

EUROSIP: seasonal forecasting with ocean-atmosphere models

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Outline

- **Seasonal forecasting**
 - Success of coupled ocean-atmosphere model forecasts
 - Limitations of model forecasts
- **EUROSIP multi-model system**
 - Benefits of multi-model approach
 - European organization
- **Implications for the models**
 - Initialization
 - Number of integrations
 - Need for and test of model quality

Sources of seasonal predictability

○ KNOWN TO BE IMPORTANT:

- El Nino variability - biggest single signal
- Other tropical ocean SST - important, but multifarious
- Climate change - especially important in mid-latitudes
- Local land surface conditions - e.g. soil moisture in spring

○ OTHER FACTORS:

- Volcanic eruptions - definitely important for large events
- Mid-latitude ocean temperatures - always controversial
- Remote soil moisture/ snow cover - not well established
- Sea ice anomalies - local effects, but remote??
- Dynamic memory of atmosphere - most likely on 1-2 months
- Stratospheric influences - various possibilities

Step 1: Build a coupled model

- **IFS (atmosphere)**

- T_L159L62 Cy31r1, 1.125 deg grid for physics (operational in Sep 2006)
- Full set of singular vectors from EPS system to perturb atmosphere initial conditions (more sophisticated than needed ...)
- Ocean currents coupled to atmosphere boundary layer calculations

- **HOPE (ocean)**

- Global ocean model, 1x1 mid-latitude resolution, 0.3 near equator
- A lot of work in developing the OI ocean analyses, including analysis of salinity, multivariate bias corrections and use of altimetry.

- **Coupling**

- Fully coupled, no flux adjustments, except no physical model of sea-ice

Step 2: Make some forecasts

- **Initialize coupled system**

- Aim is to start system close to reality. Accurate SST is particularly important, plus equatorial ocean sub-surface.

- **Run an ensemble forecast**

- Explicitly generate an ensemble on the 1st of each month, with perturbations to represent the uncertainty in the initial conditions; run forecasts for 7 months

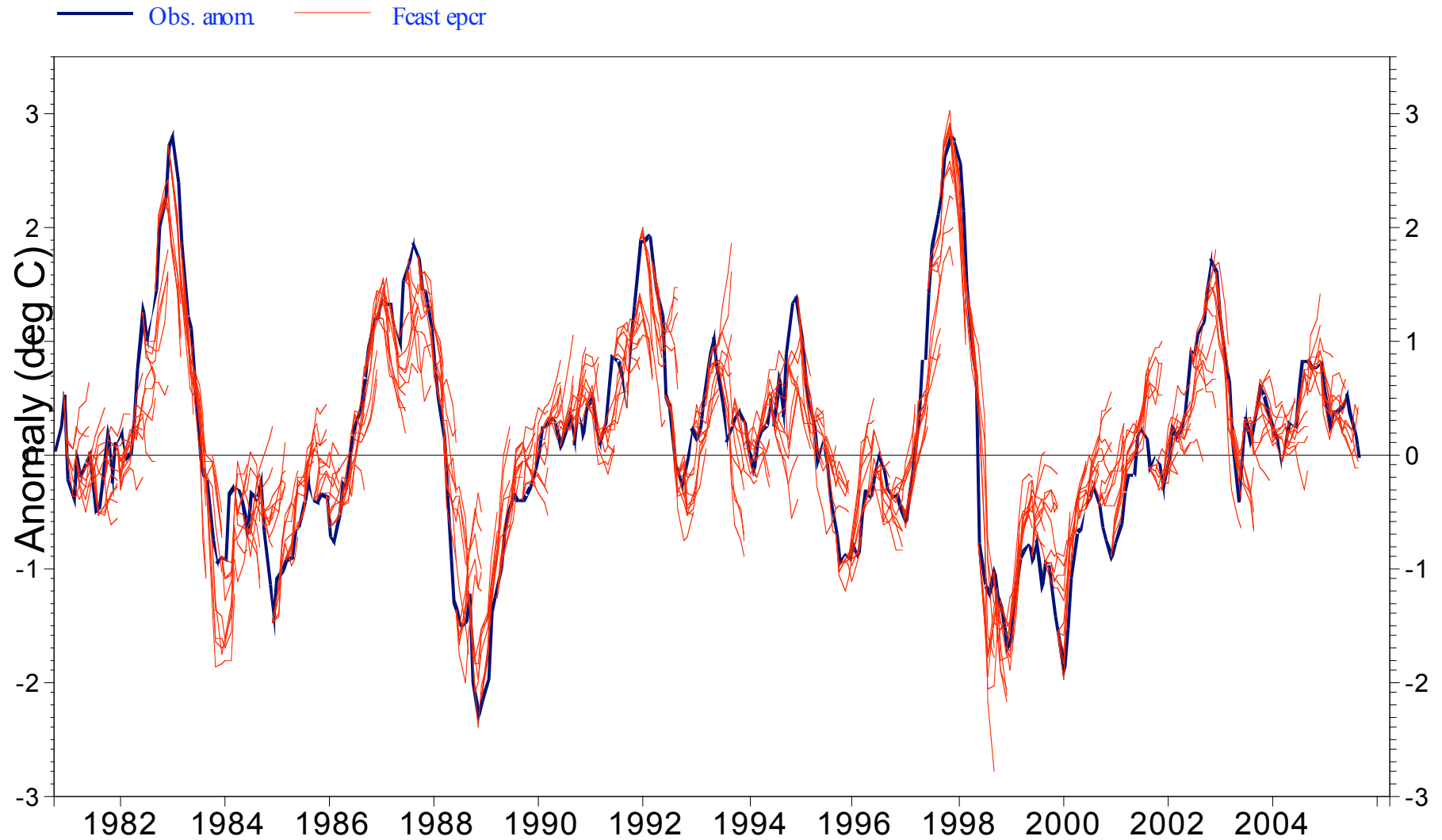
- **Remove systematic error**

- Forecasts have considerable systematic error; estimate this error from a set of **previous forecasts**, which define the model climatology.
- Model climatology is a function of date *and* forecast lead-time.
- Linear assumption is not correct, but seems to work remarkably well.

El Nino forecast performance

NINO3.4 SST forecast anomalies

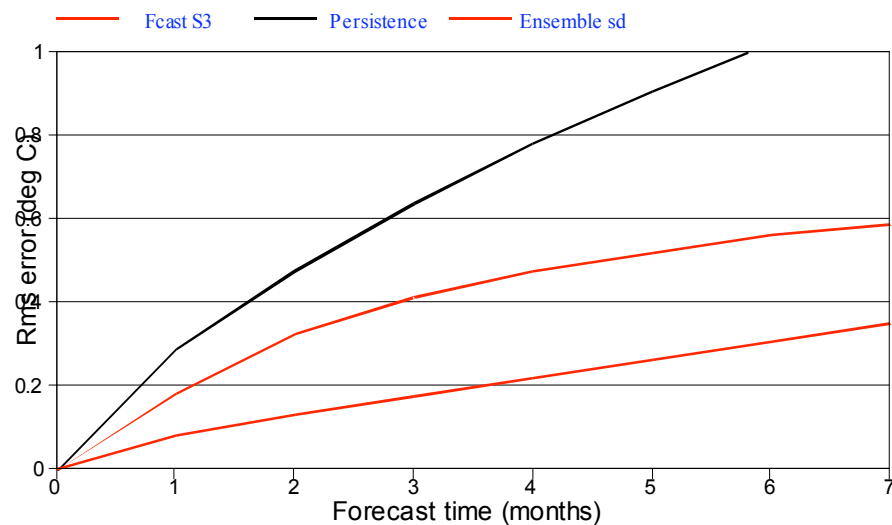
ECMWF forecasts at month 6
Ensemble size is 5 SST obs:



ECMWF System 3

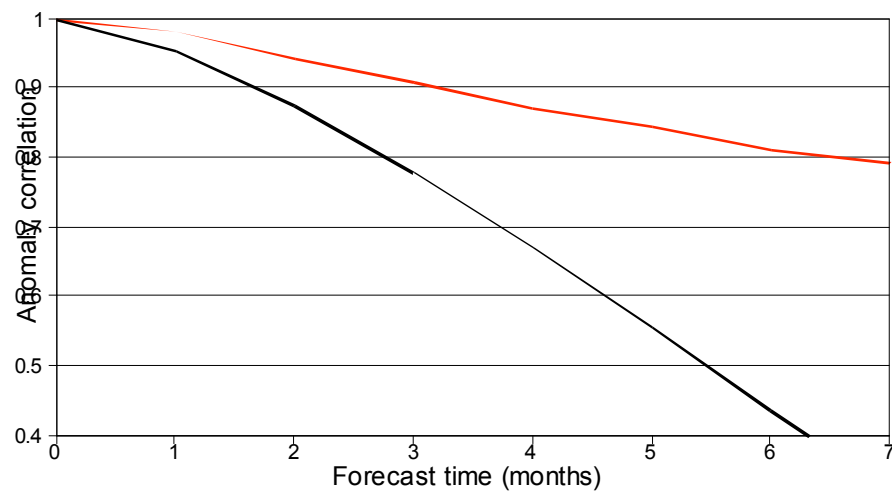
NINO3.4 SST rms errors

300 start dates from 19810101 to 20051201
Ensemble size is 11



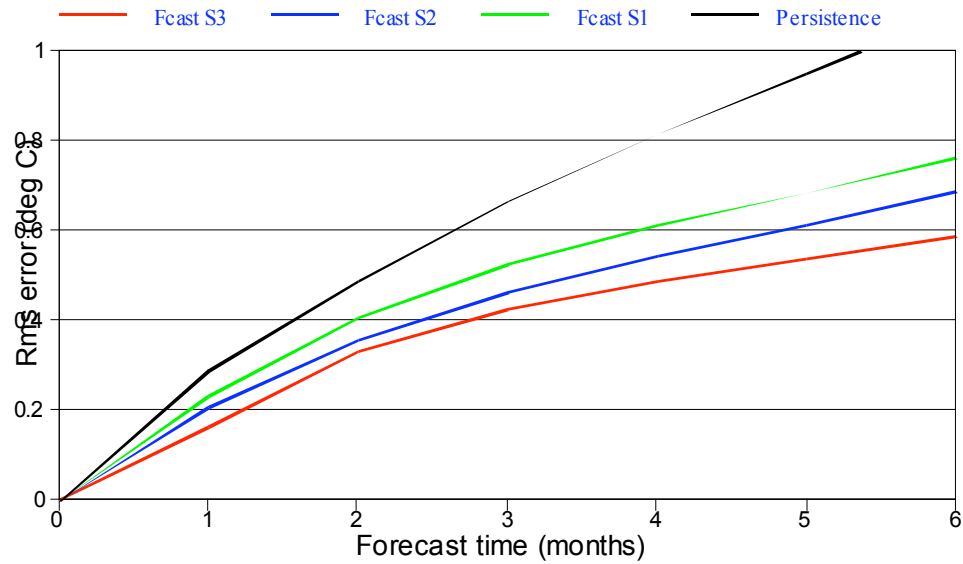
NINO3.4 SST anomaly correlation

wrt NCEP adjusted OI2 1971-2000 climatology



NINO3.4 SST rms errors

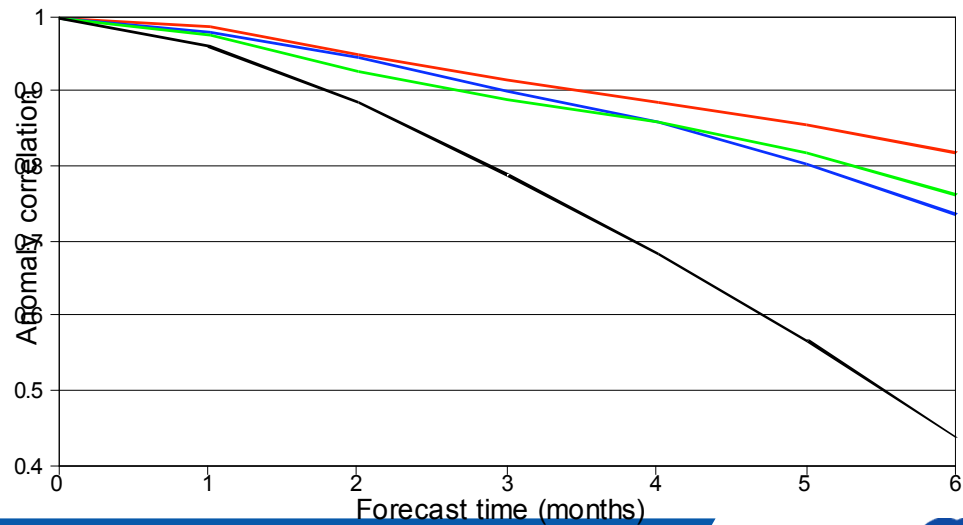
192 start dates from 19870101 to 20021201
Ensemble sizes are 11 (0001), 5 (0001) and 5 (0001)



Progress S1 -> S2 -> S3

NINO3.4 SST anomaly correlation

wrt NCEP adjusted OI2 1971-2000 climatology



S1: 1997

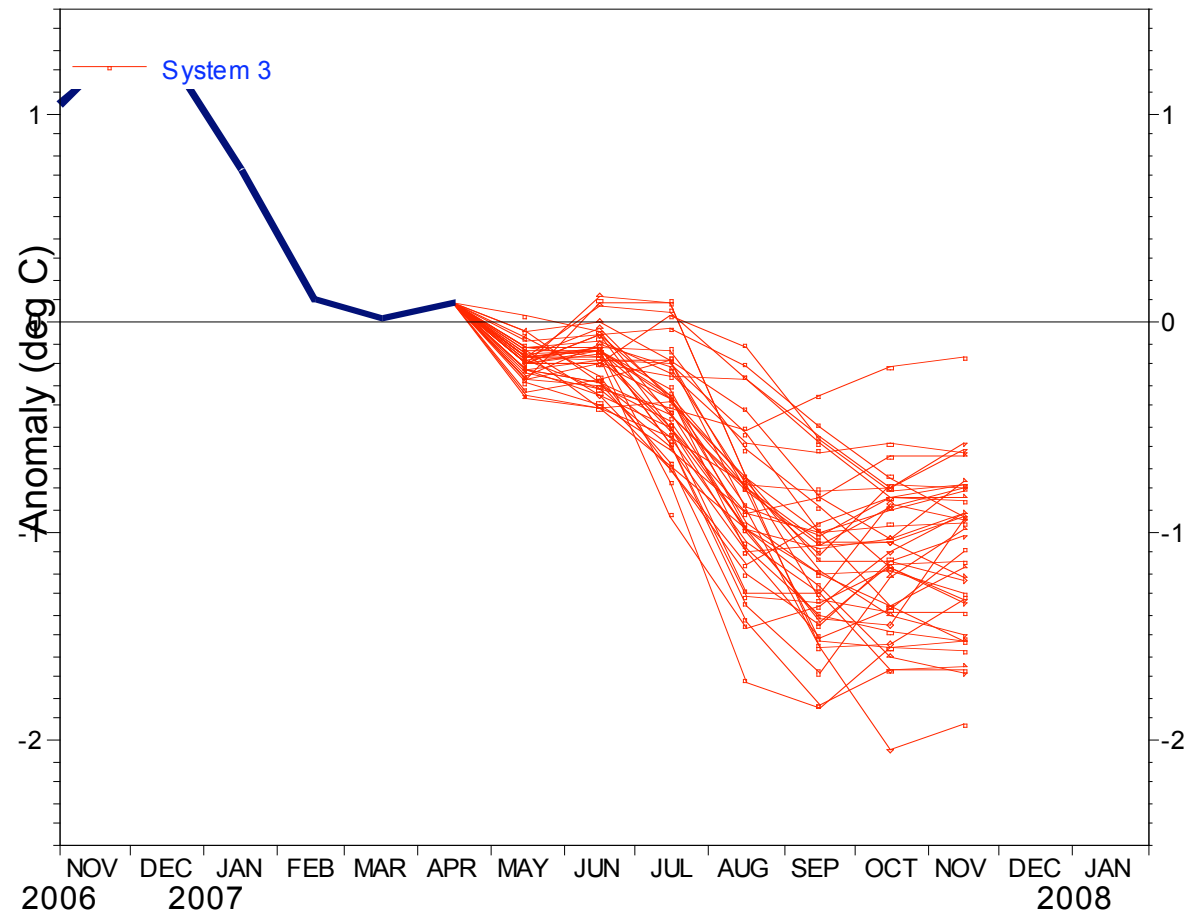
S2: 2002

S3: 2007

Latest forecast

NINO3.4 SST anomaly plume ECMWF forecast from 1 May 2007

Monthly mean anomalies relative to NCEP adjusted OI2 1971-2000 climatology



Forecast issue date: 15 May 2007



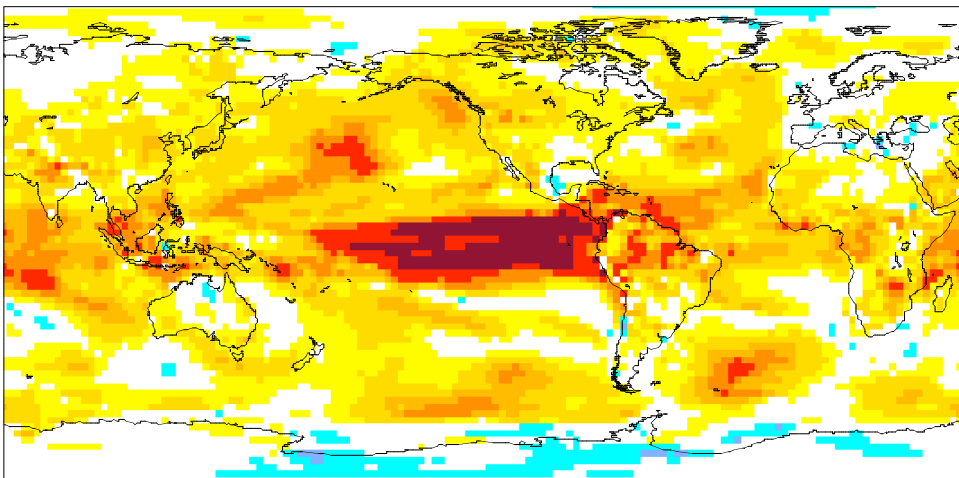
Atmospheric skill (provisional)

Deterministic skill: DJF ACC

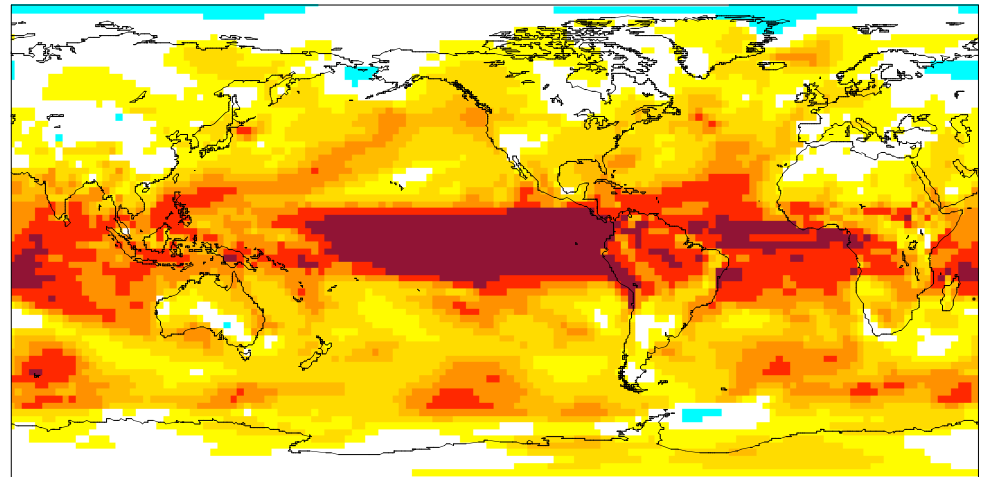
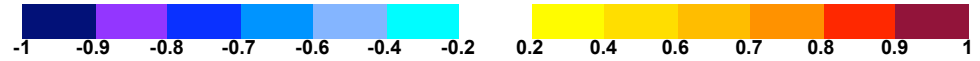
Temperature: actual forecasts

Temperature: perfect model

Anomaly Correlation Coefficient for CodOecmfE0001S003M001 with 11 ensemble members
Near-surface temperature
Hindcast period 1981-2003 with start in November and averaging period 2 to 4



Perfect-model Anomaly Correlation Coefficient for CodOecmfE0001S003M001 with 11 ensemble members
Near-surface temperature
Hindcast period 1981-2003 with start in November and averaging period 2 to 4

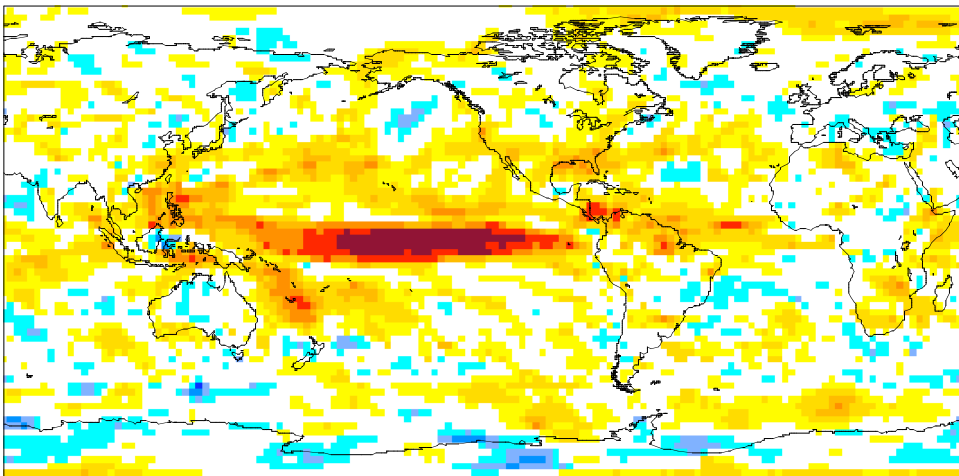


Deterministic skill: DJF ACC

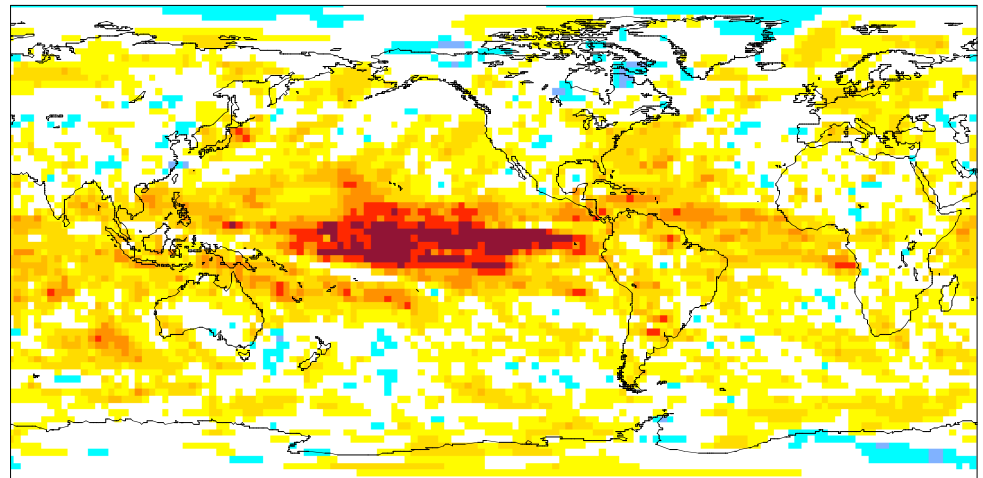
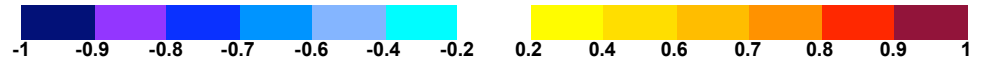
Precip: actual forecasts

Precip: perfect model

Anomaly Correlation Coefficient for CodOecmfE0001S003M001 with 11 ensemble members
Precipitation
Hindcast period 1981-2003 with start in November and averaging period 2 to 4

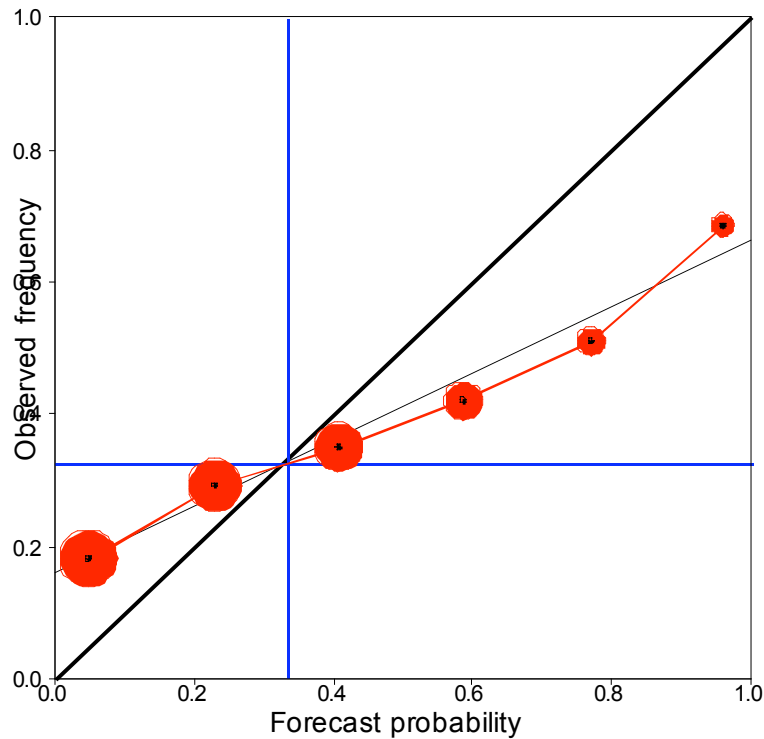


Perfect-model Anomaly Correlation Coefficient for CodOecmfE0001S003M001 with 11 ensemble members
Precipitation
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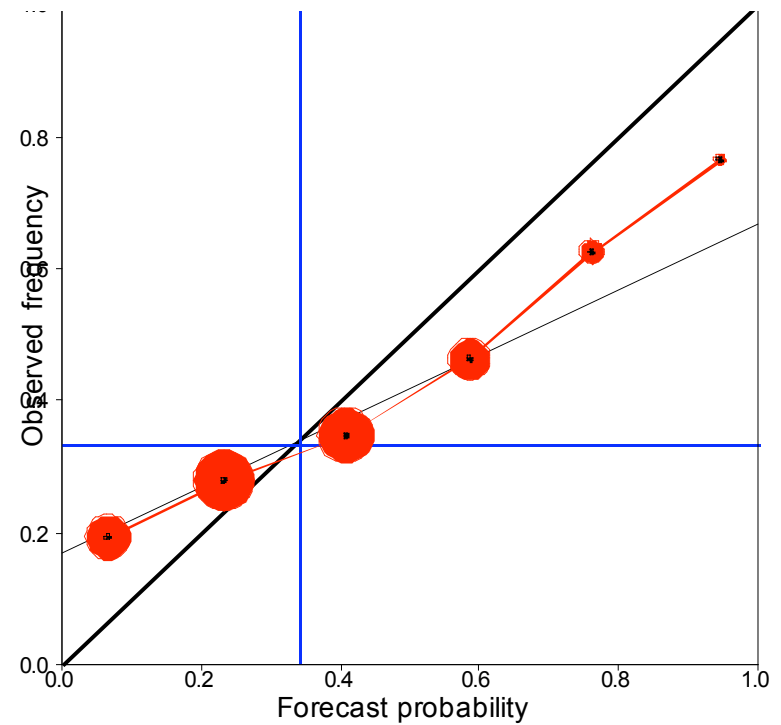


Probabilistic skill: Reliability diagrams

Tropical precip < lower tercile, JJA



NH extratrop temp > upper tercile, DJF



Why multi-model seasonal forecasts?

- **Seasonal forecasts dominated by model error**

- Ocean initialization is important too, especially for El Nino
- However, with state-of-the-art systems model error dominates even for SST forecasts
- For atmospheric response, model error even more of an issue

- **Different models have different errors**

- We do not expect errors to be orthogonal or independent
- But at least *some* part of model error will differ and can be averaged out
- Averaging over models improves both skill and reliability

- **Research shows multi-model approach works well**

- Eg DEMETER project

Theory of multi-model forecasting

- **NOT sampling equally likely outcomes**

- We expect all models to be wrong, none of them to be possible solutions
- ARE sampling model-induced errors in predicted outcomes of individual events
- If these errors are independent, then a large multi-model ensemble might have a *small* error in the ensemble mean, much smaller than the inter-model spread

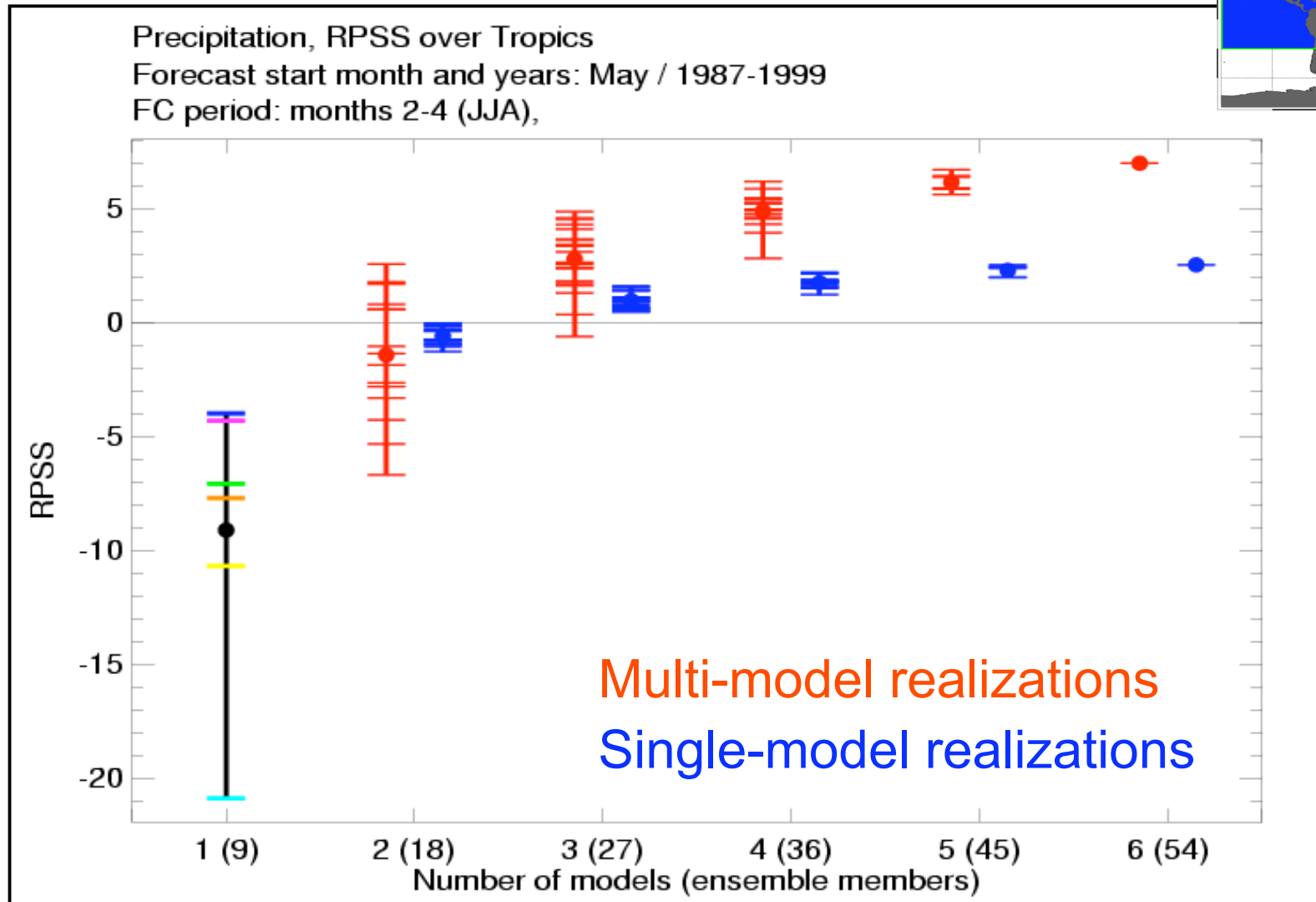
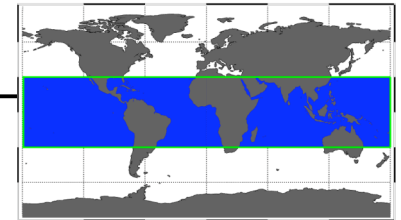
-> Optimistic view of multi-model forecasts

- **Model errors can easily be highly dependent**

- Under-represented processes (eg complexity of aerosols)
- Anything related to not having a 10km resolution ...

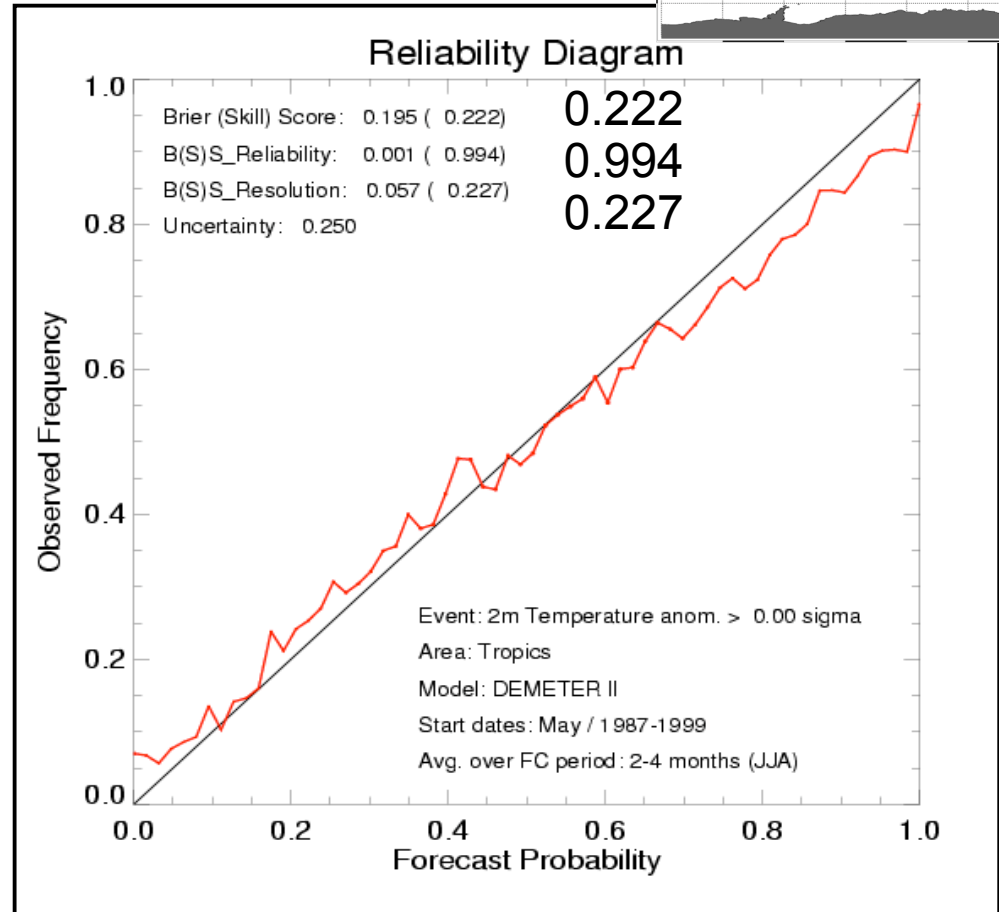
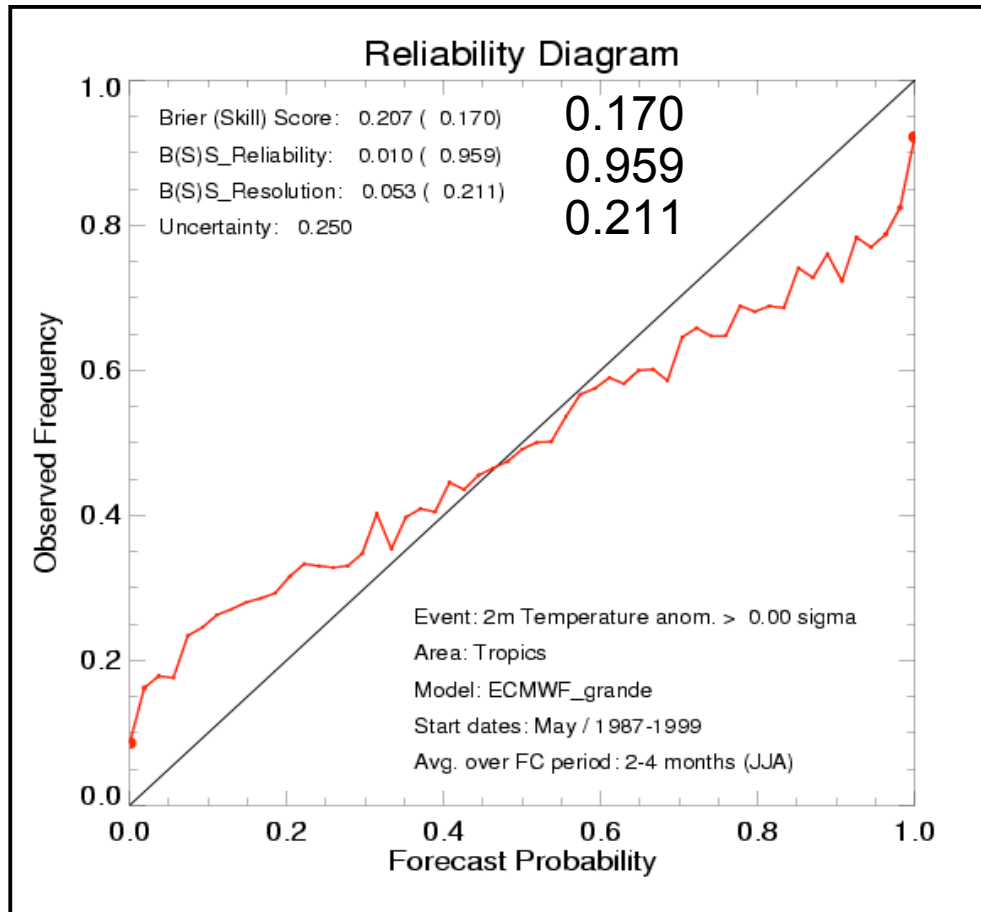
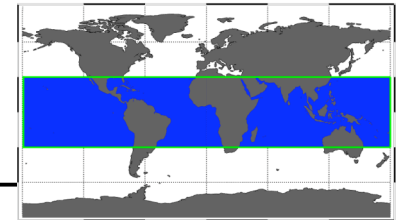
-> Pessimistic view of multi-model forecasts

DEMETER: Impact of number of models



Reliability diagram: 2m-Temp.>0


Benefit of 6 models



single-model (54 members)

multi-model

What is EUROSIP?

- **EUROSIP multi-model seasonal forecasts**
 - SIP = seasonal-interannual prediction
 - 6 month forecasts
- **Three models at present**
 - ECMWF: system 3 (IFS/HOPE)
 - Met Office: system 3 (GLOSEA – based on HADCM3)
 - Meteo-France: system 2 (Arpege/OPA)
- **Real-time forecasts since April 2005**
 - Initial agreements with Met Office and Meteo-France were made separately
 - Replaced by recently agreed comprehensive new data policy
 - German contribution now under development 

What does EUROSIP give?

- **A coordinated operational system**
 - Production schedule
 - System design (ensemble sizes, back integrations ..)
- **Unified data structures**
 - Based on ECMWF grib, MARS, fdb, and dissemination structures for seasonal forecast data – a relatively mature infrastructure
- **Performance checking of individual models**
 - Analysis of historic performance
 - Checking of real-time forecasts
- **Multi-model products**

EUROSIP multi-model seasonal forecast

Prob(2m temperature > upper tercile)

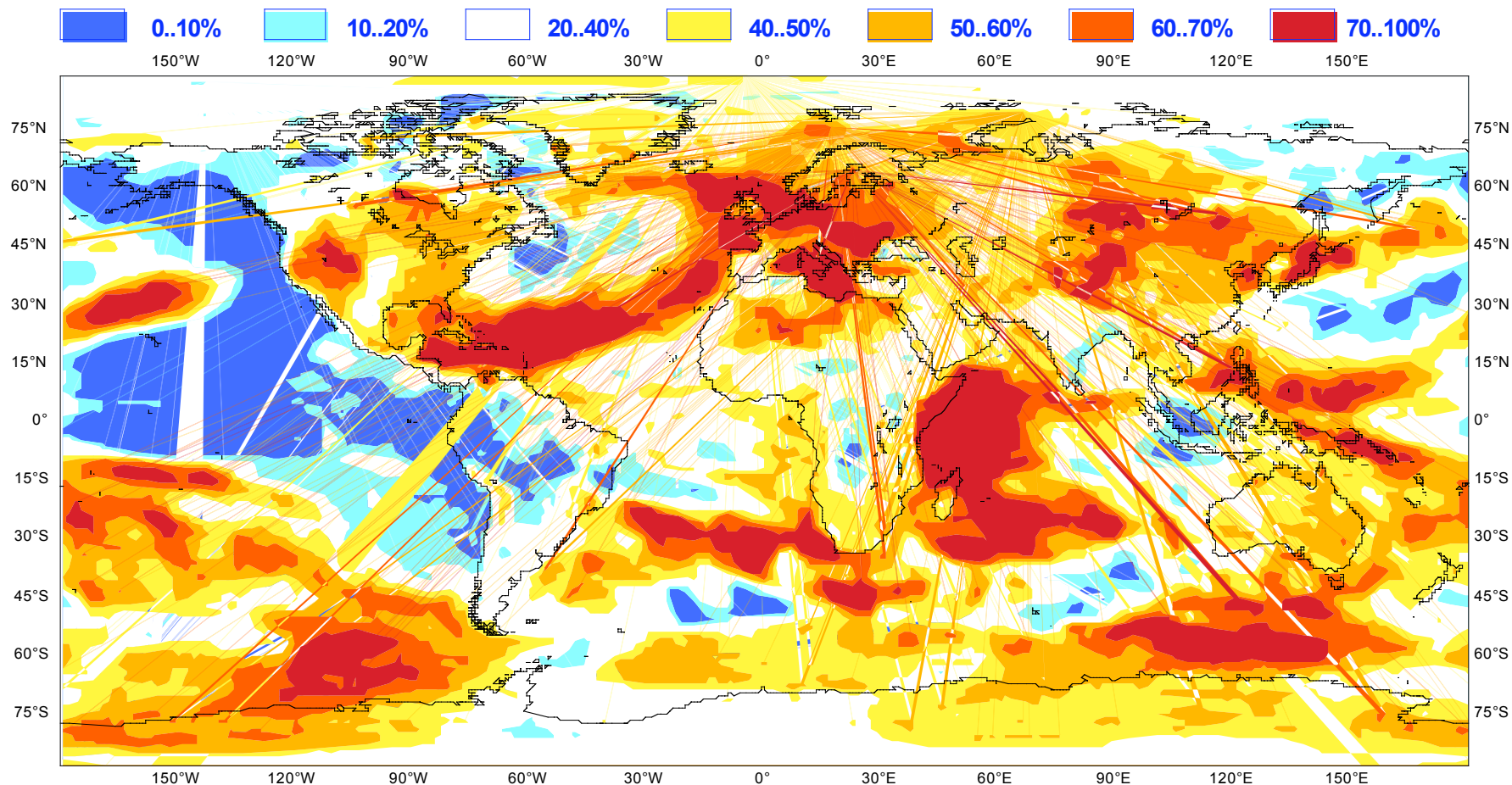
Forecast start reference is 01/05/07

Unweighted mean

ECMWF/Met Office/Météo-France

JJA 2007

No significance test applied



Forecast issue date: 15/05/2007

ECMWF

COSMOS General Assembly, Mainz 23-25 May 2007

ECMWF

Implications: initialization

- **ECMWF uses sophisticated ocean analysis**
 - Multivariate scheme, OI based
 - Special adjustments for model biases
 - Altimetry
- **But simple wind-forced ocean is also good**
 - Use decent quality wind field (eg ERA-40)
 - Strong relaxation to SST (eg NCEP OIv2)
 - SST forecasts only slightly worse than with high quality analysis
 - Impact on main atmosphere fields is very limited
 - => **Benefit of a different atmosphere/ocean model is easily obtained**

Implications: number of integrations

- **Real-time forecasts:**

- 41 member ensemble, once per month
- 6 or 7 months long => 24 model-years
- To be completed in 1 day wall-clock (but is very parallel)

- **Re-forecasts for calibration**

- 5 or more member ensemble (recommended 11)
- Every month, from 1981-present
- Say 900 -> 2000 model years
- Wall-clock time: up to a year
- Repeated for each new system
- => Do not change systems very often

Implications: model quality

- **Seasonal forecasting is demanding of model quality**
 - Biggest source of forecast error is model error
 - Tropical convection is particularly tough – multi-scale problem
 - Any error that affects the mean state can be a problem
- **Seasonal forecasting is a good test of model quality**
 - Relatively large number of cases to verify (cf climate)
 - Sensitive to most of the physics of the atmosphere, and faster coupled ocean-atmosphere processes
- **Climate**
 - Seasonal forecasting is unavoidably linked with climate change
 - Ability of models to capture observed trends (even in short range forecasts) can disappoint