

Dynamic Programming

Other Topics

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Objectives

- To explain the difference between discrete and continuous dynamic programming
- To discuss about multiple state variables
- To discuss the curse of dimensionality in dynamic programming



Discrete versus Continuous Dynamic Programming

- Discrete dynamic programming problems: Number of stages is finite
- When the number of stages tends to infinity then it is called a continuous dynamic programming problem
- Also called as infinite-stage problem
- Continuous dynamic problems are used to solve continuous decision problem
- The classical method of solving continuous decision problems is by the calculus of variations



Discrete versus Continuous Dynamic Programmingcontd.

- However, the analytical solutions, using calculus of variations is applicable only for very simple problems
- The infinite-stage dynamic programming approach provides a very efficient numerical approximation procedure for solving continuous decision problems
- For discrete dynamic programming model, the objective function value is the sum of individual stage outputs
- For a continuous dynamic programming model, summation is replaced by integrals of the returns from individual stages
- Such models are useful when infinite number of decisions have to be made in finite time interval



Multiple State Problems

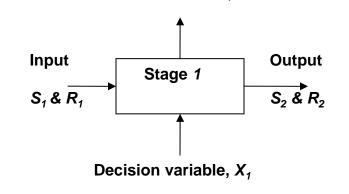
- Problems in which there are more than one state variable
- For example, consider a water allocation problem to *n* irrigated crops
- Let S_i be the units of water available to the remaining *n-i* crops
- Considering only the allocation of water, the problem can be solved as a single state problem, with S_i as the state variable
- Now, assume that L units of land are available for all these n crops
- Allocation of land also to be done to each crop after considering the units of water required for each unit of irrigated land containing each crop

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Multiple State Problems ...contd.

- Let *R_i* be the amount of land available for *n-i* crops
- An additional state variable R_i is included while sub-optimizing different stages
- Thus, in this problem two allocations need to be made: water and land.
- A single stage problem consisting of two state variables, $S_1 \& R_1$ is shown below Net Benefit, NB₁





Multiple State Problemscontd.

In general, for a multistage decision problem of *T* stages, containing two state variables S_t and R_t, the objective function can be written as

$$f = \sum_{t=1}^{T} NB_{t} = \sum_{t=1}^{T} h(X_{t}, S_{t}, R_{t})$$

where the transformation equations are given as

 $S_{t+1} = g(X_t, S_t)$ for t = 1, 2, ..., T & $R_{t+1} = g'(X_t, R_t)$ for t = 1, 2, ..., T



Curse of Dimensionality

- Limitation of dynamic programming: Dimensionality restriction
- The number of calculations needed increases rapidly as the number of variables and stages increase
- Increases the computational effort
- Increase in the number of stage variables causes an increase in the number of combinations of discrete states to be examined at each stage
- For a problem consisting of 100 state variables and each variable having 100 discrete values, the sub-optimization table will contain 100¹⁰⁰ entries



Curse of Dimensionalitycontd.

- The computation of one table may take 100⁹⁶ seconds (about 100⁹² years) even on a high speed computer
- Like this 100 tables have to be prepared for which computation is almost impossible
- This phenomenon as termed by Bellman, is known as *"curse of dimensionality" or "Problem of dimensionality"* of multiple state variable dynamic programming problems



Thank You

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