STRUCTURAL METHODS IN SUBCELLULAR IMAGE ANALYSIS

CENTRE FOR IMAGE PROCESSING AND ANALYSIS

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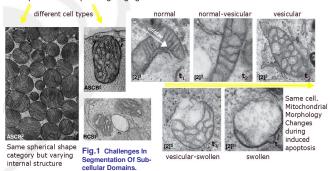






INTRODUCTION

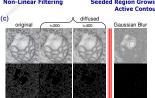
Mitochondria play a central role in cellular bioenergetics and in the regulation of apoptotic cell death. Mitochondrial morphology (shape and *cristae* architecture) is crucial to the understanding of apoptosis mechanisms and the subsequent development of therapies targeting age- and cancer-related diseases[1][2].



There is a high demand in automated segmentation tools which can provide an objective quantitative information in a reasonable time frame^[2]. The state of the art however is still dominated by manual segmentation tools^[3]. Early attempts to address the challenges shown above are based on the machine learning framework^[4].

THEORY AND METHODS

The variety of available methods^[5] reflects the complexity of the subcellular domain. We adopt a multi-stage segmentation process which consists of the removal of noise artifacts, salient feature enhancement, localization and shape extraction. Some



Joint Membrane Enhancement and Denoising: Anisotropic Diffusion [7]:

$$\begin{split} \frac{\partial I}{\partial t} &= \operatorname{div}\left(c(x,y,t)\nabla I\right) = \nabla c \cdot \nabla I + c(x,y,t)\Delta I \\ c\left(||\nabla I||\right) &= e^{-(||\nabla I||/K)^2} \quad c\left(||\nabla I||\right) = \frac{1}{1 + \left(\frac{||\nabla I||}{K}\right)^2} \end{split}$$

Fig.2 (a) Proposed approach. (b) GVF framework. (c) Anisotropic Diffusion (Perona-Malik)[7]

RESULTS I: SINGLE MITOCHONDRION

Here, the segmentation of a mitochondrion is entirely performed by means of Mathematical Morphology (MM). The nonuniform illumination is corrected in step II. Contrast Enhancement in step III, Filtering in step IV, Binary mapping in step V. Particle size filters analysis the components with area A < A_{th} in step VI. The gaps present in the outer membrane are sealed in step VII. Flood fill and shape refinement operations are given in VIII-IX and the extracted mitochondrial

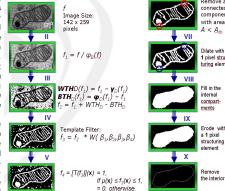


Fig.3 Case Study on the Segmentation of a Mitochondrion

RESULTS II: SEPARATED ORGANELLES

In the case of well separated organelles we apply MM tools to segment distinct objects. Here reconstruction-by-dilation, also frequently referred to as opening-by-reconstruction

on gray-scale images is being considered.

The notion of a *mask* and a *marker* along with the structuring element provide the basis for this approach.

The objective is to process the grayscale image such as to fill in the inner region of a mitochondrion. Granularity and texture analysis are used in the classification of objects.

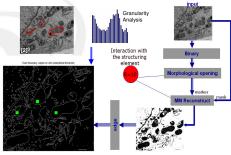


Fig.4 Segmentation of Separated Organelles with MM Method

RESULTS III: DEFORMED AND CLUSTERED

We apply a joint edge- and region-based segmentation approach which is based on the selection of appropriate localization markers and either the contour propagation or the marker-controlled region-grow. Low SNR of acquired TEM images results in disconnected membrane. Initialization and stopping criteria determine the yield.

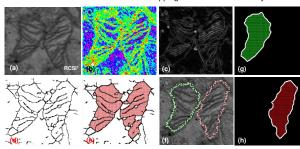
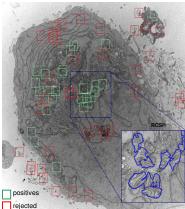


Fig. 5 (a) Original image. (b) Contour map. (c) Gradient map. (d) Binary skeleton. (e) Region

DISCUSSION AND OUTLOOK

TOWARDS the 'WHOLE CELL' AUTOSEGMENTATION



At present, our work is focused on the analysis of mammalian cells under the apoptosis treatment (sample preparation by RCSI¹). The present segmentation bottle-neck poses a considerable problem for Electron Tomography imaging. The imaging of a cell's interior typically consists of more than 100 slices and more than 30 distinct objects to be segmented¹¹].

■ Need for AUTOSEGMENTATION

The figure on the left shows our results for the STS treated cancer cell DU-145 as compared to the ground truth data. Our current and future work focuses on the development of auto-segmentation algorithms and tools.

Fig.6 STS treated DU-145 Localization and Segmentation Results.

ACKNOWLEDGMENT

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