

**A Centralized Local Algorithm
for the Sparse Spanning Graph Problem**

Christoph Lenzen, Reut Levi

**A Sublinear Tester for Outerplanarity
(and Other Forbidden Minors) With One-Sided Error**

Hendrik Fichtenberger, Reut Levi,
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Local Graph Algorithms

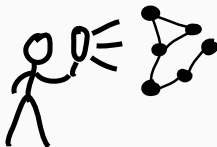


classic / global algorithm
see whole input, $\Omega(n)$ time
output solution



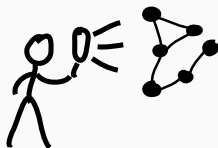
local algorithm
see only small parts, $o(n)$ time
provide query access to solution

The Local Sparse Spanning Graph Problem (LSSG)



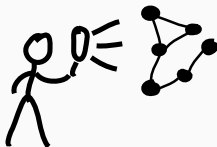
- bounded degree graph $G = (V, E)$ given, $V = [n]$
- LSSG algorithm provides query access to a spanning graph $G' = (V, E')$: “is $(7, 18) \in E'$?”

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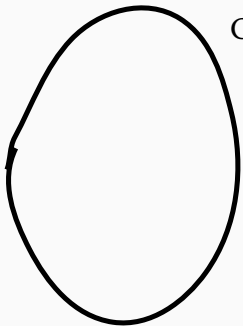
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Main Result

An LSSG algorithm with query and time complexity

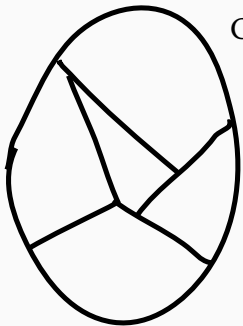
$\tilde{O}(n^{2/3}) \cdot \text{poly}(1/\epsilon)$ per query. It guarantees $|E'| \leq (1 + \epsilon)n$ w.h.p.

Graph Partitions



Given a graph G ,

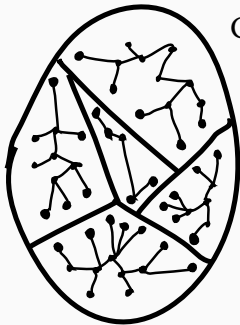
Graph Partitions



Given a graph G ,

1. partition G into small parts

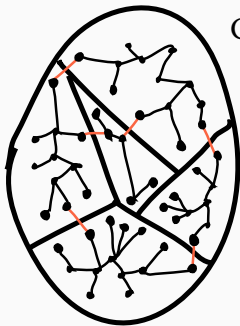
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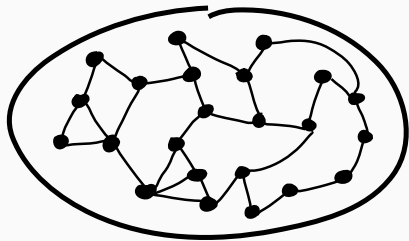
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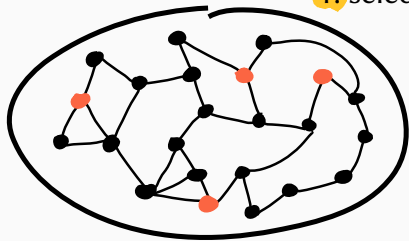
1. partition G into small parts
2. compute spanning tree inside of the parts
3. add n edges between parts to make graph connected

Voronoi Partitions

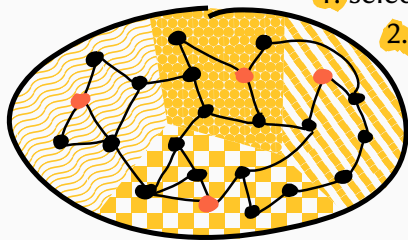


Voronoi Partitions

1. select $\Theta(n^{2/3})$ random centers



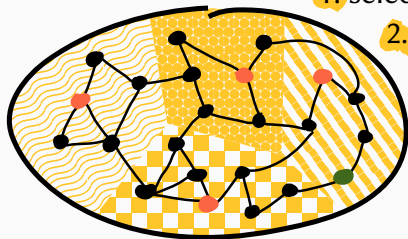
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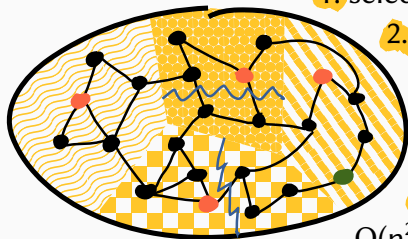


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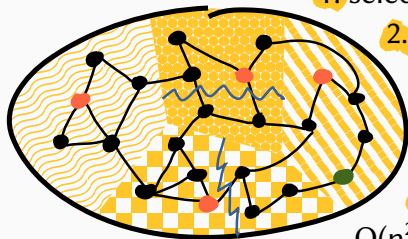
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summary: each core cluster has

- a BFS spanning tree ✓
- diameter $O(\log n)$ ✓
- size $O(n^{1/3})$ ✓

Local Construction of Core Clusters

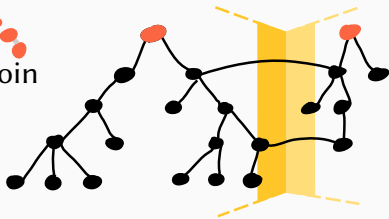


Local Construction of Core Clusters

1. each vertex flips a coin
- 
- The diagram shows a network graph with several vertices and edges. The vertices are represented by black and red circles. The edges are represented by grey lines. A single red circle is shown to the right of the main graph structure.

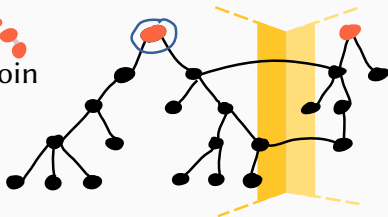
Local Construction of Core Clusters

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2. BFS exploration



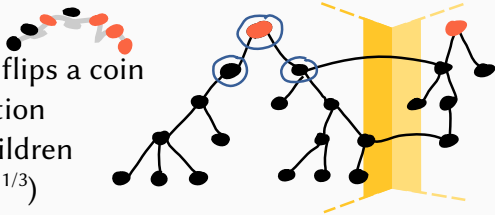
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- complexity: $O(n^{1/3})$



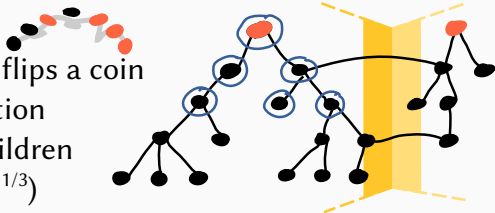
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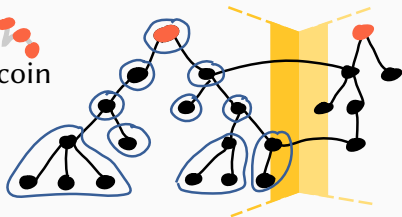
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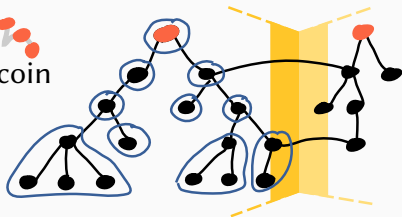
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Not in This Talk

- Partitioning of remote vertices into *remote clusters*
- Joining core clusters to reduce number of cluster pairs (\approx number of edges needed to connect clusters) to ϵn

**A Sublinear Tester for
Outerplanarity (& Other Forbidden
Minors) With One-Sided Error**

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Sublinear Graph Algorithms



classic / global algorithm

see everything, $\Omega(n)$ time

output solution

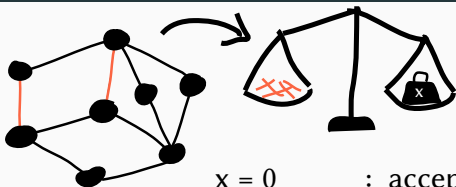


sublinear algorithm

see only small parts, $o(n)$ time

estimate solution's value

Testing Outerplanarity With One-Sided Error



$x = 0$: accept always

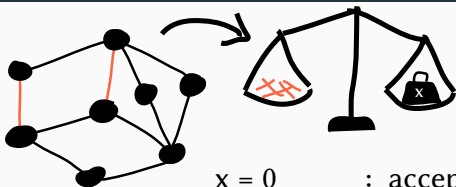
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#● < d

Testing Outerplanarity With One-Sided Error



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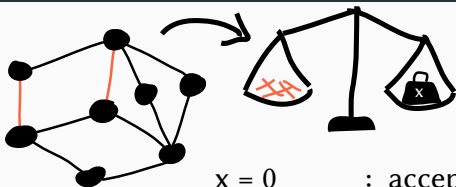
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\uparrow
 ϵ -far

\uparrow
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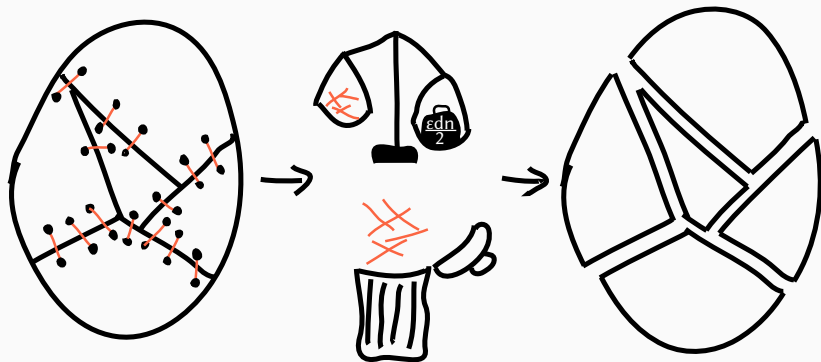
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Main Result

An \mathcal{F} -minor freeness tester for every family \mathcal{F} of forbidden minors that contains either the $K_{2,k}$, $(k \times 2)$ -grid or k -circus graph with query complexity / running time $\tilde{O}(n^{2/3}/\epsilon^5)$

Partitioning Revisited

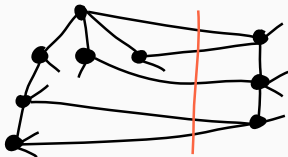


How about the cut in Voronoi partitions?

- number of cut edges involving a remote cluster is $\leq \epsilon dn/4$
- number of cut edges between core clusters might be $> \epsilon dn/4$

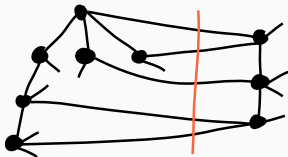
Construction of $K_{2,k}$

Theorem: cuts of size $> f$ between clusters imply $K_{2,k}$ -minors



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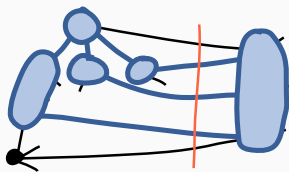
idea:

always have BFS tree,
enforce more structure
by large cut size

$$f \approx \Theta(d k \log(n))$$

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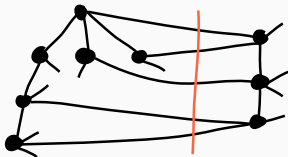
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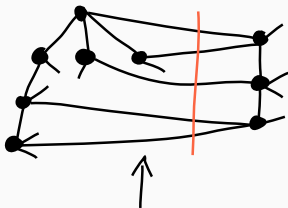
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$(k+1) \cdot \log(n)$ cut vertices on left side

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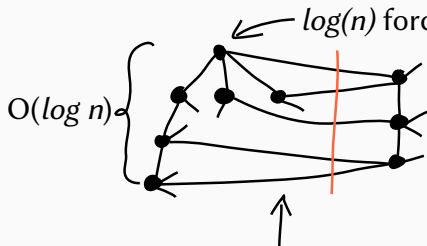
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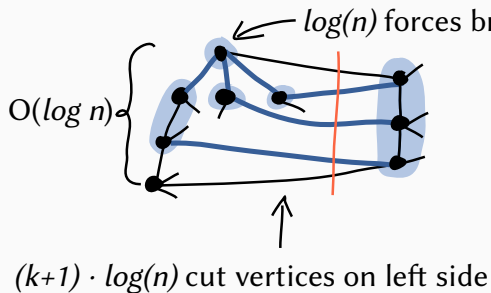
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Summary

Result: Local Spanning Graphs

An LSSG algorithm with query and time complexity $\tilde{O}(n^{2/3}) \cdot \text{poly}(1/\epsilon)$ per query. It guarantees $|E'| \leq (1 + \epsilon)n$ w.h.p.

Result: Minor-Freeness Testing

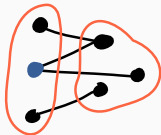
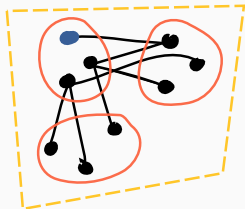
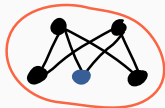
An \mathcal{F} -minor freeness tester for every family \mathcal{F} of forbidden minors that contains either the $K_{2,k}$, $(k \times 2)$ -grid or k -circus graph with query complexity / running time $\tilde{O}(n^{2/3}/\epsilon^5)$

recent progress by Kumar et al. (2018) for arbitrary \mathcal{F} : $O(n^{1/2+o(1)})$

Additional Slides

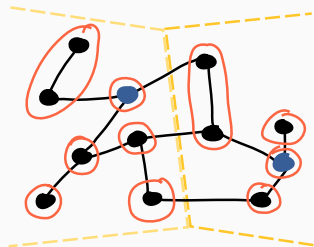
Algorithm

1. sample $O(f / \epsilon)$ edges
2. for every sampled edge (u,v) :
 - i) explore **cluster(s)** of u,v
 - ii) compute cut sizes between core cluster and remaining **Voronoi cell** of u,v
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3. reject iff minor found or some cut $> f$



Super Clusters

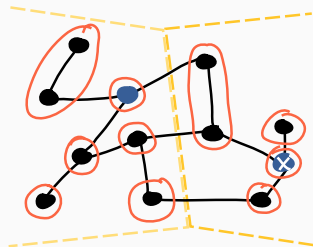
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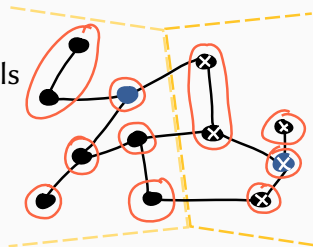
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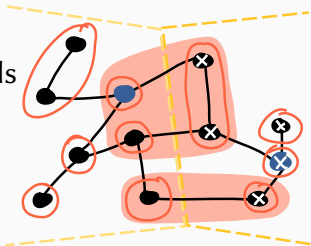
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1. mark each **Voronoi cell** w.p. $1/n^{1/3}$
2. mark each **core cluster** of marked cells
3. **join** unmarked core clusters with marked neighboring core clusters



Super Clusters

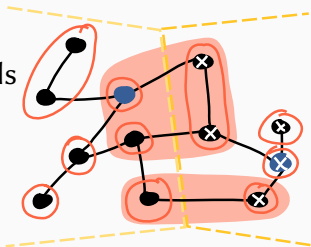
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~~locally reconstructable~~

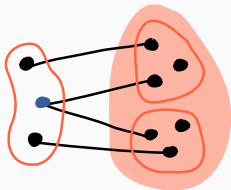
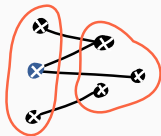
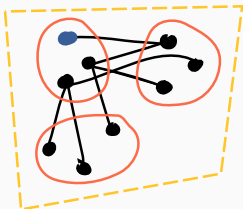
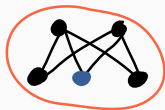
local membership queries

$f \cdot \#(\text{core clusters}) \cdot \#(\text{super clusters}) \in O(\epsilon dn)$

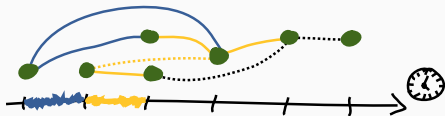


Tester With Super Clusters

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Remote Clusters [Elkin, Neiman, 2017]



1. each remote vertex picks random delay
2. after delay, start BFS: one level per time
3. construct *remote* clusters from BFS