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WCRP and ICTP Interpreting Climate Change Simulations: Capacity Building for Developing Nations Seminar

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Uncertainty in climate change prediction.

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Lecture outline

- Some basic concepts
- Sources of uncertainty in climate change prediction
- The concept of climate change prediction
- Assessing uncertainty and reliability in climate change prediction
- Most material from: Giorgi, Climatic Change, 2005

Coupled system processes

We live in a non-linearly coupled Climate System characterized by a wide range of spatial and temporal scales



Human factors





Natural factors



The earth's climate can change because of many human and natural factors Incoming solar radiation

> Absorbed by greenh

Variations of Solar radiation





Temporal scales of forcings



The concept of climate prediction (Lorenz 1975)

- Predictability of the first kind
 - Initial value problem: Prediction of the evolution of the climate system given some knowledge of the initial state
- Predictability of the second kind
 - Boundary value problem: Prediction of the response of the climate system to external forcings
- The predictability range depends on the time scale of the phenomenon of interest

Anthropogenic Climate Change Prediction

- Time scale of interest is decades to centuries
- Fundamentally, climate change prediction is a predictability problem of the second kind
 - Response of climate statistics (mean, variability, extremes) to changing anthropogenic forcings
- However, a first kind predictability component arises because of the long time scales involved in ocean, cryosphere and biosphere processes
 - The transient evolution of climate statistics may depend on the initial state of the ocean and biosphere

What is a transient Climate Change Simulation?



INTRINSIC SOURCES OF UNCERTAINTY IN CLIMATE CHANGE PREDICTION

Intrinsic sources of uncertainty in climate change prediction: Unpredictability of natural and anthropogenic forcings





It is impossible to predict major volcanic eruptions or social/technological developments (scenarios instead of predictions)



Intrinsic sources of uncertainty in Climate Change Prediction: Thresholds and feedbacks

- Feedbacks within the climate system can enhance its non-linearity and thus decrease predictability
 - Snow and sea-ice albedo feedback
 - Biogeochemical / hydrologic feedbacks
 - Adaptation / mitigation feedbacks
- Threshold behaviors also enhance nonlinearity and decrease predictability
 - Shut down of the Thermohaline Circulation
 - Melting of Greenland and Antarctic Ice Sheets

Examples of threshold or feedback behavior



Shut down of the deep oceanic circulation





Melting of Greenland and the West Antarctic ice sheet

Intrinsic sources of uncertainty in Climate Change Prediction: Regimes

- There is evidence of quasi-stationary regimes within the climate system
- External forcings can modify the frequency of occurrence (PDF) or the structure of different regimes
 - Corti et al. (1999)
 - Laird et al. (1996)

PDF of 500 Hpa Height (Corti et al. 1999)

1949 / 94

1949 / 71



1949 / 94 No EN-LN

1971/94

Intrinsic sources of uncertainty in Climate Change Prediction: Natural variability



The uncertainty of Climate Change

Because of the internal variability and non-linearity of the climate system and the random component of the external natural and anthropogenic forcings, the climate change of interest is only one realization within a range of possible realizations, each characterized by a certain likelihood to occur.

<u>Climate change is characterized by an</u> <u>intrinsic (unavoidable) uncertainty which</u> <u>needs to be fully explored</u>

"Actual" Climate Change PDF



ADDED SOURCES OF UNCERTAINTY IN CLIMATE CHANGE PREDICTION

Added sources of uncertainty in Climate Change Prediction: Initial Conditions

Climate can evolve differently depending on the initial conditions of its slow components (which we do not know with good accuracy)



Added sources of uncertainty in Climate Change Prediction: Climate Models

Climate models are imperfect representations of the climate system and different models respond differently to the same climate forcing agents.



The Climate Change Prediction Problem

The purpose of climate prediction is not to predict what will be the exact climate of the future, but to reconstruct as closely as possible the PDF of possible future climates, i.e. climate change prediction needs to be approached in a probabilistic way. The use of imperfect models and observations adds an element of uncertainty that needs to be reduced to the extent possible.

Predicted vs. "Actual" Climate Change PDF



Assessing uncertainty and reliability in climate change prediction

Assessing Uncertainty and Reliability

- We do not have specific case studies to test our anthropogenic climate change predictions, e.g. as in weather and seasonal prediction, and as a result it is critical to:
 - Understand and possibly quantify the uncertainties underlying our predictions
 - Evaluate and possibly quantify the reliability of our predictions









IPCC Emission and Concentration Scenarios

CO2 Emissions



CO2 Concentrations





Climate Simulation Segment of the Uncertainty Cascade



Model configuration uncertainty Global scale



IPCC – 2007: Global temperature change projections for the 21st century



Model configuration and scenario uncertainty contribute approximately equally to the global warming projection uncertainty

Internal system variability: Global scale



Model configuration uncertainty Regional scale



Regional Temperature Change A2, DJF, 9 AOGCMs



Regional Precipitation Change A2, DJF, 9 AOGCMs



Uncertainties in regional climate change projections: The PRUDENCE strategy



Sources of uncertainty in the simulation of temperature and precipitation change (2071-2100 minus 1961-1990) by the ensemble of PRUDENCE simulations (whole Europe) (Note: the scenario range is about half of the full IPCC range, the GCM range does not cover the full IPCC range) (Adapted from Deque et al. 2006)



Assessing reliability in climate change prediction

- Model performance in reproducing observed features of the historical climate
- Inter-model agreement in the simulation of climate change features
- Model performance in reproducing reconstructed features of past climates
- Physical soundness of the processes that lead to the simulated changes

Temperature Bias, 9 AOGCMs



Precipitation Bias, 9 AOGCMs



Reliability Criteria

- Model performance in reproducing observed features of the historical climate
- Inter-model agreement in the simulation of climate change features
- Model performance in reproducing reconstructed features of past climates
- Physical soundness of the processes that lead to the simulated changes

Inter-model agreement in the simulation of temperature change



Inter-model agreement in the simulation of precipitation change



Accounting for model performance and convergence criteria : The REA Method (Giorgi and Mearns 2002,2003)

$$\widetilde{\Delta T} = \widetilde{A}(\Delta T) = \frac{\sum_{i} R_{i} \Delta T_{i}}{\sum_{i} R_{i}},$$

$$\widetilde{\delta}_{\Delta T} = [\widetilde{A}(\Delta T_i - \widetilde{\Delta T})^2]^{1/2} = \left[\frac{\sum_i R_i (\Delta T_i - \widetilde{\Delta T})^2}{\sum_i R_i}\right]^{1/2}$$

- R_i = reliability for model i
- $R_B =$ bias reliability
- R_D = convergence reliability
- E = measure of natural
 variability

Ensemble model reliability

$$\tilde{\rho} = \tilde{A}(R) = \frac{\sum_{i} R_{i}^{2}}{\sum_{i} R_{i}}$$

$$R_{i} = \left[(R_{B,i})^{m} \times (R_{D,i})^{n} \right]^{[1/(m \times n)]}$$
$$= \left\{ \left[\frac{\epsilon_{T}}{abs(B_{T,i})} \right]^{m} \left[\frac{\epsilon_{T}}{abs(D_{T,i})} \right]^{n} \right\}^{[1/(m \times n)]}$$

Reliability Criteria

- Model performance in reproducing observed features of the historical climate
- Inter-model agreement in the simulation of climate change features
- Model performance in reproducing reconstructed features of past climates
- Physical soundness of the processes that lead to the simulated changes
 - Process understanding

Predicting climate change

- There is substantial uncertainty across models in the simulation of climate change, therefore a climate change prediction needs to be based on the compounded information from different models and, where available, different methods
- Different approaches are being developed to produce probabilistic climate change information from ensembles of model simulations

Model range with flat distribution (IPCC 2001)



Large ensembles using simplified climate models (Wigley and Raper 2001; Webster et al 2004)



Use of scaling factors to increase the ensemble size (Stott and Kettleborough 2002; Murphy et al. 2004)



Information from ensembles of AOGCM runs

(Raisanen and Palmer 2001; REA method, Giorgi and Mearns 2002,03; Bayesian approach, Tebaldi et al. 2004)



Conclusions

- Climate change prediction is characterized by an intrinsic uncertainty and by an added uncertainty due to the prediction process
- Because of its nature, climate prediction needs to be approached in a probabilistic way
- Large ensembles of simulations are needed in order to fill the phase space of possible future climates (and climate change paths) and to produce meaningful PDFs
 - Different initial conditions, model configurations, scenarios, feedback processes, regionalization techniques
 - Especially important at the regional scale and for changes of hydrologic variables
- A clear understanding of uncertainties and processes is critical
 - Understanding and reducing the climate sensitivity range



A climate change prediction strategy

Large Ensemble of Intermediate Resolution (1-2 Deg.) AOGCM Simulations (Models, Scenarios, IC, Feedbacks)

> Clustering of "Characteristic" Climate Change Simulations

Regionalization of Characteristic Climate Change Simulations (RCM, SD, etc.)