

A constant-factor approximation algorithm for the asymmetric travelling salesman problem

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Joint work with

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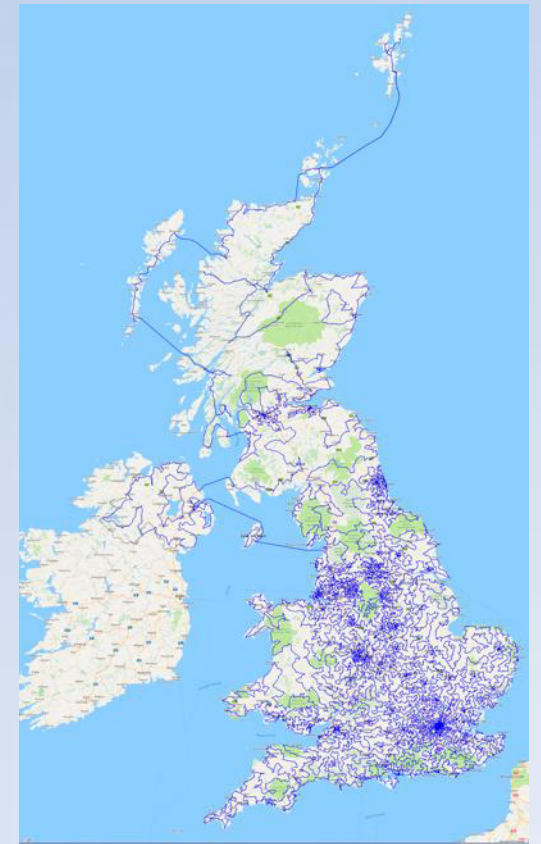
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Travelling salesman problem

Given n cities and their pairwise distances, find a shortest tour visiting all n cities.

- One of the best known **NP-hard** optimization problems
- Studied since the 19th century
- **Symmetric TSP:** $(i, j) = (j, i)$, $d_{ij} = d_{ji}$
- **Asymmetric TSP:** $(i, j) \neq (j, i)$, $d_{ij} \neq d_{ji}$ is possible



UK pub tour
[Cook et al., 2015]

Triangle inequality:

$$d_{ij} \leq d_{ik} + d_{kj}$$

Symmetric vs Asymmetric TSP

Symmetric TSP

- 1.5-approximation algorithm [Christofides '76]
- **Graphic TSP**: unweighted graph shortest path metric
 - Current best 1.4 [Sebő & Vygen '14], following
 - [Oveis Gharan, Saberi & Singh '11]
 - [Mömke & Svensson '11]
 - [Mucha '12]

Symmetric vs Asymmetric TSP

Asymmetric TSP

- **\log** -approximation algorithm [Frieze, Galbiati & Maffioli '82]
- **$0.99 \log$** [Bläser '03]
- **$0.84 \log$** [Kaplan, Lewenstein, Shafrir & Sviridenko '03]
- **$0.67 \log$** [Feige & Singh '07]
- **$\left(\frac{\log}{\log \log}\right)$** [Asadpour, Goemans, Mądry, Oveis Gharan & Saberi '10]
via thin trees.

Asymmetric TSP – recent developments

- **loglog** bound on integrality gap of LP
[Anari & Oveis Gharan '15]

Constant-factor approximations:

- Bounded genus graphs [Oveis Gharan & Saberi '11]
- Node-weighted graphs [Svensson '15]
- Graphs with 2 edge weights [Svensson, Tarnawski & V. '16]

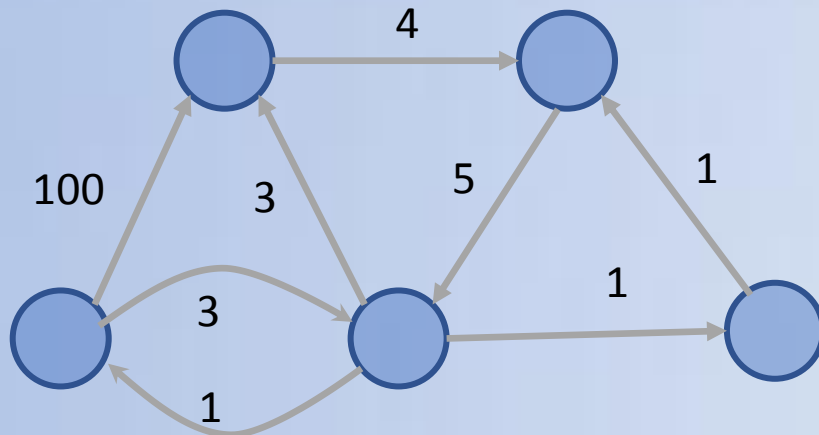
Our result: constant-factor approximation for general ATSP
with respect to the Held-Karp relaxation.

ATSP – Graphic formulation

Input: directed graph $G = (V, E)$, edge weights $w: E \rightarrow \mathbb{R}^+$
 Find a minimum weight **tour**.

- **Tour** = closed walk visiting every vertex at least once =
 = Eulerian and connected edge multiset
- **Eulerian**: $\text{In-degree}(v) = \text{Out-degree}(v)$
- **Subtour** = closed walk (not necessarily connected)

In-degree & out-degree in F

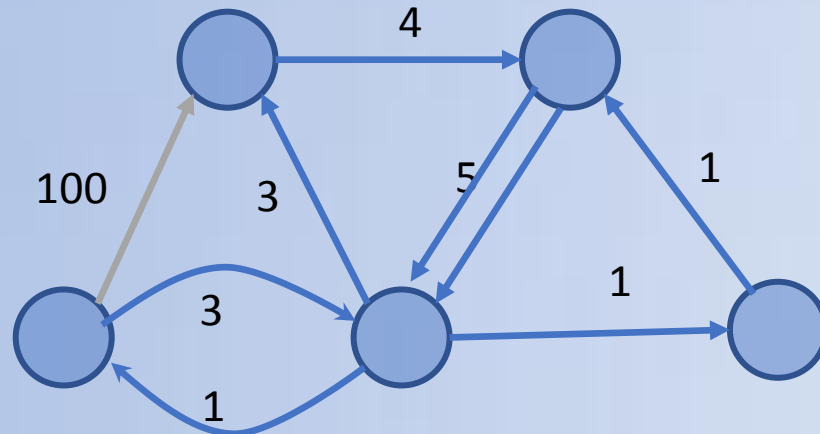


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In-degree & out-degree in F



Held-Karp relaxation

- Input: $G = (V, E)$, edge weights $w: E \rightarrow \mathbb{R}^+$.
- Variables $x: E \rightarrow \mathbb{R}^+$: multiplicity of selecting edge e .

minimize

subject to $\sum_{e \in E} x_e \cdot \begin{pmatrix} \text{deg}(v) \\ \text{deg}(w) \end{pmatrix} = \begin{pmatrix} \text{deg}(v) \\ \text{deg}(w) \end{pmatrix}$, /

Eulerian degree constraints

Subtour elimination constraints

Undirected degree:
 $\text{deg}(v) = \sum_{e \in E} x_e$

Held-Karp relaxation

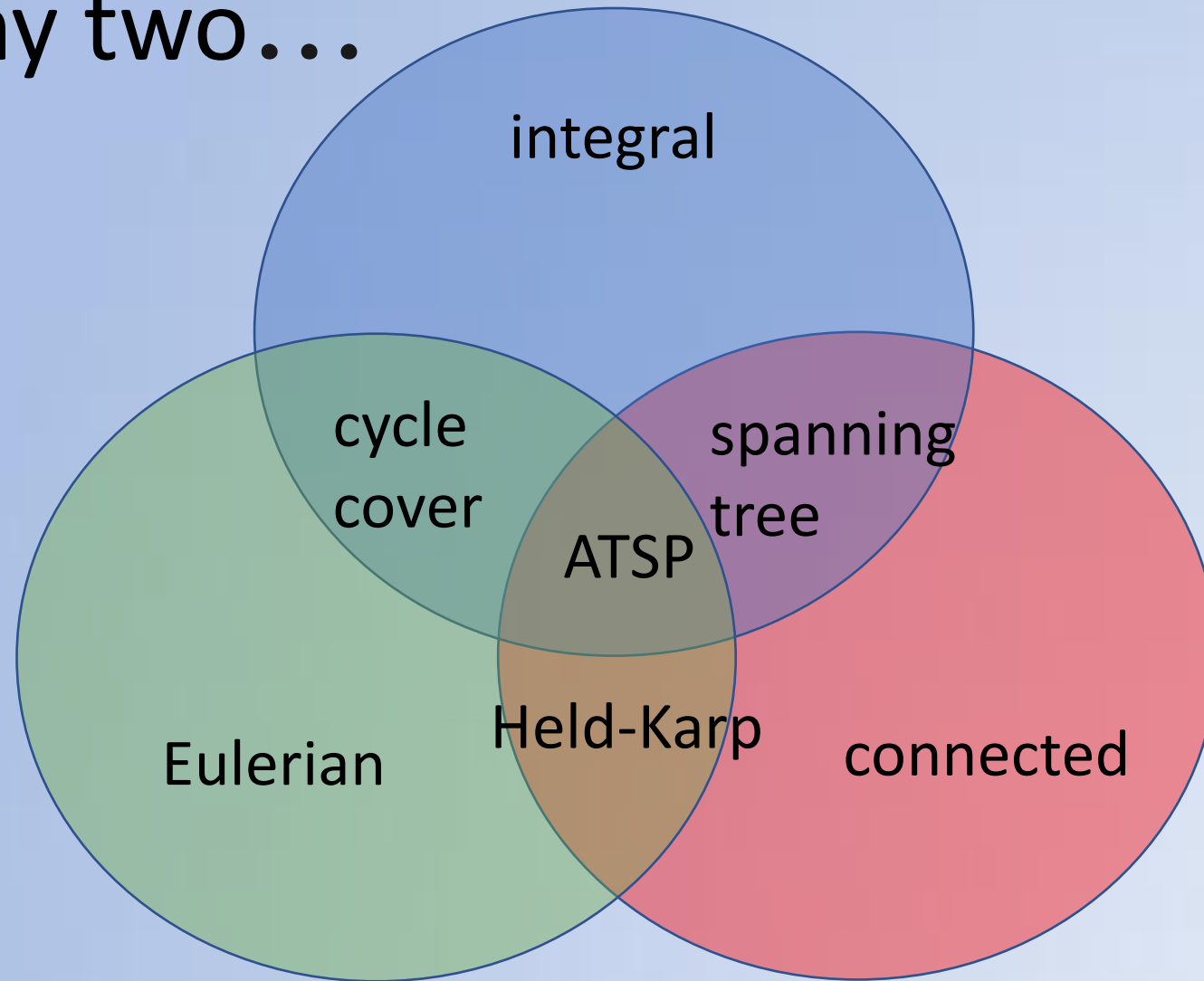
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minimize

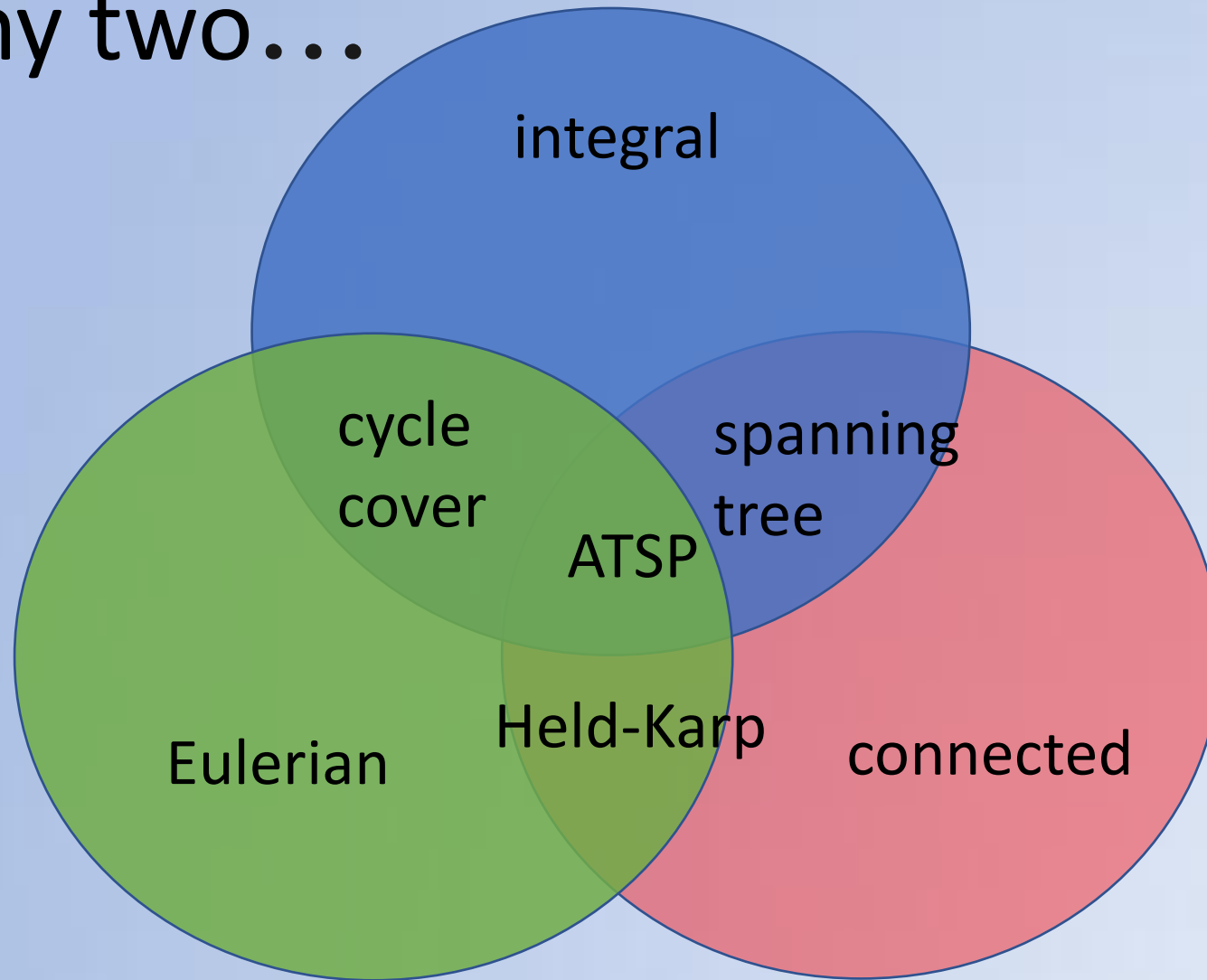
subject to $\sum_{e \in E} x_e \cdot \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} = \sum_{e \in E} x_e \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$,
 $x_e \leq 1$

- Can be solved in polynomial time
- Integrality gap [Charikar, Goemans & Karloff '06]

Pick any two...



Pick any two...



Repeated cycle cover algorithm [Frieze, Galbiati & Maffioli '82]

Relaxing connectivity:

1. Find minimum weight cycle cover
2. Contract and repeat

- Each cycle cover has cost
- Overall **log** rounds
- **log** approximation

Node-weighted case [Svensson'15]

Directed graph $G = (V, E)$, node weights $w: V \rightarrow \mathbb{R}^+$
 $(G, w) = (G_1, w_1) + (G_2, w_2)$

Local-Connectivity ATSP: relaxing connectivity constraints to local

-light algorithm for
Local-Connectivity ATSP



(G, w) -approximation
for ATSP

Theorem [Svensson'15]

There exists a polytime $(1 + \epsilon)$ -
approximation for node-weighted ATSP.

Roadmap



General ATSP

LP duality +
uncrossing

Laminarily
weighted ATSP

Irreducible
instances

Node weighted algorithm
+ contractions

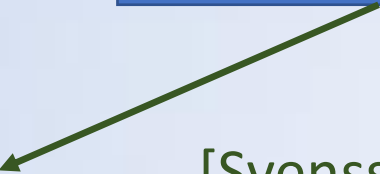
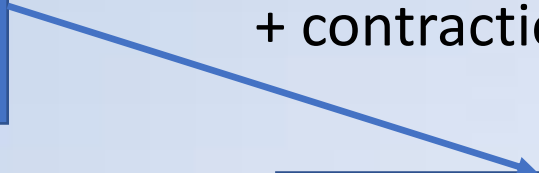
Vertebrate pairs

$O(1)$ -light lCATSP
algorithm in
vertebrate pairs

Local-connectivity
ATSP

[Svensson '15]

Graph theory:
contractions



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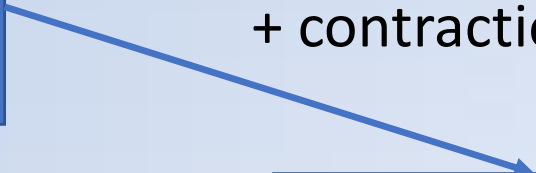
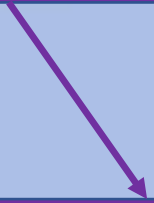
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Dual of the Held-Karp relaxation

minimize

subject to

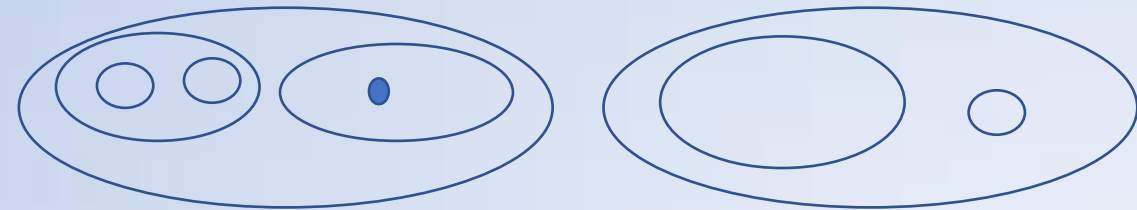
$$\begin{pmatrix} () \\ () \end{pmatrix} = \begin{pmatrix} () \\ () \end{pmatrix}$$

maximize

subject to

$$:(,) \quad + \quad - \quad (,) \quad /$$

- Dual can be solved in polynomial time.
- One can efficiently find an optimal $(,)$ such that the support of is a **laminar family** of sets.
Efficient uncrossing [Karzanov'96]



Laminarly weighted ATSP: $=$ / / /

minimize

subject to

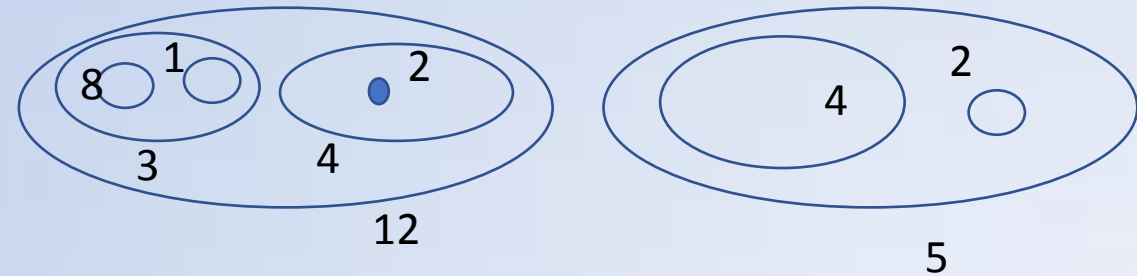
$$\begin{pmatrix} () \\ () \end{pmatrix} = \begin{pmatrix} () \\ () \end{pmatrix}$$

maximize

subject to

$$:(,) \quad + \quad - \quad (,) \quad /$$

- G : directed graph
- \mathcal{L} : laminar family of sets
- x : feasible Held-Karp solution
tight on every set in \mathcal{L} : $\begin{pmatrix} () \\ () \end{pmatrix} =$
- z : $z = x +$



Laminarly weighted ATSP: $=$ / / /

minimize

subject to

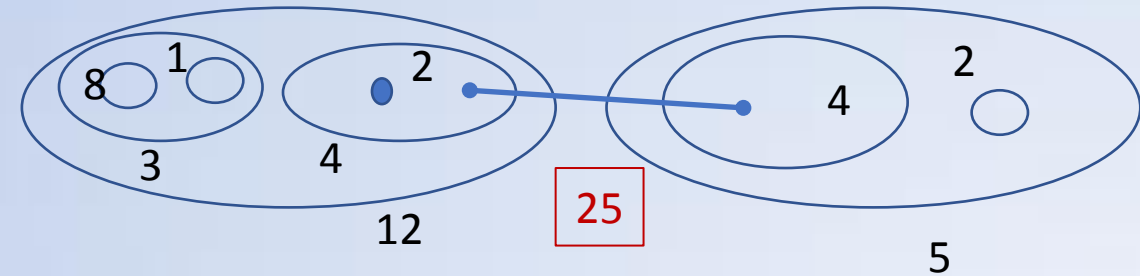
$$\begin{pmatrix} () \\ () \end{pmatrix} = \begin{pmatrix} () \\ () \end{pmatrix}$$

maximize

subject to

$$:(,) + - (,) /$$

- G : directed graph
- \mathcal{L} : laminar family of sets
- H : feasible Held-Karp solution
tight on every set in \mathcal{L} : $\begin{pmatrix} () \\ () \end{pmatrix} =$



- W : $+$

Induced weight function: $(,) = :(,)$

Laminarly weighted ATSP: $=$ / / /

minimize

subject to

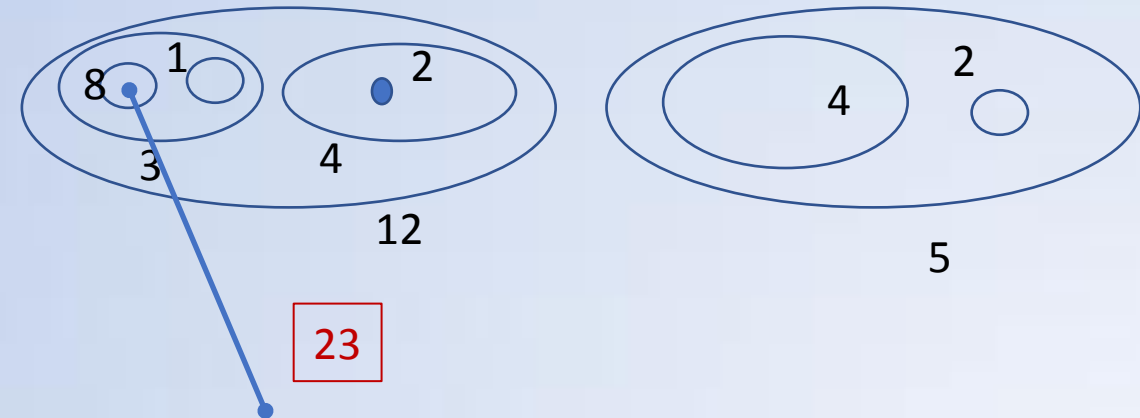
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maximize

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tight on every set in \mathcal{L} : $\begin{pmatrix} () \\ () \end{pmatrix} =$



- w : $+$

Induced weight function: $(,) = :(,)$

Reduction to laminarly weighted ATSP

- Start with any G and λ .
- Compute Held-Karp optimal solution π and dual μ supported on laminar family
- Delete all edges with $w_{ij} - \mu_i - \mu_j < 0$.

maximize

subject to

$$w_{ij} - \mu_i - \mu_j \geq 0$$

Observations:

- Optimal solutions and optimum value are the same for G and for G'

$$w_{ij} - \mu_i - \mu_j = w_{ij} - \mu_i - \mu_j + \lambda_{ij}$$

- All remaining edges have $w_{ij} - \mu_i - \mu_j \geq 0$

$$w_{ij} - \mu_i - \mu_j = \lambda_{ij}$$

Roadmap



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LP duality +
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Laminarily
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Irreducible
instances

Node weighted algorithm
+ contractions

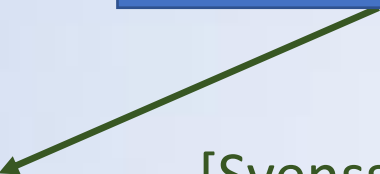
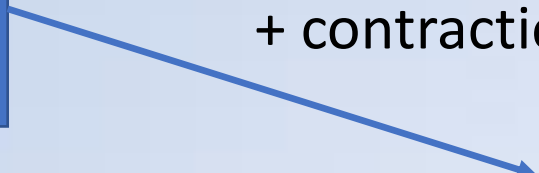
Vertebrate pairs

$O(1)$ -light
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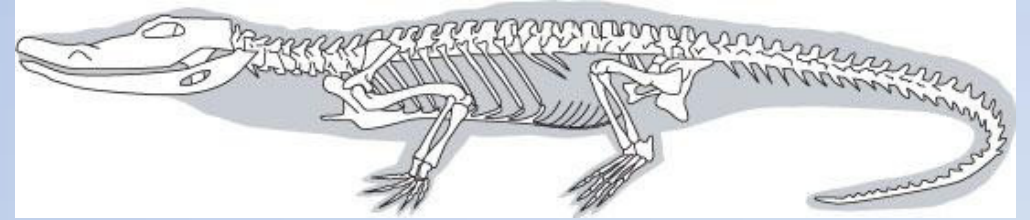
Local-connectivity
ATSP

[Svensson '15]

Graph theory:
contractions

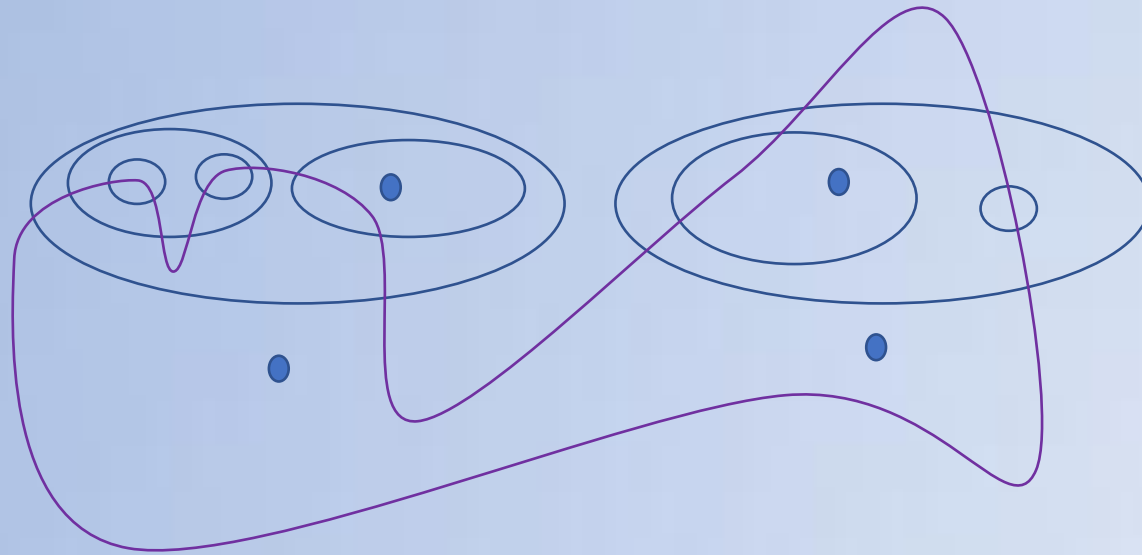


Vertebrate pairs



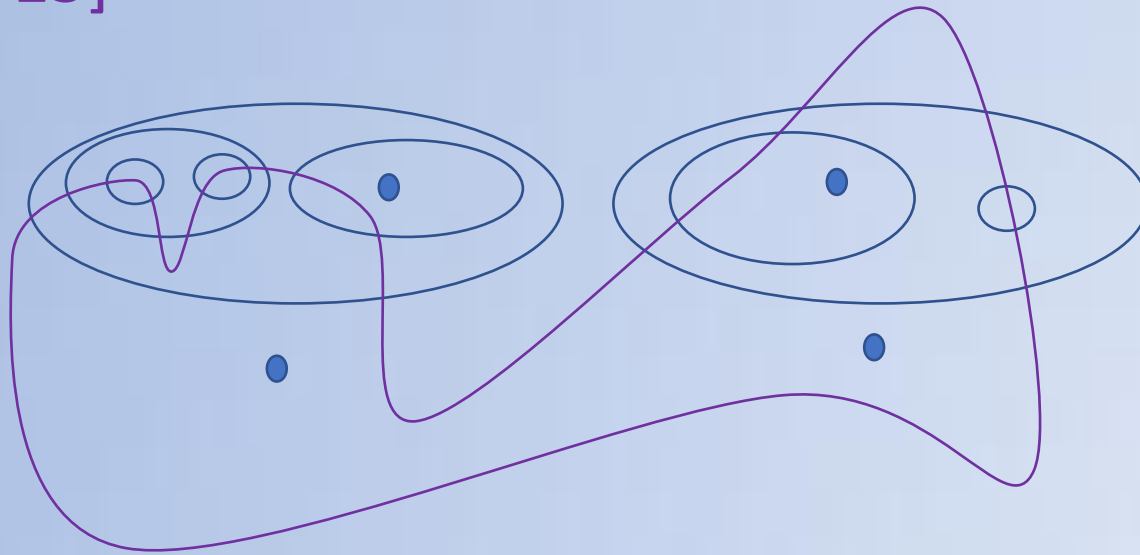
Vertebrate pair $(,)$

- $= (, , ,)$ instance
- : backbone = subtour that crosses every nonsingleton set in



Vertebrate pairs

- We will reduce general ATSP to solving ATSP for a vertebrate pair $(\mathcal{V}, \mathcal{V}')$ with $|\mathcal{V}'| = \Theta(|\mathcal{V}|)$ (more or less..)
- Solve Local-Connectivity ATSP on such instances, and apply [Svensson'15]

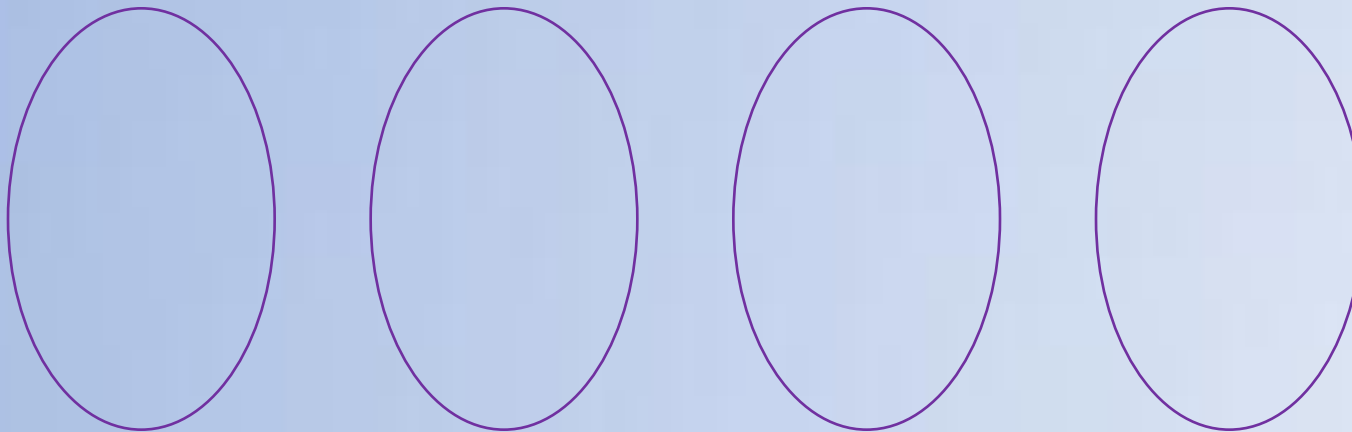


Local-Connectivity ATSP [Svensson'15]

Instance $(G, \{C_i\}, \{w_{ij}\})$ with induced weights $w_{ij} = w_{ij} +$

Lower bound function \mathbf{lb}_α with $\mathbf{lb} =$

Input: partition of the vertex set $\mathcal{C} =$



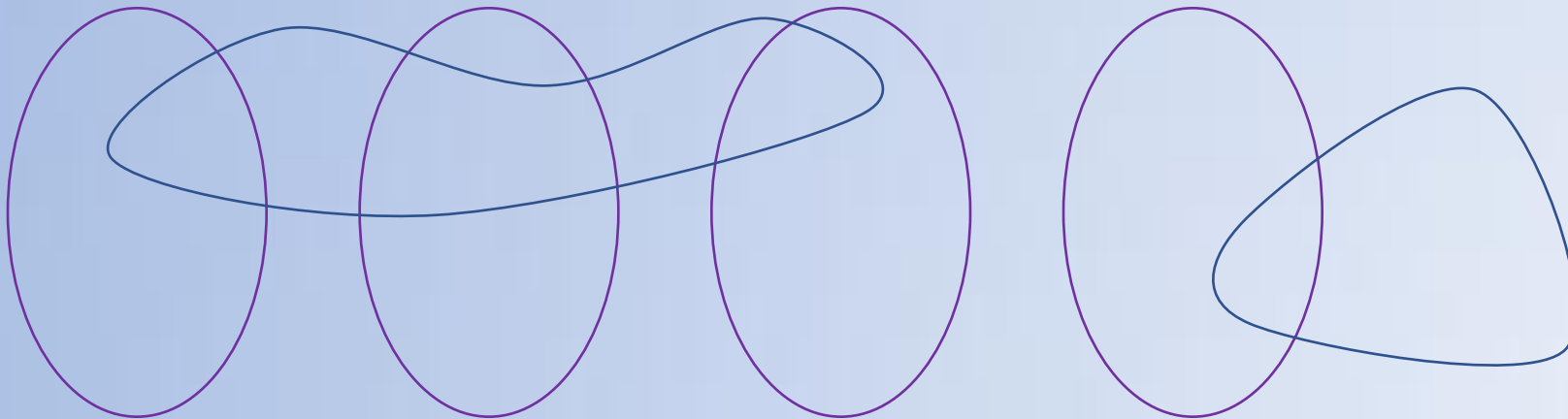
Local-Connectivity ATSP [Svensson'15]

Instance $(G, \mathcal{C}, \mathbf{w})$ with induced weights \mathbf{w}

Lower bound function \mathbf{lb} with $\mathbf{lb} =$

Input: partition of the vertex set $\mathcal{C} =$

Output: Eulerian edge set E with $|E \cap C_i| \geq \mathbf{lb}(C_i)$ for each C_i



Local-Connectivity ATSP [Svensson'15]

Instance $\mathcal{I} = (V, E, w)$ with induced weights \mathbf{w}

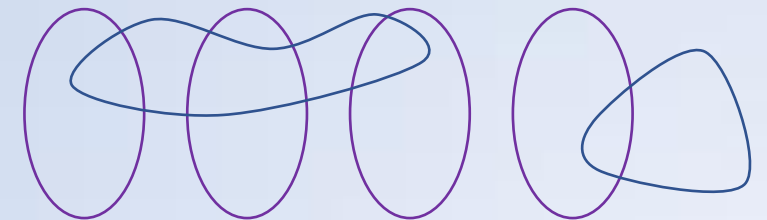
Lower bound function \mathbf{lb} with $\mathbf{lb} = \mathbf{w} + \mathbf{c}$

Input: partition of the vertex set $\mathcal{C} = \{C_1, \dots, C_k\}$

Output: Eulerian \mathcal{C} with $|E(C_i)| \geq \mathbf{lb}(C_i)$ for each C_i

-light algorithm: for every component C_i of \mathcal{C} ,

$$\frac{\mathbf{lb}(C_i)}{\mathbf{lb}(V)}$$



Every component pays for itself locally

Local-Connectivity ATSP [Svensson'15]

-light algorithm for
Local-Connectivity ATSP

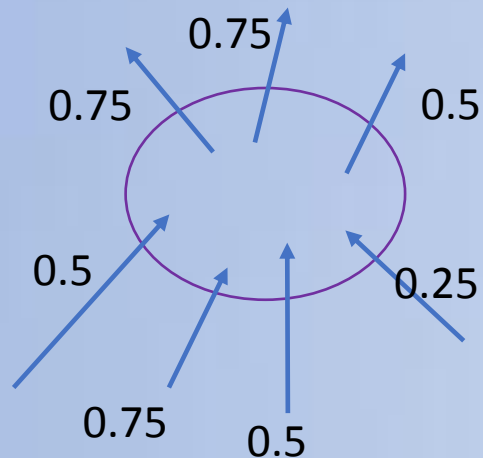


($+$)-approximation
for ATSP

Theorem [Svensson'15]
There exists a polytime $(+)$ -
approximation for node-weighted ATSP.

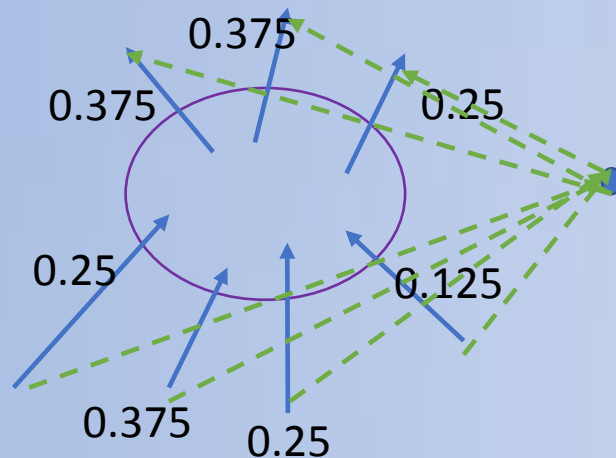
Local-Connectivity ATSP: node-weighted case

- Instance (G, w, c) , with G containing only singletons (ignore c)
 $(G, w) = \{s\} + \{t\}$
- Define $lb(s, t) = \{s, t\}$
- Partition G into \mathcal{C} all strongly connected
- Modify w and c , and solve an integer circulation problem



Local-Connectivity ATSP: node-weighted case

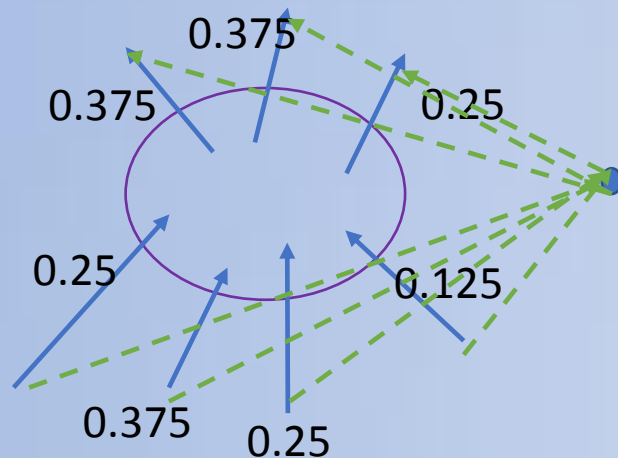
- Instance (G, w, τ) , with τ containing only singletons (ignore τ)
 $(G, w) = \{C_1\} + \{C_2\}$
- Define $lb(C_i) = \{C_i\}$
- Partition $\tau = \tau_1 \cup \tau_2$ all strongly connected
- Modify G and w , and solve an integer circulation problem



- For each C_i , create auxiliary vertex
- Reroute 1 fractional unit of incoming and outgoing flow to
- Solve integer circulation problem routing =1 unit through each
- Map back to original

Local-Connectivity ATSP: node-weighted case

- The rerouted is feasible to the circulation problem of weight
- **Flow integrality:** there exists integer solution of weight
- After mapping back, every vertex with \geq has in-degree
- For a component C , $f(C) = \sum_{e \in C} f(e) = \sum_{e \in C} \{ \} + \{ \} \{ \}$
- $\mathbb{I}b(C) = \{ \}$ **2-light algorithm**



Local-Connectivity ATSP: one nonsingular set in

- Vertebrate pair $(\mathcal{A}, \mathcal{B})$. Assume \mathcal{A} has a single non-singleton component \mathcal{C} . Thus,

$$(\mathcal{A}, \mathcal{B}) = \left(\begin{array}{c} \mathcal{C} + \mathcal{D} \\ \mathcal{C} + \mathcal{E} \end{array} \right) = \left(\begin{array}{c} \mathcal{C} \\ \mathcal{D} \end{array} \right) + \left(\begin{array}{c} \mathcal{E} \\ \mathcal{E} \end{array} \right)$$

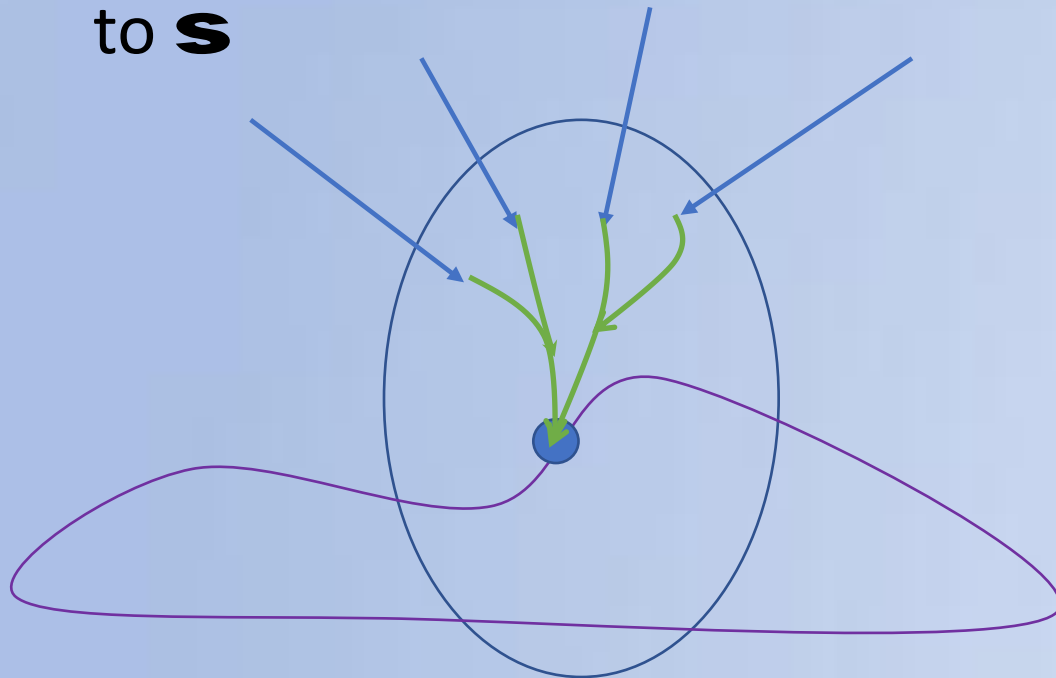
- Define

$$\mathbf{lb} = \frac{|\mathcal{C}|}{|\mathcal{A}|}$$

- $\mathbf{lb} = \frac{|\mathcal{C}|}{|\mathcal{A}|}$, since $(\mathcal{C}, \mathcal{C}) = \Theta(\mathcal{C})$

Local-Connectivity ATSP: one nonsingular set in

- By assumption, $\sum_{i \in S} \text{deg}(i) = 1$
- Backbone property: there is a node $s \in S$
- Simple flow argument: we can route the incoming 1 unit of flow to to \mathbf{s}



Local-Connectivity ATSP: one nonsingular set in

- Partition \mathcal{E} =
- Add backbone \mathcal{E}_0 into Eulerian set \mathcal{E} .
- Via flow splitting, **force** all edges entering \mathcal{E}_0 to proceed to
- Create auxiliary vertices \mathcal{V}_0 as before
- Solve integral circulation problem, and add solution to \mathcal{E}_0 .

Local-Connectivity ATSP: one nonsingular set in

Analysis

- For all components not crossing S , $\text{lb}(C) \approx \text{lb}(C)$ exactly as in the node-weighted case
- **Giant component** containing v .
 - Contains all edges crossing S
 - Has lower bound $\text{lb}(C) \approx \text{lb}(C) = \Theta(\text{lb}(C))$
 - $\text{lb}(C) \approx \text{lb}(C)$
- Therefore solution is ϵ -light.
- Same approach extends to arbitrary ϵ : enforce that every subtour crossing a set in \mathcal{S} must intersect the backbone.

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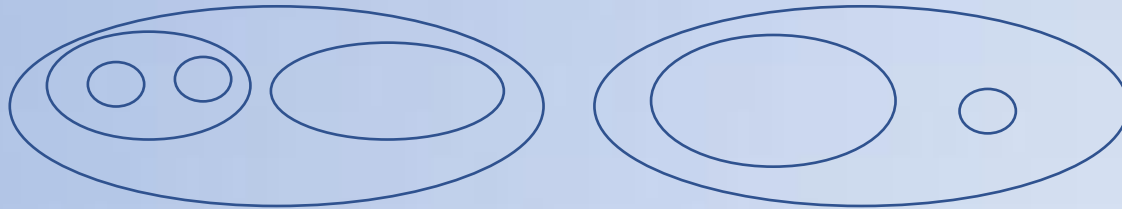
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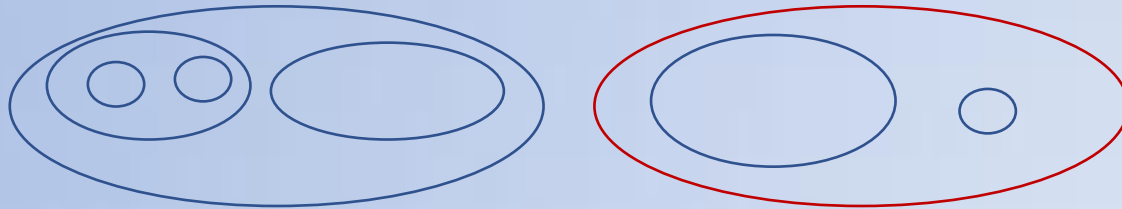
Motivation: reducing by contraction

- All sets in the family are singletons: node-weighted ATSP
- Would like to reduce the problem by contracting nonsingleton sets in



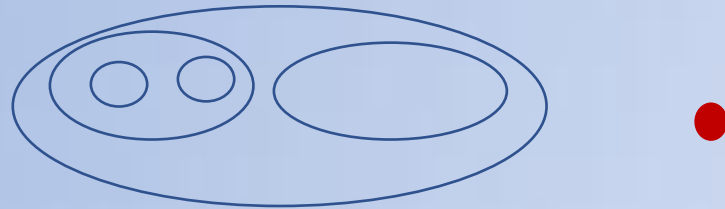
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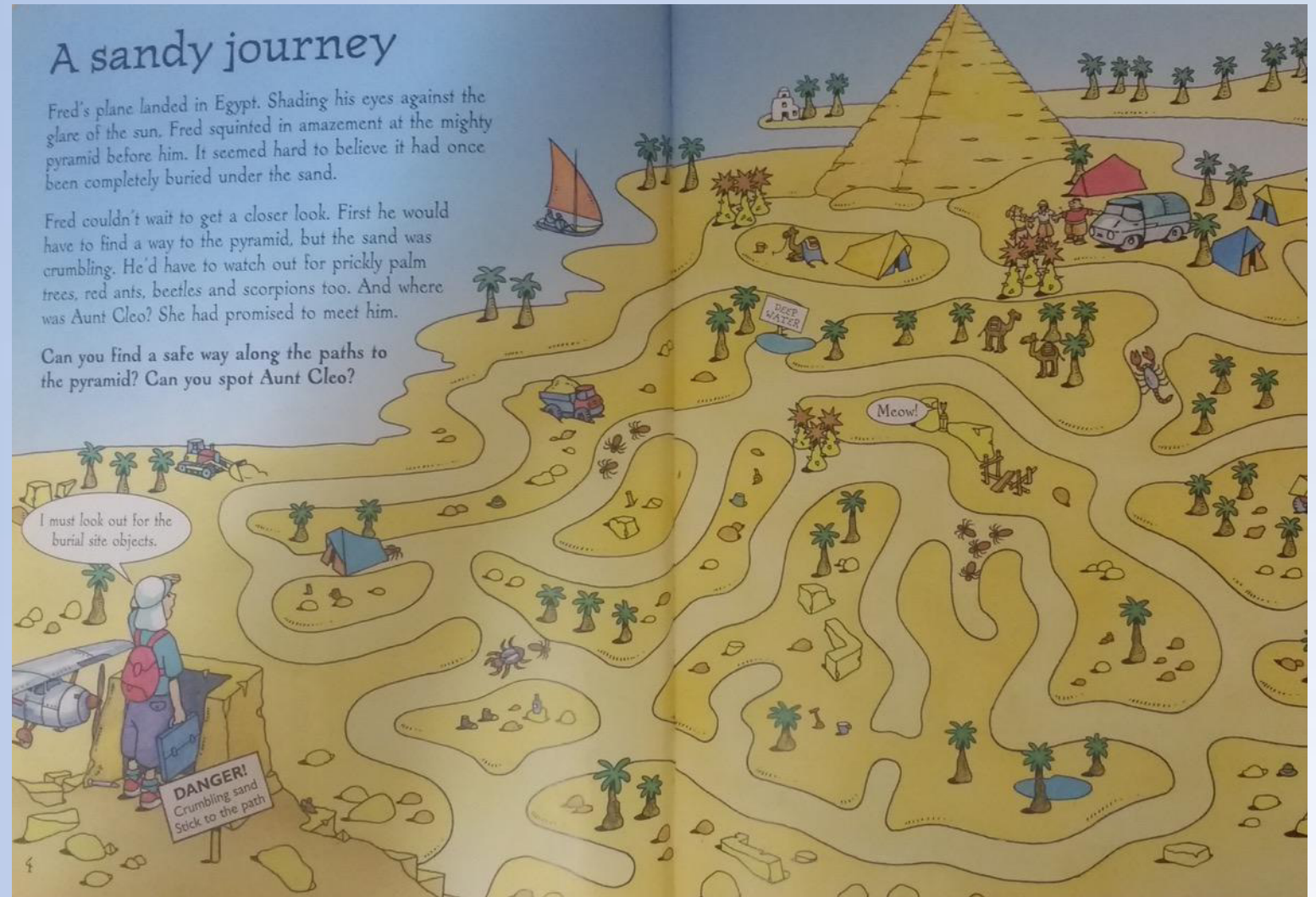
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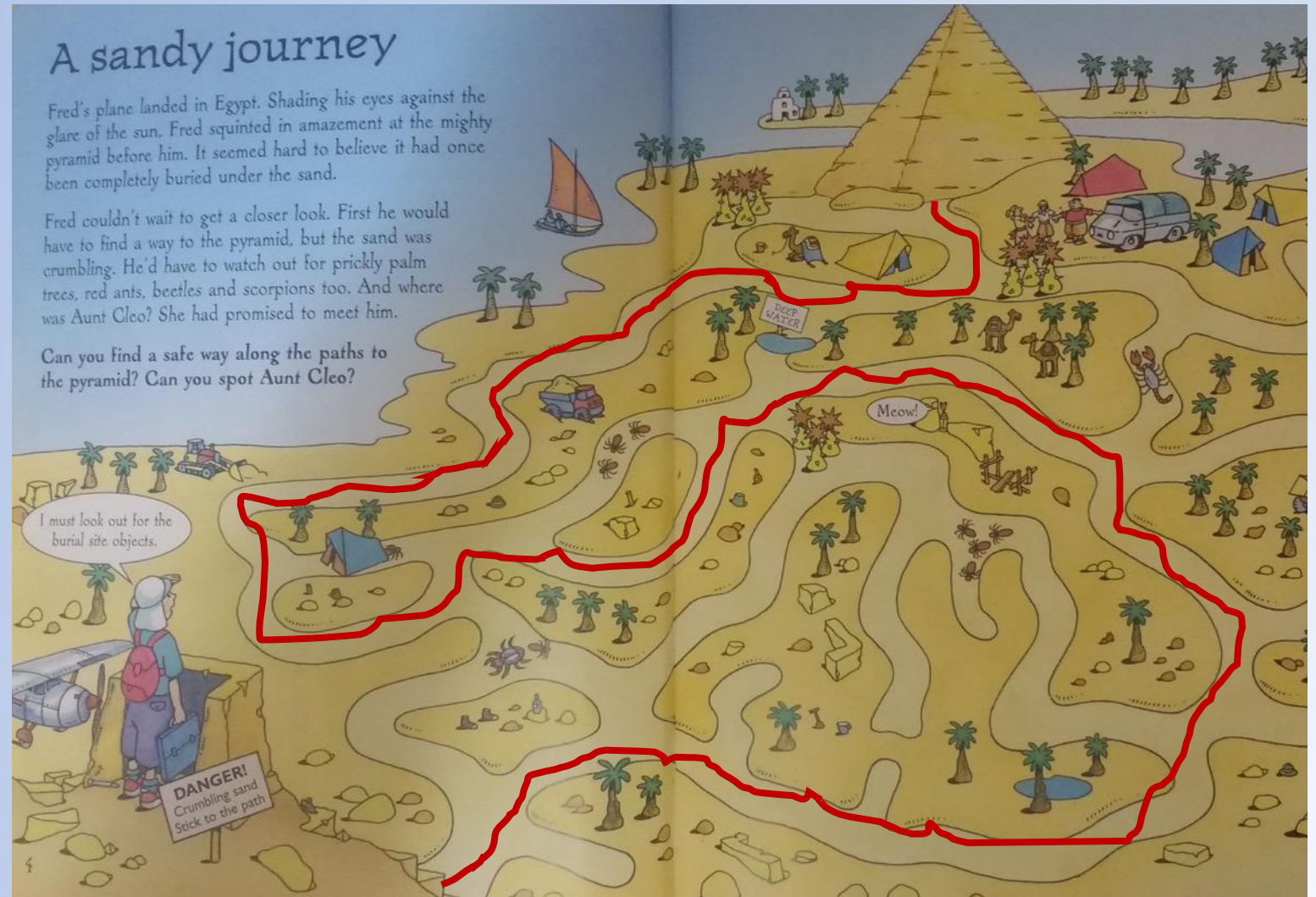
Irreducible instances

- Irreducible set S :
There exists $s, t \in S$ such that the shortest path between s and t inside S visits *almost all* sets in S .
- Irreducible instance (G, s, t) :
all sets in G are irreducible



Irreducible instances

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There exists $s, t \in S$ such that the shortest path between s and t inside S visits *almost all* sets in S .
- Irreducible instance (G, s, t) :
all sets in G are irreducible



Irreducible instances

- **Reducible set S** : For every pair u, v , there is a cheap path connecting them (if they are connected).
- Reducible sets can be contracted.

Theorem:

polytime ϵ -approximation for irreducible instances

polytime ϵ -approximation for arbitrary instances

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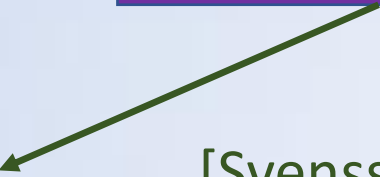
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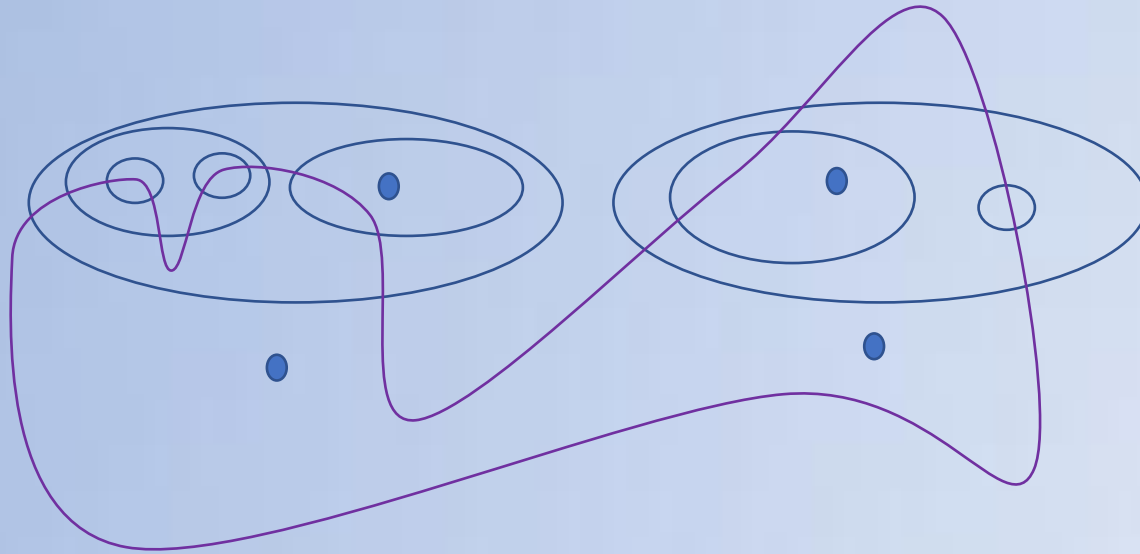
Graph theory:
contractions



Vertebrate pairs

Vertebrate pair $(\mathcal{V}, \mathcal{E})$

- $\mathcal{V} = (\mathcal{V}_1, \mathcal{V}_2, \mathcal{V}_3)$ instance
- \mathcal{E} : backbone = subtour that crosses every nonsingleton set in



Finding a vertebrate pair in an irreducible instance $= (V, E, w)$

1. Obtain a node-weighted instance by contracting all maximal sets in \mathcal{S}
2. Use [Svensson '15] to find a tour here, and blow it back to a subtour in the original instance in a pessimistic way: inside each maximal S , T crosses S .
3. If it crosses every set in \mathcal{S} , then (T, S) is a vertebrate pair
4. Otherwise, recurse by contracting all maximal sets in \mathcal{S} not crossed by T .
This works because their total weight is $\leq \frac{1}{2} \sum_{S \in \mathcal{S}} w(S)$.

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[Svensson '15]

Summary

- Via all these reductions, we obtain an **5500**-approximation algorithm for ATSP.
- Squeezing the arguments a bit more and opening up black boxes, can be probably decreased to a few hundreds.
- Still very far from lower bound 2 on the integrality gap of Held-Karp

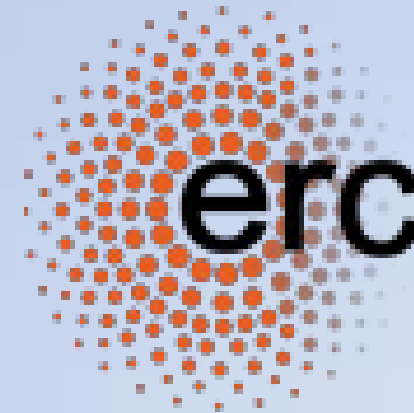
Open questions

- Improve to a constant \ll
- **Thin tree** conjecture is still open.
- Bottleneck ATSP.
- Better than $3/2$ approximation for symmetric TSP.

SCALEOPT

Scaling Methods for Discrete and Continuous Optimization

- ERC Starting Grant 2018-22
- Openings for post docs and PhD students
<http://personal.lse.ac.uk/veghl/scaleopt.html>



European
Research
Council



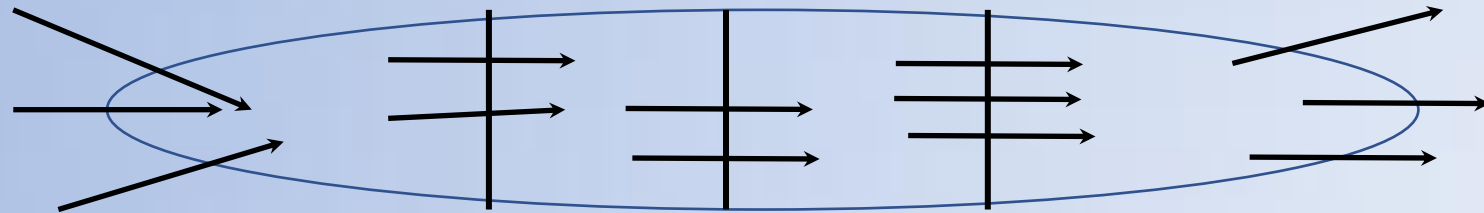
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Thank you!

Simplifying assumption *for the talk*

Assumption: all sets in the family are strongly connected in .

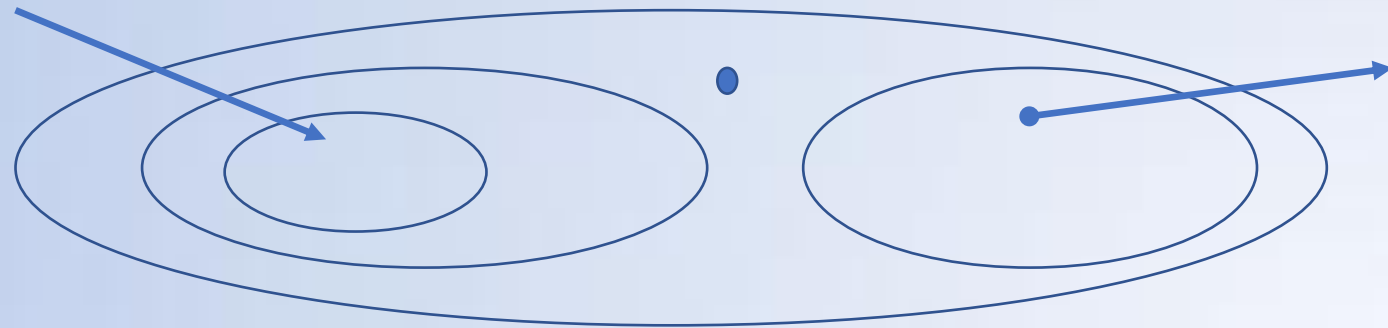
Not true in general, but the connected components have a nice path structure:



Paths traversing a set

- How much is the weight of connecting an incoming and an outgoing edge in a set **S** ?

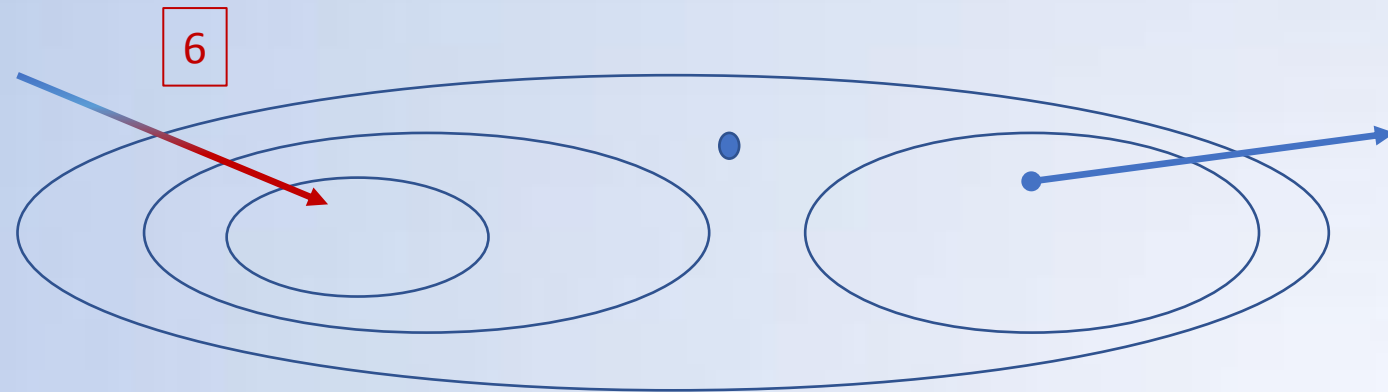
$$(,) =$$



Paths traversing a set

- How much is the weight of connecting an incoming and an outgoing edge in a set **S** ?

$$(,) = : ,$$



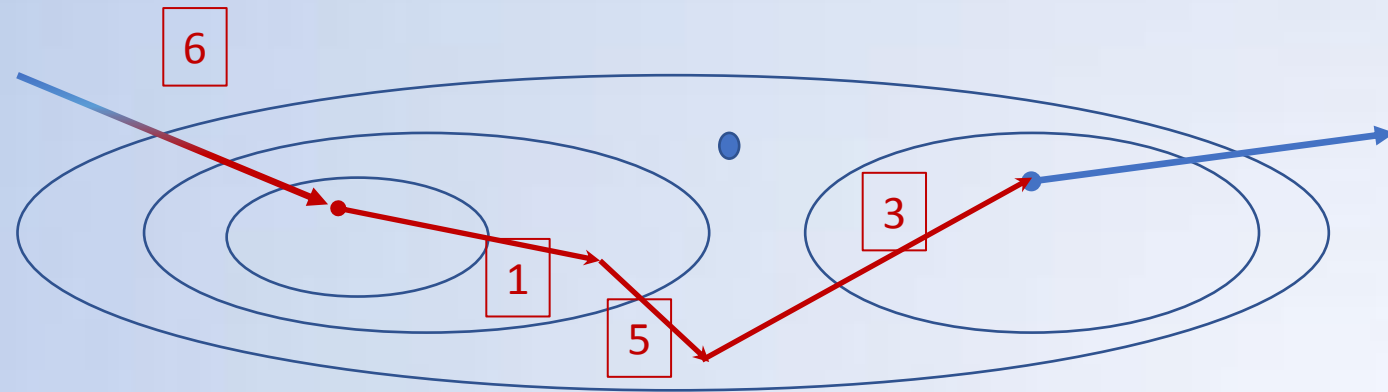
Paths traversing a set

- How much is the weight of connecting an incoming and an outgoing edge in a set S ?

$$(s, t) = \text{Min weight path inside } S + (s, t)$$

:

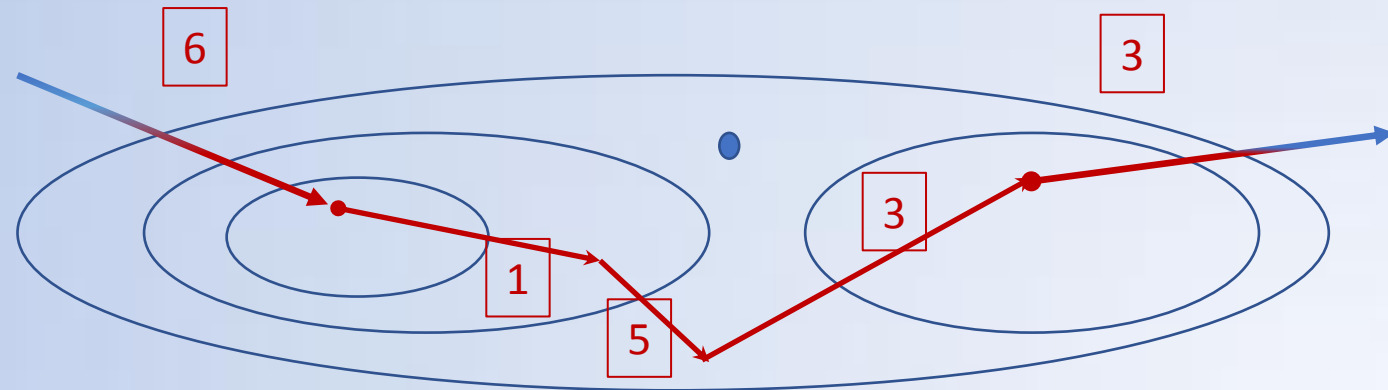
Min weight path inside .



Paths traversing a set

- How much is the weight of connecting an incoming and an outgoing edge in a set S ?

$$(s, t) = \text{weight}(s, t) + (s, t) + \text{weight}(s, t) = \text{weight}(s, t)$$



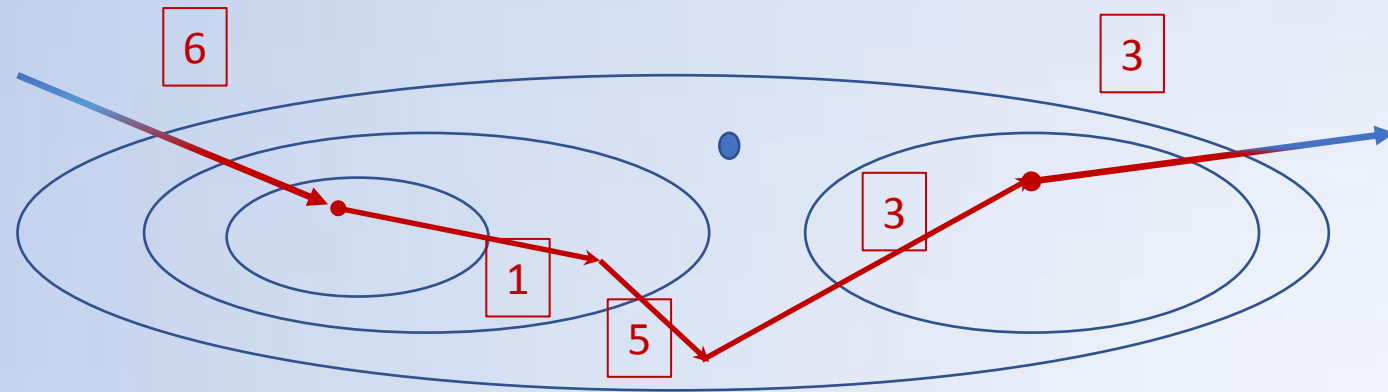
Paths traversing a set

- How much is the weight of connecting an incoming and an outgoing edge in a set S ?

$$\begin{aligned}
 (u, v) &= \text{weight}(u, v) + \text{weight}(u, v) + \dots = \\
 &: \text{weight}(u, v) : \text{weight}(u, v) : \dots
 \end{aligned}$$

Lemma:

$$(u, v) = \dots$$



Irreducible instances

- Reducible set S :

$$\text{Max}_{S'}(f, g) - \text{Max}_S(f, g)$$

- Irreducible instance (f, g) :
no set S is reducible

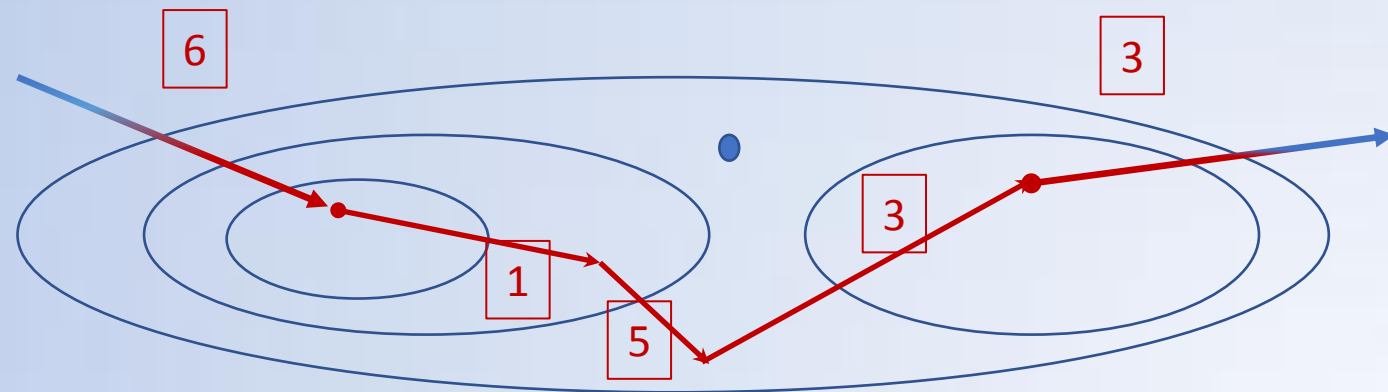
Lemma:

$$\text{Max}_{S'}(f, g) = \text{Max}_S(f, g)$$

Theorem:

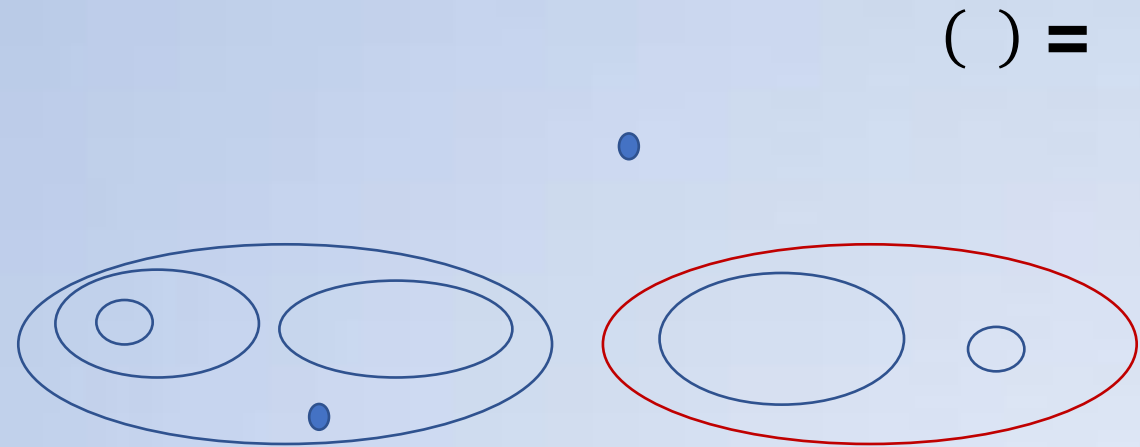
polytime ϵ -approximation for
irreducible instances

polytime ϵ -approximation for
arbitrary instances



Recursive algorithm via contractions

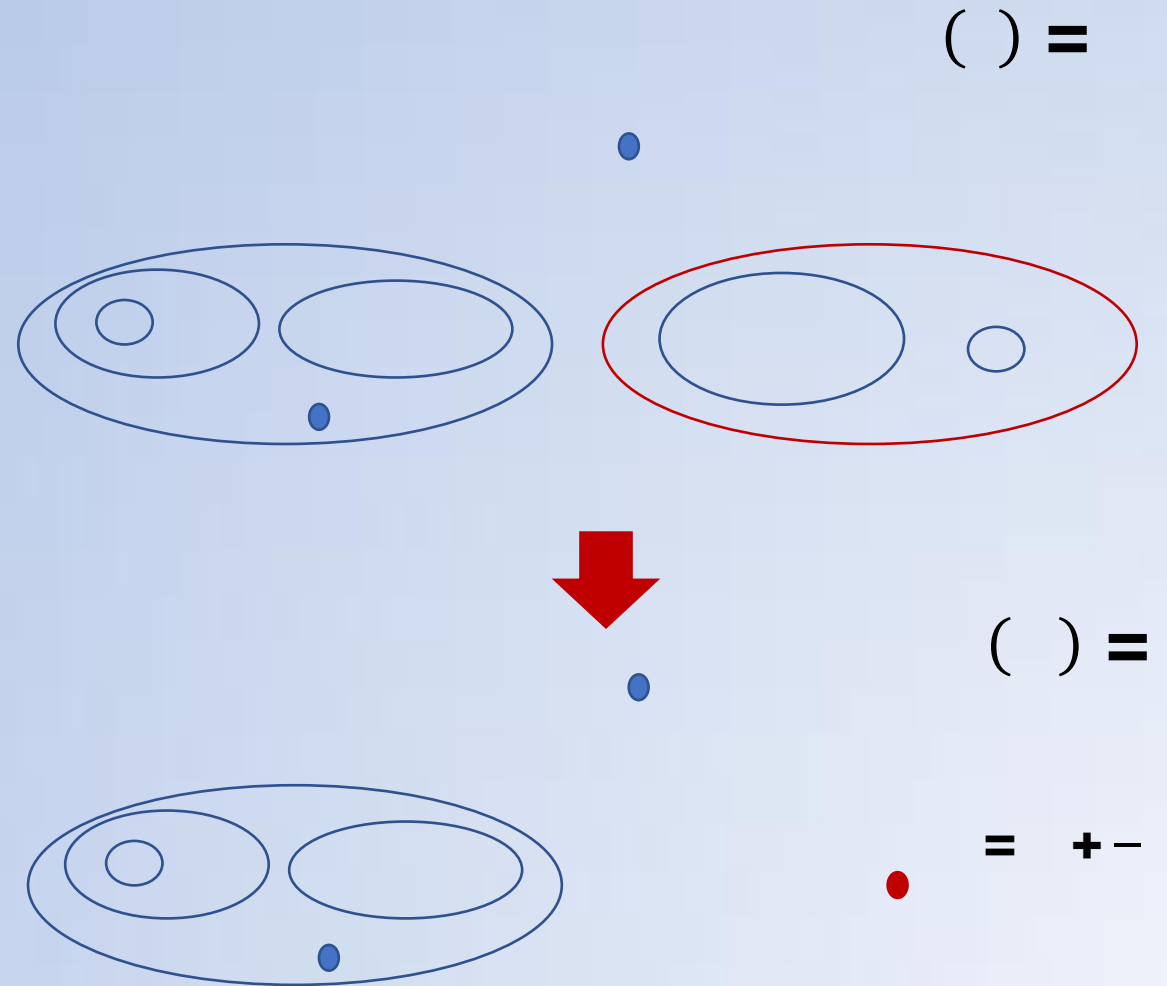
- Instance (G, s, t, c)
- $OPT(G, s, t, c) =$
 $=$ Held-Karp optimum
- R : minimal reducible set in G



-approximation for (G, s, t, c)
 =
 -approximation on instance by contracting R
 +
 -approximation of irreducible instance $(G \setminus R, s, t, c)$ inside

Recursive algorithm via contractions

- Instance (S, w)
- $(S, w) =$
 $=$ Held-Karp optimum
- S : minimal reducible set in S
- $(S, w) =$: contract in S
- $(S, w) =$: contract in S
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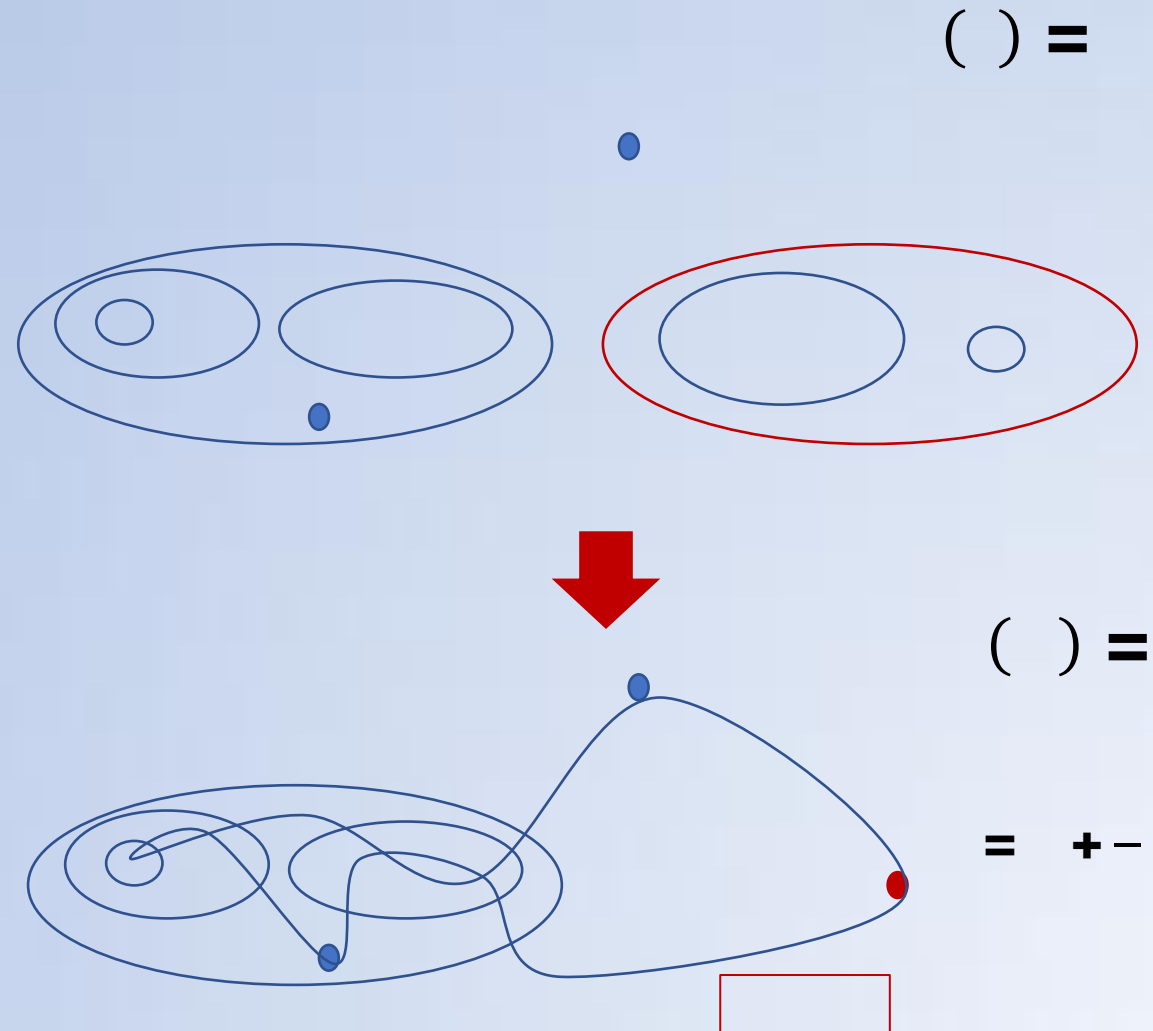


Recursive algorithm via contractions

Inductive assumption: We have a polytime ϵ -approximation for smaller instances

- Apply recursively on (S, ϵ) to obtain tour T

$$= (S, \epsilon) - - (S, \epsilon)$$



Contracting

Inductive assumption: We have a polytime ϵ -approximation for smaller instances

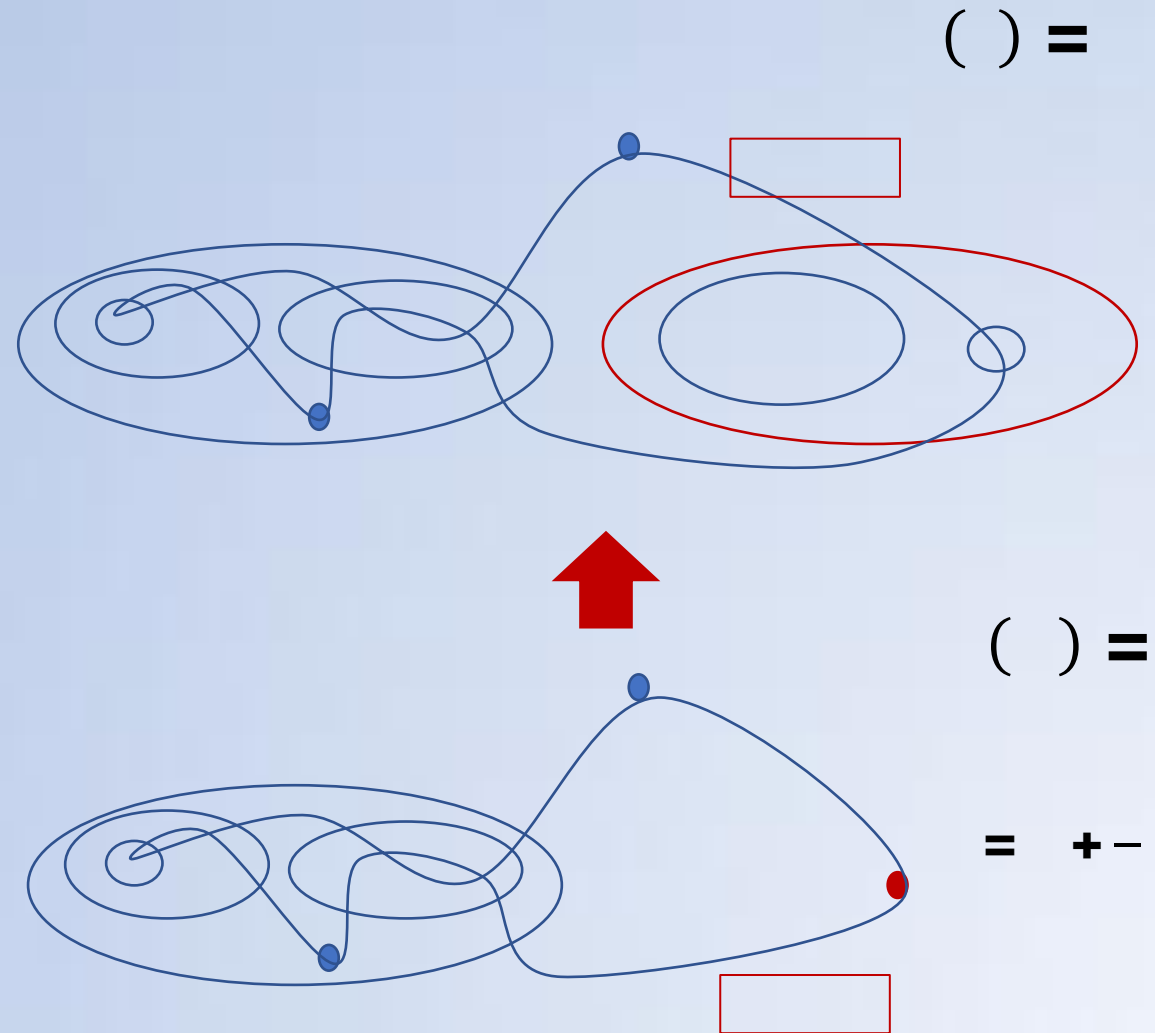
- Apply recursively on G to obtain tour T

$$T = T_{\text{subtour}} + T_{\text{contracted}}$$

$$= T_{\text{subtour}} + T_{\text{contracted}}$$

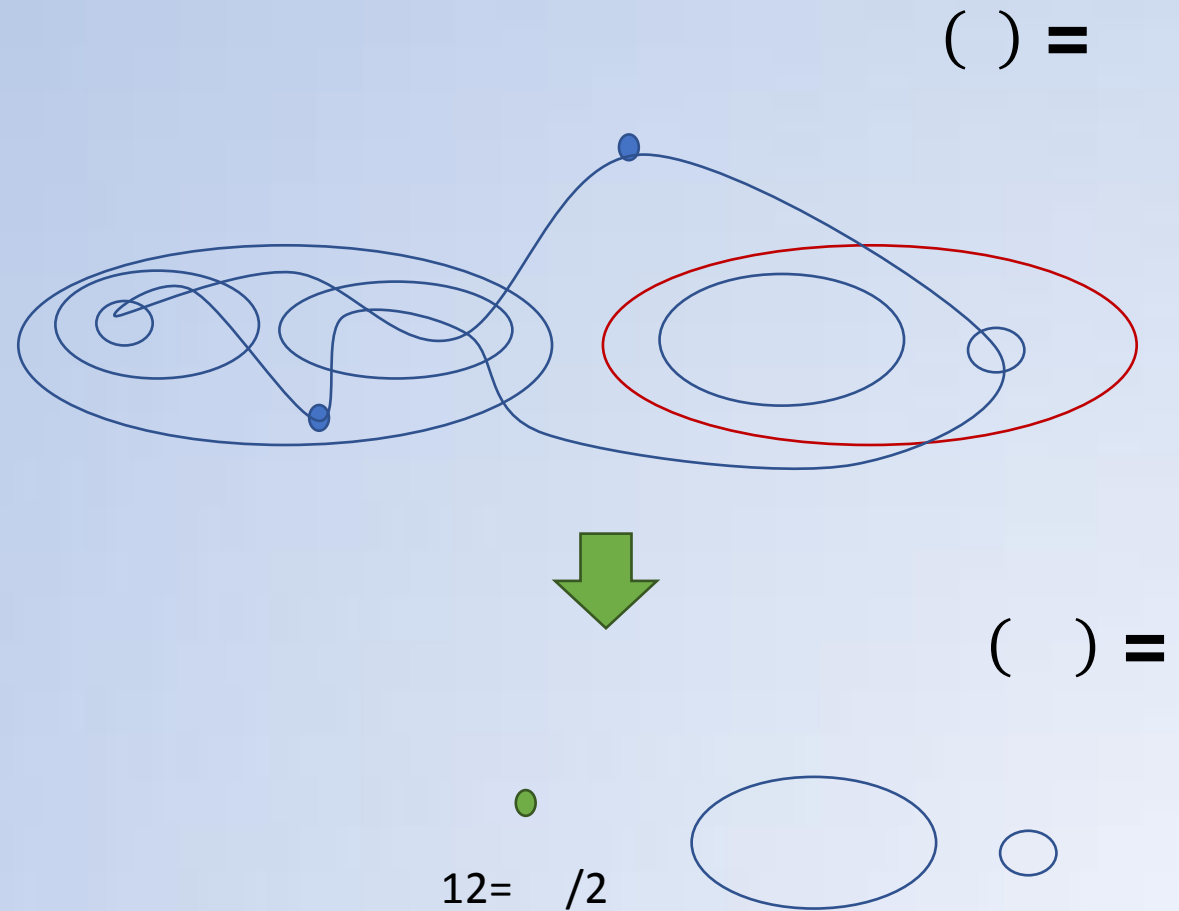
- Map back to subtour T_{subtour} in G with

$$T_{\text{subtour}}$$



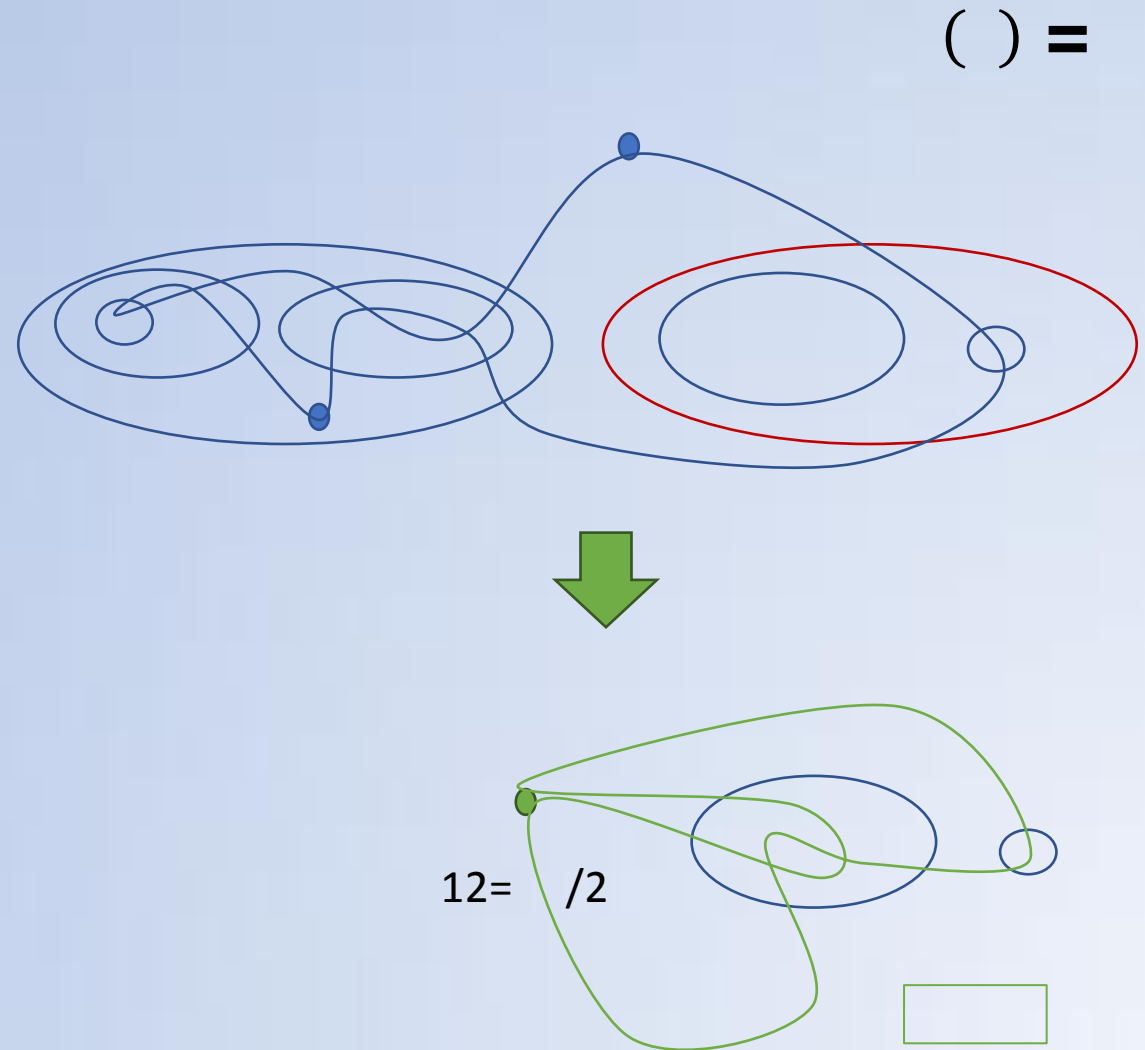
Inducing on

- We add a tour inside S , using the ϵ -approximation on irreducible instances.
- \mathcal{S} : remove S , and contract to \mathcal{S} , with $\{ \} = /$
- \mathcal{S} is irreducible.



Inducing on

- We add a tour T inside S , using the α -approximation on irreducible instances.
- \mathcal{I} : remove S , and contract to \mathcal{I}' , with $w(\mathcal{I}') = \frac{1}{\alpha} w(\mathcal{I})$.
- \mathcal{I}' is irreducible.
- Find tour T' in \mathcal{I}' with weight $w(T') = \frac{1}{\alpha} w(T)$.



Inducing on

- Find tour in with weight $() =$
- Map back to in with $() ()$

• is a tour in

$$\begin{aligned}
 & () \quad () \quad () \quad () \\
 & + \quad () = \quad () \quad ()
 \end{aligned}$$

