# HyMeX



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# Impacts of climatic oscillations on Mediterranean hydrology

<u>Wouter Dorigo<sup>1</sup>, Laura Crocetti<sup>1</sup>, Brecht Martens<sup>2</sup>,</u> Agnes Kauer<sup>1</sup>, Bernhard Bauer-Marschallinger<sup>1</sup>, Angelika Xaver<sup>1</sup>, Diego Miralles<sup>2</sup>, Victor Pellet<sup>3</sup>, Filipe Aires<sup>3</sup>

- 1) CLIMERS- Research Group Climate and Environmental Remote Sensing, Department of Geodesy and Geoinformation, TU WIEN, Vienna, Austria
- 2) Laboratory of Hydrology and Water management, Ghent University, Ghent, Belgium
- 3) Estellus, Observatoire de Paris, Paris, France

http://climers.geo.tuwien.ac.at/

wouter.dorigo@tuwien.ac.at

Picture: © NASA Visible Earth



### Atlantic Meridional Overturning Circulation (AMOC)

- Global warming is expected to weaken the AMOC due to melting of Greenland ice cap [Cheng et al., 2013; Rahmstorf et al., 2015]
- A slowing down of the AMOC would alter the course and amplitude of the Gulfstream and, hence, the path of Atlantic storm tracks [Joyce and Zhang, 2010]



![](_page_1_Picture_4.jpeg)

red: NAO, black: AMOC

![](_page_1_Figure_6.jpeg)

[Cheng et al., 2013]

![](_page_1_Picture_8.jpeg)

#### **Climate modes as drivers of climate variability**

North Atlantic Oscillation is one of the main drivers of Mediterranean climate variability

 NAO index is based on the difference of normalized sea level pressure (SLP) between Lisbon, Portugal and Stykkisholmur/Reykjavik, Iceland

![](_page_2_Figure_3.jpeg)

![](_page_2_Figure_4.jpeg)

#### **Clear link between NAO and gauge-based precipitation**

![](_page_3_Figure_1.jpeg)

Correlation between *CRU* precipitation and *NAO* for the winter season (1949-1996)

![](_page_3_Picture_3.jpeg)

[Mariotti et al., 2001]

Correlation between the winter *NAO* index and E-OBS precipitation (1950-2010)

![](_page_3_Picture_6.jpeg)

![](_page_3_Picture_7.jpeg)

# **Study objective**

Can our understanding of climate mode impacts on precipitation be improved based on satellite observations?

#### **Challenges:**

- Can we disentangle the individual impact of co-varying multiple climate modes?
- Uncertainties in the observations
- Climate modes may have impacts at various lags

![](_page_4_Figure_6.jpeg)

AMMSST: Atlantic Meriodonal Mode SST TNA: Tropcial Northern Atlantic

![](_page_4_Picture_8.jpeg)

![](_page_4_Picture_9.jpeg)

#### Target variables: WACMOS-MED multiple satellite products

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- Provides optimised estimates of water balance terms (P, ET, ΔTerrestrial Water Storage)
- Weighted sum of various precipitation datasets (CMORPH, PERSIANN, GPCP, ...)
- Constrained product  $\rightarrow$  closes the water budget at sub-basin scale
- INTegration estimate (limited temporal extent: 2004-2009)
- CALibration estimate (long temporal coverage: 1980-2012)

#### Advantages:

- minimizes uncertainties
- consistent  $\rightarrow$  better solves the water budget
- long time coverage

![](_page_5_Figure_11.jpeg)

WACMOS

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temporal resolution: monthly spatial resolution: 0.25° detrended anomalies

#### **Predictor variables: Climate Oscillation Indices**

#### Temporal resolution: **monthly** Considered lag times: **0-5 months**

![](_page_6_Figure_2.jpeg)

![](_page_6_Picture_3.jpeg)

**102 features** 

#### **CLIMATE OSCILLATION INDICES**

Atlantic Meridional Mode (AMMSST) Atlantic Multidecadal Oscillation (AMO) Dipole Mode Index (DMI) East Atlantic (EA) East Atlantic / Western Russia (EAWR) East Pacific-North Pacific (EPNP) Northern Annular Mode (NAM) North Atlantic Oscillation (NAO) Pacific Decadal Oscillation (PDO) Polar / Eurasia (**PEA**) Pacific / North American Index (PNA) Southern Annular Mode (SAM) Scandinavia (SCAND) Southern Oscillation Index (SOI) Tropical Northern Atlantic (TNA) Tropical Southern Atlantic (TSA) Western Pacific (WP)

#### LASSO - Least Absolute Shrinkage and Selection Operator

Machine learning regression method

$$\beta = \arg\min\{\sum_{i=1}^{n} (y \downarrow_i - \beta \downarrow_0 - \sum_{j=1}^{n} p = \beta \downarrow_j x \downarrow_i j \ )^2 + \alpha \sum_{j=1}^{n} p = \beta \downarrow_j j \ \}$$

- $\beta$ ... p-dimensional vector with the estimated regression coefficients
- $n \dots$  number of training samples in the dataset
- $\mathcal{V}$ *i* ... value of the target variable in sample i
- $\mathcal{P}$  ... number of features

 $\chi i j \dots$  value of feature j in sample i  $\alpha \sum j = 1 \uparrow p = |\beta \downarrow j| \dots$  regularization that minimizes the sum of the absolute values of the coefficients  $\rightarrow$  prevents overfitting of the model by reducing its complexity

- Nested 5-fold cross validation (CV)
  - 5-fold CV for determination of α
  - 5-fold CV for calculation of  $\mathbf{R}^2$  (= coefficient of determination) between y and y  $\downarrow pred$
- Significance test

![](_page_7_Picture_13.jpeg)

![](_page_7_Picture_14.jpeg)

#### Total explained variance in P anomalies based on all-year data

(1982-01 - 2012-12)

![](_page_8_Figure_2.jpeg)

![](_page_8_Picture_3.jpeg)

R<sup>2</sup>

0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40

![](_page_8_Picture_5.jpeg)

#### Total explained variance for seasonal models (1982-01 – 2012-12)

Dec Jan Feb

Jun Jul Aug

![](_page_9_Picture_3.jpeg)

#### **Total explained variance of combined seasonal models**

![](_page_10_Picture_1.jpeg)

![](_page_10_Picture_2.jpeg)

R<sup>2</sup>

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0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40

![](_page_10_Picture_4.jpeg)

#### **Temporal dynamics**

![](_page_11_Figure_1.jpeg)

#### **COI coefficients – all-year data**

![](_page_12_Figure_1.jpeg)

#### **COI coefficients – Dec-Jan-Feb model**

![](_page_13_Figure_1.jpeg)

## **Impact of lags**

Coefficients for NAO with monthly time lags (Dec-Jan-Feb model) 

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![](_page_14_Figure_2.jpeg)

![](_page_14_Picture_3.jpeg)

![](_page_14_Picture_4.jpeg)

![](_page_14_Picture_5.jpeg)

### Are Earth Ooservations a valuable alternative to gauge-based data?

- LASSO regression winter (DJF) precipitation
- Period 1982-2012

**E-OBS** 

#### WACMOS-MED P\_CAL

![](_page_15_Figure_5.jpeg)

![](_page_15_Picture_6.jpeg)

![](_page_15_Picture_8.jpeg)

![](_page_16_Figure_1.jpeg)

![](_page_16_Picture_2.jpeg)

![](_page_16_Picture_3.jpeg)

![](_page_17_Figure_1.jpeg)

![](_page_17_Picture_2.jpeg)

![](_page_17_Picture_3.jpeg)

![](_page_18_Figure_1.jpeg)

![](_page_18_Picture_2.jpeg)

![](_page_18_Picture_3.jpeg)

Impact of model cross validation

Squared correlations NAO - E-OBS (1950-2010)

![](_page_19_Figure_3.jpeg)

![](_page_19_Picture_4.jpeg)

![](_page_19_Picture_5.jpeg)

LASSO regression E-OBS without nested CV

#### Conclusions

- LASSO Regression can be used to disentangle the impact of multiple, potentially correlated COIs on Mediterranean precipitation
- All-year and seasonal models show deviating importance of individual climate modes
- In the winter months (DJF), precipitation anomalies primarily driven by:
  - North Atlantic Oscillation
  - East Atlantic Oscillation
  - East Atlantic West Russia Pattern
- Earth observations are suitable for this purpose
  - But, are limited by the temporal coverage

Thank you for your attention

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![](_page_20_Picture_11.jpeg)

wouter.dorigo@tuwien.ac.at

![](_page_20_Picture_13.jpeg)

#### coefficients – seasonal Dec-Jan-Feb

![](_page_21_Figure_1.jpeg)

![](_page_21_Picture_2.jpeg)

#### **Coefficients – entire time period**

![](_page_22_Figure_1.jpeg)