

Understanding Resolution Sensitivity in the Community Atmosphere Model

A Dissertation presented

by

Adam R. Herrington

to

The Graduate School

in Partial Fulfillment of the

Requirements

for the Degree of

Doctor of Philosophy

in

Atmospheric Sciences

Stony Brook University

May 2019

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Stony Brook University

The Graduate School

Adam R. Herrington

We, the dissertation committee for the above candidate for the
Doctor of Philosophy degree, hereby recommend
acceptance of this dissertation

Kevin A. Reed, Assistant Professor and Advisor
School of Marine and Atmospheric Sciences

Minghua Zhang, Distinguished Professor and Chairperson of Defense
School of Marine and Atmospheric Sciences

Marat F. Kharoutdinov, Professor
School of Marine and Atmospheric Sciences

Peter H. Lauritzen, Scientist III
Climate and Global Dynamics Laboratory
National Center for Atmospheric Research

Colin M. Zarzycki, Assistant Professor
Department of Meteorology and Atmospheric Science
Pennsylvania State University

This dissertation is accepted by the Graduate School

Interim Dean of the Graduate School

ABSTRACT

Understanding Resolution Sensitivity in the Community Atmosphere Model

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2019

Many next-generation atmospheric general circulation models (AGCM) allow for substantial grid flexibility, enabling the representation of a wider-range of horizontal scales at reduced computational cost through the use of variable resolution grids. AGCMs, however, are notoriously sensitive to grid resolution, and in some cases, solutions noticeably diverge across the refinement region

of variable resolution grids. The lack of scientific consensus on the reason(s) solutions diverge with increasing horizontal resolution remains an obstacle to more widespread application of variable resolution grids to regional climate problems. It is the purpose of this thesis to develop and apply a hierarchy of reduced complexity model configurations across horizontal resolutions, to isolate and understand resolution sensitivity in an AGCM, the Community Atmosphere Model (CAM). The hierarchy of configurations used in this thesis range from a dry thermal bubble test, to an aqua-planet configuration with comprehensive moist physics, to a realistic, present day simulation. The magnitude of resolved vertical motion increases substantially with resolution in response to the buoyancy produced by the stratiform cloud scheme. This sensitivity follows from the fact that higher resolution grids are able to support smaller buoyancy length scales, forcing pressure gradients and related vertical velocities to scale like the inverse of the grid spacing.

An understanding of resolution sensitivity has proven useful in two separate applications. (1) The impact of computing the physical parameterizations on a separate, coarser resolution grid than used by the dynamical core (a ‘physics grid’) is evaluated in detail. It was found that a coarser physics grid does not degrade the effective resolution of the model, with the vertical velocity still determined by the dynamical core grid spacing, rather than the physics grid spacing. Since the physical parameterizations are responsible for about half the computational cost of the conventional CAM configuration, the coarser physics grid allows for significant cost savings with little to no downside. (2) Understanding the relationship between grid-scale clouds, parameterized convection

and resolution in CAM. For 30 years, CAM and it's predecessor versions have all shown that parameterized convective precipitation decreases at the expense of stratiform precipitation and the atmosphere becomes drier and less cloudy, with increasing resolution. A strong case was presented in which the larger magnitude vertical velocities with increasing resolution progressively dries and stabilizes the atmosphere, and the stability dependent threshold in the deep convection scheme is triggered less often. The culmination of this thesis is a robust framework for understanding resolution sensitivity, which may be used to anticipate and alleviate issues arising from the increasingly diverse types of grids that are now common among AGCMs for simulating regional climate.

DEDICATION

For my beloved brother, Jesse

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ACKNOWLEDGEMENTS

This thesis is a culmination of four years of hard work, but only made possible with the support of several people. I work well independently, but this can sometimes be a character flaw —left to my own devices, I would probably go long stretches without checking in with my advisor. Dr. Kevin Reed set very clear expectations from day one that we would meet weekly to discuss progress. Kevin provided invaluable scientific and technical support, encouraged me to attend conferences and was crucial in helping me navigate and secure a postdoctoral career. For all these reasons and more, thank you Kevin. I am also extremely grateful to have participated in a year long visit to the National Center for Atmospheric Research (NCAR), as an Advance Study Program graduate student visitor under the mentorship of Dr. Peter Lauritzen. The knowledge expelled onto me from Peter cannot be overstated. I am a better scientist for being graced by his presence and I appreciate all the time he devoted to helping me understand the model development world.

I would like to also thank committee members Drs. Minghua Zhang, Marat Khairoutdinov and Colin Zarzycki. Each member provided excellent feedback throughout this process, and each have had illuminating one-on-one scientific discussions with me along the way. From a previous life, I would like to thank my Masters' advisor Dr. Chris Poulsen, who saw me through a period when

I still had my climate model training wheels on, and which enabled me to hit the ground running when I first arrived as a Ph.D. student at Stony Brook. I would like to thank my grand advisor, Dr. Christiane Jablonowski, who exposed me to the art of climate modeling as a Masters' student, providing me hope that a black box can in fact be understood.

And lastly, but certainly not least, I want to thank mother, Bobbi Herrington, sister, Rachel Herrington and grandmother, Martha Wilk for being such a wonderful and loving family.

CHAPTERS

1 Introduction: The history of resolution sensitivity in CAM and its predecessors

1.1 A brief history of Atmospheric General Circulation Models

The atmosphere consists of a large number of moving parts, operating over an impressive range of space-time scales. Despite the enormous complexity of the atmosphere, and against all odds, it was eventually realized that the general circulation of the atmosphere —referring to the large-scale overturning circulations such as the Hadley and Ferrell Cells, including synoptic scale disturbances embedded within the jet-stream— can be modeled using surprisingly simple idealizations. In the 1950’s, scientist Norman Phillips noted that “dishpan” experiments —a rotating pan of water heated on its edges with an electric coil— appear remarkably similar to large-scale weather patterns in the atmosphere (Weart, 2008),

at least the gross features of the general circulation of the atmosphere can be predicted without having to specify the heating and cooling in great detail.

Inspired by this observation, Phillips developed a 2-layer numerical model of the atmosphere using a simplified set of equations and idealized heating of the lower layer. His model produced a realistic looking jet-stream with synoptic disturbances. Like the dishpan experiments, his numerical model was able to

produce the large-scale features of the atmosphere while neglecting finer-scale processes thought to be important in the atmosphere. The Phillips model is considered the first Atmosphere General Circulation Model (AGCM) and was extremely coarse in resolution with horizontal grid spacing $\Delta x = 22.5^\circ$ (Weart, 2008).

Phillips' AGCM only applies to the mid-latitudes; models developed by Manabe et al. (1965) and Held and Hou (1980) solidified the idea that relatively simple parameterizations of diabatic heating can produce realistic tropical circulations as well. The study of Held and Hou (1980) is in particular quite remarkable in that it does not contain any explicit representation of moisture. Moist processes and radiative cooling were instead parameterized as a simple newtonian relaxation of potential temperature to an inventive reference profile. Even with the crudeness of this approximation, a realistic Hadley Cell is simulated.

Manabe et al. (1965) explicitly simulates advection of water vapor by the resolved dynamics, parameterizes convection and clouds using a critical lapse rate approach and super-saturation criteria, and radiation using a radiative transfer model. The Manabe et al. AGCM is significantly more complex than any AGCM that had come before, and is structurally similar to modern AGCMs. The horizontal grid spacing of the Manabe et al. model is $\Delta x = 4.5^\circ$.

This brief history of AGCM development is intended to provide the reader with the reason AGCMs are on grids that cannot resolve, and therefore largely ignore myriad of scales and processes. The short answer — because it still works. But with the vastly greater computational power realized over the

decades following the first AGCM models, scientists have simulated the atmosphere using finer and finer grid resolutions, and with this came a whole new set of challenges. The title of this thesis — understanding resolution sensitivity in the Community Atmosphere Model — is focused on one of these overarching challenges in one AGCM.

Standard convergence tests in the Community Atmosphere Model (CAM) and its predecessor versions result in weak, or non-convergent solutions with increasing horizontal resolution (Kiehl and Williamson, 1991; Williamson et al., 1995; Williamson, 1999, 2008; Li et al., 2011; Reed and Jablonowski, 2011b; Reed et al., 2012; O'Brien et al., 2013; Rauscher et al., 2013; Zarzycki et al., 2014b; Lu et al., 2015). While this has been a documented feature of the model since not long after its inception, from a strict computational fluid dynamics perspective this is not acceptable model behavior (Williamson, 2008). The following sections provide a review of these convergence studies, while also documenting the major model developments, which has been increasing in complexity for over thirty years.

1.2 CCM1

The National Center for Research (NCAR) Community Climate Model, version 1 (CCM1; Williamson et al., 1987) is the successor to version 0. Kiehl and Williamson (1991) is the first convergence study using a NCAR climate model. The CCM1 dynamical core solves the equations of motion and tracer advection using the global spectral-transform method, referred to as the eulerian spectral dynamical core. Explicit numerical dissipation is added using

Table 1: Kiehl and Williamson (1991) experimental design and global means.

Variable	4.5° (R15)	2.8° (T42)	1.9° (T63)	1.1° (T106)
∇^4 coefficient (m^4/s)	2.0×10^{16}	1.0×10^{16}	2.0×10^{15}	1.0×10^{15}
Δt_{phys} (s)	1800	900	600	360
Total Cloud Fraction	0.47	0.36	0.29	0.26
Total Precipitable Water (mm)	21.1	20.0	18.9	18.8

biharmonic ∇^4 diffusion, and a moisture corrective factor $q - flux$ is used to compensate for clipping of negative water vapor concentrations that arises from oscillatory errors typical of spectral methods. The model consists of 12 vertical levels using a σ coordinate. The parameterization of clouds in CCM1 is broadly similar to Manabe et al. (1965), and depends on lapse rate and grid cell relative humidity. A stable stratiform cloud forms with cloud fraction 0.95 when a grid cell is super-saturated and the lapse-rate is stable. If the lapse rate is steeper than the moist adiabat, the total cloud fraction of the unstable layer is set to 0.3, and the level above the unstable layer is set to 0.95, approximating an anvil cloud top. Radiative transfer is parameterized after Kiehl et al. (1987). A parameterization of momentum flux divergence due to stationary gravity waves is included (McFarlane, 1987).

The convergence study of Kiehl and Williamson (1991) consists of an ensemble of simulated January's. Considering the computational limitations at the time, an impressive range of resolutions are covered in this study (Table 1). The boundary conditions contain real-world topography, and prescribed seasonally varying sea-surface temperatures (SST). The only parameters that vary with resolution are the physics time-step, Δt_{phys} , the standard deviation of the sub-grid topography (for the gravity wave drag parameterization), and the biharmonic diffusion coefficient for numerical dissipation. The coefficients

of biharmonic diffusion were selected to preserve the slope of kinetic energy spectrum (see Boville, 1991). The lowest resolution simulation (R15) statistics are found by averaging the January data from 11 simulated annual cycles, and data from the next highest resolution (T42) are computed from 6 annual cycles. In higher resolution simulations (T63 and T106), three simulations each were initialized from Dec. 15th of the T42 ensembles, and integrated through the month of January.

Table 1 provides the global mean total cloud fraction and total precipitable water from the four simulations. With increasing resolution, the model generally becomes less cloudy and drier. Kiehl and Williamson (1991) show that the magnitude of the upward and downward vertical motion increases with resolution, and argues the greater subsidence is responsible for the drier and less cloudy atmosphere at higher resolutions. In the zonal mean, it was discussed that the width of ascending branch of the Hadley Cell was found to be grid limited at the lower resolutions, containing only 3 – 4 grid cells in the R15 case, but that at the higher resolutions (T63 and T106) the width had converged and reached the “natural scale of the phenomena.”

1.3 CCM2

CCM, version 2 (CCM2) is described in Hack et al. (1993) and the convergence study in Williamson et al. (1995). CCM2 uses the eulerian spectral dynamical core from CCM1, but water vapor is transported using a semi-Lagrangian method (Williamson and Rasch, 1994), since oscillatory errors from the spectral dynamics in CCM1 too often drove water vapor concentrations negative.

Table 2: Williamson et al. (1995) experimental design and global means. The modified parameter experiment pertains to the bottom half of the table.

Variable	3.75° (T31)	2.8° (T42)	1.9° (T63)	1.1° (T106)
∇^4 coefficient (m^4/s)	2.0×10^{16}	1.0×10^{16}	2.0×10^{15}	1.0×10^{15}
Δt_{phys} (s)	450	450	450	450
τ_{conv} (s)	1200	1200	1200	1200
Total Cloud Fraction	0.56	0.54	0.52	0.50
Total Precipitable Water (mm)	24.3	23.9	22.6	22.5
Convective Precipitation (mm/day)	2.62	2.58	2.56	2.55
Stratiform Precipitation (mm/day)	0.83	0.89	0.93	1.08
Δt_{phys} (s)	1200	1200	720	450
τ_{conv} (s)	3600	3600	1800	1200
Total Cloud Fraction	0.55	0.55	0.54	0.55
Total Precipitable Water (mm)	24.0	23.6	22.9	22.4
Convective Precipitation (mm/day)	2.66	2.63	2.65	2.52
Stratiform Precipitation (mm/day)	0.78	0.85	0.90	1.04

CCM2 contains 18 vertical levels, and uses a hybrid- σ vertical coordinate. The physical parameterizations have undergone a major overhaul. CCM2, unlike its predecessor, includes a diurnal cycle and atmospheric absorption of solar radiation. Convection is handled through a mass flux scheme (Hack, 1994). The planetary boundary layer scheme is described in Holtslag and Boville (1993) and includes non-local, up-gradient mixing. The stratiform cloud scheme has added dependencies on vertical motion and static stability, and introduces a height varying relative humidity threshold to distinguish mid-to-high level stratus from marine stratocumulus clouds (Kiehl et al., 1994).

Williamson et al. (1995) is unique in it that consists of two convergence tests, one with fixed cloud parameters across resolutions (“unmodified parameterizations”), and one that contains resolution dependent cloud parameters to minimize changes in total cloud fraction with resolution (“modified parameterizations”). They reasoned that the impact of drying at higher resolution on cloud fraction may be avoided through decreasing the relative humidity thresholds in the cloud parameterization (Kiehl et al., 1994).

Table 2 shows the results of the convergence tests in Williamson et al. (1995) in both the modified and unmodified parameterizations simulations. The means are computed from an ensemble of January's, and therefore directly comparable to the CCM1 study (Kiehl and Williamson, 1991). Note the use of different Δt_{phys} in the two experiments. Through comparing the T106 simulations in the two experiments, one can deduce the impact of tuning, since they use the same Δt_{phys} . While the cloud fractions are different, with the modified case containing more clouds, the total precipitable water, convective and stratiform precipitation rates are very similar.

The cloud fraction decreases with resolution in the unmodified experiments, but is approximately invariant with resolution in the modified case, as intended. Both the modified and unmodified show a similar resolution sensitivity of total precipitable water, convective and stratiform precipitation to resolution. Williamson et al. (1995) chose to preserve a relationship between the Δt_{phys} and the convective time-scale τ_{conv} used in the mass-flux convection scheme $\tau_{conv} \sim 3 \times \Delta t_{phys}$. They discuss that the amount of instability removed in a time-step by the convection scheme is proportional to Δt_{phys} , and any remaining instability is removed instantaneously by the stratiform cloud scheme. This dependency can be counteracted through changing τ_{conv} in proportion, and likely contributes to the similar values of convective precipitation in both the modified and unmodified cases.

Table 3: Williamson (2008) experimental design and global means.

Variable	2.8° (T42)	1.4° (T85)	0.7° (T170)	0.35° (T340)
∇^4 coefficient (m^4/s)	1.0×10^{16}	1.0×10^{15}	1.0×10^{14}	2.25×10^{13}
Δt_{phys} (s)	300	300	300	300
Total Precipitable Water (mm)	20.21	19.63	19.13	18.75
Convective Precipitation (mm/day)	1.71	1.59	1.44	1.36
Stratiform Precipitation (mm/day)	1.11	1.38	1.62	1.75
Δt_{phys} (s)	2400	1200	600	300
Total Precipitable Water (mm)	19.57	19.39	19.18	18.75
Convective Precipitation (mm/day)	1.85	1.76	1.55	1.36
Stratiform Precipitation (mm/day)	0.89	1.17	1.51	1.75

1.4 CAM3

The next version of the NCAR climate model lineage in which a convergence study is available¹ is referred to as the Community Atmosphere Model, version 3 (CAM3) and is described in Collins et al. (2004). Like CCM2, CAM3 uses the eulerian spectral dynamical core with semi-Lagrangian tracer transport and biharmonic diffusion. CAM3 has 26 vertical levels, and uses a hybrid- σ terrain following coordinate. The Hack (1994) convection scheme from CCM2 now only handles shallow convection; the deep convection is parameterized after Zhang and McFarlane (1995), which uses a CAPE closure. The stratiform scheme considers both liquid and ice cloud amounts (Rasch and Kristjánsson, 1998; Zhang et al., 2003), and still contains a height dependent relative humidity threshold. A more detailed treatment of the impact of aerosols on shortwave radiation and microphysics is included (Collins et al., 2004), in addition to a modification to the longwave treatment of water vapor (Collins, 2002).

Breaking with the previous convergence studies, Williamson (2008) uses an aqua-planet configuration ('CONTROL' in Neale and Hoskins, 2000), which is

¹The intervening versions CCM3 and CAM2 don't appear to have published a convergence study, to the authors knowledge.

a planet made up entirely of ocean with a fixed, zonally symmetric SST profile and in perpetual equinox. Without the seasonal cycle, aqua-planets produce equilibrium statistics from a shorter simulation, and the final 12 months of a 14 month simulation are analyzed. Williamson (2008) experiments with a variety of combinations of Δt_{phys} and grid resolution, and only a selection of the simulations are shown in Table 3.

As with the previous convergence studies, the total precipitable water and convective precipitation rate decrease with resolution, stratiform precipitation increases. In contrast to Williamson et al. (1995), τ_{conv} in both the shallow and deep convection schemes are fixed at their default values, regardless of Δt_{phys} . It is interesting to see the dependence of the convection scheme on Δt_{phys} discussed in Williamson et al. (1995) and explored in more detail in Williamson (2013). The convective precipitation rate in the T42 simulations is less using the smaller Δt_{phys} , consistent with the discussed mechanism (Table 3). Interestingly, the stratiform precipitation rate is larger and the total precipitable water indicates the atmosphere is drier, using the smaller Δt_{phys} , which is in the same direction as with increasing resolution. That is, reducing Δt_{phys} while also increasing the resolution increases the sensitivity of the model (Williamson, 2008). In addition, the study finds that reducing the diffusion coefficients with resolution impacts the precipitation rates in the same direction as increasing the resolution, and also contribute to overall resolution sensitivity.

Table 4: Experimental design and global means in (top) Herrington and Reed (2017) and (bottom) Rauscher et al. (2013).

Variable	<i>ne</i> 16 (208.5 km)	<i>ne</i> 30 (111.2 km)	<i>ne</i> 60 (55.6 km)	<i>ne</i> 120 (27.8 km)
∇^4 coefficient (m^4/s)	6.0×10^{15}	9.0×10^{14}	1.0×10^{14}	1.0×10^{13}
Cloud Fraction	0.62	0.57	0.51	0.42
Total Precipitable Water (mm)	20.08	19.65	19.38	19.10
Total Precipitation (mm/day)	2.96	3.05	3.08	3.08
Convective Precipitation (mm/day)	1.38	1.15	0.97	0.79
Stratiform Precipitation (mm/day)	1.58	1.90	2.12	2.29
Variable	240 km	120 km	60 km	30 km
∇^4 coefficient (m^4/s)	5.0×10^{15}	5.0×10^{14}	5.0×10^{13}	5.0×10^{12}
Total Precipitation (mm/day)	2.93	3.04	3.10	3.19
Convective Precipitation (mm/day)	1.40	1.23	1.06	0.90
Stratiform Precipitation (mm/day)	1.53	1.81	2.04	2.29

1.5 CAM4

CAM, version 4 (CAM4) is described in Neale et al. (2010). The major updates to the CAM4 physics are primarily the modifications to the deep convection scheme (Zhang and McFarlane, 1995) —the incorporating of convective momentum transport (Richter and Rasch, 2008) and entrainment into the CAPE trigger/closure (Raymond and Blyth, 1992; Neale et al., 2008)..

The default dynamical core has been changed to a finite-volume model after Lin (2004) (CAM-FV), however there are no convergence studies the author is aware of that uses more than two grid resolutions in CAM-FV. Instead this section will highlight the convergence study Rauscher et al. (2013) using the Model for Prediction Across Scales dynamical core (CAM-MPAS; Ringler et al., 2008) and the study in of Herrington and Reed (2017) (i.e., Chapter 2) using the spectral-element dynamical core (CAM-SE; Taylor and Fournier, 2010; Dennis et al., 2012). All aspects of the CAM-MPAS and CAM-SE convergence experiments are identical except for the dynamical cores since the model data are from the same project (Leung et al., 2013). Both dynam-

ical cores contain explicit ∇^4 hyper-diffusion, with the coefficients shown in Table 4, and Δt_{phys} is fixed at 600 s at all resolutions. The simulations use an aqua-planet configuration, although the SST profile ('QOBS' in Neale and Hoskins, 2000) is different from the CAM3 study (Williamson, 2008). The global means in Table 4 are from the final 4.5 years of a 5 year integration.

In both experiments, the convective precipitation decreases with resolution at the expense of stratiform precipitation. The convective (stratiform) precipitation is more (slightly more) sensitive to resolution in CAM-MPAS compared to CAM-SE. Total cloud fraction and total precipitable water are not available from the CAM-MPAS study, but in CAM-SE, both decrease with resolution.

1.6 CAM5

CAM, version 5 (CAM5) is documented in (Neale et al., 2012). The stratiform scheme has been changed significantly (Park et al., 2014), including stratus-radiation-turbulence interactions that improves the simulation of marine stratocumulus. Stratiform microphysics have been updated to include prognostic number concentrations of liquid and ice (Morrison and Gettelman, 2008), and is coupled with a three mode Modular Aerosol Model (Liu et al., 2012). The shallow convection scheme has been replaced with a CIN based mass flux scheme (Park and Bretherton, 2009), and the planetary boundary layer scheme replaced to incorporate moist turbulence (Bretherton and Park, 2009).

The CAM-FV dynamical core is still the default, but a convergence study using CAM-SE is discussed here. Zarzycki et al. (2014b) perform a conver-

Table 5: Zarzycki et al. (2014b) experimental design and global means.

Variable	<i>ne15</i> (222.4 km)	<i>ne120</i> (27.8 km)
∇^4 coefficient (m^4/s)	1.0×10^{16}	1.0×10^{12}
Total Cloud Fraction	0.63	0.64
Total Precipitable Water (mm)	32.8	31.7
Convective Precipitation (mm/day)	3.66	2.25
Stratiform Precipitation (mm/day)	0.49	1.85

gence study using only two grid resolutions, but was intended to evaluate a variable-resolution configuration whose grid resolution straddled these end-member resolutions. They use an aqua-planet configuration, with the same SST configuration as in Williamson (2008) ('CONTROL' in Neale and Hoskins, 2000), and Δt_{phys} is fixed at 1800 s at all resolutions. The aqua-planets are spun-up for two months, and an additional two years are simulated.

Table 5 shows that global mean statistics in the simulations. Total cloud fraction is remarkably insensitive despite an 8-fold increase in resolution. Convective precipitation decreases in the usual sense with resolution, while stratiform precipitation increases by almost a factor of four. The atmosphere becomes drier with resolution, by about 1 mm water equivalent. Comparing total precipitable water to the CAM3 simulations (Table 3), it is apparent the CAM5 atmosphere contains a lot more moisture, an increase of about 10 mm water equivalent compared to CAM3. The CAM5 convective precipitation rates are also higher than in CAM3, by about 1-2 mm/day.

1.7 CAM6

Chapter 6 is a convergence study using CAM, version 6 physics (CAM6; https://ncar.github.io/CAM/doc/build/html/users_guide/index.html) with the CAM-SE dynamical core, a new dry mass vertical coordinate (Lau-

Table 6: Experimental design and global means from Chapter 6.

Variable	<i>ne20</i>	<i>ne30</i>	<i>ne40</i>	<i>ne60</i>	<i>ne80</i>	<i>ne120</i>
∇^4 coefficient (m^4/s)	1.5×10^{15}	4.0×10^{14}	1.5×10^{14}	4.0×10^{13}	1.5×10^{13}	4.0×10^{12}
Δt_{phys} (s)	2700	1800	1350	900	675	450
Cloud Fraction	0.844	0.835	0.824	0.810	0.804	0.800
Total Precipitable Water (mm)	23.31	23.01	22.62	22.25	21.93	21.72
Convective Precipitation (mm/day)	1.91	1.83	1.68	1.47	1.29	1.08
Stratiform Precipitation (mm/day)	1.26	1.42	1.60	1.85	2.05	2.22

ritzen et al., 2018) and Conservative Semi-Lagrangian Advection Method for tracer advection (CSLAM; Lauritzen et al., 2017). The physics are evaluated on the CSLAM grid (Herrington et al., 2018). The Cloud Layers Unified by Binormals (CLUBB Golaz et al., 2002; Bogenschutz et al., 2013) is an assumed PDF higher order closure model that handles shallow convection, planetary boundary layer mixing and cloud macrophysics. The macrophyics are coupled to a two-moment bulk cloud microphysics scheme with prognostic precipitation (Gettelman et al., 2015). The gravity wave drag scheme has been replaced with an anisotropic gravity wave drag scheme that uses the orientation of topographic ridges to determine the direction of propagation.

Table 6 shows the results of the convergence study in an aqua-planet configuration ('QOBS' SST profile in Neale and Hoskins, 2000), with means computed from the last 11 months of 12 month integrations. Total precipitable water, total cloud fraction and deep convective precipitation rate decreases, while stratiform precipitation increases, monotonically with resolution. Resolution sensitivity in CAM6 is similar to all prior model versions in the CAM-lineage.

Compared to CAM4 (Table 4), the CAM6 atmosphere contains more moisture (an increase of about 3 mm water equivalent) and slightly more convective precipitation. This pattern is similar, although the magnitude of the moisten-

ing and increase of convective precipitation smaller, to the comparison of the CAM3 and CAM5 aqua-planets.

1.8 Key takeaways

A key finding of approximately 30 years of convergence studies of NCAR’s climate model lineage can be summarized as follows. In every convergence study reviewed, and where the data are available:

1. Total precipitable water decreases with resolution.
2. Total cloud fraction decreases with resolution, except in the CCM2 modified parameter experiment and the CAM5 experiment, in which case they are largely insensitive to resolution.
3. Stratiform precipitation rates increase with resolution, except in the CCM2 modified parameter experiment, where the convective precipitation rates increase from T42 to T63.
4. Convective precipitation rates decrease with resolution.

These observations form the general rules of thumb for how the model has historically responded to resolution.

2 An explanation for the sensitivity of the mean state of CAM to horizontal resolution on aqua-planets

2.1 Introduction

One of the greatest challenges to global atmospheric modeling is representing the enormous range of physically relevant spatial scales given present day limits in high performance computing. For whatever the social, economic or political reasons may be, there is a persistent societal force on technological innovation that, has in the past, and will likely continue to result in increased computing power with time. The global atmospheric modeling community exploits advances in high performance computing as an opportunity to develop models that push the limits on the range of explicitly resolved scales, while maintaining reasonable computational throughput.

State-of-the-art General Circulation Models (GCMs) currently run at horizontal resolutions of approximately 100 km to 200 km, with high-resolution configurations of about 25 km to 50 km. A standard measure of model performance in the computational fluid dynamics community is that the numerical solution of a model converges with decreasing grid spacing. As discussed in Williamson (2008), convergence of the resolved fluid dynamical core is often satisfied in dry idealized test cases, but convergence of the moist GCM, including coupling with column physics, receives less attention. With the recent interest in variable resolution grids, and with horizontal resolutions approaching

non-hydrostatic scales, there is increased awareness in the global atmospheric modeling community to develop GCMs with convergent solutions.

The Community Atmosphere Model (CAM), a GCM supported by the National Center for Atmospheric Research and the Department of Energy, has a long history of non-convergence (Kiehl and Williamson, 1991; Williamson et al., 1995; Williamson, 1999, 2008; Li et al., 2011; Reed and Jablonowski, 2011b; Reed et al., 2012; O'Brien et al., 2013; Rauscher et al., 2013; Zarzycki et al., 2014b; Lu et al., 2015). Non-convergence is strong when CAM is coupled with version 4 physics and earlier, but there are indications of non-convergence when coupled with version 5 physics (Reed et al., 2012; Reed and Medeiros, 2016), although not as strong (O'Brien et al., 2013; Zarzycki et al., 2014b). Many studies have found a strong sensitivity of large-scale simulation statistics to horizontal resolution, such as global cloud coverage, precipitation, surface pressure, vertical velocity and Hadley Cell strength (Kiehl and Williamson, 1991; Williamson et al., 1995; Williamson, 2008; Li et al., 2011; Rauscher et al., 2013; O'Brien et al., 2013; Zarzycki et al., 2014b) and more recently, the location of the eddy-driven jet (Lu et al., 2015). Although intuitively we may expect some of the variations in these large-scale statistics to be related, these studies have not made clear what relationship occurs in the simulations.

Here, we study the sensitivity of CAM to horizontal resolution in an aqua-planet configuration, identify relationships between non-converging large-scale statistics using energy and moisture budgets and propose a possible explanation for the resolution sensitivity of the mean state. Other studies have focused on the convergence of CAM with respect to statistical extremes, such as precip-

itation extremes (Williamson, 2008; Li et al., 2011; Yang et al., 2014; O'Brien et al., 2016) or tropical cyclones (Reed and Jablonowski, 2011a). Although the causes of non-convergence of mean and extreme statistics may be related, we primarily focus on the mean state in this study.

The non-convergence of CAM with increasing horizontal resolution has been attributed to the representation of moist processes (Williamson, 1999, 2008; O'Brien et al., 2013). Studies on the convergence of CAM with respect to the model time-step have come to similar conclusions (Williamson and Olson, 2003; Williamson, 2008, 2013; Wan et al., 2015). Williamson (2013) argues that since convective adjustment in CAM versions 4 and earlier is limited by fixed relaxation timescales in the closure assumptions, a greater proportion of super-saturated air remains in a grid column as the time-step is reduced. Due to the sequential coupling of the moist physics, any super-saturated air remaining after the convection scheme has been called upon is condensed locally by the cloud macrophysics scheme, resulting in an increasingly buoyant state passed to the dynamical core as the time-step is reduced. Although the inconsistency of using fixed relaxation time-scales in the convection scheme is problematic, others have shown that the strong sensitivity of CAM to horizontal resolution remains when the time-step is held fixed (Williamson, 2008; O'Brien et al., 2013; Rauscher et al., 2013) or the convection scheme is completely turned off (O'Brien et al., 2013).

It has been recognized that the resolved vertical motion is participating in the sensitivity of precipitation extremes to resolution in CAM (Li et al., 2011; Yang et al., 2014). Rauscher et al. (2016) apply the mass continuity

equation to the spectral properties of the horizontal wind in a suite of regional models to argue that resolved vertical velocity is expected to increase with resolution. The authors go on to formulate a heuristic scaling that linearly relates the resolved mass fluxes at cloud base to the precipitation rate in models, providing an explanation for the sensitivity of precipitation extremes to resolution. O'Brien et al. (2016) show that the relationship of Rauscher et al. (2016) provides an excellent fit to the precipitation rates in CAM. While the insights provided in Rauscher et al. (2016) are valuable, there is no simple explanation for the spectral slope of the horizontal wind that can inform us on what processes are responsible for the increase in resolved vertical velocity with resolution in the models.

Here we analyze the dominant balances in the large-scale circulation's response to varying horizontal resolution, and propose a possible explanation for the sensitivity of CAM to horizontal resolution at hydrostatic scales. We hypothesize that an increase in horizontal resolution leads to a reduction in horizontal scale of the diabatic forcing arising from the column physics, facilitating fine scale flow and faster resolved convective updrafts within the dynamical core, and steering the coupled system towards a new mean state. The paper is organized as follows. Section 2.2 provides an overview of the model and experimental design. In Section 2.3, an analysis of the mean state utilizing energy and moisture budgets is presented. In Section 2.4, we provide an interpretation of relationships between non-converging statistics and articulate our explanation for the sensitivity of the mean state to horizontal resolution. Section 2.5 presents our conclusions.

2.2 Methods

We use model output from the multi-institutional project “Development of Frameworks for Robust Regional Climate Modeling” aimed at improving simulations of climate at the regional scale (Leung et al., 2013). One of the goals of the project is to determine the sensitivity of global atmospheric models to quasi-uniform increases in horizontal resolution to evaluate their ability to support variable resolution meshes. A strong dependency of simulation statistics on horizontal resolution would yield poor results on a variable resolution grid, as solutions would diverge across mesh transitions. Here, we focus on the behavior of a single GCM, CAM, to variations in quasi-uniform mesh spacing.

2.2.1 Model Description

High-Order Multiscale Modeling Environment (HOMME; now commonly referred to as CAM-SE) refers to the spectral element dynamical core option in CAM version 5.0, the atmospheric component of the Community Earth System Model (CESM) version 1.0. A full documentation of HOMME is given in Neale et al. (2012) and only a brief overview is provided here. We chose the HOMME dynamical core since the spectral element method and its variable resolution capabilities (Zarzycki et al., 2014b) shows promise for more widespread use in future generation atmospheric models. HOMME was designed for massively parallel systems and demonstrates perfect strong scaling up to one element per processing core (Dennis et al., 2012).

HOMME solves the vector-invariant form of the horizontal momentum equations using a locally conservative, fourth-order accurate continuous-Galerkin

method on a quasi-uniform cubed-sphere mesh (Taylor and Fournier, 2010; Dennis et al., 2012). Explicit dissipation is applied to horizontal momentum and tracer advection using fourth-order hyper-viscous damping and enhanced second-order damping in a ‘sponge-layer’ near the model top. HOMME is a hydrostatic model and utilizes a hybrid terrain following, pressure based vertical coordinate. HOMME offers several options for the vertical discretization and time stepping. For the simulations analyzed in this paper, second-order finite differences are used in the vertical (Simmons and Burridge, 1981) and the vertical Lagrangian remap method of Lin (2004) is used for tracers to enforce monotonicity. A second-order two-stage Runge-Kutta time-stepping scheme is used to advance the dynamics, and the tracers evolve using a leap-frog scheme (Taylor and Fournier, 2010; Dennis et al., 2012).

HOMME is coupled to the CAM version 4 physics package (CAM4). Full descriptions of the calculations of moist processes, radiation, surface fluxes and turbulent mixing is documented in Neale et al. (2010). Briefly, CAM4 parameterizes deep convection using the mass flux model of Zhang and McFarlane (1995), modified by a dilute plume calculation after Raymond and Blyth (1992) and parameterized convective momentum transport from Richter and Rasch (2008). The mass-flux model of Hack (1994) is used for shallow and mid-level convection and prognostic cloud macrophysics and cloud microphysics are from Rasch and Kristjánsson (1998) and Zhang et al. (2003). Surface fluxes and a nonlocal turbulent mixing scheme for the planetary boundary layer are described in Holtslag and Boville (1993). Total precipitation rate in the model is the sum of contributions from the convection schemes and the combined effect

of the cloud macrophysics and microphysics schemes. Although sometimes referred to as stratiform precipitation in the GCM literature, the precipitation rate resulting from the macrophysics and microphysics schemes is referred to as the large-scale precipitation rate throughout the manuscript.

The physics routines are coupled to each other using a sequential splitting method. The dynamics and tracers are subcycled within a (usually) longer physics time-step, and the column physics are coupled to the dynamical core using a time-split approach (Neale et al., 2012). A fraction of the physics tendencies are applied during each dynamics time-step, such that the full physics forcing is realized over the longer physics time-step. The consequences of ‘dribbling’ the physics tendencies into the dynamical core can result in artificial energy sinks (P. Lauritzen, personal communication), but reduces the occurrence of spectral ringing in HOMME (Thatcher and Jablonowski, 2016).

2.2.2 Simulation Design

The model is run in aqua-planet configuration with a fixed, zonally symmetric sea surface temperature profile as the lower boundary condition (“Control” in Neale and Hoskins (2000)). The model top is forced with prescribed solar forcing equivalent to a diurnally varying perpetual equinox. All simulations use 26 hybrid η vertical levels, and were performed at four different horizontal resolutions of 16 (ne16), 30 (ne30), 60 (ne60) and 120 (ne120) elements along the edge of a cubed sphere face, which corresponds to a quasi-uniform grid spacing of approximately 220 km, 110 km, 55 km and 28 km, respectively. The physics time-step is fixed at 600 seconds for all simulations, while the

dynamics and tracer time-steps vary with resolution in proportion to their respective Courant numbers. The hyper-viscosity coefficients are, respectively, 6×10^{15} , 9×10^{14} , 1×10^{14} and $1 \times 10^{13} \text{ m}^4/\text{s}$ in the ne16, ne30, ne60 and ne120 simulations. The hyper-viscosity coefficients are experimentally determined by model developers (e.g., Boville, 1991), decreasing by about an order of magnitude for a halving of the grid spacing. Each simulation was run for five years. The first six months contain the spin-up period and have been omitted from all analysis in this study.

The novelty of these simulations is that only two parameters, the hyper-viscosity coefficient and the dynamics/tracers time-step, are varied as the horizontal resolution is varied. Further, the use of an aqua-planet configuration minimizes any resolution dependence of the boundary conditions, such as steep continental terrain associated with a more realistic geography. Since the physics time-step is fixed, the simulations are free of variations that might result from time-step-sensitive processes in the column physics (Wan et al., 2015), such as rate-limited convective adjustment (Williamson, 2013).

At 600 seconds, the physics time-step is relatively short. From Williamson (2013), we may expect the convective scheme to be inefficient at this time-step, resulting in a greater proportion of convective instabilities removed by the dynamical core in our simulations. It is important to distinguish our specific model configuration from scientifically validated configurations of CAM4, where the convection scheme is not restricted through the use of a longer physics time-step.

2.2.3 Convergence Metrics

To facilitate comparison across resolutions, we have adopted the convention of constructing anomalies relative to the lowest-resolution simulation (ne16). For any generic variable, convergence of that variable is implied if its anomaly converges to a finite value as the horizontal resolution is increased. In order to compare the spatial distribution of a particular field across different model resolutions, the ne30, ne60 and ne120 simulations are mapped to the ne16 grid using an area conserving interpolation.

Our convention contrasts with typical convergence tests where anomalies, or errors, are defined relative to a high-resolution reference simulation (Jablonowski and Williamson, 2006; Williamson, 2008). The high-resolution simulations (ne60 and ne120) are ‘experimental’ in the sense that the column physics in these simulations were originally tuned for a lower resolution. The mean state in these high-resolution simulations is unverified, or even degraded relative to lower-resolution scientifically validated versions of CAM (Wehner et al., 2014). However, experimental high-resolution simulations appear to produce more realistic tropical cyclones (Wehner et al., 2014) and are more skillful at simulating precipitation extremes (Wehner et al., 2014; O’Brien et al., 2016).

2.2.4 Isentropic Coordinates

Cloud parameterizations in CAM are strongly tied to the relative humidity field (Neale et al., 2010, 2012; Park et al., 2014), which can be understood using a moisture balance framework. We choose to work in isentropic coordi-

nates because they separate along and cross-isentropic mixing, which approximate different mechanisms controlling the relative humidity of the atmosphere (Schneider et al., 2006). The 26 hybrid η vertical levels were linearly interpolated to 120 isentropic surfaces between 210 K and 600 K, equally spaced in $\theta^{-1/\kappa}$, where θ is the potential temperature and κ the adiabatic exponent, which approximate equally spaced pressure levels in an isothermal atmosphere (Schneider et al., 2006).

As a starting point for our analysis of moisture transport, we utilize the isentropic meridional mass streamfunction (Held and Schneider, 1999; Schneider et al., 2006),

$$\psi(\phi, \theta) = 2\pi a \cos(\phi) \int_{\theta_b}^{\theta} \overline{\rho_\theta v^*} d\theta, \quad (1)$$

where a is the radius of Earth, v is the meridional velocity and ρ_θ is the isentropic density, $\rho_\theta = -g^{-1} \partial_\theta p \mathcal{H}(\theta - \theta_s)$ with the gravitational acceleration g , and pressure p . The Heavyside step function $\mathcal{H}(\theta - \theta_s)$ forces the isentropic density to zero on isentropes less than the instantaneous surface potential temperature, θ_s . The term $\overline{(x)}^* = \overline{(\rho_\theta x)} / \overline{\rho_\theta}$ is a Favre-filter or density-weighted mean of x , with overbars indicating a temporally and zonally averaged mean along isentropic surfaces.

Table 7: Time-mean global integrals in the simulations. Differences with respect to the ne16 simulation are indicated with a $\Delta(\cdot)$ and the standard deviations of the monthly mean anomalies are given in parentheses.

	ne16	ne30	ne60	ne120
Temperature (K)	247.10	247.49	247.54	247.37
$\Delta(\cdot)$	—	0.39 (0.25)	0.43 (0.23)	0.27 (0.24)
Cloud fraction	0.62	0.57	0.51	0.42
$\Delta(\cdot)$	—	-0.04 (0.01)	-0.11 (0.01)	-0.19 (0.01)
Total precipitable water (mm)	20.08	19.65	19.38	19.10
$\Delta(\cdot)$	—	-0.43 (0.19)	-0.70 (0.21)	-0.98 (0.22)
Total precipitation (mm day^{-1})	2.96	3.05	3.08	3.08
$\Delta(\cdot)$	—	0.09 (0.07)	0.12 (0.06)	0.12 (0.06)
Large-scale precipitation (mm day^{-1})	1.58	1.90	2.12	2.29
$\Delta(\cdot)$	—	0.32 (0.06)	0.54 (0.06)	0.71 (0.05)
Convective precipitation (mm day^{-1})	1.38	1.15	0.97	0.79
$\Delta(\cdot)$	—	-0.23 (0.02)	-0.41 (0.02)	-0.59 (0.02)

2.3 Results

2.3.1 Global Sensitivity

Some globally integrated quantities, including their anomalies from the ne16 control are displayed in Table 7 as climatological means over the final 4.5 years of the simulations. The standard deviation of the anomalies, computed from monthly means, is an indication of the significance of a particular anomaly associated with low frequency variability in the model.

Global temperatures are relatively invariant to resolution (Table 7), partially a reflection of the fixed sea surface temperatures in the aqua-planet configuration. The atmosphere becomes progressively drier with increasing resolution. Global mean total precipitable water is reduced by 0.98 ± 0.22 mm, about 5% at the highest simulated resolution (ne120) relative to the control. The progressive drying trend is robust, since the anomalies in total precipitable water exceed two standard deviations (Table 7). Global average cloud fraction is also progressively reduced with increasing resolution (Ta-

ble 7). The magnitude of the trend in cloud fraction is drastic, decreasing by approximately a third, or 0.19 ± 0.01 at the highest resolution relative to the control. The reduction in cloud fraction is split evenly between convective clouds and clouds associated with large-scale precipitation, at all resolutions (not shown). The high sensitivity of cloud fraction to resolution is well known in CAM4 (O'Brien et al., 2013; Rauscher et al., 2013; Zarzycki et al., 2014b) and earlier versions (Kiehl and Williamson, 1991; Williamson, 2008), but there is little evidence for this sensitivity in the default physics package in CAM, version 5 (CAM5; Zarzycki et al., 2014b).

Global mean total precipitation increases slightly with resolution, but appears to converge to 3.08 mm/day at the two highest resolutions. Total precipitation is the sum of the convective precipitation and the large-scale precipitation. The trends in the two forms of precipitation with resolution are progressive, large and compensating (Table 7). The global mean large-scale precipitation (convective precipitation) increases (decreases) by 0.71 ± 0.05 mm/day (0.59 ± 0.02 mm/day) in the highest resolution simulation. The increase (decrease) of the large-scale (convective) precipitation rate with increasing horizontal resolution has previously been documented in CAM5 (Zarzycki et al., 2014b), CAM4 (O'Brien et al., 2013; Rauscher et al., 2013; Zarzycki et al., 2014b) and earlier versions (Williamson et al., 1995; Williamson, 2008).

Changes to the hydrologic cycle with resolution may be expected to affect the flows of energy determining the global atmospheric energy budget. Energy fluxes at the interface of the model with the surface, $f(\eta_{SFC})$, and the top of the model, $f(\eta_{TOP})$, are provided by the model and used to compute global

energy budgets for the surface and atmosphere. A generic flux, f , may be integrated around the domain of the model to compute the globally integrated energy tendency due to the divergence of the flux. We use the convention that positive values indicate a gain of energy to the system, and positive fluxes are directed upward.

The surface and top of atmosphere energy budget, based on standard equations (Peixoto and Oort, 1992), are presented in Table 8. The large top of atmosphere energy imbalance is taken up almost entirely by the surface energy imbalance, corresponding to a net gain of energy to the surface on the order of 20 W/m² in all the simulations (Table 8). The large sink of energy to the surface is a consequence of the fixed lower boundary condition, which prevents the surface fluxes from adjusting to bring the surface into an energy balance. The globally integrated total atmospheric energy tendency per unit area, F_{TOT} , has only a slight imbalance of about -1 W/m² in all the simulations, which is roughly the same magnitude as the low frequency variability (Table 8) and together indicate the atmosphere is in a near balance on monthly timescales throughout the simulations.

The global atmospheric energy budget per unit area may be expressed as,

$$F_{TOT} = F_{CS} + F_{CD} + F_S + F_{SH} + L(P_L + P_C), \quad (2)$$

where the energy gain due to the divergence of the longwave radiative fluxes are separated into clear sky and cloud components (F_{CS} and F_{CD} , respectively), such that their sum is the all-sky longwave term. Each of the longwave terms

Table 8: Global energy budgets in the simulations.

	ne16	ne30	ne60	ne120				
Summary of energy budgets (W m^{-2} , +gain of energy)								
Top of atmosphere	15.82	(1.08)	15.96	(1.32)	17.49	(1.16)	20.83	(0.77)
Surface	16.83	(1.73)	17.05	(1.63)	18.70	(1.56)	22.09	(1.45)
Atmosphere, F_{TOT}	-1.01	(2.04)	-1.09	(2.1)	-1.22	(1.95)	-1.26	(1.64)
ΔF_{TOT}	—	—	-0.08	(2.93)	-0.21	(2.82)	-0.25	(2.61)
Top of atmosphere net longwave flux (W m^{-2} , +upward)								
$f_{\text{CS}}(\eta_{\text{TOP}})$	257.63	—	259.19	—	259.70	—	259.59	—
$\Delta f_{\text{CS}}(\eta_{\text{TOP}})$	—	—	1.55	(0.6)	2.07	(0.57)	1.95	(0.53)
$f_{\text{CD}}(\eta_{\text{TOP}})$	-35.02	—	-30.87	—	-26.90	—	-22.18	—
$\Delta f_{\text{CD}}(\eta_{\text{TOP}})$	—	—	4.16	(1.35)	8.12	(1.21)	12.85	(1.02)
Surface net longwave flux (W m^{-2} , +upward)								
$f_{\text{CS}}(\eta_{\text{SFC}})$	90.92	—	91.51	—	92.06	—	92.72	—
$\Delta f_{\text{CS}}(\eta_{\text{SFC}})$	—	—	0.60	(0.34)	1.15	(0.35)	1.80	(0.39)
$f_{\text{CD}}(\eta_{\text{SFC}})$	-33.53	—	-31.32	—	-28.10	—	-23.79	—
$\Delta f_{\text{CD}}(\eta_{\text{SFC}})$	—	—	2.21	(0.77)	5.43	(0.79)	9.74	(0.8)
Longwave heating rate (W m^{-2} , +gain of energy)								
F_{CS}	166.72	—	167.67	—	167.64	—	166.87	—
ΔF_{CS}	—	—	-0.95	(0.68)	-0.92	(0.67)	-0.15	(0.66)
F_{CD}	1.50	—	-0.45	—	-1.20	—	-1.61	—
ΔF_{CD}	—	—	-1.95	(1.56)	-2.70	(1.44)	-3.11	(1.29)
Shortwave heating rate (W m^{-2} , +gain of energy)								
F_S	70.26	—	69.61	—	68.94	—	68.03	—
ΔF_S	—	—	-0.66	(1.81)	-1.32	(1.78)	-2.24	(1.53)
Surface sensible heat (W m^{-2} , +gain of energy)								
F_{SH}	8.95	—	9.04	—	9.31	—	9.87	—
ΔF_{SH}	—	—	0.09	(0.37)	0.36	(0.38)	0.91	(0.38)
Surface latent heat (W m^{-2} , +gain of energy)								
LP_L	45.78	—	55.08	—	61.32	—	66.31	—
$L\Delta P_L$	—	—	9.30	(1.86)	15.54	(1.7)	20.53	(1.67)
LP_C	39.84	—	33.32	—	27.96	—	22.89	—
$L\Delta P_C$	—	—	-6.52	(0.7)	-11.87	(0.66)	-16.94	(0.61)

are related to the fluxes at the surface and top of atmosphere as $F = f(\eta_{\text{SFC}}) - f(\eta_{\text{TOP}})$. Note that F_S is the gain in energy associated with atmospheric absorption of solar radiation, F_{SH} is the surface sensible heat term and $L(P_L + P_C)$ is the surface latent heat term, expressed as the latent heat of condensation L multiplied by the sum of the global mean large-scale precipitation rate P_L and the global mean convective precipitation rate P_C , which are provided in Table 7.

Clear-sky longwave radiative fluxes are on the order of 1-2 W/m² higher than in the control simulation, and the trend is not monotonic with resolution.

The clear-sky longwave fluxes at the surface are monotonic with resolution, increasing by up to 1.8 ± 0.39 W/m² in the ne120 simulation. The trend in clear-sky surface fluxes indicate a reduction in downward longwave fluxes at the surface as they are sourced from higher and colder layers, consistent with increased atmospheric transmissivity in a drier atmosphere. The trends in clear sky fluxes at the top and bottom of the atmosphere compensate for one another, yielding a weak sensitivity of the clear sky longwave radiative heating rate to horizontal resolution (Table 8).

The longwave cloudy fluxes at the top of atmosphere increase monotonically with resolution by up to 12.85 ± 1.02 W/m² in the ne120 simulation relative to the control. The large magnitude changes in longwave cloud fluxes are consistent with the large reduction in global cloud fraction in Table 7. At the surface, longwave fluxes due to clouds also progressively increase with resolution, but by a smaller amount than the top of atmosphere fluxes (Table 8). The surface and top of atmosphere fluxes combine to yield a longwave effect of clouds on the atmospheric energy budget that is two orders of magnitude less than the energy tendency associated with clear-sky longwave radiation. However, the magnitude of the change in the cloud forcing term with horizontal resolution is comparable to the trend in the clear-sky term, increasing progressively with resolution by up to 3.11 ± 1.29 W/m² in the highest resolution simulation relative to the control.

The largest changes to the atmospheric energy budget with resolution occur within the latent heating term. The monotonic increase in latent heating associated with the trend in large-scale precipitation rate is large, up to a

20.53 ± 1.67 W/m² increase in the highest resolution simulation relative to the control (Table 8). These large magnitude changes to the atmospheric energy budget are mostly balanced by reductions in the global convective precipitation rate. Sensible heating is about an order of magnitude smaller than latent heating in the simulations, and are relatively insensitive to changes in resolution (Table 8).

Atmospheric absorption of solar radiation decreases with resolution by 2.24 ± 1.53 W/m² in the ne120 simulation relative to the control. Decreased shortwave absorption is due to greater transmissivity from a drier atmosphere, despite increased availability of shortwave radiation owing to large reductions in cloud fraction.

2.3.2 Regional Sensitivity

An energy budget may be rewritten for zonal columns of energy with infinitesimally thin meridional thickness. Expressed as a zonal average, the atmospheric energy budget differs from the global budget (equation 2) by a transport term (Peixoto and Oort, 1992),

$$H(\phi) = \nabla \int \bar{\mathbf{V}} s + \frac{d}{dp} \int \bar{\omega s}, \quad (3)$$

where H is the column integrated divergence of the flux of dry static energy, $s = c_p T + gZ$, for an infinitesimally thin zonal slice at latitude ϕ ; $\mathbf{V} = (u, v)$ is the horizontal wind vector, ω is the vertical pressure velocity, ∇ is the horizontal divergence operator and the terms c_p , T and Z are the specific heat

capacity of dry air at constant pressure, air temperature and geopotential height, respectively. The integral refers to a mass weighted vertical integral over the atmospheric column in pressure coordinates ($\int = \int dp/g$) where p is pressure. The overbar notation indicates a zonally and temporally averaged quantity.

Through evaluating the response of CMIP3 models to global warming emission scenarios, Muller and O’Gorman (2011) have shown that regional changes to the zonal average energy budget are well approximated by the change in the column integrated divergence of the flux of s by the mean vertical circulation, ΔH_m . This term can be decomposed into dynamic, ΔH_{DYN} , and thermodynamic components, ΔH_{THM} , as,

$$\Delta H_m = \Delta H_{DYN} + \Delta H_{THM} = \int \Delta(\bar{\omega}) \frac{d\bar{s}}{dp} + \int \bar{\omega} \Delta\left(\frac{d\bar{s}}{dp}\right), \quad (4)$$

Making the assumption $\Delta H \sim \Delta H_m$, the change in the zonally averaged atmospheric energy budget between two simulations may be expressed as,

$$\Delta F_{TOT}(\phi) = L[\Delta P_L(\phi) + \Delta P_C(\phi)] + F_{DC}(\phi) + \Delta H_{DYN}(\phi) + \Delta H_{THM}(\phi), \quad (5)$$

where $F_{DC} = F_S + F_{CL} + F_{CD} + F_{SH}$ contains the diabatic cooling terms. Figure 1a,b shows the zonal average latent heat anomalies associated with changes in convective and large-scale precipitation relative to the ne16 simulation. The overall reduction in convective precipitation manifests as a strong, monotonic reduction in latent heating with resolution at the equator, and to a lesser extent in the subtropics. The large-scale precipitation more than compensates

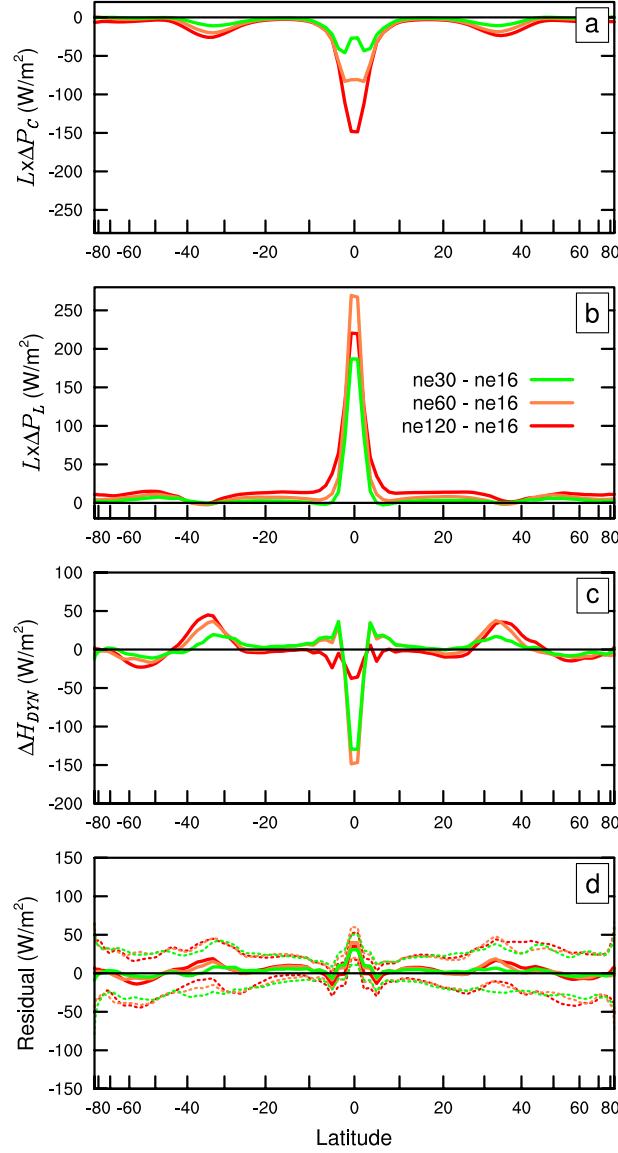


Figure 1: Departure of the components of the zonal average energy budget from the ne16 simulation. Anomalies in latent heat associated with (a) convective precipitation and (b) large-scale precipitation, (c) the dynamical component of the anomalous divergence of the flux of dry static energy by the mean vertical circulation and (d) the residual computed from the sum of (a), (b) and (c). Dotted lines refer to twice the standard deviation associated with low frequency (monthly) variability.

for the relative loss of energy due to reductions in convective precipitation at the equator. The anomalous heating due to changes in large-scale precipitation in the ne120 simulation is intermediate to the ne30 and ne60 anomalies at the equator. While the maximum anomalies at the equator are not monotonic with resolution, the width of the equatorial anomaly does become progressively wider with resolution and there is a slight increase in heating with resolution almost everywhere outside the equator.

Figure 1c shows the dynamical component, ΔH_{DYN} of the divergence of the flux of s , due to the change in the mean vertical circulation. The value of ΔH_{DYN} becomes progressively more negative at the equator in the ne30 and ne60 simulations, and positive anomalies persist just outside the equator. In contrast, the ne120 simulation remains negative just outside the equator and the magnitude of the equatorial minimum in ΔH_{DYN} is less than in the ne30 or ne60 simulations. Near the subtropics ΔH_{DYN} becomes increasingly positive with resolution (Figure 1c), indicating enhanced subsidence. Anomalous heating from greater subsiding motion is primarily balanced by reductions in subtropical convective precipitation (Figure 1).

Qualitatively, the spatial distributions of latent heat anomalies are of a similar order of magnitude as ΔH_{DYN} and appear compensatory. Figure 1d shows the sum of the three terms $L(\Delta P_L + \Delta P_C) + \Delta H_{DYN}$, which may be interpreted as the latent heating anomalies that are not balanced by ΔH_{DYN} . The sum of these three terms is generally small compared to the individual terms, suggestive of a first-order balance of the anomalous energy.

The relative change in diabatic cooling, ΔF_{DC} , although important in the

global mean (Table 8), has very little influence on large regional changes to the energy budget observed in the simulations (not shown). This result is consistent with the response of CMIP3 models to global warming emissions scenarios (Muller and O’Gorman 2011). In contrast to Muller and O’Gorman (2011), ΔH_{THM} is also small and shows minimal variation with latitude (not shown). The small contribution of ΔH_{THM} to the anomalous energy budget indicates that the distribution of s in the simulations is not very sensitive to resolution, partially a reflection of the fixed sea surface temperatures in the aqua-planet configuration. The remaining heating anomalies in Figure 1d may be balanced by neglected transport terms or energy storage.

Interestingly, the residual heating appears to converge to a similar zonal distribution by ne30 (Figure 1d). While the source of this residual may be uncertain, it indicates that the sensitivity of the zonal average energy budget to resolution in the ne30, ne60 and ne120 simulations is contained within the latent heating and ΔH_{DYN} terms. As ΔH_{DYN} is proportional to the change in the resolved vertical motion, our results indicate that an active vertical velocity field is providing a balance to the large variations in latent heating observed in the model. Yang et al. (2014) come to a similar conclusion through analysis of the moisture budget for extreme precipitation events, which together indicate that important changes to the resolved vertical velocity field are common to the sensitivity of mean and extreme statistics with resolution.

Changes to the zonal-mean time-mean resolved upward motion near the equator, along with increases in subtropical subsidence, implies a change in the large-scale Hadley Circulation with resolution. Figure 2b-d shows the

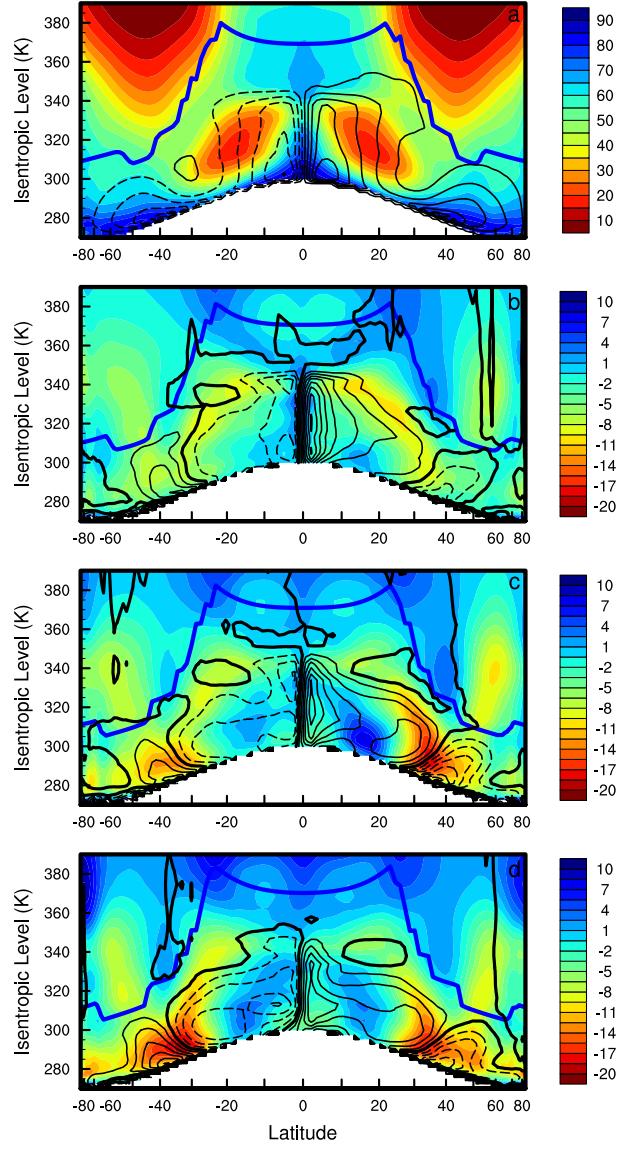


Figure 2: (a) Zonally averaged isentropic meridional streamfunction (black lines) and relative humidity field (color) in the ne16 control. Contour interval is $50 \times 10^9 \text{ kg/s}$. (b-d) Departures of the zonally averaged isentropic meridional streamfunction and relative humidity field from the ne16 control in the (b) ne30 (c) ne60 and (d) ne120 simulations. Contour interval is $10 \times 10^9 \text{ kg/s}$. Solid (dashed) lines correspond to clockwise (counter-clockwise) motion and thick solid line corresponds to the zero contour. Blue solid line is the tropopause height computed using a lapse-rate WMO definition.

departure of the mean isentropic meridional mass streamfunction in the simulations from the mean in the control simulation (Figure 2a). At the equator, an anomalous overturning cell extending into the upper-troposphere occurs in the ne30 simulation (Figure 2b). The descending branch of the anomalous equatorial circulation is split between a local branch, confined to within 10° of the equator, and a poleward traveling branch, flowing under subtropical jet core (not shown) and extending to about 40° latitude (Figure 2b). The anomalous poleward branch is unique in the sense that the mean Hadley Cell only extends to the latitude of the subtropical jet (Figure 2a), which remains in a fixed location in the simulations (Lu et al., 2015). At higher resolutions, the local descending branch becomes weaker and more confined in the meridional direction, while the poleward branch becomes more pronounced, merging with anomalous upward motion around 20° to form a partially closed anomalous circulation extending to 40° latitude (Figures 2c,d). The anomalous ascent near 20° is consistent with the anomalous off equatorial heating in Figure 1b. While generally the Hadley Cell tends to intensify with resolution, the spatial complexities of the anomalies indicate that changes in the Hadley Cell are not a simple uniform increase in intensity of the mean overturning cell depicted in Figure 2a.

2.3.3 Moisture Budget

Changes in the distribution and strength of resolved mass transport in the atmosphere have profound impacts on the distribution of water vapor, which directly influences the cloud amount diagnosed by CAM. Figure 2b-d shows

variations in zonal mean relative humidity relative to the ne16 simulation (Figure 2a). Despite increases in relative humidity in and above the upper-troposphere, and within the subtropical dry zone, there is a clear tendency towards a drier atmosphere with resolution. The reduction in global average cloud fraction with resolution (Table 7) is directly related to the reductions in relative humidity, since the macrophysics scheme computes the cloud fraction as a simple function of the relative humidity field in CAM4 (Neale et al., 2010; Park et al., 2014). The greatest reduction in clouds occurs at the overlap between where clouds are abundant and regions that experience the greatest amount of drying (not shown).

The most pronounced changes in moisture in the simulations are the progressive reductions in subtropical relative humidity with resolution in the region of elevated subsidence identified in Figure 1c. The maximum reduction in relative humidity occurs in the ne120 simulation, decreasing by a third or 20% at the 290 K level in the vicinity of 30° latitude and leading to the complete elimination of clouds in this region (not shown). The drying anomalies are aligned with the anomalous poleward branch, which descends into the region of maximum drying (Figures 2b-d). Drying anomalies also coincide with an anomalous extra-tropical circulation, aligned with an equator-ward subsiding branch that becomes stronger with increasing resolution (Figure 2b-d).

To unravel the role of different processes influencing the relative humidity changes, Figure 3 shows some components of the time-mean zonal averaged moisture budget in isentropic coordinates after Schneider et al. (2006), for the northern hemisphere only. The isentropic moisture budget naturally separates

along and cross-isentropic mixing, which approximate different mechanisms controlling the relative humidity of the atmosphere (Schneider et al., 2006). The descending branch of the Hadley Cell is a cross-isentropic process that exports dry air from upper levels to lower levels of the troposphere (Pierrehumbert, 1999). The moisture budget of the subtropics is also influenced by extra-tropical recirculation, which is an approximately adiabatic process (Galewsky et al., 2005; Schneider et al., 2006). Extra-tropical recirculation refers to moisture advected out of the subtropics, condensing along poleward trajectories and re-circulating back into the subtropics (Galewsky et al., 2005).

The zonal averaged moisture balance in isentropic coordinates is (Schneider et al., 2006),

$$\partial_t(\overline{\rho_\theta} \overline{q}^*) + \partial_y(\overline{\rho_\theta} \overline{v}^* \overline{q}^*) + \partial_\theta(\overline{\rho_\theta} \overline{Q}^* \overline{q}^*) + \partial_y(\overline{\rho_\theta} \overline{v'} \overline{q'}) + \partial_\theta(\overline{\rho_\theta} \overline{Q'} \overline{q'}) = \overline{\rho_\theta} \overline{S}^*, \quad (6)$$

where q is the specific humidity, $S = \frac{dq}{dt}$ and $Q = \frac{d\theta}{dt}$ are the specific humidity and diabatic forcing terms from the column physics. Primes denote fluctuations about the zonal and temporally averaged mean. The zonal average moisture tendency in isentropic coordinates reflects a balance between moisture sources, $\overline{\rho_\theta} \overline{S}^*$, and the divergence of mean and eddy fluxes of moisture along and across isentropic surfaces.

Figure 3b,d,f show the departure of the mean cross-isentrope, $\partial_\theta(\overline{\rho_\theta} \overline{Q}^* \overline{q}^*)$, mean along-isentrope, $\partial_y(\overline{\rho_\theta} \overline{v}^* \overline{q}^*)$, and eddy along-isentrope, $\partial_y(\overline{\rho_\theta} \overline{v'} \overline{q'})$, terms in the ne120 simulation relative to the ne16 simulation values, given in Figure 3a,c,e. The remaining terms of moisture budget could not be computed di-

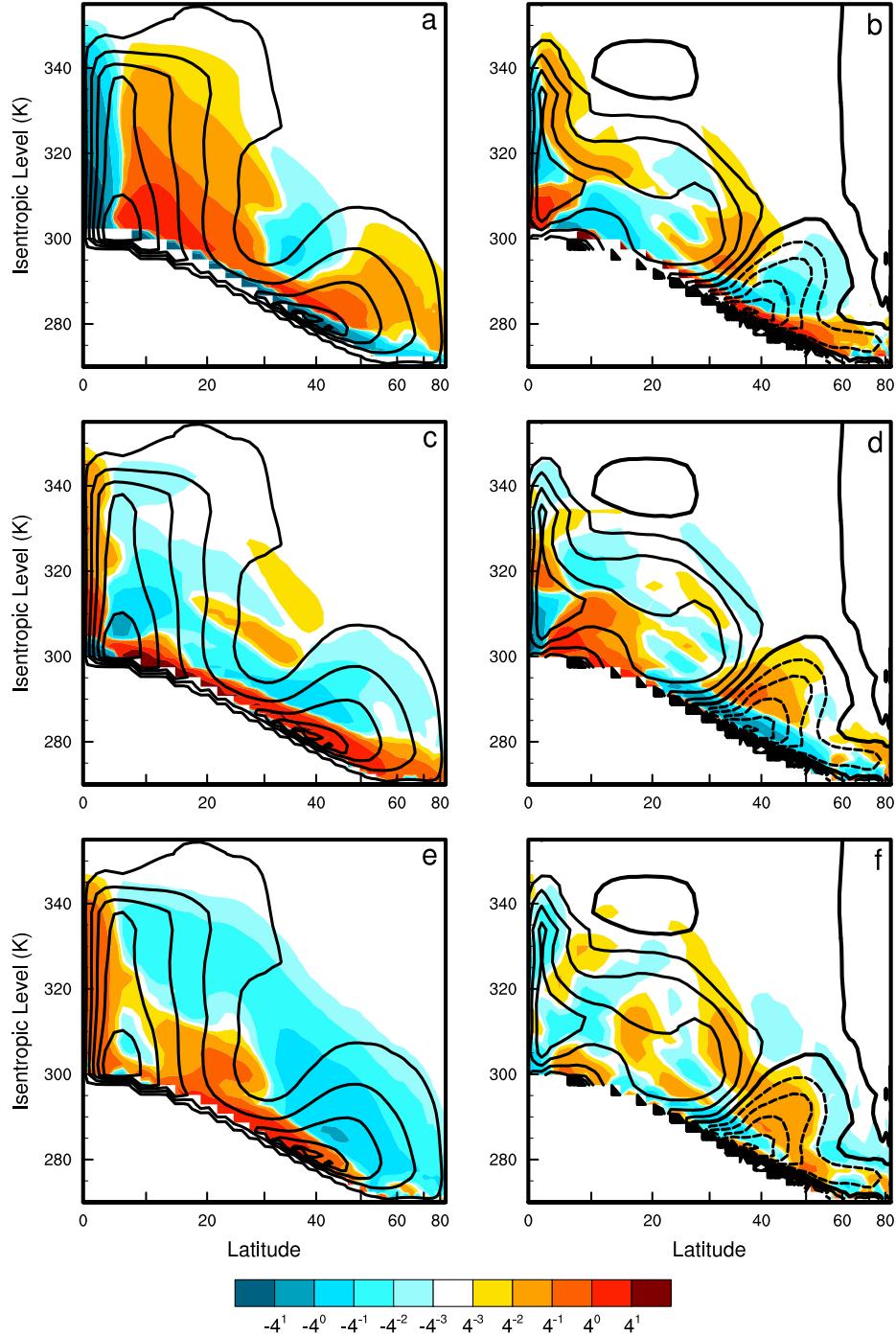


Figure 3: (a), (c) and (e) Components of the zonal average moisture budget in isentropic coordinates in the ne30 simulation. (b), (d) and (e) Departure of the components of the zonally averaged moisture budget in the ne120 simulation from the ne16 control. (a), (b) Divergence of the cross isentropic moisture transport by the mean, (c), (d) divergence of the along isentropic moisture transport by the mean and (e), (f) divergence of the along isentropic eddy transport. Units are in 10^{-6} kg/m²/K/s and the meridional streamfunction contours are as in Figure 2.

rectly from the available model output. Anomalies in the mean cross-isentropic flux of moisture near the equator are characterized by low-level divergence and convergence aloft (Figure 3b). This pattern indicates extraction of moisture from the lower troposphere into the middle troposphere due to greater equatorial ascent. Anomalies in along-isentropic fluxes (Figure 3d,f) supply the lower troposphere with moisture at the equator, maintaining the cross-isentropic anomalies.

The large subtropical drying anomaly near 40° (Figure 2d) is supported by anomalous divergence from all terms shown in Figure 3. The reduction in convective precipitation near 40° opposes drying through relative increases in moisture source (not shown). The divergence of along isentropic eddy fluxes (Figure 3c) is most closely aligned with the drying anomaly flanking the anomalous poleward flow and extending into the region of maximum drying near 40° (Figure 2d). This result is suggestive of greater eddy activity along the anomalous poleward flow. Figure 4 shows the changes in the eddy momentum fluxes, $\overline{u'v'}$, with resolution, relative to the eddy fluxes in the ne16 simulation (Figure 4a). An increase in eddy activity lies just above the anomalous poleward flow, and the magnitude of the anomalous eddy fluxes increase with resolution (Figure 4). Alignment of the anomalous poleward flow with the anomalous eddy fluxes is suggestive of a momentum balance between the Coriolis force and the divergence of the eddy fluxes. Alternatively, the increased wave activity may be related to a reduction in effective dissipation of potential vorticity with resolution (Lu et al., 2015).

The large changes in cloud fraction (Table 7) are balanced by greater over-

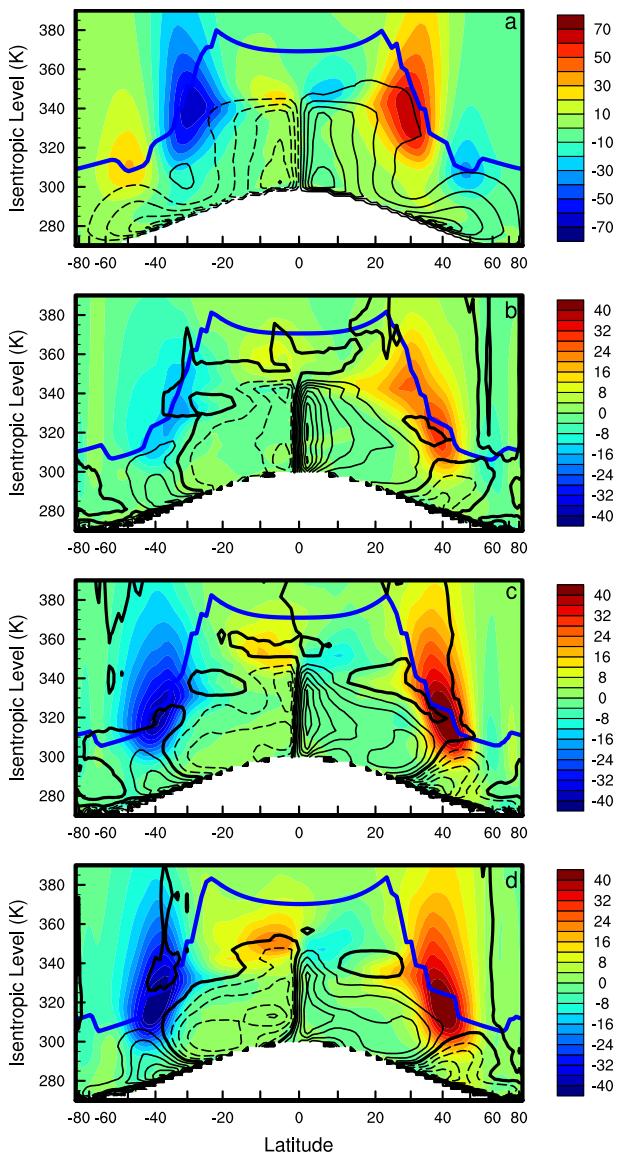


Figure 4: As in Figure 2, but for the eddy fluxes $\overline{u'v'}$ in units of m^2/s^2 .

turning of the Hadley Cell, greater isentropic mixing associated with variations in the extra-tropical circulation and a reduction in activity of the convection scheme. While this is not a particularly new finding (Kiehl and Williamson, 1991), it indicates that the same underlying sensitivity of cloud coverage to resolution has continued to persist within the CAM lineage for at least 25 years.

2.4 An Explanation

Our results have demonstrated that a progressive drying of the atmosphere with increasing resolution is balanced through tropical and extra-tropical changes to the large-scale circulation. Here, we make the important simplification that many of the complexities embedded in the sensitivity of the large-scale circulation to horizontal resolution is a non-linear response to a resolution sensitive process rooted in linear Boussinesq theory. This simplification is unverified - its purpose is to provide the simplest possible interpretation of the resolution signal that is consistent with linear theory. We recognize that there may be other important resolution sensitive processes (e.g., Lu et al., 2015), but suggest that they are not sufficient to explain the large changes to the hydrologic cycle identified in the simulations.

We begin with a non-hydrostatic, dry, non-rotating, frictionless atmosphere with no heat sources, a framework often used in the analysis of internal gravity waves (Holton, 2004). While the neglect of moist processes is clearly invalid for a moist GCM, the simplified framework may be viewed as approximating the evolution of individual dynamics time-steps in response to buoyant

perturbations passed from the column physics. For solutions to a linearized vertical velocity perturbation of the type $w = w_0 \exp[i(kx + mz + \sigma t)]$ in an unstable atmosphere, let $X^2 = -|N^2|$, where N^2 is the buoyancy frequency. The dispersion relation is,

$$\sigma^2 = X^2 \frac{k^2}{k^2 + m^2}, \quad (7)$$

where σ is the unstable growth rate, k is the horizontal wavenumber and m , the vertical wavenumber. According to equation 7, the growth rate is bounded by X since $k/\sqrt{k^2 + m^2}$ is bounded by unity. That is, convergence occurs as the horizontal scale of motion is reduced from the hydrostatic limit ($k \ll m$), to within the non-hydrostatic limit ($k \approx m$). For deep tropospheric perturbations, the non-hydrostatic limit occurs at horizontal resolutions approaching the depth of the troposphere.

At hydrostatic scales ($k \ll m$), the unstable growth rate is approximately linear in k (equation 7). The significance of k in the numerator of equation 7 can be traced back to the horizontal scale of the locally balanced hydrostatic pressure perturbation (Jeevanjee and Romps, 2016), driving horizontal convergence into the unstable air column and vertical motion. A model that permits removal of convective instability by the resolved dynamics results in an unstable growth rate, σ , that is linear in k at hydrostatic scales. This result is true of both hydrostatic and non-hydrostatic models. The equivalent dispersion relation for hydrostatic models is $\sigma^2 = X^2(k^2/m^2)$, implying that hydrostatic solutions are unbounded in scale in the non-hydrostatic regime

Table 9: Mean upward vertical pressure velocity at 600 hPa and the fraction of time the deep convection scheme is active in the simulations (FREQZM) in the region 10° N to 10° S.

	ne16	ne30	ne60	ne120
ω (hPa day ⁻¹)	-126.59	-156.40	-182.97	-225.14
FREQZM (fraction)	0.54	0.42	0.34	0.29

(Orlanski, 1981), referred to as an “ultra-violet catastrophe.” This ultra-violet catastrophe is not relevant at typical truncation scales of present day GCMs.

We hypothesize that the horizontal scale of the diabatic forcing from the column physics is proportional to the grid spacing, decreasing the horizontal scale of the hydrostatic pressure perturbations and increasing the unstable growth rate σ and resolved vertical motion as the resolution is increased. A reduction in horizontal scale of the diabatic forcing arising from the column physics was identified in CAM by Hack et al. (2006). Our hypothesis was motivated in part by the study of Williamson (1999), where convergence of the Hadley Cell was recovered in a prior version of CAM through fixing the horizontal scale of the physics forcing and increasing the resolution of the dynamical core. Our hypothesis attempts to explain the sensitivity of the resolved vertical motion in CAM with resolution identified in this study, and others (Kiehl and Williamson, 1991; Li et al., 2011; Yang et al., 2014; O’Brien et al., 2016).

Mean values of the vertical pressure velocity at the 600 hPa level, conditionally sampled for upward motion in the ascent region of the Hadley Cell (10° N to 10° S) are shown in Table 9. No interpolation was performed during

this calculation and all values were computed on their native grids. Ascent velocities associated with the Hadley Cell increase with resolution, and show no indication of convergence.

We argue that the increases in mean vertical velocities in Table 9 are due to contributions of increasingly finer scales with increasing horizontal resolution. Figure 5 shows the power spectra of 600 hPa vertical pressure velocities over the entire domain in the simulations, including upward and downward motion. Wavenumbers on the order 10 or less appear converged, potentially reflecting the ability of the model to resolve the deformation radius in all simulations and enabling realistic interactions between mid-latitude eddies and the mean flow (Williamson, 1999; Dong and Valdes, 2000; Zurita-Gotor et al., 2015). For wavenumbers on the order of 10 or larger, the general trend is for reduced power at lower wavenumbers and increased power at higher wavenumbers with increasing resolution (Figure 5).

Snapshots of the vertical pressure velocities at the equator for the four different resolutions (Figure 6) provide further evidence that a transfer of variance to finer scales contributes to the observed variations in the mean state (Table 9). Resolved convective updrafts populate the equator, and their horizontal scale decreases with increasing resolution. This transfer of velocity variance to finer scales is consistent with the model results and emergent scaling presented by Rauscher et al. (2016), as well as linear Boussinesq theory (equation 7).

The linear theory, and our hypothesis in general, is only applicable to convectively unstable conditions within the dynamical core. The occurrence of convectively unstable conditions may be thought to be infrequent in GCMs,

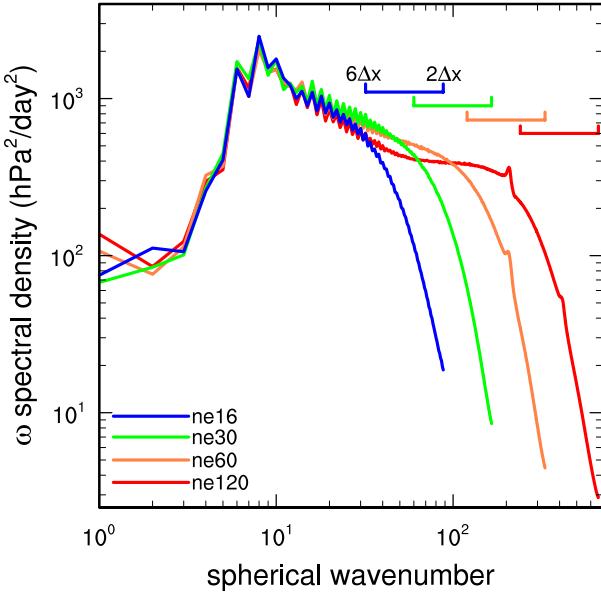


Figure 5: Power spectral density of the 600 hPa vertical pressure velocity over the entire model domain. Values computed from 360 instances of 6-hourly output randomly selected from the integration period.

as the convection schemes are designed to prevent the occurrence of resolved convection. Figure 6, however, provides evidence that resolved convection is a common occurrence in the deep tropics in CAM. Here, we propose an explanation for the occurrence of resolved convection in the presence of a convection scheme in CAM. The relevant ordering of the sub-models in our simulations, on the level of an individual physics time-step is the convection schemes, the large-scale precipitation scheme and the dynamics, respectively (Neale et al., 2010). This process ordering implies that convectively unstable air produced by the large-scale condensation scheme interacts directly with the dynamics before the convection schemes are able to consume the instability. The resolved dynamics consume buoyancy at a rate proportional to

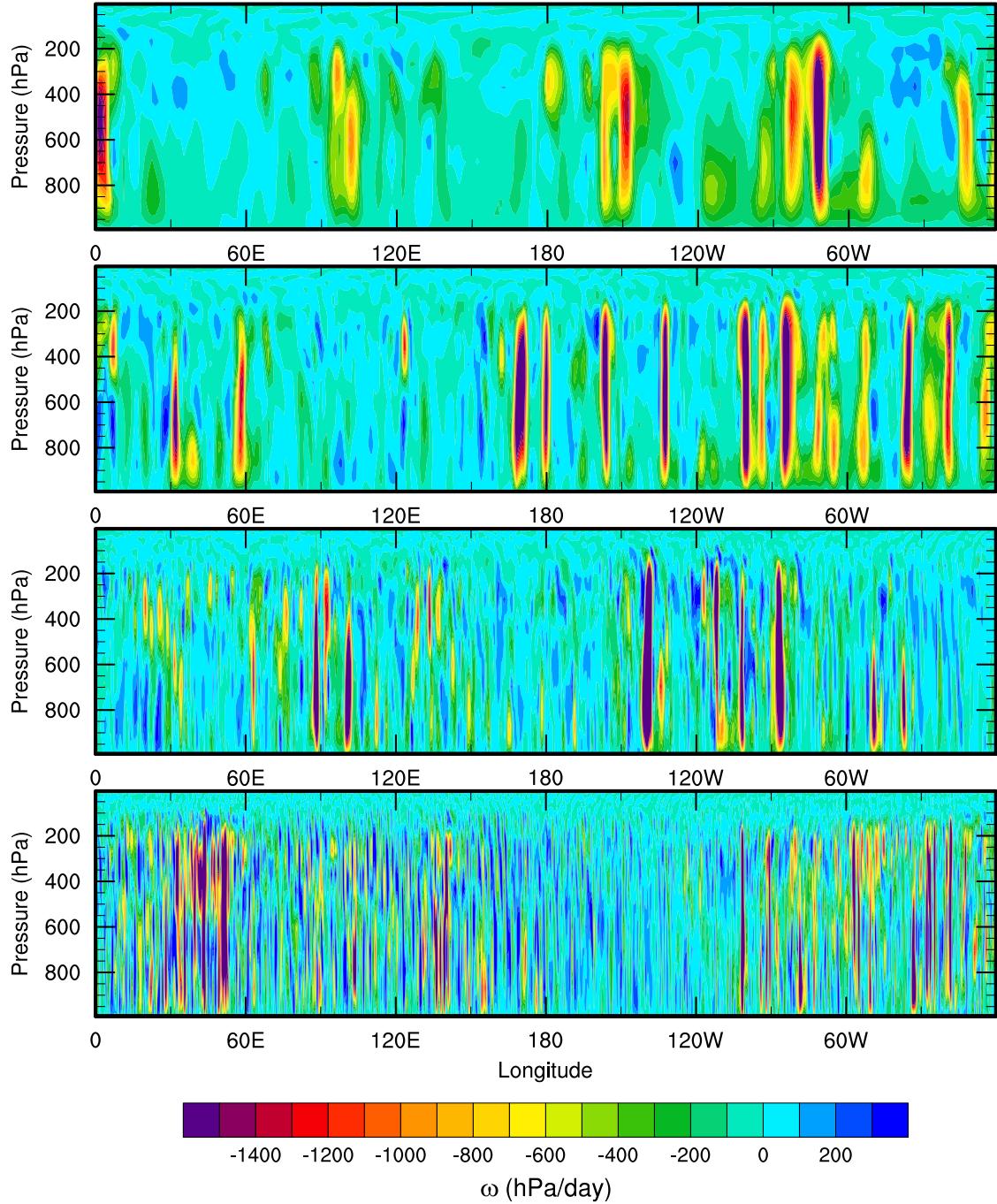


Figure 6: Longitude-height instantaneous snapshot of the vertical pressure velocity at the equator during a randomly chosen time in the (a) ne16 (b) ne30 (c) ne60 and (d) ne120 simulations.

ω in the thermodynamic equation, whose dynamics are approximated by the dispersion relation (equation 7). As the horizontal resolution of the model increases, resolved updrafts intensify with resolution, consuming buoyancy at an increasing rate. The convection scheme is then left with a smaller proportion of the available potential energy initially produced through the large-scale precipitation scheme. A caveat to this explanation is that the mean state is determined by the solution trajectory spanning many time-steps, which may not be reflective of processes occurring in a single time-step.

The activity of the deep convection scheme in the model is tied to an estimate of the available potential energy using a dilute plume calculation after Raymond and Blyth (1992). Although this quantity departs from the usual definition of the convective available potential energy (CAPE), we will refer to this quantity as CAPE throughout the discussion to be consistent with prior CAM studies. When CAPE exceeds 70 J/kg, the deep convection scheme is triggered (Neale et al. 2010). Unfortunately, CAPE is not available from the model output and we do not attempt to diagnose it from the available fields. A related quantity, the fraction of time in which the deep convection scheme is active in the simulations (FREQZM), is available from the model output and the mean values for the deep tropics are provided in Table 9. FREQZM is systematically reduced as the vertical velocities increase with resolution. We speculate that the resolved vertical motion becomes more effective at consuming available potential energy with increasing resolution, thereby reducing the frequency in which CAPE reaches its threshold value of 70 J/kg.

While we don't expect the steady-state GCM solutions of vertical velocity

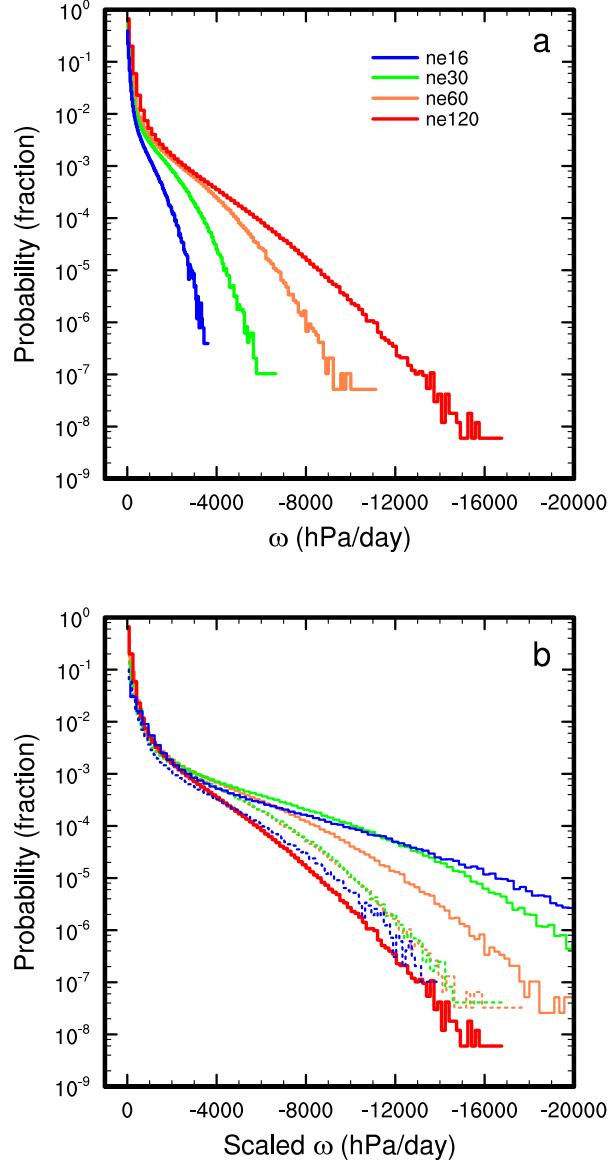


Figure 7: (a) Probability density distributions of 600 hPa vertical pressure velocity conditionally sampled for upward motion in the region 10° N to 10° S. (b) Probability density distribution of vertical pressure velocities in (a) rescaled to the ne120 simulation using a Δx^{-1} scaling (solid thin lines) and a $\Delta x^{-2/3}$ scaling (dotted thin lines). Values computed from 1460 instances of 6-hourly output randomly selected from the integration period.

to be linear in k , it is instructive to determine the degree to which the linear scaling reflects the full solution. To perform the scaling, we take the grid spacing, Δx , from the four simulations and scale the vertical velocities in the ne16, ne30 and ne60 simulations to the scale of the ne120 simulation using a $k \propto \Delta x^{-1}$ scaling. The probability density distribution of resolved upward motion in the deep tropics for all simulations are shown in Figure 7a, and the scaled distributions are provided in Figure 7b. The Δx^{-1} relationship does a poor job of explaining the sensitivity of the resolved vertical motion to horizontal resolution, systematically over predicting the vertical velocities compared to the ne120 simulation.

The Δx^{-1} scaling is an interpretation of the dispersion relation (Weisman and Klemp, 1982), but may also be derived from a scale analysis of the kinetic energy budget (Pauluis and Garner, 2006) or the Poisson equation (Jeevanjee and Romps, 2016). Because of the the poor fit of the Δx^{-1} scaling on the model solutions (Figure 7b), it might be tempting to abandon our explanation altogether. However, the physics of the Δx^{-1} scaling has a strong theoretical foundation, and instead we reconcile our explanation with a modification. We argue that the linear Boussinesq scaling is reflective of what occurs in GCMs, but that additional processes, likely associated with parameterized moist processes and resulting feedbacks onto the general circulation, steer the solution away from a Δx^{-1} scaling.

Rauscher et al. (2016) propose a shallower scaling exponent of $\Delta x^{-2/3}$. The is the scaling that results from applying mass continuity to the $-5/3$ spectral slope of the horizontal winds typical of mesoscale circulations (Rauscher et al.,

2016). For comparison, the $\Delta x^{-2/3}$ scaling is overlain in Figure 5b. While the $\Delta x^{-2/3}$ scaling overestimates the vertical velocities compared to the ne120 simulation, the overestimate is much less than any single curve using the Δx^{-1} scaling, and the scaled distributions all line up on top of one another. The Rauscher et al. (2016) scaling is clearly a more accurate representation of the sensitivity of the fully coupled solution to horizontal resolution than that provided by linear Boussinesq theory.

2.5 Conclusions

We have studied the convergence of the mean state of the spectral element version of the Community Atmosphere Model (CAM) coupled to the version 4 physics package in an aqua-planet configuration. As the horizontal resolution is increased, the atmosphere progressively dries, cloud cover decreases and the large-scale precipitation rate increases at the expense of convective precipitation. Variations in the global atmospheric energy budget are primarily a reflection of these changes, resulting in a reduction in longwave cloud forcing and an increase in clear-sky transmissivity with resolution. The regional atmospheric energy budget indicates that large-scale precipitation increases are balanced by anomalous convergence of dry-static energy transport, due to increasing mean vertical velocities in the deep tropics. An analysis of the moisture budget reveals that the drying trend in the simulations is balanced by anomalous divergence of moisture fluxes associated with an intensification of the Hadley Cell and simultaneous changes to the extra-tropical circulation. The large reductions in cloud fraction with resolution are a result of a lower

relative humidity field and reductions in activity of the convection scheme.

A possible explanation for the resolution sensitivity of the mean state was formulated, inspired by the experiments of Williamson (1999) and linear Boussinesq theory. We hypothesize that an increase in horizontal resolution leads to a reduction in horizontal scale of the diabatic forcing from the column physics, facilitating fine-scale flow and faster resolved convective updrafts within the dynamical core. Since buoyancy consumption is proportional to the magnitude of resolved upward motion, the dynamical core becomes more efficient at consuming buoyancy with resolution. As a result of the process ordering within CAM, and presumably other General Circulation Models (Beljaars, 1991), the convection scheme may only remove buoyancy produced through large-scale precipitation after the dynamics have had an opportunity to consume of some of the buoyancy, which we speculate results in a reduction in activity of the convection scheme with resolution.

It is difficult to provide definitive proof of our explanation from an analysis of the aqua-planet experiments. We suggest that the reason for this difficulty is because the implied Δx^{-1} dependence of the resolved vertical velocity elicits a non-linear response, pushing the full solution towards a new equilibrium that departs from Δx^{-1} scaling. We have developed some simple deterministic dry experiments to test our hypothesis, using idealized diabatic forcing. Our results indicate that a Δx^{-1} dependency is a robust feature of the spectral element dynamical core in CAM (Herrington and Reed, 2018). Future work will target how the inclusion of moist parameterizations effects the scaling.

The strong coupling of the large-scale precipitation scheme with the dy-

namics at high horizontal resolution, but still within the hydrostatic regime, has been shown to be important for the simulation of Tropical Cyclones (Zhao and Lin, 2012) and precipitation extremes (Li et al., 2011; Yang et al., 2014; Rauscher et al., 2016; O’Brien et al., 2016). Our explanation, if confirmed, implies that the interaction of the large-scale precipitation scheme with the dynamics is highly dependent on the horizontal resolution, and efforts to produce ‘scale-aware’ closures may need to focus on this interaction to preserve some of the benefits of high-resolution simulations.

3 An idealized test of the response of CAM to near grid-scale forcing across hydrostatic resolutions

3.1 Introduction

The implementation of static or adaptive grid refinement in global atmospheric general circulation models (AGCMs) is a desirable, and increasingly common feature of next-generation atmospheric models. Regional grid refinement permits the explicit simulation of processes at the kilometric scale in a global model, providing a unification of global, regional and, in some cases, cloud-scale weather and climate simulations within a single AGCM framework (e.g., Lorant and Royer, 2001; Walko and Avissar, 2011; Rauscher et al., 2013; Zängl et al., 2014; Zarzycki et al., 2014b; Rhoades et al., 2016a; Harris et al., 2016). This framework permits the interaction of the refined region with the globally resolved circulation, resulting in an autonomous representation of energy and enstrophy cascades (Charney, 1971; Lindborg, 1999), and global teleconnections (Wallace and Gutzler, 1981) that may affect the solution locally.

The spectral element dynamical core option (CAM-SE) in the Community Atmosphere Model (CAM) has been designed to support variable resolution grids (Zarzycki et al., 2014b; Guba et al., 2014b). While CAM-SE is equipped with higher order numerics and local conservation properties to prevent degradation of the solution across highly distorted elements (Taylor and Fournier, 2010; Zarzycki et al., 2014a; Guba et al., 2014b), the resolved vertical mo-

tion and precipitation rates exhibit an alarming sensitivity to grid-refinement (e.g., Reed et al., 2012; Yang et al., 2014; Zarzycki et al., 2014b; O’Brien et al., 2016; Herrington and Reed, 2017). Although it is commonly accepted that the sensitivity of AGCM solutions to grid-refinement are related to the parameterization of moist physics, the exact mechanisms are not clearly identified. The difficulties of deciphering causes of diverging AGCM solutions under grid-refinement lie in the significant complexity of the models themselves. For example, it is quite common for an AGCM to have three distinctly separate schemes for predicting cloud amount (e.g., Park et al., 2014). While there may be satisfactory reasons for this separation, and notwithstanding recent advances in unified cloud schemes (Golaz et al., 2002; Park, 2014), this complexity makes it difficult to unravel which component of which scheme, or interactions with other parts of the model, are contributing to the model’s behavior.

An alternative explanation for the resolution sensitivity observed in AGCMs relates to the inherent scale dependencies of the underlying equations of motion (Frierson, 2007; Herrington and Reed, 2017). These scale dependencies have been discussed before in the context of non-hydrostatic convection permitting simulations (Orlanski, 1981; Weisman et al., 1997; Pauluis and Garner, 2006; Jeevanjee and Romps, 2016), but its relevance to hydrostatic convection in AGCMs is rarely discussed (Frierson, 2007; Jeevanjee, 2017). This is hardly surprising since convection schemes are implemented in AGCMs to reduce the occurrence of resolved convection, and may lead many to question whether resolved convection occurs in AGCMs at all. However, evidence showing the

importance of resolved updrafts and downdrafts on simulated tropical convective systems in AGCMs can be found as far back as Moncrieff and Klinker (1997), and more recently by O'Brien et al. (2016).

Figure 8 illustrates the prevalence of resolved updrafts in an AGCM. The simulations are from a series of CAM-SE aqua-planet simulations (following Neale and Hoskins, 2000) at four different horizontal resolutions typical of present day AGCMs. The only parameters varied in the simulations are the dynamics time-step and hyper-viscosity coefficients, and the physics time-step is set to 1800 s for all resolutions (Table 10). All other aspects of the simulations are identical to the aqua-planet reference simulation described in Medeiros et al. (2016). Instantaneous snapshots of the vertical pressure velocity in the longitude-pressure plane at the equator for four different quasi-uniform horizontal resolutions are overlain by contours of large-magnitude, instantaneous diabatic forcing from the column physics. Resolved updrafts and downdrafts, primarily a response to grid-scale “thermals” and re-evaporation of falling condensate by the large-scale condensation scheme, populate the deep tropics and are a crucial component of the tropical circulation (Herrington and Reed, 2017).

What is also clear from Figure 8 is that the magnitude of the vertical motion, as well as the horizontal scale of the diabatic forcing and resolved convective elements as a whole, has some proportional relation to the grid spacing. The global power spectrum of the diabatic forcing arising from the column physics in the simulations (Supplementary Figure 18) provides conclusive evidence for a reduction in diabatic forcing scale with resolution. As

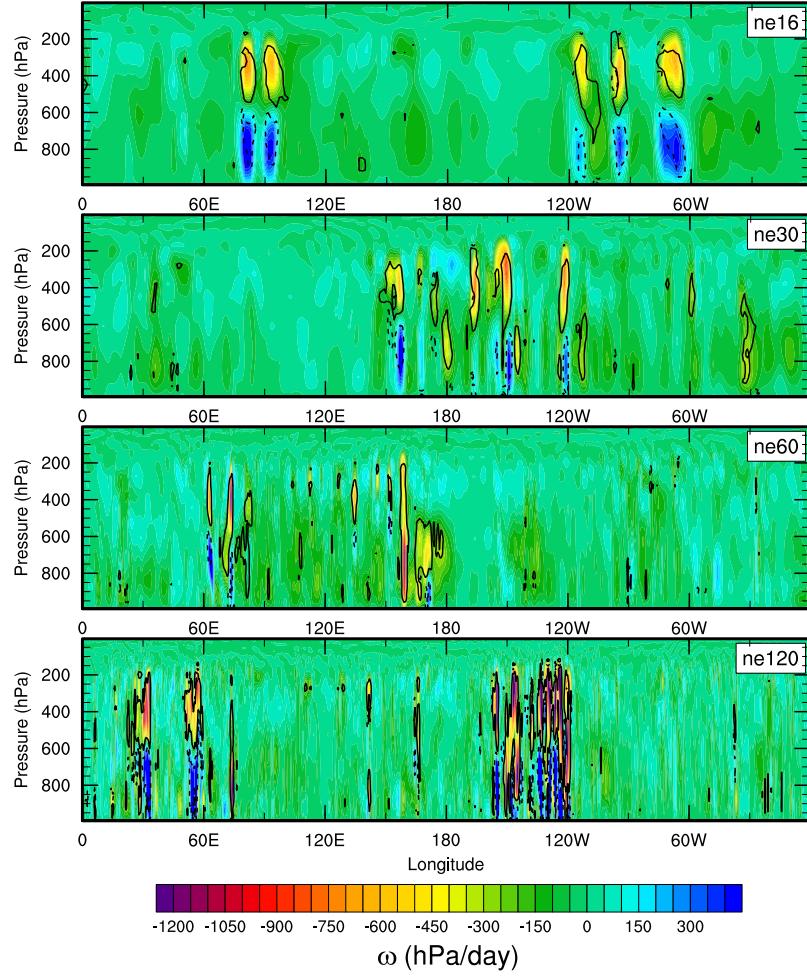


Figure 8: Instantaneous snapshot of the vertical pressure velocity (colors) and the diabatic forcing from the column physics (black contours) for a longitude-height transect at the equator from a series of aqua-planet simulations at four different horizontal resolutions. Contour interval is $+10 \text{ K day}^{-1}$ (solid) and -10 K day^{-1} (dashed).

discussed in Herrington and Reed (2017), the relationship between the horizontal scale of the buoyancy forcing, and the magnitude of the vertical velocity is consistent with the equations of motion. To illustrate this point, we proceed with a scale analysis of the non-rotating, dry equations of motion in the anelastic framework described in Jeevanjee and Romps (2016).

Jeevanjee and Romps (2016) augment the conventional anelastic approximation with a few important modifications. The first modification is that the full pressure is split between a locally balanced hydrostatic pressure, p_{HYD} , and a non-hydrostatic residual (Das, 1979). The second modification is the introduction of the effective buoyancy, β . The effective buoyancy is the vertical acceleration that would result from zeroing out the wind field, which is useful for isolating the component of vertical acceleration due to the Archimedean buoyancy. Jeevanjee and Romps (2016) then introduce the buoyancy pressure, p_β , which is the residual non-hydrostatic pressure associated with zeroing out the wind field, the vertical gradient of which is the effective buoyancy. These modifications lead to the Poisson equation (equation 8 in Jeevanjee and Romps, 2016):

$$-\nabla(p_\beta) = \nabla_h^2(p_{HYD}), \quad (8)$$

where ∇^2 is the Laplacian operator ($\nabla_h^2 + \frac{d^2}{dz^2}$) and ∇_h^2 is ($\frac{d^2}{dx^2} + \frac{d^2}{dy^2}$). Consider a vertical column of air of diameter D and height H , and constant Archimedean buoyancy B_0 throughout, then after taking the vertical derivative of equation 8, the scale analysis for β is:

$$\beta = B_0 \left[D^2 \left(\frac{1}{D^2} + \frac{1}{H^2} \right) \right]^{-1}, \quad (9)$$

(compare to equation 10 in Jeevanjee and Romps, 2016). Setting the scale of β to W/T , where W is a vertical velocity scale and $T = H/W$, and noting that at hydrostatic scales $\frac{1}{D^2} \ll \frac{1}{H^2}$, solving for W :

$$W = \sqrt{B_0 H} \frac{H}{D}. \quad (10)$$

The vertical velocity scale due to the Archimedean buoyancy is inversely proportional to the horizontal scale of the Archimedean buoyancy, at hydrostatic scales. Inspection of the Poisson equation (equation 8) indicates that the $1/D$ scaling is a result of the horizontal scale of the hydrostatic pressure perturbation, which drives convergence into the air column and vertical motion. The horizontal scale of the hydrostatic pressure perturbation is set by the horizontal scale of the Archimedean buoyancy (Jeevanjee and Romps, 2016), and determined by the horizontal scale of the diabatic forcing.

Through assuming a linear relationship between grid spacing (Δx) and diabatic forcing scale, which remains untested, but which Jeevanjee (2017) has shown evidence for at cloud-permitting resolutions, one may expect the variations in vertical velocity to go like Δx^{-1} . However, a Δx^{-1} relationship, while consistent with the general trend of greater vertical motion at higher resolution (e.g. Figure 8), overestimates the resolution sensitivity observed in a series of CAM-SE aqua-planet simulations analyzed in Herrington and Reed (2017). The simulations depicted in Figure 8 are similar – the Δx^{-1} scaling overestimates the sensitivity to horizontal resolution (Figure 17, dashed lines). The Δx^{-1} scaling may be incorrect simply because the relationship between

Table 10: Quasi-uniform horizontal grid spacing (Δx), time-stepping (Δt) and hyper-viscosity coefficients (ν) used in the aqua-planet simulations. Hyper-viscosity coefficients are after Zarzycki et al. (2014a) and are identical for the treatment of momentum, temperature and specific humidity.

	ne16	ne30	ne60	ne120
Δx (km)	208.5	111.2	55.6	27.8
Δt_{PHYS} (s)	1,800	1,800	1,800	1,800
Δt_{DYN} (s)	600	300	150	75
ν ($\text{m}^4 \text{s}^{-1}$)	8.073^{16}	10^{15}	10^{14}	10^{13}

diabatic forcing scale and Δ is non-linear. Alternatively, through maintaining the linear relationship between forcing scale and Δx , Herrington and Reed (2017) propose that feedbacks within other components of the AGCM steer the solution towards a weaker resolution sensitivity.

The goal of this paper is to test the robustness of the scaling (equation 10) in a simplified, non-rotating AGCM framework. We begin with an idealized set of rising thermal bubbles using CAM-SE without moist processes, and incrementally introduce some moist parameterizations to determine how their inclusion may affect the scaling of equation 10. We do not consider the influence of convection schemes – CAM convection schemes are incompatible with our experimental design. While this limits our ability to extend our results to the behavior of a complete AGCM, we argue that our simplified moist framework is still relevant to the behavior of a complete AGCM, and as such, will shed light on whether the moist physics or the scale dependencies of the dry equations of motion are the driving force behind the resolution sensitivity in CAM.

Rising bubble experiments are commonly used to test atmospheric models, although they are disproportionately used in non-hydrostatic models, often resembling an idealized thermal with horizontal scales on the order of a kilometer or less (Klemp and Wilhelmson, 1978; Grabowski and Clark, 1991; Bryan and Fritsch, 2002; Jeevanjee and Romps, 2016). The focus of this study is on bubbles of hydrostatic scale, and are intended as a surrogate grid-scale thermal occurring in more complex AGCM configurations (e.g., Figure 8). Section 3.2 provides an overview of the dynamical cores and moist physical parameterizations used throughout the paper, and a description of the analytical initial conditions. Section 3.3 presents the results of the experiments, Section 3.4 contains a discussion of the results and section 3.5 provides some concluding thoughts.

3.2 Materials and Methods

3.2.1 Dynamical Cores

For most of our experiments, we choose to work with the spectral element dynamical core option of the Community Atmosphere Model (CAM-SE; Dennis et al., 2012), the atmospheric component of the Community Earth System Model, version 1.2. CAM-SE solve the vector-invariant hydrostatic primitive equations using a fourth-order accurate, continuous Galerkin method on the sphere. A brief overview of the CAM-SE dynamical core is provided in Herrington and Reed (2017), and further details are available in the model documentation (Neale et al., 2012). The version of CAM-SE used in this study is more recent than for the simulations analyzed in Herrington and Reed (2017),

with the main distinction being the change to a third-order accurate, five-stage Runge Kutta time-stepping scheme (Dennis et al., 2012; Guerra and Ullrich, 2016).

CAM-SE uses an unstructured, cubed-sphere grid. For the quasi-uniform grid used in this study, the notation for the cubed-sphere grid is abbreviated by the number of elements along the edge of one of the six cubed-sphere panels. The grid used in this work contains 30 elements along the edge of a cubed-sphere face (abbreviated as ‘ne30’). Since each element contains 4 interior Gauss-Lobatto-Legendre (GLL) quadrature nodes, and 12 GLL nodes along the (shared) element boundary, the ne30 simulations correspond to an average horizontal grid-spacing of 111.2 km. The CAM-SE dynamical core contains no implicit dissipation in the horizontal, and so fourth-order hyper-viscosity operators are applied to state variables to suppress numerical artifacts (Dennis et al., 2012). The resolution dependent hyper-viscosity coefficients used throughout this study are the same as those used for the aqua-planet experiments provided in Table 10, and are from Zarzycki et al. (2014a).

Experiments are also performed using the finite-volume dynamical core in CAM (CAM-FV; Lin, 2004; Neale et al., 2012). CAM-FV solves for the horizontal dynamics using a finite-volume formulation of the vector invariant primitive equations (Lin and Rood, 1997). Like CAM-SE, CAM-FV utilizes the “Vertically Lagrangian” method (Lin, 2004), solving the horizontal dynamics within floating hybrid sigma-pressure surfaces, and periodically remapping the state back to a Eulerian, vertical reference coordinate (Neale et al., 2012). CAM-FV contains implicit dissipation of scalars quantities, which includes

vorticity, and explicit divergence damping is applied to the divergent winds. Both CAM-SE (Dennis et al., 2012), and CAM-FV (Lauritzen et al., 2011a; Whitehead et al., 2011) contain default fourth-order divergence damping, but to promote a larger diversity in design choices between the two dynamical cores, the second-order divergence damping option is used in CAM-FV (Collins et al., 2004). However, sensitivity experiments will be discussed that quantify the effect of the order of the divergence damping operator in CAM-FV. The dynamics are solved on a latitude-longitude grid, and necessitates a polar filter to stabilize solutions in polar regions (Neale et al., 2012). A standard grid with a horizontal resolution of 0.9° latitude by 1.25° longitude (nominally 1°), equivalent to an average equatorial grid spacing of 139 km, is used in all CAM-FV experiments.

The vertical resolution is the same for all grids in this study (regardless of dynamical core), consisting of 30 unequally spaced, hybrid sigma-pressure levels (Neale et al., 2012). Higher horizontal resolution grids are obtained by reducing the Earth-like planetary radius to match the resolution of other grid configurations, as in Reed and Medeiros (2016). Modifications to the Coriolis parameter are not required - only non-rotating planets are considered in this study. For all experiments described in this paper, a dynamics time-step of 75 seconds and 56.26 seconds are used in CAM-SE and CAM-FV, respectively, which satisfy the stability constraints of the highest resolution grids used throughout the study. The motivation for fixing the dynamics time-step, regardless of horizontal resolution, is in part for simplicity, but also to generalize our results to the behavior of the variable resolution version of

CAM-SE. The variable resolution version of CAM-SE requires the time-step to be equal everywhere on the grid, and must be chosen to satisfy the Courant number in the refined region (Zarzycki et al., 2014a).

3.2.2 Moist Physics

The moist experiments contain water species and parameterizations of moist processes. Water species are transported by the dynamical core, and the parameterizations are computed at the end of every third dynamics time-step in CAM-SE, and every fourth dynamics time-step in CAM-FV, equivalent to a physics time-step of 225 s. Further details on the coupling of the dynamical cores to the physical parameterizations is described in Herrington and Reed (2017) and Neale et al. (2012). We experiment with parameterizations that are intended to span the spectrum of complexity. In the least complex case, the dynamical cores are coupled to the large-scale condensation scheme of the ‘simple physics’ package of Reed and Jablonowski (2012), which we refer to as ‘simple condensation’ throughout the manuscript. This warm-rain scheme assumes all condensates are immediately removed from the column with no cloud phase.

Towards the other end of the spectrum, the Community Atmosphere Model, version 5 physics package (CAM5; Neale et al., 2012) contains a stratiform cloud scheme, with separate macrophysical (Park et al., 2014) and microphysical (Morrison and Gettelman, 2008) routines that together, constitute a more complex large-scale condensation scheme. The stratiform scheme computes ice and liquid cloud phases through prognostic equations for cloud droplet

number and concentration, and contains an explicit representation of falling hydrometeors (Morrison and Gettelman, 2008). As in the aqua-planet reference simulations presented in Medeiros et al. (2016), the cloud liquid and cloud ice concentrations are held to constant values ($100 \times 10^6 \text{ m}^{-3}$ and $0.1 \times 10^6 \text{ m}^{-3}$, respectively) to suppress dependencies on aerosol species, which are not present in our simulations. As a final level of complexity, we couple the stratiform scheme with the downgradient vertical dissipation routine used in CAM5 for the free troposphere, which diffuses momentum, temperature and specific humidity using standard mixing length and Richardson number dependent formulations (Neale et al., 2010).

Initial Conditions

The experiments are designed to isolate the scale sensitivity expressed in complex, moist AGCM resolution experiments, such as in Figure 8. The general procedure is to produce a set of approximately one-day long simulations of a rising thermal bubble in an idealized tropical environment, each bubble having its horizontal radius, r_h , and model horizontal grid-spacing, Δx , simultaneously reduced by a proportionality factor, X , to be defined later. An implicit assumption of this experimental design is that the forcing scale, $2r_h$, is linear in Δx .

The background temperature and moisture profiles are inspired by the thermodynamic structure of strongly convecting regions in the deep tropics of the aqua-planet simulations, and are similar to the neutrally stable conditions proposed by Bryan and Fritsch (2002). The background temperature

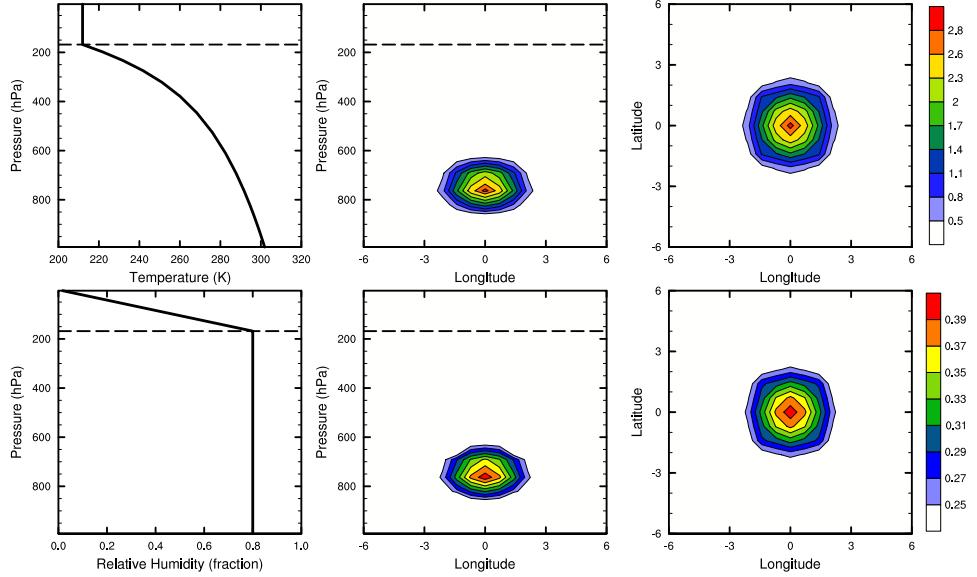


Figure 9: Initial conditions used in the moist experiments. Background temperature (top, left) and background relative humidity (bottom, left) reference profiles. A moist bubble with $r_h = X \times 350$ km, where X is the scale factor from Table 11, is depicted in the middle and right plots: potential temperature perturbation in the longitude-pressure plane (top, middle) and the latitude-longitude plane (top, right) and the relative humidity perturbation in the longitude-pressure plane (bottom, middle) and the latitude-longitude plane (bottom, right). The horizontal dashed lines denotes the location of the tropopause.

profile is a horizontally uniform, reversible moist adiabat pertaining to a surface temperature, T_0 , and constant relative humidity with height, h_0 . The tropopause pressure is the hydrostatically balanced pressure corresponding to a fixed height of 15 km, above which the temperature and specific humidity are set to the tropopause values (Figure 9).

A temperature perturbation is introduced following Klemp et al. (2015):

$$T'(R) = \begin{cases} \Delta\theta * \Pi * \cos^2(R\frac{\pi}{2}) & \text{for } R \leq 1 \\ 0 & \text{for } R > 1 \end{cases}, \quad (11)$$

with magnitude $\Delta\theta$. θ is the potential temperature, $\Pi = (p/p_0)^\kappa$ is the exner function and $\kappa = R_d/c_{pd}$, where g is the gravitational acceleration, R_d is the gas constant for dry air, c_{pd} is the specific heat capacity of dry air at constant pressure and p is the pressure.

$$R(r, z) = \left[\left(\frac{r}{r_h} \right)^2 + \left(\frac{z - z_c}{r_z} \right)^2 \right]^{\frac{1}{2}}, \quad (12)$$

where r_h is again the horizontal bubble radius, r_z the vertical radius. The great circle distance, r , from central latitude ϕ_c and central longitude λ_c is

$$r(\phi_c, \lambda_c) = a * \cos^{-1} (\sin(\phi_c) \sin(\phi) + \cos(\phi_c) \cos(\phi) * \cos(\lambda - \lambda_c)), \quad (13)$$

where a is the planetary radius. The hydrostatically balanced reference height, z , is found through vertical integration of the hydrostatic relation, and z_c is the location of the bubble center in the vertical. A relative humidity perturbation

Table 11: Scale factors (X) used to scale Earth's planetary radius to match the horizontal resolution of other grids (header), as well as the horizontal bubble radius (r_h) and hyper-viscosity coefficients (ν), for each equivalent resolution.

	ne16	ne30	ne60	ne120
Scale factor, X	15/8	1	1/2	1/4
Horizontal radius, r_h (km)	656.25	350	175	87.5
ν ($\text{m}^4 \text{s}^{-1}$)	8.073^{16}	10^{15}	10^{14}	10^{13}

is collocated with the temperature perturbation, and is of the form:

$$RH'(R) = \begin{cases} (1 - h_0) + \Delta h * \Pi * \cos^2(R\frac{\pi}{2}) & \text{for } R \leq 1 \\ 0 & \text{for } R > 1 \end{cases}, \quad (14)$$

where Δh is the magnitude of the relative humidity perturbation above saturation. The relative humidity parameters are set to $h_0 = 0.80$ and $\Delta h = 0.20$ (relative humidity is defined as a fraction as opposed to a percent, throughout this work) modeled after the aqua-planets. The definition of relative humidity in the initial conditions needs to be made consistent with the parameterization one is using. The simple condensation scheme and the CAM5 physics package use different definitions for saturation vapor pressure, and their initial conditions were made consistent with these definitions.

Through fixing the initial conditions, and multiplying the planetary radius, a , by the scale factor X , r_h is varied by the same factor as Δx . In the style of Reed and Medeiros (2016), the scale factors are defined as the coefficients required to scale the Earth's planetary radius to match the horizontal resolutions used in the aqua-planet simulations (Table 11). We set the horizontal

diameter of the moist bubbles to about seven times the grid spacing (specifically, we set $r_h = 350$ km at Earth's planetary radius) and a vertical diameter of 4 km ($r_z = 2$ km), which is comparable to the horizontal and vertical scales of diabatic forcing in the aqua-planet simulations (Figure 8).

Two additional important parameters are the vertical location of the bubble (z_c) and the magnitude ($\Delta\theta$). A smaller z_c or larger $\Delta\theta$ both lead to increased available potential energy, and potentially greater vertical motion. Figure 8 indicates that large-magnitude stratiform tendencies are common in the Tropical upper-troposphere at all resolutions. The diabatic forcing depicted in Figure 8 correspond to potential temperature perturbations on the order of 1 K. Despite the prevalence of the upper-troposphere structures, we opt for a more extreme initial condition, one that is less common in the simulations, to encourage non-linearity that could potentially break the scaling of equation 10. Plots of the initial conditions for the moist experiments are provided in Figure 9, depicting a bubble with a magnitude of 3K, a vertical extent much less than the depth of the troposphere, and located just above the nominal boundary layer. Table 12 provides a list of all parameters used to generate the initial conditions for the experiments.

3.3 Results

3.3.1 CAM-SE without moisture

Before embarking on an analysis of the moist experiments, we first consider the behavior of an experiment which contains no moist processes. The red markers in Figure 10 are the results of a dry experiment, i.e. the initial conditions are

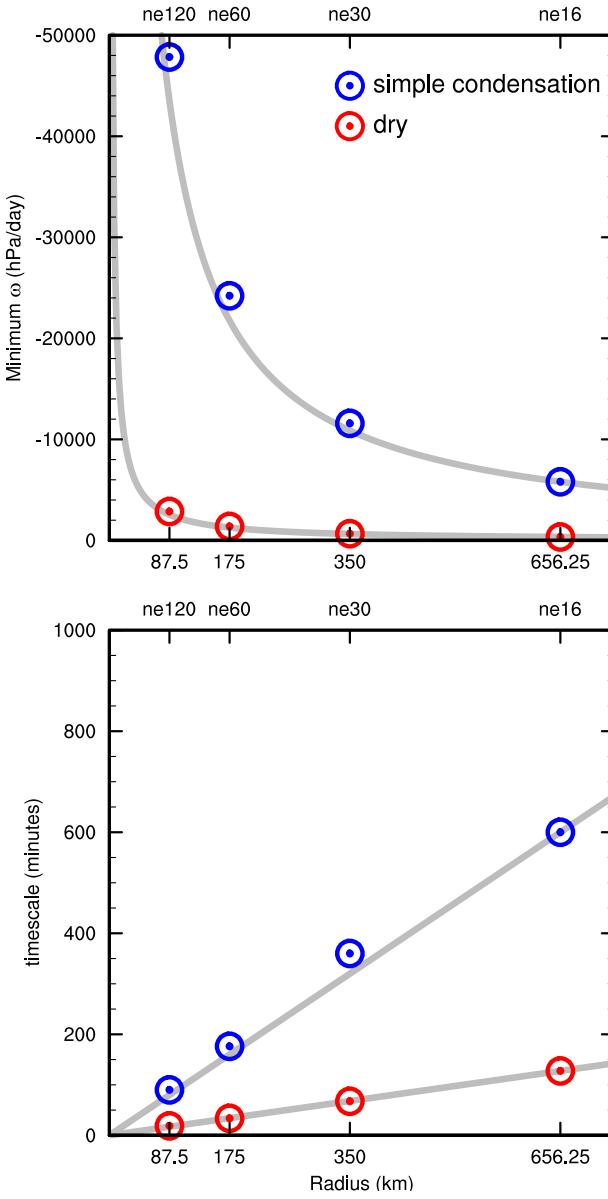


Figure 10: Results of the experiments using the simple condensation scheme and an equivalent configuration without moisture ('dry'). (Top) minimum vertical pressure velocity, ω , over the duration of each experiment as a function of bubble radius (Bottom) the model time associated with the minimum in ω as a function of bubble radius. The top axis indicates the equivalent horizontal resolution of the grid used to simulate a particular bubble radius. Grey lines represent the scaling of equation 10, scaled by the lowest resolution (ne16) solution.

Table 12: Parameters used in generating the initial conditions for the dry and moist experiments. X is the scale factor, provided in Table 11.

Parameter	Value
T_0 (K)	302.15
p_0 (Pa)	10^5
z_{TROP} (km)	15
$\Delta\Theta$ (K)	3
r_Z (km)	2
z_C (km)	3
φ_C	0
λ_C	π
h_0	0.8
Δh	0.2
a (km)	$X \times 6,371.22$

those of Figure 9, but with no specific humidity. The blue markers are the results from a moist experiment, and are discussed in the next section. Results are displayed as the minimum vertical pressure velocity, ω , over the duration of each experiment as a function of r_h . The $1/r_h$ scaling implied by equation 10 is overlain as a grey line, scaled by the lowest resolution simulation. The analytical scaling and the model experiments agree to a remarkable extent, indicating that the model dynamics are adequately described by the simple scale analysis of equation 10 for the case of a dry, rising bubble.

3.3.2 CAM-SE coupled to the Simple Condensation Scheme

Figure 10 also shows the results of the moist experiment with CAM-SE coupled to the simple condensation scheme. The simple moist experiments obey the linear scaling of equation 10, despite the inclusion of moist processes. This is not a trivial result since the analytical scaling (equation 10) was derived without any treatment of moist processes. Moisture does however amplify the

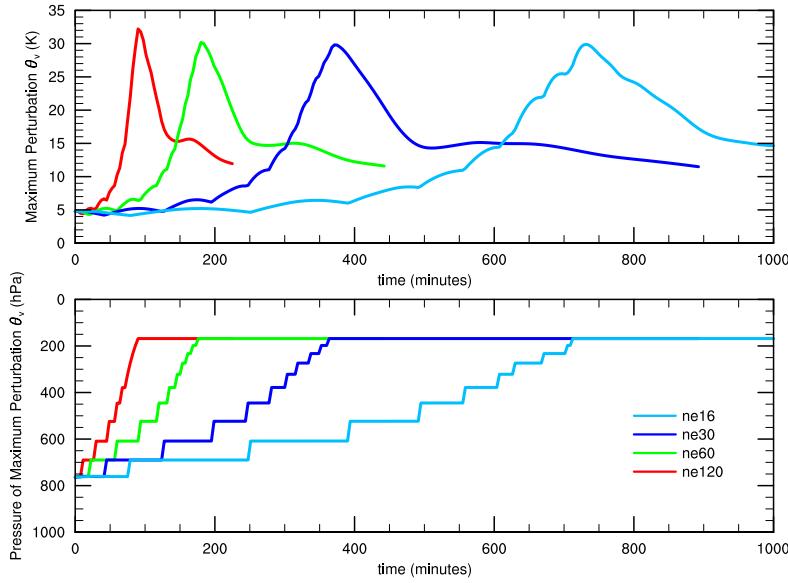


Figure 11: Time series of the maximum perturbation θ_v (top) and the pressure associated with this maximum (bottom) in the moist experiment using the simple condensation scheme.

vertical motion by about an order of magnitude compared to the dry case, with ω approaching -50×10^3 hPa day $^{-1}$, or about 20 m s $^{-1}$. Whereas the dry bubbles are unable to reach the tropopause, the moist bubbles maintain their structures all the way to the tropopause, owing to a much larger reservoir of available potential energy.

The minimum vertical pressure velocities in the moist tests are attained once the moist bubble reaches the tropopause. This can be determined by comparing the timescales in Figure 10 to the time evolution of the bubble pressure (Figure 11). Figure 11 also shows that the minimum in ω are associated with a maximum virtual potential temperature (θ_v) perturbation, defined as the difference in θ_v from the initial reference state. The maximum θ_v pertur-

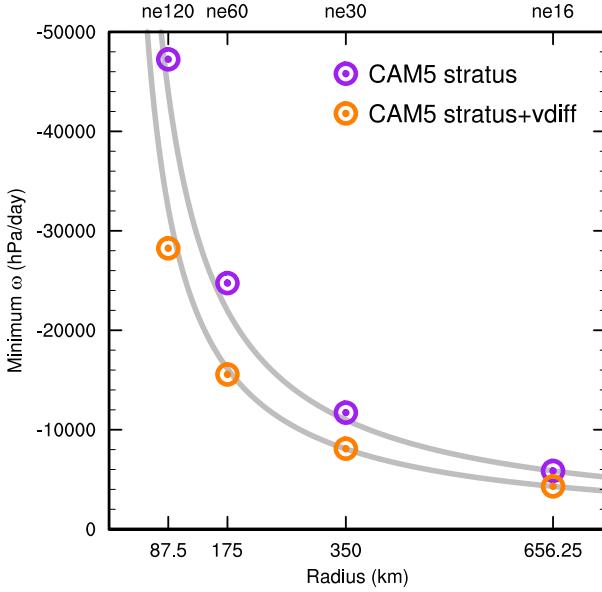


Figure 12: As in Figure 10, but for the moist experiments using the CAM5 stratiform scheme (‘stratus’) and the CAM5 stratiform scheme coupled with the CAM5 vertical diffusion routine (‘stratus+vdiff’).

bation is approximately invariant to bubble radius, indicating the difference in ω across resolutions are almost entirely a result of the $1/r_h$ scaling, as opposed to a change in available potential energy ($\sqrt{B_0 H}$ in equation 10). The scaling fits the moist experiments well since the horizontal scale of the moist bubbles do not change significantly as they ascend to their level of neutral buoyancy (i.e., the tropopause).

3.3.3 CAM-SE coupled to CAM5 Moist Physics

The moist experiments are repeated using the more complex physics modules contained in the CAM5 physics package. Despite the increased complexity of the CAM5 stratiform cloud scheme (labeled as “stratus”) compared with

the simple condensation scheme, the magnitudes of ω across the four different resolutions are nearly identical (Figures 12 and 10). The inclusion of vertical dissipation (labeled as “stratus+vdif”) reduces the magnitude of ω by about 40%, but preserves the $1/r_h$ scaling (Figure 12). This is achieved through reducing the magnitude of the Archimedean buoyancy by local downgradient vertical mixing, but the scheme is unable to affect its horizontal scale. We conclude that the incorporation of moist processes, short of a convection scheme, maintains a physical response analogous to the physics of the dry system (equation 10).

3.3.4 Sensitivity to physics time-step

CAM-SE

In all the moist experiments discussed so far, the physics time-step is set to 225 s. In practice, a longer physics time-step is used to minimize the computational overhead associated with the physics routines. For example, the default physics time-step in the ne30 configuration is 1800 s, and the experiments analyzed in Herrington and Reed (2017) use a physics time-step of 600 s. To explore the implications of using longer physics time-steps, the moist experiments are run using physics time-steps of 450 s, 900 s and 1800 s. These experiments are performed using the simple condensation scheme and the CAM5 stratiform scheme.

The results of the physics time-step experiments for the two different condensation routines are shown in Figure 13. The time-step sensitivity using only the CAM5 stratus parameterization is almost identical to the time-step

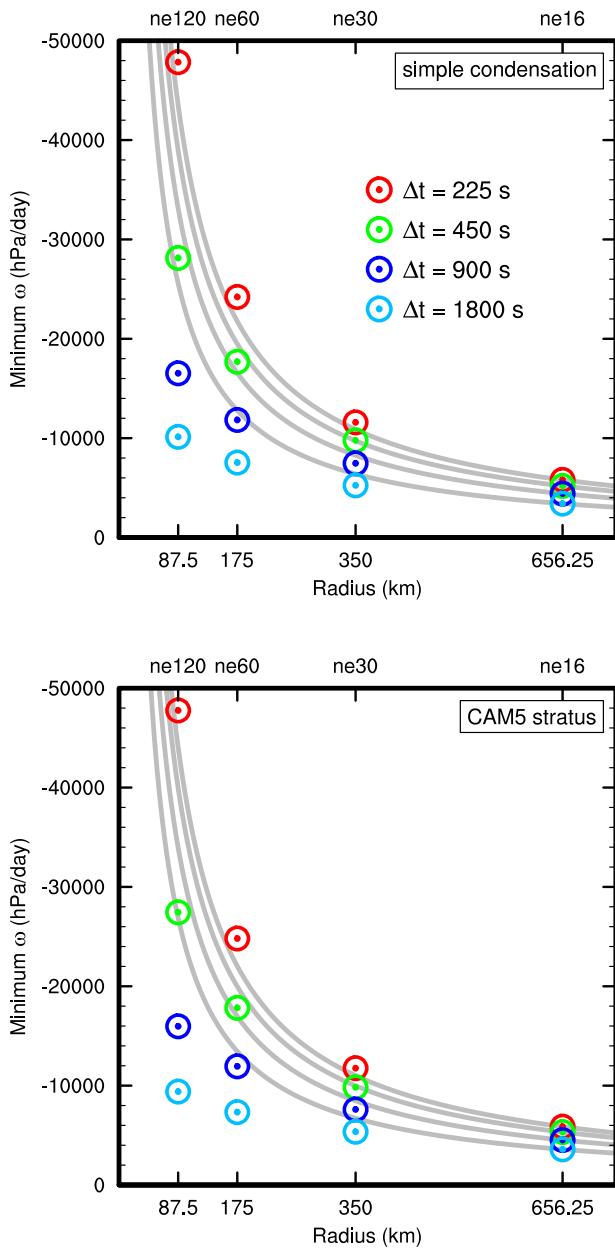


Figure 13: As in Figure 12, but for the physics time-step sensitivity experiments using the simple condensation scheme (top) and the CAM5 stratiform scheme (bottom).

sensitivity of the simple condensation scheme (Figure 13). For a physics time-step of 450 s, the simulations begin to depart from the $1/r_h$ scaling observed in the 225 s experiments, damping the vertical motion at smaller r_h . At longer physics time-steps, the departure from $1/r_h$ scaling is even greater.

CAM-FV

The time-step sensitivity experiment using the CAM5 stratiform routine has been repeated using the CAM-FV dynamical core (Figure 14). The magnitude of ω in CAM-FV is about 50% less than in CAM-SE, at all resolutions (Figures 13 and 14). This is consistent with simulations of tropical cyclones using these two dynamical cores – CAM-SE tends to have stronger storms as compared with CAM-FV (Reed and Jablonowski, 2012; Reed et al., 2012, 2015b). The qualitative behavior of the response to resolution and time-stepping is however, similar to CAM-SE – ω generally follow the scaling at small physics time-steps, but depart from the scaling at longer physics time-steps (Figure 14). Note that the ω field in the 2° runs do not follow the scaling (not shown). To observe the scaling in CAM-FV at small physics time-steps, one must scale to a resolution other than 2° (Figure 14).

The time-step sensitivity experiments in CAM-FV are repeated using the fourth-order divergence dampening operator, resulting in slightly lower magnitude vertical pressure velocities, but otherwise the experiment depicts a very similar sensitivity to physics time-step as the second order divergence damping simulation (not shown). The minimum ω in the simulation using the 225 s physics time-step and a 0.25° grid, is -20179.8 hPa/day, compared to -23008.7

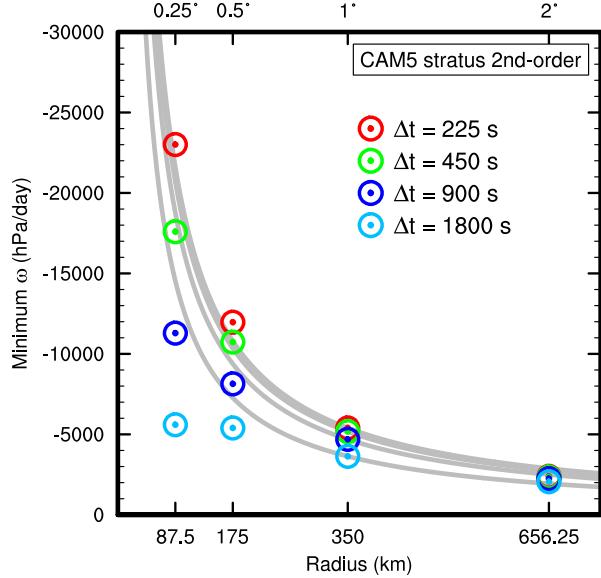


Figure 14: As in Figure 13, but using the CAM-FV with second-order divergence damping. Note the analytical scaling (grey lines) have been scaled to the 1° simulation in this plot, and the y-axis range is smaller, compared with Figure 13.

hPa/day with the second-order divergence damping. This result is intuitive – the higher the order of the divergence damping operator, the more effectively it damps near grid-scale features (Whitehead et al., 2011), such as the $\sim 7\Delta x$ moist bubbles.

The fourth-order, internally computed divergence damping coefficient was found to be about twice as large as that used by CAM-SE. An additional simulation using the 0.25° grid and 225 s time-step, in which the fourth-order divergence damping coefficient is lowered by an order of magnitude, results in an increase in the magnitude of ω by from -20179.8 hPa/day to -24689.7 hPa/day. The time-series of the minimum vertical pressure velocities in this

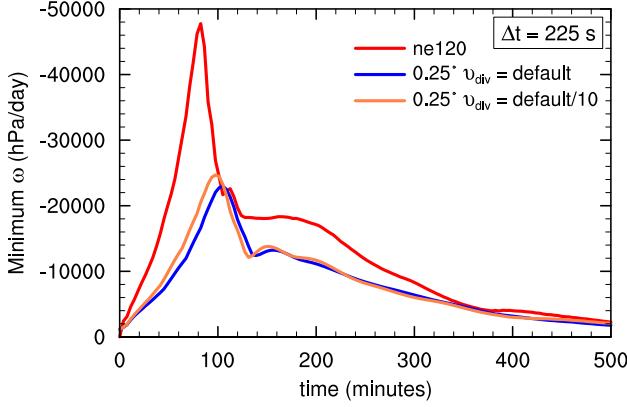


Figure 15: Time-series of the minimum in ω for three different moist bubble simulations using the CAM5 stratiform scheme and a physics time-step of 225 s, at the highest grid resolution. The CAM-SE run is labeled ‘ne120’, while a pair of CAM-FV simulations using the fourth-order divergence damping are depicted, one using the default divergence damping coefficient, and another simulation with the coefficient reduced by an order of magnitude.

simulation is compared to the equivalent CAM-SE simulation, in Figure 15, along with the CAM-FV simulation using the default fourth order divergence damping coefficient. While the larger divergence-damping coefficient used in CAM-FV may explain a portion of the differences between the CAM-SE and CAM-FV solutions, it does not explain most of the differences between the two solutions (Figure 15). The lower magnitude ω in CAM-FV may be a result of the diffusive nature of finite-volume numerics and/or limiters in the CAM-FV dynamical core (Lin and Rood, 1997), but could also be related to unphysical numerical aspects of the continuous Galerkin method, potentially resulting in artificially large magnitude in CAM-SE. Further testing is required to provide conclusive evidence of the design aspects leading to the differences in solutions between these two dynamical cores.

Sensitivity to viscosity

Through reducing the specific humidity hyper-viscosity coefficient in CAM-SE by an order of magnitude, the vertical velocity scaling is nearly recovered using a 450 s time-step (Supplementary Figure 19). Through individually reducing the hyper-viscosity coefficients on the other state variables (temperature, pressure thickness, non-divergent and divergent winds), the solutions either did not improve the scaling as effectively as the coefficient for specific humidity, or the model failed (note that the specific value of the coefficients are chosen by model developers, in part, for numerical stability). For example, model failure was common when reducing the divergence damping coefficient. In the simulations that did not crash, reducing the divergence damping coefficient by an order of magnitude lead to an increase in the magnitude of the vertical motion at all resolutions, but it's effect on the scaling is ambiguous, and not as clear as the experiment in which specific humidity coefficient is reduced (Supplementary Figure 19).

Figure 16 depicts the time dependence of the simulation in which the specific humidity hyper-viscosity coefficient is reduced by an order of magnitude. The results are presented as a time-series of the minimum ω in the highest resolution simulation, using a physics time-step of 450 s. The target solution, i.e., the solution found by scaling the lowest resolution simulation to the highest resolution using equation 10, as well as the highest resolution simulation using the default hyper-viscosity coefficients, are also shown in Figure 16. Through reducing the specific humidity coefficient, the ne120 solution approaches, but still underestimate, the target solution (Figure 16).

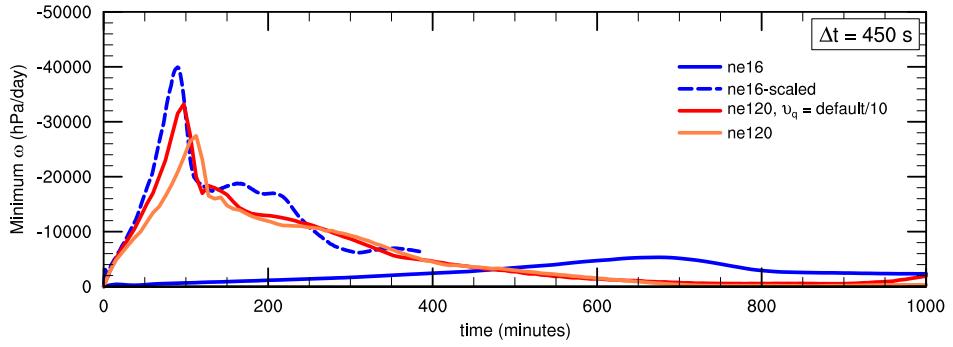


Figure 16: Time-series of the minimum ω in three different CAM-SE moist bubble simulations using the CAM5 stratiform scheme and a physics time-step of 450 s. The dashed line shows the results of an ne16 simulation scaled to the ne120 resolution using equation 10. The orange line denotes the ne120 resolution simulation, while the red line is an ne120 simulation with the specific humidity hyper-viscosity coefficient reduced by an order of magnitude.

3.4 Discussion

The idealized experiments are evidence that the scaling of equation 10 is maintained in a moist physics configuration, basically a two-component idealization of an AGCM containing only a dynamical core and large-scale condensation scheme, across multiple resolutions. Extending our results to the vertical velocity scaling observed in more complex configurations, such as aqua-planet experiments (e.g., Figure 8), may be a stretch. The CAM5 moist physics package contains shallow (Park and Bretherton, 2009) and deep (Zhang and McFarlane, 1995; Richter and Rasch, 2008; Neale et al., 2008) convection schemes. The large, predominantly stratiform thermals occurring only within the upper-troposphere in Figure 8, and the associated re-evaporation of falling condensate within the lower troposphere, are related to the deep convection scheme, and are designed to represent the effects of anvil cloud top formation in a deep

convective system (Neale et al., 2012; Park et al., 2014).

The source of moisture for the anvil tops in the model is primarily through detrainment of parameterized deep convective plumes, although moisture transport by the resolved dynamics and/or parameterized deep convective mass fluxes may be important as well. The anvil tops are confined to within the upper-troposphere because detrainment of the plume air is only permitted above the vertical level pertaining to the minimum in moist static energy (Zhang and McFarlane, 1995). The deep convection scheme is designed to remove unstable air originating in the planetary boundary layer through vertical redistribution, severely limiting its ability to reduce the unstable conditions created through anvil cloud top formation. Therefore, to understand the model response to the rather popular anvil cloud tops (Figure 8), the reduced stratiform-dynamical core system may not be a terrible approximation. The initial conditions used in the idealized moist bubble experiments contain unstable conditions only above the planetary boundary layer, and a separate run using the complete CAM5 moist physics package confirms that the deep convection scheme is not triggered by the initial conditions (not shown).

It is not common for climate models to be ran using a physics time-step as small as 225 s, potentially explaining why the aqua-planet simulations of Herrington and Reed (2017) utilizing a 600 s physics time-step, or those discussed in the introduction using an 1800 s time-step (Figure 17, dashed lines), do not follow the vertical velocity scaling of equation 10 across resolutions. To test this idea, four additional CAM-SE aqua-planet experiments are run, utilizing two different resolutions, ne30 and ne120. All four simulations are as described

in the introduction, and are run for one year. The dynamics time-step was fixed between the two resolutions, set to 75 s, and each resolution is simulated twice using physics time-steps of 1800 s and then 225 s. The results of these four simulations are depicted in Figure 17 as a probability density distribution of the negative ω at all vertical levels between 10°N and 10°S, computed from 6-hourly output over the final 11 months of the integration. The ne30 solutions are expressed as the ne30 probability density distribution scaled to the ne120 resolution following Pauluis and Garner (2006),

$$P(\omega_1) = \alpha P(\omega_0/\alpha), \quad (15)$$

where $P(\omega)$ is the probability density distribution of the vertical pressure velocity, the subscripts refer to two different resolutions, Δx_0 and Δx_1 , and α is the ratio of the velocity scale at Δx_0 to the velocity scale at Δx_1 . Assuming the characteristic buoyancy of parcels in a simulation is unchanged between two resolutions, then through assuming $D \propto \Delta x$, $\alpha = \Delta x_1 / \Delta x_0$ from equation 10.

Using an 1800 s physics time-step results in the usual scenario, with the scaled ne30 solutions generally over-predicting the occurrence of the ne120 upward ω (Figure 17), although the mismatch is not as severe as in Herrington and Reed (2017). This result is consistent with the moist bubble experiments - the over-prediction of vertical motion occurs using the 1800 s physics time-step (Figure 13). In contrast, with the 225 s physics time-step, the scaled ne30 solutions now under-predict the occurrence of large-magnitude vertical motion at high resolution (Figure 17), which does not occur in the moist bubble exper-

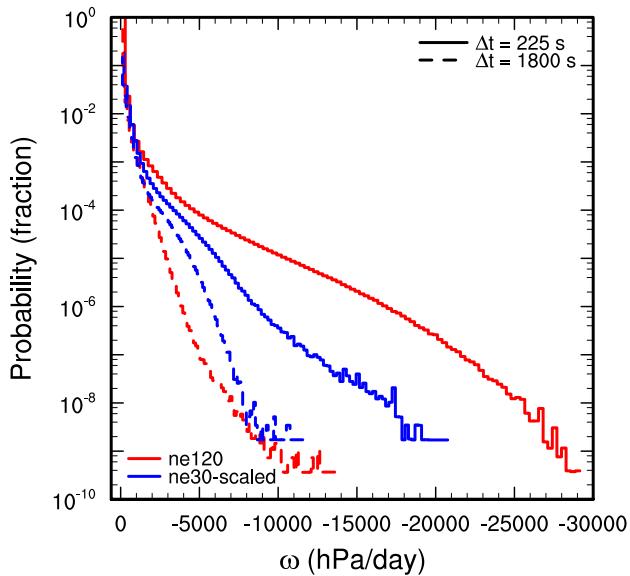


Figure 17: Probability density distribution of negative at all levels, and between 10°N and 10°S , for four different aqua-planet simulations. An ne30 and ne120 simulation are depicted with colors, and the physics time-step used in the simulations are shown in different line styles. The ne30 solutions are expressed as ne30 solutions scaled to the ne120 resolution, as described in the text.

iments (Figure 13). Reducing the physics time-step leads to larger magnitude vertical motions at both resolutions, which is apparent from the longer tails of the distributions (Figure 17). While the 225 s time-step simulations does not necessarily improve the ability of the scaling to explain the resolution sensitivity, the aqua-planet experiments are nonetheless consistent with idealized moist bubble experiments – the magnitude of the vertical motion is damped through the use of longer physics time-steps, and more-so at higher resolution.

One should bear in mind that comparing the relatively coarse time-sampling of the instantaneous ω in the aqua-planets (6-hourly) to the scaling derived from the high-frequency output (from 225 s to 1800 s) of the minimum ω in the

moist bubble tests, may be reason alone for the observed difference in scaling between the two configurations. While the authors feel that the sampling bias is mitigated to some extent through the use of a large sample size to compute the aqua-planets statistics, concerns regarding dataset inconsistency have not been thoroughly explored.

The authors speculate that the mismatch between scaled and the actual solutions at short physics time-steps in the aqua-planets are due to a non-negligible change in thermodynamic properties of thermals across resolutions. One example leading to such a change is the occurrence of grid-point storms at high resolution and small physics time-steps, reported in Williamson (2013). Compared with the ne30 simulation, the response of the ne120 simulation to the shorter physics time-step leads to a proportionally larger increase in the magnitude of ω (Figure 17). The probability density distributions presented in Williamson (2013) suggests that extreme values of ω in the tails of the distribution, such as occur in the ne120, 225 s time-step simulation (Figure 17), are consistent with grid-point storms. Grid-point storms can amplify ω through positive feedbacks on the parcel buoyancy, for example, by drawing in moisture through its own circulation, and over the course of the storms evolution (Williamson, 2013).

3.5 Conclusions

An idealized test was developed to understand the resolution sensitivity of the spectral-element (CAM-SE) and finite-volume (CAM-FV) dynamical core options of the Community Atmosphere Model, coupled with moist physics

routines of varying complexity, and in the non-rotating limit. The test consists of analytical initial conditions of a buoyant bubble in a neutral environment, modeled after strongly convecting regions in aqua-planets. The bubble may be thought of as a surrogate cloud thermal arising from the physics package in more complex AGCM simulations. Through varying the planetary radius and holding the initial conditions fixed, the radius of the bubble (r_h) is reduced by the same factor as the planetary radius, approximating the reduction in diabatic forcing scale that occurs when increasing the horizontal resolution of an AGCM. The experiments were performed in a dry framework and a moist framework. In the dry experiments, the sensitivity of the vertical velocities to bubble radius is almost exactly proportional to $1/r_h$, consistent with the scale analysis of the Poisson equation (equation 10) adapted from Jeevanjee and Romps (2016).

Through including moisture, and a simple or more complex large-scale condensation routine, the $1/r_h$ scaling is maintained in both dynamical cores. This is by no means expected given that the analytical scaling is derived from the dry equations of motion, and indicates the potential usefulness of the moist bubble test for the intercomparison of dynamical cores (e.g., Reed and Jablonowski, 2012). The greater available potential energy in the moist experiments results in vertical velocities that are an order of magnitude larger than in the dry experiments. In CAM-SE, incorporation of a downgradient vertical dissipation routine damps the Archimedean buoyancy and vertical velocity response, but still recovers the $1/r_h$ scaling. Experiments with the CAM-FV dynamical core simulate vertical pressure velocities about half as large, com-

pared with CAM-SE. Sensitivity experiments indicate that the differences in the strength of the explicit divergence damping between these two dynamical cores probably can't explain the main differences between the two models' solutions.

While it is common for CAM to be ran using physics time-steps in the range of 600 s to 1800 s (O'Brien et al., 2013; Wehner et al., 2014; Reed et al., 2015b; Zarzycki et al., 2014b; Reed and Medeiros, 2016), we have used a smaller physics time-step of 225 s. Additional experiments using more conventional, larger physics time-steps results in a severely damped vertical velocity response to r_h , with more damping at larger time-steps. The specific mechanism leading to the time-truncation errors at large physics time-steps is difficult to determine exactly. One clue came from a separate CAM-SE experiment, in which the scaling was nearly recovered with the hyper-viscosity coefficient for specific humidity dissipation reduced by an order of magnitude. We speculate that numerical dissipation from the dynamical core may result in time-truncation errors when the ratio of the physics to dynamics time-step is too large, resulting in too infrequent coupling between two fast processes, the condensation routine with its near instantaneous release of available potential energy, and the convectively unstable dynamics.

It is difficult to make an argument for or against a smaller physics time-step from this study alone. Due to our choice of horizontal resolution in this study, ‘thermals’ are represented as laminar, rising bubbles of hydrostatic scale. At cloud resolving scales, thermals are characterized as highly turbulent (Grabowski and Clark, 1991; Wicker and Skamarock, 1998; Bryan and Fritsch,

2002), consisting of $O(100$ m) turbulent eddies that mix the cloud and environment. Tompkins and Semie (2017) have illustrated the importance of the sub-grid turbulence scheme at cloud permitting resolutions, exerting strong control over moisture entrainment, the thermodynamic properties of convecting cores and large-scale organization. While a numerical filter should not be confused as a turbulence closure (Jablonowski and Williamson, 2011), unless, perhaps if it is explicitly designed for that purpose (e.g., Grinstein et al., 2007; Marras et al., 2015), there is certainly the possibility that modulating the strength of the numerical filter might act as some form of entrainment parameter for the grid-scale thermals in AGCM simulations. On the other hand, it has been argued that AGCMs are overly-dissipative (Shutts, 2005; Berner et al., 2012), implying that additional damping resulting from a longer physics time-step is not preferred. In any case, it is instructive for model users to understand that there is a large sensitivity of the vertical motion to physics time-step, especially at high resolution.

It may not be worth dismissing the relevance of this work to more complex AGCM configurations entirely due to the omission of convection schemes. Large, positive buoyancy forcing from the stratiform scheme frequently occurs above the planetary boundary layer, limiting a convection schemes ability to respond to this instability. The results of the aqua-planet experiments, which utilize the complete CAM5 physics package, generally support the notion of time-truncation errors at large physics time-step uncovered in the moist bubble experiments, suggesting there may be utility in studying the behavior of the reduced large-scale condensation-dynamical core system. This is not to sug-

gest that the convection schemes and stratiform scheme are entirely decoupled from one another - the convection schemes are crucial for pumping heat and moisture out of the boundary layer, often providing the near-saturated conditions in the free-troposphere that is eventually condensed by the stratiform scheme (Park et al., 2014). Future efforts will work towards initializing the boundary layer with unstable conditions in order to understand the influence of convection schemes on the vertical velocity scaling.

Finally, our work has implications for understanding the resolution sensitivity of AGCMs. This study, along with Herrington and Reed (2017), have shown that by assuming a linear relation between the horizontal scale of the diabatic forcing and the grid spacing, the analytical scaling derived from the dry equations of motion over-estimates the vertical velocity response to horizontal resolution in aqua-planet simulations, using conventional long physics time-steps. Through reducing the physics time-step to a less conventional 225 s, the problem appears to be reversed: the scaled solutions under-predict the magnitude of the vertical motion of the aqua-planets across resolutions. The authors speculate that this under-prediction is related to a change in characteristic parcel buoyancy across resolutions, e.g., due to the occurrence of grid-point storms at high resolution and short physics time-steps, identified in a prior version of CAM (Williamson, 2013). Additionally, the relationship between forcing scale and grid spacing may be non-linear, for example, if the forcing scale is determined by emergent constraints on the spectral properties of mesoscale circulations (Rauscher et al., 2016; O'Brien et al., 2016).

While there may be other factors contributing to the resolution sensitivity

of the aqua-planets (e.g., Lu et al., 2015), our idealized experiments has made clear that the vertical velocity across different resolutions is to first-order determined by the inverse of the forcing scale, and that departures from that scaling can occur through feedbacks with other components of the model. Rather than laying the burden of resolution sensitivity entirely on the moist physical parameterizations, this work is consistent with the idea that the inherent scale dependencies of the dynamical core are instead instigated by the horizontal scale of the buoyancy forcing the grid is able to support. The implications of the latter are that resolved updrafts and downdrafts, which are perhaps more wide-spread in AGCMs than previously thought, are very sensitive to horizontal resolution.

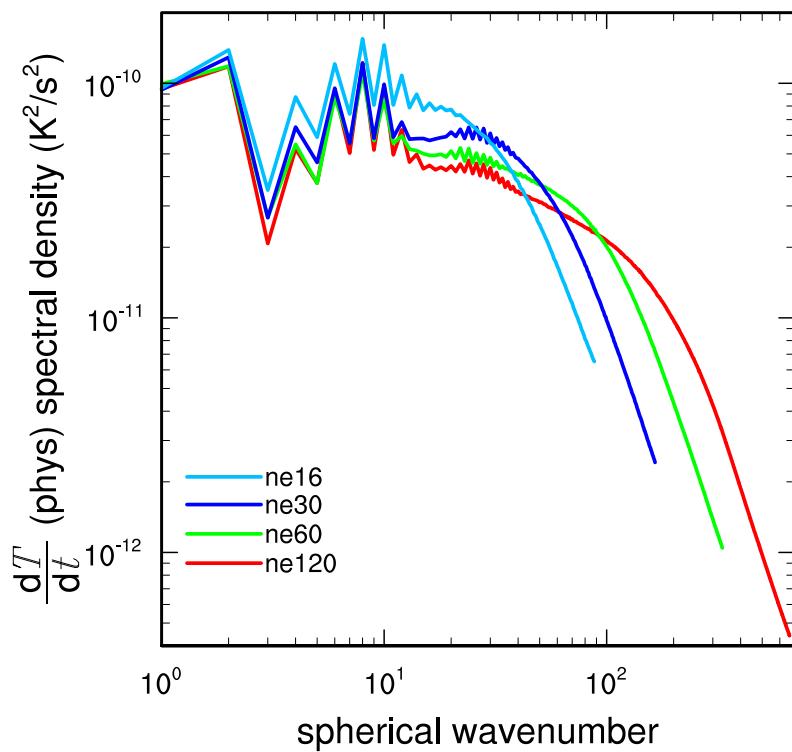


Figure 18: Global 400 hPa power spectral density of the temperature tendency arising from the CAM5 physics routines in the aqua-planet simulations.

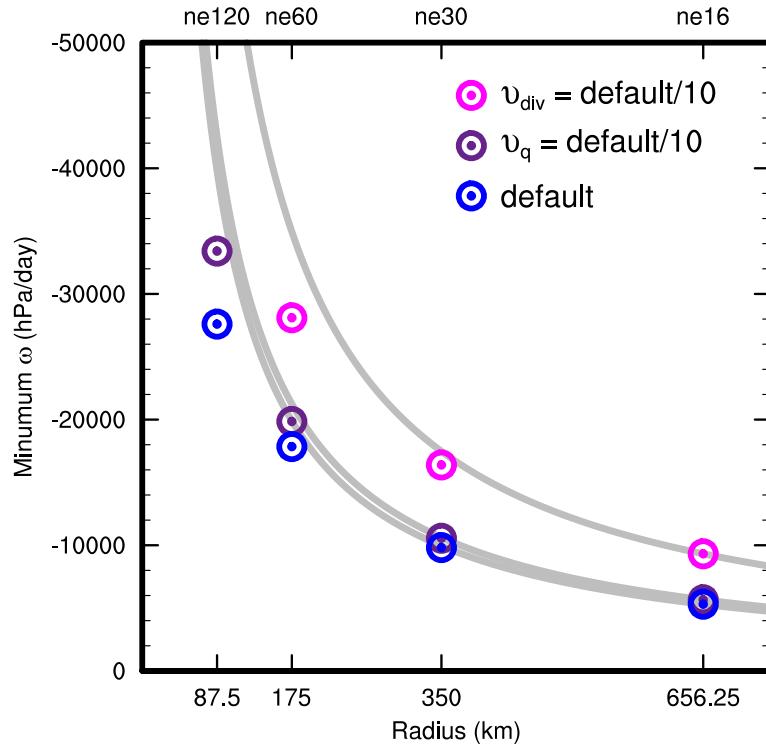


Figure 19: Results of three moist bubble experiments using CAM-SE coupled to the CAM5 stratiform scheme, using a physics time-step of 450 s. Results are expressed as the minimum vertical pressure velocity, ω , over the duration of each simulation as a function of initial bubble radius. Each experiment contains four simulations, with grid-spacing (top x-axis) and initial size of the bubble (bottom x-axis) progressively decreasing. The purple markers indicate an experiment in which the specific humidity hyper-viscosity coefficient is reduced by an order of magnitude. Similarly, the pink markers denote an experiment in which the divergence damping coefficient is reduced by an order of magnitude. The highest resolution divergence damping coefficient result is not shown since this configuration resulted in model failure.

4 Physics-dynamics coupling with element-based high-order Galerkin methods: quasi equal-area physics grid

4.1 Introduction

An increasing number of numerical methods publications in the atmospheric science literature concern transport, shallow-water, and three-dimensional models employing element-based high-order Galerkin discretizations such as finite-element and discontinuous Galerkin methods (for an introduction to these methods see, e.g., Durran, 2010; Nair et al., 2011; Ullrich, 2014). Some global models based on Galerkin methods have reached a level of maturity for which they are being considered for next generation climate and weather models due to their inherent conservation properties, high-order accuracy (for smooth problems), high parallel efficiency, high processor efficiency, and geometric flexibility facilitating mesh-refinement applications. NCAR’s Community Atmosphere Model (CAM; Neale et al., 2012) offers a dynamical core based on continuous Galerkin finite elements (Taylor and Fournier, 2010), referred to as CAM-SE (CAM Spectral Elements; Taylor et al., 2008; Dennis et al., 2012; Lauritzen et al., 2018). CAM-SE is, in particular, being used for high resolution climate modeling (e.g., Small et al., 2014; Reed et al., 2015a; Bacmeister et al., 2018) and static mesh-refinement applications (e.g., Fournier et al., 2004; Zarzycki et al., 2014a,b; Guba et al., 2014b; Rhoades et al., 2016b). Other examples of models based on high-order Galerkin methods that are be-

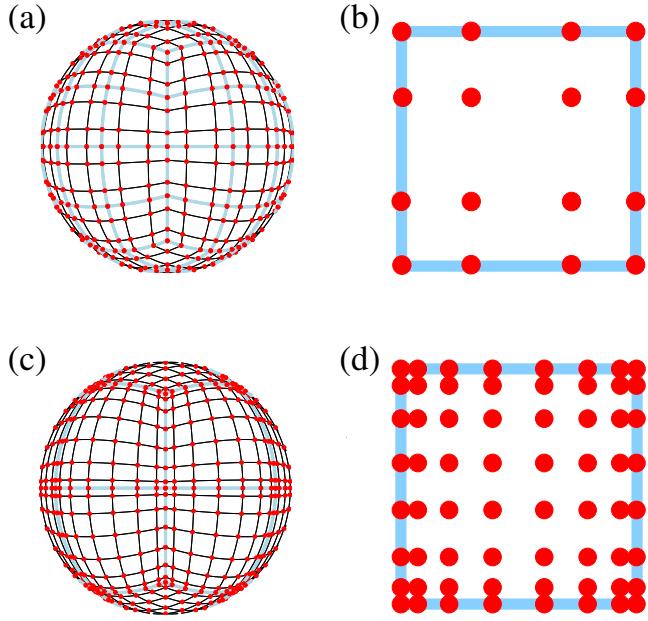


Figure 20: Example of CAM-SE GLL quadrature grids, marked with red filled circles, (a & c) on the cubed-sphere and (b & d) in an element. (a)-(b) and (c)-(d) use 4×4 ($np = 4$) and 8×8 ($np = 8$) GLL quadrature points in each element, respectively. (a) and (c) have the same average grid-spacing at the Equator (7.5°) which is obtained by using (a) 4×4 ($ne = 4$) and (b) 2×2 ($ne = 2$) elements on each cubed-sphere face/panel, respectively. The element boundaries are marked with thick light blue lines. The grid configurations shown on (a) and (c) are referred to as $ne4np4$ and $ne2np8$, respectively.

ing considered for ‘operational’ weather-climate applications are Giraldo and Restelli (2008), Nair et al. (2009), Brdar et al. (2013) and the Energy Exascale Earth System Model (<https://e3sm.org/>).

Assumptions inherent to the physical parameterizations (also referred to as *physics*) require the state passed by the dynamical core represent a ‘large-scale state’, for example, in quasi-equilibrium-type convection schemes (Arakawa and Schubert, 1974; Plant and Craig, 2008). In finite-volume methods (e.g., Lin, 2004), one may think of the dynamical core state as the average state of

the atmosphere over a control volume, and for resolutions typical of climate simulations is entirely consistent with the notion of a ‘large-scale state’. For finite-difference methods (e.g., Suarez et al., 1983) the point value is thought of as representative for the atmospheric state in the vicinity of the point value and one can usually associate a volume with the grid-point. Hence the physics grid (the grid on which the state of the atmosphere is evaluated and passed to physics) and the dynamics grid (the grid the dynamical core uses) coincide. Having the physics and dynamics grids coincide is obviously convenient since no interpolation is needed (which could disrupt conservation properties) and the number of degrees of freedom on both grids is exactly the same.

For the regular latitude-longitude, cubed-sphere and icosahedral grids the distance between the grid-points is gradually varying for finite-volume/finite-difference discretizations. Examples of models that use these grids are CAM-FV (latitude-longitude grid, Lin, 2004), FV3 (cubed-sphere grid, Putman and Lin, 2007) and ICON (icosahedral grid, Wan et al., 2013). For high-order element-based Galerkin methods, the dynamical core grid is defined by the quadrature points. In CAM-SE, these are the Gauss-Lobatto-Legendre (GLL) quadrature nodes. A unique aspect of the high-order quadrature rules is that the nodes within an element are located at the roots of the basis set, which may be irregularly spaced. For example, Figure 20 shows GLL points on an individual element of a cubed-sphere grid for degree 3 ($np \times np = 4 \times 4$ quadrature points) and degree 7 ($np \times np = 8 \times 8$ quadrature points) Lagrange polynomial basis used in CAM-SE. The higher the order of the quadrature rule, the greater variance in distance between GLL quadrature points within

an element. GLL quadrature points cluster near the edges and, in particular, the corners of the elements.

The resolved scales of motion are not determined by the distance between quadrature nodes, but rather the degree of the polynomial basis in each element. The nodes may be viewed as irregularly spaced samples of an underlying spectrally truncated state. From this perspective, one might expect the nodal solutions to be independent of location within an element. While the interior quadrature nodes are C^∞ in CAM-SE (i.e. the basis representation is infinitely smooth and all derivatives are continuous), the smoothness of boundary nodes are constrained by the need to patch neighboring solutions together to form the global basis set, an operation known as the direct stiffness summation (DSS; Maday and Patera, 1987; Canuto et al., 2007). The DSS operation is attractive because it allows for high-order accuracy with minimal communication between elements, but degrades the solution to C^0 at element boundaries (i.e., all derivatives are discontinuous). Through evaluating the physics at the nodal points, strong grid-scale forcing or oscillatory behavior near an element boundary may exacerbate the discontinuity, and our initial expectation, that the nodal solutions are independent of within-element location, is unlikely for non-smooth problems, e.g., the presence of rough topography or moist physics grid-scale forcing.

It is the purpose of this paper to document the implementation of an entirely separate, quasi-equal area finite-volume physics grid into CAM-SE. The use of a separate physics grid is not entirely unheard of; prior studies have utilized the infrastructure developed for global-spectral transform methods to

experiment with different physics grids (Williamson, 1999; Wedi, 2014). In our framework, the dynamical core state is integrated over control volumes to provide a volume averaged state to the physics, thereby minimizing the influence of any one particular nodal value on the physics forcing. Section 4.2 provides a thorough explanation of how grid imprinting manifests in high-order Galerkin methods for non-smooth problems. The implementation of the physics grid configuration into CAM-SE is presented in Section 4.3. Results from a hierarchy of idealized model configurations are presented in Section 4.4, illustrating the physics grid is effective at mitigating undesirable grid imprinting in the solution. Section 4.5 contains a discussion of results and concluding remarks.

4.2 The Quadrature Node Problem

Figure 21 is a schematic illustrating in one-dimension how grid-imprinting is enabled by the physics, when the dynamical core is built using high-order Galerkin methods. The schematic depicts a time-step, starting from smooth initial conditions (Figure 21a), and subsequently advancing the dynamics one Runge-Kutta time-step (Figure 21b). Since the boundary nodes of adjacent elements overlap one-another, there are now two solutions for each boundary node. The DSS operator, effectively a numerical flux applied to the element boundaries such that overlapping nodal values agree, is applied (Figure 21c), rendering the solutions at element boundaries C^0 ; less-smooth than neighboring C^∞ interior nodes. An element boundary discontinuity may be exacerbated if, e.g., the physics updates the state at an element boundary (Figure 21d,e),

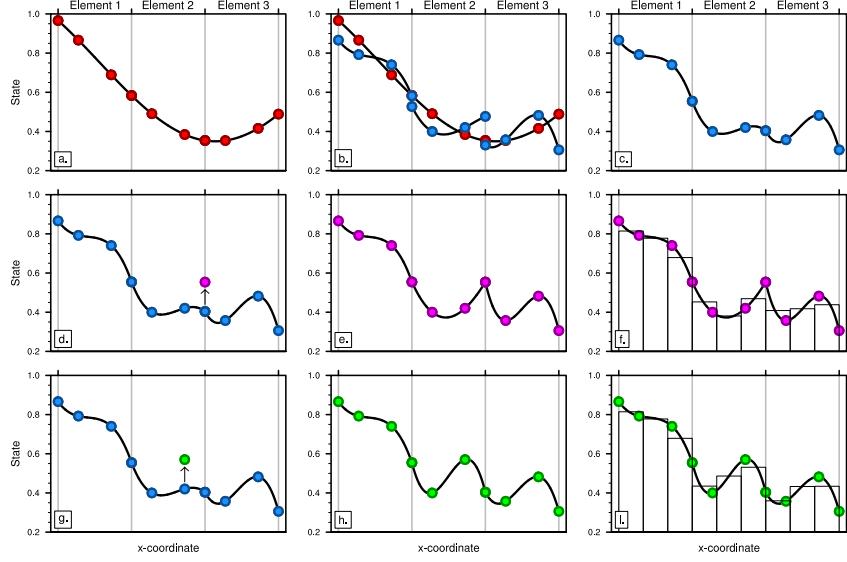


Figure 21: A one-dimensional schematic showing the relationship between the basis functions, the quadrature nodes and the proposed physics grid, over the coarse of a time-step. The filled circles are the GLL quadrature points in each element, which are connected by a Lagrange polynomials basis (curves). (a) Smooth initial condition are (b) advanced by the dynamics one Runge-Kutta step (blue), and (c) shows the solution after applying the DSS operator. Applying (d) grid-scale forcing to an element boundary node, (e) the basis representation is clearly C^0 at the element boundary. In contrast, (d) applying grid-scale forcing to an interior node (e) results in a smooth, C^∞ continuous field. (f),(i) Vertical bars pertain to the values on the physics grid, found through integrating the basis functions over the control volumes.

resulting in characteristically tighter gradients on the boundary nodes compared to if the physics forcing were applied to an interior node (Figure 21g,h).

To test the degree to which nodal solutions depend on within-element position, an aqua-planet simulation (Neale and Hoskins, 2000; Medeiros et al., 2016), which consists of an ocean covered planet in perpetual equinox, with fixed, zonally symmetric sea surface temperatures idealized after the present day climatology, is carried out using CAM-SE, using CAM, version 4 physics

(CAM4; Neale et al., 2010) and run for one year. The nominal low resolution $ne30np4$ grid is used, pertaining to an average equatorial grid spacing of 111.2km . The probability density distribution of the upward vertical pressure velocity (ω), conditionally sampled based on three categories - ‘interior’, ‘edge’ and ‘corner’ nodes - is provided in Figure 22a. The motivation for assessing noise in the ω field comes from its connection with the atmosphere’s divergent modes, as follows from the continuity equation in pressure coordinates. These modes are in turn sensitive to the within-element inhomogeneity of the pressure gradient that emerges from high-order Galerkin methods. There is an apparent dependence on nodal location, with interior nodes being characteristically sluggish, and corner and edge nodes having systematically larger magnitude vertical motion. This behavior is consistent with the smoothness properties of the different nodal locations, with discontinuous pressure gradients resulting in greater vertical motion at edge and corner nodes. The main division of solutions shown in Figure 22a is primarily between whether a node is, or is not situated on an element boundary, and is a nuanced signature of high-order element-based Galerkin methods for non-smooth problems.

If the conventional physics-dynamics coupling paradigm is applied to CAM-SE, then the physics are to be evaluated at the GLL nodes, and a volume associated with the quadrature point should be defined. One approach to construct this grid is to decompose each spectral element into $(np-1) \times (np-1)$ subcells and then take the dual grid of this subcell grid. For cubed-sphere meshes, this dual grid will have a control volume associated with each quadrature point. These control volumes will be triangles for the cube corner quadrature points

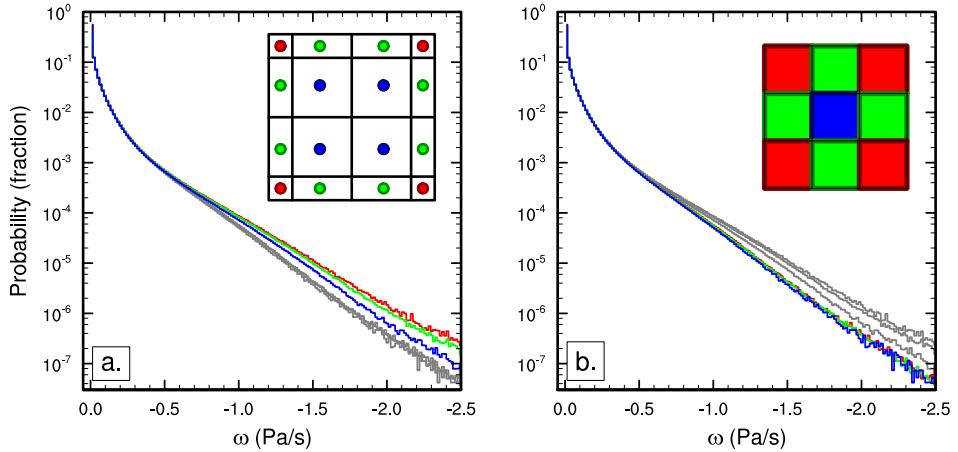


Figure 22: Probability density distribution of instantaneous upward ω in a pair of aqua-planet simulations using CAM4 physics. Figure is constructed from one year of six hourly data, at all vertical levels. (a) $ne30np4$ configuration conditionally sampled for interior, edge and corner node control volumes, and similarly (b) for the $ne30pg3$ configuration. The curves in (b) are overlain in (a) in grey, and similarly the curves in (a) are overlain in (b). Note the consistently larger magnitude ω for boundary nodes compared with interior nodes in (a), and that the bias is substantially reduced through mapping to a quasi-equal area physics grid.

and quadrilaterals for all remaining quadrature points. Newton iteration can then be used to adjust the corners of these control volumes so that their spherical area exactly match the Gaussian weight multiplied by the metric term (these weights are used for integrating the basis functions over the elements and can therefore, in this context, be interpreted as areas). For cubed-sphere meshes, the Newton iteration can be replaced by a direct method if some of the quadrilaterals are replaced by pentagons giving additional flexibility in matching the spherical area to the quadrature weights. Such a dual grid is shown in Figure 23. This grid is used in the NCAR CESM (Community Earth System Model) coupler for passing states between ocean, atmosphere and land

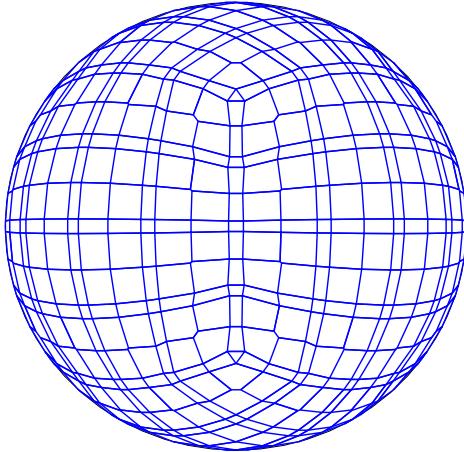


Figure 23: An example of control volumes constructed around GLL quadrature points ($ne4np4$) so that the spherical area of the control volumes exactly match the quadrature weight multiplied by the metric factor.

components since the current remapping method is finite-volume based and therefore requires control volumes (it is noted that methods exist that do not require control volumes for conservative interpolation, e.g., Ullrich and Taylor (2015)). Hence the components ‘see’ an irregular atmospheric grid. Similarly, the parameterizations in the atmosphere ‘see’ a state that is anisotropically sampled in space (see Figure 1 and 5 in Kim et al., 2008).

The quadrature grid in element-based Galerkin methods is defined to perform mathematical operations on the basis functions, e.g., computing gradients and integrals, rather than evaluating the state variables for physics-dynamics coupling. One may argue that it would be more consistent to integrate the basis functions over quasi-equal area control volumes within each element and pass those control volume average values to physics rather than irregularly spaced quadrature point values. In this case when integrating basis functions

over control volumes a grid-cell average value is more representative of the values near the extrema at the element boundary than the quadrature point value. The relationship between the nodal values, the basis functions and the proposed control volumes is illustrated schematically in one-dimension in parts (f) and (i) in Figure 21.

4.3 Methods

Here we focus on CAM-SE, however, in principle the methods apply to any element-based high-order Galerkin model. The physics grid in CAM-SE is defined by sub-dividing each element using equi-angular gnomonic coordinate lines to define the sides of the physics grid control volumes (see the Appendix 8.1) for details). Note that the element boundaries are defined by equi-angular gnomonic grid lines. The notation $pg = 3$ refers to the configuration where the elements are divided into $pg \times pg = 3 \times 3$ equi-angular physics grid cells (see Figure 24) resulting in a quasi-equal spherical area grid resembling the cubed-sphere. Defining the physics grid by sub-dividing elements makes it possible to use the same element infrastructure as already used in CAM-SE, thereby facilitating its implementation. Here we make use of the *ne30np4* and *ne30pg3* grids that use GLL quadrature point physics grid (physics and dynamics grid coincide), and the same ($pg = 3$) resolution quasi equal-area physics grids, respectively. In all configurations we use degree three Lagrange basis ($np = 4$) and $ne \times ne = 30 \times 30$ elements on each cubed-sphere panel.

A consequence of separating physics and dynamics grids is that the atmo-

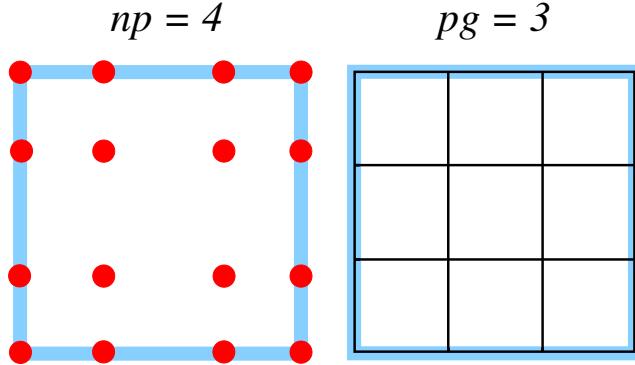


Figure 24: A schematic illustration of an element, indicating the relationship between (left) the dynamical core grid, and (right) the proposed quasi-equal area physics grid. The physics grid contains $pg \times pg = 3 \times 3$ grid cells in each element.

spheric state must be mapped to the physics grid and the physics tendencies must be mapped back to the dynamics grid. This is discussed in separate sections below. When separating physics and dynamics grids it is advantageous to use a vertical coordinate that is static during physics-dynamics coupling. This was one motivation to switch to a dry-mass vertical coordinate in CAM-SE (Lauritzen et al., 2018); since dry mass remains constant throughout physics the dry-mass vertical coordinate remains fixed during physics-dynamics coupling. The dry mass coordinate subsequently evolves as floating Lagrangian layers by the dynamics (Lin, 2004) periodically mapped back to a reference hybrid-sigma-pressure coordinate after Simmons and Burridge (1981). All variables mapped between grids are collocated, layer-mean values (Lauritzen et al., 2018).

4.3.1 Mapping state from dynamics grid (GLL) to physics grid (pg)

The dynamics state is defined on the GLL grid in terms of temperature $T^{(gll)}$, zonal wind component $u^{(gll)}$, meridional wind component $v^{(gll)}$, and dry pressure level thickness $\Delta p^{(gll)}$. In the mapping of the atmospheric state to the physics grid it is important that the following properties are met:

1. conservation of scalar quantities such as mass and dry thermal energy,
2. for tracers; shape-preservation (monotonicity), i.e., the mapping method must not introduce new extrema in the interpolated field, in particular, negatives,
3. consistency, i.e., the mapping preserves a constant,
4. linear correlation preservation, i.e., if field A is a linear function of B , this relationship is still preserved (see, e.g, equation 5 in Lauritzen and Thuburn, 2012)

Other properties that may be important, but not pursued here, includes total energy conservation and axial angular momentum conservation. Total energy is a quadratic quantity that is inherently difficult to conserve unless one maps total energy requiring one to diagnose either temperature or momentum components. For example, enforcing total energy conservation locally using, e.g., Lin (2004)'s method where total energy and velocity components are remapped and temperature is a derived variable, has proven problematic (C. Chen, personal communication). Similarly conservation of axial angular momentum is problematic. Conservation of angular momentum requires one to interpolate

the zonal and meridional components of momentum which creates large errors near the poles. To avoid the pole problem we interpolate contra-variant components of the momentum vector, which violates axial angular momentum conservation.

We argue that the most consistent method for mapping scalar state variables from the GLL grid to the physics grid is to integrate the Lagrange basis function representation (used by the SE dynamical core) over the physics grid control volumes, i.e., integrate the basis function representation of $\Delta p^{(gll)} \times T^{(gll)}$ and $\Delta p^{(gll)}$ over the physics grid control volume (see, e.g., Lauritzen et al., 2017; Ullrich and Taylor, 2015)

$$\Delta p^{(pg)} = \frac{1}{A^{(pg)}} \int_{A^{(pg)}} \Delta p^{(gll)} dA, \quad (16)$$

$$T^{(pg)} = \frac{1}{A^{(pg)} \Delta p^{(pg)}} \int_{A^{(pg)}} T^{(gll)} \Delta p^{(gll)} dA, \quad (17)$$

where $A^{(pg)}$ is the physics grid area. The integrals are numerically computed using the GLL quadrature rule on each physics grid element, which exactly (to machine precision) integrates the basis functions over the pg control volumes (Lauritzen et al., 2017). Thermal energy and dry air mass is conserved and the mapping is consistent. For the wind, which is a vector, the zonal and meridional wind components are mapped by transforming to contra-variant wind components, evaluating the basis function representation thereof at the equi-angular center of the physics grid control volumes and then transformed back to latitude-longitude coordinate system winds. All of the operations are local to the element and do not require communication between elements.

The mapping of tracers is more problematic since the SE basis function representation is oscillatory although the shape-preserving filter guarantees shape-preservation at the GLL nodes (Guba et al., 2014a). To avoid this issue we use the CAM-SE-CSLAM version of CAM-SE (Conservative Semi-Lagrangian Multi-tracer transport scheme Lauritzen et al., 2017), where tracers are advected on the $pg = 3$ physics grid using the inherently mass and linear-correlation preserving CSLAM algorithm. Note that in CAM-SE-CSLAM the dry mass internally predicted by CSLAM, $\Delta p^{(cslam)}$, is, by design, equal to $\Delta p^{(gll)}$ integrated over the CSLAM/physics grid control volume (Lauritzen et al., 2017). Since the tracer grid and physics grids are co-located and $\Delta p^{(pg)} = \Delta p^{(cslam)}$ then the mass conservation, correlation preservation, consistency and shape-preservation constraints are inherently fulfilled.

4.3.2 Mapping tendencies from physics grid (pg) to dynamics grid (GLL)

The physics tendencies are computed on the finite-volume physics grid and are denoted $f_T^{(pg)}, f_u^{(pg)}, f_v^{(pg)}$, and $f_m^{(pg)}$. Note that dry air mass is not modified by physics and hence there is no tendency for dry mass, $f_{\Delta p} \equiv 0$. Also, it is important to map tendencies and not state from the physics grid to GLL grid otherwise one will get spurious tendencies from mapping errors when the actual physics tendency is zero (unless a reversible map is used).

It is important that this process:

1. for tracers; mass tendency is conserved,
2. for tracers; in each tracer grid cell the mass tendency from physics must

not exceed tracer mass available in tracer grid cell (it is assumed that the physics tendency will not drive tracer mixing ratio negative on the physics grid),

3. linear correlation preservation,
4. consistency, i.e., the mapping preserves a constant tendency.

Other properties that may be important, but not pursued here, includes total energy conservation (incl. components of total energy) and axial angular momentum conservation. Scalar variables are mapped from the physics grid to GLL grid using a tensor-product Lagrange interpolation in two dimensions (i.e., we assume that the pressure variations in the vertical are small). The local coordinates on a cubed-sphere are discontinuous at the element edges so the interpolation requires special attention at the cube corners and edges. The details are provided in the Appendix 8.1. Lagrange interpolation preserves a constant (including zero) and linear correlations. Tracer and physics grids are co-located so tracer mass, tracer shape, and tracer correlations are trivially preserved on the tracer grid; and the inconsistency in point 2 above will not appear.

Mapping from pg to GLL grids while conserving mass was found to be difficult without excessive grid imprinting at element edges. Mass-conservation (using conventional finite-volume methods) requires a control volume to be defined around the GLL points (see Figure 23 in this paper or Figure 8b in Ullrich et al., 2016). These volumes are artificial and not consistent with the SE method. Integrating the CSLAM reconstruction of water tracers of such

artificial control volumes led to GLL node grid imprinting in the mapping and will not preserve a constant mixing ratio since the mapping of $\Delta p^{(pg)}$ to GLL will not yield the GLL node value for dry pressure-level thickness (i.e., the maps are not reversible). A reversible map requires that the number of degrees of freedom on the source mesh ($pg3$ has 9 degrees of freedom) equal the number of degrees of freedom on the target mesh ($np4$ grid has 16 degrees of freedom). This condition is violated by construction for individual elements.

It was also found important to use an interpolator that is smooth across element boundaries. Using an algorithm that only uses information from an element of control volumes will (at best) be C^0 at the element boundaries and therefore lead to boundary node grid imprinting. A stencil that extends beyond one element is therefore necessary. After much experimentation, the best results in terms of grid-imprinting were obtained with tensor-cubic interpolation (see the Appendix 8.1 for details) and by using the CAM-SE-CSLAM configuration (which requires the same boundary exchange/communication as used in CSLAM).

4.3.3 Time splitting and physics-dynamics coupling

The physics and dynamics are integrated in time using a sequential-update approach (e.g., Williamson, 2002). The dynamical core is sub-cycled over the (usually) longer physics time-step, Δt_{phys} , e.g., the vertical remapping time-step Δt_{remap} is cycled $nsplit$ times, totaling to Δt_{phys} . In CAM-SE, a fraction of the physics forcing, e.g., $f_q \times \Delta t_{remap}$ is applied at the beginning of each $nsplit$ vertical remap subcycles, such that the full forcing ($f_q \times \Delta t_{phys}$) is

realized over the course of a physics time-step. This approach of dribbling the tendencies over sub-intervals has the advantage of reducing gravity wave noise (Thatcher and Jablonowski, 2016), but may disrupt tracer mass conservation (Zhang et al., 2017). In CAM-SE-CSLAM, all but the tracer mass quantities are dribbled, with tracer mass receiving the full physics update, e.g., $f_q \times \Delta t_{phys}$, applied only at the beginning of the first remap sub-cycle, and thereby conserving tracer mass. This is the $f\text{type} = 2$ configuration described in detail in Section 3.6.3 in Lauritzen et al. (2018).

In the SE integration of the equations of motion on the GLL grid the water species are needed in the computation of the pressure gradient force and generalized expressions for heat capacity at constant pressure c_p , etc. Hence the mixing ratios for water vapor and dynamically/thermodynamically active condensates (e.g., cloud liquid and cloud ice) are needed on the GLL grid. We have chosen to advect the water species on the GLL grid using the SE method as well as on the physics grid using CSLAM. Every time physics updates the water species on the CSLAM grid, a forcing term (equal to the difference between updated CSLAM water variables and the SE values) is applied to the GLL water variables using dribbling so that the CSLAM solution and SE solution for water species are tightly coupled.

4.4 Results

A hierarchy of idealized model configurations are presented in order to elucidate the differences between CAM-SE and CAM-SE-CSLAM (available from the CESM2.1 release; <https://doi.org/10.5065/D67H1H0V>). Here, the con-

figurations are presented in order of increasing complexity, each with a pair of approximately 1° simulations, pertaining to the *ne30np4* (CAM-SE) and *ne30pg3* (CAM-SE-CSLAM) grids, and a $\Delta t_{phys} = 1800$ s.

4.4.1 Moist Baroclinic Wave

The moist baroclinic wave test case was developed as part of the ‘CESM Simple Models’ project (Polvani et al., 2017), and included in the release of CESM2. It is effectively the dry test-case of Ullrich et al. (2014), but initialized with moisture and coupled to the Kessler moist physics routine (Kessler, 1969). For more details on this test case (which was part of the 2016 Dynamical Core Model Intercomparison Project, Ullrich et al., 2017), see Section 4.1 in Lauritzen et al. (2018). A measure of the uncertainty in the reference solution, the L_2 difference norm between two high-resolution solutions using different dynamical cores, was also presented in Lauritzen et al. (2018) and provided again here in Figure 25. The L_2 norm between CAM-SE and CAM-SE-CSLAM lies below the uncertainty of the reference solution, indicating their differences are insignificant.

The flow field of the baroclinic wave test is used to drive the terminator “toy”-chemistry test of Lauritzen et al. (2015b, 2017). The terminator test is used to assess linear-correlation preservation using two reactive species advected across the terminator line. The model is initialized with species for which their weighted sum, Cl_y , is a constant (constant surface pressure and constant mixing ratio; $Cl_y = Cl + 2Cl_2 = 4 \times 10^{-6}$ kg/kg), such that if tracer correlations are preserved, then the column-integrated weighted sum of the

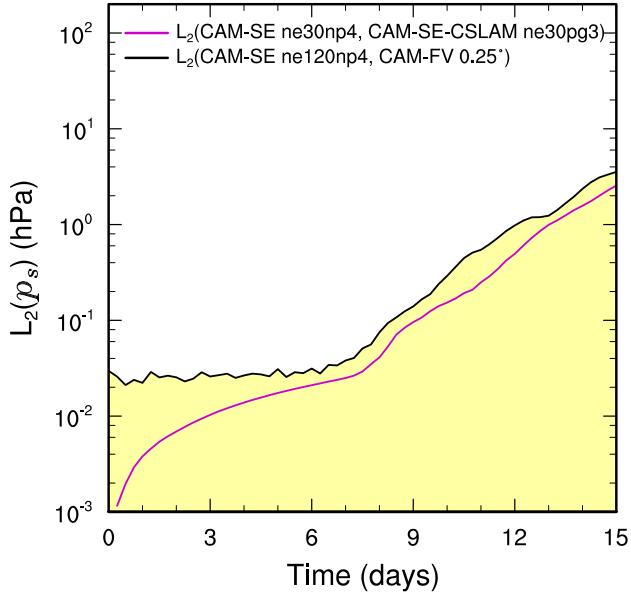


Figure 25: L_2 difference norms of the surface pressure field, p_s , in the moist baroclinic wave simulations. L_2 values lying within the yellow region fall below the estimate of the uncertainty in the reference solution (black curve), computed as the difference norm between two approximately 0.25° resolution versions of CAM, the spectral-element and finite-volume (CAM-FV) dynamical cores.

species should not vary in time. Figure 26 provides a snapshot of the vertically integrated weighted sum of species at day 15. In CAM-SE, the tracer correlations are not preserved at day 15 and the field is populated by overshoots and undershoots. In contrast, by day 15, CAM-SE-CSLAM still conserves tracer correlations to within machine precision, consistent with the previous results of this test-case initialized with a dry baroclinic wave (Lauritzen et al., 2017).

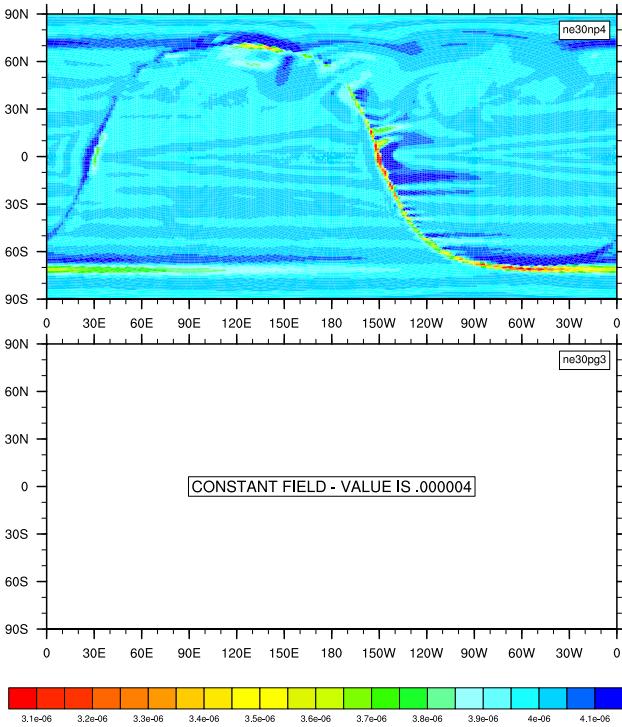


Figure 26: Results of the terminator “toy”-chemistry test. Snapshot of the total column integrated, weighted sum of the species, $\langle Cl_y \rangle = \langle Cl \rangle + \langle 2Cl_2 \rangle$, in kg/kg, at day 15 of the moist baroclinic wave test. (Top) CAM-SE, (Bottom) CAM-SE-CSLAM.

4.4.2 Aqua-planets

Two year long aqua-planet simulations are performed using CAM-SE and CAM-SE-CSLAM, using the CAM4 physics package (Neale et al., 2010), as discussed in Section 4.2. Away from the grid-scale, the mean states in the two models are very similar. Figure 27 shows the zonal-mean climatological precipitation rates in CAM-SE and CAM-SE-CSLAM. Considering how sensitive this aqua-planet configuration is to design choices in CAM-SE (Lauritzen et al., 2018), it is somewhat unexpected that the zonal means look so similar

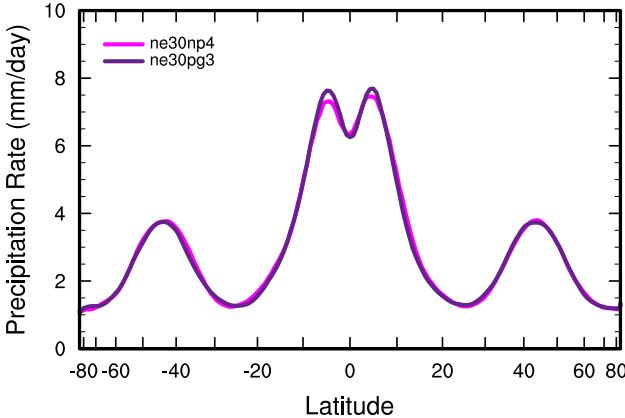


Figure 27: Climatological zonal-mean total precipitation rate in the aquaplanets, computed from a pair of year long simulations.

to one another.

A plot similar to Figure 22a is constructed for the CAM-SE-CSLAM simulation, a probability density distribution of upward ω conditionally sampled based on location within the element. Like Figure 22a, Figure 22b divides up the control volumes by corner, edge and interior cells. Through the use of the quasi-equal area physics grid, the dynamical core state appears more or less independent of location within the element, a marked improvement over CAM-SE. Since the state is approximately independent of in-element location, it follows that the physics forcing, which is evaluated from the dynamical core state, may be expected to also show an improvement in grid-imprinting.

The low-level, mean and variance of the temperature tendencies from the physics, on the GLL grid, $f_T^{(gll)}$, in the two simulations are shown in Figure 28. The mean states in the two models resemble one another, consistent with the zonal mean precipitation rates (Figure 27). The mean physics tendencies

contains modest grid imprinting in CAM-SE (barely visible near the storm-track regions), while in the variance field, grid imprinting is both ubiquitous and unmistakable. The variance is larger on boundary nodes, manifesting as a ‘stitching’ pattern resembling the cube-sphere grid. In CAM-SE-CSLAM, the grid imprinting is all but eliminated based on the mean and variance of the physics tendencies (Figure 28), consistent with our expectation.

The global mean and variance of the low-level physics tendencies are marginally lower in CAM-SE-CSLAM compared with CAM-SE on the GLL grid (by about 1% and 6% for the mean and variance, respectively; Figure 28). While these differences may be small, and potentially insignificant, they are consistent with the state on the GLL grid in the two simulations. Through re-creating Figure 22a, but using the ω field on the GLL grid in the CAM-SE-CSLAM run, the frequency of large magnitude ω values (less than -1.0 Pa/s) associated with interior, corner and edge nodes is slightly lower (not shown). This suggests that the lower magnitude physics forcing in CAM-SE-CSLAM impacts the state on the GLL grid, albeit modestly. Therefore the lower frequency of large magnitude ω in CAM-SE-CSLAM (Figure 22) may not be solely due to the smoothing effect of integrating the basis functions over control volumes, but also the lower magnitude physics tendencies feeding back onto the dynamical state.

As stated in Section 4.3, the mapping of the state to the physics grid and the reverse interpolation of physics tendencies to the GLL grid is not total energy conserving. CAM has a global energy fixer (Williamson et al., 2015) which can be used to estimate the errors associated with the mapping algorithms. To

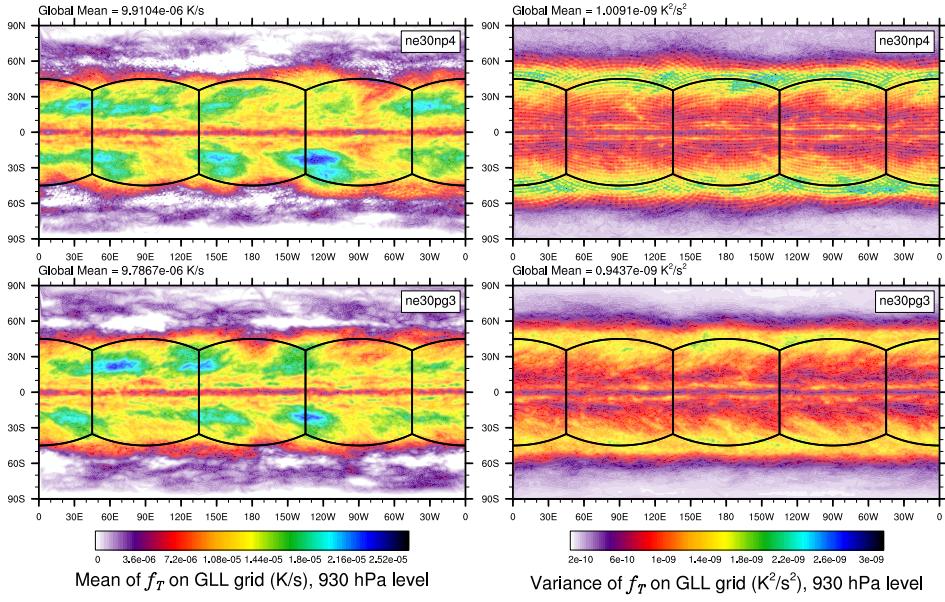


Figure 28: Mean (left) and variance (right) of the low level temperature tendencies from the physical parameterizations on the GLL grid, with the *ne30np4* configuration, (top row) and *ne30pg3* configuration (bottom row), in a pair of year-long aqua-planet simulations. Grid imprinting is observed along the element boundaries in *ne30np4*, but is absent from the *ne30pg3* simulation.

do so, it is presumed that there are no compensating mapping errors in going to and from the physics and dynamics grids, and that CAM-SE-CSLAM and CAM-SE have the same energy dissipation rates. Under these assumptions the spurious globally integrated total energy errors due to the mapping algorithm is estimated to be approximately 0.0025 W/m^2 in the aqua-planet simulations. In comparison, the dynamical core total energy dissipation is on the order of 0.1 W/m^2 (Lauritzen et al., 2018).

4.4.3 Held-Suarez with Topography

Grid imprinting associated with the flow around obstacles is more problematic than that encountered on the aqua-planets. In order to diagnose grid imprinting due to topographic flow, an idealized Held-Suarez configuration (Held and Suarez, 1994) is outfitted with real world topography after Fox-Rabinovitz et al. (2000); Baer et al. (2006), and run for two years. Figure 29 shows the mean ω at two different vertical levels in the middle troposphere. The data are presented as a raster plot on their respective unstructured grids, in order to delineate whether a particular value is associated with an interior, edge or element boundary node.

At higher latitudes (e.g., the southern Andes), the flow is smooth, conforming reasonably to the underlying topography. At lower latitudes, over the Andes (between the equator and 20° S) or the Himalayas (from 20° N to 30° N), there is a clear preference for extrema to occur at the element boundaries (Figure 29). The vertical structure of ω in regions of strong grid-imprinting indicates full-troposphere upward/downward motion (not shown). Grid imprinting is therefore more common in regions of weak stratification, such as occurs in the deep tropics, with forced up-slope flow facilitating the release of gravitational instability. Resolved updrafts/downdrafts often align with the element boundaries due to its systematically tighter pressure gradients.

Through the use of the quasi-equal area physics grid, grid imprinting due to topographic flow is reduced (Figures 29). The native topography lives on the physics grid, and the topography is mapped to the nodal points at run-time in CAM-SE-CSLAM. Mapping topography to the quadrature nodes ensures

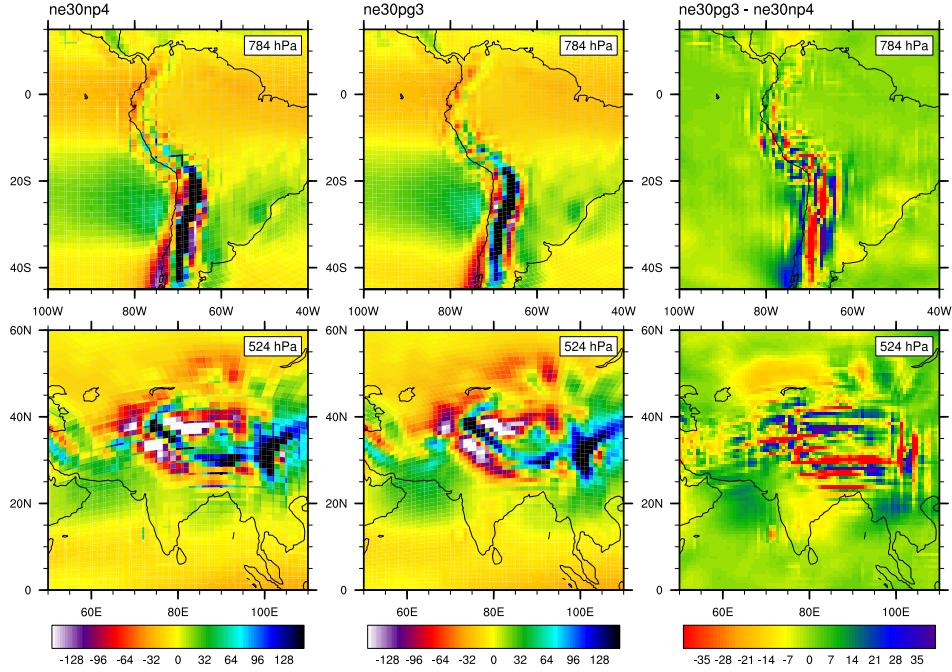


Figure 29: Mean ω at two model levels in the middle troposphere, in a Held-Suarez configuration outfitted with real world topography. (Left) CAM-SE state on the GLL grid, *ne30np4*, (Middle) CAM-SE-CSLAM state on the physics grid, *ne30pg3* and (Right) their differences computed through bi-linear interpolation to a common latitude-longitude grid. The ω fields are computed from a 1200 day Held-Suarez simulation. The data are contoured according to a ‘cell fill’ approach, in which the coupler grids (e.g, Figure 23) are used to delineate the vertices of the control volumes.

that no new extrema will be introduced to the boundary nodes, where the solution is least smooth. This effect can not be very large, since grid noise over topography is similar in CAM-SE and CAM-SE-CSLAM on the GLL grid (not shown). From the perspective of the physics grid, CAM-SE-CSLAM clearly mitigates the influence of grid-induced extrema on the state. This can be seen by comparing Figures 29a and 10b, and their differences (Figure 29c), which shows that the largest differences coincide with the element boundaries.

The reduction in grid imprinting in this modified Held-Suarez configuration appears to be almost entirely a result of the smoothing effect of integrating the basis functions over the control volumes of the physics grid.

4.4.4 AMIP type simulations

A pair of 20 year-long AMIP type simulations are performed, using CAM, version 6 physics package (CAM6) and using perpetual year 2000 SST boundary conditions (*F2000climo* compset in CESM2.0; <https://doi.org/10.5065/D67H1H0V>). Figure 30 shows the climatological precipitation fields in CAM-SE (left) and CAM-SE-CSLAM (middle), and over the same mountainous regions as in Figure 29. The plots have some similar features to the ω field in the Held-Suarez runs; the greater variance at lower latitudes, and on the windward side of the mountains are broadly similar. CAM-SE-CSLAM has a lower spatial variance, e.g., the lack of extrema over the Andes at about 15° S compared to CAM-SE, and the grid-scale precipitation peak over the Himalayas at about 30° N. The difference plot (Figure 30; right panel) is more broadly populated by blue, purple and white contours, indicating that CAM-SE has, in general, larger magnitude precipitation rates over high topography. The difference plots also highlight a couple of zonally aligned strips of anomalous precipitation, in particular, near the foot of the Himalayas in CAM-SE. These bands are in the same location as the bands of precipitation identified in CAM-SE in Lauritzen et al. (2015a) (their Figure 7), but using CAM, version 5 physics, of which they argue are spurious in nature.

To assist in identifying whether a particular precipitation pattern is spuri-

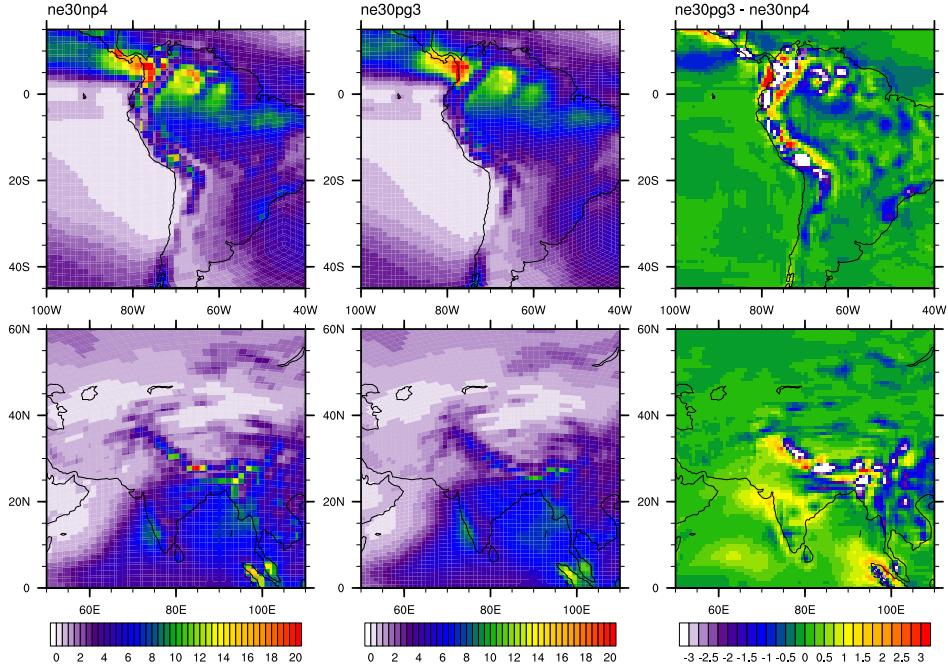


Figure 30: Climatological total precipitation rate (in mm/day) computed from the final 19 years of a pair of 20 year long AMIP type simulations. (Left) CAM-SE, (middle) CAM-SE-CSLAM and (Right) their differences.

ous, an *F2000climo* simulation is carried out using the finite-volume dynamical core that uses a regular latitude-longitude $0.9^\circ \times 1.25^\circ$ grid (CAM-FV; f09 grid; Neale et al., 2012). CAM-FV is the default low resolution model in CESM2.0, and with its smoothly varying grid, does not suffer from the Quadrature Node Problem (Section 4.2). Figure 31 shows the global precipitation fields in CAM-SE, CAM-SE-CSLAM and CAM-FV, compared to an observational dataset, the Global Precipitation Climatology Project (GPCP; 1979-2003) gridded dataset (Huffman et al., 2001). The magnitude of the precipitation rates in all three models are higher than the GPCP dataset, primarily over land in the Tropics (note the lack of red contours in the GPCP

dataset), which should be interpreted cautiously due to widely-accepted issues in constructing a reliable, gridded, global precipitation dataset. At lower latitudes, CAM-FV has lower spatial variance, and overall lower magnitudes, compared with CAM-SE. The GPCP dataset indicates that perhaps the precipitation rates in low-latitude mountainous regions in CAM-FV and CAM-SE are larger than in reality. Following suit, the reduction in magnitude and spatial variance in precipitation in these regions in CAM-SE-CSLAM may be interpreted as an improvement over CAM-SE.

4.5 Conclusions

Element-based high-order Galerkin Methods possess many of the attractive qualities recommended for next generation global atmospheric models. Among these, high-order accuracy is achieved with minimal communication between elements, allowing for near perfect scaling on massively parallel systems. Element communication amounts to a numerical flux applied to the element boundaries, reconciling overlapping solutions of adjacent elements but degrading the smoothness of the boundary nodes in the process (to C^0). For non-smooth problems, gradients are systematically tighter at the element boundaries, and local extrema often characterize the boundary nodes. This behavior is illustrated using NCAR’s Community Atmosphere Model with Spectral Elements dynamics (CAM-SE) in an aqua-planet configuration, in a Held-Suarez configuration with real-world topography and in an AMIP type configuration.

The authors argue that the conventional physics-dynamics coupling paradigm, in which the physical parameterizations are evaluated on the dynamical core

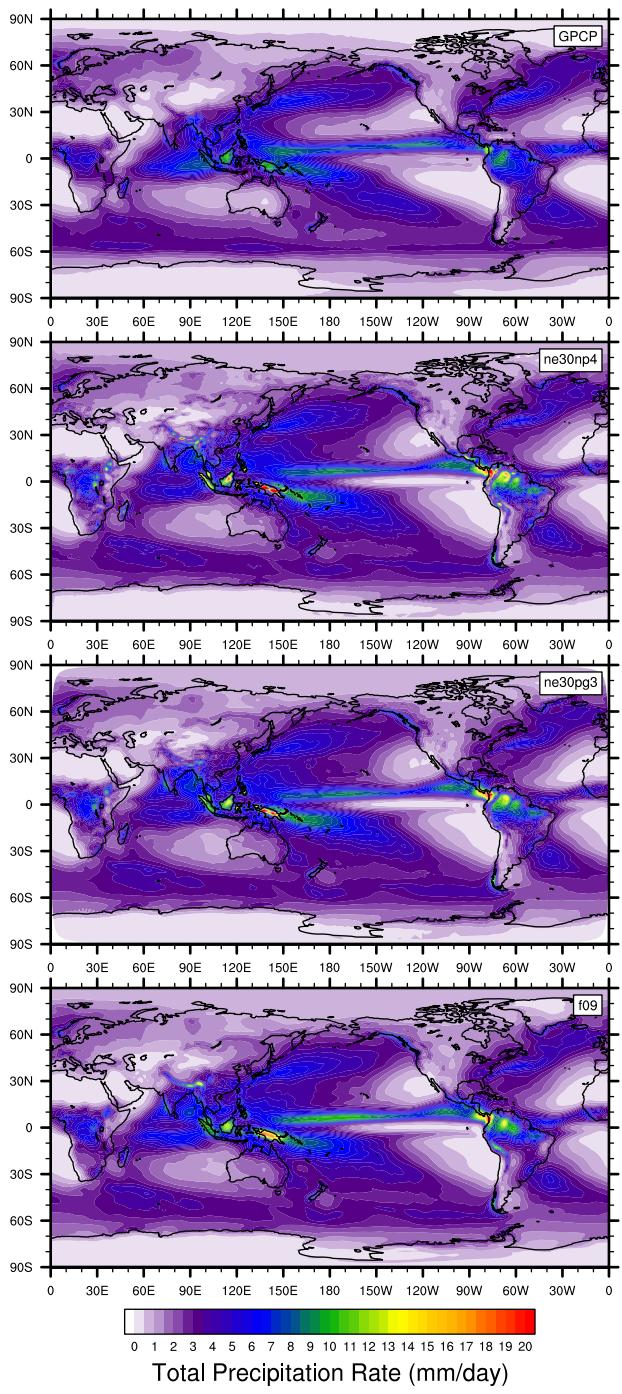


Figure 31: Climatological total precipitation rate computed from the final 19 years of a suite of 20 year long AMIP simulations, using CAM-SE (ne30np4), CAM-SE-CSLAM (ne30pg3) and CAM-FV (f09). The top plot is an observational product, the gridded GPCP climatological precipitation dataset.

grid, exacerbates grid imprinting. A separate physics grid is proposed and implemented in CAM-SE, and referred to as CAM-SE-CSLAM, through dividing the elements into quasi-equal areas with equivalent degrees of freedom. The state is mapped to the physics grid with high-order accuracy through integrating CAM-SE’s Lagrange basis functions over the control volumes. Control volumes near element boundaries now represent a state in the vicinity of the extrema produced through the boundary exchange, as opposed to the the nodal value itself. These control volumes are also compatible with a ‘large-scale state’ as required by the physical parameterizations. The physical parameterizations are evaluated on the finite volume grid, and the forcing terms are mapped back to the dynamical core grid using a cubic tensor-product Lagrange interpolation. In aqua-planet simulations, evaluating the parameterizations on the physics grid removes any obvious dependence of proximity to the element boundary, resulting in a more realistic state with negligible grid imprinting. The mapping algorithm does not conserve total energy, but it is estimated that these errors are one to two orders of magnitude less than the total energy dissipation from the dynamical core.

In CAM-SE-CSLAM, the physics grid replaces the default CAM-SE quadrature point-based coupler grid (Figure 23) to compute fluxes between model components in the Community Earth System Model (CESM). The appeal here is two-fold. Through integrating the Lagrange basis functions over control volumes, one can be certain that the fluxes computed from this grid are a volume averaged flux. The same can not be said for CAM-SE, where artificial control volumes (with sizes proportional quadrature weights) are constructed around

nodal values and assumed to represent the volume averaged state. The second advantage of the new coupler grid is that extrema occurring on boundary nodes may no longer influence other model components in simulations without rough topography. While grid imprinting is effectively eliminated in the aquaplanets, experiments with real-world topography (Held-Suarez and AMIP type configurations) reduces, but does not entirely eliminate, imprinting from the mean state. The quasi-equal area physics grid is nonetheless effective at mitigating numerical nuances associated with high-order element-based Galerkin methods, for non-smooth problems.

Future work will focus on the impact of using a coarser, $pg \times pg = 2 \times 2$ physics grid configuration. The coarser physics grid may be more effective at reducing spurious noise over regions of rough topography, while potentially reducing the computational overhead. Any advantages of using a coarser resolution physics grid will be weighed against any potential reduction in a model's effective resolution.

5 Exploring a lower resolution physics grid in CAM-SE-CSLAM

5.1 Introduction

Global atmospheric models fundamentally consist of two components. The dynamical core (*dynamics*), which numerically integrate the adiabatic equations of motion and tracer advection, and the physical parameterizations (*physics*), which compute the effects of diabatic and subgrid-scale processes (e.g., radiative transfer and moist convection) on the grid-scale. More out of convenience than anything else, the physics are evaluated on the dynamics grid, i.e., the physics and dynamics grids coincide. From linear stability and accuracy analysis of numerical methods, it is a common result that the shortest simulated wavelengths are not accurately represented by the dynamical core. Additionally, simulated downscale cascades result in an unrealistic collection of energy and/or enstrophy near the truncation scale, which may be observed from kinetic energy spectra in model simulations (Skamarock, 2011). Some form of dissipation must be incorporated into models to mitigate these numerical artifacts near the grid scale (Jablonowski and Williamson, 2011). The unrealistic nature of the grid-scale led Lander and Hoskins (1997) to speculate whether the physics should be evaluated on a grid that is more reflective of the scales actually resolved by the dynamical core.

Exploring the impact of different physics grid resolutions has so far been limited to models employing the spectral transform method (Lander and Hoskins,

1997; Williamson, 1999; Wedi, 2014). Lander and Hoskins (1997) argued that passing under-resolved states to the physics may be especially problematic in spectral transform models, since the physics are evaluated on a latitude-longitude transform grid, and contains more degrees of freedom than the spectral representation to prevent aliasing of quadratic quantities. However, Lander and Hoskins (1997) found that the spectral truncation of the physics tendencies damps errors that may result from passing an under-resolved state to the physics, although the extent to which these errors may still be present in the model is difficult to address.

Another class of spectral transform models evaluate the quadratic terms using semi-Lagrangian methods, which are implicitly diffusive, relaxing constraints on the resolution of the transform grid. Wedi (2014) experimented with different transform grid resolutions and concluded that the standard high resolution quadratic grid actually improves forecast skill over the use of a lower-resolution transform grid. They suggest that increasing the resolution of the transform grid simulates a kind of sub-grid variability on the spectral state, which is thought to be under-represented in global atmospheric models (Shutts, 2005). This is in principle the purpose of “super-parameterization,” in which a cloud resolving model is embedded in each grid cell to approximate sub-grid variability, and improves both diurnal and sub-seasonal variability in the model (Randall et al., 2003).

After the physics tendencies are transformed into spectral space, it is possible to truncate the tendencies at any particular wave number in global spectral transform models. Williamson (1999) conducted a pair of convergence tests

using a spectral transform model; a conventional convergence test and one in which the spectral truncation of the physics tendencies is held fixed and the resolution of the dynamical core increased. In contrast to the realistic weather forecasts of Wedi (2014), Williamson (1999) ran their model to equilibrium in an idealized climate configuration. When the physics and dynamics resolutions increase together, as in more typical convergence studies, the strength of the Hadley Cell increases monotonically with resolution. This sensitivity of Hadley Cell strength to horizontal resolution is a common result of global models at hydrostatic resolutions (see Herrington and Reed, 2017, and references therein). But with the truncation wave number of physics tendencies held fixed, the Hadley Cell showed very little sensitivity to dynamical core resolution, resembling the solution for which the dynamics truncation wave number is equal to that of the lower resolution physics. Herrington and Reed (2017) speculated that these results suggest the scales of motion resolved by the dynamical core may be aliased to the lower resolution physics.

Global spectral transform models, while remarkably efficient at small processor counts, do not scale well on massively parallel systems. High-order Galerkin methods are becoming increasingly popular in climate and weather applications due to their high-parallel efficiency, high-processor efficiency, high-order accuracy (for smooth problems), and geometric flexibility facilitating mesh-refinement applications (e.g., Giraldo and Restelli, 2008; Nair et al., 2009; Brdar et al., 2013, and the Energy Exascale Earth System Model; <https://e3sm.org/>). High resolution climate simulations with NCAR’s Community Atmosphere Model (CAM; Neale et al., 2012) are typically performed using

a continuous Galerkin dynamical core referred to as CAM-SE (CAM Spectral Elements; Taylor et al., 2008; Dennis et al., 2012; Lauritzen et al., 2018). CAM-SE may be optionally coupled to a conservative, semi-Lagrangian tracer advection scheme for accelerated multi-tracer transport (CAM-SE-CSLAM; Lauritzen et al., 2017). Tracer advection then evolves on an entirely separate, finite-volume grid which contains the same degrees of freedom as CAM-SE’s quadrature node grid.

Element-based Galerkin methods are susceptible to grid-imprinting, and may need be considered when contemplating a particular physics grid (Herrington et al., 2018, hereafter referred to as H18). Grid imprinting manifests at the element boundaries, since the global basis is least smooth (C^0 ; all derivatives are discontinuous) for quadrature nodes lying on the element boundaries, and the gradients (e.g., pressure gradients) are systematically tighter producing local extremes. Through computing the physics tendencies at the nodal points, element boundary extrema is also observed in the physics tendencies.

H18 has shown that through evaluating the physics on the finite-volume tracer advection grid in CAM-SE-CSLAM, element boundary errors are substantially reduced, although still problematic in regions of steep terrain, at low latitudes. Through integrating CAM-SE’s basis functions over the control volumes of the finite-volume grid, element boundary extrema is additionally weighted by the C^∞ solutions (i.e., the basis representation is infinitely smooth and all derivatives are continuous) that characterize the interior of the element, and the state is smoother. Additionally, in defining an area averaged state, the finite-volume physics grid is made consistent with assumptions inherent to

the physics, and is more appropriate for coupling to other model components (e.g., the land model), which is typically performed using finite-volume based mapping algorithms.

The CAM-SE-CSLAM finite-volume grid is defined through dividing the elements of CAM-SE’s gnomonic cubed-sphere grid with equally spaced, equi-angular coordinate lines parallel to the equi-angular element boundaries, such that there are 3×3 control volumes per element (hereafter referred to as *pg3*; see Figure 32). While the physics grid in H18 is *pg3*, i.e., the physics and dynamics grids have the same degrees of freedom, the control volumes in *pg3* encompass a region of the element in which their proximity to the element boundaries are not equal. Therefore, not every control volume in an element has the same smoothness properties. This may be avoided through defining a physics grid in which the elements are instead divided into 2×2 control volumes (hereafter referred to as *pg2*; see Figure 32). The control volumes of the *pg2* grid all have the same proximity to the element boundaries, and should mitigate the element boundary noise that remains in the *pg3* grid, and shown in H18.

In this study, we test the hypothesis that the coarser, *pg2* physics grid is effective at reducing spurious noise at element boundaries, particularly over regions of rough topography. In addition, the recent trend towards running models at ever higher resolutions is an almost prohibitive computational burden. As the physics are responsible for over half of the computational cost in CAM-SE (Lauritzen et al., 2018), the improvement in computational performance using a coarser resolution physics grid is potentially significant. However, any

advantages of using a coarser physics grid need be weighed against any potential reduction in simulation quality, e.g., possible aliasing of the resolved scales of motion by the coarser grid, as suggested by the results of Williamson (1999). Section 5.2 describes the implementation of the *pg2* grid into CAM-SE-CSLAM, and the idealized model configurations used throughout this study. Section 5.3 provides results of model simulations, to test the implementation of the mapping algorithms and identify any changes in grid imprinting, and in the resolved scales of motion, compared with the *pg3* configuration. Section 5.4 provides a discussion of the results and conclusions.

5.2 Methods

Separating dynamics, tracer and physics grids introduces the added complexity of having to map the state from dynamics and tracer grids to the physics grid; and mapping physics tracer increments back to the tracer grid and physics increments needed by the dynamical core to the dynamics grid (see Figure 32). The dynamics grid in the case of CAM-SE-CSLAM refers to the Gauss-Lobatto-Legendre (GLL) quadrature nodes used by the spectral-element method to solve the momentum equations for the momentum vector (u, v), thermodynamics equation for temperature (T), continuity equation for dry air mass ($\frac{1}{g}p$), and continuity equations for water vapor and thermodynamically and inertially active condensates (see, e.g., Lauritzen et al., 2018, for details). By tracer grid we refer to the *pg3* grid on which CSLAM performs tracer transport of water vapor, condensates and other tracers. Although water vapor and condensates are being advected by the CSLAM scheme on the

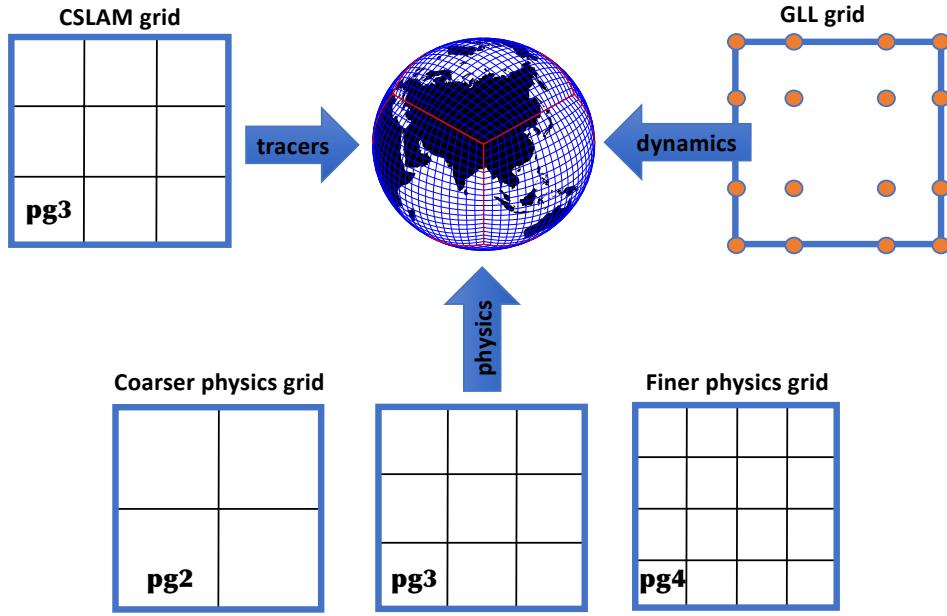


Figure 32: An overview of the different grids in CAM-SE-CSLAM.

pg3 grid, these quantities are also needed on the GLL grid for the momentum equations and thermodynamic equation. Transport of water variables is also performed by the spectral-element method on the GLL grid. To avoid decoupling of water species on the CSLAM and GLL grids, the GLL water species are overwritten by the CSLAM values every physics time-step. This is explained in detail in H18.

Similarly to the CAM-SE-CSLAM *pg3* configuration, the dynamics state (momentum vector, temperature, dry pressure) must be mapped from the *GLL* grid to the physics grid. Exactly the same algorithms as used in the *pg3* configuration apply, i.e. momentum components are interpolated by evaluating the internal Lagrange basis functions (used in the spectral-element method) at the equi-angular (gnomonic) center of the *pg2* cells and the Lagrange basis

function representations of temperature and pressure are integrated over the *pg2* control volumes. See H18 for details.

As compared to the *pg3* configuration, the extra complication with the *pg2* setup is that the tracer grid does not coincide with the physics grid, i.e. the tracer state needs to be mapped from the CSLAM grid (*pg3*) to the physics grid (*pg2*), and tracer increments computed by physics must be mapped from the physics grid back to the CSLAM grid. In order to describe the mapping algorithms between the grids some notation needs to be introduced.

The mapping algorithms are applied to each element Ω (with spherical area $\Delta\Omega$) so without loss of generality consider one element. Let $\Delta A_k^{(pg2)}$ and $\Delta A_\ell^{(pg3)}$ be the spherical area of the physics grid cell $A_k^{(pg2)}$ and CSLAM control volume $A_\ell^{(pg3)}$, respectively. The physics grid cells and CSLAM cells, respectively, span the element, Ω , without gaps or overlaps

$$\cup_{k=1}^{nphys^2} A_k^{(pg2)} = \Omega \text{ and } A_k^{(pg2)} \cap A_\ell^{(pg2)} = \emptyset \quad \forall k \neq \ell, \quad (18)$$

$$\cup_{k=1}^{nc^2} A_k^{(pg3)} = \Omega \text{ and } A_k^{(pg3)} \cap A_\ell^{(pg3)} = \emptyset \quad \forall k \neq \ell, \quad (19)$$

where $nc = 3$ is the CSLAM grid resolution parameter and $nphys = 2$ is the physics grid resolution parameter (following the Fortran code base), although the methods described here are valid for any arbitrary integer $nphys$ (e.g., $nphys = 4$ is shown in Figure 32). The overlap areas between the k -th physics grid cell and ℓ th CSLAM cell are denoted

$$A_{k\ell} = A_k^{(pg2)} \cap A_\ell^{(pg3)}, \quad (20)$$

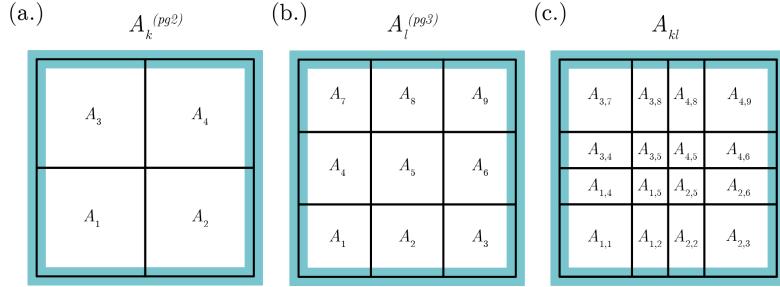


Figure 33: Indices notation for (a) the $pg2$ grid, (b) the $pg3$ grid and (c) their exchange grid.

(see Figure 33) so that

$$A_k^{(pg2)} = \cup_{l=1}^{nc^2} A_{kl}. \quad (21)$$

This overlap grid is also referred to as the *exchange grid*.

5.2.1 Mapping tracers from $A^{(pg3)}$ to $A^{(pg2)}$ (CSLAM to physics grid)

The CSLAM and physics grids are both finite-volume grids so existing CSLAM technology can be used to map the tracer state from CSLAM to physics grid. That is, compute a high-order shape-preserving reconstruction of mixing ratio m and dry air mass $\frac{1}{g}\Delta p$ per unit area in each CSLAM control volume and integrate those reconstruction functions over the overlap areas (Lauritzen et al., 2010; Nair and Lauritzen, 2010). This algorithm retains the properties of CSLAM: inherent mass-conservation, consistency (constant mixing ratio is preserved), mixing ratio shape-preservation and linear-correlation preservation.

Denote the known cell averaged values of dry pressure-level thickness and

mixing ratio $\overline{\Delta p}^{(pg3)}$ and $\overline{m}^{(pg3)}$, respectively. We consider a particular layer and for simplicity drop the layer subscript. The same procedure is applied to each layer in a column. The unknowns we would like to compute are the cell-averaged values of the same quantities on the physics grid; $\overline{\Delta p}^{(pg2)}$ and $\overline{m}^{(pg2)}$, respectively. The dry pressure level thickness integrated over the k 'th physics grid cell is given by

$$\overline{\Delta p}_k^{(pg2)} = \frac{1}{\Delta A_k^{(pg2)}} \sum_{\ell=1}^{nc^2} \langle \delta p \rangle_{k\ell}, \quad (22)$$

where $\langle \delta p \rangle_{k\ell}$ is the dry mass in a layer over overlap area $A_{k\ell}$. It is computed by integrating a high-order (2D polynomial of degree 2) reconstruction of pressure-level thickness in each CSLAM cell over the overlap area $A_{k\ell}$

$$\langle \delta p \rangle_{k\ell} = \int_{A_{k\ell}} \left[\sum_{i+j \leq 2} \mathcal{P}_{\ell}^{(ij)} x^i y^j \right] dA. \quad (23)$$

The reconstruction coefficients $\mathcal{P}_{\ell}^{(ij)}$ in CSLAM cell ℓ are computed from the cell average pressure level thicknesses on the CSLAM grid $\overline{\Delta p}^{(pg3)}$ and the numerical integration over overlap areas is done by line-integrals. The details of that are given in Lauritzen et al. (2010) and not repeated here.

The average tracer mass per unit area on the physics grid is given by

$$\overline{m \Delta p}_k^{(pg2)} = \frac{1}{\Delta A_k^{(pg2)}} \sum_{\ell=1}^{nc^2} \langle m \delta p \rangle_{k\ell}, \quad (24)$$

where $\langle m \delta p \rangle_{k\ell}$ is the tracer mass over $A_{k\ell}$ resulting from integrating a high-order reconstruction of Δp and m combined using the approach outlined in

Appendix B of Nair and Lauritzen (2010) over the overlap area $A_{k\ell}$

$$\langle m\delta p \rangle_{k\ell} = \int_{A_{k\ell}} \left[\overline{\Delta p}_\ell^{(pg3)} \sum_{i+j \leq 2} \mathcal{M}_\ell^{(ij)} x^i y^j + \overline{m}_\ell^{(pg3)} \sum_{i+j \leq 2} \widetilde{\mathcal{P}}_\ell^{(ij)} x^i y^j \right] dA, \quad (25)$$

where $\widetilde{\mathcal{P}}_\ell^{(00)} = \mathcal{P}_\ell^{(00)} - \overline{\Delta p}_\ell^{(pg3)}$ and $\widetilde{\mathcal{P}}_\ell^{(ij)} = \mathcal{P}_\ell^{(ij)}$ for $i, j > 0$, and $\mathcal{M}_\ell^{(ij)}$ are the reconstruction coefficients for the mixing ratio in CSLAM cell $A_\ell^{(pg3)}$. A shape-preserving limiter is applied to the reconstruction of mixing ratio m (Barth and Jespersen, 1989) and not Δp . This way of combining the reconstruction function for Δp and m in (25) ensures that a constant mixing ratio is preserved (consistency), tracer mass is conserved, linear-correlations are preserved and tracer shape-preservation is retained. The mixing ratio on the physics grid is then

$$\overline{m}_k^{(pg2)} = \frac{\overline{(m\Delta p)}_k^{(pg2)}}{\overline{\Delta p}_k^{(pg2)}}, \quad (26)$$

where $\overline{\Delta p}_k^{(pg2)}$ is given in (22).

Perhaps surprisingly a much more challenging problem is to map tracer increments (or state) from the physics grid to the CSLAM grid while retaining important properties such as mass-conservation, consistency, and correlation preservation. Why this mapping problem is challenging is explained in detail in Section 5.2.3 after having defined important properties for mapping physics increments/tendencies.

5.2.2 Mapping tracer increments from $A^{(pg2)}$ to $A^{(pg3)}$ (physics to CSLAM grid)

The increments from the parameterizations are computed on the physics grid. The tracer increment in physics grid cell k is denoted $\bar{f}_k^{(pg2)}$ so that the updated mixing ratio on the physics grid is $\bar{m}_k^{(pg2)} + \bar{f}_k^{(pg2)}$. The problem is how to map $\bar{f}_k^{(pg2)}$ to the CSLAM control volumes, to obtain $\bar{f}^{(pg3)}$, satisfying the following constraints:

- 1. Local mass-conservation:** At a minimum total physics mass forcing on an element computed on the physics grid should equal the element physics mass forcing on the CSLAM grid

$$\bar{f}_k^{(pg2)} \bar{\Delta p}_k^{(pg2)} \Delta A_k^{(pg2)} = \sum_{\ell=1}^{nc^2} \left[\bar{\Delta p}_{\ell}^{(pg3)} \bar{f}_{\ell}^{(pg3)} \Delta A_{k\ell} \right], \quad (27)$$

where $\bar{\Delta p}_k^{(pg2)}$ is the pressure level thickness in physics grid cell k and similarly for $\bar{\Delta p}^{(pg3)}$. We enforce a more local constraint in which only mass-increments overlapping with a particular CSLAM cell contributes to the mass-increment in that CSLAM cell.

- 2. Local shape-preservation in mixing ratio:** The increments mapped to the CSLAM grid and added to the previous CSLAM state should not produce values smaller than the updated physics grid mixing ratios, $\bar{m}_k^{(pg2)} + \bar{f}_k^{(pg2)}$, or values smaller than the existing CSLAM mixing ratios

that overlap with physics grid cell A_ℓ

$$\bar{m}_\ell^{(pg3)} + \bar{f}_\ell^{(pg3)} \geq \bar{m}_k^{(min)} = \min \left(\bar{m}_k^{(pg2)} + \bar{f}_k^{(pg2)}, \{\bar{m}_{k\ell} | \ell = 1, nc^2\} \right), \quad (28)$$

where

$$\bar{m}_{k\ell} = \frac{\langle m \delta p_{k\ell} \rangle}{\langle \delta p_{k\ell} \rangle}. \quad (29)$$

The numerator and denominator in (29) are defined in (23) and (25), respectively. In particular this means that an increment, when mapped to the $pg3$ grid, should not drive the state negative (described in detail below as the ‘negativity’ problem).

A similar definition apply for maxima

$$\bar{m}_\ell^{(pg3)} + \bar{f}_\ell^{(pg3)} \leq \bar{m}_k^{(max)} = \max \left(\bar{m}_k^{(pg2)} + \bar{f}_k^{(pg2)}, \{\bar{m}_{k\ell} | \ell = 1, nc^2\} \right), \quad (30)$$

3. Linear correlation preservation: The physics forcing must not disrupt linear tracer correlation between species on the CSLAM grid (see, e.g., Lauritzen and Thuburn, 2012), i.e. if two tracers are linearly correlated and the physics increment preserves linear correlations on the physics grid then the tracer increment on the CSLAM grid must not disrupt linear correlations.

4. Consistency: A non-zero constant mixing ratio increment from physics, $cNST$, on the physics grid, $\bar{f}_k^{(pg2)} = cNST \forall k$, must result in the same (constant) forcing on the CSLAM grid, $\bar{f}_\ell^{(pg3)} = \bar{f}_k^{(pg2)} = cNST \forall \ell$.

To motivate the algorithm that will simultaneously satisfy 1-4 it is informative to discuss how ‘standard’ mapping algorithms will violate one or more of the constraints:

5.2.3 Why ‘conventional’ conservative remapping will not work

It is helpful to analyze in detail why conventional remapping can not satisfy properties 1-4 above. Assume that one remaps the mass-increments in exactly the same way as the mapping of mixing ratio state from the CSLAM grid to the physics grid described in section 5.2.1. That is, replace m with f and map from physics grid to the CSLAM grid instead of the other way around.

Denote the mapped mass-increment $\widetilde{f\Delta p}^{(pg3)}$ and due to the properties of the mapping algorithm the mass-increment is conserved, linear correlation between mass-increments are conserved and shape in mass-increment is preserved. The problems arise when converting from mass to mixing ratio.

Conserve mass but not consistency If ones uses the known pressure-level thickness on the CSLAM grid $\overline{\Delta p}_k^{(pg3)}$ to convert from mass-increment to mixing-ratio increment

$$\overline{m}_k^{(pg3)} = \frac{\widetilde{f\Delta p}_k^{(pg3)}}{\overline{\Delta p}_k^{(pg3)}}, \quad (31)$$

a constant mixing ratio increment is not conserved. Basically the constant increment mapped to the CSLAM grid and converted to mixing ratio increment through (31) will, rather than being constant, reflect the spurious discrepancy between $\widetilde{\Delta p}_k^{(pg3)}$ and $\overline{\Delta p}_k^{(pg3)}$, where $\widetilde{\Delta p}_k^{(pg3)}$ is the pressure-level thickness mapped from the $pg2$ grid to the $pg3$ grid. That said, mass will be conserved

since the dynamical core state has $\overline{\Delta p}_k^{(pg3)}$ (unless the increment drives the mixing ratio negative - described in detail below).

Consistent but not mass-conserving Rather than converting to mixing ratio using $\overline{\Delta p}_k^{(pg3)}$, a constant increment can be preserved by using

$$\overline{m}_k^{(pg3)} = \frac{\widetilde{f} \overline{\Delta p}_k^{(pg3)}}{\widetilde{\Delta p}_k}, \quad (32)$$

instead. But now mass-conservation is lost since, again, $\widetilde{\Delta p}_k^{(pg2)} \neq \overline{\Delta p}_k^{(pg2)}$. This issue is similar to the mass-wind inconsistency found in specified dynamics applications (e.g. Jöckel et al., 2001; Lauritzen et al., 2011b).

The ‘negativity’ problem and linear correlations Even if one could derive a reversible map for mapping $\overline{\Delta p}^{(pg2)}$ from the physics grid to the CSLAM grid, there could still be problems if the increment drives the mixing ratios negative (or overshooting occurs) on the CSLAM grid. This can easily happen for tracers, such as cloud liquid amount and cloud ice amount, that are zero in most of the domain and non-zero in localized areas/points (where there are clouds). We refer to this as the ‘negativity problem’. This problem is depicted schematically in Figure 34. Consider a single element of CSLAM control volumes, containing only a single cell with mixing ratio 1.0, and 0.0 everywhere else ($\overline{m}_\ell^{(pg3)}$; Figure 34a). The mixing ratios are mapped to the pg2 grid using, for simplicity, the piecewise constant method where a constant value inside the pg2 cells is used during the integration over overlap cells ($\overline{m}_k^{(pg2)}$; Figure 34b).

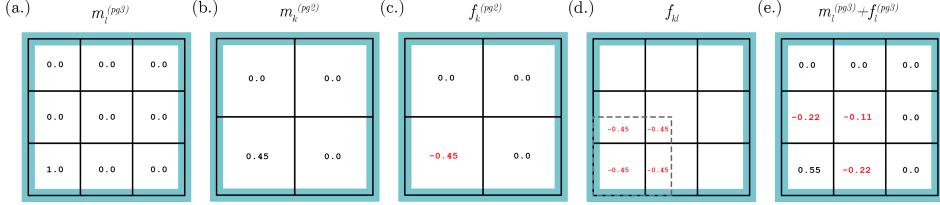


Figure 34: Schematic illustration of the ‘negativity problem’ in a single element. (a.) Initial CSLAM tracer values, (b.) mapped to $pg2$, (c) produces a tracer increment on $pg2$, (d.) with negative increments on the exchange grid overlying CSLAM cells in (a) that were initially zero and (e) driving those mixing ratios negative.

Now consider the case in which physics removes all the mass from the physics cell k : $\bar{f}_k^{(pg2)} = -\bar{m}_k^{(pg2)}$ (Figure 34c). The tracer increment is mapped from $pg2$ to $pg3$ using the piecewise constant method. Some of the non-zero increments are now in overlap areas where the original CSLAM grid cells have mixing ratio zero (\bar{f}_{kl} ; Figure 34d), and hence, the state is driven negative when adding the overlap increment to the CSLAM state (Figure 34e). This is referred to as the negativity problem although it can also happen for maxima.

The negativity issue could be avoided if one remaps the physics updated state instead of mapping increments/tendencies. In that case a shape-preserving filter will make sure that the state on the CSLAM grid is not negative (and does not overshoot). That said, if physics does not change the state and it is mapped back to the CSLAM grid then spurious tendencies (proportional to the errors introduced by mapping state from the CSLAM grid to the physics grid and back again) are introduced. Hence it is advantageous to map increments/tendencies since any reasonable algorithm will preserve a zero function.

As illustrated above a standard remapping method will NOT simultaneously satisfy 1-4 and hence a new algorithm has been derived.

5.2.4 New tendency mapping algorithm

The problem is how to map the mass-increment on the physics grid, $\bar{f}^{(pg2)} \Delta A^{(pg2)}$, to the CSLAM cells that overlap with $\Delta A^{(pg2)}$. To maintain shape-preservation, linear correlations and to avoid the negativity problem locally, it is advantageous to define a mass excess function on the exchange grid $\Delta m_{k\ell}^{(excess)}$. It is basically the maximum amount of mixing ratio that can be removed (in the case $\bar{f}^{(pg2)} < 0$) without producing new minima in the exchange grid mixing ratio $m_{k\ell}$

$$\Delta m_{k\ell}^{(excess)} = \bar{m}_{k\ell} - \bar{m}_k^{(min)}, \quad (33)$$

where $\bar{m}_{k\ell}$ is defined in (29). So the maximum amount of mass that we can be removed from the exchange grid cells that span physics grid cell A_k without violating the shape-preservation constraint ((28) and (30)) is

$$\sum_{\ell} \Delta m_{k\ell}^{(excess)} \bar{\Delta p}_{k\ell} \delta A_{k\ell}. \quad (34)$$

If physics is designed not to remove more mass than available in A_k (which should be the case for a carefully designed physics package) then it is guaranteed that

$$\sum_{\ell} \Delta m_{k\ell}^{(excess)} \bar{\Delta p}_{k\ell} \delta A_{k\ell} \geq \bar{f}^{(pg2)} \Delta p_k \Delta A^{(pg2)}. \quad (35)$$

We distribute the physics mass-forcing (assuming $\bar{f}^{(pg2)} < 0$) according to the mass excess in each overlap area by solving this equation for γ_k

$$\Delta A_k^{(pg2)} \bar{\Delta p}_k^{(pg2)} \bar{f}^{(pg2)} = \gamma_k \sum_{\ell} \Delta m_{k\ell}^{(excess)} \bar{\Delta p}_{k\ell} \delta A_{k\ell}, \quad (36)$$

and add mass increment (which in this case is negative)

$$\gamma_k \Delta m_{k\ell}^{(excess)} \bar{\Delta p}_{k\ell} \delta A_{k\ell}, \quad (37)$$

to the ℓ th CSLAM cell state $\bar{m}^{(pg3)} \bar{\Delta p}_{\ell}^{(pg3)} \Delta A_{\ell}^{(pg3)}$. This process is repeated for all physics cells. Note that this problem is a well-posed, i.e. $\gamma_k > 0$, since physics will not remove more mass than is locally available (35). The way in which the mass-forcing is distributed to the CSLAM cells using the excess function insures that the negativity problem is avoided. Mass is conserved by design and shape-preservation is obtained by using the excess function.

If the physics increment is positive (assuming $\bar{f}^{(pg2)} > 0$) we define a ‘lack’ function

$$\Delta m_{k\ell}^{(lack)} = \bar{m}_{k\ell} - \bar{m}^{(max)}, \quad (38)$$

and solve

$$\bar{\Delta p}_k^{(pg2)} \bar{f}^{(pg2)} \Delta A_k^{(pg2)} = \gamma_k \sum_{\ell} \left[\Delta m_{k\ell}^{(lack)} \bar{\Delta p}_{k\ell} \delta A_{k\ell} \right], \quad (39)$$

for γ_k and follow the same procedure as for mass excess. Since positive and negative forcing is treated in exactly the same way, linear correlations are preserved. Note how the definition of the excess/lack function insures linear

correlation preservation; for example, if one would prevent negative values and not do anything about overshoots then linear correlations would not be preserved since the minima and maxima are not treated in the same way.

While the above algorithm satisfies properties 1-4 in section 5.2.2, it is not a high-order algorithm in terms of formal accuracy. This is illustrated in Figure 50 (row 3) where a smooth analytical tendency (approximate spherical harmonic of order 32 and azimuthal wave number 16; Jones, 1999)

$$f^{(pg2)} = \frac{1}{2} + \frac{1}{2} \cos(16\lambda) \sin(2\theta)^{16}, \quad (40)$$

where (λ, θ) is latitude-longitude, is mapped from $pg2$ to $pg3$ grid using this algorithm assuming $m_\ell^{(pg3)} = 0$, $\forall \ell$. The errors in the mapping are not always aligned with large gradients in the analytical function as would be expected for a ‘traditional’ interpolation algorithm. The errors are maximum on the order of 60%. To reduce errors we therefore perform a higher-order pre-allocation of tendencies that is not mass-conserving but satisfies properties 2,3, and 4 in Section 5.2.2.

5.2.5 High-order (non-conservative) pre-allocation of tracer tendencies

A high-order tracer mass increment in overlap area $A_{k\ell}$ can be computed using the following formula

$$\langle f \delta p \rangle_{k\ell} = \int_{A_{k\ell}} \left[\overline{\Delta p}_\ell^{(pg3)} \sum_{i+j \leq 2} \mathcal{F}_k^{(ij)} x^i y^j + \overline{f}_k^{(pg2)} \sum_{i+j \leq 2} \widetilde{\mathcal{P}}_\ell^{(ij)} x^i y^j \right] dA, \quad (41)$$

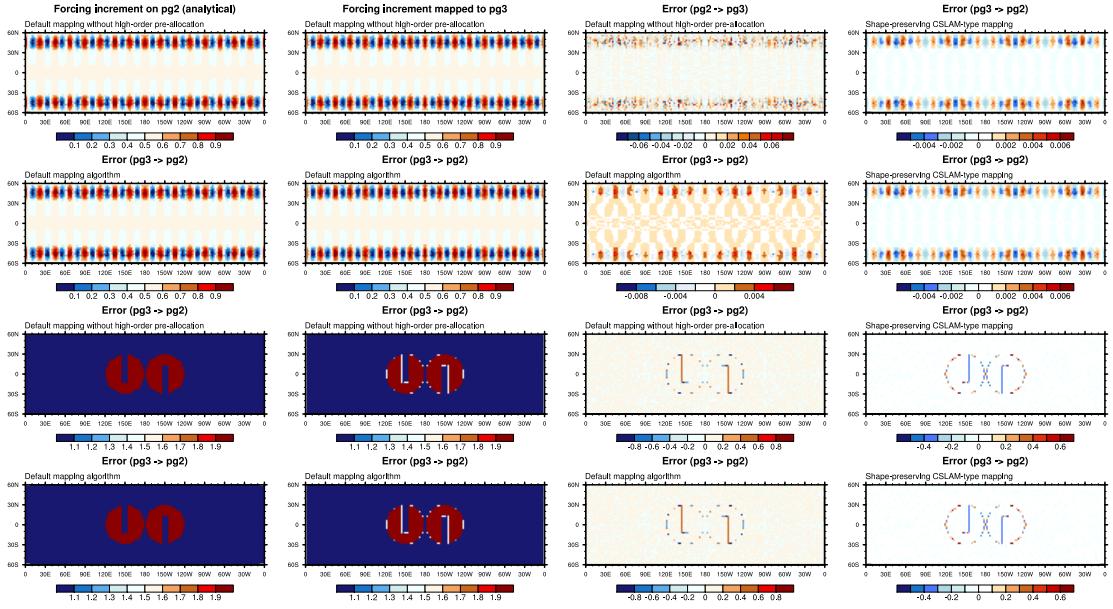


Figure 35: Mapping of idealized functions from $pg2$ (column 1) to $pg3$ (columns 2) and errors (column 3) using the new tendency algorithm only (row 1 and 3) and new tendency algorithm with the high-order pre-allocation (row 2 and 4). Column 4 shows the errors in mapping the same distributions from $pg3$ to $pg2$ using traditional remapping (CSLAM technology).

where $\mathcal{F}_k^{(ij)}$ is the forcing increment reconstruction coefficients in the k th physics grid cell and $\bar{f}_k^{(pg2)}$ is the average physics increment in the k th physics grid cell. Note that we are using the known dry pressure reconstruction coefficients on the $pg3$ grid instead of reconstructing sub-grid-scale pressure variations from the physics grid cell averaged values. We can do that since the dry pressure is not modified by physics. This highlights the importance of a dry-pressure formulation of the dynamical core when separating physics and dynamics grids (Lauritzen et al., 2018). If the physics forcing is constant then $\langle f\delta p \rangle_{kl}$ exactly equals $\langle \delta p \rangle_{kl}$ from (23); in other words, the mapping is de-

signed to be reversible in dry pressure. The physics increment in terms of mixing ratio change is given by

$$\bar{f}_{k\ell} = \frac{\langle f \delta p \rangle_{k\ell}}{\langle \delta p \rangle_{k\ell}}, \quad (42)$$

where the denominator is given by (23).

Shape-preservation, as defined by (28) and (30), is enforced by eliminating under and overshoots on the exchange grid by modifying the forcing increment $\bar{f}_{k\ell}$ so that shape-preservation is not violated in the overlap areas²

$$\bar{m}_k^{(min)} \leq \bar{m}_{k\ell} + \tilde{\bar{f}}_{k\ell} \leq \bar{m}_k^{(max)}. \quad (44)$$

While this algorithm preserves linear correlations, shape, and is consistent, is it not mass-conservative. Hence the remaining physics increment not allocated in the algorithm above is allocated using the new tendency algorithm described in Section 5.2.4.

Combining the high-order pre-allocation algorithm with the new tendency algorithm (which in this case can also be considered as a mass-fixer that does not disrupt correlation-preservation, shape and consistency) leads to an order-of-magnitude reduction in mapping errors for a smooth function (see Figure 50 row 3 and 4) while full-filling the mass-conservation, shape-preservation, linear

²In the computation of $\bar{m}_{k\ell}$ there can be small overshoots and undershoots (due to numerical integration errors) compared to the CSLAM cell average values $\bar{m}_\ell^{(pg)}$ that it overlaps with so we set

$$\bar{m}_k^{(min)} = \min \left(\bar{m}_k^{(min)}, \left\{ \bar{m}_\ell^{(pg)} \mid \ell = 1, nc^2 \right\} \right) \quad (43)$$

correlation and consistency constraint. Mass and linear correlation preservation is illustrated in the baroclinic wave test with terminator chemistry test in Section 5.3.1. Shape-preservation and consistency is demonstrated in an idealized mapping test where a smooth function, see (40), and a slotted-cylinder (see equation 12 in Lauritzen et al., 2012) are mapped to/from the *pg2* and *pg3* grids. Since the background value in the mapping of the slotted-cylinder field is preserved the mapping algorithm is consistent. Since no new over- and undershoots are produced (particularly obvious in the mapping of the slotted cylinders) the mapping is shape-preserving. We also note that the mapping errors with the default algorithm (higher-order pre-allocation with new tendency algorithm) are similar to the errors in mapping the same field from *pg3* to *pg2* using traditional remapping with CSLAM technology (column 4 in Figure 50).

5.2.6 Model Configurations

All simulations in this study are run on the Cheyenne supercomputer hosted at the NCAR-Wyoming Supercomputer Center (Computational and Information Systems Laboratory, 2017). Three model component sets (*compsets*) in the Community Earth System Model, version 2.1 (CESM2.1; <https://doi.org/10.5065/D67H1H0V>) are chosen to carry out the objectives discussed in Section 5.1. The least complex compset is a moist baroclinic wave test using a simple, Kessler microphysics scheme (*FKESSLER* compset; Lauritzen et al., 2018). The baroclinic wave setup is primarily used to evaluate the new mapping algorithms and their ability to preserve linear-correlations between two reactive tracers. The role of topography is investigated using a dry Held-

Suarez configuration (*FHS*94 compset; Held and Suarez, 1994) modified to include real world topography. H18 indicate that this configuration tends to have more grid-noise over steep terrain than in a more complex configuration using CAM, version 6 physics [CAM6; https://ncar.github.io/CAM/doc/build/html/users_guide/index.html], and is therefore a conservative choice for evaluating any change in grid imprinting between *pg3* and *pg2*.

To understand whether the resolved scales of motion are influenced by a coarser resolution physics grid, a suite of aqua-planet simulations (Neale and Hoskins, 2000; Medeiros et al., 2016) are carried out over a range of spectral-element grid resolutions, using CAM6 physics (*QPC*6 compset). The aqua-planet is an ocean covered planet in perpetual equinox, with fixed, zonally-symmetric sea surface temperatures idealized after present day Earth (*QOBS* in Neale and Hoskins, 2000). While the dynamics time-step, Δt_{dyn} , varies with resolution according to a CFL criterion, there is no established standard for how the physics time-step, Δt_{phys} , should vary across resolutions. This is further complicated by several studies indicating a high sensitivity of solutions to Δt_{phys} in CAM (Williamson and Olson, 2003; Williamson, 2013; Wan et al., 2015; Herrington and Reed, 2018).

Here, a scaling for Δt_{phys} across resolutions is proposed, based on results of the moist bubble test (Herrington and Reed, 2018) using CAM-SE-CSLAM and detailed in Appendix 8.2. The basis for the scaling is to alleviate truncation errors that arise in the moist bubble test when Δt_{phys} is too large. The

scaling is linear in grid-spacing,

$$\Delta t_{phys} = \Delta t_{phys,0} \times \frac{N_{e,0}}{N_e} s, \quad (45)$$

where $\Delta t_{phys,0}$ is taken to be the standard 1800s used in CAM-SE-CSLAM at low resolution, $N_{e,0} = 30$ (equivalent to a dynamics grid-spacing of 111.2km). N_e refers to the horizontal resolution of the grid; each of the six panels of the cubed-sphere are divided into $N_e \times N_e$ elements. Throughout the paper, spectral-element grid resolutions are denoted by an ne followed by the quantity N_e , e.g., $ne30$.

The only other parameter varied across resolutions modulates the strength of explicit numerical dissipation. The spectral element method is not implicitly diffusive, so fourth-order hyper-viscosity operators are applied to the state to suppress numerical artifacts. The scaling of the hyper-viscosity coefficients, ν , across resolutions is defined as,

$$\nu_T = \nu_{vor} = 0.30 \times \left(\frac{30}{N_e} 1.1 \times 10^5 \right)^3 \frac{m^4}{s}, \quad (46)$$

$$\nu_p = \nu_{div} = 0.751 \times \left(\frac{30}{N_e} 1.1 \times 10^5 \right)^3 \frac{m^4}{s}, \quad (47)$$

where subscripts T , vor , p , div refer to state variables the operators are applied to, temperature, vorticity, pressure and divergence, respectively. The exponent in equation (47) reduces the coefficient by about³ an order of mag-

³This is approximate. To reduce the coefficients by exactly an order of magnitude for each doubling of the resolution, the exponent should be $\frac{\ln 2}{\ln 10} \approx 3.01029$, which it has been updated to in the most recent version of CESM2.1

nitude for each doubling of the resolution (as in Lauritzen et al., 2018). No explicit dissipation of tracers (e.g., water vapor) is required since the semi-Lagrangian numerics in CSLAM are diffusive.

5.3 Results

5.3.1 Mass Conservation and Linear-Correlation Preservation

To illustrate how different the solutions look using the coarser resolution physics grid, Figure 36 shows a snapshot of the cloud liquid field of the moist baroclinic wave test on day 10, in the *ne30pg3* and *ne30pg2* configurations. The cloud liquid fields show in detail clouds forming at wave fronts. As expected, the *pg2* grid looks slightly coarser than *pg3* due to its larger control volumes. Despite this, the details of the wave patterns look reasonably similar to one another.

The models ability to preserve linear correlations is assessed using the idealized Terminator "Toy" Chemistry test (Lauritzen et al., 2015b, 2017). The tests consists of two reactive species undergoing photolysis as they are advected over the terminator line. The flow field is provided by the moist baroclinic waves test. The model is initialized with species such that their weighted sum Cl_y is a constant, i.e., $Cly = Cl + 2Cl_2 = 4 \times 10^{-6} \text{ kg kg}^{-1}$. If linear-correlations are preserved, than the column integrated weighted sum of the species, $\langle CLy \rangle$, is constant.

H18 had shown that in the *ne30pg3* configuration, $\langle CLy \rangle$ on day 15 of the terminator test is everywhere $4 \times 10^{-6} \text{ kg kg}^{-1}$, to within machine precision. While the *pg3* to *pg2* mapping algorithm in theory preserves linear correla-

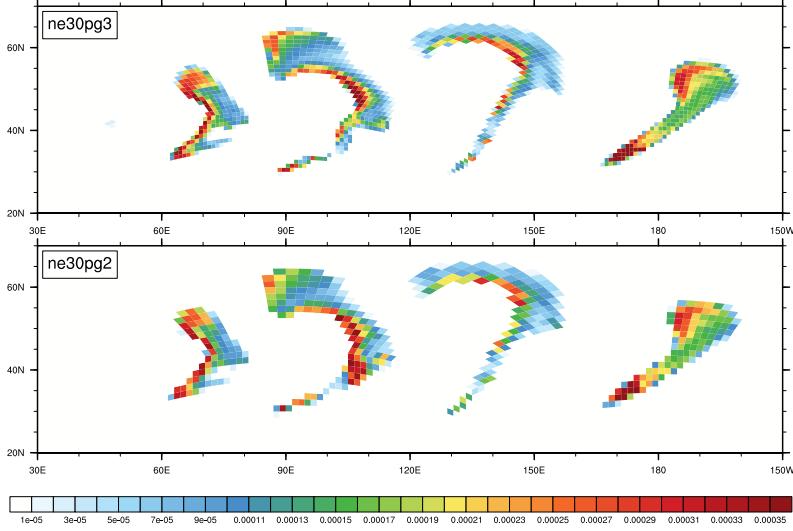


Figure 36: Snapshot of the cloud liquid field in kg kg^{-1} near the 700hPa level, on day 10 of the moist baroclinic wave test in the *ne30pg3* and *ne30pg2* configurations, displayed on the upper and lower panels, respectively. The fields are shown as a raster plot on their respective physics grids.

tions to machine precision, we found larger than round-off errors in *pg2*, likely due to *if*-logic with machine dependent thresholds in the implementation of the algorithm. Figure 37 shows $\langle CLy \rangle$ on day 15 in the *ne30pg2* configuration, which has a minimum value of $3.99936 \times 10^{-6} \text{ kg kg}^{-1}$, corresponding to a maximum relative error of 0.016%. For comparison, another terminator test is performed with the equivalent dynamics grid resolution using CAM-SE (*ne30np4*), in which tracers are advected using the spectral element method. The maximum relative error in this configuration is 31.6%, three orders of magnitude greater error than the *ne30pg2* configuration.

Tracer mass conservation is analyzed in a pair of *ne30pg2* and *ne30pg3* aqua-planet simulations, following the method of Lauritzen and Williamson

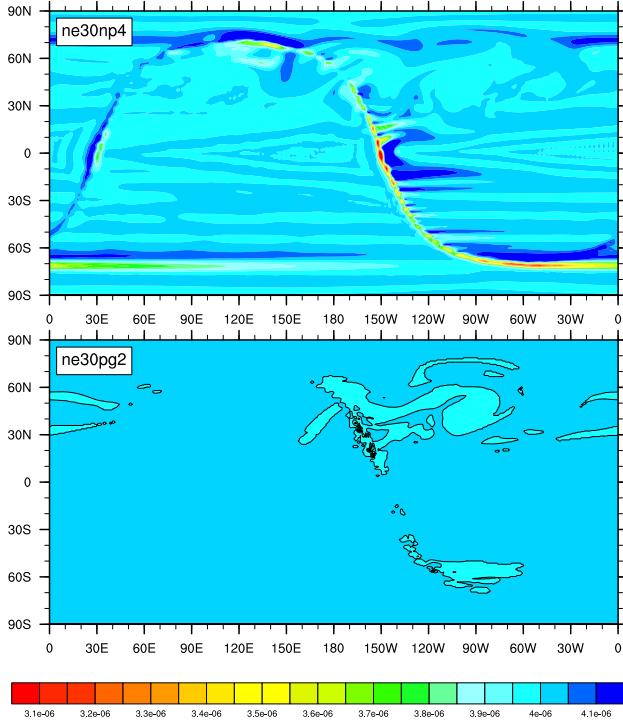


Figure 37: $\langle CLy \rangle$ in kg kg^{-1} on day 15 of the moist baroclinic wave test in the *ne30np4* and *ne30pg2* configurations, displayed on the upper and lower panels, respectively. The lower panel has a single contour level of $3.999\text{E-}6 \text{ kg kg}^{-1}$ corresponding to a relative error of 0.025%.

(2019). Energy and mass conservation due to a particular model process is assessed by model state I/O before and after each sub-process in the model. The loss of water vapor mass due to the mapping algorithms in the *ne30pg2* configuration is estimated as $1.184\text{E-}16 \text{ Pa}$ per time-step, computed as the difference between the the column integrated, global mean climatological water vapor pressure increment on the physics grid and on the tracer grid. This small error is effectively zero to within machine precision, and similar to an equivalent calculation in the *ne30pg3* simulation of $2.171\text{E-}17 \text{ Pa}$ per time-

step, which contains no mapping errors since the physics and tracer grids coincide. Negligible mapping error in the *ne30pg2* configuration is primarily a result of solving equations (36),(39) for γ_k to circumvent the ‘negativity’ problem. Re-running the *ne30pg2* aqua-planet simulation without this mass fixer, e.g., through setting $\gamma_k = 1$ and $\Delta m_{k\ell}^{(excess)} = \bar{m}_{k\ell}$ in the mass increment (37), results in a spurious loss of water vapor mass of 2.424E–07 Pa per time-step; the mass fixer is necessary for conserving tracer mass in *ne30pg2*.

5.3.2 Grid Imprinting

Flow over topography can result in significant grid imprinting using the spectral element method (Lauritzen et al., 2015a, H18). Figure 38 shows the results of the Held-Suarez with topography simulations. The middle panel is the vertical pressure velocity, ω , averaged over two years, over the Andes and Himalayan region at two different levels in the mid-troposphere, using the *ne30pg3* grid. The fields are displayed as a raster plot on the physics grid, so that individual extrema, which characterize the flow over the Andes between about $10^\circ - 20^\circ$ S, may be identified as spurious. Near the foot of the Himalayas, between about $20^\circ - 30^\circ$ N, there are parallel stripes of extrema aligned with the mountain front that appear to be spurious $2\Delta x$ oscillations.

As discussed in H18, grid imprinting over mountainous terrain tends to occur in regions of weak gravitational stability, causing extrema to extend through the full depth of the troposphere as resolved updrafts and downdrafts. Thus, grid imprinting over mountains may be alleviated through increasing the divergence damping in the model. Figure 38 (right panel) repeats the *ne30pg3*

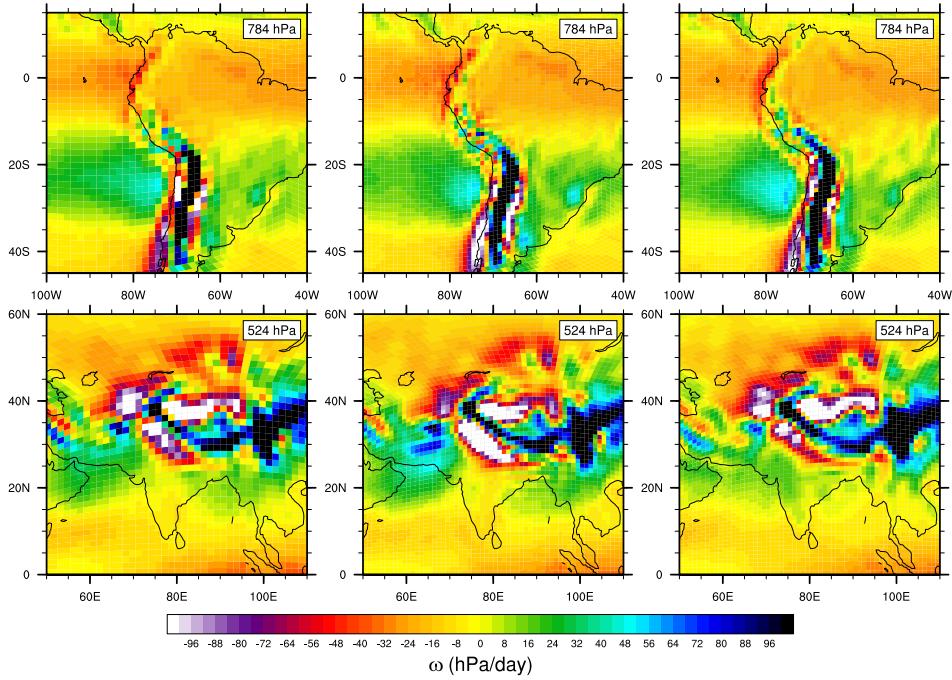


Figure 38: Mean ω at two model levels in the middle troposphere, in a Held-Suarez configuration outfitted with real world topography. (Left) *ne30pg2* (Middle) *ne30pg3* and (Right) *ne30pg3* with the divergence damping coefficient, ν_{div} , increased by an order of magnitude. The ω fields are computed from a two-year simulation. The data are presented on a raster plot in order to identify individual grid cells

simulation through increasing ν_{div} by an order of magnitude. The spurious noise over the Andes and the Himalayas are damped, and grid point extrema tend to diffuse into neighboring grid cells. The wavenumber-power spectrum of the kinetic energy due to divergent flow (Figure 39) confirms that divergent modes are damped at higher wavenumbers (greater than 30), by about an order of magnitude relative to the default *ne30pg3* simulation.

The ω field of the *ne30pg2* simulation is provided in Figure 38 (left panel). Grid cell extrema over the Andes is less prevalent than in the *ne30pg3* sim-

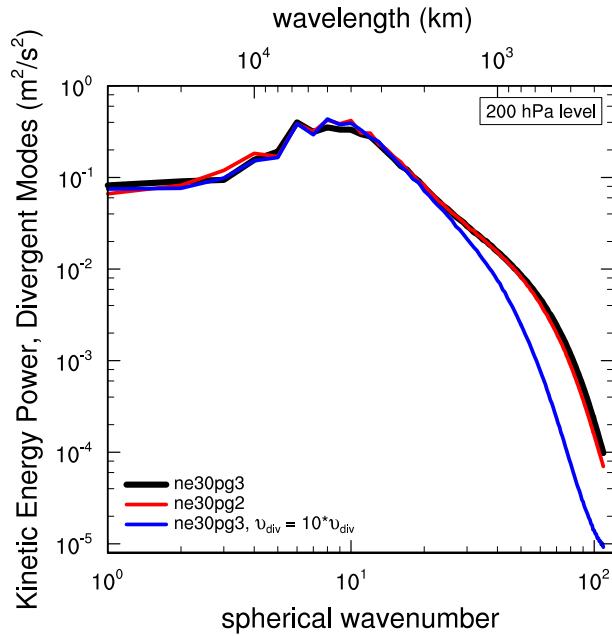


Figure 39: Kinetic energy power spectrum arising from divergent modes in *ne30pg3*, *ne30pg2* and *ne30pg3* with the divergence damping coefficient, ν_{div} , increased by an order of magnitude, in the Held-Suarez with topography simulations. Spectra computed from five months of six-hourly winds.

ulation, as seen by the reduction in large magnitude ω (e.g., red grid cells). The spurious oscillations at the foot of the Himalayas appear to have been entirely eliminated. This improvement in grid imprinting is due to the consistent smoothness properties of the control volumes in the *pg2* grid compared with the *pg3* grid discussed in Section 5.1, and these results are consistent with our hypothesis. The divergent modes are marginally damped relative to *ne30pg3* for wavenumbers greater than about 50, but are an order of magnitude larger than in the enhanced divergence damping *ne30pg3* run (Figure 39). From a scientific and model development perspective, the *pg2* configuration is preferable to the *pg3* configuration, since it eliminates grid imprinting without

placing any additional constraints on ν_{div} .

5.3.3 Impact on Resolved Scales of Motion

Tropical regions are very sensitive to horizontal resolution, primarily due to the scale dependence of resolved updrafts and downdrafts at hydrostatic scales (Weisman et al., 1997; Pauluis and Garner, 2006; Jeevanjee, 2017; Herrington and Reed, 2017, 2018). The vertical velocity of updrafts and downdrafts is related to the horizontal length scales of buoyancy the model is able to support. This can be demonstrated through a scale analysis of the Poisson equation (Jeevanjee and Romps, 2016) valid for hydrostatic scales, showing that the ratio of the scale of ω at two resolutions, due to their respective buoyancies is,

$$\frac{\omega_{\Delta x_1}}{\omega_{\Delta x_2}} = \frac{D_{\Delta x_2}}{D_{\Delta x_1}}, \quad (48)$$

where $D_{\Delta x}$ is a characteristic buoyancy horizontal length scale for grid-spacing Δx (hereafter referred to as the *forcing scale*), and it is presumed that the magnitude of the buoyancy and the vertical scale of the buoyancy is unchanged or compensating across the two resolutions. Equation (48) indicates that the magnitude of the vertical velocity scales like the inverse of the forcing scale, which was verified in a simple moist bubble configuration using CAM-SE and the CAM finite-volume dynamical core (Herrington and Reed, 2018), as well as using CAM-SE-CSLAM as configured in the present study (Appendix 8.2). It is by no means trivial that equation (48) holds for the moist bubble test, since the scaling is derived from the dry anelastic equations.

In aqua-planet simulations using CAM-SE, the forcing scale is grid-limited, varying with resolution in the range of five to ten times the grid-spacing (Herrington and Reed, 2018). From equation (48), this grid-dependence explains why the updrafts and downdrafts are so sensitive to horizontal resolution. A grid-limited forcing scale is analogous to an effective resolution, which is the characteristic length scale below which the solution becomes contaminated by numerical artifacts, and the features are overly damped due to numerical dissipation. The effective resolution may be inferred from kinetic energy spectra as the wavenumber where the slope of the spectrum becomes steeper than the observationally determined slope (Skamarock, 2011). In the CESM2 release of CAM-SE, this criterion occurs near wavenumber 60 (see Figure 6 in Lauritzen et al., 2018), a length scale of about six times the grid spacing and overlapping with the estimated forcing scale.

When the physics and dynamics grids are of different resolutions, which grid determines the models characteristic forcing scale? The remainder of Section 5.3 attempts to address this question using spectral element grids at low resolution (Section 5.3.3), high resolution (Section 5.3.3) and across all resolutions typical of present day climate models (Section 5.3.3).

Low Resolution

The question posed above may be addressed through comparing *ne30pg2*, where $\Delta x_{phys} = 166.8\text{km}$ (hereafter Δx is expressed as the average equatorial grid spacing), $\frac{3}{2}$ times larger than the dynamics grid spacing, $\Delta x_{dyn} =$

Table 13: Δx and Δt for the physics and dynamics in the low resolution simulations. Δx is computed as the average equatorial grid spacing.

Grid name	Δx_{dyn}	Δt_{dyn}	Δx_{phys}	Δt_{phys}
<code>ne20pg3</code>	166.8km	300s	166.8km	1800s
<code>ne30pg2</code>	111.2km	300s	166.8km	1800s
<code>ne30pg3</code>	111.2km	300s	111.2km	1800s

111.2km, to a simulation where both are equal to the physics grid spacing, $\Delta x_{dyn} = \Delta x_{phys} = 166.8km$ (`ne20pg3`), and another simulation where both are equal to the dynamics grid spacing, $\Delta x_{dyn} = \Delta x_{phys} = 111.2km$ (`ne30pg3`). The resolvable scales in the `ne30pg2` solution are expected to be bounded by the `ne20pg3` and `ne30pg3` solutions. Although according to equation (51), Δt_{phys} for `ne20` grids should be different from `ne30` grids, here it is set to the `ne30` value (see Table 13) in order to reduce the differences between the three configurations, and justified because lower resolution runs aren't very sensitive to this range of Δt_{phys} (Figure 52).

Figure 40 is a snapshot of the ω field in the Inter-Tropical Convergence Zone (ITCZ) in the pressure-longitude plane, in the three simulations. The ω field is overlaid with the $\pm 15K/day$ contour of the physics temperature tendencies (black), which are primarily due to stratiform cloud formation. Since the component of ω due to buoyancy is determined by the physics temperature tendencies mapped to the GLL grid, the tendencies and ω are shown on the GLL grid, $f_T^{(gll)}$ and $\omega^{(gll)}$, respectively. The white contour is intended to outline regions where the deep convection scheme is fairly active, set to the $0.0075kg/m^2/s$ value of the convective mass fluxes (note the convective mass fluxes have not been mapped to the GLL grid, and are instead shown on the

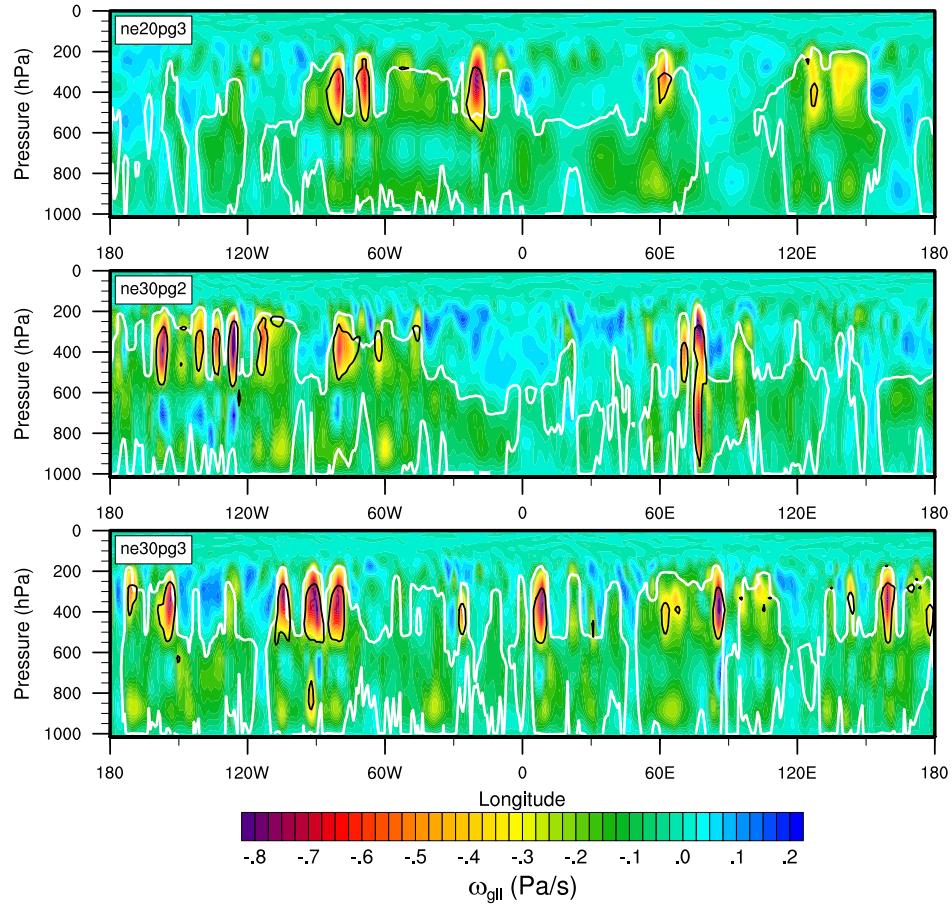


Figure 40: Snapshots in the longitude-pressure plane of $\omega^{(gll)}$ through the ITCZ region in the *ne20pg3*, *ne30pg2* and *ne30pg3* configurations, in the upper, middle and lower panels, respectively. Black is the $\pm 15 \text{ K/day}$ contour of the physics tendencies, and the white contour is the $0.0075 \text{ kg/m}^2/\text{s}$ contour of the parameterized deep convective mass fluxes.

pg grid). The figure indicates that large regions of the ITCZ are comprised of upward ω that balance the warming due to compensating subsidence produced by the deep convection scheme. Much larger magnitude ω are comprised of resolved updrafts driven by the buoyancy of stratiform clouds, and resolved downdrafts due to evaporation of condensates produced by overlying clouds (Herrington and Reed, 2018). These large buoyancy stratiform clouds tend to form in the middle-to-upper troposphere due to detrainment of moisture from the deep convection scheme (Zhang and McFarlane, 1995).

It is not obvious from the snapshots in Figure 40 whether the length scales of the stratiform clouds, which are approximately equal to the models characteristic forcing scale, are any different across the three simulations. Analogous to determining the effective resolution (Skamarock, 2011), the forcing scale may be inferred from the wave-number power spectrum of $f_T^{(gll)}$ as the maximum wavenumber prior to the steep, un-physical decline in power that characterizes the near-grid scale (hereafter $f_T^{(gll)}$ is referred to as the *forcing*). The wave-number power spectrum of the forcing in the middle-to-upper troposphere is shown in Figure 41a. Unlike kinetic energy spectra, the decline in power near the models effective resolution is more gradual, making it difficult to determine a characteristic forcing scale from the spectra. However, it is clear that the slope of the *ne20pg3* spectrum begins to steepen at smaller wavenumbers than in the *ne30pg3* spectra. Additionally, the *ne30pg2* spectra is remarkably similar to the *ne30pg3* spectra, for all wavenumbers. These spectra indicate that the characteristic forcing scale in the *ne30pg2* and *ne30pg3* simulations are similar, and that both are smaller than the *ne20pg3* forcing

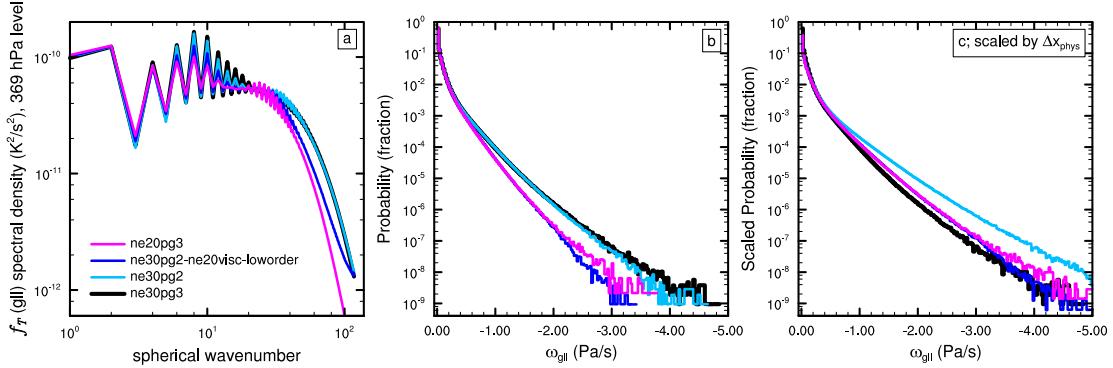


Figure 41: (Left) Wavenumber-power spectrum of the temperature tendencies from the moist physics, near the 369 hPa level, (Middle) probability density distribution and (Right) the scaled probability density distribution of upward ω everywhere in the model. The scaled distributions are scaled to *ne30pg3* using Δx_{phys} .

scale. From equation (48), it is expected that the magnitude of the vertical motion is greater in both the *ne30pg2* and *ne30pg3* simulations.

The probability density function (PDF) of upward $\omega^{(gll)}$ everywhere in the simulations is shown in Figure 41b. Large magnitude $\omega^{(gll)}$ are more frequent in the *ne30pg2* run, compared to *ne20pg3*, and the PDF is actually more similar to the *ne30pg3* distribution, consistent with their similar forcing scales. This may be further illustrated through scaling the PDF's,

$$P_s(\omega) = \alpha \times P(\omega/\alpha), \quad (49)$$

where $P_s(\omega)$ is the scaled PDF of ω and α is the ratio of ω to ω_{target} , the ω associated with the target grid resolution, Δx_{target} . Making the assumption that the forcing scale is linear in Δx , then from equation (48), $\alpha = \Delta x_{target}/\Delta x$. The target resolution is taken here to be equal to the *ne30pg3* grid resolution.

If the forcing scale of $ne30pg2$ is in fact determined by Δx_{phys} , then one sets $\Delta x = \Delta x_{phys}$ in α . This scaled PDF, however, severely overestimates the frequency of upward ω of the target resolution, $ne30pg3$ (Figure 41c). It is clear from the similarity of the un-scaled PDF's of $ne30pg2$ and $ne30pg3$ (Figure 41b), and their forcing spectra (Figure 41a), that the characteristic forcing scale in these two configurations are approximately the same. It follows that the forcing scales in $ne30pg2$ and $ne30pg3$ are determined by their common grid, Δx_{dyn} , rather than Δx_{phys} , which are different. And one can be reasonably confident in the linear framework used to approximate α - the scaled $ne20pg3$ PDF fits the $ne30pg3$ distribution quite well. It then follows that the forcing scale of $ne20$ simulations is about $\frac{3}{2}$ times that of $ne30$ simulations, the ratio of their grid spacings.

There are two reasons the $pg2$ forcing scale is determined by the GLL grid. The first being that the hyper-viscosity coefficients are a function of the GLL grid resolution (equation (47)), and the second, that the physics tendencies are mapped to the $pg3$ and GLL grids using high-order mapping, which reconstructs scales the $pg2$ grid is unable to support (see Appendix 8.3). The impact of only using low-order mapping or only using $ne20$ viscosity in a $ne30pg2$ simulation results in a forcing spectra that lies in between the default $ne30pg2$ and $ne20pg3$ runs (not shown). The combined effect of both factors on the forcing scale is illustrated through an $ne30pg2$ simulation that uses low-order mapping, and with hyper-viscosity coefficients set to $ne20$ values ($ne30pg2-ne20visc-loworder$ in Figure 41). The PDF of $\omega^{(gll)}$ and the forcing spectrum more closely resemble the $ne20pg3$ run. In the $ne30pg2-ne20visc-$

Table 14: Δx and Δt for the physics and dynamics in the high resolution simulations. Δx is computed as the average equatorial grid spacing.

Grid name	Δx_{dyn}	Δt_{dyn}	Δx_{phys}	Δt_{phys}
ne80pg3	41.7km	112.5s	41.7km	675s
ne120pg2	27.8km	75s	41.7km	450s
ne120pg3	27.8km	75s	27.8km	450s

loworder configuration, the forcing scale is more accurately determined by Δx_{phys} since the scaled PDF is in fairly good agreement with the *ne30pg3* simulation (Figure 41c).

High Resolution

The experiment described in the previous section is repeated here for a *ne120pg2* aqua-planet simulation, corresponding to an approximate grid spacing of $\Delta x_{dyn} = 27.8\text{km}$ and $\Delta x_{phys} = 41.7\text{km}$. *ne80pg3* refers to the grid in which the physics and dynamics are the same resolution as the physics of the *ne120pg2* grid, and *ne120pg3*, the grid in which the physics and dynamics are equal to the resolution of the dynamics of *ne120pg2*. At these higher resolutions, the solutions are sensitive to Δt_{phys} (Figure 52), and so the *ne80* grid uses a larger time-step than that of the *ne120* grids (Table 14), following equation (51).

Figure 42 is the same as Figure 41, but for the high resolution simulations. While the *ne80pg3* forcing spectra begins to drop off near wavenumber 100, the *ne120pg2* and *ne120pg3* drop off closer to wavenumber 200, and their spectra lie on top of one another (Figure 42a). The PDF's of (upward) $\omega^{(gl)}$ show that the *ne120* distributions lie on top of one another, and while not a perfect

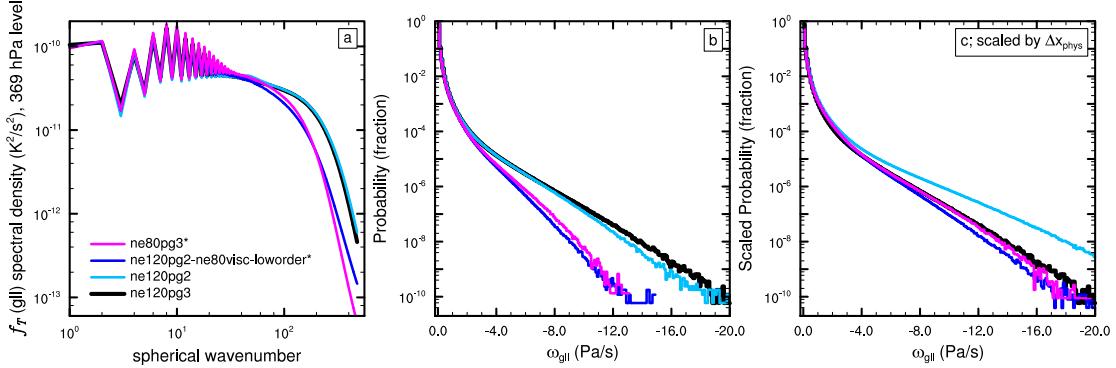


Figure 42: As in Figure 41, but for the high resolution simulations. Asterisks indicate that $\Delta t_{phys} = 675s$, which is larger than that used for the default *ne120* runs (see Table 14).

match, both *ne120* runs have substantially more frequent large magnitude vertical motion than in the *ne80pg3* run (Figure 42b). As in the low resolution runs, the similarity of the *ne120* forcing spectra and $\omega^{(gll)}$ distributions indicate that the forcing scale of the *ne120pg2* run is not determined by the physics grid spacing, but rather the dynamics grid spacing. This is also evident from the over-prediction of the frequency of large magnitude $\omega^{(gll)}$ compared with the *ne120pg3* run, through scaling the *ne120pg2* PDF and setting the forcing scale proportional to Δx_{phys} in equation (49) (Figure 42c).

In the *ne120pg2* simulation, the dynamics grid determines the forcing scale for the same two reasons found in the low resolution runs. The high-order mapping of the physics to the dynamics is important for reconstructing scales not supported on the *pg2* grid, and scaling the viscosity coefficients by the dynamics grid spacing is also important. But in order to recreate the *ne80pg3* solution using the *ne120pg2* grid, the physics time-steps must be the same

Table 15: Δx and Δt for the physics and dynamics in the high resolution simulations. Δx is computed as the average equatorial grid spacing.

Grid name	Δx_{dyn}	Δt_{dyn}	Δx_{phys}	Δt_{phys}
ne40pg3	83.4km	222.5s	83.4km	1350s
ne60pg2	55.6km	150s	83.4km	900s
ne60pg3	55.6km	150s	55.6km	900s

for these two grids. Combining all three modifications leads to an *ne120pg2* solution that resembles the *ne80pg3* run (*ne120pg2 – ne80visc – loworder** in Figure 42). The forcing spectrum and distribution of $\omega^{(gll)}$ match that of the *ne80pg3* run, and scaling the PDF by Δx_{phys} closely resembles the *ne120pg3* distribution.

Across Resolutions

Three intermediate resolution aqua-planets are run to provide a continuous representation of the solution spanning from low to high resolution (Table 15). Figure 43 is scatter plot of the climatological global mean state versus Δx_{dyn} for all model configurations listed in Tables 13–15. The fields plotted in the figure, upward ω , and the two components of precipitation, stratiform precipitation rate (CLUBB) and deep convective precipitation rate (ZM), are all sensitive to resolution. Upward ω and CLUBB precipitation decreases, and ZM precipitation increases monotonically with Δx_{dyn} . The *pg2* solutions have very similar values to the *pg3* solutions, although they are slightly offset towards the lower resolution side of the plots. The differences between the *pg2* and *pg3* solutions are much less than the differences between *pg2* and configurations where the physics and dynamics grids are both equal to the

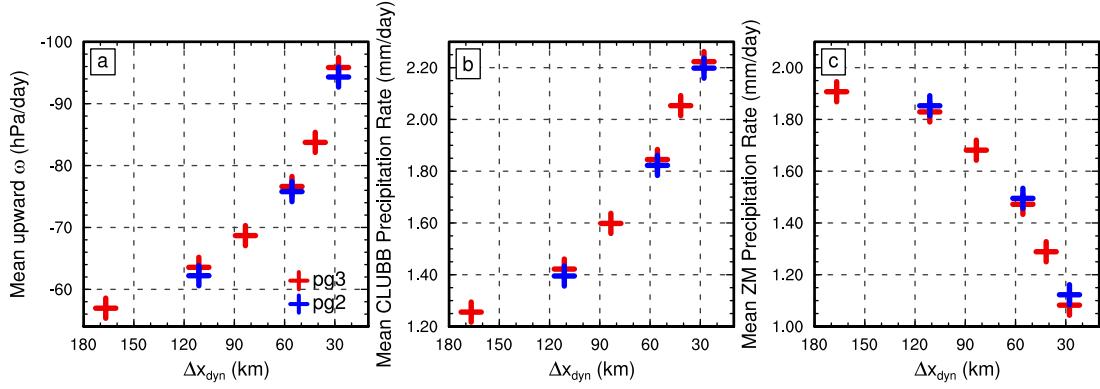


Figure 43: Global mean, time-mean (a) upward ω , (b) CLUBB precipitation rate and (c) parameterized deep convective precipitation rate. All means computed from the final 11 months of one-year simulations, and upward ω is computed using 6-hourly output.

pg2 physics grid resolution (e.g., *ne40pg3* compared with *ne60pg2*). The mean state of the configurations resembles that of the transients discussed in the previous sections; the coarser *pg2* physics grid does not appear to degrade the resolved scales of motion, which are primarily determined by the dynamics grid resolution.

5.4 Conclusions

This study documents the implementation of a coarser resolution physics grid into the Community Atmosphere Model (CAM), with spectral element dynamics (based on a dry-mass vertical coordinate) and conservative semi-Lagrangian advection of tracers (CAM-SE-CSLAM). The spectral-element and tracer advection grids are mapped to a finite-volume physics grid after Herrington et al. (2018), but containing $\frac{2}{3}$ fewer degrees of freedom in each horizontal direction.

Mapping from the coarser physics grid to the dynamics and tracer grids is performed with high-order reconstructions, and a tendency mapping algorithm is developed to ensure shape preservation, consistency, linear-correlation preservation and mass conservation. These numerical properties are verified to a high degree of precision through idealized tests.

The coarser resolution physics grid is designed to eliminate grid imprinting that manifests for non-smooth problems using element-based high-order Galerkin methods. The physics grid control volumes encompass a region of the element such that an isotropic representation of the numerics is provided to the physical parameterizations, and it was hypothesized that this method eliminates grid imprinting from the element boundaries. Using a Held-Suarez configuration modified with real-world topography, it was shown that element boundary noise over steep topography is eliminated from the coarser physics grid solution, consistent with our hypothesis.

Physical parameterizations make up a significant fraction of the total computational cost of atmosphere models, and the coarser physics grid may be used to reduce this overhead. The cost savings is due to the factor $\frac{4}{9}$ fewer grid columns in which the physics need be computed, and for CESM2.1, where CAM6 physics makes up about half the cost of the overall model (Lauritzen et al., 2018), corresponds to a potential 25% fewer core hours. The authors sought to understand whether the reduction in computational cost occurs at the expense of a degraded solution, through aliasing the dynamics to the coarser resolution physics. An exhaustive number of grids were developed and run in an aqua-planet configuration, and confirm that the resolved scales

of motion are not degraded through the use of a coarser resolution physics grid. It was found that the resolved scales are primarily determined by the effective resolution of the dynamical core. This was attributed to two factors; (1), explicit numerical dissipation by the dynamics blurs the distinction between solutions on the physics, dynamics or tracer grids, and (2), that high-order mapping of the physics tendencies to the dynamics and tracer grids reconstructs scales that are not supported on the coarser physics grid.

The coarser physics grid in CAM-SE-CSLAM provides significant cost savings with little to no downside. The coarser physics grid replicates solutions from the conventional method of evaluating the physics at the same resolution as the dynamical core, removes grid imprinting from the solution and runs efficiently on massively parallel systems. The coarser physics grid may be leveraged to reduce the computational burden as a component of increasingly expensive Earth System Models, or permit once unattainable throughputs for high-resolution climate simulations. The coarser physics grid configuration of CAM-SE-CSLAM is well positioned to address the scientific challenges ahead, as a formidable next generation climate model.

6 Parameterized convection, grid-scale clouds and resolution sensitivity in CAM-SE-CSLAM

6.1 Introduction

An increasing number of atmospheric dynamical cores are being developed to maximize efficiency on massively parallel systems, paving the way towards regionally high-resolution ($\Delta x = 50$ km or less), or even globally high resolution simulations (Skamarock et al., 2012; Zängl et al., 2014; Harris et al., 2016; Ullrich et al., 2017; Lauritzen et al., 2018). Incorporating these advances into Atmospheric General Circulation Models (AGCMs) requires the development of physical parameterizations appropriate for the diversity of grid configurations that dynamical cores are now able to support, referred to as scale-aware physics. The most common approach to understand and develop scale-aware physics has been through the lens of limited area, cloud resolving simulations (Plant and Craig, 2008; Arakawa and Wu, 2013; Song and Zhang, 2018). In filtering cloud resolving solutions to some target, lower resolution resembling an AGCM grid, and studying how the filtered moments vary as a function of target grid resolution, one may develop a more general relationship between resolved and unresolved scales. While this approach is likely necessary for developing scale-aware physics, it is not sufficient. The equations of motions of motion have inherent scale dependencies (Orlanski, 1981; Weisman et al., 1997; Pauluis and Garner, 2006; Jeevanjee, 2017), and the resolved dynamics

act accordingly to the scales a grid is able to support. Scale-aware physics should also accommodate these dependencies.

The sensitivity of the Community Atmosphere Model (CAM) to horizontal resolution (and it's predecessor versions) is well documented over the last few decades, beginning with the Community Climate Model, version 1 (Kiehl and Williamson, 1991). The total precipitable water and convective precipitation rate decreases and stratiform precipitation increases monotonically with resolution in every convergence study in the CAM-lineage that publishes these quantities (Kiehl and Williamson, 1991; Williamson et al., 1995; Williamson, 2008; Rauscher et al., 2013; Zarzycki et al., 2014b; Herrington and Reed, 2017). In these studies, total cloud fraction decreases monotonically with resolution as well, except in the CAM, version 5 study (CAM5; Zarzycki et al., 2014b). The tendency for the atmosphere to dry with resolution was connected to an observed increase in the magnitude of the resolved vertical velocities with resolution in the studies of (Kiehl and Williamson, 1991; Williamson et al., 1995), with greater subsiding motion leading to drier conditions, and lower cloud fractions.

Herrington and Reed (2017, 2018) analyzed a set of CAM5 and CAM, version 4 (CAM4) aqua-planet simulations and found that resolved updrafts dominate the vertical velocity field of tropical convecting systems at all resolutions (they considered a range typical of present day models; $208.5\text{km} \geq \Delta x \geq 27.8\text{km}$). The scale of resolved updrafts are collocated in the horizontal with the buoyancy produced by grid-scale clouds (also referred to as stratiform clouds in the literature), and are grid limited, conforming to the effective

resolution of the model of about $5 - 10\Delta x$. Assuming there is a characteristic buoyancy length scale associated with a grid resolution, and it is proportional to Δx , the ratio of the vertical velocity scale at that grid resolution W , to a high-resolution reference simulation W_{ref} is:

$$\alpha = \frac{W}{W_{ref}} = \frac{\Delta x_{ref}}{\Delta x}, \quad (50)$$

where Δx_{ref} is the grid-spacing of a high-resolution reference, and it is assumed that the magnitude and height scale of the buoyancy is unchanged or compensating across resolutions. The physical interpretation of equation 50 is that a rising column of buoyancy creates a low pressure perturbation of similar horizontal scale in its wake, and this pressure gradient scales like Δx^{-1} , facilitating convergence into the pressure minimum and the resultant vertical velocity also scales like Δx^{-1} . While Herrington and Reed (2017) found that equation 50 over-predicted the change in vertical velocity across resolutions in their aqua-planet simulations, Herrington and Reed (2017) discovered that this over-prediction was at least in-part, due to time-truncation errors arising from too small a physics time-step Δt_{phys} in the higher resolution simulations.

In this study, it is shown that through scaling Δt_{phys} in a manner which avoids large time-truncation errors at higher resolutions (Chapter 5), the analytical scaling in equation 50 does explain the change in vertical velocities across resolutions in an aqua-planet configuration using CAM. The implications of equation 50 are that for a doubling of resolution, one can get the same mass flux in half the area. Grid-scale cloud fraction and regions of ascent are

then confined to smaller areas at higher resolutions; the area and magnitude of subsiding motion increases to conserve mass in the simulations. Greater subsiding motion results in a drier and more stable atmosphere, which reduces the frequency the deep convection scheme is triggered. Section 6.2 describes the model and design of the convergence study, section 6.3 presents results and section 6.3 provides conclusions.

6.2 Methods

6.2.1 Dynamical Core

The spectral-element dynamical core option of Community Atmosphere Model (CAM), coupled with semi-Lagrangian advection method for tracer advection, and dry-mass vertical coordinate is used in this study (Lauritzen et al., 2017, 2018; Herrington et al., 2018). The dry dynamics are solved using the spectral element method in which the vector-invariant form of the hydrostatic equations of motion are solved with a degree 3 polynomial basis in each element, and the solutions are patched together at the boundaries using the direct stiffness summation. The elements are defined by a cubed-sphere grid, and the notation for horizontal grid resolution is an ‘ne’ followed by the number of elements contained on an edge of one of cubed-sphere faces, e.g., $ne30$. ∇^4 explicit numerical dissipation is applied to temperature, dry pressure thickness, rotational and divergent winds. The advection of tracers uses a semi-Lagrangian method on a finite volume grid constructed from the cubed-sphere used for the elements, and contains the same degrees of freedom and the dry dynamics. The coupled system conserves energy, mass and preserves linear correlations

between two reactive species.

The physical parameterizations (*physics*) are evaluated on the tracer advection grid, and the tendencies mapped back to the spectral element grid (Herrington et al., 2018). A coarser physics grid, containing $\frac{5}{9}$ fewer degrees of freedom than the dynamical core grid is also available as part of the CAM-SE-CSLAM package (Chapter 5). This configuration is also used in this study, but it is considered a perturbed configuration from the default configuration in this study. The dynamics time-step is sub-cycled within a longer physics time-step Δt_{phys} , and the temperature and momentum increments from the physics are then divided by the number subcycles and added to the dynamical core at the beginning of each subcycle; the full moisture increment from the physics is applied only at the start of the first subcycle (*ftype* = 2 option in Lauritzen and Williamson, 2019).

6.2.2 Moist Physical Parameterizations

The simulations use CAM, version 6 physics (CAM6; https://ncar.github.io/CAM/doc/build/html/users_guide/index.html). The Cloud Layers Unified by Binormals (CLUBB Golaz et al., 2002; Bogenschutz et al., 2013) is an assumed PDF higher order closure model that handles shallow convection, planetary boundary layer mixing and cloud macrophysics. The macrophyisics are coupled with a two-moment bulk cloud microphysics scheme with prognostic precipitation (Gettelman et al., 2015), and microphysics are coupled with a three mode Modular Aerosol Model (Liu et al., 2012). Deep convection is parameterized using a quasi-equilibrium mass flux scheme (Zhang and

McFarlane, 1995) using a dilute-plume calculation to close the mass fluxes and determine convective triggering (Neale et al., 2008). The deep convection scheme includes convective momentum transport (Richter and Rasch, 2008).

6.2.3 Experimental Design

This convergence study uses an aqua-planet configuration (Neale and Hoskins, 2000; Medeiros et al., 2016), an all ocean planet with fixed, zonally symmetric sea surface temperatures modeled after the present day Earth (*QOBS* in Neale and Hoskins, 2000). The aqua-planets are in a perpetual equinox, and aerosols are largely absent from the simulations. Each simulation is ran for one simulated year. Six different horizontal grid are used in this study, which are provided in Table 16. The horizontal hyper-viscosity operators ν vary with resolution after Chapter 5, and are also provided in the table. The values of ν are slightly larger for divergence damping and not shown in the Table. Δt_{phys} is chosen to scale with resolution, in proportion to the grid spacing,

$$\Delta t_{phys} = \Delta t_{phys,0} \times \frac{n_{e,0}}{n_e} s, \quad (51)$$

where $\Delta t_{phys,0}$ is taken to be the standard 1800s used in CAM-SE-CSLAM at low resolution, $n_{e,0} = 30$ (equivalent to a dynamics grid-spacing of 111.2km). This scaling was chosen to avoid large time-truncation errors in a rising moist bubble test (Appendix 8.2), and it is understood that it will likely lead to greater resolution sensitivity (Williamson, 2008). The convective time-scale in the deep convection scheme is fixed at 3600 s in all simulations.

Table 16: Experimental design and global means.

Variable	<i>ne20</i>	<i>ne30</i>	<i>ne40</i>	<i>ne60</i>	<i>ne80</i>	<i>ne120</i>
ν (m^4/s)	1.5×10^{15}	4.0×10^{14}	1.5×10^{14}	4.0×10^{13}	1.5×10^{13}	4.0×10^{12}
Δt_{phys} (s)	2700	1800	1350	900	675	450
Total Cloud Fraction	0.844	0.835	0.824	0.810	0.804	0.800
Total Precipitable Water (mm)	23.31	23.01	22.62	22.25	21.93	21.72
Convective Precipitation (mm/day)	1.91	1.83	1.68	1.47	1.29	1.08
Stratiform Precipitation (mm/day)	1.26	1.42	1.60	1.85	2.05	2.22

6.3 Results

Table 16 provides a list of some relevant globally averaged metrics that are typically published in CAM-lineage convergence studies. Total precipitable water, total cloud fraction and deep convective precipitation rate decreases, while stratiform precipitation increases, monotonically with resolution. Resolution sensitivity in CAM6 is similar to all prior model versions within the CAM-lineage.

The probability density function (PDF) of negative, or upward vertical pressure velocities ω in the aqua-planets is shown in Figure 44a. The magnitude of upward ω increases in a monotonic way with resolution, with positive, or downward ω behaving similarly (not shown). The PDF's may be scaled to the high-resolution *ne120* resolution through $P(\omega)_s = \alpha P(\omega/\alpha)$, where α is the scale factor equation 50. The scaled PDF's all line up on top of the high-resolution reference (Figure 44b); equation 50 explains the variation in vertical velocity across resolutions to first order.

Changes to the time-mean vertical velocity field can be understood through decomposing the mass weighted global mean ω into upward and downward components,

$$\overline{\langle \omega \rangle} = \overline{\langle f_u \rangle} \overline{\langle \omega_u \rangle} + \overline{\langle f_d \rangle} \overline{\langle \omega_d \rangle}, \quad (52)$$

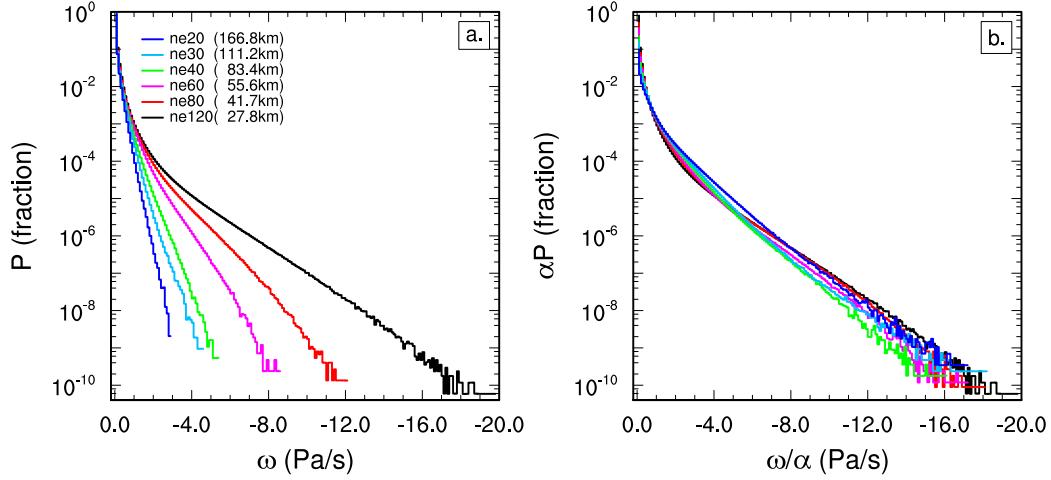


Figure 44: Probability density distribution of the upward vertical pressure velocities ω computed everywhere in the model from 1 year of 6-hourly data (a) raw values (b) scaled to the $ne120$ resolution using $P(\omega)_s = \alpha P(\omega/\alpha)$, where α is from equation 50.

where $\langle f_x \rangle$ and $\langle \omega_x \rangle$ refers to the mass weighted vertical fraction and mass weighted vertical mean ω , respectively, subscript u refers to upward motion and d , downward motion; overbars indicate time-mean.

The components of equation 52 are shown in Figure 45a,b,e,f for the six aqua-planet simulations. The magnitude of both $\overline{\langle \omega_u \rangle}$ and $\overline{\langle \omega_d \rangle}$ increase monotonically with resolution, the upward component increasing more than the downward component. In contrast, $\overline{\langle f_u \rangle}$ decreases with resolution while $\overline{\langle f_d \rangle}$ increases, although in both cases there is a reversal in this trend in the highest resolution simulation, $ne120$. Note that the magnitude of the change in $\overline{\langle f_u \rangle}$ and $\overline{\langle f_d \rangle}$ is small, a range of about 0.015 across all simulations. Figure 45e,g shows that the products $\overline{\langle f_u \rangle} \overline{\langle \omega_u \rangle}$ and $\overline{\langle f_d \rangle} \overline{\langle \omega_d \rangle}$ are equal and opposite, which is a requirement of mass conservation in the model and a convenient check

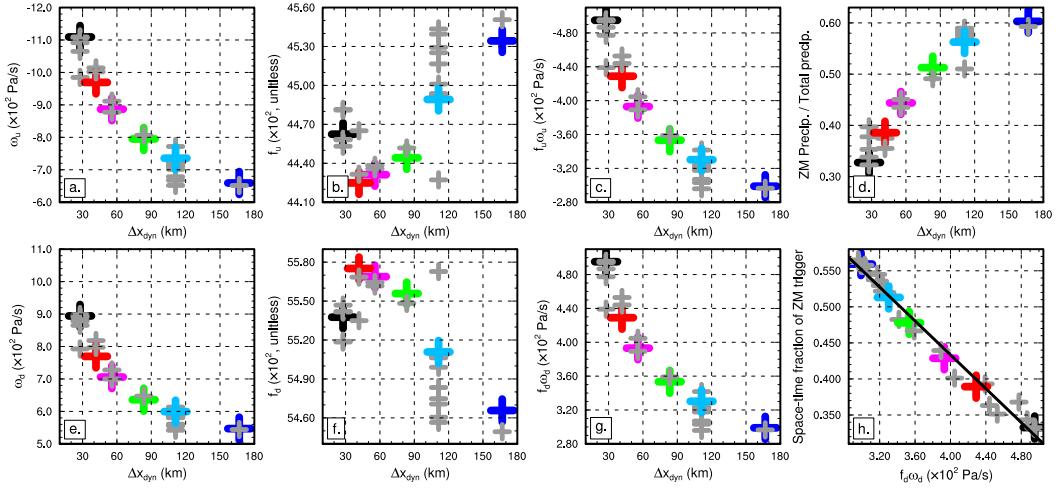


Figure 45: (a,b,c,e,f,g) Components of the global mean vertical pressure velocity $\langle \omega \rangle$ (d) ratio of global mean ZM precipitation rate to the total precipitation rate versus grid spacing Δx and (e) scatter plot of $\langle f_d \rangle \langle \omega_d \rangle$ and FREQZM, and the fitted linear regression which has a Pearson's R-value = 0.99. (a) $\langle \omega_u \rangle$ (b) $\langle f_u \rangle$ (c) $\langle f_u \rangle \langle \omega_u \rangle$ (e) $\langle \omega_d \rangle$ (f) $\langle f_d \rangle$ (g) $\langle f_d \rangle \langle \omega_d \rangle$. Colors are as in Figure 44. Grey crosses are for the perturbed parameter runs.

of the calculation. Besides the six aqua-planets described in Section 6.2, 23 additional year-long aqua-planet experiments were carried out at various resolutions, with modified parameters in the dynamical core and the physics. The spread in the components of equation 52 in the simulations with perturbed parameters is relatively large for $\langle f_x \rangle$, compared with $\langle \omega_x \rangle$.

The reduction in the Zhang and McFarlane (1995) deep convective precipitation rate (hereafter referred to as the *ZM precipitation*) is depicted in Figure 45d, which shows the global mean fraction of total precipitation arising from ZM precipitation decreasing from about 0.60 at low resolution (*ne20*) to about 0.32 at high resolution (*ne120*). This reduction in ZM precipitation manifest through triggering less frequently (not shown). The space-time

fraction the ZM precipitation scheme is triggered (hereafter referred to as the *FREQZM*) is highly negatively correlated with $\overline{\langle f_d \rangle} \overline{\langle \omega_d \rangle}$ in the 29 simulations (Pearson's R-value = 0.99; Figure 45h), more then with any individual component of equation 52.

Figure 46a,b shows the time mean zonal mean $\langle f_d \rangle \langle \omega_d \rangle$ and FREQZM. $\langle f_d \rangle \langle \omega_d \rangle$ is smallest in the deep tropics (equatorward of about 10°), increasing sharply poleward throughout the subtropics and then begins to decrease in the middle latitudes ($\geq 40^\circ$) and continuing into the polar regions. FREQZM is largely opposite of $\langle f_d \rangle \langle \omega_d \rangle$; where greater subsiding motion occurs FREQZM is small, except in the mid-latitude region between 30° and 60° where the spatial variations are quite different. The resolution sensitivity of these two quantities has no zonal dependence; $\langle f_d \rangle \langle \omega_d \rangle$ increases and FREQZM decreases everywhere with resolution, in similar proportions.

The ZM trigger uses a dilute form of CAPE, which is consistent with the negative relation between $\overline{\langle f_d \rangle} \overline{\langle \omega_d \rangle}$ and FREQZM in space, and across resolutions. In the classical, non-dilute case, CAPE can be separated into two components (Zhang, 2002); instability due to the thermodynamic state of parcels in the boundary layer and the instability generated through advection of dry static energy and moisture by the environment, i.e., the resolved flow. The latter term generally contributes positively to CAPE in regions of ascent and negatively in regions of subsidence (Song and Zhang, 2018), consistent with the negative correlation of $\langle f_d \rangle \langle \omega_d \rangle$ in space, and across resolutions.

To further support a negative relation between subsiding motion and CAPE, temperature and moisture profiles are collected from the deep tropics in the

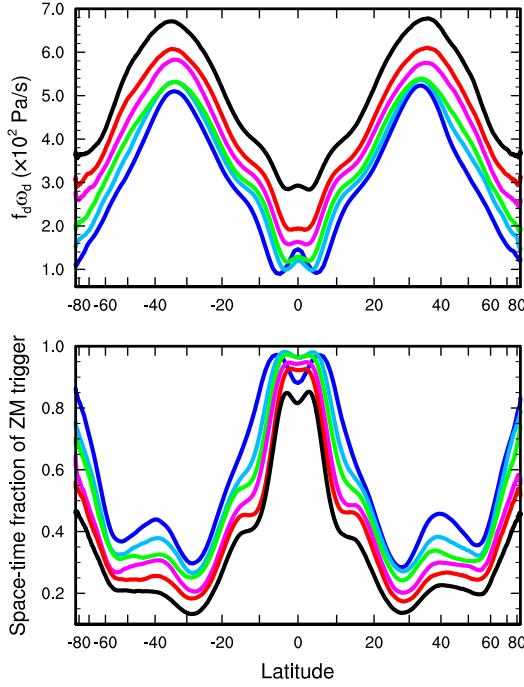


Figure 46: Time mean zonal mean (a) $\langle f_d \rangle \langle \omega_d \rangle$, (b) FREQZM. Colors are as in Figure 44.

simulations, and conditionally sampled depending on whether $\langle \omega \rangle$ is positive or negative, indicating predominantly subsiding or ascending grid columns. The mean temperature and moisture profiles of subsiding and ascending regions are then used to compute the dilute-CAPE used by the ZM scheme, offline. Figure 47b indicates that the mean profile of ascending regions are associated with large values of CAPE ($> 170 \text{ J/kg}$), and low values in subsiding regions ($< 110 \text{ J/kg}$). The mean profiles of subsiding regions have an anomalous warming layer in the $600 - 800 \text{ hPa}$ layer and an anomalous moisture deficit throughout the entire column, relative to the non-conditionally sampled mean (not shown). Figure 47a shows that fractional area of air columns in the

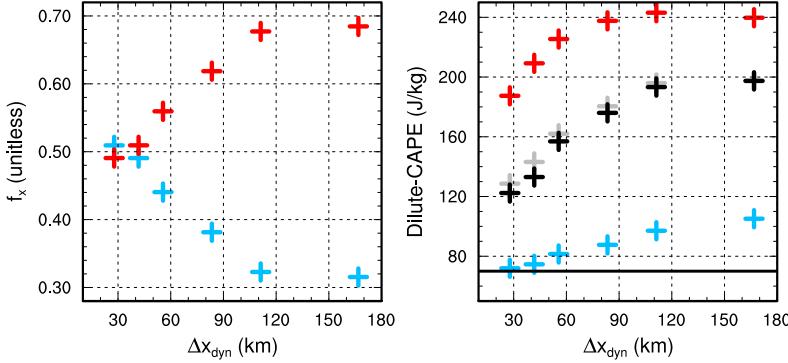


Figure 47: (a) time mean fraction of the deep tropics in the simulations with upward $\langle \omega \rangle$ (red) and downward $\langle \omega \rangle$ (blue), (b) CAPE computed from the mean temperature and moisture profiles of upward regions and downward regions. Black is for CAPE computed from the mean temperature and moisture profiles for the entire deep tropics, grey is the approximate discussed in the text.

deep tropics that are predominantly subsiding (ascending) changes drastically with resolution, from 0.3 (0.7) in the *ne20* run, and monotonically increasing (decreasing) with resolution to 0.7 (0.3) in the *ne120* run. Interestingly, the sum of the product of the fractional areas with their corresponding CAPE values gives approximately the same value as the CAPE value resulting from the non-conditionally sampled mean temperature and moisture profiles (Figure 47b). This provides strong evidence that CAPE values in the deep tropics are decreasing with resolution because a larger (smaller) fraction of subsiding (ascending) columns make up a larger (smaller) portion of the region with resolution.

To understand how the relationship between subsiding motion and the trigger in the ZM scheme manifests within the simulations themselves, a logistic regression is performed for each grid column in the simulations. Logistic re-

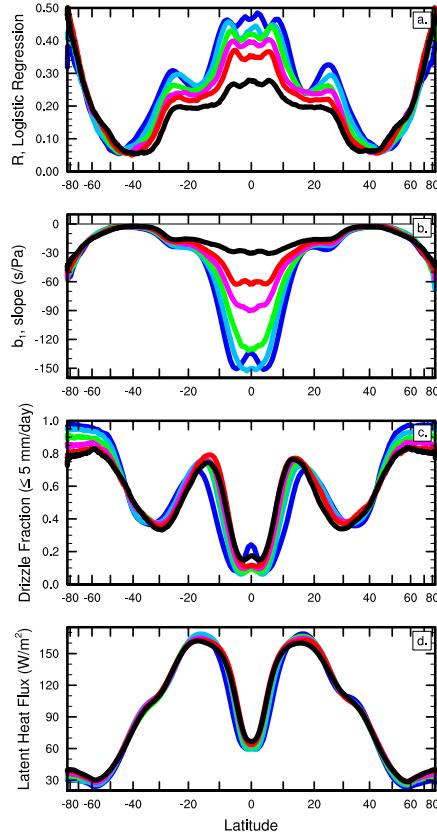


Figure 48: Zonal mean (a) R-values and (b) sensitivity parameter in the logistic regression, (c) drizzle fraction, (d) time mean surface latent heat fluxes. Colors are as in Figure 44.

gression uses an iterative method to fit a continuous variable predictor, x to a binary predictand p (Wilks, 2011),

$$p = \frac{\exp[b_0 + b_1x]}{1 + \exp[b_0 + b_1x]}, \quad (53)$$

where b_0 and b_1 are the shape parameters of the exponential. The predictor is chosen as the instantaneous $\langle f_d \rangle \langle \omega_d \rangle$ of a grid column, and the predictand is whether or not the ZM scheme is active, 1 for yes and 0 for no. Since the

aqua-planets have zonally symmetric boundary conditions, there is a zonally varying structure in the goodness of fit (R-value) and parameter b_1 (hereafter referred to as the sensitivity parameter). Figure 48a shows the zonal mean R-values indicates the goodness of fit is greatest in the deep tropics. Figure 48d shows the time mean zonal mean latent heat flux in the simulations, which is expected to contribute positively to CAPE through the component associated with the thermodynamic state of boundary layer parcels. In the deep tropics, the latent heat flux is small, and the sensitivity parameter is large and negative (Figure 48b), indicating that subsiding motion is successfully depressing CAPE, and the activity of the ZM scheme in that region.

Poleward of 10° , the R-value decreases to a local minimum between $15^\circ - 20^\circ$ latitude, and the magnitude of the sensitivity parameter steeply declines. Even though the subsiding motion is increasing polewards of 10° , this motion is less effective in depressing the convection scheme. The local minimum in the R-value corresponds with a local maximum in the latent heat fluxes, indicating that the ZM scheme is primarily responding to the instability of the boundary layer due to large surface latent heat fluxes. The CAPE values in the $10^\circ - 15^\circ$ latitude region are likely to be small, since the ZM precipitation rate consists primarily of drizzle (Figure 48c). The predominance of drizzle in this region is probably a result of the large subsiding motion (Figure 46) constraining CAPE from becoming too large. AGCMs are known to suffer from an excess drizzle bias in precisely this approximate region (Dai, 2006), and this analysis indicates that this bias is due to the use of a CAPE trigger function. In the $30^\circ - 60^\circ$ latitude region the shape parameter reaches it's global minimum and

then begins to increase in the polar regions. This is consistent with lack of spatial correspondence between $\langle f_d \rangle \langle \omega_d \rangle$ and FREQZM in the mid-latitudes, and the reemergence of this correspondence in the polar regions (Figure 46).

The sensitivity parameter becomes less negative with resolution, which is likely due to the generally greater magnitude $\langle f_d \rangle \langle \omega_d \rangle$ with resolution, which requires a lower sensitivity parameter to predict the binary of whether the ZM scheme is active. The R-values generally decrease with resolution indicating that there is degradation in the relationship with resolution. All values shown are statistically significant at the 95% level using a log-likelihood test (Wilks, 2011).

6.4 Conclusions

A convergence study is carried out in order to unravel the relationships between grid-scale clouds, parameterized convection and resolution sensitivity in an aqua-planet configuration of the Community Atmosphere Model (CAM) with version 6 physics (CAM6), the spectral-element dynamical core coupled to the Conservative Semi-Lagrangian Advection Method for advection of tracers and dry-mass vertical coordinate (CAM-SE-CSLAM). Common with just about all versions in the CAM-lineage, total cloud fraction, total precipitable water and convective precipitation decrease with resolution, while stratiform precipitation increases with resolution. The magnitude of vertical velocities everywhere in the model increase, and scale like the inverse of the grid spacing Δx^{-1} . The climatological area of regions of ascent decrease with resolution, the area of subsiding motion increases.

The product of the mass weighted vertical fraction of downward vertical pressure velocities (ω_d) with the mass weighted vertical mean of ω_d ($\langle f_d \rangle \langle \omega_d \rangle$) is highly negatively correlated with the frequency the Zhang and McFarlane (1995) deep convection scheme is triggered (FREQZM). The correlation of the climatological quantities is linear using 29 simulations containing a large range of grid resolutions, and has a goodness of fit Pearson's R-value of 0.99. A logistic regression between $\langle f_d \rangle \langle \omega_d \rangle$ and FREQZM was performed at each grid column in the simulations, and indicates that $\langle f_d \rangle \langle \omega_d \rangle$ is skillful at predicting FREQZM, with greater downward motion leading to less frequent convective triggering. This relationship is only strong in the deep tropics and in the polar regions.

The Zhang and McFarlane (1995) deep convection scheme (ZM scheme) uses a Convective Available Potential Energy (CAPE) threshold as the convective trigger, and the high negative correlation between subsiding motion and the trigger is consistent with the dependence of CAPE on vertical advection of dry static energy and moisture (Zhang, 2002). The precipitation rates in the $15^\circ - 20^\circ$ latitude region consist primarily of drizzle from the ZM scheme. The convective trigger is weakly correlated with rather strong subsiding motion in this region, but the latent heat fluxes into the boundary layer are at a global maximum. It is suggested that the downward motion is opposing the generation of CAPE by the latent heat fluxes of this region, but due to the overwhelming magnitude of the latent heat fluxes, cannot depress CAPE below the trigger threshold. The resultant CAPE values are small and ZM scheme produces drizzle most of the time. The excess simulation of drizzle

in climate models is a known model bias (Dai, 2006), and it is suggested here that this bias arises from the use of CAPE as the sole criterion for convective triggering.

The conclusion of this study builds on previous work (Herrington and Reed, 2017, 2018, Chapter 5), and it is of the following relationships between parameterized convection, grid-scale clouds and resolution sensitivity in CAM. When the ZM scheme is triggered, it moistens the middle-to-upper troposphere through detrainment (Zhang and McFarlane, 1995), moisture which is then condensed by the stratiform cloud scheme to produce grid-scale clouds. The buoyancy produced by grid-scale clouds leads to large magnitude upward motion by the dynamical core, and to conserve mass, downward motion radiating away from the grid-scale clouds as horizontally propagating gravity waves. The grid-scale clouds are grid-limited; their horizontal size is set by the effective resolution of the model. The vertical velocities associated with the buoyancy of stratiform clouds are inversely proportional to their horizontal extent, and therefore scale like Δx^{-1} . Due to the greater efficiency of vertical mass transport with increasing resolution, the area of ascending motion decreases with resolution. The area and magnitude of the subsiding regions increase with resolution in order to conserve mass, which has the effect of stabilizing and drying the atmosphere with increasing resolution. The ZM scheme, with its stability dependent CAPE trigger, responds through becoming less active with increasing resolution.

7 Conclusions

The requirement that Atmospheric General Circulation Models (AGCMs) exhibit convergent solutions with increasing horizontal resolution was not historically pursued by the community, and as a result many AGCM solutions, such as NCAR’s Community Atmosphere Model (CAM), change drastically with resolution. Convergent solutions were likely not pursued during the early AGCM development phase (1950s through the 1980s) due to computational limitations.

The expanse of computing power beginning around the 1990s permitted convergence tests that indicated AGCMs have weak convergence properties. This issue had largely been ignored, or at least had not become a priority to model developers until recently. Dynamical cores are now more commonly developed with unstructured meshes, which allows for substantial grid flexibility and bringing to forefront the issue of convergence. The pursuit of *scale-aware physics* is in essence an effort to develop closures that result in AGCM solutions to converging with increasing resolution, and that those solutions converge to some reasonable depiction of the atmosphere we live in.

7.1 Summary

This work is an effort to understand the specific reasons CAM produces very different solutions with increasing horizontal resolution —which components of the model are responding to a change in resolution, why is it sensitive to resolution and how does it feedback onto other model components. CAM is an extremely complex model. Understanding why CAM simulates a certain

feature a certain way is by no mean trivial. Issac Held authored a paper in 2005 titled *The Gap between Simulation and Understanding in Climate Modeling* (Held, 2005), advocating for the community to shift focus more towards understanding these models. He suggests using a hierarchy of models, from the simplest to the most complex, in order to truly nail down the underlying causes of a phenomenon in an AGCM. This is the approach taken in this thesis.

Chapter 2 presents the sensitivity of the mean state of CAM, version 4 to horizontal resolutions typical of present-day AGCMs in an aqua-planet configuration. Weak convergence of the mean state is characterized by a progressive drying of the atmosphere and large reductions in cloud coverage with increasing resolution. Analyses of energy and moisture budgets indicate that these trends are balanced by variations in moisture transport by the resolved circulation, and a reduction in activity of the convection scheme. In contrast, the large-scale precipitation rate increases with resolution, which is approximately balanced by greater advection of dry static energy associated with more active resolved vertical motion in the ascent region of the Hadley cell.

An explanation for the sensitivity of the mean state to horizontal resolution is proposed, based on linear Boussinesq theory. The authors hypothesize that an increase in horizontal resolution in the model leads to a reduction in horizontal scale of the diabatic forcing arising from the column physics, facilitating finescale flow and faster resolved convective updrafts within the dynamical core, and steering the coupled system toward a new mean state. This hypothesis attempts to explain the underlying mechanism driving the

variations in moisture transport observed in the simulations.

In Chapter 3, a set of idealized experiments are developed in CAM to understand the vertical velocity response to reductions in forcing scale that occurs when the horizontal resolution is increased. The test consists of a set of rising bubble experiments, in which the horizontal radius of the bubble and the model grid spacing are simultaneously reduced. The test is performed with moisture, through incorporating moist physics routines of varying complexity, although convection schemes are not considered. Results confirm that the vertical velocity in CAM is to first-order, proportional to the inverse of the horizontal forcing scale, which is consistent with a scale analysis of the dry equations of motion. In contrast, experiments in which the physics time step Δt_{phys} are relaxed back to more conventional values results in severely damped vertical motion at high resolution, degrading the scaling. A set of aqua-planet simulations using different physics time steps are found to be consistent with the results of the idealized experiments.

The following two chapters arose from work performed during a year-long visit at NCAR as an Advanced Study Program(ASP) graduate student visitor. Chapter 4 contains a detailed analysis of the physics-dynamics coupling in CAM’s spectral-element dynamical core (CAM-SE), which is uses a continuous Galerkin method, and introduces a new coupling approach. Atmospheric modeling with element-based high-order Galerkin methods presents a unique challenge to the conventional physics-dynamics coupling paradigm, due to the highly irregular distribution of nodes within an element and the distinct numerical characteristics of the Galerkin method. The conventional coupling

procedure is to evaluate the physical parameterizations (*physics*) on the dynamical core grid. Evaluating the physics at the nodal points exacerbates numerical noise from the Galerkin method, enabling and amplifying local extrema at element boundaries.

Grid imprinting may be substantially reduced through the introduction of an entirely separate, approximately isotropic finite-volume grid for evaluating the physics forcing. Integration of the spectral basis over the control-volumes provides an area average state to the physics, which is more representative of the state in the vicinity of the nodal points rather than the nodal point itself, and is more consistent with the notion of a “large-scale state” required by conventional physics packages. This study documents the implementation of a quasi-equal area physics grid into CAM-SE, and is shown to be effective at mitigating grid imprinting in the solution. The physics grid is also appropriate for coupling to other components within the Community Earth System Model, since the coupler requires component fluxes to be defined on a finite-volume grid, and one can be certain that the fluxes on the physics grid are indeed, volume-averaged.

Chapter 5 describes the implementation of a coarser resolution physics grid into CAM-SE. The dry dynamics is represented by the spectral element dynamical core and tracer transport is computed using the Conservative Semi-Lagrangian Finite Volume Method (CAM-SE-CSLAM). Algorithms are presented that map fields between the dynamics and physics grids while maintaining numerical properties ideal for atmospheric simulations such as mass conservation and mixing ratio shape and linear-correlation preservation. The

results of experiments using the lower resolution physics grid are compared to the conventional method in which the physics and dynamical grids coincide. The lower resolution physics grid consists of control volumes designed to provide an isotropic representation of the dynamics to the physical parameterizations, and eliminates grid imprinting, even in regions with steep topography.

The impact of the coarser resolution physics grid on the resolved scales of motion is analyzed in an aqua-planet configuration, across a range of dynamical core grid resolutions. Through analyzing the relationship between the physics forcing and resolved vertical motion, it was determined that the effective resolution of the model is not degraded through the use of a coarser resolution physics grid. Since the physics makes up about half the computational cost of the conventional CAM-SE-CSLAM configuration, the coarser physics grid may allow for significant cost savings with little to no downside.

Chapter 6 is the final chapter, and puts together a full picture on the causes of resolutions sensitivity in aqua-planets via a thorough analysis of the CAM-SE-CSLAM simulations from Chapter 5. Through scaling the Δt_{phys} in the aqua-planets such that the large time-truncation errors discovered in Chapter 3 may be avoided, it was found that the vertical velocity field everywhere in the model is indeed inversely proportional to the grid spacing of the model. The scaling arises because grid-scale clouds are grid limited; their horizontal extent is set by the effective resolution of the model. The dynamics reacts to grid limited clouds through pressure perturbations of the same scale, and so the pressure gradients, and likewise the vertical velocity both scale in proportion to the inverse of the grid spacing. The scaling indicates that for

a doubling of the resolution, the resolved mass flux may be achieved in half the area, which then confines the ascending regions to smaller regions of the model with increasing resolution. To conserve mass, subsiding motion both becomes faster and encompasses a larger region in the model with increasing resolution, particularly in the Tropics, which creates a more stable and drier mean state. The tendency towards atmospheric stability with increasing resolution reduces the frequency the deep convection scheme is triggered, and convective precipitation becomes a smaller fraction of the total precipitation rate.

7.2 Significance and future work

A solid understanding of the causes of resolution sensitivity in an AGCM is useful in many regards. Many AGCMs now support flexible grid structures, such as variable-resolution grids (VR). VR grids are attractive since they can resolve regional climate processes in a global model, but resolution sensitivity may degrade their value since solutions may be permitted to diverge across the refinement region. To be clear, there are many successful VR studies that do not suffer from the convergence issue, and it is generally safe to use refinement over higher-latitude regions due to relatively subdued buoyancy induced vertical velocities, compared to the tropics. But to completely reconcile this deficiency requires the development of scale-aware physics. Scale-aware physics should be built in part, to anticipate and correct the sensitivity of AGCMs to horizontal resolution, which of course, requires knowledge of the causes of resolution sensitivity.

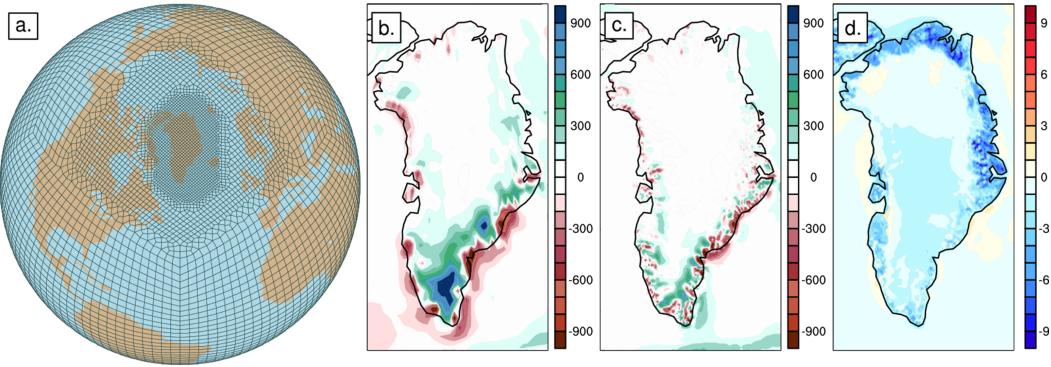


Figure 49: (a.) Variable resolution grid proposed for the pre-industrial control; grid lines delineate the element boundaries of the spectral-element dynamical core, corresponding to an approximate grid spacing of 111.2km in the coarse region and 27.8km in the refined region. CAM-SE *F* compset simulations using (b.) a uniform 111.2km grid and (c.), (d.) CAM-VR (van Kampenhout et al., 2018). Shown are the climatological (1980-1999) (b.), (c.) annually accumulated precipitation difference (mm w.e.) and (d.) JJA 2-meter air temperature difference (K) from the RACMO2 regional climate model simulation.

CAM-SE supports variable resolution grids (CAM-VR), and due to the vast modeling experience that came of this thesis, I was able to assist in a CAM-VR study with regional grid refinement over the big ice sheet in Greenland (Figure 49a; van Kampenhout et al., 2018). Recent attempts to couple NCAR’s ice sheet model to CAM results in unrealistic ice sheet expansion of the Greenland Ice Sheet due to a high-precipitation bias in southern Greenland, and a summer cold bias on the coastal margins (W. Lipscomb, personal communication). Using CAM-VR, the precipitation bias is successfully improved (Figures 49b,c), indicating the usefulness of the VR approach. Unfortunately, the summer temperature bias remains (Figure 49d). The author has plans to understand and reconcile this bias as an ASP Postdoctoral at NCAR, which would alleviate a major obstacle to realistic simulations of the

Greenland Ice Sheet.

8 APPENDICES

8.1 Mapping the physics tendencies to the GLL grid

The mapping of the physics tendencies from the physics grid to the GLL grid is done with tensor-cubic Lagrange interpolation. The elements of the cubed-sphere in SE are created from an equi-angular gnomonic projection. Consider one element $(\alpha, \beta) \in [\alpha_1^{(elem)}, \alpha_2^{(elem)}] \times [\beta_1^{(elem)}, \beta_2^{(elem)}]$, where (α, β) are central angle coordinates and $\alpha_1^{(elem)}$ and $\alpha_2^{(elem)}$ are the minimum and maximum central angles in the α -coordinate direction, respectively, and similarly for β . Let $\Delta\alpha^{(elem)} = \alpha_2^{(elem)} - \alpha_1^{(elem)}$ and $\Delta\beta^{(elem)} = \beta_2^{(elem)} - \beta_1^{(elem)}$. The physics grid cell central angle centers are located at

$$(\alpha_i^{(pg)}, \beta_j^{(pg)}) = \left[\alpha_1^{(elem)} + \left(i - \frac{1}{2}\right) \Delta\alpha^{(pg)}, \beta_1^{(elem)} + \left(j - \frac{1}{2}\right) \Delta\beta^{(pg)} \right], \quad (54)$$

where $\Delta\alpha^{(pg)} = \Delta\beta^{(pg)} = \frac{\Delta\alpha^{(elem)}}{pg} = \frac{\Delta\beta^{(elem)}}{pg}$. The interpolation is performed in central-angle coordinates using tensor product cubic interpolation. For elements located on a cubed-sphere edge or corner the coordinate system for neighboring elements may be on a different panel. To take into account this coordinate change the central angle locations of physics grid cell centers located on other panels are transformed to the coordinate system of the panel the element in question is located on (the transformations are given in, e.g., Nair et al., 2005). An illustration is given in Figure 50 for an element located in the lower left corner of a panel. The element in question is $(\xi, \chi) \in (-1, 1)^2$ where,

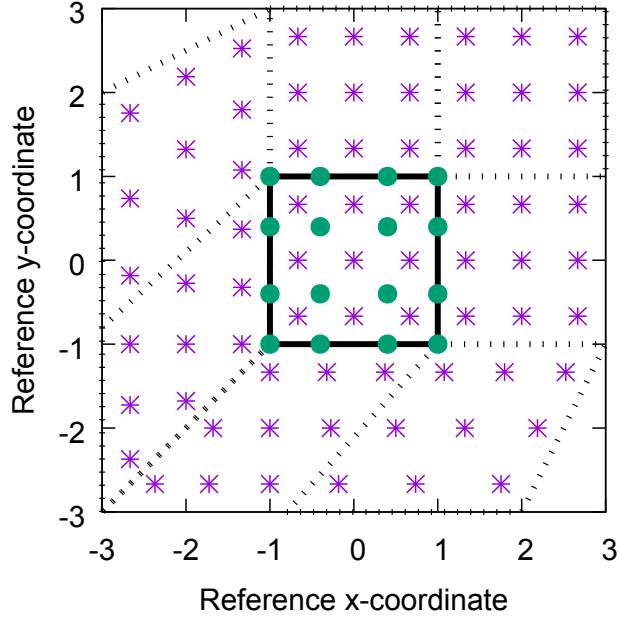


Figure 50: Schematic of the coordinate system in which the dimensionally split cubic Lagrange interpolation is computed. The physics grid centers are marked with asterisks and the GLL points, we are interpolating to, with solid filled circles. The element in which the GLL points are located is bounded by thick black lines and located in the lower left corner of a panel. The stippled lines mark the boundaries of the remaining elements. For simplicity we are using the normalized coordinate centered at the element on which the GLL points we are interpolating to are located. Note that the coordinates for points on neighboring panels (using a different local coordinate system) must be transformed to the coordinate system of the element in question.

for simplicity, we have transformed the element coordinates into normalized coordinates $(\xi, \chi) = \left(\frac{2(\alpha^{(pg)} - \alpha_1^{(elem)})}{\Delta\alpha^{(elem)}} - 1, \frac{2(\beta^{(pg)} - \beta_1^{(elem)})}{\Delta\beta^{(elem)}} - 1 \right)$; also used internally in the SE dynamical core (see, e.g., section 3.3 in Lauritzen et al., 2018). The GLL points are located at $-1, -1/\sqrt{1}, 1/\sqrt{5}$, and 1 in each coordinate direction. Near the edges/corners of an element cubic extrapolation is used if the centered stencil expands beyond the panel.

8.2 Defining Δt_{phys} across resolutions

Herrington and Reed (2018) developed a moist bubble test, which indicate that time-truncation errors are large at high resolution (about 50km or less) using more conventional values for the physics time-step. The test may be able to provide insight on a reasonable scaling of Δt_{phys} across resolutions in more complex configurations. In the test a set of non-rotating simulations are initialized with a warm, super-saturated moist bubble, and the grid spacing and bubble radius are simultaneously reduced by the same factor in each run through varying the planetary radius. The test was designed to mimic the reduction in buoyancy length scales that occur when the model resolution is increased in more complex configurations (Hack et al., 2006; Herrington and Reed, 2018).

The moist bubble test is performed with CAM-SE-CSLAM and coupled to the simple condensation routine of Kessler (1969) across five different resolutions (pertaining to the *ne30*, *ne40*, *ne60*, *ne80*, and *ne120* grids). The results are expressed as the minimum ω throughout each one day simulation, and shown in Figure 51. Two sets of simulations are performed with both *pg3* and *pg2*, one with Δt_{phys} determined by equation (51), and an equivalent set of simulations with $\Delta t_{phys} = 1800s$ for all resolutions.

With the diameters of the bubbles set proportional to Δx_{dyn} , Herrington and Reed (2018) has shown that ω converges to the scaling of equation (48) in the limit of small Δt_{phys} , where small Δt_{phys} refers to the CFL limiting time-step used by the dynamics. Equation (48) is overlain as grey lines in Figure 51, with *ne30* being the reference resolution. The solutions using Δt_{phys} from equa-

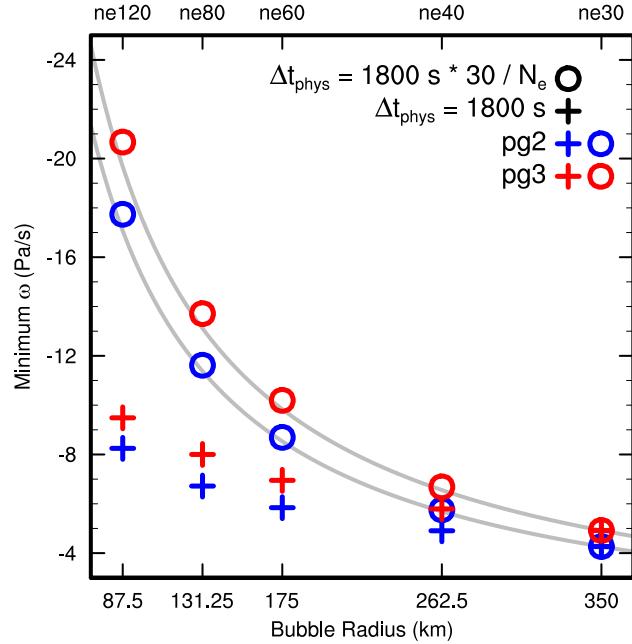


Figure 51: The magnitude of ω in the $pg3$ solutions are systematically larger than the $pg2$ solutions, which is primarily a result of the damping effect of integrating the basis functions over a larger control volume.

tion (51) follow the scaling, whereas fixing $\Delta t_{phys} = 1800s$ across resolutions damps the solution relative to the analytical solution, progressively more so at higher resolutions. If Δt_{phys} is too large, the solution has non-negligible error, which is avoided through scaling Δt_{phys} according to equation (51).

To get a handle on whether the test is useful for understanding more realistic configurations, four aqua-planet simulations are performed using the CAM6 physics package. A pair of $ne30pg2$ simulations, one in which Δt_{phys} is set to the appropriate value from equation (51) ($1800s$), and another where it is set to the Δt_{phys} corresponding to the $ne20$ resolution ($2700s$). Similarly, a pair of $ne120pg2$ simulations are performed, one with Δt_{phys} set to the value

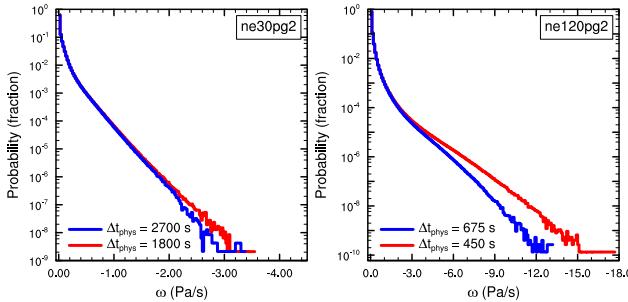


Figure 52: Probability density distribution of upward ω everywhere in the model in the aqua-planets using the *ne30pg2* grid (Left) and the *ne120pg2* grid (Right). Figure computed for one year of 6-hourly data. The different colors indicate the physics time-steps used in the runs.

from equation (51) (450s), and one with Δt_{phys} set to the *ne80* value (675s).

Figure 52 shows the PDFs of upward ω computed from a year of six-hourly data in the simulations. At lower resolution, Δt_{phys} has only a very small effect on the solution, near the tale-end of the distributions. At high-resolution, values of ω less than about $-3Pa/s$ are more frequent in the small Δt_{phys} run, with the discrepancy growing more for larger magnitudes of ω . The progressively larger errors with increasing resolution also manifests in the moist bubble tests, indicating that truncation errors arising from large Δt_{phys} do exist in more complex configurations.

8.3 The impact of high-order mapping to the dynamics grids

Figure 53a shows a close-up of the wavenumber power spectrum of the forcing on the *pg* grid (dotted), where it is computed, and on the *GLL* grid (solid), where it is has been mapped. In *ne30pg3*, the magnitudes are similar on

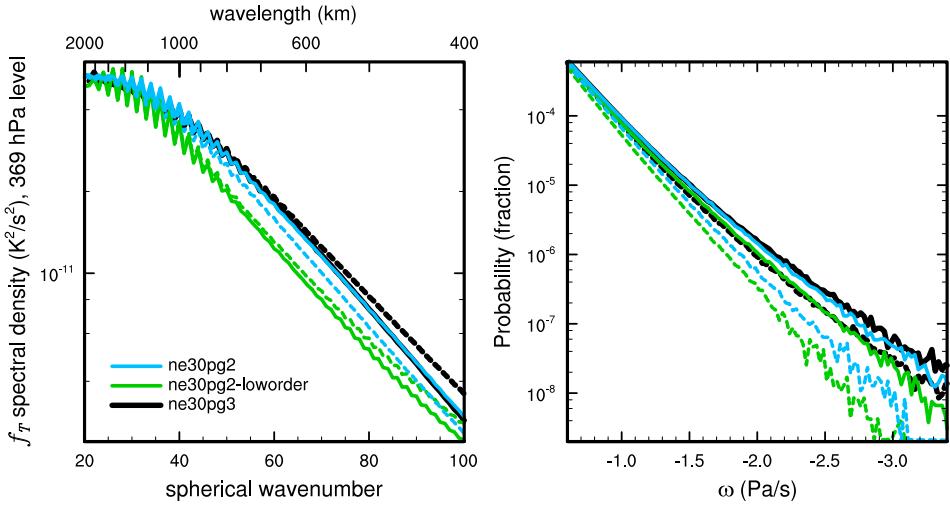


Figure 53: (Left) Wavenumber-power spectrum of the temperature tendencies from the moist physics, at the 369 hPa level, and (right) probability density distribution of upward ω , everywhere in the model, for three year-long aqua-planet simulations. Solid lines refer to values of on the *GLL* grids, and dashed lines, the fields on the *pg* grids. See text for details regarding the three simulations.

both grids, except the mapping tends to damp the high wavenumbers of the forcing on the *GLL* grid (greater than 60), but these scales are primarily below the effective resolution of the model and should not effect the solution. For *ne30pg2*, the magnitude of the forcing is actually greater after mapping to the *GLL* grid, and more similar to the forcing in the *ne30pg3* simulations. The high-order mapping can therefore replicate the scales of the physics tendencies that occur in the *pg3* simulation, even though the physics are evaluated on a coarser *pg2* grid.

The importance of the high-order mapping can be shown with an additional *ne30pg2* simulation, using low-order mapping (*ne30pg2 – loworder* in Figure 53). Specifically, low-order mapping refers to piecewise constant map-

ping between the *pg2* and *CSLAM* grids, and bi-linear mapping from *pg2* to the *GLL* grid. The forcing spectrum is now similar on both the *pg2* and *GLL* grids, although the low-order mapping tends to damp the forcing on the *GLL* grid for wavenumbers greater than about 60, scales smaller than the models effective resolution (Figure 53a). A close up of the PDF of $\omega^{(gll)}$ is provided in Figure 53b (solid lines). As expected, the frequency of large magnitude $\omega^{(gll)}$ in the low-order run is less compared to the default *ne30pg2* simulation.

The dotted lines in Figure 53b show the PDF of ω on the *pg* grids. The frequency of large magnitude ω is reduced on the *pg* grids, compared to the state on the *GLL* grids. This is primarily due to the smoothing effect of integrating the nodal point values over control volumes (H18). The larger ω values are even less frequent on the *pg2* grid due to integrating over control volumes $\frac{9}{4}$ times greater than the *pg3* control volumes.

REFERENCES

- A. Arakawa and W. H. Schubert. Interaction of a cumulus cloud ensemble with the large-scale environment, Part I. *J. Atmos. Sci.*, 31:674–701, 1974.
- A. Arakawa and C.-M. Wu. A unified representation of deep moist convection in numerical modeling of the atmosphere. part i. *Journal of the Atmospheric Sciences*, 70(7):1977–1992, 2013.
- J. T. Bacmeister, K. A. Reed, C. Hannay, P. Lawrence, S. Bates, J. E. Truesdale, N. Rosenbloom, and M. Levy. Projected changes in tropical cyclone activity under future warming scenarios using a high-resolution climate model. *Climatic Change*, 146:547–560, 2018. doi: 10.1007/s10584-016-1750-x. URL <http://dx.doi.org/10.1007/s10584-016-1750-x>.
- F. Baer, H. Wang, J. J. Tribbia, and A. Fournier. Climate modeling with spectral elements. *Mon. Wea. Rev.*, 134:3610–3624, 2006. doi: 10.1175/MWR3360.1. URL <http://dx.doi.org/10.1175/MWR3360.1>.
- T. Barth and D. Jespersen. The design and application of upwind schemes on unstructured meshes. *Proc. AIAA 27th Aerospace Sciences Meeting, Reno*, 1989.
- A. Beljaars. Numerical schemes for parameterizations. In *Proc. ECMWF Seminar on Numerical Methods in Atmospheric Models, Reading, United Kingdom, ECMWF, 1991*, volume 2, pages 1–42, 1991.
- J. Berner, T. Jung, and T. N. Palmer. Systematic model error: The impact of increased horizontal resolution versus improved stochastic and deterministic parameterizations. *Journal of Climate*, 25(14):4946–4962, 2012. doi: 10.1175/jcli-d-11-00297.1.

- P. A. Bogenschutz, A. Gettelman, H. Morrison, V. E. Larson, C. Craig, and D. P. Schanen. Higher-order turbulence closure and its impact on climate simulations in the community atmosphere model. *Journal of Climate*, 26(23):9655–9676, 2013.
- B. A. Boville. Sensitivity of simulated climate to model resolution. *J. Climate*, 4(5):469–485, 1991. doi: 10.1175/1520-0442(1991)004<0469:sosctm>2.0.co;2. URL [http://dx.doi.org/10.1175/1520-0442\(1991\)004<0469:sosctm>2.0.co;2](http://dx.doi.org/10.1175/1520-0442(1991)004<0469:sosctm>2.0.co;2).
- S. Brdar, M. Baldauf, A. Dedner, and R. Klöfkorn. Comparison of dynamical cores for nwp models: comparison of cosmo and dune. *Theoretical and Computational Fluid Dynamics*, 27(3-4):453–472, 2013. doi: 10.1007/s00162-012-0264-z.
- C. S. Bretherton and S. Park. A new moist turbulence parameterization in the community atmosphere model. *Journal of Climate*, 22(12):3422–3448, 2009.
- G. H. Bryan and J. M. Fritsch. A benchmark simulation for moist nonhydrostatic numerical models. *Monthly Weather Review*, 130(12):2917–2928, 2002. doi: 10.1175/1520-0493(2002)130<2917:absfmn>2.0.co;2.
- C. Canuto, M. Y. Hussaini, A. Quarteroni, and T. Zang. *Spectral Methods: Evolution to Complex Geometries and Applications to Fluid Dynamics*. Springer, 1 edition, 7 2007. ISBN 9783540307273.
- J. Charney. Geostrophic turbulence. *J. Atmos. Sci.*, 28:1087–1095, 1971.
- W. Collins. A new parameterization for infrared emission and absorption by water vapor in the national center for atmospheric research community

- atmosphere model. *Journal of Geophysical Research*, 107(22):1–20, 2002.
- W. D. Collins, P. J. Rasch, B. A. Boville, J. J. Hack, J. R. McCaa, D. L. Williamson, J. T. Kiehl, B. Briegleb, C. Bitz, S.-J. Lin, M. Zhang, and Y. Dai. Description of the NCAR Community Atmosphere Model (CAM 3.0). Technical Report TN-464+STR, National Center for Atmospheric Research, 2004. 214 pp.
- Computational and Information Systems Laboratory. Cheyenne: HPE/SGI ICE XA System (AMP Model Developments), 2017.
- A. Dai. Precipitation characteristics in eighteen coupled climate models. *Journal of Climate*, 19(18):4605–4630, 2006.
- P. Das. A non-archimedean approach to the equations of convection dynamics. *Journal of the Atmospheric Sciences*, 36(11):2183–2190, 1979. doi: 10.1175/1520-0469(1979)036<2183:anaatt>2.0.co;2.
- J. M. Dennis, J. Edwards, K. J. Evans, O. Guba, P. H. Lauritzen, A. A. Mirin, A. St-Cyr, M. A. Taylor, and P. H. Worley. CAM-SE: A scalable spectral element dynamical core for the Community Atmosphere Model. *Int. J. High. Perform. C.*, 26(1):74–89, 2012. doi: 10.1177/1094342011428142.
URL <http://hpc.sagepub.com/content/26/1/74.abstract>.
- B. Dong and P. J. Valdes. Climates at the last glacial maximum: Influence of model horizontal resolution. *Journal of climate*, 13(9):1554–1573, 2000.
- D. Durran. *Numerical Methods for Fluid Dynamics: With Applications to Geophysics*, volume 32 of *Texts in Applied Mathematics*. Springer, 2 edition, 2010. 516 p.
- A. Fournier, M. A. Taylor, and J. J. Tribbia. The spectral element atmo-

- sphere model (SEAM): High-resolution parallel computation and localized resolution of regional dynamics. *Mon. Wea. Rev.*, 132(3):726–748, 2004.
- M. S. Fox-Rabinovitz, G. L. Stenchikov, M. J. Suarez, L. L. Takacs, and R. C. Govindaraju. A uniform- and variable-resolution stretched-grid gcm dynamical core with realistic orography. *Mon. Wea. Rev.*, 128:1883–1898, 2000.
- D. M. W. Frierson. The dynamics of idealized convection schemes and their effect on the zonally averaged tropical circulation. *Journal of the Atmospheric Sciences*, 64(6):1959–1976, 2007. doi: 10.1175/jas3935.1.
- J. Galewsky, A. Sobel, and I. Held. Diagnosis of subtropical humidity dynamics using tracers of last saturation. *Journal of the atmospheric sciences*, 62(9):3353–3367, 2005.
- A. Gettelman, H. Morrison, S. Santos, P. Bogenschutz, and P. Caldwell. Advanced two-moment bulk microphysics for global models. part ii: Global model solutions and aerosol–cloud interactions. *Journal of Climate*, 28(3):1288–1307, 2015.
- F. Giraldo and M. Restelli. A study of spectral element and discontinuous galerkin methods for the Navier-Stokes equations in nonhydrostatic mesoscale atmospheric modeling: Equation sets and test cases. *J. Comput. Phys.*, 227(8):3849 – 3877, 2008. doi: <http://dx.doi.org/10.1016/j.jcp.2007.12.009>.
- J.-C. Golaz, V. E. Larson, and W. R. Cotton. A pdf-based model for boundary layer clouds. part i: Method and model description. *Journal of the Atmospheric Sciences*, 59(24):3540–3551, 2002. doi: 10.1175/1520-0469(2002)

- 059⟨3540:apbmfb⟩2.0.co;2.
- W. W. Grabowski and T. L. Clark. Cloud–environment interface instability: Rising thermal calculations in two spatial dimensions. *Journal of the Atmospheric Sciences*, 48(4):527–546, 1991. doi: 10.1175/1520-0469(1991)048<0527:ciirtc>2.0.co;2.
- F. F. Grinstein, L. G. Margolin, and W. J. Rider, editors. *Implicit Large Eddy Simulation: Computing turbulent fluid dynamics*. Cambridge University Press, 2007.
- O. Guba, M. Taylor, and A. St-Cyr. Optimization-based limiters for the spectral element method. *J. Comput. Phys.*, 267(0):176 – 195, 2014a. doi: <http://dx.doi.org/10.1016/j.jcp.2014.02.029>.
- O. Guba, M. A. Taylor, P. A. Ullrich, J. R. Overfelt, and M. N. Levy. The spectral element method (sem) on variable-resolution grids: evaluating grid sensitivity and resolution-aware numerical viscosity. *Geosci. Model Dev.*, 7(6):2803–2816, 2014b. doi: 10.5194/gmd-7-2803-2014.
- J. E. Guerra and P. A. Ullrich. A high-order staggered finite-element vertical discretization for non-hydrostatic atmospheric models. *Geosci. Model Dev.*, 9(5):2007–2029, 2016. doi: 10.5194/gmd-9-2007-2016. URL <https://www.geosci-model-dev.net/9/2007/2016/>.
- J. Hack, B. Boville, B. Briegleb, J. Kiehl, and D. Williamson. Description of the NCAR Community Climate Model (CCM2). *NCAR Technical Note*, NCAR/TN-382+STR, 1993. doi: 10.5065/D6QZ27XV.
- J. Hack, M. Caron, G. Danabasoglu, K. W. Oleson, C. Bitz, and J. Truesdale. Ccsm–cam3 climate simulation sensitivity to changes in horizontal

- resolution. *J. Climate*, 19(1):2269–2289, 2006. doi: 10.1175/JCLI3764.1.
- J. J. Hack. Parameterization of moist convection in the national center for atmospheric research community climate model (ccm2). *Journal of Geophysical Research: Atmospheres*, 99(D3):5551–5568, 1994.
- L. M. Harris, S.-J. Lin, and C. Tu. High-resolution climate simulations using gfdl hiram with a stretched global grid. *Journal of Climate*, 29(11):4293–4314, 2016. doi: 10.1175/jcli-d-15-0389.1.
- I. M. Held. The gap between simulation and understanding in climate modeling. *Bulletin of the American Meteorological Society*, 86(11):1609–1614, 2005.
- I. M. Held and A. Y. Hou. Nonlinear axially symmetric circulations in a nearly inviscid atmosphere. *Journal of the Atmospheric Sciences*, 37(3):515–533, 1980.
- I. M. Held and T. Schneider. The surface branch of the zonally averaged mass transport circulation in the troposphere. *Journal of the atmospheric sciences*, 56(11):1688–1697, 1999.
- I. M. Held and M. J. Suarez. A proposal for the intercomparison of the dynamical cores of atmospheric general circulation models. *Bull. Am. Meteorol. Soc.*, 73:1825–1830, 1994.
- A. Herrington and K. Reed. An idealized test of the response of the community atmosphere model to near-grid-scale forcing across hydrostatic resolutions. *J. Adv. Model. Earth Syst.*, 10(2):560–575, 2018.
- A. Herrington, P. Lauritzen, M. A. Taylor, S. Goldhaber, B. E. Eaton, J. Bacmeister, K. Reed, and P. Ullrich. Physics-dynamics coupling with

- element-based high-order galerkin methods: quasi equal-area physics grid. *Mon. Wea. Rev.*, 47:69–84, 2018. doi: 10.1175/MWR-D-18-0136.1.
- A. R. Herrington and K. A. Reed. An explanation for the sensitivity of the mean state of the community atmosphere model to horizontal resolution on aquaplanets. *J. Climate*, 30(13):4781–4797, 2017. doi: 10.1175/jcli-d-16-0069.1. URL <http://dx.doi.org/10.1175/jcli-d-16-0069.1>.
- J. R. Holton. *An Introduction to Dynamic Meteorology*. Elsevier Academic Press, Amsterdam, fourth edition edition, 2004.
- A. Holtslag and B. Boville. Local versus nonlocal boundary-layer diffusion in a global climate model. *Journal of Climate*, 6(10):1825–1842, 1993.
- G. J. Huffman, R. F. Adler, M. M. Morrissey, D. T. Bolvin, S. Curtis, R. Joyce, B. McGavock, and J. Susskind. Global precipitation at one-degree daily resolution from multisatellite observations. *J. Hydrometeorology*, 2(1):36–50, 2001.
- C. Jablonowski and D. L. Williamson. A baroclinic instability test case for atmospheric model dynamical cores. *Q. J. R. Meteorol. Soc.*, 132:2943–2975, 2006.
- C. Jablonowski and D. L. Williamson. The pros and cons of diffusion, filters and fixers in atmospheric general circulation models., in: P.H. Lauritzen, R.D. Nair, C. Jablonowski, M. Taylor (Eds.), Numerical techniques for global atmospheric models. *Lecture Notes in Computational Science and Engineering*, Springer, 80, 2011.
- N. Jeevanjee. Vertical velocity in the gray zone. *Journal of Advances in Modeling Earth Systems*, 9(6):2304–2316, 2017. doi: 10.1002/2017MS001059.

- N. Jeevanjee and D. M. Romps. Effective buoyancy at the surface and aloft. *Quart. J. Roy. Meteor. Soc.*, 142(695):811–820, 2016.
- P. Jöckel, R. von Kuhlmann, M. G. Lawrence, B. Steil, C. Brenninkmeijer, P. J. Crutzen, P. J. Rasch, and B. Eaton. On a fundamental problem in implementing flux-form advection schemes for tracer transport in 3-dimensional general circulation and chemistry transport models. *Q.J.R. Meteorol. Soc.*, 127(573):1035–1052, April 2001.
- P. W. Jones. First- and second-order conservative remapping schemes for grids in spherical coordinates. *Mon. Wea. Rev.*, 127:2204–2210, 1999.
- E. Kessler. On the distribution and continuity of water substance in atmospheric circulations. *Meteorol. Monogr.*, 10(32):88, 1969.
- J. Kiehl and D. Williamson. Dependence of cloud amount on horizontal resolution in the national center for atmospheric research community climate model. *Journal of Geophysical Research: Atmospheres*, 96(D6):10955–10980, 1991.
- J. Kiehl, J. Hack, and B. Briegleb. The simulated earth radiation budget of the national center for atmospheric research community climate model ccm2 and comparisons with the earth radiation budget experiment (erbe). *Journal of Geophysical Research: Atmospheres*, 99(D10):20815–20827, 1994.
- J. T. Kiehl, R. J. Wolski, B. P. Briegleb, and V. Ramanathan. Documentation of radiation and cloud routines in the ncar community climate model (ccm1). Technical report, NCAR Tech. Note NCAR/TN-288+ IA, 1987.
- Y.-J. Kim, F. X. Giraldo, M. Flatau, C.-S. Liou, and M. S. Peng. A sensitivity study of the kelvin wave and the Madden-Julian oscillation in aqua-

- planet simulations by the Naval Research Laboratory Spectral Element Atmospheric Model. *J. Geo. Res.: Atmospheres*, 113(D20), 2008. doi: 10.1029/2008JD009887. D20102.
- J. B. Klemp and R. B. Wilhelmson. The simulation of three-dimensional convective storm dynamics. *Journal of the Atmospheric Sciences*, 35(6): 1070–1096, 1978. doi: 10.1175/1520-0469(1978)035<1070:tsotdc>2.0.co;2.
- J. B. Klemp, W. C. Skamarock, and S.-H. Park. Idealized global non-hydrostatic atmospheric test cases on a reduced-radius sphere. *Journal of Advances in Modeling Earth Systems*, 7(3):1155–1177, 2015. doi: 10.1002/2015ms000435.
- J. Lander and B. Hoskins. Believable scales and parameterizations in a spectral transform model. *Mon. Wea. Rev.*, 125:292–303., 1997. doi: 10.1175/1520-0493.
- P. Lauritzen and J. Thuburn. Evaluating advection/transport schemes using interrelated tracers, scatter plots and numerical mixing diagnostics. *Quart. J. Roy. Met. Soc.*, 138(665):906–918, 2012. doi: 10.1002/qj.986.
- P. H. Lauritzen and D. L. Williamson. A total energy error analysis of dynamical cores and physics-dynamics coupling in the community atmosphere model (cam). *J. Adv. Model. Earth Syst.*, 2019. doi: 10.1029/2018MS001549.
- P. H. Lauritzen, R. D. Nair, and P. A. Ullrich. A conservative semi-Lagrangian multi-tracer transport scheme (CSLAM) on the cubed-sphere grid. *J. Comput. Phys.*, 229:1401–1424, 2010. doi: 10.1016/j.jcp.2009.10.036.
- P. H. Lauritzen, A. Mirin, J. Truesdale, K. Raeder, J. Anderson, J. Bacmeister, and R. B. Neale. Implementation of new diffusion/filtering operators in the

CAM-FV dynamical core. *Int. J. High Perform. Comput. Appl.*, 2011a. doi: 10.1177/1094342011410088.

P. H. Lauritzen, P. A. Ullrich, and R. D. Nair. Atmospheric transport schemes: desirable properties and a semi-Lagrangian view on finite-volume discretizations, in: P.H. Lauritzen, R.D. Nair, C. Jablonowski, M. Taylor (Eds.), Numerical techniques for global atmospheric models. *Lecture Notes in Computational Science and Engineering*, Springer, 2011, 80, 2011b. doi: 10.1007/978-3-642-11640-7_8.

P. H. Lauritzen, W. C. Skamarock, M. J. Prather, and M. A. Taylor. A standard test case suite for 2d linear transport on the sphere. *Geo. Geosci. Model Dev.*, 5:887–901, 2012.

P. H. Lauritzen, J. T. Bacmeister, P. F. Callaghan, and M. A. Taylor. Ncar global model topography generation software for unstructured grids. *Geoscientific Model Development Discussions*, 8(6):4623–4651, 2015a. doi: 10.5194/gmdd-8-4623-2015.

P. H. Lauritzen, A. J. Conley, J.-F. Lamarque, F. Vitt, and M. A. Taylor. The terminator ”toy” chemistry test: a simple tool to assess errors in transport schemes. *Geoscientific Model Development*, 8(5):1299–1313, 2015b. doi: 10.5194/gmd-8-1299-2015.

P. H. Lauritzen, M. A. Taylor, J. Overfelt, P. A. Ullrich, R. D. Nair, S. Goldhaber, and R. Kelly. CAM-SE-CSLAM: Consistent coupling of a conservative semi-lagrangian finite-volume method with spectral element dynamics. *Mon. Wea. Rev.*, 145(3):833–855, 2017. doi: 10.1175/MWR-D-16-0258.1.

P. H. Lauritzen, R. Nair, A. Herrington, P. Callaghan, S. Goldhaber, J. Den-

- nis, J. T. Bacmeister, B. Eaton, C. Zarzycki, M. A. Taylor, A. Gettelman, R. Neale, B. Dobbins, K. Reed, and T. Dubos. NCAR CESM2.0 release of CAM-SE: A reformulation of the spectral-element dynamical core in dry-mass vertical coordinates with comprehensive treatment of condensates and energy. *J. Adv. Model. Earth Syst.*, 2018. doi: 10.1029/2017MS001257.
- L. R. Leung, T. Ringler, W. D. Collins, M. Taylor, and M. Ashfaq. A hierarchical evaluation of regional climate simulations. *Eos, Transactions American Geophysical Union*, 94(34):297–298, 2013.
- F. Li, W. D. Collins, M. F. Wehner, D. L. Williamson, J. G. Olson, and C. Algieri. Impact of horizontal resolution on simulation of precipitation extremes in an aqua-planet version of community atmospheric model (cam3). *Tellus A: Dynamic Meteorology and Oceanography*, 63(5):884–892, 2011.
- S.-J. Lin. A ‘vertically Lagrangian’ finite-volume dynamical core for global models. *Mon. Wea. Rev.*, 132:2293–2307, 2004.
- S.-J. Lin and R. B. Rood. An explicit flux-form semi-Lagrangian shallow-water model on the sphere. *Q.J.R.Meteorol.Soc.*, 123:2477–2498, 1997.
- E. Lindborg. Can the atmospheric kinetic energy spectrum be explained by two-dimensional turbulence? *Journal of Fluid Mechanics*, 388:259–288, 1999. doi: 10.1017/s0022112099004851.
- X. Liu, R. C. Easter, S. J. Ghan, R. Zaveri, P. Rasch, X. Shi, J.-F. Lamarque, A. Gettelman, H. Morrison, F. Vitt, et al. Toward a minimal representation of aerosols in climate models: Description and evaluation in the community atmosphere model cam5. *Geoscientific Model Development*, 5(3):709, 2012.
- V. Lorant and J.-F. Royer. Sensitivity of equatorial convection to horizontal

- resolution in aquaplanet simulations with a variable-resolution gcm. *Monthly Weather Review*, 129(11):2730–2745, 2001. doi: 10.1175/1520-0493(2001)129<2730:soecth>2.0.co;2.
- J. Lu, G. Chen, L. R. Leung, D. A. Burrows, Q. Yang, K. Sakaguchi, and S. Hagos. Toward the dynamical convergence on the jet stream in aquaplanet agcms. *Journal of Climate*, 28(17):6763–6782, 2015.
- Y. Maday and A. T. Patera. Spectral element methods for the incompressible Navier Stokes equations. In A. K. Noor and J. T. Oden, editors, *State of the Art Surveys on Computational Mechanics*, pages 71–143. ASME, New York, 1987.
- S. Manabe, J. Smagorinsky, and R. F. Strickler. Simulated climatology of a general circulation model with a hydrologic cycle. *Mon. Wea. Rev.*, 93(12):769–798, 1965.
- S. Marras, M. Nazarov, and F. X. Giraldo. Stabilized high-order galerkin methods based on a parameter-free dynamic sgs model for les. *Journal of Computational Physics*, 301:77–101, 2015. doi: 10.1016/j.jcp.2015.07.034.
- N. McFarlane. The effect of orographically excited gravity wave drag on the general circulation of the lower stratosphere and troposphere. *J. Atmos. Sci.*, 44:1775–1800, 1987. doi: 10.1175/1520-0469.
- B. Medeiros, D. L. Williamson, and J. G. Olson. Reference aquaplanet climate in the community atmosphere model, version 5. *J. Adv. Model. Earth Syst.*, 8(1):406–424, 2016. doi: 10.1002/2015MS000593.
- M. W. Moncrieff and E. Klinker. Organized convective systems in the tropical western pacific as a process in general circulation models: A toga coare

- case-study. *Quarterly Journal of the Royal Meteorological Society*, 123(540): 805–827, 1997. doi: 10.1002/qj.49712354002.
- H. Morrison and A. Gettelman. A new two-moment bulk stratiform cloud microphysics scheme in the community atmosphere model, version 3 (CAM3). Part I: Description and numerical tests. *J. Climate*, 21:3642–3659, 2008.
- C. J. Muller and P. A. O’Gorman. An energetic perspective on the regional response of precipitation to climate change. *Nature Climate Change*, 1(5): 266–271, 2011. doi: 10.1038/nclimate1169. URL <http://dx.doi.org/10.1038/nclimate1169>.
- R. Nair, H.-W. Choi, and H. Tufo. Computational aspects of a scalable high-order discontinuous galerkin atmospheric dynamical core. *Computers & Fluids*, 38(2):309 – 319, 2009. doi: <http://dx.doi.org/10.1016/j.compfluid.2008.04.006>.
- R. D. Nair and P. H. Lauritzen. A class of deformational flow test cases for linear transport problems on the sphere. *J. Comput. Phys.*, 229:8868–8887, 2010. doi: 10.1016/j.jcp.2010.08.014.
- R. D. Nair, S. J. Thomas, and R. D. Loft. A discontinuous galerkin transport scheme on the cubed sphere. *Mon. Wea. Rev.*, 133(4):814–828, 2005. doi: 10.1175/MWR2890.1.
- R. D. Nair, M. N. Levy, and P. H. Lauritzen. Emerging numerical methods for atmospheric modeling, in: P.H. Lauritzen, R.D. Nair, C. Jablonowski, M. Taylor (Eds.), Numerical techniques for global atmospheric models. *Lecture Notes in Computational Science and Engineering*, Springer, 80, 2011.
- R. B. Neale and B. J. Hoskins. A standard test for agcms including their

- physical parametrizations: I: the proposal. *Atmos. Sci. Lett.*, 1(2):101–107, 2000. doi: 10.1006/asle.2000.0022.
- R. B. Neale, J. H. Richter, and M. Jochum. The impact of convection on ENSO: From a delayed oscillator to a series of events. *J. Climate*, 21:5904–5924, 2008.
- R. B. Neale, C.-C. Chen, A. Gettelman, P. H. Lauritzen, S. Park, D. L. Williamson, A. J. Conley, R. Garcia, D. Kinnison, J.-F. Lamarque, D. Marsh, M. Mills, A. K. Smith, S. Tilmes, F. Vitt, P. Cameron-Smith, W. D. Collins, M. J. Iacono, R. C. Easter, S. J. Ghan, X. Liu, P. J. Rasch, and M. A. Taylor. Description of the NCAR Community Atmosphere Model (CAM 4.0). NCAR Technical Note, National Center of Atmospheric Research, 2010.
- R. B. Neale, C.-C. Chen, A. Gettelman, P. H. Lauritzen, S. Park, D. L. Williamson, A. J. Conley, R. Garcia, D. Kinnison, J.-F. Lamarque, D. Marsh, M. Mills, A. K. Smith, S. Tilmes, F. Vitt, P. Cameron-Smith, W. D. Collins, M. J. Iacono, R. C. Easter, S. J. Ghan, X. Liu, P. J. Rasch, and M. A. Taylor. Description of the NCAR Community Atmosphere Model (CAM 5.0). NCAR Technical Note NCAR/TN-486+STR, National Center of Atmospheric Research, 2012.
- T. A. O’Brien, F. Li, W. D. Collins, S. A. Rauscher, T. D. Ringler, M. Taylor, S. M. Hagos, and L. R. Leung. Observed scaling in clouds and precipitation and scale incognizance in regional to global atmospheric models. *Journal of Climate*, 26(23):9313–9333, 2013. doi: 10.1175/jcli-d-13-00005.1.
- T. A. O’Brien, W. D. Collins, K. Kashinath, O. Rübel, S. Byna, J. Gu, H. Kr-

- ishnan, and P. A. Ullrich. Resolution dependence of precipitation statistical fidelity in hindcast simulations. *J. Adv. Model. Earth Syst.*, 8(2):976–990, 2016. doi: 10.1002/2016ms000671. URL <http://dx.doi.org/10.1002/2016ms000671>.
- I. Orlanski. The quasi-hydrostatic approximation. *J. Atmos. Sci.*, 38:572–582, 1981. doi: 10.1175/1520-0469(1981)038<0572:TQHA>2.0.CO;2. URL [http://dx.doi.org/10.1175/1520-0469\(1981\)038<0572:TQHA>2.0.CO;2](http://dx.doi.org/10.1175/1520-0469(1981)038<0572:TQHA>2.0.CO;2).
- S. Park. A unified convection scheme (unicon). part i: Formulation. *Journal of the Atmospheric Sciences*, 71(11):3902–3930, 2014. doi: 10.1175/jas-d-13-0233.1.
- S. Park and C. S. Bretherton. The university of washington shallow convection and moist turbulence schemes and their impact on climate simulations with the community atmosphere model. *J. Climate*, 22:3449–3469, 2009.
- S. Park, C. S. Bretherton, and P. J. Rasch. Integrating cloud processes in the community atmosphere model, version 5. *Journal of Climate*, 27(18):6821–6856, 2014. doi: 10.1175/jcli-d-14-00087.1.
- O. Pauluis and S. Garner. Sensitivity of radiative–convective equilibrium simulations to horizontal resolution. *J. Atmos. Sci.*, 63(7):1910–1923, 2006.
- J. P. Peixoto and M. Oort. *Physics of Climate*. American Institute of Physics, 1992.
- R. T. Pierrehumbert. Subtropical water vapor as a mediator of rapid global climate change. *Mechanisms of Global Climate Change at Millennial Time Scales Geophysical Monograph Series*, page 339–361, 1999. doi: 10.1029/gm112p0339.

- R. S. Plant and G. C. Craig. A stochastic parameterization for deep convection based on equilibrium statistics. *J. Atmos. Sci.*, 65:87–105, 2008. doi: 10.1175/2007JAS2263.1. URL <http://dx.doi.org/10.1175/2007JAS2263.1>.
- L. Polvani, A. Clement, B. Medeiros, J. Benedict, and I. Simpson. Opening the door to simpler climate models in the community earth system model project. *EOS*, 2017. submitted.
- W. M. Putman and S.-J. Lin. Finite-volume transport on various cubed-sphere grids. *J. Comput. Phys.*, 227(1):55–78, 2007.
- D. Randall, M. Khairoutdinov, A. Arakawa, and W. Grabowski. Breaking the cloud parameterization deadlock. *Bulletin of the American Meteorological Society*, 84(11):1547–1564, 2003.
- P. J. Rasch and J. E. Kristjánsson. A comparison of the ccm3 model climate using diagnosed and predicted condensate parameterizations. *Journal of Climate*, 11(7):1587–1614, 1998. doi: 10.1175/1520-0442(1998)011<1587:acotcm>2.0.co;2.
- S. A. Rauscher, T. D. Ringler, W. C. Skamarock, and A. A. Mirin. Exploring a global multiresolution modeling approach using aquaplanet simulations. *Journal of Climate*, 26(8):2432–2452, 2013. doi: 10.1175/jcli-d-12-00154.1.
- S. A. Rauscher, T. A. O’Brien, C. Piani, E. Coppola, F. Giorgi, W. D. Collins, and P. M. Lawston. A multimodel intercomparison of resolution effects on precipitation: simulations and theory. *Climate Dynamics*, 47(7–8):2205–2218, 2016. doi: 10.1007/s00382-015-2959-5.
- D. J. Raymond and A. M. Blyth. Extension of the stochastic mixing model to cumulonimbus clouds. *Journal of the Atmospheric Sciences*, 49(21):

- 1968–1983, 1992. doi: 10.1175/1520-0469(1992)049<1968:eotsmm>2.0.co;2.
- K. A. Reed and C. Jablonowski. Assessing the uncertainty in tropical cyclone simulations in ncars community atmosphere model. *Journal of Advances in Modeling Earth Systems*, 3(8), 2011a. doi: 10.1029/2011ms000076.
- K. A. Reed and C. Jablonowski. An analytic vortex initialization technique for idealized tropical cyclone studies in agcms. *Monthly Weather Review*, 139(2):689–710, 2011b. doi: 10.1175/2010mwr3488.1.
- K. A. Reed and C. Jablonowski. Idealized tropical cyclone simulations of intermediate complexity: A test case for agcms. *Journal of Advances in Modeling Earth Systems*, 4(2):n/a–n/a, 2012. ISSN 1942-2466. doi: 10.1029/2011MS000099. URL <http://dx.doi.org/10.1029/2011MS000099>.
- K. A. Reed and B. Medeiros. A reduced complexity framework to bridge the gap between agcms and cloud-resolving models. *Geophysical Research Letters*, 43(2):860–866, 2016. doi: 10.1002/2015gl066713.
- K. A. Reed, C. Jablonowski, and M. A. Taylor. Tropical cyclones in the spectral element configuration of the community atmosphere model. *Atmospheric Science Letters*, 13(4):303–310, 2012. doi: 10.1002/asl.399.
- K. A. Reed, J. T. Bacmeister, N. A. Rosenbloom, M. F. Wehner, S. C. Bates, P. H. Lauritzen, J. E. Truesdale, and C. Hannay. Impact of the dynamical core on the direct simulation of tropical cyclones in a high-resolution global model. *Geophys. Res. Lett.*, 42(9):3603–3608, 2015a. doi: 10.1002/2015GL063974.
- K. A. Reed, B. Medeiros, J. T. Bacmeister, and P. H. Lauritzen. Global radiative–convective equilibrium in the community atmosphere model, ver-

- sion 5. *Journal of the Atmospheric Sciences*, 72(5):2183–2197, 2015b. doi: 10.1175/jas-d-14-0268.1.
- A. M. Rhoades, X. Huang, P. A. Ullrich, and C. M. Zarzycki. Characterizing sierra nevada snowpack using variable resolution cesm. *J. Appl. Meteor. Climatol.*, 55, 2016a. doi: 10.1175/JAMC-D-15-0156.1. URL <http://dx.doi.org/10.1175/JAMC-D-15-0156.1>.
- A. M. Rhoades, X. Huang, P. A. Ullrich, and C. M. Zarzycki. Characterizing sierra nevada snowpack using variable-resolution cesm. *Journal of Applied Meteorology and Climatology*, 55(1):173–196, 2016b. doi: 10.1175/JAMC-D-15-0156.1. URL <https://doi.org/10.1175/JAMC-D-15-0156.1>.
- J. H. Richter and P. J. Rasch. Effects of convective momentum transport on the atmospheric circulation in the community atmosphere model, version 3. *J. Climate*, 21(7):1487–1499, 2008.
- T. D. Ringler, L. Ju, and M. Gunzburger. A Multiresolution Method for Climate System Modeling: Application of Spherical Centroidal Voronoi Tesselations. *Ocean Dynamics*, 2008.
- T. Schneider, K. L. Smith, P. A. O’Gorman, and C. C. Walker. A climatology of tropospheric zonal-mean water vapor fields and fluxes in isentropic coordinates. *Journal of Climate*, 19(22):5918–5933, 2006. doi: 10.1175/jcli3931.1.
- G. Shutts. A kinetic energy backscatter algorithm for use in ensemble prediction systems. *Quart. J. Roy. Meteorol. Soc.*, 131:3079–3102, 2005.
- A. J. Simmons and D. M. Burridge. An energy and angular-momentum conserving vertical finite-difference scheme and hybrid vertical coordinates.

- Mon. Wea. Rev.*, 109(4):758–766, 1981.
- W. Skamarock. Kinetic energy spectra and model filters, in: P.H. Lauritzen, R.D. Nair, C. Jablonowski, M. Taylor (Eds.), Numerical techniques for global atmospheric models. *Lecture Notes in Computational Science and Engineering*, Springer, 80, 2011.
- W. C. Skamarock, J. B. Klemp, M. G. Duda, L. Fowler, S.-H. Park, and T. D. Ringler. A multi-scale nonhydrostatic atmospheric model using centroidal Voronoi tesselations and C-grid staggering. *Mon. Wea. Rev.*, 240:3090–3105, 2012. doi: doi:10.1175/MWR-D-11-00215.1.
- R. J. Small, J. Bacmeister, D. Bailey, A. Baker, S. Bishop, F. Bryan, J. Caron, J. Dennis, P. Gent, H.-m. Hsu, M. Jochum, D. Lawrence, E. Muñoz, P. diNezio, T. Scheitlin, R. Tomas, J. Tribbia, Y.-h. Tseng, and M. Vertenstein. A new synoptic scale resolving global climate simulation using the community earth system model. *J. Adv. Model. Earth Syst.*, 6(4):1065–1094, 2014. doi: 10.1002/2014MS000363.
- F. Song and G. J. Zhang. Understanding and improving the scale dependence of trigger functions for convective parameterization using cloud-resolving model data. *Journal of Climate*, 31(18):7385–7399, 2018.
- M. J. Suarez, A. Arakawa, and D. A. Randall. The parameterization of the planetary boundary layer in the ucla general circulation model: Formulation and results. *Mon. Wea. Rev.*, 111(11):2224–2243, 1983.
- M. Taylor, J. Edwards, and A. St-Cyr. Petascale atmospheric models for the community climate system model: new developments and evaluation of scalable dynamical cores. *J. Phys.: Conf. Ser.*, 125, 2008. doi: 10.1088/

1742-6596/125/1/012023.

- M. A. Taylor and A. Fournier. A compatible and conservative spectral element method on unstructured grids. *J. Comput. Phys.*, 229(17):5879 – 5895, 2010. doi: 10.1016/j.jcp.2010.04.008.
- D. R. Thatcher and C. Jablonowski. A moist aquaplanet variant of the Held– Suarez test for atmospheric model dynamical cores. *Geoscientific Model Development*, 9(4):1263–1292, 2016. doi: 10.5194/gmd-9-1263-2016. URL <https://www.geosci-model-dev.net/9/1263/2016/>.
- A. M. Tompkins and A. G. Semie. Organization of tropical convection in low vertical wind shears: Role of updraft entrainment. *Journal of Advances in Modeling Earth Systems*, 9(2):1046–1068, 2017. doi: 10.1002/2016ms000802.
- P. A. Ullrich. A global finite-element shallow-water model supporting continuous and discontinuous elements. *Geosci. Model Dev.*, 7(6):3017–3035, 2014. doi: 10.5194/gmd-7-3017-2014.
- P. A. Ullrich and M. A. Taylor. Arbitrary-order conservative and consistent remapping and a theory of linear maps: Part I. *Mon. Wea. Rev.*, 143: 2419–2440, 2015. doi: 10.1175/MWR-D-14-00343.1.
- P. A. Ullrich, T. Melvin, C. Jablonowski, and A. Staniforth. A proposed baroclinic wave test case for deep and shallow-atmosphere dynamical cores. *Quart. J. Royal Meteor. Soc.*, 140(682):1590–1602, 2014.
- P. A. Ullrich, D. Devendran, and H. Johansen. Arbitrary-order conservative and consistent remapping and a theory of linear maps: Part ii. *Mon. Wea. Rev.*, 144(4):1529–1549, 2016. doi: 10.1175/MWR-D-15-0301.1.
- P. A. Ullrich, C. Jablonowski, J. Kent, P. H. Lauritzen, R. Nair, K. A. Reed,

- C. M. Zarzycki, D. M. Hall, D. Dazlich, R. Heikes, C. Konor, D. Randall, T. Dubos, Y. Meurdesoif, X. Chen, L. Harris, C. Kühnlein, V. Lee, A. Qaddouri, C. Girard, M. Giorgetta, D. Reinert, J. Klemp, S.-H. Park, W. Skamarock, H. Miura, T. Ohno, R. Yoshida, R. Walko, A. Reinecke, and K. Viner. "dcmip2016: A review of non-hydrostatic dynamical core design and intercomparison of participating models". *Geosci. Model Dev.*, 10:4477–4509, 2017. doi: 10.5194/gmd-10-4477-2017.
- L. van Kampenhout, A. M. Rhoades, A. R. Herrington, C. M. Zarzycki, J. T. M. Lenaerts, W. J. Sacks, and M. R. van den Broeke. Regional grid refinement in an earth system model: Impacts on the simulated greenland surface mass balance. *The Cryosphere Discuss.*, 2018. doi: 10.5194/tc-2018-257.
- R. L. Walko and R. Avissar. A direct method for constructing refined regions in unstructured conforming triangular–hexagonal computational grids: Application to olam. *Monthly Weather Review*, 139(12):3923–3937, 2011. doi: 10.1175/mwr-d-11-00021.1.
- J. M. Wallace and D. S. Gutzler. Teleconnections in the geopotential height field during the northern hemisphere winter. *Monthly Weather Review*, 109(4):784–812, 1981. doi: 10.1175/1520-0493(1981)109<0784:titghf>2.0.co;2.
- H. Wan, M. A. Giorgetta, G. Zängl, M. Restelli, D. Majewski, L. Bonaventura, K. Fröhlich, D. Reinert, P. Ripodas, L. Kornblueh, et al. "the icon-1.2 hydrostatic atmospheric dynamical core on triangular grids, part i: formulation and performance of the baseline version". *Geosci. Model Dev.*, 6: 735–763, 2013.

- H. Wan, P. J. Rasch, M. A. Taylor, and C. Jablonowski. Short-term time step convergence in a climate model. *Journal of advances in modeling earth systems*, 7(1):215–225, 2015. doi: 10.1002/2014MS000368.
- S. R. Weart. *The discovery of global warming*. Harvard University Press, 2008.
- N. P. Wedi. Increasing horizontal resolution in numerical weather prediction and climate simulations: illusion or panacea? *Philosophical Transactions of the Royal Society of London A: Mathematical, Physical and Engineering Sciences*, 372(2018), 2014. doi: 10.1098/rsta.2013.0289.
- M. Wehner, K. Reed, F. Li, J. Bacmeister, C. Chen, C. Paciorek, P. Gleckler, K. Sperber, W. Collins, A. Gettelman, and C. Jablonowski. The effect of horizontal resolution on simulation quality in the community atmospheric model, cam5. 1. *J. Adv. Model. Earth Syst.*, 6(4):980–997, 2014. doi: 10.1002/2013MS000276.
- M. Weisman and J. B. Klemp. The dependence of numerically simulated convective storms on vertical wind shear and buoyancy. *Mon. Wea. Rev.*, 110:504–520, 1982.
- M. L. Weisman, W. C. Skamarock, and J. B. Klemp. The resolution dependence of explicitly modeled convective systems. *Monthly Weather Review*, 125(4):527–548, 1997. doi: 10.1175/1520-0493(1997)125<0527:TRDOEM>2.0.CO;2.
- J. Whitehead, C. Jablonowski, R. B. Rood, and P. H. Lauritzen. A stability analysis of divergence damping on a latitude-longitude grid. *Mon. Wea. Rev.*, 139:2976–2993, 2011.
- L. J. Wicker and W. C. Skamarock. A time-splitting scheme for the

- elastic equations incorporating second-order runge–kutta time differencing. *Monthly Weather Review*, 126(7):1992–1999, 1998. doi: 10.1175/1520-0493(1998)126<1992:atssft>2.0.co;2.
- D. S. Wilks. *Statistical methods in the atmospheric sciences*, volume 100. Academic press, 2011.
- D. Williamson, J. Kiehl, V. Ramanathan, R. Dickinson, and J. Hack. Description of ncar community climate model (ccm1). ncar tech. note. Technical report, NCAR/TN-285+ STR, 1987.
- D. L. Williamson. Convergence of atmospheric simulations with increasing horizontal resolution and fixed forcing scales. *Tellus A*, 51:663–673, 1999. doi: 10.1034/j.1600-0870.1999.00009.x.
- D. L. Williamson. Time-split versus process-split coupling of parameterizations and dynamical core. *Mon. Wea. Rev.*, 130:2024–2041, 2002.
- D. L. Williamson. Convergence of aqua-planet simulations with increasing resolution in the community atmospheric model, version 3. *Tellus A*, 60(5):848–862, 2008. doi: 10.1111/j.1600-0870.2008.00339.x.
- D. L. Williamson. The effect of time steps and time-scales on parametrization suites. *Quart. J. Roy. Meteor. Soc.*, 139(671):548–560, 2013. doi: 10.1002/qj.1992.
- D. L. Williamson and J. G. Olson. Dependence of aqua-planet simulations on time step. *Q. J. R. Meteorol. Soc.*, 129(591):2049–2064, 2003.
- D. L. Williamson and P. J. Rasch. Water vapor transport in the ncar ccm2. *Tellus A*, 46(1):34–51, 1994.
- D. L. Williamson, J. T. Kiehl, and J. J. Hack. Climate sensitivity of the

- ncar community climate model (ccm2) to horizontal resolution. *Climate Dynamics*, 11(7):377–397, 1995. doi: 10.1007/s003820050082.
- D. L. Williamson, J. G. Olson, C. Hannay, T. Tonazzo, M. Taylor, and V. Yudin. Energy considerations in the community atmosphere model (cam). *J. Adv. Model. Earth Syst.*, 7(3):1178–1188, 2015. doi: 10.1002/2015MS000448. URL <http://dx.doi.org/10.1002/2015MS000448>.
- Q. Yang, L. R. Leung, S. A. Rauscher, T. D. Ringler, and M. A. Taylor. Atmospheric moisture budget and spatial resolution dependence of precipitation extremes in aquaplanet simulations. *Journal of Climate*, 27(10):3565–3581, 2014. doi: 10.1175/jcli-d-13-00468.1.
- C. M. Zarzycki, C. Jablonowski, and M. A. Taylor. Using variable-resolution meshes to model tropical cyclones in the community atmosphere model. *Mon. Wea. Rev.*, 142(3):1221–1239, 2014a. doi: 10.1175/MWR-D-13-00179.1.
- C. M. Zarzycki, M. N. Levy, C. Jablonowski, J. R. Overfelt, M. A. Taylor, and P. A. Ullrich. Aquaplanet experiments using cam’s variable-resolution dynamical core. *J. Climate*, 27(14):5481–5503, 2014b. doi: 10.1175/JCLI-D-14-00004.1.
- G. Zhang and N. McFarlane. Sensitivity of climate simulations to the parameterization of cumulus convection in the canadian climate centre general circulation model. *Atmosphere-ocean*, 33(3):407–446, 1995.
- G. J. Zhang. Convective quasi-equilibrium in midlatitude continental environment and its effect on convective parameterization. *Journal of Geophysical Research: Atmospheres*, 107(D14):ACL–12, 2002.

- K. Zhang, P. J. Rasch, M. A. Taylor, H. Wan, L.-Y. R. Leung, P.-L. Ma, J.-C. Golaz, J. Wolfe, W. Lin, B. Singh, S. Burrows, J.-H. Yoon, H. Wang, Y. Qian, Q. Tang, P. Caldwell, and S. Xie. Impact of numerical choices on water conservation in the e3sm atmosphere model version 1 (eam v1). *Geoscientific Model Development Discussions*, 2017:1–26, 2017. doi: 10.5194/gmd-2017-293.
- M. Zhang, W. Lin, C. Bretherton, J. Hack, and P. Rasch. A modified formulation of fractional stratiform condensation rate in the ncar community atmospheric model (cam2). *Journal of Geophysical Research*, 108(D1), 2003. doi: 10.1029/2002jd002523.
- I. M. H. Zhao, M. and S.-J. Lin. Some counterintuitive dependencies of tropical cyclone frequency on parameters in a gcm. *J. Atmos. Sci.*, 69:2272–2283, 2012. doi: 10.1175/JAS-D-11-0238.1.
- P. Zurita-Gotor, I. M. Held, and M. F. Jansen. Kinetic energy-conserving hyperdiffusion can improve low resolution atmospheric models. *Journal of Advances in Modeling Earth Systems*, 7(3):1117–1135, 2015. doi: 10.1002/2015ms000480.
- G. Zängl, D. Reinert, P. Rípodas, and M. Baldauf. The icon (icosahedral non-hydrostatic) modelling framework of dwd and mpi-m: Description of the non-hydrostatic dynamical core. *Quarterly Journal of the Royal Meteorological Society*, 141(687):563–579, 2014. doi: 10.1002/qj.2378.