

6.231 DYNAMIC PROGRAMMING

LECTURE 15

LECTURE OUTLINE

- Review of basic theory of discounted problems
- Monotonicity and contraction properties
- Contraction mappings in DP
- Discounted problems: Countable state space with unbounded costs
- Generalized discounted DP
- An introduction to abstract DP

DISCOUNTED PROBLEMS/BOUNDED COST

- Stationary system with arbitrary state space

$$x_{k+1} = f(x_k, u_k, w_k), \quad k = 0, 1, \dots$$

- Cost of a policy $\pi = \{\mu_0, \mu_1, \dots\}$

$$J_\pi(x_0) = \lim_{N \rightarrow \infty} E_{w_k} \left\{ \sum_{k=0}^{N-1} \alpha^k g(x_k, \mu_k(x_k), w_k) \right\}$$

with $\alpha < 1$, and for some M , we have $|g(x, u, w)| \leq M$ for all (x, u, w)

- **Shorthand notation for DP mappings** (operate on functions of state to produce other functions)

$$(TJ)(x) = \min_{u \in U(x)} E_w \left\{ g(x, u, w) + \alpha J(f(x, u, w)) \right\}, \quad \forall x$$

TJ is the optimal cost function for the one-stage problem with stage cost g and terminal cost αJ .

- For any stationary policy μ

$$(T_\mu J)(x) = E_w \left\{ g(x, \mu(x), w) + \alpha J(f(x, \mu(x), w)) \right\}, \quad \forall x$$

“SHORTHAND” THEORY – A SUMMARY

- **Cost function expressions** [with $J_0(x) \equiv 0$]

$$J_\pi(x) = \lim_{k \rightarrow \infty} (T_{\mu_0} T_{\mu_1} \cdots T_{\mu_k} J_0)(x), \quad J_\mu(x) = \lim_{k \rightarrow \infty} (T_\mu^k J_0)(x)$$

- **Bellman’s equation:** $J^* = T J^*$, $J_\mu = T_\mu J_\mu$
- **Optimality condition:**

$$\mu: \text{optimal} \quad \langle == \rangle \quad T_\mu J^* = T J^*$$

- **Value iteration:** For any (bounded) J and all x :

$$J^*(x) = \lim_{k \rightarrow \infty} (T^k J)(x)$$

- **Policy iteration:** Given μ^k ,
 - Policy evaluation: Find J_{μ^k} by solving

$$J_{\mu^k} = T_{\mu^k} J_{\mu^k}$$

- Policy improvement: Find μ^{k+1} such that

$$T_{\mu^{k+1}} J_{\mu^k} = T J_{\mu^k}$$

MAJOR PROPERTIES

- **Monotonicity property:** For any functions J and J' on the state space X such that $J(x) \leq J'(x)$ for all $x \in X$, and any μ

$$(TJ)(x) \leq (TJ')(x), \quad \forall x \in X,$$

$$(T_\mu J)(x) \leq (T_\mu J')(x), \quad \forall x \in X.$$

- **Contraction property:** For any bounded functions J and J' , and any μ ,

$$\max_x |(TJ)(x) - (TJ')(x)| \leq \alpha \max_x |J(x) - J'(x)|,$$

$$\max_x |(T_\mu J)(x) - (T_\mu J')(x)| \leq \alpha \max_x |J(x) - J'(x)|.$$

- Shorthand writing of the contraction property

$$\|TJ - TJ'\| \leq \alpha \|J - J'\|, \quad \|T_\mu J - T_\mu J'\| \leq \alpha \|J - J'\|,$$

where for any bounded function J , we denote by $\|J\|$ the sup-norm

$$\|J\| = \max_{x \in X} |J(x)|.$$

CONTRACTION MAPPINGS

- Given a real vector space Y with a norm $\| \cdot \|$ (see text for definitions).
- A function $F : Y \mapsto Y$ is said to be a *contraction mapping* if for some $\rho \in (0, 1)$, we have

$$\|Fy - Fz\| \leq \rho \|y - z\|, \quad \text{for all } y, z \in Y.$$

ρ is called the *modulus of contraction* of F .

- **Linear case, $Y = \mathfrak{R}^n$:** $Fy = Ay + b$ is a contraction (for some norm $\| \cdot \|$) if and only if all eigenvalues of A are strictly within the unit circle.
- For $m > 1$, we say that F is an *m-stage contraction* if F^m is a contraction.
- **Important example:** Let X be a set (e.g., state space in DP), $v : X \mapsto \mathfrak{R}$ be a positive-valued function. Let $B(X)$ be the set of all functions $J : X \mapsto \mathfrak{R}$ such that $J(s)/v(s)$ is bounded over s .
- The *weighted sup-norm* on $B(X)$:

$$\|J\| = \max_{s \in X} \frac{|J(s)|}{v(s)}.$$

- **Important special case:** The discounted problem mappings T and T_μ [for $v(s) \equiv 1$, $\rho = \alpha$].

A DP-LIKE CONTRACTION MAPPING

- Let $X = \{1, 2, \dots\}$, and let $F : B(X) \mapsto B(X)$ be a linear mapping of the form

$$(FJ)(i) = b(i) + \sum_{j \in X} a(i, j) J(j), \quad \forall i$$

where $b(i)$ and $a(i, j)$ are some scalars. Then F is a contraction with modulus ρ if

$$\frac{\sum_{j \in X} |a(i, j)| v(j)}{v(i)} \leq \rho, \quad \forall i$$

[Think of the special case where $a(i, j)$ are the transition probs. of a policy].

- Let $F : B(X) \mapsto B(X)$ be the mapping

$$(FJ)(i) = \min_{\mu \in M} (F_{\mu}J)(i), \quad \forall i$$

where M is parameter set, and for each $\mu \in M$, F_{μ} is a contraction from $B(X)$ to $B(X)$ with modulus ρ . Then F is a contraction with modulus ρ .

CONTRACTION MAPPING FIXED-POINT TH.

- **Contraction Mapping Fixed-Point Theorem:** If $F : B(X) \mapsto B(X)$ is a contraction with modulus $\rho \in (0, 1)$, then there exists a unique $J^* \in B(X)$ such that

$$J^* = FJ^*.$$

Furthermore, if J is any function in $B(X)$, then $\{F^k J\}$ converges to J^* and we have

$$\|F^k J - J^*\| \leq \rho^k \|J - J^*\|, \quad k = 1, 2, \dots$$

- Similar result if F is an m -stage contraction mapping.
- This is a special case of a general result for contraction mappings $F : Y \mapsto Y$ over normed vector spaces Y that are **complete**: every sequence $\{y_k\}$ that is Cauchy (satisfies $\|y_m - y_n\| \rightarrow 0$ as $m, n \rightarrow \infty$) converges.
- The space $B(X)$ is complete [see the text (Section 1.5) for a proof].

GENERAL FORMS OF DISCOUNTED DP

- **Monotonicity assumption:** If $J, J' \in R(X)$ and $J \leq J'$, then

$$H(x, u, J) \leq H(x, u, J'), \quad \forall x \in X, u \in U(x)$$

- **Contraction assumption:**
 - For every $J \in B(X)$, the functions $T_\mu J$ and TJ belong to $B(X)$.
 - For some $\alpha \in (0, 1)$ and all $J, J' \in B(X)$, H satisfies

$$|H(x, u, J) - H(x, u, J')| \leq \alpha \max_{y \in X} |J(y) - J'(y)|$$

for all $x \in X$ and $u \in U(x)$.

- We can show all the standard analytical and computational results of discounted DP based on these two assumptions (**with identical proofs!**)
- With just the monotonicity assumption (as in shortest path problem) we can still show various forms of the basic results under appropriate assumptions (like in the SSP problem)

EXAMPLES

- Discounted problems

$$H(x, u, J) = E\{g(x, u, w) + \alpha J(f(x, u, w))\}$$

- Discounted Semi-Markov Problems

$$H(x, u, J) = G(x, u) + \sum_{y=1}^n m_{xy}(u)J(y)$$

where m_{xy} are “discounted” transition probabilities, defined by the transition distributions

- Deterministic Shortest Path Problems

$$H(x, u, J) = \begin{cases} a_{xu} + J(u) & \text{if } u \neq t, \\ a_{xt} & \text{if } u = t \end{cases}$$

where t is the destination

- Minimax Problems

$$H(x, u, J) = \max_{w \in W(x, u)} [g(x, u, w) + \alpha J(f(x, u, w))]$$

RESULTS USING CONTRACTION

- The mappings T_μ and T are sup-norm contraction mappings with modulus α over $B(X)$, and have unique fixed points in $B(X)$, denoted J_μ and J^* , respectively (cf. **Bellman's equation**). *Proof*: From contraction assumption and fixed point Th.
- For any $J \in B(X)$ and $\mu \in \mathcal{M}$,

$$\lim_{k \rightarrow \infty} T_\mu^k J = J_\mu, \quad \lim_{k \rightarrow \infty} T^k J = J^*$$

(cf. **convergence of value iteration**). *Proof*: From contraction property of T_μ and T .

- We have $T_\mu J^* = T J^*$ if and only if $J_\mu = J^*$ (cf. **optimality condition**). *Proof*: $T_\mu J^* = T J^*$, then $T_\mu J^* = J^*$, implying $J^* = J_\mu$. Conversely, if $J_\mu = J^*$, then $T_\mu J^* = T_\mu J_\mu = J_\mu = J^* = T J^*$.
- **Useful bound for J_μ** : For all $J \in B(X)$, $\mu \in \mathcal{M}$

$$\|J_\mu - J\| \leq \frac{\|T_\mu J - J\|}{1 - \alpha}$$

Proof: Take limit as $k \rightarrow \infty$ in the relation

$$\|T_\mu^k J - J\| \leq \sum_{\ell=1}^k \|T_\mu^\ell J - T_\mu^{\ell-1} J\| \leq \|T_\mu J - J\| \sum_{\ell=1}^k \alpha^{\ell-1}$$

RESULTS USING MON. AND CONTRACTION I

- **Existence of a nearly optimal policy:** For every $\epsilon > 0$, there exists $\mu_\epsilon \in \mathcal{M}$ such that

$$J^*(x) \leq J_{\mu_\epsilon}(x) \leq J^*(x) + \epsilon v(x), \quad \forall x \in X$$

Proof: For all $\mu \in \mathcal{M}$, we have $J^* = TJ^* \leq T_\mu J^*$. By monotonicity, $J^* \leq T_\mu^{k+1} J^* \leq T_\mu^k J^*$ for all k . Taking limit as $k \rightarrow \infty$, we obtain $J^* \leq J_\mu$.

Also, choose $\mu_\epsilon \in \mathcal{M}$ such that for all $x \in X$,

$$\|T_{\mu_\epsilon} J^* - J^*\| = \|(T_{\mu_\epsilon} J^*)(x) - (TJ^*)(x)\| \leq \epsilon(1-\alpha)$$

From the earlier error bound, we have

$$\|J_\mu - J^*\| \leq \frac{\|T_\mu J^* - J^*\|}{1-\alpha}, \quad \forall \mu \in \mathcal{M}$$

Combining the preceding two relations,

$$\frac{|J_{\mu_\epsilon}(x) - J^*(x)|}{v(x)} \leq \frac{\epsilon(1-\alpha)}{1-\alpha} = \epsilon, \quad \forall x \in X$$

- **Optimality of J^* over stationary policies:**

$$J^*(x) = \min_{\mu \in \mathcal{M}} J_\mu(x), \quad \forall x \in X$$

Proof: Take $\epsilon \downarrow 0$ in the preceding result.

RESULTS USING MON. AND CONTRACTION II

- **Nonstationary policies:** Consider the set Π of all sequences $\pi = \{\mu_0, \mu_1, \dots\}$ with $\mu_k \in \mathcal{M}$ for all k , and define for any $J \in B(X)$

$$J_\pi(x) = \limsup_{k \rightarrow \infty} (T_{\mu_0} T_{\mu_1} \cdots T_{\mu_k} J)(x), \quad \forall x \in X,$$

(the choice of J does not matter because of the contraction property).

- **Optimality of J^* over nonstationary policies:**

$$J^*(x) = \min_{\pi \in \Pi} J_\pi(x), \quad \forall x \in X$$

Proof: Use our earlier existence result to show that for any $\epsilon > 0$, there is μ_ϵ such that $\|J_{\mu_\epsilon} - J^*\| \leq \epsilon(1 - \alpha)$. We have

$$J^*(x) = \min_{\mu \in \mathcal{M}} J_\mu(x) \geq \min_{\pi \in \Pi} J_\pi(x)$$

Also

$$T^k J \leq T_{\mu_0} \cdots T_{\mu_{k-1}} J$$

Take limit as $k \rightarrow \infty$ to obtain $J \leq J_\pi$ for all $\pi \in \Pi$.

MIT OpenCourseWare
<http://ocw.mit.edu>

6.231 Dynamic Programming and Stochastic Control
Fall 2015

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.