

Dynamical proxies, bias correction, and statistical downscaling as tools for seasonal forecast

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The MEDSCOPE Project

... will serve as a **community builder** for future **climate service** activities based on climate predictions in the **Mediterranean**, contributing to the building of a common and shared knowledge.



WP3: Forecast Calibration, Verification, and Information Synthesis



... will develop and release **advanced tools** to improve the extraction of relevant information from climate prediction systems and assess their robustness and uncertainty.

Copernicus Climate Change Service (C3S)

Seasonal Forecast

MEDSCOPE mainly focuses on the seasonal timescale using forecasts already available in C3S. Exploring also the potential of predictions at longer time-scales (multiannual).



The screenshot shows the Copernicus Climate Change Service (C3S) website. The header includes the Copernicus logo and the C3S logo. A navigation menu contains links for ABOUT C3S, NEWS & MEDIA, EVENTS, TENDERS, PRODUCTS, SERVICES, and HEL. The main content area is titled "Seasonal forecasts" and includes a breadcrumb trail "home > products". Below this, there are four panels: a line graph showing a blue line diverging into a red shaded area, a world map with green and yellow regions, a world map with yellow and orange regions, and a map of Europe with yellow and orange regions. Below the panels, there is a paragraph of text: "The Copernicus Climate Change Service (C3S) is developing seasonal forecast products, with a target publication date of 15th of each month. These products are based on data from several state-of-the-art seasonal prediction systems. The current proof-of-concept phase includes graphical forecast products for a number of variables (air and sea-surface temperature, atmospheric circulation and precipitation); the forecasts are updated every month and cover a time range of 6 months. The interface to the list of products offers links to maps or timeseries for the forecast variables, and the facility to navigate the full set of graphics. Multi-system combinations, as well as predictions from the individual component systems, are available. A number of multi-system data products, derived from the inputs provided by the participants in the C3S seasonal..."

Toolbox

- Developing methodologies to extract usable information from predictions. We will produce tools for prediction **verification, calibration, downscaling, ensemble member combination and selection** that will be publicly released via a **toolbox** and shared among partners and users.

[HOME](#)[THE PROJECT](#) ▾[TOOLBOX](#)[EVENTS](#)[PLATFORM](#)[CONTACTS](#)

MEDSCOPE

TOOLBOX

Home - TOOLBOX



Gitlab for MEDSCOPE project.

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Files (236 KB) Commits (4) Branch (1) Tags (0) Readme

master medscope / + 🔍 Find file History 📄

👤 Web address corrected in devtools::install_git f96cd89d 📄 nperez committed 2 days ago

Name	Last commit	Last Update
📁 R	Multivariate RMSE R function with documentation and vign...	3 days ago
📁 man	Multivariate RMSE R function with documentation and vign...	3 days ago



GitLab



python™

Toolbox

1) Bias correction and forecast calibration:

- Analogs using weather regimes



- ADAMONT: multivariate bias correction



- Forecast assimilation method for calibrating

- Dynamical based-bias correction



2) Statistical/stochastic downscaling

- Rainfarm: complex orography



- ADAMONT: multivariate downscaling



- Analogs



- SMOP: Mountainous regions



3) Advanced multivariable forecast scores:



- Multivariate RMSE

- s2d verification

- Indices multivariable EOFs



4) Forecast system combination and selection of sub-ensembles for applications

- ensemble calibration with selection and weighting



- weighting ensembles according NAO



- clustering techniques. EOFs and K-means



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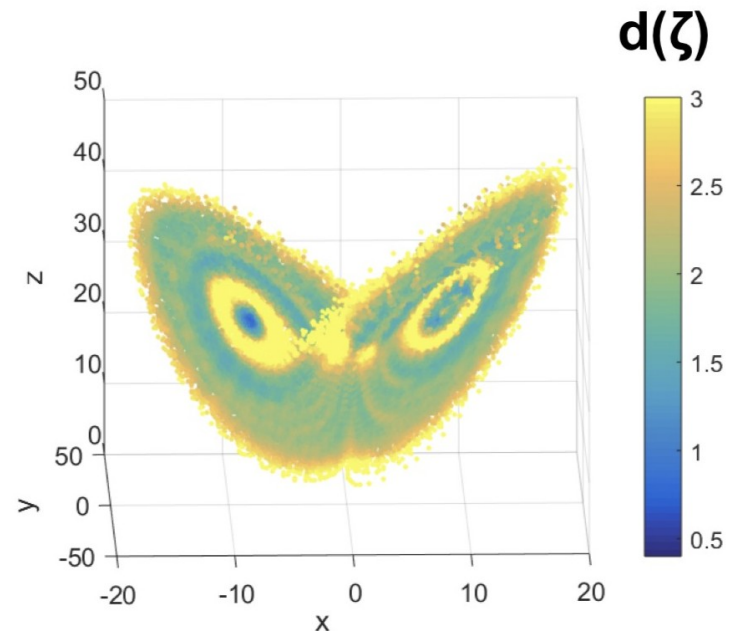


Dynamical Systems

Compute Dynamical Systems metrics to characterize atmospheric states, verifying that a long series of observations sample the underlying attractor.

Local Dimensions d

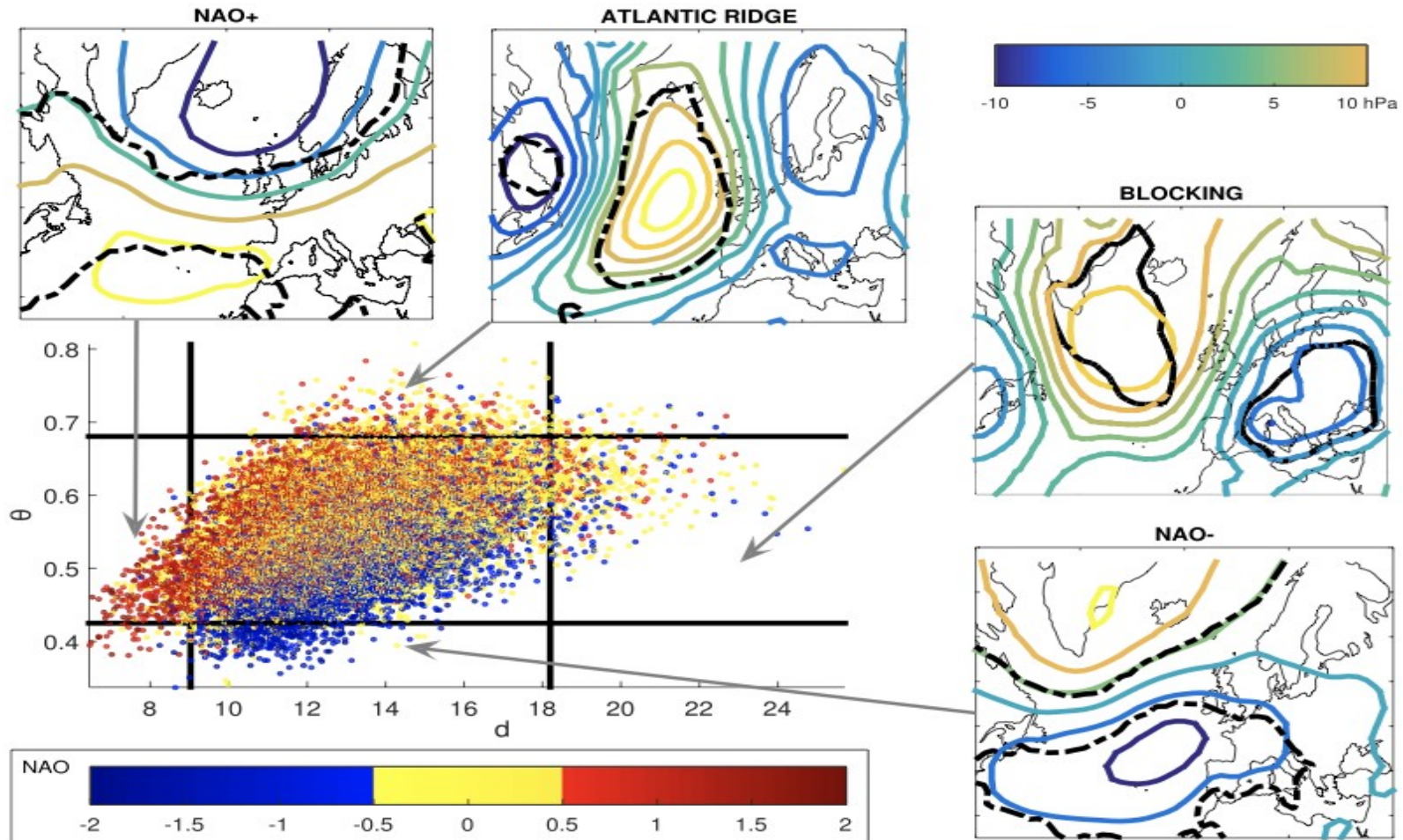
t is proportional to the number of possible configurations (**number of degrees of freedom**) originating and resulting from the atmospheric field analyzed.



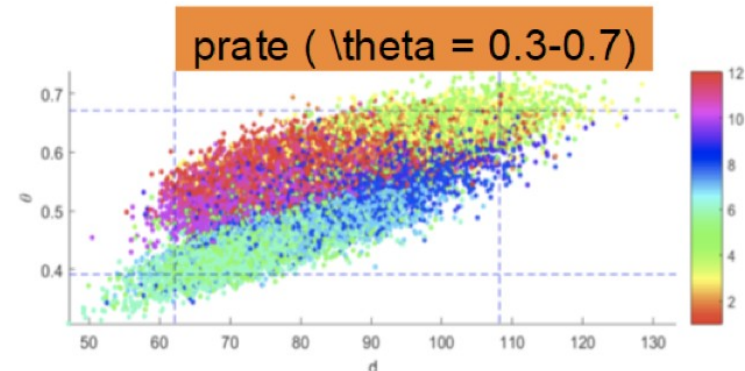
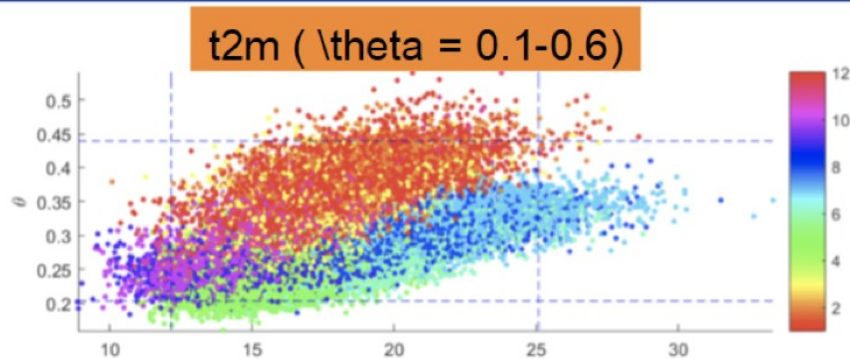
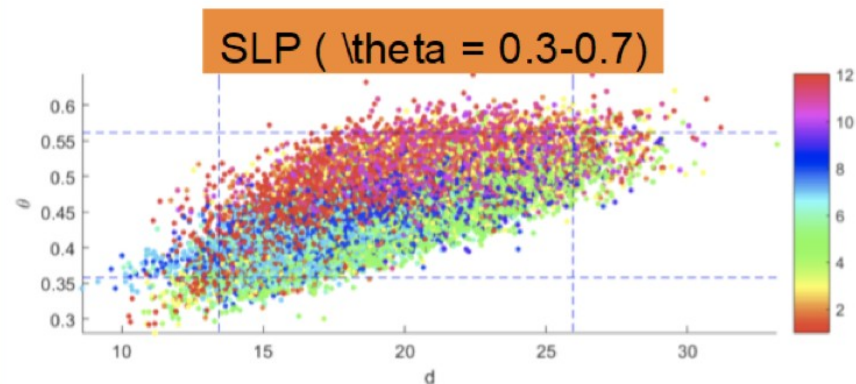
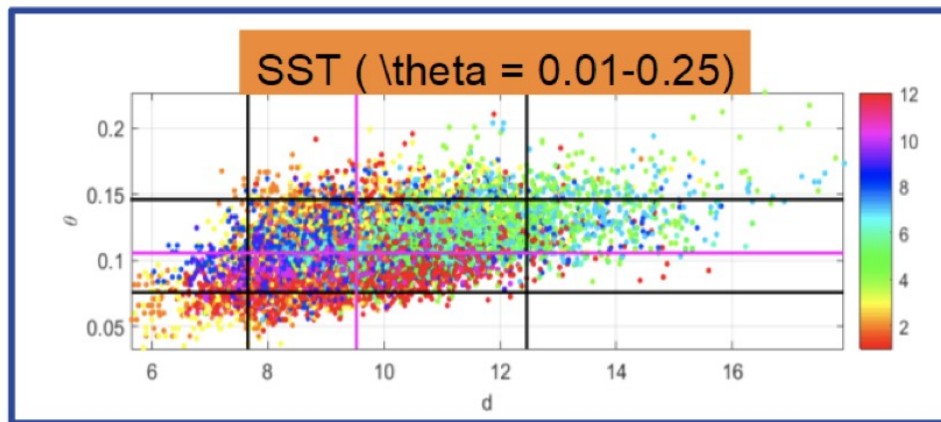
Persistence Θ

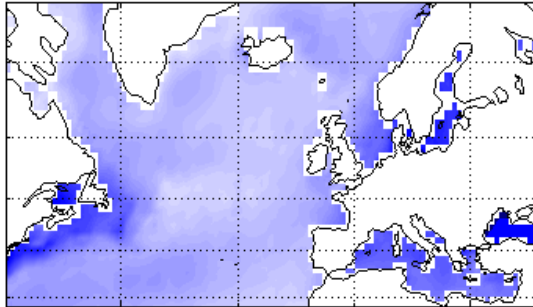
its inverse tells for **how long the atmospheric field will look like the one under examination**. For the present analysis Θ is an inverse number of persistence days.

SLP attractor for winter NCEP data. Low values of Theta and dimension are the most predictable configurations and represents NAO+/-, high values on Theta and dimension are the most unpredictable configurations and represents in SLP Blocking and AR.

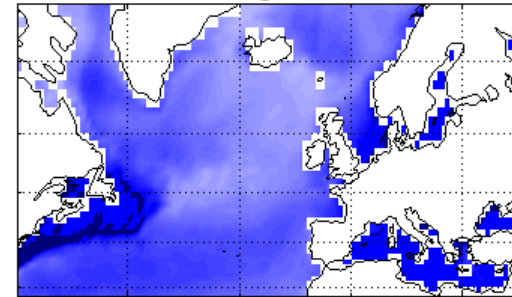


North Atlantic Attractor for different variables (SST, SLP, t2m, prate). Theta values for SST are the smallest in comparison with the rest of variables which means more predictability and the reason to use this variable to apply the bias correction method.

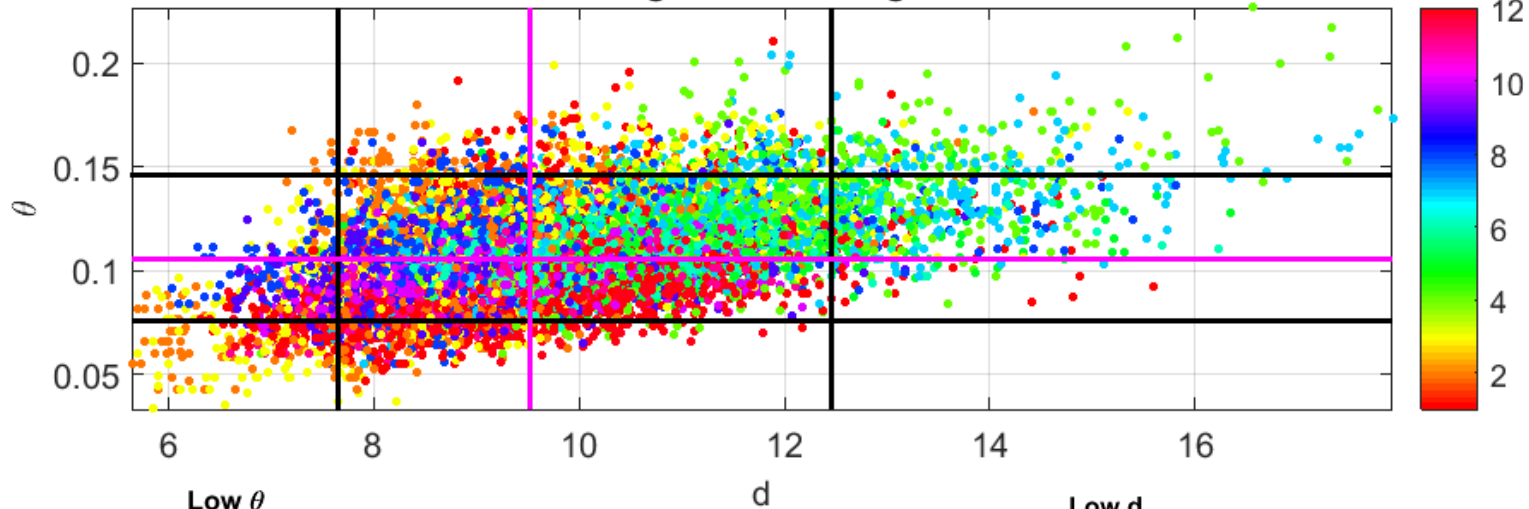
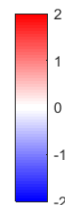
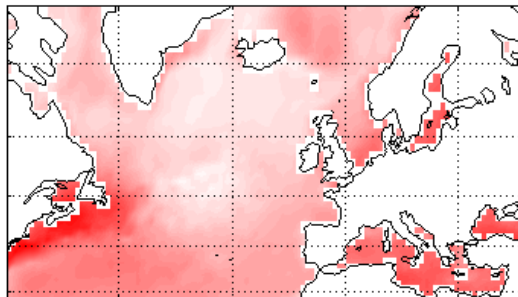


High θ (Anomalies seasonal cycle)

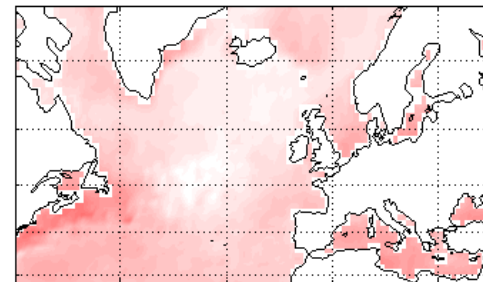
High d



NOAA sst High Res 0.25 deg 1982-2017

Low θ 

Low d

Alvarez-Castro, in
prep

Definition

A method to apply multivariate bias correction to seasonal forecast data.

It is based on a quantile mapping method for statistical adjustment of climate simulations, but also uses information from dynamical systems (Faranda et al., 2017) based on persistence and local dimension of the predictors in the North Atlantic region in order to correct the bias on precipitation and temperature in the Mediterranean area.

Data and method

- observations from NCEP and GPCP - 40 members CMCC SPSv3

- dynamical-conditioned BC method:

using an Index or "weather regimes" in a climate/seasonal model.

Quantile map is close to the observations but takes into account just the statistics, this method take into account the interannual variability that should produce time series much more realistic.

Obs: 1982-2017 (SST, Precipitation) → (NOAA, GPCC)

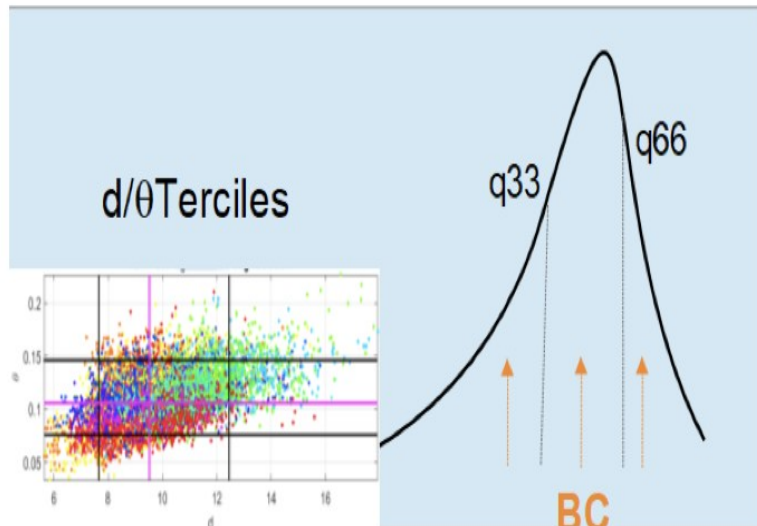
1982

Hindcast Period

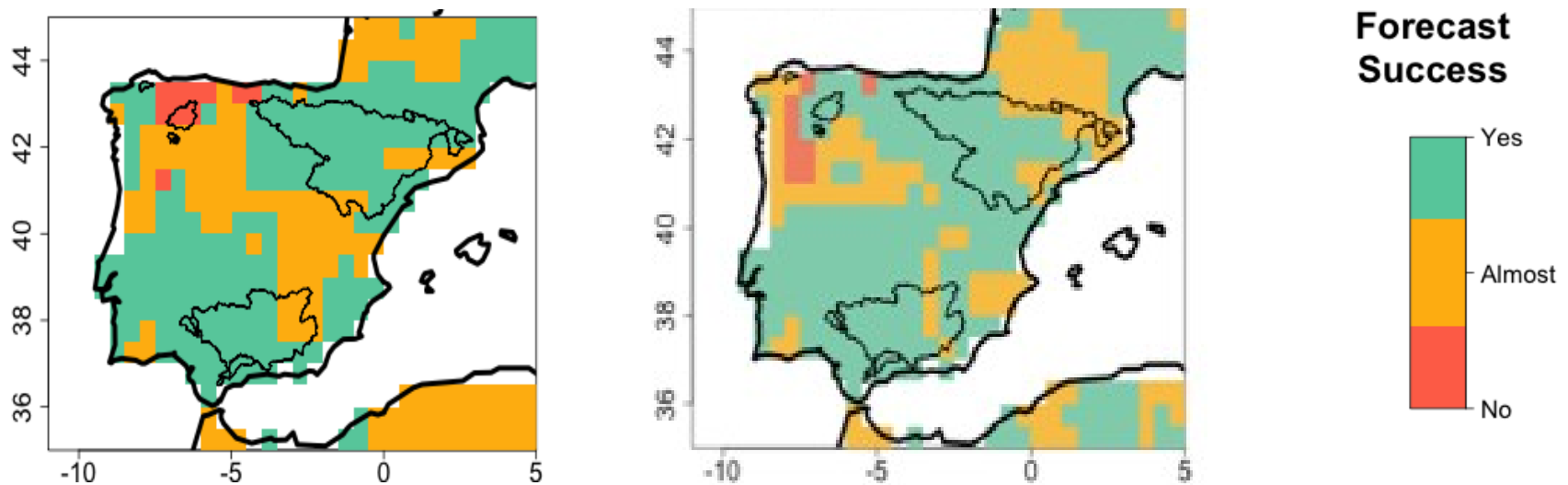
2010

2017

Forecast Period



Calibration: NCEP and CMCC-CESM model

Example for September 2018 (NMME) :

Functions for MEDSCOPE Toolbox: Bias Correction



1. R functions for the computation of d and Θ :

- a. To compute local dimension and persistence
- b. Results Visualization Tools

2. R functions for the bias correction:

- a. Quantiles based on local dimension and θ
- b. Existing quantile map functions (qmap)
- c. Computation of probabilities for ensembles

3. R functions for visualization of verification:

- a. Traffic lights figures for verification
- b. Contingency tables

North Atlantic and Mediterranean regions

Trained using:

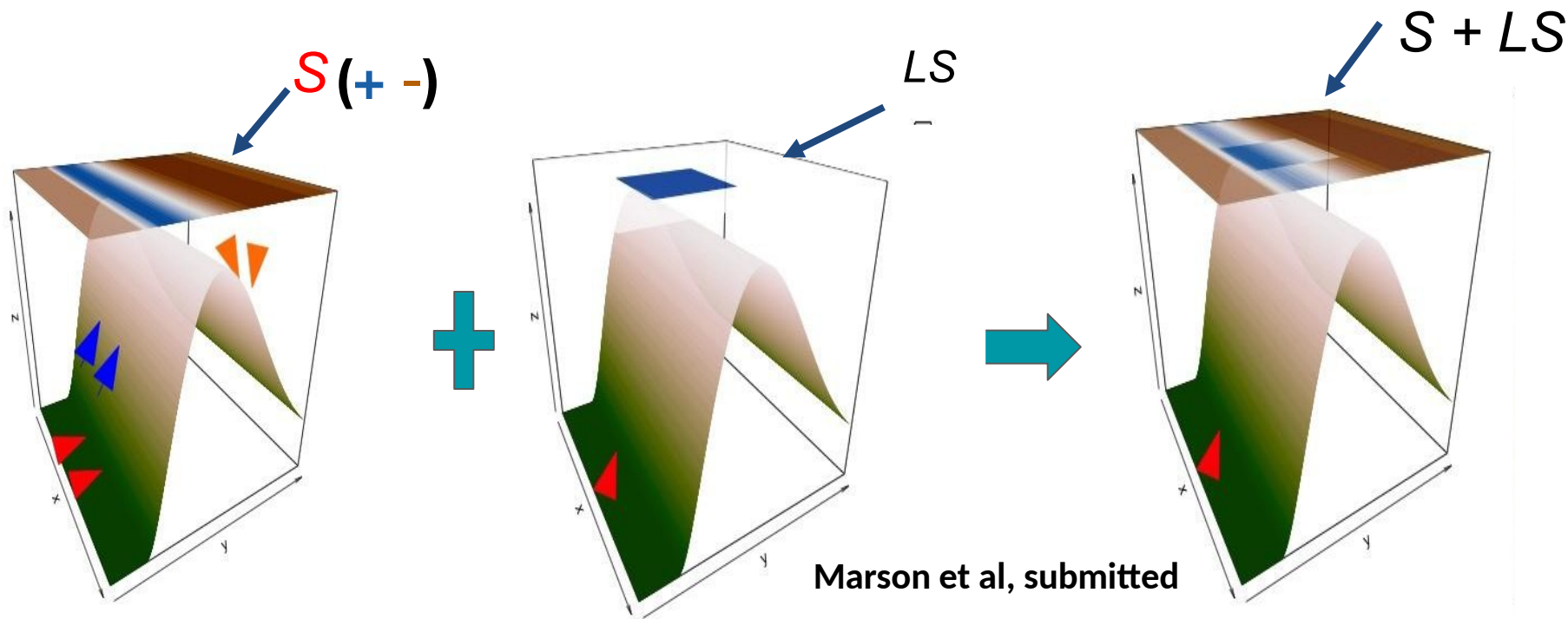
- SST NOAA high res
- NCEP
- EOBS
- CMCC-CM model
- Seasonal Forecast
- C3S-CMCC

A Process-Informed Statistical Model of Orographic Precipitation (SMOP)

1. Spatialization of precipitation fields in mountainous areas
2. Predictive Downscaling of Climate Models

Small Scale orographic source of condensation computed analitically [Smith, 2003]

Large Scale source of precipitation (frontal systems, l-s convection)

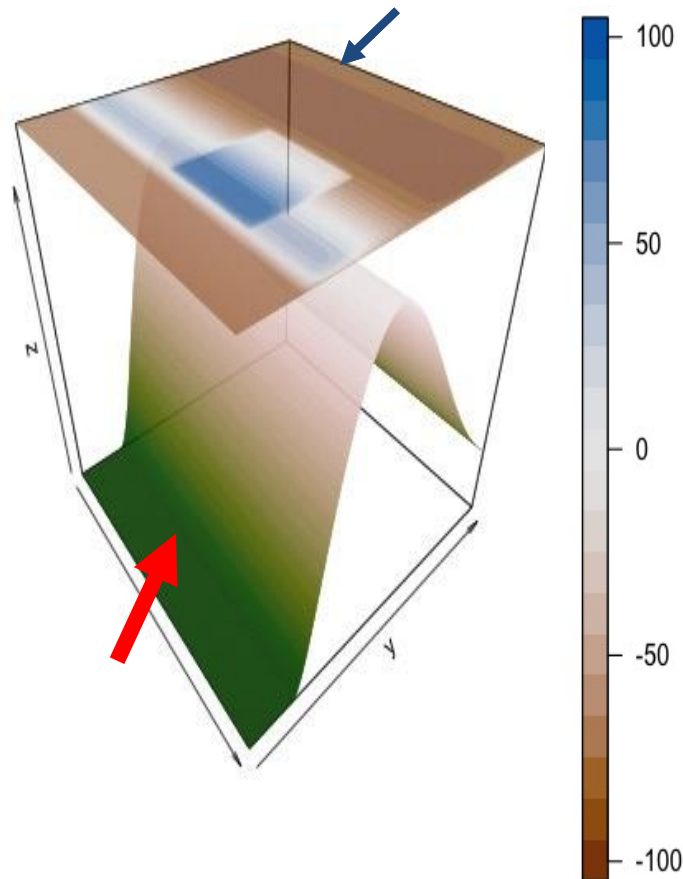


- Time scale:**

$$\tau_f = 1000s$$

The total source is then advected by the horizontal wind before precipitate on the ground
 example of the spatial pattern of “precipitation potential” produced advecting the total source for 5000, 10000, 20000, 50000 seconds

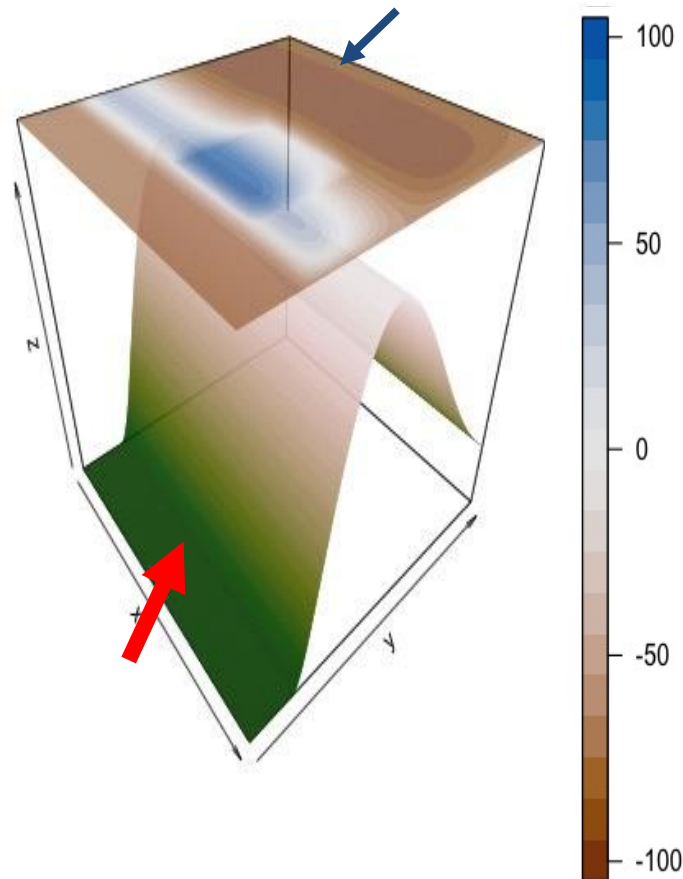
ρ^* (potential precipitation $\frac{mm}{day}$)



- **Time scale:**

$$\tau_f = 5000\text{s}$$

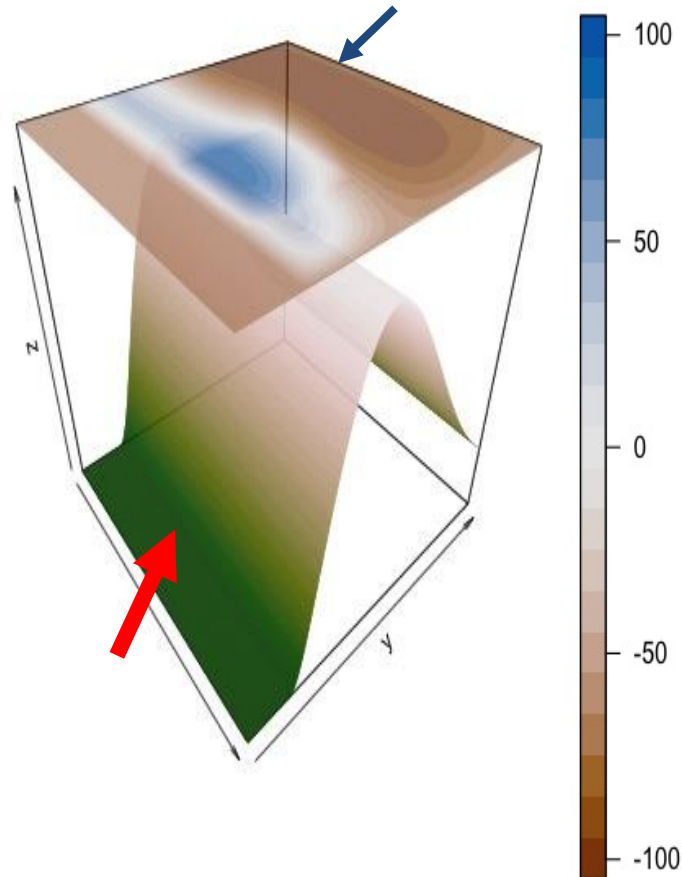
ρ^* (potential precipitation $\frac{\text{mm}}{\text{day}}$)



- **Time scale:**

$$\tau_f = 10000\text{s}$$

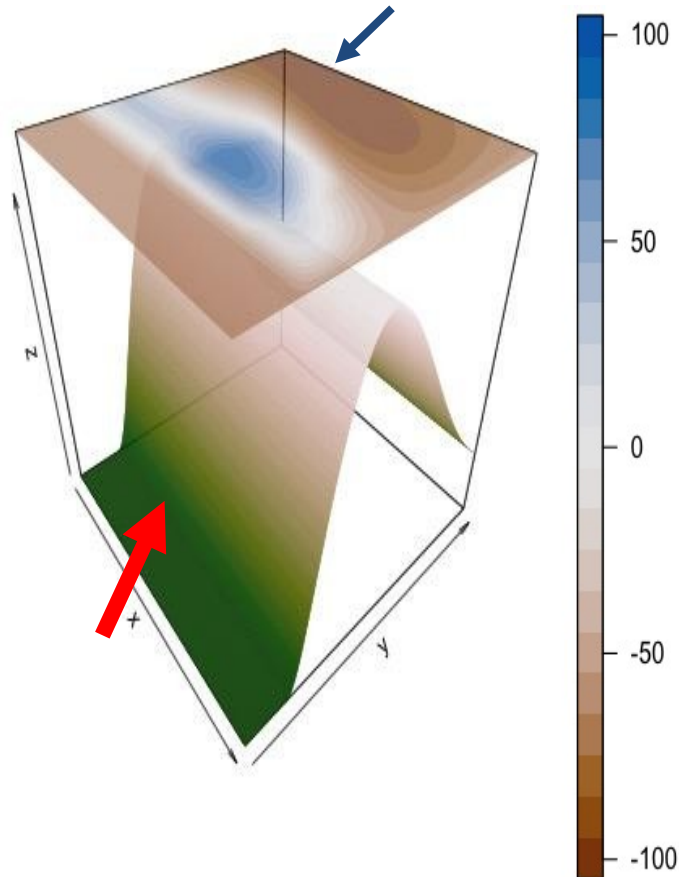
p^* (potential precipitation $\frac{\text{mm}}{\text{day}}$)



- **Time scale:**

$$\tau_f = 20000\text{s}$$

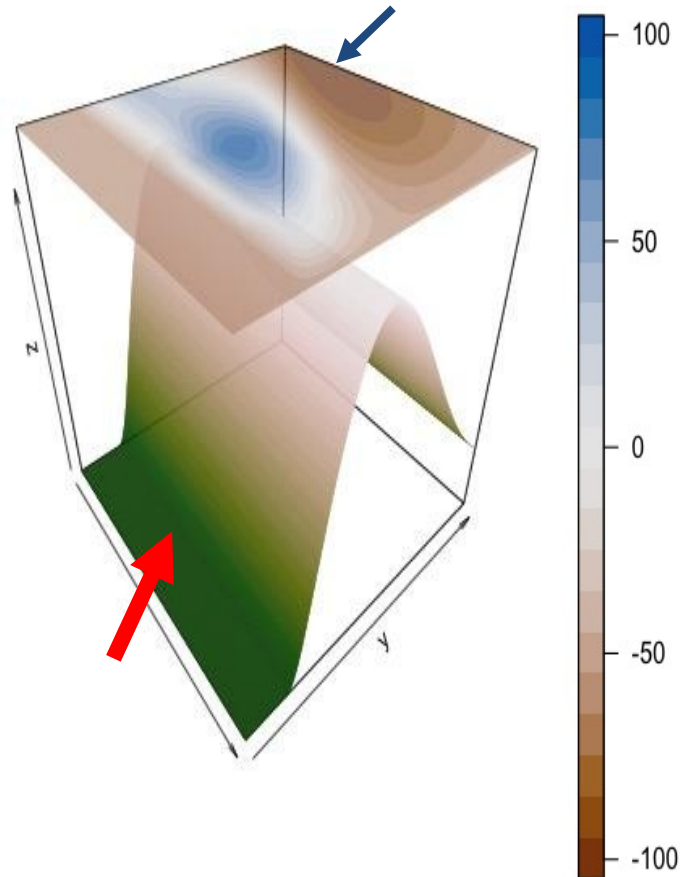
ρ^* (potential precipitation $\frac{\text{mm}}{\text{day}}$)

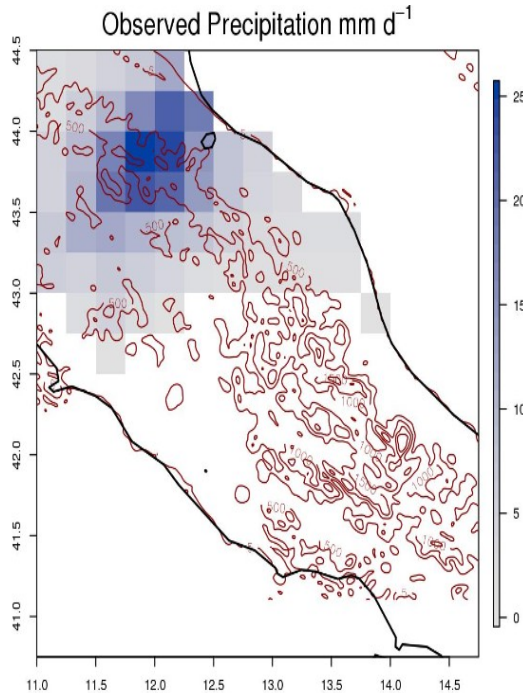


- **Time scale:**

$$\tau_f = 50000s$$

p^* (potential precipitation $\frac{mm}{day}$)





E-OBS (Haylock et al. 2008)
.25 x .25 deg Regular Grid

A Real Case Study Over Central I

Domain and Data

- Terrain: GTOPO30 (95% variance)
- ERA-Interim: (U, T, RH, LS-prec)

20

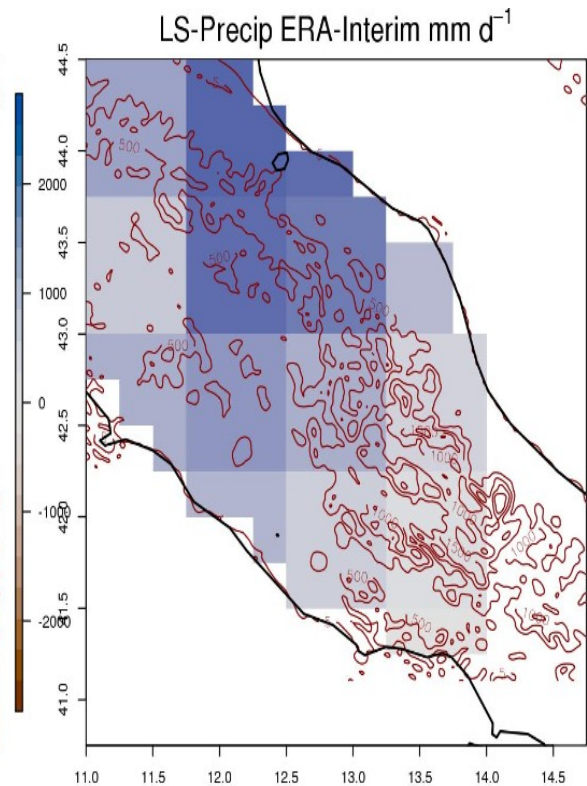
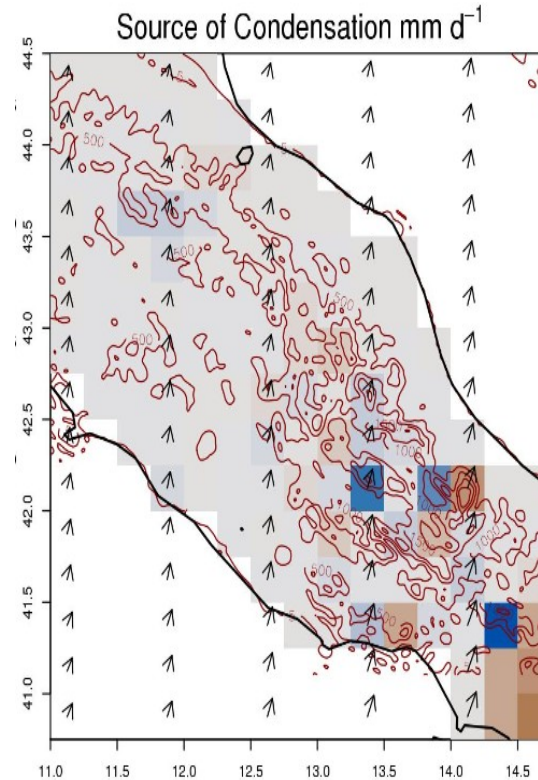
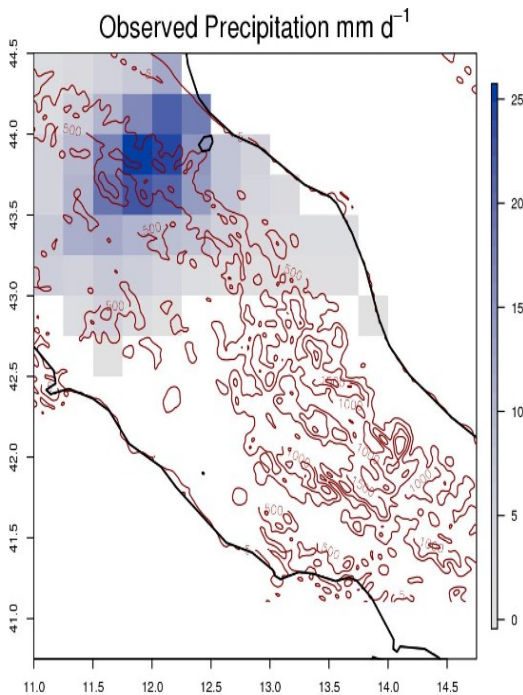
15

10

5

0

- Positive (blue) where condensation occurs and
- negative (brown) where instead evaporation occurs



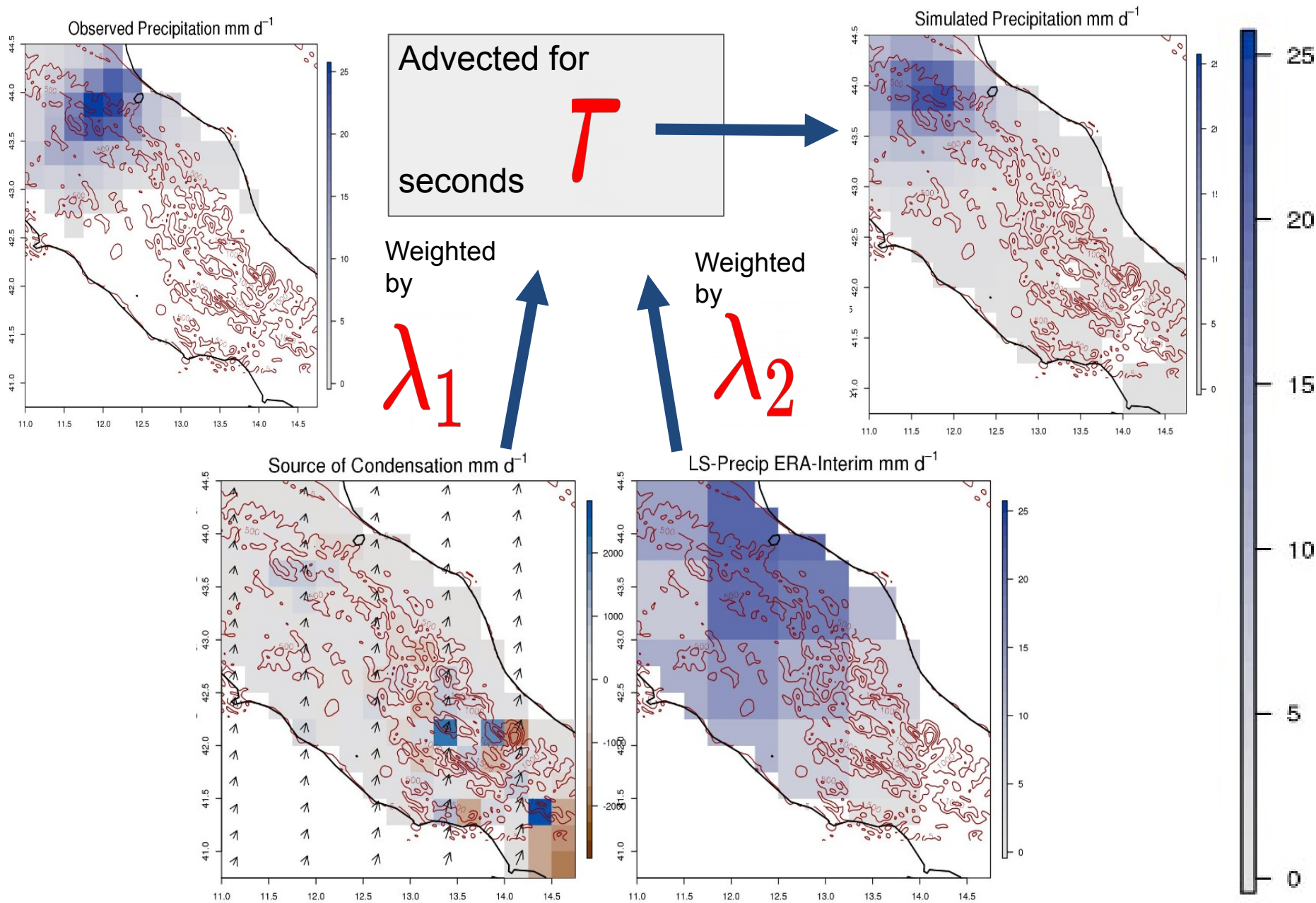
20

15

10

5

0



Predictive Downscaling

 λ_1, λ_2

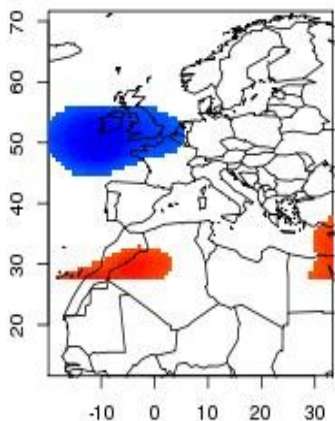
Can be predicted in the future by exploiting their relation with large-scale fields

geopotential height λ_2

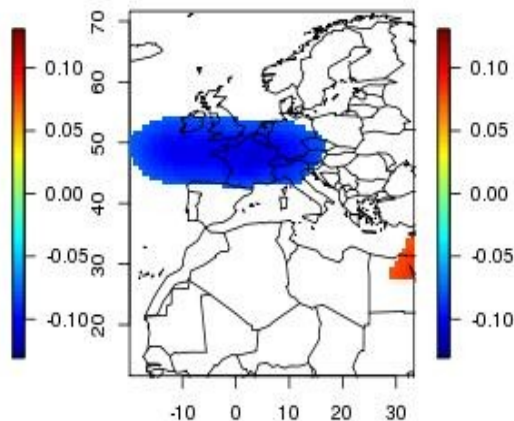
 λ_1

relative humidity

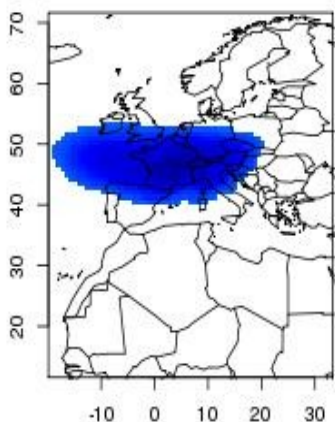
Correlation with Z at 500hPa



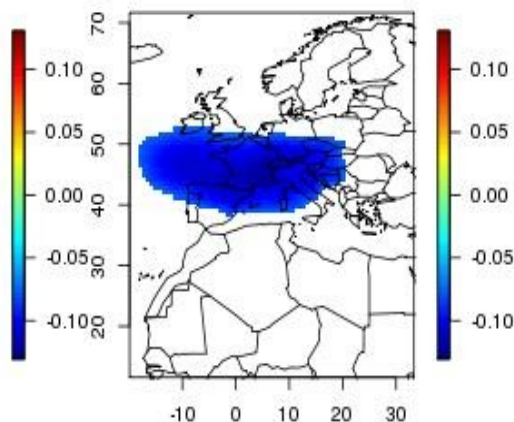
Correlation with Z at 700Pa



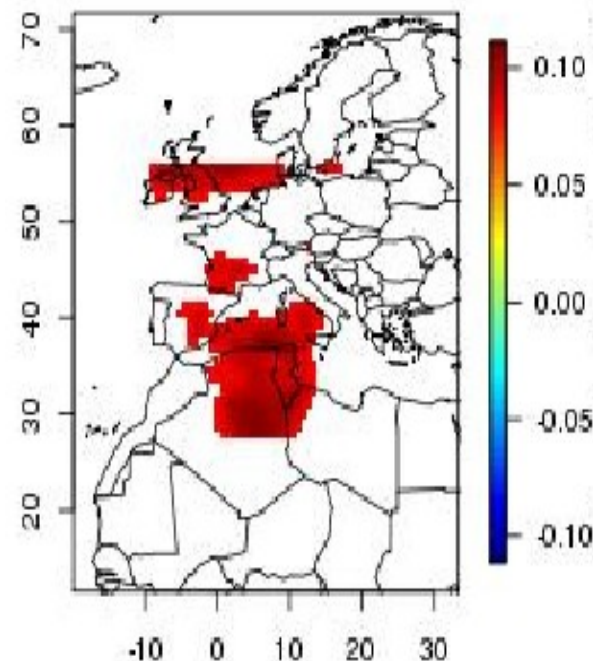
Correlation with Z at 850hPa



Correlation with Z at 1000hPa



Correlation with RH at 850hPa



Functions for MEDSCOPE Toolbox: Statistical Downscaling



1. R functions for the spatialization of precipitation:

- a. To compute local sources of condensation;
- b. To estimate $\lambda_1, \lambda_2, \tau$ from data (3 versions):
 - i. For precipitation occurrence only
 - ii. For precipitation occurrence + power transform to obtain intensities
 - iii. For precipitation occurrence and intensities simultaneously
- c. To apply the spatialization on the desired grid
- d. Results Visualization Tools

2. R functions for the predictive downscaling:

- a. To search linkages between $\lambda_1, \lambda_2, \tau$ and large-scale fields
- b. To predict in time $\lambda_1, \lambda_2, \tau$ using an “analog approach”
- c. To predict in time and spatialize the precipitation field.
- d. Results Visualization Tools

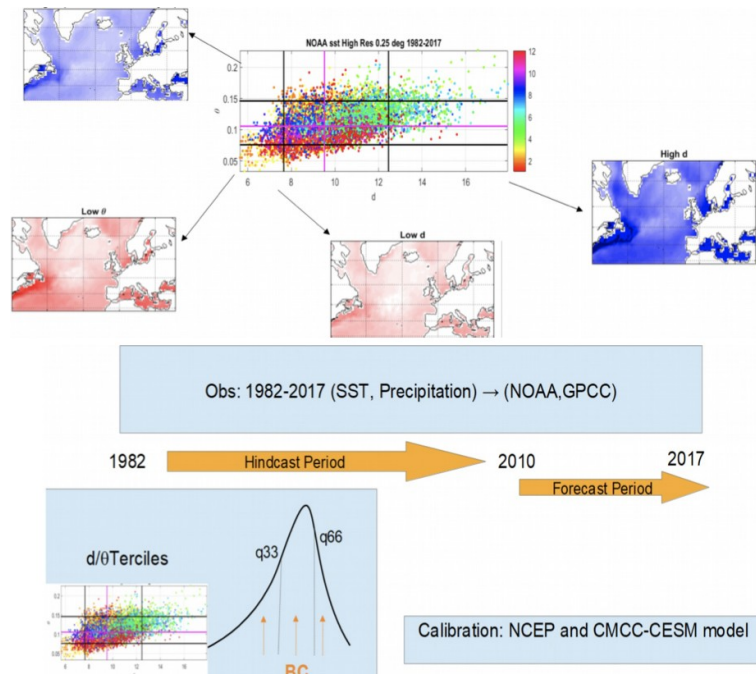
3. SMOP already trained on three geographical areas, ready for applying spatialization/downscaling:

Central Italy; Pyrennes; Southern France

Trained using:
EOBS;
ERA-Interim
Seasonal Forecast
C3S-CMCC

Bias Correction:

A new dynamical-conditioned bias correction method for seasonal forecast on precipitation and temperature in the Mediterranean Region



- New tool for WP3 Medscope toolbox developed in R and available in the upcoming months.

- We make use of recent advances in dynamical systems theory to estimate **two instantaneous dynamical properties** of the SST fields for the North Atlantic sector and the Mediterranean region: **local dimension and persistence** (Faranda et al, 2017).

- This tool combine dynamical to statistical information in order to better apply a bias correction method (as in Manzananas et al, 2018)

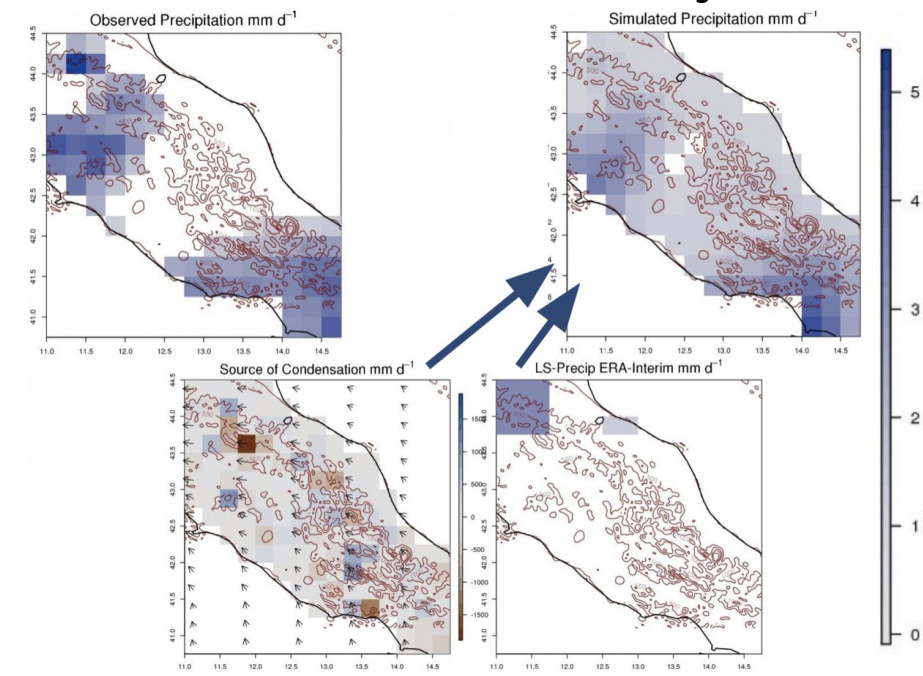
- SST in North Atlantic give us values of predictability (in terms of dynamical systems) higher than for SLP.

- **warming ocean seems to lead to more predictable configurations**

Statistical Downscaling:

A process-informed statistical framework for precipitation in mountainous regions

Winter storm in Central Italy



Sub-grid refinement by combining

- **Local scale processes causing orographic rainfall (analytical)**
- **Large scale precipitation component (from climate models)**

in a spatial autoregressive framework.

The relative contribution of local and large-scale sources is adjusted by observations. The approach may be used as kernel for predictive downscaling techniques.

Finding linkages between the parameters λ_1 , λ_2 , τ , and the large-scale fields it is possible to predict their values in the future using maps of the selected large scale fields.

Thank you



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