

# Dilation of regular polygons

## Algorithmic aspects

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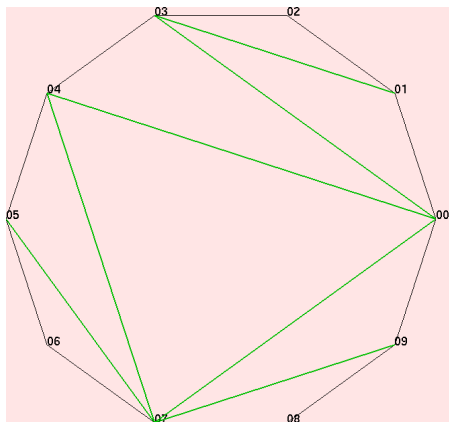
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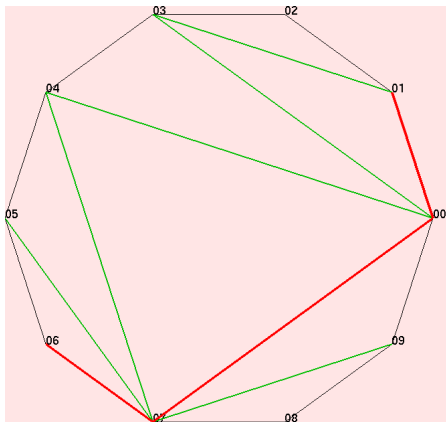


# Example of a triangulation



A triangulation  $T$  of a 10-gon. Corresponding dilation:  $\text{dil}(T) = 1.42705098$

# Example of a triangulation



The path between a critical pair for this triangulation is shown in red.

$$\text{dil}(T) = \frac{\text{total length of the red path}}{\text{euclidean distance between the endpoints}}$$



# Combinatorial explosion

- Straightforward algorithm: iterate over all possible triangulations  $T$  (see e.g. Mulzer (2004)).
- Impossible for  $n \geq 25$ : the number of triangulations of a  $n$ -gon is equal to the Catalan number  $C_{n-2}$ , where

$$C_k = \frac{1}{k+1} \cdot \binom{2k}{k}$$

( $C_{23} = 343.059.613.650$ )

→ **combinatorial explosion**







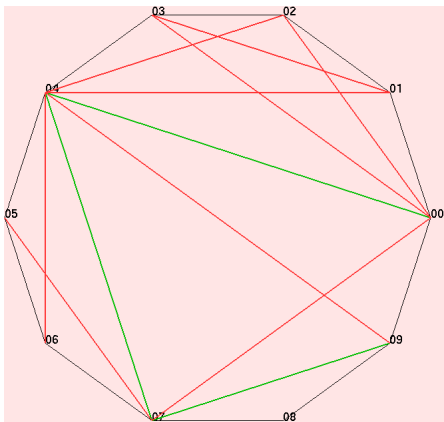




# Graphs with cliques

- Given  $P$ , we are interested in *all* triangulations containing  $P$ .
- The graph  $GC_P$  is obtained by taking all segments between points of  $S$  which do not intersect segments of  $P$ .
- “Duality”: for  $T \in \mathcal{T}$ ,  $P \subseteq T \Leftrightarrow T \subseteq GC_P$

# A graph with cliques $GC_P$



10-gon, three segments in  $P$  (shown in green),  $GC_P$ : green and red segments  
 $n_{lb}(P) = 1.42705098$

# Lower bound from a partial configuration

- Given  $P$ , “naive” lower bound on the dilation of *all* triangulations containing  $P$  given by

$$\text{nlb}(P) := \max_{\substack{p, q \in S \\ p \neq q}} \frac{d_{GC_P}(p, q)}{d_{\text{Euclidean}}(p, q)}$$

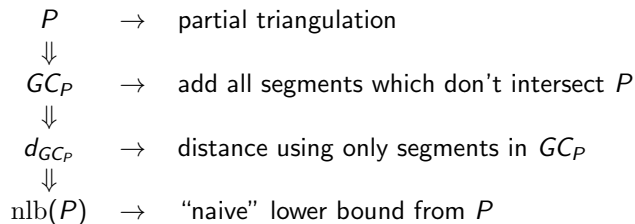
- Monotonicity:

$$P \subseteq P' \Rightarrow \text{nlb}(P) \leq \text{nlb}(P')$$

- If  $T \in \mathcal{T}$  is a triangulation,

$$\text{nlb}(T) = \text{dil}(T)$$

# Summary of the “naive” lower bound technique



# The lower bound technique

- We want a better bound  $\text{lb}(P)$  with

$$\text{nlb}(P) \leq \text{lb}(P) \leq \min_{\substack{T \in \mathcal{T} \\ P \subseteq T}} \text{dil}(T)$$

- We use  $GC_P$  (as for  $\text{nlb}$ ).

# Pairs of pairs of points

- Idea of nlb: use the inequality

$$d_{GC_p}(p, q) \leq d_{\text{Graph of } \mathcal{T}}(p, q)$$

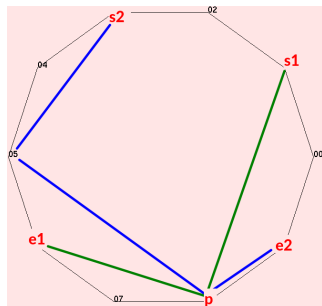
for a fixed pair of points  $p, q \in S, p \neq q$ .

- Problem: *pairs of points are considered independently.*
- Solution (inspired by Dumitrescu and Ghosh (2016)): *consider two pairs of points at once.*



# Pairs of pairs of points

Simple observation: if  $s_1, s_2, e_1, e_2 \in S$  are distinct points in clockwise order, then the paths from  $s_1$  to  $e_1$  and from  $s_2$  to  $e_2$  must intersect at some point  $p \in S$ .



# Pairs of pairs of points

- We have no idea of which  $p$  is optimal  $\rightarrow$  take the one which gives the lowest bound.
- The bound  $\text{lb}(s_1, s_2, e_1, e_2)$  associated to  $s_1, s_2, e_1, e_2 \in S$  is

$$\min_{p \in S} \max \left\{ \frac{d_{GC_P}(s_1, p) + d_{GC_P}(p, e_1)}{d_{\text{Euclidean}}(s_1, e_1)}, \frac{d_{GC_P}(s_2, p) + d_{GC_P}(p, e_2)}{d_{\text{Euclidean}}(s_2, e_2)} \right\}$$

- We obtain our better bound

$$\text{lb}(P) = \max_{\substack{s_1, s_2, e_1, e_2 \in S \\ \text{distinct and} \\ \text{in clockwise order}}} \text{lb}(s_1, s_2, e_1, e_2)$$

# What we have and what we want

- Lower bound technique: lower bound  $\text{lb}(P)$  on the dilation of triangulations which contain  $P$ .
- Our goal: find a global lower bound  $\text{glb}$  with

$$\text{glb} \leq \min_{T \in \mathcal{T}} \text{dil}(T)$$

and a sharp inequality (“=”  $\rightarrow$  dil computed).

# Algorithm to find a global lower bound

- Algorithm for glb: take

$$\text{glb} = \min_{P \in \mathcal{C}} \text{lb}(P)$$

where  $\mathcal{C} \subseteq \mathcal{P}$  is a set of partial configurations.

- Exhaustive method: case  $\mathcal{C} = \mathcal{T}$ !

# Global lower bound: which configurations should we consider?

- How does the algorithm choose  $\mathcal{C}$ ?
- Key point: good tradeoff between  $\mathcal{C}$  small (fast algo, possibly poor bound) and  $\mathcal{C}$  large (slower, better bound).

# The search tree

- Abstract “search tree” of partial configurations  $P \in \mathcal{P}$ .
- For each  $P$ , we have a bound  $\text{lb}(P)$ .
- Monotonicity is important: if  $P_0 \subseteq P_1 \subseteq \dots \subseteq P_n = T \in \mathcal{T}$ , then

$$\text{lb}(P_0) \leq \text{lb}(P_1) \leq \dots \leq \text{lb}(P_k) = \text{dil}(T)$$

# Pruning the search tree

*Pruning* is very efficient for optimisation problems on search trees  
→ need a “target value”

Lower bound, with a “target value”  $c$

Given a constant

$$c \geq \min_{T \in \mathcal{T}} \text{dil}(T)$$

return a *proven* lower bound

$$\text{glb} \leq \min_{T \in \mathcal{T}} \text{dil}(T)$$

In practice,  $c = \text{dil}(T_{\text{candidate}}) \in \mathcal{T}$  for a very good triangulation  $T$ .

# What is $c$ useful for?

- $c$  is only used for pruning purposes  
→ “cut” branches of the search tree
- $c$ , given as input to the lower bound algorithm, *does not change the result returned by the algorithm* (!)
- The speed of the proposed method depends *crucially* on the “quality” of  $c$ .
- Hope: prove that  $c$  is in fact equal to the dilation, i.e.

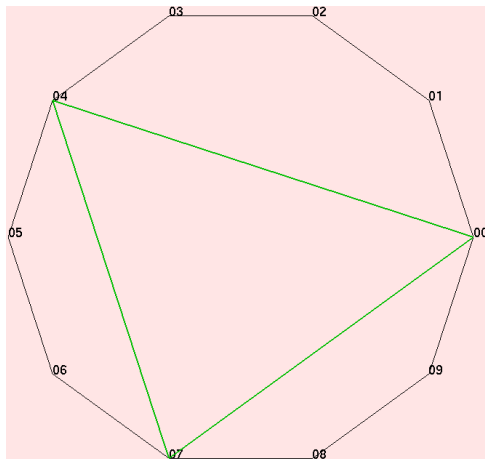
$$\text{glb} = c = \text{dil}(T_{\text{candidate}})$$



# Important edges first

- The order in which partial configurations are considered matters.
- Important to first put some edges that will cause  $lb(P)$  to be big, to cut early.
- Our program puts the *edges of the triangle which contains the center* first.
- It then puts three smaller triangles on the 3 zones delimited by the central triangle.

# Central triangle



A possible central triangle in a 10-gon.

# Putting it all together

## Lower bound algorithm

- 1 Take a positive integer  $n$  and a “target value”  $c$  as input.
- 2 Go through the search tree of partial triangulations, considering important edges first (adding triangles gradually).
- 3 Prune while going through the search tree.
- 4 Stop at a specified depth.
- 5 Return the global lower bound  $glb$ .

# Upper bound: what are we looking for?

As we saw before, we need a good target constant  $c = \text{dil}(T_{good})$  if we want our lower bound algorithm to run fast enough, and we can only conclude if

$$c = \min_{T \in \mathcal{T}} \text{dil}(T)$$

# Classical techniques

- Most articles only focus on the upper bound part: find  $T_{good}$ ,

$$\min_{T \in \mathcal{T}} \text{dil}(T) \lesssim \text{dil}(T_{good})$$

- Two typical steps:
  - Describe a class of “seemingly good” triangulations (classes with 4 and 6 parameters in Sattari and Izadi (2019)).
  - Find the optimal triangulation among the members of the class.

# Discussion of such techniques

Two main advantages:

- The number of considered configurations is polynomial in  $n$ .
- Finding the best configuration  
→ doable either with a computer or by hand.

Intrinsic issues:

- No formal justification regarding why these classes are considered, only heuristic motivations.
- (!) *No control on the sharpness of the inequality*

$$\min_{T \in \mathcal{T}} \text{dil}(T) \leq \text{dil}(T_{\text{good}})$$

# Discussion of such techniques

- Second issue: due to the nature of the methods, i.e. living in  $\mathcal{S} \subseteq \mathcal{T}$  and forgetting about the rest of  $\mathcal{T}$ .
- Lower bound algorithm  $\rightarrow$  response to the second issue.
- To avoid these issues, we will use *metaheuristics* instead to find good configurations.

# Metaheuristics

- Goal: explore the search space  $\mathcal{T}$  and find good configurations.
- Metaheuristics: generic methods to solve optimization problems.
- Here: *hill climbing*.





















