

# Diagnostics 1

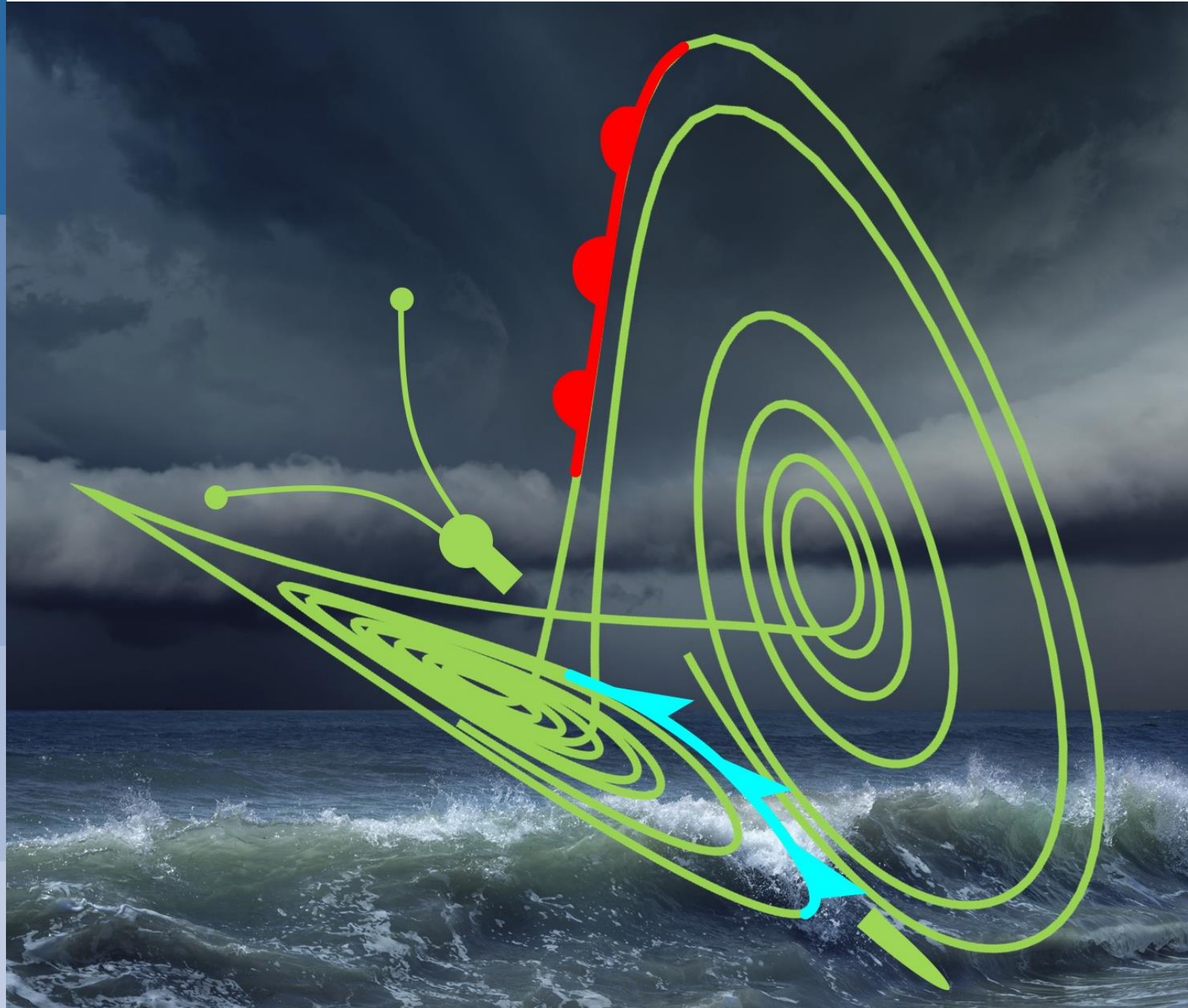
Mark Rodwell

**Collaborators**

David Richardson, Dave Parsons,  
Heini Wernli, Linus Magnusson, Elias Hólm

Training course: Predictability & ocean-atmosphere ensemble forecasting

1 March 2019, ECMWF



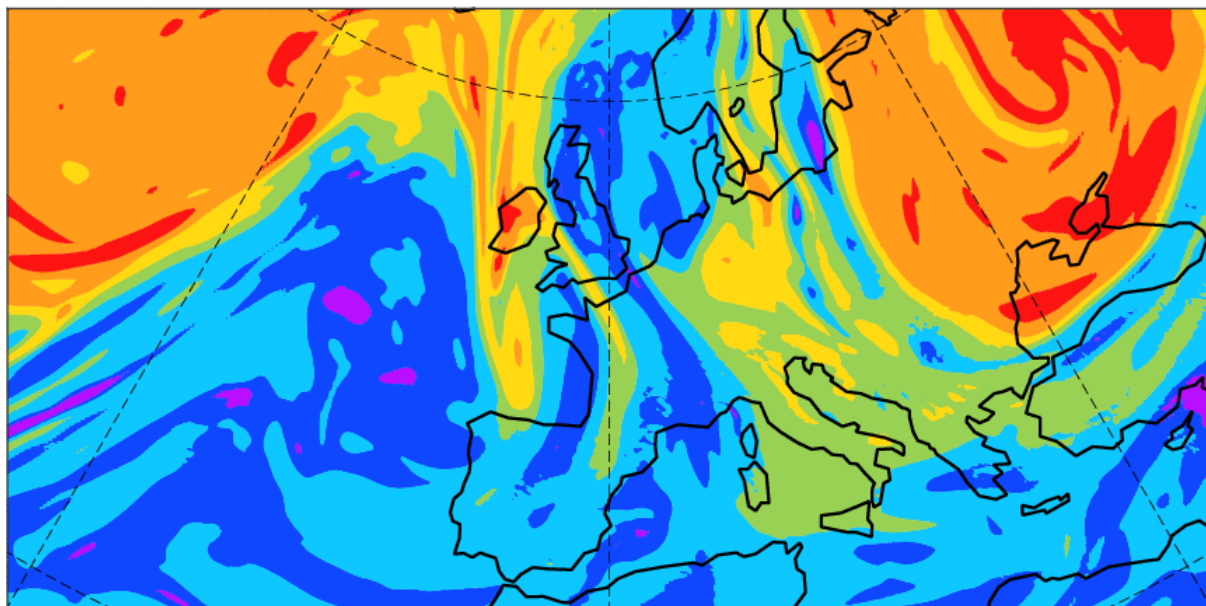


# Animation of a very poor medium-range single forecast

Potential Vorticity on the Potential Temperature = 320K surface. 20110410 0 UTC. Step (days, hours) = 0 00.0

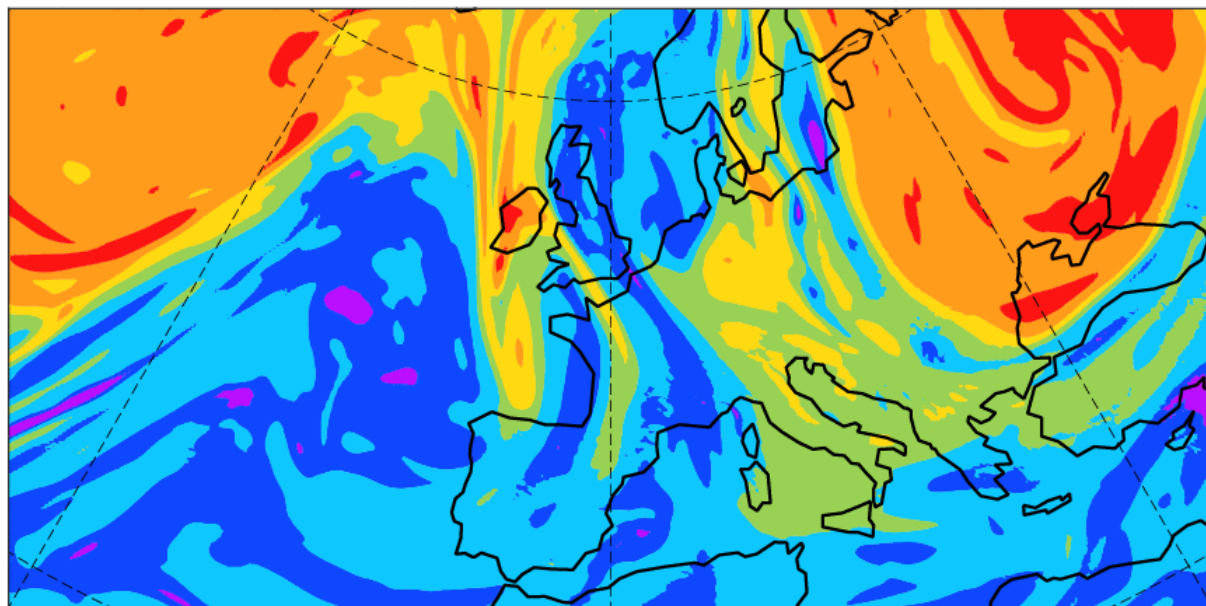
Observed

PVU



Forecast

PVU

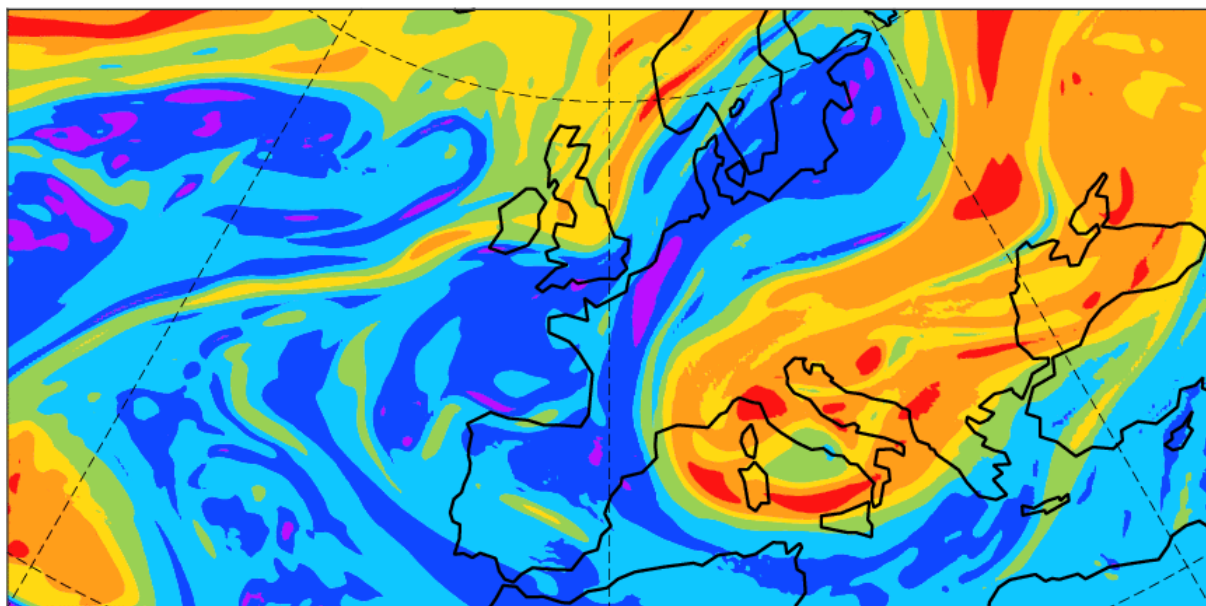


# Animation of a very poor medium-range single forecast

Potential Vorticity on the Potential Temperature = 320K surface. 20110410 0 UTC. Step (days, hours) = 6 00.0

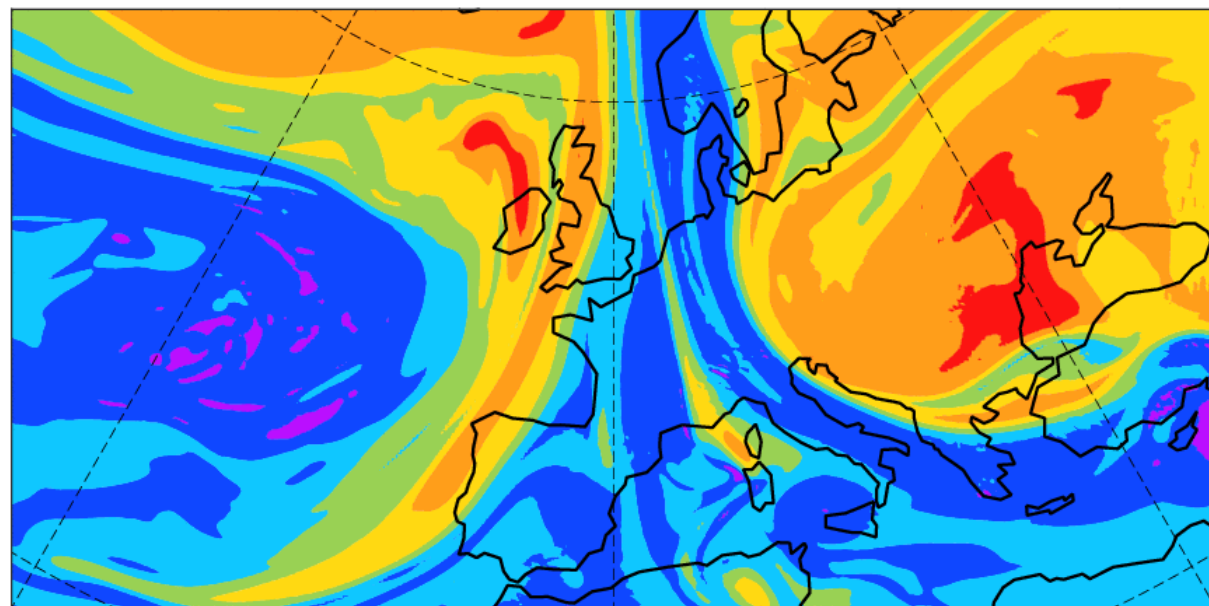
AN

PVU



HR

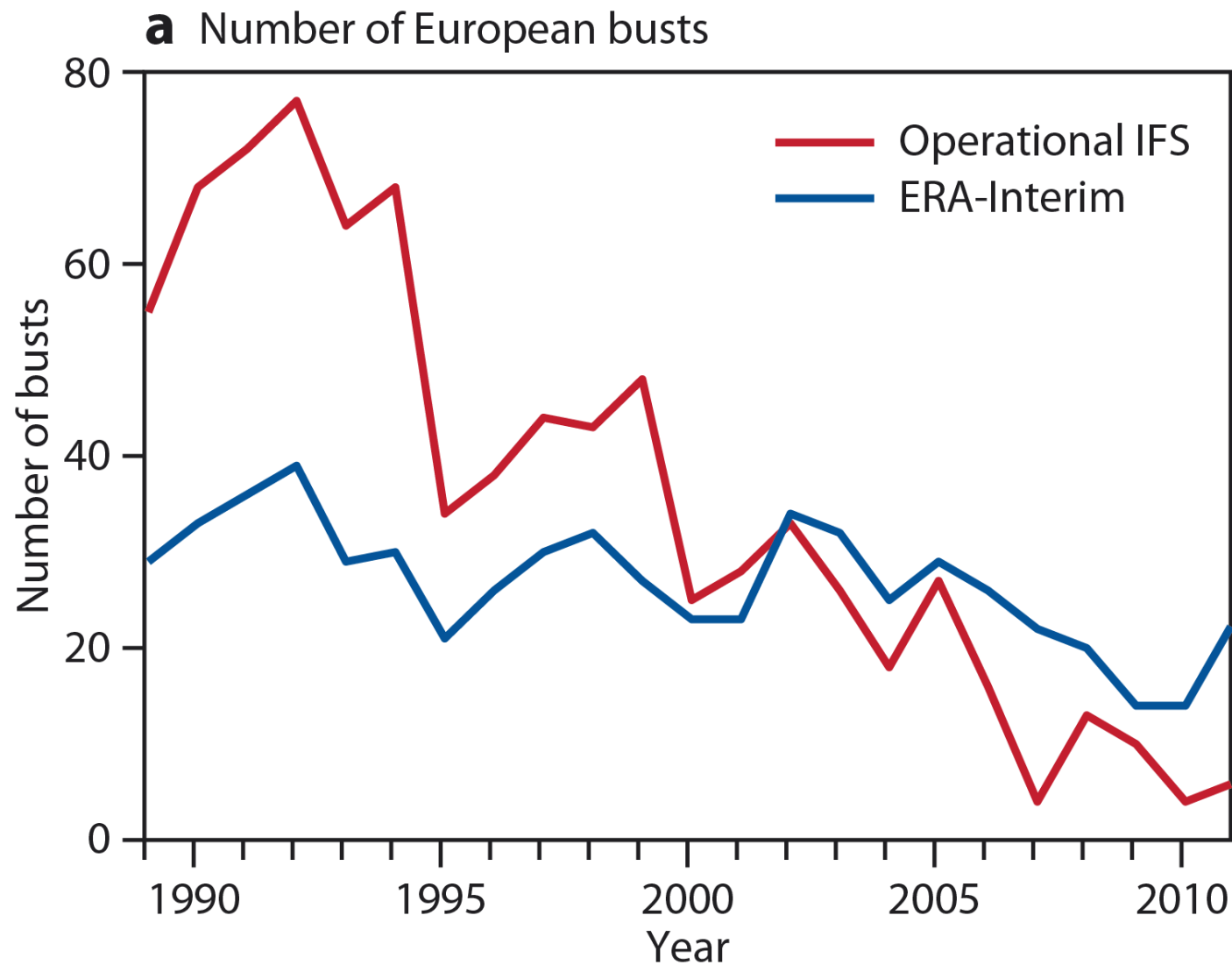
PVU



**FAIL**

We see the mixing of air masses. The eventual block (high pressure) over Northern Europe is not well predicted  
With a single forecast, it is easy to quantify the error (pointwise differences, pattern correlations etc.)

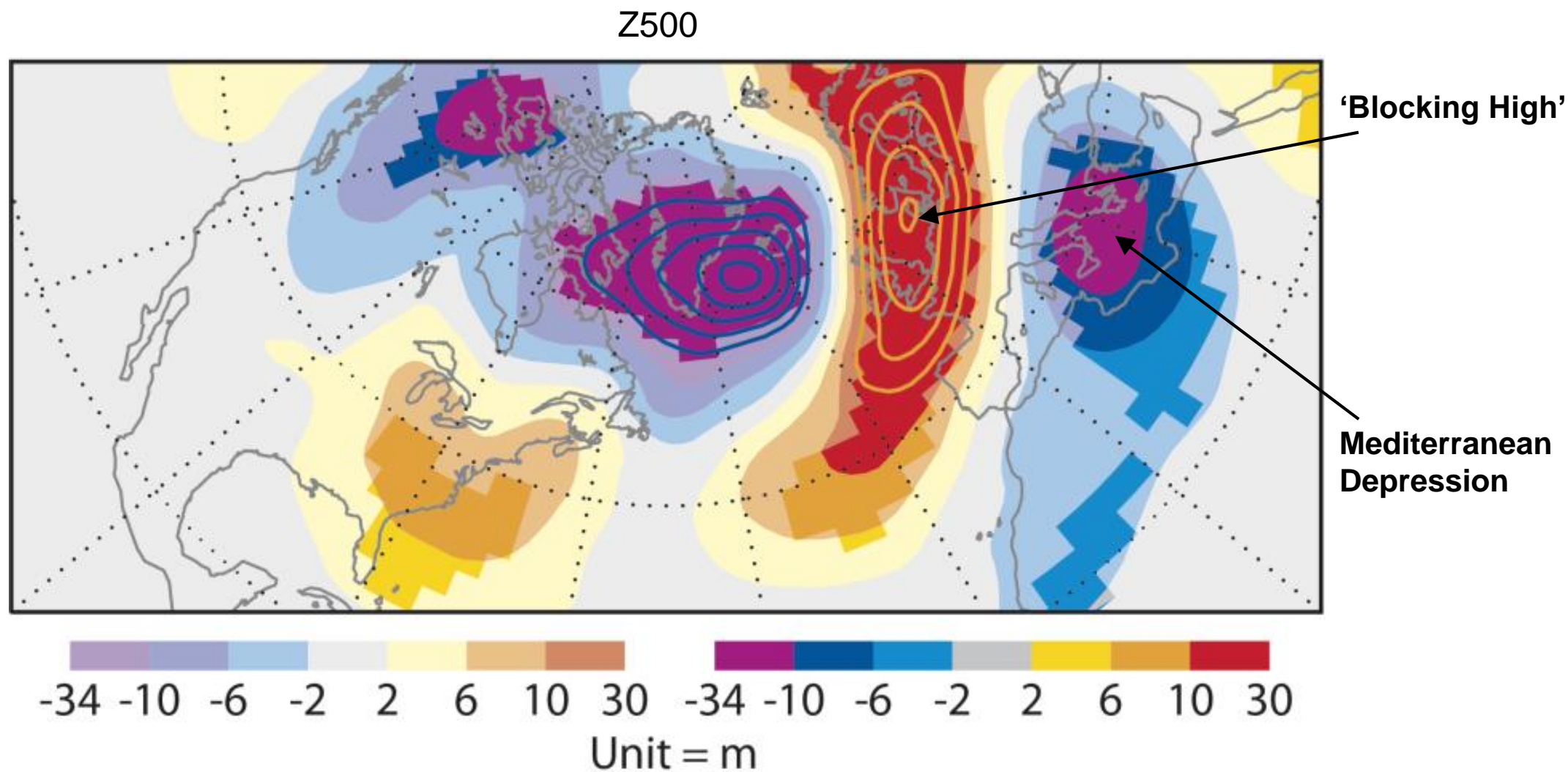
Frequency of severe busts has dropped a lot in operational forecast  
Can use busts within re-analysis to get a better picture of the salient aspects



A bust is defined as when the day-6 Z500 forecast has European RMSE > 60m and ACC < 40%

# Composite conditions during a bust

Rodwell et al, 2013, BAMS

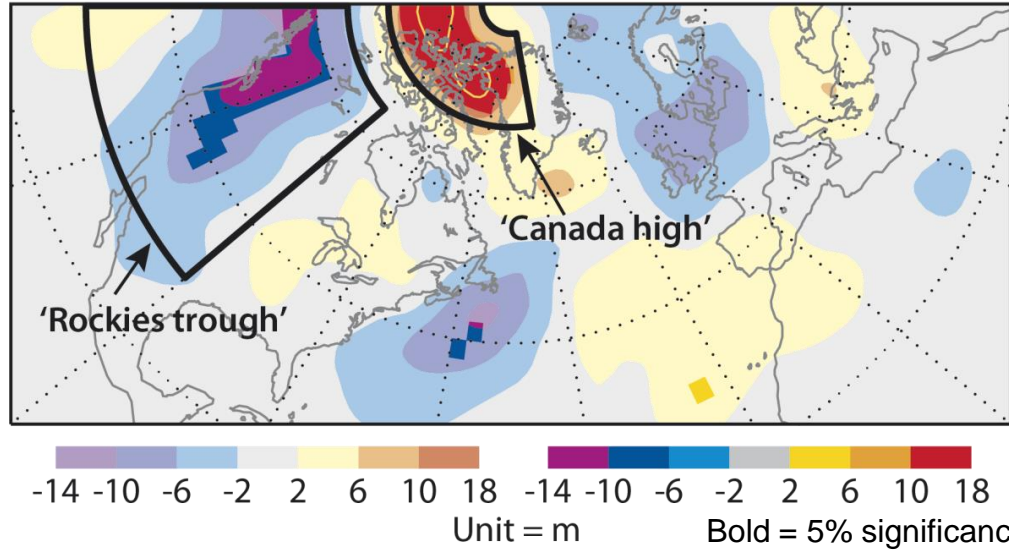


Using all 584 ERA Interim busts that occurred between the dates 1 January 1989 - 24 June 2012.

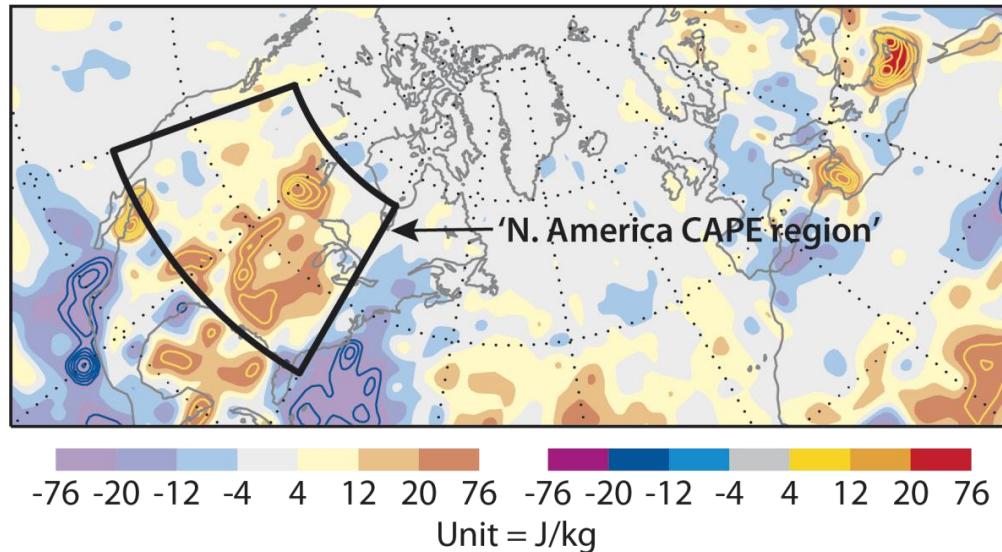
# Average initial conditions of 584 single forecast “busts” over Europe at day 6

Rodwell et al, 2013, BAMS

**a** Z500 anomaly



**b** CAPE anomaly



Trough over the Rocky mountains, with high convective potential ahead  
Conducive to the formation of mesoscale convection

Analysis highlights geographically-fixed factors. Other flow-features (extratropical transition of tropical cyclones and cyclogenesis) have also been shown to be important for some busts.

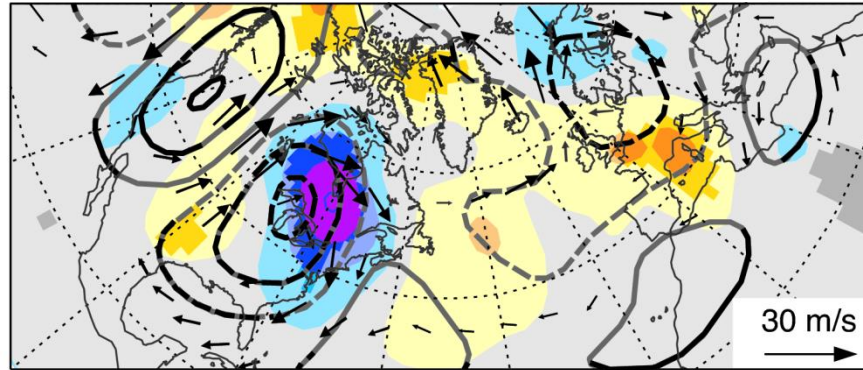
'CAPE' = Convective Available Potential Energy

# PV budget at 330K for trough/CAPE composite

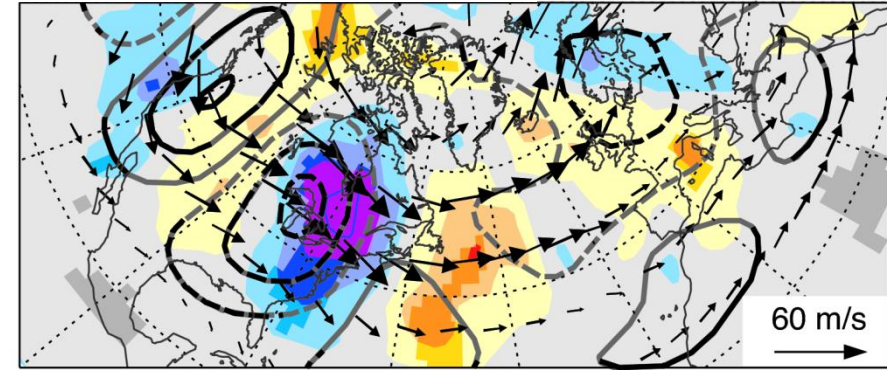
Rodwell et al, 2013, BAMS

Contours : PV anomaly  
Shading : PV tendency

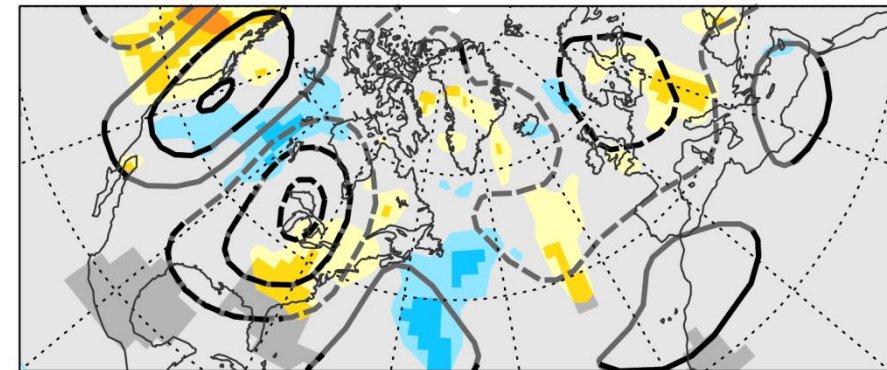
### Local PV tendency and anomalous winds



### PV advection and full winds



### Diabatic and frictional PV tendency



Unit = 0.1PVU/day



CI=0.4PVU

Largely the wave is being advected downstream  
But, by modifying the stratification, diabatic processes appear to oppose the advection term and slow the propagation of the wave  
Not just a symptom of the wave, but an active ingredient in the downstream bust

Composite over 95 trough/CAPE events 25 June 2010 – 20 March 2012 (0 or 12UTC) where operational analysis has projection onto Rockies trough > 3 and onto CAPE > 1. Contours show Potential Vorticity (PV) anomaly, shading shows (a) local tendency, (b) adiabatic advection and (c) material PV tendency (due to diabatic & frictional processes; deduced as the residual in the PV budget) on the 330K surface ( $\approx 250\text{hPa}$ )



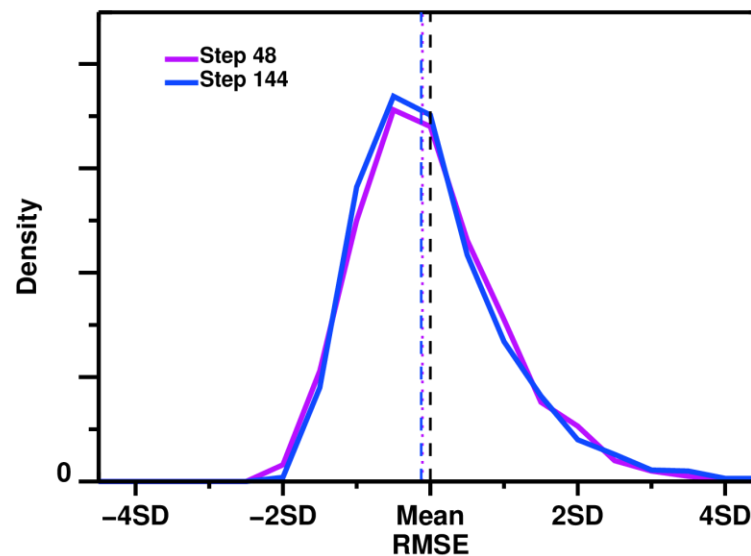
# Distribution of RMSE Z500 over Europe

With Heini Wernli

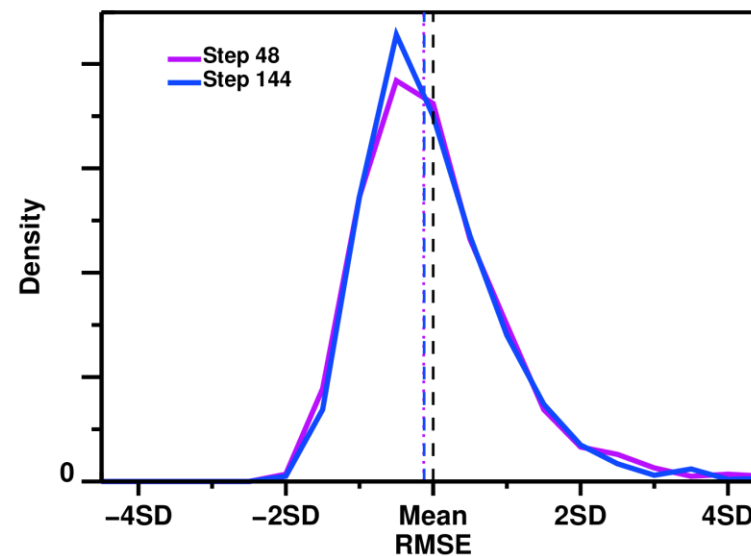
Shape of distribution is independent of lead-time  
Limited by zero error, so inevitably a longer tail at large error

How important for the seasonal-mean score are cases where the error > 2SD?

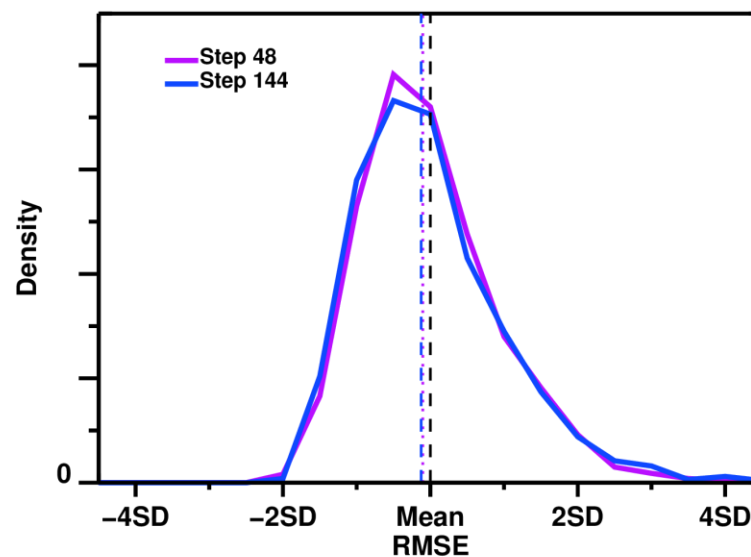
DJF distribution of RMSE for Europe 2010–2018



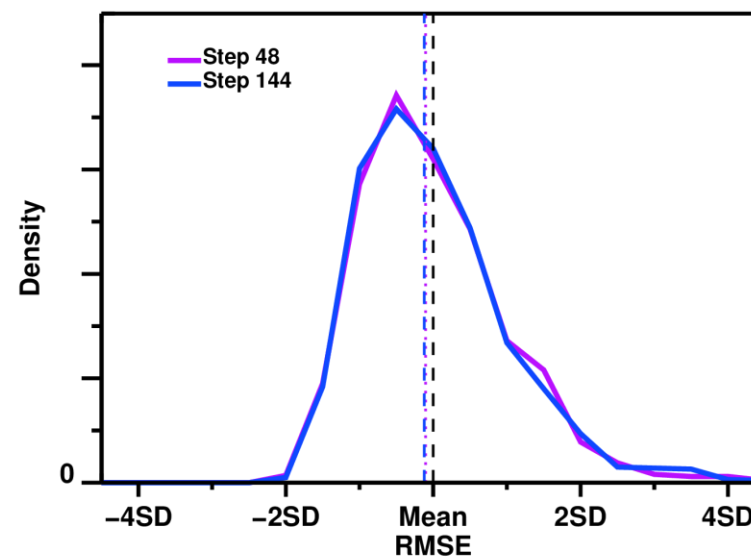
MAM distribution of RMSE for Europe 2010–2018



JJA distribution of RMSE for Europe 2010–2018



SON distribution of RMSE for Europe 2010–2018

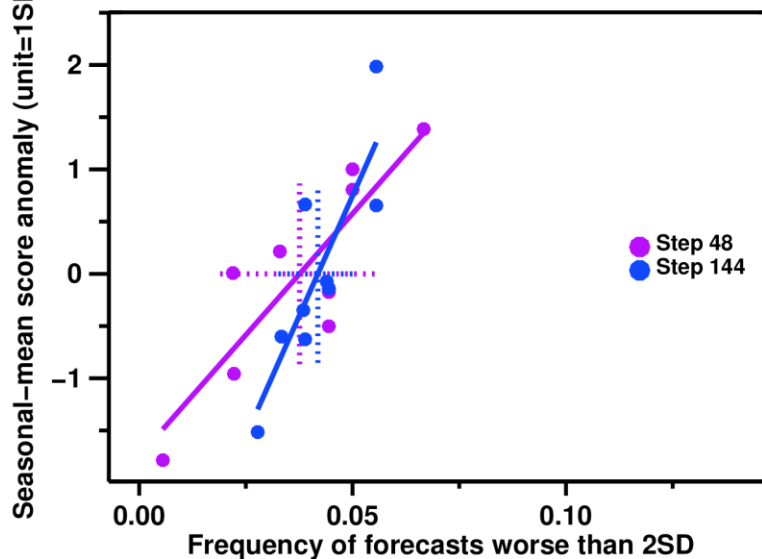


# Impact of frequency of poor forecasts on seasonal-mean Z500 RMSE over Europe

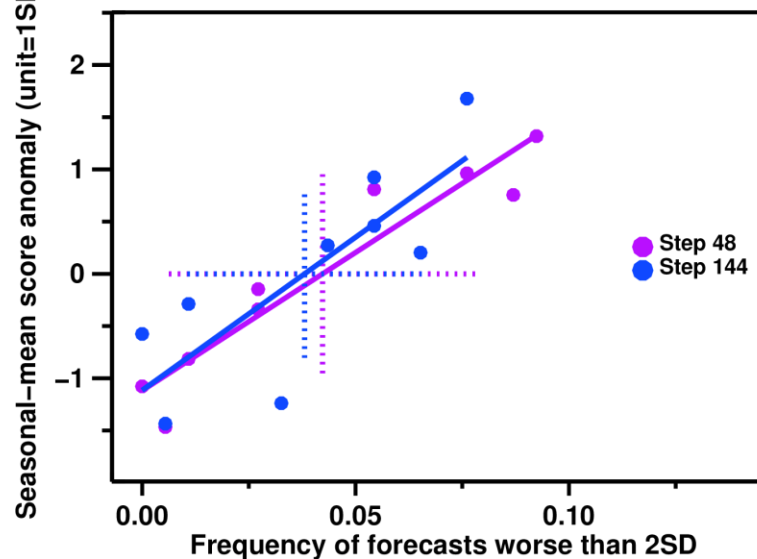
With Heini Wernli

The impact of 1SD in poor forecast frequency equates to close to 1SD in seasonal-mean score

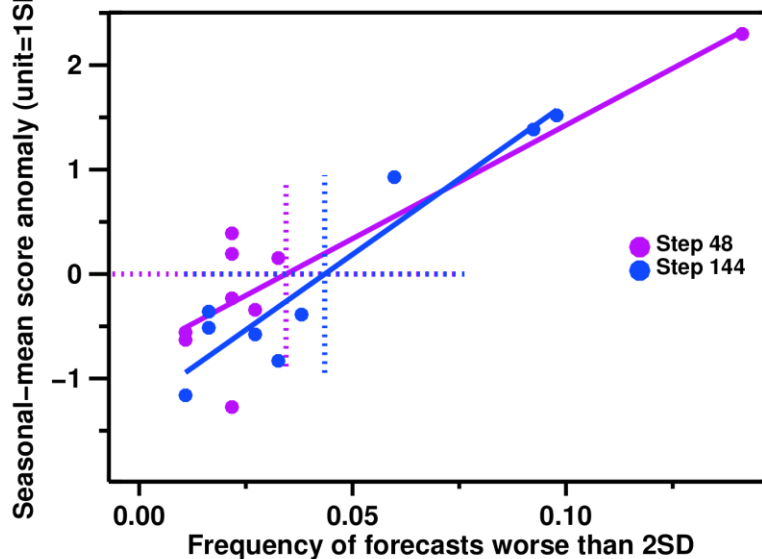
DJF scores versus busts. RMSE for Europe 2010–2018



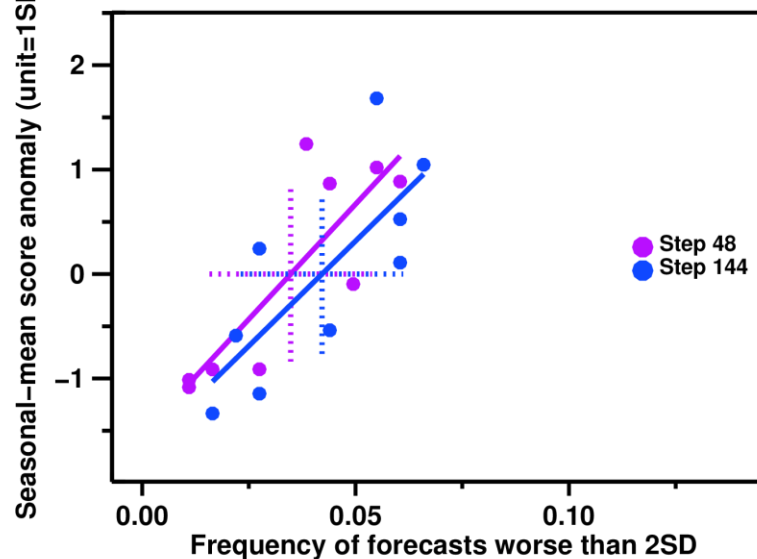
MAM scores versus busts. RMSE for Europe 2010–2018



JJA scores versus busts. RMSE for Europe 2010–2018

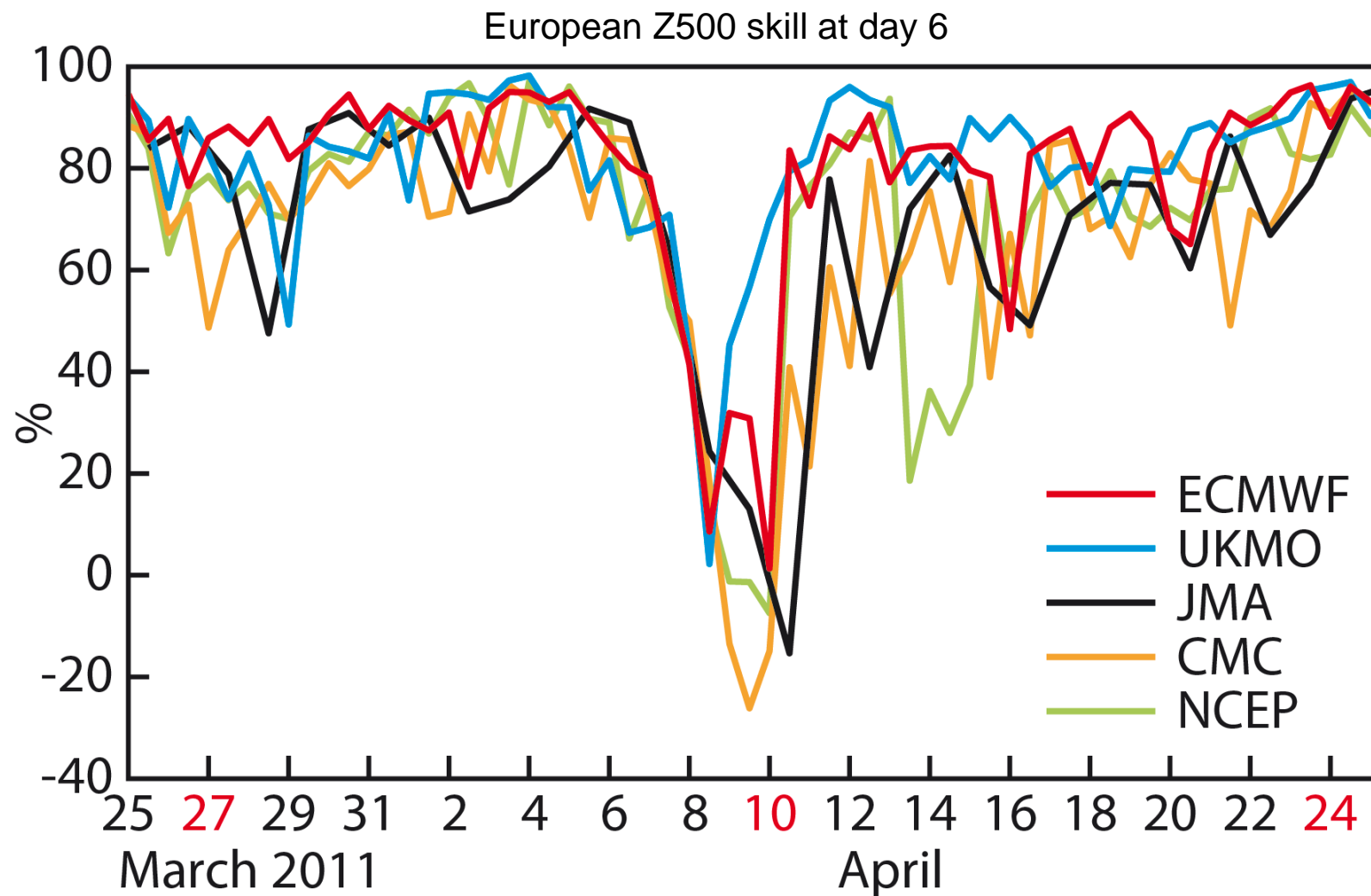


SON scores versus busts. RMSE for Europe 2010–2018



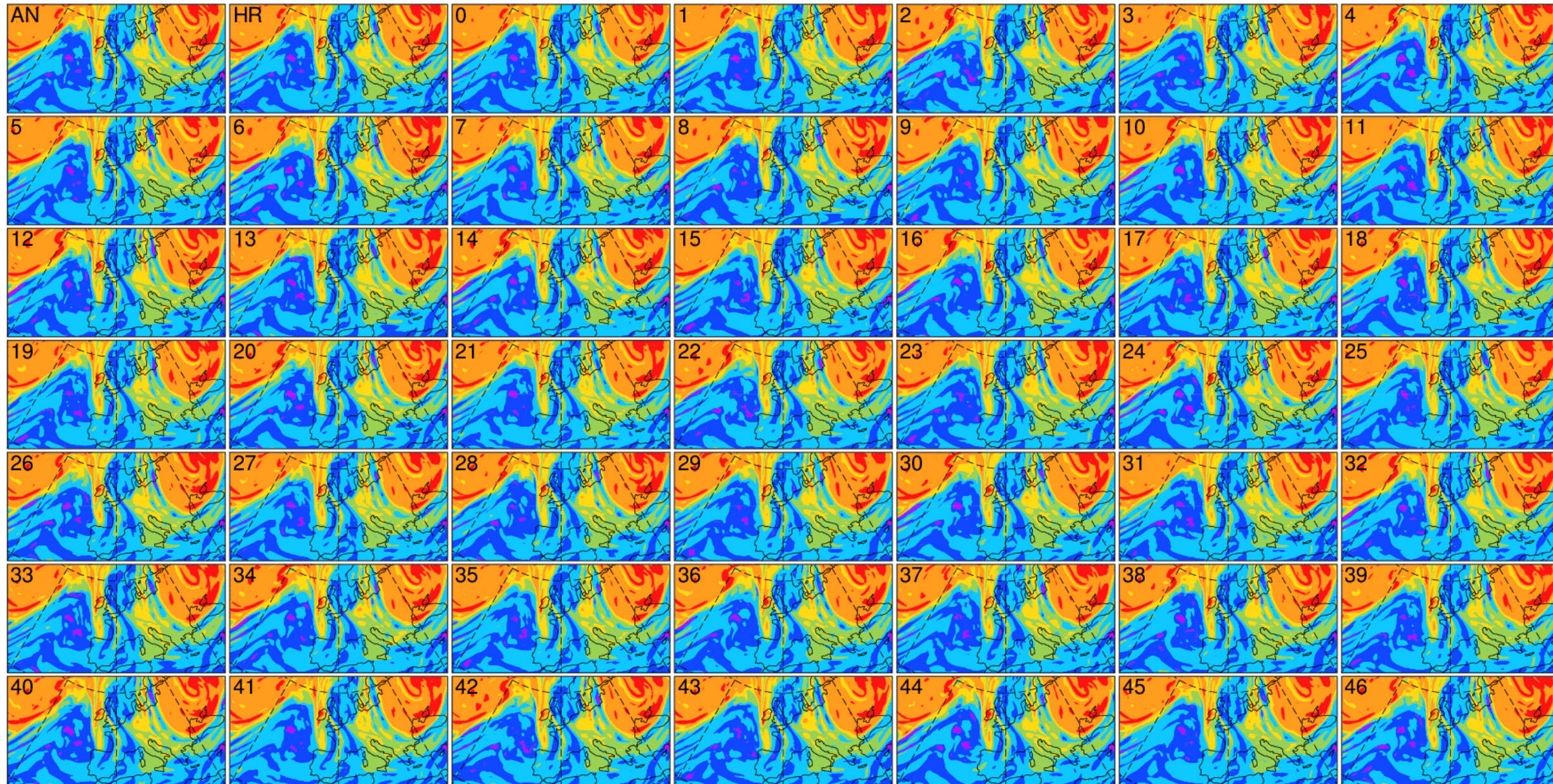
# Back to the single bust case: All forecast centres suffered

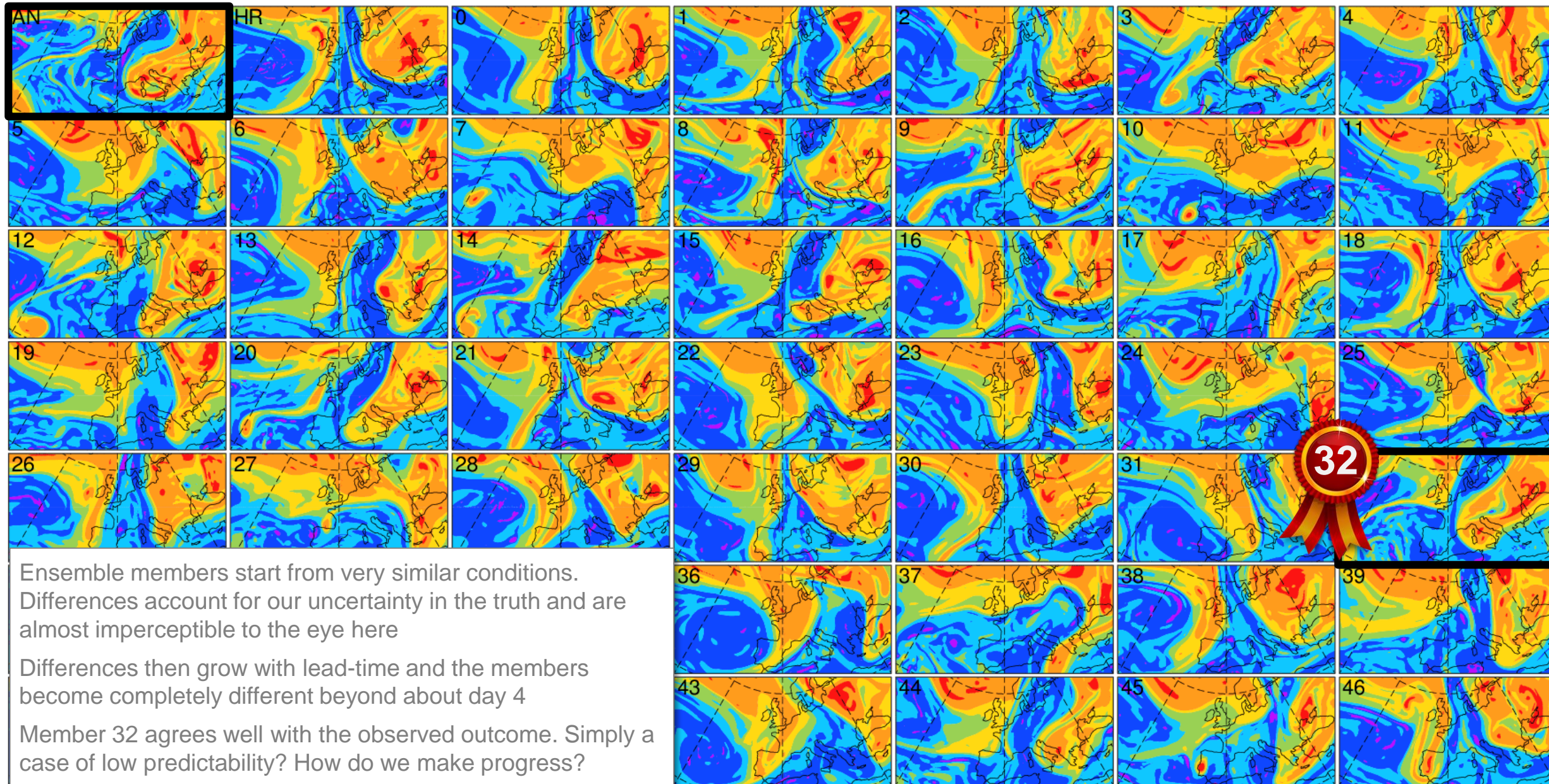
Rodwell et al, 2013, BAMS



All centres suffered  
Suggests an issue of reduced "predictability"  
Look at the ensemble

Spatial Anomaly Correlation Coefficient for 500 hPa geopotential height in [12.5°W–42.5°E, 35°N–75°N]. Date is forecast start

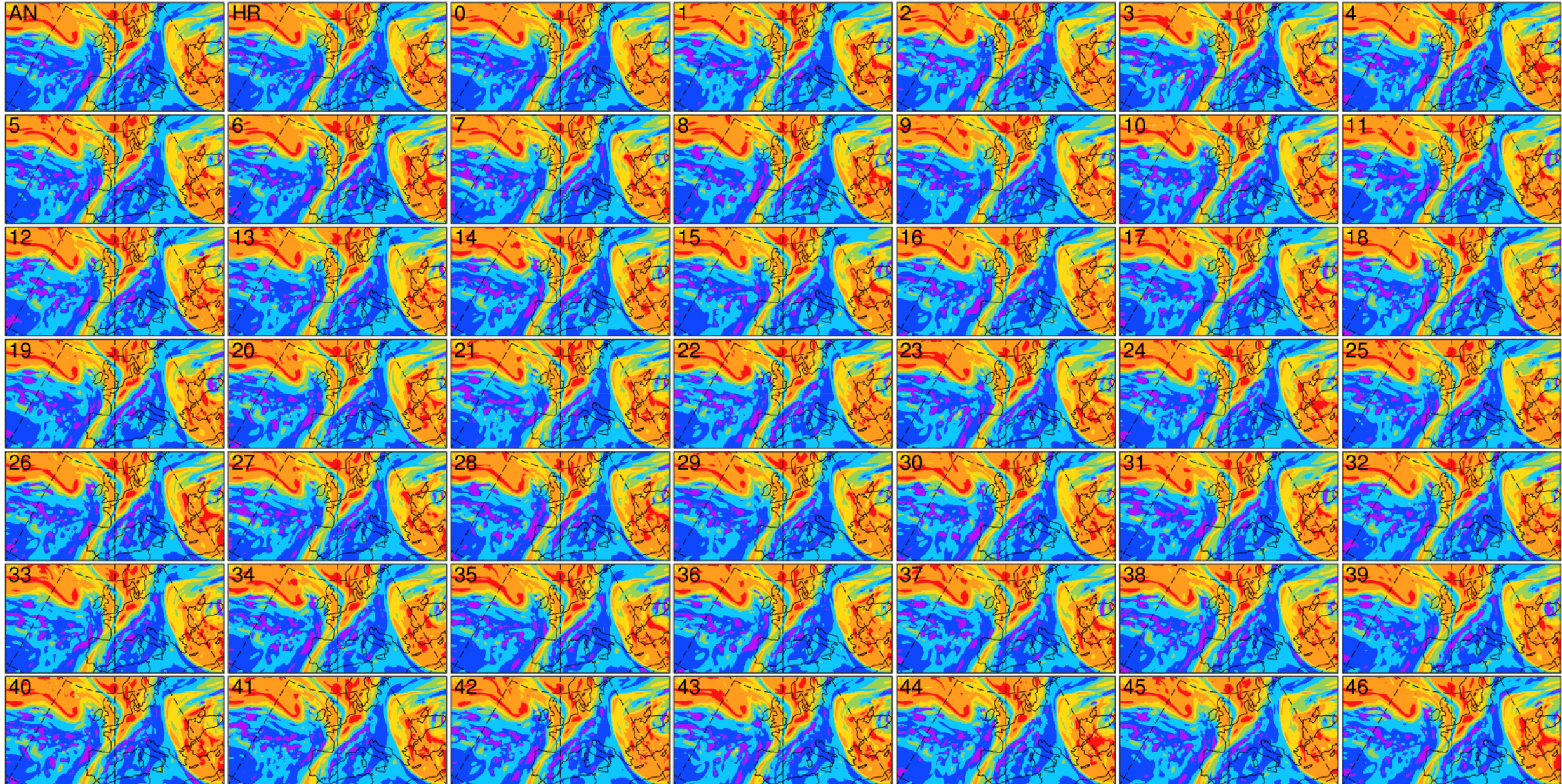


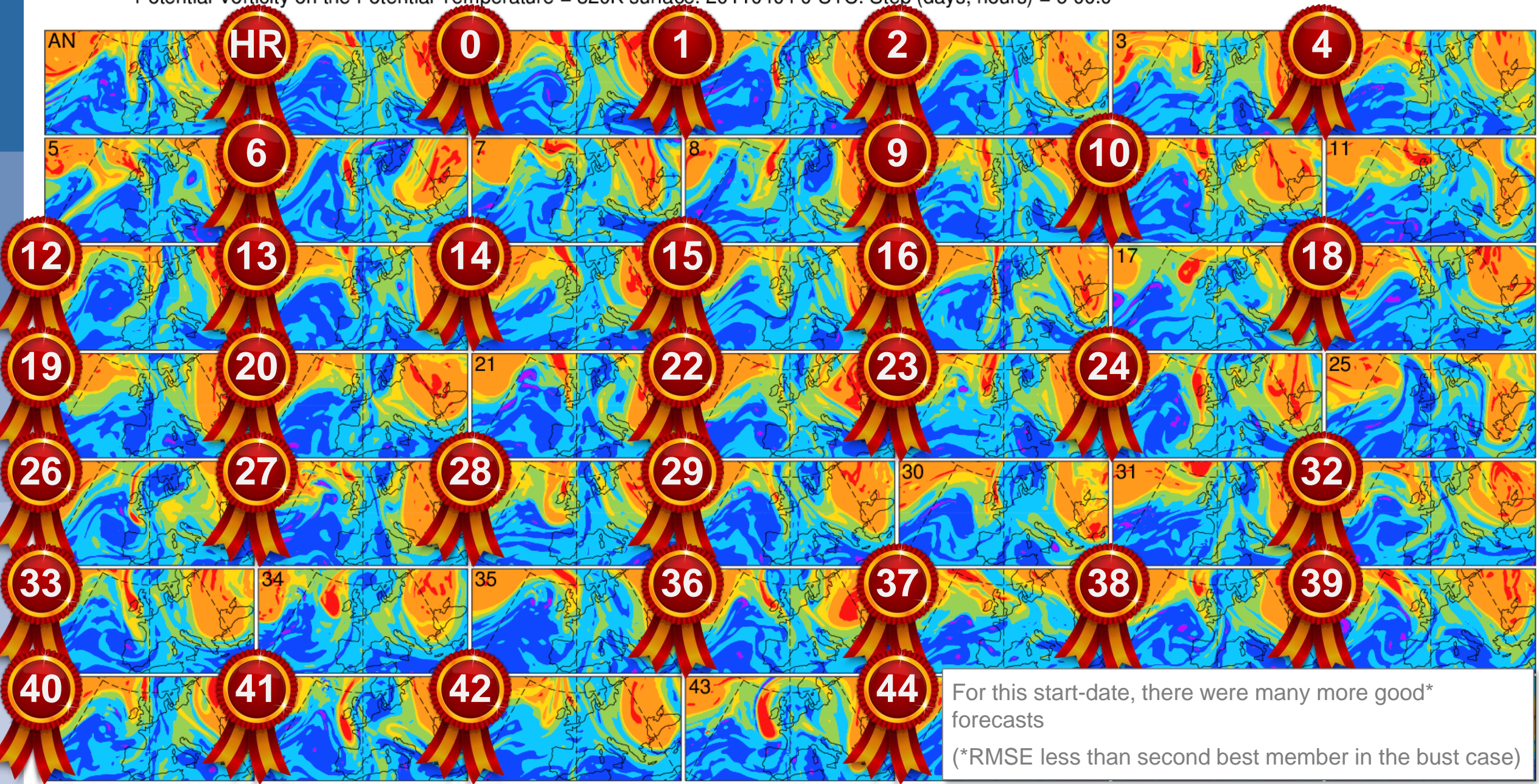


Ensemble members start from very similar conditions. Differences account for our uncertainty in the truth and are almost imperceptible to the eye here

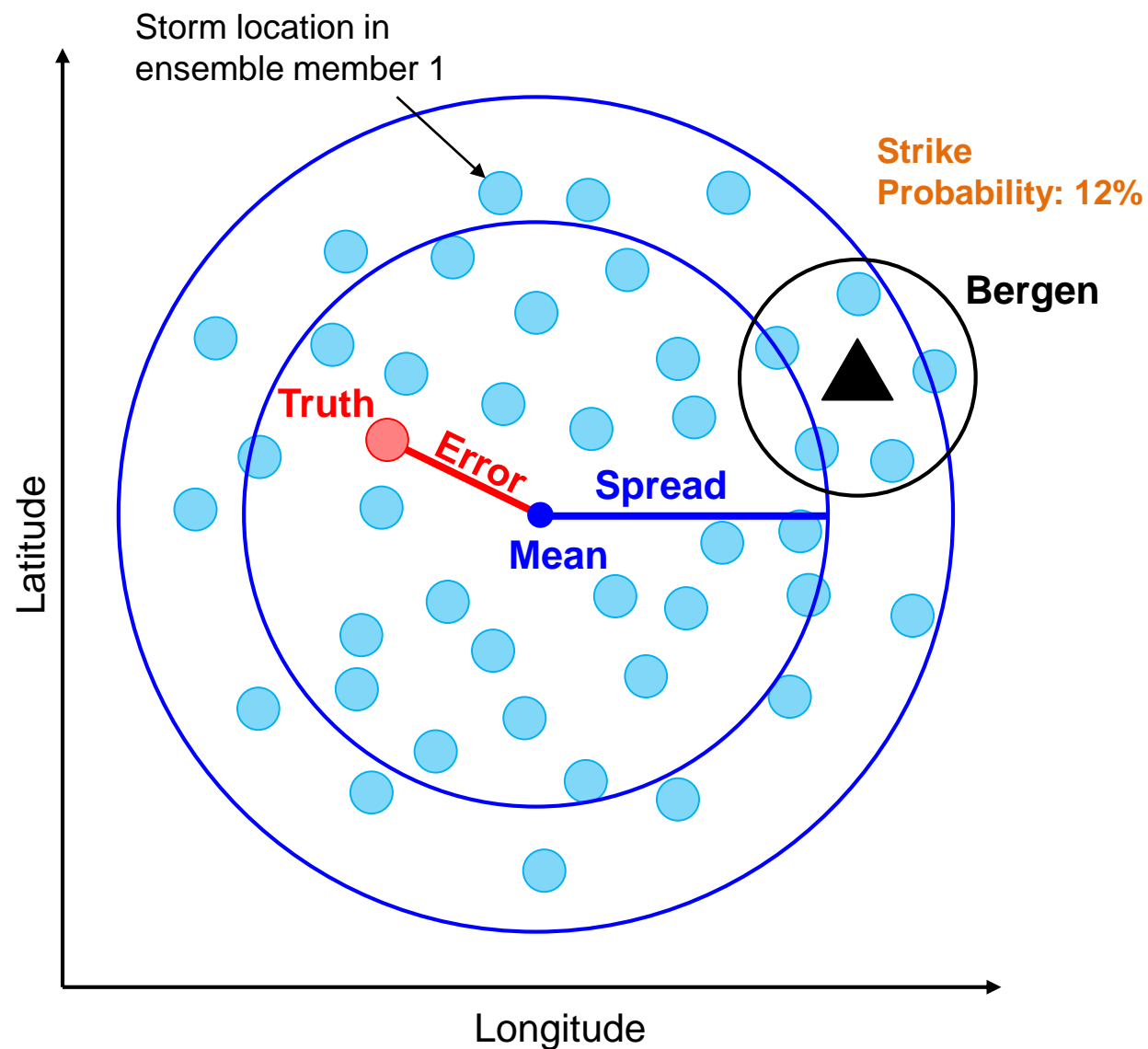
Differences then grow with lead-time and the members become completely different beyond about day 4

Member 32 agrees well with the observed outcome. Simply a case of low predictability? How do we make progress?





# Motivation: Reliability and Sharpness



In a **reliable** forecast system, the truth should be statistically indistinguishable from the individual ensemble members

Reliability is very useful: an event predicted to occur with probability 12% will happen with frequency 12%

An easily testable consequence of reliability is that

$$\overline{\text{Error}^2} = \overline{\text{Spread}^2}$$

(averaged over many forecast start dates)

“The task of NWP research is to maintain/improve reliability while decreasing spread (improving refinement)”

Q. Can we develop diagnostics which efficiently (optimally?) guide us in this task?



# Ensemble spread and error

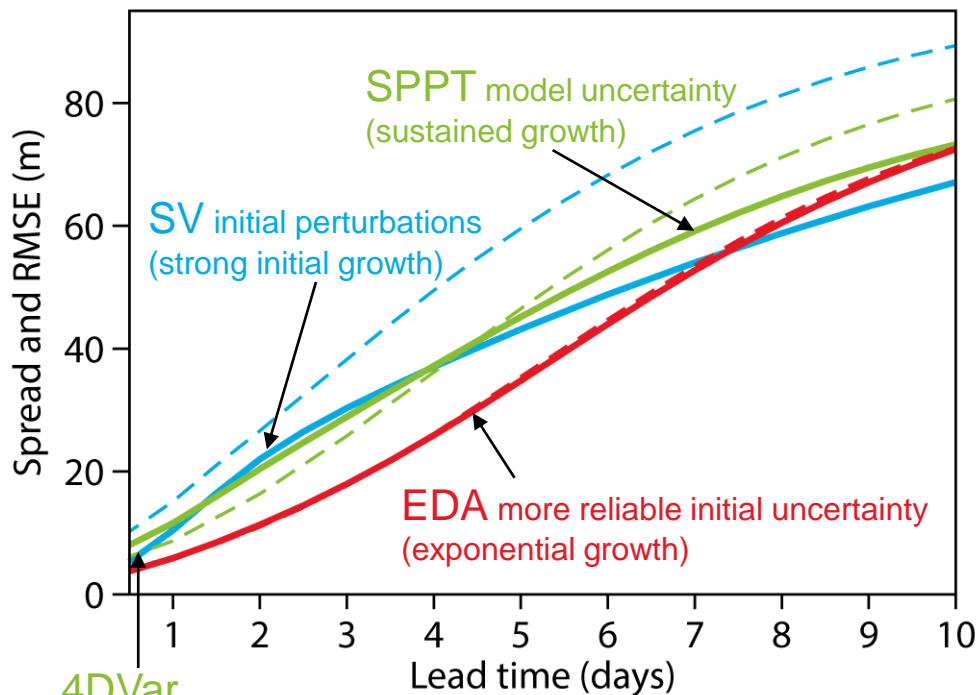
Z500

Rodwell et al. 2018, BAMS

Overall Error and Spread have reduced and come into alignment; due to better observations, initial conditions, forecast model and better representation of uncertainty

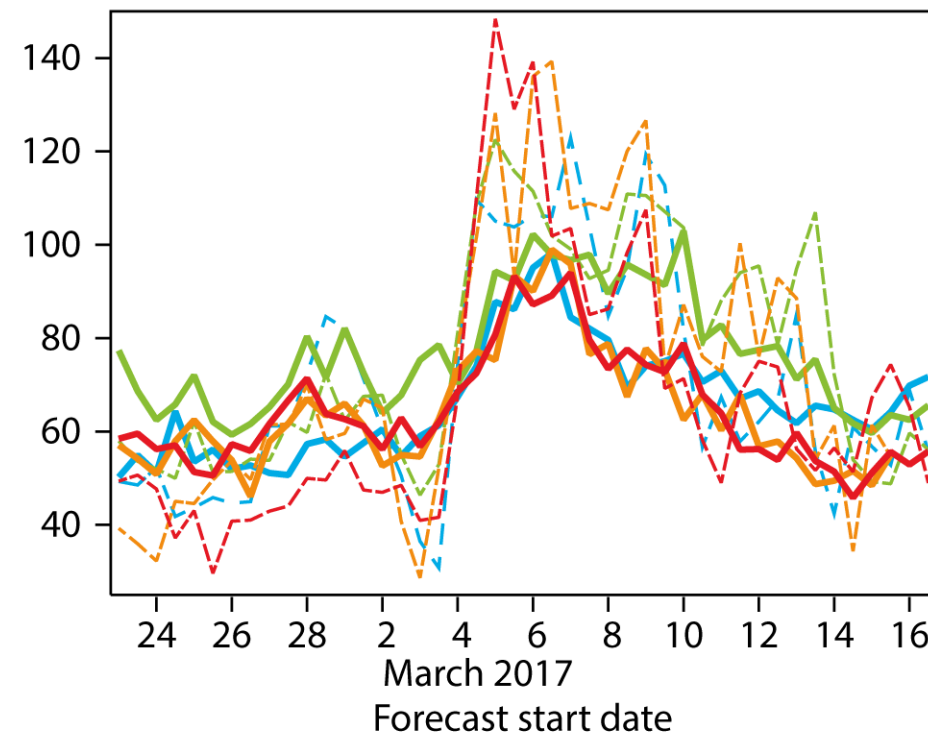
...but we make ensemble forecasts to represent the day-to-day variations in predictability and uncertainty. Can we evaluate it in our forecasts?

Annual means N.Hem. (ECMWF)



	1996	2005	2014
Spread			
Error			

Timeseries for Europe at D+6 (TIGGE)



	ECMWF	UKMO	JMA	NCEP
Spread				
Error				

500 hPa geopotential height (Z500). "Error" is RMS of ensemble-mean error  
 Spread = ensemble standard deviation (scaled to take account of finite ensemble size)

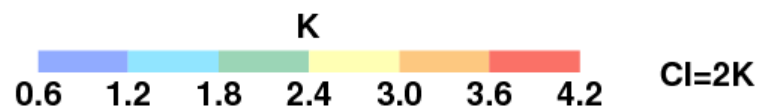
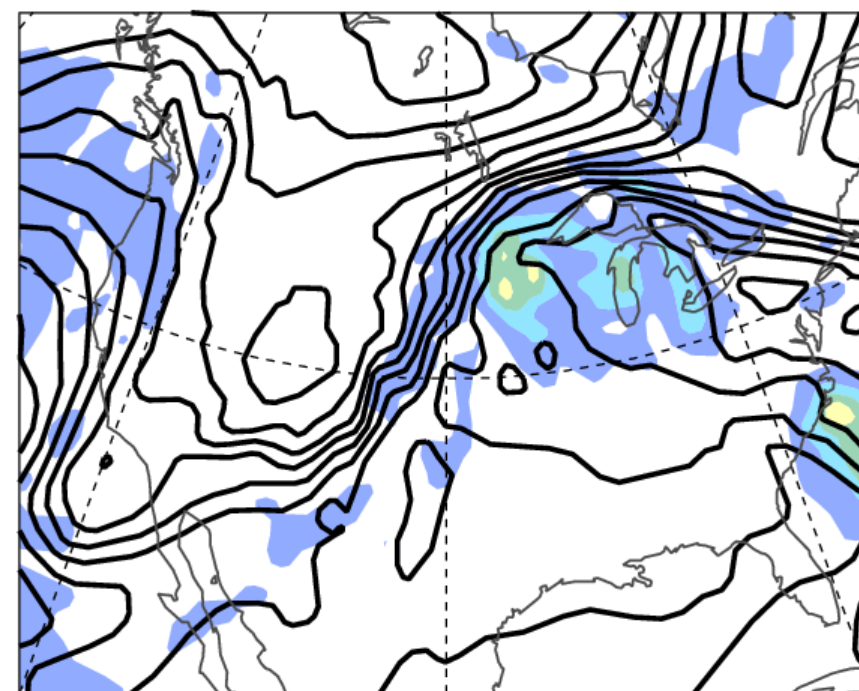
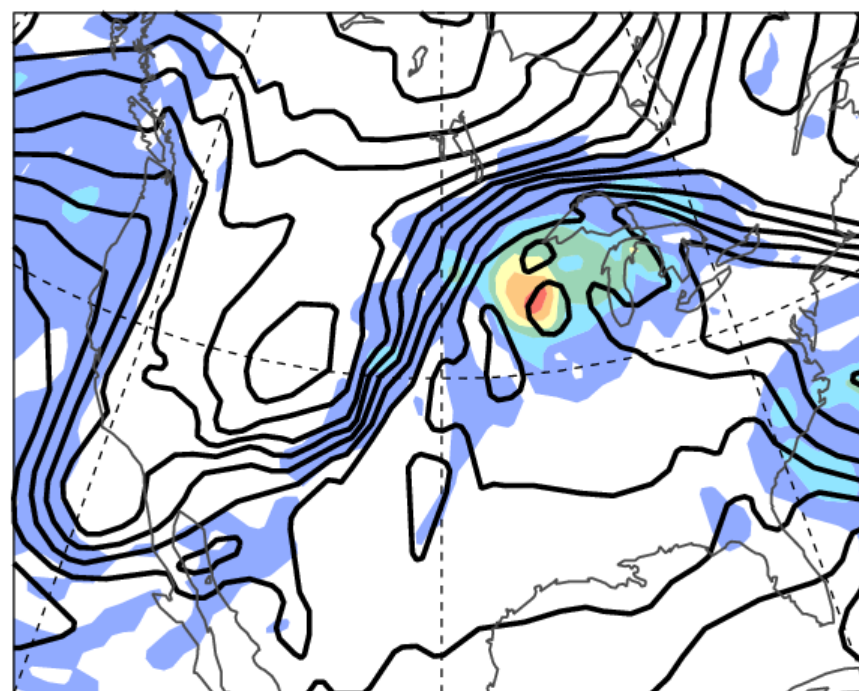
10 April T200 mean & spread

First-guesses T+12hr

Analyses

Uncertainties in the prediction of mesoscale convection magnify first-guess spread

Assimilation of new observations acts to sharpen the distribution



Data: Temperature at 200 hPa from 10-member EDA, valid at 6UTC

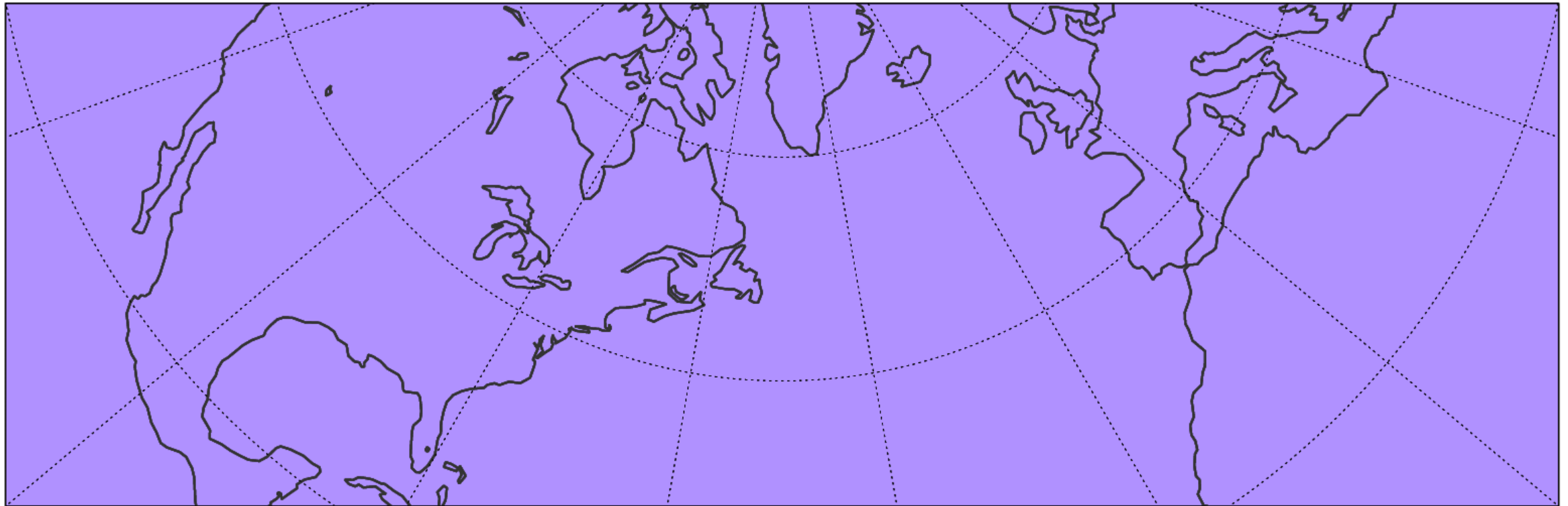
# Animation of ECMWF ensemble forecast spread 20170305 12Z D+0 to 6: $\sigma_{Z500}$

ECMWF ENS stdev  $Z_{500\text{hPa}}$  (shaded).

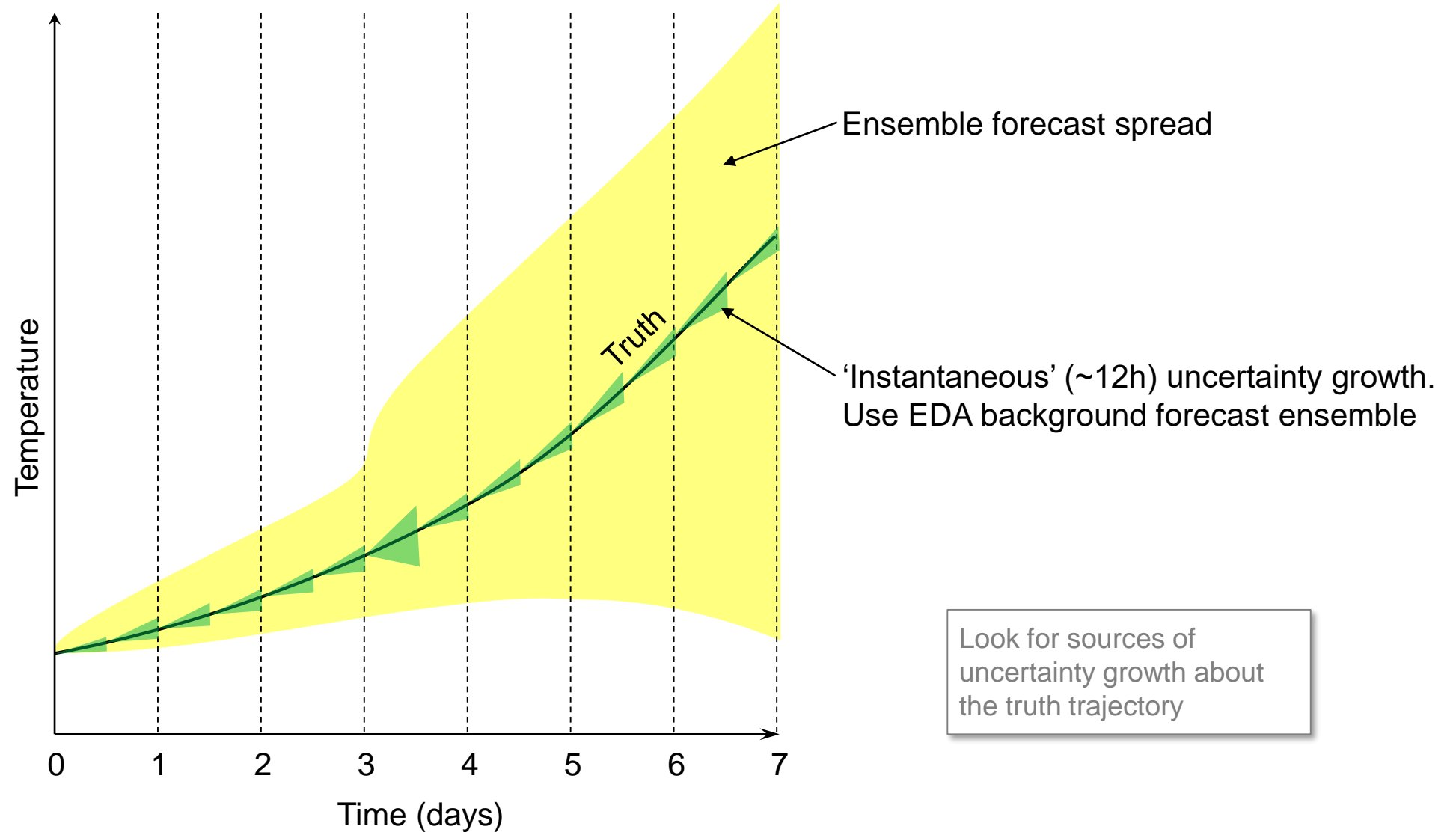
20170305 12Z

Uncertainty growing from various sources, is itself advected, and becomes large over Europe by D+6

Unit: m



# Ensemble forecast spread and the quasi-instantaneous growth of uncertainty



# “Lagrangian” growth-rate (following EnsMn horizontal flow) for EDA background $\sigma_{PV_{315}}$

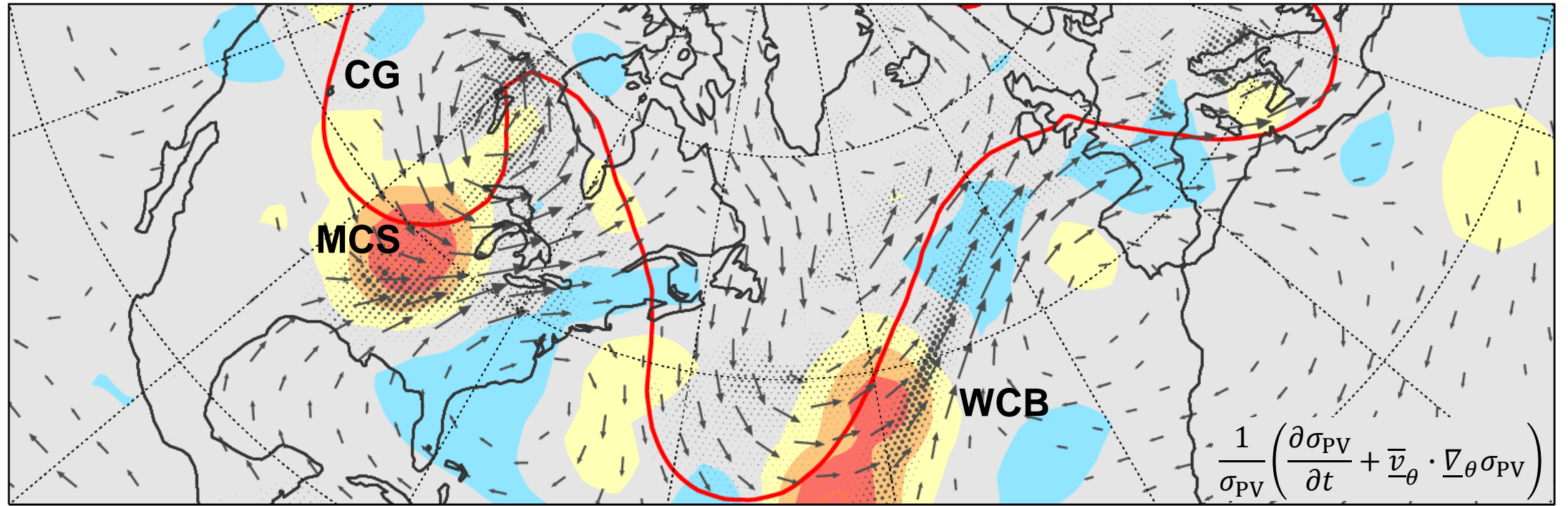
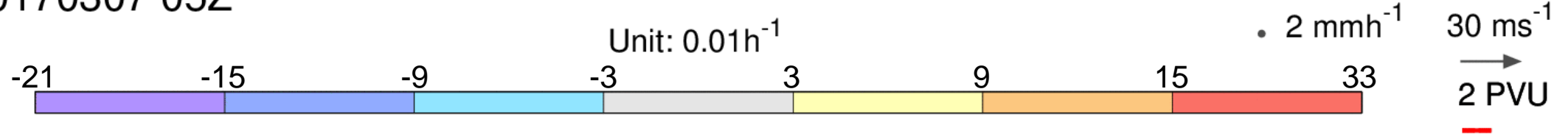
Much uncertainty growth associated with moist processes: **Warm Conveyor-Belts**, and **Meso-Scale Convection**

Interaction of uncertain features, large ENS spread & poor prediction of Euro blocking at D+6

Aim: Evaluate short-range synoptic flow-dependent representation of uncertainty

Q: Is sensitivity to moist processes real or due to deficiencies in model uncertainty representation? TIGGE?

20170307 05Z

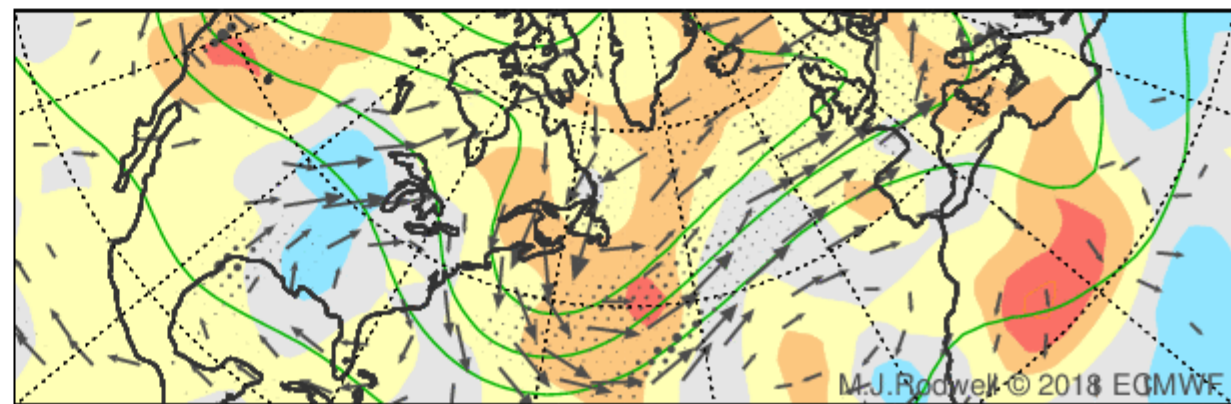
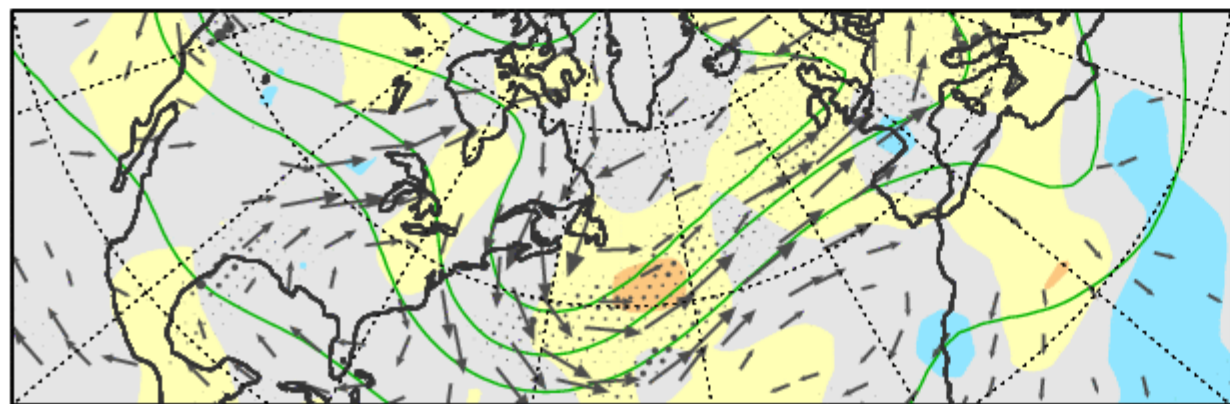
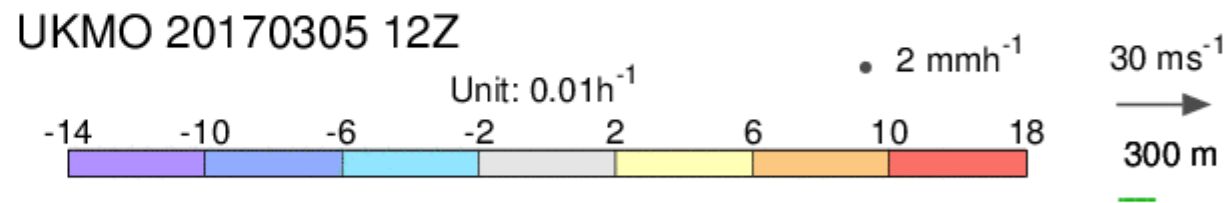
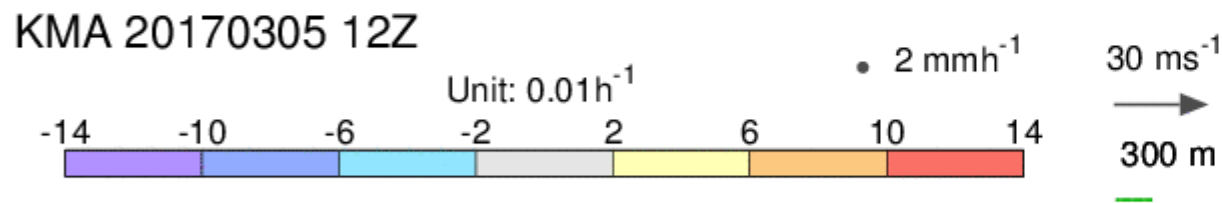
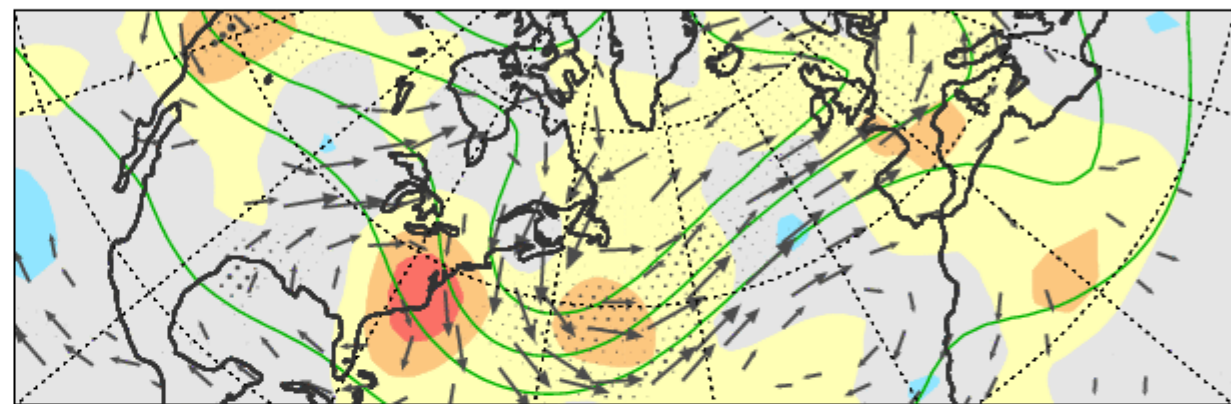
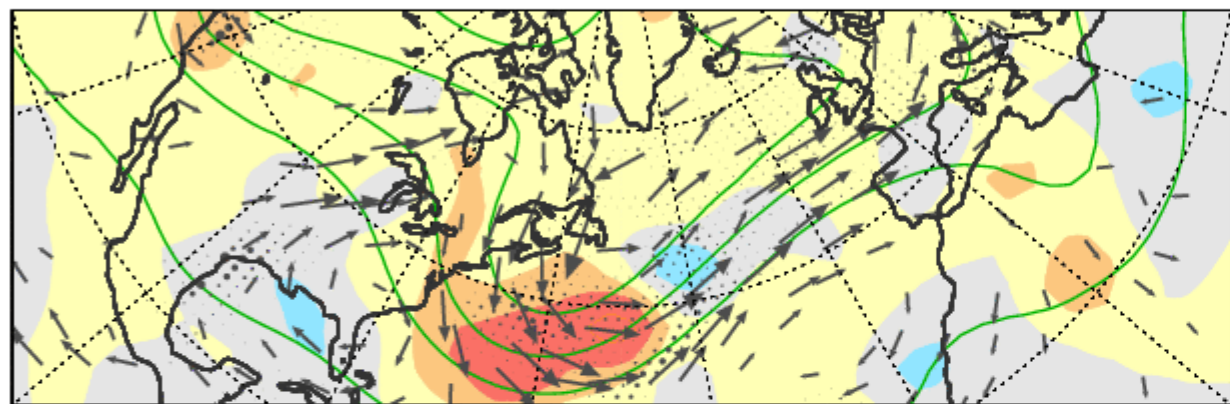
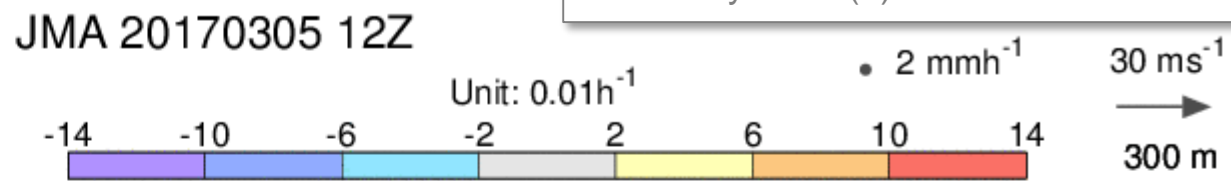
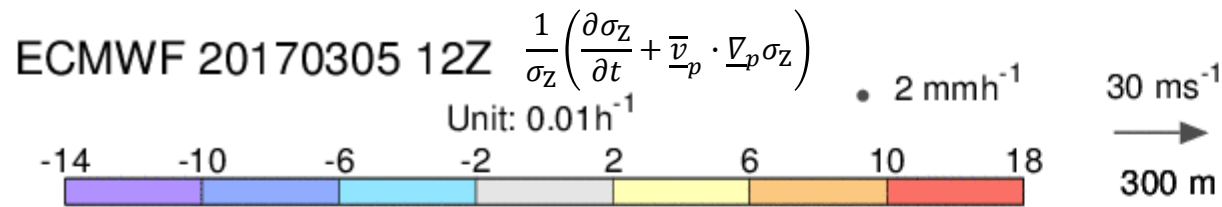


$PV_{315}=2$  &  $v_{850}$  from control forecast, precipitation is ensemble-mean. 1d running-mean gives 12h-integrated growth rate with any diurnal cycle removed. T21 smoothed

**TIGGE 12h "Lagrangian" growth rate (following Ens-mean horizontal flow) for Z<sub>250hPa</sub> (shaded)**

CF 850hPa winds (vectors), CF Z<sub>250</sub> (green contour) and EM precip (dots). 1d running-mean applied.

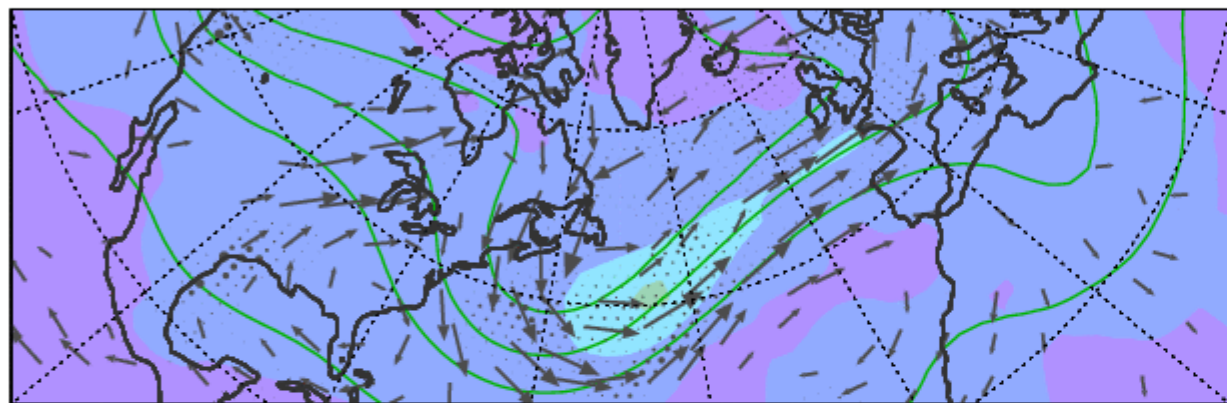
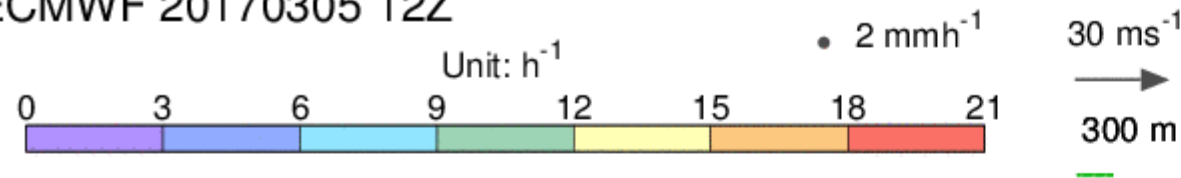
EDA(PV<sub>315K</sub>) ≈ ENS(Z<sub>250hPa</sub>) ≈ JMA.  
 UKMO: MCS less, Europe more. KMA  
 less everywhere(?) Which is best?



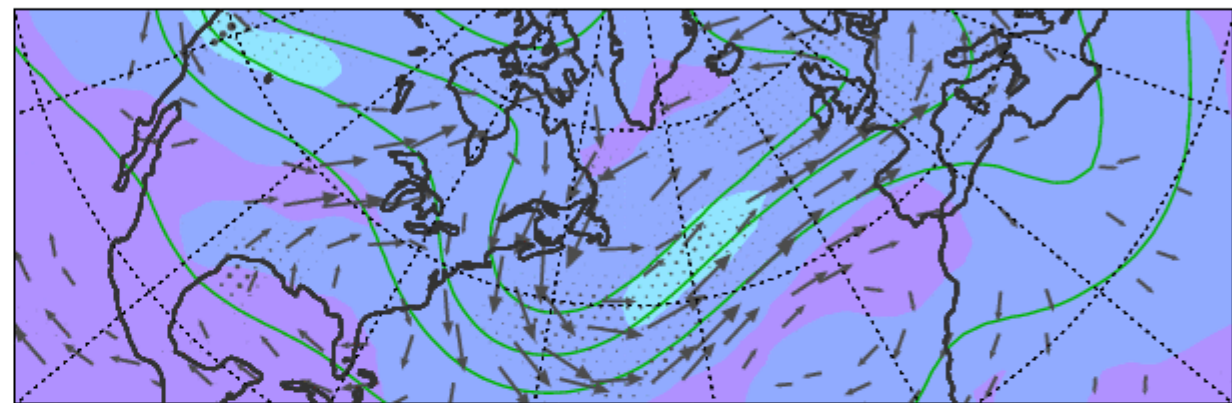
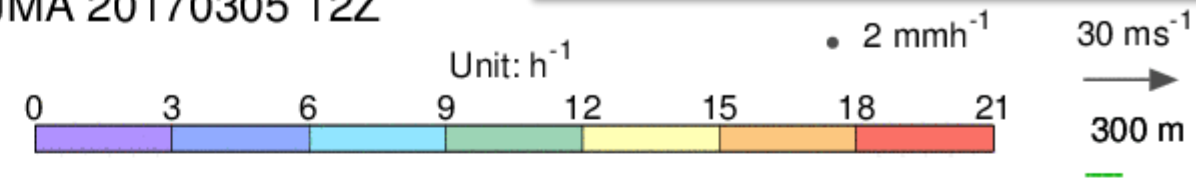
**TIGGE 12h Standard deviation in  $Z_{250\text{hPa}}$  (shaded) ~Initial uncertainty of ensemble**  
CF 850hPa winds (vectors), CF  $Z_{250}$  (green contour) and EM precip (dots). 1d running-mean applied.

Ensembles start from a different initial uncertainties, but growth-rate (in linear regime) should be unaffected

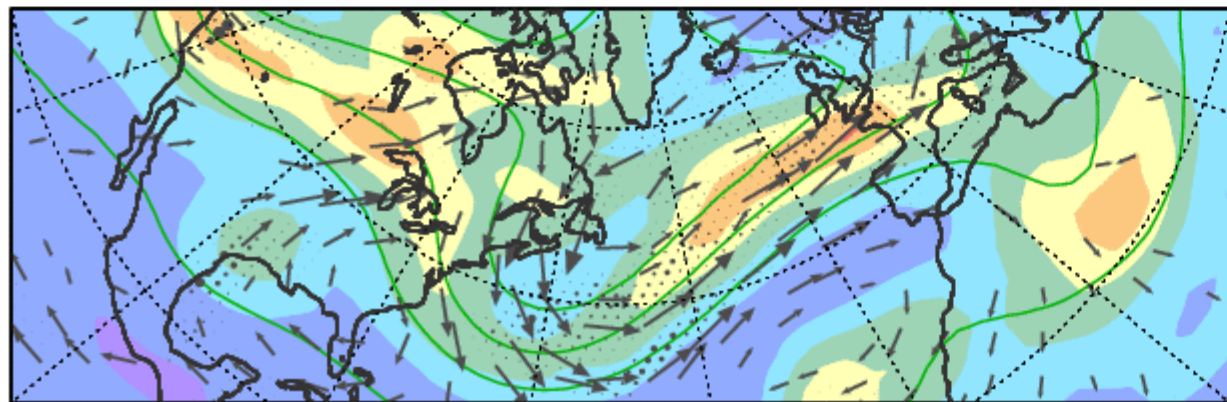
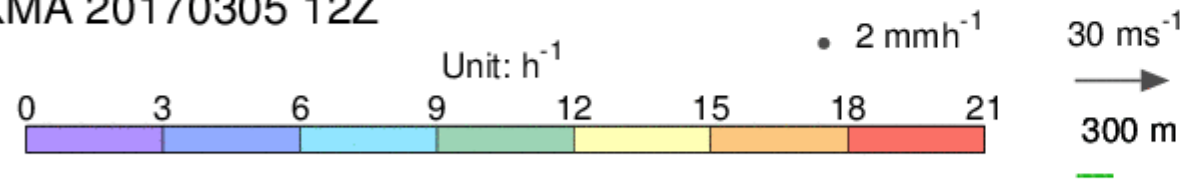
ECMWF 20170305 12Z



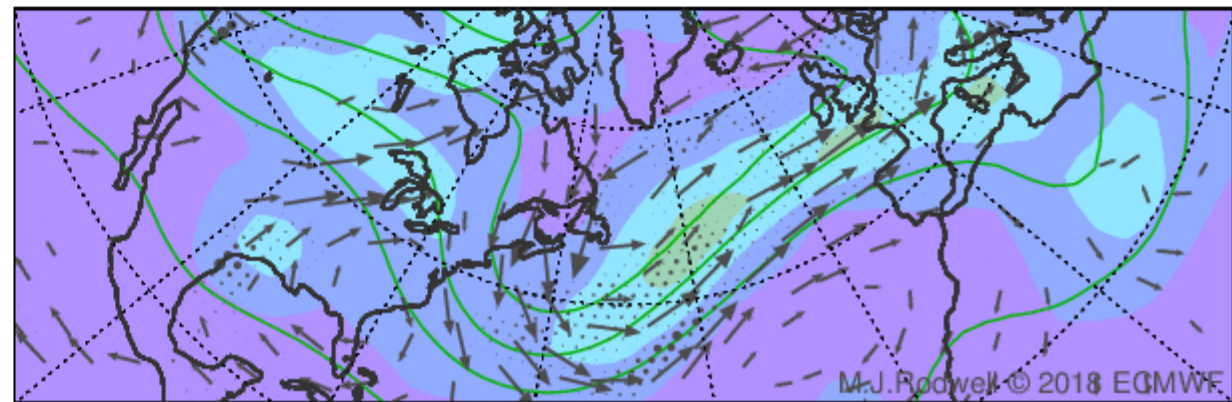
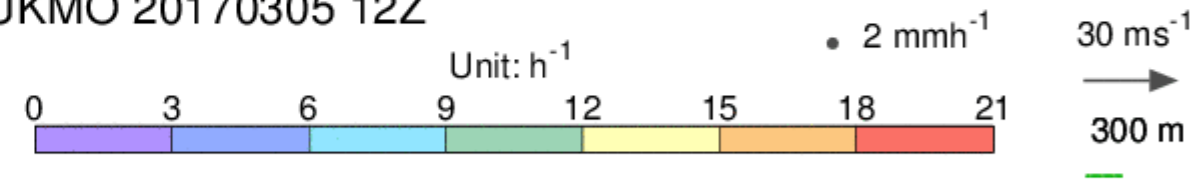
JMA 20170305 12Z



KMA 20170305 12Z

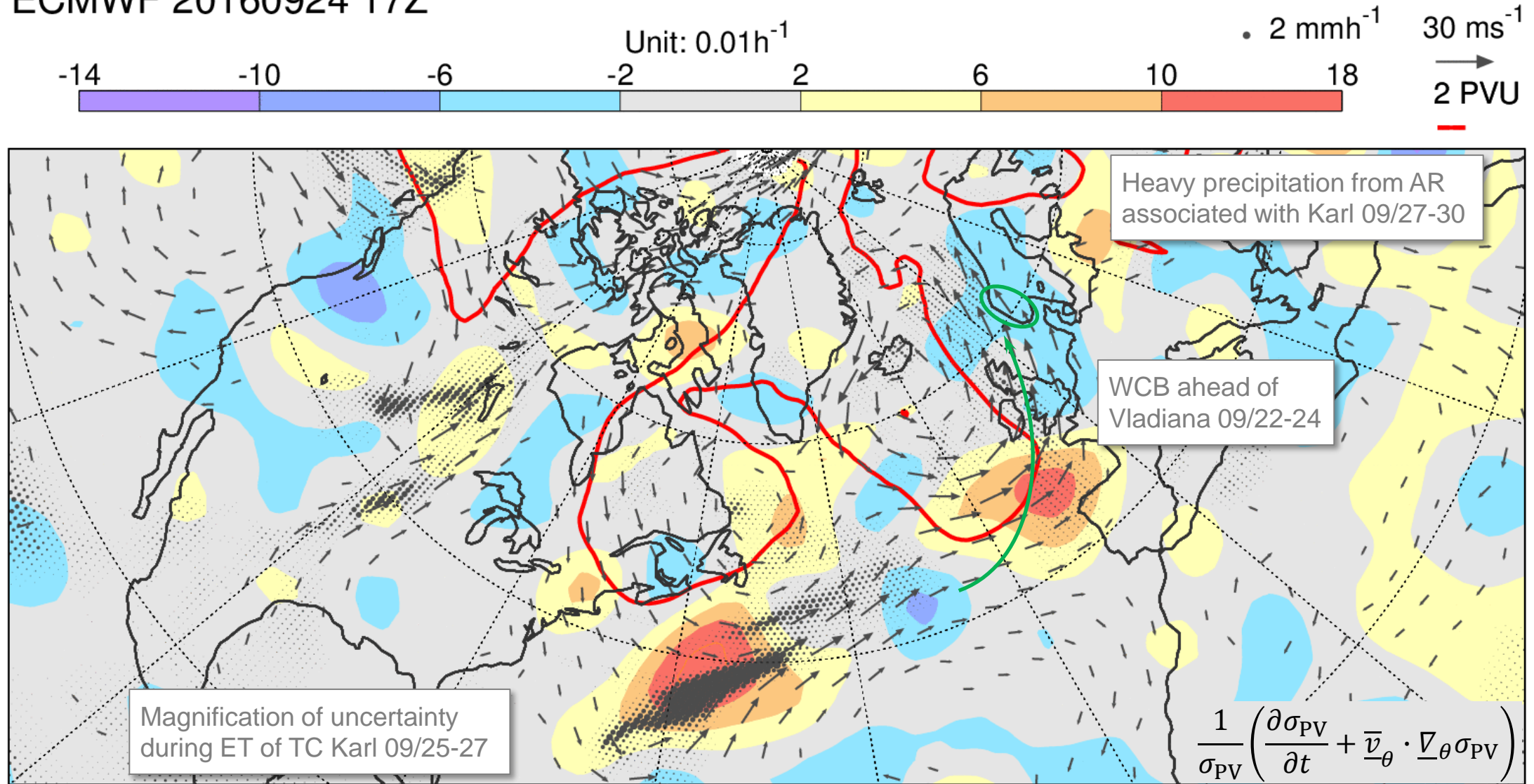


UKMO 20170305 12Z



# “Lagrangian” growth-rate for $\sigma_{PV_{315}}$ : NAWDEX Vladiana & TC Karl

ECMWF 20160924 17Z

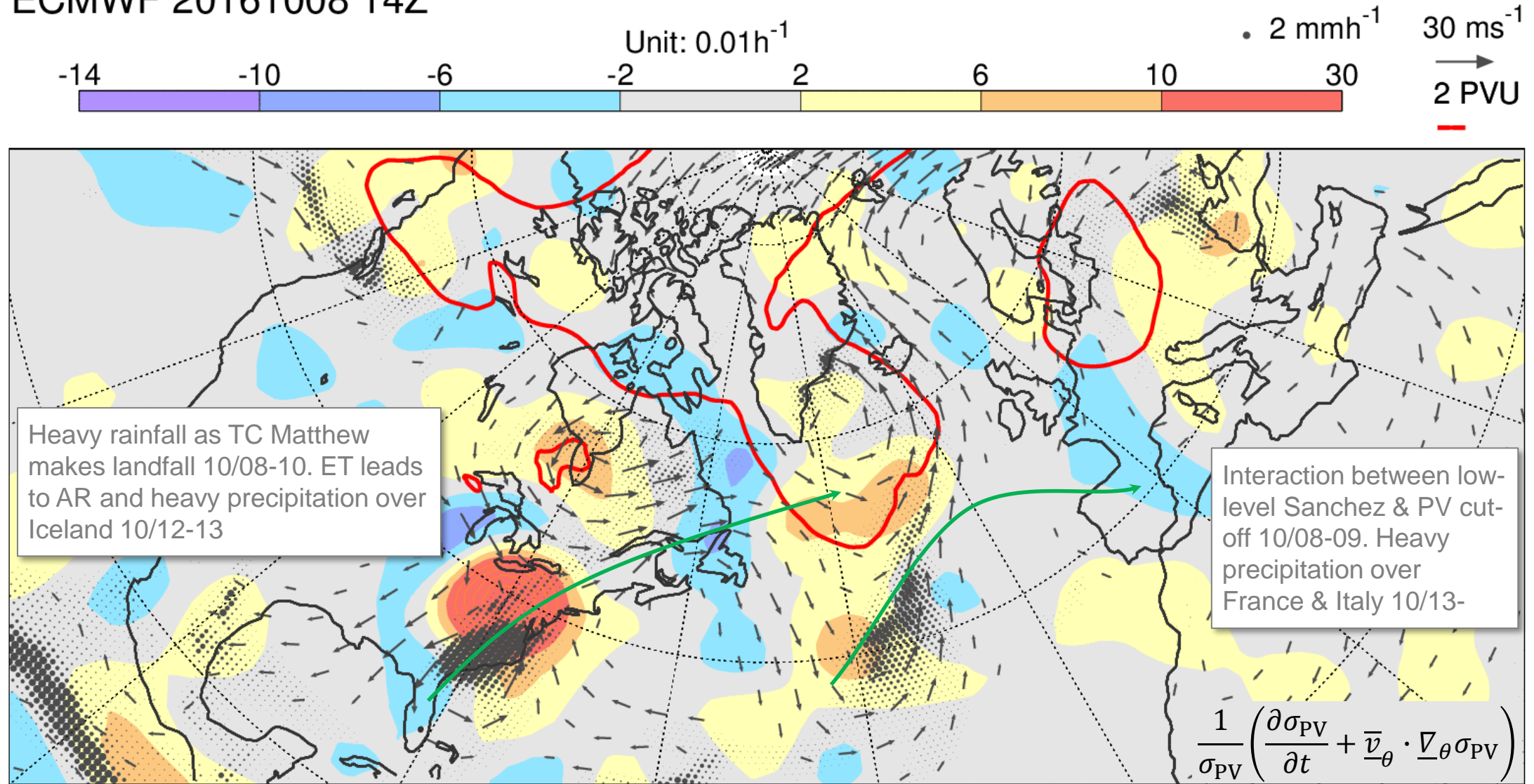


$PV_{315}=2$  &  $v_{850}$  from control forecast, precipitation is ensemble-mean. 1d running-mean gives 12h-integrated growth rate with any diurnal cycle removed. T21 smoothed



# “Lagrangian” growth-rate for $\sigma_{PV_{315}}$ : NAWDEX Sanchez

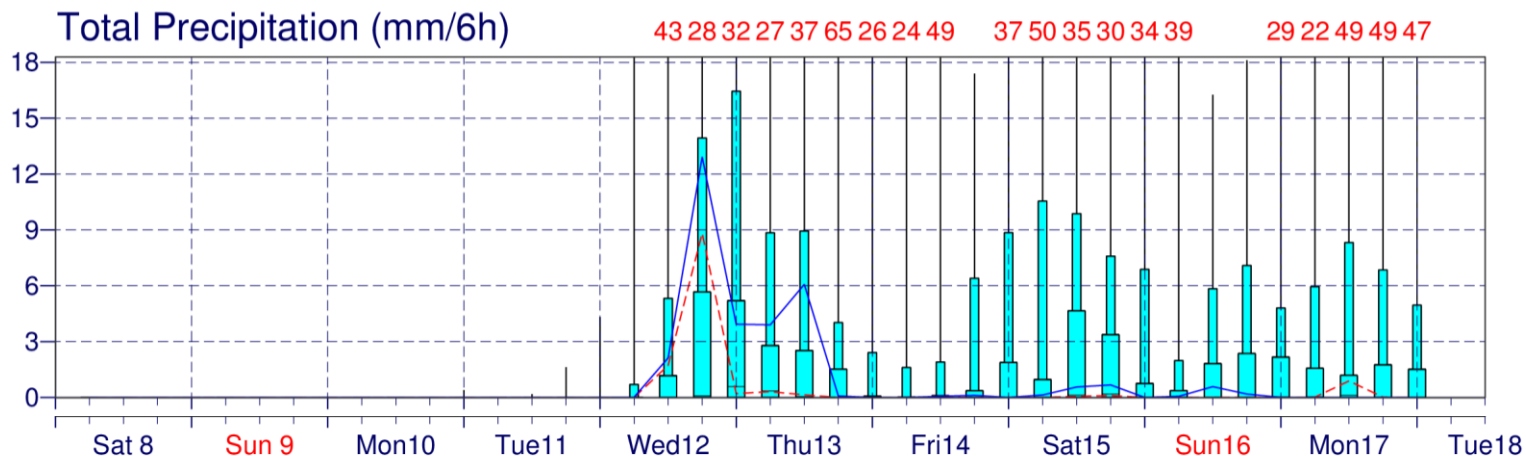
ECMWF 20161008 14Z



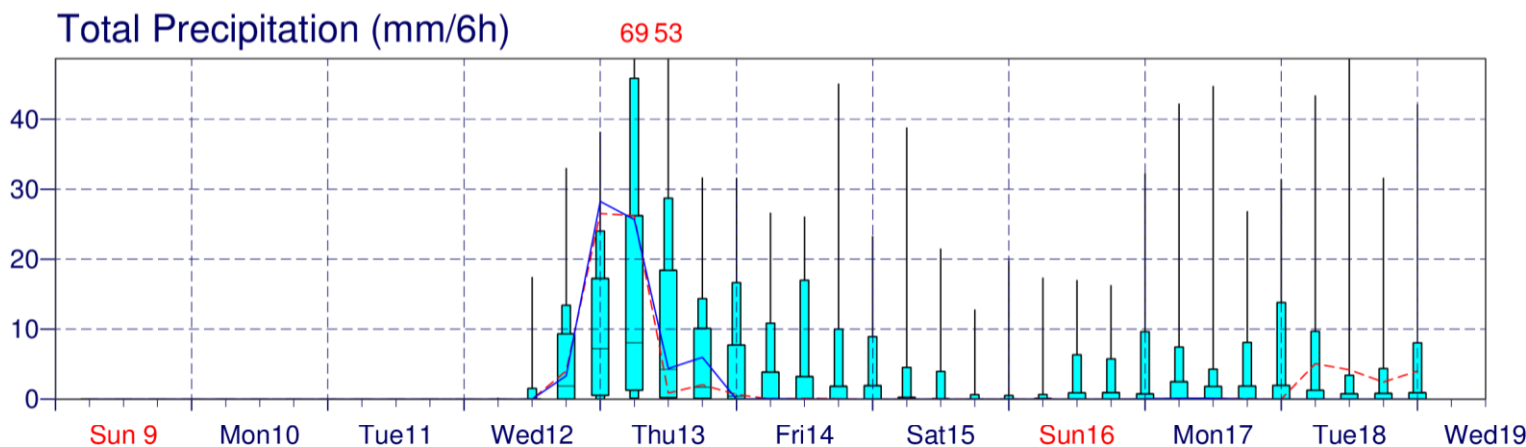
$PV_{315}=2$  &  $v_{850}$  from control forecast, precipitation is ensemble-mean. 1d running-mean gives 12h-integrated growth rate with any diurnal cycle removed. T21 smoothed

# Forecast for precipitation in Montpellier, southern France

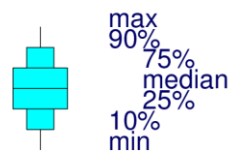
FC start 10/08



FC start 10/09



Once uncertainties associated with the interaction between Sanchez and the upper-level PV cut-off are resolved, the probability for strong precipitation over southern France firms-up. Note the different y-axes.



ENS Control(16 km) High Resolution (8 km)