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June 4, 2015

#### Abstract

# 1 Introduction

In this report, we address the problem of variable stiffness control for robust grasping. As shown in our previous work (IROS2014), variable stiffness control can effectively stabilize the grasp under disturbance. However, the stability of such controller has never been addressed before. In this work, we are trying to address this problem using a two-fingered test examples, see Fig. 1. To this end, first, a detailed formulation for the dynamics of the hand-object is derived, which takes rolling constraints and the soft fingertip into account. Then, a variable stiffness controller is presented to grasp the object stable and the upper bound for the change rate of the stiffness is derived from the proof of stability. (We follow the same notation as in Arimoto's book, add ref.)

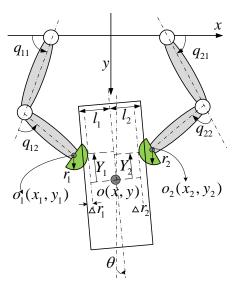


Figure 1: The hand-object system in 2D. Each finger has 2 DOFs with soft fingertips.

# 2 Dynamics

In this part, we first consider the contact model of the fingertips and the constraints involved in the hand-object system. Then the overall dynamics of the system is formulated.

## 2.1 Contact model of soft fingertip

$$f = c_1 (\Delta r)^2 + c_2 \frac{d}{dt} \Delta r \tag{1}$$

where  $c_1$  and  $c_2$  are positive constant parameters which depends on the material of the fingertip.

### 2.2 Contact constraints

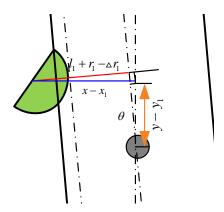


Figure 2: The constraint that the fingertip should keep contact with the object surface.

The fingertip should keep contact with the object surface, as shown in Fig. 2, which can be expressed as follows:

$$l_1 + r_1 - \Delta r_1 = (x - x_1)\cos\theta - (y - y_1)\sin\theta \tag{2}$$

$$l_2 + r_2 - \Delta r_2 = -(x - x_2)\cos\theta + (y - y_2)\sin\theta \tag{3}$$

## 2.3 Rolling constraints

The rolling constraints on each fingertip can be represented as:

$$(r_i - \Delta r_i)\frac{d}{dt}\phi_i = -\frac{d}{dt}Y_i, \quad i = 1, 2$$
(4)

where  $Y_i$  and  $\phi_i$  are given by:

$$Y_i = (x_i - x)\sin\theta + (y_i - y)\cos\theta \tag{5}$$

$$q_{11} + q_{12} + \phi_1 = \pi + \theta \tag{6}$$

$$q_{21} + q_{22} + \phi_2 = \pi - \theta \tag{7}$$

#### 2.4 Overall Dynamics

The total kinetic energy for the overall system can be described as follows:

$$K = \sum_{i=1,2} \frac{1}{2} \dot{\mathbf{q}}_i^T H_i \dot{\mathbf{q}}_i + \frac{1}{2} M(\dot{x}^2 + \dot{y}^2) + \frac{1}{2} I \dot{\theta}^2$$
 (8)

where  $\mathbf{q}_{i} = [q_{i1}, q_{i2}]^{T}$ .

The total potential energy can be given as:

$$P = \sum_{i=1,2} \int_0^{\Delta r_i} c_1 \eta^2 d\eta$$
 (9)

Then from the Hamilton's principle, we have:

$$\int_{t_0}^{t_1} \left[ \delta(K - P) - \sum_{i=1,2} c_2 \Delta \dot{r}_i \frac{\partial \Delta r_i}{\partial X^T} \delta X + \sum_{i=1,2} \lambda_i [(r_i - \Delta r_i) \frac{\partial \phi_i}{\partial X^T} + \frac{\partial Y_i}{\partial X^T}] \delta X + \sum_{i=1,2} \mathbf{u_i}^T \delta \mathbf{q}_i \right] dt = 0$$
(10)

where  $X = [\mathbf{q}_1^T, \mathbf{q}_2^T, x, y, \theta]^T$ . Then we have the overall dynamics for the object-hand system as follows:

$$H_{i}(\mathbf{q_{i}})\ddot{\mathbf{q}}_{i} + (\frac{1}{2}\dot{H}_{i} + S_{i})\dot{\mathbf{q}}_{i} + f_{i}\frac{\partial\Delta r_{i}}{\partial\mathbf{q}_{i}^{T}} - \lambda_{i}[(r_{i} - \Delta r_{i})\frac{\partial\phi_{i}}{\partial\mathbf{q}_{i}^{T}} + \frac{\partial Y_{i}}{\partial\mathbf{q}_{i}^{T}}] = \mathbf{u}_{i}$$
 (11)

$$M\ddot{x} + \sum_{i=1,2} \left[ f_i \frac{\partial \Delta r_i}{\partial x} - \lambda_i \frac{\partial Y_i}{\partial x} \right] = 0 \tag{12}$$

$$M\ddot{y} + \sum_{i=1,2} \left[ f_i \frac{\partial \Delta r_i}{\partial y} - \lambda_i \frac{\partial Y_i}{\partial y} \right] = 0 \tag{13}$$

$$I\ddot{\theta} + \sum_{i=1,2} \left[ f_i \frac{\partial \Delta r_i}{\partial \theta} - \lambda_i ((r_i - \Delta r_i) \frac{\partial \phi_i}{\partial \theta} + \frac{\partial Y_i}{\partial \theta}) \right] = 0$$
 (14)

With the identities in appendix A, the overall dynamics can be simplified

$$H_{i}(\mathbf{q_{i}})\ddot{\mathbf{q}}_{i} + \left(\frac{1}{2}\dot{H}_{i} + S_{i}\right)\dot{\mathbf{q}}_{i} - (-1)^{i}f_{i}J_{i}^{T}\begin{bmatrix}\cos\theta\\-\sin\theta\end{bmatrix} - \lambda_{i}[(r_{i} - \Delta r_{i})\begin{bmatrix}-1\\-1\end{bmatrix} + J_{i}^{T}\begin{bmatrix}\sin\theta\\\cos\theta\end{bmatrix} = \mathbf{u}_{i}$$
(15)

$$M\ddot{x} - (f_1 - f_2)\cos\theta + (\lambda_1 + \lambda_2)\sin\theta = 0 \tag{16}$$

$$M\ddot{y} + (f_1 - f_2)\sin\theta + (\lambda_1 + \lambda_2)\cos\theta = 0 \tag{17}$$

$$I\ddot{\theta} - f_1 Y_1 + f_2 Y_2 + l_1 \lambda_1 - l_2 \lambda_2 = 0 \tag{18}$$

## 3 Variable stiffness control

Motivated by the analysis of the overall system dynamics, the following control law is adopted for each finger to achieve stable grasp:

$$u_i = -D_i \dot{\mathbf{q}}_i + k J_i^T (\mathbf{x}_i - \mathbf{x}_d) \tag{19}$$

$$\mathbf{x}_{i} = \begin{bmatrix} x_{i} \\ y_{i} \end{bmatrix} \qquad \mathbf{x}_{d} = \frac{1}{2} \begin{bmatrix} x_{1} + x_{2} \\ y_{1} + y_{2} \end{bmatrix}$$
 (20)

where  $D_i$  is a diagonal positive definite matrix representing the damping gain.  $k \in \mathbb{R}^+$  represents the *variable* stiffness for each fingertip(k is the same value for the two-finger grasp to ensure force balance).

# 4 Stability Proof–1

Taking the sum of inner product of (15) with  $\dot{\mathbf{q}}_i$ , i = 1, 2, (16) with  $\dot{x}$ , (17) with  $\dot{y}$ , (18) with  $\dot{\theta}$ , we have:

$$\frac{d}{dt}E = -\sum_{i=1,2} \{\dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2\} + \frac{\dot{k}}{4} (\mathbf{x}_1 - \mathbf{x}_2)^T (\mathbf{x}_1 - \mathbf{x}_2)$$
(21)

$$E = K + V + P \tag{22}$$

$$K = \sum_{i=1,2} \frac{1}{2} \dot{\mathbf{q}}_i^T H_i \dot{\mathbf{q}}_i + \frac{1}{2} M(\dot{x}^2 + \dot{y}^2) + \frac{1}{2} I \dot{\theta}^2$$
 (23)

$$P = \sum_{i=1,2} \int_0^{\Delta r_i} c_1 \eta^2 d\eta$$
 (24)

$$V = \frac{k}{4} (\mathbf{x}_1 - \mathbf{x}_2)^T (\mathbf{x}_1 - \mathbf{x}_2)$$
(25)

As proved in Arimoto's book, the closed-loop dynamics is asymptotically stable if:

$$\frac{d}{dt}E < 0 \tag{26}$$

which leads to

$$\dot{k} < \frac{\sum\limits_{i=1,2} {\dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2}}{(\mathbf{x}_1 - \mathbf{x}_2)^T (\mathbf{x}_1 - \mathbf{x}_2)}$$

$$(27)$$

From this, we can conclude the following comments Is this conclusion correct? Any example?:

- If the object is softer,  $c_2$  is bigger, we can change the stiffness faster.
- If the object is smaller,  $(\mathbf{x}_1 \mathbf{x}_2)^T(\mathbf{x}_1 \mathbf{x}_2)$ , we can change the stiffness faster

As noticed that the change rate of the stiffness is still depending on the state variable and we still don't know how we realize the desired stiffness. Note: in our previous work of grasp adaptation, we simply change the k to  $k_d$ , which may lead to unstable behavior if  $k_d - k$  is too big. This leads us to the following section

# 5 Stability Proof

The stability proof in previous section set an upper bound for the change rate of stiffness. However, it does not take the control policy for the stiffness into account. In our IROS paper, the desired stiffness is a function of the tactile sensing S and the rest length L, which is encoded using a probabilistic model GMM as follows,

$$k = g(S, L) = \sum_{i=1}^{m} h_i [\mu_{k,i} + \Sigma_{ky,i} \Sigma_{y,i}^{-1} (y - \mu_{y,i})]$$
 (28)

Denote  $y = [L^T, S^T]^T$ , Explain all the notations here later.

To prove eqn. (21), we need to compute the change rate of stiffness, i.e.  $\dot{k}$ ,

$$\dot{k} = \frac{\partial k}{\partial y} \left( \frac{\partial y}{\partial L} \frac{\partial L}{\partial t} + \frac{\partial y}{\partial s} \frac{\partial s}{\partial t} \right) \tag{29}$$

$$= \frac{\partial k}{\partial y} \begin{bmatrix} \mathbf{x}_1^T J_1 \dot{\mathbf{q}}_1 + \mathbf{x}_2^T J_2 \dot{\mathbf{q}}_2 \\ \mathbf{x}_1^T J_1 \dot{\mathbf{q}}_1 + \mathbf{x}_2^T J_2 \dot{\mathbf{q}}_2 \\ \frac{\partial s}{\partial t} 2 \| \mathbf{x}_1 - \mathbf{x}_2 \| \end{bmatrix} \frac{1}{2 \| \mathbf{x}_1 - \mathbf{x}_2 \|}$$
(30)

From the definition of eqn. (28), we have:

$$\frac{\partial k}{\partial y} = \sum_{i=1}^{m} \left\{ \frac{\partial h_i}{\partial y} \left[ \mu_{k,i} + \Sigma_{ky,i} \Sigma_{y,i}^{-1} (y - \mu_{y,i}) \right] + h_i (\Sigma_{ky,i} \Sigma_{y,i}^{-1}) \right\}$$
(31)

Key Assumption: During the change of stiffness, we assume the weight of each gaussian components does not change with respect to y, namely  $\frac{\partial h_i}{\partial y} = 0$ .

We shall view  $h_i$  as the responsibility of the i- gaussian component, but only use the rest length to compute it. Due to the physical constraints of the grasp, during the change of stiffness, the rest length doesn't change (ignoring the effect of fingertip's deformation) and thus it is more reliable to just the rest length for computing the weighting coefficients  $h_i$ . Therefore we have  $\frac{\partial h_i}{\partial u} = 0$ .

With this assumption, we have:

$$\frac{\partial k}{\partial y} = \sum_{i=1}^{m} h_i(\Sigma_{ky,i} \Sigma_{y,i}^{-1})$$
(32)

$$:= [A_{11}, A_{12}, A_{13}] \tag{33}$$

Note that  $A_{11} \in R, A_{12} \in R, A_{13} \in R^{n_s \times 1}$  are constant values or matrix,  $n_s$  is the dimension of tactile reading. To prove  $\frac{d}{dt}E < 0$ , we only need to proove:

$$[A_{11}, A_{12}, A_{13}] * \begin{bmatrix} \mathbf{x}_1^T J_1 \dot{\mathbf{q}}_1 + \mathbf{x}_2^T J_2 \dot{\mathbf{q}}_2 \\ \mathbf{x}_1^T J_1 \dot{\mathbf{q}}_1 + \mathbf{x}_2^T J_2 \dot{\mathbf{q}}_2 \\ \frac{\partial s}{\partial t} * 2 \|\mathbf{x}_1 - \mathbf{x}_2\| \end{bmatrix} \frac{1}{2\|\mathbf{x}_1 - \mathbf{x}_2\|} * \frac{\|\mathbf{x}_1 - \mathbf{x}_2\|^2}{4} - \sum_{i=1,2} \{\dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2\} < 0$$

(34)

(37)

$$\begin{aligned}
&\Leftarrow (A_{11} + A_{12})\mathbf{x_{1}}^{T} J_{1} \dot{\mathbf{q}}_{1} + (A_{11} + A_{12})\mathbf{x_{2}}^{T} J_{2} \dot{\mathbf{q}}_{2} + A_{13} \frac{\partial s}{\partial t} 2 \|\mathbf{x}_{1} - \mathbf{x}_{2}\| < \\
\dot{\mathbf{q}}_{1}^{T} \frac{2D_{1}}{\|\mathbf{x}_{1} - \mathbf{x}_{2}\|} \dot{\mathbf{q}}_{1} + \dot{\mathbf{q}}_{2}^{T} \frac{2D_{2}}{\|\mathbf{x}_{1} - \mathbf{x}_{2}\|} \dot{\mathbf{q}}_{2} + 2 \sum_{i=1,2} \{c_{2} \Delta \dot{r}_{i}^{2}\} / \|\mathbf{x}_{1} - \mathbf{x}_{2}\| \\
&\Leftarrow A_{13} \frac{\partial s}{\partial t} < \dot{\mathbf{q}}_{1}^{T} \frac{D_{1}}{\|\mathbf{x}_{1} - \mathbf{x}_{2}\|^{2}} \dot{\mathbf{q}}_{1} - \frac{A_{11} + A_{12}}{2\|\mathbf{x}_{1} - \mathbf{x}_{2}\|} \mathbf{x}_{1}^{T} J_{1} \dot{\mathbf{q}}_{1} + \\
&\dot{\mathbf{q}}_{2}^{T} \frac{D_{1}}{\|\mathbf{x}_{1} - \mathbf{x}_{2}\|^{2}} \dot{\mathbf{q}}_{2} - \frac{A_{11} + A_{12}}{2\|\mathbf{x}_{1} - \mathbf{x}_{2}\|} \mathbf{x}_{2}^{T} J_{2} \dot{\mathbf{q}}_{2} + \sum_{i=1,2} \{c_{2} \Delta \dot{r}_{i}^{2}\} / \|\mathbf{x}_{1} - \mathbf{x}_{2}\|^{2} \\
&(36)
\end{aligned}$$

$$\Leftarrow A_{13} \frac{\partial s}{\partial t} + (\frac{A_{11} + A_{12}}{4})^2 [\mathbf{x}_1^T J_1 D_1^{-1} J_1^T \mathbf{x}_1 + \mathbf{x}_2^T J_2 D_2^{-1} J_2^T \mathbf{x}_2] < \underbrace{ \| \frac{D_1^{1/2}}{\|\mathbf{x}_1 - \mathbf{x}_2\|} \dot{\mathbf{q}}_1 - \mathbf{b}_1 \|^2 + \| \frac{D_2^{1/2}}{\|\mathbf{x}_1 - \mathbf{x}_2\|} \dot{\mathbf{q}}_2 - \mathbf{b}_2 \|^2 + \sum_{i=1,2} \{c_2 \Delta \dot{r}_i^2\} / \|\mathbf{x}_1 - \mathbf{x}_2\|^2}_{\text{denote this part as a good } \mathbf{x} \geq 0}$$

$$\Leftarrow A_{13} \frac{\partial s}{\partial t} + (\frac{A_{11} + A_{12}}{4})^2 [\mathbf{x}_1^T J_1 D_1^{-1} J_1^T \mathbf{x}_1 + \mathbf{x}_2^T J_2 D_2^{-1} J_2^T \mathbf{x}_2] < \gamma$$
(38)

Note that the first term in the left side of eqn. (38)), i.e.,  $A_{13} \frac{\partial s}{\partial t}$  represents the change of stiffness due to the change of tactile sensing, if it is negative, eqn. (38)) will keep holding and we don't need to change the damping to stabilize the system. When  $A_{13} \frac{\partial s}{\partial t} > 0$ , namely increasing the stiffness according to the tactile sensing, we need to set a proper damping term to stabilize the system. From the bound of Jacobian(Christian ott's book), we have,

$$\lambda_{max}(J_i J_i^T) < \sigma_{max}^2, i = 1, 2 \tag{39}$$

Also, from the physical constraint of the finger,  $\|\mathbf{x}_i\| < L_f$ , i = 1, 2. Assume we use the same damping at each joint,  $D_1 = D_2 = diag\{d\}$ , then we have:

$$d > \frac{2\sigma_{max}^2 L_f^2 (\frac{A_{11} + A_{12}}{4})^2}{\gamma - A_{13} \frac{\partial s}{\partial t}}$$
 (40)

#### Remark:

•  $\gamma$  is a parameter that is determined heuristically. If  $\gamma \uparrow$ , then  $d \downarrow$ . From the definition of  $\gamma$ , it will increase if the softness of fingertip  $c_2$  increases

or the size of object decreases, i.e.,  $\|\mathbf{x}_1 - \mathbf{x}_2\| \downarrow$ . (This intuitively makes sense. If finger is softer, less damping is required.)

- The value  $\frac{\partial s}{\partial t}$  is determined by the sensitivity of the tactile sensor. In practice, most often a filter is used to preprocess the sensory data. The rate of change of tactile sensing  $\frac{\partial s}{\partial t}$  can be directly obtained from the filer and the damping can be set accordingly. Another possibility is to assume a bound for  $||A_{13}\frac{\partial s}{\partial t}|| < \gamma_s$  and then damping d can be set using the maximal value  $\gamma_s$ .
- Note that the bound for d is still very conservative, due to the assumption made on  $\gamma$  and the neglect of relative high friction at each joint of the fingers. (For all the experiments we have done so far, we didn't observe any unstable behavior.)

# 6 Stability Proof–2

This is another possibility to ensure the stability of system by observing the energy dissipation, but it is also conservative and an observer for the joint velocity is required. Intuitively, this method use a virtual tank to adaptively set the rate of change of the stiffness.

As noticed in Eq. 21, the dissipated energy  $\sum_{i=1,2} \{\dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2\}$  set an

upper bound for the energy that we can inject into the system. In order to stably vary the stiffness, we keep track of the energy dissipated by the system using a virtual tank(add ref.). The control signal is designed as follows:

$$u_i = -D_i \dot{\mathbf{q}}_i + J_i^T(\omega z) \tag{41}$$

The new term  $z \in \mathbb{R}$  is the state variable associated with the energy in the tank and,

$$T(z) = \frac{1}{2}z^2\tag{42}$$

$$\dot{z} = \frac{\sigma}{z} \sum_{i=1,2} \{ \dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2 \} - \sum_{i=1,2} \omega^T J_i \dot{\mathbf{q}}_i$$
 (43)

 $\omega(t) \in \mathbb{R}^3$  is the control variable through which it is possible to exchange the energy between the impedance components and the tank.  $\sigma$  is a design parameter that opens/closes the tank for energy exchange:

$$\sigma = \begin{cases} 1 & T(z) < T_{lim} \\ 0 & otherwise \end{cases}$$

Now with the new controller and tank energy, we can derive:

$$E = K + V + T \tag{44}$$

$$\frac{d}{dt}E = -\sum_{i=1,2} \{ \dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2 \} + \sigma \sum_{i=1,2} \{ \dot{\mathbf{q}}_i^T D_i \dot{\mathbf{q}}_i + c_2 \Delta \dot{r}_i^2 \} \le 0$$
 (45)

Recalling Eq. 19, to get the desired stiffness  $k_d$ , we can exploit the energy stored in the tank by setting:

$$\omega(t) = \begin{cases} K_d(\mathbf{x}_i - \mathbf{x}_d)/z & T(z) > T_{low} \\ 0 & otherwise \end{cases}$$

The stability proof is similar by following Arimoto's method from Eq. 45.

## 7 simulation

In this section, we will use an 1D example to example the idea of virtual tank for implementing the varying stiffness.

# 8 appendix

## A Identities

From equation (2)(3)(5)(6)(7), we have the following identities:

$$\begin{split} \frac{\partial \Delta r_1}{\partial \mathbf{q}_1^T} &= J_1^T \left[ \cos \theta, -\sin \theta \right]^T \\ \frac{\partial \phi_1}{\partial \mathbf{q}_1^T} &= \left[ -1, -1 \right]^T \\ \frac{\partial Y_1}{\partial \mathbf{q}_1^T} &= J_1^T \left[ \sin \theta, \cos \theta \right]^T \\ \frac{\partial \Delta r_1}{\partial x} &= J_2^T \left[ \sin \theta, \cos \theta \right]^T \\ \frac{\partial \Delta r_1}{\partial x} &= -\cos \theta \\ \frac{\partial Y_1}{\partial x} &= -\sin \theta \\ \frac{\partial \Delta r_1}{\partial y} &= \sin \theta \\ \frac{\partial Y_1}{\partial y} &= -\cos \theta \\ \frac{\partial \Delta r_1}{\partial \theta} &= -Y_1 \\ \frac{\partial \phi_1}{\partial \theta} &= 1 \\ \frac{\partial Y_2}{\partial \theta} &= \left( x_1 - x \right) \cos \theta - \left( y_1 - y \right) \sin \theta \\ \frac{\partial Y_2}{\partial q_2^T} &= -J_2^T \left[ \cos \theta, -\sin \theta \right]^T \\ \frac{\partial \phi_2}{\partial \mathbf{q}_2^T} &= \left[ -1, -1 \right]^T \\ \frac{\partial Y_2}{\partial \mathbf{q}_2^T} &= J_2^T \left[ \sin \theta, \cos \theta \right]^T \\ \frac{\partial Y_2}{\partial q_2^T} &= \cos \theta \\ \frac{\partial Y_2}{\partial x} &= -\cos \theta \\ \frac{\partial Y_2}{\partial x} &= -\sin \theta \\ \frac{\partial Y_2}{\partial y} &= -\sin \theta \\ \frac{\partial Y_2}{\partial y} &= -\sin \theta \\ \frac{\partial Y_2}{\partial y} &= -\cos \theta \\ \frac{\partial Y_2}{\partial y} &= -\cos \theta \\ \frac{\partial Y_2}{\partial \theta} &= 1 \\ \frac{\partial \phi_2}{\partial \theta} &= -1 \\ \frac{\partial Y_2}{\partial \theta} &= (x_2 - x) \cos \theta - (y_2 - y) \sin \theta \end{split}$$