

# Identification for Control

A rather extreme example in helicopter vibration control

# Problem statement



Vibration from rotors leads to pilot fatigue

Passive vibration damping system needs active control

Feedback from accelerometers near pilot to active dampers

Hydraulic system coupled with spring plates

Under-actuated but expect improvement over passive control

# System information a priori

Range of vibration frequencies is very limited  $17\text{Hz} \pm 5\%$

An under-damped mode lies around 40Hz

System is stable in open-loop

No control produces no craziness

The vibration frequency changes only slightly in flight

The airframe dynamics change with configuration and load

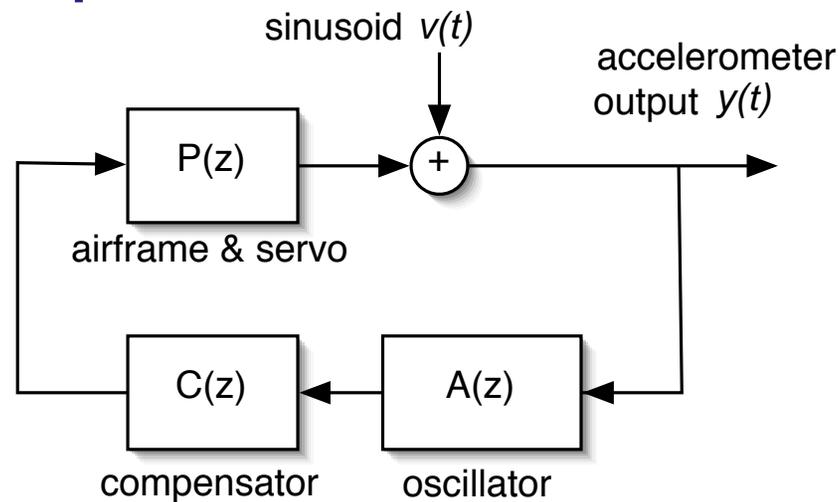
Look for a stable model with a simple parametrization

Control design will use open-loop stability - low gain solution

Adaptive solution for changing dynamics and frequency

Slow changes

# The putative control solution



$$y(t) = [1 + PCA]^{-1} v(t)$$

Performance: need  $1+PCA$  very large at vibration frequency 17Hz

Need  $CA$  small near the unmodeled resonance at 40Hz

Robust stability given by  $CA$  almost zero

Look for a perturbation on  $CA=0$

$A(z)$  an oscillator

$C(z)$  a phase compensator

Stability tied to the phase of  $C(z)$  at the oscillation frequency

# Model requirements

## Low-order simple model

High frequency detail managed by the controller not modeling

Parametric model is good for adaptation

Want few parameters to estimate

Data is very sinusoidal - not informative

## The model should be stable just like the real plant

A low-gain control strategy is adopted

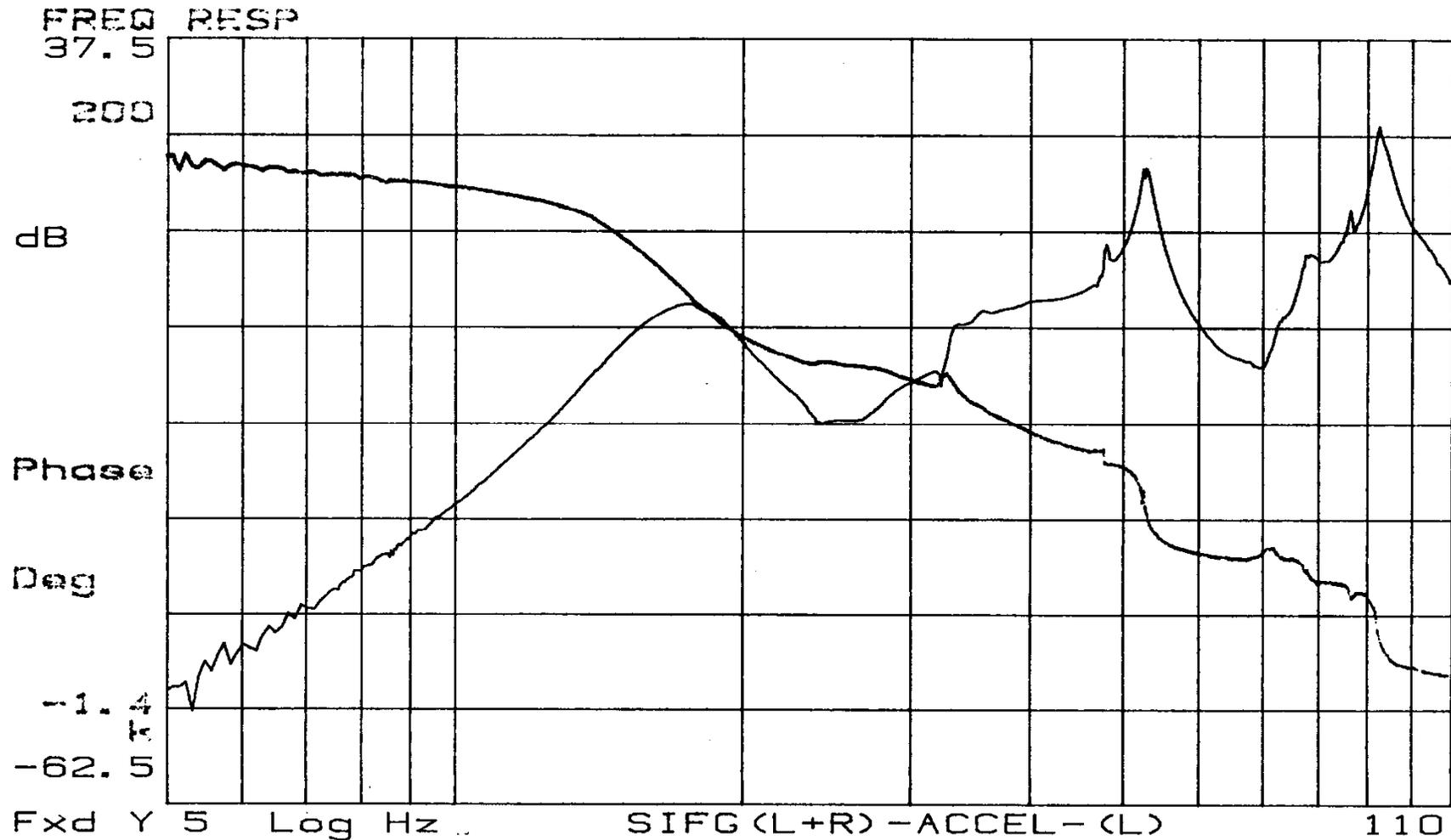
Matches the robust stability approach

## The fit in the neighborhood of 15Hz to 20Hz is most important

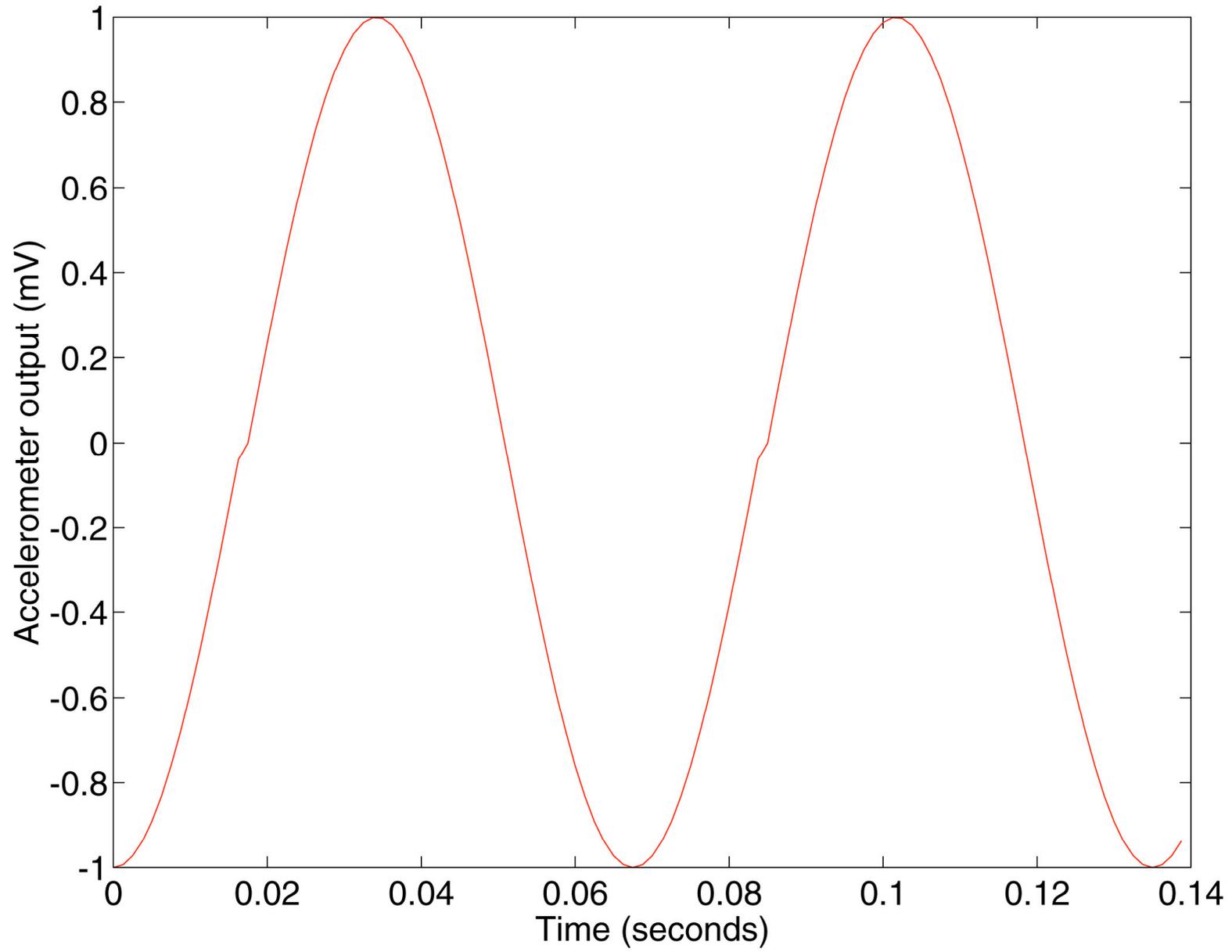
Actuation incapable of addressing other modes

Accommodation of model mismatch by the controller

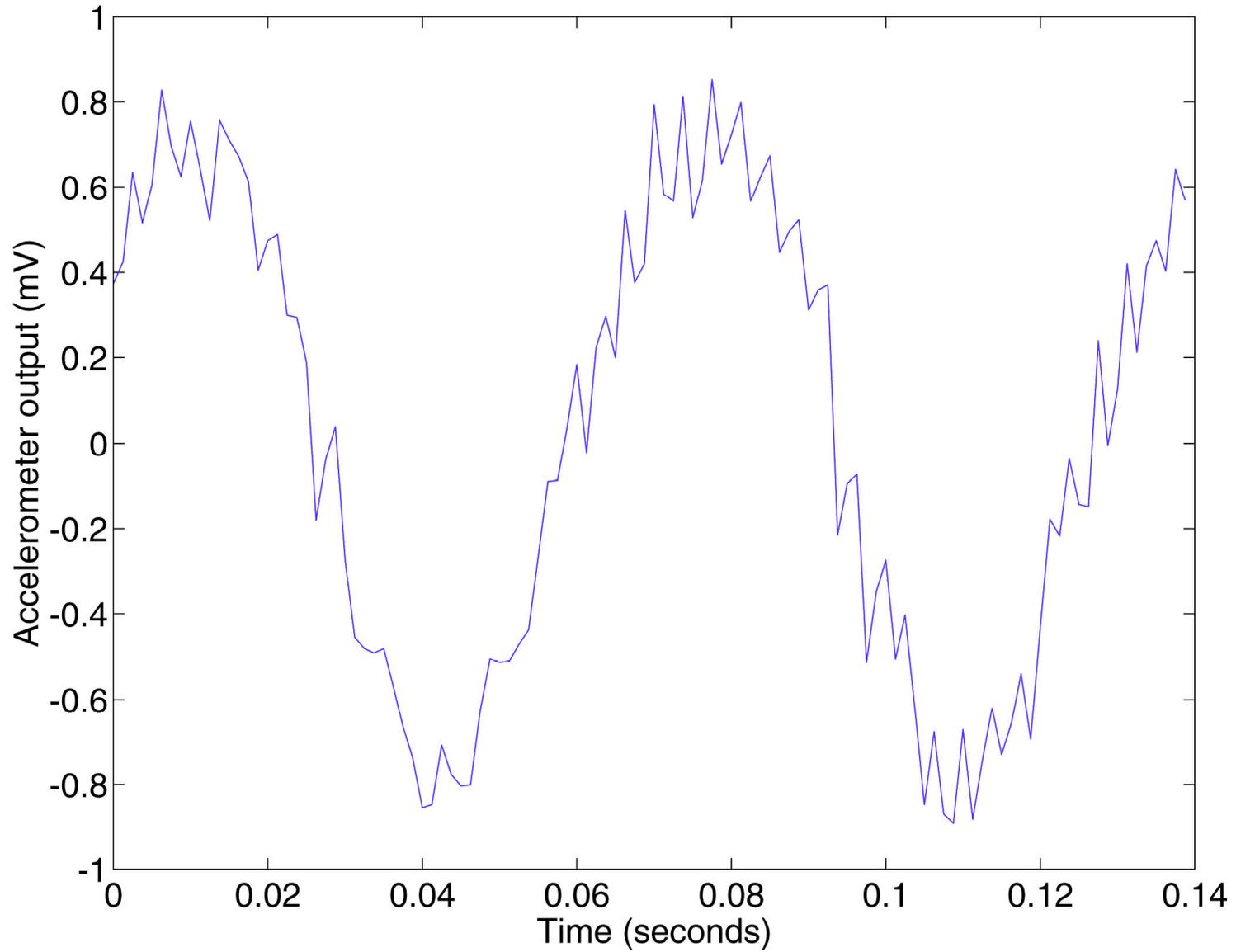
# Non-parametric model data



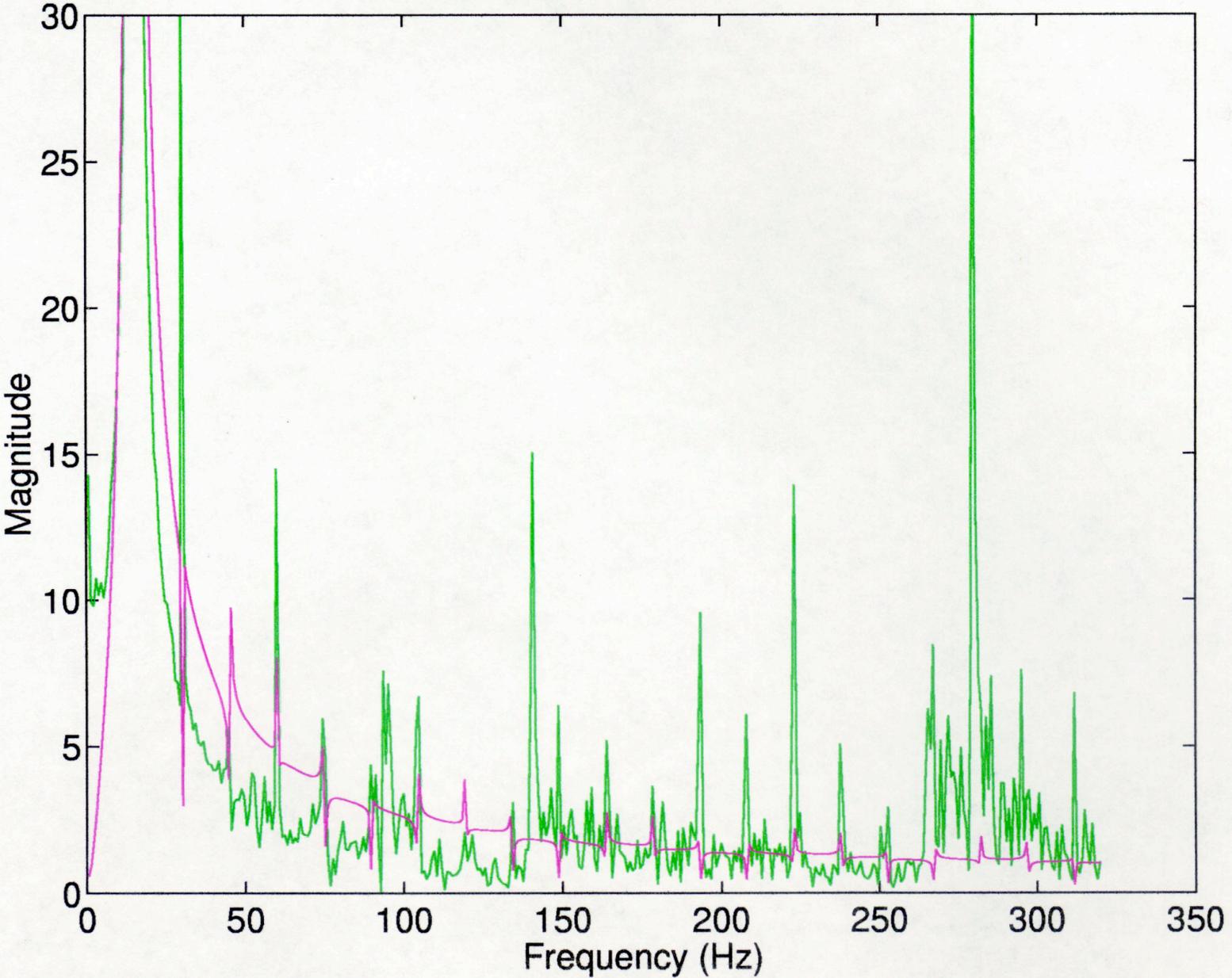
Input 15Hz excitation



Measured output with 15Hz excitation



DFT magnitudes of 15Hz excited output



## Cleaning the data - filtering

$$\hat{P}(e^{j\omega_i}) = \frac{\sum_{k=1}^{1000} y_{i,k} \exp(j\omega_i k/T)}{\sum_{k=1}^{1000} u_{i,k} \exp(j\omega_i k/T)}$$

Sinusoidal experimental data collected at 21 frequencies

Input sinusoid contains some small harmonic distortion

Accelerometer signal contains very significant harmonic distortion

Pass both through a very narrow-band-pass filter

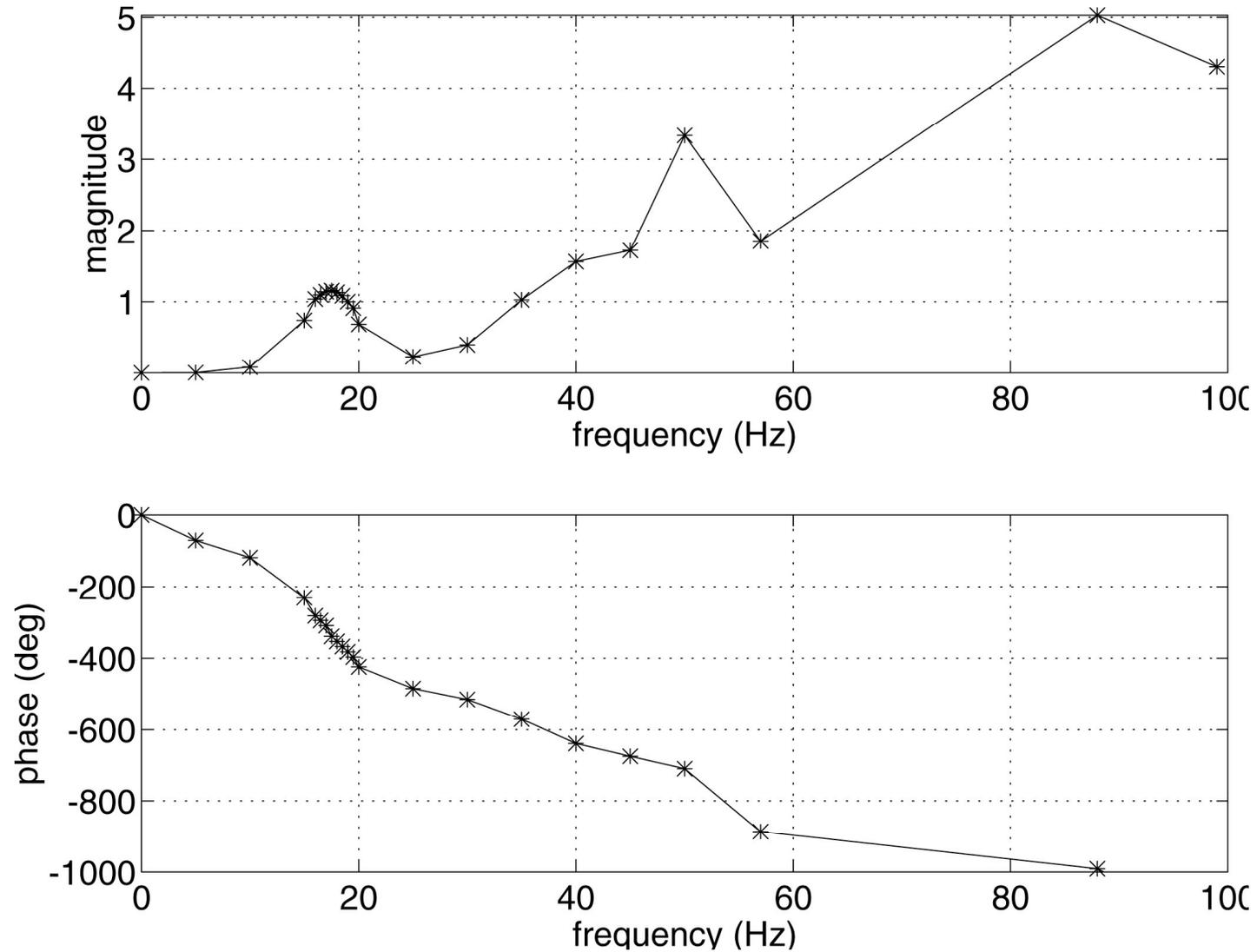
This is achieved with the above computations

This is akin to the discrete Fourier transform

Divide the the results to get the frequency response values

21 complex frequency response estimates

# Estimated frequency response



# Parametric model development

Start with 21 frequency response values centered around 17Hz

Fit a stable low-order model to these values

Direct frequency-domain fit

Difficult to guarantee stability

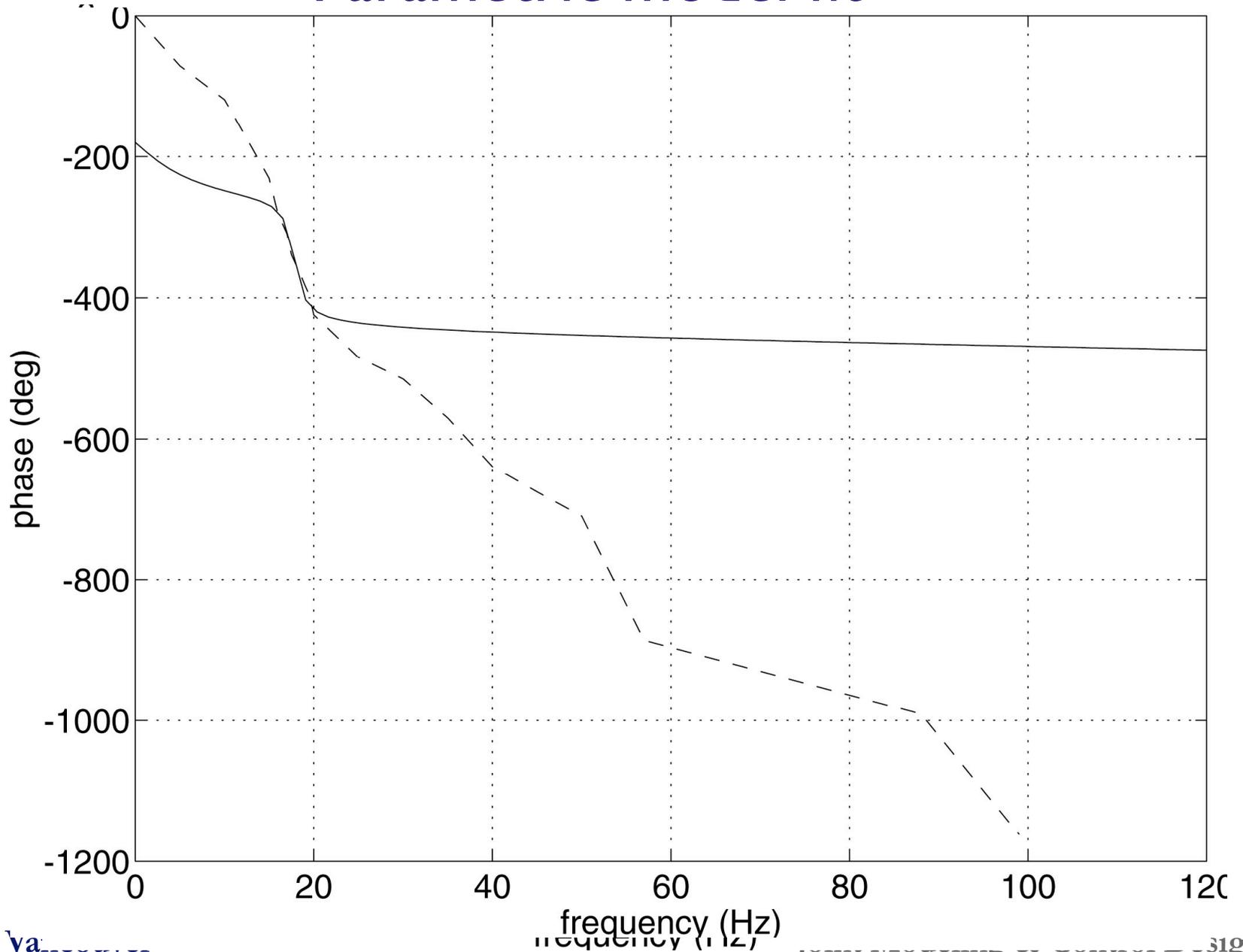
We actually want a weighted fit to emphasize 17Hz

Use the estimated frequency response values to cook up some fake data with the correct frequency distribution.

Fit the model using this data record and Output Error model structure

Guaranteed stable

# Parametric model fit

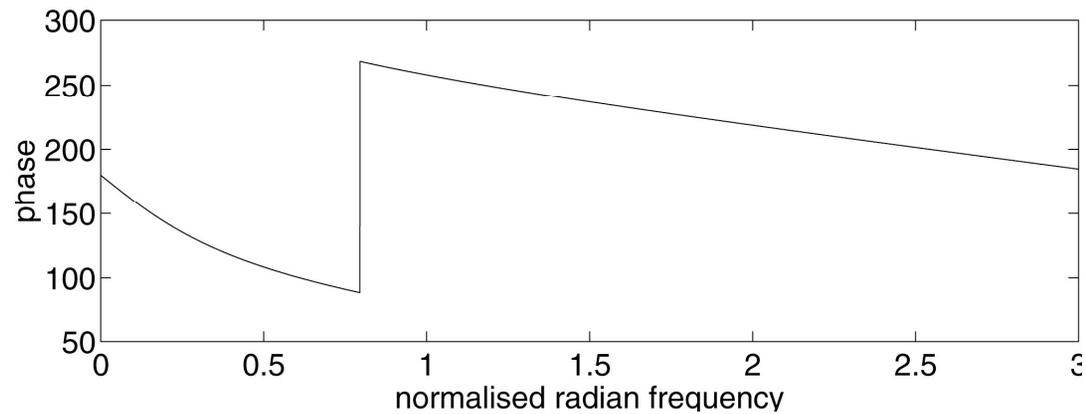
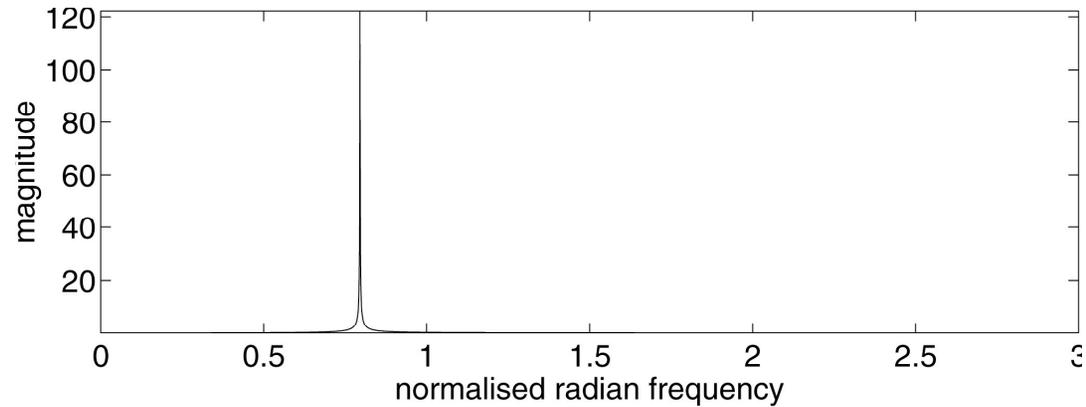


APC04 va

frequency (Hz)

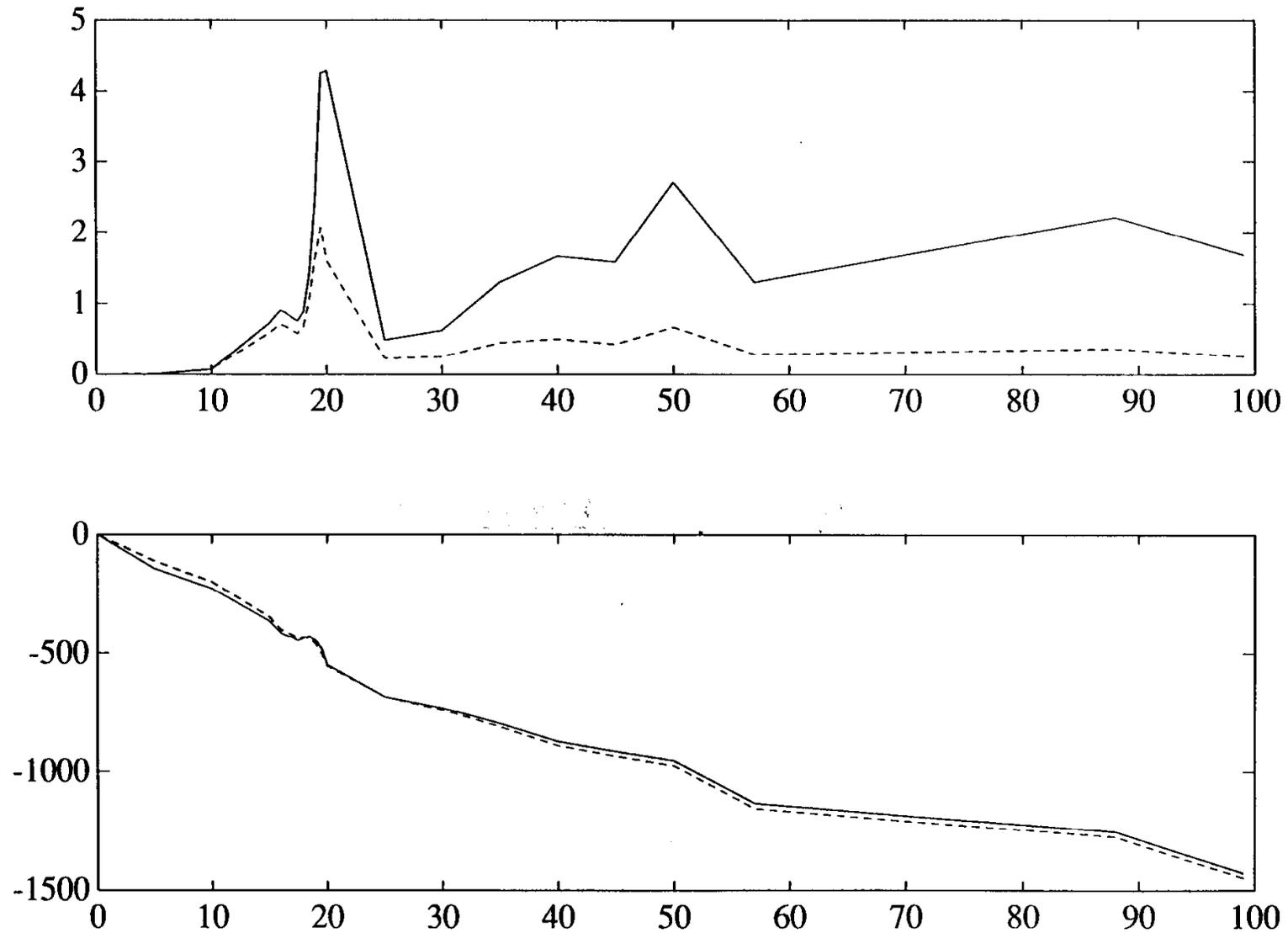
sign

# Model-based control design



Frequency-weighted Linear Quadratic Gaussian Control for disturbance rejection

# Controller robustness check



# Conclusion

The problem specification and our prejudice about the solution have colored the whole modeling and control design

Modeling has reflected the robust control requirements

Experiment design

Data concentrated where phase accuracy needed

Sinusoidal excitation avoided rate limiters

Data preparation before modeling

Harmonic distortion needed to be removed via filtering

Parametric model has low order and is stable

Heavy frequency weighting in model fit stage

Control design

Frequency weighted against model mismatch at HF

Don't care at LF