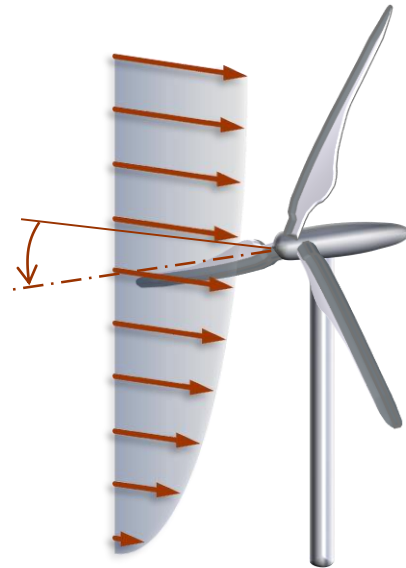


IMPROVED KNOWLEDGE OF WIND CONDITIONS FOR WIND TURBINE AND WIND FARM CONTROL

Carlo L. Bottasso and **Stefano Cacciola**,
TU München

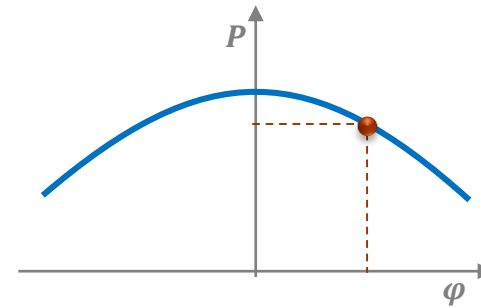
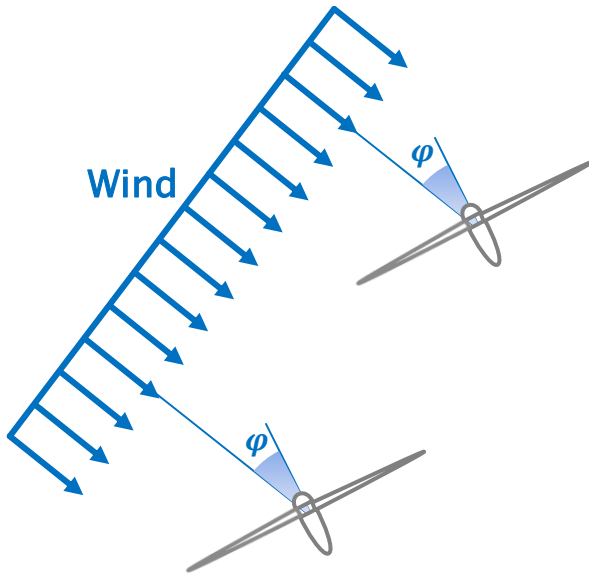


4th MSE colloquium
München, Germany, July 3, 2014

Motivation

Operating in **yawed conditions**:

- **Reduces power** as $\cos^3(\text{yaw})$



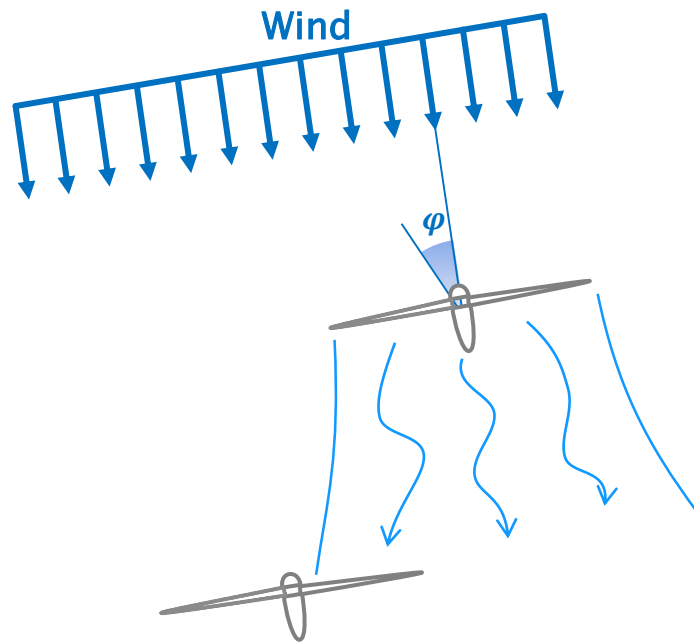
- Causes **vibrations** and excites low-damped side-side modes
- Changes airfoil AoA, possible **performance degradations** (e.g., dynamic stall)



Motivation

Sometimes yawing a machine is helpful:

Due to the presence of the **wake** of the first turbine, the downstream turbine feels:



- Lower **mean wind speed** over the rotor disk (**less power available**)
- Higher **turbulence intensity** and periodic loads (**fatigue** problems)
- **Performance degradations** (e.g., dynamic stall)

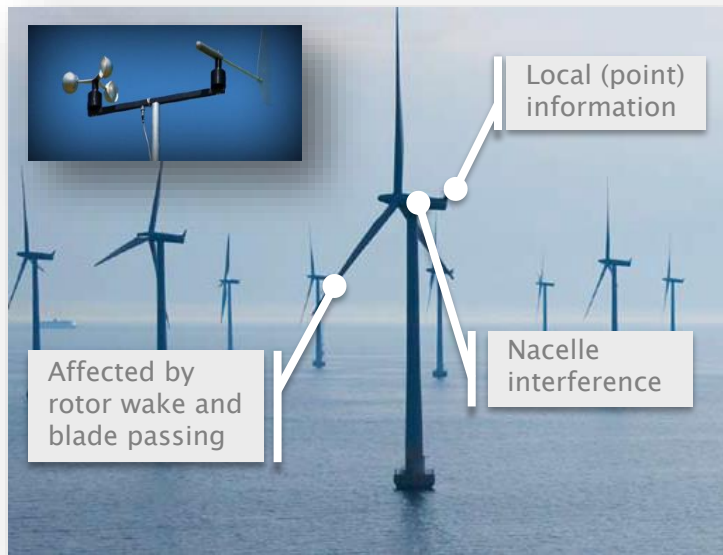
One could yaw a turbine to **improve performances** of **downstream turbines**



Motivation

Reliable yaw measurements are **difficult** to obtain.

Nacelle anemometer



Lidar (lased doppler anemometer)

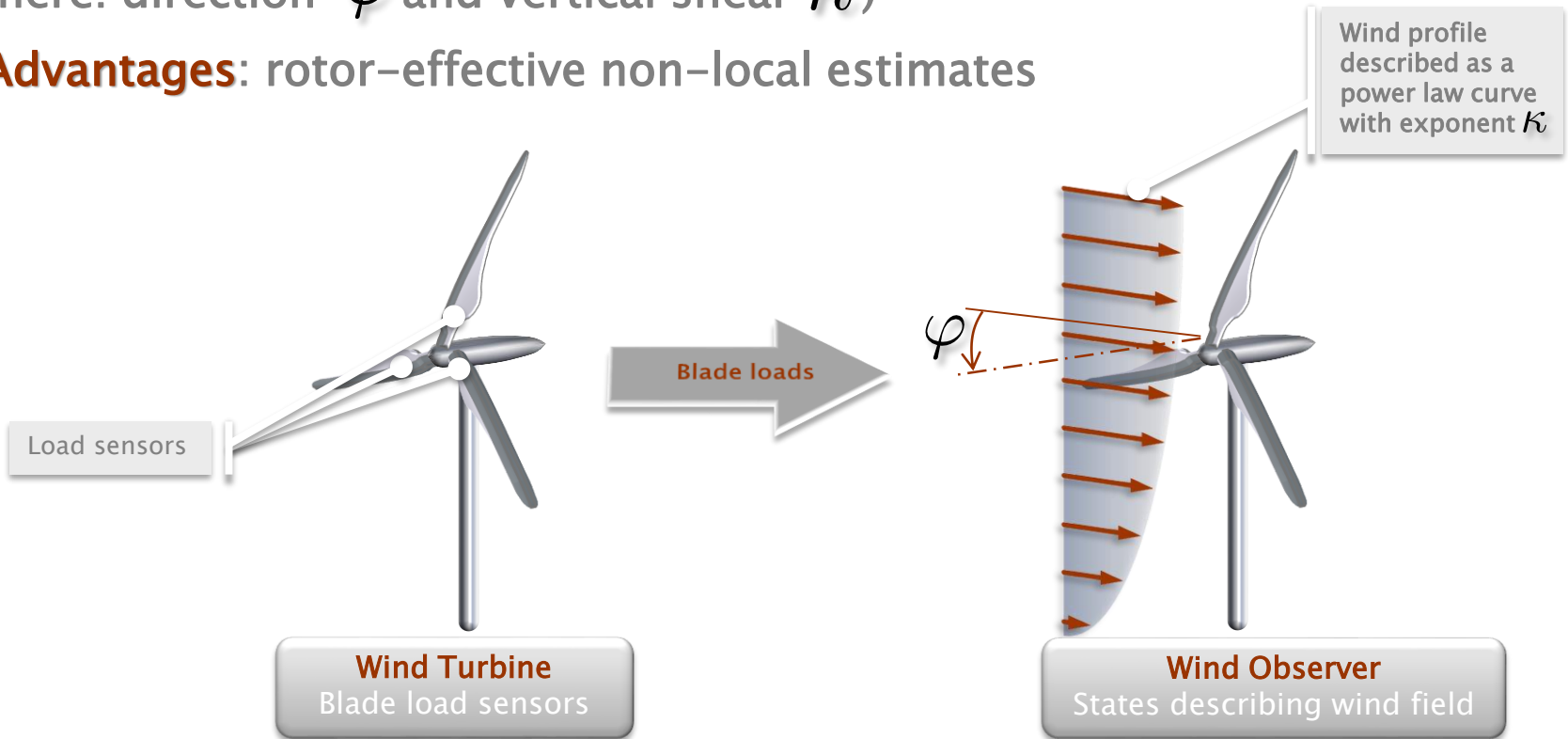


The Concept in a Nutshell

Any **anisotropy** of the wind generates **periodic loads**

By interpreting the rotor response, one can infer desired wind states
(here: direction φ and vertical shear κ)

Advantages: rotor-effective non-local estimates



The rotor is the **ultimate anemometer**



Outline

- Formulation of a general observation model
 - Model structure from an analytical blade response model
 - Observer synthesis by identification
 - Implementation
- Results
 - Testing in a high-fidelity simulation environment
 - Testing with an aeroelastically-scaled wind tunnel model
 - Field testing on NREL CART3 wind turbine
- Conclusions and outlook



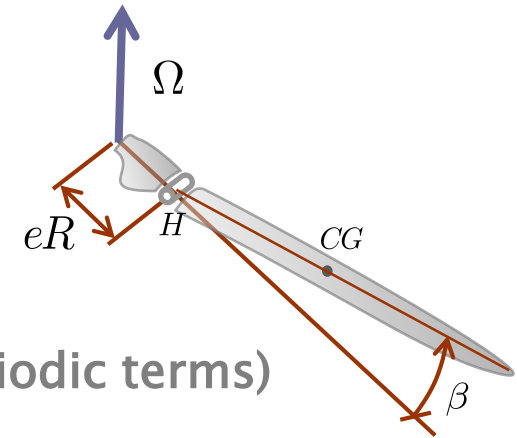
Observation Model Structure

Inspired by rigid flapping blade (Eggleston & Stoddard 1987):

- Assume 1P harmonic solution

$$\beta = \beta_0 + \beta_{1s} \sin \psi + \beta_{1c} \cos \psi$$

- Insert into blade dynamics, drop h.o.t.'s
- Solve for wind misalignment and shear (compute misalignment and shear from 1P periodic terms)



Remarks:

- Linear relationship between misalignment/shear and blade 1P
- Misalignment and shear are **independent** and **observable**
- **Wind-dependent** coefficients
- **Gyroscopic effects** during yawing (to be considered)

Not useful for **practical applications**, due to limitations/simplifications of flapping blade model problem

A General Observation Model

Linear input-output wind-scheduled model:

Wind-speed-dependent coefficients

$$\begin{Bmatrix} \varphi \\ \kappa \end{Bmatrix} = \mathbf{A}(V)\mathbf{m} + \mathbf{b}(V)$$

Driving input (blade root loads):

$$\bar{\mathbf{m}} = (m_{1c}^{\text{OP}}/m_0^{\text{OP}}, m_{1s}^{\text{OP}}/m_0^{\text{OP}}, m_{1c}^{\text{IP}}/m_0^{\text{IP}}, m_{1s}^{\text{IP}}/m_0^{\text{IP}})^T$$

OP: out-of-plane, IP: in-plane

1P load harmonics by **multiblade Coleman-Feingold transformation:**

$$\begin{Bmatrix} m_0 \\ m_{1c} \\ m_{1s} \end{Bmatrix} = \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 2 \cos \psi_1 & 2 \cos \psi_2 & 2 \cos \psi_3 \\ 2 \sin \psi_1 & 2 \sin \psi_2 & 2 \sin \psi_3 \end{bmatrix} \begin{Bmatrix} m_1 \\ m_2 \\ m_3 \end{Bmatrix}$$

Model Identification

N observations of wind parameters/associated blade response harmonics:

$$\mathbf{W} = \mathbf{T}(V)\mathbf{M}$$

where

$$\mathbf{W} = \left[\left\{ \begin{array}{c} \varphi_1 \\ \kappa_1 \end{array} \right\}, \left\{ \begin{array}{c} \varphi_2 \\ \kappa_2 \end{array} \right\}, \dots, \left\{ \begin{array}{c} \varphi_N \\ \kappa_N \end{array} \right\} \right]$$

$$\mathbf{M} = \left[\left\{ \begin{array}{c} \mathbf{m}_1 \\ 1 \end{array} \right\}, \left\{ \begin{array}{c} \mathbf{m}_2 \\ 1 \end{array} \right\}, \dots, \left\{ \begin{array}{c} \mathbf{m}_N \\ 1 \end{array} \right\} \right]$$

$$\mathbf{T}(V) = [\mathbf{A}(V), \mathbf{b}(V)]$$

Compute unknown model coefficients by **least-squares**:

$$\mathbf{T}(V) = \mathbf{W}\mathbf{M}^T(\mathbf{M}\mathbf{M}^T)^{-1}$$

Wind scheduling: identify observation model at different wind speeds V_k to cover entire operating envelope of the wind turbine

Linearly interpolate at run time: $\mathbf{T}(V) = (1 - \xi)\mathbf{T}(V_k) + \xi\mathbf{T}(V_{k+1})$



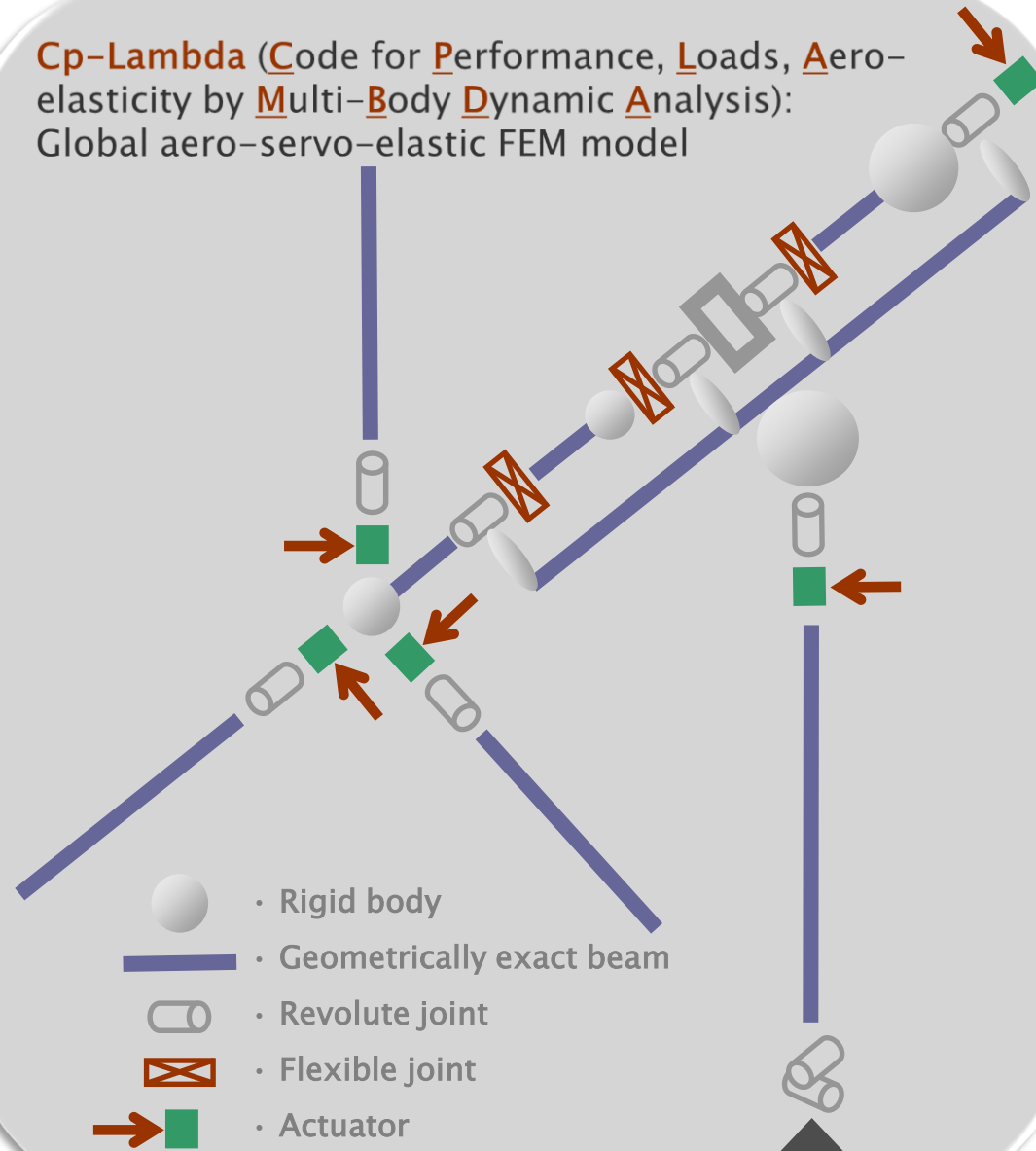
Testing in a Simulation Environment

3MW high-fidelity HAWT model

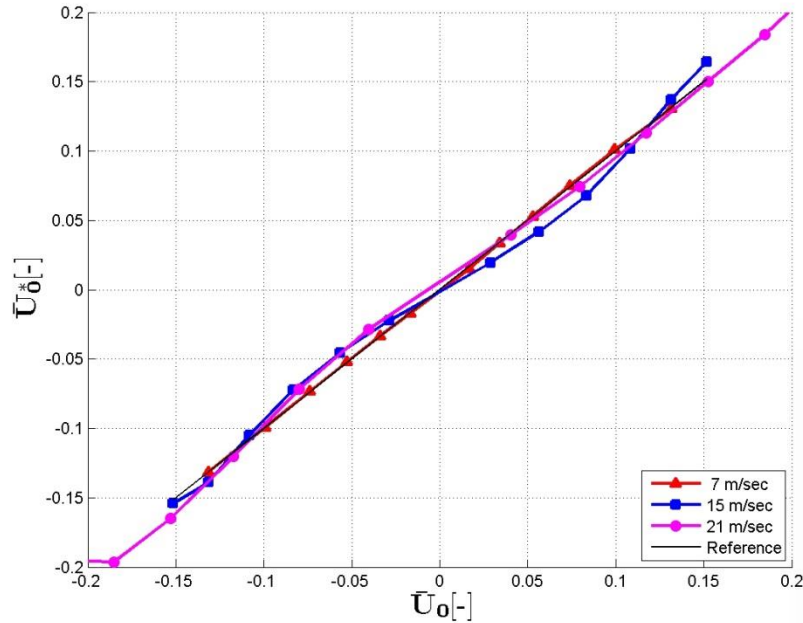
Cp-Lambda highlights:

- Geometrically exact composite-ready beam models
- Generic topology (Cartesian coordinates+Lagrange multipliers)
- Dynamic wake model (Peters-He, yawed flow conditions)
- Efficient large-scale DAE solver
- Non-linearly stable time integrator
- Fully IEC 61400 compliant (DLCs, wind models)

Cp-Lambda (Code for Performance, Loads, Aero-elasticity by Multi-Body Dynamic Analysis):
Global aero-servo-elastic FEM model

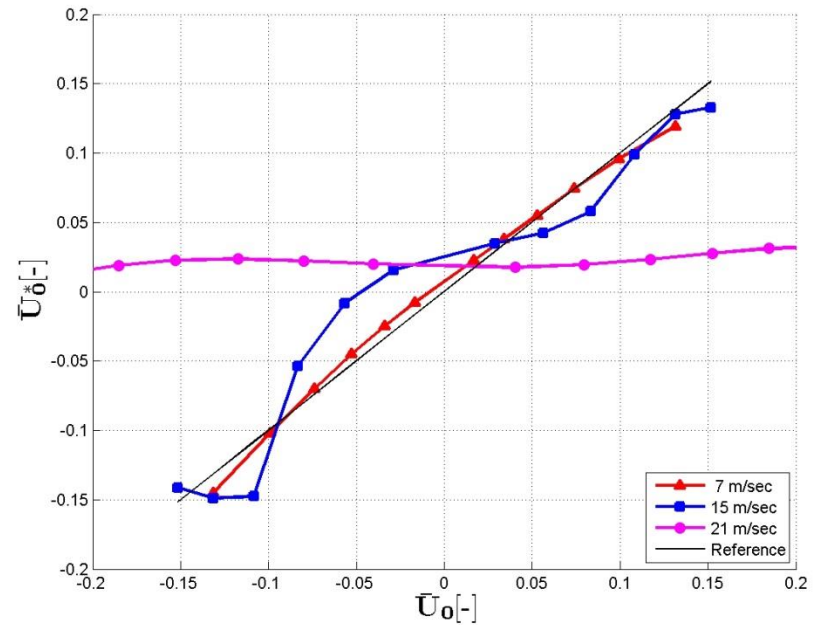


Verification of Observability

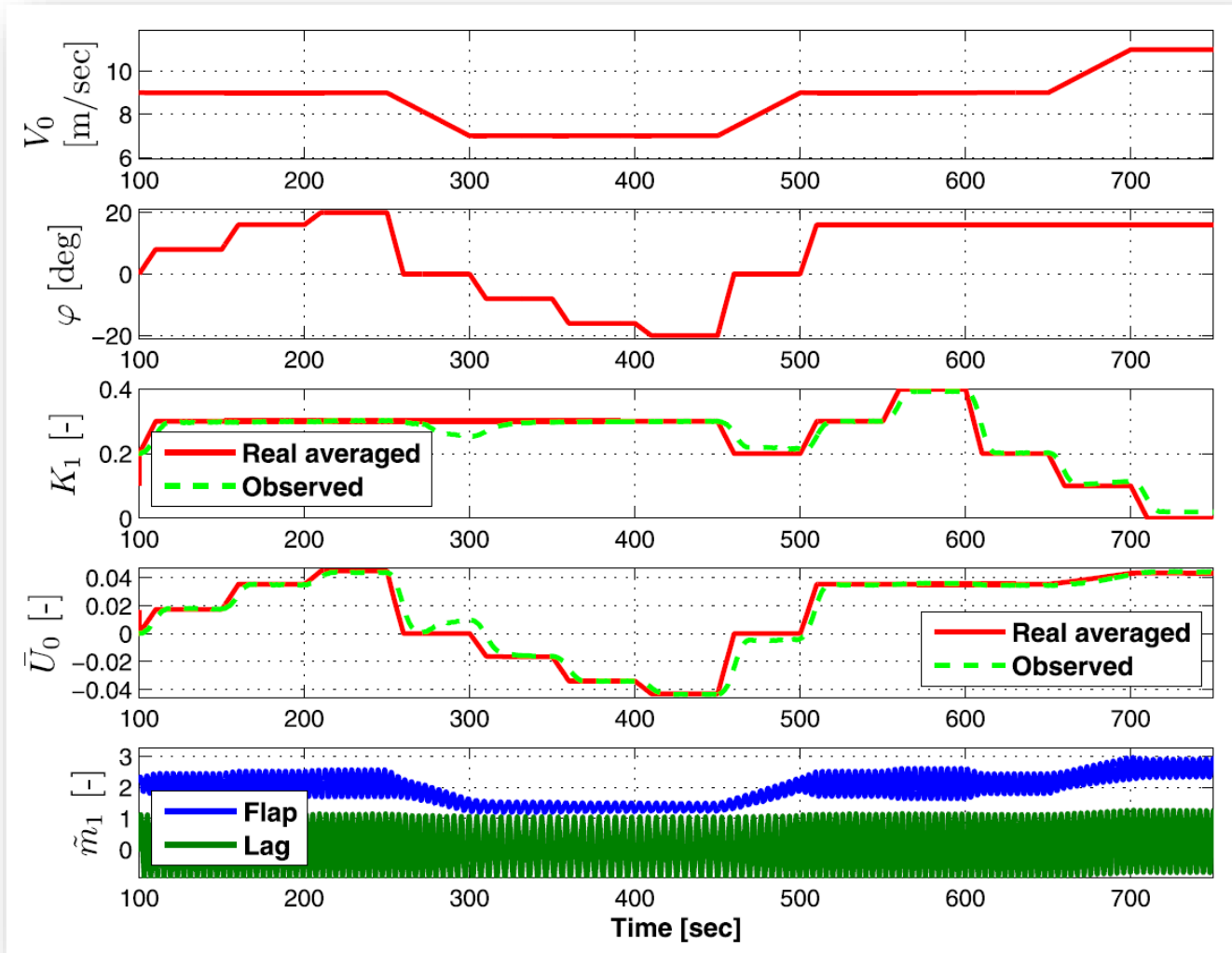


▲ OP and IP loads

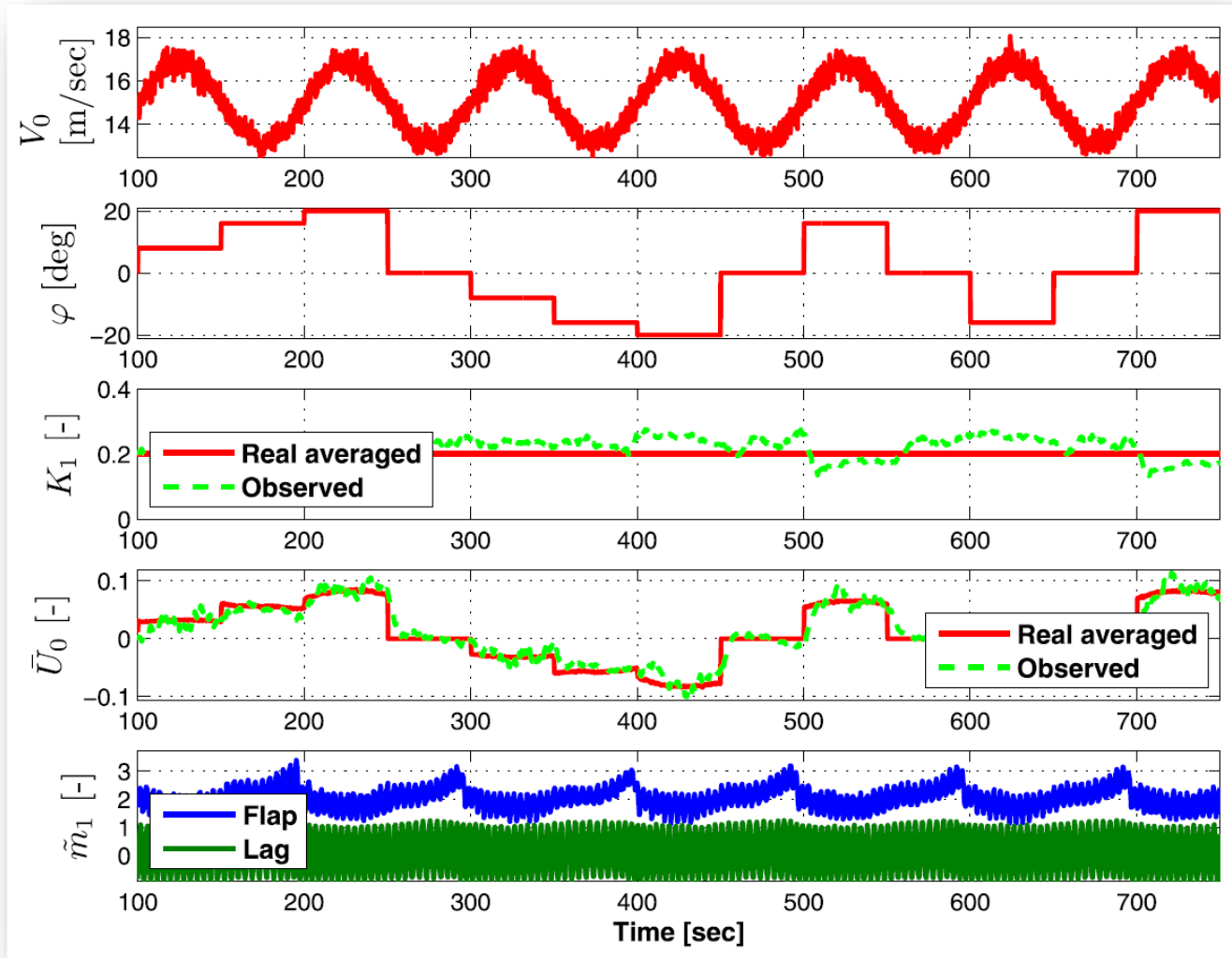
▼ OP loads only



Yaw Observation with Varying Shear



Yaw Observation with 10% Turbulence and Varying Mean Wind Speed

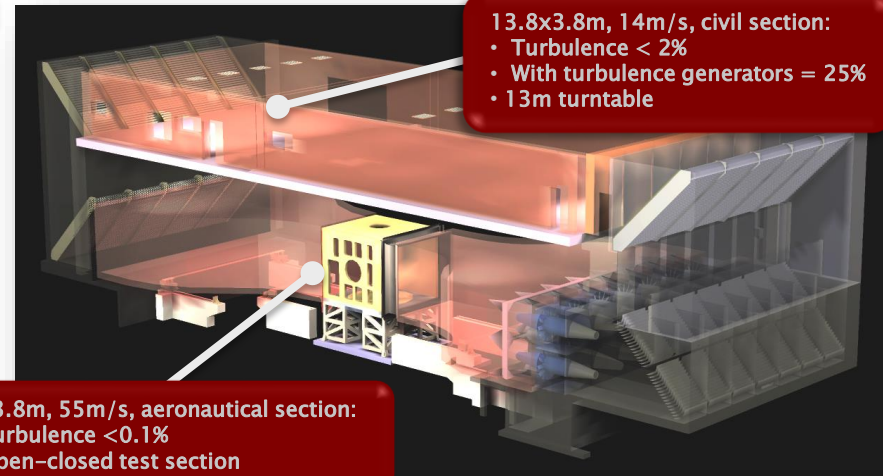


Wind Tunnel Testing

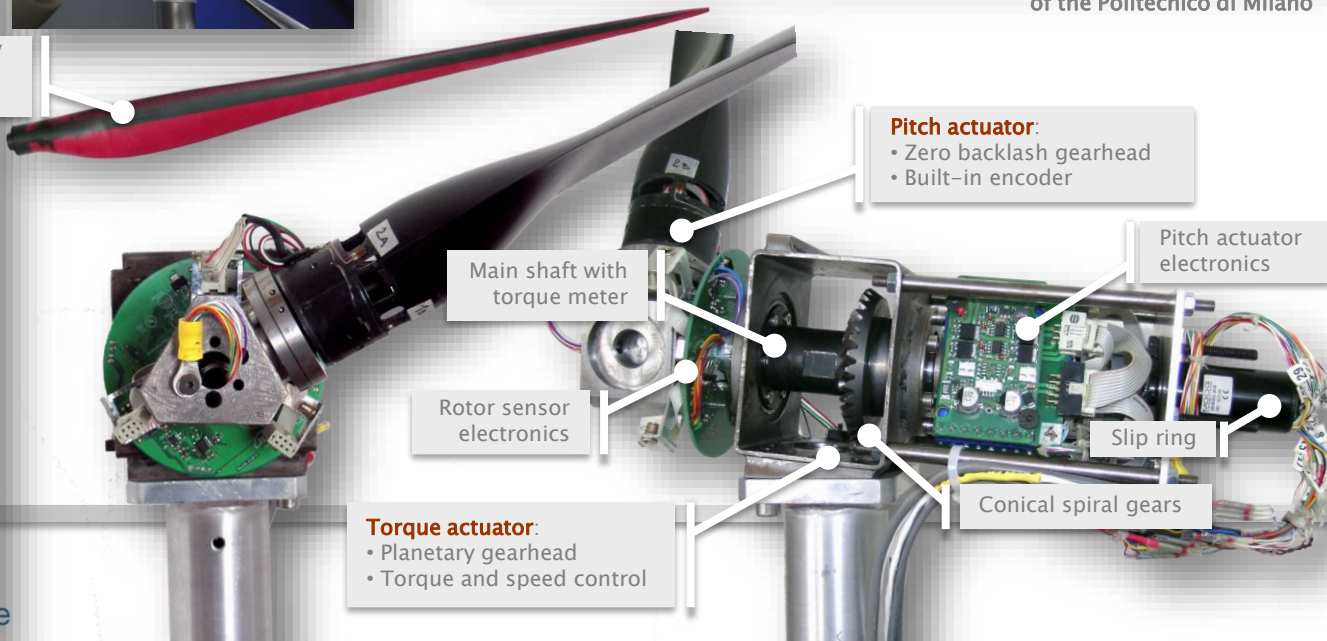
WT²: aeroelastically-scaled wind tunnel model of the Vestas V90 wind turbine with individual blade pitch and torque control

Applications:

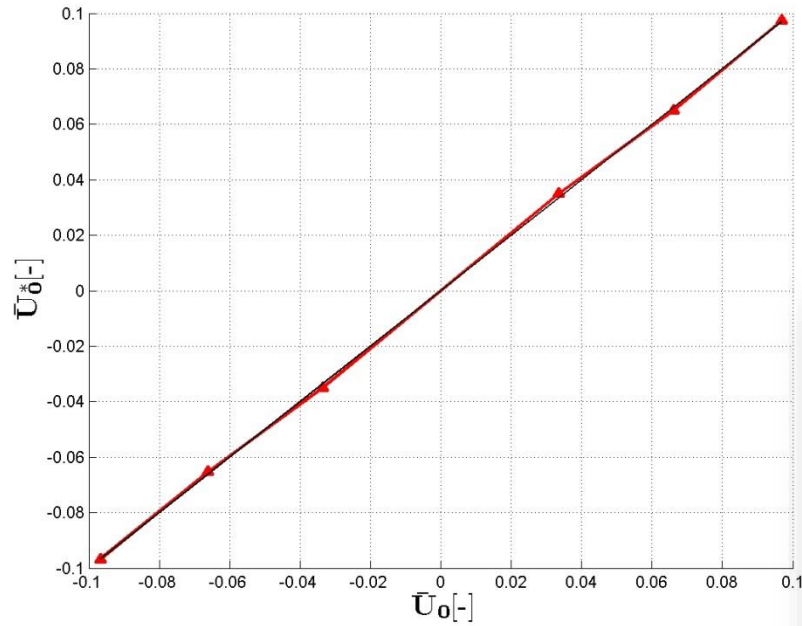
- Testing of advanced control laws and supporting technologies
- Testing of extreme operating conditions
- Tuning of mathematical models
- Aeroelasticity and system identification of wind turbines
- Multiple wind turbine interactions
- Off-shore wind turbines (moving platform actuated by hydro-structural model)



Civil-Aeronautical Wind Tunnel of the Politecnico di Milano

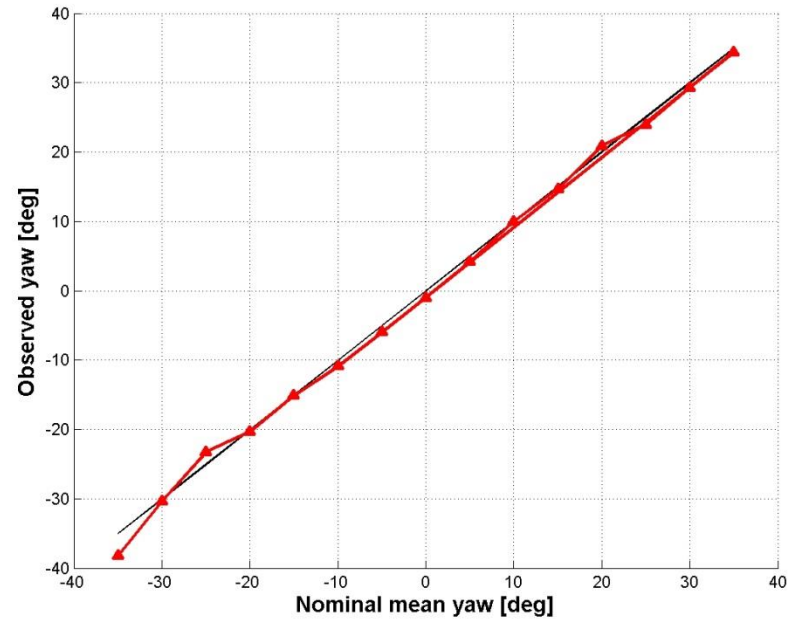


Wind Tunnel Testing



▲ Identification data set

▼ Verification data set

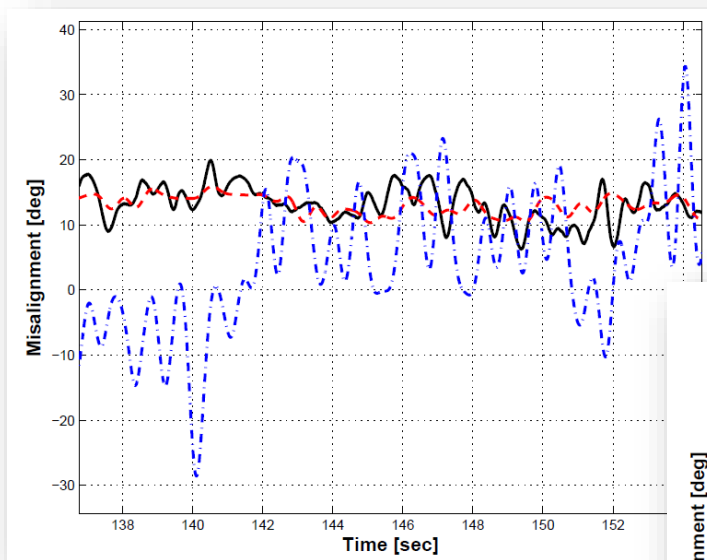


Field Testing

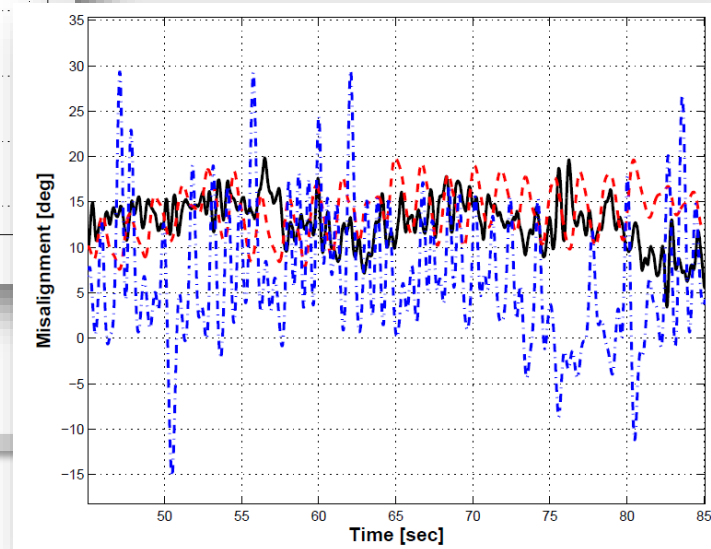
NREL CART3 wind turbine

Model identified from real field measurements
(elimination of outliers by **RANSAC**)

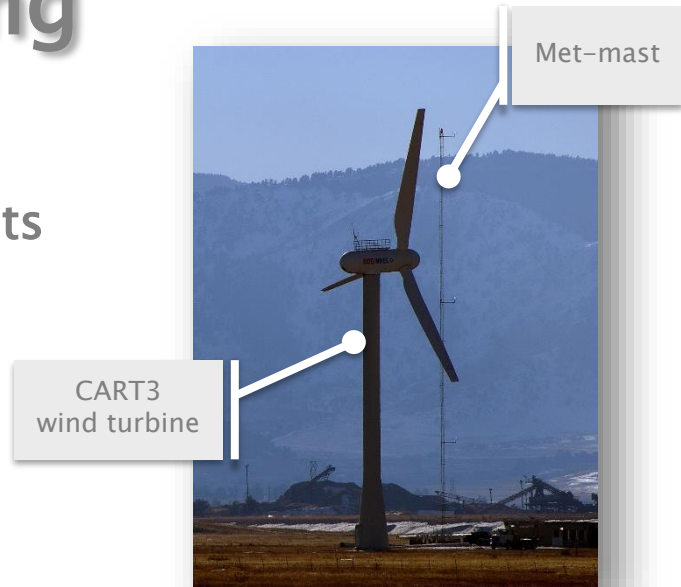
Two typical time histories:



Good match at the low frequencies
(what needed for yaw control)



Black solid: met mast
Blue dash-dotted: wind vane
Red dashed: observer



Conclusions

- Successful verification in **simulation, wind tunnel** and **field testing**
- Simple **model-free** identification
- Good quality of the estimates, **superior** to on-board wind vanes
- **Negligible computational cost**

Outlook:

- Further testing on larger machines, should see even better results
- Field testing of shear observer
- On-board use of observed wind states



Acknowledgements

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THANK YOU FOR YOUR ATTENTION