

1. Introduction

It is well recognized the strong influence of large-scale climatic patterns in the variability of many components in the hydrological cycle (Markovic, et al. 2009). Previous studies in the Iberian Peninsula identified the influence of atmospheric circulations in time series of precipitation (Muñoz, et al. 2004; Gallego, et al. 2005; Gonzalez-Hidalgo, et al. 2009; Rodriguez-Puebla et al. 2010), streamflow (Trigo, et al. 2004; Moreno, et al. 2006; Morán, et al. 2010) and piezometric heights (Luque-Espinar, et al. 2008). However, the influence in the maximum streamflow has not been addressed.

In the present work we addressed the question, How is the influence of large scale climate patterns in the maximum streamflow in the Peninsular Spain?, we evaluate this linkage between this variables by mean of Granger causality test, discrete spectral analysis and continuous spectral analysis.

2. Objectives

In general, the principal objectives of this work are:

- Evaluate changes in a climatic variables have an impact on changes in the maximum hydrological variables in the study area.
- Identified the reproduction of climatic cycles in the maximum monthly time series in a discrete approach.
- Explore changes in variance and linkage in phases between the variables in short and long-term.

3. Study Area and Information

Spain is situated in a complex meteorologically area, between mid latitudes and the north sub-tropics, between two important mass water (Atlantic Ocean and Mediterranean Sea), as well as very rugged orography, represented one of the most extreme cases of variability natural indicator within Europe.



Figure 1. Study area with main hydrological divisions and principal mountain system. Figure 2. Patterns of wet days in some regions within the zone of study (base period 1960-1990).

The analysis is based on two types of information : Maximum streamflow time series and macroclimatic indices. We use 80 maximum monthly streamflow time series, distributed along the study area with at least 40 years (Fig.3).

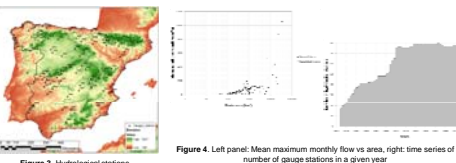


Figure 3. Hydrological stations. Figure 4. Left panel: Mean maximum monthly flow vs area, right: time series of number of gauge stations in a given year.

We selected five teleconnection patterns NAO (North Atlantic Oscillation), AO (Arctic Oscillation), MO (Mediterranean Oscillation), WeMO (Western Mediterranean Oscillation) and Multivariate ENSO Index.

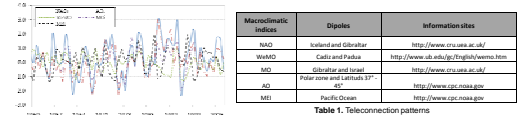


Figure 5. Evolution of the winter climate indices in the last years

4. Methods

Granger Causality Test

This test is useful for evaluate when changes in a variable X will have an impact on changes other variable Y and provide valuable information about the power of the variable X in the future values Y (Granger, 1969). Let X and Y time series. We assume the null hypothesis H0 that X does not cause Y (H0: β1 = β2 =...=βp). To evaluate the null hypothesis, one first finds the proper lagged values of Y to include in a univariate autoregression of Y:

Yt = α0 + α1Yt-1 + α2Yt-2 + + αpYt-p + εt (1)

Here Yti is retained in the regression if and only if it has a significant t-statistic; m is the greatest lag length for which the lagged dependent variable is significant. Next, the autoregression is augmented by including lagged values of X:

Yt = α0 + α1Yt-1 + α2Yt-2 + + αpYt-p + β0 + β1Xt-1 + β2Xt-2 + + βpXt-p + εt (2)

We obtain RSS (full model) from Ec.2 and RSS(restricted model) from Ec.1. According to F-test evaluate the explanatory power.

F[(N - k)/q] = [(RSSrestricted - RSSfull)/RSSfull] (3)

N: number of observations; k: number of parameters from full model; q: number of parameters from restricted model The null hypothesis is rejected if F-value > critical value from the F-distribution, then we can say X causes Y.

Spectral Analysis

The first approach in the analysis of the linkage between climatic oscillation and hydrological cycles we used classic spectral analysis. There are many methods for estimate the spectral density, we decided to use the Blackman-Tukey method, where this is calculate from the covariance function (Chatfield, 1991):

S(ω) = 1/π * {w(0)C(0) + Σ k=1 to m {w(k)C(k)cos(ek)} (4)

Where s(ω) is the estimated spectral density for frequency ω, C(k) in the function of covariance for k-ésimo value and w(k) weighting function, know a "lag-window", which is used to give less weight to the covariance estimates as the lag increases. The lag window used was the Tukey window:

w(k) = 1/2 * {1 + cos(πk/m)} 0 ≤ k ≤ m (5)

m is the maximum number of lags for the covariance function used in the spectral estimation, the maximum number of lags are n-1, where n is the number of data. We identified the principal cycles for different confidence levels 99%, 95%, >95% and no detected by mean of the method propose by Luque, et al (2008).

Continuous Wavelet

Continuous approach in the linkage of climatic and hydrology variables we used Wavelet analysis and two useful tool for this analysis cross wavelet transform and wavelet coherence transform. The continuous wavelet transform (Torrence and Compo, 1998) allow decompose a time series in the time-frequency domain, y we will identified the principal oscillations and how this change in time. We selected for this work the Morlet wavelet:

ψ0(η) = π^-1/2 * e^iω0η * e^-η^2/2 (6)

Where ω0 is the dimensionless frequency and η dimensionless time. The cross wavelet transform (CWT) defined :

W^xy(s,t) = W^x(s,t) * W^y(s,t) (7)

this allow identified when the time series oscillated in a common frequency in time, besides detect the intermittent coupling. Another tool is how coherent is the cross wavelet transform, who is the wavelet coherence transform (WCT). Which is a measure of the intensity covariance between the timer series. According with Torrence and Webster (1999) is defined:

Rc^2(s) = [S(w^xw^y(s))] / [S(w^xw^x(s)) * S(w^yw^y(s))] (8)

5. Results

The results obtained with Granger causality test, allow to identified the important impact of the macroclimatic indices in the behavior of the maximum streamflow. It is clear the influence of the NAO and AO indices in the time series of the Atlantic basins and principally for 90 and 95%, the MO index show important impact in time series of the Atlantic and the Ebro confederation for 90 and 95%. The MO index show significance result in timer series from the façade Mediterranean principally in basins from Jucar and in the lower of Ebro.



Figure 6. Maps of statistically significance for Granger causality test between monthly maximum streamflow and monthly circulation patterns

Discrete Spectral Approach

After that, we used the classical spectral analysis for explored the connection between climatic cycles and oscillations in the maximum hydrologics. We identified the principal oscillation modes in the macroclimatic series for confidences levels of 90 and 95% (Table 2). Figure 7 show power spectrum of monthly time series with bands of confidence levels.

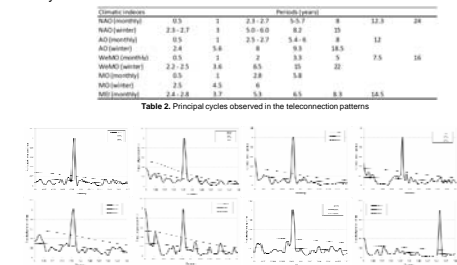


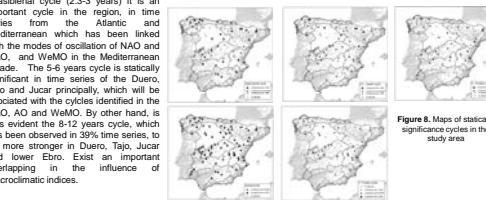
Figure 7. Maximum monthly streamflow power spectrum

In order to observe the spatial distribution of principal cycles in the study area, we elaborated maps according different confidences levels, we consider semi-annual, annual e interannual cycles (2.3-2.8 years, 5-6 years y 8-12 years).Table 3 show the results of significant time series in percentage for different confidence levels (90%, 95%, <95% and no detectable). Figure 8 show maps of statically significance cycles.

Table 2. Principal cycles observed in the teleconnection patterns

| Teleconnection | Period (years) | Confidence level (%) |
|----------------|----------------|----------------------|
| NAO | 2.3-2.7 | 90 |
| AO | 2.3-2.7 | 90 |
| MO | 2.3-2.7 | 90 |
| WeMO | 2.3-2.7 | 90 |
| MEI | 2.3-2.7 | 90 |

Table 2. Principal cycles observed in the teleconnection patterns



Continuous Spectral Approach

□ Short - term

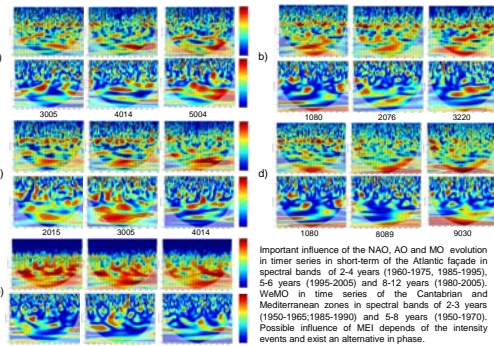


Figure 9. Cross-wavelet transforms and squared coherence, between monthly maximum streamflow and macroclimatic indices (a)NAO; b)AO; c)MO; d)WeMO and e)MEI. The thick black contour designates the 95% confidence level against noise and the cone of influence, where the effects become important. The relative phase relationship is shown as arrows (where in phase pointing right and anti-phase left)

□ Long - term

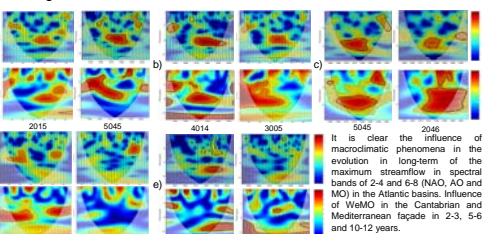


Figure 10. Cross-wavelet transforms and squared coherence, between annual maximum streamflow and winter macroclimatic indices (a)NAO; b)AO; c)MO; d)WeMO and e)MEI. The thick black contour designates the 95% confidence level against noise and the cone of influence, where the effects become important. The relative phase relationship is shown as arrows (where in phase pointing right and anti-phase left)

6. Conclusions

The results allow to evidence the important influence of the climatic patterns in the evolution of the maximum streamflow, as well as the spatial variation of this influence in the study area. Its clear that influence is limited by the principal mountain system. The granger test show the significant impact of the climatic variables in the hydrological variables. The spectral analysis show the linkage between the variables in different cycles and the continuous analysis evidence intermittent association along the time. Exist overlapping areas in the influence of this phenomena and it is important to considerate the connection between the atmospheric and ocean configuration. In general, the results provide valuable information about the climatic indices for improve forecasting and use in non-stationary framework.

Acknowledgments

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