

Looking for Local Monthly Wave Climate Indices

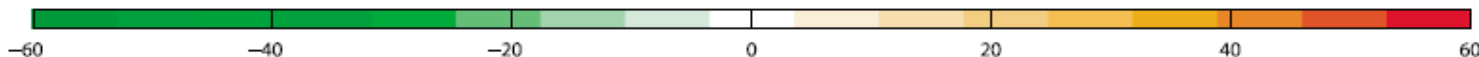
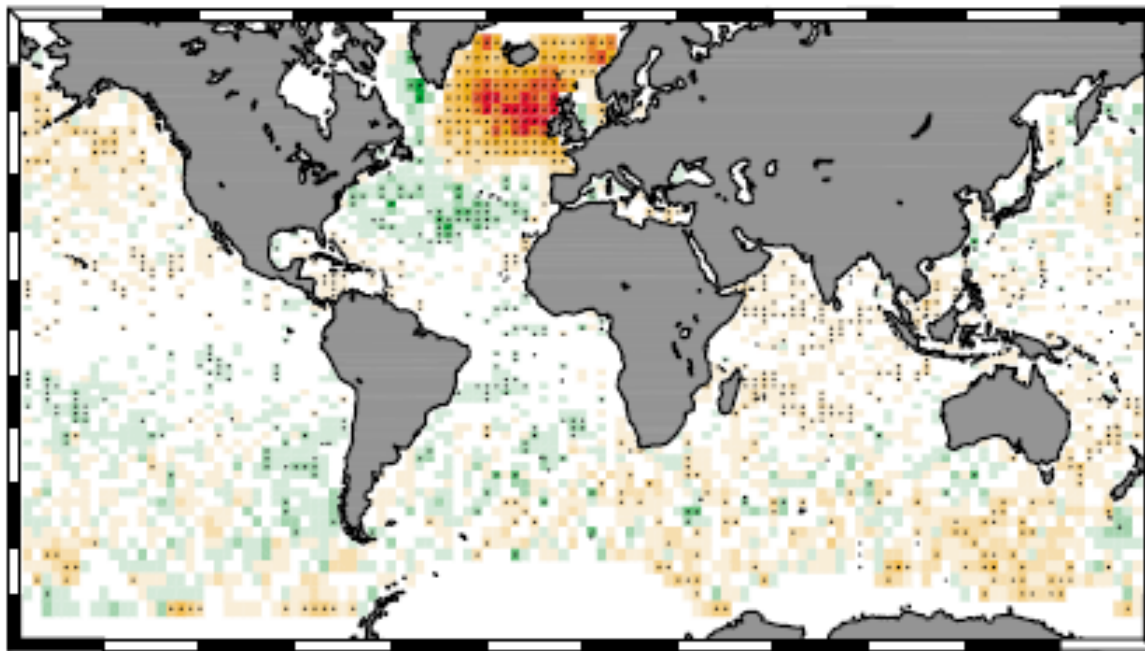
Paula Camus, Fernando Méndez, Iñigo Losada, Melisa Menéndez,
Antonio Espejo, Jorge Pérez, Ana Rueda, Yanira Guanche

Wave Climate and Climate Change Group
Environmental Hydraulics Institute “IHCantabria”

Well-known climate indices (PNA, SOI, Niño3, NAO, AO, SAM,..) have been used to explain wave climate variability

Gulev and Griegorieva (2006)
Hemer (2010)
Izaguirre et al. (2010, 2011)
Sterl and Caires (2005)
Wang and Swail (2001)
Caires et al.(2006)
Semedo et al. (2011)
Stopa et al.(2012)

(e) β_{NAO}

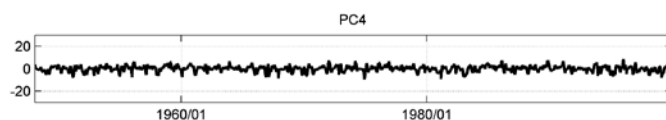
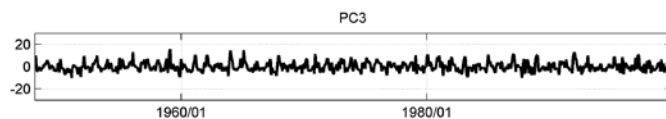
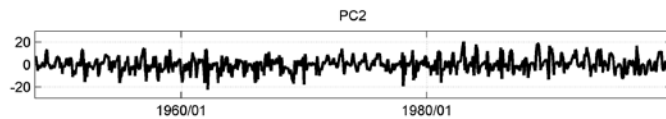
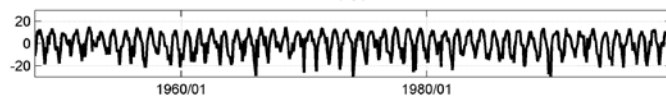


Spatial variability of the influence of climate indices in the location parameter of the wave extreme model
(Izaguirre et al., 2011)

To find temporal coefficients of modes of atmospheric variability strongly linked to wave climate at particular location



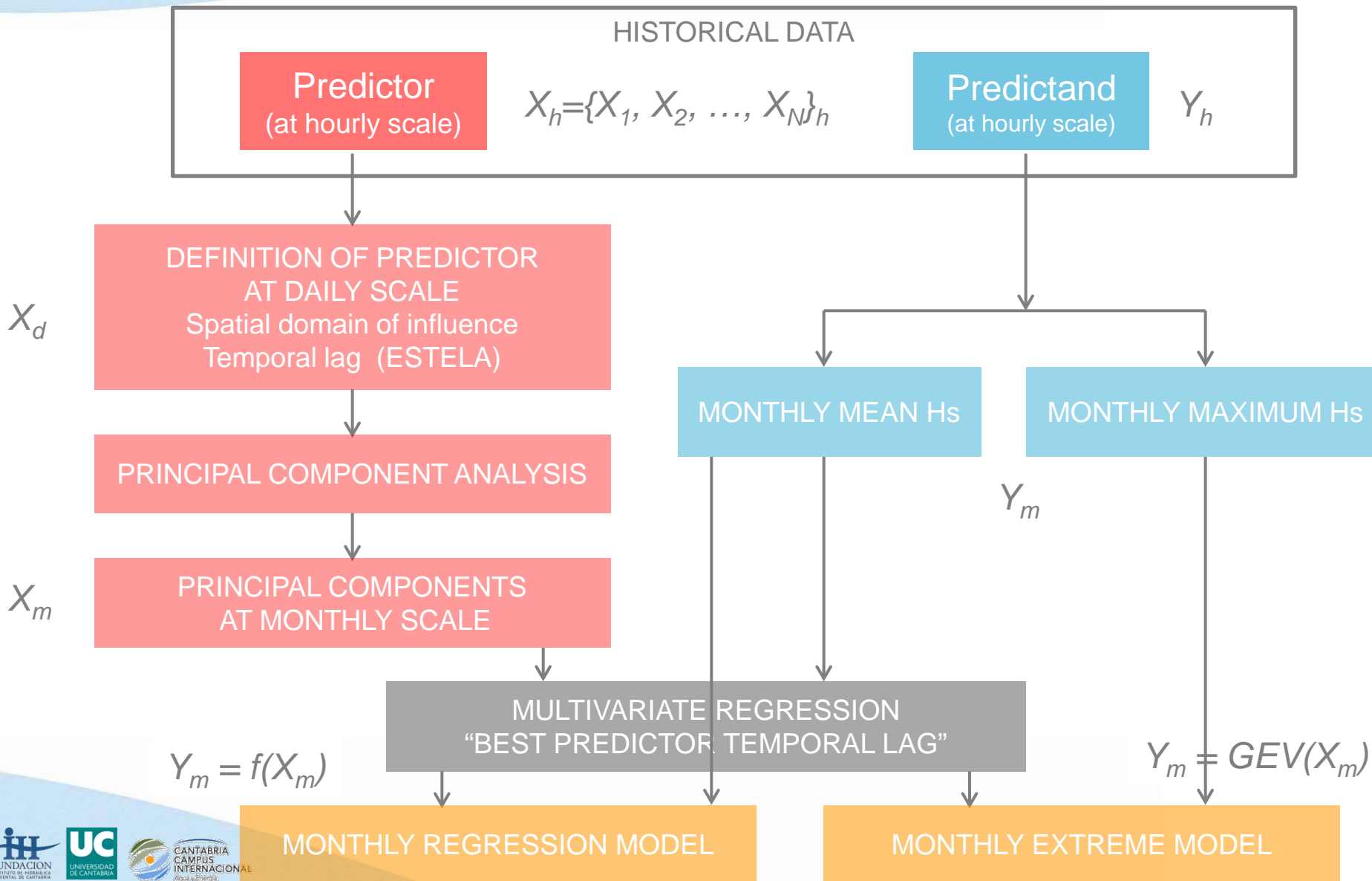
$$X_m = \{X_1, X_2, \dots, X_N\}_m$$



$$Y_m$$



$$Y_m = f(X_m)$$



HISTORICAL DATA

Predictor

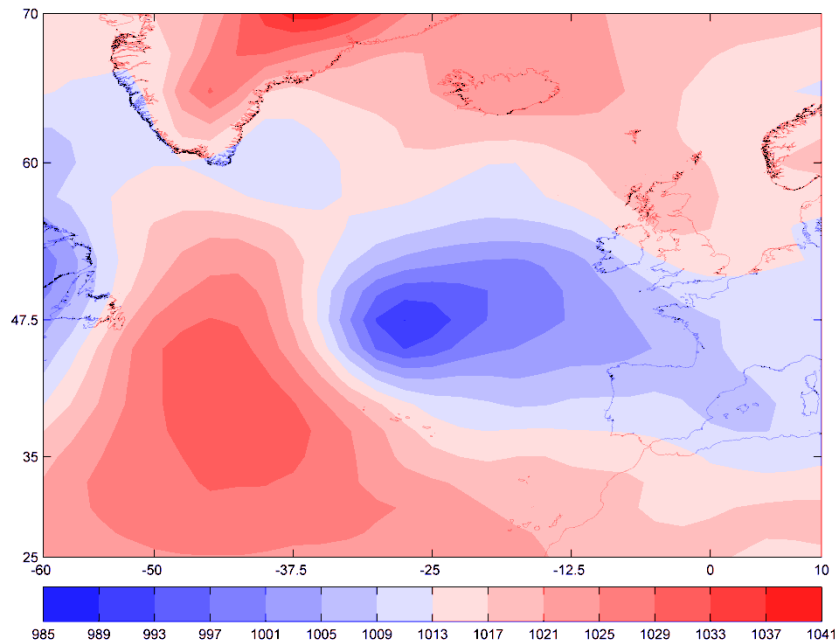
Predictand

NCEP reanalysis (SLP), 1948-2010

Time resolution: 6 h

Spatial resolution: 2.5°

Kalnay et al., 1996

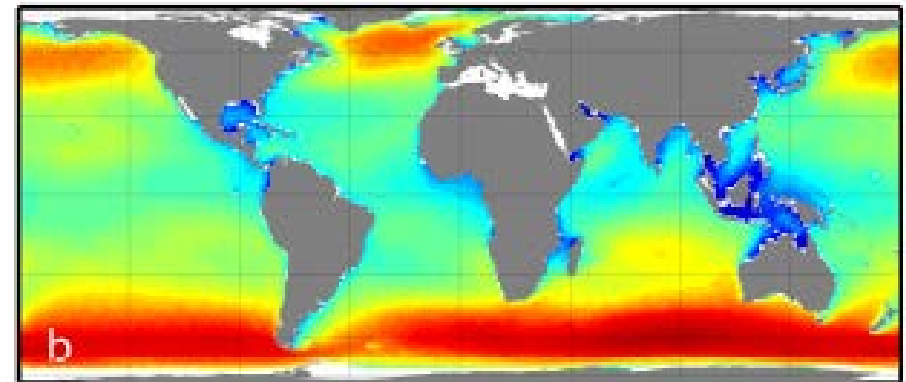


GOW reanalysis, 1948-2010

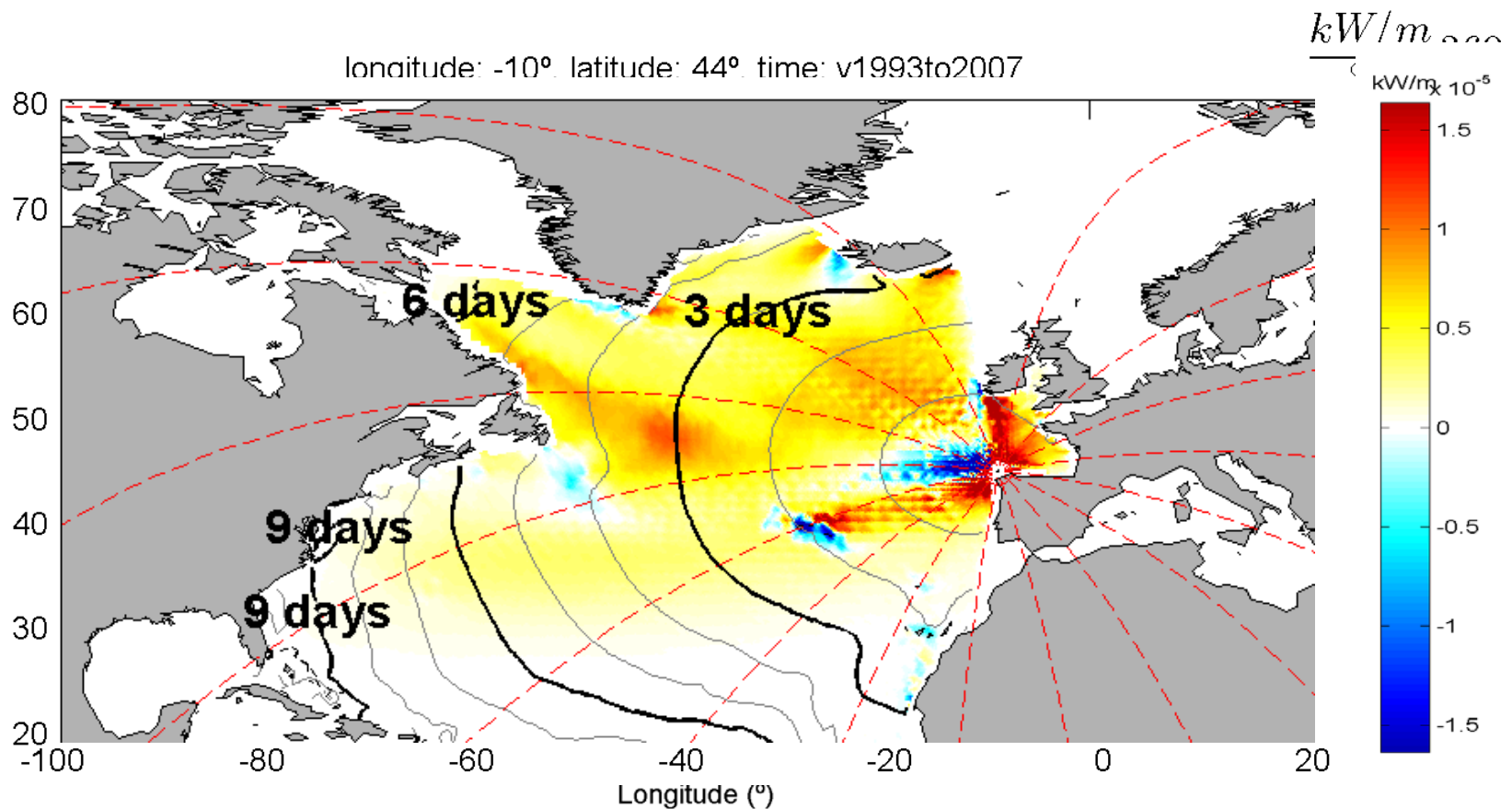
Time resolution: hourly

Spatial resolution: 1.5° in longitude and 1.0° in latitude

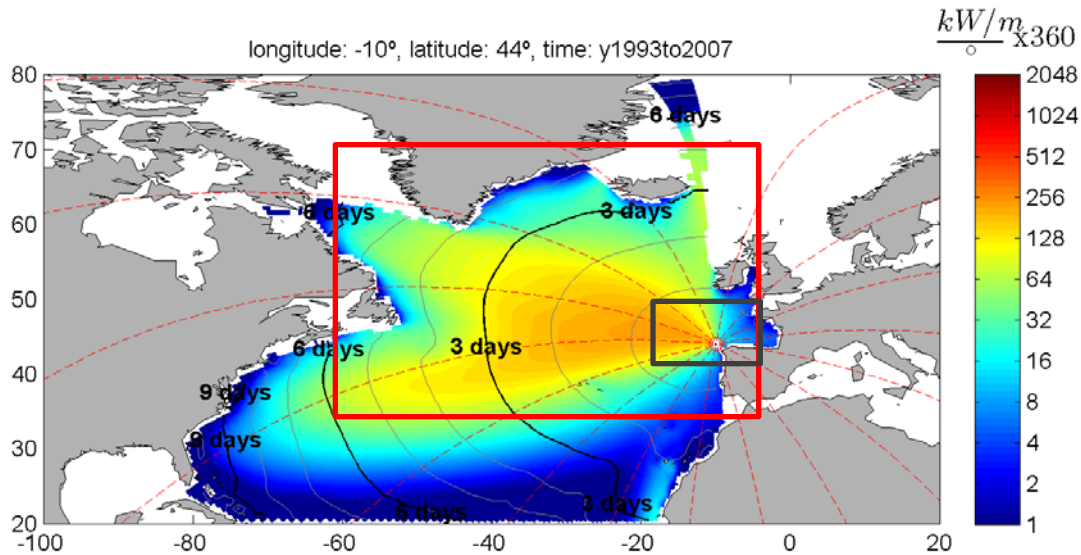
Reguero et al., 2012



DEFINITION OF PREDICTOR AT DAILY SCALE: Spatial domain and temporal lag ESTELA (Pérez et al., 2013)

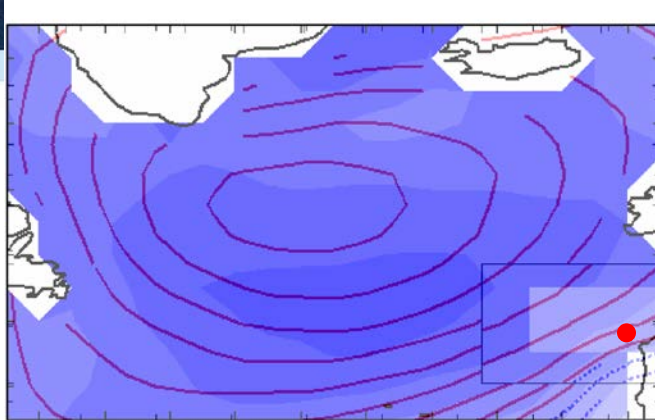


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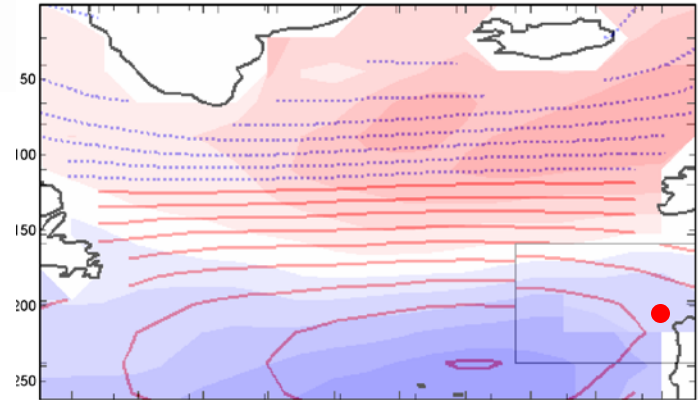
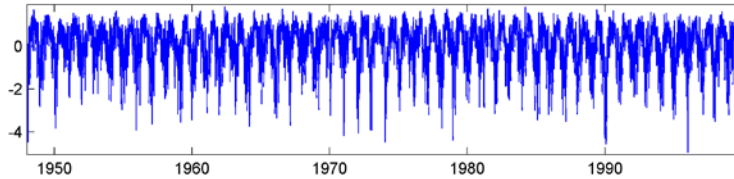


time scale

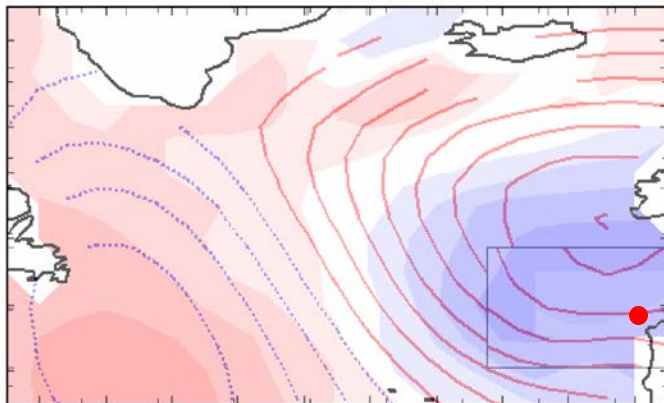
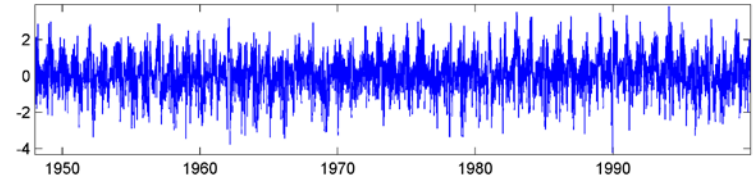
PRINCIPAL COMPONENT ANALYSIS



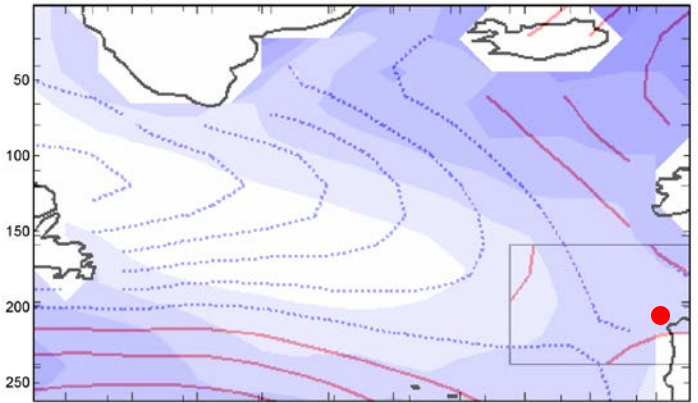
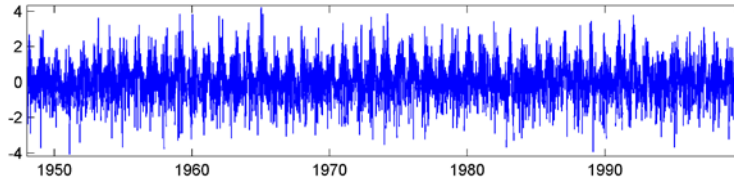
PC1 - var=26.7614 %



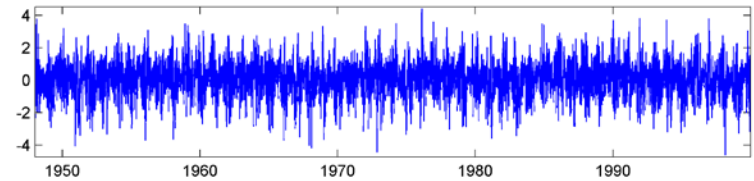
PC2 - var=17.7753 %



PC3 - var=10.5323 %

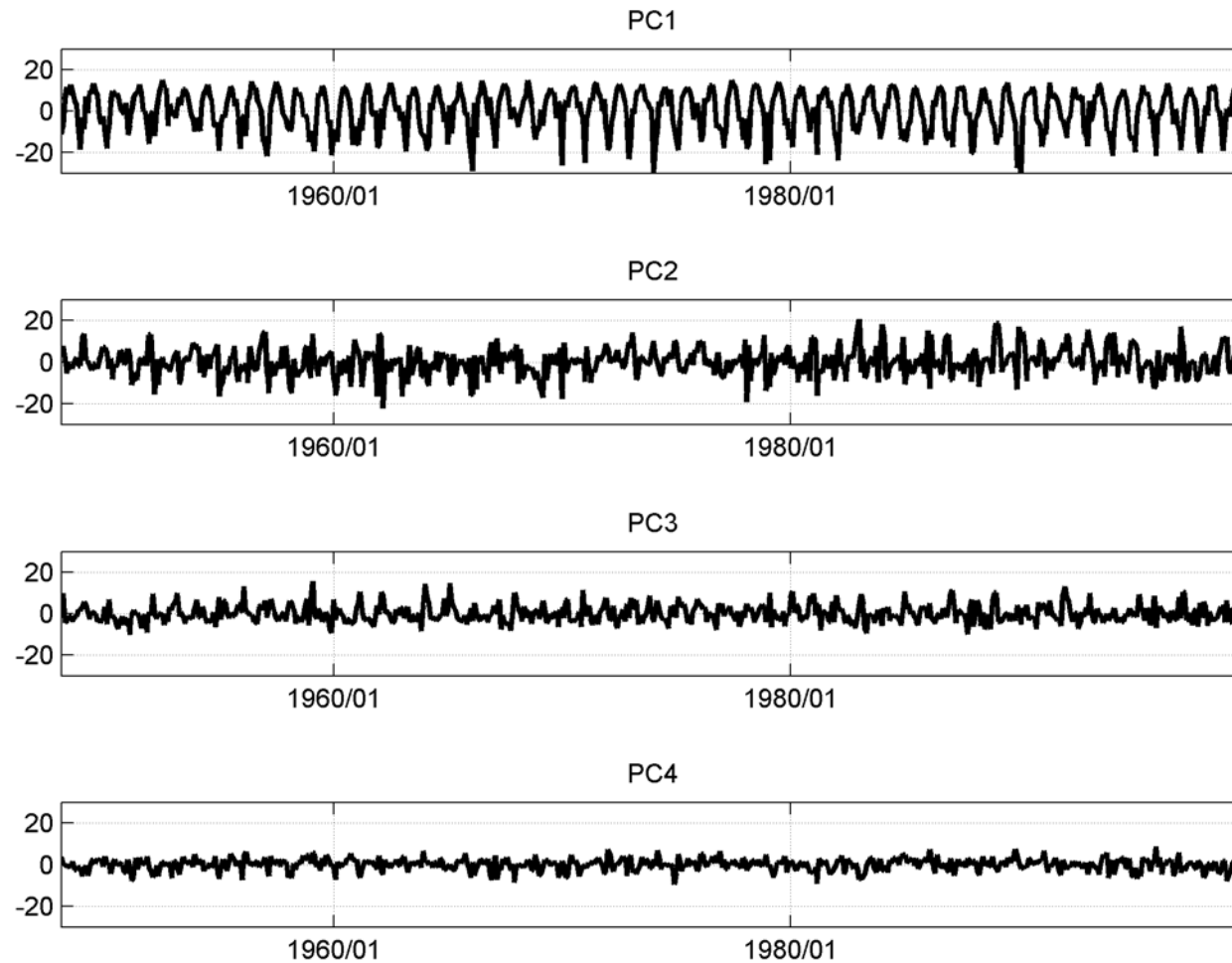


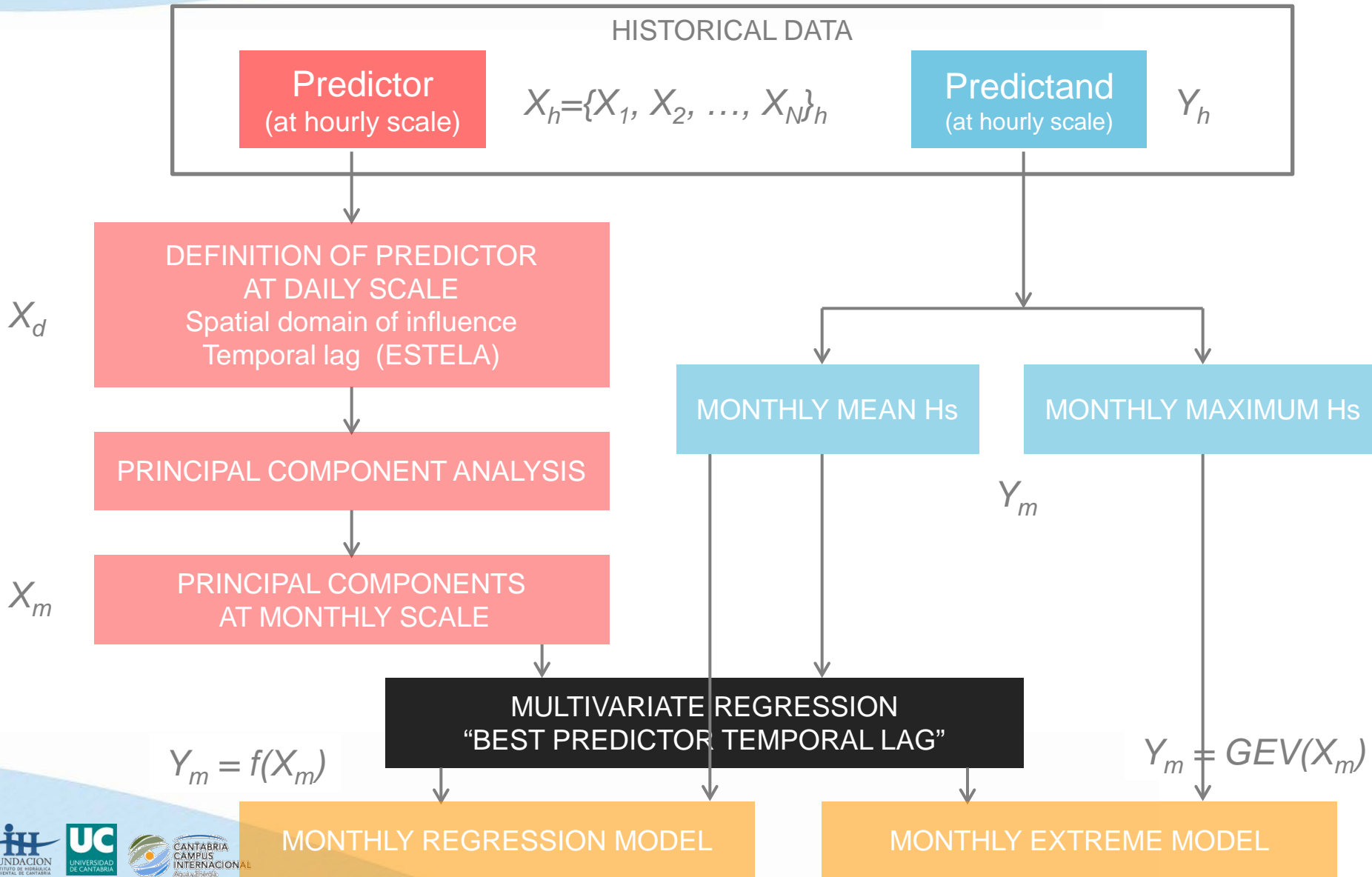
PC4 - var=5.7049 %



PRINCIPAL COMPONENTS AT MONTHLY SCALE

$$X_m = \{ PC1_m, PC2_m, \dots, PCN_m \}$$





MULTIVARIATE LINEAR REGRESSION

Predictors are selected in a forward procedure (Wang et al., 2010)

$$Y_m = f(X_m)$$

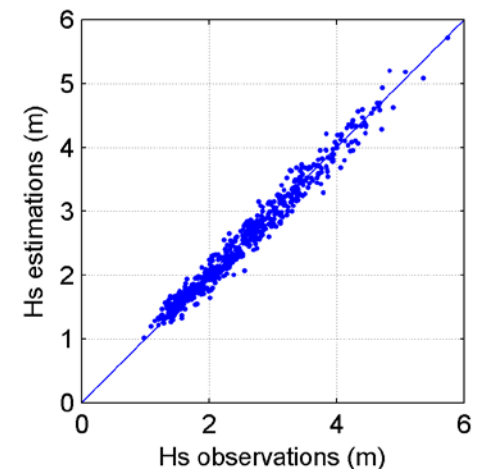
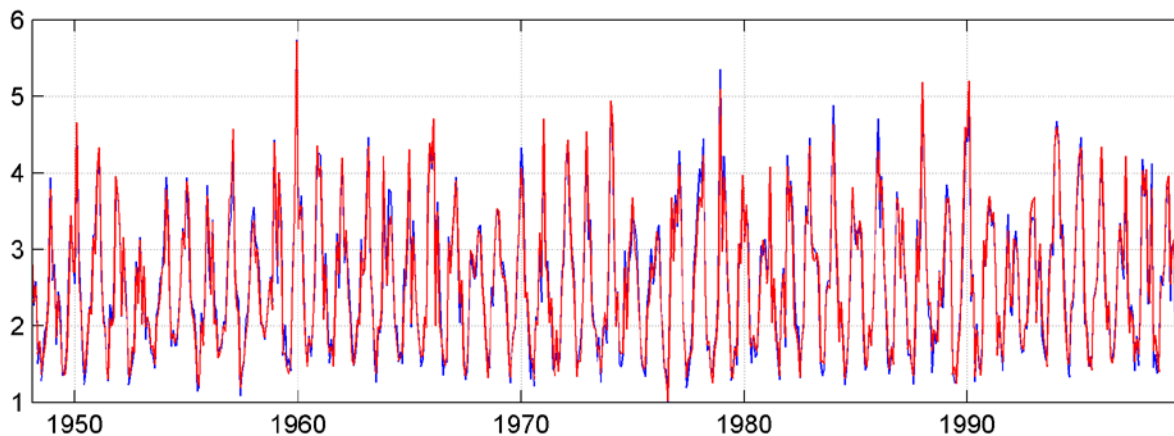
Calibration period: 1948 - 1999

$$M_1 : Y_m(t) = a + b_1 PC_m^1(t)$$

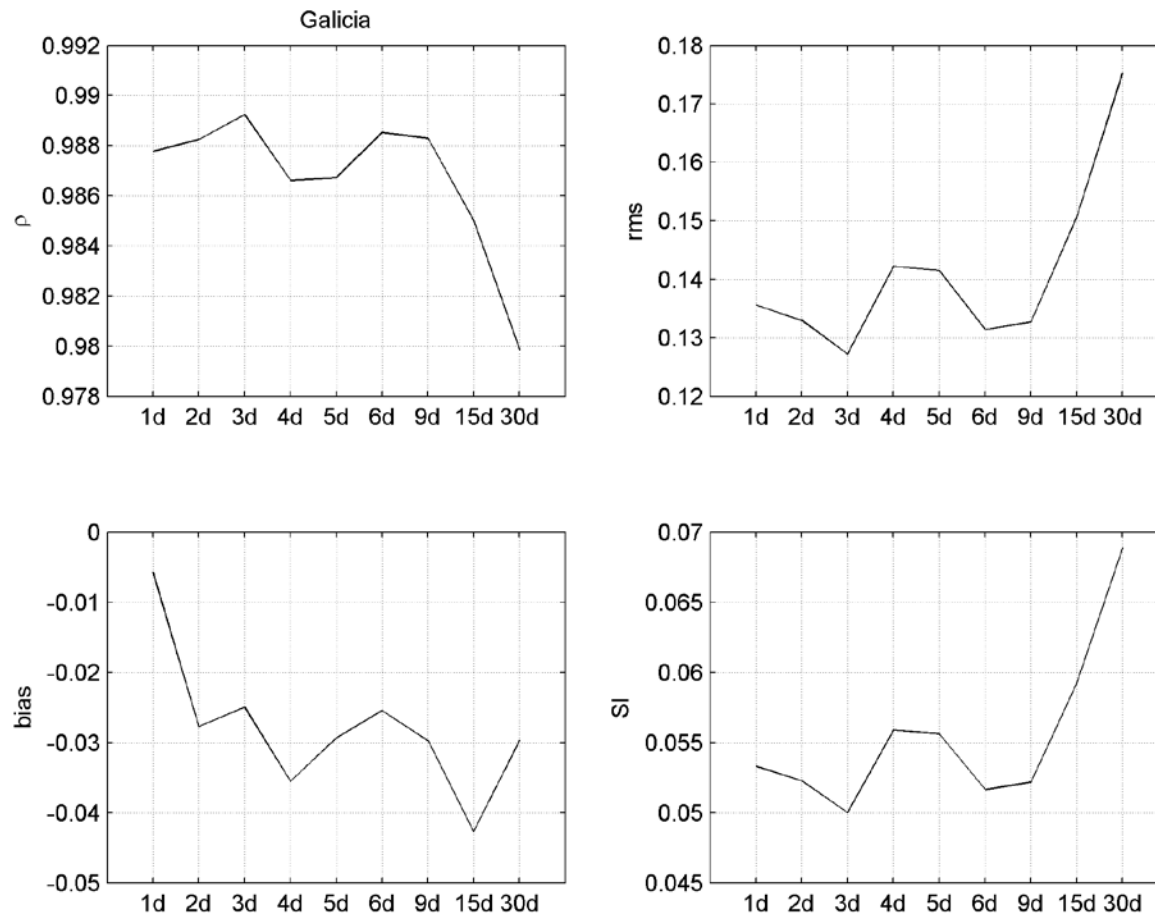
$$M_2 : Y_m(t) = a + b_1 PC_m^1(t) + b_2 PC_m^2(t)$$

...

$$M_i : Y_m(t) = a + \sum_{j=1}^i b_j PC_m^j(t)$$



Sensitivity analysis - BEST TEMPORAL LAG

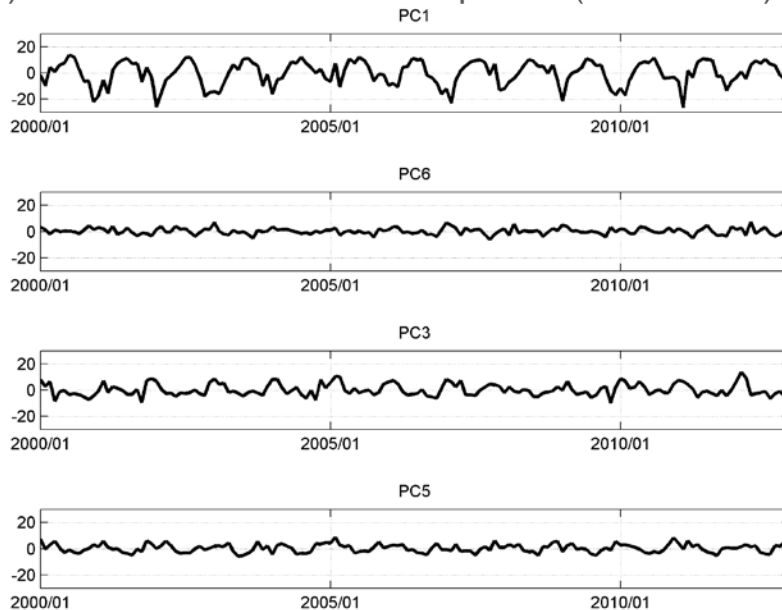


MONTHLY REGRESSION MODEL

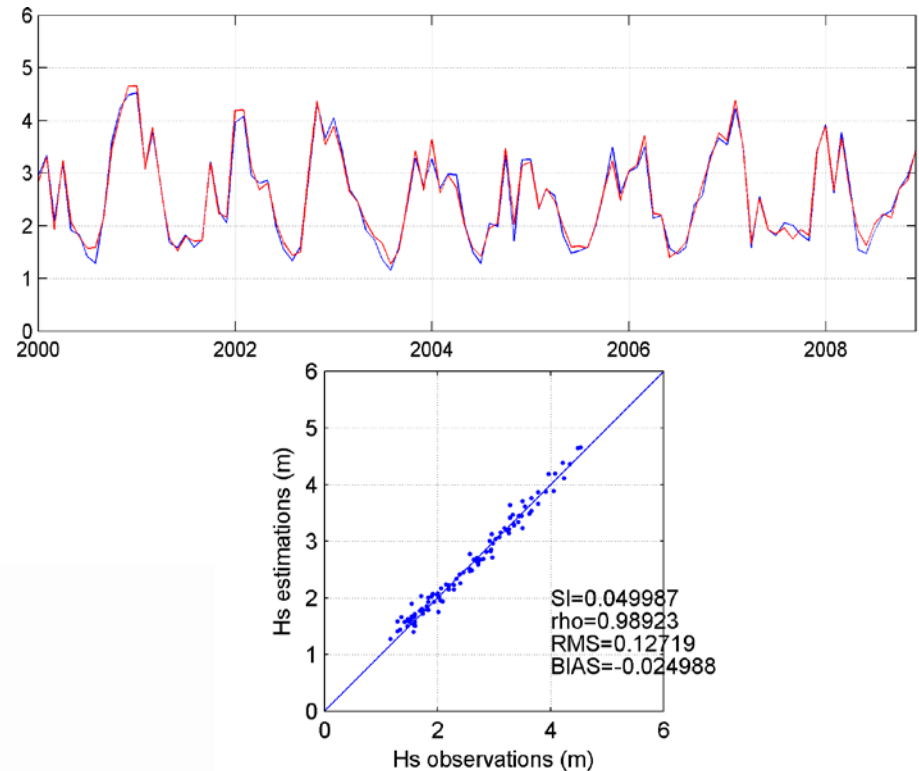
Validation of the regression model with the “best predictor”

$$Y_m(t) = 2.8424 - 0.8462 * PC1_m(t) + 0.2666 * PC6_m - 0.2343 * PC3_m + 0.1830 * PC5_m + \dots$$

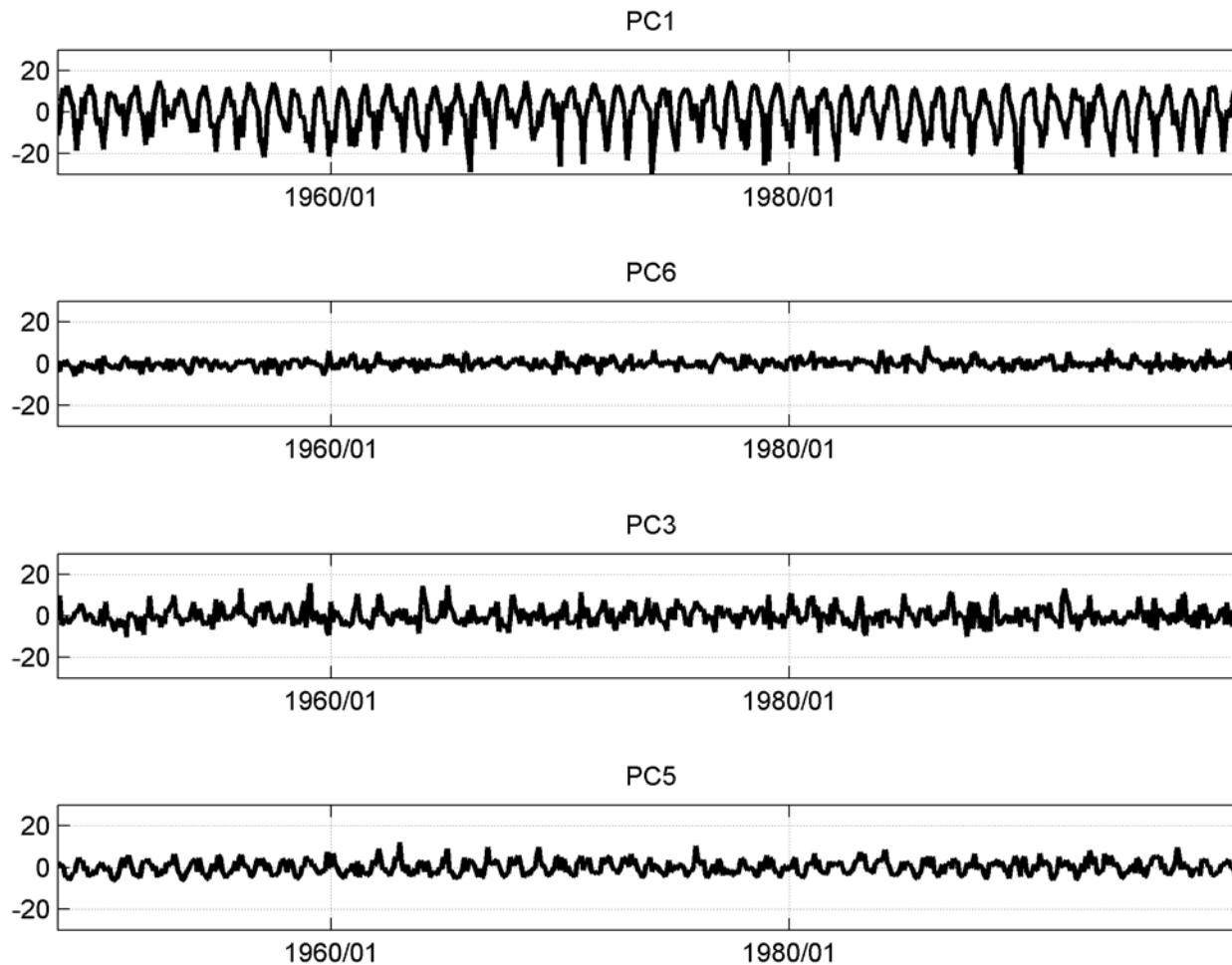
a) Predictor in the validation period (2000-2010)



b) Results in the validation period (2000-2010)



MONTHLY REGRESSION MODEL - LOCAL MONTHLY WAVE CLIMATE INDICES

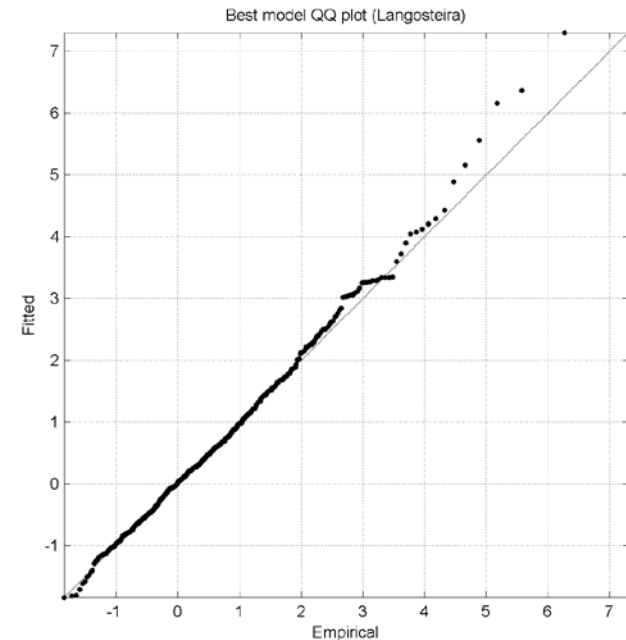
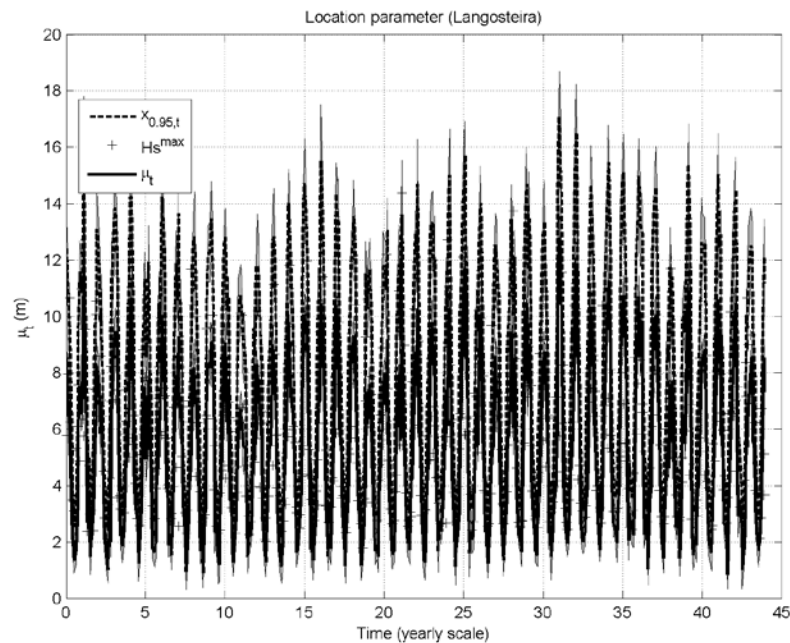
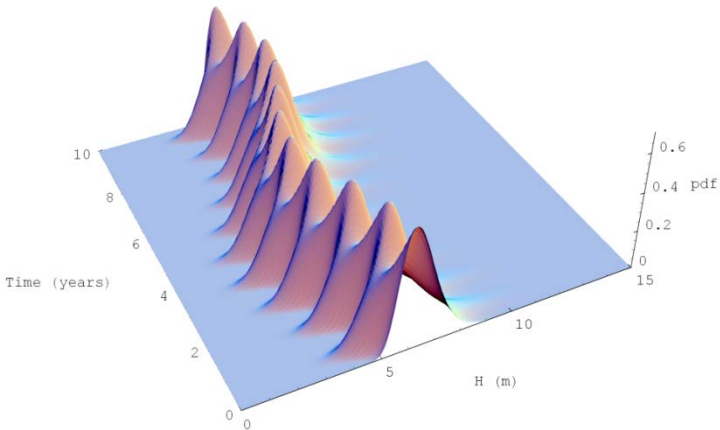


MONTHLY EXTREME MODEL

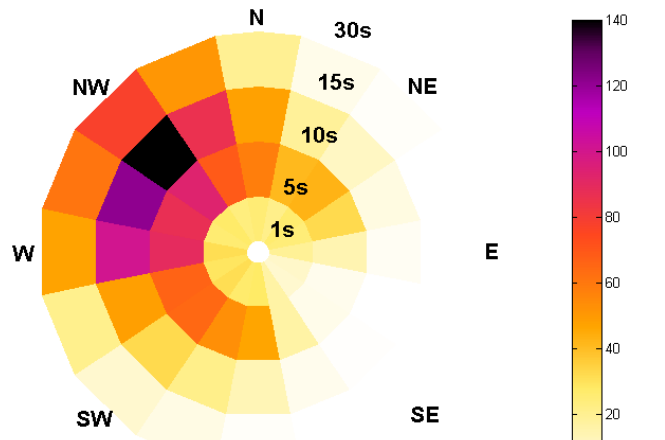
$$F(x; \theta) = \exp \left\{ - \left[1 + \xi \left(\frac{x - \mu}{\psi} \right) \right]^{-1/\xi} \right\}$$

Mínguez et al., 2010

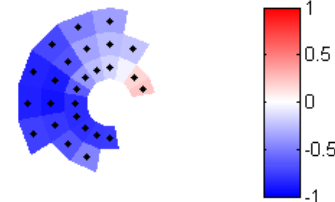
$$\begin{aligned} \mu(t) &= \beta_0 + \beta_1 * PC1_m + \beta_2 * PC2_m + \dots \\ \psi(t) &= \alpha_0 + \alpha_1 * PC1_m + \alpha_2 * PC2_m + \dots \\ \xi(t) &= \gamma_0 + \gamma_1 * PC1_m + \gamma_2 * PC2_m + \dots \end{aligned}$$



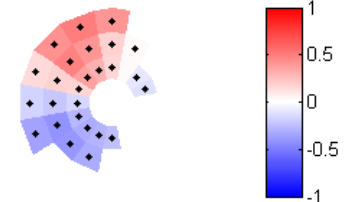
CORRELATION OF LOCAL MONTHLY INDICES WITH LOCAL WAVE CLIMATE



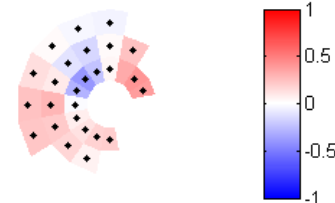
p PC1 - Galicia



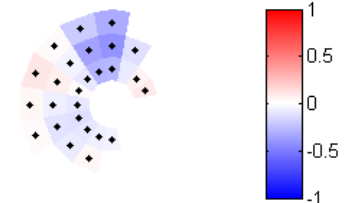
p PC2 - Galicia



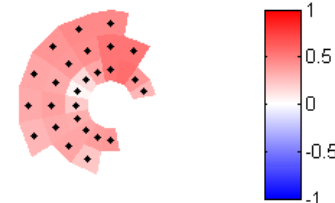
p PC3 - Galicia



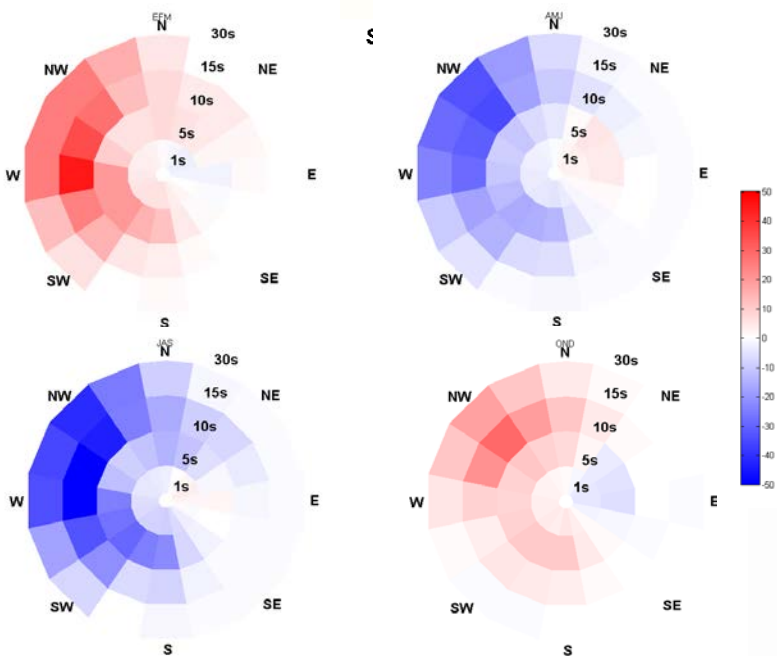
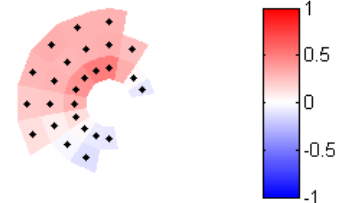
p PC4 - Galicia



p PC5 - Galicia



p PC6 - Galicia

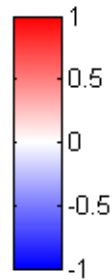


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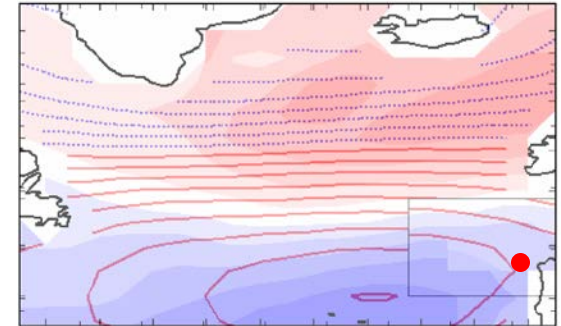
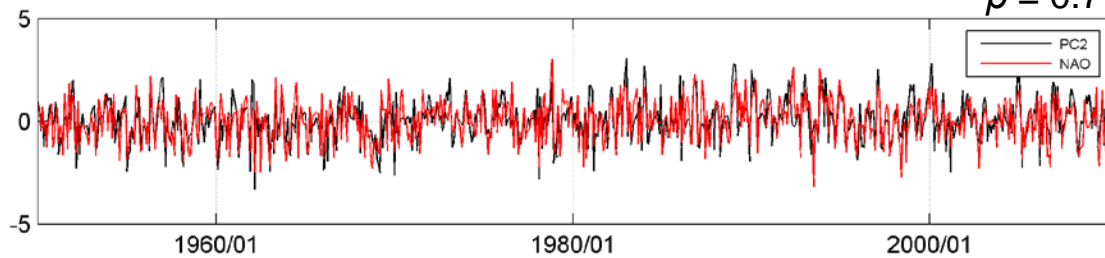
ρ PC2 - Galicia



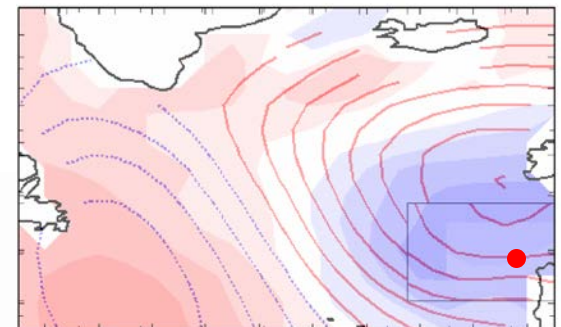
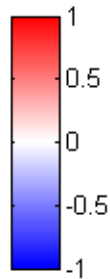
ρ NAO - Galicia



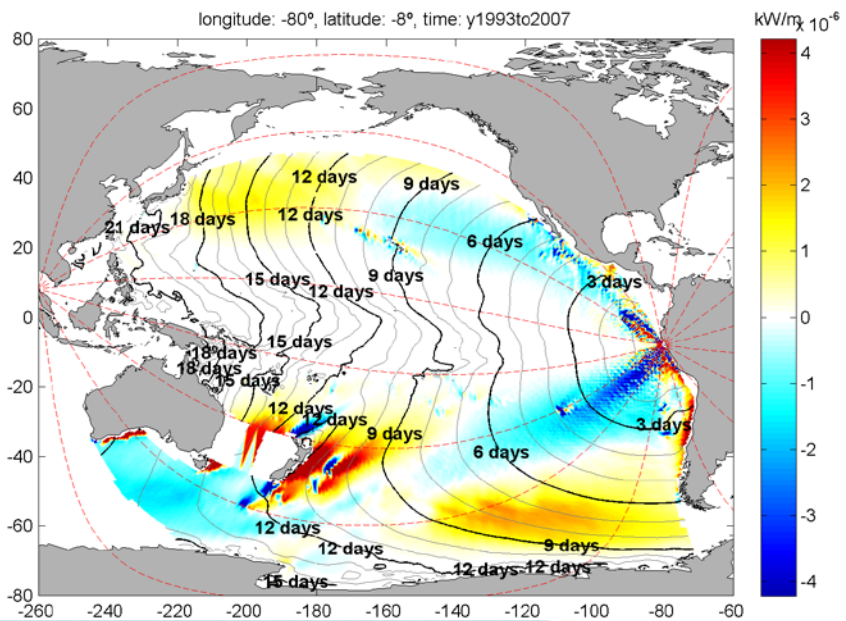
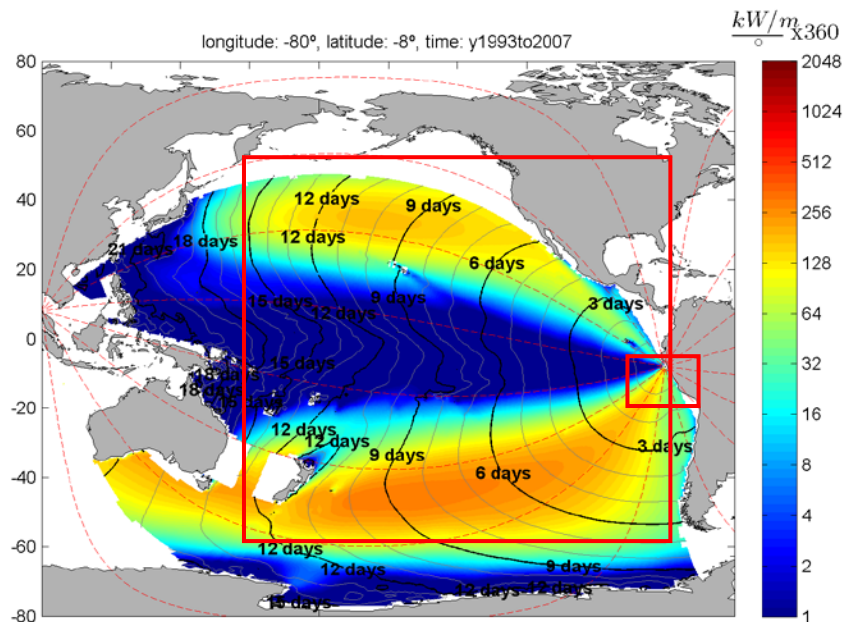
$\rho = 0.7$



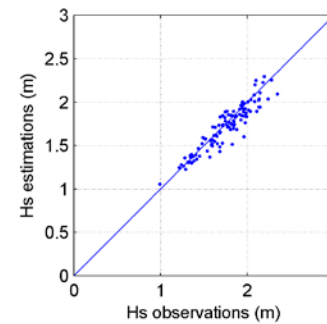
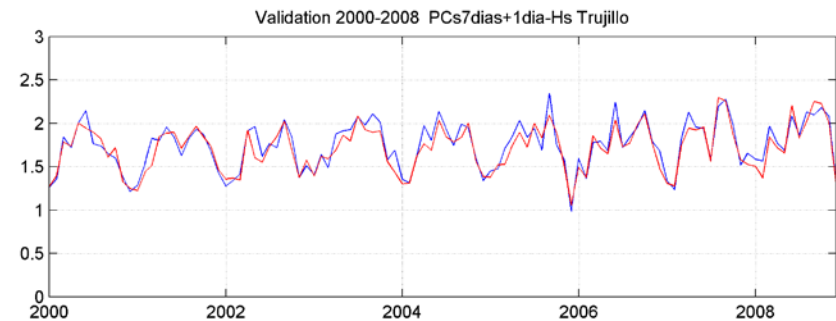
ρ PC3 - Galicia



TRUJILLO (PERÚ)

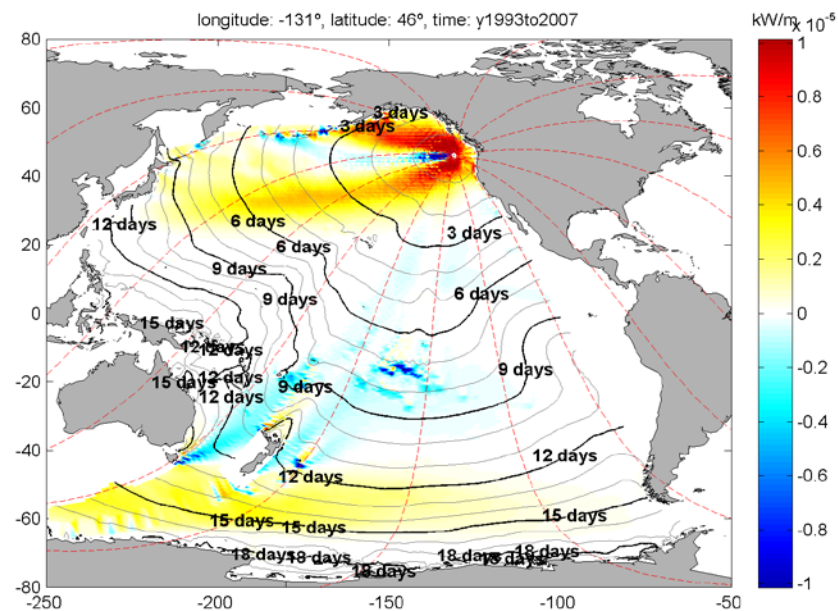
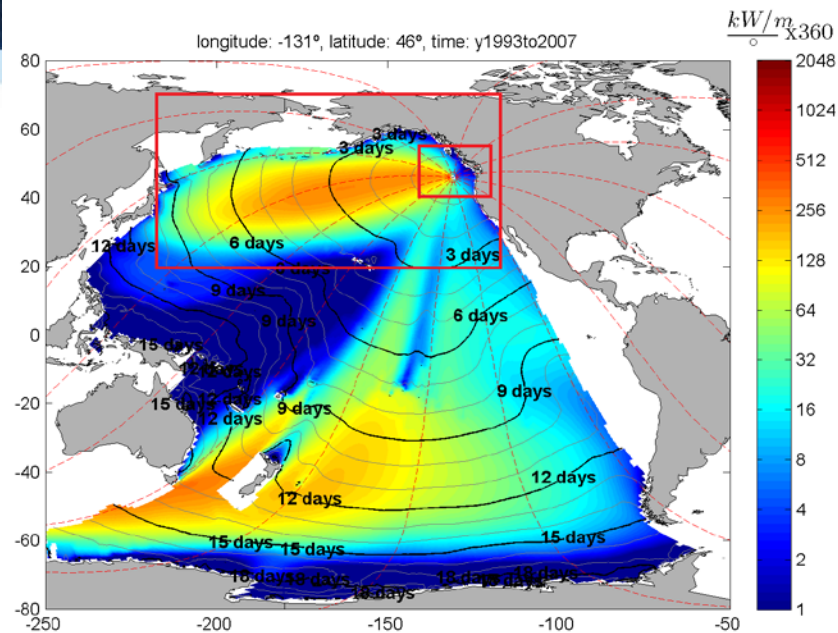


$$Y_m = f(X_m)$$

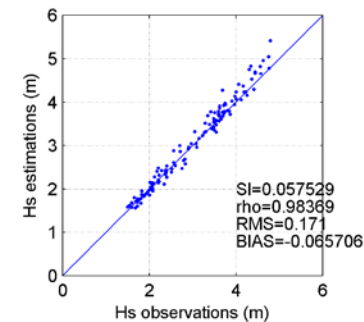
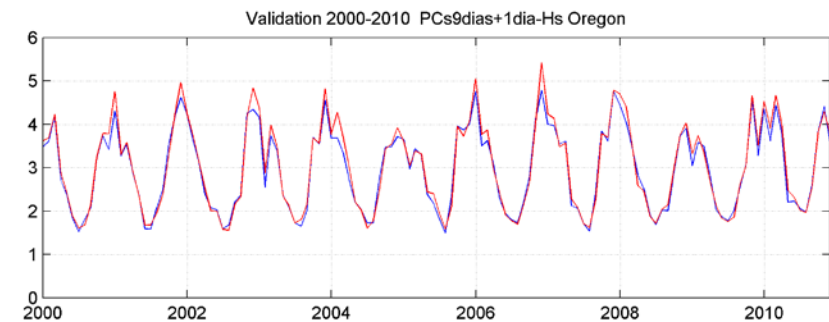


SI=0.06109
rho=0.93343
RMS=0.10673
BIAS=0.036409

OREGON (USA)



$$Y_m = f(X_m)$$



- A methodology to obtain local monthly wave climate indices has been proposed
- The methodology is based on the statistical relation between the atmospheric predictor (sea level pressure) and the local wave climate at a monthly scale
- The spatial domain and the temporal lag are the key properties of the wave predictor. The ESTELA method (Pérez et al., 2013) has been implemented to characterize the footprint of the wave climate at the particular location of interest and to automate the definition of the predictor spatial domain.
- The optimal temporal lag of the predictor in the area of influence of the local wave climate at particular location is obtained as the best solution of a multivariate regression model
- The monthly values of the principal components of predictor at a daily scale are the local indices.
- The local monthly indices obtained with this methodology improve the statistical model fitting, both the monthly mean wave height and the monthly maximum wave height.
- This statistical downscaling approach can be used to project, for different CMIP5 climate models and in extratropical areas, monthly wave climate (long-term distributions, long-term extreme value distributions)
- Further research is needed to extend this methodology to tropical regions (tropical cyclones...)

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