



Rogier Floors & Bjarke Olsen

Using mesoscale and microscale models for wind resource assessment

DTU

Introduction

- For wind resource assessment the wind distribution has to be known
- Mesoscale models are frequently used for assessing wind resources
- Typical resolutions are 1-3 km, higher resolution very expensive for multiyear runs.
- This is typically not enough to resolve microscale flow features, particularly over land
- Here we show:
 - Illustrations of microscale effects on the flow near a coastal site
 - A tool chain that couples these models and a validation



Sanz Rodrigo, J., Chávez Arroyo, R.A., Moriarty, P., Churchfield, M., Kosović, B., Réthoré, P.-E., Hansen, K.S., Hahmann, A., Mirocha, J.D. and Rife, D. (2017), Mesoscale to microscale wind farm flow modeling and evaluation. WIREs Energy Environ, 6: e214. <u>https://doi.org/10.1002/wene.214</u>







Normalized wind speed at 50 m agl

1,3

0,7

WRF setup NEWA

- WRF 3.8.1
- CORINE land cover
- SRTM elevation
- MYNN PBL scheme
- Boundary conditions ERA5
- 3 x 3 km horizontal res
- 61 vertical levels
- Ref: Hahmann et al (2020)

Two meteorological masts

- One mast 241 m north of row and one 241 m mast in the southerly end
- Equipped with cup and sonic anemometers





Balcony experiment NEWA

- 2 Scanning lidars installed at 50 m above the ground
- Scanning using 45 line-of-sight in 90 degree plain
- Range gates varying from 105 to 7000 m
- 1 scan of full plain in 45 sec
- 2016-04-12 to 2016-06-17, 66 days (1584 hours)
- Wind speed and direction reconstructed on a ~7x7 km area with 100 m resolution
 Ref: Karagali et al. (2018)





Normalized wind speed from wind lidars at 50 m agl



1,3

0,7

~4.5 km

235

25 25 (12.5

15

35 225 3

12.5

15

ST5

Elevation contours

Danmarks Højdemodel (lidar based vector data) Ref: QGIS WAsP plugin (2022)

03 7.5



Normalized wind speed from wind lidars at 50 m agl

15 15 (12.5

35 CHAS (5)

125

00



1,3

05 07.5

Roughness contours

CORINE 2018 (vector data)
Ref: QGIS WAsP plugin (2022)



Normalized wind speed from wind lidars at 50 m agl

0.4 0.03

1.2 0.05

0.05

0.4

0.03

0.03 1.2



1,3



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Normalized wind speed at 50 m agl using microscale model WAsP





Climatological behaviour of the wind

 Averaged over a sufficiently long time, for a particular wind direction sector near always resembles a Weibull distribution

$$f_u(u) = \frac{k}{A} \left(\frac{u}{A}\right)^{k-1} \exp\left(-\left(\frac{u}{A}\right)^k\right)$$

- A is the scale parameter
- K is the shape parameter

The higher the k, the more narrow the distribution is.



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Transforming wind distribution with height and roughness





Validating the NEWA downscaling method



Comparison of:

- 1. NEWA microscale model chain (Hahmann, 2020)
- 2. New model chain developed based on PyWAsP and using stability data from mesoscale model (Floors, 2023)

- 26 masts with time series were selected within the NEWA Central Europe domain
- Elevation data NASADEM 30 m resolution
- CORINE vector data for roughness maps



Power density downscaled WRF vs. obs.

- Plain comparison between measurement campaigns (typically 1-5 yrs)
- The WRF simulations are based on 30 yrs, so year-toyear variability might contribute to errors.





Power density downscaled WRF + LTC vs. obs.

 Measurements corrected to same 30 yr period using variance ratio method





Conclusions

- Microscale effects contribute to +- 30% of variation in mean wind speed at 50 m over a relatively flat forested coastal site
- Microscale model was able to capture areas of most speed-up and speed-down
- Incoorporating a newly developed stability model improved the power density predictions of a model chain involving WRF and WAsP.
- Year-to-year variations explained part of the mismatch between model results and measurements.

Next steps:

- Checking model chain for 'correct' periods without LCT
- Look at more micrometeorological 'complex' sites



References

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