

*Computer Vision*  
*CAP 5416*  
*Fall 2005*  
*Term Project Report*

## Implementation of B-snakes

### Introduction:

Active contours or Snakes are used heavily in computer vision for boundary delineation or edge detection. A **snake** is defined as an energy minimizing spline - whose energy depends on its shape and location within the image. Shape of the snake is controlled by the internal forces (the implicit model) and external forces (image data). Internal force acts as smoothing constraint for the snake and the external force guide the snake towards the features in the image.

Let  $v(s) = (x(s), y(s))$  is the parametric representation of the snake where the value of  $s$  is within the range  $(0,1)$ . Then total energy of the snake can be represented as –

$$\begin{aligned} E_{\text{snake}} &= \int_0^1 E_{\text{snake}}(\mathbf{v}(s)) ds \\ &= \int_0^1 E_{\text{int}}(\mathbf{v}(s)) + E_{\text{ext}}(\mathbf{v}(s)) \end{aligned}$$

Where  $E_{\text{int}}$  is the internal energy of the snake and  $E_{\text{ext}}$  is the external energy of the snake. Here the objective is to find such  $v(s)$  so that the total energy of the snake is minimized. Minimization of the above equations involves calculus of variation and solving Euler equation. For discrete case we get the following equations –

$$\mathbf{Ax} + \mathbf{f}_x(\mathbf{x}, \mathbf{y}) = 0$$

$$\mathbf{Ay} + \mathbf{f}_y(\mathbf{x}, \mathbf{y}) = 0$$

In order to evolve the snake, the expression of the snake can be given as a function time by –

$$\mathbf{x}_t = (\mathbf{A} + \gamma\mathbf{I})^{-1}(\gamma\mathbf{x}_{t-1} - \mathbf{f}_x(x_{t-1}, y_{t-1}))$$

$$\mathbf{y}_t = (\mathbf{A} + \gamma\mathbf{I})^{-1}(\gamma\mathbf{y}_{t-1} - \mathbf{f}_y(x_{t-1}, y_{t-1}))$$

In this project I have implemented B-spline representation of the snake. This B-spline representation of snake is called B-snake. B-snake is represented by a linear combination of B-spline basis functions and a set of control vertices  $V_i = (X_i, Y_i)$ .

B-snake representation is compact, exhibits local control, and it is possible to include corners by allowing multiple knots. I have used uniform cubic spline basis functions to represent the B-snake.

Formulation of the problem and the algorithm:

Now I will describe how B-snake is implemented in this project. This involves two steps which are given below -

Step1: initialization of the B-snake

We assume that there are (p+1) points on the snake curve and we use (m+1) control points to denote the snake. Here  $m < p$ . From the given (p+1) points we want to find an approximate B-spline such that the distance between the curve and the spline approximation is minimized. This distance equation is given by the following equation -

$$R = \sum_{j=0}^p |Q(u_j) - P_j|^2 = \sum_{j=0}^p ((x(u_j) - x_j)^2 + (y(u_j) - y_j)^2)$$

As the above equation is quadratic equation, the minimum value of R will at points  $X_i$  and  $Y_i$  such that the following derivatives are zero.

$$\begin{cases} \frac{\partial R}{\partial X_i} = 0 \\ \frac{\partial R}{\partial Y_i} = 0 \end{cases}$$

By solving the derivative we get the following system of linear equations in  $X_i$  and  $Y_i$  where  $x$  is from 0 to 1.

$$\begin{aligned} \sum_{i=0}^m X_i \sum_{j=0}^p B_i(u_j) B_l(u_j) &= \sum_{j=0}^p x_j B_l(u_j) \\ \sum_{i=0}^m Y_i \sum_{j=0}^p B_i(u_j) B_l(u_j) &= \sum_{j=0}^p y_j B_l(u_j) \end{aligned}$$

We solve the above system of linear equation and get the initial B-snake.

Step 2: - Minimization of the snake energy

In this step we minimize the total energy of the spline curve. Total energy of the spline curve is given by ---

$$E = \sum_{j=0}^p \left\{ \frac{1}{2} \alpha(u_j) \left[ \left( \sum_{i=0}^m X_i B_i'(u_j) \right)^2 + \left( \sum_{i=0}^m Y_i B_i'(u_j) \right)^2 \right] \right. \\ \left. + \frac{1}{2} \beta(u_j) \left[ \left( \sum_{i=0}^m X_i B_i''(u_j) \right)^2 + \left( \sum_{i=0}^m Y_i B_i''(u_j) \right)^2 \right] \right. \\ \left. + F(v(u_j)) \right\}$$

Here we want to find such control points which minimize the energy of the curve by satisfying -

$$\forall i \in \{0, \dots, m\} \begin{cases} \frac{\partial E}{\partial X_i} = 0 \\ \frac{\partial E}{\partial Y_i} = 0 \end{cases}$$

This yields the following system of linear equations for  $X_i$  s-

$$\sum_{i=0}^m X_i \left[ \sum_{j=0}^p \alpha(u_j) B_i'(u_j) B_i'(u_j) + \sum_{j=0}^p \beta(u_j) B_i''(u_j) B_i''(u_j) \right] + \sum_{j=0}^p B_i(u_j) F_x(v(u_j)) = 0$$

This system of linear equation can be solved iteratively by the way similar to the original snake algorithms.

$$\begin{cases} X_{t+1} = (A_b + \gamma I)^{-1} (\gamma X_t - G_x(x_t, y_t)) \\ Y_{t+1} = (A_b + \gamma I)^{-1} (\gamma Y_t - G_y(x_t, y_t)) \end{cases}$$

Thus we get a new set of control points. By that minimize the total energy of the B-spline approximation of the snake. When the total energy is minimized we get our final snake. We stop the iteration when the total energy of the snake at two successive steps is same.

### Implementation:

I have implemented the project in MATLAB. The program at first take inputs from user. The inputs include name of the image file, the parameters alpha, beta, gamma and number of control points to be used to approximate the snake. Then it initializes the snake by forming a B-spline approximation of the snake curve. After that the control points are evolved and the snake finally goes to its desired location where its energy is minimum.

### Results:

Output of my program justifies that the B-spline snake deforms to find the boundary in the image and evolves. Figures 1 to 6 in the next page show how B-snake finds the image boundary in the images.

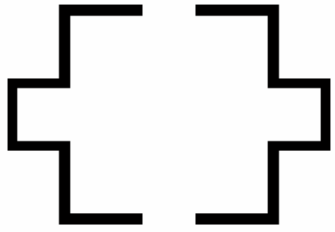


Fig1. Image room.pgm.

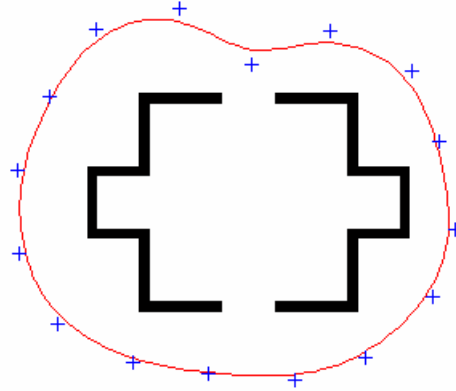


Fig2. Initial B-snake with control points

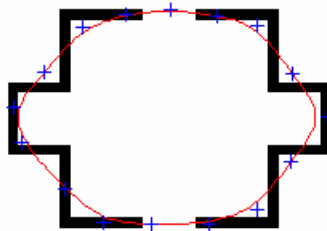


Fig 3. Final B-spline of minimum energy



Fig 4. Image U64.pgm

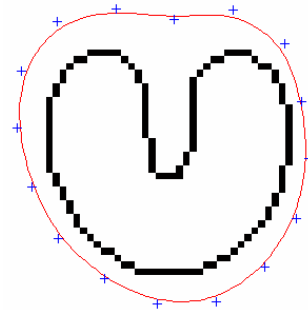


Fig 5. Initial B-snake with control points

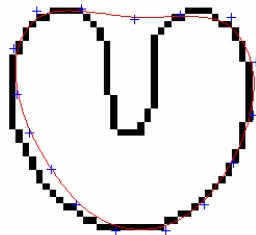


Fig 6. Final B-snake of minimum energy

### Discussion:

For B-spline snake we use the control points only to evolve the snake. That significantly reduces the computation time of calculating the stiffness matrix. Numbers of control points that determine the shape of the snake have influence of the performance of the algorithm. If we use few points then the snake does not evolve well. Here the step size gamma is large with respect to the value used in normal snake algorithm. Moreover B-snake converges quickly then the traditional snake algorithm. By using non-uniform B-spline and allowing multiple knots, we can create the snake which can find corners in the image.

### Conclusion

B snake is a good approximation of the traditional snake. B snake represent the snake compactly, allow corners, and converge fast. Thus it can be applied instead of snake where we need efficient implementation.

### Reference:

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