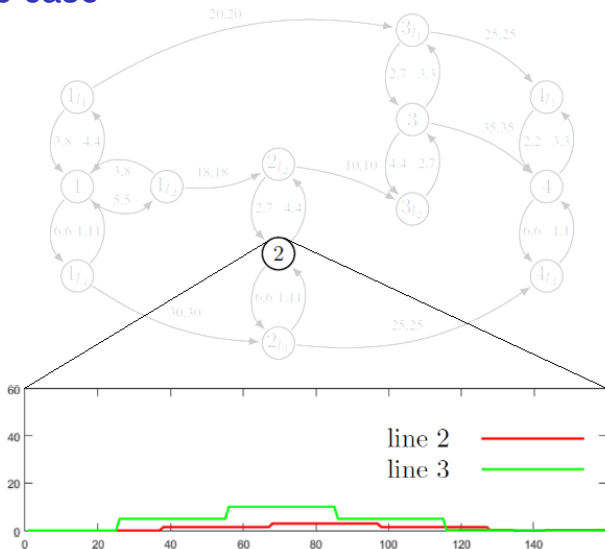


# Illustrative case

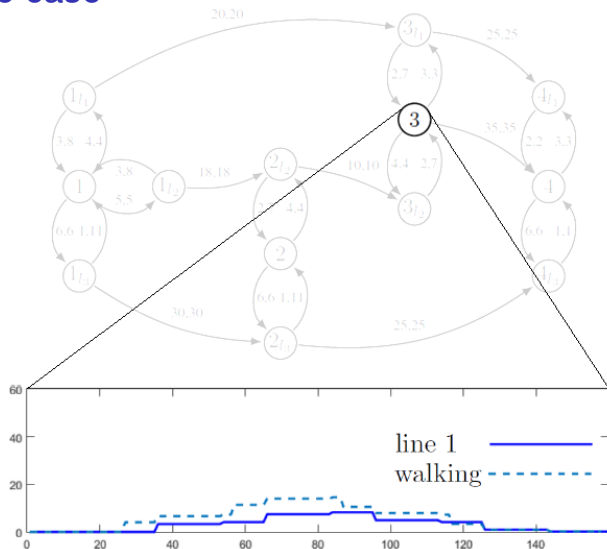




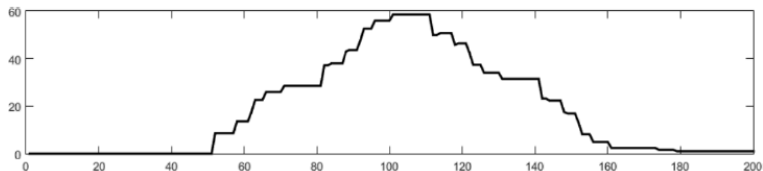
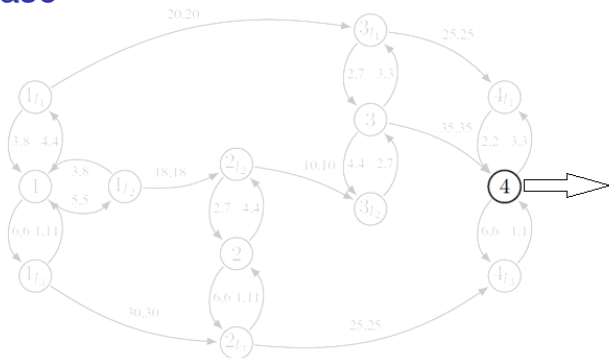
## Illustrative case



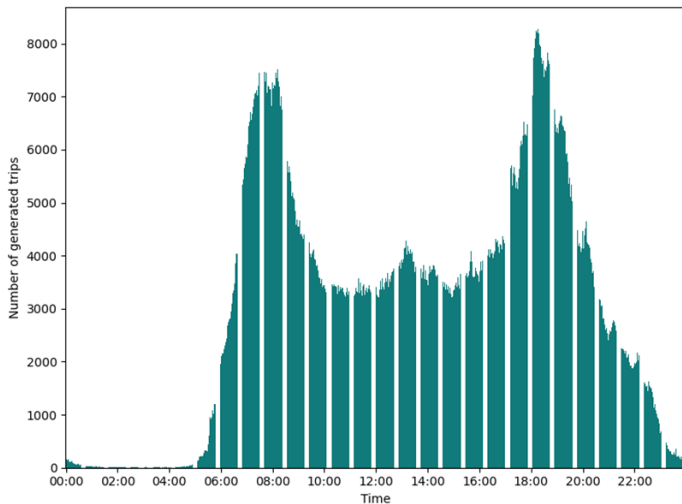
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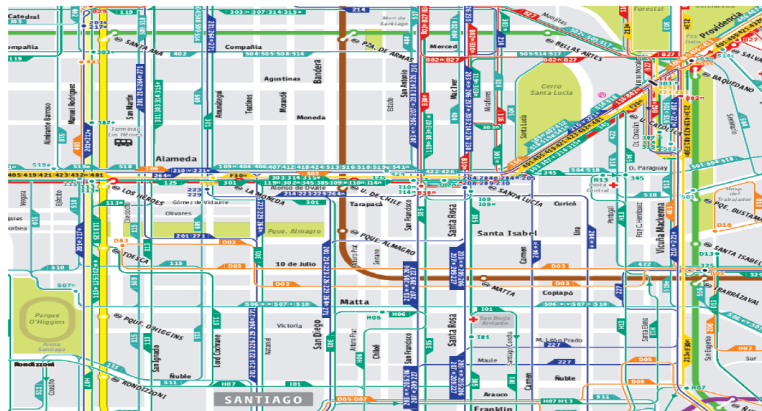
## Illustrative case



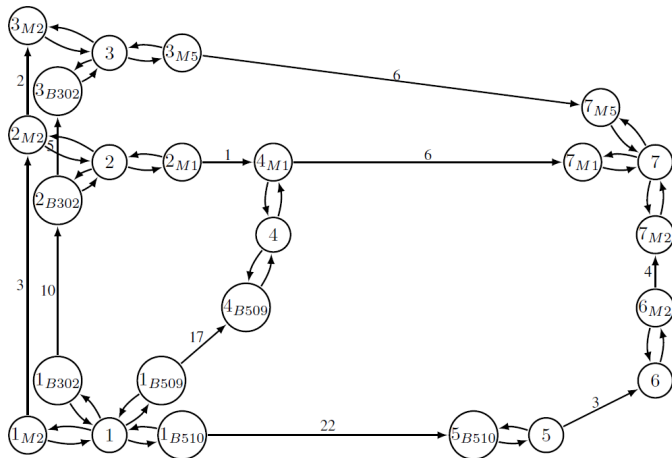
## Real data: Demand estimation with smartcard data



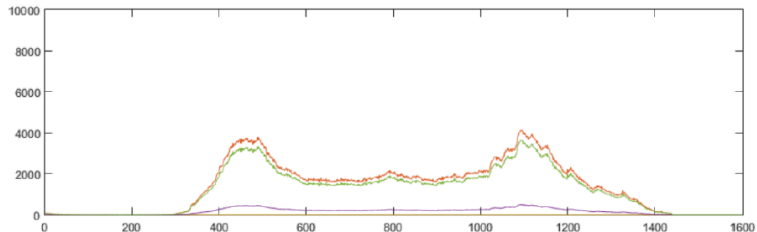
## Real data: Subnetwork A of Santiago



## Real data: Reasonable arcs for subnetwork A

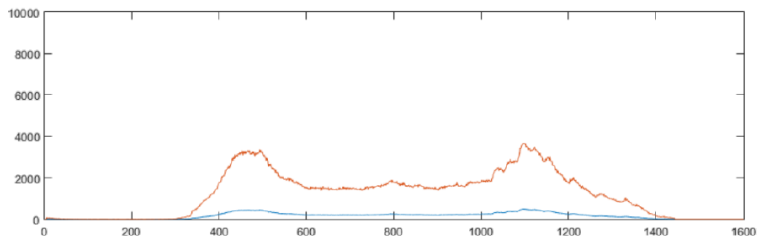


## Real data: Inflows from Parque Ohiggins



B506 (green), L2 (red) and B505 (purple) from Parque O'higgins

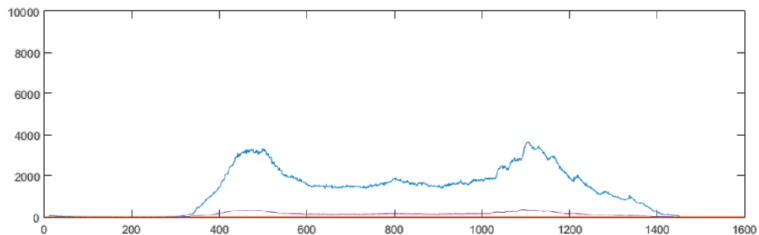
## Real data: Inflows from Los Héroes



L1 (red) and L2 (blue) from Los Héroes

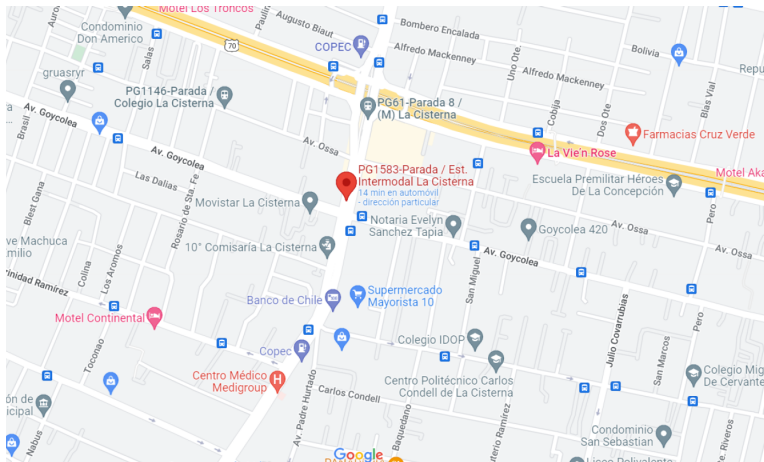


## Real data: Inflows from Irarrázabal



L5 (blue) and walking (purple) from Irarrázabal

# Real data: Subnetwork B of Santiago



## Real data: Subnetwork B of Santiago

### Characteristics

- Intermodal La Cisterna as the single destination
- Associated with 38 services
- Underlying digraph originally has 1035 nodes and 1648 arcs

## Real data: Subnetwork B of Santiago

### Characteristics

- Intermodal La Cisterna as the single destination
- Associated with 38 services
- Underlying digraph originally has 1035 nodes and 1648 arcs

### After applying reasonability concept

- 1084 arcs
- More than 34 % of reductions of arcs

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## Features

### The MDTrA model

- Adapts an MTE's arc-choice model to a dynamic and stochastic assignment context
- Allows working overlapping routes
- Does not require independence of route costs
- Respects FIFO rule

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### The MDTrA model

- Adapts an MTE's arc-choice model to a dynamic and stochastic assignment context
- Allows working overlapping routes
- Does not require independence of route costs
- Respects FIFO rule

### The MDTrA algorithm

- Does not require routes enumeration
- Allows initialization with non-empty networks
- Efficiently runs network loading though dynamic programming

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## Future works

The present work and its results open research opportunities on:

- Extend the reasonability concept
- Study initialization with non-empty networks
- Use the algorithms' outputs as initial solution and then apply improvement methods

# Thank you for your attention

Questions? feedback? Please, reach us

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## Bibliography

Addison and Heydecker (1996). *Transportation and Traffic Theory (article)*, 1996.

Addison and Heydecker (1998). *Transportation Networks (chapter)*, 1998.

Baillon and Cominetti (2005). *Mathematical Programming (article)*, 2008.

Han (2003). *Transportation Research Part B (article)*, 2003.

Heydecker and Addison (1997).

Heydecker and Addison (2005). *Transportation Science (article)*, 2005.

Koch and Skutella (2005). *Theory of Computing Systems (article)*, 2011.

Lim and Heydecker (2005). *Transportation Research Part B (chapter)*, 2005.

Merchant and Nemhauser (1978). *Transportation Science (article)*, 1988.

Sheffi (1985). *Urban Transportation Networks*, 1985.

Robert B. Dial (1971). *A probabilistic multipath traffic assignment which obviates enumeration*.