## Set-point Regulation of an Uncertain 6-DOF Magnetically Levitated Positioning Stage

Shai Arogeti

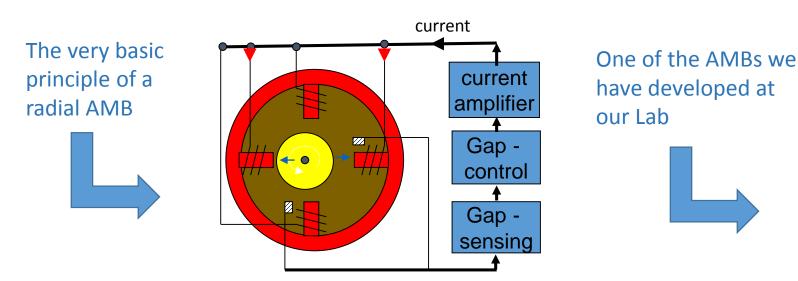
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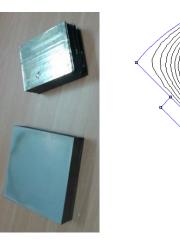
### Outline

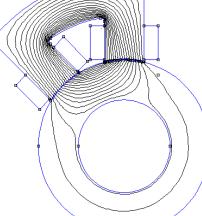
- Background some of our activities in the field of AMB
- Six-DOF Precision Positioning Stage (mechanical structure and dynamical model)
- Iterative Output Control Law (theoretical)
- Iterative Output Control Law (practical)
- Experimental Results
- Summary and Conclusions

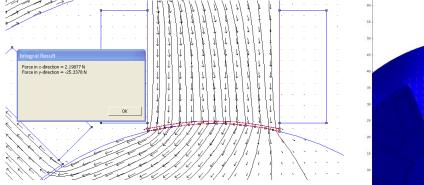
- Active Magnetic Bearings (AMB) allows rotation with no friction.
- It uses electromagnetic forces to prevent mechanical contact between the static (stator) and the moving (rotor) parts.
- Applications of AMBs include very high rotating systems, such as turbomolecular pumps and Flywheel Energy Storage Systems.

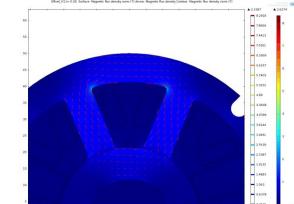


The design includes magnetic and mechanical analysis (using finite elements software)







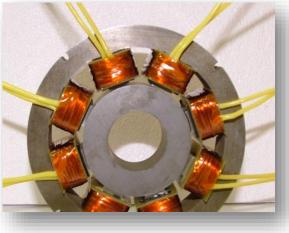


#### Our interests includes:

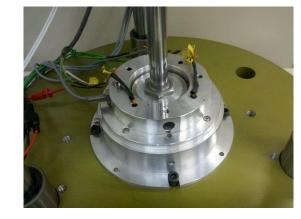
- Optimal design for minimum losses
- Adaptive control (unknown imbalance)
- AMB control, the case of elastic shaft.

Our AMBs are produced (**in-house**) from raw materials (e.g., of electrical transformers)











A 5 DOF AMB system at our Lab, it includes:

- Two radial AMBs
- One axial AMB (works against gravity)
- High speed brushless motor (up to 60,000 RPM)



#### Many in-house skills have been acquired

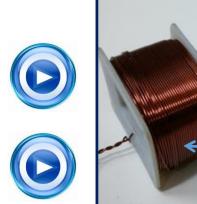
Besides AMBs design and control we have developed a winding machine at our Lab.

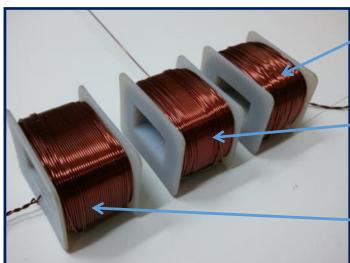
- The wire tension is closed-loop controlled.
- Very slow winding allows maximum number of windings in a given volume.
- The bobbins (spools) are 3-D printed. •









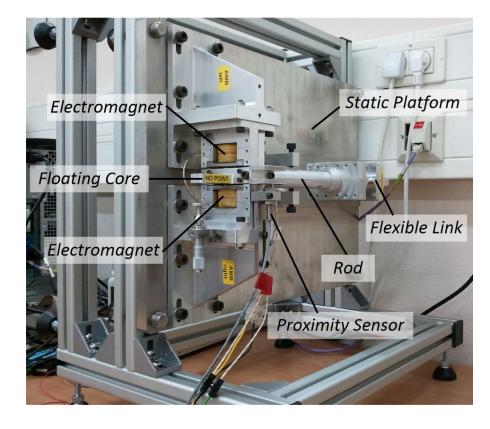


Industrial Machine Prod. Time (?)

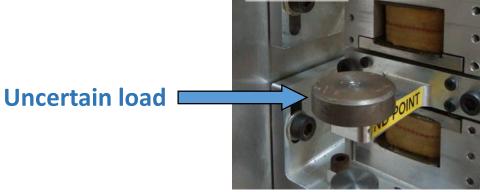
**Designed Machine** Prod. Time: ~15min

**Designed Machine** Prod. Time: ~50min

A single degree of freedom "AMB" (imbalance effects can be added by a small rotating eccentric mass)

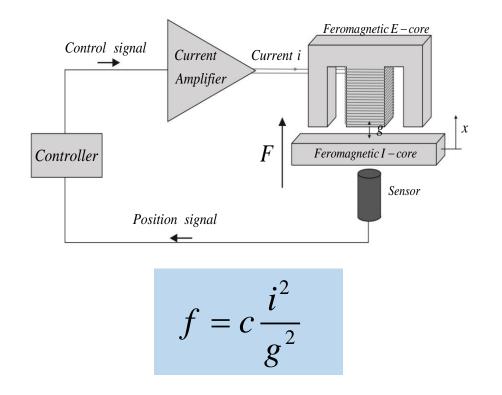


This system can be placed vertically or horizontally (depending on the gravity effect we want to achieve)



m=0.66kg

The simplest model of the electromagnetic force (commonly utilized for control design)



Systems consist of these actuators are **unstable** 

A single DOF electro-magnetic actuator includes two E cores and a single I core. Force can be applied in both directions (usually by linearizing the system around a bias current)

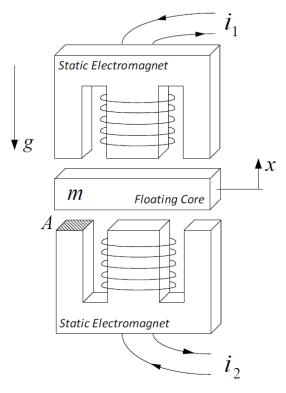


Fig. 1: SDOF AMB positioning system scheme







One very talented **MSc student**: Sergei Basovich



6 DOF magnetically levitated positioning stage





All of this brings us to the subject of this talk:

# Set-point Regulation of an Uncertain 6-DOF Magnetically Levitated Positioning Stage

## Shai Arogeti

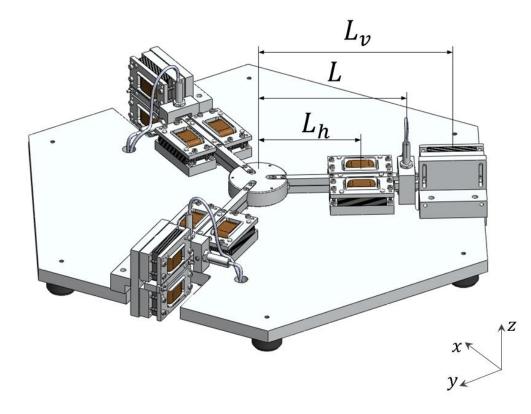
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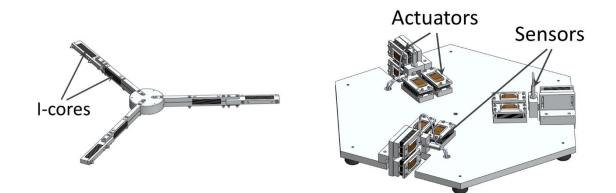
So, I could have started here, but . . .

#### **Stage Structure**

# The stage consists of six electromagnetic actuators (Three are vertical and the other are horizontal)



The levitated-part consists of three arms connected in a joint, where each arm serves as a support for two I-cores.

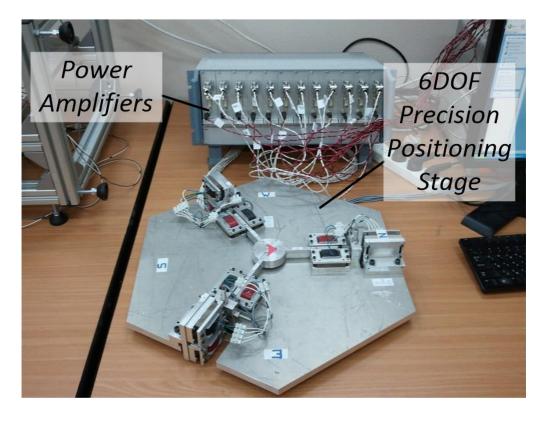


The air gaps in all six actuators are measured by six proximity (eddy current) sensors to obtain information about the stage position and orientation.

The traveling range is  $\pm 450 \times 10^{-6} [m] \pm 1500 \times 10^{-6} [rad]$ 

#### **Stage Structure**

# The mechanical data of the stage is . . . (products of inertia are negligible)



#### TABLE I: Stage Parameters

$I[kgm^2]$ - inertia of the platen			
$I_{xx}$	$I_{yy}$	$I_{zz}$	
5.4336e-3	5.4336e-3	1.0844e-2	
$I_{xy}$	$I_{yz}$	$I_{zx}$	
2.5717e-17	9.1320e-22	1.1316e-21	
Platen nominal mass	Electromagnetic coef.	Nominal air gap	
m[kg]	$c[Nm^2/A^2]$	$l_0[m]$	
0.648	1.0809e-5	450e-6	
Relevant dimensions [m]			
$L_h$	L	$L_v$	
0.113	0.163	0.213	

The stage (levitated part) is modeled as a rigid body with 6 DOF

$$M \ddot{q} + C(\dot{q})\dot{q} + w + \zeta = \Phi_{LF} \cdot f \qquad , \label{eq:mass_states}$$

$$q = [x, y, z, \varphi, \theta, \psi]^T$$

#### (small angles and small displacements are assumed)

	The inertia matrix is	$M = diag\left\{m_t, m_t, m_t, I_{xx}, I_{yy}, I_{zz}\right\}$
	The actuator forces	$f = [f_1, f_2, f_3, f_4, f_5, f_6]^T$
Assumed	The gravity force	$w = [0, 0, m_t g, 0, 0, 0]^T$
unknown	Torque due to a shifted c.g. $\zeta = [0_{1\times 3}, -\Delta \tau^T]^T$ (because of the payload)	

Transformation from actuator forces to body forces

$$\Phi_{LF} = \begin{bmatrix} Y & P \\ L_h P & -L_v Y \end{bmatrix} \quad \text{where} \quad Y = \begin{bmatrix} \sqrt{3/2} & 0 & -\sqrt{3/2} \\ -1/2 & 1 & -1/2 \\ 0 & 0 & 0 \end{bmatrix} , \quad P = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

Transformation from body coordinates to actuator coordinates

$$l = \Phi_{LQ} q \qquad \text{where} \qquad \Phi_{LQ} = \begin{bmatrix} Y^T & L_h P^T \\ P^T & -L_v Y^T \end{bmatrix} = \Phi_{LF}^T$$

Transformation from sensor coordinates to body coordinates

$$q = \Phi_{QS} s \qquad \text{where} \qquad \Phi_{QS}^{-1} = \begin{bmatrix} -Y^T & -LP^T \\ -P & LY^T \end{bmatrix}$$

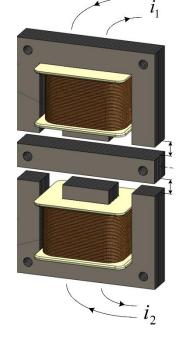
The actuator forces  $f_k = c \left( \frac{i_{k1}^2}{\left(l_0 + l_f - l_k(q)\right)^2} - \frac{i_{k2}^2}{\left(l_0 + l_f + l_k(q)\right)^2} \right)$ , k = 1, 2, 3, 4, 5, 6

 $l_f$  represents an additional length due to final permeability

Then, control currents ( $i_{k1}$  and  $i_{k2}$ ) are applied based on:

$$\begin{split} i_{k1} &= \begin{cases} \left( l_0 + l_f - l_k(q) \right) \sqrt{f_k / c}, & f_k > 0 \\ 0 & otherwise \end{cases} \\ i_{k2} &= \begin{cases} \left( l_0 + l_f + l_k(q) \right) \sqrt{-f_k / c}, & f_k < 0 \\ 0 & otherwise \end{cases}, \ k = 1, 2, 3, 4, 5, 6 \end{cases} \end{split}$$

$$\kappa = 1, 2, 3, 4, 3, 0$$



 $l_0 = 450 \times 10^{-6} [m]$  $l_f = 1.8634 \times 10^{-6} [m]$ 

The matrix  $C(\dot{q})$  of the term  $C(\dot{q})\dot{q}$  is given as,

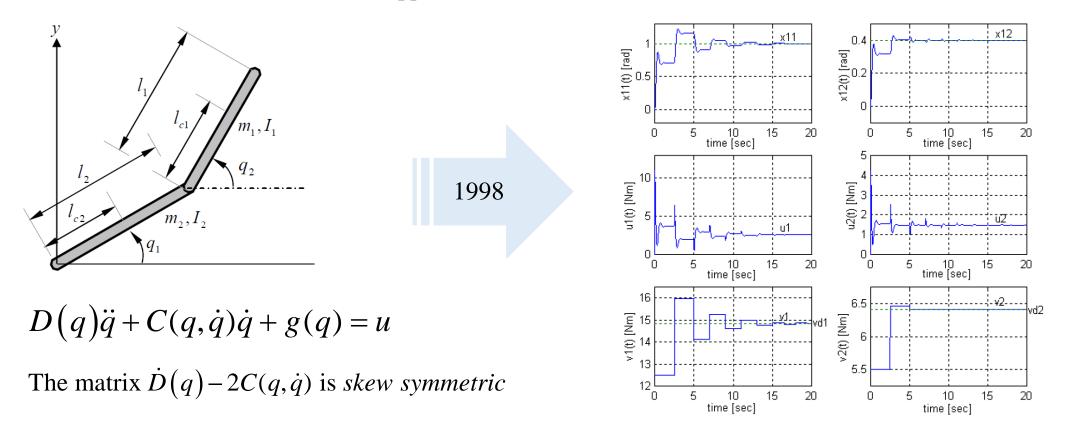
$$C = \begin{bmatrix} 0_{3\times3} & 0_{3\times3} \\ 0_{3\times3} & C_0 \end{bmatrix} \quad \text{Where,} \quad C_0 = \begin{bmatrix} 0 & I_{zz}\dot{\psi} & -I_{yy}\dot{\theta} \\ -I_{zz}\dot{\psi} & 0 & I_{xx}\dot{\phi} \\ I_{yy}\dot{\theta} & -I_{xx}\dot{\phi} & 0 \end{bmatrix}$$

It is important to note that the matrix  $\dot{M} - 2C(\dot{q})$ is a skew symmetric matrix

#### **Iterative Output Control Law**

#### Some old results from robotics

A. Ailon, "**Output controller based on iterative schemes for set-point regulation of uncertain flexible-joint robot models**," *Automatica*, vol. 32, no. 10, pp. 1455-1461, 1996.



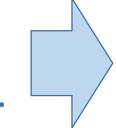
### Iterative Output Control Law (Six-DOF Positioning Stage)

We define the uncertain term as  $p \triangleq w + \zeta$ 

The state space representation,

$$\dot{x}_{1} = x_{2}$$
  
$$\dot{x}_{2} = M^{-1}(-C(x_{2})x_{2} - p + \Phi_{LF} \cdot f)$$
  
where,  $x_{1} = q$ ,  $x_{2} = \dot{q}$ 

For the 6DOF stage, we propose the following **iterative controller-observer** 



$$f = \Phi_{LF}^{-1} \left( -C_1 (x_1 - x_1^d) - C_2 \dot{z} + v \right)$$
$$\dot{z} = -K(z - x_1)$$

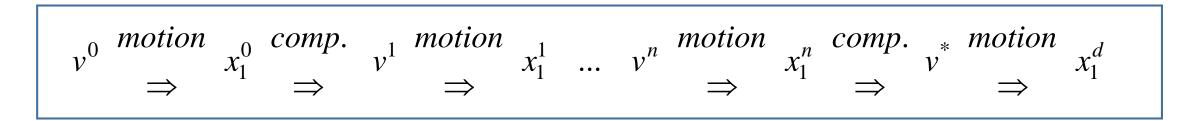
 $C_1, C_2, K$  positive definite & diagonal

The compensating piecewise function v (constant at each iteration) is defined by the following update law

$$v^{n+1} = v^n - S\left(x_1^n - x_1^d\right)$$

Where, 
$$S \triangleq C_1(I - \alpha I)$$
,  $0 < \alpha < \frac{1}{2}$ 

#### This can be schematically represented by the following process



**Lemma 1** Let the system with the uncertain term  $p \in \mathcal{R}^6$  be controlled by the controller-observer with an arbitrary  $v \in \mathcal{R}^6$ . Then, the equilibrium point  $x_1 = \overline{x}_1$  of the closed loop system is asymptotically stable.

## Proof Define a scalar function $H(x_1, x_2, z)$ as, $H(x_1, x_2, z) = \frac{1}{2} [x_2^T M x_2 + (x_1 - z)^T C_2 K(x_1 - z) + (x_1 - x_1^d - C_1^{-1} v)^T C_1 (x_1 - x_1^d - C_1^{-1} v)] + U_p(x_1)$ where $U_p(x_1)$ satisfies $dU_p(x_1)/dx_1 = p$ . Evaluating $\frac{dH(x_1, 0, z)}{dx_r} = 0$ where $x_r = [x_1^T, z^T]^T$ , yields the steady state equations of the closed loop. $p + C_1(x_1 - x_1^d) + C_2 K(x_1 - z) - v = 0$ $-C_2 K(x_1 - z) = 0$

#### Proof (cont.)

Evaluating the Hessian ( 
$$d^2 H(x_1, 0, z) / d^2 x_r$$
 ) we obtain,  $\begin{bmatrix} C_1 + C_2 K & -C_2 K \\ -C_2 K & C_2 K \end{bmatrix}$ 

which can be shown to be positive definite for  $C_1, C_2, K > 0$ .

Therefore the scalar function  $H(x_1, x_2, z)$  is a convex function and it has a global minimum at  $\overline{x}_r = [\overline{x}_1^T, \overline{z}^T]^T$  for a given constant vector v

Thus, a Lyapunov candidate function can be defined as  $V = H(x_1, x_2, z) - H(\overline{x}_1, 0, \overline{z})$ 

and its time derivative is  $\dot{V} = -C_2 K \dot{z}^2 \le 0$ 

Hence, invoking the LaSalle's invariance principle, asymptotic stability of the equilibrium point  $\bar{x}_r = [\bar{x}_1^T, \bar{z}^T]^T$  is concluded.

**Lemma 2** Consider the stage model and define the map  $T(v): \mathcal{R}^6 \to \mathcal{R}^6$  as,

$$T(v) = v - S\left(x_1 - x_1^d\right)$$

Then, the map T(v) is a global contraction, i.e., there exists exactly one  $v^*$  such that.

 $T(v^*) = v^*$ 

Proof

For a given couple of vectors  $v^1$  and  $v^2$  (of the series  $\{v^n\}$  generated by T(v))

$$T(v^{1}) - T(v^{2}) = v^{1} - v^{2} - C_{1}(I - \alpha I)(x_{1}^{1} - x_{1}^{2})$$

From the equilibrium equations, it follows that  $x_1^1 - x_1^2 = C_1^{-1}(v^1 - v^2)$ 

Since  $C_1$  is diagonal, we have (from the last two equations)  $||T(v^1) - T(v^2)|| = \alpha ||v^1 - v^2||$ 

Hence, the map T(v) is contraction, with  $T(v^*) = v^*$ , and  $\left\|v^* - v^n\right\| \leq \frac{\alpha^n}{1 - \alpha} \left\|T(v^0) - v^0\right\|$ 

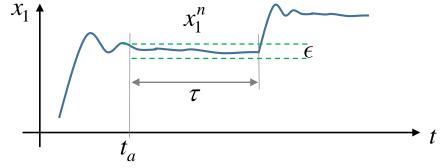
- Since the convergence of the closed-loop system to the desired setpoint  $x_1^d$  involves an infinite time process, this algorithm is **impractical**.
- For a real control task, the controller should be used with a decision module that (within a single iteration) concludes convergence to a sufficiently close vicinity of the intermediate equilibrium point.
- As a result, due to the differences between the theoretical and practical intermediate equilibrium points, the last term of the practical process will slightly deviate from the desired set point x<sub>1</sub><sup>d</sup>.
- How close to  $x_1^d$  can we get ? Lets put this in a mathematical framework

#### Assumption

For a particular control task, positive constants  $\epsilon$  and  $\tau$  can be selected in such a way that from satisfaction of

$$||x_1^n(t) - x_1^n(t_a)|| \le \epsilon, \quad \forall t \in [t_a, t_a + \tau], t_a \ge 0$$

during the n-th iteration, it can be concluded that



$$\|\varphi_{v}(t) - \overline{\varphi}_{v}(v)\| \leq \zeta(\epsilon), \quad \forall t \in [t_{a}, \infty)$$

where  $\varphi_v(t) = [x_1(t)^T, x_2(t)^T, z(t)^T]^T$  is the system trajectory,

 $\overline{\varphi}_{v}(v) = [x_{1}^{nT}, 0, z_{1}^{nT}]^{T}$  is the *n*-th equilibrium point,

and  $\zeta(\epsilon)$  is a constant (corresponding to a chosen  $\epsilon$  )

Practically, we are not using the map  $T(v) = v - S(x_1 - x_1^d)$ 

hence, we define the practical map  $E(v): \mathcal{R}^6 \to \mathcal{R}^6$ 

$$E(v) = v - S\left(x_1 + \Delta(v) - x_1^d\right)$$

where the error term  $\Delta(v)$  satisfies

$$\|\Delta(v)\| \le \epsilon, \forall v \in \mathcal{R}^6$$

and we use following lemma.

**Lemma 3** Consider the map E(v). Then, for any pair of vectors  $\{v^1, v^2\}$  satisfying

$$\left\|v^{1}-v^{2}\right\| \geq \theta \triangleq \frac{2\lambda_{max}(S)\epsilon}{\alpha}$$

the following holds

$$\left\| E\left(v^{1}\right) - E\left(v^{2}\right) \right\| \leq 2\alpha \left\| v^{1} - v^{2} \right\|$$

Proof

Expanding 
$$E(v^1) - E(v^2)$$
 we obtain

$$E(v^{1}) - E(v^{2}) = T(v^{1}) - T(v^{2}) + S\Delta(v^{2}) - S\Delta(v^{1})$$

and 
$$\left\| E\left(v^{1}\right) - E\left(v^{2}\right) \right\| = \left\| T\left(v^{1}\right) - T\left(v^{2}\right) + S\Delta\left(v^{2}\right) - S\Delta\left(v^{1}\right) \right\|$$

#### Proof (cont.)

For the right-hand side of the last equation, by the triangle inequality, we have

$$\left\| T\left(v^{1}\right) - T\left(v^{2}\right) + S\Delta\left(v^{2}\right) - S\Delta\left(v^{1}\right) \right\| \leq \left\| T\left(v^{1}\right) - T\left(v^{2}\right) \right\| + \left\| S\Delta\left(v^{2}\right) - S\Delta\left(v^{1}\right) \right\|$$

For the right-hand side of the last equation,

$$\left\|T\left(v^{1}\right)-T\left(v^{2}\right)\right\|+\left\|S\Delta\left(v^{2}\right)-S\Delta\left(v^{1}\right)\right\|\leq\alpha\left\|v^{1}-v^{2}\right\|+2\lambda_{max}(S)\epsilon^{2}$$

For the right-hand side of the last equation, using the Lemma condition,

$$\alpha \left\| v^1 - v^2 \right\| + 2\lambda_{max}(S)\epsilon \le 2\alpha \left\| v^1 - v^2 \right\|$$

So, as long as 
$$\left\|v^1 - v^2\right\| \ge \theta \triangleq \frac{2\lambda_{max}(S)\epsilon}{\alpha}$$

We have, 
$$||E(v^1) - E(v^2)|| \le 2\alpha ||v^1 - v^2||$$
,  $0 < \alpha < \frac{1}{2}$ 

#### and the (practical) map E(v) can be considered contraction.

Now suppose that for the sequence  $\{v^n\}$  generated by  $v^{n+1} = E(v^n)$ , n = 0, 1...there exists a minimal integer  $m(v^0)$  for which  $\left\|v^{m(v^0)-1} - v^{m(v^0)}\right\| < \theta$ 

The  $m(v^0)$ -th iteration is the final iteration.

It is very important to be able to estimate the deviation from  $x_1^d$  after the final iteration

For that, we have **Lemma 4** and **Lemma 5** (in our paper), which are not presented here.

The final conclusion from these Lemmas is that,

$$||x_1^d - x_1^{m(v^0)}|| \le 3 \frac{\lambda_{max}(C_1)}{\lambda_{min}(C_1)} \epsilon$$



The upper bound of the steady state error norm (after the last iteration) can be made as small as desired.

- The **traveling range** of the proposed positioning stage **is relatively** restricted.
- If not all the terms of the practical intermediate equilibrium point sequence are found inside the operational area, the **steady state** equations are no longer valid.
- Hence, another important practical aspect is the **boundedness** of the intermediate steady state response.
- To provide that, we introduce the **initialization phase** which augments the iterative process.

The augmented iterative process is represented as,

where,  $\chi_1^n \triangleq x_1^n + \Delta(v^n)$  is the *n*-th term of the series of *practical equilibrium points* generated during the practical process.

- The initialization phase represents the response of the system with  $x_1^d = 0$  and v = 0.
- As a result of the initialization, the system will move to  $\chi_1^i$ .
- The initialization phase assures that the update mechanism starts acting when  $\chi_1^i$  is found inside the traveling range of the stage.
- The initial input  $v^0$  is determined by  $v^0 = -S(\chi_1^i x_1^d)$

and for the rest of the process we use  $v^{n+1} = v^n - S(\chi_1^n - \chi_1^d)$ 

**Lemma 6** Consider the system with  $x_1^d = 0$  and v = 0, and let  $\delta^i$  be the air-gap vector corresponding to the practical equilibrium point  $\chi_1^i$ . Then, for  $C_1$  satisfying

$$\left\| \Phi_{LQ} C_1^{-1} \right\|_2 \cdot \beta < l_0 - \left\| \Phi_{LQ} \right\|_2 \cdot \epsilon \qquad (*)$$

where the scalar  $l_0$  represents the nominal air gap value in each actuator and  $\beta$  is the upper bound of P, the following holds,

$$\left\|\delta^{i}\right\| < l_{0}$$

**Lemma 7** Consider the system with the practical update law. Let  $\delta^n$  and  $l^d$  be the air-gap vectors, corresponding to  $\chi_1^n$  and  $x_1^d$  respectively. Then, for  $C_1$  satisfying (\*) and for,  $v^0 = -S(\chi_1^i + \chi_1^d)$ 

the following holds,

 $|\delta_k^n| < l_0, \quad k = 1, 2, 3, 4, 5, 6$ 

Assumptions required for the proofs of **Lemma 6** and **Lemma 7** (can be found in our paper),

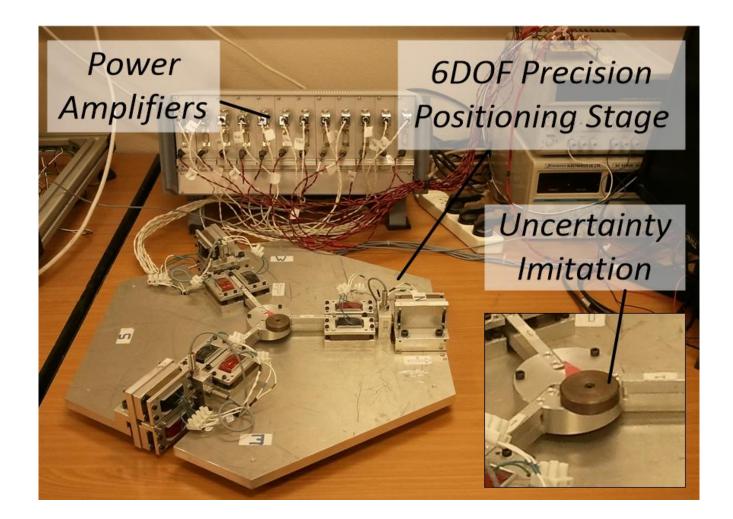
- For the uncertain term p, a positive constant  $\beta \triangleq \sup \|p\|$ exists and it is known.
- The constants  $\epsilon$  and  $\alpha$  can be selected such that  $\epsilon < \frac{1-2\alpha}{2 \cdot \left\| \Phi_{LQ} \right\|_2} l^0$
- All the component of the air gap vector  $l^d$ , corresponding to the desired set-point vector  $x_1^d$ , satisfies

$$|l_k^d| < l_0 - \frac{2}{1-\alpha} \left\| \Phi_{LQ} \right\|_2 \cdot \epsilon, \quad k = 1, 2, 3, 4, 5, 6$$

The presented algorithm was verified experimentally.

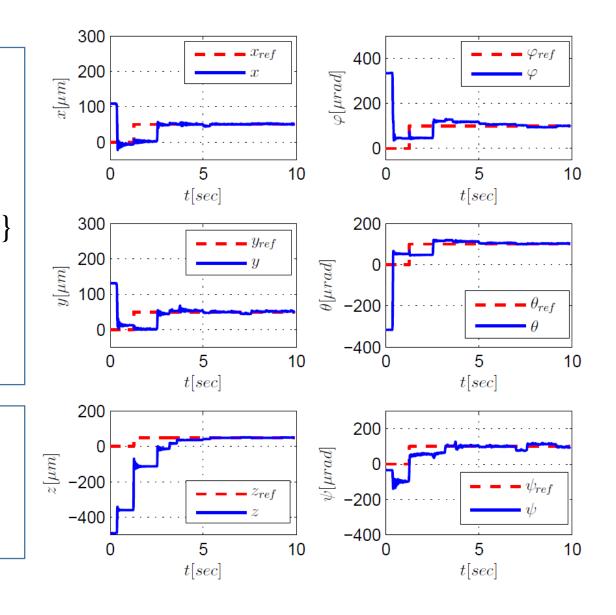
To imitate the uncertainty we attached a steel payload of m = 0.07 kg to the stage platen.

besides the additional negative force (w.r.t., z), it causes uncertain torques (w.r.t., x and y ).



To implement the controller we selected,  $C_{1} =$ 1000 · *diag* {32.5, 32.5, 32.5, 16.8, 16.8, 16.8}  $C_2 = diag\{25, 25, 31, 11.5, 11.5, 7.3\}$  $\alpha = 0.1, \quad \epsilon = 1 \times 10^{-6}, \quad \tau = 0.2$ 

while the required set point  $x_1^d = 1 \times 10^{-6} \{50, 50, 50, 100, 100, 100\}^T$ 

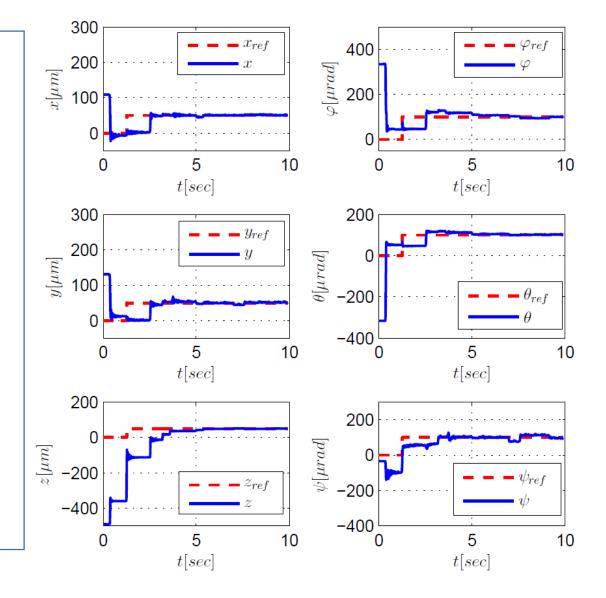


All required assumptions for the initialization phase have verified.

At the time slot 0 < t < 0.35 the system stays at initial conditions.

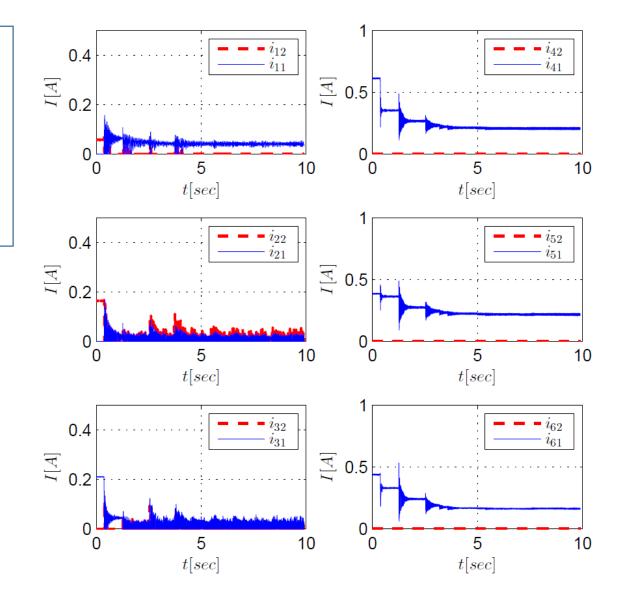
At the time slot 0.35 < t < 1.26 it undergoes the initialization phase.

At the time section 1.26 < t < 2.53the system responses to  $v^0$  and  $x_1^d$ .



These are the control currents.

For the vertical actuators, only the upper coils were activated.

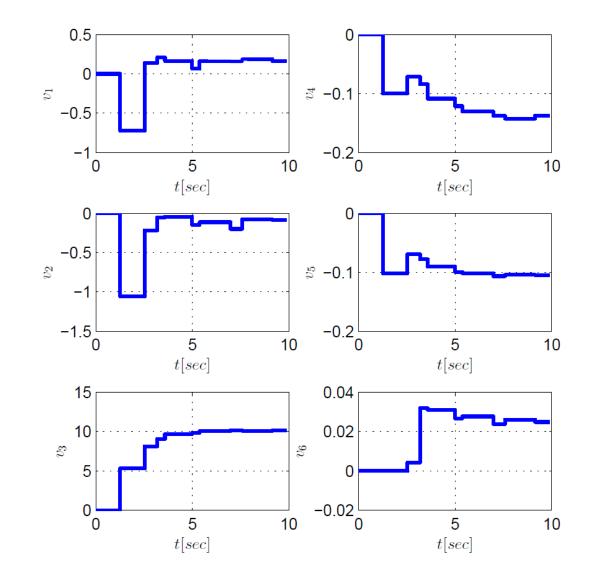


For the attached payload the uncertain vector p is estimated as,  $p = \{0, 0, 7.044, -0.019, -0.011, 0\}^T$ 

For this *p* , we have defined  $\beta = 7.1$ 

In equilibrium condition, for  $x_1 = x_1^d$ , we suppose to have v = p

We understand that, v here is compensating for uncertainties that are not considered in the model



### **Summary and Conclusions**

- A brief introduction to our activities in the field of AMB has been presented.
- A 6-DOF precision positioning stage, based purely on magnetic levitation principles (developed at our Lab) was presented
- Some old results from robotics have been modified to suit the case of magnetic levitation application (to the set point goal).
- An unknown payload was assumed (and hence, an unknown c.g.)
- The results were demonstrated experimentally.
- The proposed algorithm can be used as an identification routine, allowing realization of a simple controller (that is not based on iterations).