Recursive Methods

Introduction to Dynamic Optimization

Outline Today's Lecture

- neoclassical growth application: use all theorems
- constant returns to scale
- homogenous returns
- unbounded returns



Restrictions

- since F is unbounded is the $\sup < \infty$? is the max well defined?
- can we apply the Principle of Optimality?

1. restrict $\Gamma:$ for some α such that $\gamma\beta<1:$

$$y \in \Gamma(x) \implies ||y|| \le \alpha ||x||$$

"state can't grow too fast"

2. restrict F : for some $0 < B < \infty$

 $|F(x,y)| \le B(||x|| + ||y||) \text{ all } (x,y) \in A$

"some weak boundedness condition: only allow unboundedness along rays"

Implications

 $||x_t|| \leq \alpha^t ||x_0||$ for $x \in \Pi(x_0)$ all $x_0 \in X$

Thus:

$$|u_{n}(x) - u_{n-1}(x)| = \beta^{t} |F(x_{t}, x_{t+1})|$$

$$\leq \beta^{t} B(||x_{t}|| + ||x_{t+1}||)$$

$$= \beta^{t} B(\alpha^{t} ||x_{0}|| + \alpha^{t+1} ||x_{0}||)$$

$$= (\beta \alpha)^{t} B(1 + \alpha) ||x_{0}|| \to 0$$

so $u_n(x)$ is Cauchy $\implies u_n(x) \rightarrow u(x)$ So we have A1 and A2 \implies theorems 4.2 and 4.4

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supremum's properties

- we established that $v^*:X\to R$
- note that $u(\lambda x) = \lambda u(x)$ and $x \in \Pi(x_0) \implies \lambda x \in \Pi(\lambda x_0)$
- v^* must be homogenous of degree 1

$$v^{*} (\lambda x_{0}) = \sup_{\substack{x \in \Pi(\lambda x_{0})}} u(x)$$
$$= \sup_{\substack{\frac{x}{\lambda} \in \Pi(x_{0})}} u\left(\lambda \frac{x}{\lambda}\right)$$
$$= \lambda \sup_{\tilde{x} \in \Pi(x_{0})} u(\tilde{x})$$
$$= \lambda v^{*} (x_{0})$$

$$\begin{aligned} |u(x)| &= \left| \sum_{t=0}^{\infty} \beta^{t} F(x_{t}, x_{t+1}) \right| \\ &\leq \left| \sum_{t=0}^{\infty} \beta^{t} \left| F(x_{t}, x_{t+1}) \right| \\ &\leq \left| B \sum_{t=0}^{\infty} \beta^{t} \left(\alpha^{t} \| x_{0} \| + \alpha^{t+1} \| x_{0} \| \right) \right| \\ &\leq \left| B \| x_{0} \| \sum_{t=0}^{\infty} (\beta \alpha)^{t} (1 + \alpha) \right| \\ &= \left| \left[B \frac{1 + \alpha}{1 - \beta \alpha} \right] \| x_{0} \| \\ \Longrightarrow \left| v^{*}(x) \right| \leq c \| x_{0} \| \text{ for some } c \in R \end{aligned}$$

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What Space to Use?

 $H(X) = \begin{cases} f: X \to R: f \text{ is continuous and homogenous of degree 1} \\ \text{and } \frac{f(x)}{\|x\|} \text{ is bounded} \end{cases}$ $\|f\| = \sup_{\substack{x \in X \\ \|x\|=1}} |f(x)| = \sup_{x \in X} \frac{|f(x)|}{\|x\|}$

- H(X) is complete
- define operator $T:H(X) \rightarrow H(X)$

$$Tf(x) = \max_{y \in \Gamma(x)} \left\{ F(x, y) + \beta f(y) \right\}$$

Properties

• Operator $T: H(X) \to H(X)$

$$Tf(x) = \max_{y \in \Gamma(x)} \left\{ F(x, y) + \beta f(y) \right\}$$

• note that for any
$$v \in H(X)$$

$$\beta^{t} |v(x_{t})| \leq \beta^{t} c ||x_{t}|| \leq (\alpha \beta)^{t} c ||x_{0}|| \to 0$$

thus $\beta^t v(x_t) \to 0$ for all feasible plans (Theorems 4.3 and 4.5 apply) $\implies T$ has unique fixed point $v \in H(X)$

• is T is a contraction?

Is T a contraction?

- Modify Blackwell's condition (bounded functions) to show that T it is a contraction; approach in SLP
- Note that

$$\frac{\Gamma f}{|x||} = \max_{y \in \Gamma(x)} \left\{ \frac{1}{\|x\|} F(x, y) + \beta \frac{1}{\|x\|} f\left(\frac{y}{\|y\|} \|y\|\right) \right\} \\
= \max_{y \in \Gamma(x)} \left\{ F\left(\frac{x}{\|x\|}, \frac{y}{\|x\|}\right) + \beta \frac{\|y\|}{\|x\|} f\left(\frac{y}{\|y\|}\right) \right\}$$

- Idea: study related operator on functions space of continuous functions defined for $\|x\|=1$



Yes, T is a contraction!

- since \tilde{T} is a contraction of modulus $\alpha\beta < 1$

$$\sup_{\tilde{x}\in\tilde{X}} \left| \tilde{T}f - \tilde{T}g \right| \le \alpha\beta \sup_{\tilde{x}\in\tilde{X}} |f - g|$$

• for
$$f \in H(X)$$

$$\tilde{T}f = \frac{Tf}{\|x\|}$$

(note that $f \in H(X)$

• Thus

$$\sup_{x \in X} |Tf - Tg| = \|x\| \sup_{\tilde{x} \in \tilde{X}} \left| \tilde{T}f - \tilde{T}g \right| \le \alpha\beta \sup_{\tilde{x} \in \tilde{X}} |f - g| = \alpha\beta \sup_{x \in X} |f - g|$$

so T is a contraction on H(X)

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Nr. 12

Renormalizing

- studying a related operator is convenient in practice
 → reduces dimensionality!
- ||x|| = 1 not necessarily most convenient normalization ...
- ... another normalization (much used) if $x = (x^1, x^2) \in \mathbb{R}^n$ and $x^1 \in \mathbb{R}$ then use $x^1 = 1$

Homogenous Returns of Degree $\boldsymbol{\theta}$

similar tricks work (see Alvarez and Stokey, JET)

• rough idea for: $\theta > 0$

 $F\left(\lambda x,\lambda y\right) = \lambda^{\theta}F\left(x,y\right)$

 $\left|F\left(x,y\right)\right| \leq B\left(\left\|x\right\| + \left\|y\right\|\right)^{\theta} \text{ all } (x,y) \in A$

- Γ as before but now α such that $\gamma\equiv\beta\alpha^{\theta}<1$
- arguments are exactly parallel
- in particular, T is a contraction of modulus γ
- for $\theta < 0$ and $\theta = 0$ several complications with origin... but they can be surmounted

Unbounded Returns and Monotonicity

- numerically cannot handle unbounded returns
- idea: T may not be a contraction
 but all is not lost: it still is monotonic

Theorem 4.14

- 1. Start from $v_0 \ge v^*$
- 2. IF $Tv_0 = v_1 \leq v_0$ then define $v_n = T^n v_0$ (decreasing sequence)
- 3. **IF** $\lim_{n\to\infty} v_0(x_n) \leq 0$ all $x \in \Pi(x_0)$ all x_0 then clearly $v_n(x) \to v(x)$ for all $x \in X$, for some $v: X \to \overline{R}$
- 4. IF Tv = v (is this implied by $v_n \to v$?)

THEN $v = v^*$

• can be used for quadratic returns

Unbounded Returns and Monotonicity

Squeezing argument: 1. suppose $v_L(x) \le v^*(x) \le v^U(x)$ 2. and $T^n v^U(x) \rightarrow v$ and $T^n v^U(x) \rightarrow v$ THEN $v = v^*$

Next Class

- we're done with Chapter 4
- next class: deterministic dynamics
- Chapter 6
- Boldrin-Montruccio 1986 paper