

# Experimental validation of two semi-implicit homogeneous discretized differentiators on a cable-driven parallel robot

SYNOBS day - 6th December 2023 - CNAM Paris

L. Michel, M. Métillon, S. Caro, M. Ghanes, F. Plestan, J. P. Barbot,  
and Y. Aoustin

LS2N UMR CNRS 6004, Nantes Université, Ecole Centrale de Nantes

# Interest of the discretized differentiators with projectors

## The objectives

- Real-time discrete signal differentiation from a measured signal
- An alternative solution to the usual backward difference scheme to reduce
  - Chattering effect
  - Noise
  - Disturbances

## Some application domains in engineering

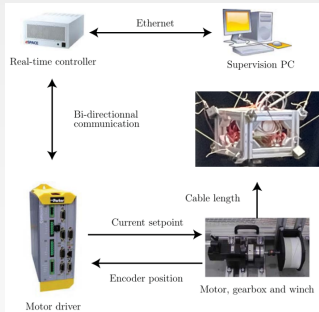
- Robotics
- Automatic control
- Signal processing
- and so on...

# The CRAFT parallel robot 1/2

## A cable-driven parallel robot



a)



b)

- Base frame of CRAFT: 4 m long, 3.5 m wide, and 2.7 m high
- Suspended moving-platform (MP): 3-DoF translational, 3-DoF rotational motions, and 8 cables; 0.28 m long, 0.28 m wide and 0.2 m high; 5 kg mass

## The CRAFT parallel robot 2/2

### Dynamic model

$$\mathbb{I}_p \ddot{\mathbf{p}} + \mathbf{C} \dot{\mathbf{p}} - \mathbf{w}_g = \mathbf{W} \boldsymbol{\tau} + \mathbf{w}_e \quad (1)$$

where:

•

$$\dot{\mathbf{p}} = \begin{bmatrix} \dot{\mathbf{t}} \\ \boldsymbol{\omega} \end{bmatrix} \quad \ddot{\mathbf{p}} = \begin{bmatrix} \ddot{\mathbf{t}} \\ \boldsymbol{\alpha} \end{bmatrix}, \quad (2)$$

$\dot{\mathbf{t}} = [\dot{t}_x, \dot{t}_y, \dot{t}_z]^\top$  and  $\ddot{\mathbf{t}} = [\ddot{t}_x, \ddot{t}_y, \ddot{t}_z]^\top$  MP linear velocity and acceleration, respectively;  $\boldsymbol{\omega} = [\omega_x, \omega_y, \omega_z]^\top$  and  $\boldsymbol{\alpha} = [\alpha_x, \alpha_y, \alpha_z]^\top$  MP angular velocity and acceleration

- $\mathbf{W}$  wrench matrix that maps the cable tension vector  $\boldsymbol{\tau}$  exerted by the cables onto MP
- External wrench  $\mathbf{w}_e$ , a 6-dimensional vector / Wrench  $\mathbf{w}_g$  due to gravity
- Matrix  $\mathbb{I}_p$  is the spatial inertia of the platform
- $\mathbf{C}$  is the matrix of the centrifugal and Coriolis wrenches

## Problem statement 1/4

### Continuous-time state model systems

- $p(t)$  is a bounded perturbation, unknown such as:

$$p_M > 0 \text{ such that } |p(t)| < p_M \text{ for all } t > 0. \quad (3)$$

- The continuous model under consideration:

$$\Sigma : \begin{cases} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= p(t) \\ y &= x_1 \end{cases} \quad (4)$$

$x_1$  and  $x_2$  are respectively the angular variable and velocity;  $y$  is the measure of  $x_1$  with additional noise  $\eta$

## Problem statement 2/4

### Discretized state model systems

- Following notation for the discretized variable:

$$\begin{aligned}\bullet(t = (k + 1)h) &= \bullet^+ \\ \bullet(t = kh) &= \bullet.\end{aligned}\tag{5}$$

the implicit Euler discretization of the continuous-time model can be written

$$\begin{cases} x_1^+ = x_1 + h x_2^+ = x_1 + h(x_2 + hp^+) \\ x_2^+ = x_2 + hp^+ \\ y = x_1 \end{cases}.\tag{6}$$

where  $p(t)$  is assumed to be a constant parameter or a slowly variable

## Problem statement 3/4

### Homogeneous continuous-time differentiator

- Homogeneity approach, interesting: Due to the dilatation, this local property to global settings  $\Rightarrow$  continuous-time homogeneous differentiator is therefore chosen under the assumption  $|p(t)| < p_M$

$$\begin{cases} \dot{z}_1 = z_2 + \lambda_1 \mu \lceil \epsilon_1 \rceil^\alpha \\ \dot{z}_2 = \lambda_2 \mu^2 \lceil \epsilon_1 \rceil^{2\alpha-1} \\ \hat{y} = z_1 \end{cases} \quad (7)$$

where

- $\alpha \in ]0.5 \ 1[$ ,  $\epsilon_1 = y - z_1$ , and the notation  $\lceil \bullet \rceil^\alpha = |\bullet|^\alpha \text{sgn}(\bullet)$
- If  $\alpha = 0.5$ , it becomes the classical super-twisting algorithm
- If  $\alpha \searrow$  then accuracy  $\nearrow$  but noise rejection  $\nearrow$

## Problem statement 4/4

### Existing Euler discretization schemes

- **Explicit method:**  $z_i$  and  $\dot{z}_i$ , known at  $t = kh$ ,  $z_i^+$  deduced from  $z_i^+ = z_i + h\dot{z}_i$

$$\begin{cases} z_1^+ = z_1 + h(z_2 + \lambda_1 \mu [\epsilon_1]^\alpha) \\ z_2^+ = z_2 + h\lambda_2 \mu^2 [\epsilon_1]^{2\alpha-1} \end{cases} \quad (8)$$

Chattering effect remains  $\Rightarrow$  the numerical solution is not attractive

- **Implicit method:**  $z_i$  is known,  $\dot{z}_i^+$  is *calculated* such as  $z^+$  is equal to  $z_i^+ = z_i + h\dot{z}_i^+$ .

$$\begin{cases} z_1^+ = z_1 + h(z_2^+ + \lambda_1 \mu [\epsilon_1^+]^\alpha) \\ z_2^+ = z_2 + h\lambda_2 \mu^2 [\epsilon_1^+]^{2\alpha-1} \end{cases} \quad (9)$$

If  $\epsilon_1^+ = 0$ ,  $z_2^+ = 0 \Rightarrow z_2 = 0$  Hence, the two correction terms  $\lambda_1 [\epsilon_1^+]^\alpha$  and  $\lambda_2 [\epsilon_1^+]^{2\alpha-1}$  with  $\epsilon_1^+ = 0$  become inoperative



# Semi-implicit Homogeneous Euler differentiators 1/2

## Semi-implicit Euler homogeneous differentiator (SIHD-1 version)

The first scheme SIHD1 allows to overcome the drawbacks of these two previous numerical schemes

$$\begin{cases} z_1^+ = z_1 + h(z_2^+ + \lambda_1 \mu |\epsilon_1|^\alpha \mathcal{N}_1) \\ z_2^+ = z_2 + E_1^+ h \lambda_2 \mu^2 |\epsilon_1|^{2\alpha-1} \mathcal{N}_1 \end{cases} \quad (10)$$

The def. of the single projector  $\mathcal{N}_1$  (associated to the enabling flag  $E_1^+$ ) reads:

$$\mathcal{N}_1(\epsilon_1) := \begin{cases} \epsilon_1 \in SD \rightarrow \mathcal{N}_1 = \frac{\lceil \epsilon_1 \rceil^{1-\alpha}}{\lambda_1 \mu h}, & E_1^+ = 1 \\ \epsilon_1 \notin SD \rightarrow \mathcal{N}_1 = \text{sign}(\epsilon_1), & E_1^+ = 0 \end{cases} \quad (11)$$

## Semi-implicit Homogeneous Euler differentiators 2/2

### Semi-implicit homogeneous Euler discretization (SIHD-2 version)

$$\begin{cases} z_1^+ = z_1 + h(z_2^+ + \lambda_1 \mu |\epsilon_1|^\alpha \mathcal{N}_1) \\ z_2^+ = z_2 + E_1^+ h \lambda_2 \mu^2 |\epsilon_1|^{2\alpha-1} \mathcal{N}_2 \end{cases} \quad (12)$$

with the projector  $\mathcal{N}_1$  and the flag  $E_1^+$  defined in (11) and when  $\epsilon_1 \in SD$   $\epsilon_1 = h \epsilon_2$  holds,  $\mathcal{N}_2$  reads as:

$$\mathcal{N}_2 := \begin{cases} \epsilon_1 \in SD' \rightarrow \mathcal{N}_2 = \frac{[\epsilon_1]^{2(1-\alpha)}}{\lambda_2 h^2 \mu^2} \\ \epsilon_1 \notin SD' \rightarrow \mathcal{N}_2 = \text{sign}(\epsilon_1) \end{cases} \quad (13)$$

$$SD' = \{ \epsilon_1 \in SD / |\epsilon_1| \leq (\lambda_1 \mu^2 h^2)^{\frac{1}{2(1-\alpha)}} \equiv |\epsilon_2| \leq (\lambda_1 \mu^2)^{\frac{1}{2(1-\alpha)}} h^{\frac{\alpha}{1-\alpha}} \}.$$

## Experimental validation 1/4

### Condition of data capture

- For each of the eight electrical motors an encoder sensor measures the angular variable of its shaft
- The eight motors with a gearbox reducer of ratio  $n = 8$
- The measured value of the angular position at the output shaft of the gearbox reducer
- The sampling period of the acquisition data is equal to 1 ms
- The recording data in position are processed off-line in order to apply the semi-implicit homogeneous Euler discretized differentiators SIHD-1 and SIHD-2

## Experimental validation 2/4

### Attenuation noise projectors (SIHD<sub>θ</sub>-1)

Measured angular positions noisy  $y \Rightarrow y_m = x_1 + \eta$  where  $\eta$  is a measurement noise (the output corrective term  $e_1$  becomes  $e_{1m} = y_m - z_1$ )

$\Rightarrow$  a modified projector including a new parameter  $\theta$  to extend SIHD-1 and SIHD-2 in order to mitigate the influence of noise

$$\begin{cases} z_1^+ = z_1 + h(z_2^+ + \lambda_1 \mu |\epsilon_{1m}|^\alpha \mathcal{N}_{\theta_1}) \\ z_2^+ = z_2 + E_{\theta_1}^+ h \lambda_2 \mu^2 |\epsilon_{1m}|^{2\alpha-1} \mathcal{N}_{\theta_1} \end{cases} \quad (14)$$

$$\mathcal{N}_{\theta_1} := \begin{cases} (1 - \theta) |\epsilon_{1m}|^{1-\alpha} < \lambda_1 \mu h \rightarrow \mathcal{N}_{\theta_1} = \frac{(1 - \theta) |\epsilon_{1m}|^{1-\alpha}}{\lambda_1 h \mu} \\ (1 - \theta) |\epsilon_{1m}|^{1-\alpha} \geq \lambda_1 \mu h \rightarrow \mathcal{N}_{\theta_1} = \text{sign}(\epsilon_{1m}) \end{cases}$$

## Experimental validation 3/4

### Attenuation noise projectors (SIHD<sub>θ</sub>-2)

$$\begin{cases} z_1^+ = z_1 + h \left( z_2^+ + \lambda_1 \mu |\epsilon_{1m}|^\alpha \mathcal{N}_{\theta_1} \right) \\ z_2^+ = z_2 + E_{\theta_1}^+ h \lambda_2 \mu^2 |\epsilon_{1m}|^{2\alpha-1} \mathcal{N}_{\theta_2} \end{cases} \quad (15)$$

$$\mathcal{N}_{\theta_2} := \begin{cases} (1 - \theta) |\epsilon_{1m}|^{2(1-\alpha)} < \lambda_2 \mu^2 h^2 \rightarrow \mathcal{N}_{\theta_2} = \frac{(1 - \theta) [\epsilon_{1m}]^{2(1-\alpha)}}{\lambda_2 h^2 \mu^2} \\ (1 - \theta) |\epsilon_{1m}|^{2(1-\alpha)} \geq \lambda_2 \mu^2 h^2 \rightarrow \mathcal{N}_{\theta_2} = \text{sign}(\epsilon_{1m}) \end{cases}$$

## Experimental validation 4/4

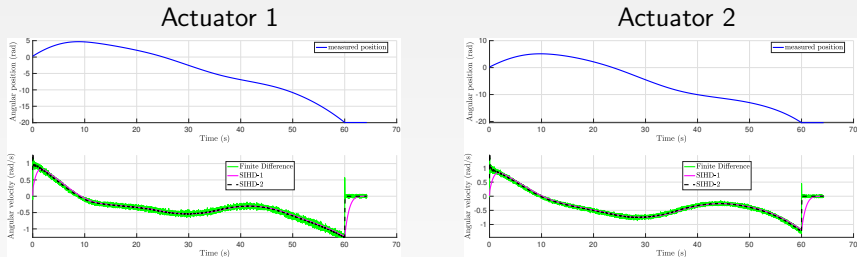
### Parameter setting

- $\lambda_i$ ,  $i = 1, 2$  parameters chosen such as the linear part stable
- Value of homogeneous exponent  $\alpha$  is chosen between the coefficient of Levant's differentiator ( $\alpha = 0.5$ ) and the linear solution of the discretized differentiators SIHD-1 and SIHD-2 ( $\alpha = 1$ )
- The parameter  $\theta$  is chosen by numerical test trial and error allowing a good filtering of the noise *i.e.*  $0.5 < \theta < 1$
- Numerical values:

$$\lambda_1 = 2 \cdot 10^4, \quad \lambda_2 = 1 \cdot 10^4, \quad \alpha = 0.81, \quad \theta = 0.9, \quad \mu = 1 \quad (16)$$

# Experimental Results 1/5

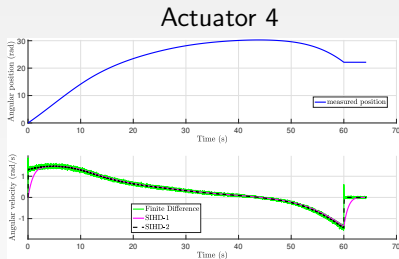
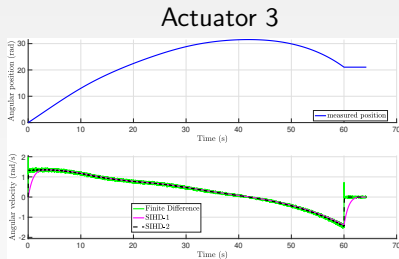
## Comparison between back. difference method, SIHD-1 and SIHD-2



**Figure:** Recording angular positions (blue color), reference velocity calculated by backward difference (green color), estimated velocities with SIHD-1 (solid line red color) and estimated velocities with SIHD-2 (dashed line black color).

## Experimental Results 2/5

### Comparison between back. difference method, SIHD-1 and SIHD-2

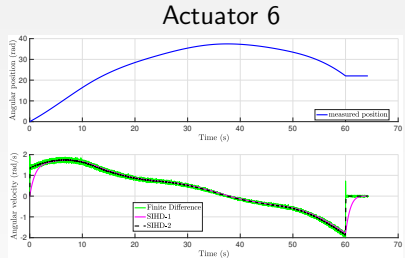
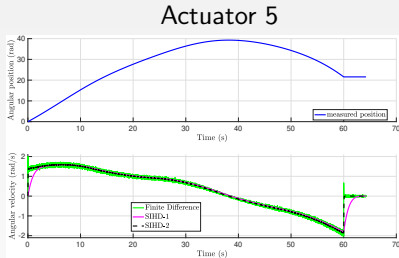


**Figure:** Recording angular positions (blue color), reference velocity calculated by backward difference (green color), estimated velocities with SIHD-1 (solid line red color) and estimated velocities with SIHD-2 (dashed line black color).



## Experimental Results 3/5

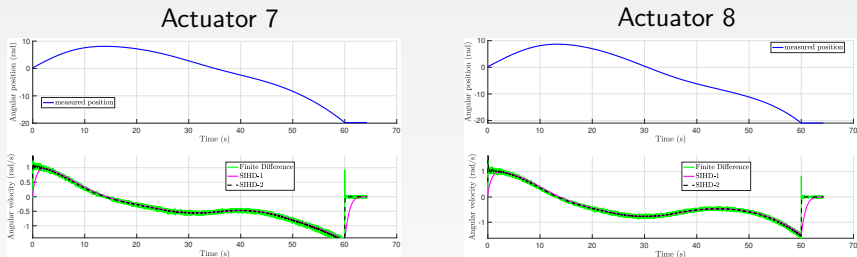
### Comparison between back difference method, SIHD-1 and SIHD-2



**Figure:** Recording angular positions (blue color), reference velocity calculated by backward difference (green color), estimated velocities with SIHD-1 (solid line red color) and estimated velocities with SIHD-2 (dashed line black color).

## Experimental Results 4/5

### Comparison between back difference method, SIHD-1 and SIHD-2



**Figure:** Recording angular positions (blue color), reference velocity calculated by backward difference (green color), estimated velocities with SIHD-1 (solid line red color) and estimated velocities with SIHD-2 (dashed line black color).

## Experimental Results 5/5

### Discussion

- Performances of these differentiators are almost uniform whatever the motor
- Angular velocities are smoother *i.e.* less noisy than the reference velocities obtained by backward difference
- Dynamics of the three signals are similar / Transient behavior of the velocity with SIHD-2 is better than with Euler differentiator SIHD-1
- No tachymeter sensor on the motor shaft  $\Rightarrow$  difficult to consider the backward difference signal as the reference velocity
- Evaluation of the sensitivity to noise, for example with motor 4

	angular velocity (rad/s)		
	$\sigma$ , BD	$\sigma$ , SIHD-1	$\sigma$ SIHD2
motor 4	0.032	0.017	0.017

# Conclusion, perspectives

## Conclusion

- Cable-driven parallel robot *CRAFT* : a complex mechanical system
- *CRAFT* promising for handling, rescue, or personal assistance
- Two new semi-implicit homogeneous differentiators applied with success to estimate the angular velocity of the output shaft of the eight motors of *CRAFT*
- Good experimental results: less noisy than the one calculated with backward difference

## Perspectives.

- *Cascaded* utilization to estimate the acceleration ;-)
- Semi-implicit homogeneous differentiators for identification tasks of model parameters
- Co-manipulation between its effector and human thanks to a force sensor