

Algorithms - CheatSheet

IN BA4 - Ola Nils Anders Svensson

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This is a cheat sheet for the Algorithms midterm exam. For suggestions, contact me on Telegram ([elazdi.al](#)) or via EPFL email (ali.elazdi@epfl.ch).

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Master Theorem If $T(n) = aT(\frac{n}{b}) + f(n)$, $a \geq 1$, $b > 1$, and $f(n)$ asymptotically positive Case 1: If $f(n) = O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$. Case 2: If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \log n)$. Case 3: If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$, and if $a f(\frac{n}{b}) \leq c f(n)$ for some $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$. Common case - If $f(n) = \Theta(n^d)$ for some exponent d: - If $\frac{d}{\log b} < 1$ (or $d < \log_b a$), then $T(n) = \Theta(n^d)$. - If $\frac{d}{\log b} = 1$ (or $d = \log_b a$), then $T(n) = \Theta(n^d \log n)$. - If $\frac{d}{\log b} > 1$ (or $d > \log_b a$), then $T(n) = \Theta(n^{\log_b a})$.	
Queue Operations Queue-Empty(Q): Time: $O(1)$, Auxiliary Space: $O(1)$ 1. Returns TRUE if the queue is empty ($Q.head = Q.tail$). 2. Returns FALSE otherwise. Enqueue(Q, x): Time: $O(1)$, Auxiliary Space: $O(1)$ 1. Adds element x to the rear of queue Q . 2. $Q.Qtail \leftarrow x$ 3. $Q.Qtail \leftarrow Q.Qtail + 1$ (or wrap around if using circular array) Dequeue(Q): Time: $O(1)$, Auxiliary Space: $O(1)$ 1. If Queue-Empty(Q), return error "underflow". 2. Otherwise, remove and return the element at the front. 3. $x = Q.Qhead$ 4. $Q.Qhead = Q.Qhead + 1$ (or wrap around) 5. Return x Queue Implementation: 1. Q.head: Index of the front element 2. Q.tail: Index where next element will be inserted 3. In a circular array, indices wrap around 4. Leave one slot empty to distinguish full/empty states Overall Space Complexity: $O(n)$ for a queue of capacity n	
Merge Sort 1. Divide: Split the array evenly into two smaller subarrays, and continue dividing recursively. 2. Sort (Recursively): Apply merge sort recursively on each subarray until each has only one element (base case). MERGE-SORT(A, p, r) $n_1 \leftarrow q - p + 1$ $n_2 \leftarrow r - q$ let $L[1 \dots n_1 + 1]$ and $R[1 \dots n_2 + 1]$ be new arrays for $i \leftarrow 1$ to n_1 do $L[i] \leftarrow A[p + i - 1]$ for $j \leftarrow 1$ to n_2 do $R[j] \leftarrow A[q + j]$ $L[n_1 + 1] \leftarrow \infty$ $R[n_2 + 1] \leftarrow \infty$ $i \leftarrow 1$ $j \leftarrow 1$ for $k \leftarrow p$ to r do if $L[i] \leq R[j]$ $A[k] \leftarrow L[i]$ $i \leftarrow i + 1$ else $A[k] \leftarrow R[j]$ $j \leftarrow j + 1$	
Priority Queue Maintains a dynamic set of elements with associated priority values (keys). Maximum(S): Return element of S with highest priority (return $A[1]$, complexity $O(1)$) Insert(S, x): Insert element x into set S 1. Increment the heap size 2. Insert a new node in the last position in the heap, with key $-\infty$ 3. Increase the $-\infty$ value to key using Heap-Increase-Key Extract-Max(S): Remove and return element of S with highest priority 1. Make sure heap is not empty 2. Make a copy of the maximum element (the root) 3. Make the last node in the tree the new root 4. Re-heapify the heap, with one fewer node 5. Return the copy of the maximum element Increase-Key(S, x, k): Increase the value of element x 's key to the new value k 1. Make sure key $\geq A[i]$ 2. Update $A[i]$'s value to key 3. Traverse the tree upward comparing new key to the parent and swapping if necessary Time Complexity: Insert, Extract-Max, Increase-Key: $O(\log n)$ Maximum: $O(1)$ Space Complexity: $O(n)$	
Stack Operations Stack-Empty(S): Time: $O(1)$, Auxiliary Space: $O(1)$ 1. Returns TRUE if the stack is empty. 2. Returns FALSE otherwise. Push(S, x): Time: $O(1)$, Auxiliary Space: $O(1)$ 1. Adds element x to the top of stack S . 2. Increments the stack pointer. Pop(S): Time: $O(1)$, Auxiliary Space: $O(1)$ 1. If Stack-Empty(S), return error "underflow". 2. Otherwise, remove and return the top element. 3. Decrements the stack pointer. Stack Implementation: 1. Elements are stored in a simple array 2. S.top: Index of the topmost element 3. An empty stack has S.top = 0 or S.top = -1 (implementation dependent) Overall Space Complexity: $O(n)$ for a stack of size n	
Dynamic Programming Problem: Optimal solutions to overlapping subproblems Fibonacci Sequence: Top-Down (Memoization): 1. Create memo array $F[0 \dots n]$ initialized to NIL. 2. Base cases: $F[0] = 0$, $F[1] = 1$ 3. Recursive with memo: - Return $F[n]$ if already computed - Otherwise compute $F[n] = F[n-1] + F[n-2]$ - Store result in $F[n]$ and return MEMOIZED-FIB(n) Let $r = [0 \dots n]$ be a new array for $i = 0$ to n do $r[i] \leftarrow \infty$ return MEMOIZED-FIB-AUX(n, r) MEMOIZED-FIB-AUX(n, r) if $r[n] \geq 0$ return $r[n]$ if $n = 0$ or $n = 1$ ans $\leftarrow n$ else ans \leftarrow MEMOIZED-FIB-AUX($n-1$, r) + MEMOIZED-FIB-AUX($n-2$, r) $r[n] \leftarrow$ ans return $r[n]$ Bottom-Up (Tabulation): 1. Create array $F[0 \dots n]$ 2. Base cases: $F[0] = 0$, $F[1] = 1$ 3. For $i = 2$ to n : - Compute $F[i] = F[i-1] + F[i-2]$ 4. Return $F[n]$ EXTENDED-BOTTOM-UP-CUT-ROD(p, n) let $r[0 \dots n]$ and $s[0 \dots n]$ be new arrays for $i \leftarrow 0$ to n do $r[i] \leftarrow \infty$ for $j \leftarrow 1$ to i do $q \leftarrow \infty$ for $k \leftarrow 1$ to j do if $k < p[j] + r[j-k]$ $q \leftarrow p[j-k] + r[j-k]$ $r[i] \leftarrow q$ $s[i] \leftarrow j$ return r, s	
Cut-Rod Problem: Find optimal way to cut rod to maximize revenue Top-Down (Memoization): 1. Create memo array $r[0 \dots n]$ with $r[0] = 0$ 2. For uncalculated $r[j]$, compute: $r[j] = \max_{1 \leq i \leq j} (p[i] + r[j-i])$ 3. Return $r[n]$ MEMOIZED-CUT-ROD-AUX(p, n, r) if $r[n] \geq 0$ return $r[n]$ if $n = 0$ $q = 0$ else $q = \infty$ for $i = 1$ to n do $q = \max(q, p[i] + r[n-i])$ return q MEMOIZED-CUT-ROD(p, n) for $i = 0$ to n do $r[i] \leftarrow \infty$ return MEMOIZED-CUT-ROD-AUX(p, n, r) Bottom-Up (Tabulation): 1. Create array $r[0 \dots n]$ with $r[0] = 0$ 2. For $j = 1$ to n : - Compute $r[j] = \max_{1 \leq i \leq j} (p[i] + r[j-i])$ 3. Return $r[n]$ EXTENDED-BOTTOM-UP-CUT-ROD(p, n) let $r[0 \dots n]$ and $s[0 \dots n]$ be new arrays for $i \leftarrow 0$ to n do $r[i] \leftarrow \infty$ for $j \leftarrow 1$ to i do $q \leftarrow p[j] + r[i-j]$ $s[i] \leftarrow j$ return r, s	

Akra-Bazzi, Ali Najib Variation: For recurrence $T(n) = \alpha T(an) + \beta T(bn) + \Theta(n^d)$, where $\alpha, \beta > 0$, $a, b \in (0, 1)$, $d \geq 0$, and unique p with $\alpha a^p + \beta b^p = 1$. Then: $p > d \Rightarrow T(n) = \Theta(n^p)$ $p = d \Rightarrow T(n) = \Theta(n^d \log n)$ $p < d \Rightarrow T(n) = \Theta(n^d)$ If p is not easily found, compare $\alpha a^d + \beta b^d$ with 1: $\alpha a^d + \beta b^d < 1 \Rightarrow T(n) = \Theta(n^d)$ $\alpha a^d + \beta b^d = 1 \Rightarrow T(n) = \Theta(n^d \log n)$ $\alpha a^d + \beta b^d > 1 \Rightarrow T(n) = \Theta(n^p)$, with p determined by $\alpha a^p + \beta b^p = 1$.	
Insertion Sort. 1 - Select the key Begin with the second element (at index 1) as the key. 2 - Compare and Shift Compare the key with elements in the sorted section (to its left). 3 - Shift Elements If an element is greater than the key, shift that element one position to the right. 4 - Insert the Key Once an element less than or equal to the key is found (or you reach the start), insert the key immediately after that element. 5 - Repeat Move forward to the next element, treating it as the new key, and repeat until the array is sorted. Time Complexity: Worst-case $O(n^2)$, Best-case $O(n)$. Space Complexity: $O(1)$.	
Strassen's Matrix Multiplication Divide: Partition each of A, B, C into four $\frac{n}{2} \times \frac{n}{2}$ submatrices: $\begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \cdot \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$ Conquer: Compute 7 products (recursively on $\frac{n}{2} \times \frac{n}{2}$ matrices): $M_1 = (A_{11} + A_{22})(B_{11} + B_{22})$ $M_2 = (A_{21} + A_{22})B_{11}$ $M_3 = A_{11}(B_{12} - B_{22})$ $M_4 = A_{22}(B_{21} - B_{11})$ $M_5 = (A_{11} + A_{12})B_{22}$ $M_6 = (A_{21} - A_{11})(B_{11} + B_{12})$ $M_7 = (A_{12} - A_{22})(B_{21} + B_{22})$ Combine: Assemble the resulting submatrices to form C : $C_{11} = M_1 + M_4 - M_5 + M_7$ $C_{12} = M_2 + M_4$ $C_{21} = M_3 + M_5$ $C_{22} = M_1 + M_3 - M_2 + M_6$ Time Complexity: $O(n^{\log_2 7}) \approx O(n^{2.81})$ Space Complexity: $O(n^2)$	
Binary Search Trees (BST) BST-Search 1. Start at root 2. If NULL, return NULL 3. If key = root's key, return root 4. If key < root's key, search left 5. If key > root's key, search right TREE-SEARCH(x, k) if $x = \text{NIL}$ or $k = \text{key}[x]$ return x else if $k < x.\text{key}$ return TREE-SEARCH(x.left, k) else return TREE-SEARCH(x.right, k) Time: $O(\log n)$ avg, $O(h)$ worst Space: $O(n)$ for tree, $O(h)$ auxiliary BST-Minimum 1. Start at root 2. If NULL, return NULL 3. Follow left pointers until no left child 4. Return leftmost node TREE-MINIMUM(x) while $x.\text{left} \neq \text{NIL}$ do $x \leftarrow x.\text{left}$ return x Time: $O(h)$ Space: $O(n)$ for tree, $O(1)$ auxiliary BST-Maximum 1. Start at root 2. If NULL, return NULL 3. Follow right pointers until no right child 4. Return rightmost node TREE-MAXIMUM(x) while $x.\text{right} \neq \text{NIL}$ do $x \leftarrow x.\text{right}$ return x Time: $O(h)$ Space: $O(n)$ for tree, $O(1)$ auxiliary BST-Successor 1. If right subtree exists: Return minimum in right subtree 2. Otherwise: Find first ancestor where node is in left subtree TREE-SUCCESSOR(x) if $x.\text{right} \neq \text{NIL}$ return TREE-MINIMUM(x.right) y $\leftarrow x.p$ while $y \neq \text{NIL}$ and $x = y.\text{right}$ do $x \leftarrow y$ $y \leftarrow y.p$ return y Time: $O(h)$ Space: $O(n)$ for tree, $O(1)$ auxiliary BST-Insert 1. Create new node z with key 2. Start at root, track parent $y = \text{NIL}$ 3. Move down tree (left if key $<$ node's key, right otherwise) 4. Once NULL found, link z as child of y 5. If y is NIL, z becomes root 6. Otherwise, insert z as left or right child based on key comparison TREE-INSERT(T, z) y $\leftarrow \text{NIL}$ T.root $\leftarrow z$ while $x \neq \text{NIL}$ do $y \leftarrow x$ if $x.\text{key} < z.\text{key}$ if $y \neq \text{NIL}$ $y.\text{left} \leftarrow z$ else $T.root \leftarrow z$ else $y.\text{right} \leftarrow z$ return T Time: $O(h)$ Space: $O(n)$ for tree, $O(1)$ auxiliary BST-Delete 1. If z has no left: transplant right 2. If z has no right: transplant left 3. With both children: a. Find successor y b. Handle y 's children c. Replace z with y TREE-DELETE(T, z) if $z.\text{left} = \text{NIL}$ TRANSPLANT(T, z, z.right) // z has no left child else if $z.\text{right} = \text{NIL}$ TRANSPLANT(T, z, z.left) // z has just a left child else z has two children y \leftarrow TREE-MINIMUM(T.right) // y is z 's successor if $y \neq z$ TRANSPLANT(T, y, y.right) y.right $\leftarrow z$ TRANSPLANT(T, z, y) y.left $\leftarrow z$ y.right $\leftarrow y$ Time: $O(h)$ Space: $O(n)$ for tree, $O(1)$ auxiliary	
Properties: - Left subtree: all keys $<$ node's key - Right subtree: all keys $>$ node's key - Left and right subtrees are also BSTs - A Node has: key (value), left & right (child pointers), parent (optional) - Tree height h : length of longest path from root to leaf	

Big-O

If $\exists c > 0$ and $\exists n_0 > 0$, $0 \leq f(n) \leq c \cdot g(n) \forall n \geq n_0$, $f(n) = O(g(n))$.

Big-Omega

If $\exists c > 0$ and $\exists n_0 > 0$, $0 \leq c \cdot g(n) \leq f(n) \forall n \geq n_0$, $f(n) = \Omega(g(n))$.

Big-Theta

If $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$, $f(n) = \Theta(g(n))$.

Little-o

If $\forall c > 0 \exists n_0 > 0$, $0 \leq f(n) < c \cdot g(n) \forall n \geq n_0$, $f(n) = o(g(n))$.

Relations

$f(n) = o(g(n)) \implies f(n) = O(g(n))$

Comparison of Common Functions (Ascending Order)

$O(1) < O(\log n) < O(n) < O(n^2) < O(n^3) < O(n^k) < O(n^{\log n}) < O(n^n)$.

Heap

Root is $A[1]$ Left(i) = $2i$ Right(i) = $2i + 1$ Parent(i) = $\lfloor i/2 \rfloor$

Max-Heapify (heapify subtree rooted at i)

1. Starting at the root

2. Compare $A[i]$, $A[\text{Left}(i)]$, $A[\text{Right}(i)]$

3. If necessary, swap $A[i]$ with the largest of the two children

4. Max-Heapify the swapped child

5. Continue comparing and swapping down the heap until subtree rooted at i is max-heap

Time Complexity: $O(\log n)$

Space Complexity: $O(n)$ including heap array, $O(1)$ auxiliary

Max-Heap-Insert (insert new key into heap)

1. Increase heap size: $A.\text{heap-size} = A.\text{heap-size} + 1$

2. Set the last element to negative infinity: $A[A.\text{heap-size}] = -\infty$

3. Call Max-Heap-Increase-Key to update to the correct value

Time Complexity: $O(\log n)$

Space Complexity: $O(n)$ including heap array, $O(1)$ auxiliary

Max-Heap-Increase-Key (increase key at position i)

1. Ensure new key is larger than current: if key $< A[i]$ then error

2. Set $A[i] = \text{key}$

3. Compare with parent and swap if necessary:

while $i \neq 1$ and $A[\text{Parent}(i)] < A[i]$

4. Exchange $A[i]$ with $A[\text{Parent}(i)]$

5. Set $i = \text{Parent}(i)$ and continue upward

Time Complexity: $O(\log n)$

Space Complexity: $O(n)$ including heap array, $O(1)$ auxiliary

Build-Max-Heap (build a max-heap from an array)

1. Start from the last non-leaf node at index $\lfloor \frac{n}{2} \rfloor - 1$

2. Move upwards to the root (index 0) and:

a. Max-Heapify the current node

b. Ensure the subtree rooted here satisfies max-heap property

3. Repeat until the root node is processed

4. After completion, array A represents a valid max heap

Time Complexity: $O(n)$

Space Complexity: $O(n)$ including heap array, $O(1)$ auxiliary

Heap Sort

1. Build a Max Heap:

a. Convert the given array into a max heap

b. Start from the last non-leaf node and heapify upwards

c. Ensure each parent node is greater than its children

2. Extract Maximum Elements:

a. Swap the root (maximum value) with the last element

b. Reduce heap size by one to exclude the last element

c. Heapify the root to maintain max heap property

d. Repeat until heap size becomes 1

3. Final Sorted Array:

a. After extraction, the sorted array in ascending order is obtained

b. Maximum elements are placed at the end

Time Complexity: $O(n \log n)$

Space Complexity: $O(n)$ including array, $O(1)$ auxiliary

Maximum Subarray Problem

Problem: Find contiguous subarray with largest sum

Divide and Conquer Approach:

1 - Divide.

Split array at midpoint

$\text{mid} = \lfloor (\text{low} + \text{high}) / 2 \rfloor$

2 - Conquer.

Find maximum subarrays recursively

Left max: in $A[\text{low} \dots \text{mid}]$

Right max: in $A[\text{mid} + 1 \dots \text{high}]$

Crossing max: spans the midpoint

3 - Combine: Return the largest of the three max (left_max, right_max, crossing_max)

4 - Finding the Crossing Maximum:

a. Find maximum suffix in left half (from mid down to low)

b. Find maximum prefix in right half (from mid+1 up to high)

c. Crossing max = max suffix + max prefix

Crossing Subarray: The maximum subarray that crosses the midpoint combines the maximum suffix of the left half with the maximum prefix of the right half.

Time Complexity: $\Theta(n \log n)$ due to $T(n) = 2T(n/2) + \Theta(n)$

Space Complexity: $O(\log n)$ for recursion stack

Find-Max-Subarray(A , low, high)

if high == low

return (low, high, $A[\text{low}]$)

else

$\text{mid} = \lfloor (\text{low} + \text{high}) / 2 \rfloor$

$(\text{left-low, left-high, left-sum}) = \text{Find-Max-Subarray}(A, \text{low}, \text{mid})$

$(\text{right-low, right-high, right-sum}) = \text{Find-Max-Subarray}(A, \text{mid} + 1, \text{high})$

$(\text{cross-low, cross-high, cross-sum}) = \text{Find-Max-Crossing-Subarray}(A, \text{low}, \text{mid}, \text{high})$

if left-sum \geq right-sum and left-sum \geq cross-sum

return (left-low, left-high, left-sum)

else if right-sum \geq left-sum and right-sum \geq cross-sum

return (right-low, right-high, right-sum)

else

return (cross-low, cross-high, cross-sum)

Find-Max-Crossing-Subarray(A , low, mid, high)

$\text{left-sum} \leftarrow -\infty$, $\text{sum} \leftarrow 0$

for $i = \text{mid} + 1$ to low do

$\text{sum} \leftarrow \text{sum} + A[i]$

if $\text{sum} > \text{left-sum}$

$\text{left-sum} \leftarrow \text{sum}$, $\text{max-left} \leftarrow i$

$\text{right-sum} \leftarrow -\infty$, $\text{sum} \leftarrow 0$

for $j = \text{mid} + 1$ to high do

$\text{sum} \leftarrow \text{sum} + A[j]$

if $\text{sum} > \text{right-sum}$

$\text{right-sum} \leftarrow \text{sum}$, $\text{max-right} \leftarrow j$

return (max-left, max-right, left-sum + right-sum)

Data Structure Operations Summary

Priority Queue Operations

Operation

Description

Time

Space

Construction

Create priority queue from array of n elements

$O(n)$

$O(n)$

Maximum(S)

Returns element with highest priority

$O(1)$

$O(1)$

Extract-Max(S)

Removes and returns element with highest priority

$O(\log n)$

$O(1)$

Insert(S, x)

Inserts element x into set S

$O(\log n)$

$O(1)$

Increase-Key(S, x, k)

Increases priority of element x to k

$O(\log n)$

$O(1)$

Stack Operations

Operation

Description

Time

Space

Construction

Create stack from array of n elements

$O(n)$

$O(n)$

Stack-Empty(S)

Returns TRUE if stack is empty, FALSE otherwise

$O(1)$

$O(1)$

Push(S, x)

Adds element x to top of stack S

$O(1)$

$O(1)$

Pop(S)

Removes and returns top element from stack S

$O(1)$

$O(1)$

Heap Operations

Operation

Description

Time

Space

Construction

Create heap from array of n elements

$O(n)$

$O(n)$

Max-Heapify(A , i)

Maintains max-heap property at node i

$O(\log n)$

$O(1)$

Build-Max-Heap(A , n)

Converts array A of n elements into max heap

$O(n)$

$O(1)$

Heap-Sort(A , n)

Sorts array A of n elements using heap structure

$O(n \log n)$

$O(1)$

Max-Heap-Insert(A , k)

Inserts key k into heap A

$O(\log n)$

$O(1)$

Heap-Extract-Max(A)

Returns and removes largest element from heap A

$O(\log n)$

$O(1)$

Heap-Increase-Key(A , i , k)

Increases key at index i to new value k

$O(\log n)$

$O(1)$

BST Operations

Operation

Description

Time

Space

Construction

Create BST from array of n elements

Best: $O(n \log n)$, Worst: $O(n^2)$

$O(n)$

BST-Search(T , k)

Finds node with key k in tree T

$O(h)$

$O(h)$

BST-Minimum(T)

Returns node with smallest key in T

$O(h)$

$O(1)$

BST-Maximum(T)

Returns node with largest key in T

$O(h)$

$O(1)$

BST-Successor(x)

Returns node with smallest key greater than x 's key

$O(h)$

$O(1)$

BST-Insert(T , z)

Inserts node z into BST T

$O(h)$

$O(1)$

BST-Delete(T , z)

Removes node z from BST T

$O(h)$

$O(1)$

BST-Inorder(T)

Visits all nodes in sorted order

$O(n)$

$O(h)$

BST-Preorder(T)

Visits root before its children

$O(n)$

$O(h)$

BST-Postorder(T)

Visits children before root

$O(n)$

$O(h)$

Queue Operations

Operation

Description

Time

Space

Construction

Create queue from array of n elements

$O(n)$

$O(n)$

Queue-Empty(Q)

Returns TRUE if queue is empty, FALSE otherwise

$O(1)$

$O(1)$

Enqueue(Q , x)

Adds element x to rear of queue Q

$O(1)$

$O(1)$

Dequeue(Q)

Removes and returns front element from queue Q

$O(1)$

$O(1)$

Matrix Chain Multiplication:

Find optimal parenthesization to minimize multiplications

Bottom-Up (Tabulation):

1. Create table $m[1..n, 1..n]$ with $m[i, i] = 0$
- $m[i, j]$ stores minimal cost of multiplying matrices i through j
2. For $l = 2$ to n (chain length):
3. For $i = 1$ to $n - l + 1$:
 - Set $j = i + l - 1$
 - Compute $m[i, j] = \min_{i \leq k < j} \{m[i, k] + m[k + 1, j] + p_{i-1} p_k p_j\}$
 - Store k in $s[i, j]$ that achieved minimum cost
4. Return $m[1, n]$

```
MATRIX-CHAIN-ORDER( $p$ )
 $n \leftarrow p.\text{length} - 1$ 
let  $m[1..n, 1..n]$  and  $s[1..n, 1..n]$  be new tables
for  $i = 1$  to  $n$  do
     $m[i, i] \leftarrow 0$ 
for  $\ell = 2$  to  $n$  do //  $\ell$  is the chain length
    for  $i = 1$  to  $n - \ell + 1$  do
         $j \leftarrow i + \ell - 1$ 
         $m[i, j] \leftarrow \infty$ 
        for  $k = i$  to  $j - 1$  do
             $q \leftarrow m[i, k] + m[k + 1, j] + p_{i-1} p_k p_j$ 
            if  $q < m[i, j]$ 
                 $m[i, j] \leftarrow q$ 
                 $s[i, j] \leftarrow k$ 
return  $m, s$ 
```

Complexity:

- Time: $O(n^3)$
- Space: $O(n^2)$ (includes matrix dimensions and tables)
- Table $s[i, j]$ stores index of last matrix in first parenthesized group