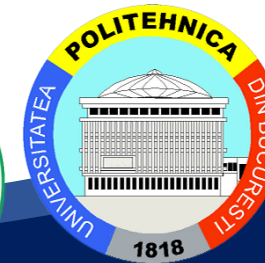




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Advanced Optimization: Techniques and Industrial Applications



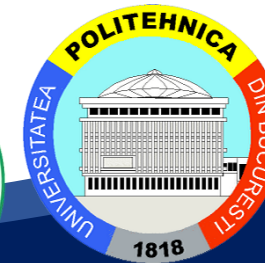
Curriculum Development
of Master's Degree Program in
Industrial Engineering for Thailand Sustainable Smart Industry



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Session 1.5: Dynamic Programming



Curriculum Development
of Master's Degree Program in
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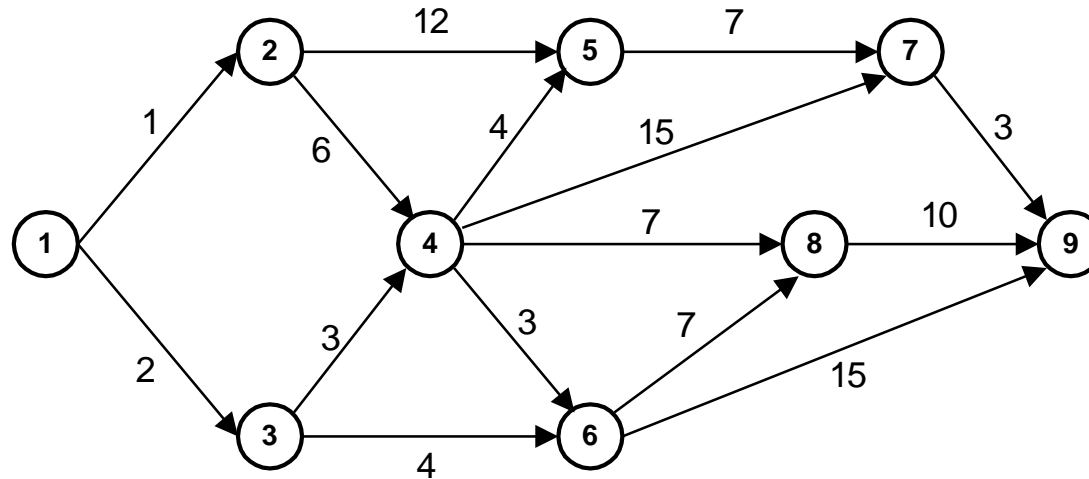
Introduction

- A specific type of mathematical programming in which the optimal solution of the original problem is found by solving *a chain of subproblems*.
- In dynamic programming, the optimal solution of one subproblem will be used as input to the next subproblem. When the last subproblem is solved, the optimal solution for the *entire* problem is achieved, which includes the solution of the original problem.
- Linkage between the stages of a DP problem is performed through *recursive computations*. Depending on the nature of the problem at hand, *forward recursive equation* or *backward recursive equation* will be developed for finding solution.



Example 1: Shortage Path Problem

Find the shortage path from node 1 to node 9 of the network:



Define:

f_i : minimum total travel time from node i to node 9.

t_{ij} : travel time through the directed arc (i, j) .



On an arbitrary arc (i, j) , it can be seen that:

$$f_i \leq t_{ij} + f_j \quad i \neq 9, \forall j$$

Hence:

$$f_i \leq \min_j \{t_{ij} + f_j\} \quad i \neq 9$$

However, the shortage path from node i to node 9 should include some intermediate node j (if these intermediate nodes exist). Then,

$$f_i = \min_j \{t_{ij} + f_j\} \quad i \neq 9$$

The above equation is the *recursive equation* (or functional equation) of the shortage path problem in the *backward* form.

Based on the recursive equation, the optimal solution can be found as follows:

$$f_9 = 0$$

$$f_8 = t_{89} + f_9 = 10 + 0 = 10$$

$$f_7 = t_{79} + f_9 = 3 + 0 = 3$$

$$f_6 = \min \begin{cases} t_{68} + f_8 \\ t_{69} + f_9 \end{cases} = \min \begin{cases} 7 + 10 \\ 15 + 0 \end{cases} = 15$$

$$f_5 = t_{57} + f_7 = 7 + 3 = 10$$

$$f_4 = \min \begin{cases} t_{45} + f_5 \\ t_{46} + f_6 \\ t_{47} + f_7 \\ t_{48} + f_8 \end{cases} = \min \begin{cases} 4 + 10 \\ 3 + 15 \\ 15 + 3 \\ 7 + 10 \end{cases} = 14$$



Introduction

$$f_3 = \min \begin{cases} t_{34} + f_4 \\ t_{36} + f_6 \end{cases} = \min \begin{cases} 3 + 14 \\ 4 + 15 \end{cases} = 17$$

$$f_2 = \min \begin{cases} t_{24} + f_4 \\ t_{25} + f_5 \end{cases} = \min \begin{cases} 6 + 14 \\ 12 + 10 \end{cases} = 20$$

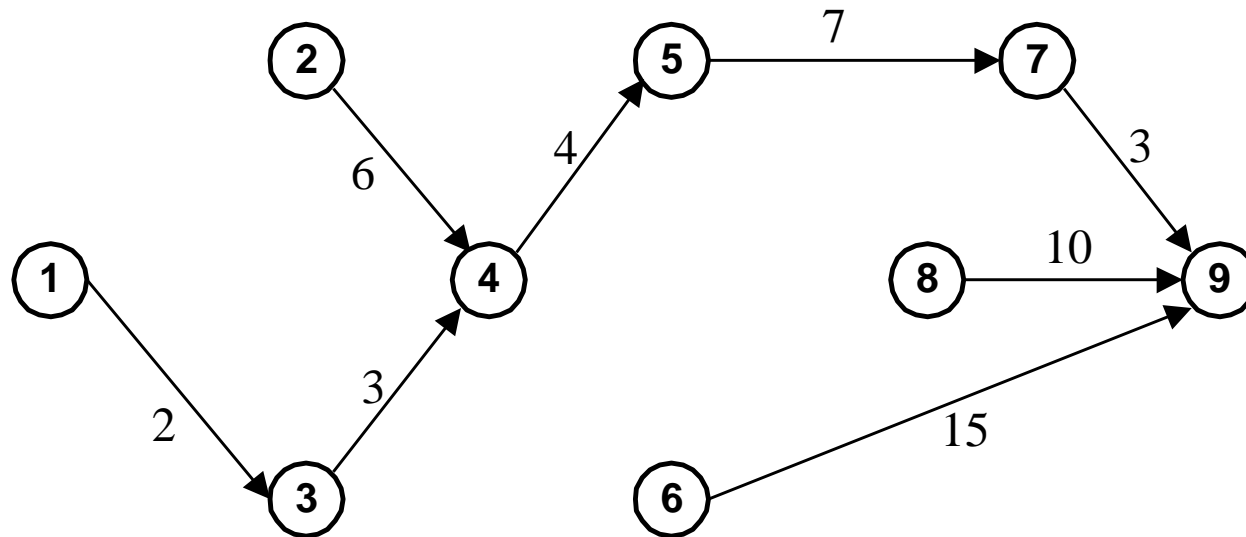
$$f_1 = \min \begin{cases} t_{12} + f_2 \\ t_{13} + f_3 \end{cases} = \min \begin{cases} 1 + 20 \\ 2 + 17 \end{cases} = 19$$

Shortage path from node 1 to node 9: 1-3-4-5-7-9 with total travel time = 19.



Introduction

It is noted that each subproblem is associated with a network's node and when the shortage path from node 1 to node 9 is determined, we also know the shortage paths from **every nodes of the network to node 9**.



Introduction

The above problem can also be solved by use of *forward* recursive equation as presented below
Define:

f_j : minimum total travel time from node 1 to node j .

t_{ij} : travel time through the directed arc (i, j) .

On an arbitrary arc (i, j) , it can be seen that:

$$f_j \leq t_{ij} + f_i \quad j \neq 1, \forall i$$

Hence:

$$f_j \leq \min_i \{t_{ij} + f_i\} \quad j \neq 1$$

However, the shortest path from node 1 to node j should include some intermediate node i (if these intermediate nodes exist). Then,

$$f_j = \min_i \{t_{ij} + f_i\} \quad j \neq 1$$



Solution can be found recursively as follows:

$$f_1 = 0$$

$$f_2 = t_{12} + f_1 = 1 + 0 = 1$$

$$f_3 = t_{13} + f_1 = 2 + 0 = 2$$

$$f_4 = \min \begin{cases} t_{24} + f_2 \\ t_{34} + f_3 \end{cases} = \min \begin{cases} 6 + 1 \\ 3 + 2 \end{cases} = 5$$

$$f_5 = \min \begin{cases} t_{25} + f_2 \\ t_{45} + f_4 \end{cases} = \min \begin{cases} 12 + 1 \\ 4 + 5 \end{cases} = 9$$

$$f_6 = \min \begin{cases} t_{36} + f_3 \\ t_{64} + f_4 \end{cases} = \min \begin{cases} 4 + 2 \\ 3 + 5 \end{cases} = 6$$



$$f_7 = \min \begin{cases} t_{47} + f_4 \\ t_{57} + f_5 \end{cases} = \min \begin{cases} 15 + 5 \\ 7 + 9 \end{cases} = 16$$

$$f_8 = \min \begin{cases} t_{48} + f_4 \\ t_{68} + f_6 \end{cases} = \min \begin{cases} 7 + 5 \\ 7 + 6 \end{cases} = 12$$

$$f_9 = \min \begin{cases} t_{69} + f_6 \\ t_{79} + f_7 \\ t_{89} + f_8 \end{cases} = \min \begin{cases} 15 + 6 \\ 3 + 16 \\ 10 + 12 \end{cases} = 19$$

The solution from forward recursive equation gives the shortage paths from **node 1 to every other nodes** of the network, not only to node 9.



Note:

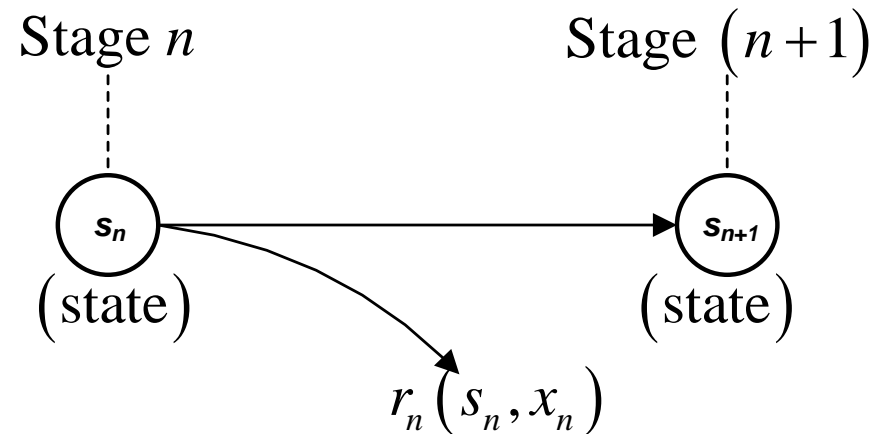
Not all DP programs can be solved by both forward and backward recursive techniques. The use of backward recursion or forward recursion depends on the specific structure of the problem under consideration.

Introduction

Bellman's Principle of Optimality

An optimal policy has the property that whatever the initial state and the initial decision are, the remaining decisions must constitute an optimal policy with regard to the state resulting from the first decision

The basic DP approach can be illustrated by the following diagram:



Introduction

Bellman's Principle of Optimality

- At state s_n in stage n , if decision x_n is taken the current state s_n will be transferred to a new state s_{n+1} in stage $(n + 1)$.
- A revenue $r_n(s_n, x_n)$ will be obtained by decision x_n taken at state s_n
- The new state s_{n+1} is also a function of s_n and x_n , and can be expressed in form of a transformation function: $s_{n+1} = t_n(s_n, x_n)$.

Introduction

Bellman's Principle of Optimality

In case of **maximization problem** and **backward recursive technique** is employed, if we denote $f_n(s_n)$ as the **maximum total revenue** obtained when the system moves from stage n to stage N (the last stage), given the observed state at stage n is s_n , then:

$$f_n(s_n) = \max_{x_n \in D(s_n)} \left\{ r_n(s_n, x_n) + f_{n+1}(t_n(s_n, x_n)) \right\}$$

In which $D(s_n)$ is the set of all possible decisions of a *given* state s_n at stage n (decision set).

Introduction

Bellman's Principle of Optimality

Similarly, when the **forward recursive** technique is employed, if we denote $f_n(s_n)$ as the **maximum total revenue** when the system move from stage 1 to stage n , given the observed state at stage n is s_n , then:

$$f_{n+1}(s_{n+1}) = \max_{x_n \in D(s_{n+1})} \left\{ r_n(s_n, x_n) + f_n(s_n) \right\}$$

In which $D(s_{n+1})$ is the set of all possible decisions x_n at stage n such that these decisions will help to transfer the states s_n 's at stage n to a **predefined** state s_{n+1} in stage $(n + 1)$, i.e., $s_{n+1} = t_n(s_n, x_n)$.





Introduction

Bellman's Principle of Optimality

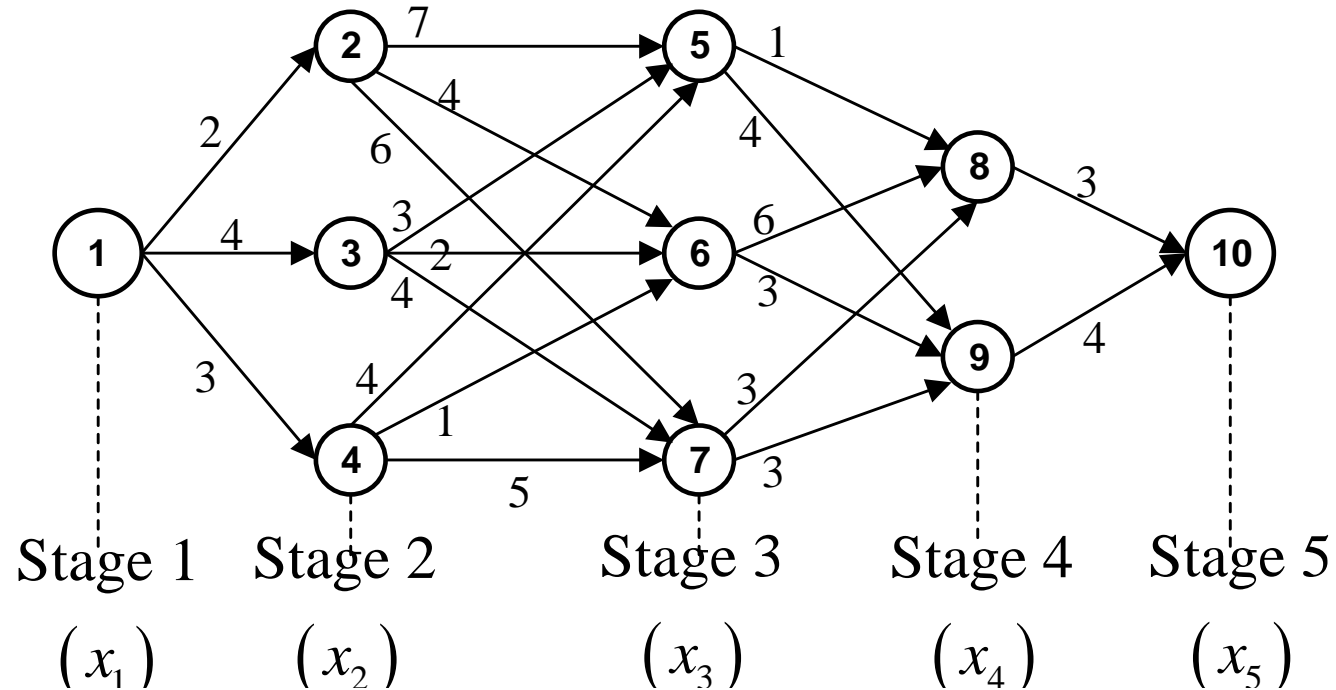
The Bellman' optimality principle can help to establish recursive equations when the structure of the problem can be arranged in stages.



Introduction

Bellman's Principle of Optimality

Example 2: Find the shortest path from node 1 to node 10 of the network



Introduction

Bellman's Principle of Optimality

Applying backward recursive approach, the solution can be found as follows:

Stage 4:

s_4	x_4	$r_4(s_4, x_4) + f_5(t_4(s_4, x_4))$	$f_4(s_4)$	x_4^*
		10		
Node 8		3	3	10
Node 9		4	4	10

In stage 4, the state s_4 can be *node 8* or *node 9*, and the only decision (i.e., x_4) that can be taken is *go to node 10*.



Introduction

Bellman's Principle of Optimality

When the state is *node 8*:

$$r_4(s_4, x_4) = r_4(8, 10) = 3 \quad f_5(t_4(s_4, x_4)) = f_5(10) = 0$$
$$\Rightarrow f_4(s_4) = f_4(8) = 3$$

When the state is *node 9*:

$$r_4(s_4, x_4) = r_4(9, 10) = 4 \quad f_5(t_4(s_4, x_4)) = f_5(10) = 0$$
$$\Rightarrow f_4(s_4) = f_4(9) = 4$$

Due to the fact that there exists only one possible decision, that decision is the optimal decision.



$$f_3^*(s_3, x_3) + f_4(t_3(s_3, x_3))$$

MSE 4.0

Introduction

Bellman's Principle of Optimality

Stage 3:

s_3	x_3	$r_3(s_3, x_3) + f_4(t_3(s_3, x_3))$	$f_3(s_3)$	x_3^*
		8	9	
Node 5	1 + 3 = 4	4 + 4 = 8	4	8
Node 6	6 + 3 = 9	3 + 4 = 7	7	9
Node 7	3 + 3 = 6	3 + 4 = 7	6	8

In this stage, the state s_3 can be *node 5*, *node 6* or *node 7*. The decision x_3 can be *go to node 8* or *go to node 9*.



Introduction

Bellman's Principle of Optimality

When the state is *node 5*:

If the decision is *go to node 8*:

$$r_3(s_3, x_3) = r_3(5, 8) = 1$$

$$f_4(t_3(s_3, x_3)) = f_4(8) = 3$$

If the decision is *go to node 9*:

$$r_3(s_3, x_3) = r_3(5, 9) = 4$$

$$f_4(t_3(s_3, x_3)) = f_4(9) = 4$$

$$\Rightarrow f_3(s_3) = f_3(5) = \text{Min}\{1 + 3, 4 + 4\} = 4$$

\Rightarrow Optimal decision: *go to node 8*



Introduction

Bellman's Principle of Optimality

When the state is *node 6*:

If the decision is *go to node 8*:

$$r_3(s_3, x_3) = r_3(6, 8) = 6$$

$$f_4(t_3(s_3, x_3)) = f_4(8) = 3$$

If the decision is *go to node 9*:

$$r_3(s_3, x_3) = r_3(6, 9) = 3$$

$$f_4(t_3(s_3, x_3)) = f_4(9) = 4$$

$$\Rightarrow f_3(s_3) = f_3(6) = \text{Min}\{6 + 3, 3 + 4\} = 7$$

\Rightarrow Optimal decision: *go to node 9*



Introduction

Bellman's Principle of Optimality

When the state is node 7:

If the decision is *go to node 8*:

$$r_3(s_3, x_3) = r_3(7, 8) = 3$$

$$f_4(t_3(s_3, x_3)) = f_4(8) = 3$$

If the decision is *go to node 9*:

$$r_3(s_3, x_3) = r_3(7, 9) = 3$$

$$f_4(t_3(s_3, x_3)) = f_4(9) = 4$$

$$\Rightarrow f_3(s_3) = f_3(7) = \text{Min}\{3 + 3, 3 + 4\} = 6$$

\Rightarrow Optimal decision: *go to node 8*



Introduction

Bellman's Principle of Optimality

Stage 2:

s_2	x_2	$r_2(s_2, x_2) + f_3(t_2(s_2, x_2))$			Min $f_2(s_2)$	x_2^*
		5	6	7		
Node 2		$7 + 4 = 11$	$4 + 7 = 11$	$6 + 6 = 12$	11	5 or 6
Node 3		$3 + 4 = 7$	$2 + 7 = 9$	$4 + 6 = 10$	7	5
Node 4		$4 + 4 = 8$	$1 + 7 = 8$	$5 + 6 = 11$	8	5 or 6

In this stage, the state s_2 can be *node 2*, *node 3* or *node 4*. The decision x_2 can be *go to node 5*, *go to node 6*, or *go to node 7*.



Introduction

Bellman's Principle of Optimality

When the state is *node 2*:

If the decision is *go to node 5*:

$$r_2(s_2, x_2) = r_2(2, 5) = 7$$

$$f_3(t_2(s_2, x_2)) = f_3(5) = 4$$

If the decision is *go to node 6*:

$$r_2(s_2, x_2) = r_2(2, 6) = 4$$

$$f_3(t_2(s_2, x_2)) = f_3(6) = 7$$

If the decision is *go to node 7*:

$$r_2(s_2, x_2) = r_2(2, 7) = 6$$

$$f_3(t_2(s_2, x_2)) = f_3(7) = 6$$

$$\Rightarrow f_2(s_2) = f_2(2) = \text{Min}\{7 + 4, 4 + 7, 6 + 6\} = 11$$

\Rightarrow Optimal decision: *go to node 5* or *go to node 6*



Introduction

Bellman's Principle of Optimality

When the state is *node 3*:

If the decision is *go to node 5*:

$$r_2(s_2, x_2) = r_2(3, 5) = 3$$

$$f_3(t_2(s_2, x_2)) = f_3(5) = 4$$

If the decision is *go to node 6*:

$$r_2(s_2, x_2) = r_2(3, 6) = 2$$

$$f_3(t_2(s_2, x_2)) = f_3(6) = 7$$

If the decision is *go to node 7*:

$$r_2(s_2, x_2) = r_2(3, 7) = 4$$

$$f_3(t_2(s_2, x_2)) = f_3(7) = 6$$

$$\Rightarrow f_2(s_2) = f_2(3) = \text{Min}\{3 + 4, 2 + 7, 4 + 6\} = 7$$

\Rightarrow Optimal decision: *go to node 5*



Introduction

Bellman's Principle of Optimality

When the state is *node 4*:

If the decision is *go to node 5*:

$$r_2(s_2, x_2) = r_2(4, 5) = 4$$

$$f_3(t_2(s_2, x_2)) = f_3(5) = 4$$

If the decision is *go to node 6*:

$$r_2(s_2, x_2) = r_2(4, 6) = 1$$

$$f_3(t_2(s_2, x_2)) = f_3(6) = 7$$

If the decision is *go to node 7*:

$$r_2(s_2, x_2) = r_2(4, 7) = 5$$

$$f_3(t_2(s_2, x_2)) = f_3(7) = 6$$

$$\Rightarrow f_2(s_2) = f_2(3) = \text{Min}\{4 + 4, 1 + 7, 5 + 6\} = 8$$

\Rightarrow Optimal decision: *go to node 5* or *go to node 6*



Introduction

Bellman's Principle of Optimality

Stage 1:

s_1	x_1	$r_1(s_1, x_1) + f_2(t_1(s_1, x_1))$	$f_1(s_1)$	x_1^*		
		2	3	4		
Node 1		$2 + 11 = 13$	$4 + 7 = 11$	$3 + 8 = 11$	11	3 or 4

In this stage, the state s_1 is *node 1*. The decisions x_1 can be *go to node 2*, *go to node 3*, or *go to node 4*.



Introduction

Bellman's Principle of Optimality

If the decision is *go to node 2*:

$$r_1(s_1, x_1) = r_1(1, 2) = 2$$

$$f_2(t_1(s_1, x_1)) = f_2(2) = 11$$

If the decision is *go to node 3*:

$$r_1(s_1, x_1) = r_1(1, 3) = 4$$

$$f_2(t_1(s_1, x_1)) = f_2(3) = 7$$

If the decision is *go to node 4*:

$$r_1(s_1, x_1) = r_1(1, 4) = 3$$

$$f_2(t_1(s_1, x_1)) = f_2(4) = 8$$

$$\Rightarrow f_1(s_1) = f_1(1) = \text{Min}\{2 + 11, 4 + 7, 3 + 8\} = 11$$

\Rightarrow Optimal decision: *go to node 3* or *go to node 4*





Introduction

Bellman's Principle of Optimality

The optimal solution has been found. There exist three shortage paths from node 1 to node 10:

1-3-5-8-10, 1-4-5-8-10, 1-4-6-9-10



Introduction

Bellman's Principle of Optimality

Example 3:

Five medical teams will be dispatched to 3 regions to help improve medical care. The performance is measured by the expected additional person-years of life. The estimated performance measures are given in the table:

No. of Teams	Additional person-years life (in 1000 units)		
	Region 1	Region 2	Region 3
0	0	0	0
1	45	20	50
2	70	45	70
3	90	75	80
4	105	110	100
5	120	150	130



Introduction

Bellman's Principle of Optimality

The problem is to allocate the medical teams so that the total additional person-years of life can be maximized.

Denote:

x_n : number of teams to be allocated to region n ($n = 1, 2, 3$).

s_n : number of teams available for allocation to the regions
 $n, n + 1, \dots, 3$.

$p_n(x_n)$: the measure of performance from allocation x_n teams to region n .

$f_n(s_n)$: Total maximum performance obtained when s_n teams are
allocated to regions $n, n + 1, \dots, 3$.

Introduction

Bellman's Principle of Optimality

The problem can be formulated as follows:

$$\begin{aligned} f_1(5) &= \text{Max } p_1(x_1) + p_2(x_2) + p_3(x_3) \\ \text{s.t.} \quad &x_1 + x_2 + x_3 \leq 5 \\ &x_j \geq 0 \text{ and integer } \forall j = 1, 2, 3 \end{aligned}$$

The above problem can be considered as embedded in the following chains of subproblems:

$$\begin{aligned} f_n(s_n) &= \text{Max } \sum_{i=n}^3 p_i(x_i) \\ \text{s.t.} \quad &\sum_{i=n}^3 x_i \leq s_n \\ &x_j \geq 0 \text{ and integer } \forall j = n, \dots, 3 \end{aligned}$$



Introduction

Bellman's Principle of Optimality

The backward recursive equation can be developed as:

$$f_n(s_n) = \max_{\substack{0 \leq x_n \leq s_n \\ x_n \text{ integer}}} \{p_n(x_n) + f_{n+1}(s_n - x_n)\}$$

Noting that $f_4(s_4) = 0$, the solution can be obtained as follows:

Introduction

Bellman's Principle of Optimality

$$\underline{n = 3}$$

No. of teams	$p_3(x_3) + f_4(s_4) = p_3(x_3) + f_4(s_3 - x_3)$						Max $f_3(s_3)$	x_3^*
	$x_3 = 0$	$x_3 = 1$	$x_3 = 2$	$x_3 = 3$	$x_3 = 4$	$x_3 = 5$		
$s_3 = 0$	0	-	-	-	-	-	0	0
$s_3 = 1$	0	50	-	-	-	-	50	1
$s_3 = 2$	0	50	70	-	-	-	70	2
$s_3 = 3$	0	50	70	80	-	-	80	3
$s_3 = 4$	0	50	70	80	100	-	100	4
$s_3 = 5$	0	50	70	80	100	130	130	5



Introduction

Bellman's Principle of Optimality

$$\underline{n = 2}$$

No. of teams	$p_2(x_2) + f_3(s_3) = p_2(x_2) + f_3(s_2 - x_2)$						Max $f_2(s_2)$	x_2^*
	$x_2 = 0$	$x_2 = 1$	$x_2 = 2$	$x_2 = 3$	$x_2 = 4$	$x_2 = 5$		
$s_2 = 0$	0+0 = 0	-	-	-	-	-	0	0
$s_2 = 1$	0+50 = 50	20+0 = 20	-	-	-	-	50	0
$s_2 = 2$	0+70 = 70	20+50 = 70	45+0 = 45	-	-	-	70	0-1
$s_2 = 3$	0+80 = 80	20+70 = 90	45+50 = 95	75+0 = 75	-	-	95	2
$s_2 = 4$	0+100 = 100	20+80 = 100	45+70 = 115	75+50 = 125	110+0 = 110	-	125	3
$s_2 = 5$	0+130 = 130	20+100 = 120	45+80 = 125	75+70 = 145	110+50 = 160	150+0 = 150	160	4



Introduction

Bellman's Principle of Optimality

$$\underline{n = 1}$$

No. of teams	$p_1(x_1) + f_2(s_2) = p_1(x_1) + f_2(s_1 - x_1)$						Max $f_1(s_1)$	x_1^*
	$x_1 = 0$	$x_1 = 1$	$x_1 = 2$	$x_1 = 3$	$x_1 = 4$	$x_1 = 5$		
$s_1 = 0$	*	-	-	-	-	-	*	*
$s_1 = 1$	*	*	-	-	-	-	*	*
$s_1 = 2$	*	*	*	-	-	-	*	*
$s_1 = 3$	*	*	*	*	-	-	*	*
$s_1 = 4$	*	*	*	*	*	-	*	*
$s_1 = 5$	0+160 = 160	45+125 = 170	70+95 = 165	90+70 = 160	115+50 = 165	120+0 = 120	170	1

(*: no need to determine those values)

Optimal solution: $x_1^* = 1, x_2^* = 3, x_3^* = 1$; optimal objective function 170.



Introduction

Bellman's Principle of Optimality

Example 4: Resource Allocation Problem

Consider the single resource allocation problem to produce N products:

$$\begin{aligned} \text{Max} \quad & p_1(x_1) + p_2(x_2) + \cdots + p_N(x_N) \\ \text{s.t.} \quad & c_1(x_1) + c_2(x_2) + \cdots + c_N(x_N) \leq K \\ & x_j \in \Omega_j \quad \forall j = 1, 2, \dots, N \end{aligned}$$

In which:

$p_j(x_j)$: profit obtained by producing x_j units of product j .

$c_j(x_j)$: units of the resource consumed for producing x_j units of product j .

Ω_j : the set of possible production levels for product j .

The above problem can be solved by dynamic programming



Introduction

Bellman's Principle of Optimality

The embedded problem in backward recursive form

Define:

- (n, y) : state - y units of resource are allocated to produce products from n through N .
- $f_n(y)$: maximum total profit obtained from products n through N , when y units of resource are allocated to them.
- $f_1(K)$: the optimal value to be determined.



Introduction

Bellman's Principle of Optimality

Notes:

1. $f_n(y)$ can be expressed as:

$$\begin{aligned} \text{Max} \quad & p_n(x_n) + \cdots + p_N(x_N) \\ \text{s.t.} \quad & c_n(x_n) + \cdots + c_N(x_N) \leq y \\ & x_j \in \Omega_j \quad \forall j = n, \dots, N \end{aligned}$$

The problem $f_1(K)$ is embedded in the above problems: $f_n(y)$ for $n = 1, 2, \dots, N$ and $y = 0, 1, 2, \dots, K$



Introduction

Bellman's Principle of Optimality

2. Boundary conditions:

$$\begin{aligned} \text{Max } f_N(y) &= \text{Max } p_N(x_N) \\ \text{s.t. } c_N(x_N) &\leq y \\ x_N &\in \Omega_N \end{aligned}$$



Introduction

Bellman's Principle of Optimality

Backward recursive equation:

$$f_n(y) = \max_{\substack{c_n(x_n) \leq y \\ x_n \in \Omega_n}} \{p_n(x_n) + f_{n+1}(y - c_n(x_n))\}$$

In this case, we have:

$$s_n = (n, y); D(s_n) = \{x \in \Omega_n \mid c_n(x) \leq y\}; \text{ and} \\ s_{n+1} = t_n(s_n, x) = (n + 1, y - c_n(x))$$



Introduction

Bellman's Principle of Optimality

The embedded problem in forward recursive form

Define:

- (n, y) : state - y units of resource are allocated to produce products from 1 to n .
- $f_n(y)$: maximum total profit obtained from products 1 through n , when y units of resource are allocated to them.
- $f_N(K)$: the optimal value to be determined.



Introduction

Bellman's Principle of Optimality

Forward recursive equation:

$$f_n(y) = \max_{\substack{c_n(x_n) \leq y \\ x_n \in \Omega_n}} \{p_n(x_n) + f_{n-1}(y - c_n(x_n))\}$$

In this case, we still have:

$$s_n = (n, y); D(s_n) = \{x \in \Omega_n \mid c_n(x) \leq y\}; \text{ and}$$
$$s_{n-1} = t_n(s_n, x) = (n - 1, y - c_n(x)).$$

