



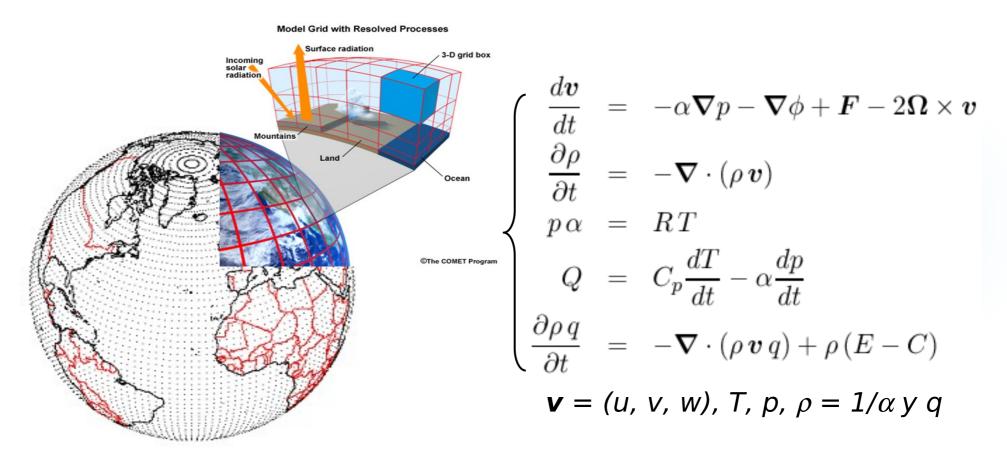


Downscaling and Bias Correction in climate4R

University of Buenos Aires, 31 August 2022



Introduction



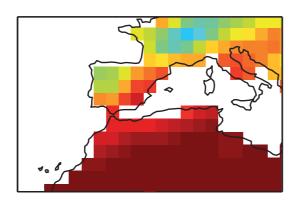
General Circulation Models (GCM) simulate the evolution of the climate system by solving numerically a set of partial differential equations over a spatio-temporal grid. These simulations are forced with different possible emission scenarios of greenhouse gases up to the end of the 21st century.

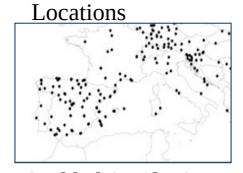


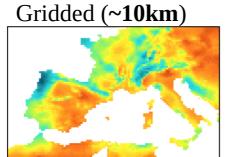
Motivation









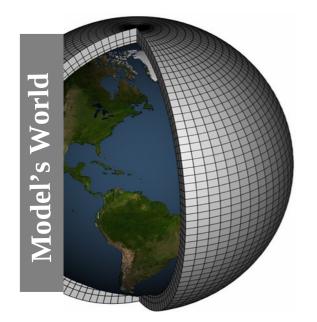




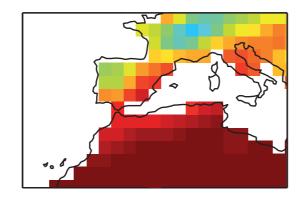
The gap in resolution and the existence of model biases prevents climate data from GCMs to be usable by certain socio-economic sectors (e.g., energy, agriculture, hydrology)

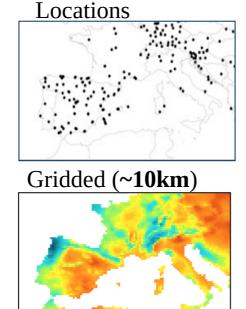


Motivation





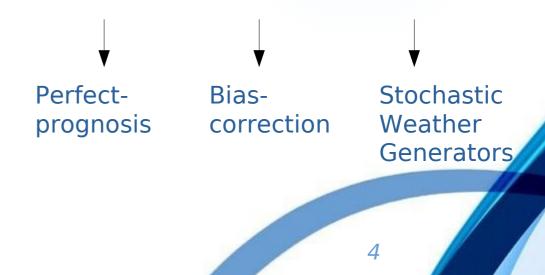




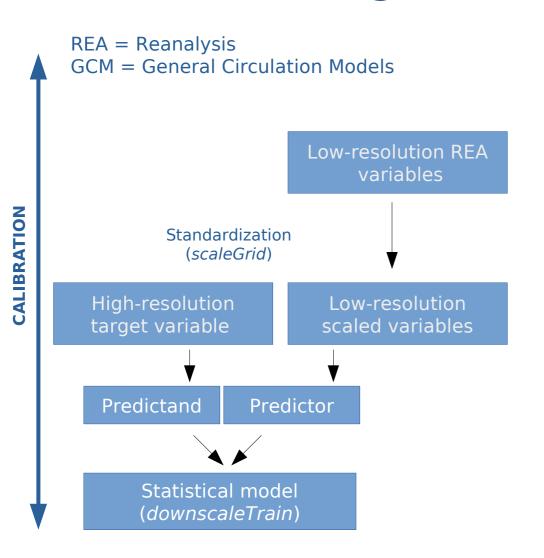
Downscaling bridges the scale gap between the model and the real world. There are 2 conceptually different downscaling approaches:

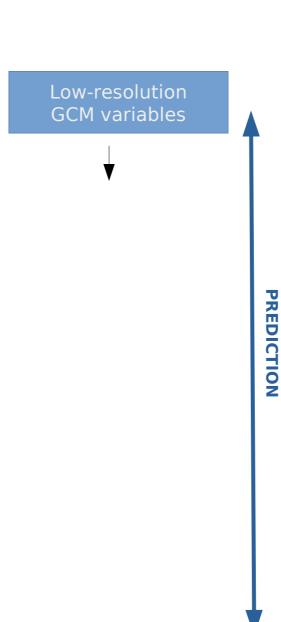
- Dynamical downscaling
- Statistical downscaling

Statistical downscaling (SD) establishes empirical relationships between large-scale atmospheric (predictors) and local (predictands) variables with large observational records.

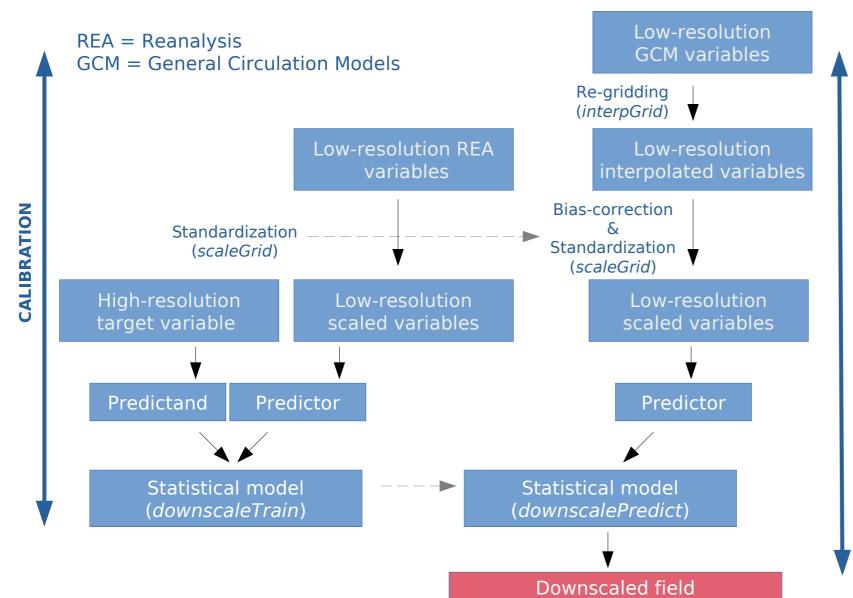






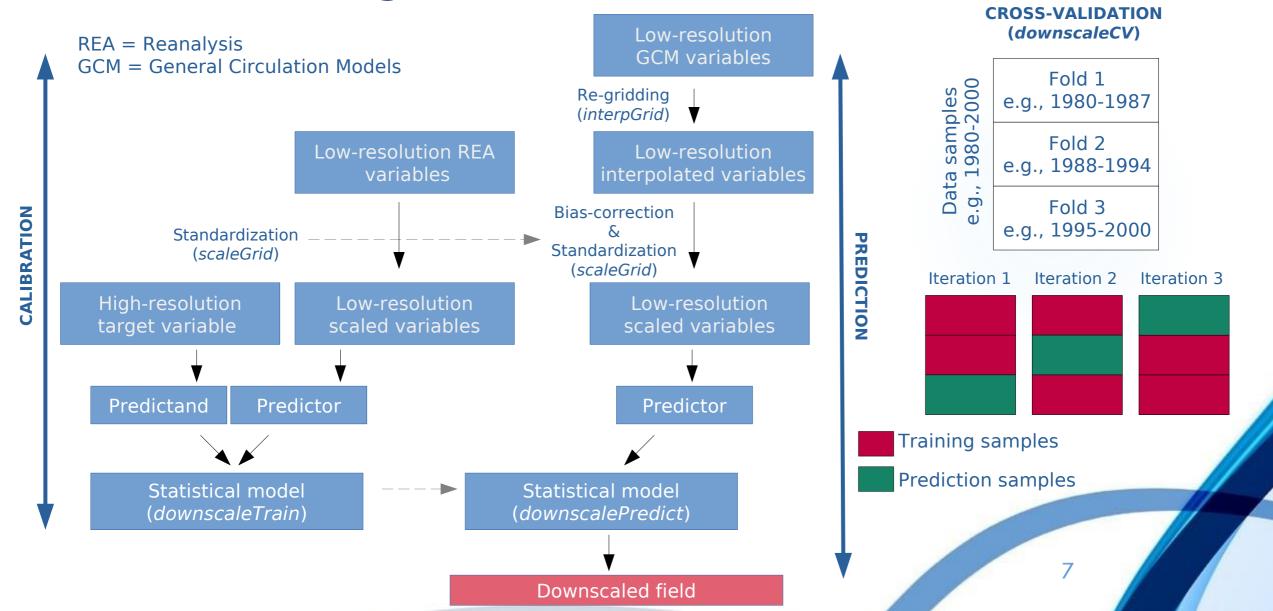




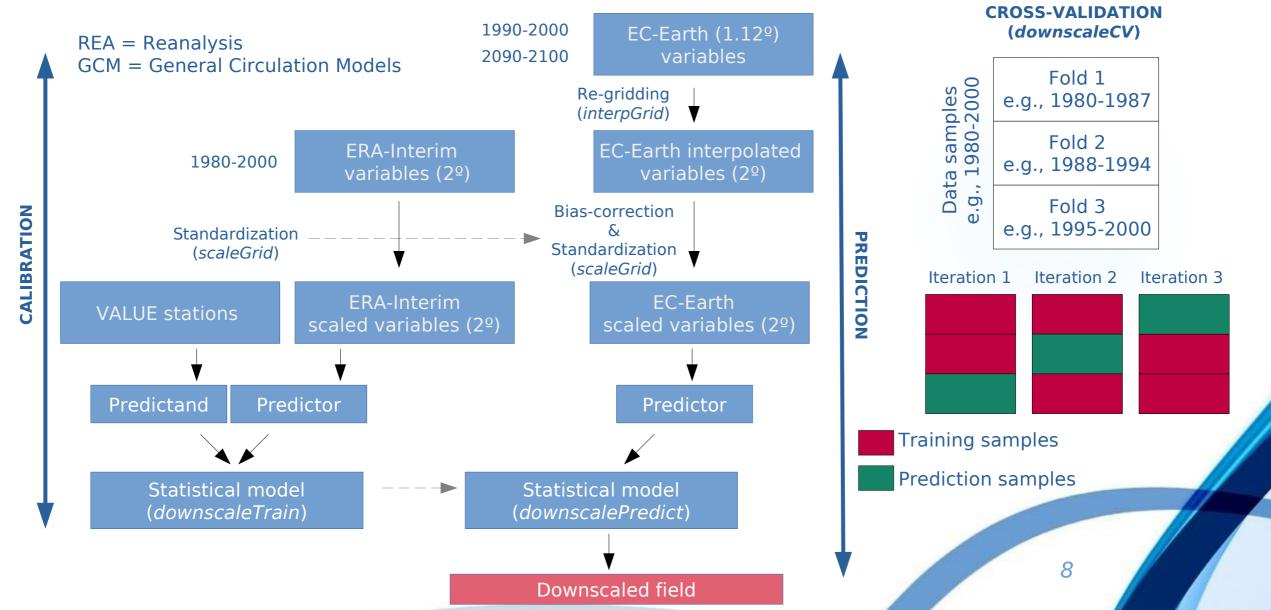


PREDICTION







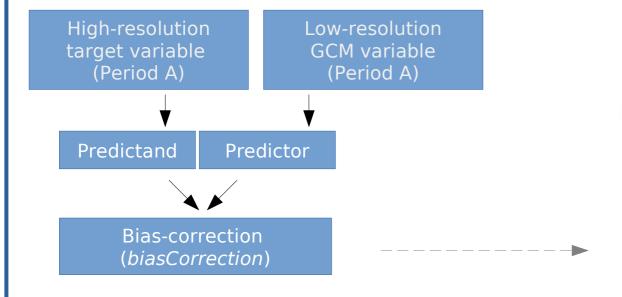


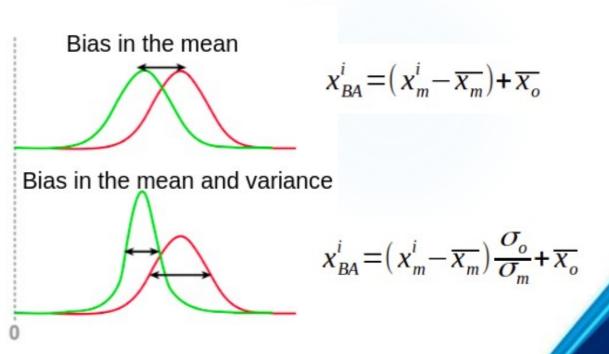


Bias-correction

GCM = General Circulation Models

Period A = temporal period within the historical scenario







Bias-correction

CALIBRATION

GCM = General Circulation Models Period A = temporal period within the historical scenario Period B = temporal period within the future scenario Low-resolution High-resolution Low-resolution target variable GCM variable GCM variable (Period A) (Period A) (Period B) **PREDICTION Predictand** Predictor Predictor Bias-correction Bias-correction (biasCorrection) (biasCorrection)