

# Large-Scale Mitochondria Segmentation

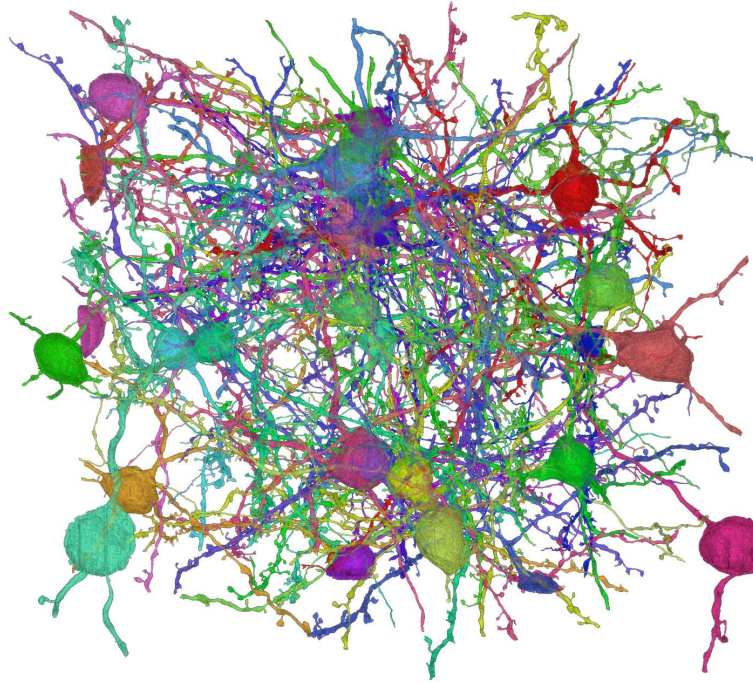
# Overview

1. Motivation
2. Problem Statement
3. Challenges
4. Related Work
5. Method
6. Post-processing
7. Results

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# Connectomics



# Need for Connectomics

1. Correlation between neuron wirings with
  - a. Cognition
  - b. Intelligence
  - c. Consciousness
2. How memory is stored
3. Evolution of wirings with age
4. Diseases

# Mitochondria

1. Powerhouse of the cell
2. Responsible for
  - a. Neuronal development <sup>1</sup>
  - b. Developmental Brain Diseases<sup>1</sup>
3. Relevant mitochondria features
  - a. **Morphology**
  - b. **Number density**
  - c. Dysfunctionality and more

<sup>1</sup>*Geurim Son and Jinju Han, “Roles of mitochondria in neuronal development” in BMB Reports 2018.*

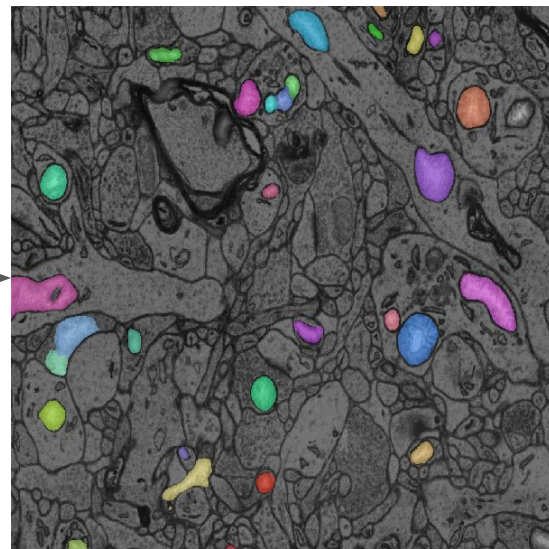
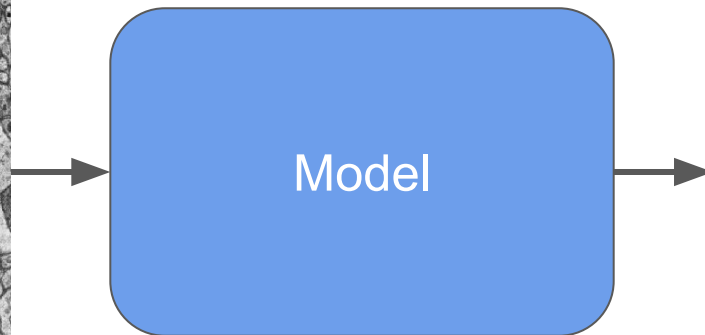
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# Mitochondria Segmentation



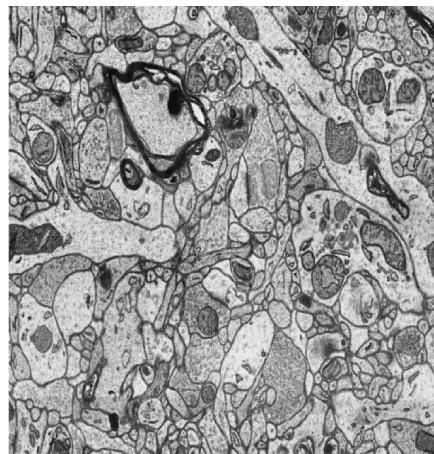
EM Image



Segmentation Map

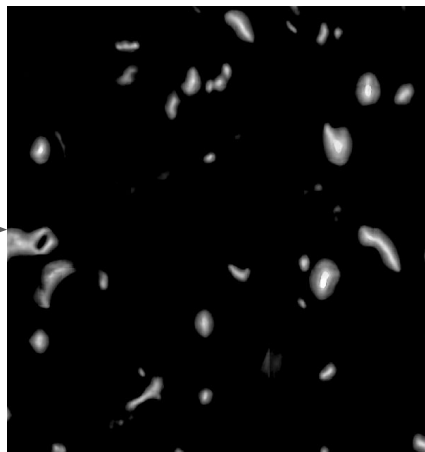


# Mitochondria Segmentation



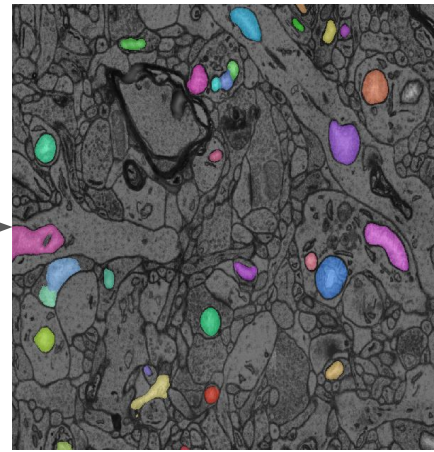
EM Image

Neural Network



Intermediate  
Representation

Inference



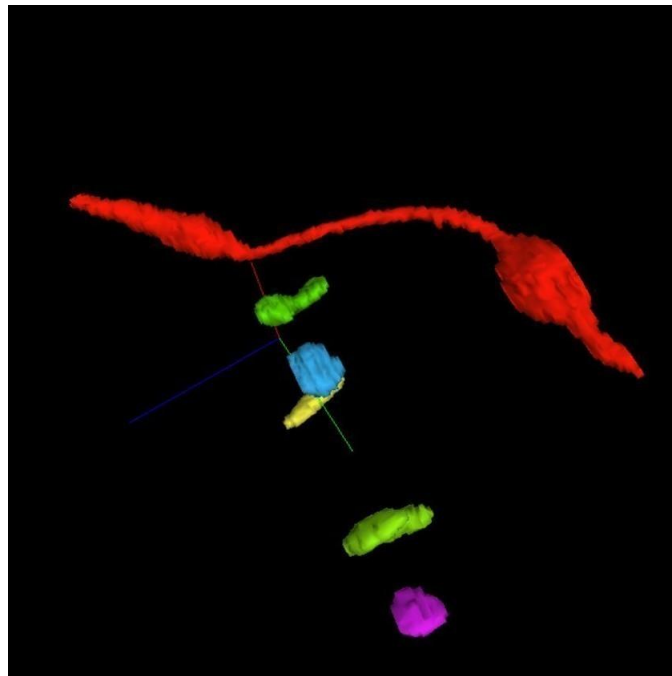
Segmentation  
Map

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# Challenges

1. Non-convex Morphology
2. Mitochondria in close vicinity
3. Annotated data and Generalization
  - a. Training data: 1x6Kx4Kx300
  - b. Test data: 49X4KX4KX1K



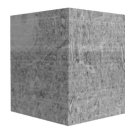
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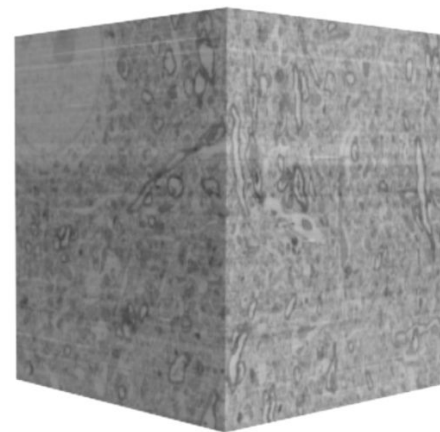


# Challenges

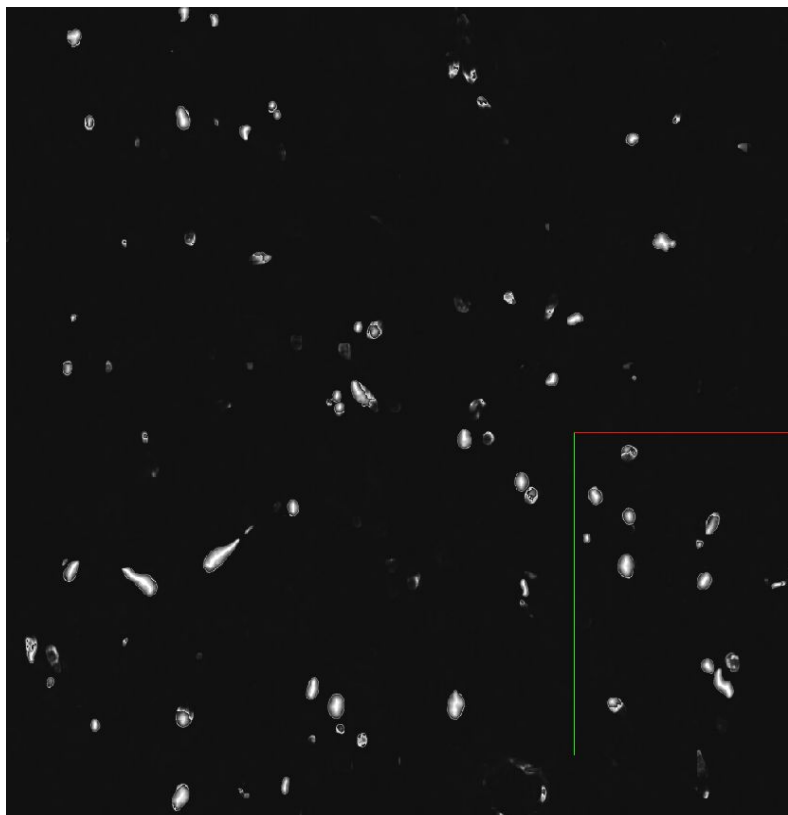
1. Non-convex Morphology
2. Mitochondria in close vicinity
3. Annotated data and Generalization
  - a. Training data:  $1 \times 6 \times 4 \times 300$
  - b. Test data:  $49 \times 4 \times 4 \times 1 \text{k}$



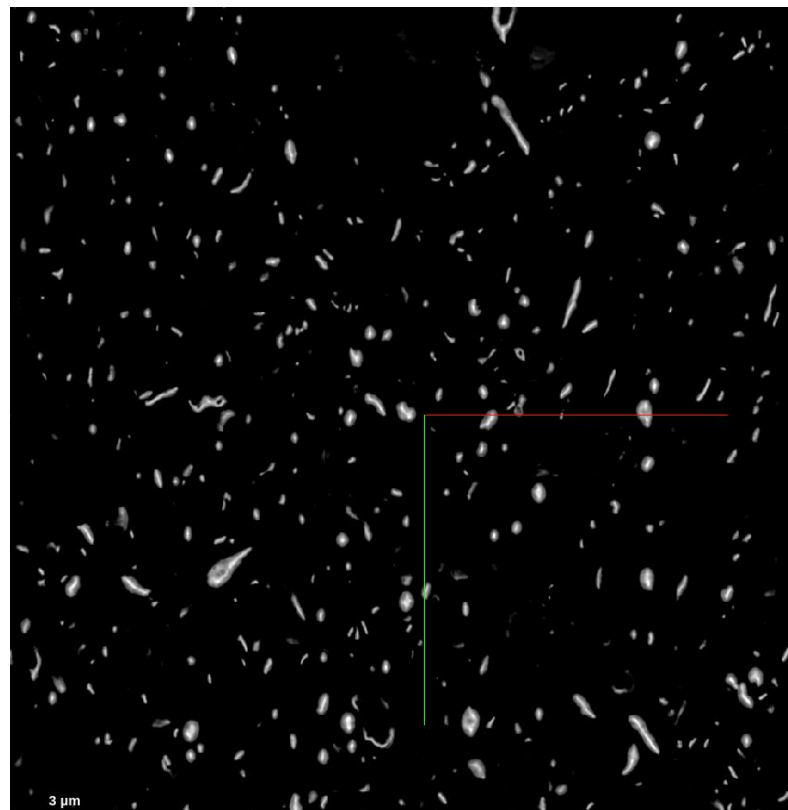
Train



Test



Output



Expected Output

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# Related Work

## 1. Boundary Prediction

- a. Distance Transform
- b. Seed Generation
- c. Flood filling rule
- d. False merges

## 2. Deep Watershed<sup>2</sup>

- a. Discretized watershed maps with the boundaries at the same level
- b. Cut at specific energy level
- c. Poor connectivity

<sup>2</sup>*M. Bai and R. Urtasun*, “Deep Watershed Transform for Instance Segmentation”. In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, pages 5221–5229, 2017. 2, 3, 5



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# Connectivity-Aware Watershed Energy

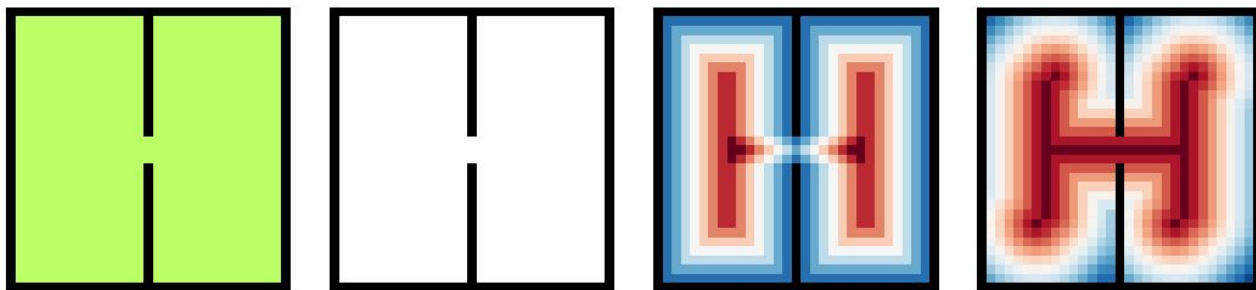
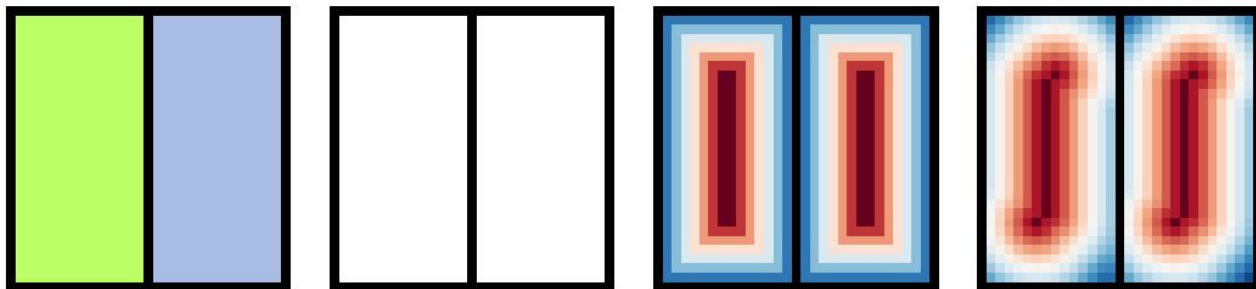
## 1. Ground Truth -> Energy Map

- a. Recursive erosion of boundary pixels
- b. Distance transform using skeleton pixels
- c. Normalization of distance transform

## 2. Track changes in connectivity

- a. Augment model with loss functions
- b. More weight for keypoints (skeleton ends and branch points) in loss function

# Connectivity-Aware Watershed Energy



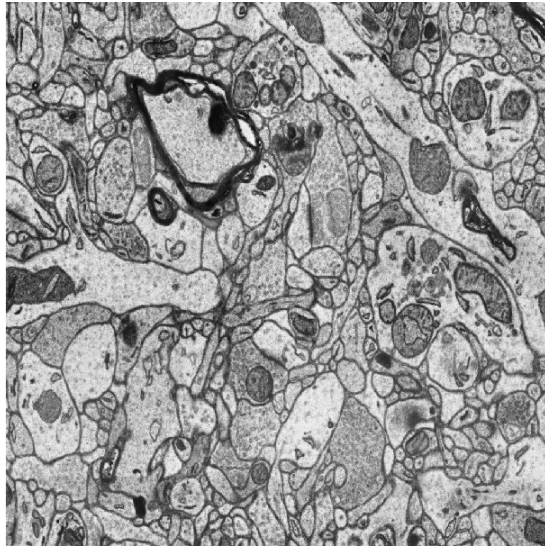
(a) Instance  
Masks

(b) Edge  
 $L^2 = 0.06$

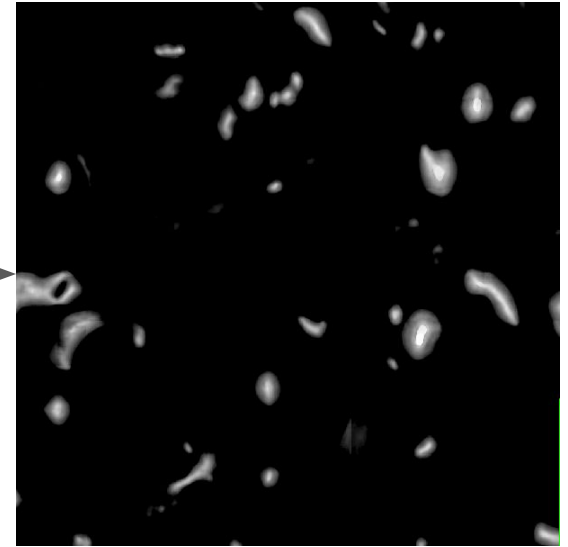
(c) DWT  
 $L^2 = 0.09$

(d) **Ours**  
 $L^2 = \mathbf{0.19}$

# Connectivity-Aware Watershed Energy



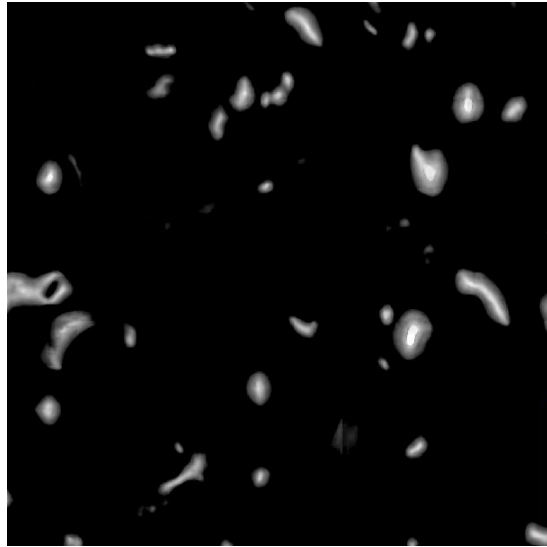
EM Image



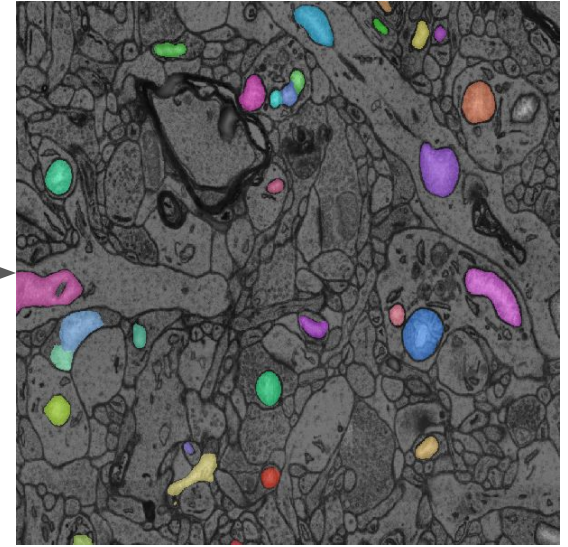
Distance Transform Map  
(Energy Map)

<sup>3</sup>O. Ronneberger and P. Fischer and T. Brox, "U-Net: Convolutional Networks for Biomedical Image Segmentation" in MICCAI 2015, vol. 9351, pp. 234-241, 2015

# Connectivity-Aware Watershed Energy



Distance Transform Map  
(Energy Map)



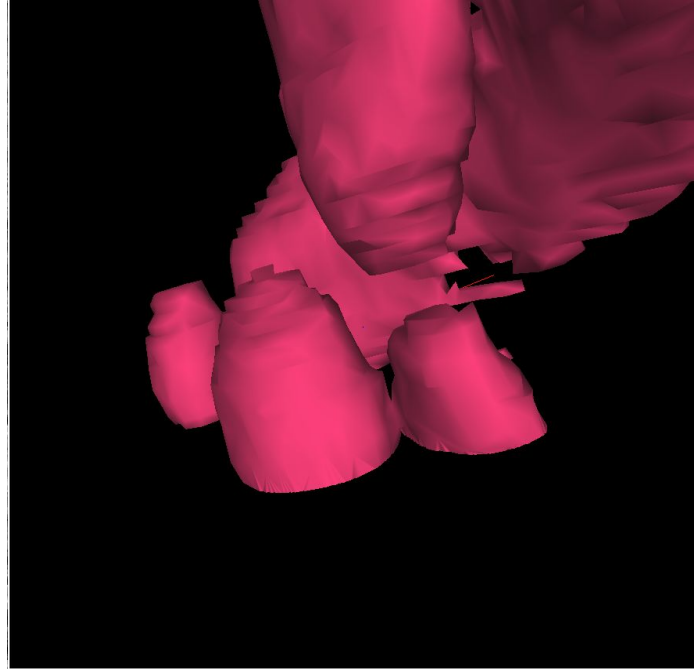
Segmentation Map

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# Previous Approaches

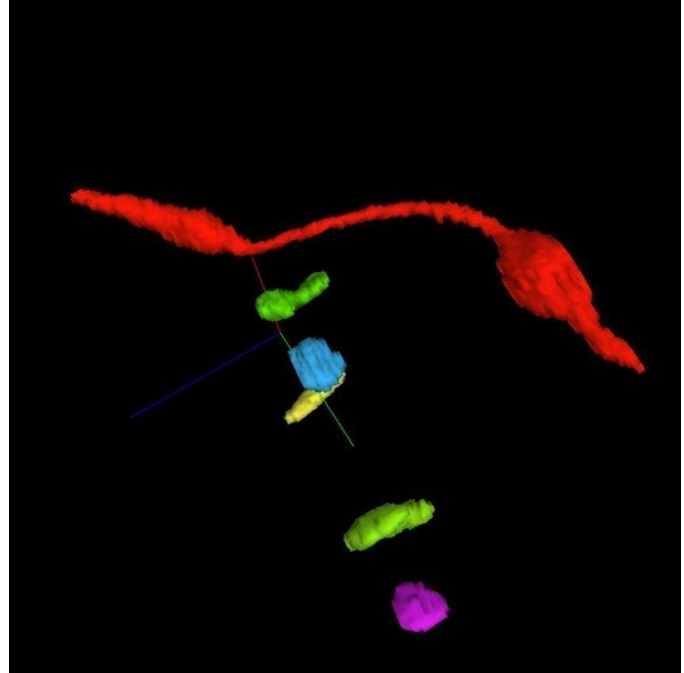
1. Connected Components
  - a. False Merges
2. CC with Binary Erosion/Dilation
  - a. False Splits
3. Watershed
  - a. Oversegmentation





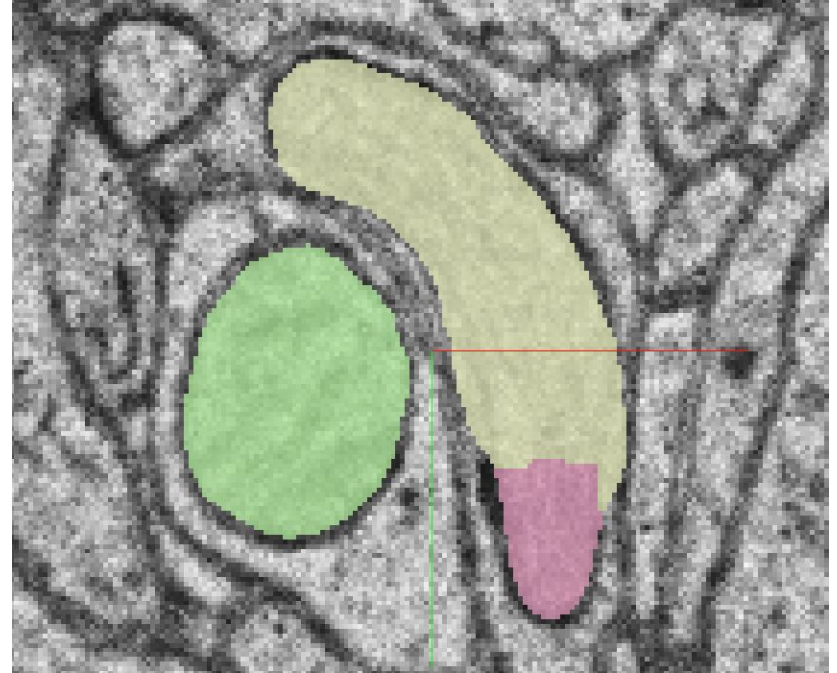
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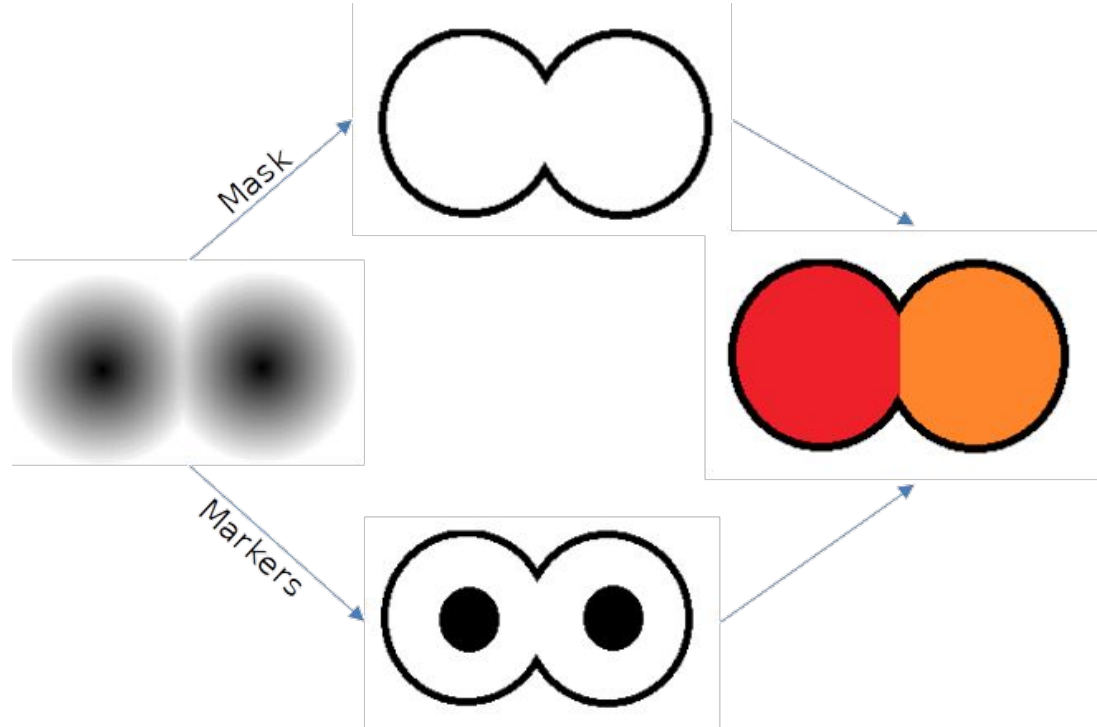
# Previous Approaches

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# Marker Generation for Watershed

1. Threshold Energy Map
  - a. Marker generation
2. Binarize CC output
  - a. Mask Generation
3. Watershed



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# Results (Annotated Cell Body)

**Matching Criteria: > 0.8 IoU**

	True Positives	False Positives	False Negatives
Initial Prediction	49	12	20
After Error Correction	59	12	10

**Precision/Recall (by pixel): 94% / 96%**