

# Outline Today's Lecture

- finish Euler Equations and Transversality Condition
- Principle of Optimality: Bellman's Equation
- Study of Bellman equation with bounded  $F$
- contraction mapping and theorem of the maximum

## Infinite Horizon $T = \infty$

$$V^*(x_0) = \sup_{\{x_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t F(x_t, x_{t+1})$$

subject to,

$$x_{t+1} \in \Gamma(x_t) \quad (1)$$

with  $x_0$  given

- $\sup \{ \}$  instead of  $\max \{ \}$
- define  $\{x'_{t+1}\}_{t=0}^{\infty}$  as a plan
- define  $\Pi(x_0) \equiv \{ \{x'_{t+1}\}_{t=0}^{\infty} \mid x'_{t+1} \in \Gamma(x'_t) \text{ and } x'_0 = x_0 \}$

# Assumptions

A1.  $\Gamma(x)$  is non-empty for all  $x \in X$

A2.  $\lim_{T \rightarrow \infty} \sum_{t=0}^T \beta^t F(x_t, x_{t+1})$  exists for all  $x \in \Pi(x_0)$   
then problem is well defined

# Recursive Formulation: Bellman Equation

- value function satisfies

$$\begin{aligned} V^*(x_0) &= \max_{\substack{\{x_{t+1}\}_{t=0}^{\infty} \\ x_{t+1} \in \Gamma(x_t)}} \left\{ \sum_{t=0}^{\infty} \beta^t F(x_t, x_{t+1}) \right\} \\ &= \max_{x_1 \in \Gamma(x_0)} \left\{ F(x_0, x_1) + \max_{\substack{\{x_{t+1}\}_{t=1}^{\infty} \\ x_{t+1} \in \Gamma(x_t)}} \sum_{t=1}^{\infty} \beta^t F(x_t, x_{t+1}) \right\} \\ &= \max_{x_1 \in \Gamma(x_0)} \left\{ F(x_0, x_1) + \beta \max_{\substack{\{x_{t+1}\}_{t=1}^{\infty} \\ x_{t+1} \in \Gamma(x_t)}} \sum_{t=0}^{\infty} \beta^t F(x_{t+1}, x_{t+2}) \right\} \\ &= \max_{x_1 \in \Gamma(x_0)} \{ F(x_0, x_1) + \beta V^*(x_1) \} \end{aligned}$$

**continued...**

- idea: use BE to find value function  $V^*$  and policy function  $g$  [Principle of Optimality]

# Outline Today's Lecture

- housekeeping: ps#1 and recitation day/ theory general / web page
- finish Principle of Optimality:  
Sequence Problem  $\iff$  solution to Bellman Equation  
(for values and plans)
- begin study of Bellman equation with bounded and continuous  $F$
- tools: contraction mapping and theorem of the maximum

# Sequence Problem vs. Functional Equation

- Sequence Problem: (SP)

$$V^*(x_0) = \sup_{\{x_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t F(x_t, x_{t+1})$$

s.t.  $x_{t+1} \in \Gamma(x_t)$   
 $x_0$  given

- ... more succinctly

$$V^*(x_0) = \sup_{\tilde{x} \in \Pi(x_0)} u(\tilde{x}) \quad (\text{SP})$$

- functional equation (FE) [this particular FE called Bellman Equation]

$$V(x) = \max_{y \in \Gamma(x)} \{F(x, y) + \beta V(y)\} \quad (\text{FE})$$

# Principle of Optimality

IDEA: use BE to find value function  $V^*$  and optimal plan  $x^*$

- **Thm 4.2.**  $V^*$  defined by SP  $\Rightarrow V^*$  solves FE
- **Thm 4.3.**  $V$  solves FE and .....  $\Rightarrow V = V^*$
- **Thm 4.4.**  $\tilde{x}^* \in \Pi(x_0)$  is optimal  
 $\Rightarrow V^*(x_t^*) = F(x_t^*, x_{t+1}^*) + \beta V^*(x_{t+1}^*)$
- **Thm 4.5.**  $\tilde{x}^* \in \Pi(x_0)$  satisfies  $V^*(x_t^*) = F(x_t^*, x_{t+1}^*) + \beta V^*(x_{t+1}^*)$   
and .....  
 $\Rightarrow \tilde{x}^*$  is optimal

# Why is this Progress?

- **intuition:** breaks planning horizon into two: 'now' and 'then'
- **notation:** reduces unnecessary notation (especially with uncertainty)
- **analysis:** prove existence, uniqueness and properties of optimal policy (e.g. continuity, monotonicity, etc...)
- **computation:** associated numerical algorithm are powerful for many applications

## Proof of Theorem 4.3 (max case)

Assume for any  $\tilde{x} \in \Pi(x_0)$

$$\lim_{T \rightarrow \infty} \beta^T V(x_T) = 0.$$

BE implies

$$\begin{aligned} V(x_0) &\geq F(x_0, x_1) + \beta V(x_1), \text{ all } x_1 \in \Gamma(x_0) \\ &= F(x_0, x_1^*) + \beta V(x_1^*), \text{ some } x_1^* \in \Gamma(x_0) \end{aligned}$$

Substituting  $V(x_1)$ :

$$\begin{aligned} V(x_0) &\geq F(x_0, x_1) + \beta F(x_1, x_2) + \beta^2 V(x_2), \text{ all } x \in \Pi(x_0) \\ &= F(x_0, x_1^*) + \beta F(x_1^*, x_2^*) + \beta^2 V(x_2^*), \text{ some } x^* \in \Pi(x_0) \end{aligned}$$

Continue this way

$$\begin{aligned} V(x_0) &\geq \sum_{t=0}^n \beta^t F(x_t, x_{t+1}) + \beta^{n+1} V(x_{n+1}) \text{ for all } x \in \Pi(x_0) \\ &= \sum_{t=0}^n \beta^t F(x_t^*, x_{t+1}^*) + \beta^{n+1} V(x_{n+1}^*) \text{ for some } x^* \in \Pi(x_0) \end{aligned}$$

Since  $\beta^T V(x_T) \rightarrow 0$ , taking the limit  $n \rightarrow \infty$  on both sides of both expressions we conclude that:

$$V(x_0) \geq u(\tilde{x}) \text{ for all } \tilde{x} \in \Pi(x_0)$$

$$V(x_0) = u(\tilde{x}^*) \text{ for some } \tilde{x}^* \in \Pi(x_0)$$

# Bellman Equation as a Fixed Point

- define operator

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

- $V$  solution of BE  $\iff V$  fixed point of  $T$  [i.e.  $TV = V$ ]

## Bounded Returns:

- if  $\|F\| < B$  and  $F$  and  $\Gamma$  are continuous:  $T$  maps continuous bounded functions into continuous bounded functions
- bounded returns  $\implies T$  is a Contraction Mapping  $\implies$  unique fixed point
- many other bonuses

# Needed Tools

- Basic Real Analysis (section 3.1):
  - {vector, metric, noSLP, complete} spaces
  - cauchy sequences
  - closed, compact, bounded sets
- Contraction Mapping Theorem (section 3.2)
- Theorem of the Maximum: study of RHS of Bellman equation (equivalently of  $T$ ) (section 3.3)

# Bellman Equation: Principle of Optimality

- Principle of Optimality idea: use the functional equation

$$V(x) = \max_{y \in \Gamma(x)} \{F(x, y) + \beta V(y)\}$$

to find  $V^*$  and  $g$

- note: nuisance subscripts  $t, t + 1$ , dropped
- a solution is a function  $V(\cdot)$  the same on both sides
- **IF** BE has unique solution then  $V^* = V$
- more generally the “right solution” to (BE) delivers  $V^*$

# Outline Today's Lecture

- study Functional Equation (Bellman equation) with bounded and continuous  $F$
- tools: contraction mapping and theorem of the maximum

# Bellman Equation as a Fixed Point

- define operator

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

- $V$  solution of BE  $\iff V$  fixed point of  $T$  [i.e.  $TV = V$ ]

## Bounded Returns:

- if  $\|F\| < B$  and  $F$  and  $\Gamma$  are continuous:  $T$  maps continuous bounded functions into continuous bounded functions
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# Our Favorite Metric Space

$$S = \left\{ f : X \rightarrow R, f \text{ is continuous, and } \|f\| \equiv \sup_{x \in X} |f(x)| < \infty \right\}$$

with

$$\rho(f, g) = \|f - g\| \equiv \sup_{x \in X} |f(x) - g(x)|$$

**Definition.** A linear space  $S$  is complete if any Cauchy sequence converges. For a definition of a Cauchy sequence and examples of complete metric spaces see SLP.

**Theorem.** The set of bounded and continuous functions is Complete. See SLP.

# Contraction Mapping

**Definition.** Let  $(S, \rho)$  be a metric space. Let  $T : S \rightarrow S$  be an operator.  $T$  is a contraction with modulus  $\beta \in (0, 1)$

$$\rho(Tx, Ty) \leq \beta \rho(x, y)$$

for any  $x, y$  in  $S$ .

# Contraction Mapping Theorem

**Theorem (CMThm).** If  $T$  is a contraction in  $(S, \rho)$  with modulus  $\beta$ , then (i) there is a unique fixed point  $s^* \in S$ ,

$$s^* = Ts^*$$

and (ii) iterations of  $T$  converge to the fixed point

$$\rho(T^n s_0, s^*) \leq \beta^n \rho(s_0, s^*)$$

for any  $s_0 \in S$ , where  $T^{n+1}(s) = T(T^n(s))$ .

## CMThm – Proof

for (i) **1st step:** construct fixed point  $s^*$   
take any  $s_0 \in S$  define  $\{s_n\}$  by  $s_{n+1} = Ts_n$  then

$$\rho(s_2, s_1) = \rho(Ts_1, Ts_0) \leq \beta \rho(s_1, s_0)$$

generalizing  $\rho(s_{n+1}, s_n) \leq \beta^n \rho(s_1, s_0)$  then, for  $m > n$

$$\begin{aligned} \rho(s_m, s_n) &\leq \rho(s_m, s_{m-1}) + \rho(s_{m-1}, s_{m-2}) + \dots + \rho(s_{n+1}, s_n) \\ &\leq [\beta^{m-1} + \beta^{m-2} + \dots + \beta^n] \rho(s_1, s_0) \\ &\leq \beta^n [\beta^{m-n-1} + \beta^{m-n-2} + \dots + 1] \rho(s_1, s_0) \\ &\leq \frac{\beta^n}{1 - \beta} \rho(s_1, s_0) \end{aligned}$$

thus  $\{s_n\}$  is cauchy. hence  $s_n \rightarrow s^*$

**2nd step:** show  $s^* = Ts^*$

$$\begin{aligned}\rho(Ts^*, s^*) &\leq \rho(Ts^*, s_n) + \rho(s^*, s_n) \\ &\leq \beta\rho(s^*, s_{n-1}) + \rho(s^*, s_n) \rightarrow 0\end{aligned}$$

**3rd step:**  $s^*$  is unique.  $Ts_1^* = s_1^*$  and  $s_2^* = Ts_2^*$

$$0 \leq a = \rho(s_1^*, s_2^*) = \rho(Ts_1^*, Ts_2^*) \leq \beta\rho(s_1^*, s_2^*) = \beta a$$

only possible if  $a = 0 \Rightarrow s_1^* = s_2^*$ .

Finally, as for (ii):

$$\rho(T^n s_0, s^*) = \rho(T^n s_0, Ts^*) \leq \beta\rho(T^{n-1} s_0, s^*) \leq \dots \leq \beta^n \rho(s_0, s^*)$$

**Corollary.** Let  $S$  be a complete metric space, let  $S' \subset S$  and  $S'$  close. Let  $T$  be a contraction on  $S$  and let  $s^* = Ts^*$ . Assume that

$$T(S') \subset S', \quad \text{i.e. if } s' \in S', \text{ then } T(s') \in S'$$

then  $s^* \in S'$ . Moreover, if  $S'' \subset S'$  and

$$T(S'') \subset S'', \quad \text{i.e. if } s' \in S'', \text{ then } T(s') \in S''$$

then  $s^* \in S''$ .

### **Blackwell's sufficient conditions.**

Let  $S$  be the space of bounded functions on  $X$ , and  $\|\cdot\|$  be given by the sup norm. Let  $T : S \rightarrow S$ . Assume that (i)  $T$  is monotone, that is,

$$Tf(x) \leq Tg(x)$$

for any  $x \in X$  and  $g, f$  such that  $f(x) \geq g(x)$  for all  $x \in X$ , and (ii)  $T$  discounts, that is, there is a  $\beta \in (0, 1)$  such that for any  $a \in R_+$ ,

$$T(f + a)(x) \leq Tf(x) + a\beta$$

for all  $x \in X$ . Then  $T$  is a contraction.

**Proof.** By definition

$$f = g + f - g$$

and using the definition of  $\|\cdot\|$ ,

$$f(x) \leq g(x) + \|f - g\|$$

then by monotonicity i)

$$Tf \leq T(g + \|f - g\|)$$

and by discounting ii) setting  $a = \|f - g\|$

$$Tf \leq T(g) + \beta \|f - g\|.$$

Reversing the roles of  $f$  and  $g$ :

$$Tg \leq T(f) + \beta \|f - g\|$$

$$\Rightarrow \|Tf - Tg\| \leq \beta \|f - g\|$$

# Bellman equation application

$$(Tv)(x) = \max_{y \in \Gamma(x)} \{F(x, y) + \beta v(y)\}$$

Assume that  $F$  is bounded and continuous and that  $\Gamma$  is continuous and has compact range.

**Theorem.**  $T$  maps the set of continuous and bounded functions  $S$  into itself. Moreover  $T$  is a contraction.

**Proof.** That  $T$  maps the set of continuous and bounded functions to itself follows from the Theorem of Maximum (we do this next)

That  $T$  is a contraction follows since  $T$  satisfies the Blackwell sufficient conditions.

$T$  satisfies the Blackwell sufficient conditions. For monotonicity, notice that for  $f \geq v$

$$\begin{aligned}Tv(x) &= \max_{y \in \Gamma(x)} \{F(x, y) + \beta v(y)\} \\ &= F(x, g(x)) + \beta v(g(x)) \\ &\leq \{F(x, g(x)) + \beta f(g(x))\} \\ &\leq \max_{y \in \Gamma(x)} \{F(x, y) + \beta f(y)\} = Tf(x)\end{aligned}$$

A similar argument follows for discounting: for  $a > 0$

$$\begin{aligned}T(v + a)(x) &= \max_{y \in \Gamma(x)} \{F(x, y) + \beta (v(y) + a)\} \\ &= \max_{y \in \Gamma(x)} \{F(x, y) + \beta v(y)\} + \beta a = T(v)(x) + \beta a.\end{aligned}$$

# Theorem of the Maximum

- want  $T$  to map continuous function into continuous functions

$$(Tv)(x) = \max_{y \in \Gamma(x)} \{F(x, y) + \beta v(y)\}$$

- want to learn about optimal policy of RHS of Bellman

$$G(x) = \arg \max_{y \in \Gamma(x)} \{F(x, y) + \beta v(y)\}$$

- First, continuity concepts for correspondences
- ... then, a few example maximizations
- ... finally, Theorem of the Maximum

# Continuity Notions for Correspondences

assume  $\Gamma$  is non-empty and compact valued (the set  $\Gamma(x)$  is non empty and compact for all  $x \in X$ )

**Upper Hemi Continuity (u.h.c.) at  $x$ :** for any pair of sequences  $\{x_n\}$  and  $\{y_n\}$  with  $x_n \rightarrow x$  and  $x_n \in \Gamma(y_n)$  there exists a subsequence of  $\{y_n\}$  that converges to a point  $y \in \Gamma(x)$ .

**Lower Hemi Continuity (l.h.c.) at  $x$ :** for any sequence  $\{x_n\}$  with  $x_n \rightarrow x$  and for every  $y \in \Gamma(x)$  there exists a sequence  $\{y_n\}$  with  $x_n \in \Gamma(y_n)$  such that  $y_n \rightarrow y$ .

**Continuous at  $x$ :** if  $\Gamma$  is both upper and lower hemi continuous at  $x$

# Max Examples

$$h(x) = \max_{y \in \Gamma(x)} f(x, y)$$

$$G(x) = \arg \max_{y \in \Gamma(x)} f(x, y)$$

**ex 1:**  $f(x, y) = xy$ ;  $X = [-1, 1]$ ;  $\Gamma(x) = X$ .

$$G(x) = \begin{cases} \{-1\} & x < 0 \\ [-1, 1] & x = 0 \\ \{1\} & x > 0 \end{cases}$$

$$h(x) = |x|$$

continued...

**ex 2:**  $f(x, y) = xy^2$ ;  $X = [-1, 1]$ ;  $\Gamma(x) = X$

$$G(x) = \begin{cases} \{0\} & x < 0 \\ [-1, 1] & x = 0 \\ \{-1, 1\} & x > 0 \end{cases}$$
$$h(x) = \max\{0, x\}$$

# Theorem of the Maximum

Define:

$$h(x) = \max_{y \in \Gamma(x)} f(x, y)$$

$$\begin{aligned} G(x) &= \arg \max_{y \in \Gamma(x)} f(x, y) \\ &= \{y \in \Gamma(x) : h(x) = f(x, y)\} \end{aligned}$$

**Theorem.** (Berge) Let  $X \subset \mathbb{R}^l$  and  $Y \subset \mathbb{R}^m$ . Let  $f : X \times Y \rightarrow \mathbb{R}$  be continuous and  $\Gamma : X \rightarrow Y$  be compact-valued and continuous. Then  $h : X \rightarrow \mathbb{R}$  is continuous and  $G : X \rightarrow Y$  is non-empty, compact valued, and u.h.c.

# $\lim \max \rightarrow \max \lim$

**Theorem.** Suppose  $\{f_n(x, y)\}$  and  $f(x, y)$  are concave in  $y$  and  $f_n \rightarrow f$  in the sup-norm (uniformly). Define

$$g_n(x) = \arg \max_{y \in \Gamma(x)} f_n(x, y)$$

$$g(x) = \arg \max_{y \in \Gamma(x)} f(x, y)$$

then  $g_n(x) \rightarrow g(x)$  for all  $x$  (pointwise convergence); if  $X$  is compact then the convergence is uniform.

# Uses of Corollary of CMThm

Monotonicity of  $v^*$

**Theorem.** Assume that  $F(\cdot, y)$  is increasing, that  $\Gamma$  is increasing, i.e.

$$\Gamma(x) \subset \Gamma(x')$$

for  $x \leq x'$ . Then, the unique fixed point  $v^*$  satisfying  $v^* = Tv^*$  is increasing. If  $F(\cdot, y)$  is strictly increasing, so is  $v^*$ .

# Proof

By the corollary of the CMThm, it suffices to show  $Tf$  is increasing if  $f$  is increasing. Let  $x \leq x'$  :

$$\begin{aligned} Tf(x) &= \max_{y \in \Gamma(x)} \{F(x, y) + \beta f(y)\} \\ &= F(x, y^*) + \beta f(y^*) \text{ for some } y^* \in \Gamma(x) \\ &\leq F(x', y^*) + \beta f(y^*) \end{aligned}$$

since  $y^* \in \Gamma(x) \subset \Gamma(x')$

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

If  $F(\cdot, y)$  is strictly increasing

$$F(x, y^*) + \beta f(y^*) < F(x', y^*) + \beta f(y^*).$$

## Concavity (or strict) concavity of $v^*$

**Theorem.** Assume that  $X$  is convex,  $\Gamma$  is concave, i.e.  $y \in \Gamma(x)$ ,  $y' \in \Gamma(x')$  implies that

$$y^\theta \equiv \theta y' + (1 - \theta) y \in \Gamma(\theta x' + (1 - \theta) x) \equiv \Gamma(x^\theta)$$

for any  $x, x' \in X$  and  $\theta \in (0, 1)$ . Finally assume that  $F$  is concave in  $(x, y)$ . Then, the fixed point  $v^*$  satisfying  $v^* = T v^*$  is concave in  $x$ . Moreover, if  $F(\cdot, y)$  is strictly concave, so is  $v^*$ .

# Differentiability

- can't use same strategy: space of differentiable functions is not closed
- many envelope theorems
- Formula: if  $h(x)$  is differentiable and  $y$  is interior then

$$h'(x) = f_x(x, y)$$

right value... but is  $h$  differentiable?

- one answer (Demand Theory) relies on f.o.c. and assuming twice differentiability of  $f$
- won't work for us since  $f = F(x, y) + \beta V(y)$  and we don't even know if  $f$  is once differentiable! → going in circles

# Benveniste and Sheinkman

First a Lemma...

**Lemma.** Suppose  $v(x)$  is concave and that there exists  $w(x)$  such that  $w(x) \leq v(x)$  and  $v(x_0) = w(x_0)$  in some neighborhood  $D$  of  $x_0$  and  $w$  is differentiable at  $x_0$  ( $w'(x_0)$  exists) then  $v$  is differentiable at  $x_0$  and  $v'(x_0) = w'(x_0)$ .

**Proof.** Since  $v$  is concave it has at least one subgradient  $p$  at  $x_0$  :

$$w(x) - w(x_0) \leq v(x) - v(x_0) \leq p \cdot (x - x_0)$$

Thus a subgradient of  $v$  is also a subgradient of  $w$ . But  $w$  has a unique subgradient equal to  $w'(x_0)$ .  $\square$

# Benveniste and Sheinkman

Now a Theorem

**Theorem.** Suppose  $F$  is strictly concave and  $\Gamma$  is convex. If  $x_0 \in \text{int}(X)$  and  $g(x_0) \in \text{int}(\Gamma(x_0))$  then the fixed point of  $T, V$ , is differentiable at  $x$  and

$$V'(x) = F_x(x, g(x))$$

**Proof.** We know  $V$  is concave. Since  $x_0 \in \text{int}(X)$  and  $g(x_0) \in \text{int}(\Gamma(x_0))$  then  $g(x_0) \in \text{int}(\Gamma(x))$  for  $x \in D$  a neighborhood of  $x_0$  then

$$W(x) = F(x, g(x_0)) + \beta V(g(x_0))$$

and then  $W(x) \leq V(x)$  and  $W(x_0) = V(x_0)$  and  $W'(x_0) = F_x(x_0, g(x_0))$  so the result follows from the lemma.  $\square$