



Lidar-Assisted Control for Wind Turbines II

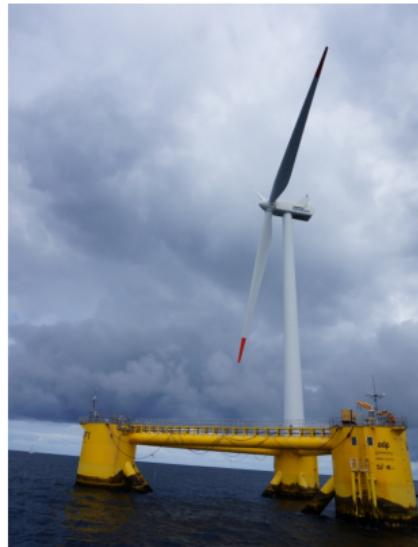
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10.09.2024

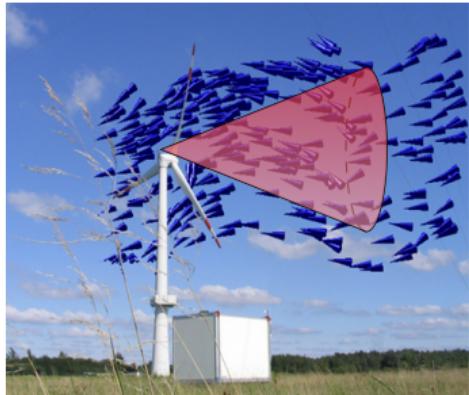
Lecture #11 of Course
"Controller Design for Wind
Turbines and Wind Farms"

Schedule

- 02.09. 1 Controller Design Objectives and Modeling
- 03.09. 2 Baseline Generator Torque Controller
- 04.09. 3 Collective Pitch Controller
- 05.09. 4 Filter Design
- 06.09. 7 Wind Field Generation
- 09.09. 10 Lidar-Assisted Control I
- 10.09. 11 Lidar-Assisted Control II**
- 11.09. 5 Tower Damper
- 12.09. 6 Advanced Torque Controller
- 12.09. 8 Steady State Calculations
- 16.09. 9 Individual Pitch Control
- 17.09. 12 Wind Farm Effects
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Lidar-Assisted Control for Wind Turbines



Motivation

- ▶ wind is changing over space and time
- ▶ conventional control reacts after impact
- ▶ lidar technology provides wind preview
- ▶ better control performance is expected



Main questions

- ▶ How can useful wind preview signals be extracted from lidar data?
- ▶ How can these signals be used to improve wind turbine control?

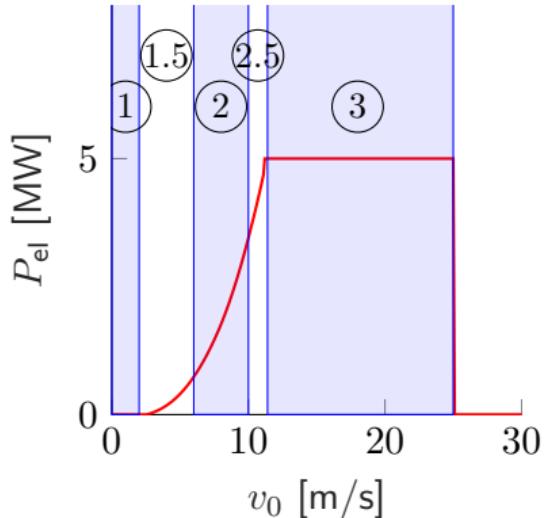
Contents

1. Collective Pitch Feedforward Control

2. Other Ideas and Conclusions

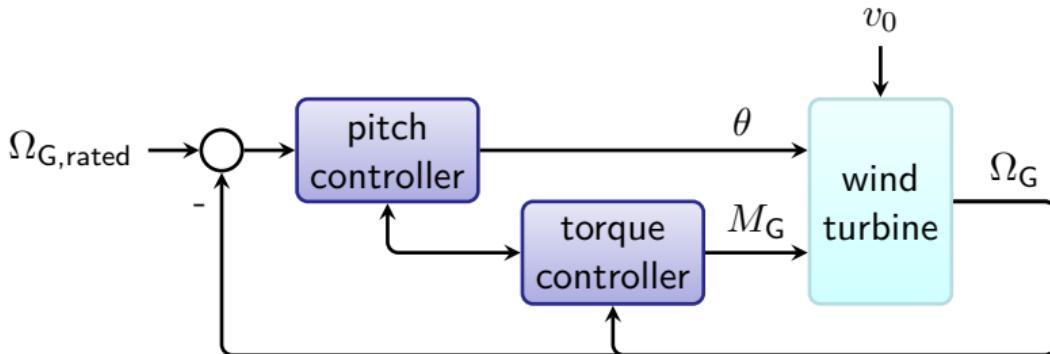


Wind Turbine Control Regions



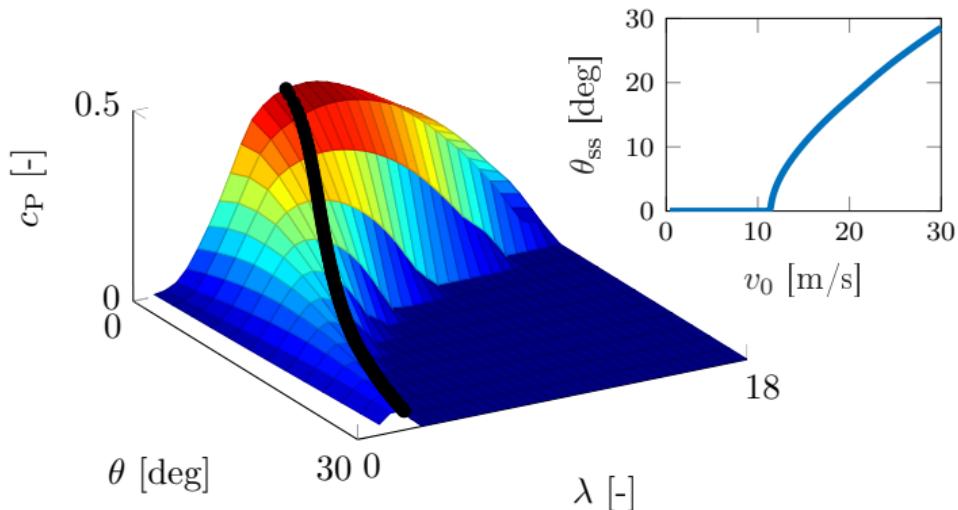
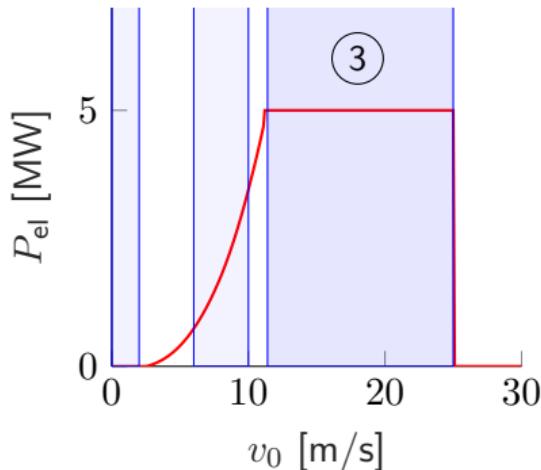
- ▶ Region 1: wind speed too low, no power generation.
 - ▶ Region 2: control goal: maximize energy yield.
 - ▶ Region 3: Maximum/rated power is reached.
Control goal: maintain rated power and generator speed, reduce structural loads.
 - ▶ Transition regions: 1.5 and 2.5.

Baseline Wind Turbine Controller



- ▶ Torque controller optimizes power below and maintains rated power above rated wind speed by adjusting generator torque M_G based on generator speed Ω_G .
 - ▶ Pitch control maintains rated generator speed $\Omega_{G,\text{rated}}$ by adjusting blade pitch angle θ .
 - ▶ Interactions for transition between above and below rated wind very important!

Strategy Baseline Pitch Controller



- ▶ Maximum power and rotor speed is reached → controller in region 3 aims to maintain $P_{\text{el,rated}} = \frac{1}{2}\rho\pi R^2 c_p(\lambda, \theta)v_0^3 \eta_{\text{el}}$ and rated rotor speed.
 - ▶ This results in static pitch angle θ_{ss} over wind speed v_0 to reduce power coefficient c_p .

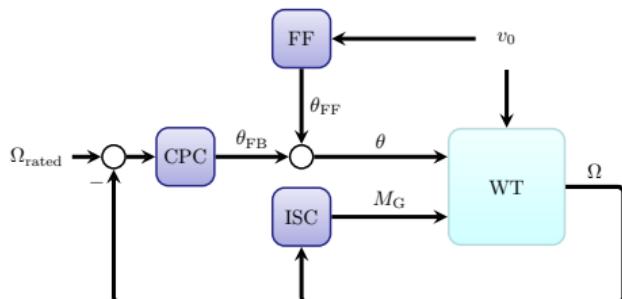
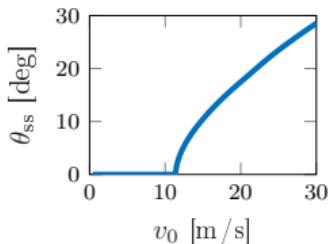
Dinamicamente

 - ▶ Compromise between speed regulation, reduction of structural load, and pitch activity.

Collective Pitch Feedforward Controller Design

$$J\dot{\Omega} = M_a(\theta, \Omega, v_0) - M_G \stackrel{!}{=} 0$$

Objetivo: mantener el pitch en el valor deseado para que el sistema no se acelere



Control in full load operation

- ▶ pitch controller (CPC) maintains rated speed
 - ▶ torque controller (ISC) maintains rated power

Collective pitch feedforward [1]

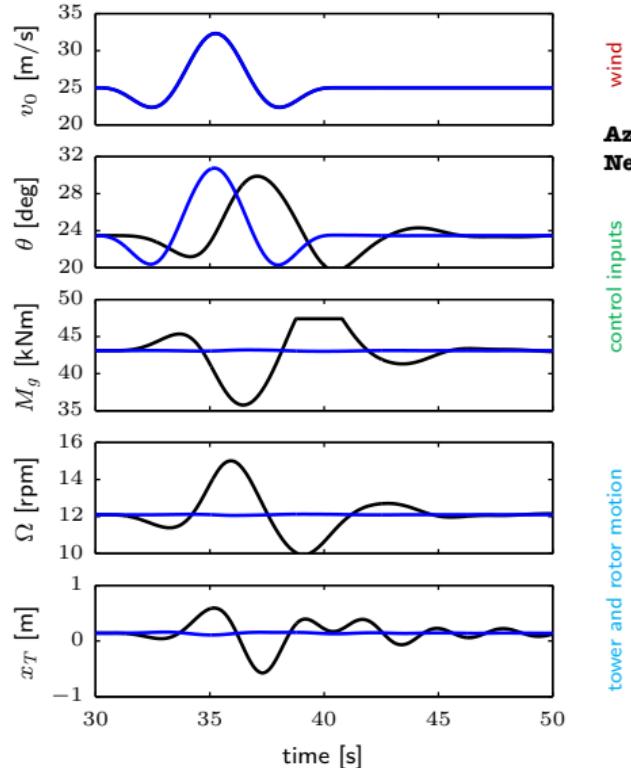
pitch follows static curve θ_{ss} to cancel out effects of changing wind speed v_0 in aerodynamics

Advantages

- ▶ simple update to conventional FB
 - ▶ works over entire full load
 - ▶ little model information
 - ▶ guaranteed stability

Garantiza la estabilidad ya que si el control realimentado es estable el agregar un control anticipativo estable el sistema continua siendo estable.

Comparison with Perfect Wind Preview

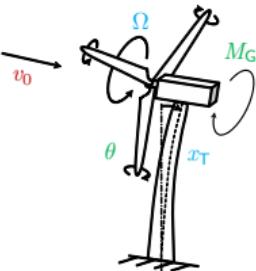


wind

Azul - Control Anticipativo
Negro - Control estandar

control inputs

tower and rotor motion



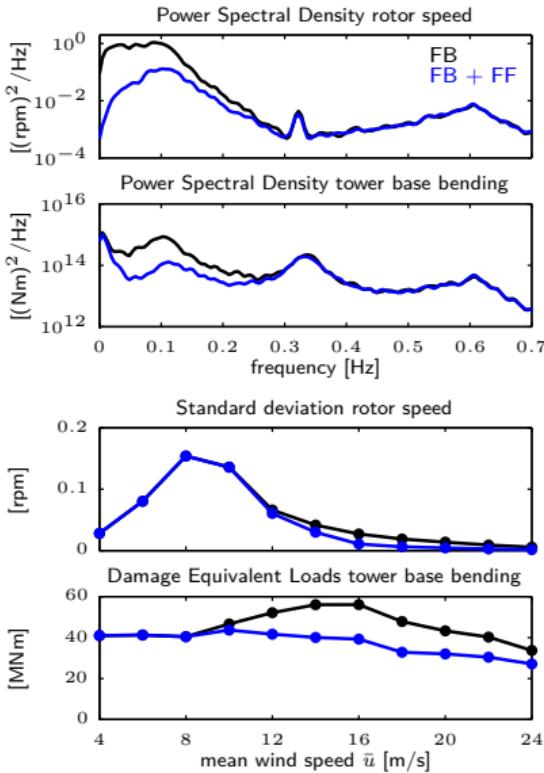
Environment

- ▶ FAST model + coherent wind
 - ▶ EOG at 25 m/s

Results

- ▶ pitch θ follows wind speed v_0
 - ▶ constant generator torque M_G
 - ▶ no variation in rotor speed Ω
 - ▶ also less tower top motion x_T
 - ▶ robust with respect to model uncertainties over large range

Comparison with Realistic Preview



Power Spectral Densities

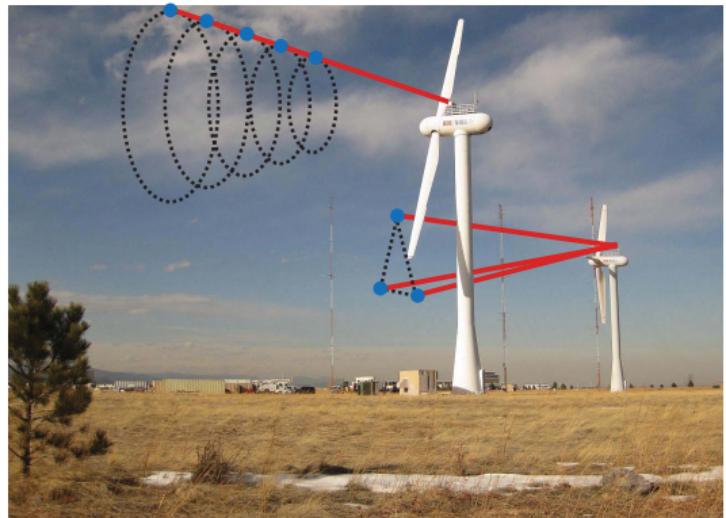
- ▶ FAST model + turbulent wind + lidar simulator
 - ▶ reconstruction + adaptive filter
 - ▶ reduction at low frequencies for rotor speed + tower bending

Life time analysis

- ▶ simulations over full range
 - ▶ beneficial for high wind speeds
 - ▶ DEL on tower reduced by $\approx 20\%$
 - ▶ benefits for blade + shaft loads
 - ▶ only marginal energy increase
 - ▶ robust with respect to measurement uncertainties

Damage equivalent load

Field Testing 2012 - Setup



Control Advanced Research Turbines

- ▶ National Wind Technology Center
 - ▶ 42 m rotor diameter
 - ▶ 600 kW rated power
 - ▶ heavily instrumented

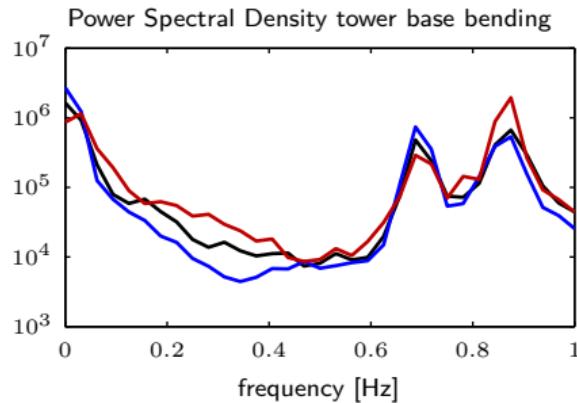
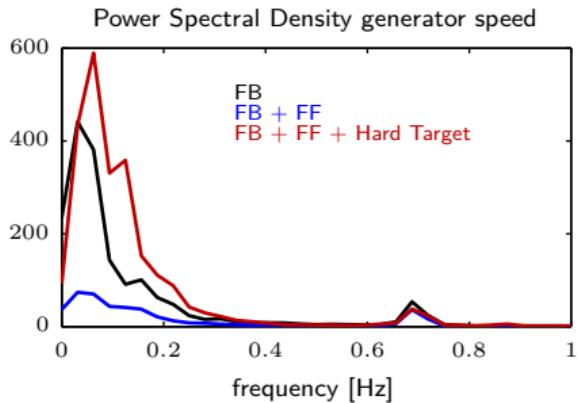
Lidar systems

- ▶ commercial system on CART3
 - ▶ SWE scanning lidar on CART2

Organization

- ▶ cooperation between SWE and NREL
 - ▶ focus on rotor speed regulation

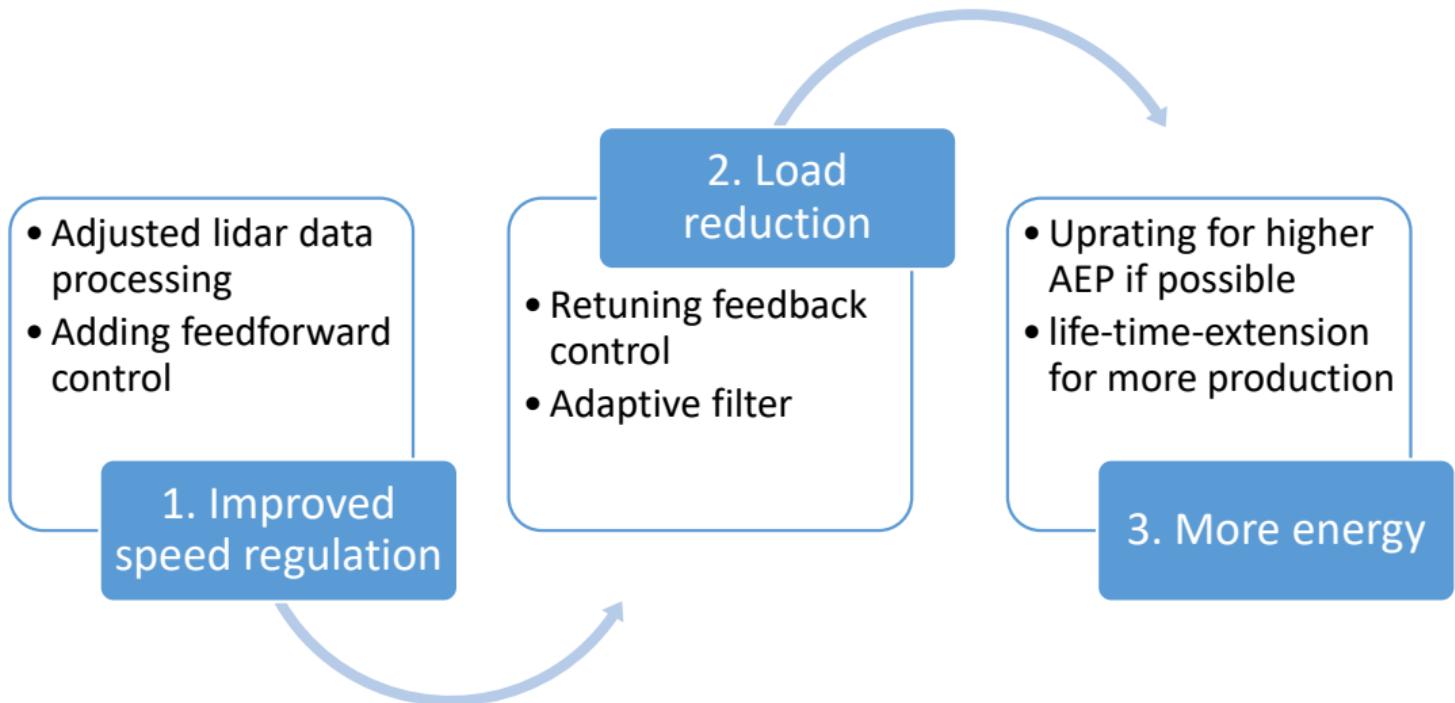
Field Testing 2012 - Results



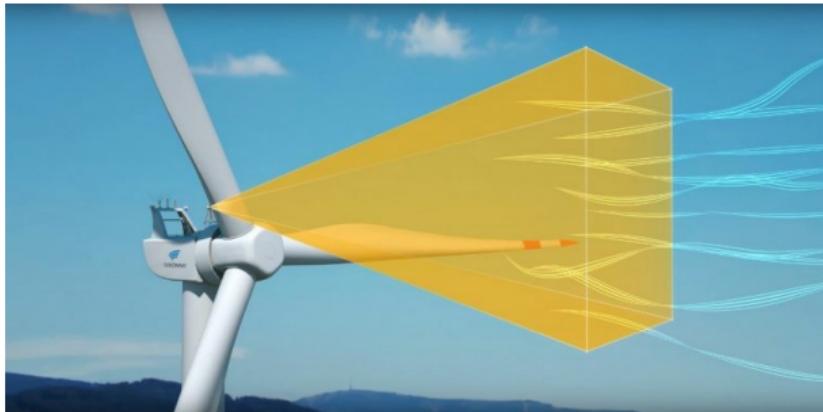
Main results

- ▶ reduction standard deviation of generator speed of 30% at low frequencies
 - ▶ but increase of 30% before solving a measurement problem due to met mast
 - ▶ similar behavior for the tower base bending moment and other loads
 - checking correlation online and more adaptivity necessary

Chain of Benefit for Lidar-Assisted Control



Current Status of Lidar-Assisted Control



[Goldwind]



Status

- ▶ >2000 wind turbines (\approx 6 GW) running with LAC (Goldwind supported by sowento)
 - ▶ Further industrial collaboration with SGRE, Movelaser etc.

Challenges

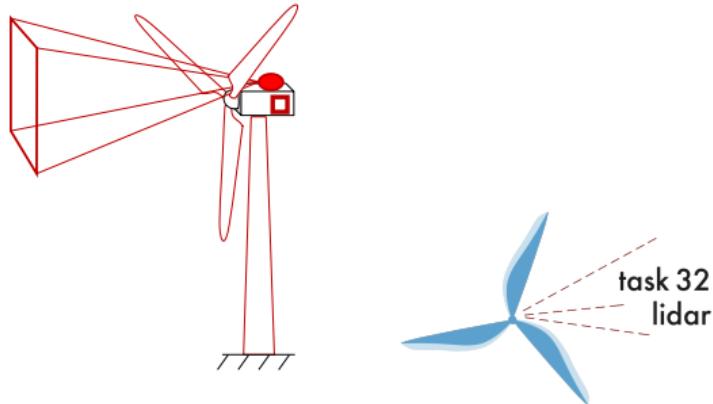
- ▶ complexity of economic benefit
 - ▶ robustness / adaptivity
 - ▶ certification

Systems Engineering



Systems Engineering

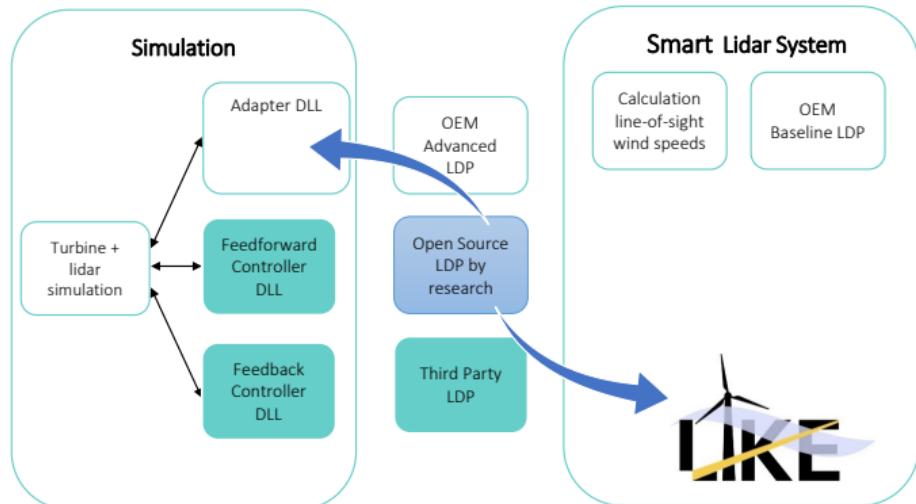
- ▶ used for multidisciplinary optimization
 - ▶ challenge: set up optimization problem
 - ▶ typically applied to aircraft design



Optimization Lidar-Assisted Control

- ▶ Which economic benefit can be obtained by LAC?
 - ▶ IEA Wind Task 32 Workshop [2]

The “Smart Lidar” concept



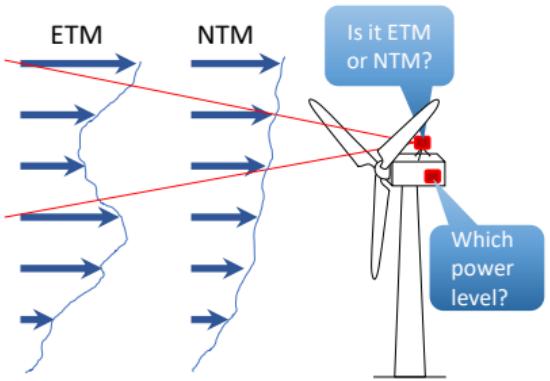
Main Idea

- ▶ enables cooperation
 - ▶ algorithms can be adjusted for specific conditions
 - ▶ algorithms can be tested in aero-elastic simulations
 - ▶ LDP outside safety-relevant turbine controller
 - ▶ great for certification

PhD Project in Flensburg

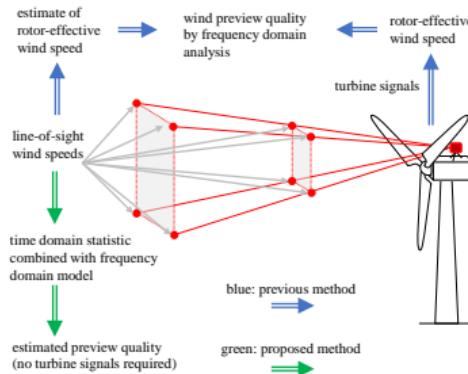
- ▶ Adaptive lidar systems for wind turbine control
 - ▶ Lidar Knowledge Europe

Possible “Apps” for a Smart Lidar



Turbulence Intensity Estimation [3]

- ▶ provides estimate similar to point measurement
 - ▶ uses a spectral model to relate variance of line-of-sight wind speeds to variance of longitudinal wind speed



Wind Preview Quality Estimation [4]

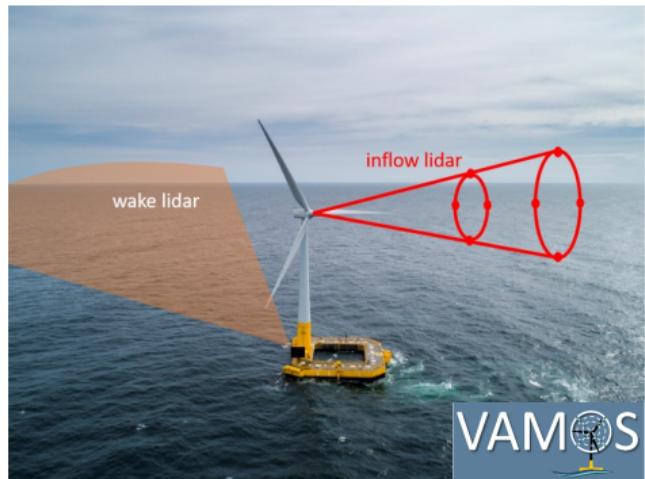
- ▶ provides a correlation coefficient or measurement coherence bandwidth
 - ▶ uses a spectral model to relate covariance between line-of-sight measurements to correlation coefficient

Smart Lidar concept tested offshore!

Validation, Measurement and Optimization of Floating Wind Energy Systems

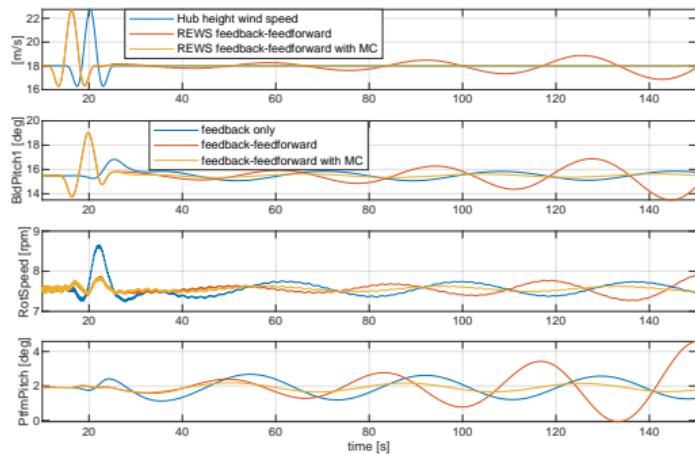
- ▶ German-French research project
 - ▶ measurements on Floatgen demonstrator since summer 2020
 - ▶ simultaneous measurement of inflow and wake on floating wind turbine with lidar
 - ▶ power curve assessment
 - ▶ monitoring with model-based observer
 - ▶ real-time lidar data processing for feedforward control

→ DLL [5] runs on PC (gateway)



[Press release: www.uni-stuttgart.de/en/research/]

A Tutorial on LAC for Floating Offshore Wind Turbines



Open Source Tools of IEA Wind Task 52

- ▶ Detailed Lidar Simulator
 - ▶ Source code of lidar data processing, feedback and feedforward controller
 - ▶ Scripts to run examples

→ Used in LAC Summer Games

The LAC Summer Games 2024 – Participation is everything!

Motivation

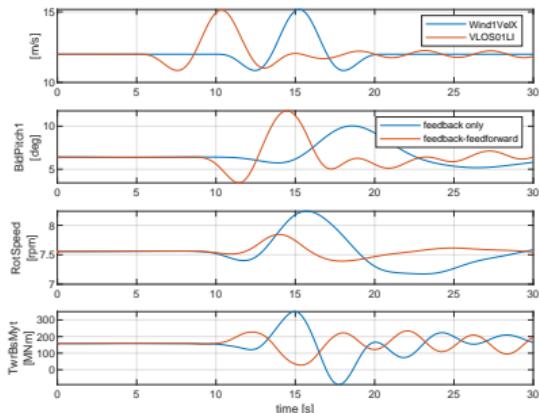
- ▶ Share knowledge and make application easier!
- ▶ Share enthusiasm with students!

Three disciplines

- ▶ The 30 s sprint **Great for controls students!**
- ▶ The 18 m/s hurdles
- ▶ The DLC 1.2 Marathon

How to participate?

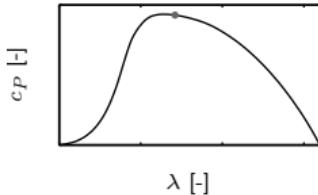
- ▶ Try the code on [Github!](#)
- ▶ Check details in the [Announcement at Zenodo!](#)
- ▶ Join the next webinar and discussion on github!
- ▶ Mobilize your students and colleagues!



Other Ideas

Direct Speed Control

- ▶ tracking of optimal λ at partial load can be improved
 - ▶ marginal energy gain, higher loads and power variation



Lidar-Assisted Yaw Control

- ▶ lidar-estimation of misalignment might be better as point measurement
 - ▶ some energy gain, depends on inhomogeneity and control strategy

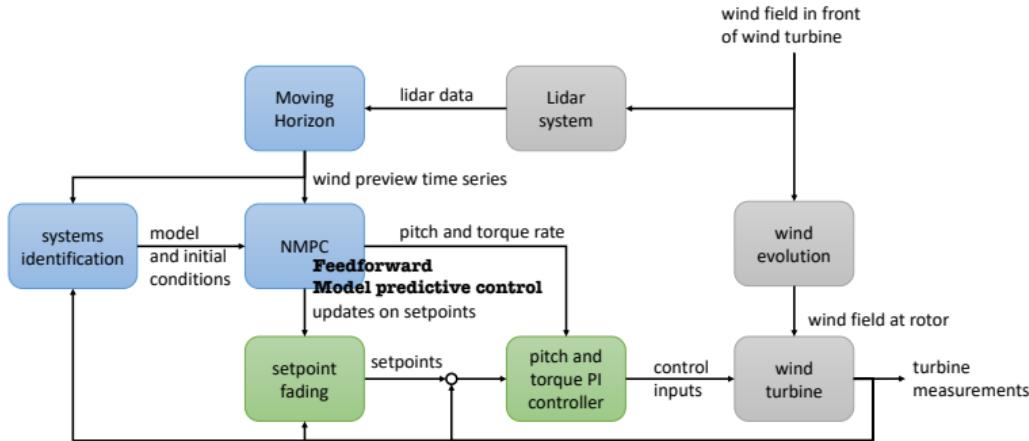
Multivariable Feedforward Control

- ▶ collective pitch and generator torque feedforward
 - ▶ advantages between partial and full load
 - ▶ can be realized as Model Predictive Control (optimizing control action over time horizon)

Lidar-Assisted Individual Pitch Control

- ▶ blade load reduction from shear or blade-effective wind speed measurement
 - ▶ shear estimation critical, advantage over feedback questionable

Nonlinear Model Predictive Control Field

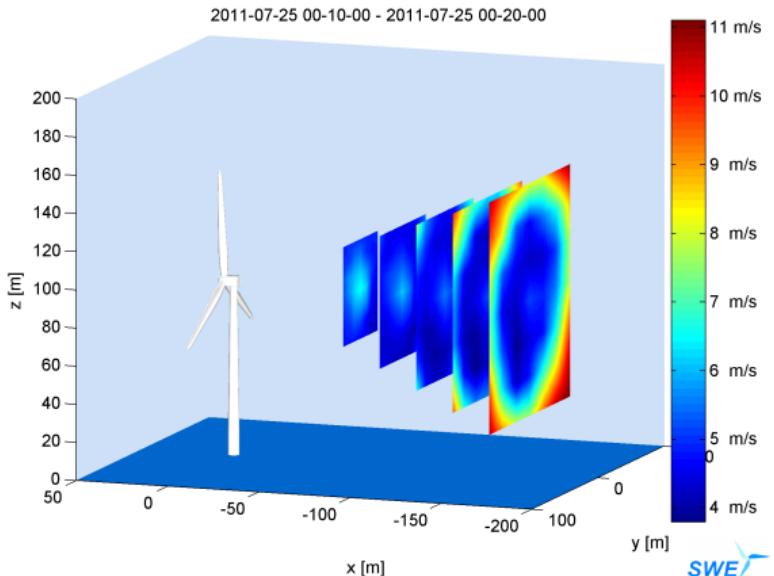
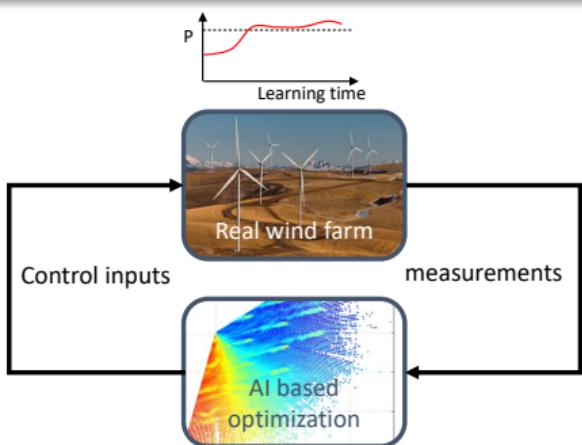


- ▶ Moving Horizon lidar data processing developed by sowento
 - ▶ Provides Wiener-filtered wind speed over several seconds
 - ▶ Currently tested at 6 MW wind turbine with NMPC from IAV
 - ▶ Feedforward concept will be tested on 2 MW WETI research turbine

Lidar-Assisted Wind Farm Control

Lidar systems can be used for:

- ▶ wake model verification
 - ▶ farm control monitoring
 - ▶ track wind turbine wakes
 - ▶ optimize wind farm operation



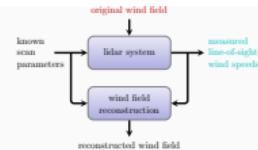
Conclusions

Main questions

- ▶ How can useful wind preview signals be extracted from lidar data?

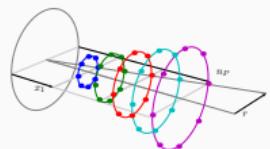
Nacelle-based lidar needs new model-based wind field reconstruction!

- ▶ Main problem: limitation to line-of-sight wind speeds.
- ▶ Good estimates can be provided with adequate wind models!
- ▶ For lidar-assisted control we can assume perfect alignment and the Frozen Turbulence Hypothesis to condense all line-of-sight to an estimate for the rotor-effective wind speed.



Correlation between lidar and rotor very important!

- ▶ Rotor-effective wind speed can be also estimated from turbine data.
- ▶ We can model spectra, transfer function and coherence based on Fourier transform of measurement equations and using turbulence models.
- ▶ Lidar data needs to be filtered according to the transfer function.
- ▶ We can use the correlation model for scan optimization and filter design.



Conclusions

Main questions

- ▶ How can these signals be used to improve wind turbine control?

Load reduction is possible, transformable into more energy!

- ▶ nonlinear feedforward controller as add-on to conventional feedback
 - ▶ in region 3 collective pitch feedforward control based on static pitch curve
 - ▶ design assuming perfect wind preview and then adding adaptive filter
 - ▶ currently first commercial applications

References

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- [4] F. Guo and D. Schlipf. "Lidar Wind Preview Quality Estimation for Wind Turbine Control". In: *American Control Conference*. New Orleans, LA, USA, May 2021. DOI: 10.23919/ACC50511.2021.9483442.
- [5] D. Schlipf, F. Lemmer, and S. Raach. "Multi-Variable Feedforward Control for Floating Wind Turbines Using Lidar". In: *International Ocean and Polar Engineering Conference*. Oct. 2020. DOI: 10.18419/opus-11067. URL:
<https://onepetro.org/ISOPEIOPEC/proceedings-abstract/ISOPE20/A11-ISOPE20/ISOPE-I-20-1174/446367>.

Please let me know if you have further questions!

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