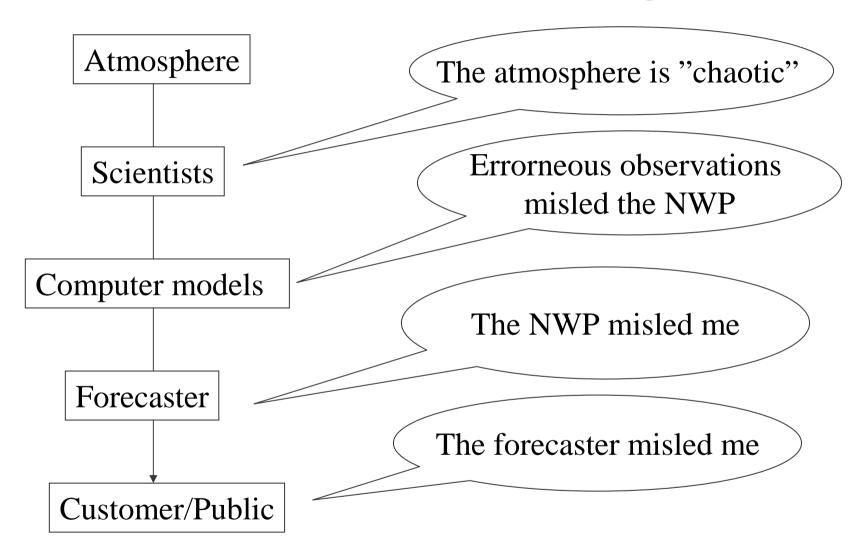
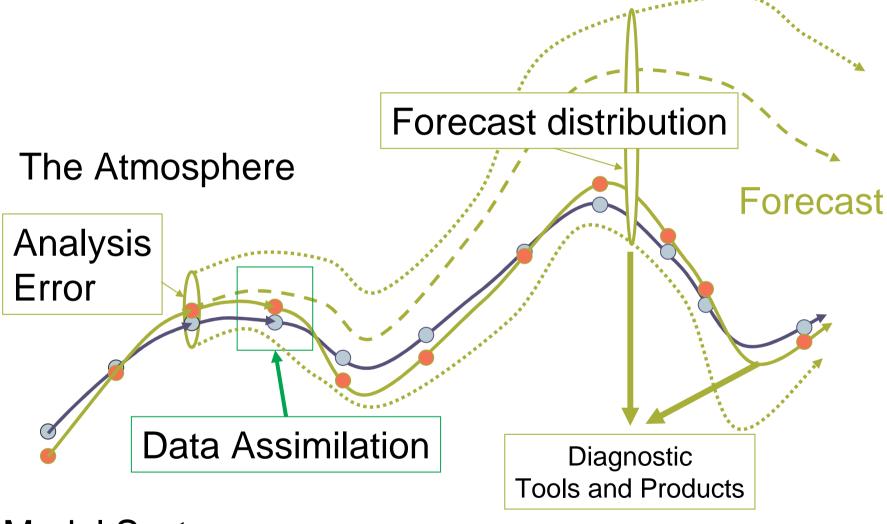
## Ensemble Prediction Systems and Probabilistic Forecasting

#### "The Blame Game" or "The Passing of The Buck"



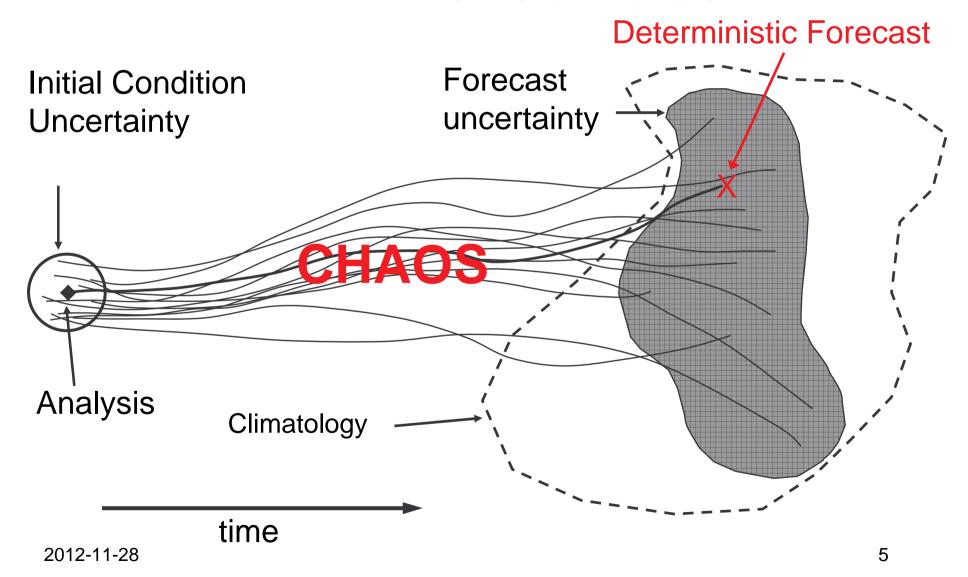
## Understanding chaos

#### **Numerical Weather Prediction**



Model System

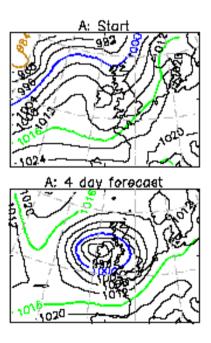
## Quantifying uncertainty with ensembles

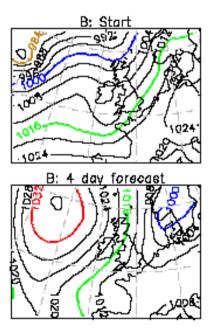


#### The Effect of Chaos

- We can *usually* forecast the general pattern of the weather up to about 3 days ahead.
- Chaos then becomes a major factor

Tiny errors in our analysis of the current state of the atmosphere lead to large errors in the forecast – these are both equally valid 4-day forecasts.

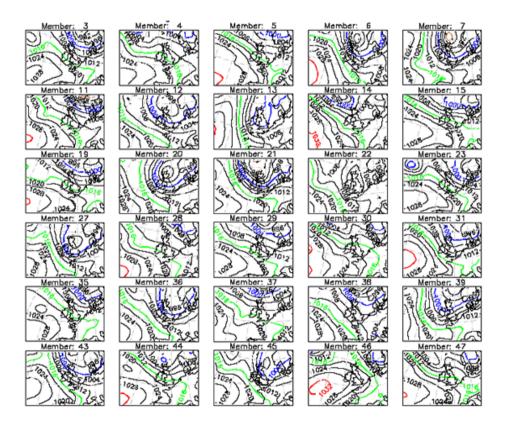




Fine details (eg rainfall) have shorter predictability

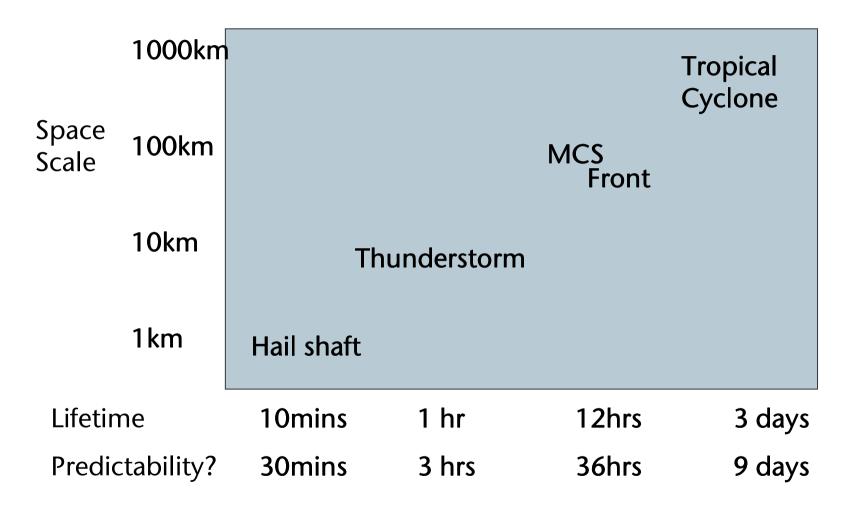
#### **Ensembles**

- In an ensemble forecast we run the model many times from slightly different initial conditions
- This provides a range of likely forecast solutions which allows forecasters to:
  - assess possible outcomes;
  - estimate risks
  - gauge confidence.



# Reminder on scales and predictability

### **Temporal Resolution**



### **MOGREPS**

#### Short-range Ensembles

## ECMWF EPS has transformed the way we do Medium-Range Forecasting

- Uncertainty also in short-range:
  - Rapid Cyclogenesis often poorly forecast deterministically
  - Uncertainty of sub-synoptic systems (eg thunderstorms)
  - Many customers most interested in short-range
- Assess ability to estimate uncertainty in local weather
  - QPF
  - Cloud Ceiling, Fog
  - Winds etc

#### Initial conditions perturbations

- Perturbations centred around 4D-Var analysis
- Transforms calculated using same set of observations as used in 4D-Var (including all satellite obs) within +/- 3 hours of data time
- Ensemble uses 12 hour cycle (data assimilation uses 6 hour cycle)

#### Initial conditions perturbations

#### **Differences with ECWMF Singular Vectors:**

- It focuses on errors growing during the assimilation period, not growing period:
  - Suitable for Short-range!
- Calculated using the same resolution than the forecast
- ETKF includes moist processes
- Running in conjunction with stochastic physics to propagate effect

#### Model error: parameterisations

#### Random parameters •Initial stoch. Phys. Scheme for the UM

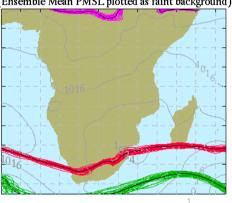
- ■QUMP (Murphy et al., 2004)
- (Arribas, 2004)

Parameter	Scheme	min/std/Max
Entrainment rate	CONVECTION	2/3/5
Cape timescale	CONVECTION	30 / 30 / 120
RH critical	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

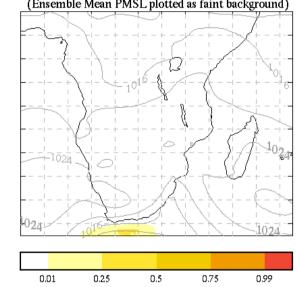
2012-11-28 <del>1</del>4

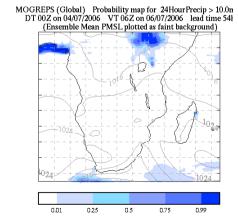
## MOGREPS products

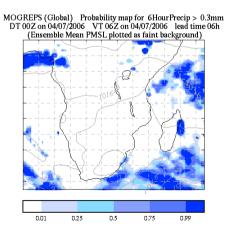
MOGREPS (Global) Spaghetti chart for 1000-500hPaThickness DT 00Z on 04/07/2006 VT 18Z on 04/07/2006 lead time 18h 510/528/546/564/582 dam (Black lines represent Control member) (Ensemble Mean PMSL plotted as faint background)



MOGREPS (Global) Probability map for 10mWindSpeed > 34.0knots DT 00Z on 04/07/2006 VT 06Z on 06/07/2006 lead time 54h (Ensemble Mean PMSL plotted as faint background)







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## Using probabilities

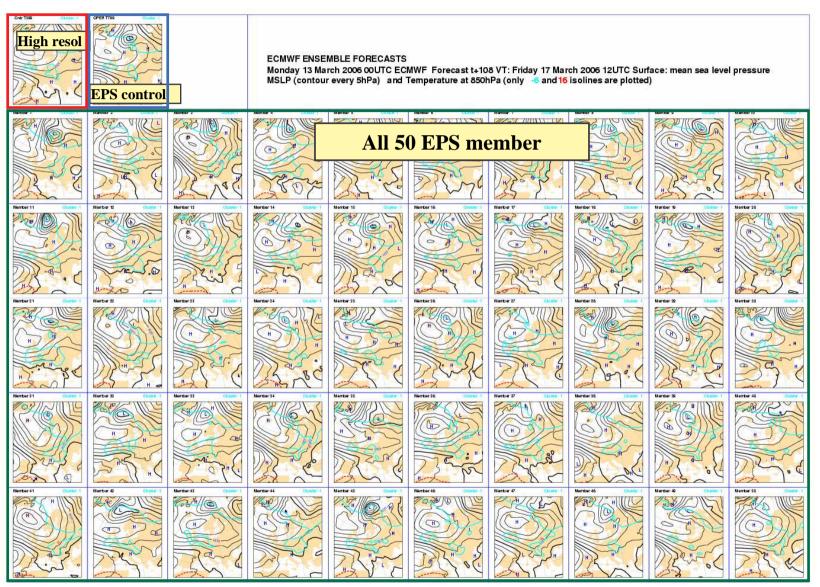
### Using probabilities

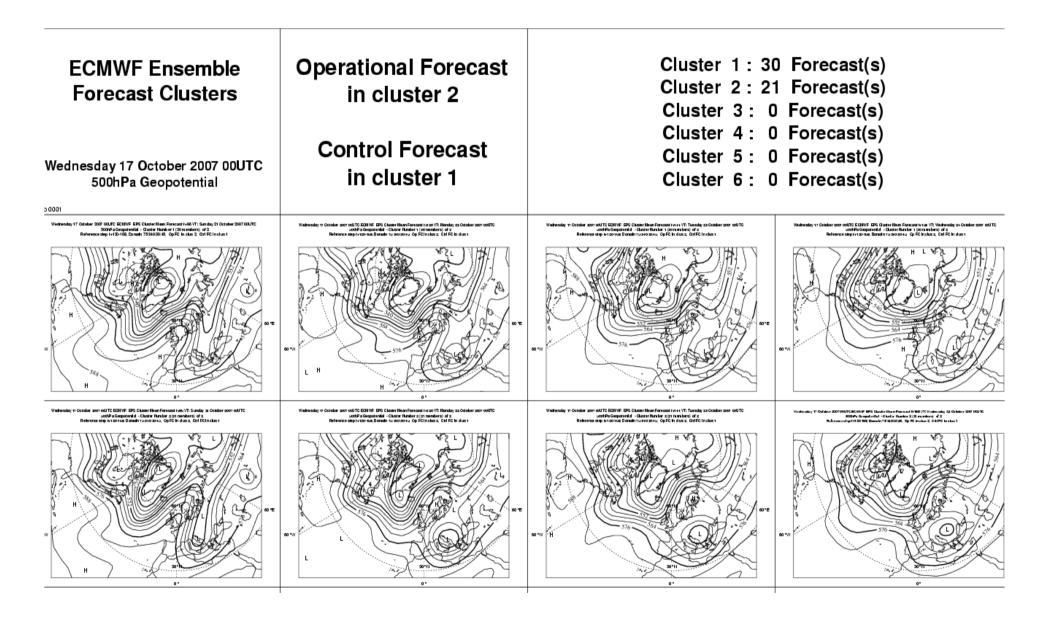
- Recipients of forecasts & warnings are sensitive to different levels of risk: reflecting cost of mitigation vs expected loss
- An intelligent response to forecasts & warnings depends on risk analysis, requiring knowledge of impact probability
- Use of ensembles to estimate probability at longer lead times is well established in meteorology

## Ensemble forecast products

# Stamp maps and clusters

#### Products: Stamp maps

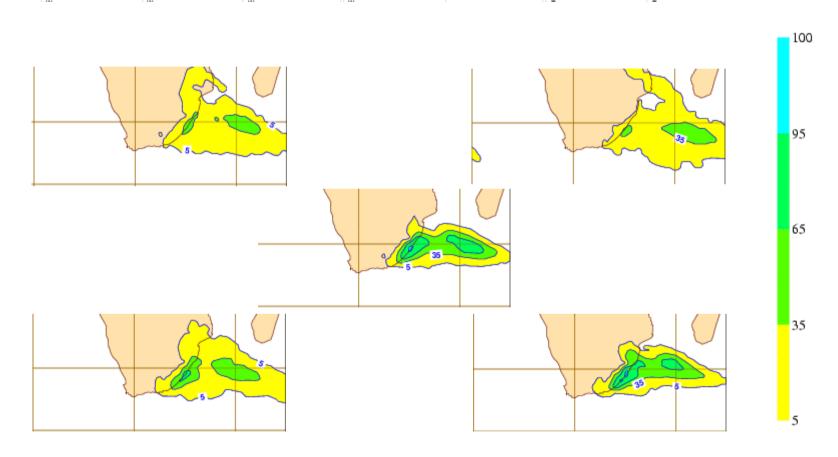




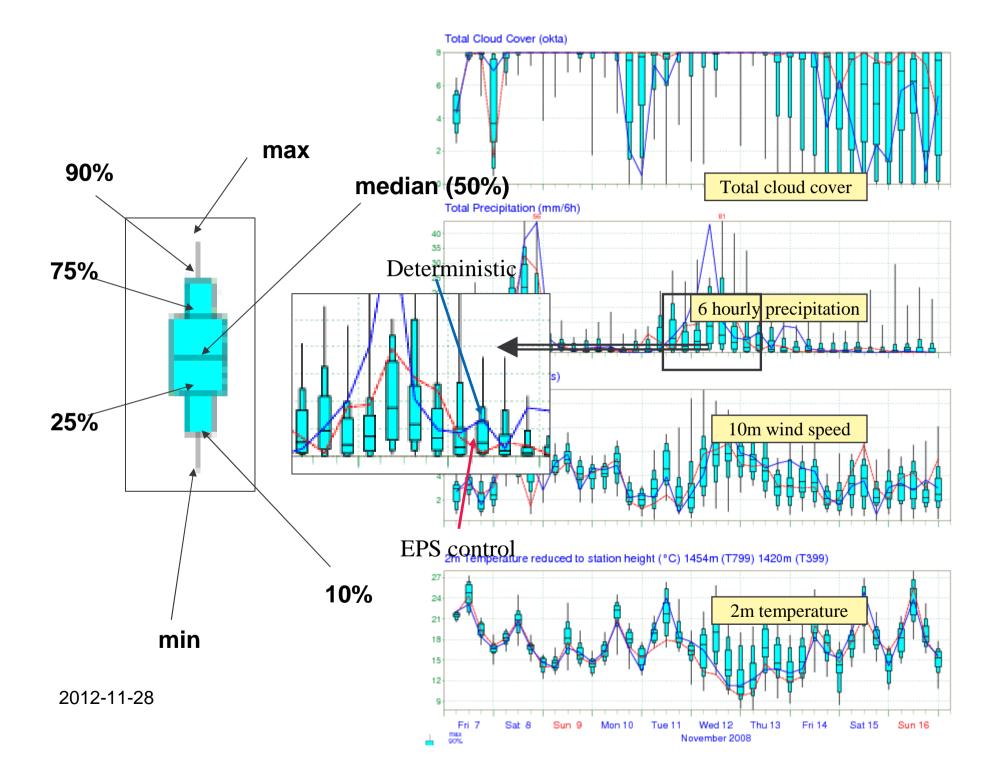
## **Probability maps**

#### EC Total ppn prob > 20mm 12z Tue - 12z Wed

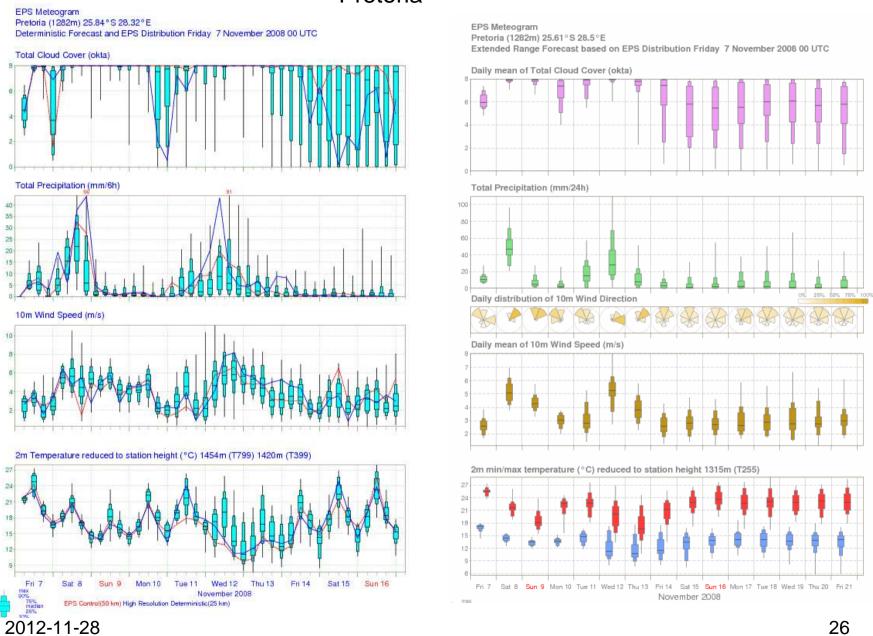
Monday 2 October 2006 00 UTC ©ECMWF Forecast probability t+036-060 VT: Tuesday 3 October 2006 12 UTC - Wednesday 4 October 2006 12 UTC Surface: Total precipitation probability > 20.0 mm



## **EPSgrammes**



#### Pretoria



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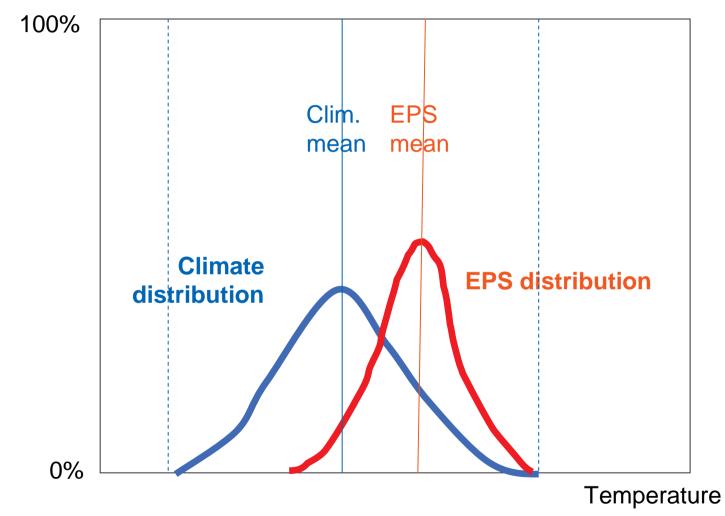
# The Extreme Forecast Index (EFI)

#### Extreme forecast index (EFI)

- EFI measures the distance between the EPS cumulative distribution and the model climate distribution
- Takes values from –1 (all members break climate minimum records) and +1 (all beyond model climate records)
- The main idea is to have an index that can be conveniently mapped – removing the effect from different climatologies – to use as an "alarm bell"

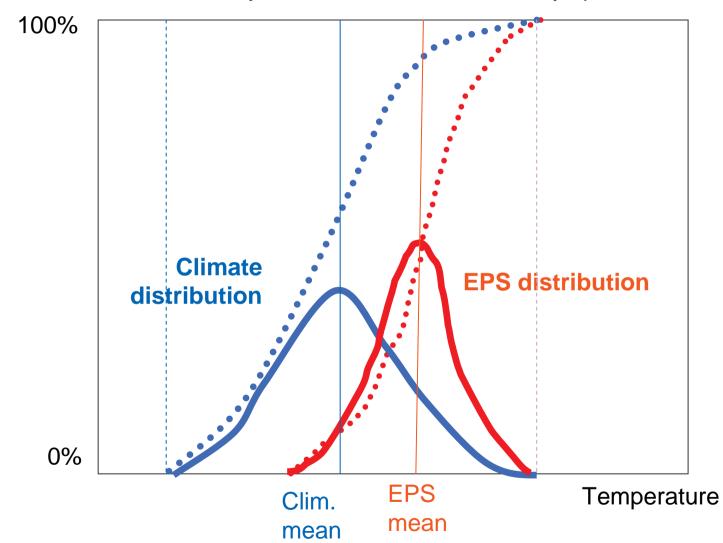
#### Advantages with probability density functions

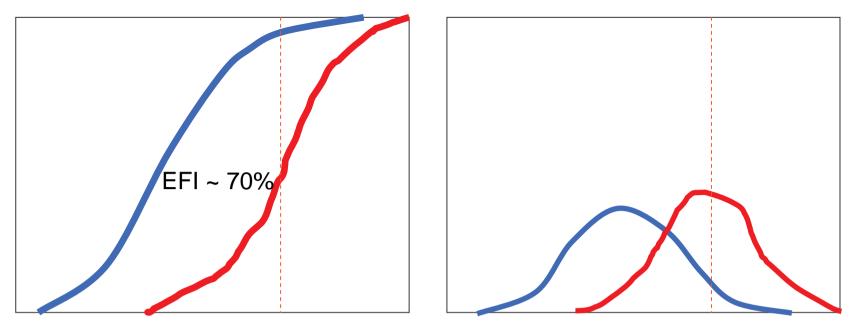
Means and asymmetric variances are easily spotted



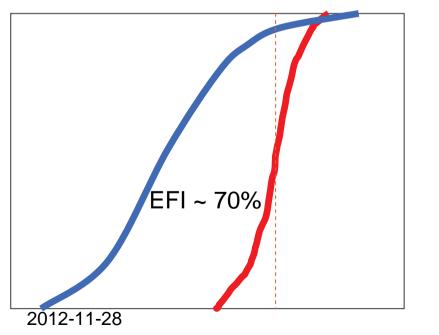
#### Advantages with probability density functions

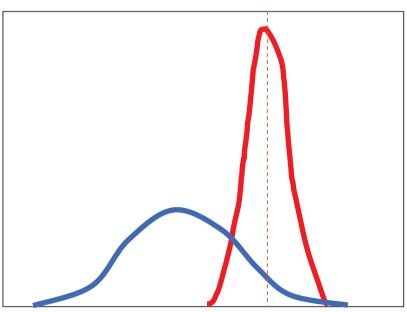
Means and asymmetric variances are easily spotted

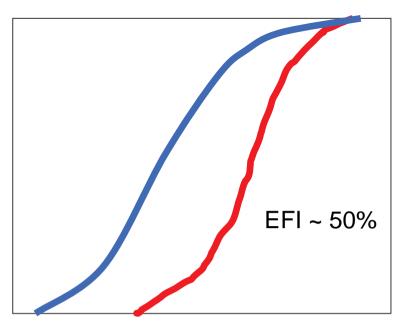


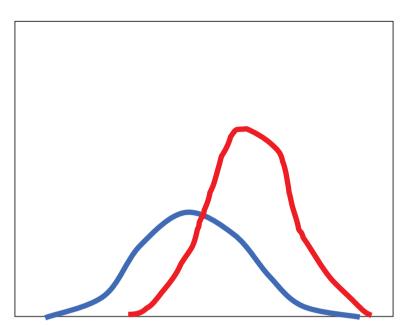


The EFI generally does not take the probability into account

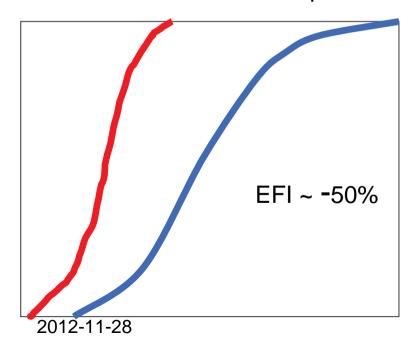


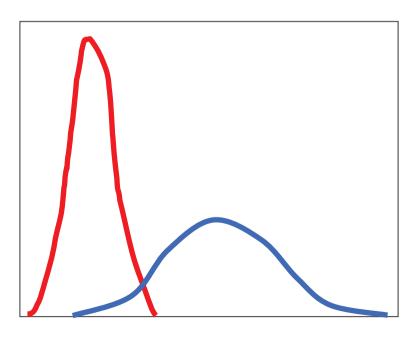




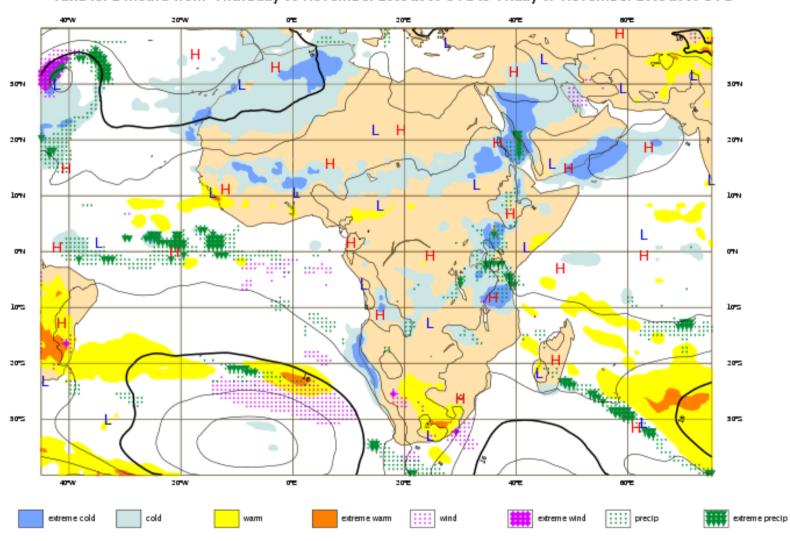


For temperature the EFI can take values < 0

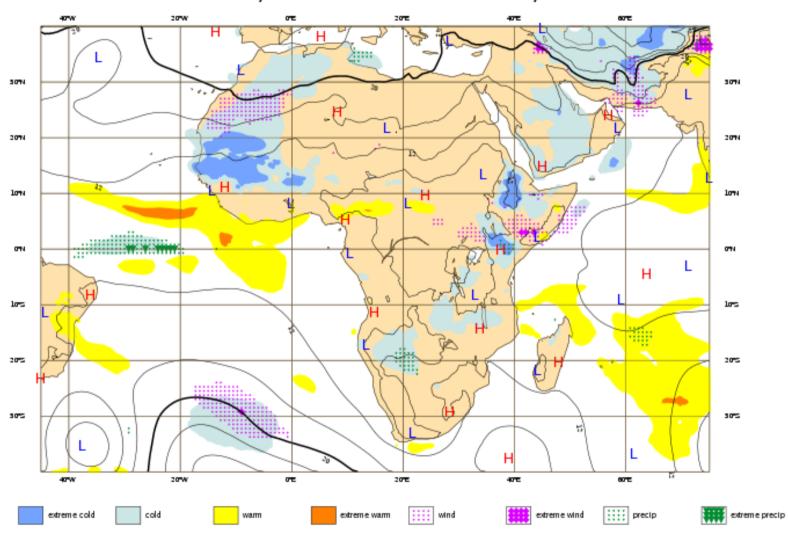




Weather anomalies predicted by EPS: Thursday 06 November 2008 at 00 UTC 1000 hPa Z ensemble mean (Thursday 06 November 2008 at 12 UTC) and EFI values for 24h TP, 10m wind gust and 2m temperature valid for 24hours from Thursday 06 November 2008 at 00 UTC to Friday 07 November 2008 at 00 UTC

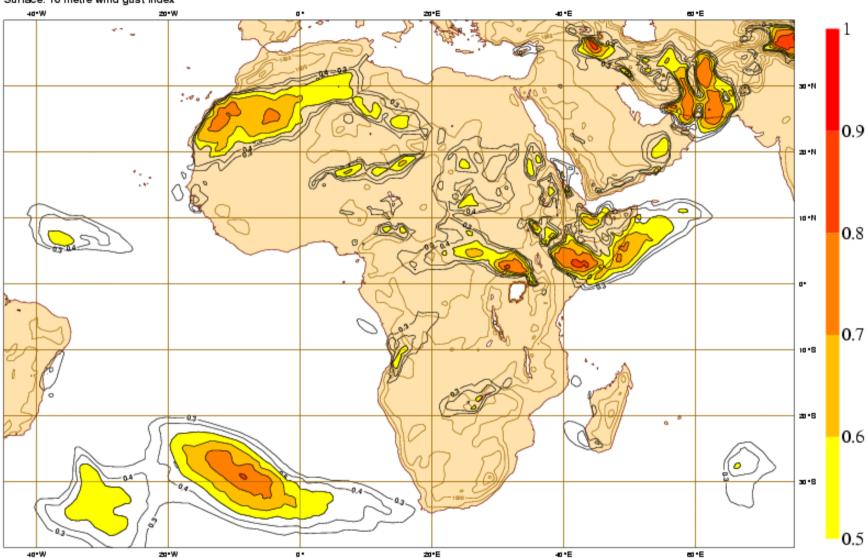


Weather anomalies predicted by EPS: Thursday 06 November 2008 at 00 UTC 1000 hPa Z ensemble mean (Monday 10 November 2008 at 12 UTC) and EFI values for 24h TP, 10m wind gust and 2m temperature valid for 24hours from Monday 10 November 2008 at 00 UTC to Tuesday 11 November 2008 at 00 UTC

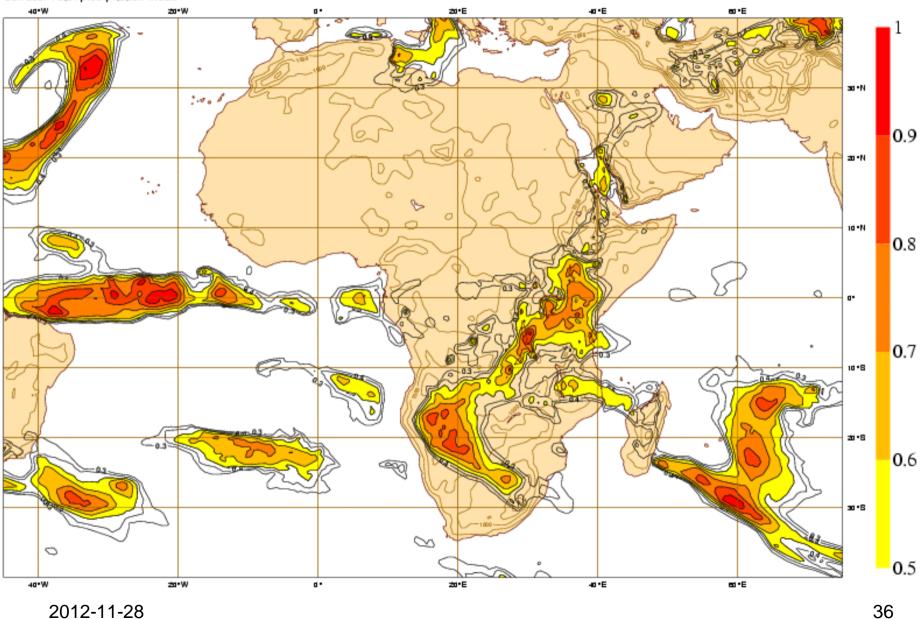


#### Product under development

Thursday 6 November 2008 00UTC ©ECMWF Extreme forecast index t+096-120 VT: Monday 10 November 2008 00UTC - Tuesday 11 November 2008 00UTC Surface: 10 metre wind gust index

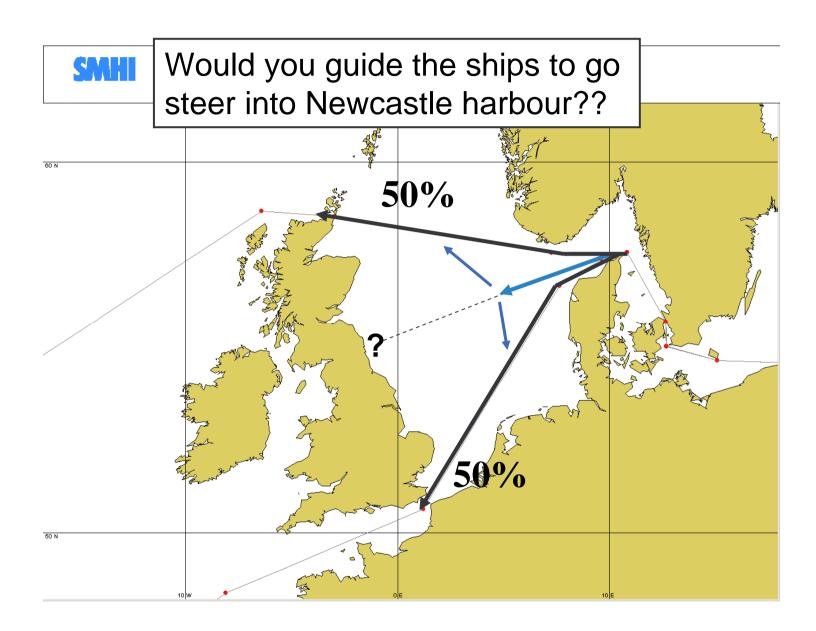


Product under development
Thursday 6 November 2008 00 UTC ©E CMWF Extreme forecast index t+000-120 VT: Thursday 6 November 2008 00 UTC - Tuesday 11 November 2008 00 UTC Surface: Total precipitation index



## Working with the EPS

- -Ensemble mean acts as a dynamic filter and removes normally unpredictable features
- -The removed features are put back in a consistent way as probabilities



#### **Questions & Answers**