

## Homework 6

(Due: Wednesday, November 29 at the *start* of the class)

Note: Study groups *discussing* the problems are strongly encouraged. But please write your own answers and submit your own programs (no copy and past from your classmate's program!).

### 1 Structural Dynamic Discrete Choice Model

Consider a dynamic model of import decision:

$$V(\omega_{t-1}, d_{t-1}) = \max_{d_t \in \{0,1\}} W(\omega_{t-1}, d_{t-1}, d_t) + \xi_t(d_t), \quad (1)$$

$$W(\omega_{t-1}, d_{t-1}, d_t) = E[\tilde{\pi}_\theta(\omega_t, d_t) - FC_\theta(d_t, d_{t-1}) + \beta V(\omega_t, d_t) | \omega_{t-1}, d_{t-1}]. \quad (2)$$

where  $\omega_t$  is productivity,  $(\xi_t(0), \xi_t(1))'$  is import cost shock,  $d_t$  is discrete import decision,  $\pi_t(\cdot)$  is the profit after maximizing out the variable inputs,

$$FC_\theta(d_t, d_{t-1}) = (\delta_1 + \delta_2(1 - d_{t-1}))d_t$$

is the fixed cost of importing materials where  $\delta_1$  is per-period fixed cost while  $\delta_2$  is one-time sunk cost of importing.

The stochastic process of  $\omega_t$  and  $(\xi_t(0), \xi_t(1))$  are given by

$$\omega_t = \rho\omega_{t-1} + \epsilon_t,$$

$$\epsilon_t \sim_{iid} N(0, \sigma_\epsilon^2),$$

$$(\xi(0), \xi(1)) \sim_{iid} \text{Type 1 Extreme value distribution.}$$

We specify  $\tilde{\pi}_\theta(\omega_t, d_t) = \exp(\tilde{\alpha}_0 + \tilde{\alpha}_1\omega_t + \alpha_2d_t)$  and

$$E[\tilde{\pi}_\theta(\omega_t, d_t) | \omega_{t-1}, d_{t-1}] = E[\exp(\tilde{\alpha}_0 + \tilde{\alpha}_1\omega_t + \alpha_2d_t) | \omega_{t-1}, d_{t-1}] = \exp(\alpha_0 + \alpha_1\omega_{t-1} + \alpha_2d_t) \equiv \pi_\theta(\omega_{t-1}, d_t).$$

with  $\alpha_0 = \tilde{\alpha}_0 + 0.5\sigma_\epsilon^2$  and  $\alpha_1 = \tilde{\alpha}_1\rho$ , where the last equality follows from  $\omega_t = \rho\omega_{t-1} + \epsilon_t$  and  $E[\exp(\epsilon_t)] = \exp(0.5\sigma_\epsilon^2)$ .

Using the property of extreme value distributed variable, (2) is written as

$$V(\omega_{t-1}, d_{t-1}) = \text{Euler's constant} + \ln \left( \sum_{d \in \{0,1\}} \exp(W(\omega_{t-1}, d_{t-1}, d_t)) \right) \quad (3)$$

$$W(\omega_{t-1}, d_{t-1}, d_t) = \pi_\theta(\omega_{t-1}, d_t) - FC_\theta(d_t, d_{t-1}) + \beta \int V(\rho\omega_{t-1} + \epsilon', d_t)(1/\sigma_\epsilon)\phi(\epsilon'/\sigma_\epsilon)d\epsilon. \quad (4)$$

Once the fixed point of (7) is solved, then we can compute the conditional choice probabilities by the logit formula as:

$$\Pr_\theta(d_t = 1 | \omega_{t-1}, d_{t-1}) = \frac{\exp(\pi_\theta(\omega_{t-1}, 1) - FC_\theta(1, d_{t-1}) + \beta E_{\epsilon'}[V(\rho\omega_{t-1} + \epsilon', 1)])}{\sum_{d' \in \{0,1\}} \exp(\pi_\theta(\omega_{t-1}, d') - FC_\theta(d', d_{t-1}) + \beta E_{\epsilon'}[V(\rho\omega_{t-1} + \epsilon', d')])}. \quad (5)$$

Read “HW6\_data” into Matlab, which contains  $N \times (T + 1)$  panel data for firm’s discrete import decisions (“ $d_m$ ”) and productivity shock (“ $\omega$ ”), where there are  $N = 2000$  firms across  $T + 1 = 7$  years. Estimate the AR(1) process for  $\omega_{it}$  and let  $\hat{\rho}$  and  $\hat{\sigma}_\epsilon^2$  be the estimated parameters [You should get something close to  $\hat{\rho} = 0.6$  and  $\hat{\sigma}_\epsilon^2 = 0.07$ ]. In this exercise, you are asked to estimate the parameters  $\theta = (\alpha_0, \alpha_1, \alpha_2, \delta_1, \delta_2)'$  while fixing  $\hat{\rho}$  and  $\hat{\sigma}_\epsilon^2$  at the estimates of the AR(1) process by maximizing the log likelihood function:

$$\mathcal{L}(\theta | \{d_{it}, \omega_{it}\}_{i,t}) = \sum_{i=1}^N \sum_{t=1}^T \ln(\Pr_\theta(d_{it} | \omega_{i,t-1}, d_{i,t-1})), \quad (6)$$

where  $\Pr_\theta(d_{it} = 1 | \omega_{i,t-1}, d_{i,t-1})$  is given by (8).

1. Solve the fixed point of (7) by approximate the state space of  $\omega$  by a set of  $N_\omega = 121$  discrete points:  $\mathcal{W} = \{-1.5, -1.475, -1.45, \dots, -0.025, 0, 0.025, \dots, 1.45, 1.475, 1.5\}$  (i.e., set  $\mathcal{W} = [-1.5 : 0.025 : 1.5]$ ). Let  $\omega^1 = -1.5$ ,  $\omega^2 = -1.475$ , ...,  $\omega^{120} = 1.475$ , and  $\omega^{N_\omega} = 1.5$ . Let  $q^j = (\omega^j + \omega^{j+1})/2$  for  $j = 1, \dots, N_\omega - 1$  so that  $q^j$  is the middle point between  $\omega^j$  and  $\omega^{j+1}$ . The AR(1) process of  $\omega$  is then approximated on the grid as:

$$\Pr(\omega_t = \omega^j | \omega_{t-1} = \omega^i) = \begin{cases} \Phi((q^1 - \rho\omega^i)/\sigma_\epsilon), & \text{if } j = 1 \\ \Phi((q^j - \rho\omega^i)/\sigma_\epsilon) - \Phi((q^{j-1} - \rho\omega^i)/\sigma_\epsilon), & \text{if } 1 < j < N_\omega \\ 1 - \Phi((q^{N_\omega} - \rho\omega^i)/\sigma_\epsilon), & \text{if } j = N_\omega. \end{cases}$$

Compute  $\Pr(\omega_t = \omega^j | \omega_{t-1} = \omega^i)$  for  $i, j = 1, \dots, N_\omega$  and construct a  $N_\omega \times N_\omega$  transition matrix, called  $M_\omega$ , of which  $(i, j)$ -th element is  $\Pr(\omega_t = \omega^j | \omega_{t-1} = \omega^i)$ . We set  $\beta = 0.9$ .

2. Once the state space is discretized, both the value function  $V$  and the conditional choice probabilities  $\Pr_\theta(d_t = 1 | \omega_{t-1}, d_{t-1})$  can be expressed by  $N_\omega \times 2$  matrices. Write a function m-file which takes  $\theta$ ,  $M_\omega$ , and  $\{\omega^1, \dots, \omega^{N_\omega}\}$  as inputs and then produces the fixed point  $V$  and the conditional choice probability  $\Pr_\theta(d_t = 1 | \omega^j, d_{t-1})$  for  $j = 1, \dots, N_\omega$ .
3. “Discretize” the data on productivity shock,  $\{\omega_{it}\}$ , by setting  $\omega_{it}$  equal to the closest grid point among  $\{\omega^1, \dots, \omega^{N_\omega}\}$ , i.e.,

$$\omega_{it}^* = \begin{cases} \omega^1, & \text{if } \omega_{it} \leq q^1 \\ \omega^j, & \text{if } q^j < \omega_{it} \leq q^{j+1} \text{ for some } 1 < j < N_\omega \\ \omega^{N_\omega}, & \text{if } q^{N_\omega} < \omega_{it}. \end{cases}$$

4. Write a function m-file which takes  $\theta$ ,  $M_\omega$ , and  $\{\omega^1, \dots, \omega^{N_\omega}\}$ ,  $\{\omega_{it}^*\}$ , and  $\{d_{it}\}$  as inputs and produces the negative value of log-likelihood function (6) as output. Estimate  $\theta$  by the maximum likelihood. Report the estimates and their standard errors. You can also use bootstrap.
5. The conditional choice probabilities  $\Pr_\theta(d_t = 1 | \omega_{t-1}^*, d_{t-1})$  and the transition matrix of  $\omega_t^*$  jointly characterize the stochastic process of  $(d_t, \omega_t^*)$ . Using the estimate  $\hat{\theta}$  and the transition matrix  $M_\omega$ , compute the stationary joint distribution of  $(d_t, \omega_t^*)$  as well as the conditional distribution of  $d_t$  given  $\omega_t^*$  in the long-run.
6. Compute the empirical transition matrix of  $d_t$  and report it in the  $2 \times 2$  table, where the  $(1, 1)$ -th element is a fraction of firms with  $d_t = 0$  given  $d_{t-1} = 0$ . Compute the predicted transition matrix of  $d_t$  under the estimated parameters and compare it with the empirical transition probability. Does the estimated model successfully replicate the dynamic patterns of import status?
7. Suppose that a government unexpectedly and permanently introduces import subsidies in the form of one-time lump-sum transfer for the “first-time” importers at  $t = 0$  and, as a result,

the value of  $\delta_2$  decreases by 50 %. Analyze how a fraction of importers change over time after the introduction of import subsidies starting from the stationary distribution of  $(\omega_t^*, d_t)$  by plotting a fraction of importers as y-axis and time as x-axis.<sup>1</sup>

8. Now, suppose that a firm makes an import decision *after* observing  $\omega_t$  so that the Bellman equation is given by:

$$\begin{aligned} V(\omega_t, d_{t-1}) &= \max_{d_t \in \{0,1\}} W(\omega_t, d_{t-1}, d_t) + \xi_t(d_t), \\ W(\omega_t, d_{t-1}, d_t) &= \tilde{\pi}_\theta(\omega_t, d_t) - FC_\theta(d_t, d_{t-1}) + \beta E[V(\omega_{t+1}, d_t) | \omega_t, d_{t-1}]. \end{aligned}$$

so that

$$\begin{aligned} V(\omega_t, d_{t-1}) &= \text{Euler's constant} + \ln \left( \sum_{d_t \in \{0,1\}} W(\omega_t, d_{t-1}, d_t) \right) \\ W(\omega_t, d_{t-1}, d_t) &= \tilde{\pi}_\theta(\omega_t, d_t) - FC_\theta(d_t, d_{t-1}) + \beta \int V(\rho\omega_t + \epsilon', d_t) (1/\sigma_\epsilon) \phi(\epsilon'/\sigma_\epsilon) d\epsilon. \end{aligned}$$

The conditional choice probabilities are given by the logit formula as:

$$\Pr_\theta(d_t = 1 | \omega_t, d_{t-1}) = \frac{\exp(\tilde{\pi}_\theta(\omega_t, 1) - FC_\theta(1, d_{t-1}) + \beta E_{\epsilon'}[V(\rho\omega_t + \epsilon', 1)])}{\sum_{d' \in \{0,1\}} \exp(\tilde{\pi}_\theta(\omega_t, d') - FC_\theta(d', d_{t-1}) + \beta E_{\epsilon'}[V(\rho\omega_t + \epsilon', d')])}.$$

Estimate a parameter vector  $\theta = (\tilde{\alpha}_0, \tilde{\alpha}_1, \alpha_2, \delta_0, \delta_1)'$  and report the standard errors. [Note that the program for solving the Bellman equation is the same as before. When you evaluate the likelihood, you will replace  $\omega_{t-1}$  with  $\omega_t$ .]

## 2 Reduced Form vs. Structural Model

Consider the following model of demand and supply:

$$D_t = \alpha_1 p_t + \alpha_2 y_t + u_{Dt} \tag{7}$$

$$S_t = \beta_1 p_t + \beta_2 w_t + u_{St}. \tag{8}$$

---

<sup>1</sup>You can do this either by generating the “counterfactual” data set with the estimated model as in homework 2 or use the estimated transition probability of  $(d_t, \omega_t)$ .

We assume that  $(u_{Dt}, u_{St})$  is independent of  $(y_t, w_t)$ , and independently distributed over time. In equilibrium, we have  $q_t = D_t = S_t$ . We observe the random sample of  $\{q_t, p_t, y_t, w_t\}_{t=1}^T$  generated from this equilibrium model. For your reference, we may define

$$\Gamma = \begin{pmatrix} 1 & -\alpha_1 \\ 1 & -\beta_1 \end{pmatrix} \quad \text{and} \quad \Gamma^{-1} = \begin{pmatrix} \frac{-\beta_1}{\alpha_1 - \beta_1} & \frac{\alpha_1}{\alpha_1 - \beta_1} \\ \frac{-1}{\alpha_1 - \beta_1} & \frac{1}{\alpha_1 - \beta_1} \end{pmatrix}.$$

1. Suppose that we consider the following reduced-form specification:

$$q_t = \pi_{11}y_t + \pi_{12}w_t + \tilde{u}_{qt}. \tag{9}$$

Express  $\pi_{11}$ ,  $\pi_{12}$ , and  $\tilde{u}_{qt}$  in terms of the structural parameter  $\theta = (\alpha_1, \alpha_2, \beta_1, \beta_2)$  and the iid demand and supply shocks,  $u_{Dt}$  and  $u_{St}$ .

2. Consider the effect of counterfactual policy of consumption tax so that the consumer faces  $(1 + \tau)p_t$  instead of  $p_t$  in their demand equation while the supply equation does not change. The tax revenue is thrown away by the government. Let  $q_t^{CF}$  be the counterfactual equilibrium quantity, where we assume that  $(y_t, w_t, u_{Dt}, u_{St})$  in this counterfactual remains the same as the actual value. Let  $q_t^{CF} - q_t$  be the effect of introducing consumer tax.

- (a) Briefly discuss if the reduced-form parameter in (9) is enough to analyze the effect of the counterfactual consumption tax on equilibrium quantity without knowing the relationship between the reduced-form parameter  $(\pi_{11}, \pi_{12})$  and the structural parameter  $\theta$ .
- (b) Derive  $q_t^{CF} - q_t$  in terms of the structural parameter  $\theta$  and the iid demand and supply shocks,  $u_{Dt}$  and  $u_{St}$ .
- (c) Discuss how to compute the standard error for  $E[q_t^{CF} - q_t]$ .