

# **Nonlinear Observability Analysis of Micro-machined Electrostatic Actuators Using Self-Sensing**

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# Agenda

- 1. Background
- 2. Observability Analysis
- 3. Verifications
- 4. Conclusions

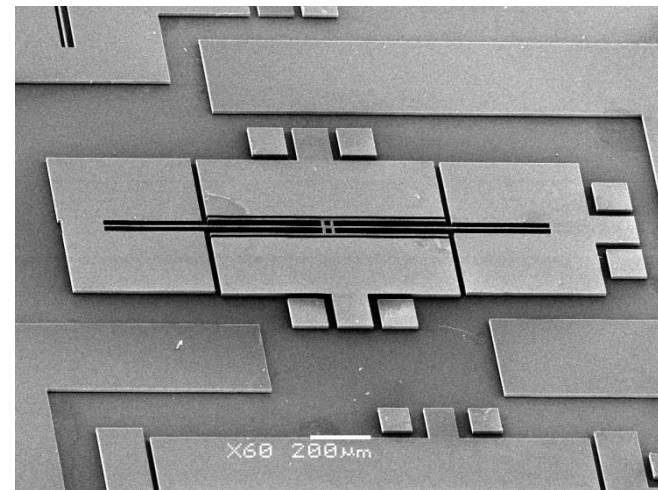
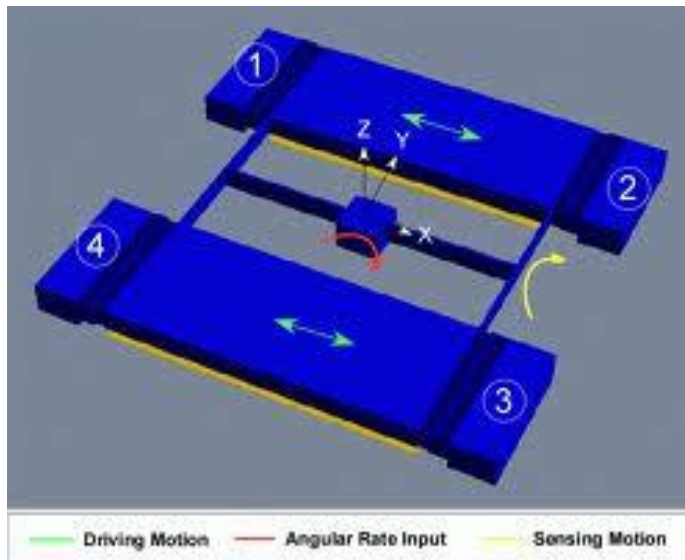
# 1. Background

Model of PPAs

Problem Statement

# Applications of PPAs

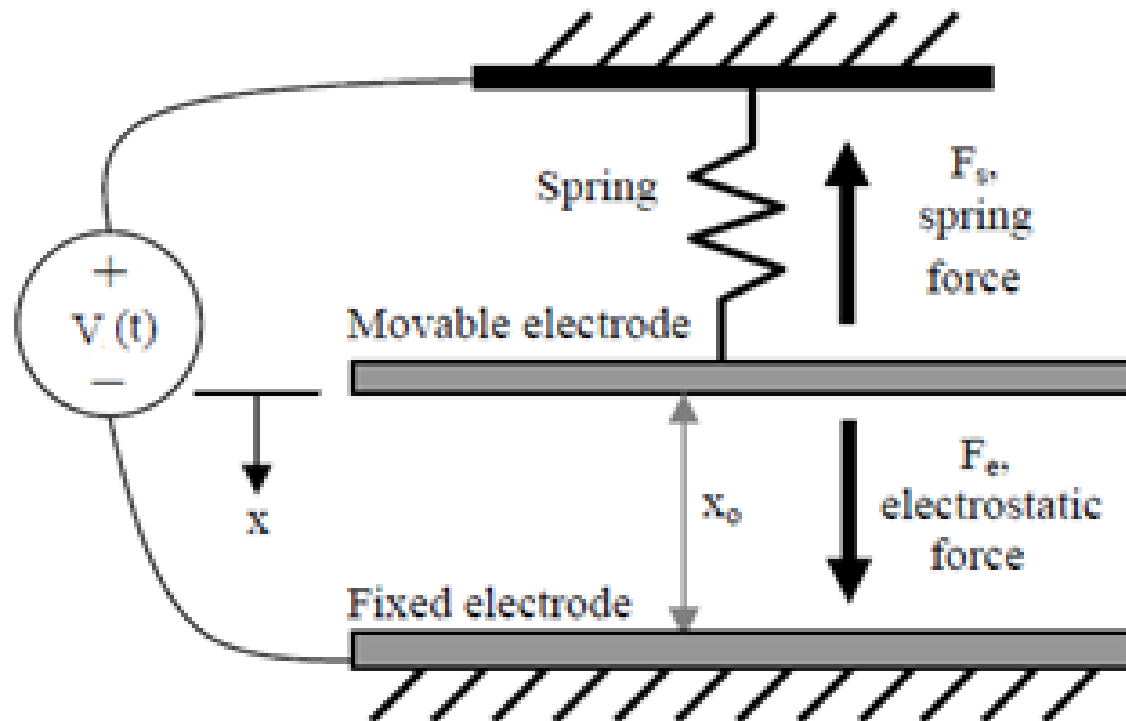
- Electrostatic parallel-plate actuators (PPAs) are used in many types of microelectromechanical systems (MEMS) devices, such as accelerometers, variable capacitors, RF devices and micromirrors.



# Parallel Plate Actuators

Ignoring gravity and damping force, at equilibrium condition:

Spring force = electrostatic force

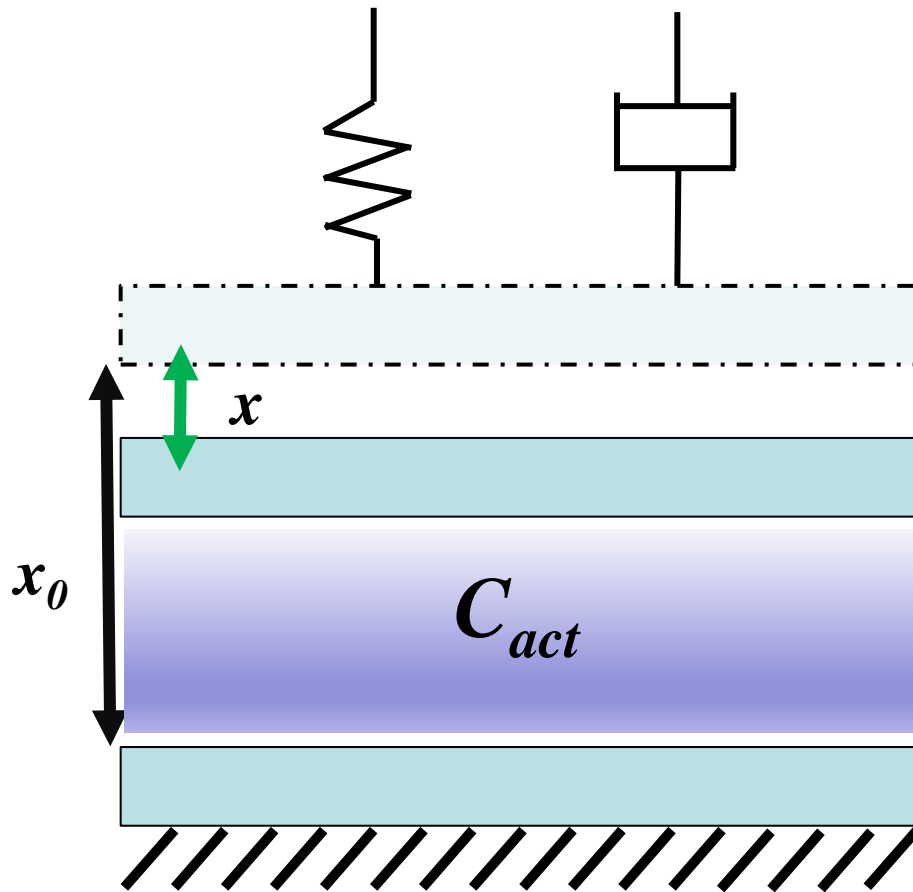


# System Dynamics' Equation (1)

$$m\ddot{x} + c\dot{x} + kx = f(x, V) = \frac{\epsilon A V^2}{2(x_o - x)^2}$$

- A: areas of two electrodes
- m: mass
- c: damping coefficient
- k: spring constant
- $x_0$ : initial gap
- $\epsilon_0 \epsilon_r$ : permittivity  $\approx 8.854$  pf/m

# System Dynamics' Equation (2)



The capacitance between the electrodes is:

$$C_{act} = \frac{\epsilon_0 \epsilon_r A}{x_0 - x}$$

Its resonant frequency is:

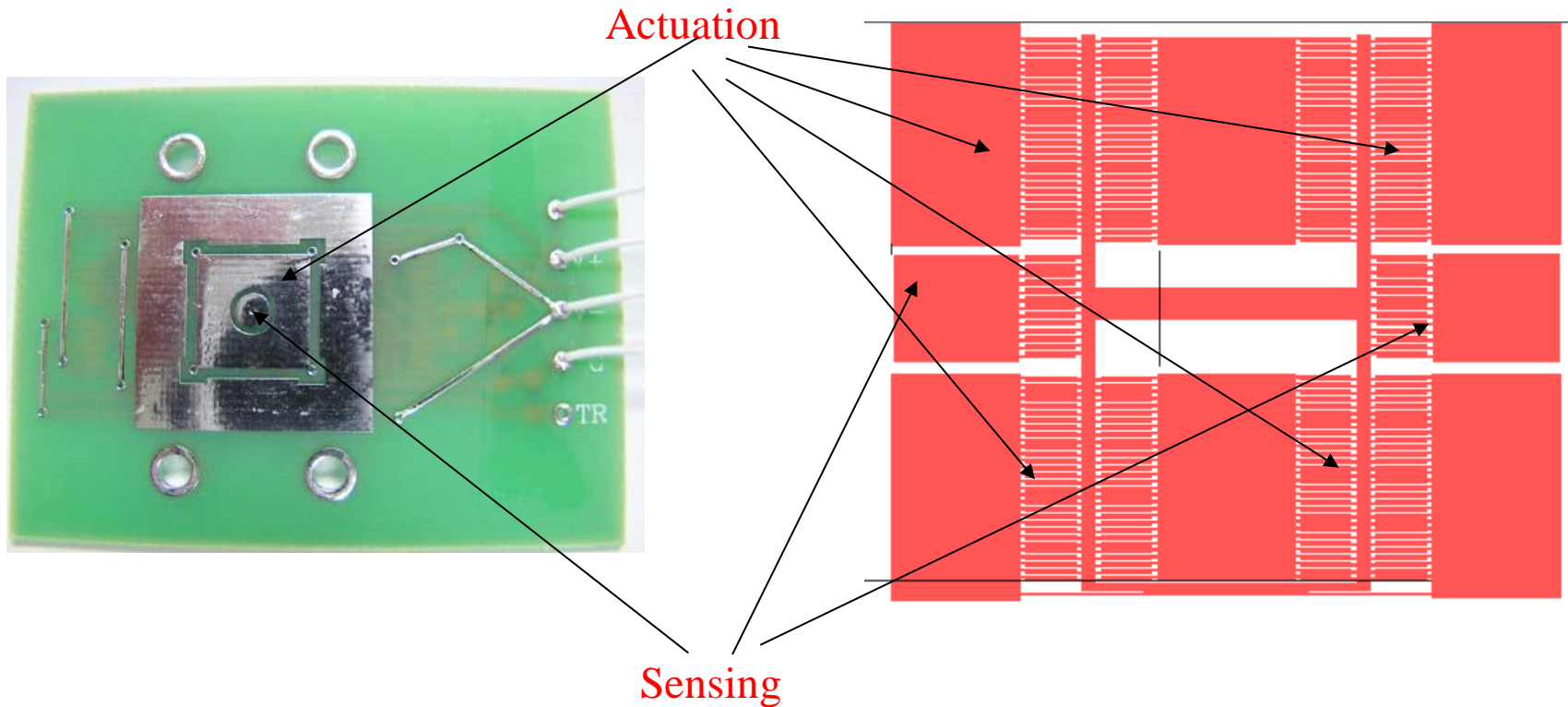
$$\omega_0 = \sqrt{\frac{k}{m}}$$

# Main Applications' Issue

- Limited stable range (less than  $1/3$ )
- Typically open loop control just for on/off applications
- Feedback control techniques are required in some advanced applications

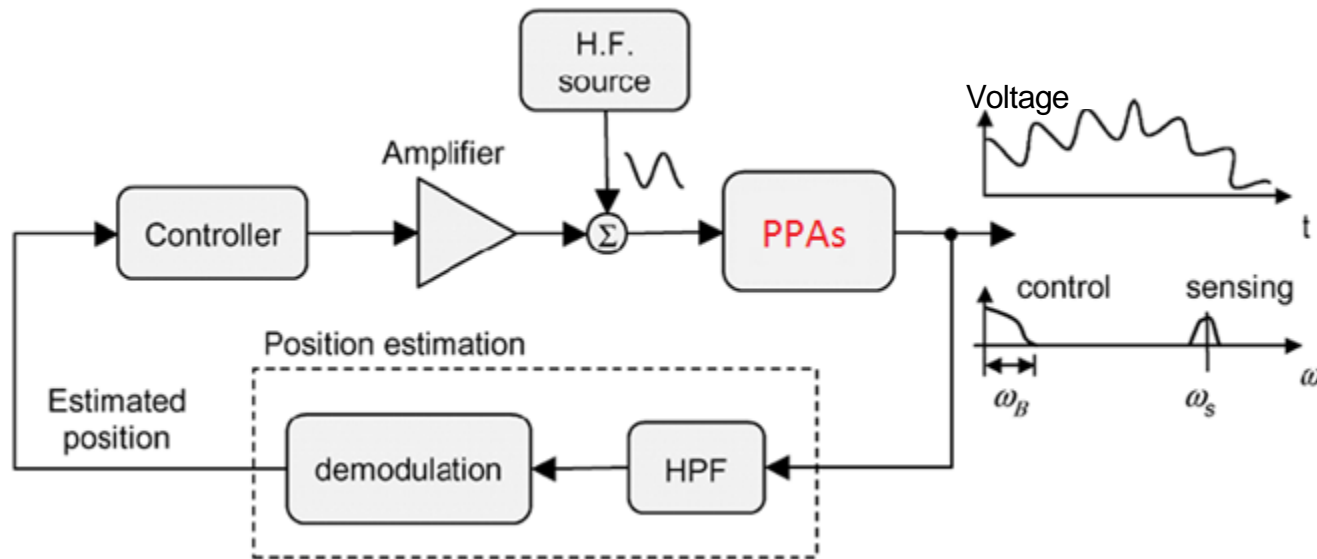
# External Sensing for PPAs (1)

- Additional sensing structures



# External Sensing for PPAs (2)

- Additional Excitation Source

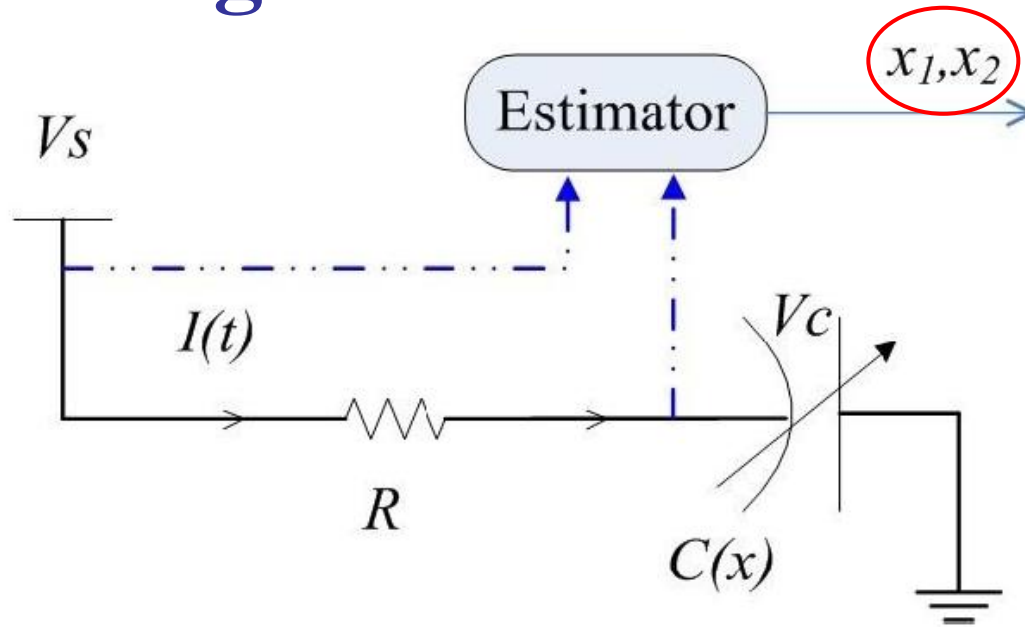


- It needs more electronics.
- May introduce harmonics.

## 2. Observability Analysis

- Nonlinear Observability Analysis
- Linearized Observability Analysis
- Observability Enhancement

# System Configuration

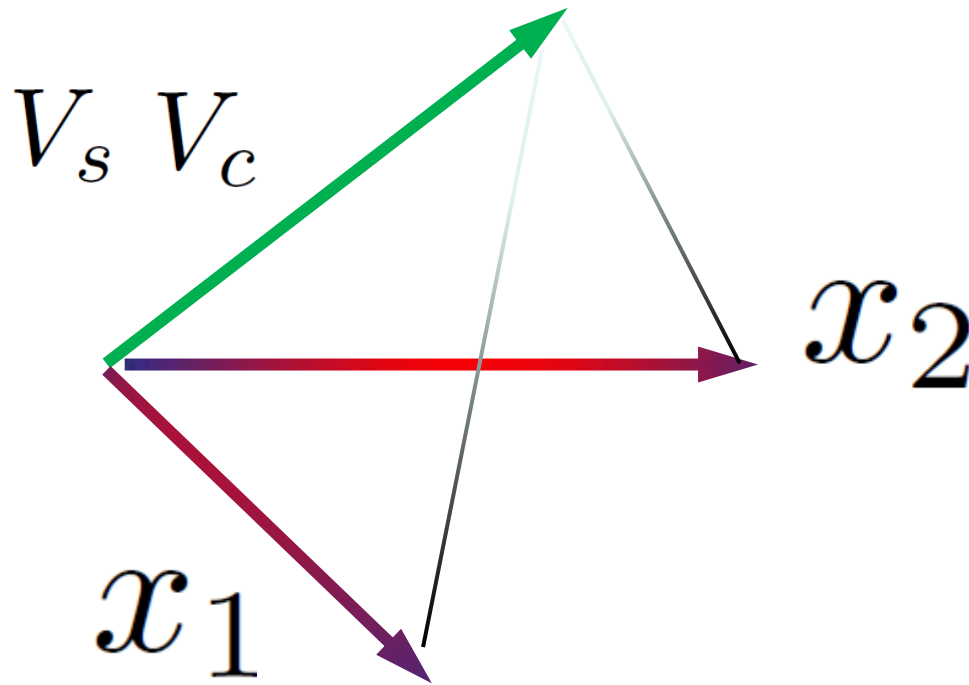


$$m\dot{x}_2 = -cx_2 - kx_1 + \frac{\epsilon_0\epsilon_r AV_c^2}{2(x_0 - x_1)^2}$$

$$\dot{V}_c = \frac{1}{c(x_1)} \left( -\frac{V_c}{R} - \frac{V_c\epsilon_0\epsilon_r A}{2(x_0 - x_1)^2} x_2 + \frac{V_s}{R} \right)$$

# Can we really recovery the information?

- Can we determine the position and velocity by measuring  $V_s$  and  $V_c$ ?
- Is it observable?



# Nonlinear Observability Criteria

- The given nonlinear system

$$\dot{x} = f(x, u)$$

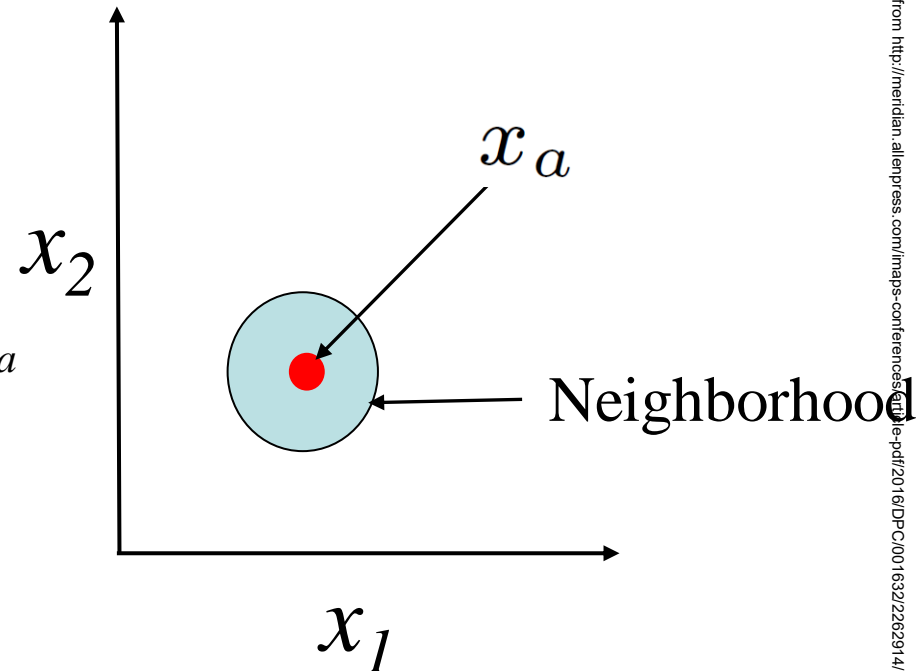
$$y = h(x).$$

is locally observable around  $x_a$   
if  $O_n$  has full rank

$$O_n = \left. \frac{\partial l(x)}{\partial x} \right|_{x=x_a}$$

where  $l(x) = \begin{bmatrix} \mathcal{L}_f^0 h \\ \mathcal{L}_f^1 h \\ \mathcal{L}_f^2 h \\ \vdots \\ \mathcal{L}_f^{n-1} h \end{bmatrix}$

and  $\mathcal{L}_f h$  is the Lie derivative



# Determining the Observability (1)

- Calculating  $O_n$  for the System:

$$O_n = \left. \frac{\partial l(x)}{\partial x} \right|_{x=x_a} = \begin{bmatrix} 0 & 0 & 1 \\ O_{n21} & O_{n22} & O_{n23} \\ O_{n31} & O_{n32} & O_{n33} \end{bmatrix}$$

$$O_{n21} = \frac{V_c - V_s}{R\epsilon_0\epsilon_r A} - \frac{V_c}{2(x_0 - x_1)^2} x_2$$

$$O_{n22} = -\frac{V_c}{2(x_0 - x_1)}$$

$$O_{n23} = -\frac{(x_0 - x_1)}{R\epsilon_0\epsilon_r A} - \frac{1}{2(x_0 - x_1)} x_2$$

# Determining the Observability (2)

$$O_{n31} = \frac{1}{4A^2(\varepsilon_0\varepsilon_r)^2mR^2(x_0 - x_1)^4} \\ (-3A^3(\varepsilon_0\varepsilon_r)^3R^2V_c^3 - 8m(V_c - V_s)(x_0 - x_1)^5 \\ + (2A^2(\varepsilon_0\varepsilon_r)^2R^2V_cx_0 - x_1) \\ (kx_0(x_0 - x_1) - x_2(c(-x_0 + x_1) + mx_2)))) \\ O_{n32} = \frac{m(4V_c - 3V_s)(x_0 - x_1)^2}{2A\varepsilon_0\varepsilon_rmR(x_0 - x_1)^2} \\ + \frac{A\varepsilon_0\varepsilon_rRV_c(c(x_0 - x_1) - mx_2)}{2A\varepsilon_0\varepsilon_rmR(x_0 - x_1)^2}$$

- Hard to tell the rank by calculating its determinant:

$$|O_n| = O_{21}O_{n32} - O_{n22}O_{n31};$$

# Linear Observability (1)

Linearize the system around a equilibrium point  $(x_{1e}, 0, V_{ce})$  and  $V_{ce} = V_s$

$$\delta \dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ -k/m + \frac{\varepsilon_0 \varepsilon_r D V_{ce}^2}{m(x_0 - x_{1e})^3} & -c/m & \frac{\varepsilon_0 \varepsilon_r D V_{ce}}{m(x_0 - x_{1e})^2} \\ 0 & -\frac{1}{2(x_0 - x_{1e})} V_{ce} & -\frac{x_0 - x_{1e}}{R \varepsilon_0 \varepsilon_r D} \end{bmatrix} \delta x$$

$$\delta \dot{x} = \begin{bmatrix} 0 & 1 & 0 \\ a_{21} & a_{22} & a_{23} \\ 0 & a_{32} & a_{33} \end{bmatrix} \delta x = A \delta x$$

# Linear Observability (2)

- The system is observable if and only if the following matrix has full rank

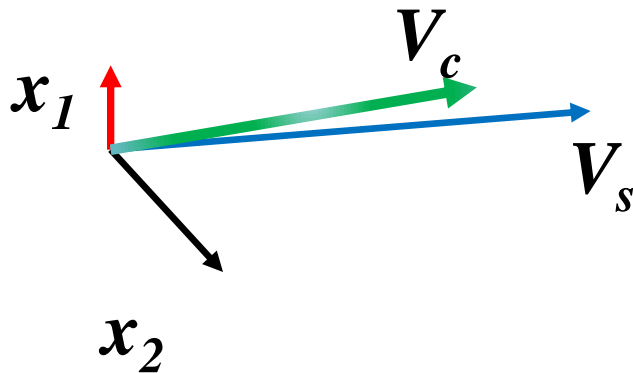
$$Q_o = \begin{bmatrix} 0 & 0 & 1 \\ 0 & a_{32} & a_{33} \\ a_{21}a_{32} & a_{22}a_{32} + a_{32}a_{33} & a_{33}^2 + a_{23}a_{32} \end{bmatrix}$$

- It has full rank if and only if  $|Q_o| = a_{21}a_{32}^2$  is not zero.
- The system is observable unless  $x_I = x_o/3$

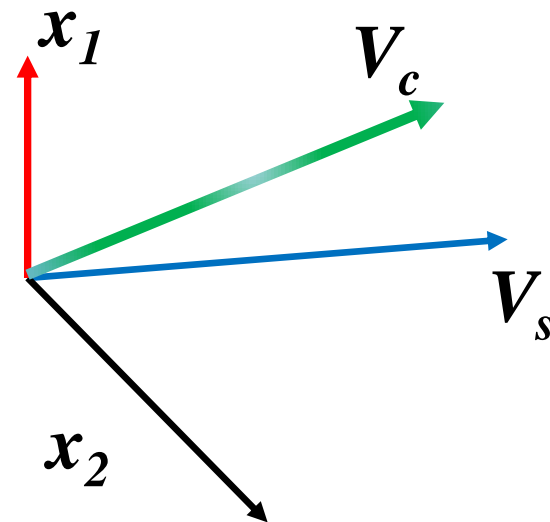
# Observable Quality (1)

- Though it's observable, the quality of the observability is uncertain.
- Can we make a reliable PPA to fulfill this technique?

Bad Observability



Good Observability



# Observable Quality (2)

- It's observable if and only if  $Q_o$  has full rank, but the quality of the observability is uncertain.

$$Q_o = \begin{bmatrix} 0 & 0 & 1 \\ 0 & a_{32} & a_{33} \\ a_{21}a_{32} & a_{22}a_{32} + a_{32}a_{33} & a_{33}^2 + a_{23}a_{32} \end{bmatrix}$$

- An example of a matrix that has full rank, but is weakly full rank:

$$\begin{bmatrix} 1 & 1 \\ 1 & 0.99 \end{bmatrix}$$

# Observability Enhancement (1)

- Using singular value decomposition (SVD) to analysis  $Q_o$ .

$$Q_o = U\Sigma V^T$$

- where  $\Sigma = \text{diag}(\sigma_1, \sigma_2, \sigma_3)$  is a diagonal matrix, and  $\sigma_1 \geq \sigma_2 \geq \sigma_3$  are the singular values of  $Q_o$ ; U and V are orthogonal matrices.

- Define the observability index as:

$$O_2 = \frac{\sigma_1}{\sigma_3}$$

- The observability is better when  $O_2$  becomes smaller.

# Observability Enhancement (2)

- Directly solving  $O_2$  is difficult.
- The upper bound of  $O_2$  is:

$$\frac{\|Q_o\|_F^3}{|Q_o|} \geq \frac{\sigma_1}{\sigma_3}$$

- where  $\|Q_o\|_F$  is the Frobenius norm of  $Q_o$  and

$$|Q_o| = \sigma_1 \sigma_2 \sigma_3$$

Thus, if  $|Q_o| = a_{21}a_{32}^2$  is increasing and  $\|Q_o\|_F$  is decreasing, the observability  $O_2$  will decrease, which means that actuator's observability is improved (less sensitive to noise and finite word length).

# Observability Enhancement (3)

- This observability can be improved if an element in  $G$  changes which satisfies:

$$\frac{\partial \|Q_o\|_F^3}{\partial a_{ij}} \leq \frac{\partial |Q_o|}{\partial a_{ij}}$$

- Larger series resistance, smaller initial gap and smaller spring constant will improve the observability.

# 3. Validation

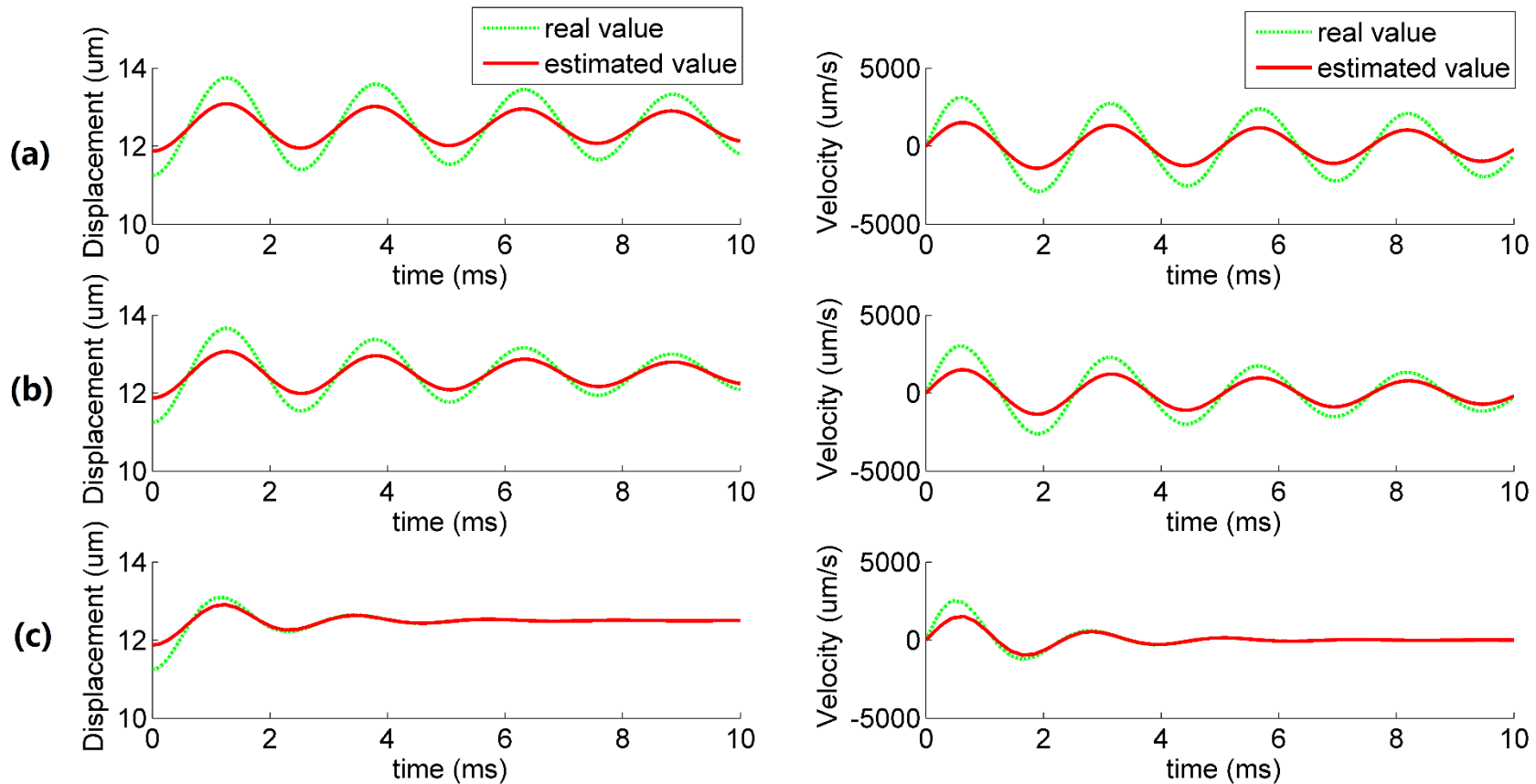
- Feasibility of Self-Sensing
- Performances Using Different Parameters

# System Parameters

- A simulation study was performed using the given parameters

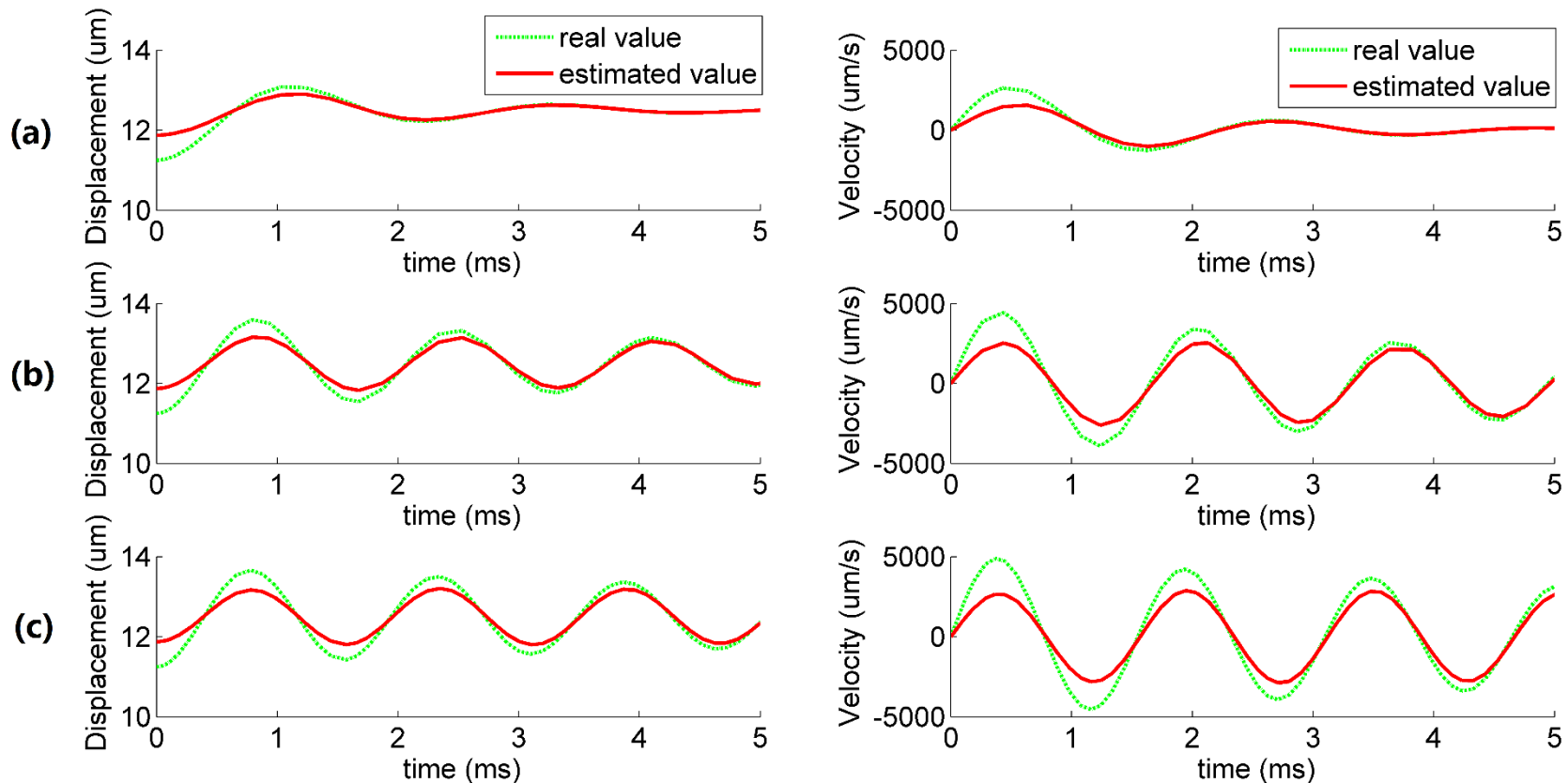
Symbol	Description	Value	Unit
k	spring constant	150	N/m
c	damping coefficient	$7.5 \times 10^{-4}$	N · s/m
m	proof mass	7.5	$\mu\text{g}$
A	overlap area of the electrodes	$2.9 \times 2.9$	mm
R	series resistance	100 M	$\Omega$
$x_0$	initial gap distance	10	$\mu\text{m}$
$\epsilon_0$	permittivity of free space	$8.854 \times 10^{-12}$	F/m
$\epsilon_r$	permittivity of air	1.0006	N/A

# Simulation Results (1)



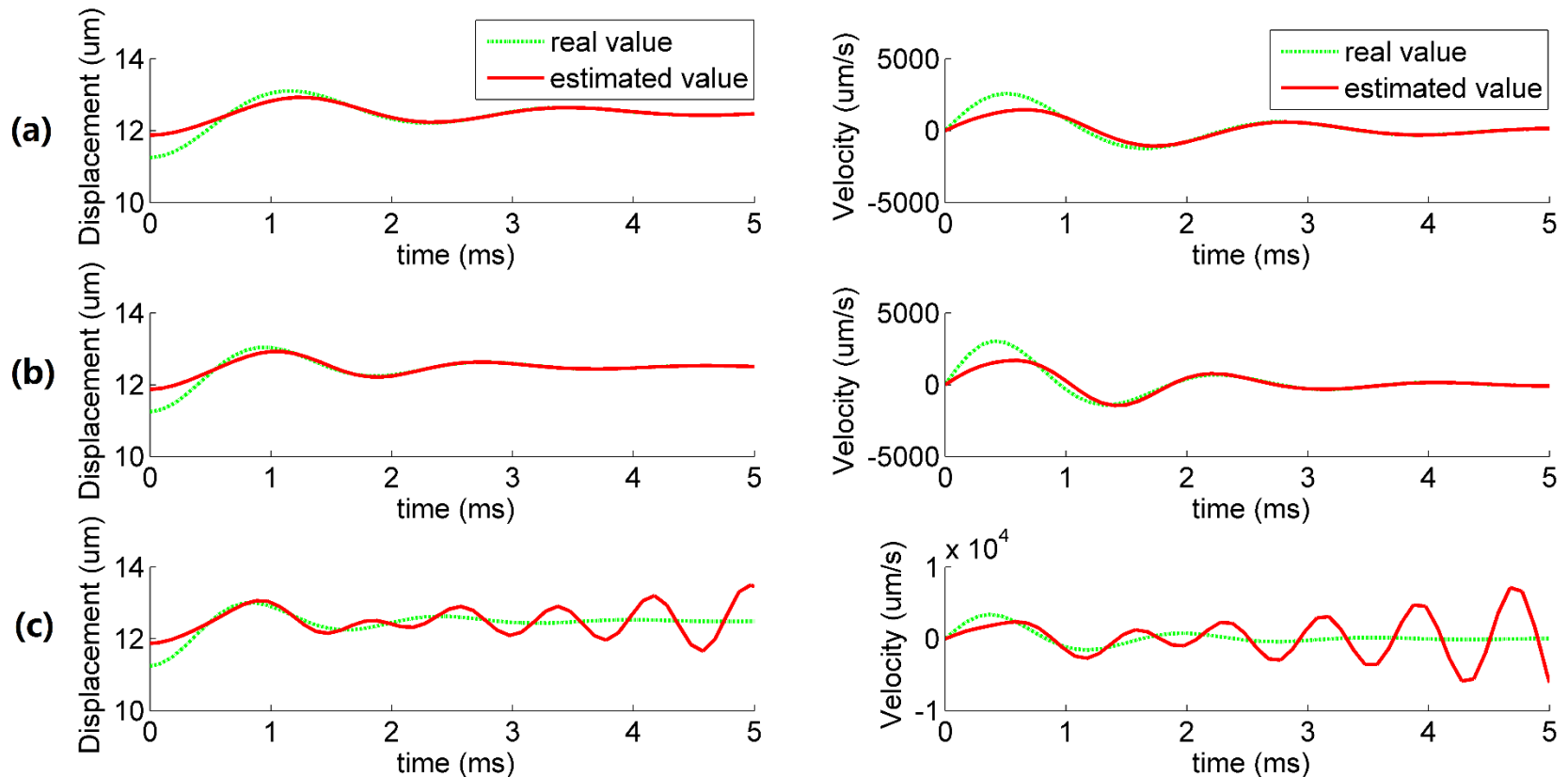
(a)  $R = 1\text{M } \Omega$ , (b)  $R = 10\text{M } \Omega$  and (c)  $R = 100\text{M } \Omega$

# Simulation Results (2)



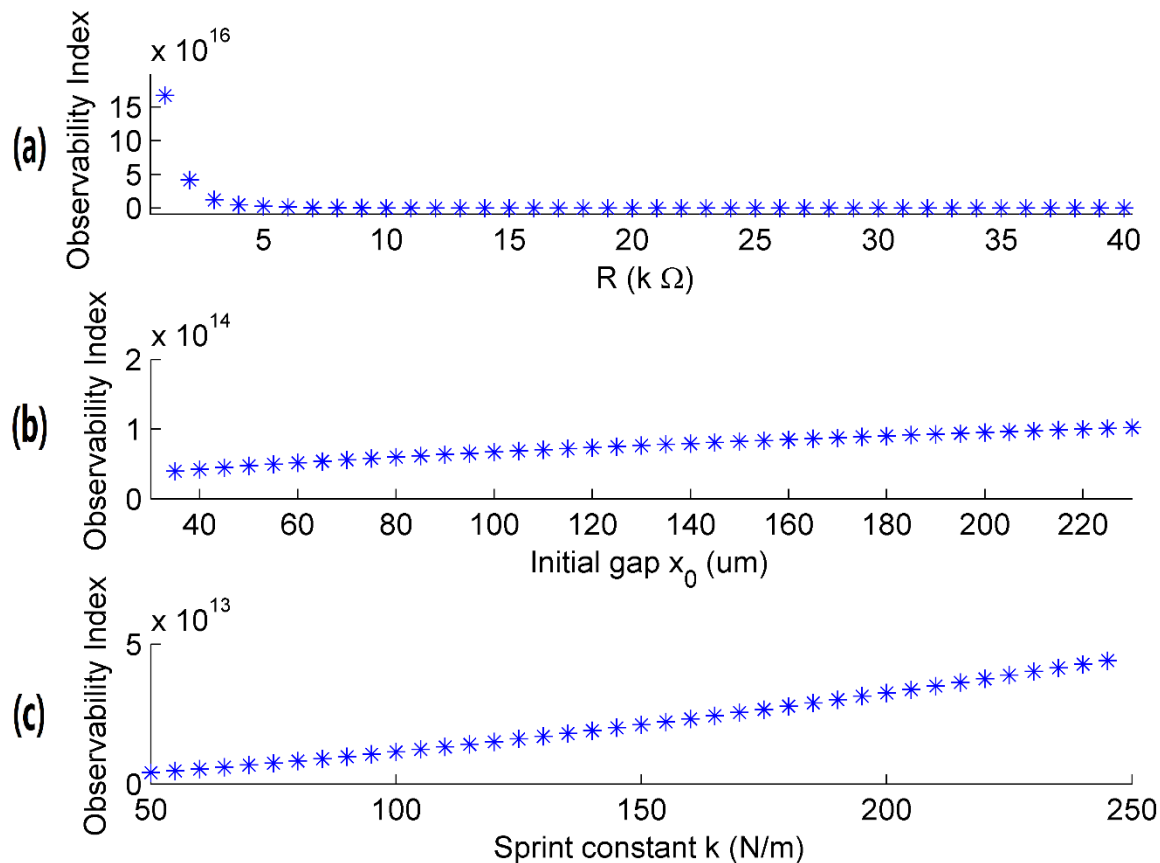
(a)  $x_0 = 50 \mu\text{m}$ , (b)  $x_0 = 100 \mu\text{m}$  and (c)  $x_0 = 150 \mu\text{m}$

# Simulation Results (3)



(a)  $k = 150$  N/m, (b)  $k = 200$  N/m and (c)  $k = 250$  N/m

# Simulation Results (4)



## 4. Conclusions

- The self-sensing technique to estimate the mechanical states of a class of MEMS electrostatic actuators is feasible when a reasonably soft spring exists and the displacement is nonzero.
- Variation of the actuators' parameters also affect the observability. This technique is improved with a larger series resistance and a smaller initial gap.

# Thank You for Your Attention.

## Questions?