

Urban transportation

* Transport economics: Partial equilibrium, share of commuting options.

* Urban way: Commuting mode choice from the urban economics perspective.

→ Full eq; but not as well suited for the data.

Commuting options: walk $w(l - b_w r) = z + SR(r)$

$$\text{bus} \quad w(l - X - b_B r) = z + SR(r)$$

Time cost important ← drive $w(l - b_D r) - c = z + SR(r)$

for high w

Pecuniary cost important

for low w .

$b_w > b_B > b_D$ Time to travel
1 unit of distance.

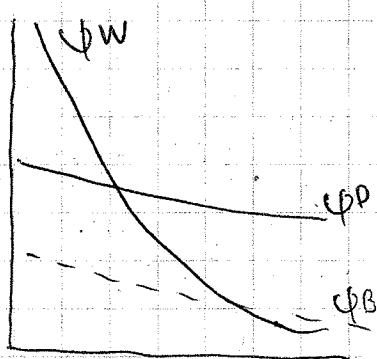
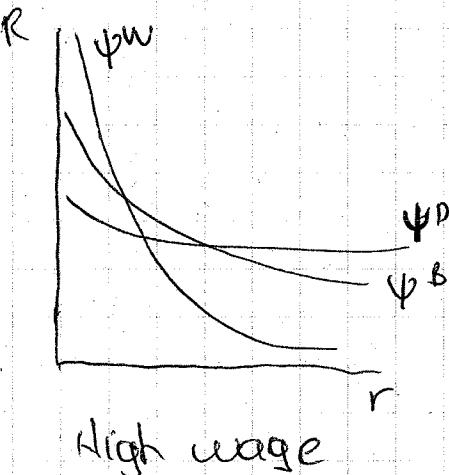
$X \rightarrow$ Fixed time cost for bus

$c \rightarrow$ Fixed pecuniary cost of driving.

Slopes of bid-rents

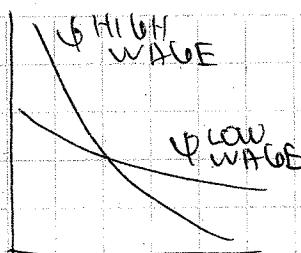
$$-\frac{w_{bw}}{s}, -\frac{w_B}{s}, -\frac{w_D}{s}$$

→ Seems to match configuration of big cities...



* Why do the poor live in cities? Should be outbid by the rich.

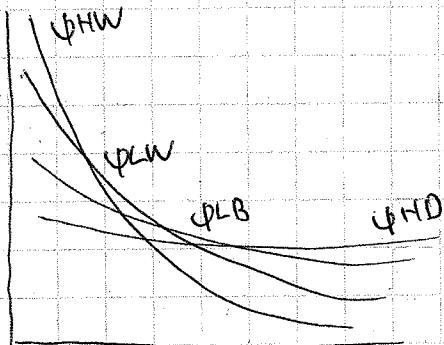
→ Closier paper X



$$\text{if } \frac{d\ln s}{d\ln I} < 1$$

& should hold within each commuting mode.

Combining the two



* Shows forces at play, but does not suit data well

Modelling individual commuting mode choice

- McFadden (1974)

Random utility set-up

$$U(s, x) = v(s, f(x)) + \epsilon(s, x)$$

↑
Final choice
choice chars ↓
deterministic

↗ indirect utility
iid. draw prob of utility
ne greater than other choices.

$$\Pr[x_i(s, B)] = \prod_{i \in \text{choice set}} \Pr[v(s, x_i) > v(s, x_j)]$$

$$= \Pr[v(s, x_i) \geq \max_j v(s, x_j)]$$

Need to choose distribution with closed form for max.

~~Read~~ $f(y) = p(\epsilon | y) = e^{-e^{-(y-\mu)/b}}$ Extreme value type I (Gumbel)

$$\begin{aligned} \text{mode} &= \mu \\ \text{mean} &= \mu + 0.58 \\ \text{sol} &= \pi/\sqrt{6} \end{aligned}$$

$$\Pr[\text{choose 1}] = \Pr[v + \epsilon_1 \geq \max_{j>1} [v_j + \epsilon_j]]$$

Distribution of max for EV type I: $\Pr[\epsilon_2 \leq v - \epsilon_1] \Pr[\epsilon_3 \leq v - \epsilon_2]$

$$\Pr[\max_{j>1} [v_j + \epsilon_j] \leq v] = \prod_{j=2}^J e^{-e^{-(v-\epsilon_j)}}$$

$$= e^{-\sum_j e^{-(v_j - \epsilon_j)}} = e^{-e^{-v}} \sum_j e^{v_j}$$

$$\Pr[\max_{j \geq 1} [v_j + \epsilon_j] \leq u] = e^{-e^{-u} \sum_{j \geq 1} v_j}$$

$$\text{So } \Pr[v_1 + \epsilon_1 \geq \max_{j \geq 1} [v_j + \epsilon_j]] = e^{-e^{-(v_1 + \epsilon_1)} \sum_{j \geq 1} v_j}$$

Unconditional probability of choosing 1

$$= \int_{-\infty}^{\infty} f(\epsilon_1) \Pr(v_1 + \epsilon_1 \geq \max_{j \geq 1} [v_j + \epsilon_j]) d\epsilon_1$$

$$= \frac{e^{-(e^{v_1} (e^{v_2} + e^{v_3} + \dots + e^{v_K}) + 1)} e^{-\epsilon_1}}{e^{v_1} (e^{v_2} + e^{v_3} + \dots + e^{v_K}) + 1} \cdot \frac{e^{v_1}}{\sum_{j \geq 1} e^{v_j}}$$

q switch in notation

$$\begin{matrix} p_{ij} \\ \text{of choice } i \end{matrix} = \frac{\exp(\phi_i^T s_i + \theta^T c_{ij} s_j)}{\sum_k \exp(\phi_k^T s_j + \theta^T c_{kj} s_j)}$$

Estimate by ML

$$\hat{\phi}, \hat{\theta} = \underset{j}{\operatorname{arg\,max}} \left[\sum_i p_{ij} D_{ij} \right] = \begin{cases} 1 & \text{if } j \text{ chooses } i \\ 0 & \text{otherwise} \end{cases}$$

$j \mapsto$ "What units? Households, individuals?"

$i \mapsto$ IIA, criticism.

IIA may not hold given the information we have about choices.

\rightarrow Intra-household allocation

Identification: Not all ϕ, θ identified, only can identify them relative to a reference choice.

$$\begin{matrix} p_{ij} = \exp(\phi_i^T s_i + \theta^T (c_{ij} - c_{1j}) s_j) \\ 1 + \sum_{k \geq 1} \exp(\phi_k^T s_i + \theta^T (c_{kj} - c_{1j}) s_j) \end{matrix}$$

What to do to get at IIA?

\rightarrow Normal disturbances

Probit random coefficients model

Probit random coefficients model

* Correlation in errors and coefficients across choices.

Suppose $U_i = Z_i \beta + n_i$ for choice i \rightarrow can be relaxed.
 $n_i \sim N(0, \sigma^2_i)$, $n_i \perp\!\!\!\perp z_i, n_j, \beta$

$$\beta \sim N(\bar{\beta}, \Sigma_\beta)$$

$$U_i = Z_i \bar{\beta} + Z_i (\beta - \bar{\beta}) + n_i$$

Suppose there are three choices

If $U_1 \geq U_2$

$$U_1 - U_2 = (Z_1 - Z_2) \bar{\beta} + (Z_1 - Z_2)(\beta - \bar{\beta}) + n_1 - n_2 \geq 0$$

$$\text{Var}(U_1 - U_2) = (Z_1 - Z_2) \Sigma_\beta (Z_1 - Z_2)^T + \sigma^2_1 + \sigma^2_2$$

$$\text{Cov}(U_1 - U_2, U_1 - U_3) = (Z_1 - Z_2) \Sigma_\beta (Z_1 - Z_3)^T + \sigma^2_1$$

$$\rho = \text{corr}(U_1 - U_2, U_1 - U_3) = \frac{\text{cov}(U_1 - U_2, U_1 - U_3)}{\sqrt{\text{Var}(U_1 - U_2)} \sqrt{\text{Var}(U_1 - U_3)}}$$

Want to know $\Pr[U_1 \geq t_1, U_2 \geq t_2]$

$$= \Pr[U_1 - U_2 \geq 0 \text{ and } U_1 - U_3 \geq 0]$$

$$= \Pr[(Z_1 - Z_2)\bar{\beta} + \sqrt{\text{Var}(U_1 - U_2)} t_1 \geq 0$$

$$\text{and } (Z_1 - Z_3)\bar{\beta} + \sqrt{\text{Var}(U_1 - U_3)} t_2 \geq 0]$$

specify correlation later.

t_1, t_2 joint

standard normal

$$\Pr[\text{choose } i] = \int \int f(t_1, t_2) dt_2 dt_1$$

\downarrow joint std normal

Limits

$$= \int_{-\infty}^{\frac{(Z_1 - Z_2)\bar{\beta}}{\sqrt{\text{Var}(U_1 - U_2)}}} \int_{-\infty}^{\frac{(Z_1 - Z_3)\bar{\beta}}{\sqrt{\text{Var}(U_1 - U_3)}}} f(t_1, t_2) dt_2 dt_1$$

$$\int(\bar{\beta}, \Sigma_\beta, \sigma^2_1, \sigma^2_2, \sigma^2_3) = \prod_j \prod_i [\Pr[\text{choose } i]]^{D_{ij}}$$

Small / Winston / Yan

"Revealed preference"

Not enough variation.

F or X
cost on X
relative time
time of day → relative 80-50 percentile
in time

"Stated preference"

Variables:

c_{it} : toll difference
 R_{it} : unreliability difference
 T_{it} : time difference

Values of travel time

$$\frac{\partial V/\partial T}{\partial V/\partial C} = \frac{\partial V/\partial T}{\lambda} = \frac{\text{utils/time}}{\text{utils/$}} = \$/\text{time}$$

$$V_1 = v_1 + \epsilon_1$$

$$V_2 = v_2 + \epsilon_2$$

$$\Pr(1|1,2) = \frac{e^{V_1}}{e^{V_1} + e^{V_2}} = \frac{e^{v_1 - v_2}}{1 + e^{v_1 - v_2}}$$

$$\Pr(2|1,2) = \frac{1}{1 + e^{v_1 - v_2}}$$

$$v_1 - v_2 = u_i - n_i$$

$u_i \sim \text{FEV}$
 $n_i \rightarrow \text{Logit.}$

$$\Pr(V_1 > V_2) = \Pr(\dots)$$

$$u_{it} = \theta_i + \beta_i K_{it}$$

choice chars

$$VOT_i = \frac{\partial u_{it}}{\partial T_{it}} = \frac{\partial u_{it}}{\partial v_{it}}$$

$$VOR_i = \frac{\partial u_{it}}{\partial R_{it}} = \frac{\partial u_{it}}{\partial v_{it}}$$

Heterogeneity: $\theta_i = \bar{\theta} + \phi u_i + \epsilon_i \sim N(0, \sigma^2_\epsilon)$
 $\beta_i = \bar{\beta} + r z_i + \beta_i \sim N(0, \sigma^2_\beta)$

Some more assumptions on error structure, final system

Estimation.

+ express \rightarrow freeway.

Example for two choices: logit $i = X \quad z = F$

$$U_1 = X_{1i}\beta + \eta_1 \rightarrow \Pr[X] = \Pr[X_{1i} - X_{2i}\beta > n_z - n_1] \\ U_2 = X_{2i}\beta + \eta_2 = \frac{e^{(X_1 - X_2)\beta}}{1 + e^{(X_1 - X_2)\beta}}$$

$$f = \prod_j \left[\frac{e^{(X_{1j} - X_{2j})\beta}}{1 + e^{(X_{1j} - X_{2j})\beta}} \right]^{D_j} \left[\frac{1}{1 + e^{(X_{1j} - X_{2j})\beta}} \right]^{1 - D_j}$$

Their model

$$L = \prod_j \int_{\Theta} \Pr[X | \phi, \theta]^{D_j} \Pr[F | \phi, \theta]^{1 - D_j} f(\theta) d\theta$$

n choices independent
conditional on
the random effects
from Θ .

Individuals obs
within individual

$$\Pr[X | \phi, \theta] = F[\theta^{\text{BR}} + \phi^{\text{BK}} w_i + \beta_i^{\text{BK}} X_i^{\text{BK}} + v_i^{\text{BR}}]$$

or
 $F[\beta^{\text{BS}}]$.

c does not enter integral, only one obs, not correlated.

Algorithm

(1) θ^0
(2) Simulate L , get θ^+ by gradient based method.

Stop when $|\theta^{k+1} - \theta^k| < \text{stopping crit.}$

Policy paper \rightarrow NOV simulations