

KliWiSt – Predicting future wind speeds based on climate projections and MCP methods

Fraunhofer IWES

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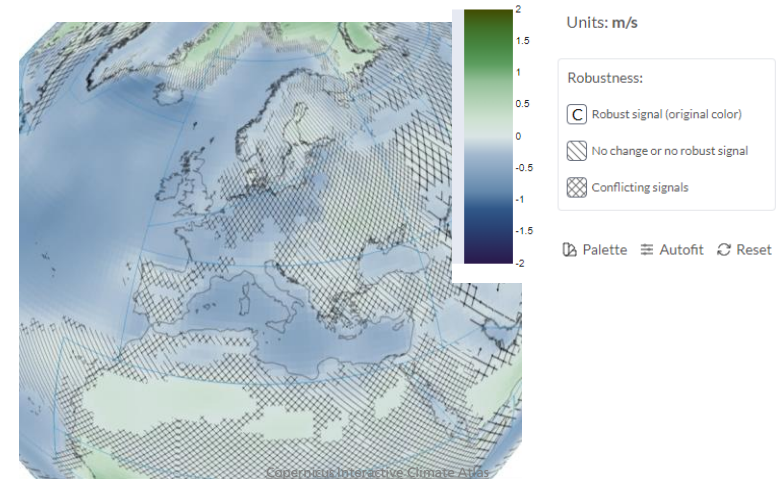
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Motivation

High resolution time series-based analysis

- State-of-the-Art (TR6 – Rev.10):
Assumption of persistence of past wind climate
(+ maybe additional uncertainty)
 - Still valid?
 - **Timing of energy production** (e.g. by **time of day** and **season**) becomes more crucial
 - How to improve?
 - **Include future climate information** in wind resource assessment
 - Perform **time series analysis** in addition to mean values
- Data based uncertainty estimation
- Analysis of seasons and time of day possible



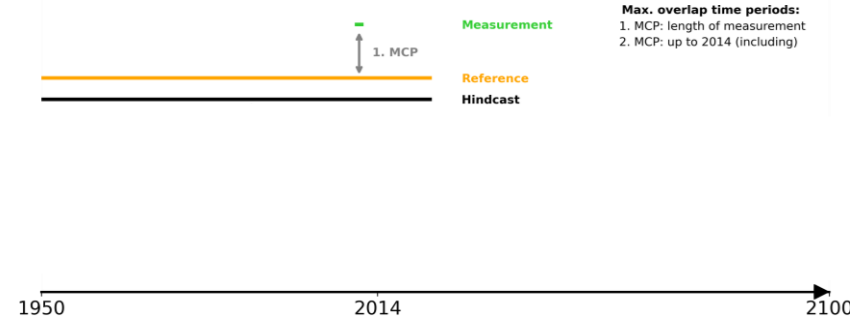
Method

Handling climate model data in wind resource assessment

Borowski et al. - Part 1 (in internal review) on the impact of inter-annual variability on long-term wind speed predictions based on the MCP method

Base: State-of-the-Art method to predict hindcast wind speed (using Measure-Correlate-Predict (MCP) Methodology)

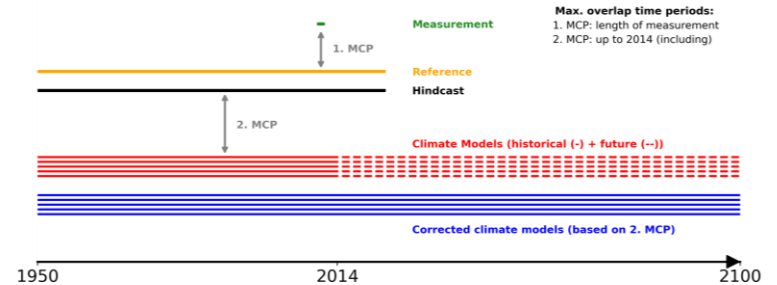
- Input:
 - Short-term on-site data (One-year measurement)
 - Long-term reference data (e.g. reanalysis)
- Output:
 - **Hindcast:** Prediction of wind speeds of the past
 - Including: Site specific wind characteristics and long-term wind variability
- Method:
 - Statistical relationship (correlation): e.g. linear regression



Method

Handling climate model data in wind resource assessment

- Input:
 - Hindcast
 - Ensemble of climate models (historical)
- Output:
 - Corrected climate models (ensemble): Prediction of the wind climate from past to future
- Method to link input data:
 - Statistical relationship (correlation): e.g., linear regression
 - **Issue:** Time inconsistency between Hindcast and climate models data sets



Issue: Excuse – Weather forecast VS Climate model

Initial value – Boundary value problem

Weather Model/Reanalysis ! Connection not time consistent ! Climate Model

- Short-term predictions (**hours to days**)
- High precision thanks to accurate **initial conditions**
- Long-term trends and averages (**decades to centuries**)
- Focus on **boundary conditions** and **external drivers** (e.g., CO₂ emissions, solar cycle, Ocean currents, Weather Systems (e.g. NAO))

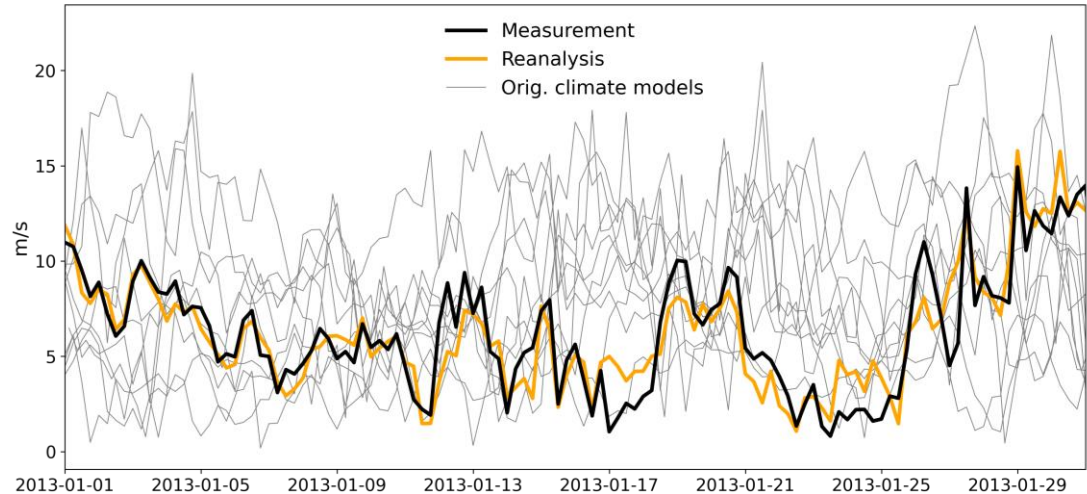
The **uncertainties** in the **initial conditions** grow **strongly** over **longer periods** of time, which makes it **impossible** to make **precise long-term predictions** based on initial values alone.

!!! Climate model data represent averages/trends/statistical values **BUT NOT** a precise prediction for a certain time step **!!!**

Problem: Excuse – Weather forecast VS Climate model

Visualization

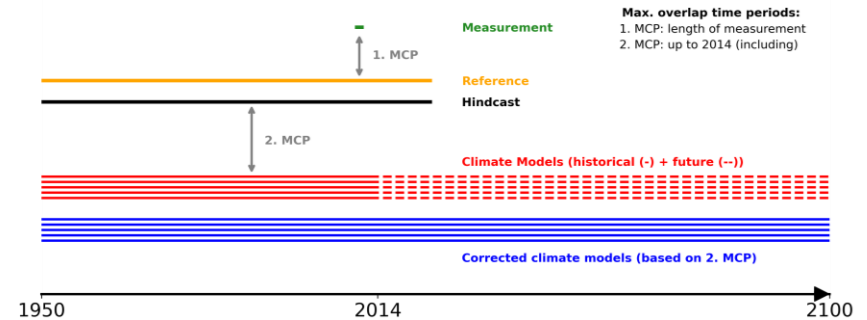
- Weather model/reanalysis:
 - Exact values at the specified time steps!
- Climate models:
 - Represent the probabilistic value of the time step based on the surrounding decades!



Solution

Handling climate model data in wind resource assessment

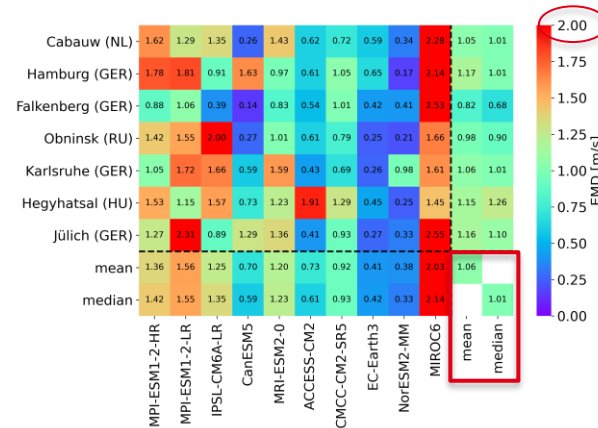
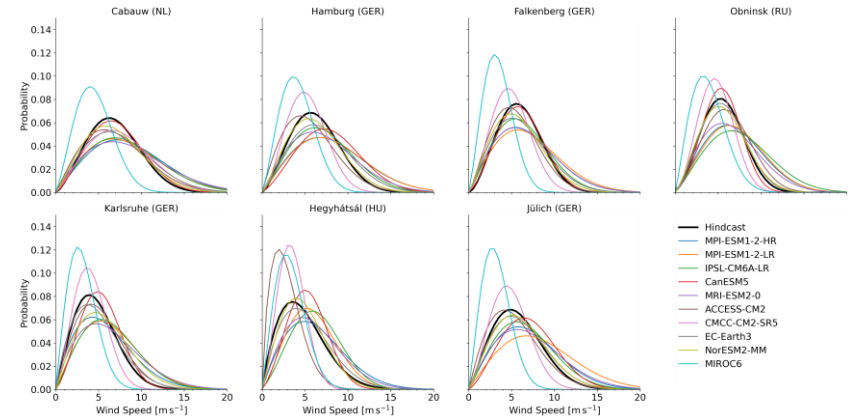
- Input:
 - **Hindcast**
 - **Ensemble of climate models (historical)**
- Output:
 - **Corrected climate models (ensemble):** Prediction of the wind climate from past to future
- Method to link input data:
 - Statistical relationship (correlation): e.g., linear regression
 - **Solution:** Methods that do not require time consistency



Historical: Validation

Original climate models (CMs)

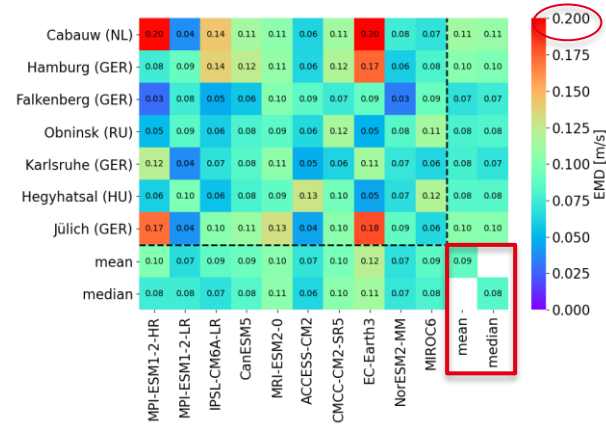
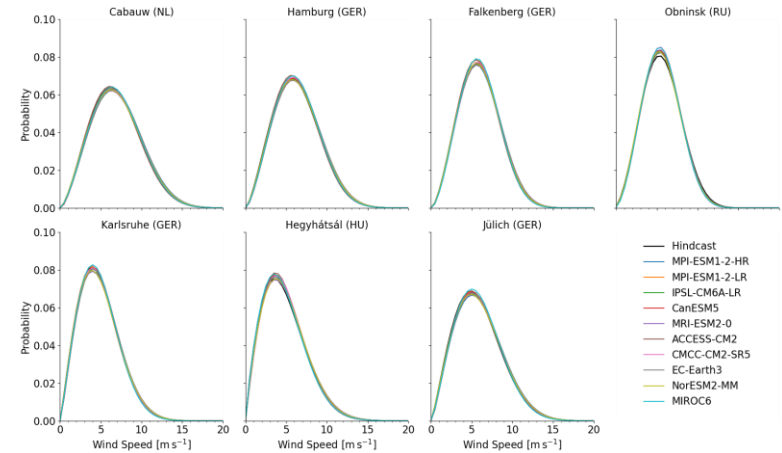
- Earth Mover's Distance (EMD) used here to compare two histograms
- Performance of climate models in the past:
 - Site dependent
 - Mean of EMD: 1.06 m/s
 - High-resolution models fit better to hindcasts than low-resolution models at most sites (e.g., MPI-ESM1-2-HR/LR)



Historical: Validation

Corrected climate models (CMs)

- Method is **beneficial**
 - Reduction of EMD: 0.09 m/s (vs 1.06 m/s)
 - Correction depending on site
 - Regardless of resolution (low/high): models can be **fitted quite well**
- Seasonal cycle?



Historical: Validation

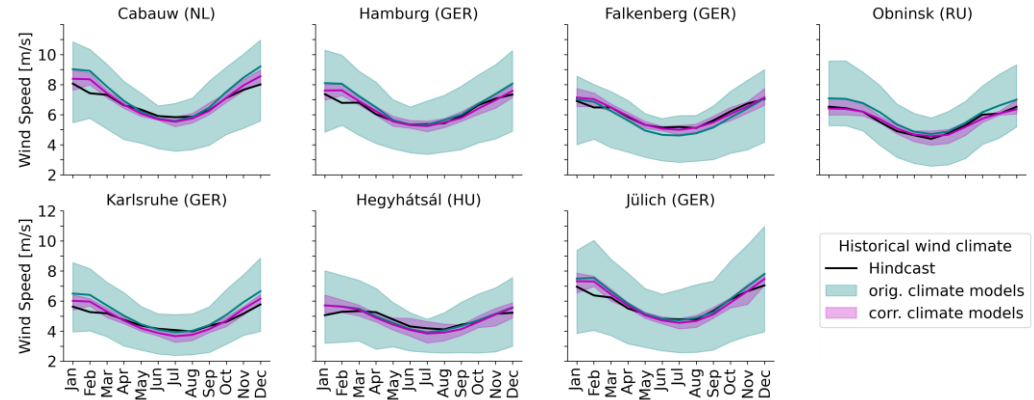
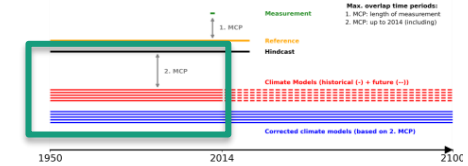
Seasonal cycle

- Improvement:

- Monthly mean of corrected CMs fit better to Hindcast
- Reduction of spread of CMs
- Greatest adjustment: Winter

- Further improvements:

- Usage of downscalings/regional models

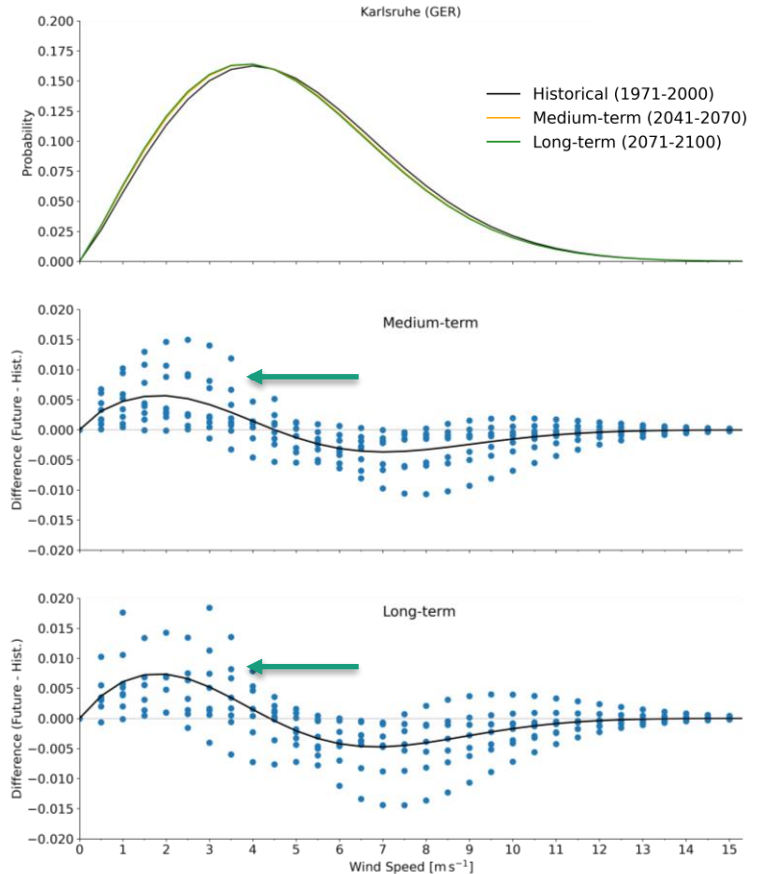


Future Wind Resource

Karlsruhe: Historical – medium – long-term

Scenario: Worst case scenario (SSP5 – 8.5)

- Minor differences between historical and future
- More frequent: Low wind speeds (< 4 m/s)
 - More frequent shutdowns
- Less frequent: 5 – 10 m/s
- Spread of CMs: Slightly higher for long-term



Seasonal Future Wind Resource

Karlsruhe: Historical – medium – long-term

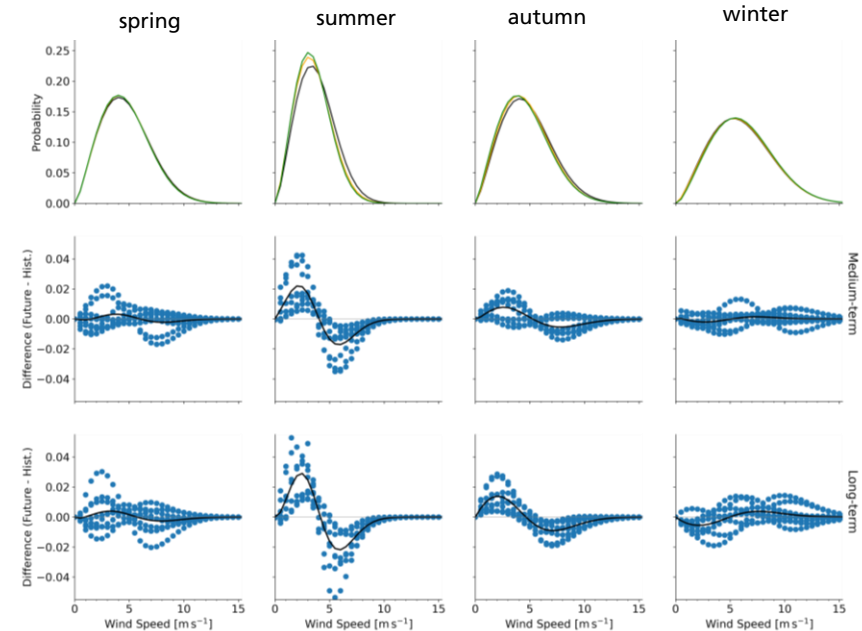
Scenario: Worst case scenario (SSP5 – 8.5)

- Difference: Future – Historical
- Degree of deviation varies by season
- Partly conflicting signals:

Wind speed change	Higher to lower	Lower to higher
Clear signal	Summer Autumn + long-term	
Conflicting signal	Spring Autumn + medium-term	Winter + (medium)/long-term

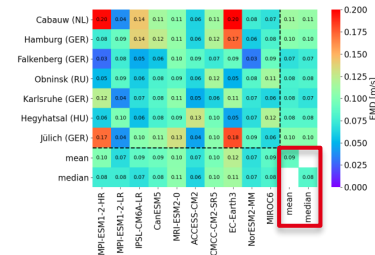
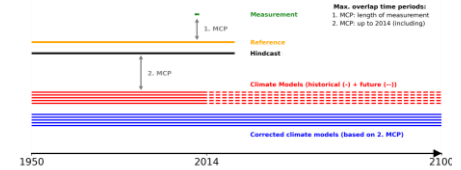
Borowski et al. – Part 2 (in preparation): Predicting future wind speeds based on climate projections and MCP methods

— Historical (1971-2000)
— Medium-term (2041-2070)
— Long-term (2071-2100)



Conclusion

- Workflow to incorporate climate models at a specific site**
 - High resolution: **6 hourly** instead of daily/monthly/annual means
 - Method** can be used with **every climate model** data set **on any site**
- Future wind speed changes:
 - Tendency to lower wind speeds** at some sites
 - Seasonal differences
 - Differences in medium- and long-term
- Further improvement: Usage of downscaling/regional climate models
 - CMIP6 downscaled models** (in preparation) will provide hourly resolution



Thank you for your attention

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