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B-Spline Level Set For Drosophila Image Segmentation

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Conclusion and Perspectives

1. Introduction



1. Introduction

Segmentation algorithms

DNA, F-actin and GFP-Rac are used for nuclei and cytoplasm detection

□ Fuzzy C-means based Multi-Threshold segmentation and Sharpening technology

U Watershed algorithm based on simulation of flooding process

Deep-Learning with different Network Structures









Conclusion and Perspectives

2. Problem statement

Most of algorithms cited ab**Havato** based on the discrete level set innerleonne et thison

However

requires a periodical re-nipical blattloh step to reshape the contour which increases the computational cost



Continuous representation of level set function using the B-spline coefficients is proposed











Conclusion and Perspectives

3. B-spline Level Set for Image Segmentation

The key idea of the B-spline level set is

Expressing the level set \u03c6 as a linear combination of Bspline coefficients

$$\phi(x) = \sum_{K \in d} c \begin{bmatrix} k \end{bmatrix} \beta^n \left(\frac{x}{h} - k \right)$$

The energy function J is minimized with regard to B-spline coefficients ci

$$\nabla_{c}J = \frac{\partial J}{\partial c[k]} = \int_{\Omega} \omega(x) \beta_{h}^{n} (x - hk) dx$$

3. B-spline Level Set for Image Segmentation



3. B-spline Level Set for Image Segmentation



Less computational cost





B-Spline Level Set For Image Segmentation

Segmentation Results



Conclusion and Perspectives

Properties

• Drosophila RNAi experiments.

- Presence of locally low contrast values
- Large number of cells to be segmented

• Clustered cells. **Drosophila images** What kind or images are used in this work ?

- Dimension : 686×518 pixels
- Image Preprocessing : Local Normalization
- Source : Valrose institute of Biology (iBV)





In this work:

Results are presented for fixed parameters

Results are presented for varing parameters h and initialization

Results are presented for noisy images



Good segmentation performance is clearly noticed





➤ Calculates the degree of similarity between the segmented image and the ground truth.

DICE

>Ground truth are given with manual segmentation (provided by biologist)

> 1== identical regions 0 == different regions

	B-spline level set	Level set	Marker Controlled Watershed
Drosophila 1	0.7861	0.7279	0.6501
Drosophila 2	0.7422	0.6423	0.6819
Drosophila 3	0.7993	0.6999	0.6975
Drosophila 4	0.7423	0.6256	0.6422
FMI2	0.7648	0.6698	

DICE coefficient for Drosophila images and FMI 2



Scale h

h controls smoothing of level set With increasing h, the high level of smoothing makes cells blurry





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DICE coefficient as a function of the scale parameter for varying noise levels









Conclusion and Perspectives

5. Conclusion et perspectives

Conclusion

Improved version of B-spline level set is proposed based on Chan-Vese model

Takes in consideration intensities inhomogeneity Within Drosophila images

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B-spline level set outperforms the conventional level set and Marker Controlled Watershed

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5. Conclusion et perspectives



