



**University of Athens**

**School of Physics**

**Atmospheric Modeling and Weather Forecasting Group**

**<http://forecast.uoa.gr>**

Evaluation of high resolution wave simulations with  
SAR-observations and estimation of the wave power  
spatial and temporal distribution

**George Kallos**

With contribution from

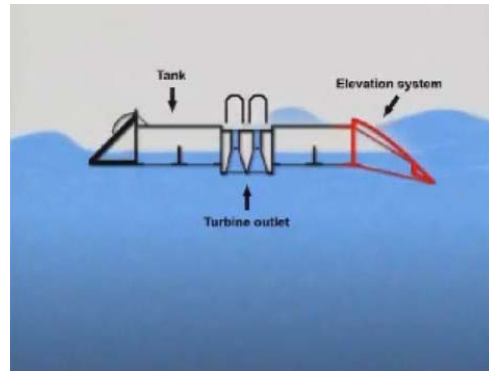
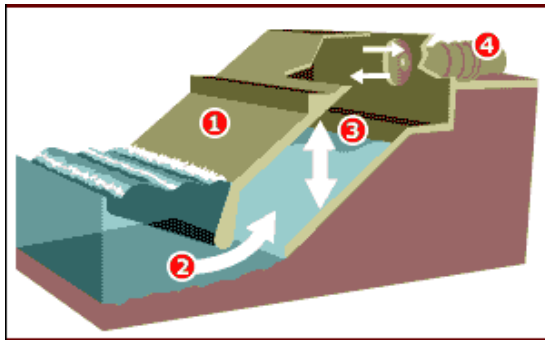
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# Renewable Energy Resources - Wave Energy

- Over the last years, the use of renewable resources for energy production is receiving increased attention as a result of the threat posed by climate change and the strict environmental policies regarding the production of greenhouse gases.
- Within this framework, wave energy (the energy that can be captured by sea waves) is a promising alternative energy resource with critical advantages:
  - Low variability (easier integration to the general grid)
  - High predictability
  - Good seasonal load for the most energetic seas (NW Europe)
  - It can be produced even in the case of low winds by exploiting the swell component of waves
  - Ocean energy technologies produce no emissions of harmful pollutants or greenhouse gases

# What kind of information the energy production industry needs ?



Wave Energy is dependent on the joint distribution of Significant Wave Height ( $H_s$ ) and Wave Energy Period ( $T_e$ ):

$$P_w = \frac{\rho g^2}{64\pi} H_s^2 T_e \quad [W/m]$$

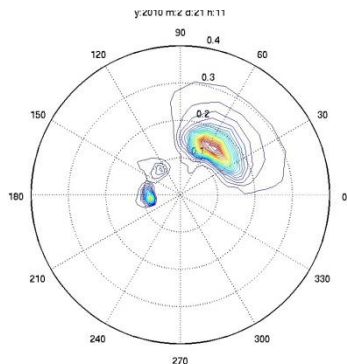
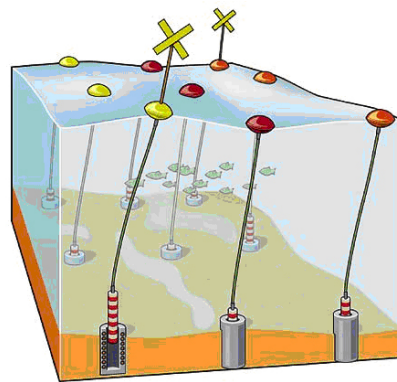
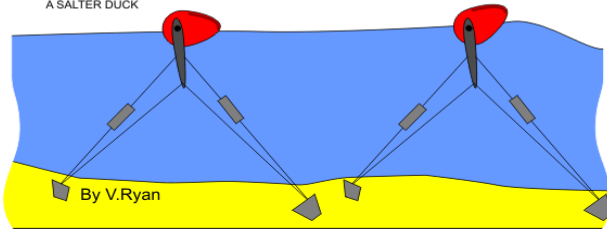
Different types of WECs are suitable to different combinations of wave height and period.

The use of the full wave spectrum is critical for the estimation of the wave energy potential over deep and shallow water areas:



INTERNAL VIEW OF DEVICE SIMILAR TO A SALTER DUCK

The wave device opposite incorporates an electricity generating system based on a pendulum connected to a generator. As the Salter Duck 'bobs' up and down on the waves, the pendulum swings forwards and backwards generating electricity.



$$P_W = \rho g \int_0^{2\pi} \int_0^{\infty} c_g(f, h) E(f, \theta) df d\theta$$

where  $c_g$  is the group velocity:

$$c_g(f, h) = \begin{cases} \frac{g}{4\pi} f^{-1}, & \text{for deep water} \\ \sqrt{gh}, & \text{for shallow water} \end{cases}$$

# Wave Resource Assessment-Methodology

- Resource mapping
- Resource analysis identifying areas where “hot spots” of high intensity exist
  - Mean wave conditions ( $H_s$ ,  $T$ ,  $\theta$ ) – 10 year mean values
  - Spatial and temporal variability of wave parameters at different time scales
    - Inter-annual
    - Seasonal
    - Monthly
  - Statistical measures for the asymmetry and the impact of non frequent values of the analysis results
  - Multivariate distribution fitting
- Analysis taking into account constraints such as bathymetry, distance from shore, marine structures, local commercial activities, fisheries, military areas, ship routes, etc.

# Wave Resource Assessment-Numerical Models

- Difficulties in obtaining wide coverage (spatial and temporal) of **observed wave data** over sea areas.
- The main tool for accurate environmental predictions is today the use of **Numerical Weather Prediction** (NWP) models that simulate successfully the general weather conditions with average accuracy reaching 80-90%.
- Such models are able to provide accurate short or long term forecasts for environmental parameters that are crucial for wave energy estimation:
  - Wave Height and Direction
  - Swell Height and Direction
  - Wave Period
  - Extreme Values

# Wave Resource Assessment-Resource Mapping

The EU Projects

**Marina Platform** (<http://www.marina-platform.info/>) and

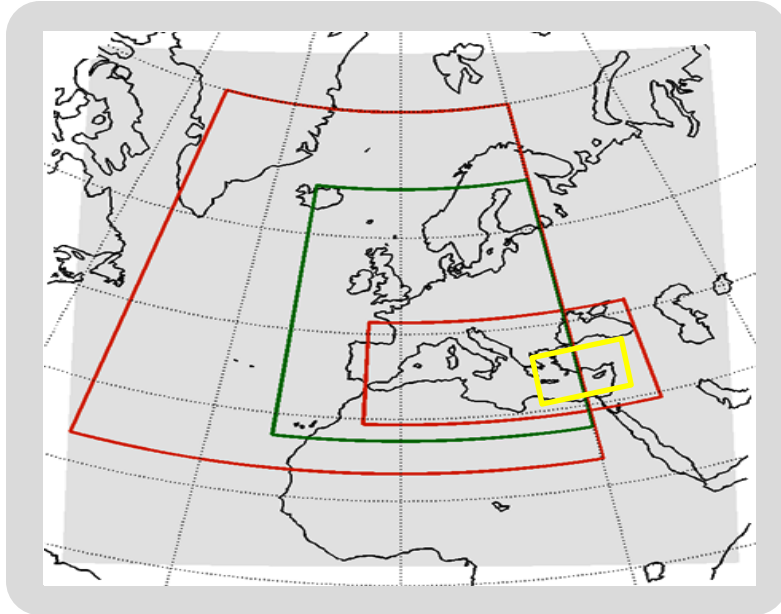
**E-Wave** (<http://www.oceanography.ucy.ac.cy/ewave/>)

focus on the monitoring and exploitation of the wave energy potential over the Atlantic and Mediterranean coastline of Europe.

A 10-year (2001-2010) high spatial and temporal resolution reanalysis data-set was derived for wind, wave and tidal parameters providing a wealth of information for marine resource assessment for the entire coastline of Europe.



# The models used



- The wave model includes **data assimilation systems** that can utilize satellite altimetry data.
- A **new advection scheme** (Corner Transport Upstream) has been adopted providing a more uniform propagation in all directions
- The **maximum wave height** is estimated by means of the probability distribution of sea surface elevation

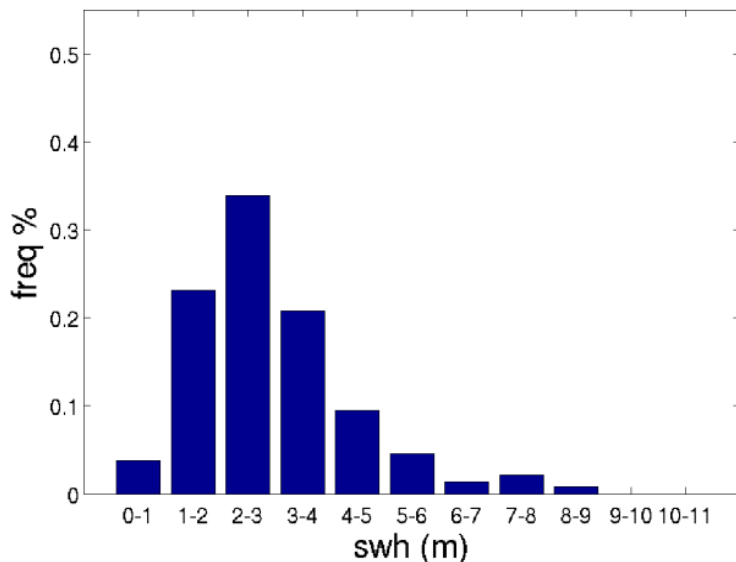
Atmospheric model Characteristics	<b>SKIRON</b>
Horizontal Resolution	0.05° x 0.05°
Initial and Boundary Conditions	High Resolution Reanalysis (15x15 Km)
Vertical Levels	45 (up to 50hPa)
Output at:	10, 40, 80, 120, 180 m a.s.l
Timestep	15 sec
<b>Full set of meteorological variables - every 1h</b>	

Wave model Characteristics	<b>WAM ECMWF CY33R1 Marina Platform</b>	<b>WAM ECMWF CY33R1 E-WAVE</b>
Model's domain	20–75°N, 50°W–30°E	30–41°N, 15°–37°E
Horizontal Resolution	0.05° x 0.05° (1601 x 1101 grid Points)	1/60° x 1/60° (1.667 km approximately)
Frequencies	25 (min 0.055Hz)	25 0.0417-0.54764Hz logarithmically spaced
Directions	24 (equally spaced)	24 (equally spaced)
Timestep	45 sec	45 sec

# Seasonal evaluation against ASAR records

## UK Atlantic coastline

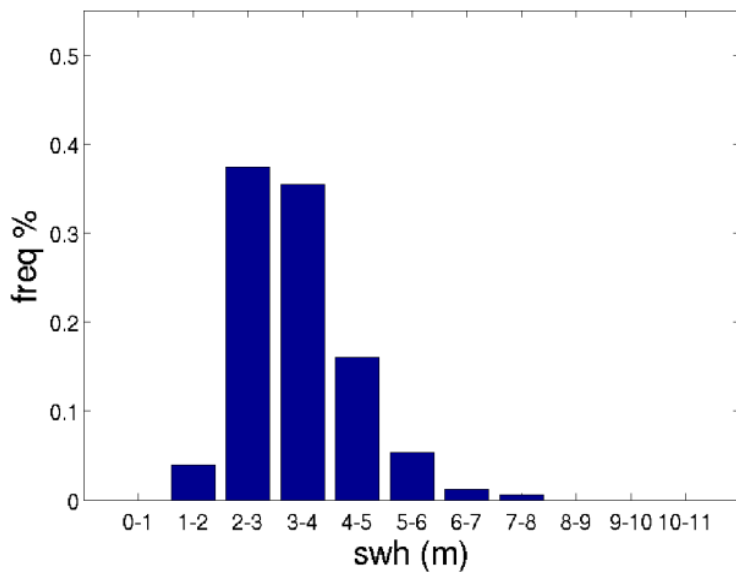
### WAM England 2010



Model results

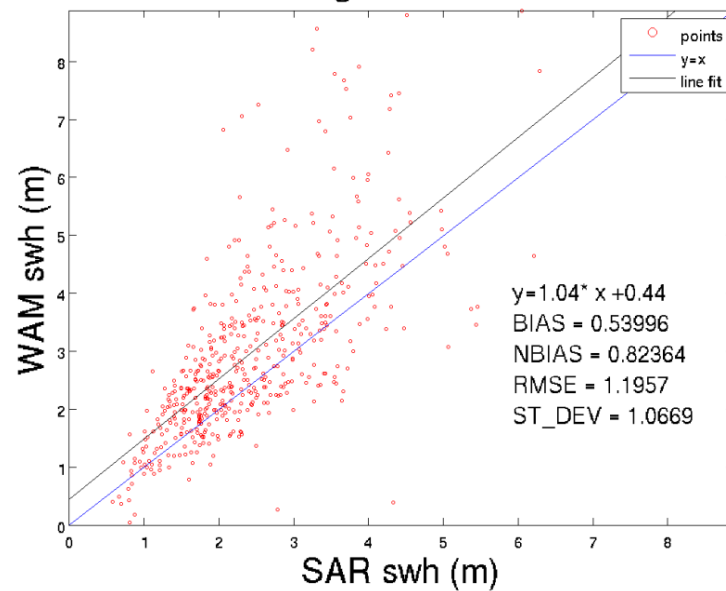


### Satellite England 2010



SAR records

### England 2010





# Seasonal evaluation against ASAR records

## UK Atlantic coastline

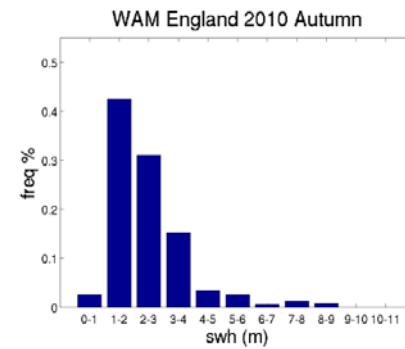
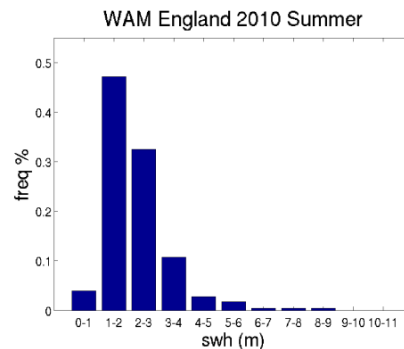
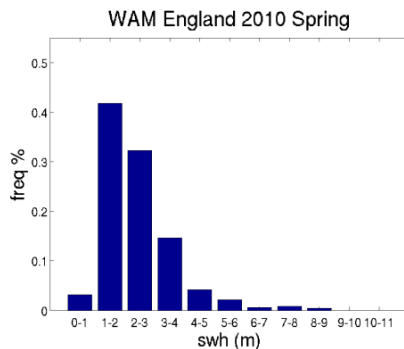
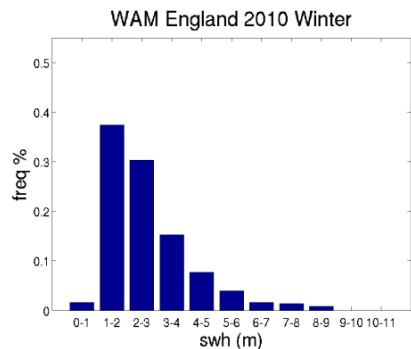
Dec – Feb

March – May

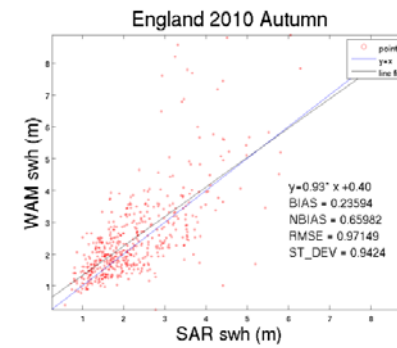
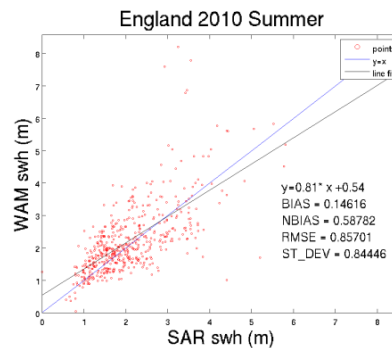
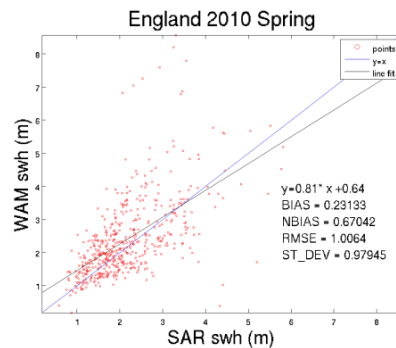
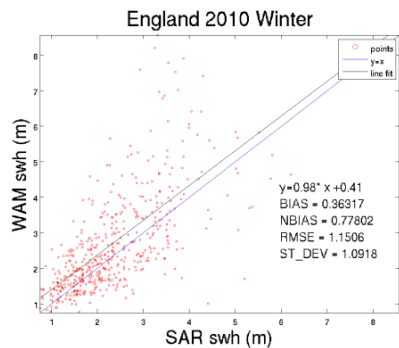
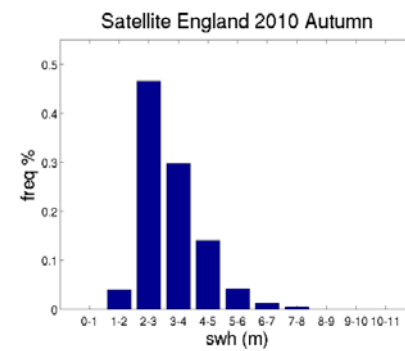
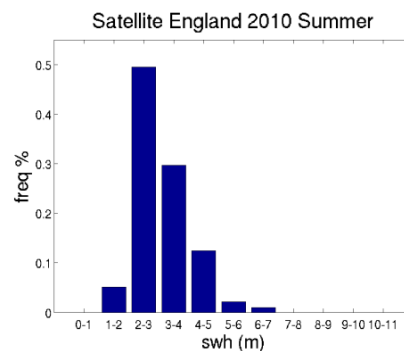
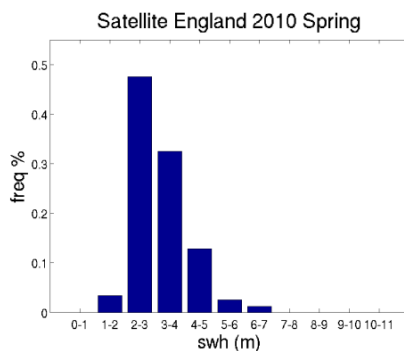
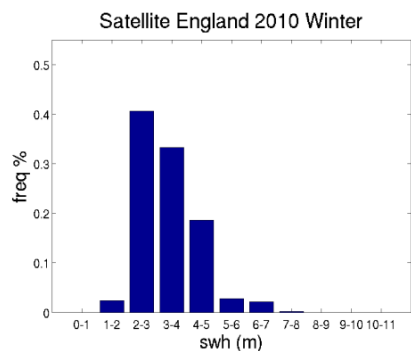
June – Aug

Sep - Nov

WAM



SAR

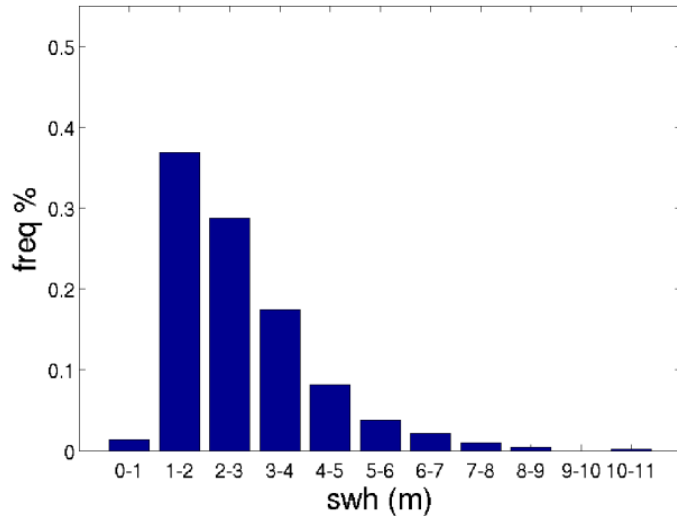


# WAM evaluation against ASAR records (annual)

## Spanish Atlantic coastline

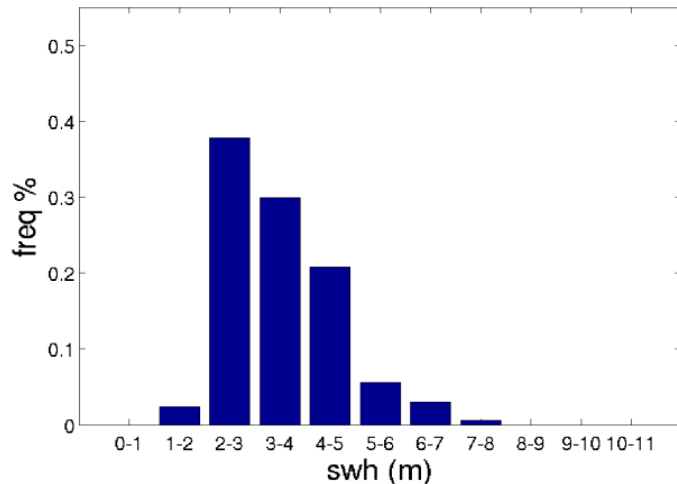
### Model results

WAM Spain 2010

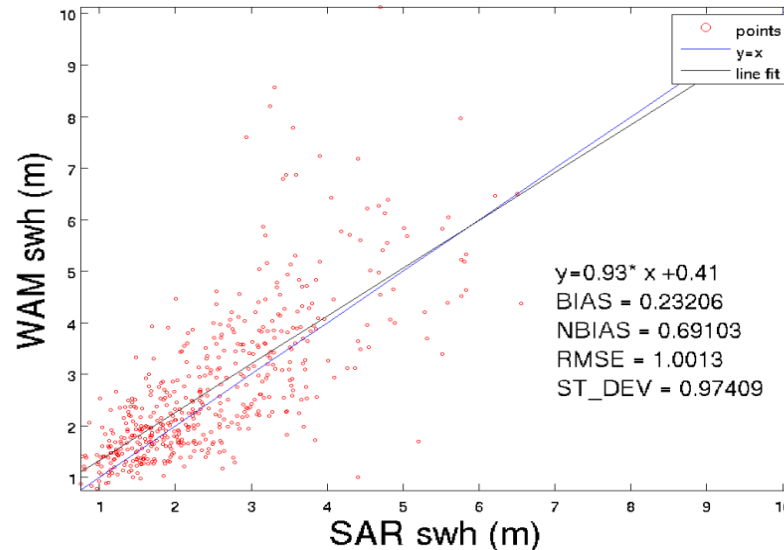


### SAR records

Satellite Spain 2010



Spain 2010



More uniform correlation between WAM modeled data and ASAR records is recorded over the Atlantic Spanish coastline

# Seasonal evaluation Spanish Atlantic coastline

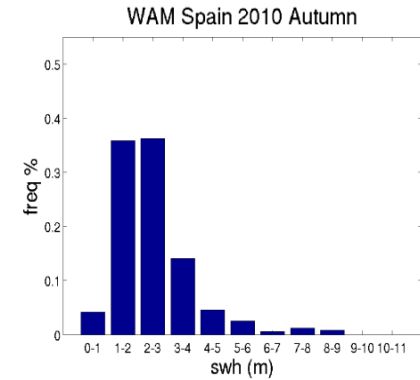
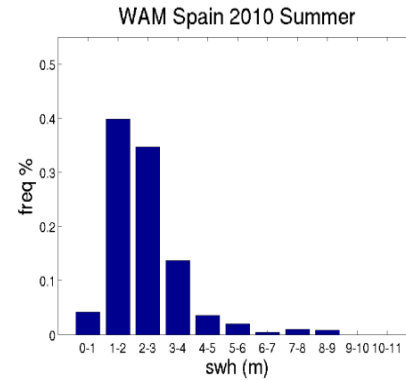
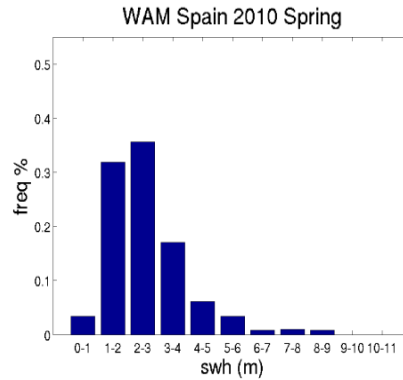
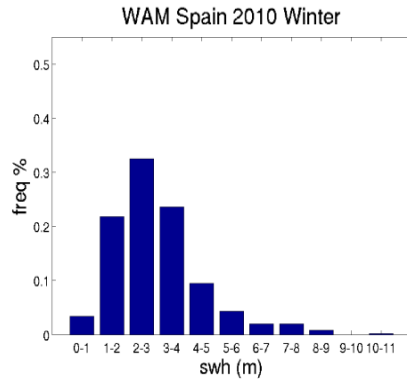
Dec – Feb

March – May

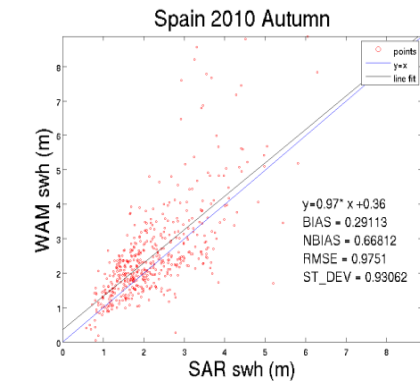
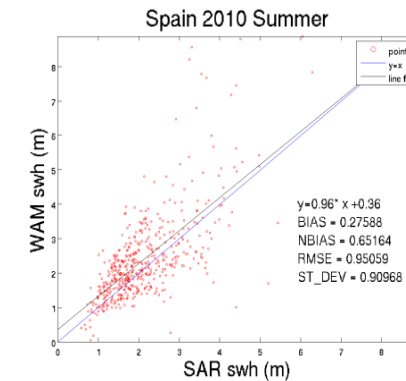
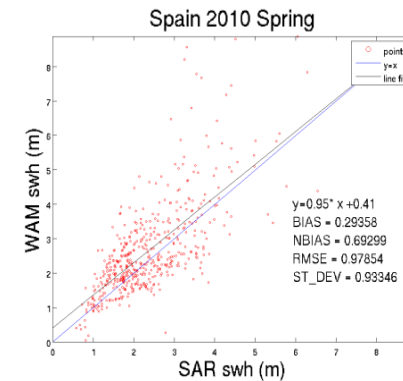
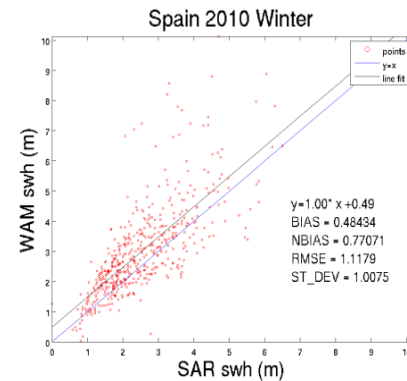
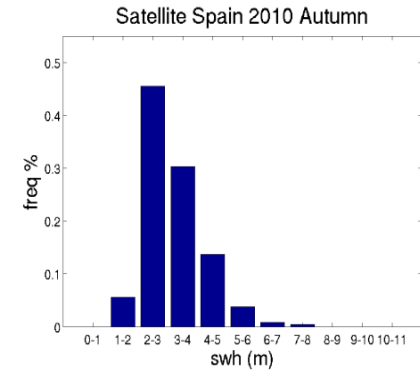
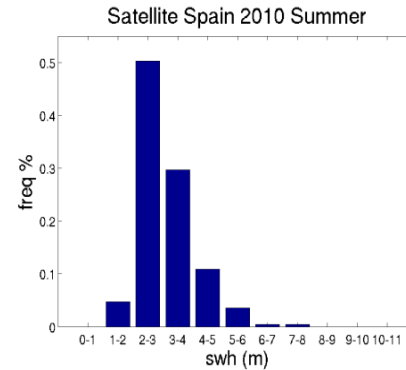
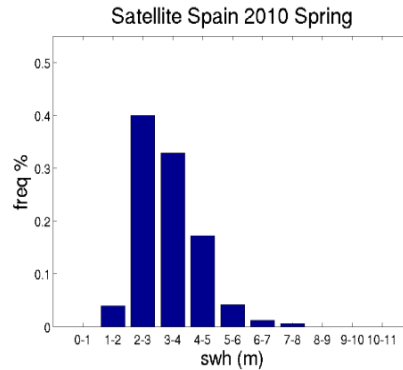
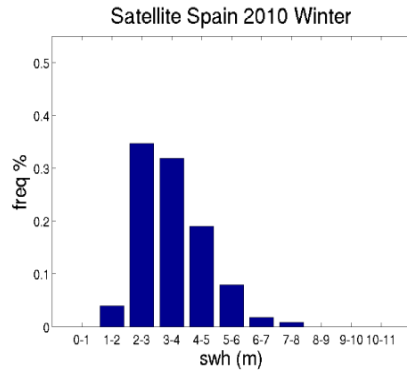
June – Aug

Sep - Nov

WAM

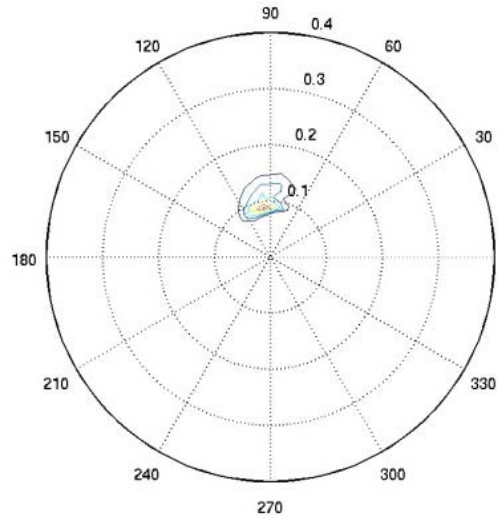


SAR

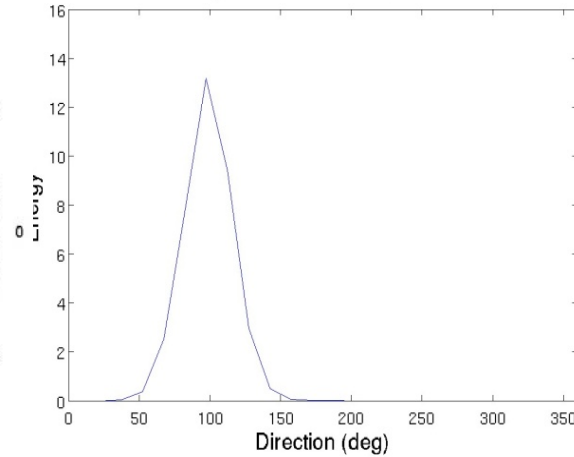


# Evaluation based on the full wave spectra

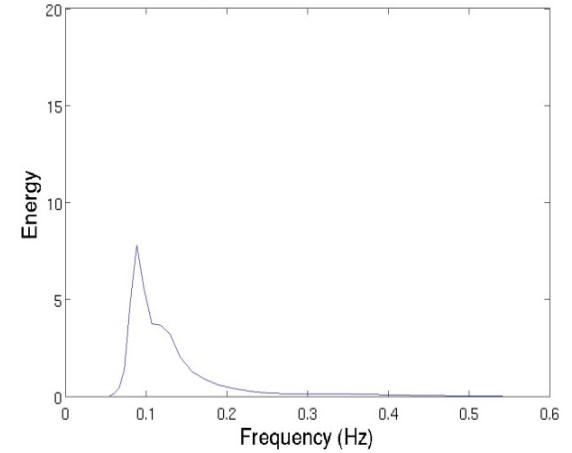
**WAM**



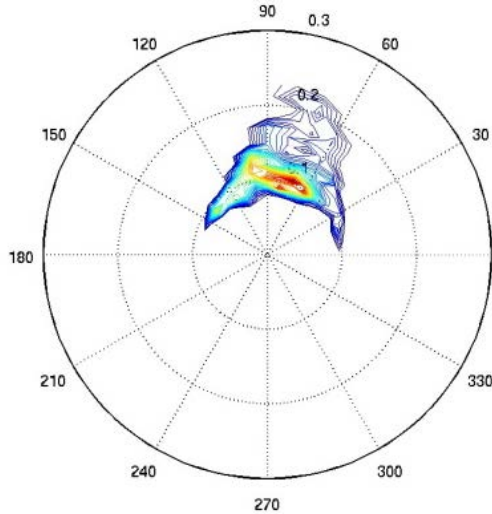
WAM Data: lat 53.05 lon -9.8  
y:2010 m:6 d:16 h:22



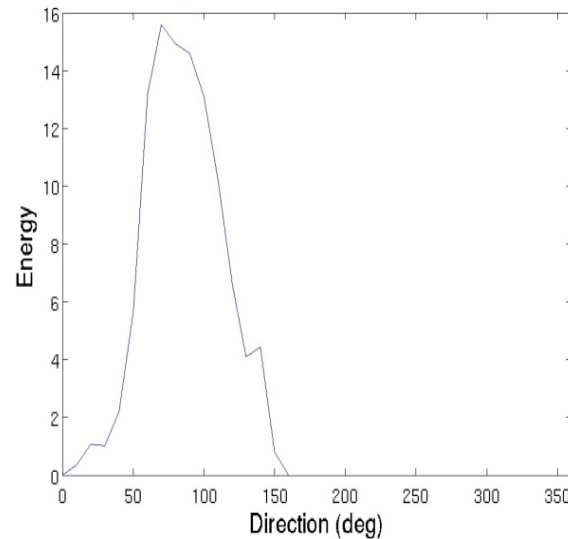
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y:2010 m:6 d:16 h:22



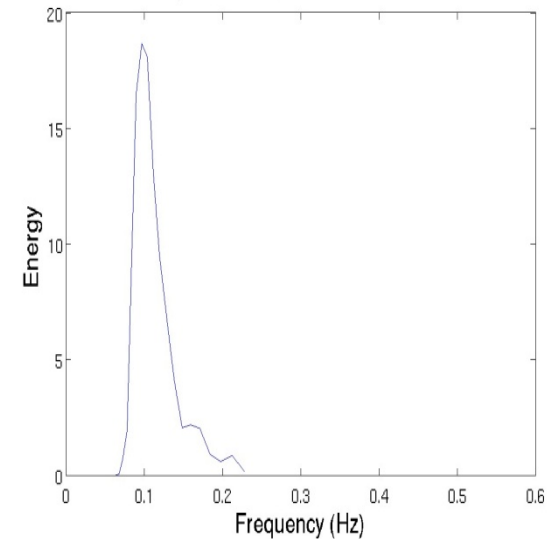
**SAR**



Satellite Data: lat 53.1359 lon -9.8654  
y:2010 m:6 d:16 h:22 f:11



Satellite Data: lat 53.1359 lon -9.8654  
y:2010 m:6 d:16 h:22 f:11



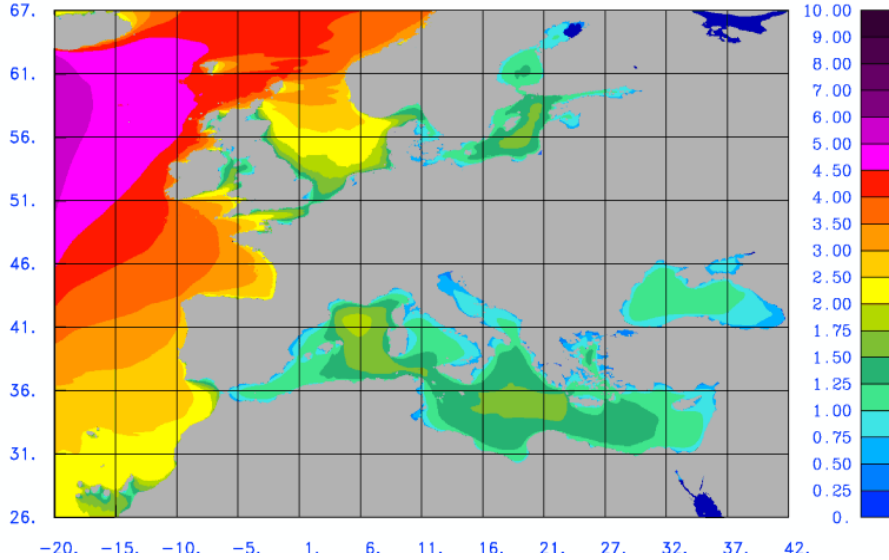
The frequency band of the SAR records is steeper than the corresponding model results while the directional distribution of modeled and recorded data are comparable

# Seasonal Distribution of Significant Wave Height

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DEC-JAN-FEB

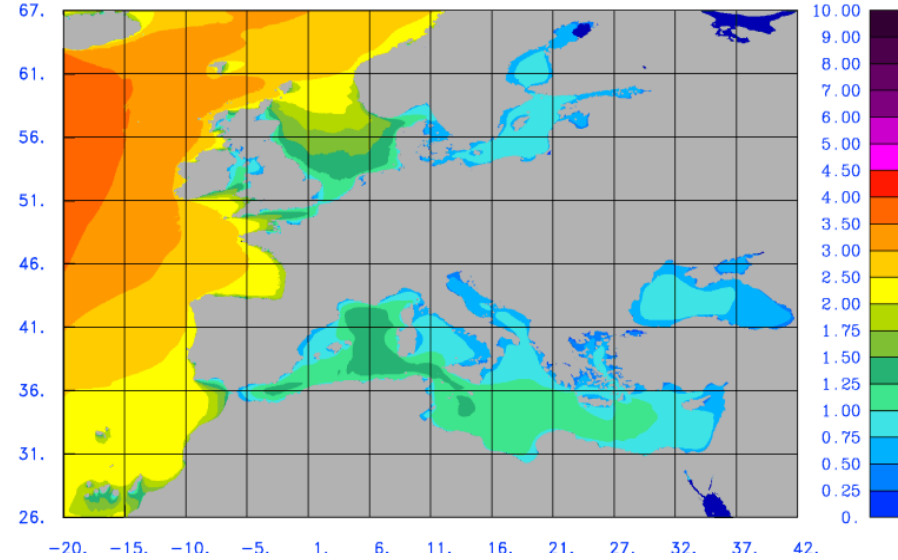
Significant wave height(m) mean value



University of Athens

MAR-APR-MAY

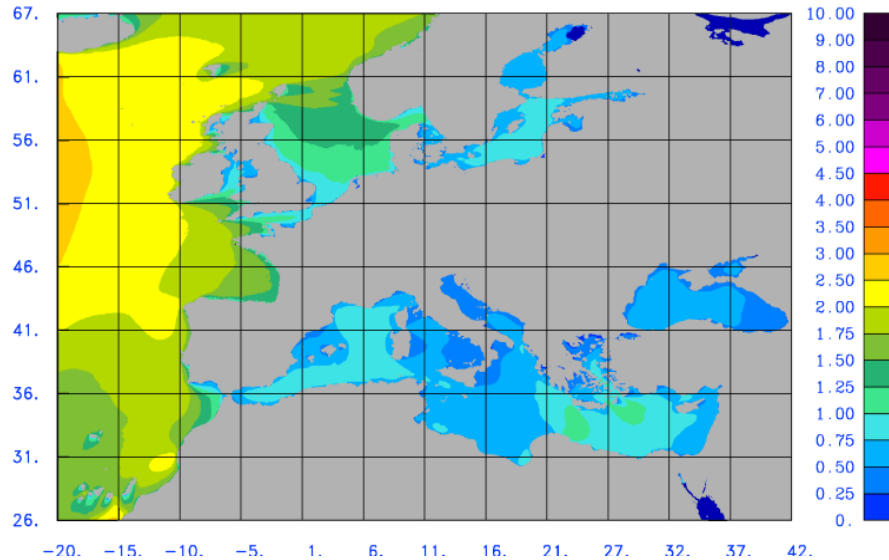
Significant wave height(m) mean value



University of Athens

JUN-JUL-AUG

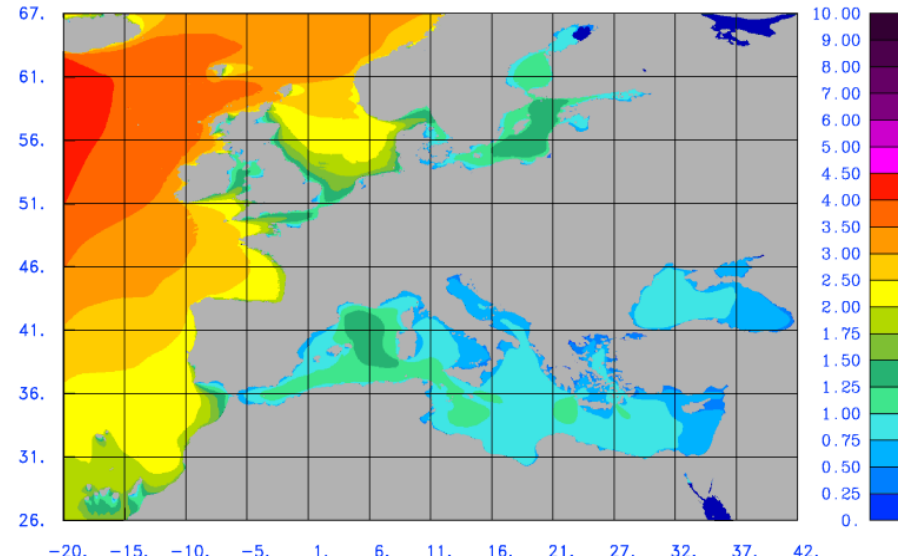
Significant wave height(m) mean value



University of Athens

SEP-OCT-NOV

Significant wave height(m) mean value

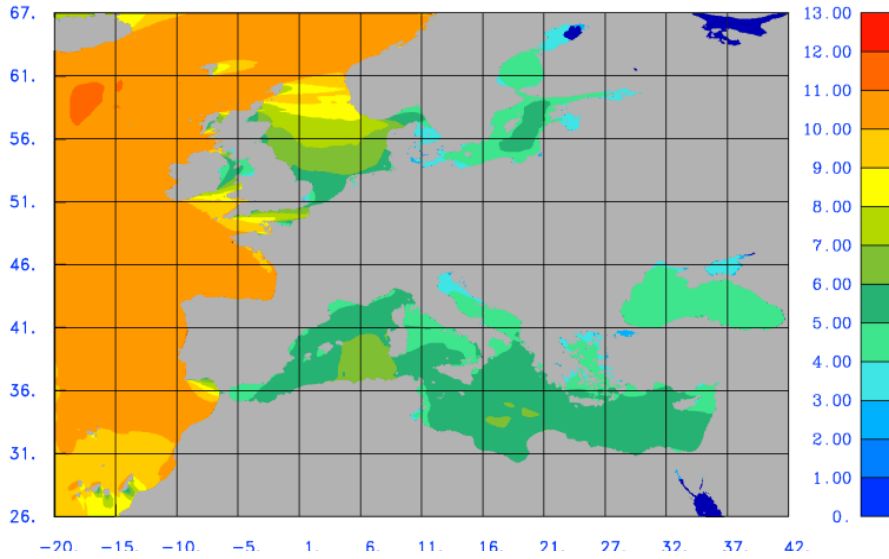


# Seasonal Distribution of Wave Period

University of Athens

DEC-JAN-FEB

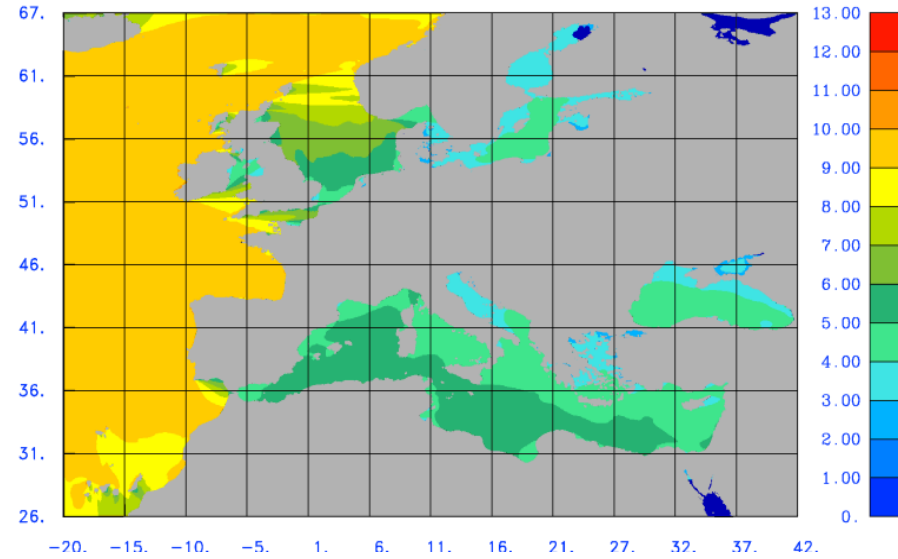
Mean Wave Period (sec)



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MAR-APR-MAY

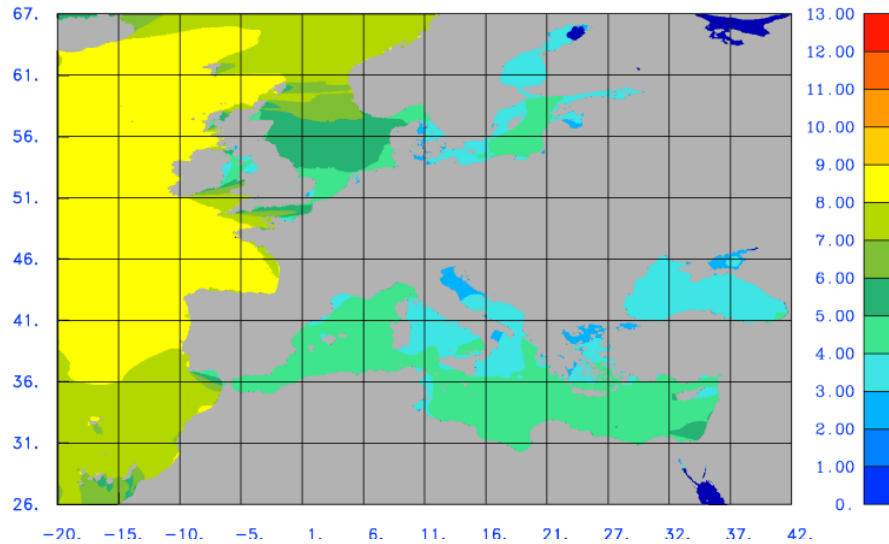
Mean Wave Period (sec)



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JUN-JUL-AUG

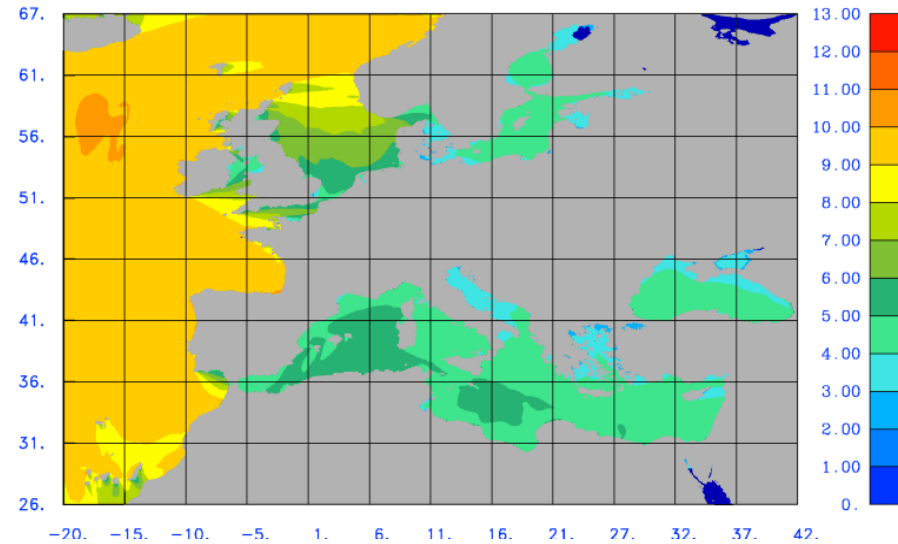
Mean Wave Period (sec)



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SEP-OCT-NOV

Mean Wave Period (sec)

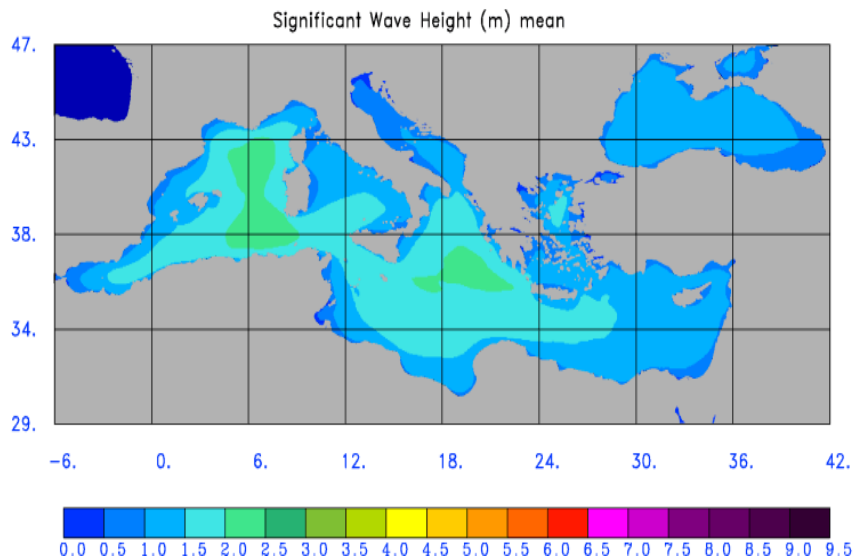


# Wave climatology: Is it enough for supporting efficiently the resource assessment?

Monthly variability of the Mean Hs in Med Sea

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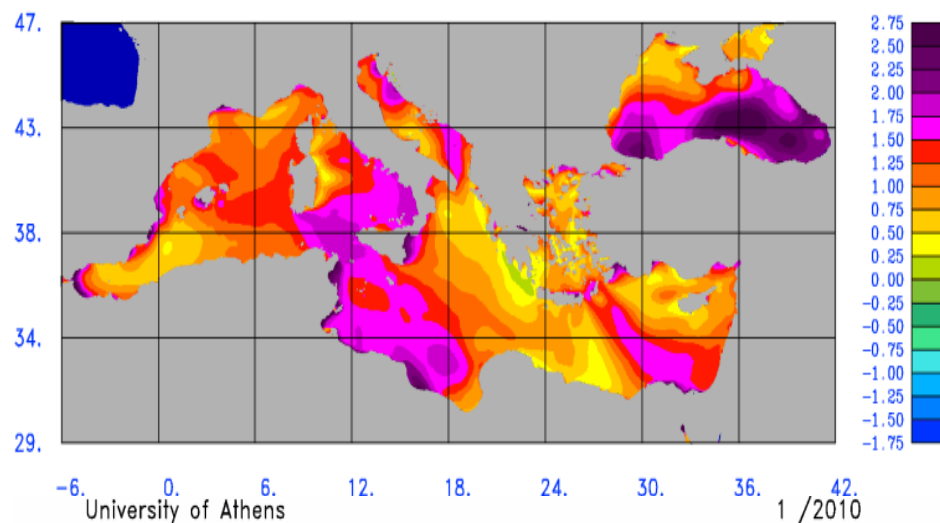
1 /2010



University of Athens

1 /2010

Skewness of Significant Wave Height

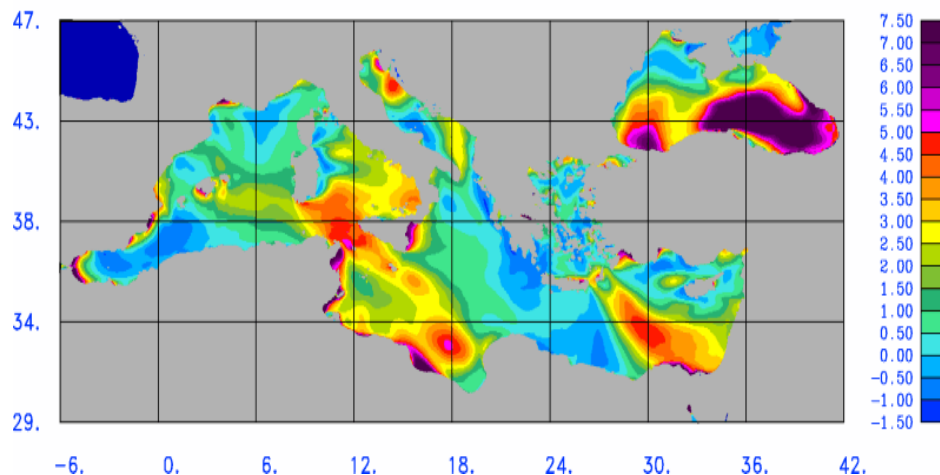


## Statistical measures for the asymmetry and the

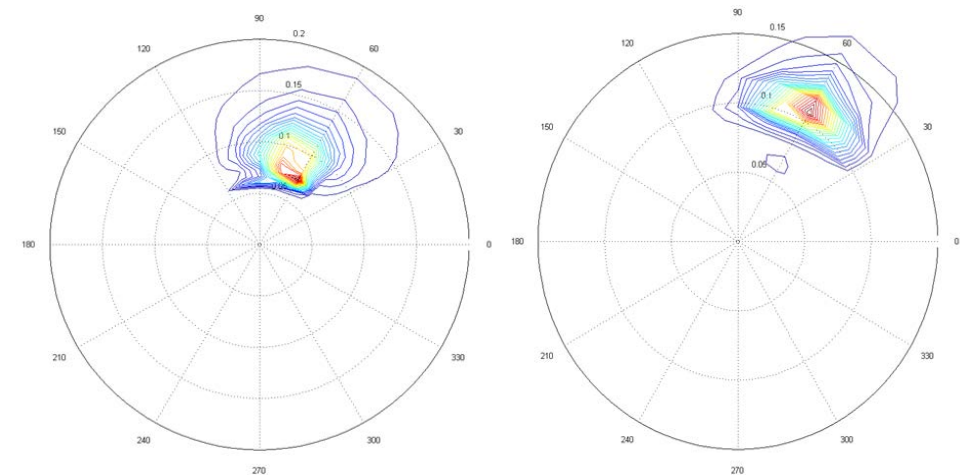
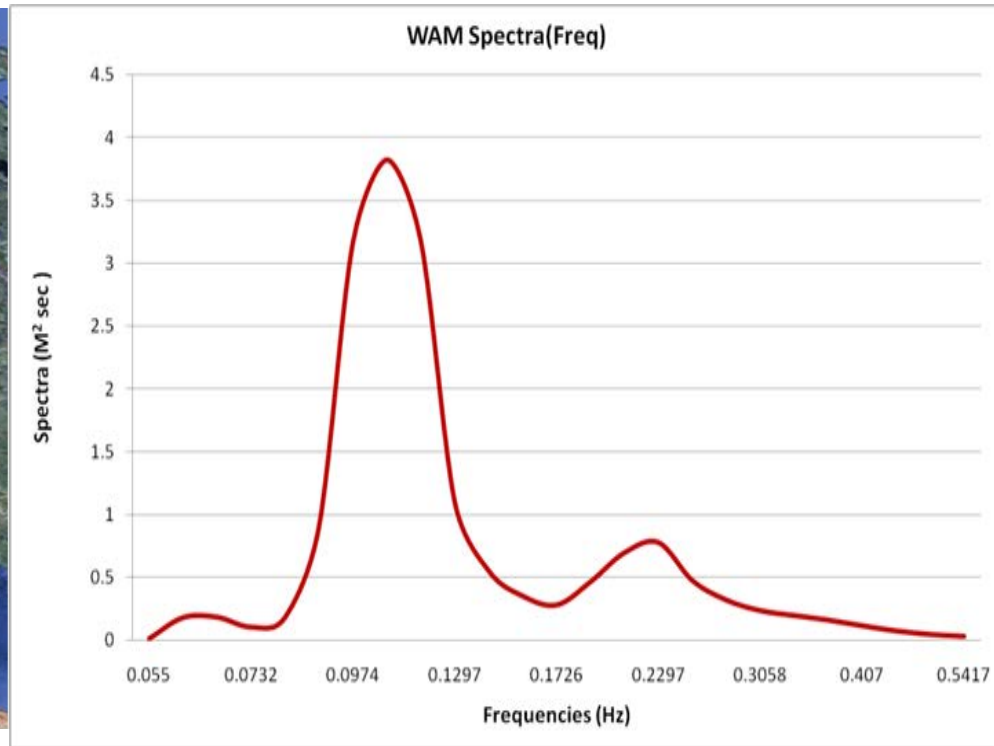
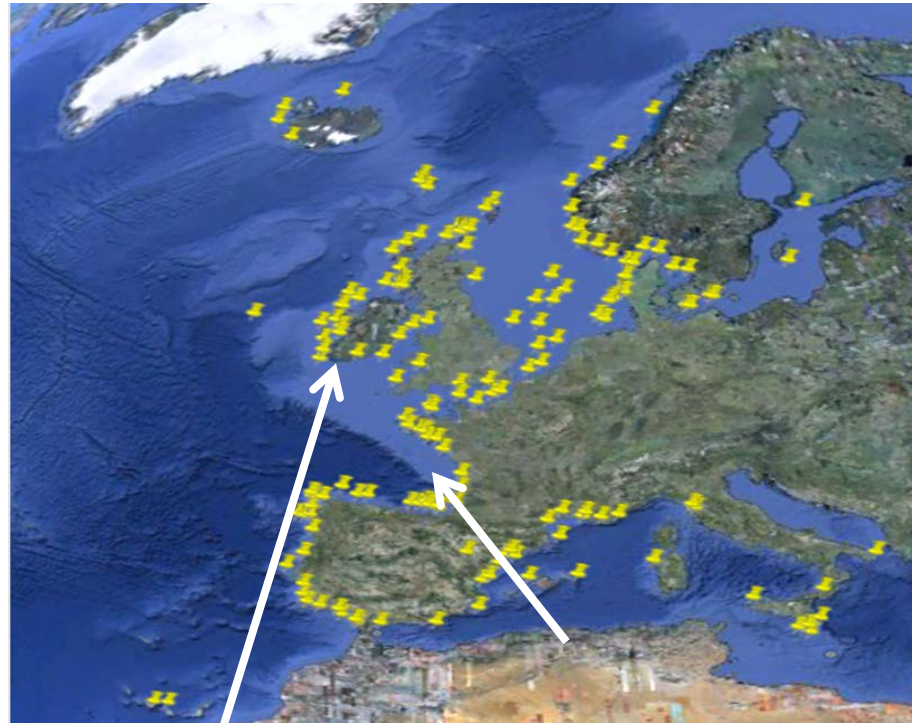
**kurtosis** of the data could be essential

- **Skewness** (3<sup>rd</sup> standardized moment) provides information for the tails of the distributions
- Areas with potential impact from extreme values can be spotted based on the **kurtosis** (4<sup>th</sup> standardized moment)

Kurtosis of Significant Wave Height



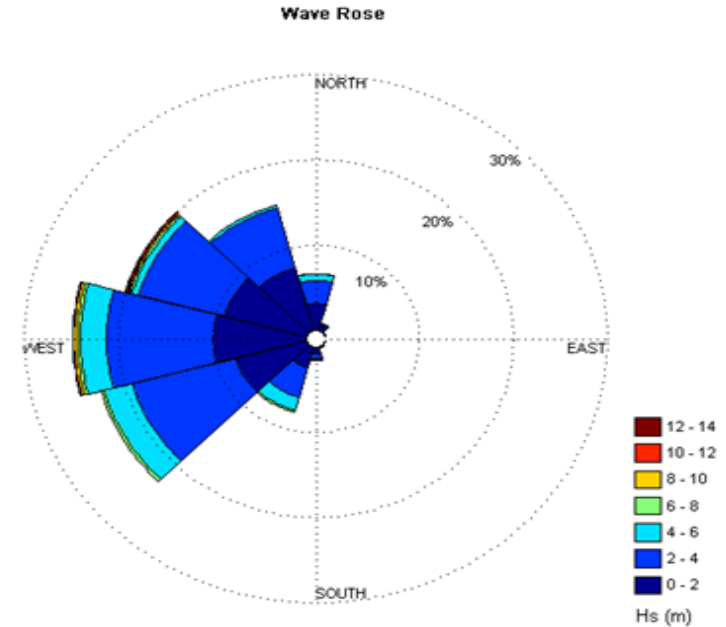
# On site statistical analysis



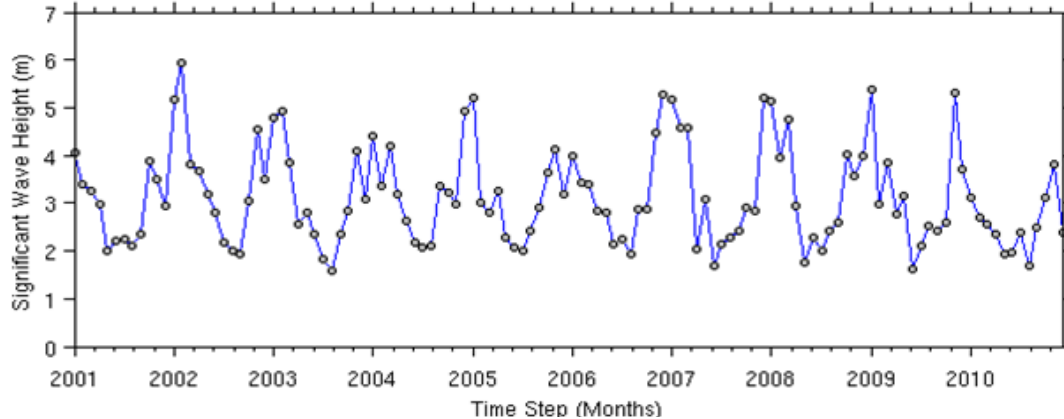
For a set of preselected points of interest along the coastline of Europe, the full wave spectrum derived without adopting any preselected standard forms, like e.g. JONSWAP, has been stored to provide the full package of information needed



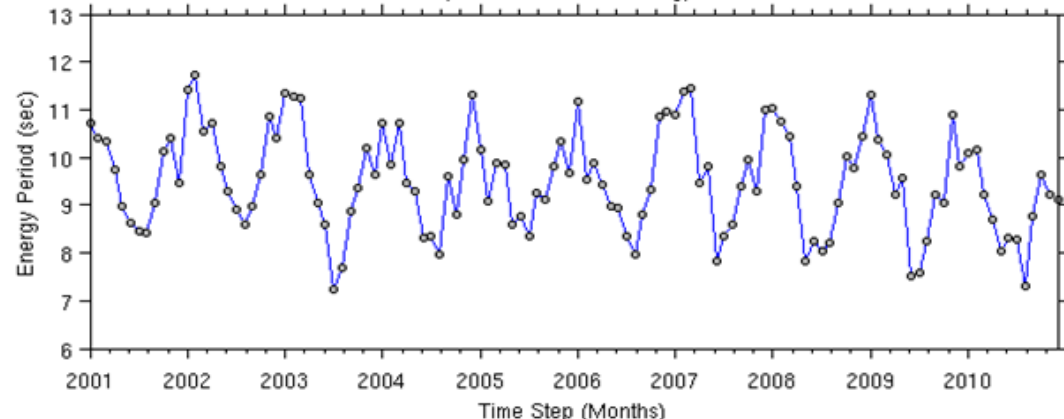
# On site statistical analysis



Monthly Mean Values of Significant Wave Height

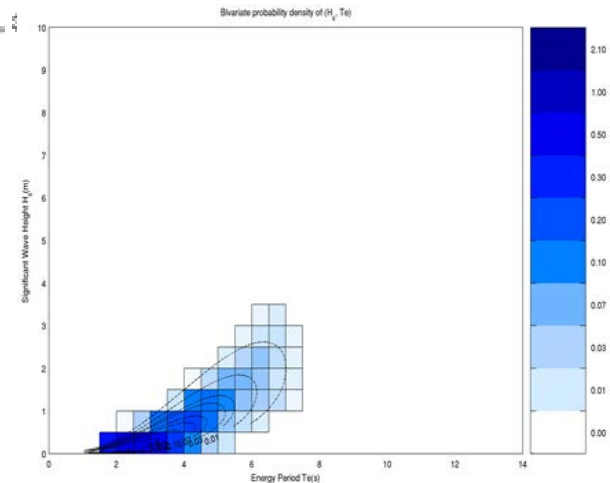
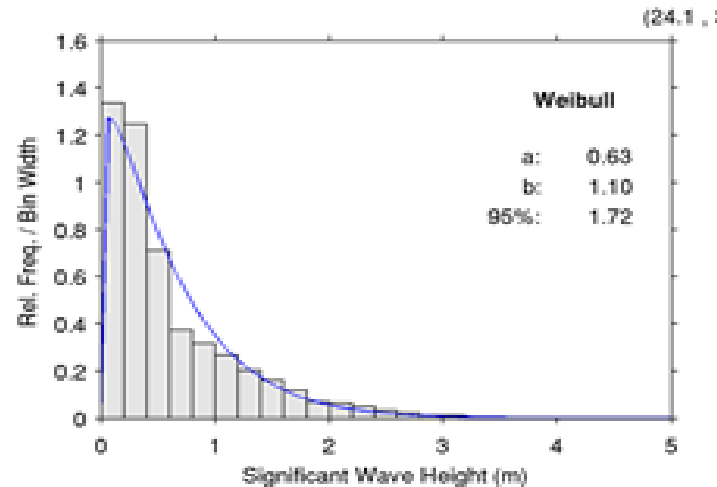
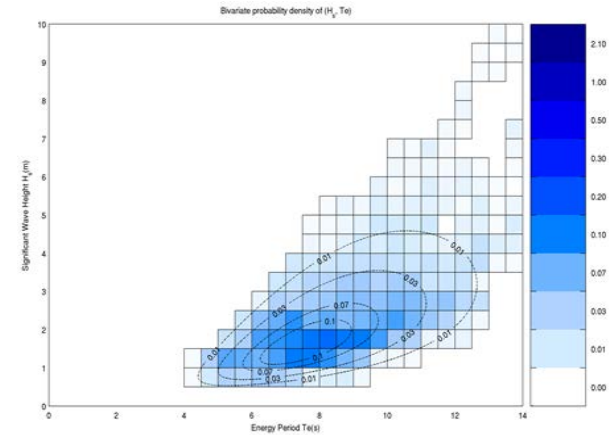
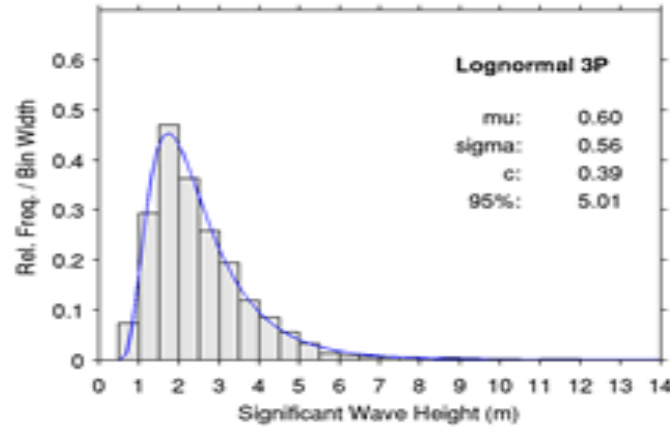


Monthly Mean Values of Energy Period



- The **time distribution** of crucial parameters over the whole 10-year study period may reveal trends and (seasonal or other) periodicities
- The **directionality** of the local wave parameters could characterize the wave climate of the area under study

# On site statistical analysis

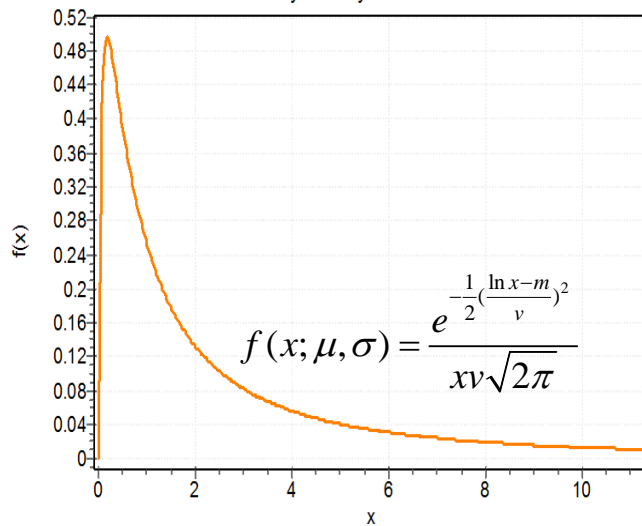


- The optimal distribution with its corresponding parameters are spatially dependent and not predefined/fixed by any way
- Weibull distribution could be a good choice for fitting wind speed and significant wave height values but Lognormal 3P provides an interesting alternative with even better convergence.
- Thresholds for extreme values are equivalently estimated by the two PDFs as the corresponding 95-percentiles.
- The joint  $H_s/T_e$  distribution is a statistical information of **primary importance** for wave resource characterization
- Different local wave climatology is depicted in the bivariate plots

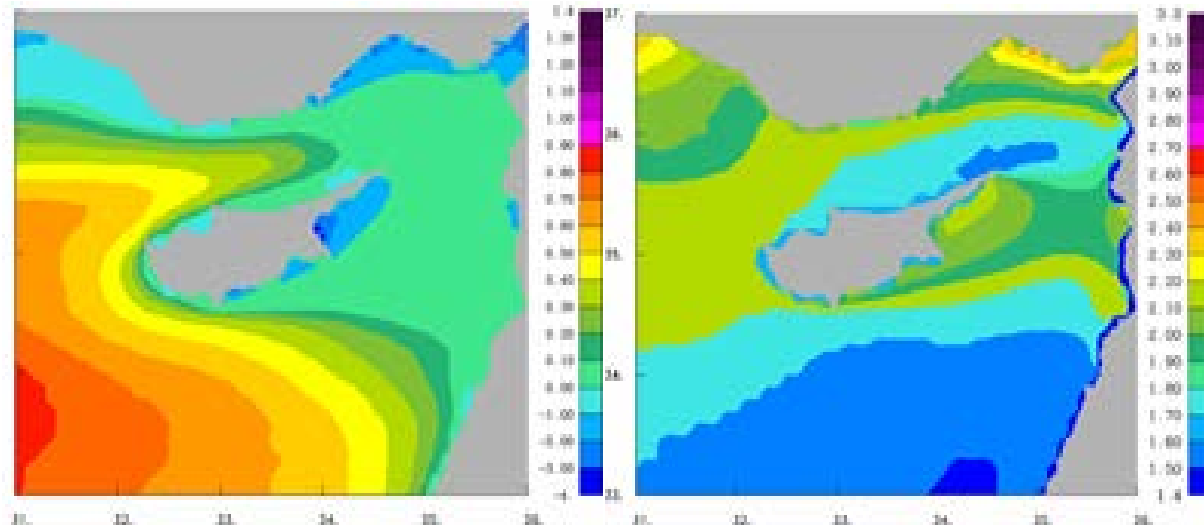
# Wave Power Potential distribution

- The wave energy potential can be also analyzed by a PDF fitting point of view.
- In the present work, a series of independent statistical tests proved that the **Lognormal distribution** optimally fits the modeled data.
- **Equally good fit** can be also succeeded by the **Generalized Extreme Value PDF**.
- **The corresponding parameters have a non trivial spatial distribution** and provide information of potential value for grid designers and researchers.

Probability Density Function



— Lognormal (1.5; 0.5)



Lognormal m-parameter

Lognormal v-parameter



# Some Concluding Remarks

- ❖ The estimation of the wave energy available potential is not as straight forward as in wind power case being directly dependent on two wave parameters ( $H_s$ - $T_e$ ).
- ❖ The lack of a dense observational network over sea areas poses further difficulties revealing the increased role that satellite data may have.
- ❖ Numerical wind/wave models, with optimization post processes, is considered as a good approach.
- ❖ The suitability of an area for wave energy exploitation cannot be based on a Yes/No answer.
- ❖ The local wave characteristics and the corresponding energy potential should be analyzed on different time scales and by employing statistical indexes measuring not only averages but also the variation, asymmetry and potential impact of extreme values as well as the 1 or 2-D optimally fitted distributions.
- ❖ The use of SAR information could be critical in energy estimation and monitoring since the wave directional spectrum gives full-package information avoiding averaging and/or smoothing over frequencies and directions.
- ❖ The specific characteristics of the technology that will be employed for the translation of the wave energy to power are crucial and should be taken into account.

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