

Connected Component Tree

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FOUR-DAY COURSE
on
Mathematical Morphology in image analysis
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ESIEE
ENGINEERING

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— PARIS-EST



Motivation

- Connectivity plays important role for image processing
- Morphological connected filters
 - Morphological filters acting on components of a set
 - Morphological filters acting on components of the flat zones of a function
- Graphs are adapted for handling digital connectivity

Problem

- **How can connected filters be efficiently implemented?**

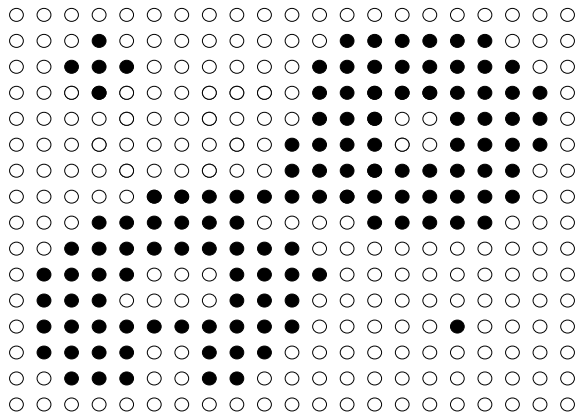
Outline

- Connected filters on (weighted) graphs
 - Connected components
 - Component tree
 - Example of application
- Algorithms for connected filters on (weighted) graphs
 - Union of disjoint set
 - Connected component labeling
 - Component tree extraction

Structuring a discrete set

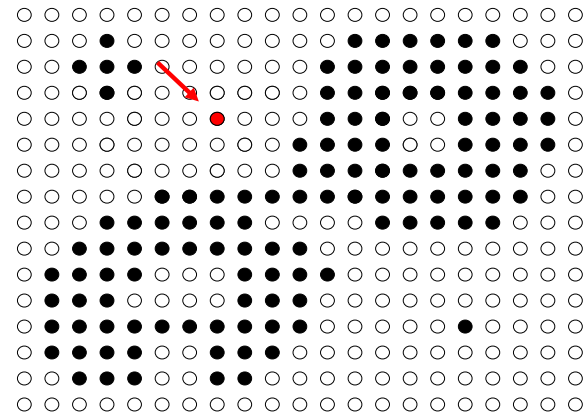
- Let V be a set
 - (eg, the set of all image pixels)
- Let X be a subset of V
 - (eg, a binary object to be processed)

Discrete set



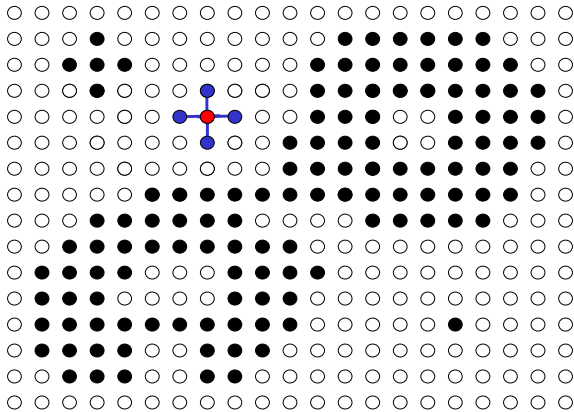
V (white & black dots) and X (black dots)

Discrete set



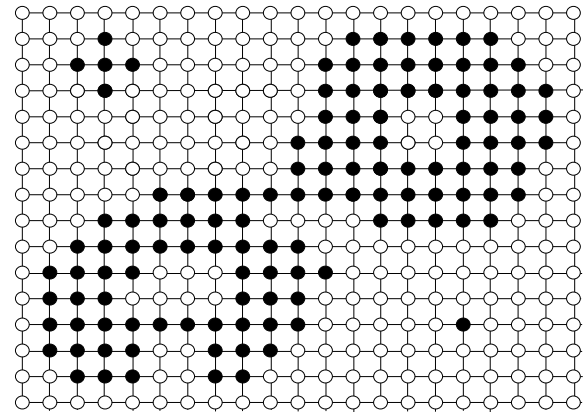
V (white & black dots) and X (black dots)

Graph



Neighborhood (in blue) of a point (in red)

Graph



Vertex set (dots) and edge set E (line segments) of a graph (V,E)

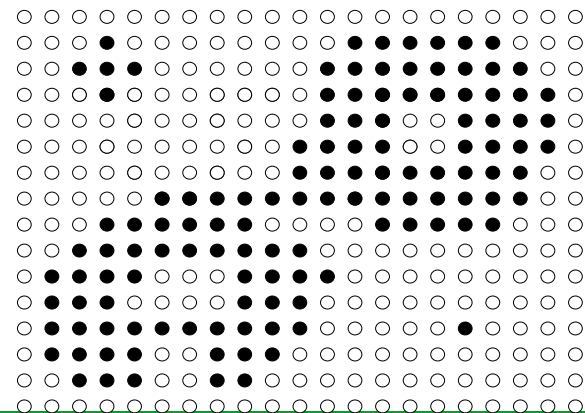
Connected sets

- Let $G = (V, E)$ be a (undirected) graph
- Let X be a subset of V

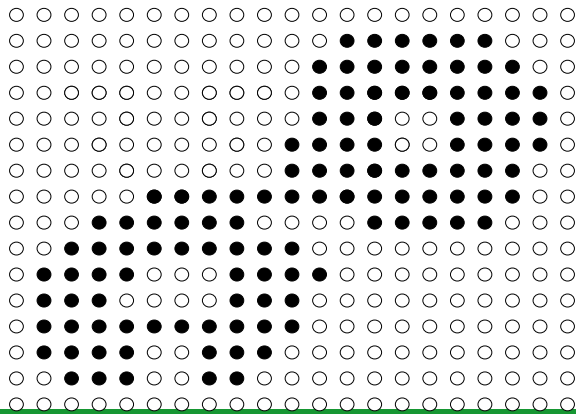
Definition

- X is **connected** if
 - there exists a path between any two points of X

Example: Discrete set that is not connected



Example: Connected discrete set



Connected components

- Let $G = (V, E)$ be a (undirected) graph.
- Let X be a subset of V
- Let Y be a subset of X

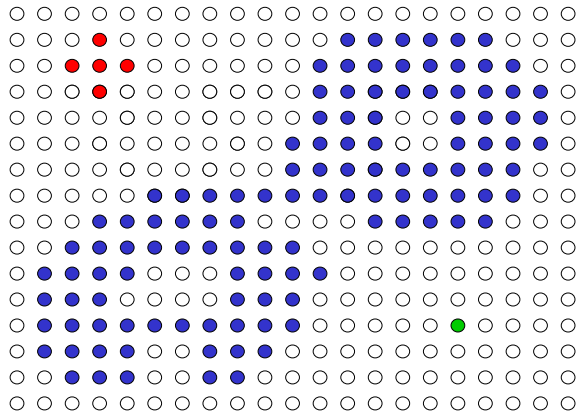
Definition

- Y is a **(connected) component of X** if
 - Y is connected
 - no proper superset of Y included in X is connected

Property

- Y is a **component of X** iff for any x in Y
 - Y is the union of all connected subsets of X that contains x

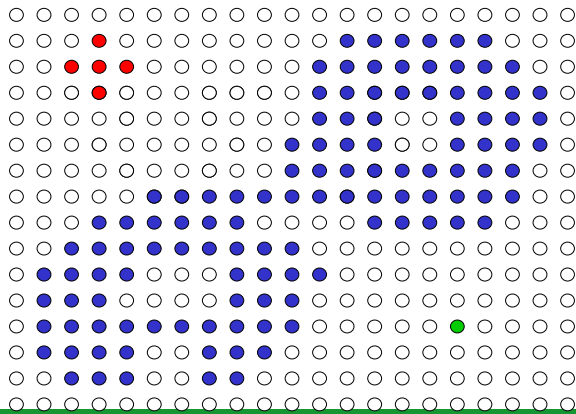
Example : Connected components partition



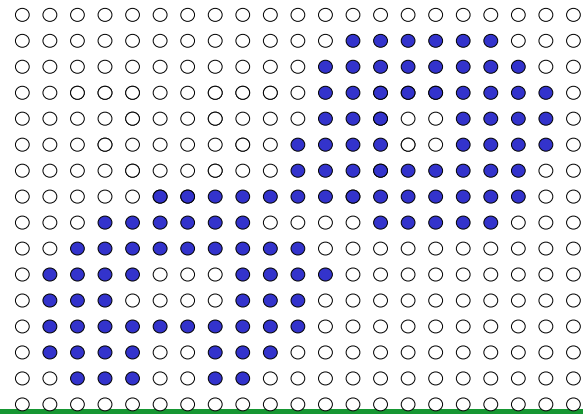
Connected opening a discrete set

- According to a given criterion (e.g., the size), keep the most important components of X

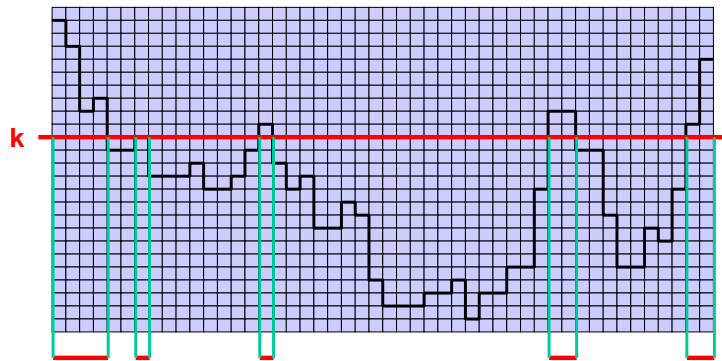
Example: area filtering a discrete set



Example: area filtering a discrete set



From maps to sets

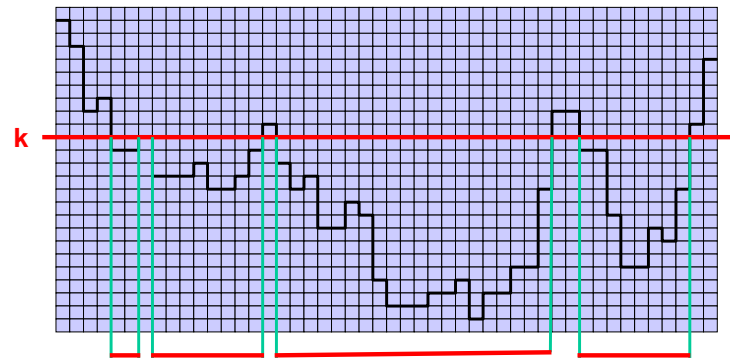


$$F_k = \{x \in E, F(x) \geq k\}, \text{ i.e., points above } k$$

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From maps to sets



$$\bar{F}_k = \{x \in E, F(x) < k\}, \text{ i.e., points below } k$$

J. Serra, J. Cousty, B.S. Daya Sagar: Course on Math. Morphology

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Structuring a discrete map: level sets decomposition

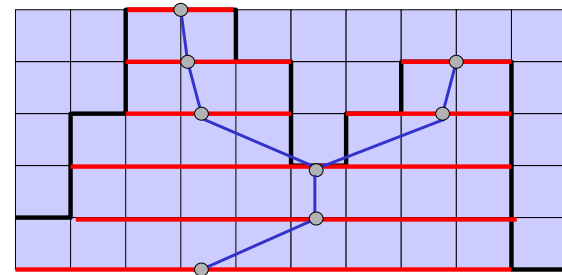
Definition

- Let F be a map from V to N and let k be an integer
- The **(upper) level-set of F at level k** is the set made of the points of V whose value is greater than or equal to k
 - $F_k = \{x \in E, F(x) \geq k\}$
- Any connected component of any (upper) level set of F is called a **(upper) component of F**

Remark

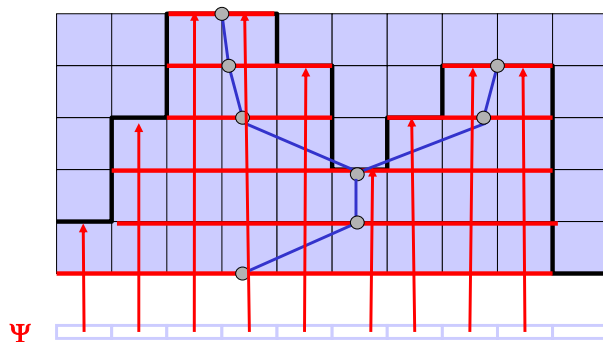
- Any two components of F are either nested or disjoint

Structuring a discrete map: **(upper)** Component tree

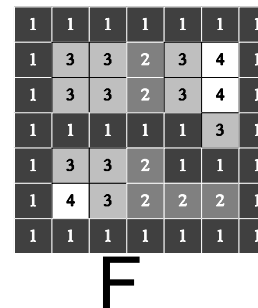


Components + inclusion relation \rightarrow tree structure

Structuring a discrete map: (*upper*) Component tree



Level-sets, components: the 2D case



Level-sets, components: the 2D case

1	1	1	1	1	1	1
1	3	3	2	3	4	1
1	3	3	2	3	4	1
1	1	1	1	1	3	1
1	3	3	2	1	1	1
1	4	3	2	2	2	1
1	1	1	1	1	1	1

F

1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1
1	1	1	1	1	1	1

F_1

Level-sets, components: the 2D case

1	1	1	1	1	1	1
1	3	3	2	3	4	1
1	3	3	2	3	4	1
1	1	1	1	1	3	1
1	3	3	2	1	1	1
1	4	3	2	2	2	1
1	1	1	1	1	1	1

F

0	0	0	0	0	0	0
0	1	1	1	1	1	0
0	1	1	1	1	1	0
0	0	0	0	0	1	0
0	1	1	1	0	0	0
0	1	1	1	1	1	0
0	0	0	0	0	0	0

F_2

Level-sets, components: the 2D case

1	1	1	1	1	1	1
1	3	3	2	3	4	1
1	3	3	2	3	4	1
1	1	1	1	1	3	1
1	3	3	2	1	1	1
1	4	3	2	2	2	1
1	1	1	1	1	1	1

F

0	0	0	0	0	0	0
0	1	1	0	1	1	0
0	1	1	0	1	1	0
0	0	0	0	0	1	0
0	1	1	0	0	0	0
0	1	1	0	0	0	0
0	0	0	0	0	0	0

F₃

Level-sets, components: the 2D case

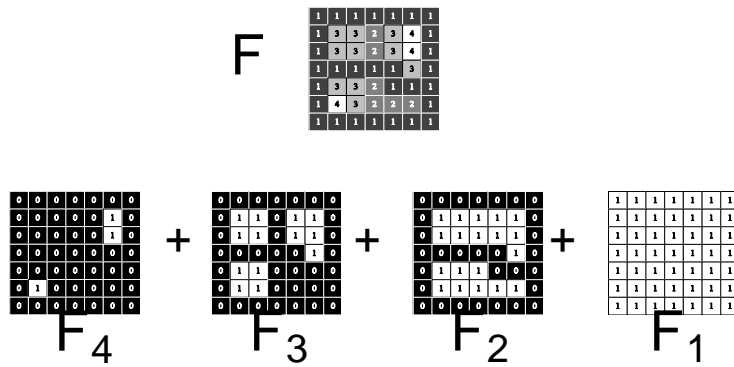
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1	3	3	2	3	4	1
1	3	3	2	3	4	1
1	1	1	1	1	3	1
1	3	3	2	1	1	1
1	4	3	2	2	2	1
1	1	1	1	1	1	1

F

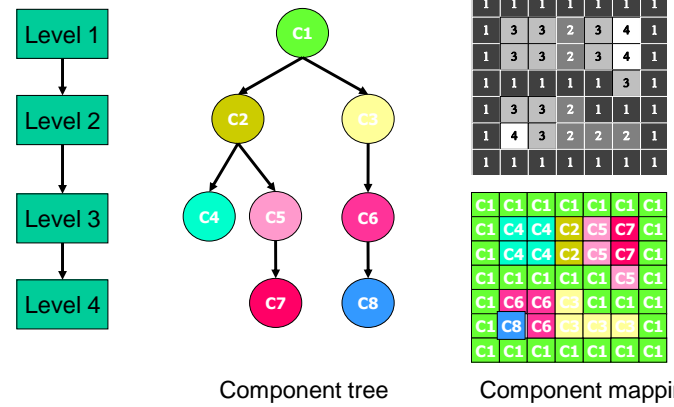
0	0	0	0	0	0	0
0	0	0	0	0	1	0
0	0	0	0	0	1	0
0	0	0	0	0	0	0
0	0	0	0	0	0	0
0	1	0	0	0	0	0
0	0	0	0	0	0	0

F₄

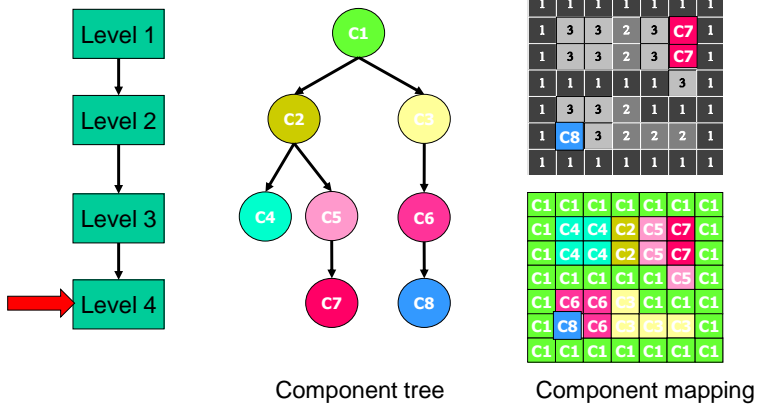
Cross-sections, components: the 2D case



Component tree: the 2D case

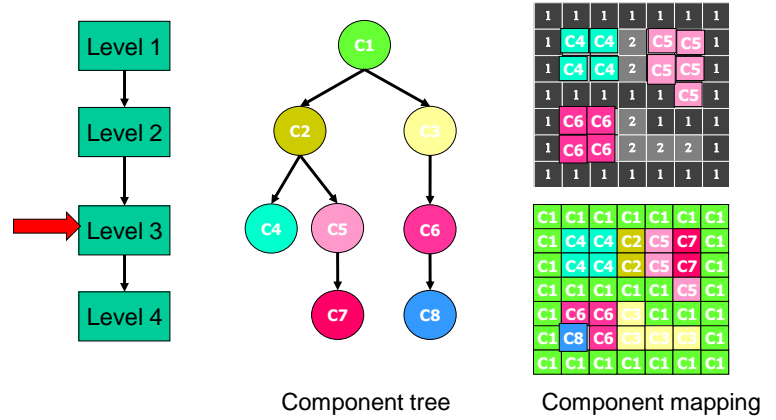


Component tree: the 2D case



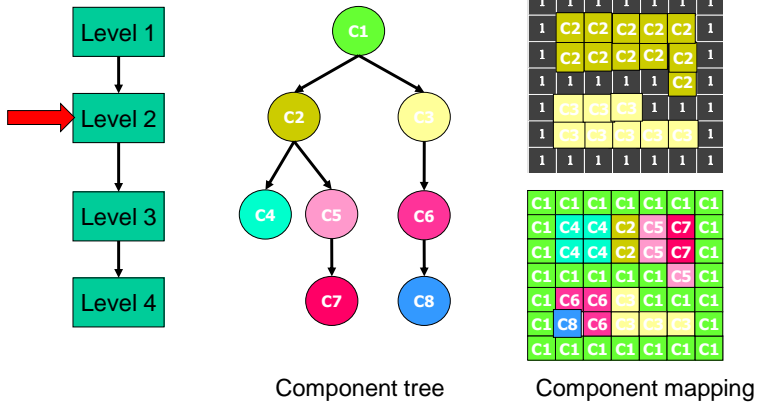
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Component tree: the 2D case

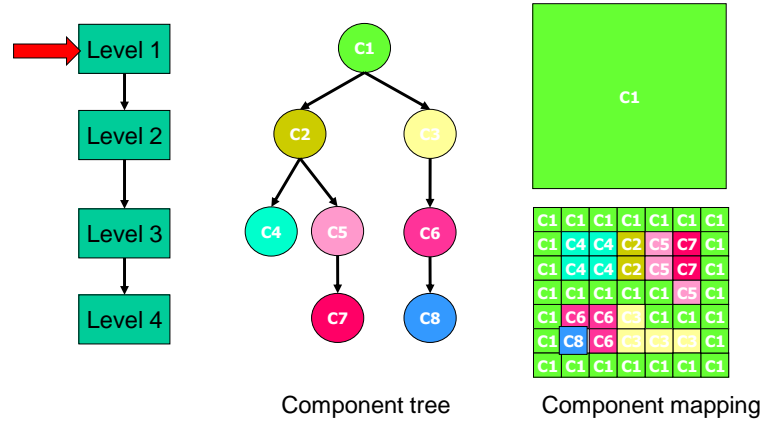


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Component tree: the 2D case



Component tree: the 2D case

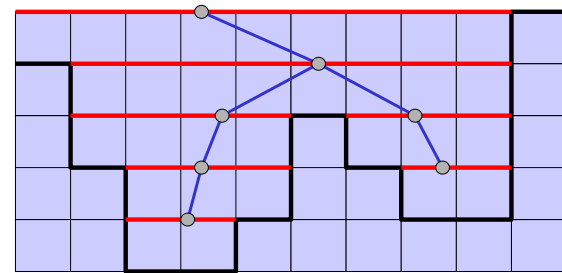


Filtering a discrete map : flooding

- A leaf of the component tree T of F is:
 - a **regional minimum of F** , if T is a lower component tree
 - a **regional maximum of F** , if T is an upper component tree
- According to a given criterion, we can iteratively remove the leaf components of F
- Such a removal is called a flooding

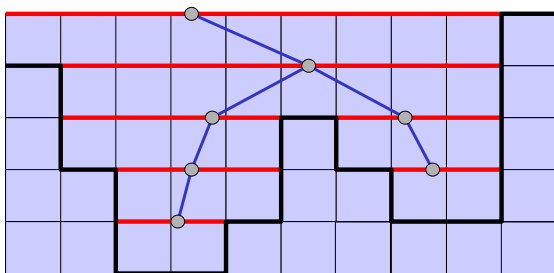
Flooding: intuitive example

step 1



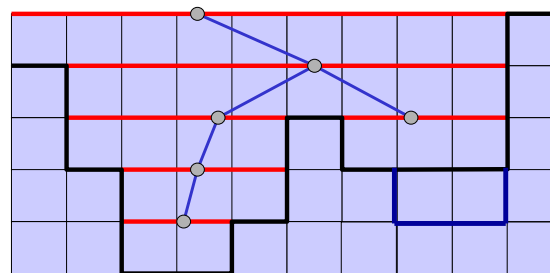
Flooding: intuitive example

step 1



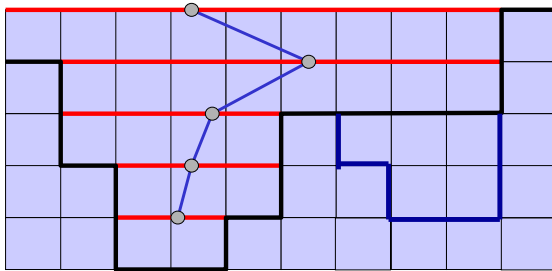
Flooding: intuitive example

step 2



Flooding: intuitive example

step 3



We remove one branch of T
We flood one "lobe" of F

Flooding: formal definition

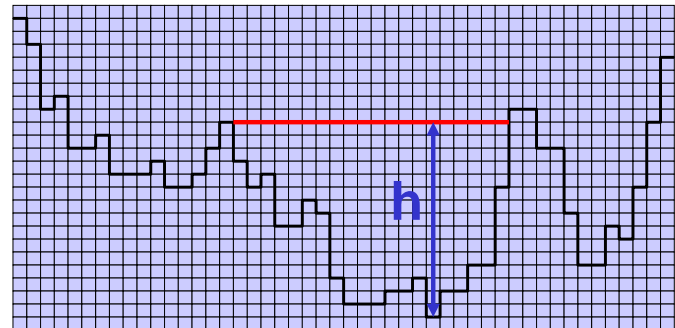
Definition

- Let x in V
- The **elementary flooding of F at x** is defined by:
 - $\eta_x F(y) = F(y) + 1$ if x and y belong to the same minimum
 - $\eta_x F(y) = F(y)$ otherwise
- Any map obtained from F by composition of elementary floodings is called a **flooding of F**

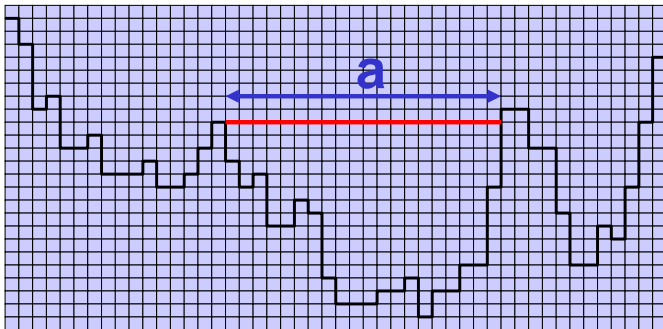
Filtering a grayscale map

- What criterion (attribute) to select the non-significant minima to be flooded?

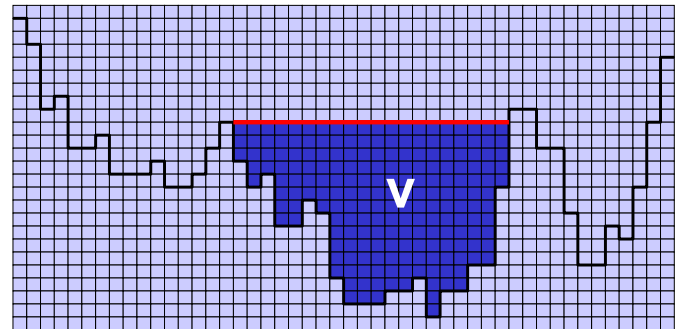
Attributes: height



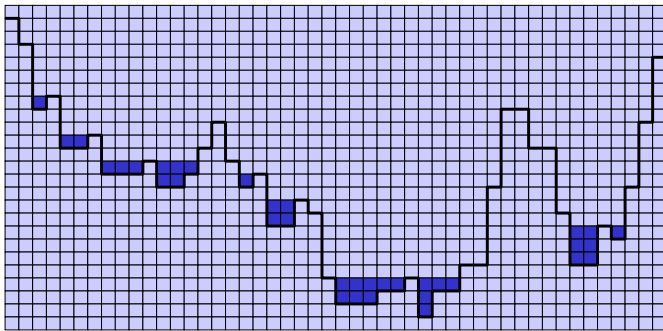
Attributes: area



Attributes: volume

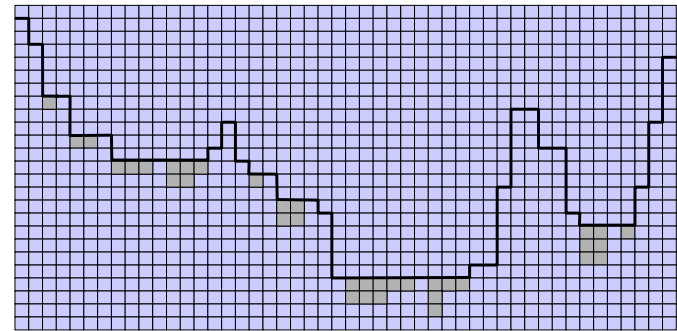


Filtering by volume



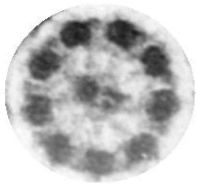
Regional minima

Filtering by volume

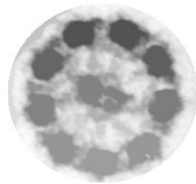
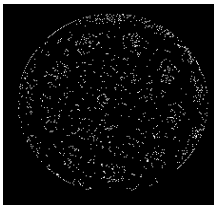


Regional minima

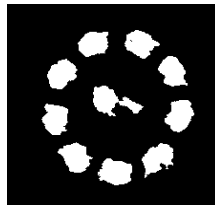
Application: keep the 10 most volumic lobes



Original image (top) and
its many minima (bottom)



Filtered image (top) and its
10 minima (bottom)



Algorithms requested for achieving this result

- **Component tree algorithm**
- Flooding the minima with attribute below a given threshold
- Or alternatively, keeping only the k most significant minima

Algorithms for the component tree

- There exist several algorithms for the component tree
 - Breen & Jones 1996
 - Salembier, Oliveras & Garido 1998
 - Mattes & Demongeot 2000
 - **Coupric & Najman 2006**
- **Tarjan's union find**
- **Connected components labeling**
- **Quasi-linear time component tree extraction**

Disjoint set problem

- Maintaining a collection of **disjoint subsets** of a set V under the **operation of union**
- Each subset X is represented by a unique element of X called the *canonical element of X*
- Three operations allow the collection to be managed:
 - **MakeSet(x)**: add the set $\{x\}$ to the collection
 - **Find(x)**: return the canonical element of the set containing x
 - **Link(x,y)**:
 - Remove the two sets whose canonical elements are x and y
 - Add their union

Tarjan's union-find algorithm

Procedure `MakeSet`(*element* x)
`Fth`(x) := x ; `Rnk`(x) := 0;

Function `element Find` (*element* x)
`if` (`Fth`(x) $\neq x$) `then` `Fth`(x) := `Find`(`Fth`(x));
`return` `Fth`(x);

Function `element Link` (*element* x , *element* y)
`if` (`Rnk`(x) > `Rnk`(y)) `then` `exchange`(x , y);
`if` (`Rnk`(x) == `Rnk`(y)) `then` `Rnk`(y) := `Rnk`(y) + 1;
`Fth`(x) := y ;
`return` y ;

- **Quasi-linear algorithm:** $O(m \times \alpha(m, |V|))$
 - for m operations
 - α : inverse of Ackerman function (in practice never greater than 4)

Connected components labeling with union-find

Algorithm 1: `BuildConnectedComponents`

Data: (E, Γ) - graph
Data: A set $X \subseteq E$
Result: M - map from X to E

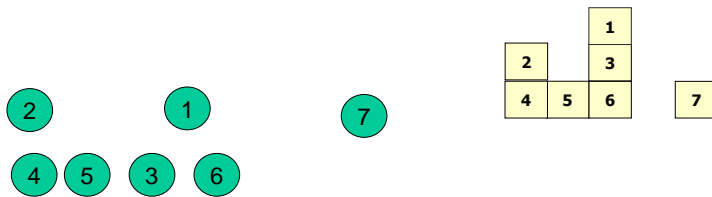
- 1 `foreach` $p \in X$ `do` `MakeSet`(p);
- 2 `foreach` $p \in X$ `do`
- 3 `compp` := `Find`(p);
- 4 `foreach` $q \in \Gamma(p)$ *such that* $q \in X$ `do`
- 5 `compq` := `Find`(q);
- 6 `if` (`compp` \neq `compq`) `then` `compp` := `Link`(`compq`, `compp`);
- 8 `for` $p \in X$ `do` `M`(p) := `Find`(p);

Connected components labeling with union-find

Algorithm 1: BuildConnectedComponents

```

Data:  $(E, \Gamma)$  - graph
Data: A set  $X \subseteq E$ 
Result:  $M$  - map from  $X$  to  $E$ 
1 foreach  $p \in X$  do MakeSet( $p$ );
2 foreach  $p \in X$  do
3   comp $p$  := Find( $p$ );
4   foreach  $q \in \Gamma(p)$  such that  $q \in X$  do
5     comp $q$  := Find( $q$ );
6     if (comp $p$   $\neq$  comp $q$ ) then comp $p$  := Link(comp $q$ , comp $p$ );
8 for  $p \in X$  do  $M(p)$  := Find( $p$ );
    
```

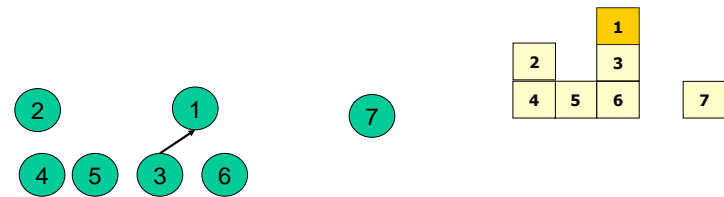


Connected components labeling with union-find

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Connected components labeling with union-find

Algorithm 1: BuildConnectedComponents

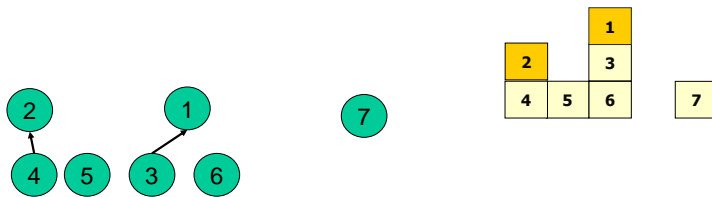
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Connected components labeling with union-find

Algorithm 1: BuildConnectedComponents

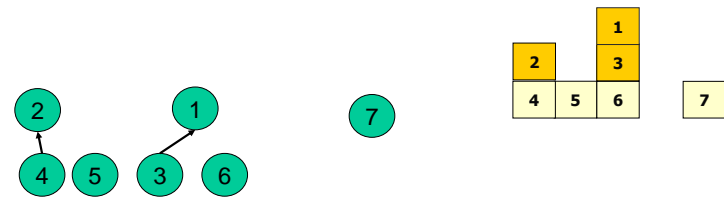
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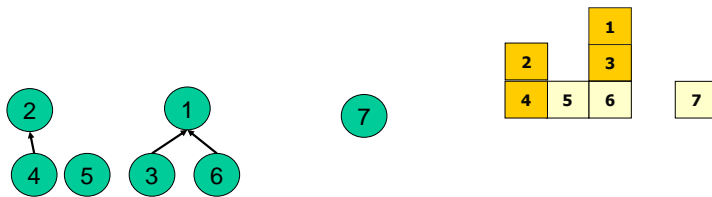


Connected components labeling with union-find

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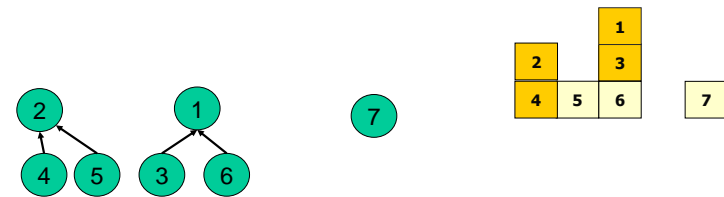


Connected components labeling with union-find

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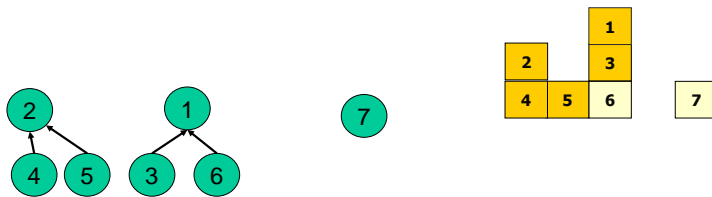


Connected components labeling with union-find

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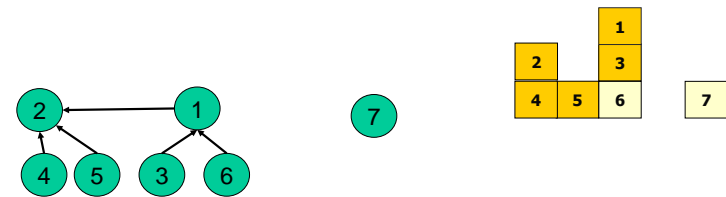


Connected components labeling with union-find

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Data: A set  $X \subseteq E$ 
Result:  $M$  - map from  $X$  to  $E$ 
1 foreach  $p \in X$  do MakeSet( $p$ );
2 foreach  $p \in X$  do
3   comp $p$  := Find( $p$ );
4   foreach  $q \in \Gamma(p)$  such that  $q \in X$  do
5     comp $q$  := Find( $q$ );
6     if (comp $p$   $\neq$  comp $q$ ) then comp $p$  := Link(comp $q$ , comp $p$ );
8 for  $p \in X$  do  $M(p)$  := Find( $p$ );
    
```



Connected components labeling with union-find

Algorithm 1: BuildConnectedComponents

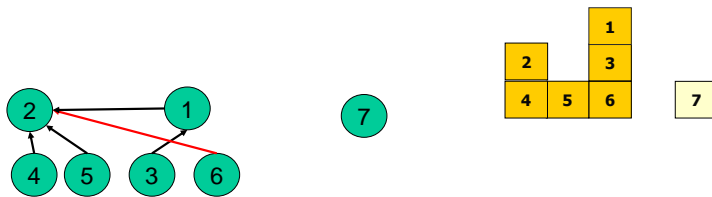
Data: (E, Γ) - graph

Data: A set $X \subseteq E$

Result: M - map from X to E

```

1 foreach  $p \in X$  do MakeSet( $p$ );
2 foreach  $p \in X$  do
3    $comp_p := \text{Find}(p)$ ;
4   foreach  $q \in \Gamma(p)$  such that  $q \in X$  do
5      $comp_q := \text{Find}(q)$ ;
6     if ( $comp_p \neq comp_q$ ) then  $comp_p := \text{Link}(comp_q, comp_p)$ ;
8 for  $p \in X$  do  $M(p) := \text{Find}(p)$ ;
    
```



Connected components labeling with union-find

Algorithm 1: BuildConnectedComponents

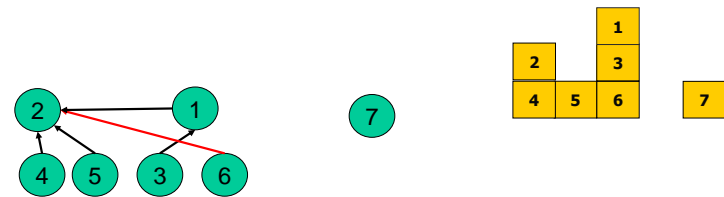
Data: (E, Γ) - graph

Data: A set $X \subseteq E$

Result: M - map from X to E

```

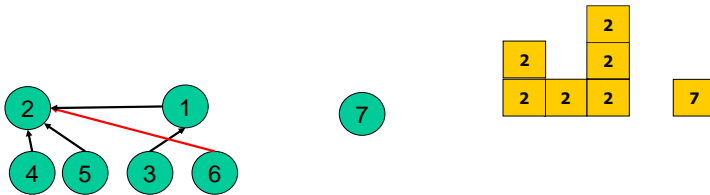
1 foreach  $p \in X$  do MakeSet( $p$ );
2 foreach  $p \in X$  do
3    $comp_p := \text{Find}(p)$ ;
4   foreach  $q \in \Gamma(p)$  such that  $q \in X$  do
5      $comp_q := \text{Find}(q)$ ;
6     if ( $comp_p \neq comp_q$ ) then  $comp_p := \text{Link}(comp_q, comp_p)$ ;
8 for  $p \in X$  do  $M(p) := \text{Find}(p)$ ;
    
```



Connected components labeling with union-find

Algorithm 1: BuildConnectedComponents

```
Data:  $(E, \Gamma)$  - graph  
Data: A set  $X \subseteq E$   
Result:  $M$  - map from  $X$  to  $E$   
1 foreach  $p \in X$  do MakeSet( $p$ );  
2 foreach  $p \in X$  do  
3    $comp_p := \text{Find}(p)$ ;  
4   foreach  $q \in \Gamma(p)$  such that  $q \in X$  do  
5      $comp_q := \text{Find}(q)$ ;  
6     if ( $comp_p \neq comp_q$ ) then  $comp_p := \text{Link}(comp_q, comp_p)$ ;  
8 for  $p \in X$  do  $M(p) := \text{Find}(p)$ ;
```



Component tree algorithm

- General case
 - Two collections of disjoint subsets of V
- Particular case: all values are distinct
 - One collection of disjoint subsets of V

Component tree algorithm: all values distinct

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

Component tree algorithm: all values distinct (init)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

Image

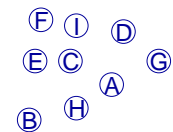
A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

Lowest Node

A	B	C
D	E	F
G	H	I

nodes

Disjoint sets



Component tree algorithm: all values distinct (step 1)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = C$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

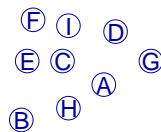
Lowest Node

A	B	C
D	E	F
G	H	I

nodes

Ⓒ

Disjoint sets



Component tree algorithm: all values distinct (step 2)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = G$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

Lowest Node

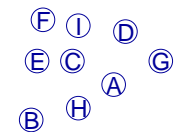
A	B	C
D	E	F
G	H	I

nodes

Ⓒ

Ⓒ

Disjoint sets



Component tree algorithm: all values distinct (step 3)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = A$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

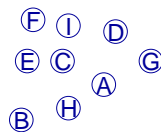
Lowest Node

A	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 4)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = D$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

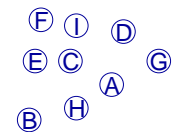
Lowest Node

A	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 4)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = D$
 $y = A$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

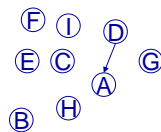
Lowest Node

D	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 4)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = D$
 $y = G$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

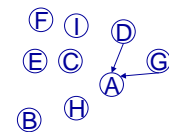
Lowest Node

D	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 5)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = H$
 $y = G$

Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

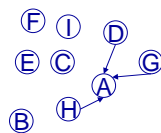
Lowest Node

H	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 6)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = E$
 $y = D$

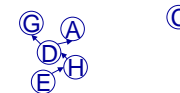
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

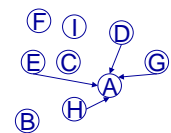
Lowest Node

E	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 6)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = E$
 $y = H$

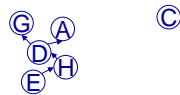
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

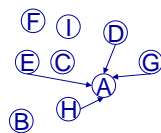
Lowest Node

E	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 7)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = I$
 $y = H$

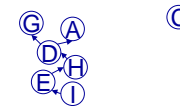
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

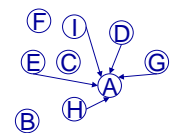
Lowest Node

I	B	C
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 8)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = F$
 $y = C$

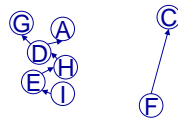
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

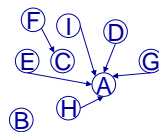
Lowest Node

I	B	F
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 7)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = F$
 $y = E$

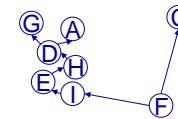
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

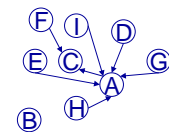
Lowest Node

I	B	F
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 7)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = F$
 $y = I$

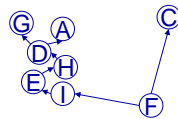
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

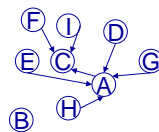
Lowest Node

I	B	F
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 9)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = B$
 $y = A$

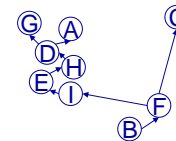
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

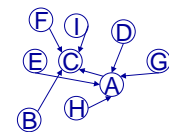
Lowest Node

I	B	B
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct (step 10)

- **foreach** point x in V in decreasing order of altitude **do**
 - **foreach** neighbor y of x **do**
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do**
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$

$x = B$
 $y = C$

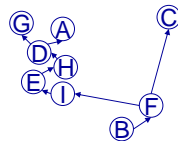
Image

A:7	B:1	C:9
D:6	E:4	F:2
G:8	H:5	I:3

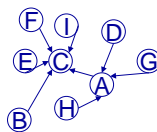
Lowest Node

I	B	B
D	E	F
G	H	I

nodes



Disjoint sets



Component tree algorithm: all values distinct

- **foreach** point x in V in decreasing order of altitude **do** $O(|V|)$
 - **foreach** neighbor y of x **do** $O(|V|+|E|)$
 - If $(F(y) > F(x))$ and $(\text{Find}(x) \neq \text{Find}(y))$ **do** $O(|E| \cdot \alpha(|E|, |V|))$
 - $\text{adjNode} := \text{lowestNode}[\text{Find}(y)]$ $O(|E|)$
 - $\text{nodes}[x].\text{addChild}(\text{nodes}[\text{adjNode}])$ $O(|E|)$
 - $\text{curNode} := \text{Link}(x, \text{adjNode})$ $O(|E| \cdot \alpha(|E|, |V|))$
 - $\text{lowestNode}[\text{curNode}] := \text{nodes}[x]$ $O(|E|)$

- **Time complexity : quasi-linear**

- with respect to the number of edges and vertices
- Requires a sorting pre-processing step

Component tree algorithm: general case

- Two union-find implementations needed
 - One for building the union of connected components over height
 - Similar to the one in the case where all values were distinct
 - One for building the connected components of a given altitude

Build the component tree

Algorithm 2: BuildComponentTree

Data: (E, F, F) - weighted graph with N points.
Result: N_n - number of nodes of the component tree ($\leq N$).
Result: $nodes$ - array $[0 \dots N - 1]$ of nodes.
Result: $Root$ - Root of the component tree
Result: C - map from E to $[0 \dots N - 1]$ (component mapping).
Local: $subtreeRoot$ - map from $[0 \dots N - 1]$ to $[0 \dots N - 1]$.

```

1  $N_n := N$ ; Sort the points in decreasing order of level for  $F$ ;
2 foreach  $p \in E$  do {MakeSet1( $p$ ); MakeSet2( $p$ );  $nodes[p] := MakeNode(F(p))$ ;  $subtreeRoot[p] := p$ };
3 foreach  $p \in E$  in decreasing order of level for  $F$  do
4    $curCanonicalElt := Find1(p)$ ;
5    $curNode := Find2(subtreeRoot[curCanonicalElt])$ ;
6   foreach already processed neighbor  $q$  of  $p$  with  $F(q) \geq F(p)$  do
7      $adjCanonicalElt := Find1(q)$ ;
8      $adjNode := Find2(subtreeRoot[adjCanonicalElt])$ ;
9     if ( $curNode \neq adjNode$ ) then
10      if ( $nodes[curNode] \rightarrow level == nodes[adjNode] \rightarrow level$ ) then
11         $curNode := MergeNodes(adjNode, curNode)$ ;
12      else
13        //We have  $nodes[curNode] \rightarrow level < nodes[adjNode] \rightarrow level$ 
14         $nodes[curNode] \rightarrow addSon(nodes[adjNode])$ ;
15         $curCanonicalElt := Link1(adjCanonicalElt, curCanonicalElt)$ ;
16         $subtreeRoot[curCanonicalElt] := curNode$ ;
17
18  $Root := subtreeRoot[Find1(Find2(0))]$ ;
19 foreach  $p \in E$  do  $C(p) := Find2(p)$ ;

```

Auxiliary functions

Function node MakeNode (int level)

allocate a new node n with an empty list of sons;
 $n \rightarrow \text{level} := \text{level};$
return n;

Function int MergeNodes (int node1, int node2)

$\text{tmpNode} := \text{Link2}(\text{node1}, \text{node2});$
if ($\text{tmpNode} == \text{node2}$) **then**
 | Add the list of sons of $\text{nodes}[\text{node1}]$ to the list of sons of $\text{nodes}[\text{node2}]$;
else
 | Add the list of sons of $\text{nodes}[\text{node2}]$ to the list of sons of $\text{nodes}[\text{node1}]$;
 $N_n := N_n - 1;$
return tmpNode ;

Example

110	90	100	0	1	2	12;	[120]
50	50	50	3	4	5	0;	[110]
40	50	50	6	7	8	2;	[100]
50	50	50	9	10	11	1;	[90]
120	70	80	12	13	14	14;	[80]
						13;	[70]
						3; 4; 5; 8; 9; 10; 11;	[50]
						6;	[40]
						7	[20]

step 1

110	90	100
50	50	50
40	50	50
50	50	50
120	70	80

0	1	2
3	4	5
6	7	8
9	10	11
12	13	14

Fth1

0	1	2
3	4	5
6	7	8
9	10	11
12	13	14

Fth2

0	1	2
3	4	5
6	7	8
9	10	11
12	13	14

SubtreeRoot

[12] 120

Step 2

110	90	100
50	50	50
40	50	50
50	50	50
120	70	80

0	1	1
3	4	5
6	7	8
9	10	11
12	13	14

Fth1

0	1	2
3	4	5
6	7	8
9	10	11
12	13	14

Fth2

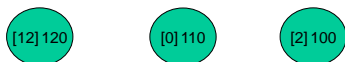
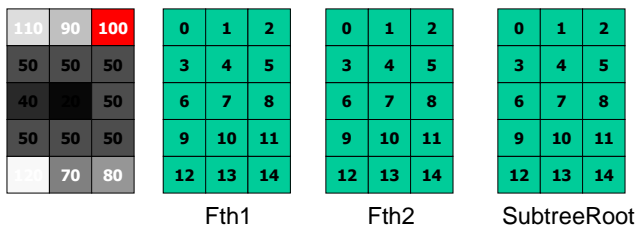
0	1	2
3	4	5
6	7	8
9	10	11
12	13	14

SubtreeRoot

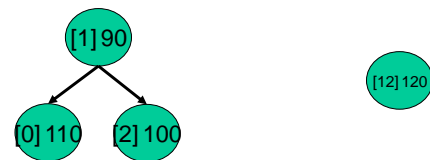
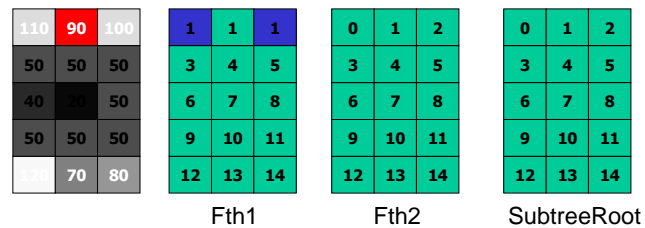
[12] 120

[0] 110

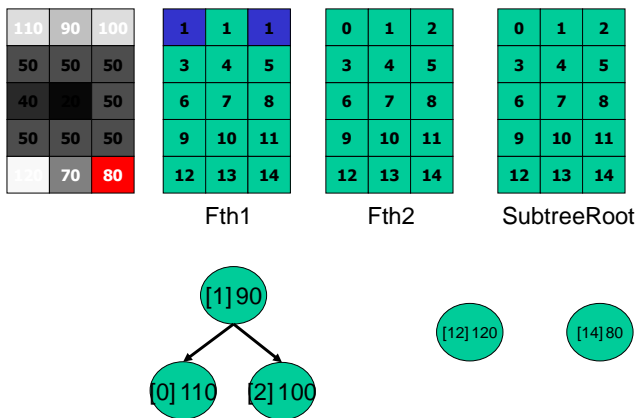
step 3



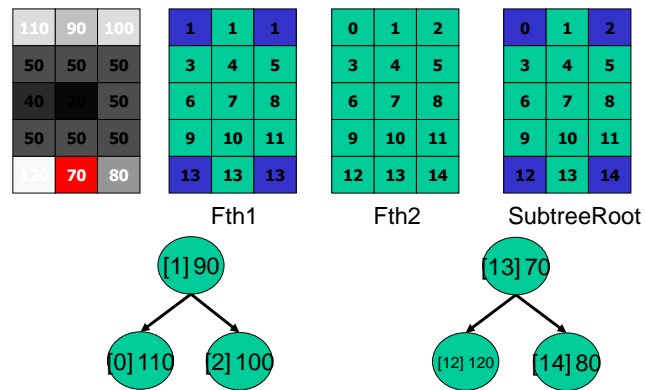
step 4



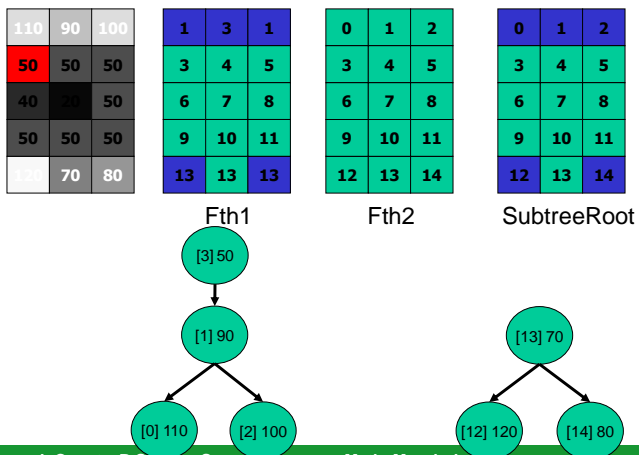
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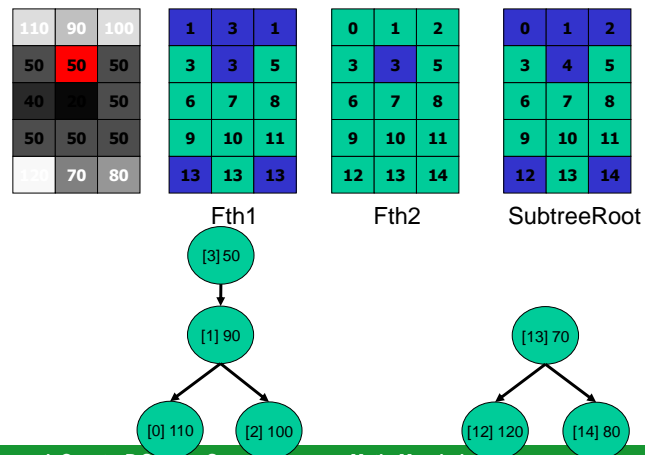
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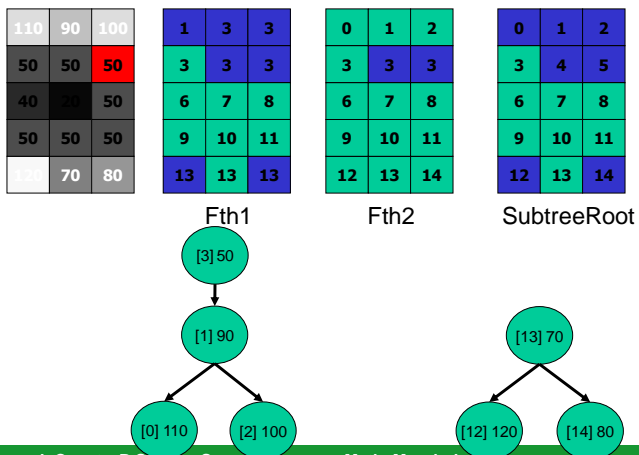
step 7



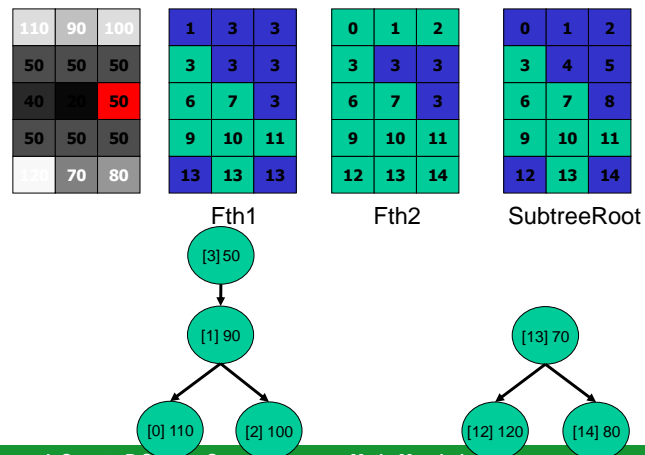
step 8



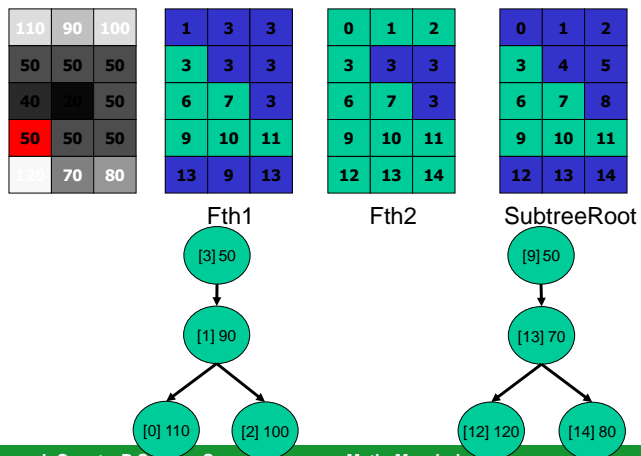
step 9



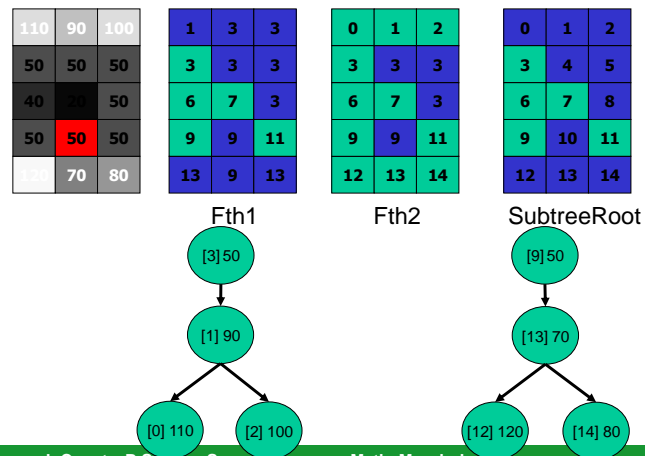
step 10



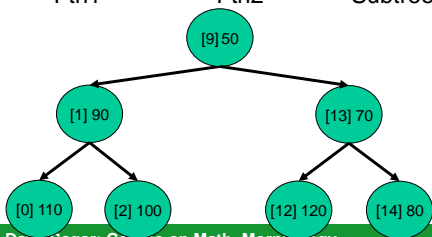
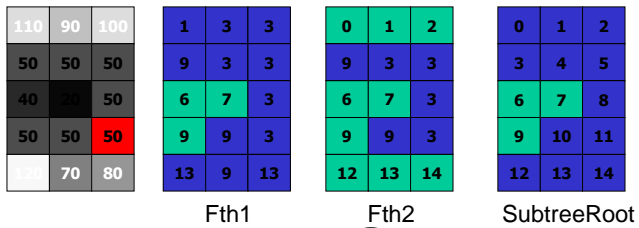
step 11



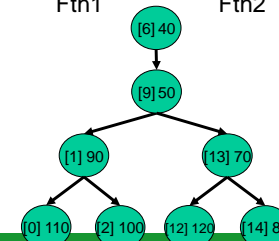
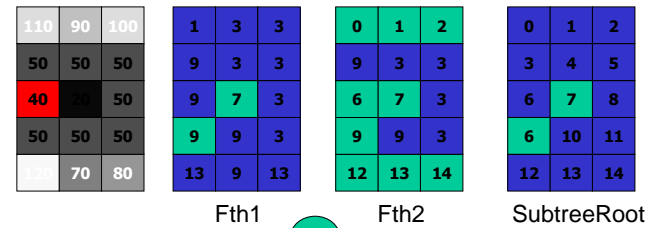
step 12



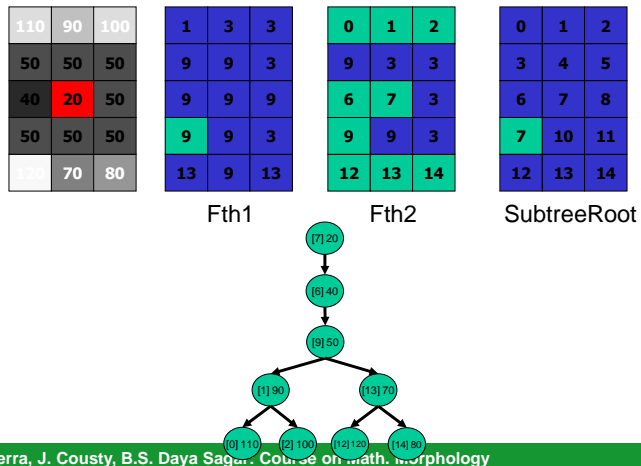
step 13



step 14



step 15



Component tree: conclusion

- Another insight into connected filters (level sets stacking)
- Useful to compute floodings
 - then attribute openings & closings
- Efficient algorithm
 - Quasi-linear time
 - Easy implementation
- Many applications in signal processing and image analysis
- Important for computing connection values
- There exist self-dual trees