

Re-initialization Free Level Set Evolution via Reaction Diffusion-Supplementary Materials

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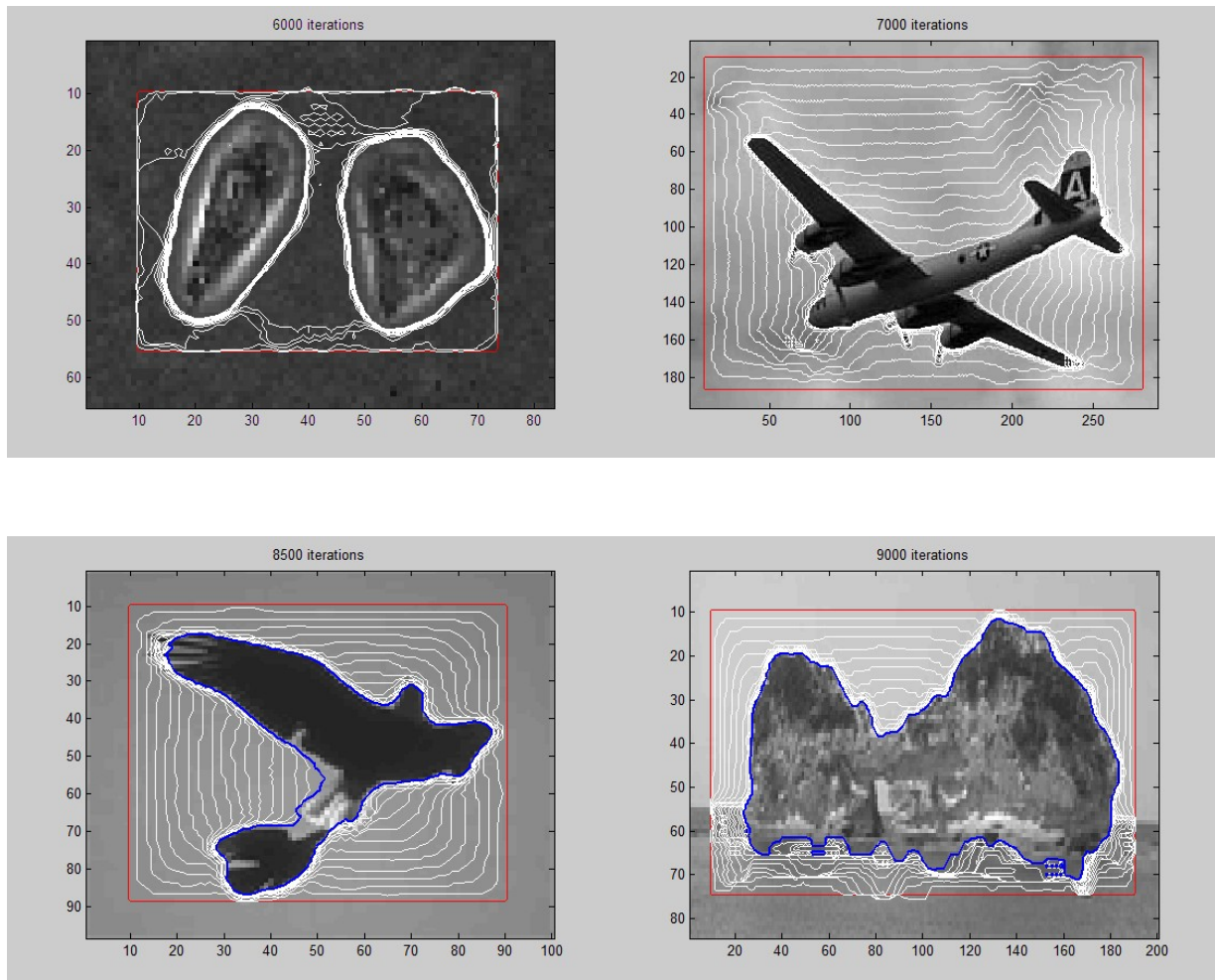
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Some Experimental Results in our paper:

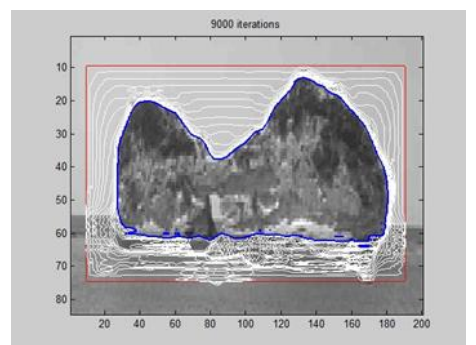
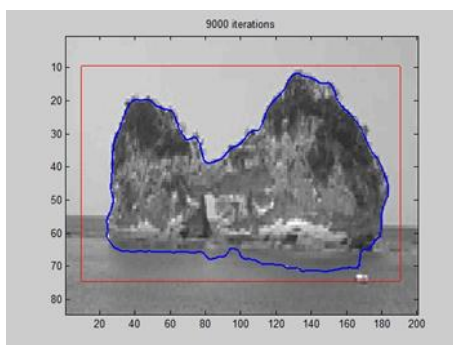
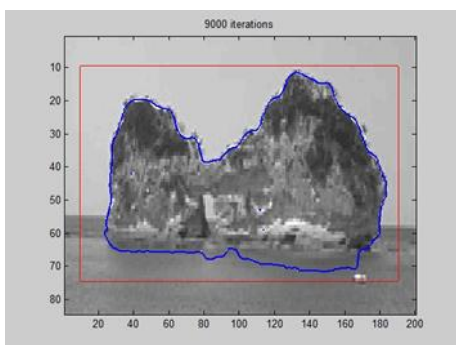
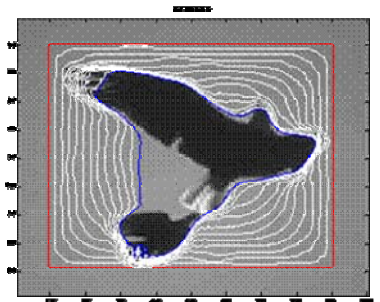
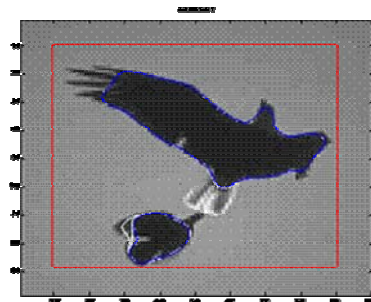
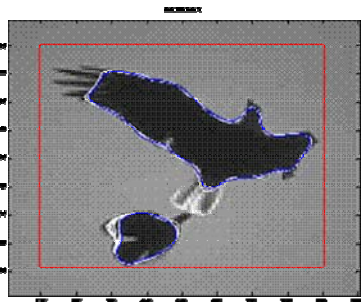
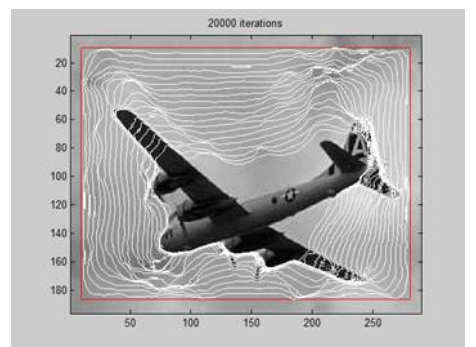
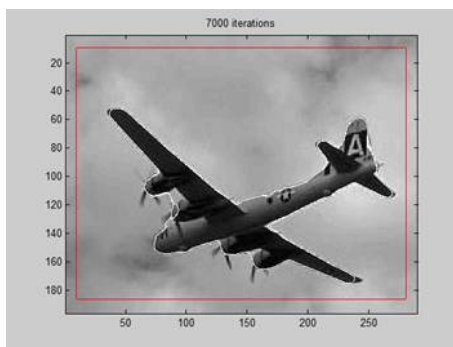
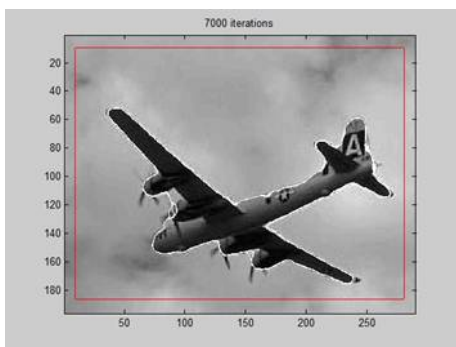
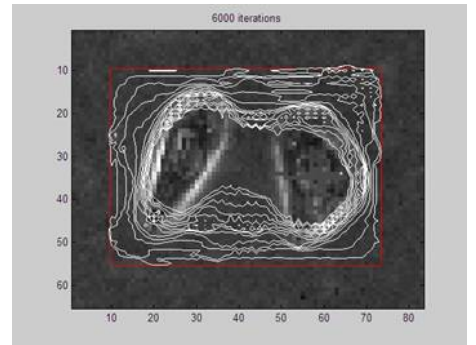
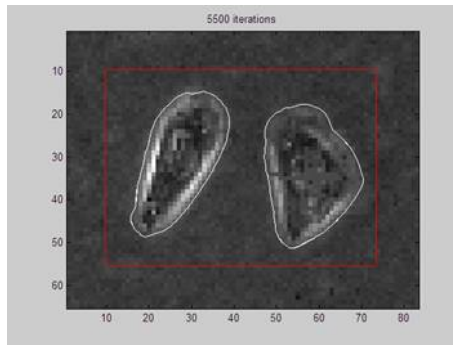
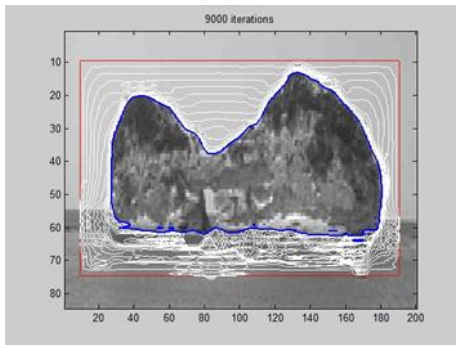
1. Segmentation of example images with weak edges

(a) $\delta_{1,\rho}(\phi)$ is used.

Results by our RD method [1].



Results by GDRLE₁ [10], GDRLE₂ [10] and GDRLE₃ [6], respectively. For GDRLE₁ and GDRLE₂, time step is set as $\Delta t=5$ as in the original paper.



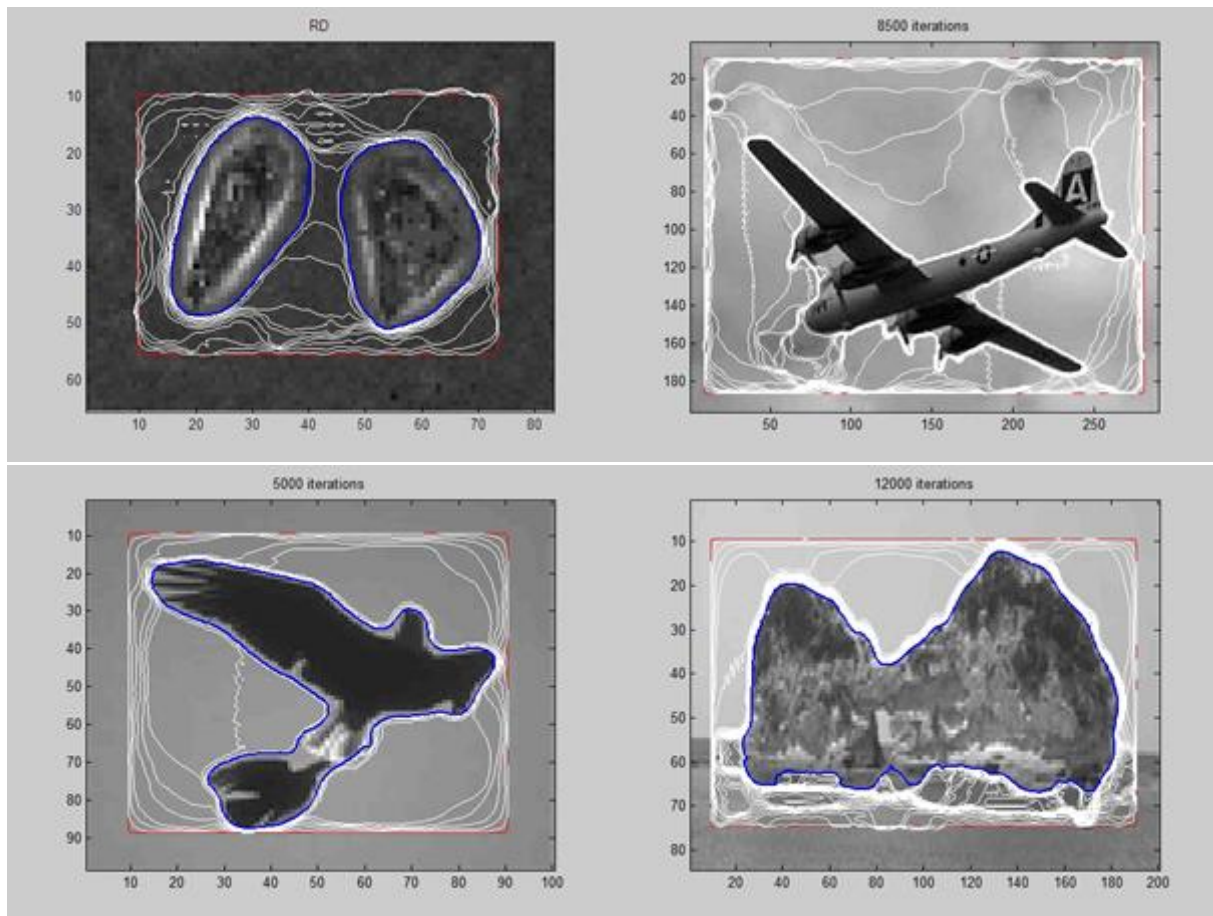
GDRLE₁

GDRLE₂

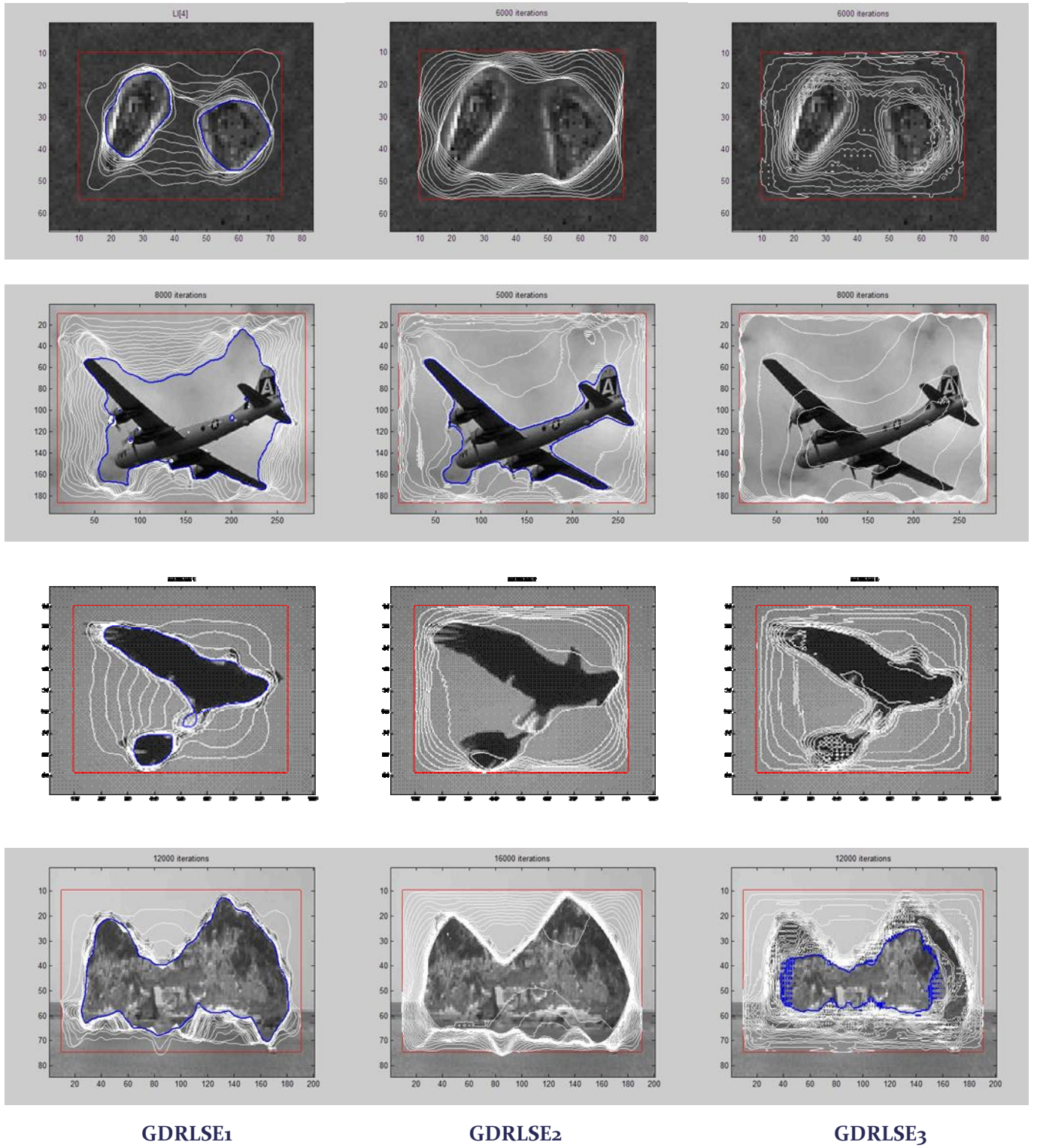
GDRLE₃

(b) $\delta_{2,\rho}(\phi)$ is used.

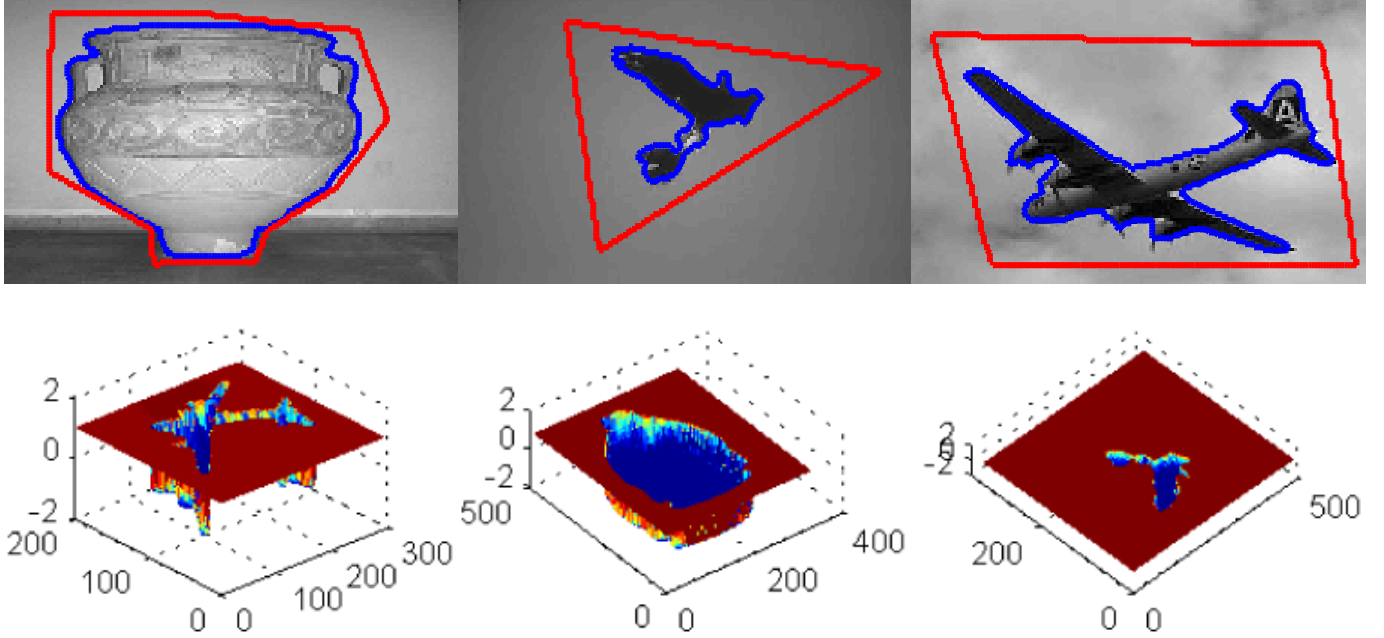
Results by our RD method [1].



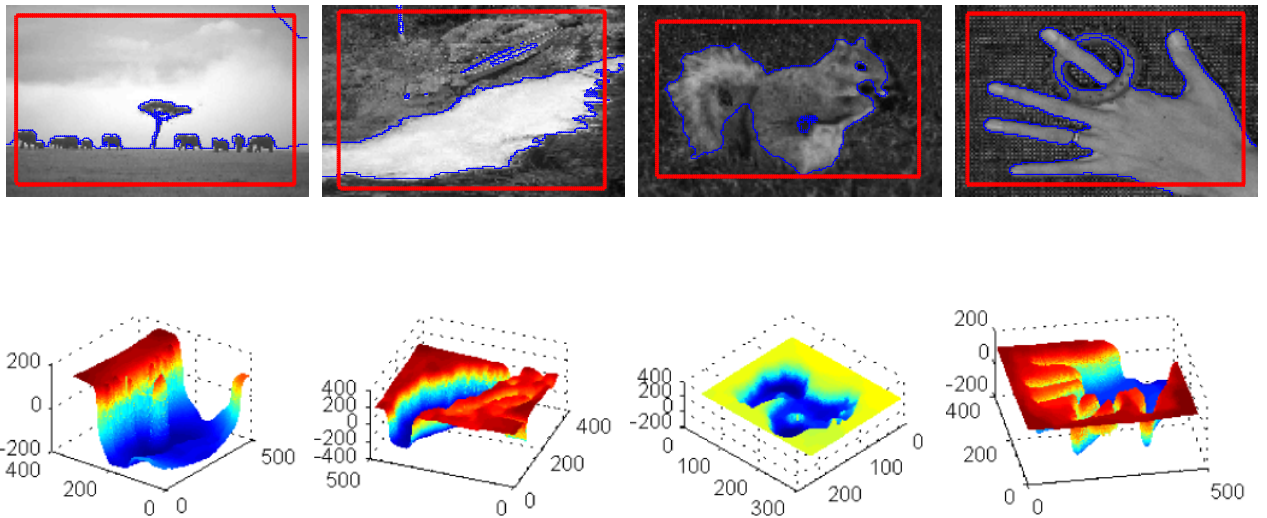
Results by GDRLSE₁ [10], GDRLSE₂ [10] and GDRLSE₃ [6], respectively.



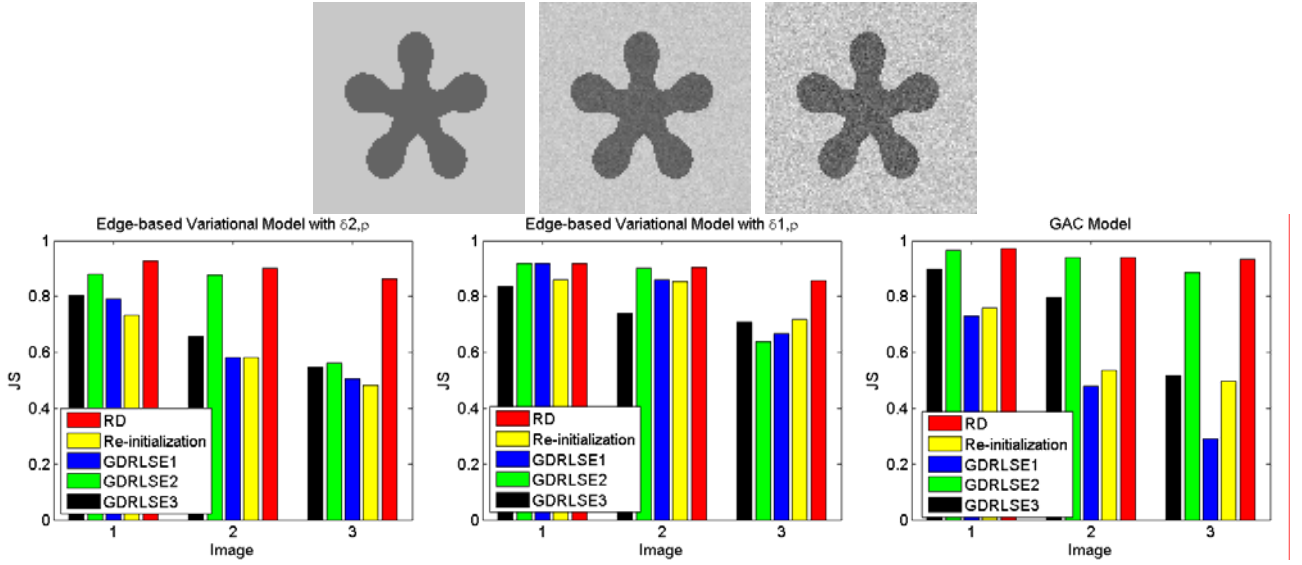
2. Segmentation results on three images in the Corel dataset (downloaded from [46]). Top row: initial contours (red lines) and final contours (blue lines). Bottom row: final LSFs. Parameter settings: Left-most column and right-most column: $\Delta t_1=0.1$, $\Delta t_2=0.001$, $\nu=0.5$; Middle column: $\Delta t_1=0.1$, $\Delta t_2=0.001$, $\nu=0.1$.



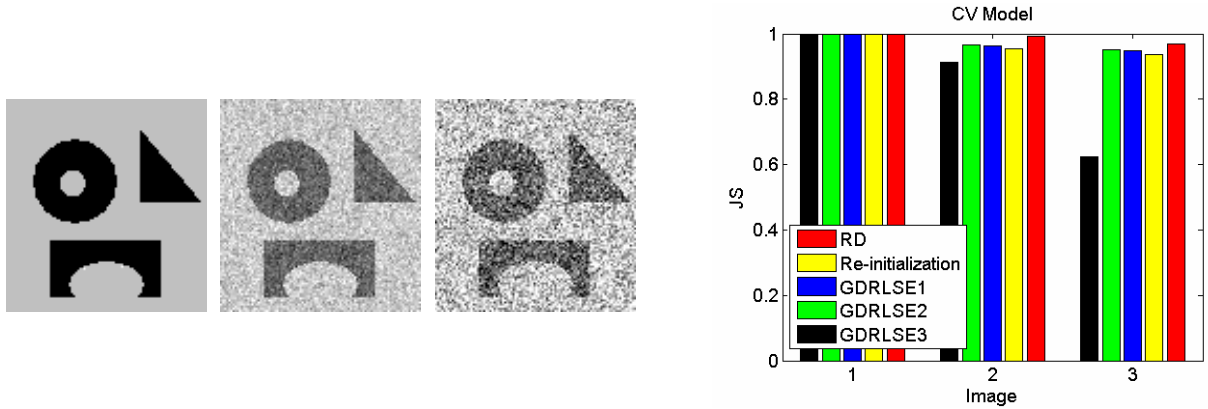
3. Segmentation results on four real images (top two: downloaded from [46]; bottom two: downloaded from [47]). Top row: initial contours (red lines) and final contours (blue lines). Bottom row: final LSFs. We set the parameters $\Delta t_1 = 0.1$, $\Delta t_2 = 0.01$, $\mu = 0.2 \times 2552$, $\nu = 0$, $\lambda_1 = \lambda_2 = 1$ for all images except for the squirrel image, for which we set $\mu = 0.1 \times 2552$.



4. Quantitative comparisons among RD and GDRLSE methods for edge-based models. Top row, from left to right: clean image (image 1), noisy image (image 2) (Gaussian noise with zero mean and standard deviation $\sigma=0.001$), and noisy image (image 3) (Gaussian noise with zero mean and standard deviation $\sigma=0.005$). Bottom row, from left to right: the JS values using the edge-based variational model with δ_2, ρ and δ_1, ρ in Section 5.3, and GAC model in Section 5.4, respectively. For edge-based variational models, we set $\Delta t_1=0.1$, $\Delta t_2=0.001$, $\alpha=0.2$, $\lambda=1$, $\nu=0.05$ for all the three images. For the GAC model, we set $\Delta t_1=0.1$, $\Delta t_2=0.001$, $\alpha=0.2$, $\lambda=1$ for all the three images, and we set $\nu=0.05$, $\nu=0.2$, $\nu=0.5$ for the images from left to right.

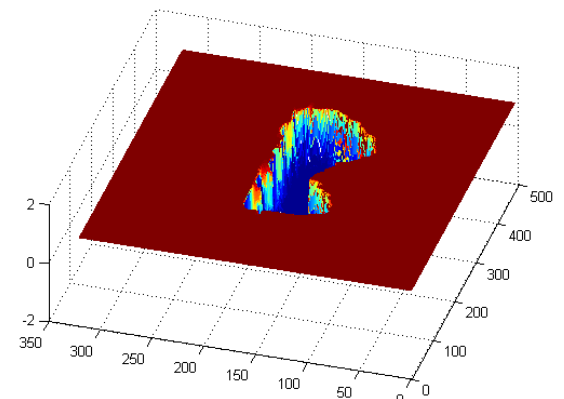
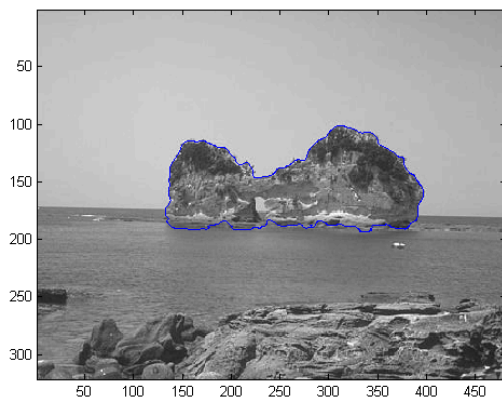
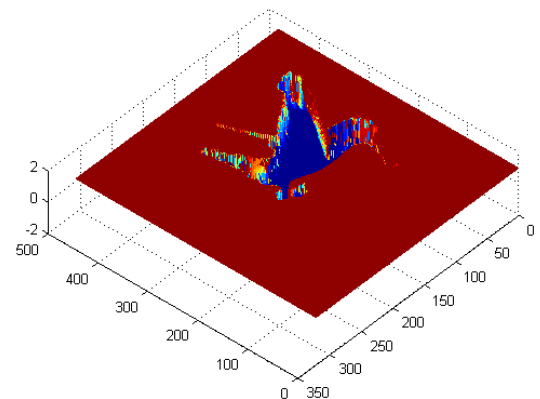
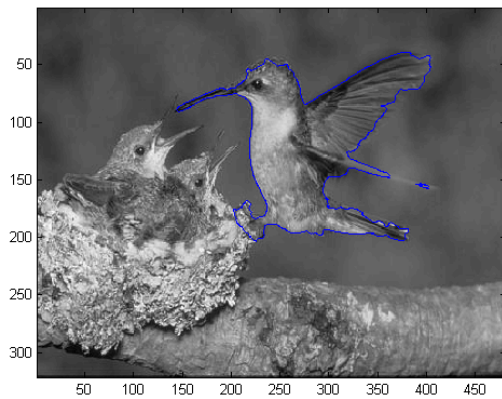
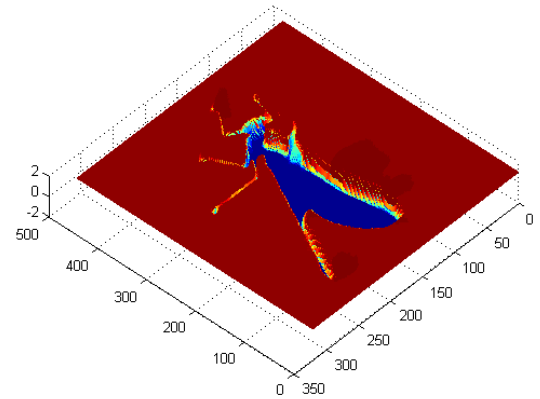
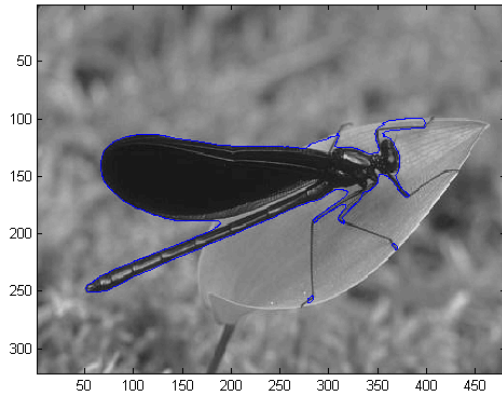


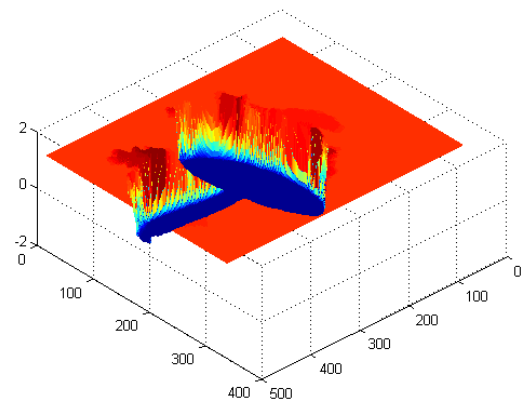
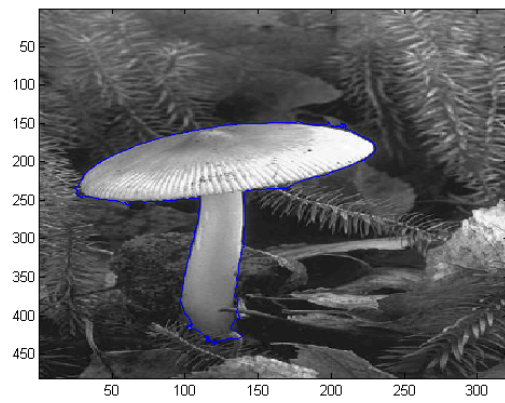
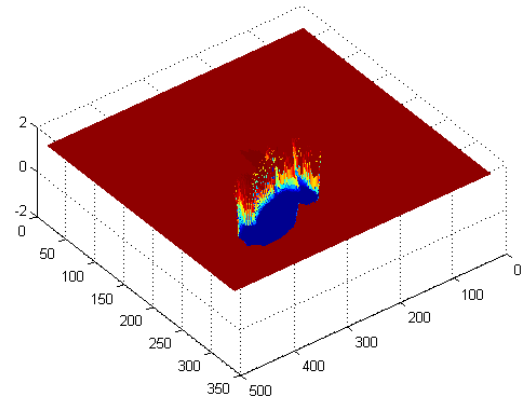
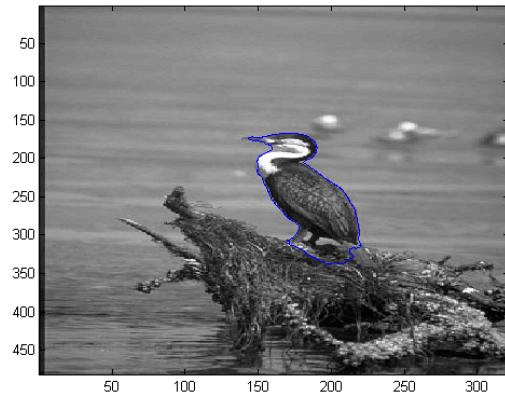
5. Quantitative comparisons among RD, re-initialization, and GDRLSE methods for the CV model [18]. Left three images: clean image, images with Gaussian noise of zero mean and standard deviation $\sigma=0.01$, $\sigma=0.05$, respectively. Right image: the JS values by competing methods. We set $\Delta t_1=0.1$, $\Delta t_2=0.01$, $\mu=0.1 \times 2552$, $\nu=0$, $\lambda_1=\lambda_2=1$ for all three images.



More Experimental Results :

1. Apply to GAC model [7].

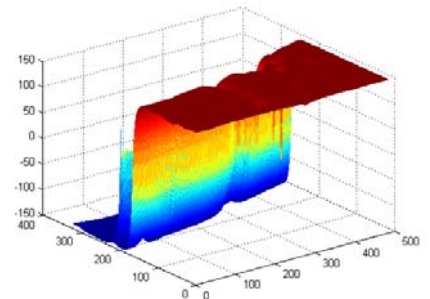
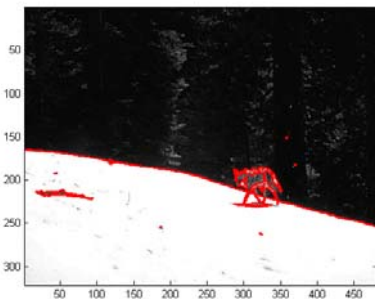
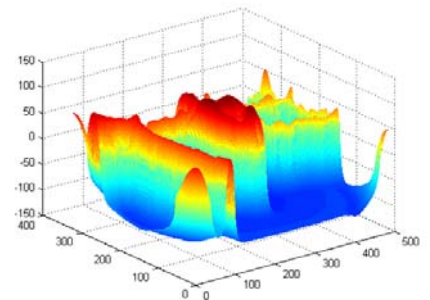
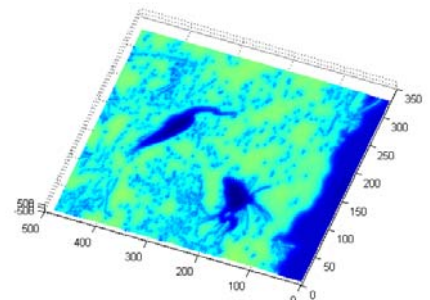
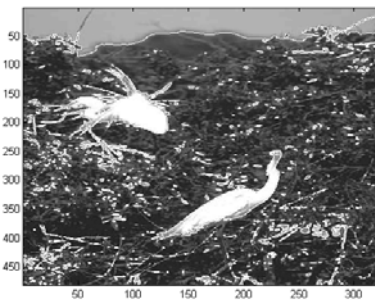
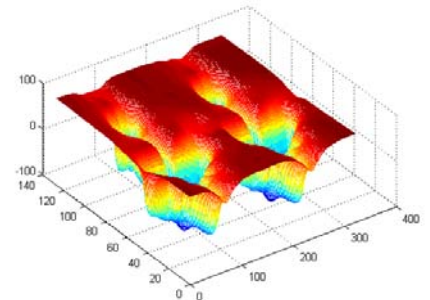
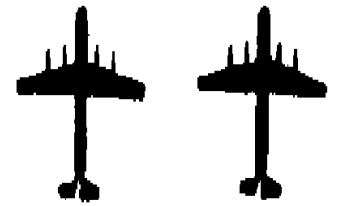
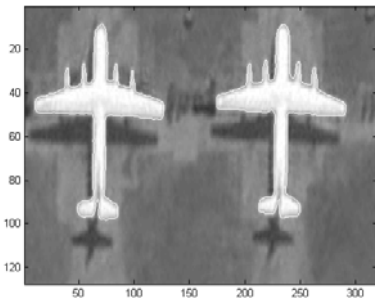
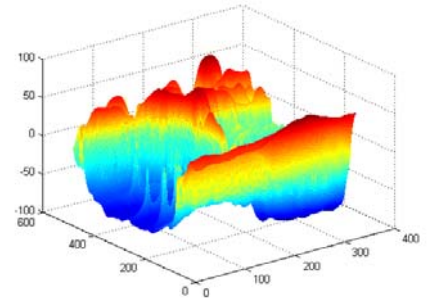
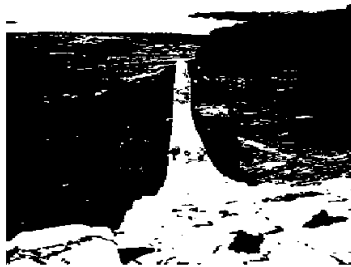
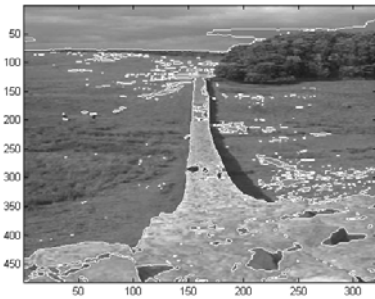


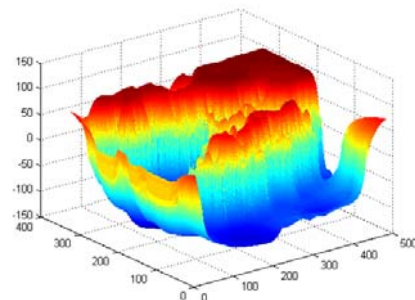
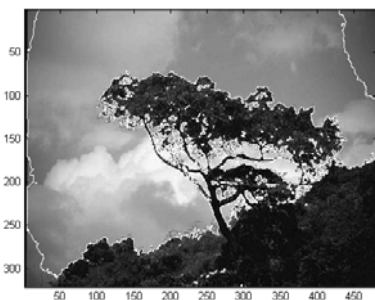
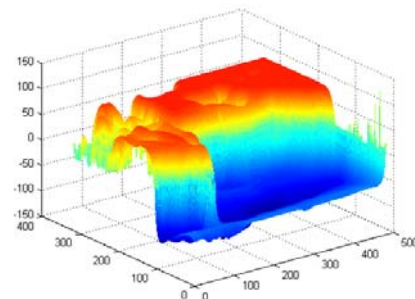
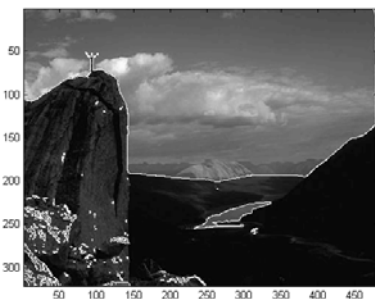
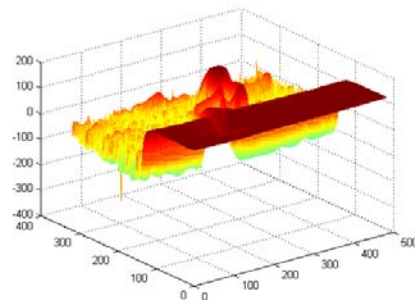
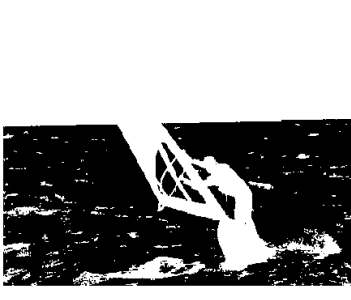
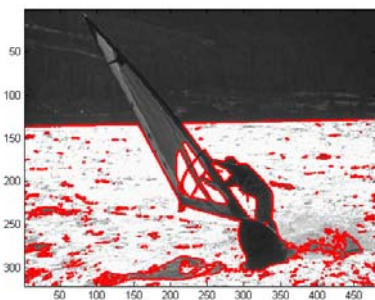
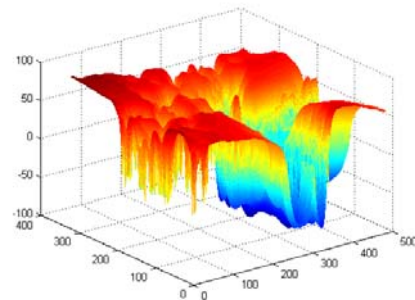
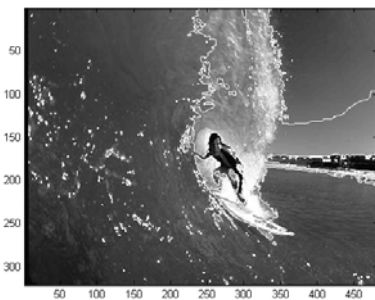


Segmentation results

Final LSF

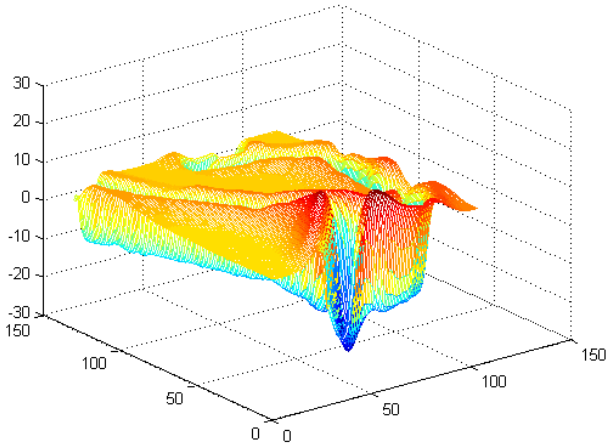
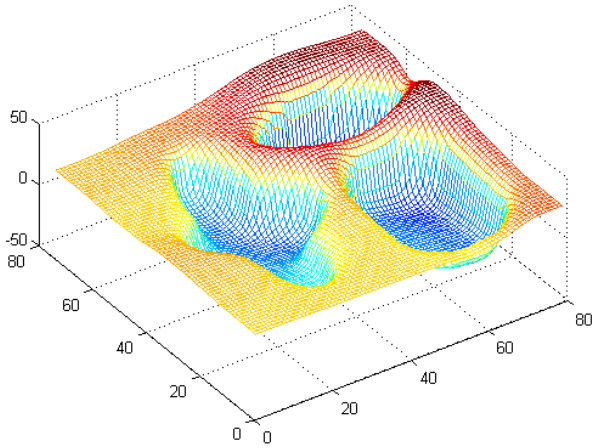
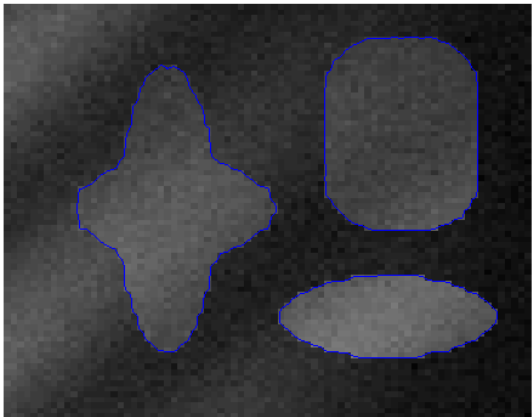
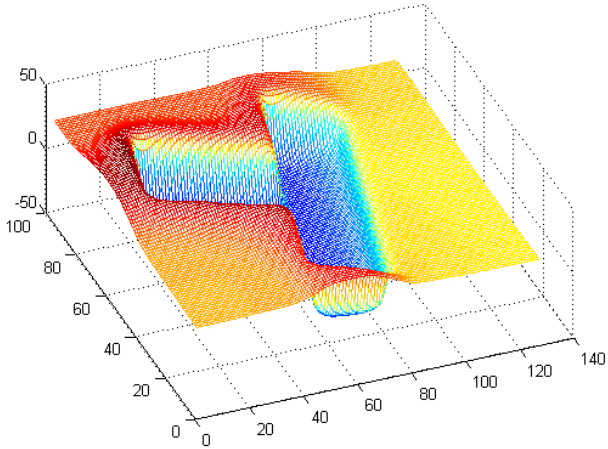
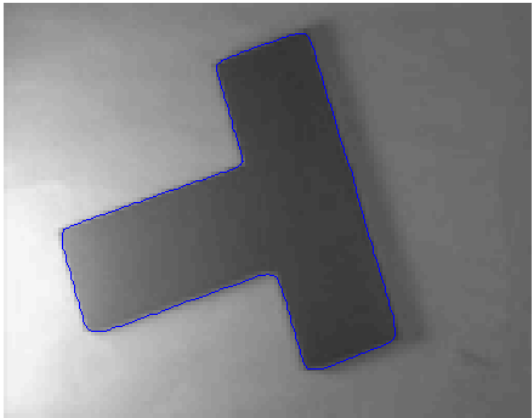
2 . Apply to CV model [8].

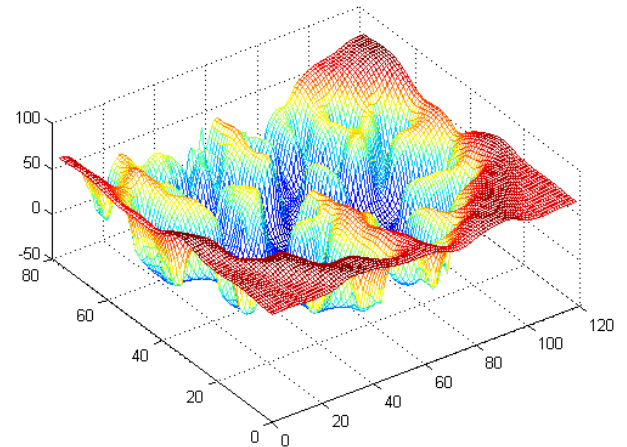
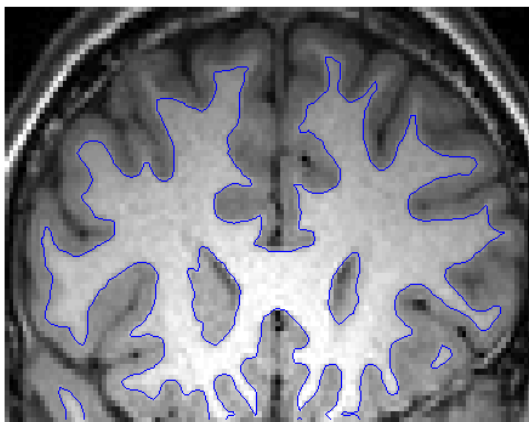
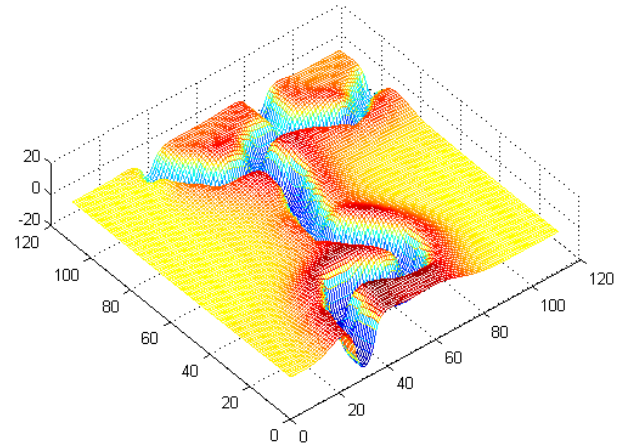
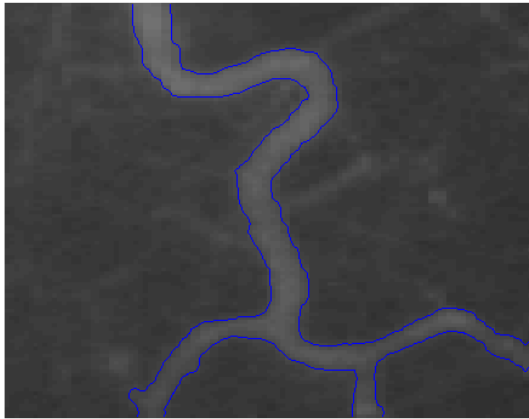




(a) Results represented by boundaries (b) Results represented by regions (c) Final LSF

3 . Apply to LBF model for images with intensity inhomogeneity [9].





(a) Final Results

(b) Final LSF

References:

- [1] K.Zhang, L.Zhang, H.Song, and D.Zhang, Re-initialization Free Level Set Evolution via Reaction Diffusion. IEEE Trans. On Image Processing.

Our former related work

- [2] K.Zhang, L.Zhang, H.Song and W. Zhou, "Active contours with selective local or global segmentation: a new formulation and level set method," Image and Vision Computing, vol. 28, issue 4, pp. 668-676, April 2010.
- [3] K.Zhang, H.Song, and L.Zhang, "Active contours driven by local image fitting energy," Pattern recognition, vol.43, no.4, pp.1199-1206, 2010.
- [4] K.Zhang, L.Zhang and S.Zhang, "A VARIATIONAL MULTIPHASE LEVEL SET APPROACH TO SIMULTANEOUS SEGMENTATION AND BIAS CORRECTION," ICIP 2010.

Other related work

- [5] C. Li, C. Xu, C. Gui, and M. D. Fox, "Level set evolution without re-initialization: A new variational formulation," Proc. IEEE Conf. Computer Vision and Pattern Recognition, vol. 1, pp. 430-436, 2005.

- [6] X.Xie, "Active Contouring Based on Gradient Vector Interaction and Constrained Level Set Diffusion," IEEE Trans. Image Processing, vol. 19, no. 1, pp. 154-164, 2010.
- [7] V.Caselles, R.Kimmel, and G.Sapiro, "Geodesic Active Contours," Int. J. Comput. Vis., vol.22, no.1 pp. 61-79,1997.
- [8] T. Chan and L. Vese, "Active contours without edges," IEEE Trans. Image Process, vol. 10, no. 2, pp. 266-277, Feb. 2001.
- [9] C. Li, C. Kao, J. Gore, and Z. Ding, "Implicit Active Contours Driven by Local Binary Fitting Energy," Proc. IEEE Conf. Computer Vision and Pattern Recognition, pp. 1-7, 2007.
- [10]C. Li, C. Xu, C. Gui, and M. D. Fox, "Distance Regularized Level Set Evolution and Its Application to Image Segmentation," IEEE Trans. Image Processing, vol. 19, no. 12, pp. 154-164, Dec. 2010.