# Estimate Dynamic Models with CCP

## Problem Set 1 Empirical Industrial Organization 2025 Srping

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Utility (Profit) The profit function of the Harold Zurcher is

$$\pi(s, Y) = (-RC)Y + (-\mu s)(1 - Y) = (-RC + \mu s)Y - \mu s$$

The realized profit is

$$\Pi(s, Y, \nu_0, \nu_1) = u(s, Y) + \nu_Y$$

Since only the difference of profit matters for the replacement decision, we can normalize such that

$$u(s, Y) = (-RC + \mu s)Y$$
  

$$v(s, Y, \nu) = \Pi(s, 1, \nu_0, \nu_1) - \Pi(s, 0, \nu_0, \nu_1)$$
  

$$= (-RC + \mu s)Y + \nu$$

Variables For observed variables, we have

- s: the state variable, the mileage at the end of period t
- Y: the decision variable, the decision to replace the bus at the end of period t

For **unobserved variables**, we have  $\nu$ .

**State transition** The state variable s evolves according to this transition function

$$s_{t+1} = s_t + \epsilon_{t+1} \quad \epsilon_{t+1} \sim N(\rho, \sigma_{\rho}^2)$$

**Parameters** The profit function parameters:

- $\mu$ : the cost of maintaining the bus
- RC: the replacement cost of the bus

The state transition parameters:  $\rho$  and  $\sigma_{\rho}^2$ . Discount factor  $\beta = 0.99$  is given.

Value function The value function is

$$V(s,\nu) = \max_{Y} \{ v(s,Y,\nu) + \beta \mathbb{E}[V(s',\nu')|s,Y] \}$$
  
=  $\max_{Y} \{ u(s,Y) + \nu_{Y} + \beta \int \mathbb{E}_{\nu'}[V(s',\nu')]f(s'|s)ds' \}$ 

To understand this value function, notice that at the end of each period t Harold Zurcher is essentially comparing the **continuation value** from replacing Y = 1 and that from not replacing Y = 0. We define the **continuation value** as  $\tilde{v}(s, Y, \nu)$ . The value function is the maximum of them.

$$\tilde{v}(s, 1, \nu) = \tilde{u}(s, 1) + \nu = (-RC + \mu s) + \nu + \beta \int \mathbb{E}_{\nu'} [V(s', \nu')|s, 1] f(s'|s) ds'$$

$$\tilde{v}(s, 0, \nu) = \tilde{u}(s, 0) = \beta \int \mathbb{E}_{\nu'} [V(s', \nu')|s, 0] f(s'|s) ds'$$

Therefore the value function can be rewritten as

$$V(s,\nu) = \max_{V} \left\{ \tilde{v}(s,1,\nu), \tilde{v}(s,0,\nu) \right\}$$

If we know the  $\tilde{u}(s, 1)$  and  $\tilde{u}(s, 0)$ , and assume  $\nu$  are iid from T1EV, we are back to the static logit discrete choice model such that

$$\Pr(Y = 1|s) = \frac{\exp(\tilde{u}(s,1))}{\exp(\tilde{u}(s,1)) + \exp(\tilde{u}(s,0))}$$

We follow Bajari et al. (2007) to approximate the continuation value  $\tilde{v}$  by

$$\tilde{v}(s,1,\nu) = \tilde{u}(s,1) + \nu = x_{i1}^1 \beta_1 + x_{i1}^2 \beta_2 + x_{i1}^3 + \nu$$
$$\tilde{v}(s,0,\nu) = \tilde{u}(s,0) = x_{i0}^1 \beta_1 + x_{i0}^2 \beta_2 + x_{i0}^3$$

where

- $x_i^1 = \frac{1}{K} \sum_{k=1}^K \sum_{t=1}^T Y_{it}$
- $x_i^2 = \frac{1}{K} \sum_{k=1}^K \sum_{t=1}^T s_{it} Y_{it}$
- $x_i^3 = \frac{1}{K} \sum_{k=1}^K \sum_{t=1}^T s_{it}$

Therefore, we obtain  $RC = -\beta_1, \mu = \beta_2$ .

The optimal  $\theta^*$  is the one that maximizes the likelihood function

$$\max_{\theta} \sum_{i=1}^{N} \sum_{t=1}^{T} \ln \Pr(Y_{it}|s_{it}, \theta)$$

### References

Bajari, P., Benkard, C. L., and Levin, J. (2007). Estimating dynamic models of imperfect competition. *Econometrica*, 75(5):1331–1370.