2<sup>nd</sup> International Workshop on Waves, Storm Surges and

#### **Coastal Hazards**

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# Determination of Oceanic Extremes using a Spatial Ensemble of Satellite Data



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- 4 Conclusion

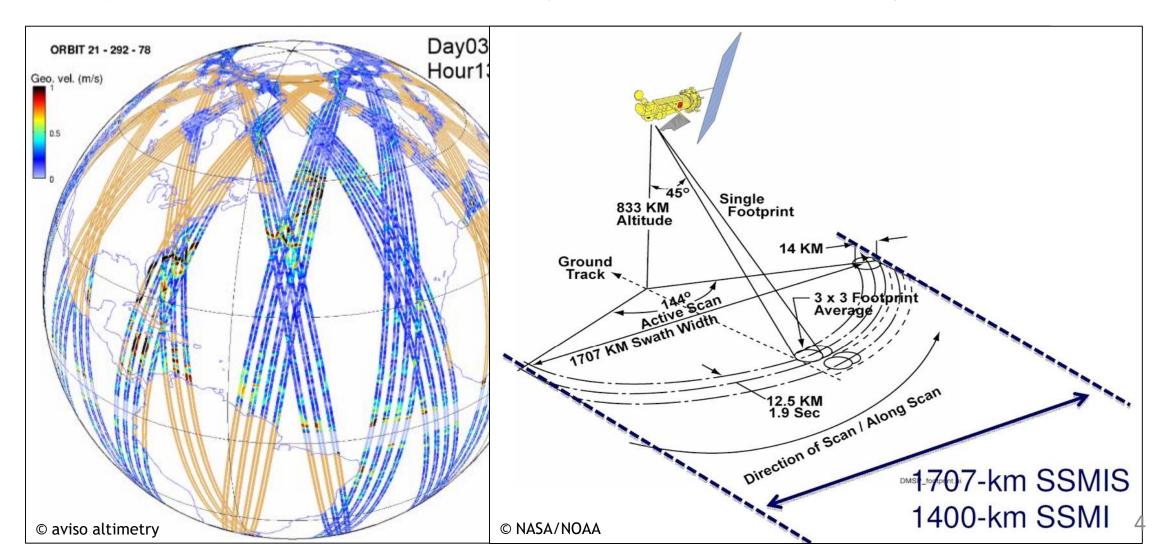
1 Global distribution of extreme wave height and wind speed from satellite data (Takbash et al. 2019)

• Satellite data (Young et al., 2017):

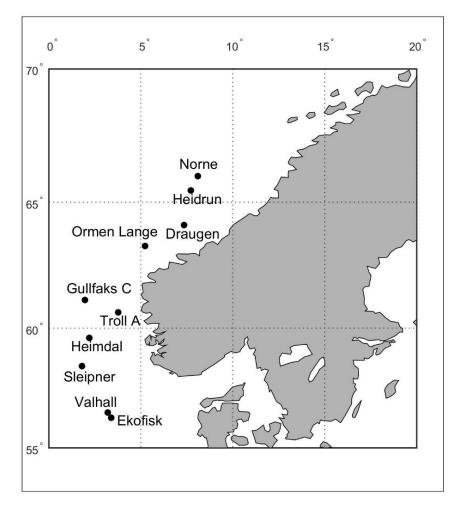
Altimeter/Radiometer  $\rightarrow$  Long duration records & global coverage

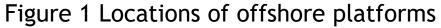
- $\circ~$  Two general approaches:
  - Initial Distribution Method (IDM) & Peaks over Threshold (PoT)
- $\circ\,$  Focus:
  - How well does the PDF fits the data?
  - How well is the tail of the PDF defined?

1 Global distribution of extreme wave height and wind speed from satellite data (Takbash et al. 2019)



#### Correction of the Radiometer data





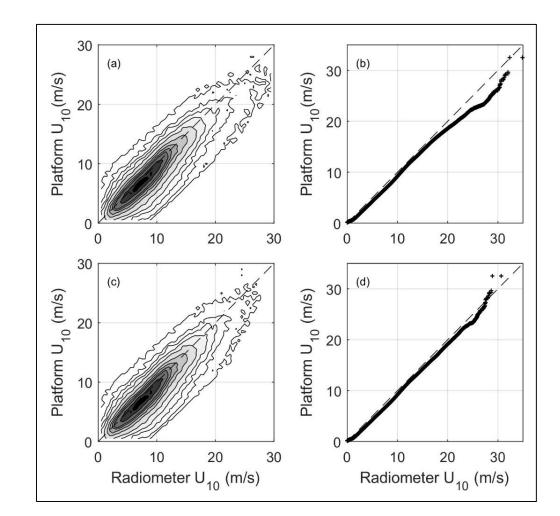
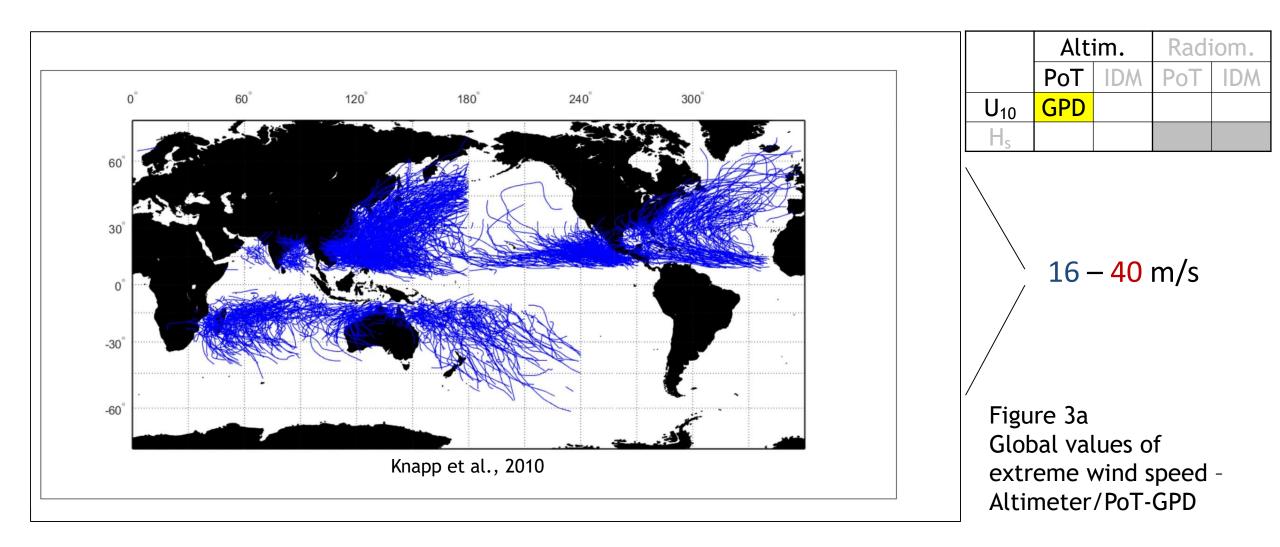
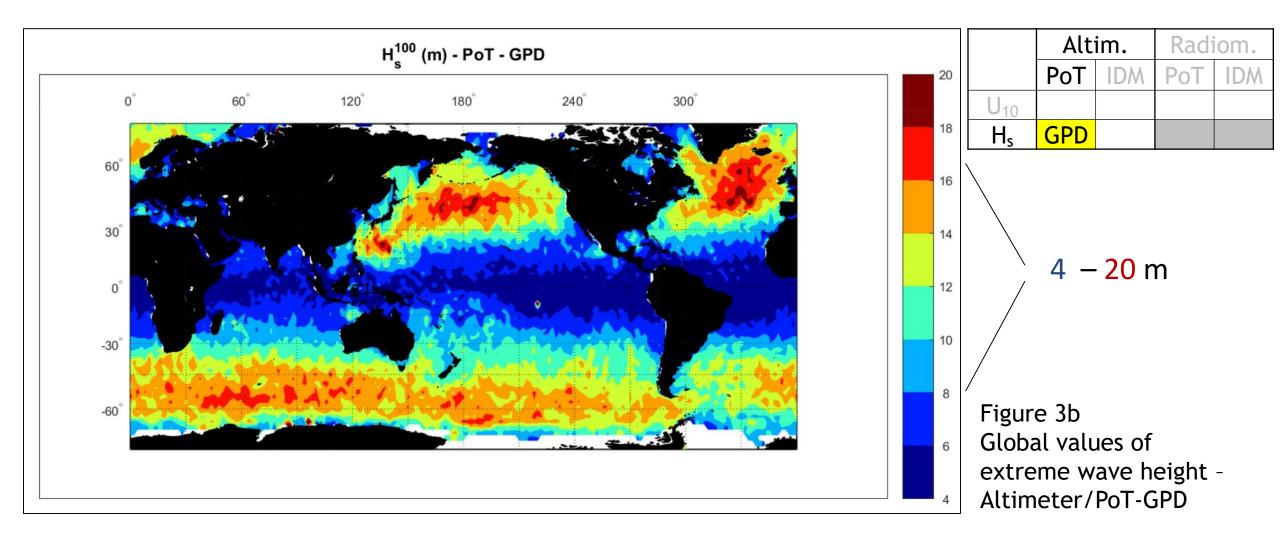


Figure 2 Radiometer-platform comparisons

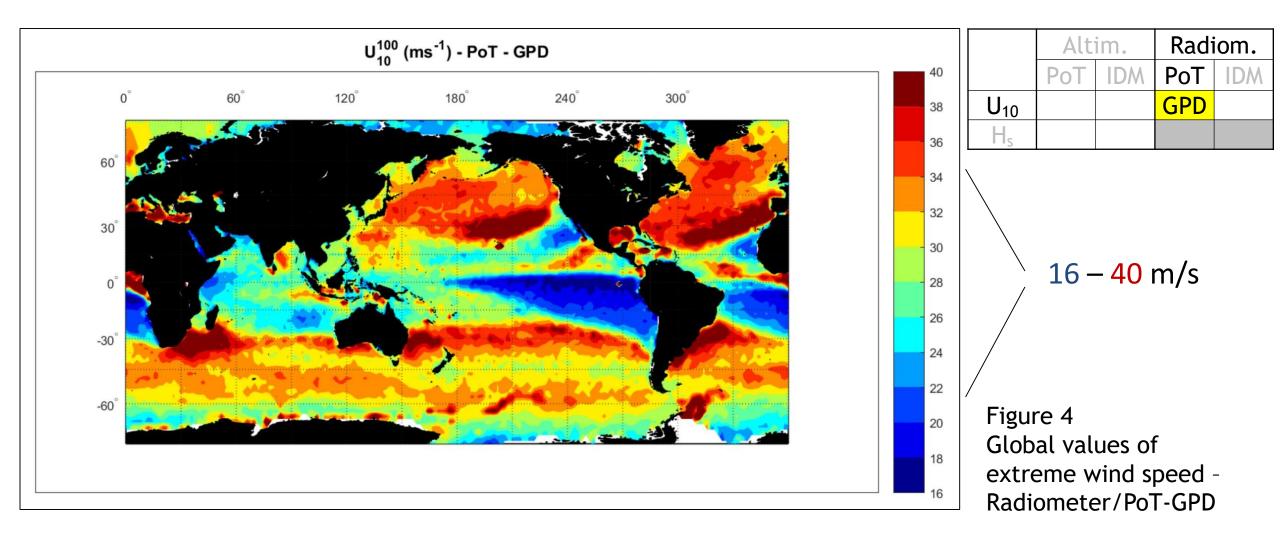
#### Global distribution of Extremes - Altimeter PoT



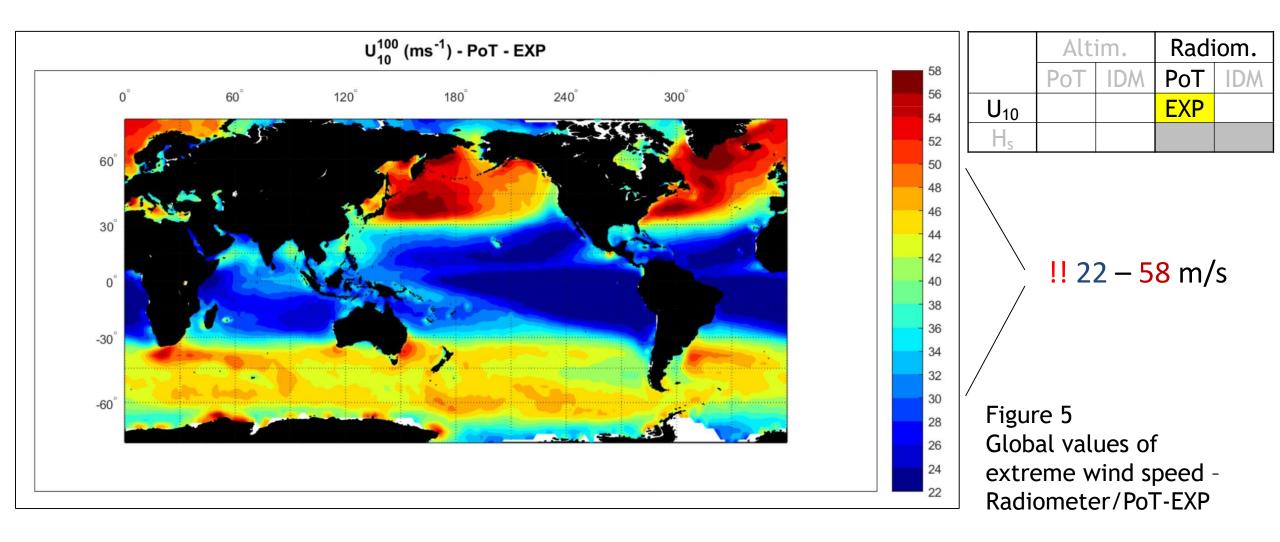
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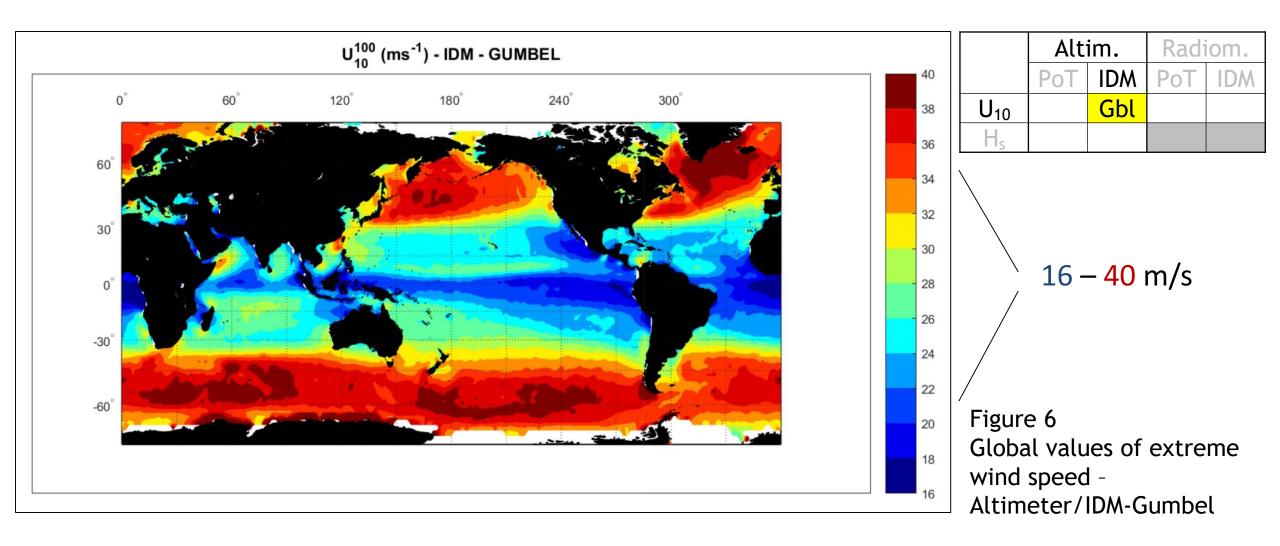
#### Global distribution of Extremes - Radiometer PoT



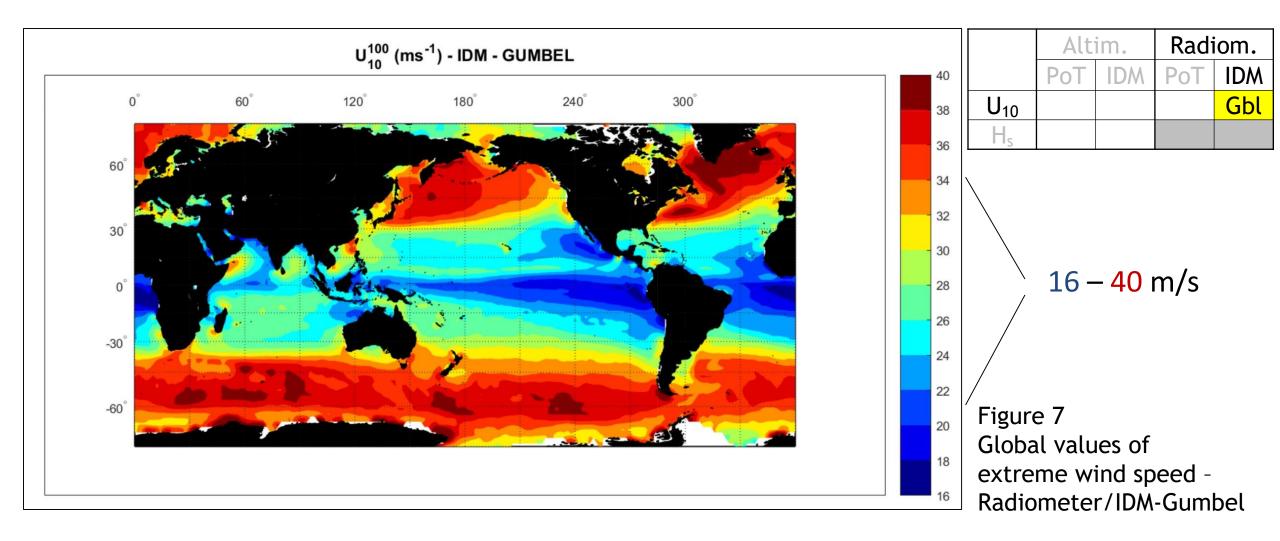
#### Global distribution of Extremes - Radiometer PoT



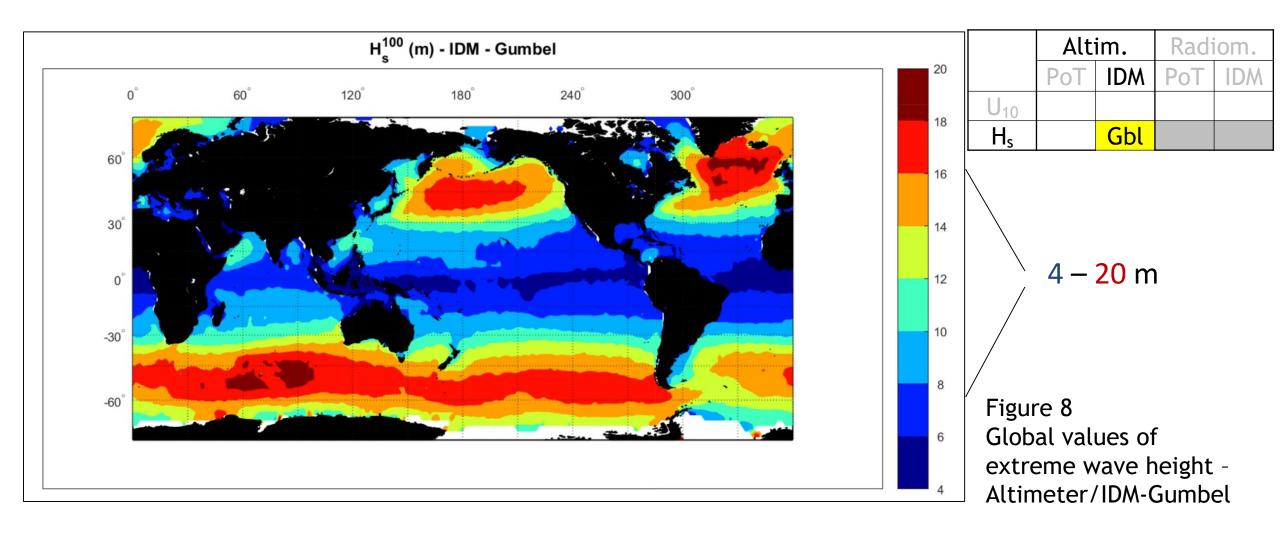
#### Global distribution of Extremes - Altimeter IDM



#### **Global distribution of Extremes - Radiometer IDM**



#### **Global distribution of Extremes - Altimeter IDM**



# 2 Conclusion

☐ The new Satellite data enables PoT analysis for the first time for

- Altimeter:
  - Values consistent with buoy and previous numerical model data
  - Much greater fine scale structure
- Radiometer:
  - Unacceptable "fair-weather" bias
  - $\rightarrow$  Unusable for PoT

□ IDM yield quite biased estimates of extremes and their spatial distribution

→ Comparing to PoT, little reason to use IDM in the future!

3 Global extreme wave height from spatial ensemble data (Takbash and Young 2019)

Aim:

reduce potential errors and **the size of confidence limits** on the resulting estimates of extremes when applying the **PoT approach** to **altimeter** data

- → Potential method: increase the amount of data points by combining data from adjacent regions
  - $\rightarrow$  spatial ensemble of aggregated (pooled) data
    - Equivalent to a longer duration time series!
    - Method has been tested by Breivik et al. (2013; 2014) in the time domain

3 Global extreme wave height from spatial ensemble data (Takbash and Young 2019)

Two criteria must be satisfied:

- 1. The areas must be far enough apart that the data points of these areas are **independently distributed** (i.e. uncorrelated)
- 2. The areas must still be representative of a similar wave climate by showing a **common parent distribution**

 $\rightarrow$  Assessment: ERA-Interim reanalysis data (continues in spatial/time domain)

3 Global extreme wave height from spatial ensemble data (Takbash and Young 2019)

To satisfy the **first criterion**:

Determine the **spatial coherence of wave height** on a global basis using the Pearson correlation coefficient

- Removed the **monthly variation** (monthly mean)
  - Deseasonalised time series
  - Irregular variations (storms and long-term trend)

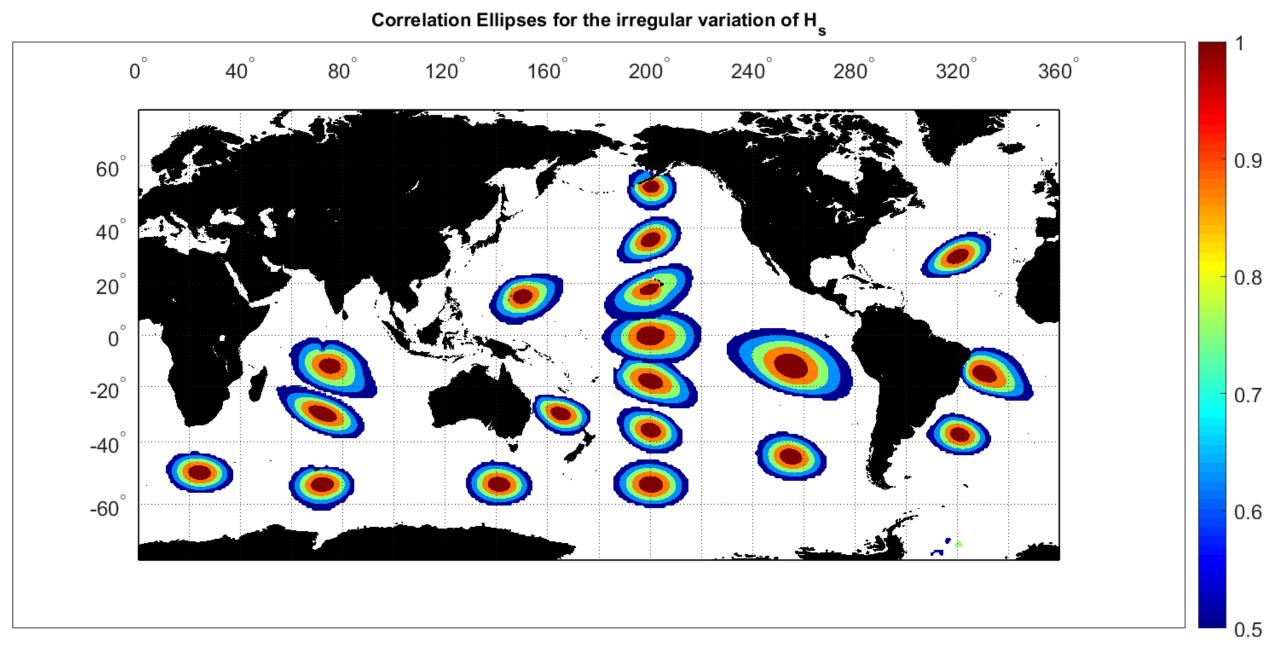


Figure 9 Correlation Ellipses - irregular variation of H<sub>s</sub> [monthly means subtracted from the time series]

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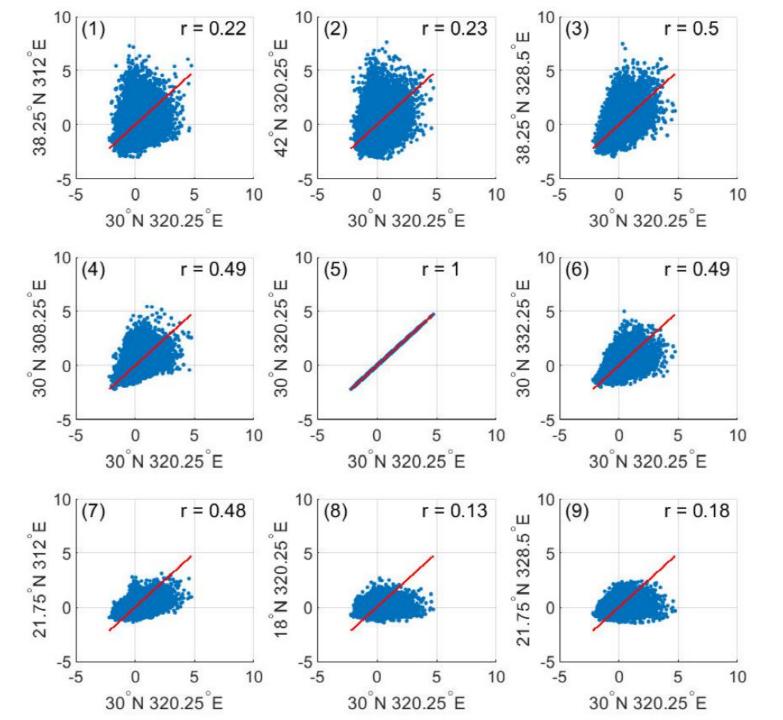


Figure 10 Scatter plots of deseasonalised  $H_s$  between a location at 30°N, 320.25°E (North Atlantic) and surrounding locations (12° radius)

- 3 Global extreme wave height from spatial ensemble data (Takbash and Young 2019)
- → Assessment of spatial coherence between **areas in zonal direction**!

#### Second criterion:

The relative percentage difference (RPD) between

- monthly means (averaged over the duration of the records)
- averaged monthly 99<sup>th</sup> percentile values

→ Wave climate is similar when RPD is less than 10 % for both parameter!

$$\begin{split} \overline{\text{RPD}}(i,j) &= \frac{1}{12} \sum_{k=1}^{12} \frac{|\overline{H}_s(i,k) - \overline{H}_s(j,k)|}{\overline{H}_s(i,k)}, \\ \text{RPD}^{99}(i,j) &= \frac{1}{12} \sum_{k=1}^{12} \frac{|H_s^{99}(i,k) - H_s^{99}(j,k)|}{H_s^{99}(i,k)}. \end{split}$$

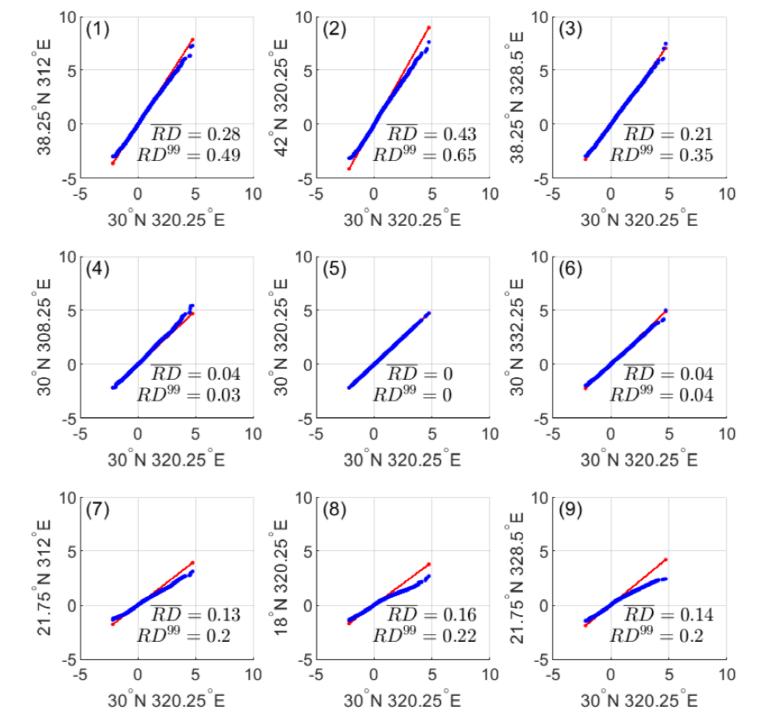


Figure 11 QQ plots of deseasonalised  $H_s$  between a location at 30 °N, 320.25 °E (North Atlantic) to surrounding locations (12° radius)

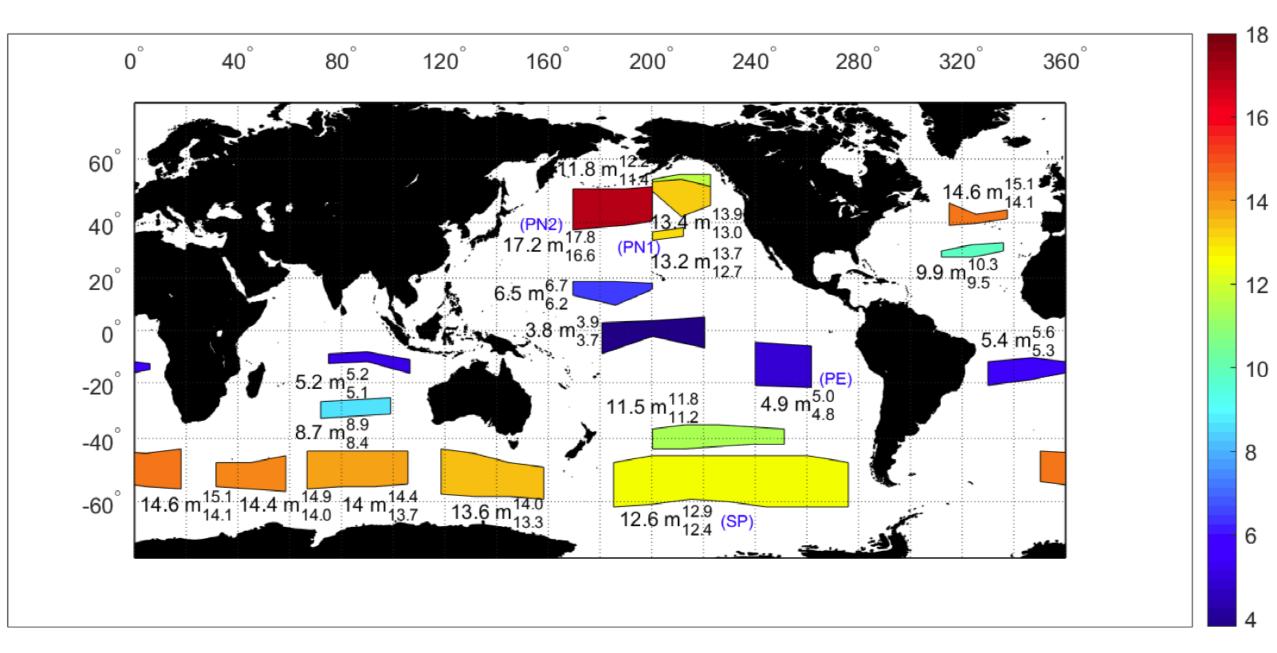
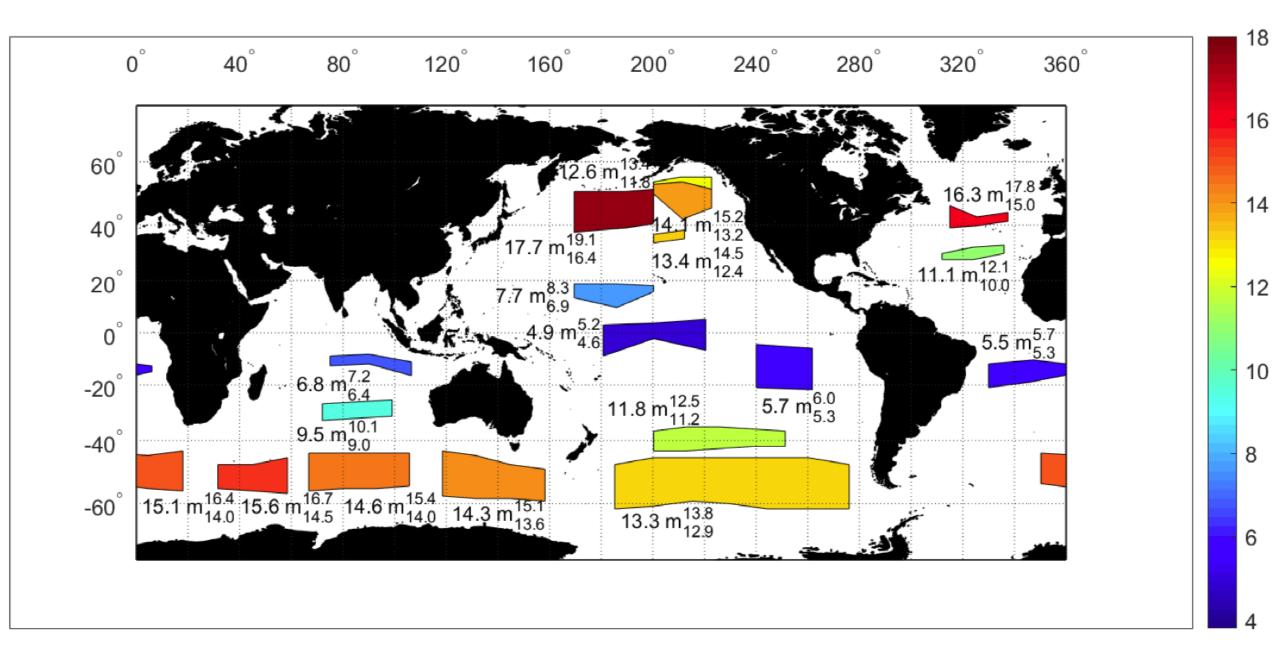


Figure 12 Ensemble spatial regions for ERA-Interim data



## 4 Conclusion

For ERA-Interim data we created spatial ensemble equivalent to:
60 yrs (N Pacific) to 210 yrs (S Pacific)

□ For Altimeter data we created spatial ensemble equivalent to: 54 yrs (N Pacific) to 189 yrs (S Pacific)

 $\rightarrow$  Reduced size of the confidence interval by ~30% - 60% depending on the size of the spatial ensemble data