

# Linear-Time Computation of a Linear Kernel for Dominating Set on Planar Graphs

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joint work with

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# Dominating Set

## DOMINATING SET

**Input:** A graph  $G = (V, E)$  and a natural number  $k$ .

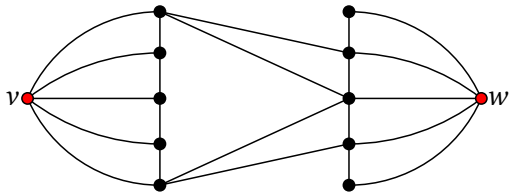
**Question:** Is there a dominating set  $D \subseteq V$  with  $|D| \leq k$  for  $G$ ?

Here,  $D$  is a dominating set  $\iff V \subseteq N[D]$ .

Size of a minimum dominating set for  $G$  is denoted by  $\gamma(G)$ .

DOMINATING SET is NP-hard even when restricted to planar graphs.

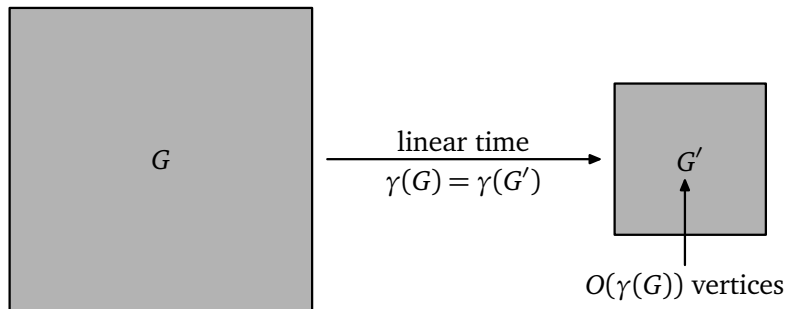
## Example for Dominating Set



The set  $\{v, w\}$  dominates the shown graph.

## Problem Kernels

We show an  $O(\gamma(G))$ -vertex problem kernel for DOMINATING SET on planar graphs that is computable in linear time.



## Why Problem Kernels Matter

In general, problem kernels yield equivalent smaller instances in polynomial time. Hence, used as a preprocessing step, they can

- ▶ speed up algorithms that exactly solve NP-hard problems,
- ▶ speed up slow but effective approximation algorithms,
- ▶ speed up slow but effective kernelization algorithms,
- ▶ speed up heuristics often used in practice.

Combining fast kernelization algorithms with effective ones yields small kernels fast.

In contrast, it is unclear how a fast approximation algorithms can be combined with an effective one to get an approximation algorithm that is both fast and effective.

# The Kernel Size Race

Problem kernels are currently the hottest topic in parameterized algorithms, resulting in a kernel size race.

## An Example:

**FEEDBACK VERTEX SET:** delete at most  $k$  vertices to transform a graph into a forest (make it cycle-free).

Burrage, Estivill-Castro, Fellows, and Langston, IWPEC 2006:  
 $O(k^{11})$ -vertex kernel

Bodlaender and van Dijk, TCS 2010:  
 $O(k^3)$ -vertex kernel

Thomassé, ACM Trans. Algorithms 2010:  
 $O(k^2)$ -vertex kernel

# The Kernel Size Race

Problem kernels are currently the hottest topic in parameterized algorithms, resulting in a kernel size race.

## Another Example:

**CLUSTER EDITING:** delete or add at most  $k$  edges to transform a graph into a disjoint union of cliques.

Gramm, Guo, Hüffner, and Niedermeier, TCS 2005:  
 $O(k^2)$ -vertex kernel

Fellows, Langston, Rosamond, and Shaw, FCT 2007:  
 $O(k)$ -vertex kernel

Guo, TCS 2009:  
 $4k$ -vertex kernel

Chen and Meng, COCOON 2010:  
 $2k$ -vertex kernel

# Problem Kernels for Dominating Set

## Focusing on Kernel Size:

Alber, Fellows, and Niedermeier, J. ACM, 2004:

$335\gamma$ -vertex kernel in  $O(n^3)$  time on planar graphs.

Chen, Fernau, Kanj, and Xia, SIAM J. Comput., 2007:

$67\gamma$ -vertex kernel in  $O(n^3)$  time on planar graphs.

## Focusing on Larger Graph Classes:

Fomin and Thilikos, ICALP 2004:

$O(\gamma + g)$ -vertex kernel in  $O(gn^3)$  time on graphs of genus  $g$ .

Philip, Raman, and Sikdar, ESA 2009:

$O(\gamma^{2(d+1)^2})$ -vertex kernel in  $O(2^d dn^2)$  time on  $d$ -degenerate graphs.

Fomin, Lokshantov, Saurabh, and Thilikos, SODA 2010:

$O(\gamma)$ -vertex kernel in polynomial time on apex-minor free graphs.



# Linear-Time Problem Kernel for Dominating Set

## Focusing on Running Time:

van Bevern, Hartung, Kammer, Niedermeier, Weller, Manuscript 2011 (submitted):  
 $O(\gamma)$ -vertex kernel in  $O(n)$  time.

## Small Kernel Fast:

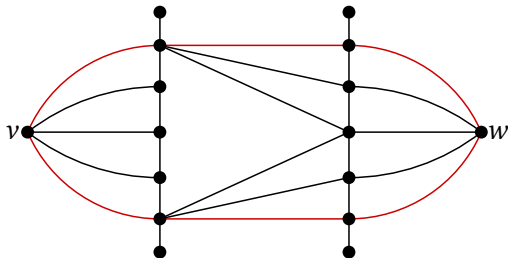
$67\gamma$ -vertex kernel in  $O(\gamma^3 + n)$  time using the above kernelization as preprocessing step for the following result:

Chen, Fernau, Kanj, Xia, SIAM J. Comput., 2007:

$67\gamma$ -vertex kernel in  $O(n^3)$  time on planar graphs.

# Tools for Obtaining Dominating Set Problem Kernels

Many results exploit a result by Alber, Fellows, and Niedermeier, J. ACM 2004: a graph is decomposable into  $O(\gamma)$  regions:

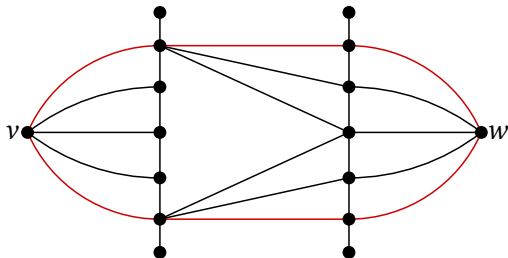


A region  $R(v, w)$  between two vertices  $v$  and  $w$  is a closed bounded subset of the plane such that:

1. the boundary of  $R(v, w)$  is formed by two simple paths between  $v$  and  $w$ , each of which has length at most three and
2. all vertices inside the region  $R(v, w)$  are from  $N[v] \cup N[w]$ .

# Problem Kernel Analysis of Alber et al., J. ACM, 2004

A graph is decomposable into  $O(\gamma)$  regions:



Problem kernel is obtained as follows:

1. shrink each region to  $O(1)$  vertices and
2. reduce number of vertices outside of regions to  $O(\gamma(G))$ .

## Our Problem Kernel

Re-use the region framework and data reduction rules by Alber et al. with slight modifications, but find and update the structures to be reduced in linear time.

### Example: Finding and Shrinking Regions

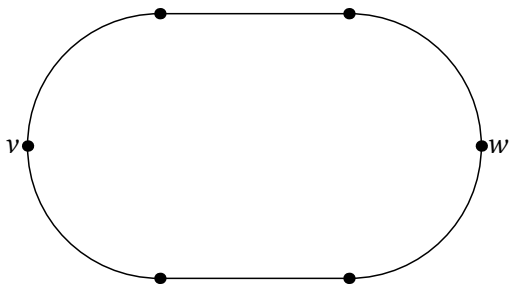
Problem: graph is decomposable into  $O(\gamma)$  regions, but we do not actually *have* them  $\rightsquigarrow$  shrink everything that *could* be a region.

Alber et al. delete vertices from  $N[v] \cup N[w]$  for all vertices  $v, w$ .  
 $\rightsquigarrow \Omega(n^2)$  time. Rules applied  $O(n)$  times  $\rightsquigarrow O(n^3)$  time.

Assuming that we know how to shrink regions: how to find them?

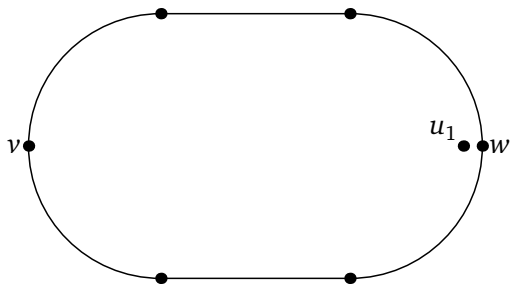
- ▶ Largely exploit the restricted structure of regions and
- ▶ know when to stop: do not apply data reduction rules more often than necessary to obtain the problem kernel.

## Shrinking Regions in Linear Time



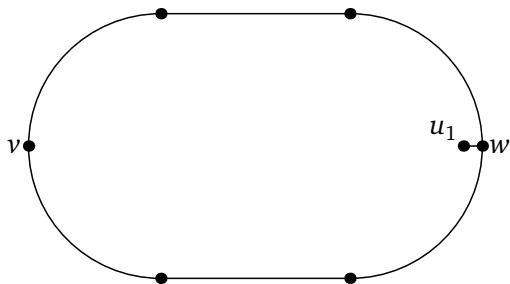
Consider this region  $R(v, w)$ .

## Shrinking Regions in Linear Time



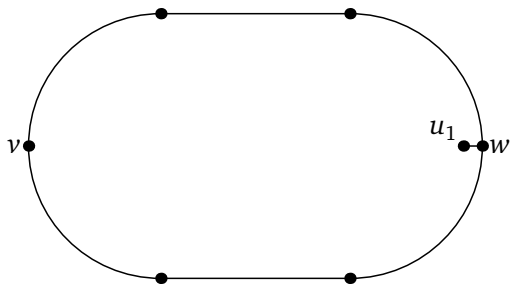
Assume that  $u_1 \in R(v, w)$ . Then,  $u_1 \in N[v, w]$ .

## Shrinking Regions in Linear Time



W.l. o. g., assume that  $u_1 \in N(w)$ .

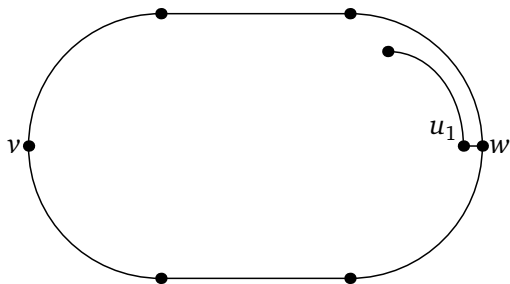
## Shrinking Regions in Linear Time



If  $\deg(u_1) \leq 1$ : delete  $u_1$  and remember  $w$  to be dominating.

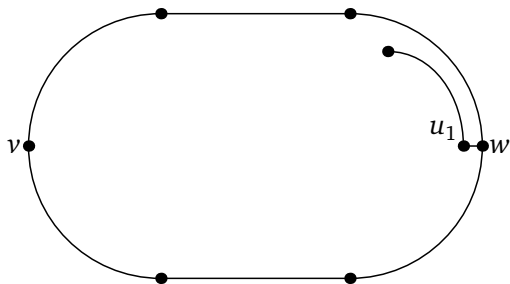


## Shrinking Regions in Linear Time



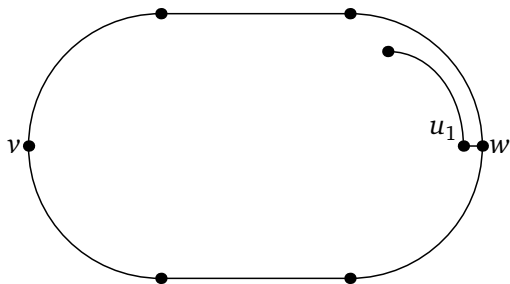
Hence,  $u_1$  has a neighbor, which, due to planarity, is in  $R(v, w)$ .

## Shrinking Regions in Linear Time



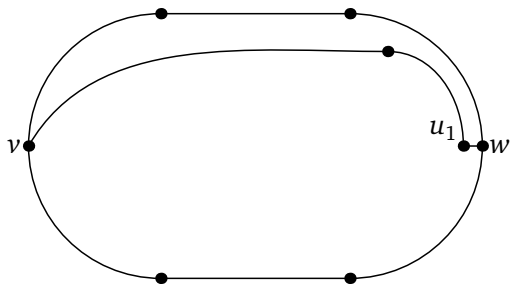
If  $N[N[u_1]] \subseteq N[w]$ , delete  $u_1$  (remember  $w$  to be dominating).

## Shrinking Regions in Linear Time



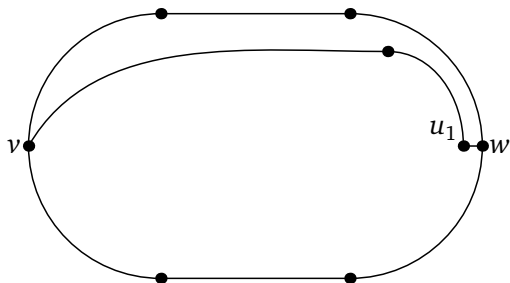
Assume that  $u_1$ 's neighbor is nonadjacent to  $w$ .

## Shrinking Regions in Linear Time



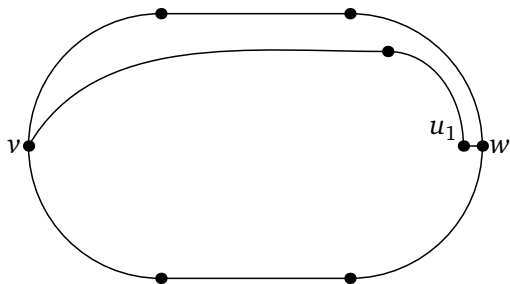
Then,  $u_1$ 's neighbor is adjacent to  $v$ .

## Shrinking Regions in Linear Time



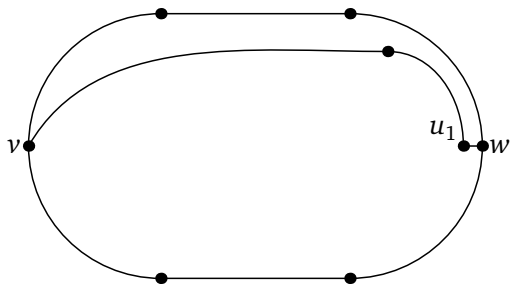
Observation: inner vertices of  $R(v, w)$  lie on short  $v$ - $w$ -paths.

## Shrinking Regions in Linear Time



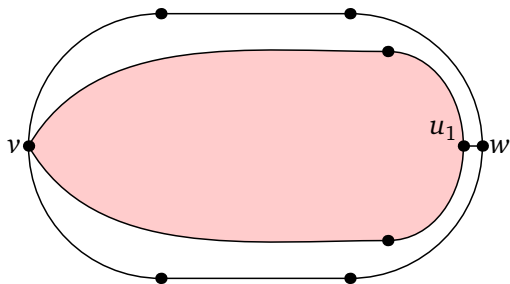
Can quickly list such paths if vertices of  $R(v, w)$  have small degree.

## Shrinking Regions in Linear Time



If  $u_1$  has large degree, then  $N(u_1) \cap N(v)$  or  $N(u_1) \cap N(w)$  is large.

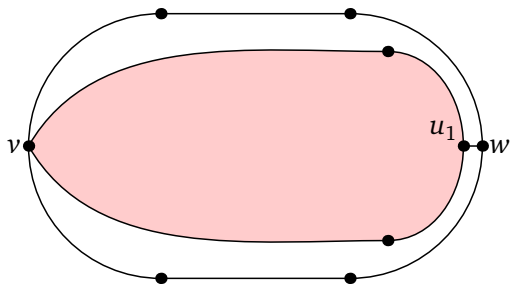
## Shrinking Regions in Linear Time



Assume that  $N(u_1) \cap N(v)$  is large  $\rightsquigarrow$  red sub-region  $R(v, u_1)$ .

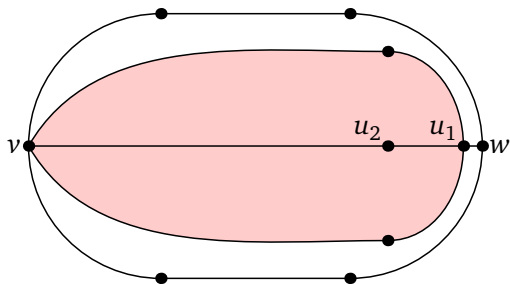


## Shrinking Regions in Linear Time



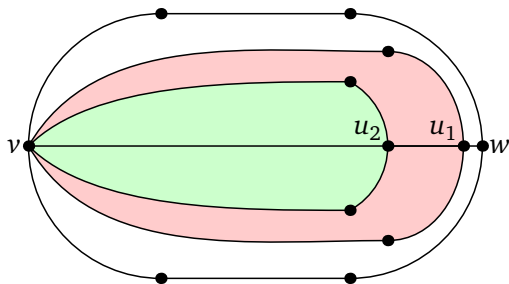
Can shrink  $R(v, u_1)$  if it only contains vertices with small degree.

## Shrinking Regions in Linear Time



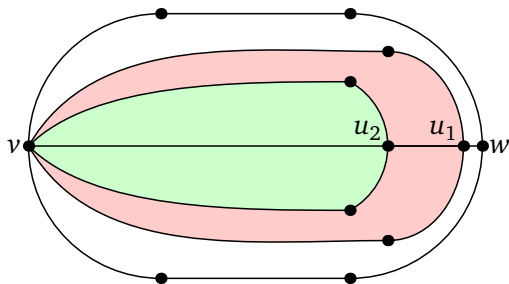
Assume that  $u_2 \in R(v, u_1)$ . By planarity,  $|N(u_2) \cap N(1)| \in O(1)$ .

## Shrinking Regions in Linear Time



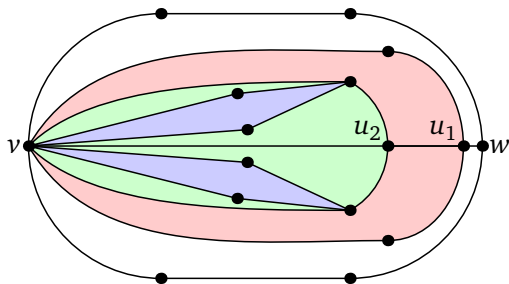
Hence, if  $u_2$  has large degree, then  $N(u_2) \cap N(v)$  (green) is large.

## Shrinking Regions in Linear Time



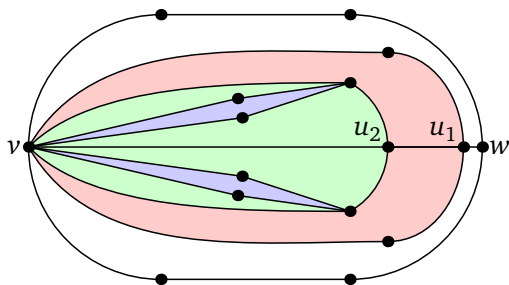
Can shrink  $R(v, u_2)$  if it only contains vertices with small degree.

## Shrinking Regions in Linear Time



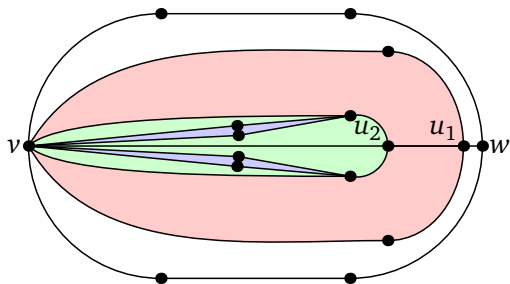
This is the case since vertices in blue regions are safely deletable.

## Shrinking Regions in Linear Time



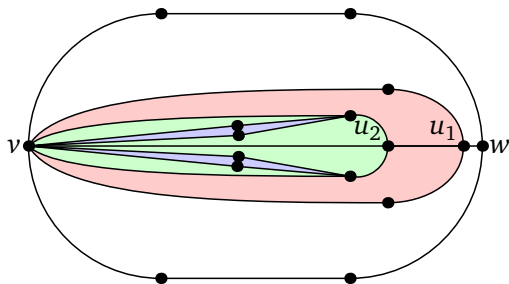
Shrink blue regions  $\rightsquigarrow$  vertices in  $R(v, u_2)$  have low degree.

## Shrinking Regions in Linear Time



Shrink  $R(v, u_2) \rightsquigarrow$  vertices in  $R(v, u_1)$  have low degree.

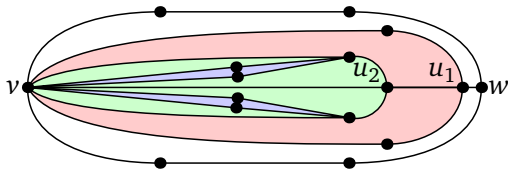
## Shrinking Regions in Linear Time



Shrink  $R(v, u_1) \rightsquigarrow$  vertices in  $R(v, w)$  have low degree.



## Shrinking Regions in Linear Time



We can now find and shrink  $R(v, w)$ .

# Shrinking Regions in Linear Time

Three times, basically apply the following algorithm:

1. Preprocess (e. g. degree-one deletion rule)
2. Find deletion candidates regions:
  - 2.1 For each  $v \in V$ , in  $O(\deg(v))$  time, list all short paths starting at  $v$ , having length at most four, and only crossing vertices with constant degree.
  - 2.2 For each such path  $p$ , let  $v_p, w_p$  be its endpoints.
  - 2.3 If  $p$  contains only vertices of  $N[v_p] \cup N[w_p]$ , then  $p$  is a deletion candidate (could be part of a region to shrink).
3. Shrink regions, which contain  $O(n)$  deletion candidates in total, since only  $\sum_{v \in V} \deg(v) = O(n)$  paths are generated.

## Conclusion

Not only the size of problem kernels, but also the running time of kernelization algorithms has to be optimized.

Optimize one of size and speed; get the other for free:

- ▶ can combine speed-optimized and size-optimized kernelization algorithms to obtain small kernels fast
- ↪  $67\gamma$ -vertex kernel in  $O(\gamma^3 + n)$  time for DOMINATING SET in planar graphs.<sup>1</sup>
- ↪ faster algorithms for solving hard problems.

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<sup>1</sup>Executing the algorithm by Chen et al. (SIAM J. Comput., 2007) after ours