# Dynamic Structural Models for Policy Evaluation

## Sergi Quintana

Universitat Autònoma de Barcelona Barcelona School of Economics

BSE Summer School, 2024





### **Practical Sessions Outline**

- 1. Solving Value Functions
- 2. Estimating the model
- 3. Conditional Choice Probability (CCP) estimation
- 4. CCP estimation finite dependence





# Session 1: Solving Value Functions

BSE Summer School, 2024



### Session Outline

Introduction

Infinite Horizon

Alternative Scenarios



#### General framework

- ightharpoonup Discrete time, t = 0, 1, ..., T
- ightharpoonup Every period t individual chooses one of J mutually exclusive alternatives:

$$d_t = \{j : j \in D = \{1, 2, ..., J\}\}$$

where 
$$d_{jt} = 1\{d_t = j\}$$
 and  $\sum_{j=1}^{J} d_{jt} = 1$ 

ightharpoonup Payoff at t depends on state variables:

$$oldsymbol{s}_t = \{oldsymbol{x}_t, oldsymbol{arepsilon}_t\}$$

where  $\varepsilon_t$  is vector of **unobservables** and  $s_t$  follows **Markovian** process  $s_{t+1} \sim F(s_{t+1}|s_t, d_t)$ 

► Individual's intertemporal payoff:

$$\mathbb{E}\left[\sum_{\tau=0}^{T-t} \beta^{\tau} U(\boldsymbol{s}_{t+\tau}, d_{t+\tau})\right]$$



#### Assumptions

**Assumption 1. Additive Separability.** Utility function additively separable between observable and unobservable components:

$$U(d_t, \boldsymbol{x}_t, \boldsymbol{\varepsilon}_t) = u(d_t, \boldsymbol{x}_t) + \varepsilon_t(d_t)$$

Assumption 2. IID Unobservables. Unobserved variables  $\varepsilon_t$  are iid across agents and over time given  $x_t$ .

Assumption 3. Conditional Independence. Conditional on current decision and observable state variables, next period observables do not depend on  $\varepsilon_t$ :

$$F_x(\boldsymbol{x}_{t+1}|d_t, \boldsymbol{x}_t, \boldsymbol{\varepsilon}_t) = F_x(\boldsymbol{x}_{t+1}|d_t, \boldsymbol{x}_t)$$

$$F(\boldsymbol{x}_{t+1}, \boldsymbol{\varepsilon}_{t+1}|d_t, \boldsymbol{x}_t, \boldsymbol{\varepsilon}_t) = F_x(\boldsymbol{x}_{t+1}|d_t, \boldsymbol{x}_t)F_{\varepsilon}(\boldsymbol{\varepsilon}_{t+1})$$



### Assumptions

Assumption 4. Conditional Logit. Unobservables  $\varepsilon_{jt}$  are independent across alternatives and type-I extreme value distributed.

 $V_t(\boldsymbol{x}_t)$  - ex-ante value in period t, before  $\varepsilon_t$  is revealed.

$$V_t(\boldsymbol{x}_t) \equiv \mathbb{E}_{t-1} \left[ \sum_{\tau=0}^{T-t} \sum_{j \in D} \beta^{\tau} d_{jt+\tau}^*(u_{jt+\tau}(\boldsymbol{x}_{t+\tau}) + \varepsilon_{jt+\tau}) | \boldsymbol{x}_t \right]$$

where  $d_{jt+\tau}^* = 1$  if j is an optimal decision at period  $t + \tau$ .



### Empirical example

- ▶ Maintenance of electric bikes' batteries in Barcelona
- ▶ Bikes can be either **replaced** (R) or **maintained** (M)
- $ightharpoonup a_t$  age of the bike battery
- ► Utility:

$$u(a_t, d_t, \epsilon_t) = \begin{cases} -\theta_R + \epsilon_{1t} & \text{if } d_t = 1\\ -\theta_{M1} a_t - \theta_{M2} a_t^2 + \epsilon_{0t} & \text{if } d_t = 0 \end{cases}$$

where  $\epsilon_{0t}$  and  $\epsilon_{1t}$  are unobserved by econometrician, iid with all Rust assumptions

▶ A is a number of states (if A = 5 then  $a_t = 1, ..., 5$ ) and T is time horizon



### **Infinite Horizon**

#### Value functions

- 1. Formulate dynamic programming problem for the firm (assumptions, Bellman equation)
- 2. Solve the model for A = 5. Write system of 5 equations with 5 unknowns.
- 3. Find a fixed point for each of the equation.
- 4. Write down probabilities used later to construct Maximum Likelihood estimation.





### Infinite Horizon

#### Fixed Point Algorithm

- Well-known, basic algorithm for dynamic programming
- ► It always (sometimes slowly) converge
- ► Bellman operator:

$$\Gamma(V)(a) = \max_{d \in D} \left[ u(a, d) + \beta \int V(a') dF(a'|a) \right]$$

▶ Step 1. Specify  $V^0$  and apply Bellman operator:

$$V^{1}(a) = \max_{d \in D} [u(a, d) + \beta \int V^{0}(a')dF(a'|a)]$$

ightharpoonup Step 2. Iterate K times until convergence:

$$V^{K}(a) = \max_{d \in D} \left[ u(a, d) + \beta \int V^{K-1}(a') dF(a'|a) \right]$$





### Infinite horizon

#### Conditional value functions

Conditional value function given our assumptions:

$$v_j(a) = u_j(a_t) + \beta \int V(a_{t+1}) dF_a(a_{t+1}|a_t, j)$$

Given Type-I Extreme Value assumption Emax can be written as a function of  $v_i(a_t)$ :

$$V_t(a) = \ln \sum_{j \in D} \exp\{v_j(a)\} + \gamma$$

where  $\gamma$  is the Euler's constant.





### Infinite horizon

#### Empirical example

▶ In our case, depending on the state  $a_t$ :

$$V(1) = \ln(\exp\{v_1\} + \exp\{v_0(1)\}) + \gamma$$

$$V(2) = \ln(\exp\{v_1\} + \exp\{v_0(2)\}) + \gamma$$
...
$$V(5) = \ln(\exp\{v_1\} + \exp\{v_0(5)\}) + \gamma$$

▶ In our model, given 5 possible states (A=5) we have:

$$\begin{split} v_1 &= -\theta_R + \beta \ln(\exp\{v_0(1)\} + \exp\{v_1\}) + \beta \gamma \\ v_0(1) &= -\theta_{M1} - \theta_{M2} + \beta \ln(\exp\{v_0(2)\} + \exp\{v_1\}) + \beta \gamma \\ v_0(2) &= -2\theta_{M1} - 4\theta_{M2} + \beta \ln(\exp\{v_0(3)\} + \exp\{v_1\}) + \beta \gamma \\ \dots \\ v_0(5) &= -5\theta_{M1} - 25\theta_{M2} + \beta \ln(\exp\{v_0(5)\} + \exp\{v_1\}) + \beta \gamma \end{split}$$



# Alternative Scenarios

#### Examples

► Absorbing State

Example: Buying a house or marriage.

► Finite Horizon

Example: Lifecycle models.

► Finite Horizion + Absorbing State

Example. Lifecylce with retirement decisions or education models with drop out decisions.

They all have one thing in common: there is a terminal continuation value! Therefore, we will **not** solve these models by value function iteration.



### Finite Horizon

#### Empirical example

- ▶ Suppose now at period 5 we stop modeling the choices and agents get a terminal value depending on their state.
- ▶ Now the continuation value should be obtained by backwards induction.
- ▶ At period t = 5 we have  $V_t(a) = T(a)$ .
- ightharpoonup Then at period t:

$$v_{jt}(a) = u_{jt}(a) + \beta V_{t+1}(a)$$

► And we can iterate backwarsd using:

$$V_{t+1}(a) = \log(\exp(v_{t1}(a')) + \exp(v_{t0}(a')))$$

 $\blacktriangleright$  Where a' is next period a depening on today's choice.





## **Absorbing State**

#### Empirical example

- ▶ Suppose now, instead of replacing the batery we replace the bike.
- ▶ Bikes can be either **replaced** (R) or **maintained** (M)
- ▶ Due to safety regulation each bike has to be replace after 4 years
- ► This means that at period 5 all bikes have exited the model.
- $ightharpoonup a_{t+1} = a_t + 1$  if maintained
- $ightharpoonup a_{t+1} = 0$  if replaced
- ►  $a_5 = 0$
- ► The rest of the model remains unchanged





## **Absorbing State**

#### Solving for the continuation value

- $\blacktriangleright$  First assume a terminal continuation. In this case it is the terminal payoff that we get for changing the bike. T.
- $\blacktriangleright$  Write the conditional value functions related to T.

$$\begin{split} v_1 &= -\theta_R + \beta T \\ v_0(1) &= -\theta_{M1} - \theta_{M2} + \beta \ln(\exp\{v_0(2)\} + \exp\{v_1\}) + \beta \gamma \\ v_0(2) &= -2\theta_{M1} - 4\theta_{M2} + \beta \ln(\exp\{v_0(3)\} + \exp\{v_1\}) + \beta \gamma \\ v_0(3) &= -3\theta_{M1} - 9\theta_{M2} + \beta \ln(\exp\{v_1\}) + \beta \gamma \end{split}$$



