

Theorem C.14 (again!)

Consider (X, d_X) and (Y, d_Y) . If (Y, d_Y) is complete, then $(B(X, Y), d_\infty)$ and $(CB(X, Y), d_\infty)$ are complete.

Proof $B(X, Y)$ is complete:

Let $f_n \in B(X, Y)$ be a Cauchy sequence.

Pick any $x \in X$. Then $y_n = f_n(x)$ is a Cauchy sequence in (Y, d_Y) . Since (Y, d_Y) is complete, $y_n \rightarrow y^*$ for some $y^* \in Y$.

Define $f^*(x) = y^*$, i.e. $f^*(x) = \lim_{n \rightarrow \infty} f_n(x)$.

By continuity of d_Y , we know

$$d_X(f^*(x), f_n(x)) = \lim_{m \rightarrow \infty} d_Y(f_m(x), f_n(x)), \text{ for all } x.$$

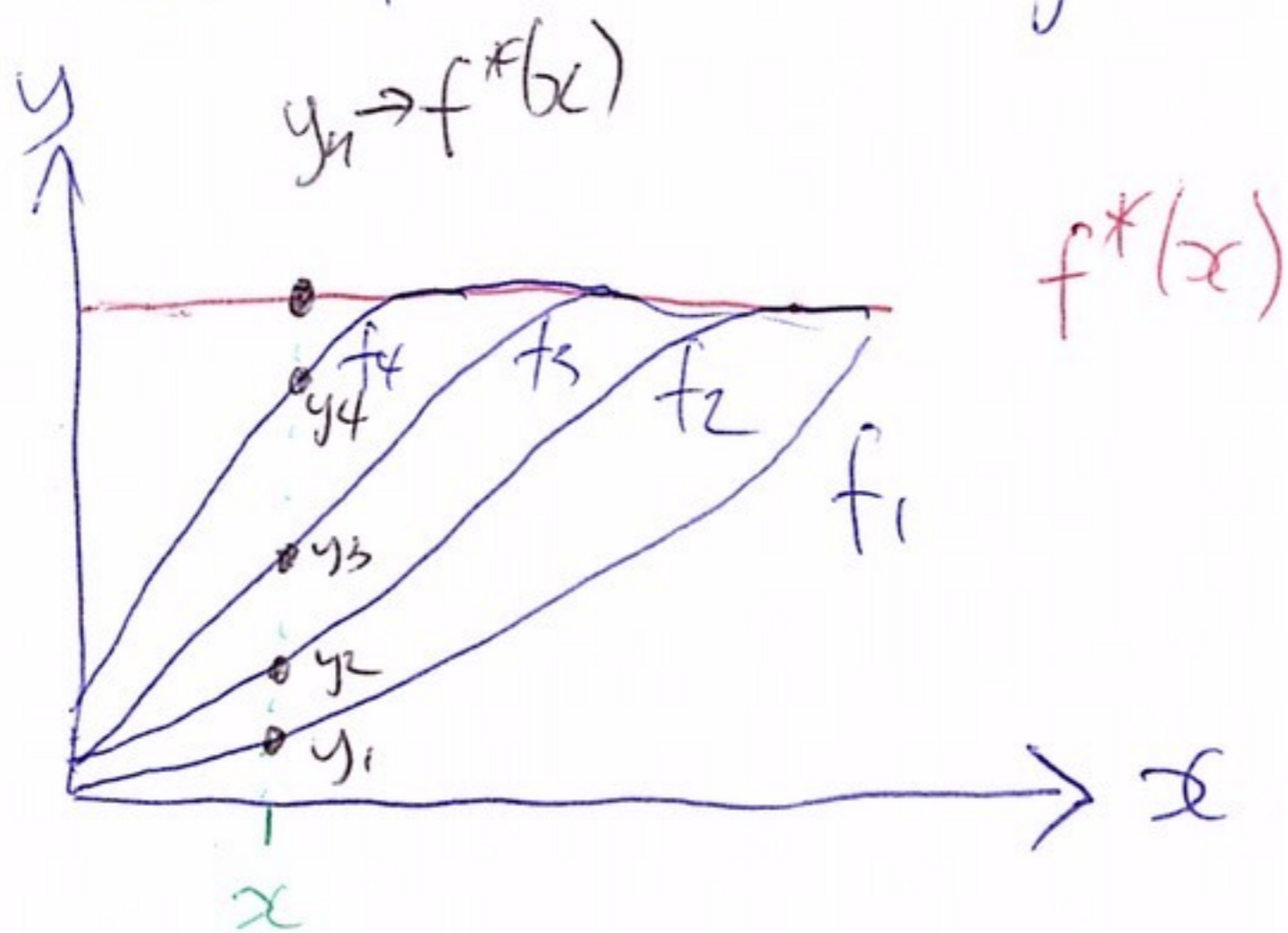
$$\leq \lim_{m \rightarrow \infty} \sup_{x' \in X} d_Y(f_m(x'), f_n(x')), "$$
$$= \lim_{m \rightarrow \infty} d_\infty(f_m, f_n), \text{ for all } x.$$

$$\text{So } \sup_{x \in X} d_Y(f^*(x), f_n(x)) \leq \lim_{m \rightarrow \infty} d_\infty(f_m, f_n).$$

as $n \rightarrow \infty$, " $\rightarrow 0$."

Conclusion: $d_\infty(f^*, f_n) \rightarrow 0$ which implies $f_n \rightarrow f^*$. □

What could have gone wrong?



Back to dynamic programming in section 2.4

$$\pi(p; w) = \max_{x \in \mathbb{R}_+^{N-1}} p f(x) - w \cdot x$$

$$\pi(p; w) = \max_{y \in \mathbb{R}_+} p y - c(y; w) \leftarrow \text{Bellman equation}$$

$$\text{where } c(y; w) = \min_{x \in \mathbb{R}_+^{N-1}} w \cdot x \\ \text{s.t. } f(x) \geq y.$$

Lemma (Principle of Optimality)

The definitions of π agree.

Proof $\max_{x \in \mathbb{R}_+^{N-1}} p f(x) - w \cdot x$

$= \max_{y \in \mathbb{R}_+} p y - w \cdot x$

point-less!

$$= \max_{\substack{y \in \mathbb{R}_+, \\ x \in \mathbb{R}_+^{N-1}}} p f(x) - w \cdot x$$

s.t. $f(x) = y$ ← slack ("pointless") constraint

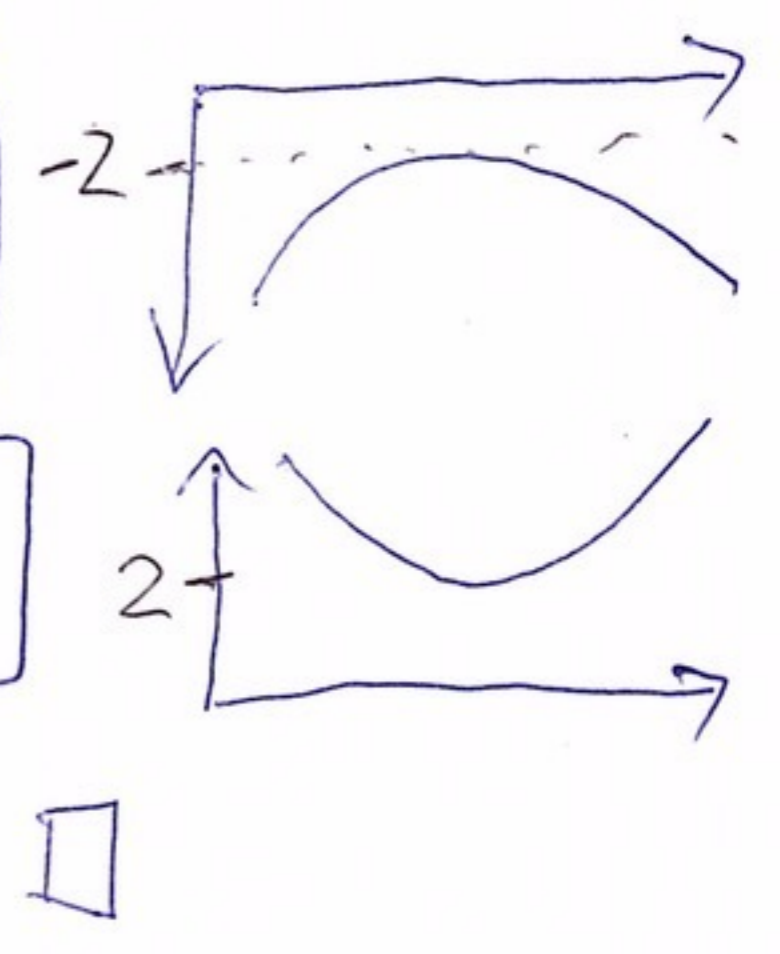
$$= \max_{y \in \mathbb{R}_+} \left[\max_{\substack{x \in \mathbb{R}_+^{N-1} \\ \text{s.t. } f(x) = y}} p f(x) - w \cdot x \right]$$

$$= \max_{y \in \mathbb{R}_+} \left[\max_{\substack{x \in \mathbb{R}_+^{N-1} \\ \text{s.t. } f(x) = y}} p y - w \cdot x \right]$$

$$= \max_{y \in \mathbb{R}_+} p y + \left[\max_{\substack{x \in \mathbb{R}_+^{N-1} \\ \text{s.t. } f(x) = y}} -w \cdot x \right]$$

$$= \max_{y \in \mathbb{R}_+} p y - \left[\min_{\substack{x \in \mathbb{R}_+^{N-1} \\ \text{s.t. } f(x) = y}} w \cdot x \right]$$

$$= \max_{y \in \mathbb{R}_+} p y - c(y; w).$$



□

Theorem If $y(p; w)$ is an optimal supply policy, then

$$p = \left[\frac{\partial c(y; w)}{\partial y} \right]_{y = y(p; w)}$$

Appendix G: Infinite Horizon

Dynamic Programming

Cake of size k .

$$V_t(k) = \max_{(x_s)_{s=1}^{\infty}} \sum_{s=t}^{\infty} \beta^{s-t} u(x_s)$$

$$\text{s.t. } \sum_{s=t}^{\infty} x_s = k.$$

Bellman equation:

$$V_t(k) = \max_{x_t, k_{t+1} \geq 0} u(x_t) + \beta V_{t+1}(k_{t+1})$$

$$\text{s.t. } x_t + k_{t+1} = k.$$

Since V_t does not depend on t :

$$V(k) = \max_{x, k' \geq 0} u(x) + \beta V(k')$$

\square means "tomorrow"

recursive
Bellman equation

$$\text{s.t. } x + k' = k.$$

Theorem (Principle of Optimality)

Suppose there is a solution to the cake eating problem. Then the value function V_0 is a solution to the Bellman equation.

Proof: $V_0(k) = \max_{(x_s)_{s=0}^{\infty}} \sum_{s=0}^{\infty} \beta^s u(x_s) \text{ s.t. } \sum_{s=0}^{\infty} x_s = k$

$$= \max_{x_0, k_1} \sum_{s=1}^{\infty} \beta^s u(x_s) \text{ s.t. } x_0 + k_1 = k \text{ and } \sum_{s=1}^{\infty} x_s = k_1$$

$$= \max_{x_0, k_1} \left[\begin{array}{l} \max_{(x_s)_{s=1}^{\infty}} \sum_{s=0}^{\infty} \beta^s u(x_s) \\ \text{s.t. } \sum_{s=1}^{\infty} x_s = k_1 \end{array} \right]$$

$$= \max_{x_0, k_1} \left[\begin{array}{l} \max_{(x_s)_{s=1}^{\infty}} u(x_0) + \sum_{s=1}^{\infty} \beta^s u(x_s) \\ \text{s.t. } \sum_{s=1}^{\infty} x_s = k_1 \end{array} \right]$$

$$= \max_{x_0, k_1} \left[\begin{array}{l} u(x_0) + \beta \left[\max_{(x_s)_{s=1}^{\infty}} \sum_{s=1}^{\infty} \beta^{s-1} u(x_s) \right] \\ \text{s.t. } \sum_{s=1}^{\infty} x_s = k_1 \end{array} \right]$$

$$= \max_{x_0, k_1} u(x_0) + \beta V_0(k_1)$$

s.t. $x_0 + k_1 = k$

$$= \max_{x, k'} u(x) + \beta V_0(k')$$

s.t. $x + k' = k$

□

Theorem (Banach fixed point theorem)

Let (X, d) be a complete metric space.

If $f: X \rightarrow X$ is a contraction of

degree a , then

(i) f has a unique fixed point x^* ,

(ii) Given any $x_0 \in X$, then the sequence defined by ~~$x_{n+1} = f(x_n)$~~ $x_{n+1} = f(x_n)$ converges to x^* ,

(iii) $d(x_n, x^*) \leq \frac{a^n}{1-a} d(x_0, x_1)$ for all n .

Proof (i) uniqueness: Suppose both x^* and x^{**}

are fixed points of f , where $x^* \neq x^{**}$.

Since they are fixed points, $x^* = f(x^*)$ and

$x^{**} = f(x^{**})$. Applying the contraction

property,

$$d(f(x^*), f(x^{**})) \leq a d(x^*, x^{**})$$

$$\Rightarrow d(x^*, x^{**}) \leq a d(x^*, x^{**})$$

$$\Rightarrow d(x^*, x^{**}) < d(x^*, x^{**}) \quad \text{⚡}$$

(i) existence: We first show x_n is a Cauchy

sequence. If we repeatedly apply the contraction property, we get:

$$d(f(x_0), f(x_m)) \leq a d(x_0, x_m)$$

$$d(f(f(x_0)), f(f(x_m))) \leq a^2 d(x_0, x_m)$$

$$\textcircled{*} d(f^n(x_0), f^n(x_m)) \leq a^n d(x_0, x_m)$$

This implies

$$\begin{aligned}
 d(x_0, x_m) &\leq d(x_0, x_1) + d(x_1, x_2) + \dots + d(x_{m-1}, x_m) \\
 &\leq d(x_0, x_1) + a d(x_0, x_1) + a^2 d(x_0, x_1) + \dots \\
 &= d(x_0, x_1) [1 + a + a^2 + \dots] \\
 &= \frac{1}{1-a} d(x_0, x_1).
 \end{aligned}$$

triangle inequality $n=1, m=1$ $n=m-1, m=1$

Rewriting the formula (*) above:

$$\begin{aligned}
 d(x_n, x_{n+m}) &\leq a^n d(x_0, x_m) \\
 &\leq a^n \left\{ \frac{1}{1-a} d(x_0, x_1) \right\} \text{ for all } n, n.
 \end{aligned}$$

$$\Rightarrow d(x_n, x_m) \leq \frac{a^N}{1-a} d(x_0, x_1) \text{ for all } n, m \geq N. \quad (***)$$

So x_n is a Cauchy sequence.

Since (X, d) is a complete metric space, $x_n \rightarrow x^*$ for some $x^* \in X$. Let $y_n = f(x_n)$.

Since f is continuous, and $x_n \rightarrow x^*$ we conclude $y_n \rightarrow f(x^*)$. But y_n is a subsequence of

x_n (since $y_n = x_{n+1}$), we conclude $y_n \rightarrow x^*$.

This means $x^* = f(x^*)$. So x^* is a fixed point of f .

(iii) Approximation bound: By continuity of

$$d(x_n, x^*) = \lim_{m \rightarrow \infty} d(x_n, x_m) \leq \frac{a^n}{1-a} d(x_0, x_1).$$

since d is continuous

from (***) \square

Back to the cake-eating problem

"tomorrow's value function"

Bellman operator:

$$F(V')(k) = \max_{x, k' \geq 0} u(x) + \beta V'(k')$$

s.t. $x + k' = k$.

$$F: B(\mathbb{R}_+) \rightarrow B(\mathbb{R}_+)$$

Our goal is to use Banach's fixed point theorem to find a fixed point of F . Algorithm:

- ① Start with a guess V_0 .
- ② $V_{n+1} = F(V_n)$
- ③ Stop at V_N , where N is chosen to ensure $d_\infty(V_N, V)$ is small enough.