



Norwegian University of  
Science and Technology



# Identification of turbine dynamics using PMUs

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# Outline



- Background
- Previous work
- Theoretical validation
- Results
- Conclusions and further work

# Background Power Systems

- Large interconnected system



**Figure:** Nordic power system[ENTSO-e]

# Background

## Power Systems

Figure: Nordic power system[ENTSO-e]

- Large interconnected system
- Balancing challenge

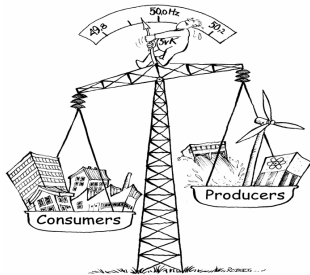


Figure: Balancing challenge[Statnett]

# Background

The power system is dynamic

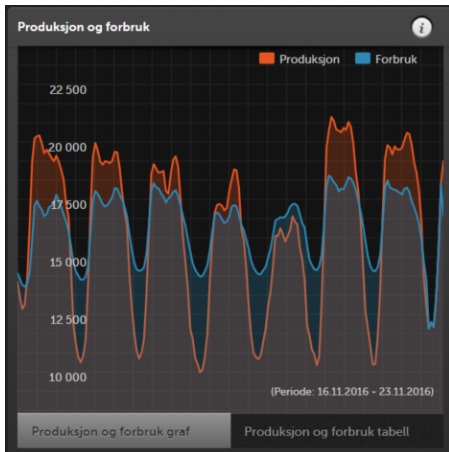
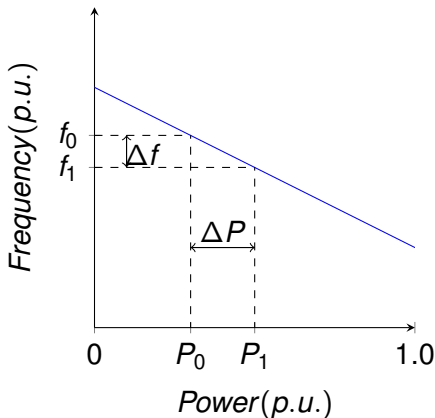


Figure: Production and consumption [statnett.no]

# Background

## Frequency containment reserves (FCR)

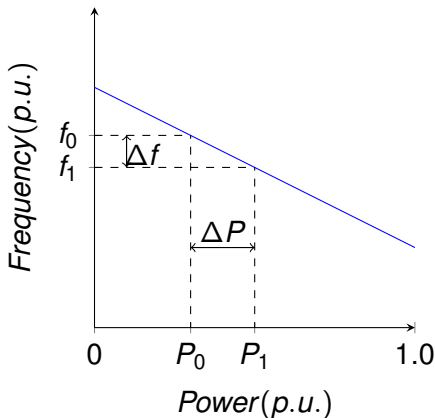
- Power balance/frequency containment control (FCC) is mainly determined by governor response.
- Activation of primary reserves is determined by the governor droop settings.



# Background

## Frequency containment reserves (FCR)

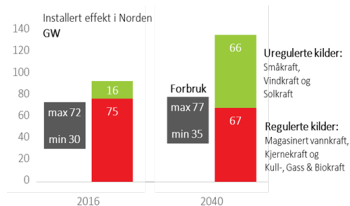
- Power balance/frequency containment control (FCC) is mainly determined by governor response.
- Activation of primary reserves is determined by the governor droop settings.
- In steady state



# Background

## Challenges in operation

- Towards 100% renewable electricity generation
  - Larger variability
  - More uncertainty
  - Increasing complexity



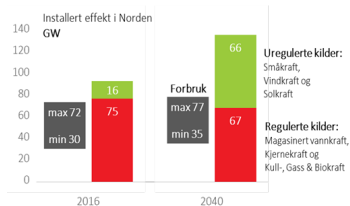
**Figure:** Present and future energy mix[Statnett]



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## Challenges in operation

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**Figure:** Present and future energy mix[Statnett]

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- Less time for actions

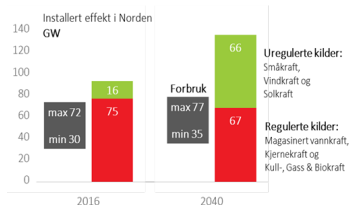


Figure: Present and future energy mix[Statnett]

# Background

## Challenges in operation

- Towards 100% renewable electricity generation
  - Larger variability
  - More uncertainty
  - Increasing complexity
- More dynamics
- Less time for actions
- **Hydropower** is the main resource for balancing

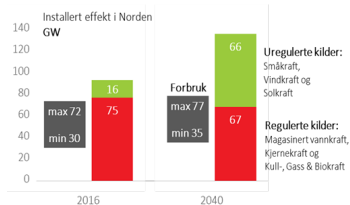
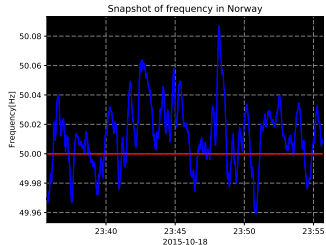


Figure: Present and future energy mix[Statnett]

# Background

## Frequency quality in the Nordics

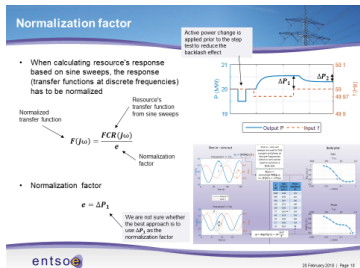
- From 2008 the time the frequency has been outside its allowed band has increased
- The performance of hydro turbine governors play an important role



# Background

## New requirements on FCR

- Nordic TSOs are developing new requirements on FCR
- This includes offline testing and verification of performance



# Background

## Research question



1. Can we do the tests only using PMUs?

# Background

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2. What is the best way to do the tests?

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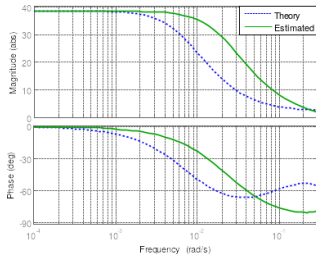


1. Can we do the tests only using PMUs?
2. What is the best way to do the tests?
3. Is there anything to gain from combining traditional tests at the plant with PMU measurements?



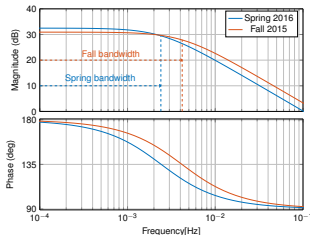
# Previous work

- Governor dynamics were identified using the ARX model structure



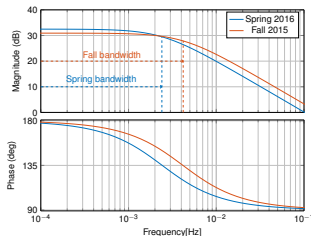
# Previous work

- Governor dynamics were identified using the ARX model structure
- Governor dynamics were identified using time domain vector fitting



# Previous work

- Governor dynamics were identified using the ARX model structure
- Governor dynamics were identified using time domain vector fitting
- However, no theoretical validation was made.



## Previous work

Work presented in this here



- Development of test system for governor and turbine identification.

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- Theoretical validation.

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Work presented in this here



- Development of test system for governor and turbine identification.
- Theoretical validation.
- Investigation of assumptions made in the validation.

# Theoretical validation

## System identification basic

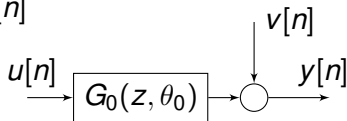


- Assume that a data set  $Z^N = \{u[n], y[n] | n = 1 \dots N\}$  has been collected.
- The dataset  $Z^N$  is assumed generated by

$$\mathcal{S} : y[n] = G_0(z, \theta_0)u[n] + H_0(z, \theta_0)e[n] \quad (1)$$

- Using the data set  $Z^N$  we want to find the parameter vector  $\theta^N$  minimizing

$$\hat{\theta}_N = \arg \min_{\theta} \frac{1}{N} \sum_{n=1}^N \epsilon^2(n, \theta) \quad (2)$$





# Theoretical validation

## Consistency



- A consistent estimate means that the true parameter vector  $\theta_0$  is the unique solution to the asymptotic prediction error criterion.

$$\theta^* = \arg \min_{\theta} \bar{E}\epsilon^2(n, \theta) \quad (3)$$

with

$$\bar{E}\epsilon^2(n, \theta) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N E\epsilon^2(n, \theta) \quad (4)$$

and

$$\epsilon(n, \theta) = H_1^{-1}(z, \theta)(y[n] - G_1(z, \theta)u[n]) \quad (5)$$

# Theoretical validation

## System identification basics take away



- Define what one wants to identify.

# Theoretical validation

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- Define the input and outputs of the system.

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- Prove that one will obtain a consistent estimate using the selected inputs and outputs.

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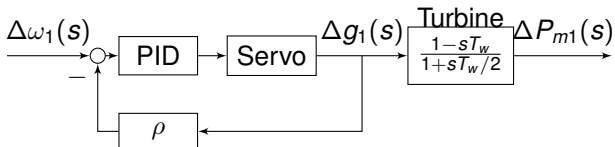
## System identification basics take away



- Define what one wants to identify.
- Define the input and outputs of the system.
- Prove that one will obtain a consistent estimate using the selected inputs and outputs.
- *The input and output have to be modeled to do this*

# Definition of identification problem

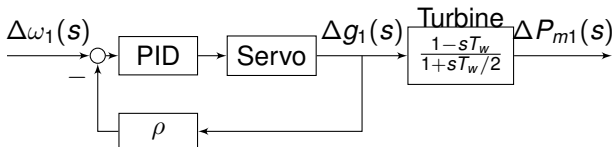
Choice of input and output to the identification problem



- Preferably we would use:

# Definition of identification problem

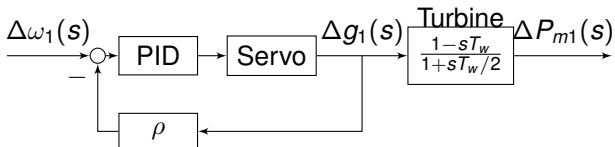
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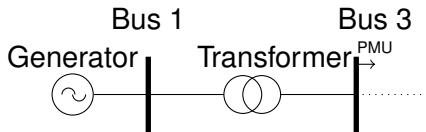


- Preferably we would use:
  - $\Delta\omega_1[n]$  as input and,
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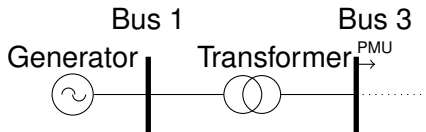
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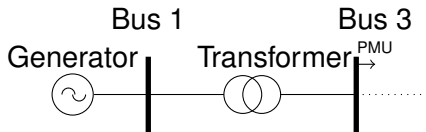
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# Definition of identification problem

## Choice of input and output to the identification problem



- Preferably we would use:
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  - $\Delta P_{m1}[n]$  as output.
- However, the TSO has only access to
  - $\Delta\omega_3[n]$  and,
  - $\Delta P_{e3}[n]$ .

## Definition of identification problem

### Assumptions regarding input and output



- We assume that the PMU is situated sufficiently close to the generator such that:
  - $\Delta\omega_1[n] \approx \Delta\omega_3[n]$  and,
  - $\Delta P_{e1}[n] \approx \Delta P_{e3}[n]$ .
- The electrical power is related to the mechanical power by the swing equation:

$$\Delta\omega_1(s) = \frac{\Delta P_{m1}(s) - \Delta P_{e1}(s)}{2\mathcal{H}_1 s + K_{d1}} \quad (6)$$

## Definition of identification problem

Transfer function that can be identified using PMUs



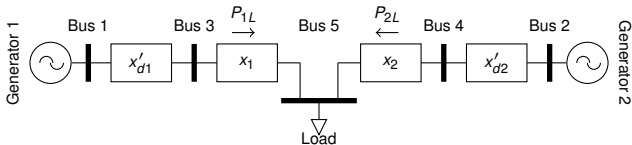
- We now introduce the following transfer functions:
  - The turbine and governor dynamics are described by  $G_{t1}(s)$  and,
  - $G_{J1}(s) = 1/(2\mathcal{H}_1 s + K_{d1})$
- We can now write the angular speed as:

$$\Delta\omega_1(s) = -\frac{G_{J1}(s)}{1 + G_{J1}(s)G_{t1}(s)}\Delta P_{e1}(s) + v_1(s) \quad (7)$$

- The transfer function  $G_1(s)$  we can identify is therefore:

$$G_1(s) = -\frac{G_{J1}(s)}{1 + G_{t1}(s)G_{J1}(s)} \quad (8)$$

# Introduction of test system



- *We need to model relation between  $P_{e1}[n]$  and  $\Delta\omega_1[n]$ .*
- We therefore introduce a small test system consisting of:
  - The plant we want to identify.
  - an aggregated plant,
  - an aggregated load
  - the subsynchronous reactances and,
  - the line reactances.

# Test system for identification

## Model of the load



- We assume the following model for the load

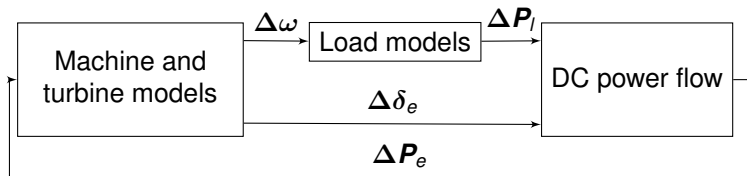
$$\Delta P_{load} = \Delta P_f + \Delta P_s \quad (9)$$

where:

- $\Delta P_f$ : is frequency dependent part of the load
- $\Delta P_s$ : is the stochastic part of the load assumed to be filtered white noise.

# Test system for identification

## Connecting the elements together



- To connect the elements together we will use the dc power flow.
  - It is simple.
  - Strong coupling between active power and frequency.



# Test system for identification

## DC power flow

- We start by organizing the DC power flow in terms of loads and generators

$$\begin{bmatrix} \Delta \mathbf{P}_e \\ \Delta \mathbf{P}_l \end{bmatrix} = \begin{bmatrix} \mathbf{B}_{11} & \mathbf{B}_{12} \\ \mathbf{B}_{21} & \mathbf{B}_{22} \end{bmatrix} \begin{bmatrix} \Delta \delta_e \\ \Delta \delta_l \end{bmatrix} \quad (10)$$

- The angle of the non generator buses can now be calculated as:

$$\Delta \delta_l = \mathbf{B}_{22}^{-1} (\Delta \mathbf{P}_l - \mathbf{B}_{21} \Delta \delta_e) \quad (11)$$

- The power injections at the generator buses are:

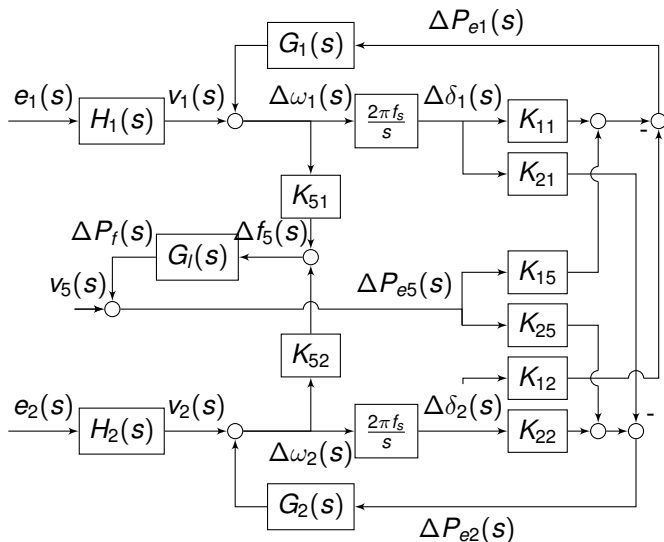
$$\Delta \mathbf{P}_e = \mathbf{B}_{11} \Delta \delta_e + \mathbf{B}_{12} \Delta \delta_l \quad (12)$$

- Finally, we substitute (16) into (12) and rearrange to obtain.

$$\Delta \mathbf{P}_e = \begin{bmatrix} \mathbf{B}_{11} - \mathbf{B}_{12} \mathbf{B}_{22}^{-1} \mathbf{B}_{21} & \mathbf{B}_{12} \mathbf{B}_{22}^{-1} \end{bmatrix} \begin{bmatrix} \Delta \delta_e \\ \Delta \mathbf{P}_l \end{bmatrix} \quad (13)$$

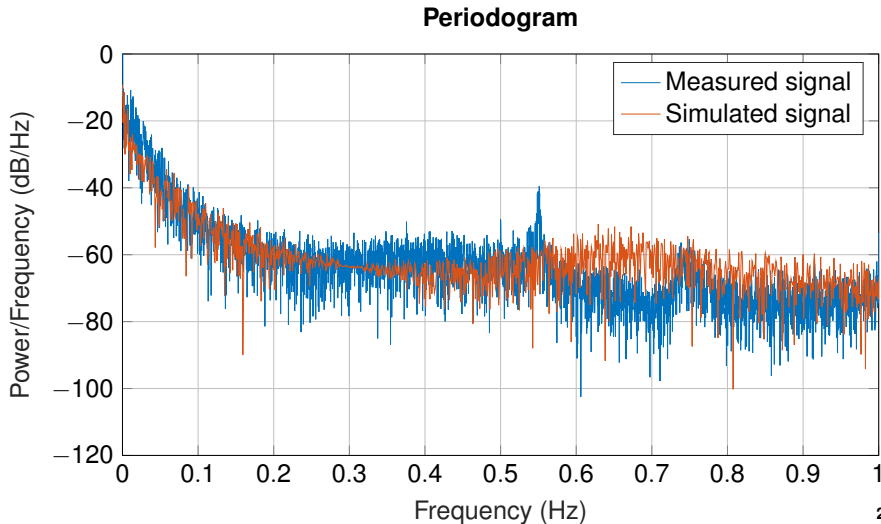
# Test system for identification

## Putting it all together



# Test system for identification

## System frequency response



# Results

## Results from the theoretical validation



- A consistent estimate of the closed loop transfer function of the turbine and electromechanical dynamics can be obtained by using:

# Results

## Results from the theoretical validation



- A consistent estimate of the closed loop transfer function of the turbine and electromechanical dynamics can be obtained by using:
  - Measured PMU frequency as the output  $u[n]$

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- A consistent estimate of the closed loop transfer function of the turbine and electromechanical dynamics can be obtained by using:
  - Measured PMU frequency as the output  $u[n]$
  - Measured PMU power as the input  $y[n]$

# Results

## Results from the theoretical validation



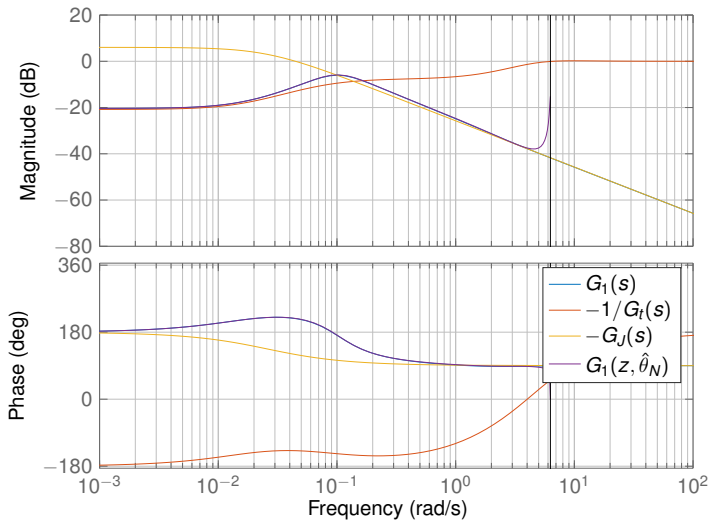
- A consistent estimate of the closed loop transfer function of the turbine and electromechanical dynamics can be obtained by using:
  - Measured PMU frequency as the output  $u[n]$
  - Measured PMU power as the input  $y[n]$
- The proof was done with the following assumptions.
  - The system is excited by a load acting as a white noise process
  - The measurement error of the electrical power is negligible.
  - The measured frequency is a good estimate of the generator speed.

# Results

## Results from simulations



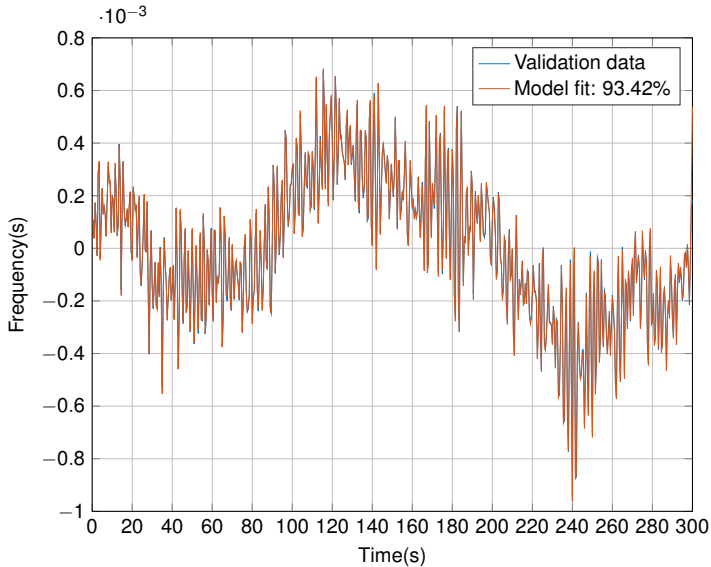
Bode Diagram





# Results

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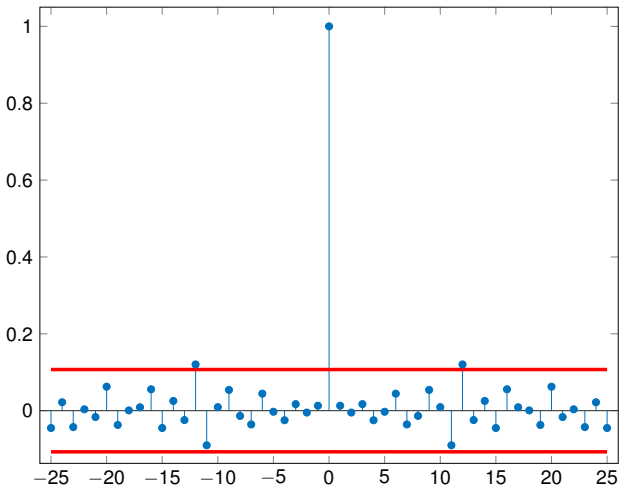


# Results

## Results from simulations



Sample Autocorrelation with 99% Confidence Intervals



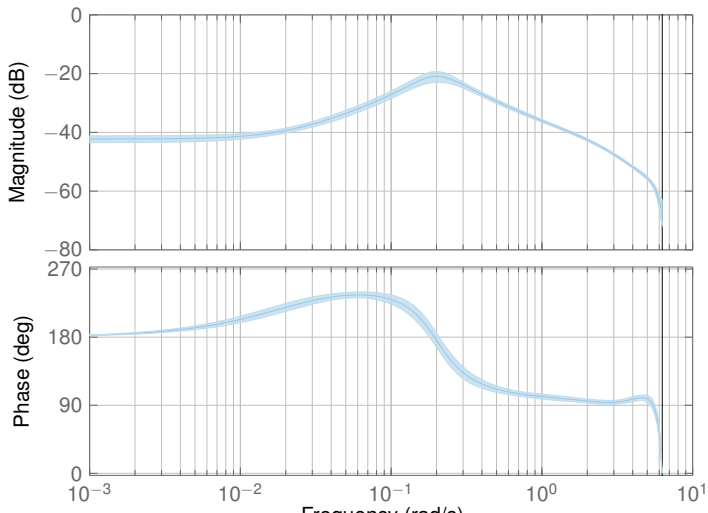
# Results

## Results from the power system



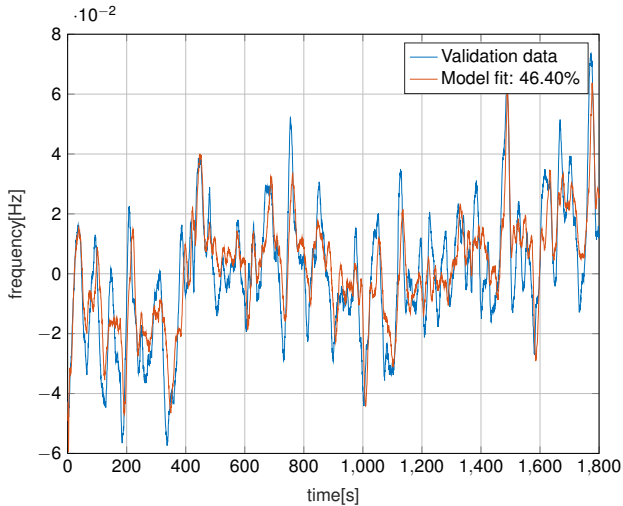
### Bode Diagram

From: u1 To: y1



# Results

## Results from the power system

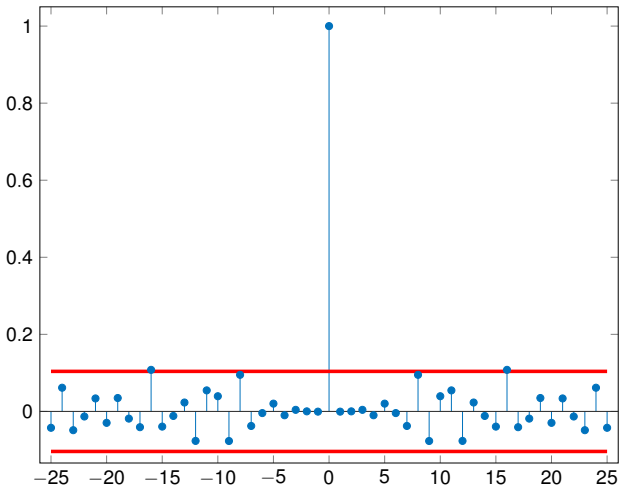


# Results

## Results from the power system



Sample Autocorrelation with 99% Confidence Intervals



## Conclusion on the validation



- A consistent estimate can indeed be obtained.
- The approach gives information on the frequency bias
- The assumption regarding  $\Delta\omega_1 \approx \Delta\omega_3$  should be investigated.

## Assumption that $\Delta\omega_1[n] \approx \Delta\omega_3[n]$



- In reality  $\Delta\omega_1[n] \approx \Delta\omega_3[n]$  is only valid for a certain frequency range.
- We need an expression for the angular velocity at bus 3.
- The two standard options would be:
  - The time derivative of the voltage angle at the bus.
  - The centre of inertia equation.
- We will instead use the frequency divider(FD) formula (Milano 2017).

## Assumption that $\Delta\omega_1[n] \approx \Delta\omega_3[n]$

### Derivation of the FD formula

- Start by the DC power flow assumption assuming the load changes to be negligible.

$$\begin{bmatrix} \Delta \mathbf{P}_e \\ \mathbf{0} \end{bmatrix} = \begin{bmatrix} \mathbf{B}_{11} & \mathbf{B}_{12} \\ \mathbf{B}_{21} & \mathbf{B}_{22} \end{bmatrix} \begin{bmatrix} \Delta \delta_e \\ \Delta \delta_l \end{bmatrix} \quad (14)$$

- Then we rearrange

$$\Delta \delta_l = -\mathbf{B}_{22}^{-1} \mathbf{B}_{21} \Delta \delta_e \quad (15)$$

- We now take the time derivative to obtain.

$$\Delta \omega_l = -\mathbf{B}_{22}^{-1} \mathbf{B}_{21} \Delta \omega_e \quad (16)$$



# Assumption that $\Delta\omega_1[n] \approx \Delta\omega_3[n]$

## FD example

$$\mathbf{B} = \left[ \begin{array}{cc|cc} b'_{d1} & 0 & -b'_{d1} & 0 \\ 0 & b'_{d2} & 0 & -b'_{d2} \\ \hline -b'_{d1} & 0 & b'_{d1} + b_1 & 0 \\ 0 & -b'_{d2} & 0 & b'_{d2} + b_2 \\ 0 & 0 & -b_1 & -b_2 \\ & & & b_1 + b_2 \end{array} \right] \quad (17)$$

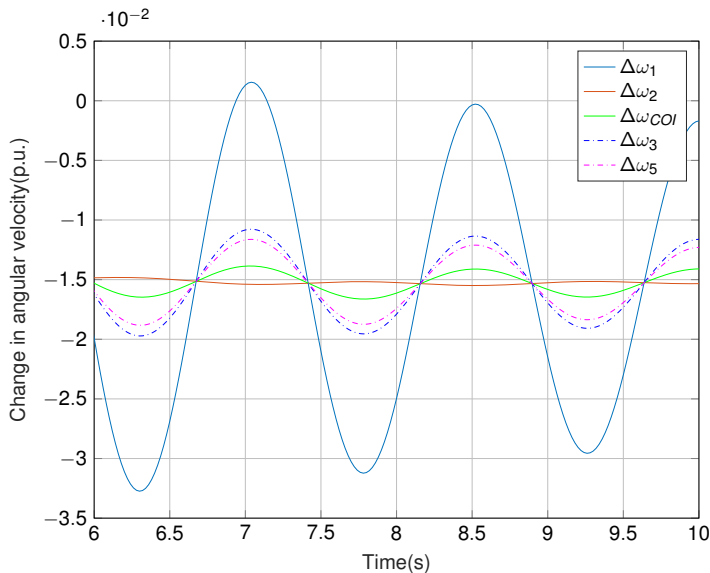
$$\Delta\omega_3 = \frac{(b_1 b_{d2} + b_1 b_2 + b_2 b_{d2}) b_{d1}}{|\mathbf{B}_{22}|} \Delta\omega_1 + \frac{b_1 b_2 b_{d2}}{|\mathbf{B}_{22}|} \Delta\omega_2 \quad (18)$$

$$\Delta\omega_5 = \frac{b_1 b'_{d1} (b'_{d2} + b_2)}{|\mathbf{B}_{22}|} \Delta\omega_1 + \frac{b_2 b'_{d2} (b'_{d1} + b_1)}{|\mathbf{B}_{22}|} \Delta\omega_2 \quad (19)$$

$$\Delta\omega_{COI} = \frac{1}{\mathcal{H}_1 + \mathcal{H}_2} (\mathcal{H}_1 \Delta\omega_1 + \mathcal{H}_2 \Delta\omega_2) \quad (20)$$

# Assumption that $\Delta\omega_1[n] \approx \Delta\omega_3[n]$

## FD example



**Assumption that  $\Delta\omega_1[n] \approx \Delta\omega_3[n]$**

**Assumptions further work**



- Continue with the dc power flow assumption,

**Assumption that  $\Delta\omega_1[n] \approx \Delta\omega_3[n]$**

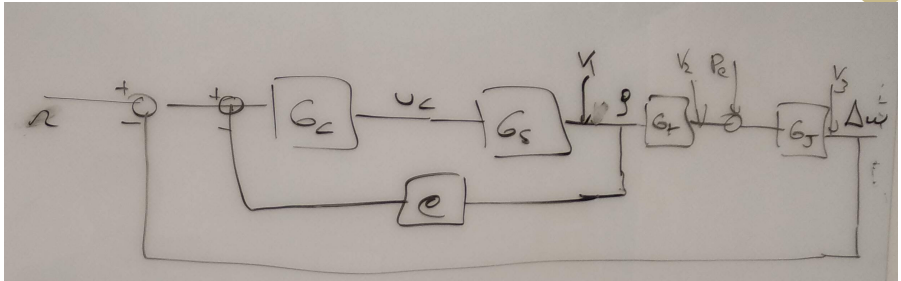
**Assumptions further work**



- Continue with the dc power flow assumption,
- however, remove the FD assumption of  $\Delta P_l = 0$ .

# Future and ongoing work

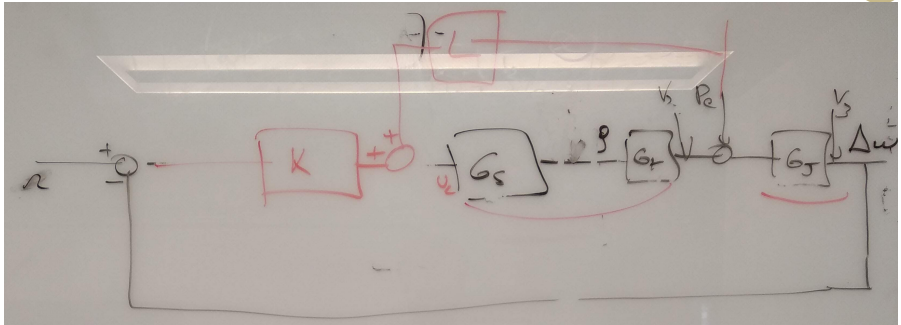
## Identification of turbine dynamics at the plant



- Which transfer functions can we consistently identify without adding external excitation?
- What do we gain in terms of having access to a PMU?
- How do we best design an external excitation signal?

# Future and ongoing work

## Using the identified models for control



- Use the identified models to find optimal PID parameters.
- Use PMU and identified models to design a feedforward controller.