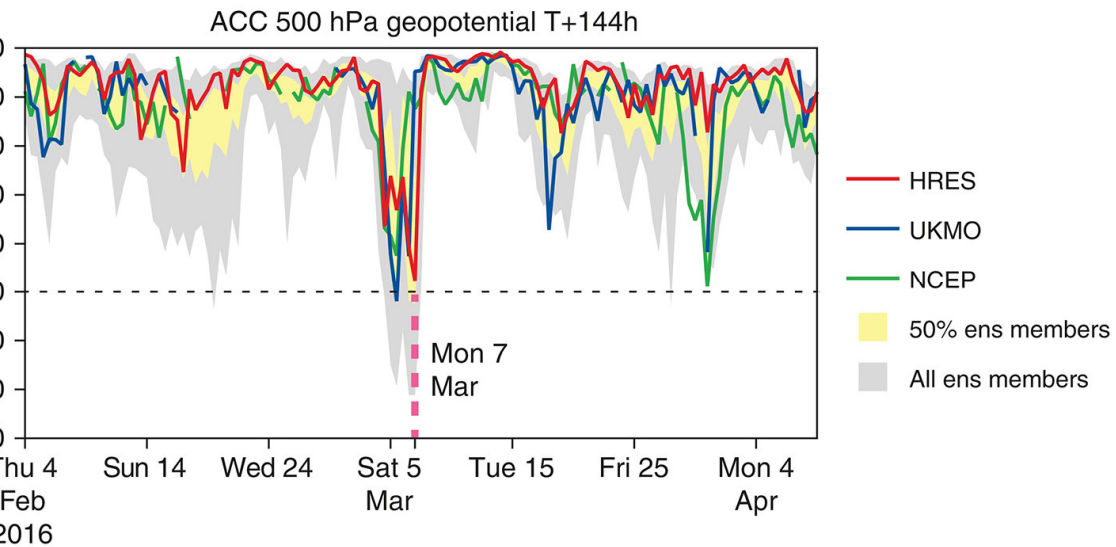


# Diagnostic methods for understanding the origin of forecast errors and uncertainties

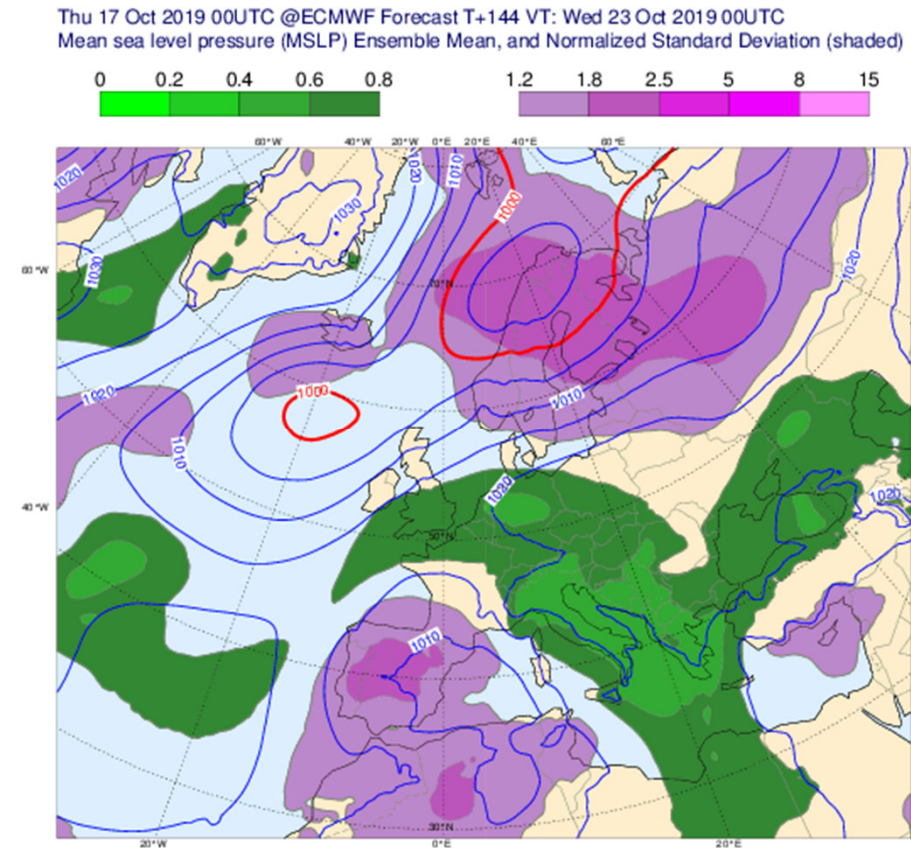
Linus Magnusson

Based on experiences from:  
Rodwell, Magnusson, et al. (2013)  
Magnusson (2017)  
Grams, Magnusson and Madonna (2018)  
Magnusson et al. (2019)

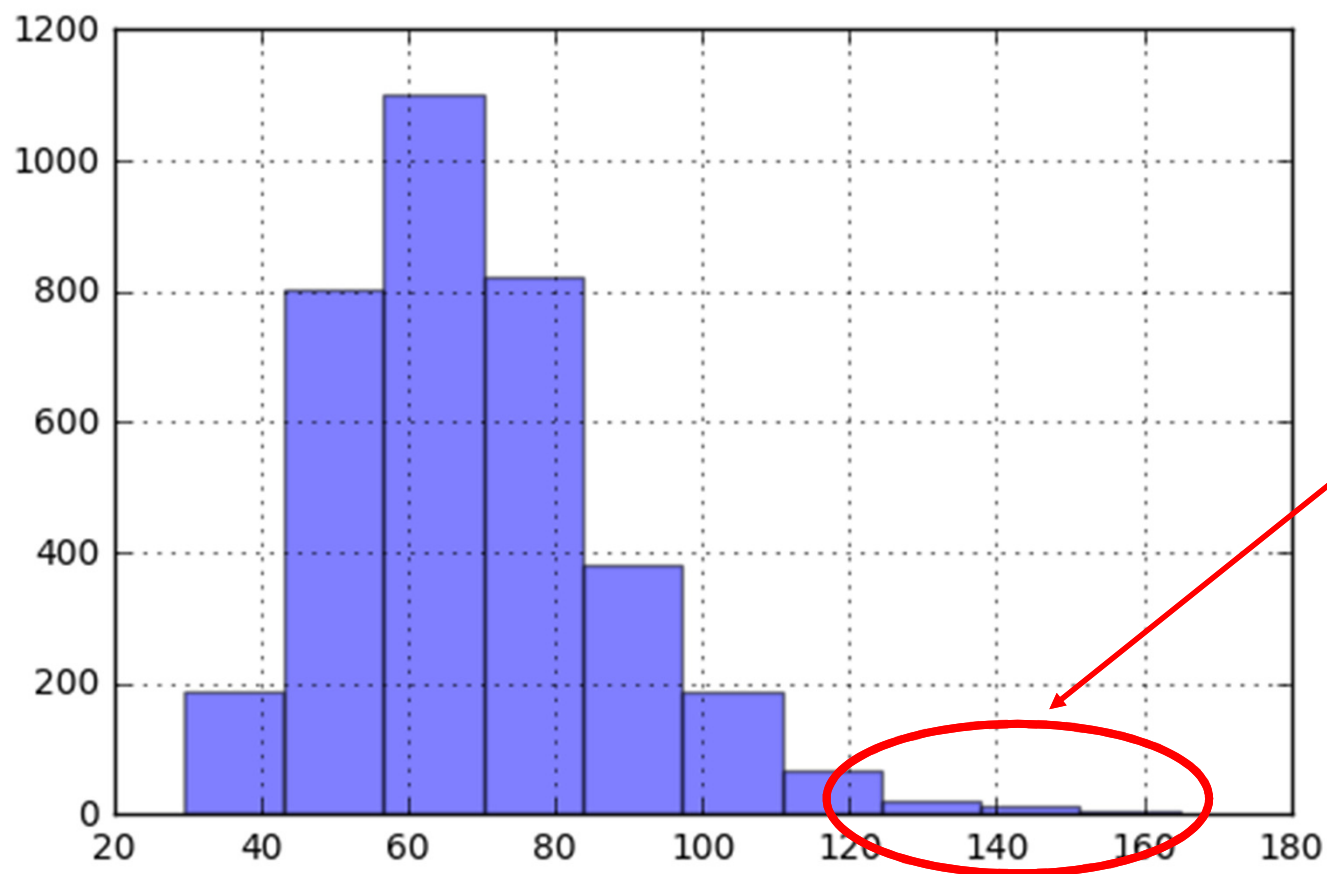
## Case of large forecast error



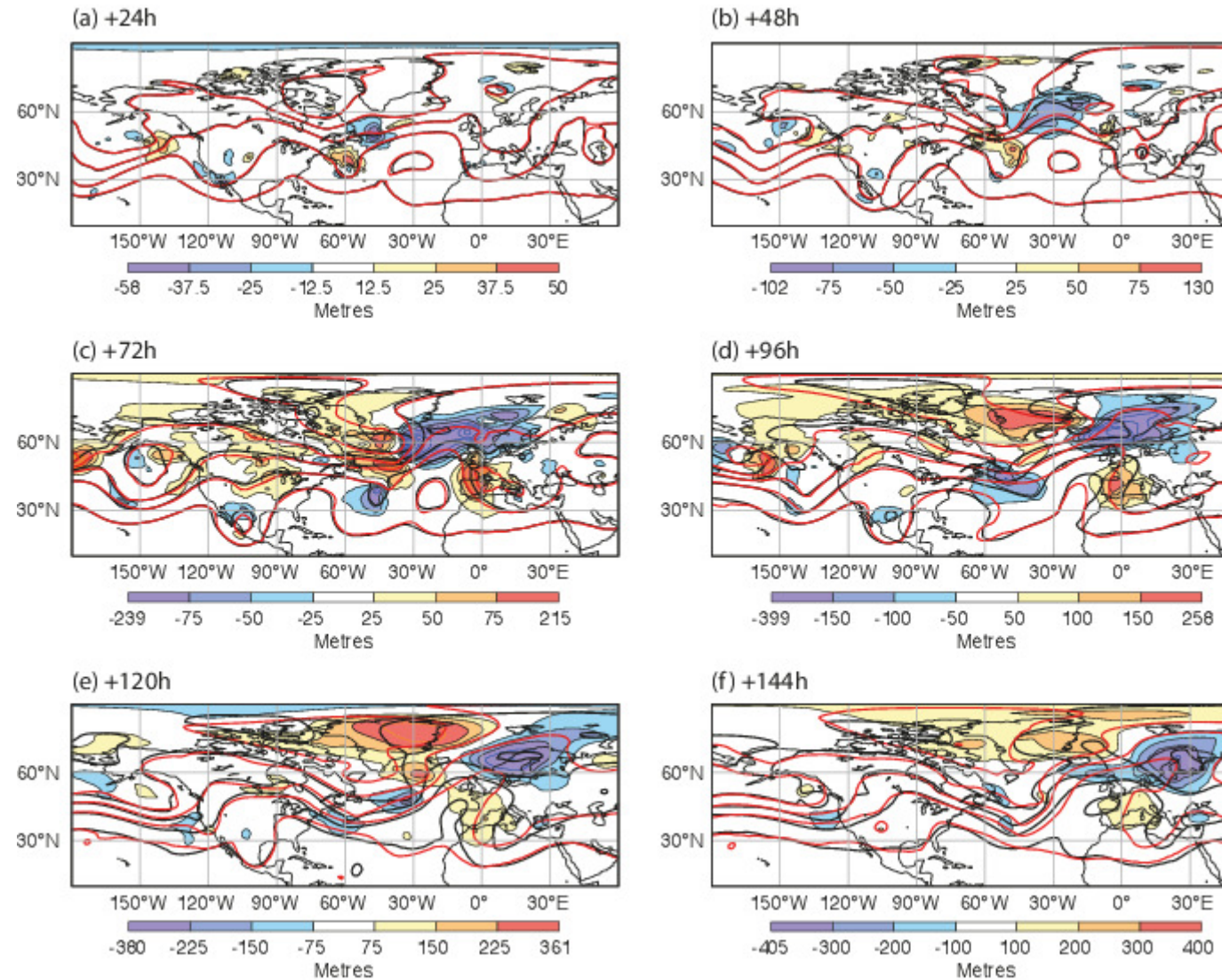
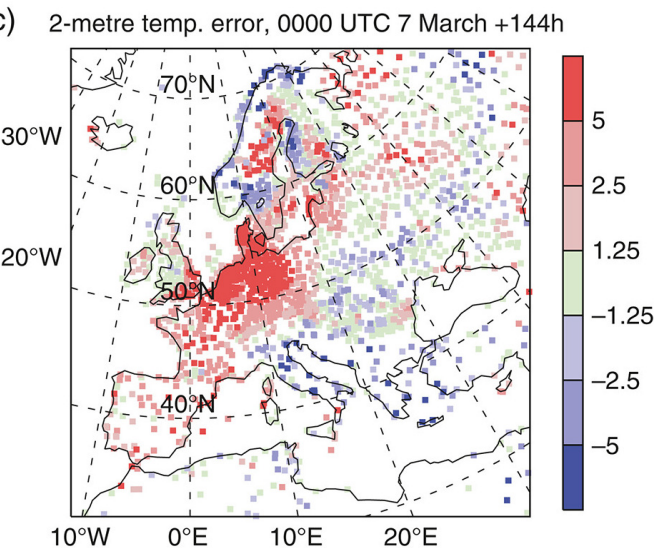
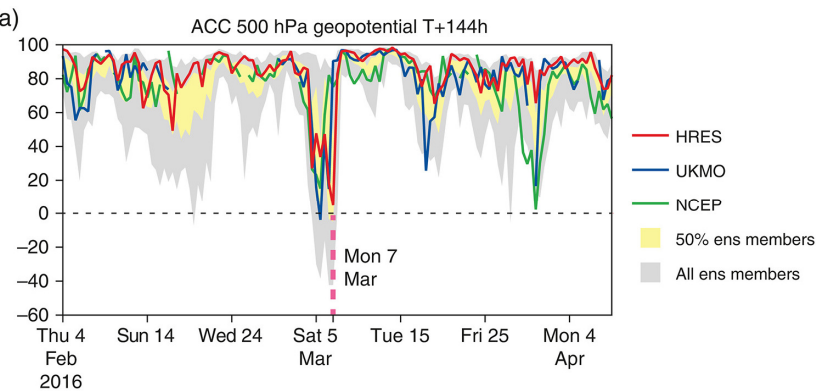
## Case of large forecast uncertainties



## Distribution of Day 6, z500 RMSE over 4 years

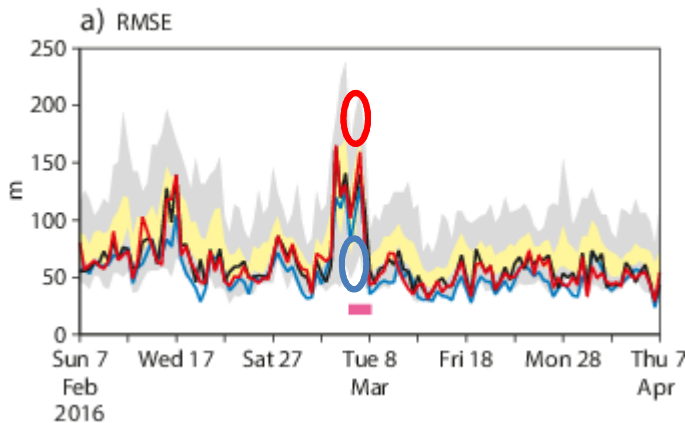


## Errors in z500 forecasts from HRES



Black – forecast, red – analysis (EC), error - shade

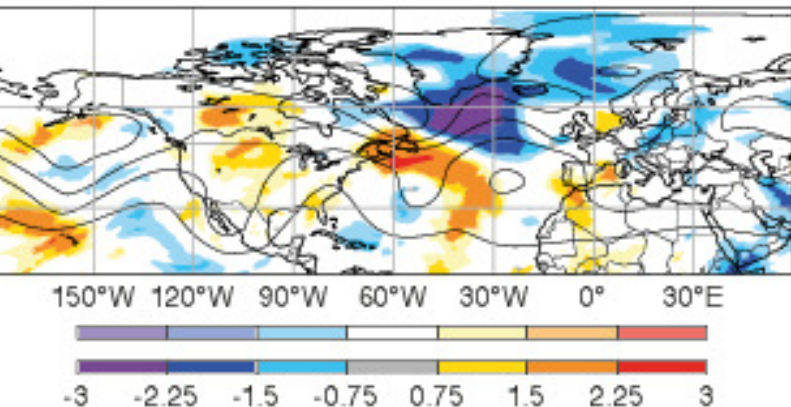
# Ensemble sensitivity – rank method



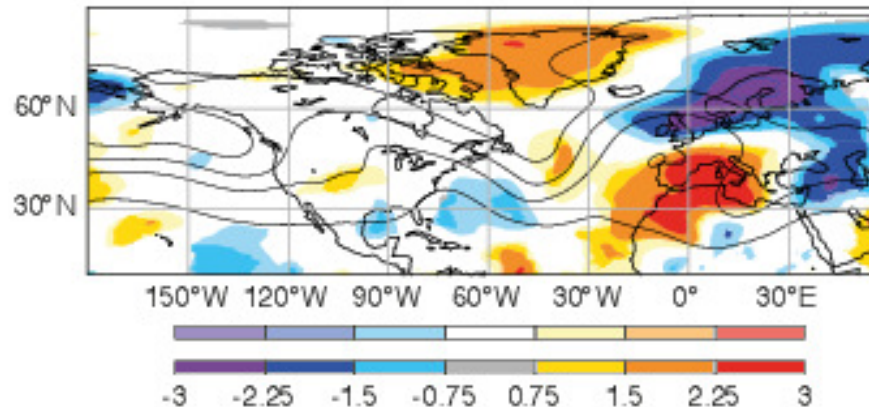
$$\frac{\text{mean(bad)} - \text{mean(good)}}{\text{stdev(all)}}$$

Z500 ensemble sensitivity

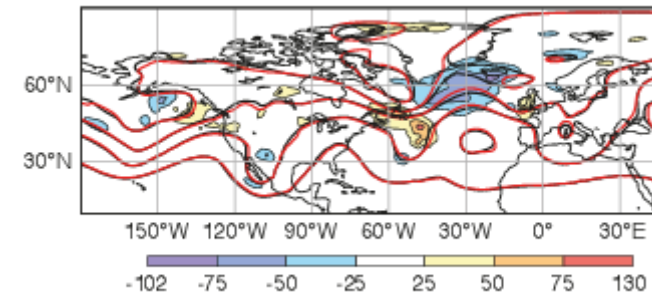
(c) Sens. Rank +48h



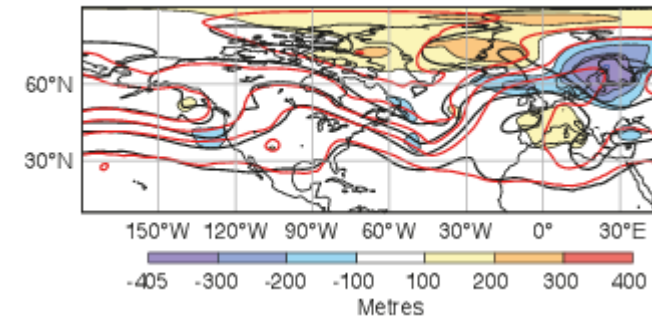
(d) Sens. Rank +144h



(b) +48h



(f) +144h

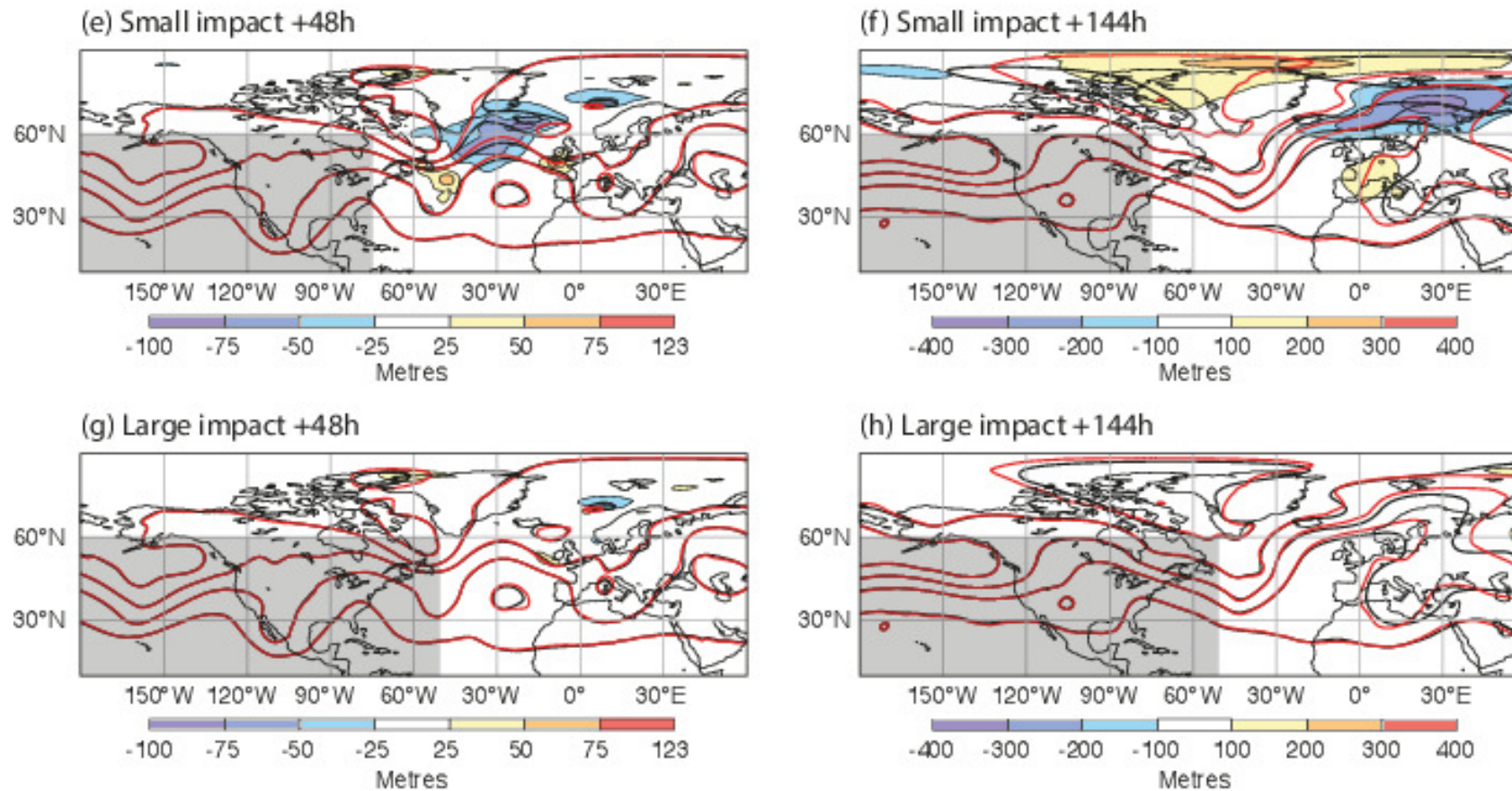


See:

Ancell and Hakim (2000)  
 Torn and Hakim (2000)  
 Zheng et al. (2013)  
 Torn et al. (2015)  
 Lamberson et al. (2015)  
 Magnusson (2017)

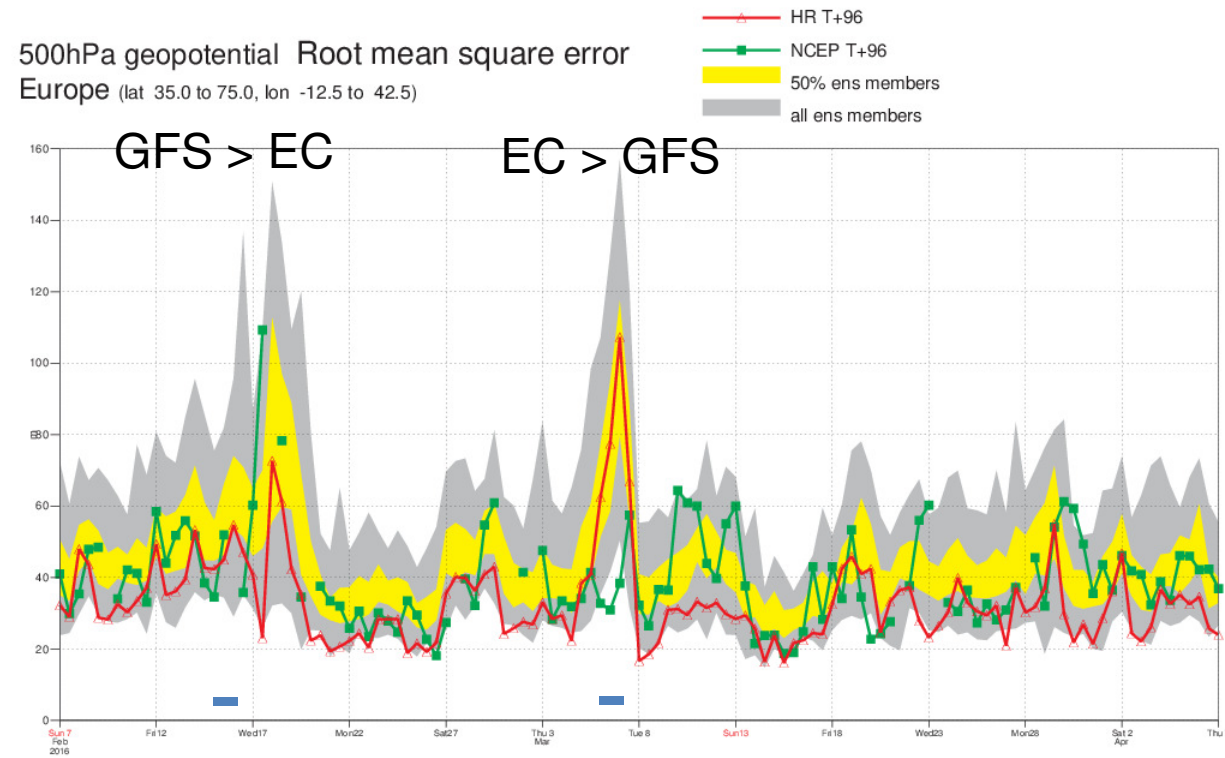
## Errors in relaxation experiments

$$-\lambda (X - X_{\text{ref}}) \quad 20 \text{ ensemble members}$$



Black – ensemble mean, red – analysis, error - shade

# RMSE, day 4, Europe, February-April 2016



## Comparison between FV3, GFS and IFS

- Initialisation scheme developed at GFDL to initialise FV3 from GFS and IFS initial conditions (Chen et al., AMS Hurricanes 2017)
- Create opportunity to compare forecasts with “same” initial conditions but different models:  
Compare model biases, predictability, etc

	GFS	FV3, 2018 GFDL version	IFS
Dynamical core	Spectral, Hydrostatic, S-L advection	Finite-volume, Non-hydrostatic, cube-square grid	Spectral, Hydrostatic, S-L advection
Physics	GFS	Modified GFS (GFDL microphysics, YSU PBL scheme, mixed-layer ocean model)	IFS
Resolution	13 km / 64 levels	13 km / 91 levels	9 km / 137 levels
Initial conditions	GFS	GFS, IFS	IFS

Same initial condition

Same model

Same initial condition

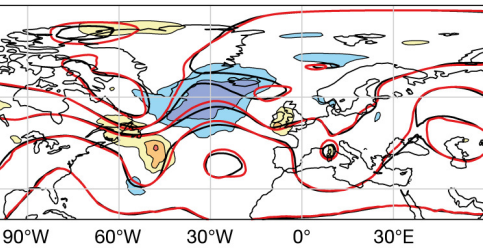
1-year sample: Initialised every 5<sup>th</sup> day from 15 August 2015 to 15 August 2016  
Hurricane season: Initialised every 12<sup>th</sup> hour from 1 August to 30 October 2017

# Z500 error pattern for bust case

ECMWF initial conditions

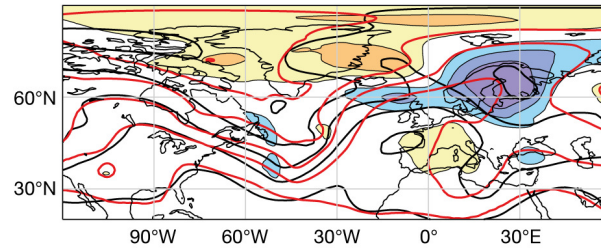
2-day error

+ 2 days



6-day error

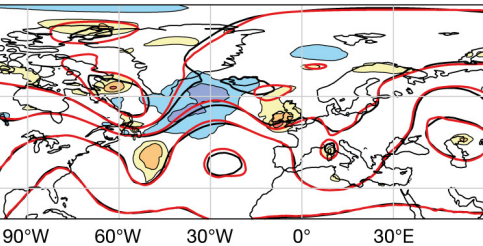
**b** EC + 6 days



GFS initial conditions

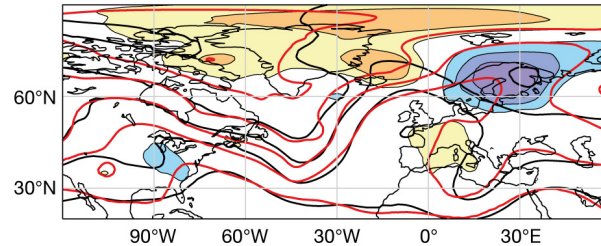
2-day error

ec + 2 days

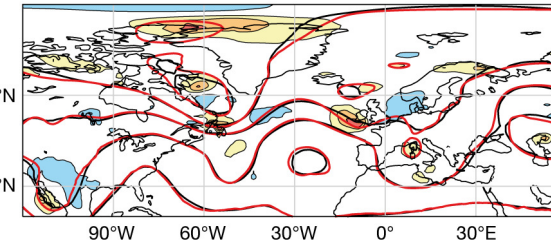


6-day error

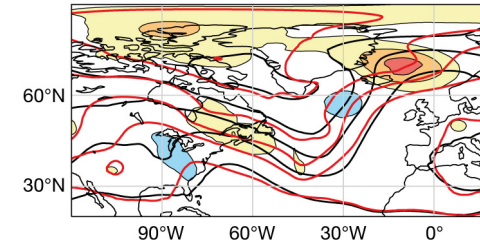
**d** FV3ec + 6 days



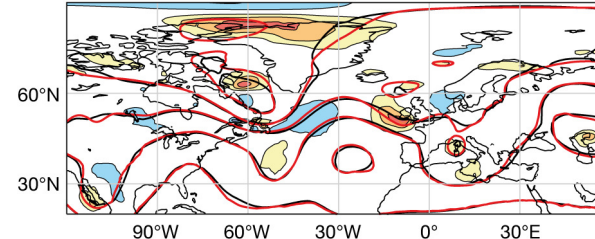
**e** FV3gfs + 2 days



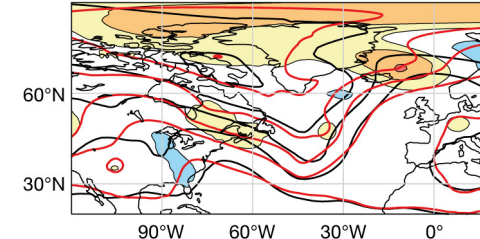
**f** FV3gfs + 6 days



**g** GFS + 2 days



**h** GFS + 6 days

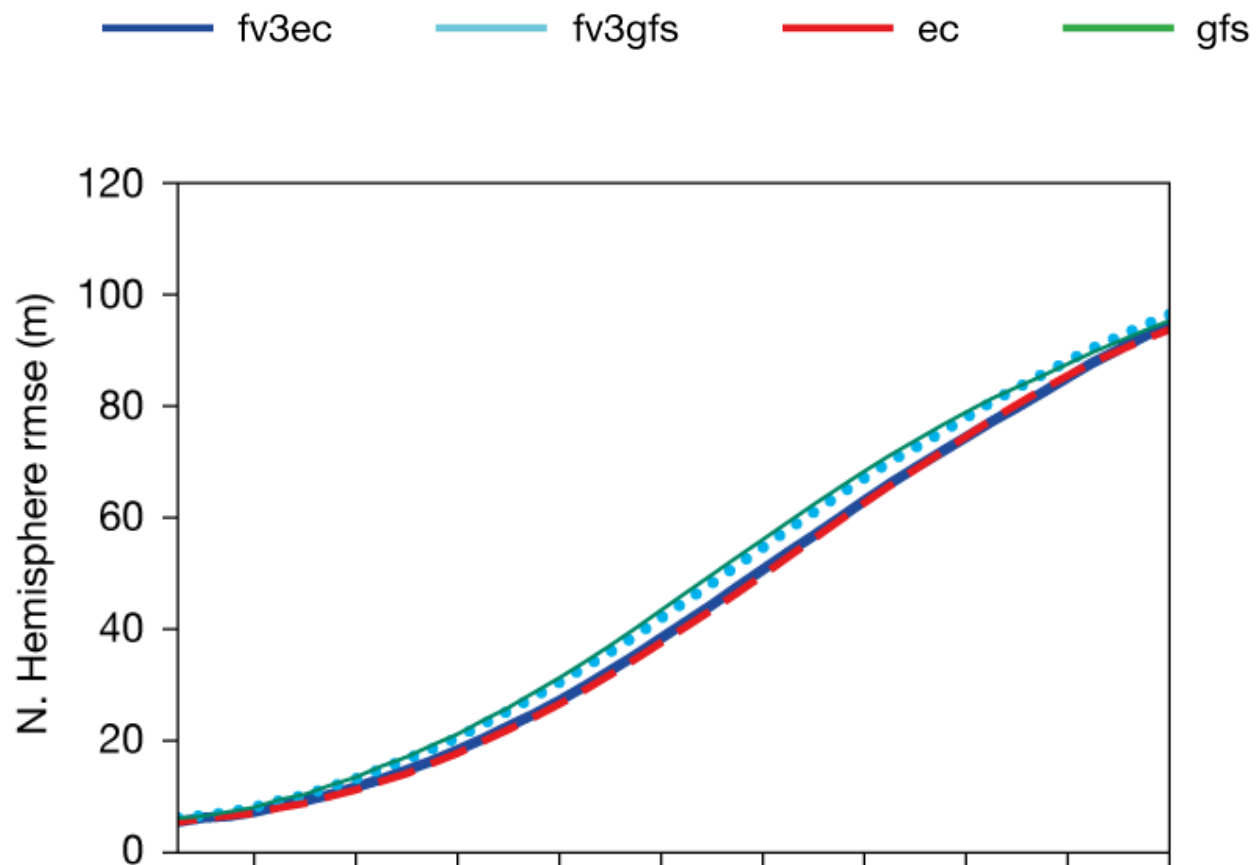


-100 -75 -50 -25 25 50 75 100  
(metres)

-400 -300 -200 -100 100 200  
(metres)

# MSE for z500, Northern hemisphere

1 year sampled every 5<sup>th</sup> day, Verified against UKMO analysis



Magnusson et. al. (2019, Q.

# Tracking uncertainties in ensemble forecasts

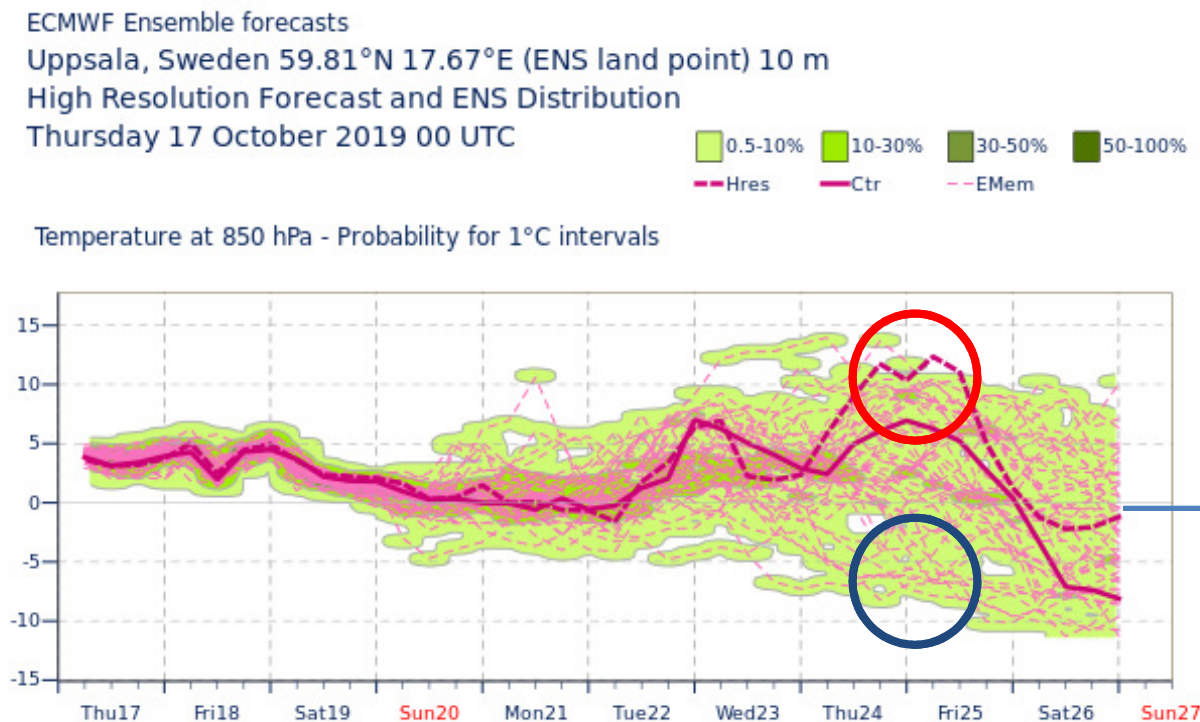
## - Use of ensemble sensitivities

- From where originates the uncertainties in the forecast? When can we expect the uncertainties to be eliminated?

$$\frac{\text{mean(upper tail)} - \text{mean(lower tail)}}{\text{stdev(all)}}$$

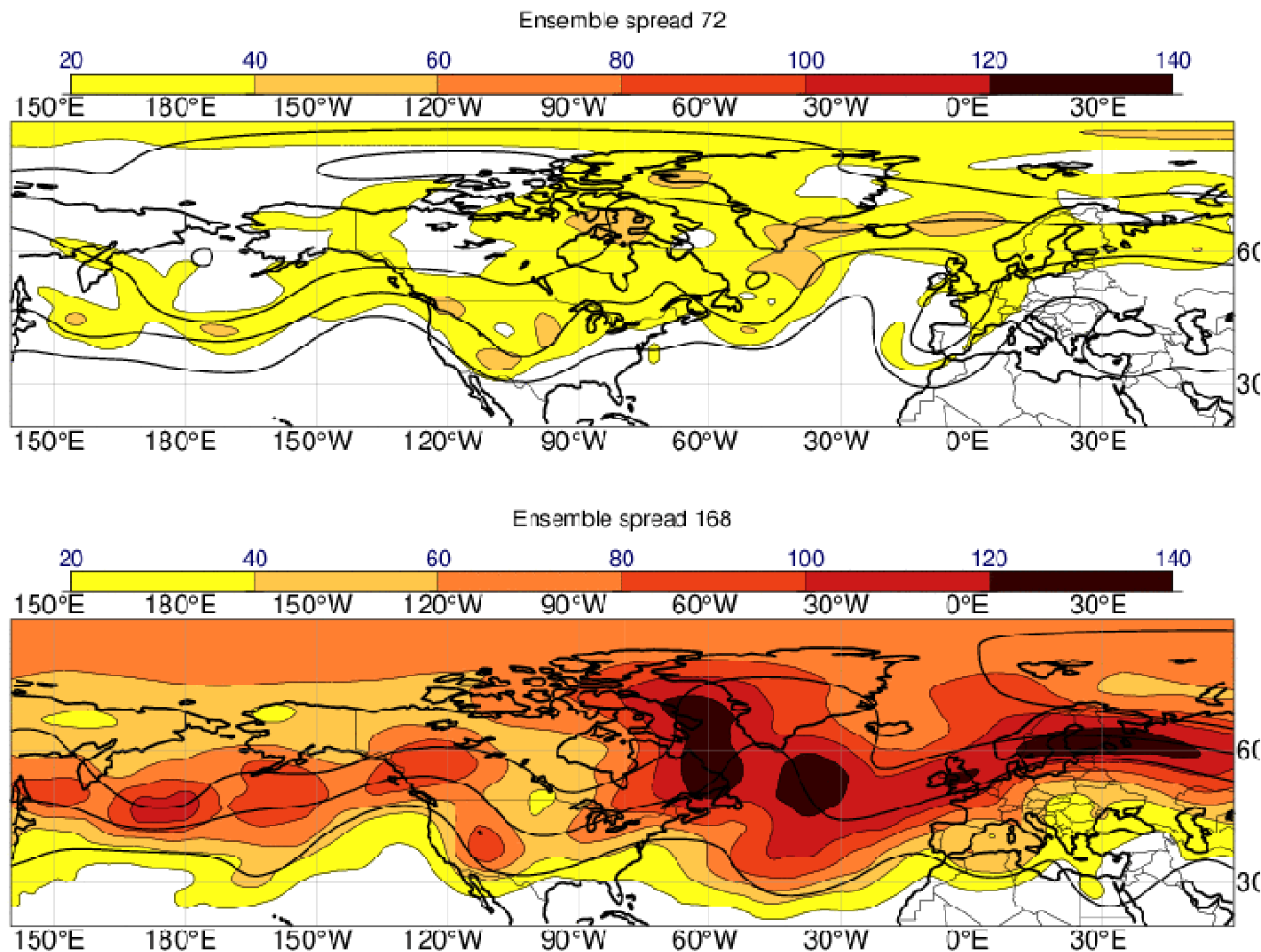
For example:

Mean of the 5 coldest members minus the mean of the 5 warmest members



Map into another variable, are the groups separated?

# Z500 ensemble spread in forecast from 17 Oct 12UTC + 18 Oct 12UTC



$$\frac{\text{mean(upper tail)} - \text{mean(lower tail)}}{\text{stdev(all)}}$$

Define a metric e.g:

Temperature in Uppsala

Forecast error (in hindsight)

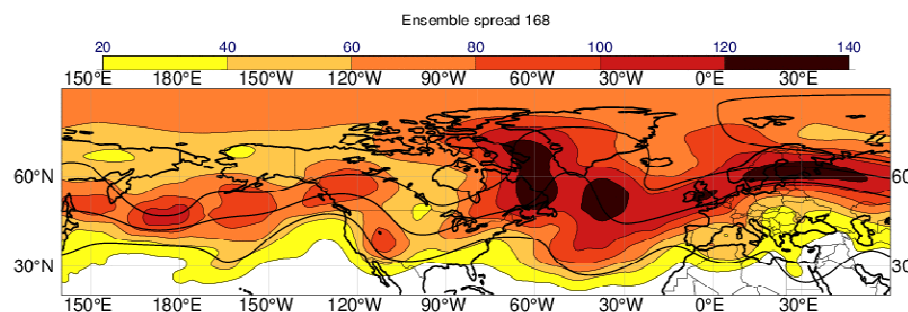
Tropical cyclone track

....

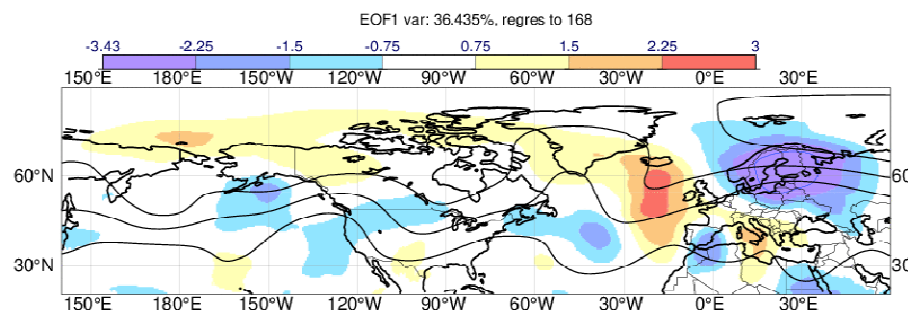
Projection on leading spread EOF

# 18 October 00UTC +168h and 17 October 12UTC +190h

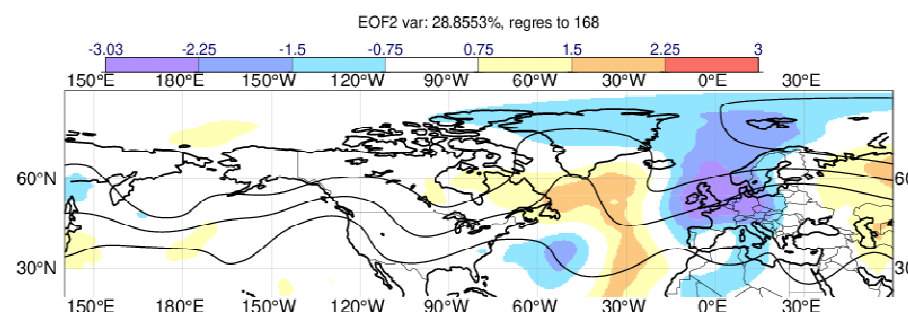
spread



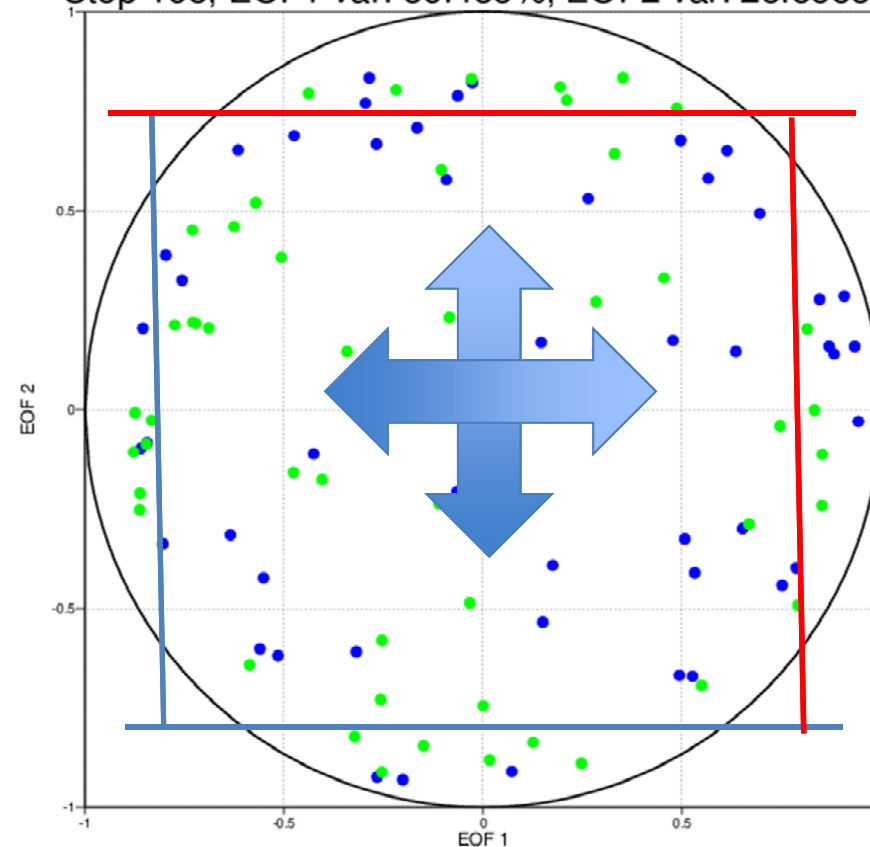
ster  
ference  
F1



ster  
ference  
F2



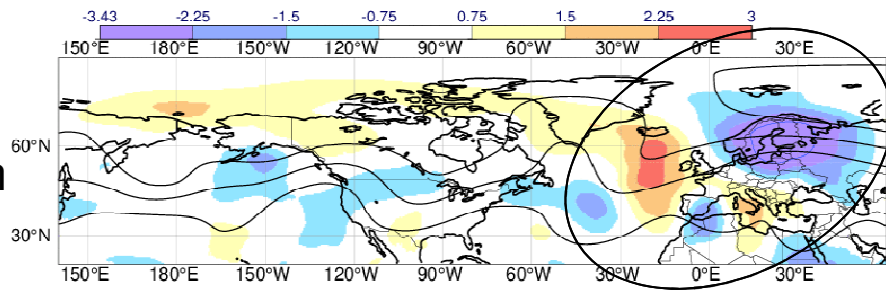
Step 168, EOF1 var: 36.435%, EOF2 var: 28.8553%



# Ensemble sensitivities for different lead times

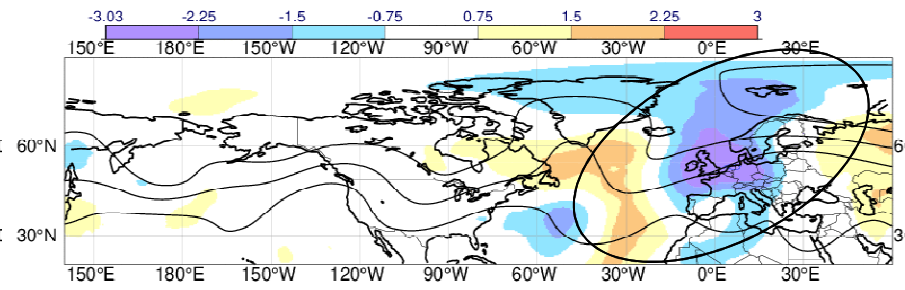
## EOF1

EOF1 var: 36.435%, regres to 168

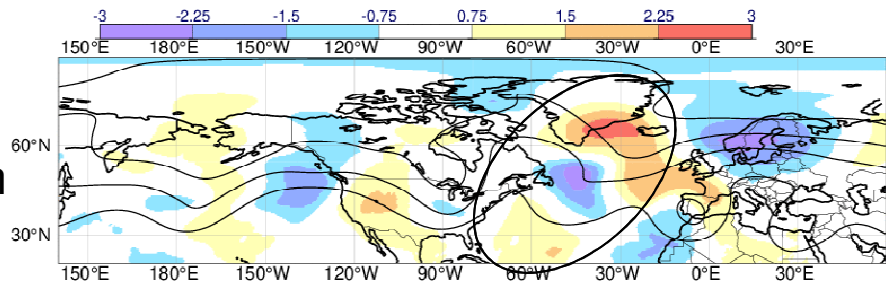


## EOF2

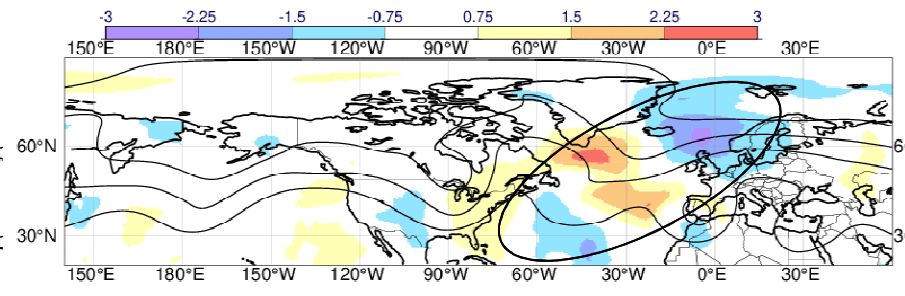
EOF2 var: 28.8553%, regres to 168



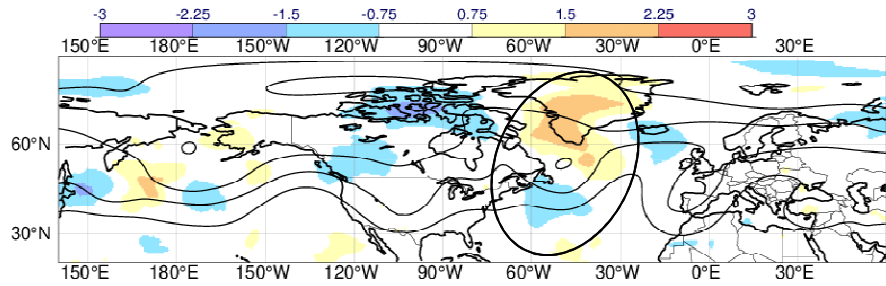
EOF1 var: 36.435%, regres to 120



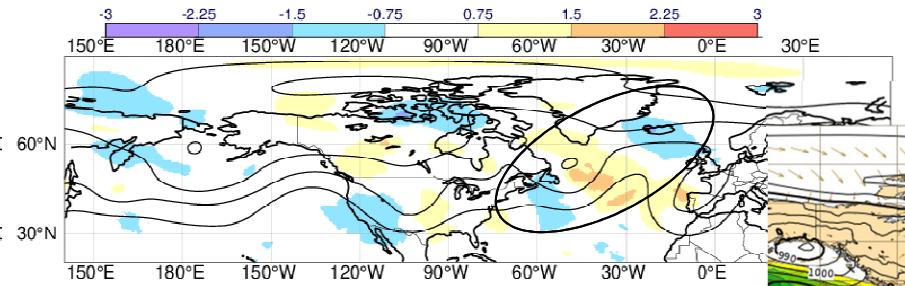
EOF2 var: 28.8553%, regres to 120



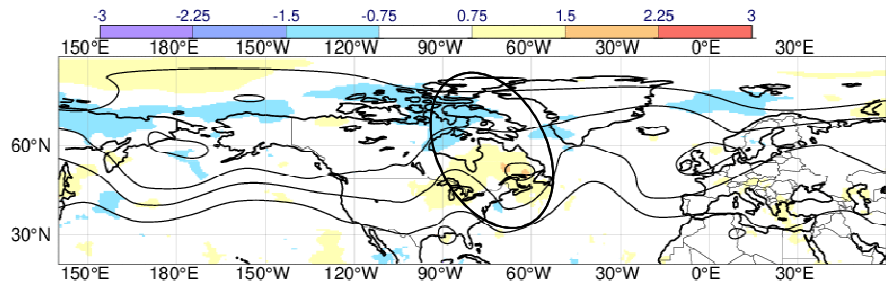
EOF1 var: 36.435%, regres to 72



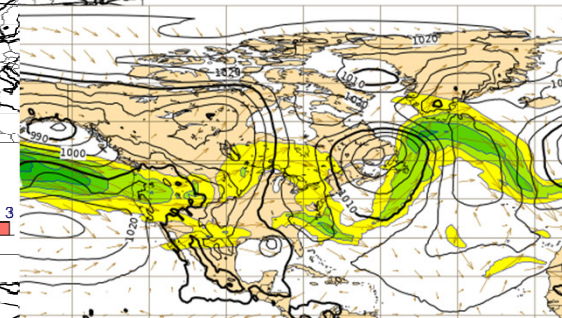
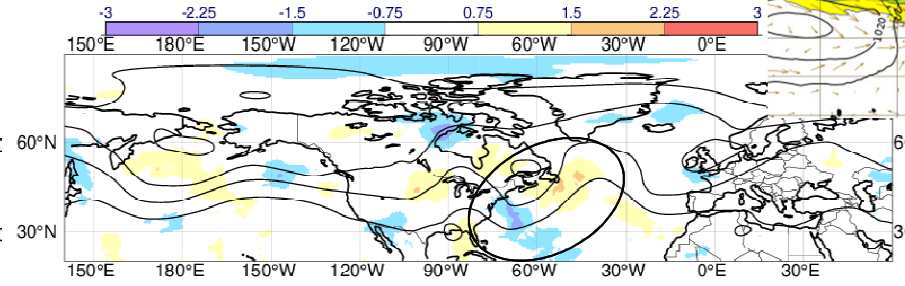
EOF2 var: 28.8553%, regres to 72



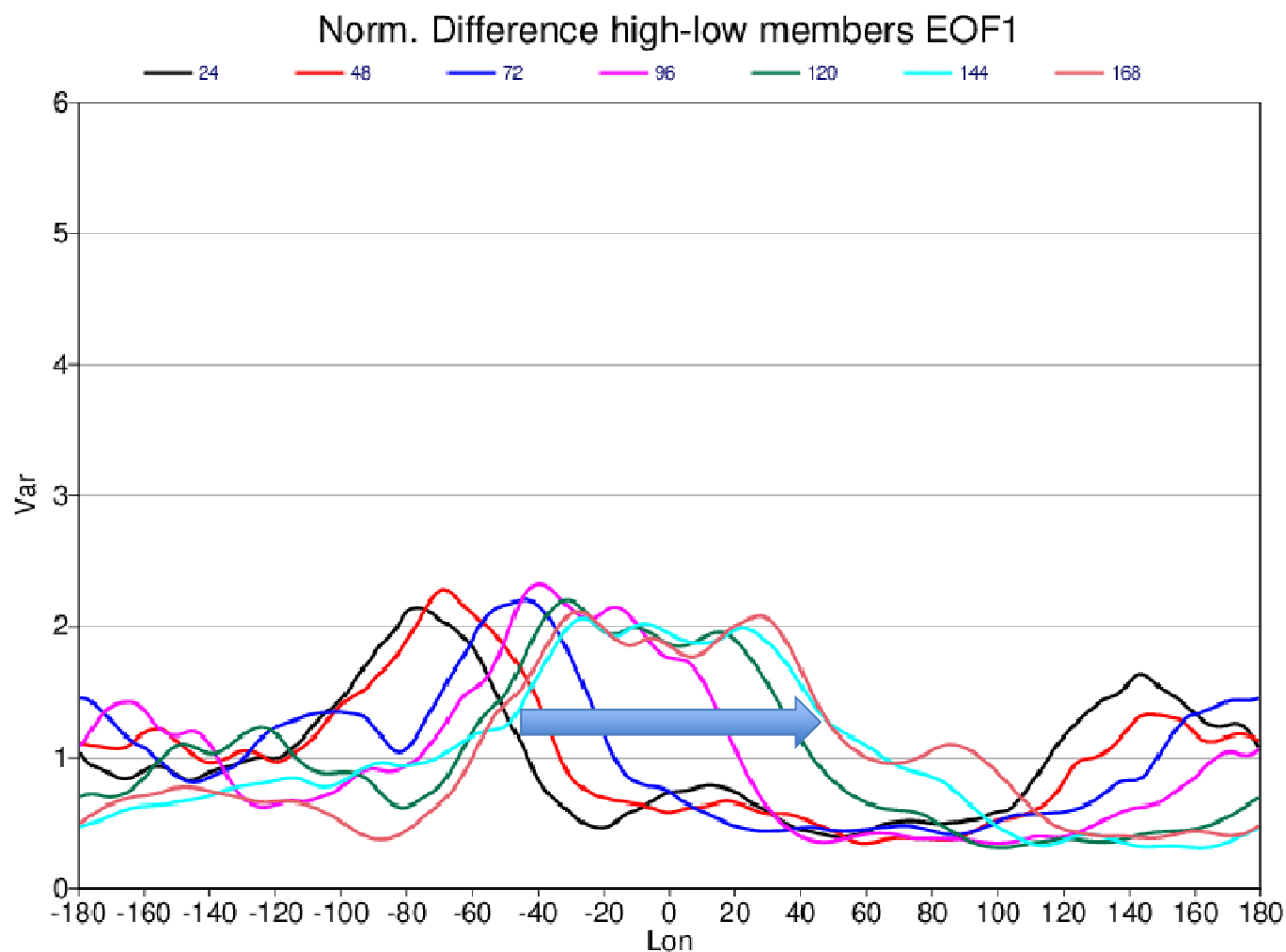
EOF1 var: 36.435%, regres to 24



EOF2 var: 28.8553%, regres to 24



# Meridional propagation of ensemble sensitivities



## Summary ensemble sensitivities

1. Select function for calculating the sensitivities (rank method, correlation, linear regression)
2. Define final metric (e.g EOFs of ensemble spread)
3. Select parameter to follow the sensitivities

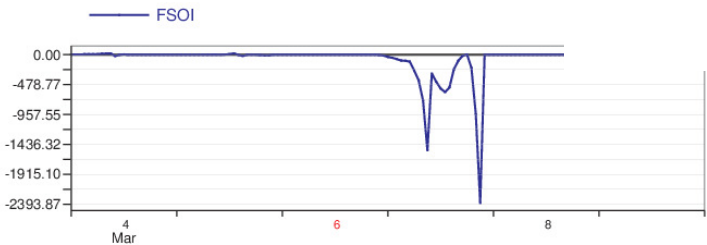


# Observation impact

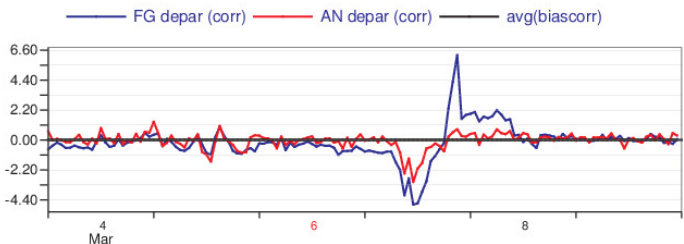
## One SYNOP

Surface pressure (hPa) from station ID 4  
Active data, EXP =0001

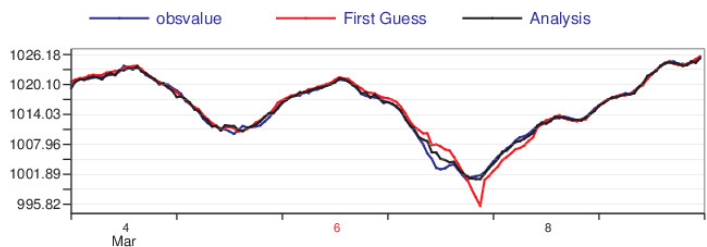
FSOI



Departures



Values



## MSLP and surface heat-flux

20160307 0z +12 Net upward heat flux

