

Approximation Algorithms (Part II)

Traveling Salesperson Problem (TSP)

Input: Edge Weighted graph $G=(V, E, w)$

Output: A Hamiltonian cycle of min weight
—
w.l.o.g. we can assume G is the complete graph.

Thm: TSP is NP-complete.

Pf: A solution can be checked to be a valid H.C. of the claimed weight in polytime.

Given an instance of H.C. problem $H=(V, E_H)$ we construct a weighted graph $G=(V, \binom{V}{2})$ where for a pair (u, v) :

$$w_{(u,v)} = \begin{cases} 1 & (u,v) \in E_H \\ 2 & (u,v) \notin E_H \end{cases}$$

Now if H has a H.C., then there is a TSP sol'n of weight n . O/w, TSP sol'n must be $\geq n+1$. $\square$

Thm: For any function $f(n)$ that's computable in $\text{poly}(n)$ time, no $f(n)$ -apx of TSP can be obtained in $\text{poly}(n)$ time unless $P=NP$.

Pf: In the proof above, set the weights of $(u,v) \notin E_H$ to be $f(n)$. $\square$

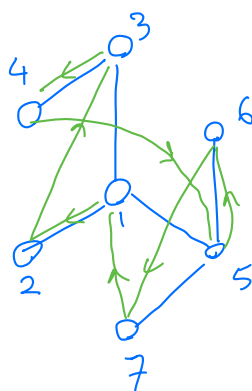
Good News

Metric Assumption: For any 3 disjoint vertices u, v, w , we have

$$w(u, w) \leq w(u, v) + w(v, w).$$

Thm: A 2-apx of metric TSP can be obtained in $O(n^2 \log n)$ time.

Pf: Let M be a minimum spanning tree of the input G which takes $O(n^2 \log n)$ time. We run a DFS on M and return the vertices in the order visited by DFS.



output: 1 2 3 4 5 6 7

↑
A

* Obs: $w(M) \leq \text{OPT}$.

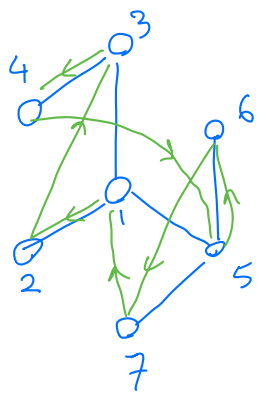
Pf: Take OPT, remove one edge and the result is a spanning tree.

* $w(A) \leq 2 \cdot w(M)$.

Putting the two together implies that

$$w(A) \leq 2w(M) \leq 2 \text{OPT}.$$

Note that a DFS over M goes over each edge of M exactly twice (considering backtrack edges). This implies there is a tour (i.e., a walk that visits every vertex at least once and goes back to the original vertex) of weight $2 \cdot w(M)$.



tour:

1 2 1 3 4 3 1 5 6 5 7 5 1
 x x x x x x

output: 1 2 3 4 5 6 7

The tour can be turned into a Hamiltonian cycle by shortcutting over already seen vertices. This doesn't increase the cost of the tour because of the metric assumption. $\square$

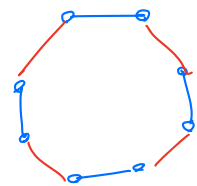
Suppose that all distances are either 1 or 2. (AKA $(1,2)$ -metric)

Thm: A 1.5-approx of TSP can be obtained in $(1,2)$ -metrics.

pf: Let M be a perfect matching of min total weight. We return an arbitrary H.C. consisting of all edges of M .

Observe that M has weight at most $\frac{TSP}{2}$: Take TSP, it can be decomposed into two perfect matchings by coloring its edges alternatively as "blue" and "red".

One of the two matchings must have weight $\leq \frac{TSP}{2}$, therefore



The min cost p.m. also has weight $\leq \frac{TSP}{2}$. Therefore the output is

$$\text{of weight} \leq \frac{TSP}{2} + \frac{n}{2} \cdot 2.$$

We also know $TSP \geq n$.

Therefore our sol'n is $\leq 1.5 TSP$. $\square$

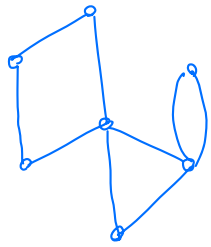
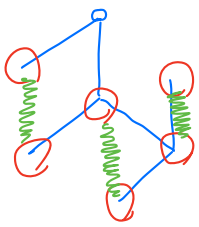
Thm: (Christofides '76): A 1.5-approx of TSP can be obtained in polytime even in general metrics.

pf sketch:

Take an MST M of G .

Find a min cost perfect matching P between the odd-degree vertices of M .

Obs: $\underbrace{P \cup M}_{\text{multiset}}$ is a subgraph where every vertex has even degree.



Fact: Any graph with all even-degree vertices has a tour that visits every edge exactly once.

This tour can be turned into H.C. of weight at most $w(P) + w(M)$ using the metric assumption & shortcutting as before.

We already know $w(M) \leq \text{OPT}$

we also have $w(P) \leq \frac{\text{OPT}}{2}$. $\square$

In 2021 Karlin, Klein, Oveis Gharan showed that a $1.5 - 10^{-36}$ apx of metric TSP can be found in polytime.

Remark: There is fixed $\epsilon > 0$ s.t.

obtaining a $(1+\epsilon)$ -apx of metric TSP

is NP-hard.

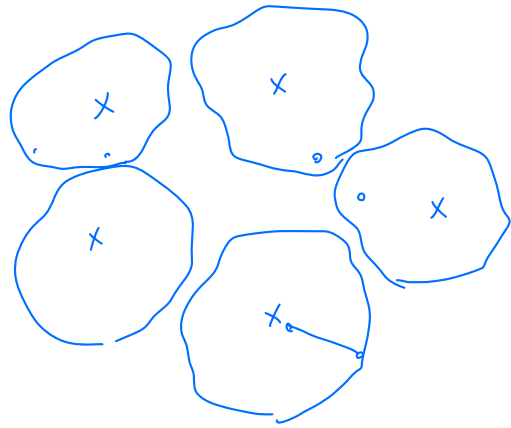
k-center clustering

We are given a complete weighted graph $G = (V, \binom{V}{2}, w)$ and parameter k .

The goal is to select a subset $U \subseteq V$ of size $|U| = k$ such that

$$\max_{v \in V} \min_{u \in U} w(v, u)$$

is minimized.



Remark: The problem is NP-hard to apx within a factor of ~ 1.8 even in metric spaces.

Thm [Gonzalez & Teofila '85]: metric k-center can be 2-approximated in poly time.

Alg: start with an arbitrary center $u_1 \in V$.

For $i=2$ to k :

$$u_i \leftarrow \arg \max_{u \in V} \min_{u_j} (u, u_j)$$