

# Development of a wave ensemble system at the Met Office

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11th Wave Workshop, Halifax, 18-23/10/2009



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- Address weakness of deterministic models
  - No indication of possible alternatives
  - Can miss extreme events
  - No measure of situation-specific uncertainty (general uncertainty info available from routine validation)
  - Doesn't allow to quantify risk
  - To quantify confidence in the forecast as it varies between forecast cycles (flow-dependent uncertainty)
  - To obtain reliable probabilities of wave events happening
    - By sampling from a range of possible forecasts in a way which is consistent with the error structure of the observations and of the model



Investigate feasibility of running a short-range wave ensemble system on the MO computer architecture

- Run test cases to determine ideal configuration of a trial system:
  - influence of initial conditions
  - Influence of boundary conditions
- Develop an ensemble suite based on operational wave model: global EPS (T+72) and nested regional EPS (T+54)
- 3 months trial period from Feb-Apr 2009
- Validation: spread-skill relationship and reliability of probabilities
- Development of trial products



- Optimal run configuration determined consisting of global 90km and regional 24km ensemble
- Tested over a 3 month period
- Forcing winds are good quality and provide a suitable forcing to the system
- Ensemble mean performance very close to that of the control run
- Spread is a good indicator of forecast uncertainty but further work required (e.g. perturbing model physics)
- Forecast probabilities are good (esp. 1m and 2m thresholds)
- Potential products have been developed to aid interpretation of the results



#### Description of the models

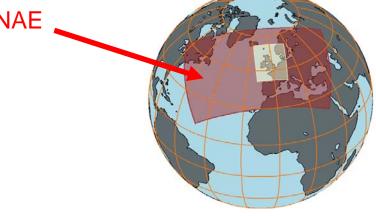
Atmospheric ensemble (MOGREPS) and wave model



#### MOGREPS – The Met Office shortrange ensemble

- 24-member ensemble designed for short-range forecasting
  - Global ensemble (~90km resolution, 38 levels) to T+72
    - Also runs to 15 days at ECMWF for multi-model ensemble research
  - Regional ensemble over N. Atlantic and Europe (NAE) (24km resolution, 38 levels) to T+54
  - ETKF for initial condition perts
  - Stochastic physics
  - Global run at 0Z and 12Z.
    Regional run at 6Z & 18Z







#### WAVEWATCH III

- Operational version of WAVEWATCH III with 2<sup>nd</sup> order advection scheme
- Spectral resolution: 24 directions X 25 frequencies
- Tolman-Chalikov source terms
- Snl evaluated with DIA scheme



## Development of the wave ensemble system



#### System design

Met Office

Reliable system can be obtained through a choice of the following:

- Atmospheric Forcing (MOGREPS)
- Initial Conditions: start from single field
  - use each previous run for IC's

- perturb initial conditions

- Boundary Conditions : determ. BC for NAE
   ensemble BC for NAE
- Perturbed physics in the model
  - Outside the scope of this study

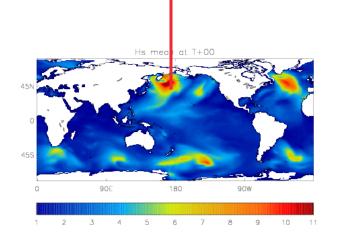
Our aim is to obtain a reliable Wave EPS for a regional domain (NAE)

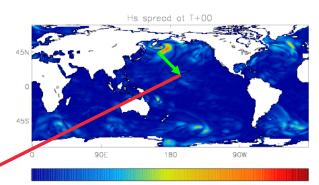


### Initial conditions (1)

#### Swell generated here at T+0

- System not expected to depend strongly on IC due to the weakly nonlinear, highly dissipative nature of the wave equations
- Tested in 2 phases:
  - 1. Starting from identical condition, run model with 24 different forcing winds for five days (using first 12 hours of each MOGREPS cycle)
  - 2. Allow system to relax under identical forcing winds
- Results: most of the spread vanishes within 12-24 hours
- Memory of initial wind-sea almost totally lost after 12-24 hours
- Memory of swell retained up to T+60 and beyond





0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1 1.1 1.2 1.3 1.4 1.5 1.6 1.7 1.8 1

Residual spread in swell field at T+60



### Initial conditions (2)

Three options:

- Perturb the initial wave field using an algorithm such as ETKF or SV – requires either DA system or linearised tangent operators
- Use the latest T+0 "analysis" field for all ensemble members =>no spread present in the system initially, but start from a best guess field
- 3. Restart each ensemble member from the previous cycles' forecast => start with an initial spread

Option 3 was selected to maintain spread at low lead times



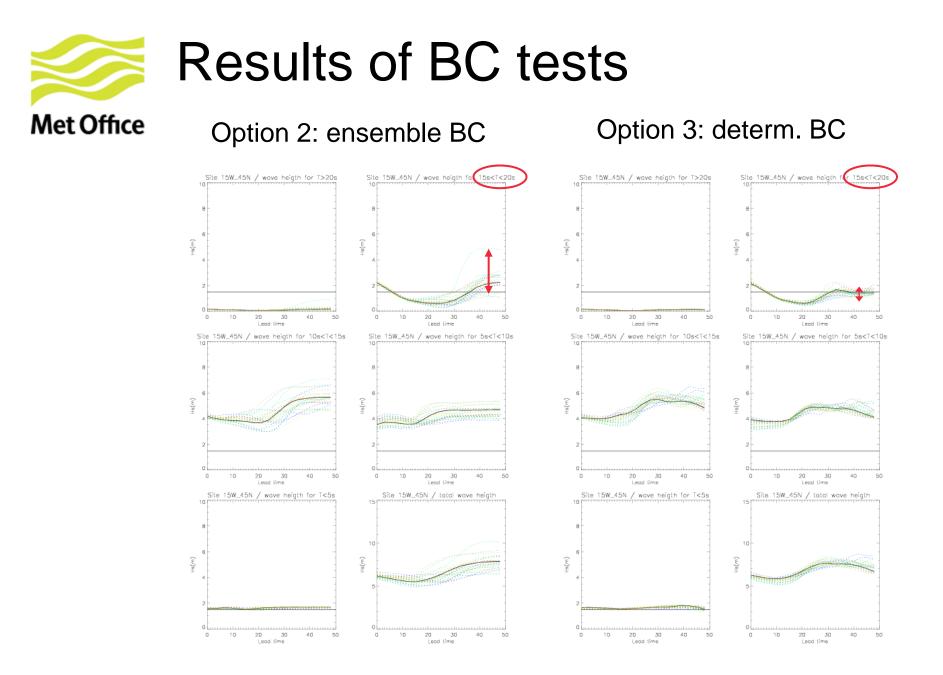
### Boundary conditions (NAE)

Two possible approaches:

- Use the same boundary condition for all regional ensemble members – computationally less expensive but potential loss of spread in the swell part of the spectrum
- 2. Use BC generated by a global ensemble

Similar overall statistics (will only be noticeably different in swell dominated seas)

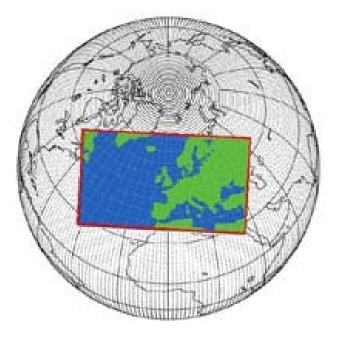
Quite different when looking at certain periods (15s<Tm<20s) Option 2 was selected





#### Met Office

- Global ensemble:
  - 24 members
  - each member restarts from previous cycle's T+12 field
  - 90km run
  - 24 directions X 25 freq.
  - run to T+72
  - Provides BC for NAE
- NAE ensemble:
  - 24 members
  - each member restarts from previous cycle's T+12 field
  - 24km run
  - same spectral res.
  - run to T+54
- All forcing from MOGREPS



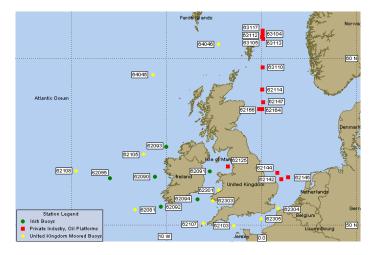


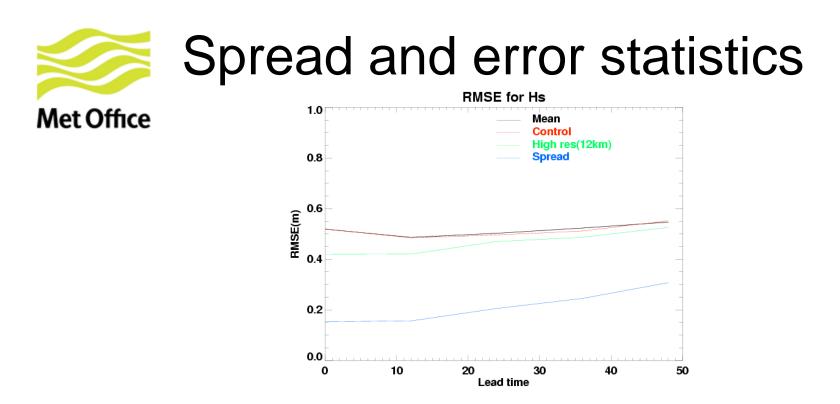
#### Verification results (focus on NAE)



#### Aims of the verification

- Assess deterministic value of ensemble mean
- Check if the ensemble spread is a good indicator of forecast uncertainty
- Verify the probabilities derived from the system
- All this work has been done using wave buoy data

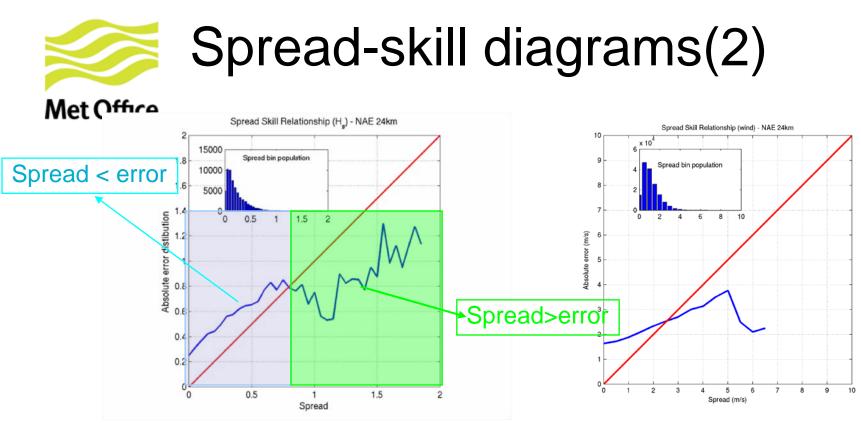




- Ensemble mean fc expected to outperform control member => not clear
- Highres initially has lowest RMS error but gap closes with incr. lead time
- Spread increases with lead time as expected and explains ~50% error
- Contributions to error:
  - 1. Forecast uncertainties (IC,forcing) -> spread
  - 2. Model uncertainties (physics, not considered here)
  - 3. Observational errors (~10% of observed Hs)



- How do we check if the spread is a good indicator of forecast uncertainty?
- Can be verified with spread-skill diagrams
  - X-axis: spread
  - Y-axis: 80<sup>th</sup> percentile of the error for each spread bin
  - Ideal spread-skill: 1-1 relationship
  - Residual errors linked to imperfect model/obs and finite size of the ensemble



- Spread is good predictor of forecast error (80<sup>th</sup> percentile of absolute error on the ensemble mean)
- Main population in lower spread bins
- When the spread in the FC is low, the spread underestimates FC error (more confident in this)
- When the spread in the FC is high, the spread overestimates FC error
   ⇒ Prudent approach

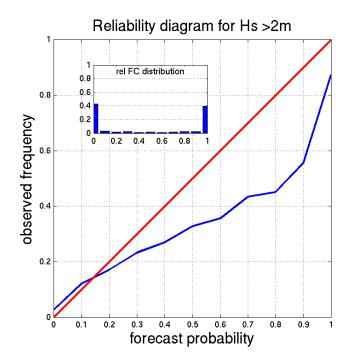


### Verifying probabilities

- Deterministic models: predict event and verify against the observational outcome
- Probabilistic models: no true or false forecast
- "If we predict that an event will happen with a probability of 60% and the event happens, where we right or wrong?"
- Right if in 60% of those cases, the event actually happens => reliability diagrams
- Requires long period to verify



- Graph of observed frequency of an event vs. forecast probability of the event
- If event probability is 60%, it should occur 60% of the time in the long run
- Graph should be a diagonal for a perfect system
- Ref: Wilks 1995

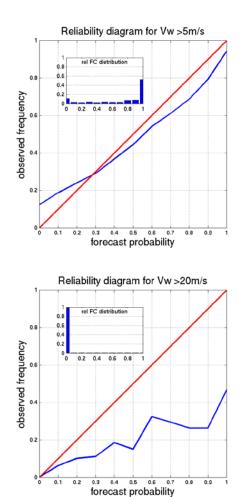


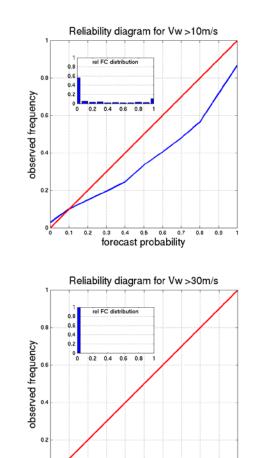


### Reliability diagrams (Vw)

• Tendency to underestimate probabilities

- Pretty good performance for lower wind speeds
- Lack of extreme events in trial period





0.3 0.4 0.5 0.6 0.7

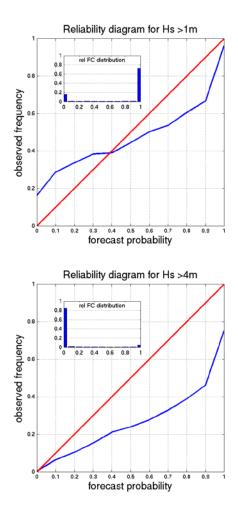
forecast probability

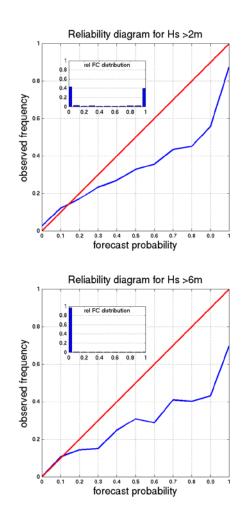
0.8 0.9

0 0.1 0.2



#### Reliability diagrams (Hs)







### Potential products



# Challenges in conveying probabilistic information

- People (and some journalists) don't understand probabilities (e.g. BBQ summer coverage)
- Customers want to make a yes/no decision

The office now claims that it is "66% certain" that next winter will be warmer and wetter than last. The figure is an ominously precise advance on the 65% certainty of a warm summer. The information is useless without knowing the likelihood of the "66%" being correct.

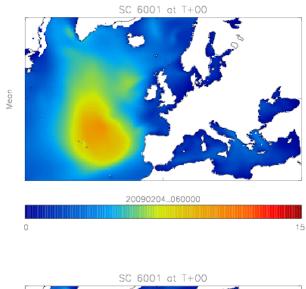
# Met Office

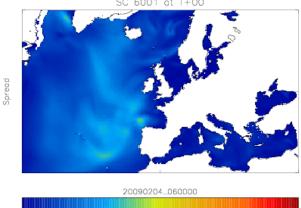
### Products – Mean/Spread plots

0

#### • Shows mean forecast (top)

- Ensemble spread (bottom) shows where the forecast uncertainties are located geogr.
- Two options:
  - Location of interest is in low spread zone -> forecast reliable -> use as det. Forecast but with higher confidence
  - Location of interest is in high spread zone -> more detailed look required ->see next slides



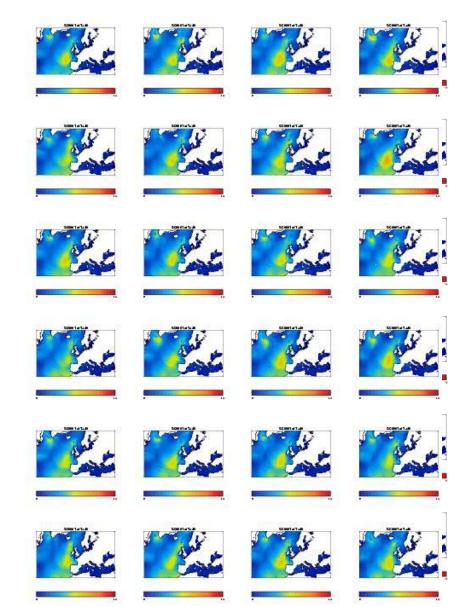




#### Products – Postage stamps

#### Shows all forecasts

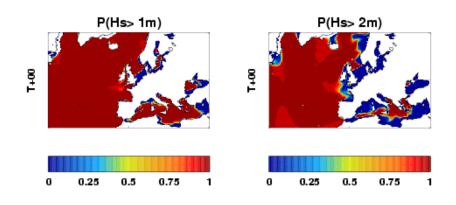
- Overview of each ensemble member
- Allows visual clustering
- Differences for instance at T+40 south of Greenland

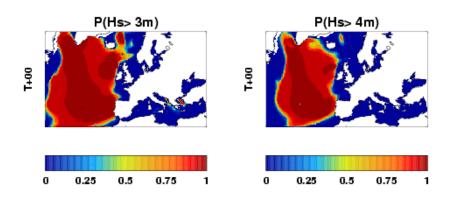


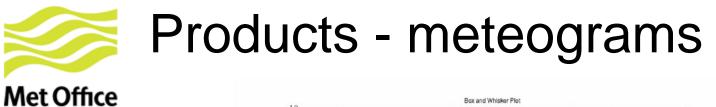


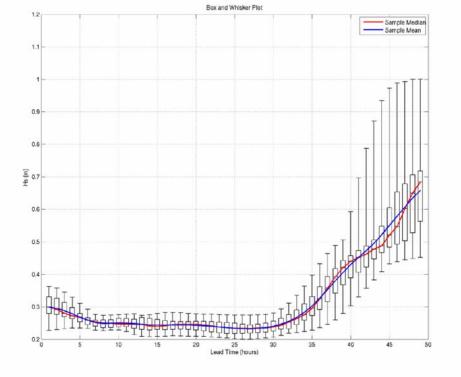
#### Products – Probability maps

- Shows probability of  ${}^{\bullet}$ exceeding different thresholds [1m,2m,3m,4m]
- Help for operational decisions (e.g. operation with 4m thresh for RN)



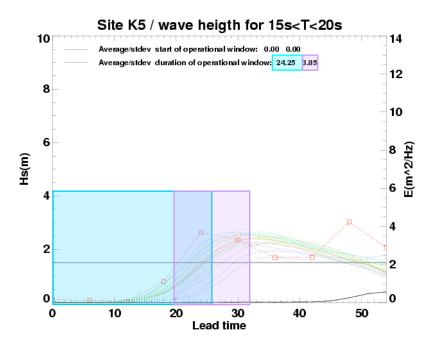






- Site-specific product
- Shows probabilistic forecast for all lead times
- Shows mean, median and quartiles

Products – Operational windows Met Office



- Some offshore operations are mainly sensitive to energy in certain frequency bands (e.g. heavy-lift and low period swells) => select band and operational threshold
- Integrate E spectrum over that band to obtain an equivalent waveheight
- Check for operational window where threshold is not exceeded



#### Conclusions / Future work



- Optimal run configuration determined consisting of glbal and regional ensemble
- Tested over a 3 month period
- Forcing winds are good quality and provide a suitable forcing to the system
- Ensemble mean performance very close to that of the control run
- Spread is a good indicator of forecast uncertainty but further work required (e.g. perturbing model physics)
- Forecast probabilities are good (esp. 1m and 2m thresholds)
- Potential products have been developed to aid interpretation of the results



- Investigate ways of representing model uncertainty in the forecast (e.g. perturbed physics schemes)
- Extend validation to include satellite data to improve our understanding of the performance of the system away from coastal waters
- Implement Relative Economic Value score to understand how valuable the system would be
- Target a period with more severe conditions to improve stats on severe sea state events