UGP: Essays on Dynamic Programming

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§1. A Multi-Stage Allocation Process

1.1. Introduction

1.1.1. Notation

The standard Dynamic Programming problem desribed throughout will be in metric space X.

Given, $\Phi: \mathbb{X} \times \mathbb{X} \to \mathbb{R}$ and $\Gamma: \mathbb{X} \rightrightarrows \mathbb{R}$

For a fixed x^0 , find a sequence $x_m = x^1, ..., x^m, ...$ such that

$$\max \Phi(x^0, x^1) + \sum_{i=1}^{\infty} \delta^i \Phi(x^i, x^{i+1}), \text{ where } \delta \in (0, 1) \text{ and } x^1 \in \Gamma(x^0), x^{i+1} \in \Gamma(x^i), m = 1, 2, 3, \dots$$

Here δ is called the discount factor.

1.1.2. Functional Equations

Considering a space of Real numbers \mathbb{R} , we can define a functional equation as follows:

$$f(x) = \max_{0 \le y \le x} [g(y) + h(x - y) + f(ay + b(x - y))]$$

where $a, b \in \mathbb{R}$ and g, h are given functions.

This can be seen as an **insurance problem** where functions g and h represent the different places where and initial premium x is invested. The function f represents the value of the policy at the end of the period.

We can define, $R_1(x,y) = g(y) + h(x-y)$, initially the aim is to determine $\max_{0 \le y \le x} R_1(x,y)$. The value of x and y obtained from this maximization can be used to generate the next element of the sequence x_m .

i.e.
$$x_1 = ay + b(x - y) = y_1 + (x_1 - y_1)$$

Further we define $R_2(x, y, y_1) = g(y) + h(x - y) + g(y_1) + h(x_1 - y_1)$ where $(y, y_1) \in [0, x] \times [0, x_1]$

This can be generalized to $R_m(x,y,y_1,...,y_{m-1})=g(y)+h(x-y)+g(y_1)+h(x_1-y_1)+g(y_2)+h(x_2-y_2)+...+g(y_{m-1})+h(x_{m-1}-y_{m-1})$

and the maximization problem becomes

$$\max R_m(x, y, y_1, ..., y_{m-1})$$
 such that

$$\begin{array}{ll} ay + b(x - y) = x_1 & 0 \le y \le x \\ ay_1 + b(x_1 - y_1) = x_2 & 0 \le y_1 \le x_1 \\ \vdots & \vdots & \vdots \\ ay_{m-1} + b(x_{m-1} - y_{m-1}) = x_m & 0 \le y_{m-1} \le x_{m-1} \end{array}$$

Here, $f_N(x) = \max_{t,y_i} R_N(x,y_1,...,y_N)$ and $f_N(x)$ can be interpreted as the maximum return obtained from an N-stage decision with initial premium x and $x \ge 0$.

$$\therefore f_N(x) = \max_{0 \le y \le x} [g(y) + h(x - y) + f_{N-1}(ay + b(x - y))]$$

or

$$f(x) = \max_{0 \le y \le x} [g(y) + h(x - y) + f(ay + b(x - y))] \text{ for } N \to \infty$$

1.2. Existence and Uniqueness Theorems

Under the following assumptions, the functional equation has a unique solution, which is continuous at x=0 and has a value 0 at x=0. Moreover the solution is continuous at $x\geq 0$.

The assumptions are:

1. g and h are continuous on $[0, \infty)$ and g(0) = h(0) = 0

2. If
$$m(x) = \max_{0 \le y \le x} [\max\{|g(y)|, |h(y)|\}]$$
 and $c = \max\{a, b\}$, then $\sum_{n=0}^{\infty} m(c^n x) < \infty \forall x \ge 0$

Proof. We define a mapping T(f,y) = g(y) + h(x-y) + f(ay+b(x-y)) and $f_{N+1}(x) = \max_{0 \le y \le x} T(f_N,y)$, assuming g and h to be non-negative.

The sequence $\{f_N(x)\}$ is an increasing sequence of functions such that $f_N \to f$

Prooving that *T* is bounded above would suffice.

Method 1: We know that $x_1 \leq cx$

$$\therefore ax_1 + b(x_1 - y) \le cx_1 \le c^2 x$$

$$\implies T(f_N, y) \le 2(m(x) + m(cx) + \dots + m(c^N x))$$

Which proves that *T* is bounded above by a constant.

Method 2:

$$f_{N+1}(x) = \sup_{0 \le y \le x} T(f_N, y)$$

$$f(x) \ge \sup_{0 \le y \le x} T(f_N, y)$$

$$f(x) \ge T(f_N, y) \forall y \in [0, x]$$

$$f(x) \ge T(f, y) \text{ as } N \to \infty$$

$$\therefore f(x) = \sup_{0 \le y \le x} T(f, y)$$

Also,

$$T(f_N, y) = g(y) + h(x - y) + f_N(ay + b(x - y))$$

$$\leq g(y) + h(x - y) + f(ay + b(x - y))$$

$$= T(f, y)$$

$$f_{N+1}(x) = \sup_{0 \le y \le x} T(f_N, y)$$
$$\le \sup_{0 \le y \le x} T(f, y)$$
$$f_{N+1}(x) \le \sup_{0 \le y \le x} T(f, y)$$

As $N \to \infty$,

$$f(x) \le \sup_{0 \le y \le x} T(f, y)$$

 $\therefore f(x) = \sup_{0 \le y \le x} T(f, y)$ using the above two inequalities

1.2.1. Existence and Uniqueness with g and h not necessarily non-negative

We can write,

$$\begin{split} f_1(x) &= \max_{0 \leq y \leq x} [g(y) + h(x-y)] \\ &= \max_{y \in [0,x]} [g(y) + h(x-y) + 0] \\ &= \max_{y \in \Gamma(x)} [g(y) + h(x-y)] \end{split} \qquad \Gamma(x) = [0,x] \text{ a set valued map} \end{split}$$

Digression

We will use a generalized form of Berge Maximum Theorem in the following proof, which is described here. **Berge Maximum Theorem (Generalized Form)**

Let Θ and $\mathbb X$ be two metric spaces and $\Gamma:\Theta\rightrightarrows\mathbb X$ is a compact valued correspondence. Let $\phi:\mathbb X\times\Theta\to\mathbb R$ be a continuous function.

Let
$$\sigma(\theta) = argmax\{\phi(x,\theta): x \in \Gamma(\theta)\}$$
 and let $\phi^*(\theta) = \max_x \{\phi(x,\theta): x \in \Gamma(\theta)\}$

Let $\Gamma:\Theta\rightrightarrows\mathbb{X}$ is continuous at some $\theta\in\Theta$. Then,

- 1. $\sigma:\Theta \rightrightarrows \mathbb{X}$ is compact-valued, upper semi-continuous and closed at θ .
- 2. $\phi^*: \Theta \to \mathbb{R}$ is continuous at θ .

Some definitions that follow from above are:

- 1. **Upper semi-continuous Map:** Let $\Gamma: \mathbb{X} \rightrightarrows \mathbb{Y}$, then Γ is upper semi-continuous at $x \in \mathbb{X}$ if for any $\{x^m\} \in \mathbb{X}$ and $\{y^m\} \in \mathbb{Y}$ with $x_0^m \to x$ and $y^m \in \Gamma(x^m)$, the sequence $\{y^m\}$ has a convergent subsequence such that its limit is in $\Gamma(x)$.
- 2. Compact valued Map: $\Gamma : \mathbb{X} \rightrightarrows \mathbb{Y}$ is a compact valued correspondence which is upper semi-continuous if for a compact subset $S \subset \mathbb{X}$, the set $\Gamma(S)$ is compact.
- 3. Closed Map: $\Gamma: \mathbb{X} \rightrightarrows \mathbb{Y}$ is closed at $x \in \mathbb{X}$ if for any $\{x^m\} \in \mathbb{X}$ with $x^m \to x$ and $\{y^m\} \in \mathbb{Y}$ with $y^m \to y$ such that $y^m \in \Gamma(x^m)$ for each m, we have $y \in \Gamma(x)$.

By applying the generalized form of Berge Maximum Theorem, we can show that $f_1(x)$ is continuous.

$$T(f_N, y) = g(y) + h(x - y) + f_N(ay + b(x - y))$$
 and $f_{N+1}(x) = \max_{0 \le y \le x} T(f_N, y)$.

$$f_{N+1}(x) = T(f_N, y_N) \ge T(f_N, y_N + 1)$$

$$f_{N+2}(x) = T(f_N + 1, y_N + 1) \ge T(f_N + 1, y_N)$$

Define,

$$u_N(x) = \sup_{0 \le z \le x} |f_N(z) - f_N + 1(z)|$$

and

$$f_{N+1}(x) - f_{N+2}(x) \ge T(f_N, y_{N+1}) - T(f_N + 1, y_{N+1})$$
$$f_{N+1}(x) - f_{N+2}(x) \le T(f_N, y_N) - T(f_N + 1, y_N)$$

From these two inequalities, we get

$$|f_{N+1}(x) - f_{N+2}(x)| \le \max\{|T(f_N, y_{N+1}) - T(f_N + 1, y_{N+1})|, |T(f_N, y_N) - T(f_N + 1, y_N)|\}$$

$$|T(f_N, y_{N+1}) - T(f_N + 1, y_{N+1})| = |f_N(ay_N + b(x - y_N)) - f_{N+1}(ay_{N+1} + b(x - y_{N+1})) \le u_N(cx)$$

$$|T(f_N, y_N) - T(f_N + 1, y_N)| = |f_N(ay_N + b(x - y_N)) - f_{N+1}(ay_N + b(x - y_N)) \le u_N(cx)$$

$$\therefore ay + b(x - y) \le cx \text{ where } c = \max\{a, b\}$$

$$\therefore u_{N+1}(x) \le u_N(cx)$$
 for $N \in \mathbb{N}$

We now need to estimate $u_1(x)$

$$f_2(x) = T(f_1, y_1)$$

$$f_1(x) - f_2(x) \ge g(y_1) + h(x - y_1) - T(f_1, y_1)$$

$$f_2(x) - f_1(x) \le f_1(ay_1 + b(x - y_1))$$

Let y_0 be the maximizing point to obtain $f_1(x)$ i.e. $f_1(x) = g(y_0) + h(x - y_0)$ and $f_2(x) \ge T(f_1, y_0)$

$$f_1(x) - f_2(x) \ge g(y_0) + h(x - y_0) - T(f_1, y_0)$$

$$= -f_1(ay_0 + b(x - y_0))$$

$$f_2(x) - f_1(x) \ge f_1(ay_0 + b(x - y_0))$$

$$|f_2(x) - f_1(x)| \le \max\{f_1(ay_1 + b(x - y_1)), |f_1(ay_0 + b(x - y_0))|\} \implies |f_2(x) - f_1(x)| \le 2m(cx) \forall x \in [0, x]$$

$$\therefore u_1(x) \le 2m(cx)$$

$$u_{N+1}(x) \le u_N(cx) \le u_{N-1}(c^2x) \le \dots \le u_1(c^Nx) \le 2m(c^{N+1}x)$$

$$\therefore u_{N+1}(x) \le 2m(c^{N+1}x) \text{ and } u_1(x) \le 2m(cx) \text{ for } N \in \mathbb{N}$$

$$\implies \boxed{u_N(x) \le 2m(c^Nx) \text{ for } N \in \mathbb{N}}$$

From the above result we can say that,

$$\sum_{N=1}^{\infty} u_N(x) \le \sum_{N=1}^{\infty} 2m(c^N x)$$

that is

$$\sum_{N=1}^{\infty} u_N(x) \le \infty \text{(convergent)}$$

$$\implies u_N(x) \to 0 \text{ as } N \to \infty$$

 $\implies f_N(x)$ is a cauchy sequence (which then implies that $f_N - f_M$ can be expressed as a sum of differences).

$$\therefore f(x) = \lim_{N \to \infty} f_N(x)$$
 exists and is continuous $\forall x \ge 0$

Now we know that,

$$f_{N+1}(x) \ge T(f_N, y)$$

 $\ge g(y) + h(x - y) + f_N(ay + b(x - y)) \text{ for } y \in [0, x]$

For large N,

$$f(x) \ge g(y) + h(x - y) + f(ay + b(x - y))$$

$$\ge \max_{0 \le y \le x} T(f, y) \tag{A}$$

For a bounded sequence $\{y_N\}$ with $y_N \to y^*$ for some $y^* \in [0, x]$ and $N \to \infty$ we have,

$$f(x) = T(f, y^*)$$

$$\leq \max_{0 \leq y \leq x} T(f, y)$$
(B)

using (A) and (B),

$$f(x) = \max_{0 \le y \le x} T(f, y)$$

1.2.2. Uniqueness

By proving the existence, we know that,

$$f(x) = \max_{0 \le y \le x} T(f, y)$$

Assume on the contrary that there are two such solutions f&F

$$f(x) = T(f, y) \ge T(f, w)$$

$$F(x) = T(F, w) \ge T(F, y)$$

Now,

$$\begin{split} |f(x) - F(x)| &\leq \max\{|T(f,y) - T(F,y)|, |T(f,w) - T(F,w)|\} \\ &\leq \max\{|f(ay + b(x-y) - F(ay + b(x-y))|, |f(aw + b(x-w) - F(aw + b(x-w))|\} \end{split}$$

From previous definition, $u(x) = \sup_{0 \le z \le x} f(z) - F(z)$ We know that $u(x) \ge 0$ and $u(x) \to 0$ as $x \to \infty$.

$$\therefore 0 \ge u(x) \ge |f(x) - F(x)|$$

$$\implies |f(x) - F(x)| \to 0 \text{ as } x \to \infty$$

$$\implies f(x) = F(x) \text{ for all } x \ge 0$$

... The solution obtained is unique.

1.3. Approximations

Theorem 1.1. Let $f_0(x)$ satisfy the following conditions:

- $f_0(x)$ is continuous for $x \ge 0$
- $f_0(0) = 0$

Then if the conditions of existence are fulfilled, the sequence defined by

$$f_{n+1}(x) = \max_{0 \le y \le x} T(f_n, y) \text{ for } n \in \mathbb{W}$$

converges to the solution f(x) of the problem, uniformly in any finite interval.

1.3.1. Approximation in the policy space

We call a sequence of allocations, i.e. a sequence of admissible choices of y, a policy and a policy which yields f(x) an optimal policy.

The duality that exists in the theory of dynamic programmingarises from the interconnection between the functions f(x) which measure the maximum return and the policies which yield these maximum returns. Actually a policy is a function, since a policy is a determination of y as a function of x. If the policy is not unique, y will not be a single-valued function of x.

Just as we can approximate in the space of the functions f(x), so we can approximate in the space of policies, y(x).

Theorem 1.2. Let $f_0(x)$ be the result of an initial approximation in the policy space, that is, $f_0(x) = T(f_0, y_0(x))$ where $y_0(x)$ is any continuous function of x satisfying $0 \le y_0(x) \le x$, the sequence defined by

$$f_{n+1}(x) = \max_{0 \le y \le x} T(f_n, y) \text{ for } n \in \mathbb{N}$$

converges to the solution f(x), uniformly and in a finite interval.

Proof. By the definition of $f_n(x)$, we have

$$f_{n+1} \ge f_n \ge f_0 \ \forall \ \mathbb{N}$$

Proving that $f_0(x)$ is continuous, will imply continuity of $f_n(x)$.

$$f_0(x) = g(y_0) + h(x - y_0) + \cdots$$

where g and h are continuous functions.

Hence, $f_0(x)$ which is iteratively obtained from g(x) and h(x) is also continuous in a finite interval.

We can construct examples for such a solution, say

Example. f(x) = T(f, y(x)) be an iterative form of the solution.

If the optimal policy consisted of the choice y=0 continually, the solution can be represented by the equation f(x)=h(x)+f(bx), with the value of b such that $h(b^nx)<\infty$ i.e.,

$$f(x) = h(x) + f(bx)$$

$$= h(x) + h(bx) + f(b^{2}x)$$

$$= h(x) + h(bx) + h(b^{2}x) + f(b^{3}x)$$

$$= \cdots$$

$$= \sum_{n=0}^{\infty} h(b^{n}x)$$

Here, $h(b^n x) < \infty$ for all $n \in \mathbb{N}$, so the series converges and f(x) is a continuous function of x.

1.4. Properties of the solution

1.4.1. Convexity

Theorem 1.3. If, along with existence assumptions, we impose the conditions that g and h be convex functions of x, then f(x) will be a convex function, and for each value of x, y will equal 0 or x.

Proof. g and *h* are convex functions and,

$$f_1(x) = \max_{0 \le y \le x} (g(y) + h(x - y))$$

In the interval [0, x], g and h are convex, hence g(y) + h(x - y) is also convex.

 \therefore Maximum of a convex function must occur at the endpoints of the interval, i.e. y = 0 or y = x. We can say that,

$$f_1(x) = \max(g(y), h(x))$$

Then,

$$f_2(x) = \max_{0 \le y \le x} (g(y) + h(x - y) + f_1(ay + b(x - y)))$$

similarly follows,

$$f_2(x) = \max(g(x) + f_1(ax), h(x) + f_1(bx))$$

Therefore we can inductively conclude that $f_n(x)$ is convex thus, the limit of the sequence f(x) is also convex.

1.4.2. Concavity

Theorem 1.4. [Weak Version] If, along with existence assumptions, we impose the conditions that g and h be strictly concave functions of x, then f(x) will be a strictly concave function, with an optimal policy which is unique.

Lemma 1.5. If G(x,y) is a concave function of x and y for $x,y \ge 0$, then f(x) as defined by $f(x) = \max_{0 \le y \le x} G(x,y)$ is a concave function of x for x > 0.

Theorem 1.6. [Strong Version] Assuming that,

- 1. g(x) and h(x) are both strictly concave for x > 0, monotone increasing with continuous derivatives and that g(0) = h(0) = 0.
- 2. $\frac{g'(0)}{(1-a)} > \frac{h'(0)}{(1-b)}, h'(0) > g'(\infty), b > a.$

Then the optimal policy has the following form:

- 1. y = x for $0 < x < \bar{x}$, where \bar{x} is the root of $h'(0) = g'(x) + (b a)g'(ax) + (b a)ag'(a^2x) + \cdots$
- 2. y = y(x) for $x > \bar{x}$ where y(x) is a function satisfying the inequalities O < y(x) < x, and y(x) is the solution of g'(y) h'(x-y) + (a-b)f'(ay+b(x-y)) = 0

Proof. It is known that,

$$f_1(x) = \max_{0 \le y \le x} (g(y) + h(x - y))$$

 $g'(0) > h'(0) \implies g'(y) - h'(x-y) > 0$ for $y \in [0,x]$ with the roots at y = x and in the interval $[0,x'], \forall x' > y$. At y = x, the value x^* for which $g'(x^*) - h'(0) = 0$ is the root of the equation g'(x) - h'(0) = 0.

This equation has precisely one solution due to the fact that g and h are strictly concave.

For $x \ge x^*$, let $y_1(x)$ be the unique solution of g'(y) = h'(x - y). Then, $f_1(x) = g(y_1) + h(x - y_1)$ and $f'(x) = [g'(y_1) - h'(x - y_1)] \frac{dy_1}{dx} + h'(x - y_1) = h'(x - y_1)$ for $x > x_1$. Further,

$$f_2(x) = \begin{cases} g'(x), & \text{if } 0 \le x \le x_2 \\ h'(x - y_2) + bf_1(ay_2 + b(x - y_2)), & \text{if } x > x_2 \end{cases}$$

where $y_2(x)$ is the unique solution of $g'(y) = h'(x-y) + (b-a)f_1(ay+b(x-y))$ Inductively,

$$f_{n+1}(x) = \max_{0 \le y \le x} \left\{ g(y) + h(x-y) + f_n(ay + b(x-y)) \right\}$$

The solution space hence can be divided into three parts from, $[0, x_2]$, $[x_2, x_1]$ and $[x_1, \infty)$, wherein the inequality $f'_{n+1}(x) \ge f'_n(x)$ holds for all x. For n = 1,

$$f_2'(x) = \frac{bg'(y_2) - ah'(x - y_2)}{b - a}$$
 for $x \ge x_2$

and

$$f'_1(x) = \frac{bg'(y_1) - ah'(x - y_1)}{b - a}$$
 for $x \ge x_1$

For $[x_1, \infty)$, and f' is monotonically decreasing and $y_2 < y_1$, $f'_2(x) > f'_1(x)$. In, $[0, x_2]$, $f'_2(x) = f'_1(x)$.

In $[x_2, x_1]$,

$$f_1'(x) = g'(x)$$

$$f_2'(x) = \frac{bg'(y_2) - ah'(x - y_2)}{b - a} \text{ for } x \ge x_2$$

Hence, in the interval, $0 \le y_2 \le x$ for $x \in [x_2, x_1]$,

$$f_2'(x) \ge \frac{bg'(x) - ah'(0)}{b - a} > g'(x)$$

And we assume that, $g'(x) \geq h'(0)$.: $f_2'(x) > f_1'(x)$ Or, for $x_1 > x_2 > x_3 > \dots > x_n > 0$, $f_1'(x) \leq f_2'(x) \leq f_3'(x) \leq \dots \leq f_n'(x) \leq \dots$

Since $f_n(x)$ converges to f(x), $f'_n(x)$ to f'(x), $y_n(x)$ to y(x) and x_n to \bar{x} , the solution has the form,

$$f(x) = \max_{0 \le y \le x} \{g(y) + h(x - y) + f(ay + b(x - y))\}\$$

$$\frac{df}{dx} = 0 \implies g'(y) - h'(x - y) + (a - b)f'(ay + b(x - y)) = 0$$

[Bel57]

References

[Bel57] Richard Bellman. Dynamic Programming. Dover Publications, 1957.