

Referee #1

Q1. The manuscript "Extreme waves and climatic patterns of variability in the Eastern North Atlantic and Mediterranean basins" by Morales-Márquez et al. describes an analysis of the 99percentile of significant wave height focused during winter months over NE Atlantic and Mediterranean Sea. The study is interesting and some outcomes show novel information and wave climate variability patterns (eg. the contribution of seasonality into variability, historical trends over the 30-year period used, correlation with climate indices).

Reply #1. We deeply thank Referee's comments and the effort that she/he made in reviewing carefully our work. We have introduced in the new version of the Manuscript (hereinafter Ms) all points raised in the review. We sincerely think that the new version of the Ms has been improved, thanks to this discussion.

Q2. From my point of view, however, the manuscript requires an improvement in the wording to better describe some technical steps. One of the main weaknesses of the work is that it seems that the SWH has not been validated against observations, neither the SHW99 magnitude nor its variations.

Reply #2. As suggested, all technical aspect raised by the Reviewer have been modified following her/his suggestions. In particular we have included a comparison between the wave reanalysis and available in situ buoys. All Referee's concerns are explained in detail in the following replies.

SPECIFIC COMMENTS

Q3. English grammar needs to be checked.

Reply #3. The Ms. has been revised in detail and typos corrected.

Q4. Lines 11-12 "Besides, extreme waves influence the upper ocean by enhancing vertical mixing through the Stokes layer". A reference of the statement would be desirable.

Reply #4. As suggested by the Reviewer, a new reference has been included in this sentence of the manuscript. Lines #14-15 read,

"Besides, extreme waves influence the upper ocean by enhancing vertical mixing through the Stokes layer (Polton et al., 2005).

Q5. Line 13. I do not completely understand the message about the role of the extreme waves in coastal flooding at 'intra-annual' scale. Please, consider rewriting.

Reply #5. Coastal flooding depends on the combined effect of Sea Level, Storm Surges and Wave Setup. While the first impact is a large-scale effect, surges and wave setup have a clear seasonality (see Figure 2 panels c and d). Thus, larger impacts in flooding are related with maximum waves and large surges that are intra-annual processes.

Q6. References of the journal articles are odd worded at the end of the manuscript, making difficult find the referenced information sources.

Reply #6. We followed the Ocean Science template for the References.

Q7. I suggest removing 'methods' of the title of section 2 since EOFs, correlation significance estimation, composites, etc. methods are not described in this section.

Reply #7. We modified the Section title as "Data and extreme wave values".

Q8. Section 2.1 "Waves and Atmospheric Data" requires organisation and adding relevant details for the analysis of extreme wave climate. I am confused by some aspects that are ambiguously mentioned. Some

examples: a) What is the time resolution of the used winds to generate the waves by forcing WaveWatch model? Are winds fields at 0.5deg are used over the Mediterranean Sea? Is this spatial resolution enough to simulate wave extremes? b) What is the time resolution of SWH from the database? Averaged 3hourly values? One hourly value each 3 hours? c) Has the used wave data source been validated in the study area? A comparison against buoy records in the analyzed domains shown in figure 1 is crucial to validate the further analysis of extreme waves.

Reply #8.

a) The temporal resolution of the winds forcing the wave model is 1 hour. It is defined at lines #57-60 of the Ms.:

“This dataset (i.e. WAVEWATCH III 30-year Hindcast Phase 2, (Chawla et al., 2012)) has been generated by forcing the “state-of-the-art” wave model WAVEWATCH III (Tolman, 2009) with 10-m height high-resolution wind fields from the NCEP Climate Forecast System Reanalysis and Reforecast (CFSRR) a 30-year homogeneous data set of hourly 1/2° spatial resolution winds.”

Regarding the wind resolution at the Mediterranean Sea the hindcast fits better in this basin than in the North Atlantic Ocean according to the buoys data from CMEMS (see Table 1 of the Ms), having largely been used in the North Atlantic to analyze the mean and the extreme waves (see for instance Forte, M.F. et al., 2012; and Gonçalves, M. et al., 2018).

b) The wave model provides outputs every 3 hours, which is largely accepted to be a sea state.

c) Thanks to referee’s suggestions we compare the hindcast at the Mediterranean Sea and NE Atlantic Ocean with buoys available at Copernicus (<https://marine.copernicus.eu/>). In addition, we write the following sentences in order to include this comparison in the Ms. (Lines #74-82):

“The hindcast is validated at the two basins with available buoys following the methodology described in Morales Márquez et al. (2018). For 13 buoys in the Atlantic and 11 in the Mediterranean Sea, we compute the correlation coefficient (R^2), the Scatter Index (SCI) and the Relative Bias (RB) defined as:

$$R^2 = \frac{\text{Cov}(b, m)}{\sigma_b \sigma_m},$$

$$SCI = \frac{\text{rms}_{m-b}}{\max(\text{rms}_b, |\langle b \rangle|)},$$

$$RB = \frac{\langle m - b \rangle}{\max(\text{rms}_b, |\langle b \rangle|)},$$

where m is the hindcast at buoy location and b the buoy dataset.

Table 1 shows the comparison between the WAVEWATCH III 30 hindcast and the buoys from Copernicus Marine environment monitoring service (CMEMS, <https://marine.copernicus.eu/>). See Fig. 1 for buoy locations.

Table 1. Statistical comparison between the WAVEWATCH III 30 hindcast and CMEMS buoys.

Basin	R^2 (%)	SCI	Relative bias
Mediterranean Sea	73.09 ± 0.17	0.39 ± 0.00	-0.13 ± 0.00
North Atlantic	72.67 ± 0.96	0.33 ± 0.02	0.16 ± 0.02

Table 1 shows the comparison between the WAVEWATCH III 30-year Hindcast Phase 2 and the buoys from Copernicus Marine environment monitoring service (CMEMS, <https://marine.copernicus.eu/>). See Fig. 1 buoys locations.”

Additionally, Figure 1 has been modified in order to show location of buoys.

Q9. I wonder how robust figure 2c and 2d are, as they are calculated from the maximum 99percentile value of a month and year. Are the monthly spatial patterns preserved for the averaged month of highest 99percentile value?

Reply #9. Figure 2 shows the maximum value of all monthly 99 percentiles. Panels a and b correspond to the value and panels c and d to the month when it occurs. The monthly patterns are not preserved (each grid point has its maximum at a specific month). The interest of this Figure is thus the homogeneity in the obtained months since the most of extreme waves are concentrated in few months (Dec-Mar).

Q10. Is semi-annual cycle statistically significant in the regression model? (eq.1). Panels of figure 3 do not show clear semiannual cycles for points 1 & 2. Maybe the variance reduction is only due to annual cycle

Reply #10. We compute the level of significance for the annual and semiannual cycle being both significant in the whole basins. As Referee points out, most of the variance reduction is due to the annual cycle but the semiannual variation is also significant both in the North Atlantic Ocean and in the Mediterranean Sea. In addition, we show that the intra-annual frequencies of the SWH₉₉ signal have to be removed if we want to analyze the long-term variability of the extreme waves.

Q11. Lines 90-93. Please, review the grammar. I do not completely understand the insights in the sentences. How 'generation wave areas' have been estimated/detected? What does development wave area mean?

Reply #11. Thank you for your advice. We rewrite the lines #102-105 of the Ms. as,

"In the Mediterranean Sea, there are two different areas in terms of seasonality. One is located in the central basin where seasonality explains up to a 70% of extreme waves and the other located in the Gulf of Genova and the Alboran Sea, where seasonality explains less than 10% of the signal (Fig. 3b). These areas are very active in terms of cyclogenetic activity (Trigo et al., 2002) and thus the seasonal signal is relatively less important here.

Regarding the estimation of the generation areas, we distinguish between Sea and Swell waves the first generated by local winds and the second including intensity, duration and distance (waves travelling across large areas). Swell waves are more ordered, presenting a more regular appearance with larger periods and wavelengths in a narrower frequency range than Sea type waves. The wave generation areas can be estimated analyzing wave features (SWH and wavelength, or period since they are related through the dispersion relationship) and local winds.

In these lines of the Ms., we explain how the local wind dominates the areas where the seasonality describes a smaller value of the extreme wave; corresponding with the generated waves areas. On the other hand, the higher values of the seasonality take place in regions where the waves can be developed better because they travel large distances. In these areas the dominant sea state corresponds to Swell.

Q12. I do not see the relationship (alignment) of the historical estimated SWH₉₉ trends (1979-2009) with future projections of surface wind (referenced from Gallagher et al. 2016). Please, clarify this statement.

Reply #12. Figure 3a and b of Gallagher et al, 2016., show the projected changes for the period of 2070-2099 relative to 1980-2009. There, they obtain a similar wind variation pattern than the trend of SWH₉₉ map that we present. We want to remark that the tendency of the SWH₉₉ most likely will follow the same pattern that we have calculated.

We have rewritten these lines in order to clarify it, in #111-113 of the Ms. as,

"This aligns with the obtained results in Gallagher et al. 2016., where the future projections of mean surface wind show an average decrease over the North Atlantic Ocean for winter

season, so likely the extreme waves continue with the same pattern in terms of long-term variability.”.

Q13. Line 116. Please, provide details about how the 5yr-periodicity is estimated. I cannot see clearly in Fig.5-1.

Reply #13. We modified the Ms and we now explain that we computed it through a FFT analysis of the PC. For that, we add this information in #129-130 lines of Ms. as,

“The first EOF, which explains a 28.5% of the winter SWH₉₉, presents a periodicity in its PC around 5 years (calculated through FFT analysis of Fig. 5-1).”

Q14. White color of colorbar in figures 7 & 9 must be centered on zero.

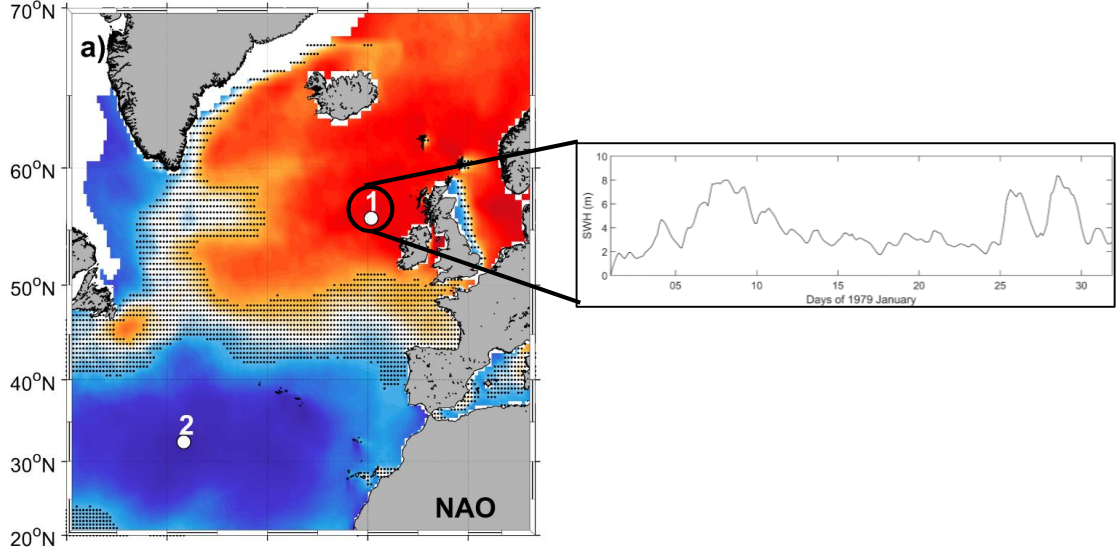
Reply #14. This has been fixed. In addition, we change the colors of the colorbar in order to distinguish better the values of the correlation. We modify also the colorbar of the Fig. 4 in order to have the same colors tones in all the figures.

Q15. Line 186-87. Please, clarify the step 2 to build the composite. I do not understand what the authors refer as ‘time steps’ and why only 2 values per month are obtained.

Reply #15. We clarify the methodology used for obtaining composites with the following example for the positive NAO phase. The steps are the following:

1. *We select the grid point with maximum correlation (both positive and negative values) between SWH₉₉ and the climatic index.*

In this case point #1 presents the maximum positive correlation with the NAO and for simplicity we plot SWH for January 1979 (the process is done for the whole time series, repeating this methodology for all the months).

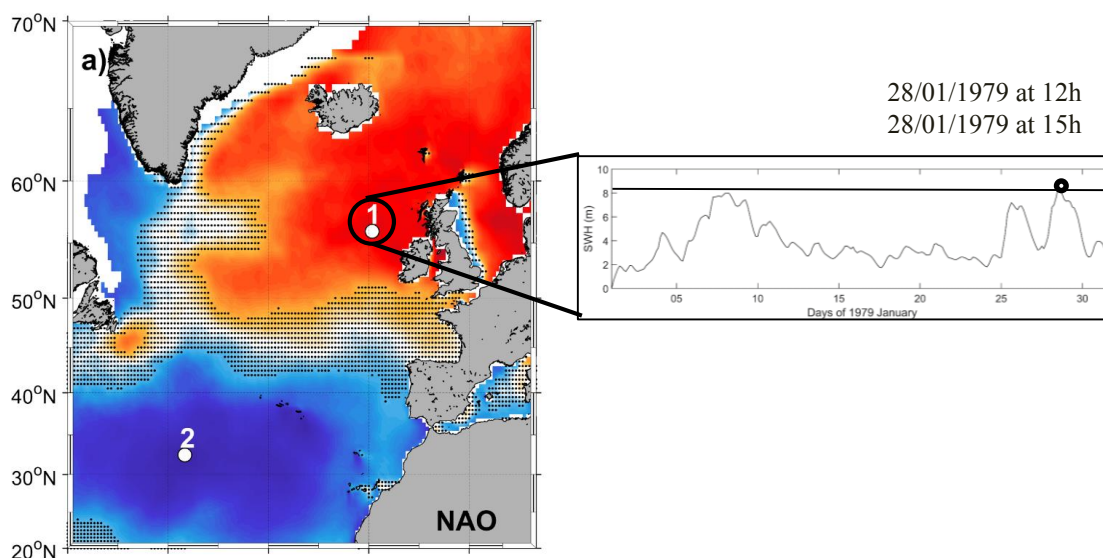


2. *From the 3-hourly dataset the following data are selected:*

$$SWH_{3\text{-hours}} \geq SWH_{99\text{ month}}$$

As the SWH time series are of 3-hourly time step, we have 224 data in months of 28 days, 232 data in months with 29 days, 240 data in months with 30 days, and 248 in months with 31 days. The 99th percentile is defined as the value that only the 1% of the time series exceed it. And with our time step, the 1% of the number of data for each month those are higher than SWH₉₉ is only 2 values.

For January 1979 and for the positive phase NAO composite is concerned, the values that exceeded than SWH₉₉ for that month occur at 12h and at 15h on 28th of January 1979.



3. The U10, SLP and SWH values for the chosen moments in point #2 of this methodology are selected in whole study area.

For our example, we take the values of U10, SLP and SWH for the 12h and 15h on 28th of January 1979 in all the grid points.

4. The mean value of the selected data in point #3 of this methodology is calculated for each spatial point. The results are shown in figures 9-12 of the Ms.

We calculate the mean value for U10, SLP and SWH on 28th January 1979 at 12h and 15h in all the grid points.

In addition, we have added this sentence in order to help to understanding better the fact of referee comments, in lines #202-203 of the Ms.

“- We select the time steps for which the original 3-hourly SWH time series at points #1 and #2 exceed SWH₉₉ (2 values each month are selected because the monthly number of data are 224-248 depending on the month of the year).”

Q16. Lines 189-91. As far as I understand how the composite is built, the composite maps only represent spatial patterns for a target location with a high correlation between winter SHW99 and a climate index. They are not synoptic maps (a map for a given moment of time).

Reply #16. Yes. The composites represent the mean situations of U10, SLP and SWH that take place when the extreme waves and the different climatic indices have the higher value of correlation with each specific index. We use the term 'synoptic' for the spatial dimension (similar to the 'mesoscale' term largely used in oceanography) where a map is obtained for the whole basin.

Q17. Figures 9 & 10. The magnitude of the wind vector and SLP would help the reader to better understand the resulting maps.

Reply #17. The wind speed reference is defined in the bottom left part of each subplot as a red vector of 5 m/s. And regarding the reference of SLP, if we put the values of each contour, the visualization of the wind arrows will be difficult, for this reason we decided to remove them. In addition, the important point of these figures is to know the low and high pressure distribution, and how the gradient pressure determines the wind speed and therefore the extreme waves.

Referee #2: David Woolf

Q1. The study investigates the relationship of extreme wave heights to atmospheric modes based on a high-quality wave hindcast. The study is described concisely and clearly. There are not any great revelations, but a useful study is reported fairly and competently in an appropriate form and to a suitable journal. I have some detailed comments, which are given below, but I am content for a revised version to be published after reasonable attention to all reviewer comments.

Reply #1. We sincerely thank reviewer for his constructive comments and the effort that he made in reviewing carefully our work. We deeply think that thanks to this revision the new version of the manuscript (hereinafter Ms.) improved significantly.

SPECIFIC COMMENTS

Q2. I do not have a problem with calculating the linear trends by a simple method and reporting these (Figure 4 and associated text), But I'd urge caution in interpretation. Firstly, there is likely to be some autocorrelation in the atmospheric forcing (and thus wave heights), which makes the independence of values assumed in simple regression doubtful. Secondly, the particular time period, 1979-2009 is pertinent; different periods would show different patterns.

Reply #2. We totally agree with the referee's comment. We consider relevant to write a new sentence in the Ms. in order to clarify this issue (lines #116-117):

"These calculations are restricted to the time period corresponding to the atmospheric forcing of the WAVEWATCH III and different patterns could be obtained for different periods".

Q3. A similar weakness in estimation of statistical significance is apparent in the use of "t-values" for the relationship of PCs to climate indices (lines 130-136), but again that is a minor objection and should not discourage publication. It is possible to take the statistical analysis further, for example through a wavelet analysis of the wave height - climate index relationship (I have seen this for sea level, but I am not aware of such an analysis for wave heights), but that is for another paper.

Reply #3. We agree with the referee's comment again. The lines #151-152 of the Ms. have been rewritten in order to note this issue, as:

"These significance values are particular for this study since they depend on the data used and the analyzed time period."

We agree that a wavelet analysis would complement these kind of studies. In order to keep the focus of the work we prefer to perform them in a new Ms. However, we pointed this comment in the Summary and conclusions section. Lines #291-293 of Ms. read:

"For future studies, a wavelet coherence analysis (Torrence and Compo, 1998) between the main climatic indices and the extreme waves will provide additional information on the dominant modes of variability and how they vary in time."

Q4. My general impression is that, as for some previous studies, the relationship of extreme waves in the North Atlantic to NAO is very convincing, but (albeit with calculated significance) the other relationships are rather weak (possibly with the exception of SCAND).

Reply #4. Yes, the NAO index is the most relevant climatic mode of the North Atlantic Ocean, being noteworthy the effect of the SCAND index. However, we include the rest of climatic indices (EA and EA/WR) analysis because they have a significant influence on the extreme waves in the Mediterranean basin. And, as we mention in the Ms., we use the North Atlantic Ocean study as a "validation" of the methodology developed and then applying it in the Mediterranean Sea, since the North Atlantic wave climate has been largely analyzed in previous studies.

Q5. It seems to me to be open to debate if we understand the extreme wave heights better from modest correlations to atmospheric indices. The relationship to composite was more interesting and in some respects was more convincing. For example, the relationship to EA is physically sensible and quite satisfying in this form.

Reply #5. We agree with the referee in this aspect, we think that the composite analysis is the most innovative and interesting study in this paper, being a more suitable way to relation the climatic modes with the extreme waves. However, the proposed methodology needs to compute the correlation in order to select the spatial point where the climatic patterns have a higher influence. In addition, it can be useful for the better understanding of the relation of the climatic indices and the extreme waves for the reader.

Q6. The sections from lines 193-217 and from 219-246 are rather monotonous and not very effective in communication. I suggest finding a more engaging method of communicating this information, perhaps a Table?

Reply #6. We are not sure if the referee talks about the description of the composites or about the relation between the composites and the EOFs. If it is the first case, we consider these sections as descriptive (we only describe the composite results; Fig. 9-12), being the composite figures the clearer and concise way to transmit the information. If, on the other hand, he refers to the relation with the EOFs and PCs, the table which show that is the Table 2 (in new Ms., old Table 1) where the correlation between the main climatic indices and the principal modes appear. We want to remark that we obtain the same results if we compare the signals of PCs with the different climatic indices and the EOFs maps with the composites.

Q7. I do not have any strong objections to the content of “Summary and conclusions” though I can be counted as a sceptic regarding simple projections of NAO behaviour and their utility in projections of extreme waves.

Reply #7. We think that the climatic indices signal is more stable than the extreme waves, and it can be a suitable way to know any probable behavior of them. However, we are aware that when we mention the future projections, we are talking about statistics, but if we improve somehow the extreme waves prognostic, this paper will have met its target.

Q8. The abstract adequately describes the topic and principal results, but gives no explanation of the methodology beyond “31-year wave model hindcast”. I suggest adding another sentence.

Reply #8. The following sentence has been added in lines # 7-10 of the new Ms.,

“A new methodology for analyzing the atmospheric signature associated with extreme waves is proposed. The method obtains the composites of Significant Wave Height (SWH), mean sea level pressure (MSLP) and 10 m-height wind velocity (U10) using the instant when specific climatic indices have the stronger correlation with extreme waves.”

Q9. Line 64-65. “we assume that wave climate is constant for 3 hours”. I interpret that phrase as an assumption of an autocorrelation period of 3 hours, is that correct? Was the data analysed to reach this conclusion? Does it have any implications beyond simply informing using 3-hour data?

Reply #9. For a limited period of time and in a particular geographical region, wave conditions vary in a stationary way. For this reason, it is commonly assumed that a sea state remains stationary for 3-6 hours. The objective of the sentence in the Ms. is only to explain that the time step is suitable in order to do the presented analysis in this paper, since we follow the common physical assumption of the wave state where the waves features are stationary during 3 hours (the time step of our dataset).

Q10. Line 76 and following: There are ~240 3-hour values in each calendar month. How exactly is the 99th percentile calculated? (An interpolation between the second and third highest values for each month?)

Reply #10. The 99th percentile is calculated as the value that only the 1% of the monthly data exceed it. Therefore, this value is interpolated between the second and third highest values for

each month since we have 1 data each 3 hours (with ~240 data/month only 2.4 data exceed the 99th percentile value).

Q11. Line 82. How good is a fit of annual and semiannual sinusoidal to the seasonality? Was there any analysis for higher harmonics?

Reply #11. Fitting the seasonality of the monthly SWH99 series as a cosine function through a least squares adjustment, we remove the annual and semiannual frequencies in the both basins. We check this method with a fast Fourier transform analysis, comparing both signal (with and without seasonality) obtaining that the annual and semiannual periods are removed. In addition, we verify that there is any dominant higher frequency in time series. This is because when computing the extreme waves any periodicity inside a month is erased.

Q12. Line 93. spelling “cyclogenetic”

Reply #12. This has fixed.

Q13. Line 95. Perhaps change “adjusted though a first order polynomial : : :” to “fitted by a linear regression in time”?

Reply #13. This has been fixed.

Q14. Line 101. Change “northern” to “north”.

Reply #14. This has been fixed.

Q15. Line 116. “periodicity : : : around 5 years” Not wrong, but perhaps risky? I would generally avoid talking about periodicity unless there is a very strong case.

Reply #15. We modified the Ms and we now explain that we computed it through a FFT analysis of the PC. For that, we add this information in #129-130 lines of Ms. as,

“The first EOF, which explains a 28.5% of the winter SWH99, presents a periodicity in its PC around 5 years (calculated through FFT analysis of Fig. 5-1).”

Q16. Line 145, “: : : being the rest of correlations marginal”. I could not make sense of this line!

Reply #16. We want to remark that the rest of the correlations are not significant. However, we agree with the referee’s comment in that this line is misleading, and thus we remove this part of the sentence.

Q17. Line 196 “: : : leads a wind jet”. I suggest “this composite is characterized by a strong westerly wind stream : : :” N.B. A similar relationship was demonstrated dynamically by Wolf and Woolf (2006; GRL 33(6)).

Reply #17. This has been changed.

Q18. Add gratitude to NCEP and NOAA CPC for data in Acknowledgments?

Reply #18. Thank for the referee’s advice. We write the following sentence in lines #311-312 of Ms:

“In addition, authors thank NCEP and NOAA CPC for the free available data that have been used in this article.”

Extreme waves and climatic patterns of variability in the Eastern North Atlantic and Mediterranean basins

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Abstract. The spatial and temporal variability of extreme wave climate in the North Atlantic Ocean and the Mediterranean Sea is assessed using a 31-year wave model hindcast. Seasonality accounts for 50% of the extreme wave height variability in North Atlantic Ocean and up to 70% in some areas of the Mediterranean Sea. Once seasonality is filtered out, the North Atlantic Oscillation and the Scandinavian Index are the dominant large-scale atmospheric patterns that control the interannual variability of extreme waves during winters in the North Atlantic Ocean; and to a lesser extent, the East Atlantic Oscillation also modulates extreme waves in the central part of the basin. In the Mediterranean Sea, the dominant modes are the East Atlantic and East Atlantic/Western Russia modes which act strongly during their negative phases.

A new methodology for analyzing the atmospheric signature associated with extreme waves is proposed. The method obtains the composites of Significant Wave Height (SWH), mean sea level pressure (MSLP) and 10 m-height wind velocity (U10) using the instant when specific climatic indices have the stronger correlation with extreme waves.

Copyright statement.

1 Introduction

The accurate assessment of extreme wind-wave conditions is essential for human activities e.g., maritime traffic and wave energy generation and is a major source of coastal hazards.

Besides, extreme waves influence the upper ocean by enhancing vertical mixing through the Stokes layer (Polton et al., 2005).

Extreme waves reaching port areas, determine the design and operation of coastal and offshore infrastructures, they are also responsible for coastal flooding at intra-annual scales (Orejarena-Rondón et al., 2019). Waves are the ocean surface response to the wind stress acting over it and therefore there is a direct connection between surface atmospheric circulation and waves (Lin et al., 2019).

The study of extreme waves at different temporal scales has been extensively addressed in several works (Wang and Swail, 2001, 2002; Caires et al., 2006; Méndez et al., 2006; Menéndez et al., 2008, 2009; Izaguirre et al., 2010, 2012; Young et al.,

20 2012; Weiss et al., 2014; Sartini et al., 2017). Most of these studies focused on the spatio-temporal distribution of extreme waves rather than on the atmospheric conditions producing them.

Over the North Atlantic Ocean and the Mediterranean Sea, atmospheric circulation is driven by the temperature gradient between the North Pole and the Equator that organizes three-cell system associated with the Equatorial-Low, the subtropical Azores-High, the Iceland-Low and the North Pole-High pressure centres (Martínez-Asensio et al., 2016). Atmospheric cir-
25 culation can be characterized by specific modes of variability with defined characteristics that may also have effects over a region or remote areas through atmospheric teleconnections (Wallace and Gutzler, 1981). The main patterns of atmospheric variability on the North Atlantic and Europe are the North Atlantic Oscillation (NAO), the East Atlantic pattern (EA), the Scandinavian pattern (SCAND) and the East Atlantic/Western Russia (EA/WR) patterns (Barnston and Livezey, 1987). NAO is the leading mode of variability in the North Atlantic and is often defined as the sea level pressure difference between the
30 Iceland Low and the Azores High (Hurrell et al., 2003). NAO controls the strength and direction of westerly winds reaching the European coasts, as well as the location of the storm tracks across the North Atlantic (Marshall et al., 2001). The EA is the second predominant mode of low frequency variability in the North Atlantic area. It consists of a north-south dipole of anomaly over the North Atlantic, with a strong multidecadal variability. The SCAND pattern consists of a primary circulation center over Scandinavia, with weaker centers of opposite sign over western Europe. The EA/WR pattern consists of four main
35 anomaly centers; its positive phase is associated with positive wave height anomalies located over Europe and negative wave height anomalies over the central North Atlantic (<https://www.cpc.ncep.noaa.gov/data/teledoc/telecontents.shtml>).

There have been a number of studies that have tried to unravel the relation between wave climate and large scale atmospheric patterns. Woolf et al. (2002) found a strong connection between interannual wave climate variability in North Atlantic Ocean and NAO, and in a lesser degree with EA index. Castelle et al. (2018) examined the relation between winter-mean wave height,
40 detailing a high correlation with the NAO index and with the Western Europe Pressure Anomaly (WEPA) index; this is a new definition of a climatic pattern which based on the sea level pressure gradient between the stations Valentia (Ireland) and Santa Cruz de Tenerife (Canary Islands). Izaguirre et al. (2010) detected a relation between the NAO and EA indices with the extreme wave climate in the North-East Atlantic Ocean. Izaguirre et al. (2012) evaluated the synoptic atmospheric patterns associated to the extreme Significant Wave Height (SWH) finding a higher interannual variability of the extreme SWH in the northern
45 part of the Atlantic Ocean. In the Mediterranean Sea, clear relations between extreme waves and the negative phases of EA and the EA/WR indices have been also reported (Izaguirre et al., 2010).

In this paper, we extend earlier studies, analyzing the short and long term variability of extreme waves in the North Atlantic Ocean and in the Mediterranean Sea, not only for diagnostic purposes but also to be able to provide statistical prognostics of extremes waves associated to the most important climatic indices with some anticipation. The paper is structured as follows.
50 In Section 2, the data used and the description of the extreme waves are presented. In Section 3, we present the spatial and temporal distribution of the extreme waves as well as the relation between the four patterns of climatic variability and the spatial distribution of extreme waves during winter. Finally, Section 4 concludes the work.

2 Data and extreme waves values

2.1 Waves and atmospheric data

- 55 Wave data is obtained from a high resolution global hindcast from the National Center for Environmental Prediction (NCEP) with a temporal sampling of 1 hour and different spatial resolutions. This dataset (i.e., *WAVEWATCH III 30-year Hindcast Phase 2*, (Chawla et al., 2012)) has been generated by forcing the “state-of-the-art” wave model WAVEWATCH III (Tolman, 2009) with 10-m height high-resolution wind fields from the NCEP Climate Forecast System Reanalysis and Reforecast (CFSRR) a 30-year homogeneous data set of hourly $1/2^\circ$ spatial resolution winds.
- 60 The wave model consists of global and regional nested grids, developed by the presence of currents and bathymetry (Amante and Eakins, 2009), taking into account the conservation of action density (Janssen, 2008). In addition, the dissipation and physical terms parameterization formulated in Ardhuin et al. (2010) is used in this work.

The hindcast has been validated using the National Data Buoy Center (NDBC-NOAA) as well as with altimeter database provided by the Institut Français de Recherche pour l’Exploitation de la Mer (IFREMER) database (Chawla et al., 2011, 2012, 65 2013).

- The simulation spans a time period of 31 years from 1979 to 2009 with hourly outputs, although since we assume that wave climate is constant for 3 hours, we only use 1 data for this period; and with a spatial resolution varying according to the study area. In all the grids, the full resolution ETOPO1 bathymetry is used in regular spherical grids. The North Atlantic domain spans from 20°N to 70°N in latitude and 60°W to 10°E in longitude at 0.5° resolution (Fig. 1a). The Mediterranean Sea covers 70 from 30°N to 48°N in latitude and 7°W to 43°E in longitude with a spatial resolution of 0.167° (see Fig. 1b). Sea level pressure and wind velocity at 10 meters height are provided by the NCEP-CFSR forcing with a resolution of 0.5° for the same period (Saha, 2009).

The hindcast is validated at the two basins with available buoys following the methodology described in Morales Márquez et al. (2018)). For 13 buoys in the Atlantic and 11 in the Mediterranean Sea, we compute the correlation coefficient (R^2), Scatter Index (SCI) and the Relative Bias (RB) defined as:

$$R^2 = \frac{\text{Cov}(b, m)}{\sigma_b \sigma_m}, \quad (1)$$

$$SCI = \frac{\text{rms}_{m-b}}{\max(\text{rms}_b, |\langle b \rangle|)}, \quad (2)$$

$$RB = \frac{\langle m - b \rangle}{\max(\text{rms}_b, |\langle b \rangle|)}, \quad (3)$$

being m the modeled dataset by *WAVEWATCH III 30-year Hindcast Phase 2* and b the *in situ* buoys dataset.

Table 1.

Statistical comparison between the *WAVEWATCH III 30-year Hindcast Phase 2* and CMEMS buoys.

Basin	R^2 (%)	SCI	Relative bias
Mediterranean Sea	73.09 ± 0.17	0.39 ± 0.00	-0.13 ± 0.00
North Atlantic	72.67 ± 0.96	0.33 ± 0.02	0.16 ± 0.02

Table 1 shows the comparison between the *WAVEWATCH III 30-year Hindcast Phase 2* and the buoys from Copernicus Marine environment monitoring service (CMEMS, <https://marine.copernicus.eu/>). See Fig. 1 buoys locations.

75 Leading climatic modes of variability, namely NAO, EA, EA/WR and SCAND (see introduction Section for a description of these modes) have been downloaded from the NOAA Climate Prediction Centre. Indices are constructed through a rotated principal component analysis of the monthly mean standardized 500-mb height anomalies in the Northern Hemisphere ensuring the independence between modes, at a monthly scale, due to orthogonality (Barnston and Livezey, 1987).

2.2 Extreme wave climate

80 Extreme wave climate is here defined in terms of the monthly 99th percentile of SWH (hereinafter SWH₉₉). Over the North Atlantic, maximum values of SWH₉₉ during the whole period of time analyzed (1979-2009), are observed at mid to high latitudes with values reaching 13 m (see Fig. 2a). This situation is very similar to the obtained in Vinoth and Young (2011), where there are higher values of SWH in the northern part of the study area. These maximum values befall predominantly during winter season (DJFM), with an 81.2% of occurrence (Fig. 2c). Over the Mediterranean Sea, maximum values of SWH₉₉
85 are at most 8 m (Fig. 2b) with a 91.06% of occurrence during winter season (DJFM) (Fig. 2d).

Seasonality is assessed by fitting a cosine function to the monthly SWH₉₉ series through a least squares adjustment (Menéndez et al., 2009),

$$f(t) = \sum_{i=1}^2 A_i \cos\left(\frac{2\pi}{T_i}(t - \phi_i)\right) : \quad (4)$$

where, $i = 1, 2$ are the annual and semiannual cycle, A_i the amplitude, ϕ_i the phase, $T_{1,2} = 365.25$ and 182.63 and t time in
90 days. The monthly SWH₉₉, for the location with the largest variance reduction in the North Atlantic (point 1, Fig. 3a), is shown in black in the top panel of Fig. 3 (the same for the Mediterranean Sea, point 2 in Fig. 3b in the bottom panel), while the time series after removing seasonality by fitting Eq. (4) are shown in blue. Seasonality in the North Atlantic accounts on average for a 50% of the variance of the signal (Fig. 3a). It means that the half of the extreme waves signal is explained with the annual and semiannual cycle.

In the Mediterranean Sea, there are two different areas in terms of seasonality. One is located in the central basin where seasonality explains up to a 70% of extreme waves and the other located in the Gulf of Genova and the Alboran Sea, where seasonality explains less than 10% of the signal (Fig. 3b). These areas are very active in terms of cyclogenetic activity (Trigo et al., 2002)) and thus the seasonal signal is relatively less important here.

To analyze the long term trend of SWH_{99} during winter months (DJFM) where almost most of extreme waves occur, the temporal series is fitted by a linear regression in time at the 90% confidence level in each spatial point (see Fig. 4). Locations where the trend is not significant are represented with a dot. The 31-year trend in the North Atlantic (Fig. 4a) displays an area with significant positive values (up to 2.5 cm/year) from Portugal coasts to Canada. The rest of the basin presents a negative value of tendency, with maximum values around 3.5 cm/year in Bay of Biscay, Labrador Sea and between United Kingdom and Iceland.

This aligns with the obtained results in Gallagher et al. (2016), where the future projections of mean surface wind show an average decrease over the North Atlantic Ocean for winter season, so likely the extreme waves continue with the same pattern in terms of long-term variability.

In the Mediterranean Sea, the values of SWH_{99} tendency during winter months (DJFM) are substantially smaller (see Fig. 4b). Only the center part, north of Cyprus (with negative values up to 2.4 cm/year and 1 cm/year, respectively) and the Aegean Sea (with positive values of trend around 1 cm/year) present a trend statistically significant.

These calculations are restricted to the time period corresponding to the atmospheric forcing of the *WAVEWATCH III 30-year Hindcast Phase 2* and different patterns could be obtained for different periods.

These results differ from the trend which is calculated in Young et al. (2011), since in that study they considered all the months of the year in order to calculate the monthly SWH_{99} 's trend, and in this work, we analyze the tendency of the monthly SWH_{99} only during winter season (DJFM). We have verified that there is a positive trend of SWH_{99} during the summer season (it is not shown in this study), however the values in this season are considerably lower than those found during winters.

In this paper the winter extreme wave climate is studied in order to remove the seasonality, because maximum values of SWH_{99} take place during the winter season.

3 Spatio-temporal patterns of extreme waves

The spatio-temporal variability of the SWH_{99} is assessed computing EOFs of the winter (DJFM) fields. Prior to the computation of the EOFs the spatial mean winter SWH_{99} is removed and the analyses have been performed onto anomalies with respect to the mean values (Ponce de León et al., 2016). Mean fields are mapped in Fig. 5a and 6a.

The first three EOFs for the North Atlantic are shown in Fig. 5 (b, c and d) and their principal components (PCs) together with their explained variance in Fig. 5 (1, 2 and 3).

The first EOF, which explains a 28.5% of the winter SWH₉₉, presents a periodicity in its PC around 5 years (calculated through a FFT analysis of Fig. 5-1).

120 This first mode shows a spatial dipole with opposite values in the north and south of the basin. The second mode which explains a 15.5% of the winter variability shows an area in the central basin separating two zones at the north and south with different sign (Fig. 5c). Values for the central part are three times larger than the ones obtained for the northern and southern sides, indicating that the contribution of this EOF is to increase/decrease the winter extremes in the central Atlantic when its PC is negative/positive. The third EOF which explains a 8.3% of the winter variability displays also three different zones with
125 a central area shifted to the East-West direction and extending from Bay of Biscay to the Celtic Sea and at the north and at the south zones displaying opposite sign during winter (Fig. 5d).

For the Mediterranean Sea, the first three EOF modes for winter are shown in Fig. 6 (b, c and d) and their PCs in Fig. 6 (1, 2 and 3). The first EOF explaining a 38.0% of the total variance, represents a spatially coherent increase/decrease of SWH₉₉ over the entire basin. The second EOF explaining the 15.1% of the variance shows differences between the eastern and western
130 basins. The contribution of this mode is to increase/decrease SWH₉₉ in the western Mediterranean and simultaneously a decrease/increase in the eastern Mediterranean according to the amplitude of the PC. Finally, the third EOF explaining a 7.3% of the variance displays two zones, the Tyrrhenian Sea and the southern part of the Gulf of Lion with the same sign and the rest of the basin with opposite behavior.

The relationship between extreme waves and the climatic modes of variability in the North Atlantic and Mediterranean
135 Seas is explored and quantified as follows. Winter averages of climate indices are first correlated with the corresponding PCs described above for each basin. The significance level is set at 90% with a t -value adjusted as,

$$t = |c| \sqrt{\frac{N-2}{1-c^2}}, \quad (5)$$

where c the correlation coefficient and N the length of the time series. If t is equal or higher than the t -value of a Student's t -distribution of $N-2$ degrees of freedom then the correlation is assumed to be statistically significant at the predefined 90%
140 confidence level.

These significance values are particular for this study since they depend on the data used and the analyzed time period.

3.1 Correlations between winter extreme waves and climatic modes of variability

The correlation between the four climate indices and the first three SWH₉₉ PCs are shown in Table 2 where (and hereinafter) bold indicates statistically significant correlations at the 90% confidence level. The major correlation in the North Atlantic is
145 obtained with the NAO and the SCAND through the first PC1 (correlation of 82.6% and -63.3%, respectively). This is not surprising, as winter NAO and SCAND indices are correlated themselves (note that, although monthly indices are orthogonal, this does not necessarily hold for seasonal or yearly averages). NAO teleconnection not only dominates the extreme values

Table 2. Correlation between main climate indices and amplitudes of the three first modes of the averaged monthly SWH₉₉ series of the North Atlantic Ocean and Mediterranean Sea for winters.

	North Atlantic Ocean			Mediterranean Sea		
	PC1	PC2	PC3	PC1	PC2	PC3
NAO	0.826	−0.138	0.323	0.242	−0.193	0.297
EA	−0.120	-0.459	−0.042	0.298	-0.361	0.026
EA/WR	0.057	0.171	−0.127	0.093	-0.342	0.366
SCAND	-0.633	−0.126	−0.165	−0.091	0.347	-0.390

of SWH during winter season; but also the mean SWH, wave period and peak wave direction magnitudes for wintertime in this region Gallagher et al. (2014). To a lesser extent, EA is correlated with the second PC (explaining around 16% of the winter variability). These results are in accordance with those obtained by Izaguirre et al. (2010) and Gleeson et al. (2019) where they show that extreme waves in the North Atlantic are related to the positive phase of NAO and with the negative of EA and SCAND. For the Mediterranean Sea, both the NAO and the EA are correlated with the winter SWH₉₉ through the first PC and some of the variability correlated by the negative phases of EA and EA/WR through the second PC. However, the values of the correlations in the Mediterranean Sea are significantly lower than those in the North Atlantic, because the main climatic patterns consist in some strong poles located at Atlantic Ocean which drive zonal flows toward Europe. These climatic situations generate a weak circulation into the Mediterranean Sea, which is not as related to the higher values of waves. In addition, wave climate depends on wind regimes and on land-sea distribution. In other words, waves need fetch to develop and at the Mediterranean Sea, the available distance is more restricted Lionello and Sanna (2005).

Atlantic ocean

Correlation maps for winter SWH₉₉ in the North Atlantic and the four climate indices are displayed in Fig. 7. Some of these spatial correlations present similarities with the EOFs patterns shown in Fig. 5. In particular, the correlation map between NAO and SWH₉₉ (Fig. 7a) mimics the first SWH₉₉-EOF for winter (Fig. 5b) with correlation values consistent with the one obtained using PC1 (see Table 2). The correlation between EA and SWH₉₉ (Fig. 7b) shows large similarities with the second SWH₉₉-EOF for winter extreme waves (see Fig. 5c), but with the opposite sign. Finally, the correlation between SCAND and SWH₉₉ (Fig. 7d) shows similarities with the first SWH₉₉-EOF (Fig. 5b). During winter, the northern part of North Atlantic Ocean has a positive correlation with the NAO and negative at the south with maximum values close to 0.8 (Fig. 7a). Correlation map for the EA index shows positive correlations with maximum values close to 0.75 in the central part of the basin and negative at the north and south with maximum values of 0.47 (Fig. 7b). Correlation map for the EA/WR displays an area of negative correlation extending from the Bay of Biscay to Greenland and also near the west coast of Africa and at the northern and central basin appear two zones with positive correlation, with maximum values around 0.55 (Fig. 7c). Finally, correlation map between

SCAND and SWH₉₉ shows negative correlations in the central Atlantic with maximum correlations of 0.74 and positive in the central Atlantic with maximum value of 0.60 (Fig. 7d).

Mediterranean Sea

Correlation maps between winter SWH₉₉ in the Mediterranean Sea and the four climatic indices are shown in Fig. 8. Contrary to what is found in the North Atlantic, maps of correlations between extreme waves and climate modes are not clearly linked with the EOFs patterns of the wave field. The NAO index presents negative correlation in the whole Mediterranean basin (Fig. 8a). At the eastern side of the domain, the Adriatic and the Aegean Sea, present maximum correlations with values around 0.50. Correlation is positive only in the Ligurian Sea, with a value around 0.30. The EA index displays also negative correlation in the whole domain (Fig. 8b) with larger values over the west with correlation around 0.60. The correlation map between EA/WR and SWH₉₉ shows negative correlations in the Tyrrhenian Sea, the Adriatic and the Ionian Seas with maximum correlation of 0.40 (Fig. 8c). Positive correlation is obtained in the Aegean Sea with maximum values around 0.60. Finally, the correlation map between SCAND and SWH₉₉ shows large positive correlations in the Gulf of Lions, at the southern and central Mediterranean Sea and in the Adriatic Sea with values of around 0.50.

3.2 Synoptic atmospheric composites associated to extreme wave patterns

The analysis of the atmospheric signature associated with extreme SWH is performed by computing the composites of extreme SWH, atmospheric mean sea level pressure (MSLP) and 10 m wind velocity (U10). The objective is to find the atmospheric pattern that is associated with the extreme winter waves. The procedure to build the composites is as follows:

- First, we select the locations with the highest correlations between SWH₉₉ and each of the atmospheric indices (points labeled as # 1 for maximum positive correlation and as #2 for maximum negative correlation in Fig. 7 and Fig. 8 for the North Atlantic and Mediterranean Sea respectively).
- We select the time steps for which the original 3-hourly SWH time series at points #1 and #2 exceed SWH₉₉ (2 values each month are selected because the monthly number of data are 224-248 depending on the month of the year).
- Finally, we compute the composites for SWH, U10 and MSLP over the whole domain for all selected dates.

Note that the locations numbered as #1 and #2 in each map represent the largest positive and negative correlations with the corresponding index. The composite maps are thus interpreted as the synoptic patterns associated to positive and negative phases (respectively) of the climate index, leading to extreme waves.

Atlantic ocean

The composite for U₁₀ and MSLP built using location #1 (positive correlation between SWH₉₉ and NAO) (Fig. 9a) shows the typical configuration associated with the positive NAO phase that is characterized by low pressures across high lati-

tudes in the North Atlantic and high pressures over the central North Atlantic, the eastern United States and western Europe.

200 This composite is characterized by a strong westerly wind stream crossing the central North Atlantic whose fetch generates large waves at the western part of the British Islands and south of Iceland ($SWH > 9$ m).

A similar relationship was demonstrated dynamically by Wolf and Woolf (2006).

By contrast, in the south of the North Atlantic Ocean, in the Azores region, positive NAO phase results in low SWH_{99} values ($SWH \approx 4$ m). This pattern corresponds to the first EOF (Fig. 5a and Fig. 5-1 for the spatial mode and its amplitude
205 respectively) when, for positive values of the PC, positive anomalies are presented in the north part of the basin, and negative anomalies in the central part (the opposite for negative values of the PC). Composite for the positive phase of EA shows a similar structure than the one obtained for the positive phase of NAO, but with the North Atlantic cyclone shifted southwards and with the high pressures covering the entire Atlantic at 30° N (Fig. 9b). Maximum waves associated with the EA are obtained in the central Atlantic as the result of the southwards winds blowing from Greenland. This pattern corresponds to
210 the second EOF (Fig. 5c and Fig. 5-2). Composite for the EA/WR positive phase shows the low pressure system at 40° W, with the maximum extreme waves located to the east of Newfoundland (Fig. 9c). Finally, the atmospheric composite for the positive SCAND phase shows a cyclone (at 40° N) generating extreme waves smaller than the obtained with the previous three composites -values of SWH below 5m- Fig. 9d. This pattern is associated with the first EOF when its amplitude takes negative values (see Fig. 5b and Fig. 5-1) corresponding also with the atmospheric situation related with the negative phase
215 of NAO (see Fig. 10a). For the negative EA phase, the cyclone is located between Greenland and Iceland, generating a strong wind jet from the coast of Canada to Ireland (Fig. 10b). At this point, we want to remark that since here we are analyzing the negative correlations, the values displayed in Table 2 have to be changed in sign. The second EOF (Fig. 5c) according to Table 2 is related with the composite built for the maximum negative correlation between SWH_{99} and the EA index (Fig. 10b). Composites for the negative phase of EA/WR, are characterized by a western shift of the Iceland low that generates strong
220 zonal winds between 50° N and 60° N with maximum extreme waves located between south of Ireland and North of Spain (Fig. 10c). The Iceland low for the negative SCAND phase, is shifted northeast of Iceland with winds blowing southwestwards and extreme waves located between Iceland and Great Britain (Fig. 10d). This composite is associated with the first EOF (Fig. 5b), thus having a correlation with the positive NAO phase (see also Fig. 9a for comparison).

Mediterranean Sea

225 In the Mediterranean Sea, for the positive correlations between indices and extreme waves, we choose locations near the coast since negative correlations dominate the entire basin (Fig. 8). The atmospheric composite for the positive NAO phase displays a low pressure system in the north of Italy with associated eastwards winds in the western and central basins (Fig. 11a). These conditions are strongly associated with the atmospheric situations discussed in Trigo et al. (1999) for the cyclogenetic activity during winters. This composite is related with the distribution of extreme waves shown by the third EOF (Fig. 6d). Note that
230 the amplitude of this mode is positively correlated with NAO according to Table 2. The composite for the positive EA phase shows an intense cyclogenetic activity in the eastern Mediterranean Sea with its center of action over Cyprus which generates

strong winds and waves north of Egypt (Fig. 11b). This index, as shown in Table 2, is negatively correlated with the amplitude of the second EOF whose pattern presents large values for the extreme wave anomalies over the eastern Mediterranean Sea (Fig. 6c); in other words, a positive EA results in larger SWH₉₉, in the eastern basin as displayed in Fig. 11b. Regarding the positive EA/WR phase, the resulting composite presents a very similar pattern for surface pressure, winds and waves than the one obtained for the positive EA phase (see Fig. 11b and Fig. 11c). The EA/WR index is also negatively correlated with the amplitude of the second EOF (Table 2) resulting in the same distribution of extreme waves as previously explained regarding the EA. Finally, the composite for the positive SCAND index displays a cyclonic structure on the northwestern part of the Mediterranean Sea -between Corsica and Sardinia- with winds blowing southwards at the Gulf of Lion (Fig. 11d). The SCAND index is positively correlated with the amplitude of the second EOF (Table 2), indicating larger/smaller SWH in the western/eastern Mediterranean during its positive phase.

For the negative phases during winters, the composite for the NAO displays a weak cyclone over the Ligurian Sea (Fig. 12a). In this situation, however, the pressure gradient is weaker and due to the small fetch the resulting extreme waves are small (below 3m in SWH). Regarding the negative phase of the EA index, the composite shows a strong cyclone centered over Italy with a large pressure gradient over the northwestern Mediterranean Sea. This situation generates strong winds between the Balearic Islands and Corsica and Sardinia, generating large waves in this area. EA index is negatively correlated with the amplitude of the second EOF (see Fig. 6c). For the negative phase of EA/WR, composite shows a low pressure system over the Ionian Sea with a strong pressure gradient between Sicily and Tunisia, resulting in large extreme waves in this passage (Fig. 12c). This index is negatively correlated with the amplitude of the second EOF (Table 2) suggesting that for positive anomalies of SWH₉₉ (see Fig. 6c) a negative phase of the EA/WR index results in an increase of extreme waves. Finally, the composite for the negative phase of the SCAND index displays a low pressure system over north of Italy. Although the pressure gradient associated with this low system is intense north of Corsica, the small fetch impels the formation of extreme waves (Fig. 12d).

4 Summary and conclusions

This work presents a new methodology to study the extreme wave climate and the atmospheric synoptic conditions responsible for the extreme waves. Winds and pressure are obtained by computing the composites corresponding to the monthly values of the 99th percentile of the significant wave height. As a result, it is possible to infer changes in the location and intensity of extreme waves through the understanding of the variability of climatic patterns (widely studied). This approach could be of interest for all the activities related to the prognostic of extreme waves, such as, the design of offshore structures, among others.

The study is focused on the North Atlantic Ocean and the Mediterranean Sea, although the methodology can be extrapolated to any region in order to get a deeper insight on the seasonal and interannual variability of extreme wave climate. In the present work, the interannual variability has been analyzed using Empirical Orthogonal Functions, which have been correlated against the four main climate indices of variability in the area, i.e., NAO, EA, EA/WR and SCAND. Finally, the most reliable atmospheric situation associated with each climatic pattern is discussed using the described methodology.

265 The extreme waves climate has a large seasonal signal both in the North Atlantic Ocean and in the Mediterranean Sea. Our results also indicate a large intra-annual variability in the central part of the Mediterranean Sea and lower in the Alboran and in the Ligurian sub-basins, in agreement with Sartini et al. (2017) where they exhibited different degrees of seasonality depending on the main mesoscale meteorological features of the locations analyzed. Concerning long term trend of extreme waves, it is predominantly negative although there are some areas, as in the center of the North Atlantic Ocean or in the Aegean Sea, where the value of the tendency is positive. These results are not in line with the studies of Young et al. (2011) and Young and Ribal (2019) because here, we assess only the extreme waves values during the winter season, when most of the maximum SWH occur.

Regarding climatic modes of variability, we found that the NAO and the SCAND indices are the leading modes of climatic variability affecting extreme waves in the North Atlantic Ocean during winters. The positive NAO phase increases extreme waves in the northern North Atlantic while the negative NAO phase results in an increase of extreme waves in the southern North Atlantic in accordance with Hurrell et al. (2003). By contrast, a positive SCAND index increases extreme waves in the southern North Atlantic while a negative SCAND index increases the extreme waves in the north part of the North Atlantic Ocean, as Martínez-Asensio et al. (2016) also pointed out. To a lesser extent, the EA also influences the extreme waves in the North Atlantic Ocean, as Izaguirre et al. (2010) concluded, too. While the positive EA phase drives extreme wave climate in the central North Atlantic, the negative phase controls extreme wave climate at higher and lower latitudes (see Fig. 7b).

For future studies, a wavelet coherence analysis (Torrence and Compo, 1998) between the main climatic indices and the extreme waves will provide additional information on the dominant modes of variability and how they vary in time.

The interannual variability of extreme waves during winters in the Mediterranean Sea is dominated, to a large extent, by the negative phase of EA with larger effect in the western basin. Positive NAO phase has also influence on extreme waves although they are smaller in the whole Mediterranean Sea.

285 Caires et al. (2006) reported that the wave climate is expected to change by a small amount in response to climate change (below 5% between 1990 and 2080). The results presented here could be used to project climate, and to develop appropriate studies for coastal protection improving numerical models and to define long term wave energy conversion strategies; since the climatic patterns of NAO will dominate in a larger extent the extreme waves climate at the North Atlantic Ocean in future scenarios, according to Gleeson et al. (2017).

290 *Data availability.* All data are accessible from <https://polar.ncep.noaa.gov/waves/hindcasts/nopp-phase2.php> and from <https://www.cpc.ncep.noaa.gov/data/teledoc/telecontents.shtml>

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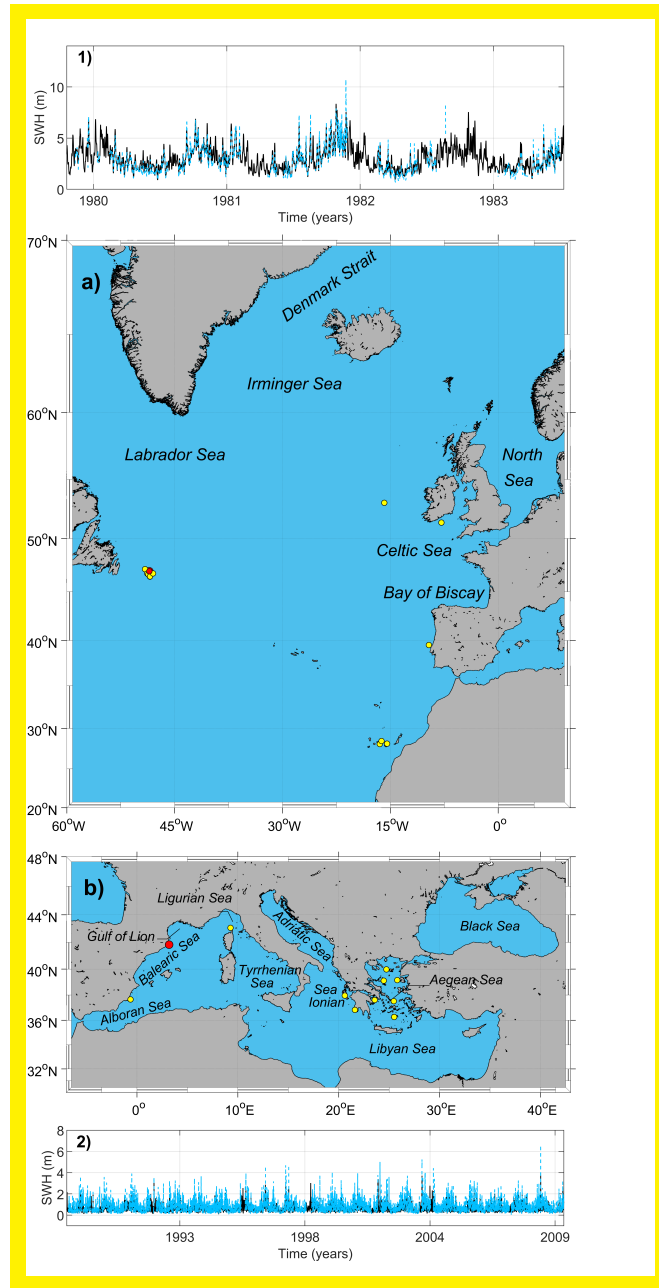


Figure 1.

Location of study zones. a) Eastern North Atlantic Ocean. b) Mediterranean Sea. The yellow points are the locations of the used buoys in the comparison with the modeled data by *WAVEWATCH III 30-year Hindcast Phase 2*. Panels 1 and 2: SWH series of hindcasts (black line) and a representative buoy (dashed blue line) of North Atlantic Ocean and Mediterranean Sea, respectively. The red points are the locations of the representative buoys.

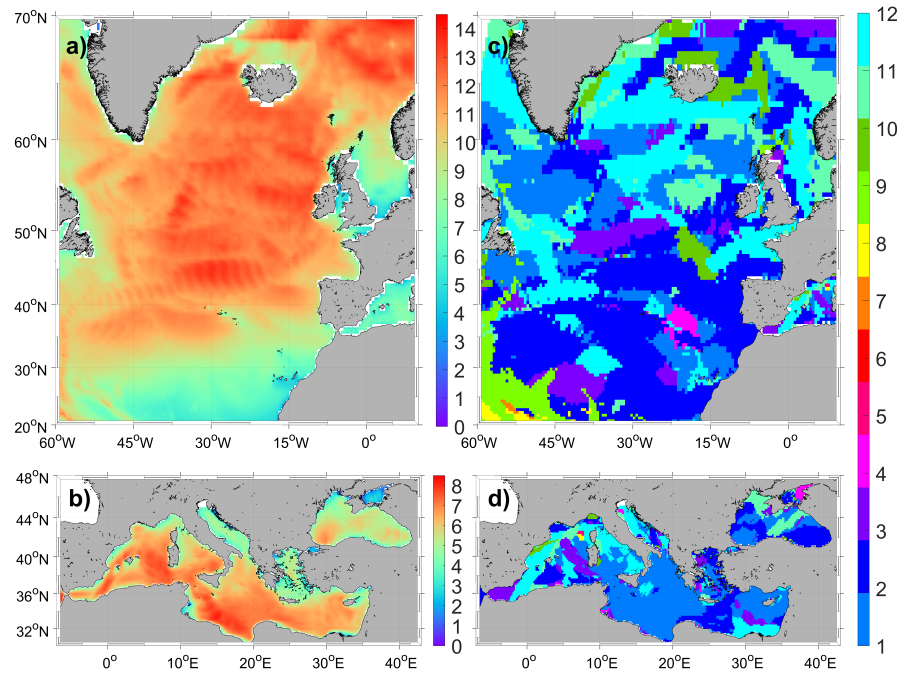


Figure 2. Maximum value of monthly 99th percentile SWH in m for a) North Atlantic Ocean and b) Mediterranean Sea. And the month of the year [from January (1) to December (12)] when there is the maximum value of 99th percentile SWH for c) North Atlantic Ocean and d) Mediterranean Sea.

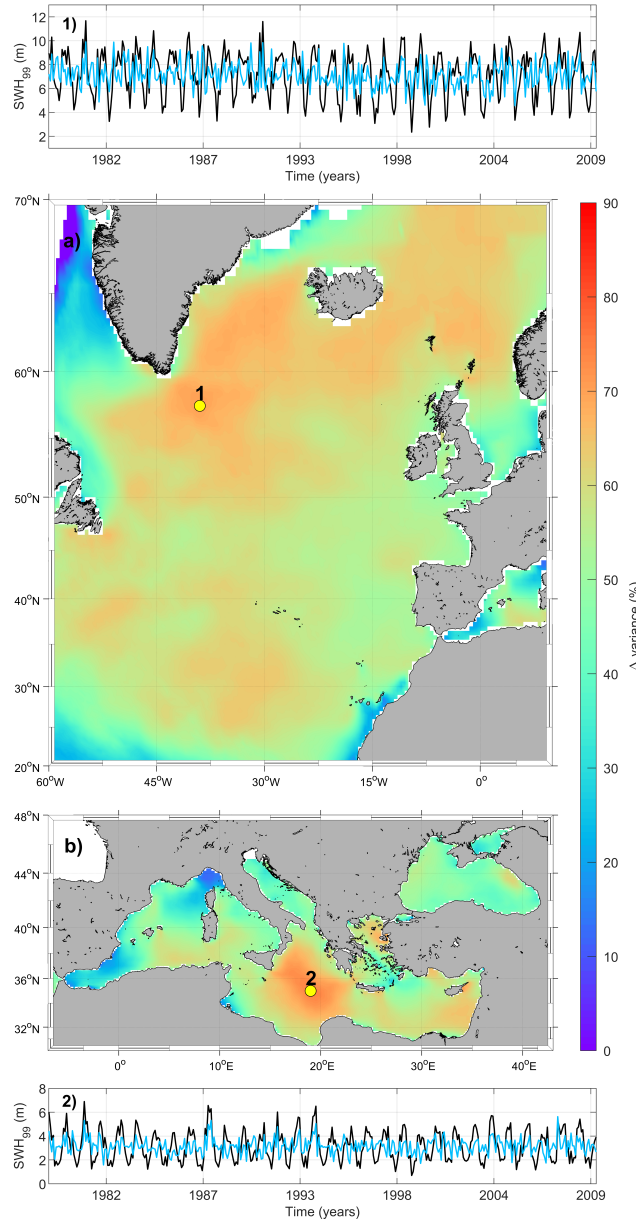


Figure 3. Variance reduction in percentage if the seasonality is removed of the monthly 99th percentile SWH series for a) North Atlantic Ocean and b) Mediterranean Sea. Panels 1 and 2: monthly 99th percentile SWH series (black line) and monthly 99th percentile SWH series without seasonality (blue line) for a point of North Atlantic Ocean and Mediterranean Sea, respectively.

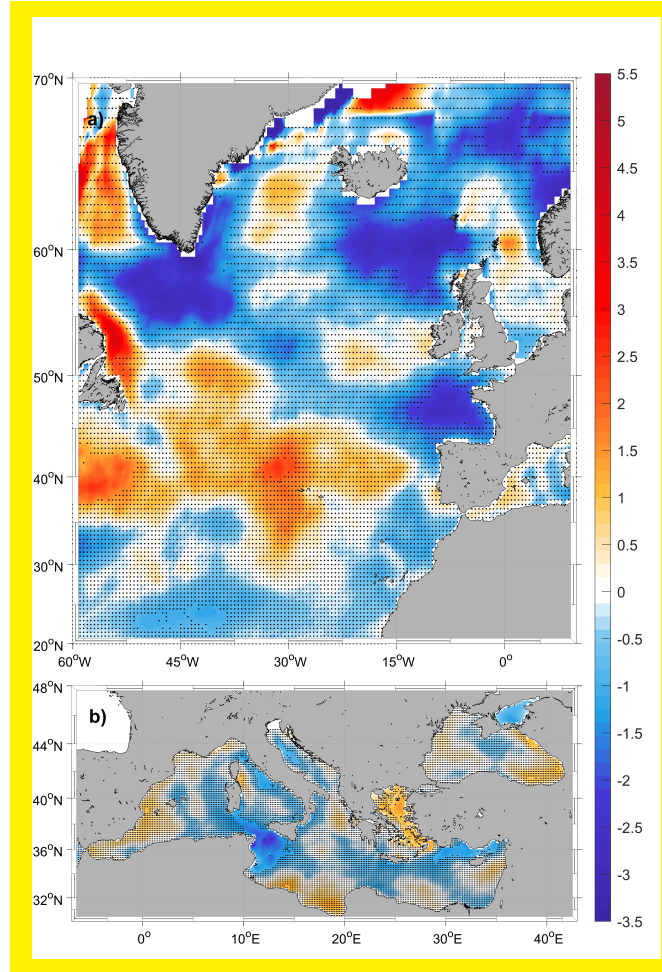


Figure 4. Trend of the monthly 99th percentile SWH during winters (DJFM) in cm/year. No significant values at the 90% confidence interval are dotted.

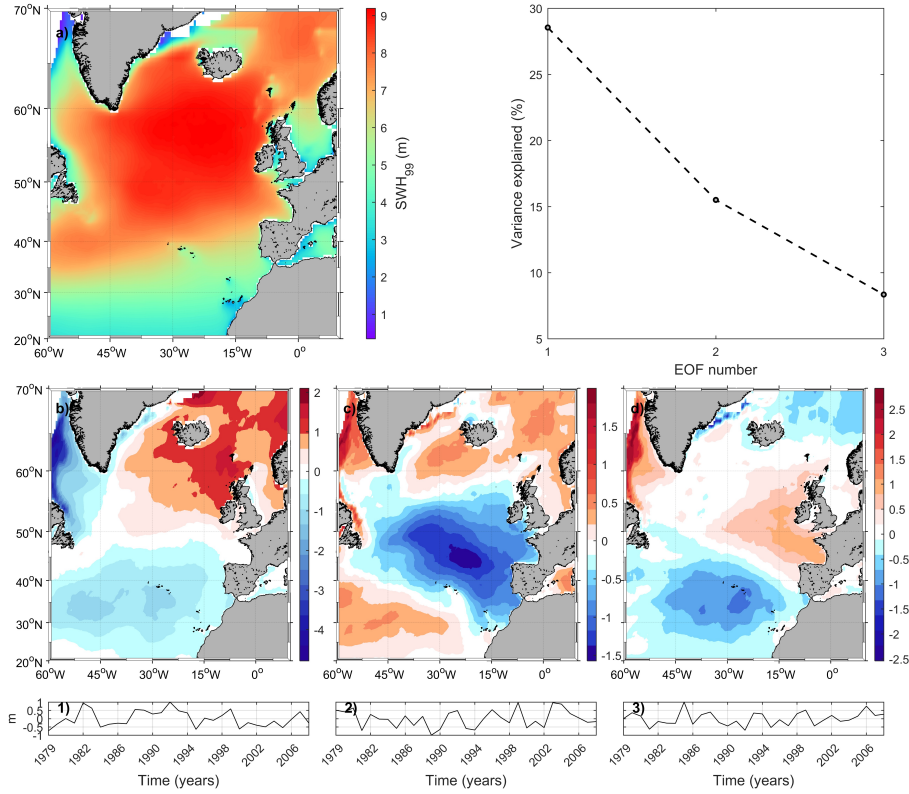


Figure 5. a) Mean field of winter SWH 99th percentile over the North Atlantic. EOF analysis of SWH₉₉ anomalies, showing: the explained variance of the three first EOFs; b-d) spatial patterns of the EOFs 1-3; 1-3) principal components of EOFs above.

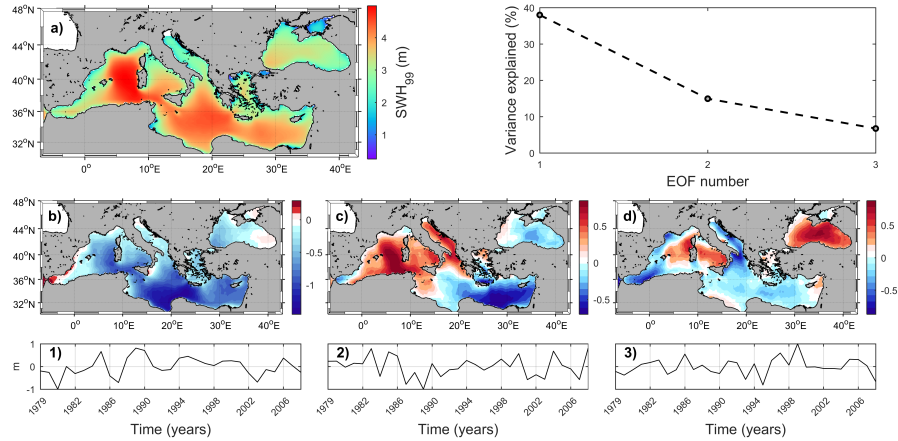


Figure 6. a) Mean field of winter SWH 99th percentile over the Mediterranean Sea. EOF analysis of SWH₉₉ anomalies, showing: the explained variance of the three first EOFs; b-d) spatial patterns of the EOFs 1-3; 1-3) principal components of EOFs above.

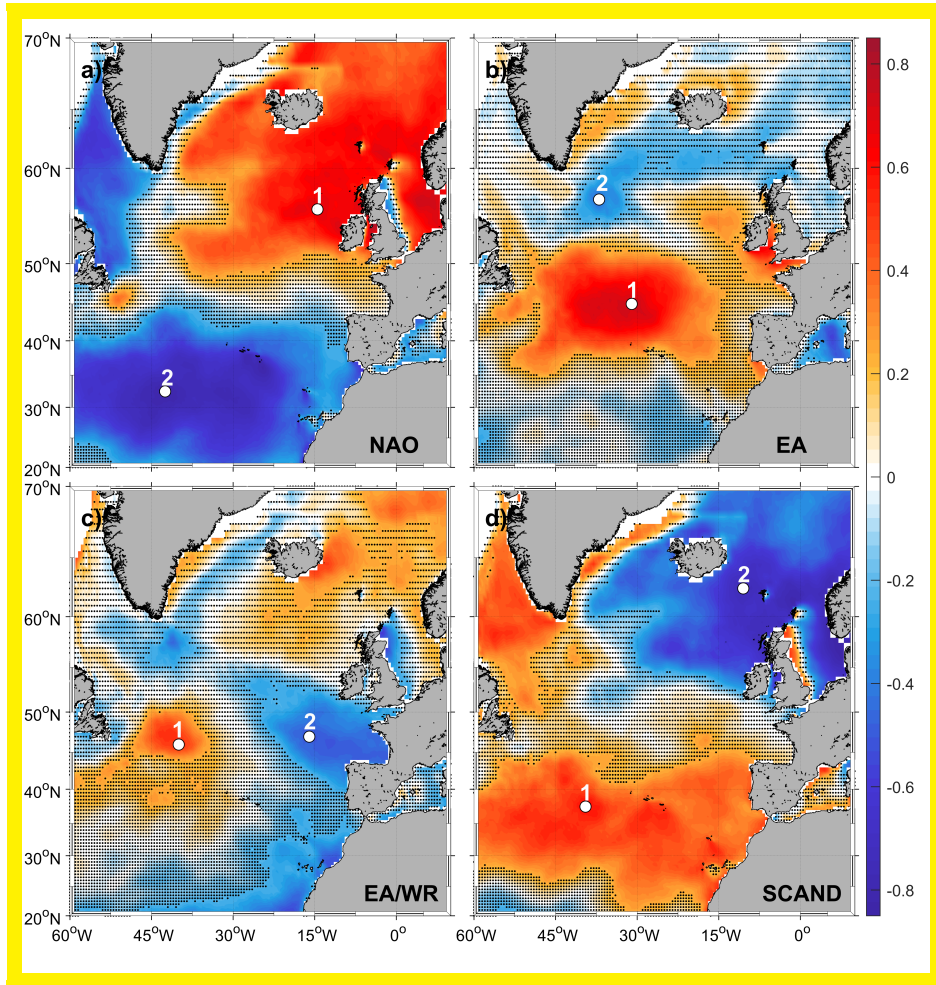


Figure 7. Pearson correlation coefficient of winter mean 99th percentile SWH North Atlantic series and a) NAO, b) EA, c) EA-WR and d) SCAND winter mean indices. No significant values at the 90% confidence interval are dotted. The white points show 1) the maximum positive and 2) the maximum negative value of correlation coefficient.

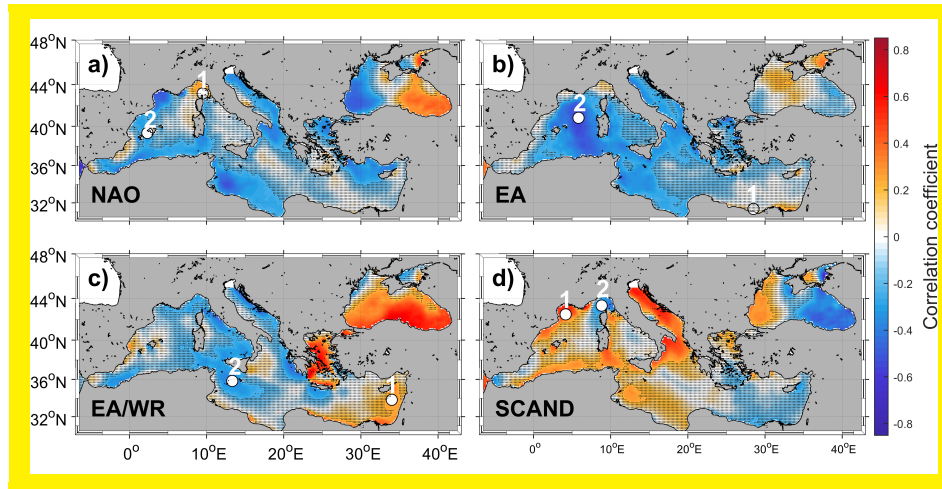


Figure 8. Pearson correlation coefficient of winter mean 99th percentile SWH Mediterranean Sea series and a) NAO, b) EA, c) EA-WR and d) SCAND winter mean indices. No significant values at the 90% confidence interval are dotted. The white points show 1) the maximum positive and 2) the maximum negative value of correlation coefficient.

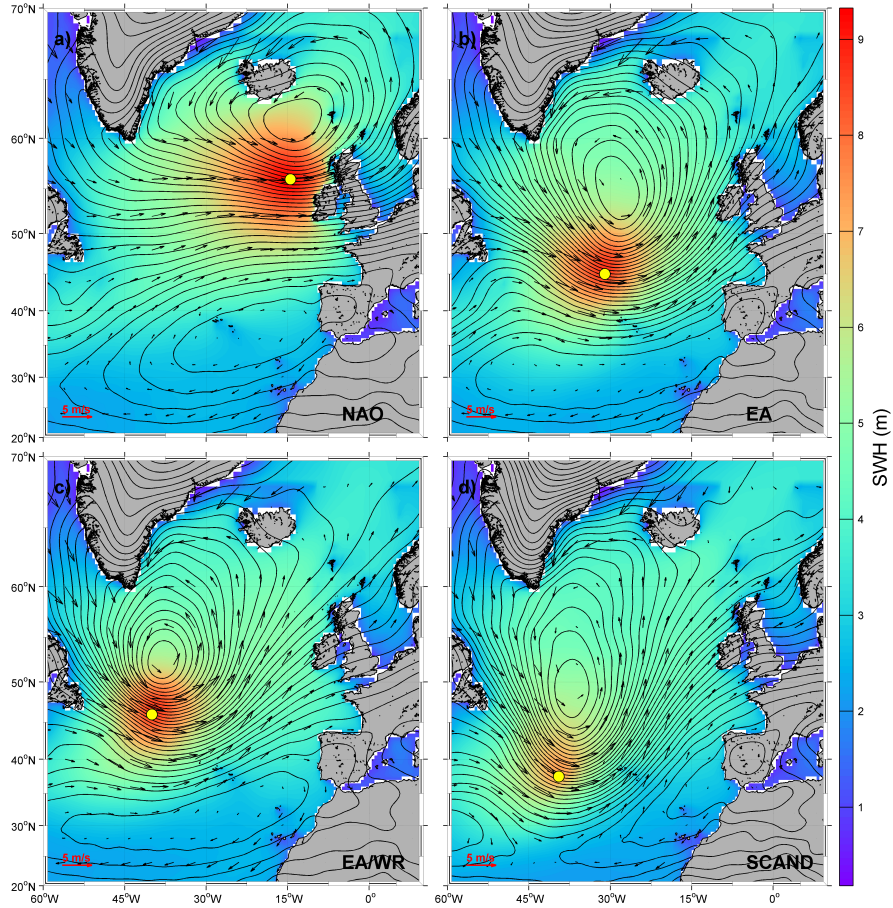


Figure 9. Winter atmospheric situations for the positive phase of a) NAO, b) EA, c) EA-WR and d) SCAND indices in the North Atlantic Ocean. The vectors represent the 10 m wind speed in m/s; the contours, the sea level pressure in Pa and the color range is the mean value of SWH in m. The red left bottom arrow represents the wind scale.

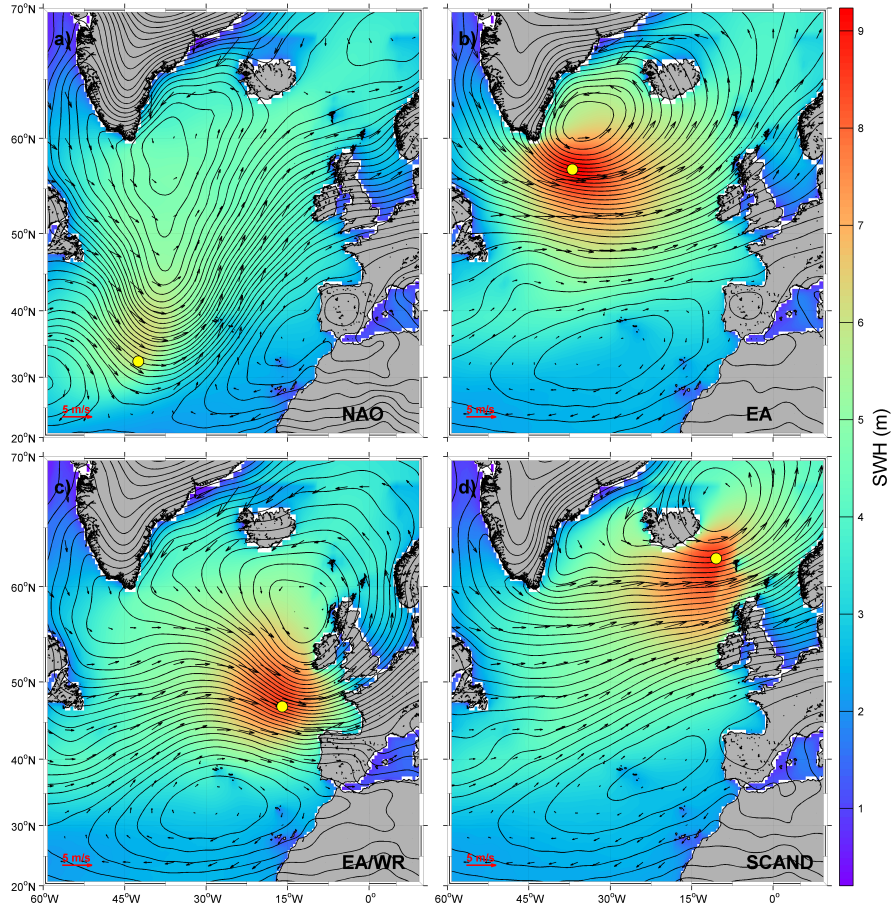


Figure 10. Winter atmospheric situations for the negative phase of a) NAO, b) EA, c) EA-WR and d) SCAND indices in the North Atlantic Ocean. The vectors represent the 10 m wind speed in m/s; the contours, the sea level pressure in Pa and the color range is the mean value of SWH in m. The red left bottom arrow represents the wind scale.

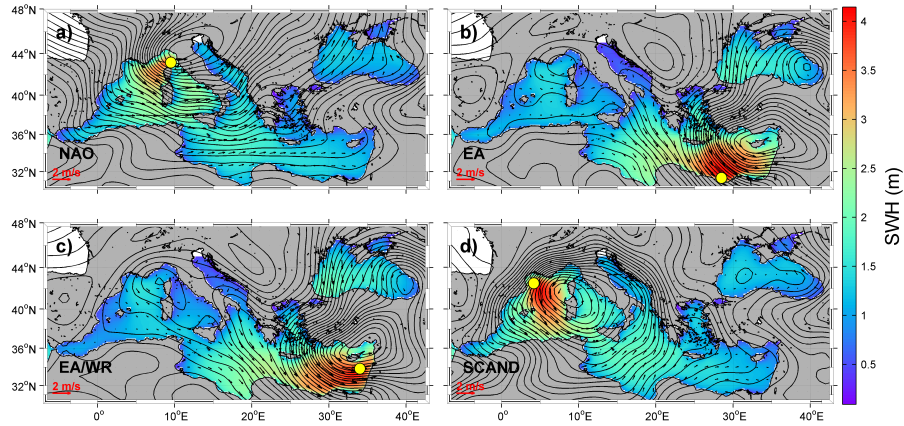


Figure 11. Winter atmospheric situations for the positive phase of a) NAO, b) EA, c) EA-WR and d) SCAND indices in the Mediterranean Sea. The vectors represent the 10 m wind speed in m/s; the contours, the sea level pressure in Pa and the color range is the mean value of SWH in m. The red left bottom arrow represents the wind scale.

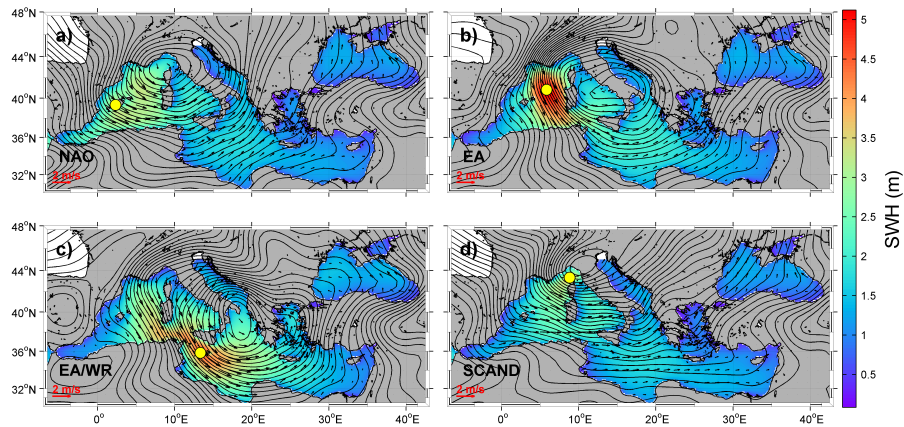


Figure 12. Winter atmospheric situations for the negative phase of a) NAO, b) EA, c) EA-WR and d) SCAND indices in the Mediterranean Sea. The vectors represent the 10 m wind speed in m/s; the contours, the sea level pressure in Pa and the color range is the mean value of SWH in m. The red left bottom arrow represents the wind scale.