Structural Microeconometric Methods:

Problem Set 3: Estimating an Optimal Stopping Model of Schooling

Data

- The Data come from *Génération 98*, a large scale survey conducted in France by Céreq.
- It was already described in the Problem Set devoted to the analysis of the grade progression model (hazard function analysis)

Model Structure:

- Two states
 - school (every individual starts there)
 - Full Time work
 - t = 1, 2, ...T periods to allocate (T=60 years)
- ullet Schooling is constructed as follows: s=1,2,S grade levels
 - -*s=1:Lycee (entre), duree= 1 an
 - * s = 2: lycee (CAP/BEP), duree=1an

- * s = 3 : lycee (bacc), duree=1an
- * s = 4: Enseignement sup (sans diplome), duree=1
- * s = 5:Bacc+2, duree= 1 an
- * s = 6: Bacc+3: duree = 1 an
- * s = 7: Bacc+4, duree= 1an
- * s = 8 : Bacc+5 (ou plus)=1 an
- \bullet For each grade level, s,
 - Assume a standard # of years for duration of grade s : In general, it will be 1 year.

- Transform each grade into a quantitative measure (say Bacc+2=?? years of schooling)
- Assume the "normal" age level of entry in the market (A_s) : In other words, and individual who completes grade s, must enter the labor market at age A_s (in case he does not go further)
- In general, at any date/age t, accumulated experience is $t-A_s$.
- At final date T, Grade-level-specific maximum) experience, is $T-A_s=60-A_s$
- Before estimating the model: estimate the log(earnings) regression using earnings data (schooling, experience, repeat indicator)

The Utility Equations

$$U_i^s(.) = X_i'\delta_x + \delta_L \cdot I(repeat) + \varepsilon_i^s$$

with

$$\varepsilon_i^s \sim i.i.dN(0, \sigma = 1)$$

$$U_i^w(s,t) = \ln(w_i). = \varphi_1 + \varphi_2 \cdot educ(years) + \varphi_3 \cdot (t - A_s) + \varphi_4 \cdot I(repeat)$$

where the variable t denotes age.

In last period, the utility of work is:

$$U_i^w(s,60) = \ln(w_i). = \varphi_1 + \varphi_2 \cdot educ(years) + \varphi_3 \cdot (60 - A_s) + \varphi_4 \cdot I(repeat)$$

Note that because we are using numbers that have been estimated prior to estimation of the model, $U_i^w(s, 60), U_i^w(s, 59), \dots$ contains only known constants.

Value of Attending Grade s:

$$V^{s}(s) = X'_{i}\delta_{x} + \delta_{L} \cdot I(repeat) + \varepsilon_{i}^{s}$$

+\beta E(Max\{V^{s}(s+1), V^{w}(s, A_{s})\})

Value of entering the labor market with grade s, at age A_s

$$V^{w}(s, A_{s}) = \varphi_{1} + \varphi_{s} \cdot I(s) + \varphi_{4} \cdot I(repeat)$$

$$+ \beta \{\varphi_{1} + \varphi_{2} \cdot s + \varphi_{3} \cdot (t - A_{s}) + \varphi_{4} \cdot I(repeat)$$

$$+ \dots$$

$$+ \beta^{(60 - A_{s} - 1)} \{\varphi_{1} + \varphi_{2} \cdot s + \varphi_{3} \cdot (T - A_{s}) + \varphi_{4} \cdot I(repeat)$$

$$= \sum_{j=A_{s}}^{60} \beta^{j-A_{s} - 1} \{\varphi_{1} + \varphi_{2} \cdot s + \varphi_{3} \cdot (j - A_{s}) + \varphi_{4} \cdot I(repeat) \}$$

• start from s = 8

– it is clear that the value function of entering grade 8 (and automatically starting full-time work at age A_8 since grade 8 is terminal), $V^s(8)$, has a trivial structure, since no Emax(.) needs to be taken:

$$V^{s}(8) = X'_{i}\delta_{x} + \delta_{L} \cdot I(repeat) + \varepsilon_{i}^{8} + V^{w}(8, A_{8})$$

• The future utility associated to entering grade 8, $V^w(8, A_8)$, is trivial (sum of life cycle earnings in logs from age A_8 to age 60.

• When contemplating entering grade 8, the choice requires a comparison between $V^s(8)$ and $V^w(7, A_7)$

• So, the choice probability is:

$$\begin{aligned} \Pr(d_8 &= 1) = \Pr V^s(8) > V^w(7, A_7) \\ &= \Pr(X_i' \delta_x + \delta_L \cdot I(repeat) + \varepsilon_i^8 + V^w(8, A_8) > V^w(7, A_7) \\ &= \Pr\{\varepsilon_i^8 > V^w(7, A_7) - V^w(8, A_8) - X_i' \delta_x - \delta_L \cdot I(repeat)\} \\ &= \Phi\{V^w(7, A_7) - V^w(8, A_8) - X_i' \delta_x - \delta_L \cdot I(repeat)\} \end{aligned}$$

Obviously, the choice of not attending grade 8 (stopping at grade 7) is

$$\Pr(d_8 = 0) = \Pr V^s(8) \le V^w(7, A_7)$$

$$= 1 - \Phi\{V^w(7, A_7) - V^w(8, A_8) - X_i'\delta_x - \delta_8 - \delta_L \cdot I(repeat)\}$$

- When we will solve the problem recursively, the first non-trivial exercise will be when contemplating grade 7:
 - you will need to compute $EMax\{V^s(s+1), V^w(s)\}$
 - or, in other words, $EMax\{V^s(8), V^w(7, A_7)\}$
 - to reduce notation burden, denote $EMax\{V^s(8), V^w(7, A_7)\}$ by EV(8) since this choice is made at the beginning of period 8,

$$EV(8) = pr(d_8 = 1) \cdot E(V^s(8) | V^s(8) > V^w(7, A_7))$$

$$+pr(d_8 = 0) \cdot E(V^w(7, A_7) | V^w(7, A_7) > V^s(8))$$

$$= pr(d_8 = 1) \cdot E\{V^s(8) | V^s(8) > V^w(7, A_7) \}$$

$$+pr(d_8 = 0) \cdot E(V^w(7, A_7)$$

Note that

- $V^w(7,A_7)$ is non stochastic, so $E(V^w(7,A_7)=V^w(7,A_7)$ and is already known.
- only need to evaluate $E(V^{s}(8) | V^{s}(8) > V^{w}(7, A_{7})$
- Formally, this is

$$E(V^{s}(8) \mid V^{s}(8) > V^{w}(7, A_{7}) = E\{X'_{i}\delta_{x} + \delta_{L} \cdot I(repeat) + \varepsilon_{i}^{8} + V^{w}(8, A_{8}) \mid X'_{i}\delta_{x} + \delta_{L} \cdot I(repeat) + \varepsilon_{i}^{8} + V^{w}(8, A_{8}) > V^{w}(7, A_{7})$$

This is equal to

$$= X_i'\delta_x + \delta_8 + \delta_L \cdot I(repeat) + V^w(8, A_8)$$

+ $E(\varepsilon_i^8 \mid \varepsilon_i^8 > V^w(7, A_7) - V^w(8, A_8) - X_i'\delta_x - \delta_L \cdot I(repeat)$ }

- Recall that $\varepsilon_i^8 \sim N(\mu, \sigma^2)$
- $\mu = 0, \sigma^2 = 1$
- $E(\varepsilon_i^8 \mid \varepsilon_i^8 > c) = \mu + \frac{\phi(c-\mu)/\sigma}{1-\Phi(c-\mu)/\sigma}$

- where $c = V^w(7, A_7) V^w(8, A_8) X_i'\delta_x \delta_8 \delta_L \cdot I(repeat)$
- Keep EV(8) in memory
- Proceed recursively until you reach the initial period (say the minimum schooling).
- In the preceding period/choice, we get
 - Compare $V^s(7)$ with $V^w(6, A_6)$
 - where $V^s(7)$ is simply

$$V^{s}(7) = X'_{i}\delta_{x} + \delta_{L} \cdot I(repeat) + \varepsilon_{i}^{7} + E \max\{V^{s}(8), V^{w}(7, A_{7})\}$$
$$= X'_{i}\delta_{x} + \delta_{L} \cdot I(repeat) + \varepsilon_{i}^{7} + EV(8)$$

where EV(8) is now given

Compute

*
$$Pr(d_7 = 1)$$

*
$$Pr(d_7 = 0)$$

*
$$E(V^s(7) \mid V^s(7) > V^w(6, A_6)$$

- Compare $V^s(6)$ with $V^w(5, A_5)$
 - etc.....

- Estimation is achieved by maximum likelihood.
- ullet the probability of having reached grade level \tilde{s} , is

$$L^{s} = Pr[(d_{1} = 1), (d_{2} = 1)....(d_{\tilde{s}} = 1), (d_{\tilde{s}+1} = 0)]$$

= $Pr(d_{1} = 1) \cdot Pr(d_{2} = 1).... Pr(d_{\tilde{s}} = 1) \cdot Pr(d_{\tilde{s}+1} = 0)$

except for the probability of having reached grade level 8, which is

$$L^s = Pr[(d_1 = 1), (d_2 = 1)....(d_{\tilde{s}} = 1), (d_8 = 1)]$$

= $Pr(d_1 = 1) \cdot Pr(d_2 = 1).... Pr(d_{\tilde{s}} = 1) \cdot Pr(d_8 = 1)$

To estimate wages, we assume that observed wages (earnings) are expressed as:

 $\ln(w_i^{obs}) = \varphi_1 + \varphi_2 \cdot educ + \varphi_3 \cdot (60 - A_s) + \varphi_4 \cdot I(repeat) + \varepsilon_{it}^m$ where ε_{it}^m is measurement error (i.i.d. across time and individuals).

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Due date: May 2017

- 1. Estimate the structural parameters $\{\delta_x, \delta_L, \beta\}$ by maximum likelihood. Compute the mean level of schooling in the population and the predicted distribution by years (grade)?
- 2. Answer the following question: Suppose that following technological change, the effect of education on log wages is multiplied by 2, what will happen to changes in mean schooling?