



**National Oceanography  
Centre, Southampton**

UNIVERSITY OF SOUTHAMPTON AND  
NATURAL ENVIRONMENT RESEARCH COUNCIL

# Extreme Wave Heights from Satellites

Peter Challenor

National Oceanography Centre,  
Southampton, UK

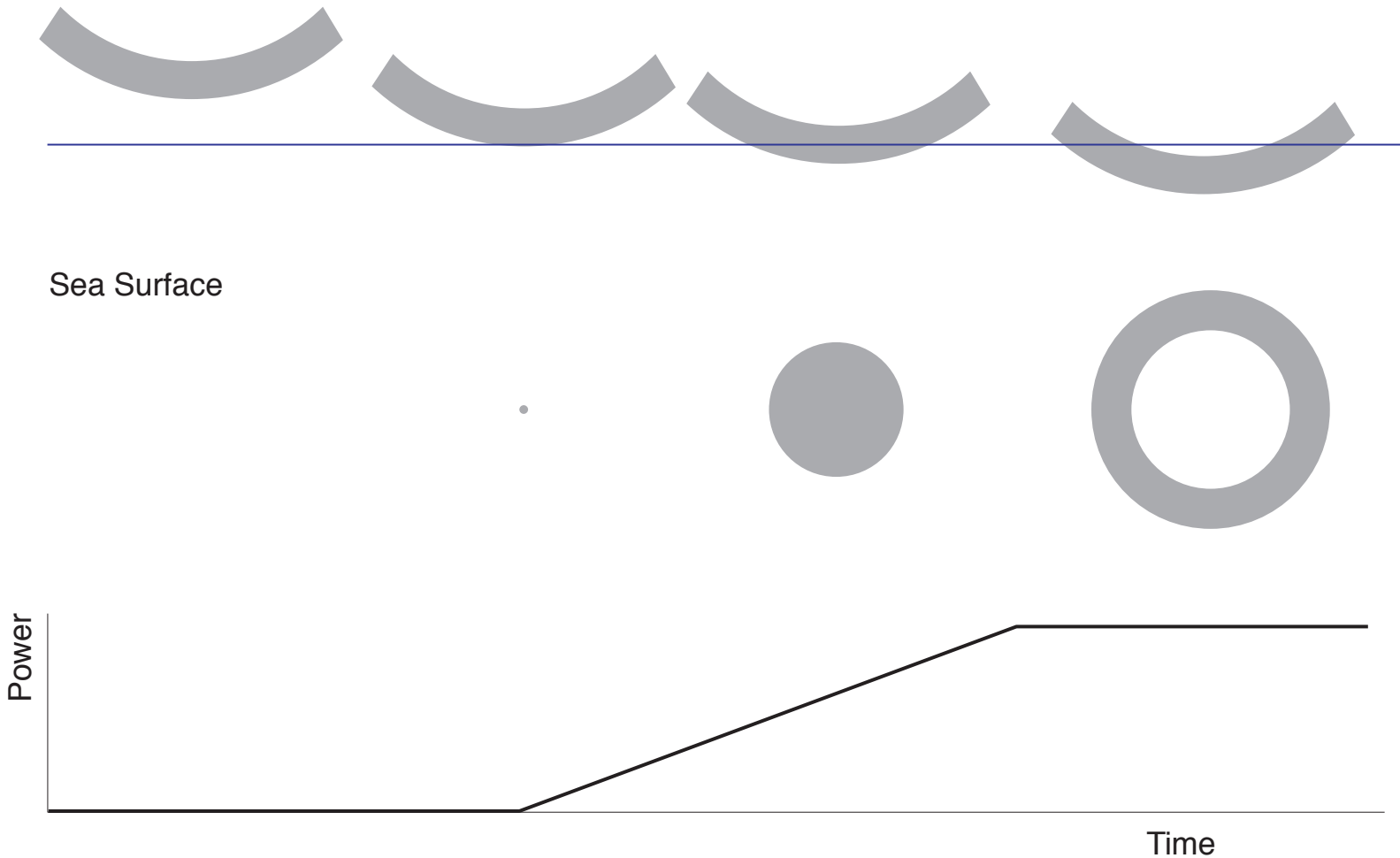
# Outline

- Why are we interested in extremes from satellites?
- Waves from radar altimeters
- Extremes and altimeter data
- Classical Methods
- Bayesian Hierarchical models

# Waves from satellites

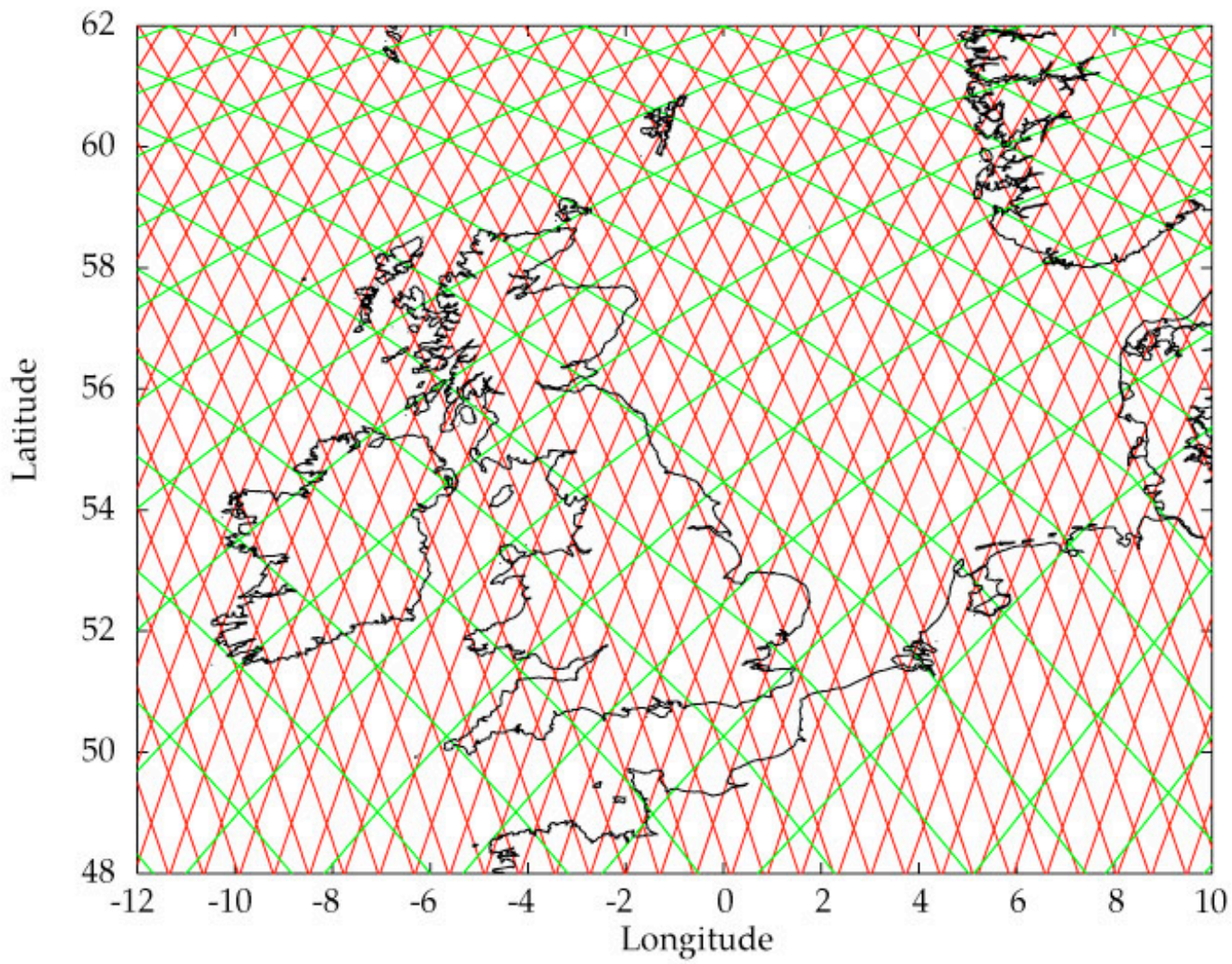
- Limited coverage from buoy measurements
- Models are models
- Satellites give a direct measurement of wave conditions
- Radar altimeters and SAR's

# How an altimeter works



# Altimeter waves

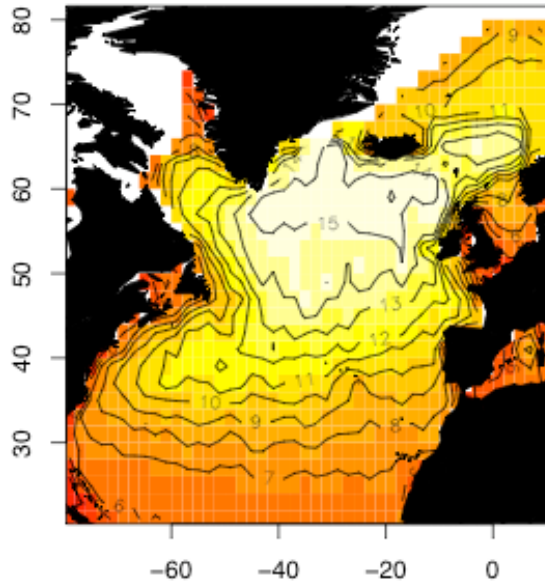
- Direct measurement of significant wave height  
(Can also measure wave period)
- Accuracy comparable (or better than) buoys
- Calibration of instruments needed to produce a consistent data set between instruments
- Only measures directly below the satellite



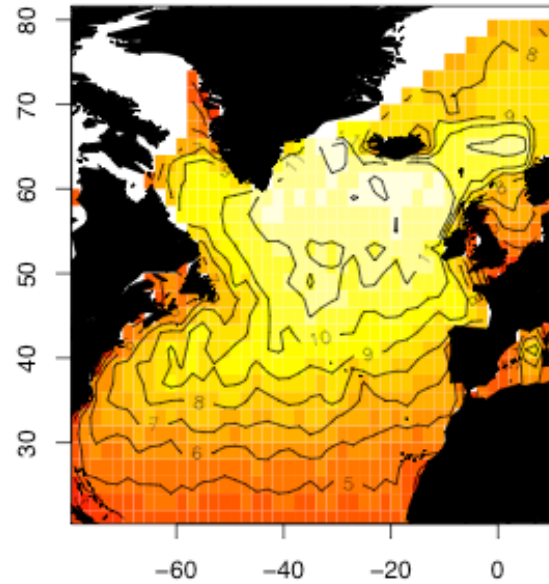
# Data

- TOPEX and ERS-1/2 data from 1992 -2004 from the NOC GAPS database
- Data calibrated to US data buoy network
- (Also processed data from the TUD RADS database - similar results - not presented here)

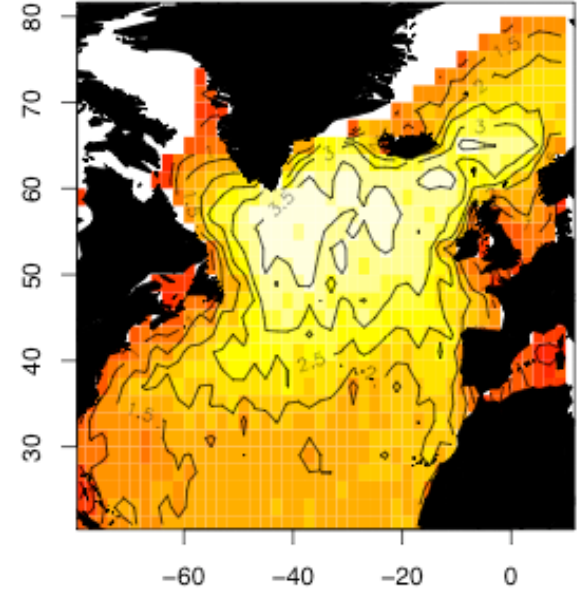
FT-1 (Gumbell)



Weibull

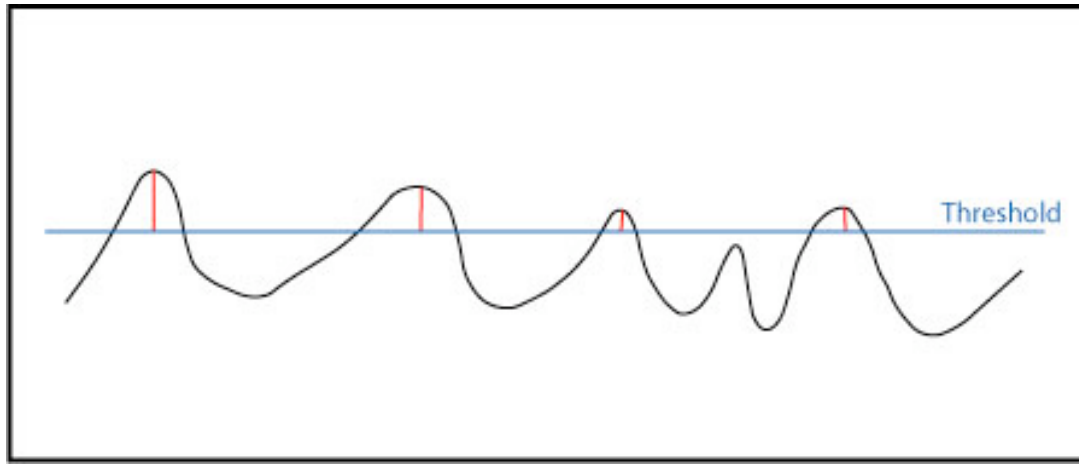


Difference in  $H_{50}$



- Different distributions give very different answers.





1. Choose a high threshold ( $u$ )
2. Take all exceedences above that threshold
3. Fit a GPD distribution

$$P(X < x | x > \mu) = 1 - \left[1 + \frac{\eta \cdot x}{\sigma}\right]^{-\frac{1}{\eta}}$$

# POT for altimeter data

- Altimeter data do not record the biggest exceedences at a point
- But the distribution of any exceedence = probability of the largest
- The temporal sampling is poor
- Undersampling will lead to an underestimate of the extremes by about 10-15% (Robinson and Tawn, 2000)

# Altimeter Processing

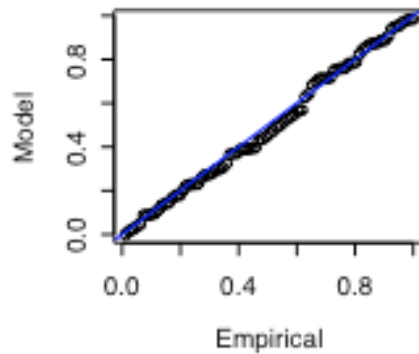
- For each  $2^\circ$  square in the North Atlantic take each altimeter pass across the square.
- Replace each pass by its median (declustering)

# Setting the Threshold

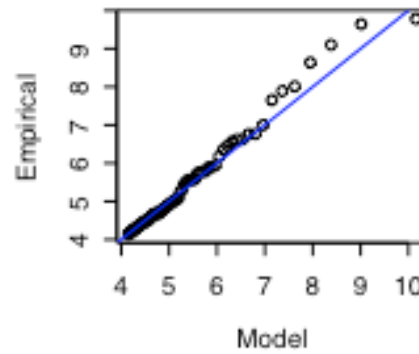
- With *in situ* data we set the threshold by hand - start high and lower the threshold until nothing changes
- We cannot do that so we set the threshold to the 90th quantile i.e 10% of the data are exceedences

# An Example Fit

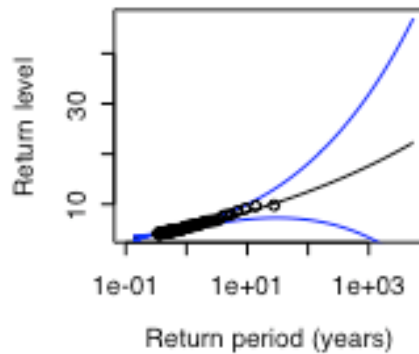
Probability Plot



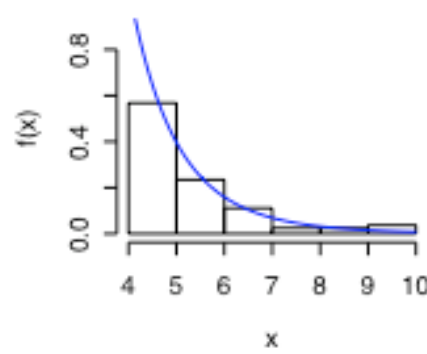
Quantile Plot



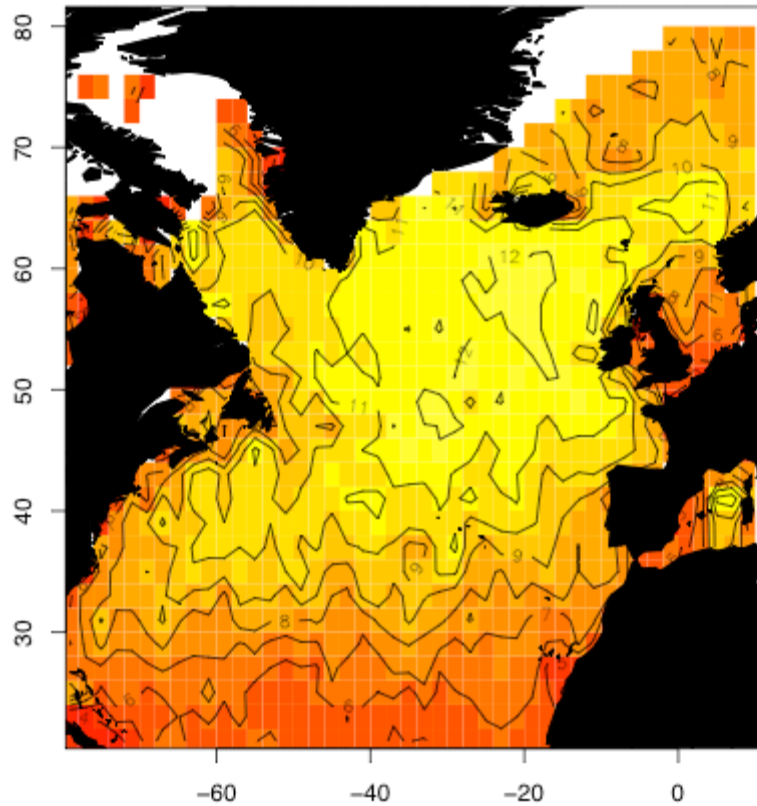
Return Level Plot



Density Plot



# 50-year Return Value



Max value = 14.9m

# Non-stationarity

- So far we have assumed that the wave climate is stationary i.e. that the statistics of exceedences are the same throughout the year
- This is not true - storms are bigger in the winter

# Variable Threshold

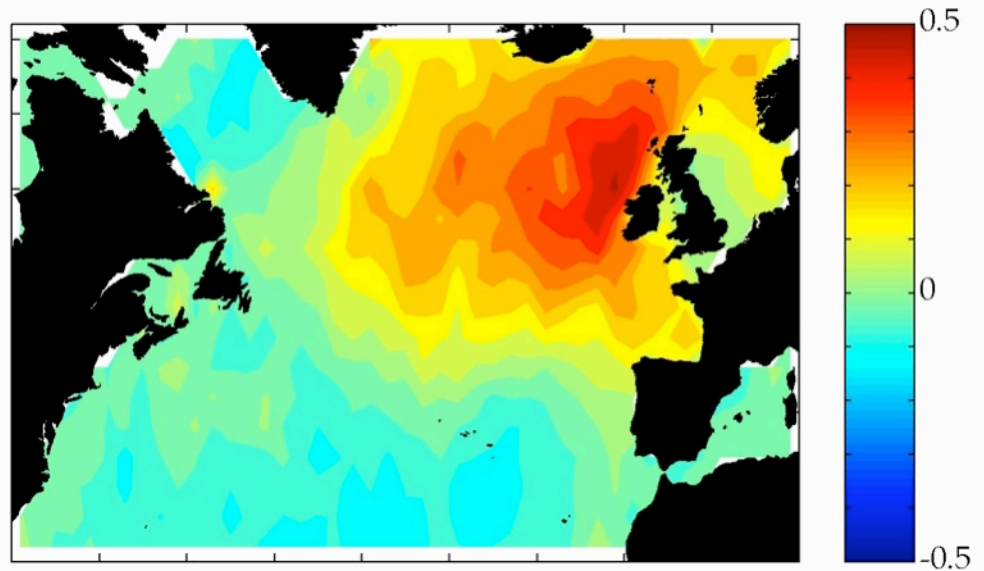
- Use a different threshold value every month
- We can check whether this more complex model is 'better' (likelihood ratio test)
- It is significantly better everywhere



# The North Atlantic Oscillation

- We know that the NAO affects mean wave height
- Does it also affect the extremes?

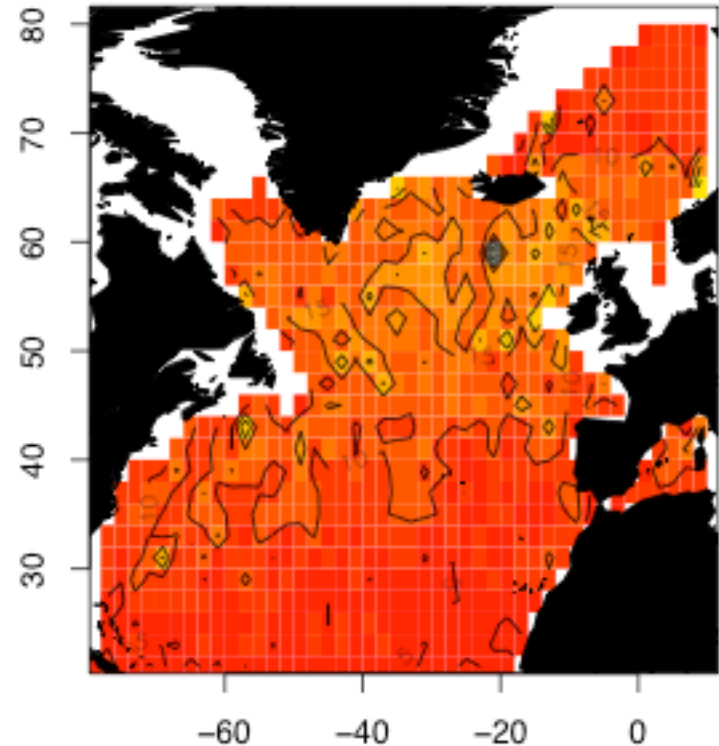
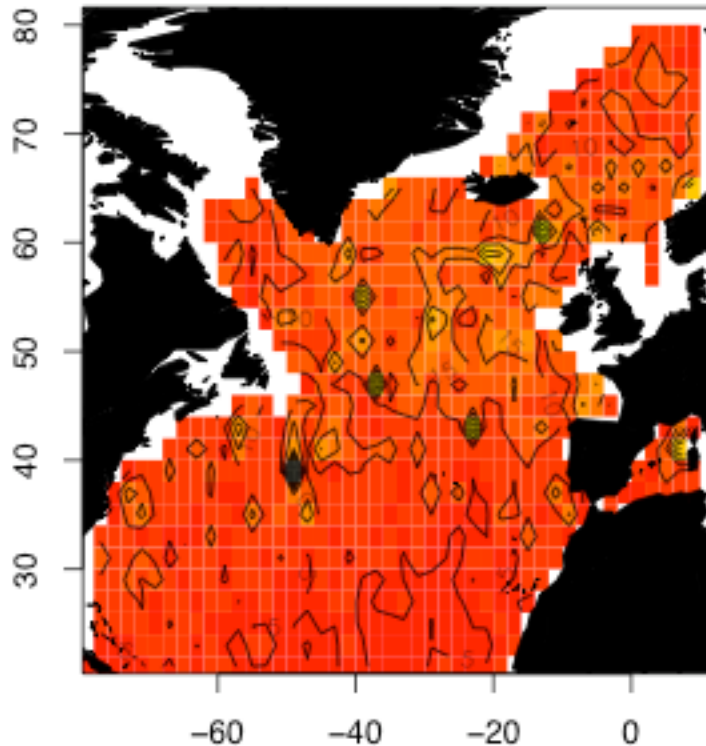
Sensitivity of Average Winter Wave Height to NAO (metres/unit index)



Adding monthly values of the NAO to the scale parameter does little to improve the fit.

If we add an interaction term - a different response to the NAO each month we get better results

These plots show a low (left) and high (right) NAO January  $H_{50}$



# Interim Conclusions

- We have found an NAO response, but it is noisy.
  - Perhaps we need to use the winter rather than monthly NAO
  - Or vary the shape parameter
  - Or we may need more data

# Spatial Models for Extremes

- So far we have assumed that the extremes in each  $2^\circ$  square are independent of the extremes in the squares around it.
- This is clearly not true
- Two ways to approach this
  - Use a continuous model in space (similar to kriging or optimal interpolation)
  - Treat each  $2^\circ$  square separately but build in some correlation structure between them
- For computational reasons we are investigating the latter

# A Bayesian hierarchical model

- Conditional on the values of the parameters in each  $2^\circ$  square the exceedences are given by a GPD
- The parameters are either fixed ( $\mu$ ) or have priors attached to them.
- The calculations are done in WinBUGS
- square are a weighted average of the parameters in adjacent squares.
- Computations still underway

# Conditionally Autoregressive Model

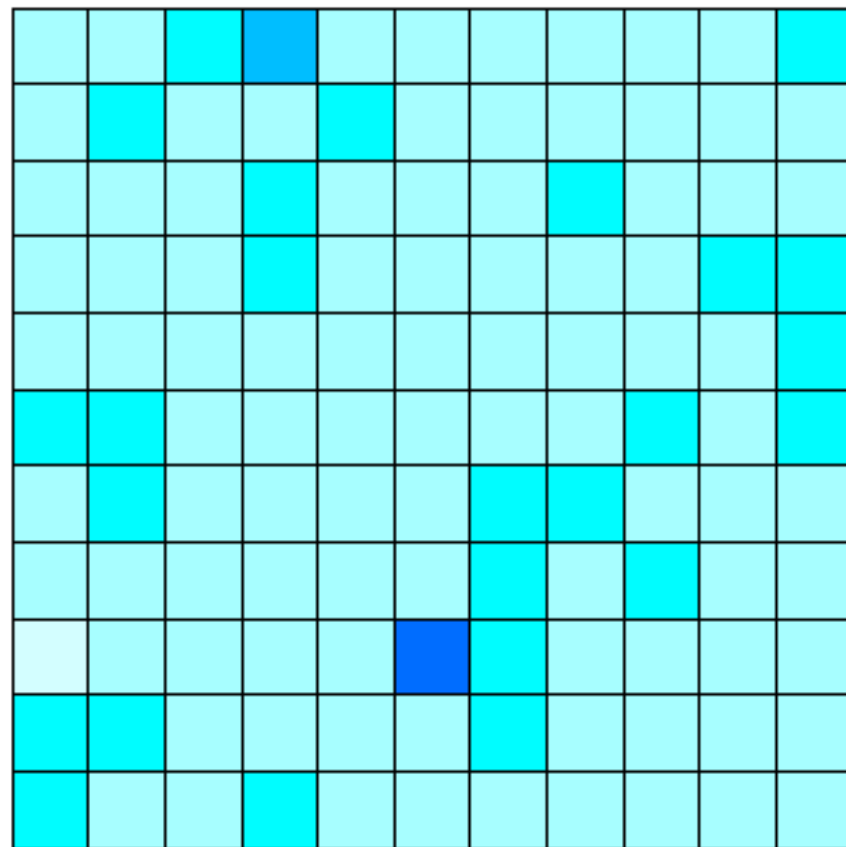
- For a spatial model we make the prior for  $\sigma$ , say, follow a 'conditional autoregressive' (CAR) model.
- Essentially this means that prior for  $\sigma$  in one square is a weighted average of the prior values in adjacent squares.
- Bermudez, Mendes, Pereira and Turkman (2008)

# Non Spatial

- Use 11x11 square for test purposes
- 12-22°N 24-34°W
- 20 largest values in each square
- Threshold = 21st
- Gamma Prior on  $s$
- Normal Prior on  $\eta$
- Run Gibbs sampler for 1000 iterations to burn in

[samples]means for H50

N  
↑



- Light Blue [1] < 10.0
- Light Blue [93] 10.0 - 12.
- Light Blue [25] 12.5 - 15.
- Light Blue [1] 15.0 - 17.5
- Dark Blue [1] >= 17.5

0.01 km



# Spatial

- CAR prior on  $s$
- constant  $\eta$ ?
- CAR on  $\eta$

# Things to do

- Get it working!
- Larger area; better choice of data
- Proper priors
- Covariates - NAO etc