# Graph-Theoretic Solutions to Computational Geometry Problems 

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## Historically, many connections from graph-theoretic algorithms to computational geometry...

1. Geometric analogues of classical graph algorithm problems

Typical issue: using geometric information to speed up naive application of graph algorithms
E.g., Euclidean minimum spanning tree
= Spanning tree of complete graph with Euclidean distances Solved in O( $\mathrm{n} \log \mathrm{n}$ ) time by Delaunay triangulation [Shamos 1978]


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2. Geometric approaches to graph-theoretic problems

How many different minimum spanning trees can a graph with linearly varying edge weights form? $\mathrm{O}\left(\mathrm{m} \mathrm{n}^{1 / 3}\right)$ via crossing number inequality [Dey, DCG 1998] $\Omega(m \alpha(n))$ via lower envelopes of line segments [E., DCG 1998]


## Historically, many connections from graph-theoretic algorithms to computational geometry...

Today: 3. Graph-theoretic approaches to geometric problems
Geometry leads to auxiliary graph
Special properties of auxiliary graph lead to algorithm
Algorithm on auxiliary graph leads to solution


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## Outline

## Art gallery theorems

Partition into rectangles
Minimum diameter clustering
Bend minimization
Mesh stripification
Angle optimization of tilings
Metric embedding into stars

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## The Art Gallery Problem

Input: a simple polygon (no holes, no self-crossings), the floor plan of an art gallery

Output: a small set of points (places for guards to stand) from which whole gallery visible

Exact optimization is NP-hard
Approximation algorithms known
Today: what is worst-case \#guards as a function of gallery complexity?


This art gallery can be guarded from four points Claudio Rocchini, GFDL image on Wikimedia commons,

## Chvátal's Art Gallery Theorem [Chvátal, JCTB 1975]

Every n-vertex simple polygon requires at most floor(n/3) guards
For every $\mathrm{n} \geq 3$, some simple polygons require exactly floor(n/3) guards


Each guard can see at most one tooth of the comb

Fisk's Proof of the Art Gallery Theorem [Fisk, JCTB 1978]
I. Triangulate the polygon


Fisk's Proof of the Art Gallery Theorem [Fisk, JCTB 1978]
II. 3-color the (maximal outerplanar) graph of the triangulation


Dual graph is a tree
Remove a leaf (degree-two vertex of triangulation), recurse

Fisk's Proof of the Art Gallery Theorem [Fisk, JCTB 1978]
III. Select the vertices with the least-frequently-used color


The points within any triangle can be seen from the triangle vertex of the selected color

## Algorithmic implications

Graph coloring is hard in general, but...
Finding a set of floor(n/3) guards can be performed in linear time

> Linear time triangulation of simple polygons [Chazelle, DCG 1991]

Linear time optimal coloring of maximal outerplanar graphs
(easy using greedy coloring: maximal outerplanar graphs are chordal, chordal graphs are perfectly orderable)

## Orthogonal Art Galleries

At most floor(n/4) guards needed [Kahn, Klawe, Kleitman, SIAM ADM 1983]

Partition into convex quadrilaterals (non-trivial)

Squaregraph: planar graph, all interior faces quadrilaterals, all interior vertex degrees $\geq 4$ (bipartite, median graph)

Kinggraph: add diagonals of quads Guaranteed to be 4-chromatic

4-color, use smallest color class


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## Rectangle partition

Input: orthogonal polygon (sides parallel to axes, possibly with holes)

Output: partition into minimum \# of rectangles

Applications include
Bitmap data compression
VLSI mask fabrication
DNA array design
Reconnaissance planning



17 rectangles


16 rectangles

## Rectangle partition

Input is partitioned into rectangles
if and only if

Each concave vertex is adjacent to an internal segment
\# Rectangles $\geq$ \# Segments + 1

(equality when no degree-4 internal vertices)

So \# Rectangles = \# Concave vertices + 1 regardless of the partition???


17 rectangles


17 rectangles


## Key observation

[Lipski et al., Fund. Inf. 1979; Ohtsuke, ISCAS 1982; Ferrari et al., CVGIP 1984]

Some segments can cover two concave vertices at once
\# Rectangles =
\# Concave vertices -
\# Two-vertex segments + 1
Problem becomes one of finding maximum non-intersecting set of two-vertex segments
(Shared endpoint counts as intersection)

## König's theorem

[D. König, Mat. Fiz. Lapok 1931]
In a bipartite graph, independence number $=\mathrm{n}-|\mathrm{M}|$ where $M$ is a maximum matching

Upper bound:
For arbitrary graphs, \# MIS $\leq$ \# vertices - \# match (MIS can only use one vertex from every matched pair)

Lower bound:
Even levels of alternating path decomposition starting from unmatched vertices form a large independent set


## Rectangle partition algorithm

Find line segments that could cover two concave polygon vertices
Form their (bipartite) intersection graph
Use matching algorithms to find a maximum independent set
Add additional line segments to cover the remaining concave vertices

Time (using geometric data structures to speed up matching steps):

$$
O\left(n^{3 / 2} \log n\right)
$$

[Lipski, Networks 1983 \& IPL 1984; Imai \& Asano, SICOMP 1986]

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# Partition into rectangles <br> Minimum diameter clustering 

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## Minimum diameter clustering problems

[Aggarwal et al, J. Algorithms 1991; E. \& Erickson, DCG 1994]
Diameter $=$ max distance among points
Given n points ...
Find k of them with minimum diameter
Find as many as possible with diameter D
Test whether there exists a subset of $k$ points with diameter $D$

Since there are $O\left(\mathrm{n}^{2}\right)$ possible diameters, these are all equivalent to each other (up to log factors) via binary search

## Easy (but unhelpful) reduction to a graph problem

Maximizing \#points in a diameter-D cluster = finding a maximum clique in a two-dimensional unit disk intersection graph

Can be solved in polynomial time [Clark, Colbourne, \& Johnson, Disc. Math. 1990] but translation into a graph is too direct to provide insight into solution


## More insightful reduction

[Clark et al, Aggarwal et al]

For each pair of points $p, q$ that might be a diameter (purple):

The points within $\operatorname{dist}(p, q)$ of both $p$ and $q$ form a lune (intersection of two circles)

Any incompatible pair of points (too far apart to be in cluster)
belong to opposite half-lunes

Max \# points having pq as
 their diameter = max ind. set of bipartite graph

## Dynamic graph algorithm for multiple bipartite MIS's

[E. \& Erickson 1994]
To find max cluster size for given diameter D:
For each input point p:
Let $q$ be a point at distance $D$ from $p$ (not necessarily in input), generate the lune from p and q , and find maximum independent set in the bipartite graph of lune.

Rotate q continuously around p;
whenever the rotation changes the set of points within the lune:
Do a single alternating path search to update MIS
Time $0\left(n^{3} \log n\right)$
To find min D given cluster size k :
Use binary search for D among input distances Limit subproblem size via $k$-nearest-neighbor graph Time O(n $\log n+k^{2} n \log ^{2} k$ )

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## Rectilinear cartograms

Transform map into stylized diagram in which areas represent non-geographic data about regions


Raisz, Geog. Rev. 1934:
"It should be emphasized that the statistical cartogram is not a map."
Diagram modifed from CC-BY-SA image by Brianski, Canuckguy, Zaparojdik, Madman2001, Roke, \& Ssolbergj, online at http://commons.wikimedia.org/wiki/File:Blank_Map_of_Europe_-w_boundaries.svg

## Algorithmic issues in cartogram construction

Fitting given numerical quantities
to areas of regions
[E. et al, SoCG 2009]
Making region adjacencies match their geographic orientations
[E. \& Mumford, WADS 2009]
Today:
Minimizing \# bends


Originally studied as a technique for graph drawing
[Tamassia, SICOMP 1987; Tamassia et al., Trans. Sys. Man. Cyb. 1988;
Tamassia et al., SPDP 1991; Fößmeier \& Kauffman, GD 1995]

## Modeling bends by flow

Create a vertex in a graph for every junction and region

One unit of flow $=90$ degrees
Send four units of flow from junctions to adjacent regions

Send flow across any bend between two regions

Junction has four outgoing units
Region with k junctions has 4-2k outgoing units or $2 \mathrm{k}-4$ incoming units ( $2 \mathrm{k}+4$ incoming units for the exterior region)

## Bend minimization as min-cost flow

Vertex per junction and region Additional circulation vertex (not shown)

Junction-region edge: min capacity 1 , cost 0

Region-region edge: min capacity 0 , cost 1

Circulation-junction edge: min\&max capacity 4

Region-circulation edge: min\&max capacity 2 k - 4 (or $2 \mathrm{k}+4$ for exterior region)

Minimum cost integer circulation gives
 minimum bend layout

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## Problem: sort triangles of a geometric model into a single contiguous triangle strip



Communicate only vertex per triangle to graphics hardware instead of three Part of schemes for data compression of mesh topology \& geometry
Space-filling curve from strip useful for dithering, mesh simplification, etc.
[E. \& Gopi, Eurographics 2004]

## As with min-diameter clustering, a direct reduction to a graph is unhelpful

Form dual graph
One vertex for every triangle
Edge connecting two adjacent triangles
Single-loop triangle strip
= Hamiltonian cycle in dual graph
But even in dual graphs of planar triangular meshes, Hamiltonian cycle is NP-complete and may not even exist


## Less-direct matching-based approach

Dual graph is 3-regular and bridgeless; therefore, it has a perfect matching [Peterson 1891; efficient algorithms due to Biedl et al., J. Algorithms 2001]

Complementary edges to matching form set of cyclic strips


## Less-direct matching-based approach (II)

In many cases, local moves allow number of cycles to be reduced


## Less-direct matching-based approach (III)

In remaining cases, subdividing two triangles allows local move to merge cycles


Result: single strip with same geometry as original model at most $3 / 2$ as many triangles as original

In practice, increase only 1-2\%; some models require 39/37 factor increase

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For each vertex of a line arrangement, form convex centrally-symmetric tile - Unit length sides

- Edges perpendicular to lines

These tiles fit together into a tiling of a convex polygon (or, for infinite arrangements, a tiling of the Euclidean plane)


This idea was used by De Bruijn [Indag. Math. 1981] to form a Penrose rhomb tiling from five sets of parallel lines

Same method also works for hyperbolic arrangements
(tiles a non-convex subset of the Euclidean plane)
[E., GD 2004]

This arrangement has no triangles and requires five colors if crossing lines must be different colors (the max for triangle-free hyperbolic arrangements)
[Ageev, Disc. Math. 1996]

## The tiling formed from Ageev's arrangement

Result is a squaregraph (planar, all faces quadrilaterals, all interior vertices have degree $\geq 4$ ) that cannot be embedded into the product of fewer than five trees
[Bandelt, Chepoi, E., 2009]
Can we make the drawing more legible by adjusting slopes of sides of tiles?
$\qquad$


## Optimization of angular resolution [E. \& Wortman, 2009]

Angular resolution = sharpest angle of drawing [Malitz \& Papakostas, STOC 1992]
Given tiling by symmetric polygons find combinatorially equivalent tiling with optimal angular resolution


## Transformation to parametric shortest path problem

Graph in which edges have as weights linear functions of $\lambda$

Concept in tiling:
"Zone" $z_{i}$ of parallel line segments
Angular resolution $\alpha$
$\alpha$ is a feasible resolution
Amount to adjust angle of $z_{i}$
Angle $\mathrm{z}_{\mathrm{i}} \mathrm{z}_{\mathrm{j}}$ is at least $\alpha$
Interior angles are convex

Corresponding concept in graph:

Vertex $\mathrm{v}_{\mathrm{i}}$<br>Parameter value $\lambda$

Graph has no negative cycles for $\lambda$
Distance from start vertex $s$ to $v_{i}$
Edge $v_{i} v_{j}$ with weight $\theta_{i}-\theta_{j}-\lambda$
Edge $v_{i} v_{j}$ with weight $\pi+\theta_{i}-\theta_{j}$

## The optimized drawing:

An algorithm of Karp \& Orlin [Disc. Appl. Math. 1981] can solve parametric negative cycle detection for edge weights const, const + $\lambda$ in time $0\left(\mathrm{n}^{3}\right)$
(Implementation used binary search + Bellman-Ford)

Translation to and from graph problem can be done within the same time bound


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## Metric space

Set $X$ of points, and function d from pairs of points to real numbers satisfying

Symmetry: for all $x$ and $y, d(x, y)=d(y, x)$
Positivity: for all $x$ and $y, d(x, y) \geq 0$, with equality iff $x=y$
Triangle inequality: for all $x, y$, and $z, d(x, y)+d(y, z) \geq d(x, z)$ (if equal, then $y$ is "between" $x$ and $z$ )

## Examples:

Shortest path lengths in weighted undirected graphs
Euclidean distance between points in $\mathbf{R}^{\mathbf{d}}$

## Embeddings and distortion

Embedding: 1-1 function from the points of one metric space to another
dilation of a single distance:

$$
\frac{d(f(x), f(y))}{d(x, y)}
$$

distortion of embedding $=$ worst ratio of dilations

$$
\max _{x \neq y} \frac{d(f(x), f(y))}{d(x, y)} / \min _{u \neq v} \frac{d(f(u), f(v))}{d(u, v)}
$$

if embedding scaled so all distances nondecreasing, distortion = max dilation

$$
\max _{x \neq y} \frac{d(f(x), f(y))}{d(x, y)}
$$

## Structured vs unstructured metrics

Unstructured: any metric possible $\mathrm{O}\left(\mathrm{n}^{2}\right)$ degrees of freedom in specifying distances

Distance matrix
Graph shortest path distances
n -dimensional Linfinity metric
Structured: constrained subset of metrics O(n) degrees of freedom

Low-dimensional Euclidean or $L_{p}$ metric
Tree shortest path distances

## Metric embedding problem

Given an unstructured metric space find a low-distortion embedding into a structured space

or, sometimes (not today)<br>find a random family of embeddings in which any individual distance has low expected distortion

Many applications in which
approximation algorithm designed for structured space can be extended to arbitrary metric spaces

## Minimum dilation star problem [E. \& Wortman, WADS 2009]

Find minimum distortion embedding
Target structured space is a star:
There exists a point (not necessarily in the image of the embedding) that is between every other pair of points

Equivalently, graph shortest path metric on a tree with one new non-leaf node


## Solution idea

Form a graph, with two vertices per point of the input metric space

Downward edge weight = negative distance

Upward edge weight = param times distance

Optimal distortion =
 minimum parameter value s.t. graph has no negative cycles (no longer Karp-Orlin, but still polynomial)

Distance from star hub to leaf point in embedding = 1/2 (difference between distances from s to two vertices for that point) always non-negative because of upwards length-0 edge

## Conclusions

Graph-theoretic point of view is useful in many non-graph problems
The graph algorithms used for these problems are often classical... maximum independent set and maximum clique maximum or perfect matching maximum or minimum-cost flow graph shortest paths
...but sometimes with a twist
parametric negative cycle detection
Special classes of graphs and their structure is often important
maximal outerplanar graphs
squaregraphs and kinggraphs
bipartite graphs
intersection graphs of unit circles
planar graphs
bridgeless 3-regular graphs
Much more likely remains to be discovered


[^0]:    Minimum-diameter clustering via maximum independent sets in bipartite graphs (more detail later in talk)

