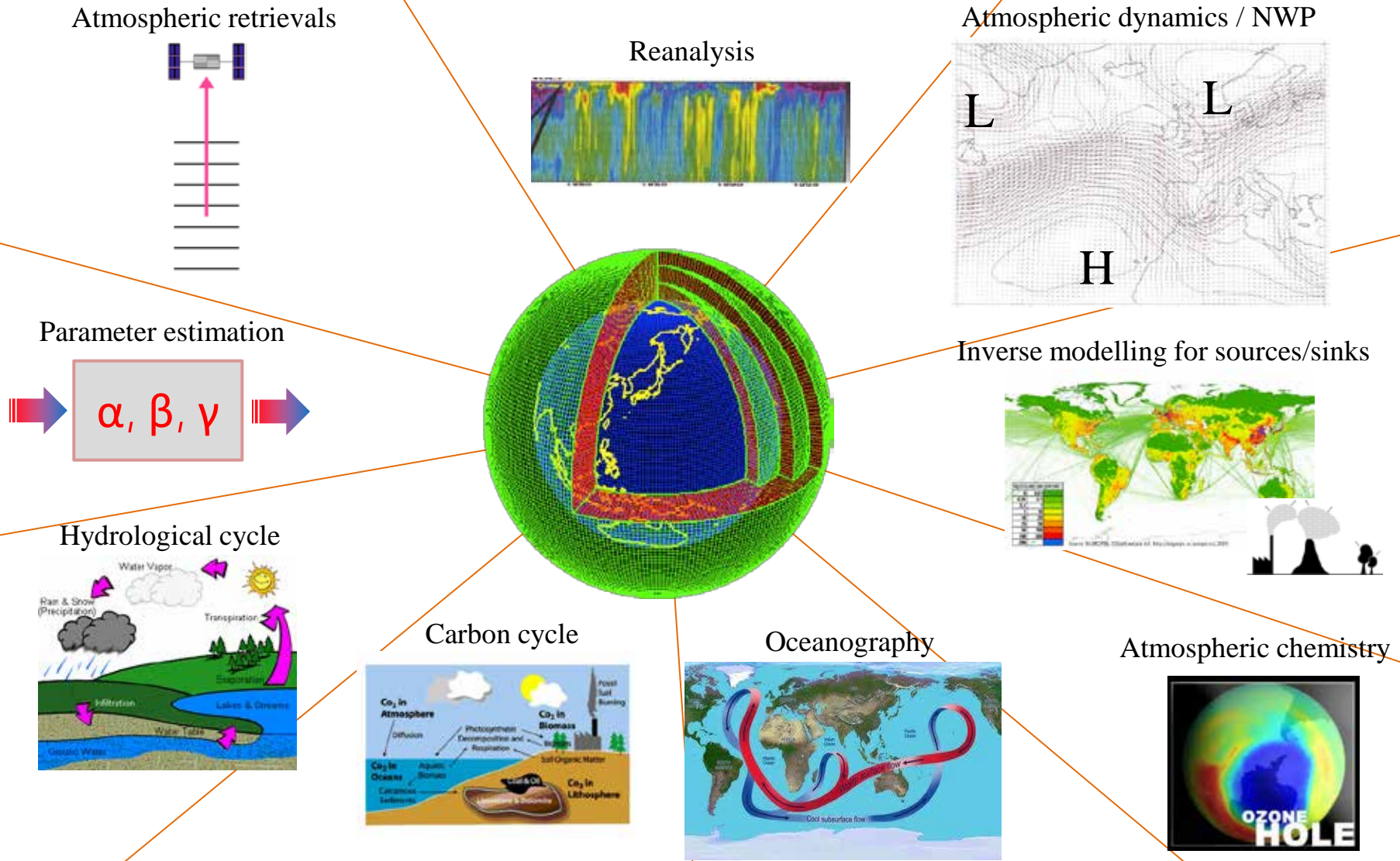


# Applications of Data Assimilation in Earth System Science

Alan O'Neill

National Centre for Earth Observation

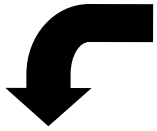
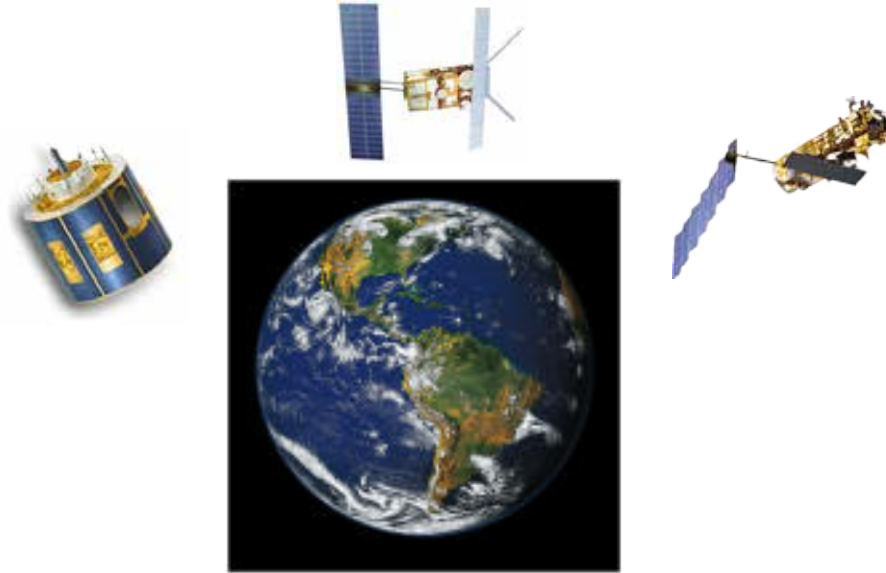
# Applications of data assimilation in the geosciences



# 2050 VISION

- By 2050 the Earth will be viewed from space with better than 1km/1min resolution
- Computer power will be over 100,000 times greater than it is today
- To exploit this technological revolution, the world must be digitised

# Digital World



What users get

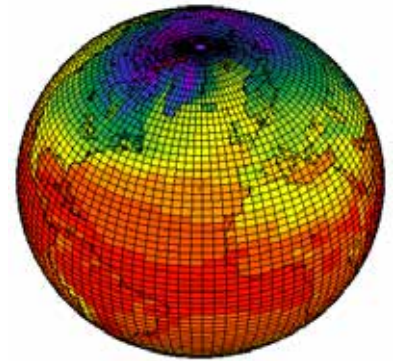
High-tech Sampler

What end-users want

Level 2 for individual sensor

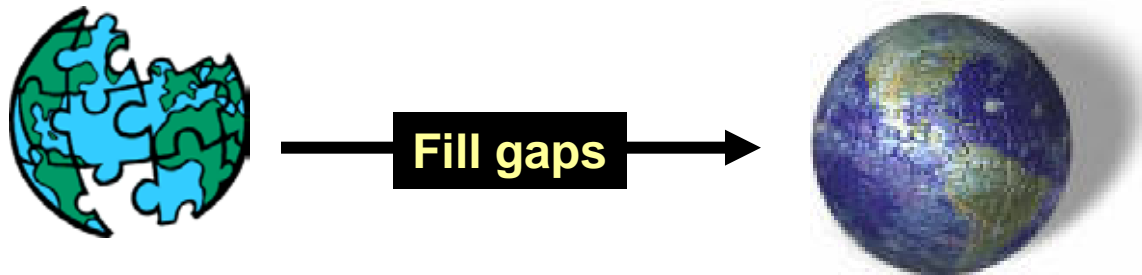
Synthesis via Assimilation of EO data into Earth System Model

4D digital movie of the Earth System



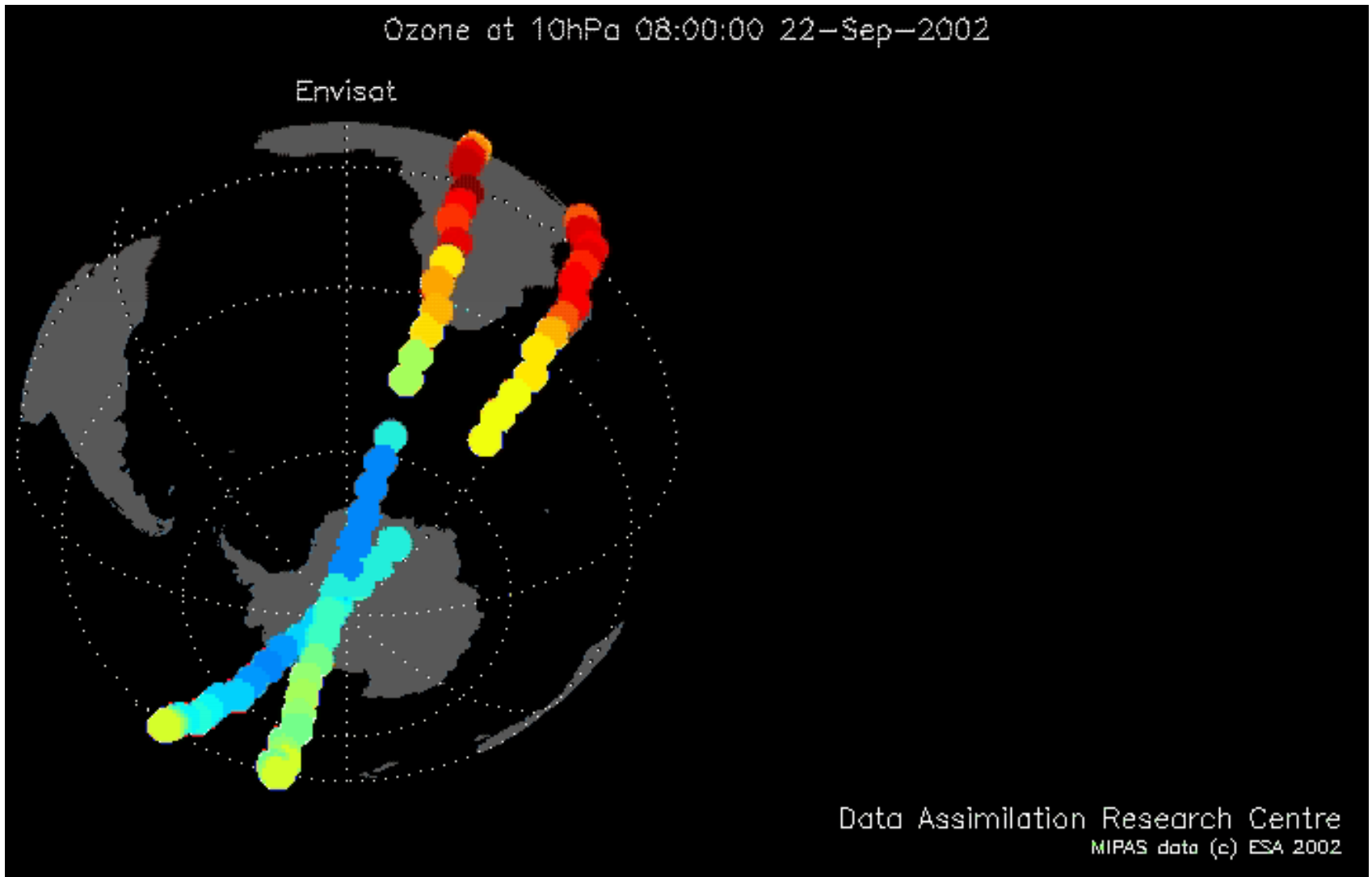
# State Estimations & Physical Interpolation

EO data provide a global view .....  
but have a limited & sequential  
sampling .....



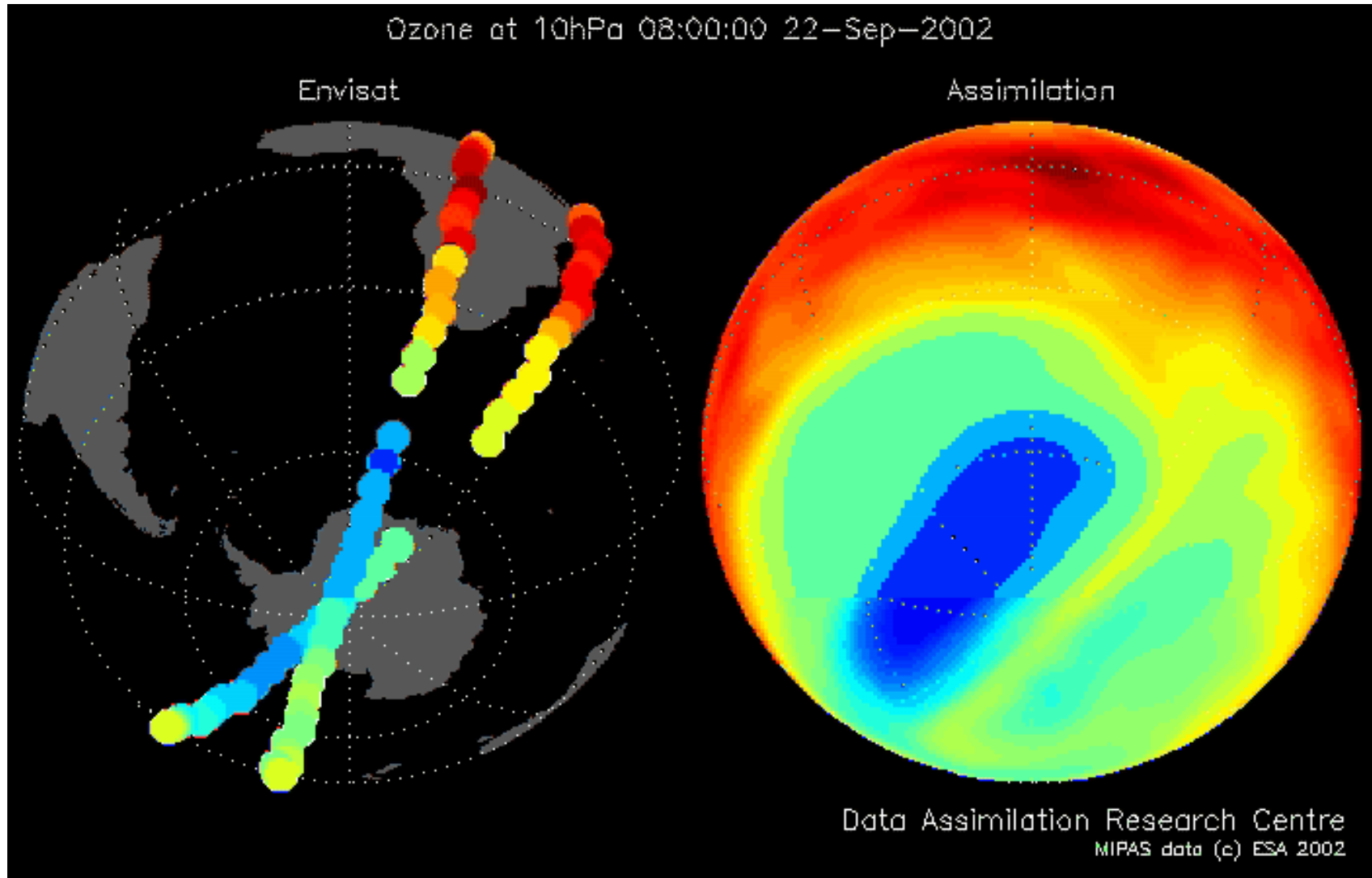
Assimilation of data into models  
provides an **optimal synthesis** of  
heterogeneous observations  
taking account of **errors** and  
**dynamical** principles ...

# Chemical analysis



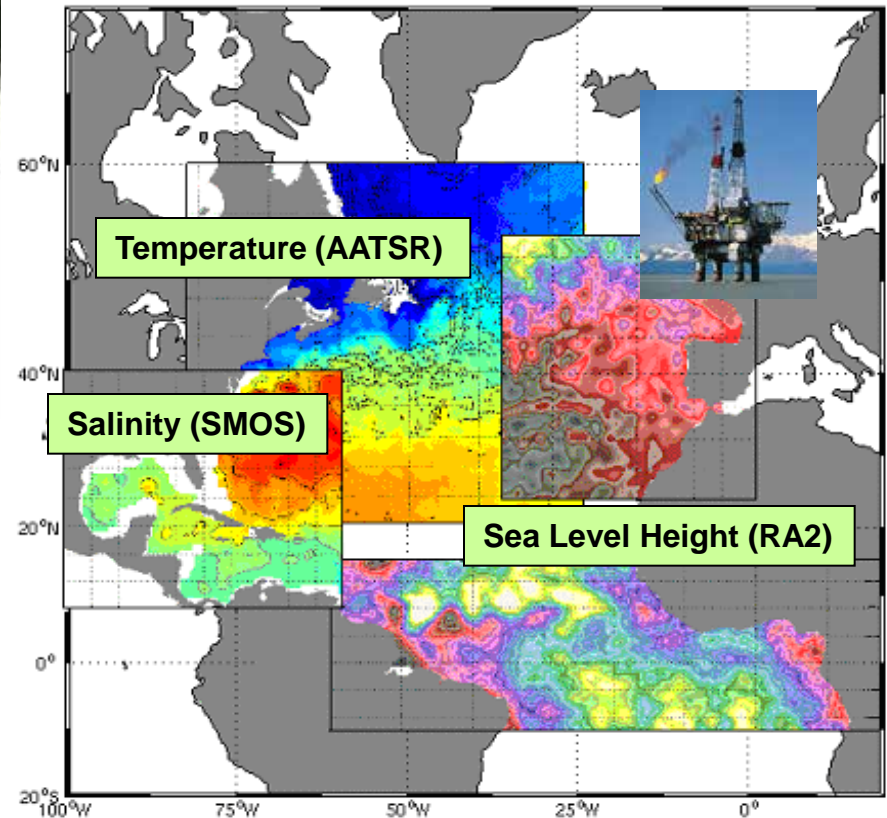
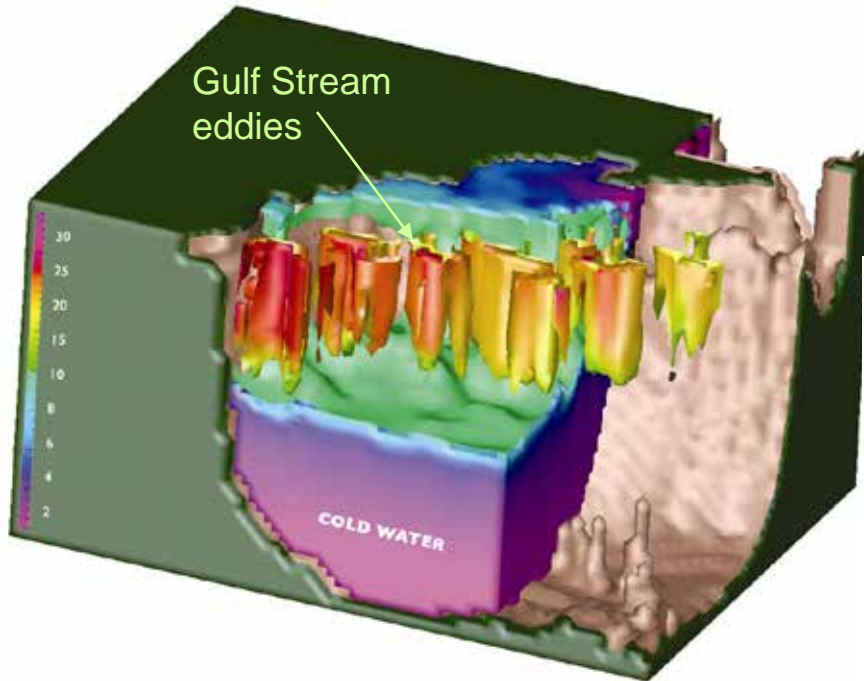
$O_3$  measured by MIPAS/Envisat

# Assimilation of O3 data into GCM



# Operational oceanography

Model dynamics transports EO information from surface (data-rich region) to depth (data-poor region).

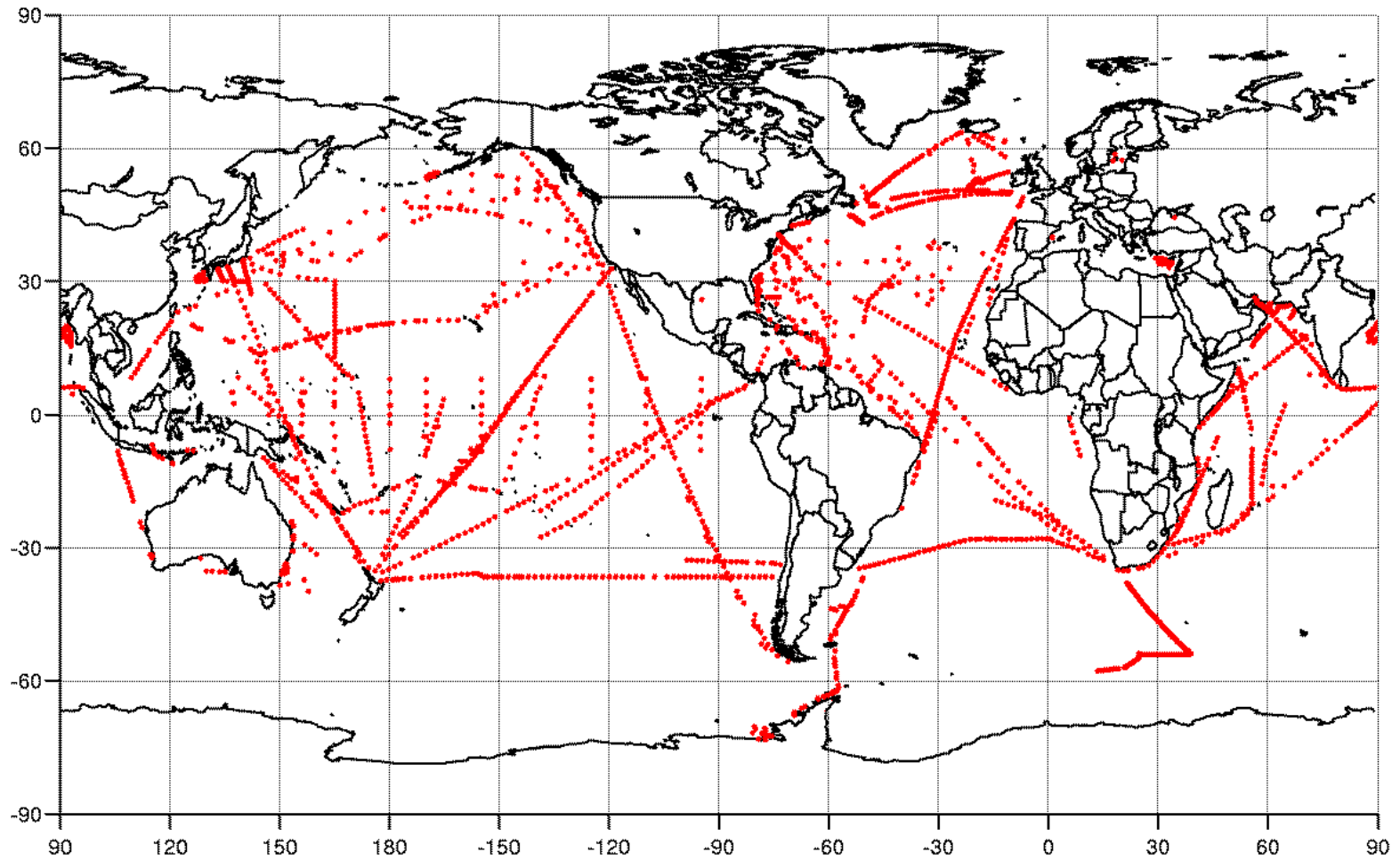


Assimilation of EO data into ocean models provides the best available quantitative picture of the ocean state.

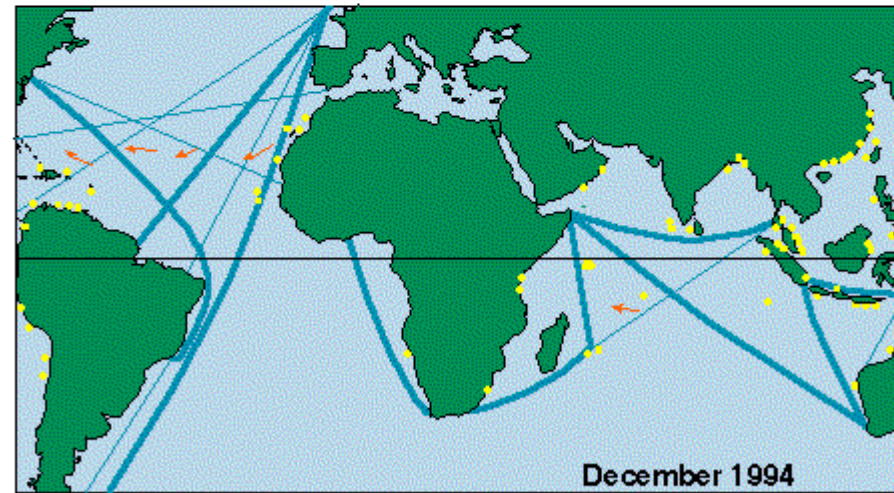
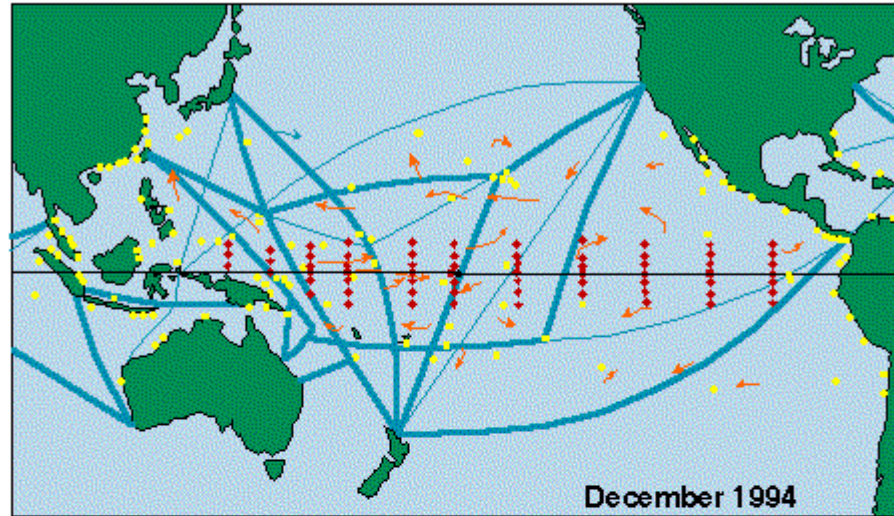
Essential building block for the development of operational marine services (ROSES).



# XBT data assimilated in March 1996.



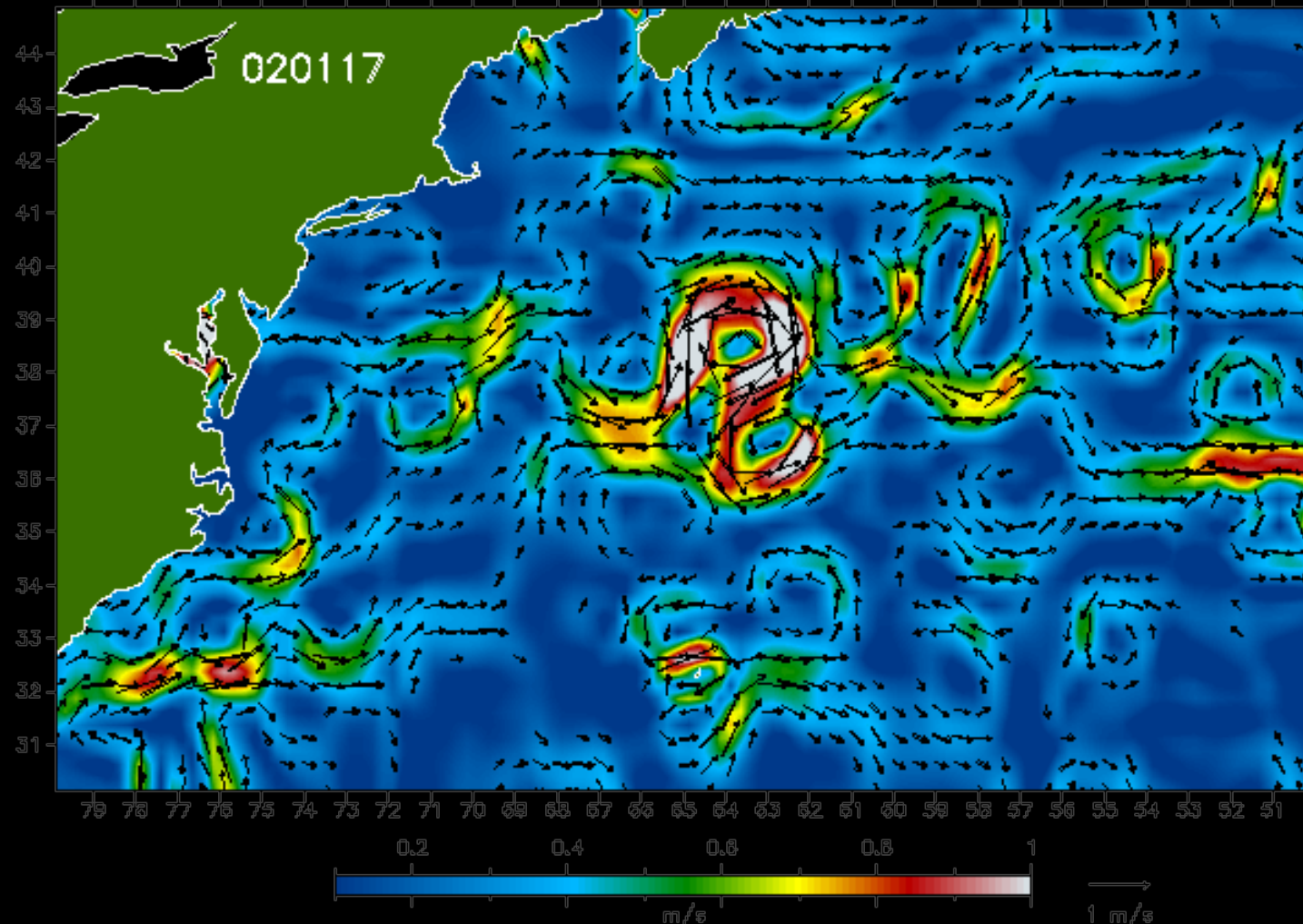
# TOGA In Situ Ocean Observing System Global Tropics



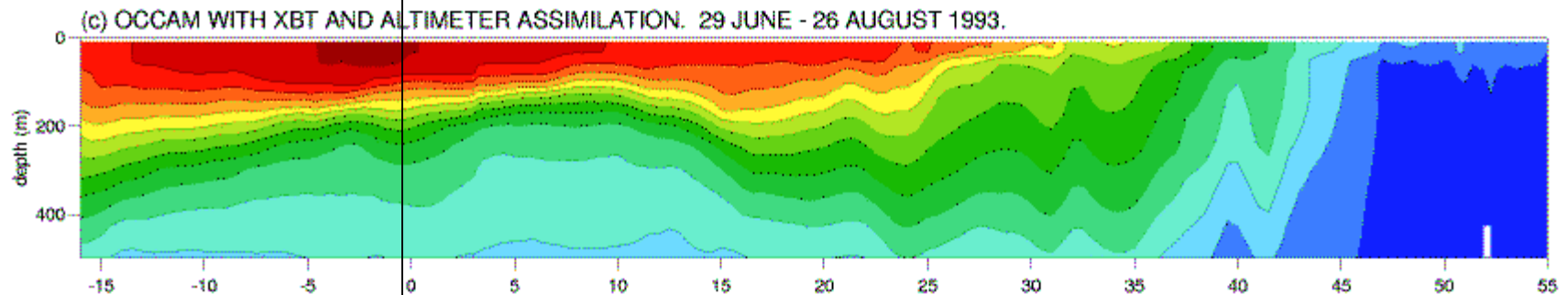
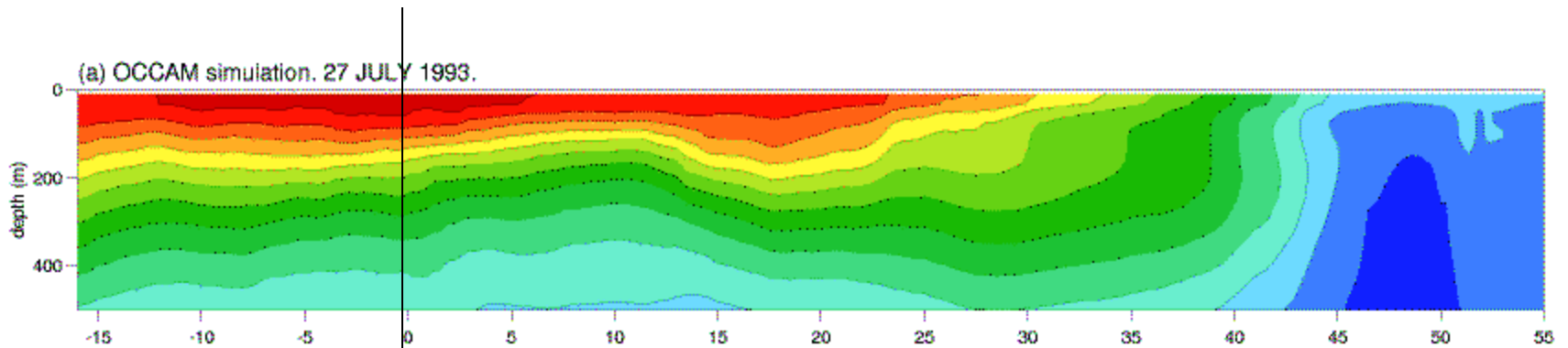
- Note that absolute currents (the non-variable field) will be measurable once the ocean Geoid is known

# Geoid from GOCE

The variable part of currents in the N Atlantic (Gulf Stream)



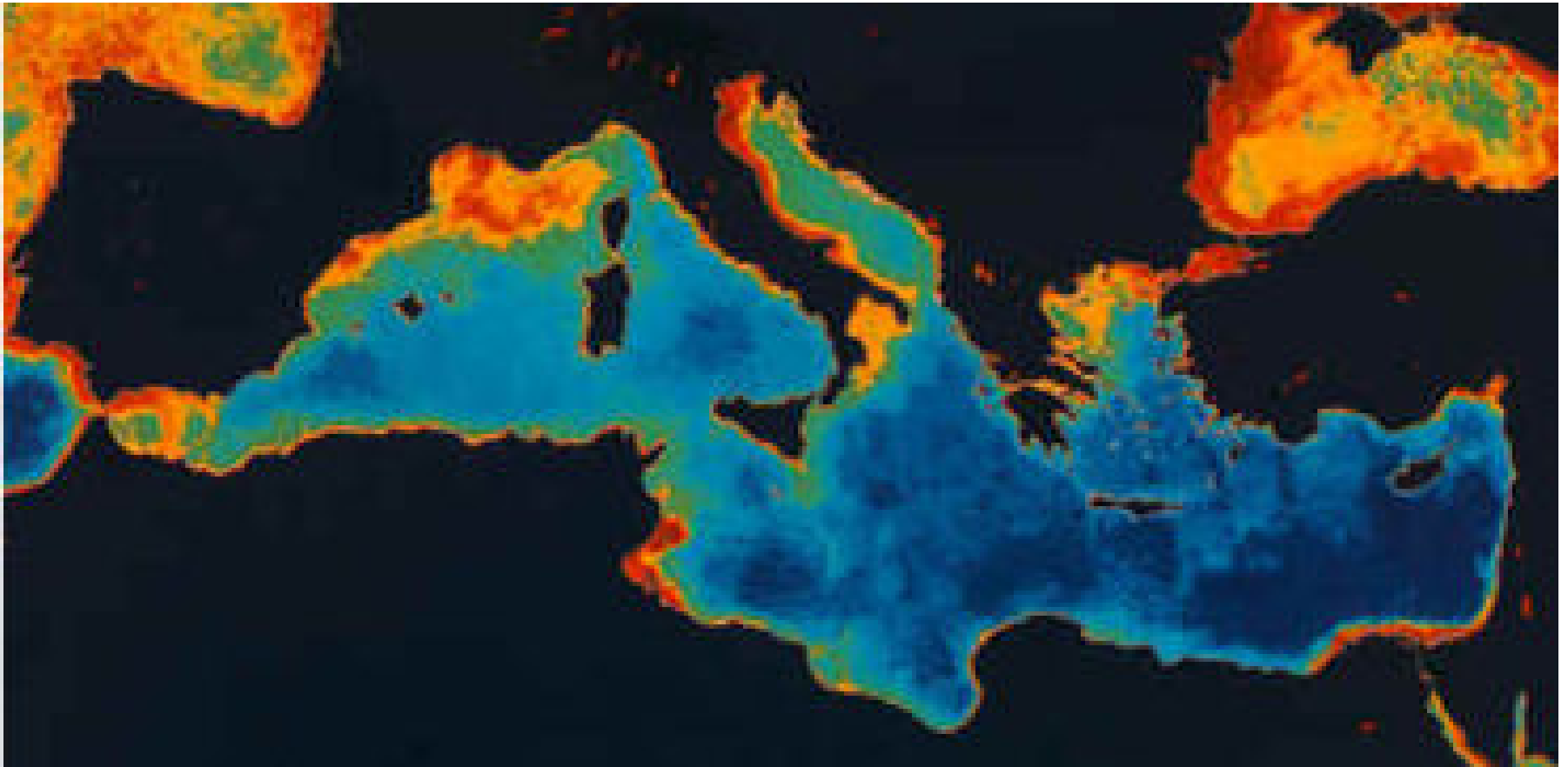
# Cross-section in mid Atlantic



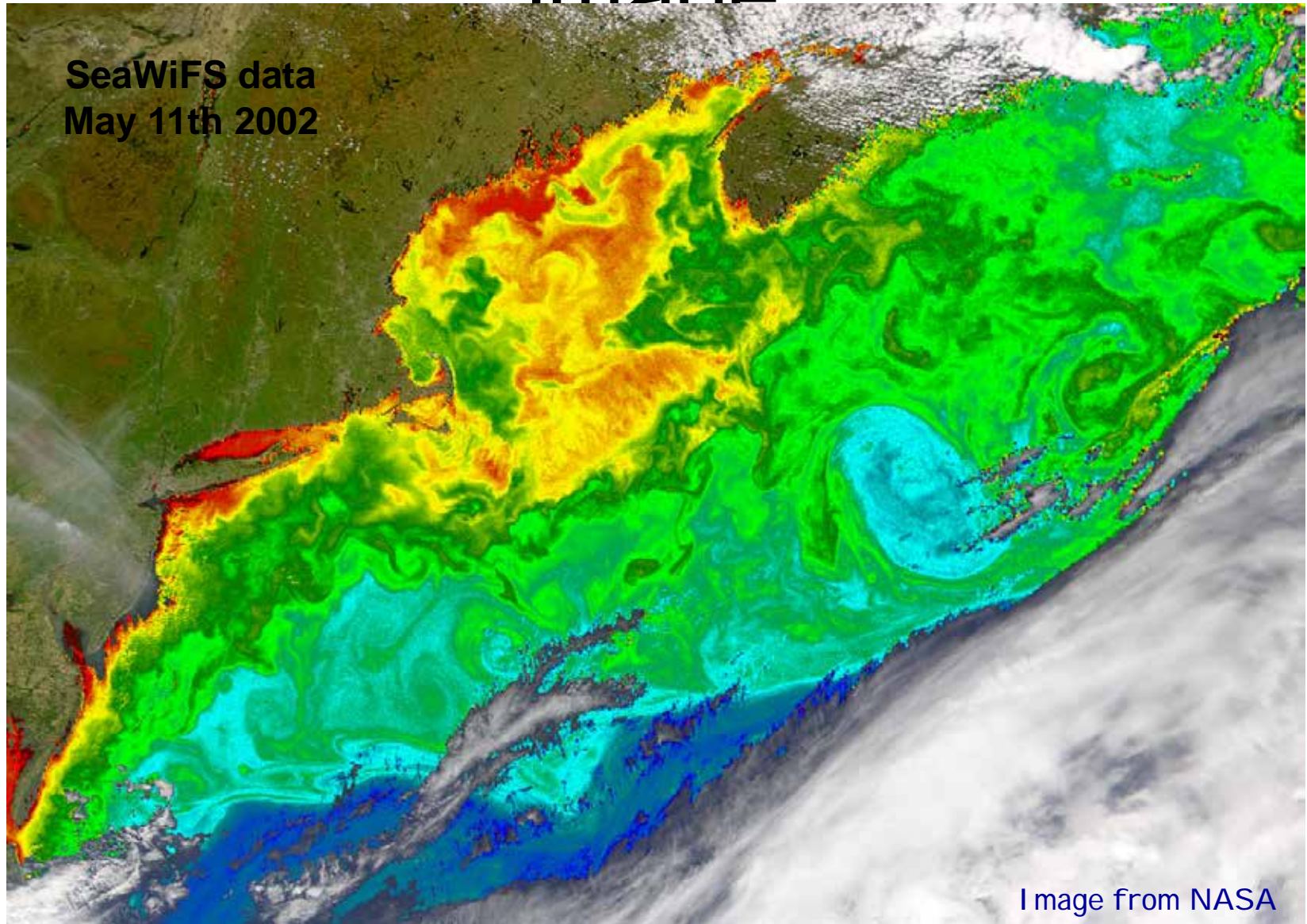
**south**

**north**

# MERIS ocean colour



# Ocean eddies in a chlorophyll image

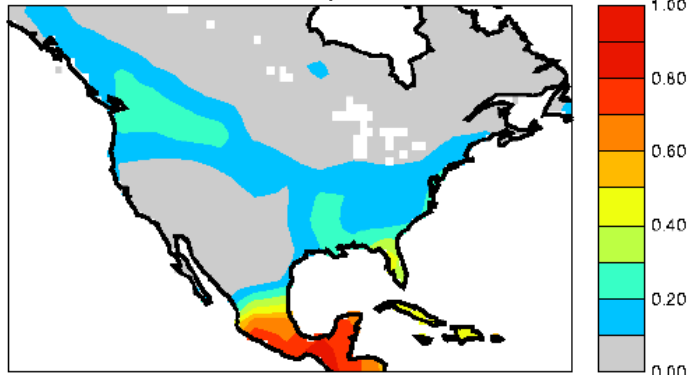


# Land Initialization: Motivation

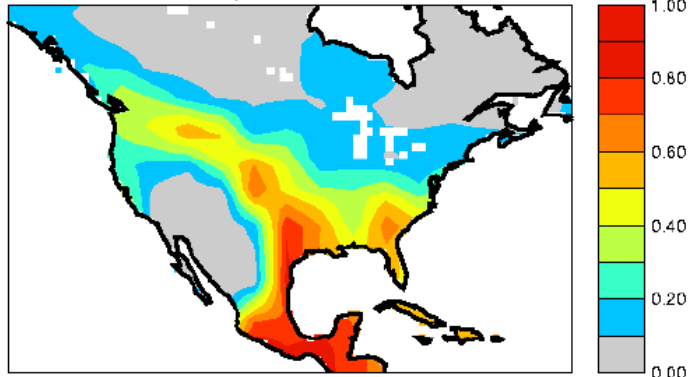
- Knowledge of soil moisture has a greater impact on the predictability of summertime precipitation over land at mid-latitudes than Sea Surface Temperature (SST).

Index of Precipitation Predictability (JJA):

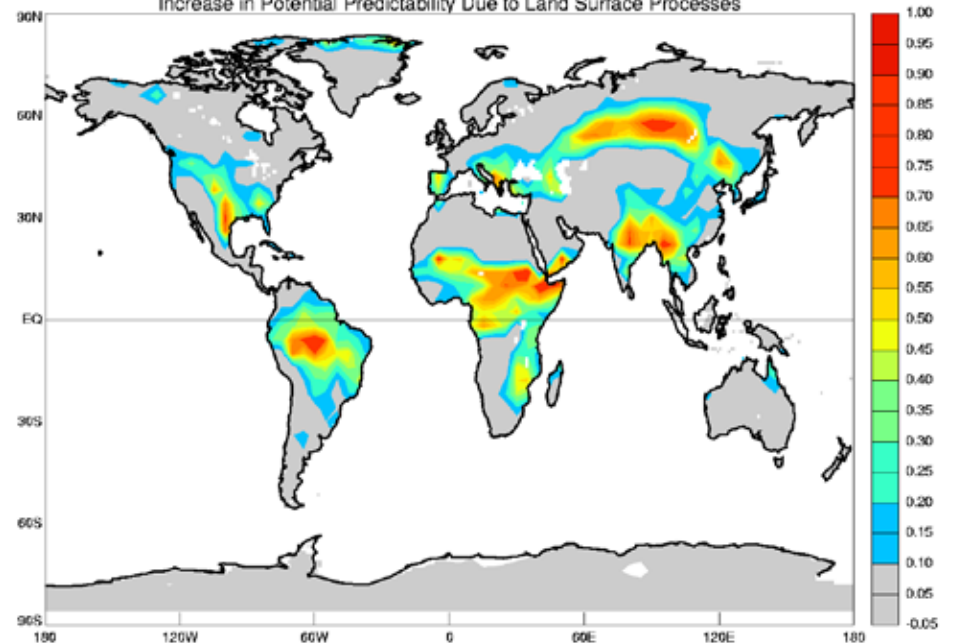
Given Predictability of SSTs



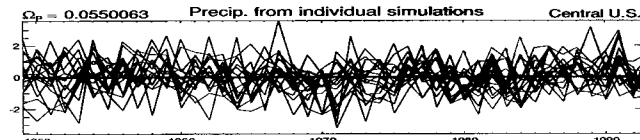
Given Predictability of SSTs and Land Moisture



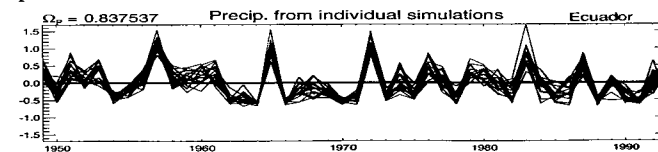
Increase in Potential Predictability Due to Land Surface Processes



$\Omega_P$  near 0: P timeseries different in different simulations:



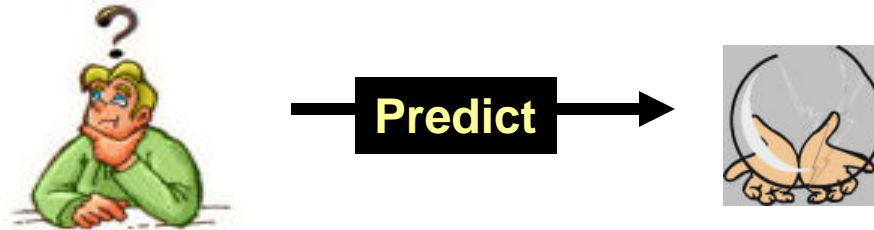
$\Omega_P$  near 1: P timeseries similar in different simulations:



# Environmental Forecasting

EO data are critical for monitoring the global environment but managing risks requires forecasts

.....



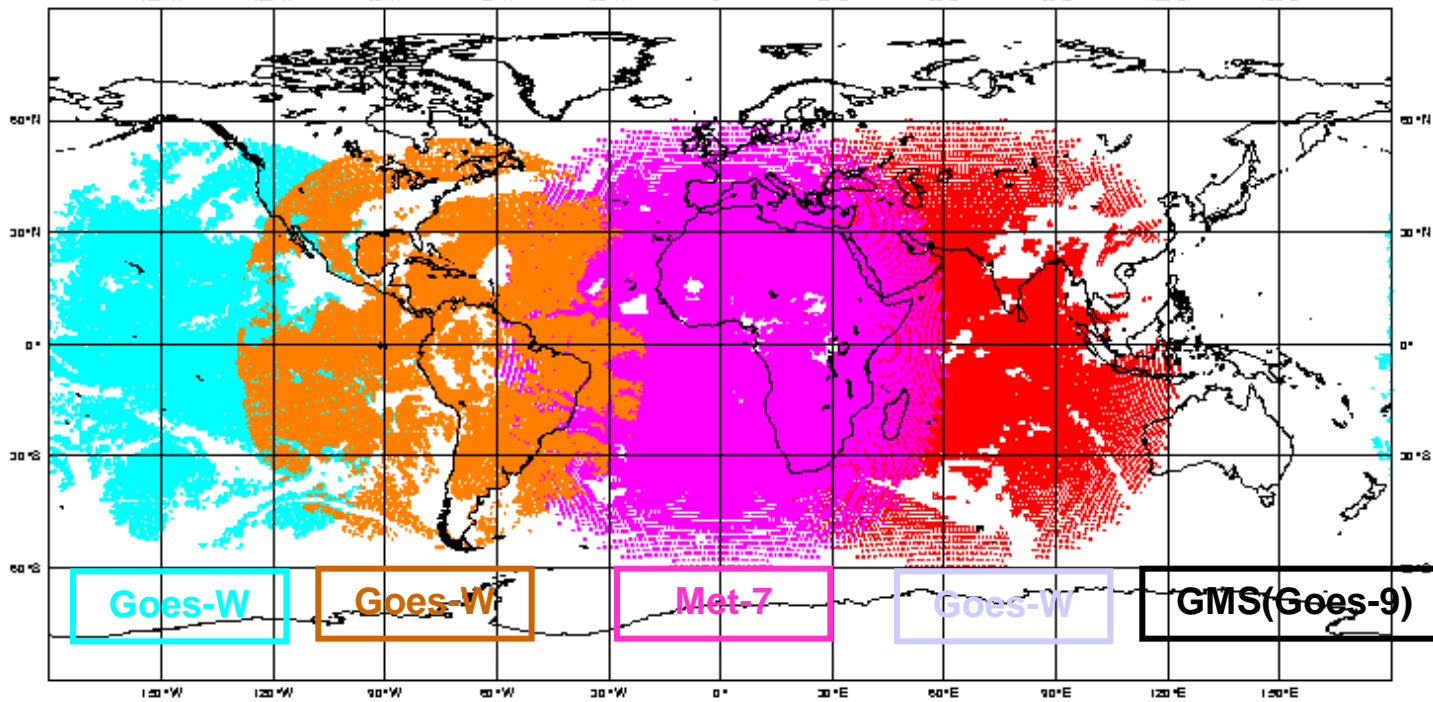
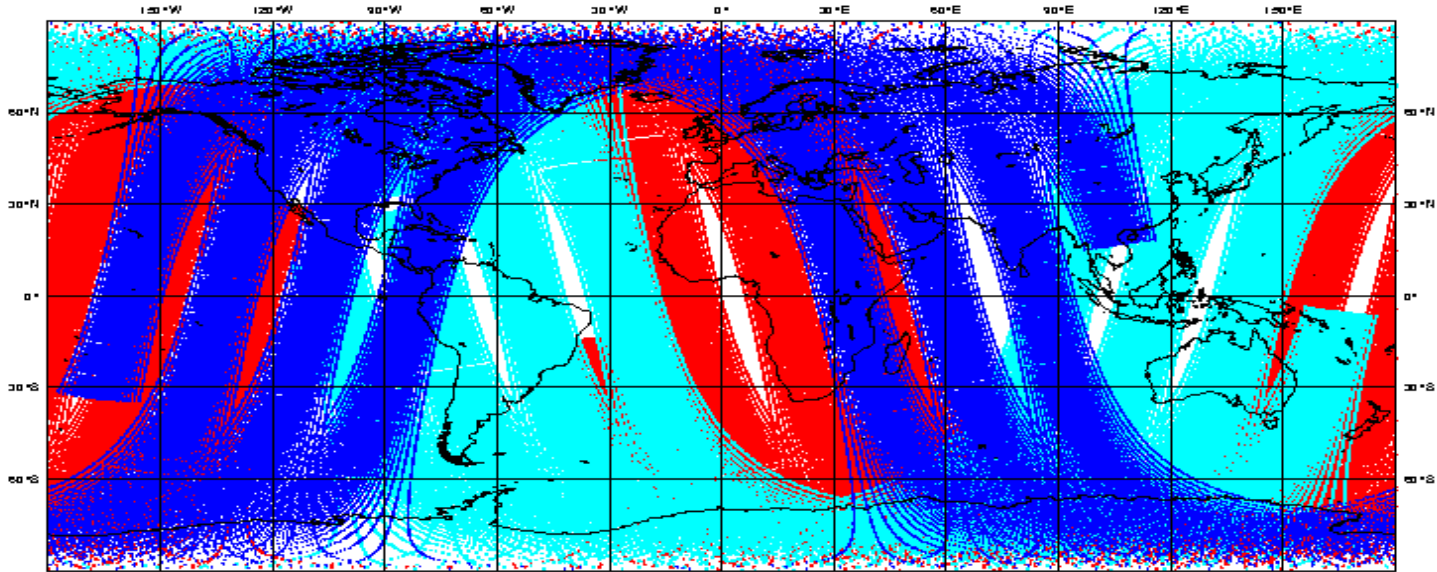
Assimilation of data into models is at the heart of **operational prediction** .....



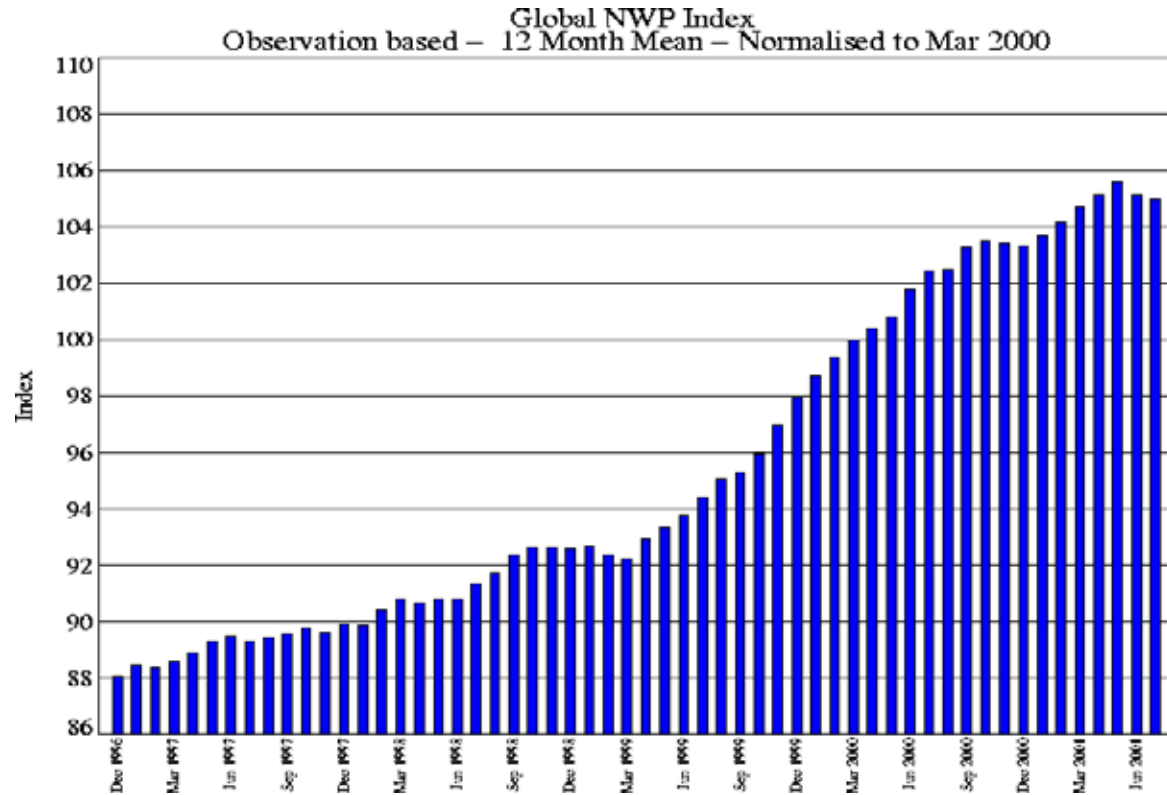
NOAA-15

NOAA-16

NOAA-17



# Impact on NWP at the Met Office



Mar 99. 3D-Var  
and ATOVS

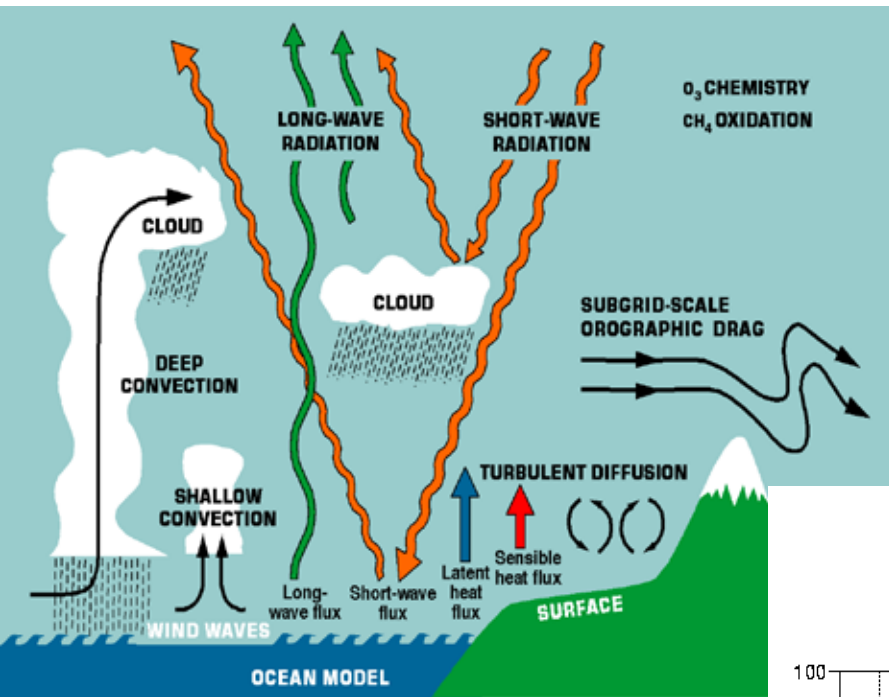
Jul 99. ATOVS over Siberia,  
sea-ice from SSM/I

Oct 99. ATOVS as radiances,  
SSM/I winds

May 00. Retune  
3D-Var

Feb/Apr 01. 2nd satellites,  
ATOVS + SSM/I

# Weather forecasting

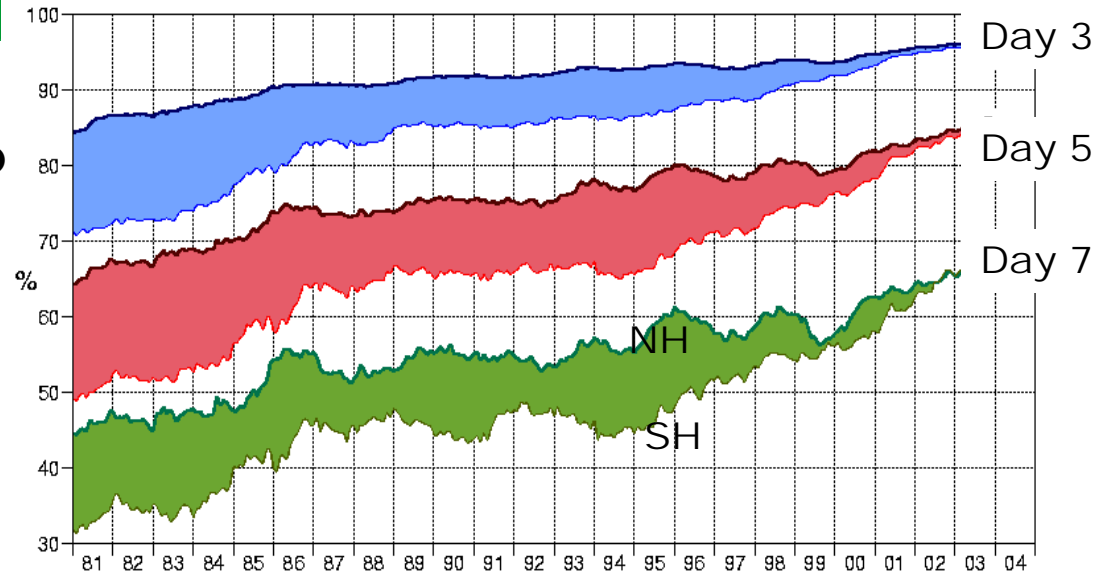


## Numerical Weather Prediction:

- ∅ Sophisticated atmospheric models.
- ∅ Most mature assimilation techniques (able to ingest sounding radiances).
- ∅ Very big user of EO data.

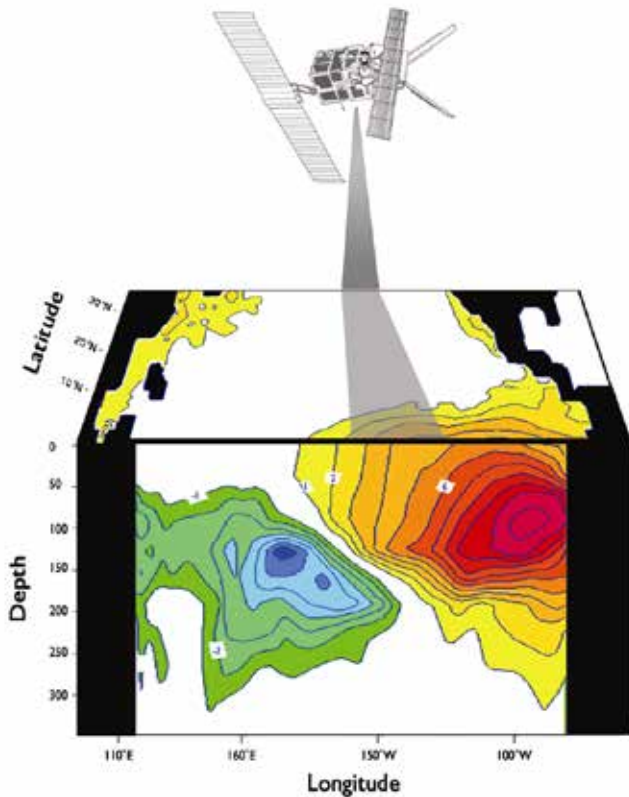
Satellite data have contributed to the continuous improvement of forecast quality with enormous benefits for society.

## NWP Forecast Skill

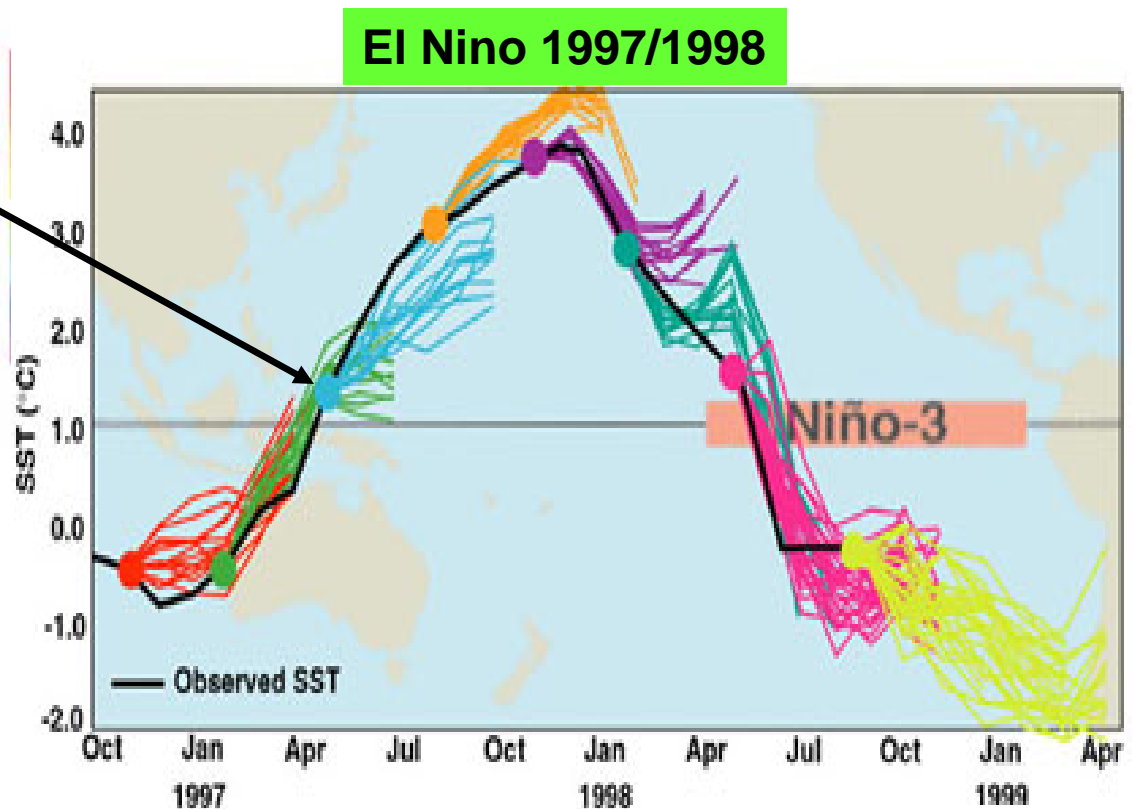


# Seasonal Prediction

Coupled models are now routinely used to make probabilistic prediction of the mean state of climate several months ahead. Enormous potential for EO!

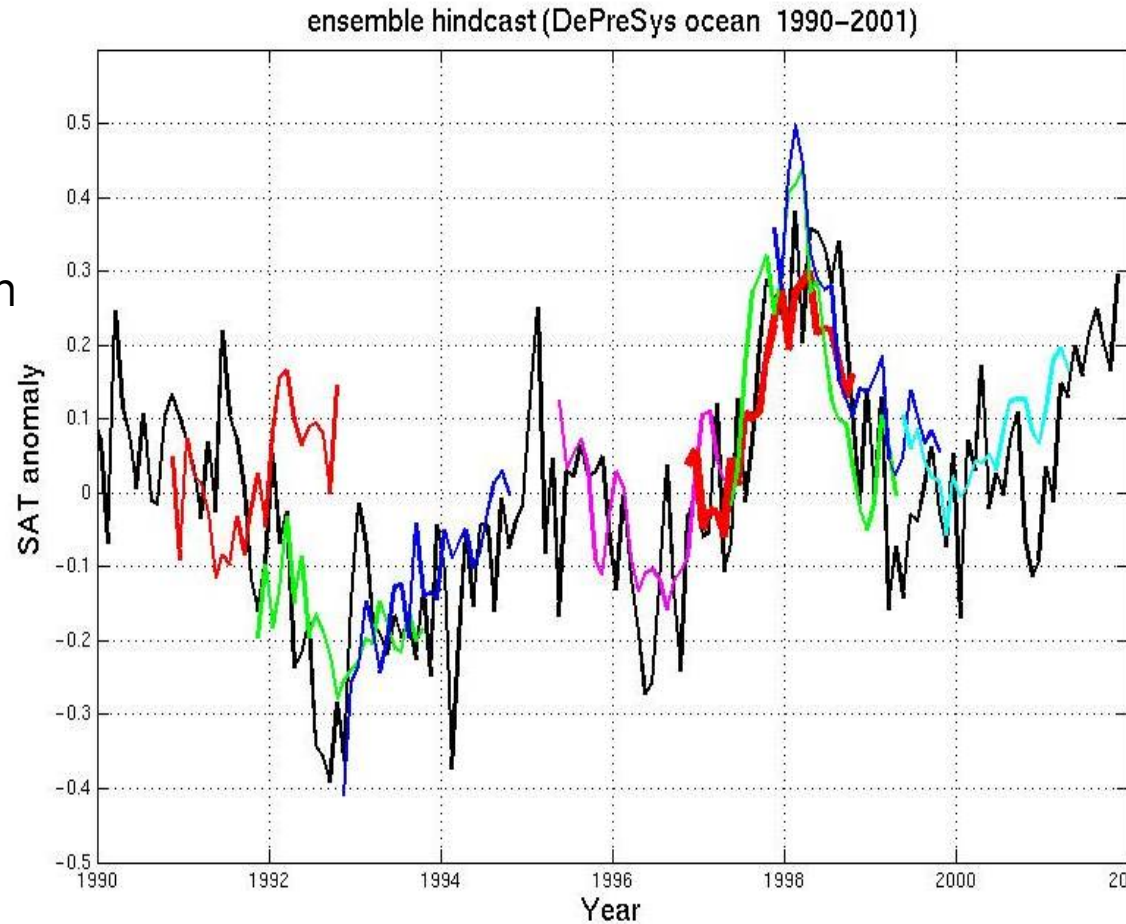


High-quality ocean analysis are vital to initialise coupled models and make accurate ensemble predictions.



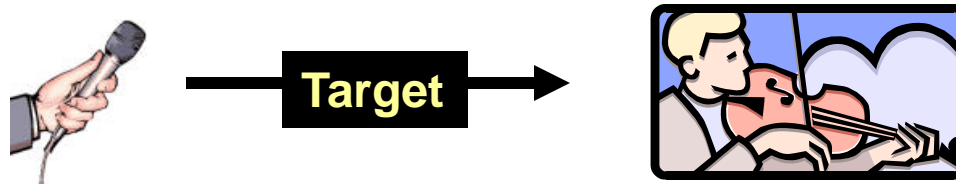
# Initialised Climate predictions

- Global Surface Air Temperature hindcasts from HadCM3 (following Smith et al 2007 Science)
  - Black line Obs.
  - Ensemble mean of Nov 2-year hindcasts



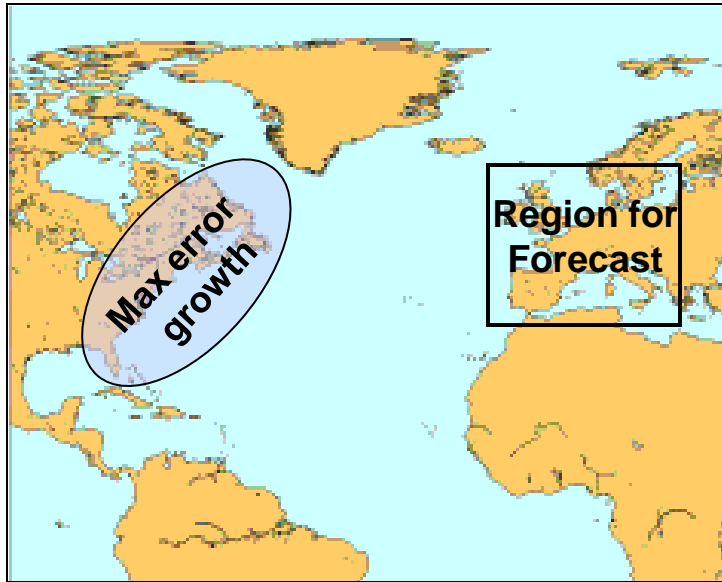
# Observing System Design

Observing systems should help to advance our current state of knowledge .....



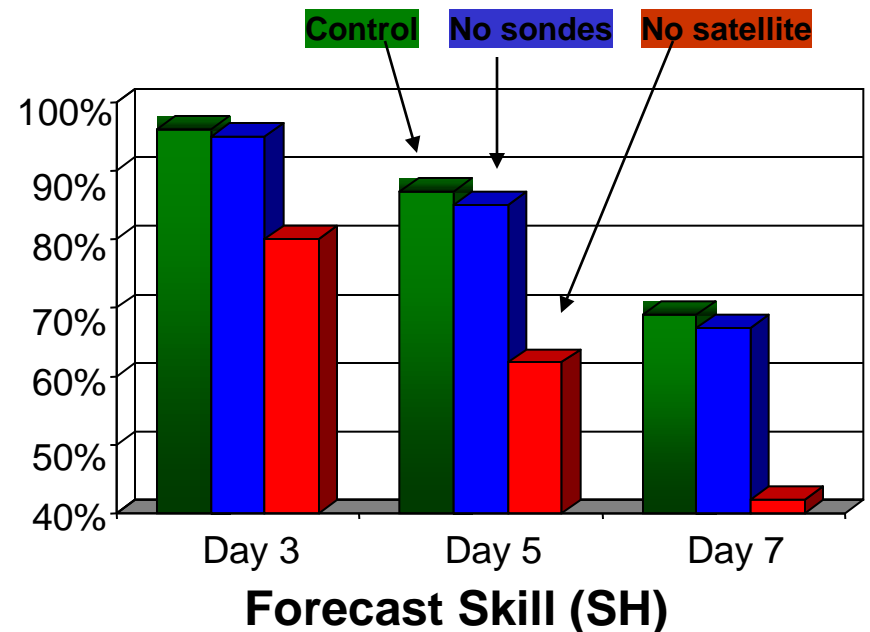
Model **sensitivity** experiments allow us to **target** observations and to **evaluate** objectively the incremental value of EO data. Potential cost savings!

# Numerical Laboratory



## *Where/What should we measure?*

Data Assimilation helps to identify sensitive regions where observations would maximise benefits for forecast.



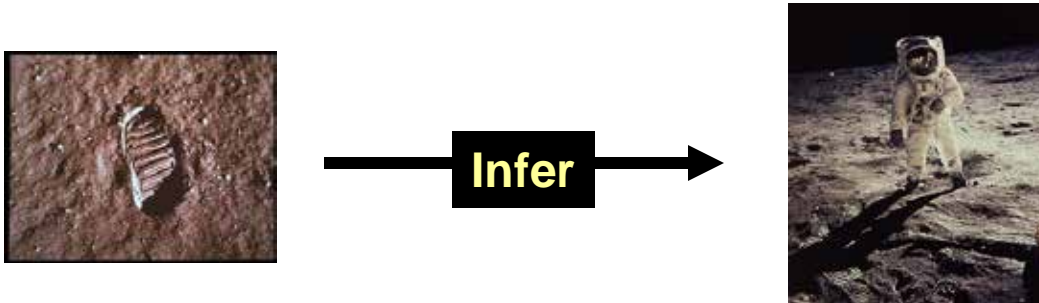
Courtesy ECMWF

## *What is the added value of EO?*

Observing System experiments help to quantify the impact of withdrawing various (synthetic) data streams on forecast skill (e.g. evaluation of Swift mission before launch!).

# Inverse Modelling

EO provides an indirect measure of the quantity of interest .....

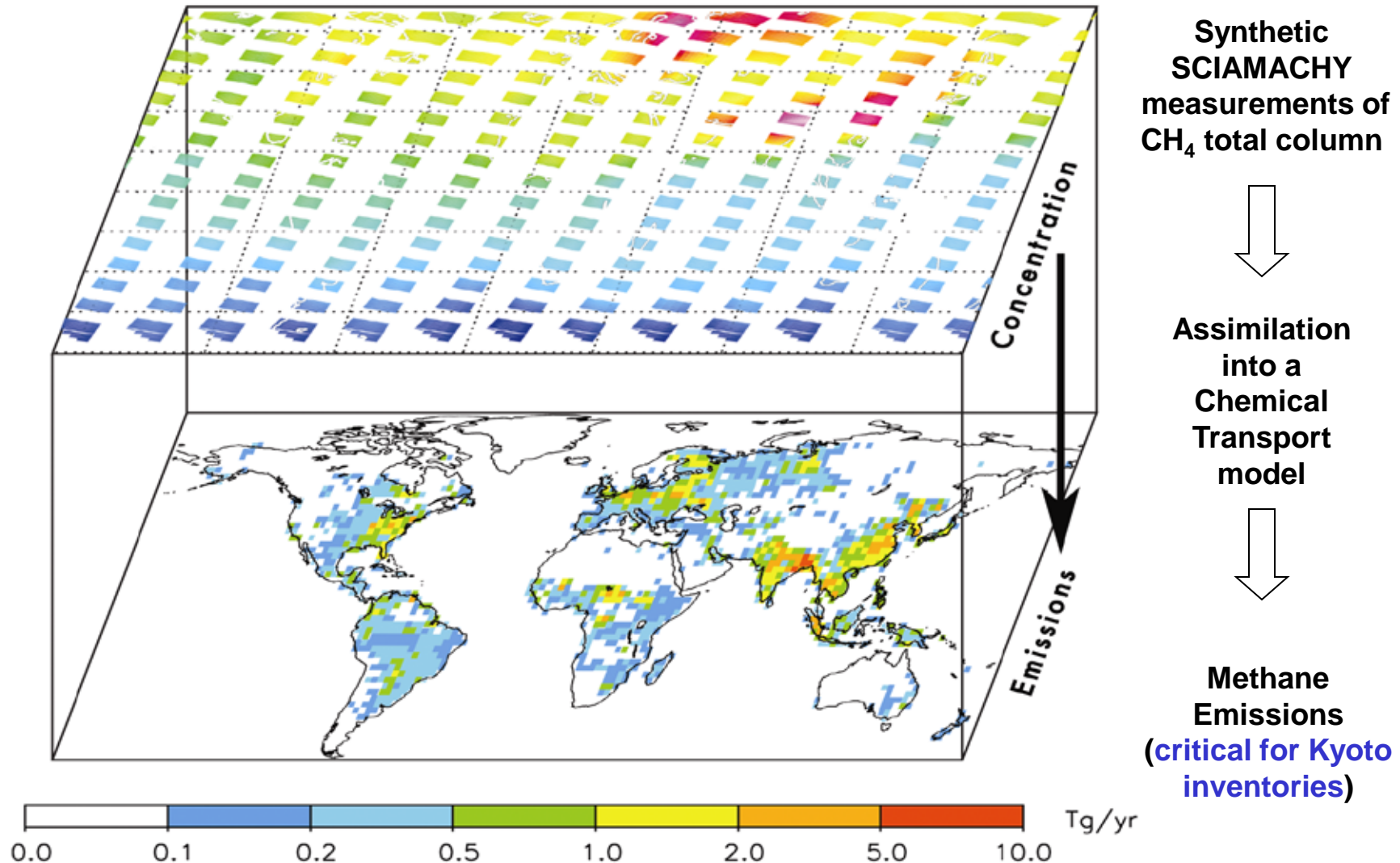


Assimilation of data into models enables to one to infer [**non-observable**] geophysical quantities of interest by exploiting physical/chemical **linkages** in the system



# GHG sources & sinks

Models play a diagnostic role by helping to interpret observations (e.g. causal relationships)



# Re-analysis

# Reanalysis

## The need for long atmospheric/oceanic data sets

- Researchers need long, consistent, global 4-d data sets for scientific studies.
- Most researchers do not have access to NWP systems to make data sets for specific needs.

## Why reanalysis?

- Over time, models, assimilation systems and available observations change.
- Use observation sets from history and assimilate with one state-of-the-art system.

## Reanalysis is good for

- Large-scale variability studies (e.g.)
  - El-Nino, La-Nina, MJO, NAO, monsoons, storm tracks.
- Studies of quantities well modelled and well observed.

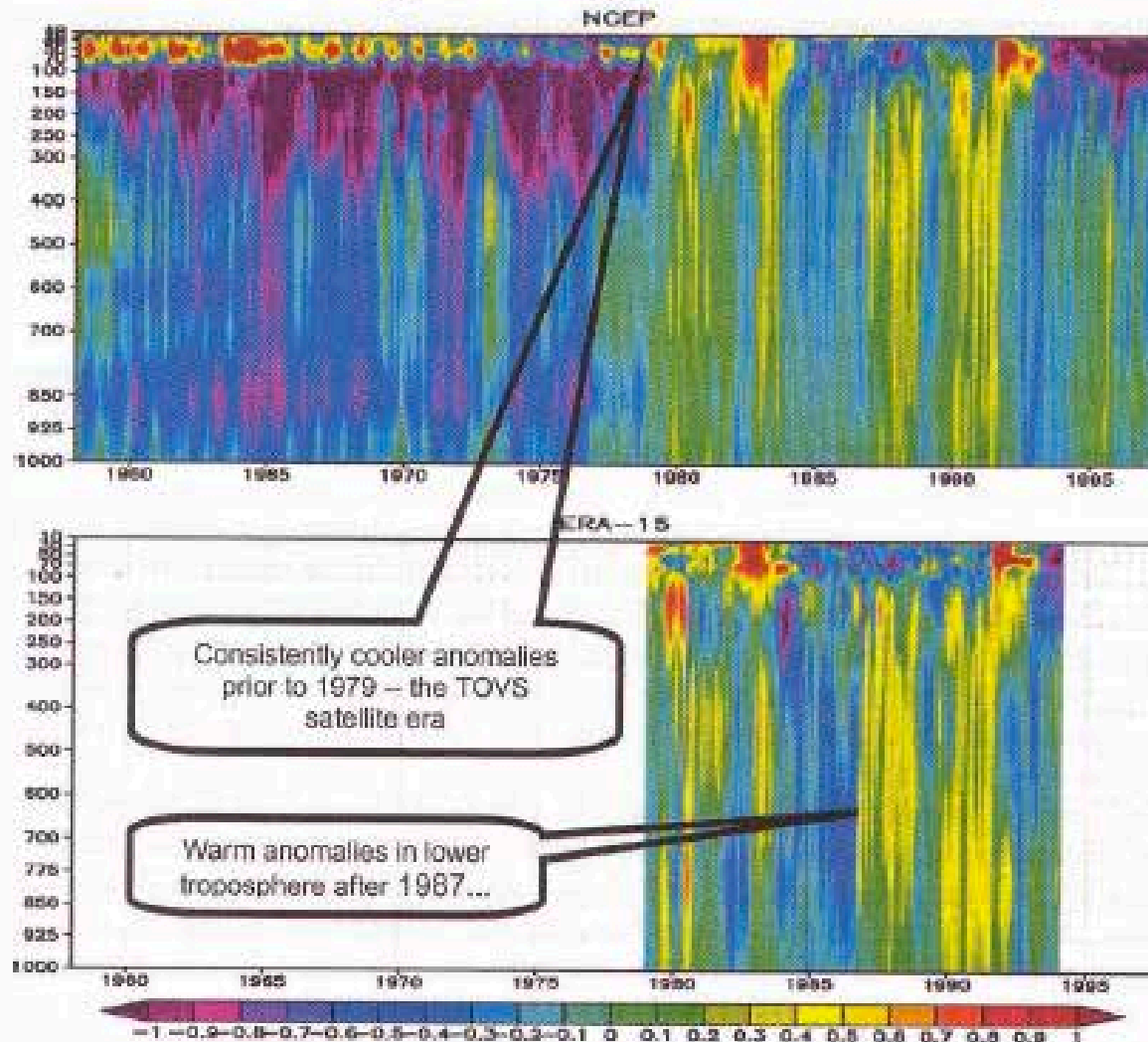
## Reanalysis is not good for

- Trend studies (due to changes in observing systems).
- Studies of poorly observed quantities (e.g.)
  - stratospheric water vapour.
- Studies of derived quantities, not constrained by observations (e.g.)
  - divergent wind, clouds, hydrological cycle, surface fluxes, vertical wind.

## Leading reanalysis data sets

- NCEP/DOE - atmosphere.
- ECMWF (ERA-15, ERA-40) - atmosphere.
- NASA/DAO - atmosphere.
- JMA (JRA-25) - atmosphere.
- GODAE - planned ocean reanalysis.

# Biases cause problems with reanalysis 1



*Courtesy, R.Rood (2003)*

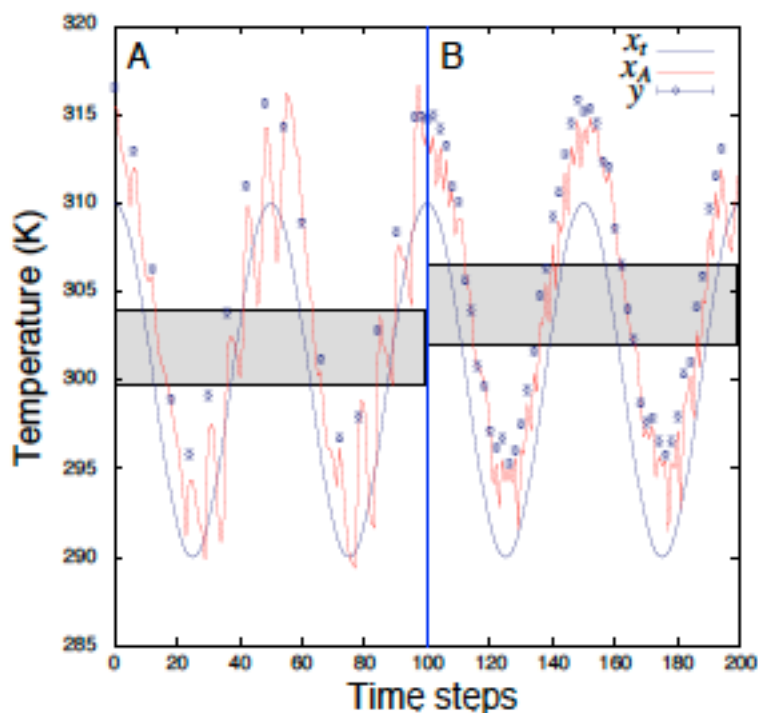
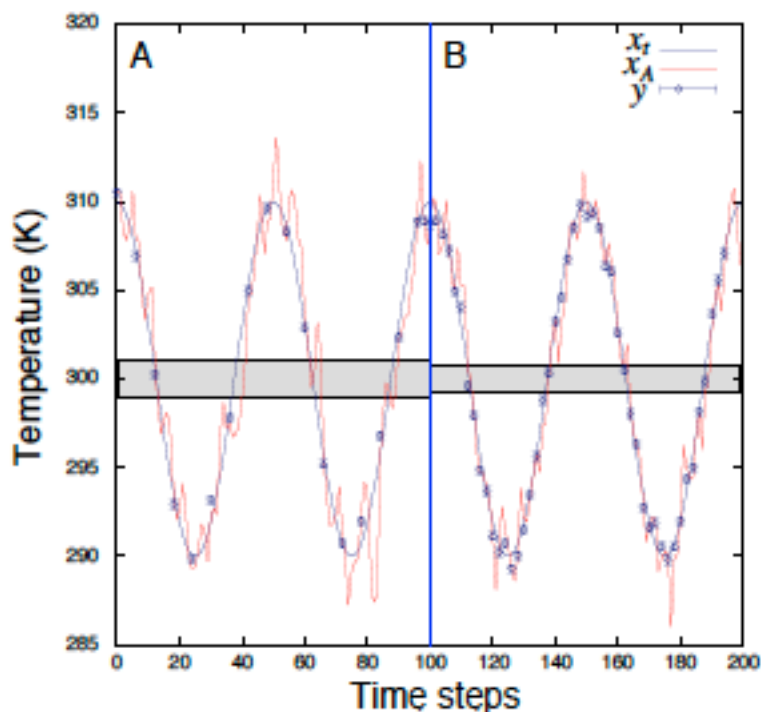
*Reanalysis is inappropriate for climate trend studies*

## Biases cause problems with reanalysis 2

Consider a jump in frequency of the assimilation of an observation type

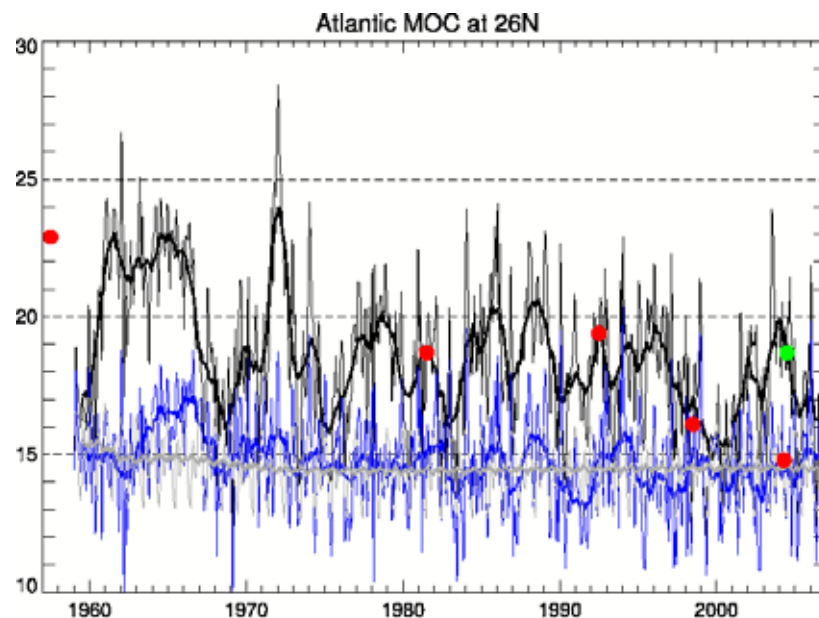
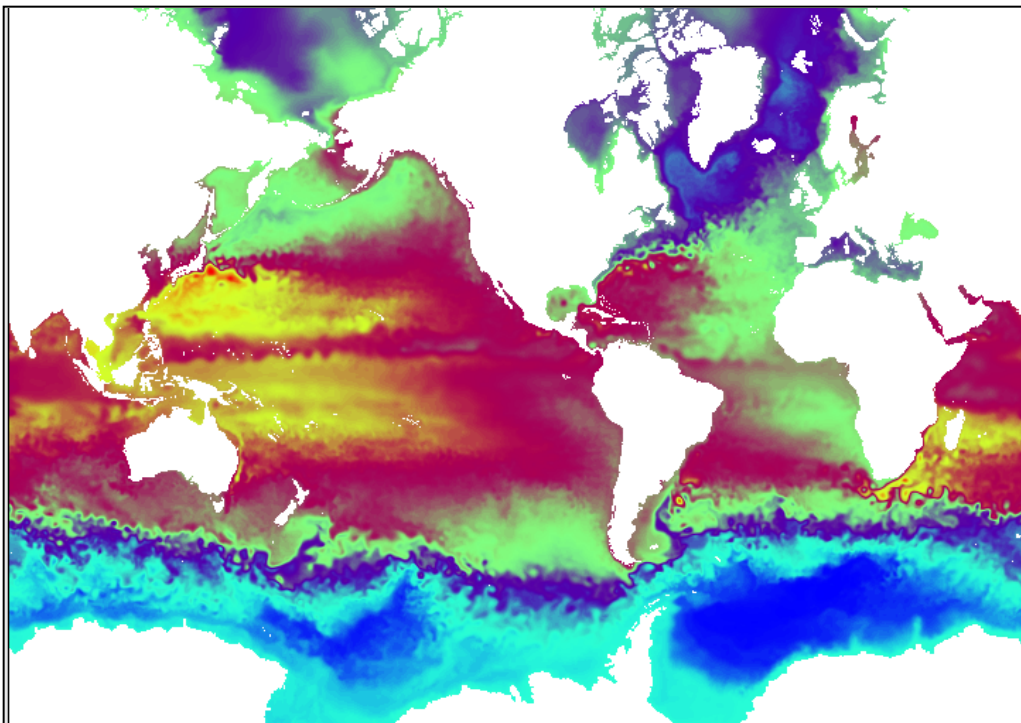
If the obs are unbiased, the mean error reduces.

If the obs are biased (6K), the mean error can increase.



*Biases (in observations or models) can lead to artefacts (e.g. apparent trends).*

# Uses of Ocean Reanalysis



Analysed sea level 26-31 Dec 2004

Global  $\frac{1}{4}$  NEMO 18 yr Synthesis with assimilation

Eg. Better Gulf Stream separation here aids Bryden section based annual

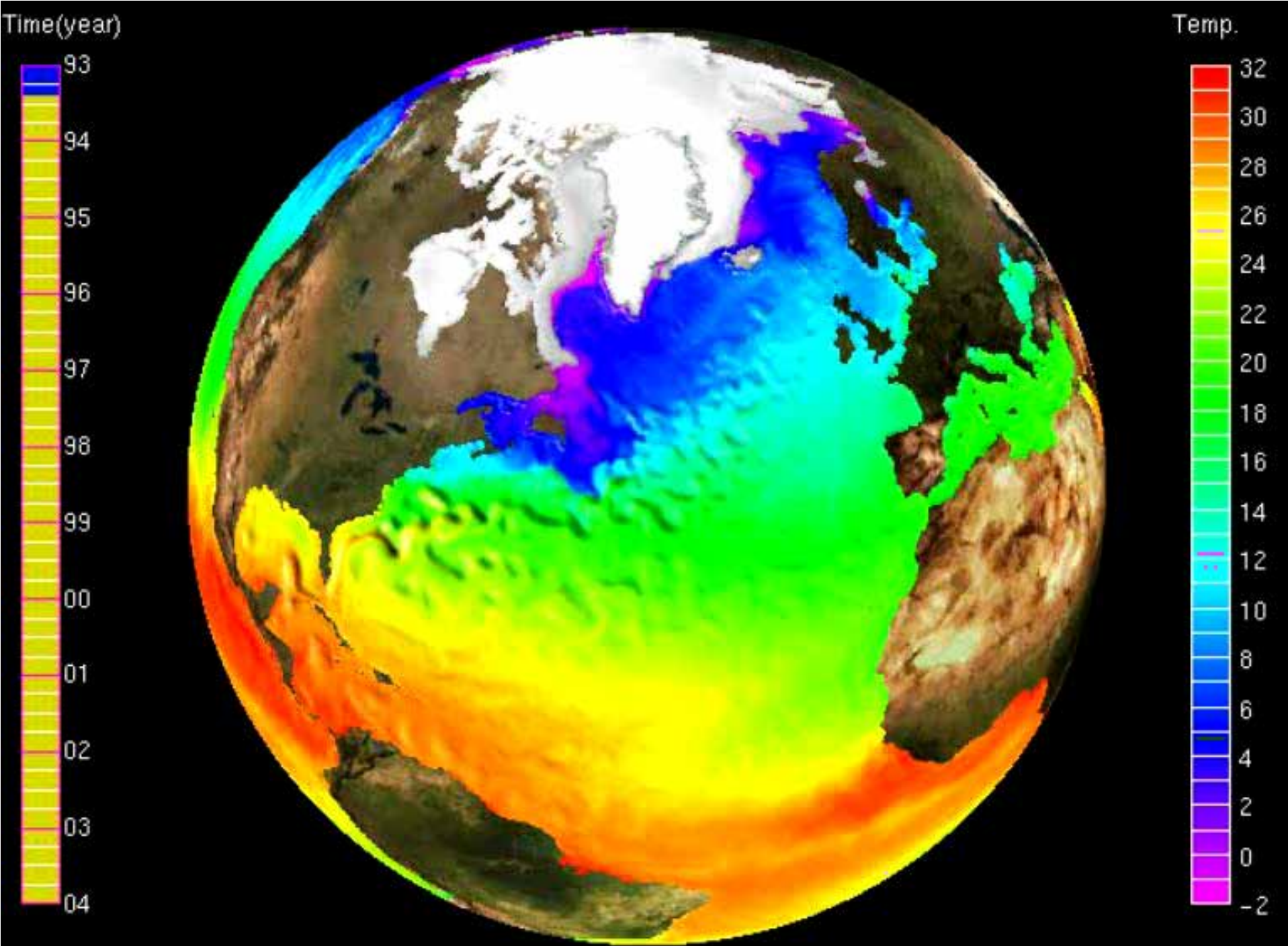
Altimeter and GOCE assimilation for NCEO) (Black line is assimilation)

Thermohaline MOC transports

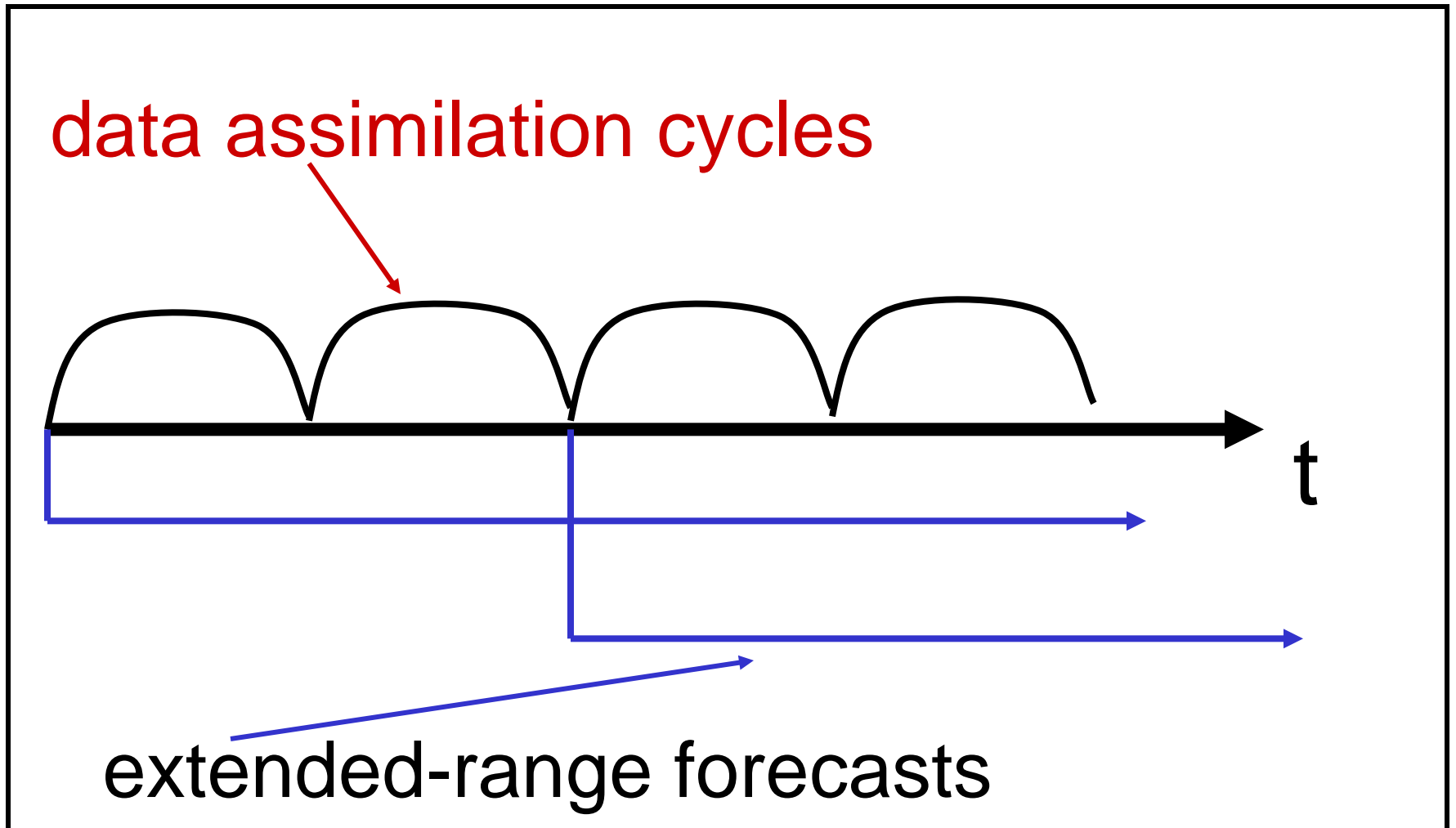
ECMWF reanalysis compared

Bryden section based annual

Balmaseda et al 2007



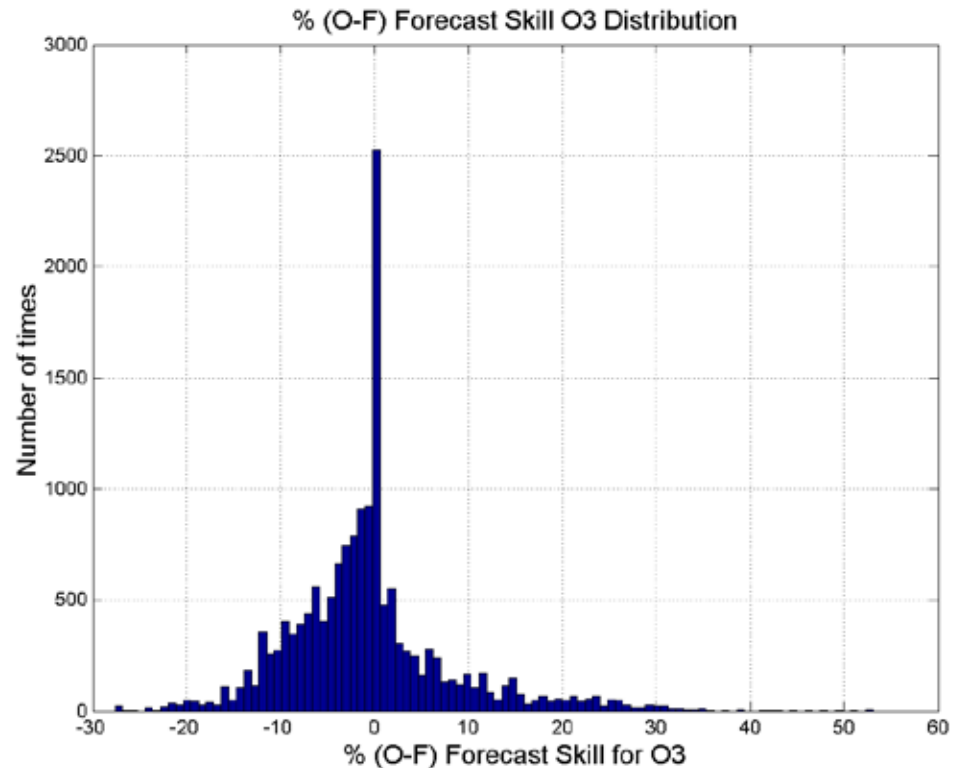
# Testing Earth System Models





# Skill Measures: Observation Increment, (O-F)

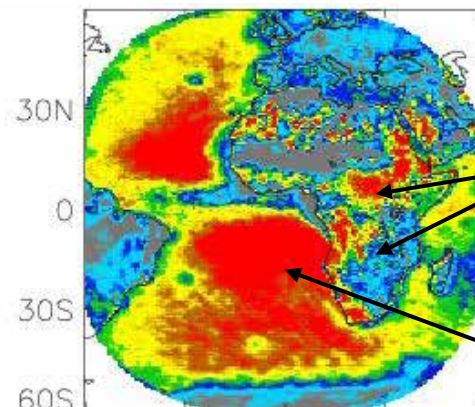
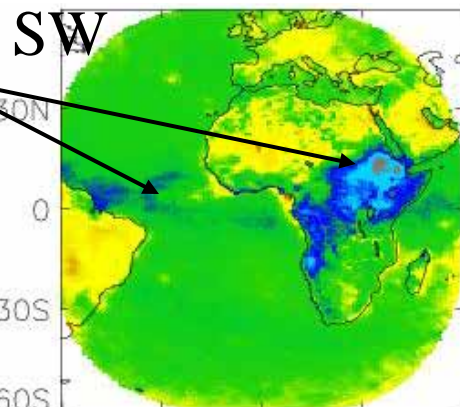
- The difference between the forecast from the first guess,  $F$ , and the observations,  $O$ , also known as observed-minus-background differences or the innovation vector.
- This is probably the best measure of forecast skill.



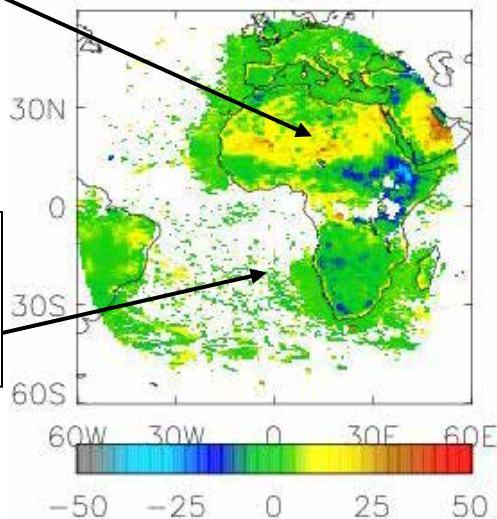


# Systematic biases in the UK's Unified Model

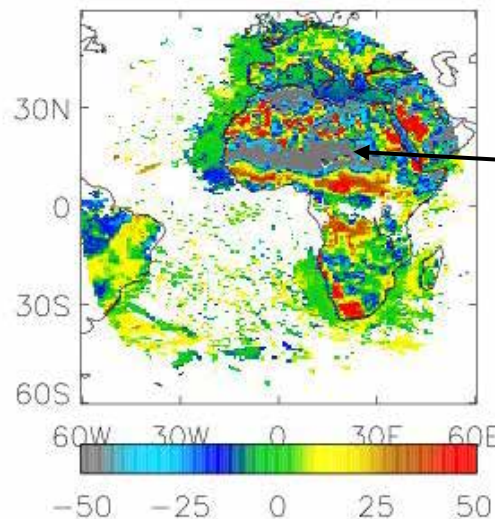
LW Model minus GERB 1200 UTC



Model-GERB OLRcs 12 UTC



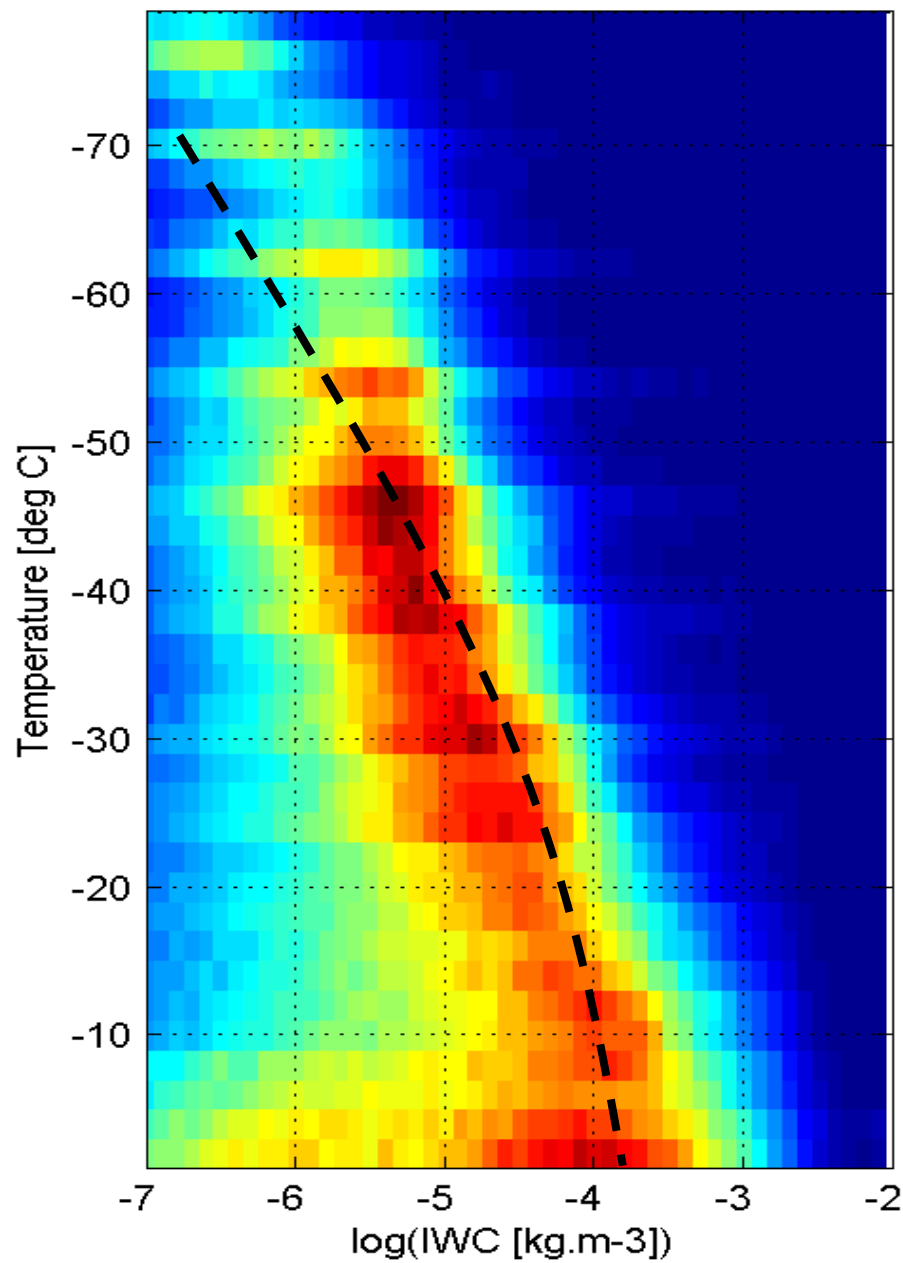
Model-GERB RSWcs 12 UTC



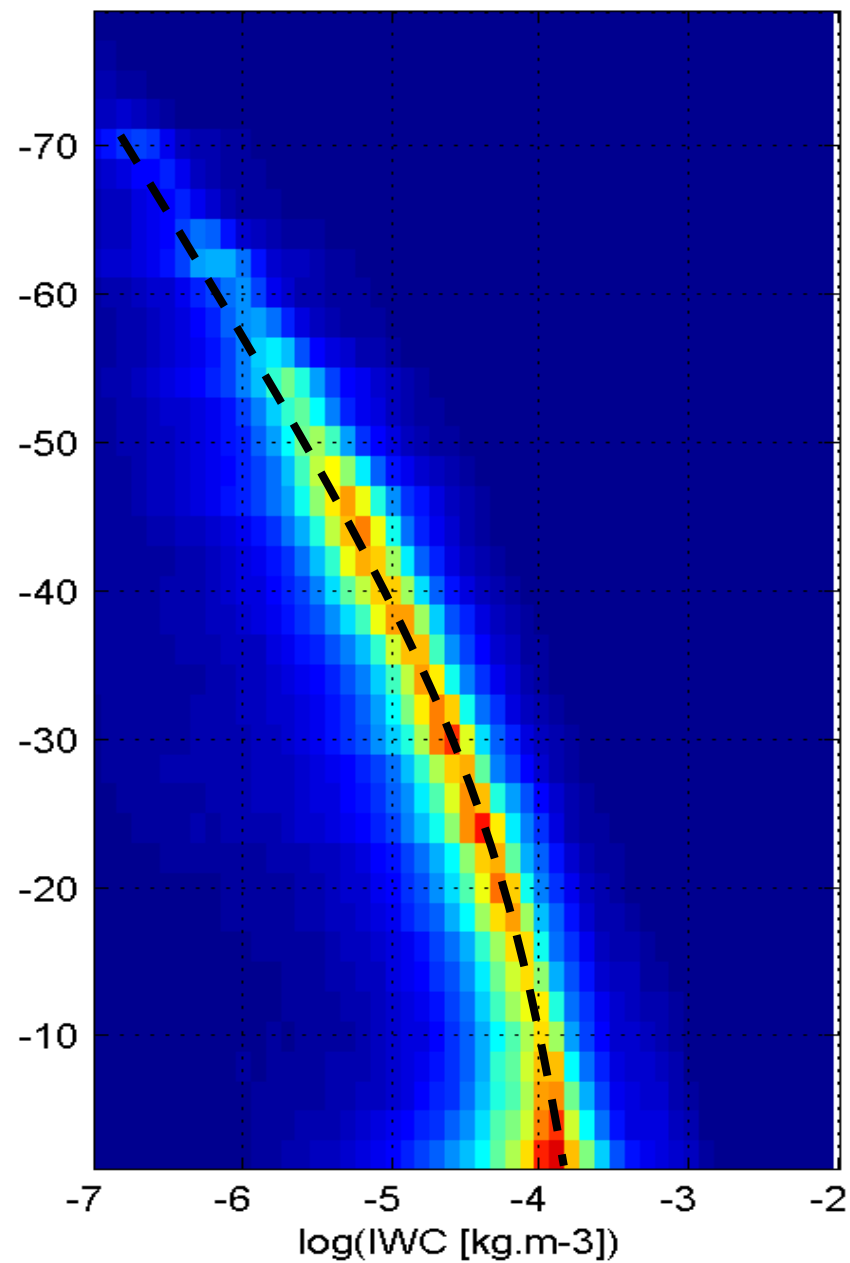
Top: all-sky differences

Bottom: clear-sky differences

Cloudsat-CALIPSO



Global UK met-office model



Hogan &

# Conclusion

