

Norwegian University of Science and Technology



Identification of turbine dynamics using PMUs

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Outline



Background

Previous work

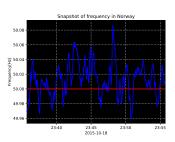
Theoretical validation

Results

Conclusions and further work

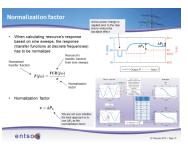
Frequency quality in the Nordics

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



New requirements on FCR due to frequency quality

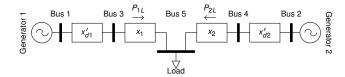
- Nordic TSOs are developing new requirements on FCR
- This includes offline testing and verification of performance



Idea on monitoring the FCR online

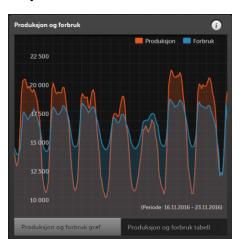


Less intrusive



Idea on monitoring the FCR online

- Less intrusive
- The system is dynamic





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



Figure: Present and future energy mix[Statnett]

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

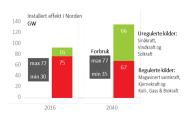


Figure: Present and future energy mix[Statnett]

- Towards 100% renewable electricity generation
 - Larger variability
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 - Increasing complexity
- More dynamics
- Less time for actions



Figure: Present and future energy mix[Statnett]

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

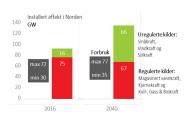


Figure: Present and future energy mix[Statnett]



1. Do the transmission system operator (TSO) know whether or not the hydropower plants deliver the FCR they are supposed to?



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- 2. Can the TSO measure it online?

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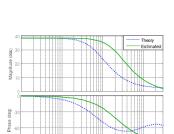
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Previous articles

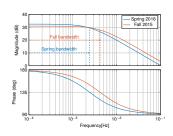
 Governor dynamics were identified using the ARX model structure



Frequency (rad/s)

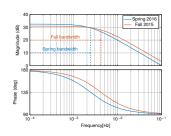
Previous articles

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

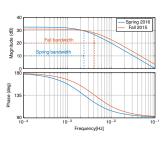


Previous articles

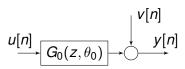
- Governor dynamics were identified using the ARX model structure
- Governor dynamics were identified using time domain vector fitting
- There are also other papers in the literature using other methods for online identification, however, mostly relying on data from disturbance recordings.



 Why do we get different results?



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a+b

- Why do we get different results?
- The signals we use are corrupted by noise.
- Using system identification techniques we can estimate the variance of the covariance matrix of the parameter vector
- However, first we have to prove that we will get consistent results.

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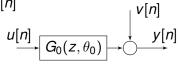
System identification basic

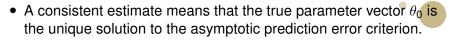
- Assume that a data set
 Z^N = {u[n], y[n]|n = 1...N}
 has been collected.
- The dataset Z^N is assumed generated by

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

 Using the data set Z^N we want to find the parameter vector θ^N minimizing

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





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

with

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

and

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

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 - Measured PMU frequency as the output u[n]



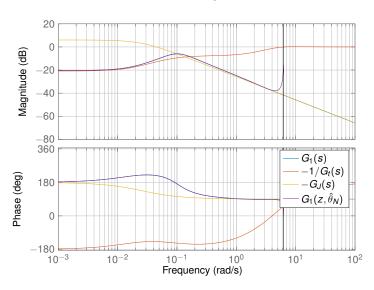
- A consistent estimate of the closed loop transfer function of the turbine and electromechanical dynamics can be obtained by using:
 - Measured PMU frequency as the output u[n]
 - Measured PMU power as the input y[n]



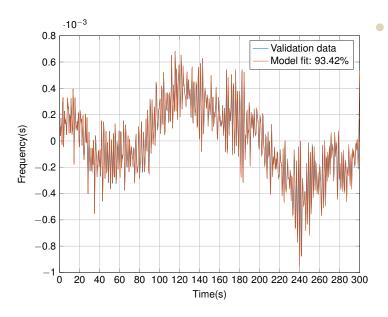
- A consistent estimate of the closed loop transfer function of the turbine and electromechanical dynamics can be obtained by using:
 - Measured PMU frequency as the output u[n]
 - Measured PMU power as the input y[n]
- The proof was done with the following assumptions.
 - The system is excited by a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

Results from simulations



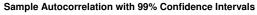


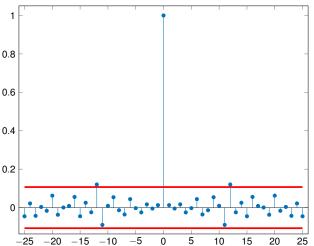
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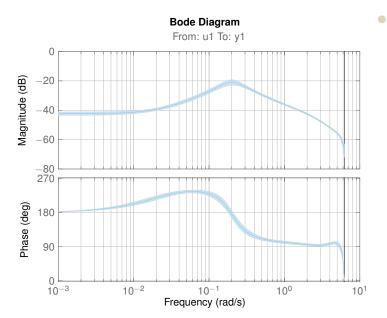
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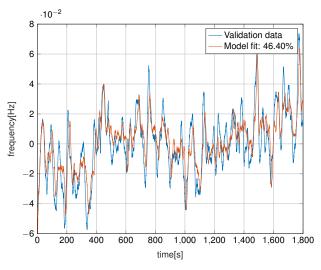


Results from the power system



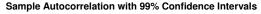
Results from the power system

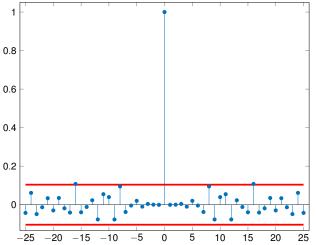




Results from the power system







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- It is indeed possible to identify the turbine dynamics(closed loop with electromechanical dynamics) using PMU measurements.
- Look into the assumptions