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# A DYNAMIC DISCRETE-CONTINUOUS CHOICE MODEL TO EXPLAIN CAR OWNERSHIP, USAGE AND FUEL TYPE DECISIONS

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Aurélie Glerum EPFL

Emma Frejinger Université de Montréal

Anders Karlström KTH

Muriel Beser Hugosson KTH

Michel Bierlaire EPFL

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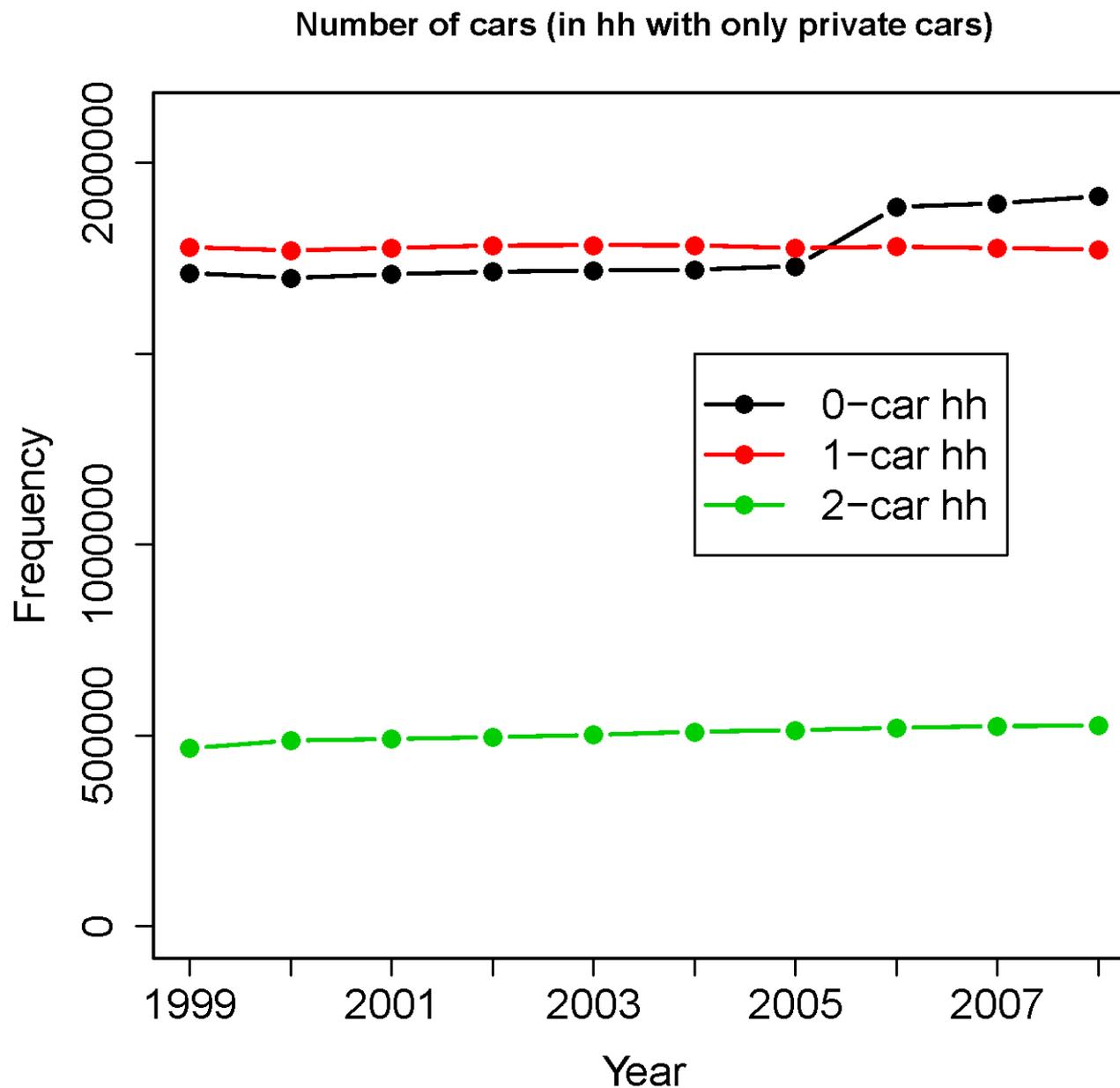
- Introduction
- Background and data
- Dynamic discrete-continuous choice model
- Illustrative example
- Estimation on synthetic data
- Conclusion and future works

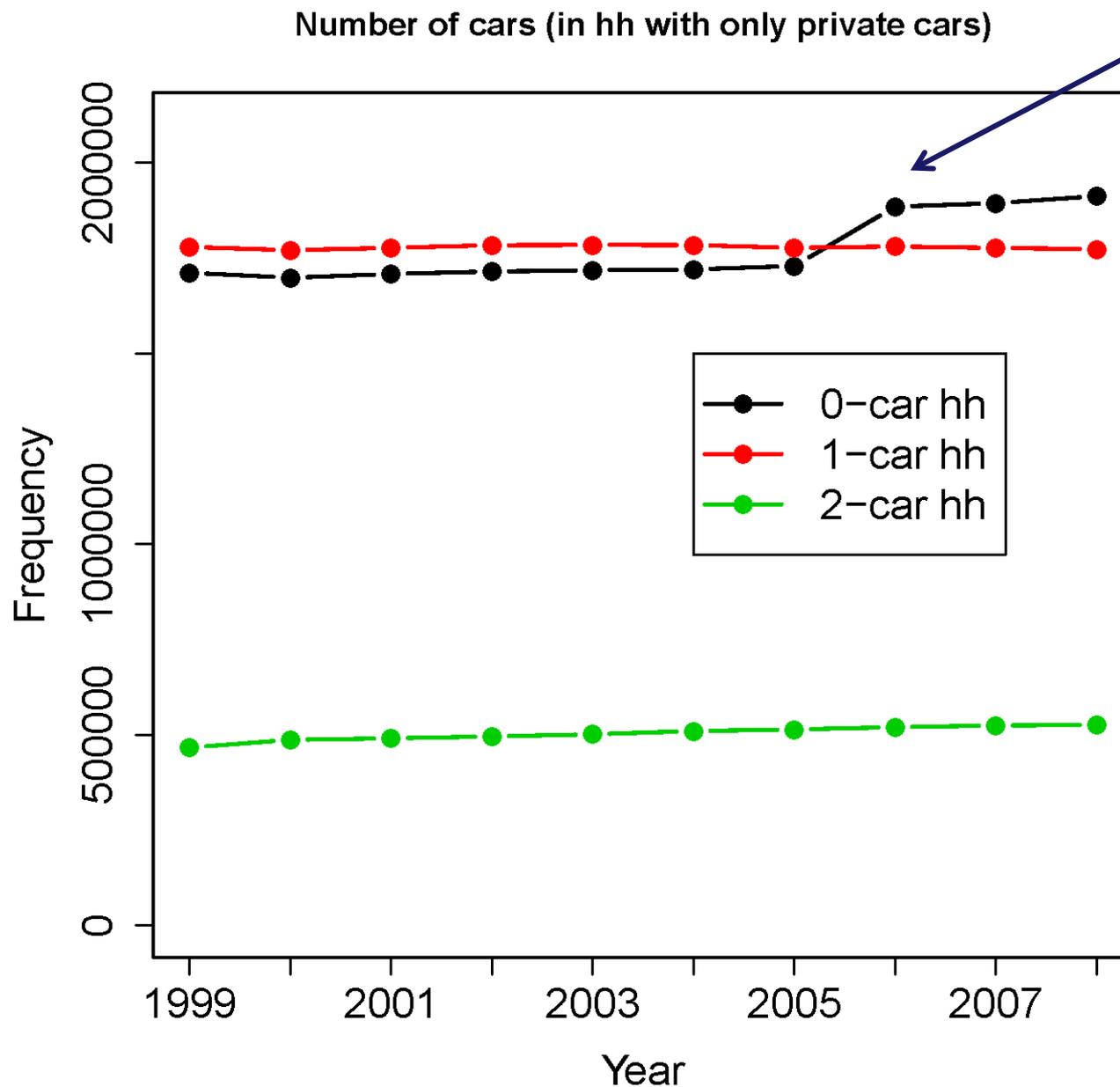
## Aim of the research:

- Model dynamics of car transactions, usage and choice of fuel type in the Swedish car fleet
  - Motivations
    - Governmental policies to reduce carbon emissions / car usage:
      - Stockholm congestion tax
      - Independence of fossil fuels
    - Technology changes:
      - Increase of alternative-fuel vehicles
    - Economical features:
      - Financial crisis
      - Fuel price changes
- ⇒ Car ownership and usage vary importantly over time.
- ⇒ Model needed to analyze and predict impact of policies on ownership and usage

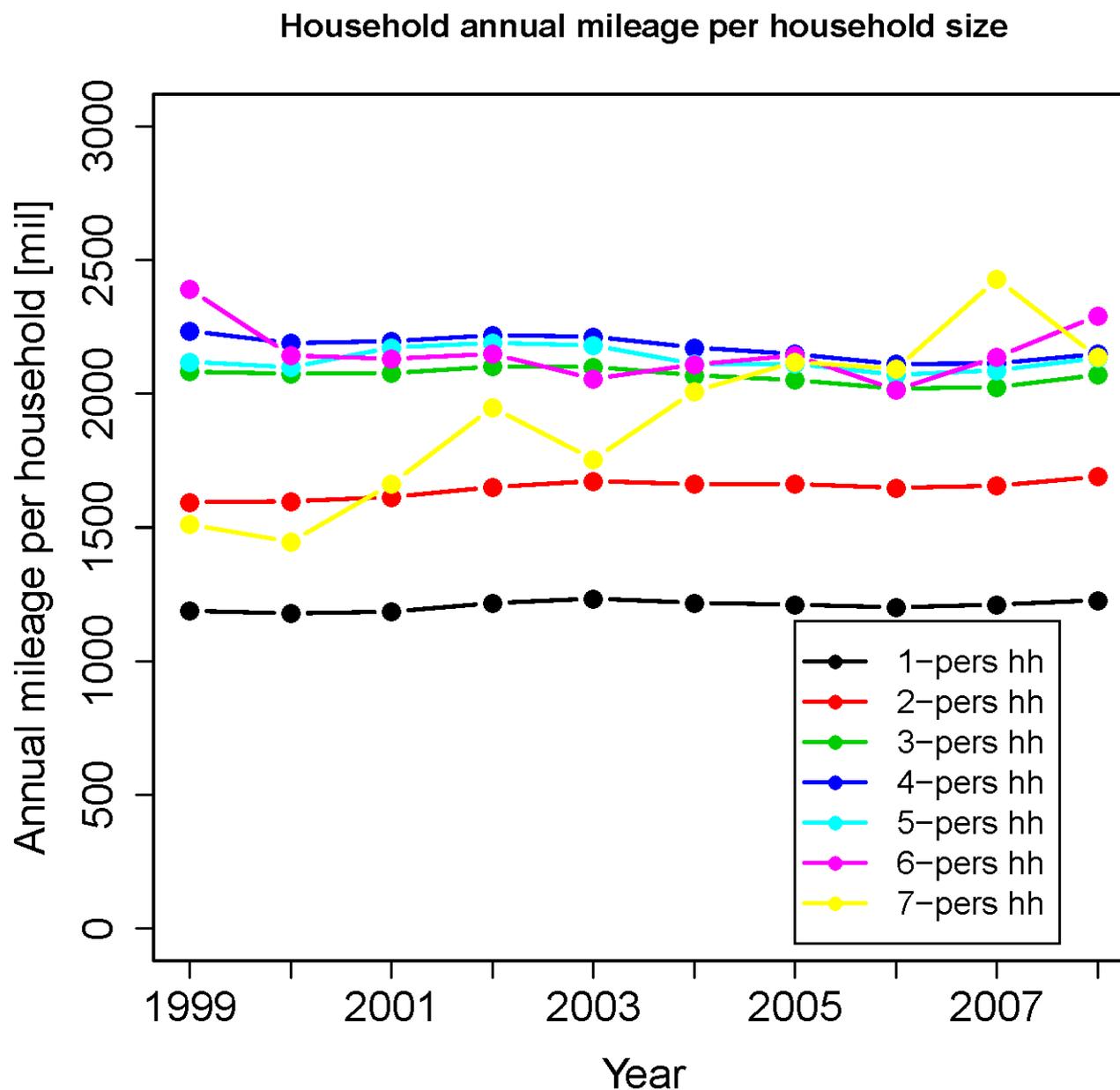
## Register data of Swedish population and car fleet:

- **Data from 1998 to 2008**
- **All individuals**
  - **Individual information:** socio-economic information on car holder (age, gender, income, home/work location, employment status/sector, etc.)
  - **Household information:** composition (families with children and married couples)
- **All vehicles**
  - Privately-owned cars, cars from privately-owned company and **company cars**
  - Vehicle **characteristics** (make, model, fuel consumption, fuel type, age)
  - **Annual mileage** from odometer readings
  - Car bought **new or second-hand**





OWNERSHIP  
Congestion tax  
Stockholm



- **Car ownership models in transportation literature:**
  - Discrete choice models (DCM) widely used, but mostly **static models**.
    - Main drawback: do not account for forward-looking behavior
    - Important aspect to account for since car is a durable good
  - **Econometric literature: dynamic programming (DP) + DCM**
  - Recently, **dynamic discrete choice models (DDCM)** starting to be applied in transportation field (Cirillo and Xu, 2011; Schiraldi, 2011)

- **Joint models of car ownership and usage:**
  - **Duration models** and regression techniques for car holding duration and usage (De Jong, 1996)
  - Vehicle type, usage and replacement decisions using **dynamic programming, discrete-continuous**, mixed logit (Schjerning, 2008, and Munk-Nielsen, 2012)
  - **Discrete-continuous model** of vehicle choice and usage based on register data: includes expectation of fuel prices & car future resale price (Gillingham, 2012)
- **Wide literature on car ownership and usage models**

- **Models for discrete-continuous choices:**
  - **One of the base references:** Joint choice of energy portfolio and energy consumption (Dubin and McFadden, 1984)
  - **Joint choice of vehicle type and usage** (Munk-Nielsen, 2012; Gillingham, 2012)
  - Other references (Hanemann, 1984; Bhat, 2005, 2013)

- Car are durable goods  $\implies$  Need to account for **forward-looking behavior** of agents
- Difficulty of modeling a **discrete-continuous choice** when jointly modeling car ownership and usage
- Many models focus on individual decisions, but choices regarding car ownership and usage made at **household level**

- Car are durable goods  $\Rightarrow$  Need to account for **forward-looking behavior** of agents
- Difficulty of modeling a **discrete-continuous choice** when jointly modeling car ownership and usage
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### Proposed methodology:

- Attempt to address these issues by applying **dynamic discrete-continuous choice model (DDCCM)**

Large **register data** of all **individuals** and **cars** in Sweden



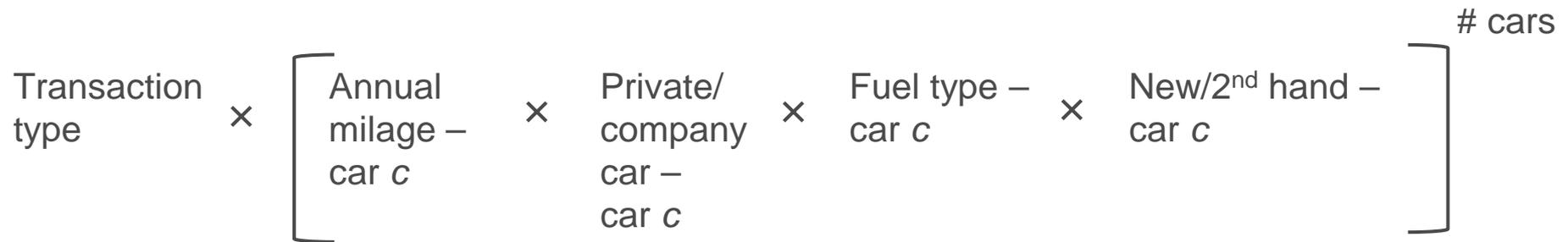
## MAIN FEATURES

- In the area of **dynamic choice modeling**
  - Choices modeled at **household level**
  - **Up to two cars** allowed (only 4% households with > 2 cars in 2007)
- **Constant elasticity of substitution (CES) utility** to model annual driving distance for 2-car households
- Several **choices modeled simultaneously**

### Objective

Model simultaneously car ownership, usage and fuel type.

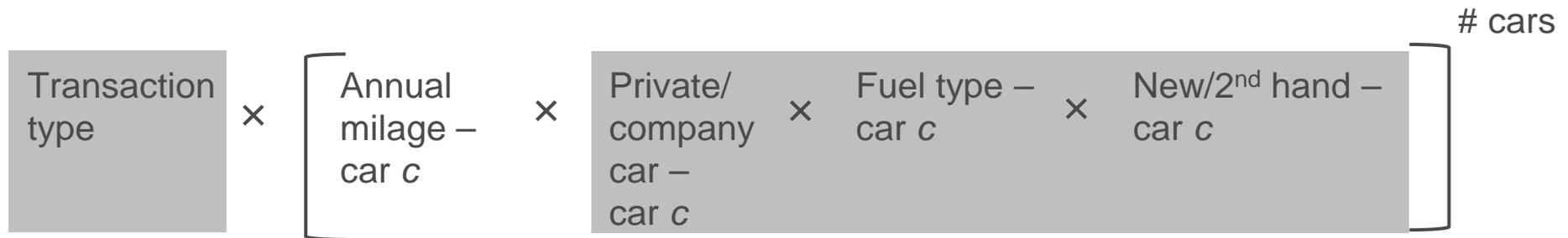
**In details:** model **simultaneous choice** of



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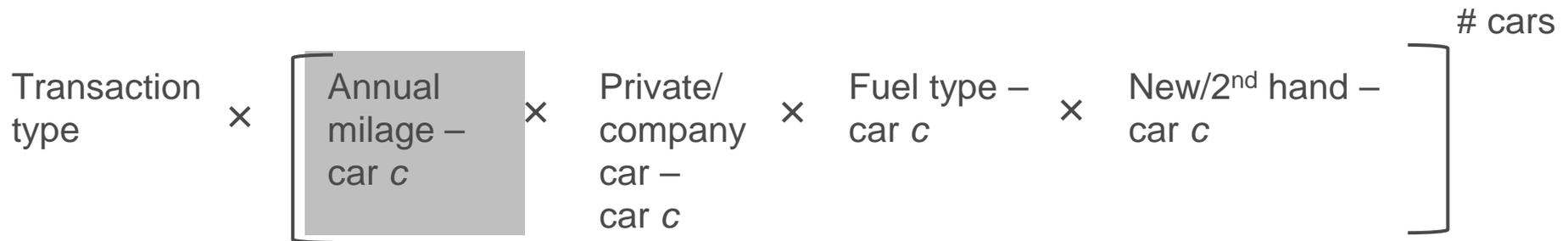


**Discrete variables**

### Objective

Model simultaneously car ownership, usage and fuel type.

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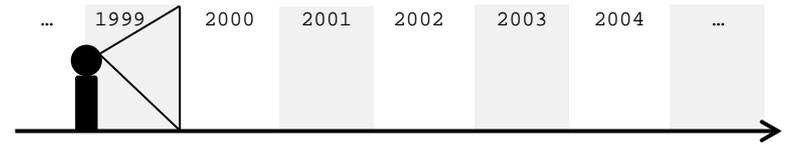


### Continuous variables

1. **Choice at household level:** up to 2 cars in household
2. **Strategic choice of:**
  - Transaction
  - Type(s) of ownership (company vs private car)
  - Fuel type(s)
  - Car state(s) (new vs 2<sup>nd</sup>-hand)

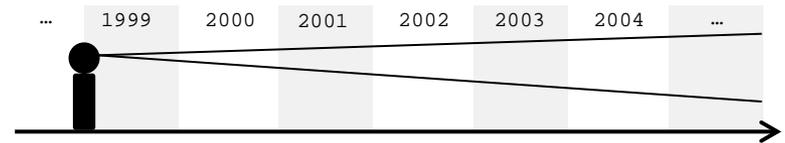
⇒ Account for forward-looking behavior of households
3. **Myopic choice of:**
  - Annual mileage(s)
4. **Choice of mileage conditional** on choice of discrete variables

### Myopic choice (**static** case)



$$P(\text{action}) = \frac{\exp\{\text{instantaneous utility}\}}{\sum_{\text{all poss. actions}} \exp\{\text{instantaneous utilities}\}}$$

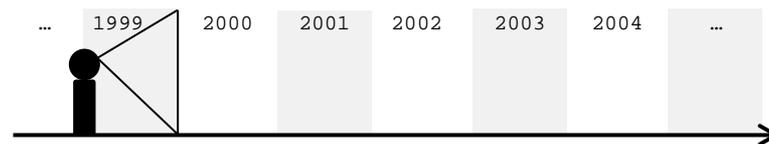
### Strategic choice (**dynamic** case)



$$P(\text{action}) = \frac{\exp\{\text{instantaneous utility} + \text{expected discounted utility of future choices}\}}{\sum_{\text{all poss. actions}} \exp\{\text{instantaneous utilities} + \text{expected discounted utilities of future choices}\}}$$

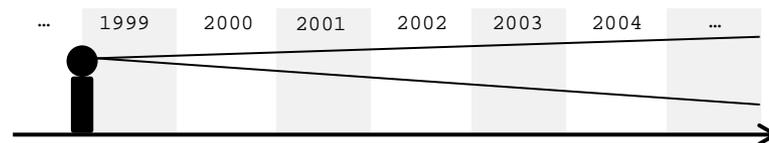
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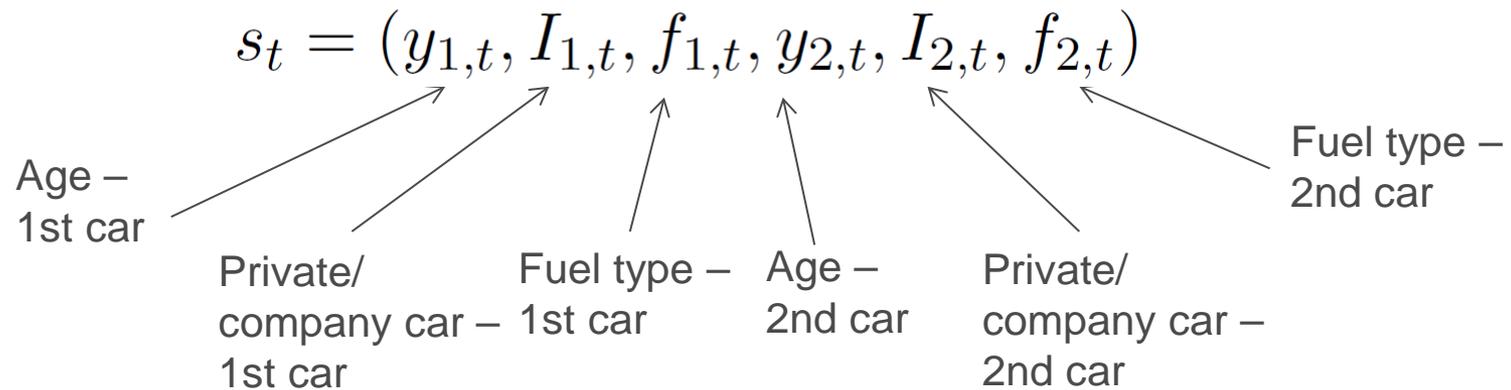


Embeds a **choice model** into a **dynamic programming framework**

### Components of the DDCCM:

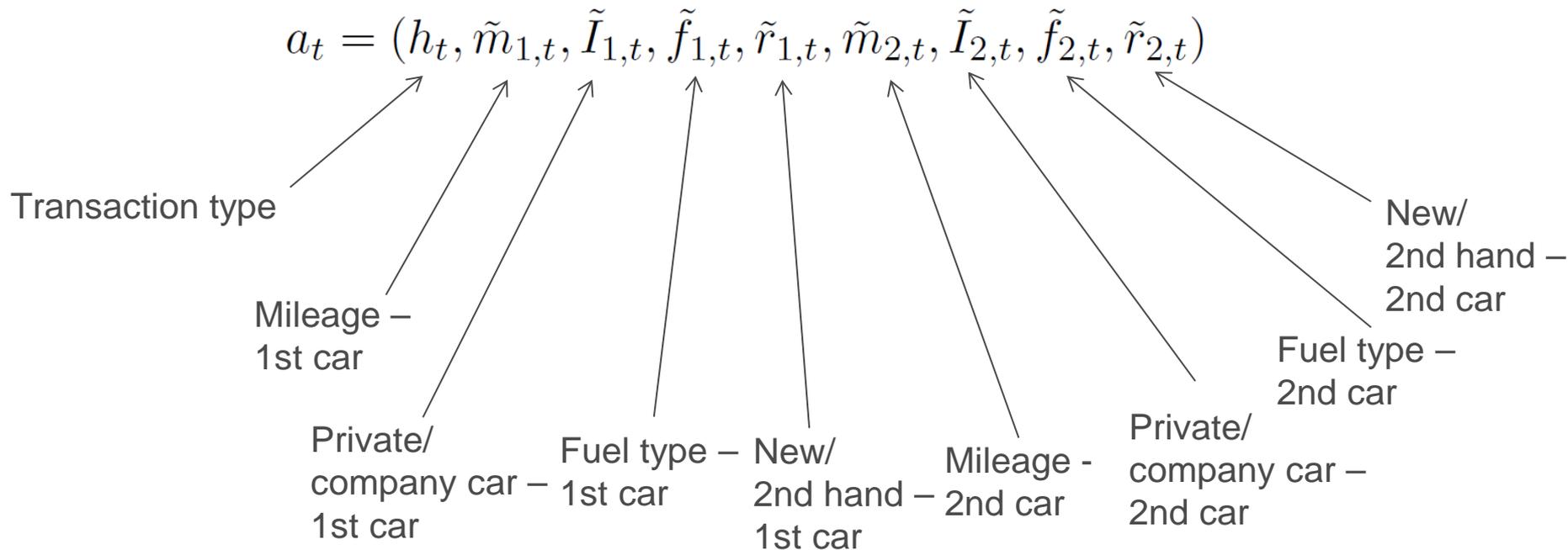
- Agent
- Time step
- State space
- Action space
- Transition rule
- Instantaneous utility function

- **Agent:** household
- **Time step**  $t$ : year
- **State space**  $S$



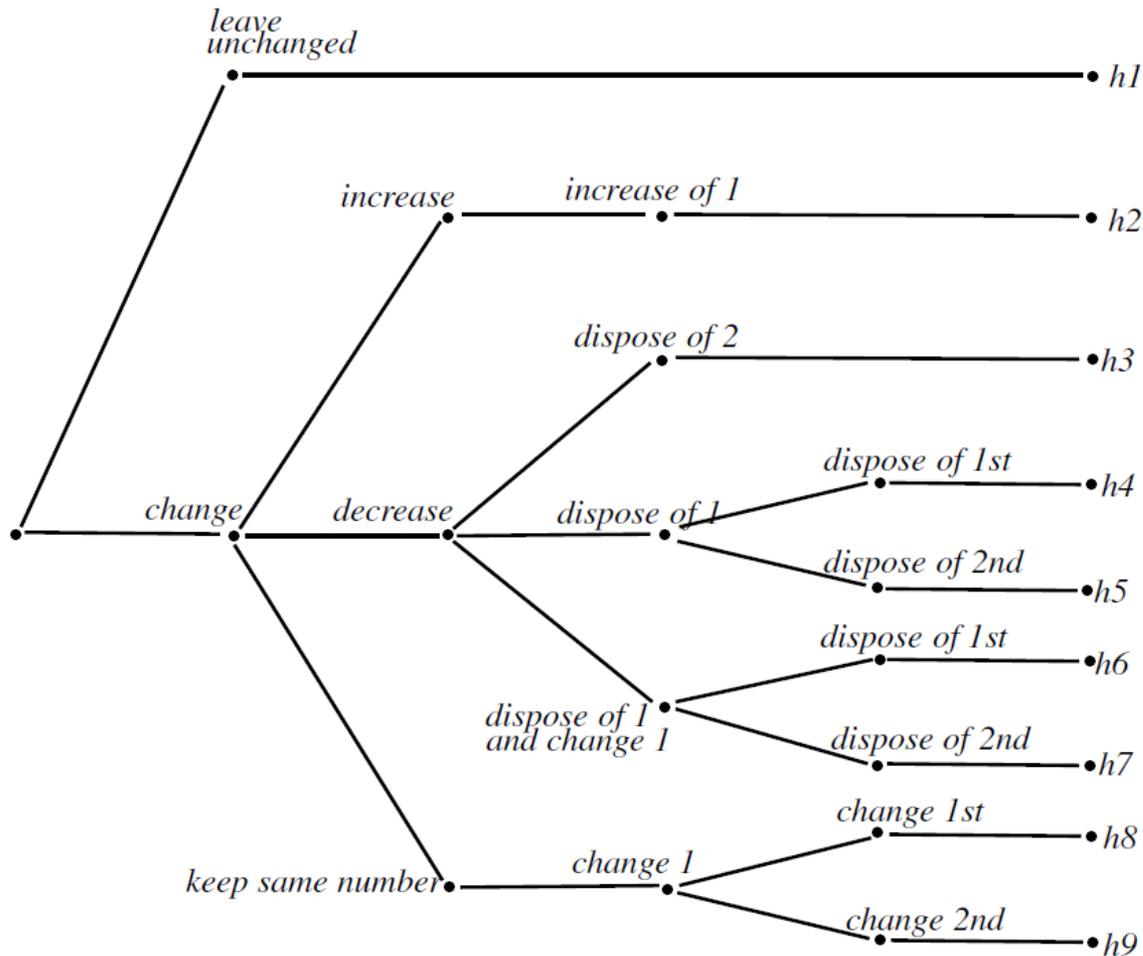
## DEFINITION OF THE COMPONENTS

- Action space  $A$



- Action space  $A$

*Transaction types*



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- **Transition rule:** deterministic rule: each state  $s_{t+1}$  can be inferred exactly once  $s_t$  and  $a_t$  are known.

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- **Instantaneous utility function:**

$$u(s_t, a_t^C, a_t^D, x_t, \theta) = \underbrace{v(s_t, a_t^C, a_t^D, x_t, \theta)}_{\text{Deterministic term}} + \underbrace{\varepsilon_D(a_t^D)}_{\text{Random term for discrete choices}}$$

Assume additive **deterministic utility** for simplicity (see also Munk-Nielsen, 2012):

$$v(s_t, a_t^C, a_t^D, x_t, \theta) = \underbrace{v_t^D(s_t, a_t^D, x_t, \theta)}_{\text{Utility for discrete actions}} + \underbrace{v_t^C(s_t, a_t^D, a_t^C, x_t, \theta)}_{\text{Utility for continuous actions}}$$

- Instantaneous utility function

- Utility for continuous actions:

**Constant elasticity of substitution (CES)** utility function (e.g. Zabalza, 1983):

$$v_t^C(s_t, a_t^D, a_t^C, x_t, \theta) = (m_{1,t}^{-\rho} + \alpha \cdot m_{2,t}^{-\rho})^{-1/\rho}$$

- Captures substitution patterns between the choice of both annual driving distances
  - $\rho$  = elasticity of substitution
  - $\alpha$  = share parameter
- Formulation could be extended to introduce randomness in  $\alpha$ .

- Parameters obtained by **maximizing likelihood**:

$$\mathcal{L} = \prod_{n=1}^N \prod_{t=1}^{T_n} P(a_{n,t}^D | s_{n,t}, x_{n,t}, \theta)$$

- Optimization algorithm**: Rust's **nested fixed point algorithm (NFXP)** (Rust, 1987):
  - Outer optimization algorithm**: search algorithm to obtain parameters maximizing likelihood
  - Inner value iteration algorithm**: solves the dynamic programming problem for each parameter trial
- Plan to investigate variants of NFXP to speed up computational time (e.g. swapped algorithm from Aguirregabiria and Mira, 2002)

### Outer algorithm:

- Standard estimation procedure (as for static discrete choice models)
- Here: BHHH algorithm

### Inner algorithm:

#### Two steps

1. Finding the optimal value(s) of annual mileage conditional on the discrete choices
2. Finding the expected discounted utility of future choices (= value function)

### 1. Finding the optimal value(s) of mileage

- Maximization of the continuous utility:  $\max_{m_{1,t}, m_{2,t}} v_t^C$   
 s.t.  $p_{1,t}m_{1,t} + p_{2,t}m_{2,t} = \text{Inc}_t$

- Find analytical solutions:  $m_{1,t}^*$  and  $m_{2,t}^*$

$$m_{2,t}^* = \frac{\text{Inc}_t \cdot p_{2,t}^{(-1/(\rho+1))}}{p_{2,t}^{(\rho/(\rho+1))} + p_{1,t}^{(\rho/(1+\rho))} \alpha^{(-1/(\rho+1))}}$$

$$m_{1,t}^* = \frac{\text{Inc}_t}{p_{1,t}} - \frac{p_{2,t}}{p_{1,t}} m_{2,t}^*$$

- Optimal continuous utility  $v_t^{C*}(s_t, a_t^D, a_t^{C*}, x_t, \theta)$

## 2. Finding the expected discounted utility of future choices (= value function)

- **Logsum** formula used in the completely discrete case (DDCM) (Aguirregabiria and Mira, 2010; Cirillo and Xu, 2011)
- Logsum can be applied here given the **key assumptions**:
  - Choice of mileage(s) is conditional on discrete actions
  - Choice of mileage(s) is myopic

$$\bar{V}(s_t, x_t, \theta) = \log \sum_{a_t^D} \exp\{v_t^D(s_t, a_t^D, x_t, \theta) + v_t^{C^*}(s_t, a_t^D, a_t^{C^*}, x_t, \theta) + \beta \sum_{s_{t+1} \in S} \bar{V}(s_{t+1}, x_{t+1}, \theta) f(s_{t+1} | s_t, a_t)\}$$

- Iterate on **Bellman equation** to find integrated value function  $\bar{V}$

# ILLUSTRATIVE EXAMPLE

## Assumptions for the example:

- Deterministic utility function

$$v_t^D(s_t, a_t^D, x_t, \theta) = C(s_t) + \tau(a_t^D) + \beta_{\text{Age}}(a_t^D, s_t) \cdot \max(\text{Age1}_t, \text{Age2}_t)$$

Constant for  
households with  
at least one car

Transaction  
costs

Transaction-dependant  
parameters relative to age  
of oldest car

- Chose arbitrary values for parameters

# ILLUSTRATIVE EXAMPLE

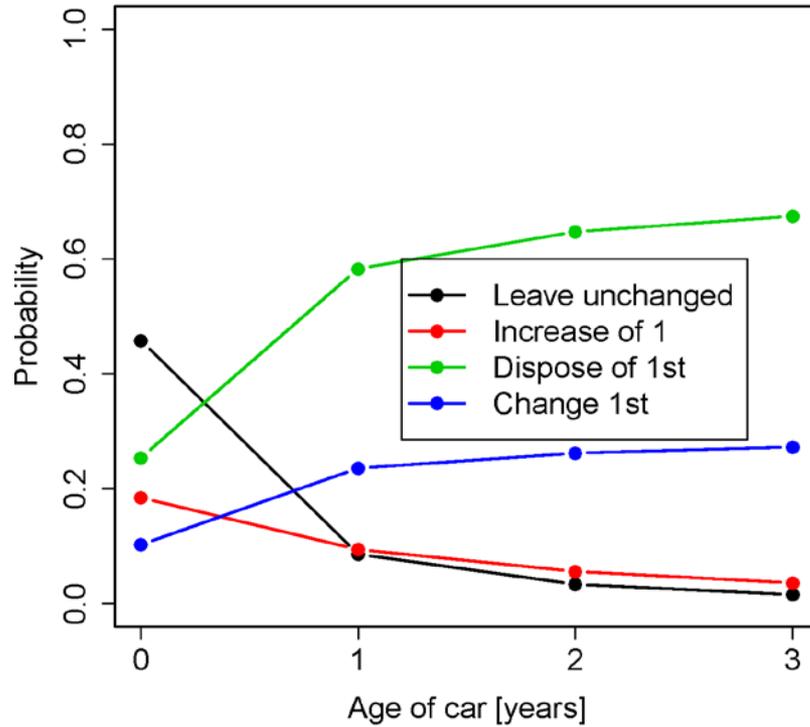
Transaction name	Case	$\beta_{Age}$			$\tau$
		0 car	1 car	2 cars	all households
$h_1$ : leave unchanged		0	-1	-1	0
$h_2$ : increase 1		0	0	-	-3
$h_3$ : dispose 2		-	-	1	0
$h_4$ : dispose 1st	1st car is oldest	-	1.5	1.5	0
	2nd car is oldest	-	-	0	0
$h_5$ : dispose 2nd	1st car is oldest	-	-	0	0
	2nd car is oldest	-	-	1.5	0
$h_6$ : dispose 1st and change 2nd		-	-	0	-4
$h_7$ : dispose 2nd and change 1st		-	-	0	-4
$h_8$ : change 1st	1st car is oldest	-	1.5	1.5	-4
	2nd car is oldest	-	-	0	-4
$h_9$ : change 2nd	1st car is oldest	-	-	0	-4
	2nd car is oldest	-	-	1.5	-4

$C = 5$ , for 1- or 2-car households

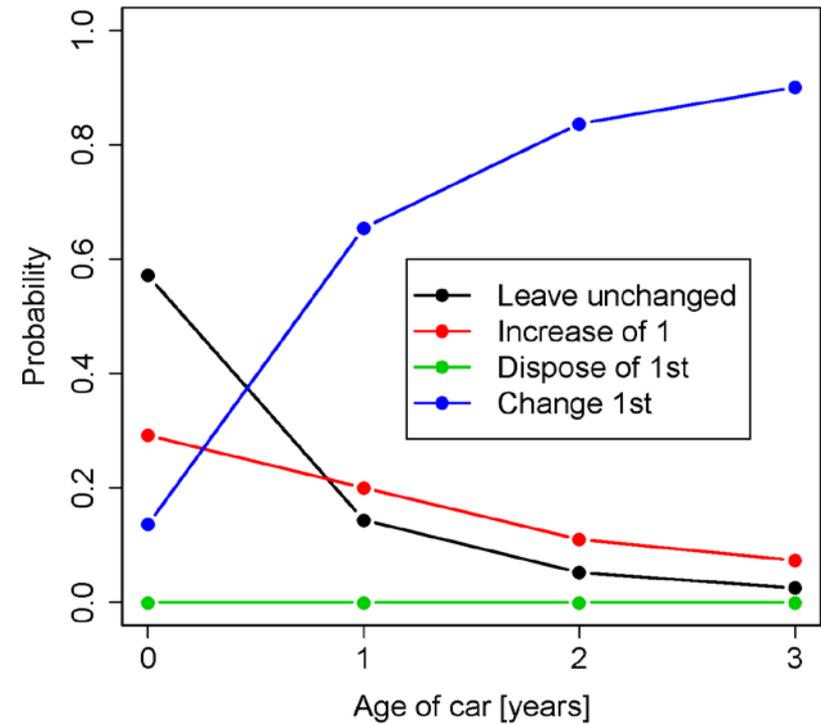
## Assumptions for the example:

- Visualize choice probabilities for one observation:
  - 1-car household
  - Annual income = 530'000 SEK ( $\approx$  60'200 €, 74'200 CHF)
  - 8% expenses on fuel

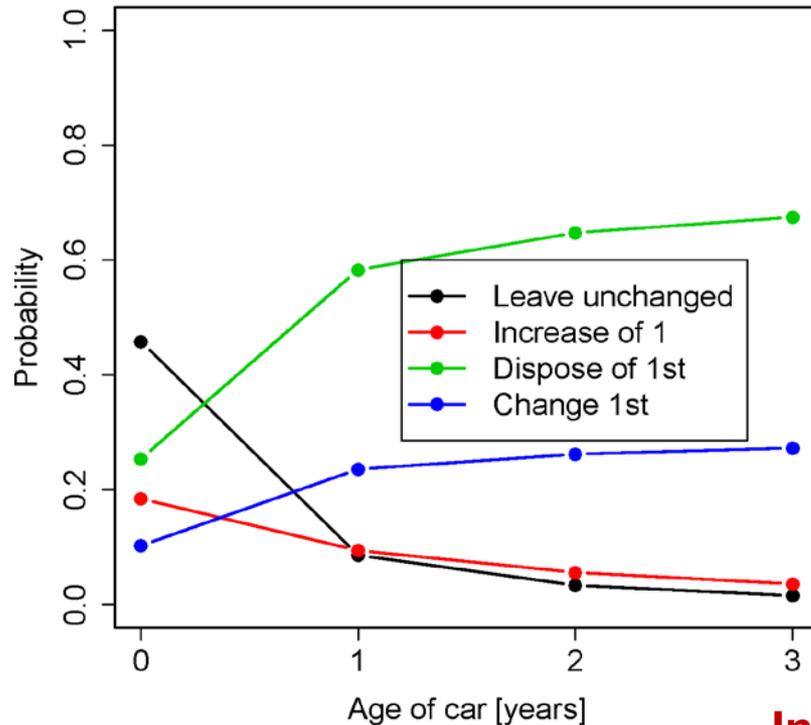
## Static



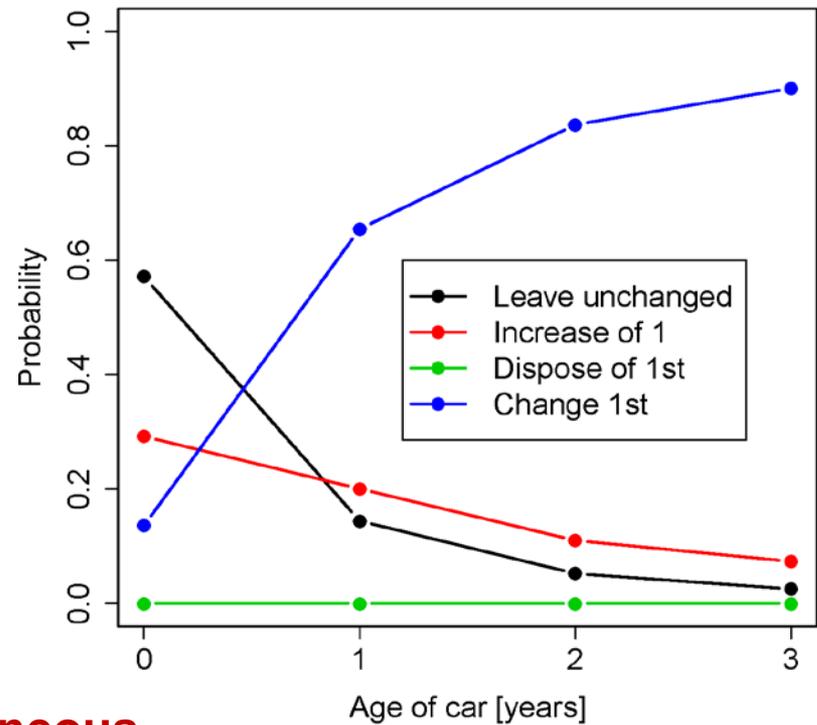
## Dynamic



## Static



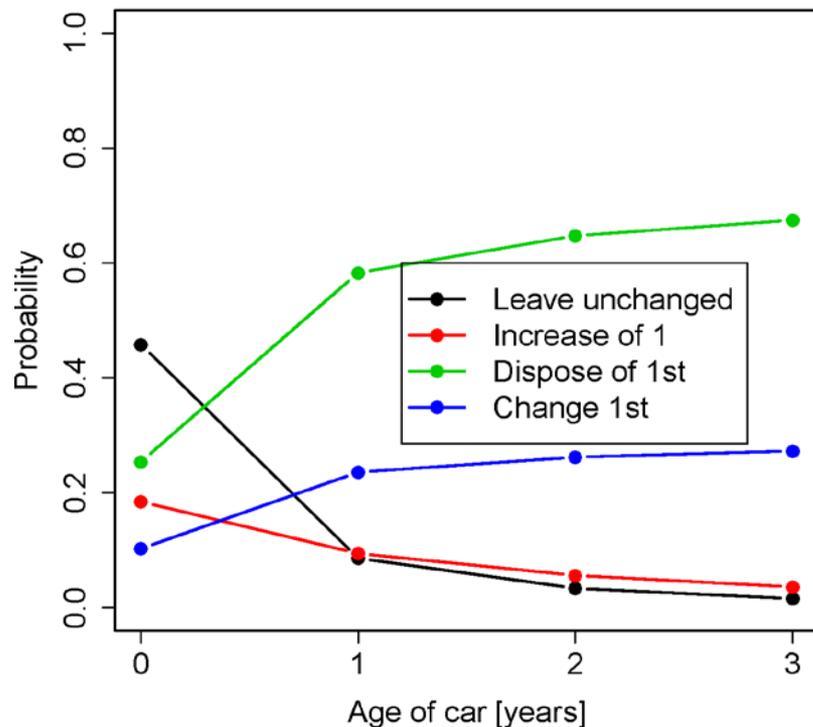
## Dynamic



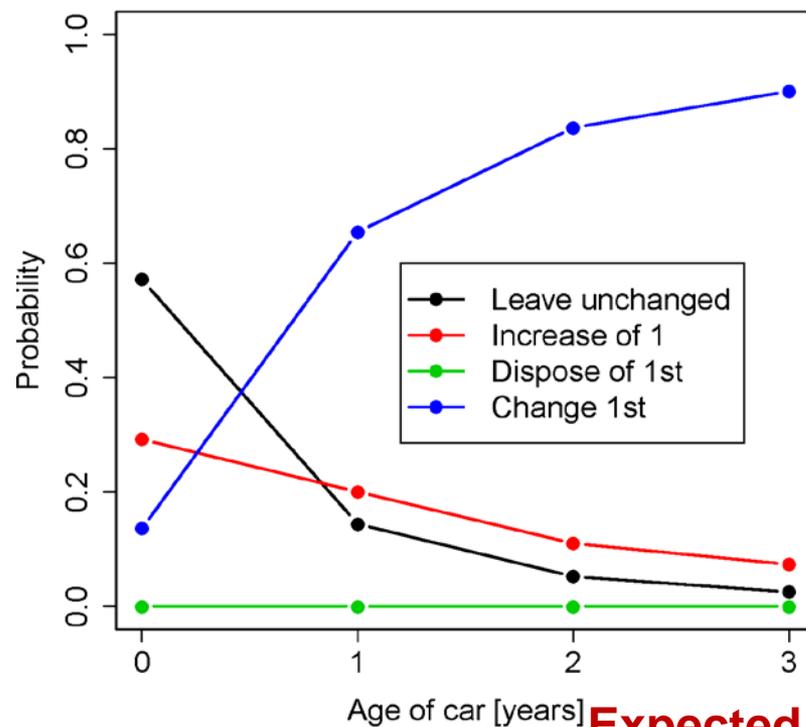
**Instantaneous utility**

$$P(a_{n,t}^D | s_{n,t}, x_{n,t}, \theta) = \frac{v_{n,t}^D + v_{n,t}^{C*} + \beta \sum_{s_{n,t+1} \in S} \bar{V} f}{\sum_{a_{n,t}^{\tilde{D}}} \left\{ v_{n,t}^D + v_{n,t}^{C*} + \beta \sum_{s_{n,t+1} \in S} \bar{V} f \right\}}$$

## Static



## Dynamic



**Expected  
discounted  
utility**

$$P(a_{n,t}^D | s_{n,t}, x_{n,t}, \theta) = \frac{v_{n,t}^D + v_{n,t}^{C*} + \beta \sum_{s_{n,t+1} \in S} \bar{V} f}{\sum_{a_{n,t}} \left\{ v_{n,t}^D + v_{n,t}^{C*} + \beta \sum_{s_{n,t+1} \in S} \bar{V} f \right\}}$$

## Approach to validate the model framework

- Generate 100 observations based on **distributions of variables** in the **Swedish register data**
- **Generate choice** (for each observation) based on postulated parameters
- **Estimation** of model
- Approach validated once **postulated parameters** are **retrieved**

## Conclusion:

- Methodology to model choice of car ownership and usage dynamically
- Example of application shows how static & dynamic cases can differ
- Currently validating the approach (using synthetic data)

## Next steps:

- Model estimation on register data
- Scenario testing:
  - Validation of policy measures taken during the years available in the data
  - Test policy measures that are planned to be applied in future years

# Thank you!

