

Shorter Tours by Nicer Ears

7/5-approximation for graphic TSP,
3/2 for the path version,
and 4/3 for two-edge-connected subgraphs

Jens Vygen

(joint work with András Sebő)

September 21, 2012

Metric TSP

Given a complete graph G and metric weights $c : E(G) \rightarrow \mathbb{R}_{\geq 0}$, find a Hamiltonian circuit in G with minimum total weight.

- ▶ NP -hard
- ▶ best known approximation ratio $\frac{3}{2}$ (Christofides [1976])
- ▶ no $\frac{185}{184}$ -approximation algorithm exists unless $P = NP$ (Lampis [2012])
- ▶ integrality ratio of subtour relaxation between $\frac{4}{3}$ and $\frac{3}{2}$ (Wolsey [1980])

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- ▶ approximation ratio $1.5 - \epsilon$ (Gharan, Saberi, Singh [2011])
- ▶ approximation ratio 1.461 (Mömke, Svensson [2011])
- ▶ approximation ratio 1.445 (Mucha [2012])

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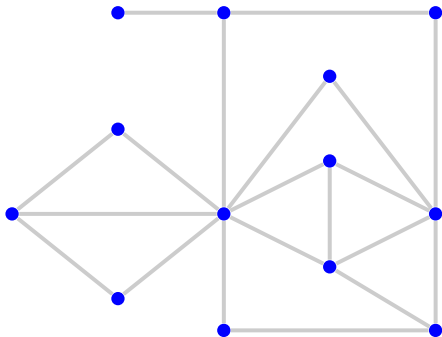
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We will show an approximation ratio of 1.4.

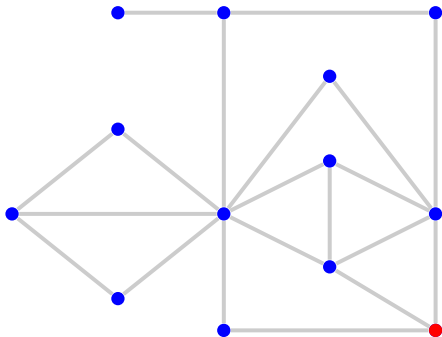
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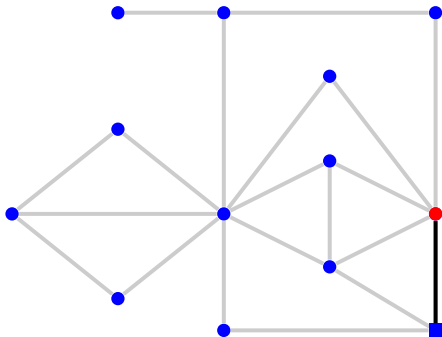
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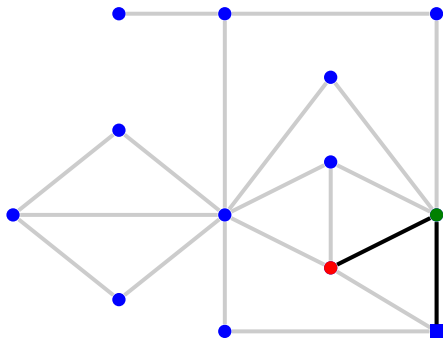
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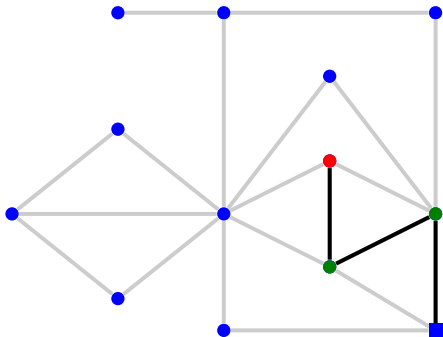
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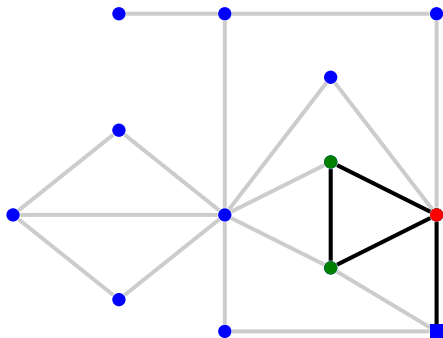
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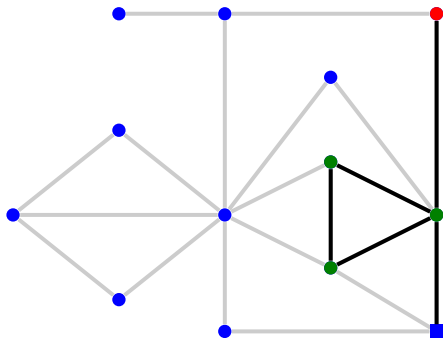
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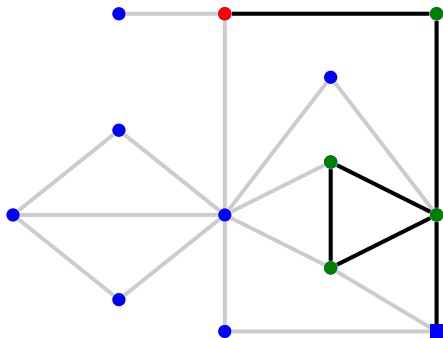
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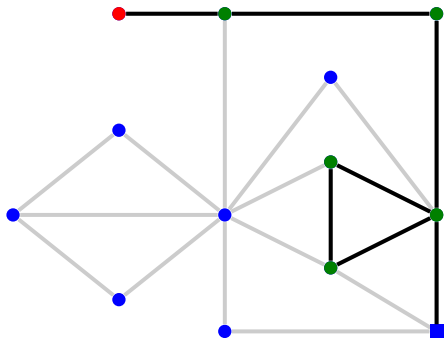
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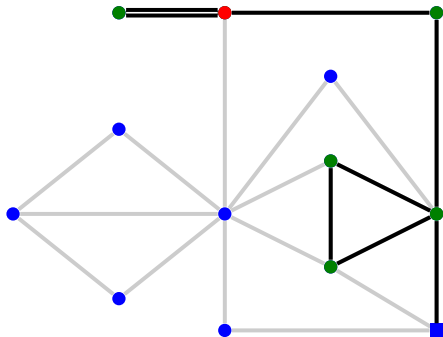
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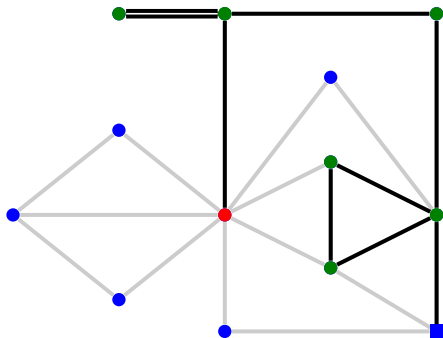
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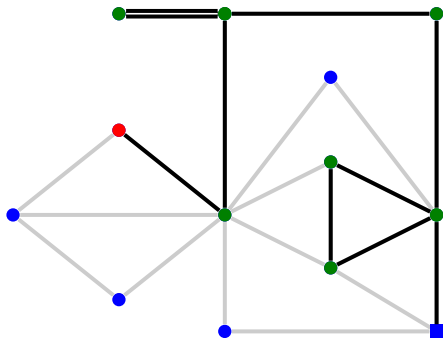
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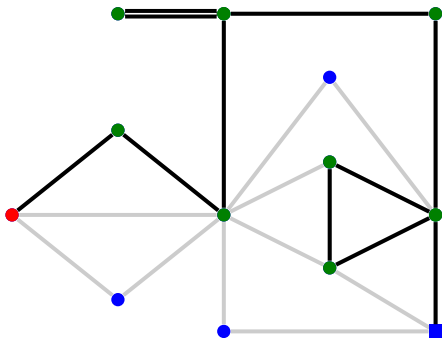
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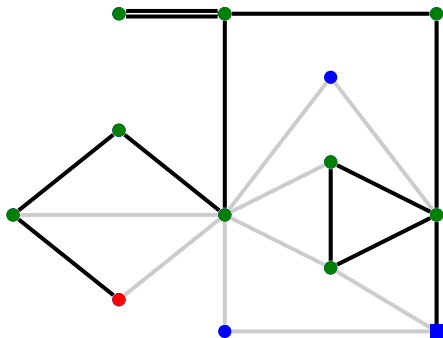
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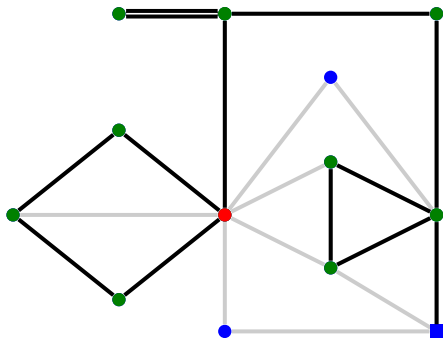
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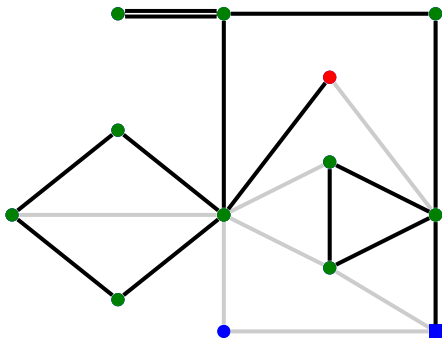
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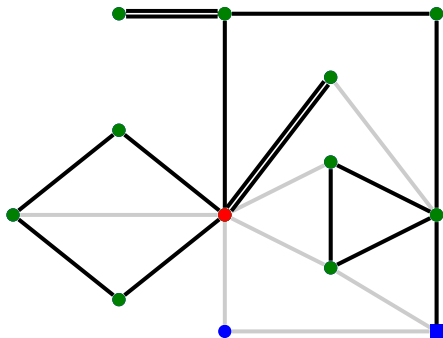
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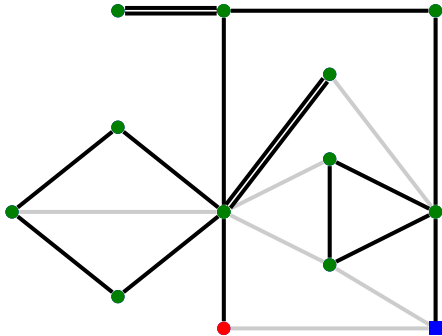
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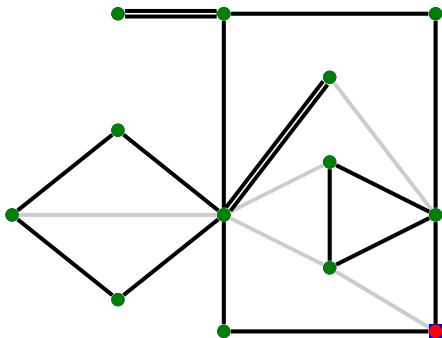
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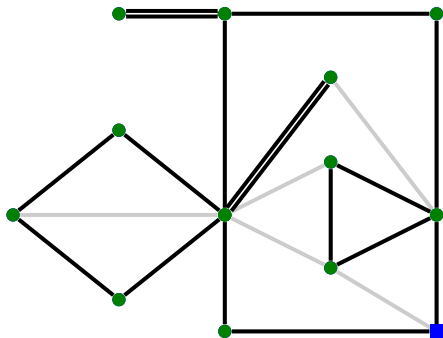
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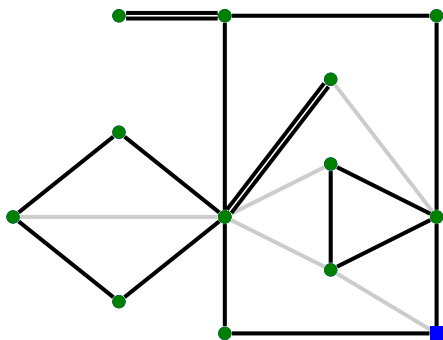
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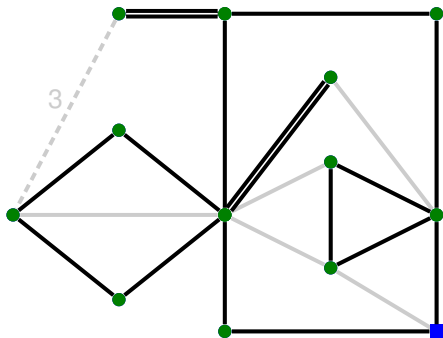


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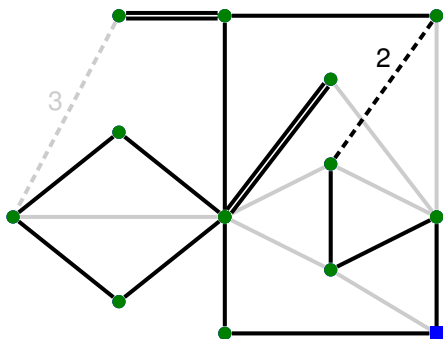


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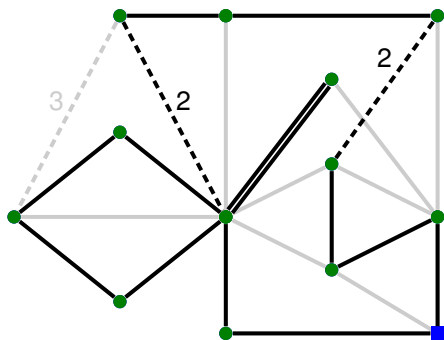


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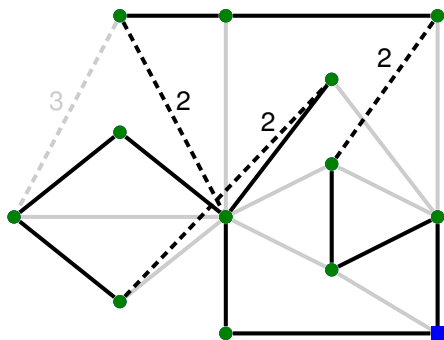


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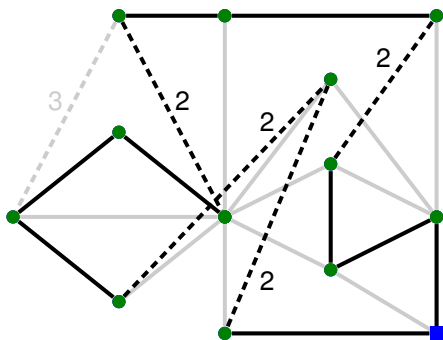


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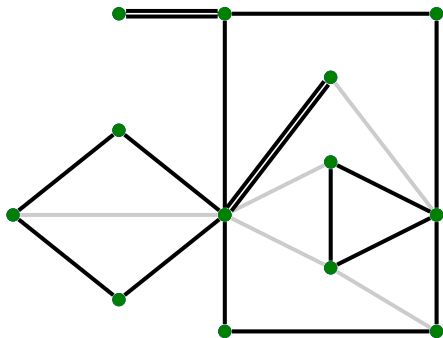


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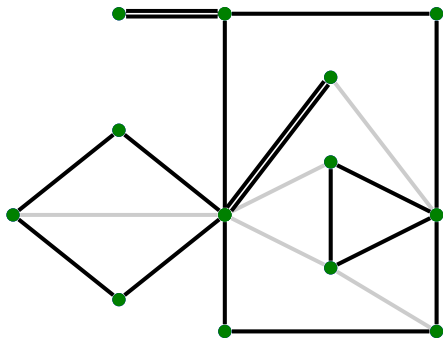


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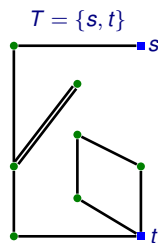
its edge set is called **tour** or **connected- \emptyset -join** of G

Main results

Let G be a connected graph and $T \subseteq V(G)$ with $|T|$ even.

A **connected- T -join** of G (aka **T -tour**) is a set $F \subseteq E(2G)$ such that

- ▶ $(V(G), F)$ is connected, and
- ▶ $v \in T \Leftrightarrow |F \cap \delta(v)|$ odd.

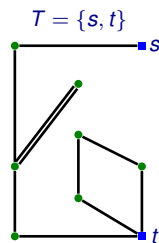


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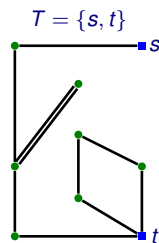
- ▶ **Graphic TSP** (smallest tour = connected- \emptyset -join):
from $\frac{13}{9}$ (Mucha [2012]) to $\frac{7}{5}$
- ▶ **Connected- T -join** (smallest connected- T -join):
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1.578 for $T = \{s, t\}$ (An, Kleinberg, Shmoys [2012]), to $\frac{3}{2}$
- ▶ **2ECSS** (smallest 2-edge-connected spanning subgraph):
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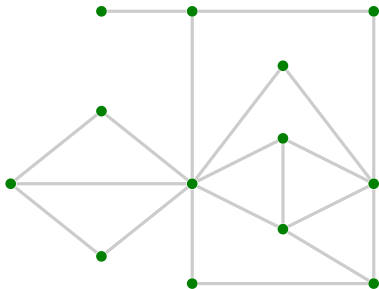


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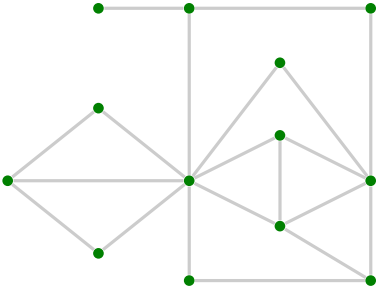
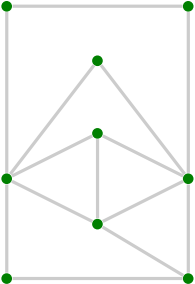
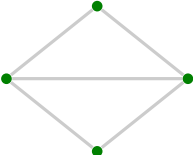
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Note that doubling edges is necessary (except for 2ECSS) and tripling edges does not help.

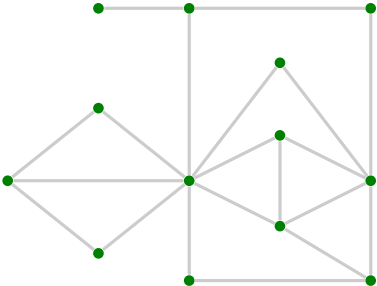
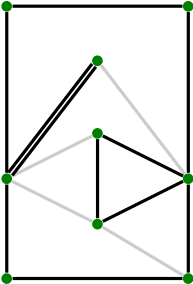
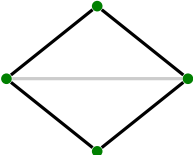
Consider blocks separately



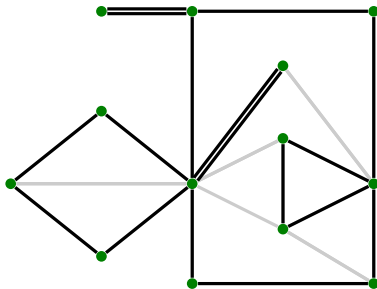
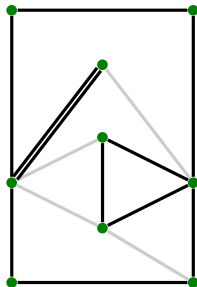
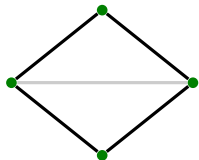
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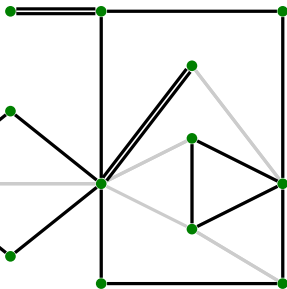
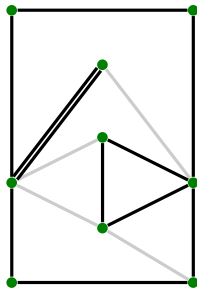
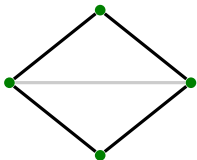
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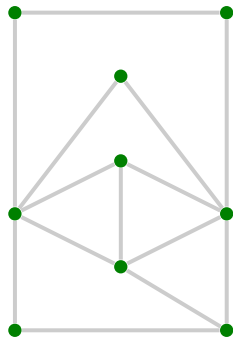


So we may assume that the input graph 2-vertex-connected.

Ear-decompositions

Write $G = P_0 + P_1 + \cdots + P_k$, where P_0 is a single vertex, and each P_i ($i = 1, \dots, k$) is either

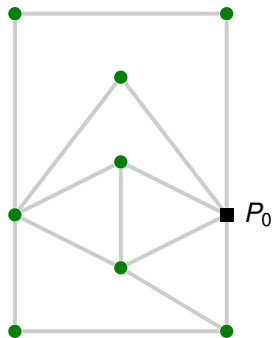
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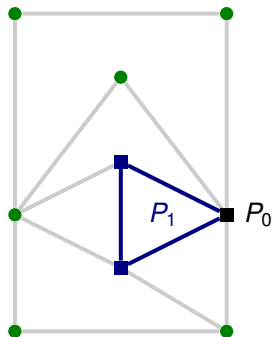
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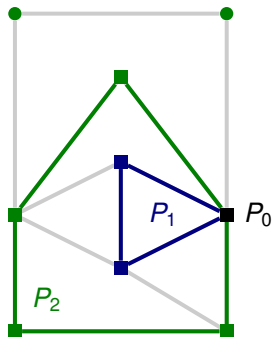
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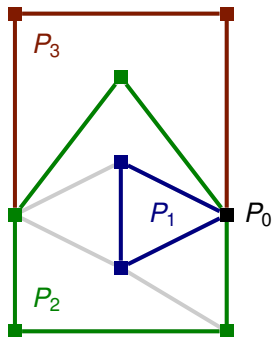
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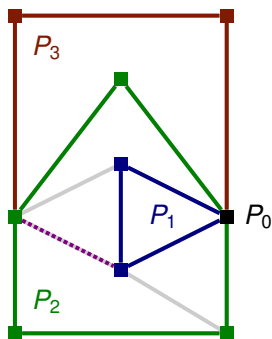
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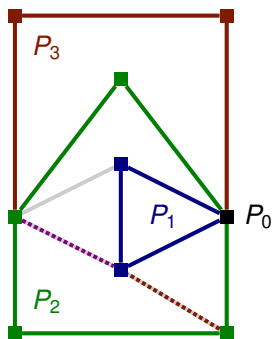
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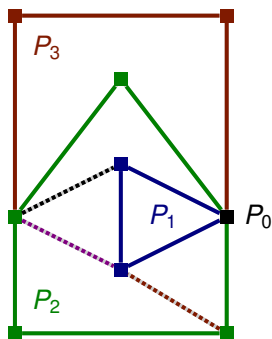
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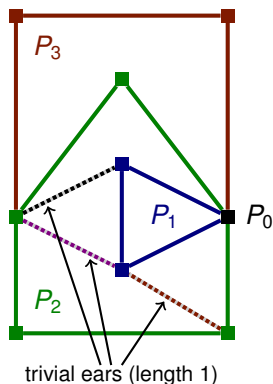
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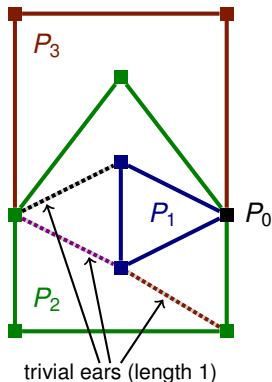
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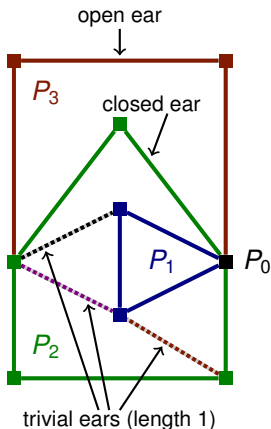


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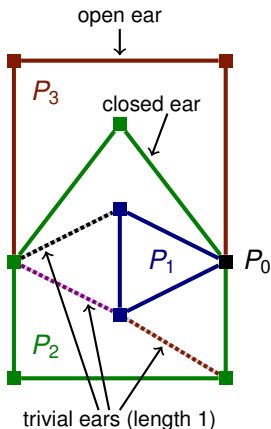


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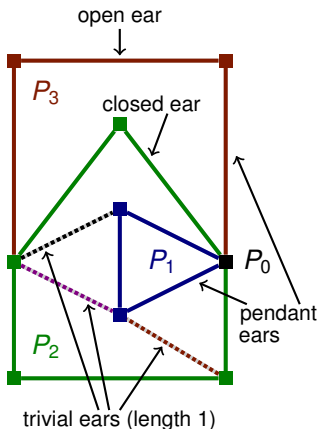


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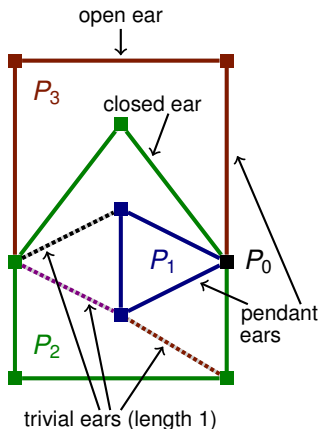


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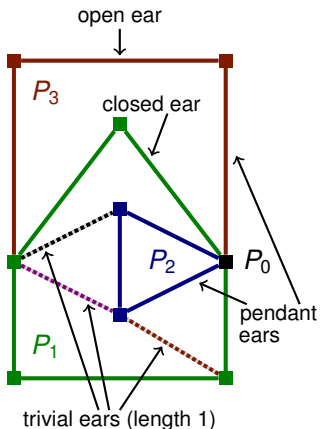


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Ear-decompositions with fewest even ears

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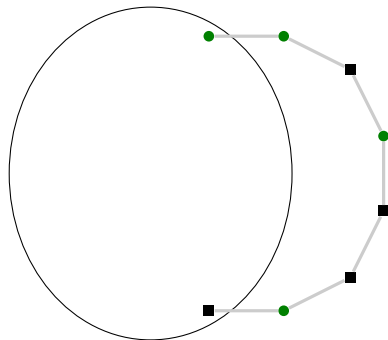
$$\frac{|V(G)|-1+\varphi(G)}{2} = \max\left\{\min\{|J| : J \text{ is a } T\text{-join}\} : T \subseteq V(G), |T| \text{ even}\right\}.$$

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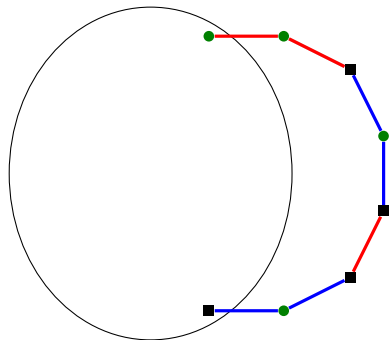
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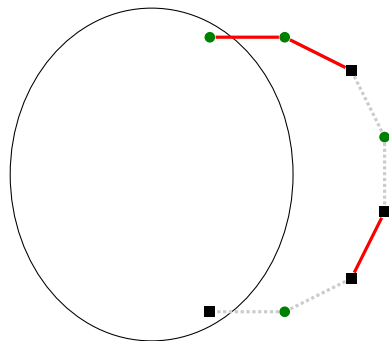
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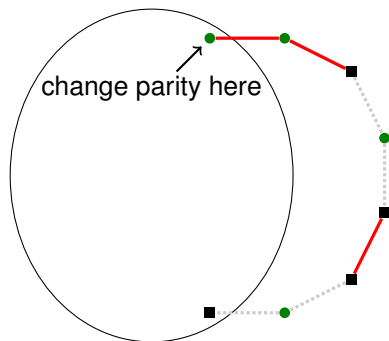
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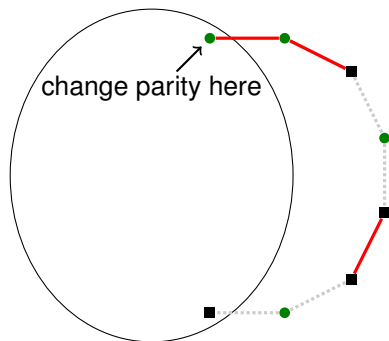
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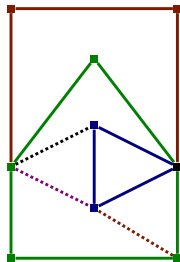
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- ▶ This yields a T -join with $\leq \frac{1}{2}(|V(G)| - 1 + k_{\text{even}})$ edges

□

Ear-decompositions for 2ECSS

Simple algorithm for 2ECSS: delete all trivial ears.

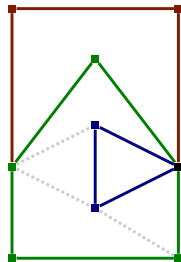


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where k_{even} is the number of even ears,
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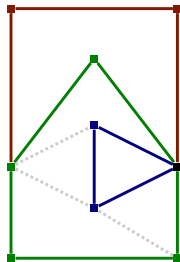


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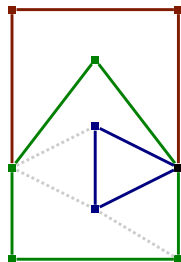
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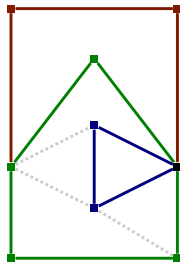
By Frank's theorem, we get $k_{\text{even}} = \varphi(G)$. Note $k_3 \leq \frac{1}{2}(|V(G)| - 1)$.

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This immediately yields a $\frac{3}{2}$ -approximation for 2ECSS

(was improved to $\frac{17}{12}$ by [Cheriyán, Sebő and Szigeti \[2001\]](#))

New algorithm for 2ECSS

- ▶ Compute a nice ear-decomposition.
- ▶ Optimize short ears so that they serve best for connectivity.

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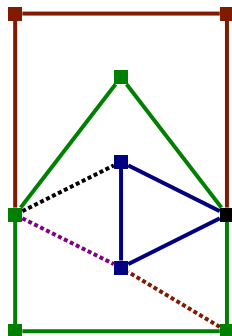
Alternative yields an 2ECSS with at most $\frac{5}{4}L + \frac{1}{2}\pi$ edges.

→ The better of the two 2ECSSs has at most $\frac{4}{3}L$ edges.

Nice ear-decompositions

An ear-decomposition is called **nice** if

- (i) the number of even ears is minimum,
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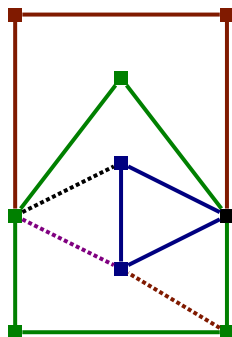
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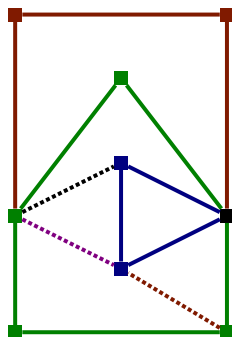
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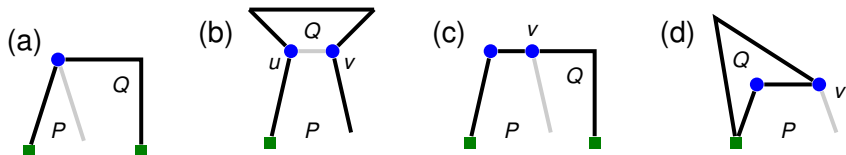
Sketch of Proof:

- ▶ Compute an ear-decomp. with fewest even ears (Frank [1993])
- ▶ Subdivide one edge of each even ear (\Rightarrow factor-critical graph)
- ▶ Compute an open odd ear-decomp. (Lovász, Plummer [1986])
- ▶ Undo subdivisions \Rightarrow open ear-decomp. with fewest even ears
- ▶ Replace non-pendant short ears
- ▶ Replace adjacent short ears



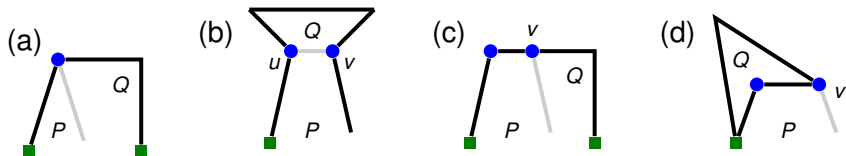
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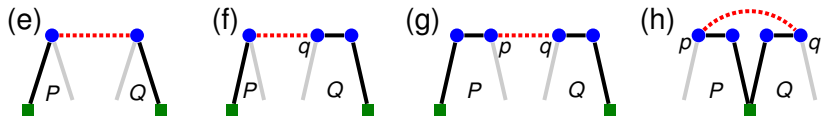


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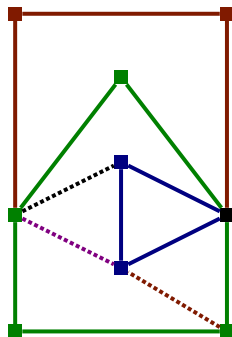


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Optimizing short ears

- ▶ Adding all short ears leaves some number of connected components

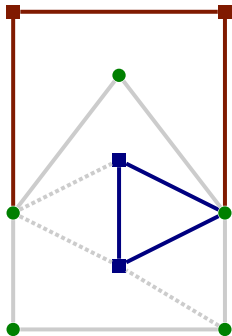


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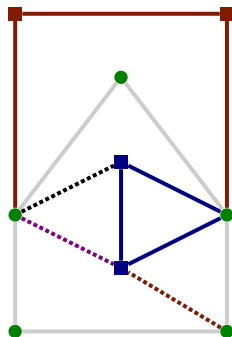


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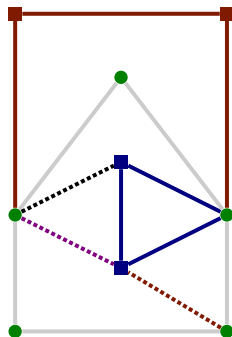


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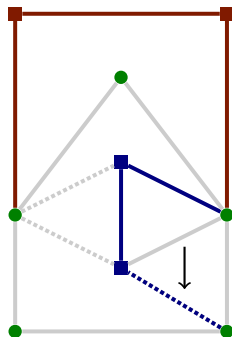


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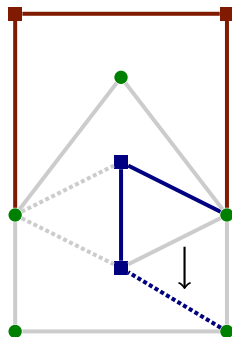


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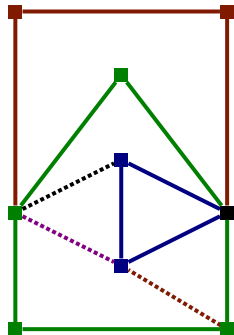
Note: Replacing some short ears by other ears (with the same internal vertices) will maintain a nice ear-decomposition.

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Eardrums and earmuffs

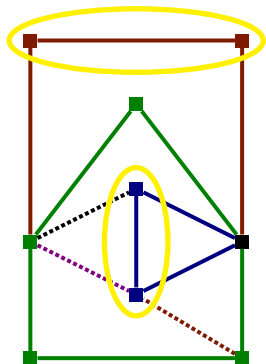
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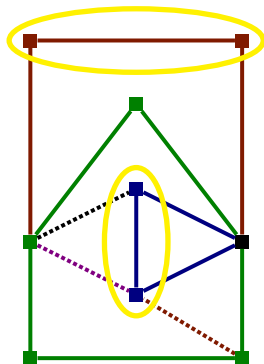


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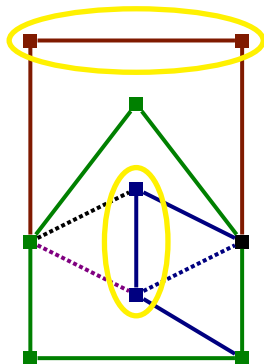
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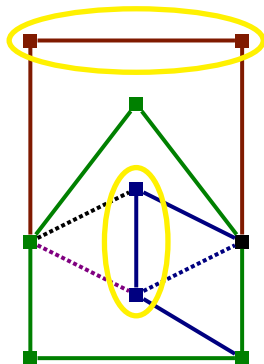
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Theorem

A maximum earmuff can be computed in polynomial time.

Call the maximum $\mu(G, M)$.



First solution: matroid intersection

- ▶ Represent each path $P \in \mathcal{P}_f$ ($f \in M$) by the set e_P of its two endpoints; let $E_f := \{e_P : P \in \mathcal{P}_f\}$.
- ▶ Let r be the rank function of the cycle matroid of the complete graph on $V(G)$.

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Theorem (Rado [1942])

Let E be a finite set and r the rank function of a matroid on E .

Let $E_1, E_2, \dots, E_k \subseteq E$. Then

$$\max\{r(\{e_1, \dots, e_k\}) : e_i \in E_i \ (i = 1, \dots, k)\} = \min\{r(\bigcup_{i \in I} E_i) + k - |I| : I \subseteq \{1, \dots, k\}\}.$$

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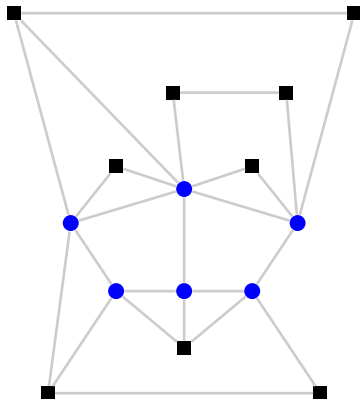
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Note: Special case of matroid intersection.

Earmuff maximization: example

■ vertex in V_M

● vertex in $V(G) \setminus V_M$

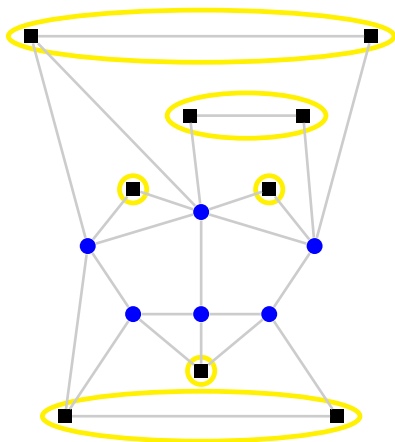


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○ earmuff



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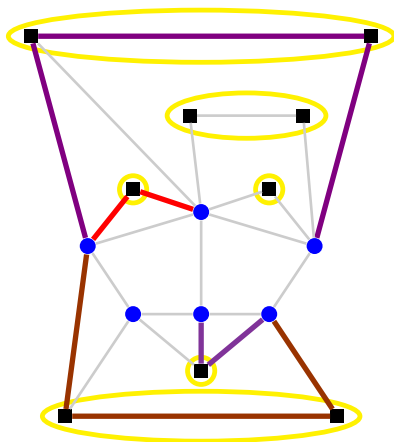


maximum earmuff



other edges

○ earmuff



Earmuff maximization: example

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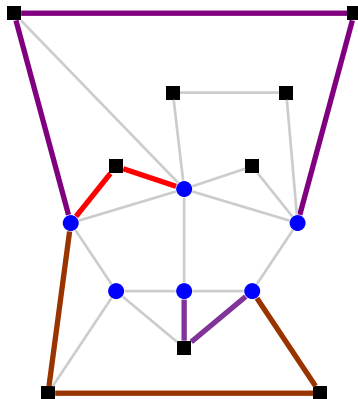
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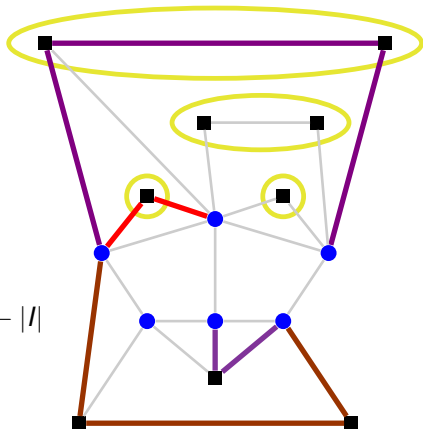
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 $\mu(G, M) = r(\bigcup_{i \in I} E_i) + |M| - |I|$




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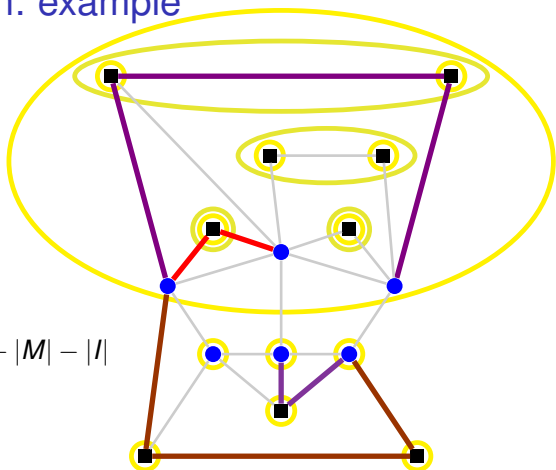
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⊙ cuts in dual solution




No edge belongs to more than 2 cuts. Number of cuts \geq
 $|V(G)| + |I| - r(\bigcup_{i \in I} E_i) - 1 = |V(G)| - 1 + |M| - \mu(G, M)$

Earmuff maximization: example

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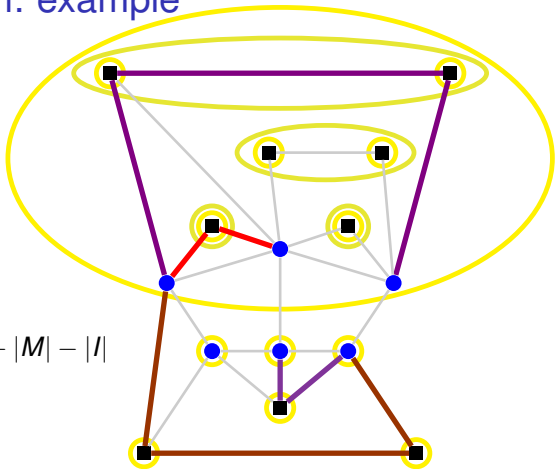
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Theorem

Any 2ECSS has at least $L_\mu := |V(G)| - 1 + |M| - \mu(G, M)$ edges. \square

New algorithm for 2ECSS

- ▶ Compute a nice ear-decomposition.
- ▶ Optimize short ears so that they serve best for connectivity.

Note: number of even ears is minimum, all short ears are pendant

- ▶ Take all edges of pendant ears.
 - ▶ Add edges to obtain connectivity.
 - ▶ Add edges to correct parity.
- Alternatively:**
- ▶ Take all edges of nontrivial ears.

Theorem

The new algorithm yields a tour with at most $\frac{3}{2}L - \pi$ edges, where L is a lower bound on the number of edges in any 2ECSS, and π is the number of pendant ears (after optimization).

Alternative yields an 2ECSS with at most $\frac{5}{4}L + \frac{1}{2}\pi$ edges.

→ The better of the two 2ECSSs has at most $\frac{4}{3}L$ edges.

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New algorithm for 2ECSS: proof of main theorem

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} $\frac{1}{2}(|V_0| - 1 + \varphi_0(G))$ edges.

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Proof: Since $|V_0| \leq |V(G)| + \varphi_\pi(G) - 2\pi_{\text{short}} - 4\pi_{\text{long}}$,
correcting parity needs at most $\frac{1}{2}L_\varphi - \pi - \pi_{\text{long}}$ edges.



New algorithm for TSP

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New algorithm for TSP

- ▶ Compute a nice ear-decomposition.
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Theorem

Mömke-Svensson yields a tour with at most $\frac{4}{3}L + \frac{2}{3}\pi$ edges.

→ The better of the two tours has at most $\frac{7}{5}L$ edges.

The Mömke-Svensson lemma

Definition (Mömke and Svensson [2011])

Let G be a 2-vertex-connected graph.

A **removable pairing** in G consists of a set R of **removable edges** and a set of pairwise disjoint **pairs** of elements of R such that

- ▶ deleting any edge set $S \subseteq R$ that contains at most one edge out of each pair does not disconnect the graph
- ▶ the two edges of any pair share a vertex, and this vertex is incident to another edge

Theorem (Mömke and Svensson [2011])

Given a 2-vertex-connected graph G and a removable pairing (R, \mathcal{P}) . Then one can find a tour with at most $\frac{4}{3}|E(G)| - \frac{2}{3}|R|$ edges in polynomial time.

Mömke-Svensson applied to ear-decompositions

Theorem

Given a 2-vertex-connected graph G with an ear-decomposition in which all ears are nontrivial. Then one can find a tour with at most $\frac{4}{3}(|V(G)| - 1) + \frac{2}{3}\pi$ edges in polynomial time.

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Proof: We define the **removable pairing** as follows:

- ▶ For each pendant ear, R will contain exactly one of its edges.
- ▶ For each non-pendant ear, R will contain a pair of edges incident to an internal vertex that is endpoint of another ear.

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So $|R| = 2(|E(G)| - (|V(G)| - 1)) - \pi$.

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Proof of Mömke-Svensson lemma

Theorem (Mömke and Svensson [2011])

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Proof: Find an odd join J containing at most one edge of each pair.

Add a second copy of each edge in $J \setminus R$.

Delete the edges in $J \cap R$.

We get a tour with $|E(G)| + c(J)$ edges,

where $c(e) = -1$ for $e \in R$ and $c(e) = 1$ for $e \notin R$.

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Construct G' as follows. For each pair P with edges $\{u, v\}, \{v, w\}$, add a vertex v_P and an edge $\{v_P, v\}$ of weight 0, and replace the two edges in P by $\{u, v_P\}, \{v_P, w\}$.

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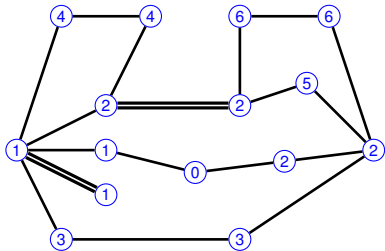
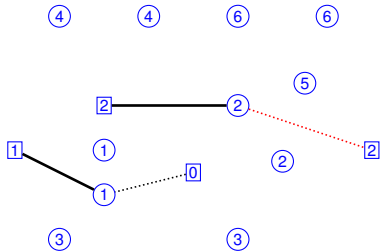
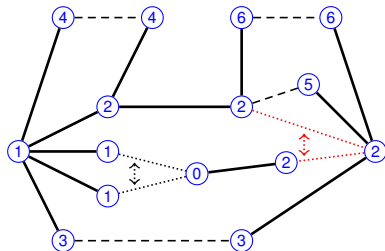
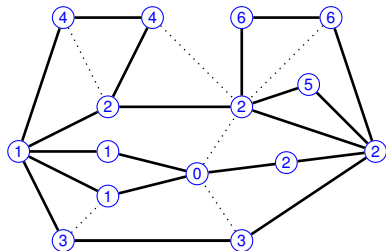
The all- $\frac{1}{3}$ -vector is in the odd join polytope of G' , and in its face defined by $x(\delta(v_P)) = 1$ for all pairs P .

Hence there is an odd join J as required with

$$c(J) \leq \frac{1}{3}c(E(G')) = \frac{1}{3}|E(G)| - \frac{2}{3}|R|.$$



Example: application of Mömke-Svensson lemma



New algorithm for TSP

- ▶ Compute a nice ear-decomposition.
- ▶ Optimize short ears so that they serve best for connectivity.
- ▶ Delete all 1-ears. In each of the resulting blocks:
- ▶ Take all edges of pendant ears. Alternatively:
- ▶ Add edges to obtain connectivity. ▶ Apply lemma of Mömke-Svensson.
- ▶ Add edges to correct parity.

Theorem

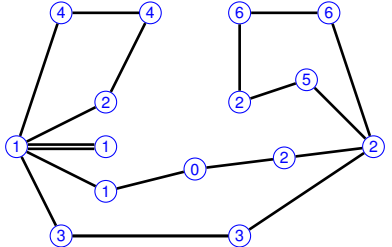
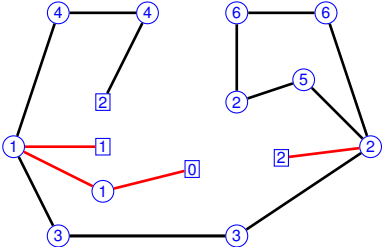
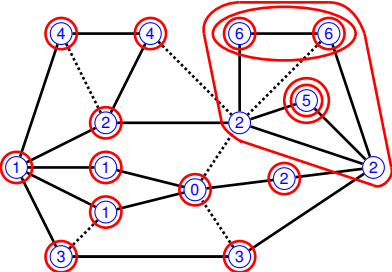
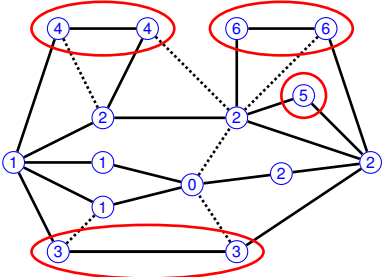
In each block, this algorithm yields a tour with at most $\frac{3}{2}L - \pi$ edges, where L is a lower bound on the number of edges in any 2ECSS, and π is the number of pendant ears (after optimization).

Theorem

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→ The better of the two tours has at most $\frac{7}{5}L$ edges.

Example: Shorter tour by nicer ears



New algorithm for TSP

- ▶ Compute a nice ear-decomposition.
- ▶ Optimize short ears so that they serve best for connectivity.
- ▶ Delete all 1-ears. In each of the resulting blocks:
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New algorithm for connected- T -joins

- ▶ Compute a nice ear-decomposition.
- ▶ Optimize **clean** ears so that they serve best for connectivity.

(**Clean** ears are short ears without an internal vertex in T .)

- ▶ Take all edges of clean ears.
- ▶ Apply ear induction to pendant but not clean ears.
- ▶ Add edges to obtain connectivity.
- ▶ Add edges to correct parity.

Alternatively:

- ▶ Apply ear induction to all ears.

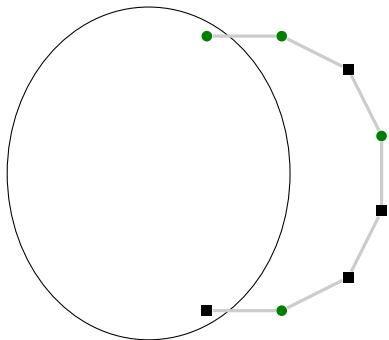
Theorem

The new algorithm yields a connected- T -join with at most $\frac{3}{2}L + \frac{1}{2}\varphi(G) - \pi$ edges, where L is a lower bound and π is the number of pendant ears (after optimization).

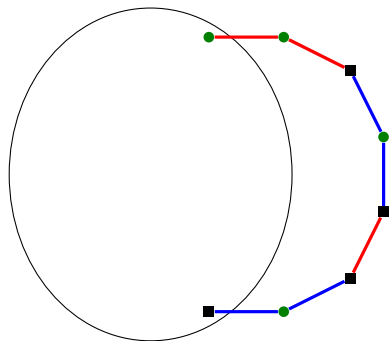
Alternative yields at most $\frac{3}{2}L - \frac{1}{2}\varphi(G) + \pi$ edges.

→ The better of the two has at most $\frac{3}{2}L$ edges.

Tool for connected- T -joins: ear induction

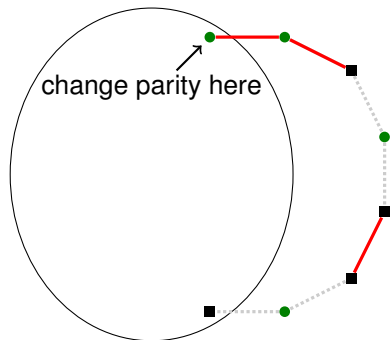


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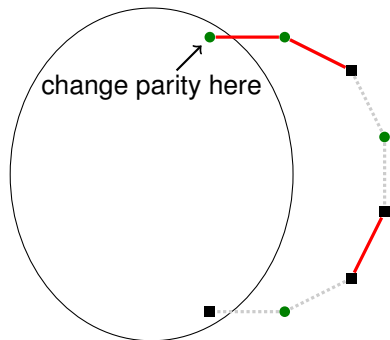
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Tool for connected- T -joins: ear induction



- ▶ Split pendant ear at vertices in T (that have wrong parity so far)
- ▶ Take smaller part for obtaining a T -join

Tool for connected- T -joins: ear induction

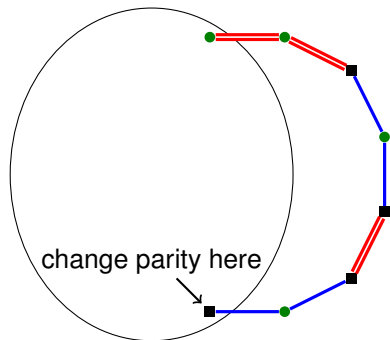


- ▶ Split pendant ear at vertices in T (that have wrong parity so far)
- ▶ Take smaller part for obtaining a T -join

This yields a

- ▶ T -join with $\leq \frac{1}{2}(|V(G)| - 1 + k_{\text{even}})$ edges

Tool for connected- T -joins: ear induction

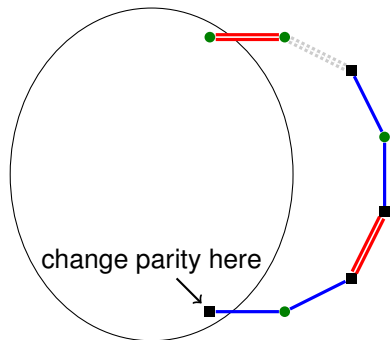


- ▶ Split pendant ear at vertices in T (that have wrong parity so far)
- ▶ Take smaller part for obtaining a T -join
- ▶ Double smaller part for obtaining a connected- T -join

This yields a

- ▶ T -join with $\leq \frac{1}{2}(|V(G)| - 1 + k_{\text{even}})$ edges

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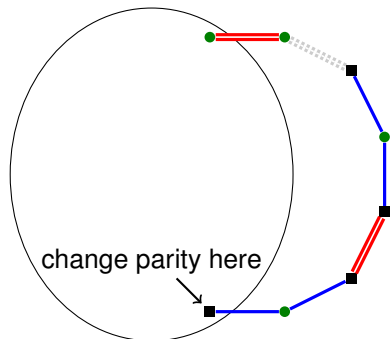


- ▶ Split pendant ear at vertices in T (that have wrong parity so far)
- ▶ Take smaller part for obtaining a T -join
- ▶ Double smaller part for obtaining a connected- T -join
- ▶ May delete one pair of parallel edges (if there is one)

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This yields a

- ▶ T -join with $\leq \frac{1}{2}(|V(G)| - 1 + k_{\text{even}})$ edges
- ▶ connected T -join with $\leq \frac{3}{2}(|V(G)| - 1) + \pi_{\text{clean}} - \frac{1}{2}k_{\text{even}} - k_{\text{odd}}$ edges

$$\leq \frac{3}{2}(|V(G)| - 1) + \frac{1}{2}k_{\text{even}} - \pi_{\text{notclean}} \quad \text{and} \quad \leq \frac{3}{2}(|V(G)| - 1) - \frac{1}{2}k_{\text{even}} + \pi$$

New algorithm for connected- T -joins

- ▶ Compute a nice ear-decomposition.
- ▶ Optimize **clean** ears so that they serve best for connectivity.

Clean ears are short ears without an internal vertex in T .

- ▶ Take all edges of clean ears.
- ▶ Apply ear induction to pendant but not clean ears.
- ▶ Add edges to obtain connectivity.
- ▶ Add edges to correct parity.

Alternatively:

- ▶ Apply ear induction to all ears.

Theorem

The new algorithm yields a connected- T -join with at most $\frac{3}{2}L + \frac{1}{2}\varphi(G) - \pi$ edges, where L is a lower bound and π is the number of pendant ears (after optimization).

Alternative yields at most $\frac{3}{2}L - \frac{1}{2}\varphi(G) + \pi$ edges.

→ The better of the two has at most $\frac{3}{2}L$ edges.

Using nicer ears, more generally and formally

Definition

Let G be a graph with a nice ear-decomposition, $T \subseteq V(G)$, $|T|$ even. An ear is **clean** if it is short and T contains none of its internal vertices. Let M contain for each clean ear the set of its internal vertices.

Theorem

Let G, T, M be as above. Suppose the given ear-decomposition contains a maximum earmuff for M (of cardinality μ).

Then a connected- T -join of cardinality at most $L_\mu + \frac{1}{2}L_\varphi - \pi$ can be constructed in $O(|V(G)|^3)$ time.

Recall:

π = number of pendant ears

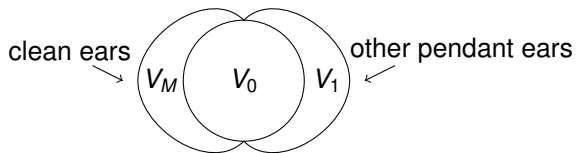
L_φ = $|V(G)| - 1 + \varphi(G)$

L_μ = $|V(G)| - 1 + |M| - \mu$

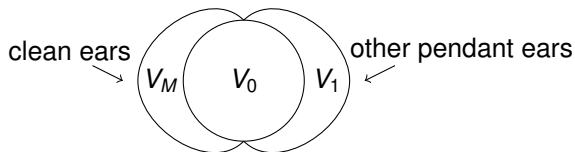
L_μ is a lower bound on the optimum (in fact, on the LP value).

L_φ is a lower bound if $T = \emptyset$.

Proof



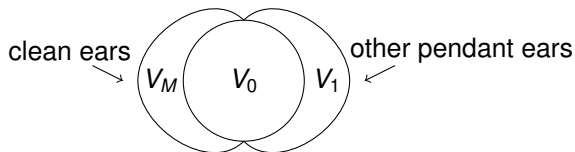
Proof



- ▶ take all edges of clean ears

$$\frac{3}{2}|V_M| + \frac{1}{2}\varphi_M$$

Proof

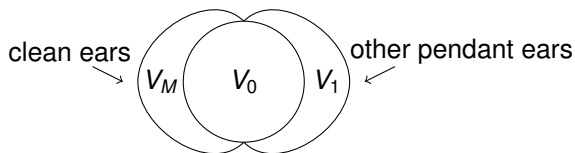


- ▶ take all edges of clean ears
- ▶ add edges of $G[V_0]$ so that $(V_M \cup V_0, E_1 \cup E_2)$ is connected

$$\frac{3}{2}|V_M| + \frac{1}{2}\varphi_M$$

$$|V_0| - 1 - \mu$$

Proof



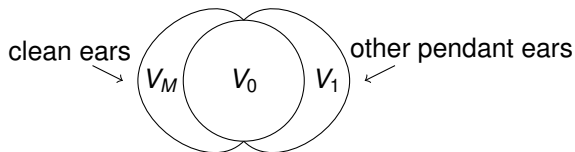
- ▶ take all edges of clean ears
- ▶ add edges of $G[V_0]$ so that $(V_M \cup V_0, E_1 \cup E_2)$ is connected
- ▶ apply ear induction to non-clean pendant ears

$$\frac{3}{2}|V_M| + \frac{1}{2}\varphi_M$$

$$|V_0| - 1 - \mu$$

$$\frac{3}{2}|V_1| + \frac{1}{2}\varphi_1 - (\pi - |M|)$$

Proof



- ▶ take all edges of clean ears
- ▶ add edges of $G[V_0]$ so that $(V_M \cup V_0, E_1 \cup E_2)$ is connected
- ▶ apply ear induction to non-clean pendant ears
- ▶ Correct parities on $G[V_0]$

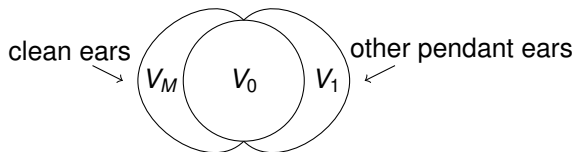
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$$\frac{1}{2}(|V_0| + \varphi_0 - 1)$$

Proof



- ▶ take all edges of clean ears $\frac{3}{2}|V_M| + \frac{1}{2}\varphi_M$
- ▶ add edges of $G[V_0]$ so that $(V_M \cup V_0, E_1 \cup E_2)$ is connected $|V_0| - 1 - \mu$
- ▶ apply ear induction to non-clean pendant ears $\frac{3}{2}|V_1| + \frac{1}{2}\varphi_1 - (\pi - |M|)$
- ▶ Correct parities on $G[V_0]$ $\frac{1}{2}(|V_0| + \varphi_0 - 1)$

Adding up, using $\varphi(G) = \varphi_0 + \varphi_1 + \varphi_M$, yields

$$\frac{3}{2}(|V(G)| - 1) + |M| - \mu + \frac{1}{2}\varphi(G) - \pi = L_\mu + \frac{1}{2}L_\varphi - \pi$$



Second solution to earmuff maximization

Let M be a earmuff with $V_M \cap T = \emptyset$ (e.g., contain for each clean ear the set of its internal vertices).

- ▶ Let $U = V(G) \setminus V_M$, where $V_M = \bigcup M$.
- ▶ Represent \mathcal{P}_f ($f \in M$) by the set U_f of its endpoints.
- ▶ Sufficient to find a maximum cardinality subset $F \subseteq M$ with a **forest representative system** $(e_f)_{f \in F}$ of $(U_f)_{f \in F}$, i.e.,
 - ▶ $e_f \in \binom{U_f}{2}$ for all $f \in F$,
 - ▶ $e_f \neq e_{f'}$ for $f \neq f'$, and
 - ▶ the graph $(U, \{e_f : f \in F\})$ is a forest.

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Theorem (\approx Lovász [1970])

This maximum is $\mu =$

$$\min \left\{ |M| - \sum_{W \in \mathcal{W}} (|\{f \in M : U_f \subseteq W\}| - (|W| - 1)) : \mathcal{W} \text{ is a partition of } U \right\}.$$

(This holds for any finite sets $U, M, (U_f)_{f \in M}$ with $\emptyset \neq U_f \subseteq U$.)

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Note: We have an algorithm with running time $O(|V(G)||E(G)|)$

Lower bounds and LP relaxations

Theorem (Cheriyán, Sebő, Szigeti [2001])

$$L_\varphi := |V(G)| - 1 + \varphi(G) \leq \text{LP}(G) :=$$
$$\min \left\{ x(E(G)) : x \in \mathbb{R}_{\geq 0}^{E(G)}, x(\delta(W)) \geq 2 \text{ for all } \emptyset \neq W \subset V(G) \right\}$$

Proof (Sketch): Frank's theorem $\Rightarrow T \Rightarrow$ 2-packing of T -cuts. \square

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Proof (Sketch): Frank's theorem $\Rightarrow T \Rightarrow 2$ -packing of T -cuts. \square

Theorem

$$L_\mu := |V(G)| - 1 + |M| - \mu \leq \text{LP}(G, T) := \\ \min \left\{ x(E(G)) : x \in \mathbb{R}_{\geq 0}^{E(G)}, \right. \\ \left. x(\delta(W)) \geq 2 \text{ for all } \emptyset \neq W \subset V(G) \text{ with } |W \cap T| \text{ even,} \right. \\ \left. x(\delta(\mathcal{W})) \geq |\mathcal{W}| - 1 \text{ for all partitions } \mathcal{W} \text{ of } V(G) \right\}$$

Proof (Sketch): 2-packing of cuts, including the partition from min-max theorem for forest representation systems. \square

Summary of Results

We obtained an improved approximation ratio of:

- ▶ $\frac{7}{5}$ for **Graphic TSP**
 - ▶ $\frac{3}{2}$ for **Connected- T -join**
 - ▶ $\frac{4}{3}$ for **2ECSS**
-
- ▶ All algorithms combinatorial, running time $O(|V(G)|^3)$
 - ▶ These bounds are tight.
 - ▶ These are also upper bounds on the integrality ratios of the natural LPs for unit weights.

Open problems

- ▶ improve approximation ratios (to $\frac{4}{3}$ for Graphic TSP?)
- ▶ extend to weights (general metrics)
- ▶ extend to directed graphs

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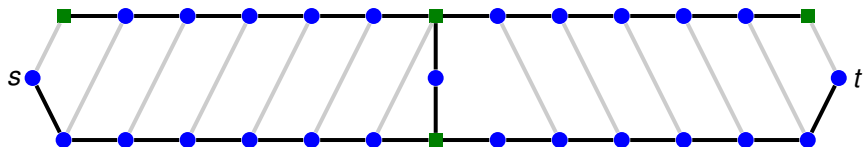
Thank you!

Open problems

- ▶ improve approximation ratios (to $\frac{4}{3}$ for Graphic TSP?)
- ▶ extend to weights (general metrics)
- ▶ extend to directed graphs

Thank you!

Tight example for connected- T -join



$$|V(G)| = 8k + 5$$

(Here $k = 3$.)

$$T = \{s, t\}$$

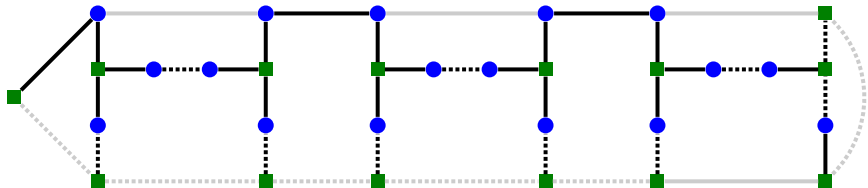
$$OPT = 8k + 4$$

$$\varphi(G) = 2$$

$$\pi = 1 = \frac{1}{2}\varphi(G).$$

Algorithm computes solution with $12k + 6$ edges.

Tight example for graphic TSP



$$|V(G)| = OPT = 10k + 1$$

(Here $k = 3$.)

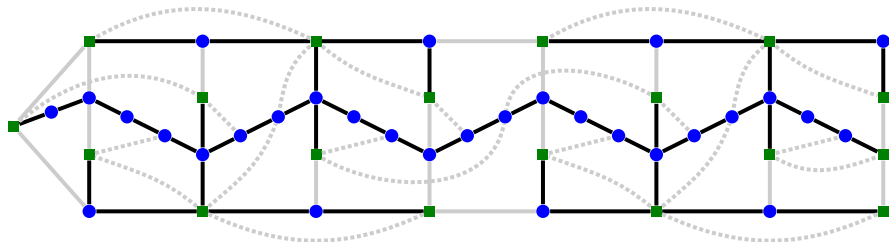
$$\varphi(G) = 0$$

$$L = 10k$$

$$\pi = k = \frac{1}{10}L.$$

Algorithm computes solution with $14k$ edges.

Tight example for 2ECSS



$$L = |V(G)| = OPT = 24k$$

(Here $k = 2$.)

$$\varphi(G) = 1$$

$$\pi = 4k = \frac{1}{6}L.$$

Algorithm computes solution with $32k - 1$ edges.