# The Traveling Salesman: Classical Tools and Recent Advances

Lecture 4: Symmetric s-t Path TSP

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# Symmetric s-t Path TSP



Recall: s-t Path TSP

#### Input:

- A complete undirected graph G = (V, E);
- Start vertex  $s \in V$ , end vertex  $t \in V$ ;
- Edge costs  $c(e) \equiv c(i,j) \ge 0$  for all  $e = (i,j) \in E$ ;
- Edge costs satisfy the **triangle inequality**:  $c(i,j) \le c(i,k) + c(k,j)$  for all i,j,k.

**Goal:** Find a min-cost path from *s* to *t* that visits all other vertices in between.

#### Recall: LP relaxation

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

Call  $\delta(S)$  an  $\underline{s-t}$  cut if  $s \in S, t \notin S$  (or  $s \notin S, t \in S$ ). Call  $\delta(S)$  a  $\underline{\text{non } s-t}$  cut if  $s, t \notin S$  (or  $s, t \in S$ ).

$$\text{Min } \sum_{e \in E} c(e)x(e)$$
 
$$x(\delta(i)) = \begin{cases} 1, & \forall i = s, t, \\ 2, & \forall i \neq s, t, \end{cases}$$
 
$$x(\delta(S)) \geq \begin{cases} 1, & \forall s\text{-}t \text{ cuts } \delta(S), \\ 2, & \forall \text{non } s\text{-}t \text{ cuts } \delta(S), \\ 0 \leq x(e) \leq 1, & \forall e \in E. \end{cases}$$

#### Recall: No Even Narrow Cuts - No Problem

Let  $x^*$  be an optimal solution for the s-t path TSP LP.

#### **Definition**

A cut  $\delta(S)$  is <u>narrow</u> if  $x^*(\delta(S)) < 2$ .

#### **Observation**

Let T be a spanning tree, and  $W_T = Odd_T \triangle \{s, t\}$ .

If  $c(T) \leq OPT_{LP}$ , and there is no narrow cut  $\delta(S)$  for which  $|\delta(S) \cap T|$  is even, then adding a minimum-cost  $W_T$ -matching to T gives an s-t traveling salesman path of cost at most  $\frac{3}{2}OPT_{LP}$ .

#### Recall: Narrow Cuts Are Nested

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

If  $\delta(S_1), \delta(S_2)$  are narrow cuts,  $S_1 \neq S_2$ , then either  $S_1 \subset S_2$  or  $S_2 \subset S_1$ .

So the narrow cuts look like  $s \in S_1 \subset S_2 \subset \cdots \subset S_k \subset V$ .



Each narrow cut  $\delta(S_i)$  is indicated by a gray line;  $S_i = \bigcup_{j=1}^i C_j$  is all nodes to the left of the line.

#### An idea!?

Can we just use Gao's algorithm again??!?

✓ Yes, we can find a Gao-tree  $T_{\mathrm{Gao}}$ :

$$|T_{\mathrm{Gao}} \cap \delta(S)|$$
 is odd for all cuts  $\delta(S)$  with  $x^*(\delta(S)) < 2$ .

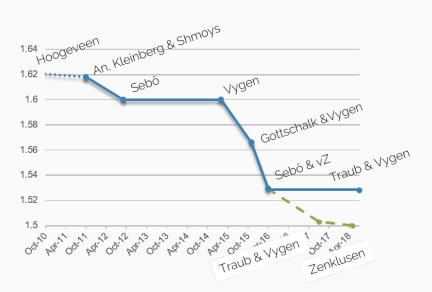
✓ That implies that the "cost of parity correction" is at most  $\frac{1}{2}OPT_{LP}$ :

$$c(M) \leq \frac{1}{2}OPT_{LP},$$

for a minimum-cost  $W_{T_{Gao}}$ -matching M.

**X** But, unfortunately, Gao (2015) shows that  $c(T_{\mathrm{Gao}}) \not \leq \mathit{OPT}_{\mathit{LP}}...$ 

## **Recent Developments**



## Many new ideas

Many interesting ideas in these recent developments:

- An, Kleinberg, Shmoys (2012): Best-of-many trees algorithm;
- Gottschalk, Vygen (2018): Choosing better trees;
- Sebő, vZ (2019): Best-of-many with deletion;
- Traub, Vygen (2019), Zenklusen (2019): Dynamic programming.

We will talk about the first three ideas in this lecture, giving high level ideas and simplified proofs. In tomorrow's lecture, we will describe the last result (giving a full analysis).

# **Best-of-Many Trees**



## Convex Combination of Spanning Trees<sup>1</sup>

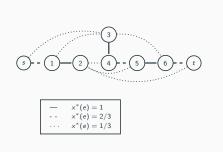
Because an optimal LP solution  $x^*$  is in the <u>spanning tree polytope</u> (feasible for the spanning tree LP), we can compute a convex combination of spanning trees

$$x^* = \sum_{i=1}^k \lambda_i \chi_{\mathcal{T}_i}.$$

$$\left(\sum_{i=1}^{k} \lambda_i = 1, \lambda_i \ge 0 \text{ for } i = 1, \dots, k.\right)$$

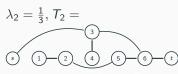
<sup>&</sup>lt;sup>1</sup>Recall: The characteristic vector of T has  $\chi_T(e) = 1$  if  $e \in T$ ,  $\chi_T(e) = 0$  if  $e \notin T$ .

## **Example: Convex Combination of Spanning Trees**



$$\lambda_1 = \frac{1}{3}, T_1 =$$

$$\begin{array}{c} & & & & \\ & & \\ & & & \\ & & \\ & & & \\ & &$$



$$\lambda_3 = \frac{1}{3}, T_3 =$$

$$\begin{array}{c} 3 \\ \hline & \\ & \\ \end{array}$$

## **Best-of-Many Algorithm**

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

$$x^* = \sum_{i=1}^k \lambda_i \chi_{\mathcal{T}_i}.$$

For each spanning tree  $T_i$ :

- Let  $W_{T_i} = Odd_{T_i} \triangle \{s, t\}$  be the set of vertices whose degree parity needs fixing.
- Let  $M_i$  be a minimum-cost  $W_{T_i}$ -matching.
- Find s-t traveling salesman path by shortcutting Eulerian path of  $(V, T_i \sqcup M_i)$ .

Return the shortest traveling salesman path found over all i.

#### **Probabilistic View**

Since the algorithm returns the best solution among the solutions based on  $T_1, \ldots, T_k$ , the cost of the algorithm's solution is at most the (weighted) average cost of these solutions.

For convenience, we view the weighted average cost as an  $\underline{\text{expected}}$   $\underline{\text{value}}$ , by considering a random spanning tree  $\mathbb{T}$  where  $P(\mathbb{T}=T_i)=\lambda_i$  for  $i=1,\ldots,k$ , and adding  $\mathbb{M}$ , a min-cost  $\mathbb{W}_{\mathbb{T}}$ -matching for the random spanning tree  $\mathbb{T}$ .

The cost of the algorithm's solution is at most

$$\mathbb{E}(c(\mathbb{T})) + \mathbb{E}(c(\mathbb{M})),$$

where  $\mathbb{E}(\cdot)$  indicates the expectation.

**Observation:**  $\mathbb{P}(e \in \mathbb{T}) = x^*(e)$ , and  $\mathbb{E}(|\mathbb{T} \cap \delta(S)|) = x^*(\delta(S))$ .

#### **Theorem**

#### **Theorem**

The Best-of-Many algorithm returns a solution of cost at most  $\frac{13}{8}$   $OPT_{LP} = 1.625 OPT_{LP}$ .

We will prove the theorem, by proving the following two lemmas.

#### Lemma (Connectivity Cost)

$$\mathbb{E}(c(\mathbb{T})) = OPT_{LP}.$$

#### **Lemma (Parity Correction Cost)**

$$\mathbb{E}(c(\mathbb{M})) \leq \frac{5}{8} OPT_{LP}.$$

## **Connectivity Cost**

$$\mathbb{E}(c(\mathbb{T})) = \mathit{OPT}_{\mathit{LP}}.$$

## **Analyzing the Parity Correction Cost**

Remember that the obstacle in our analysis is even narrow cuts, i.e., s-t cuts  $\delta(S)$  with  $x^*(\delta(S)) < 2$  and  $|T \cap \delta(S)|$  even.

#### Lemma

For a narrow cut  $\delta(S)$ ,

$$P(|\mathbb{T} \cap \delta(S)| \text{ is odd}) \geq 2 - x^*(\delta(S)).$$

## **Analyzing the Parity Correction Cost**

Let the narrow cuts be  $\delta(S_1), \delta(S_2), \ldots, \delta(S_\ell)$ .

**Approach**: Given a tree T, construct a vector  $z_T$  that is feasible for the  $W_T$ -matching LP:

$$z_T = \frac{1}{2}x^* + \text{ additional vectors},$$

one for each narrow cut  $\delta(S_j)$  such that  $|T \cap \delta(S_j)|$  is even

$$x^*(\delta(S_j)) = 5/3$$

## **Analyzing the Parity Correction Cost**

Let the narrow cuts be  $\delta(S_1), \delta(S_2), \dots, \delta(S_\ell)$ . For each narrow cut, let  $f_j = 2 - x^*(\delta(S_j))$ .

Let  $e_j$  be the cheapest edge in  $\delta(S_j)$ . For a tree T, we define

$$z_T = \frac{1}{2}x^* + \sum_{j:|T \cap \delta(S_j)| \text{ is even}} \frac{1}{2}f_j \chi_{e_j}.$$

$$x^*(\delta(S)) = 5/3$$



$$z_T(\delta(S)) = \frac{1}{2} \cdot \frac{5}{3} + \frac{1}{2}(2 - \frac{5}{3}) = 1$$
  
(if  $|T \cap \delta(S)|$  is even)

# **Proof of Second Lemma:** $\mathbb{E}(c(\mathbb{M})) \leq \frac{5}{8}OPT_{LP}$

Since

$$z_T = \frac{1}{2}x^* + \sum_{j:|T \cap \delta(S_j)| \text{ is even}} \frac{1}{2}f_j \chi_{e_j}$$

is feasible for the  $W_T$ -matching LP for any tree T, we have

$$\begin{split} \mathbb{E}(c(\mathbb{M})) & \leq \mathbb{E}(\sum_e c(e)z_{\mathbb{T}}(e)) \\ & = \frac{1}{2}\sum_{e \in E} c(e)x^*(e) + \sum_{j=1}^{\ell} \frac{1}{2}f_jc(e_j)P(|\mathbb{T} \cap \delta(S_j)| \text{ even}). \end{split}$$

We showed that  $P(|\mathbb{T} \cap \delta(S_j)| \text{ is odd}) \ge 2 - x^*(\delta(S_j)) = f_j$ , so  $\mathbb{E}(c(\mathbb{M})) \le$ 

$$\frac{1}{2} OPT_{LP} + \sum_{j=1}^{\ell} \frac{1}{2} f_j c(e_j) (1 - f_j) \leq \frac{1}{2} OPT_{LP} + \frac{1}{8} \sum_{j=1}^{\ell} c(e_j).$$

#### One Last Lemma

We need one last ingredient.

#### Lemma (Gao (2015))

$$\sum_{i=1}^{\ell} c(e_i) \leq OPT_{LP}.$$

#### Proof.

Consider the MST T. We will show that we can assign each narrow cut  $\delta(S_j)$  an edge  $e_{T,j} \in T \cap \delta(S_j)$  in such a way that no edge in T is assigned to more than one cut.

So 
$$\sum_{j=1}^{\ell} c(e_j) \leq \sum_{j=1}^{\ell} c(e_{T,j}) \leq c(T) \leq OPT_{LP}$$
.

## Assigning Edges of T to Narrow Cuts

Consider (V, T). Contract  $S_1$  to  $v_1$ ,  $V \setminus S_\ell$  to  $v_\ell$ , and  $S_j \setminus S_{j-1}$  to  $v_j$ , for  $j = 2, ..., \ell - 1$ .

Graph is connected; remove edges from  $\,T\,$  if necessary to ensure  $\,T\,$  is a spanning tree of contracted graph.

For  $j = 1, \ldots, \ell$ :

• Let  $e_{T,j}$  be the edge incident on  $v_j$  on the unique path from  $v_j$  to  $v_{j+1}$ . Remove  $e_{T,j}$  from T, and contract  $v_j, v_{j+1}$ .

## **Improved Analysis**

We showed that the solution returned by the Best-of-Many algorithm has cost at most  $\frac{13}{8}OPT_{LP}=1.625OPT_{LP}$ .

An, Kleinberg and Shmoys give a more refined analysis, showing the following result.

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

The Best-of-Many algorithm returns a solution of cost at most  $\frac{1+\sqrt{5}}{2}$   $OPT_{LP} \leq 1.618 OPT_{LP}$ .

Their analysis was further improved by Sebő:

#### Theorem (Sebő (2013))

The Best-of-Many algorithm returns a solution of cost at most  $1.60PT_{LP}$ .

# **Choosing Better Trees**



## **Reassembling Trees**

Vygen shows that exchanging edges in pairs of spanning trees of the convex combination can improve their properties under certain conditions.

#### Theorem (Vygen (2015))

The Best-of-Many algorithm "with Reassembling of Trees" returns a solution of cost at most  $1.599OPT_{LP}$ .

Analysis is complicated, but the idea of reassembly led to the next idea: a Gao-like (or layered) convex combination.

#### **Gao-like Convex Combinations**

Given a convex combination  $x^* = \sum_{i=1}^k \lambda_i \chi_{T_i}$ , and a narrow cut  $\delta(S)$ , we previously showed that

$$P(|\mathbb{T} \cap \delta(S)| \text{ is odd}) \ge \sum_{i:|T_i \cap \delta(S)|=1} \lambda_i \ge 2 - x^*(\delta(S)).$$

Call a narrow cut <u>lonely</u> in tree T if  $|T \cap \delta(S)| = 1$ . Let  $f_S = 2 - x^*(\delta(S))$ .

The above says that each narrow cut is lonely in an at least an " $f_S$  fraction" of the trees in the convex combination.

Gottschalk and Vygen showed that we can find a convex combination such that  $\delta(S)$  is lonely in the "first"  $f_S$  fraction of the trees; and that this holds simultaneously for all narrow cuts  $\delta(S)$ .

## **Layered Convex Combination**

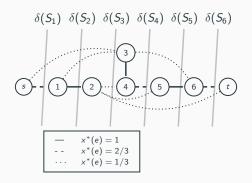
# Theorem (Gottschalk, Vygen (2018), Schalekamp, Sebő, Traub, vZ (2018))

There exists spanning trees  $T_1, \ldots, T_k$  and multipliers  $\lambda_1, \ldots, \lambda_k \geq 0$  such that

$$x^* = \sum_{i=1}^k \lambda_i \chi_{\mathcal{T}_i},$$

and for any narrow cut  $\delta(S)$ , there exists  $\ell$  such that  $|T_i \cap \delta(S)| = 1$  for  $1 \le i \le \ell$  and  $\sum_{j=1}^{\ell} \lambda_i \ge 2 - x^*(\delta(S))$ .

## **Example**



Narrow cuts  $\delta(S)$  indicated by gray lines; S is the set of vertices to the left of the line.

$$x^*(\delta(S_j)) = \frac{5}{3}$$
 for  $j = 2, 3, 4, 5$   $\rightarrow$  must be lonely in first  $2 - \frac{5}{3} = \frac{1}{3}$  fraction of trees,  $x^*(\delta(S_1)) = x^*(\delta(S_6)) = 1$   $\rightarrow$  must be lonely in first  $2 - 1 = 1$  fraction of trees.

## **Layered Convex Combination**

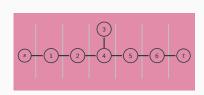
#### "Layer 1":

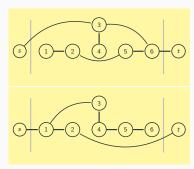
- All cuts  $\delta(S)$  with  $x^*(\delta(S)) < 2$  are lonely in trees in layer 1.
- Weight of layer 1 is  $\phi_1$ .

#### "Layer 2":

- All cuts  $\delta(S)$  with  $x^*(\delta(S)) < 2 \phi_1$  are lonely in layer 2.
- Weight of layer 2 is  $\phi_2$ .

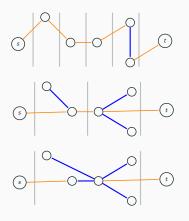
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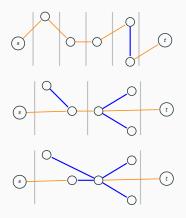
### **Layered Trees and Matroids**

Schalekamp, Sebő, Traub and vZ: Trees in a given layer are bases of a matroid:



- Spanning tree in each "level set", plus
- one edge per lonely cut.
- $\Rightarrow$  simpler proof of the theorem of Gottschalk and Vygen.
- ⇒ we can use greedy algorithm to find minimum-cost tree for each layer (instead of computing convex combination).

## Why Layered Set of Trees May Help in the Analysis



As we go down the layered set:

- Trees are less restrictive
   → tree cost decreasing;
- More narrow cuts may be even → parity correction cost increasing.

## **Best-of-Many on Layered Trees**

#### Theorem (Gottschalk, Vygen (2018))

The Best-of-Many algorithm on a Layered Set of Trees returns a solution of cost at most  $1.566OPT_{LP}$ .

# **Best-of-Many with Deletion**



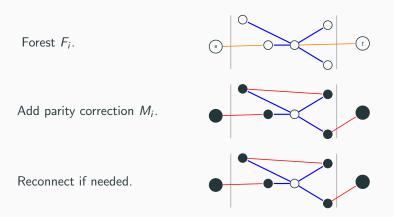
## **Best-of-Many with Deletion**

Sebő and vZ (2016) propose the <u>Best-of-Many with Deletion (BOMD)</u> algorithm: given optimal LP solution  $x^*$ , and a layered set of trees for  $x^*$ , for each spanning tree  $T_i$ :

- Delete the edges in the layer's lonely cuts to get a forest  $F_i$ .
- Let  $W_{F_i} = \text{Odd}_{F_i} \triangle \{s, t\}$  be the set of vertices whose degree parity needs fixing, and let  $M_i$  be a minimum-cost  $W_{F_i}$ -matching.
- Add doubled edges D<sub>i</sub> in lonely cuts if needed to reconnect (V, F<sub>i</sub> ⊔ M<sub>i</sub>).
- Find s-t traveling salesman path by shortcutting Eulerian path of  $(V, F_i \sqcup M_i \sqcup D_i)$ .

Return the shortest traveling salesman path found over all i.

## First Example of BOMD



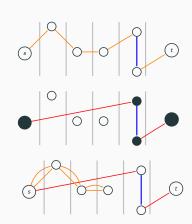
Parity correction reconnected the forest! (and we show this happens often)

## Second Example of BOMD

Forest  $F_i$ .

Add parity correction  $M_i$ .

Reconnect if needed.



## Analysis of BOMD

Since we start with a forest instead of a tree, we "save" compared to starting with a spanning tree. The analysis of the cost of parity correction is similar to before. We can prove that parity correction often reconnects the forest, so that the cost of reconnection is small on average.

#### Theorem (Sebő, vZ (2016))

The Best-of-Many with Deletion algorithm returns a solution of cost at most  $\left(\frac{3}{2}+\frac{1}{34}\right)$   $OPT_{LP}<1.5294OPT_{LP}.$ 

The analysis was improved to:

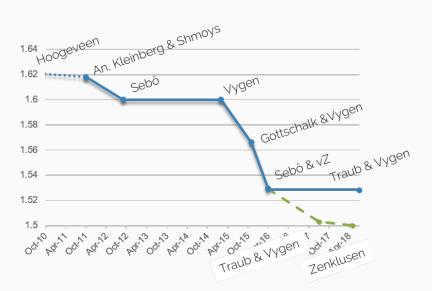
### Theorem (Traub, Vygen (2016))

The Best-of-Many with Deletion algorithm returns a solution of cost at most  $1.5284OPT_{LP}$ .

# **Summary**



## **Recent Developments**



# "ALL IDEAS GROW OUT OF OTHER IDEAS"

- ANISH KAPOOR