

1. MOTIVATIONS

Arcs and segments are the most appearing primitives in images

- Detection of shapes
 - medical imaging
 - technical images
 - manual drawings
- Character recognition
 - manual sketches
 - scanned documents



Fig 1: Arc and segment detection of an image

This decomposition allows a simple descriptor of object contours and facilitates their manipulation in image processing applications.

2. CONCEPTS & METHOD

The proposed method consists in decomposing the contours extracted from images, called *discrete curves* into arcs and segments. In particular, the decomposing algorithm aggregates mainly three following concepts:

1. *Adaptive tangential cover* (ATC) [1]: a discrete geometry tool allowing to study the geometrical characteristics of a curve
2. *Dominant points* (DP) [2]: the characteristic points of a curve allowing an approximation of the curve by a polygon
3. *Tangent space* (TS) [3]: a representation of the polygon issued by the dominant points of the curve allowing to detect the points on arcs and segments of the polygon

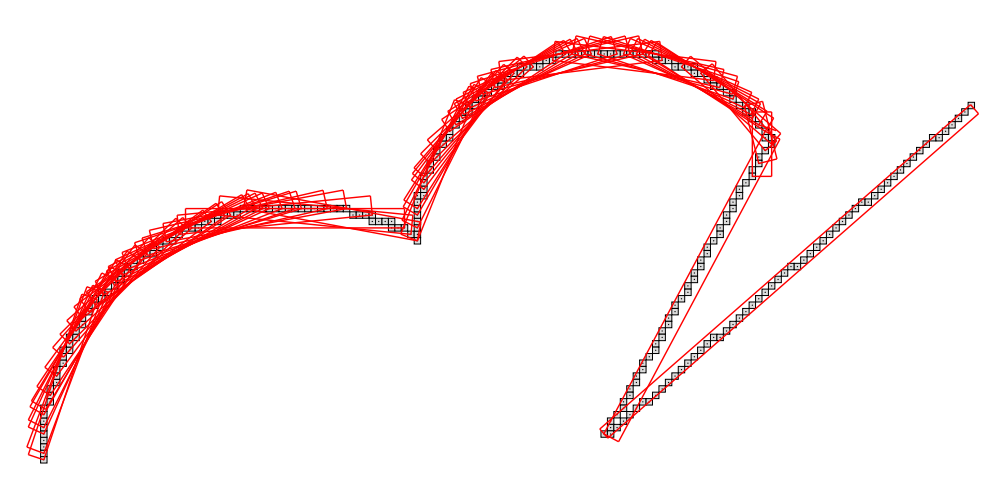


Fig 2: ATC

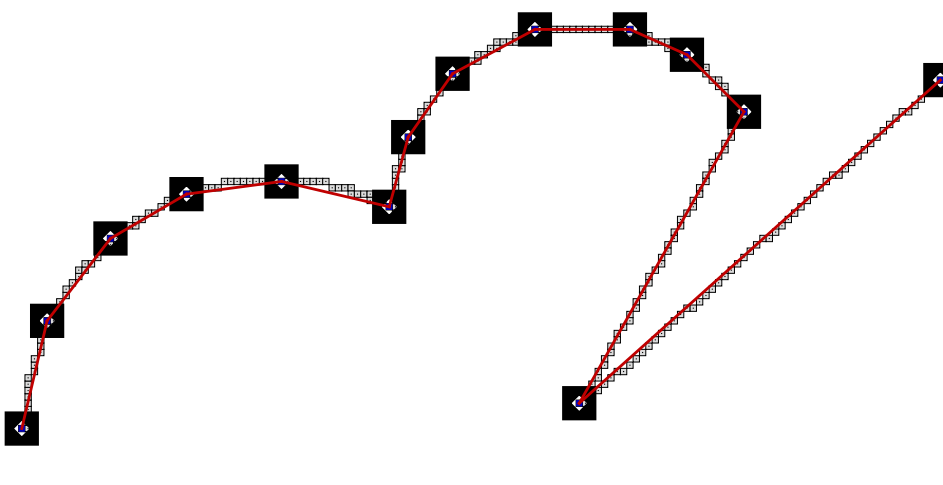


Fig 3: DP

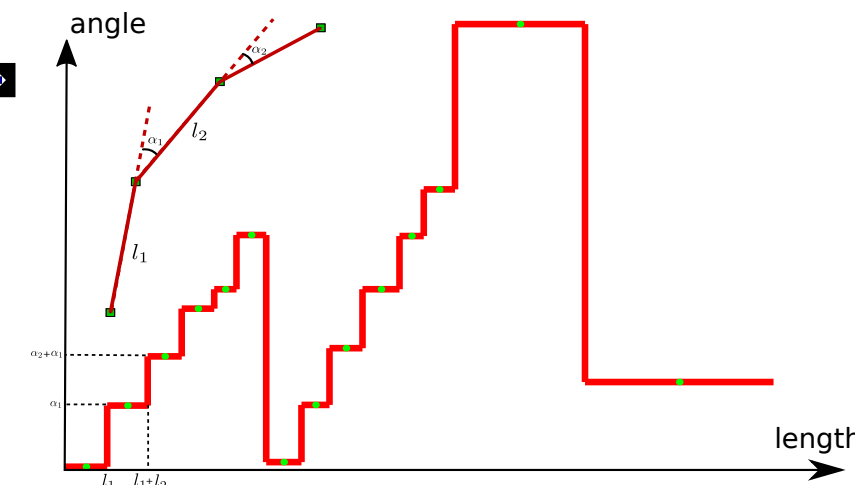


Fig 4: TS

REFERENCES

- [1] P. Ngo, H. Nasser, I. Rennesson-Debled, and B. Kerautret. Adaptive tangential cover for noisy digital contours. In *Proceeding of DGCI*, volume LNCS 9647, 2016.
- [2] P. Ngo, H. Nasser, and I. Rennesson-Debled. Efficient dominant point detection based on discrete curve structure. In *Proceeding of IWCIA*, volume LNCS 9448, pages 143–156, 2015.
- [3] T. P. Nguyen and I. Rennesson-Debled. Arc segmentation in linear time. In *Proceeding of CAIP*, volume LNCS 6854, pages 84–92, 2011.

3. PROPOSED ALGORITHM

Algorithm 1: Curve Decomposition into Arcs and Segments

```

Input:  $C = (C_i)_{0 \leq i \leq n-1}$  a discrete curve of  $n$  points
      thickness width of MBS for collinear test in the tangent space
       $\alpha_{Max}$  threshold of admissible angle in the tangent space
       $nbCirclePoint$  threshold of midpoints belonging to an arc
       $isseTol$  threshold of approximation by an arc or segments
Output:  $ARCS$  and  $SEGs$  sets of arcs and segments of  $C$ 
1  $ARCS \leftarrow \emptyset, SEGs \leftarrow \emptyset, pARC \leftarrow \emptyset, MBS_v \leftarrow \emptyset$ 
2 Detect the dominant point  $D$  of  $C$ 
3 Transform  $D$  into the tangent space  $T(D)$ 
4 Construct the midpoint curve  $\{M_i\}_{i=0}^{m-1}$  of  $T(D)$ 
5 for  $i \leftarrow 1$  to  $m-2$  do
6    $C_b, C_e$  is the part of  $C$  corresponding to  $M_i$ 
7   if  $(|M_i.y - M_{i-1}.y| > \alpha_{Max}) \& (|M_i.y - M_{i+1}.y| > \alpha_{Max})$  then
8      $SEGs \leftarrow SEGs \cup \{C_b, C_e\}$ 
9      $MBS_v \leftarrow \emptyset$ 
10  else
11    if  $MBS_v \cup \{M_i\}$  is a MBS of thickness then
12       $MBS_v \leftarrow MBS_v \cup \{M_i\}$ 
13       $pARC \leftarrow pARC \cup \{C_b, C_e\}$ 
14    else
15      if  $|MBS_v| \geq nbCirclePoint$  then
16         $C_b, C_e$  is the part of  $C$  corresponding to  $pARC$ 
17         $Arc(O, R, \beta_b, \beta_e)$  is the best fitting arc for  $C_b, C_e$ 
18        if  $ISSE(C_b, C_e, Arc(O, R, \beta_b, \beta_e)) <$ 
19           $isseTol \sum_{C_i \in C_b, C_i \in pARC} ISSE(C_b, C_e, \overline{C_b, C_e})$  then
20           $ARCS \leftarrow ARCS \cup \{Arc(O, R, \beta_b, \beta_e)\}$ 
21        else
22           $SEGs \leftarrow SEGs \cup \{C_b, C_e \mid C_b, C_e \in pARC\}$ 
23      else
24         $SEGs \leftarrow SEGs \cup \{C_b, C_e \mid C_b, C_e \in pARC\}$ 
25       $pARC \leftarrow \emptyset$ 

```

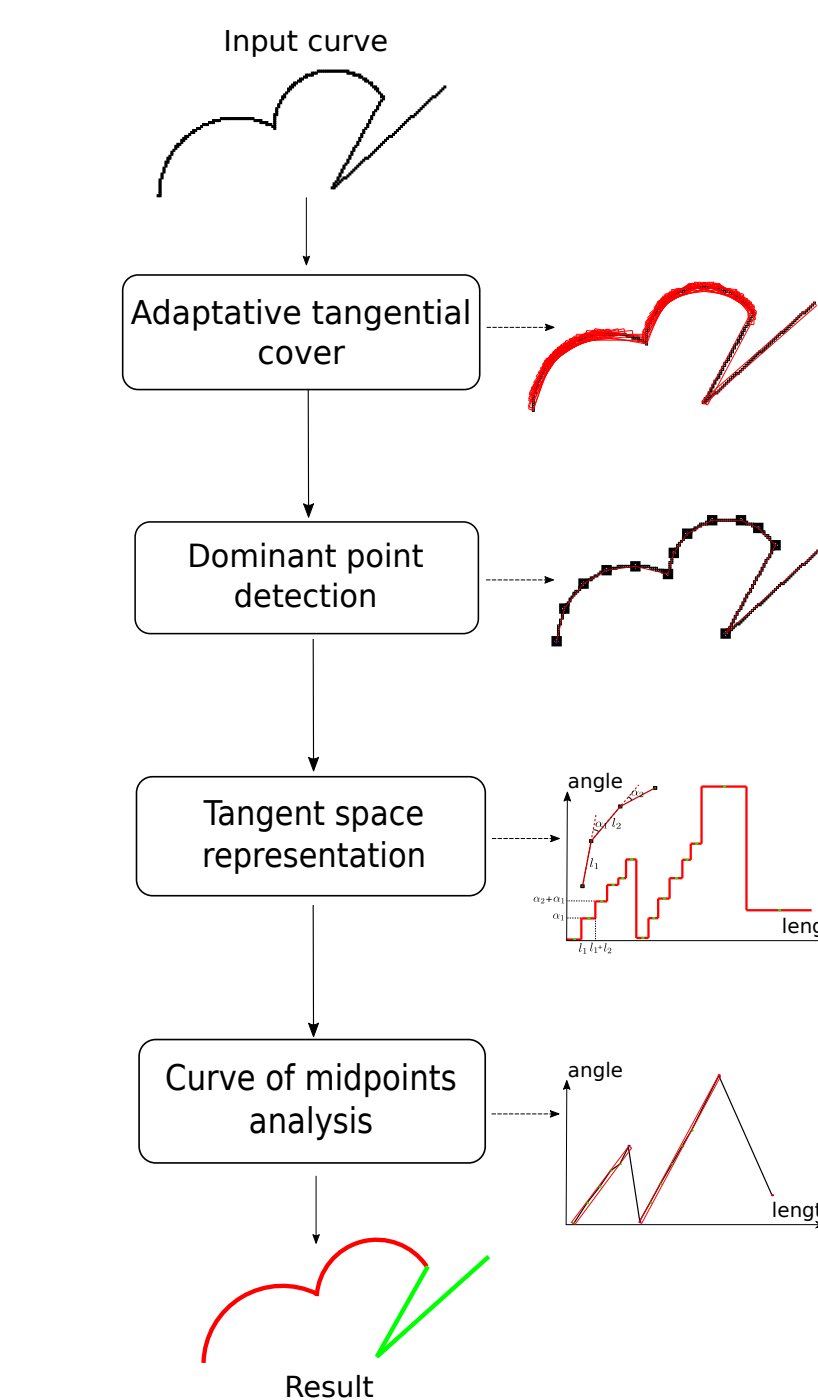


Fig 5: Algorithm's flowchart

Parameter (default value)	Signification
$samplingStep$ (1)	Sampling step of noise detection
$maxScale$ (15)	Maximal thickness of noise detection
$thickness$ (0.2)	Width for the quasi-collinear test
α_{Max} (0.78)	Admissible angle in the tangent space
$nbCirclePoint$ (3)	Threshold of midpoints belonging to an arc
$isseTol$ (4)	Threshold of approximation by an arc or segments

4. SOURCE CODE

Download and installation:

<https://github.com/ngophuc/CurveDecomposition>



Online demonstration:

http://ipol-geometry.loria.fr/~phuc/ipol_demo/RRPR_demo

Input: A file in sdp format containing several contours

```

x0 y0 x1 y1 ... xn yn # Points of contour 1
xn+1 yn+1 xn+2 yn+2 ... xm ym # Points of contour 2

```

Output: Several files in svg (or eps) format

```

__OutPts.svg      File of input points
__ATC.svg        Result of ATC computation
__DP.svg         Result of dominant point simplification
__OnlyArcSeg.svg Arc and segment decomposition result

```

Code: Execution is named *testContourDecom*, to run the program on *contour.sdp* with *samplingStep=1.0*, *maxScale=10*, *alphaMax=0.78*, *thickness=0.2*, *nbPointCircle=3* and *isseTol=4.0*

```

./testContourDecom -i contour.sdp -d IMAGENEDIRECTORY
--samplingStep 1.0 --maxScale 10 -a 0.78 -t 0.2 -n 3 -s 4.0

```

4. EXPERIMENTAL RESULTS

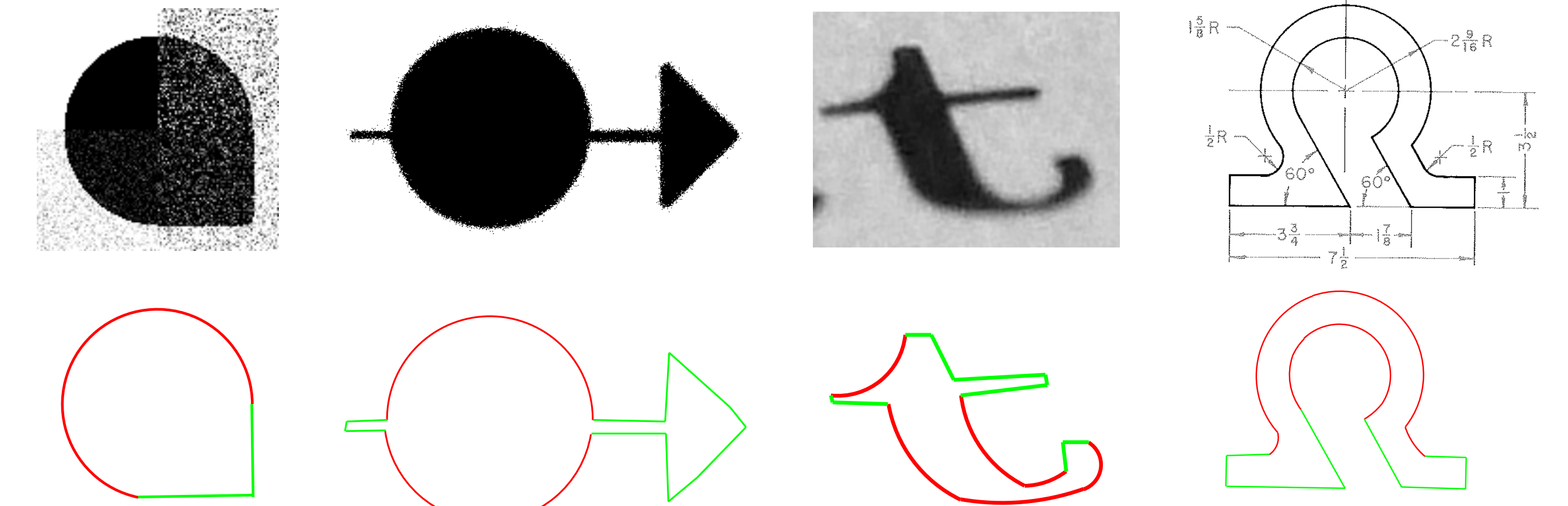
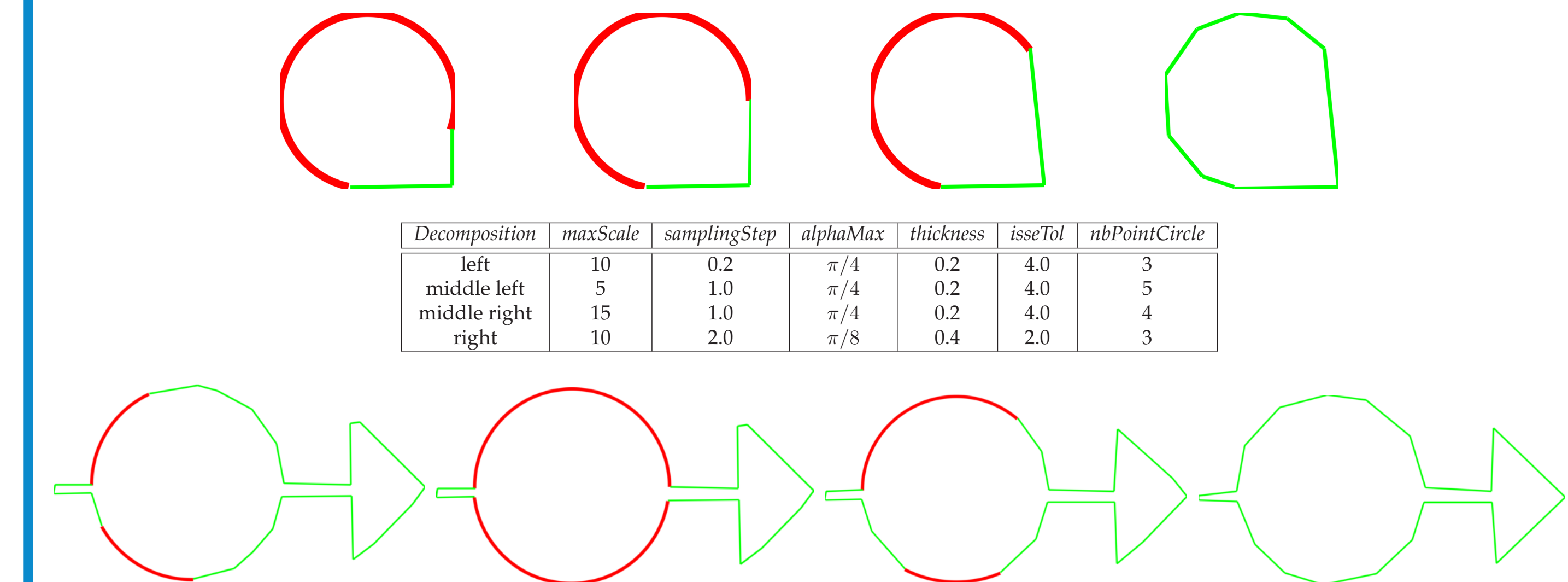


Fig 6: Experiments on different types of noise and images



Decomposition	maxScale	samplingStep	alphaMax	thickness	isseTol	nbPointCircle
left	10	0.2	$\pi/4$	0.2	4.0	3
middle left	5	1.0	$\pi/4$	0.2	4.0	5
middle right	15	1.0	$\pi/4$	0.2	4.0	4
right	10	2.0	$\pi/8$	0.4	2.0	3

Fig 7: Experiments on the parameter sensibility

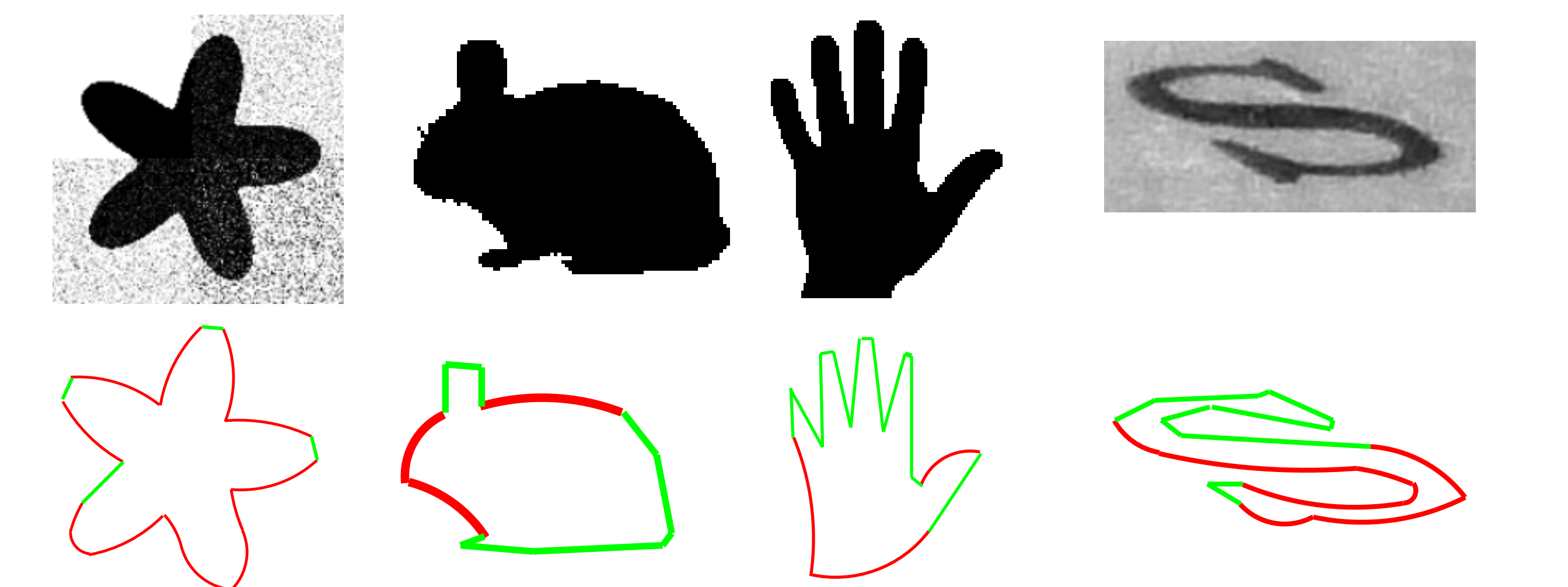


Fig 8: Borderline cases using default parameters

5. CONCLUSION & PERSPECTIVES

- Algorithm to decompose a noisy curve into arcs and segments
- A better descriptor of the curve comparing to dominant points
- Extension to other types of primitive: ellipse, parabola, ect.
- Extension to 3D curves
- Integration of topological information into the decomposition
- Application to image processing tasks: shape reconstruction, character recognition, ect.