

Stable Matching, Complexity, and **Representative Problems**

> Winter 2003 Paul Beame



Women get the raw deal in the G-S algorithm

- $S^*=\{(m,best(m): m\widehat{I}M\}$
- For each **w**, **worst**(**w**)= lowest rated man among all valid partners of **w**
- Claim: S*={(worst(w),w): w∈ W}
- - Suppose (m,w)∈ S*, m¹worst(w)=m'
 Consider stable matching S' s.t. (m',w)∈ S'
 - must exist since m'=worst(w) ■ In S', m is paired with w'¹w=best(m)
 - Therefore (m,w'),(m',w)Î S' but
 - m >_w m' and w >_m w' so (m,w) would prefer each other, contradicting stability of S'

Measuring efficiency: The RAM model

- RAM = Random Access Machine
- Time ≈ # of instructions executed in an ideal assembly language
 - each simple operation (+,*,-,=,if,call) takes one time step
 - each memory access takes one time step



Complexity analysis

- Problem size N
 - Worst-case complexity: max # steps algorithm takes on any input of size N
 - Best-case complexity: min # steps algorithm takes on any input of size N
 - Average-case complexity: avg # steps algorithm takes on inputs of size N



Stable Matching

- Problem size
 - N=2n² words
 - 2n people each with a preference list of length n
 - 2n²log n bits
 - specifying an ordering for each preference list takes nlog n bits
- Brute force algorithm
 - Try all n! possible matchings
- Gale-Shapley Algorithm
 - n² iterations, each costing constant time
 - For each man an array listing the women in preference
 - For each woman an array listing the prefences indexed

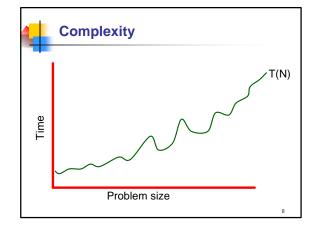
Complexity

- The complexity of an algorithm associates a number **T(N)**, the best/worst/average-case time the algorithm takes, with each problem size N.
- Mathematically,
 - T is a function that maps positive integers giving problem size to positive real numbers giving number of steps.



Efficient = Polynomial Time

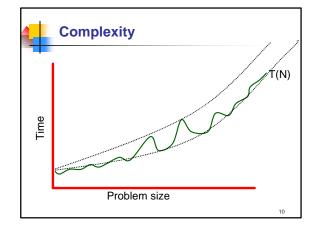
- Polynomial time
 - Running time T(N) £ cNk+d for some c,d,k>0
- Why polynomial time?
 - If problem size grows by at most a constant factor then so does the running time
 - $T(2N) £ c(2N)^k+d £ 2^k(cN^k+d)$
 - Polynomial-time is exactly the set of running times that have this property
 - Typical running times are small degree polynomials, mostly less than N3, at worst N6, not



O-notation etc

- Given two positive functions f and g
- f(N) is O(g(N)) iff there is a constant c>0 so that f(N) is eventually always £ c g(N)
- f(N) is o(g(N)) iff the ratio f(N)/g(N) goes to 0 as N gets large
- f(N) is W(g(N)) iff there is a constant e>0 so that f(N) is 3 e g(N) for infinitely many values of N
- f(N) is Q(g(N)) iff f(N) is O(g(N)) and f(N) is W(g(N))

Note: The definition of \mathbf{W} is the same as " $\mathbf{f}(\mathbf{N})$ is **not** $\mathbf{o}(\mathbf{g}(\mathbf{N}))$ "





5 Representative Problems

- Interval Scheduling
 - Single resource
 - Reservation requests
 - Of form "Can I reserve it from start time s to finish time f?"
 - s < f</p>
 - Find: maximum number of requests that can be scheduled so that no two reservations have the resource at the same time

Interval scheduling

- Formally
 - Requests 1,2,...,n
 - request i has start time s_i and finish time f_i > s_i
 - Requests i and j are compatible iff either
 - request i is for a time entirely before request j • f, £ s,
 - or, request j is for a time entirely before
 - request i
 - f_j£s_j
 Set A of requests is compatible iff every pair of requests i,j∈ A, i¹j is compatible
 Goal: Find maximum size subset A of compatible



Interval Scheduling

- We shall see that an optimal solution can be found using a "greedy algorithm"
 - Myopic kind of algorithm that seems to have no look-ahead
 - These algorithms only work when the problem has a special kind of structure
 - When they do work they are typically very efficient

13



Weighted Interval Scheduling

- Same problem as interval scheduling except that each request i also has an associated value or weight w;
 - w_i might be
 - amount of money we get from renting out the resource for that time period
 - amount of time the resource is being used
- Goal: Find compatible subset A of requests with maximum total weight

14



Weighted Interval Scheduling

- Ordinary interval scheduling is a special case of this problem
 - Take all w_i =1
- Problem is quite different though
 - E.g. one weight might dwarf all others
- "Greedy algorithms" don't work
- Solution: "Dynamic Programming"
 - builds up optimal solutions from smaller problems using a compact table to store them

15



Bipartite Matching

- A graph G=(V,E) is bipartite iff
 - V consists of two disjoint pieces X and Y such that every edge e in E is of the form (x,y) where xî X and yî Y
 - Similar to stable matching situation but in that case all possible edges were present
 - MÍE is a matching in G iff no two edges in M share a vertex
 - Goal: Find a matching M in G of maximum possible size

6



Bipartite Matching

- Models assignment problems
 - X represents jobs, Y represents machines
 - X represents professors, Y represents courses
- If |X|=|Y|=n
 - G has perfect matching iff maximum matching has size n
- Solution: polynomial-time algorithm using "augmentation" technique
 - also used for solving more general class of network flow problems

17



Independent Set

- Given a graph G=(V,E)
 - A set IÍ V is independent iff no two nodes in I are joined by an edge
- Goal: Find an independent subset I in G of maximum possible size
- Models conflicts and mutual exclusion

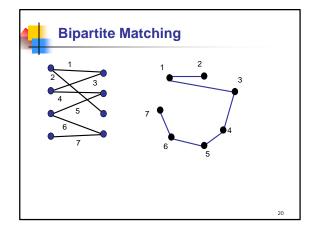
18



Independent Set

- Generalizes
 - Interval Scheduling
 - Vertices in the graph are the requests
 - Vertices are joined by an edge if they are **not** compatible
 - Bipartite Matching
 - Given bipartite graph G=(V,E) create new graph G'=(V',E') where
 - V'=E
 - Two elements of V' (which are edges in G) are joined if they share an endpoint in G

19





Independent Set

- No polynomial-time algorithm is known
 - But to convince someone that there was a large independent set all you'd need to do is show it to them
 - they can easily convince themselves that the set is large enough and independent
 - Convincing someone that there isn't one seems much harder
- We will show that Independent Set is NP-complete
 - Class of all the hardest problems that have the property above

21



Competitive Facility Location

- Two players competing for market share in a geographic area
- e.g. McDonald's, Burger King
- Rules:
 - Region is divided into n zones, 1,...,n
 - Each zone i has a value b_i
 - Revenue derived from opening franchise in that zone
 - No adjacent zones may contain a franchise
 - i.e., zoning regulations limit density
- Players alternate opening franchises
- Find: Given a target total value B is there a strategy for the second player that always achieves ≥ B?

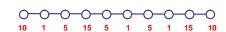
22



Competitive Facility Location

- Model geography by
 - A graph G=(V,E) where
 - V is the set {1,...,n} of zones
 - E is the set of pairs (i,j) such that i and j are adjacent zones
- Observe:
 - The set of zones with franchises will form an independent set in G

Competitive Facility Location



Target B = 20 achievable ?

What about B = 25?

24



Competitive Facility Location

- Checking that a strategy is good seems hard
 You'd have to worry about all possible responses at each round!
 - a giant search tree of possibilities
- Problem is PSPACE-complete
 - Likely strictly harder than NP-complete problems
 - PSPACE-complete problems include
 - Game-playing problems such as n×n chess and checkers
 - Logic problems such as whether quantified boolean expressions are always true
 - Verification problems for finite automata