Fast Processes in Large Scale Models: Their Parameterization and Evaluation

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What is Life? with Mind and Matter and Autobiographical Sketches

ERWIN SCHRÖDINGER

"I can see no other escape from this dilemma than that some of us should venture to embark on a synthesis of facts and theories, albeit with second-hand and incomplete knowledge of some of them – *at the risk of making fools of ourselves.*"

I venture to take the risk of making fools of myself ...



Climate Model, Fast Physics and Parameterization



Virtually Unchanged Large Uncertainty of Model Climate Sensitivity through Ages

Estimates of central value and uncertainty range from major national and international assessments



The wide spread of model climate sensitivity has been attributed to parameterization of cloud-related fast processes.

The pace of progress has been frustratingly slow.

"...The modeling of clouds is one of the weakest links in the general circulation modeling efforts."

--Charney et al., National Academy Report, 1979



Deficiencies in the representation of cloud processes in climate models drive much of the uncertainty surrounding predictions of climate change.

This was true 30 years ago, it's true now, and at the rate we are going it will still be true 30 years from now. (D. Randall, 2008)

Why is progress so slow ? How can we accelerate it ?

4M-1E Complexity and Components Involved



- 4M scientific
 - -- Multibody
 - -- Multiscale
 - -- Multitype
 - -- Multi-dimension
- 1E engineering -- coordination among both investigators and components





Aerosol

Droplet

Turbulent Eddies

Convection Clusters

Global

Major Parameterized Fast Processes in Climate Models



Major Parameterized Fast Processes in Climate Models





Model/Data Multiscale Hierarchy



Aerosol Droplet Turbulent Eddies S. Cu Clusters Global

DNS = Direct Numerical Simulation *LES* = *Large Eddy* Simulation CRM = Cloud-**Resolving Model** WRF = Weather **Research** and **Forecast Model** GCM = Global**Climate Model** RCM = RegionalClimate Model

GCRM = Global CRM

NWP = Numerical Weather Forecasting

SCM = Single Column Model

Lack of adequate observations is another reason for the slow progress.



Cloudnet Sites
ARM/NOAA Sites
ARM Mobile Facility

Surface-Based Cloud-Profiling Measurements Complementary to Satellites and In-Situ



Critical Experiment and Evaluation

•<u>Critical experiments</u> test competitive theories/models and help determine the correct one.

• Complexity and nature of GCM fast physics calls for critical evaluation.

• **Critical evaluation** calls for FASTER project to *focus and constantly* deal with multiscale modeling, observations, and process hierarchies, with 6M strategy.

6M Strategy: multi-objectives, m-tasks, m-approaches, m-tools, m-disciplines, and m-institutions





What is FASTER?

- Represent FAst-physics System TEstbed and Research
 - Treat fast physics as a system
 - Combine testbed and research
 - Evolve with GCMs
 - Aim at faster "realtime" evaluation of fast physics
- FASTER is a DOE effort to *bridge ESM and ASR* sciences by utilizing *ARM* measurements to accelerate/improve evaluation and parameterization of cloud-related fast processes in climate models.





FASTER Project Pyramid



Continuous Evaluation of Fast Processes in Climate Models Using ARM Measurements



Facility/Model Development



Web-Based Fast-Physics Testbed







NWP-Testbed



The NWP-testbed is built on the EU Cloudnet project and integrates with the other FASTER modeling components, esp SCM

High-Resolution Modeling Activities



WRF-FASTER: A New CRM/LES

WRF reconfigured as a CRM/LES model; validated with wellknown models and well-tested cases, FASTER warm-up cases, and CGIL; shown here is cumulus clouds on 21 June 1997 at ARM SGP.



Novie time: 12 -13 local time Resolution: $\Delta x = \Delta y = 100$ m; $\Delta z = 40$ m; $\Delta t = 0.1$ S

(Endo et al., 2012: Mon Wea Rev)



Multiscale Data Assimilation System



GSI = Grid Space Interpolation, NCEP-3DVAR scheme Multi-Scale GSI = GSI + JPL Multiscale DA System

Li et al, 2011: Experiments with a multi-scale data assimilation system. Mon Wea Rev

Data Integration

* Infuse available data and tailor to FASTER needs

- * Generate new data as needed:
- -- CAPE (Convective Available Potential Energy)
- -- CIN (Convective Inhibition)
- -- Stratiform-Convective Partition
- -- Aerosol Properties
- * Model Data
- * 4D visualization

Multiscale and Multi-Sources



Beyond Fried Egg: Multiscale VES







- Other models e.g., NWP
- Evaluation metrics/features

• 3D

- Long-term & real time
- Collaboration with computer scientists at BNL and SBU
- Growth into a user facility?

Evaluation Package, Strategy and Uniqueness

- •Long-term statistical , regime, and case study
- Emphasis on coupling and relationship
- Consideration of observational spread
- Metric development
- Model and process ranking/weakest link
- Comprehensive: GCM/SCM, NWP and HRM







Science Snapshots

Better Use of ARM Radiation Measurements and New Method for Inferring Cloud Albedo

• Analytic relationship between SRCF, cloud fraction and cloud albedo: SRCF = fraction x albedo

• New approach to infer cloud albedo from surface-based radiation measurement;

• Application to other ARM sites;

• PI product to release to the ARM/ASR community

Hourly data from 1997 to 2009 at ARM SGP



Reference: Liu et al., 2011: Relationship between cloud radiative forcing, cloud fraction and cloud albedo, and new surface-based approach for determining cloud albedo. Atmos. Chem. Phys., 11, 7155-7170.

Evaluation of NWP Reanalysis & Analysis

- Widespread use
- Use in ARM largescale forcing and merged sounding VAP
- Cloud properties not assimilated; errors from model physics;

•Three major reanalysis products evaluated;



 Model negative biases in cloud properties.
Coupling between cloud properties and near surface meteorology, especially relative humidity likely via PBL processes (not shown)

Ref: Wu et al, 2012: Observational-based evaluation of major NWP reanalyses against ARM measurements over the SGP site. J. Geophys. Res (revised)



Evaluation of NWP Forecasts

- SGP 2001 to 2010
- Darwin 2005 to 2009
- SGP2004 example
- Qualitative agreement
- Underestimated clouds, esp., low-level clouds
- Consistent with reanalysis



Barrett et al 2009: Geophys. Res. Lett, 36, L17811.

SCM-NWP Integration

2004-2008 Annual Mean Cloud Fraction over SGP



Relation between Cloud Albedo and Fraction

• 12 year monthly data (1997 -2009) at the ARM SGP site

• Two sets of measurements: surface-based Solar Infrared Radiation System (SIRS) and GOES satellite

• Joint occurrence of cloud fraction and cloud albedo

• Conspicuous positive correlation between cloud albedo and cloud fraction for both GOES and SIRS



Do GCMs simulate the relationship, and how well ?

Comparison between GCMs and Observations

 General positive correlation

 Large inter-model spread

Difference between
TOA and surface based
cloud albedo





SCM Investigation

✓ Different physics: SCAM3, SCAM4, and SCAM5;

✓ SCAMs produce general positive correlation, with SCAM5 best simulating the observations;





SCM Investigation

✓ Different physics: SCAM3, SCAM4, and SCAM5;

✓ SCAMs produce general positive correlation, with SCAM5 best simulating the observations;

 ✓ Different cloud overlaps: standard (solid) and random (dashed);

✓ Differences can arise from model physics and/or cloud overlap assumptions.



The relationship between cloud albedo and cloud fraction, along with the fact that model physics and/or overlap assumption can lead to comparable results, calls for consideration of coupling and consistency issues in development of fast physics parameterizations!

Tuning and Compensating Errors — Analysis

• Tuning to the same TOA energy budget leads to an inverse relationship between cloud fraction f and cloud albedo α :

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\alpha f ~ constant.
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• A more accurate expression is

$$\frac{\Delta \alpha}{\alpha} = -\frac{\Delta f}{f}$$

(Liu et al., ACP, 11, 7155-7170, 2011)

Tuning and Compensating Errors — Evidence

19 IPCC AR4 GCM Results



These results demonstrate that "tuning" parameterizations to observations lead to serious compensating errors, even distinct cloud regimes; we should derive parameterizations from first principles and reduce the number of tunable parameters as much as possible, and meantime look for smart objective "tuning" !!

Compensating Errors in Precipitation

 $\mathbf{P} = \overline{\mathbf{p}} \mathbf{N}$









Crucial message: large scale forcing controls SCM total precipitation more, however, from different compensating errors in different GCMs >> convection trigger vs strength?

Offline Evaluation of Surface Flux Parameterization Using 7-Year ARM



(Liu et al, Mon Wea Rev, 2012)

Three Levels of Parameterization

Fast Processes



Mean-field parameterization

Resolved slaves subgrid

Stochastic parameterization

Subgrid affects resolved

Unified parameterization

Interacting subgrid processes

(self-consistency issues)

Resolved Grid Variables

Parameterization is not just practical necessity, but deep theoretical underpinning of scale-interactions within the multiscale system.

Systems theories for representing all aerosol/cloud/precipitation processes



Three-moment scheme for aerosol-cloud-precipitation continuum Liu et al., (2002, 2004, 2006, 2007, 2009) on clouds and precipitation.



A journey of thousand miles starts with a single step

Suggestions & Collaborations ?



Thanks so much !

Evaluation of SCMs with ARM Measurements

- 4 major SCMs;
- Different versions;
- 3 year (1999 -2001) runs with hourly resolution
- Model performance as a function of surface temperature
- Deficient parameterizations related to convection
- More results next



Ref: Song et al, 2012: Observational-based evaluation of major NWP reanalyses against ARM measurementsP site. J. Geophys. Res (to be submitted)



Entrainment-Mixing Processes & Microphysics



Homogeneous Mixing Fraction



Further parameterization of the scale number leads to a much needed parameterization for homogeneous mixing fraction.

Lu et al 2011: Examination of turbulent entrainment-mixing mechanisms using a combined approach. J. Geophys. Res.; 2012: Relationship between homogeneous mixing fraction and transition scale number, Environ. Res. Lett.

Entrainment-Mixing Processes: New Approach for Estimating Entrainment Rate

- Elimination of need for in-cloud measurements of temperature and water vapor
- Smaller uncertainty
- Potential for a parameterization that directly links microphysical effects of entrainment mixing with entrainment rate
- Potential for a remote sensing technique to measure entrainment rate for ARM



Lu et al 2012: A new approach for estimating entrainment rate. Geophys. Res. Lett.

New Metric - Relative Euclidean Distance

• Traditional statistical measures: bias, standard deviation difference, correlation coefficient r

• A new metric that (1) measures overall performance and (2) can be used to compare not just different models but different quantities: Relative Euclidean Distance:

$$\mathbf{D} = \sqrt{\left[\frac{\left(\overline{\mathbf{M}} - \overline{\mathbf{O}}\right)}{\overline{\mathbf{O}}}\right]^2} + \left[\frac{\left(\sigma_{\mathbf{M}} - \sigma_{\mathbf{O}}\right)}{\sigma_{\mathbf{O}}}\right]^2 + (1 - r)^2$$



Decrease of D from cloud fraction to precipitation is consistent with that SCM precipitation is better constrained by large scale forcing.

Evaluation of Surface Flux Schemes

- 5 major schemes in GCM and WRF
- 7 year SGP measurements
- Model performance in terms of relation between daily mean and standard deviation
- Parameterizations can be wrong in mean, standard deviation, or both



Daily Mean

Liu et al 2012: Evaluation of surface flux parameterizations using long-term ARM measurements. Mon. Wea. Rev. (Revised)

Evaluation of Aerosol Cloud Interactions

Overview:

- Observations from ARM IOPs are being utilized to evaluate the interactions between clouds and aerosols in the NASA GISS ModelE.
- Parameterizations of droplet activation, droplet effective radius, and relationships between surface aerosol and cloud properties are tested.

Highlights:

- Simulated droplet activation generally follows observations.
- Effective radius parameterizations result in significantly different values the impact of these differences on climate are currently being evaluated.





Reference:

de Boer, G., S. Menon, S.E. Bauer, T. Toto, A. Vogelmann and M. Cribb (2012): Evaluation of aerosol-cloud interactions in the GISS ModelE using ARM Observations, *Atmos. Phys. Chem.*, in preparation

Evaluation of SCMs with ARM Measurements

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Convective/Stratiform Rain Partitioning at SGP



Three Definitions of Homogeneous Mixing Fraction --- Ψ_1





Dependence of Homogeneous Mixing Fraction on Transition Scale Number



Validation with LES Results

A benchmark case over the SGP site simulated by an LES model, WRF-FASTER (Endo et al., 2011)



Aerosol indirect effects constitute the major uncertainty in climate forcing!



Cloud Properties vs 2-m RH (monthly)



Obs ERA-Interim NCEP/NCAR NCEP/DOE

The cloud properties strongly link to the relative humidity (RH):

>Obs/ERA-Interim: strongest, with correlation [0.62, 0.80]

R2: slightly stronger than R1 on the link between cloud fraction (or SRCF) and the RH

R1/R2: relatively weak on the link between cloud albedo and the RH

!!! Strong link between the cloud properties and RH **!!!**

Standard Deviation vs Mean (monthly)



Obs ERA-Interim NCEP/NCAR NCEP/DOE

The standard deviation and mean of the cloud properties :

>Obs: overall largest mean/std for the cloud properties

>ERA-interim: overall second largest mean/std for cloud fraction, and second largest std for SRCF and cloud albedo

R1/R2: overall similar mean/std, except R2 cloud fraction (albedo) std is slightly (significantly) larger than R1

!!! Observations show the largest mean/std !!!

Three Definitions of Homogeneous Mixing Fraction --- Ψ_3

$$\psi_{3} = \frac{\ln N - \ln N_{i}}{\ln N_{h} - \ln N_{i}} = \frac{\ln r_{v}^{3} - \ln r_{vi}^{3}}{\ln r_{vh}^{3} - \ln r_{vi}^{3}}$$

This definition, Ψ_3 , turns out to be related to α :

$$\psi_3 = 1 - \alpha$$

where α was defined by Morrison and Grabowski (2008):

$$N = N_0 \left(\frac{q}{q_0}\right)^{\alpha}$$

Two Transition Scale Numbers (2)

τ_{react} is based on:



- A: a function of pressure and temperature; **B:** a function of pressure, temperature and droplet number concentration (N_a or N_0).



Scale Number NLa

Explicit Mixing Parcel Model (EMPM)



Domain size: $20 \text{ m} \times 0.001 \text{ m} \times 0.001 \text{ m}$; **Adiabatic Number Concentration:** 102.7, 205.4, 308.1, 410.8, 513.5 c **Relative humidity:** 11%, 22%, 44%, 66%, 88%; **Dissipation rate:** 1e-5, 5e-4, 1e-3, 5e-3, 1e-2, 5e-2 m Mixing fraction of dry air:

Krueger (2008)

0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9.

FASTER Team



- 12 Institutions
- 24 + investigators with combined areas of expertise needed
- Major GCMs/SCMs
- Major NWP models
- WRF model
- CRM/LES models
- Observations

Uncertainty/Spread of ARM Measurements

•Large differences, esp. effective radius;

• Large difference for cloud fraction products

• Consideration of product spread in model evaluation;

• Possible way out:



Consistent definition of cloud fraction



References: Huang et al., 2012: An intercomparison of radar-based liquid cloud microphysics retrievals and implication for model evaluation studies. Atmos. Mea. Tech, submitted. In press