Vaa3d Workshop

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Background

• Swc is not on the signal due to the inconsistent image resolution during the labeling process

Our problems

I 3D View [D:/Refinement/refinement_0810/long/Img_x_9498.54_y_24132.3_z_4085.64.v3draw]





I 3D View [D:/Refinement/refinement_0810/distal/Img_x_8857.54_y_20346.8_z_4959.v3draw]

Mean-shift

• Move the signal to the centroid of the brightness of the picture

$$m(x_i) = \frac{\sum_{i=1}^{n} x_i g(||\frac{x-x_i}{h}||)^2}{\sum_{i=1}^{n} g(||\frac{x-x_i}{h}||)^2} - x$$

Pros and cons:

• Meanshift:

Advantage: work well on dentrites (bold light signal) make the swc close to the centerline

Disadvantage: rely on threshold and window radius may easily make mistakes

Remaining problem:

II 3D View [D:/Refinement/refinement_0810/long/Img_x_9498.54_y_24132.3_z_4085.64.v3draw]













GD(graph-augmented deformable model)

• Step1: initialize a "shortest" path based on Euclidean distance and image intensity between given point A and B.

$$e(v_0, v_1) = \|v_0 - v_1\| \cdot \left(\frac{g_I(v_0) + g_I(v_1)}{2}\right)$$

• Step2: optimize the objective function

$$L(C) = \int_C g(p, I[\Theta(p, r)]) \|dp\|$$

Step1: divide neuron into segments

Step2: shift branch point on signal

Step3: run GD twice





Pros and cons:

• GD:

Advantage: do a much better refinement job

Disadvantage:

lean on given points and might be disturbed by other signalsbad on dentrites (bold light signal)There is no guarantee that swc is on the centerlineSwc may become rugged





Current plan

- Step1: Divide swc into segments according to branch points
- Step2: Run GD on each segment
- Step3: Smooth and shift swc to centerline (meanshift)

To do:

- There needs to be a quantitative method to evaluate the effect of refinement.
- Ensure that the swc is in the center of the signal
- Further refinement