

# A Working Example of an Adaptive Controller With the Slotine Li Algorithm

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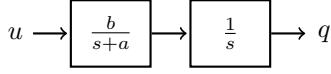


Fig. 2. Simplified servomechanism model

**Abstract**—The abstract goes here.

## I. INTRODUCTION

THE intention of this paper is to provide a simple implementation of the Slotine Li Algorithm to control a robotic manipulator with an adaptive control law.

## II. PRELIMINARIES

### A. Servomechanism Model

The servomechanism used in this experiment consists of a DC motor with a brass disc on it's axis, connected to a power amplifier, so as to have a block diagram representation like the one in the figure 1.

Where  $\beta$  is the gain of the power amplifier because of the armature current feedback loop,  $K_C$  is a pre amplification stage,  $K$  is the current-torque relation,  $Kb$  is related to the counter electromotive force. Let's consider a high gain in the power amplification stage due to the current feedback loop, we can simplify this model to a much more simpler one:

$$\ddot{q}(t) + a\dot{q}(t) = bu(t) \quad (1)$$

Where  $a$  and  $b$  are defined as:

$$a = \frac{B_m}{J_m} \quad (2)$$

$$b = \frac{K}{K_C J_m} \quad (3)$$

We consider  $\beta$  of great magnitude, so that the time constant of the electric servomechanism is smaller than the mechanical one.

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Manuscript delivered April 20, 2015

### B. Parameter Identification

In previous works it has been established that the values of the constants  $a$  and  $b$  are:

$$a = 0.45 \quad (4)$$

$$b = 31.0 \quad (5)$$

So, when presented the results of the estimation of this parameters, they will be compared against this results.

### C. Slotine Li Algorithm for Non-linear Robot Manipulators

The Slotine-Li algorithm for robotic manipulator considers a dynamic equation for the robot like:

$$H(q)\ddot{q} + C(q, \dot{q}) + g(q) = \tau \quad (6)$$

Which can be rewritten like this:

$$\tilde{H}(q)\ddot{q}_r + \tilde{C}(q, \dot{q})\dot{q}_r + \tilde{g}(q) = Y(q, \dot{q}, \ddot{q}_r)\tilde{a} \quad (7)$$

Where  $\hat{a}$  is the parameter estimated through the adaptation law, and  $\hat{H}(q)$  is the matrix  $H(q)$  substituted with the estimated parameter  $\hat{a}$ . More over, we define the error in the estimation of the parameter  $\tilde{a} = a - \hat{a}$  and  $Y$  only depends on dynamic parameters. The error in the estimation of the state of the system is defined as  $\tilde{q}(t) = q(t) - q_d(t)$  and we define a new state as:

$$\dot{q}_r = \dot{q}_d - \Lambda\tilde{q} \quad (8)$$

so that when we use the proposed control law:

$$\tau = \hat{H}(q)\ddot{q}_r + \hat{C}(q, \dot{q})\dot{q}_r + \hat{g}(q) - K_D s \quad (9)$$

with the adaptation law:

$$\dot{\hat{a}} = -\Gamma Y^T s \quad (10)$$

$\Gamma$  being a constant positive definite matrix, and  $s$  being defined as:

$$s = \dot{q} - \dot{q}_r = \dot{\tilde{q}} + \Lambda\tilde{q} \quad (11)$$

the equation of the closed loop system will be as stated in 7.

### D. Trajectories Generation

The algorithm needs not only the signal for the trajectory to track, but the derivative and the second derivative, both of them can be achieved with a second order filter like the one in the figure3.

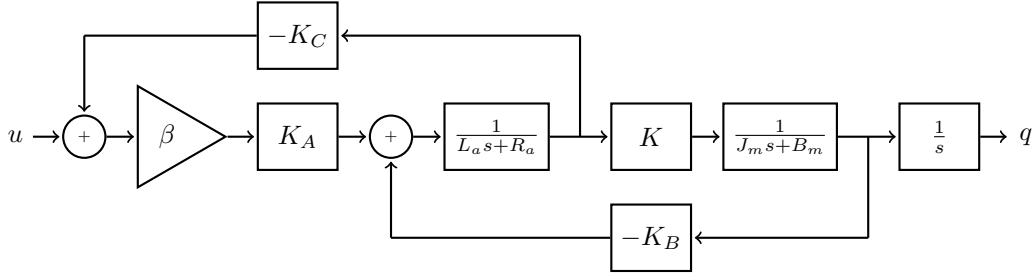


Fig. 1. Servomechanism model

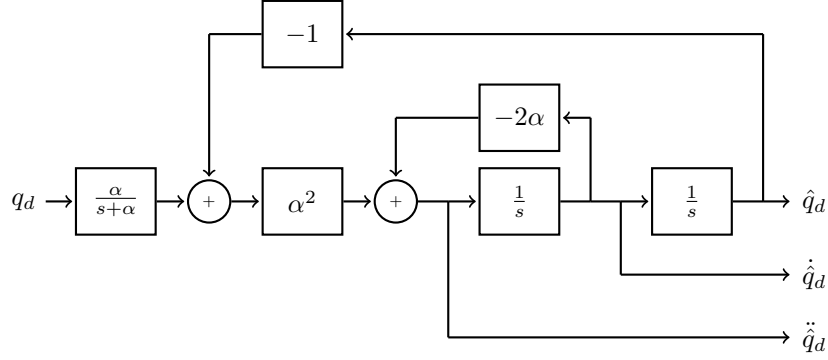


Fig. 3. Second order filter for trajectory generation

### III. CONTROL LAW DERIVATION

Since our model is linear and the control law is designed for a non-linear system, it is necessary to derive a simplification for the control law. Rewriting 1 to look like the non-linear manipulator model we have:

$$\frac{1}{b}\ddot{q} + \frac{a}{b}\dot{q} = u \quad (12)$$

And the proposed control law then translates to:

$$u = \frac{1}{b}\ddot{q}_r + \frac{a}{b}\dot{q}_r - K_D s$$

Where we will define  $\beta = \frac{1}{b}$  and  $\alpha = \frac{a}{b}$  to handle this values easier, so that we will have:

$$u = \beta\ddot{q}_r + \alpha\dot{q}_r - K_D (\dot{q} - \dot{q}_r) \quad (13)$$

And the adaptation law remains:

$$\dot{\hat{a}} = -\Gamma Y^T s$$

Where the vector of parameters to estimate will be:

$$\hat{a} = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \quad (14)$$

The vector of known signals is:

$$Y = \begin{pmatrix} \dot{q}_r \\ \ddot{q}_r \end{pmatrix} \quad (15)$$

The control and adaptation laws are implemented in the figures 4 and 5.

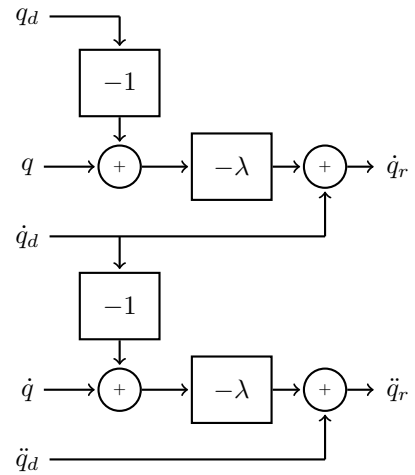


Fig. 4. Slotine-Li algorithm implementation

### IV. EXPERIMENT PARAMETERS

### V. EXPERIMENTAL RESULTS

Subsection text here.

### VI. CONCLUSION

The conclusion goes here.

### APPENDIX A

### PROOF OF THE FIRST ZONKLAR EQUATION

Appendix one text goes here.

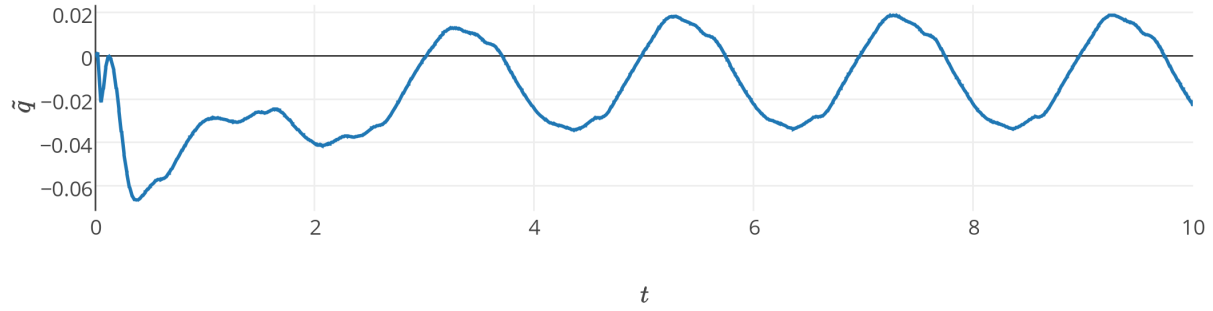


Fig. 7. Error signal

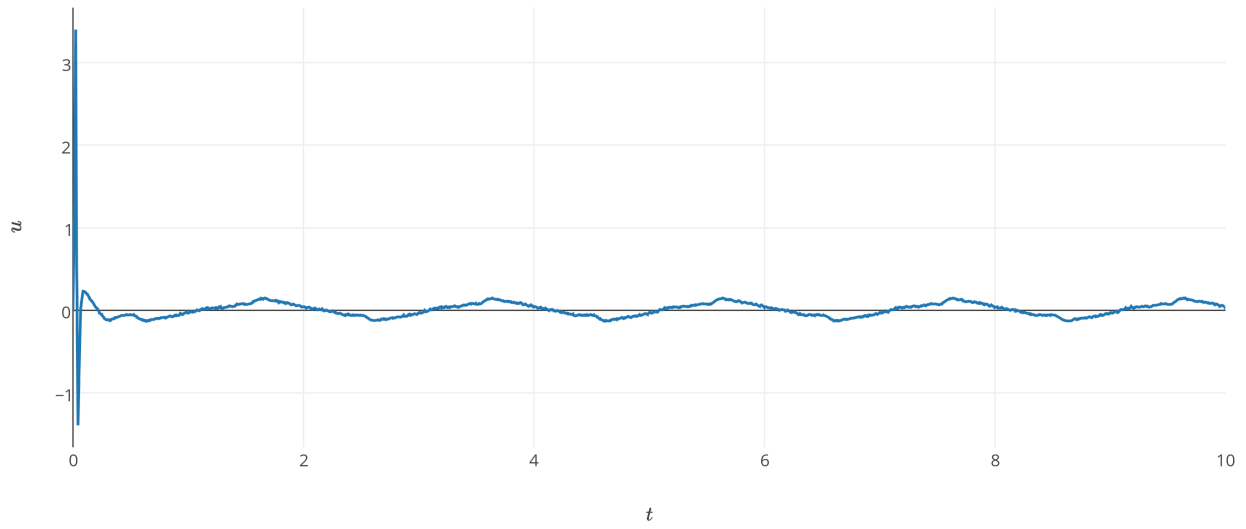


Fig. 9. Control law signal

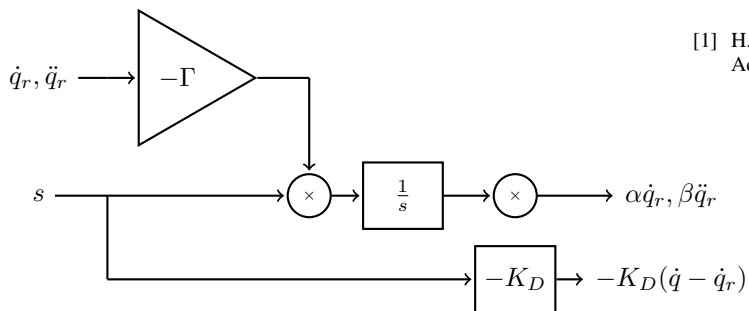


Fig. 5. Adaptation law implementation

## REFERENCES

- [1] H. Kopka and P. W. Daly, *A Guide to L<sup>A</sup>T<sub>E</sub>X*, 3rd ed. Harlow, England: Addison-Wesley, 1999.

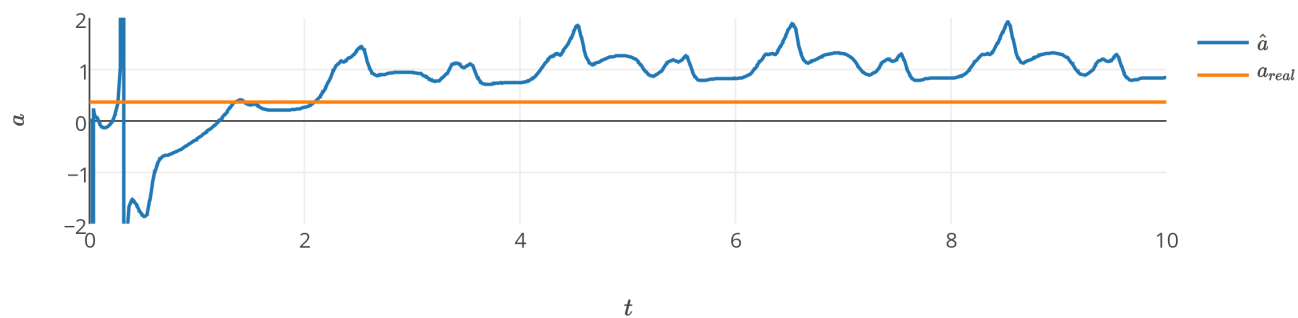


Fig. 11. Estimated signal of parameter a

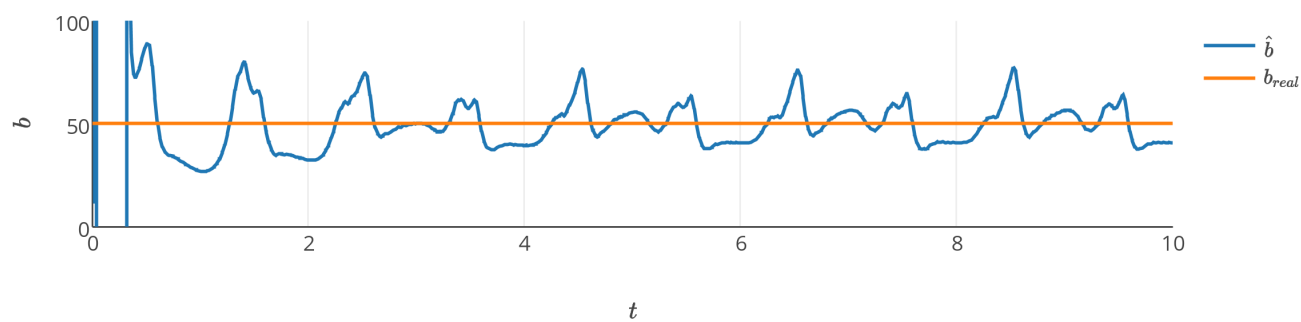


Fig. 12. Estimated signal of parameter b

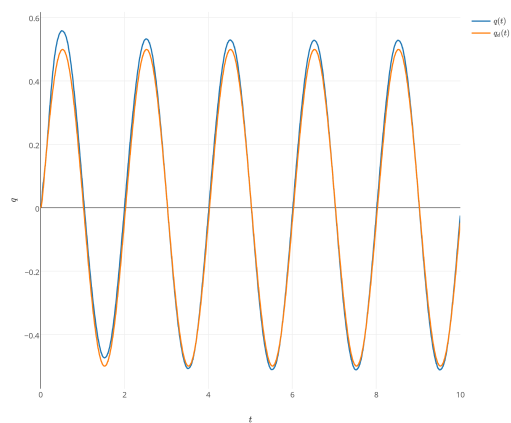


Fig. 6. Servomechanism response against generated signal

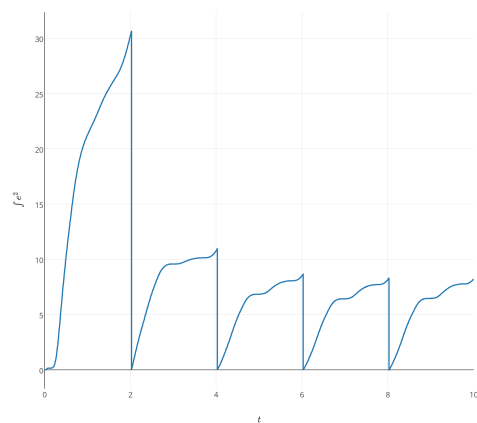


Fig. 8. Integrated square error

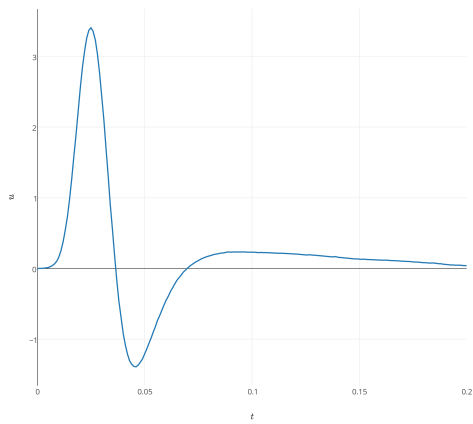


Fig. 10. Control signal first impulse