

# A Decade of Dynamics of Residential Location, Car Ownership, Activity, Travel and Land Use in the Seattle Metropolitan Region

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# Overview

- Briefly on Triggers
- Data Used
- Multilevel Cluster Models
  - Persons within households
  - Households over time
- Attitudes vs Longitudinal Travel Behavior
- Next Steps?

# General Motivation of this Research

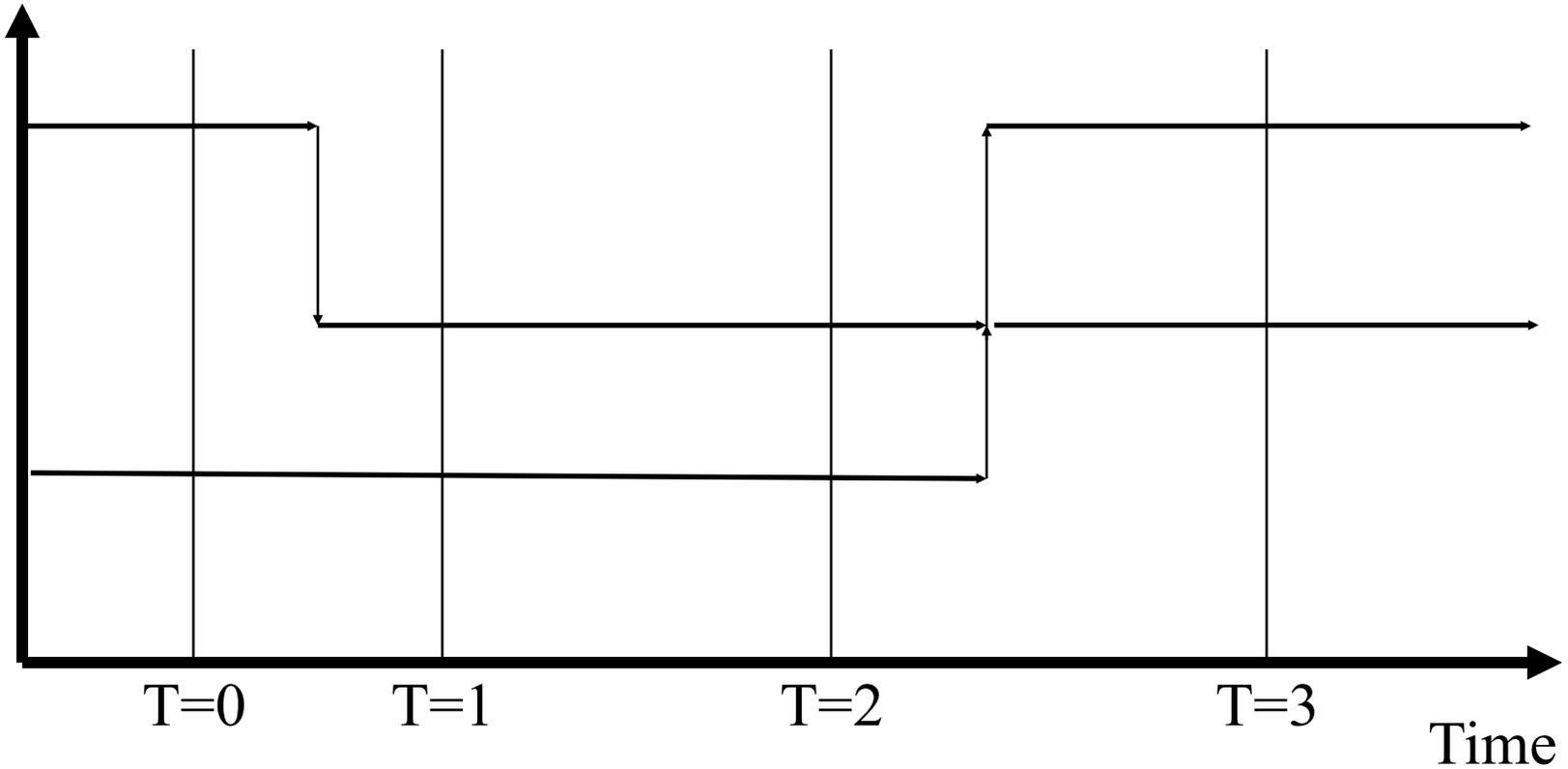
- We change land use to change behavior (e.g., higher density and diversity of opportunities)
- We change car ownership rules to change behavior (e.g., incentives for hybrid electric cars)
- We extrapolate from cross sectional differences changes over time (e.g., rich and poor) – we know it is wrong!
- We NEED a longitudinal tool to tell us what happens when “things” change inside and outside the household
- We need models that can be used in large scale dynamic activity simulators

The tool used here

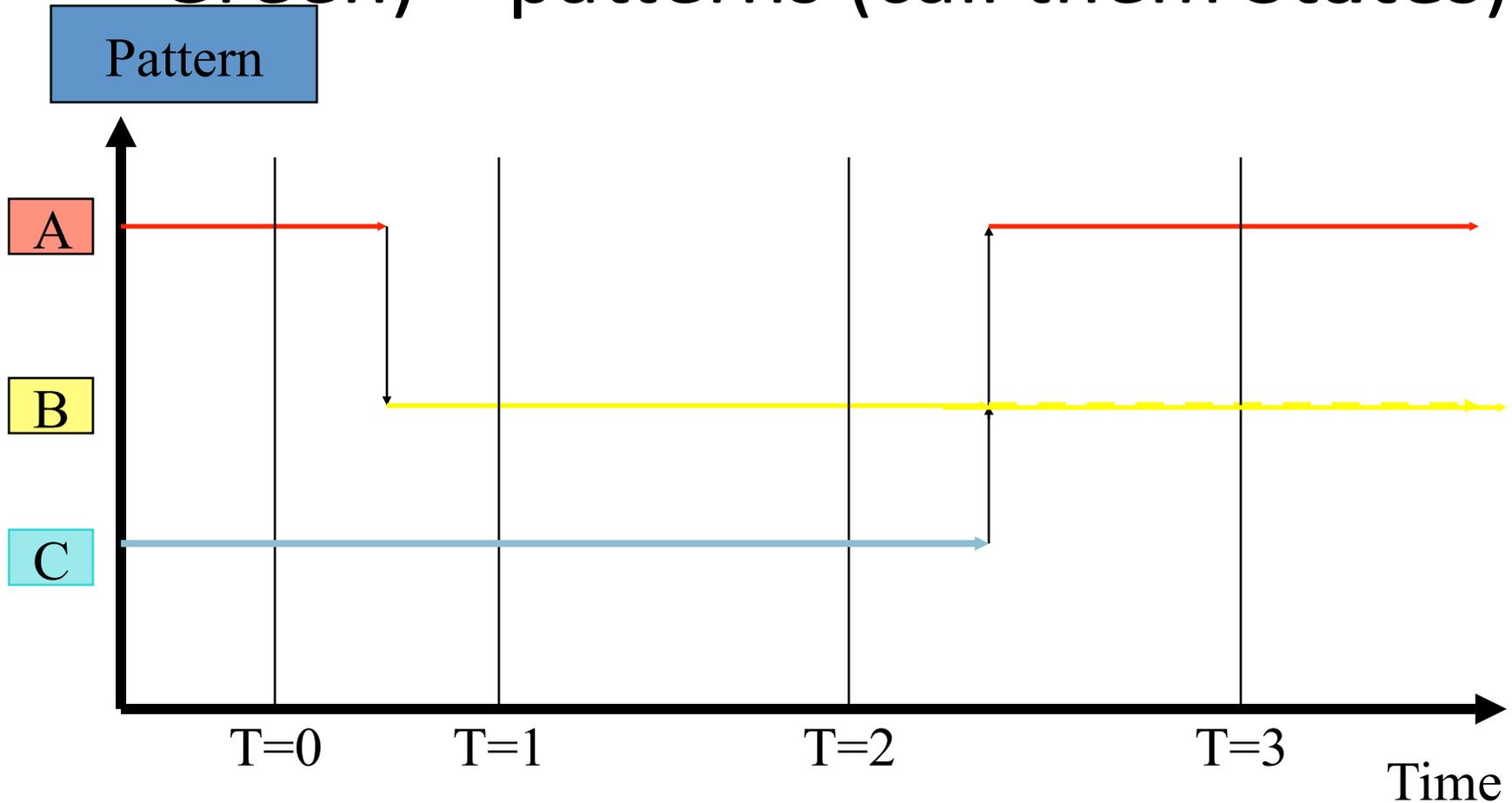
# **MIXED MARKOV LATENT CLASS & PAST FINDINGS**

# We Observe 2 households

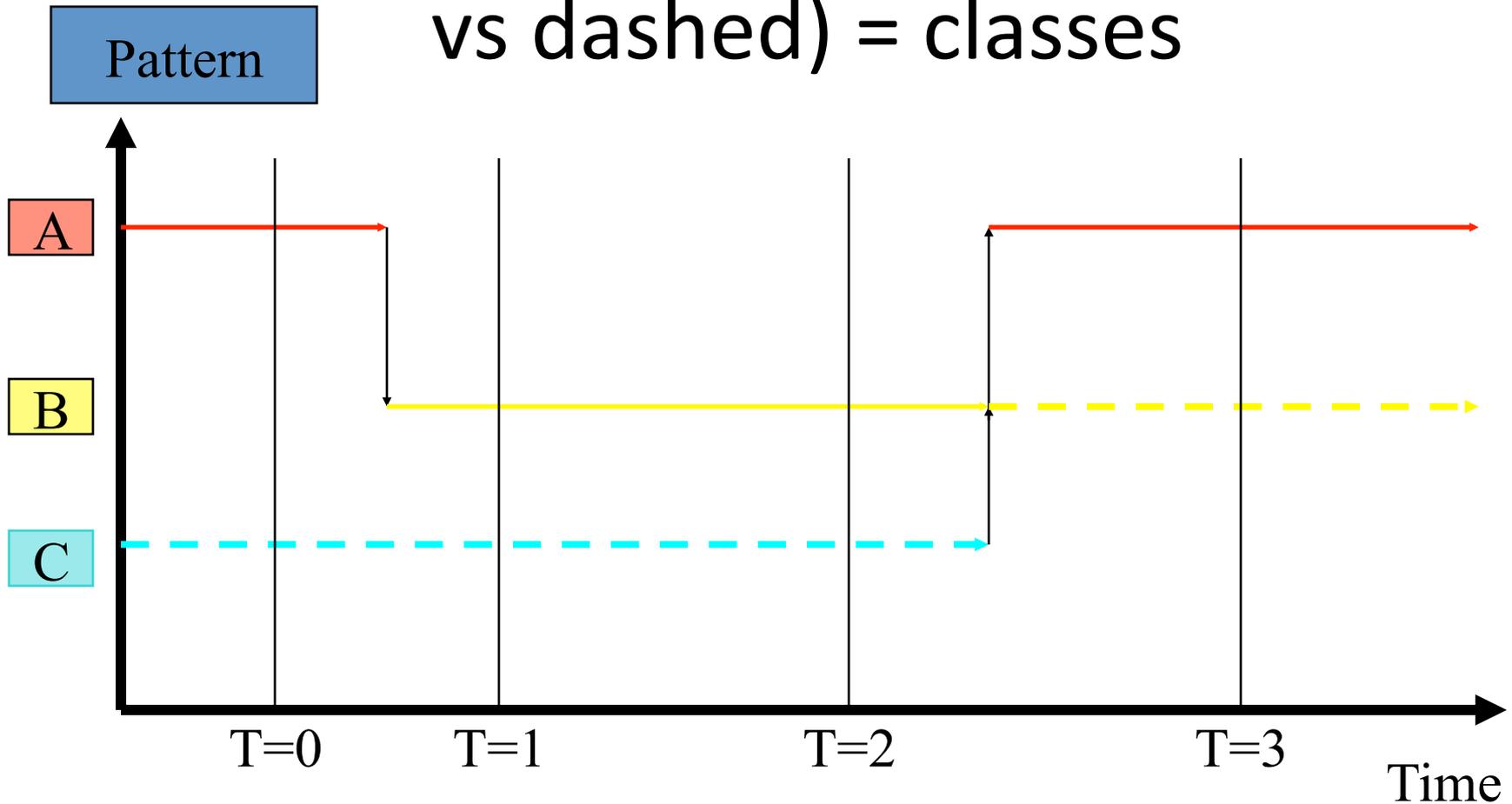
Number of cars or trips



# Derive Clusters of Behavior (Red-Yellow-Green) = patterns (call them **States**)



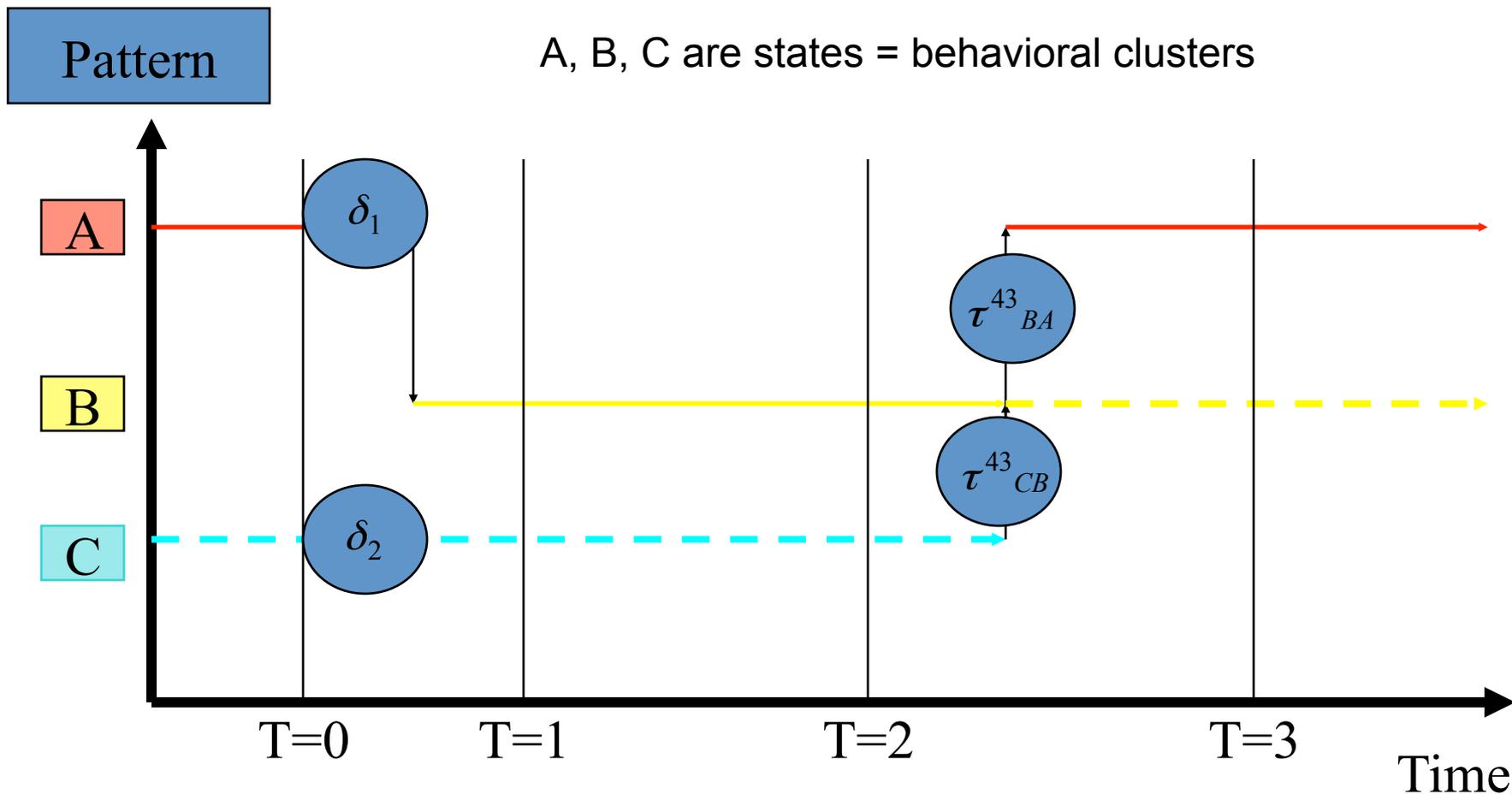
# Identify Groups of Households of Different Sequences of Patterns (solid vs dashed) = classes



Solid Line = Household 1

Dashed Line = Household 2

# Parameters of Dynamics = switching among patterns by different classes & probabilities of switching



Solid Line = Person 1

Dashed Line = Person 2

# MMLC

- includes a multi-category latent variable ( $w$ ) representing household heterogeneity and its categories called **classes**;
- uses many “dependent” or response variables ( $y$ ) forming another set of categorical latent variables ( $x$ ) at 10 time points and its categories are called **states**;
- **transitions** from one state at one time to another state at another time point are estimated using a probability model and they are functions of **triggers**;
- uses and tests the effect of **covariates (Zs)** of many different specifications of the models; and
- is a model-based clustering approach providing probabilistic membership of observations in clusters of **classes**, clusters of **states**, and probabilities of **transitions from state to state over time**.

# Transitions

Difference in  
sociodemographics  
and land use between  
 $t-1$  and  $t$

Difference in  
sociodemographics  
and land use  
between  $t$  and  $t+1$

Change from State A to State B at time  $t$

```
graph TD; A["Difference in sociodemographics and land use between t-1 and t"] --> C["Change from State A to State B at time t"]; B["Difference in sociodemographics and land use between t and t+1"] --> C; D["Elapsed Time and Time Squared from 1989"] --> C; E["Average Intra household Age"] --> C;
```

Elapsed Time and  
Time Squared from  
1989

Average Intra  
household Age

# Significant Triggers

(variables that are significant in the transition models)

- Workers (mixed causality)
- Children 6-17 (increase activity)
- Mean Age (aging decreases activity)
- Adults (increase activity)
- Increase in young children no impact

# Significant Triggers

(variables that are significant in the transition models)

- Density (Number of Business Establishments and Employees)
  - *Significant but mixed impacts*
  - *We get mixed indications from the 0.5 miles and 2.0-2.5 mile annulus*
- Diversity (Shannon index)
  - *Land use div. increase within 0.5 more likely to stay at same behavior*
  - *Diversity decreases travel*

**GOOD BUT NOT ENOUGH!**

# Specific Questions of the Presentation (bring attitudes in longitudinal models)

- Do people with different attitudes live together?
- Do attitudes early in a person's life persist in its effect in later years?
- Can we identify distinct attitudinal (x-section) and behavioral (over time) groups and study systematic differences among them?

# DATA USED

# Puget Sound region

- 3.7 million residents
- 4 counties
- 82 cities and towns
- 26 regional growth centers

Source: [www.psrc.org](http://www.psrc.org)



# Data

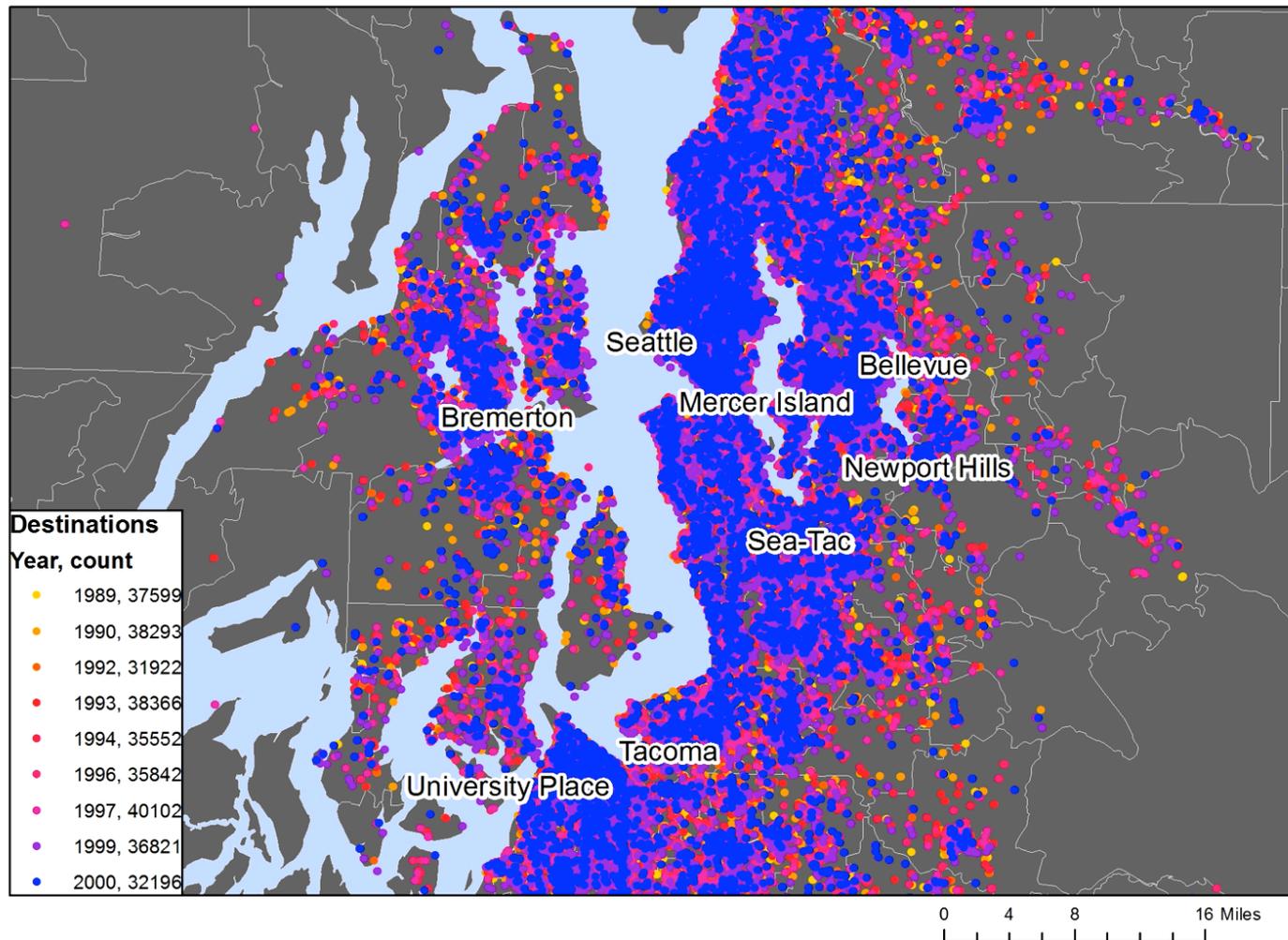
- Puget Sound Transportation Panel, 1989 – 2002
  - Indicators of travel behavior over time
- National Establishments Time Series, 1990 – 2002
  - Indicators of land use over time

# Puget Sound Transportation Panel (PSTP)

- The first general-purpose trans. household panel survey in USA
- Administered on the same households and persons within each household repeatedly over time.
- Started in 1989 with irregular intervals – 10 waves to 2002
- Collected
  - Household demographics,
  - Persons' social & economic information,
  - Travel behaviour through two-day travel diary
  - Attitudes about different aspects of trans. system
- Added *Technology Questions* in 1997, 1999, 2000, 2002

# Puget Sound Transportation Panel

Destinations span the entire region



# Two-part Analysis

- Attitudes about carpooling, transit, and private car use
- 23 attitudinal questions
- 2,472 persons in 1,362 households
- Test homophily (birds of a feather hypothesis)
- Use multilevel cluster analysis to classify people and households jointly
- Behavior over time 1989 - 2002
- 230 households that participated in all ten waves of PSTP
- Use the information about attitudes to explain trajectories of behavior
- Use changes in land use to explain transitions from one type of behavior to another
- Use changes in demographics to explain transitions of behavior from one type to another

# **ATTITUDE GROUPS (CLUSTERS)**

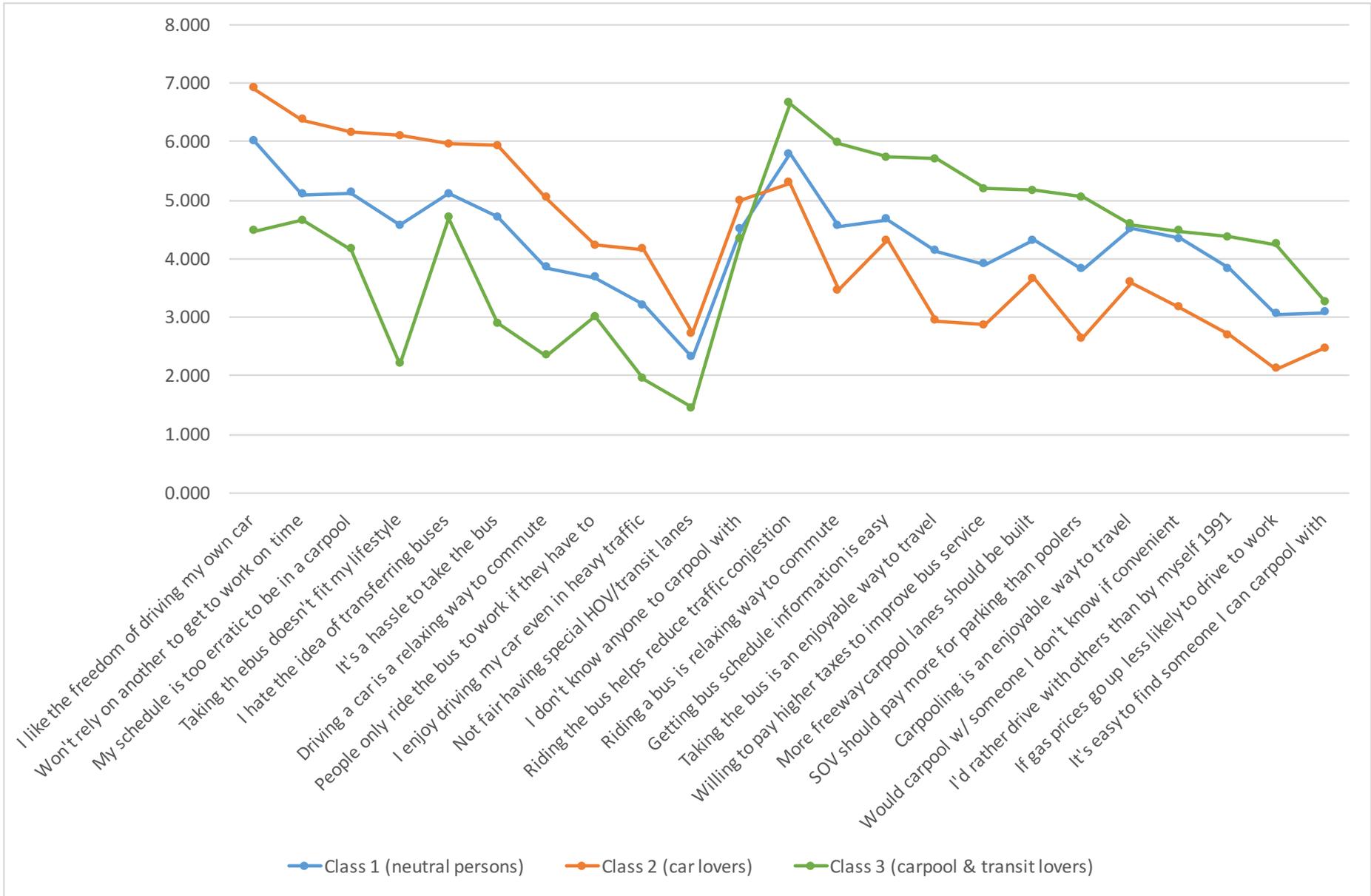
# Multi-level Latent Class Clustering (persons in households)

$$f(Y_k) = \sum_{h=1}^H P(w_k = h) \prod_{j=1}^{n_k} f(Y_{kj} | w_k = h)$$
$$f(Y_{kj} | w_k = h) = \sum_{l=1}^L P(x_{kj} = l | w_k = h) \prod_{i=1}^I f(y_{kji} | x_{kj} = l, w_k = h)$$

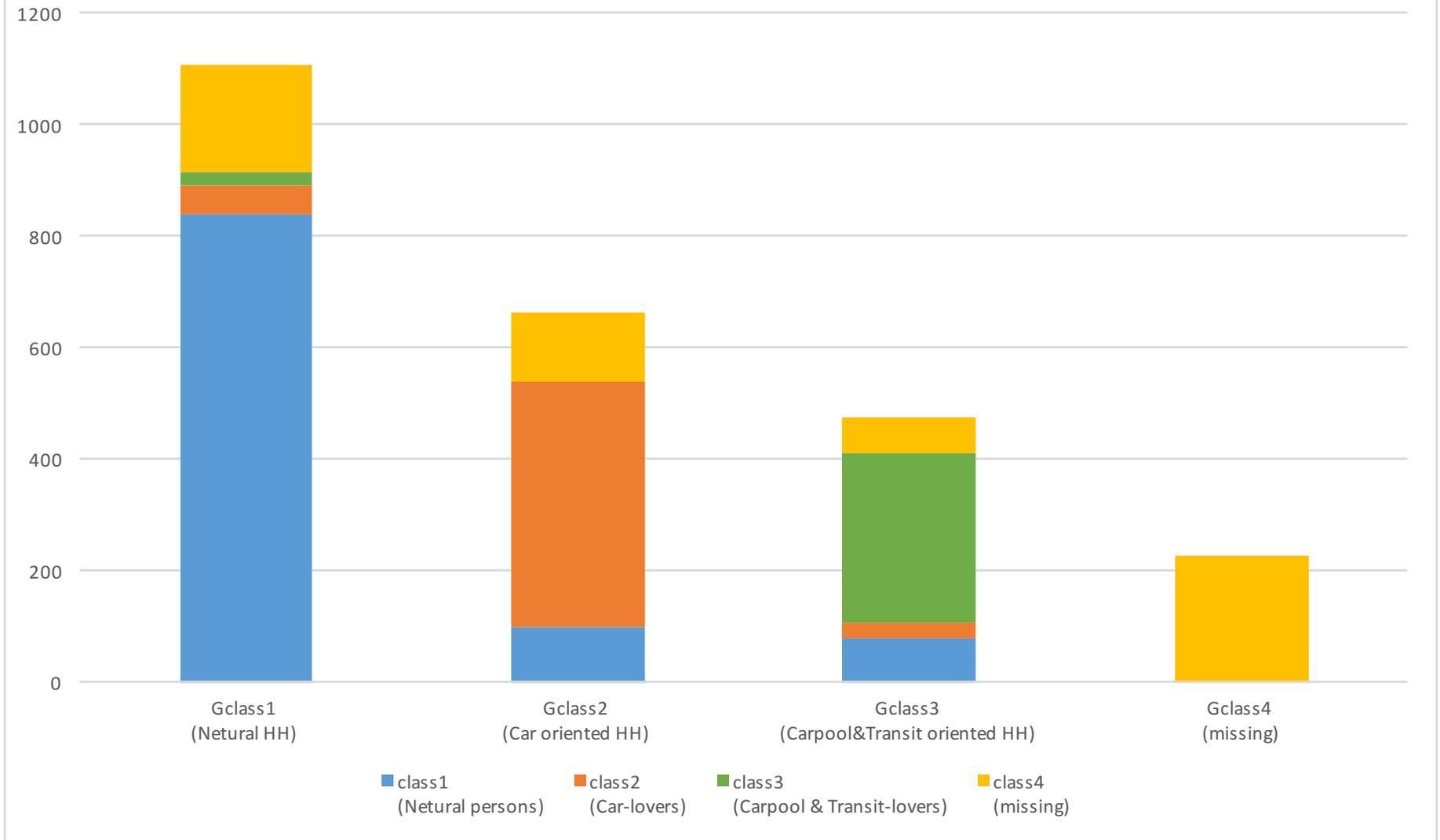
index  $l$  ( $l=1, \dots, L$ ) is used for the person groups and the index  $h$  ( $h=1, \dots, H$ ) for the household groups.  $f(y_{kji} | x_{kj}=l, w_k=h)$  is the conditional density for variable  $i$  of individual  $j$  in household  $k$  given his/her membership in person-level class  $l$  and household level class  $h$ .

$P(x_{kj}=l | w_k=h)$  is the probability that individual  $j$  of household  $k$  belongs to latent class  $l$  given that his/her household belongs to latent class  $h$ .  $P(w_k=h)$  is the probability that household  $k$  belongs to latent class  $h$ .

1: Very strongly disagree, 2: Strongly disagree, 3: Somewhat disagree, 4: Neutral, 5: Somewhat agree, 6: Strongly agree, 7: Very strongly agree



### Person Attitude Groups



Household Attitude Groups and Intra-household Distribution

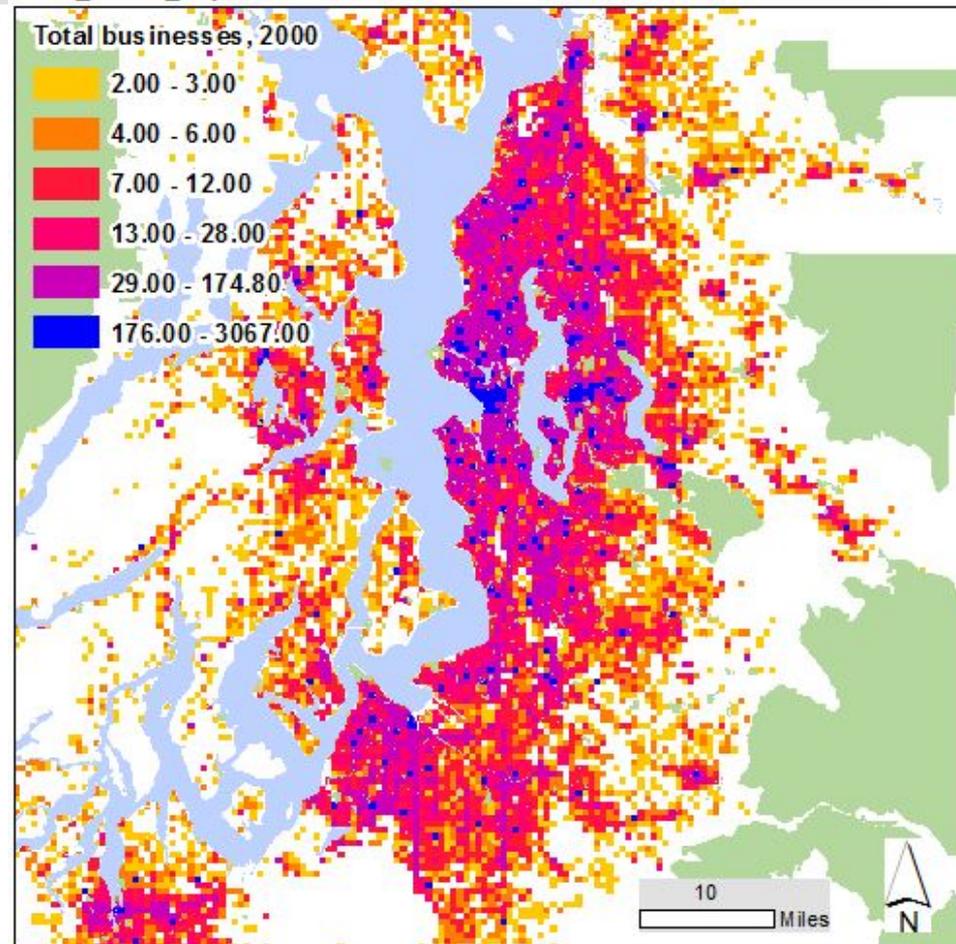
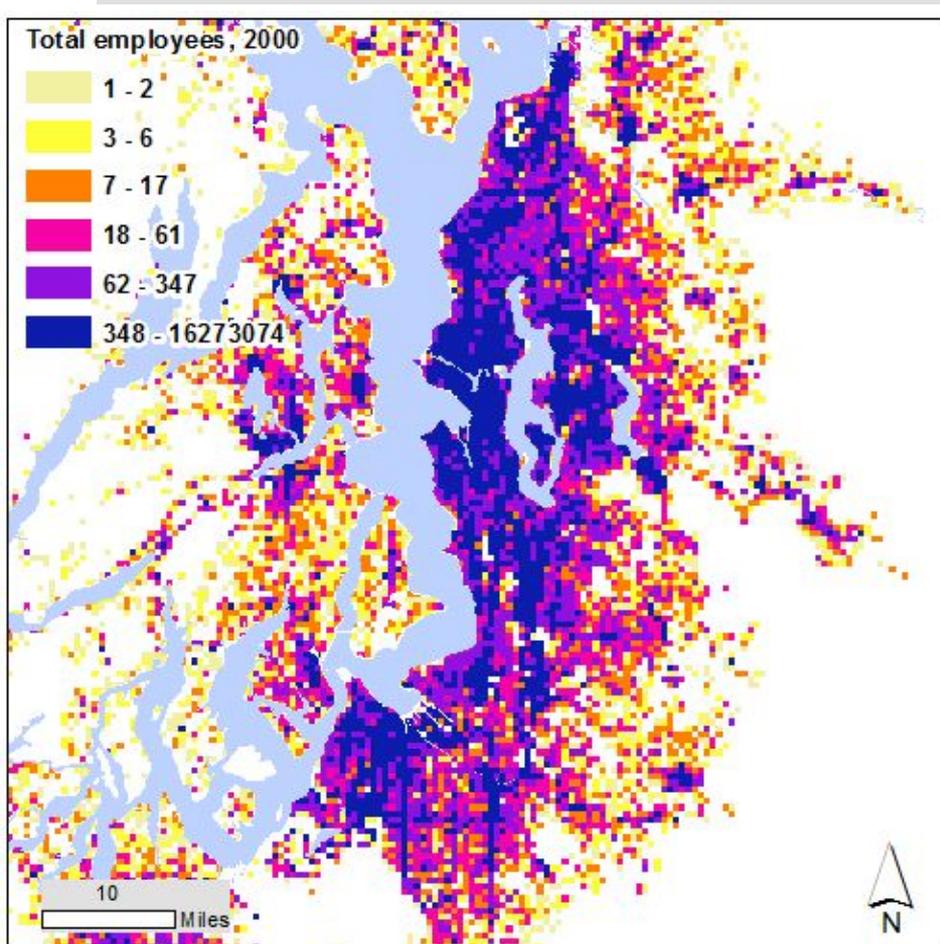
# Findings in Part 1 of Analysis

We find mix:

- Car lovers tend to live with car lovers or with neutrals
- Carpool/transit lovers tend to live with carpool/transit lovers, neutrals, but also car lovers
- Lack of strong homophily (reject the bird of a feather hypothesis)

# TRAVEL BEHAVIOR

# National Establishments Time Series



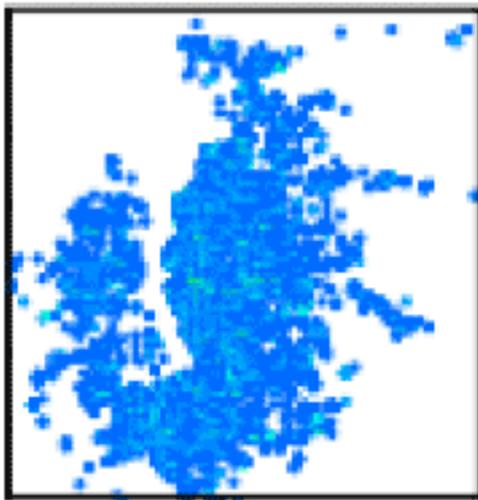
- National Establishments Time Series (1990- 2000+2002)
  - Geocoded establishments, aggregated to 750m<sup>2</sup> grid cell
  - Industry type, no. of employees, total sales

# Density and Diversity

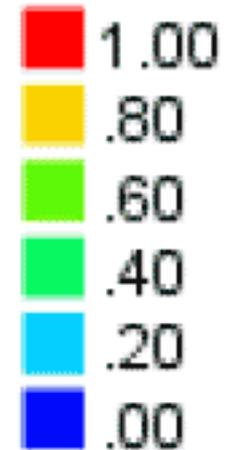
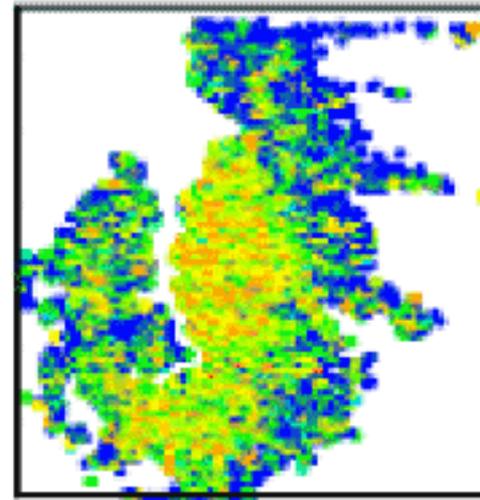
- Density (# of employees)

- Diversity (employees per industry type)

1990

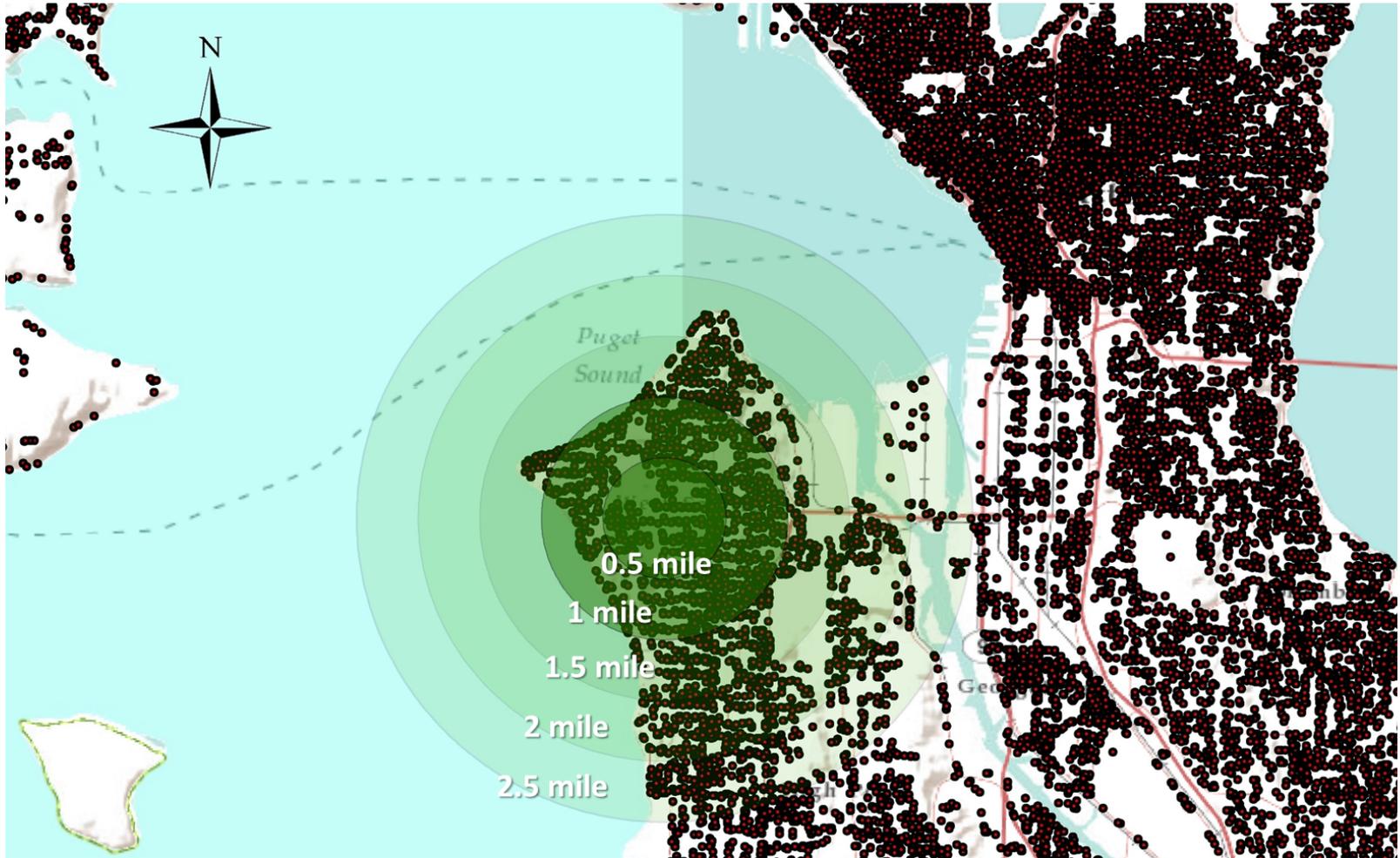


1990

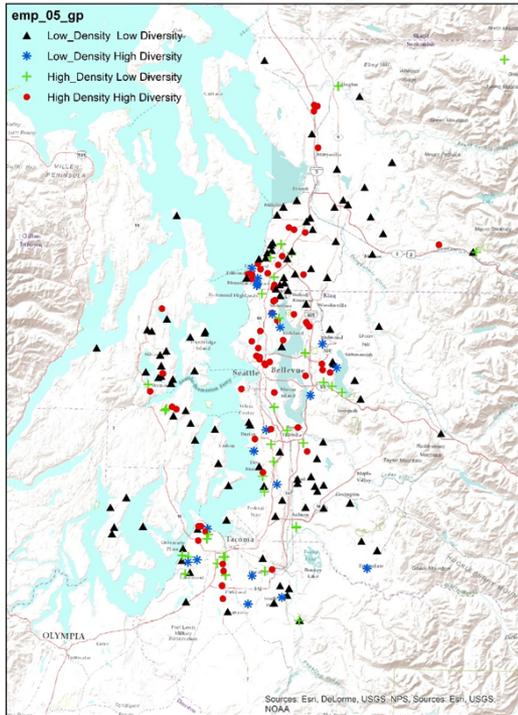


$$H' = - \sum_{i=1}^R p_i \ln p_i$$

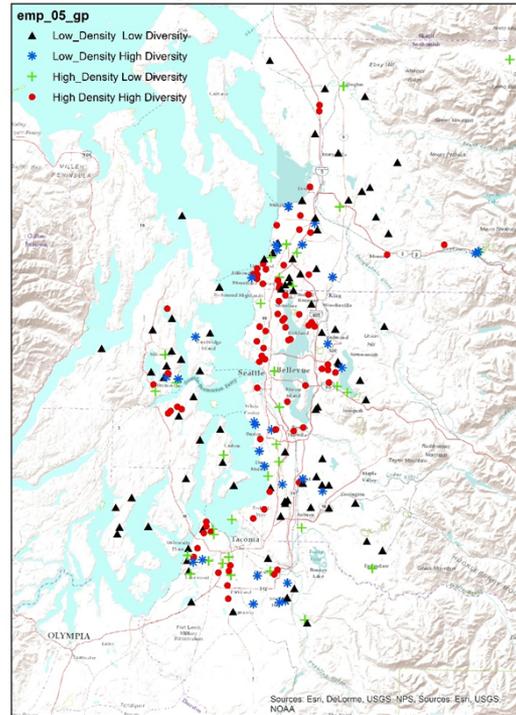
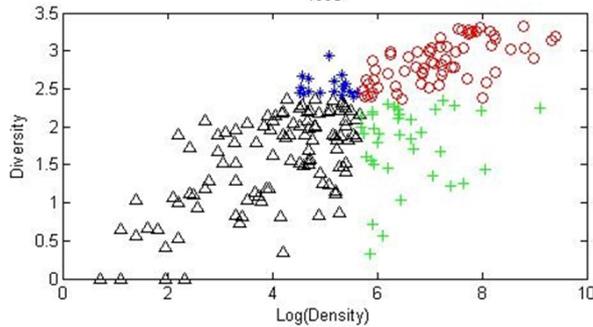
# Annuli Surrounding Residences



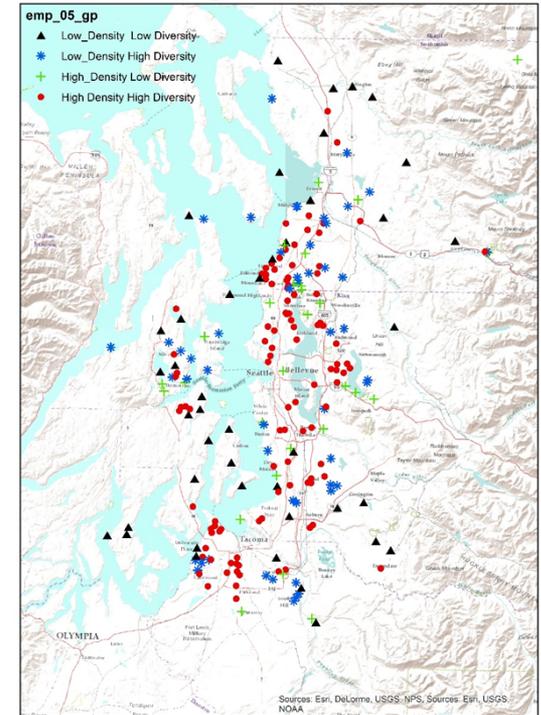
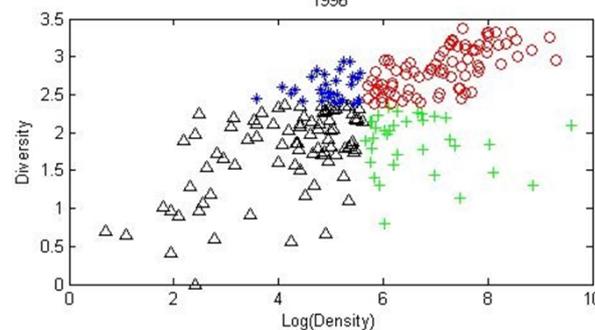
# Land use Characteristics Surrounding Home (<0.5 mile) in 1990, 1996, 2002



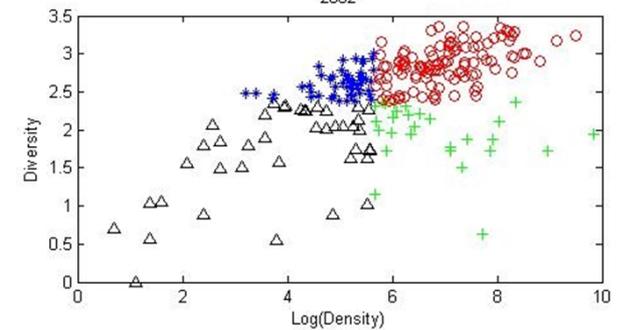
1990



1996

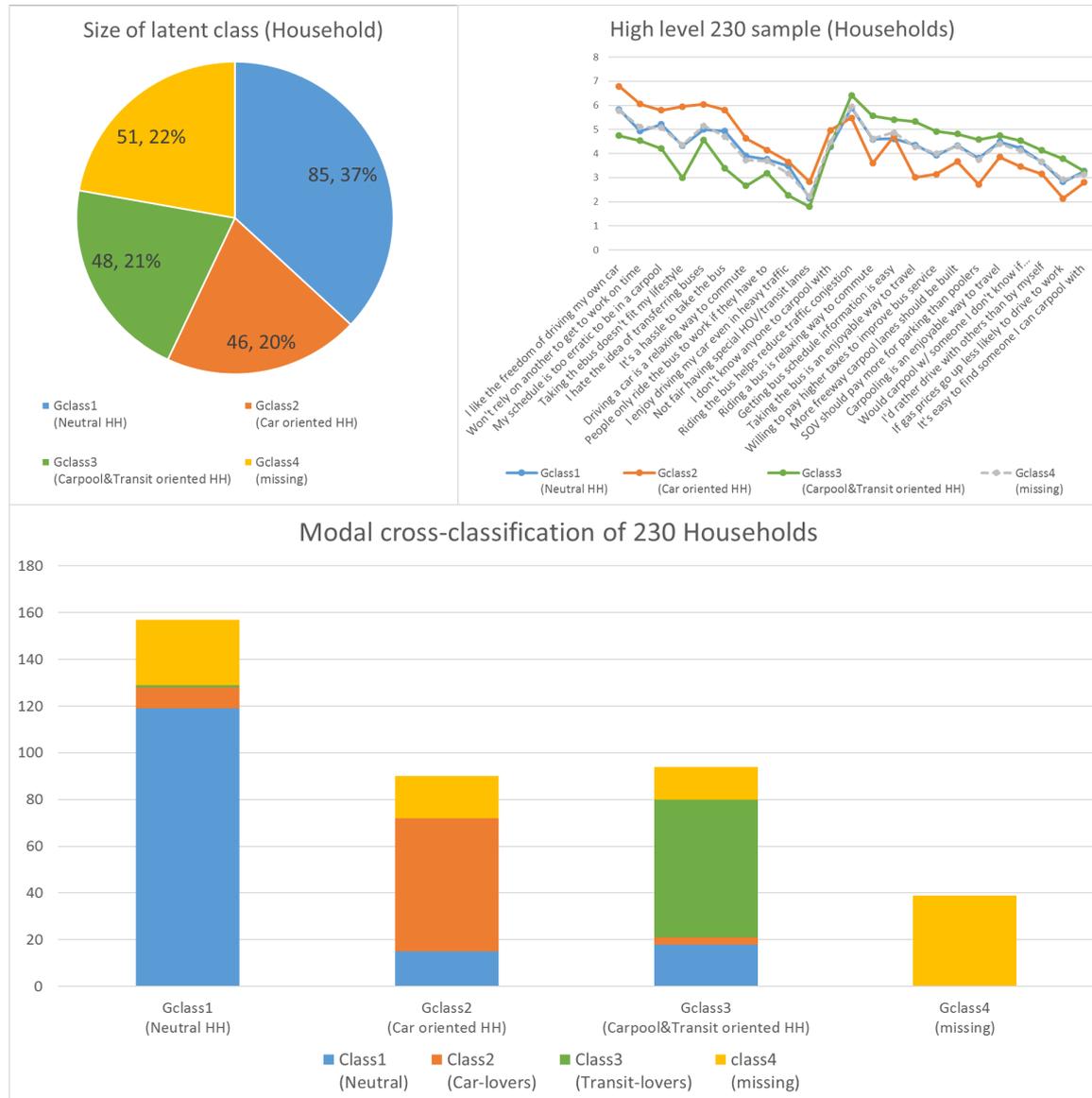


2002

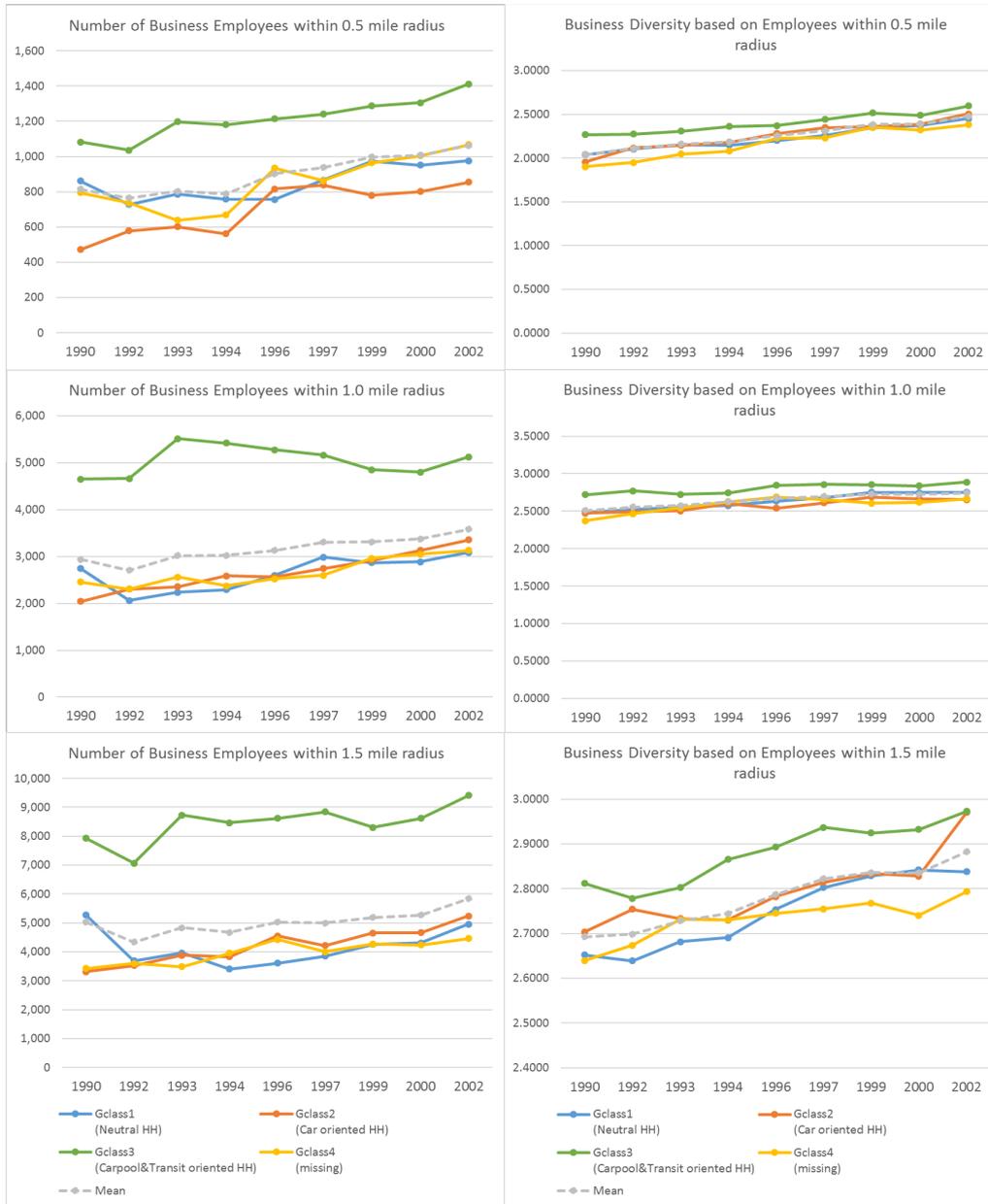


Use the attitude classification from  
part 1 for both persons and  
households just to verify if we can see  
differences in behavior

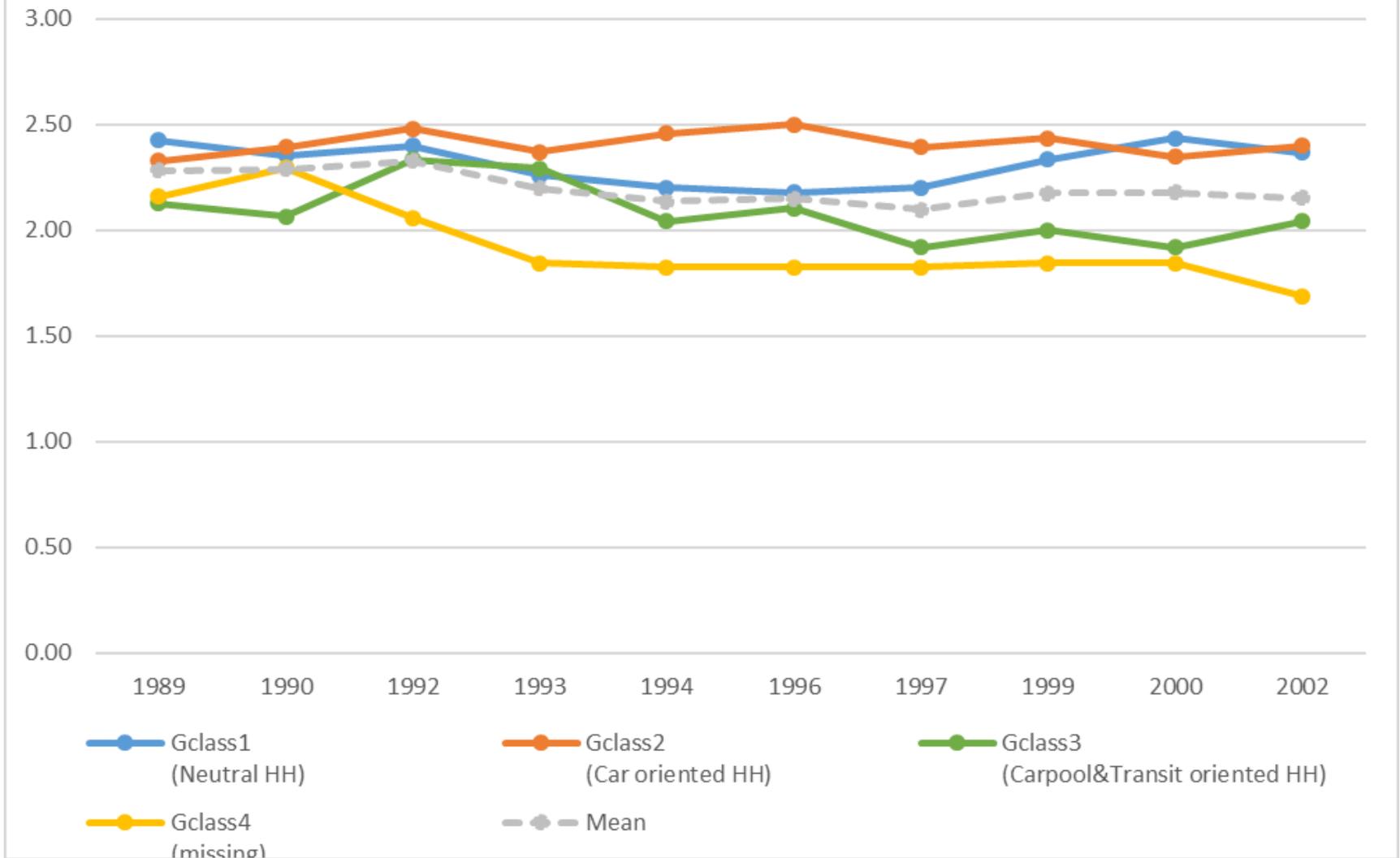
(my note: skip to 37 if running out of time)

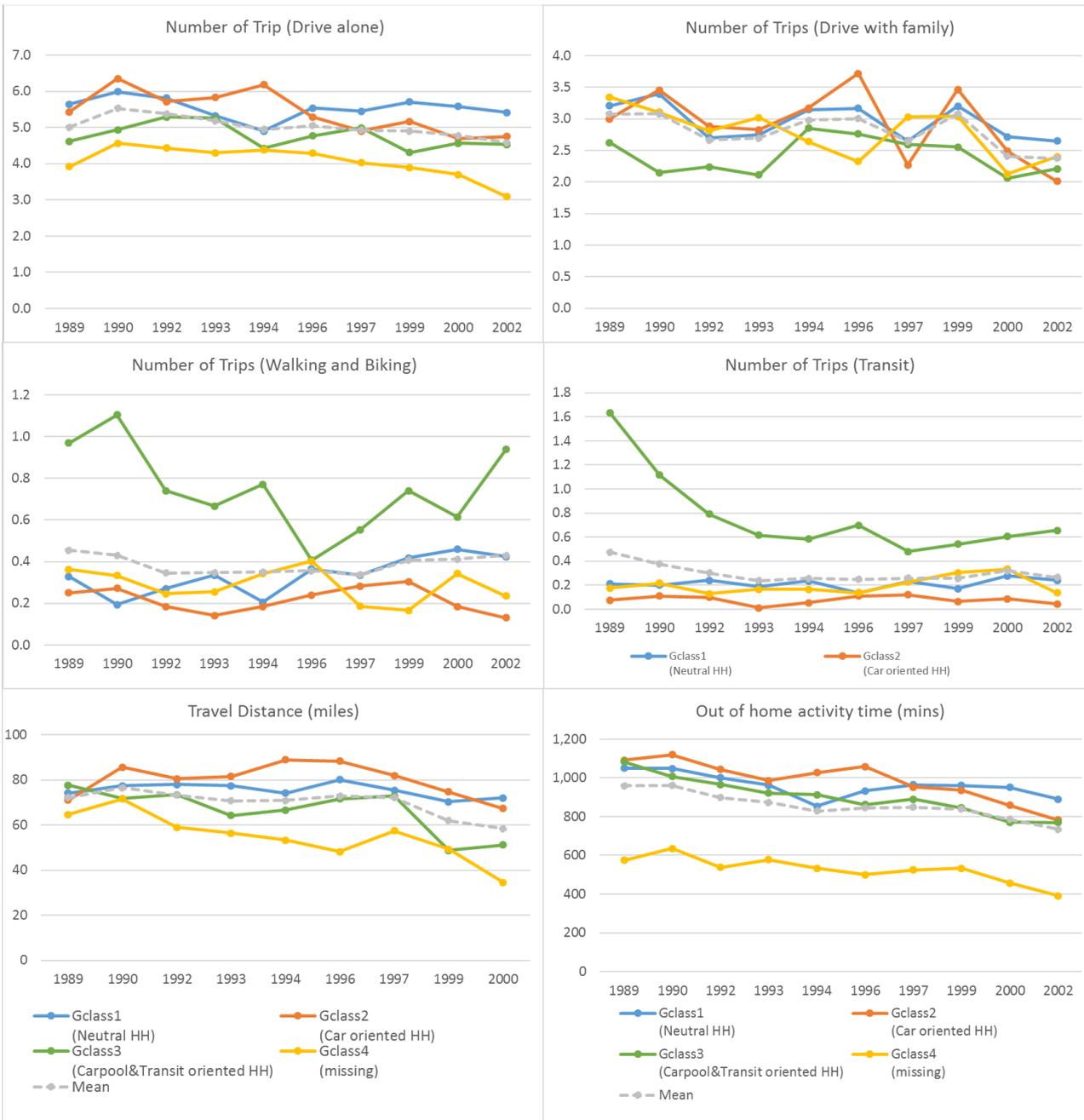


# What happens around their homes?

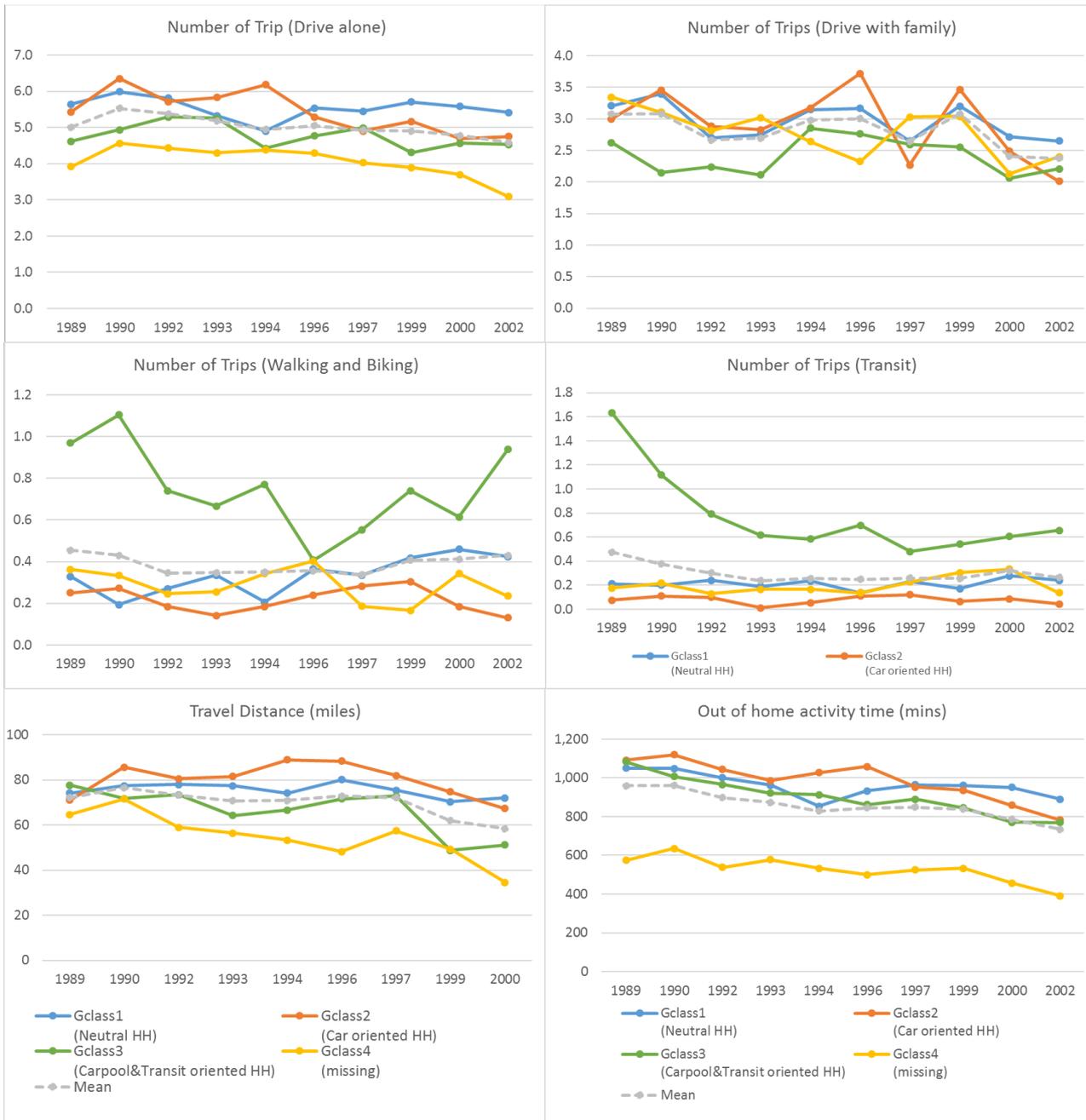


### Changes in Number of Vehicles





Missing attitudes people are least active (in surveys are also a group that does not participate)



Consistency of transit lovers and their behavior!

# **PART 2**

# The Multilevel Longitudinal Latent Class Model (time points within persons)

$$f(Y_k) = \sum_{h=1}^H P(v_k = h) \prod_{t=1}^{n_k} f(Y_{kt} | v_k = h)$$
$$f(Y_{kt} | v_k = h) = \sum_{l=1}^L P(x_{kt} = l | v_k = h) \prod_{q=1}^Q f(y_{ktq} | x_{kt} = l, v_k = h)$$

230 households at 10 time points

Groups (latent classes) based on *drive alone trips, car sharing trips, transit trips, bike and walk trips, number of cars, employment density around the household residence, and employment diversity around the household residence.*

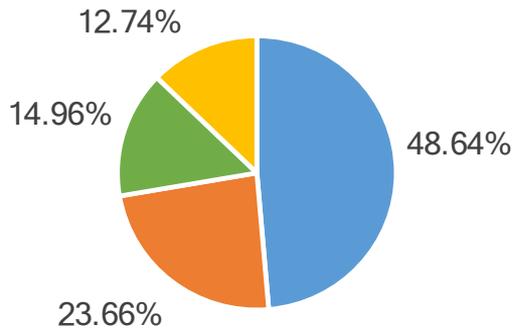
Seven variables ( $q=1, \dots, 7$ ) and

$k$  for the households ( $k=1, \dots, n_k$ ).  $n_k = 230$ .

$l$  for the behavioral variables,

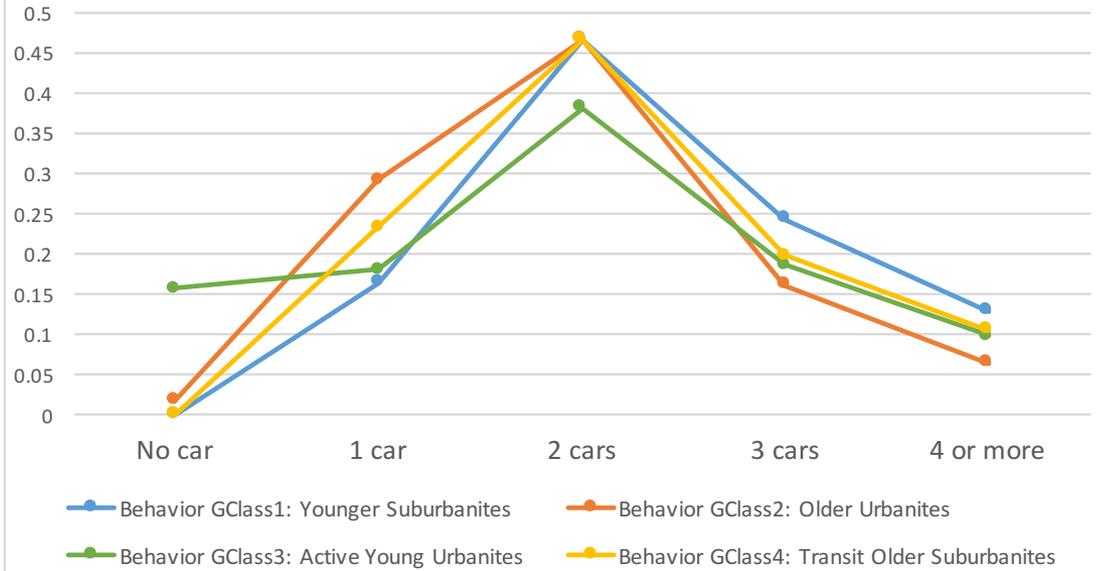
$t$  for the 10 time points (Equations 3 and 4). The index  $k$  goes from 1 to 230. The sum of all  $n_k$  is 2,300 households that provide answers in the 10 waves of the panel.

### Behavioral Gclass Size

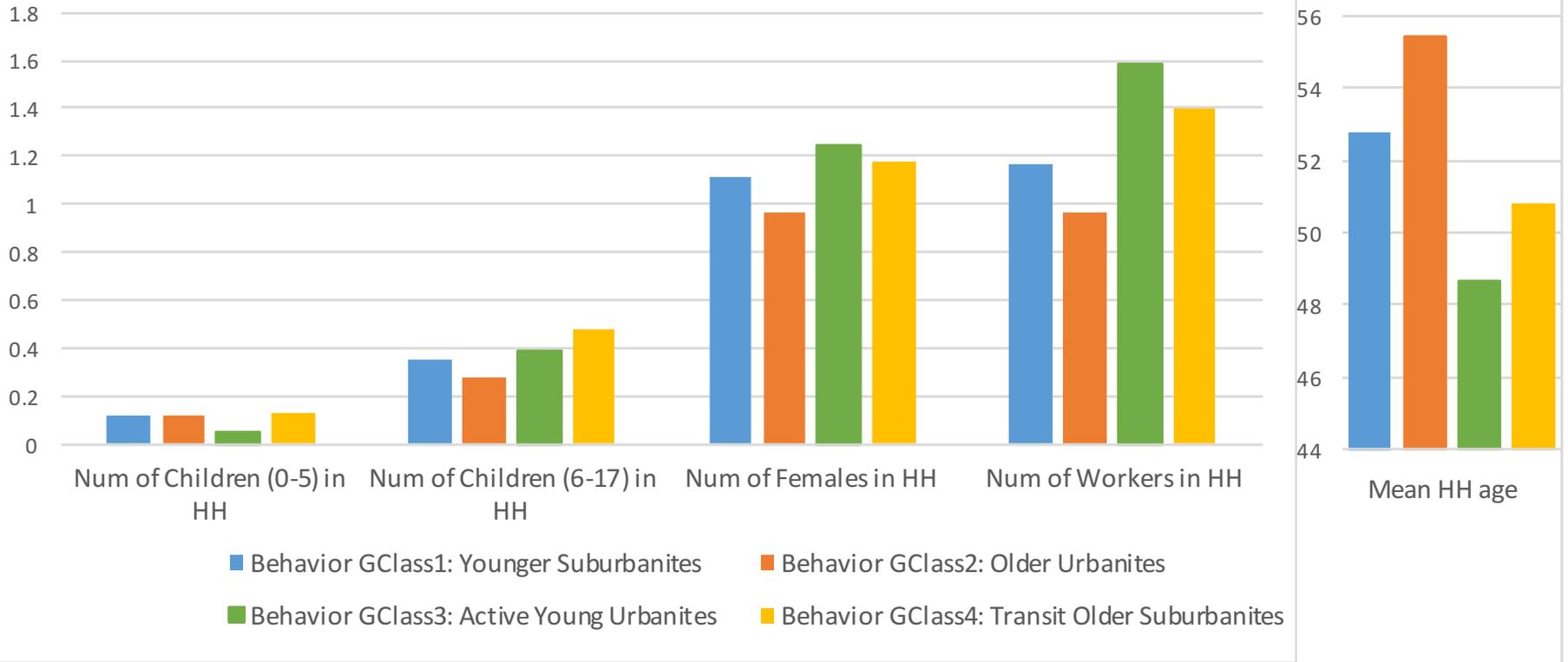


- Behavior GClass1: Younger Suburbanites
- Behavior GClass2: Older Urbanites
- Behavior GClass3: Active Young Urbanites
- Behavior GClass4: Transit Older Suburbanites

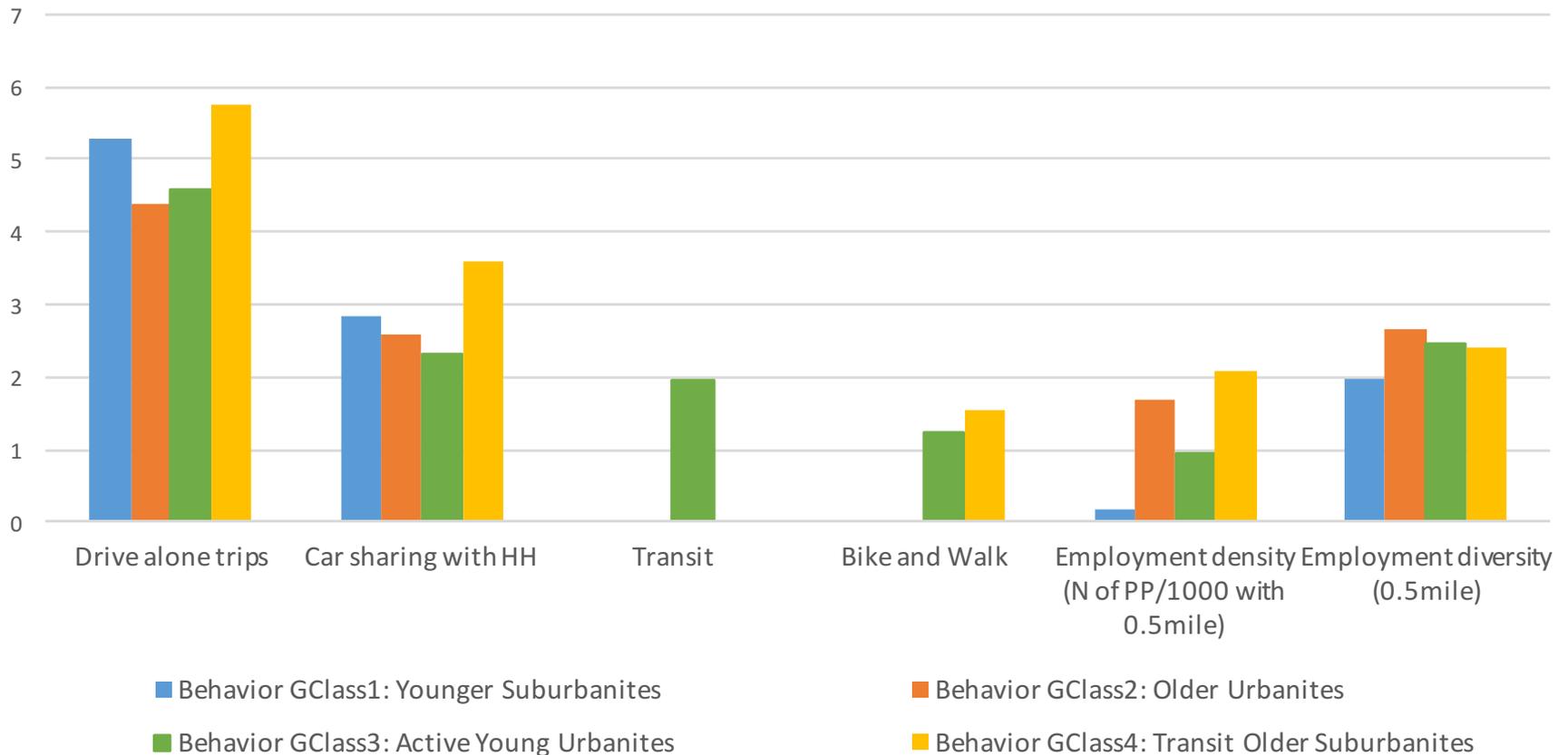
### Number of Cars in Each Behavioral Gclass



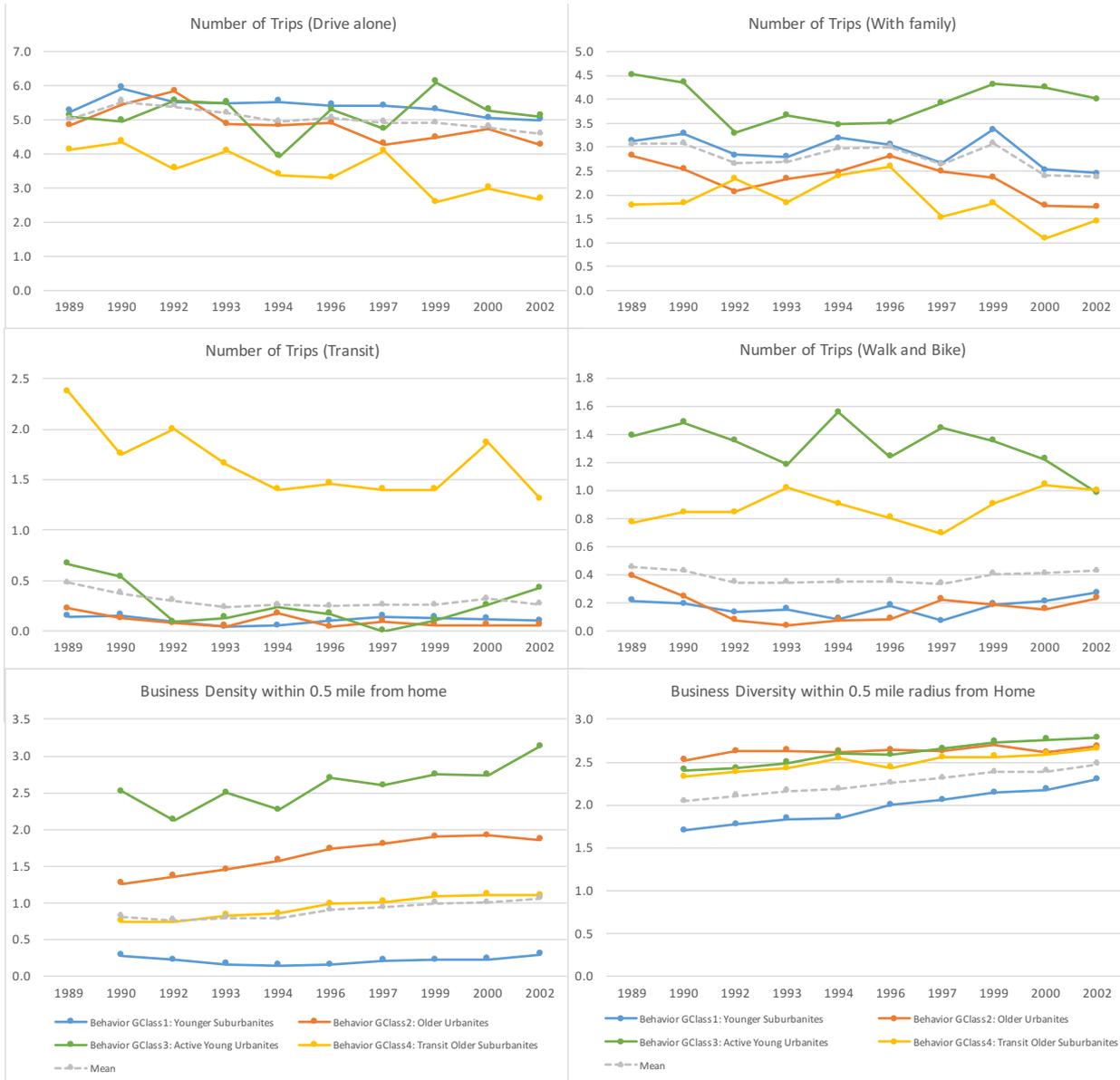
Average Demographics Across the Ten Waves



## Travel Behavior Indicators Across the Ten Waves

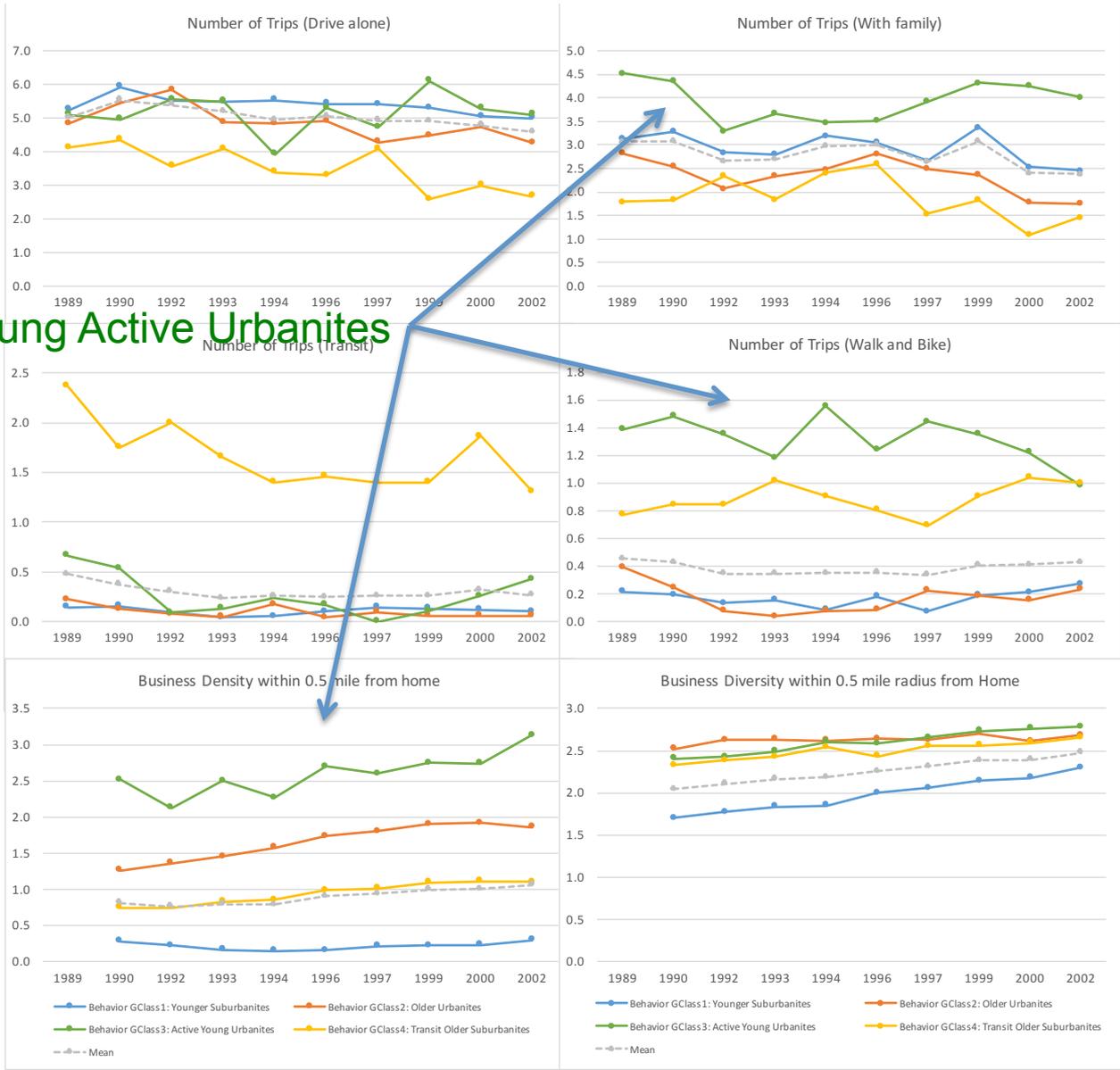


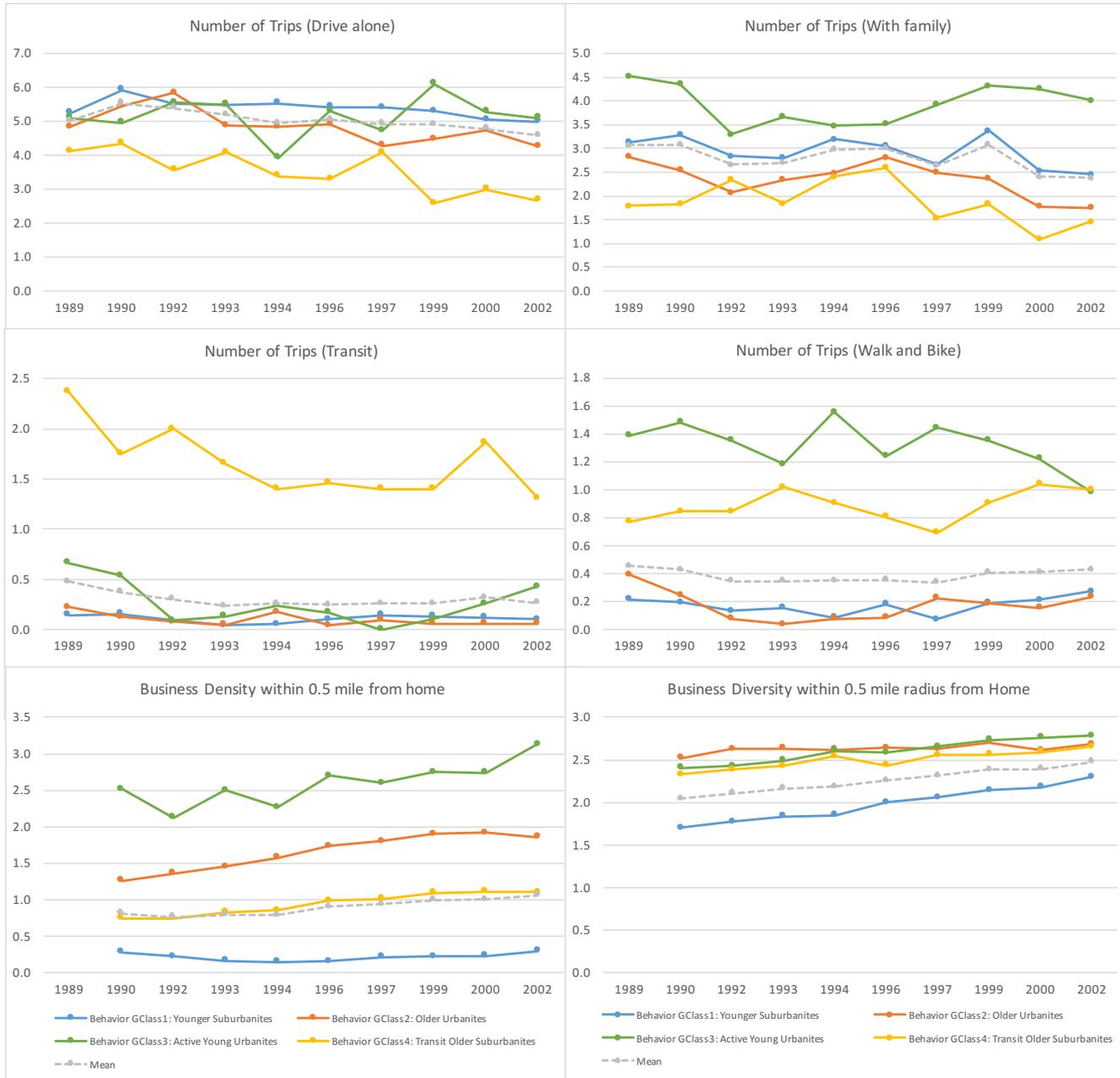
# Distinct groups behaviorally over time



# Distinct groups behaviorally over time

Green is Young Active Urbanites





**Younger suburban dwellers** more likely to be neutral (42%), car-oriented (23%) and with missing attitudes (22%).

The **transit older suburbanites** are car-pool and transit oriented households (65.4%). This is also the behavioral class with the least car driving alone and car sharing with relatives trips.

		Behavior GClass				Total	
		Younger Suburbanites (Blue)	Older Urbanites (Orange)	Active Young Urbanites (Green)	Transit Older Suburbanites (Yellow)		
<b>Chi square:</b> 44.874, df:9, Contingency Coefficient = 0.404							
Attitude GClass	Neutral	Count	53	17	9	6	85
		% within Attitude GClass	62.4%	20.0%	10.6%	7.1%	100.0%
		% within Behavior GClass	42.1%	33.3%	33.3%	23.1%	37.0%
	Car-oriented HH	Count	29	13	4	0	46
		% within Attitude GClass	63.0%	28.3%	8.7%	0.0%	100.0%
		% within Behavior GClass	23.0%	25.5%	14.8%	0.0%	20.0%
	Carpooling& Transit-oriented HH	Count	16	6	9	17	48
		% within Attitude GClass	33.3%	12.5%	18.8%	35.4%	100.0%
		% within Behavior GClass	12.7%	11.8%	33.3%	65.4%	20.9%
	Missing	Count	28	15	5	3	51
		% within Attitude GClass	54.9%	29.4%	9.8%	5.9%	100.0%
		% within Behavior GClass	22.2%	29.4%	18.5%	11.5%	22.2%
Total	Count	126	51	27	26	230	
	% within Attitude GClass	54.8%	22.2%	11.7%	11.3%	100.0%	
	% within Behavior GClass	100.0%	100.0%	100.0%	100.0%	100.0%	

**Active young urbanites** and suburbanites have a substantial presence of neutral attitudes households (about 33% in both) but the

**Young active suburbanites** has 33% of their households displaying positive attitudes towards carpooling and transit. These are also the households with the highest number of car sharing with family trips.

# Disagreement in terms of attitudes within a household does not equate to different behavior than the households displaying the same within household attitudes

**Chi square:** 0.118, df:3.  
**Contingency Coefficient =** 0.023

		Behavior GClass				Total	
		Younger Suburbanites (Blue)	Older Urbanites (Orange)	Active Young Urbanites (Green)	Transit Older Suburbanites (Yellow)		
Homophily	All others (missing, disagree)	Count	98	40	21	21	180
		% within Homophily	54.4%	22.2%	11.7%	11.7%	100.0%
		% within Behavior GClass	77.8%	78.4%	77.8%	80.8%	78.3%
	Same Attitude in Household either 1,2,3	Count	28	11	6	5	50
		% within Homophily	56.0%	22.0%	12.0%	10.0%	100.0%
		% within Behavior GClass	22.2%	21.6%	22.2%	19.2%	21.7%
Total		Count	126	51	27	26	230
		% within Homophily	54.8%	22.2%	11.7%	11.3%	100.0%
		% within Behavior GClass	100.0%	100.0%	100.0%	100.0%	100.0%

# Specific Questions of the Paper

- Do people with different attitudes live together?
- **Yes but lack symmetry**
- Do attitudes early in a person's life persist in its effect in later years?
- **Yes**
- Can we identify distinct attitudinal (x-section) and behavioral (over time) groups and study systematic differences among them?
- **Yes but within hh differences don't matter**

# Related Papers

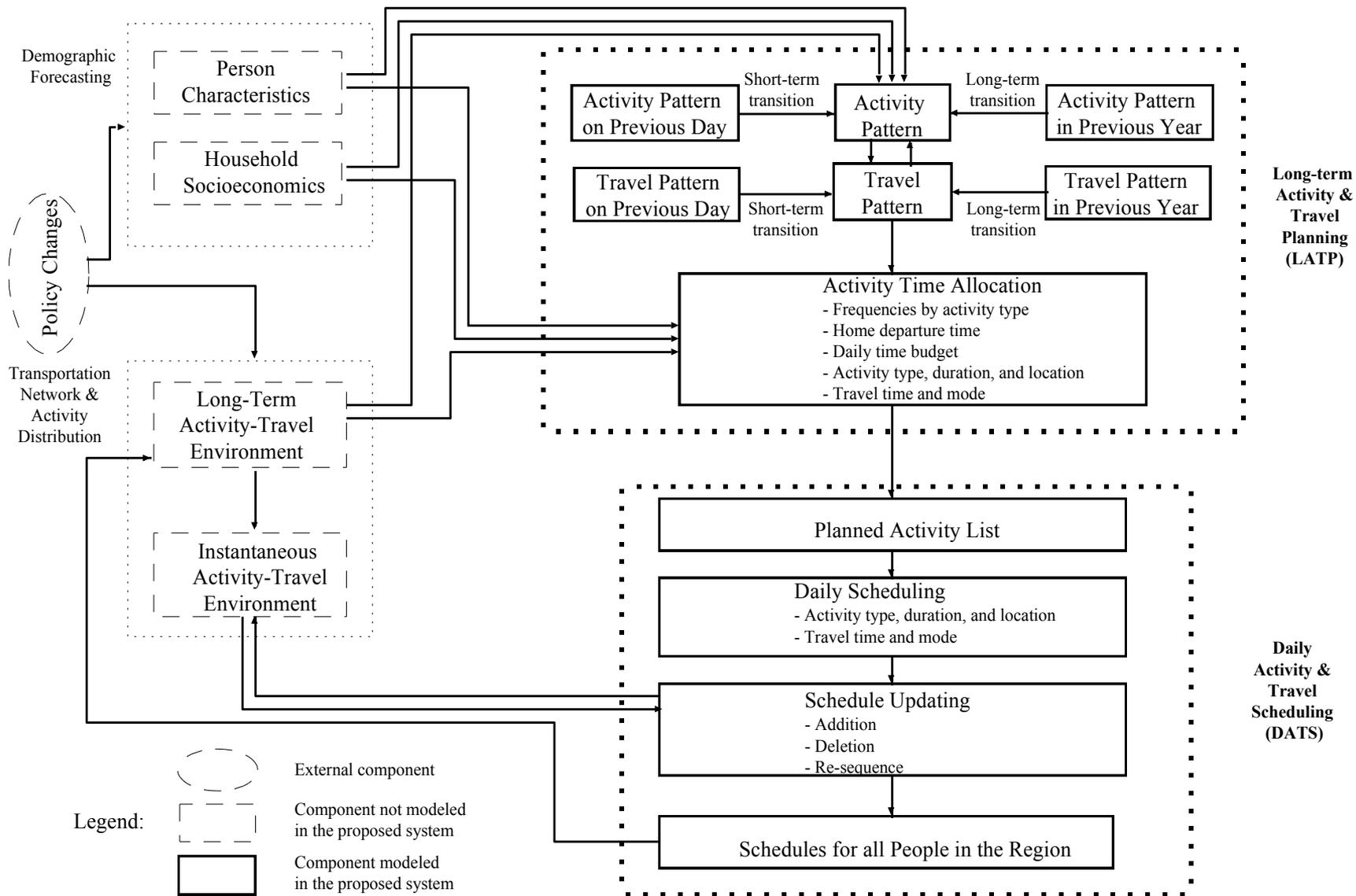
- Lee J. H. and K.G. Goulias (2017) A Decade of Dynamics of Residential Location, Car Ownership, Activity, Travel and Land Use in the Seattle Metropolitan Region. Paper to be presented at the 22<sup>nd</sup> International Symposium on Transportation and Traffic Theory, July 2017, Northwestern University, Evanston, IL.
- McBride E., J.H. Lee, A. Lundberg, A.W. Davis. and K. G. Goulias, (2016). Behavioral micro-dynamics of car ownership and travel in the Seattle metropolitan region from 1989 to 2002. *European Journal of Transport and Infrastructure Research*, 16(4), 735-753.
- Lee, J. H., Davis, A. W., & Goulias, K. G., 2015. Triggers of behavioral change. Paper presented at the International Choice Modelling Conference 2015, Austin, May 10-13. GEOTRANS REPORT 2015-05-01.
- Goulias K. G., J.H. Lee, and A.W. Davis (2015) Longitudinal Mixed Markov Latent Class Analysis of the 1989 to 2002 Puget Sound Transportation Panel Data. Paper presentation at the *94th Annual Meeting of the Transportation Research Board, Washington, D.C.*, January 11-15, 2015. Also published as GEOTRANS Report 2014-8-04, Santa Barbara, CA.



- Many Thanks  
&
- Questions?

# **EXTRA NOTES**

# The June Ma Model



# **BEHAVIORAL FRAMEWORK**

# Principles of Behavioral Dynamics

- **The household and its members in their surrounding environment are in a continuous process of adjustment**
- Examples:
  - Life cycle and turning points in life
  - Day to day scheduling of activities
  - Time allocation among household members and others in the social network (social capital)
  - Activity opportunities and travel time

# Principles of Behavioral Dynamics

- **Processes of Adjustment with Lags and Leads**
  - Information awareness and use = short and long lags depending on context
  - Searching for alternatives = length of search is function of behavioral style (optimizing vs satisficing)
  - Experimentation = effort to settle and transaction costs
  - Capability constraints = purchasing and time budgets
  - Institutional constraints = external and internal to household obligations

# What are Some Triggers?

- *Physiological alterations* (e.g., hormonal changes that alter physical and social selves)
- *Transitions* (e.g., *age-graded movement* into and out of social roles such as school grades or loss of a parent)
- *Turning points* (e.g., events that cause reorientation of priorities and lasting alterations of a person's developmental trajectory)

# What Happens?

- Triggers create barriers or offer new opportunities.
- Lead to changes in roles, self-concepts, lifestyles, worldviews, and dispositions towards other people.
- Impact behavior depending on their timing and duration, the socio-economic characteristics of the individual such as gender and sexual orientation, and ethnic and social class.

# Examples of Triggers

- Marriage
- divorce
- building a family and birth of children
- entering a new intimate relationship
- separation
- entering school
- choosing occupation
- engaging in nonoccupational studies
- graduation
- continuing studies
- dropping out of school
- job seeking
- job loss
- retirement
- starting first job
- starting private enterprise/practice
- Change is constant!
- declaring bankruptcy
- moving to another community
- leaving home
- travelling somewhere far away
- moving temporarily to another place
- entering military or community service
- loss due to death of close member
- getting a new apartment
- getting a vacation home
- change in leisure activities and hobbies
- drug use and abuse
- committing crime(s)
- religious engagement
- psychological crises
- own illness
- illness of close member
- accidents

# Principles of Behavioral Dynamics

- **Heterogeneity\***

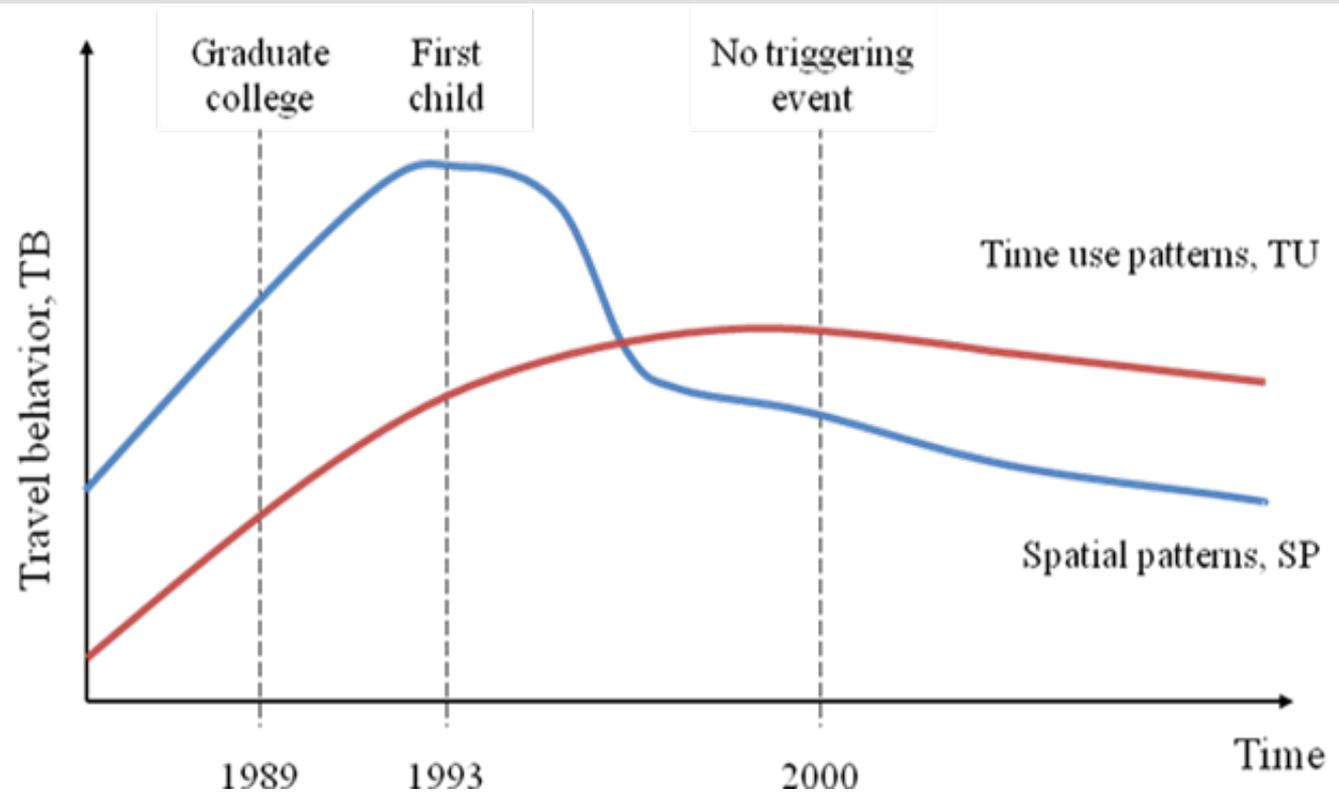
- Differences of lifestyles & roles (observed, unobserved, and unobservable)
- Differences in types of relationships (observed, unobserved, and unobservable)
- Differences in sensitivity to changes in the environment surrounding the household (observed, unobserved, and unobservable)

\*Note the unobserved heterogeneity effects on illusions of state dependence/habit

# Decision Making in Context

- Inertia and “stickiness” of habits = after searching and experimentation households avoid transaction costs
- Thresholds = decisions to change behavioral pattern based on a cost-benefit analysis of breaking habits
- Satisficing = habitual behavior may be at a lower point than optimum
- Properties:
  - Partial (gradual) adjustments possibly observed during search and experimentation
  - Speed of adjustment depends on the decision and context
  - Path dependency in adjustments that depends on knowledge
  - Asymmetry and hysteresis of change (path to increase different than path to decrease – e.g., car ownership and job loss)

# Possible Evolution



- Associate changes in behavior to **lifecycle** turning points
  - Shift in social roles
  - Shift in long-term plans

# (Longitudinal)Data to Test Dynamic Hypotheses

- Panel Surveys\*
  - Life cycle changes
  - Long vs short term decisions
  - Leads and lags
- Before After Experiments
  - Impose a change
  - Ask questions tailored to decision and context
- Direct tracking
  - GPS family
  - Combinations with panels and experiments

\*Note the long research record on panel survey design and countering of panel-specific issues (non-response, attrition, fatigue, conditioning, incomplete change observed)

# Data We Use (background)

Approach	Design	No. of Measurements per Sample Unit	Type of Variation Measured		
			Among Sample Units	Within Sample Units	In the Population Over time
Cross-Sectional	One time	One	Yes	No	No
	Repeated	One	Yes	No	Yes
Panel (Longitudinal)	Panel (PSTP)	Two or More	Yes	Yes	Usually No
	Rotating Panel	Two or more	Yes	Yes	Yes

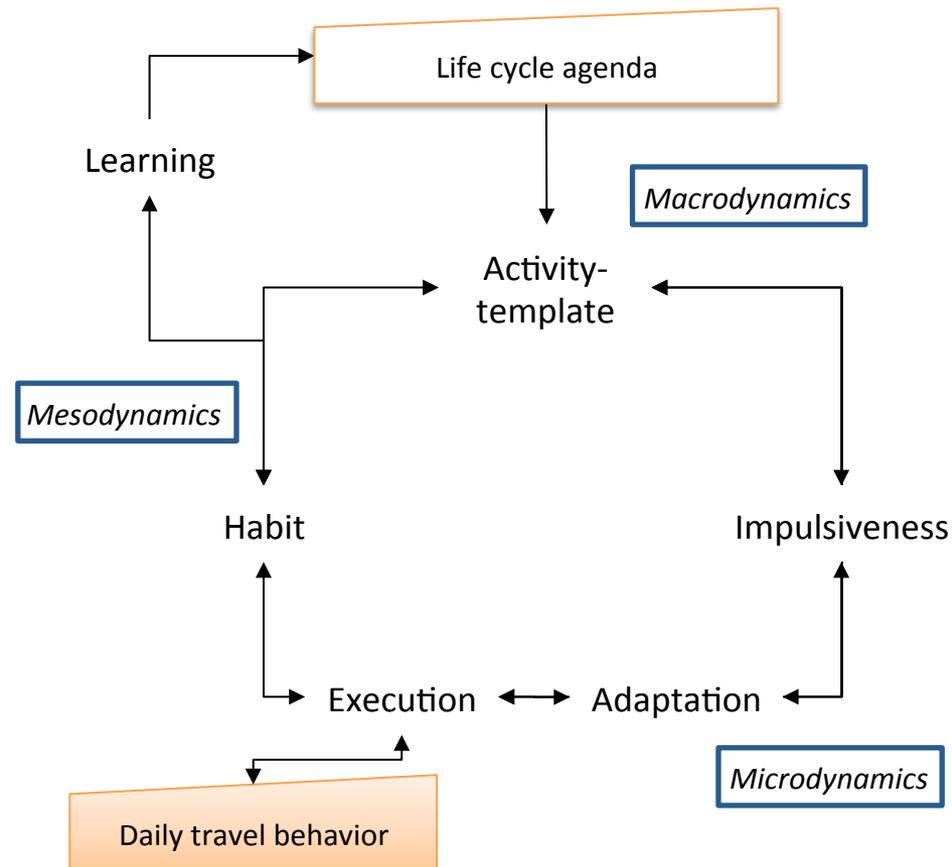
# Analytical Tools

- Stochastic Processes
  - Renewal processes
  - Markov renewal processes
  - Markov processes
  - Markov chains
- Discrete time regression models
  - Linear
  - Distributed lags
  - Lagged dependent variables
  - Non-linear models
  - Dynamic models

Note: Structural equations models and computational process models could also be included here – not in RKs review but used by his group and collaborations <sup>63</sup>

# Travel behavior dynamics (Pam Dalal Dissertation)

- Behavioral dynamics in activity scheduling
- *Macrodynamics*
  - moving through life cycle stages
- *Mesodynamics*
  - lags and leads in behavioral change
- *Microdynamics*
  - planning for a day



# Life course theory

- Over a life course, how does behavioral change manifest over **time**
- Life course theory
  - Stability and change in long-term behavior
- Drivers for behavioral change
  - Turning points
    - Triggers (death, planned retirement)
- Process of behavioral change
  - Habit (lag)
  - Socialization (lead)

