# Nano-Hexapod - Test Bench

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This document is dedicated to the experimental study of the nano-hexapod shown in Figure 0.1.

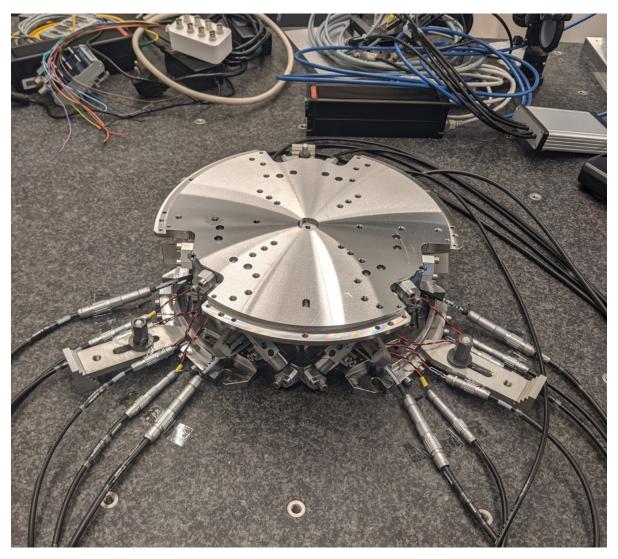


Figure 0.1: Nano-Hexapod

## Note

Here are the documentation of the equipment used for this test bench (lots of them are shwon in Figure 0.2):

- Voltage Amplifier: PiezoDrive PD200
- Amplified Piezoelectric Actuator: Cedrat APA300ML
- DAC/ADC: Speedgoat IO313
- Encoder: Renishaw Vionic and used Ruler
- Interferometers: Attocube

In Figure 0.3 is shown a block diagram of the experimental setup. When possible, the notations are consistent with this diagram and summarized in Table 0.1.

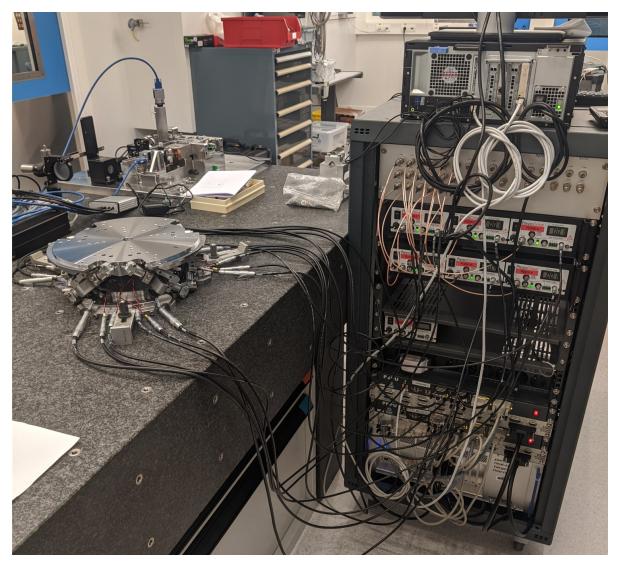


Figure 0.2: Nano-Hexapod and the control electronics

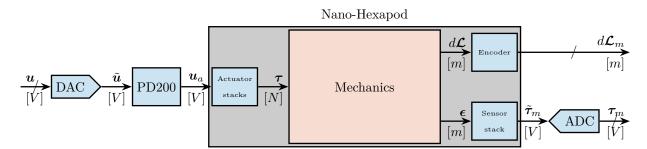


Figure 0.3: Block diagram of the system with named signals

Table 0.1: List of signals

	Unit	Matlab	Vector	Elements
Control Input (wanted DAC voltage)	[V]	u	u	$u_i$
DAC Output Voltage	[V]	u	$ ilde{m{u}}$	$ ilde{u}_i$
PD200 Output Voltage	[V]	ua	$oldsymbol{u}_a$	$u_{a,i}$
Actuator applied force	[N]	tau	au	$ au_i$
Strut motion	[m]	dL	$d\mathcal{L}$	$d\mathcal{L}_i$
Encoder measured displacement	[m]	dLm	$d\mathcal{L}_m$	$d\mathcal{L}_{m,i}$
Force Sensor strain	[m]	epsilon	$\epsilon$	$\epsilon_i$
Force Sensor Generated Voltage	[V]	taum	$ ilde{oldsymbol{ au}}_m$	$ ilde{ au}_{m,i}$
Measured Generated Voltage	[V]	taum	$oldsymbol{ au}_m$	$ au_{m,i}$
Motion of the top platform	[m,rad]	dX	$d\mathcal{X}$	$d\mathcal{X}_i$
Metrology measured displacement	[m,rad]	dXm	$d{m {\cal X}}_m$	$d\mathcal{X}_{m,i}$

This document is divided in the following sections:

- Section 1: the dynamics of the nano-hexapod when the encoders are fixed to the struts is studied.
- Section 2: the same is done when the encoders are fixed to the plates.
- Section 3: a decentralized HAC-LAC strategy is studied and implemented.

# 1 Encoders fixed to the Struts - Dynamics

In this section, the encoders are fixed to the struts.

It is divided in the following sections:

- Section 1.1: the transfer function matrix from the actuators to the force sensors and to the encoders is experimentally identified.
- Section 1.2: the obtained FRF matrix is compared with the dynamics of the simscape model
- Section 1.3: decentralized Integral Force Feedback (IFF) is applied and its performances are evaluated.
- Section 1.4: a modal analysis of the nano-hexapod is performed

# 1.1 Identification of the dynamics

## 1.1.1 Load Measurement Data

```
Matlab

**% Load Identification Data
meas_data_lf = {};

for i = 1:6
    meas_data_lf(i) = {load(sprintf('mat/frf_data_exc_strut_%i_noise_lf.mat', i), 't', 'Va', 'Vs', 'de')};
    meas_data_hf(i) = {load(sprintf('mat/frf_data_exc_strut_%i_noise_hf.mat', i), 't', 'Va', 'Vs', 'de')};
end
```

## 1.1.2 Spectral Analysis - Setup

```
% Setup useful variables
% Sampling Time [s]
Ts = (meas_data_lf{1}.t(end) - (meas_data_lf{1}.t(1)))/(length(meas_data_lf{1}.t)-1);
% Sampling Frequency [Hz]
Fs = 1/Ts;
% Hannning Windows
win = hanning(ceil(1*Fs));
% And we get the frequency vector
[~, f] = tfestimate(meas_data_lf{1}.Va, meas_data_lf{1}.de, win, [], [], 1/Ts);
```

```
i_lf = f < 250; % Points for low frequency excitation
i_hf = f > 250; % Points for high frequency excitation
```

## 1.1.3 Transfer function from Actuator to Encoder

First, let's compute the coherence from the excitation voltage and the displacement as measured by the encoders (Figure 1.1).

```
Matlab

%% Coherence
coh_dvf = zeros(length(f), 6, 6);

for i = 1:6
    coh_dvf_lf = mscohere(meas_data_lf{i}.Va, meas_data_lf{i}.de, win, [], [], 1/Ts);
    coh_dvf_hf = mscohere(meas_data_hf{i}.Va, meas_data_hf{i}.de, win, [], [], 1/Ts);
    coh_dvf(:,:,i) = [coh_dvf_lf(i_lf, :); coh_dvf_hf(i_hf, :)];
end
```

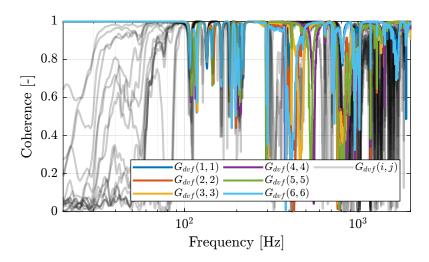


Figure 1.1: Obtained coherence for the DVF plant

Then the 6x6 transfer function matrix is estimated (Figure 1.2).

```
Matlab

%% DVF Plant (transfer function from u to dLm)

G_dvf = zeros(length(f), 6, 6);

for i = 1:6
    G_dvf_lf = tfestimate(meas_data_lf{i}.Va, meas_data_lf{i}.de, win, [], [], 1/Ts);
    G_dvf_hf = tfestimate(meas_data_hf{i}.Va, meas_data_hf{i}.de, win, [], [], 1/Ts);
    G_dvf(:,:,i) = [G_dvf_lf(i_lf, :); G_dvf_hf(i_hf, :)];
end
```

## 1.1.4 Transfer function from Actuator to Force Sensor

First, let's compute the coherence from the excitation voltage and the displacement as measured by the encoders (Figure 1.3).

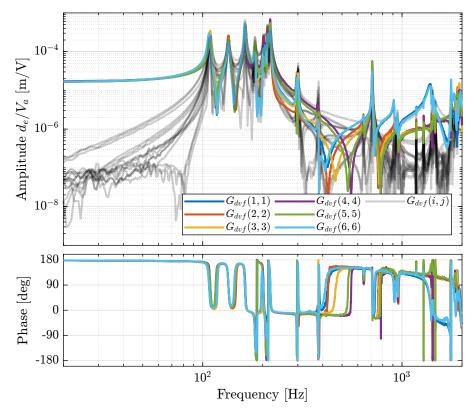


Figure 1.2: Measured FRF for the DVF plant

```
Matlab

%% Coherence for the IFF plant
coh_iff = zeros(length(f), 6, 6);

for i = 1:6
    coh_iff_lf = mscohere(meas_data_lf{i}.Va, meas_data_lf{i}.Vs, win, [], [], 1/Ts);
    coh_iff_hf = mscohere(meas_data_hf{i}.Va, meas_data_hf{i}.Vs, win, [], [], 1/Ts);
    coh_iff(:,:,i) = [coh_iff_lf(i_lf, :); coh_iff_hf(i_hf, :)];
end
```

Then the 6x6 transfer function matrix is estimated (Figure 1.4).

```
Matlab

G_iff = zeros(length(f), 6, 6);

for i = 1:6
    G_iff_lf = tfestimate(meas_data_lf{i}.Va, meas_data_lf{i}.Vs, win, [], [], 1/Ts);
    G_iff_hf = tfestimate(meas_data_hf{i}.Va, meas_data_hf{i}.Vs, win, [], [], 1/Ts);
    G_iff(:,:,i) = [G_iff_lf(i_lf, :); G_iff_hf(i_hf, :)];
end
```

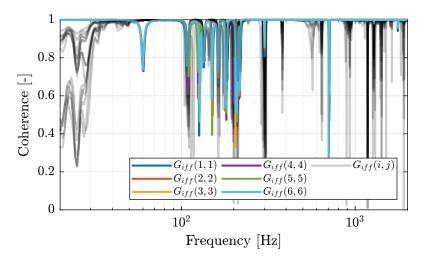


Figure 1.3: Obtained coherence for the IFF plant

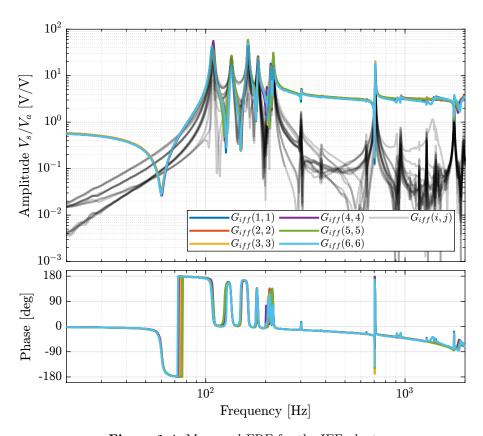


Figure 1.4: Measured FRF for the IFF plant

## 1.1.5 Save Identified Plants

```
save('matlab/mat/identified_plants_enc_struts.mat', 'f', 'Ts', 'G_iff', 'G_dvf')
```

# 1.2 Comparison with the Simscape Model

In this section, the measured dynamics is compared with the dynamics estimated from the Simscape model.

## 1.2.1 Load measured FRF

## 1.2.2 Dynamics from Actuator to Force Sensors

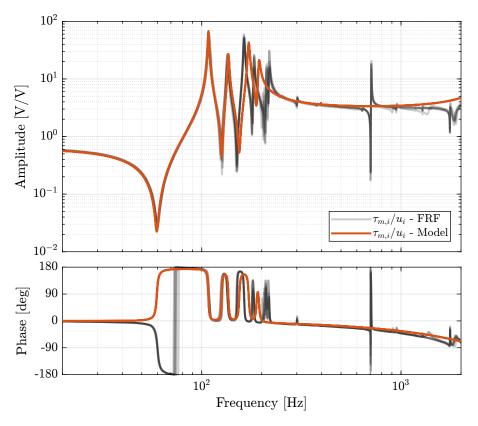
```
Matlab

%% Identify the IFF Plant (transfer function from u to taum)

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dum'], 1, 'openoutput'); io_i = io_i + 1; % Force Sensors

Giff = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

## 1.2.3 Dynamics from Actuator to Encoder



 $\textbf{Figure 1.5:} \ \ \textbf{Diagonal elements of the IFF Plant}$ 

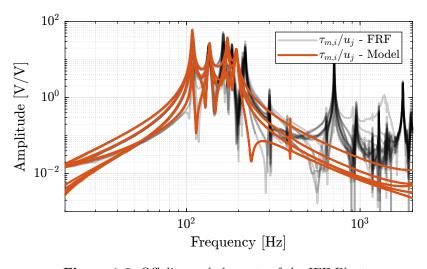


Figure 1.6: Off diagonal elements of the IFF Plant

```
Matlab

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/D'], 1, 'openoutput'); io_i = io_i + 1; % Encoders

Gdvf = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

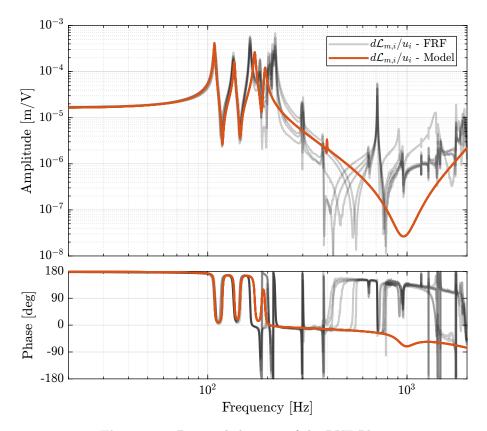


Figure 1.7: Diagonal elements of the DVF Plant

# 1.2.4 Effect of a change in bending damping of the joints

```
%% Tested bending dampings [Nm/(rad/s)]
cRs = [1e-3, 5e-3, 1e-2, 5e-2, 1e-1];

%% Identify the DVF Plant (transfer function from u to dLm)
clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/b'], 1, 'openoutput'); io_i = io_i + 1; % Encoders
```

Then the identification is performed for all the values of the bending damping.

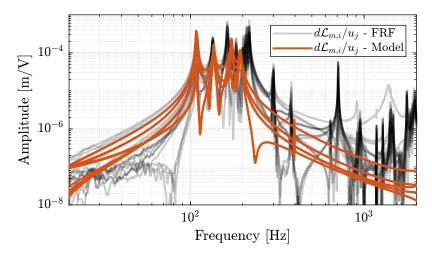


Figure 1.8: Off diagonal elements of the DVF Plant

- Could be nice
- Actual damping is very small

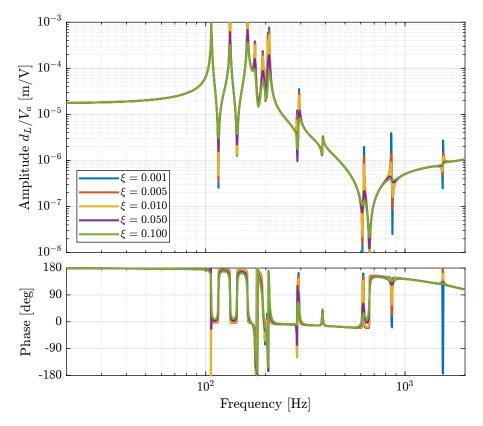
# 1.2.5 Effect of a change in damping factor of the APA

```
Matlab

%% Tested bending dampings [Nm/(rad/s)]
xis = [1e-3, 5e-3, 1e-2, 5e-2, 1e-1];

Matlab

%% Identify the DVF Plant (transfer function from u to dLm)
clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/D'], 1, 'openoutput'); io_i = io_i + 1; % Encoders
```



**Figure 1.9:** Effect of the APA damping factor  $\xi$  on the dynamics from u to  $d\mathcal{L}$ 

## Important

Damping factor  $\xi$  has a large impact on the damping of the "spurious resonances" at 200Hz and 300Hz.

#### Question

Why is the damping factor does not change the damping of the first peak?

## 1.2.6 Effect of a change in stiffness damping coef of the APA

## 1.2.7 Effect of a change in mass damping coef of the APA

```
Matlab

### Matlab

### Matlab

### Matlab

### Matlab

### Identify the DVF Plant (transfer function from u to dLm)

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1;  ### Actuator Inputs
io(io_i) = linio([mdl, '/D'], 1, 'openoutput'); io_i = io_i + 1;  ### Encoders
```

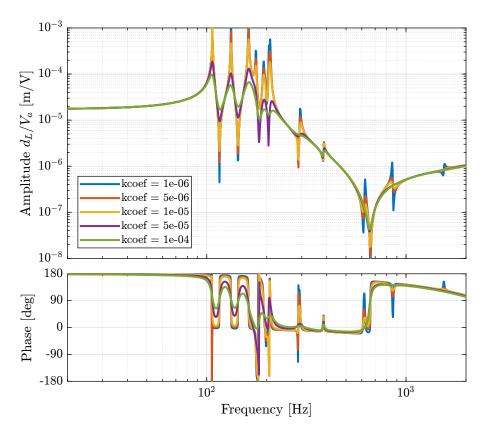
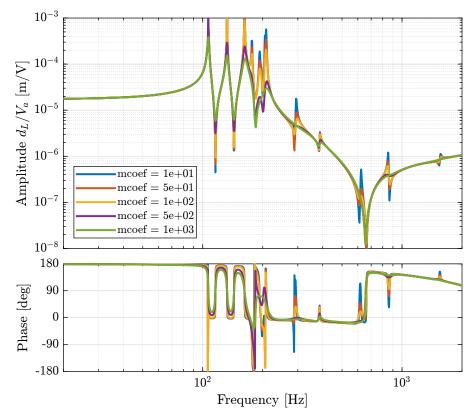


Figure 1.10: Effect of a change of the damping "stiffness coeficient" on the transfer function from u to  $d\mathcal{L}$ 



**Figure 1.11:** Effect of a change of the damping "mass coeficient" on the transfer function from u to  $d\mathcal{L}$ 

# 1.2.8 Using Flexible model

#### ${f Question}$

Why do we have smaller resonances when using flexible APA? On the test bench we have the same resonance as the 2DoF model. Could it be due to the compliance in other dof of the flexible model?

```
Matlab

Clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/D'], 1, 'openoutput'); io_i = io_i + 1; % Encoders

Gdvf = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

```
Matlab

Clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dum'], 1, 'openoutput'); io_i = io_i + 1; % Force Sensors

Giff = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

## 1.2.9 Flexible model + encoders fixed to the plates

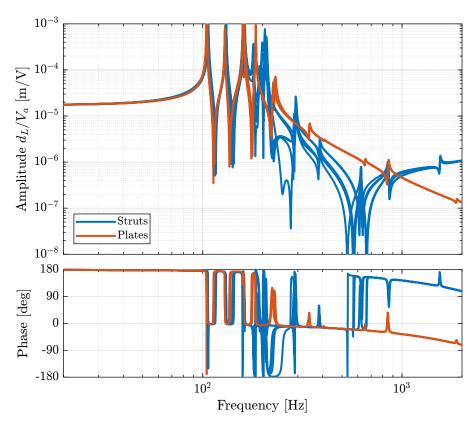


Figure 1.12: Comparison of the dynamics from  $V_a$  to  $d_L$  when the encoders are fixed to the struts (blue) and to the plates (red). APA are modeled as a flexible element.

# 1.3 Integral Force Feedback

In this section, the Integral Force Feedback (IFF) control strategy is applied to the nano-hexapod. The main goal of this to add damping to the nano-hexapod's modes.

The control architecture is shown in Figure 1.13 where  $K_{\text{IFF}}$  is a diagonal  $6 \times 6$  controller.

The system as then a new input u', and the transfer function from u' to  $d\mathcal{L}_m$  should be easier to control than the initial transfer function from u to  $d\mathcal{L}_m$ .

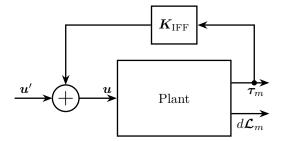


Figure 1.13: Integral Force Feedback Strategy

This section is structured as follow:

- Section 1.3.1: Using the Simscape model (APA taken as 2DoF model), the transfer function from  $\boldsymbol{u}$  to  $\boldsymbol{\tau}_m$  is identified. Based on the obtained dynamics, the control law is developed and the optimal gain is estimated using the Root Locus.
- Section 1.3.2: Still using the Simscape model, the effect of the IFF gain on the transfer function from u' to  $d\mathcal{L}_m$  is studied.
- Section 1.3.3: The same is performed experimentally: several IFF gains are used and the damped plant is identified each time.
- Section 1.3.4: The damped model and the identified damped system are compared for the optimal IFF gain. It is found that IFF indeed adds a lot of damping into the system. However it is not efficient in damping the spurious struts modes.
- Section 1.3.5: Finally, a "flexible" model of the APA is used in the Simscape model and the optimally damped model is compared with the measurements.

## 1.3.1 IFF Control Law and Optimal Gain

Let's use a model of the Nano-Hexapod with the encoders fixed to the struts and the APA taken as 2DoF model.

The transfer function from  $\boldsymbol{u}$  to  $\boldsymbol{\tau}_m$  is identified.

```
Matlab

Clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dum'], 1, 'openoutput'); io_i = io_i + 1; % Force Sensors

Giff = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

The IFF controller is defined as shown below:

Then, the poles of the system are shown in the complex plane as a function of the controller gain (i.e. Root Locus plot) in Figure 1.14. A gain of 400 is chosen as the "optimal" gain as it visually seems to be the gain that adds the maximum damping to all the suspension modes simultaneously.

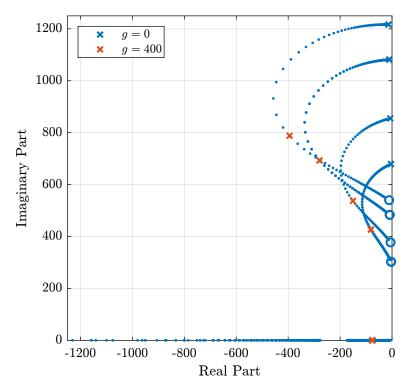


Figure 1.14: Root Locus for the IFF control strategy

Then the "optimal" IFF controller is:

And it is saved for further use.

```
______Matlab ______
```

The bode plots of the "diagonal" elements of the loop gain are shown in Figure 1.15. It is shown that the phase and gain margins are quite high and the loop gain is large arround the resonances.

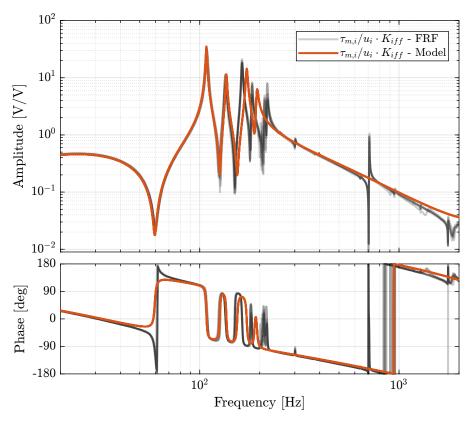


Figure 1.15: Bode plot of the "decentralized loop gain"  $G_{\text{iff}}(i,i) \times K_{\text{iff}}(i,i)$ 

## 1.3.2 Effect of IFF on the plant - Simulations

Still using the Simscape model with encoders fixed to the struts and 2DoF APA, the IFF strategy is tested.

The following IFF gains are tried:

And the transfer functions from u' to  $d\mathcal{L}_m$  are identified for all the IFF gains.

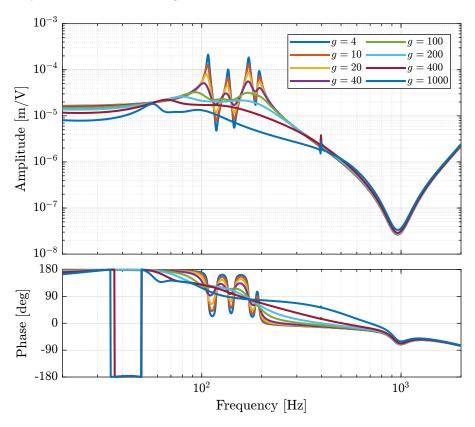
```
Matlab

%% Identify the (damped) transfer function from u to dLm for different values of the IFF gain
Gd_iff = {zeros(1, length(iff_gains))};

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Strut Displacement (encoder)

for i = 1:length(iff_gains)
    Kiff = iff_gains(i)*Kiff_g1*eye(6); % IFF Controller
    Gd_iff(i) = {exp(-s*Ts)*linearize(mdl, io, 0.0, options)};
    isstable(Gd_iff{i})
end
```

The obtained dynamics are shown in Figure 1.16.



**Figure 1.16:** Effect of the IFF gain g on the transfer function from  $\tau$  to  $d\mathcal{L}_m$ 

## 1.3.3 Effect of IFF on the plant - Experimental Results

The IFF strategy is applied experimentally and the transfer function from u' to  $d\mathcal{L}_m$  is identified for all the defined values of the gain.

#### **Load Data**

First load the identification data.

```
Matlab

*** Load Identification Data
meas_iff_gains = {};

for i = 1:length(iff_gains)
    meas_iff_gains(i) = {load(sprintf('mat/iff_strut_1_noise_g_%i.mat', iff_gains(i)), 't', 'Vexc', 'Vs', 'de', 'u')};
end
```

#### Spectral Analysis - Setup

And define the useful variables that will be used for the identification using the tfestimate function.

```
Matlab

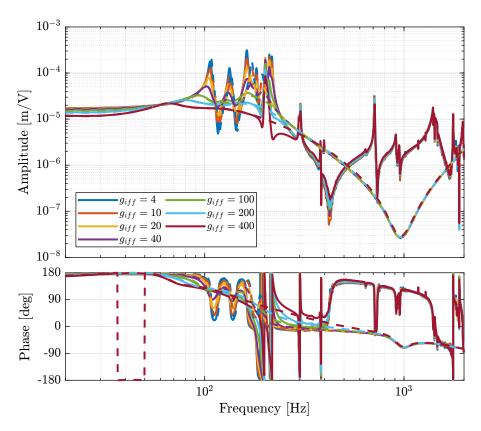
%% Setup useful variables
% Sampling Time [s]
Ts = (meas_iff_gains{1}.t(end) - (meas_iff_gains{1}.t(1)))/(length(meas_iff_gains{1}.t)-1);
% Sampling Frequency [Hz]
Fs = 1/Ts;
% Hannning Windows
win = hanning(ceil(1*Fs));
% And we get the frequency vector
[~, f] = tfestimate(meas_iff_gains{1}.Vexc, meas_iff_gains{1}.de, win, [], [], 1/Ts);
```

#### **DVF Plant**

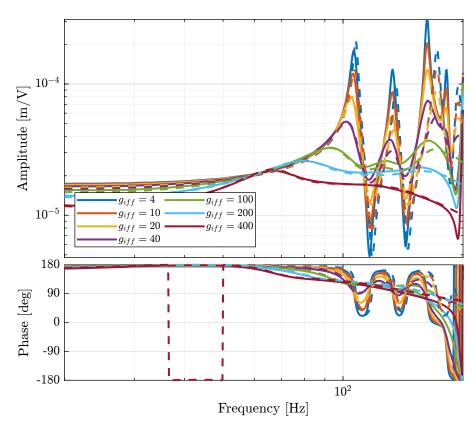
The transfer functions are estimated for all the values of the gain.

The obtained dynamics as shown in the bode plot in Figure 1.17. The dashed curves are the results obtained using the model, and the solid curves the results from the experimental identification.

The bode plot is then zoomed on the suspension modes of the nano-hexapod in Figure 1.18.



**Figure 1.17:** Transfer function from u to  $d\mathcal{L}_m$  for multiple values of the IFF gain



**Figure 1.18:** Transfer function from u to  $d\mathcal{L}_m$  for multiple values of the IFF gain (Zoom)

## Important

The IFF control strategy is very effective for the damping of the suspension modes. It however does not damp the modes at 200Hz, 300Hz and 400Hz (flexible modes of the APA). Also, the experimental results and the models obtained from the Simscape model are in agreement concerning the damped system (up to the flexible modes).

#### Experimental Results - Comparison of the un-damped and fully damped system

The un-damped and damped experimental plants are compared in Figure 1.19 (diagonal terms).

It is very clear that all the suspension modes are very well damped thanks to IFF. However, there is little to no effect on the flexible modes of the struts and of the plate.

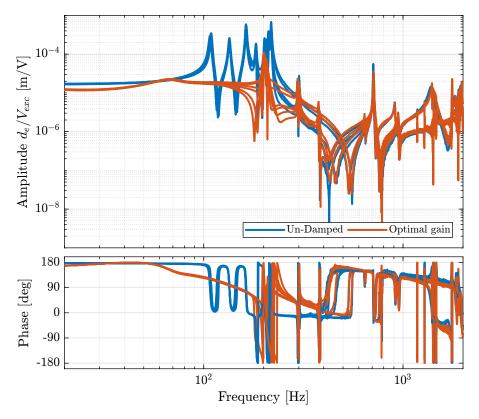


Figure 1.19: Comparison of the diagonal elements of the transfer function from u to  $d\mathcal{L}_m$  without active damping and with optimal IFF gain

## 1.3.4 Experimental Results - Damped Plant with Optimal gain

Let's now look at the  $6 \times 6$  damped plant with the optimal gain g = 400.

#### Load Data

The experimental data are loaded.

```
Matlab

%% Load Identification Data
meas_iff_struts = {};

for i = 1:6
    meas_iff_struts(i) = {load(sprintf('mat/iff_strut_%i_noise_g_400.mat', i), 't', 'Vexc', 'Vs', 'de', 'u')};
end
```

## Spectral Analysis - Setup

And the parameters useful for the spectral analysis are defined.

```
Matlab

%% Setup useful variables
% Sampling Time [s]
Ts = (meas_iff_struts{1}.t(end) - (meas_iff_struts{1}.t(1)))/(length(meas_iff_struts{1}.t)-1);
% Sampling Frequency [Hz]
Fs = 1/Ts;
% Hannning Windows
win = hanning(ceil(1*Fs));
% And we get the frequency vector
[~, f] = tfestimate(meas_iff_struts{1}.Vexc, meas_iff_struts{1}.de, win, [], [], 1/Ts);
```

### **DVF Plant**

Finally, the  $6 \times 6$  plant is identified using the tfestimate function.

```
Matlab

%% DVF Plant (transfer function from u to dLm)
G_iff_opt = {};

for i = 1:6
    G_iff_opt{i} = tfestimate(meas_iff_struts{i}.Vexc, meas_iff_struts{i}.de, win, [], [], 1/Ts);
end
```

The obtained diagonal elements are compared with the model in Figure 1.20.

And all the off-diagonal elements are compared with the model in Figure 1.21.

#### Important

With the IFF control strategy applied and the optimal gain used, the suspension modes are very well damped. Remains the un-damped flexible modes of the APA (200Hz, 300Hz, 400Hz), and the modes of the plates (700Hz).

The Simscape model and the experimental results are in very good agreement.

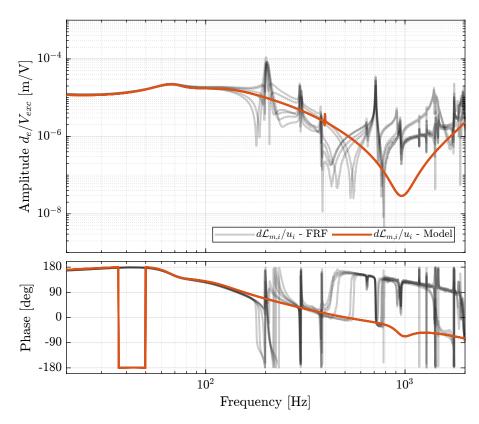


Figure 1.20: Comparison of the diagonal elements of the transfer functions from  $\boldsymbol{u}$  to  $d\boldsymbol{\mathcal{L}}_m$  with active damping (IFF) applied with an optimal gain g=400

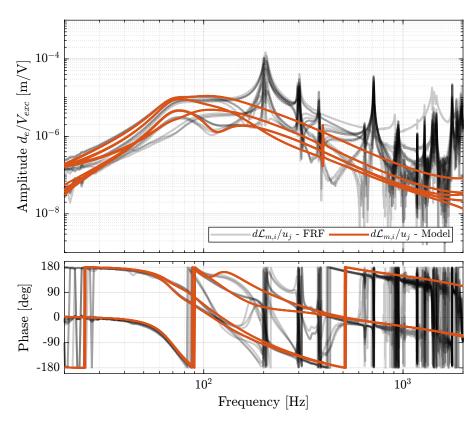


Figure 1.21: Comparison of the off-diagonal elements of the transfer functions from  $\boldsymbol{u}$  to  $d\boldsymbol{\mathcal{L}}_m$  with active damping (IFF) applied with an optimal gain g=400

## 1.3.5 Comparison with the Flexible model

When using the 2-DoF model for the APA, the flexible modes of the struts were not modelled, and it was the main limitation of the model. Now, let's use a flexible model for the APA, and see if the obtained damped plant using the model is similar to the measured dynamics.

First, the nano-hexapod is initialized.

And the "optimal" controller is loaded.

```
%% Optimal IFF controller load('Kiff.mat', 'Kiff');
```

The transfer function from u' to  $d\mathcal{L}_m$  is identified using the Simscape model.

```
Matlab

%% Linearized inputs/outputs
clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Strut Displacement (encoder)

%% Identification of the plant
Gd_iff = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

The obtained diagonal elements are shown in Figure 1.22 while the off-diagonal elements are shown in Figure 1.23.

## Important

Using flexible models for the APA, the agreement between the Simscape model of the nanohexapod and the measured FRF is very good.

Only the flexible mode of the top-plate is not appearing in the model which is very logical as the top plate is taken as a solid body.

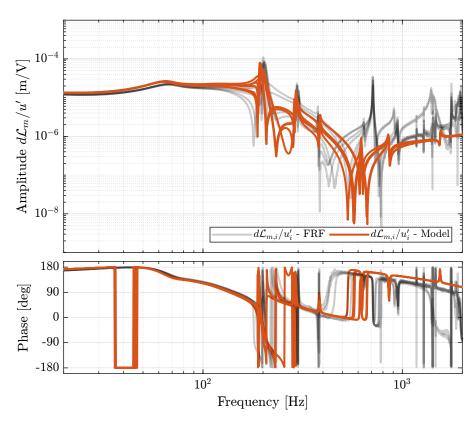


Figure 1.22: Diagonal elements of the transfer function from u' to  $d\mathcal{L}_m$  - comparison of the measured FRF and the identified dynamics using the flexible model

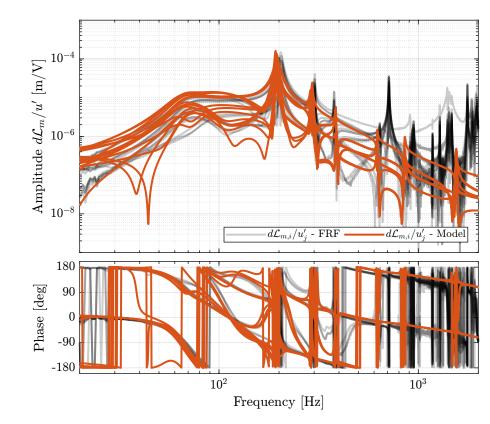


Figure 1.23: Off-diagonal elements of the transfer function from u' to  $d\mathcal{L}_m$  - comparison of the measured FRF and the identified dynamics using the flexible model

## 1.3.6 Conclusion

## Important

The decentralized Integral Force Feedback strategy applied on the nano-hexapod is very effective in damping all the suspension modes.

The Simscape model (especially when using a flexible model for the APA) is shown to be very accurate, even when IFF is applied.

# 1.4 Modal Analysis

Several 3-axis accelerometers are fixed on the top platform of the nano-hexapod as shown in Figure 1.28.

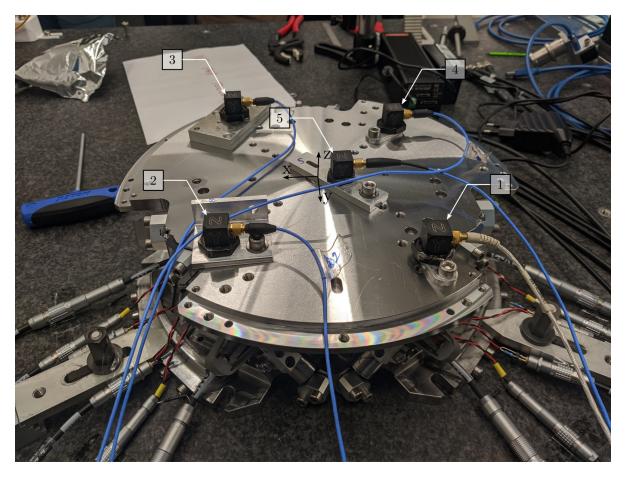


Figure 1.24: Location of the accelerometers on top of the nano-hexapod

The top platform is then excited using an instrumented hammer as shown in Figure 1.25.

From this experiment, the resonance frequencies and the associated mode shapes can be computed (Section 1.4.1). Then, in Section 1.4.2, the vertical compliance of the nano-hexapod is experimentally



 ${\bf Figure~1.25:~Example~of~an~excitation~using~an~instrumented~hammer}$ 

estimated. Finally, in Section 1.4.3, the measured compliance is compare with the estimated one from the Simscape model.

#### 1.4.1 Obtained Mode Shapes

We can observe the mode shapes of the first 6 modes that are the suspension modes (the plate is behaving as a solid body) in Figure 1.26.

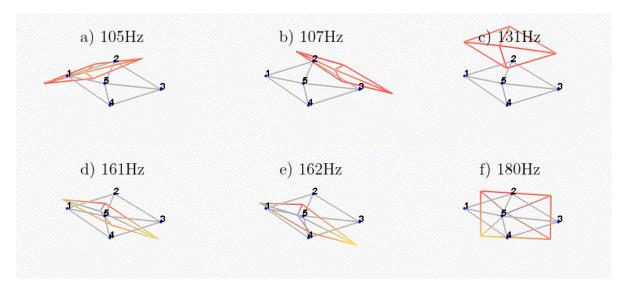


Figure 1.26: Measured mode shapes for the first six modes

Then, there is a mode at 692Hz which corresponds to a flexible mode of the top plate (Figure 1.27).



Figure 1.27: First flexible mode at 692Hz

The obtained modes are summarized in Table 1.1.

#### 1.4.2 Nano-Hexapod Compliance - Effect of IFF

In this section, we wish to estimated the effectiveness of the IFF strategy concerning the compliance.

The top plate is excited vertically using the instrumented hammer two times:

1. no control loop is used

Table 1.1: Description of the identified modes

Mode	Freq. [Hz]	Description
1	105	Suspension Mode: Y-translation
2	107	Suspension Mode: X-translation
3	131	Suspension Mode: Z-translation
4	161	Suspension Mode: Y-tilt
5	162	Suspension Mode: X-tilt
6	180	Suspension Mode: Z-rotation
7	692	(flexible) Membrane mode of the top platform

#### 2. decentralized IFF is used

The data is loaded.

```
frf_ol = load('Measurement_Z_axis.mat'); % Open-Loop
frf_iff = load('Measurement_Z_axis_damped.mat'); % IFF
```

The mean vertical motion of the top platform is computed by averaging all 5 accelerometers.

```
Matlab

%% Multiply by 10 (gain in m/s^2/V) and divide by 5 (number of accelerometers)

d_frf_ol = 10/5*(frf_ol.FFT1_H1_4_1_RMS_Y_Mod + frf_ol.FFT1_H1_7_1_RMS_Y_Mod + frf_ol.FFT1_H1_13_1_RMS_Y_Mod + frf_ol.FFT1_H1_16_1_RMS_Y_Mod)./(2*pi*frf_ol.FFT1_H1_16_1_RMS_X_Val).^2;

d_frf_iff = 10/5*(frf_iff.FFT1_H1_4_1_RMS_Y_Mod + frf_iff.FFT1_H1_7_1_RMS_Y_Mod + frf_iff.FFT1_H1_10_1_RMS_Y_Mod + frf_iff.FFT1_H1_13_1_RMS_Y_Mod + frf_iff.FFT1_H1_16_1_RMS_Y_Mod)./(2*pi*frf_iff.FFT1_H1_16_1_RMS_X_Val).^2;
```

The vertical compliance (magnitude of the transfer function from a vertical force applied on the top plate to the vertical motion of the top plate) is shown in Figure 1.28.

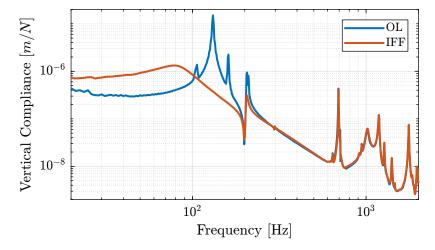


Figure 1.28: Measured vertical compliance with and without IFF

#### Important

From Figure 1.28, it is clear that the IFF control strategy is very effective in damping the suspensions modes of the nano-hexapod. It also has the effect of (slightly) degrading the vertical compliance at low frequency.

It also seems some damping can be added to the modes at around 205Hz which are flexible modes of the struts.

#### 1.4.3 Comparison with the Simscape Model

Let's now compare the measured vertical compliance with the vertical compliance as estimated from the Simscape model.

The transfer function from a vertical external force to the absolute motion of the top platform is identified (with and without IFF) using the Simscape model. The comparison is done in Figure 1.29. Again, the model is quite accurate!

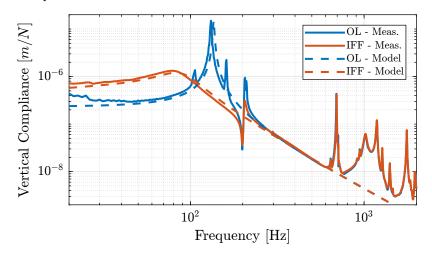


Figure 1.29: Measured vertical compliance with and without IFF

#### 1.5 Conclusion

#### Important

From the previous analysis, several conclusions can be drawn:

- Decentralized IFF is very effective in damping the "suspension" modes of the nano-hexapod (Figure 1.19)
- Decentralized IFF does not damp the "spurious" modes of the struts nor the flexible modes of the top plate (Figure 1.19)
- Even though the Simscape model and the experimentally measured FRF are in good agreement (Figures 1.22 and 1.23), the obtain dynamics from the control inputs u and the

encoders  $d\mathcal{L}_m$  is very difficult to control

Therefore, in the following sections, the encoders will be fixed to the plates. The goal is to be less sensitive to the flexible modes of the struts.

## 2 Encoders fixed to the plates - Dynamics

In this section, the encoders are fixed to the plates rather than to the struts as shown in Figure 2.1.

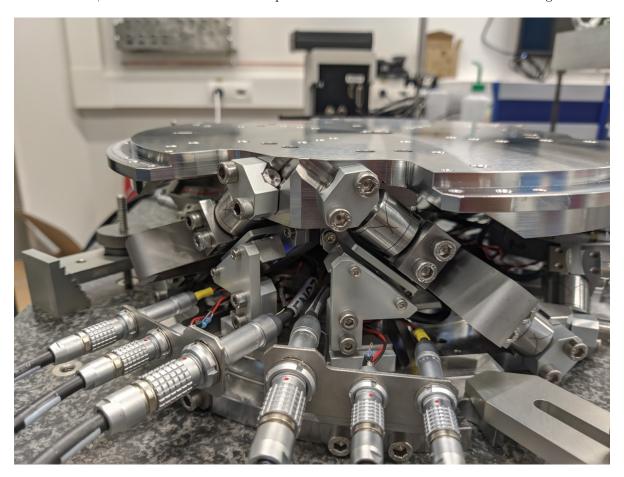


Figure 2.1: Nano-Hexapod with encoders fixed to the struts

It is structured as follow:

- Section 2.1: The dynamics of the nano-hexapod is identified.
- $\bullet$  Section 2.2: The identified dynamics is compared with the Simscape model.
- Section 2.3: The Integral Force Feedback (IFF) control strategy is applied and the dynamics of the damped nano-hexapod is identified and compare with the Simscape model.

## 2.1 Identification of the dynamics

In this section, the dynamics of the nano-hexapod with the encoders fixed to the plates is identified.

First, the measurement data are loaded in Section 2.1.1, then the transfer function matrix from the actuators to the encoders are estimated in Section 2.1.2. Finally, the transfer function matrix from the actuators to the force sensors is estimated in Section 2.1.3.

#### 2.1.1 Data Loading and Spectral Analysis Setup

The actuators are excited one by one using a low pass filtered white noise. For each excitation, the 6 force sensors and 6 encoders are measured and saved.

#### 2.1.2 Transfer function from Actuator to Encoder

Let's compute the coherence from the excitation voltage u and the displacement  $d\mathcal{L}_m$  as measured by the encoders.

```
Matlab

%% Coherence
coh_dvf = zeros(length(f), 6, 6);

for i = 1:6
    coh_dvf(:, :, i) = mscohere(meas_data_lf{i}.Va, meas_data_lf{i}.de, win, [], [], 1/Ts);
end
```

The obtained coherence shown in Figure 2.2 is quite good up to 400Hz.

Then the 6x6 transfer function matrix is estimated.

```
Matlab

%% DVF Plant (transfer function from u to dLm)
G_dvf = zeros(length(f), 6, 6);

for i = 1:6
    G_dvf(:,:,i) = tfestimate(meas_data_lf{i}.Va, meas_data_lf{i}.de, win, [], [], 1/Ts);
end
```

The diagonal and off-diagonal terms of this transfer function matrix are shown in Figure 2.3.

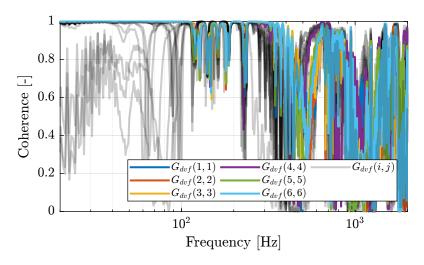


Figure 2.2: Obtained coherence for the DVF plant

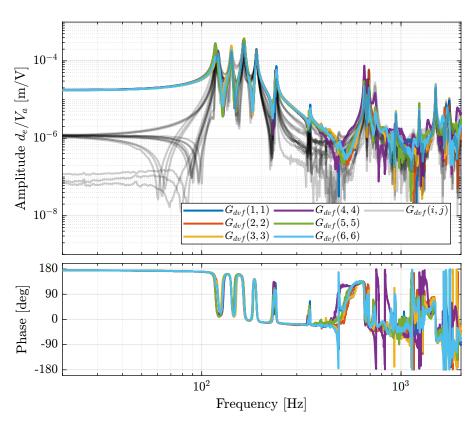


Figure 2.3: Measured FRF for the DVF plant

#### Important

From Figure 2.3, we can draw few conclusions on the transfer functions from u to  $d\mathcal{L}_m$  when the encoders are fixed to the plates:

- the decoupling is rather good at low frequency (below the first suspension mode). The low frequency gain is constant for the off diagonal terms, whereas when the encoders where fixed to the struts, the low frequency gain of the off-diagonal terms were going to zero (Figure 1.2).
- the flexible modes of the struts at 226Hz and 337Hz are indeed shown in the transfer functions, but their amplitudes are rather low.
- the diagonal terms have alternating poles and zeros up to at least 600Hz: the flexible modes of the struts are not affecting the alternating pole/zero pattern. This what not the case when the encoders were fixed to the struts (Figure 1.2).

#### 2.1.3 Transfer function from Actuator to Force Sensor

Let's now compute the coherence from the excitation voltage u and the voltage  $\tau_m$  generated by the Force senors.

The coherence is shown in Figure 2.4, and is very good for from 10Hz up to 2kHz.

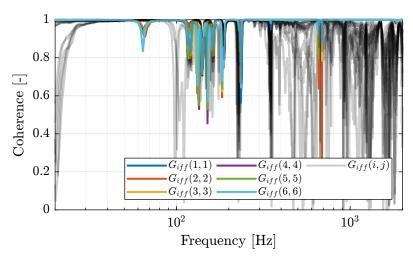


Figure 2.4: Obtained coherence for the IFF plant

Then the 6x6 transfer function matrix is estimated.

```
%% IFF Plant
G_iff = zeros(length(f), 6, 6);
for i = 1:6
    G_iff(:,:,i) = tfestimate(meas_data_lf{i}.Va, meas_data_lf{i}.Vs, win, [], [], 1/Ts);
end
```

The bode plot of the diagonal and off-diagonal terms are shown in Figure 2.5.

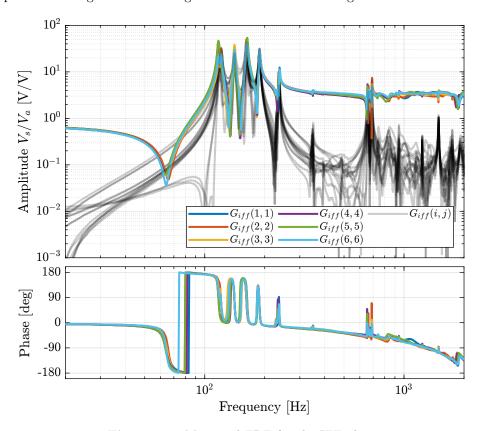


Figure 2.5: Measured FRF for the IFF plant

#### Important

It is shown in Figure 2.6 that:

- The IFF plant has alternating poles and zeros
- $\bullet$  The first flexible mode of the struts as 235Hz is appearing, and therefore is should be possible to add some damping to this mode using IFF
- The decoupling is quite good at low frequency (below the first model) as well as high frequency (above the last suspension mode, except near the flexible modes of the top plate)

#### 2.1.4 Save Identified Plants

The identified dynamics is saved for further use.

```
save('mat/identified_plants_enc_plates.mat', 'f', 'Ts', 'G_iff', 'G_dvf')
```

### 2.2 Comparison with the Simscape Model

In this section, the measured dynamics done in Section 2.1 is compared with the dynamics estimated from the Simscape model.

#### 2.2.1 Identification Setup

The nano-hexapod is initialized with the APA taken as flexible models.

#### 2.2.2 Dynamics from Actuator to Force Sensors

Then the transfer function from u to  $\tau_m$  is identified using the Simscape model.

```
Matlab

%% Identify the IFF Plant (transfer function from u to taum)
clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/Fm'], 1, 'openoutput'); io_i = io_i + 1; % Force Sensors

Giff = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

The identified dynamics is compared with the measured FRF:

- Figure 2.6: the individual transfer function from  $u_1$  (the DAC voltage for the first actuator) to the force sensors of all 6 struts are compared
- Figure 2.7: all the diagonal elements are compared
- Figure 2.8: all the off-diagonal elements are compared

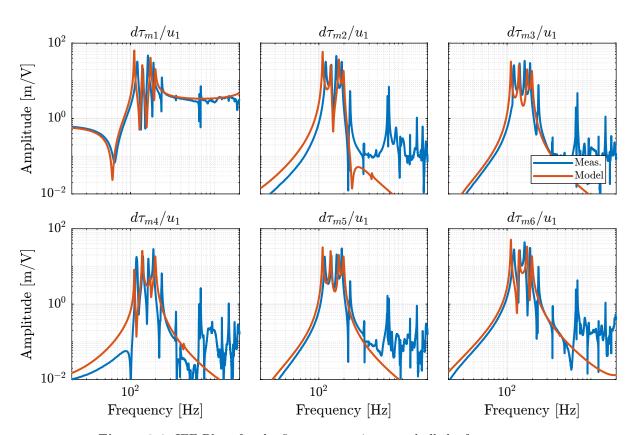
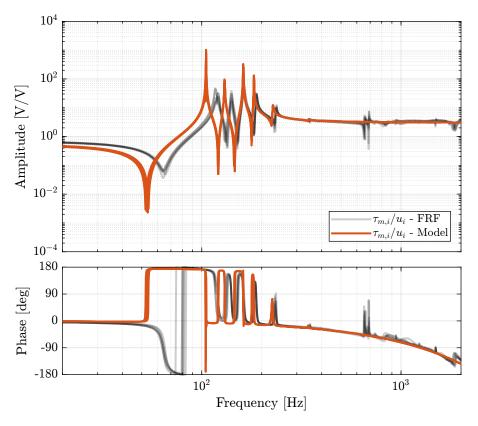


Figure 2.6: IFF Plant for the first actuator input and all the force senosrs



 ${\bf Figure~2.7:~Diagonal~elements~of~the~IFF~Plant}$ 

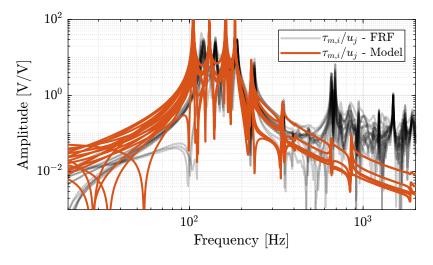


Figure 2.8: Off diagonal elements of the IFF Plant

#### 2.2.3 Dynamics from Actuator to Encoder

Now, the dynamics from the DAC voltage u to the encoders  $d\mathcal{L}_m$  is estimated using the Simscape model.

```
Matlab

Clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Encoders

Gdvf = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

The identified dynamics is compared with the measured FRF:

- Figure 2.9: the individual transfer function from  $u_3$  (the DAC voltage for the actuator number 3) to the six encoders
- Figure 2.10: all the diagonal elements are compared
- Figure 2.11: all the off-diagonal elements are compared

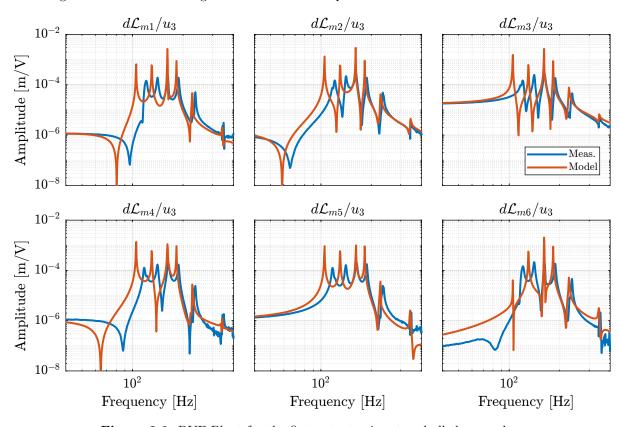


Figure 2.9: DVF Plant for the first actuator input and all the encoders

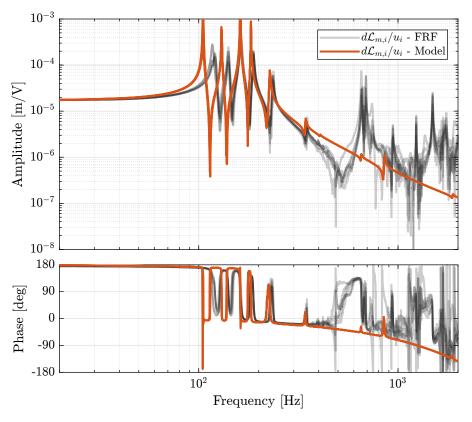


Figure 2.10: Diagonal elements of the DVF Plant

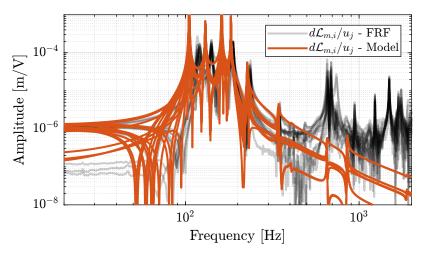


Figure 2.11: Off diagonal elements of the DVF Plant

#### 2.2.4 Flexible Top Plate

```
Matlab

Clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Encoders

Gdvf = linearize(mdl, io, 0.0, options);
```

```
size(Gdvf)
isstable(Gdvf)
```

```
_____Matlab

[sys,g] = balreal(Gdvf); % Compute balanced realization

elim = (g<1e-4); % Small entries of g are negligible states

rsys = modred(sys,elim); % Remove negligible states

size(rsys)
```

```
Matlab

%% Identify the IFF Plant (transfer function from u to taum)

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/Fm'], 1, 'openoutput'); io_i = io_i + 1; % Force Sensors

Giff = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

#### 2.2.5 Conclusion

#### Important

The Simscape model is quite accurate for the transfer function matrices from u to  $\tau_m$  and from u to  $d\mathcal{L}_m$  except at frequencies of the flexible modes of the top-plate. The Simscape model can therefore be used to develop the control strategies.

## 2.3 Integral Force Feedback

In this section, the Integral Force Feedback (IFF) control strategy is applied to the nano-hexapod in order to add damping to the suspension modes.

The control architecture is shown in Figure 2.12:

- $\tau_m$  is the measured voltage of the 6 force sensors
- $K_{\text{IFF}}$  is the  $6 \times 6$  diagonal controller
- *u* is the plant input (voltage generated by the 6 DACs)
- u' is the new plant inputs with added damping

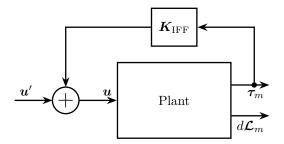


Figure 2.12: Integral Force Feedback Strategy

• Section 2.3.1

#### 2.3.1 Effect of IFF on the plant - Simscape Model

The nano-hexapod is initialized with flexible APA and the encoders fixed to the struts.

The same controller as the one developed when the encoder were fixed to the struts is used.

The transfer function from u' to  $d\mathcal{L}_m$  is identified.

```
Matlab
%% Identify the (damped) transfer function from u to dLm for different values of the IFF gain
clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Plate Displacement (encoder)
```

First in Open-Loop:

```
% Transfer function from u to dL (open-loop)
Gd_ol = exp(-s*Ts)*linearize(mdl, io, 0.0, options);
```

And then with the IFF controller:

It is first verified that the system is stable:

```
isstable(Gd_iff)

Results
```

The diagonal and off-diagonal terms of the  $6 \times 6$  transfer function matrices identified are compared in Figure 2.13. It is shown, as was the case when the encoders were fixed to the struts, that the IFF control strategy is very effective in damping the suspension modes of the nano-hexapod.

#### 2.3.2 Effect of IFF on the plant - FRF

The IFF control strategy is experimentally implemented. The (damped) transfer function from u' to  $d\mathcal{L}_m$  is experimentally identified.

The identification data are loaded:

```
Matlab

%% Load Identification Data
meas_iff_plates = {};

for i = 1:6
    meas_iff_plates(i) = {load(sprintf('mat/frf_exc_iff_strut_%i_enc_plates_noise.mat', i), 't', 'Va', 'Vs', 'de', 'u')};
end
```

And the parameters used for the transfer function estimation are defined below.

```
________Matlab

% Sampling Time [s]
Ts = (meas_iff_plates{1}.t(end) - (meas_iff_plates{1}.t(1)))/(length(meas_iff_plates{1}.t)-1);

% Hannning Windows
win = hanning(ceil(1/Ts));
```

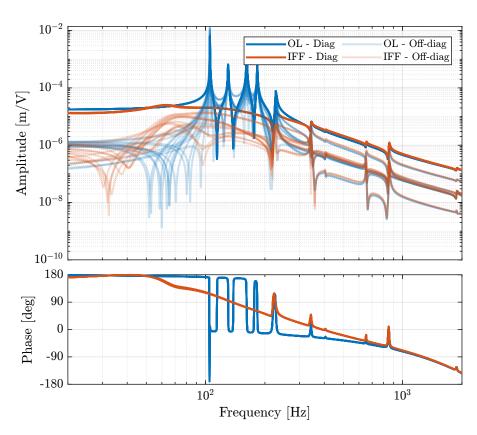


Figure 2.13: Effect of the IFF control strategy on the transfer function from  $\tau$  to  $d\mathcal{L}_m$ 

```
% And we get the frequency vector [~, f] = tfestimate(meas_iff_plates{1}.Va, meas_iff_plates{1}.de, win, [], [], 1/Ts);
```

The estimation is performed using the tfestimate command.

```
Matlab
%% Estimation of the transfer function matrix from u to dL when IFF
is applied
G_enc_iff_opt = zeros(length(f), 6, 6);

for i = 1:6
    G_enc_iff_opt(:,:,i) = tfestimate(meas_iff_plates{i}.Va, meas_iff_plates{i}.de, win, [], [], 1/Ts);
end
```

The obtained diagonal and off-diagonal elements of the transfer function from u' to  $d\mathcal{L}_m$  are shown in Figure 2.14 both without and with IFF.

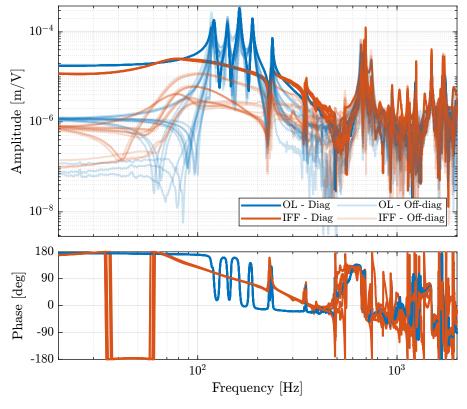


Figure 2.14: Effect of the IFF control strategy on the transfer function from  $\tau$  to  $d\mathcal{L}_m$ 

#### Important

As was predicted with the Simscape model, the IFF control strategy is very effective in damping the suspension modes of the nano-hexapod. Little damping is also applied on the first flexible mode of the strut at 235Hz. However, no damping is applied on other modes, such as the flexible modes of the top plate.

#### 2.3.3 Comparison of the measured FRF and the Simscape model

Let's now compare the obtained damped plants obtained experimentally with the one extracted from Simscape:

- $\bullet$  Figure 2.15: the individual transfer function from  $u_1'$  to the six encoders are comapred
- Figure 2.16: all the diagonal elements are compared
- Figure 2.17: all the off-diagonal elements are compared

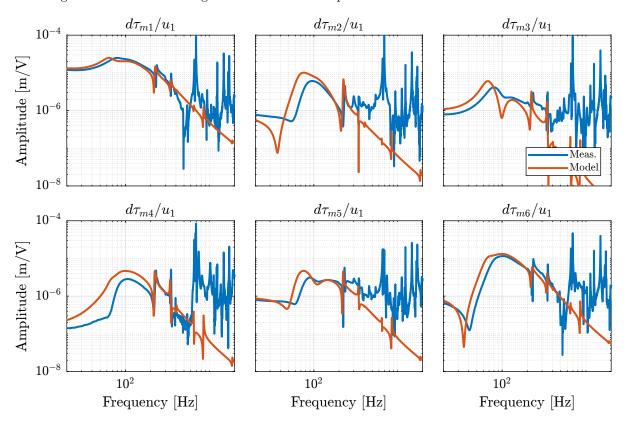


Figure 2.15: FRF from one actuator to all the encoders when the plant is damped using IFF

#### Important

From Figures 2.16 and 2.17, it is clear that the Simscape model very well represents the dynamics of the nano-hexapod. This is true to around 400Hz, then the dynamics depends on the flexible modes of the top plate which are not modelled.

#### 2.3.4 Save Damped Plant

The experimentally identified plant is saved for further use.

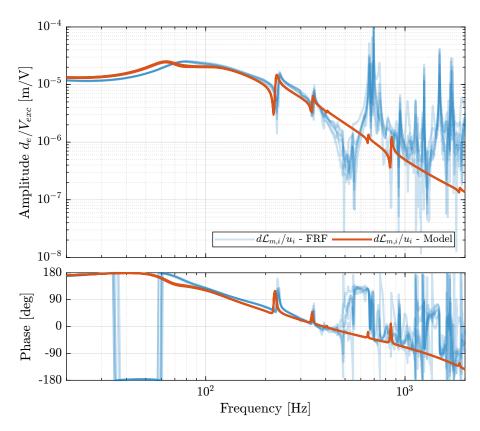


Figure 2.16: Comparison of the diagonal elements of the transfer functions from  $\boldsymbol{u}$  to  $d\mathcal{L}_m$  with active damping (IFF) applied with an optimal gain g=400

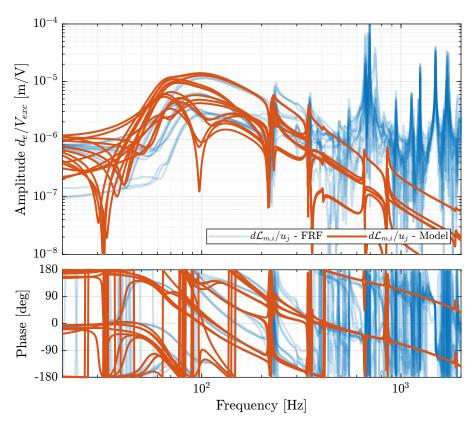


Figure 2.17: Comparison of the off-diagonal elements of the transfer functions from  $\boldsymbol{u}$  to  $d\boldsymbol{\mathcal{L}}_m$  with active damping (IFF) applied with an optimal gain g=400

```
save('matlab/mat/damped_plant_enc_plates.mat', 'f', 'Ts', 'G_enc_iff_opt')
```

```
______Matlab _____save('mat/damped_plant_enc_plates.mat', 'f', 'Ts', 'G_enc_iff_opt')
```

## 2.4 Effect of Payload mass - Robust IFF

In this section, the encoders are fixed to the plates, and we identify the dynamics for several payloads. The added payload are half cylinders, and three layers can be added for a total of around 40kg (Figure 2.18).

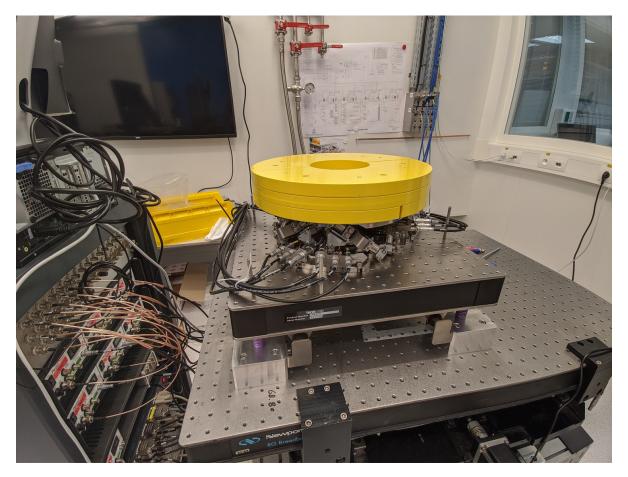


Figure 2.18: Picture of the nano-hexapod with added mass

First the dynamics from u to  $d\mathcal{L}_m$  and  $\tau_m$  is identified. Then, the Integral Force Feedback controller is developed and applied as shown in Figure 2.19. Finally, the dynamics from u' to  $d\mathcal{L}_m$  is identified and the added damping can be estimated.

# 

Figure 2.19: Block Diagram of the experimental setup and model

#### 2.4.1 Measured Frequency Response Functions

The identification is performed without added mass, and with one, two and three layers of added cylinders.

```
i_masses = 0:3;
```

The following data are loaded:

- Va: the excitation voltage (corresponding to  $u_i$ )
- Vs: the generated voltage by the 6 force sensors (corresponding to  $au_m$ )
- de: the measured motion by the 6 encoders (corresponding to  $d\mathcal{L}_m$ )

The window win and the frequency vector f are defined.

```
Matlab

Ts = (meas_added_mass{1,1}.t(end) - (meas_added_mass{1,1}.t(1)))/(length(meas_added_mass{1,1}.t)-1);

Hannning Windows
win = hanning(ceil(1/Ts));

And we get the frequency vector
[~, f] = tfestimate(meas_added_mass{1,1}.Va, meas_added_mass{1,1}.de, win, [], [], 1/Ts);
```

Finally the  $6 \times 6$  transfer function matrices from u to  $d\mathcal{L}_m$  and from u to  $\tau_m$  are identified:

```
Matlab _
  %% DVF Plant (transfer function from u to dLm)
G_dL = {};
  for i_mass = i_masses
                                                     G_dL(i_mass+1) = \{zeros(length(f), 6, 6)\};
                                                                                                   \texttt{G\_dL}\{i\_mass+1\}(:,:,i\_strut) = \texttt{tfestimate}(\texttt{meas\_added\_mass}\{i\_strut,\ i\_mass+1\}. \\ \texttt{Va},\ \texttt{meas\_added\_mass}\{i\_strut,\ i\_mass+1\}. \\ \texttt{de},\ 
                                             win, [], [], 1/Ts);
                                                  end
end
  %% IFF Plant (transfer function from u to taum)
G_{tau} = \{\};
  for i_mass = i_masses
                                                  G_{tau}(i_{mass+1}) = \{zeros(length(f), 6, 6)\};
for i_{strut} = 1:6
                                                                                              \texttt{G\_tau}\{i\_\texttt{mass+1}\}(:,:,i\_\texttt{strut}) = \texttt{tfestimate}(\texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{mass+1}\}. \\ \texttt{Va},\ \texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{mass+1}\}. \\ \texttt{Vs},\ \texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\}. \\ \texttt{Vs},\ \texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\}. \\ \texttt{Vs},\ \texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\{i\_\texttt{strut},\ i\_\texttt{meas\_added\_mass}\}. \\ \texttt{Vs},\ \texttt{
                                                  win, [], [], 1/Ts);
                                                     end
end
```

The identified dynamics are then saved for further use.

```
save('mat/frf_vib_table_m.mat', 'f', 'Ts', 'G_tau', 'G_dL')
Matlab
```

#### 2.4.2 Transfer function from Actuators to Encoders

The transfer functions from  $u_i$  to  $d\mathcal{L}_{m,i}$  are shown in Figure 2.20.

#### Important

From Figure 2.20, we can observe few things:

- The obtained dynamics is changing a lot between the case without mass and when there is at least one added mass.
- Between 1, 2 and 3 added masses, the dynamics is not much different, and it would be easier to design a controller only for these cases.
- The flexible modes of the top plate is first decreased a lot when the first mass is added (from 700Hz to 400Hz). This is due to the fact that the added mass is composed of two half cylinders which are not fixed together. Therefore is adds a lot of mass to the top plate without adding a lot of rigidity in one direction. When more than 1 mass layer is added, the half cylinders are added with some angles such that rigidity are added in all directions (see Figure 2.18). In that case, the frequency of these flexible modes are increased. In practice, the payload should be one solid body, and we should not see a massive decrease of the frequency of this flexible mode.
- Flexible modes of the top plate are becoming less problematic as masses are added.
- First flexible mode of the strut at 230Hz is not much decreased when mass is added. However, its apparent amplitude is much decreased.

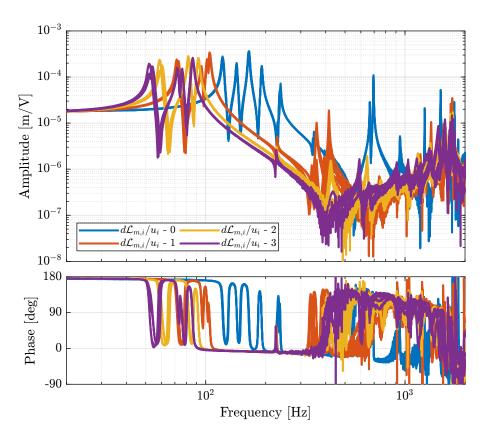


Figure 2.20: Measured Frequency Response Functions from  $u_i$  to  $d\mathcal{L}_{m,i}$  for all 4 payload conditions

#### 2.4.3 Transfer function from Actuators to Force Sensors

The transfer functions from  $u_i$  to  $\tau_{m,i}$  are shown in Figure 2.21.

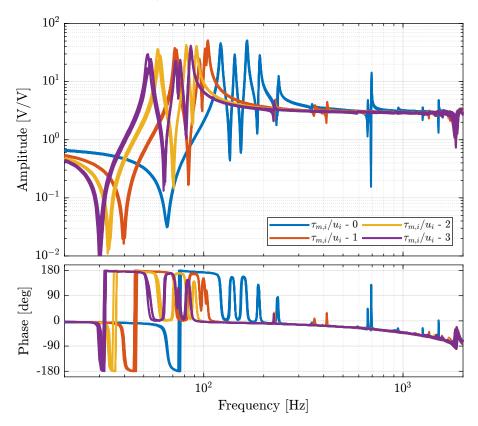


Figure 2.21: Measured Frequency Response Functions from  $u_i$  to  $\tau_{m,i}$  for all 4 payload conditions

#### Important

From Figure 2.21, we can see that for all added payloads, the transfer function from  $u_i$  to  $\tau_{m,i}$  always has alternating poles and zeros.

## 2.5 Comparison with the Simscape model

#### 2.5.1 System Identification

Let's initialize the simscape model with the nano-hexapod fixed on top of the vibration table.

support.type = 1; % On top of vibration table
Matlab

The model of the nano-hexapod is defined as shown bellow:

And finally, we add the same payloads as during the experiments:

```
payload.type = 1; % Payload / 1 "mass layer"
```

First perform the identification for the transfer functions from u to  $d\mathcal{L}_m$ :

```
Matlab

%% Identify the DVF Plant (transfer function from u to dLm)

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Encoders

%% Identification for all the added payloads
G_dL = {};

for i = i_masses
    fprintf('i = %i\n', i)
    payload.type = i;
    G_dL(i+1) = {exp(-s*frf_ol.Ts)*linearize(mdl, io, 0.0, options)};
end
```

```
Matlab

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/Fm'], 1, 'openoutput'); io_i = io_i + 1; % Force Sensors

%% Identification for all the added payloads
G_tau = {};

for i = 0:3
    fprintf('i = %i\n', i)
    payload.type = i;
    G_tau(i+1) = {exp(-s*frf_ol.Ts)*linearize(mdl, io, 0.0, options)};
end
```

The identified dynamics are then saved for further use.

#### 2.5.2 Transfer function from Actuators to Encoders

The measured FRF and the identified dynamics from  $u_i$  to  $d\mathcal{L}_{m,i}$  are compared in Figure 2.22. A zoom near the "suspension" modes is shown in Figure 2.23.

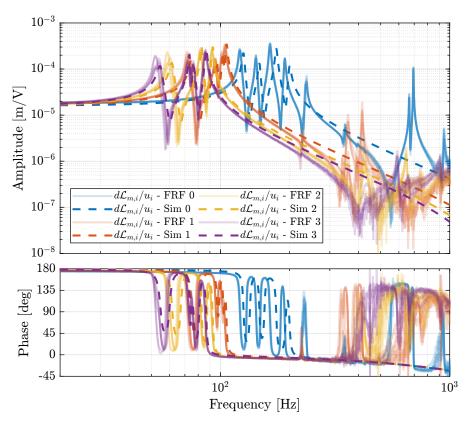


Figure 2.22: Comparison of the transfer functions from  $u_i$  to  $d\mathcal{L}_{m,i}$  - measured FRF and identification from the Simscape model

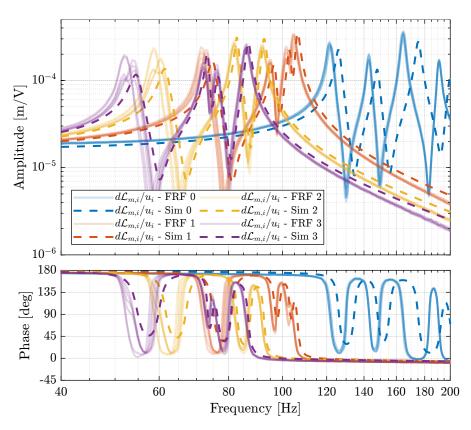


Figure 2.23: Comparison of the transfer functions from  $u_i$  to  $d\mathcal{L}_{m,i}$  - measured FRF and identification from the Simscape model (Zoom)

#### Important

The Simscape model is very accurately representing the measured dynamics up. Only the flexible modes of the struts and of the top plate are not represented here as these elements are modelled as rigid bodies.

#### 2.5.3 Transfer function from Actuators to Force Sensors

The measured FRF and the identified dynamics from  $u_i$  to  $\tau_{m,i}$  are compared in Figure 2.24. A zoom near the "suspension" modes is shown in Figure 2.25.

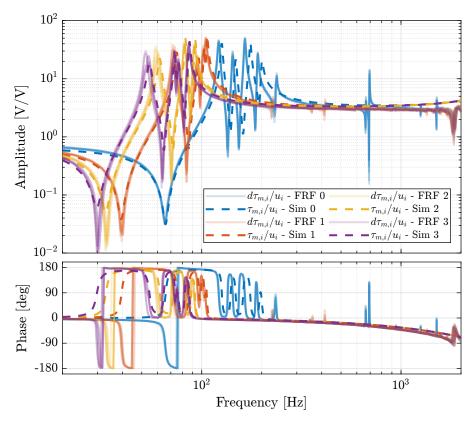


Figure 2.24: Comparison of the transfer functions from  $u_i$  to  $\tau_{m,i}$  - measured FRF and identification from the Simscape model

## 2.6 Integral Force Feedback Controller

#### 2.6.1 Robust IFF Controller

Based on the measured FRF from  $u_i$  to  $\tau_{m,i}$ , the following IFF controller is developed:

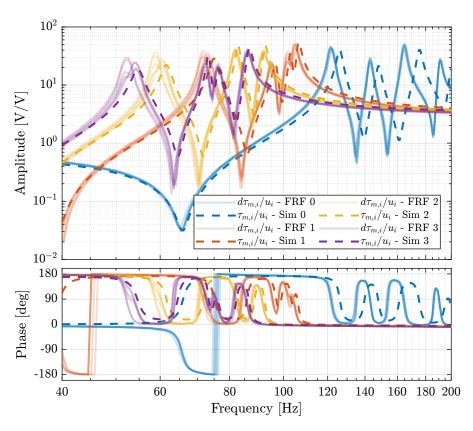


Figure 2.25: Comparison of the transfer functions from  $u_i$  to  $\tau_{m,i}$  - measured FRF and identification from the Simscape model (Zoom)

Then, the Root Locus plot of Figure 2.26 is used to estimate the optimal gain. This Root Locus plot is computed from the Simscape model.

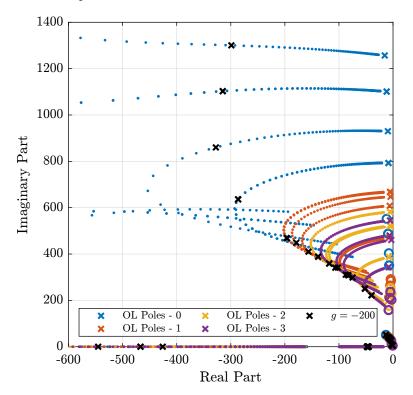


Figure 2.26: Root Locus for the IFF control strategy (for all payload conditions).

The found optimal IFF controller is:

```
Matlab

%% Optimal controller
g_opt = -2e2;
Kiff = g_opt*Kiff_g1*eye(6);
```

It is saved for further use.

```
save('mat/Kiff_opt.mat', 'Kiff')

Matlab
```

The corresponding experimental loop gains are shown in Figure 2.27.

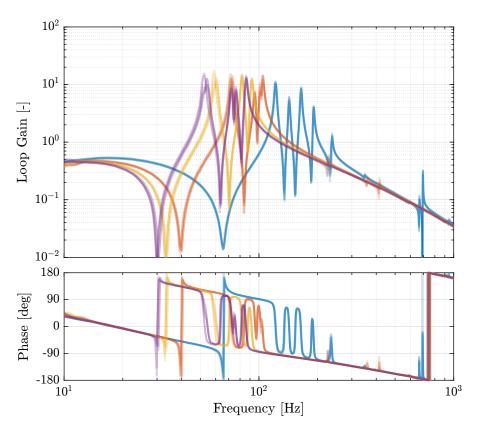


Figure 2.27: Loop gain for the Integral Force Feedback controller

#### Important

Based on the above analysis:

- The same IFF controller can be used to damp the suspension modes for all payload conditions
- The IFF controller should be robust

#### 2.6.2 Estimated Damped Plant from the Simscape model

Let's initialize the simscape model with the nano-hexapod fixed on top of the vibration table.

```
support.type = 1; % On top of vibration table
```

The model of the nano-hexapod is defined as shown bellow:

And finally, we add the same payloads as during the experiments:

```
payload.type = 1; % Payload / 1 "mass layer"
```

```
Matlab

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Plate Displacement (encoder)

%% Identify for all add masses
G_dL = {};

for i = i_masses
    payload.type = i;
    G_dL(i+1) = {exp(-s*frf_ol.Ts)*linearize(mdl, io, 0.0, options)};
end
```

The identified dynamics are then saved for further use.

```
save('mat/sim_iff_vib_table_m.mat', 'G_dL');

Matlab
```

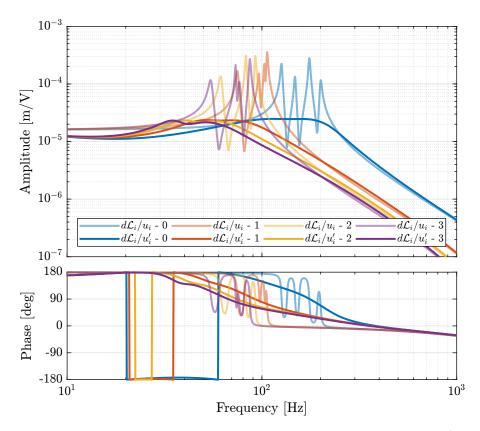


Figure 2.28: Transfer function from  $u_i$  to  $d\mathcal{L}_{m,i}$  (without active damping) and from  $u'_i$  to  $d\mathcal{L}_{m,i}$  (with IFF)

# 2.6.3 Compute the identified FRF with IFF

The identification is performed without added mass, and with one, two and three layers of added cylinders.

```
i_masses = 0:3;
```

The following data are loaded:

- $\bullet$  Va: the excitation voltage for the damped plant (corresponding to  $u_i'$ )
- de: the measured motion by the 6 encoders (corresponding to  $d\mathcal{L}_m$ )

```
for i_mass = i_masses
    for i_strut = 1:6
        meas_iff_mass(i_strut, i_mass+1) = {load(sprintf('frf_data_exc_strut_%i_iff_vib_table_%im.mat', i_strut, i_mass), 't',
        'Va', 'de')};
    end
end
```

The window win and the frequency vector f are defined.

```
% Sampling Time [s]
Ts = (meas_iff_mass{1,1}.t(end) - (meas_iff_mass{1,1}.t(1)))/(length(meas_iff_mass{1,1}.t)-1);

% Hannning Windows
win = hanning(ceil(1/Ts));

% And we get the frequency vector
[~, f] = tfestimate(meas_iff_mass{1,1}.Va, meas_iff_mass{1,1}.de, win, [], [], 1/Ts);
```

Finally the  $6 \times 6$  transfer function matrix from u' to  $d\mathcal{L}_m$  is estimated:

```
Matlab

With DVF Plant (transfer function from u to dLm)

G_dL = {};

for i_mass = i_masses
    G_dL(i_mass+1) = {zeros(length(f), 6, 6)};
    for i_strut = 1:6
        G_dL{i_mass+1}(:,:,i_strut) = tfestimate(meas_iff_mass{i_strut, i_mass+1}.Va, meas_iff_mass{i_strut, i_mass+1}.de, win,

I], [], 1/Ts);
    end
end
```

The identified dynamics are then saved for further use.

```
save('mat/frf_iff_vib_table_m.mat', 'f', 'Ts', 'G_dL');

Matlab
```

## 2.6.4 Comparison of the measured FRF and the Simscape model

The following figures are computed:

- Figure 2.29: the measured damped FRF are displayed
- Figure 2.30: the open-loop and damped FRF are compared (diagonal elements)
- Figure 2.31: the obtained damped FRF is compared with the identified damped from using the Simscape model

#### Important

The IFF control strategy effectively damps all the suspensions modes of the nano-hexapod whatever the payload is. The obtained plant is easier to control (provided the flexible modes of

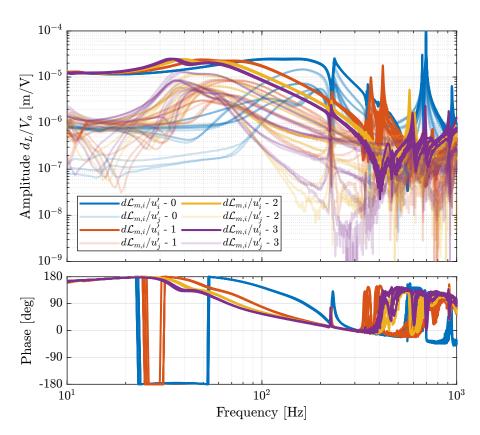
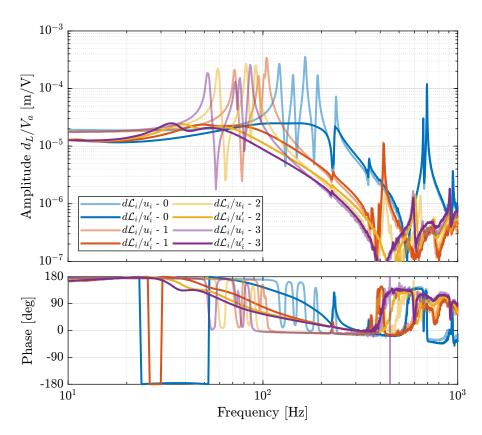


Figure 2.29: Diagonal and off-diagonal of the measured FRF matrix for the damped plant



 ${\bf Figure~2.30:~Damped~and~Undamped~measured~FRF~(diagonal~elements)}$ 

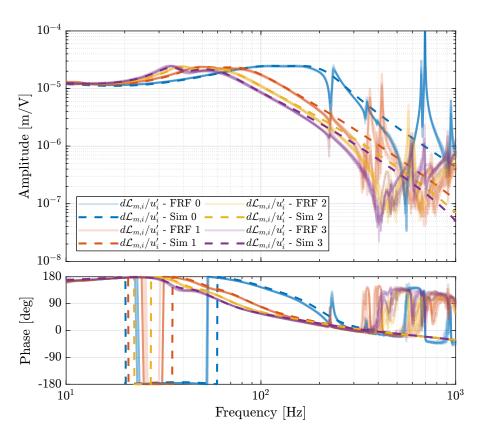


Figure 2.31: Comparison of the measured FRF and the identified dynamics from the Simscape model

# 2.6.5 Change of coupling with IFF

The added damping using IFF reduces the coupling in the system near the suspensions modes that are damped. It can be estimated by taking the ratio of the diagonal-term and the off-diagonal term.

This is shown in Figure 2.32.

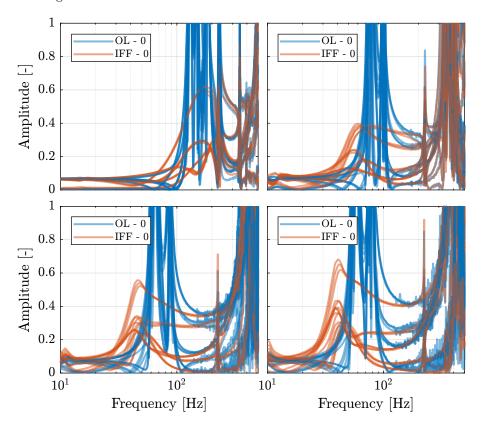


Figure 2.32: Comparison of the coupling with and without IFF

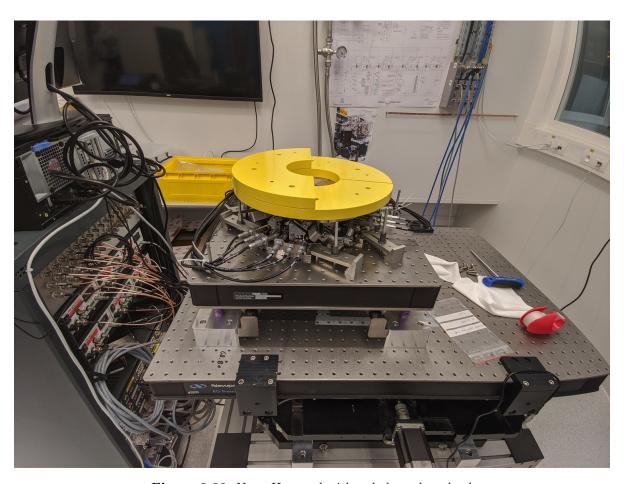
## 2.7 Un-Balanced mass

## 2.7.1 Introduction

# 2.7.2 Compute the identified FRF with IFF

The following data are loaded:

• Va: the excitation voltage for the damped plant (corresponding to  $u_i'$ )



 ${\bf Figure~2.33:~Nano-Hexapod~with~unbalanced~payload}$ 

• de: the measured motion by the 6 encoders (corresponding to  $d\mathcal{L}_m$ )

```
Matlab

*** Load Identification Data
meas_added_mass = {zeros(6,1)};

for i_strut = 1:6
    meas_iff_mass(i_strut) = {load(sprintf('frf_data_exc_strut_%i_iff_vib_table_1m_unbalanced.mat', i_strut), 't', 'Va',
    'de')};
end
```

The window win and the frequency vector f are defined.

```
% Sampling Time [s]
Ts = (meas_iff_mass{1}.t(end) - (meas_iff_mass{1}.t(1)))/(length(meas_iff_mass{1}.t)-1);

% Hannning Windows
win = hanning(ceil(1/Ts));

% And we get the frequency vector
[~, f] = tfestimate(meas_iff_mass{1}.Va, meas_iff_mass{1}.de, win, [], [], 1/Ts);
```

Finally the  $6 \times 6$  transfer function matrix from u' to  $d\mathcal{L}_m$  is estimated:

```
Matlab

%% DVF Plant (transfer function from u to dLm)

G_dL = zeros(length(f), 6, 6);
for i_strut = 1:6
    G_dL(:,:,i_strut) = tfestimate(meas_iff_mass{i_strut}.Va, meas_iff_mass{i_strut}.de, win, [], [], 1/Ts);
end
```

The identified dynamics are then saved for further use.

```
_____Matlab ______
save('mat/frf_iff_unbalanced_vib_table_m.mat', 'f', 'Ts', 'G_dL');
```

## 2.7.3 Effect of an unbalanced payload

The transfer functions from  $u_i$  to  $d\mathcal{L}_i$  are shown in Figure 2.34. Due to the unbalanced payload, the system is not symmetrical anymore, and therefore each of the diagonal elements are not equal. This is due to the fact that each strut is not affected by the same inertia.

## 2.8 Conclusion

### Important

In this section, the dynamics of the nano-hexapod with the encoders fixed to the plates is studied. It has been found that:

• The measured dynamics is in agreement with the dynamics of the simscape model, up to

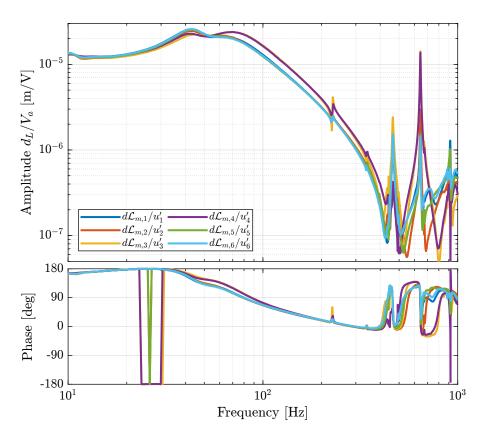


Figure 2.34: Transfer function from  $u_i$  to  $d\mathcal{L}_i$  for the nano-hexapod with an unbalanced payload

- the flexible modes of the top plate. See figures 2.7 and 2.8 for the transfer function to the force sensors and Figures 2.10 and 2.11for the transfer functions to the encoders
- The Integral Force Feedback strategy is very effective in damping the suspension modes of the nano-hexapod (Figure 2.14).
- The transfer function from u' to  $d\mathcal{L}_m$  show nice dynamical properties and is a much better candidate for the high-authority-control than when the encoders were fixed to the struts. At least up to the flexible modes of the top plate, the diagonal elements of the transfer function matrix have alternating poles and zeros, and the phase is moving smoothly. Only the flexible modes of the top plates seems to be problematic for control.

# 3 Decentralized High Authority Control with Integral Force Feedback

In this section is studied the HAC-IFF architecture for the Nano-Hexapod. More precisely:

- The LAC control is a decentralized integral force feedback as studied in Section 2.3
- The HAC control is a decentralized controller working in the frame of the struts

The corresponding control architecture is shown in Figure 3.1 with:

- $r_{\mathcal{X}_n}$ : the  $6 \times 1$  reference signal in the cartesian frame
- $r_{d\mathcal{L}}$ : the 6 × 1 reference signal transformed in the frame of the struts thanks to the inverse kinematic
- $\epsilon_{d\mathcal{L}}$ : the 6 × 1 length error of the 6 struts
- u': input of the damped plant
- $\bullet$  u: generated DAC voltages
- $\tau_m$ : measured force sensors
- $d\mathcal{L}_m$ : measured displacement of the struts by the encoders

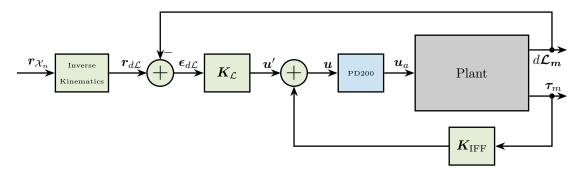


Figure 3.1: HAC-LAC: IFF + Control in the frame of the legs

This part is structured as follow:

- Section 3.1: some reference tracking tests are performed
- Section 3.2: the decentralized high authority controller is tuned using the Simscape model and is implemented and tested experimentally

- Section 3.3: an interaction analysis is performed, from which the best decoupling strategy can be determined
- Section 3.4: Robust High Authority Controller are designed

# 3.1 Reference Tracking - Trajectories

In this section, several trajectories representing the wanted pose (position and orientation) of the top platform with respect to the bottom platform are defined.

These trajectories will be used to test the HAC-LAC architecture.

In order to transform the wanted pose to the wanted displacement of the 6 struts, the inverse kinematic is required. As a first approximation, the Jacobian matrix J can be used instead of using the full inverse kinematic equations.

Therefore, the control architecture with the input trajectory  $r_{\mathcal{X}_n}$  is shown in Figure 3.2.

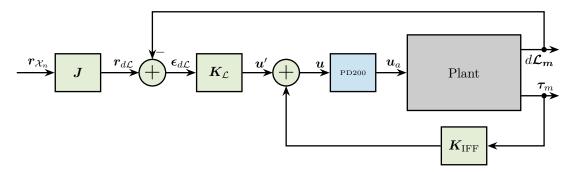


Figure 3.2: HAC-LAC: IFF + Control in the frame of the legs

In the following sections, several reference trajectories are defined:

- Section 3.1.1: simple scans in the Y-Z plane
- Section 3.1.2: scans in tilt are performed
- Section 3.1.3: scans with X-Y-Z translations in order to draw the word "NASS"

## 3.1.1 Y-Z Scans

A function <code>generateYZScanTrajectory</code> has been developed (accessible here) in order to easily generate scans in the Y-Z plane.

For instance, the following generated trajectory is represented in Figure 3.3.

```
Matlab

%% Generate the Y-Z trajectory scan
Rx_yz = generateYZScanTrajectory(...
'y_tot', 4e-6, ... % Length of Y scans [m]
```

```
'z_tot', 4e-6, ... % Total Z distance [m]
'n', 5, ... % Number of Y scans
'Ts', 1e-3, ... % Sampling Time [s]
'ti', 1, ... % Time to go to initial position [s]
'tw', 0, ... % Waiting time between each points [s]
'ty', 0.6, ... % Time for a scan in Y [s]
'tz', 0.2); % Time for a scan in Z [s]
```

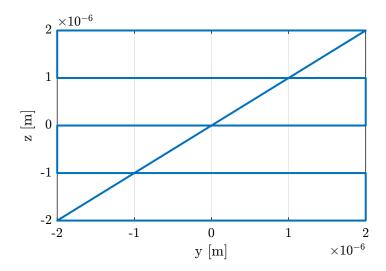


Figure 3.3: Generated scan in the Y-Z plane

The Y and Z positions as a function of time are shown in Figure 3.4.

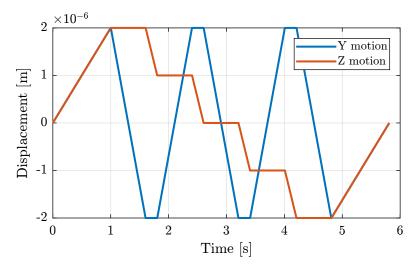


Figure 3.4: Y and Z trajectories as a function of time

Using the Jacobian matrix, it is possible to compute the wanted struts lengths as a function of time:

$$\boldsymbol{r}_{d\mathcal{L}} = \boldsymbol{J} \boldsymbol{r}_{\mathcal{X}_n} \tag{3.1}$$

```
%% Compute the reference in the frame of the legs
dL_ref = [J*Rx_yz(:, 2:7)']';
```

The reference signal for the strut length is shown in Figure 3.5.

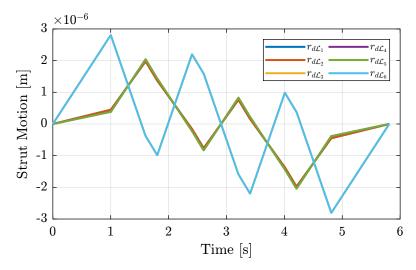


Figure 3.5: Trajectories for the 6 individual struts

#### 3.1.2 Tilt Scans

A function generalSpiralAngleTrajectory has been developed in order to easily generate  $R_x, R_y$  tilt scans.

For instance, the following generated trajectory is represented in Figure 3.6.

```
%% Generate the "tilt-spiral" trajectory scan
R_tilt = generateSpiralAngleTrajectory(...
    'R_tot', 20e-6, ... % Total Tilt [ad]
    'n_tturn', 5, ... % Number of scans
    'Ts', 1e-3, ... % Sampling Time [s]
    't_turn', 1, ... % Turn time [s]
    't_end', 1); % End time to go back to zero [s]
```

The reference signal for the strut length is shown in Figure 3.7.

# 3.1.3 "NASS" reference path

In this section, a reference path that "draws" the work "NASS" is developed.

First, a series of points representing each letter are defined. Between each letter, a negative Z motion is performed.

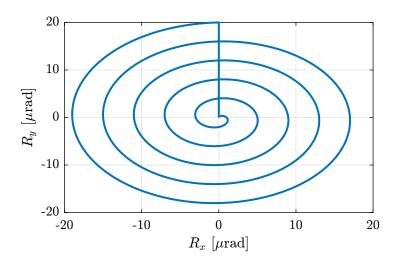


Figure 3.6: Generated "spiral" scan

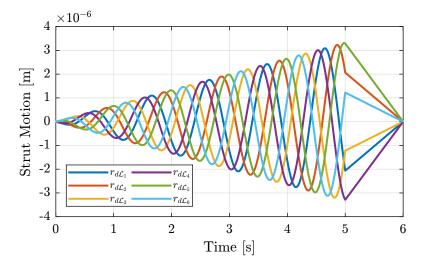


Figure 3.7: Trajectories for the 6 individual struts - Tilt scan

```
Matlab

ref_path = [ ...
    0, 0,0; % Initial Position
    0,0,1; 0,4,1; 3,0,1; 3,4,1; % N
    3,4,0; 4,0,0; % Transition
    4,0,1; 4,3,1; 5,4,1; 6,4,1; 7,3,1; 7,2,1; 4,2,1; 4,3,1; 5,4,1; 6,4,1; 7,3,1; 7,0,1; % A
    7,0,0; 8,0,0; % Transition
    8,0,1; 11,0,1; 11,2,1; 8,2,1; 8,4,1; 11,4,1; % S
    11,4,0; 12,0,0; % Transition
    12,0,1; 15,0,1; 15,2,1; 12,2,1; 12,4,1; 15,4,1; % S
    15,4,0;
    ];

%% Center the trajectory arround zero
    ref_path = ref_path - (max(ref_path) - min(ref_path))/2;

%% Define the X-Y-Z cuboid dimensions containing the trajectory
    X_max = 10e-6;
    Y_max = 4e-6;
    Z_max = 2e-6;
    ref_path = ([X_max, Y_max, Z_max]./max(ref_path)).*ref_path; % [m]
```

Then, using the generateXYZTrajectory function, the  $6 \times 1$  trajectory signal is computed.

```
%% Generating the trajectory
Rx_nass = generateXYZTrajectory('points', ref_path);
```

The trajectory in the X-Y plane is shown in Figure 3.8 (the transitions between the letters are removed).

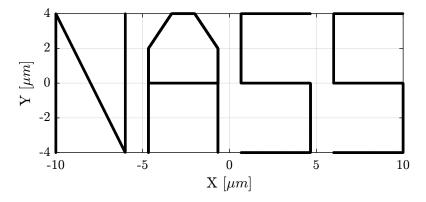


Figure 3.8: Reference path corresponding to the "NASS" acronym

It can also be better viewed in a 3D representation as in Figure 3.9.

# 3.2 First Basic High Authority Controller

In this section, a simple decentralized high authority controller  $K_{\mathcal{L}}$  is developed to work without any payload.

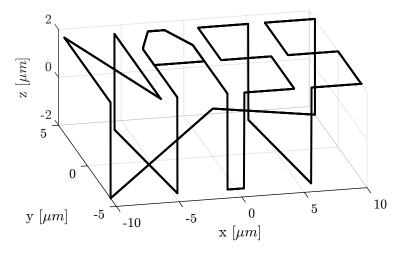


Figure 3.9: Reference path that draws "NASS" - 3D view

The diagonal controller is tuned using classical Loop Shaping in Section 3.2.1. The stability is verified in Section 3.2.2 using the Simscape model.

## 3.2.1 HAC Controller

Let's first try to design a first decentralized controller with:

- a bandwidth of 100Hz
- sufficient phase margin
- simple and understandable components

After some very basic and manual loop shaping, A diagonal controller is developed. Each diagonal terms are identical and are composed of:

- A lead around 100Hz
- A first order low pass filter starting at 200Hz to add some robustness to high frequency modes
- A notch at 700Hz to cancel the flexible modes of the top plate
- A pure integrator

```
Matlab

%% Lead to increase phase margin
a = 2; % Amount of phase lead / width of the phase lead / high frequency gain
wc = 2*pi*100; % Frequency with the maximum phase lead [rad/s]

H_lead = (1 + s/(wc/sqrt(a)))/(1 + s/(wc*sqrt(a)));

%% Low Pass filter to increase robustness
H_lpf = 1/(1 + s/2/pi/200);

%% Notch at the top-plate resonance
gm = 0.02;
xi = 0.3;
wn = 2*pi*700;
```

This controller is saved for further use.

The experimental loop gain is computed and shown in Figure 3.10.

```
______ Matlab 
L_hac_iff_struts = pagemtimes(permute(frf_iff.G_dL{1}, [2 3 1]), squeeze(freqresp(Khac_iff_struts, frf_iff.f, 'Hz')));
```

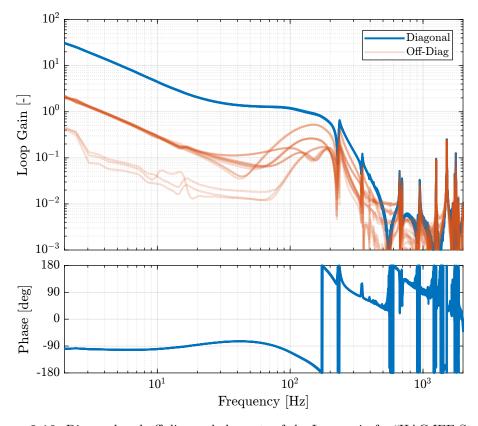


Figure 3.10: Diagonal and off-diagonal elements of the Loop gain for "HAC-IFF-Struts"

## 3.2.2 Verification of the Stability using the Simscape model

The HAC-IFF control strategy is implemented using Simscape.

```
Matlab

%% Identify the (damped) transfer function from u to dLm

clear io; io_i = 1;
io(io_i) = linio([mdl, '/du'], 1, 'openinput'); io_i = io_i + 1; % Actuator Inputs
io(io_i) = linio([mdl, '/dL'], 1, 'openoutput'); io_i = io_i + 1; % Plate Displacement (encoder)
```

We identify the closed-loop system.

```
Matlab

%% Identification
Gd_iff_hac_opt = linearize(mdl, io, 0.0, options);
```

And verify that it is indeed stable.

## 3.2.3 Experimental Validation

Both the Integral Force Feedback controller (developed in Section 2.3) and the high authority controller working in the frame of the struts (developed in Section 3.2) are implemented experimentally.

Two reference tracking experiments are performed to evaluate the stability and performances of the implemented control.

The position of the top-platform is estimated using the Jacobian matrix:

```
Matlab

%% Pose of the top platform from the encoder values
load('jacobian.mat', 'J');

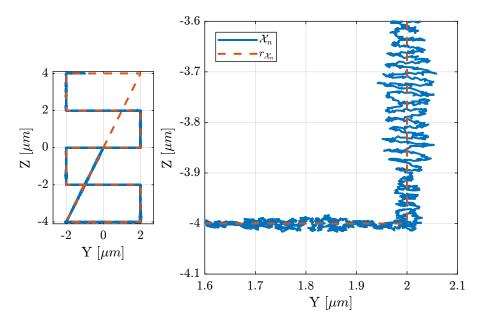
Xe = [inv(J)*de']';
```

```
Matlab

%% Generate the Y-Z trajectory scan

Rx_yz = generateYZScanTrajectory(...
    'y_tot', 4e-6, ... % Length of Y scans [m]
    'z_tot', 8e-6, ... % Total Z distance [m]
    'n', 5, ... % Number of Y scans
    'Ts', 1e-3, ... % Sampling Time [s]
    'ti', 1, ... % Time to go to initial position [s]
    'tw', 0, ... % Waiting time between each points [s]
    'ty', 0.6, ... % Time for a scan in Y [s]
    'tz', 0.2); % Time for a scan in Z [s]
```

The reference path as well as the measured position are partially shown in the Y-Z plane in Figure 3.11.



**Figure 3.11:** Measured position  $\mathcal{X}_n$  and reference signal  $r_{\mathcal{X}_n}$  in the Y-Z plane - Zoom on a change of direction

## Important

It is clear from Figure 3.11 that the position of the nano-hexapod effectively tracks to reference signal. However, oscillations with amplitudes as large as 50nm can be observe.

It turns out that the frequency of these oscillations is 100Hz which is corresponding to the crossover frequency of the High Authority Control loop. This clearly indicates poor stability margins. In the next section, the controller is re-designed to improve the stability margins.

# 3.2.4 Controller with increased stability margins

The High Authority Controller is re-designed in order to improve the stability margins.

The bode plot of the new loop gain is shown in Figure 3.12.

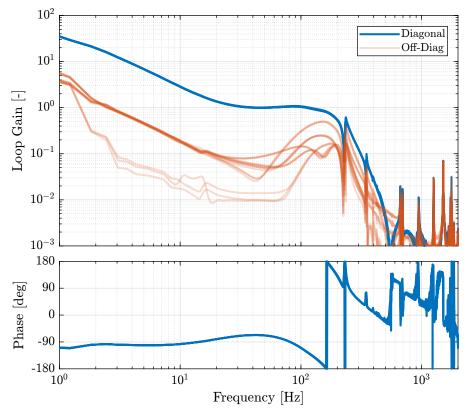


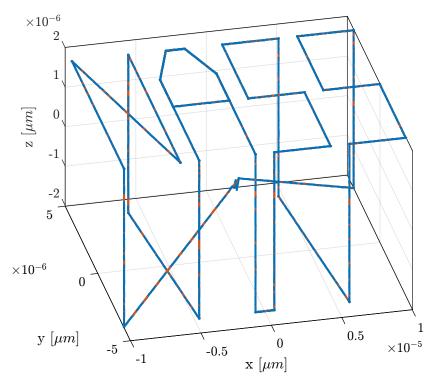
Figure 3.12: Loop Gain for the updated decentralized HAC controller

This new controller is implemented experimentally and several tracking tests are performed.

The pose of the top platform is estimated from the encoder position using the Jacobian matrix.

```
%% Compute the pose of the top platform
load('jacobian.mat', 'J');
Xe = [inv(J)*de']';
```

The measured motion as well as the trajectory are shown in Figure 3.13.



**Figure 3.13:** Measured position  $\mathcal{X}_n$  and reference signal  $r_{\mathcal{X}_n}$  for the "NASS" trajectory

The trajectory and measured motion are also shown in the X-Y plane in Figure 3.14.

The orientation errors during all the scans are shown in Figure 3.15.

## Important

Using the updated High Authority Controller, the nano-hexapod can follow trajectories with high accuracy (the position errors are in the order of 50nm peak to peak, and the orientation errors 300nrad peak to peak).

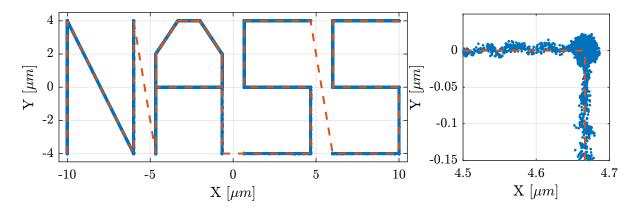


Figure 3.14: Reference path and measured motion in the X-Y plane  $\,$ 

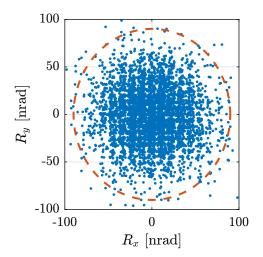


Figure 3.15: Orientation errors during the scan

# 3.3 Interaction Analysis and Decoupling

In this section, the interaction in the identified plant is estimated using the Relative Gain Array (RGA) [2, Chap. 3.4].

Then, several decoupling strategies are compared for the nano-hexapod.

The RGA Matrix is defined as follow:

$$RGA(G(f)) = G(f) \times (G(f)^{-1})^T$$
(3.2)

Then, the RGA number is defined:

$$RGA-num(f) = ||I - RGA(G(f))||_{sum}$$
(3.3)

In this section, the plant with 2 added mass is studied.

#### 3.3.1 Parameters

```
wc = 100; % Wanted crossover frequency [Hz]
[~, i_wc] = min(abs(frf_iff.f - wc)); % Indice corresponding to wc

/// Plant to be decoupled
frf_coupled = frf_iff.G_dL{2};
G_coupled = sim_iff.G_dL{2};
```

## 3.3.2 No Decoupling (Decentralized)



Figure 3.16: Block diagram representing the plant.

## 3.3.3 Static Decoupling

The DC gain is evaluated from the model as be have bad low frequency identification.

-4373.6	-4016.5	660.7	4299.3	3910.6	-62011.5
4281.6	640.2	-4019.2	-4356.8	-61991.2	3914.4
653.8	4295.8	3914.6	-62004.5	-4370.5	-4020.0
-4008.9	-4366.5	-62012.2	3903.3	4292.4	660.9
3919.7	-62006.0	-4377.8	-4025.8	655.6	4302.8
-62019.3	3906.8	4303.7	668.6	-4013.2	-4377.9

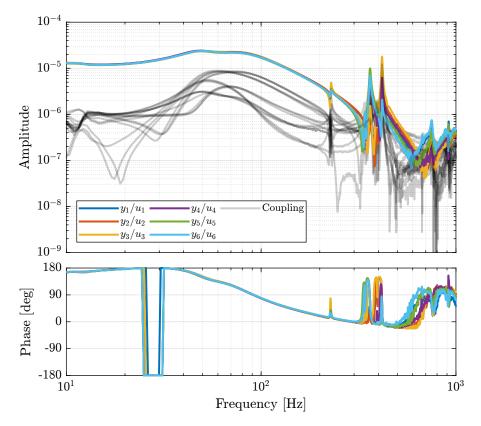


Figure 3.17: Bode Plot of the decentralized plant (diagonal and off-diagonal terms)

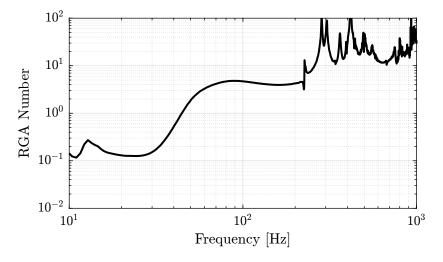


Figure 3.18: RGA number for the decentralized plant

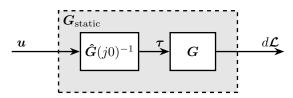


Figure 3.19: Decoupling using the inverse of the DC gain of the plant

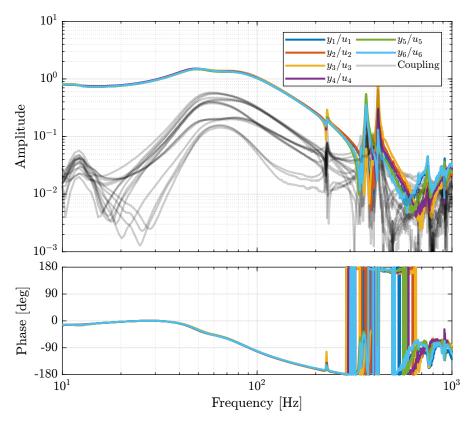


Figure 3.20: Bode Plot of the static decoupled plant

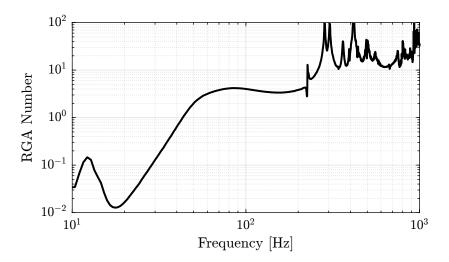
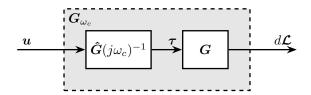


Figure 3.21: RGA number for the statically decoupled plant

# 3.3.4 Decoupling at the Crossover



**Figure 3.22:** Decoupling using the inverse of a dynamical model  $\hat{G}$  of the plant dynamics G

67229.8	3769.3	-13704.6	-23084.8	-6318.2	23378.7
3486.2	67708.9	23220.0	-6314.5	-22699.8	-14060.6
-5731.7	22471.7	66701.4	3070.2	-13205.6	-21944.6
-23305.5	-14542.6	2743.2	70097.6	24846.8	-5295.0
-14882.9	-22957.8	-5344.4	25786.2	70484.6	2979.9
24353.3	-5195.2	-22449.0	-14459.2	2203.6	69484.2

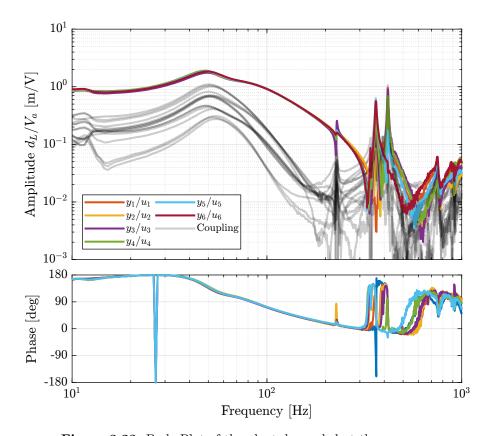


Figure 3.23: Bode Plot of the plant decoupled at the crossover

```
for i = 1:size(RGA_wc, 1)
    RGA_wc_sum(i) = sum(sum(abs(eye(6) - squeeze(RGA_wc(i,:,:)))));
end
```

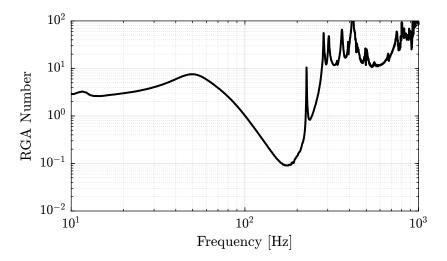


Figure 3.24: RGA number for the plant decoupled at the crossover

# 3.3.5 SVD Decoupling

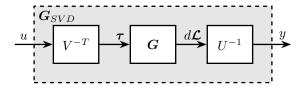


Figure 3.25: Decoupling using the Singular Value Decomposition

```
Matlab

%% Compute the RGA matrix for the SVD decoupled plant
RGA_svd = zeros(size(frf_coupled));
for i = 1:length(frf_iff.f)
    RGA_svd(i,:,:) = squeeze(G_dL_svd(i,:,:)).*inv(squeeze(G_dL_svd(i,:,:))).';
end

%% Compute the RGA-number
RGA_svd_sum = zeros(size(RGA_svd, 1), 1);
for i = 1:length(frf_iff.f)
    RGA_svd_sum(i) = sum(sum(abs(eye(6) - squeeze(RGA_svd(i,:,:)))));
end
```

```
Matlab

%% RGA Number for the SVD decoupled plant
figure;
plot(frf_iff, f, RGA_svd_sum, 'k-');
set(gca, 'XScale', 'log'); set(gca, 'YScale', 'log');
xlabel('Frequency [Hz]'); ylabel('RGA Number');
xlim([10, 1e3]); ylim([1e-2, 1e2]);
```

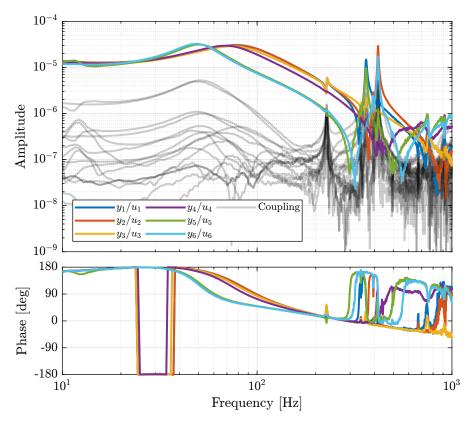


Figure 3.26: Bode Plot of the plant decoupled using the Singular Value Decomposition

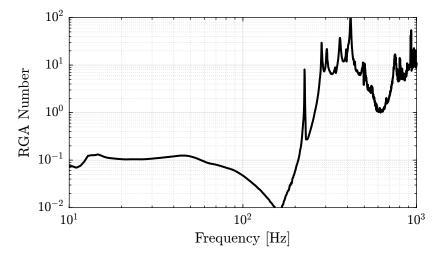
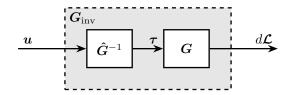


Figure 3.27: RGA number for the plant decoupled using the SVD  $\,$ 

# 3.3.6 Dynamic decoupling



**Figure 3.28:** Decoupling using the inverse of a dynamical model  $\hat{G}$  of the plant dynamics G

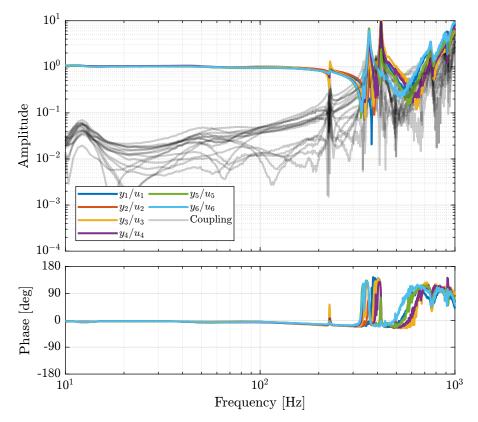


Figure 3.29: Bode Plot of the dynamically decoupled plant

# 3.3.7 Jacobian Decoupling - Center of Stiffness

## 3.3.8 Jacobian Decoupling - Center of Mass

# 3.3.9 Decoupling Comparison

Let's now compare all of the decoupling methods (Figure 3.37).

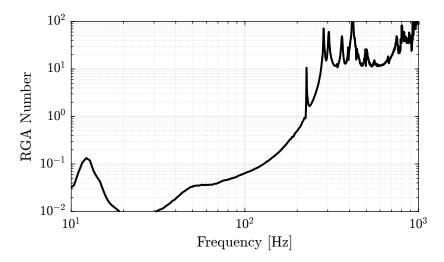


Figure 3.30: RGA number for the dynamically decoupled plant

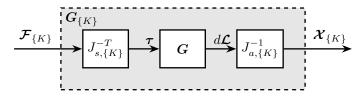


Figure 3.31: Decoupling using Jacobian matrices evaluated at the Center of Stiffness

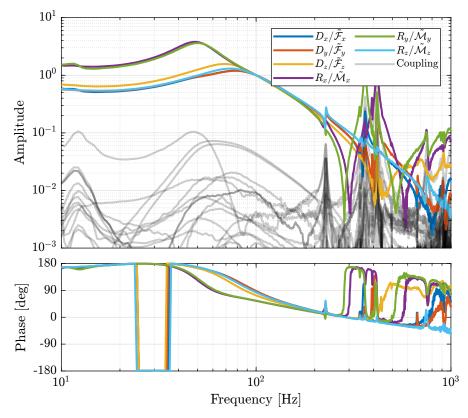


Figure 3.32: Bode Plot of the plant decoupled using the Jacobian evaluated at the "center of stiffness"

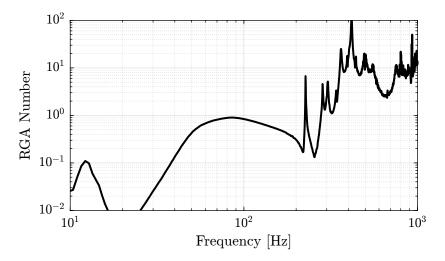


Figure 3.33: RGA number for the plant decoupled using the Jacobian evaluted at the Center of Stiffness

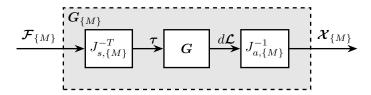


Figure 3.34: Decoupling using Jacobian matrices evaluated at the Center of Mass

## Important

From Figure 3.37, the following remarks are made:

- Decentralized plant: well decoupled below suspension modes
- Static inversion: similar to the decentralized plant as the decentralized plant has already a good decoupling at low frequency
- **Crossover inversion**: the decoupling is improved around the crossover frequency as compared to the decentralized plant. However, the decoupling is increased at lower frequency.
- **SVD decoupling**: Very good decoupling up to 235Hz. Especially between 100Hz and 200Hz.
- **Dynamic Inversion**: the plant is very well decoupled at frequencies where the model is accurate (below 235Hz where flexible modes are not modelled).
- Jacobian Stiffness: good decoupling at low frequency. The decoupling increases at the frequency of the suspension modes, but is acceptable up to the strut flexible modes (235Hz).
- Jacobian Mass: bad decoupling at low frequency. Better decoupling above the frequency of the suspension modes, and acceptable decoupling up to the strut flexible modes (235Hz).

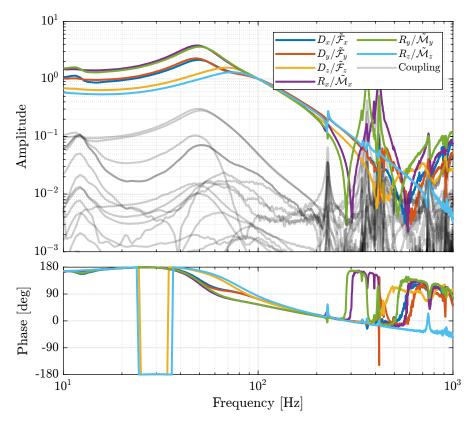


Figure 3.35: Bode Plot of the plant decoupled using the Jacobian evaluated at the Center of Mass

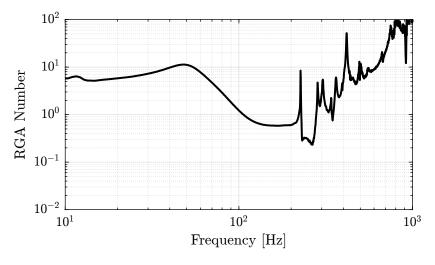


Figure 3.36: RGA number for the plant decoupled using the Jacobian evaluted at the Center of Mass

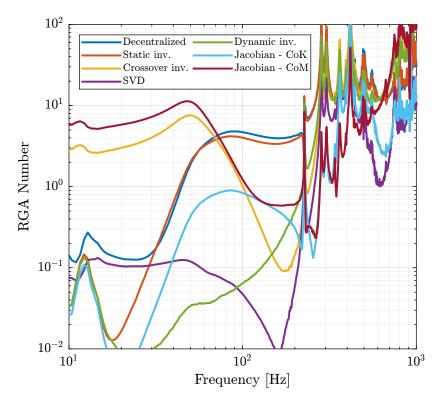


Figure 3.37: Comparison of the obtained RGA-numbers for all the decoupling methods

# 3.3.10 Decoupling Robustness

Let's now see how the decoupling is changing when changing the payload's mass.

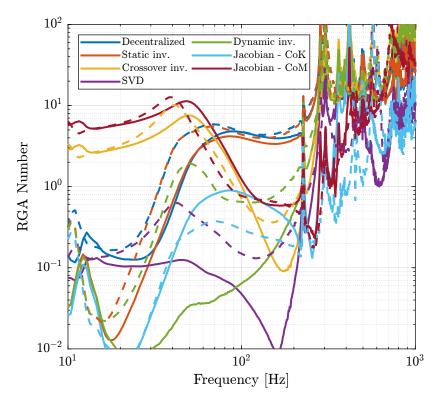
\_\_\_\_\_\_Matlab \_\_\_\_\_\_ frf\_new = frf\_iff.G\_dL{3};

The obtained RGA-numbers are shown in Figure 3.38.

## Important

From Figure 3.38:

• The decoupling using the Jacobian evaluated at the "center of stiffness" seems to give the most robust results.



**Figure 3.38:** Change of the RGA-number with a change of the payload. Indication of the robustness of the inversion method.

## 3.3.11 Conclusion

## ${\bf Important}$

Several decoupling methods can be used:

- SVD
- Inverse
- Jacobian (CoK)

Table 3.1: Summary of the interaction analysis and different decoupling strategies

Method	RGA	Diag Plant	Robustness
Decentralized	_	Equal	++
Static dec.	_	Equal	++
Crossover dec.	-	Equal	0
SVD	++	Diff	+
Dynamic dec.	++	Unity, equal	-
Jacobian - CoK	+	Diff	++
Jacobian - CoM	0	Diff	+

# 3.4 Robust High Authority Controller

In this section we wish to develop a robust High Authority Controller (HAC) that is working for all payloads.

[1]

# 3.4.1 Using Jacobian evaluated at the center of stiffness

## **Decoupled Plant**



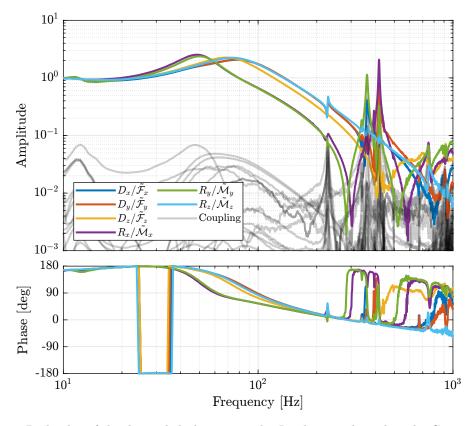


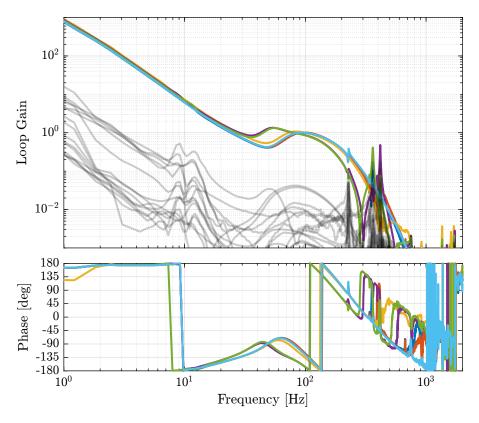
Figure 3.39: Bode plot of the decoupled plant using the Jacobian evaluated at the Center of Stiffness

## SISO Controller Design

As the diagonal elements of the plant are not equal, several SISO controllers are designed and then combined to form a diagonal controller. All the diagonal terms of the controller consists of:

- A double integrator to have high gain at low frequency
- A lead around the crossover frequency to increase stability margins
- Two second order low pass filters above the crossover frequency to increase the robustness to high frequency modes

# **Obtained Loop Gain**

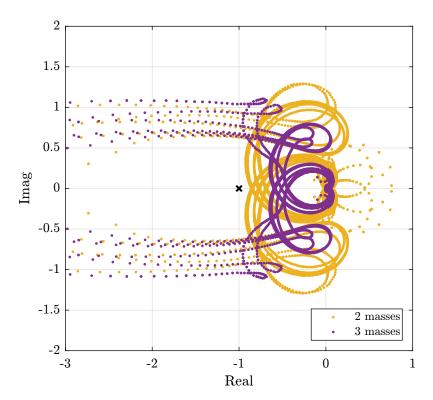


**Figure 3.40:** Bode plot of the Loop Gain when using the Jacobian evaluated at the Center of Stiffness to decouple the system

```
%% Controller to be implemented
Kd = inv(J_cok')*input_normalize*ss(Kd_diag)*inv(Js_cok);
```

# Verification of the Stability

Now the stability of the feedback loop is verified using the generalized Nyquist criteria.



**Figure 3.41:** Loci of  $L(j\omega)$  in the complex plane.

# Save for further analysis

```
save('mat/Khac_iff_struts_jacobian_cok.mat', 'Kd')

Matlab
```

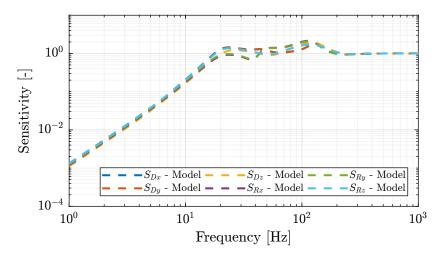
# Sensitivity transfer function from the model

The results are shown in Figure 3.42.

# 3.4.2 Using Singular Value Decomposition

# **Decoupled Plant**

```
G_nom = frf_iff.G_dL{2}; % Nominal Plant Matlab
```



**Figure 3.42:** Estimated sensitivity transfer functions for the HAC controller using the Jacobian estimated at the Center of Stiffness

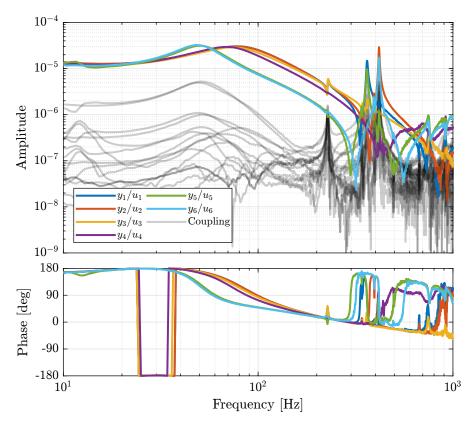


Figure 3.43: Bode plot of the decoupled plant using the SVD

# Controller Design

# Loop Gain

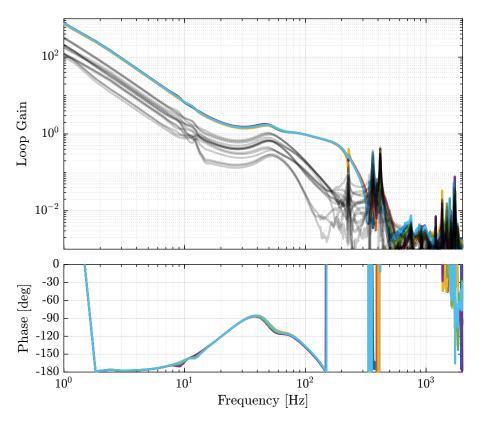


Figure 3.44: Bode plot of Loop Gain when using the SVD

# **Stability Verification**

```
Matlab

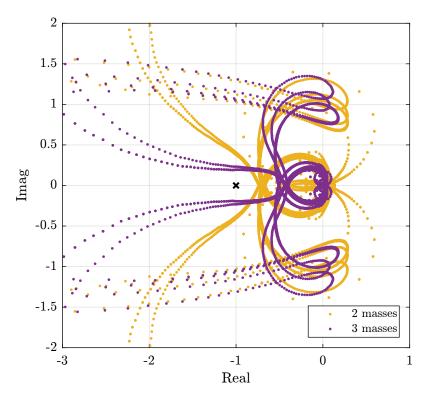
%% Compute the Eigenvalues of the loop gain
Ldet = zeros(3, 6, length(frf_iff.f));

for i = 1:3
    Lmimo = pagemtimes(permute(frf_iff.G_dL{i}, [2,3,1]),squeeze(freqresp(Kd, frf_iff.f, 'Hz')));
    for i_f = 2:length(frf_iff.f)
        Ldet(i,:, i_f) = eig(squeeze(Lmimo(:,:,i_f)));
    end
end
```

# Save for further analysis

```
save('mat/Khac_iff_struts_svd.mat', 'Kd')

Matlab
```



**Figure 3.45:** Locis of  $L(j\omega)$  in the complex plane.

#### Measured Sensitivity Transfer Function

The sensitivity transfer function is estimated by adding a reference signal  $R_x$  consisting of a low pass filtered white noise, and measuring the position error  $E_x$  at the same time.

The transfer function from  $R_x$  to  $E_x$  is the sensitivity transfer function.

In order to identify the sensitivity transfer function for all directions, six reference signals are used, one for each direction.

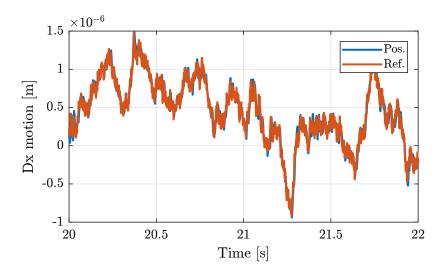
An example is shown in Figure 3.46 where both the reference signal and the measured position are shown for translations in the x direction.

The sensitivity transfer functions estimated for all directions are shown in Figure 3.47.

#### Important

From Figure 3.47:

- The sensitivity transfer functions are similar for all directions
- The disturbance attenuation at 1Hz is almost a factor 1000 as wanted
- The sensitivity transfer functions for  $R_x$  and  $R_y$  have high peak values which indicate poor stability margins.



 ${\bf Figure~3.46:~Reference~position~and~measured~position}$ 

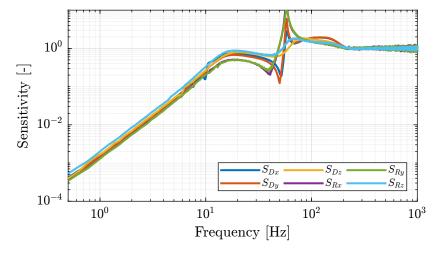


Figure 3.47: Measured diagonal elements of the sensitivity transfer function matrix.

# Sensitivity transfer function from the model

The sensitivity transfer function is now estimated using the model and compared with the one measured.

The results are shown in Figure 3.48. The model is quite effective in estimating the sensitivity transfer functions except around 60Hz were there is a peak for the measurement.

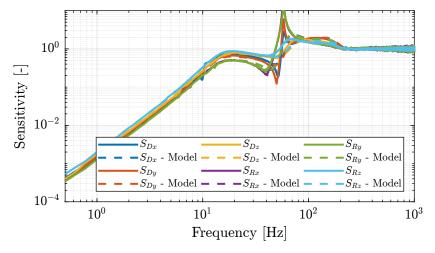


Figure 3.48: Comparison of the measured sensitivity transfer functions with the model

# 4 Nano-Hexapod fixed on the Spindle

# 4.1 Change of dynamics when fixed on the Spindle

#### 4.1.1 Measured Frequency Response Functions

The identification only performed without any payload.

The following data are loaded:

- Va: the excitation voltage (corresponding to  $u_i$ )
- ullet Vs: the generated voltage by the 6 force sensors (corresponding to  $oldsymbol{ au}_m$ )
- de: the measured motion by the 6 encoders (corresponding to  $d\mathcal{L}_m$ )

```
Matlab

%% Load Identification Data
meas_added_mass = {};

for i_strut = 1:6
    meas_added_mass(i_strut) = {load(sprintf('frf_data_exc_strut_%i_spindle_0m.mat', i_strut), 't', 'Va', 'Vs', 'de')};
end
```

The window win and the frequency vector f are defined.

```
% Sampling Time [s]
Ts = (meas_added_mass{1}.t(end) - (meas_added_mass{1}.t(1)))/(length(meas_added_mass{1}.t)-1);

% Hannning Windows
win = hanning(ceil(1/Ts));

% And we get the frequency vector
[~, f] = tfestimate(meas_added_mass{1}.Va, meas_added_mass{1}.de, win, [], [], 1/Ts);
```

Finally the  $6 \times 6$  transfer function matrices from u to  $d\mathcal{L}_m$  and from u to  $\tau_m$  are identified:

```
Matlab

%% DVF Plant (transfer function from u to dLm)

G_dL = zeros(length(f), 6, 6);

for i_strut = 1:6
    G_dL(:,:,i_strut) = tfestimate(meas_added_mass{i_strut}.Va, meas_added_mass{i_strut}.de, win, [], [], 1/Ts);
end

%% IFF Plant (transfer function from u to taum)

G_tau = zeros(length(f), 6, 6);

for i_strut = 1:6
```

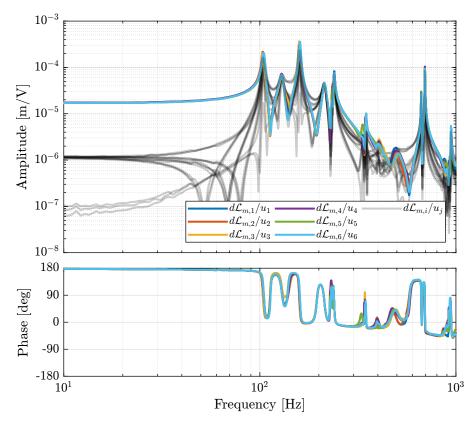
```
 \texttt{G\_tau(:,:,i\_strut)} = \texttt{tfestimate(meas\_added\_mass\{i\_strut\}.Va, meas\_added\_mass\{i\_strut\}.Vs, win, [], [], 1/Ts); \\ \textbf{end}
```

The identified dynamics are then saved for further use.

```
save('mat/frf_spindle_m.mat', 'f', 'Ts', 'G_tau', 'G_dL')
Matlab
```

# 4.1.2 Transfer function from Actuator to Encoder

The transfer functions from  $u_i$  to  $d\mathcal{L}_{m,i}$  are shown in Figure 4.1.

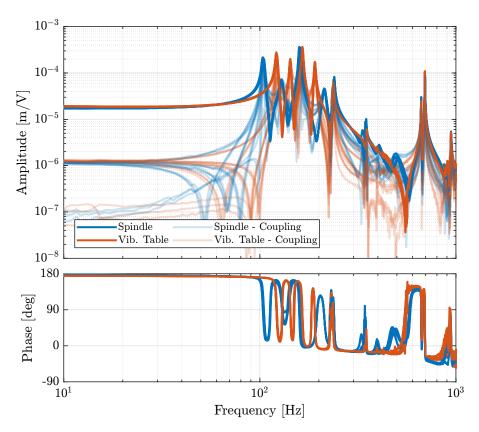


**Figure 4.1:** Measured Frequency Response Functions from  $u_i$  to  $d\mathcal{L}_{m,i}$  when the nano-hexapod is fixed to the Spindle

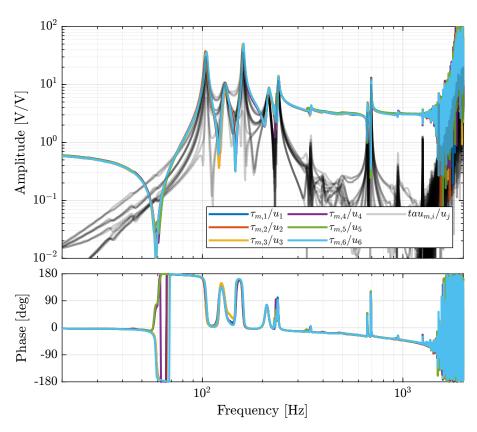
The dynamics of the nano-hexapod when fixed on the Spindle is compared with the dynamics when the nano-hexapod is fixed on the "vibration table" in Figure 4.2.

# 4.1.3 Transfer function from Actuator to Force Sensor

The transfer functions from  $u_i$  to  $\tau_m$  are shown in Figure 4.3.

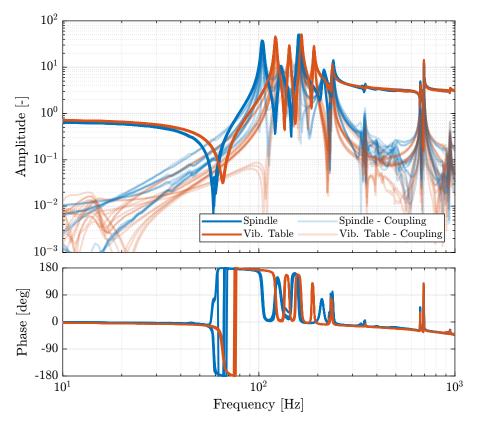


**Figure 4.2:** Comparison of the dynamics from u to  $d\mathcal{L}$  when the nano-hexapod is fixed on top of the Spindle and when it is fixed on top of the "Vibration Table".



**Figure 4.3:** Measured Frequency Response Functions from  $u_i$  to  $\tau_{m,i}$  when the nano-hexapod is fixed to the Spindle

The dynamics of the nano-hexapod when fixed on the Spindle is compared with the dynamics when the nano-hexapod is fixed on the "vibration table" in Figure 4.4.



**Figure 4.4:** Comparison of the dynamics from u to  $d\mathcal{L}$  when the nano-hexapod is fixed on top of the Spindle and when it is fixed on top of the "Vibration Table".

#### 4.1.4 Conclusion

#### Important

The dynamics of the nano-hexapod does not change a lot when it is fixed to the Spindle. The "suspension" modes are just increased a little bit due to the added stiffness of the spindle as compared to the vibration table.

# 4.2 Dynamics of the Damped plant

As the dynamics is not much changed when the nano-hexapod is fixed on top of the Spindle, the same IFF controller is used to damp the plant.

#### 4.2.1 Measured Frequency Response Functions

The identification is performed without added mass, and with one, two and three layers of added cylinders.

```
i_masses = 0:3;
```

The following data are loaded:

- Va: the excitation voltage (corresponding to  $u_i$ )
- Vs: the generated voltage by the 6 force sensors (corresponding to  $\tau_m$ )
- de: the measured motion by the 6 encoders (corresponding to  $d\mathcal{L}_m$ )

```
Matlab

### Load Identification Data
meas_added_mass = {};

for i_mass = i_masses
    for i_strut = 1:6
        meas_added_mass(i_strut, i_mass+1) = {load(sprintf('frf_data_exc_strut_%i_spindle_%im_iff.mat', i_strut, i_mass), 't',
        'Va', 'Vs', 'de')};
    end
end
```

The window win and the frequency vector f are defined.

```
% Sampling Time [s]
Ts = (meas_added_mass{1,1}.t(end) - (meas_added_mass{1,1}.t(1)))/(length(meas_added_mass{1,1}.t)-1);

% Hannning Windows
win = hanning(ceil(1/Ts));

% And we get the frequency vector
[~, f] = tfestimate(meas_added_mass{1,1}.Va, meas_added_mass{1,1}.de, win, [], [], 1/Ts);
```

Finally the  $6 \times 6$  transfer function matrices from u to  $d\mathcal{L}_m$  and from u to  $\tau_m$  are identified:

```
Matlab

Windows DVF Plant (transfer function from u to dLm)

G_dL = {};

for i_mass = i_masses
    G_dL(i_mass+1) = {zeros(length(f), 6, 6)};
    for i_strut = 1:6
        G_dL(i_mass+1)(:,:,i_strut) = tfestimate(meas_added_mass{i_strut, i_mass+1}.Va, meas_added_mass{i_strut, i_mass+1}.de,

win, [], [], 1/Ts);
    end
end
```

The identified dynamics are then saved for further use.

# 4.2.2 Effect of Integral Force Feedback

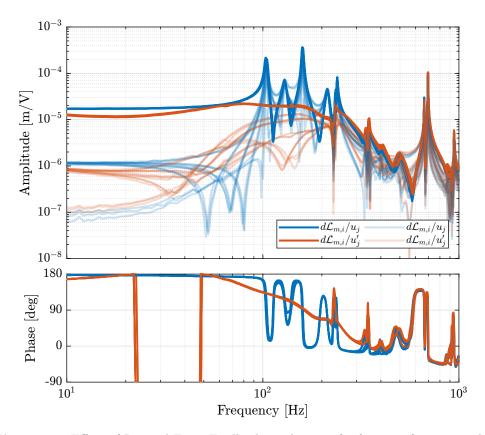


Figure 4.5: Effect of Integral Force Feedback on the transfer function from  $u_i$  to  $d\mathcal{L}_i$ 

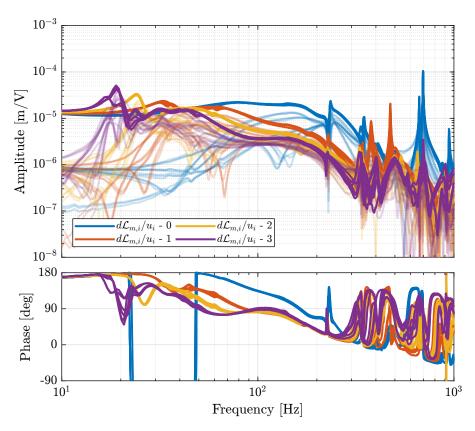
# 4.2.3 Effect of the payload

#### Important

From Figure 4.6 we can see that the coupling is quite large when payloads are added to the nano-hexapod. This was not the case when the nano-hexapod was fixed to the vibration table.

#### Question

What is causing the resonances at 20Hz, 25Hz and 30Hz when there is some added payload? Why the coupling is much larger than when the nano-hexapod was on top of the isolation table?



**Figure 4.6:** Effect of the payload on the transfer functions from  $u_i'$  to  $d\mathcal{L}_i$ 

# 4.2.4 Effect of rotation

# Important

The identified plants with and without spindle's rotation are compared in Figure 4.7. It is shown that the rotational speed as little effect on the plant dynamics.

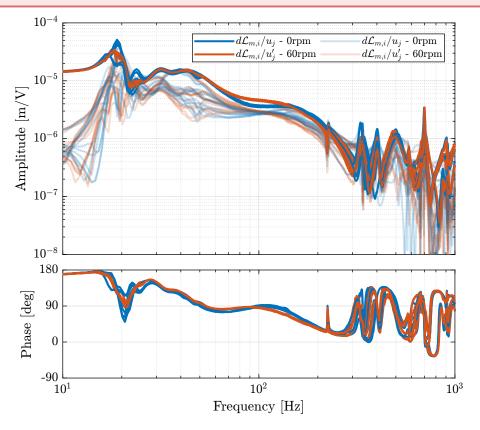


Figure 4.7: Comparison of the damped plant when the spindle is not rotating and when it is rotating at 60 RPM

# 5 Functions

# 5.1 generateXYZTrajectory

# Function description

```
function [ref] = generateXYZTrajectory(args)
% generateXYZTrajectory -
%
% Syntax: [ref] = generateXYZTrajectory(args)
%
% Inputs:
% - args
%
% Outputs:
% - ref - Reference Signal
```

# **Optional Parameters**

```
arguments
args.points double {mustBeNumeric} = zeros(2, 3) % [m]

args.ti (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % Time to go to first point and after last point [s]
args.tw (1,1) double {mustBeNumeric, mustBeNonnegative} = 0.5 % Time wait between each point [s]
args.tm (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % Motion time between points [s]

args.Ts (1,1) double {mustBeNumeric, mustBePositive} = 1e-3 % Sampling Time [s]
end
```

#### Initialize Time Vectors

```
time_i = 0:args.Ts:args.ti;
time_w = 0:args.Ts:args.tw;
time_m = 0:args.Ts:args.tm;
Matlab
```

# **XYZ** Trajectory

```
% Go to initial position
xyz = (args.points(1,:))'*(time_i/args.ti);

% Wait
xyz = [xyz, xyz(:,end).*ones(size(time_w))];

% Scans
for i = 2:size(args.points, 1)
    % Go to next point
    xyz = [xyz, xyz(:,end) + (args.points(i,:)' - xyz(:,end))*(time_m/args.tm)];
    % Wait a litle bit
    xyz = [xyz, xyz(:,end).*ones(size(time_w))];
end

% End motion
xyz = [xyz, xyz(:,end) - xyz(:,end)*(time_i/args.ti)];
```

# Reference Signal

```
t = 0:args.Ts:args.Ts*(length(xyz) - 1);

ref = zeros(length(xyz), 7);

ref(:, 1) = t;
 ref(:, 2:4) = xyz';
Matlab
```

# 5.2 generateYZScanTrajectory

# Function description

```
function [ref] = generateYZScanTrajectory(args)
% generateYZScanTrajectory -
%
% Syntax: [ref] = generateYZScanTrajectory(args)
%
% Inputs:
% - args
%
% Outputs:
% - ref - Reference Signal
```

# **Optional Parameters**

```
arguments
args.y_tot (1,1) double {mustBeNumeric, mustBePositive} = 10e-6 % [m]
```

```
args.z_tot (1,1) double {mustBeNumeric, mustBePositive} = 10e-6 % [m]
args.n (1,1) double {mustBeInteger, mustBePositive} = 10 % [-]

args.Ts (1,1) double {mustBeNumeric, mustBePositive} = 1e-4 % [s]

args.ti (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % [s]
args.tw (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % [s]
args.ty (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % [s]
args.tz (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % [s]
args.tz (1,1) double {mustBeNumeric, mustBeNonnegative} = 1 % [s]
end
```

#### Initialize Time Vectors

```
time_i = 0:args.Ts:args.ti;
time_w = 0:args.Ts:args.tw;
time_y = 0:args.Ts:args.ty;
time_z = 0:args.Ts:args.tz;
Matlab
```

#### Y and Z vectors

```
% Go to initial position
z = (time_i/args.ti)*(args.z_tot/2);

% Wait
z = [z, z(end)*ones(size(time_w))];

% Scans
for i = 1:args.n
    z = [z, z(end)*ones(size(time_y))];

if i < args.n
    z = [z, z(end) - (time_z/args.tz)*args.z_tot/(args.n-1)];</pre>
```

```
end
end

% Wait a litle bit
z = [z, z(end)*ones(size(time_w))];

% End motion
z = [z, z(end) - z(end)*time_i/args.ti];
```

# Reference Signal

```
t = 0:args.Ts:args.Ts*(length(y) - 1);

Matlab

ref = zeros(length(y), 7);

ref(:, 1) = t;
 ref(:, 3) = y;
 ref(:, 4) = z;
```

# 5.3 generateSpiralAngleTrajectory

# Function description

```
function [ref] = generateSpiralAngleTrajectory(args)
% generateSpiralAngleTrajectory -
%
% Syntax: [ref] = generateSpiralAngleTrajectory(args)
%
% Inputs:
% - args
%
% Outputs:
% - ref - Reference Signal
```

# **Optional Parameters**

```
arguments

args.R_tot (1,1) double {mustBeNumeric, mustBePositive} = 10e-6 % [rad]

args.n_turn (1,1) double {mustBeInteger, mustBePositive} = 5 % [-]

args.Ts (1,1) double {mustBeNumeric, mustBePositive} = 1e-3 % [s]

args.t_turn (1,1) double {mustBeNumeric, mustBePositive} = 1 % [s]

args.t_end (1,1) double {mustBeNumeric, mustBePositive} = 1 % [s]

end
```

#### Initialize Time Vectors

```
time_s = 0:args.Ts:args.n_turn*args.t_turn;
time_e = 0:args.Ts:args.t_end;
Matlab
```

# Rx and Ry vectors

```
Rx = sin(2*pi*time_s/args.t_turn).*(args.R_tot*time_s/(args.n_turn*args.t_turn));
Ry = cos(2*pi*time_s/args.t_turn).*(args.R_tot*time_s/(args.n_turn*args.t_turn));

Rx = [Rx, 0*time_e];
Ry = [Ry, Ry(end) - Ry(end)*time_e/args.t_end];
```

# Reference Signal

```
matlab

ref = zeros(length(Rx), 7);

ref(:, 1) = t;
 ref(:, 5) = Rx;
 ref(:, 6) = Ry;
Matlab
```

# 5.4 getTransformationMatrixAcc

# Function description

# Transformation matrix from motion of the solid body to accelerometer measurements

Let's try to estimate the x-y-z acceleration of any point of the solid body from the acceleration/angular acceleration of the solid body expressed in  $\{O\}$ . For any point  $p_i$  of the solid body (corresponding to an accelerometer), we can write:

$$\begin{bmatrix} a_{i,x} \\ a_{i,y} \\ a_{i,z} \end{bmatrix} = \begin{bmatrix} \dot{v}_x \\ \dot{v}_y \\ \dot{v}_z \end{bmatrix} + p_i \times \begin{bmatrix} \dot{\omega}_x \\ \dot{\omega}_y \\ \dot{\omega}_z \end{bmatrix}$$
 (5.1)

We can write the cross product as a matrix product using the skew-symmetric transformation:

$$\begin{bmatrix} a_{i,x} \\ a_{i,y} \\ a_{i,z} \end{bmatrix} = \begin{bmatrix} \dot{v}_x \\ \dot{v}_y \\ \dot{v}_z \end{bmatrix} + \underbrace{\begin{bmatrix} 0 & p_{i,z} & -p_{i,y} \\ -p_{i,z} & 0 & p_{i,x} \\ p_{i,y} & -p_{i,x} & 0 \end{bmatrix}}_{P_{i,|\times|}} \cdot \begin{bmatrix} \dot{\omega}_x \\ \dot{\omega}_y \\ \dot{\omega}_z \end{bmatrix}$$
(5.2)

If we now want to know the (scalar) acceleration  $a_i$  of the point  $p_i$  in the direction of the accelerameter direction  $\hat{s}_i$ , we can just project the 3d acceleration on  $\hat{s}_i$ :

$$a_{i} = \hat{s}_{i}^{T} \cdot \begin{bmatrix} a_{i,x} \\ a_{i,y} \\ a_{i,z} \end{bmatrix} = \hat{s}_{i}^{T} \cdot \begin{bmatrix} \dot{v}_{x} \\ \dot{v}_{y} \\ \dot{v}_{z} \end{bmatrix} + (\hat{s}_{i}^{T} \cdot P_{i,[\times]}) \cdot \begin{bmatrix} \dot{\omega}_{x} \\ \dot{\omega}_{y} \\ \dot{\omega}_{z} \end{bmatrix}$$

$$(5.3)$$

Which is equivalent as a simple vector multiplication:

$$a_{i} = \begin{bmatrix} \hat{s}_{i}^{T} & \hat{s}_{i}^{T} \cdot P_{i,[\times]} \end{bmatrix} \begin{bmatrix} \dot{v}_{x} \\ \dot{v}_{y} \\ \dot{v}_{z} \\ \dot{\omega}_{x} \\ \dot{\omega}_{y} \\ \dot{\omega}_{z} \end{bmatrix} = \begin{bmatrix} \hat{s}_{i}^{T} & \hat{s}_{i}^{T} \cdot P_{i,[\times]} \end{bmatrix}^{O} \vec{x}$$

$$(5.4)$$

And finally we can combine the 6 (line) vectors for the 6 accelerometers to write that in a matrix form. We obtain Eq. (5.5).

#### Important

The transformation from solid body acceleration  $\vec{O}\vec{x}$  from sensor measured acceleration  $\vec{a}$  is:

$$\vec{a} = \underbrace{\begin{bmatrix} \hat{s}_{1}^{T} & \hat{s}_{1}^{T} \cdot P_{1,[\times]} \\ \vdots & \vdots \\ \hat{s}_{6}^{T} & \hat{s}_{6}^{T} \cdot P_{6,[\times]} \end{bmatrix}}_{M} O \vec{x}$$
(5.5)

with  $\hat{s}_i$  the unit vector representing the measured direction of the i'th accelerometer expressed in frame  $\{O\}$  and  $P_{i,[\times]}$  the skew-symmetric matrix representing the cross product of the position of the i'th accelerometer expressed in frame  $\{O\}$ .

Let's define such matrix using matlab:

# 5.5 getJacobianNanoHexapod

# Function description

```
function [J] = getJacobianNanoHexapod(Hbm)
% getJacobianNanoHexapod -
%
% Syntax: [J] = getJacobianNanoHexapod(Hbm)
%
% Inputs:
%    - Hbm - Height of {B} w.r.t. {M} [m]
%
% Outputs:
%    - J - Jacobian Matrix
```

# Transformation matrix from motion of the solid body to accelerometer measurements

# Bibliography

- [1] Marina Indri and Roberto Oboe. *Mechatronics and Robotics: New Trends and Challenges*. CRC Press, 2020.
- [2] Sigurd Skogestad and Ian Postlethwaite. *Multivariable Feedback Control: Analysis and Design Second Edition*. John Wiley, 2007. ISBN: 978-0470011683.