## A Constant-Factor Approximation Algorithm for the Asymmetric Traveling Salesman Problem

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## What's the cheapest way to visit all 24727 pubs in the UK?

 45,495,239 meters

Cook, Espinoza, Goycoolea, Helsgaun (2015)

## Find the shortest tour that visits $n$ given cities



## Traveling Salesman Problem

- Variants studied in mathematics by Hamilton and Kirkman already in the 1800's
- Benchmark problem:
- one of the most studied NP-hard optimization problems
- yet our understanding is quite incomplete


> What can be accomplished with efficient computation (approximation algorithms)?

## Two basic versions

Symmetric: distance(u,v) = distance(v,u)

2-approximation is trivial
1.5-approximation [Christofides'76] taught in undergrad courses, still unbeaten

Asymmetric: more general, no such assumption is made


## Asymmetric Traveling Salesman Problem

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Output: a minimum-weight tour that visits each vertex at least once


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Equivalently could have:



- Complete graph with $\triangle$-inequality
- Visit each vertex exactly once


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Variables: $\quad x_{u v}=$ \#times we traverse edge $(u, v)$

Minimize:
Subject to:


$$
x(\delta(S)) \geq 2
$$

$$
x \geq 0
$$

## Integrality gap of the


i.e. how far off is that particular algorithm?

## Pick any two...



## Two natural approaches: begin with...

Output: a minimum-weight consected Eutgrian multigraph

## Add Eulerian graphs until connected

$\log _{2} n$-approximation via repeated cycle covers [Frieze, Galbiati, Maffioli'82]
$0.99 \log _{2} n$-approximation
[Bläser'03]
$0.84 \log _{2} n$-approximation
[Kaplan, Lewenstein, Shafrir, Sviridenko'05]
$0.67 \log _{2} n$-approximation
[Feige, Singh'07]

## Local-Connectivity ATSP

- Defined new, easier problem
- Reduced $O$ (1)-approximation of ATSP to it
- Solved it for unweighted graphs (easy part)
[Svensson'15]
...
Solved it for graphs with two edge weights [Svensson, T., Vegh'16]


## Start with spanning tree, then make Eulerian

$O(\log n / \log \log n)$-approximation via thin trees [Asaur, Goemans, Mądrỳ, Qxeis Gharan, Saberi' 10]
$O$ (1)-approximation for planar \& bounded-genus graphs
[Oveis Gharan, Saberi'11]
Integrality gap poly $(\log \log n)$
via generalization of Kadison-Singer
[Anari, Oveis Gharan'14]

NP-hard to approximate within $1+\frac{1}{74}$
[Papadimitriou, Vempala‘00, Karpinski, Lampis, Schmied'13]
Integrality gap $=2$
[Charikar, Goemarts, Karloff'02]

## Theorem:



## Outline of reductions

## Laminarly-weighted instances



Vertebrate pairs



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Minimize: $\quad \sum_{u v \in E} w(u, v) x_{u v}$
Subject to: $x\left(\delta^{+}(v)\right)=x\left(\delta^{-}(v)\right)$ for all $v \in V$

$$
\begin{aligned}
x(\delta(S)) & \geq 2 \quad \text { for all } \mathrm{S} \subset V \\
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\end{aligned}
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1. Solve LP to obtain solution depicted in black
2. Forget edges with LP-value $=0$

- Doesn't change LP-value
- Any tour is smaller instance is a tour in original instance

3. Now all edges have positive LP-value

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## Do these edges have structure?

By complementarity slackness, each remaining edge corresponds to tight constraint in dual

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Maximize: $\sum_{s c V} 2 \cdot y_{S}$
Subject to:

$$
\begin{gathered}
\Sigma_{s:(u, v) \in \delta(S)} y_{S}+\alpha_{u}-\alpha_{v} \leq \boldsymbol{w}(u, v) \text { for all }(u, v) \in E \\
/ \mathrm{y} \geq 0
\end{gathered}
$$

Sum of $y$-values cutting ( $u, v$ ) + tail potential - head potential
is at most the edge-weight

## Dual has variables:

- $\alpha_{v}$ - vertex potential for each v
- $y_{S}$ - value for each cut $S$

Minimize: $\quad \sum_{u v \in E} w(u, v) x_{u v}$
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Dual value $=L P$-value $=22$


Maximize: $\sum_{s \subset V} 2 \cdot y_{S}$
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$$
\begin{gathered}
\sum_{s:(u, v) \in \delta(S)} y_{s}+\alpha_{u}-\alpha_{v} \leq w(u, v) \\
4+3-6=1
\end{gathered}
$$

Minimize: $\sum_{u v \in E} w(u, v) x_{u v}$
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Dual value $=L P$-value $=22$


By complementarity slackness:

$$
\sum_{s:(u, v) \in \delta(S)} y_{S}+\alpha_{u}-\alpha_{v}=w(u, v)
$$

for every edge ( $u, v$ ) (since we only kept positive edges)

Minimize: $\quad \sum_{u v \in E} w(u, v) x_{u v}$
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\sum_{s:(u, v) \in \delta(S)} y_{S}=w(u, v)-\alpha_{u}+\alpha_{v}
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\sum_{s:(u, v) \in \delta(S)} y_{S}=w(u, v)-\alpha_{u}+\alpha_{v}=: w^{\prime}(u, v)
$$

for every edge ( $u, v$ ) (since we only kept positive edges)

## Observation:

For any Eulerian edge set F

$$
w(F)=w^{\prime}(F)
$$

$$
\text { (A) } \quad \begin{aligned}
w^{\prime}(F) & =w(A, B)+\alpha_{A}-\alpha_{B} \\
& +w(B, C)+\alpha_{B}-\alpha_{C} \\
& +w(C, A)+\alpha_{c}-\alpha_{A} \\
& =w(F)
\end{aligned}
$$

Minimize: $\quad \sum_{u v \in E} w(u, v) x_{u v}$
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## Observation:

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w(F)=w^{\prime}(F)
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Thus equivalent to consider weight function $w^{\prime}$ :

$$
w^{\prime}(u, v)=\sum_{s:(u, v) \in \delta(S)} y_{S}
$$

So normalize and forget about vertex potentials

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Maximize: $\sum_{S \subset V} 2 \cdot y_{S}$
Subject to:

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\begin{gathered}
\sum_{S:(u, v) \in \delta(S)} y_{\boldsymbol{S}}+\alpha_{u}-\alpha_{v} \leq \boldsymbol{w}(u, v) \text { for all }(u, v) \in E \\
y \geq 0
\end{gathered}
$$

## What happened?

Something complicated with no structure


A lot of structure:

$$
w(e)=\sum_{s:(u, v) \in \delta(S)} y_{S}
$$

1. Drop 0-edges
2. Complementarity slackness
3. Normalize with vertex potentials

Minimize: $\sum_{u v \in E} w(u, v) x_{u v}$
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## Subject to:

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\Sigma_{S:(u, v) \in \delta(S)} \boldsymbol{y}_{S}+\boldsymbol{\alpha}_{u}-\boldsymbol{\alpha}_{v} \leq \boldsymbol{w}(u, v) \text { for all }(u, v) \in E
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$$
y \geq 0
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Let $\mathcal{L}=\left\{S: y_{S}>0\right\}$ be support of dual solution

Minimize: $\quad \sum_{u v \in E} w(u, v) x_{u v}$
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Let $\mathcal{L}=\left\{S: y_{S}>0\right\}$ be support of dual solution

Again by complementarity slackness

$$
x(\delta(S))=2 \text { for every } S \in \mathcal{L}
$$

So every $S \in \mathcal{L}$ is a tight set!

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x(\delta(S))=2 \text { for every } S \in \mathcal{L}
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So every $S \in \mathcal{L}$ is a tight set!

By "standard" uncrossing techniques:
$\mathcal{L}$ is a laminar family
Any two sets are either disjoint or one is a subset of the other


No two sets intersect non-trivially

## Laminarly-weighted

Laminarly-weighted instance $\mathcal{J}=(G, \mathcal{L}, x, y)$ :

- $x, y$ primal and dual solutions (that will be optimal by definition)


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- $\mathcal{L}=\left\{S: y_{S}>0\right\}$ is a laminar family of tight sets (LP says that we should visit each such set once)
- weights induced by $\mathcal{L}$ and $y$ :

$$
w(e)=\sum_{S \in \mathcal{L}: e \in \delta(S)} y_{S} \quad \text { for every edge } e
$$

Held-Karp lower bound $=\mathrm{OPT}=2 \cdot \sum_{S \in \mathcal{L}} y_{S} \quad$ (=28 in example)

## Theorem:

A $\rho$-approximation algorithm for laminarly-weighted instances yields a $\rho$-approximation algorithm for general ATSP


## Reduced our task to:



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## Let's take a detour

## Repeated cycle cover

[Frieze, Galbiati, and Maffioli'82]

Find min-cost cycle cover
"Contract"
Repeat until graph is connected


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## Repeated cycle cover

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Find min-cost cycle cover
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Worst case: all cycles have length 2 so need to repeat $\log _{2} n$ times (each time cost $O P T_{L P}$ )


Cost of cycle cover $\leq \boldsymbol{O P T}$
Cost of cycle cover $\leq$ OPT
Cost of cycle cover $\leq \boldsymbol{O P T}$

Total cost $\leq \mathbf{3} \cdot \boldsymbol{O P T}$

## $\log _{2} n$-approximation

## Recursive algorithm fine if value drops

Each time we take a cycle cover we make some progress

What if the value of OPT drops by say a factor 9/10 each time?

Then total cost would be

$$
\sum_{i=0}^{\log _{2} n}\left(\frac{9}{10}\right)^{i} O P T \leq \sum_{i=0}^{\infty}\left(\frac{9}{10}\right)^{i} O P T=10 \cdot O P T
$$



No one has been able to pursue this strategy with cycle cover approach
We pursue it using the structure of laminarly-weighted instances

## Le retour

## Laminarly-weighted

Laminarly-weighted instance $\mathcal{J}=(G, \mathcal{L}, x, y)$ :


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Held-Karp lower bound $=$ OPT $=2 \cdot \sum_{S \in \mathcal{L}} \boldsymbol{y}_{\boldsymbol{S}} \quad$ (=28 in example)

## Contraction and lift

## Contraction of tight sets in $\mathcal{L}$



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Remains to specify $y$-value of new vertex/set

## Lifting a tour in the contracted instance



## Lifting a tour in the contracted instance



Lift tour in contracted instance to subtour in original instance

## Lifting a tour in the contracted instance


$\because$
What to do?
Simply add a shortest path

Lift tour in contracted instance to subtour in original instance

maximuneost over all possible ways to enter and exit the original set



Set y-value of new set to pay for maximum cost over all possible ways to enter and exit the original set

In example:

$$
?=5+2+2+1+4+3=17 \quad \text { (path crosses every tight set) }
$$



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In example:

$$
?=5+2+2+1+4+3=17 \quad \text { (path crosses every tight set) }
$$

Fact: No matter how we enter and exit, there exists a path that enters and exits each set at most once => contraction does not increase LP-value

Generalization of the fact: if there is a path from $u$ to $v$ then there is one without cycles


Change of cost in example:


## By design:

Fact: Lift no more expensive than tour in contracted instance

## Facts about contraction



Fact: No matter how we enter and exit, there exists a path that enters and exits each set at most once => contraction does not increase LP-value

Fact: Lift no more expensive than tour in contracted instance


Lift is a subtour but may not be a tour:
it visits all vertices outside contracted set but not inside

However, if contraction causes significant decrease in value, then we can use remaining budget to complete the lift into tour

## Implementing recursive strategy

## (Ir)reducible sets in $\mathcal{L}$

DEF: A set $S \in \mathcal{L}$ is reducible if worst way to enter/exit crosses at most a weighted $\frac{3}{4}$ fraction of the sets strictly inside $S$


Total value inside $S=2+2+1+4+3=12$
So worst way to enter/exit should cross sets of value at most 9 to be reducible

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Worst way to enter/exit crosses sets of value $=12$
IRREDUCIBLE

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DEF: A set $S \in \mathcal{L}$ is reducible if worst way to enter/exit crosses at most a weighted $\frac{3}{4}$ fraction of the sets strictly inside $S$

We say that an instance is irreducible if no set in $\mathcal{L}$ is reducible


Total value inside $S=2+2+1+4+3=12$
So worst way to enter/exit should cross sets of value at most 9 to be reducible

Worst way to enter/exit crosses sets of value $=9$

## Theorem:

A $\rho$-approximation algorithm for irreducible instances yields a $\mathbf{8} \rho$-approximation algorithm for laminarly-weighted instances, and thus for general ATSP

Let $\mathcal{A}$ be a $\rho$-approximation algorithm for irreducible instances...

## Alg for reducible instances

If instance is irreducible, simply run $\mathcal{A}$
Otherwise select minimal reducible set $S \in \mathcal{L}$
Recursively find tour $T$ in instance with $S$ contracted
Complete lift of $T$ to a tour in original instance using $\mathcal{A}$


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Recursive call returns $8 \rho$-approximate solution $T$ on smaller instance:

$$
w(T) \leq 8 \rho\left(O P T-\frac{1}{4}\left(2 \cdot \sum_{R \in \mathcal{L : R} \subset S} y_{R}\right)\right)=8 \rho O P T-2 \rho\left(2 \cdot \sum_{R \in \mathcal{L : R} \subset S} y_{R}\right)
$$

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$$

Remaining task: complete lift to a tour using $\mathcal{A}$ while paying at most the above

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Task: complete to tour while paying at most $2 \rho\left(2 \cdot \sum_{R \in \mathcal{L}: R \subset S} y_{R}\right)$

- We need to only connect unvisited vertices inside $S$


## Simplifying assumption:

instance obtained by restricting to vertices inside $S$ is feasible

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An irreducible instance since $S$ was a minimal reducible set

Held-Karp value $=2$ times dual values

$$
=2 \cdot \sum_{R \in \mathcal{L}: R \subset S} y_{R}
$$

Solve this instance with $\mathcal{A}$ to find tour on $S$ of weight

$$
\leq \rho \cdot\left(2 \cdot \sum_{R \in \mathcal{L}: R \subset S} y_{R}\right)
$$

Better by a factor 2 than needed


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Contract and recursively find lift (subtour) of weight

$$
\leq 8 \rho O P T-2 \rho\left(2 \cdot \sum_{R \in \mathcal{L : R \subset S}} y_{R}\right)
$$

Under simplifying assumption, find tour on $S$ of weight

$$
\leq \rho\left(2 \cdot \sum_{R \in \mathcal{L : R} \subset S} y_{R}\right)
$$

Final tour has value at most

$$
\leq 8 \rho O P T-\rho\left(2 \cdot \sum_{R \in \mathcal{L}: R \subset S} y_{R}\right)
$$

Simplifying assumption not true in general:
We define the operation of inducing on $S$ for ATSP in paper. Makes us lose another factor of 2

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$$
\leq 8 \rho O P T-2 \rho\left(2 \cdot \sum_{R \in \mathcal{C : R} \subset S} y_{R}\right)
$$

## Eulerian set of edges

Under simplifying assumption, find tour on $S$ of weight

$$
\leq \rho\left(2 \cdot \sum_{R \in L: R \subset S} y_{R}\right) * 2
$$

Final tour has value at most

$$
\leq 8 \rho O P T
$$

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## Theorem:

A $\rho$-approximation algorithm for irreducible instances yields a 8p-approximation algorithm for laminarly-weighted instances, and thus for general ATSP


## Simplifying assumptions

- $\mathcal{L}$ contains all singletons (every vertex has a node-weight)
- The instance is perfectly irreducible:
the contraction of any set causes no decrease in LP-value


When contracting a set, the LP-decrease is proportional to \#sets not crossed by path in worst way to enter/exit

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## Alg for perfect irreducible

Contract all maximal sets in $\mathcal{L}$
Resulting instance is node-weighted, use Svensson'15 to obtain a 28-approximate tour

Obtain lift of tour and rewire first visit so as to make sure to visit worst enter/exit path


Node-weighted instance Use 28-approximation by Ola

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Cost of tour:

$$
\begin{aligned}
& w(\text { lift })+w(\text { paths }) \\
& w(\text { lift }) \leq 28 \cdot O P T
\end{aligned}
$$

We add 3 paths per maximal set Cost of each path bounded by the LP-value inside that set

$$
w(\text { paths }) \leq 3 \cdot O P T
$$

Total cost $\leq 31 \cdot$ OPT

In general not perfect irreducibility:
Worst enter/exit path only crosses most sets in $\mathcal{L}$

We further reduce to the case when we are given subtour $B$ such that:

- $w(B) \leq 31 \cdot O P T$
- $\quad B$ crosses all non-singleton sets of $\mathcal{L}$
(to get this, we contract the sets it doesn't cross, and solve them recursively; it's okay because there are few)

$B$ is called the backbone and together with the instance they form a vertebrate pair



## Vertebrate pairs

Vertebrate pair ( $\mathcal{J}, B$ )

- J $=(G, \mathcal{L}, x, y)$ instance
- $B$ : backbone = subtour that crosses every non-singleton set in $\mathcal{L}$



## Vertebrate pairs

- We have reduced general ATSP to solving ATSP for a vertebrate pair $(\mathcal{J}, B)$ with $w(B)=\Theta(O P T)$
- We want to solve Local-Connectivity ATSP on such instances and apply the reduction by (Svensson 2015)



## Local-Connectivity ATSP (Svensson 2015)

Instance $\mathcal{J}=(G, \mathcal{L}, x, y)$ with induced weights $w: E \rightarrow \mathbb{R}_{+}$ Lower bound function $\mathrm{lb}: V \rightarrow \mathbb{R}_{+}$with $\sum_{v \in V} \mathrm{lb}(v)=O P T$ Input: partition of the vertex set $V=V_{1} \cup V_{2} \cup \cdots \cup V_{k}$

$V_{1}$


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Output: subtour $F$ that crosses each $V_{i}$
$\alpha$-light algorithm: for every component $C$ of $F$,

$$
w(E(C)) \leq \alpha \operatorname{lb}(V(C))
$$


"Every component locally pays for itself"

## Local-Connectivity ATSP (Svensson 2015)


(Svensson 2015)
27-approximation for node-weighted ATSP

Want:
$\mathrm{O}(1)$-light algorithm for vertebrate instances

## Local-Connectivity ATSP: node-weighted case

- Instance $\mathcal{J}=(G, \mathcal{L}, x, y)$, with $\mathcal{L}$ containing only singletons (ignore $B$ )

$$
w(u, v)=y_{\{u\}}+y_{\{v\}}
$$

- Define $\operatorname{lb}(u)=2 y_{\{u\}} \forall u \in V$
- Partition $V=V_{1} \cup V_{2} \cup \cdots \cup V_{k}$
- Modify $G$ and $x$, and solve an integer circulation problem



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- For each $V_{i}$, create auxiliary vertex $a_{i}$
- Reroute 1 fractional unit of incoming and outgoing flow $x$ to $a_{i}$
- Solve integer circulation problem routing $=1$ unit through each $a_{i}$ (and $\leq 1$ unit through each $v$ with $y_{v}>0$ )
- Map back to original $G$


## Local-Connectivity ATSP: node-weighted case

- The rerouted $x$ is feasible for the circulation problem, of weight $O P T$



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## Local-Connectivity ATSP: node-weighted case

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- Flow integrality: there exists also integer solution of weight $\leq O P T$
- After mapping back, every vertex (with $y_{v}>0$ ) has in-degree $\leq 2$
- For a component $C, w(E(C))=\sum_{(u, v) \in E(C)} y_{\{u\}}+y_{\{v\}} \leq 4 \sum_{v \in C} y_{\{v\}}$



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- Flow integrality: there exists also integer solution of weight $\leq O P T$
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- For a component $C, w(E(C))=\sum_{(u, v) \in E(C)} y_{\{u\}}+y_{\{v\}} \leq 4 \sum_{v \in C} y_{\{v\}}$
- $\operatorname{lb}(V(C))=2 \sum_{v \in C} y_{\{\mathrm{v}\}} \Rightarrow 2$-light algorithm



## Local-Connectivity ATSP: one non-singleton set in $\mathcal{L}$

- Vertebrate pair (J, B).

Assume $\mathcal{L}$ has a single non-singleton component $S$.
Thus,

$$
w(u, v)= \begin{cases}y_{\{u\}}+y_{\{v\}}+y_{S} & \text { if }(u, v) \in \delta(S) \\ y_{\{u\}}+y_{\{v\}} & \text { if }(u, v) \notin \delta(S)\end{cases}
$$

- Define $\mathrm{lb}(u)=2 y_{\{u\}}$ as before, but on one backbone vertex $u \in V(B)$ put $\operatorname{lb}(u)=w(B)$ instead
- $\sum_{v \in V} \operatorname{lb}(v)=\Theta(O P T)$, since $w(B)=\Theta(O P T)$


## Local-Connectivity ATSP: one non-singleton set in $\mathcal{L}$

- By assumption, $x\left(\delta^{\text {in }}(S)\right)=x\left(\delta^{\text {out }}(S)\right)$
- Backbone property: there is a node $s \in V(B) \cap S$
- Flow argument: we can route the incoming 1 unit of flow of $S$ to $s$ (within $x$ )



## Local-Connectivity ATSP: one non-singleton set in $\mathcal{L}$

- Partition $V=V_{1} \cup V_{2} \cup \cdots \cup V_{k}$
- Add backbone $B$ as initial content of the Eulerian output set $F$
- Via flow splitting [Svensson, T., Vegh'16] "force" all edges entering $S$ to proceed to $s \in V(B)$
- Create auxiliary vertices $a_{i}$ as before
- Solve integral circulation problem, and add solution to $F$


## Local-Connectivity ATSP: one non-singleton set in $\mathcal{L}$

## Analysis

- For all components $C$ not crossing $S$, $w(E(C)) \leq 2 \mathrm{lb}(V(C))$ exactly as in the node-weighted case
- Giant component $C_{0}$ containing $B$ :
- Contains all edges in $F$ crossing $S$
- Has lower bound $\operatorname{lb}\left(V\left(C_{0}\right)\right) \geq \operatorname{lb}(u)=\Theta(O P T)$
- $w\left(E\left(C_{0}\right)\right) \leq w(F) \leq O(O P T)$
- Therefore solution is $O(1)$-light.

- Same approach extends to arbitrary $\mathcal{L}$ : enforce that every subtour crossing a non-singleton set in $\mathcal{L}$ must intersect the backbone.

Summary and open problems...


## Theorem:

A $O(1)$-approximation algorithm with respect to Held-Karp relaxation

## Sequence of reductions




Laminarly-weighted instances


Irreducible instances behave like node-weighted


Complete backbone to tour using circulations and Local-Connectivity ATSP


## Open questions

- Is the right ratio 2 ?
- $\quad$ Unoptimized constant $=5500$
- By optimizing our approach, we believe we can get an upper bound in the hundreds. New ideas are needed to get close to lower bound of 2
- Bottleneck ATSP: find tour with minimum max-weight edge
- Thin tree conjecture: Is there a tree $T$ such that for every $S \subset V$

$$
|\delta(S) \cap T| \leq O(1) x(\delta(S))
$$

(would also imply apx for Bottleneck ATSP [An, Kleinberg, Shmoys'10])

