

**How to Reject Unknown Disturbances Using
Adaptive Feedback Control**
Application to Active Vibration Control

I.D. Landau, A. Constantinescu
Laboratoire d'Automatique de Grenoble(INPG/CNRS), France

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Outline

- Rejection of unknown stationary disturbances
- Indirect adaptive control
- Direct adaptive control
- Rejection of unknown narrow band disturbances in active vibration control. Real-time results
- Conclusions

Unknown disturbance rejection

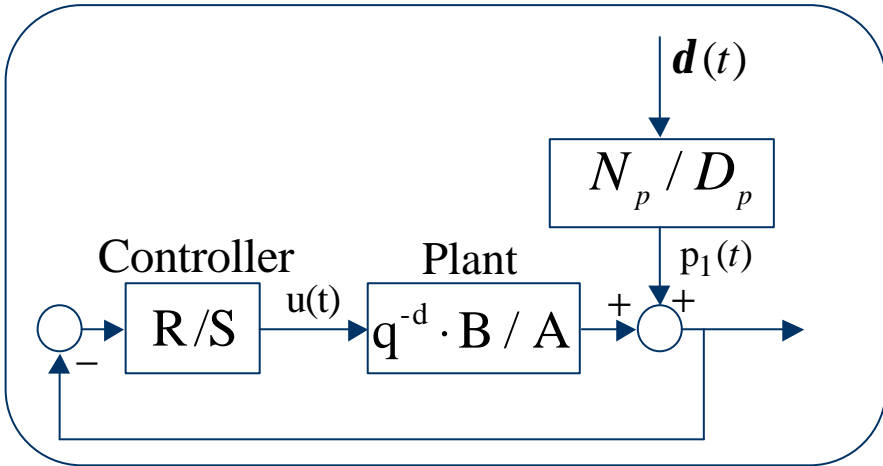
- **Problem:** Attenuation of unknown and/or variable stationary disturbances
- **Solution:** Adaptive feedback control
- **Methodology:** Based on the *Internal Model Principle*
 - Indirect adaptive control algorithm
 - Direct adaptive control algorithm
- **Objective:** Computation of a controller with an adaptive internal model of the disturbance

Rem:

Stationary disturbances models have poles on the unit circle

Hypothesis: Plant model parameters are constant and known

Closed loop system. Notations



$$p_1(t) = \frac{N_p(q^{-1})}{D_p(q^{-1})} \cdot \mathbf{d}(t) : \text{deterministic disturbance}$$

$D_p \rightarrow$ poles on the unit circle; $d(t) = \text{Dirac}$

Controller:

$$R(q^{-1}) = R'(q^{-1}) \cdot H_R(q^{-1});$$

$$S(q^{-1}) = S'(q^{-1}) \cdot H_S(q^{-1}).$$

Internal model principle: $H_S(z^{-1}) = D_p(z^{-1})$

$$\text{Output: } y(t) = \frac{A(q^{-1})S(q^{-1})}{P(q^{-1})} \cdot p_1(t) = S_{yp}(q^{-1}) \cdot p_1(t)$$

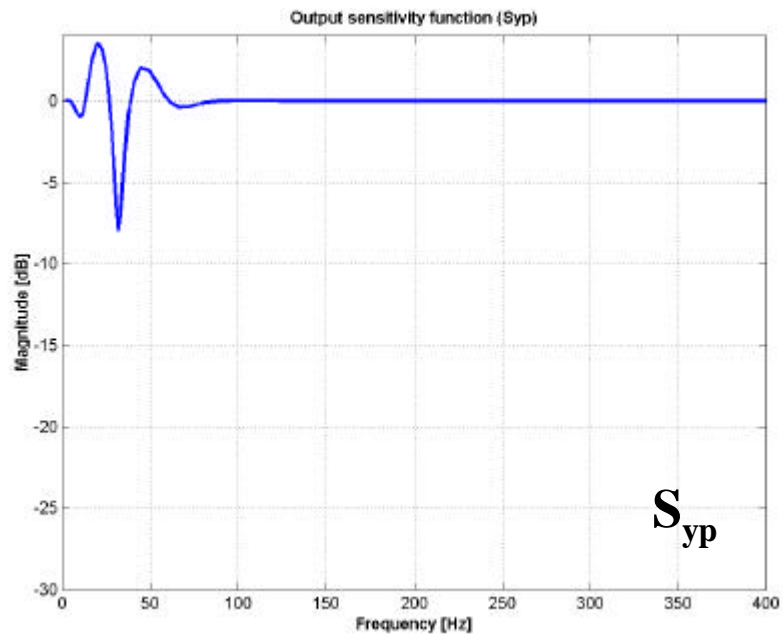
$$\text{CL poles: } P(q^{-1}) = A(q^{-1})S(q^{-1}) + z^{-d}B(q^{-1})R(q^{-1})$$

$$y(t) = \frac{A(q^{-1})H_S(q^{-1})S'(q^{-1})N_p(q^{-1})}{P(q^{-1})} \cdot \frac{1}{D_p(q^{-1})} \cdot \mathbf{d}(t)$$

Internal Model Principle: $H_S(z^{-1})=D_p(z^{-1})$

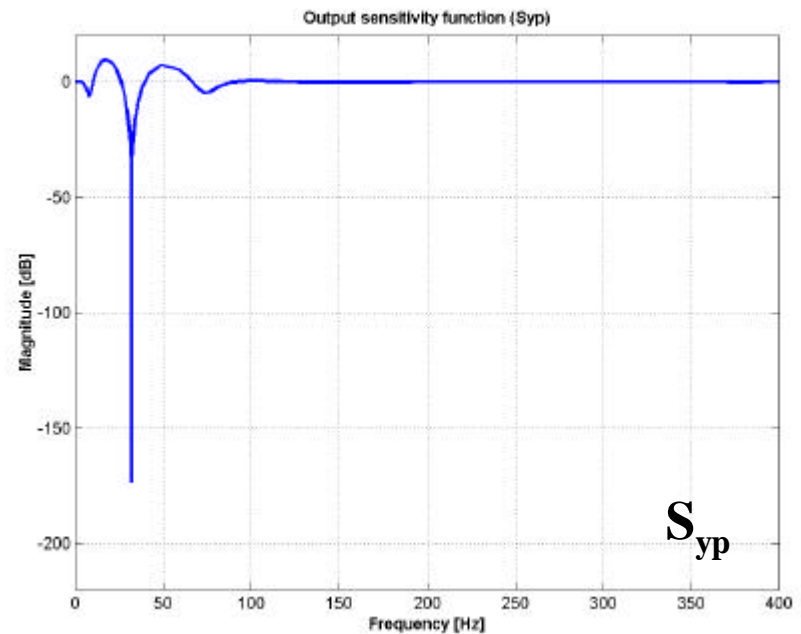
Controller without internal model

$$H_S=1$$



Controller with internal model at 32 Hz

$$H_S: (f=32 \text{ Hz}, \zeta=0)$$



Indirect adaptive control

Two-step methodology:

1. Identification of the disturbance model, $D_p(q^{-1})$
2. Computation of the controller, considering $H_s(q^{-1}) = \hat{D}_p(q^{-1})$

Indirect adaptive control

Example : Sinusoidal (narrow band) disturbance

Step I : Estimation of D_p

Sinusoid \rightarrow ARMA with $n_{D_p} = 2$: $D_p(q^{-1}) = 1 + d_{p_1}q^{-1} + d_{p_2}q^{-2}$;
 $N_p(q^{-1}) = 1 + n_{p_1}q^{-1}$.

Output : $y(t+1) = -d_{p_1}y(t) - d_{p_2}y(t-1) + n_{p_1}e(t) + e(t+1)$,
 where $e(t) =$ gaussian white noise.

A priori predictor : $\hat{y}^0(t+1) = \hat{\mathbf{q}}^T(t) \cdot \mathbf{f}(t)$; *A posteriori* predictor : $\hat{y}(t+1) = \hat{\mathbf{q}}^T(t+1) \cdot \mathbf{f}(t)$,

where $\hat{\mathbf{q}}^T(t) = [\hat{d}_{p_1}(t) \quad \hat{d}_{p_2}(t) \quad \hat{n}_{p_1}(t)]$;

$\mathbf{f}^T(t) = [-y(t) \quad -y(t-1) \quad \mathbf{e}(t)]$.

A priori error: $\mathbf{e}^0(t+1) = y(t+1) - \hat{y}^0(t+1)$; *A posteriori* error: $\mathbf{e}(t+1) = y(t+1) - \hat{y}(t+1)$

Parametric adaptation algorithm \rightarrow
$$\begin{cases} \hat{\mathbf{q}}(\mathbf{t}+1) = \hat{\mathbf{q}}(\mathbf{t}) + \mathbf{F}(\mathbf{t}+1)\mathbf{f}(\mathbf{t})\mathbf{e}^0(\mathbf{t}+1); \\ \mathbf{F}(\mathbf{t}+1) = \mathbf{adaptation\ gain}. \end{cases}$$

Indirect adaptive control

Step II: Computation of the controller

Solving Bezout equation (for S' and R)

$$H_S = \hat{D}_p$$

$$A\hat{D}_p S' + q^{-d} BR = P$$

$$S = D_p S'$$

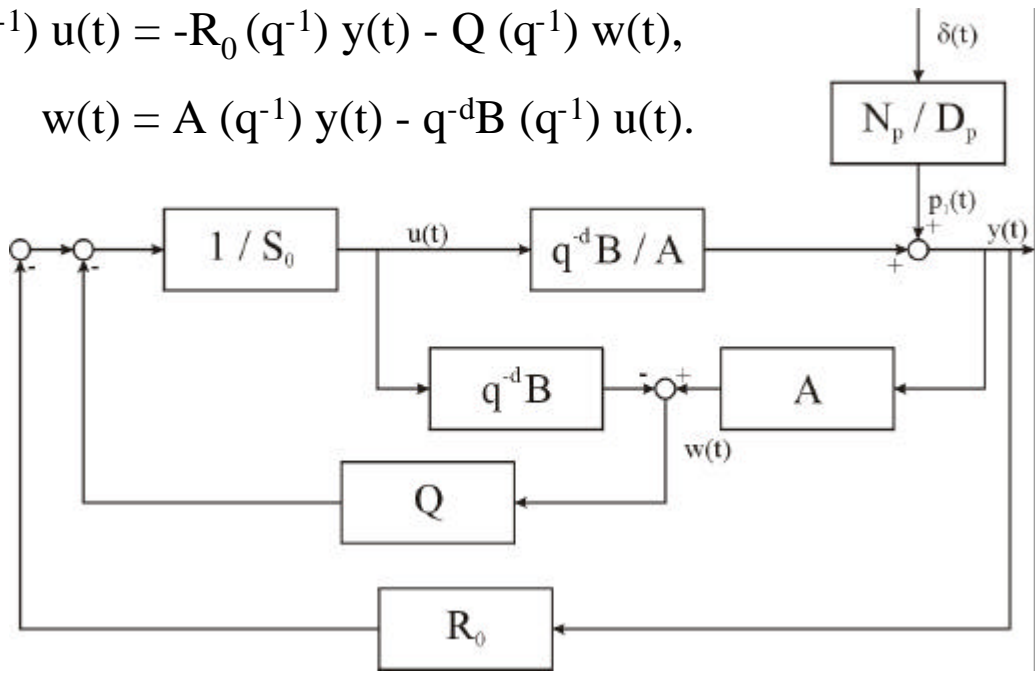
Direct adaptative control (Q-parameterization)

Nominal contr: $[R_0(q^{-1}), S_0(q^{-1})]$.
 Bezout: $P(q^{-1}) = A(q^{-1})S_0(q^{-1}) + q^{-d}B(q^{-1})R_0(q^{-1})$.
 Control: $S_0(q^{-1}) u(t) = -R_0(q^{-1}) y(t)$

Q-parameterization :
 $R(z^1) = R_0(q^{-1}) + A(q^{-1})Q(q^{-1});$
 $S(q^{-1}) = S_0(z^{-1}) - q^{-d}B(q^{-1})Q(q^{-1}).$
 $Q(q^{-1})$ computed such as $[R(q^{-1}), S(q^{-1})]$
contain the internal model of the disturb.

$$S_0 - q^{-d} BQ = MD_p$$

Control: $S_0(q^{-1}) u(t) = -R_0(q^{-1}) y(t) - Q(q^{-1}) w(t),$
 where $w(t) = A(q^{-1}) y(t) - q^{-d}B(q^{-1}) u(t).$



Direct Adaptive Control (unknown D_p)

(Based on an idea of Y. Z. Tsytkin)

Hypothesis: Identified (known) plant model (A,B,d).

Goal: minimize $y(t)$.

Consider $p_1(t) = \frac{N_p(q^{-1})}{D_p(q^{-1})} \cdot \mathbf{d}(t)$: deterministic disturbance.

$$y(t) = \frac{A(q^{-1}) [S_0(q^{-1}) - q^{-d} B(q^{-1}) Q(q^{-1})]}{P(q^{-1})} \cdot \frac{N_p(q^{-1})}{D_p(q^{-1})} \cdot \mathbf{d}(t)$$

⇓

$$\mathbf{e}(t) = \frac{S_0(q^{-1})}{P(q^{-1})} \cdot w(t) - \frac{q^{-d} B(q^{-1})}{P(q^{-1})} Q(q^{-1}) \cdot w(t).$$

Let $\hat{Q}(t, q^{-1})$ be an estimated value of $Q(q^{-1})$

We can show that

$$\mathbf{e}(t+1) = [Q(q^{-1}) - \hat{Q}(t+1, q^{-1})] \cdot \frac{q^{-d} B^*(q^{-1})}{P(q^{-1})} \cdot w(t) + v(t+1)$$

($v(t+1)$ = disturbance term $\rightarrow 0$)

The Algorithm

A priori adaptation error: $\mathbf{e}^0(t+1) = w_1(t+1) - \hat{\mathbf{q}}^T(t)\mathbf{f}(t);$

A posteriori adaptation error: $\mathbf{e}(t+1) = w_1(t+1) - \hat{\mathbf{q}}^T(t+1)\mathbf{f}(t),$

where

$$\hat{\mathbf{q}}^T(t) = [\hat{q}_0(t) \quad \hat{q}_1(t)]; \quad \mathbf{f}^T(t) = [w_2(t) \quad w_2(t-1)], \quad (\text{for } n_Q = 1)$$

and

$$w_1(t+1) = \frac{S_0(q^{-1})}{P(q^{-1})} \cdot w(t+1);$$

$$w_2(t) = \frac{q^{-d} B^*(q^{-1})}{P(q^{-1})} \cdot w(t);$$

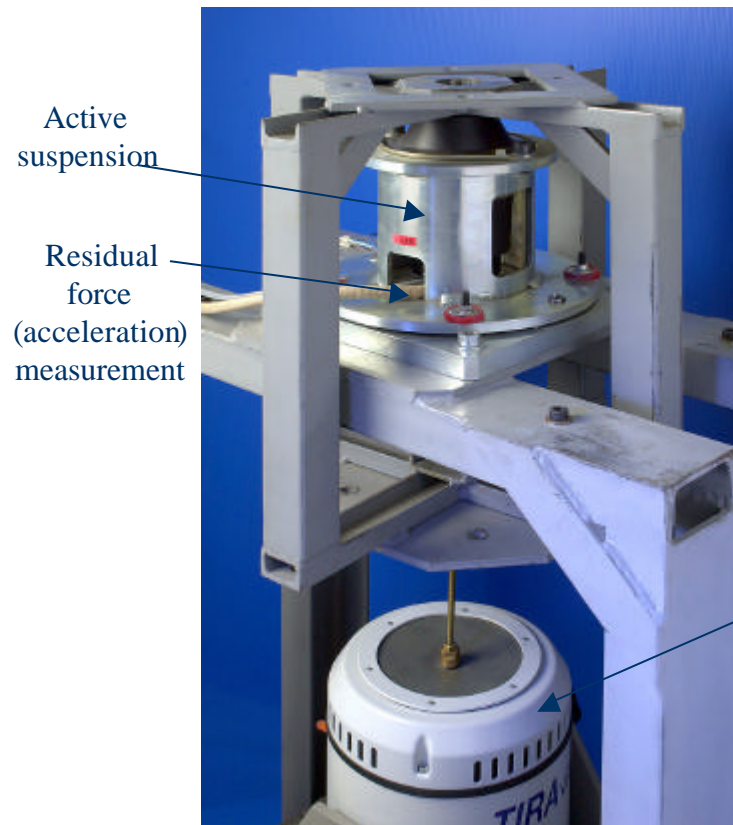
$$w(t+1) = A(q^{-1}) \cdot y(t+1) - q^{-d} B^*(q^{-1}) \cdot u(t);$$

$$B(q^{-1}) \cdot u(t+1) = B^*(q^{-1}) \cdot u(t).$$

Parameter adaptation algorithm:

$$\begin{cases} \hat{\mathbf{q}}(t+1) = \hat{\mathbf{q}}(t) + \mathbf{F}(t+1)\mathbf{f}(t)\mathbf{e}^0(t+1); \\ \mathbf{F}^{-1}(t+1) = \mathbf{I}_1(t)\mathbf{F}^{-1}(t) + \mathbf{I}_2(t)\mathbf{f}(t)\mathbf{f}^T(t). \end{cases}$$

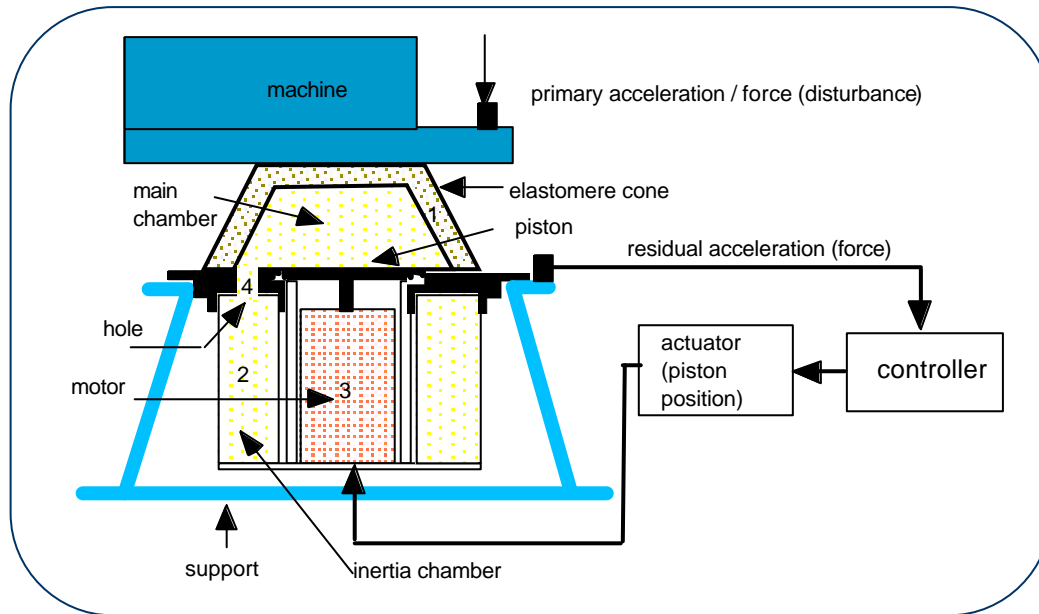
The Active Suspension



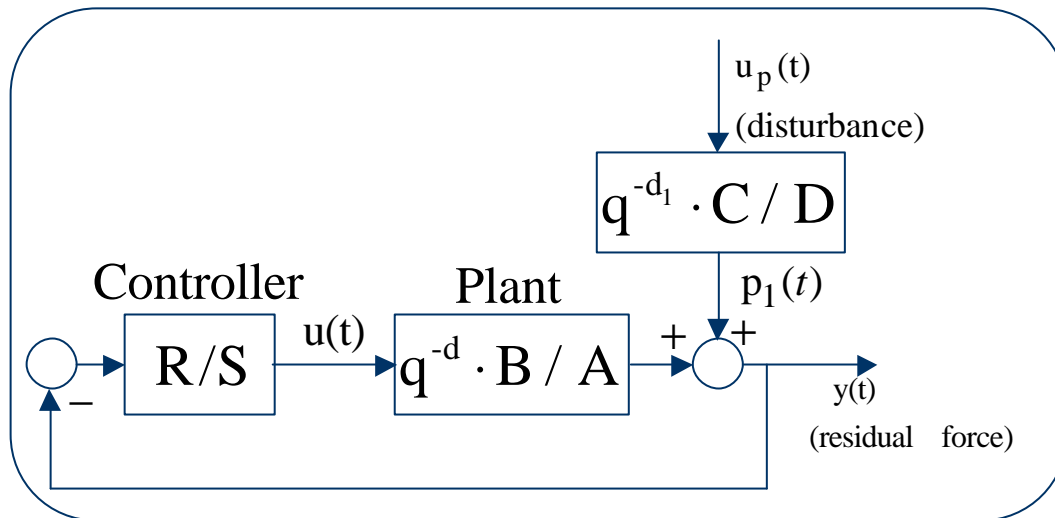
Primary force (acceleration) (the shaker)



The Active Suspension System



- Two paths :
- Primary
 - Secondary (double differentiator)

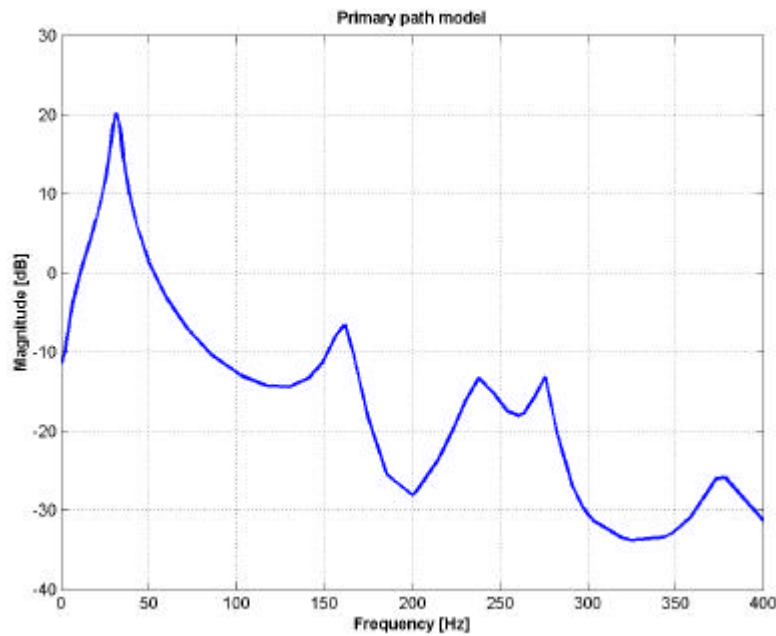


$$T_s = 0.00125 \text{ s}$$

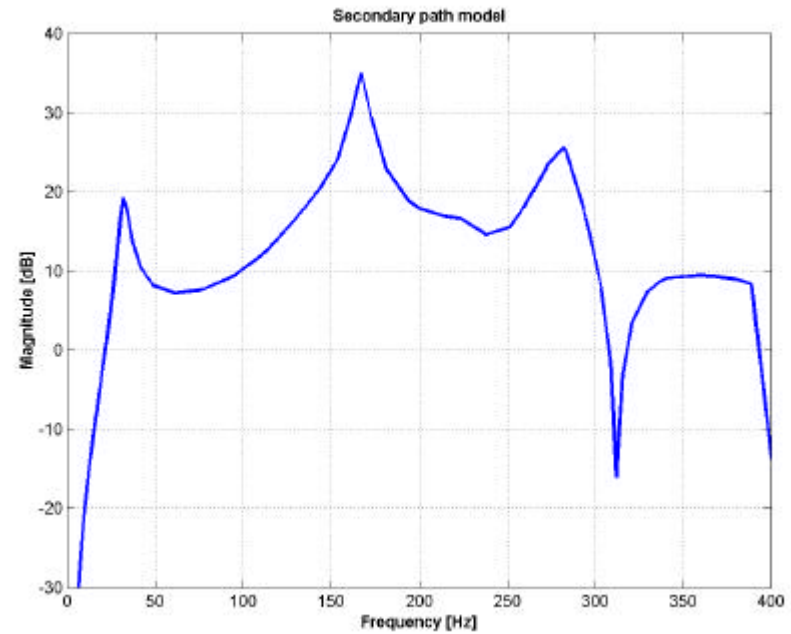
Active Suspension

Frequency Characteristics of the Identified Models

Primary path



Secondary path



$$n_A = 14; n_B = 16; d = 0$$

V. Real-time results

*Narrow band disturbances = variable frequency sinusoid \mathbf{P} $n_Q = 1$
Frequency range: 25 , 47 Hz*

Evaluation of the two algorithms in **real-time**

Nominal controller $[R_0(q^{-1}), S_0(q^{-1})]$: $n_{R0}=14$, $n_{S0}=16$

Implementation protocol 1: Self-tuning

- The algorithm stops when it converges and the controller is applied.
- It restarts when the variance of the residual force is bigger than a given threshold.
- As long as the variance is not bigger than the threshold, the controller is constant.

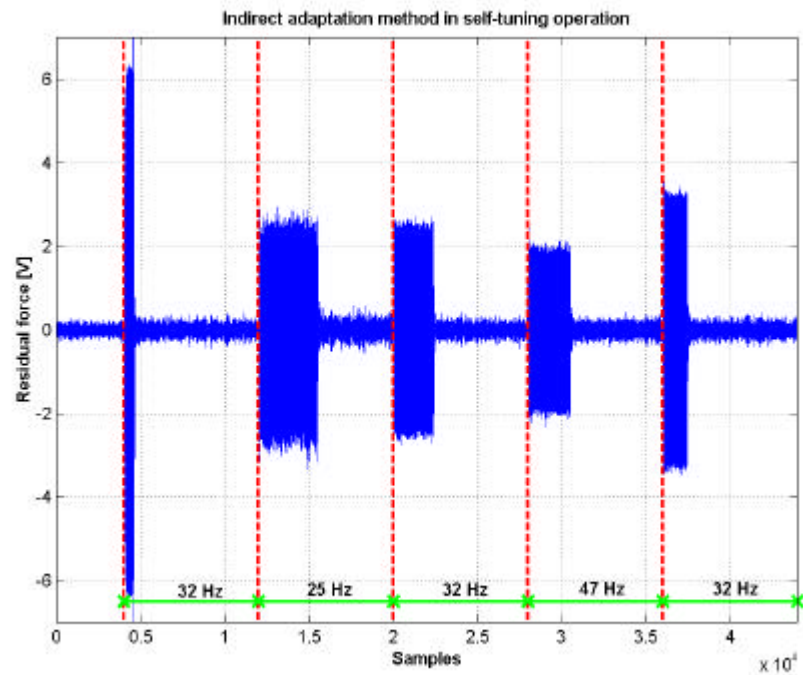
Implementation protocol 2: Adaptive

- The adaptation algorithm is continuously operating
- The controller is updated at each sample

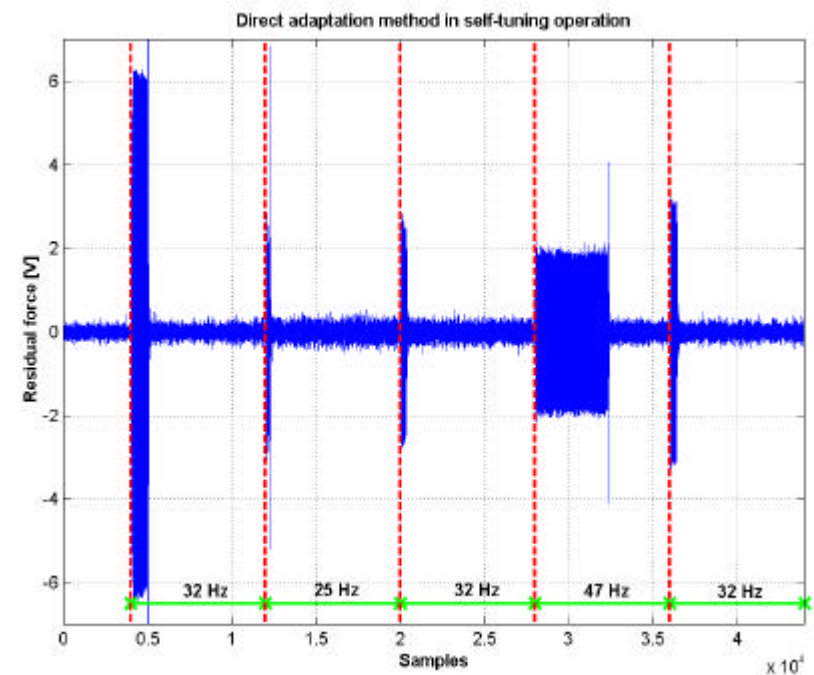
Time Domain Results

Self-tuning Operation

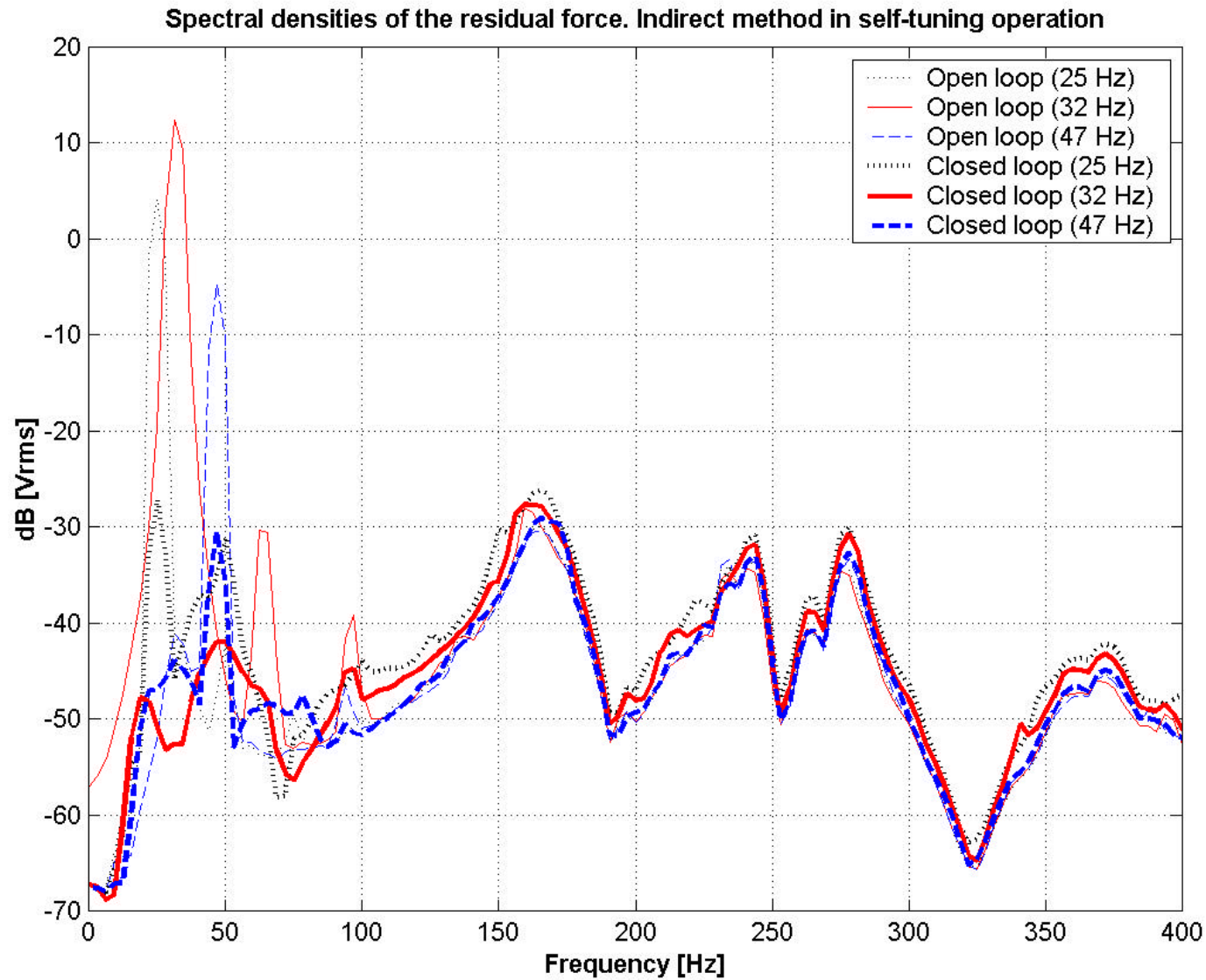
Indirect adaptive method



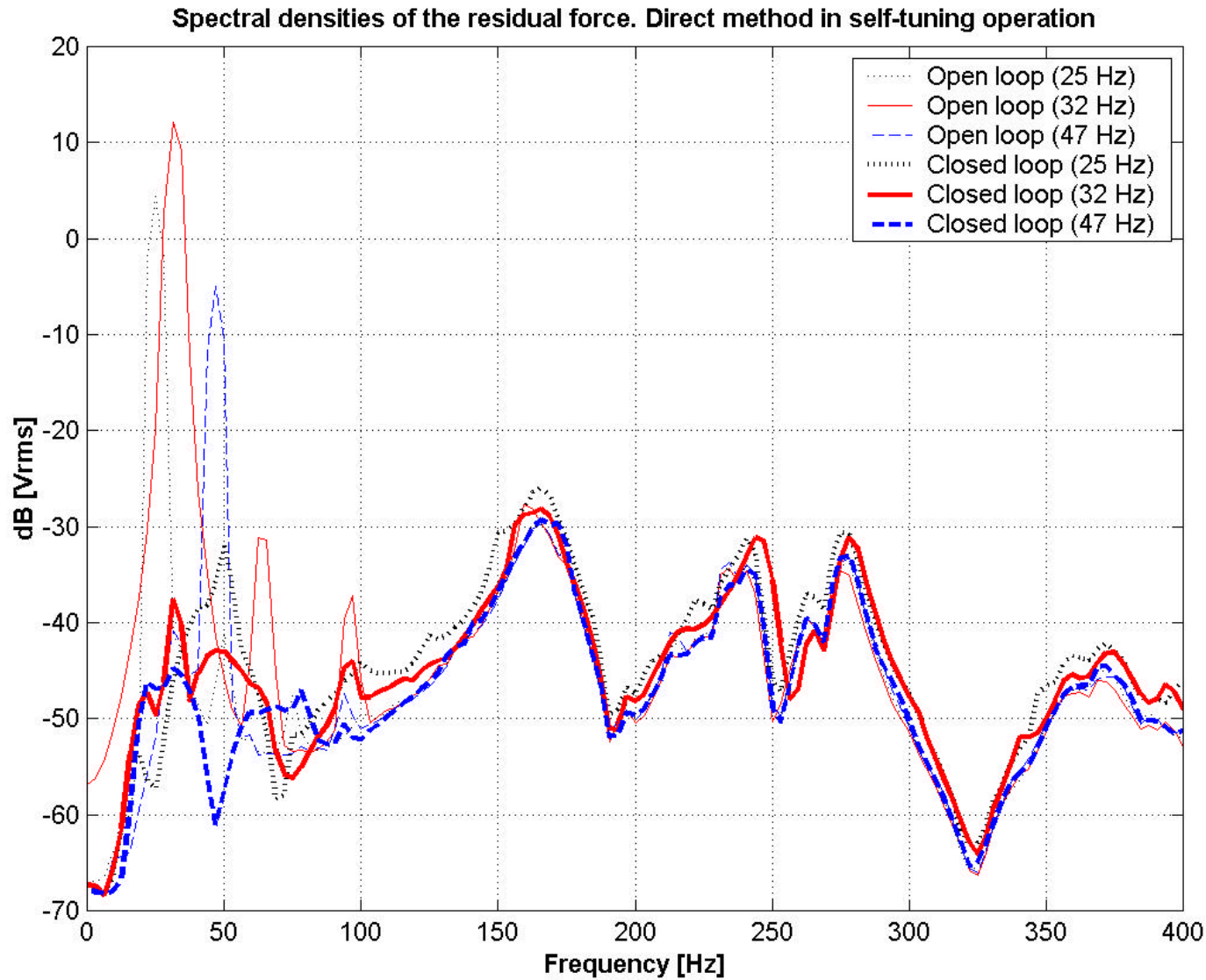
Direct adaptive method



Frequency domain results – indirect adaptive method



Frequency domain results – direct adaptive method

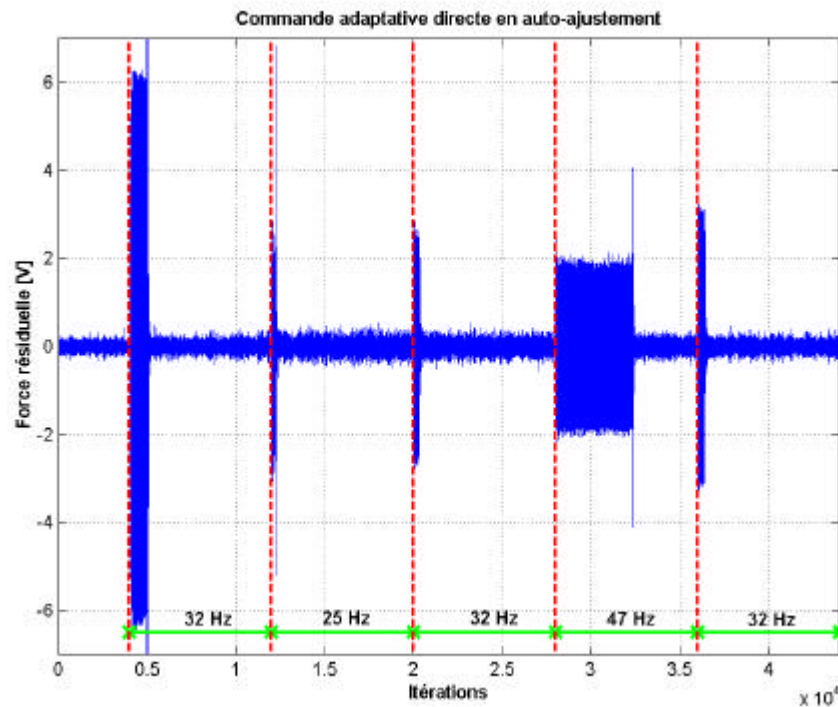


Real-time performances – indirect and direct methods

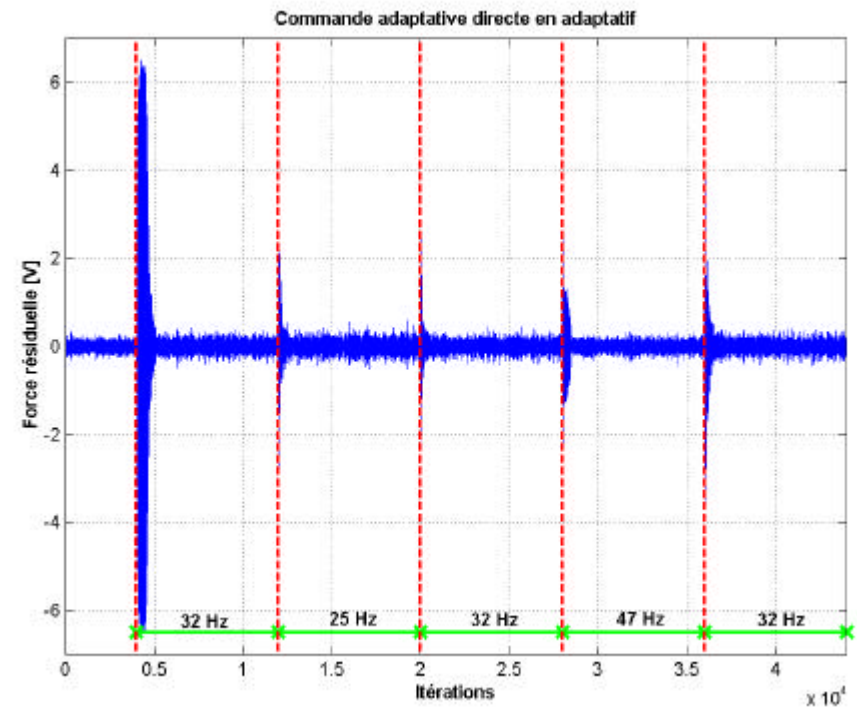
Method	Indirect			Direct		
Frequency [Hz]	25	32	47	25	32	47
Attenuation [dB]	31.32	64.23	25.72	61.65	49.64	55.79

Direct Adaptive Control

Self-tuning Mode



Adaptative Mode



- *Direct adaptive control in adaptive mode operation gives better results than direct adaptive control in self-tuning mode*
- *Direct adaptive control leads to a much simpler implementation than indirect adaptive control*

Conclusions

- Using internal model principle, adaptive control solutions can be provided for the rejection of unknown disturbances
- Both direct and indirect solutions can be provided
- Two modes of operation can be used : self-tuning and adaptive
- Direct adaptive control is the simplest to implement
- Direct adaptive control offers better performance
- The methodology has been extensively tested on an active suspension system