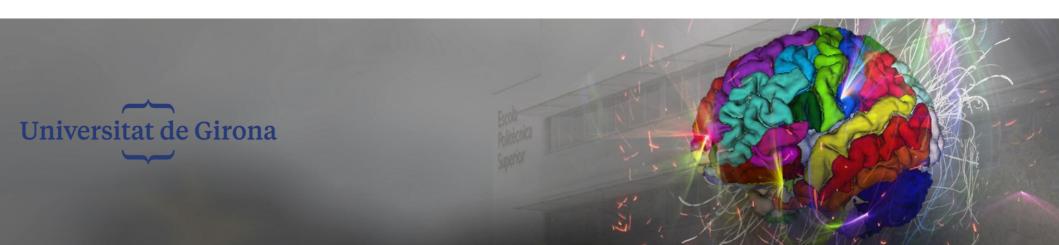


Computer Aided Diagnosis (CAD)

Deformable Models Segmentation





Part 1: Template matching & deformable template matching

Part 2: Active shape & Active appearance models

Part 3: Active contours & level sets





- Active contours (or snakes) segmentation.
- No prior information of the object to be segmented (although it can be added in posterior improvements).
- An initial boundary represented by points is iteratively modified by applying shrink/expansion operations.





Active contours (or snakes) segmentation.



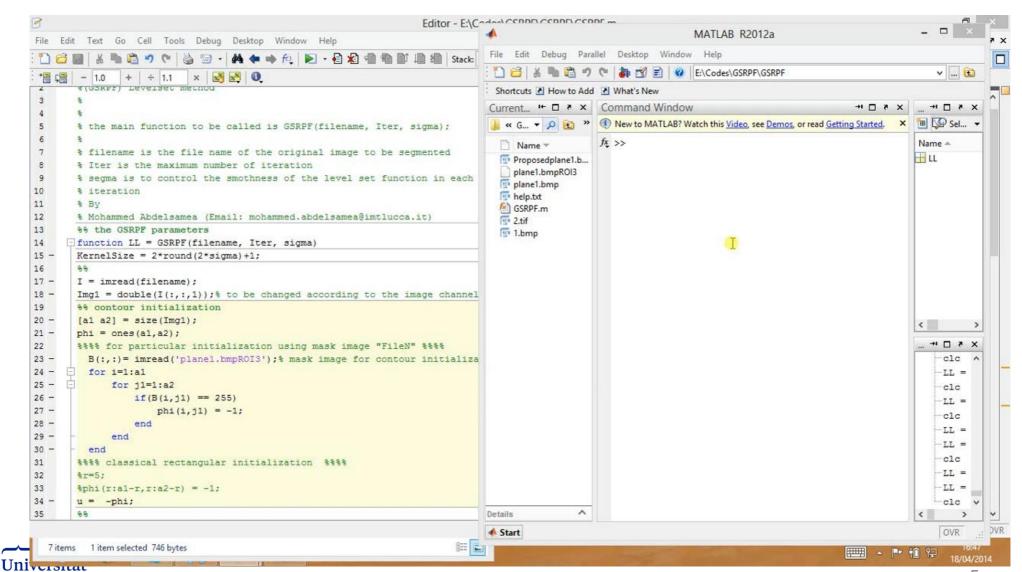




de Girona

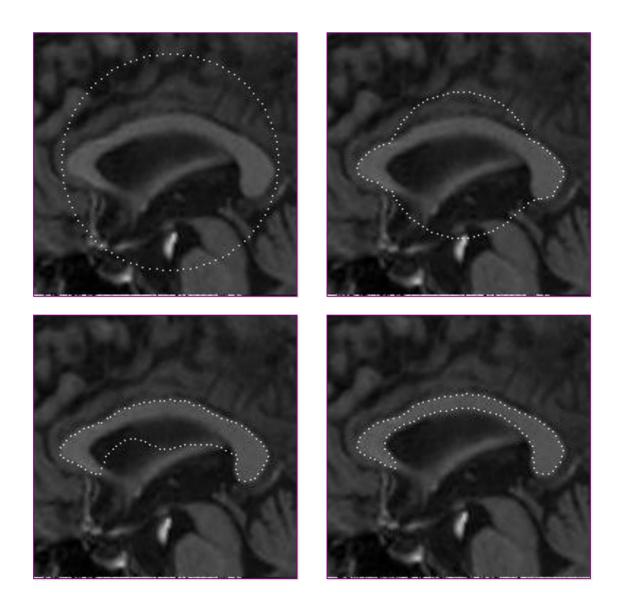
Active contours

Active contours (or snakes) segmentation. Actually, a level set.



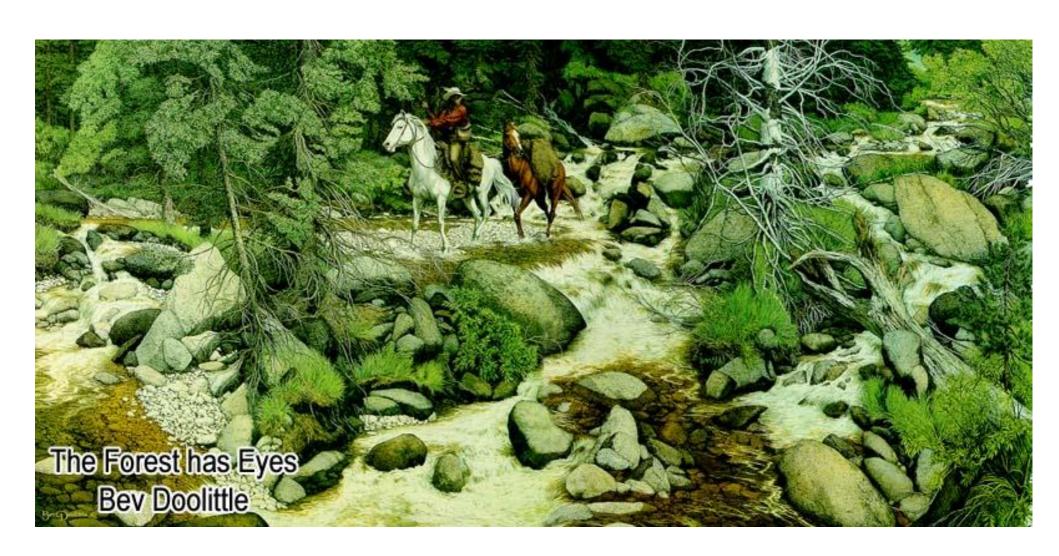


Segmentation of the corpus callosum (brain MRI)



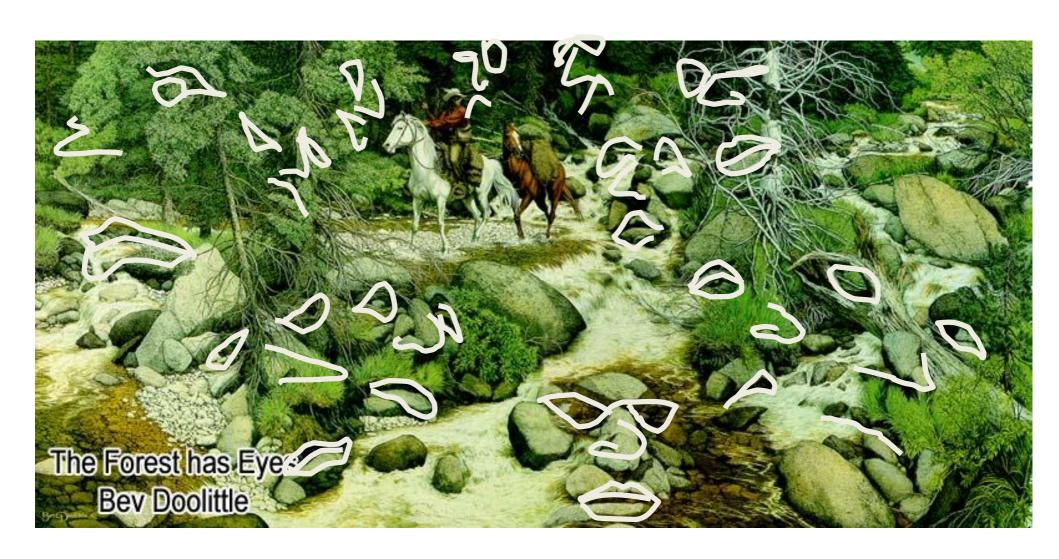














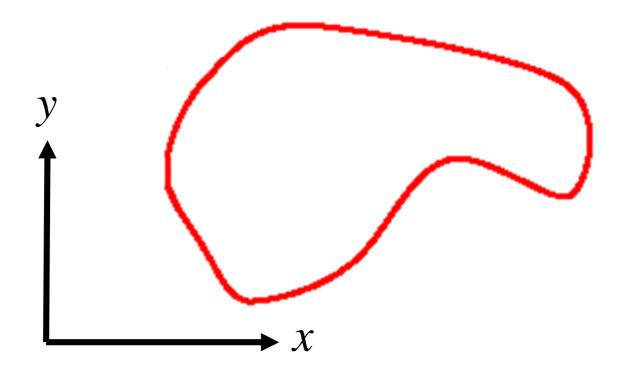


- It is an approach based on "energies"
- The objective is to find a parameterised curve that minimises the weighted sum of internal energy and external energy.
 - Internal forces: hold the curve together (elasticity forces) and keep it from bending too much (bending forces).
 - External energy (potential energy) attract the curve towards the desired object boundaries.
- Parametric curves are initialised within the image domain and forced to move toward the energy minima.





 Curves are a generalisation of functions (functions have a single y for each x).

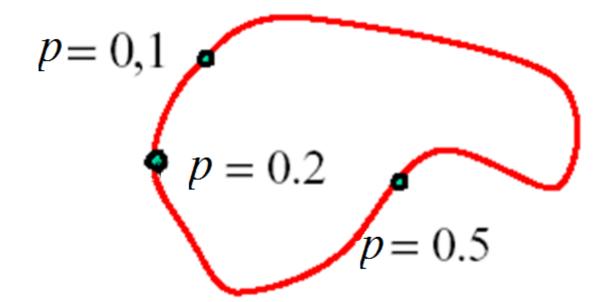






Curves. Parametric representation:

$$C(p) = (x(p), y(p)) \qquad 0 \le s \le 1$$

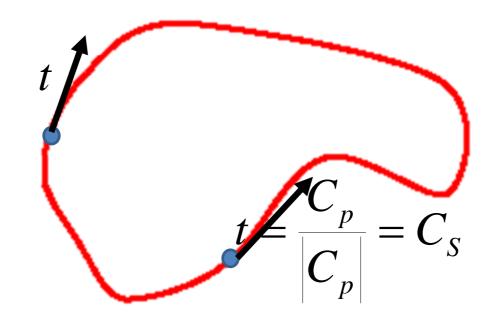


• Closed contour: C(0) = C(1)





- The tangent is the derivative of the curve in that point.
- We normalise it to unit length.

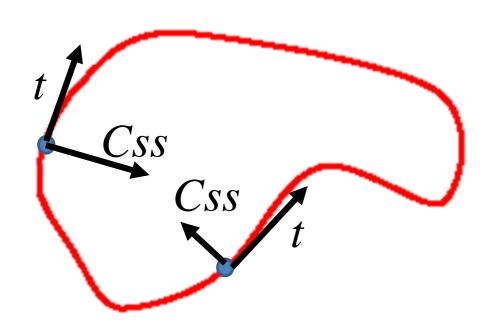


$$C_{p} = \frac{\partial C}{\partial p} = \left(\frac{\partial x}{\partial p}, \frac{\partial y}{\partial p}\right) = \left(x_{p}, y_{p}\right)$$





- The normal is orthogonal to the tangent.
- The derivative of the tangent is orthogonal.



$$egin{aligned} ig|C_Sig|&=1 \ ig\langle C_S,C_Sig
angle &=1 \ rac{\partial}{\partial s}ig\langle C_S,C_Sig
angle &=rac{\partial}{\partial s}1 \ 2ig\langle C_S,C_{SS}ig
angle &=0 \ C_S\perp C_{SS} \end{aligned}$$





 The curvature relates the derivative of the tangent with the normal:

$$C_{SS} = \kappa N$$

- The curvature of a straight line is 0 (tangent do not modify).
- The curvature of a circle is constant (actually, for a circle of radius r , $\kappa = \frac{1}{r}$
- As sharper the boundary, larger the curvature

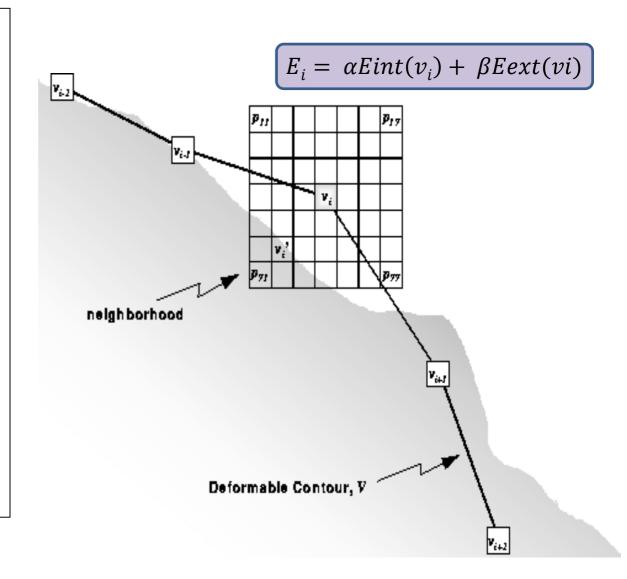




Active contour: compute the energy at each point of the curve

Each point v_i is moved to the point v_i' corresponding to the location of the minimum value in E_i .

If the energy functions are chosen correctly, the contour V should approach and stop at the object boundary.







Active contours: the energies

Original formulation (Kass, Witkin, Terzopoulos, 1987):

$$E[(C)(p)] = \alpha \int_0^1 E_{int}(C(p)) dp + \beta \int_0^1 E_{img}(C(p)) dp + \gamma \int_0^1 E_{con}(C(p)) dp$$

- Internal forces: hold the curve together (elasticity forces) and keep it from bending too much (bending forces).
- External energy (potential energy) attract the curve towards the desired object boundaries.
- The last term is optional, and can be used to account for user-defined constraints, or prior knowledge on the structure to be recovered

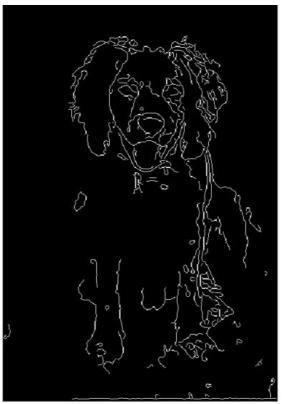




Active contours: internal energy

• In principle, we want to favor smooth shapes, contours with low curvature, contours similar to a known shape, etc. to balance what is actually observed (i.e., in the gradient image).





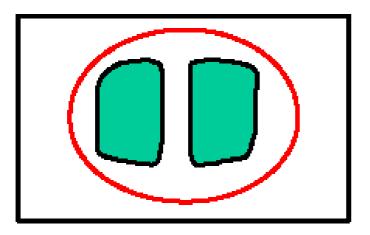


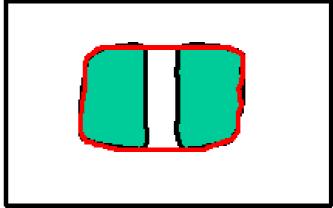




Active contours: internal energy

- There are many different approaches according the ways the energies are defined.
- Internal energies define the "bending" energy:









Active contours: internal energy

Internal energies usually have two terms:

$$E_{internal}(v(s)) = \alpha \frac{\left| \frac{dv}{ds} \right|^2}{\left| \frac{d^2v}{ds} \right|^2} + \beta \frac{\left| \frac{d^2v}{ds} \right|^2}{\left| \frac{d^2v}{ds} \right|^2}$$
The more the curve bends \Rightarrow the larger this energy value is.

The weights α and β dictate how much influence each.

The weights α and β dictate how much influence each component has.

Internal energy for whole curve:

$$E_{internal} = \int_{0}^{1} E_{internal}(v(s)) ds$$



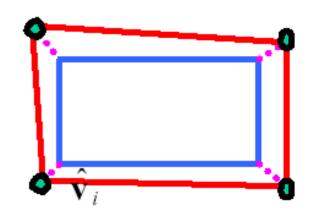


Active contours: adding priors

 If the object to segment is some smooth variation on a known shape, we can use a term that will penalize deviation from that shape:

$$E_{con} = \alpha \cdot \sum_{i=0}^{n-1} (\nu_i - \hat{\nu}_i)^2$$

where $\{\hat{V}_i\}$ are the points of the known shape.







Active contours: external energy

- The external energy measures how well the curve matches the image data.
- We want to "attract" the curve toward the image features: an energy expression proportional to the gradient magnitude will attract the contour to any edge.

$$\begin{split} E_{external}(v(s)) &= -\nabla (I(v(s))^2) = -(|G_x(v(s))|^2 + |G_y(v(s))|^2) \\ E_{external}(v(s)) &= -\nabla (G_\sigma(v(s)) * I(v(s))) \\ E_{external}(v(s)) &= \frac{1}{1 + \nabla (G_\sigma * I)} \end{split}$$

External energy for the whole curve:

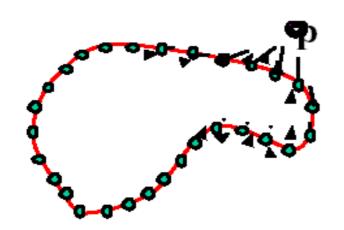


$$E_{external} = \int_{0}^{1} E_{external}(v(s)) ds$$



Active contours: interactive energy

- The energy function can be altered online based on user input (interactive segmentation).
- The cursor allows to push/pull the snake away from a point.

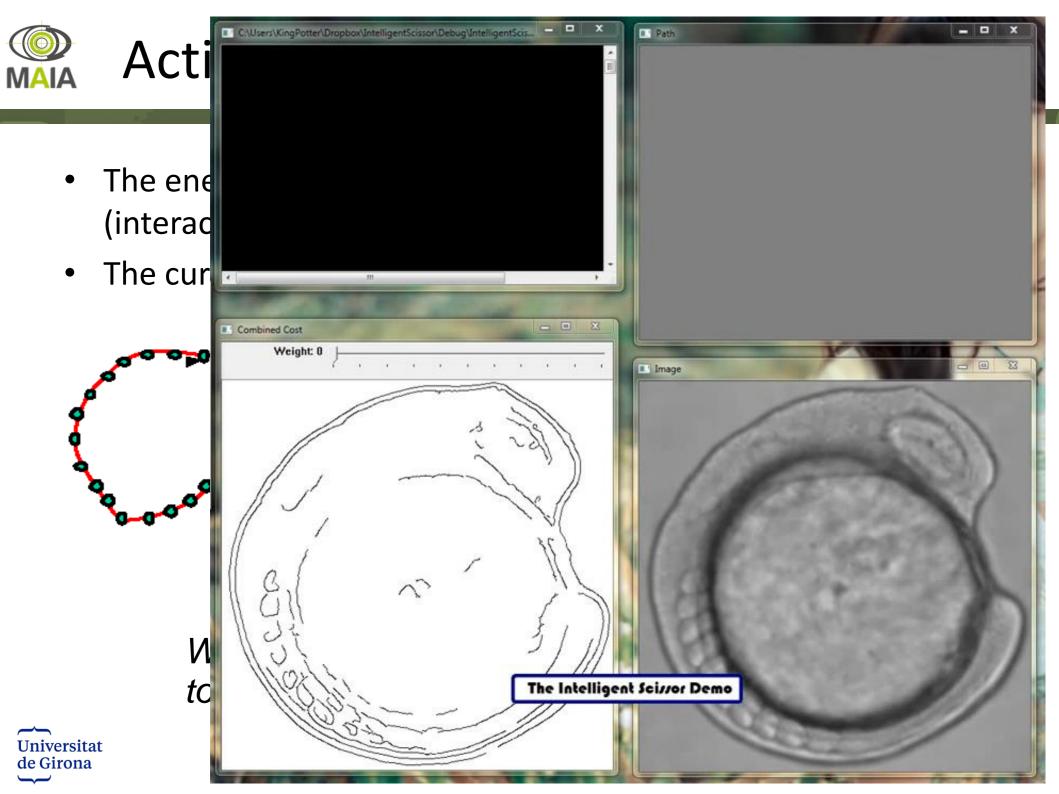


$$E_{push} = + \sum_{i=0}^{n-1} \frac{r^2}{|\nu_i - p|^2}$$

Nearby points get pushed hardest

What expression could we use to pull points towards the cursor position?







Active contours: energy minimisation

$$E[(C)(p)] = \alpha \int_0^1 E_{int}(C(p)) dp + \beta \int_0^1 E_{img}(C(p)) dp + \gamma \int_0^1 E_{con}(C(p)) dp$$

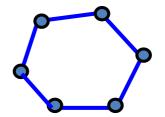
- There are different ways to minimise the energy:
 - Greedy search
 - Dynamic programming
 - Gradient descent



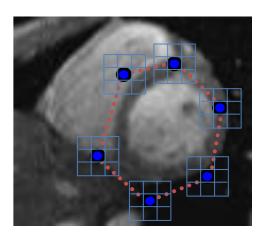


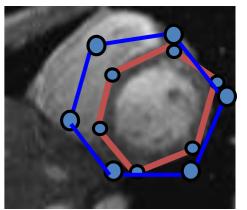
Active contours: summary

- A simple elastic snake is defined by
 - A set of n points,
 - An internal elastic energy term
 - An external edge based energy term



- To use this to locate the outline of an object
 - Initialize in the vicinity of the object
 - Modify the points to minimize the total energy: for each point, search window around it and move to where energy function is minimal
 - Stop when predefined number of points have not changed in last iteration, or after max number of iterations









Pros

- Low complexity.
- Easy to introduce prior knowledge.
- Can account for open as well as closed structures.
- A well established technique, numerous publications it works.
- User Interactivity.

Cons

- Selection on the parameter space and the sampling rule affects the final segmentation result.
- Estimation of the internal geometric properties of the curve in particular higher order derivatives.
- Quite sensitive to the initial conditions.
- Changes of topology.





- Let's implement two examples:
 - Follow the tutorial of Cris Luengo: http://www.crisluengo.net/index.php/archives/217
 - Follow the procedure of Emmanuel Trucco:
 (pdf in the moodle webpage)





To know more...

The origins of snakes is:

Kass, M.; <u>Witkin, A.</u>; <u>Terzopoulos, D.</u> (1988). <u>"Snakes: Active contour models"</u> (PDF). International Journal of Computer Vision.

Many improvements:

- Laurent D. Cohen, On active contour models and balloons,
 CVGIP: Image Understanding, Volume 53, Issue 2, March 1991
- Chenyang Xu and Jerry L. Prince. "Snakes, Shape, and Gradient Vector Flow". IEEE Transactions on Image Processing, vol. 7, no. 3, pp. 359-368, March1998
- Cremers, D.; Schnorr, C.; Weickert, J. "Diffusion-snakes: combining statistical shape knowledge and image information in a variational framework". Proceedings. IEEE Workshop on. 50: 137–144.
- Blake & Isard, Active contours.





To know more...

- The origins of level sets are:
 - Osher, S.; Sethian, J. A. (1988), "Fronts propagating with curvature-dependent speed: Algorithms based on Hamilton– Jacobi formulations"
 - V. Caselles, R. Kimmel, G. Sapiro, Geodesic Active Contours,

