

## C3S Energy Webinar GLOBAL CLIMATE INDICATORS 23<sup>rd</sup> April 2024

Processing methodologies and selected examples of global climate indicators for the C3S Global Energy Climate Service

> Letizia Lusito **Inside Climate Service**













Climate

Change

## Outline

**Motivation:** the C3S Enhanced Operational Global Service for the Energy Sector and the need for

**Climate indicators** 

Wind profile power law for the wind speed scaling

#### Global climate indicators, their utility, applications and examples:

Historical climate Seasonal forecasts Climate projections

#### Bias adjustment (b.-a.) methods, applications and examples:

Delta approach Quantile matching (QM) CDF-transform (CDFt)

**Conclusions and future outlook** 









# Wind profile power law for the wind speed scaling









## Wind speed scaling

#### The need for wind speed scaling:

wind speed measurements are traditionally available (anemometers installation heights) or referred to the standard height of **10 m** above ground level in order to clear the measurements from the effects of terrain, buildings, surroundings (open ground)

https://library.wmo.int/viewer/41755/ download?file=wmotd\_452\_en.pdf&type=pdf&navigator=1

## Wind speed scaling for the energy sector

How can we use the 10 m winds to determine wind speeds at the heights relevant for wind turbines?



190 metres tall wind turbines (Hornsea 1 offshore wind farm)

Winds are driven by large-scale pressure gradients in the atmosphere :

 (> 1000 m or so above the surface) winds are driven mainly by geostrophic balance between Coriolis and local pressure gradient forces.

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(< 1000 m) frictional effects distort the wind field; wind speed and direction become dependent upon elevation above the mean surface, roughness of the surface, air-sea temperature differences

## Wind profile (or shear) power law

https://en.wikipedia.org/wiki/Hornsea\_Wind\_Farm

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#### Wind shear power law

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The **wind profile power law** is a relationship between the wind speeds  $v_r$  at one (reference) height  $z_r$  and the wind speed v at another height z.

$$\frac{v}{v_r} = \left(\frac{z}{z_r}\right)^{\alpha}$$

The exponent  $\alpha$  is an empirically derived coefficient that varies dependent upon the stability of the atmosphere. For neutral stability conditions, it is approximately 1/7 = 0.143.

## Application

The wind shear exponent  $\alpha$  can be used to estimate the vertical wind profile and derive wind speed information at different heights above ground in context and applications connected to the energy sector, e.g., wind farms modelling and climate modelling for RE resources assessments.







## Wind speed scaling for the energy sector

Climate In the C3S Energy contracts, a location dependent exponent is computed to account for local features as represented by the model (ERA5). Temporal variations are accounted for too, by stratifying  $\alpha$  according to the time of day and month.



The power law exponent will be used to derive the climate indicator "wind speeds at 100 m" when not directly available, e.g., for CMIP6 projections, for seasonal forecast and, for coherence, also for the past climate data (1950-2023)



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# Global climate indicators, their utility, applications and examples









## Historical Climate Indicators

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## Based on ECMWF Reanalysis v5 (ERA5)

Climate reanalyses combine past observations with models to generate consistent time series of multiple climate variables.

The most relevant climate indicators for the Energy Sector have been processed in a more ready-to-use formats by the C3S Energy Team

#### Spatial resolution:

- Gridded level (NetCDF format) covering the entire Globe
- Aggregated information at ADM0/ADM1 level

#### Temporal resolution: daily, soon to be hourly (except Precipitation)

ERA5 is the latest climate reanalysis produced by ECMWF, providing hourly data on many atmospheric, land-surface and sea-state parameters together with estimates of uncertainty.

ERA5 data are available in the Climate Data Store on regular latitude-longitude grids at 0.25° x 0.25° resolution, with atmospheric parameters on 37 pressure levels.

- Temperature
- Precipitation
- Wind speed at 10/100 m AGL
- Wind direction at 10/100 m AGL
- Solar radiation at Surface













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## Historical Climate Indicators

#### **Selected examples:**

- maps from gridded products : useful to compare climate indicators across geographical • areas/selected locations over a certain period of time, e.g. for resource assessments
- time series from aggregated products : useful to compare climate indicators across countries over a certain period of time

June 2022, Hour (UTC): 0





## Seasonal forecast climate indicators

## Seasonal forecast models for the C3S2 Energy community

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**Simulations of possible weather scenarios:** probabilities of how likely it is that a season will be wetter, drier, warmer or colder compared to the average for that period of the year. This is possible because some components of the Earth system evolve more slowly than the "chaotic" atmosphere in a predictable way.

Model	Original spatial res.	Ensemble Members	Initialisation Approach	Lead time	Link to institutions
ECMWF SYS 5.1	1° x 1°	25 hindcasts 51 forecasts	Burst	215 days	https://www.ecmwf.int/en /forecasts/documentation- and-support/long-range
CMCC SYS 3.5	1° x 1°	40 hindcasts 50 forecasts	Burst	6 calendar months	https://sps.cmcc.it
DWD SYS 2.1	1° x 1°	30 hindcasts 50 forecasts	Burst	6 calendar months	https://www.dwd.de/EN/o urservices/seasonals_forec asts/start.html









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Application: Seasonal forecast climate indicators for the C3S2 Energy community (other models follow the same schema)

Variable	Time period	Source	Temporal resolution	Туре	Units
Temperature (TA)	Hindcasts: 1993-2016 Forecasts: From 2023-01	ECMWF SYS 5.1	Gridded: 6 hours ADMO: 1 month	Instantaneous	°К
Precipitation (TP)	Hindcasts: 1993-2016 Forecasts: From 2023-01	ECMWF SYS 5.1	Gridded: 1 day ADMO: 1 month	Daily cumulated	m
Solar Radiation at Surface (GHI)	Hindcasts: 1993-2016 Forecasts: From 2023-01	ECMWF SYS 5.1	Gridded: 1 day ADMO: 1 month	Daily mean	W m <sup>-2</sup>
Wind speed 10 m (WS10)	Hindcasts: 1993-2016 Forecasts: From 2023-01	ECMWF SYS 5.1	Gridded: 6 hours ADMO: 1 month	Instantaneous	m <sup>-1</sup>
Wind speed 100 m (WS100)	Hindcasts: 1993-2016 Forecasts: From 2023-01	ECMWF SYS 5.1	Gridded: 6 hours ADMO: 1 month	Instantaneous	m-1









## Climate Projections

Climate The World Climate Research Programme (WCRP) Working Group on Coupled Modelling oversees the Change Coupled Model Intercomparison Project, which is now in its 6th phase (CMPI6)

- For climate model improvements
- For support national and international assessments of climate change.

#### CO2 emissions in CMIP6 scenarios

	-			
SSP	Scenario	Estimated warming (2041–2060)	Estimated warming (2081–2100)	Very likely range in °C (2081–2100)
SSP1- 1.9	very low GHG emissions: CO <sub>2</sub> emissions cut to net zero around 2050	1.6 °C	1.4 °C	1.0 - 1.8
SSP1- 2.6	low GHG emissions: CO <sub>2</sub> emissions cut to net zero around 2075	1.7 °C	1.8 °C	1.3 – 2.4
SSP2- 4.5	intermediate GHG emissions: CO <sub>2</sub> emissions around current levels until 2050, then falling but not reaching net zero by 2100	2.0 °C	2.7 °C	2.1 - 3.5
SSP3- 7.0	high GHG emissions: CO <sub>2</sub> emissions double by 2100	2.1 °C	3.6 °C	2.8 - 4.6
SSP5- 8.5	very high GHG emissions: CO <sub>2</sub> emissions triple by 2075	2.4 °C	4.4 °C	3.3 - 5.7



Eyring, V., Bony, S., Meehl, G. A., Senior, C. A., Stevens, B., Stouffer, R. J., and Taylor, K. E.: Overview of the Coupled Model Intercomparison Project Phase 6 (CMIP6) experimental design and organization, Geosci. Model Dev., 9, 1937–1958, https://doi.org/10.5194/gmd-9-1937-2016, 2016.







## Climate Projections

Climate Change

#### Application: CMIP6 projection models to be delivered within C3S-Energy

Selection based on model components independence, availability of reasonable horizontal and temporal resolution, maximum variability in the Equilibrium Climate Sensitivity (ECS)

#### C3S2 Energy selection:

(data further interpolated at 0.25°x 0.25° horizontal resolution, daily resolution)



Model	Time resolution	Spatial resolution	Scenarios	Variant label	Calendar
CMCC-CM2-SR5	3 hours	100 km	historical, ssp245, ssp370	r1i1p1f1	365_day
EC-Earth3	3 hours	100 km	historical, ssp245, ssp370	r1i1p1f1	proleptic_gregorian
MPI-ESM1-2-HR	3 hours	100 km	historical, ssp245, ssp370	r1i1p1f1	proleptic_gregorian
BCC-CSM2-MR	3 hours	100 km	historical, ssp245, ssp370	r1i1p1f1	365_day

Boé, J., Terray, L. Projections climatiques globales de nouvelle génération : résultats sur la France et éléments pour la sélection des modèles, Report for EDF, 2023









# Climate Projections

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## Climate projections indicators

Variable	Time period	Source	Temporal Resolution	Spatial Resolution	Spatial Aggregation	Units
Temperature (TA)						°K
Precipitation (TP)			daily for			М
Wind speed 10 m (WS10)	2015-2065	models x	and ADM0,	0.25° x0.25°	Gridded & ADM0, ADM1	m s <sup>-1</sup>
Wind speed 100 m (WS <u>100)*</u>		2 3385	ADM0/ADM1			m s⁻¹
Solar Radiation at Surface (GHI)						W m <sup>-2</sup>







## Indicators in the Projection Stream

Climate Change **Selected examples:** time series for a particular Country for combinations of models/scenarios: useful to compare the influence of scenarios, fixing a model or to compare the projections offered by different models and have a better estimation of the uncertainty











# Bias adjustment (b.-a.) methods, applications and examples









## The need for the bias adjustment

Climate Change Fully physically based climate simulation chain formed by General Circulation Models (GCMs) dynamically downscaled through Regional Climate Models (RCMs), are powerful tools for describing general climate conditions,

Their direct use in climate change impact or adaptation studies, risk assessment or other analyses requiring climate projections at a regional or local scale is still **challenging**.

These data present **systematic biases** (systematic model errors caused by imperfect conceptualization, discretization, coarse representation of regional features, and spatial averaging within model grid cells) when compared to observations.

**Bias adjustments**: the most adopted method to provide 'corrected' climate scenarios; it consists in the calibration of an empirical transfer function between simulated and observed distribution over a historical period that can be used to adjust the climate model output over a future period.

This statistical post-processing step adjusts **selected statistics** (mean, variance, distribution) of the so-called "raw" model simulations to better match observed time series over the reference period (Bartok et al., 2019).



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Bartók, Tobin, Vautard, Vrac, Jin, Levavasseur,

Denvil, Dubus, Parey, Michelangeli, Troccoli, Saint-

Drenan, A climate projection dataset tailored for

the European energy sector, Climate Services,

Volume 16, 2019, 100138, ISSN 2405-8807, https://doi.org/10.1016/j.cliser.2019.100138.

## Overview of bias adjustment methods

Clima Chan	<b>Delta</b> $F = \frac{\langle PROJ \rangle}{\langle ERA5 \rangle}$ $PROJ_{B.A.} = \frac{PROJ}{F}$	Adjustment of a "source" dataset, based on a target variable simply calculating the change factors (or anomalies or deltas) from the averaged datasets The averages of the source and target dataset are matched by performing a scaling adjustment of the source dataset with the delta factors	$O_{REF}$ $T_{BC}$ $T_{Correction}$ $T_{REF}$ $T_{RAW}$	Navarro-Racines, C., Tarapues, J., Thornton, P. <i>et al.</i> High-resolution and bias-corrected CMIP5 projections for climate change impact assessments. <i>Sci Data</i> <b>7</b> , 7 (2020). https://doi.org/10.1038/s41597-019- 0343-8
	Quantile mapping $P_o = h(P_m)$ $P_o = F_0^{-1}(F_m(P_m))$	Application of the probability integral transform that is designed to adjust the distribution of modelled data, such that it matches observed climatologies by <b>mapping</b> the modelled cumulative distribution function ( <b>CDF</b> ) of the variable of interest onto the observed CDF	CDF of Observations CDF of Observations CDF of Climate model (before bias correction) Model $\frac{F_m(x_m)}{(x_m)}$ Obs quantile $\frac{F_m(x_m)}{(x_m)}$ Obs Quantile $\frac{F_m(x_m)}{(x_m)}$ Obs Quantile $\frac{F_m(x_m)}{(x_m)}$ Obs	Gudmundsson, L., Bremnes, J.B., Haugen, J.E., Engen Skaugen, T., 2012. Technical note: downscaling RCM precipitation to the station scale using quantile mapping—a comparison of methods. Hydrol. Earth Syst. Discuss. 9, 6185–6201, http://dx.doi.org/10.5194/hessd- 9-6185-2012.
	<b>CDFt</b> $T(F_{Gh}(x)) = F_{Sh}(x)$ $T(u) = F_{Sh}(F_{Gh}^{-1}(u))$ $F_{Sf}(x) = T(F_{Gf}(x))$ $F_{Sf}(x) =$ $F_{Sh}(F_{Gh}^{-1}(F_{Gf}(x)))$	A probabilistic downscaling method: statistical properties (CDFs) between a target and a source (biased projection model or GCMs) datasets are modelled. Non-parametric correspondences between the predictor and predictand CDFs. $F_{Sh}(x)$ the CDF of the target for the calibration (h, historical) period and $F_{Gh}(x)$ the CDF of the source/to-be-adjusted for the calibration period. Thus, T can be used to model the relationship between $F_{Sf}(x)$ and $F_{Gf}(x)$ , <i>i.e.</i> , same CDFs but for a future (different) period, assuming that $F_{Gf}(x)$ is known.	Orrected	Michelangeli, Vrac, Loukos (2009) Probabilistic downscaling approaches: Application to wind cumulative distribution functions. Geo. Res. Letters, 36, L11708, <doi:10.1029 2009gl038401=""> Vrac, Drobinski, Merlo, Herrmann, Lavaysse, Li, Somot (2012) Dynamical and statistical downscaling of the French Mediterranean climate: uncertainty assessment. Nat. Hazards Earth Syst. Sci., 12, 2769- 2784<doi:10.5194 nhess-12-2769-<br="">2012&gt;.</doi:10.5194></doi:10.1029>

	Bias-ac	djustment method	s for C3S Energy
Climate Change	Variable	Seasonal Forecasts	Climate Projections
	ТА	Quantile mapping with ERA5 as reference	Moving windows <b>Cumulative</b> <b>Distribution Function transform (CDFt)</b> with ERA5 as reference
	ТР	Quantile mapping with ERA5 as reference	Cumulative Distribution Function transform (CDFt) with ERA5 as reference
	WS10	Quantile mapping with ERA5 as reference	Cumulative Distribution Function transform (CDFt) with bias adj. ERA5 as reference
2	GHI	Quantile mapping with ERA5 as reference	Delta method with ERA5 as reference
-			











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## Conclusions and future outlook

#### Added value of the C3S2 Energy contract:

- wind shear power law used to derive wind speeds at height relevant for the wind power simulations, also for climate projections, where they are not normally available to user communities
- bias adjustment used to minimize biases in climate projections making them useful to renewable energy assessments but preserving the climate change trends
- improvement in the delivery of climate indicators strongly based on the continuous engagement and feedback with users, stakeholders and ECMWF/CDS communities
- Next steps:
  - $\circ$  Further assessment of ERA5 and CMPI6 climate indicators
  - Publication of data and procedures
  - $\circ$  Further engagement with users/stakeholders and training activities









### BACKUP









## BIAS ADJUSTMENT

Climate Process of reducing biases in climate models at post processing phase Change

Even though climate models (CMs), both regional and global, have gained impressive skill in recent decades, model errors and biases do still exist. <u>Model biases are systematic differences between the simulated climate statistic and the corresponding real-world statistic over a historic period where observations exist (Maraun 2016).</u>

These biases could be due to inaccurately resolved topography of the model, due to misrepresentation of convection processes, due to misplacement of large scale processes such as the midlatitude circulation or many other factors. These model errors then manifest as discrepancies in the values and statistical properties of meteorological variables.

Bias adjustment' methods, which have now become a standard pre-processing step for climate impact studies, calibrate an empirical transfer function between simulated and observed distribution over a historical period that can be used to adjust the climate model output over a future period.









#### Seasonal Forecasts

Climate Change Simulations of possible weather scenarios: probabilities of how likely it is that a season will be wetter, drier, warmer or colder compared to the average for that period of the year. This is possible because some components of the Earth system evolve more slowly than the "chaotic" atmosphere in a predictable way.

#### The CMCC Seasonal Prediction System model CMCC-SPS3.5

**Coupled Model**: several independent but fully coupled model components simultaneously simulating the Earth's atmosphere, ocean, land, sea ice and river routing, together with a central coupler/driver component for data synchronization and exchange

Operated monthly in Ensemble seasonal mode (6-month predictions) Monthly ensemble hindcasts (1993-2016) for performance evaluation and bias adjustments to operational forecasts Forecast ensemble size: 50 in operational, 40 in hindcast mode













## The need for climate data

The need for climate data

Major transformations in the ENERGY sector:

- increasingly higher share of power supply from variable renewable energy (RE) sources, (wind and solar)<sup>1</sup>
- taking place against a variable and changing climate

#### **C3S Enhanced Energy** operational service:

to deliver an enhanced operational energy service at the global scale covering data about the **past climate**, **multi-model seasonal forecasts** and **multi-model climate projections** 

<sup>1</sup>most of the electricity production sector should be decarbonized by 2050 in order to fulfil the 2°C/1.5°C, warming target of the international agreements.



**Climate Indicators** 



## Wind speed scaling for the energy sector

Month: 1, Hour: 12 Climate Month: 1, Hour: 0 Month: 1, Hour: 6 Change -0.80 -0.48 -0.16 0.16 0.16 0.48 0.80 -0.80 -0.48 -0.16 alpha alpha -0.48 0.48 -0.80 -0.16 0.16 0.80 alpha Month: 7, Hour: 12 Month: 7, Hour: 6 Month: 7, Hour: 0 -0.80 -0.48 -0.16 0.16 0.48 0.80 -0.80 -0.48 0.48 -0.16 0.16 0.80 -0.80 -0.48 -0.16 alpha alpha

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0.48

0.80

0.16

alpha

0.48

0.80

### Use of the $\alpha$ exponent in C3S Energy

#### Validation

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The power law exponent will be used to derive the climate indicator "wind speeds at 100 m" when not directly available, e.g., for CMIP6 projections, for seasonal forecast and, for coherence, also for the past climate data (1950-2023)





ECMWF



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## Climate Projections

The World Climate Research Programme (WCRP) Working Group on Coupled Modelling oversees the

#### **Coupled Model Intercomparison Project, which is now in** its 6th phase (CMPI6)

a collaborative framework designed to improve knowledge of climate change since 1995.

It is developed in phases to foster the climate model improvements but also to support national and international assessments of climate change.



#### CO2 emissions in CMIP6 scenarios

- Historical - SSP1-1.9 - SSP1-2.6 - SSP4-3.4 -- SSP5-3.40S - SSP2-4.5 - SSP4-6.0 - SSP3-70 - SSP5-8 5

SSP

SSP1-

1.9

SSP1-

2.6

SSP2-

4.5

SSP3-

7.0

SSP5-

8.5



(2041 - 2060)(2081 - 2100)(2081-2100) very low GHG emissions: 1.6 °C 1.4 °C 1.0 - 1.8CO<sub>2</sub> emissions cut to net zero around 2050 low GHG emissions 1.7 °C 1.8 °C 1.3 - 2.4CO2 emissions cut to net zero around 2075 intermediate GHG emissions: CO2 emissions around current levels until 2050, then falling but not reaching net 2.0 °C 2.7 °C 2.1 - 3.5zero by 2100 high GHG emissions: 2.1 °C 3.6 °C 2.8 - 4.6CO<sub>2</sub> emissions double by 2100 verv high GHG emissions: 2.4 °C 4.4 °C 3.3 - 5.7CO<sub>2</sub> emissions triple by 2075

Scenario

Estimated

warming

Estimated

warming

Very likely range

in °C

Coupled models are computer-based models of the Earth's climate, in which different parts (such as atmosphere, oceans, land, ice) are "coupled" together, and interact in simulations.

Eyring, V., Bony, S., Meehl, G. A., Senior, C. A., Stevens, B., Stouffer, R. J., and Taylor, K. E.: Overview of the Coupled Model Intercomparison Project Phase 6 (CMIP6) experimental design and organization, Geosci. Model Dev., 9, 1937–1958, https://doi.org/10.5194/gmd-9-1937-2016, 2016.

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