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ISGAN – International Smart Grid Action Network

Matthias Lange, Quentin Libois, Remco Verzijlbergh

23 March 2021

ISGAN Academy webinar #27 Recorded webinars available at: <u>https://www.iea-isgan.org/our-work/annex-8/</u>





ISGAN in a Nutshell

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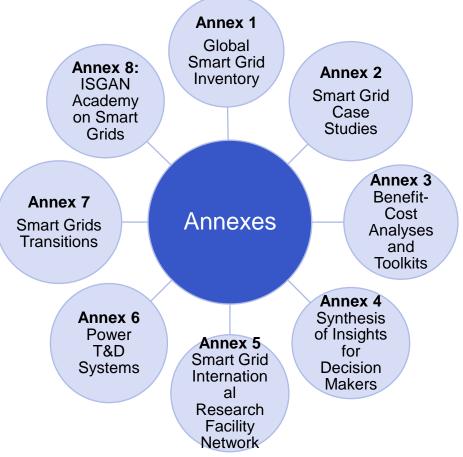


the Implementing Agreement for a Co-operative Programme on Smart Grids

an initiative of the Clean Energy Ministerial (CEM)

Strategic platform to support high-level government knowledge transfer and action for the accelerated development and deployment of smarter, cleaner electricity grids around the world ISGAN INTERNATIONAL SMART GRID ACTION NETWORK

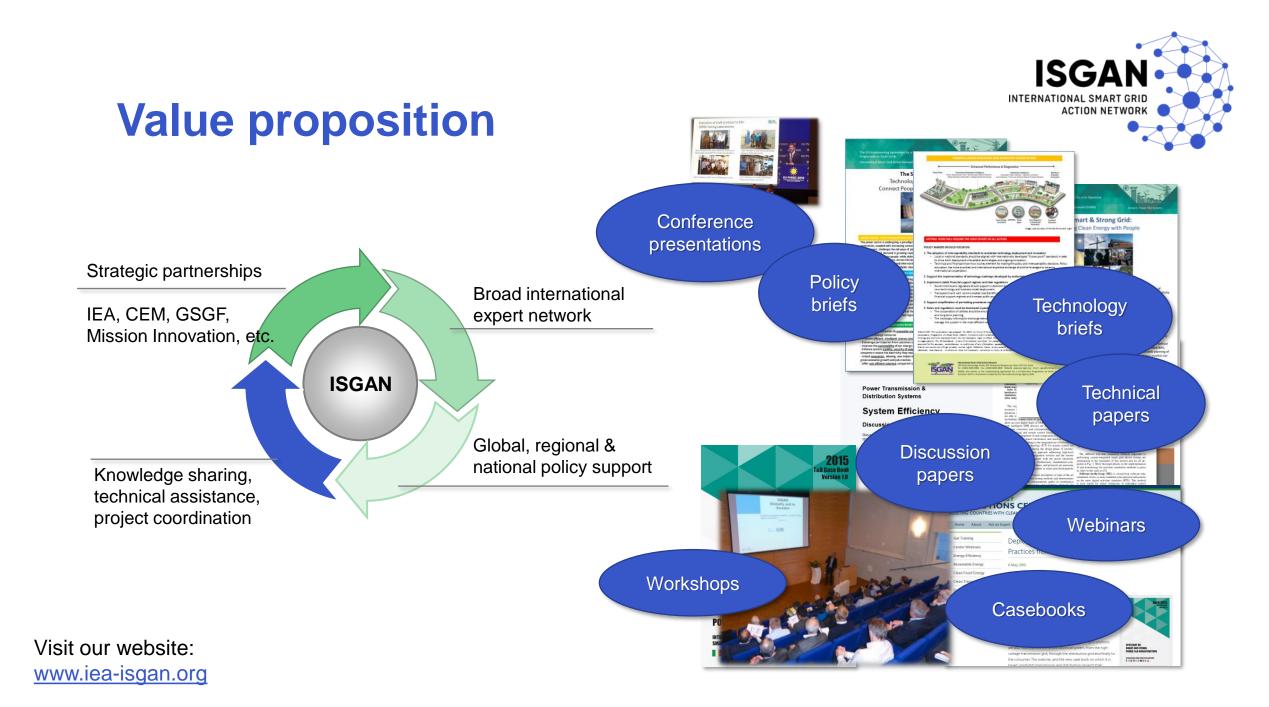
International Smart Grid Action Network is the only global government-togovernment forum on smart grids.





ISGAN's worldwide presence





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Advanced weather forecasting for RES applications

Smart4RES developments towards high-resolution and Numerical Weather Prediction solutions to improve RES forecasting models

Matthias Lange, Quentin Libois, Remco Verzijlbergh

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Agenda

- Smart4RES in a nutshell
- Challenges in forecasting of renewables
- Towards Numerical Weather Prediction models dedicated to Renewables Energy Sources
- High-resolution weather models



Smart4RES in a nutshell

Smart4RES in a nutshell

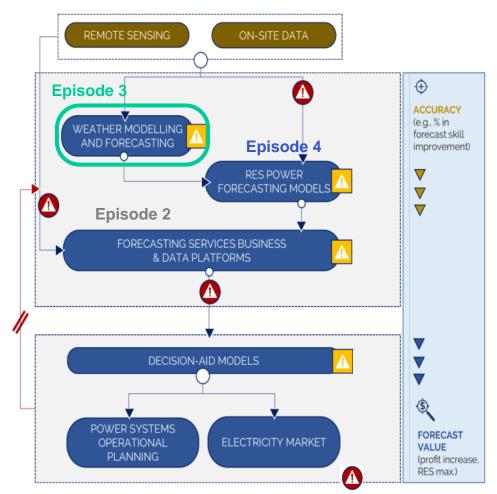




- RES forecasting is a mature technology with operational tools and commercial services used by different actors
- However, we want to make progress to improve the forecasting accuracy and to reduce costs of RES integration

Smart4RES vision

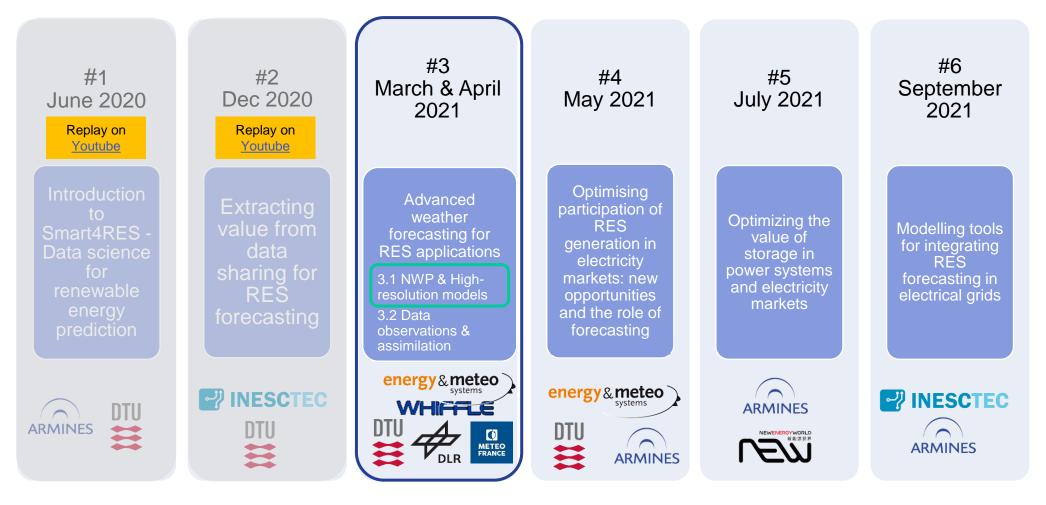
Science and industry closely co-operate to achieve outstanding improvements of RES forecasting by considering the whole model and value chain.





Smart4RES webinar series

Season1: Towards a new Standard for the entire RES forecasting value chain





Smart4RES consortium





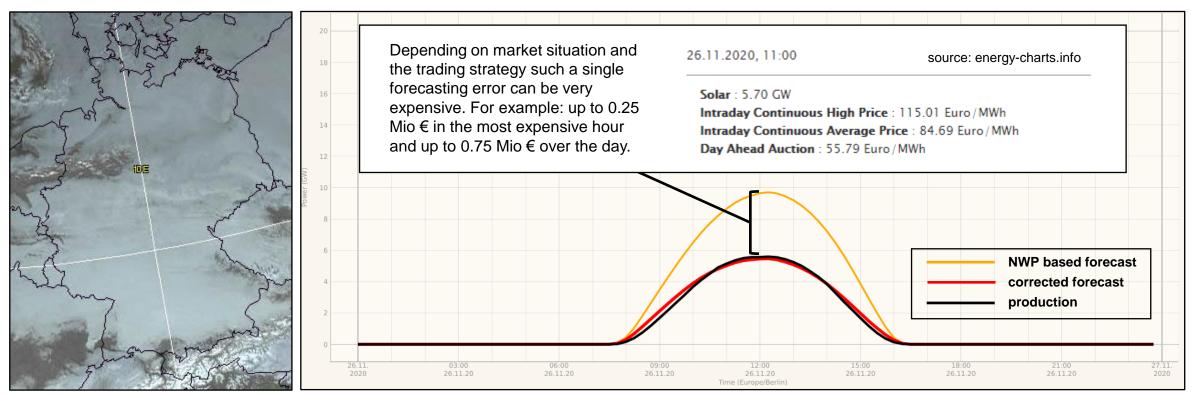
Challenges in forecasting of renewables

Challenges in forecasting of renewables

• Solar power prediction: the fog situation

fog seen from satellite

impact of fog on solar power forecast for Germany: individual weather models with large forecasting errors



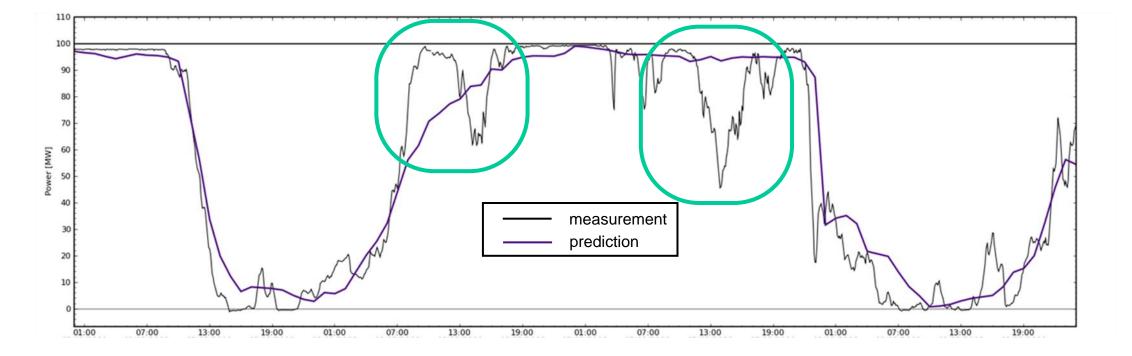




Challenges in forecasting of renewables

• Wind power prediction: fluctuations and ramps

State-of-the-art weather models provide a time resolution of 1 hour and do not predict fast changes





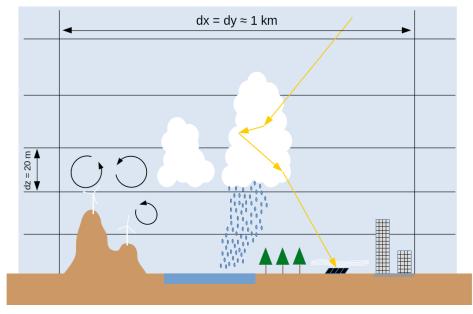
Towards Numerical Weather Prediction (NWP) models dedicated to Renewable Energy Sources (RES)

Innovative weather forecasts for RES – Challenges and Solutions



Why is weather forecasting so difficult ?

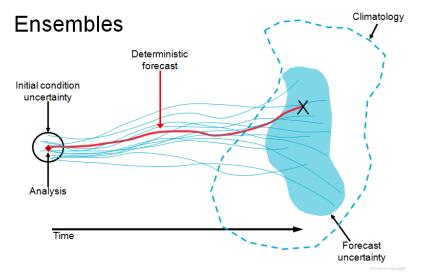
- atmosphere is **chaotic** (*initial state is critical*)
- NWP models have coarse resolution while physical processes occur at very small scales (clouds, rain, wind gusts, etc.)
 → Parameterizations needed



Many physical processes occur at scales smaller than the model grid, and are not explicitly simulated. They are parameterized based on explicitly simulated large scale variables

What are the solutions ?

- using **ensembles** to explore all possible states
- increasing spatial resolution to get rid of parameterizations
- improving parameterizations



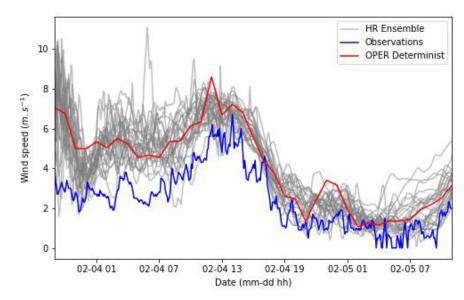
Schematic illustration of the use of ensembles in weather forecasts (UK Met Office)

Towards NWP models dedicated to RES

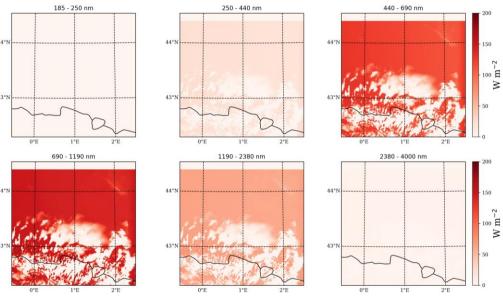


Paths to improved RES forecasts

- providing higher temporal resolution outputs (NWP models have a ~1 min time step but standard output is 1 hour)
- extracting additional relevant variables for RES (cloud optical thickness, direct/diffuse partition, spectral distribution of radiation, rapid wind fluctuations)
- evaluating and calibrating NWP models accounting for RES scores and developing specific forecasters capabilities



Comparison between 1-hour (operational) and 5-min resolution outputs for wind speed forecasts *(courtesy B. Alonzo)*

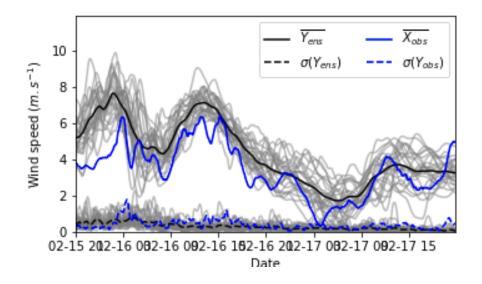


Direct solar radiation in 6 spectral bands simulated by AROME around Toulouse (France)

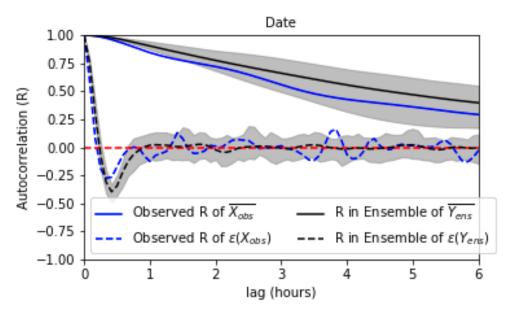
High temporal resolution outputs



- High frequency fluctuations in simulated wind speed are consistent with observations
- Slight underestimation of these fast fluctuations in the ensemble mean but some members do capture them better
- Autocorrelation demonstrates the forecast capability



Low frequency variations and intra-hourly standard deviation of wind speed. Comparison between ensemble simulations and observations (courtesy B. Alonzo)

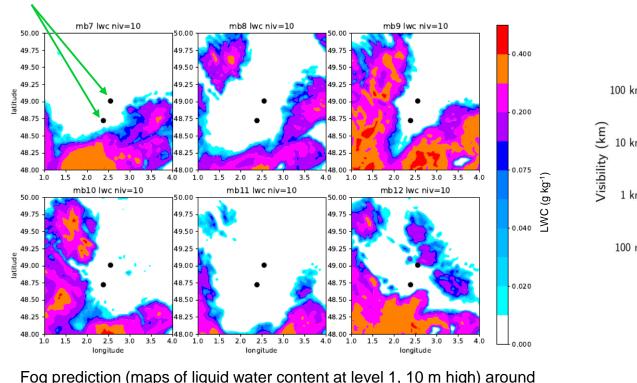


Autocorrelation functions of low and high frequency wind fluctuations *(courtesy B. Alonzo)*

The power of ensembles – fog



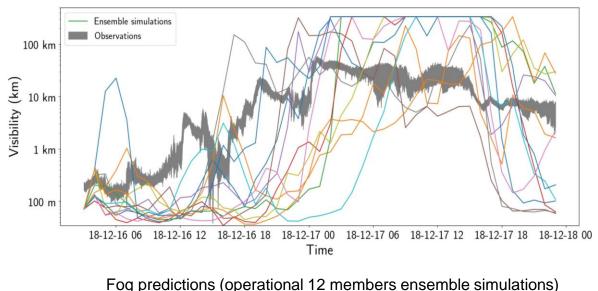
Ensembles (perturbed initial conditions, lateral forcing, surface or parameterizations) highlight how the atmospheric state can vary a lot from a member to another



Paris for 6 members of an ensemble (courtesy T. Bergot)

Paris airports

Different members can have very distinct behaviours

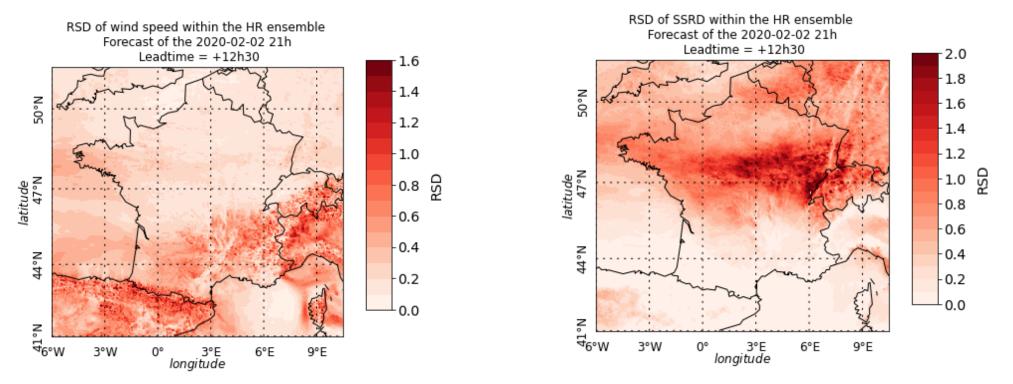


Fog predictions (operational 12 members ensemble simulations) and observations (12 measurements) at CdG airport (Paris) in December 2018 *(courtesy T. Bergot, R. Lestringant)*

The power of ensembles – wind, solar



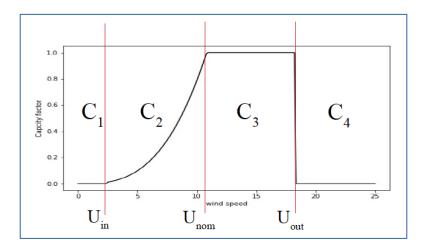
- Ensembles are useful to estimate the uncertainty associated with a forecast
- Uncertainties in RES relevant variables can be significant



Relative standard deviation (standard deviation/mean) for wind (left) and global horizontal irradiance (right) across an ensemble of 25 members (*courtesy B. Alonzo*)

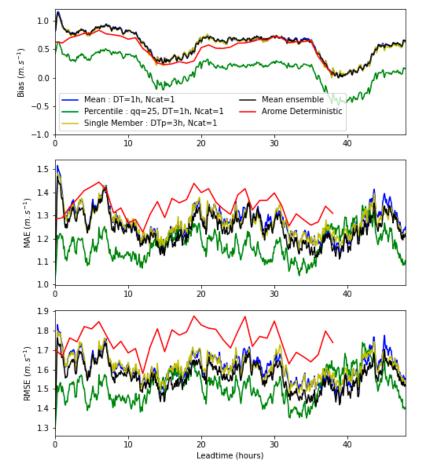
The power of ensembles – Pseudo-deterministic simulations

- Ensembles can be used to build pseudo-deterministic simulations (that look like deterministic simulations to the enduser)
- These pseudo-deterministic simulations can be built in various ways, and can be very specific to the end-user needs
- Pseudo-deterministic simulations can exceed ensemble mean forecasting performances



Definition of categories for wind speed, used to build pseudodeterministic runs for wind power forecasts. Over each period, the most represented category across the ensemble is used.



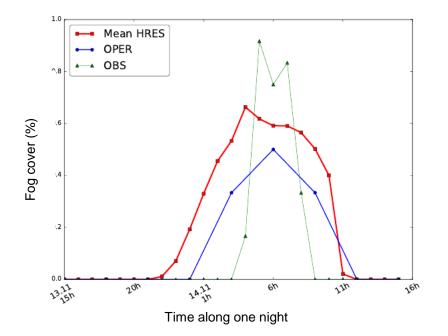


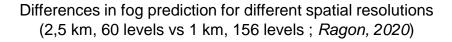
Performances of various simulations with respect to observations. Here, the dedicated pseudo-deterministic simulation performs best (*courtesy B. Alonzo*)

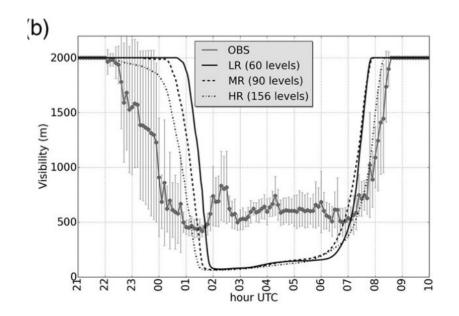
Increasing spatial resolution



- Fog formation is very sensitive to the state of the atmosphere close to the surface
- Increasing vertical and horizontal resolution modifies turbulence and dynamics







Higher vertical resolution better reproduces fog daily cycle near Paris airport (*Philip et al., 2016*)



High-resolution weather models Large Eddy Simulation (LES): the future

Operational LES model of the HornsRev offshore wind farm





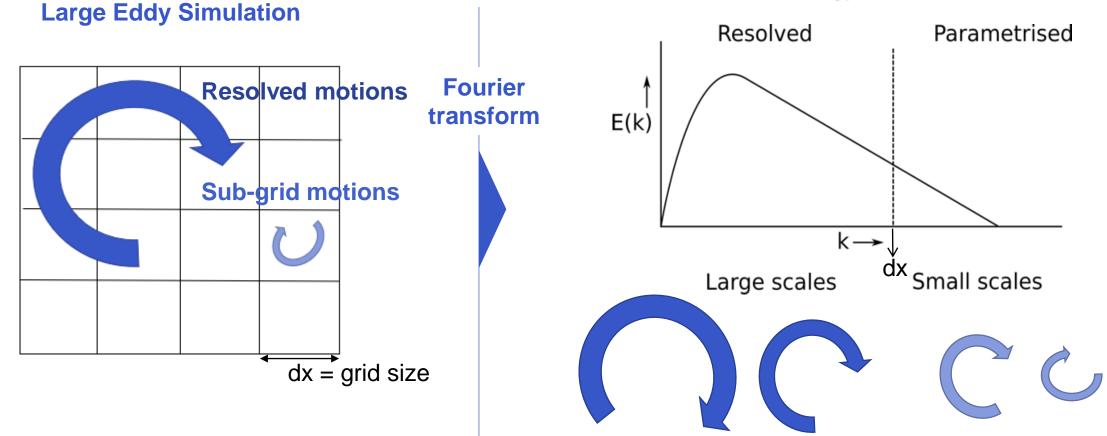
https://vimeo.com/whiffle/hornsrev

Gilbert, C., Messner, J. W., Pinson, P., Trombe, P., Verzijlbergh, R., Dorp, P. Van, & Jonker, H. (2019). Statistical Postprocessing of Turbulence-resolving Weather Forecasts for Offshore Wind Power Forecasting. *Wind Energy*, 1–16. https://doi.org/10.1002/we.2456

The energy spectrum of turbulent flow in Large Eddy Simulation (LES)



A turbulent spectrum: energy as function of wavenumber

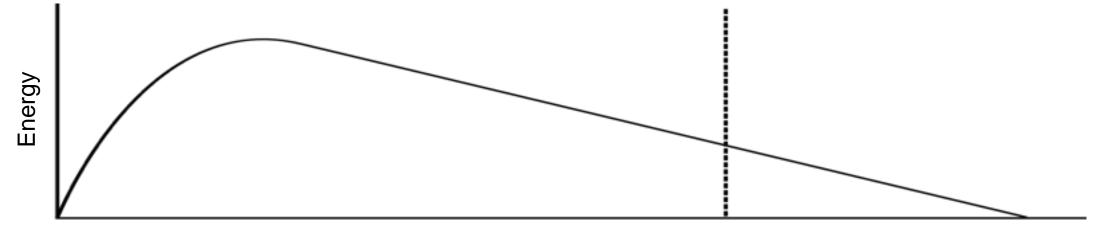


Model grid in *physical space* showing schematically the turbulent motions (eddies)

In *spectral space* the energy of different eddies peaks around a certain size (small wavenumber *k* are the largest eddies)

The energy spectrum of turbulent flow in Large Eddy Simulation (LES)





Wavenumber k



Flow dependent

Flow independent

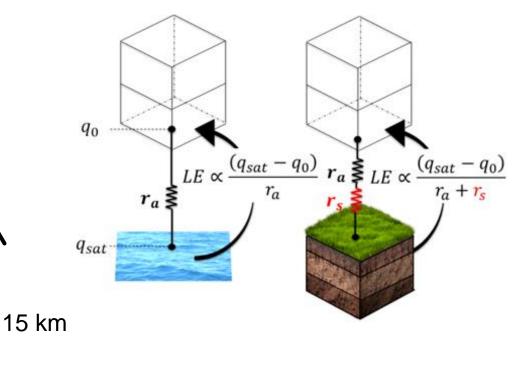
LES resolves all relevant physical processes and only parametrizes homogeneous turbulence

High-resolution allows to resolve the surface in detail

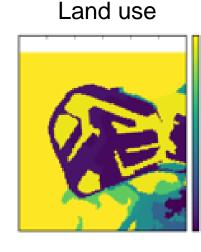


- Forest and vegetation resolved in more detail (canopy parameterization)
- Explicitly model obstacles like buildings and wind turbines

Surface scheme controls momentum, heat and moisture exchange



Tiggelen, M. Van. (2018). Towards improving the land- surfaceatmosphere coupling in the Dutch Atmospheric Large- Eddy Simulation model (DALES). MSc Thesis, Delft University of Technology.



Terrain and buildings

107.31

71.5

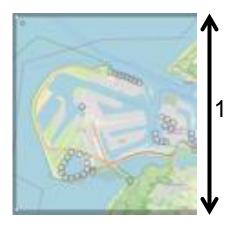
13.6

17.8

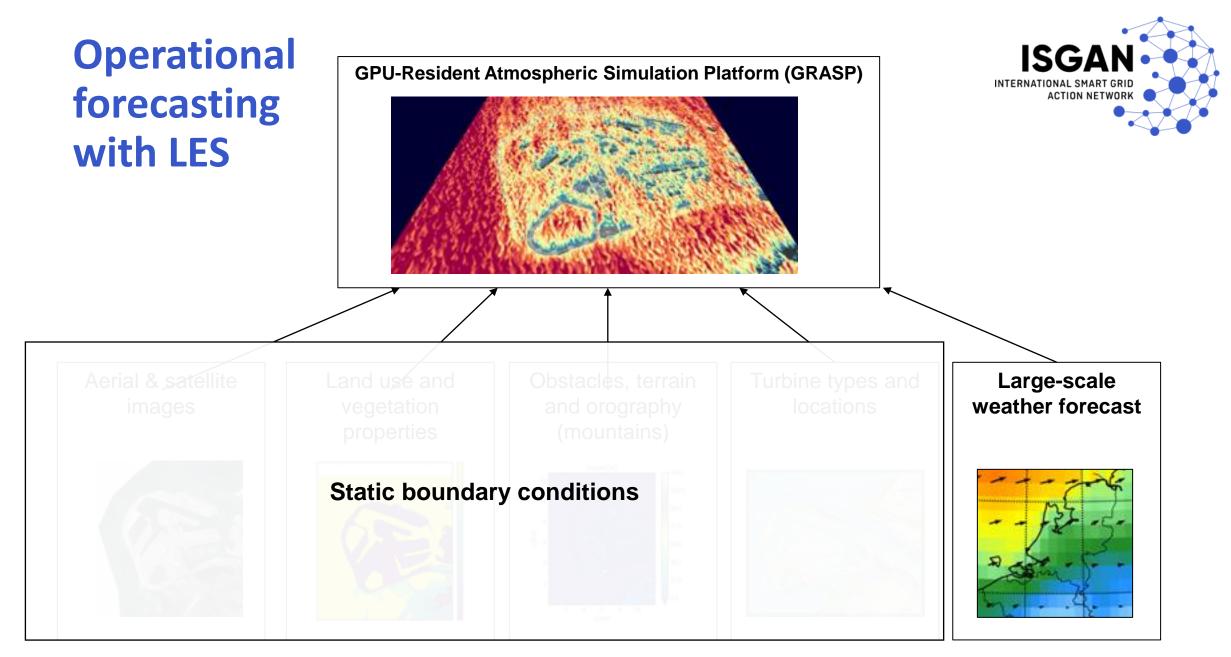
Height (m)

a (ked)







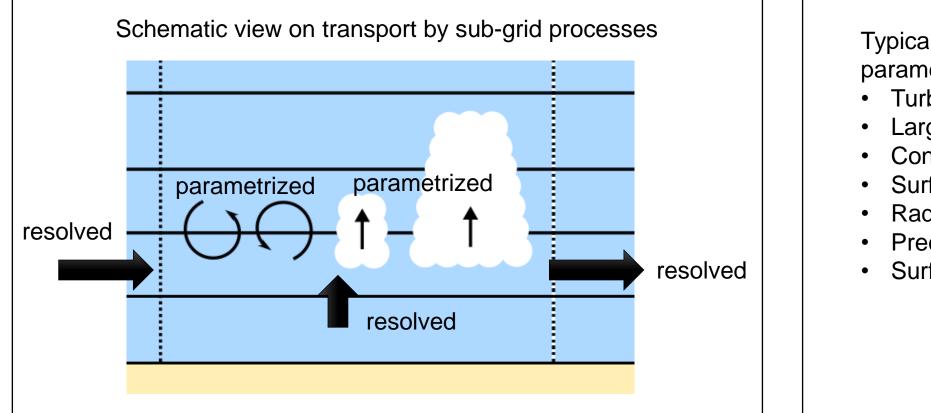


A schematic representation of different inputs to a the LES forecasting model

Physics parameterizations in NWP



Parametrization: expressing the sub-grid processes in terms of resolved quantities



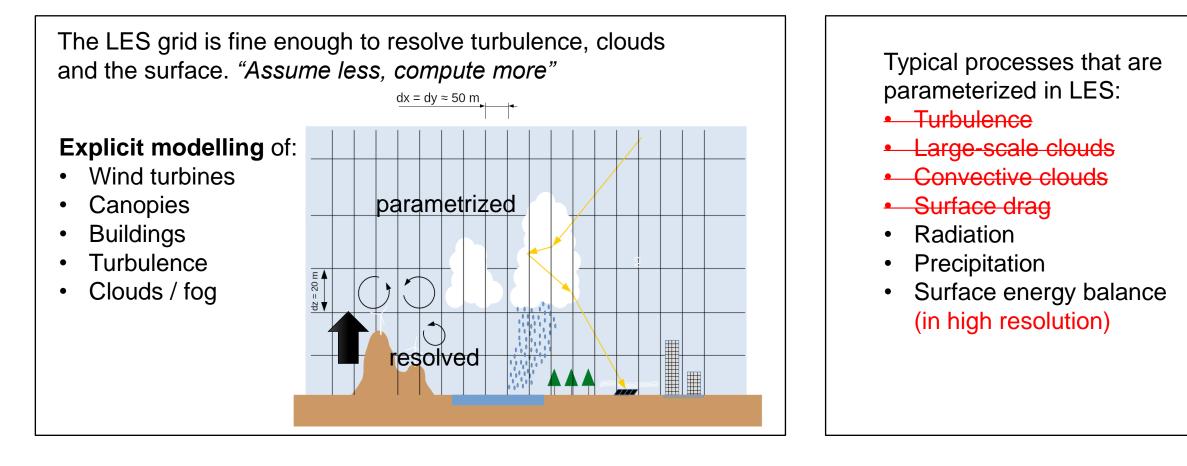
Typical processes that are parameterized in NWP:

- Turbulence
- Large-scale clouds
- Convective clouds
- Surface drag
- Radiation
- Precipitation
- Surface energy balance

Physics parameterizations in LES



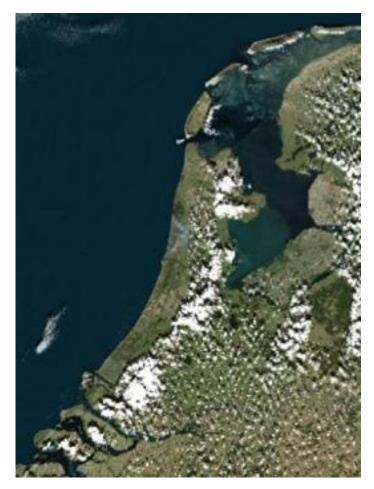
Parametrization: expressing the sub-grid processes in terms of resolved quantities



Country scale LES already possible on multi-GPU systems

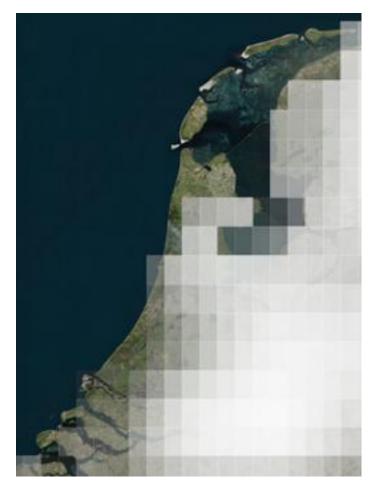


Satellite image





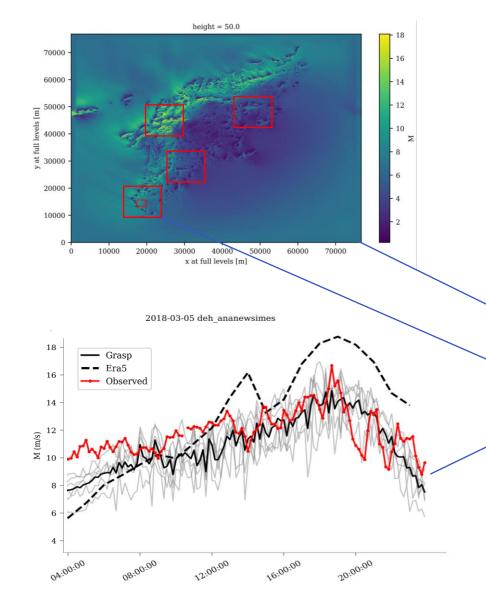
Traditional weather forecast





Smart4RES use case: Rhodes





Important forecast characteristics for smarter grid management:

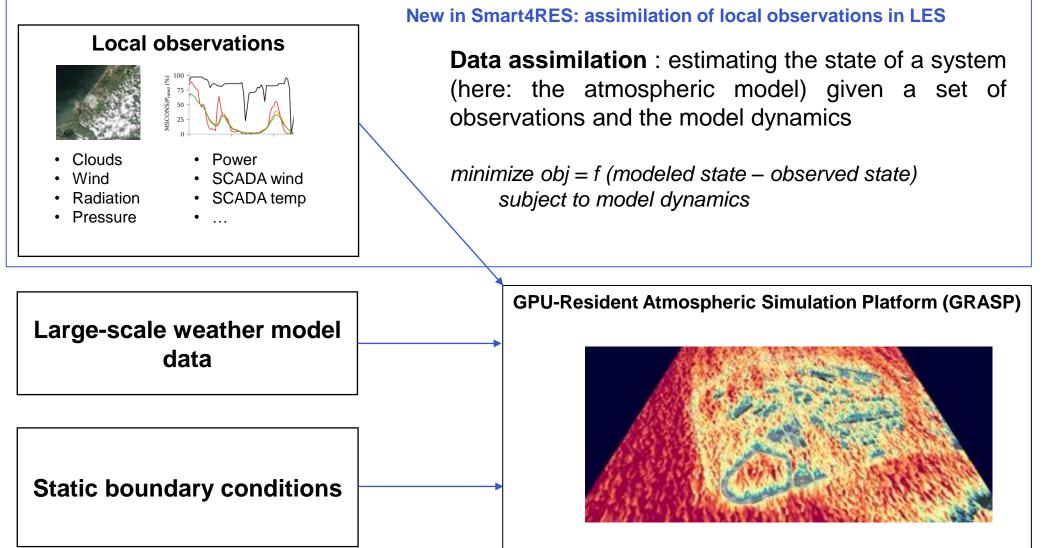
- Fast fluctuations in wind power
- Spatio-temporal patterns
- Good forecast accuracy on complex terrain

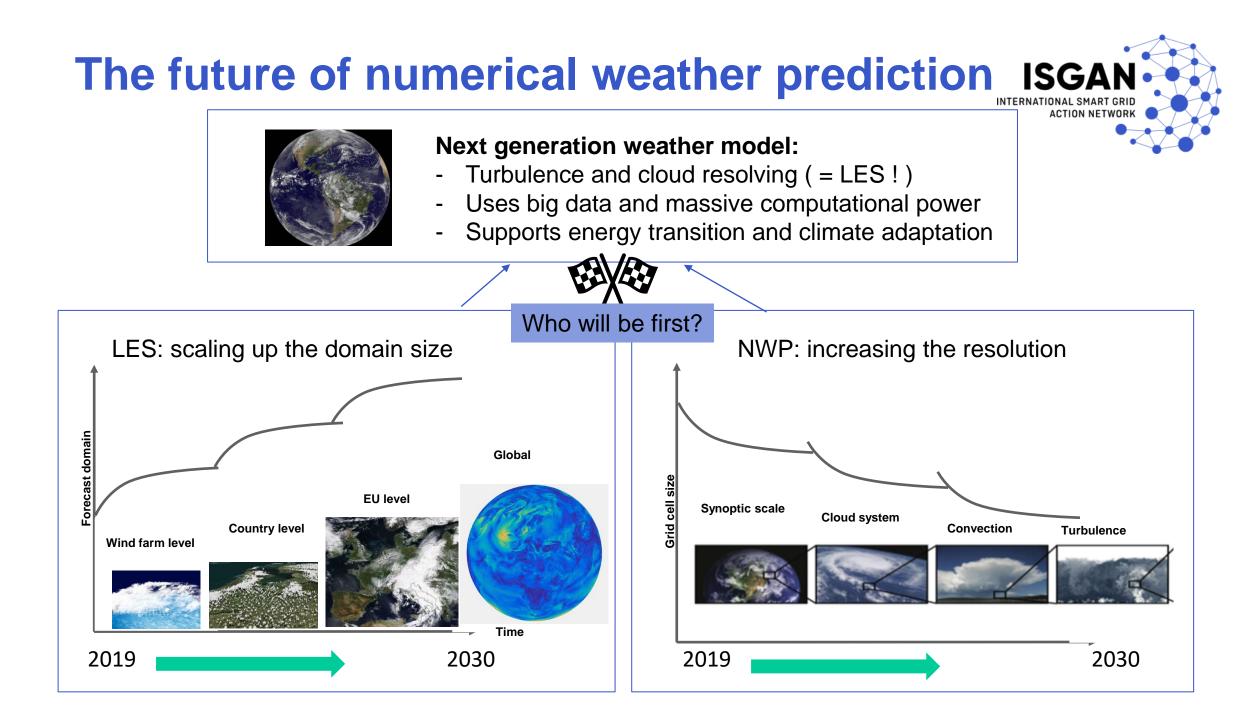
A nested LES forecast is produced

- Outer domain captures Rhodes ~ 100m resolution
- 4 inner domains on wind farms: ~ 40m resolution
- Figure shows variability in individual wind turbine forecasts
 - Compare with smooth signal of large-scale weather model (era5)

Planned innovations in Smart4RES









Take-away messages

- Numerical models contain much more than operational outputs
- The natural evolution of NWP is to increase spatial resolution

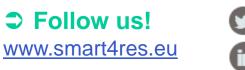
 \rightarrow Now at a stage called **grey-zone turbulence** in between parameterizations and LES (~ 500 m resolution)

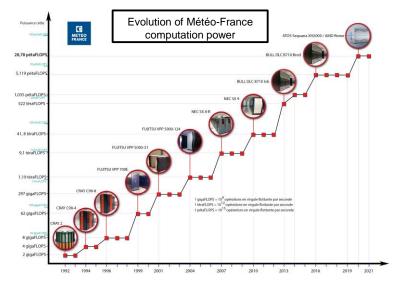
- \rightarrow LES might be the future of NWP
- Ensembles are becoming the norm to explore uncertainties
- Increasing amount of observations, including non-conventional, are likely to improve forecasts via data assimilation

Attend our next webinar!

→ for very short-term, direct observations remain very powerful because models give a statistical view of the atmosphere, not perfectly punctual in space and time

 \rightarrow combination of various inputs is promising, in particular with increased use of artificial intelligence





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Further reading

[1] Gilbert, C., Messner, J. W., Pinson, P., Trombe, P. J., Verzijlbergh, R., van Dorp, P., & Jonker, H. (2020). Statistical post-processing of turbulence-resolving weather forecasts for offshore wind power forecasting. Wind Energy, 23(4), 884–897. <u>https://doi.org/10.1002/we.2456</u>

[2] Lindsay, N., Libois, Q., Badosa, J., Migan-Dubois, A., Bourdin, V. (2020). Errors in PV power modelling due to the lack of spectral and angular details of solar irradiance inputs. Solar Energy,197, 266-278 <u>https://doi.org/10.1016/j.solener.2019.12.042</u>

[3] Schalkwijk, J., Jonker, H. J. J., Siebesma, A. P., & Van Meijgaard, E. (2015). Weather forecasting using GPU-based large-Eddy simulations. Bulletin of the American Meteorological Society, 96(5), 715–723. <u>https://doi.org/10.1175/BAMS-D-14-00114.1</u>

[4] E.J. Wiegant, P. Baas, R.A. Verzijlbergh, B.Reijmerink, and S.Caires. The new frontier in numerical metocean modelling: coupled high-resolution atmosphere wave interactions. In Wind Europe Offshore, 2019.

[5] R.A.Verzijlbergh, H.J.J. Jonker, P. van Dorp, E.J. Wiegant, P. Baas, B.M. Meijer, and J. Coeling. Breakthrough weather prediction technology enables wind turbine resolving resource assessments. In Wind Europe, 2019.

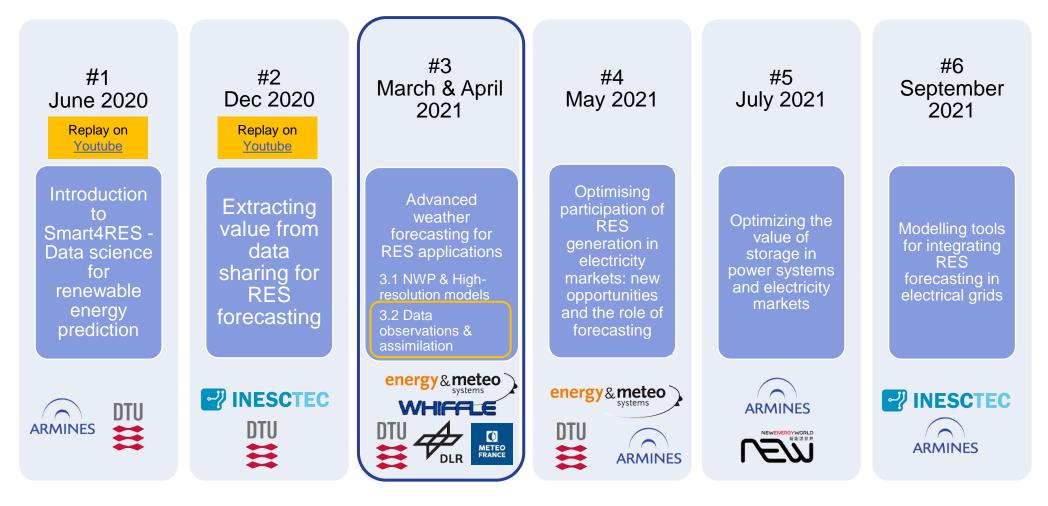






Smart4RES webinar series

Season1: Towards a new Standard for the entire RES forecasting value chain



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Thank you

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Back-up slides

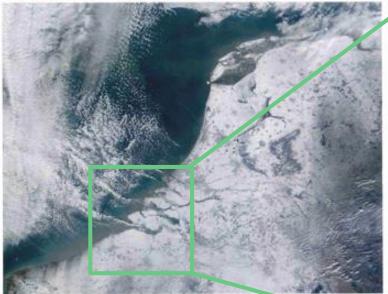




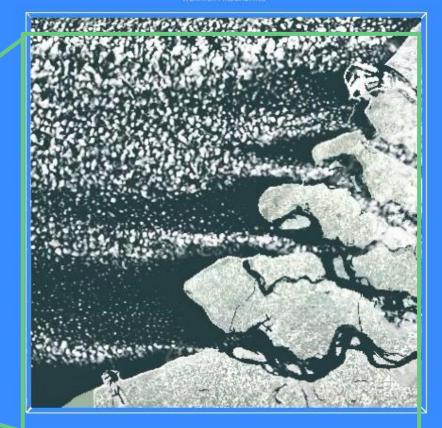




METEOROLOGICA



Wolkenstraten bij Zeeland weerspiegelen de verdeling van sneeuwoudskiekt land en zee



https://vimeo.com/whiffle/cloudstreetswinter