

Midterm exam

Empirical Industrial Organization

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This exam is inspired by Luco (2019).

Individuals in their working age make savings decisions for retirement. We approach the worker's decision problem in a static manner, i.e., the worker makes a one-time decision.

One Level Decision

The worker must choose one of J pension fund administrators (PFA) to manage her retirement savings. The worker is mandated to save 10% of her salary, y_i . Each PFA charges a percentage fee, p_j , over the worker's salary. PFAs differ in their return on investment, R_j . Finally, ε_{ij} is an i.i.d. Type I Extreme Value preference shock. With all this in hand, we can write down the indirect utility that worker i obtains from enrolling in PFA j as

$$u_{ij} = \alpha_i (y_i - 0.1y_i - p_j y_i) + \beta_i R_j + \varepsilon_{ij}, \quad (1)$$

where α_i and β_i denote random coefficients. Moreover, let $\gamma_i = (\alpha_i, \beta_i)$, and $\gamma = (\alpha, \beta)$. We assume,

$$\gamma_i = \gamma + \Gamma D_i + \nu_i, \quad (2)$$

where D_i is a $d \times 1$ vector of demographic variables, Γ is a $2 \times d$ matrix of coefficients that measure how taste varies with demographics, and ν_i is a 2×1 vector of unobserved individual characteristics determining taste, where $\nu_i \sim F_\nu(\cdot)$, with $F_\nu(\cdot)$ a distribution function.

Following Train (2009, 3.1), we can write the probability that worker i chooses PFA j , s_{ij} , as follows:

$$s_{ij} = P(u_{ij} \geq u_{ik}, \forall j \neq k) \quad (3)$$

$$= P(\varepsilon_{ij} - \varepsilon_{ik} \geq \omega_{ik} - \omega_{ij}) \quad (4)$$

$$= \frac{\exp(\omega_{ij})}{\sum_{j \in J} \exp(\omega_{ik})} \quad (5)$$

where $\omega_{ij} = \beta_i R_j - \alpha_i p_j y_i$.

Additionally, we can express the price elasticity of the demand for PFA j , η_{j,p_j} , and the cross return elasticity of the demand for PFA j with respect to the return of PFA k (with $j \neq k$), η_{j,R_k} , like so

$$\eta_{j,p_j} = \frac{p_j}{s_{ij}} \cdot \frac{\partial s_{ij}}{\partial p_j} \quad (6)$$

$$= p_j(1 - s_{ij}) \cdot \frac{\partial \omega_{ij}}{\partial p_j} \quad (7)$$

$$= -\alpha_i p_j y_i (1 - s_{ij}), \quad (8)$$

$$(9)$$

$$\eta_{j,R_k} = \frac{R_k}{s_{ij}} \cdot \frac{\partial s_{ij}}{\partial R_k} \quad (10)$$

$$= -R_k s_{ik} \cdot \frac{\partial \omega_{ik}}{\partial R_k} \quad (11)$$

$$= -\beta_i R_k s_{ik}. \quad (12)$$

as derived in Train (2009, 3.6).

Subsequently, with this setup, we can describe the log-likelihood of the model, $\mathcal{L}(\cdot)$, such that

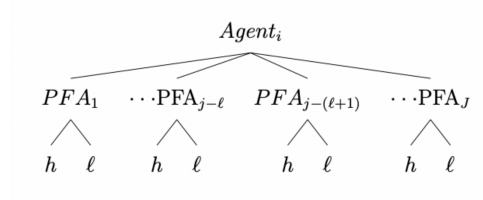
$$\mathcal{L}(\cdot) = \sum_i \sum_j \mathbf{1}_{ij} \log(s_{ij}) \quad (13)$$

$$= \sum_i \sum_j \left(\frac{\exp(\omega_{ij})}{\sum_{k \in J} \exp(\omega_{ik})} \right). \quad (14)$$

Two Level Decision

Now, suppose that each PFA offers two portfolios that workers can choose where to invest their savings in: a high-return, high-risk portfolio, h ; and low-return, low-risk portfolio, l .

We model the worker's choice of PFA and portfolio as a sequential decision, where individuals first choose the PFA, and then, conditional on the PFA, they choose the portfolio. The following diagram depicts the sequential two-level decision of the worker,



In this context, we model the indirect utility of choosing PFA $j = 1, \dots, J$ and portfolio $g = h, l$ as,

$$u_{ijg} = \alpha(y_i - 0.1y_i - p_j y_i) + \beta R_{jg} + \theta C_{jg} + \varepsilon_{ijg}, \quad (15)$$

where now, for simplicity, α and β are not random but homogeneous coefficients across individuals. The variable R_{jg} denotes the return of portfolio g in PFA j and, similarly, C_{jg} denotes the risk of portfolio g in PFA j . Finally, ε_{ijg} is an i.i.d. GEV preference shock such that the model is a two-level nested logit.

Note that the indirect utility can be rewritten as $u_{ijg} = u_{ij} + u_{ig|j} + \varepsilon_{ijg}$, where

$$u_{ij} = \alpha(y_i - 0.1y_i - p_i y_i) \quad (16)$$

is the indirect utility associated to choosing a PFA (i.e., first level), and

$$u_{ig|j} = \beta R_{jg} + \theta C_{jg} \quad (17)$$

is the indirect utility associated to choosing a portfolio conditional on a particular PFA (i.e., second level). Note that $u_{ig|j} = u_{g|j}$, $\forall i$.

Furthermore, as shown in Train (2009, 4.2.3), the probability of choosing PFA j and portfolio g , s_{ijg} , can be expressed as the product of a marginal probability of choosing PFA j and a conditional probability of choosing portfolio g conditional on PFA j . That is, $s_{ijg} = s_{ij} \cdot s_{ig|j}$, where

$$s_{ij} = \frac{\exp(u_{ij} + \lambda_j I_{ij})}{\sum_j \exp(u_{ij} + \lambda_j I_{ij})} \quad (\text{upper level model}) \quad (18)$$

and

$$s_{ig|j} = \frac{\exp\left(\frac{u_{ig|j}}{\lambda_j}\right)}{\sum_g \exp\left(\frac{u_{ig|j}}{\lambda_j}\right)} \quad (\text{lower level model}) \quad (19)$$

with

$$I_{ij} = \log \sum_g \exp\left(\frac{u_{ig|j}}{\lambda_j}\right) \quad (\text{inclusive value}) \quad (20)$$

Bear in mind that λ_j must be in the $[0, 1]$ interval as a sufficient condition that characterizes the correlation of utilities that a consumer experiences among the products in the same group j .

Finally, we can express the price elasticity of the demand for PFA j , η_{j,p_j} , and the cross price elasticity of the demand for PFA j with respect to the price of PFA k (with $j \neq k$), η_{j,p_k} , like so

$$\eta_{j,p_j} = \frac{p_j}{s_{ijg}} \cdot \frac{\partial s_{ijg}}{\partial p_j} \quad (21)$$

$$= \frac{p_j}{s_{ijg}} \left(s_{ig|j} \cdot \frac{\partial s_{ij}}{\partial p_j} + s_{ij} \cdot \frac{\partial s_{ig|j}}{\partial p_j} \right) \quad (22)$$

$$= p_j \left(\frac{s_{ig|j}}{s_{ijg}} \right) \frac{\partial s_{ij}}{\partial p_j} \quad (23)$$

$$= p_j \left(\frac{s_{ig|j}(1 - s_{ij})s_{ij}}{s_{ijg}} \right) \frac{\partial u_{ij}}{\partial p_j} \quad (24)$$

$$= -\alpha p_j y_i (1 - s_{ij}), \quad (25)$$

$$(26)$$

$$\eta_{j,p_k} = \frac{p_k}{s_{ijg}} \cdot \frac{\partial s_{ijg}}{\partial p_k} \quad (27)$$

$$= \frac{p_k}{s_{ijg}} \left(s_{ig|j} \cdot \frac{\partial s_{ij}}{\partial p_k} + s_{ij} \cdot \frac{\partial s_{ig|j}}{\partial p_k} \right) \quad (28)$$

$$= p_k \left(\frac{s_{ig|j}}{s_{ijg}} \right) \frac{\partial s_{ij}}{\partial p_k} \quad (29)$$

$$= -p_k \left(\frac{s_{ig|j} \cdot s_{ij} \cdot s_{ik}}{s_{ijg}} \right) \frac{\partial u_{ik}}{\partial p_k} \quad (30)$$

$$= \alpha p_k y_i s_{ik}. \quad (31)$$