

Playing with Kruskal: algorithms for morphological trees in edge-weighted graphs

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Highlights

Results

- A quasi-linear algorithm that computes a *binary partition tree by altitude ordering*
- Three linear post-processing algorithms that compute
 - hierarchy of quasi-flat zones
 - also known as the α -tree
 - also known as the Fuzzy Connectedness hierarchy
 - (hierarchies of) watershed cuts
 - hierarchies by increasing attributes
 - constrained connectivity hierarchies or
 - watershed-based hierarchies.

Outline

1 Binary Partition Tree and Minimum Spanning Tree

2 Post-Processing the binary tree

- Quasi-flat zones hierarchy
- Watershed-cut hierarchy
- Attribute-based hierarchies

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Minimum Spanning Tree

The Minimum Spanning Tree (MST) T is a connected spanning graph of the graph G such that the weight of T :

$$F(T) := \sum_{e \in E(T)} F(e)$$

is the least possible weight for a connected spanning subgraph of G .

Kruskal algorithm for MST (High-Level View)

- create a forest \mathcal{F} (a set of trees), where each vertex in the graph is a separate tree
- create a set S containing all the edges in the graph
- while S is nonempty and \mathcal{F} is not yet a single tree
 - remove an edge with minimum weight from S
 - if that edge connects two different trees, then add it to the forest, combining two trees into a single tree
 - otherwise discard that edge.

At the termination of the algorithm, the forest has only one component and forms a minimum spanning tree of the graph.

The *disjoint set* problem

The disjoint set problem consists in maintaining a collection \mathcal{Q} of disjoint sets under the operation of union.

Each set Q in \mathcal{Q} is represented by a unique element of Q , called the *canonical element*.

- `MakeSet(q_1)`
- `FindCanonical(q_1)`
- `Union(q_1, q_2)`

Kruskal algorithm for MST (Implementation)

Data: An edge-weighted graph (V, E, F) .

Result: A minimum spanning tree MST

Result: A collection Q

// Collection Q is initialized to \emptyset

```
1 e := 0
2 for all  $x_i \in V$  do MakeSet( $i$ );
3 for all edges  $\{x, y\}$  by (strict) increasing weight  $F(\{x, y\})$  do
4    $c_x := Q.$ FindCanonical( $x$ );  $c_y := Q.$ FindCanonical( $y$ )
5   if  $c_x \neq c_y$  then
6      $Q.$ Union( $c_x, c_y$ );
7     MST[ $e$ ] :=  $\{x, y\}$ ;  $e := e + 1$ 
8   else DoSomething( $\{x, y\}$ )
```


Main question in Kruskal implementation

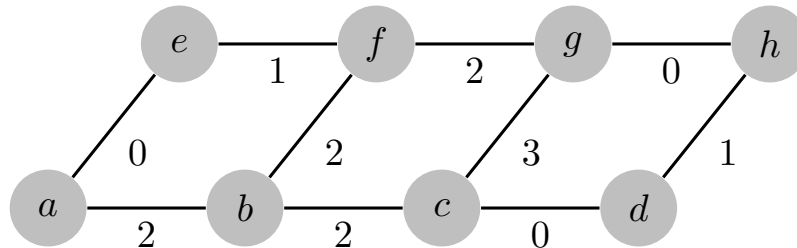
Question

How to represent and implement the collection \mathcal{Q} .

Answer

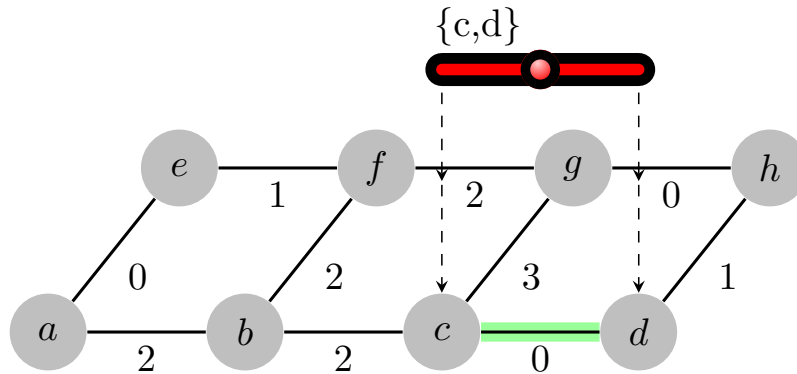
A good representation for \mathcal{Q} is as a set of trees.

Binary Partition Tree by altitude ordering



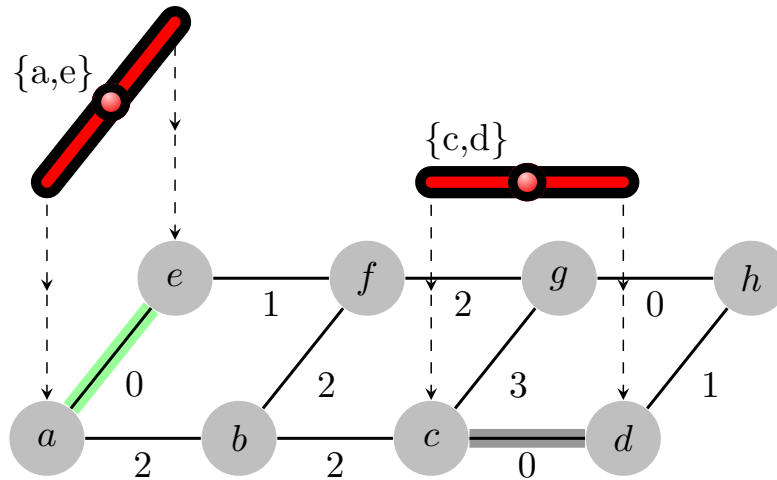
Edge-weighted graph

Binary Partition Tree by altitude ordering



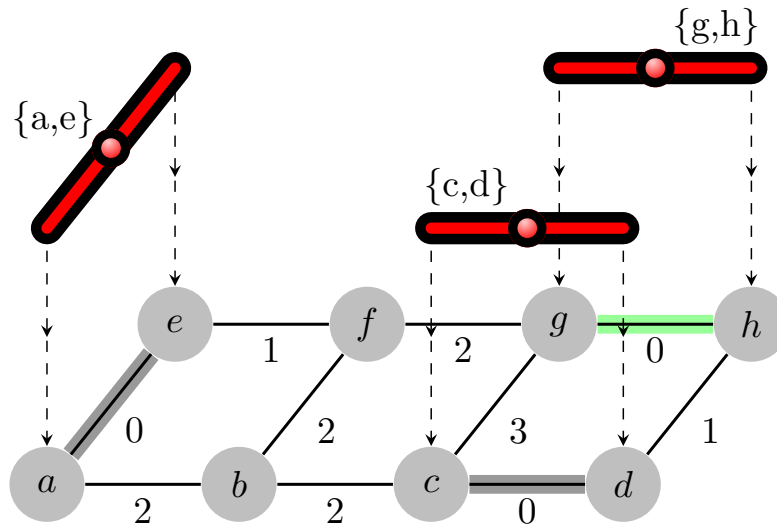
First edge-node

Binary Partition Tree by altitude ordering



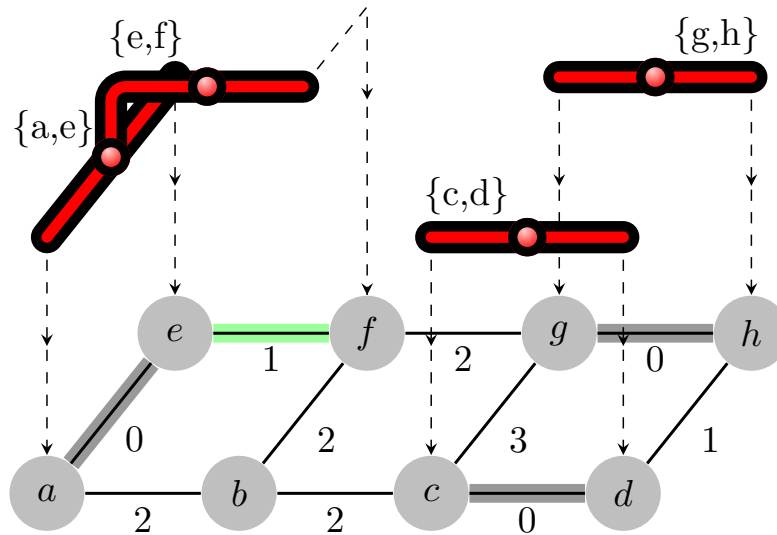
Second edge-node

Binary Partition Tree by altitude ordering



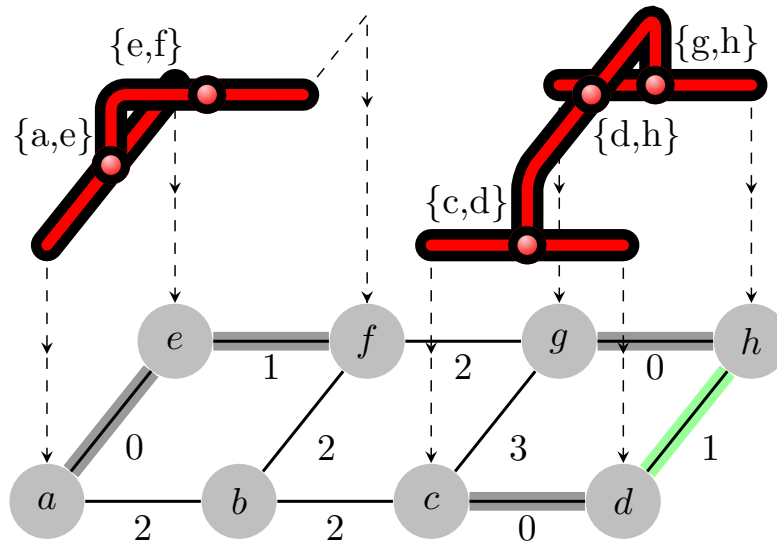
Third edge-node

Binary Partition Tree by altitude ordering



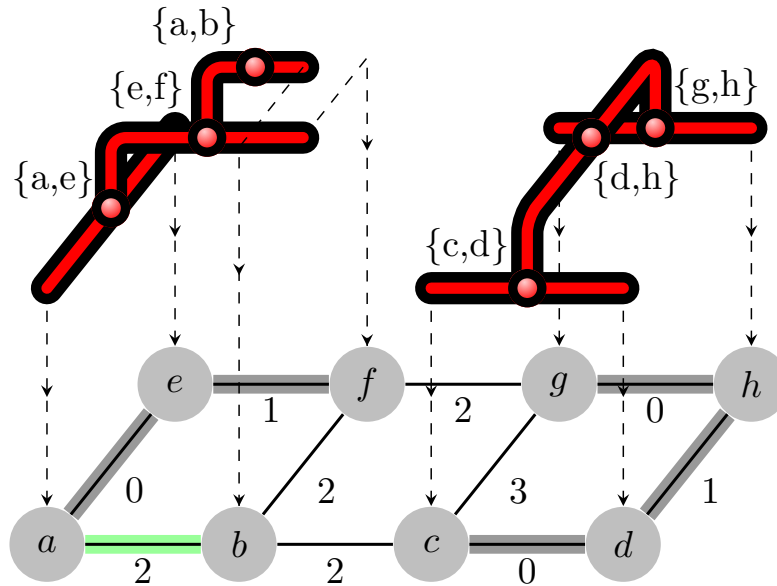
Fourth edge-node

Binary Partition Tree by altitude ordering



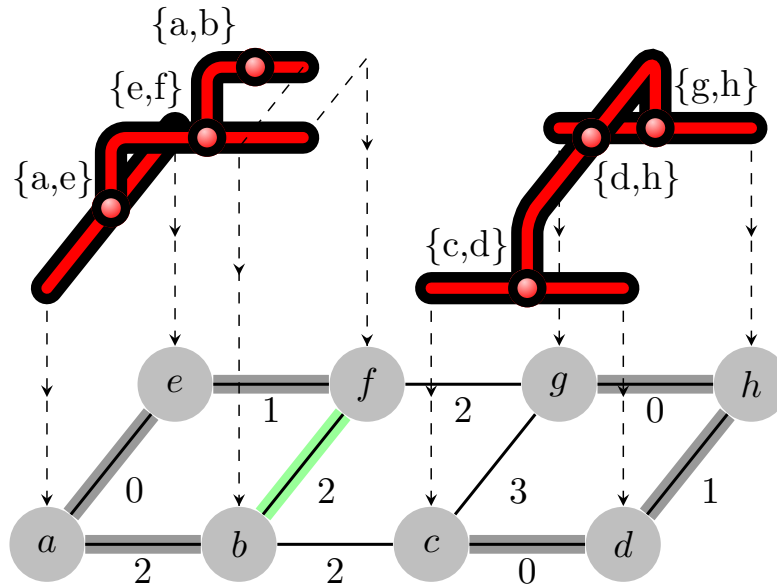
Fifth edge-node

Binary Partition Tree by altitude ordering



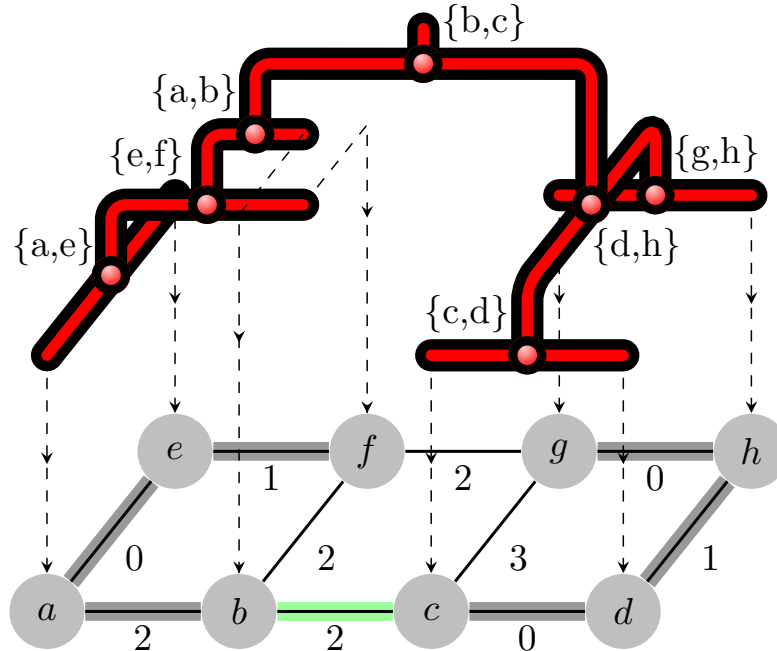
Sixth edge-node

Binary Partition Tree by altitude ordering



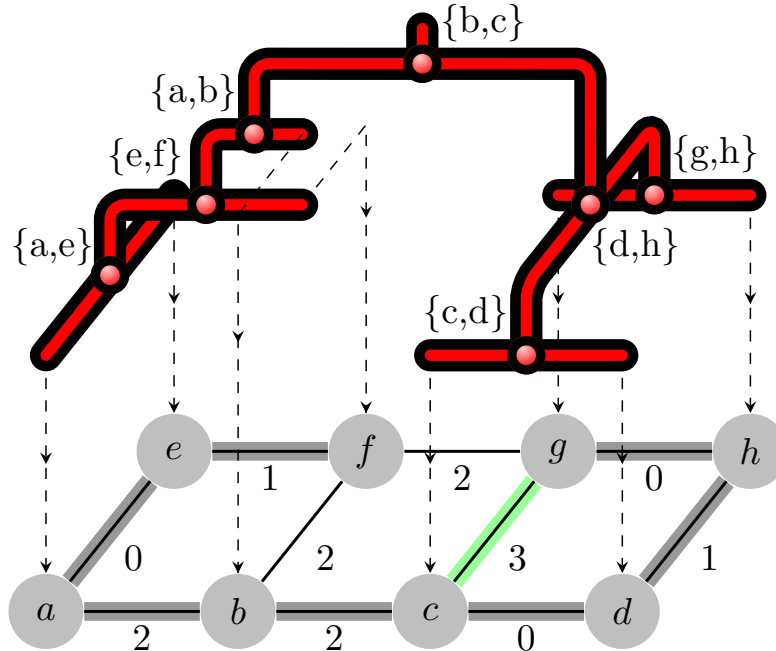
No new node

Binary Partition Tree by altitude ordering



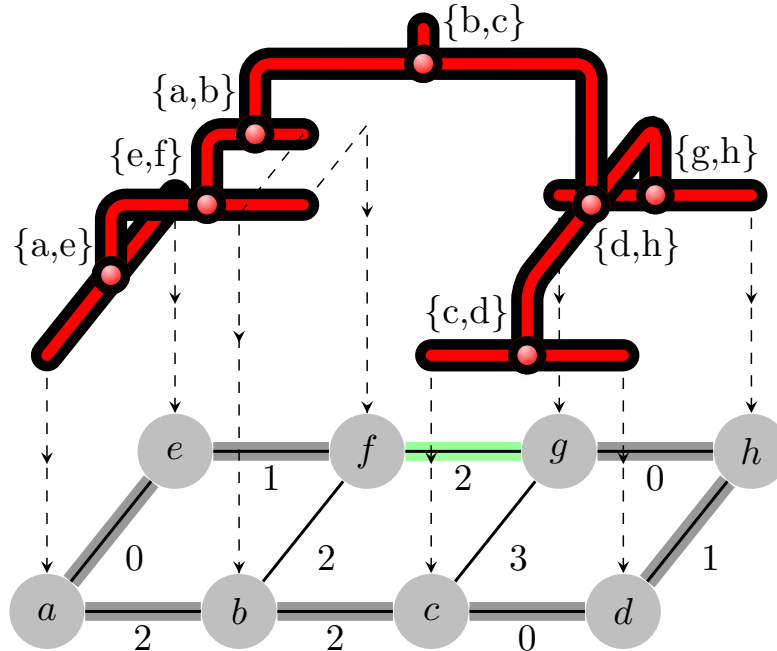
Seventh edge-node

Binary Partition Tree by altitude ordering



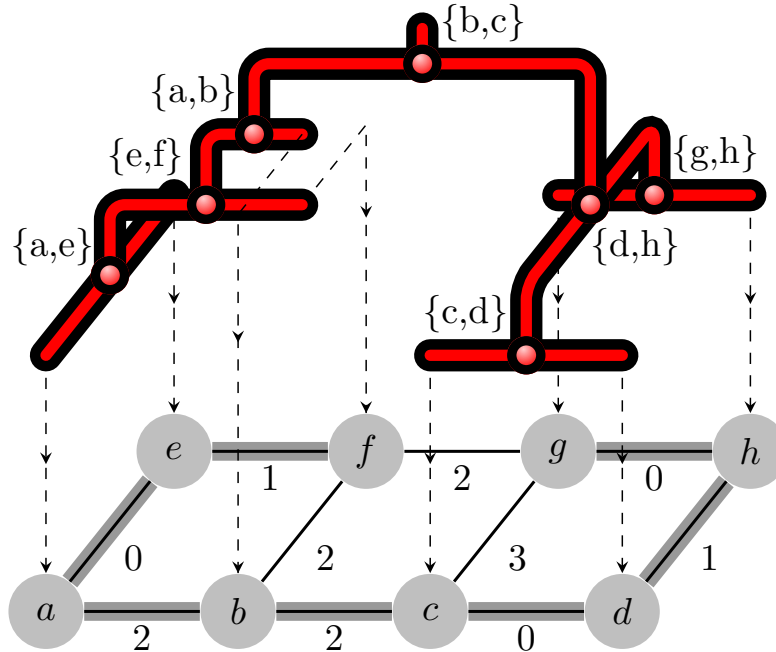
No new node

Binary Partition Tree by altitude ordering



No new node

Binary Partition Tree by altitude ordering



Final Q_{BT}

Q_{BT} Union-Find

Procedure $Q_{BT}.\text{MakeSet}(q)$

1 $Q_{BT}.\text{parent}[q] := -1; Q_{BT}.\text{size} += 1;$

Function $Q_{BT}.\text{FindCanonical}(q)$

1 **while** $Q_{BT}.\text{parent}[q] \geq 0$ **do** $q := Q_{BT}.\text{parent}[q];$
2 **return** $q;$

Function $Q_{BT}.\text{Union}(c_x, c_y)$

1 $Q_{BT}.\text{parent}[c_x] := Q_{BT}.\text{size}; Q_{BT}.\text{parent}[c_y] := Q_{BT}.\text{size};$
2 $Q_{BT}.\text{MakeSet}(Q_{BT}.\text{size});$
3 **return** $Q_{BT}.\text{size}-1;$

Q_{BT} Union-Find

Interest

The produced tree is useful

Drawback

The algorithm is slow : $O(|V|^2)$

Tarjan Union-Find

Interest

Quasi-linear complexity

Drawback

The produced tree is not useful for our purpose

Tarjan Union-Find

Procedure $Q_T.$ MakeSet(q)

1 $Q_T.$ parent[$Q_T.$ size] := -1; $Q_T.$ Rnk[$Q_T.$ size] := 0; $Q_T.$ size += 1;

Function $Q_T.$ FindCanonical(q)

1 $r := q$;
2 **while** $Q_T.$ parent[r] ≥ 0 **do** $r := Q_T.$ parent[r];
3 **while** $Q_T.$ parent[q] ≥ 0 **do** $tmp := q$; $q := Q_T.$ parent[q];
 $Q_T.$ parent[tmp] := r ;

Function $Q_T.$ Union(c_x , c_y)

1 **if** ($Q_T.$ Rnk[c_x] $>$ $Q_T.$ Rnk[c_y]) **then** swap(c_x , c_y);
2 **if** ($Q_T.$ Rnk[c_x] == $Q_T.$ Rnk[c_y]) **then** $Q_T.$ Rnk[c_y] += 1;
3 $Q_T.$ parent[c_x] := c_y ;
4 **return** c_y ;

Q_{EBT} : Efficient Q_{BT} Union-Find

Interest

- Combination of both Q_{BT} and Q_T .
- Quasi-linear complexity.
- One of the produced trees, Q_{BT} , is useful.

Q_{EBT} : Efficient Q_{BT} Union-Find

Procedure $Q_{EBT}.MakeSet(q)$

1 $Q_{EBT}.Root[q] := q; Q_{BT}.MakeSet(q); Q_T.MakeSet(q);$

Function $Q_{EBT}.Union(c_x, c_y)$

1 $t_u := Q_{EBT}.Root[c_x]; t_v := Q_{EBT}.Root[c_y];$
2 $Q_{BT}.parent[t_u] := Q_{BT}.parent[t_v] := Q_{BT}.size;$
3 $Q_{BT}.children[Q_{BT}.size].add(\{t_u\});$
4 $Q_{BT}.children[Q_{BT}.size].add(\{t_v\});$
5 $c := Q_T.Union(c_x, c_y); // \text{Union in } Q_T \text{ (with compression)}$
6 $Q_{EBT}.Root[c] := Q_{BT}.size; // \text{Update the root of } Q_{EBT}$
7 $Q_{BT}.MakeSet(Q_{BT}.size);$
8 **return** $Q_{BT}.size-1;$

Function $Q_{EBT}.FindCanonical(q)$

1 **return** $Q_T.FindCanonical(q);$

Outline

1 Binary Partition Tree and Minimum Spanning Tree

2 Post-Processing the binary tree

- Quasi-flat zones hierarchy
- Watershed-cut hierarchy
- Attribute-based hierarchies

Some helper functions

Function `getEdge(n)`

Data: a (non-leaf) node n of Q_{BT}

Result: the edge e of the MST corresponding to the n^{th} node

1 **return** $n - |V|;$

Function `weightNode(n)`

Data: a (non-leaf) node of the tree

Result: the weight of the MST edge associated with the n^{th} node
of Q_{BT}

1 **return** $F(MST[getEdge(n)]);$

Outline

1 Binary Partition Tree and Minimum Spanning Tree

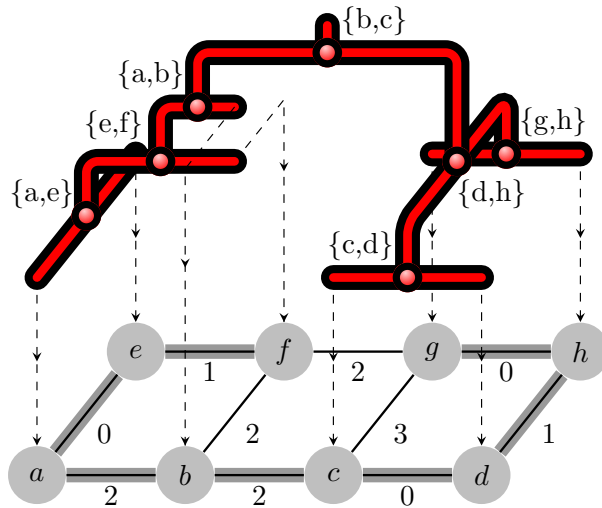
2 Post-Processing the binary tree

- Quasi-flat zones hierarchy
- Watershed-cut hierarchy
- Attribute-based hierarchies

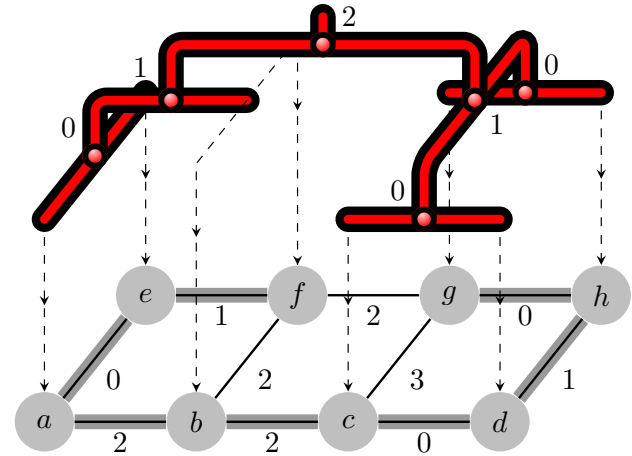
Q_{CT} : Quasi-flat zones hierarchy

- Also known as the α -tree.
- Also known as the Fuzzy Connectedness hierarchy.
- A quasi-linear algorithm: min-tree of the MST

Q_{CT} : Quasi-flat zones hierarchy



Q_{BT}



Q_{CT}

Quasi-flat zones hierarchy

Procedure Canonize Q_{BT}

Data: Q_{BT}

Result: Q_{CT} , a canonized version of Q_{BT}

```

1 for all nodes  $n$  of  $Q_{BT}$  do  $Q_{CT}.parent[n] := Q_{BT}.parent[n]$ ;  $Q_{CT}.size += 1$ ;
2 for each non-leaf and non-root node  $n$  of  $Q_{BT}$  by decreasing order do
3      $p := Q_{CT}.parent[n]$ ;
4     if ( $weightNode(p) == weightNode(n)$ ) then
5         for all  $c \in Q_{BT}.children[n]$  do  $Q_{CT}.parent[c] := p$ ;
6          $Q_{CT}.parent[n] := n$ ; // Delete node  $n$  of  $Q_{CT}$ 
// If needed, build the list of children
7 for all nodes  $n$  of  $Q_{CT}$  do
8      $p := Q_{CT}.parent[n]$ ; if  $p \geq 0$  and  $p \neq n$  then  $Q_{CT}.children[p].add(n)$ ;
```

Quasi-flat zones hierarchy

Q_{BT} or Q_{CT} ?

- It is possible to merge the min-tree algorithm with Kruskal's MST to obtain Q_{CT} in one step
- Q_{BT} contains more information than Q_{CT}
- The rest of the talk shows that computing Q_{CT} is not needed

Outline

1 Binary Partition Tree and Minimum Spanning Tree

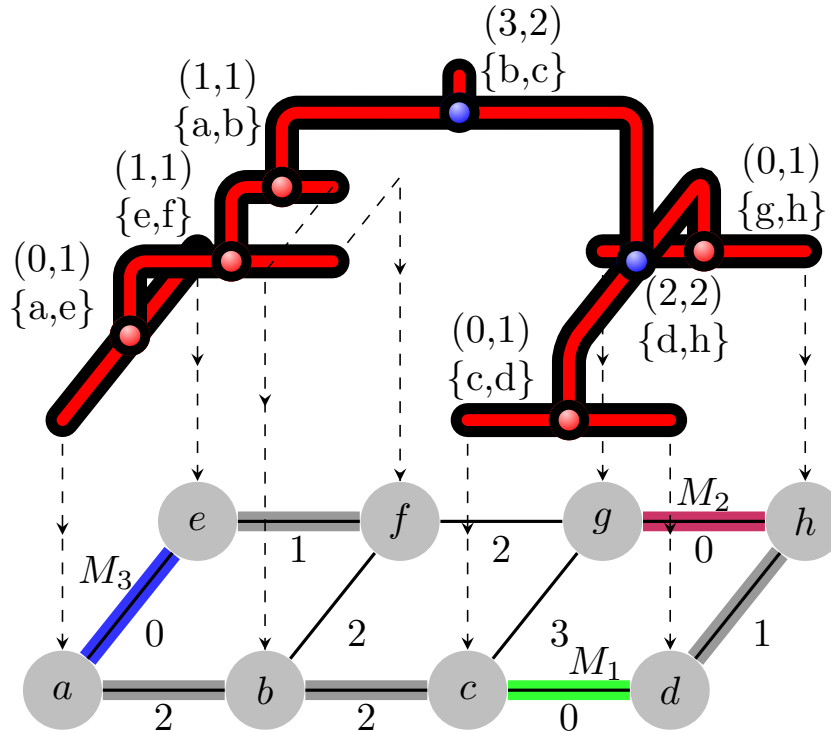
2 Post-Processing the binary tree

- Quasi-flat zones hierarchy
- **Watershed-cut hierarchy**
- Attribute-based hierarchies

Watershed cuts

- Partitions defined thanks to the *drop of water* principle
- Difficulty: non-uniqueness on flat zones (hence a choice)
- Also leads to a hierarchy (of watershed-cut partitions)
 - hierarchy by pass/connection value
 - (also known as Fuzzy connectedness)

Watershed cuts



Watershed cuts

Function watershed

Data: Q_{BT}

Result: A binary array ws indicating which MST edges are watershed

```
1 for all leaf-nodes  $n$  of  $Q_{BT}$  do minima[ $n$ ]:=0;
2 for each non-leaf node  $n$  of  $Q_{BT}$  by increasing order do
3   flag := TRUE;  $nb := 0$ ;
4   for all  $c \in Q_{BT}.children[n]$  do
5      $m := minima[c]$ ;  $nb := nb + m$ ;
6     if ( $m == 0$ ) then flag := FALSE;
7   ws[getEdge( $n$ )] := flag;
8   if ( $nb \neq 0$ ) then minima[ $n$ ] :=  $nb$ ;
9   else
10    if ( $n$  is the root of  $Q_{BT}$ ) then minima[ $n$ ] := 1;
11    else
12      $p := Q_{BT}.parent[n]$ ;
13     if ( $weightNode[n] < weightNode[p]$ ) then minima[ $n$ ] := 1;
14     else minima[ $n$ ]:=0;
```

Outline

1 Binary Partition Tree and Minimum Spanning Tree

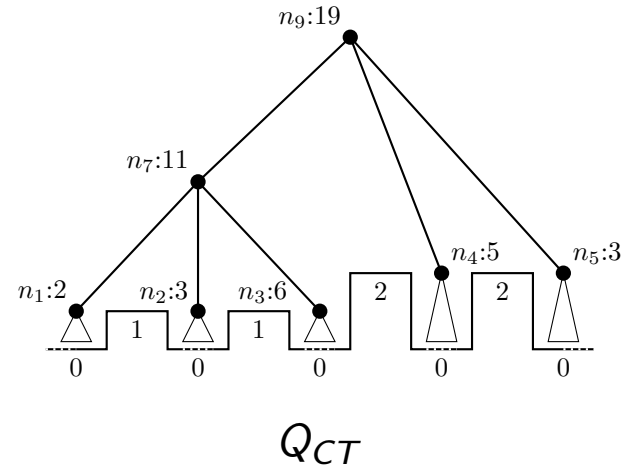
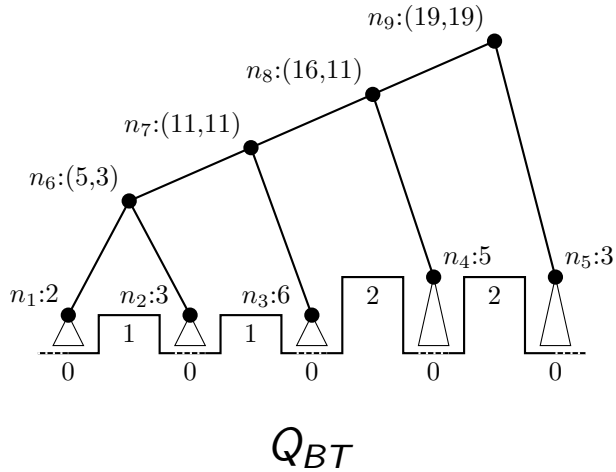
2 Post-Processing the binary tree

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- Attribute-based hierarchies

Attribute-based hierarchies

- Increasing attributes
- Watershed-cut framework
- Constrained-connectivity framework
 - The range criterion is indeed increasing
- Area-base, depth-based, volume-based hierarchies. . .
 - either from a watershed-cut hierarchy
 - or from a quasi-flat zone hierarchy

Attribute-based hierarchies



Attribute-based hierarchies

Function `getAttribute(n)`

Data: A node n of Q_{BT}

Result: The attribute at the time of the merging

```
1 if (n is the root) or (weightNode(parent[n])  $\neq$  weightNode(n)) then
2   |   for all c children of n do getAttribute(c);
3   |   attribute[n] := attributeComp[n];
4 else
5   |   max:=0;
6   |   for all children c of n do
7   |   |   v:=getAttribute(c);
8   |   |   if  $v > \text{max}$  then max := v;
9   |   |   attribute[n] := max;
10 return attribute[n];
```

Attribute-based hierarchies

Procedure ComputeMergeAttributeMST

Data: Q_{BT}

Result: a reweighted MST G corresponding to the attribute-based hierarchy

```
1 for any non-leaf node  $n$  of  $Q_{BT}$  do
2    $a_1 := \text{attribute}[\text{children}[n].\text{left}];$ 
3    $a_2 := \text{attribute}[\text{children}[n].\text{right}];$ 
4    $G[\text{getEdge}(n)] := \min(a_1, a_2);$ 
```

Conclusion

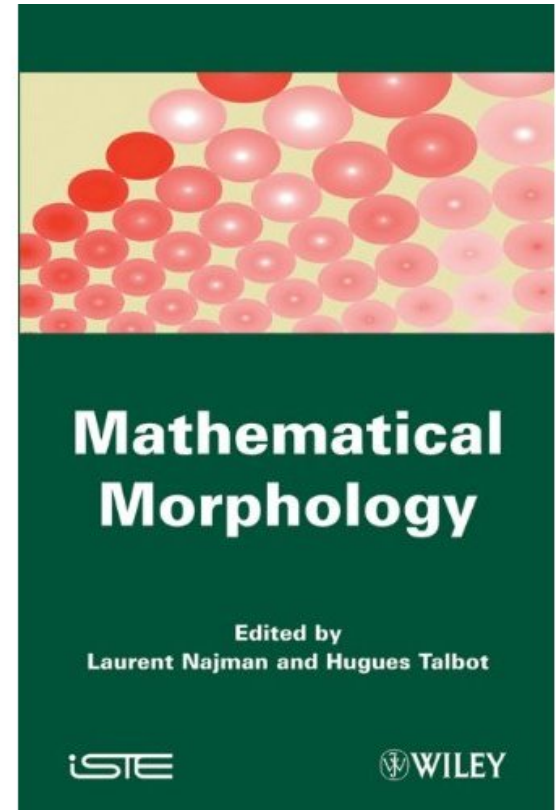
- Several elegant yet efficient algorithms for morphological trees
- Based on the Minimum Spanning Tree
- Other approaches than Kruskal can be used
- Unification theory in: Cousty, J., Najman, L., Perret, B.: *Constructive links between some morphological hierarchies on edge-weighted graphs*. (ISMM 2013).
- Source code at <http://www.esiee.fr/~info/sm/>

Summary of the poster content

	$\mathcal{PH}(G)$	$\mathcal{MH}(G)$	$\mathcal{PH}(T)$	$\mathcal{MH}(T)$	\mathcal{Q}	\mathcal{B}_{\prec}	\mathcal{H}_S
$\mathcal{PH}(G)$	\Leftrightarrow	\Leftrightarrow	\Rightarrow	\Rightarrow	\Rightarrow	\times	\times
$\mathcal{MH}(G)$	\Leftrightarrow	\Leftrightarrow	\Rightarrow	\Rightarrow	\Rightarrow	\times	\times
$\mathcal{PH}(T)$	\Leftarrow	\Leftarrow	\Leftrightarrow	\Leftrightarrow	\Rightarrow	\Leftarrow	\times
$\mathcal{MH}(T)$	\Leftarrow	\Leftarrow	\Leftrightarrow	\Leftrightarrow	\Rightarrow	\Leftarrow	\times
\mathcal{Q}	\Leftarrow	\Leftarrow	\Leftarrow	\Leftarrow	\Leftrightarrow	\Rightarrow	\times
\mathcal{B}_{\prec}	\times	\times	\Rightarrow	\Rightarrow	\Rightarrow	\Leftrightarrow	\Rightarrow
\mathcal{H}_S	\times	\times	\times	\times	\times	\Leftarrow	\Leftrightarrow

Table 1: Summary of the main results.

Thank for your attention !



Pink: <http://pinkhq.com>

Olena: <http://www.lrde.epita.fr/cgi-bin/twiki/view/Olena>