

Divide and Conquer: Divide into disjoint subproblems, then combine

- Merge Sort: in $O(n \log n)$

- Count inversion: by augmenting Merge Sort

- Closest Pair: by merging center with delta and 11 closest y points

- Karatsuba: By replacing $x_1y_2 + x_2y_1$ with $(x_1 + x_2)(y_1 + y_2) - x_1y_1 - x_2y_2$; evaluate polynomial at different points

- Strassen: Karatsuba for matrix multiplication

Greedy: Local optimal yields global optimal without memory; Prove optimality by contradiction or induction

- Interval Scheduling: Earliest finish time (replace non EFT by EFT)

- Interval Partitioning: Earliest start time (add one if not capable, argue optimal by depth is precisely the solution)

- Minimizing Lateness: Earliest deadline (argue removing inversion)

- Huffman (Lossless compression): Build prefix free tree by merging two smallest freqs (argue combining 2 smallest freqs remains optimal reversely)

Dynamic Programming: Greedy with memory; Optimal substructure and overlapping subproblems

Let $\text{OPT}(i)$ be the optimal solution for the first i items

Let $S(i)$ denote solution for the first i items

Bellman Equation: $\text{OPT}(i) = \text{Base Case}$; Recursive Case

Solution: $S(i) = \text{emptyset}$ for Base Case; $S(i - 1)$ or augmenting depending on Bellman Equation

Top-down: When not all subproblems are needed

Bottom-up: When all subproblems are needed; Prevents recursive call overhead; free memory early

- Weighted Interval Scheduling: Sort by finish time;

- Knapsack: Any sort; 2D array with weight and items; $O(nW)$

- Single Source Shortest Path: 2D array with ending node and number of edges; $O(n^3)$

- Chain Matrix Product: 2D array with starting and ending matrix; $O(n^3)$

- Edit Distance: 2D array with position of matching; $O(nm)$ time and space

- Travelling Salesman: 2D array with remaining node set and starting node; $O(n^2 2^n)$ with

space $O(n 2^n)$

Network Flow: Max flow = Min cut; Augmenting path; Residual graph; Ford-Fulkerson
Source: s ; Sink: t ; Non-negative capacities c ; directed graph

s-t flow is $f : E \rightarrow \mathbb{R}_{\geq 0}$; $0 \leq f(e) \leq c(e)$; sum of incoming flow equals outgoing flow, $f^{in}(v) = f^{out}(v)$

s-t cut (A, B) is: $s \in A, t \in B$, A and B partition vertices

- Ford-Fulkerson: While residual has path. Let $\text{bottleneck}(P)$ be the minimum capacity of edges in path P of residual graph, Augment flow f by sending $\text{bottleneck}(P)$ flow along path P ; $O((m + n)C)$;

- Edmonds-Karp: Ford-Fulkerson with BFS to find shortest path; $O(nm^2)$

$v(f) := f^{in}(t) = f^{out}(s)$

$v(f) = f^{out}(A) - f^{in}(A)$ for all A (prove by summing each a in A , cancel in and out within A)

$\text{Cap}(A, B)$: Sum of capacities of edges from A to B ; $v(f) \leq \text{Cap}(A, B)$ for any st-cut (A, B) by above

- Max Flow Min Cut: $v(f) = \text{Cap}(A, B)$ for some st-cut (A, B) (Prove by letting A be reachable nodes from s and B be remaining for final G_f graph of FF, then outgoing edges are saturated (equal to capacity), and entering edges have zero flow otherwise in A , so $v(f) = f^{out}(A) = \text{Cap}(A, B)$)

Integrality Theorem: If all capacities are integers, then there exists an integral max flow, same for variants.

For application of network flow, first show 1-1 correspondence between feasible solution and flow by double implication, then

- Bipartite Matching: For bipartite graph, connect s to left, t to right, all edges have capacity 1; Max flow is max matching; $O(nm)$ (flow with value k corresponds to a matching with cardinality k)

- Hall's Marriage Theorem: Bipartite graph G has perfect matching iff $|N(S)| \geq |S|$ for each $S \subseteq U$ (edges with s or t have capacity 1, otherwise infinity, then cut must isolate s or t)

- Edge-Disjoint Paths (Find max num of edge-disjoint paths): (paths are flows by $f(e) = 1$ if edge e is in a path otherwise 0, then with capacities 1, exists unique flow; reversely

construct paths by extracting edges from flow);

- Menger's Theorem: The maximum number of edge-disjoint paths from s to t is equal to the minimum number of edges (resp. vertices) whose removal disconnects s and t

- Multiple sources and sinks: Use master source and sink to connect the sources and sinks (with capacities infinity)

- Circulation: Negative supply node v with value k sends $d(v) = k$ more flow than it receives; positive demand node v with value k receives $d(v) = k$ more flow than it sends; zero node is transshipment node; Connect supply nodes with s , demand nodes with t , then circulation exists iff max flow value equals the sum of supplies.

- Circulation with Lower Bounds: For lower bound k between u and v , add k to $d(u)$ and $-k$ to $d(v)$, subtract k from upper bound

- Survey Design: Connect s to each customer with bounds $[c_i, c'_i]$, connect each customer to each survey with $[0, 1]$, connect each survey to t with bounds $[p_j, p'_j]$, edge from s to t with $[0, \infty]$ (so that all nodes $d(v) = 0$)

- Image Segmentation: Find min cut to segment image into two parts

Master Theorem: (O, Theta, Omega)

➤ **Theorem:** Let $a \geq 1$ and $b > 1$ be constants, $f(n)$ be a function, and $T(n)$ be defined on non-negative integers by the recurrence $T(n) \leq a \cdot T\left(\frac{n}{b}\right) + f(n)$, where n/b can be $\lfloor \frac{n}{b} \rfloor$. Let $d = \log_b a$. Then:

- If $f(n) = O(n^{d-\epsilon})$ for some constant $\epsilon > 0$, then $T(n) = O(n^d)$.
- If $f(n) = O(n^d \log^k n)$ for some $k \geq 0$, then $T(n) = O(n^d \log^{k+1} n)$.
- If $f(n) = O(n^{d+\epsilon})$ for some constant $\epsilon > 0$, then $T(n) = O(f(n))$.

Residual Graph:

- Suppose the current flow is f
- Define the residual graph G_f to be the following
 - The vertices are the same
 - For each edge $e = (u, v)$ in G , G_f has at most two edges
 - A forward edge $e = (u, v)$ with capacity $c(e) - f(e)$
 - How much additional flow can we send on e
 - A backward edge $e^{rev} = (v, u)$ with capacity $f(e)$
 - How much we can reduce the flow on e
 - Only add edges of capacity > 0

