

# Initial Comparisons Between MDCEV model and OASIS Framework

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23<sup>rd</sup>, STRC, Ascona, 2023

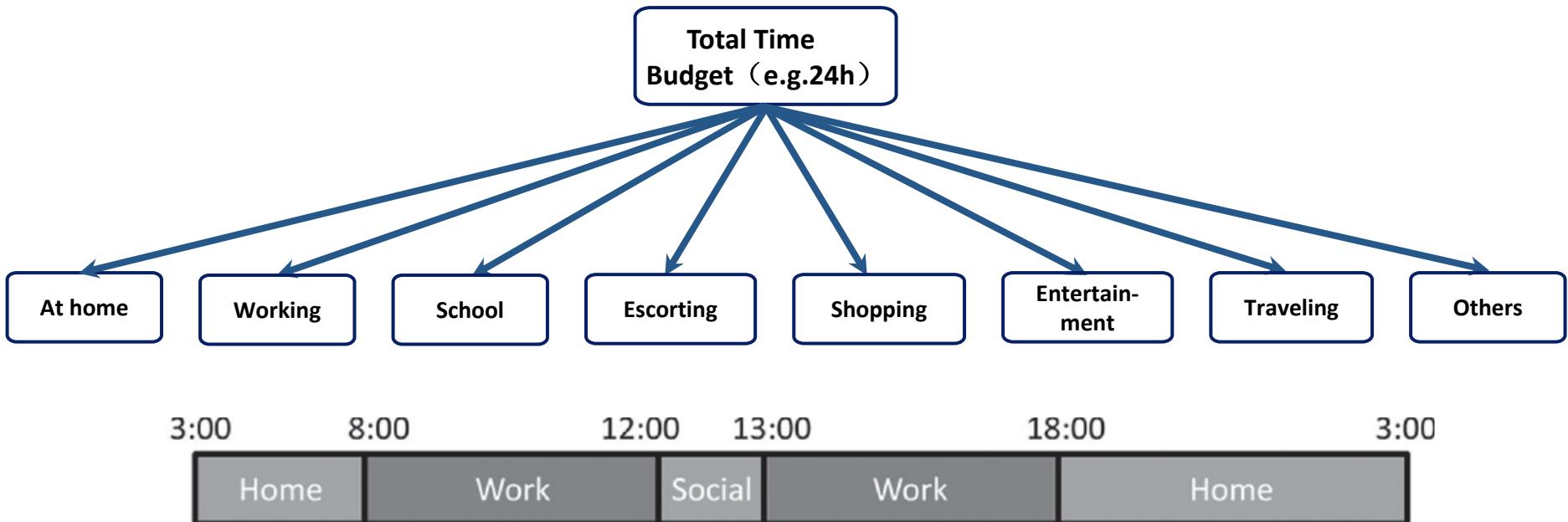


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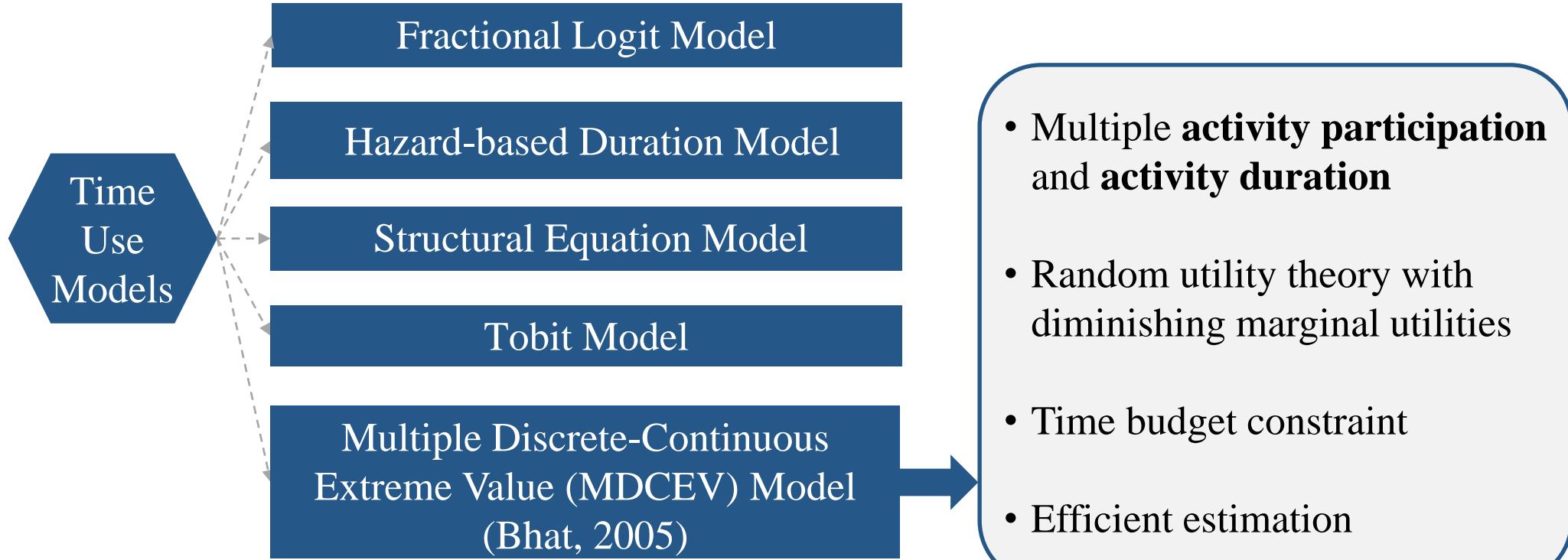
- Introduction
- Methodology Review
- Data Source and Sample Formation
- Model Estimation
- Model Simulation
- Limitations and Future Directions

## Activity-based Modeling (ABM)

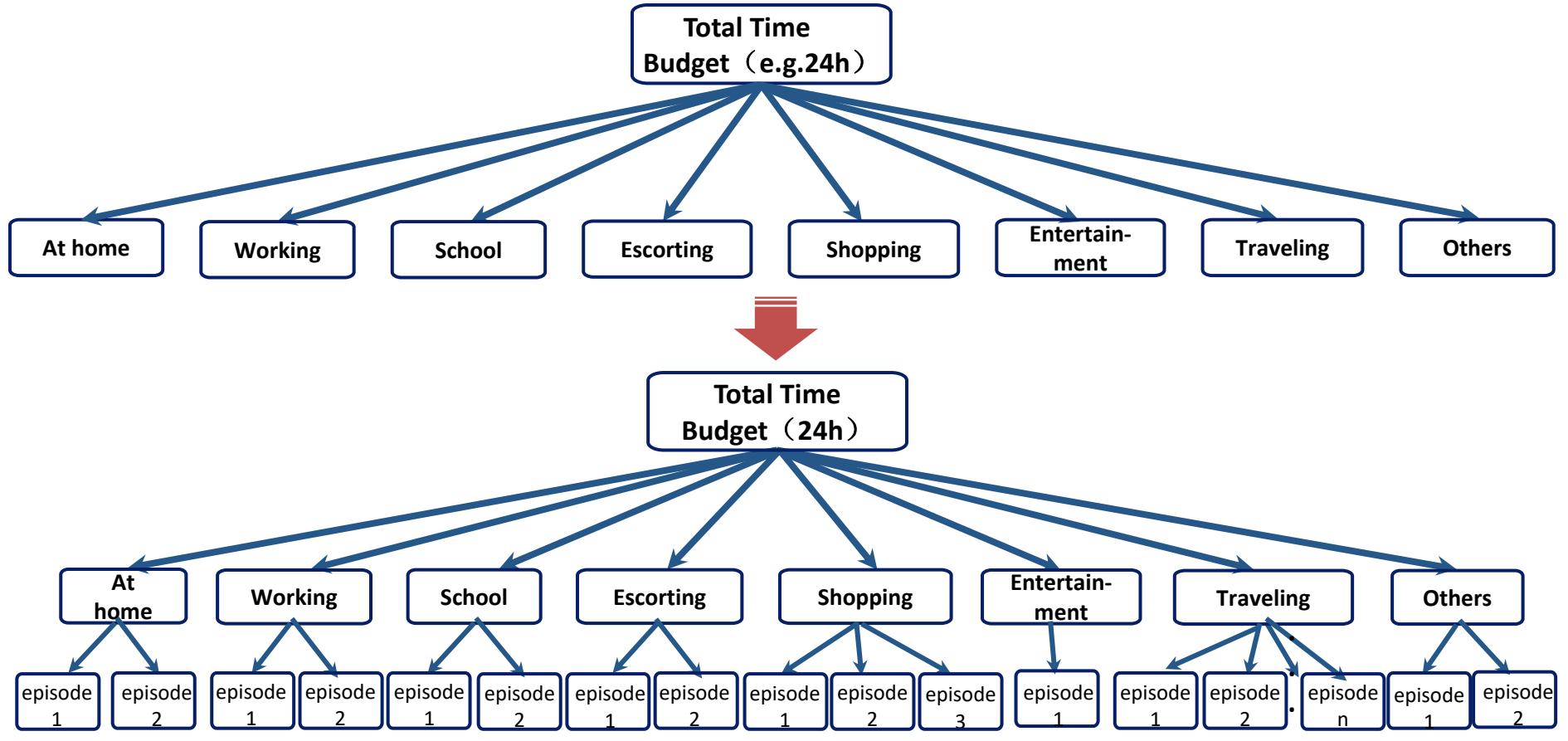
- Time is treated as an important limited resource.
- Time-use and scheduling process are critical and central components.



# Existing Methods in Time Use



# Limitation of MDCEV model

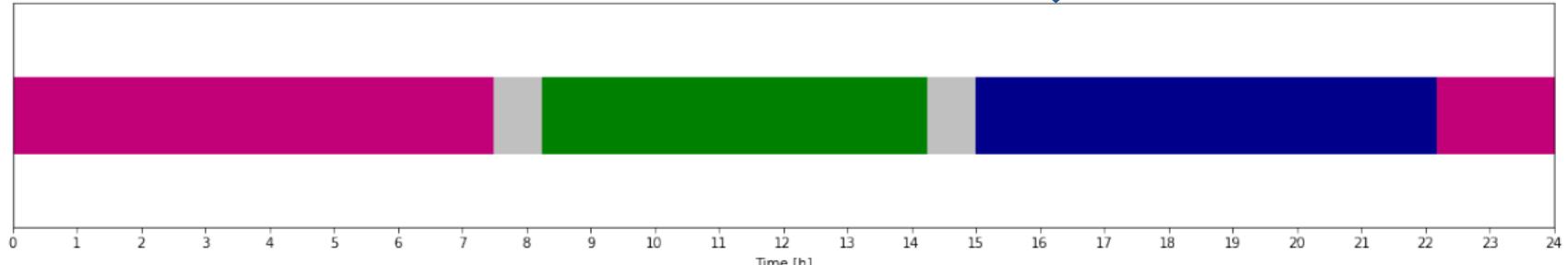


# OASIS Framework (Pougala et al., 2022)

Dimensions:

- Activity participation;
- Activity duration;
- Activity location;
- Time of day;
- Travel mode

| Person | Schedule  | Activity      | Start time (hh:mm) | Duration (hh:mm) | Location | Mode    | Participation mode |
|--------|---|---------------|--------------------|------------------|----------|---------|--------------------|
| Sara   | Isolated<br>indiv./<br>Fam. of<br>2/ Fam.<br>of 3 | Sleep_morn    | 00:00              | 6:30             | Home     | No trip | Solo               |
|        |   | Sleep_morn    | 00:00              | 6:30             | Home     | PT      | Solo               |
|        |   | Sleep_morn    | 00:00              | 6:30             | Home     | Driving | Solo               |
|        |   | Personal care | 6:30               | 2:00             | Home     | No trip | Solo               |
|        |   | Personal care | 6:30               | 2:00             | Home     | PT      | Solo               |
|        |   | Personal care | 6:30               | 2:00             | Home     | Driving | Solo               |
|        |   | Work          | 8:30               | 6:00             | Work     | PT      | Solo               |
|        |   | Work          | 8:30               | 6:00             | Work     | Driving | Solo               |
|        |   | Work          | 8:30               | 6:00             | Home     | No trip | Solo               |
|        |   | Work          | 8:30               | 6:00             | Home     | PT      | Solo               |
|        |   | Work          | 8:30               | 6:00             | Home     | Driving | Solo               |
|        |   | Home care     | 14:30              | 7:40             | Home     | No trip | Solo               |
|        |   | Home care     | 14:30              | 7:40             | Home     | PT      | Solo               |
|        |   | Home care     | 14:30              | 7:40             | Home     | Driving | Solo               |
|        |   | Sleep_night   | 22:10              | 1:50             | Home     | No trip | Solo               |



# Motivation

- It is valuable to analyze strengths, weaknesses and application scenarios of **MDCEV** and **OASIS**.
- Currently, there is no **comparative study** of them in the literature.
- An initial attempt to compare the **forecasting performance** in two **common dimensions** of the models: **activity participation** and **activity duration**.

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# Comparison List

| Features                |  | MDCEV  | OASIS  |
|-------------------------|--|--|--|
| <b>Model Structure</b>  | <b>Principle</b>   | Constrained optimization problem with random utility theory  |  |
|                         | Input  | Individual and household attributes; Model parameters; Error terms   | A set of considered activities with locations and transport modes; Scheduling preferences; Activity flexibility; Model parameters; Error terms     |
|                         | Output   | Time allocation of different activities  | Activity schedule for whole day  |
|                         | Modeling Dimensions  | Activity participation; Activity duration, Activity frequency; Chronological order of episodes from the same activity category | Activity participation; Activity start time; Activity duration; Activity frequency; Activity sequence; Activity location; Travel mode; Travel time |
|                         | Utility Function   | Non-linear   | Linear   |
|                         | Explanatory Variables  | Individual and household attributes  | Activity-travel attributes   |
|                         | Choice Set   | All activity categories  | All valid schedules  |
| <b>Model Estimation</b> | Maximum likelihood estimation  |  | Maximum likelihood estimation relied on sampled choice set   |
| <b>Model Simulation</b> | Specific algorithm by certain rules  |  | Standard mathematical programming algorithm  |
| <b>Model Extension</b>  | In-home activities; Household level; Different distributions of error terms; Random coefficients                           |  |  |
| <b>Research Focus</b>   | Exploring time allocation mechanism by sophisticated utility functions; Describing influence factors by estimation results |  | Exploring daily scheduling mechanism by integrating all dimensions; Capturing trade-offs between different choices by model simulation             |

# Model Structure—Principle

## MDCEV

$$\max U(\mathbf{t}) = \sum_{k=1}^K \frac{\gamma_k}{\alpha_k} \psi_k \left\{ \left( \frac{t_k}{\gamma_k} + 1 \right)^{\alpha_k} - 1 \right\}$$

$$\text{s.t. } \sum_{k=1}^K t_k = T$$

Where:  $\psi_k$ —baseline marginal utility;

$$(\psi_k > 0, \psi(\mathbf{z}_k, \varepsilon_k) = \exp(\boldsymbol{\beta}' \mathbf{z}_k + \varepsilon_k))$$

$\gamma_k$  and  $\alpha_k$ —satiation parameters.  
 $(\gamma_k > 0, 0 < \alpha_k < 1)$

## OASIS

$$\begin{aligned} & \max_{\omega, z, x, \tau} U(\omega, z, x, \tau, \epsilon) + \sum_{a=0}^A \omega_a (V_a^1 + V_a^2 + V_a^3) + \sum_{a=0}^A \sum_{b=0}^A z_{ab} (V_{ab}^4 + V_{ab}^5), \\ & \text{s.t.} \quad \left[ \begin{array}{l} \sum_a \sum_b (\omega_a \tau_a + z_{ab} \rho_{ab}) = T, \\ \sum_a \sum_b (\omega_a c_a + z_{ab} \kappa_{ab}) \leq B, \\ \omega_{\text{dawn}} = \omega_{\text{dusk}} = 1, \\ \tau_a \geq \omega_a \tau_a^{\min}, \quad \forall a \in A, \\ \tau_a \leq \omega_a T, \quad \forall a \in A, \\ z_{ab} + z_{ba} \leq 1, \quad \forall a, b \in A, a \neq b, \\ z_{a,\text{dawn}} = z_{\text{dusk},a} = 0, \quad \forall a \in A, \\ \sum_a z_{ab} = \omega_b, \quad \forall b \in A, b \neq \text{dawn}, \\ \sum_b z_{ab} = \omega_a, \quad \forall a \in A, a \neq \text{dusk}, \\ (z_{ab} - 1)T \leq x_a + \tau_a + z_{ab} \rho_{ab} - x_b, \quad \forall a, b \in A, a \neq b, \\ (1 - z_{ab})T \geq x_a + \tau_a + z_{ab} \rho_{ab} - x_b, \quad \forall a, b \in A, a \neq b, \\ \sum_{a \in G_k} \omega_a \leq 1 \quad k = 1, \dots, K, \\ \alpha_a^m = 1 \quad \forall a \in G_{\text{home}}, \\ \omega_a \leq \alpha_a^m \quad \forall a \in A^m, \\ \omega_a \geq \omega_b + z_{ab} - 1 \quad \forall a \in A, b \in A \setminus G_{\text{home}} \end{array} \right] \end{aligned}$$

Where, the decision variables are:

$\omega_a$ —Whether activity a is selected;

$z_{ab}$ —Whether activity b is after a;

$x_a$ —Start time of activity a;

$\tau_a$ —Duration of activity a;

$\alpha_a^m$ —Whether private mode m is

available for activity a.

# Model Estimation

## MDCEV

$$\max U(\mathbf{t}) = \sum_{k=1}^K \frac{\gamma_k}{\alpha_k} \psi_k \left\{ \left( \frac{t_k}{\gamma_k} + 1 \right)^{\alpha_k} - 1 \right\}$$

$$\text{s.t. } \sum_{k=1}^K t_k = T$$

$$\mathcal{L} = \sum_k \frac{\gamma_k}{\alpha_k} [\exp(\beta' z_k + \varepsilon_k)] \left\{ \left( \frac{x_k}{\gamma_k} + 1 \right)^{\alpha_k} - 1 \right\} - \lambda \left[ \sum_{k=1}^K x_k - B \right]$$



$$P(x_1^*, x_2^*, x_3^*, \dots, x_M^*, 0, 0, \dots, 0) = [\prod_{i=1}^M c_i] \left[ \sum_{i=1}^M \frac{1}{c_i} \right] \left[ \frac{\prod_{i=1}^M e^{V_i}}{\left( \sum_{j=1}^K e^{V_j} \right)^M} \right] (M-1)!$$

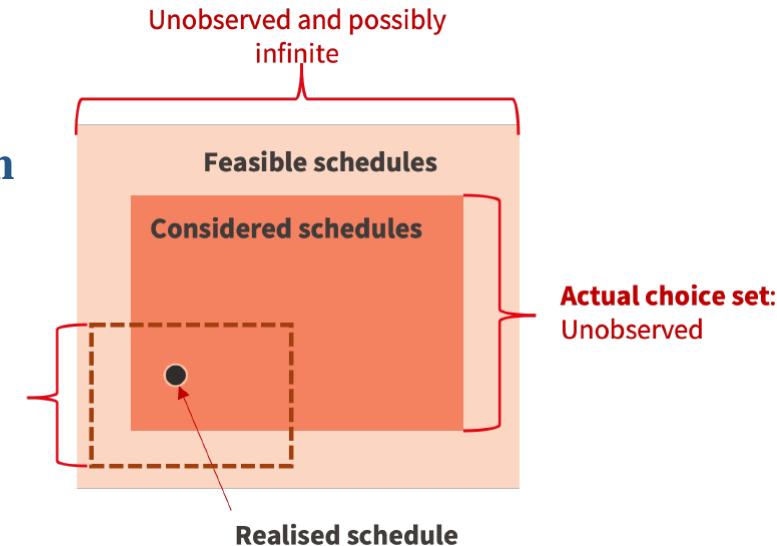
Maximum likelihood estimation

## M-H algorithm



**Estimation choice set:**  
sample of feasible  
schedules generated for  
estimation purposes

## OASIS



$$P(i_n | \mathcal{C}_n) = \frac{\exp [V_{in} + \ln q(\mathcal{C}_n | i_n)]}{\sum_{j \in \mathcal{C}_n} \exp [V_{jn} + \ln q(\mathcal{C}_n | j)]}$$

Maximum likelihood estimation  
based on sampled choice set

# Model Simulation

MDCEV

If 'i' is a chosen alternative and  
'j' is not a chosen alternative.

KT  
conditions

$$\left\{ \begin{array}{l} \psi_i \left( \frac{t_i^*}{\gamma_i} + 1 \right)^{\alpha_i - 1} = \lambda, \quad t_i^* > 0 \\ \psi_j \left( \frac{t_j^*}{\gamma_j} + 1 \right)^{\alpha_j - 1} < \lambda, \quad t_j^* = 0 \end{array} \right.$$



$$\psi_j < \lambda < \psi_i$$

Specific algorithm by certain rules

OASIS

$$\begin{aligned} & \max_{\omega, z, x, \tau} U(\omega, z, x, \tau, \epsilon) + \sum_{a=0}^A \omega_a (V_a^1 + V_a^2 + V_a^3) + \sum_{a=0}^A \sum_{b=0}^A z_{ab} (V_{ab}^4 + V_{ab}^5), \\ & \sum_a \sum_b (\omega_a \tau_a + z_{ab} \rho_{ab}) = T, \\ & \sum_a \sum_b (\omega_a c_a + z_{ab} \kappa_{ab}) \leq B, \\ & \omega_{\text{dawn}} = \omega_{\text{dusk}} = 1, \\ & \tau_a \geq \omega_a \tau_a^{\min}, \quad \forall a \in A, \\ & \tau_a \leq \omega_a T, \quad \forall a \in A, \\ & z_{ab} + z_{ba} \leq 1, \quad \forall a, b \in A, a \neq b, \\ & z_{a,\text{dawn}} = z_{\text{dusk},a} = 0, \quad \forall a \in A, \\ & \sum_a z_{ab} = \omega_b, \quad \forall b \in A, b \neq \text{dawn}, \\ & \sum_b z_{ab} = \omega_a, \quad \forall a \in A, a \neq \text{dusk}, \\ & (z_{ab} - 1)T \leq x_a + \tau_a + z_{ab} \rho_{ab} - x_b, \quad \forall a, b \in A, a \neq b, \\ & (1 - z_{ab})T \geq x_a + \tau_a + z_{ab} \rho_{ab} - x_b, \quad \forall a, b \in A, a \neq b, \\ & \sum_{a \in G_k} \omega_a \leq 1 \quad k = 1, \dots, K, \\ & \alpha_a^m = 1 \quad \forall a \in G_{\text{home}} \\ & \omega_a \leq \alpha_a^m \quad \forall a \in A^m \\ & \omega_a \geq \omega_b + z_{ab} - 1 \quad \forall a \in A, b \in A \setminus G_{\text{home}} \end{aligned}$$

Where:  $\omega_a, z_{ab}, x_a, \tau_a, \alpha_a^m$   
are decision variables.

Standard mathematical  
programming algorithm

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## Date Source and Sample Formation

■ **Data Source:** 2015 Mobility and Transport Microcensus (MTMC)

■ **Sample Formation :**

- Select full samples of Lausanne;
- Remove those who didn't finish their daily schedule at home;
- Delete samples who have negative activity durations or total duration is more than 24 hours;
- Remove those who are at home all the day;
- Final sample size: 1016 (700 for estimation; 316 for simulation)

■ **Activity Type:**

home, work, education, shopping, leisure, escort, errands and services, business trip

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# MDCEV Model

## ■ Different profiles

- $\gamma$ -profile: 
$$\max U(\mathbf{t}) = \sum_{k=1}^K \gamma_k \psi_k \ln\left(\frac{t_k}{\gamma_k} + 1\right)$$
- $\alpha$ -profile: 
$$\max U(\mathbf{t}) = \sum_{k=1}^K \frac{\psi_k}{\alpha_k} \{(t_k + 1)^{\alpha_k} - 1\}$$
- Parabola: 
$$\max U(\mathbf{t}) = \sum_{k=1}^K -0.5\psi_k(t_k - m_k)^2$$



**Traditional MDCEV  
(Bhat, 2008)**



**Parabolic MDCEV  
(Wang, M. and X. Ye., 2023)**

## ■ Parameters to be estimated (activity-specific)

|            | $\gamma$ -profile  | $\alpha$ -profile  | Parabola  |
|------------|--|--|---|
| parameters | Baseline marginal utility: $\psi_k$<br>Satiation parameter: $\gamma_k$ | Baseline marginal utility: $\psi_k$<br>Satiation parameter: $\alpha_k$ | Opening size : $\psi_k$<br>Extreme point: $m_k$ |

# MDCEV Model

## Traditional MDCEV models

| Parameters                           | $\gamma$ -Profile |             | $\alpha$ -Profile |             |
|--------------------------------------|-------------------|-------------|-------------------|-------------|
|                                      | Value             | t-Statistic | Value             | t-Statistic |
| <i><math>\psi_k</math> component</i> |                   |             |                   |             |
| <i>Constants</i>                     |                   |             |                   |             |
| Working                              | 0.6141            | 6.043       | 0.5892            | 5.903       |
| Education                            | - 0.7964          | - 6.087     | - 0.7504          | - 5.770     |
| Shopping                             | 0.5426            | 5.500       | 0.4390            | 4.601       |
| Leisure                              | 0.9144            | 9.558       | 1.0627            | 11.186      |
| Others                               | —                 | —           | —                 | —           |
| <i>Satiation parameters</i>          |                   |             |                   |             |
| Working                              | 1.0000            | (fixed)     | 1.0000            | (fixed)     |
| Education                            | 1.0000            | (fixed)     | 1.0000            | (fixed)     |
| Shopping                             | 0.6557            | 17.771      | 1.0000            | (fixed)     |
| Leisure                              | 1.8053            | 14.018      | 0.4257            | 1.971       |
| Others                               | 0.8302            | 12.758      | 1.0000            | (fixed)     |
| <b>Summary statistics</b>            |                   |             |                   |             |
| Number of cases                      | 700               |             | 700               |             |
| Final Log-likelihood                 | - 2775.5          |             | - 2703.6          |             |

## Parabolic MDCEV model

| Parameters                           | Value      | t-Statistic |
|--------------------------------------|------------|-------------|
| <i><math>m_k</math> component</i>    |            |             |
| <i>Constants</i>                     |            |             |
| Working                              | - 1.9531   | - 14.027    |
| Education                            | - 1.1320   | - 21.744    |
| Shopping                             | - 0.7094   | - 33.436    |
| Leisure                              | 0.0000     | (fixed)     |
| Others                               | - 1.1403   | - 15.374    |
| <i><math>\psi_k</math> component</i> |            |             |
| <i>Constants</i>                     |            |             |
| Working                              | - 1.0348   | - 48.763    |
| Education                            | 1.6480     | 8.197       |
| Shopping                             | 2.2063     | 12.400      |
| Leisure                              | —          | —           |
| Others                               | 2.2156     | 5.337       |
| <b>Summary statistics</b>            |            |             |
| Number of cases                      | 700        |             |
| Final Log-likelihood                 | - 4571.119 |             |

# OASIS Model

## ■ Parameters to be estimated (activity-specific)

| Parameter                  | Notation        | Activities                     |
|----------------------------|-----------------|--------------------------------|
| Activity-specific constant | $ASC_{a,n}$     |                                |
| Penalty start time (early) | $\theta_F^e$    |                                |
| Penalty start time (late)  | $\theta_F^\ell$ | All ( <i>reference: home</i> ) |
| Penalty duration (short)   | $\beta_F^s$     |                                |
| Penalty duration (long)    | $\beta_F^\ell$  |                                |

- Choice set size: 25 per observation
- Travel parameters are not included
- Desired start times and durations:  
drawn from pre-determined distributions

$$U_S = U + \sum_{a=0}^{A-1} \left( U_a^1 + U_a^2 + U_a^3 + \sum_{b=0}^{A-1} (U_{a,b}^4 + U_{a,b}^5) \right)$$

$$U_a^1 = \beta_{\text{cost}} * c_a + \varepsilon_1$$

$$U_a^2 = \theta_a^e \max(0, x_a^- - x_a) + \theta_a^\ell \max(0, x_a - x_a^+)$$

$$U_a^3 = \beta_{a,a}^e \max(0, \tau_a^- - \tau_a) + \beta_a^\ell \max(0, \tau_a - \tau_a^+)$$

$$U_{a,b}^4 = \beta_{t,\text{cost}} * c_t + \varepsilon_4$$

$$U_{a,b}^5 = \theta_t \rho_{ab}$$

# OASIS Model

| Parameters       | Value   | t-Statistic |
|------------------|---------|-------------|
| <i>Constants</i> |         |             |
| Working          | 12.800  | 8.54        |
| Education        | 9.160   | 8.69        |
| Shopping         | 10.100  | 13.20       |
| Leisure          | 10.400  | 11.00       |
| <i>Early</i>     |         |             |
| Working          | - 1.380 | - 4.95      |
| Education        | - 1.490 | - 7.85      |
| Shopping         | - 1.060 | - 12.00     |
| Leisure          | - 0.254 | - 1.46      |
| <i>Late</i>      |         |             |
| Working          | - 0.532 | - 3.01      |
| Education        | - 0.575 | - 3.42      |
| Shopping         | - 0.685 | - 4.46      |
| Leisure          | - 1.030 | - 8.75      |
| <i>Long</i>      |         |             |
| Working          | - 0.746 | - 3.95      |
| Education        | - 5.290 | - 4.37      |
| Shopping         | - 1.150 | - 4.56      |
| Leisure          | - 0.434 | - 5.13      |
| <i>Short</i>     |         |             |
| Working          | - 1.240 | - 4.85      |
| Education        | - 0.336 | - 2.10      |
| Shopping         | - 7.910 | - 2.67      |
| Leisure          | - 0.566 | - 2.32      |

## Summary statistics

Number of observations = 700

$$\underline{L}(0) = -1875.052$$

$$\underline{L}(\beta) = -218.583$$

$$\bar{\rho}^2 = 0.873$$

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## Statistical Analysis

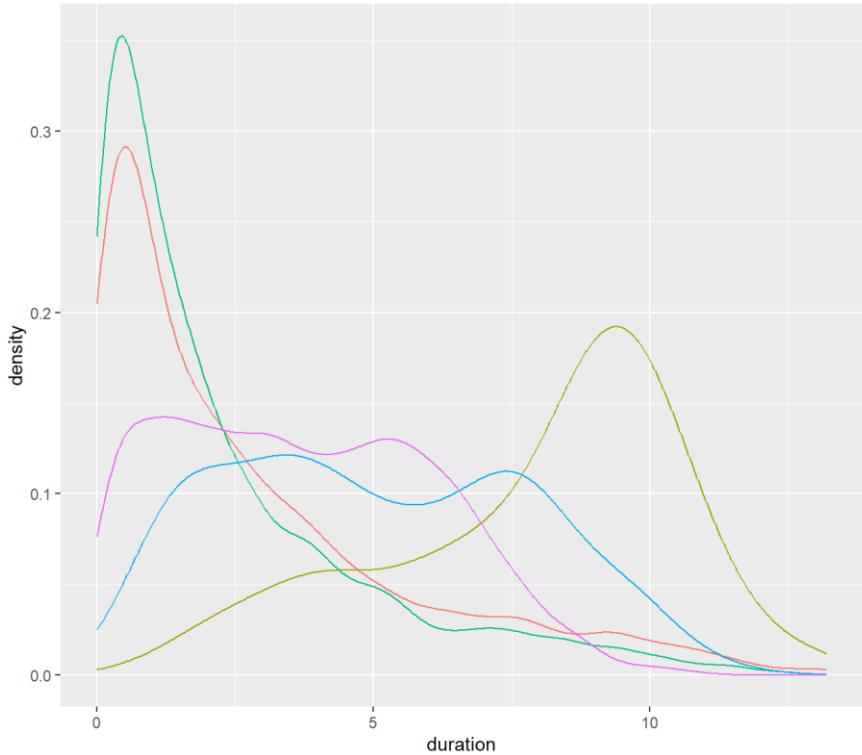
- Average time spent out-of-home, in hh:min (20 schedules for each sample)

| <b>Activity</b> | <b>Data</b> | <b>MDCEV</b>      |                   |           | <b>OASIS</b> |
|-----------------|-------------|-------------------|-------------------|-----------|--------------|
|                 |             | $\gamma$ -Profile | $\alpha$ -Profile | Parabolic |              |
| Working         | 02:36       | 01:15             | 01:15             | 02:53     | 01:03        |
| Education       | 01:07       | 00:21             | 00:22             | 00:20     | 00:31        |
| Shopping        | 00:13       | 00:42             | 01:07             | 00:27     | 00:14        |
| Leisure         | 01:25       | 02:55             | 02:19             | 01:50     | 01:30        |

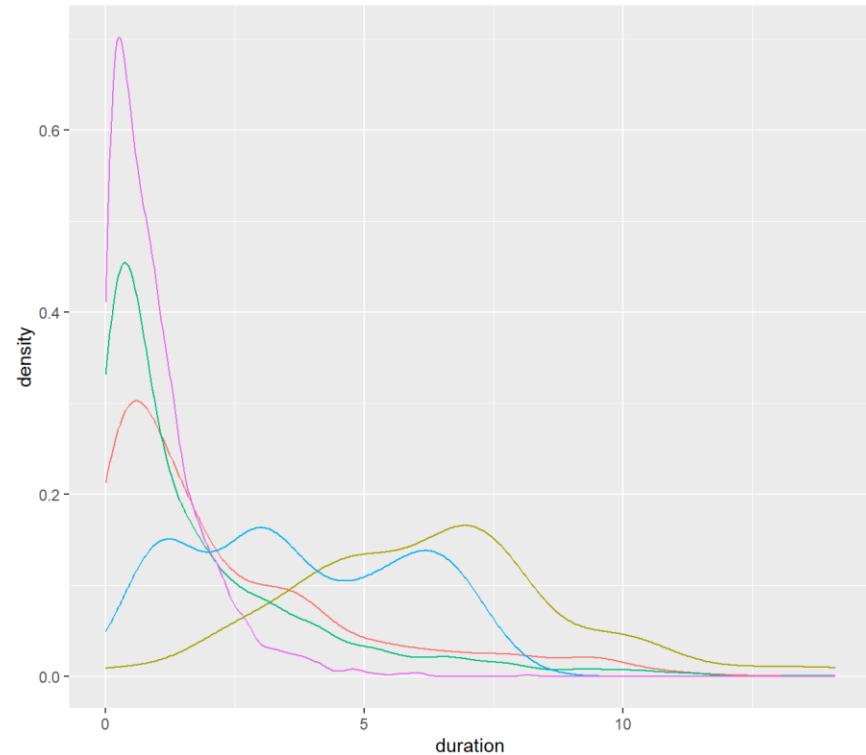
- Proportion of schedules containing each activity (20 schedules for each sample)

| <b>Activity</b> | <b>Data</b> | <b>MDCEV</b>      |                   |           | <b>OASIS</b> |
|-----------------|-------------|-------------------|-------------------|-----------|--------------|
|                 |             | $\gamma$ -Profile | $\alpha$ -Profile | Parabolic |              |
| Working         | 0.33        | 0.54              | 0.45              | 0.77      | 0.21         |
| Education       | 0.18        | 0.19              | 0.16              | 0.33      | 0.14         |
| Shopping        | 0.33        | 0.49              | 0.42              | 0.43      | 0.17         |
| Leisure         | 0.58        | 0.73              | 0.64              | 0.86      | 0.75         |

# Duration Distributions

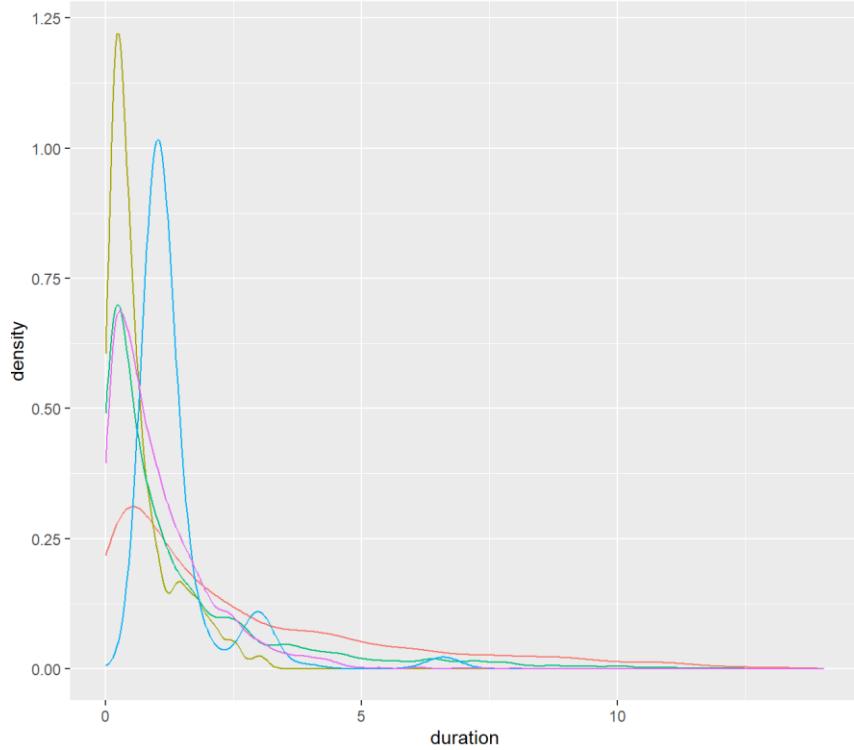


Working

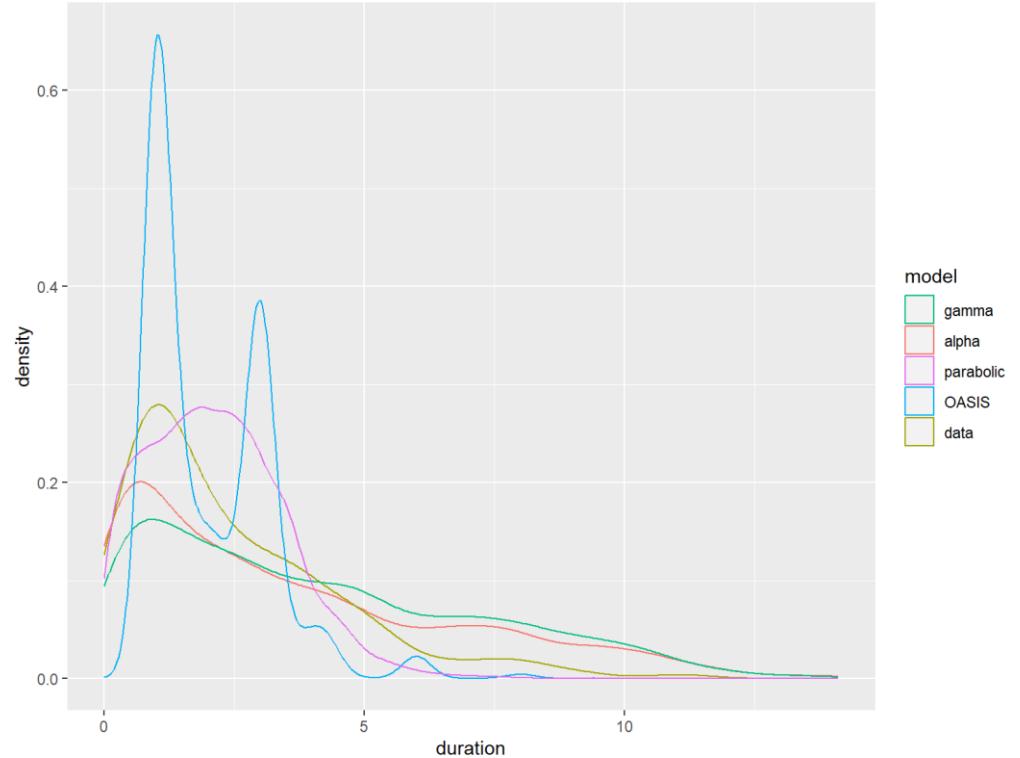


Education

# Duration Distributions



Shopping



Leisure

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## Limitations and Future Directions

- Socio-demographics and more sophisticated utility functions are to be considered;
- Further comparative study is to be carried out when the MDCEV model is combined with extra scheduling algorithm;
- More other datasets are to be tested.

Thanks for listening!

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