

MOGREPS Met Office Global and Regional Ensemble Prediction System

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Ensemble Forecasting Manager





- Introduction to Ensemble Forecasting
- Examples from
 - ECMWF Medium-range EPS
 - Met Office short-range ensemble (MOGREPS)
- Probability forecasts
 - What do they mean
 - Decision-making

The Met Office has a World-leading forecasting system, but nevertheless...

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- All forecasts are uncertain
- High-profile forecast failures are now rare, but do still occur (eg Dec '99 European storms)
- Less severe errors are much more common, e.g.
 - medium-range forecasts
 - finer details such as timing of rainfall
 - E-W position of snow over recent weekend
- Ensembles turn weather forecasts into Risk Management tools



- The atmosphere is a chaotic system: "... one flap of a seagull's wing may forever change the future course of the weather", (Lorenz, 1963)
- Up to about 3 days ahead we can usually forecast the general pattern of the weather quite accurately
- Beyond 3 days Chaos becomes a major factor

Tiny errors in how we analyse the current state of the atmosphere lead to large errors in the forecast – these are both equally valid 4-day forecasts!









Fine details (eg rainfall) have shorter predictability

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Ensembles – estimating risk



By running model(s) many times with small differences in initial conditions (and model formulation) we can:

- take account of uncertainty
- estimate probabilities and risks
 - eg. 10 members out of 50 = 20%

ECMWF ENSEMBLE FORECAST Data Time : 02/11/2004 12z D+ B Valid at : 10/11/2004 Significant Wave Height in m







5.5m

ECMWF Ensemble (EPS)



51 members

- Control (unperturbed) + 25 pairs formed by adding and subtracting a perturbation
- T_L319 Resolution (approx 60km)
- Designed for use beyond 48h
- Perturbations are linear combinations of Forward and Evolved Singular Vectors
- Includes Stochastic Perturbations to model physics

ECMWF Ensemble prediction System (EPS)





Carlisle storm, Jan 05, from ECMWF 51-member medium-range ensemble

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MOGREPS – The Met Office short-range ensemble



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





MOGREPS has successfully completed a 1-year Operational Trial. Scheduled to become operational by Dec 2007.

MOGREPS Operational System diagram



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Desirable Properties and how we achieve them

Probability forecasting



Desirable properties of an ensemble:

- All members must be equally likely
- RMS spread ~ RMS error of Ens.mean
- Ensemble spread should include observations (most of the time! ~ 2(100/n)%)
- Spread-skill relationship:
 - Small spread should indicate high probability ... but large spread not necessarily mean low skill!

If these criteria are met the ensemble may be used to estimate probabilities

Desirable Properties of Ensembles

 RMS Spread of members similar to the RMS Error of the control forecast



Verification of EPS: Verifying quality



We can verify the spread of the ensemble The Talagrand diagram

With only one ensemble member (|) all observations (•) will fall "outside"

With two ensemble members two out of three observations (2/3=67%) will fall outside

With three ensemble members two out of four observations (2/4=50%) will fall outside

• | • | • | •

• •

• • •

Two observations out of N will always fall outside yielding a proportion of 2 / N outside





The Rank Histogram



Initial Condition Perturbations



- Getting different ICs that estimate the analysis error is not enough ...
 - not all errors in the analysis are likely to grow
 - Limited computing resources

We need to focus on those perturbations that rapidly diverge

Initial conditions: Bad sampling!



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time

The ensemble capture reality less often than it should

Dangerous: false sense of security!





Singular Vectors (ECMWF; Molteni et al., 1996)

- SV try to identify the dynamically most unstable regions of the atmosphere by calculating where small initial uncertainties would affect a 48 hour forecast most rapidly. It needs an adjoint model.
 - SV perturbations are very small at initial time grow rapidly
 - Expensive to calculate done at low resolution (T63)

ECMWF ENSEMBLE PERTURBATIONS. 7/ 2/2001 12z.

Mean Sea Level Pressure - hPa

Shading shows magnitude of perturbations.

SV Perturbations

Each perturbation is a linear combination of:

- 25 NHem SVs
- 25 SHem SVs
- 25 Tropical moist SVs targetted on
 - Caribbean
 - TCs
- Evolved SVs
 - Calculated 48h previously
 - Iarger





0.3 0.5 0.7 0.9 1.1

Perturbation:





0.3 0.5 0.7 0.9 1.1





Error Breeding



Start with random perturbation - allow to grow in forecast

Rescale bred mode to analysis errors (fixed climatological rescaling factor)

Use for perturbation in next cycle

Cycle "breeds" the rapidly growing modes in the analysis cycle



Toth and Kalnay (1997), MWR 125, 3297-3319

Analysis Perturbations - Error Breeding





T+12 perturbed forecast

)*F

T+12 control forecast





Perturbation structure

Perturbation Structures – Mean and spread PMSL







Perturbation Structures – Mean and spread PMSL



 Spread tends to be concentrated around fronts and sharp gradients

 Perturbation is nonzero everywhere (in contrast to SVs)





Stochastic Physics



Stochastic physics in MOGREPS



MOGREPS employs three schemes to address different sources of model error:

- Random Parameters (RP)
 - Error due to approximations in parameterisation
- Stochastic Convective Vorticity (SCV)
 - Unresolved impact of organised convection (MCSs)
- Stochastic Kinetic Energy Backscatter (SKEB)

Excess dissipation of energy at small scales Impact is propagated to next cycle through the ETKF

Model error: using a single-model



Random parameters

QUMP (Murphy et al., 2004)

 Initial stoch. Phys. Scheme for the UM (Arribas, 2004)

Parameter	Scheme	min/std/Max
Entrainment rate	CONVECTION	2/3/5
Cape timescale	CONVECTION	30 / 30 / 120
Rhcrit	LRG. S. CLOUD	0.6 / 0.8 / 0.9
Cloud to rain (land)	LRG. S. CLOUD	1E-4/8E-4/1E-3
Cloud to rain (sea)	LRG. S. CLOUD	5E-5/2E-4/5E-4
Ice fall	LRG. S. CLOUD	17 / 25.2 / 33
Flux profile param.	BOUNDARY L.	5 / 10 / 20
Neutral mixing length	BOUNDARY L.	0.05 / 0.15 / 0.5
Gravity wave const.	GRAVITY W.D.	1E-4/7E-4/7.5E-4
Froude number	GRAVITY W.D.	2/2/4



2004012700Z - T+72



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Stochastic Kinetic Energy Backscatter (SKEB)

- Based on original idea and previous work by Shutts (2004)
- Related to new scheme for ECMWF EPS
- Aim: To backscatter (stochastically) into the forecast model some of the energy excessively dissipated by it at scales near the truncation limit
- In the case of the UM, a total dissipation of 0.75 Wm-2 has been estimated from the Semi-lagrangian and Horizontal diffusion schemes. (Dissipation from Physics to be added later on)
- Each member of the ensemble is perturbed by a different realization of this backscatter forcing © Crown copyright 2007
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SKEB



Streamfunction forcing:

$$F_{\psi} = \frac{1}{2} \alpha \cdot K \cdot R(\lambda, \phi) \frac{\sqrt{\Delta \tau \cdot D}}{\Delta \tau}$$

K.- Kinetic En.; *R*.- Random field; $\Delta \tau D$.- Dissipated en. in a time-step

R is designed to reproduce some statistical properties found with CRMs





Preliminary results:

Positive increase in spread (comparable to that seen at ECMWF)



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Product Examples






Example MOGREPS 33h Rainfall forecast



 Ensemble offers much more information on

> areas at risk

intensity



Example MOGREPS 33h 10m WS forecast







MOGREPS (Regional) Probability map for 6HourPrecip > 5.0mm DT 18Z on 27/02/2007 VT 00Z on 28/02/2007 lead time 06h (Ensemble Mean PMSL plotted as faint background)



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Products for the Risk Manager

Total Precipitation (mm/6 hr) 60 50 40 30 20 10 8 65 4 3 2 .8 65 4 .3 2 00 00 06 12 18 do 06 12 18 00 06 12 00 WED MON TUE THU FRI 10 11 12 13 14 OCTOBER 2005



Lat 58.50 / Lon 1.50





- Plot of ensemble spread
 - Range of uncertainty

- Probability graph for multiple severity thresholds
 - Example of use for risk management in offshore oil industry

Synoptic Features (from Cyclone Database)



MOGREPS (Global)





















Member 6



Member 20







Member 9





Member 21 -00



DT 12Z on 05/09/2006 VT 12Z on 08/09/2006





Member 16











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Click on Feature at T+0...





Forecast plumes for feature characteristics



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Hurricane Katrina





End-to End Outcome Forecasting



- An ensemble weather forecast can be used to drive an ensemble of outcome models, eg.:
 - Tidal surge
 - Ocean waves
 - Wind power output
 - Energy demand
 - Hydrology flood risk
 - Ship or aircraft routes

Prediction of UK & Wales Electricity Demand (June '99) 41 39 **Giga Watts** 32 32 32 33 31 29 08 09 11 02 03 04 05 07 10 06 June 1999



Storm surge Ensemble – contract for EA

CS3 Storm surge model coupled to MOGREPS NAE







Probability of surge elevation > 1.0m DT 06Z on 17/01/2007 VT 07:00 on 17/01/2007 T+01h00



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(green/blue: predicted/observed surge, solid/dashed red; alert - harmonic/model tide)

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Using Ensembles

Met Office Operations Centre







Ops Centre forecaster uses the ensemble to assess the *most probable outcome* before creating the medium-range forecast charts...

...and assess risks



ECMWF ENSEMBLE FORECAST 17/ 2/2004 12z. T+120 Valid at : 22/ 2/2004

ean Sea Level Pressure Member with highest mean:

Example – low spread, high confidence



- Snowy period in SW, Nov 2005
- ECMWF EPS very high confidence of blocking breakdown
 - Allowed issue of high confidence of return to mild conditions on Wed 30th 5 days ahead
 - Analysis of 0600 on 30th confirms this was correct



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Variable Predictability in EPS





- ACC>0.6 indication of useful forecast
- Two graphs show variable predictability
- Many EPS members more skilful than control (solid line)
- Need to develop ways to extract information from best members, without knowing which they are
- Ensemble prediction systems (EPS) allow us to assess the *flow-dependent* predictability

Figure from Molteni et al, 1996





Forecasters have always dealt with uncertainty.



Uncertainty expressed in many ways, mostly subjective.

Probabilities - getting quantitative



How can we improve on subjective description?

People in London have a 50% chance of seeing a light shower this afternoon...



...but in the NW you have an 80% risk of a shower with a 20% chance of over an inch

This is much more informative, but only if the figures are meaningful!

Use of probability forecasts - Clear Definition

What does this mean?

30% probability of rain in Scotland

- 30% in Edinburgh City Centre?
- 30% somewhere along the M8?
- or 30% "somewhere in Scotland"?
- How much?
 - A trace? 5mm? heavy downpour?

Probabilities must be *unambiguous* and *relevant* to the end user

Clear Definition



Examples of well-defined probability forecasts could be:

- 30% probability of more than 5mm of rain at Edinburgh Airport between 1200 and 1800.
- 70% probability of wind reaching gale force in at least one place in Scotland on Tuesday.
- 10% probability of wind sufficient to cause severe structural damage in Edinburgh overnight.
- The last example shows how a low probability can give useful warning of a serious event

Communicating uncertainty



Improved presentation of meteogram information



A single Probability Forecast cannot be right or wrong. Consider:

- Probability of X is 30%
- If X happens, is this right? Or wrong?

But... out of 100 such forecasts, X should happen 30 times.Verification must be done over many forecasts

Verification









- Reliability diagram shows how well probability relates to frequency of occurrence
 - Demonstrates benefit of bias correction (red) compared to raw (green) for ECMWF wind forecasts



Severe Weather Warnings

Example: Early Warnings of Severe Weather

Met Office issues Early Warnings up to 5 days ahead - when probability ≥60% of disruption due to:

- Severe Gales
- Heavy rain
- Heavy Snow
- Forecasters Provided with alerts and guidance from EPS
- Challenges:
 - Severe events not fully resolved
 - Combination of effects
 - Few events so difficult to verify



Verification of Heavy Rainfall Warnings





f/c prob

Can we use low probabilities?



Most extreme events are inherently improbable how should we respond to low probabilities? Event probability must be related to "climatology" for decision-making, eg.

- 5% risk that a plane will crash would you board it?
- 5% risk of rain would you play golf?
- 50% prob of heavy snow in London tomorrow
- Decisions must be based on user's Cost/Loss

ratio

users with low C/L should protect at low probabilities

MOGREPS Verification (Time permitting!)



Temp SON06 (79 sites UK & Europe)



Temp<2C T+24





Temp<-2C T+24









Brier Skill Score components for Temperature Met Office BSS Res 0.8 & Rel for temperature D.6 Temp>10C 0.4 0.2 \vee 0.0 ensemble 10 obal[®] ensemble ⁴ BSS EC ensemble deg 0.8 Res -0.2 12 24 60 36 48 0 0 Lead time (h) & Rel for temperature 0.6 0.4 79 sites UK & Europe 6 Nov 2006 – 28 Feb 2007 0.2 \vee 0.0 NAE ensemble <u>5</u> Globalio ensemble i -e ensemble deg Temp>15C -0.2 12 24 60 72

0

0

36

Lead time (h)

48

Reliability diagram for surface temperature

- Reliability diagram for Temp>10C
- 79 sites UK & Europe
- 6 Nov 2006 –
 28 Feb 2007






Kalman-filter bias-corrected forecasts.



Brier Skill Score components for Precipitation



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Reliability diagram for precipitation

- Reliability diagram for 12h precip>5mm
- 79 sites UK & Europe
- 1 July 2006 –
 31 March 2007



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6hr precip > 0.3mm against gridded analysis





Verification against Analysis for Reduced old NAE Model area (up to June 07), 1.0 deg grid - NAE EPS model area (area mean method)

•Reliability and sharpness diagram for T+36 forecast.

•6h precip > 0.3mm

•Verification against Nimrod Analysis over the UK at 1.0 degree resolution.

6hr precip 5mm against gridded analysis





Forecast Probability Verification against Analysis for Reduced old NAE Model area (up to June 07), 1.0 deg grid - NAE EPS model area (area mean method) •Reliability and sharpness diagram for T+36 forecast.

•6h precip > 5mm

•Verification against Nimrod Analysis over the UK at 1.0 degree resolution.

6hr precip 25mm against gridded analysis



•Reliability and sharpness diagram for T+36 forecast.

•6h precip > 5mm

•Verification against Nimrod Analysis over the UK at 1.0 degree resolution.

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Reliability diagram for wind speed



- Reliability diagram for wind speed >F5 at T+30
- 79 sites UK & Europe
- 6 Nov 2006 –
 31 March 2007



Wind speed at least gale force 8





•Reliability and sharpness diagram for T+36 forecast.

- •10m Wind > F8
- •Verification against surface obs over UK and Europe.
- •1 Jan 06 28 Feb 07

Wind speed at least severe gale force 9





•Reliability and sharpness diagram for T+36 forecast.

- •10m Wind > F9
- •Verification against surface obs over UK and Europe.
- •1 Jan 06 28 Feb 07

Wind speed at least storm force 10





•Reliability and sharpness diagram for T+36 forecast.

•10m Wind > F10

Verification against surface obs over UK and Europe.
1 Jan 06 – 28 Feb 07

Spread-skill relationship - wind speed



- Spread-skill for windspeed binned into equal population bins by spread
 - Skill corrected for observation error
- Blue MOGREPS
- Pink No Skill
- Green Perfect



Spread-skill relationship – temperature Met Office 3.5 MOGREPS Grad = 0.91 3.0 No Skill Spread-skill for Perfect Spread 2.5 temperature binned (X) 2.0 BSW3 1.5 into equal population DJF Grad = 0.63bins by spread 1.0 Grad = 0.03Skill corrected for 0.5 observation error 0.0 0.0 0.5 1.0 1.5 2.0 2.5 3.0 3.5 Average Standard Deviation (K) Blue – MOGREPS 5 MOGREPS Grad = 0.96Pink – No Skill No Skill Perfect Spread Green – Perfect JJA RMSE (K) 2 Grad = 0.08 Grad = 0.011 0 2 0 1 3 4 5

Average Standard Deviation (K)

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- Ensemble forecasting is now a mature tool for medium-range forecasting
 - New development in the short-range
- Ensembles provide extra information on
 - Uncertainty

Risks, particularly for High Impact weather
We are learning how to use probability forecasts for improved decision-making

Useful Web Addresses



- MOGREPS Operational display system:
- http://www-nwp/~fren/MOGREPS/products/home.shtml
- Met Office ECMWF Ensemble display system (PREVIN): http://ukmet/OPER_PRODUCTS/ensembles/ensembles.html
- ECMWF Website forecasts: http://www.ecmwf.int/products/forecasts/d/charts
- Ensemble pages on external web: http://www.metoffice.com/research/nwp/ensemble/index.html
- NWP Gazette article on MOGREPS: http://www.metoffice.com/research/nwp/publications/nwp_gazette/feb06/mogreps

.html

- Lecture notes from ECMWF Predictability Training Course: http://www.ecmwf.int/services/training/rcourse_notes/general_circulation.html
- Slides from 2002 ECMWF Seminar on Predictability: http://www.ecmwf.int/publications/library/ecpublications/proceedings/seminar200 2_predictability/index.html
- Ken Mylne's page of www sites (includes links to ECMWF and NCEP pages): http://www-nwp/~frkm/Ensemble_sites.html
- Ken Mylne's home page: http://www-nwp/~frkm/index.html
- Any questions? Email: ken.mylne@metoffice.com

Accreditation







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Questions & Answers