

Dynamic Programming

Recursive Equations



Introduction and Objectives

Introduction

- Recursive equations are used to solve a problem in sequence
- These equations are fundamental to the dynamic programming

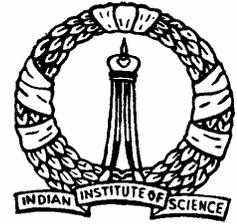
Objectives

- To formulate recursive equations for a multistage decision process
 - In a backward manner and
 - In a forward manner



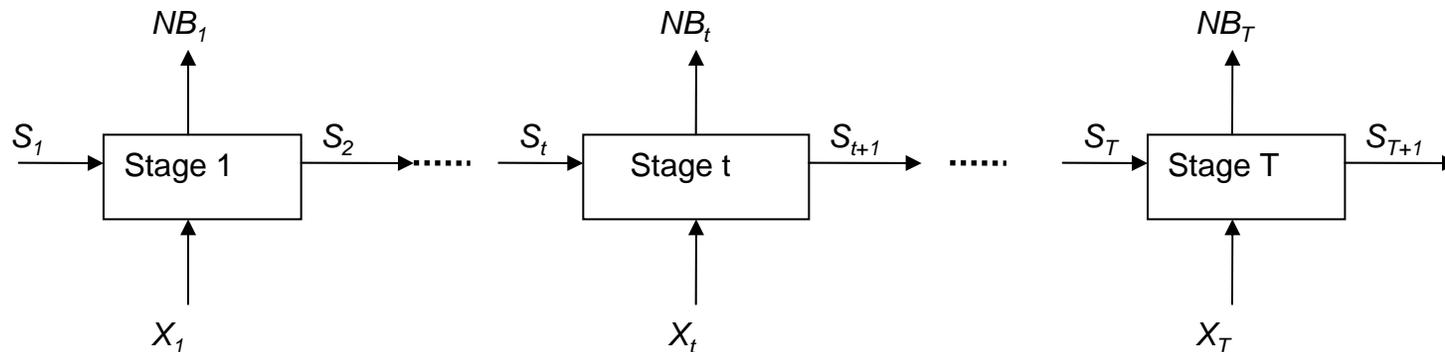
Recursive Equations

- Recursive equations are used to structure a multistage decision problem as a sequential process
- Each recursive equation represents a stage at which a decision is required
- A series of equations are successively solved, each equation depending on the output values of the previous equations
- A multistage problem is solved by breaking into a number of single stage problems through recursion
- Approached can be done in a backward manner or in a forward manner



Backward Recursion

- A problem is solved by writing equations first for the final stage and then proceeding backwards to the first stage
- Consider a serial multistage problem



- Let the objective function for this problem is

$$f = \sum_{t=1}^T NB_t = \sum_{t=1}^T h_t(X_t, S_t)$$
$$= h_1(X_1, S_1) + h_2(X_2, S_2) + \dots + h_t(X_t, S_t) + \dots + h_{T-1}(X_{T-1}, S_{T-1}) + h_T(X_T, S_T) \quad \dots(1)$$



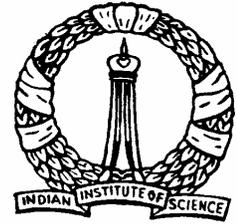
Backward Recursion ...contd.

- The relation between the stage variables and decision variables are
$$S_{t+1} = g(X_t, S_t), \quad t = 1, 2, \dots, T.$$
- Consider the final stage as the first sub-problem. The input variable to this stage is S_T .
- Principle of optimality: X_T should be selected such that $h_T(X_T, S_T)$ is optimum for the input S_T
- The objective function f_T^* for this stage is

$$f_T^*(S_T) = \text{opt}[h_T(X_T, S_T)]$$

- Next, group the last two stages together as the second sub-problem. The objective function is

$$f_{T-1}^*(S_{T-1}) = \text{opt}_{X_{T-1}, X_T} [h_{T-1}(X_{T-1}, S_{T-1}) + h_T(X_T, S_T)]$$



Backward Recursion ...contd.

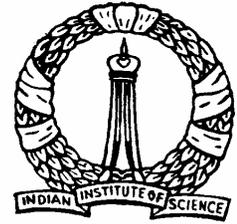
- By using the stage transformation equation, $f_{T-1}^*(S_{T-1})$ can be rewritten as
$$f_{T-1}^*(S_{T-1}) = \underset{X_{T-1}}{\text{opt}} [h_{T-1}(X_{T-1}, S_{T-1}) + f_T^*(g_{T-1}(X_{T-1}, S_{T-1}))]$$

- Thus, a multivariate problem is divided into two single variable problems as shown
- In general, the $i+1^{\text{th}}$ sub-problem can be expressed as

$$f_{T-i}^*(S_{T-i}) = \underset{X_{T-i}, \dots, X_{T-1}, X_T}{\text{opt}} [h_{T-i}(X_{T-i}, S_{T-i}) + \dots + h_{T-1}(X_{T-1}, S_{T-1}) + h_T(X_T, S_T)]$$

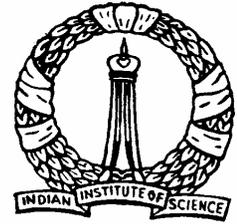
- Converting this to a single variable problem

$$f_{T-i}^*(S_{T-i}) = \underset{X_{T-i}}{\text{opt}} [h_{T-i}(X_{T-i}, S_{T-i}) + f_{T-(i-1)}^*(g_{T-i}(X_{T-i}, S_{T-i}))]$$



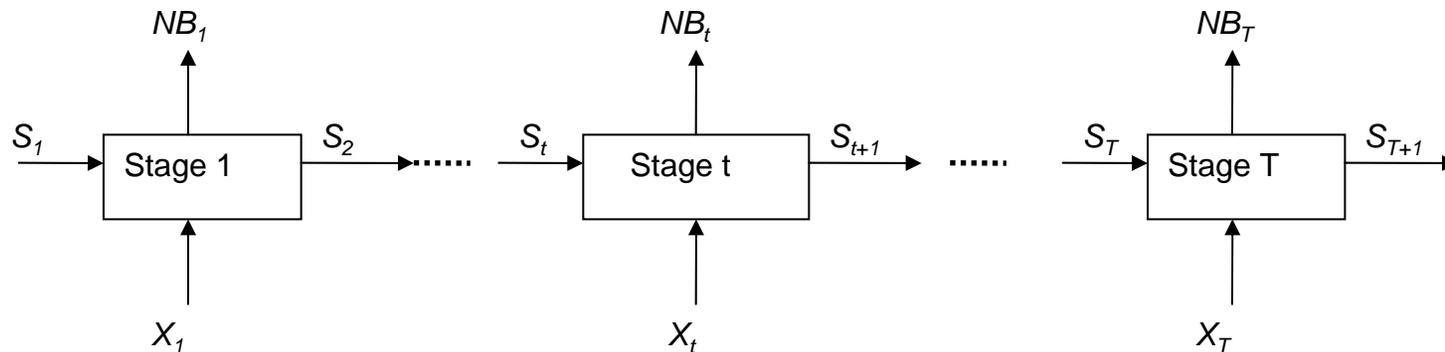
Backward Recursion ...contd.

- $f_{T-(i-1)}^*$ denotes the optimal value of the objective function for the last i stages
- Principle of optimality for backward recursion can be stated as,
 - *No matter in what state of stage one may be, in order for a policy to be optimal, one must proceed from that state and stage in an optimal manner sing the stage transformation equation*



Forward Recursion

- The problem is solved by starting from the stage 1 and proceeding towards the last stage
- Consider a serial multistage problem



- Let the objective function for this problem is

$$f = \sum_{t=1}^T NB_t = \sum_{t=1}^T h_t(X_t, S_t)$$
$$= h_1(X_1, S_1) + h_2(X_2, S_2) + \dots + h_t(X_t, S_t) + \dots + h_{T-1}(X_{T-1}, S_{T-1}) + h_T(X_T, S_T) \quad \dots(1)$$



Forward Recursion ...contd.

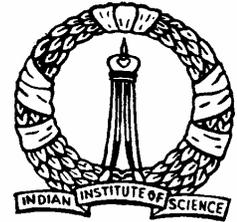
- The relation between the stage variables and decision variables are

$$S_t = g'(X_{t+1}, S_{t+1}) \quad t = 1, 2, \dots, T$$

where S_t is the input available to the stages 1 to t

- Consider the stage 1 as the first sub-problem. The input variable to this stage is S_1
- Principle of optimality: X_1 should be selected such that $h_1(X_1, S_1)$ is optimum for the input S_1
- The objective function f_1^* for this stage is

$$f_1^*(S_1) = \underset{X_1}{opt}[h_1(X_1, S_1)]$$



Backward Recursion ...contd.

- Group the first and second stages together as the second sub-problem. The objective function is

$$f_2^*(S_2) = \underset{X_2, X_1}{opt} [h_2(X_2, S_2) + h_1(X_1, S_1)]$$

- By using the stage transformation equation, $f_2^*(S_2)$ can be rewritten as

$$f_2^*(S_2) = \underset{X_2}{opt} [h_2(X_2, S_2) + f_1^*(g_2'(X_2, S_2))]$$

- In general, the i^{th} sub-problem can be expressed as

$$f_i^*(S_i) = \underset{X_1, X_2, \dots, X_i}{opt} [h_i(X_i, S_i) + \dots + h_2(X_2, S_2) + h_1(X_1, S_1)]$$

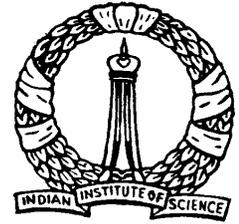


Backward Recursion ...contd.

- Converting this to a single variable problem

$$f_i^*(S_i) = \underset{X_i}{\text{opt}} [h_i(X_i, S_i) + f_{(i-1)}^*(g_i'(X_i, S_i))]$$

- f_i^* denotes the optimal value of the objective function for the last i stages
- Principle of optimality for forward recursion can be stated as,
 - *No matter in what state of stage one may be, in order for a policy to be optimal, one had to get to that state and stage in an optimal manner*



Thank You