Approximation algorithms 2 TSP, k-Center, Knapsack

CS240

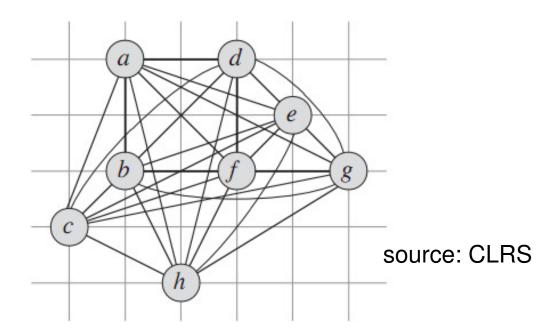
Spring 2023

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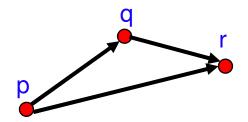
Traveling Salesman Problem

- Input A complete graph with weights on the edges.
- Output A cycle that visits each city once.
- Goal Find a cycle with minimum total weight.



Metric TSP

- TSP is NP-hard. In fact, it's even NP-hard to approximate when weights can be arbitrary.
- However, TSP is approximable for special types of weights.
- A weighted graph satisfies the triangle inequality if for any 3 vertices p, q, r, we have d_{pq}+d_{qr} ≥ d_{pr}.
 - □ I.e., direct path is always no worse than a roundabout path.
 - ☐ This is called a metric TSP.
- There is a 1.5-approx algorithm for TSP in graphs with the triangle inequality.
 - □ Let's look at a simpler 2-approx first.



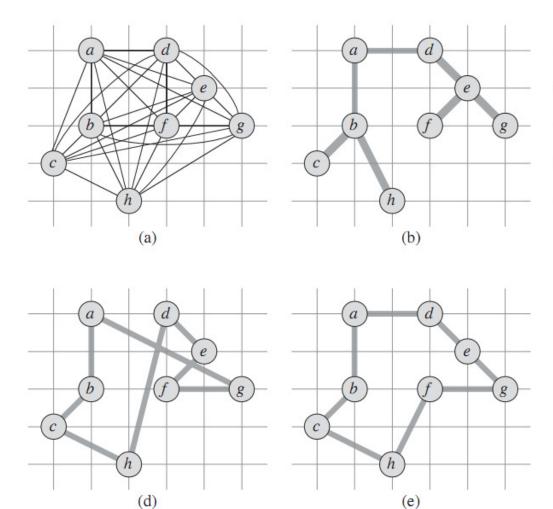


A 2-approximation for TSP

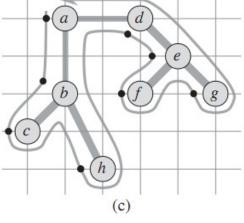
- Construct a minimum spanning tree T on G.
- Use depth-first traversal to visit all the vertices in T, starting from an arbitrary vertex.
- Convert this depth-first traversal T' to a cycle H that doesn't revisit any vertex.
- Return H as the TSP tour.



Example



source: CLRS

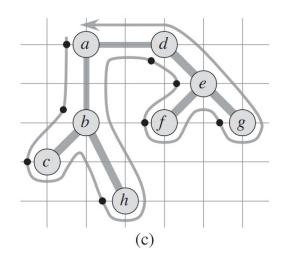


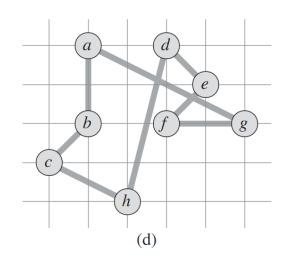
- (b) The MST T.
- (c) visit T in order abcbhbadefegeda.
- (d) converts the tour from
 - (c) to a Hamiltonian cycle, that doesn't revisit any vertices.
- (e) is the optimal TSP.



Making the tour Hamiltonian

- To go from (c) to (d), we need to make a tour T' that revisits vertices into a cycle H that doesn't revisit vertices.
- We use shortcutting.
 - If we revisit a vertex in T', we directly jump to the next vertex in T' we haven't visited.
 - We allow revisiting the first vertex.
 - □ The sequence of vertices we now visit is H.
 - □ Ex abcbhbadefegeda → abchdefga.





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Making the tour Hamiltonian

- Lemma If H is the shortcut of T', then c(H)≤c(T').
- Proof We formed H from T' by skipping over some vertices. E.g. we directly went from c to h, skipping over b.
 - □ But by the triangle inequality, $d_{cb}+d_{bh} \ge d_{ch}$.
 - So shortcutting from c to h didn't increase the distance.
 - □ The same thing applies to all our shortcuts.
 - □ So H is no longer than T'.

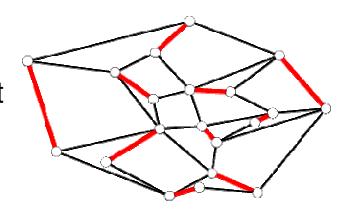
Proof of 2-approximation

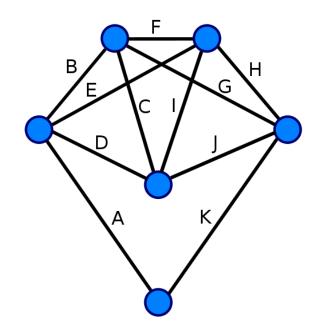
- Let H* be an optimum TSP.
- If we delete an edge from H*, we get a spanning tree.
- Since T is an MST, $c(T) \le c(H^*)$.
- Call the path from the depth-first traversal T'.
 - □ T' crosses each edge in T twice.
 - \square So c(T') = 2 c(T).
- Let H be the outcome of shortcutting T'.
 - □ H is a Hamiltonian cycle. It visits all the vertices, and ends where it started.
 - \Box c(H) \leq c(T'), by the lemma.
 - \Box c(H) \leq c(T') = 2 c(T) \leq 2 c(H*).
- So H is a 2-approximation.



Matchings and Euler cycles

- A matching in a graph is a set of nonintersecting edges.
 - A perfect matching is a matching that includes every vertex.
- An Euler tour of a graph is a path that starts and ends at the same vertex, and visits every edge once.
 - □ Hamiltonian tour visits every vertex once.
- Thm (Euler) A graph has an Euler tour if and only if all vertices have even degree.
- Note how deciding if graph has Euler tour is trivial, but deciding if it has Hamiltonian tour is NPC!

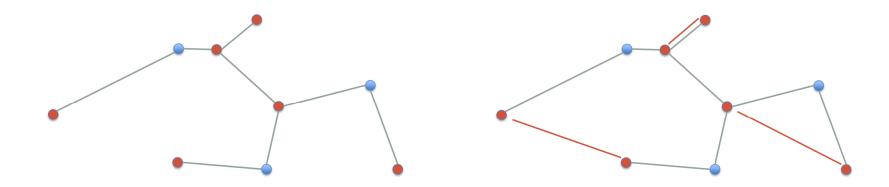






Christofides 3/2-approx algorithm

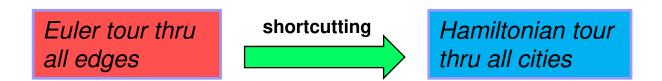
- ❖ A 3/2-approximation for TSP with triangle inequality.
- Construct a minimum spanning tree T on G.
- Find the set V' of odd degree vertices in T.
- Construct a minimum cost perfect matching M on V'.
- Add M to T to obtain T'.
- Find an Euler tour T" in T'.
- Shortcut T" to obtain a Hamiltonian cycle H. Output as the TSP.





Why Christofides works well

- In the 2-approx, we found a TSP by "doubling" the MST to an Euler tour, then shortcutting.
 - □ We need to start with Euler tour before shortcutting to ensure we visit all cities.



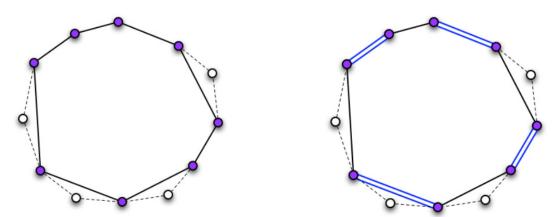
- Key to Christofides is to find a shorter Euler tour, without doubling the MST.
 - □ A graph with only even degree vertices always has Euler tour.
 - So we want to modify the MST to have all even degrees, by adding a matching.



- Lemma T' has an Euler tour.
- Proof There are an even number of vertices in V', because the total degree of T is even.
 - □ Since G is a complete graph and |V'| is even, there's a perfect matching on V'.
 - The min cost perfect matching can be found in O(n²) time using the blossom algorithm.
 - □ The degree of every node in M is odd. Since V' are the odd degree nodes in T, adding M to T makes all nodes in T' have even degree.
 - □ T' has Euler tour by Euler's theorem.



- Lemma Let H* be an optimal TSP on G, and let m be the cost of M. Then m ≤ c(H*)/2.
- Proof Let H' be the optimal TSP on V'.
 - \Box c(H') \leq c(H*) because H' is an optimal TSP on fewer vertices.
 - □ H' is a cycle on V', so it consists of two matchings on V'. The cheaper one has cost m' \leq c(H')/2 \leq c(H*)/2.
 - \square m \leq m' because M has min cost.





Proof of 3/2-approximation

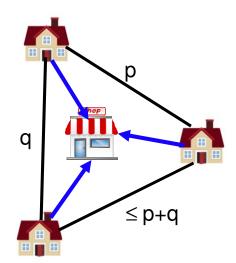
- Thm Let H be the TSP output by Christofides and let H* be an optimal TSP. Then c(H) ≤ 3/2*c(H*).
- Proof
 - \Box c(T) \leq c(H*) because T is an MST.
 - \Box c(T') = c(M) + c(T) \le c(H*)/2 + c(H*) = 3/2*c(H*).
 - \Box c(H) \leq c(T') because H is the shortcut of T'.

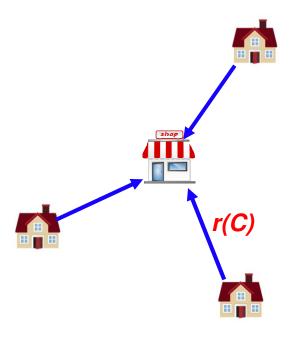
- □ Construct a minimum spanning tree T on G.
- ☐ Find set V' of odd-degree vertices in T.
- □ Construct a minimum cost perfect matching M on V'.
- \square Add M to T to obtain T'.
- ☐ Shortcut T' to obtain a Hamiltonian cycle. Output as the TSP.



k-Center problem

- Given a city with n sites, we want to build k centers to serve them.
 - □ Let S be set of sites, C be set of centers.
- Each site uses the center closest to it.
 - □ Distance of site s from the nearest center is $d(s,C) = \min_{c \in C} d(s,c)$.
- Goal is to make sure no site is too far from its center.
 - We want to minimize the max distance that any site is from its closest center.
 - Minimize r(C)=max_{s∈S} min_{c∈C} d(s,c).
 - C is called a cover of S, and r is called C's radius.
 - Where should we put centers to minimize the radius?
- Assume distances satisfy triangle inequality.





Gonzalez's algorithm

- k-Center is NP-complete.
- We'll give a simple 2-approximation for it.
- Idea Say there's one site that's farthest away from all centers. Then it makes the radius large. We'll put a center at that site, to reduce the radius.
 - □ Note we allow putting center at same location as site.



Gonzalez's algorithm

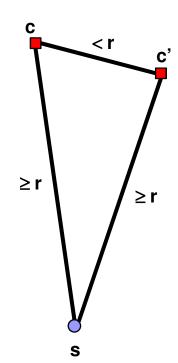
C is set of centers, initially empty.

- □ repeat k times
 - choose site s with maximum d(s,C)
 - □add s to C
- □ return C

■ Note The centers are located at the sites.



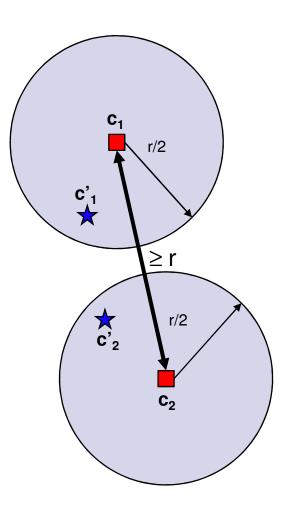
- Let C be the algorithm's output, and r be C's radius.
 - \Box r = max_{s∈S} min_{c∈C} d(s,c)
- Lemma 1 For any c,c'∈ C, d(c,c')≥r.
- Proof Since r is the radius, there exists a point s∈ S at distance ≥ r from all the centers.
 - \square If there's no such s, then C's radius < r.
 - \square So s is distance \ge r from c and c'.
 - □ Suppose WLOG c' is added to C after c.
 - □ If d(c,c')<r, then algorithm would add s to C instead of c', since s is farther.



b/A

- Cor There exist k+1 points mutually at distance $\geq r$ from each other.
 - □ By the lemma, the k centers are mutually ≥ r distance apart.
 - \square Also, there's an $s \in S$ at distance $\ge r$ from all the centers.
 - Otherwise C's covering radius is < r.
 - □ So the k centers plus s are the k+1 points.
- Call these k+1 points D.

- Let C* be an optimal cover with radius r*.
- Lemma 2 Suppose r > 2r*. Then for every c∈ D, there exists a corresponding c'∈ C*. Furthermore, all these c' are unique.
- Proof Draw a circle of radius r/2 around each c∈ D.
 - □ There must be a c'∈ C* inside the circle, because
 - c is at most distance r* away from its nearest center, since r* is C*'s radius.
 - r/2>r*.
 - □ Given $c_1, c_2 \in D$, let $c'_1, c'_2 \in C^*$ be inside c_1 and c_2 's circle, resp.
 - \Box c₁ and c₂'s circles don't touch, because $d(c_1,c_2) \ge r$.
 - □ So $c'_1 \neq c'_2$.



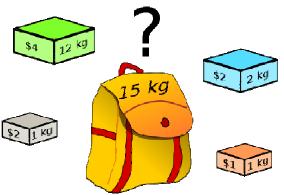


- Thm Let C be the output of Gonzalez's algorithm, and let C* be an optimal kcenter. Then r(C) ≤ 2r(C*).
- Proof By Lemma 2, if r(C)>2r(C*), then for every c∈ D, there is a unique c'∈ C*.
 - \square But there are k+1 points in D, by the corollary.
 - □ So there are k+1 points in C*. This is a contradiction because C* is a k-center.



The knapsack problem

- We have a set of items, each having a weight and a value.
- We have a knapsack that can carry up to W amount of weight.
- We want to put items in the knapsack to maximize the total value, but not exceed the weight limit.
- Ex Items 3 and 4 are the highest value items with weight ≤ 11.
- Assume all items have weight ≤ W, i.e. any single item fits in knapsack.





W = 11

Item	Value	Weight
1	1	1
2	6	2
3	18	5
4	22	6
5	28	7

b/A

A dynamic program for knapsack

- Let OPT(i,v) = minimum weight of a subset of items 1,...,i that has value ≥ v.
- If optimal solution uses item i.
 - □ Then we pay w_i weight for item i, and need to achieve value $\geq v-v_i$ using items 1,...,i-1 using min weight.
 - □ So OPT(i,v)= W_i +OPT(i-1,v- V_i).
- If optimal solution doesn't use item i.
 - \square Then we need to achieve value $\ge v$ using items 1,...,i-1.
 - □ So OPT(i,v)=OPT(i-1,v).
- Choose the case that gives smaller weight.

■ OPT(i,v) = 0 if v=0

$$\infty$$
 if i=0, v>0
min(OPT(i-1,v), w_i+OPT(i-1,v-v_i)) otherwise

NA.

Running time of dynamic program

- Say there are n items, and the largest value of any item is v*.
- The max value we can pack into the knapsack is nv*, where v* is the largest v value.
- Solve all subproblems of the form OPT(i,v), where $i \le n$ and $v \le nv^*$.
 - \square This is a total of $O(n^2v^*)$ subproblems.
- The solution to Knapsack is the max value V that can be packed with weight ≤ W.
- Having solved all the subproblems, we can find V by finding the subproblem with the largest value that has optimum weight ≤ W.
 - $\square V = \max_{v \le nv^*} OPT(n,v) \le W.$
- So solving Knapsack takes total time $O(n^2v^*)$.



Running time of dynamic program

- The DP gives an optimal solution to Knapsack and takes O(n²v*) time. Have we found a polytime algorithm for an NPcomplete problem?
- No. The problem size is O(n log(v*)), because it takes log(v*) bits to express each item's value. But O(n²v*) is not polynomial in n log(v*).
- To make this DP fast, we have to make the largest value small.

PTAS

- Let ε>0 be any number. We'll give a (1+ε)approximation for knapsack.
- By setting ε sufficiently small, we can get as good an approximation as we want!
 - ☐ This type of algorithm is called a polynomial time approximation scheme, or PTAS.
- Contrast this with earlier algs we studied, which had worse approx ratios, e.g. 2 or log n.
- But the running time will be $O(n^3/\epsilon)$. Hence we can't set ϵ =0 get the optimal solution.
- We're trading accuracy for time. The more accurate (smaller ε), the more time the algorithm takes.



Main idea: rounding

- Since we only need an approximate solution, we can change the values of the items a little (round the values) and not affect the solution much.
- We scale and round the original values to make them small.
- The previous DP took O(n²v*) time. So if the rounded values are small, this DP is fast.

W = 11

Item	Value	Weight
1	134,221	1
2	656,342	2
3	1,810,013	5
4	22,217,800	6
5	28,343,199	7

W = 11

Item	Value	Weight
1	2	1
2	7	2
3	19	5
4	223	6
5	284	7
4	223	6

Rounding

- Let ε >0 be the precision we want.
- Set $\theta = \frac{\epsilon v^*}{2n}$ to be a scaling factor.
 - \square v* is the largest value of any item.
- Scale all values down by θ then round up.
 - \square \lor '= $\lceil \lor/\theta \rceil$.
- Make a problem where each value v_i is replaced by v'_i.
 - □ Call this the scaled rounded problem.
- Let v^ be max value in the scaled rounded problem. Then $v^* = [v^*/\theta] = [v^*/(\epsilon v^*/2n)] = [2n/\epsilon]$.
- Running time of DP on scaled rounded problem is $O(n^2v^{\wedge}) = O(n^3/\epsilon)$.

Solving the original problem

- Make another new problem in which each value v_i is replaced by $u_i = [v_i/\theta]^*\theta$.
 - □ Call this the rounded problem.
 - \square We have $u_i \ge v_i$, and $u_i \le v_i + \theta$.
- Note u values are equal to v' values multiplied by θ .
 - ☐ Thus, the optimal solution for the rounded problem and the scaled rounded problem are the same.
- We now have 3 problems, the original problem, the scaled rounded problem, and the rounded problem.
- Let S be the optimal solution to the scaled rounded problem, which we can find in time O(n³/ε). S is also optimal for the rounded problem.
- We'll show S is a 1+ε approximation for the original problem.

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Correctness

■ Thm Let S* be the optimal solution to the original problem. Then $(1+\varepsilon)\sum_{i\in S}v_i \geq \sum_{i\in S^*}v_i$. Hence S is a $(1+\varepsilon)$ -approximate solution.

Proof

$$\sum_{i \in S^*} v_i \le \sum_{i \in S^*} u_i \qquad \qquad U_i \ge V_i$$

$$\leq \sum_{i \in S} u_i$$
 S is opt soln for rounded problem

$$\leq \sum_{i \in S} (v_i + \theta)$$
 $U_i \leq V_i + \theta$

$$\leq \sum_{i \in S} v_i + n \theta$$
 $|S| \leq n$

Correctness

- Suppose item j has the largest value, so $v^*=v_j$. Then $n\theta = \frac{\varepsilon}{2}v_j \le \frac{\varepsilon}{2}u_j \le \frac{\varepsilon}{2}\sum_{i \in S}u_i$
 - □ Last inequality because item j itself is feasible solution, so opt solution S is no smaller.
- So $\sum_{i \in S} v_i \ge \sum_{i \in S} u_i n\theta \ge \left(\frac{2}{\varepsilon} 1\right) n\theta$, where first inequality comes inequalities on last page.
- Assuming $\varepsilon \le 1$, then $n \theta \le \varepsilon \sum_{i \in S} v_i$
- Finally, we have

$$\sum_{i \in S^*} v_i \le \sum_{i \in S} v_i + n \theta \le \sum_{i \in S} v_i + \varepsilon \sum_{i \in S} v_i = (1 + \varepsilon) \sum_{i \in S} v_i$$



Summary

- We gave a DP for Knapsack.
- We scale and round to reduce number of different item values.
- Running the DP on the scaled rounded problem and using the solution for the original problem leads to an arbitrarily good approximation for Knapsack, a PTAS.
- There are PTAS's for a number of other problems.
 - Multiprocessor scheduling.
 - □ Bin packing.
 - □ Euclidean TSP.
- However, there are also many problems for which PTAS's do not exist, unless P=NP.