

# Parametric and Kinetic Minimum Spanning Trees

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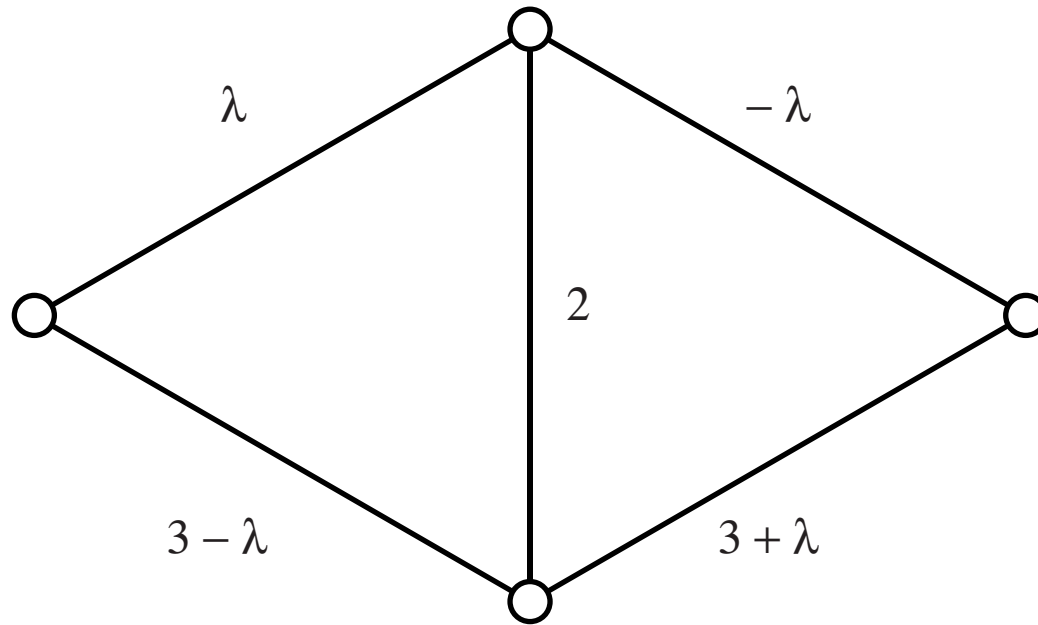
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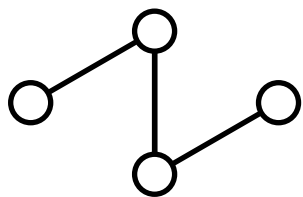
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# Parametric Minimum Spanning Tree:

Given graph, edges labeled by linear functions

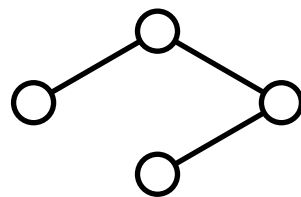


Find MST for each possible value of  $\lambda$



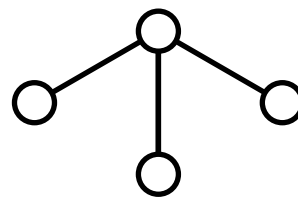
$$\lambda < -2$$

$$5 + 2\lambda$$



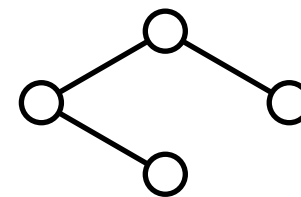
$$-2 < \lambda < -1$$

$$3 + \lambda$$



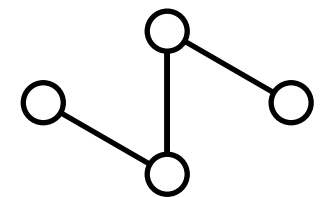
$$-1 < \lambda < 1$$

$$2$$



$$1 < \lambda < 2$$

$$3 - \lambda$$

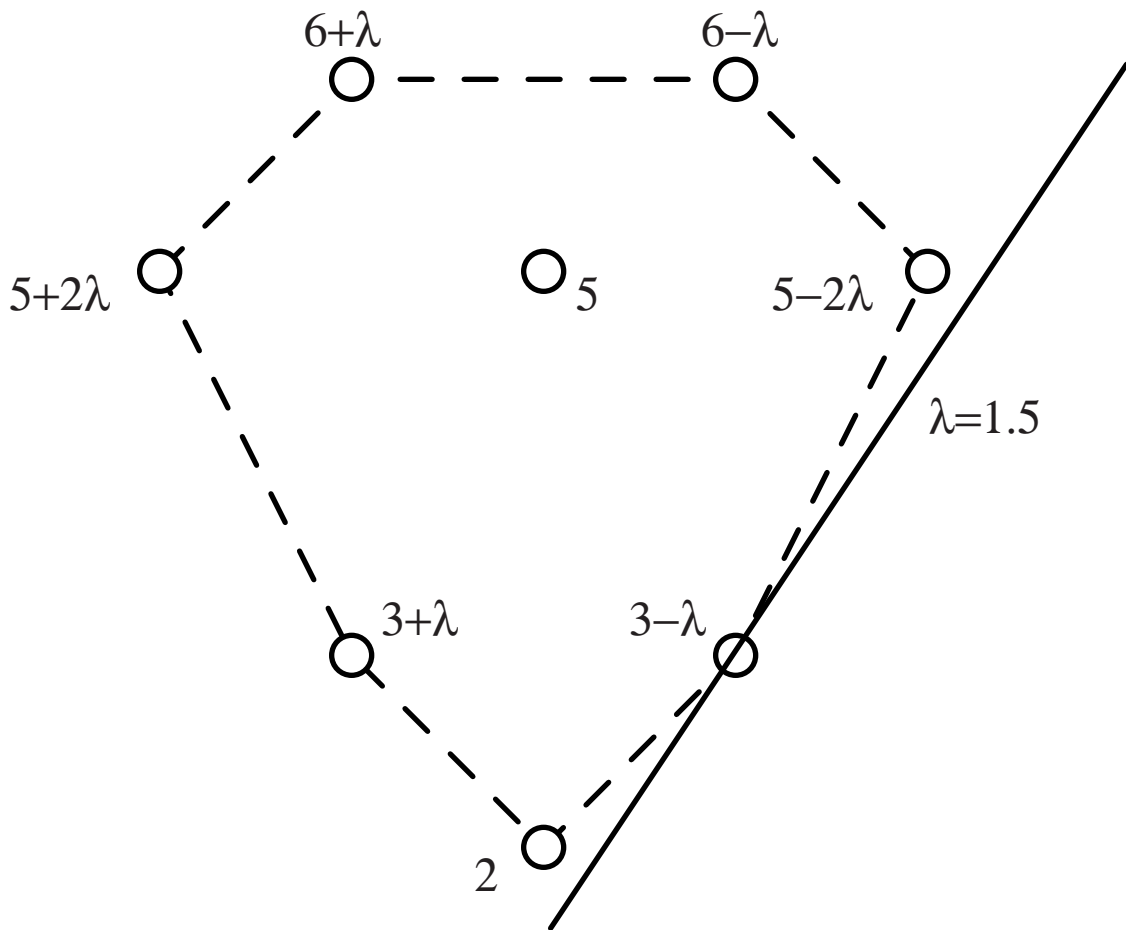


$$2 < \lambda$$

$$5 - 2\lambda$$

## Geometric Interpretation:

Point  $(-B, A)$  for tree  $w/\text{weight } A + B\lambda$



Then  $\text{MST}(\lambda) = \text{tangent to line } w/\text{slope } \lambda$

so parametric MST = lower convex hull

# Applications

For any quasiconcave function  $f(A, B)$   
optimum tree must be a convex hull vertex

Tree w/minimum cost-reliability ratio  
( $A = \text{cost}$ ,  $B = -\log \text{probability all edges exist}$ ):

$$f(A, B) = A \exp(B)$$

Tree w/minimum variance in total weight  
(if edge weights independent random variables):

$$f(A, B) = A - B^2$$

Tree with high probability of low total weight  
(if edge weights independent Gaussian variables):

$$f(A, B) = A + \sqrt{B}$$

So each of these optima can be found  
from parametric MST solution

## Previous Results on Parametric MST

Number of breakpoints:

- $O(mn^{1/3})$  [Dey 1997]
- $\Omega(m\alpha(n))$  [Eppstein 1995]

Time to compute all trees:

- $O(mn \log n)$   
[Fernández-Baca, Slutzki, Eppstein 1996]

# Dynamic Minimum Spanning Tree

An alternate form of time-varying data:  
Weighted graph subject to discrete updates  
(like parametric w/piecewise constant functions)

Many algorithms known

[Sleator, Tarjan 1983]

[Frederickson 1985]

[Eppstein 1991]

[Eppstein, Galil, Italiano, Nissenzweig 1992]

[Eppstein, Galil, Italiano, Spencer 1993]

[Henzinger, King 1997]

[Holm, de Lichtenberg, Thorup 1998]

Current best time:  $O(\log^4 n)$  per update  
better for restricted updates or planar graphs

Idea: apply dynamic graph algorithm techniques  
to parametric MST problem

# How to combine parametric and dynamic? Kinetic Algorithms!

Interpret  $\lambda$  as time parameter  
start with parametric problem, small  $\lambda$   
increase  $\lambda$  and perform updates  
maintaining correct MST at each point in process

Idea: model short-term predictability  
and long-term unpredictability  
of real-world applications

Two kinds of updates possible:

**structural:**

edge insertions and deletions

**functional:**

relabel existing edge w/new function

## Other Kinetic Algorithms

[Basch, Guibas, Hershberger 1997]

[Basch, Guibas, Zhang 1997]

[Guibas 1998]

[Agarwal, Erickson, Guibas 1998]

[Basch, Erickson, Guibas,  
Hershberger, Zhang 1999]

Basic data structures

(Priority queue)

Computational geometry

(Convex hull, closest pair,  
binary space partition,  
polygon intersection)

Typical time bounds are  $\text{polylog} \times$  worst case  
number of changes to solution



## New Results

### General graphs:

- $O(m^{2/3} \log^{4/3} m)$  per output change
- $O(n^{2/3} \log^{4/3} n)$  times worst case # changes

### Minor-closed graph families (including planar graphs):

- $O(n^{1/2} \log^{3/2} n)$  per output change

### Minor-closed families with only functional updates:

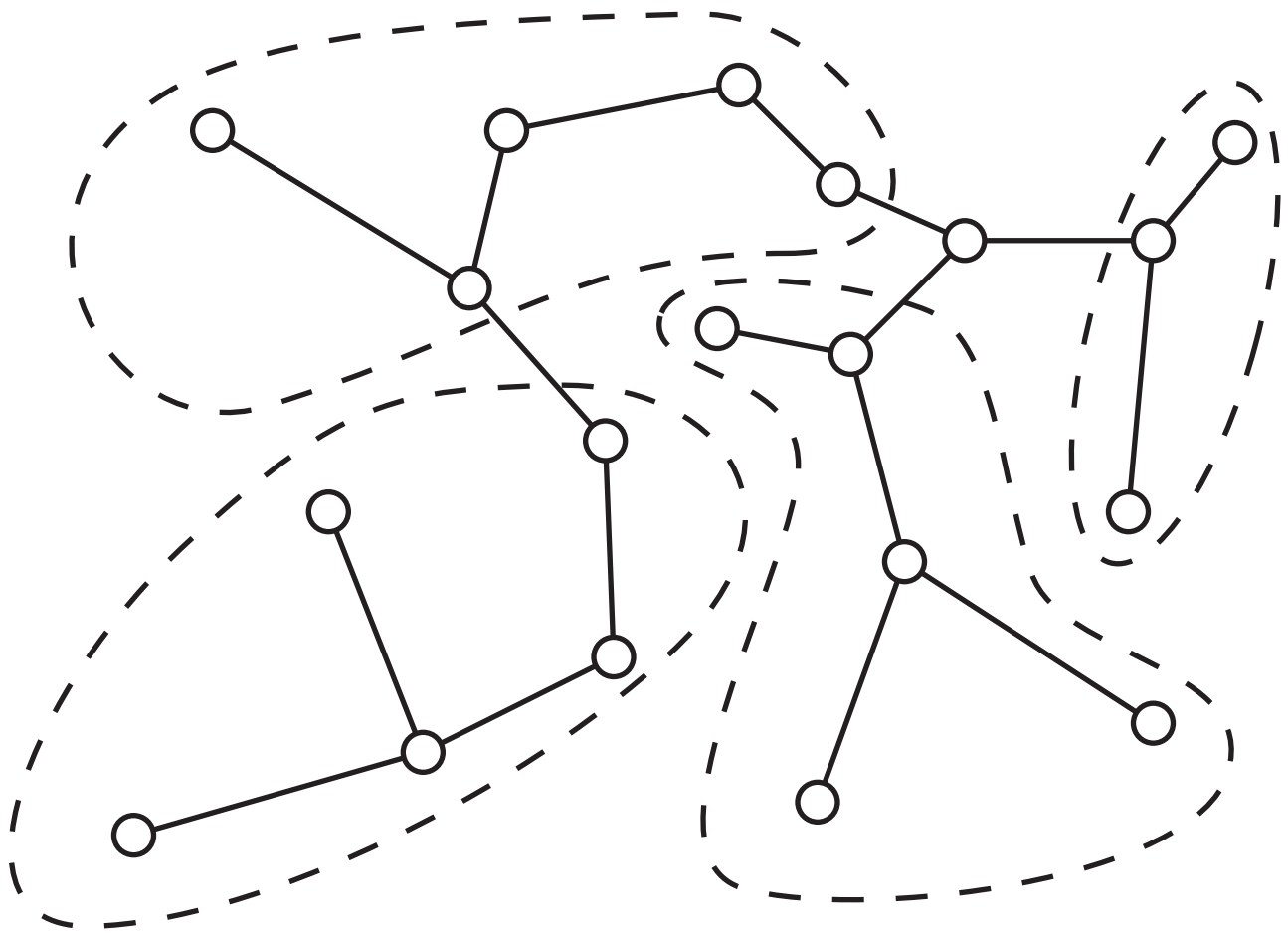
- $O(n^{3/2})$  preprocessing (nonplanar graphs only)
- $O(n^{1/4} \log^{3/2} n)$  times worst case # changes

### Some randomized improvements to polylogs

## Idea I: Clustering

Expand vertices so graph has degree three, then...

Group MST into  $k$  clusters of  $O(n/k)$  edges,  
at most two edges crossing each cluster boundary  
[Frederickson 1985]



Form bundles of non-tree edges, according to the  
clusters containing their endpoints

adjust clusters as MST changes

## Classification of MST changes

MST always changes by swap:  
insert non-tree edge, delete tree edge

Three types of swap:

- Intra-cluster swap: tree edge belongs to cluster containing both non-tree edge endpoints
- Dual-cluster swap: tree edge belongs to cluster containing both non-tree edge endpoints
- Inter-cluster swap: tree edge and non-tree edge are in disjoint clusters

## Finding Intra-Cluster Swaps

Use Megiddo's parametric search  
to find last value of  $\lambda$   
for which the cluster has same MST

Decision oracle is (static) MST algorithm

Time:  $\tilde{O}(m/k)$  per changed cluster  
Each update changes  $O(1)$  clusters  
so  $\tilde{O}(m/k)$  total

## Finding Inter-Cluster Swaps

Collapse each bundle or cluster to superedge

Weight of bundle superedge = min in bundle

Weight of cluster superedge = max in path

Handle weight queries using convex hull of coefficients of edge labels in bundle or cluster

Find swap by parametric search in collapsed graph

Time for parametric search:  $\tilde{O}(\# \text{ bundles})$

Time to rebuild convex hulls:

$\tilde{O}(m/k)$  per changed cluster

## Finding Dual-Cluster Swaps

“Ambivalent data structure” [Frederickson 1997]

For each non-tree edge endpoint, there are two tree paths inside the cluster to the two cluster exits.

Non-tree edge stores a candidate swap per exit  
Found by traversing MST within cluster  
querying dynamic convex hull of path edges

Each bundle stores a candidate swap per exit  
the best among all swaps stored by its edges

Best dual-cluster swap found by checking which candidate is correct for each bundle, picking the best of the correct candidates

Time to update edge and bundle candidates:  
 $\tilde{O}(m/k)$  per changed cluster

Time to find best swap:  $\tilde{O}(\# \text{ bundles})$

# Analysis of Clustering

General graphs:

Total time  $\tilde{O}(m/k + k^2)$

Optimal  $k = \tilde{O}(m^{1/3})$

$\tilde{O}(m^{2/3})$  per MST change

Sparse (minor-closed) graph families:

Total time  $\tilde{O}(n/k + k)$

Optimal  $k = \tilde{O}(n^{1/2})$

$\tilde{O}(n^{1/2})$  per MST change

## Idea II: Sparsification

[Eppstein, Galil, Italiano, Nissenzweig 1992]

[Fernández-Baca, Slutzki, Eppstein 1996]

Split edges of graph into two subsets

$$G = G_1 \cup G_2$$

Maintain MST of each subset  
(two smaller kinetic problems)

Combine to get MST of overall graph  
(one sparse structurally kinetic problem)

$$MST(G) = MST(T_1 \cup T_2)$$



## Sparsification Analysis

Replaces factors of  $m$  by factors of  $n$   
in any general graph MST algorithm

But subproblems changes may not  
propagate to global MST

so also replaces factors of actual MST changes  
with worst-case # changes

Therefore: general graph kinetic MST  
 $\tilde{O}(n^{2/3})$  times worst-case # changes

# Separator Based Sparsification

[Eppstein, Galil, Italiano, Spencer 1993]

Given *functionally* kinetic problem

Form separator decomposition of sparse graph

Solve MST problems on each side of separator  
(two smaller *functionally* kinetic problems)

Use solutions to form *compact certificate*  
(graph with  $O(\sqrt{n})$  vertices  
having same kinetic behavior as original subgraph)

Combine certificates  
(one very small *structurally* kinetic problem)

Total time:  $\tilde{O}(n^{1/4})$  per worst-case change

## Conclusions and Open Problems

New kinetic MST algorithms

Some improvement to parametric MST  
especially in the planar case (now  $O(n^{19/12})$ )  
but for general graphs, still not  $o(mn)$

Planar graph algorithm uses clustering  
in sparsified subproblems —  
can we instead use sparsification recursively?

Geometric kinetic MST?

Edge weights become quadratic instead of linear