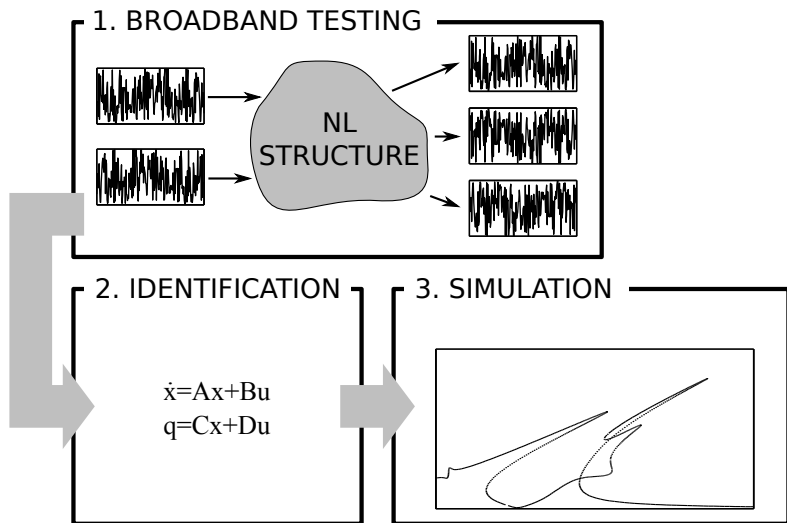


# Nonlinear frequency response curve from broadband testing

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# Three step procedure



# Feedback interpretation of nonlinearities

## General nonlinear system

$$M\ddot{q}(t) + C\dot{q}(t) + Kq(t) + f(q(t), \dot{q}(t)) = p(t)$$

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$$f(q(t), \dot{q}(t)) = \sum_{j=1}^s \mu_j L_j g_j(q(t), \dot{q}(t))$$

# Feedback interpretation of nonlinearities

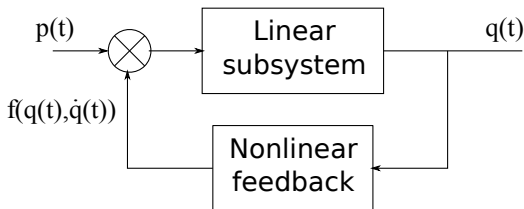
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## Feedback interpretation



# From physical to state-space domain

## State and extended input vectors

$$x(t) = \begin{bmatrix} q(t) \\ \dot{q}(t) \end{bmatrix}, \quad e(p(t), q(t), \dot{q}(t)) = \begin{bmatrix} p(t) \\ g_1(q(t), \dot{q}(t)) \\ \vdots \\ g_s(q(t), \dot{q}(t)) \end{bmatrix}$$

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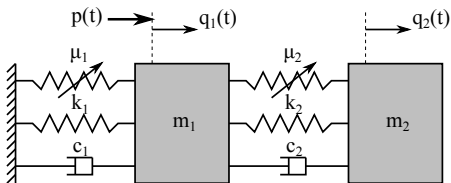
## Nonlinear state space system

$$\begin{aligned} \dot{x}(t) &= A_c x(t) + B_c e(p(t), q(t), \dot{q}(t)) \\ q(t) &= C x(t) + D e(p(t), q(t), \dot{q}(t)) \end{aligned}$$

$$A_c = \begin{bmatrix} 0 & I \\ -M^{-1}K & -M^{-1}C \end{bmatrix}, \quad C = [I \quad 0]$$

$$B_c = \begin{bmatrix} 0 & 0 & \dots & 0 \\ M^{-1}L_f & -\mu_1 M^{-1}L_1 & \dots & -\mu_s M^{-1}L_s \end{bmatrix}, \quad D = 0$$

# Identification statement

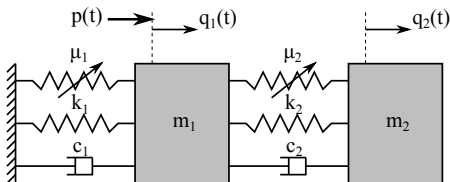


$$e(t) = \begin{bmatrix} p(t) \\ q_1(t)^3 \\ (q_1(t) - q_2(t))^3 \end{bmatrix}, L_f = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$L_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, L_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$



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## Given

- Measured output  $q(t)$
- Measured input  $p(t)$
- Set of basis function  $g_j(q(t), \dot{q}(t))$

## Determine

- Order of the system  $n$
- System matrices  $A_c, B_c, C, D$
- Nonlinear coefficients  $\mu_j$

# Subspace identification method

## Discrete time translation

$$\begin{aligned}x(k+1) &= A_d x(k) + B_d e(k) \\ q(k) &= C x(k) + D e(k)\end{aligned}$$

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$$\begin{aligned}x(k+1) &= A_d x(k) + B_d e(k) \\ q(k) &= Cx(k) + De(k)\end{aligned}$$

## Block Hankel matrices

$$Q_i = \begin{array}{c} \left. \begin{array}{c} \uparrow \\ ri \\ \downarrow \end{array} \right\} \begin{array}{c} \left[ \begin{array}{cccc} q(1) & q(2) & \dots & q(j) \\ q(2) & q(3) & \dots & q(j+1) \\ \vdots & \vdots & \ddots & \vdots \\ q(i) & q(i+1) & \dots & q(i+j-1) \end{array} \right] \end{array}$$

$\xleftarrow{j}$

$$E_i = \begin{bmatrix} e(1) & e(2) & \dots & e(j) \\ e(2) & e(3) & \dots & e(j+1) \\ \vdots & \vdots & \ddots & \vdots \\ e(i) & e(i+1) & \dots & e(i+j-1) \end{bmatrix}$$

# Input-State-Output equation

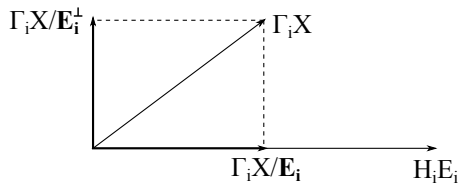
$$Q_i = \Gamma_i X + H_i E_i$$

with

$$\Gamma_i = [C^T \quad (CA_d)^T \quad (CA_d^2)^T \quad \dots \quad (CA_d^{i-1})^T]^T$$
$$H_i = \begin{bmatrix} D & 0 & 0 & \dots & 0 \\ CB_d & D & 0 & \dots & 0 \\ CA_d B_d & CB_d & D & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ CA_d^{i-2} B_d & CA_d^{i-3} B_d & CA_d^{i-4} B_d & \dots & D \end{bmatrix}$$

# Projection algorithm

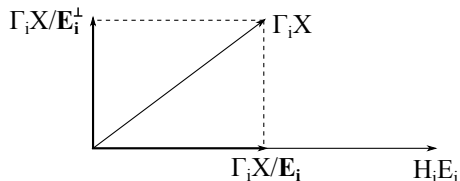
## Orthogonal projection



$$Q_i = \Gamma_i X + H_i E_i$$
$$Q_i / E_i^\perp = \Gamma_i X / E_i^\perp = \mathcal{O}_i$$

# Projection algorithm

## Orthogonal projection



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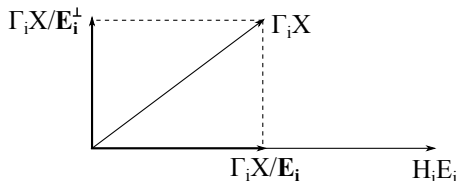
$$Q_i / E_i^\perp = \Gamma_i X / E_i^\perp = \mathcal{O}_i$$

## Singular value decomposition

$$\mathcal{O}_i = [U_1 \quad U_2] \begin{bmatrix} S_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} V_1^T \\ V_2^T \end{bmatrix}$$

# Projection algorithm

## Orthogonal projection



$$Q_i = \Gamma_i X + H_i E_i$$

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## Singular value decomposition

$$\mathcal{O}_i = [U_1 \quad U_2] \begin{bmatrix} S_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} V_1^T \\ V_2^T \end{bmatrix}$$

$$\begin{cases} \Gamma_i = U_1 S_1^{1/2} T \\ X / \mathbf{E}_i^\perp = T^{-1} S_1^{1/2} V_1^T \end{cases}$$

# Extraction of system matrices

$A_d$  and  $C$

$$\Gamma_i = \begin{bmatrix} \hat{C} \\ \hat{C}\hat{A}_d \\ \hat{C}\hat{A}_d^2 \\ \vdots \\ \hat{C}\hat{A}_d^{i-1} \end{bmatrix}, \quad \underline{\Gamma}_i = \hat{A}_d \overline{\Gamma}_i$$

$$A_d = T \hat{A}_d T^{-1}, \quad B_d = T \hat{B}_d, \quad C = \hat{C} T^{-1}, \quad D = \hat{D}$$

$B_d$  and  $D$

$$Q_i = \Gamma_i X + H_i E_i$$

$$\Gamma_i^\perp Q_i = \Gamma_i^\perp H_i E_i$$



# Extraction of system matrices

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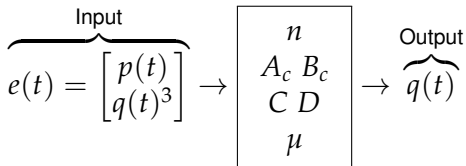
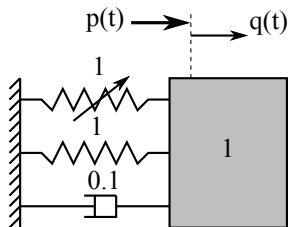
$$\Gamma_i^\perp Q_i = \Gamma_i^\perp H_i E_i$$

## Nonlinear coefficients (Optional)

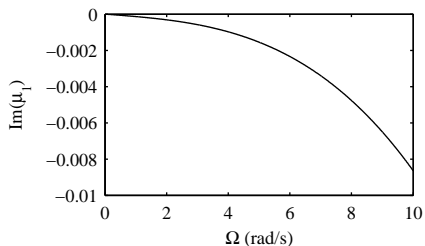
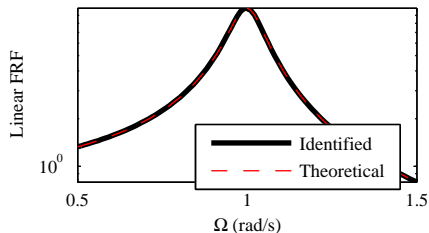
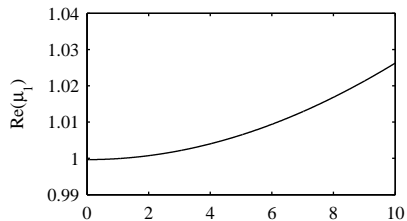
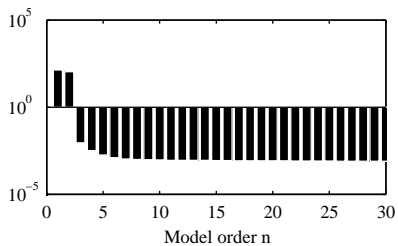
$$H_E(\omega) = \underbrace{\hat{D} + \hat{C}(j\omega I - \hat{A}_c)^{-1} \hat{B}_c}_{\text{Invariant with } T} = \begin{bmatrix} H & -H\mu_1 L_1 & \dots & -H\mu_s L_s \end{bmatrix}$$

$H$ : FrF of underlying linear system

# Application : Duffing oscillator



# Result of identification



# Harmonic balance method

## Continuation in state space domain

$$\begin{aligned}\dot{x}(t) &= A_c x(t) + B_c e(p(t), q(t), \dot{q}(t)) \\ q(t) &= C x(t) + D e(p(t), q(t), \dot{q}(t))\end{aligned}$$

# Harmonic balance method

## Continuation in state space domain

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## Truncated Fourier series expansion

$$x(t) = \frac{X_0}{\sqrt{2}} + \sum_{j=1}^N X_{cj} \cos(k_j \Omega t) + X_{sj} \sin(k_j \Omega t)$$

$$q(t) = \frac{Q_0}{\sqrt{2}} + \sum_{j=1}^N Q_{cj} \cos(k_j \Omega t) + Q_{sj} \sin(k_j \Omega t) \quad \text{Unknown Fourier coeffs } Q$$

$$e(t) = \frac{E_0}{\sqrt{2}} + \sum_{j=1}^N E_{cj} \cos(k_j \Omega t) + E_{sj} \sin(k_j \Omega t) \quad \text{Fourier coeffs } E \text{ Depend on } Q$$

# Harmonic balance method

## Residue equation

$$h(Q, \Omega) \equiv \underbrace{Q - G(\Omega)E(Q)}_{\text{Invariant with } T} = 0$$

with

$$G(\Omega) = (I \otimes C)\Lambda^{-1}(I \otimes B_c) + (I \otimes D)$$

$$\Lambda = \Omega(\nabla \otimes I) - (I \otimes A_c)$$

# Harmonic balance method

## Residue equation

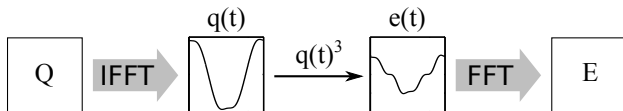
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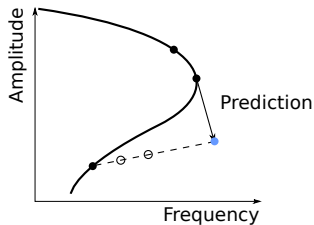
$$\Lambda = \Omega(\nabla \otimes I) - (I \otimes A_c)$$

## Fourier coefficients of nonlinear terms



# Tracking periodic solutions

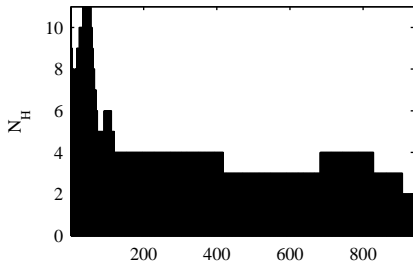
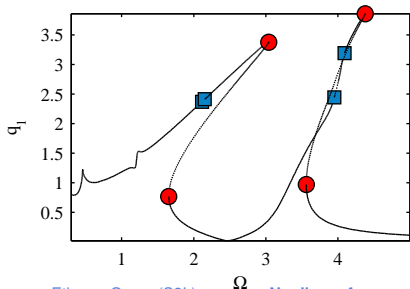
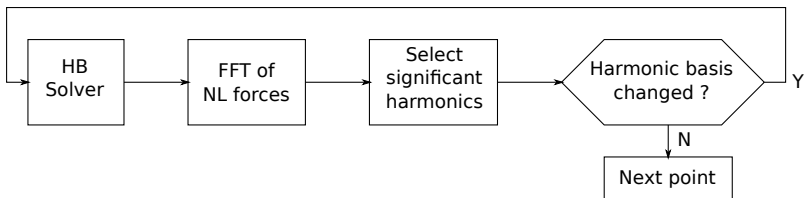
## Pseudo-arclength continuation



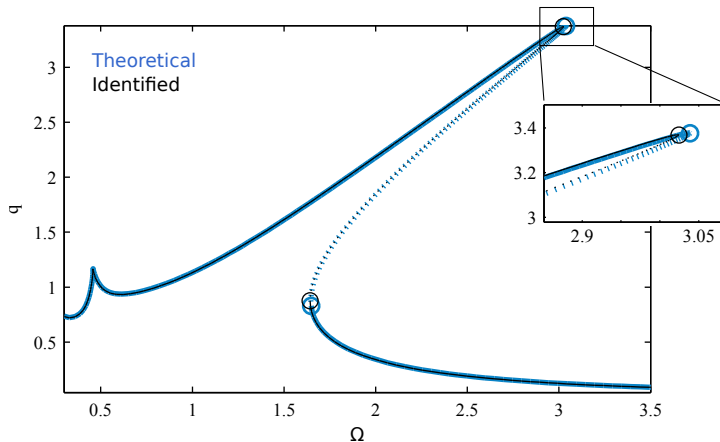
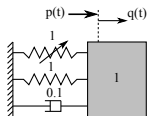


# Robust implementation

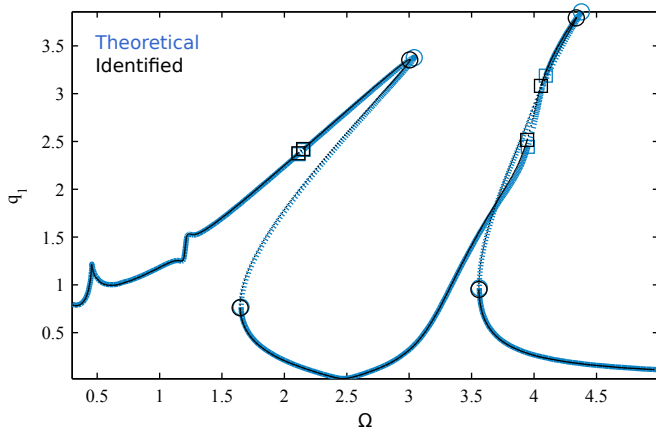
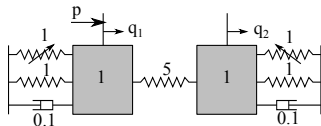
## Automatic harmonic selection



# Application 1 : single dof Duffing



# Application 2 : two dof Duffing



## Conclusion

- Nonlinear subspace identification + Continuation
- Good results on simple systems

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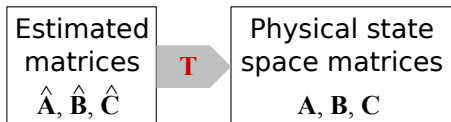
- Nonlinear subspace identification + Continuation
- Good results on simple systems

## Perspectives

- Effect of noise on the detection of bifurcation
- Experimental validation

# Error identification

## Going back to physics



$$T \begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix} = \begin{bmatrix} 0 & I \\ -M^{-1}K & -M^{-1}C \end{bmatrix} T$$
$$[C_1 \quad C_2] = [I \quad 0] T$$

This gives

$$T = \begin{bmatrix} \hat{\mathbf{C}} \\ \hat{\mathbf{C}}\hat{\mathbf{A}} \end{bmatrix}$$

# Identified matrices in "physical space"

$$A_d = T\hat{A}_d T^{-1}, \quad B_d = T\hat{B}_d, \quad C = \hat{C}T^{-1}, \quad D = \hat{D}$$

Uncorrect  $B_c$  matrix

$$B_c = \begin{bmatrix} B_1 & B_2 \\ M^{-1}L_f & -M^{-1}L_1 \end{bmatrix}$$

This gives for the Duffing

$$\ddot{q} + \frac{c}{m}\dot{q} + \frac{k}{m}q + \frac{\mu}{m}q^3 - B_2\dot{q}q^2 = \frac{p}{m} + B_1\dot{p}$$

