



Technical University of Cluj - Napoca  
Computer Science Department

# **Image Processing**

**(Year III, 2-nd semester)**

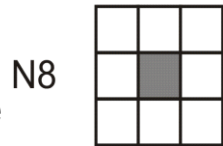
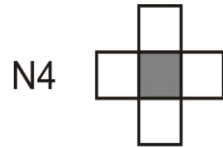
**Binary Images: Object Labeling. Contour Tracing (III)**



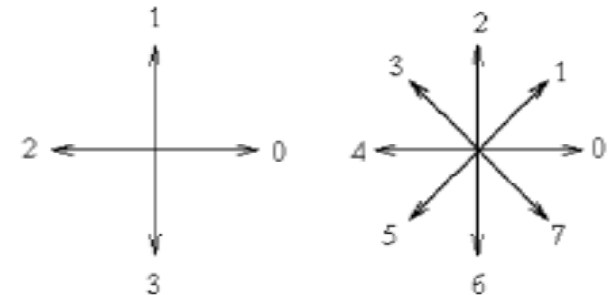
# Binary Algorithms - Definitions

## 1. Neighbors

- two pixels are **4-neighbors** or **d-neighbors** if they share a common side
- two pixels are **i-neighbors** if they share a common corner
- the **neighbour** concept includes both **d-neighbour** and **i-neighbour**
- two pixels are neighbors or **8-neighbors** if they share at least a corner
- the N-neighbour is the neighbour in direction N, where N is a direction code and  $0 \leq N \leq 3$  or  $0 \leq N \leq 7$



A pixel is said to be 4-connected to its 4-neighbors and 8-connected to its 8-neighbors.



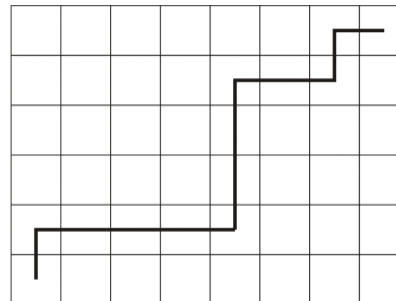
## 2. Path

Path  $(p[i_0, j_0] \Rightarrow p[i_n, j_n]) := \{[i_0, j_0], [i_1, j_1], \dots, [i_n, j_n] \mid [i_k, j_k] N_{4/8} [i_{k+1}, j_{k+1}] \forall k = 0 \dots n-1\}$

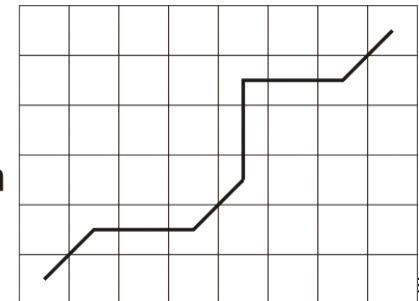
N4  $\Rightarrow$  4-path or d-path

N8  $\Rightarrow$  8-path or i-path

4-path



8-path





# Binary Algorithms - Definitions

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**3. Foreground**  $S := \{ p[i,j] \mid p[i,j] = 1 \}$

**4. Connectivity**  $p \leftrightarrow q$  (connected) if  $\exists \text{ Path } (p \Rightarrow q) \subset S, p \in S, q \in S$ .

**5. Connected points**  $\{p_i \in S, i = 1 \dots n \mid p_k \leftrightarrow p_j, \forall (p_k, p_j) \in S, k, j = 1 \dots n\}$

**6. Boundary**  $\text{Boundary}(S) := S' = \{ p \in S \mid \exists q \in N_{4/8}(p), q \in C(S) \}$   
 $C(S)$  – complement of  $S$

**7. Interior**  $\text{Interior}(S) = S - S'$

**8. Connected component**

Maximal set of connected points  $\{p_i \in S, i = 1 \dots n \mid p_k \leftrightarrow p_j, \forall (p_k, p_j) \in S, k, j = 1 \dots n\}$

**9. Background** set of all connected components belonging to  $C(S)$  that have points on the border of an image. All other components of the image belonging to  $C(S)$  are called holes.



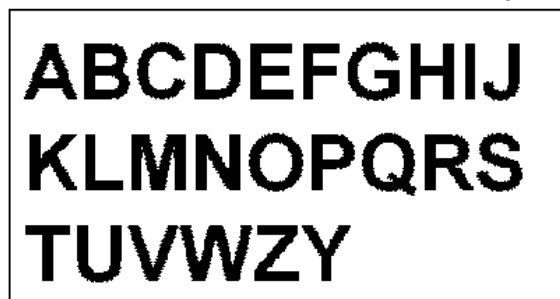
# Labeling Connected Components

**Connected component** : maximal set of connected points

$$\{p_i \in S, i = 1 \dots n \mid p_k \leftrightarrow p_j, \forall (p_k, p_j) \in S, k, j = 1 \dots n\}$$

One way to label the objects in a discrete binary image is to choose a point where  $b_{ij} = 1$  and assign a label to this point and its neighbors. Next, label all the neighbors of these neighbors, and so on.

- When this recursive procedure terminates, one component will have been labeled completely, and we can continue by choosing another start point.
- To find a new start point, we can simply scan through the image in a systematic way, starting a labeling operation whenever an unlabeled point is found where  $b_{ij} = 1$ .



ABCDEFGHIJ  
KLMNOPQRS  
TUVWZY

$\Rightarrow$   
Labeling



ABCDEFGHIJ  
KLMNOPQRS  
TUVWZY



# Sequential Labeling

## Iterative Algorithm (Haralick 1981)

- no auxiliary storage to produce the labeled image from the binary image.
- useful in environments whose storage is severely limited.

1. initialization step

2. repeat

    top-down & left-right label propagation

    bottom-up & right-left label propagation

until no changes occur

**procedure** Iterate;

    “Initialization of each 1-pixel to a unique label”

**for** L:=1 to NLINES **do**

**for** P:=1 to NCOLUMNS **do**

**if** I(L,P) =1

**then** LABEL(L,P):=NEWLABEL()

**else** LABEL(L,P):=0

**end for**

**end for;**



# Sequential Labeling

a	b	c
d	e	

“Top-down passes;

	e	d
c	b	a

Bottom up passes;

	e	

8-connected neighborhood”

“**procedure** Iterate – page 2”

“Iteration of top-down followed by bottom-up passes”

**repeat**

“Top-down passes”

CHANGE:=false;

**for** L:=1 to N\_LINES **do**

**for** P:=1 to N\_COLUMNS **do**

**if** LABEL(L,P) <> 0 **then**

**begin**

M:=MIN(LABELS(NEIGHBORS(L,P)U(L,P)));

**if** M <> LABEL(L,P)

**then** CHANGE:=true;

LABEL(L,P):=M

**end**

**end for**

**end for;**

IMAGE PROCESSING



# Sequential Labeling

“**procedure** Iterate – page 3”

“Bottom-up pass”

```
for L:= NLINES to 1 by -1 do  
    for P:= NCOLUMNS to 1 by -1 do  
        if LABEL(L,P)<>0 then  
            begin  
                M:=MIN(LABELS(NEIGHBORS(L,P)U(L,P)));  
                if M<> LABEL(L,P)  
                then CHANGE:=true;  
                LABEL(L,P):=M  
            end  
        end for  
    end for;  
  
    until CHANGE:=false  
  
end Iterate
```



# Sequential Labeling

## Example (N4)

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

1. Initial image

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

2. Initialization

	1	1		3	3	
	1	1		3	3	
	1	1	1	1	1	

3. Top-down & left-right  
label propagation

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

4. Bottom-up & right-left  
label propagation



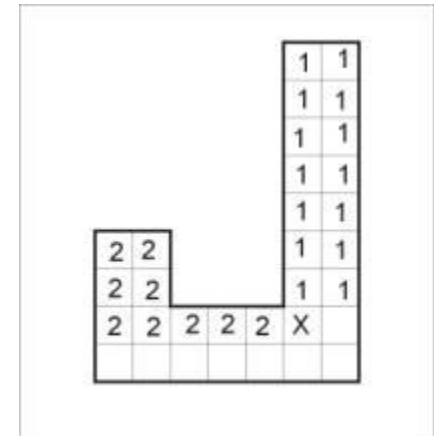


# Classical Algorithm

- Based on the classical connected components algorithm for graphs.
- 2 passes through the image but requires a large global table for recording equivalences.

## 1. First pass: performs label propagation, much as described above.

- Whenever a situation arises in which two different labels can propagate to the same pixel, the smaller label propagates and each such equivalence found is entered in an equivalence table (e.g. (1,2) → EqTable).
- Each entry in the equivalence table consists of an ordered pair, the values of its components being the labels found to be equivalent.
- After the first pass, the equivalence classes are found.
- Each equivalence class is assigned a unique label, usually the minimum (or oldest) label in the class.



## 2. A second pass through the image performs a translation, assigning to each pixel the label of the equivalence class of its 1-st pass label.



# Classical Algorithm

**procedure** Classical

“Initialize global equivalence table”

EQTABLE:=CREATE();

“Top-down pass 1”

**for** L:= 1 to NLINES **do**

“Initialize all labels on line L to zero”

**for** P:= 1 to NCOLUMNS **do**

LABEL(L,P):=0

**end for**

“Process the line”

**for** P:=1 to NCOLUMNS **do**

**if** I(L,P):= 1 **then**

**begin**

A:= NEIGHBORS((L,P));

**if** ISEMPY(A)

**then** M:=NEWLABEL()

**else** M:= MIN(LABELS(A));

LABEL(L,P):=M;

**for** X in LABELS(A) and X<>M

ADD(X, M, EQTABLE)

**end for**;

**end**

**end for**

**end for**;

IMAGE PROCESSING



# Classical Algorithm

“Find the equivalence classes”

EQCLASSES:=Resolve(EQTABLE);

“Find the equivalence label of an equivalence class”

**for** E in EQCLASSES

    EQLABEL(E):= min(LABELS(E))

**end for**;

“Top-down pass 2”

**for** L:= 1 to NLINES **do**

**for** P:= 1 to NCOLUMNS **do**

**if** I(L,P) = 1

**then** LABEL(L,P):=EQLABEL(CLASS(LABEL(L,P)))

**end for**

**end for**

**end** Classical

- **RESOLVE** - algorithm for finding the connected components of the graph structure, defined by the set of equivalences (**EQTABLE**) defined in pass 1.
- The main problem with the classical algorithm is the global equivalence table (large images with many regions, the equivalence table can become very large)



# Classical Algorithm

## Example (N4)

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. Initial image

1					2	2
		3	3			2
		3				2
		3				2
4	4	3				2
	4	3		5		2
	4	3	3	3		2
				3	3	2

2. Top down (pass 1)

### EQTABLE:

(4, 3), (3, 5), (3, 2) ...

### EQCLASSES:

1: {4, 3, 5, 2}

2: {6, 8, 9, ...}

....

n: {.....}

### EQLABEL:

1, 2

2, 6

...

n, x



# Classical Algorithm (improvement)

## A Space-Efficient Two-Pass Algorithm That Uses a Local Equivalence Table

⇒ use of a small local equivalence table that stores only the equivalences detected from the current and previous lines

Maximum number of equivalences = number of pixels / line.

1. First pass:

the equivalences from one line are used in the propagation step to the next line.

1. Second pass is required:

The new equivalence classes and labels finding followed by assigning the final labels.

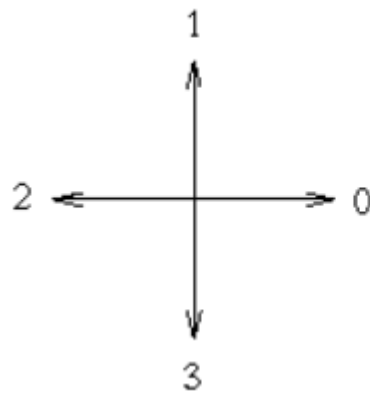


# Contour Tracing

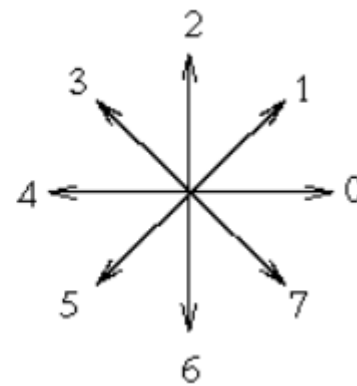
## Boundary/contour

$$\text{Contour}(R) = \{ p \in R \mid \exists q \in N_{4/8}(p), q \notin R \}$$

- The **contour** or **i-contour of R** (where R is a connected set of pixels) is defined as the set of all pixels in R which have at least one **d-neighbor** not in R.
- The **d-contour of R** is the set of all pixels in R, which have at least one neighbor not in R.
- *N-neighbor (chain-code / direction codes): c*  
(numerical operations on c are assumed to be modulo 4 or 8 )



4-neighbour



8-neighbour

1



# Contour Tracing

## Single contour tracing

- The algorithm can be described in terms of an observer who walks counterclockwise along pixels belonging to the set and selects the rightmost pixel available.
- The tracing terminates when the current pixel is the same as the initial pixel.
- The TRACER algorithm must be applied once for each hole of a region, in addition to one application for the external contour.
- Therefore, it must be combined with a search algorithm for locating holes in the interior of the region.
- Description of the contours (output):  $\{ A(x_0, y_0, c_0), C_i(x_i, y_i, c_i), i=1 \dots n \}$

## TRACER Procedure

Notations:

- A: the starting point of the contour of the set R (can be found in a number of ways, including a top-to-bottom, left-to-right scan);
- C: the current point whose neighborhood is examined;
- S : the search direction in the terms of the direction codes;
- first: is a flag that is true only when the tracing starts;
- found: is a flag that is true when a next point on the contour is found;

```
if ((A:=NEXT_START_ELEMENT())!=NULL)
{
  Q:=CREATE_LIST();
  TRACER(A,Q);
}
```



# Contour Tracing

**procedure** TRACER(A,Q)  
**begin**

first:=TRUE;

C:=A;

S:=6;

**while**((C!=A) or (first=TRUE))

{

found:=FALSE;

count:=0;

**while**((found=FALSE) and (count=<3))

{

**if** (B, the (S-1)-neighbor of C, is in R) **then**

{ APPEND((C,S-1),Q), C:=B, S:=S-2, found:=TRUE;}

**else if** (B, the S-neighbor of C, is in R) **then**

{ APPEND((C,S),Q), C:=B, found:=TRUE;}

**else if** (B, the (S+1)-neighbor of C, is in R) **then**

{APPEND((C,S+1),Q), C:=B, found:=TRUE;}

**else**

{S:=S+2, count:=count+1;}

**endif**

**endif**

**endif**

}

first:=FALSE;

}

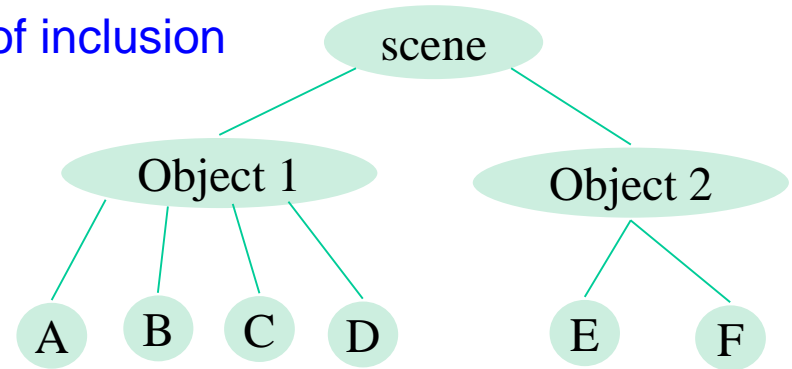
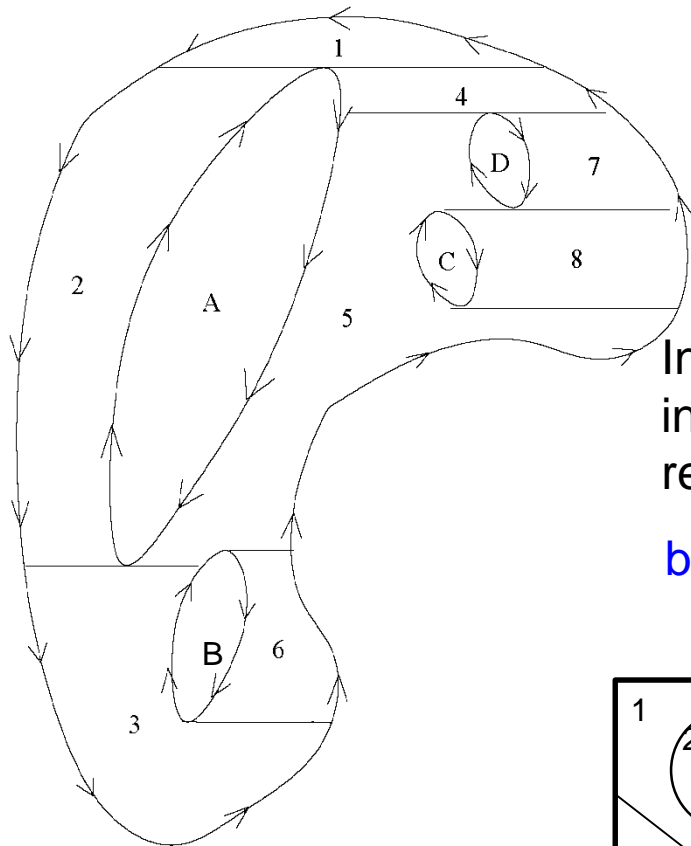
**end**





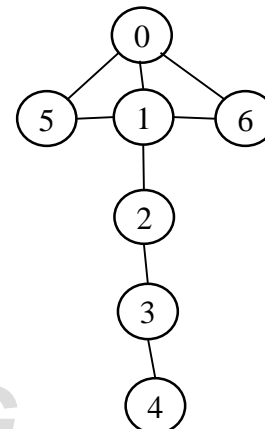
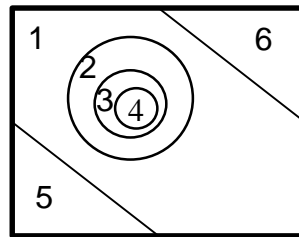
# Traversal of all the contours of an image

Structural description of the scene: a) Tree of inclusion



In this tree the nodes represent the connected regions in the image and the edges represent the inclusion relations. The root models the whole scene.

b) Adjacency graph



The nodes represent the connected regions in the image and the edges link adjacent regions.



# Traversal of all the contours of an image

---

## Traversal of All the Contours of a Region

⇒ closed i-paths and follows external contours counterclockwise (TRACER()) and contours of hole clockwise (TRACER1()).

Changes:

- When a pixel is marked as the current point, its value is incremented by 1. Then, at the end of the tracing, pixels of the contour will have values 2 or greater. These values are used when the interior is searched for holes.
- After the external contour has been found and placed in the queue Q, we start examining the contents of the latter. If we find a point located on a downward arc, we start a search to the right. Such pixels can be characterized easily by the requirement that the previous element of the chain code must have values 4 to 7 while the next element should be in the range 5 to 7.
- While scanning along the horizontal direction one must search for either the start of a hole or the other side of the outside contour.
- The TRACER1() procedure is called when an unmarked start of a hole point is found.
- The extracted contours of the holes are inserted into the Q list.
- The content of the list is examined until its end.



# Traversal of all the contours of an image

---

## MULTI\_TRACE Procedure

Notations:

P: current pixel with x, y coordinate;

c: direction code; value 8 is used to specify the beginning of a new contour;

c0: initial direction code;

Q: {(P, c)}

```
If ((A:=NEXT_START_ELEMENT())!=NULL)
{
    Q:=CREATE_LIST();
    MULTI_TRACE(A,Q);
}
```



# Traversal of all the contours of an image

procedure MULTI\_TRACE(A,Q)  
begin

TRACER(A,Q);

while (Q !=NULL)

{

(P,c):=Remove(Q);

if (c=8)

{

c0:=c;

(P,c):=Remove(Q);

}

if ((c0 ∈ {4 : 7}) and (c ∈ {5 : 7}))

{

Starting from P search in the x-direction and examine triplets of successive pixels, A, B, C.

if ((A!=0) and (B=1) and (C=0))

{

TRACER1(B,Q);

goto LABEL1 ;

}

If ((A=1) and (B=2) and (C=0))

goto LABEL1;

}

LABEL1: c0=c;

}

end

IMAGE PROCESSING



# Polygonal Approximation

## Polygonal approximation of contours

Curve  $C: f(x,y)=0 \Rightarrow$  polygon that closely approximates  $C$  with an error smaller than  $\varepsilon$  and having a number of vertices as small as possible:



- Any polygonal fitting algorithm requires that the data points be subdivided into groups, each one of them to be approximated by a side of the polygon.
- The first simplification of the polygon fit problem is to draw a line between the endpoints of each group rather than search for the optimal solution.
- If the approximation error is too big the group could be split in two and so on until the error becomes acceptable.
- Let  $Q$  a contour consisting of  $P_i (x_i, y_i)$  where  $i=1, 2, \dots, n$ , and  $\varepsilon$  the error threshold.



# Polygonal Approximation

## Procedure POLIGONAL\_APROX(Q)

### begin

A:=Create\_List();

B:=Create\_List();

i=Index\_of\_first\_point(Q);

j=Index\_of\_the\_most\_far\_point(Q);

Insert(j,A); Insert(j,B);

Insert(i,A);

while((A!=NULL)

{

Let  $k$  and  $l$  the indexes of the last elements of the lists  $A$  and  $B$ ;

Let  $P_k P_l$  the segment generated by these two points;

Let  $m$  the index of the most far point to  $P_k P_l$  segment among the contour points starting with  $P_k$  and ending with  $P_l$ , when the contour is scanned in counterclockwise direction.

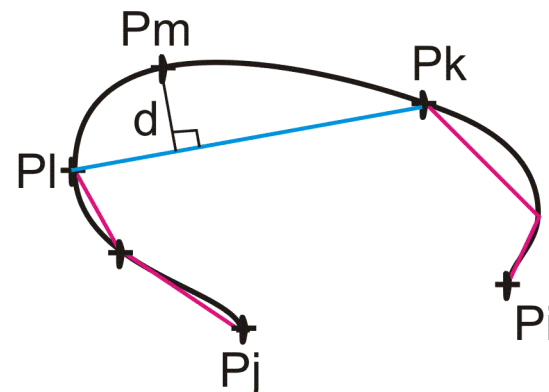
if ((d=Distance( $P_k P_l$ ,  $P_m$ ) >  $\varepsilon$ )

then Insert( $m$ , A)

else { Delete( $k$ , A)  
Insert( $k$ , B); }

}

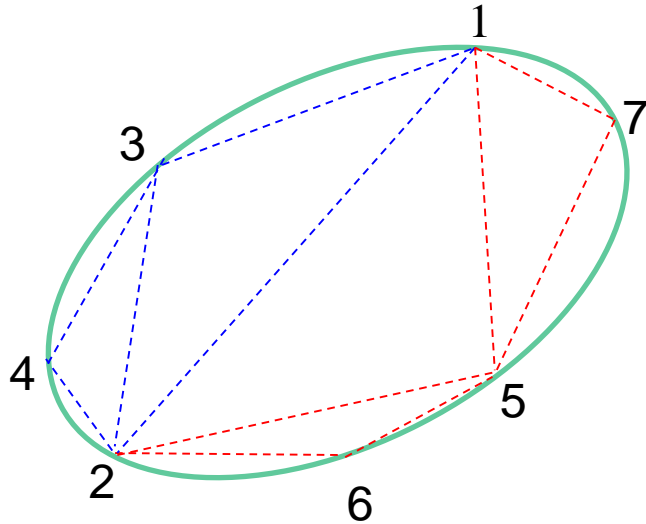
end



$$\text{distance}(P_1, P_2, (x_0, y_0)) = \frac{|(y_2 - y_1)x_0 - (x_2 - x_1)y_0 + x_2y_1 - y_2x_1|}{\sqrt{(y_2 - y_1)^2 + (x_2 - x_1)^2}}.$$



# Polygonal Approximation – Example



A	B
2	2
1	
3	
4	
2	2
1	4
3	
2	2
1	4
	3
2	2
	4
	3
	1
2	2
5	4
	3
	1

A	B
2	2
5	4
7	3
	1
2	2
5	4
	3
	1
	7
2	2
	4
	3
	1
	7
	5
2	2
6	4
	3
	1
	7
	5

A	B
2	2
	4
	3
	1
	7
	5
	6
	2
	4
	3
	1
	7
	5
	6
	2