

How to rebuild a binary image from its multilevel description based on generalized salient pixels

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#### Outline





#### Line convexity

A binary image F is **row-convex** if for every pair of pixels A and B in the same row of F, all the pixels from [AB] are in F

A binary image F is **column-convex** if for every pair of pixels A and B in the same column of F, all the pixels from [AB] are in F

A binary image F is **hv-convex** if it is both **row and columnconvex** 







#### Q-convex binary image

A binary image F is **quadrant-convex** if for every quadruple of pixels A,B,C,D, each in a different quadrant, all the (yellow) pixels "inbetween" are in F



#### How does it look like?



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### Salient pixels: S(F)



x of F such that there is a background quadrant in x for  $F \{x\}$ Z  $\cap F \{x\} = \phi$ 

### Q-convex hull : Q(S(F))

Fill F such that it is Q-convex

F is Q-convex iff F=Q(F)=Q(S(F))

# What if not Q-convex: more "complex" binary images





# What if not Q-convex: more "complex" binary images







### Generalized salient pixels (GSp)

- If F is Q-convex, then Q(F)=F
- If F is not Q-convex, then look at Q(F)\F !
- Compute the salient pixels of Q(F)\F
- If not empty, then iterate!



#### The algorithm to compute the GS matrix



k=3

k = 1

0 0 1 0 2 0 1 0 0 1 2 3 3 2 0 0 1 0 0 4 4 3 0 1 1 2 3 4 5 4 0 2 0 0 0 0 0 0 3 2 0 1 0 0 0 2 0 1 0 0 2 3 3 0 1 0 0 0 1 2 2 1 0 0

k=2

GS-matrix associated to the binary image

k = 5

It provides a k-level description of the image

k = 4



Every binary image can be decomposed in the k-multi level representation k is greater for more 'complex' binary images

k is maximum for the chessboard (all its pixels are gsp)

#### Binary image and its GS matrix





k = 1

0	0	1	1	U	1	1	0
0	1	U	1	1	0	1	0
1	1	0	0	0	1	1	1
1	0	1	0	1	0	υ	U
0	1	0	1	1	1	1	0
0	1	0	1	1	0	1	1
0	0	0	1	1	1	1	0
0	0	1	0	0	1	0	0

k=2

0	0	1	1	0	1	1	0	
0	1	0	1	1	0	1	0	
1	1	0	0	0	1	1	1	
1	0	1	0	1	0	0	0	
0	1	0	1	1	1	1	0	
0	1	0	1	1	U	1	1	
0	0	0	1	1	1	1	0	
0	0	1	0	0	1	0	0	

=3

0	0			0			U
0	1	0	1	1	0	1	0
1	1	0	U	U	1	1	1
1	0	1	0	1	U	0	0
0	1	0	1	1	1	1	0
0	1	0	1	1	0	1	1
0	0	0	1	1	1	1	0
0	0	1	0	0	1	0	0

0 0 1 1 0 1 1 0

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

k=5

0 0 1 1 0 1 1 0



Every binal y image can be decomposed in the k-multi level representation
The image at each level is O-convex
By projection, images are ordered by inclusion

# The algorithm to rebuild the binary image from its GS-matrix

```
1: procedure (B)
        for each b_{ij} = 0 do
2:
3:
           for each t=0,1,2,3 do
4:
               find b the maximum item of B in Z_t(b_{ij})
               store b in z_{ii}^t of matrix Z^t
5:
           end for
6:
        end for
7:
        for each b_{ii} = 0 do
8:
           h \leftarrow \min(z_{ij}^0, z_{ij}^1, z_{ij}^2, z_{ij}^3)
if h is odd then f_{ij} \leftarrow 1
9:
10:
             else f_{ij} \leftarrow 0
11:
12:
             end if
13:
         end for
       for each b_{ii} \neq 0 do
14:
            if b_{ij} is odd then f_{ij} \leftarrow 1
15:
16:
            else f_{ii} \leftarrow 0
17:
             end if
18:
         end for
19: end procedure
```

## How does it change the image by modifying the GS-matrix?





### Erosion and delation operations



#### Erosion-shrink the image.

The erosion operator takes two pieces of data as inputs: the first is the image which is to be eroded; the second is a structuring element (also known as a kernel). The kernel determines the precise effect of the erosion on the input image.



#### Dilation- grow the image.

It is the dual of erosion.

Repeat



- Remove pixels of the binary image corresponding to items 1 in the GS
- Compute the GS of the obtained binary image

Repeat



- Remove pixels of the binary image corresponding to items 1 in the GS
- Compute the GS of the obtained binary image

Repeat



- Remove pixels of the binary image corresponding to items 1 in the GS
- Compute the GS of the obtained binary image

### 2223

Repeat



- Remove pixels of the binary image corresponding to items 1 in the GS
- Compute the GS of the obtained binary image

 $\sim$ 

#### What about dilation?

 There is not a unique way to add 1's in the GS matrix



### Medial axis transform (MAT)



Similarly, a grey level image
 such that the intensity of its
 pixels representes their 'distance'
 to the salient pixels, iteratively by erosion

The MAT is a grey level image where each pixel on the skeleton has an intensity which represents its distance to a boundary in the original object.



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#### **Conclusions and perspectives**



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