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# A Depth-Based Algorithm for Manipulating Deformable Objects Using Smooth Parametric Surfaces and Energy Minimisation

David Navarro-Alarcon and Omar Zahra

**Abstract**—In this brief work, we present a new method for controlling deformations of soft objects by using parametric surfaces as a new type of deformation feedback features. This new approach allows us to actively deform objects into complex 3D shapes. A kinematic-based motion controller is derived using an energy minimisation strategy.

## I. THE METHOD

Consider a set of  $R$  robotic manipulators (or fingers) that rigidly grasp a deformable object  $O$ . We denote the 3D position of the  $i$ th robot (measured at the time instant  $t$ ) with the vector  $\mathbf{r}_t^i$ . The total position vector of the  $R$  robots is denoted by  $\mathbf{r}_t = [\mathbf{r}_t^1, \dots, \mathbf{r}_t^R]^\top \in \mathbb{R}^{3R}$ . To measure the object's deformations, we use a static RGB-D sensor (e.g. Kinect) that continuously observes the manipulated object. Let us denote by  $\mathbf{p}$  a 3D point located over the portion of object's surface that can be observed by the image sensor. It is assumed that this surface is smooth and twice differentiable, therefore, its 3D coordinates can be parameterised as  $\mathbf{p} = [x, y, f(x, y)]$ . However, since the object's deformation model is unknown, the analytical structure of  $z = f(x, y)$  is also not known.

The object's feedback surface can be approximated with the following parametric function:

$$f(x, y) = o + \sum_{j=0}^{M-1} \sum_{i=0}^{N-1} [a^{i,j} \quad b^{i,j} \quad c^{i,j} \quad d^{i,j}].$$

$$[\cos(ix\pi/l_x) \quad \sin(ix\pi/l_x) \quad \cos(jy\pi/l_y) \quad \sin(jy\pi/l_y)] = \mathbf{g}(x, y) \cdot \mathbf{s} \quad (1)$$

where  $N, M > 0$  is the number of harmonics taken into account,  $\mathbf{s} = [a^{i,j}, b^{i,j}, c^{i,j}, d^{i,j}, o]^\top$  represent the vector of constant surface parameters, and  $l_x, l_y > 0$  are scaling factors. Note that the vectorial function  $\mathbf{g}(x, y)$  is known and can be constructed with feedback data.

To approximate the unknown parameters  $\mathbf{s}$ , we first collect  $L$  data points of  $\mathbf{p}^k = [x^k, y^k, z^k]$  at various locations over the surface, and then construct the long vector and matrix

$$\vec{\mathbf{z}} = [z^1 \quad \dots \quad z^L]^\top \quad (2)$$

$$\vec{\mathbf{G}} = [\mathbf{g}(x^1, y^1)^\top \quad \dots \quad \mathbf{g}(x^L, y^L)^\top]^\top \quad (3)$$

With this data, the vector  $\mathbf{s}_t$  of surface parameters can be computed at every time instant  $t$  as follows:

$$\mathbf{s}_t = \left( \vec{\mathbf{G}}^\top \vec{\mathbf{G}} \right)^{-1} \vec{\mathbf{G}}^\top \vec{\mathbf{z}} \quad (4)$$

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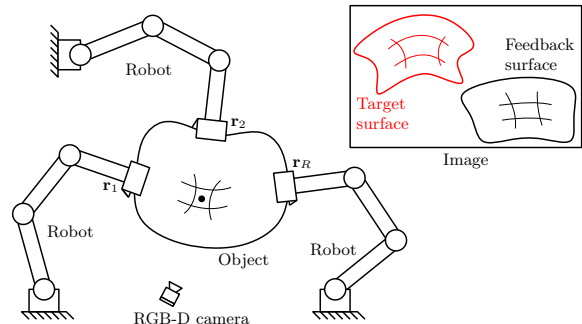


Fig. 1. The robotic system manipulating a deformable object.

The vector  $\mathbf{s}_t$  can be used to characterise the object's feedback surface. To manipulate/deform the object into a target configuration, we must find the relation between the (input) motion of the robots and the (output) surface parameters [1], [2]. We approximate this unknown model with the following linearly parametrisable expression:

$$\mathbf{s}_{t+1} = \mathbf{s}_t + \mathbf{A}_t \mathbf{u}_t \quad (5)$$

where the mapping matrix  $\mathbf{A}_t$  represents the deformation properties of the object and the vector  $\mathbf{u}_t = \mathbf{r}_t - \mathbf{r}_{t-1}$  stands for the controllable motions of the robots. This unknown deformation matrix  $\mathbf{A}_t$  can be computed online by using the method described in [3].

Given a desired surface  $f_d(x, y)$  described by the parameters  $\mathbf{s}^*$ , a controller is derived by minimising the following quadratic cost function:

$$\mathcal{P} = \langle \mathbf{s}_{t+1} - \mathbf{s}^*, \mathbf{s}_{t+1} - \mathbf{s}^* \rangle_{\mathbf{H}} + \langle \mathbf{u}_t, \mathbf{u}_t \rangle_{\mathbf{Q}} \quad (6)$$

for  $\mathbf{H} = \mathbf{H}^\top > 0$  and  $\mathbf{Q} = \mathbf{Q}^\top > 0$  as weight matrices. Computing the extremum  $\nabla \mathcal{P}(\mathbf{u}_t) = \mathbf{0}$  and solving for  $\mathbf{u}_t$  yields the motion control input:

$$\mathbf{u}_t = -(\mathbf{A}_t^\top \mathbf{H} \mathbf{A}_t + \mathbf{Q})^{-1} \mathbf{A}_t^\top \mathbf{H} (\mathbf{s}_t - \mathbf{s}^*) \quad (7)$$

that iteratively minimises  $\mathcal{P}$ . Note, however, that this method only guarantees local minimisation of the cost function.

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