

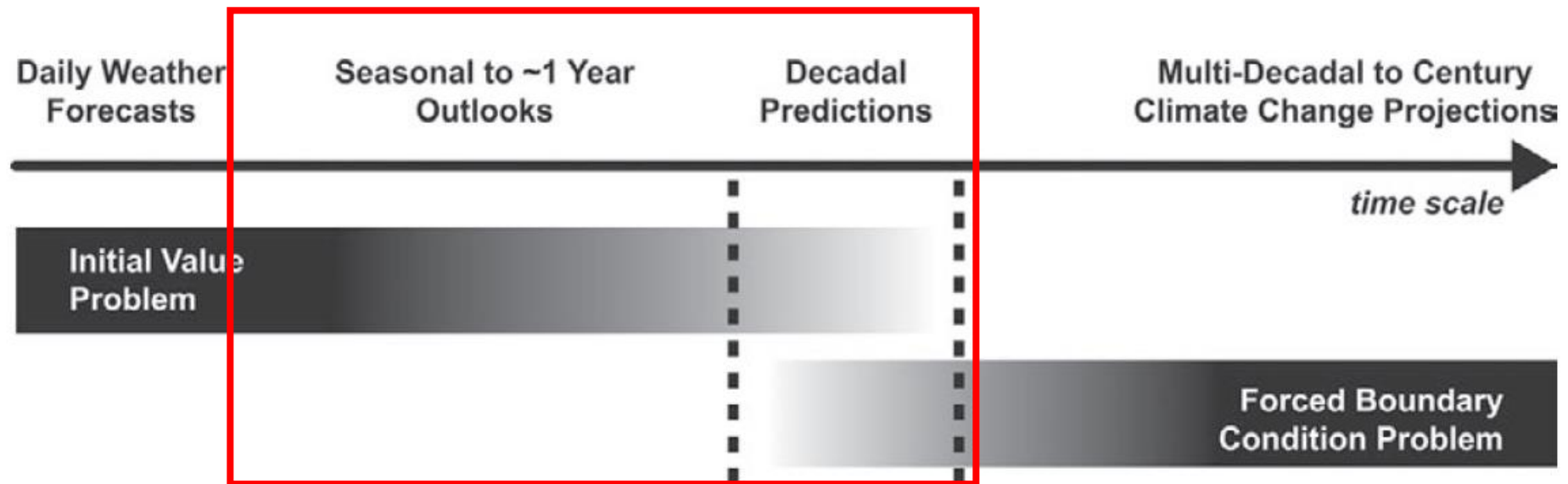
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# Climate Forecasting or the Continuous Adaptation to Climate Change

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Barcelona, Spain

# Climate time scales

Progression from initial-value problems with weather forecasting at one end and multi-decadal to century projections as a forced boundary condition problem at the other, with climate prediction (**sub-seasonal, seasonal and decadal**) in the middle. Prediction involves initialization and systematic comparison with a **simultaneous** reference.



Meehl et al. (2009)

# Sources of seasonal predictability

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- Important:

- o ENSO
  - biggest single signal
- o Other tropical ocean SST
  - difficult
- o Climate change
  - important in mid-latitudes
- o Local land surface conditions
  - soil moisture, snow
- o Atmospheric composition
  - difficult

- Other factors:

- o Volcanic eruptions
  - important for large events
- o Mid-latitude ocean temperatures
  - still somewhat controversial
- o Remote soil moisture/snow cover
  - not well established
- o Sea-ice anomalies
  - at least local effects
- o Stratospheric influences
  - various possibilities
- o Remote tropical atmospheric teleconnections

- Unknown or Unexpected

# Methods of seasonal forecasting

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- Empirical forecasting

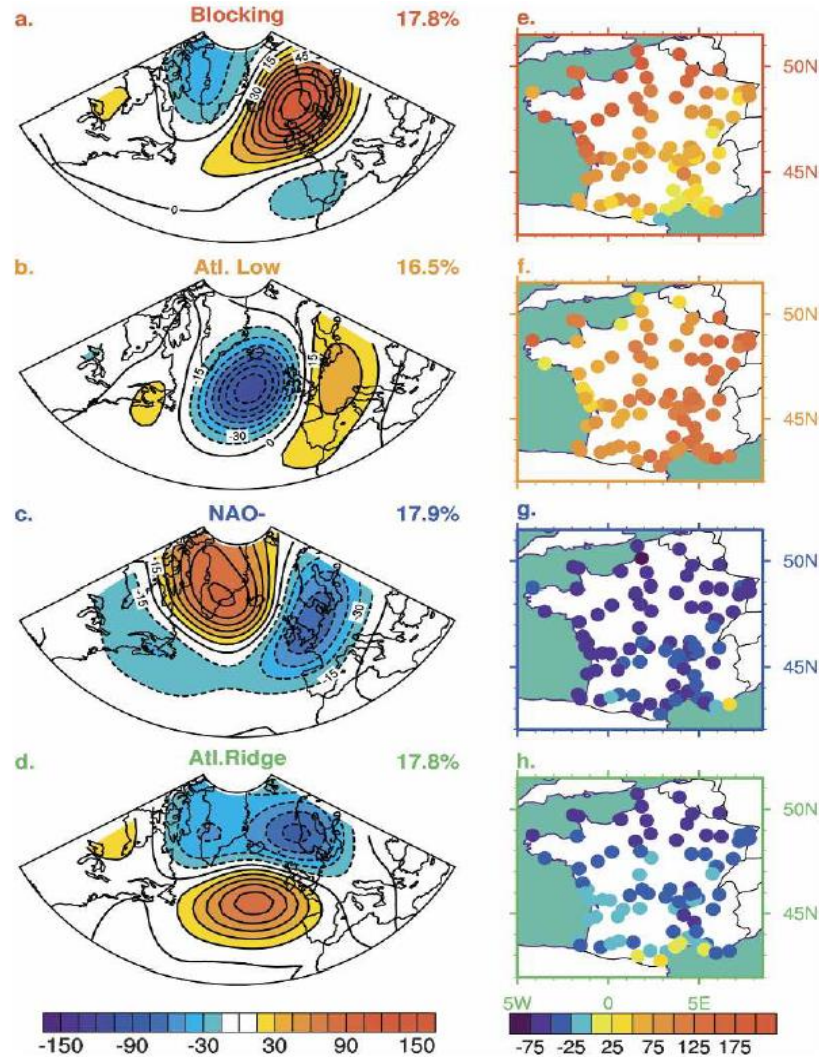
- o Use past observational record and statistical methods ✓
- o Works with reality instead of error-prone numerical models ✓
- o Limited number of past cases ✗
- o A non-stationary climate is problematic ✗
- o Can be used as a benchmark ✓

- Single-tier GCM forecasts

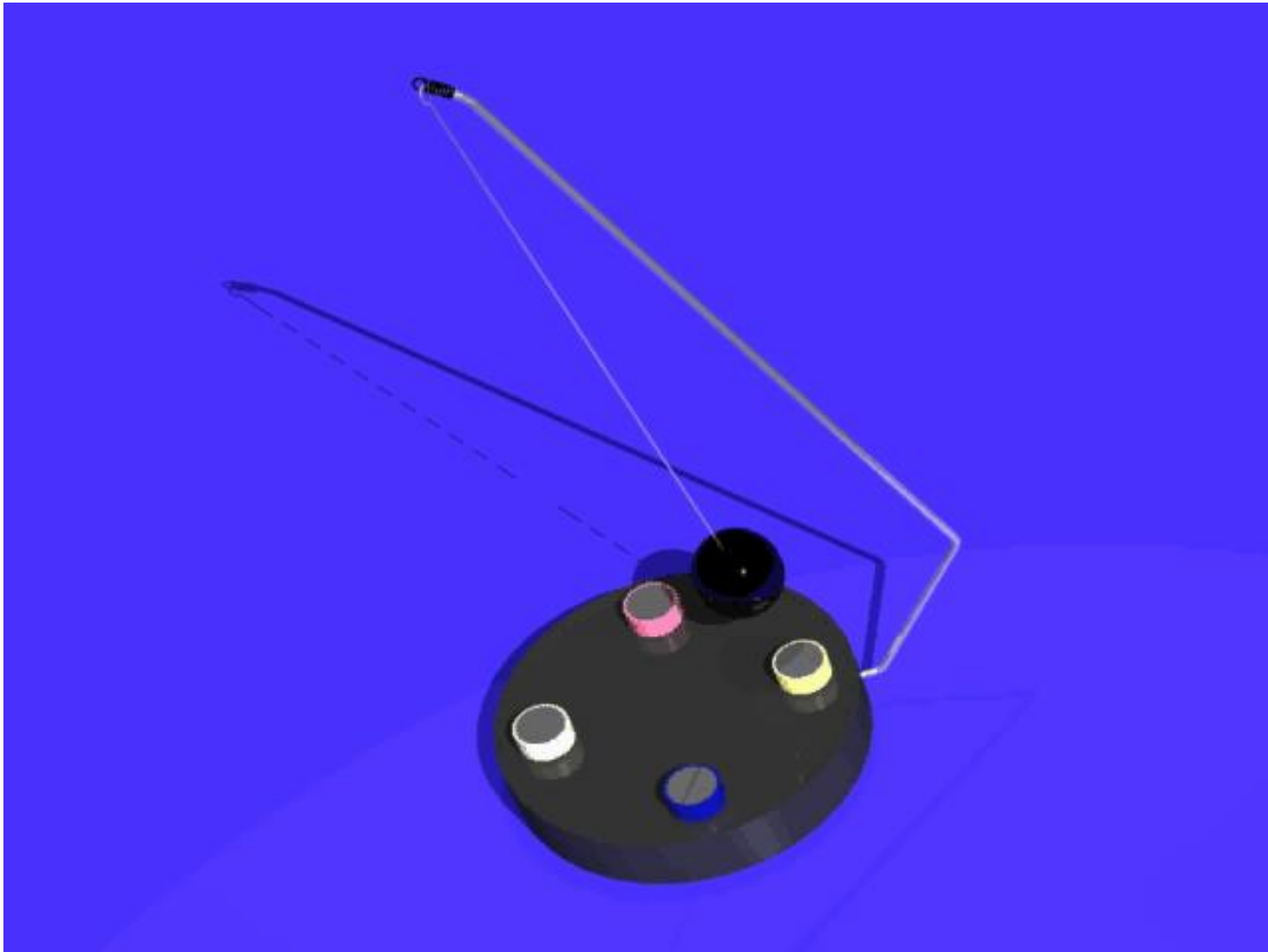
- o Include comprehensive range of sources of predictability ✓
- o Predict joint evolution of ocean and atmosphere flow ✓
- o Includes a large range of physical processes ✓
- o Includes uncertainty sources, important for prob. Forecasts ✓
- o Systematic model error is an issue! ✗

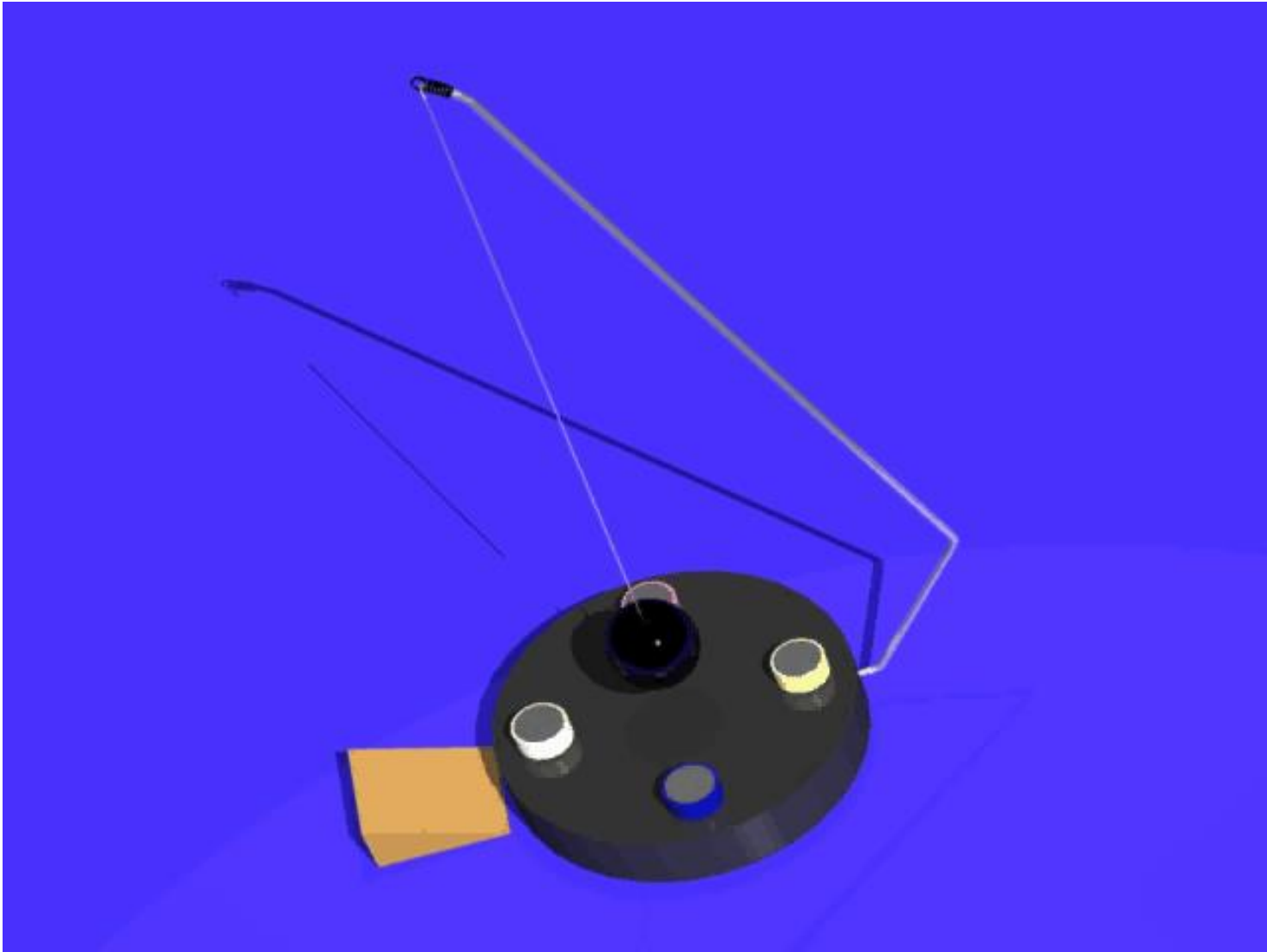
# A simile: Weather types

Z500 summer weather types and frequency change (%) of warm days

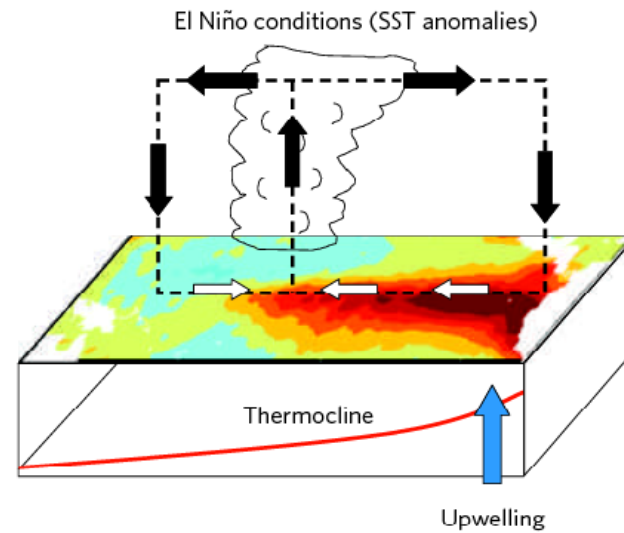
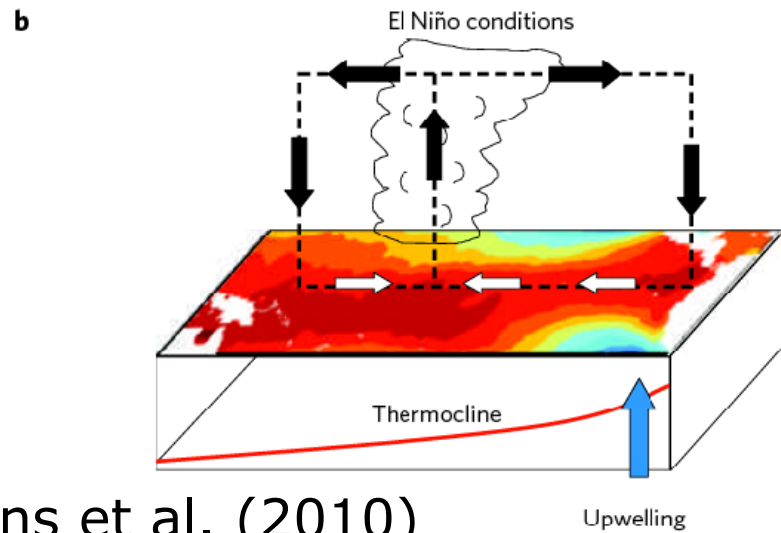
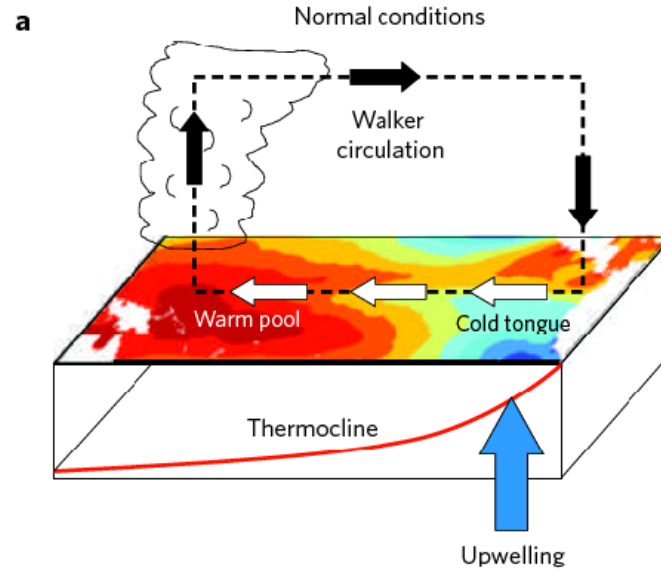


Cassou et al. (2005)





# The wedge: ENSO in the tropical Pacific

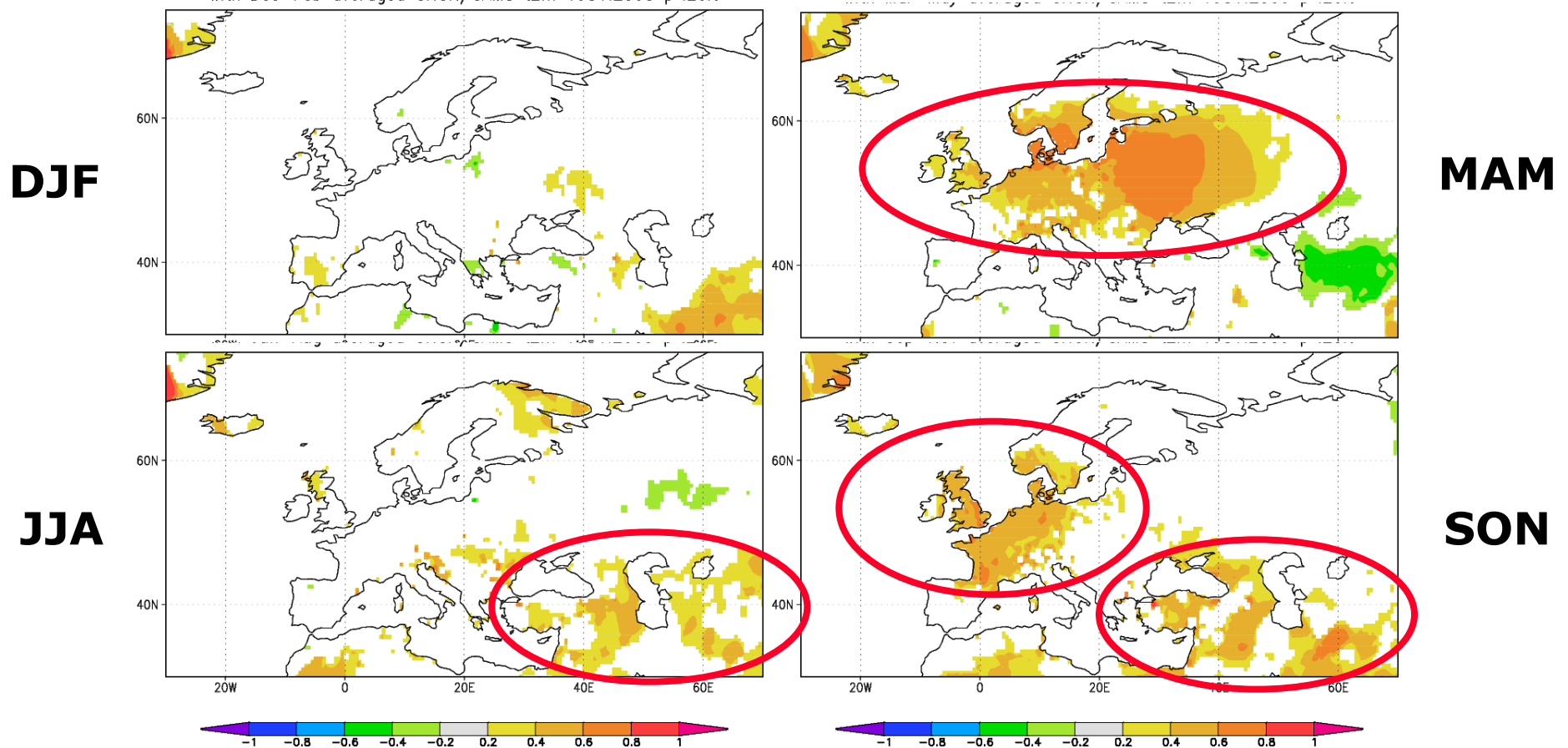


Collins et al. (2010)



# Temperature skill: persistence

Correlation of GHCN temperature of one-month lead anomaly persistence over 1981-2005. Only values statistically significant with 80% confidence are plotted.



# To produce dynamical forecasts

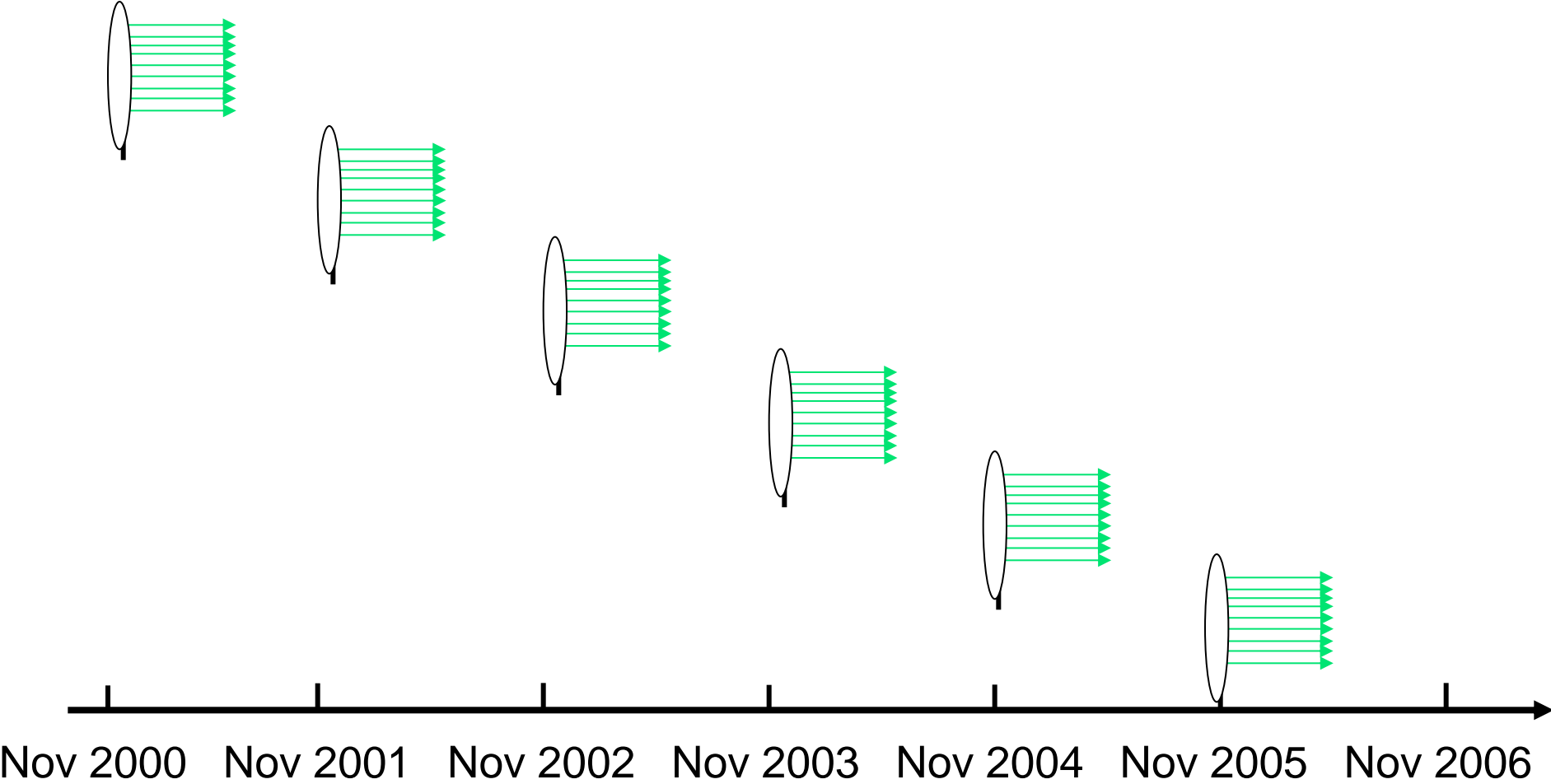
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- Build a coupled model
- Prepare initial conditions
- Initialize coupled system
  - The aim is to start the system close to reality. Accurate SST is particularly important, plus ocean sub-surface. Usually, worry about “imbalances” a posteriori.
- Run an ensemble forecast
  - Explicitly generate an ensemble on the e.g. 1st of each month, with perturbations to represent the uncertainty *in the initial conditions*; run forecasts for several months.
- Produce probability forecasts from the ensemble
- Apply calibration and combination if significant improvement is found, for which **hindcasts** are required

# Ensemble initialized climate predictions

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Assume an ensemble forecast system with an initialized ESM



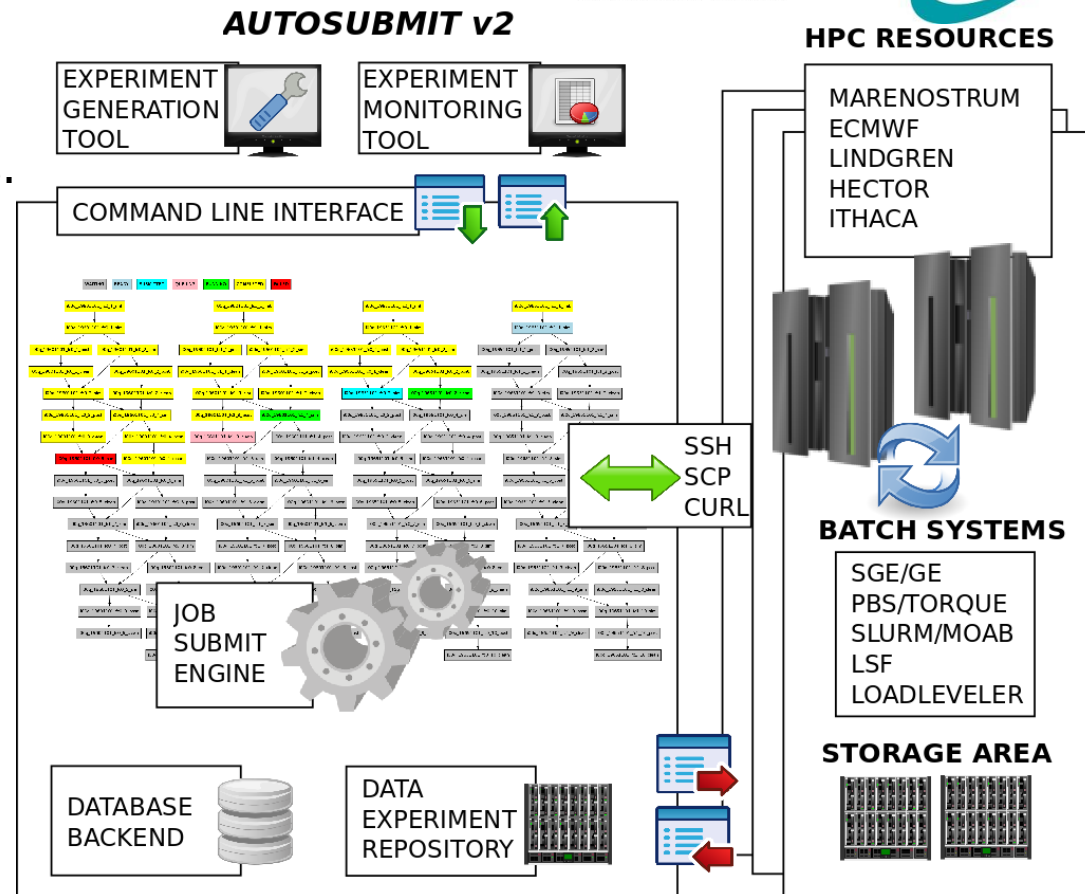
# Autosubmit

Autosubmit acts as a wrapper to run a climate experiment on a HPC. The experiment is a sequence of jobs that it submits, manages and monitors. When a job is complete, the next one can be executed.



- Divided in 3 phases: ExpID assign, experiment creation, run.
- Separation experiment/autosubmit codes.
- Config files for autosubmit and experiment.
- Database to store experiment information.
- **Common templates for all platforms.**
- Recovery after crashes.
- **Dealing with a list of schedulers and communication protocols.**

Each job has a colour in the monitoring tool: yellow=completed, green=running, blue=pending, etc.

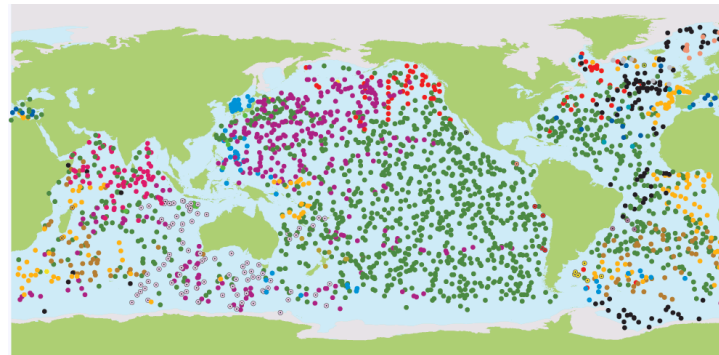
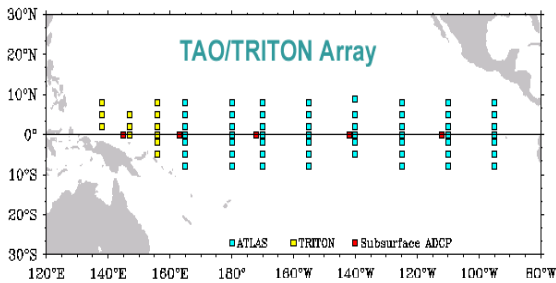
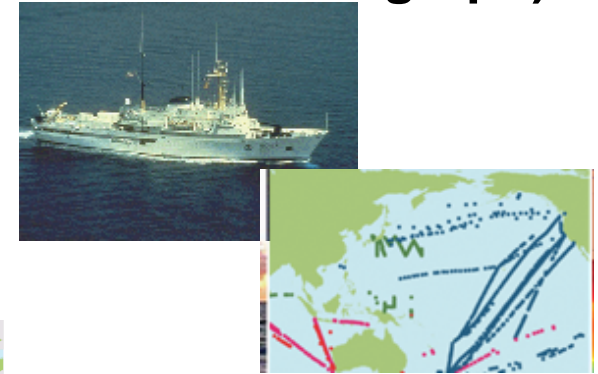
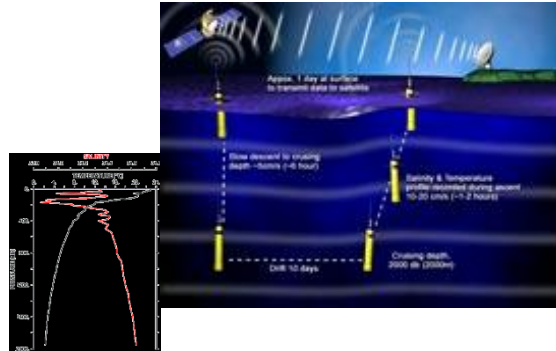


# Real-time ocean observations

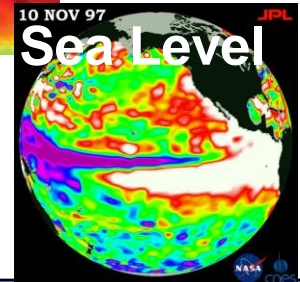
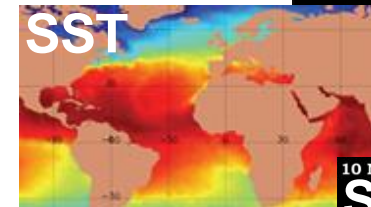
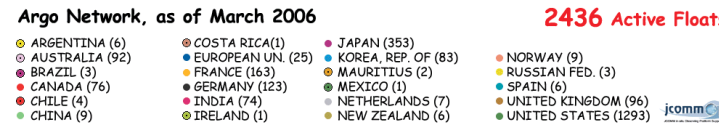
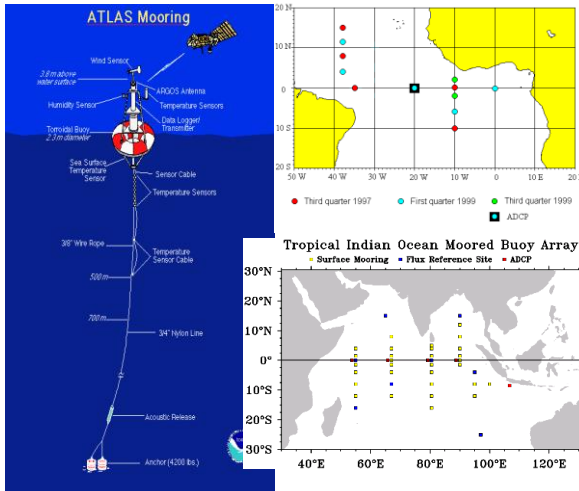
## Moorings

## ARGO floats

## XBT (eXpendable BathiThermograph)

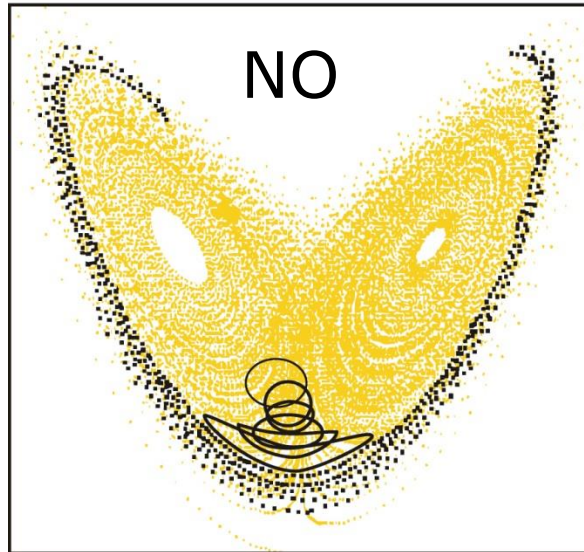
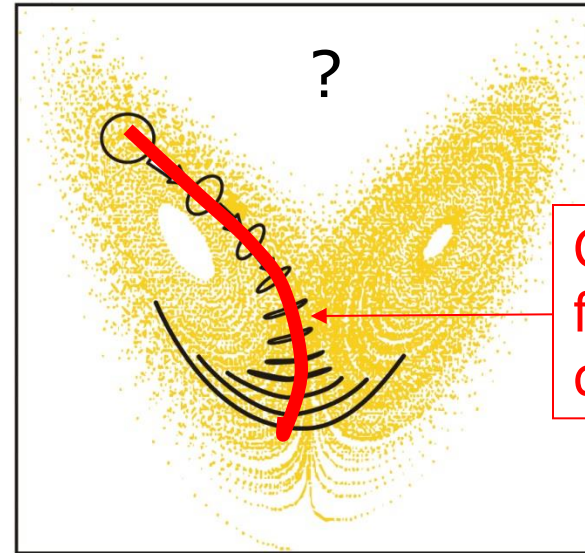
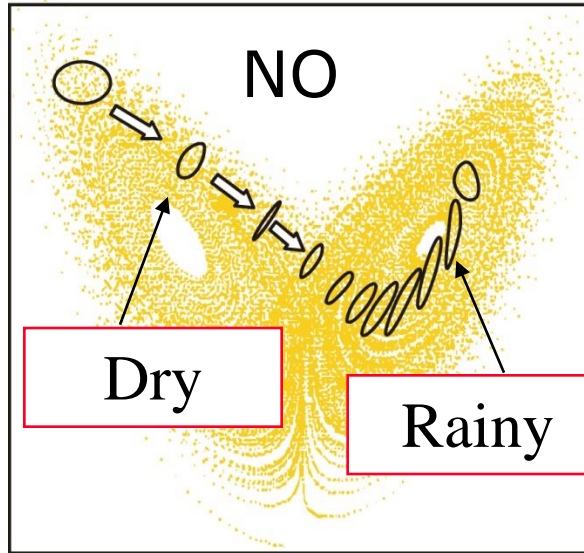


## Satellite



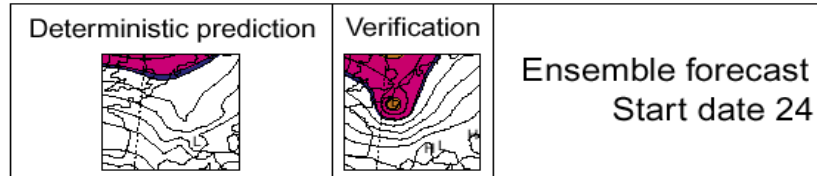
# Why running several forecasts

A farmer is planning to spray a crop tomorrow

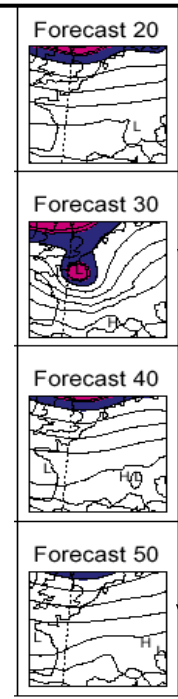
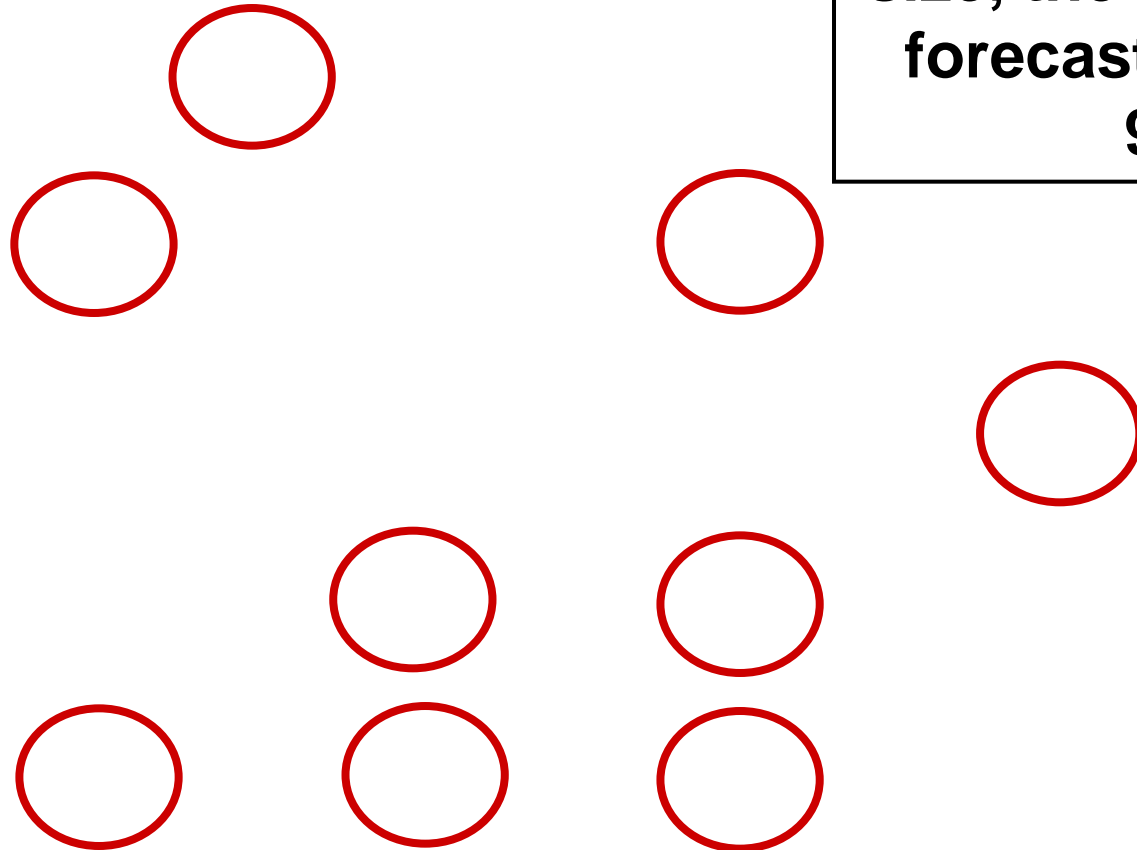


# How many members: ensemble size

ECMWF forecasts (D+42) for the storm Lothar



**With a large enough ensemble size, the very rare event can be forecast with a probability of 9/50, i.e. ~20%**



# And there are systematic errors

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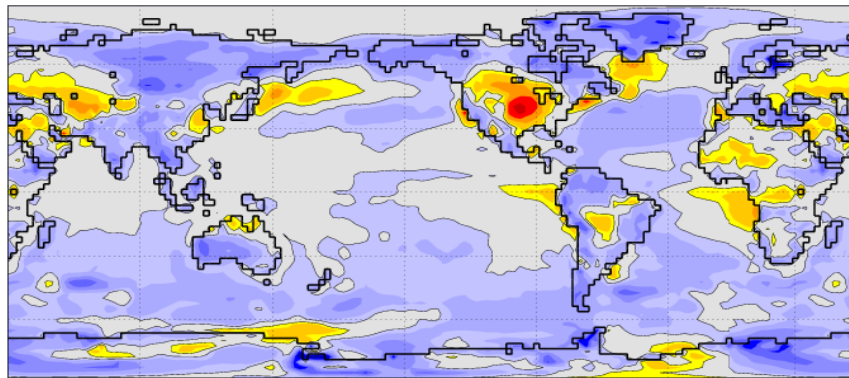
- Model drift is typically comparable to signal
  - Both SST and atmosphere fields
- Forecasts are made *relative* to past model integrations
  - Model climate estimated from 25 years of forecasts (1981-2005), all of which use a 11 member ensemble. Thus the climate has 275 members.
  - Model climate has both a mean and a distribution, allowing us to estimate eg tercile boundaries.
  - Model climate is a function of start date and forecast lead time.
- Implicit assumption of linearity
  - We implicitly assume that a shift in the model forecast relative to the model climate corresponds to the expected shift in a true forecast relative to the true climate, despite differences between model and true climate.
  - Most of the time, the assumption seems to work pretty well. But not always.



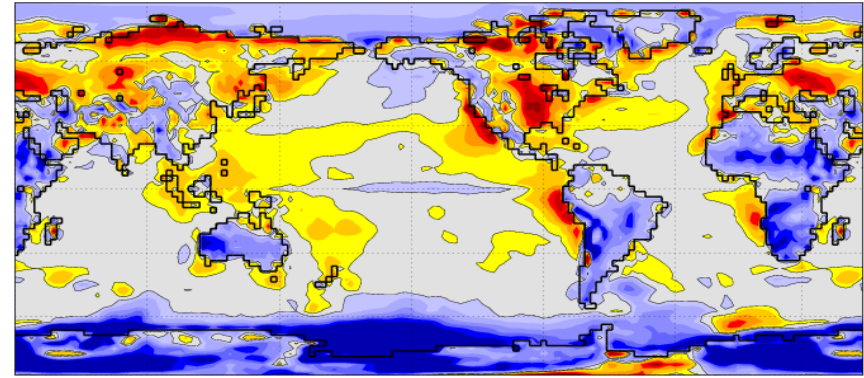
# Mean error

Mean biases (JJA 2mT over 1993-2005) are often comparable in magnitude to the anomalies which we seek to predict

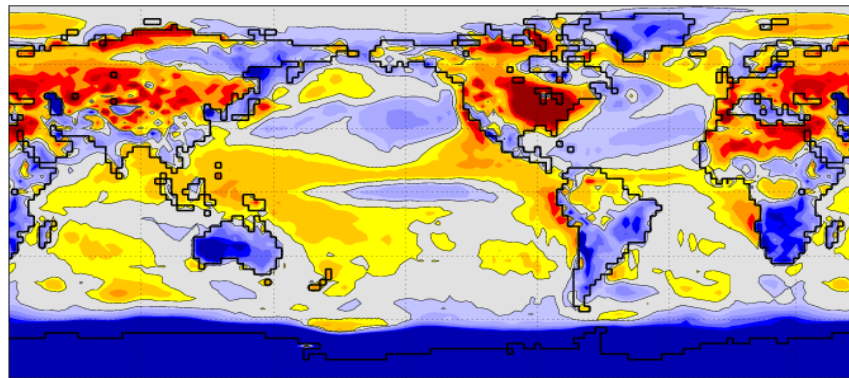
**ECMWF**



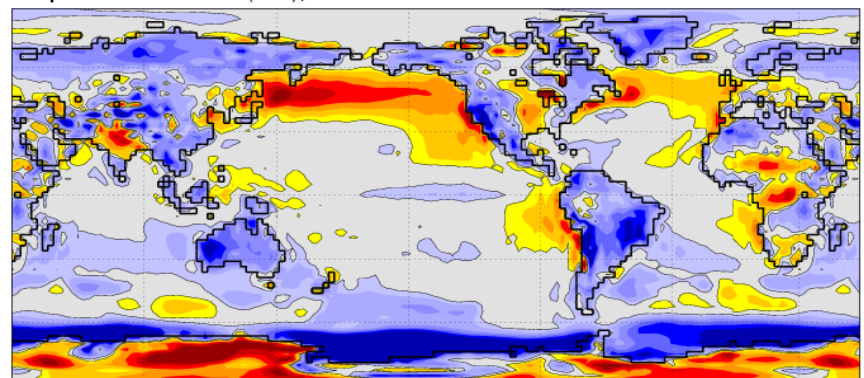
**Met Office**



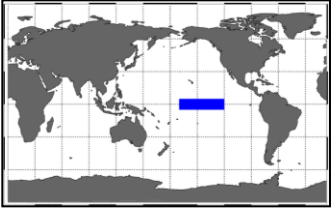
**Météo-France**



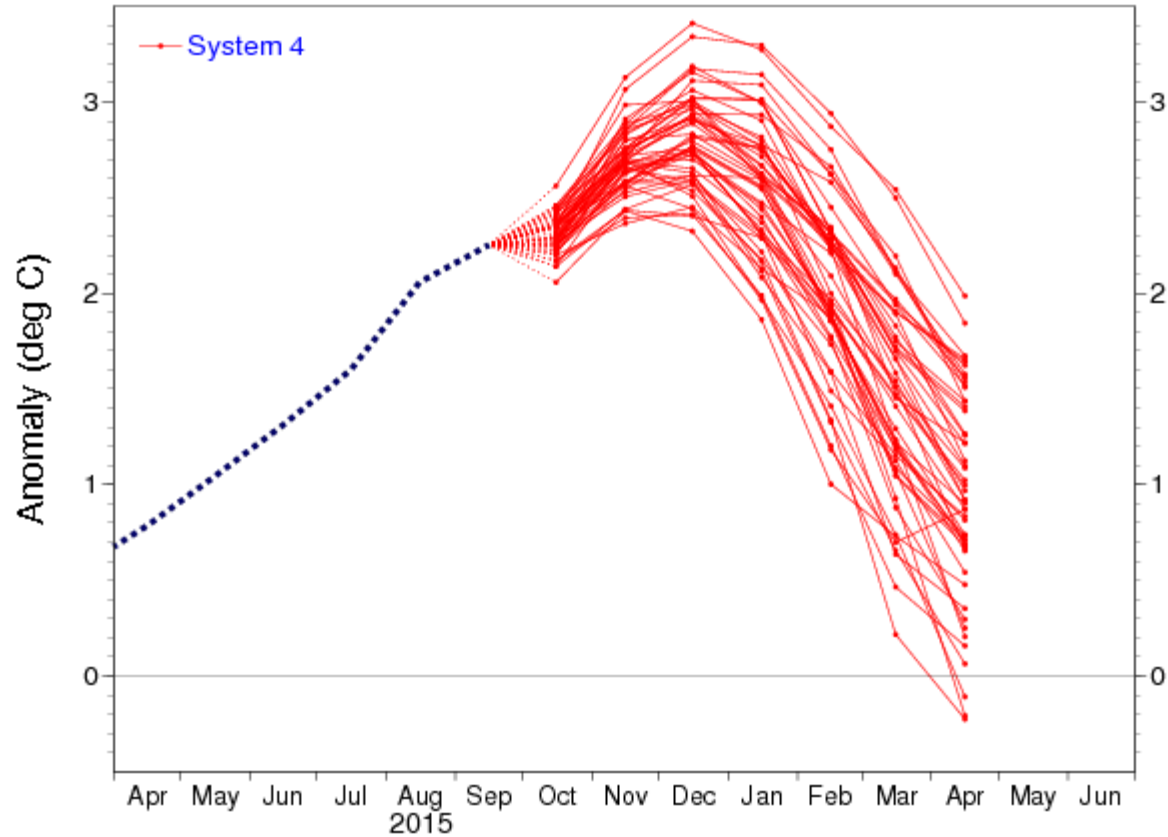
**CFS**



# ENSO ensemble predictions

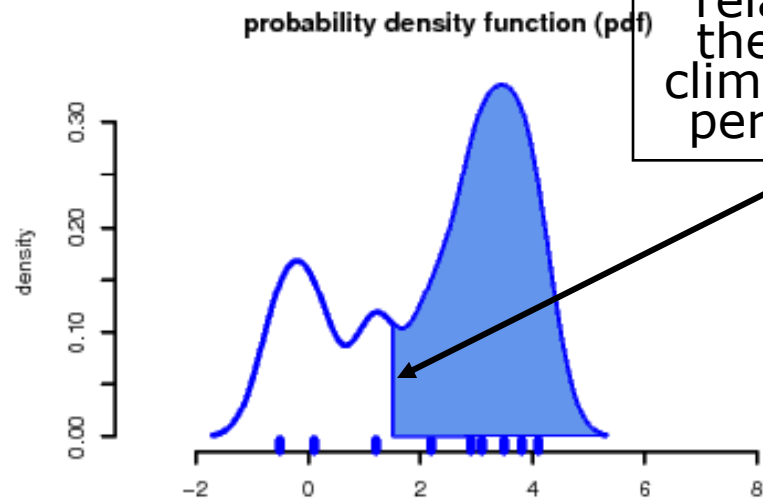
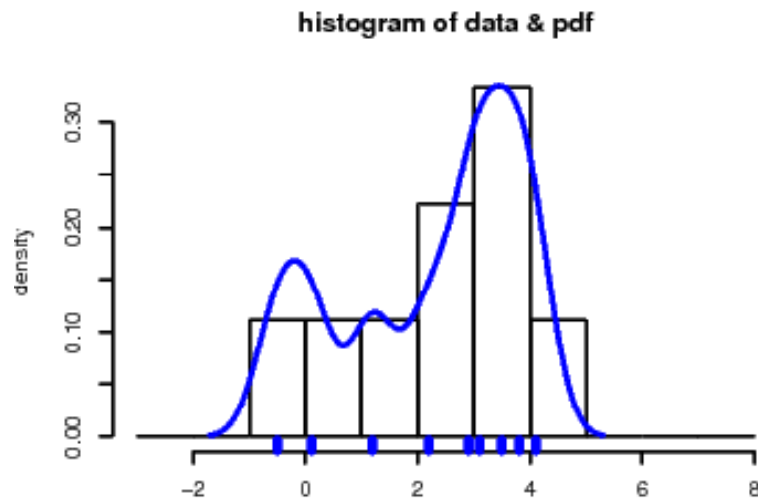
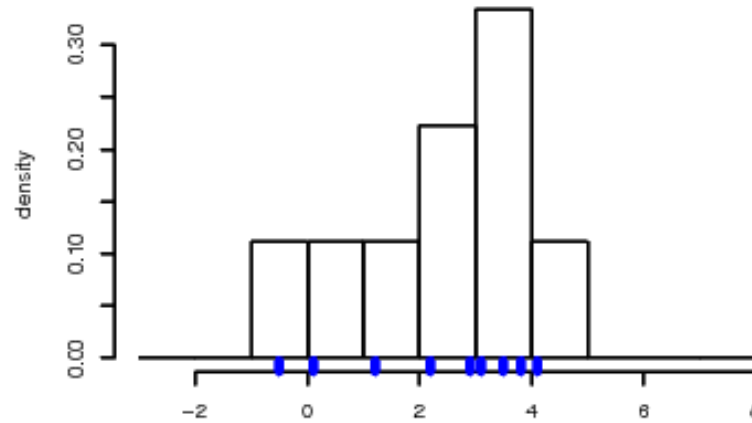


NINO3.4 SST anomaly plume  
ECMWF forecast from 1 Oct 2015  
Monthly mean anomalies relative to NCEP OIv2 1981-2010 climatology



# From ensembles to probability forecasts

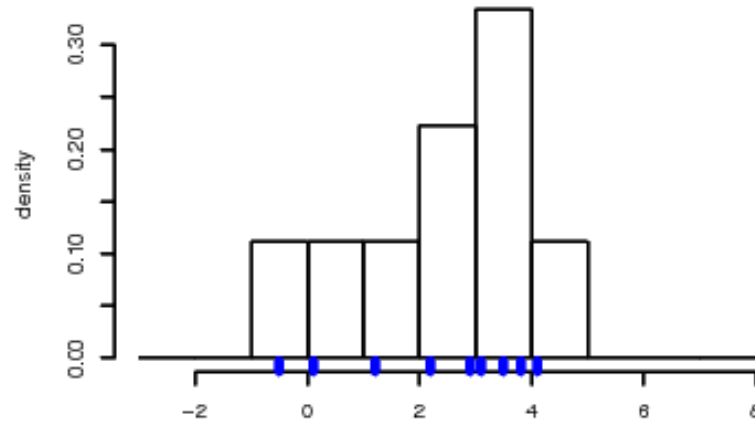
Constructing a probability forecast from a nine-member ensemble



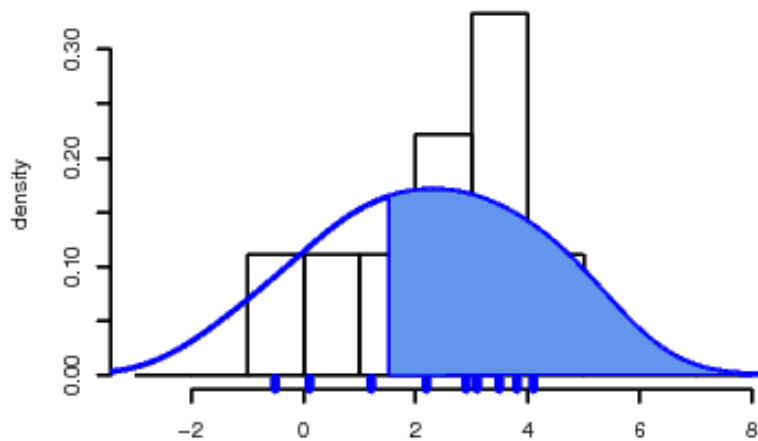
A threshold relative to the model climate (e.g. percentile)

# From ensembles to probability forecasts

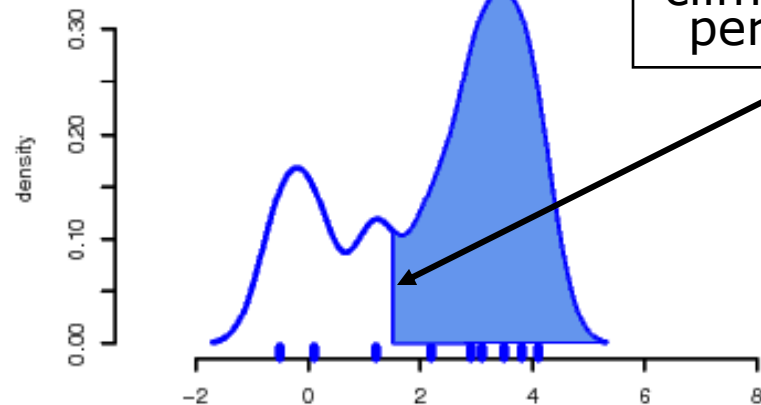
Constructing a probability forecast from a nine-member ensemble



histogram of data & normal-pdf



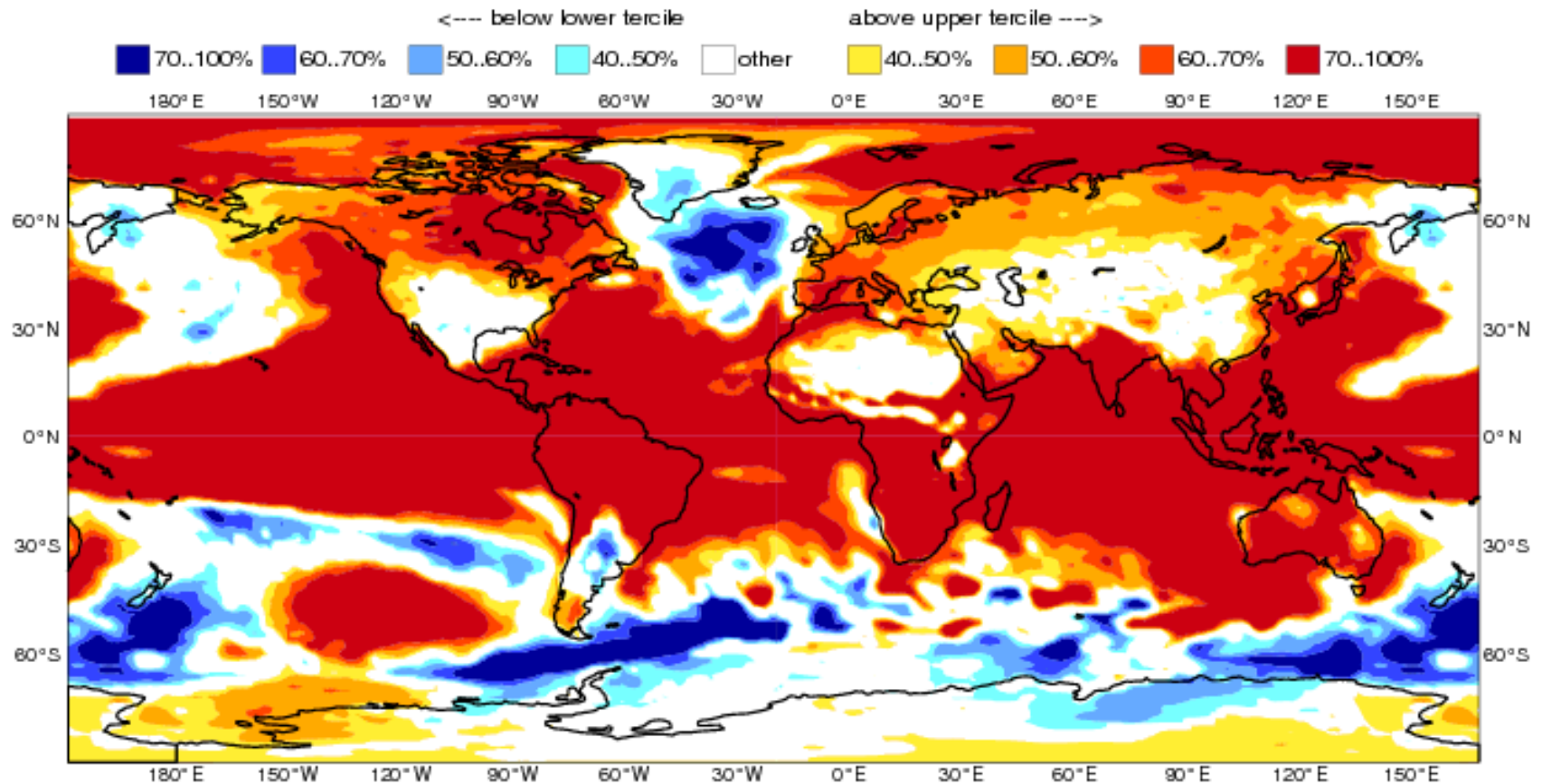
probability density function (pdf)



# Probabilistic prediction

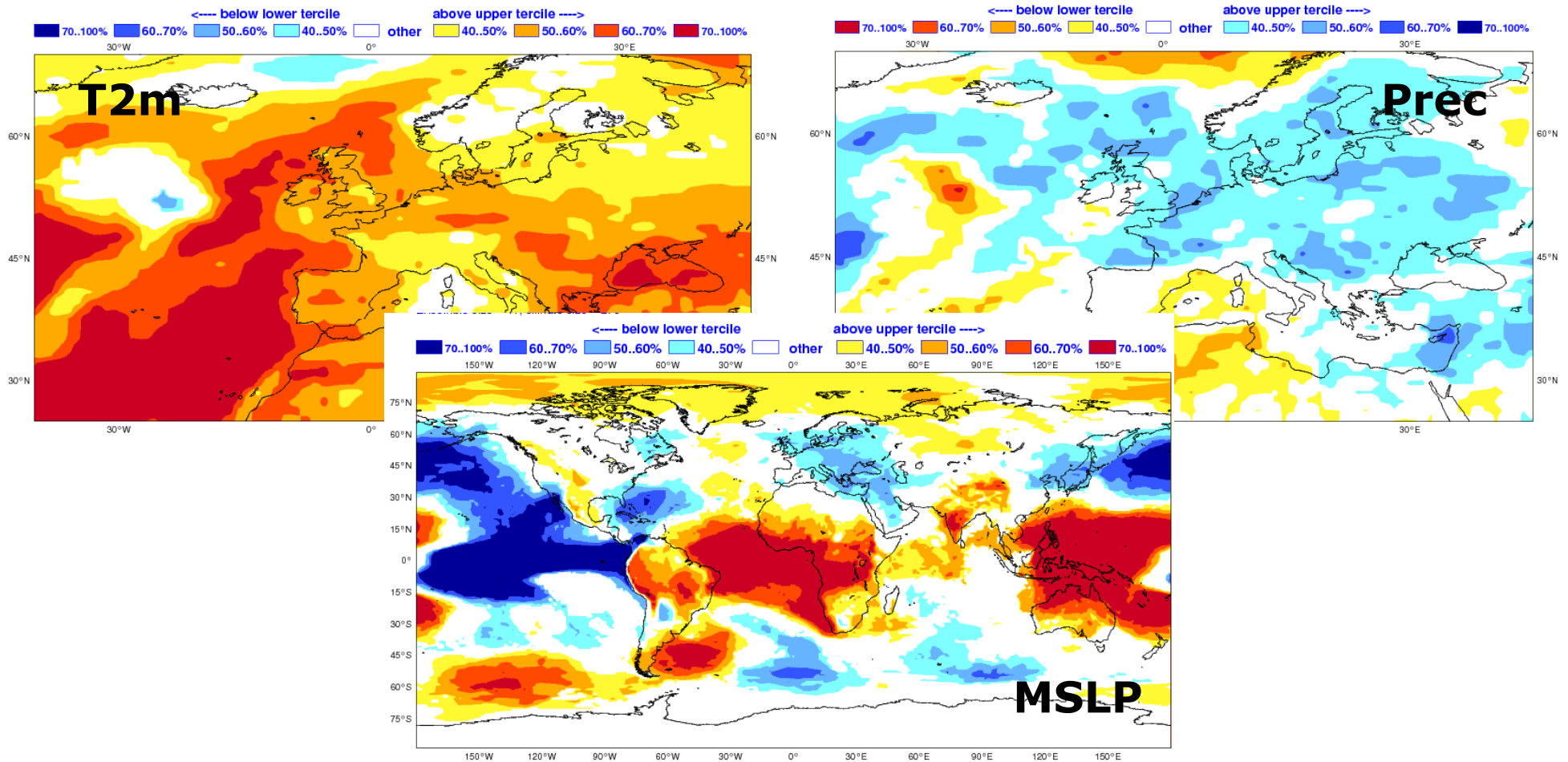
ECMWF Seasonal Forecast  
Prob(most likely category of 2m temperature)  
Forecast start reference is 01/10/15  
Ensemble size = 51, climate size = 450

System 4  
NDJ 2015/16



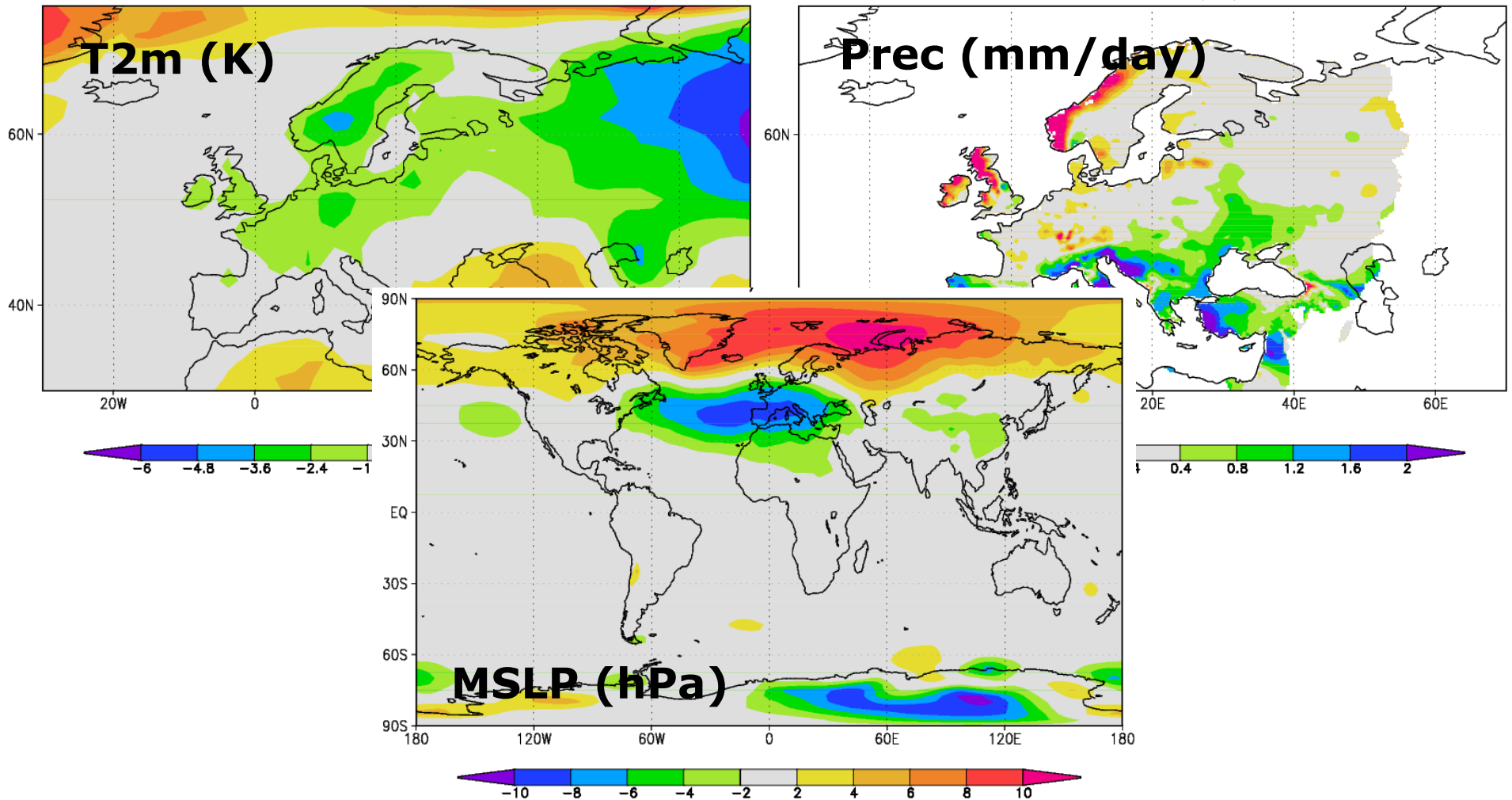
# Probabilistic prediction

## One-month lead DJF 2009-10 System 3 seasonal forecasts: tercile summary



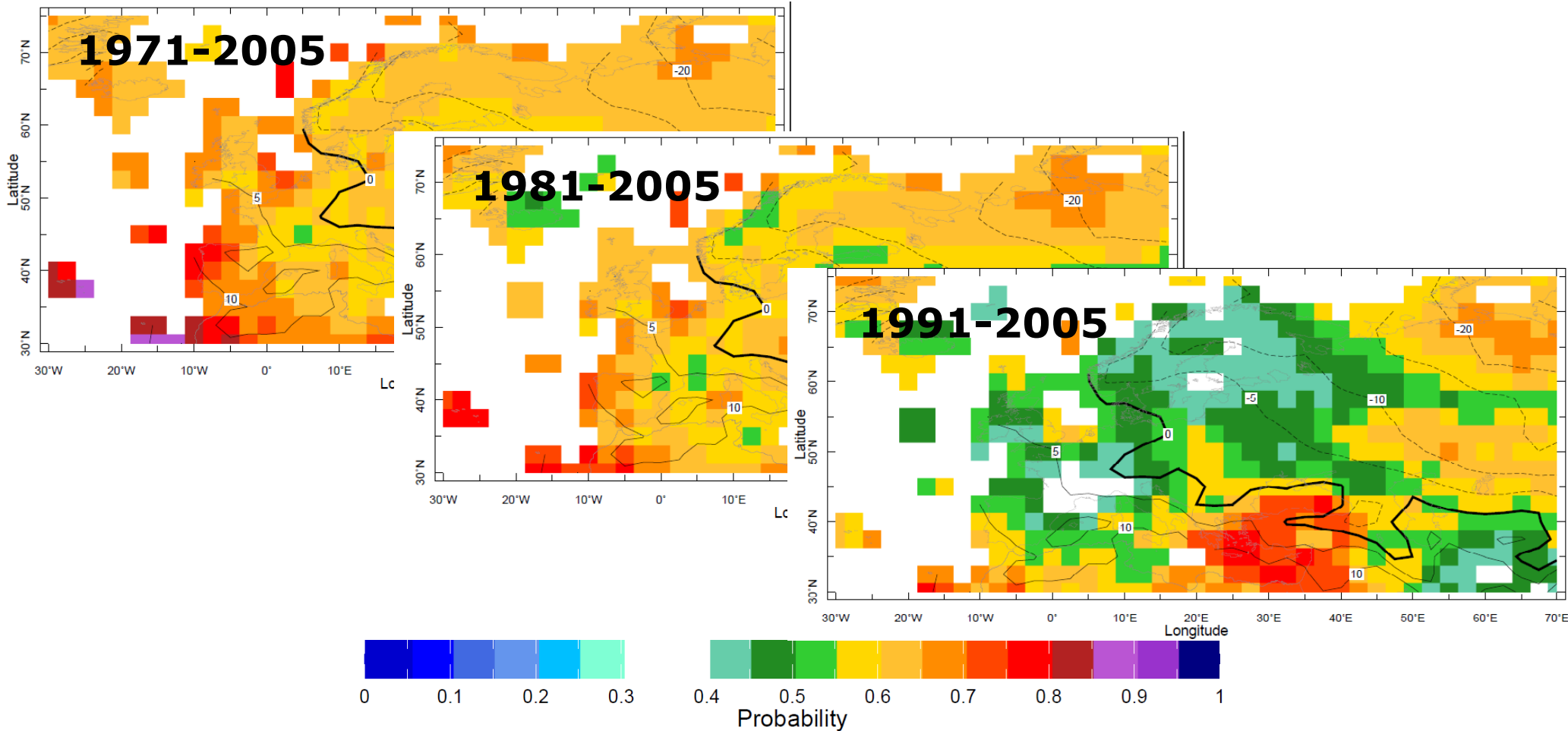
# References: what actually happened

DJF 2009-10 seasonal anomalies wrt 1981-2005.



# Impact of the reference period

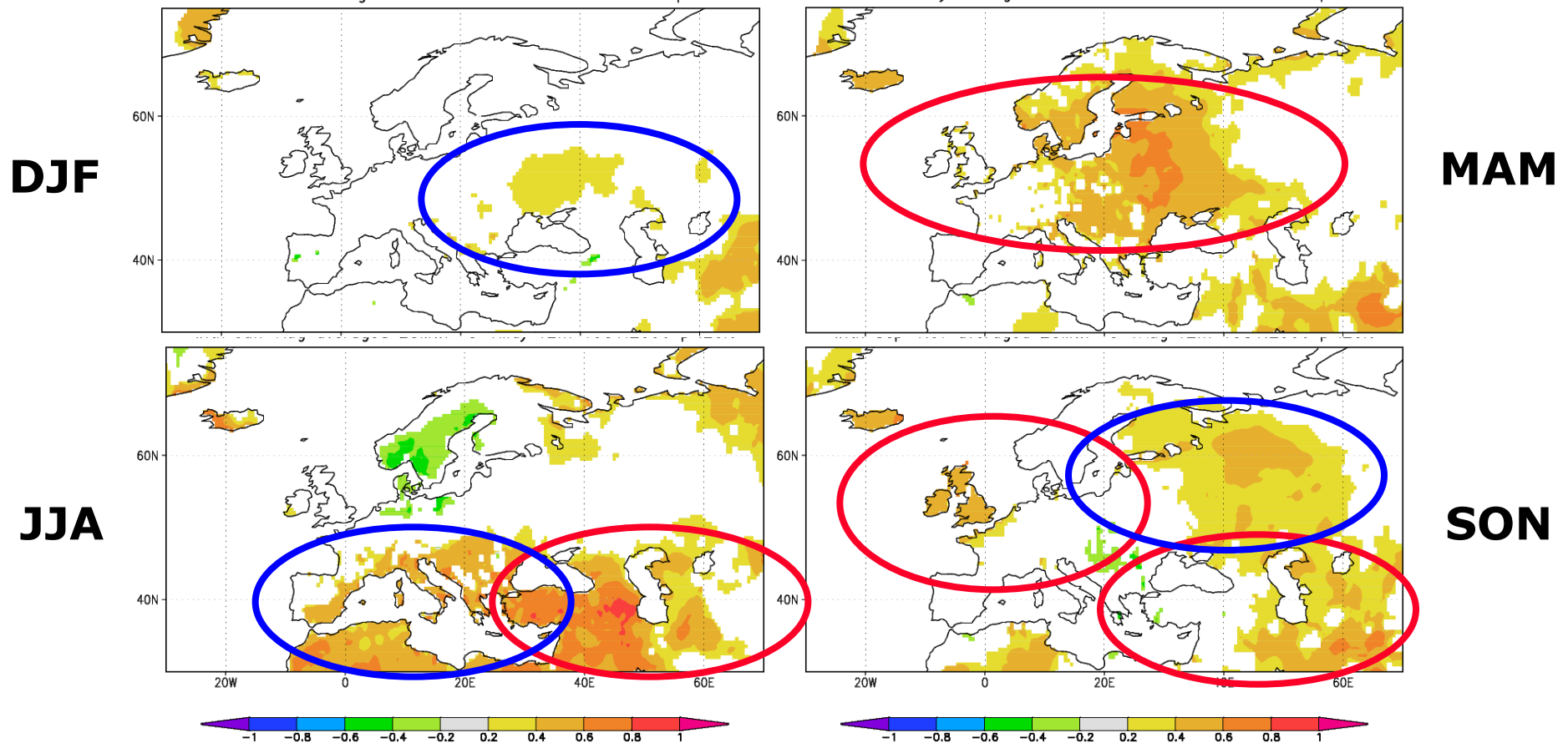
One-month lead DJF 2009-10 IRI (flexible format) temperature forecasts for anom. above the upper tercile





# Regional skill: System 4

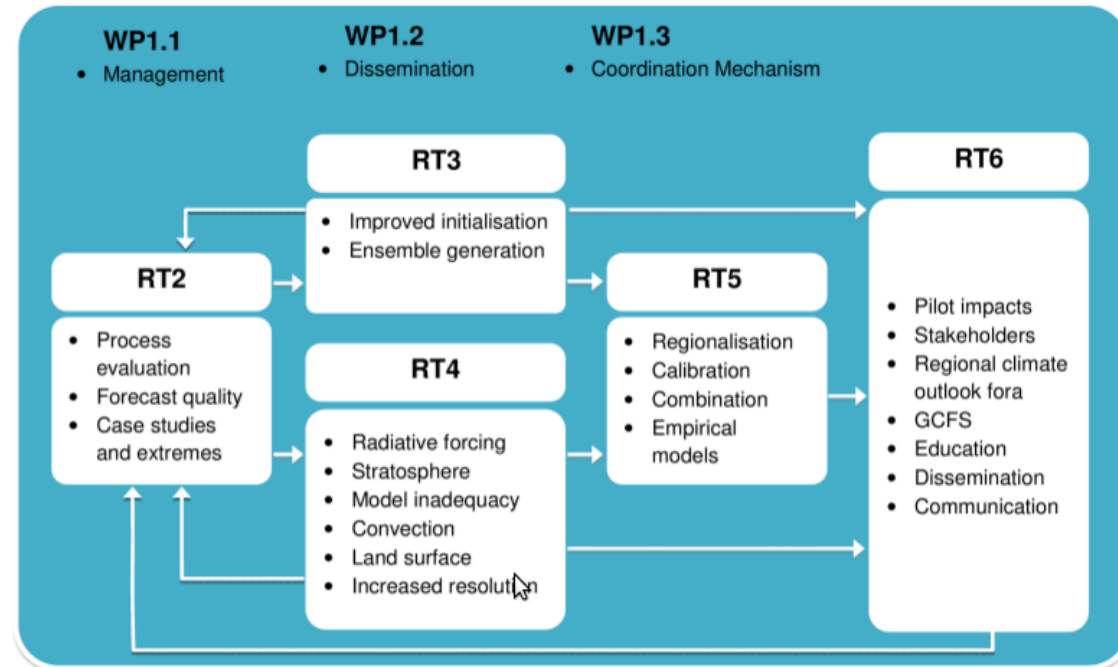
Correlation of System 4 seasonal forecasts of temperature wrt GHCN over 1981-2010. Only values statistically significant with 80% confidence are plotted.



# SPECS FP7, overall strategy

SPECS will deliver *a new generation of European climate forecast systems, including initialised Earth System Models (ESMs) and efficient regionalisation tools to produce quasi-operational and actionable local climate information over land at seasonal-to-decadal time scales with improved forecast quality and a focus on extreme climate events, and provide an enhanced communication protocol and services to satisfy the climate information needs of a wide range of public and private stakeholders.*

Forecast System	Project Partners
CNRM-CM5	CNRM, CERFACS
EC-Earth	KNMI, SMHI, IC3, ENEA
IFS/NEMO	ECMWF, UOXF
IPSL-CM5	CNRS
MPI-ESM	MPG, UniHH
UM	UKMET



WP1.1: Management  
 WP1.2: Dissemination  
 WP1.3: Coordination across EUPORIAS, NACLIM & SPECS  
 RT2: Evaluation of current s2d forecast systems

RT3: Forecast strategies  
 RT4: Improved systems  
 RT5: Calibrated predictions at the local scale

# Summary

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- **Work on initialisation**: initial conditions for all components (including better ocean), better ensemble generation, etc. Link to observational and reanalysis efforts.
  - **Model improvement**: leverage knowledge and resources from modelling at other time scales, drift reduction. More efficient codes and adequate computing resources.
  - **Calibration and combination**: empirical prediction (better use of current benchmarks), local knowledge.
  - **Forecast quality assessment**: scores closer to the user, reliability as a main target, process-based verification.
  - **Improving many processes**: sea ice, projections of volcanic and anthropogenic aerosols, vegetation and land, ...
  - **More sensitivity to the users' needs**: going beyond downscaling, better documentation (e.g. use the IPCC language), demonstration of value and outreach.
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