

Recent Progress in Approximation Algorithms for the Traveling Salesman Problem

Lecture 3: The s-t path TSP

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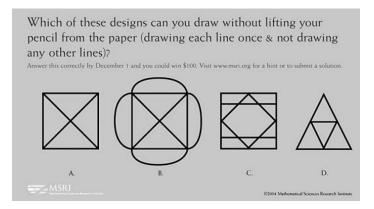
Recall the *s*-*t* path TSP: Usual TSP input plus $s, t \in V$, find a min-cost path from *s* to *t* visiting all other nodes in between (an *s*-*t* Hamiltonian path).

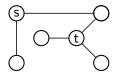
Eulerian path

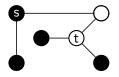
There is an Eulerian path that starts at s, ends at t, and visits every edge exactly once iff s and t have odd-degree and all other vertices have even degree.

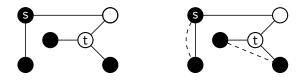
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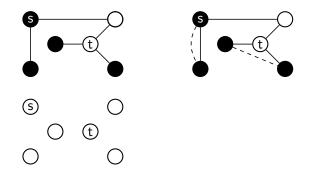
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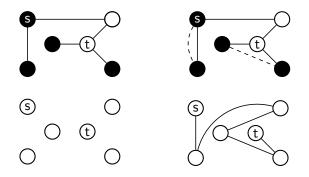




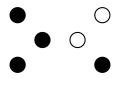




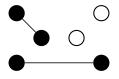




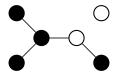




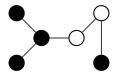










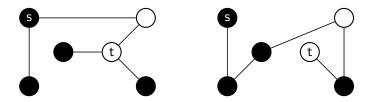


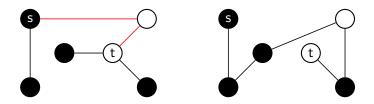
Let F be the min-cost spanning tree. Let T be the set of vertices whose parity needs changing. Then find a minimum-cost T-join J. Find Eulerian path on $F \cup J$; shortcut to an *s*-*t* Hamiltonian path.

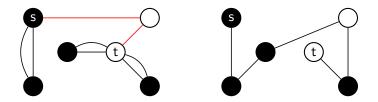
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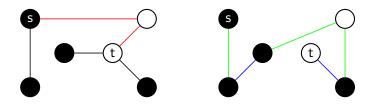
Theorem

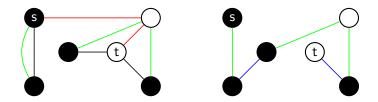
Hoogeveen's algorithm is a $\frac{5}{3}$ -approximation algorithm.

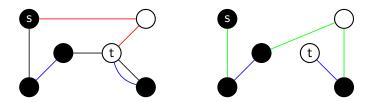


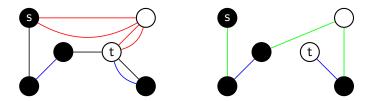


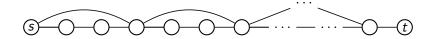


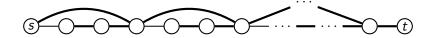




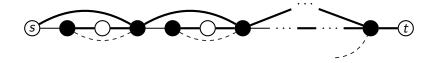












No improvement on Hoogeveen's algorithm for s-t path TSP, until just the last few years.

An, Kleinberg, Shmoys	2012	1.618
Sebő	2013	1.6
Vygen	2015	1.599
Gottschalk and Vygen	2015	1.56
Sebő and Van Zuylen	2016	1.52

Goal: Understand the An et al. algorithm and analysis; will sketch some of the ideas of the improvements.

A Linear Programming Relaxation

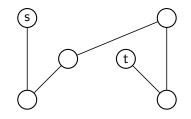
$$\begin{array}{ll} {\sf Min} & \sum_{e\in {\sf E}} c_e x_e \\ {\sf subject to:} & x(\delta(v)) = \left\{ & \\ & x(\delta(S)) \geq \left\{ & \\ & 0 \leq x_e \leq 1, \qquad \forall e \in {\sf E}, \end{array} \right. \end{array}$$

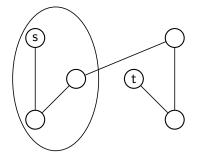
where $\delta(S)$ is the set of edges with exactly one endpoint in S, and $x(E') \equiv \sum_{e \in E'} x_e$.

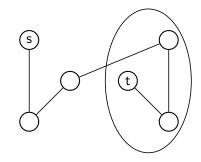
A Linear Programming Relaxation

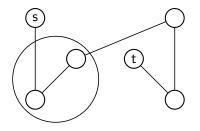
$$\begin{array}{lll} \text{Min} & \sum_{e \in E} c_e x_e \\ \text{subject to:} & x(\delta(v)) = \left\{ \begin{array}{ll} 1, & v = s, t, \\ 2, & v \neq s, t, \end{array} \right. \\ & x(\delta(S)) \geq \left\{ \begin{array}{ll} 1, & |S \cap \{s, t\}| = 1, \\ 2, & |S \cap \{s, t\}| \neq 1, \end{array} \right. \\ & 0 \leq x_e \leq 1, \qquad \forall e \in E, \end{array} \right. \end{array}$$

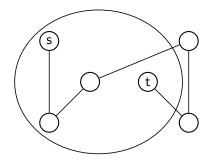
where $\delta(S)$ is the set of edges with exactly one endpoint in S, and $x(E') \equiv \sum_{e \in E'} x_e$.









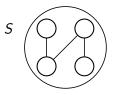


The spanning tree polytope

The spanning tree polytope (convex hull of all spanning trees) is defined by the following inequalities:

$$egin{aligned} & x(E) = |V| - 1, \ & x(E(S)) \leq |S| - 1, \ & \forall |S| \subseteq V, |S| \geq 2, \ & x(e) \geq 0, \ & \forall e \in E, \end{aligned}$$

where E(S) is the set of all edges with both endpoints in S.



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The LP relaxation and spanning trees

Lemma

Any solution x feasible for the s-t path TSP LP relaxation is in the spanning tree polytope.

Proof

$$\begin{aligned} x(\delta(v)) &= \begin{cases} 1, & v = s, t, \\ 2, & v \neq s, t, \end{cases} \\ x(\delta(S)) &\geq \begin{cases} 1, & |S \cap \{s, t\}| = 1, \\ 2, & |S \cap \{s, t\}| \neq 1, \end{cases} \\ 0 &\leq x(e) \leq 1, \quad \forall e \in E. \end{aligned}$$

$$\begin{split} & x(E) = |V| - 1, \\ & x(E(S)) \leq |S| - 1, \quad \forall |S| \subseteq V, |S| \geq 2, \\ & x(e) \geq 0, \quad \forall e \in E. \end{split}$$

A warmup to the improvements

Let OPT_{LP} be the value of an optimal solution x^* to the LP relaxation.

Theorem (An, Kleinberg, Shmoys (2012))

Hoogeveen's algorithm returns a solution of cost at most $\frac{5}{3}OPT_{LP}$.

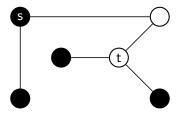
An extremely useful lemma

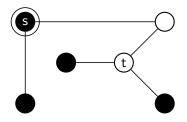
Let F be a spanning tree, and let T be the vertices whose parity needs fixing in F.

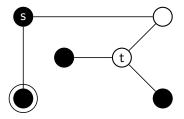
Definition

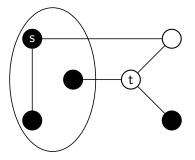
S is an *odd set* if $|S \cap T|$ is odd.

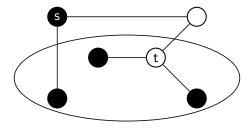
Lemma



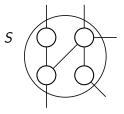








Proof of lemma



$$\sum_{v \in S} deg_F(v) = 2|E(S) \cap F| + |\delta(S) \cap F|$$

T-join LP

The solution to the following linear program is the minimum-cost T-join for costs $c \ge 0$:

subject to:

$$\begin{array}{ll} \mathsf{Min} & \sum_{e \in E} c_e x_e \\ & x(\delta(S)) \geq 1, \qquad \forall S \subseteq V, |S \cap T| \text{ odd} \\ & x_e \geq 0, \qquad \qquad \forall e \in E. \end{array}$$



$$\sum_{v \in S} deg_J(v) = 2|E(S) \cap J| + |\delta(S) \cap J|$$

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Proof of theorem

Theorem (An, Kleinberg, Shmoys (2012))

Hoogeveen's algorithm returns a solution of cost at most $\frac{5}{3}OPT_{LP}$.

Lemma

$$\begin{array}{ll} \mathsf{Min} & \sum_{e \in E} c(e) x(e) \\ & x(\delta(S)) \geq 1, \qquad \quad \forall S \subseteq V, |S \cap T| \text{ odd} \\ & x(e) \geq 0, \qquad \quad \forall e \in E. \end{array}$$

Convex combination

Let x^* be an optimal LP solution. Let χ_F be the *characteristic* vector of a set of edges F, so that

$$\chi_F(e) = \begin{cases} 1 & e \in F \\ 0 & e \notin F \end{cases}$$

Since x^* is in the spanning tree polytope, can write x^* as a convex combination of spanning trees F_1, \ldots, F_k :

$$x^* = \sum_{i=1}^k \lambda_i \chi_{F_i},$$

such that $\sum_{i=1}^{k} \lambda_i = 1$, $\lambda_i \ge 0$.

Best-of-Many Christofides' Algorithm

An, Kleinberg, Shmoys (2012) propose the *Best-of-Many Christofides*' algorithm: given optimal LP solution x^* , compute convex combination of spanning trees

$$x^* = \sum_{i=1}^k \lambda_i \chi_{F_i}.$$

For each spanning tree F_i , let T_i be the set of vertices whose parity needs fixing, let J_i be the minimum-cost T_i -join. Find *s*-*t* Hamiltonian path by shortcutting $F_i \cup J_i$. Return the shortest path found over all *i*.

Best-of-Many Christofides' Algorithm

$$x^* = \sum_{i=1}^k \lambda_i \chi_{F_i}.$$

For each spanning tree F_i , let T_i be the set of vertices whose parity needs fixing, J_i be the minimum-cost T_i -join. Find *s*-*t* Hamiltonian path by shortcutting $F_i \cup J_i$. Return the shortest path found over all *i*.

Theorem

The Best-of-Many Christofides' algorithm returns a solution of cost at most $\frac{5}{3}OPT_{LP}$.

Proof

To do better, we need to improve the analysis for the costs of the T_i -joins; recall that we use that

$$y_i = \frac{1}{3}\chi_{F_i} + \frac{1}{3}x^*$$

is feasible for the T_i -join LP.

Consider

$$y_i = \alpha \chi_{F_i} + \beta x^*.$$

Then the cost of the best s-t Hamiltonian path is at most

$$(1 + \alpha + \beta)OPT_{LP}.$$

Proof that y_i feasible for T_i -join LP had two cases. Assume S odd $(|S \cap T_i| \text{ odd})$.

If $|S \cap \{s,t\}|
eq 1$, then

$$y_i(\delta(S)) = \alpha |F_i \cap \delta(S)| + \beta x^*(\delta(S)) \ge \alpha + 2\beta.$$

We will want $\alpha + 2\beta \ge 1$, so the T_i -join LP constraint is satisfied.

If $|S \cap \{s, t\}| = 1$, then

 $y_i(\delta(S)) = \alpha |F_i \cap \delta(S)| + \beta x^*(\delta(S)) \ge 2\alpha + \beta x^*(\delta(S)).$

If $|S \cap \{s, t\}| = 1$, then

$$y_i(\delta(S)) = \alpha |F_i \cap \delta(S)| + \beta x^*(\delta(S)) \ge 2\alpha + \beta x^*(\delta(S)).$$

Since we assume $\alpha+2\beta\geq 1$, we only run into problems if

$$x^*(\delta(S)) < rac{1-2lpha}{eta}.$$

Note that $\alpha = 0$, $\beta = \frac{1}{2}$ works if $x^*(\delta(S)) \ge 2$ for all $S \subset V$, and gives a tour of cost at most $\frac{3}{2}OPT_{LP}$.

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So focus on cuts for which $x^*(\delta(S)) < 2$, and add an extra "correction" term to y_i to handle these cuts.

$$\tau\text{-Narrow}$$
 Cuts

Definition

S is τ -narrow if $x^*(\delta(S)) < 1 + \tau$ for fixed $\tau \leq 1$.

Only S such that $|S \cap \{s, t\}| = 1$ are τ -narrow.

Definition

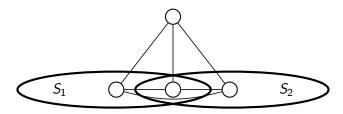
Let C_{τ} be all τ -narrow cuts S with $s \in S$.



The τ -narrow cuts in C_{τ} have a nice structure.

Theorem (An, Kleinberg, Shmoys (2012)) If $S_1, S_2 \in C_{\tau}$, $S_1 \neq S_2$, then either $S_1 \subset S_2$ or $S_2 \subset S_1$. First need to show that

$$x^*(\delta(S_1)) + x^*(\delta(S_2)) \ge x^*(\delta(S_1 - S_2)) + x^*(\delta(S_2 - S_1)).$$



Proof of theorem

Theorem (An, Kleinberg, Shmoys (2012))

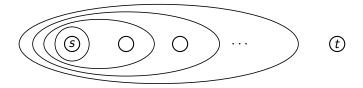
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Proof of theorem

Theorem (An, Kleinberg, Shmoys (2012))

If $S_1, S_2 \in \mathcal{C}_{\tau}$, $S_1 \neq S_2$, then either $S_1 \subset S_2$ or $S_2 \subset S_1$.

So the τ -narrow cuts look like $s \in Q_1 \subset Q_2 \subset \cdots \subset Q_k \subset V$.



Correction Factor

Let e_Q be the minimum-cost edge in $\delta(Q)$. Then consider the following (from Gao (2014)):

$$y_i = \alpha \chi_{F_i} + \beta x^* + \sum_{Q \in \mathcal{C}_{\tau}, |Q \cap T_i| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(Q))) \chi_{e_Q}$$

for $\alpha, \beta, \tau \geq 0$ such that

$$\alpha + 2\beta = 1$$
 and $\tau = \frac{1-2\alpha}{\beta} - 1.$

Theorem

 y_i is feasible for the T_i -join LP.

Proof

Two Lemmas

Recall $x^* = \sum_{i=1}^k \lambda_i \chi_{F_i}$, with $\sum_{i=1}^k \lambda_i = 1$ and $\lambda_i \ge 0$. So λ_i is a probability distribution on the trees F_i ; probability of F_i is λ_i .

Lemma

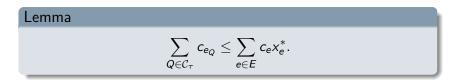
Let \mathcal{F} be a randomly sampled tree F_i , and \mathcal{T} the corresponding vertices T_i . Let $Q \in C_{\tau}$ be a τ -narrow cut. Then

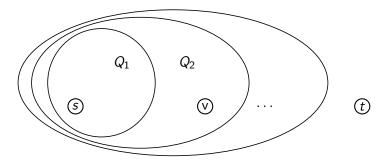
$$\Pr[|\delta(Q) \cap \mathcal{F}| = 1] \ge 2 - x^*(\delta(Q))$$

$$\Pr[|Q \cap \mathcal{T}| \text{ odd}] \le x^*(\delta(Q)) - 1.$$

Two Lemmas

Recall e_Q is the cheapest edge crossing a au-narrow cut $Q \in \mathcal{C}_{ au}$.





An-Kleinberg-Shmoys

Theorem (An, Kleinberg, and Shmoys (2012))

Best-of-Many Christofides' is a $\frac{1+\sqrt{5}}{2}$ -approximation algorithm for s-t path TSP.

Proof of AKS

For the proof, recall that e_Q is min-cost edge in $\delta(Q)$, C_{τ} are the cuts Q with $x^*(\delta(Q)) < 1 + \tau$,

$$y_i = \alpha \chi_{F_i} + \beta x^* + \sum_{Q \in \mathcal{C}_{\tau}, |Q \cap \mathcal{T}_i| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(Q))) \chi_{e_Q}$$

is feasible for the T_i -join LP, and

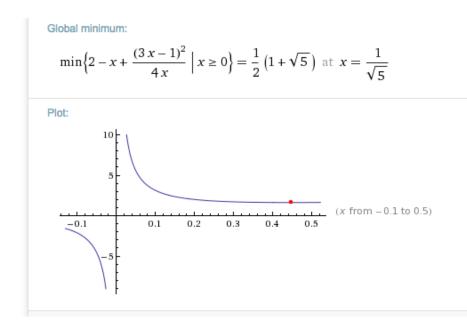
Lemma

Let \mathcal{F} be a randomly sampled tree F_i , and \mathcal{T} the corresponding vertices T_i . Let $Q \in C_{\tau}$ be a τ -narrow cut. Then

$$\begin{aligned} \mathsf{Pr}[|\delta(Q) \cap \mathcal{F}| = 1] &\geq 2 - x^*(\delta(Q)) \\ \mathsf{Pr}[|Q \cap \mathcal{T}| \ \textit{odd}] &\leq x^*(\delta(Q)) - 1. \end{aligned}$$

Lemma

$$\sum_{Q\in\mathcal{C}_{\tau}}c_{e_Q}\leq\sum_{e\in E}c_ex_e^*.$$



A $\frac{3}{2}$ -approximation algorithm for *s*-*t* TSP path in graph TSP instances.