

# Measuring Unobservables in Behavioral Models

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# Outline

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- Research agenda
- Methodology overview
- Applications
  1. Residential location
  2. Airline itineraries
  3. Environmentalism
- Conclusion

# Research Agenda

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- Incorporate latent variables and psychometric measurements into discrete choice models
  - Quantify attitudes, motivation, lifestyle, etc.
  - “Behavioral mixing”
- Approach: Integrate discrete choice models and latent variable models

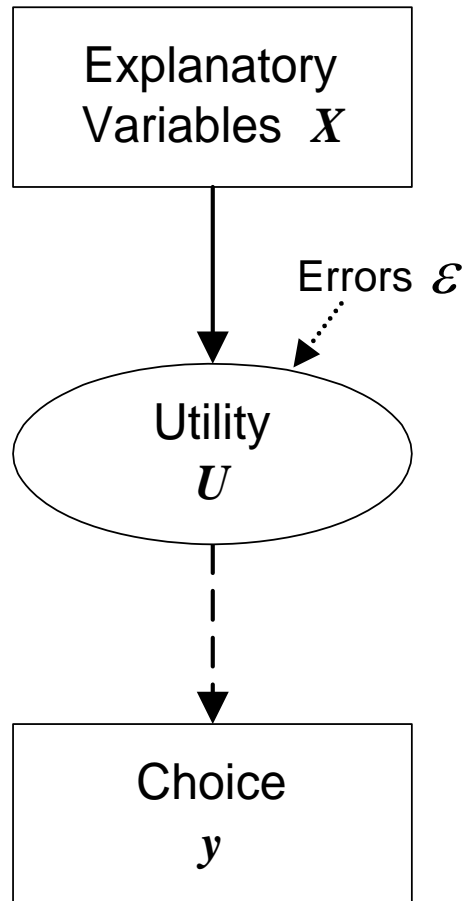
# This presentation

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- Overview
- 3 Applications
  - Demonstrate that behavioral mixing produces more intuitively appealing models
  - Leading to improved prediction and policy analysis

# Discrete Choice Model

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Random Utility

$$U = \beta' X + \varepsilon$$

*structural  
equation*

Choice Indicator  
(*Utility Maximization*)

$$y = f(U)$$

*measurement  
equation*

# Choice Probability

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## □ Distribution of $\varepsilon$

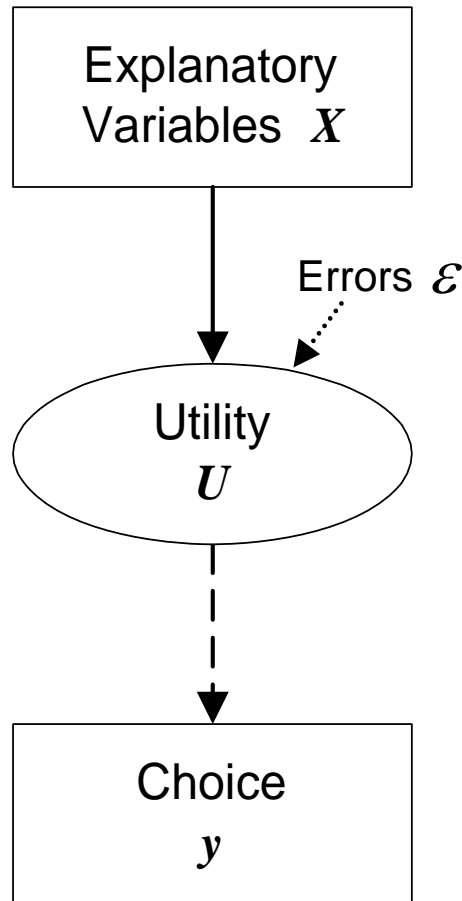
- Logit (iid Extreme Value)

$$P(i | X) = \frac{e^{\beta' X_i}}{\sum_{j \in C} e^{\beta' X_j}}$$

- Probit (Normal)
- Mixtures (Random  $\beta$ )

# Discrete Choice Model

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Random Utility

$$U = \beta' X + \varepsilon$$

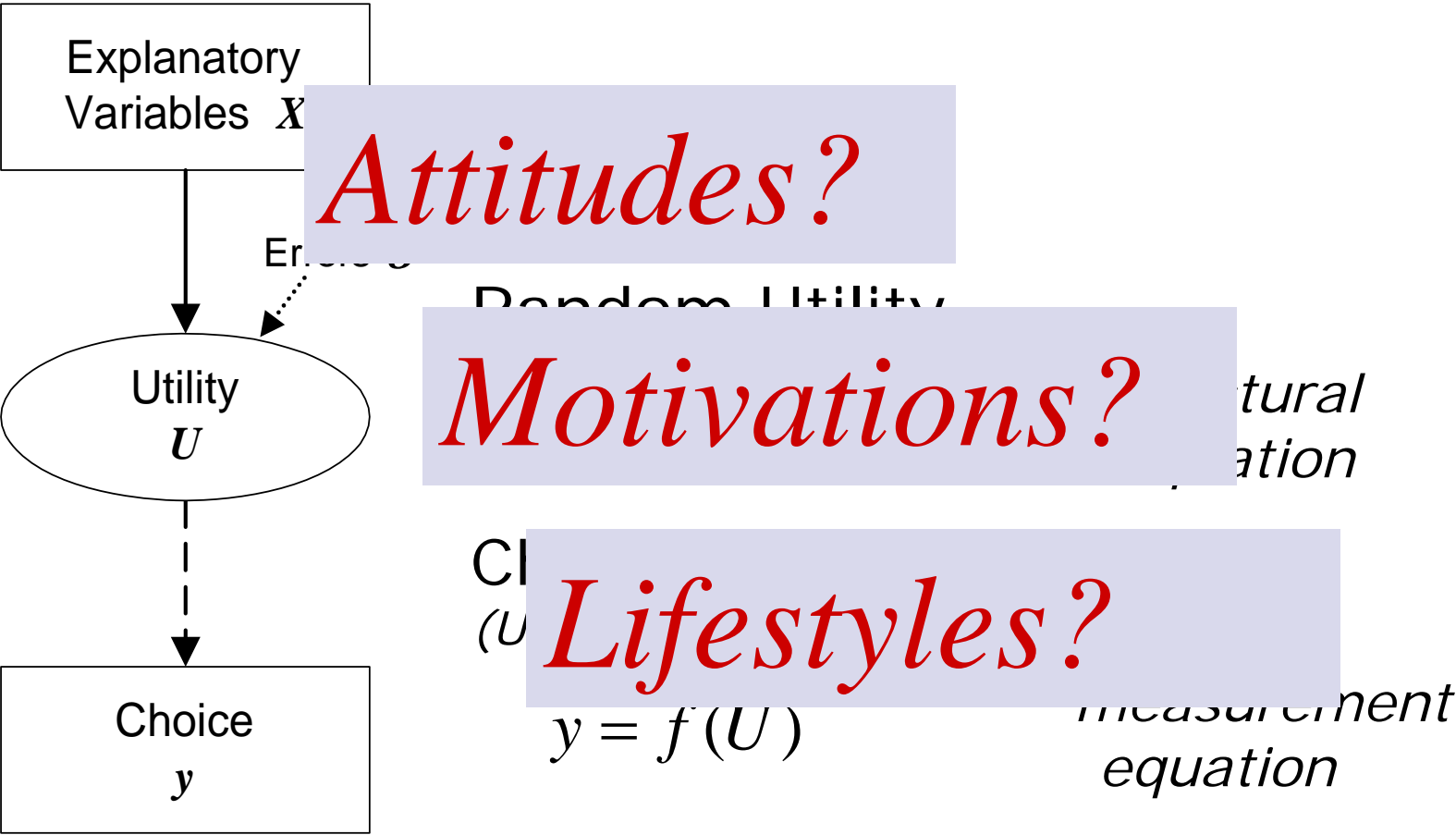
*structural  
equation*

Choice Indicator  
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$$y = f(U)$$

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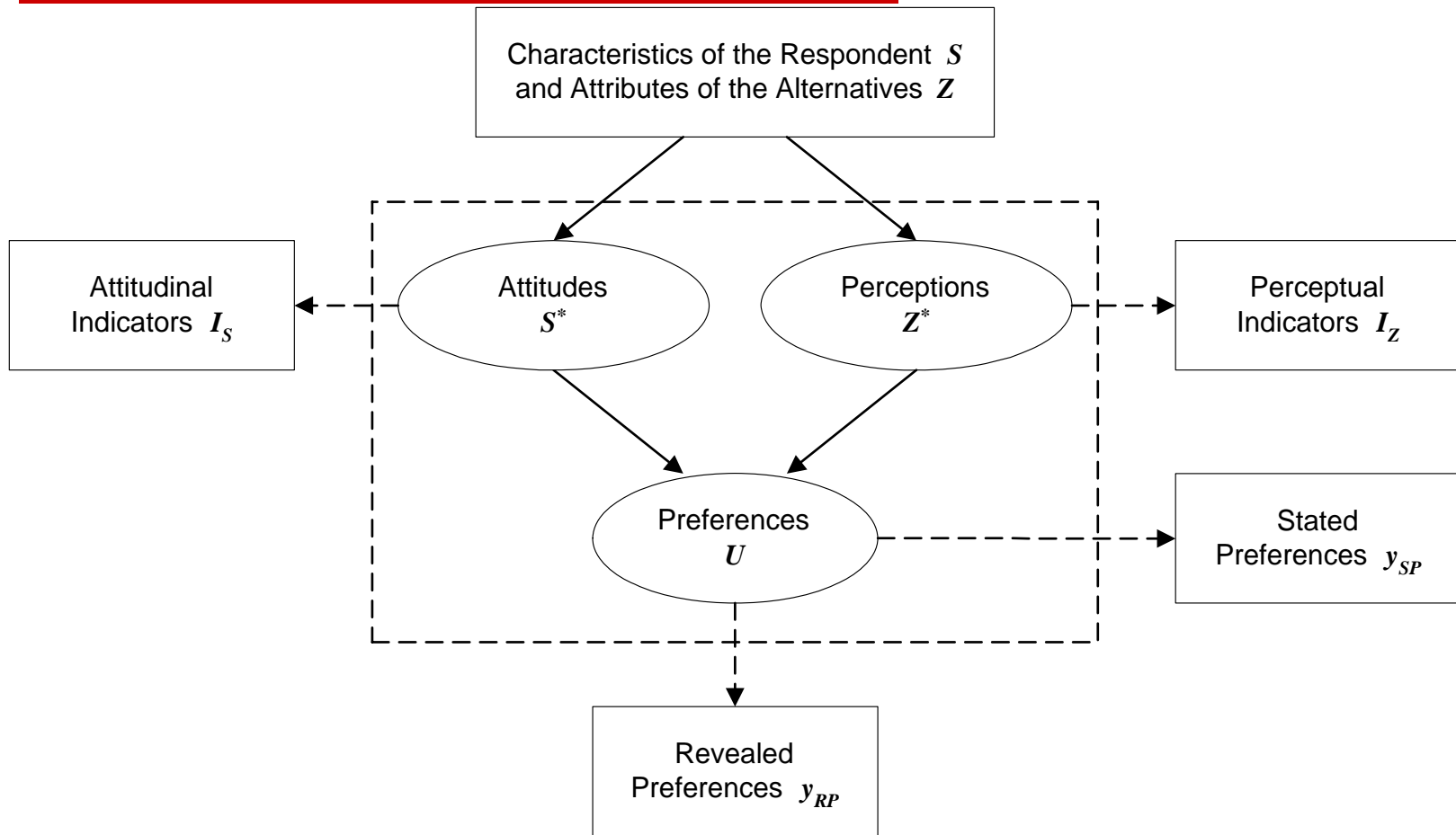
# Discrete Choice Model





# Opening the Black Box

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# APPLICATION I

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## Residential Location & Lifestyle Segmentation (with Li)



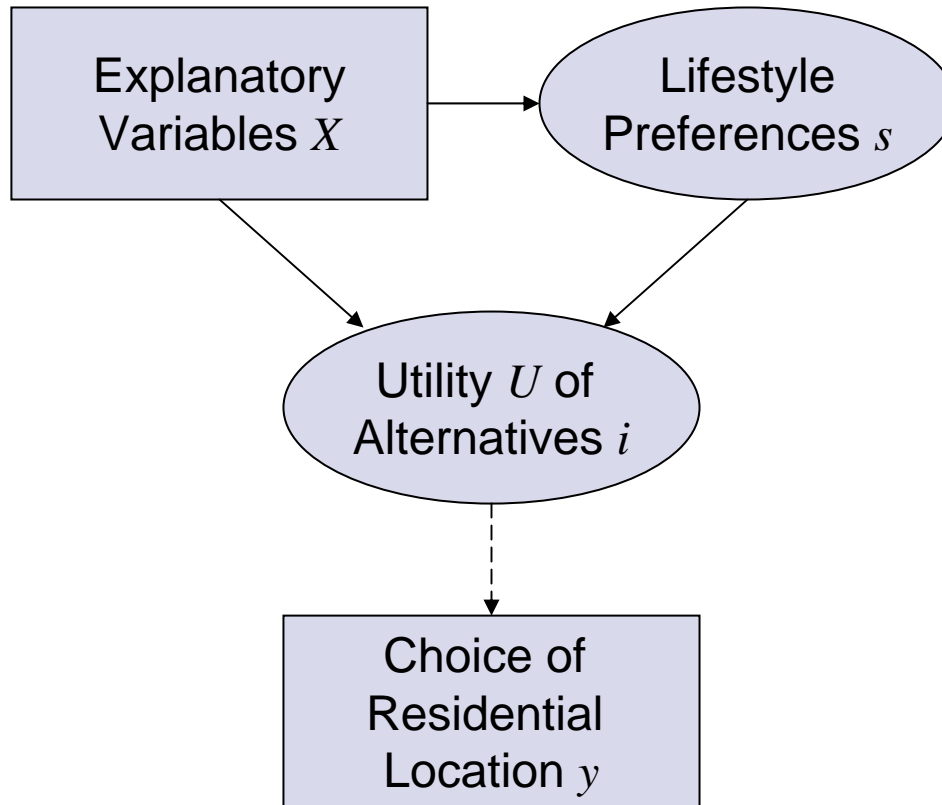
# Approach

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- ❑ Objective: Introduce “lifestyle” in models
- ❑ Hypothesis
  - ❑ Lifestyle preferences exist
  - ❑ Lifestyle differences lead to differences in considerations, criterion, and preferences for residential location choices.
- ❑ Infer “lifestyle” preferences from choice behavior using latent class choice model.
  - ❑ Latent classes = lifestyle
  - ❑ Choice model = location decisions

# Residential Location Choice Model with Latent Lifestyle Segmentation

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# Choice Experiment Example

(Alternative 1)

(Alternative 2)

(Alternative 3)

(Alternative 4)

(Alt. 5)

	Buy Single Family	Buy Multi-Family	Rent Single Family	Rent Multi-Family	
Type of Dwelling :	<i>single house</i>	<i>apartment</i>	<i>duplex / row house</i>	<i>condominium</i>	Move out of the Metro Area
Residence Size :	<i>&lt; 1,000 sq. ft.</i>	<i>500-1,000 sq. ft.</i>	<i>1,500 - 2,000 sq. ft.</i>	<i>&lt; 500 sq. ft.</i>	
Lot Size :	<i>&lt; 5,000 sq. ft.</i>	<i>n/a</i>	<i>5,000 - 7,500 sq. ft.</i>	<i>n/a</i>	
Parking :	<i>street parking only</i>	<i>street parking only</i>	<i>driveway, no garage</i>	<i>reserved, uncovered</i>	
Price or Monthly Rents :	<i>&lt; \$75K</i>	<i>\$50K - \$100K</i>	<i>&gt; \$1,200</i>	<i>\$300 - \$600</i>	
Community Type :	<i>mixed use</i>	<i>mixed use</i>	<i>rural</i>	<i>urban</i>	
Housing Mix :	<i>mostly single family</i>	<i>mostly multi-family</i>	<i>mostly multi-family</i>	<i>mostly multi-family</i>	
Age of Development :	<i>10-15 years</i>	<i>0-5 years</i>	<i>10-15 years</i>	<i>0 - 5 years</i>	
Mix of Residential Ownership :	<i>mostly own</i>	<i>mostly own</i>	<i>mostly rent</i>	<i>mostly own</i>	
Shops/Services/Entertainment :	<i>community square</i>	<i>basic shops</i>	<i>community square</i>	<i>basic, specialty shops</i>	
Local Parks :	<i>none</i>	<i>yes</i>	<i>none</i>	<i>none</i>	
Bicycle Paths :	<i>none</i>	<i>yes</i>	<i>yes</i>	<i>yes</i>	
School Quality :	<i>very good</i>	<i>very good</i>	<i>fair</i>	<i>fair</i>	
Neighborhood Safety :	<i>average</i>	<i>average</i>	<i>average</i>	<i>average</i>	
Shopping Prices Relative to Avg :	<i>20% more</i>	<i>20% more</i>	<i>same</i>	<i>10% more</i>	
Walking Time to Shops :	<i>20-30 minutes</i>	<i>20-30 minutes</i>	<i>&lt; 10 minutes</i>	<i>10 - 20 minutes</i>	
Bus Fare, Travel Time to Shops :	<i>\$1.00, 15-20 minutes</i>	<i>\$1.00, &gt; 20 minutes</i>	<i>\$0.50, 5 - 10 minutes</i>	<i>\$0.50, &lt; 5 minutes</i>	
Travel Time to Work by Auto :	<i>&gt; 20 minutes</i>	<i>15-20 minutes</i>	<i>15 - 20 minutes</i>	<i>&lt; 10 minutes</i>	
Travel Time to Work by Transit :	<i>&gt; 45 minutes</i>	<i>30-45 minutes</i>	<i>30 - 45 minutes</i>	<i>15 - 30 minutes</i>	

# Residential Location Choice Model with Latent Lifestyle Segmentation

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- Location choice model conditional on lifestyle preferences  $P_n(i | X_n, s)$

*Prob (Location  $i$  / “Suburban Lifestyle”, other explanatory variables)*

- Model of lifestyle preferences  $P_n(s | X_n)$

*Prob (“Suburban Lifestyle” / Income, Age, etc. )*

- Joint estimation

$$P_n(i | X_n) = \sum_{s=1}^S P_n(i | X_n, s) P_n(s | X_n)$$

# Extension 1: Multiple Responses per Person

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## □ Single response

$$\begin{aligned} P_n(i | X_n) &= \sum_{s=1}^S P_n(i | X_n, s) P_n(s | X_n) \\ &= \sum_{s=1}^S P_n(s | X_n) P_n(i | X_n, s) \end{aligned}$$

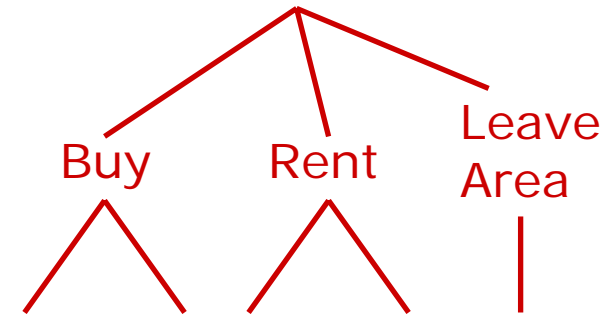
## □ Multiple responses

$$P_n(i_1, \dots, i_T | X_n) = \sum_{s=1}^S P_n(s | X_n) \prod_{t=1}^T P_n(i_t | X_{nt}, s)$$

# Extension 2: Continuous Mixing for Nesting

## □ Error Components

$$\begin{aligned}
 U_{nts}^{BSF} &= \beta_s X_{nt}^{BSF} + \sigma^B \eta_n^B + \varepsilon_{nts}^{BSF} \\
 U_{nts}^{BMF} &= \beta_s X_{nt}^{BMF} + \sigma^B \eta_n^B + \varepsilon_{nts}^{BMF} \\
 U_{nts}^{RSF} &= \beta_s X_{nt}^{RSF} + \sigma^R \eta_n^R + \varepsilon_{nts}^{RSF} \\
 U_{nts}^{RMF} &= \beta_s X_{nt}^{RMF} + \sigma^R \eta_n^R + \varepsilon_{nts}^{RMF} \\
 U_{nts}^{LEAVE} &= \beta_s X_{nt}^{LEAVE} + \sigma^L \eta_n^L + \varepsilon_{nts}^{LEAVE}
 \end{aligned}$$



$$\eta_n^B, \eta_n^R, \eta_n^L \sim iid N(0,1) \quad \varepsilon_{nts}^{BSF}, \dots, \varepsilon_{nts}^{LEAVE} \sim iid \text{ Extreme Value}$$

## □ Choice Model

$$P_n(i_1, \dots, i_T | X_n) = \int \left[ \sum_{s=1}^S P_n(s | X_n) \prod_{t=1}^T P_n(i_t | X_{nt}, s, \eta) \right] f(\eta) d\eta$$

*Estimated with Latent Gold Choice by Statistical Innovations*



# Overview of Estimation Results

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	WITHOUT Lifestyle Segmentation	WITH Lifestyle Segmentation		
Number of classes	1	2	3	4
Number of parameters	37	76	115	155
Rho-bar-square BIC	0.210	<b>0.213</b>	0.211	0.204
Rho-bar-square AIC	0.222	0.238	0.248	<b>0.254</b>

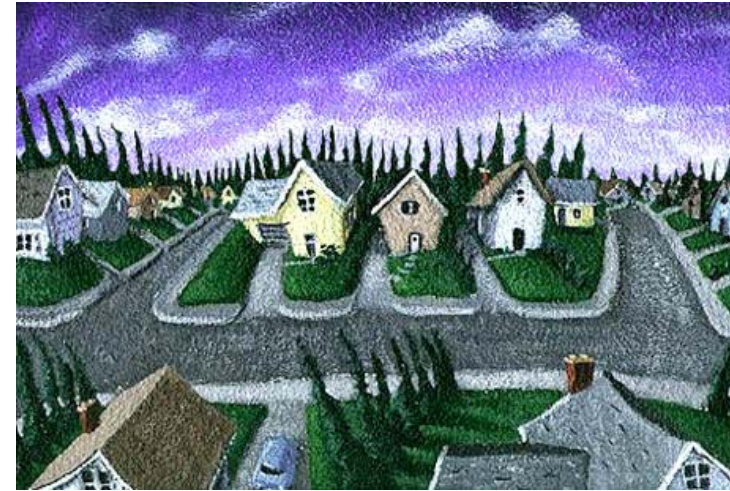
↑  
Chosen Model

# Lifestyle Segmentation Results

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## *Latent Segment 1* →

suburban, school, auto... (shopping)  
affluent, more established families



## ← *Latent Segment 2*

transit, school... (suburban)  
less affluent, younger families



## *Latent Segment 3* →

high density, urban activity... (car)  
older, non-family, professionals



# “Have it All” mentalities

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- Class 1
  - large homes and auto-oriented
  - local high-end shopping
- Class 2
  - suburban lifestyle
  - convenience of transit for work
- Class 3
  - auto-oriented
  - urban
- Relationship with development styles
  - Suburban? Mixed use? Transit-oriented?

# Policy Implications

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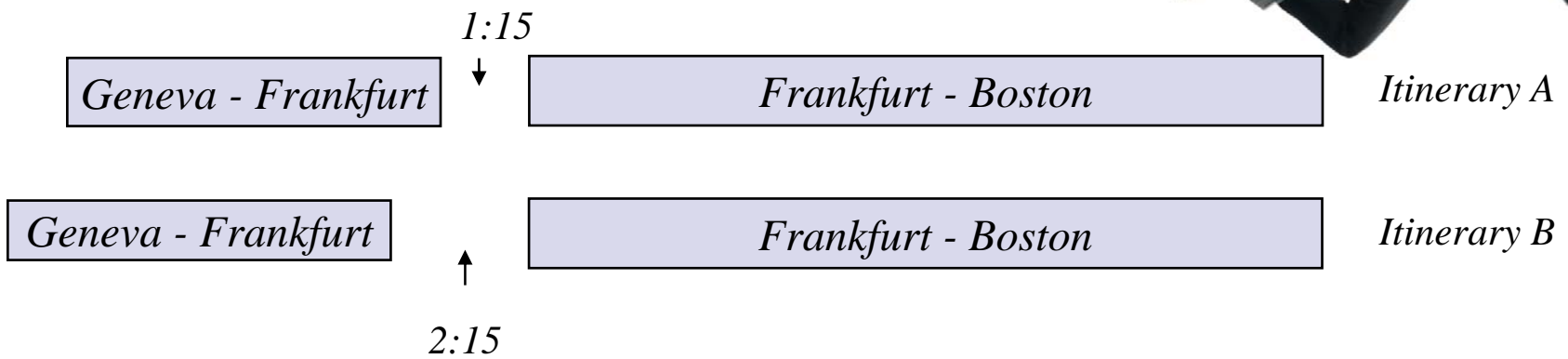
- How to encouraging more environmentally-positive behavior?
- Design for having it all
  - Supply demand for low-density residential, but support with viable transit to work. (Park-n-ride)
  - Accommodate car in urban (making the lifestyle more appealing to people who otherwise would reject) while making it easier not to have to use the car.

# APPLICATION II

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## Airline Operations (with Theis)



# Attitudes in Airline Itinerary Choice

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- Hypothesis
  - Travelers may choose longer connection time
  - Attitudes: RISK, RUSH, TRUST
  
- Motivated by airline operating assumptions
  - Demand: added minutes decreases market share
  - Supply: Depeaking lowers supply-side costs

# 2005 Survey of U.S. Domestic Passengers



Airline

Aircraft Type

Departure Airport

Flight Times

**Min. Connecting Time**

**Buffer Time**

Number of Connections

On-time Performance

Round Trip Fare

Which would you choose for a trip to Jacksonville, FL?

		Your Current Flight	Alternate Flight
AIRLINE		Delta	Continental
AIRCRAFT TYPE		Regional Jet	Standard Jet
DEPARTURE	AIRPORT	Logan International Airport, Boston MA	Burlington International Airport, Burlington VT
	TIME	8:00 AM	5:00 PM
ARRIVAL	AIRPORT	Jacksonville International	Jacksonville International
	TIME	12:00 PM	10:00 PM
LAYOVER TIME		1 hr. (your connecting airport requires a minimum of 40 mins. to connect)	40 mins. (the connecting airport requires a minimum of 40 mins. to connect)
TOTAL TRAVEL TIME		4 hrs.	5 hrs.
NUMBER OF CONNECTIONS		1	1
ON-TIME PERFORMANCE		80% of these flights are on time	90% of these flights are on time
ROUND TRIP FARE		\$250	\$188
I would choose:		<input type="radio"/> my current flight	<input type="radio"/> the alternate flight

# Psychometric Indicators

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- 14 statements rated on a 5 point scale  
“strongly disagree” to “strongly agree”

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I like to take my time when connecting between flights

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It's hard for me to find my way through airports

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Given two itineraries that only differ in connecting time, I always choose the one with shorter connecting time

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I don't think time at airports is wasted because I can shop, eat, or work at airports

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I'm willing to accept the risk of a missed connection if this gets me to my destination earlier most of the time

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I usually arrive at the check-in counter just before the check-in deadline

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Airlines sometimes underestimate the time needed to connect between flights

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It is the passenger's responsibility to plan for a sufficient transfer time when booking a connecting itinerary

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I don't mind being rushed at a connecting airport if this means I'll arrive at my final destination earlier

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Airlines only sell connections that they expect passengers could make

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I try to avoid short connections because of the risk of either me or my luggage missing the connecting flight

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I enjoy time having extra time at airports

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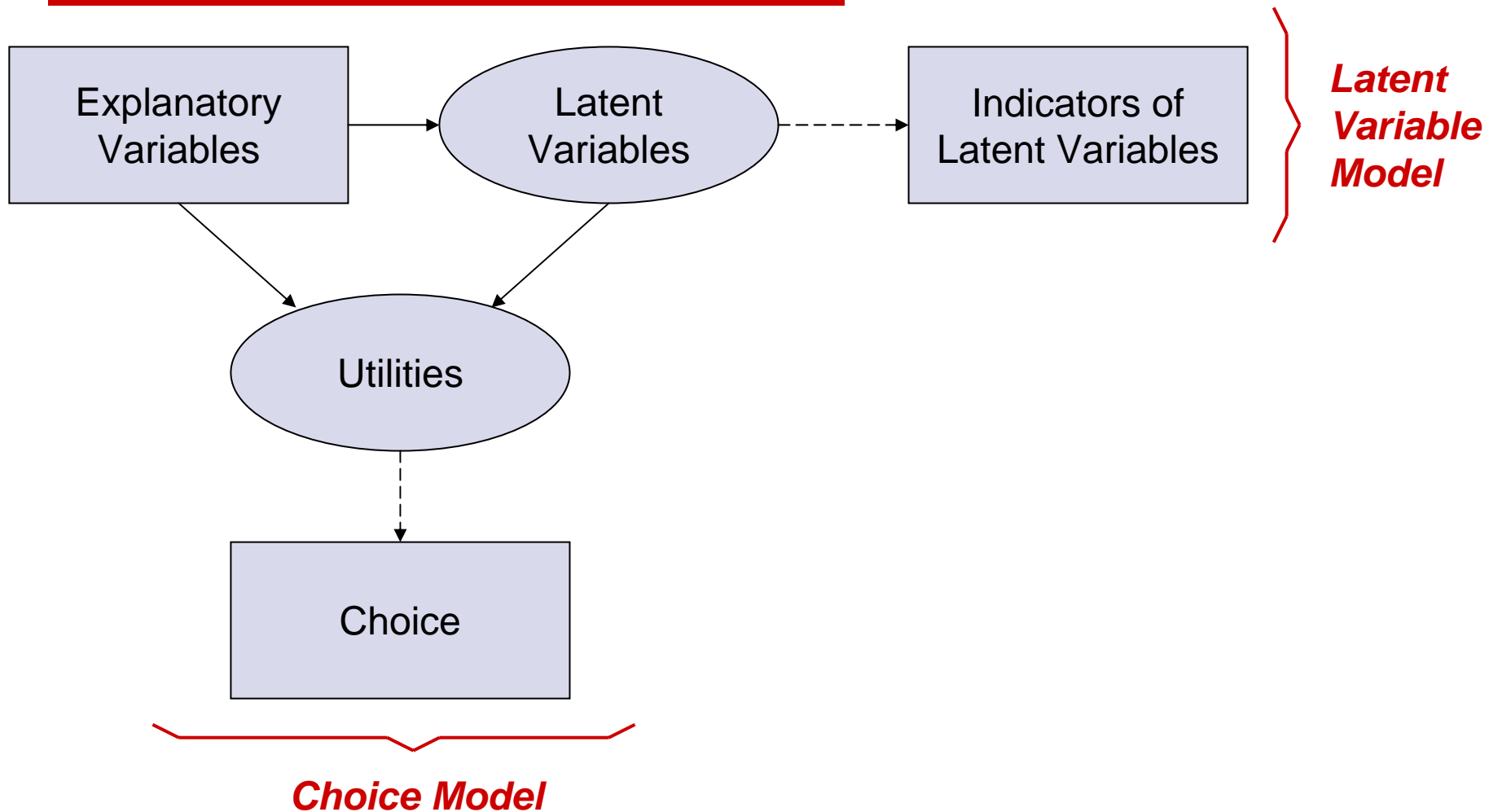
I make sure that the planned connecting time is adequate for me when booking a connecting itinerary

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# Using Psychometric Indicators

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# Formulation

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- Standard choice model (no latent variables)

$$Prob(y|X;\beta)$$

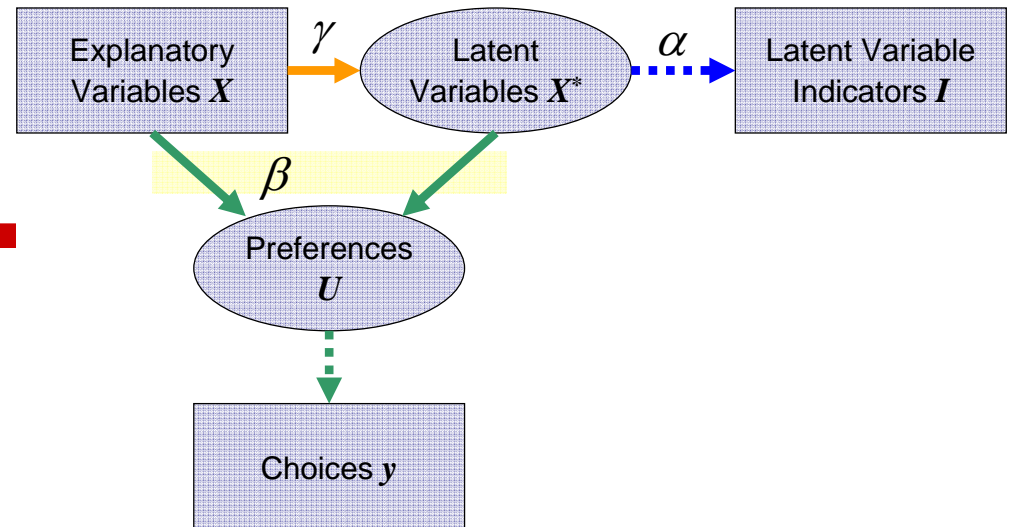
- Choice model with latent variables

$$Prob(y|X, X^*; \beta)$$

- The latent variables are unknown

$$Prob(y|X;\beta, \gamma) = \int_{X^*} Prob(y|X, X^*; \beta) f(X^*|X; \gamma) dX^*$$

# Estimation with Indicators



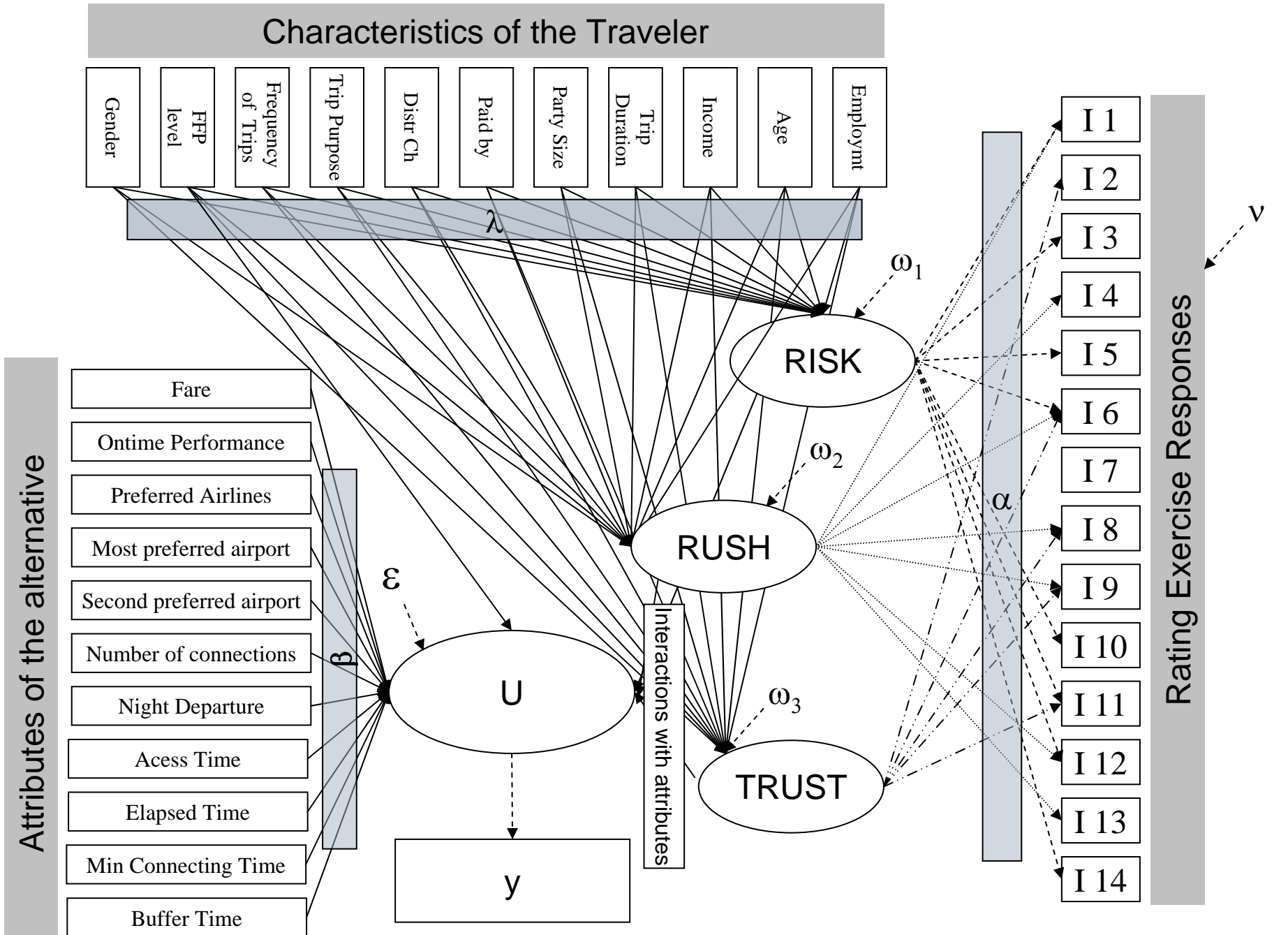
$$f(y, I | X; \alpha, \beta, \gamma)$$

$$= \int_{X^*} P(y | X, X^*; \beta) f(I | X^*; \alpha) f(X^* | X; \gamma) dX^*$$

Choice Model

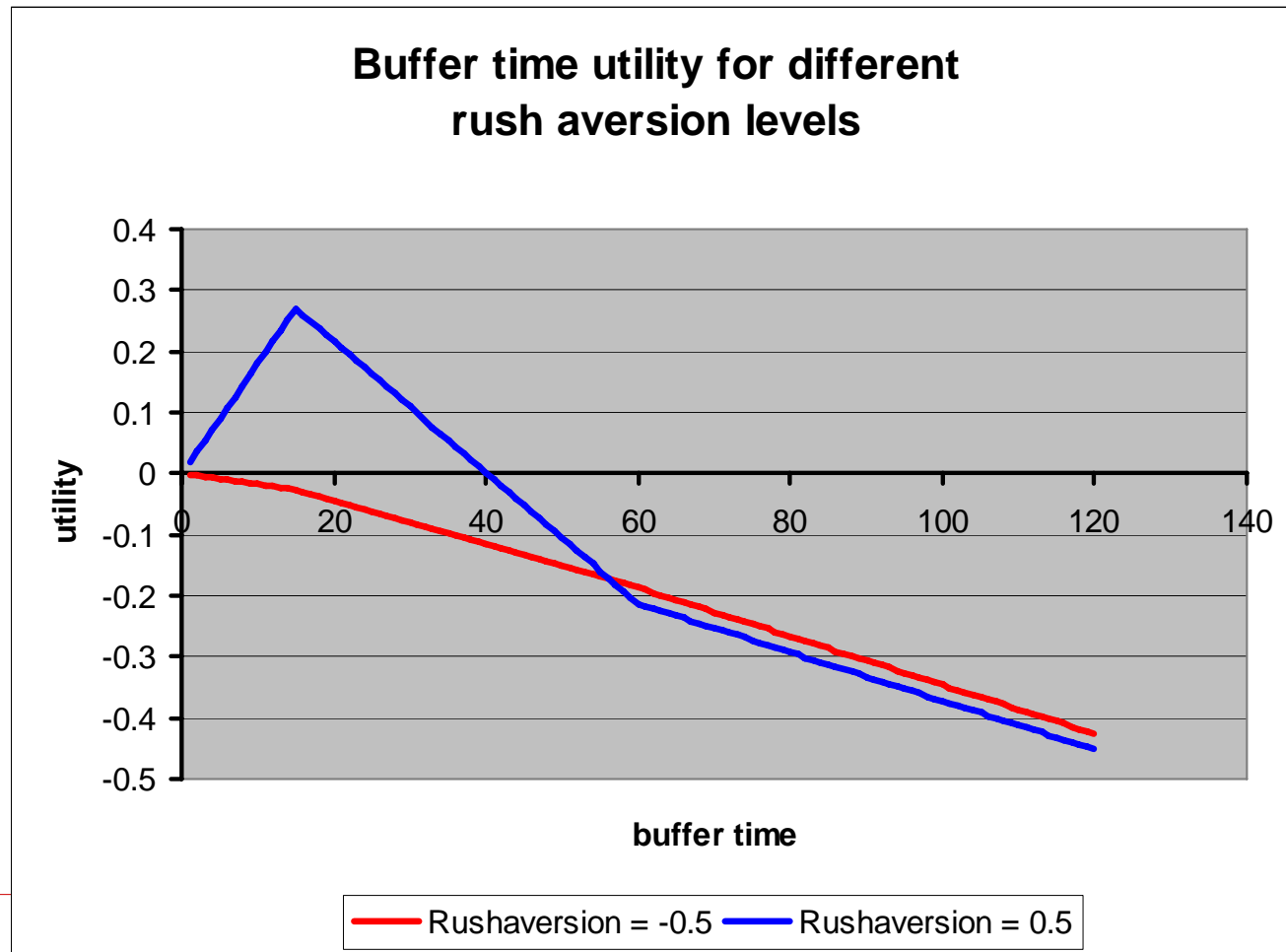
Latent Variable  
Measurement Model

Latent Variable  
Structural Model



Estimated with program from Denis Bolduc

# Influence of Rush Aversion on Perception of Buffer Time



# Who Tends Towards Rush Aversion?

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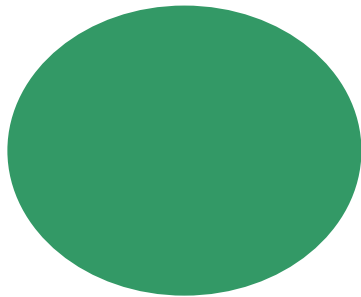
- Demographics
  - Females
  - Low/middle income
  - Employed persons
- Trip characteristics
  - Business travelers
  - Checked bags
  - Short trips (< 3 nights)
- Traveling history
  - Not missed flight in last 6 months
  - Not elite travelers

# APPLICATION III

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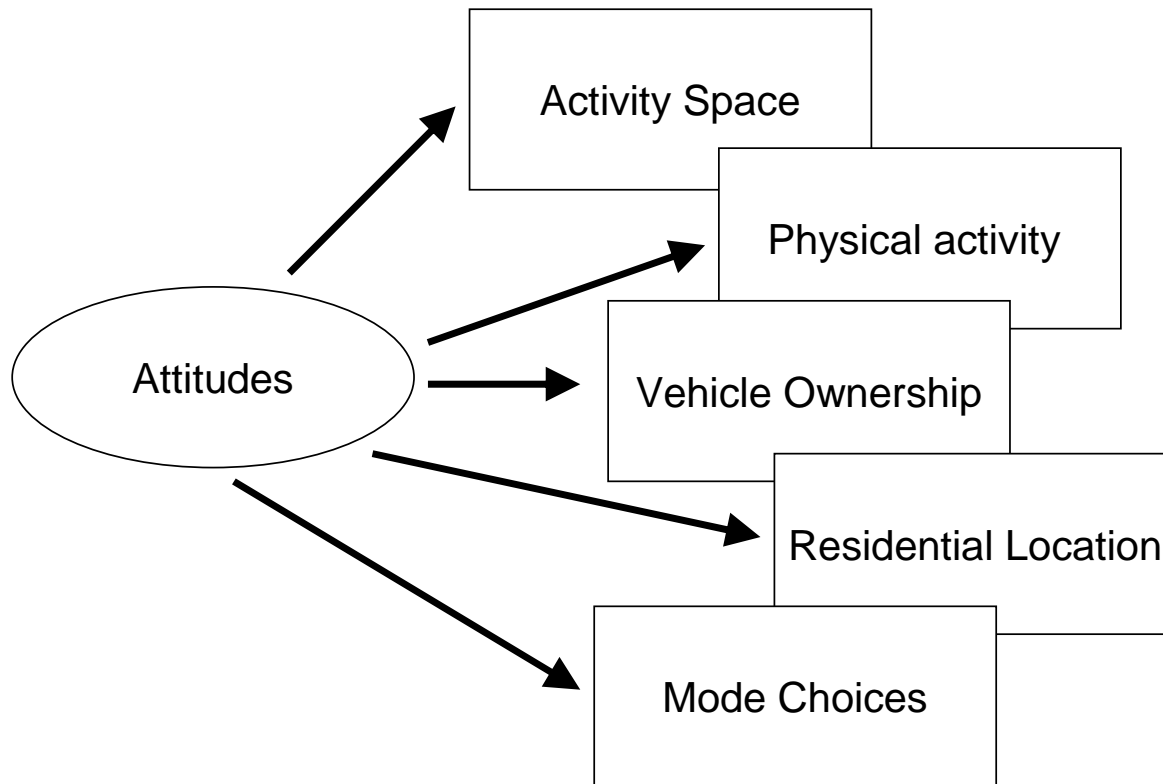
## Environmentalism and Behavior

(with Mokhtarian and Schwanen)



# Hypothesis: Intrinsic Attitudes Influencing Related Behaviors

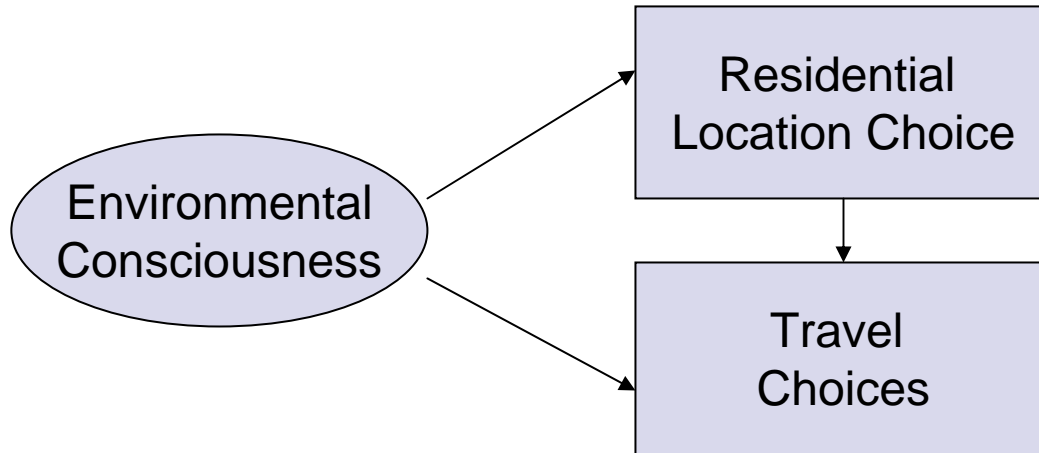
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# Application

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- 1993 household travel and activity survey from San Francisco Bay Area<sup>1</sup>
- Multiple neighborhoods sampled
  - "Traditional" – North San Francisco
  - "Suburban" – Concord & San Jose

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<sup>1</sup>Sponsored by California Air Resources Board.  
Described in Kitamura, Laidet, Mokhtarian, Buckinger, Gianelli (1994).  
Using data processed by Michael Bagley.

# Choices

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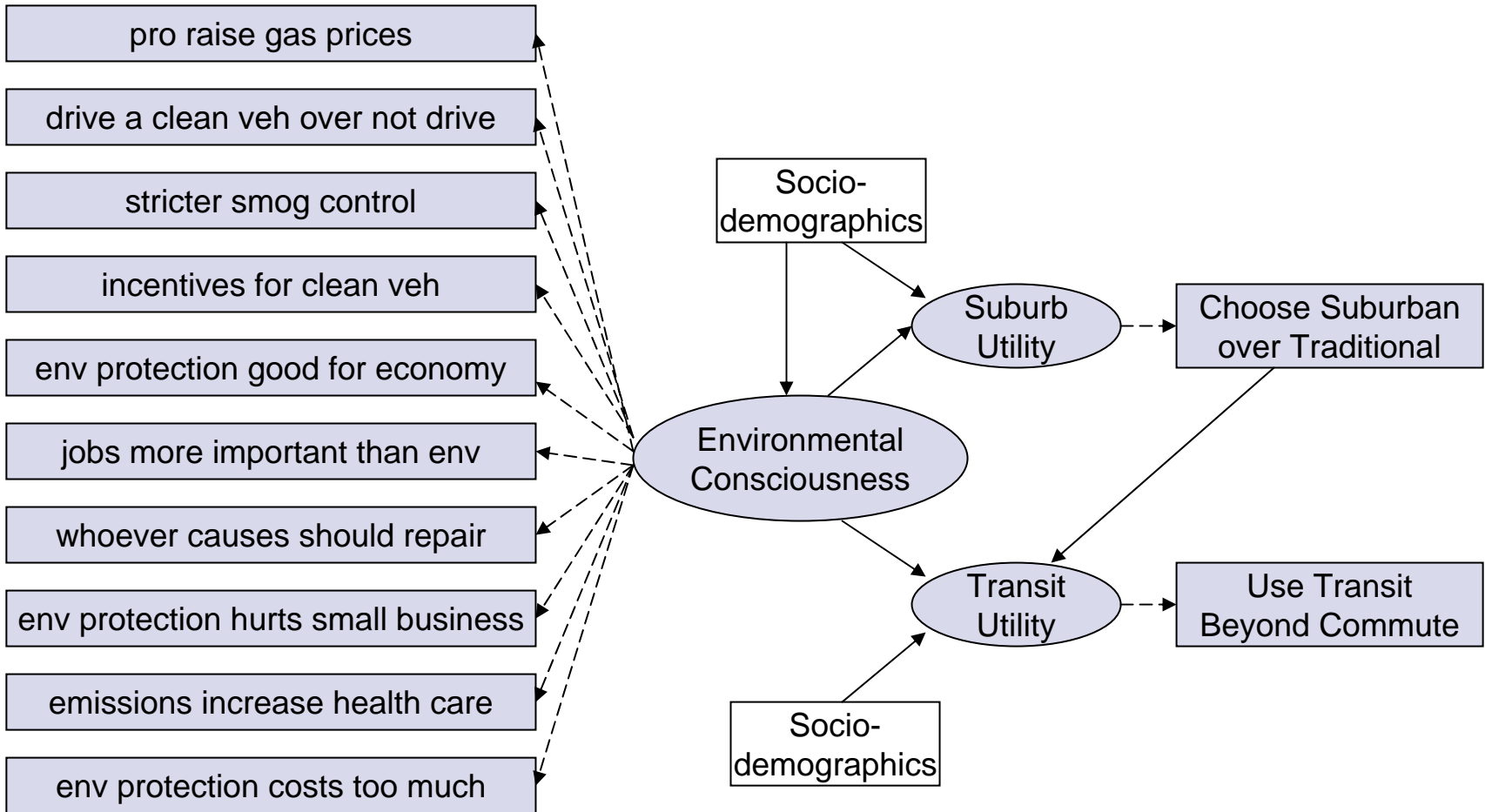
- Residential Location
  - Suburban Neighborhood
  - Traditional Neighborhood
- Travel
  - Use of transit beyond commuting
  - No use of transit beyond commuting

# Indicators of Environmental Consciousness

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- We should **raise the price of gasoline** to reduce congestion and air pollution.
- I would rather **drive an electric** or other clean-fuel vehicle than give up driving.
- Stricter vehicle smog control laws** should be introduced and enforced.
- We should provide **incentives to people who use electric** or other clean-fuel vehicles
- Environmental protection is **good for California's economy**.
- People and **jobs are more important** than the environment.
- Whoever causes** environmental damage **should repair** the damage.
- Environmentalism **hurts minority and small businesses**.
- Vehicle emissions **increase the need for health care**.
- Environmental protection **costs too much**.

# Model Framework



# Formulation

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## □ Residential location – binary logit

$$\frac{U_{Suburb}}{U_{Traditional}} = \beta_0 + \beta_1 \left( \begin{array}{l} \text{Environmental} \\ \text{Consciousness} \end{array} \right) + \beta_x \left( \begin{array}{l} \text{Socio-} \\ \text{Demographics} \end{array} \right) + \varepsilon_{Suburb} - \varepsilon_{Traditional}$$

## □ Transit use – binary logit

$$\frac{U_{Transit}}{U_{NotTransit}} = \alpha_0 + \alpha_1 \left( \begin{array}{l} \text{Environmental} \\ \text{Consciousness} \end{array} \right) + \alpha_2 (\text{Suburb}) + \alpha_x \left( \begin{array}{l} \text{Socio-} \\ \text{Demographics} \end{array} \right) + \varepsilon_{Transit} - \varepsilon_{NotTransit}$$

## □ Environmentalism – set of linear eqs

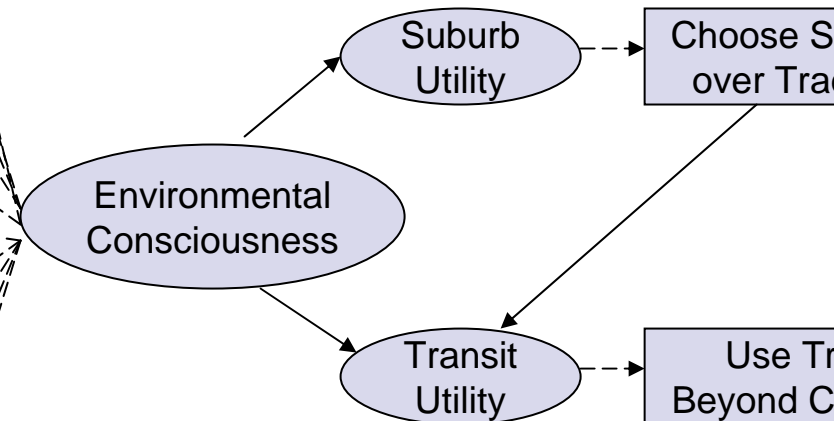
$$Indicator_k = \lambda_k \left( \begin{array}{l} \text{Environmental} \\ \text{Consciousness} \end{array} \right) + \varepsilon_k \quad k = 1, \dots, 10$$

$$\left( \begin{array}{l} \text{Environmental} \\ \text{Consciousness} \end{array} \right) = \gamma \left( \begin{array}{l} \text{Socio-} \\ \text{Demographics} \end{array} \right) + \varepsilon_{EC}$$

# Estimation Results (1)

Loading t-stat

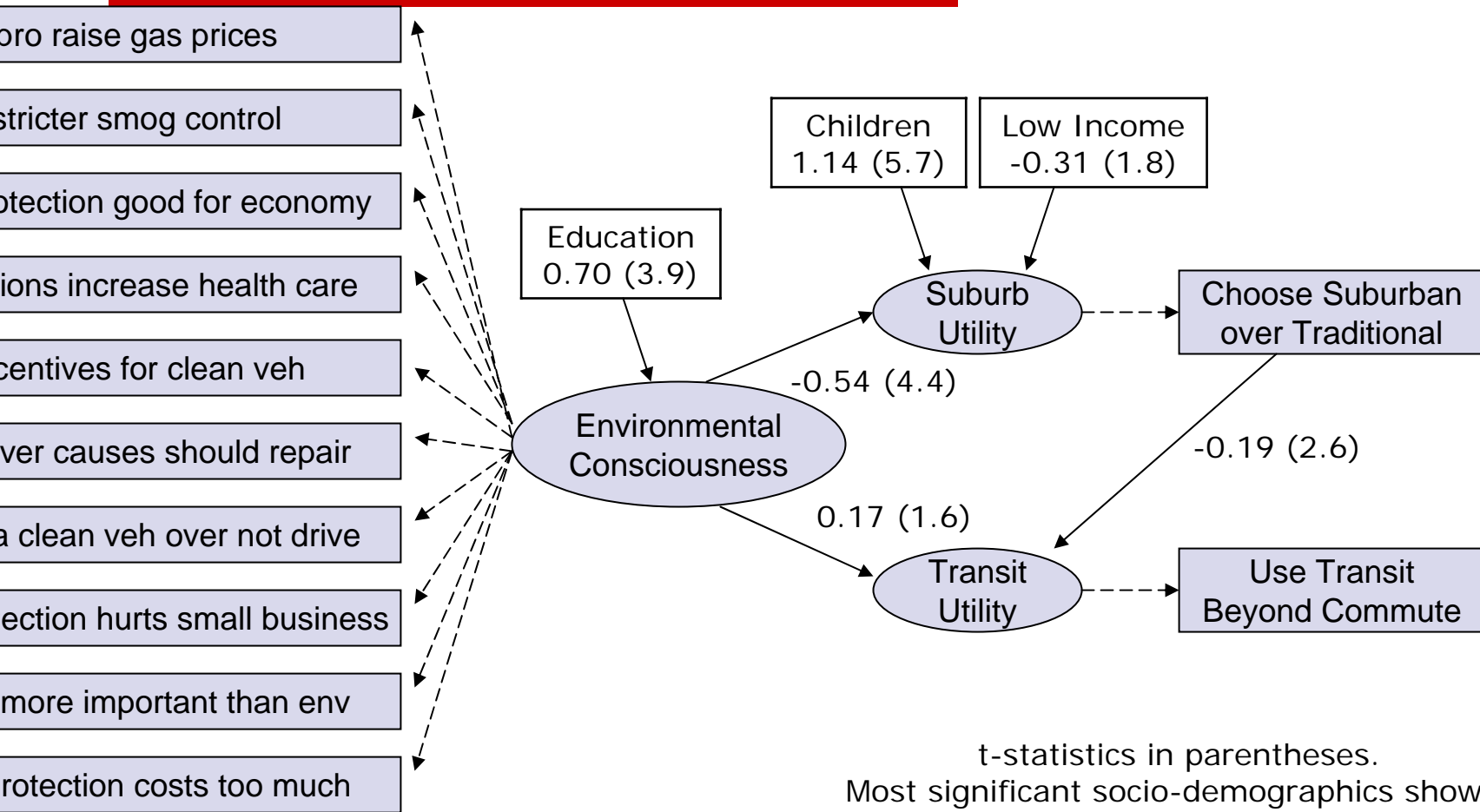
Loading	t-stat	
1.00	--	pro raise gas prices
0.93	(8.5)	stricter smog control
0.92	(7.9)	env protection good for economy
0.62	(8.4)	emissions increase health care
0.57	(7.6)	incentives for clean veh
0.30	(6.1)	whoever causes should repair
0.19	(3.0)	drive a clean veh over not drive
-0.79	(8.2)	env protection hurts small business
-0.86	(8.6)	jobs more important than env
-1.13	(8.3)	env protection costs too much



468 Observations

*Estimated with Mplus*

# Estimation Results (2)



t-statistics in parentheses.  
Most significant socio-demographics shown.

# CONCLUSION

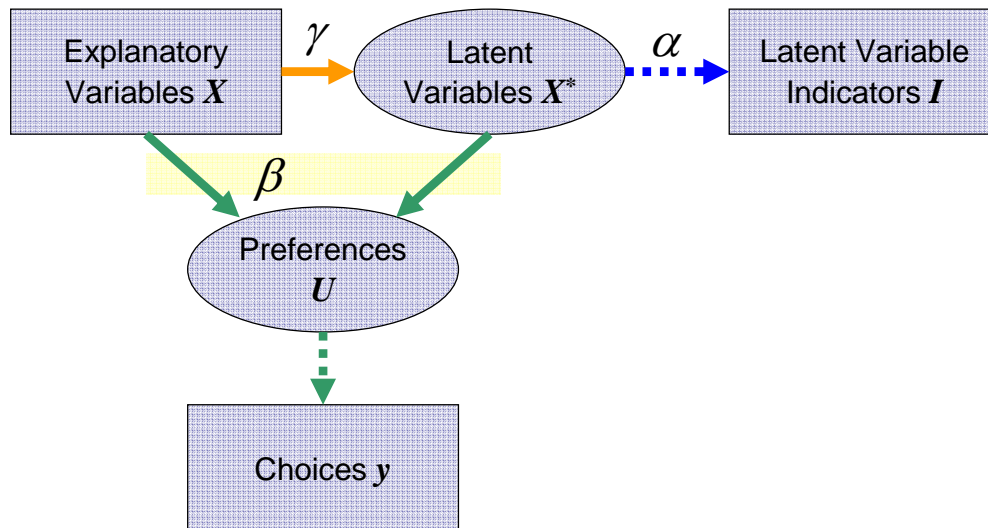
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# Method

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- Integrated choice and latent variable models provide powerful & practical method to enrich behavioral representation in discrete choice models



# Comparison with State of the Art

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## 1. Mixture Models

- Sophisticated models of the covariance
  - Random parameters, error components

## 2. Behavioral Mixture Models

- Model covariance structure via explicit latent variable constructs
- Provide behavioral rational to mixtures

*Fit?*  
*Temporal stability over time?*  
*Policy implications?*

# Applications Demonstrate More Intuitive Models

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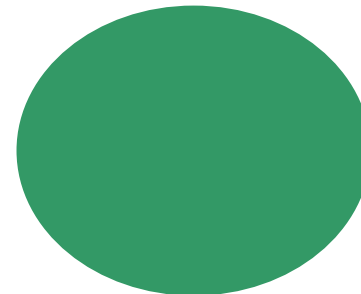
## 1. Lifestyle and Residential Choices



## 2. Risk, Rush and Airline Itinerary Choices



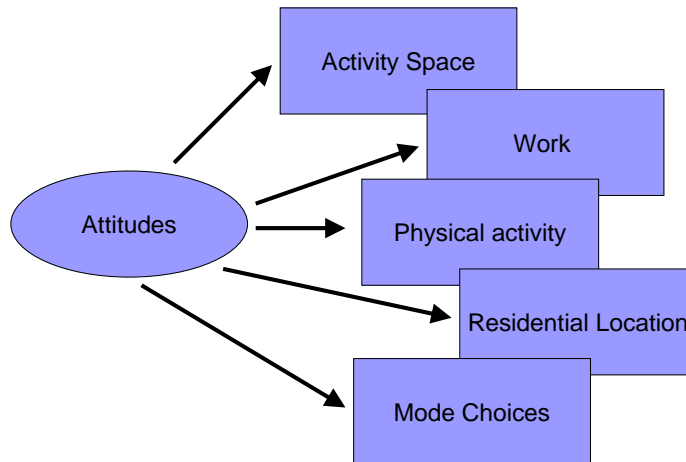
## 3. Role of Environmentalism



# Future Directions

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## □ Multi-contextual modeling



## □ Attitudinal trends

## □ Validation with forecasts