

Review session 2

Lecture 666

April 2, 2013

Dynamic Programming

- 1 Find a “smart” recursion for the problem in which the number of distinct subproblems is small; polynomial in the original problem size.
- 2 Eliminate recursion and find an iterative algorithm to compute the problems bottom up by storing the intermediate values in an appropriate data structure; need to find the right way or order the subproblem evaluation.
- 3 Estimate the number of subproblems, the time to evaluate each subproblem and the space needed to store the value. Evaluate the total running time.
- 4 Optimize the resulting algorithm further

Dynamic programming...

- 1 Longest increasing subsequence.
- 2 Computing the solution itself (not only its value).
- 3 Maximum Weight Independent Set in Trees.
- 4 Dynamic programs can be also solved as problems on DAGs.
- 5 Edit distance: $O(nm)$ [but linear space!].
- 6 Floyd-Warshall: $O(n^3)$.
- 7 Knapsack: $O(nW)$ (pseudo-polynomial).
- 8 TSP: $O(n^3 2^n)$ time and $O(n^2 2^n)$ space.

Greedy algorithms...

Greedy has its place, but be careful not to be too greedy!

- 1 Must prove correctness of greedy algorithms.
- 2 Interval scheduling (so many variants that do not work).
Proved correctness by showing that one can map the greedy solution to optimal.
- 3 Interval Partitioning/Coloring.
Proved the depth of instance was $\#$ colors used by greedy.
- 4 Scheduling to Minimize Lateness.

Minimum spanning tree

- 1 Algorithms can be interpreted as being greedy.
- 2 **Prim**: **T** maintained by algorithm will be a tree. Start with a node in **T**. In each iteration, pick edge with least attachment cost to **T**.
- 3 **Reverse delete**: Delete edges keeping connectivity. Deleting edges from most expensive to cheapest.
- 4 **Kruskal**: Add edges in increasing price. Add edge only if merges two trees in the current forest.
- 5 **Borůvka's**: Every vertex pick cheapest edge out of it. Collapse connected components of chosen edges. Repeat till have a single tree.

Why MST algorithms work?

Definition

An edge $e = (u, v)$ is a **safe** edge if there is some partition of V into S and $V \setminus S$ and e is the unique minimum cost edge crossing S (one end in S and the other in $V \setminus S$).

Definition

An edge $e = (u, v)$ is an **unsafe** edge if there is some cycle C such that e is the unique maximum cost edge in C .

Proposition

If edge costs are distinct then every edge is either safe or unsafe.

Lemma

If e is a safe edge then every minimum spanning tree contains e .

Why MST algorithms work?

Even more

Lemma

Let G be a connected graph with distinct edge costs, then the set of safe edges form a connected graph.

Corollary

Let G be a connected graph with distinct edge costs, then set of safe edges form the *unique* MST of G .

Lemma

If e is an unsafe edge then no MST of G contains e .

Data structures for MST

- 1 Heap.
- 2 Fibonacci heap.
- 3 Union-find - path compression and union by rank.
(Amazing running time - $O(\alpha(m, n))$ per operation,)

Randomized algorithms

- 1 Basic concepts in discrete probability:
Random variable, probability, expectation, linearity of expectation, independent events, conditional probability, indicator variables.
- 2 Types of randomized algorithms: Las Vegas and Monte Carlo.
- 3 Why randomization works - concentration of mass.
- 4 Proved:

Theorem

Let X_n be the number heads when flipping a coin independently n times. Then

$$\Pr\left[X_n \leq \frac{n}{4}\right] \leq 2 \cdot 0.68^{n/4} \text{ and } \Pr\left[X_n \geq \frac{3n}{4}\right] \leq 2 \cdot 0.68^{n/4}$$

Randomized algorithms

- 1 Proved **QuickSort** has $O(n \log n)$ expected running time.
- 2 Proved **QuickSort** has $O(n \log n)$ running time with high probability.
- 3 Proved **QuickSelect** has $O(n)$ expected running time.
- 4 Hashing.
 - 1 Why randomization is a must.
 - 2 **2-universal** hash functions families.
 - 3 Showed/proved a **2-universal** hash family.
Guess two random numbers α and β . Hash function is
$$h(x) = (\alpha x + \beta) \bmod p.$$

Network Flow

- 1 Definitions.
- 2 Edge flow \Leftrightarrow path flow.
- 3 Max-flow problem.
- 4 Cuts and minimum-cut.
- 5 flow \leq cut capacity.
- 6 Max-flow Min-cut Theorem.
- 7 Residual network.
- 8 Augmenting paths.
- 9 Ford-Fulkerson Algorithm.
- 10 Proved correctness of Ford-Fulkerson Algorithm if capacities are integral.

Network Flow II

- 1 Ford-Fulkerson running time is $O(mC)$.
- 2 Mentioned the strongly polynomial time algorithm by Edmonds-Karp.
- 3 Computing minimum cut from max-flow.
- 4 One can convert a flow to an acyclic flow.
- 5 A flow can be decomposed into paths from the source to the target + cycles.
- 6 Computing edge-disjoint paths using flow.
- 7 Computing vertex-disjoint paths using flow.
- 8 Menger's theorem ($\#$ edge to cut = $\#$ edge disjoint paths).
- 9 Multiple sinks/sources.
- 10 Matching in bipartite graph.
- 11 Perfect matching.

Network Flow III

- 1 Deciding if a specific team can win the Pennant using network flow.
- 2 Project scheduling.
- 3 Mentioned extensions to min-cost flow, and lower bounds on flow.
- 4 Circulations.
- 5 Survey design (using lower/upper bounds on flow).

Notes

Notes