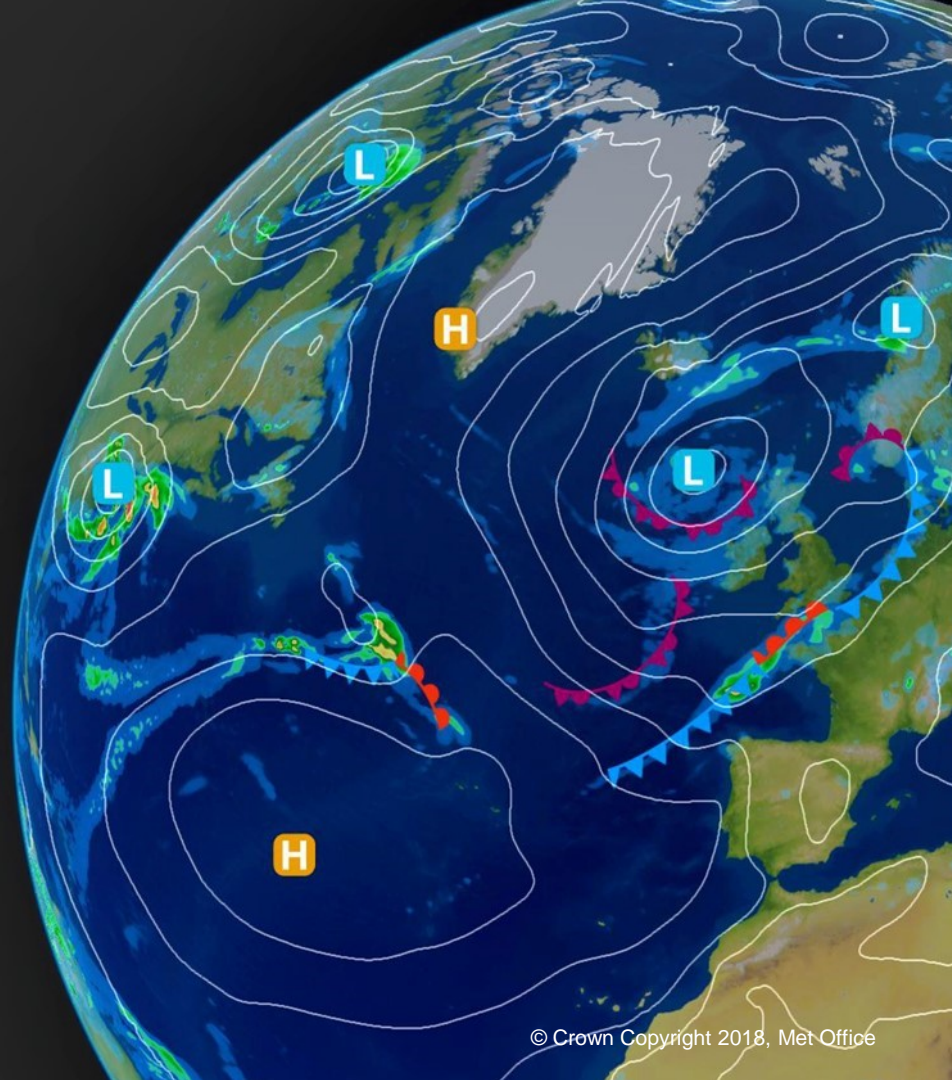


MOGREPS Ensemble Systems

Ken Mylne

Head of Verification, Impacts and Post-Processing, Weather Science

Met Office, Exeter



Outline

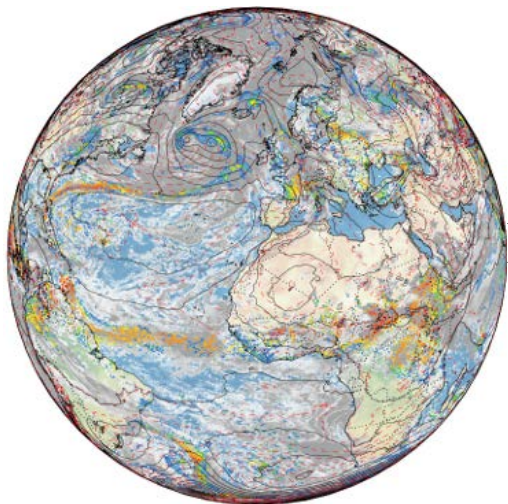
An integrated ensemble-based NWP system

- MOGREPS-G Global Ensemble
 - A few user application tools
- MOGREPS-UK Convective Scale Ensemble
- IMPROVER blended probabilistic post-processing
 - More info on posters

Deterministic models, global and UK, remain an important part of our NWP, but are heavily integrated with MOGREPS through hybrid 4D-Var, and are blended in IMPROVER.

Two model Strategy: Global and UK

Strategy defined ~10 years ago, almost complete



N x Global predictions
at ~20km with lead
times of days to weeks:
Synoptic drivers



<N x Regional predictions
at ~1km:
Local meteorology



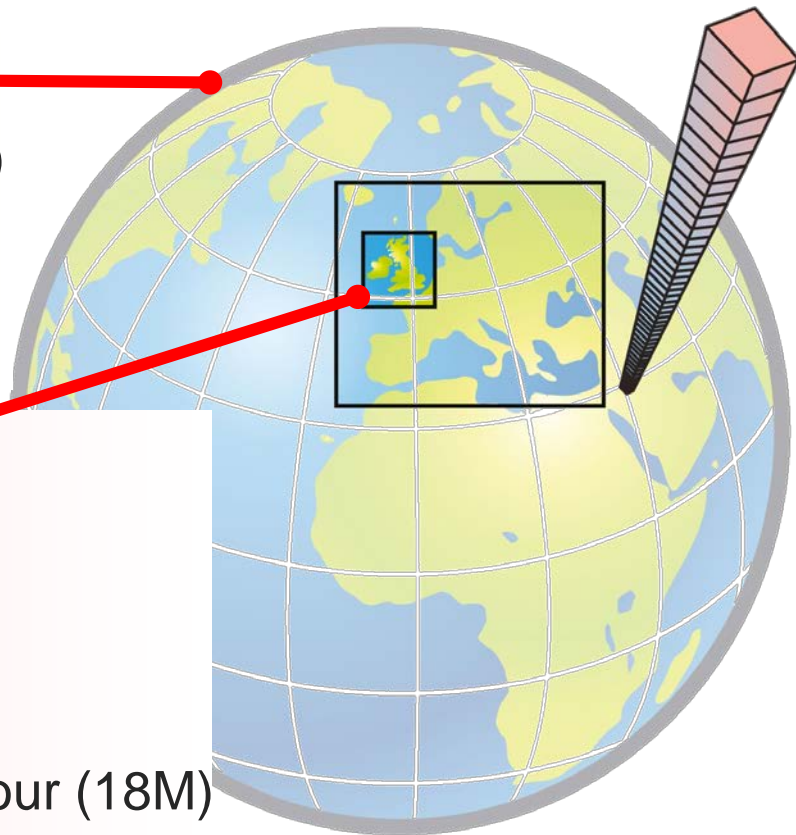
PDF of local
hazard:
Impacts

Global NWP:

- 10/20km deterministic/MOGREPS-G (36M)
- 70 vertical levels (80km top)
- Hybrid 4D Variational Data Assimilation
- Forecasts to T+48 or T+192hr every 6 hrs

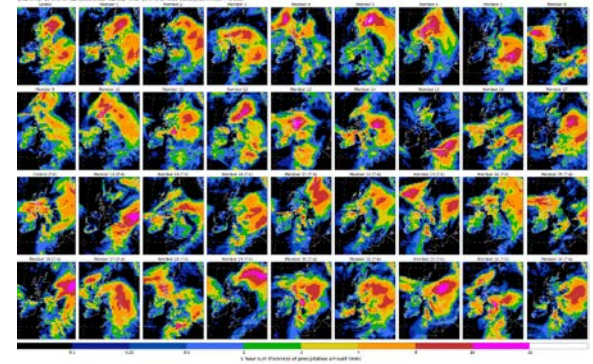
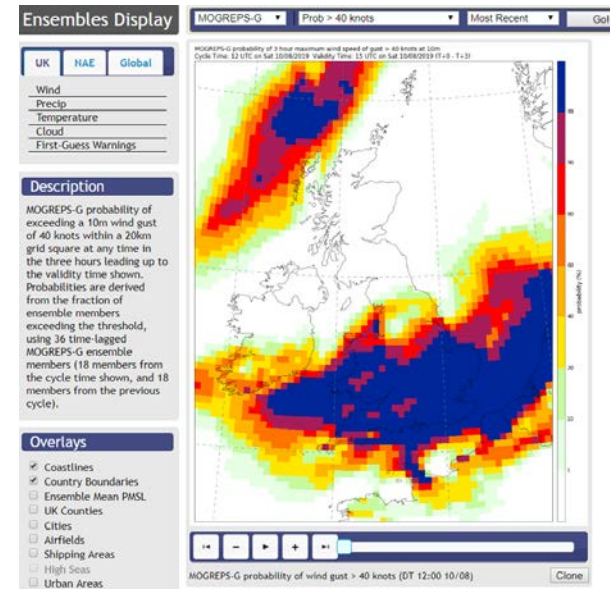
UK NWP:

- 1.5/2.2km deterministic/MOGREPS-UK
- 70 vertical levels (40km top)
- Hourly 4DVar Analysis and 12h forecast
- Forecasts to T+54h every 3hr
- MOGREPS-UK to T+120 updated every hour (18M)



Brief description MOGREPS-G

- 36-member time-lagged Global ensemble with UM at ~20km grid resolution
 - 18 members run every 6 hours with last two cycles combined to provide 36 member forecasts
 - Forecasts to 7 days
- Initial perturbations by ETKF with localisation
- Stochastic physics by SKEB and SPPT (details to follow)
- Wide-ranging post-processing applications...

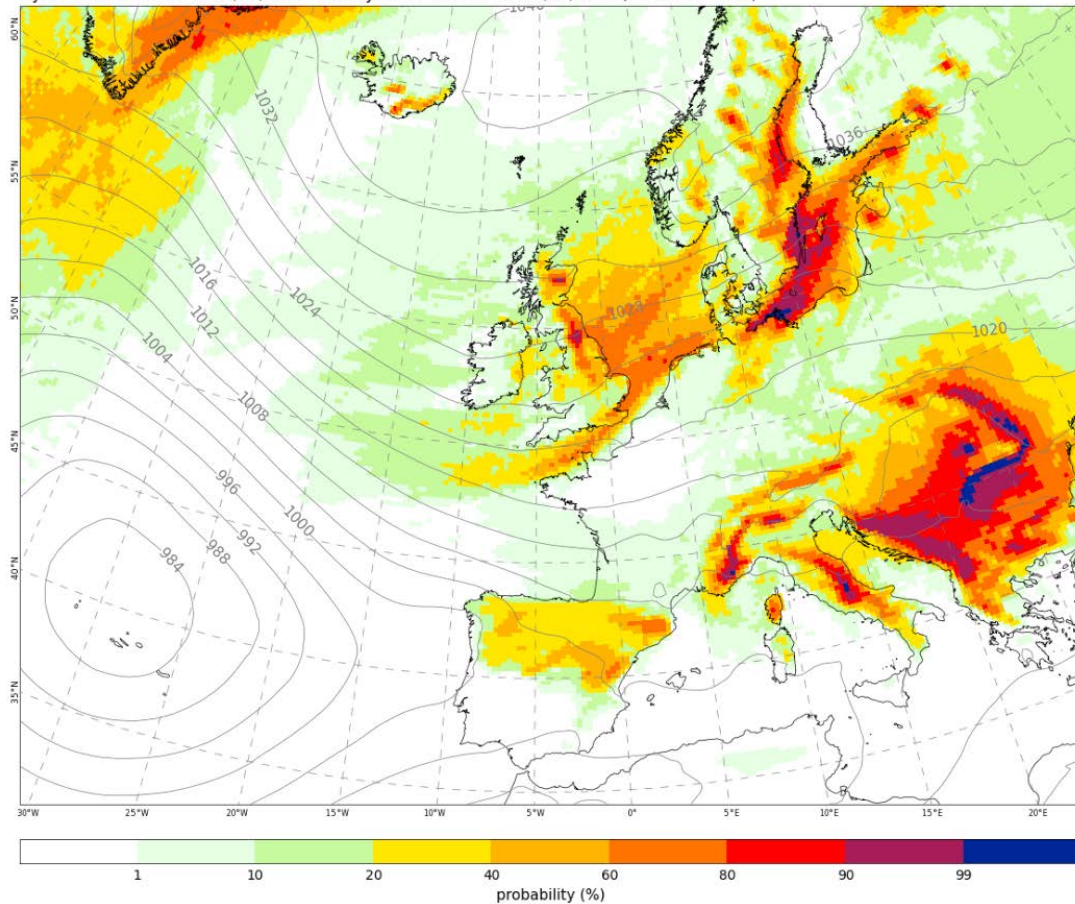


Major Snow event, Mar 2018

Probability 24 hour snowfall > 1cm

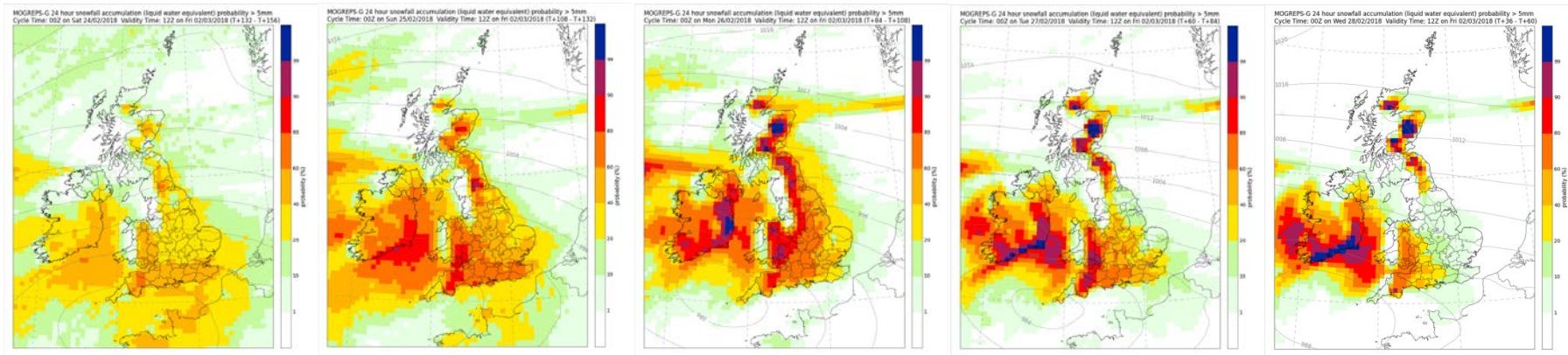
High probabilities showing up over the North Sea and eastern side of the country at a **6 to 7 day lead time**, signalling the first snow of this cold spell.

MGREPS-G 24 hour snowfall accumulation (liquid water equivalent) probability > 1mm
Cycle Time: 06Z on Tue 20/02/2018 Validity Time: 06Z on Tue 27/02/2018 (T+144 - T+168)



MOGREPS-G probability forecast evolution

Probability 24 hour snowfall > 5 cm, valid 12:00 Thu 1st Mar to 12:00 Fri 2nd March 2018



6 day forecast

5 day forecast

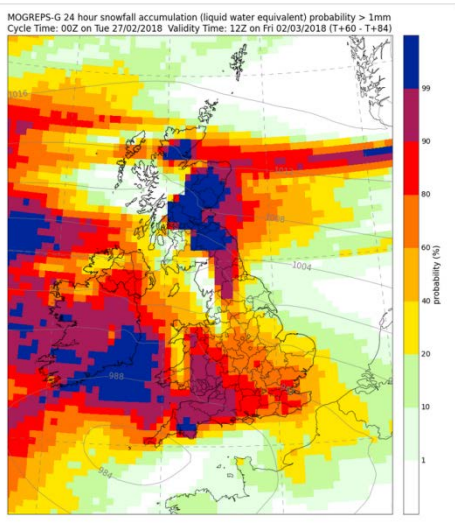
4 day forecast

3 day forecast

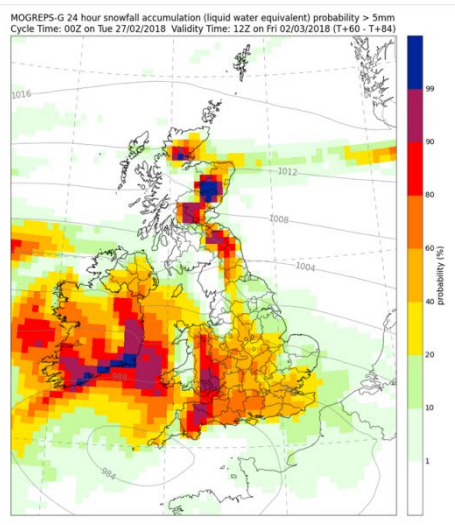
2 day forecast

MOGREPS-G probability forecasts for different snow amounts (SW snow event)

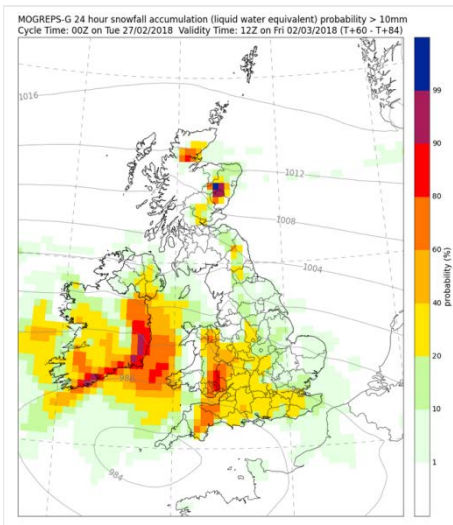
Valid 12:00 Thu 1st Mar to 12:00 Fri 2nd March 2018 – 3 day lead time



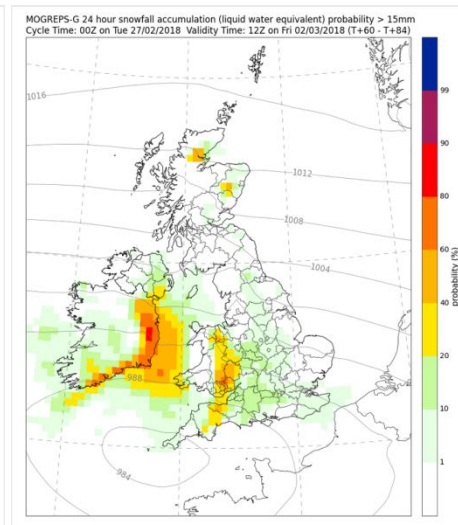
Probability > 1 cm



Probability > 5 cm



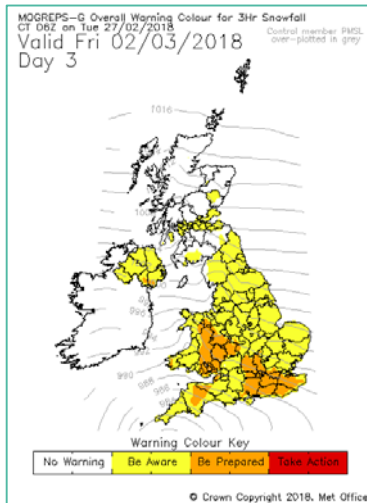
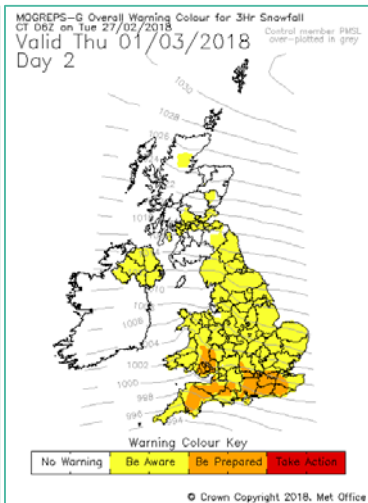
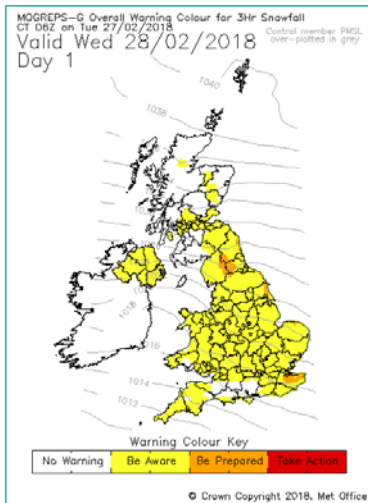
Probability > 10 cm



Probability > 15 cm

Viewing probabilities for several snowfall thresholds helps pinpoint the areas most at risk

MOGREPS-G first guess warnings for snow

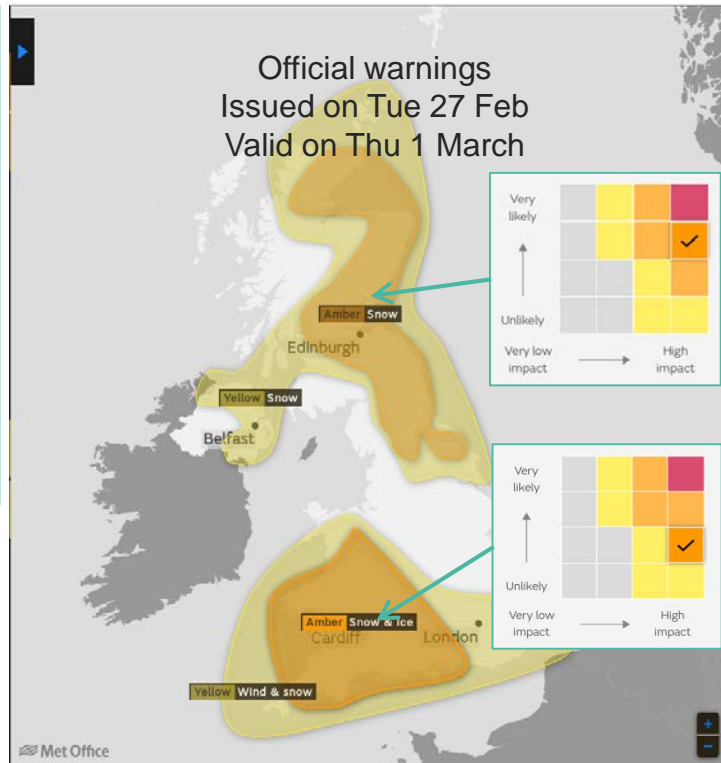


MOGREPS-G first-guess warnings

Issued on Tue 27 Feb and valid on

Wed 28th Feb (left), Thu 1st March (middle) and Fri 2nd March (right)

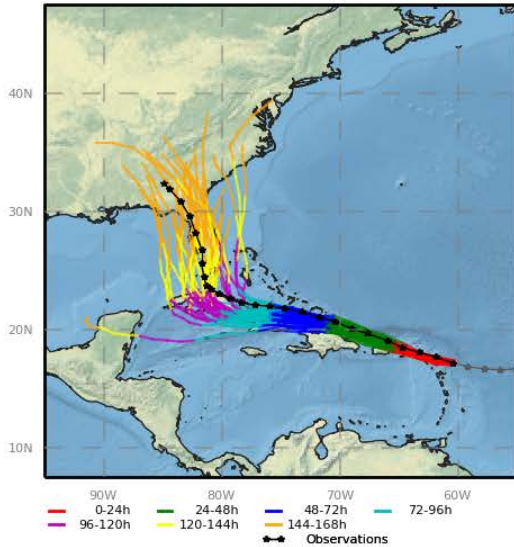
MOGREPS-G is not resolving the heaviest snow showers in the NE, but is capturing the frontal snowfall in the south much better. The UKV (1.5 km) and MOGREPS-UK (2.2 km) are able to help with the showers.



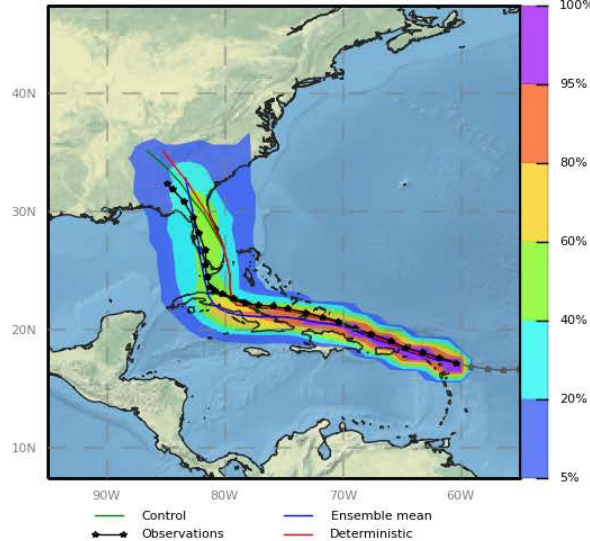
Tropical cyclone tracking and storm-following meteograms



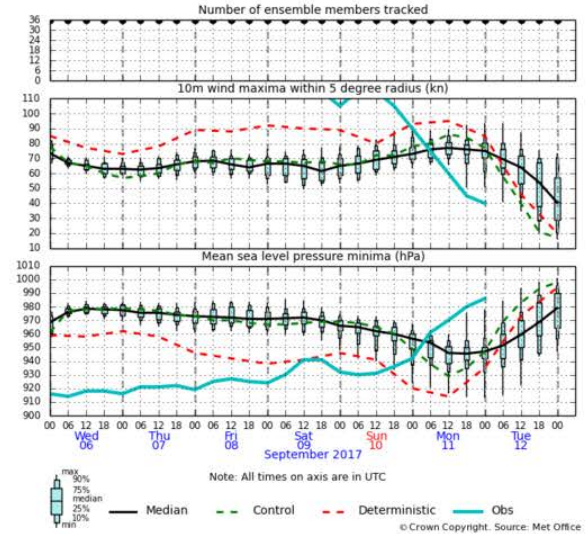
MOGREPS-G: Forecast tropical storm tracks for IRMA from 00UTC 06/09/2017



MOGREPS-G: Forecast tropical storm strike probability for IRMA from 00UTC 06/09/2017



MOGREPS-G: Tropical Cyclone storm-following meteogram IRMA (17.2N 60.4W) from 00UTC 06 September 2017



Top = 5 day forecast for 4.4 km

Bottom = 7 day MORGREPS-G forecast (18 not 36 members)

5v7 day forecast complicates direct comparison of plots.

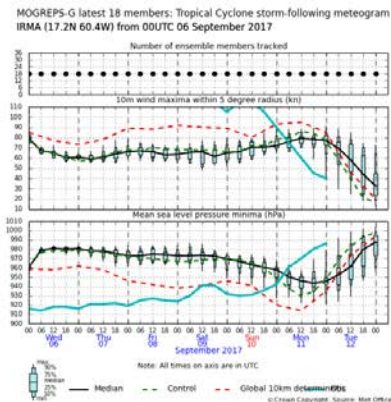
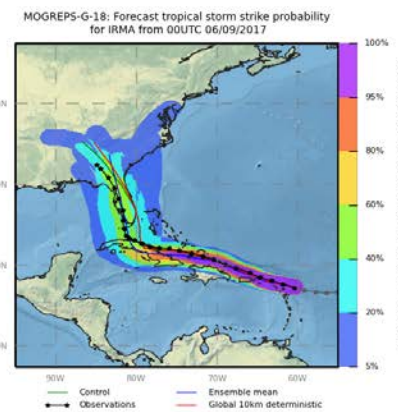
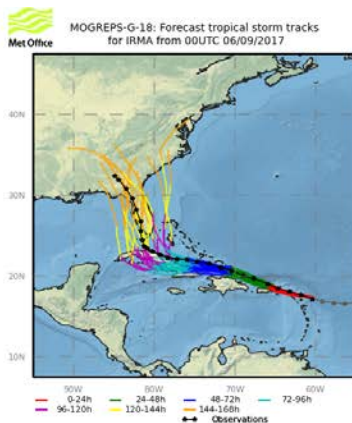
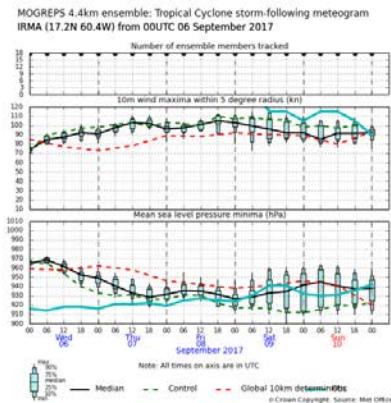
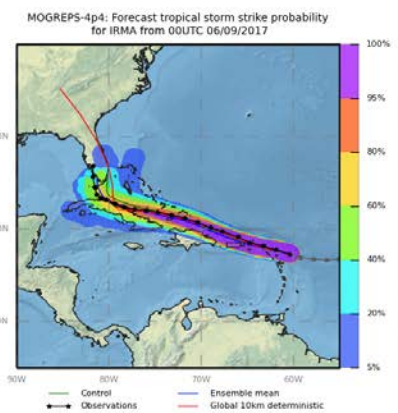
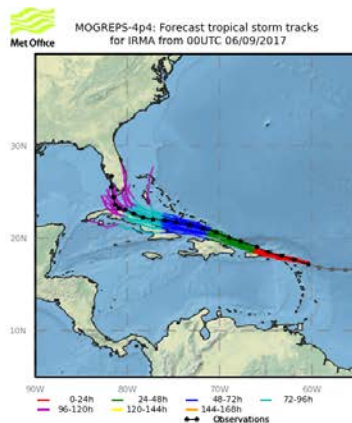
Actual track is overlaid in black

Tracks broadly similar

Propagation speed better in 4.4 km ensemble (compare location of purple lines)

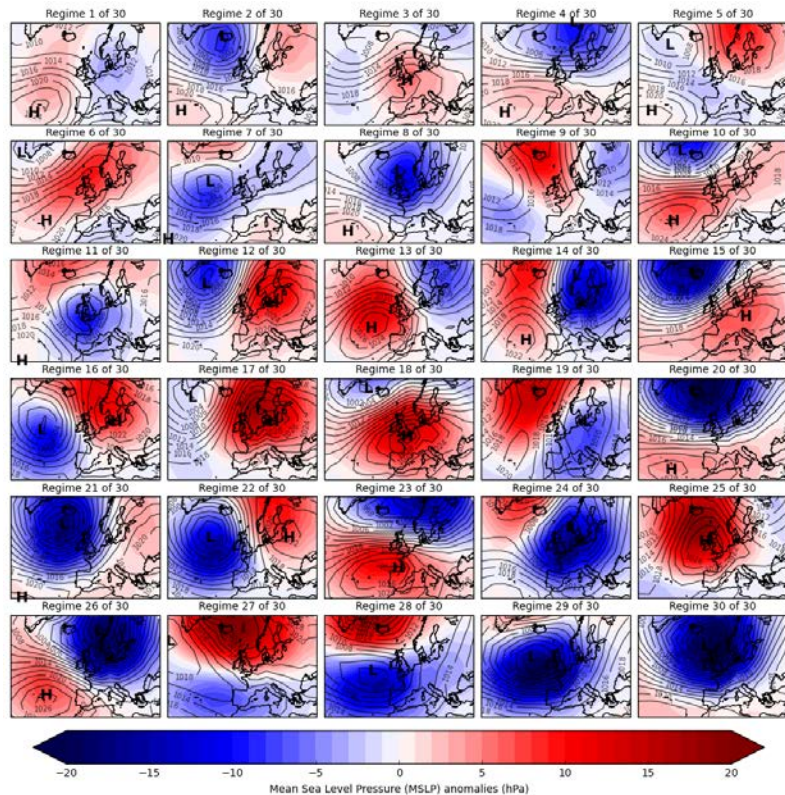
PMSL intensities in 4.4 km ensemble encompass observed value after T+48.

“Usual” peak 10 m wind underestimation



Seamless ensemble clustering:

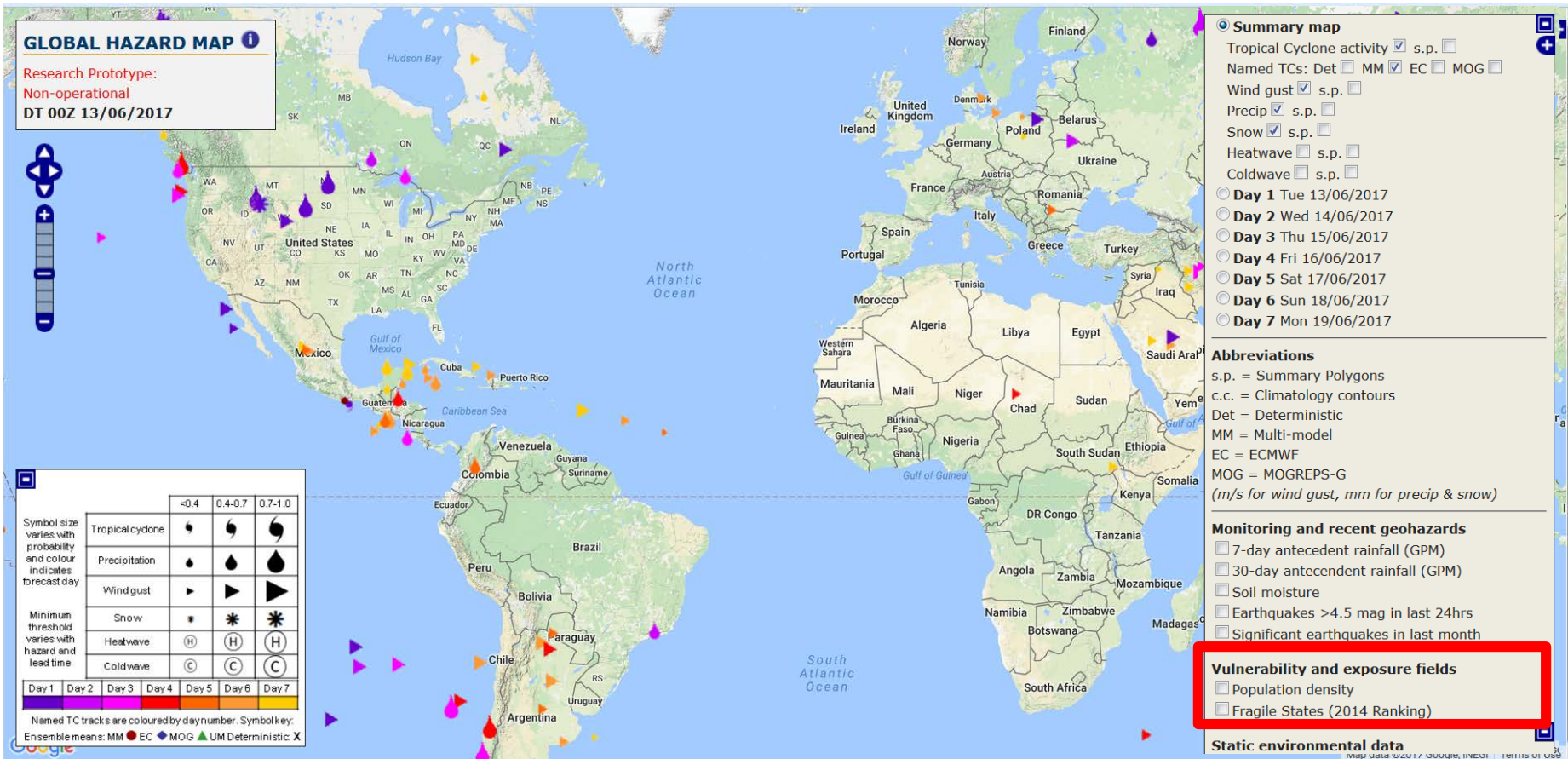
Monthly with Glosea; Medium-Range using MOGREPS-G and ECMWF ENS



	Wed 2 Dec	Thu 3 Dec	Fri 4 Dec	Sat 5 Dec	Sun 6 Dec	Mon 7 Dec	Tue 8 Dec	Wed 9 Dec	Thu 10 Dec	Fri 11 Dec	Sat 12 Dec	Sun 13 Dec	Mon 14 Dec	Tue 15 Dec	Wed 16 Dec	Regime Descriptions (UK)	Historic Occurrence N/D/J
Regime 1											2			2		Unbiased NWly	2.0%
Regime 2											4	4	2			Cyclonic W-SWly, returning Pm airmass	2.8%
Regime 3								2	2		4		4			Anticyclonic SWly, ridge over N France	2.3%
Regime 4									2	2		6		2	4	Unbiased Wly	2.7%
Regime 5																Unbiased S-SEly, high over Scandinavia	2.7%
Regime 6																Anticyclonic, Azores high ext.	2.8%
Regime 7												2	6	4	4	Cyclonic SWly, low WNW of Ireland	2.2%
Regime 8										2	4	6	4	2	2	Cyclonic W-NWly, low near Shetland	3.1%
Regime 9										2	4	4				Anticyclonic N-NEly, high near Iceland	2.6%
Regime 10													2		2	Anticyclonic W-SWly, slight Azores ridge	3.4%
Regime 11													4	2		Cyclonic, low centred over southern UK	2.4%
Regime 12														2	2	Anticyclonic Sly, high over Poland	4.2%
Regime 13							2			6	16	10	6	8	2	Anticyclonic NWly, high SW of Ireland	4.4%
Regime 14											10	12	12	8	6	Cyclonic N-NWly, low near S Sweden	4.0%
Regime 15	100	51	84		2	18			6	12	4	6	6	8	4	Unbiased SWly, very windy NW Britain	4.6%
Regime 16												2	2	2	2	Anticyclonic S-SEly, high E of Denmark	2.7%
Regime 17									4	6	2	4			2	Anticyclonic E-SEly high over Denmark	4.3%
Regime 18							6	18	55	39	25	12	10		6	Anticyclonic SWly, high over N France	4.7%
Regime 19														4	4	Cyclonic Nly, low E of Denmark	4.0%
Regime 20			10	100	76	31			2	2		6	8	8	4	Cyclonic Wly, intense low near Iceland	4.1%
Regime 21										2	4	2	6	14	12	Cyclonic SWly, deep low S of Iceland	3.8%
Regime 22												2			2	Cyclonic Sly, low W of Ireland	3.2%
Regime 23		49	6		18	41	30	25	27	31	20	12	6	2	4	Unbiased Nly, windy in N	4.1%
Regime 24												2	12	14	14	Cyclonic Nly, low in N Sea	3.2%
Regime 25								6		6	6	6	2		4	Anticyclonic Nly, high centre Irish Sea	3.6%
Regime 26					2	4				6	8	8	2	6	14	Cyclonic NWly, low near Norway, windy	3.4%
Regime 27																Anticyclonic Ely, high in Norwegian Sea	3.7%
Regime 28														2		Cyclonic SEly, low SW of UK	2.8%
Regime 29												2	2	4	2	Cyclonic S-SWly, deep low W of Ireland	2.9%
Regime 30					2						2	2	8	12	4	Cyclonic W-SWly, deep low SE of Iceland	3.0%
Total Members	51	51	51	51	51	51	51	51	51	51	51	51	51	51	51	---	---

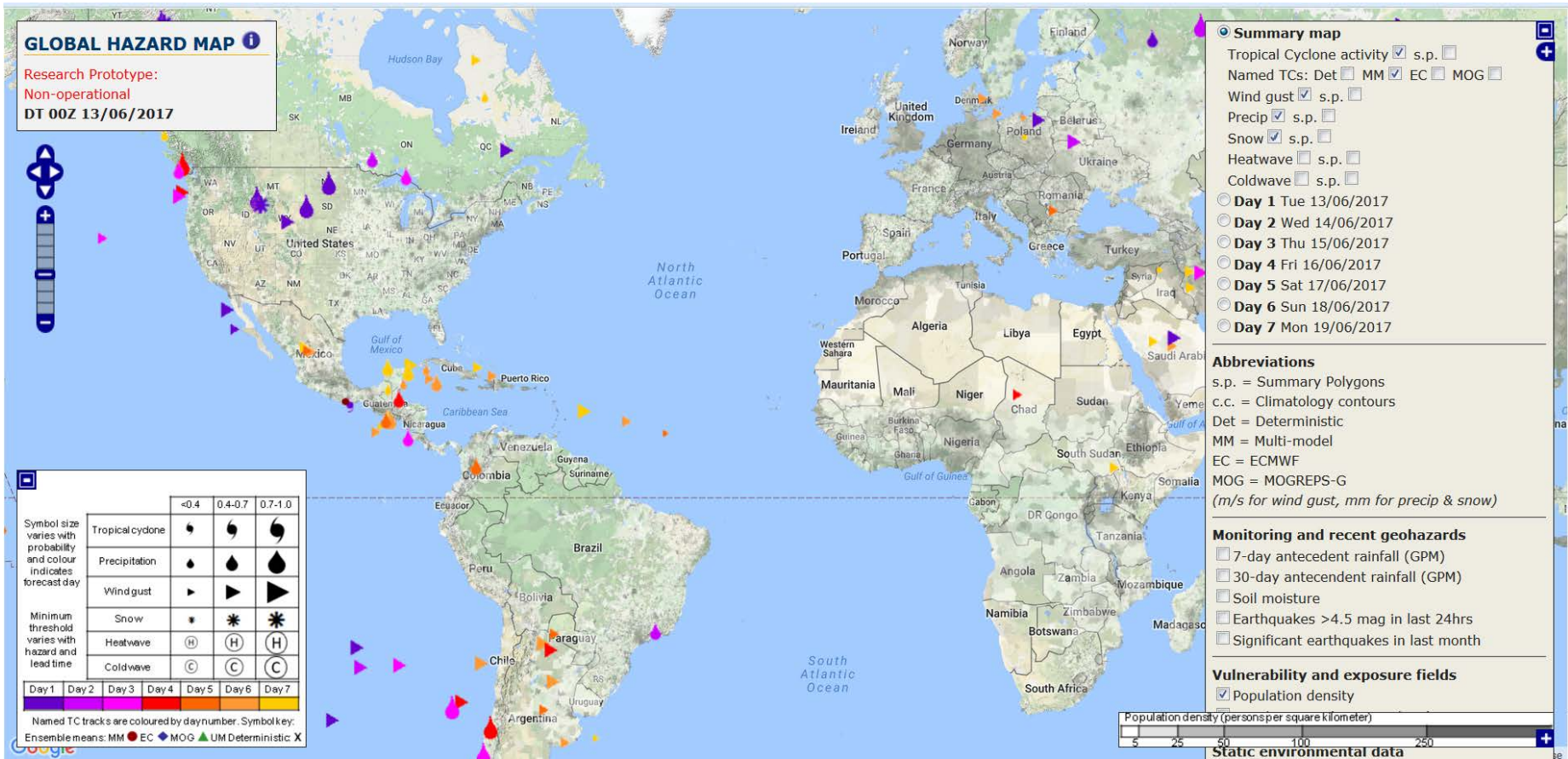
Global Hazard Map

Ensemble forecasts of extreme hazard probabilities ...



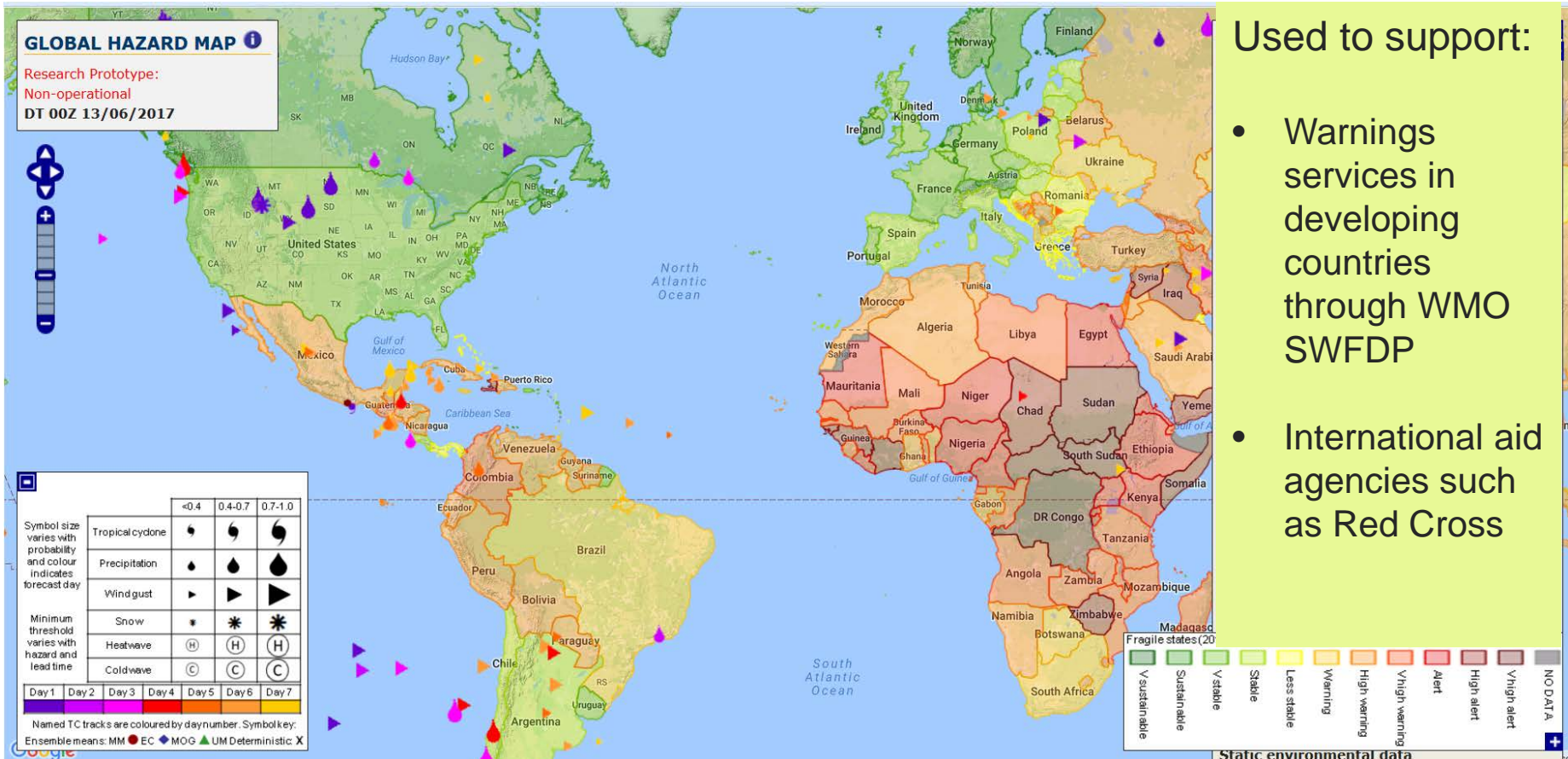
Global Hazard Map

Ensemble hazard probabilities overlaid against global scale vulnerability datasets – population density



Global Hazard Map

Ensemble hazard probabilities overlaid against global scale vulnerability datasets - fragile states



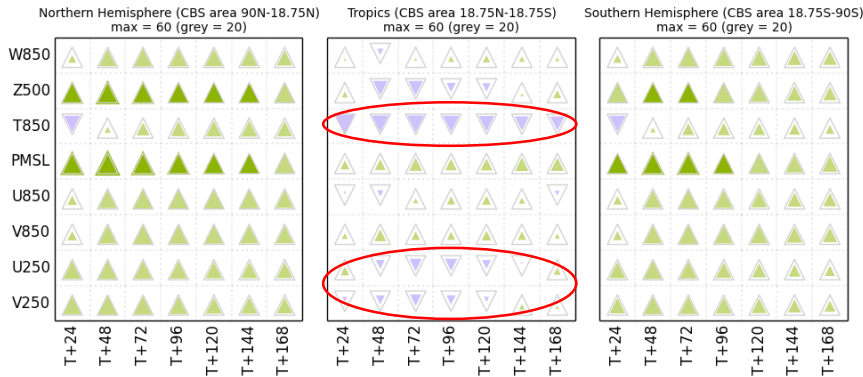
Ensemble comparison with other centres

Data source: Japan Meteorological Agency – JMA – website <http://epsv.kishou.go.jp/EPsV>

Scorecards vs individual centres

CRPS scorecard MOGREPS-G vs JMA Oct 2018. Up-green (down-purple) triangles show better (worse) performance in MOGREPS-G.

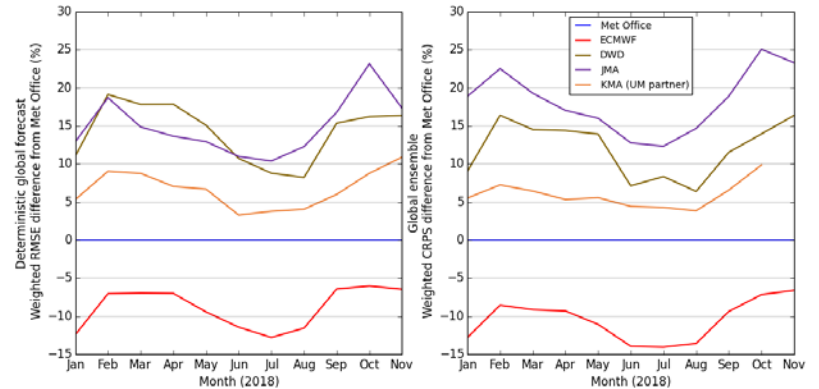
- Weaker tropical 250hPa winds improved in PS41, but 850hPa temperature bias degraded



Index time-series vs all centres

High-res deterministic RMSE Index (left) and ensemble CRPS Index (right) time-series with Met Office as zero (blue) line.

- Ensemble performance close to deterministic, but also highlights specific features linked to the ensemble

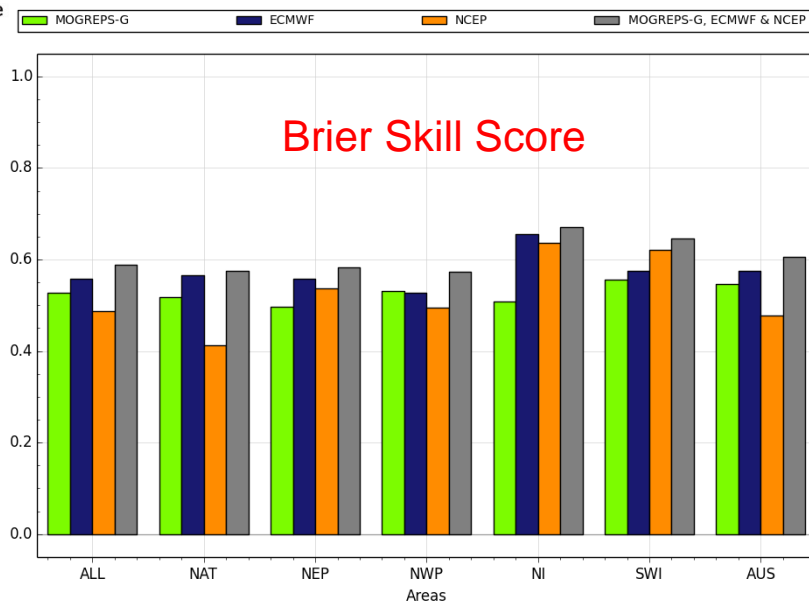


Tropical Cyclone 7-day strike probability forecasts

88 named storms in Jul 2017 – Jun 2018



Tropical Cyclone Probability July 2017 - June 2018: Brier Skill Score



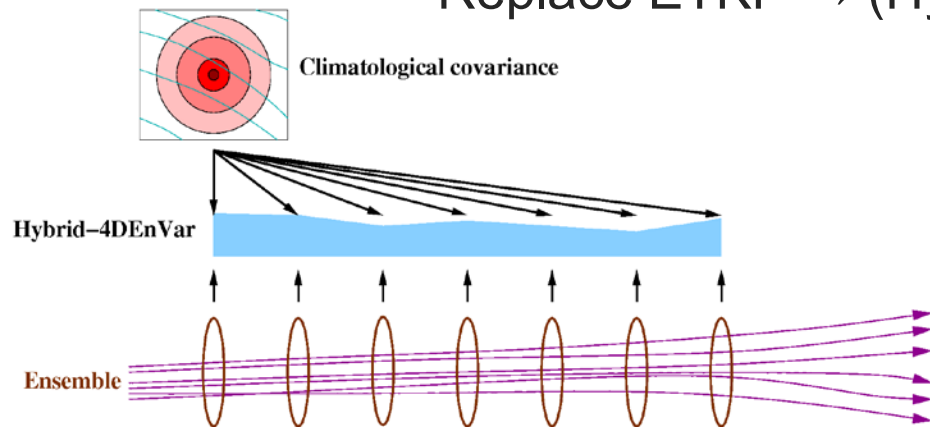
Rebecca Bowyer (nee Stretton)

© Crown Copyright 2018. Source: Met Office

- No single ensemble system is best in all areas for strike probability
- MOGREPS-G in 2nd place overall
- Multi-model system always best for this metric

PS43 Ensemble upgrade (Sep 2019)

Replace ETKF → (Hybrid-)En-4DEnVar

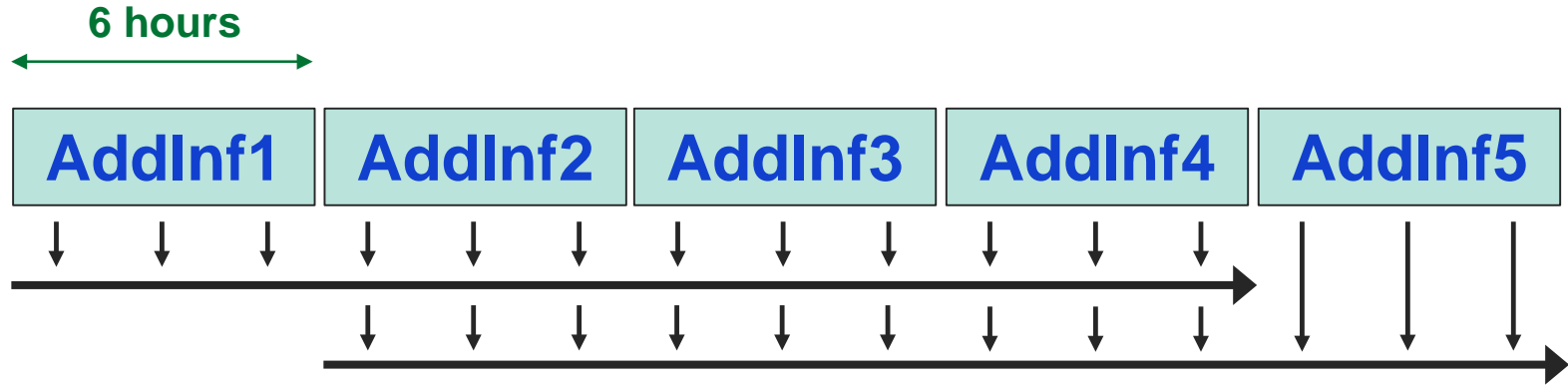


- 4DEnVar does not need a linear PF model (simplifies code maintenance) because it uses ensemble information throughout assimilation window
- Still uses climatological background error covariances (hybrid scheme)
- The 4DEnVar executable can also do an **ensemble of analyses: En-4DEnVar:**
 - (For each member, create increments relative to its own background trajectory)
- Routines available to deal with inflation, perturbed observations, etc.

Stochastic Physics and Model Error schemes

Scheme	Complexity/Cost	Maintenance	Effectiveness
SKEB (Stochastic Kinetic Energy Backscatter)	<ul style="list-style-type: none"> Theory not fully robust Widely used, but in decline Moderate cost 	Well established in UM, no significant cost	++
SPT (or SPPT) (Stochastic Perturbation of Tendencies) <i>(Impl. PS41 in 2018)</i>	<ul style="list-style-type: none"> Widely used Simple and flexible Moderate cost 	Part of GA7 (as with SKEB), stable performance	++
Additive Inflation (using DA Incr archive) <i>(Impl. PS43 in 2019)</i>	<ul style="list-style-type: none"> New idea, less tested Code lacks documentation Large offline IO & storage 	Needs regular updating of operational Analysis Increment archive	+++
Perturbed parameters <i>(Retired in PS41 2018)</i>	<ul style="list-style-type: none"> Very simple Minimal cost Can be expanded to 2D 	Needs to keep track with physics developments. Magnitude and correlation of perturbations complex.	+

Additive inflation



Create a year-long archive of data assimilation increments, then at each time-step during the model forecast add:

1. a randomly selected historical analysis increment (with 50% scaling) per six-hour period, retaining the selection for the same validity time in subsequent forecast cycles
2. a three-month mean increment, as a bias correction (with no scaling)

Note: the random selection and mean are both from the same three-month period, corresponding to the day/month of the forecast to account for seasonal dependencies

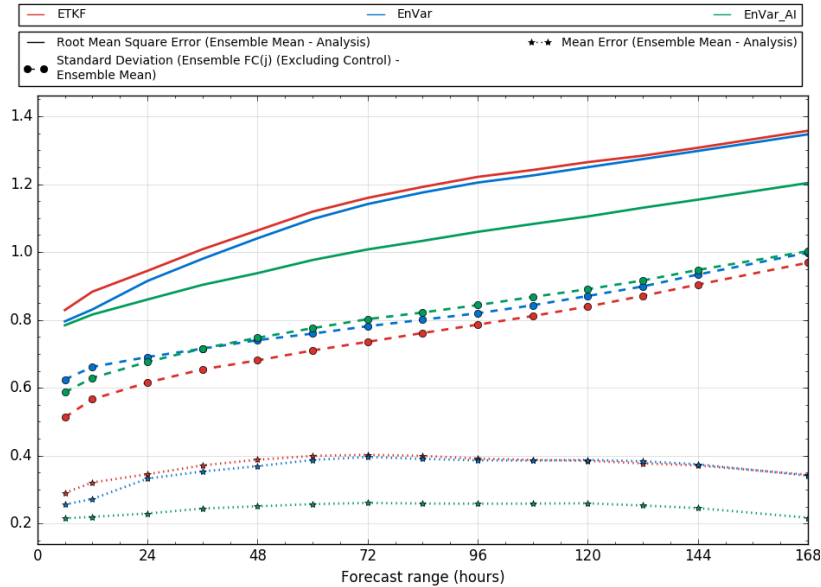
En-4dEnVar (early low-resolution results)

Improves spread, bias and error

Period: 1 Sep – 15 Oct 2018; Truth = ECMWF Analyses

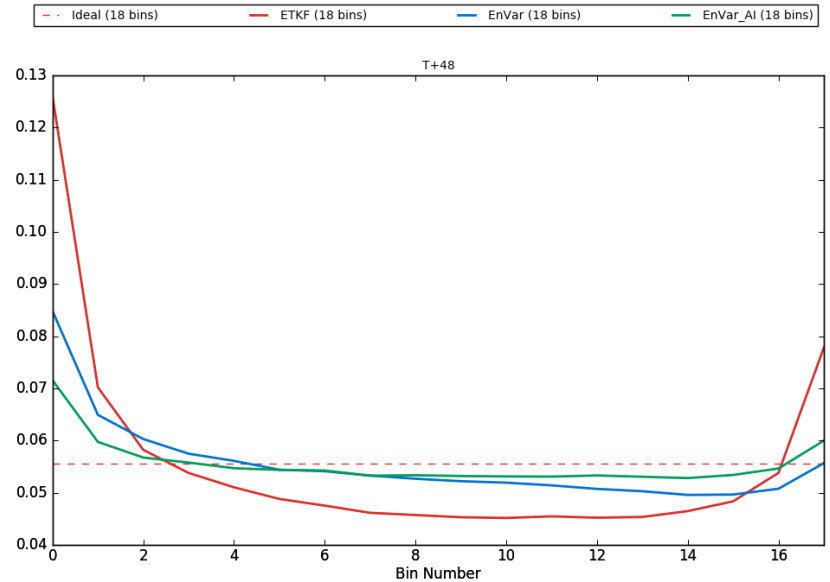
RMSE, spread, bias – T850 (tropics)

- AI improves bias and RMSE



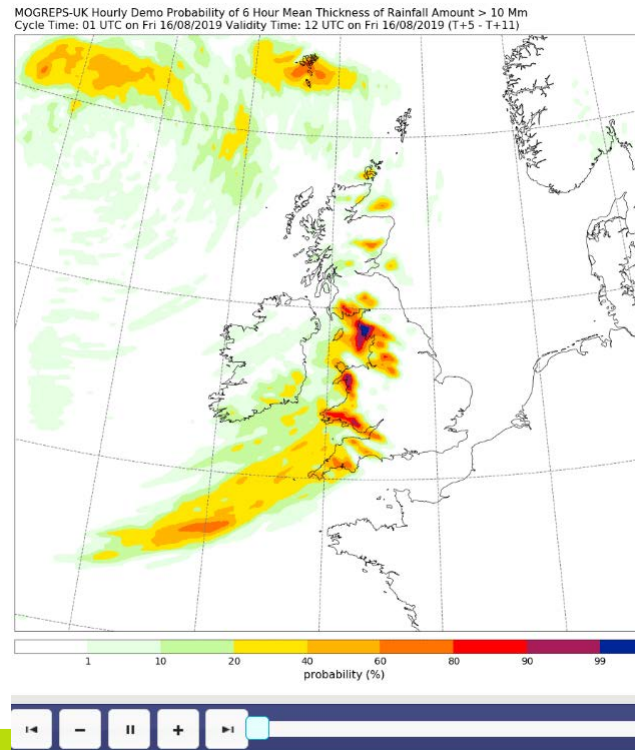
Rank histogram (T+48h) – T500 (NH)

- En-4DEnVar improves under-dispersion and AI improves bias



Brief description MOGREPS-UK

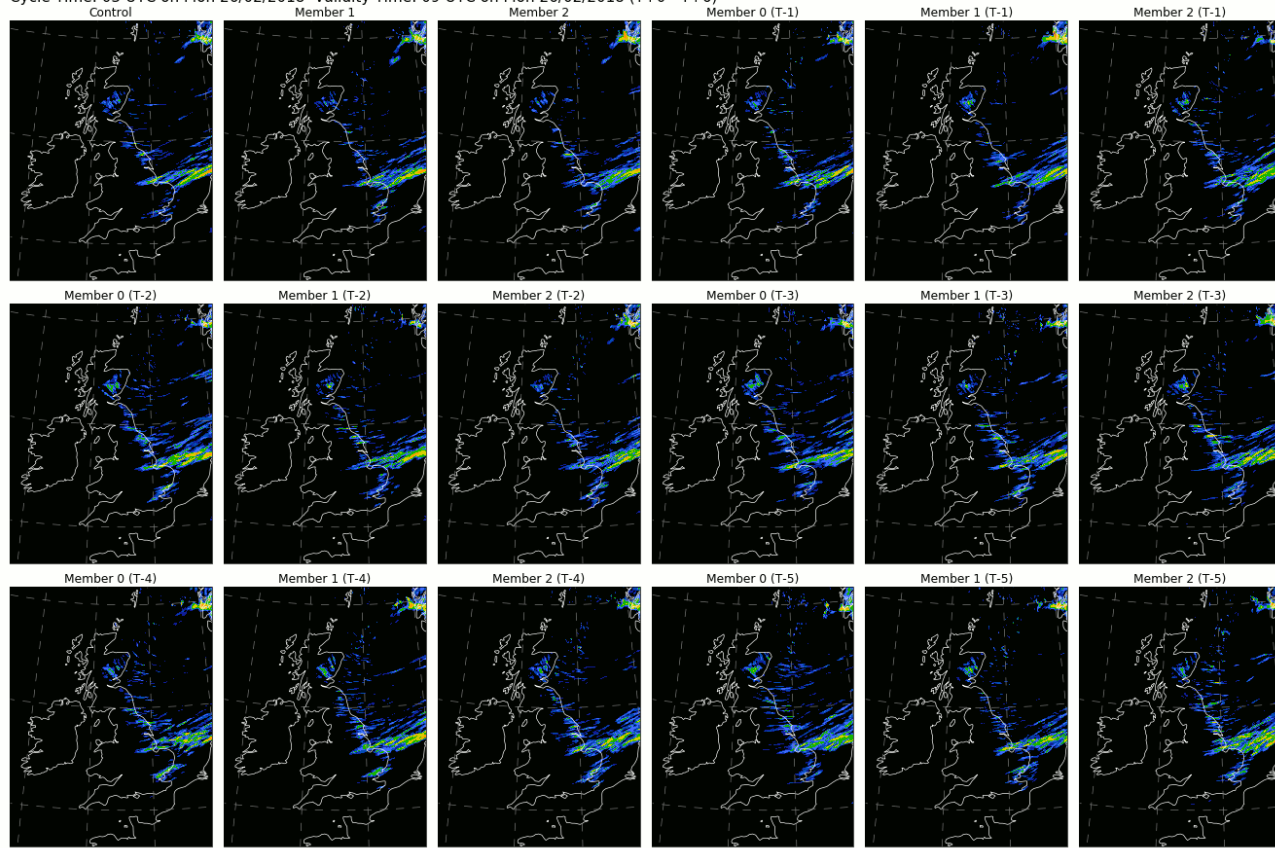
- 18-member time-lagged UK ensemble with UM at ~2.2km grid resolution
 - 3 members run every hour with last 6 cycles combined to provide 18 member forecasts
- Forecasts to **5 days**
- Downscaled global perturbations from corresponding MOGREPS-G members added to latest hourly 4D-Var UK analysis
- Stochastic physics by SKEB and SPPT



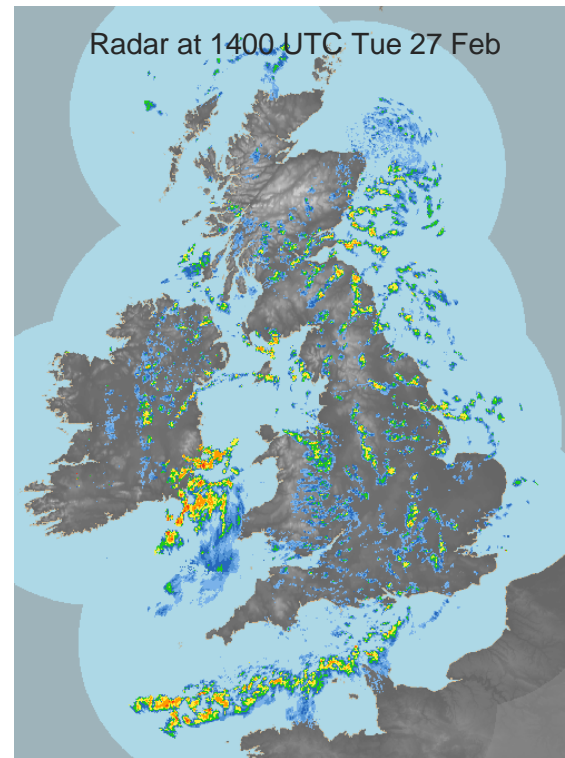
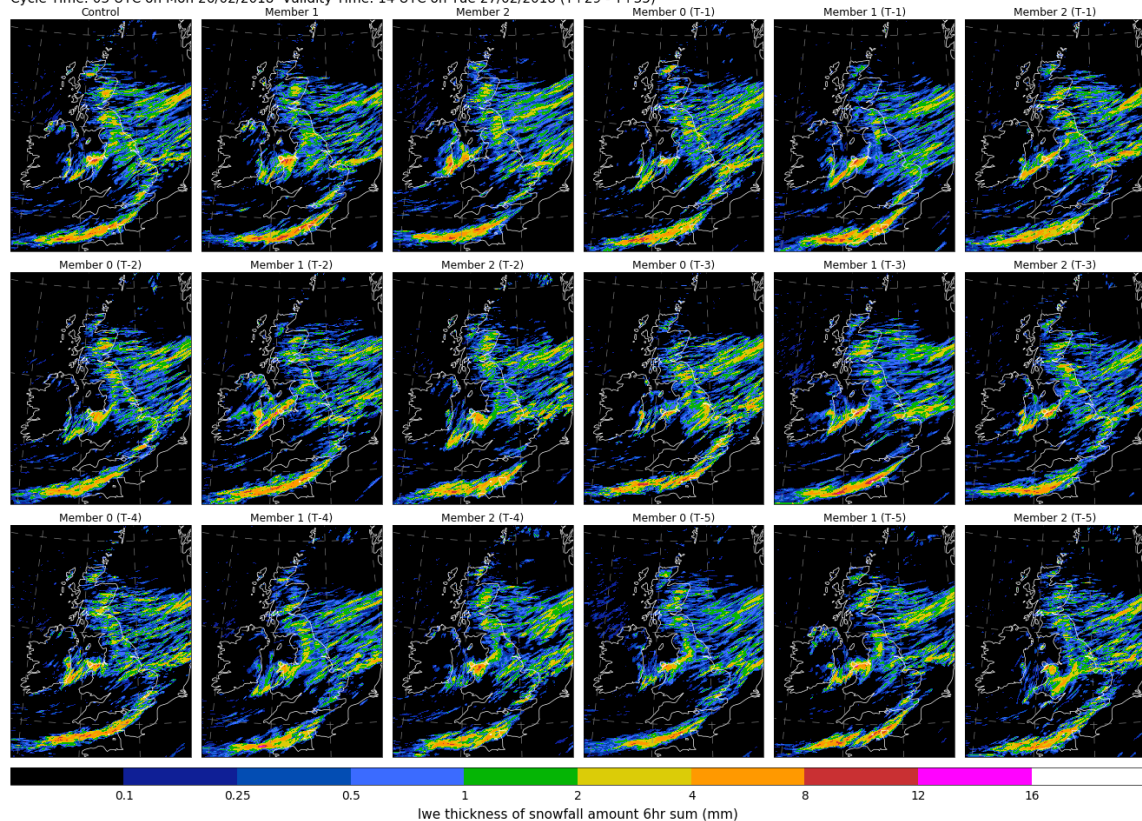
MOGREPS-UK snowfall postage stamps animation

MOGREPS-UK Iwe thickness of snowfall amount 6hr sum

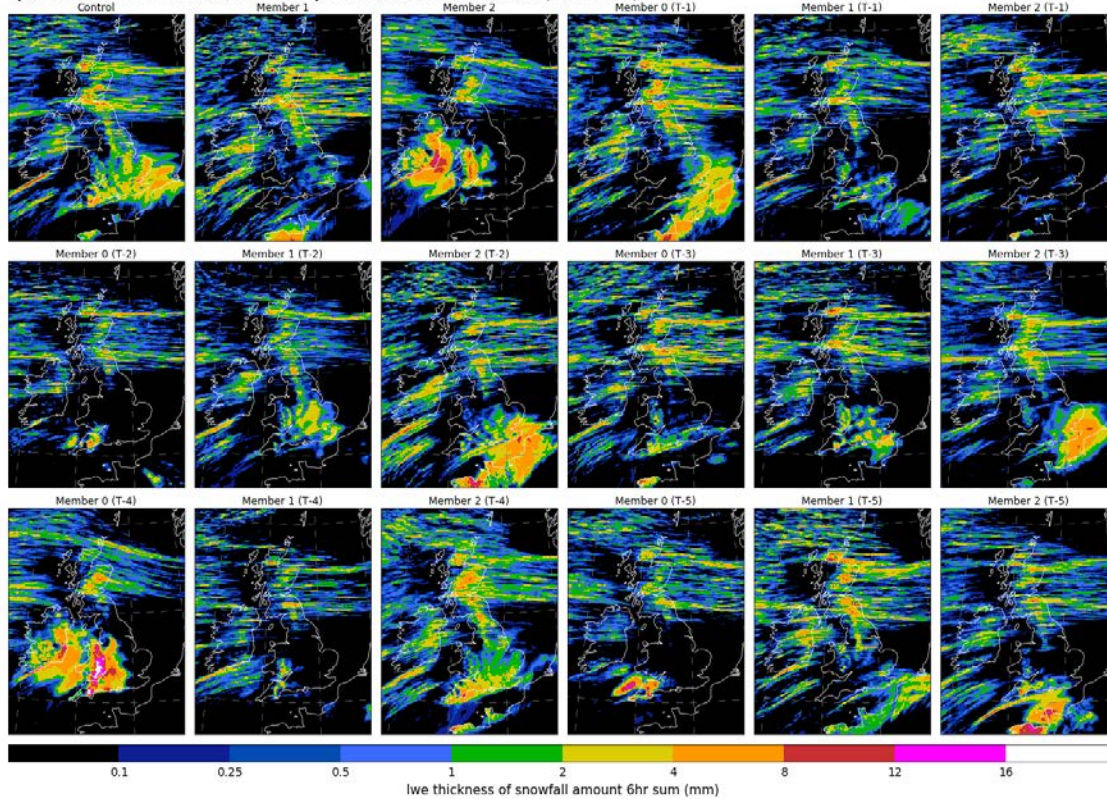
Cycle Time: 03 UTC on Mon 26/02/2018 Validity Time: 09 UTC on Mon 26/02/2018 (T+0 - T+6)



MOGREPS-UK lwe thickness of snowfall amount 6hr sum
Cycle Time: 03 UTC on Mon 26/02/2018 Validity Time: 14 UTC on Tue 27/02/2018 (T+29 - T+35)



MOGREPS-UK lwe thickness of snowfall amount 6hr sum
Cycle Time: 03 UTC on Mon 26/02/2018 Validity Time: 18 UTC on Thu 01/03/2018 (T+81 - T+87)



T+81 to 87
Valid Thu 1st March

Large spread in frontal
snowfall positioning in the
south and south west.

Lots of snow shower activity
in the NE.

The Big Data Challenge: huge volumes and rapid updates

Forecast systems throwing out vast quantities of data for forecasters and users to assimilate:

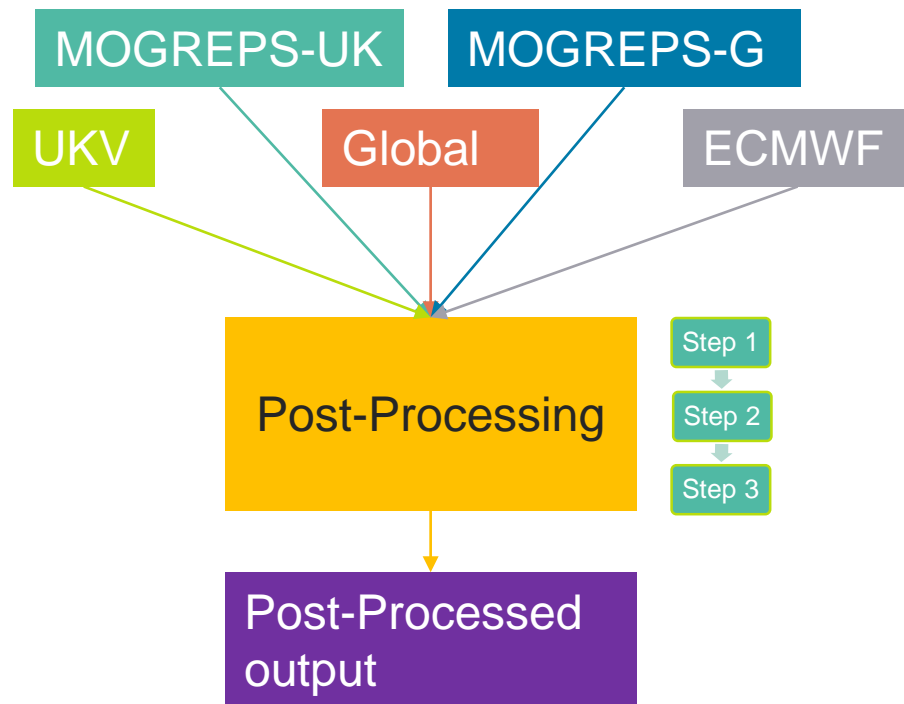
- Global models: 690 Gbytes per day
- UK models: 3200 Gbytes per day
- **TOTAL: 3.9 Tbytes per day**

Forecasts updated every hour or more –
Drinking from a Firehose of data!



Blended Probabilistic Post-Processing

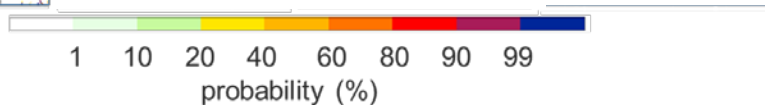
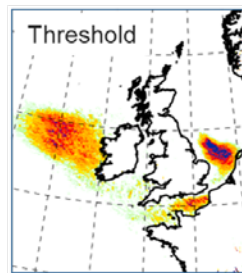
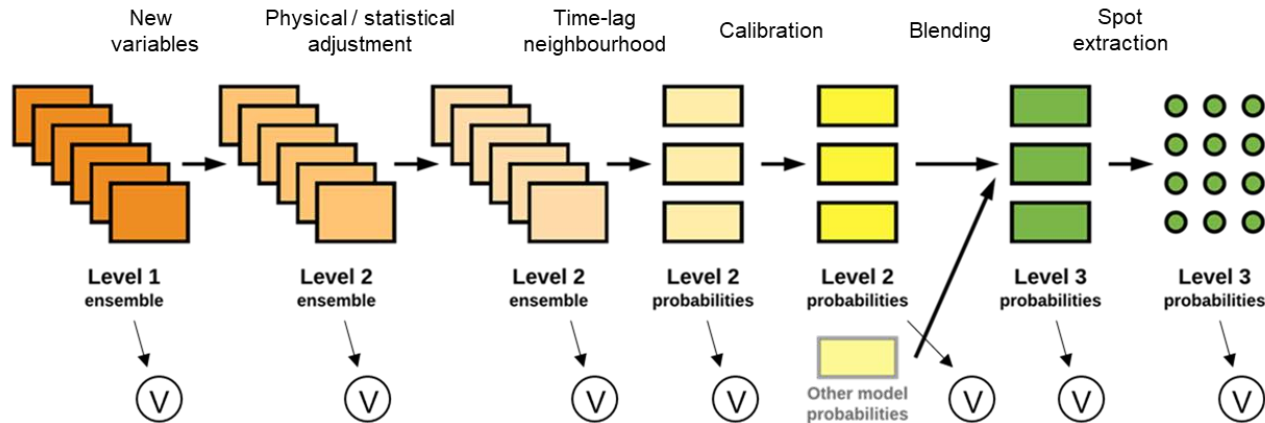
- Multiple forecasts from different models
 - Difficult for a user or operational meteorologist to keep track
- Single automated best-estimate forecast, blending the most appropriate model forecasts for any place and forecast time



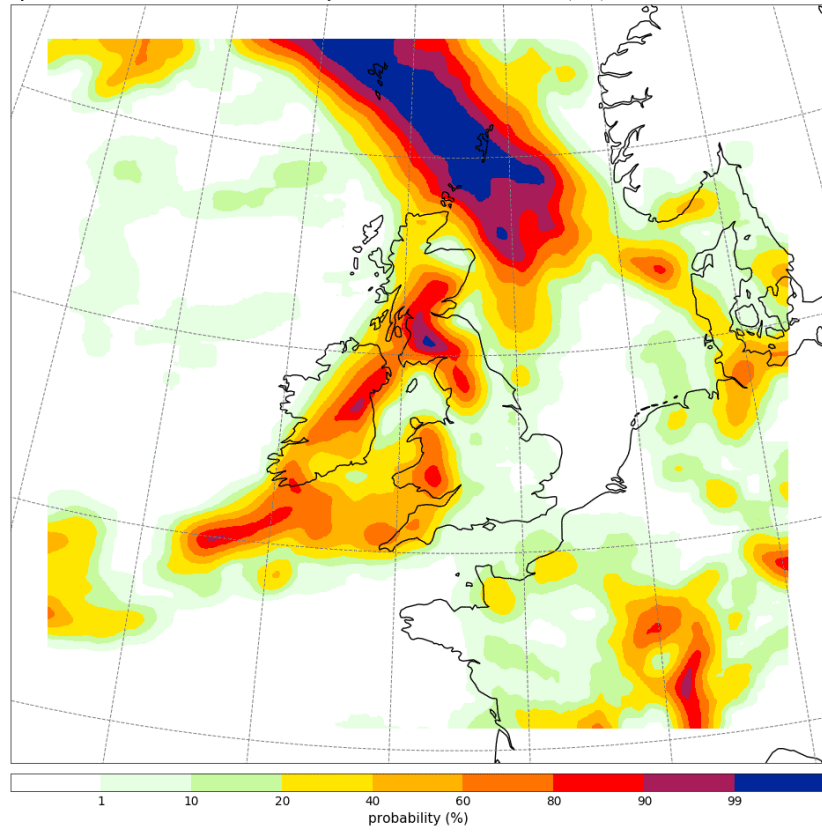
IMPROVER chains



temperature, rain,
snow, cloud, visibility,
wind ...

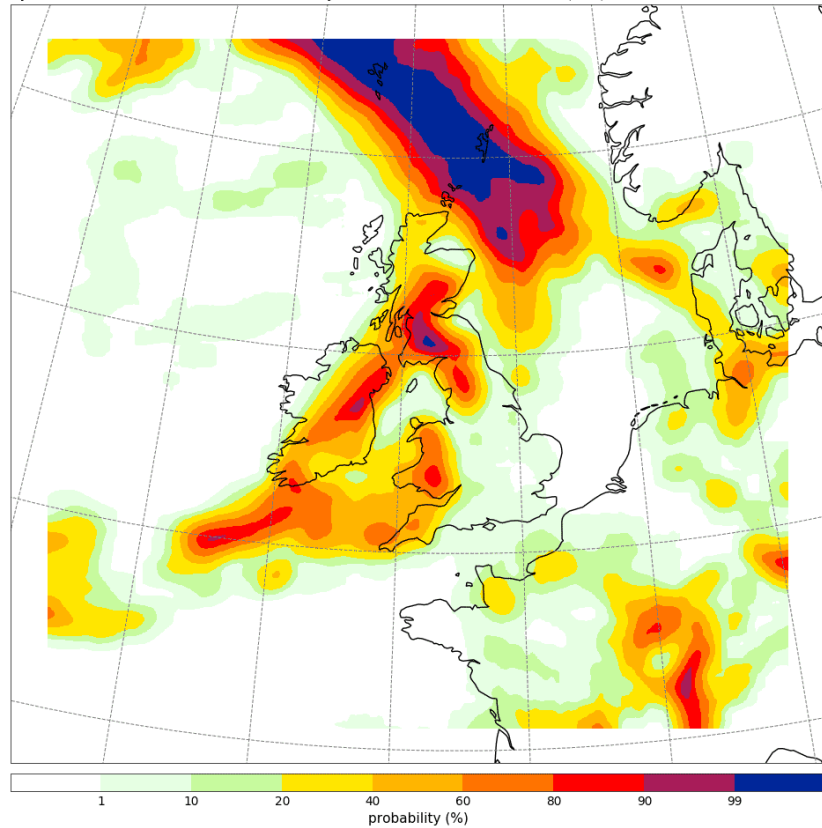


Unknown Model probability_of_rainfall_rate > 0.03 mm/hr
Cycle Time: 11 UTC on Thu 06/12/2018 Validity Time: 11 UTC on Thu 06/12/2018 (T+0)

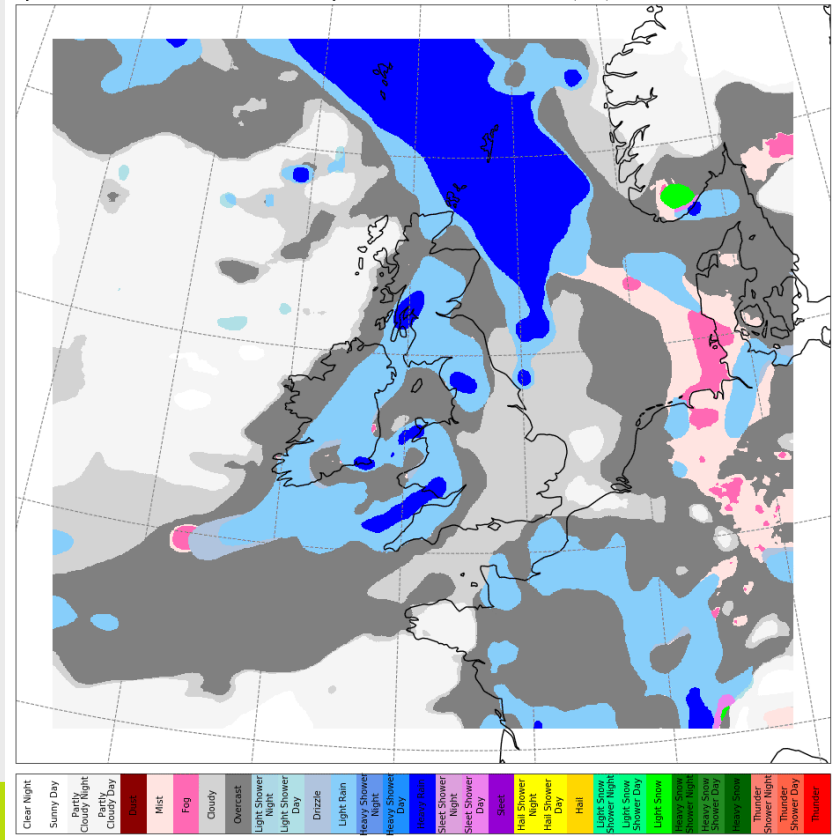


- Blended probability of rainfall
 - Showing prob (rain rate) > 0.03mm/hr
- Similar processing chains for snow, cloud, visibility, winds, temperatures, etc
- Combine to determine appropriate weather symbol...

Unknown Model probability of rainfall_rate > 0.03 mm/hr
 Cycle Time: 11 UTC on Thu 06/12/2018 Validity Time: 11 UTC on Thu 06/12/2018 (T+0)

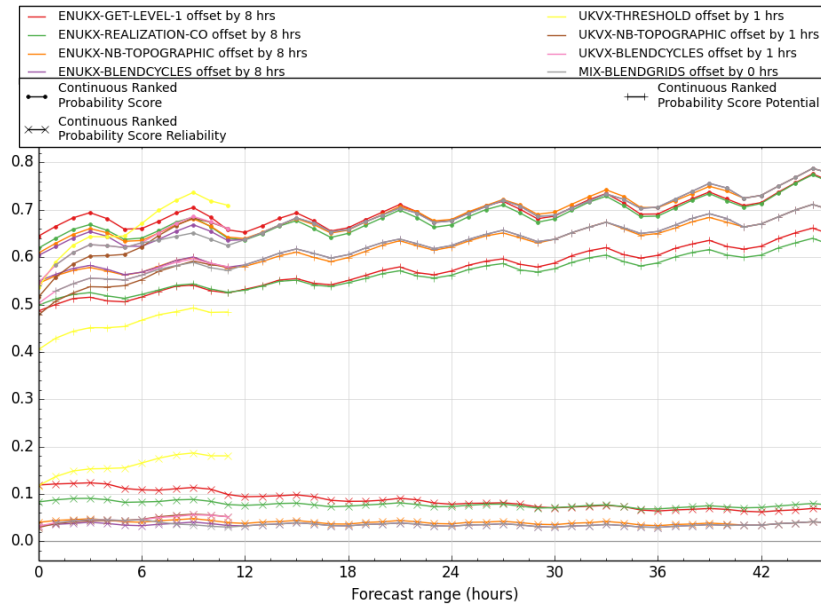


Unknown Model weather_code
 Cycle Time: 11 UTC on Thu 06/12/2018 Validity Time: 11 UTC on Thu 06/12/2018 (T+0)



Example of verification at each step: Temperature: CRPS

Surface (1.5m) Temperature, Reduced MOGREPS-UK Model area,
Equalized and Meaned between 20171101 00:00 and 20171130 00:00, Surface Obs



- CRPS (•) measures quality of full PDF. Lower is better
- **Nbhood** reduces reliability penalty (x) at the expense of 'potential' (+)
- **Nbhood** more net benefit to UKV (no prior spread)

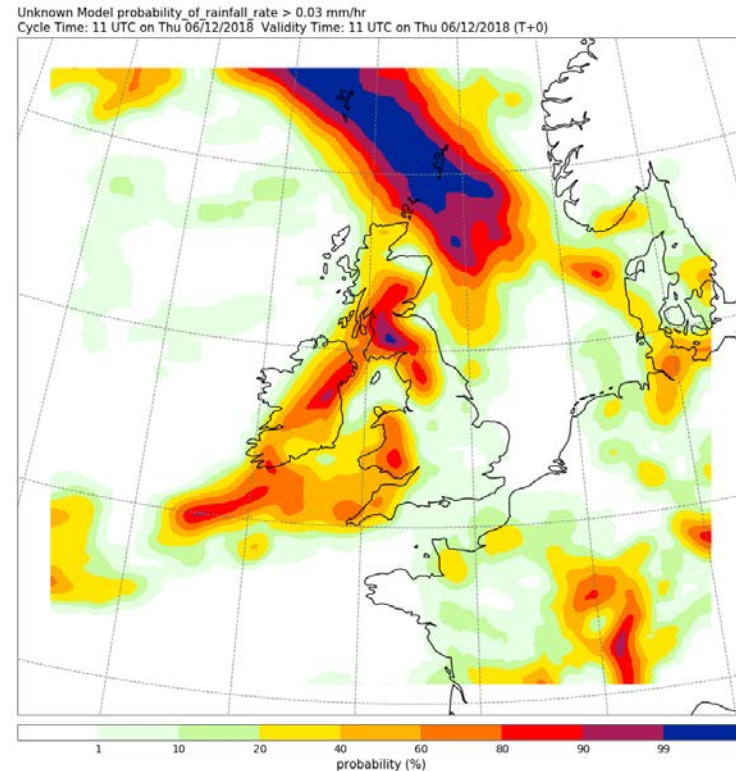
The challenge to exploit ensemble prediction

ECMWF and NCEP ensembles introduced in 1992; MOGREPS in 2005

Scientific case for ensembles overwhelming, but exploitation in services challenging:

- *...the public don't understand probabilities...*
- *...uncertainty is too difficult to communicate...*
- *...just tell me what will happen...I just need to make a decision...*

Risk Management Tool
 $Risk = Probability \times Impact$



Example: Probability of Precipitation from new IMPROVER blend of models (UKV and MOGREPS-UK) – see Roberts, MOSAC 2019.

Exploiting ensembles is a common challenge

My old friend and fellow ensemble campaigner Steve Tracton said to me about this workshop:

“Can you believe after all these years the w/shop announcement starts with
The workshop’s central theme is to support NWS in its transition from single-value deterministic to ensemble-based probabilistic forecasting and to convey uncertainty to users...”?

I’d like to say we had made more progress, but here is a quote from the Met Office’s emerging Science Strategy:

Theme: Producing and Exploiting Ensembles

Goal 1: Transform operational prediction to be based on ensembles...

