### 6.231 DYNAMIC PROGRAMMING

## **LECTURE 15**

# LECTURE OUTLINE

- Rollout algorithms
- Cost improvement property
- Discrete deterministic problems
- Sequential consistency and greedy algorithms
- Sequential improvement

# **ROLLOUT ALGORITHMS**

• One-step lookahead policy: At each k and state  $x_k$ , use the control  $\overline{\mu}_k(x_k)$  that

 $\min_{u_k \in U_k(x_k)} E\{g_k(x_k, u_k, w_k) + \tilde{J}_{k+1}(f_k(x_k, u_k, w_k))\}\},\$ 

where

$$- \tilde{J}_N = g_N.$$

 $- \tilde{J}_{k+1}$ : approximation to true cost-to-go  $J_{k+1}$ 

• Rollout algorithm: When  $\tilde{J}_k$  is the cost-to-go of some heuristic policy (called the *base policy*)

• Cost improvement property (to be shown): The rollout algorithm achieves no worse (and usually much better) cost than the base heuristic starting from the same state.

• Main difficulty: Calculating  $\tilde{J}_k(x_k)$  may be computationally intensive if the cost-to-go of the base policy cannot be analytically calculated.

- May involve Monte Carlo simulation if the problem is stochastic.
- Things improve in the deterministic case.

# EXAMPLE: THE QUIZ PROBLEM

• A person is given N questions; answering correctly question i has probability  $p_i$ , with reward  $v_i$ .

• Quiz terminates at the first incorrect answer.

• Problem: Choose the ordering of questions so as to maximize the total expected reward.

• Assuming no other constraints, it is optimal to use the *index policy*: Questions should be answered in decreasing order of the "index of preference"  $p_i v_i/(1-p_i)$ .

• With minor changes in the problem, the index policy need not be optimal. Examples:

- A limit (< N) on the maximum number of questions that can be answered.
- Time windows, sequence-dependent rewards, precedence constraints.

• Rollout with the index policy as base policy: Convenient because at a given state (subset of questions already answered), the index policy and its expected reward can be easily calculated.

#### **COST IMPROVEMENT PROPERTY**

• Let

 $\overline{J}_k(x_k)$ : Cost-to-go of the rollout policy

 $H_k(x_k)$ : Cost-to-go of the base policy

- We claim that  $\overline{J}_k(x_k) \leq H_k(x_k)$  for all  $x_k$  and k
- Proof by induction: We have  $\overline{J}_N(x_N) = H_N(x_N)$  for all  $x_N$ . Assume that

$$\overline{J}_{k+1}(x_{k+1}) \le H_{k+1}(x_{k+1}), \quad \forall x_{k+1}.$$

Then, for all  $x_k$ 

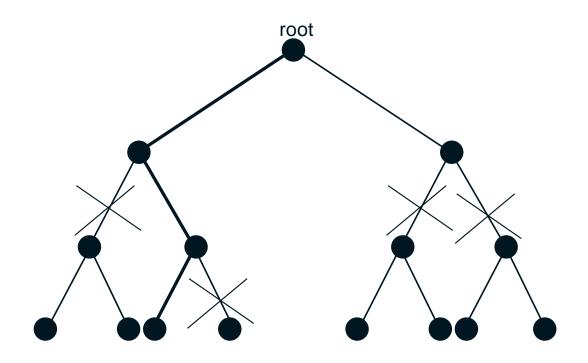
$$\overline{J}_{k}(x_{k}) = E\left\{g_{k}\left(x_{k}, \overline{\mu}_{k}(x_{k}), w_{k}\right) + \overline{J}_{k+1}\left(f_{k}\left(x_{k}, \overline{\mu}_{k}(x_{k}), w_{k}\right)\right)\right\}$$

$$\leq E\left\{g_{k}\left(x_{k}, \overline{\mu}_{k}(x_{k}), w_{k}\right) + H_{k+1}\left(f_{k}\left(x_{k}, \overline{\mu}_{k}(x_{k}), w_{k}\right)\right)\right\}$$

$$\leq E\left\{g_{k}\left(x_{k}, \mu_{k}(x_{k}), w_{k}\right) + H_{k+1}\left(f_{k}\left(x_{k}, \mu_{k}(x_{k}), w_{k}\right)\right)\right\}$$

$$= H_{k}(x_{k})$$

#### **EXAMPLE: THE BREAKTHROUGH PROBLEM**



- Given a binary tree with N stages.
- Each arc is either free or is blocked (crossed out in the figure).

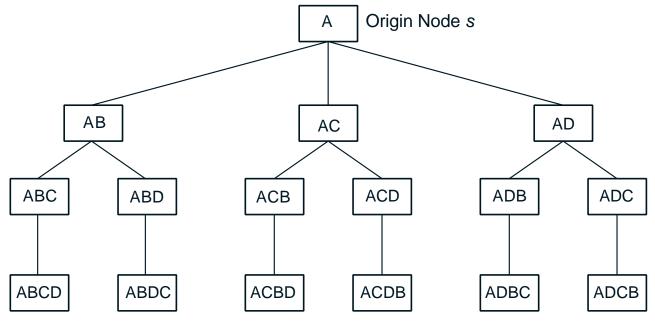
• Problem: Find a free path from the root to the leaves (such as the one shown with thick lines).

• Base heuristic (greedy): Follow the right branch if free; else follow the left branch if free.

• For large N and given prob. of free branch: the rollout algorithm requires O(N) times more computation, but has O(N) times larger prob. of finding a free path than the greedy algorithm.

# DISCRETE DETERMINISTIC PROBLEMS

- Any discrete optimization problem (with finite number of choices/feasible solutions) can be represented as a sequential decision process by using a tree.
- The leaves of the tree correspond to the feasible solutions.
- The problem can be solved by DP, starting from the leaves and going back towards the root.
- Example: Traveling salesman problem. Find a minimum cost tour that goes exactly once through each of *N* cities.



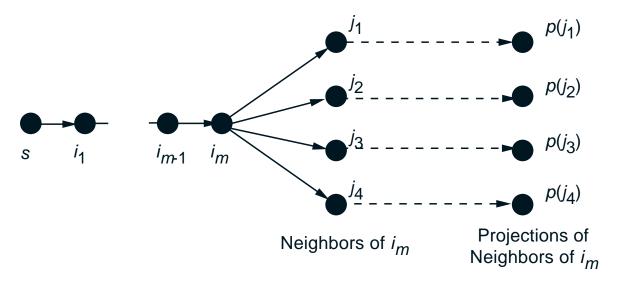
Traveling salesman problem with four cities A, B, C, D

## A CLASS OF GENERAL DISCRETE PROBLEMS

- Generic problem:
  - Given a graph with directed arcs
  - A special node s called the *origin*
  - A set of terminal nodes, called *destinations*, and a cost g(i) for each destination *i*.
  - Find min cost path starting at the origin, ending at one of the destination nodes.

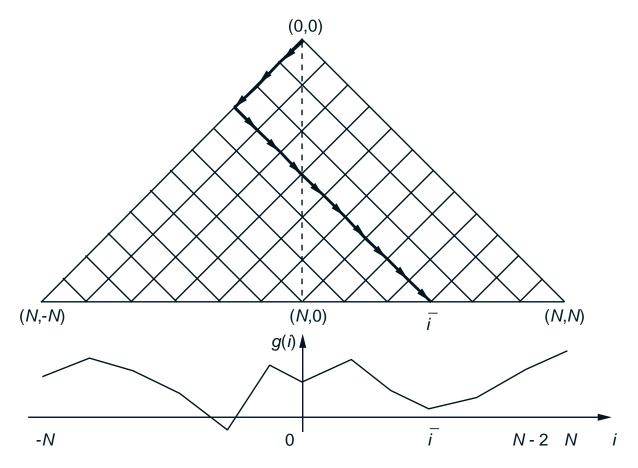
• Base heuristic: For any nondestination node i, constructs a path  $(i, i_1, \ldots, i_m, \overline{i})$  starting at i and ending at one of the destination nodes  $\overline{i}$ . We call  $\overline{i}$  the *projection* of i, and we denote  $H(i) = g(\overline{i})$ .

• Rollout algorithm: Start at the origin; choose the successor node with least cost projection



### **EXAMPLE: ONE-DIMENSIONAL WALK**

• A person takes either a unit step to the left or a unit step to the right. Minimize the cost g(i) of the point *i* where he will end up after *N* steps.



• Base heuristic: Always go to the right. Rollout finds the rightmost *local minimum*.

• Base heuristic: Compare always go to the right and always go the left. Choose the best of the two. Rollout finds a *global minimum*.

# SEQUENTIAL CONSISTENCY

• The base heuristic is sequentially consistent if for every node i, whenever it generates the path  $(i, i_1, \ldots, i_m, \overline{i})$  starting at i, it also generates the path  $(i_1, \ldots, i_m, \overline{i})$  starting at the node  $i_1$  (i.e., all nodes of its path have the same projection).

• Prime example of a sequentially consistent heuristic is a greedy algorithm. It uses an estimate F(i)of the optimal cost starting from i.

• At the typical step, given a path  $(i, i_1, \ldots, i_m)$ , where  $i_m$  is not a destination, the algorithm adds to the path a node  $i_{m+1}$  such that

$$i_{m+1} = \arg\min_{j \in N(i_m)} F(j)$$

• If the base heuristic is sequentially consistent, the cost of the rollout algorithm is no more than the cost of the base heuristic. In particular, if  $(s, i_1, \ldots, i_{\bar{m}})$  is the rollout path, we have

$$H(s) \ge H(i_1) \ge \dots \ge H(i_{\bar{m}-1}) \ge H(i_{\bar{m}})$$

where H(i) = cost of the heuristic starting from *i*.

### SEQUENTIAL IMPROVEMENT

• We say that the base heuristic is *sequentially improving* if for every non-destination node *i*, we have

$$H(i) \ge \min_{j \text{ is neighbor of } i} H(j)$$

• If the base heuristic is sequentially improving, the cost of the rollout algorithm is no more than the cost of the base heuristic, starting from any node.

- Fortified rollout algorithm:
  - Simple variant of the rollout algorithm, where we keep the best path found so far through the application of the base heuristic.
  - If the rollout path deviates from the best path found, then follow the best path.
  - Can be shown to be a rollout algorithm with sequentially improving base heuristic for a slightly modified variant of the original problem.
  - Has the cost improvement property.