Improving Christofides with Randomization & LP

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Traveling Salesman Problem

- (Circuit) Traveling Salesman Problem
 - Given a weighted graph G = (V, E) ($c : E \to \mathbb{R}_+$), find a minimum Hamiltonian circuit

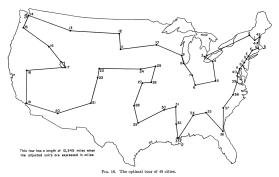
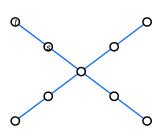


Figure from [Dantzig, Fulkerson, Johnson 1954].

- Christofides' algorithm
 - Find a minimum spanning tree \mathcal{T}_{\min}
 - Let T be the set of vertices with "wrong" parity of degree:
 i.e., T is the set of even-degree endpoints and other odd-degree vertices in Tmin
 - Find a minimum T-join J
 - Find an s-t Eulerian path of $\mathcal{T}_{min} \cup J$
 - Shortcut it into an s-t Hamiltonian path

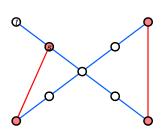
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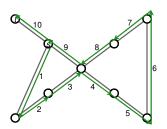
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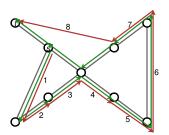
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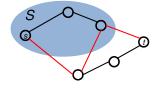
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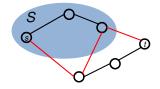
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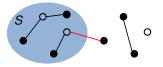
- For $x \in \mathbb{R}_+^E$ and $F \subset E$,
- $\circ x(F) := \sum_{f \in F} x_f$
- ∘ Incidence vector of F is $(\chi_F)_e := \begin{cases} 1 & \text{if } e \in F \\ 0 & \text{otherwise} \end{cases}$

Polyhedral Characterization of *T*-joins

Definition

For $T \subset V$, $J \subset E$ is a T-join if the set of odd-degree vertices in G' = (V, J) is T

 Polyhedral characterization of T-joins [Edmonds, Johnson 1973]



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 Call a feasible solution a fractional T-join; its cost upper-bounds c(J)

Path-variant Held-Karp Relaxation

• Path-variant Held-Karp relaxation For G = (V, E) and $s, t \in V$,

$$\begin{cases} \sum_{e \in \delta(\{s\})} x_e = \sum_{e \in \delta(\{t\})} x_e = 1 \\ \sum_{e \in \delta(\{v\})} x_e = 2, & \forall v \in V \setminus \{s, t\} \\ \sum_{e \in \delta(S)} x_e \ge 1, & \forall S \subsetneq V, |\{s, t\} \cap S| = 1 \\ \sum_{e \in \delta(S)} x_e \ge 2, & \forall S \subsetneq V, |\{s, t\} \cap S| \ne 1, S \ne \emptyset \\ 0 \le x_e \le 1 & \forall e \in E \end{cases}$$

$$x \in \mathbb{R}^E$$

Algorithm of An, Kleinberg, & S

- Best-of-Many Christofides' Algorithm
 - Compute an optimal solution x* to the Held-Karp relaxation
 - Rewrite x^* as a convex comb. of spanning trees $\mathcal{T}_1, \ldots, \mathcal{T}_k$
 - For each \mathcal{T}_i :
 - Let T_i be the set of vertices with "wrong" parity of degree: i.e., T_i is the set of even-degree endpoints and other odd-degree vertices in S_i
 - Find a minimum T_i -join J_i
 - Find an *s-t* Eulerian path of $\mathcal{T}_i \cup J_i$
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- Randomized algorithm for simpler analysis
- Sampling Christofides' Algorithm
 - Compute an optimal solution x^* to the Held-Karp relaxation
 - Rewrite x^* as a convex comb. of spanning trees $\mathcal{I}_1, \ldots, \mathcal{I}_k$: $x^* = \sum_{i=1}^k \lambda_i \chi_{\mathcal{I}_i}, \sum_{i=1}^k \lambda_i = 1$
 - Sample \mathscr{T} by choosing \mathscr{T}_i with probability λ_i
 - Let T be the set of vertices with "wrong" parity of degree: i.e., T is the set of even-degree endpoints and other odd-degree vertices in S
 - Find a minimum T-join J
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- $\Pr[e \in \mathscr{T}] = x_e^*$
 - $\mathsf{E}[c(\mathscr{T})] = \sum_{e \in E} c_e x_e^* = c(x^*)$
 - The rest of the analysis focuses on bounding c(J)

• Want: a fractional T-join y with $E[c(y)] \le \frac{2}{3}c(x^*)$ $x^* :=$ optimal path-variant Held-Karp solution $y := \alpha \chi_{\mathscr{T}} + \beta x^*$

Lemma

An s-t cut (U, \bar{U}) that is odd w.r.t. T (i.e., $|U \cap T|$ is odd) has at least two tree edges in it.

	$\chi_{\mathscr{T}}$	<i>X</i> *	
LB on T-odd s-t cut capacities	2	1	
LB on nonseparating cut capacities	1	2	

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LB on nonseparating cut capacities	1	2	$\alpha + 2\beta$

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$$\mathbf{y} := \alpha \chi_{\mathscr{T}} + \beta \mathbf{x}^*$$

- $y := \alpha \chi_{\mathscr{T}} + \beta x^*$ • Choose $\alpha = \beta = \frac{1}{2}$
- $\mathsf{E}[c(y)] = \alpha \mathsf{E}[c(\chi_{\mathscr{T}})] + \beta c(x^*) = (\alpha + \beta)c(x^*)$
- $\mathsf{E}[c(H)] \le \mathsf{E}[c(\mathscr{T})] + \mathsf{E}[c(J)] \le (1 + \alpha + \beta)c(x^*)$

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The given algorithm is a $(1 + \alpha + \beta)$ -approximation algorithm

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 Analysis also works for the original path-variant Christofides' algorithm

	$\chi_{\mathscr{T}}$	X^*	У
LB on T-odd s-t cut capacities	2	1	$2\alpha + \beta$
LB on nonseparating cut capacities	1	2	$\alpha + 2\beta$

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 - \bullet In particular, decrease α by $\mathbf{2}\epsilon$ and increase β by ϵ

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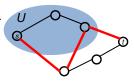
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- $E[c(y)] = (\alpha + \beta)c(x^*)$ decreases by $\epsilon c(x^*)$
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- T-odd s-t cuts with small capacity may become violated
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Definition

For $0< au\leq 1$, a au-narrow cut (U,\bar{U}) is an s-t cut with $\sum_{e\in\delta(U)}x_e^*<1+ au$



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Corollary

Each s-t cut (U, \bar{U}) with $x^*(\delta(U)) = 1$ is never odd w.r.t. T

$$(\textit{T-join}) \quad \begin{cases} \sum_{e \in \delta(\mathcal{S})} y_e \geq 1, & \forall \mathcal{S} \subset V, |\mathcal{S} \cap \mathcal{T}| \text{ odd} \\ y \in \mathbb{R}_+^{\mathcal{E}} \end{cases}$$

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Proof.

Expected number of tree edges in the cut is equal to $x^*(\delta(U))$:

$$\mathsf{E}[|\delta(U)\cap\mathscr{T}|] = \sum_{e\in\delta(U)}\mathsf{Pr}[e\in\mathscr{T}] = \sum_{e\in\delta(U)}x_e^* = 1$$

So $|\delta(U) \cap \mathcal{T}|$ is identically 1.

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Proof.

- τ -narrow \Rightarrow Expected # of tree edges in the cut is < 1 + τ
- If (U, \bar{U}) is odd w.r.t. T, there must be \geq 2 tree edges in it $\Rightarrow \Pr[|U \cap T| \text{ odd}] \leq \Pr[|\delta(U) \cap (T)| \geq 2]$
- Expt. # of tree edges in cut is $\geq 1 + \Pr[|\delta(U) \cap (T)| \geq 2] \geq 1 + \Pr[|U \cap T| \text{ odd}] \Rightarrow \Pr[|U \cap T| \text{ odd}] < \tau$



- Nonseparating cuts & s-t cuts with high capacities are safe
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- Suppose edge sets of τ -narrow cuts were disjoint
- For each τ -narrow cut (U, \bar{U}) , define "correction vector" f_U defined as the Held-Karp solution restricted to $\delta(U)$

$$(f_U)_e = \begin{cases} x_e^* & \text{if } e \in \delta(U) \\ 0 & \text{otherwise} \end{cases}$$

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$$y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}, |U\cap T| \text{ odd}} d \cdot f_U$$

$$\mathsf{E}\left[c(\sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}, |U\cap T| \text{ odd}} d \cdot f_U)\right]$$

$$\leq c\left(\sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}} \mathsf{Pr}[|U\cap T| \text{ odd}] \cdot d \cdot f_U\right)$$

$$\leq d\tau c\left(\sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}} f_U\right) \leq d\tau c(x^*)$$

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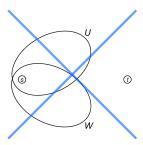
• The present algorithm is a 1.6572-approximation algorithm if τ -narrow cuts were disjoint: $E[c(y)] \le (\alpha + \beta + d\tau)c(x^*)$

 \bullet τ -narrow cuts are not disjoint

• τ -narrow cuts are not disjoint, but "almost" disjoint

Lemma

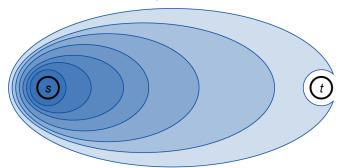
au-narrow cuts do not cross: i.e., for au-narrow cuts (U, \bar{U}) and (W, \bar{W}) with $s \in U, W$, either $U \subset W$ or $W \subset U$.



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Proof.

Suppose not. Neither $U \setminus W$ nor $W \setminus U$ is empty.

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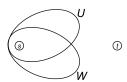
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 $x^*(\delta(U)) + x^*(\delta(W)) \ge x^*(\delta(U \setminus W)) + x^*(\delta(W \setminus U)) \ge 2 + 2$

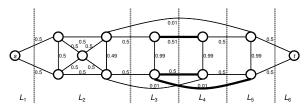


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Corollary

There exists a partition L_1, \ldots, L_ℓ of V such that

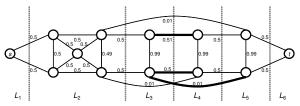
- $L_1 = \{s\}, L_\ell = \{t\}$, and
- $\{U|(U,\bar{U}) \text{ is } \tau\text{-narrow, } s \in U\} = \{U_i|1 \leq i < \ell\}, \text{ where } U_i := \cup_{k=1}^i L_k$



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Thick edges show F₃

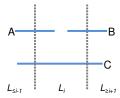
- We choose "representative edge set" $F_i := E(L_i, L_{\geq i+1})$ for each $\delta(U_i)$. We claim:
 - F_i's are disjoint
 - F_i has large capacity

Lemma

$$x^*(F_i) \geq 1 - \frac{\tau}{2}$$

$$A := x^*(E(L_{\leq i-1}, L_i)), B := x^*(E(L_i, L_{\geq i+1})),$$

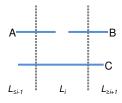
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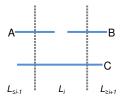


$$B+A \geq 2$$

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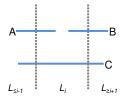
 $B+C \geq 1$

Lemma

$$x^*(F_i) \geq 1 - \frac{\tau}{2}$$

$$A := x^*(E(L_{\leq i-1}, L_i)), B := x^*(E(L_i, L_{\geq i+1})),$$

 $C := x^*(E(L_{< i-1}, L_{> i+1})).$



$$B+A \geq 2$$

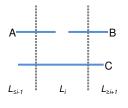
 $B+C \geq 1$
 $1+\tau > A+C$

Lemma

$$x^*(F_i) \geq 1 - \frac{\tau}{2}$$

Proof.

$$A := x^*(E(L_{\leq i-1}, L_i)), B := x^*(E(L_i, L_{\geq i+1})), C := x^*(E(L_{\leq i-1}, L_{\geq i+1})).$$



$$B+C \geq 1$$

$$1+\tau > A+C$$

$$2B > 2-\tau$$

$$x^*(F_i) = B > 1-\frac{\tau}{2}$$

 $B+A \geq 2$

- \bullet au-narrow cuts are the only cuts that may potentially be violated
- For τ -narrow cuts,
 - deficiency is at most $d = O(\epsilon)$
 - probability that the cut is odd w.r.t. T is at most $\tau = O(\epsilon)$
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- (Re)define f_U as Held-Karp solution restricted to F_i
- $y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow},|U\cap T| \text{ odd }} d \cdot f_U$ $y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow},|U\cap T| \text{ odd }} d \cdot \frac{1}{1-\frac{\tau}{2}} \cdot f_U$

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- $E[c(y)] \le (\alpha + \beta + d\tau)c(x^*)$ $E[c(y)] \le (\alpha + \beta + \frac{d\tau}{1-\frac{\tau}{2}})c(x^*) \le 0.6577c(x^*)$
- The present algorithm is a 1.6577-approximation algorithm

Summary of Feasible Fractional T-join Used

- Let $\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$ be τ -narrow cuts
- There exists a partition L_1, \ldots, L_ℓ of V such that
 - $L_1 = \{s\}, L_\ell = \{t\}, \text{ and }$
 - $\{U|(U,\bar{U}) \text{ is } \tau\text{-narrow, } s \in U\} = \{U_i|1 \leq i < \ell\},$ where $U_i := \bigcup_{k=1}^i L_k$
- We choose "representative edge set" $F_i := E(L_i, L_{\geq i+1})$ for each $\delta(U_i)$
- For each τ -narrow cut (U_i, \bar{U}_i) , define "correction vector" f_U as Held-Karp solution restricted to $F_i \subseteq \delta(U_i)$:

$$(f_U)_e = \begin{cases} x_e^* & \text{if } e \in F_i \\ 0 & \text{otherwise} \end{cases}$$

•
$$\mathbf{y} := \alpha \chi_{\mathscr{T}} + \beta \mathbf{x}^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}, |U\cap T| \text{ odd }} \mathbf{d} \cdot \frac{1}{1-\frac{\tau}{2}} \cdot \mathbf{f}_U$$

Gao's Feasible Fractional T-join

- Let $\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$ be τ -narrow cuts
- We choose "representative edge" e(i) for each $\delta(U_i)$ as cheapest in $\delta(U_i)$
- For each au-narrow cut (U_i, \bar{U}_i) , define "correction vector" f_U $(f_U)_e = \begin{cases} 1 & \text{if } e = e(i) \\ 0 & \text{otherwise} \end{cases}$
- Let T be the wrong parity of degree set for the random spanning tree T:

$$y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}, |U\cap T| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(U))) f_U$$

Gao's Feasible Fractional T-join

• Recall we set $\alpha + 2\beta = 1$; now set τ s.t. $2\alpha + \beta(1 + \tau) = 1$

Claim

For each τ -narrow cut U, $1 - 2\alpha - \beta x^*(\delta(U)) \ge 0$.

Proof.

Equivalent to have $2\alpha + \beta x^*(\delta(U)) \le 1$. Since $x^*(\delta(U)) < 1 + \tau$, the claim follows.

Lemma

$$y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow}, |U\cap T| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(U))) f_U$$

is a feasible fractional T-join,

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Proof.

Consider a set S such that $|S \cap T|$ is odd.

Case 1. If *S* is not an *s*-*t* cut (i.e., $|S \cap \{s,t\}| \neq 1$), then

$$y(\delta(S)) \ge \alpha |\mathcal{T} \cap \delta(S)| + \beta x^*(\delta(S))$$

 $\ge \alpha + 2\beta = 1$

Recall we set $\alpha + 2\beta = 1$

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Proof.

Consider a set S such that $|S \cap T|$ is odd.

Case 2a. If S is an s-t cut (i.e., $|S \cap \{s,t\}| = 1$) and S is not τ -narrow, then

$$y(\delta(S)) \ge \alpha |\mathcal{T} \cap \delta(S)| + \beta x^*(\delta(S))$$

 $\ge 2\alpha + \beta(1+\tau) = 1$

Recall we set $\alpha + 2\beta = 1$ and set τ s.t. $2\alpha + \beta(1 + \tau) = 1$

$$y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow},|U\cap T| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(U))) f_U$$
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Proof.

Consider a set S such that $|S \cap T|$ is odd.

Case 2b. If S is an s-t cut (i.e.,
$$|S \cap \{s,t\}| = 1$$
) and S is τ -narrow, then

$$\begin{aligned} y(\delta(S)) &\geq \alpha |\mathscr{T} \cap \delta(S)| + \beta x^*(\delta(S)) + \\ &\sum_{U:(U,\bar{U}) \text{ is } \tau - \text{narrow}, |U \cap T| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(U))) f_U(\delta(S)) \\ &\geq 2\alpha + \beta x^*(\delta(S)) + (1 - 2\alpha - \beta x^*(\delta(S))) \\ &= 1 \end{aligned}$$

An s-t cut (U, \bar{U}) that is odd w.r.t. T (i.e., $|U \cap T|$ is odd) has at least two tree edges in it

Corollary

Each s-t cut (U, \bar{U}) with $x^*(\delta(U)) = 1$ is never odd w.r.t. T

Corollary

For any τ -narrow cut (U, \overline{U}) , $\Pr[|U \cap T| \text{ odd}] < \tau$

- τ -narrow \Rightarrow Expected # of tree edges in the cut is < 1 + τ
- If (U, \bar{U}) is odd w.r.t. T, there must be \geq 2 tree edges in it $\Rightarrow \Pr[|U \cap T| \text{ odd}] \leq \Pr[|\delta(U) \cap (T)| \geq 2]$
- Expt. # of tree edges in cut is $\geq 1 + \Pr[|\delta(U) \cap (T)| \geq 2] \geq 1 + \Pr[|U \cap T| \text{ odd}] \Rightarrow \Pr[|U \cap T| \text{ odd}] < \tau$

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Corollary

For any τ -narrow cut (U, \overline{U}) , $\Pr[|U \cap T| \text{ odd}] \leq x^*(\delta(U)) - 1$

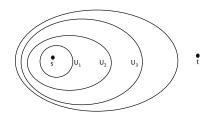
- Expected # of tree edges in the cut is $x^*(\delta(U))$
- If (U, \overline{U}) is odd w.r.t. T, there must be \geq 2 tree edges in it $\Rightarrow \Pr[|U \cap T| \text{ odd}] \leq \Pr[|\delta(U) \cap (T)| \geq 2]$
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Lemma

Let
$$\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$$
 be τ -narrow cuts. $\sum_{i=1}^{\ell-1} c_{e(i)} \leq c(\mathscr{T}_{min}) \leq \sum_e c_e x_e^*$

Proof.

Compute an injective map from U_i to edges of \mathscr{T}_{min}

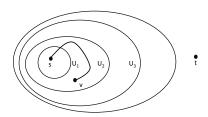


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Proof.

Compute an injective map from U_i to edges of \mathscr{T}_{min} Pick $v \in U_2 - U_1$ and consider s-v path in \mathscr{T}_{min}

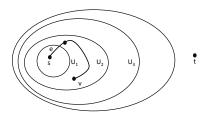


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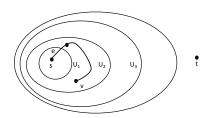


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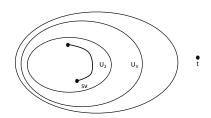


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Theorem

Best-of-many Christofides' algorithm is a deterministic ϕ -approximation algorithm for the s-t path TSP for the general metric, where $\phi=\frac{1+\sqrt{5}}{2}<1.6181$ is the golden ratio

•
$$y := \alpha \chi_{\mathscr{T}} + \beta x^* + \sum_{U:(U,\bar{U}) \text{ is } \tau-\text{narrow},|U\cap \mathcal{T}| \text{ odd}} (1 - 2\alpha - \beta x^*(\delta(U))) f_U$$

- Let $\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$ be au-narrow cuts
- Need to bound $E[c(y)] = \alpha E[c(\mathcal{T})] + \beta c(x^*) + \sum_{i=1}^{\ell-1} Pr[|U_i \cap T| \text{ odd}] (1 2\alpha \beta x^*(\delta(U_i))) c_{e(i)}$ $\leq \alpha c(x^*) + \beta c(x^*) + \sum_{i=1}^{\ell-1} (x^*(\delta(U_i)) - 1) (1 - 2\alpha - \beta x^*(\delta(U_i))) c_{e(i)}$

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- $E[c(y)] \le (\alpha + \beta)c(x^*) + \sum_{i=1}^{\ell-1} (x^*(\delta(U_i)) 1)(1 2\alpha \beta x^*(\delta(U_i)))c_{e(i)}$

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 $\leq (\alpha + \beta)c(x^*) + \max_{z:0 < z < \tau} z[1 - 2\alpha - \beta(1+z)] \sum_{i=1}^{\ell-1} c_{e(i)}$
 $\leq (\alpha + \beta + \max_{z:0 < z < \tau} z[1 - 2\alpha - \beta(1+z)])c(x^*)$
and set $\alpha = 1 - \frac{2}{\sqrt{\epsilon}}$ and $\beta = \frac{1}{\sqrt{\epsilon}}$ with $\tau = 3 - \sqrt{5}$

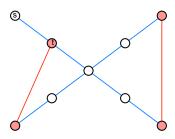
Exercise 1: Prove that Hoogeveen's algorithm for the s-t path TSP is not only a 5/3-approximation algorithm, it finds a solution of cost at most $(5/3)c(x^*)$, where x^* is the LP optimum.

Exercise 2: For any τ -narrow cut (U, \bar{U}) , $Pr[|\delta(U) \cap \mathcal{T}| = 1] \ge 2 - x^*(\delta(U))$

Theorem

Best-of-many Christofides' algorithm is a deterministic 1.6-approximation algorithm for metric s-t path TSP.

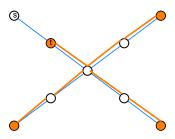
- Let \mathcal{T}_i^{st} denote the path between s and t in \mathcal{T}_i
- New Idea: [Guttmann-Beck, Hassin, Khuller, and Raghavachari, 2000] $\mathcal{T}_i \mathcal{T}_i^{st}$ is a T_i -join



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Corollary

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Proof.

$$x^*(\delta(U)) = E[|\delta(U) \cap \mathcal{T}]$$

$$\geq Pr[|\delta(U) \cap \mathcal{T}| = 1] + 2 \cdot Pr[|\delta(U) \cap \mathcal{T}| \geq 2]$$

But

$$Pr[|\delta(U) \cap \mathcal{T}| = 1] + Pr[|\delta(U) \cap \mathcal{T}| \ge 2] = 1 \Rightarrow Pr[|\delta(U) \cap \mathcal{T}| = 1] > 2 - x^*(\delta(U))$$

Lemma

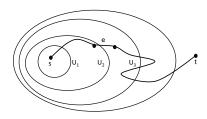
Let $\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$ be τ -narrow cuts. $E[c(\mathscr{T}^{st})] \geq \sum_{i=1}^{\ell-1} (2 - x^*(\delta(U_i))) c_{e(i)}$

Proof.

Consider each realization \mathcal{T}_i of \mathcal{T} .

Consider each U_i such that $|\delta(U_i) \cap \mathscr{T}_j| = 1$

Call this edge $\bar{e}_i \ \Rightarrow \ \bar{e}_i$ on s-t path



Lemma

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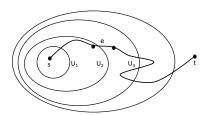
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Component of $\mathscr{T}_j - \{\bar{e}_i\}$ with s is graph induced by U_i



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Component of $\mathcal{T}_i - \{\bar{e}_i\}$ with s is graph induced by U_i

 \Rightarrow Distinct τ -narrow cuts have distinct unique edges

$$c(\mathscr{T}_j^{st}) \geq \sum_{U_i: |\delta(U_i) \cap \mathscr{T}_j| = 1} c(\bar{e}_i) \geq \sum_{U_i: |\delta(U_i) \cap \mathscr{T}_j| = 1} c(e_i)$$

Lemma follows by taking expectations and using bound on

$$Pr[|\delta(U_i) \cap \mathscr{T}| = 1].$$

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Proof.

- Let $\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$ be τ -narrow cuts
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David Shmoys

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- $E[c(y)] \le (\alpha + \beta)c(x^*) +$

$$\sum_{i=1}^{\ell-1} \frac{(x^*(\delta(U_i))-1)(1-2\alpha-\beta x^*(\delta(U_i)))}{(2-x^*(\delta(U_i)))} (2-x^*(\delta(U_i))) c_{e(i)}$$

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• Set $\alpha = 1/9$, $\beta = 4/9$, $\tau = 3/4$ with max at z = 1/2

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- Set $\alpha = 1/9$, $\beta = 4/9$, $\tau = 3/4$ with max at z = 1/2
- $E[c(y)] \leq \min \left[E[c(\mathscr{T} \mathscr{T}^{st})], \frac{5}{9}c(x^*) + \frac{1}{9}E[c(\mathscr{T}^{st})] \right]$
- Set terms = and use that $E[c(\mathcal{T} \mathcal{T}^{st})] + E[c(\mathcal{T}^{st})] = c(x^*)$

Unit-Weight Graphical Metric for s-t Path TSP

Improved algorithms for the graphical metric case

- [Oveis Gharan, Saberi, Singh 2011],
 [Mömke, Svensson 2011],
 [Mucha 2011]
- [An, Kleinberg, S 2012]Algorithmic use of τ -narrow cuts
- Methods above yield 1.5780-approximation algorithm
- [Sebő & Vygen 2012] 1.5-approximation algorithm
- [Gao 2013] simple 1.5-approximation algorithm & analysis

Gao's Algorithm for Unit-Weight Graphical Metric

- Given metric by G = (V, E)
- Include 0, 1, or 2 copies of each edge $e \in E$ to yield Eulerian graph
- Write LP based on this variation: $x_e =$ number of copies
- Partition constraints (as in spanning tree characterization)
- Non s-t cuts must have capacity at least 2
- The algorithm (just a new choice of spanning tree):
 - Solve LP and compute all narrow cuts to get layers L_i
 - Compute spanning tree in each layer
 - Compute T by adding 1 edge between L_i and L_{i+1}
 - Augment by minimum cost T-join
- Must prove support of x* always has such a spanning tree
- Analysis: $y = x^*/2$ is feasible fractional T-join

Gao's LP Relaxation

- Let \mathcal{P} be set of partitions of V
- For partition $S = (S_1, ..., S_k)$, let $\delta(S)$ be set of edges with endpoints in different parts

For
$$G = (V, E)$$
, minimize $\sum_{e \in E} x_e$ subject to

$$\begin{split} \sum_{\boldsymbol{e} \in \delta(\mathcal{S})} & x_{\boldsymbol{e}} \geq k-1, & \forall \mathcal{S} = (S_1, \dots, S_k) \in \mathcal{P} \\ & \sum_{\boldsymbol{e} \in \delta(\mathcal{S})} & x_{\boldsymbol{e}} \geq 2, & \forall \mathcal{S} : \ \emptyset \subset \mathcal{S} \subset \mathcal{V}, \ |\mathcal{S} \cap \{s,t\}| \neq 1 \\ & 0 \leq x_{\boldsymbol{e}} \leq 2, & \forall \boldsymbol{e} \in \mathcal{E} \end{split}$$

The Structure of the 1-Narrow Cuts

- Let $\{s\} = U_1 \subset U_2 \subset \cdots \subset U_\ell = V$ be 1-narrow cuts.
- Let $L_i = U_i U_{i-1}$, where $U_0 = \emptyset$.

Claim

Consider the support graph of x^* . For each $1 \le j \le k \le \ell$, the graph induced on the layers L_j, \ldots, L_k is connected.

Proof.

Suppose 1 < j and $k < \ell$ - other cases are easier.

Suppose claim is false.

Partition these layers into 2 disconnected sets S_1 and S_2

$$x^*(\delta(S_i)) \ge 2$$
, for each $i \in \{1, 2\}$

Let S_0 and S_3 denote union of lower & higher layers, resp.

These are 1-narrow cuts -
$$x^*(\delta(S_i)) < 2$$
, $i \in \{0,3\}$.

$$4 \le x^*(\delta(S_1)) + x^*(\delta(S_2)) \le x^*(\delta(S_0)) + x^*(\delta(S_3)) < 4$$

The Analysis of Gao's Algorithm

Claim

 $x^*/2$ is a fractional T-join, where T is the wrong parity of degree set for the spanning tree \mathcal{T}

Proof.

Suppose there exists a set S with $x^*(\delta(S)) < 2$ with $|S \cap T|$ odd.

S must be s-t cut. (Otherwise, LP enforces \geq 2.)

S must be 1-narrow. (Otherwise, LP again \geq 2.)

But the number of tree edges crossing each cut U_i is 1.

(By construction.)

But Key Lemma \Rightarrow odd cut has \geq 2 edges crossing it.

Some Bigger Picture Questions

- Sampling Christofides' Algorithm
 - Analysis required only that $Pr[e \in \mathcal{T}] = x_e^*$
 - [Asadpour et al.] and [Oveis Gharan et al.] focus on maximum entropy distribution:
 - for each spanning tree \mathcal{T}_i , give weight λ_i to achieve marginal distribution x^* so that λ has maximum entropy (among all such weight functions)
 - What extra power does this distribution provide?
 - Are there other more appropriate sampling approaches?
- Started with approximation algorithm + randomization ⇒ improved performance
- Are there other paradigms where this is possible?
 E.g., GRASP [Resende] one example is [Buchbinder, Feldman, Naor, Schwartz, SODA 14]

Applications & open questions

- Circuit TSP
 - Is there a better than 3/2-approximation algorithm?
 - In fact, do we already know such an algorithm for the circuit TSP (and just need the proof)?

