

Activity-Based Travel Demand Modeling

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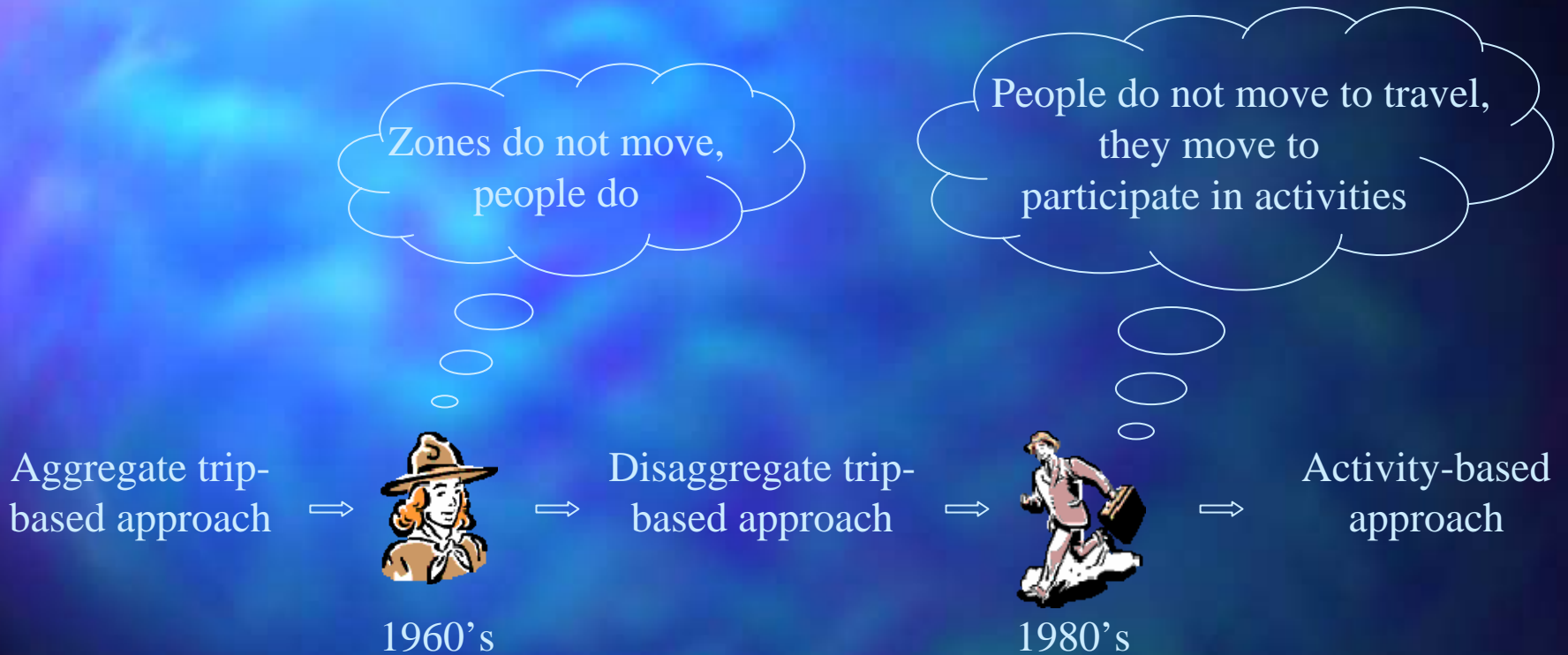
Presentation Structure

- Background
- Need for activity-based travel analysis
- Fundamental concepts and structural framework
- Possible modeling framework
- Recent methodological advances
- Summary and conclusions

Background

- Travel demand models are used to analyze travel characteristics
- Travel forecasts are important determinants of capital decisions
- Improved models can enhance quality of decisions

Evolution of Travel Demand Forecasting Techniques



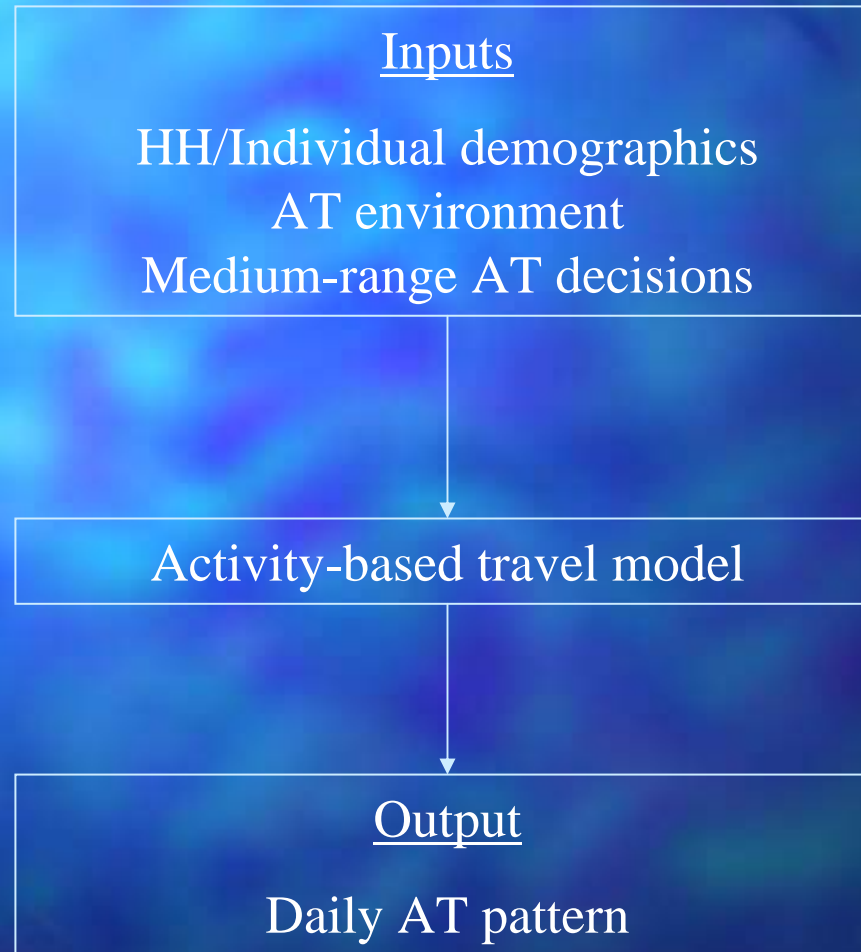
Practical Need for Activity-Based Travel Analysis

- Tour mode ↔ number of stops ▶
 - Substitution effects ▶
 - Time of day
 - ◆ TCM analysis
 - ◆ Link volumes
 - Number of trip starts
 - Travel time duration
 - ◆ Cold and hot starts ▶
- ⇒ Activity-travel patterns along a continuous time domain
- Time-space interactions

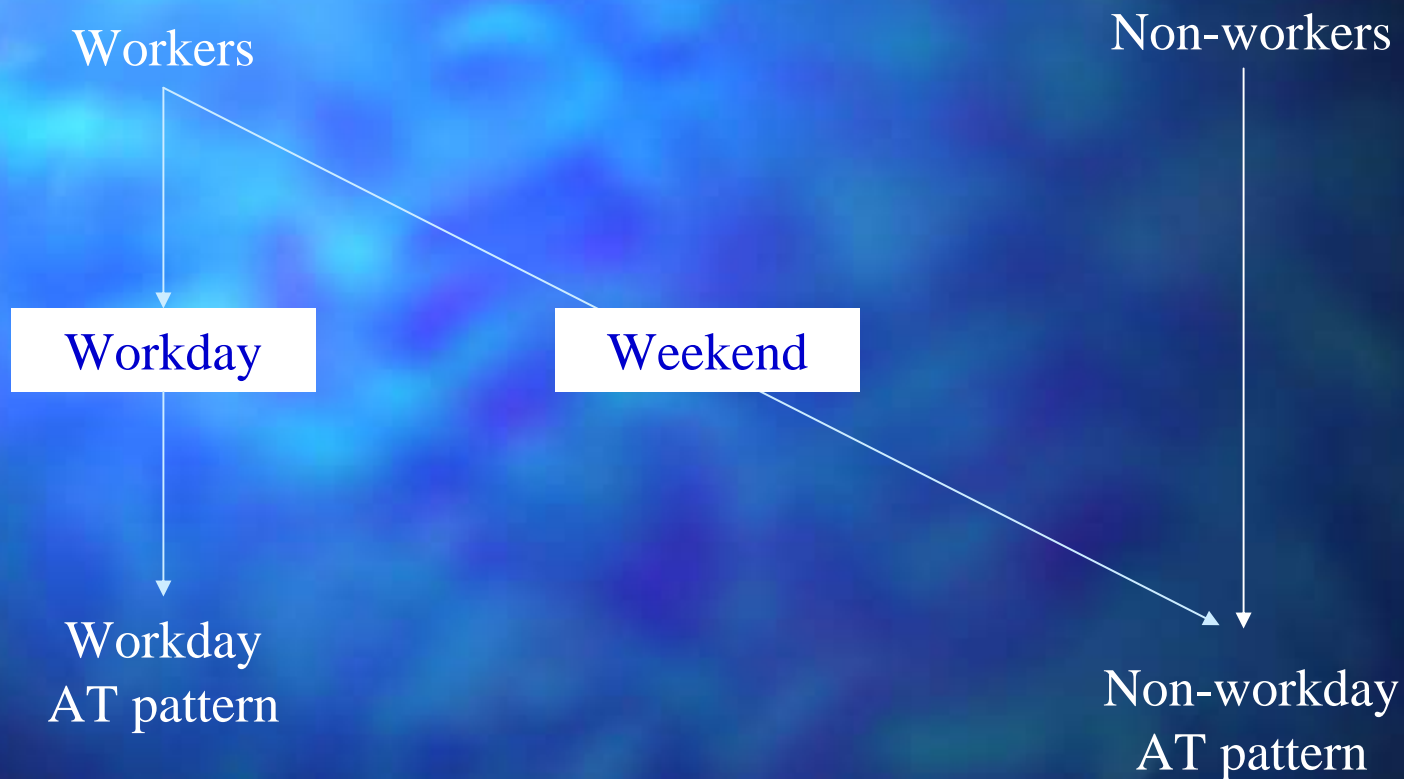
Fundamental Concepts

- Travel is a derived demand
- Complex interactions necessitate analyzing sequences or patterns of activity behavior
- Unit of analysis: whole day or longer time period
- Time is a continuous entity
- Interactions exist among individuals

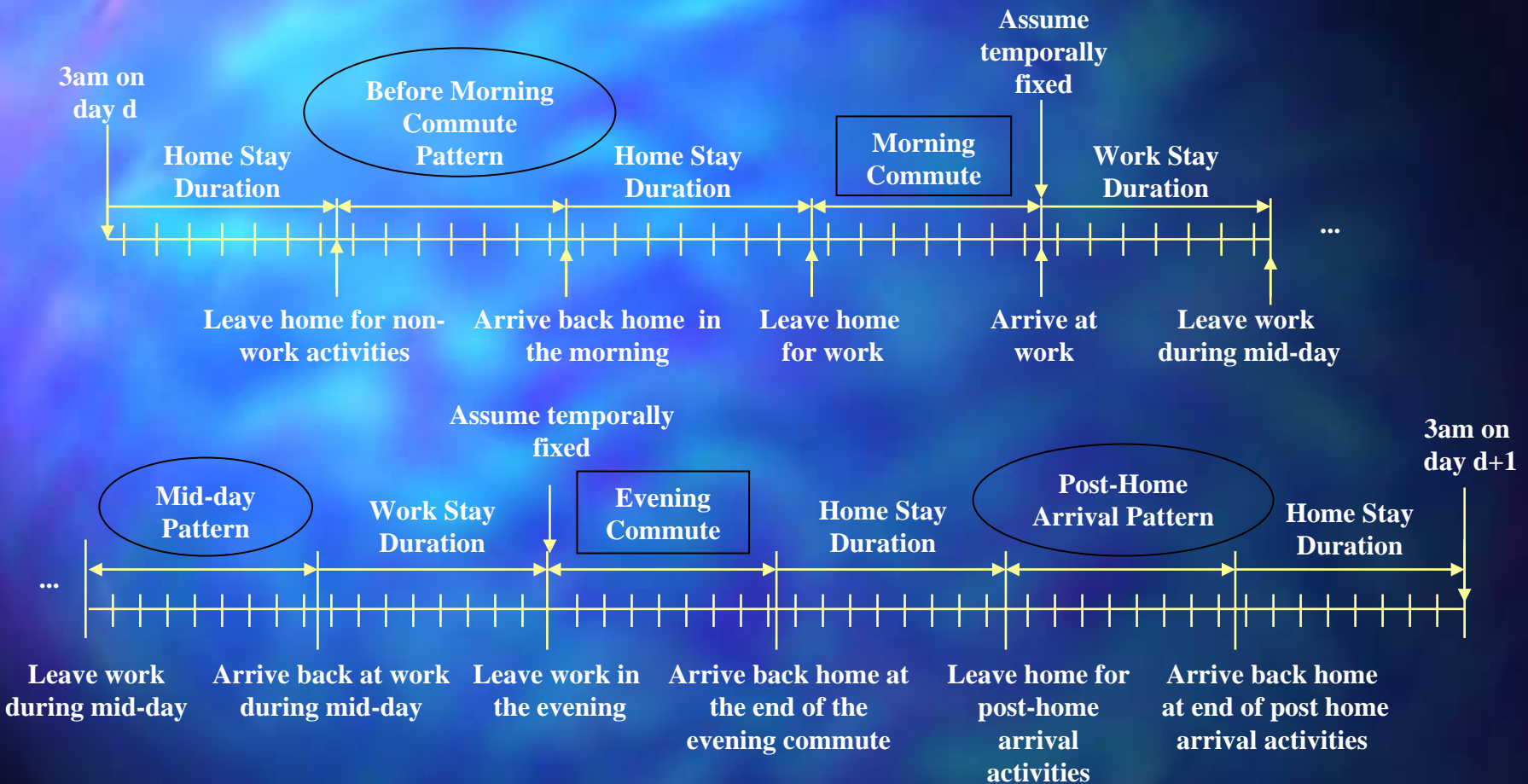
Activity-Travel (AT) Analysis: Structural Framework



Daily AT Pattern



Worker's Daily Activity Travel Pattern



Note:

Morning Commute +
 Evening Commute =
 Work Commute Plan

Attributes of Workday AT Pattern (1)

- Pattern-level attributes
 - ◆ Number of tours (before morning commute pattern, midday pattern, post home-arrival pattern)
 - ◆ Home-stay duration before morning commute

Attributes of Workday AT Pattern (2)

- Tour-level attributes
 - ◆ Travel mode
 - ◆ Number of stops
 - ◆ Home-stay duration before each tour in the before morning commute and post home-arrival patterns
 - ◆ Work stay duration before each tour in midday pattern
 - ◆ Sequence of tour in pattern

Attributes of Workday AT Pattern (3)

- Stop-level attributes
 - ◆ Activity type
 - ◆ Travel time to stop from previous stop
 - ◆ Location of stop
 - ◆ Sequence of stop in tour

Distribution of Number of Tours

Boston data **Bay Area data**

# of tours	Percentage of each number of tours in ...					
	Before work pattern		Mid-day pattern		Post-home arrival pattern	
0	96.9	96.9	65.6	74.0	67.6	79.7
1	3.0	2.9	29.2	22.5	28.5	18.2
2	0.1	0.2	4.0	2.8	3.4	1.9
≥ 3	0.0	0.0	1.2	0.7	0.5	0.3

- Focus on presence/absence of a first tour in each pattern and then model presence/absence of second tour
- Consider interactions in first tour only across different periods of the day

Distribution of Number of Stops

# of stops	Percentage of each number of stops in ...									
	Before work pattern		Morning Commute		Mid-day pattern		Evening Commute		Post-home arrival pattern	
0	96.9	96.9	85.3	85.2	65.6	74.0	64.8	74.0	67.6	79.7
1	2.3	2.2	11.3	11.5	24.6	17.8	24.1	17.8	23.1	14.2
2	0.5	0.5	2.8	2.5	6.4	4.4	7.9	5.1	6.8	3.9
3	0.2	0.3	0.5	0.6	2.4	2.0	2.4	1.9	2.0	1.4
4	0.1	0.1	0.1	0.2	0.5	0.9	0.7	0.9	0.5	0.5
≥5	0.0	0.0	0.0	0.0	0.5	0.9	0.1	0.3	0.0	0.3

- Midday, evening commute, post-home arrival stops
 ⇒ morning commute stops ⇒ before work stops
- First stop in each period ⇒ second stop ⇒ ...

Interaction in Stop-Making Across Different Times of Day

Control Variable	Value Label	Percentage of individuals making a stop during ...					
		Mid-day		Evening commute		Post-home arrival period	
Made a mid-day stop?	Yes	-	-	36.7	28.0	32.8	23.2
	No	-	-	34.4	25.3	32.2	20.5
Made an evening commute stop?	Yes	35.9	27.6	-	-	25.3	13.2
	No	33.6	25.4	-	-	36.3	22.8
Made a post-home arrival stop?	Yes	34.8	27.4	27.5	18.1	-	-
	No	34.2	25.6	38.9	28.0	-	-

Distribution of Activity Type of Stops

Activity type	Percentage of stops for each activity type during ...					
	Mid-day		Evening commute		Post-home arrival period	
Home	11.6	8.9	0.0	0.0	0.0	0.0
Pick-up/drop off	1.5	2.8	10.9	14.4	11.7	11.0
Work-related	18.4	19.3	2.6	10.1	1.8	2.8
School	1.5	0.3	1.6	2.3	1.8	4.2
Shopping	9.6	5.5	30.1	27.2	25.2	24.3
Social/recreational	2.9	3.7	16.5	15.6	32.8	27.3
Eat-out	39.1	41.9	7.0	4.8	11.0	12.4
Personal Business	15.4	13.5	31.3	25.6	15.7	15.7

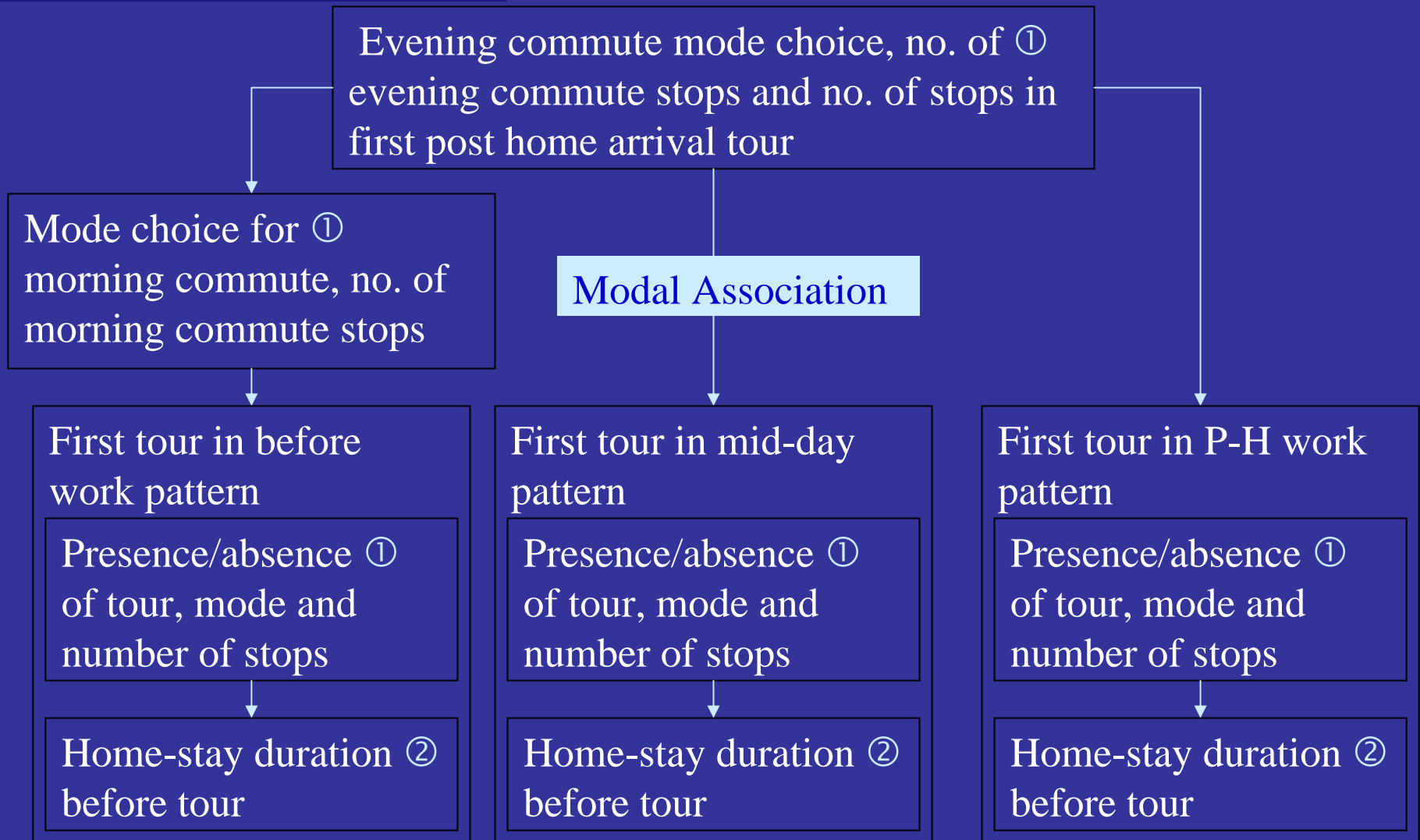
- Model number of stops in evening commute and first post home arrival tour jointly
- Model midday stop making independently

Evening Commute Mode Choice and Number of Stops

Category	Drive alone	Non-Drive Alone
% making commute stops	37.8	11.7
% making post-home arrival stops	38.6	36.7

- Model evening commute mode choice, number of evening commute stops and number of stops in first post-home arrival tour jointly
- Model travel mode for first post-home arrival tour conditional on number of stops in the tour

Modeling Framework for Pattern/Tour Level Attributes



Modeling Framework for Stop-Level Attributes



Recent Methodological Advances (1)

Motivated by:

- Refinements in discrete choice modeling
- Increasing realization of the need to model travel as part of a more holistic activity-travel pattern

Topics discussed

- Discrete choice models
- Duration modeling
- Limited-dependent variable models

Recent Methodological Advances (2)

Topics not discussed

- Detailed information on model structure/estimation
- Developments in survey data collection or imputation
- Advances in joint estimation from RP-SP data
- Panel data methods
- Computational process models

Discrete Choice Modeling

- Multinomial logit (MNL) model has been most widely used
- MNL model assumptions include:
 - ◆ IID type I extreme-value error structure across alternatives
 - ◆ Homogeneity in responsiveness to attributes of alternatives
 - ◆ Identical error variance-covariance matrix across individuals

Relaxing IID (Across Alternatives)

- Identical, but non-independent random components
- Non-identical, but independent random components
- Non-identical, non-independent random components

Identical, but Non-Independent Random Components

- Normal or type I extreme-value
- Type I extreme-value preferred
- GEV class
 - ◆ NL
 - ◆ PCL (Chu, Koppelman & Wen)
 - ◆ CNL (Vovsha, 1996)
 - ◆ OGEV (Small, 1987)
 - ◆ MNL-OGEV (Bhat, 1997)
- Advantage: closed-form
- Limitation: restrictions on dissimilarity parameters

Non-Identical, but Independent Random Components

- NE (Daganzo, 1979)
- Oddball model (Recker, 1995)
- HEV (Bhat, 1995)

Non-Identical, Non-Independent Random Components

- Error-components logit or probit (Train, 1995; Ben-Akiva & Bolduc, 1996; Bhat, 1997)

$$\begin{aligned}U_i &= V_i + \zeta_i \\ &= V_i + \mu'z_i + \varepsilon_i\end{aligned}$$

- MNP (Bunch & Kitamura, 1993; Mahmassani, 1996; marketing)

Relaxing Homogeneity in Responsiveness

Variation in coefficients

- Random coefficients with no systematic coefficient variation (Fischer & Nagin, 1985; Revelt & Train, 1996; Mehndiratti, 1996)
- Random coefficients with systematic variation (Bhat, 1996)

$$U_{qi} = \alpha_i + \delta'_i z_q + \varepsilon_{qi} + \eta'_q x_{qi}$$

$$\eta_{qk} = \pm \exp(\gamma_k + v_{qk})$$

$$\eta_{qk} = \pm \exp(\gamma_k + \beta'_k \omega_{qk} + v_{qk})$$

Segmentation Approaches

- Exogenous segmentation
 - ◆ Identify mutually-exclusive segments

- Endogenous segmentation (Swait, 1994; Gopinath & Ben-Akiva, 1995; Bhat, 1997; marketing)
 - ◆ Probabilistic assignment

Relaxing Error Variance-Covariance Structure Homogeneity Across Individuals

- Variance relaxation (Swait & Adamowicz, 1996; McMillen, 1995; Steckel & Vanhonacker, 1988)
- Covariance relaxation (Bhat, 1997; Kamakura et al., 1996)
- Variance and covariance relaxation

Hazard Duration Models (1)

- Modeling activity duration
- Focus on end-of-duration given that duration has lasted to some specified time
- Two specification issues
 - ◆ Distributional assumption regarding duration
 - ◆ Assumptions about unobserved heterogeneity

Hazard Duration Models (2)

Baseline hazard distribution

- Parametric (exponential, weibull, log-logistic, gamma, log-normal)
- Nonparametric

Unobserved heterogeneity

- Parametric (gamma, normal)
- Nonparametric
- Mostly ignored in transportation field (but see Bhat, 1996 and Hensher, 1994)

Multiple Duration Process

- More interesting and realistic
- Independence among risks (Gilbert, 1992)
- Competing risks model (Han and Hausman, 1991; Hensher, 1994)
- Generalized multiple-durations model (Bhat, 1996)

Limited-Dependent Variable Models

- Discrete/Continuous (Barnard and Hensher, 1992; Bhat, 1996)
- Discrete/Ordinal (Bhat, 1997)

Summary and Conclusions

The substantial progress can be traced to:

- Need for realistic representation
- Ability to provide micro-level demographic inputs
- Better tools for data storage/processing
- Advent of simulation techniques to approximate multi-dimensional integrals