

# Day 17 Notes

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## 1 Agenda

- Quiz
- Review of Dynamic Programming

## 2 Dynamic Programming

- Problem: order of matrix multiplication
- Key concepts
  - All solutions must have a final multiplication
  - Final multiplication cost is LHS+RHS+combining
  - For optimal soln, LHS, RHS must each be optimal
  - This is true recursively
- Notation
  - $m[i, j]$  - optimal sub-product cost  $\prod_{k=i}^j A_k$
  - $m[i, j] = 0$  if  $i == j$ , else  $\operatorname{argmin}_k (m[i, k] + m[k + 1, j] + p_{i-1}p_kp_j)$

*# Recursive Matrix Chain*

*# Computes  $m[i, j]$*

```
def RMC(p, i, j):
    if i == j:
        return 0
    else:
        m[i, j] = MAXINT
        for k in range(i, j-1):
            q = RMC(p, i, k) + RMC(p, k+1, j) + p[i-1]*p[k]*p[j]
            if q < m[i, j]:
                m[i, j] = q
    return m[i, j]
```

- Quiz question: modify this algorithm to keep track of best k at each step!
- Problem: Same sub-problems are solved over-and-over
- Just like recursive Fibonacci

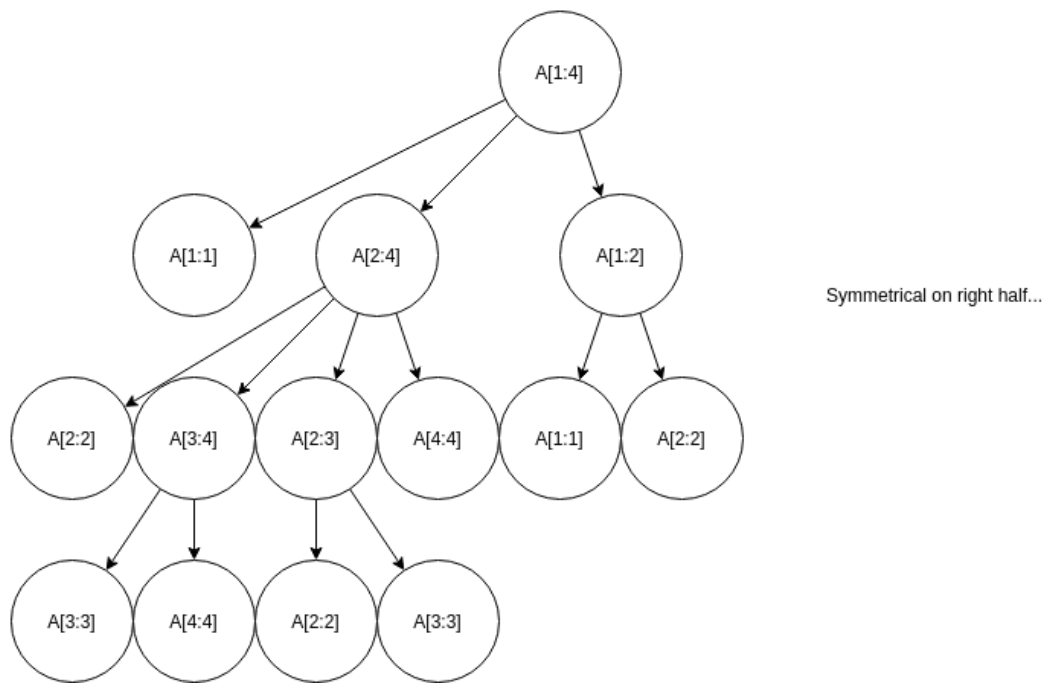


Figure 1: Example RMC solution tree for  $A_{1..4}$

- Solution: make  $m[]$  global, and check if soln to subproblem has already been found. Use answer if it has.
- **memoization**: This process of saving already-computed solutions

## Complexity

- $m$  matrix has  $n^2$  entries
- Each entry computed exactly once
- Work to compute each entry (or at least one half (diagonal symmetry)):
  - Each func call takes  $O(n)$
  - Total function takes  $O(n * n^2) = O(n^3)$

## Optimization

- Idea: instead of discovering tree recursively, why not just go through matrix in order and solve each subproblem?
- Organize as triangle (half matrix) w/  $i, j$  along edges
- Build values from trivial diagonal to corner
- **dynamic programming** technically refers to this triangle approach, not the memoization

## Dynamic TSP Example

- Is set of consecutive indices in HC also a HC? NO!
- Instead, given a subset of nodes,  $s$ , and some  $k \in s$ ,  $c(s, k) = \min$  cost to start at node 1, visit all nodes in  $S$ , and end at  $k$ .
- $c()$  can be expressed in terms of subproblems easily.
- Remove  $k^{th}$  node from the set and repeat with new  $k$
- Try all  $k$  and pick optimal
- Example
  - $c(\{2,4,6,7\}, 2)$  is  $\min(c(\{4,6,7\},4)+w[4,2], c(\{4,6,7\}, 6)+w[6,2], \dots)$
  - In general,  $c(S, k) = \min_{m \in S - \{k\}} c(s - \{k\}, m) + w[m, k]$