



## **Nacelle lidar for power perf. – the UniTTe approach to retrieve V**

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# Nacelle lidar for power perf. – the UniTTe approach to retrieve $V_\infty$

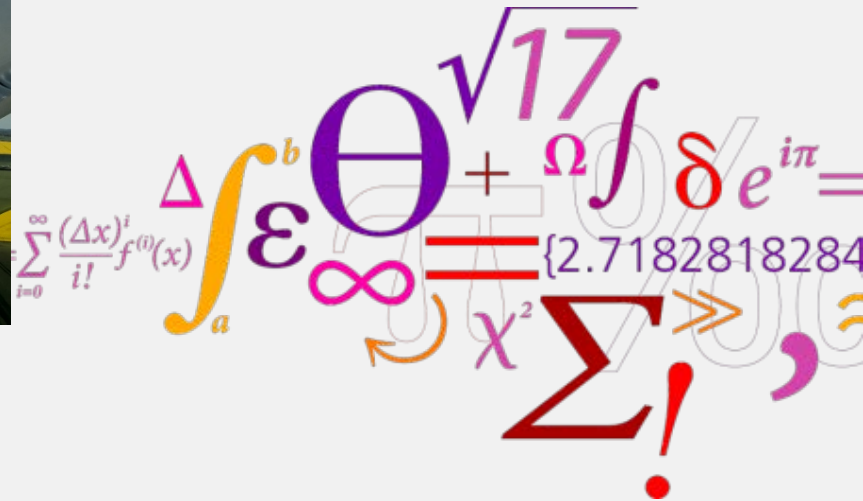


A. Borraccino, R. Wagner

IEA Wind task 32 – workshop nacelle lidars  
27<sup>th</sup> September 2017

**DTU Wind Energy**  
Department of Wind Energy

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# Power performance testing

## The modern ways

Remote sensing instruments

—

Future/Now: use of **nacelle-based wind lidars**



ZephIR Dual Mode  
(scanning)  
by *ZephIRLidar*



Wind Iris  
(4-beam)  
by *AventLidar*

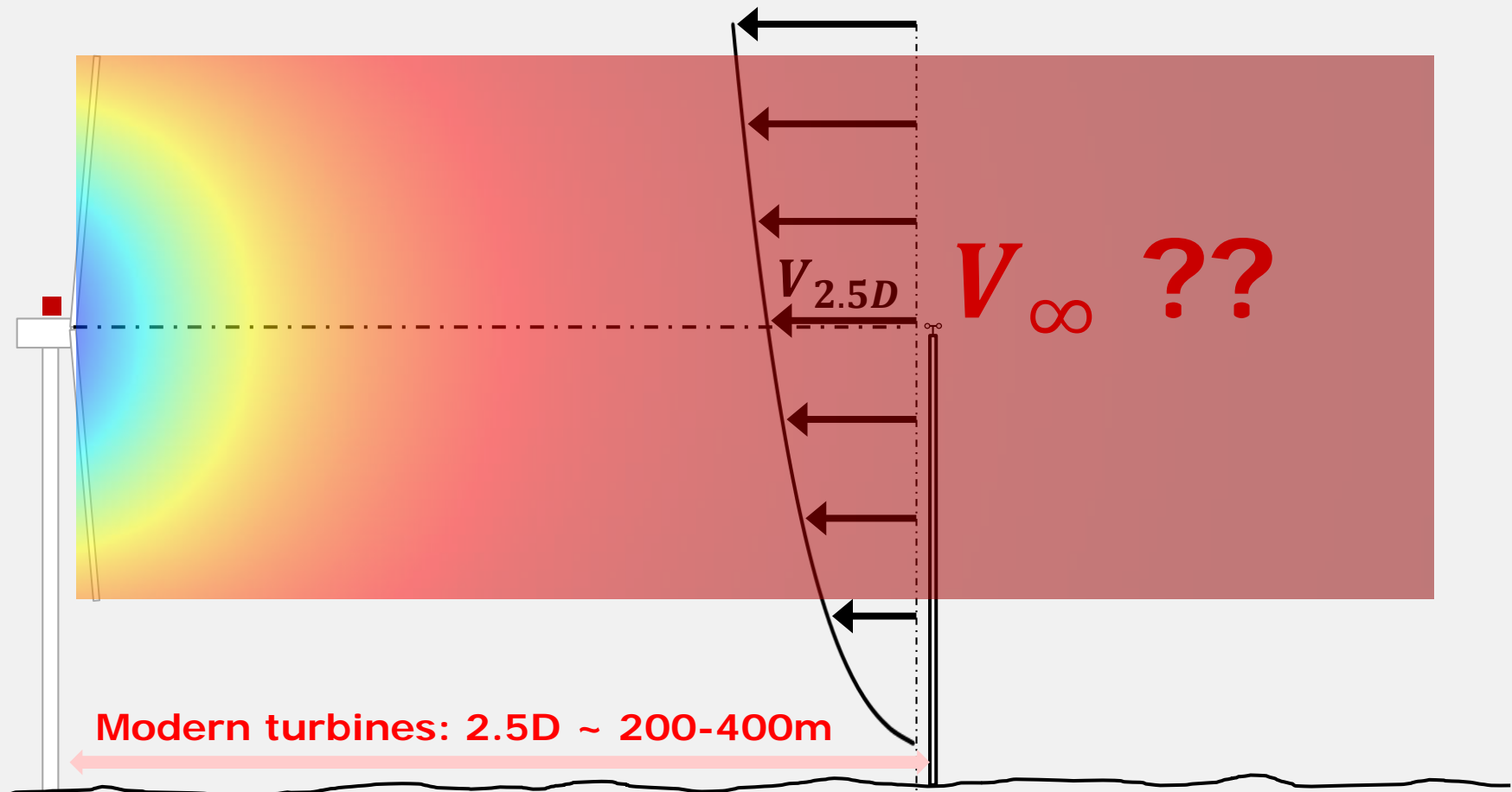


Wind Eye  
(4-beam)  
by *Windar Photonics*



Diabrezza  
(9-beam)  
by *Mitsubishi Electric*

# Searching for free stream wind speed

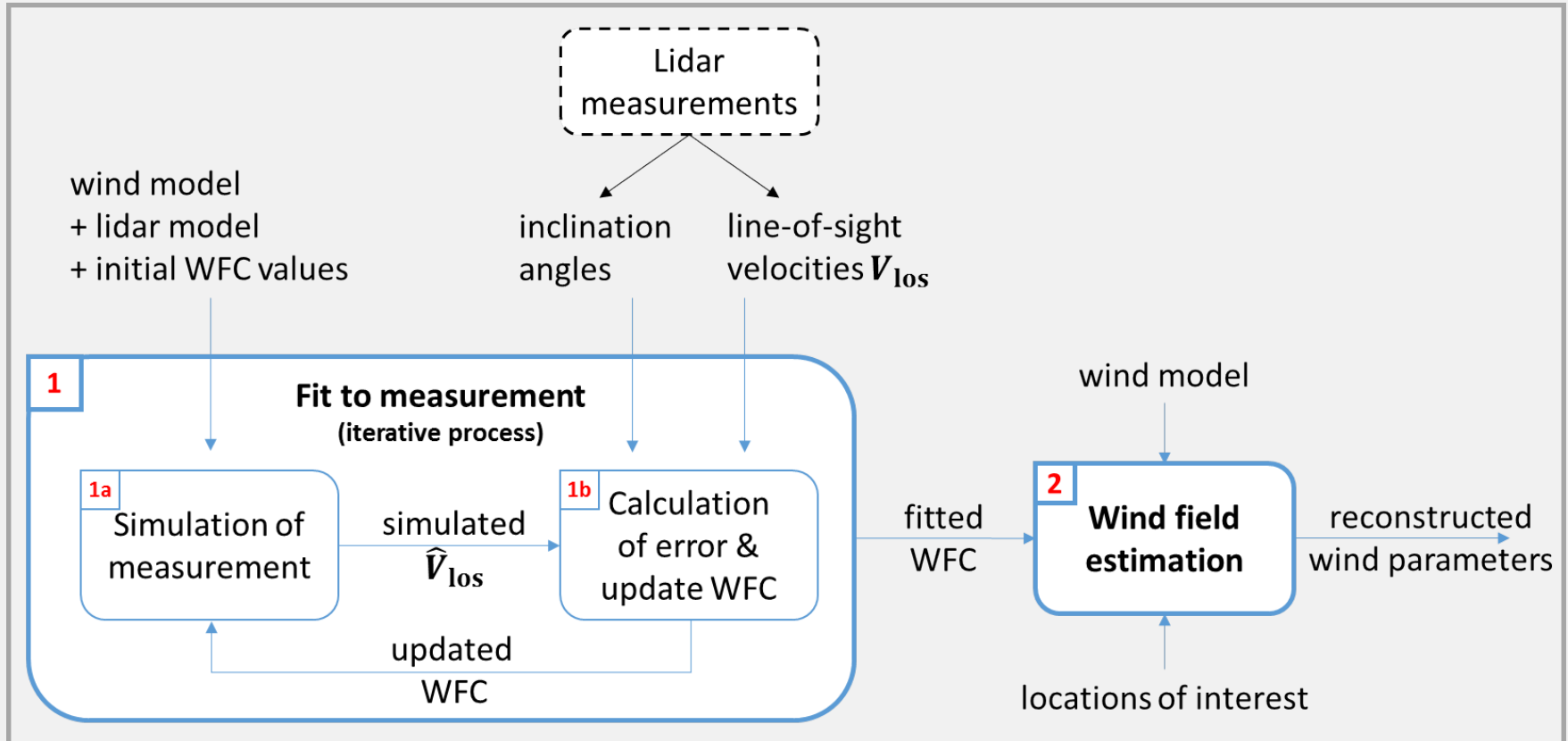


- Decorrelation WSpeed / power
- Hub height speed insufficient?
- 2.5D not really free wind ...

# Model-fitting Wind Field Reconstruction

- **Method is (not new...)**

Schlipf D., Rettenmeier A., Haizmann F., Hofsäß M., Courtney M. and Cheng, P. W.:  
 “Model Based Wind Vector Field Reconstruction from Lidar Data”, DEWEK, 2012.



- **need new “wind models” for profiling nacelle lidars, suitable for power performance testing**

# What's wrong with 2.5D?

- **Lidar range capabilities**

- Soon not sufficient for very large turbines
- Or systems will become more expensive

- **Measurement/beam locations**

- Are/will be too far away
  - to accept WFR assumptions: inhomogeneity, lack of coherence, etc
  - nac. lidars measure wind less and less representative to what the turbine feels
- Affected by e.g. terrain or site features

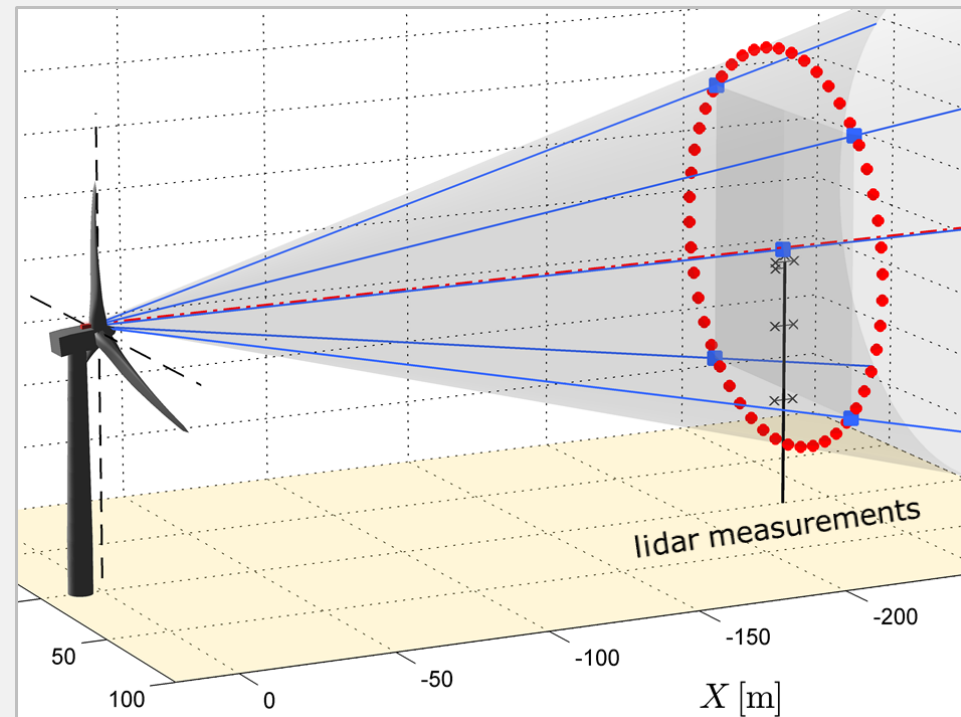
- **"Decorrelation" issues might come back**

- **AND...**

- **2.5D really is NOT free stream!**
- Under-estimation of  $V_\infty$  by about 0.7%
  - According to models, confirmed by measurements
  - Should be accounted into AEP calculations...

# Wind model accounting for shear

- Use lidar measurements at 2.5 rotor diameters
- “static” model: stationarity assumed
- Assumes horizontal homogeneity and power law shear profile
- **Fits three wind characteristics**
  - wind speed  $V_0$  (@ $H_{hub}$ )
  - + relative wind dir.  $\theta_r$  (yaw misalignment)
  - + shear exponent  $\alpha_{exp}$





# Combined wind-induction model

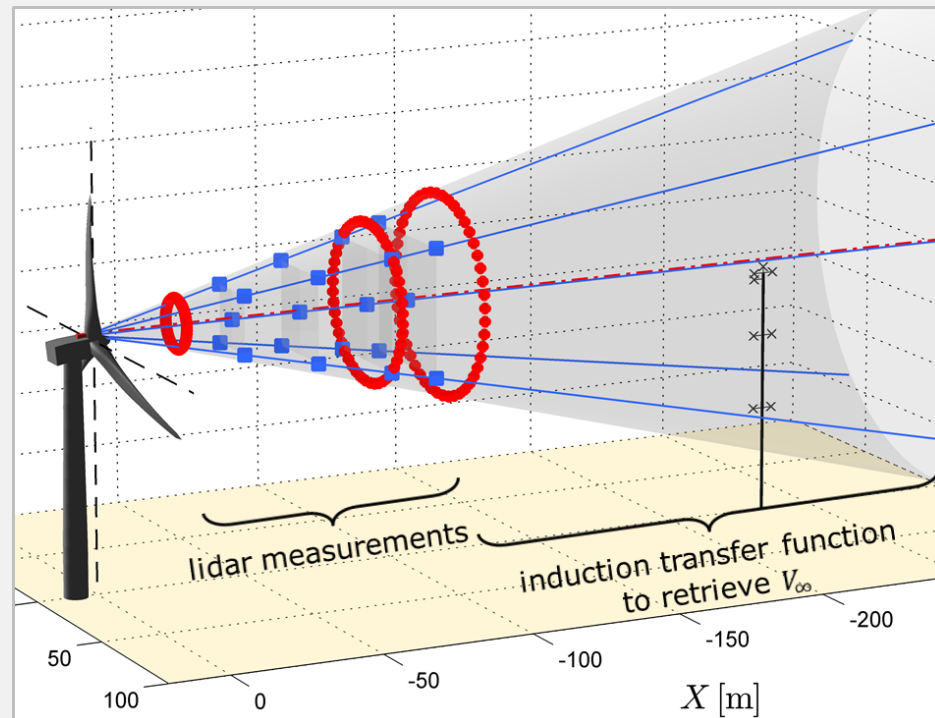
- Use lidar measurements at multiple distances close to rotor
- Additionally assumes simple induction model:

(from actuator disk and vortex sheet theory)

$$\frac{U(x)}{U_\infty} = 1 - a_{ind} \left( 1 + \frac{\xi}{\sqrt{1 + \xi^2}} \right)$$

- **Fits four wind characteristics**

- ➔ Free stream wind speed  $V_\infty$  (@ $H_{hub}$ )
- + relative wind dir.  $\theta_r$
- + shear exponent  $\alpha_{exp}$
- + induction factor  $a_{ind}$



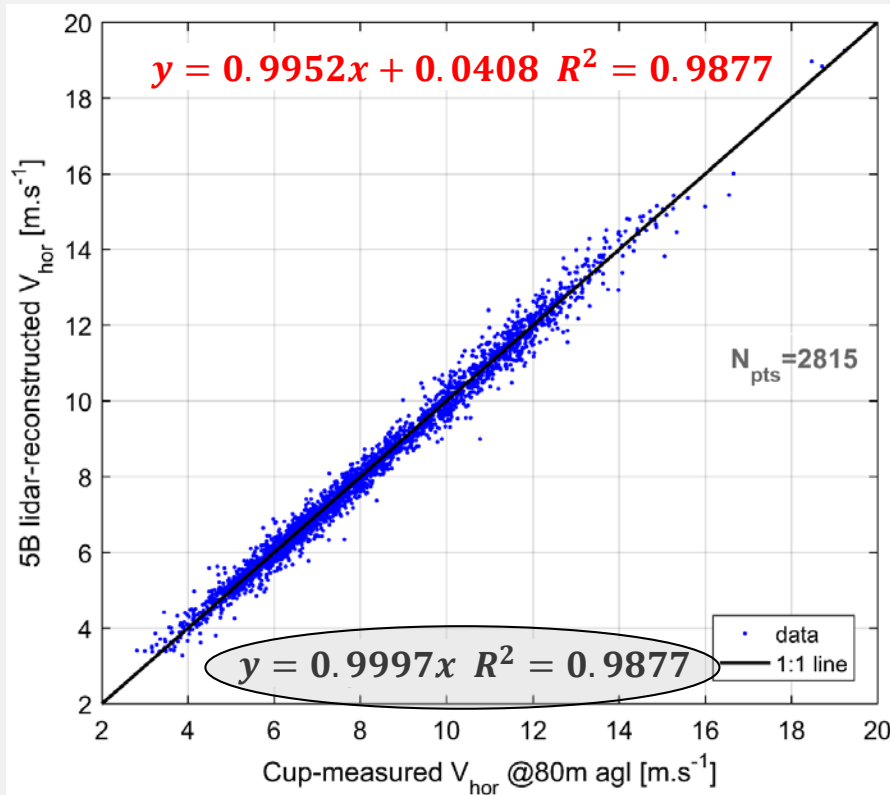


# Wind speed results

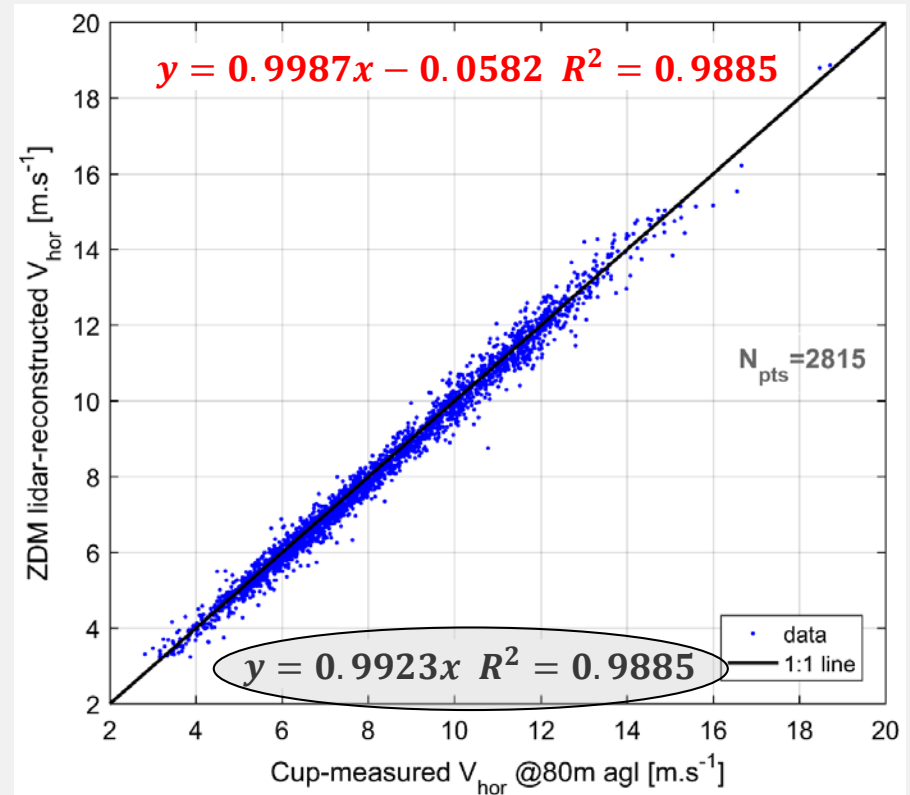
Mast comparison, WFR using the **wind-induction model**

- horizontal speed estimated @hub height and 2.5D\_rot
- IEC “free sector”: [110°,219°]

**5B-demo: use the 5 LOS**  
**4 dist, from 0.5 to @1.2D\_rot**



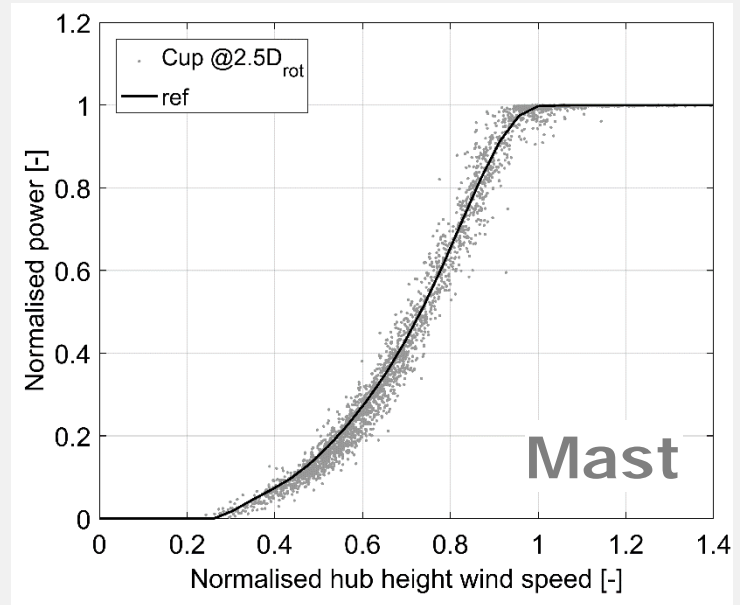
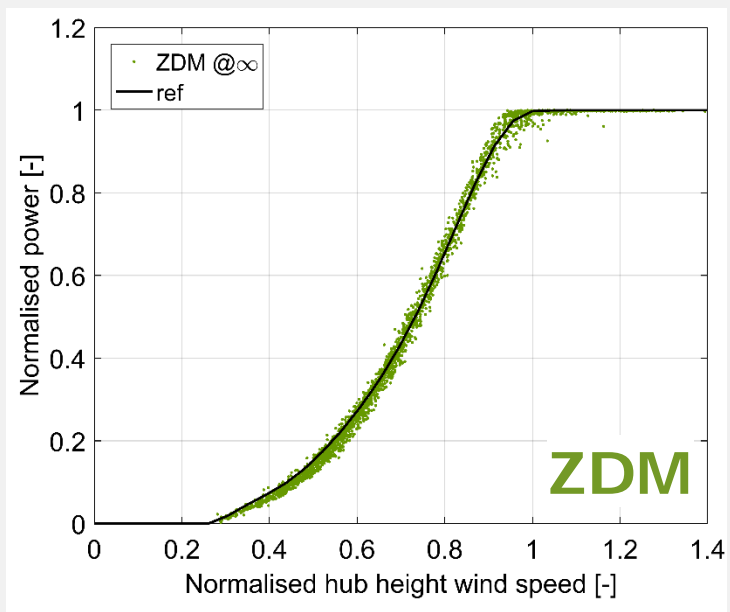
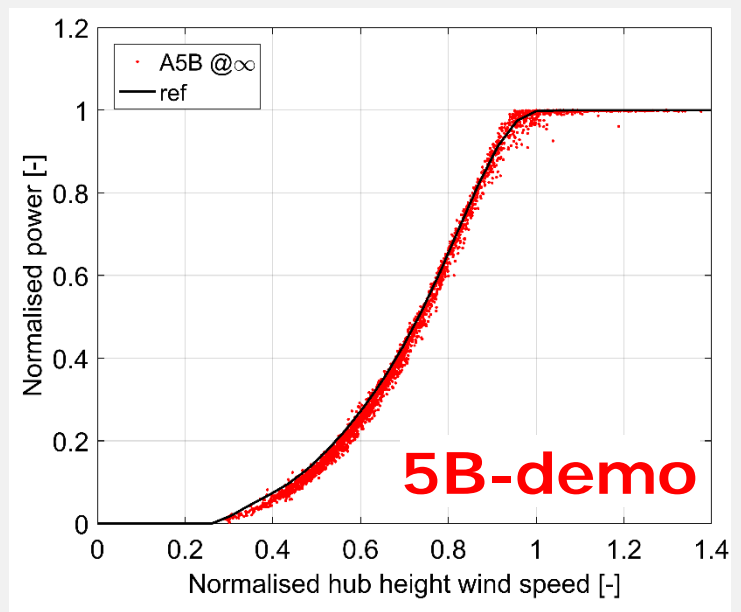
**ZDM: use 6 LOS**  
**3 dist., from 0.3 to 1.2D\_rot**





# Measured Power curves (scatter)

WFR using **wind-induction model**



# Challenges in PCV with nacelle lidars

- **Need to give directions to methods for wind field reconstruction**
  - WFR model is of critical importance for accurate wind estimates
  - What kind of shear/veer model?
  - How to quantify model inadequacy? (e.g. fitting residuals)
- **Rotor equivalent wind speed**
  - By integration of shear profile?
  - Some geometrical issues...
- **Accounting for terrain**
  - Elevation data integrated as inputs to the WFR codes?
  - Classification of terrain: different WFR models to recommend?
- **Practical questions**
  - integration of brackets into turbine design
  - alignment of nacelle lidar to rotor axis

# Thanks for your attention!

Scientific article: *Wind Energy Science*

## Research articles

### **Wind Field Reconstruction from Nacelle-Mounted Lidars Short Range Measurements**

Antoine Borraccino<sup>1</sup>, David Schlipf<sup>2</sup>, Florian Haizmann<sup>2</sup>, and Rozenn Wagner<sup>1</sup>

<sup>1</sup>DTU Wind Energy, Roskilde, Denmark

<sup>2</sup>Stuttgart Wind Energy, University of Stuttgart, Germany

My PhD thesis:

*Remotely measuring the wind using turbine-mounted lidars*

More info:

- website [www.unitte.dk](http://www.unitte.dk)
- contact:  
[borr@dtu.dk](mailto:borr@dtu.dk), [rozn@dtu.dk](mailto:rozn@dtu.dk)



# Does this make it any easier?



Flow disturbed by turbine wakes !

(very) complex terrain

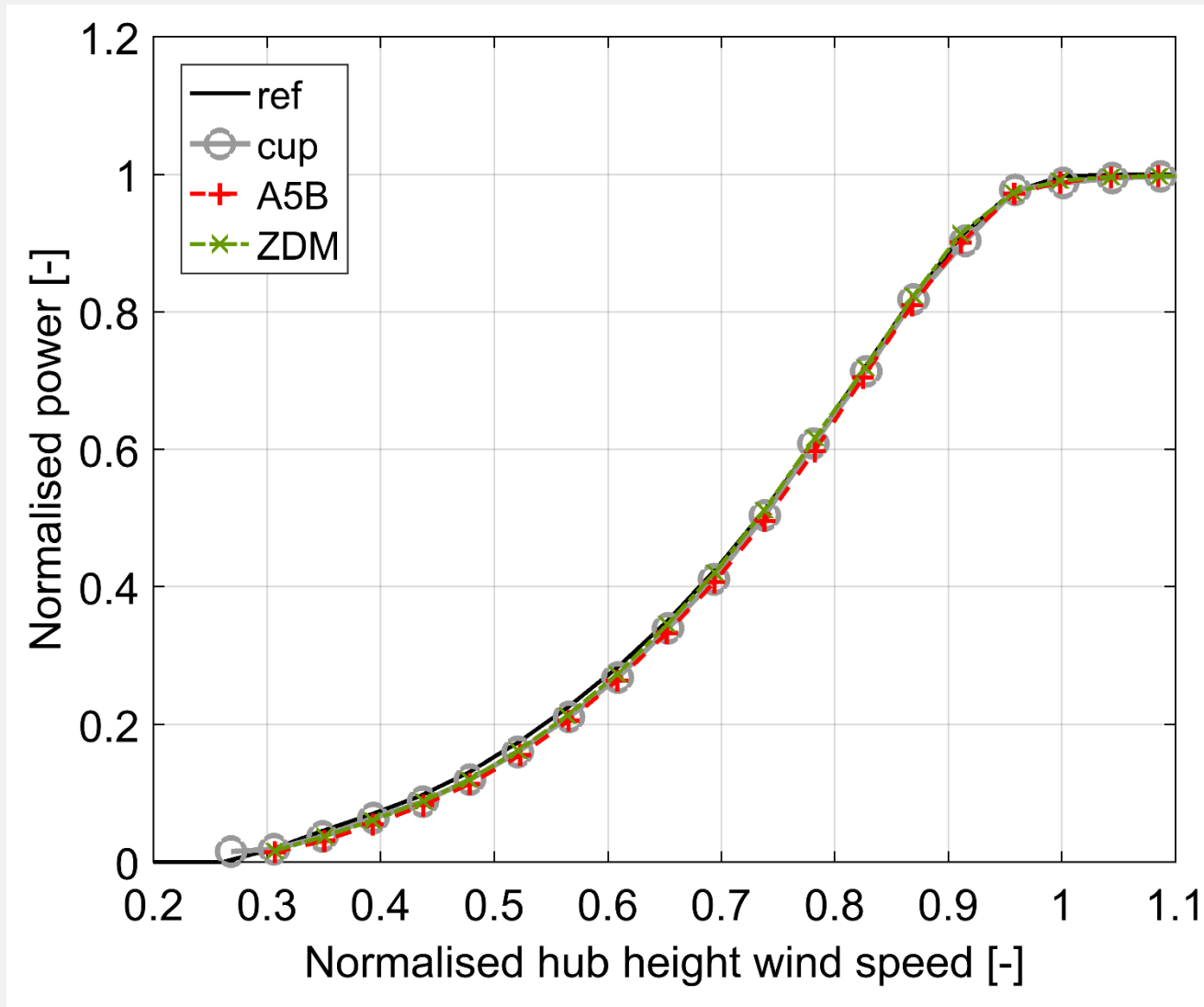


Perdigão.  
credit: N. Vasiljevic

# Preparing for questions - Wind Field Reconstruction

# Measured Power curves (binned)

WFR using **wind-induction model**

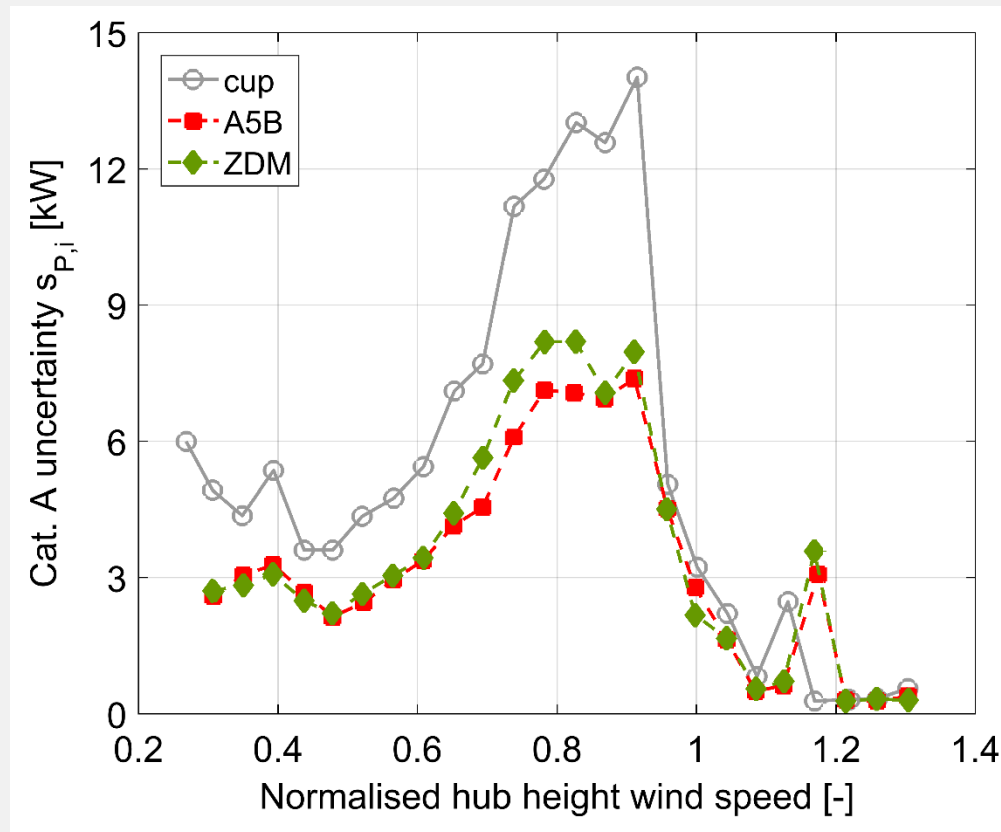




# Power curve uncertainties: power, type A

## WFR using wind-induction model

- **Clear reduction of scatter in power curve**  
→ nacelle lidars yield smaller type A (statistical) power uncertainty



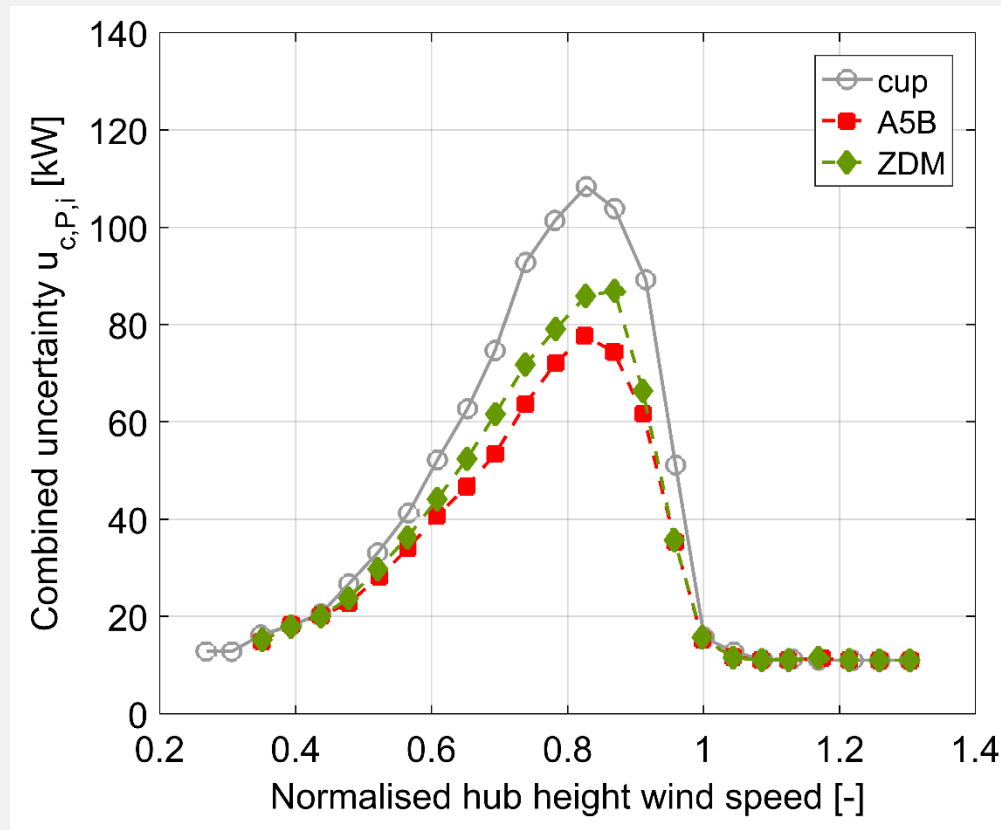
# Power curve uncertainties: combined

## WFR using wind-induction model

- Results are mostly dependent on type B wind speed uncertainty

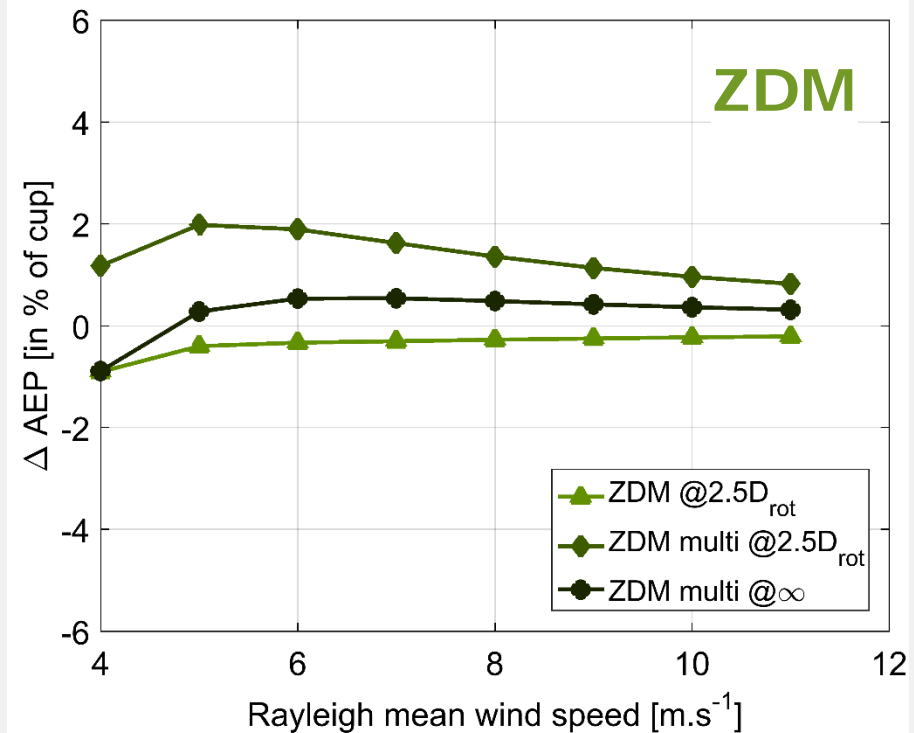
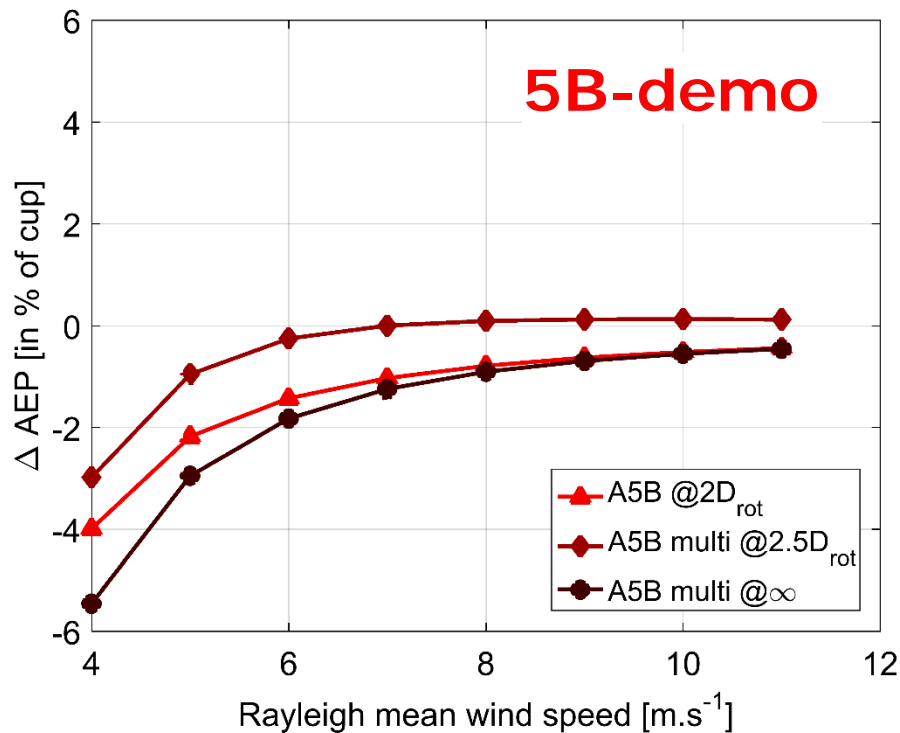
→ very sensitive to the "terrain uncertainty"

→ lidar uncertainties are smaller only due to this component...

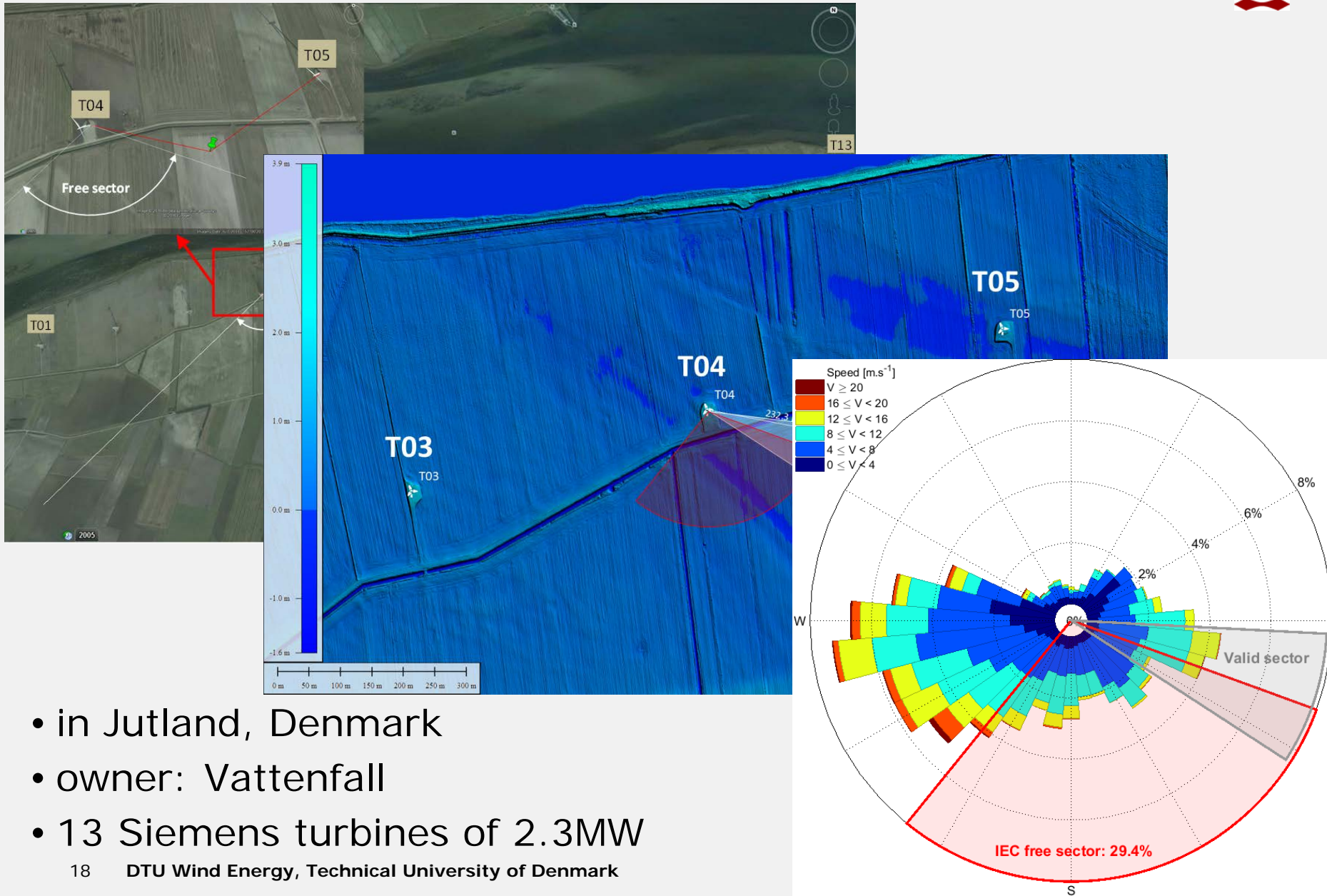


# Annual Energy production

- Derived as percentage of AEP using "mast power curve"
- 3 methods:
  - Wind model
  - Combined wind-induction
    - Wind speed estimated at  $2.5D$
    - fitted free stream wind speed ( $V_\infty$ )



# Full-scale campaign: Nørrekær Enge



- in Jutland, Denmark
- owner: Vattenfall
- 13 Siemens turbines of 2.3MW

# Wind speed results: summary table

Data filtering		Reconstruction case		Forced linear regressions results			
Case	Direction sector	Dataset	Lidar	Input measurement ranges	gain	$R^2$	Number of periods
1	[93°, 123°]	Joint	5B-Demo, 5 LOS	2.0 $D_{rot}$	1.0146	0.9936	885
			ZDM, 6 LOS	2.5 $D_{rot}$	1.0090	0.9938	
			5B-Demo, 5 LOS	from 0.5 to 1.15 $D_{rot}$	1.0063	0.9944	
			ZDM, 6 LOS	from 0.3 to 1.25 $D_{rot}$	0.9961	0.9947	

- Overestimation of 1-1.5% with the wind model
- Better performance of wind-induction model using the lidars' short-range measurements
- Lidar-to-lidar: 5B-Demo about 0.5-1% higher than ZDM

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2	[93°, 123°]	disjoint	5B-Demo, 5 LOS	2.0 $D_{rot}$	1.0133	0.9953	1476
			ZDM, 6 LOS	2.5 $D_{rot}$	1.0080	0.9942	2143
			5B-Demo, 5 LOS	from 0.5 to 1.15 $D_{rot}$	1.0057	0.9961	1123
			ZDM, 6 LOS	from 0.3 to 1.25 $D_{rot}$	0.9965	0.9962	2659

- Disjoint datasets: similar observations
- Increased number of valid data points (2-3x more)
- $R^2$  enhanced slightly

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3	[110°, 219°] (IEC free sector)	Joint	5B-Demo, 5 LOS	2.0 $D_{rot}$	1.0059	0.9848	2815
			ZDM, 6 LOS	2.5 $D_{rot}$	1.0028	0.9841	
			5B-Demo, 5 LOS	from 0.5 to 1.15 $D_{rot}$	0.9997	0.9877	
			ZDM, 6 LOS	from 0.3 to 1.25 $D_{rot}$	0.9923	0.9885	

- Better agreement between lidar and mast
- Much larger scatter (“signal decorrelation”)
- Still 5B-Demo above ZDM (about 0.5%)



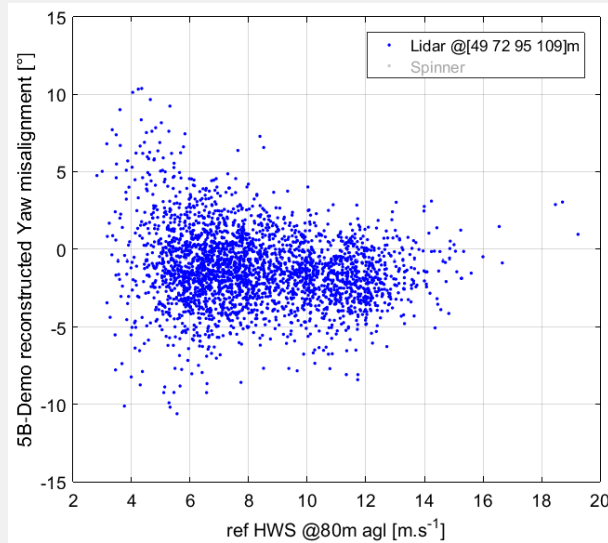
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			ZDM, 6 LOS	from 0.3 to 1.25 $D_{rot}$	0.9923	0.9885	
4	[110°, 219°] (IEC free sector)	disjoint	5B-Demo, 5 LOS	2.0 $D_{rot}$	1.0041	0.9840	4588
			ZDM, 6 LOS	2.5 $D_{rot}$	1.0038	0.9860	5615
			5B-Demo, 5 LOS	from 0.5 to 1.15 $D_{rot}$	0.9988	0.9888	4099
			ZDM, 6 LOS	from 0.3 to 1.25 $D_{rot}$	0.9935	0.9897	6199

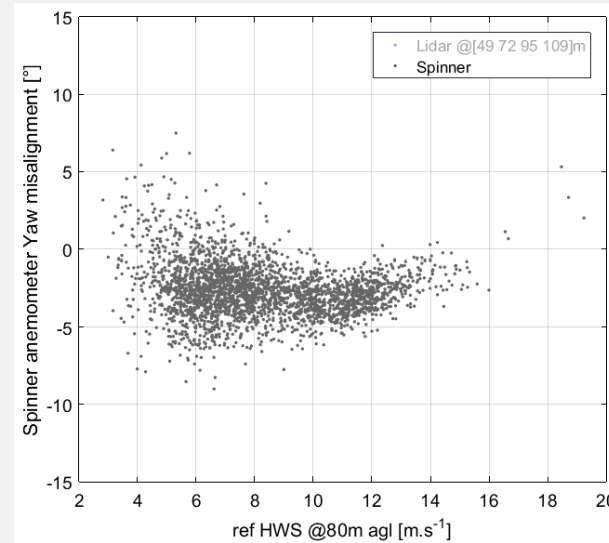
# Yaw misalignment results:

## WFR using the **wind-induction model**

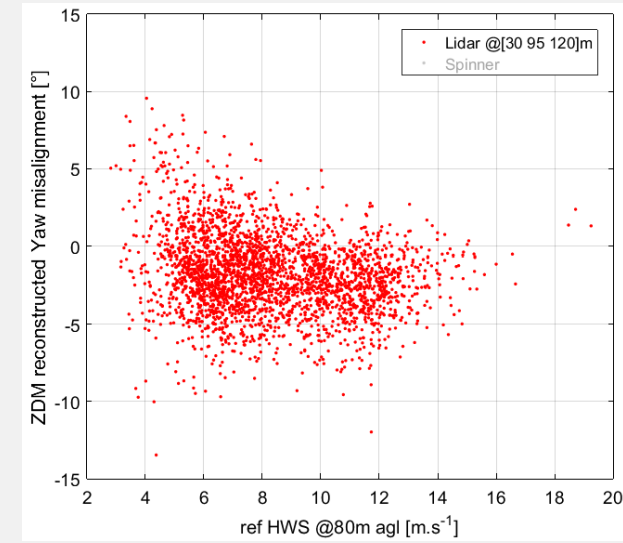
- Wind sector:  $[110^\circ, 219^\circ]$  (joint datasets)
- “Ref.” yaw misalignment from spinner anemometer



**5B-demo: 4 dist,  
from 0.5 to @1.2D\_rot**



**Spinner  
anemometer**



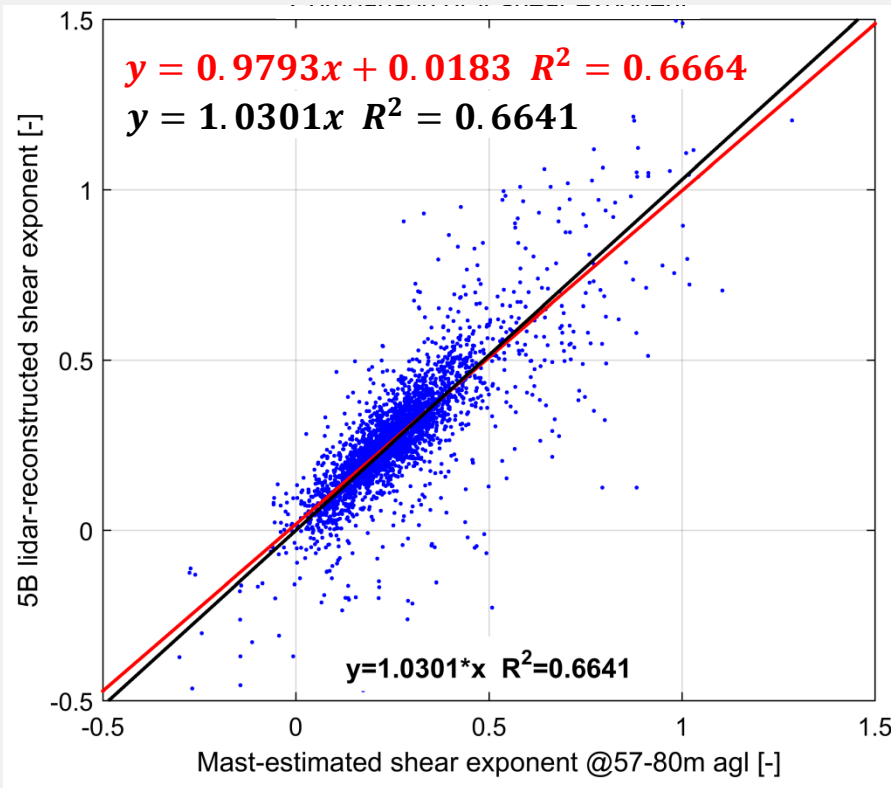
**ZDM: 3 dist.  
From 0.3 to 1.2D\_rot**

- ➔ Higher scatter with lidars than spinner
- ➔ “mean” yaw misalignment:  $\approx -3^\circ$
- ➔ The two nacelle lidars seem to provide similar results

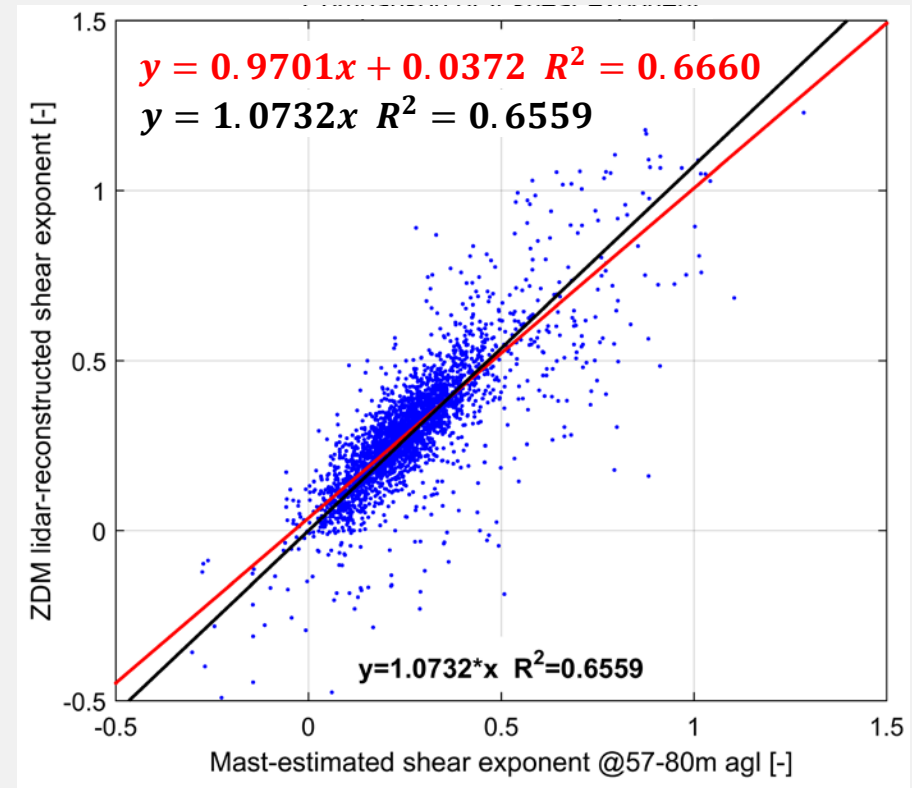
# Shear exponent results:

WFR using the **wind-induction model**

- Wind sector : [110°,219°] (joint datasets)
- “Ref.” shear exponent: from mast, using cups at 80 and 57m agl



**5B-demo: 4 dist,  
from 0.5 to @1.2D\_rot**



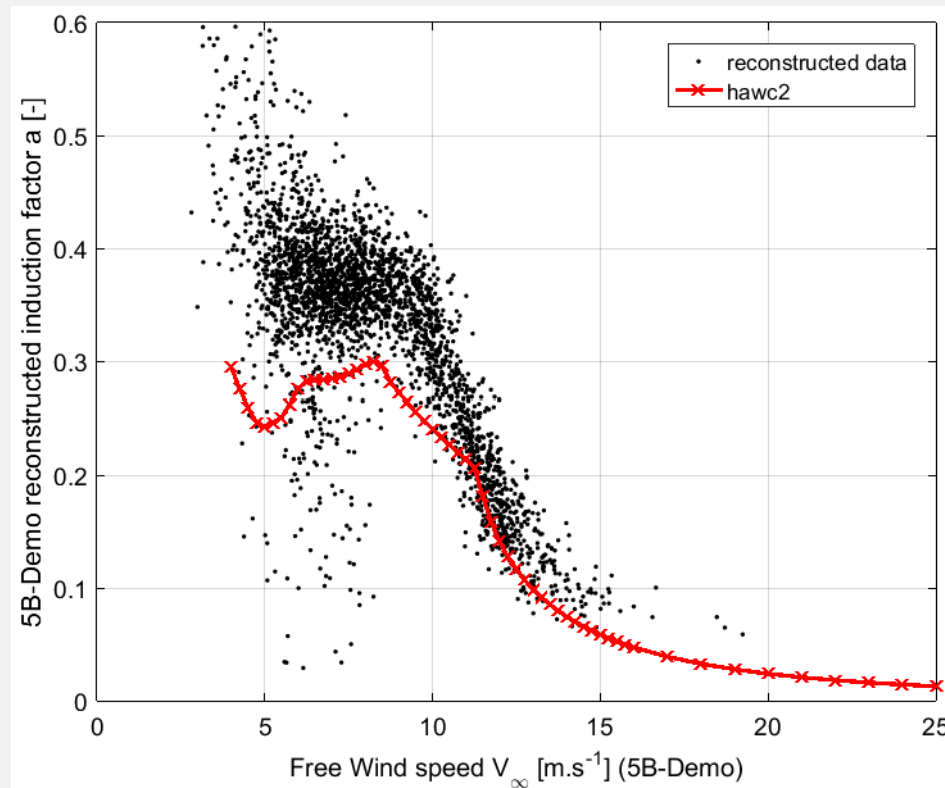
**ZDM: 3 dist.  
From 0.3 to 1.2D\_rot**

➔ Slight overestimation vs. mast ➔ Similar results between the two lidars

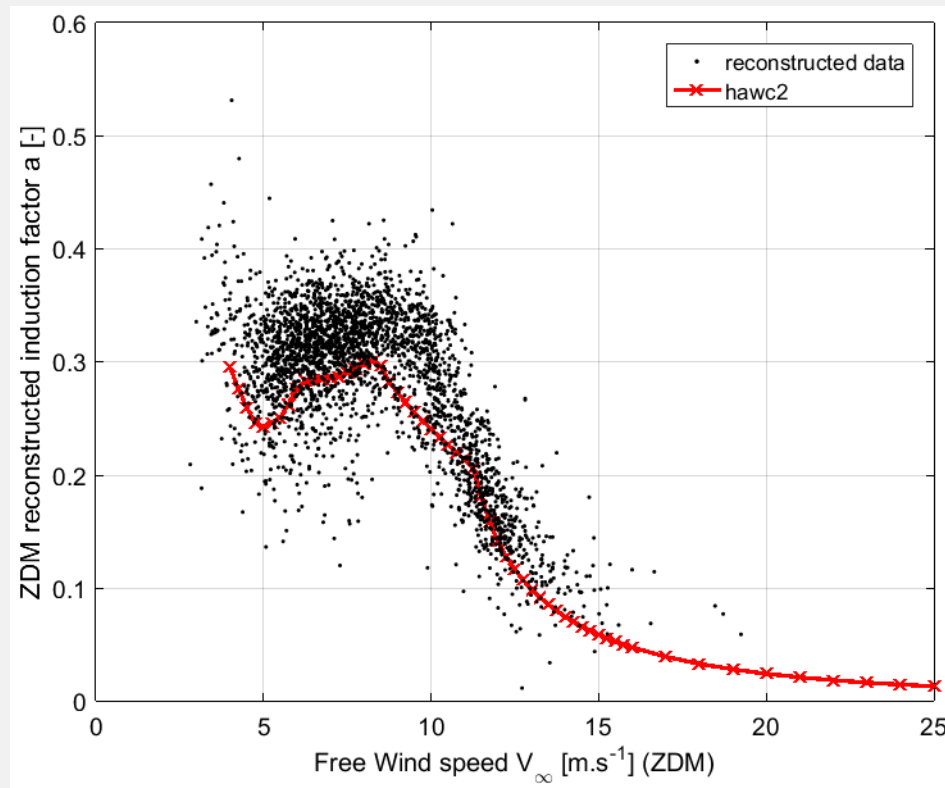
# Induction factor results:

## WFR using the wind-induction model

- Wind sector : [110°, 219°] (joint datasets)
- "Ref." induction factor:  $C_T$  from "HAWC2" simu,  $a = 0.5 \cdot (1 - \sqrt{1 - C_T})$

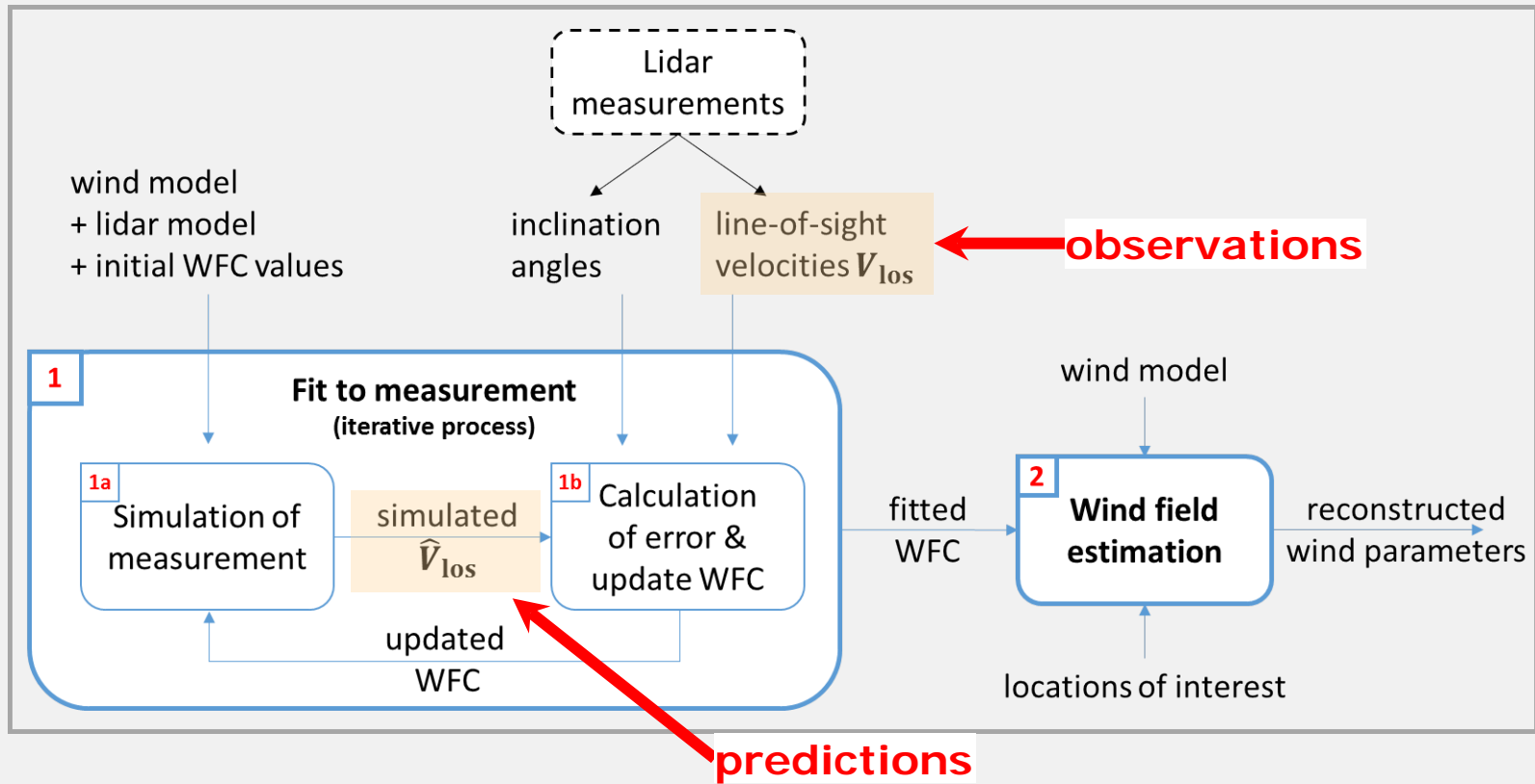


**5B-demo: 4 dist,  
from 0.5 to @1.2D\_rot**



**ZDM: 3 dist.  
From 0.3 to 1.2D\_rot**

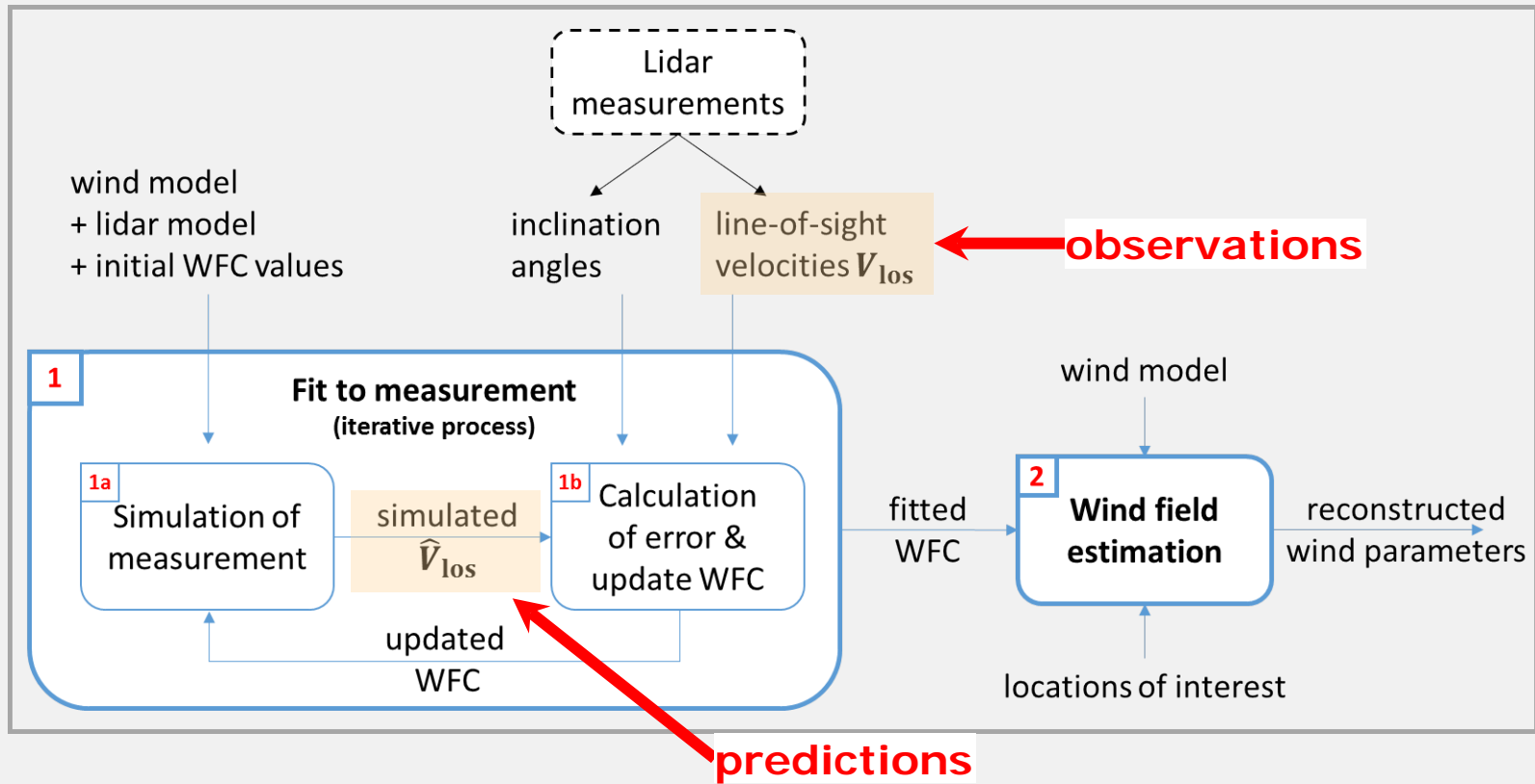
# LOS velocity fitting residuals



## • Definitions:

- $V_{los}$  and  $\hat{V}_{los}$  are column vectors of length = N meas. points (e.g. 5B-Demo = 4 dist\*5 los = 20; ZDM = 3 dist\*6 los = 18)
- "bias" =  $V_{los} - \hat{V}_{los}$  ; "error": =  $abs(V_{los} - \hat{V}_{los})$

# LOS velocity fitting residuals



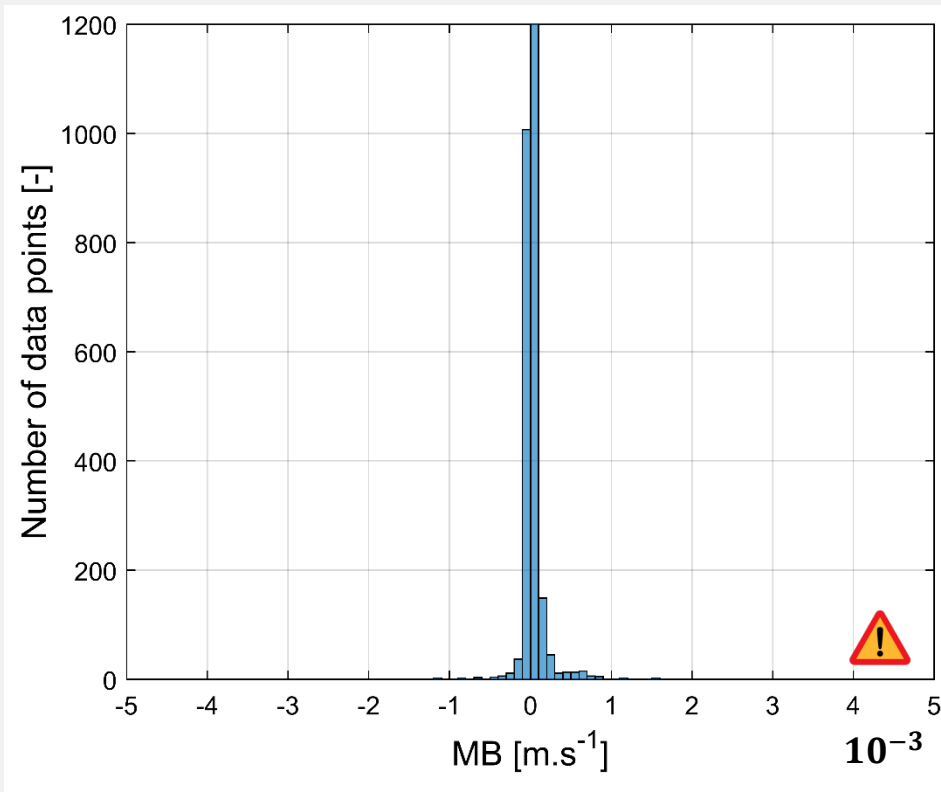
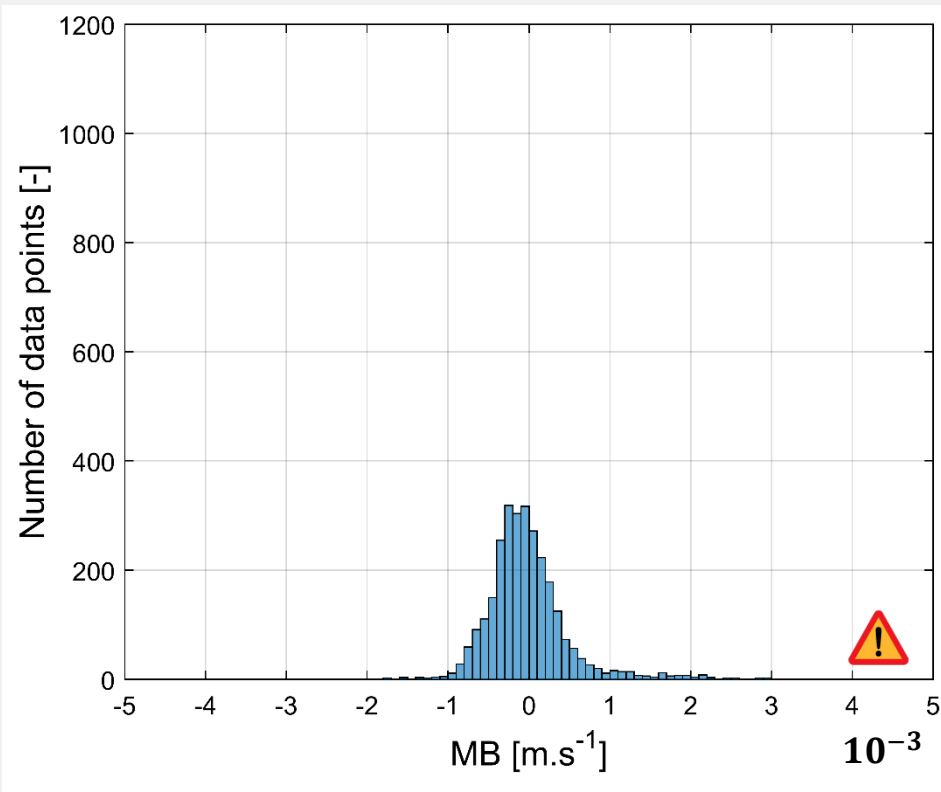
## • Computed stats:

- M: mean, N: normalised; F: fractional;
- S: squared; R: root; SS: sum of squares
- **MB**, ME, NMB, NME, MFB, MFE, SSE, MSE, **RMSE**, NMSE

# V<sub>los</sub> fitting residuals: mean bias

WFR using the **wind-induction model**

- Wind sector : [110°, 219°] (joint datasets)



**5B-demo**

**4 dist. from 0.5 to @1.2D<sub>rot</sub>**

➔ MB show very low values;

➔ Histogram centered on zero: the used model is “unbiased”

**ZDM**

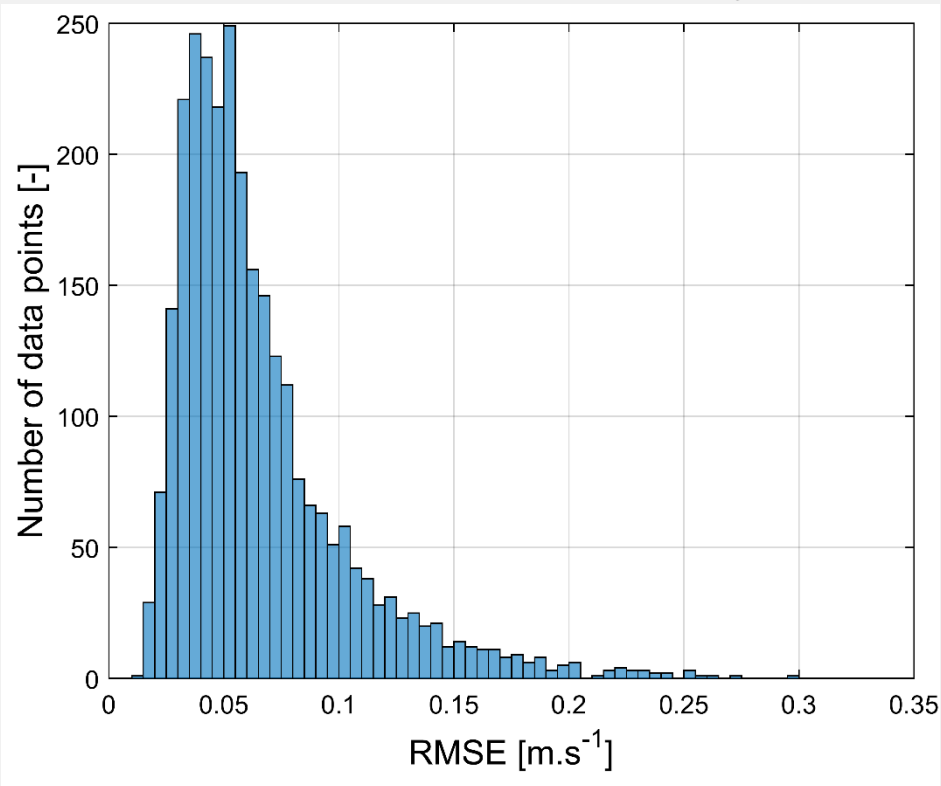
**3 dist. from 0.3 to 1.2D<sub>rot</sub>**



# V<sub>los</sub> fitting residuals: mean bias

## WFR using the wind-induction model

- Wind sector : [110°, 219°] (joint datasets)

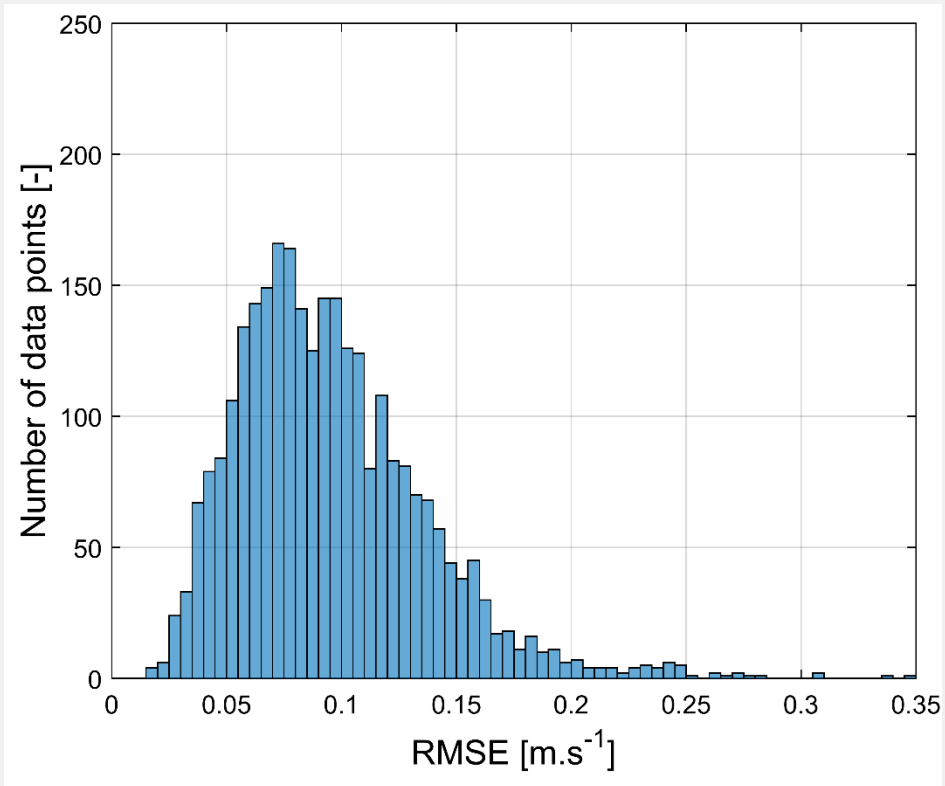


**5B-demo**

**4 dist. from 0.5 to @1.2D<sub>rot</sub>**

➔ RMSE values between 0 and 0.25 m/s

➔ Similar distributions for both lidars, with a slightly larger mean for ZDM



**ZDM**

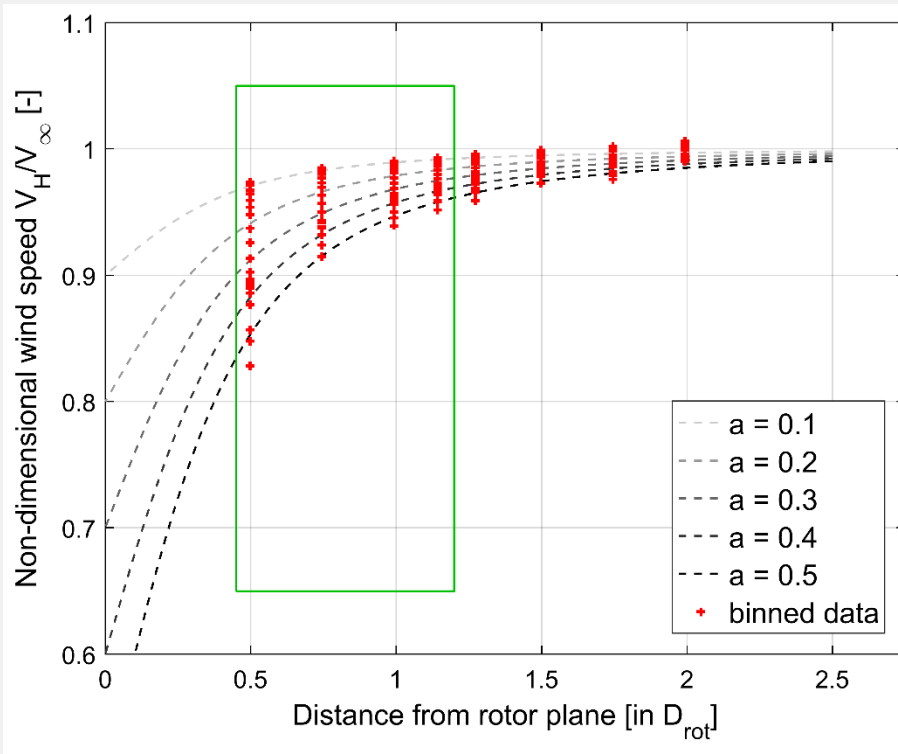
**3 dist. from 0.3 to 1.2D<sub>rot</sub>**

# A simple induction model

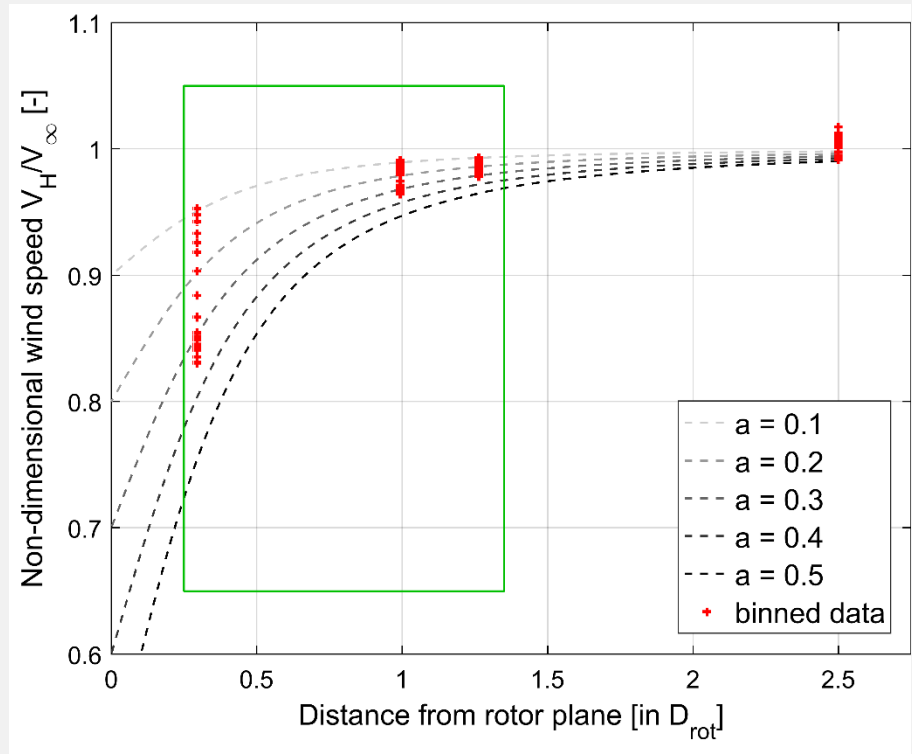
- **Derived from the Biot-Savart law**

- see *The upstream flow of a wind turbine: blockage effect*
- two parameters: induction factor  $a$ , free wind speed  $U_\infty$

$$\frac{U}{U_\infty} = 1 - a \left( 1 + \frac{\xi}{\sqrt{1+\xi^2}} \right), \text{ with } \xi = \frac{x_W}{R_{rot}}$$



**5B-demo**



**ZDM**

# Simple induction models

- One- or two- dimensional?

