

Lessons learned about the use of downscaled climate model results for practical use

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The objective

To derive **robust** and **reliable** information about future climatic outlooks on **regional** and **local** scales

"Actionable" information - still not widely discussed. E.g. often only Euro-CORDEX ensemble (RCMs)



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Relevant factors for practitioners

Sensitive to methodology (models' **minimum skillful scale)**

Nondeterministic ("chaotic") internal variability

Climate sensitivity - GCM-dependent - skill?

Different future emissions



"... at least for present day climates, dynamical downscaling methods provide little advantage over statistical techniques." (Fowler et al. 2007)

"RCMs ... do not include all possible Earth system processes and therefore neglect important processes ... which may influence regional climate projections." Chapter 10, IPCC AR6 (2021) "Disc adjustice a statistical past processing

"Bias adjustment is a statistical post-processing technique used to pragmatically reduce the mismatch between the statistics of climate model output and observations." Chapter 10, IPCC AR6 (2021)

"Several problems have been identified that may arise from an uncritical use of bias adjustment, and that may result in misleading impact assessments... In the light of these issues, several authors dismiss the use of bias adjustment for climate change studies" Cross-Chapter Box 10.2, IPCC AR6 (2021)

Empirical-statistical downscaling

Weather predictions and climate projections require different strategies

"Downscaling **weather**" (actual outcomes)

"Downscaling **climate**" (statistics)

Not the same as bias adjustment - grab large-scale climatic aspects that GCMs skillfully reproduce and add information about dependencies linking local climate.

Multiple-variable predictors

Suitable for weather, but not for climate.



Parding et al. (2019) https://doi.org/10.1175/JAMC-D-17-0348.1



Both dynamical and empirical-statistical downscaling

Different sets of assumptions

Different strengths and weaknesses

Different tools



Regional climate projections

Deser et al. (2012)

One single model - a **range** of different non-deterministic outlooks.

The *law of small numbers* becomes relevant

PERSPECTIVE

Temperature trend (°C per 55 years)

Mazatlán, Mexico Phoenix, Arizona Average Warmest United States 2 -Coolest 1910 1930 1950 1970 1990 2010 2030 2050 0 5 10 15 20 25 30 -2 -1 0

Year

NATURE CLIMATE CHANGE DOI: 10.1038/NCLIMATE1562

Number of model runs

Regional temperature

5-25°E/58-70°N





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Making use of available information

 Combine dynamical downscaling with empirical-statistical downscaling:

Bias-corrected Euro-CORDEX

VS

Empirical-statistical downscaling (ESD)

- Many climate model simulations
- Extends the RCM outputs to cover the full ensemble of opportunities



Figure: Downscaling of annual mean temperature anomalies over Norway assuming CMIP5 scenarios based on empirical-statistical and bias-corrected dynamical downscaling. The ensembles include the full set and subsets of CMIP5 global model simulations.

Information - as opposed to data

Summary statistics

Approximately normally distributed? pdf ~ $N(\mu,\sigma^2)$

The *law of small numbers* - minimum ensemble size



CMIP6

Summary & connection to society

A basis for providing climate information to society

Adoption of climate information for real-word decisions? Overcoming *the law of small numbers* and models' minimum skillful scales

Understand user needs and capacities...?

Capacity building: training, education and interaction







References



Linking climate change modelling to impacts studies: recent advances in downscaling techniques for hydrological modelling <u>https://doi.org/10.1002/joc.1556</u>

Communication of the role of natural variability in future North American climate https://doi.org/10.1038/nclimate1562

A Norwegian Approach to Downscaling <u>https://doi.org/10.5194/gmd-2021-176</u>

Various ways of using empirical orthogonal functions for climate model evaluation <u>https://doi.org/10.5194/gmd-16-2899-2023</u>

Statistical projection of the North Atlantic storm tracks <u>https://doi.org/10.1175/JAMC-D-17-0348.1</u>

Sub-sampling impact on the climate change signal over Poland based on simulations from statistical and dynamical downscaling <u>https://doi.org/10.1175/JAMC-D-18-0179.1</u>

A strategy to effectively make use of large volumes of climate data for climate change adaptation <u>https://doi.org/10.1016/j.cliser.2017.06.013</u>