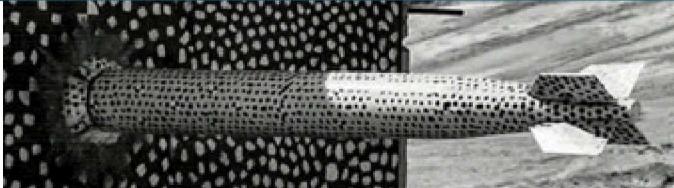


# Reducing the cost of radiative transfer in multi-scale atmosphere models



Benjamin R. Hillman



Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

# Two paths toward a cloud-resolving E3SM



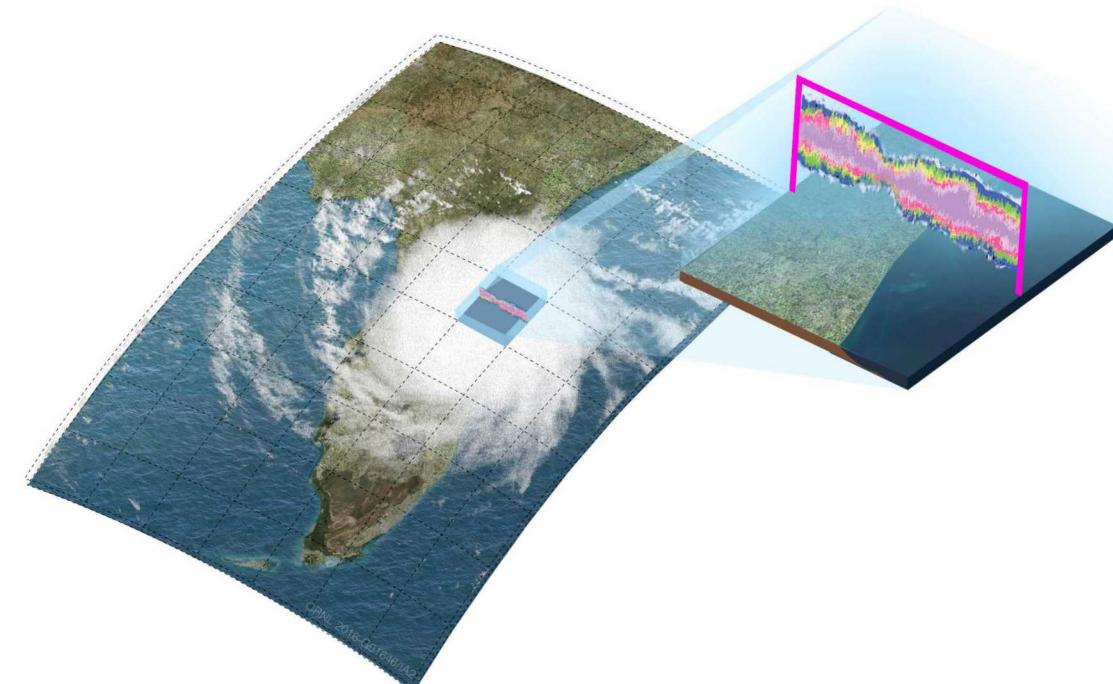
## E3SM with the Multi-scale Modeling Framework (E3SM-MMF)

- Explicitly resolve the large-scale and cloud-scale dynamics *separately*
- Embedded cloud-resolving model within each physics column
- Capture *some aspects* of cloud-resolving simulation, at lower computational cost (climate-scale)

## Simplified Cloud-Resolving E3SM Atmosphere Model (SCREAM)

- Push E3SM horizontal grid to cloud-resolving resolutions
- Simplified physics: P3 micro, Simplified Higher-Order Closure (SHOC), no deep convection
- *NOT A CLIMATE MODEL!*

# Multi-scale Modeling Framework



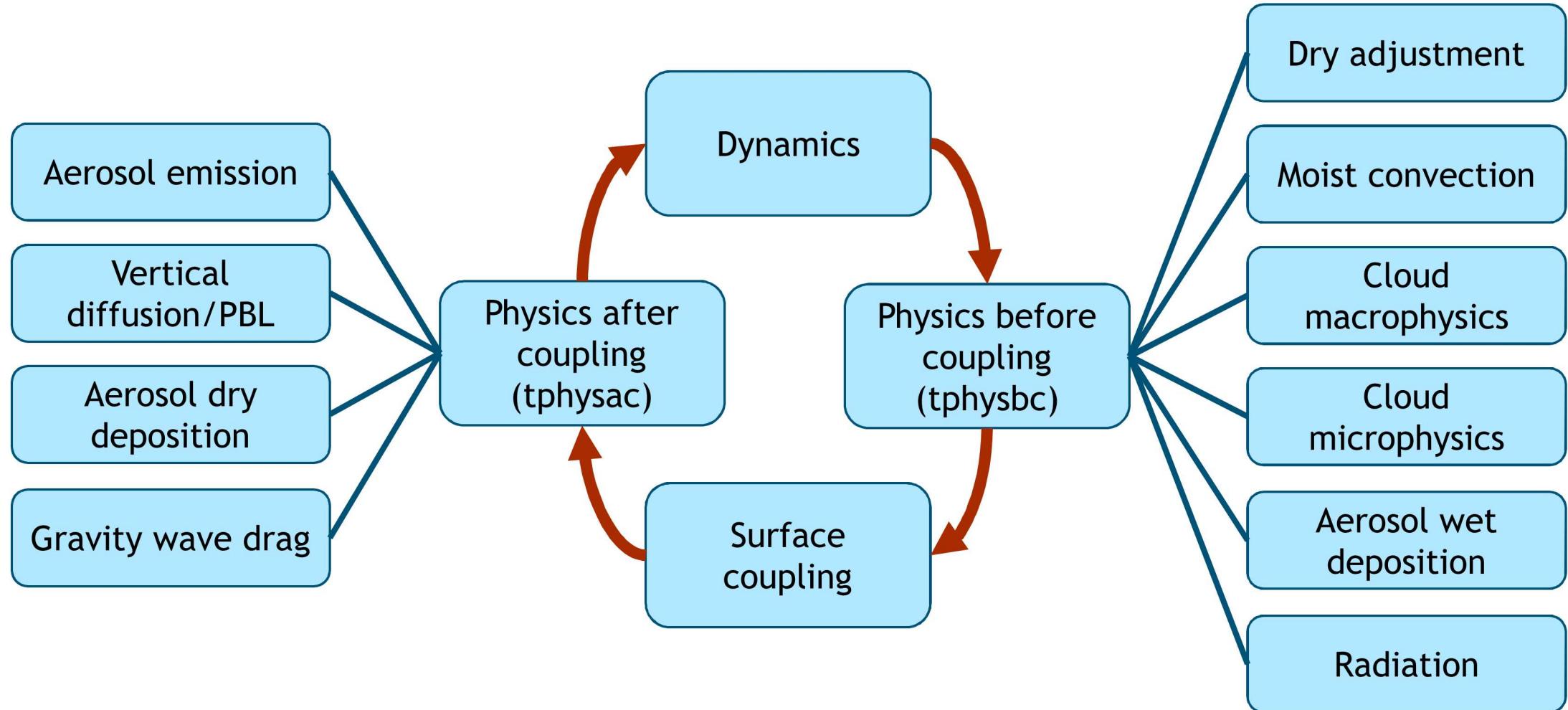
Traditional parameterizations introduce *structural uncertainty* in current global climate models

Pushing to higher resolutions allows us to drop more physical parameterizations, but this is *expensive*

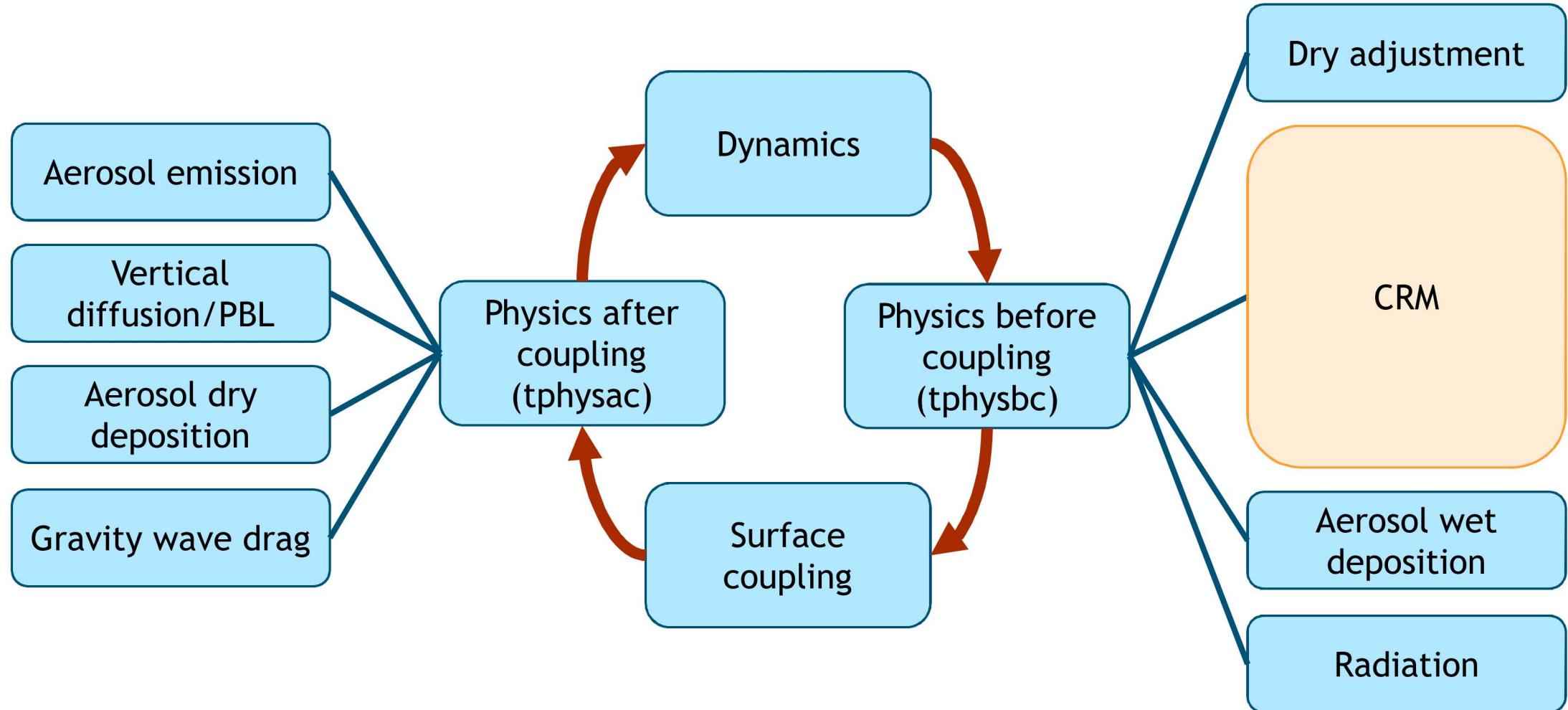
Replace cloud and convective parameterizations in a traditional global climate model with embedded cloud resolving models in each column

Exascale computers + MMF will make it possible to perform traditional climate simulations with *some aspects* of cloud resolving simulations

# E3SM Atmosphere Model schematic



# E3SM-MMF Atmosphere Model schematic



# E3SM-MMF model specifics



“Host” GCM: E3SM Atmosphere Model

- HOMME spectral element dynamical core
- v1 physics, minus convection and cloud macro and micro physics

Embedded CRM: System for Atmospheric Modeling

- Currently using single-moment microphysics scheme
- Prescribed aerosol

Target throughput: 5 simulated years per day

# Computational speed-ups

GPU port of CRM code (Matt Norman)

Mean State Acceleration (Chris Jones)

Reduced radiation resolution/frequency (Ben Hillman and Walter Hannah)

## GPU port of CRM code



Entirety of the time-stepping loop within the CRM ported using OpenACC directives-based approach

CRM code refactored to include  $ncol$  dimension from global model to expose more parallelism; do multiple CRMs at once

15-16x speed-up on summit (two P9s vs six Volta per node)

Benchmark on Summit for Gordon Bell submission used 4,600 nodes of Summit, achieved 2.5% peak double precision flop/s, throughput of 1.8 SYPD (with Mean State Acceleration and reduced radiation)

## 9 Mean State Acceleration (MSA)

Reduce number of timesteps required for CRM integration by artificially introducing a “mean-state” tendency

Push the CRM faster towards mean state

Rationale: turbulent eddies spin-up fast relative to evolution of mean state

Jones, Bretherton, and Pritchard (*JAMES*, 2015, <https://doi.org/10.1002/2015MS000488>)

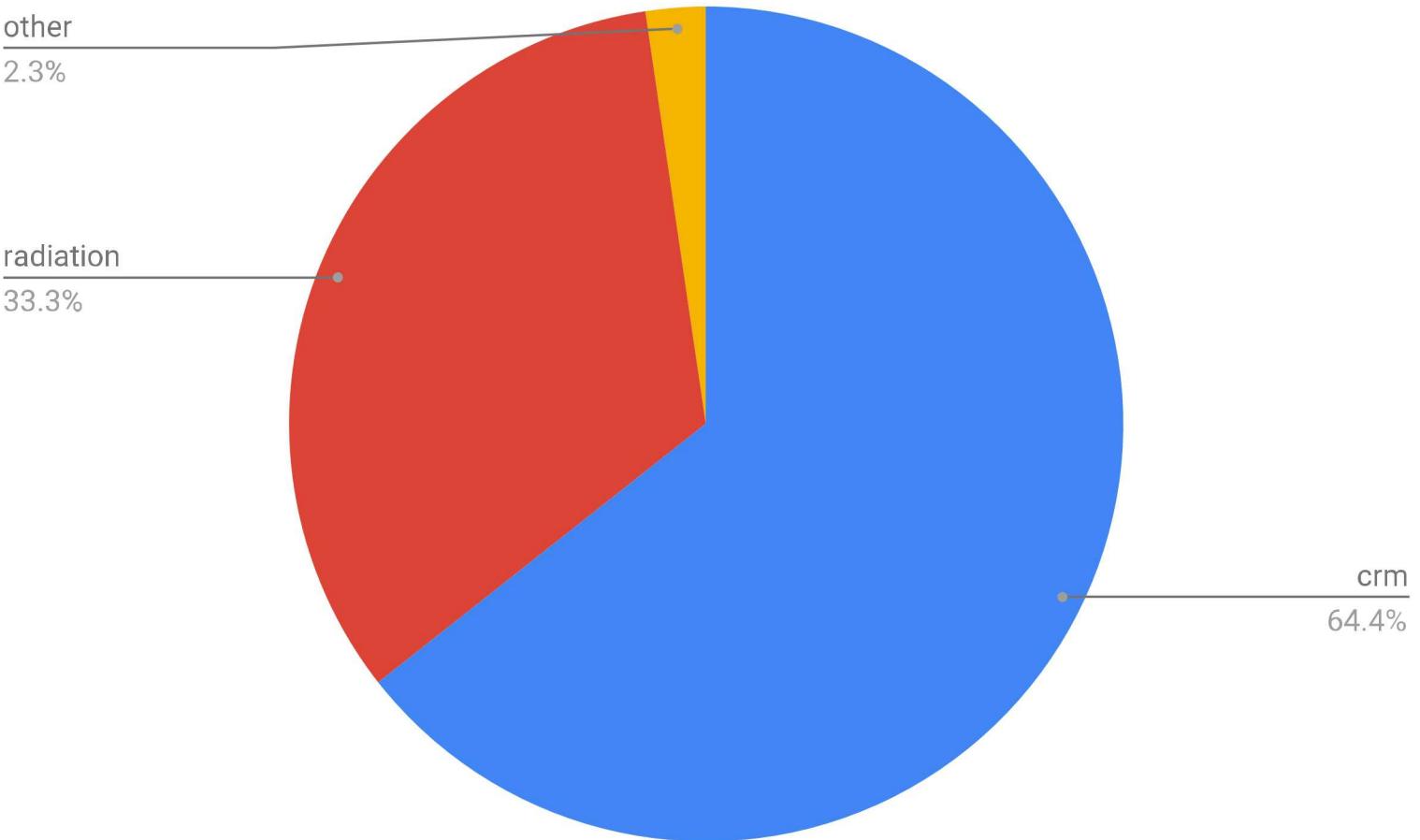
Currently an acceleration factor of about 4 appears to be stable

# Radiation cost

Radiation is expensive!

Reduce computational cost of radiation by reducing frequency and number of columns

Balance between efficiency and accuracy



Relative cost of physics packages on Intel Sandy Bridge (chama)

# Why is radiation so expensive

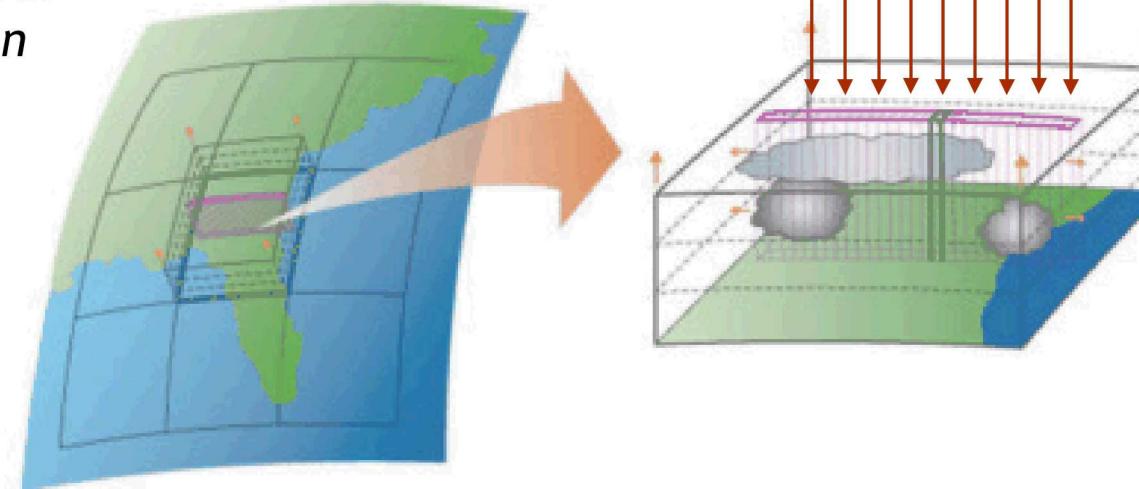


Fluxes and heating rates: integrated quantities over many spectral intervals

Lots of columns, lots of bands...lots of calculations

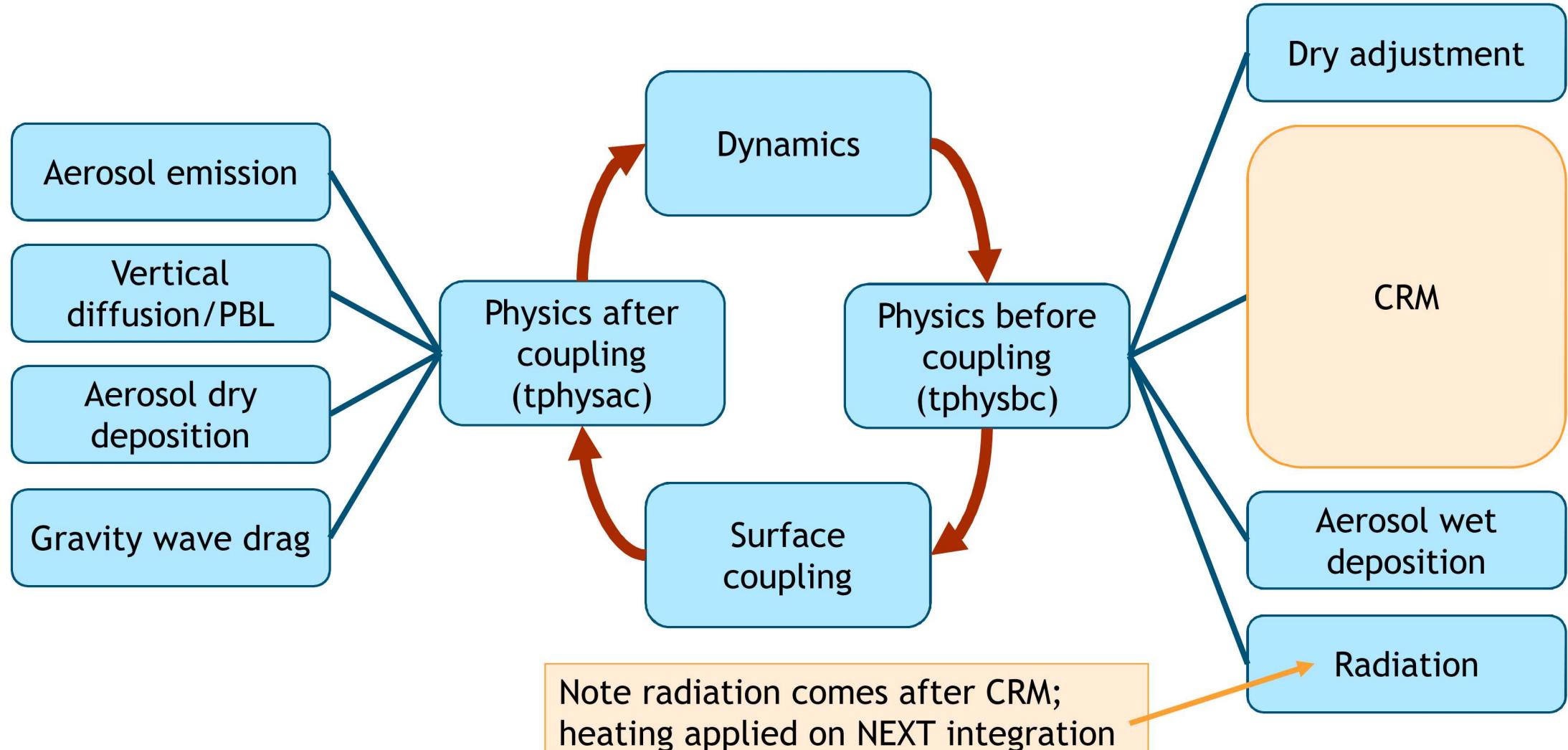
Exacerbated in models using the MMF: need to calculate fluxes and heating rates on *each CRM column*

Need to calculate fluxes for a large number of spectral intervals for *each column*



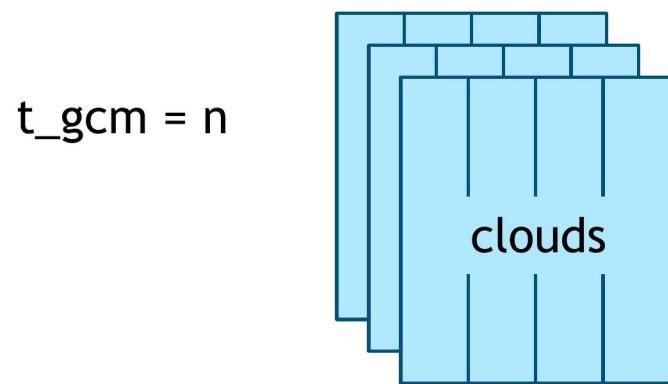
Now need to calculate fluxes for a large number of spectral intervals for *each CRM column* within each GCM column!

# E3SM-MMF Atmosphere Model schematic

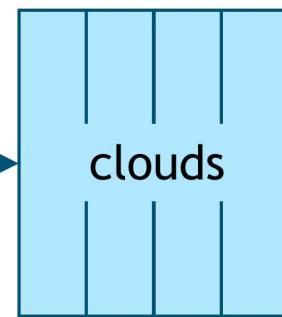


# Radiative coupling in E3SM-MMF

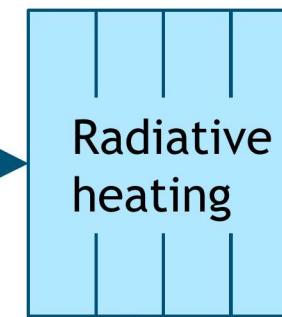
Cloud properties within CRM  
at  $t_{\text{crm}} = 1, \dots, dt_{\text{gcm}}$



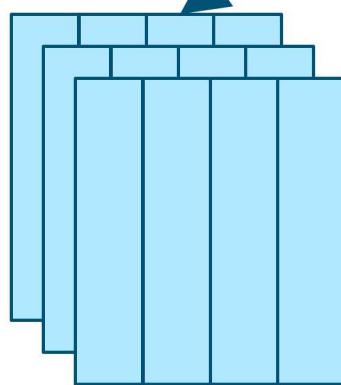
Cloud properties *time-averaged*  
within the CRM, then passed  
back to global model



Radiative heating calculated  
from *time-averaged* profiles,  
but at each CRM column



$t_{\text{gcm}} = n + 1$

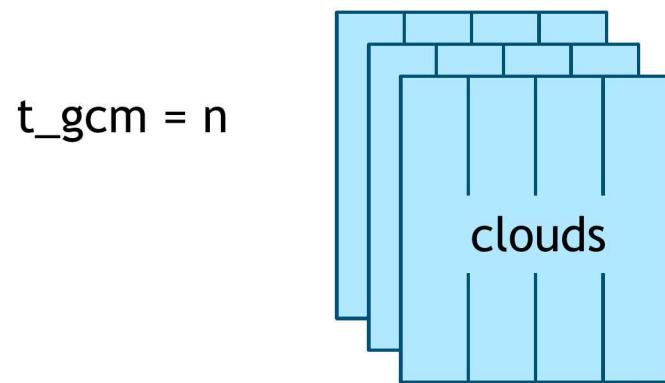


Radiative heating is  
applied during the  
*next* CRM integration

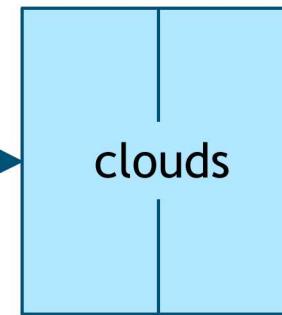
NOTE: This step is  
expensive!

# Radiative coupling in E3SM-MMF (reduced radiation)

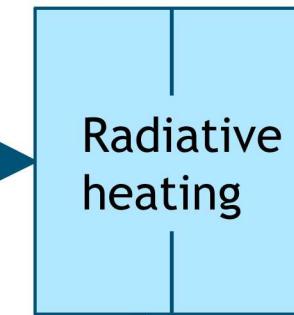
Cloud properties within CRM  
at  $t_{\text{crm}} = 1, \dots, dt_{\text{gcm}}$



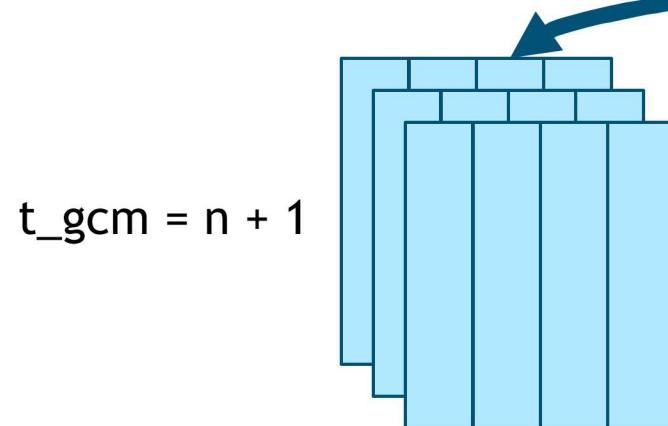
Cloud properties *time-averaged*  
within the CRM, then passed  
back to global model



Radiative heating calculated  
from *time-averaged* profiles,  
now on group-averages of CRM  
columns



*Spatially averaged*



Radiative heating is  
applied during the  
*next CRM integration*

To save cost, calculate and  
apply heating on *grouped*  
*columns*

## Minimizing biases resulting from using less columns

Using spatially-averaged cloud properties removes cloud-scale features

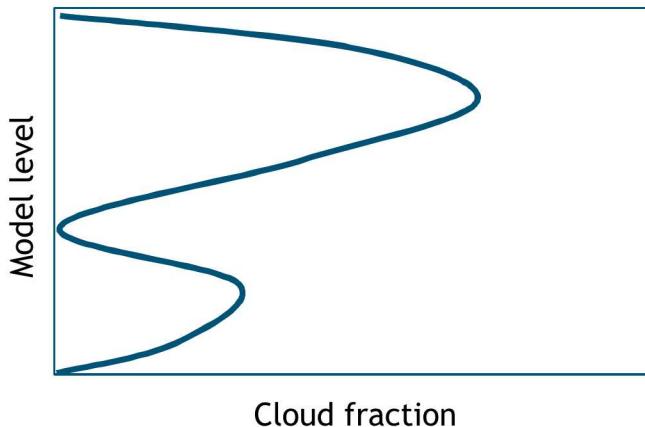
Individual columns no longer just “clear” or “cloudy” but have *partial cloud fraction*  $C: [0, 1]$

This is accounted for in the GCM through subcolumn sampling in the “MCICA” approach: reconstruct psuedo-cloudy/clear elements from domain averages with stochastic subcolumn sampling

Enabling for the CRM adds no cost (already being done internally but with trivial inputs)

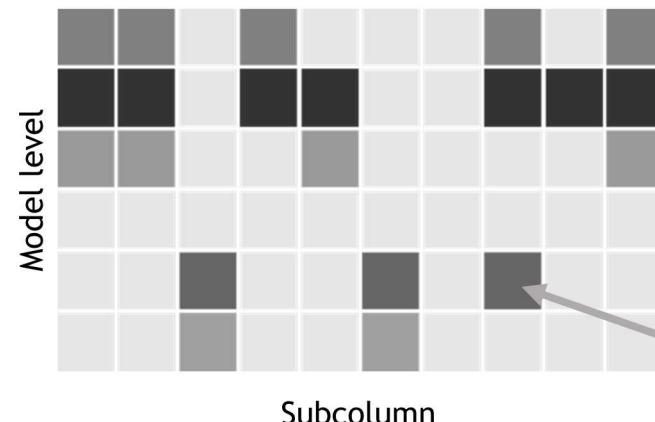
# Subcolumn sampling

Start with column-mean profiles of cloud fraction (and condensate amount)



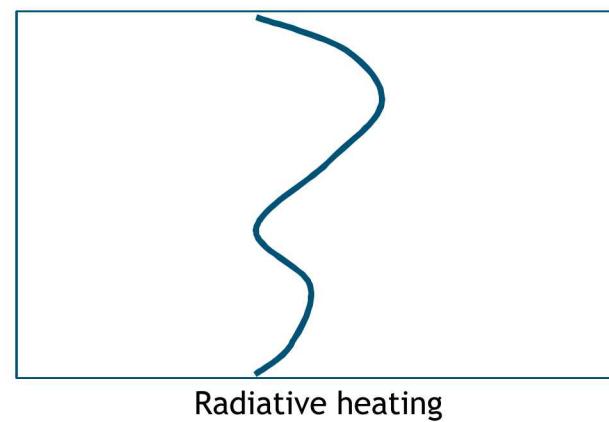
(For GCM, this is a single gridbox average; for MMF, this is a single time-averaged CRM column)

Stochastic subcolumn sampling with an overlap assumption to generate pseudo-resolved clear/cloudy elements



Note for CRM, this is subsampling *within a single CRM column*

A *single subcolumn* chosen for each spectral band; calculate band-resolved fluxes, return broadband heating



Colors: condensate amount

# Experiment setup



Small, rapid test configurations (ne4 resolution, 1 month duration)

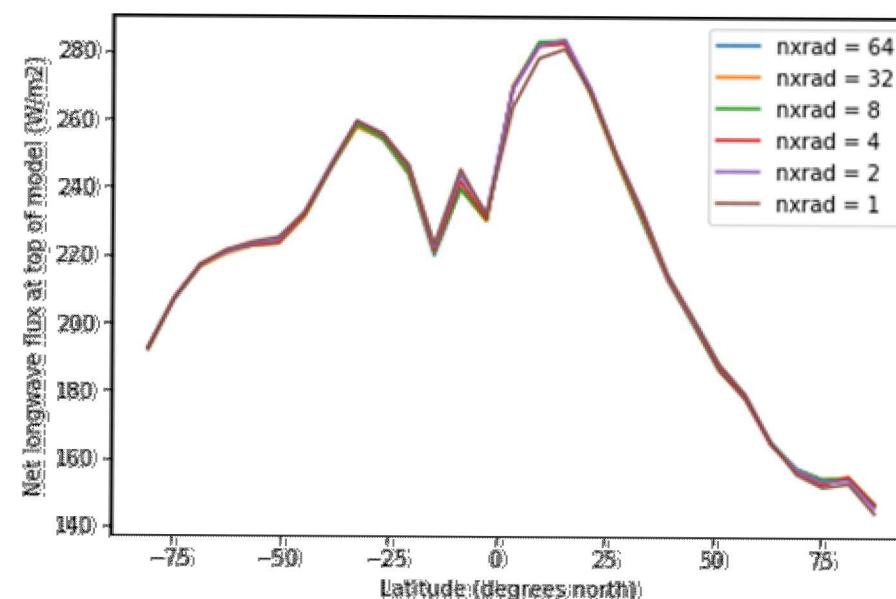
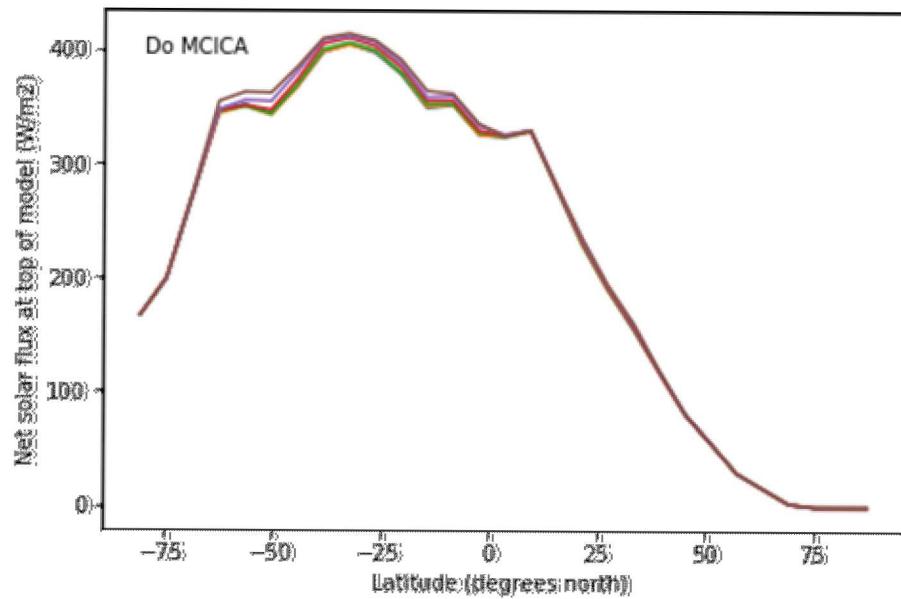
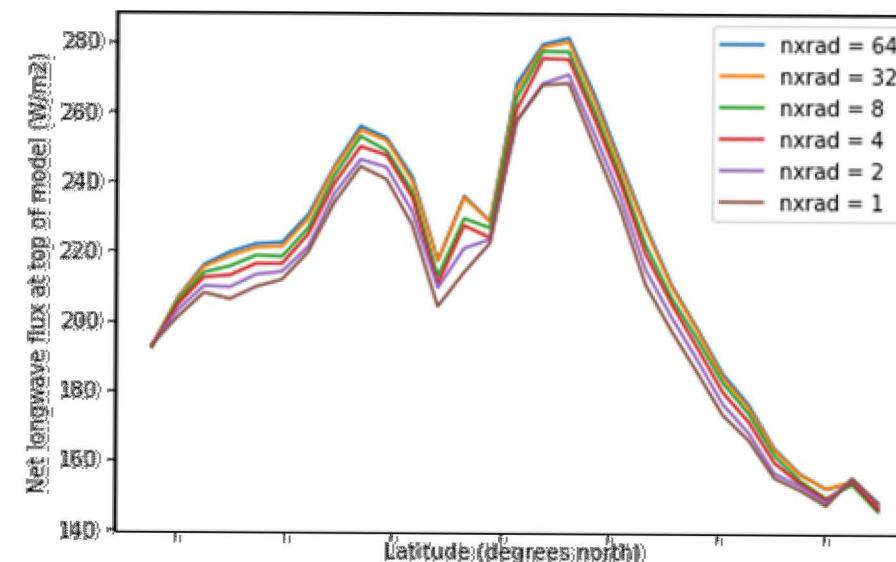
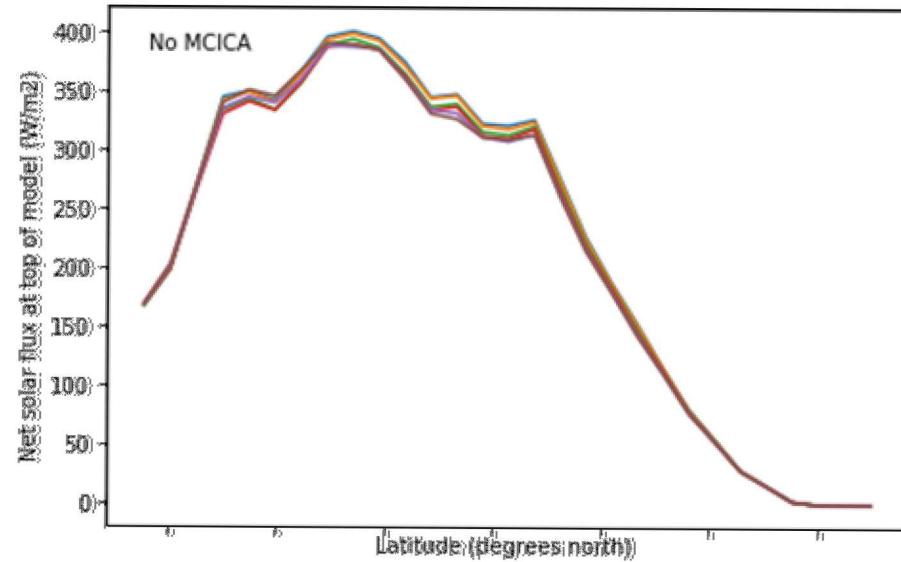
- Not a realistic resolution for climate, but sufficient for informing us about the affects of these changes since they largely affect the CRM itself, not the host model

CRM uses a 2D domain with 64 columns (oriented north-south) with a grid spacing of 1 km

Different cases using 1, 2, 4, 8, 32, and 64 columns for the radiation (i.e., group sizes of 64, 32, 16, 8, 2, and 1)

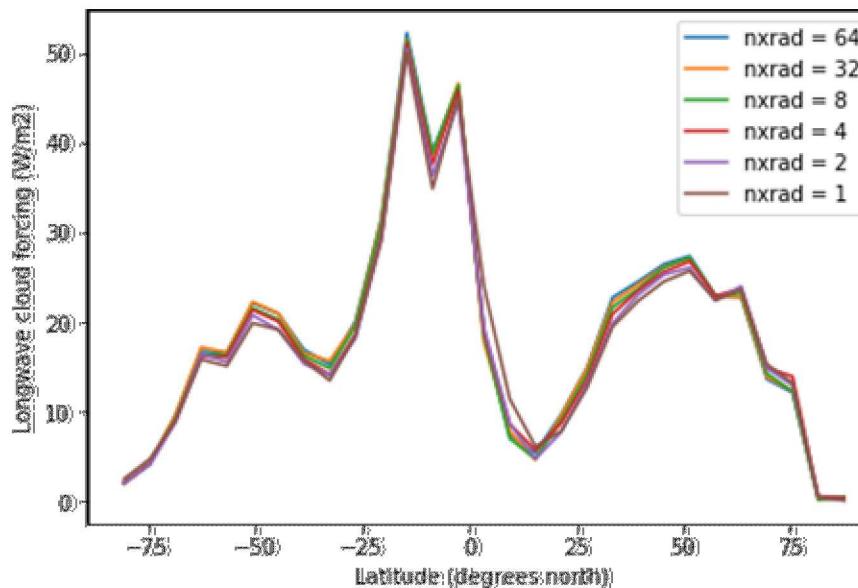
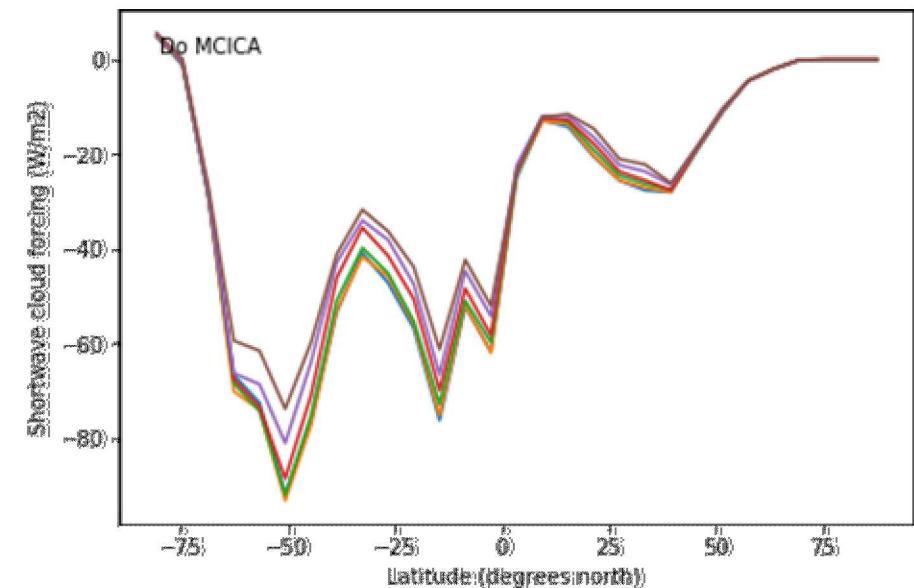
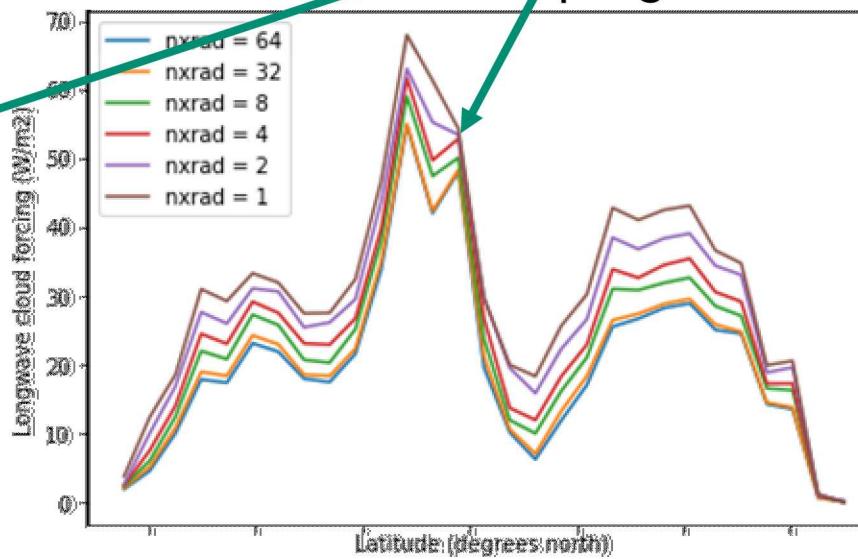
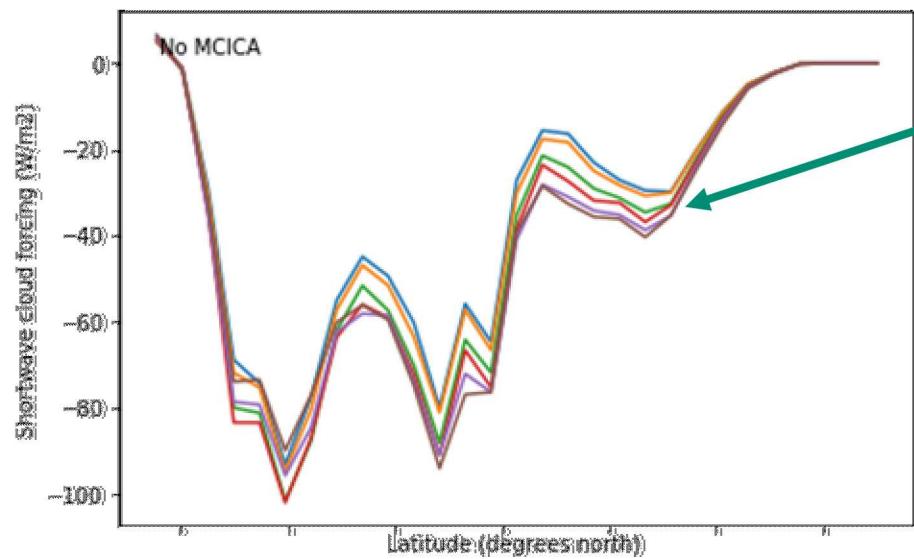
Identical tests with and without subcolumn sampling for overlap

# Errors in TOA fluxes

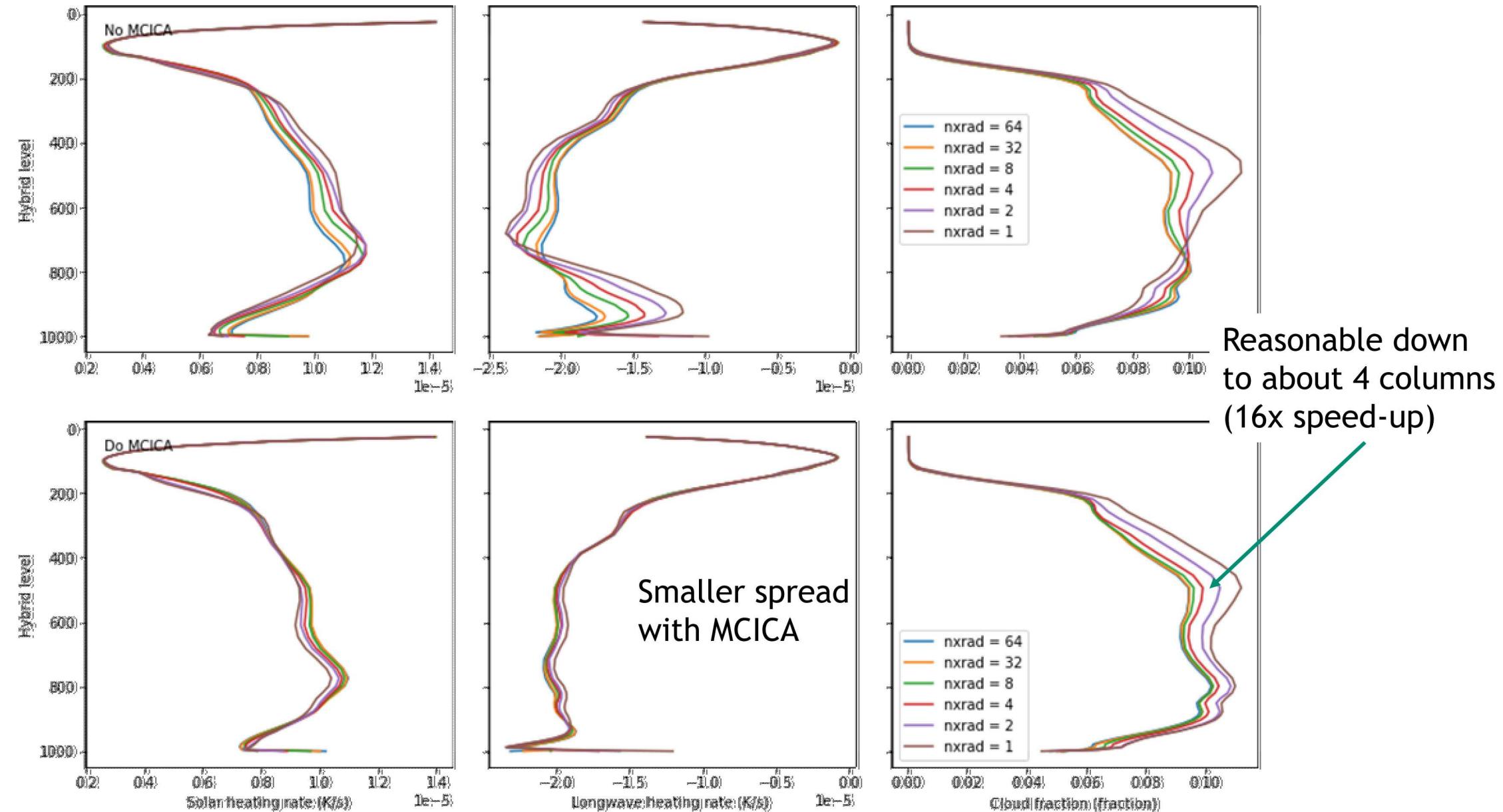


# Errors in cloud radiative effects

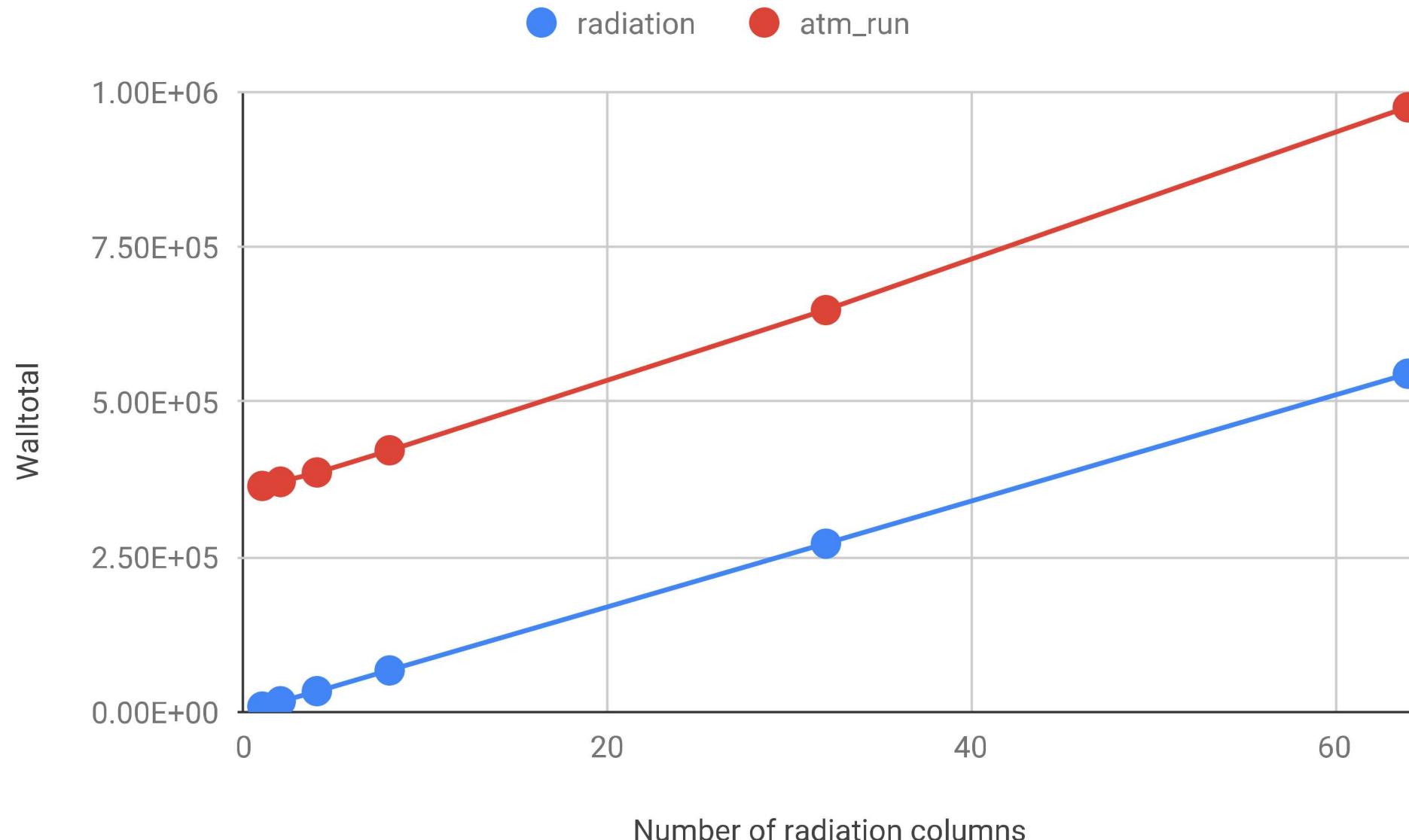
Large sensitivity with no MCICA subcolumn sampling



# Errors in heating and cloud profiles



# Speed-up from reduced radiation



## Can we just use domain averages?

It would be simple and cheap to just reduce down to *one* radiation column per CRM, and let the GCM handle radiation (even have nice ways of representing the variability)

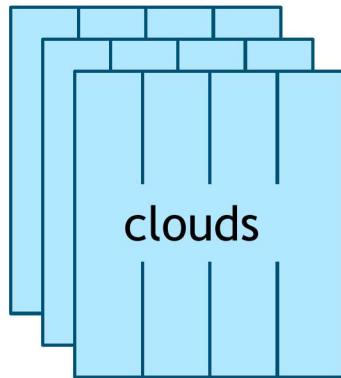
To test, calculate heating rates on full CRM resolution, but homogenize before applying them

*Tests the question: if we had a perfect subcolumn sampling scheme, could we get back the same solution when reducing down to one radiation column?*

# Homogenized heating test

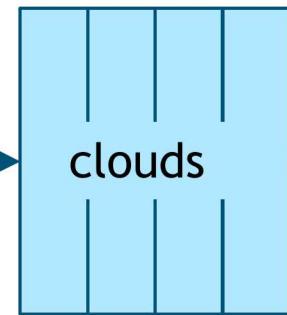


Cloud properties within CRM  
at  $t_{\text{crm}} = 1, \dots, dt_{\text{gcm}}$

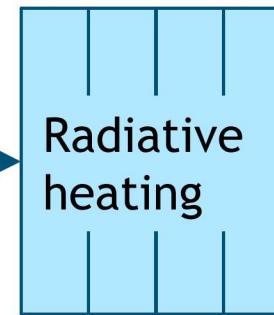


$t_{\text{gcm}} = n$

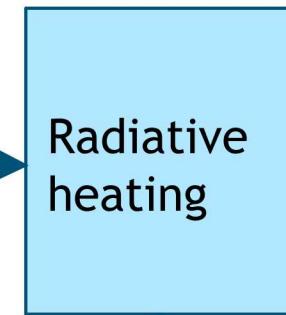
Cloud properties *time-averaged*  
within the CRM, then passed  
back to global model



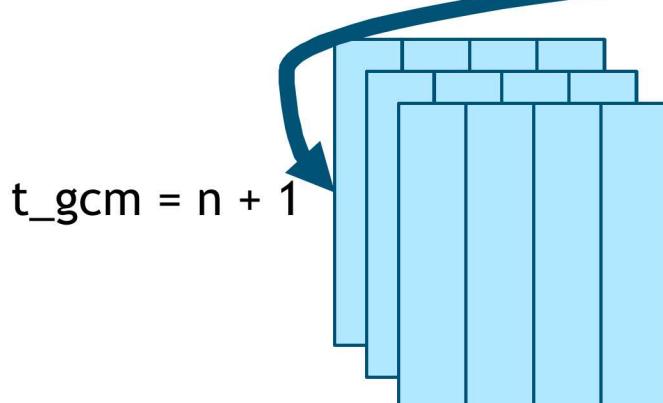
Radiative heating  
calculated from *time-averaged*  
profiles, but at  
each CRM column



Average heating  
rates before  
passing to CRM

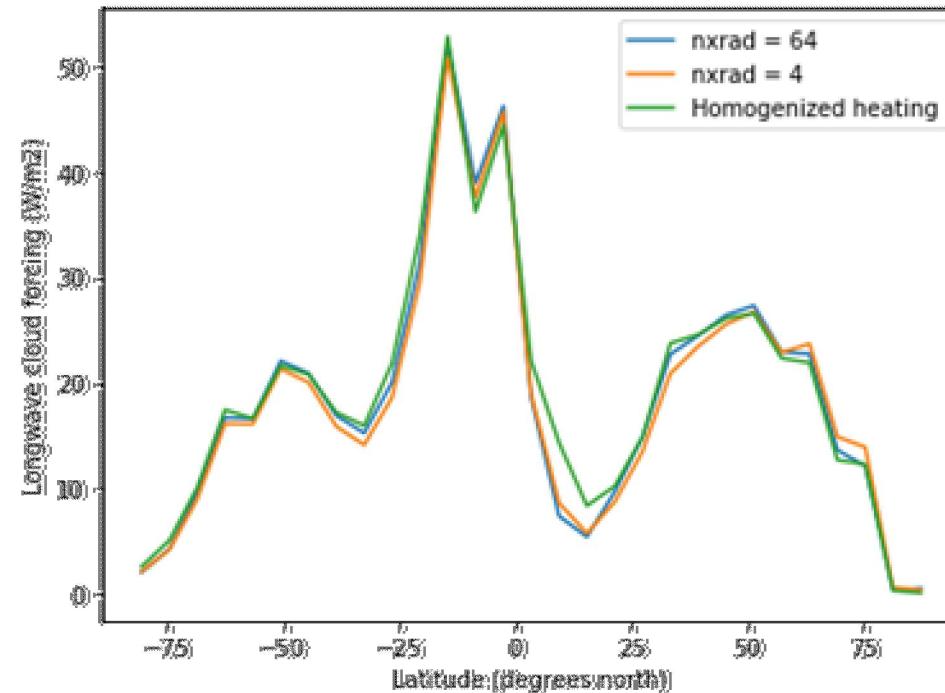
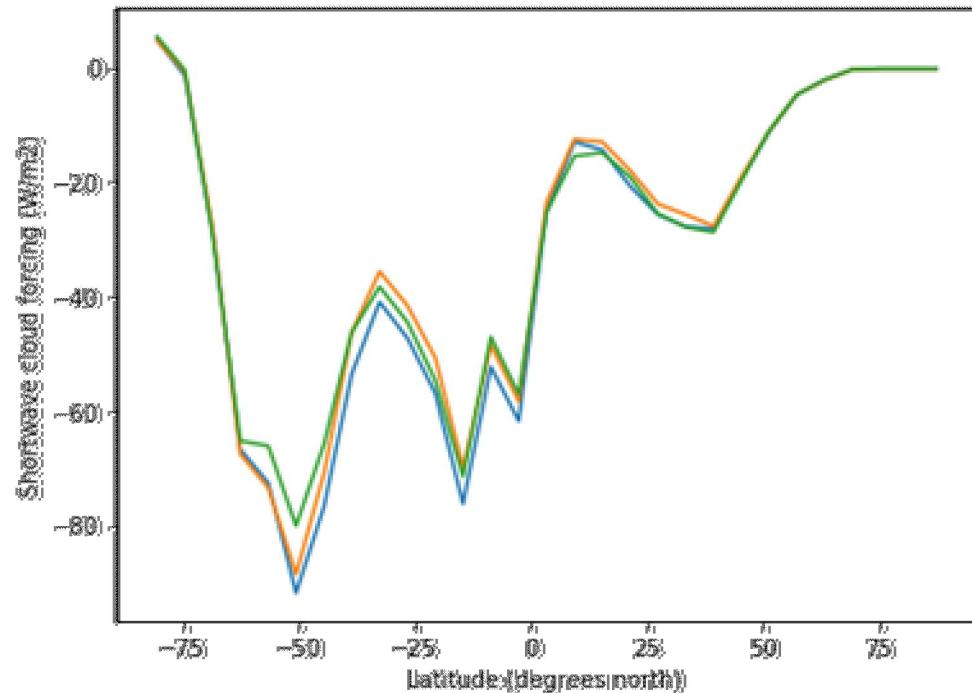


Radiative heating is  
applied during the  
*next* CRM integration

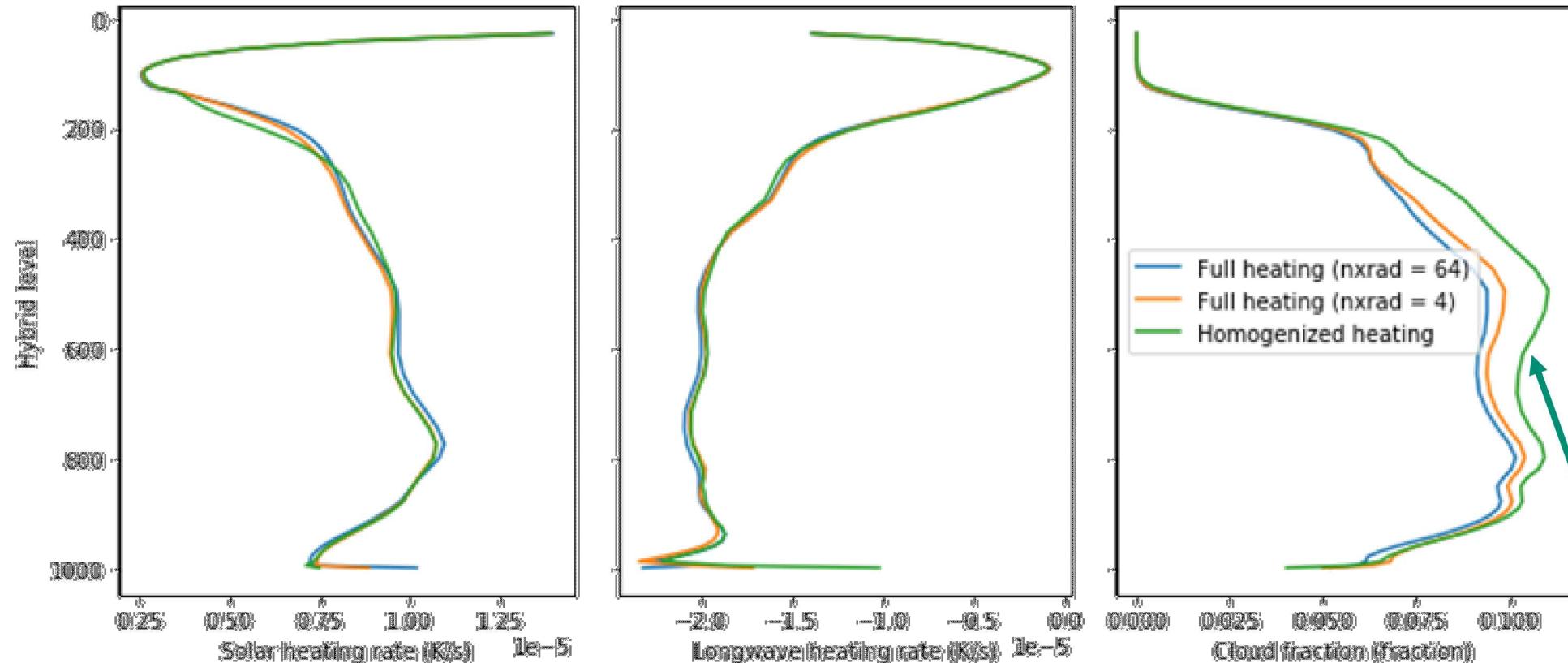


Calculate heating with full CRM  
resolution, but homogenize before  
applying tendencies to CRM

# Homogenized heating test



# Homogenized heating test



Effect of feedbacks  
from cloud-scale  
variability in  
heating

## Summary of reducing spatial resolution

Reduced radiation alone has large impact on solution

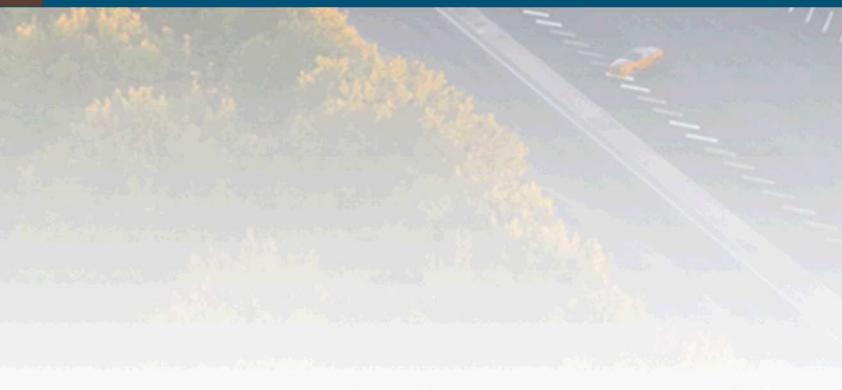
Using subcolumn sampling reduced that impact

Can get significant speedup, with tolerable differences from baseline

*Cannot* reduce this all the way down to a single radiation column: lose impact of small-scale variability in heating on cloud evolution/micro-circulations



Reducing temporal frequency



# How often do we need to update the heating?

Previous MMF implementations called radiation at every physics timestep

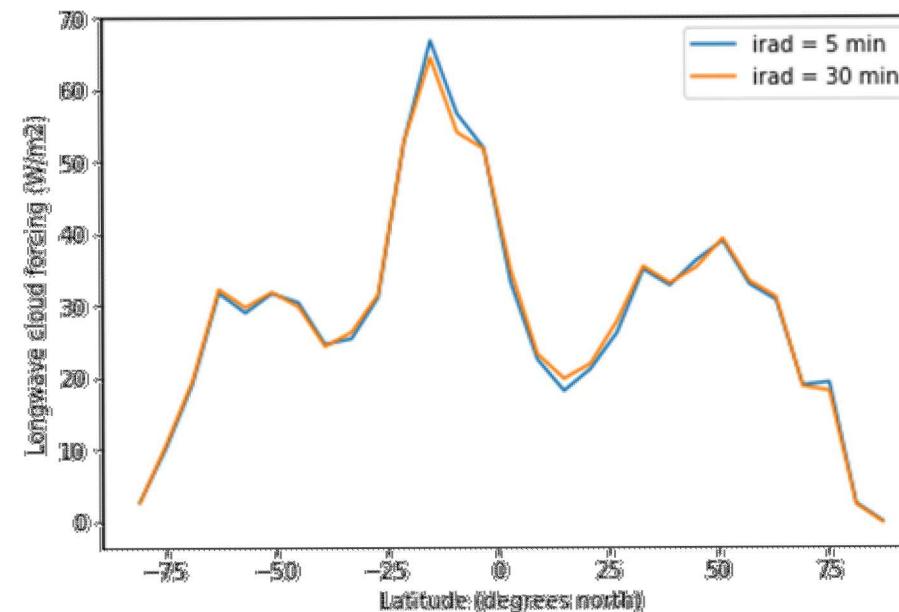
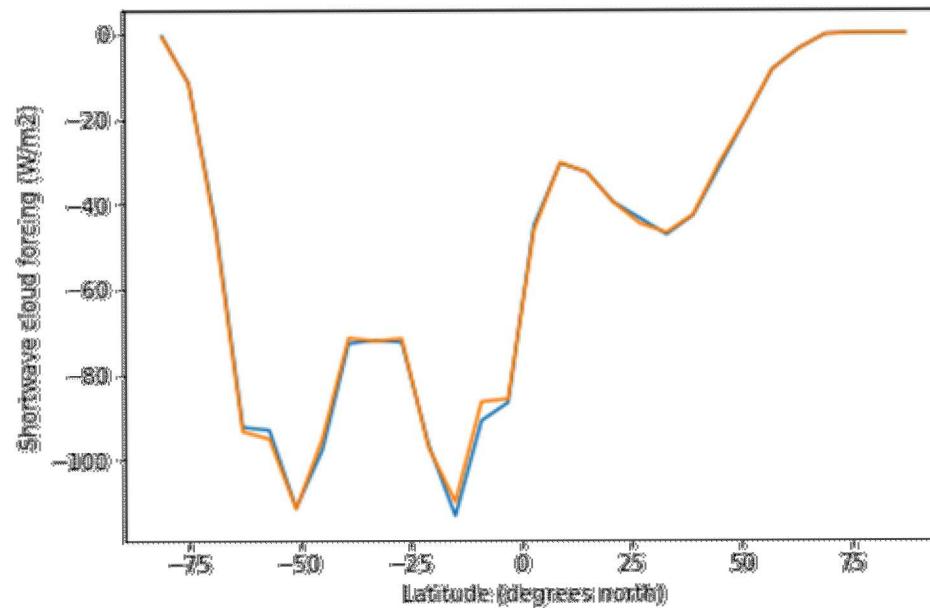
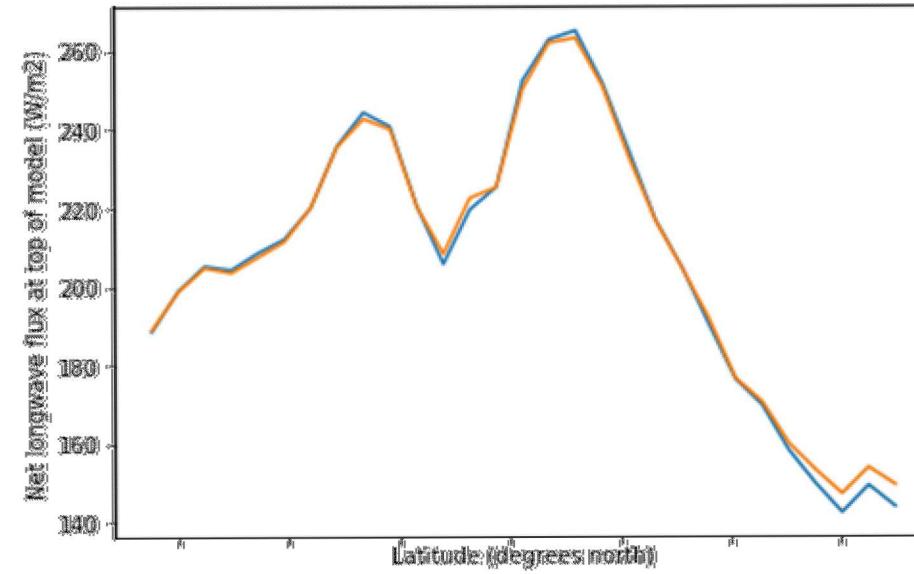
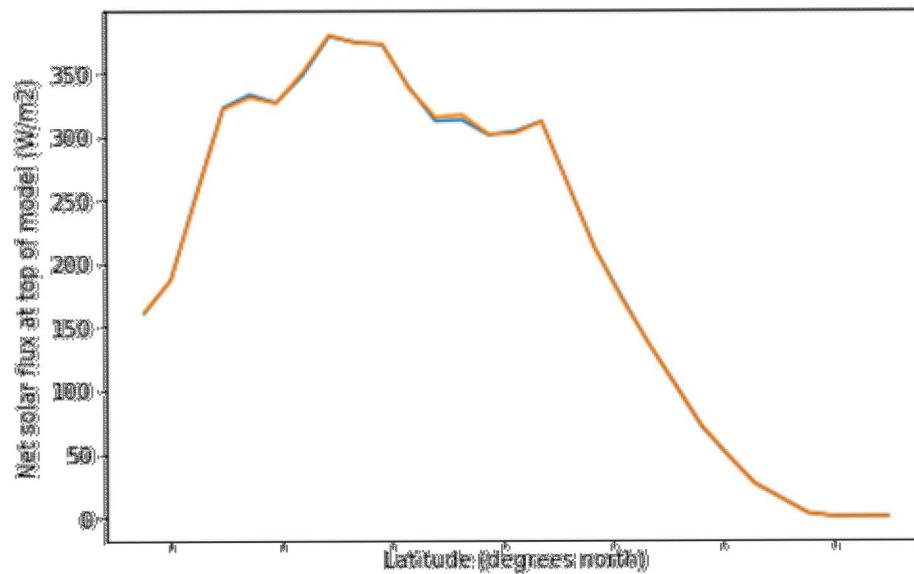
In E3SM, we usually only update the radiative heating once every 30 minutes

Can we get away with updating radiation less frequently in E3SM-MMF?

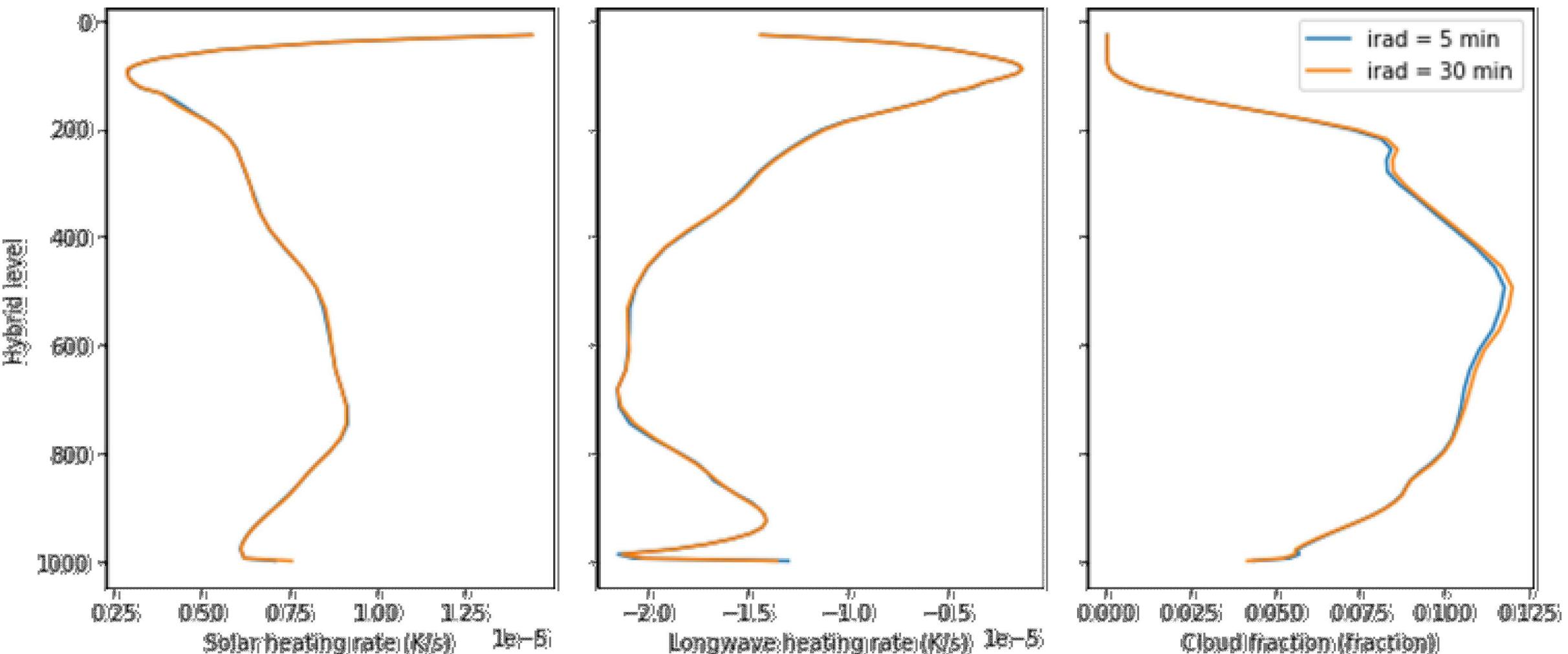
Experiment:

- Baseline: radiative heating updated every 5 minutes (every physics timestep)
- Test: radiative heating updated only every 30 minutes (every sixth timestep)
- Note that heating is still applied every CRM timestep

# Biases due to reduced temporal frequency



# Bias due to reduced temporal frequency



NOTE: different physics timestep  
and CRM dx than previous tests

## Summary of techniques for reducing cost

Reduced spatial resolution can provide significant speed-up, at cost of manageable errors in heating

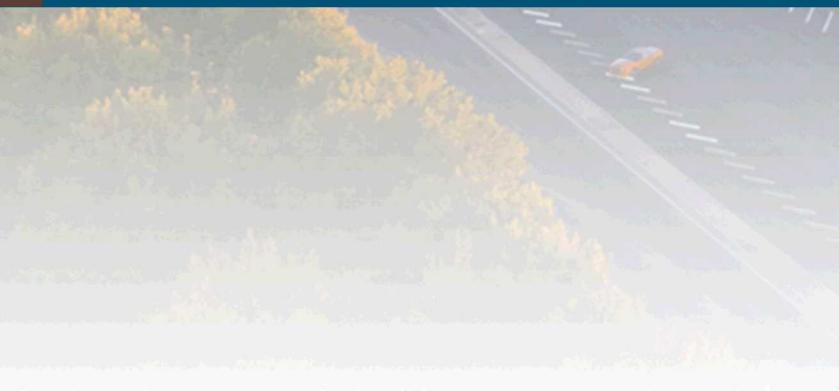
Reduced temporal resolution provides additional speed-up, with little impact on simulation (although this needs further exploration)

Will this become unnecessary on the GPU?

Thank you!



# Extra slides



# Monte Carlo Independent Column Approximation

$$\langle F^{ICA} \rangle = (1 - A_c) \sum_k^K w(\lambda_k) S(\lambda_k) F^{clr}(\lambda_k) + A_c \sum_k^K w(\lambda_k) S(\lambda_k) \sum_j^J p(s_j) F(s_j, \lambda_k)$$

Do this for every column

Sum over bands  
(order 10^2)

Sum over possible cloud states (subcolumns)

$$\langle F^{McICA} \rangle \approx (1 - A_c) \sum_k^K w(\lambda_k) S(\lambda_k) F^{clr}(\lambda_k) + A_c \sum_k^K w(\lambda_k) S(\lambda_k) F(s_{rnd}, \lambda_k)$$

Grab a random subcolumn cloud state for each spectral interval, rather than calculate flux for each spectral interval for each cloud state