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Automatic Angiography Segmentation based on improved Graph-cuts

Motivation

Angiography image



Multi-Sliced Computed Tomography (MSCT)







However, registration methods are sensitive to the background noise present in angiography images. In order to obtain a more precise registration we need a pre-processing step.

Chronic total occlusions (CTO) are obstructions of native coronary arteries. Recanalization of a CTO still remains a challenge for invasive cardiologists [1]. MSCT is a valuable technique for the non-invasive visualization of both the lumen and the features of the arterial wall of coronary vessels.

Registration of CT to X-Ray images is a valuable tool to provide complete and high quality 3D information in addition to the poor X-ray images [2].

Segmentation!

Automatic Vessel Segmentation

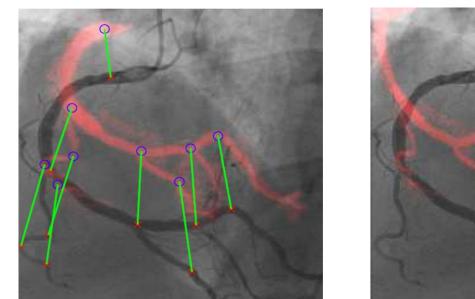
Our automatic vessel segmentation method is based on the graph-cuts [3] energy minimization framework. In this framework, an energy function is minimized in order to find the optimal segmentation of the image, using region and context information.

 $\mathbf{E}(\underline{\alpha}, \mathbf{k}, \underline{\theta}, \mathbf{z}) = U(\underline{\alpha}, \mathbf{k}, \underline{\theta}, \mathbf{z}) + V(\underline{\alpha}, \mathbf{z})$

Region potential

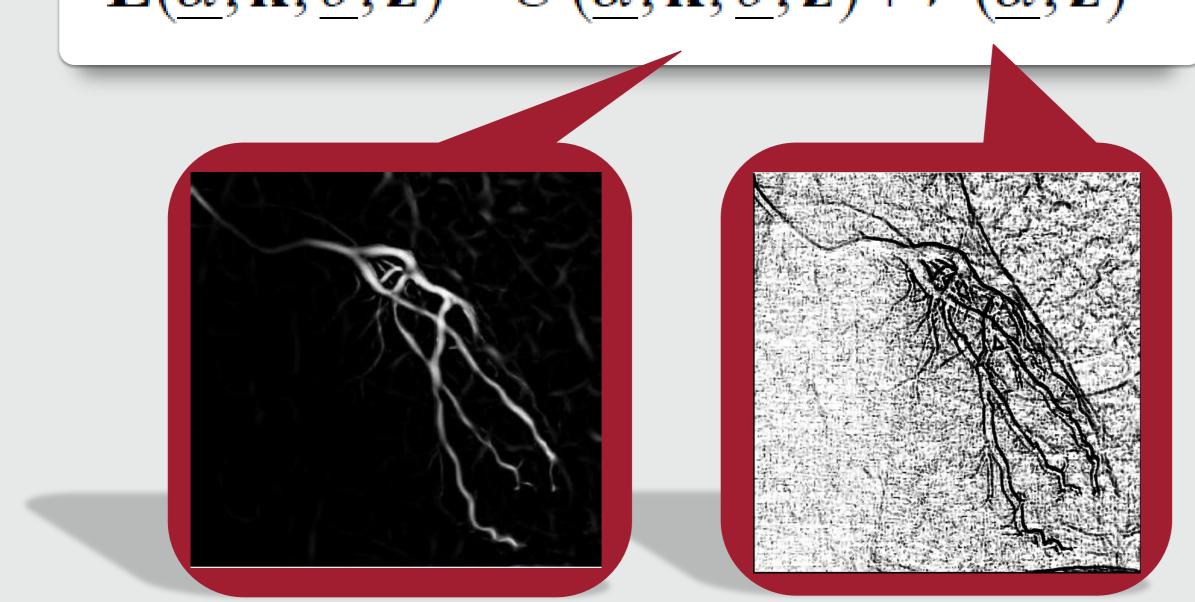
This potential encodes local information, computing a vessel probability or vesselness [4] at each pixel. This vesselness is quite poor at vessel bifurcations, so we add geodesic paths information [5] in order to improve this region information at those critical points.

Registration [6]



Not registered

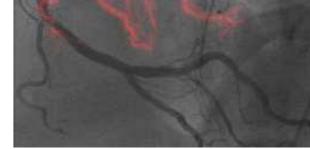




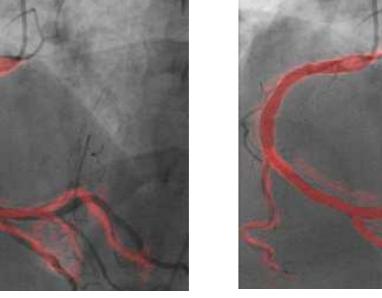
Region potential

Boundary potential

With this potential we provide contrast information to the segmentation framework. Simple pixel differences are computed, but the result is quite noisy. We propose a multi-scale approach in order to smooth this noise.



Gray-level



Vesselness

Segmented

Results



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Boundary potential

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