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APPLICATION OF MATHEMATICAL MORPHOLOGY ON NANOSTRUCTURE IMAGE PROCESSING

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Abstract: Mathematical morphology is a very effective tool on image segmentation. This paper presents selected methods of mathematical morphology that are applied to the microscopic images of nanomaterials. The first part briefly describes the mathematical background of the fundamental morphological methods, including their application on the test image. The second part contains an application both of binary and grayscale morphological methods on the nanostructure images. The influence of various structure elements is investigated, as well.

Keywords: binary morphology, nanostructure image processing, grayscale morphology

1. INTRODUCTION

Image processing is applied as an effective tool for various image analysis in many branches of science, Kuehn (1998), Wu (1995), Sorzanoa (2004). Presented paper is focused on analysis of nanomaterial images.

These nanomaterial seems to be one of the options of surface bioactivation of titanium implants that are used in medicine, Fojt (2010). Nanomaterial properties assessment is one of the essential issues in the process of material development while the tube diameter should be evaluated as well as a side thickness and other quantities. The material properties quantities could be based on the processing of microscopic images, Hodneland (2006), Bunes (2008), Kiang (1998), Gommes (2003). A lot of various methods are used for segmentation and analysis of given microscopy images, such as a low-pass filtering, application of watershed transform, adaptive thresholding, mathematical morphology Dougherty (2003), Heijmans (1998), Horgan (1998), convex hull and others.

Methods of mathematical morphology could be used for processing both of binary and grayscale images, Gonzalez (2002), Hlavac (2002), Hanbury (2001). This paper looks out only a part of the complex algorithm of given data processing that concerns of an application of binary mathematical morphology methods.

2. MATHEMATICAL BACKGROUND

Mathematical morphology provides a tool for extraction of those image components that are useful in the representation and description of a region shape. The most common use of mathematical morphology is in image enhancement, segmentation and restoration, edge detection, texture and component analysis, curve filling, noise reduction etc. A grayscale digital image generally can be represented as a set of elements, values of which are vectors of the 3-D integer space $Z^3$. The first two components represent coordinates $(x, y)$ of the pixel and the third one its discrete graylevel value. A binary image $A$ can be considered as a set of $n$ white pixels $A = \{(x_1, y_1), (x_2, y_2), \ldots, (x_n, y_n)\}$ where vector $\{x_i, y_i\} \in Z^2$ corresponds to pixel coordinates.

Morphological operators are given by the relationships between a set of points of an original image $A$ and another point set, which is called a structural element $B$. Various types of structural elements can be used. Examples of the most common structural elements are presented in Fig. 1.

Dilation, erosion, opening and closing are the basic morphological transformations.

2.1 Binary Morphology

Morphological operations can be defined with Minkowski operators, other description can be used, as well, Gonzalez (2002), Farouki (2001).
2.2 Grayscale Morphology

Basic binary morphological operation can be extended to grayscale images, Gonzalez (2002), Hanbury (2001). Operations deal with the input image function \( f(x, y) \) and a structural element \( b(x, y) \).

**Grayscale dilation** of \( f \) by \( b \), \( f \oplus b \), is defined as

\[
(f \oplus b)(s, t) = \max\{f(s-x, t-y) + b(x, y) \mid (s-x), (t-y) \in D_f; (x, y) \in D_b\}
\]

where \( D_f \) and \( D_b \) are the domains of \( f \) and \( b \).

After application of dilation on a grayscale image, output image is brighter than an input one if all the values of the structural element are positive. Dark details are reduced or eliminated.

**Grayscale erosion** is based on the minimum value of \((f-b)\) choosing in the interval defined by structural element \( b \). The Erosion of \( f \) by \( b \), \( f \ominus b \), is defined as

\[
(f \ominus b)(s, t) = \min\{f(s+x, t+y) - b(x, y) \mid (s+x), (t+y) \in D_f; (x, y) \in D_b\}
\]

where \( D_f \) and \( D_b \) are the domains of \( f \) and \( b \).

Output image after grayscale erosion application is darker than input one if all elements of the structural element are positive. The bright details that are smaller than the structural element are reduced.

**Grayscale opening and closing** have the same form as their binary alternates. The opening of grayscale image \( f \) by the structural element \( b \), \( f \ominus b \), is defined as

\[
f \ominus b = (f \ominus b) \ominus b
\]

The grayscale closing of \( f \) by \( b \), \( f \oplus b \), is defined as

\[
f \oplus b = (f \oplus b) \ominus b
\]

**Top-hat transformation** \( h \) of an image \( f \) is defined as

\[
h = f - (f \ominus b)
\]

where \( f \) is an original image and \( b \) is a structural element. This transformation is useful for enhancing detail in background.

**Morphological gradient**, denoted \( g \), highlight sharp gray-level transitions in the input image. It is defined as

\[
g = (f \oplus b) - (f \ominus b)
\]

3. NANOSTRUCTURE IMAGE PROCESSING

Nanostructure image processing features a very important part of classification process of nanomaterials quality. There is presented a selected cut of image of titanium dioxide tubes, Fojt (2010), in Fig. 4a. One of the important factor representing the material quality is an inner tube diameter. Methods of image processing seems to be a suitable tool to obtain the desired value. The essential step of the nanomaterial image processing is a separation of the inner area of tubes from the background. Preprocessing of these images based on digital filters application is
necessary, as well. The filter size and cutoff frequency of lowpass filter used depend on image character. The filtration reduces a level of noise in the image and it improves results of segmentation in the following step, Fig. 4b.

In the next step the image is divided into segments by means of watershed transform Eq. (10), Fig. 4c. After that, an adaptive thresholding is performed inside of each segment. Adaptive thresholding is used because of very different grayscale level in each part of the image. Setting of one thresholding value for whole the image is unsuitable in this case. Although the adaptive thresholding application provides satisfying results, there are still undesired artifacts in the resulting image, Fig. 4d. Some of them imply from the inner area selection only, without segments border. Application of mathematical morphology seems

Figure 4. Image preprocessing before morphology use: (a) An original image, (b) Lowpass filter application, (c) Watershed segmentation, (d) Adaptive thresholding result

Figure 5. Selected results of closing operation by means of various types of structural element in zoomed cut of an image: (a) Part of thresholded image, (b) Closing by disk size 1, (c) Closing by disk size 3, (d) Closing by square size [2,2], (e) Closing by square size [4,4], (f) Closing by diamond size 3

Figure 6. Closing of thresholded image cut by various structural element: (a) Thresholded image, (b) Closing by disk 1, (c) Closing by structural element type disk size 1, (c) Closing by structural element type square size [2,2], (d) Closing by structural element type square size [4,4]

Figure 7. Convex area (red) and non-convex area (blue) detected by various SE: (a) without morphology, (b) SE disk size 1, (c) SE square size [2,2], (d) SE square size [4,4]
to be an effectual tool for elimination of these undesired effects, Fig. 6. Closing operation can remove small holes inside the objects and it smooths out the objects contours. The result of this operation depends on the shape of structural element used, of course. Various structural elements $B$ were tested on the real image set. A selected zoom cut processed by mean of various structural element is presented in Fig 5. Only the symmetric structural elements have been considered, as it is possible to suppose an isotropic material properties.

As the tubes are supposed to have a convex cutout, non-convex areas are excluded from the next processing and the area of convex segment is considered only, Fig. 7.

Results obtained with the application of various SE are presented in Tab. 1. Sample of selected structural element is $Disk_1=\{(−1,0),(0,−1),(0,0),(1,0),(0,1)\}$ or $Square[2,2]=\{(0,0),(0,1),(1,0),(1,1)\}$.

The application of mathematical morphology methods implicates reduction of a number of objects found in the image, see Tab. 1. Results obtained by application of various structural elements are comparable. It’s not possible to appoint the best structural element explicitly. The elements in form of disk and square seems to be useful for this type of images.

Table 1. Results obtained with various structural elements use

<table>
<thead>
<tr>
<th>Structural Element</th>
<th>The Number of Tubes</th>
<th>Tube Area Percentage of the Image</th>
<th>Average Tube Area (px)</th>
<th>Standard Deviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>--</td>
<td>248</td>
<td>32.21</td>
<td>85.11</td>
<td>147.33</td>
</tr>
<tr>
<td>disk 1</td>
<td>100</td>
<td>27.95</td>
<td>183.16</td>
<td>254.08</td>
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<tr>
<td>disk 3</td>
<td>50</td>
<td>21.27</td>
<td>278.84</td>
<td>304.34</td>
</tr>
<tr>
<td>square [2,2]</td>
<td>83</td>
<td>27.49</td>
<td>217.06</td>
<td>262.94</td>
</tr>
<tr>
<td>square [4,4]</td>
<td>58</td>
<td>25.32</td>
<td>288.34</td>
<td>294.39</td>
</tr>
<tr>
<td>diamond 3</td>
<td>61</td>
<td>21.51</td>
<td>231.13</td>
<td>293.58</td>
</tr>
</tbody>
</table>

4. CONCLUSION

Nanostructure image processing is a very important part of classification process of nanomaterials quality. Application of mathematical morphology methods can improve object detection in these images. The best structural element for this detection cannot be exactly defined, as the image resolution seems to be a basic factor for structural element size and shape selection, too. Obtained results depend on an original image quality, as well. Based on testing, structural element in the form disk and square seems to be suitable for processing of the given images.

Used images originate from electron microscope TESCAN VEGA 3 SBU. The algorithm was developed in MATLAB environment, ver. 2010a, and it was tested on computer with following parameters: processor Intel(R) Core(TM) 2CPU, 2.40 GHz, RAM 4GB, OS Win7 32bit.

Application of various type of a structural element has similar demands concerning a runtime. There is a comparison of runtime values for selected structural elements use, Tab. 2.

Table 2. Runtime obtained with an application of various structural elements

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<td>0,118</td>
<td>0,234</td>
<td>0,031</td>
<td>0,027</td>
<td>0,006</td>
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</tr>
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</table>

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REFERENCES


