Image segmentation

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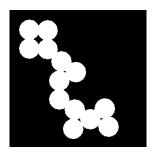
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Introduction

Goal

We want to segment an image f through the Chan-Vese Segmentation.

An example of image to segment:



Chan-Vese segmentation

The Chan-Vese segmentation finds the curve \mathcal{C} that is the boundary of the segmentation. The way to find that curve is minimizing the following functional:

$$\operatorname{arg\ min}_{c_1,\ c_2,\ C}\ \mu \operatorname{Lenght}(C) + \nu \operatorname{Area}(\operatorname{inside}(C)) +$$

$$\lambda_1 \int_{inside(C)} |f(x) - c_1|^2 dx + \lambda_2 \int_{outside(C)} |f(x) - c_2|^2 dx,$$

where C is the boundary of a closed set and c1, c2 are the values of u respectively inside and outside of C.

Level set functions

As it is difficult to manipulate C we will use a function ϕ and the C will be the zero crossing of ϕ that is

$$C = \{x \in \Omega : \phi(x) = 0\}.$$

With this the functional is rewritten as:

$$\arg \min_{c_1,c_2,\phi} \mu \int_{\Omega} \delta(\phi(x)) |\nabla \phi(x)| dx + \nu \int_{\Omega} H(\phi(x)) dx +$$

$$\lambda_1 \int_{\Omega} |f(x) - c_1|^2 H(\phi(x)) dx + \lambda_2 \int_{\Omega} |f(x) - c_2|^2 (1 - H(\phi(x))) dx,$$



Level set functions

In the previous formula H denotes the Heaviside function and δ the Dirac mass, its distributional derivative:

$$H = \left\{ egin{array}{ll} 1 & t \geq 0, \\ 0 & t < 0 \end{array}
ight., \quad \delta(t) = rac{d}{dt}H(t).$$

Note that we cannot derive H(t). Because of that, in the implementation we take the Heaviside function as

$$H_{\epsilon}(t) = rac{1}{2}igg(1 + rac{2}{\pi}rctanigg(rac{t}{\epsilon}igg)igg)$$

and the Dirac mass as

$$\delta_{\epsilon} = \frac{\epsilon}{\pi(\epsilon^2) + t^2)}$$



Implementation

Now we have to minimize the functional respect to c_1 , c_2 and ϕ . The way to do it is the following:

• Update c_1 and c_2 as

$$c_1 =$$

and

 c_2

.

f 2 Evolve ϕ using the semi-implicit gradient descent

$$\phi_{i,j} =$$

.



Results