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Key Points:

- The CESM2.0 release of the spectral element dynamical core (CAM-SE) is documented
- Model has comprehensive treatment of condensates and energy
- The CAM-SE model has been sped up significantly compared to its predecessor CAM-HOMME

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NCAR Release of CAM-SE in CESM2.0: A Reformulation of the Spectral Element Dynamical Core in Dry-Mass Vertical Coordinates With Comprehensive Treatment of Condensates and Energy

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Abstract It is the purpose of this paper to provide a comprehensive documentation of the new NCAR (National Center for Atmospheric Research) version of the spectral element (SE) dynamical core as part of the Community Earth System Model (CESM2.0) release. This version differs from previous releases of the SE dynamical core in several ways. Most notably the hybrid sigma vertical coordinate is based on dry air mass, the condensates are dynamically active in the thermodynamic and momentum equations (also referred to as condensate loading), and the continuous equations of motion conserve a more comprehensive total energy that includes condensates. Not related to the vertical coordinate change, the hyperviscosity operators and the vertical remapping algorithms have been modified. The code base has been significantly reduced, sped up, and cleaned up as part of integrating SE as a dynamical core in the CAM (Community Atmosphere Model) repository rather than importing the SE dynamical core from High-Order Methods Modeling environment as an external code.

1. Introduction

Over the last two decades or so, the spectral element (SE) method has been considered as a numerical method for the fluid flow solver in global weather/climate models (Baer et al., 2006; Choi & Hong, 2016; Fournier et al., 2004; Giraldo et al., 2013; Kelly & Giraldo, 2012). The main motivations were the SE methods' near-perfect scalability (Dennis et al., 2012), GPU (Graphics Processing Unit) acceleration (e.g., Abdi, Giraldo, et al., 2017; Abdi, Wilcox, et al., 2017), high-order accuracy for smooth problems, and mesh refinement capabilities. For some time the Community Earth System Model (CESM; Hurrell et al., 2013) has supported a SE dynamical core option in the atmosphere component CAM (Community Atmosphere Model; Neale et al., 2012) discretized on a cubed-sphere grid (Figure 1a). The SE dynamical core in CAM supports uniform resolution grids based on the equiangular gnomonic cubed-sphere grid (Rančić et al., 1996) as well as a mesh refinement capability with local increases in resolution through conformal mesh refinement as shown in Figure 1b (Baer et al., 2006; Fournier et al., 2004; Zarzycki, Jablonowski, et al., 2014; Zarzycki, Levy, et al., 2014). The dynamical core code resided in the High-Order Methods Modeling environment (HOMME), a framework for developing new generation computationally efficient and petascale capable dynamical cores based on the SE method (Taylor et al., 2008; Thomas & Loft, 2000), the discontinuous Galerkin method (Nair et al., 2009), and finite-volume method (Erath et al., 2012; Lauritzen et al., 2017). Previously, the SE dynamical core was imported into CAM as an external code base. Consequently, any updates to the HOMME code base would have to be imported into CAM and pass regression tests in both HOMME and CAM. We refer to this setup as CAM-HOMME, which has been extensively tested in AMIP-style climate simulations (e.g., Bacmeister et al., 2016; Evans et al., 2012; Gettelman et al., 2018; Reed et al., 2015; Rhoades et al., 2016; Zarzycki & Jablonowski, 2014). CAM-SE's code base is in CAM and not HOMME. CAM-SE and HOMME codes have diverged significantly due to a massive code cleanup to satisfy

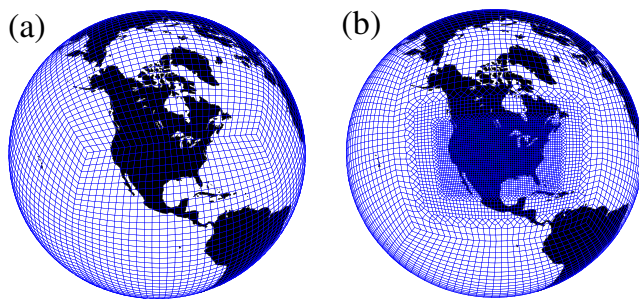


Figure 1. (a) The gnomonic equiangular cubed-sphere grid that defines the elements in CAM-SE. (b) Elements of a conformal mesh refinement grid referred to as the CONUS-grid (Contiguous United States) used in CAM-SE. CAM = Community Atmosphere Model; SE = spectral element; HOMME = High-Order Methods Modeling environment.

software engineering standards in CAM (and make the code more user-friendly for the community) as well as considerable reformulations of the dynamical core described herein.

In this paper we present a version of the SE dynamical core using a dry-mass vertical coordinate that includes condensate loading, and the continuous equations of motion conserve a comprehensive moist total energy containing all prognostic water variables and their respective heat capacities in the thermodynamic equation. The basic SE method has not been changed, but we present changes to the details of how hyperviscosity is applied and the vertical remapping that are not specific to the dry-mass vertical coordinate. As this paper serves as documentation for the CESM2.0 version of SE, we also provide details on the SE method and viscosity operators that have not been comprehensively documented in previous publications.

The SE dynamical core is the default dynamical core for high-resolution CESM applications, in particular, the configuration in which the average distance between grid points is approximately 28 km (Bacmeister et al., 2016). At such resolutions the effects of condensates (such as cloud liquid and rain) may have a significant effect on the dynamics (Bacmeister et al., 2012). CAM-HOMME does not represent the effect of condensates in the thermodynamic and momentum equations (also referred to as condensate loading). Including the thermodynamic and mass effect of condensates in the dynamical core using a dry-mass hybrid sigma vertical coordinate is mathematically simpler due to the clear separation of dry air, water vapor, and condensates in the discretization. This is the initial motivation for using a dry-mass vertical coordinate. That said, certain parts of the implementation of a dry-mass coordinate in the dynamical core are slightly more complicated since moist pressure is a diagnostic in a dry-mass vertical coordinate formulation (mass effects of moisture and condensates need to be explicitly added) whereas in a moist coordinate pressure is prognostic.

A second motivation for switching to a dry-mass vertical coordinate is the consistent coupling with the physical parameterizations. CAM physics assumes that the (moist) pressure levels are constant during the parameterization updates. Consequently, the moist pressure levels stay constant even when moisture leaves the column (e.g., rains out). At the very end of CAM physics the change in water vapor in each column is taken into account by scaling the mixing ratios for all tracers that are based on specific/moist mixing ratios so that dry air mass and tracer mass are conserved (see section 3.1.6 in Neale et al., 2012). This scaling does not guarantee shape preservation but changes the total energy. In addition, the pressure field in CAM physics does not take into account the mass of condensates. When using a dry-mass vertical coordinate, the coordinate surfaces (assuming dry mass is constant) remain constant throughout the physics updates and there is no need to adjust tracer mixing ratios and one can more easily take into account the work performed by water variables in the context of the energy cycle.

The third motivation for using a dry-mass vertical coordinate relates to total energy conservation. Currently, the energy fixer in CAM is based on a dry total energy (Williamson et al., 2015) that uses the same heat capacity for dry air and water vapor and does not include the effect of condensates. To move toward a more accurate treatment of energy in CAM, a first step is to develop a dynamical core based on equations of motion conserving an energy that more accurately represents water vapor as well as condensates. This is most easily done when using a dry-mass vertical coordinate so that the energies associated with all water variables are clearly separated. Similarly for axial angular momentum (AAM) which is an important conserved quantity of the continuous equations of motion (e.g., Lebonnois et al., 2010). The changes to physics needed to make it consistent with the dynamical core (discussed in this paragraph and the previous paragraph) is necessary for a fully consistent model. This adaptation is, however, not treated in this paper.

The fourth motivation for a dry-mass formulation is the consistent coupling between CAM-SE and CSLAM (Conservative Semi-Lagrangian Multi-tracer scheme; Lauritzen et al., 2010) in a moist atmosphere. The consistent coupling between the SE continuity equation for air and the continuity equation for tracers (solved with CSLAM) is much more consistent in a dry-mass coordinate as explained in the following. In a dry-mass formulation CSLAM transports water tracers while SE solves the continuity equation for dry air.

The two are consistently coupled by making sure that if a tracer has a value of 1 (in which case CSLAM predicts the evolution of dry air mass) then the CSLAM dry mass field is identical to the SE dry mass field integrated over CSLAM control volumes (how this is done is explained in detail in Lauritzen et al., 2017). Had one used a moist vertical coordinate, then this coupling between SE and CSLAM would be complicated by the fact that water tracers would implicitly be predicted by SE through solving the continuity equation for moist air mass (using a moist vertical coordinate). CAM-SE-CSLAM in a moist atmosphere is the subject of a separate paper.

This paper presents the reformulation of the SE dynamical core alone, while a reformulation of the physical parameterizations, as necessary for a fully consistent model in dry-mass vertical coordinates, is a separate endeavor. The paper is organized as follows. In section 2 the continuous equations of motion are derived which involves a detailed discussion of moist thermodynamics in the presence of condensates. The AAM and total energy conservation properties of continuous system of equations are also discussed. In section 3 the discretized equations of motion with focus on the vertical discretization in dry-mass vertical coordinates are derived. Details on the horizontal SE discretization on the cubed-sphere are also presented. Section 4 presents some results, validating the new dynamical core in idealized configurations, first, a moist baroclinic wave with simple warm-rain microphysics and second in a CAM6 aqua-planet configuration. The computational performance of CAM-SE is presented in section 4. The paper ends with summary and conclusions in section 5.

2. Continuous Equations

Before writing the continuous equations of motion using a dry-mass vertical coordinate (section 2.5), we first need to discuss the representation of water variables (section 2.1), discuss the ideal gas law (section 2.2), and derive the thermodynamic equation for a mixture containing all water species (section 2.3). The discussion of the equations of motion in the presence of water vapor, cloud liquid, ice, rain, and snow closely follows Staniforth et al. (2006). While the appropriate thermodynamics of moist air is well understood (e.g., Emanuel, 1994), it is unclear how to best represent moist dynamical effects in numerical models (Bannon, 2003; Ooyama, 1990, 2001; Satoh, 2003). Therefore, the derivation of the equations of motion including non-gas components is discussed in detail.

Thereafter, the dry-mass vertical coordinate is defined in section 2.4. Details on viscosity and frictional heating are presented in section 2.6 and global conservation properties derived in section 2.7.

2.1. Representation of Water Phases in Terms of Dry and Wet (Specific) Mixing Ratios

Equation (1) defines the dry mixing ratios for the water variables (vapor *wv*, cloud liquid *cl*, cloud ice *ci*, rain *rn*, and snow *sw*)

$$m^{(\ell)} \equiv \frac{\rho^{(\ell)}}{\rho^{(d)}}, \text{ where } \ell = d, wv, cl, ci, rn, sw, \quad (1)$$

where $\rho^{(d)}$ is the mass of dry air per unit volume of moist air and $\rho^{(\ell)}$ is the mass of the water substance of type ℓ per unit volume of moist air. Note that the mixing ratio for dry air is unity: $m^{(d)} \equiv \frac{\rho^{(d)}}{\rho^{(d)}} = 1$. Moist air refers to air containing dry air, water vapor, cloud liquid, cloud ice, rain amount, and snow amount. For notational purposes define the set of all components of air

$$\mathcal{L}_{\text{all}} = \{d, wv, cl, ci, rn, sw\}, \quad (2)$$

a set only referring to all water variables,

$$\mathcal{L}_{\text{water}} = \{wv, cl, ci, rn, sw\}, \quad (3)$$

and a set referring to all condensates (non-gas components of water)

$$\mathcal{L}_{\text{cond}} = \{cl, ci, rn, sw\}. \quad (4)$$

The density of a unit volume of moist air is related to the dry air density through

$$\rho = \rho^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right). \quad (5)$$

Mixing ratios can also be specified in terms of density per density of moist air; in other words, specific/moist mixing ratios

$$q^{(\ell)} \equiv \frac{\rho^{(\ell)}}{\rho}, \quad (6)$$

where, in particular, $q^{(wv)}$ is the specific humidity.

It is straightforward to convert between moist and dry mixing ratios

$$m^{(\ell)} = \frac{q^{(\ell)}}{1 - \sum_{\ell \in \mathcal{L}_{\text{water}}} q^{(\ell)}}, \quad (7)$$

$$q^{(\ell)} = \frac{m^{(\ell)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}}. \quad (8)$$

Note that if water vapor undergoes a phase change to rain and leaves the column, then the specific/wet mixing ratios change but the dry mixing ratios do not.

2.2. Ideal Gas Law and Virtual Temperature

In this section, the ideal gas law is derived for moist air containing condensates. In a parcel of moist air of volume V , the gaseous components of moist air (dry air and water vapor) occupy volume $V^{(\text{gas})}$ and the condensed phases occupy volume $V^{(\text{cond})} = V - V^{(\text{gas})}$. The ideal gas law applies to the gaseous component of air only. In that case, the partial pressure of dry air is (the pressure dry air would exert if it alone would occupy the volume)

$$p^{(d)} V^{(\text{gas})} = N^{(d)} k_B T, \quad (9)$$

where k_B is the Boltzmann constant and $N^{(d)}$ is the number of molecules of dry air. Now

$$N^{(d)} = \frac{V \rho^{(d)}}{\mathcal{M}^{(d)}}, \quad (10)$$

where $\mathcal{M}^{(d)}$ is the molar mass of dry air, so the ideal gas law for the partial pressure of dry air can be written as

$$p^{(d)} V^{(\text{gas})} = V \rho^{(d)} R^{(d)} T, \quad (11)$$

where $R^{(d)} \equiv \frac{k_B}{\mathcal{M}^{(d)}}$ is the dry air gas constant. Similarly for the partial pressure of water vapor,

$$p^{(wv)} V^{(\text{gas})} = V \rho^{(wv)} R^{(wv)} T, \quad (12)$$

where $R^{(wv)} \equiv \frac{k_B}{\mathcal{M}^{(wv)}}$ is the gas constant for water vapor. According to Dalton's law of partial pressures, the total pressure exerted is equal to the sum of the partial pressures of the individual gases:

$$p = \frac{V}{V^{(\text{gas})}} (\rho^{(d)} R^{(d)} T + \rho^{(wv)} R^{(wv)} T). \quad (13)$$

Note that the condensates do not exert a gas pressure (see section 2.4.2). Moving T , $R^{(d)}$, and $\rho^{(d)}$ outside the parenthesis on the right-hand side of (13) yields

$$p = \frac{V}{V^{(\text{gas})}} \rho^{(d)} R^{(d)} \left(1 + \frac{1}{\epsilon} m^{(wv)} \right) T, \quad (14)$$

where $\epsilon \equiv \frac{R^{(d)}}{R^{(wv)}}$. Multiplying (14) with ρ/ρ and simplifying using (1) and (5) yields

$$p = \frac{V}{V^{(\text{gas})}} \rho R^{(d)} \left(\frac{1 + \frac{1}{\epsilon} m^{(wv)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}} \right) T. \quad (15)$$

By defining the virtual temperature

$$T_v = T \left(\frac{1 + \frac{1}{\epsilon} m^{(wv)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}} \right) \quad (16)$$

and assuming that the volume of the condensates is 0, $V = V^{(\text{gas})}$, then (15) can be written as

$$p = \rho R^{(d)} T_v. \quad (17)$$

2.3. Thermodynamic Equation

Let S denote the total entropy of moist air

$$S = \sum_{\ell \in \mathcal{L}_{\text{all}}} S^{(\ell)}, \quad (18)$$

where $S^{(\ell)}$ is the entropy of component ℓ of moist air. From the chain rule applied to each component separately we get

$$dS = \sum_{\ell \in \mathcal{L}_{\text{all}}} dS^{(\ell)}, \quad (19)$$

$$= \sum_{\ell \in \mathcal{L}_{\text{all}}} \left(\frac{\partial S^{(\ell)}}{\partial E^{(\ell)}} \right) dE^{(\ell)} + \sum_{\ell \in \mathcal{L}_{\text{gas}}} \left(\frac{\partial S^{(\ell)}}{\partial V^{(\text{gas})}} \right) dV^{(\text{gas})} + \sum_{\ell \in \mathcal{L}_{\text{cond}}} \left(\frac{\partial S^{(\ell)}}{\partial V^{(\ell)}} \right) dV^{(\ell)}, \quad (20)$$

where $V^{(\ell)}$ is the volume occupied by component ℓ of moist air and $E^{(\ell)}$ is the internal energy of ℓ . If we assume that the condensates are incompressible, $dV^{(\ell)} = 0$ for $\ell \in \mathcal{L}_{\text{cond}}$, then (20) becomes

$$dS = \sum_{\ell \in \mathcal{L}_{\text{all}}} \left(\frac{\partial S^{(\ell)}}{\partial E^{(\ell)}} \right) dE^{(\ell)} + \sum_{\ell \in \mathcal{L}_{\text{gas}}} \left(\frac{\partial S^{(\ell)}}{\partial V^{(\text{gas})}} \right) dV^{(\text{gas})}. \quad (21)$$

Using that $\frac{\partial S^{(\ell)}}{\partial V^{(\text{gas})}} = \frac{p^{(\ell)}}{T}$ for $\ell \in \mathcal{L}_{\text{gas}}$, $\frac{\partial S^{(\ell)}}{\partial E^{(\ell)}} = \frac{1}{T}$, and that

$$dE^{(\ell)} = C_V^{(\ell)} dT, \quad (22)$$

where $C_V^{(\ell)}$ is the heat capacity of ℓ at constant volume, then (21) can be written as

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} C_V^{(\ell)} dT = T \sum_{\ell \in \mathcal{L}_{\text{all}}} dS^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} p^{(\ell)} dV^{(\text{gas})}. \quad (23)$$

Note that (23) holds for each component ℓ . Multiply each component version of (23) with $\frac{N^{(\ell)} \mathcal{M}^{(\ell)}}{N^{(\ell)} \mathcal{M}^{(\ell)}}$, and using that $\rho^{(\ell)} = \frac{N^{(\ell)} \mathcal{M}^{(\ell)}}{V}$, then equation (23) can be written as

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} \rho^{(\ell)} C_V^{(\ell)} dT = T \sum_{\ell \in \mathcal{L}_{\text{all}}} \rho^{(\ell)} dS^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} \frac{\rho^{(\ell)} p^{(\ell)}}{N^{(\ell)} \mathcal{M}^{(\ell)}} dV^{(\text{gas})}, \quad (24)$$

where $c_V^{(\ell)} = \frac{C_V^{(\ell)}}{N^{(\ell)} \mathcal{M}^{(\ell)}}$ is the specific heat capacity for ℓ and $s^{(\ell)} = \frac{S^{(\ell)}}{N^{(\ell)} \mathcal{M}^{(\ell)}}$ is specific entropy for ℓ . To rewrite the last sum on the right-hand side of (24), we note that for gaseous components of moist air, $\ell \in \mathcal{L}_{\text{gas}}$,

$$\frac{V^{(\text{gas})}}{N^{(\ell)} \mathcal{M}^{(\ell)}} = \frac{V^{(\text{gas})} + V^{(\text{cond})}}{N^{(\ell)} \mathcal{M}^{(\ell)}} - \frac{V^{(\text{cond})}}{N^{(\ell)} \mathcal{M}^{(\ell)}}, \quad (25)$$

$$= \alpha^{(\ell)} - \frac{V^{(\text{cond})}}{V} \frac{V}{N^{(\ell)} \mathcal{M}^{(\ell)}}, \quad (26)$$

$$= (1 - \chi^{(\text{cond})}) \alpha^{(\ell)}, \quad (27)$$

where $\alpha^{(\ell)} = \frac{V}{N^{(\ell)} \mathcal{M}^{(\ell)}}$ the specific volume of component ℓ and $\chi^{(\text{cond})} \equiv \frac{V^{(\text{cond})}}{V}$. Taking the differential of (27) and assuming no phase changes ($N^{(\ell)}$ constant so $dN^{(\ell)} = 0$), then (24) can be rewritten as

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} \rho^{(\ell)} c_V^{(\ell)} dT = T \sum_{\ell \in \mathcal{L}_{\text{all}}} \rho^{(\ell)} ds^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} \rho^{(\ell)} p^{(\ell)} d[(1 - \chi^{(\text{cond})}) \alpha^{(\ell)}]. \quad (28)$$

Using (1), equation (28) can be written as

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} c_V^{(\ell)} dT = T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} p^{(\ell)} d[(1 - \chi^{(\text{cond})}) \alpha^{(\ell)}]. \quad (29)$$

Henceforth, we assume that the volume of condensates is 0, $V^{(\text{cond})} = 0$, that is, $\chi^{(\text{cond})} = 0$. Consequently, we can rewrite (29) as

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} c_v^{(\ell)} dT = T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} p^{(\ell)} d\alpha^{(\ell)}, \quad (30)$$

$$= T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} d[p^{(\ell)} \alpha^{(\ell)}] + \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} \alpha^{(\ell)} dp^{(\ell)}, \quad (31)$$

$$= T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)} - \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} d[R^{(\ell)} T] + \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} \alpha^{(\ell)} dp^{(\ell)}, \quad (32)$$

again using the chain rule and the ideal gas law (13) with $\frac{V}{V^{(\text{gas})}} = 1$. Using that $c_p^{(\ell)} = R^{(\ell)} + c_v^{(\ell)}$ and rearranging terms in (32) yields

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} c_p^{(\ell)} dT - \sum_{\ell \in \mathcal{L}_{\text{gas}}} m^{(\ell)} \alpha^{(\ell)} dp^{(\ell)} = T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)}, \quad (33)$$

which can be written as

$$\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} c_p^{(\ell)} dT - \frac{1}{\rho_d} \sum_{\ell \in \mathcal{L}_{\text{gas}}} dp^{(\ell)} = T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)}, \quad (34)$$

since $m^{(\ell)} \alpha^{(\ell)} = \frac{m^{(\ell)}}{\rho^{(\ell)}} = \rho^{(d)}$. Dividing (34) with $\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}$ and using (5) results in

$$\frac{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} c_p^{(\ell)} dT}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}} - \frac{1}{\rho} dp = \frac{T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}}. \quad (35)$$

If we define

$$R = \frac{\sum_{\ell \in \mathcal{L}_{\text{all}}} R^{(\ell)} m^{(\ell)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}}, \quad (36)$$

$$c_v = \frac{\sum_{\ell \in \mathcal{L}_{\text{all}}} c_v^{(\ell)} m^{(\ell)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}}, \quad (37)$$

$$c_p = \frac{\sum_{\ell \in \mathcal{L}_{\text{all}}} c_p^{(\ell)} m^{(\ell)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}}, \quad (38)$$

where $R^{(\ell)} \equiv 0$ for the condensates; then using (13), the thermodynamic equation can be written in compact form

$$\delta T - \frac{RT}{c_p p} \delta p = \frac{dQ}{c_p}, \quad (39)$$

where $dQ = \frac{T \sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} ds^{(\ell)}}{\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)}}$ is the amount of heat per unit mass that is supplied reversibly to moist air. Note that (39) includes the effect of condensates but assumes that the condensates are incompressible and that their volume is zero.

2.4. Vertical Coordinate

2.4.1. Definition

Let $M_s^{(d)}$ be the mass of a column of dry air per unit area at the surface (i.e., the weight of dry air at the surface per unit area is $g M_s^{(d)}$, where g is the gravitational acceleration; assumed constant) and $M_t^{(d)}$ is the mass of air per unit area in the column above the model top. Note that weight differs from mass in that weight constitutes the force exerted by the matter when it is in a gravitational field, whereas mass is the amount of matter (which is invariant and does not depend on g). The SI unit for $M_s^{(d)}$ is kg/m^2 , and the weight $g M_s^{(d)}$ is $\frac{1}{s^2} \frac{\text{kg}}{\text{m}} \equiv \text{Pa}$ (pressure). We assume that the composition of dry air is constant and that there is no moisture above the model top so $p_t = g M_t^{(d)}$. Consider a general, terrain-following, vertical coordinate $\eta^{(d)}$ that is a function of the dry air mass $M^{(d)}$

$$\eta^{(d)} = h(M^{(d)}, M_s^{(d)}), \quad (40)$$

where $h(M_s^{(d)}, M_s^{(d)}) = 1$ and $h(M_t^{(d)}, M_s^{(d)}) = 0$. Note that by removing the superscript (d) from the equations above (so that the dry variables represent moist variables) and assume that there are no condensates, then the vertical coordinate is the usual hybrid pressure coordinate widely used in hydrostatic global modeling (Simmons & Burridge, 1981). The top and bottom boundary conditions are that $\dot{\eta}(M_s^{(d)}, M_s^{(d)}) = 0$ and $\dot{\eta}(M_t^{(d)}, M_s^{(d)}) = 0$. Note that using a dry-mass vertical coordinate simplifies the coupling to physics since the dry-mass coordinate remains constant even if water leaves the column.

2.4.2. Partial Pressure of Dry Air and Mass of Dry Air

The observant reader will have noticed that we denote the mass of dry air per unit surface $M^{(d)}$, which exerts pressure $gM^{(d)}$, and not the dry air partial pressure $p^{(d)}$. As explained below this is because $gM^{(d)} \neq p^{(d)}$ in the presence of condensates.

The hydrostatic (moist) pressure at given height z can be computed from the hydrostatic balance

$$p(z) = g \int_{z'=z}^{z'=\infty} \rho \, dz', \quad (41)$$

$$= g \int_{z'=z}^{z'=\infty} \rho^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right) dz', \quad (42)$$

$$= g \sum_{\ell \in \mathcal{L}_{\text{all}}} M^{(\ell)}(z), \quad (43)$$

where the right-hand side of (43) is the mass of dry air, water vapor, cloud liquid, cloud ice, rain, and snow per unit area above height z , respectively:

$$M^{(\ell)}(z) = \int_{z'=z}^{z'=\infty} \rho^{(d)} m^{(\ell)} \, dz'. \quad (44)$$

Using Dalton's law of partial pressures on the left-hand side of (43), one obtains the partial pressure at a certain height z is

$$p^{(d)}(z) + p^{(\text{wv})}(z) = g \sum_{\ell \in \mathcal{L}_{\text{all}}} M^{(\ell)}(z). \quad (45)$$

From (45) it is clear that in the presence of condensate one cannot equate $p^{(d)}(z)$ with $gM^{(d)}(z)$ nor $p^{(\text{wv})}(z)$ with $gM^{(\text{wv})}(z)$. The partial pressures of dry air and water vapor are both affected by the mass of the condensates even though the condensates do not exert a gas pressure. Hence, the partial pressure of dry air $p^{(d)}(z)$ at height z is different from force exerted by the mass of dry air in Earth's gravitational field $gM^{(d)}(z)$ and similar for moisture. A physical explanation is that when hydrometeors are falling at terminal velocity, the gravitational force pulling the hydrometeors downward is compensating by the upward frictional force of the gaseous atmosphere on the hydrometeors. This compensating force adds to the atmospheric pressure.

The hydrostatic balance for dry air mass, written in terms of differentials, is given by

$$dM^{(d)}(z) = -g\rho^{(d)} \, dz, \quad (46)$$

whereas in a moist atmosphere (with condensates), a dry air partial pressure hydrostatic equation does not hold

$$dp^{(d)}(z) \neq -g\rho^{(d)} \, dz. \quad (47)$$

The partial pressure of dry air $p^{(d)}(z)$ at height z will increase in the presence of condensates, whereas the mass of dry air does not. The differential of the (moist) pressure can be written in terms of the dry air mass (under the hydrostatic assumption) though

$$dp(z) = -g\rho \, dz = -g\rho^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right) dz = g \, dM^{(d)}(z) \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right). \quad (48)$$

In all, $gM^{(d)}(z)$ is equal to $p^{(d)}(z)$ at height z only if there are no condensates present at higher levels (i.e., air is a gas only) and similarly for water vapor. Henceforth, we drop the notation of the vertical dependence (in this section we used z) on $M^{(\ell)}(z) = M^{(\ell)}$. Since $\eta^{(d)}$ is the vertical coordinate, $M^{(\ell)}$ will henceforth refer to the mass

per unit area of water form ℓ above $\eta^{(d)}$. The weight of dry air per unit area $gM^{(d)}$ is not a directly measurable quantity but a theoretical construction for the dry-mass vertical coordinate. Only moist pressure (or simply pressure) is directly measurable.

2.5. Equations of Motion

The $\eta^{(d)}$ -coordinate adiabatic and frictionless atmospheric primitive equations assuming floating Lagrangian vertical coordinates (Lin, 2004; Starr, 1945) can be written in vector invariant form as

$$\frac{\partial \mathbf{v}}{\partial t} + (\zeta + f) \hat{k} \times \mathbf{v} + \nabla_{\eta^{(d)}} \left(\frac{1}{2} \mathbf{v}^2 + \Phi \right) + \frac{1}{\rho} \nabla_{\eta^{(d)}} p = 0, \quad (49)$$

$$\frac{\partial T}{\partial t} + \mathbf{v} \cdot \nabla_{\eta^{(d)}} T - \frac{1}{c_p \rho} \omega = 0, \quad (50)$$

$$\frac{\partial}{\partial t} \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} m^{(\ell)} \right) + \nabla_{\eta^{(d)}} \cdot \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} m^{(\ell)} \mathbf{v} \right) = 0, \quad \ell \in \mathcal{L}_{\text{all}}, \quad (51)$$

where Φ is the geopotential height ($\Phi = g z$), $\hat{k}$ is the unit vector normal to the surface of the sphere, $\mathbf{v} = (u, v)$ is the velocity vector with u being the zonal velocity component and v the meridional velocity component, $\zeta = \hat{k} \cdot \nabla \times \mathbf{v}$ is vorticity, f Coriolis parameter, and $\omega = dp/dT$ is the (moist) pressure vertical velocity with $d/dt = \frac{\partial}{\partial t} + \mathbf{v} \cdot \nabla_{\eta^{(d)}}$ being the material/total derivative along $\eta^{(d)}$.

The prognostic equations for $\mathbf{v}$, the temperature T , dry air mass $\frac{\partial M^{(d)}}{\partial \eta^{(d)}}$, and tracer mass $\frac{\partial M^{(d)}}{\partial \eta^{(d)}} m^{(\ell)}$ are solved with the diagnostic equation for geopotential height (hydrostatic balance)

$$\frac{\partial \Phi}{\partial \eta^{(d)}} = - \frac{R^{(d)} T_v}{p} \frac{\partial p}{\partial \eta^{(d)}}, \quad (52)$$

where

$$\frac{\partial p}{\partial \eta^{(d)}} = g \frac{\partial M^{(d)}}{\partial \eta^{(d)}} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right). \quad (53)$$

For diagnosing vertical pressure velocity ω we note that

$$\omega(\eta^{(d)}) = \frac{dp}{dT}(\eta^{(d)}), \quad (54)$$

$$= \int_{\eta^{(d)}}^{\eta^{(d)=0}} \frac{d}{dT} \left(\frac{\partial p}{\partial \eta^{(d)}} \right) d\eta^{(d)}, \quad (55)$$

$$= \int_{\eta^{(d)}}^{\eta^{(d)=0}} \frac{\partial}{\partial t} \left(\frac{\partial p}{\partial \eta^{(d)}} \right) d\eta^{(d)} + \int_{\eta^{(d)}}^{\eta^{(d)=0}} \mathbf{v} \cdot \nabla_{\eta^{(d)}} \left(\frac{\partial p}{\partial \eta^{(d)}} \right) d\eta^{(d)}. \quad (56)$$

2.6. Hyperviscosity and Frictional Heating

The SE method does not have implicit diffusion. Hyperviscosity operators are applied to the prognostic variables to dissipate energy near the grid scale (Dennis et al., 2012). Hyperviscosity also damps the propagation of spurious grid-scale modes (Ainsworth & Wajid, 2009) and, in particular, smoothes the solution at element boundaries where the basis functions are least smooth (C^0 continuous). For the uniform resolution configuration constant hyperviscosity coefficients are used on all elements, whereas the variable resolution configuration uses either a scaling of the coefficients according to individual element length scale (Zarzycki, Levy, et al., 2014) or a tensor-based hyperviscosity operator approach (Guba, Taylor, Ullrich, et al., 2014).

2.6.1. Viscosity Operator

On the right-hand side of the equations of motion hyperviscosity terms are added. For the momentum equation (49) the viscous terms are as follows. By using the vector identity $\nabla^2 \mathbf{v} = \nabla(\nabla \cdot \mathbf{v}) - \nabla \times (\nabla \times \mathbf{v})$, the viscosity term splits into two terms. The first term damps divergent modes, and the latter term damps rotational modes. Thereby one can damp divergent modes more (or less) than rotational modes by having different coefficients (ν_{div} and ν_{vor} , respectively) in front of the respective terms

$$\nu_{\text{div}} \nabla(\nabla \cdot \mathbf{v}) - \nu_{\text{vor}} \nabla \times (\nabla \times \mathbf{v}), \quad (57)$$

The fourth-order hyperviscosity operator is computed by iteratively applying the Laplacian operator (57) (for a detailed derivation with metric terms see, e.g., Ullrich, 2014).

The dry air mass layer thickness is damped with $\nu_p \nabla^4 \left\{ \frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right\}$, temperature with $\nu_T \nabla^4 T$, and tracers with $\nu_q \nabla^4 q$. The damping coefficients for divergence (ν_{div}), vorticity (ν_{vor}), level thickness (ν_p), temperature (ν_T), and tracers ν_q are resolution dependent and provided in Appendix A.3. These coefficients were determined empirically for stability and may have to be increased for specific applications such as data assimilation cycling where additional damping to remove short wavelengths due to imbalances may be necessary. Note that the viscosity coefficient for pressure, ν_p , and tracers, ν_q , should be the same; otherwise, the model is no longer *free-stream* preserving (i.e., a constant mixing ratio is preserved, also referred to as mass-wind consistency in the literature).

The dispersion properties of CAM-SE with hyperviscosity are similar to A-grid models. There are no computational modes, but the grid-scale modes are erratic with large phase errors (Ainsworth, 2014; Ainsworth & Wajid, 2009). Empirically, we have found that increasing ν_{div} and ν_p compared to ν_T and ν_{vor} is effective at damping grid-scale modes and noise resulting from the SE basis functions being least smooth, C^0 , at element edges, while not making the total kinetic energy spectrum too dissipative at the high wavenumbers.

The horizontal hyperviscosity operator can be applied on η_d -surfaces, $\nabla^4 = \nabla_{\eta_d}^4$, but it may be advantageous to apply the hyperviscosity operator on approximate dry-mass surfaces

$$\nu \nabla^4 \Xi = \nu \nabla_{\eta_d}^4 \Xi - \nu \frac{\partial \Xi}{\partial M^{(d)}} \nabla_{\eta_d}^4 M^{(d)}, \quad \Xi = \mathbf{v}, T \quad (58)$$

(p.58 in Neale et al., 2012) to reduce spurious diffusion over steep topography. This *correction* is currently not applied to tracers. In theory the damping of dry-mass layer thickness should be zero if hyperviscosity is applied on dry-mass surfaces. However, for stability it is necessary to damp dry-mass layer thickness, but instead of applying ∇^4 to $\frac{\partial M^{(d)}}{\partial \eta^{(d)}}$ it is applied to the difference between $\frac{\partial M^{(d)}}{\partial \eta^{(d)}}$ and a smoothed version of $\frac{\partial M^{(d)}}{\partial \eta^{(d)}}$ referred to as $\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right)^{(ref)}$. The reference/smoothed dry-mass layer thickness is defined in Appendix A.2.

In the top three layers second-order diffusion (Laplacian operator) is applied to the prognostic variables to provide a sponge layer. The sponge layer plays an important role in controlling the polar night jet in low-top models (see, e.g., Lauritzen et al., 2011).

2.6.2. Frictional Heating

Let $\delta \mathbf{v}$ be the change in the velocity vector due to diffusion of momentum. Then the change in kinetic energy due to hyperviscosity applied to $\mathbf{v}$ is $\frac{1}{2} \rho \mathbf{v} \cdot \delta \mathbf{v}$. This kinetic energy is converted to a heating rate by adding a heating term $\delta \mathcal{T}$ in the thermodynamic equation corresponding to the kinetic energy change

$$\rho c_p \delta \mathcal{T} = -\frac{1}{2} \rho \mathbf{v} \cdot \delta \mathbf{v} \Rightarrow \delta \mathcal{T} = -\frac{1}{2c_p} (\mathbf{v} \cdot \delta \mathbf{v}), \quad (59)$$

(p.71 in ; Neale et al., 2012). As shown in the results section 4.2 this term is rather large and therefore important for good energy conservation characteristics of the dynamical core.

Before discretizing the equations of motion some important conservation properties of the equations are discussed. For models intended for long simulations, it is beneficial to have good energy conservation properties to minimize nonlocal restoration of global energy via global energy fixers (e.g., Thuburn, 2008). As an aside, it is noted that the global energy fixers take up a significant fraction of runtime at ultrahigh resolution as the operations require global communications. Good AAM conservation may be important for the simulation of the Quasi-Biennial Oscillation although the accurate simulation of the Quasi-Biennial Oscillation also depends on vertical resolution, location of model top, model dissipation (numerical and physical), and parameterizations (such as nonorographic gravity wave drag and convection parameterization; Richter et al., 2014). For the simulation of superrotating atmospheres (e.g., Venus), however, the conservation of angular momentum is crucial (Lebonnois et al., 2012). Below we discuss the conservation properties of the moist equations of motion used in this paper.

2.7. Global Conservation: AAM and Energy

The conservation law for AAM

$$\mathcal{M} = (u + \Omega r \cos \varphi) r \cos \varphi, \quad (60)$$

(where r is the mean radius of Earth, Ω angular velocity, and φ latitude) integrated over the entire domain is derived in Appendix B, and the final equation is

$$\frac{\partial}{\partial t} \int_{\eta=0}^{\eta=1} \iint_S \left[g \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right) \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) \mathcal{M} \right] dA d\eta^{(d)} = - \iint_S \left[p_s \frac{\partial z_s}{\partial \lambda} \right] dA, \quad (61)$$

where S is the global domain and $dA = r^2 \cos \varphi d\lambda d\varphi$ is an infinitesimal surface area element on the sphere. The equation clearly separates the AAM of each component of moist air. The term in the square brackets on the right-hand side of (61) is referred to as the mountain torque. In the absence of topography, $z_s = 0$ m, (61) states that the angular momentum integrated over the entire domain is constant for the continuous equations of motion. Note that the AAM can be separated into a part ($\mathcal{M}_r$) associated with the motion of the atmosphere relative to Earth's surface (also known as wind AAM) and another part ($\mathcal{M}_\Omega$) associated with the angular velocity Ω of Earth's surface (referred to as mass AAM)

$$\mathcal{M} = \mathcal{M}_r + \mathcal{M}_\Omega = (ur \cos \varphi) + (\Omega r^2 \cos^2 \varphi). \quad (62)$$

The total energy equation integrated over the global domain is also derived in Appendix B. The final equation is

$$\frac{\partial}{\partial t} \int_{\eta=0}^{\eta=1} \iint_S \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) \sum_{\ell \in \mathcal{L}_{\text{all}}} \left[m^{(\ell)} \left(K + c_p^{(\ell)} T + \Phi_s \right) \right] dA d\eta^{(d)} = 0, \quad (63)$$

where $K = \frac{1}{2} \mathbf{v} \cdot \mathbf{v}$. Note that the energy terms (inside square brackets) in (63) separate into contributions from each component of moist air. The total energy equation (63) shows that the equations of motion used in this paper conserve a moist total energy that includes condensates. That said, the CAM physics package energy fixer assumes that the *perfect* adiabatic dynamical core conserves an energy where $c_p^{(\text{wv})} = c_p^{(d)}$, $c_p^{(\ell)} = 0$ for $\ell \in \mathcal{L}_{\text{cond}}$ and $m^{(\ell)} = 0$ for $\ell \in \mathcal{L}_{\text{cond}}$ in (63) (Williamson et al., 2015) in which case the integrand in (63) becomes

$$\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) (1 + m^{(\text{wv})}) \left[\left(K + c_p^{(d)} T + \Phi_s \right) \right]. \quad (64)$$

The discrepancy between the more comprehensive energy formula (63) and the CAM physics formula for total energy is about 0.5 W/m² (Taylor, 2011). By only including dry air and water vapor in ρ and setting $c_p^{(\text{wv})} = c_p^{(d)}$ in the equations of motion, the dynamical core (in the absence of truncation errors) will conserve the energy used in CAM physics. CAM-SE's total energy can be made consistent with CAM physics (described above), and it is enabled/disabled in the model code with the logical parameter `lcp_moist`.

3. Discretized Equations of Motion

3.1. Vertical Discretization

In the vertical the atmosphere is discretized into n_{lev} floating Lagrangian layers. The vertical index is 1 for the upper most level and n_{lev} in the lower most level. The level interfaces are referred to as half-levels so that layer k is bounded by interface level $k + 1/2$ and $k - 1/2$. Since we are using a dry-mass vertical reference coordinate, the dry air mass per unit area at the layer interfaces is defined in terms of the hybrid coefficients A and B that are only a function of level index

$$M_{k+1/2}^{(d)} = A_{k+1/2} M_t^{(d)} + B_{k+1/2} M_s^{(d)} \quad (65)$$

and similarly for full-levels k where $A_k = \frac{1}{2} (A_{k+1/2} + A_{k-1/2})$ and $B_k = \frac{1}{2} (B_{k+1/2} + B_{k-1/2})$. Note that if the d is removed from the above equation and multiplied by gravity, then the levels would be based on (moist) pressure, that is, the vertical coordinate becomes the usual hybrid vertical coordinate used in many global hydrostatic models (assuming no condensates). In every `se_rsplitt` dynamics time steps the prognostic variables on the floating Lagrangian levels are remapped to the dry-mass reference coordinate (65) (see section 3.1.5).

CAM-SE is based on a Lorenz vertical staggering—the full-level prognostic variables are $\mathbf{v}_k$ (velocity vector), T_k (temperature), $\Delta M_k^{(d)} = M_{k+1/2}^{(d)} - M_{k-1/2}^{(d)}$ (dry air mass per unit area in layer k), and $\Delta M_k^{(d)} m_k^{(\ell)}$ (tracer mass per unit area in a layer). At the level interfaces, the geopotential, (moist) pressure, and vertical velocity are defined. In the equations of motion, the full-level (moist) pressure, geopotential height, vertical velocity, and density

are needed and choices must be made on how these variables are derived (in discretized space) from the prognostic variables. Here we use the energy and angular momentum conserving method of Simmons and Burridge (1981) to compute these quantities at full levels. For simplicity, we do not consider the discretization in the horizontal in this section.

3.1.1. Pressure

The half-level (moist) pressure is

$$p_{k+1/2} = p_t + g \sum_{j=1}^k \Delta M_j^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m_j^{(\ell)} \right), \quad (66)$$

where p_t is the pressure at the model top, $\eta^{(d)} = 0$. The full-level (moist) pressure is obtained by averaging (Simmons & Burridge, 1981)

$$p_k = \frac{p_{k+1/2} + p_{k-1/2}}{2}. \quad (67)$$

3.1.2. Geopotential Height

Discretizing (52) in the vertical yields

$$\Phi_{k+1/2} = \Phi_s + R^{(d)} \sum_{j=k}^{\text{nlev}} \left[\frac{(T_v)_j}{p_j} \right] \Delta p_j, \quad (68)$$

where the (moist) pressure p_k is given in (67). The half-level geopotential is computed by averaging

$$\Phi_k = \frac{\Phi_{k+1/2} + \Phi_{k-1/2}}{2}. \quad (69)$$

3.1.3. Vertical Pressure Velocity

The vertical pressure velocity ω is obtained by discretizing (56). The first term on the right-hand side of (56) can be computed by using the continuity equations for dry air mass and water masses in each layer (51)

$$\frac{\partial}{\partial t} \left[\sum_{j=1}^k \Delta M_j^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m_j^{(\ell)} \right) \right] = - \sum_{j=1}^k \nabla_{\eta^{(d)}} \cdot \left[\Delta M_j^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m_j^{(\ell)} \right) \mathbf{v}_j \right], \quad (70)$$

so that the vertical pressure velocity at half-levels is given by

$$\omega_{k+1/2} = -g \sum_{j=1}^k \nabla_{\eta^{(d)}} \cdot \left[\Delta M_j^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m_j^{(\ell)} \right) \mathbf{v}_j \right] + g \sum_{j=1}^k \mathbf{v}_k \cdot \nabla_{\eta^{(d)}} \left[\Delta M_j^{(d)} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m_j^{(\ell)} \right) \right], \quad (71)$$

and full-level ω is

$$\omega_k = \frac{\omega_{k+1/2} + \omega_{k-1/2}}{2}. \quad (72)$$

3.1.4. Density

Full-level density is computed from the ideal gas law (17)

$$\rho_k = \frac{p_k}{R^{(d)} T_k^{(v)}}, \quad (73)$$

where the (moist) pressure p_k is computed as described in section 3.1.1. The virtual temperature is based on prognostic variables defined at the layer centers, so simple substitution into (16) yields $(T^{(v)})_k$ and similarly for the computation of $(c_p)_k$.

3.1.5. Vertical Remapping

To avoid excessive deformation or even crossing of the floating Lagrangian levels, the prognostic variables defined at

$$M_{k+1/2}^{(d)} = M_t^{(d)} + \sum_{j=1}^k \Delta M_j^{(d)} \quad (74)$$

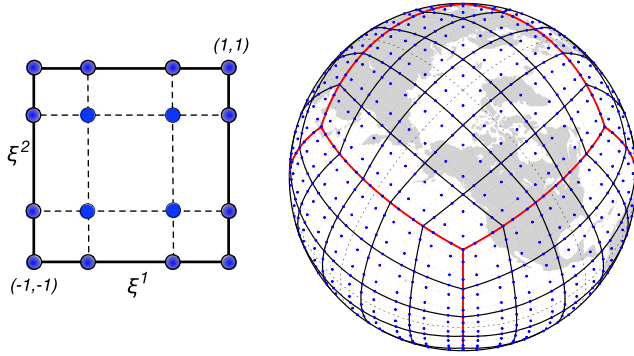


Figure 2. The left panel shows the Gauss-Lobatto-Legendre (GLL) grid with $N_p \times N_p$ quadrature points defined on a standard element $[-1, 1]^2$, where $N_p = 4$. The right panel shows the cubed-sphere (S) grid system tiled with $6N_e^2$ spectral elements Ω_e , where N_e is the number of elements in each coordinate direction on a panel (in this case $N_e = 5$). Each element Ω_e on S has the GLL grid structure.

are remapped back to the (Eulerian) reference levels given in (65) every `se_rsplit` dynamics time steps (Lin, 2004). In the remapping process we enforce conservation of mass by mapping $m_k^{(e)} \Delta M_k^{(d)}$ using the piecewise-parabolic method (PPM; Colella & Woodward, 1984) and applying a standard shape-preserving limiter to avoid unphysical (in particular negative) mixing ratios in the remapping process. The internal energy is also conserved during the remapping process by mapping $\sum_{\ell \in \mathcal{L}_{\text{all}}} c_p^{(\ell)} m_k^{(e)} T_k \Delta M_k^{(d)}$. Note that temperature must be recovered from the internal energy using the remapped tracer values for $m_k^{(e)}$. A shape-preserving filter is also used for the remapping of internal energy so that an isothermal profile remains isothermal if the limiter is active on one of the water variables, $m_k^{(e)}$. Note that if the limiter is active at the same points for more than one of the water species, then one cannot guarantee preservation of an isothermal atmosphere when nonlinear limiting filters are turned on (see, e.g., section 2.5 in Lauritzen & Thuburn, 2012).

The moist mass-weighted velocity components, $\sum_{\ell \in \mathcal{L}_{\text{all}}} m_k^{(e)} \Delta M_k^{(d)} u_k$ and $\sum_{\ell \in \mathcal{L}_{\text{all}}} m_k^{(e)} \Delta M_k^{(d)} v_k$ respectively, are remapped separately. Mapping the moist mass weighted velocity components conservatively leads to an AAM

conserving vertical remapping algorithm. For the velocity components we do not enforce shape preservation in the vertical remapping process to reduce the amount of kinetic energy dissipation.

3.2. Horizontal Discretization Using the SE Method

The CAM-SE uses a cubed-sphere geometry originally introduced by Sadourny (1972) to represent the planet Earth. The spherical surface S is a patched domain, which is partitioned into nonoverlapping quadrilateral elements Ω_e such that $S = \cup \Omega_e$ (see Figure 2). On S each 2-D element $\Omega_e(x^1, x^2)$ defined in terms of central (gnomonic) projection angles $x^1, x^2 \in [-\pi/4, \pi/4]$, which serve as the independent variables in the computational domain. The mapping from cube to sphere results in a nonorthogonal curvilinear coordinate system on S , with the metric tensor G_{ij} and analytic Jacobian $\sqrt{G} = |G_{ij}|^{1/2}$, $i, j \in \{1, 2\}$. A physical vector quantity such as the wind vector $\mathbf{v} = (u, v)$, defined on S in orthogonal lat-lon coordinates, can be uniquely expressed in tensor form using conventional notations as the covariant (u_1, u_2) and contravariant (u^1, u^2) vectors using the 2×2 transformation matrix $\mathbf{D}$ associated with the gnomonic mapping such that $\mathbf{D}^T \mathbf{D} = G_{ij}$ (see Nair et al. (2005) for the details):

$$\begin{bmatrix} u \\ v \end{bmatrix} = \mathbf{D} \begin{bmatrix} u^1 \\ u^2 \end{bmatrix} = \mathbf{D}^{-T} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}. \quad (75)$$

The governing equations defined in familiar vector form can also be expressed in general tensor form. In order to describe the SE discretization process in simple terms, we consider the following conservation law on S for an arbitrary scalar ϕ :

$$\frac{\partial \phi}{\partial t} + \nabla \cdot \mathbf{F}(\phi) = S(\phi), \quad (76)$$

where

$$\nabla \cdot \mathbf{F}(\phi) = \frac{1}{\sqrt{G}} \left[\frac{\partial \sqrt{G} F^1}{\partial x^1} + \frac{\partial \sqrt{G} F^2}{\partial x^2} \right]. \quad (77)$$

In the special case of the flux-form transport equation the contravariant fluxes $(F^1, F^2) = (u^1 \phi, u^2 \phi)$, and $S(\phi)$ is an arbitrary source term.

3.3. SE Spatial Discretization in 2D

The SE solution process involves casting the partial differential equation in Galerkin form, that is, by multiplying (76) with a test (weight) function ψ and integrating over the domain S ,

$$\int_S \psi \left[\frac{\partial \phi}{\partial t} + \nabla \cdot \mathbf{F}(\phi) - S(\phi) \right] dS = 0. \quad (78)$$

A computational form of (78) is obtained by applying Green's theorem, resulting in the weak Galerkin form, as follows:

$$\int_S \psi \frac{\partial \phi}{\partial t} dS = \int_S \nabla \psi \cdot \mathbf{F}(\phi) dS + \int_S \psi S(\phi) dS, \quad (79)$$

where the approximation to the solution ϕ and the test function belong to a polynomial space V^N . The SE method consists of partitioning the domain into nonoverlapping elements and solving the global problem locally on each element, where the solution is approximated by using a set of basis (polynomial) functions of prescribed order N . A basic assumption used in SE (or continuous Galerkin) method is that the global basis corresponding to (79) is C^0 continuous. Therefore, the problem (79) can be solved locally for each element Ω_e , if there is a mechanism by which the solution maintains C^0 continuity at the element boundaries as required by the SE discretization.

For efficient evaluation of the integral equation (79), the SE method employs the Gauss-Lobatto-Legendre (GLL) quadrature rule for integrals and collocation differentiation for derivative operators. All the corresponding numerical operations are performed on a square $[-1, 1]^2$ known as the standard (or reference) element. In order to facilitate local mesh refinement, the SEs Ω_e on S are defined as arbitrary spherical quadrilaterals in the CAM-SE grid system, which should be mapped onto the standard element. A direct way to address this problem is establishing a transformation $\mathcal{J}_e : \Omega_e \rightarrow [-1, 1]^2$, where $\mathcal{J}_e$ may be considered as a composite mapping combining the gnomonic and the quadrilateral to standard element mapping. Let the Jacobian associated with the composite mapping be $J_e = J_e(\sqrt{G})$. Then an arbitrary surface integral on Ω_e can be expressed in terms of local coordinates $\xi^1, \xi^2 \in [-1, 1]$ and the Jacobian J_e

$$\int_{\Omega_e} \psi(x^1, x^2) d\Omega_e = \int_{-1}^1 \int_{-1}^1 J_e(\xi^1, \xi^2) \psi(\xi^1, \xi^2) d\xi^1 d\xi^2 \approx \sum_{k=0}^N \sum_{l=0}^N w_k w_l J_e(\xi_k^1, \xi_l^2) \psi(\xi_k^1, \xi_l^2), \quad (80)$$

where w_k, w_l are the Gauss quadrature weights.

In the case of GLL quadrature rule, the nodal points $\xi_k, k = 0, 1, \dots, N$, are the roots of the polynomial $(1 - \xi^2)P'_N(\xi) = 0, \xi \in [-1, 1]$, and the corresponding GLL quadrature weights are given by

$$w_k = \frac{2}{N(N+1)[P_N(\xi_k)]^2},$$

where $P_N(\xi)$ is the Legendre polynomial of degree N .

For the SE discretization it is customary to use Lagrange polynomials $h_k(\xi)$, with roots at the GLL quadrature points ξ_k , as basis functions. This setup provides discrete orthogonality for the basis function $h_k(\xi)$, which is formally defined as

$$h_k(\xi) = \frac{(\xi^2 - 1)P'_N(\xi)}{N(N+1)P_N(\xi_k)(\xi - \xi_k)}. \quad (81)$$

Note that there are $N+1 = N_p$ GLL quadrature points in 1-D, and $N_p \times N_p$ GLL points are needed for 2-D spectral elements Ω_e . Figure 2 shows the GLL grid with $N_p = 4$ on the left panel, and the right panel shows the cubed-sphere grid S tiled with elements Ω_e , each with the GLL grid points.

A semidiscrete form of (79) on an element Ω_e can be obtained by approximating the solution as a tensor product of 1-D Lagrange basis $\{h_k(\xi)\}_{k=0}^N$ such that

$$\phi|_{\Omega_e} \approx \phi^e(\xi^1, \xi^2, t) = \sum_{k=0}^N \sum_{l=0}^N \phi_{kl}^e(t) h_k(\xi^1) h_l(\xi^2), \quad (82)$$

where $\phi_{kl}^e(t) = \phi^e(\xi_k^1, \xi_l^2, t)$ are the nodal grid point values of the solution and defining the test function as $\psi(\xi^1, \xi^2) = h_k(\xi^1) h_l(\xi^2)$.

By using (80) and the discrete orthogonality property of $h_k(\xi)$, we get a completely decoupled system of ODEs (Ordinary Differential Equation) on Ω_e , for each grid point (k, l)

$$\tilde{M}_{kl}^e \frac{d}{dt} \phi_{kl}^e(t) = A_{kl}^e + S_{kl}^e \quad (83)$$

$$\tilde{M}_{kl}^e = \int_{-1}^1 \int_{-1}^1 J_e h_k(\xi^1) h_l(\xi^2) d\xi^1 d\xi^2 = J_e(k, l) w_k w_l \quad (84)$$

$$A_{kl}^e = \sum_{i=0}^N J_e^{(1)}(i, l) F_{il}^1 D_{ik}^{(1)} w_l w_i + \sum_{i=0}^N J_e^{(2)}(k, i) F_{ki}^2 D_{li}^{(2)} w_k w_i \quad (85)$$

$$S_{kl}^e = J_e(k, l) w_k w_l S(U_{kl}) \quad (86)$$

where $J_e^{(i)} = J_e \partial \xi^i / \partial x^i$ is the metric term and $D_{ik}^{(i)}$ is the derivative matrix $h_k'(\xi_j^i)$, along the x^i direction and $i \in \{1, 2\}$.

The ODEs (83) can be written in a formal matrix-vector form for Ω_e following Karniadakis and Sherwin (2013):

$$\tilde{\mathbf{M}}^e \frac{d}{dT} \Phi^e = \mathbf{A}^e + \mathbf{S}^e + \mathbf{B}^e, \quad (87)$$

where $\tilde{\mathbf{M}}^e$ is the so-called mass matrix, which is a diagonal matrix with entries $\tilde{M}_{kl}^e$. $\mathbf{B}^e$ indicates the boundary terms for the element Ω_e , which is a key component linking the local and global problem (79) and enforcing C^0 continuity for solutions across element boundaries.

The global matrices associated with (79) can be obtained by summing the contributions from elemental matrices, and this procedure is known as the direct stiffness summation (DSS). However, the global matrices are not explicitly constructed. In practice, the DSS operation replaces interface values of two contiguous elements sharing the same physical location by the weighted sum (average) so that the boundary nodes get unique values, which maintains the continuity of the global solution across the element edges. This strategy has been adopted in CAM-SE. Note that the DSS operation does not affect interior nodal values of any element and preserves global conservation of the SE discretization for (76). The elemental discretization (87) combined with the DSS operation leads to the time-dependent system of ODE corresponding to (76),

$$\frac{d}{dT} \phi(t) = \text{DSS}(\phi). \quad (88)$$

The details of the discretization of the dissipation operators are provided in Appendix A.

Note that nodes on the element boundary are shared between elements and that after each Runge-Kutta (RK) step the shared nodes will have different values. The elements are coupled by averaging the two solutions at the shared nodes (so the halo communicated between elements is only one node/point deep). This is referred to as the DSS operation. The basis function representation of the solution is therefore only C^0 at element boundaries. The ability to obtain high-order accuracy with only edge point communication is an attractive feature of the spectral element method (Canuto et al., 2007; Maday & Patera, 1987). Another aspect of the spectral element is that the basis function representation is spectral on each element so if, for example, one of the quadrature values is changed, then it affects the basis functions throughout the element (except at the GLL nodes).

3.4. Mimetic Discretization

A numerical method is mimetic (or compatible) if key integral properties of divergence, gradient, and curl operators are mimicked in discretized space. The CAM-SE discretization satisfies the divergence/gradient adjoint relation

$$\int \phi \nabla \cdot \mathbf{v} dS + \int \mathbf{v} \cdot \nabla \phi dS = 0 \quad (89)$$

in discretized space (Taylor & Fournier, 2010). This property can be used to show the inherent conservation properties of CAM-SE in the horizontal discretization (see Taylor, 2011, for details). Mass is conserved, and, in the absence of viscosity terms, energy is conserved with exact time stepping.

Since the CAM-SE discretization is mimetic, the adiabatic, frictionless discretization of the equations of motion conserves total moist energy to within time-truncation errors. The equivalent internal energy change due to hyperviscosity damping of the velocity vector is added as frictional heating in the thermodynamic equation so the energy budget for the viscosity on the momentum equations is closed. The dissipation of energy due to hyperviscosity on temperature and dry mass, however, is not energy conserving. The vertical remapping does not conserve total energy either, but it does conserve moist internal energy and AAM. Hyperviscosity reduces variance but is a mass-conserving operation in CAM-SE.

3.5. Temporal Discretization

In a typical CAM setup where the model top is approximately 40 km, the maximum stable time step for solving the continuity equation for water species ℓ , $\ell \in \mathcal{L}_{\text{water}}$ (referred to as tracer advection) is limited by the maximum horizontal advective wind speed. The maximum stable time step for the remaining equations of motion (thermodynamic equation, momentum equations, and the continuity equation for dry air) is limited by the fastest gravity waves. Since the maximum advective winds are typically slower than the fastest gravity waves, different time-stepping methods are used for tracer advection and the remaining equations of motion. While the overall time step is the same, different RK methods are used to maintain stability.

The SE tracer advection algorithm uses a three-stage RK strong stability preserving time-stepping method, ensuring that the time step will preserve any shape-preserving properties preserved by the underlying spatial discretization (Spiteri & Ruuth, 2002). The shape-preserving filter used is described in Guba, Taylor, and St-Cyr (2014). The shape-preserving SE tracer advection algorithm is formally second-order accurate.

The momentum, thermodynamic, and dry air continuity equations are solved using an explicit nonlinearly third-order accurate five-stage RK method based on Kinnmark and Gray (1984) and described in Guerra & Ullrich (2016; see their equation (56)). For sake of completeness, this scheme is described in Appendix C and will be referred to as KG53. The choice of a five-stage large time step RK method for dynamics and a three-stage RK strong stability-preserving method allows both dynamics and tracers to use the same time step. For mass/tracer-mass consistency, the dynamics computes a mean dry-air mass flux averaged over the five RK stages. This mean flux is then used by all three tracer RK stages. The inviscid equations of motion are advanced one dynamics time step using the KG53 method followed by an application of the hyperviscosity operators through solving the advection-diffusion equation of the updated prognostic variables using forward Euler time stepping. Note that it is necessary to subcycle the hyperviscosity step for stability. The viscosity terms are computed as described in section 2.6.1 and discretized using the SE method presented in Appendix A.

The time steps in CAM-SE are controlled with name list variables `se_nsplrit`, `se_rsplrit`, `se_qsplrit`, and `se_hypervis_subcycle` so that the time steps for vertical remapping Δt_{remap} , tracer advection Δt_{trac} (RK3), KG53 time stepping for nontracers Δt_{dyn} , and hyperviscosity Δt_{hyper} are

$$\Delta t_{\text{remap}} = \frac{\Delta t_{\text{phys}}}{\text{se_nsplrit}}, \quad (90)$$

$$\Delta t_{\text{tracer}} = \frac{\Delta t_{\text{phys}}}{\text{se_nsplrit} * \text{se_qsplrit}}, \quad (91)$$

$$\Delta t_{\text{dyn}} = \frac{\Delta t_{\text{phys}}}{\text{se_nsplrit} * \text{se_rsplrit} * \text{se_qsplrit}}, \quad (92)$$

$$\Delta t_{\text{hyper}} = \frac{\Delta t_{\text{phys}}}{\text{se_nsplrit} * \text{se_rsplrit} * \text{se_hypervis_subcycle} * \text{se_qsplrit}}, \quad (93)$$

where Δt_{phys} is the time step used for computing physics tendencies. For the 1° configuration ($N_e = 30$, $N_p = 4$ where N_e is the number of elements in each coordinate direction on a panel and N_p is the number of quadrature points along each coordinate direction in an element) the physics time step is $\Delta t_{\text{phys}} = 30$ min, and the subcycling parameters are `se_nsplrit=2`, `se_rsplrit=3`, `se_qsplrit=1`, and `se_hypervis_subcycle=3`. See Table 1 for other resolutions.

3.6. Coupling to Physics

CAM-SE uses a time-split approach in which dynamics advances the model state and the physics tendencies are based on the dynamics updated state. When CAM-SE starts a simulation, the physics are called first and then the dynamics. The question is then how to add the physics tendencies in the dynamical core. CAM-SE supports several physics-dynamics coupling methods. Let F_X be the physics tendency for prognostic variable X in level k (for notational simplicity the vertical index is dropped). The different coupling methods are detailed below and are identified with the rather arbitrary name *ftype*. A fuller discussion with results about different coupling methods is the content of a separate paper. Some results are provided by Thatcher and Jablonowski (2016).

Table 1
Physics Time Step (Column 3), Subcycling Parameters (Columns 4–6) as a Function of Resolution (Columns 1 and 2)

Horizontal resolution	Δx_{ave}	Δt_{phys}	se_nsplitt	se_rsplitt	se_hypervis_ subcycle
ne16np4	~208 km	1,800 s	1	3	3
ne30np4	~111 km	1,800 s	2	3	3
ne60np4	~56 km	900 s	2	3	3
ne120np4	~28 km	450 s	2	3	3
ne240np4	~14 km	225 s	2	3	3
ne0np4CONUS30 × 8	~111 → ~14 km	600 s	5	3	4

Note. The horizontal resolution is specified in the format *ne30np4* which refers to $N_e = 30$ and $N_p = 4$, similar for other resolutions. $N_e = 0$ denotes a variable resolution configuration. The approximate average quadrature node spacing at the equator, Δx_{ave} (in kilometers), is also listed (column 2). For all configurations presented in this paper $\text{se_qsplitt} = 1$.

3.6.1. ftype=0 Configuration (Default)

In this case, compute the physics tendencies and add $\Delta t_{\text{remap}} F_X$ to the state of X , advance the dynamics Δt_{remap} seconds based on the updated state, add $\Delta t_{\text{remap}} F_X$ to the dynamics updated state of X , advance the dynamical core, and so on. In other words, the forcing is split into $\frac{\Delta t_{\text{phys}}}{\Delta t_{\text{remap}}}$ equal chunks and added throughout the dynamics. The ftype=0-option is used by default in CAM-SE and used for all simulations presented in this paper.

The CAM parameterization package returns mixing ratio tendencies for tracers. We convert the mixing ratio tendencies to mass tendencies $F_X \Delta M_k^{(d)}$ since the prognostic variable for tracers in the dynamical core is $\Delta M_k m_k^{(d)}$.

Generally, modelers do not allow the tracer tendencies to drive the mixing ratio of tracers negative; however, this may happen using ftype=0. If the tendency drives a mixing ratio negative, the mixing ratio is set to 0. In this case, only a fraction of the entire tracer tendency is added, leading to an inconsistency in how much mass the physics package wants to remove and the amount of tracer mass actually removed in the physics-dynamics coupling code in the dynamical core. This leads to energy conservation errors, and it has recently been shown to lead to hydrological cycle biases in coupled climate simulations (Zhang et al., 2017).

3.6.2. ftype=1 Configuration

In this configuration the entire physics forcing is added to the dynamics state $\Delta t_{\text{phys}} F_X$ which is equivalent to having the physics module update the model state. This coupling method is used in CAM-FV (Lin, 2004). Note that contrary to the ftype=0 configuration, this configuration always provides a closed mass budget for tracers in terms of physics tendencies being fully applied in the dynamical core. The disadvantage of using ftype=1 in CAM-SE is that the use of a long physics time step results in the appearance of spurious gravity waves due to large physics tendencies.

3.6.3. ftype=2 Configuration

This configuration constitutes a hybrid approach where mass variables (tracers) use the ftype=1 physics-dynamics coupling method and all other variables use the ftype=0 method.

4. Results

The evaluation of the SE dynamical core with CAM6 physics in an AMIP configuration simulation is the subject of a separate paper. Here the new dynamical core is evaluated in simpler configurations. First, the new dynamical core version is compared with the old version using an idealized moist baroclinic wave, thereby avoiding the complexity of a full physics parameterization suite that may make it harder to distinguish between cause and effect. It is confirmed that the new SE dynamical core version converges to within the uncertainty of high-resolution reference solutions computed with the old SE dynamical core (that has been extensively validated) and the current *workhorse* dynamical core for 1° climate simulation; CAM-FV (finite volume; Lin, 2004). Second, the dynamical core is evaluated in an aqua-planet setup (Medeiros et al., 2016; Neale & Hoskins, 2000; Williamson et al., 2012) using CAM6 physics. In this context the conservation properties AAM and total moist energy is discussed. Thirdly the computational efficiency of CAM-SE is evaluated.

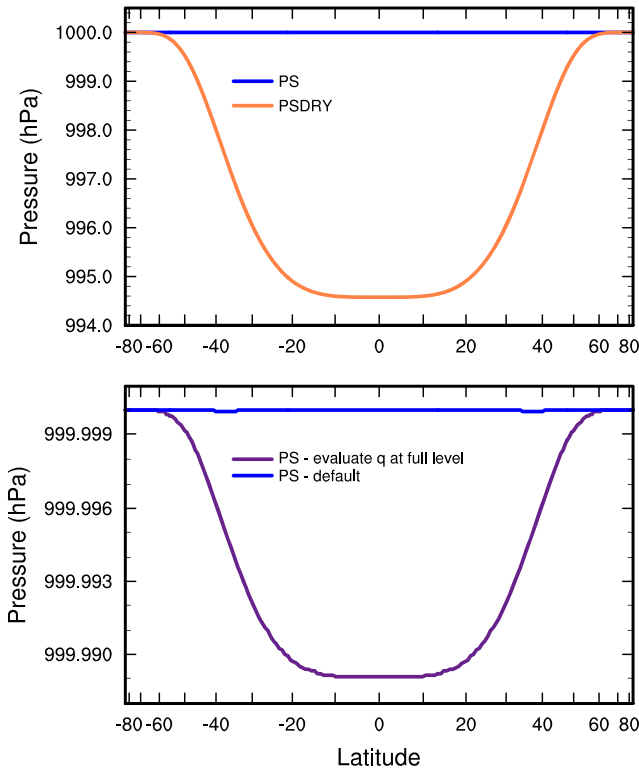


Figure 3. (top) Zonally averaged weight of dry air per unit area (orange) and moist (blue) surface pressure as a function of latitude for the numerically computed initial condition for the moist baroclinic wave. Due to increased water vapor toward the equator, the dry mass decreases, whereas the moist surface pressure is constant. (bottom) Same as the top plot but showing moist surface pressure where water vapor has been initialized by evaluating the analytic specific humidity formula at full levels (purple) and initializing the mixing ratio for water vapor in terms of moist and dry mass coordinates at half levels which effectively integrates humidity over the layer (blue). The bottom plot uses 30 levels (Community Atmosphere Model 5). Note that the top and bottom plots have different scales on the y axis.

4.1. Idealized Moist Baroclinic Wave With Kessler Microphysics

For the validation of the new dynamical core version in a simplified setup, we use a moist variant of the dry baroclinic wave of Ullrich et al. (2014) with Kessler microphysics (Kessler, 1969). This test case configuration was part of the Dynamical Core Model Intercomparison Project (DCMIP) 2016 test case suite (Ullrich et al., 2017). The initialization of the atmospheric state for the moist baroclinic wave is based on analytic expressions for T_v , $\mathbf{v}$, p , and $q^{(wv)}$ as a function of latitude and height, (φ, z) . The surface geopotential is constant, $\Phi_s = 0$, and the moist surface pressure is constant $p_s = 1,000$ hPa. The analytical expressions for temperature, velocity components, (moist) pressure, and specific humidity, denoted $T_v(\varphi, z)$, $\mathbf{v}(\varphi, z)$, $p(\varphi, z)$, and $q^{(wv)}(\varphi, z)$, respectively, are given in Appendix D. When using a moist vertical hybrid pressure coordinate, the full-level pressures p_k of the initial condition are known from the hybrid coefficients, A_k and B_k , and one can iteratively solve for z_k given (moist) pressure: $p_k = p(\varphi, z_k)$ (see Ullrich et al., 2014). Once the full-level heights, $z_k(\varphi)$, are known, then the specific humidity, virtual temperature, and velocity components can be computed by evaluating the analytical expressions at (φ, z_k) . In the DCMIP 2016 test case documentation the virtual temperature is converted to temperature using $T_v = T(1 + \epsilon q^{(wv)})$. For a dry-mass vertical coordinate model the initialization procedure is more complicated as described below.

4.1.1. Initialization of the Moist Baroclinic Wave Using a Dry-Mass Vertical Coordinate

The challenge is to extract dry mass from the initial state defined in terms of (moist) pressure and preserve a balanced initial condition. Since the initial condition does not contain condensates we can assume that moist air is a gas that will simplify the derivations. First, we write the dry atmosphere hydrostatic relation (46) in terms of specific humidity, virtual temperature, and (moist) pressure

$$\frac{\partial M^{(d)}}{\partial z} = -\rho^{(d)}, \quad (94)$$

$$= -\frac{\rho}{1 + m^{(wv)}}, \quad (95)$$

$$= -\frac{p}{R^{(d)}T_v} \frac{1}{(1 + m^{(wv)})}, \quad (96)$$

$$= -\frac{p}{R^{(d)}T_v} (1 - q^{(wv)}), \quad (97)$$

where we have substituted the ideal gas law and used $\left(\frac{1}{1+m^{(wv)}}\right) = (1 - q^{(wv)})$ for an atmosphere not containing any condensates. Hence, the dry mass as a function of latitude, φ , and height, z , is

$$M^{(d)}(\varphi, z) = M_t^{(d)} - \int_z^{z_t} \frac{p(\varphi, z)}{R^{(d)}T_v(\varphi, z)} (1 - q^{(wv)}(\varphi, z)) dz, \quad (98)$$

where z_t is the height of the model top computed by iteratively solving (D4) with $p(\varphi, z) = p_t = g M_t^{(d)}$ (assuming that there is no moisture above the model top). To initialize the dry mass levels, $M_s^{(d)}$ is needed. It is computed by integrating (98) with lower integral bound $z = z_s = 0$ m and approximating the integral using 20-point Gaussian quadrature (high-order quadrature is necessary to reduce integral errors). The weight of dry air per unit area is shown in Figure 3. The full dry-mass levels are given in terms of the hybrid coefficients $M_k^{(d)}(\varphi) = A_k M_t^{(d)} + B_k M_s^{(d)}(\varphi)$. The height, $z_k(\varphi)$, of the full dry-mass levels is computed by iteratively solving (98) with $M^{(d)}(\varphi, z_k) = M_k^{(d)}(\varphi)$. Once the heights are known the virtual temperature and velocity components can be computed by evaluating the analytical expressions at (φ, z_k) as is done for the moist vertical coordinate initialization. If we do the same for specific humidity, then the moist surface pressure (which is a diagnostic when using dry-mass vertical coordinates) deviates more than 1 Pa in the tropics from the analytical value of 1,000 hPa (see Figure 3). As a result a spurious zonal signal in surface pressure with the same amplitude

Day 10 moist baroclinic wave

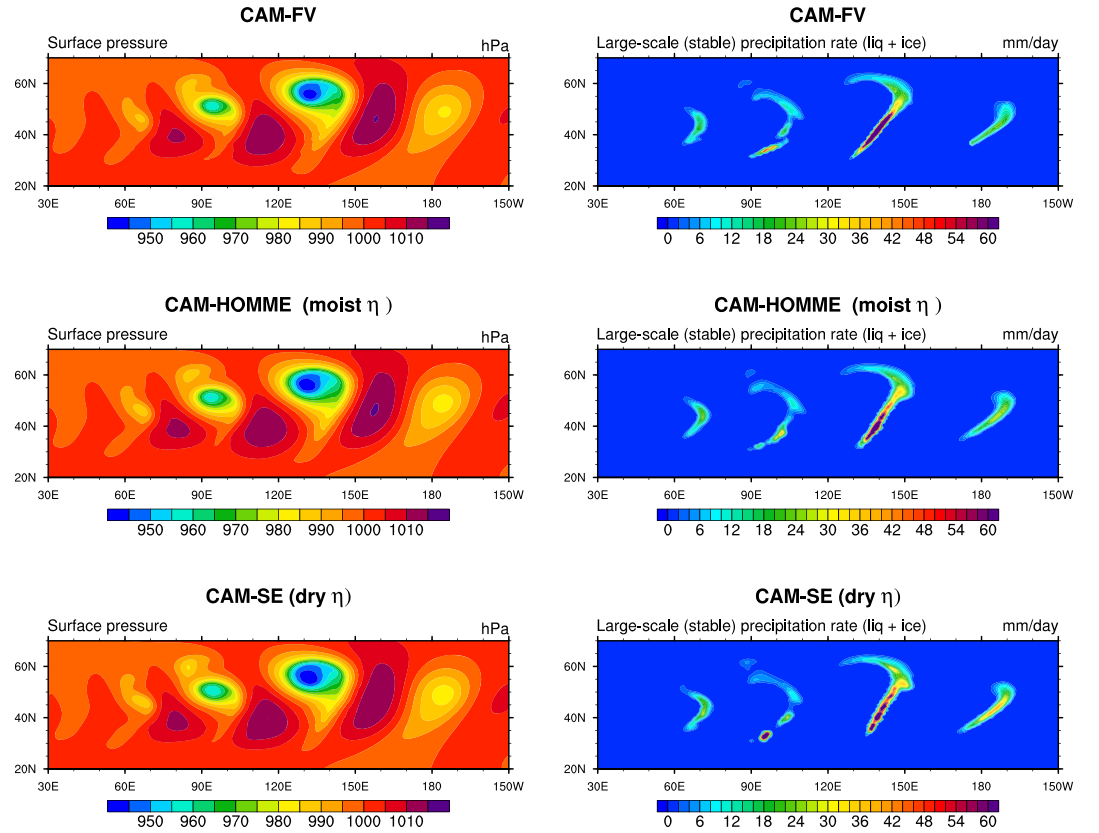


Figure 4. Left column shows moist surface pressure at day 10 for the moist baroclinic wave test case for (row 1), the finite-volume dynamical core, (row 2) CAM-HOMME version of the SE dynamical core based on a (moist) pressure vertical coordinate, and (row 3) the dry-mass vertical coordinate version of SE presented in this paper. The right column is the same as the left but for large-scale precipitation rate. All results on this figure are at 1° horizontal resolution. CAM = Community Atmosphere Model; FV = finite volume; HOMME = High-Order Methods Modeling environment; SE = spectral element.

as the initial gravity waves appears in the simulations. Hence, specific humidity must be initialized more carefully in order to obtain a more balanced initial condition which is discussed in the next paragraph.

As for $M^{(d)}$, the mass of water vapor $M^{(wv)}$ per unit area can be written as

$$M^{(wv)}(\varphi, z) = - \int_z^{z_t} \frac{p(\varphi, z)}{R^{(d)} T_v(\varphi, z)} q^{(wv)}(\varphi, z) dz, \quad (99)$$

so that the (moist) pressure at half levels can be computed as the sum of the weight of dry air and moisture per unit area since there are no condensates present in the initial condition

$$p(\varphi, z_{k+1/2}) = g M^{(d)}(\varphi, z_{k+1/2}) + g M^{(wv)}(\varphi, z_{k+1/2}), \quad (100)$$

where $z_{k+1/2}$ is computed by the same iterative procedure as for full levels. An integrated value for water vapor in a layer can now be computed from

$$m^{(wv)}(\varphi, z_k) = \frac{p(\varphi, z_{k+1/2}) - p(\varphi, z_{k-1/2})}{g M^{(d)}(\varphi, z_{k+1/2}) - g M^{(d)}(\varphi, z_{k-1/2})} - 1. \quad (101)$$

Using this method for initializing the mixing ratio for water vapor, the moist surface pressure is within 0.01 Pa of the analytical value of 10^5 Pa (see Figure 3). Once $m^{(wv)}(\varphi, z_k)$ is computed, the temperature $T(\varphi, z_k)$ can be recovered from $T_v(\varphi, z_k)$ by using (16).

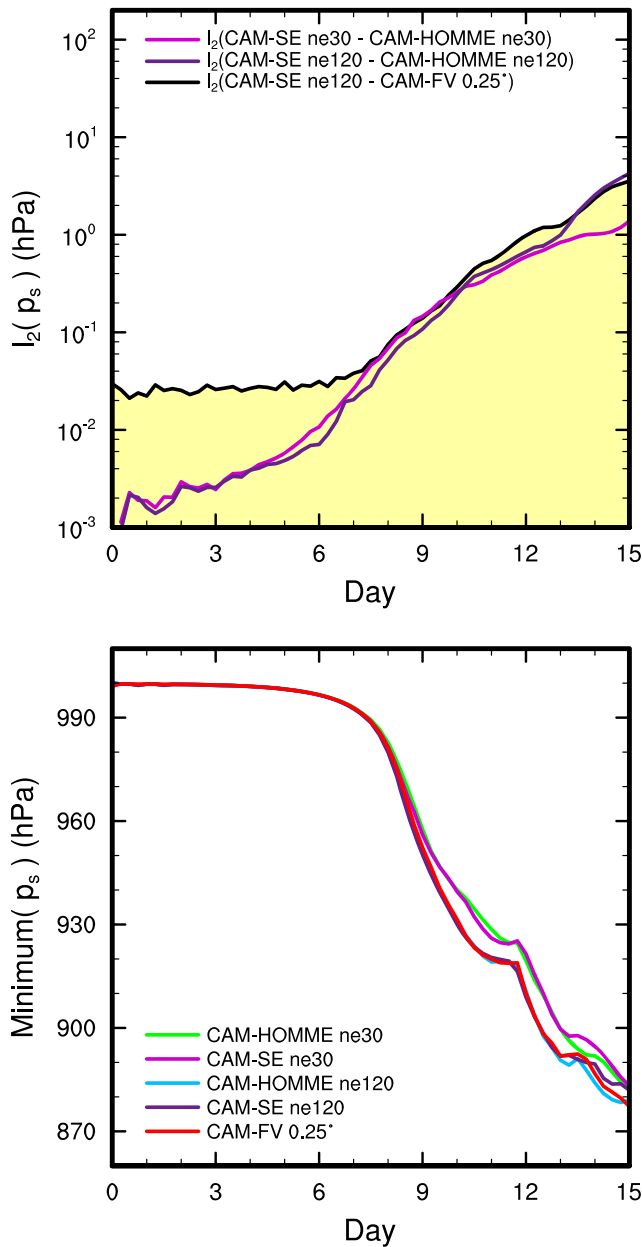


Figure 5. (top) l_2 difference norms of p_s in the moist baroclinic wave simulations. l_2 values lying within the yellow region fall below the estimate of the uncertainty in the reference solution (black curve). (bottom) Global minimum p_s in the moist baroclinic wave simulations. CAM = Community Atmosphere Model; SE = spectral element; HOMME = High-Order Methods Modeling environment; FV = Finite Volume.

cores, CAM-SE, and CAM-FV. At a $\frac{1}{4}^\circ$ resolution, the moist baroclinic wave solutions are converged to within a tolerable level of error (not shown), and therefore, comparison between dynamical cores serves as an estimate of the uncertainty in the reference solutions arising from model imperfections.

The black curve in Figure 5 is the l_2 difference norm between CAM-SE and CAM-FV at approximately a $\frac{1}{4}^\circ$ resolution. l_2 values that fall below the uncertainty in the reference are considered insignificant. The l_2 between CAM-SE and CAM-HOMME generally lies near or below the uncertainty in the reference solution, indicating that the differences between CAM-SE and CAM-HOMME are insignificant for the case of the moist baroclinic wave. This is consistent with the corresponding time-varying, global minimum in p_s for all the simulations,

4.1.2. Simulation Results

As part of the *CESM simpler model* effort started by Polvani et al. (2017), the moist baroclinic wave setup has been implemented rigorously in the CESM in the sense that the configuration easily runs from CESM without code configurations using the *FKESSLER* compset. For instructions on how to run the moist baroclinic wave with Kessler microphysics see Lauritzen and Goldhaber (2017). Since the test case configuration has been implemented in the full CESM, the dynamical core interacts with the physics module as in full climate model simulations. Hence, the global energy fixer is invoked (Williamson et al., 2015) and for the dynamical cores using a (moist) pressure vertical coordinate there is an adjustment of specific humidity to conserve water after the moist physics updates (see section 3.1.6 in Neale et al., 2012). The intent of this implementation is to evaluate the dynamical core in simplified setup but exactly as the dynamical core is configured for comprehensive climate simulations.

Figure 4 depicts day 10 of the moist baroclinic wave test using the CAM-FV (Lin, 2004) dynamical core, CAM-HOMME dynamical core based on a moist vertical coordinate and the new dynamical core version (CAM-SE). The CAM-HOMME and CAM-SE (CESM2.0-SE) dynamical cores differ not only in terms of vertical coordinates but also in terms of hyperviscosity and the formula used for the heat-capacity in the thermodynamic equation. CAM-HOMME uses $c_p^{(d)}$, whereas CAM-SE (CESM2.0-SE) uses the comprehensive formula (38) that include the heat capacity of water vapor.

To provide a more quantitative measure of the difference between the moist baroclinic wave simulations, the l_2 difference norm of p_s between two simulations is computed as the time-varying global integral in spherical coordinates:

$$l_2(p_s(t)) = \left[\frac{1}{4\pi} \int_0^{2\pi} \int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} (p_{s_1}(\lambda, \varphi, t) - p_{s_0}(\lambda, \varphi, t))^2 \times \cos(\varphi) d\varphi d\lambda \right]^{\frac{1}{2}}. \quad (102)$$

The l_2 difference norm between CAM-SE and CAM-HOMME is shown in Figure 5 for the 1° ($N_e = 30$ and $N_p = 4$) and $\frac{1}{4}^\circ$ ($N_e = 120$ and $N_p = 4$) resolution simulations with 30 vertical levels (CAM5 level locations). The l_2 norms for the 1° and $\frac{1}{4}^\circ$ simulations have similar time-varying magnitudes. As the baroclinic waves evolve, the l_2 norms grow to a maximum on the order of 1 hPa by day 15. To assess the significance of the l_2 norms, an l_2 norm that serves as an estimate of the uncertainty of a high-resolution reference simulation is computed following Jablonowski and Williamson (2006). The uncertainty in the reference is taken as the l_2 between a pair of $\frac{1}{4}^\circ$ resolution moist baroclinic wave simulations using different dynamical

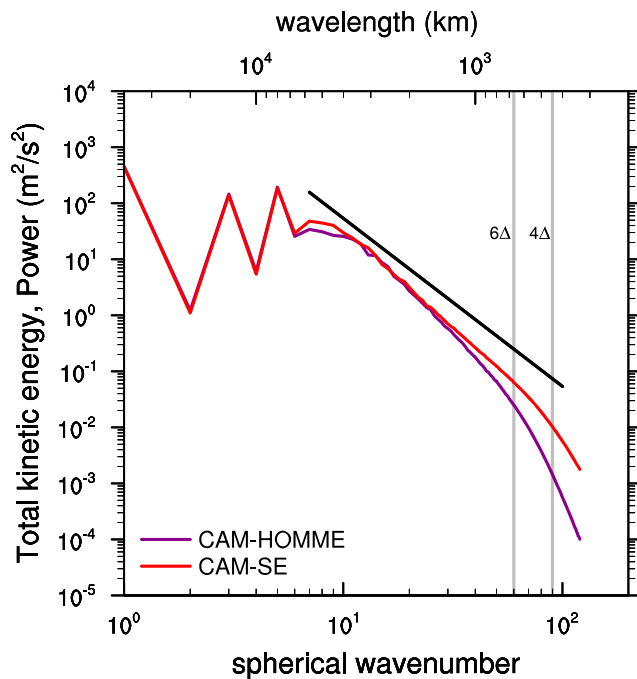


Figure 6. Total kinetic energy spectrum of the horizontal winds at the 200-hPa level in CAM-HOMME and CAM-SE at 1° horizontal resolution ($N_e = 30$ and $N_p = 4$), computed as the mean spectra from 30 days of 6-hourly instantaneous spectra. Black line is the κ^{-3} reference scaling, where κ is wavenumber. CAM = Community Atmosphere Model; HOMME = High-Order Methods Modeling environment; SE = spectral element.

which illustrates that the differences between CAM-SE and CAM-HOMME are smaller than the differences arising from increasing the horizontal resolution.

4.2. CAM6 Aqua-Planet Simulations

The CESM *simpler models* effort also supports aqua-planet configurations (Medeiros et al., 2016). The aqua-planet configurations (Neale & Hoskins, 2000) refer to an ocean-covered planet with no axial tilt—a planet in a perpetual equinox and devoid of continents. The aqua-planet compsets designate a fixed, zonally symmetric SST distribution, modeled after the present-day SST (Sea Surface Temperature) distribution on Earth (QOBS in Neale & Hoskins, 2000). The lack of a seasonal cycle allows one to compute robust statistics from a shorter simulation, and the absence of land removes any influence of discretized topography or interactions with the land model from the simulations. The aqua-planet configurations are therefore an indispensable tool for testing design choices in global atmospheric models.

To understand how the design choices adopted by CAM-SE influence the aqua-planet solutions compared to its predecessor, CAM-HOMME, we ran a pair of simulations using the CAM6 physics package for 4.5 years at 1° horizontal resolution ($N_e = 30$ and $N_p = 4$) and 32 levels in the vertical (model top is at approximately 3.6 Pa). Figure 6 shows the total kinetic energy spectrum at the 200-hPa level in the two simulations. Compared to CAM-HOMME, the slope of the kinetic energy spectrum in CAM-SE is shallower for wavenumbers larger than 30, bringing the solutions closer to the empirically (Nastrom & Gage, 1985) and theoretically (Charney, 1971) determined slope of -3 at synoptic scales. The increased kinetic energy at smaller scales is due to reducing the explicit diffusion coef-

ficients used in CAM-SE compared to CAM-HOMME rather than the reformulation of the dynamical core in dry-mass vertical coordinates or a more comprehensive treatment of moist energy. Reducing the viscosity coefficients without increasing noise in the solutions (especially in the presence of *real-world* topography) has been made possible by damping momentum and temperature on approximate pressure levels and only damping the deviation of pressure-level thickness from a smoother reference pressure-level thickness.

Figure 7a shows the zonally averaged total precipitation rate in CAM-HOMME (purple) and CAM-SE (red) averaged over the final 4 years of the simulations. The differences between the two simulations are provided as the purple curve of Figure 7b. CAM-SE has increased precipitation at the equator, with a peak difference of 3 mm/day. An approximate 1 mm/day reduction in precipitation occurs on the flanks of the equator (Figure 7). CAM-HOMME has twin Intertropical Convergence Zone (ITCZ) straddling the equator, which is a common feature of the tropics in general circulation models (Bellucci et al., 2010; Medeiros et al., 2016), but appears absent from the CAM-SE simulations (Figure 7). The total precipitation rate in the model is the sum of the precipitation from the convective parameterizations and that due to large-scale condensation. The deep convective precipitation rate indicates that the double ITCZ does exist in CAM-SE (not shown) and is masked by the large increase in total precipitation rate at the equator (Figure 7b) primarily due to an increase in large-scale condensation (not shown).

An analysis of the vertically integrated, zonally averaged dry static energy budget indicates that the latent heating due to the change in precipitation rate is balanced by the anomalous dry static energy flux convergence due to the increase in the mean resolved vertical upward motion at the equator (the dynamic component from equation (3) in Muller & O’Gorman, 2011; not shown). This balance also holds for the reduction in precipitation rate on the flanks of the equator (not shown). It is likely that an increase in resolved vertical upward motion in CAM-SE drives the increase in large-scale condensation rate observed in CAM-SE, consistent with a prior analysis of CAM-HOMME (O’Brien et al., 2016). In this scenario, greater downward motion observed on the flanks of the equator is simply compensating for the increased mass flux at the equator, which then acts to stabilize the column and reduce the precipitation rate locally.

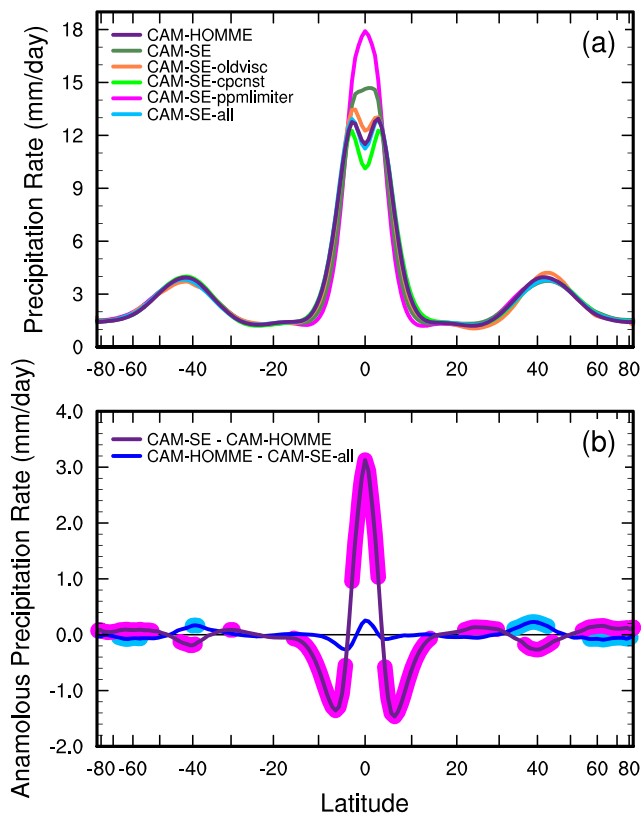


Figure 7. (a) The zonally averaged total precipitation rates in the aqua-planet simulations, averaged over the final 4 years of a 4.5-year simulation. Labels are defined in the text. (b) The change in the total precipitation rate between two simulations denoted by the label. The shading indicates where the differences are significant at the 95% confidence level. CAM = Community Atmosphere Model; HOMME = High-Order Methods Modeling environment; SE = spectral element.

The authors have identified three design aspects of CAM-SE that explain the large changes in precipitation observed in the tropics. These three design aspects are the use of a lower divergence damping coefficient, a thermodynamically consistent definition of c_p (see (38)), and the removal of the limiter in the vertical remapping of the horizontal winds. The individual effect of each of the three design choices is illustrated in Figure 7a, showing the total precipitation rate from three additional CAM-SE simulations, each simulation having one of the three modifications reverted back to the CAM-HOMME design. By using the larger divergence damping coefficients from CAM-HOMME in CAM-SE (*CAM-SE-oldvisc*), the precipitation rate at the equator is reduced by about 1 mm/day compared with CAM-SE. The CAM-SE simulation in which the limiter is turned on (*CAM-SE-ppmlimiter*), as it is in CAM-HOMME, results in a dramatic increase in equatorial precipitation of 3–4 mm/day compared with CAM-SE (Figure 7a). Through reverting the definition of c_p back to the CAM-HOMME definition ($c_p = c_p^{(d)}$; *CAM-SE-cpcnst*), the near-equatorial precipitation rates are dramatically reduced by 3–4 mm/day (Figure 7a). A fourth simulation was performed containing all three of the aforementioned modifications (*CAM-SE-all*). The simulated total precipitation rates in CAM-SE-all are indistinguishable from the CAM-HOMME simulation in the zonal mean (Figure 7b).

The influence of the new definition of c_p and the PPM limiter on tropical precipitation are not nearly as intuitive as the influence of the divergence damping coefficient. An increase in divergence damping reduces the overall magnitude of the vertical pressure velocities (Figure 8), likely explaining the reduction in mean precipitation in that simulation. It is plausible, however, that a less intuitive relationship between divergence damping and precipitation may emerge in a more realistic configuration, acknowledging that Zhao et al. (2012) have reported on a counterintuitive relationship to tropical storm genesis. In our aqua-planet simulations, the ITCZ, in part, consists of underresolved, deep convective towers simulated by the resolved dynamics, consistent with a prior study of CAM-HOMME (Herrington & Reed, 2017). The base of the convective towers often originates within the boundary layer, with convergent flow-driving resolved

upward mass fluxes through the cloud base. In the CAM-SE-ppmlimiter simulation, the magnitude of the horizontal velocities in the lowest model level of the equatorial convergence zone is larger (Figure 9b). The authors speculate that the limiter will select a larger magnitude horizontal wind in the lowest model level in a convergent flow regime, resulting in greater convergence, vertical motion, and therefore precipitation.

To explain how the limiter may influence mass convergence, consider a vertical profile of the horizontal wind, u_k , across the lowest model level, $k = nlev$, and the second lowest model level, $k = nlev - 1$ (Figure 9a). Assume that level $nlev$ is undergoing horizontal mass convergence, which over the course of a vertical remapping time step raises the top of the lowest model level. With the limiter off, the value of the wind at the new time step, u_{nlev} , is evaluated through integrating the subgrid PPM reconstruction (pink line in Figure 9a) from the surface to the top of the raised model level, such that the contribution of u_{nlev-1} to u_{nlev} is shown as the vertical green line in Figure 9a. The limiter, on the other hand, forces the contribution of u_{nlev-1} to u_{nlev} to be exactly equal to its previous value of u_{nlev-1} . The schematic wind profile in Figure 9a, indicating mass convergence in a region of positive du/dz , is characteristic of the equatorial convergence zones and, through turning the limiter on, increases the magnitude of the lowest level winds by systematically shifting the contribution of the raised level interface to the right in Figure 9a.

The influence of a thermodynamically consistent definition of c_p is to increase precipitation near the equator by 3–4 mm/day (Figure 7a). Note that the $c_p = c_p^{(d)}$ change affects not only the energy conversion term in the thermodynamic equation (equation (50)) but also the frictional heating term (equation (59)) and the vertical remapping of temperature (section 3.1.5). An additional aqua-planet simulation reveals that the frictional heating term has a nonnegligible contribution to the c_p sensitivity. A simulation in which $c_p = c_p^{(d)}$ only

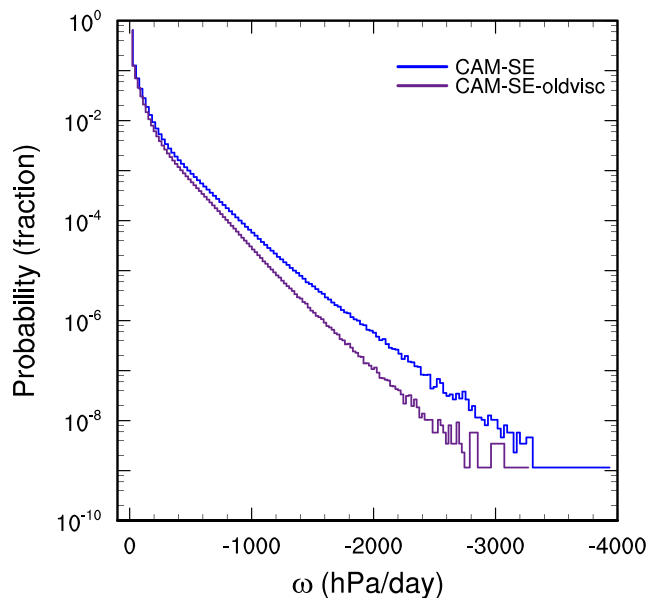


Figure 8. Probability density distribution of the upward vertical pressure velocities (ω), everywhere in the model for CAM-SE and CAM-SE-oldvisc. Figure computed from a single year of 6-hourly output. CAM-SE-oldvisc has a divergence damping coefficient 6.25 times that of CAM-SE, resulting in a reduction in the magnitude of ω . CAM = Community Atmosphere Model; SE = spectral element.

in the frictional heating term (equation (59)) results in a reduction in equatorial precipitation of 1–2 mm/day (not shown). This result is intuitive; since the term $(\mathbf{v} \cdot \delta \mathbf{v})$ is always negative, the heating rate is inversely proportional to c_p . In CAM-SE, c_p is always greater than $c_p^{(d)}$, and so there is more frictional heating in CAM-SE than in CAM-HOMME, all other aspects being equal. In the equatorial regions, additional heating likely facilitates buoyancy, which increases resolved vertical motion, leading to an increase in precipitation.

One of the motivations for moving to a dry-mass vertical coordinate is to make the coupling between the physics and dynamics more consistent. While the developments presented in this work conclude a significant portion of the effort required to achieve consistent coupling, at present, the CAM physics packages have not yet been made consistent with a dry-mass vertical coordinate. The authors speculate that some of the sensitivity to c_p shown in Figure 7a is related to the inconsistent definitions of internal energy between CAM-SE and the physics packages. An inconsistent energy definition between the dynamical core and the physics package can result in inconsistent forcing terms being passed back and forth. It was decided that this complication be left to a future sensitivity analysis, once the physics package has been made consistent with the dry-mass vertical coordinate and a comprehensive treatment of condensates and energy. But as a final point of caution, the authors recognize that while there is a surprising sensitivity of aqua-planet solutions to dynamical core design choices, this sensitivity may be different, or even nonexistent, in more realistic Earth-like configurations. These remaining uncertainties are left to be explored in future work.

4.2.1. Conservation

To assess the AAM conservation properties of the CAM-SE dynamical core, the diagnostics used in Lauritzen et al. (2014) are applied to the aqua-planet simulation described in section 4.2. In the dynamical core the column-integrated wind and mass AAM is written to history files and the total integrals of AAM are computed as a postprocessing step. The torques are obtained by subtracting the global integrals of AAM divided by the time-increment between the outputs of the AAM. For details on the discretized AAM diagnostics see Lauritzen et al. (2014). These diagnostic outputs are part of the CESM2 release and are controlled with namelist variables.

As discussed in section 2.7 the total AAM torque from the dynamical core, in the absence of topography, should be small (ideally zero) compared to the torque from the parameterization which is not spurious. CAM-SE uses the vector invariant form of the momentum equations, and the discretization thereof is not

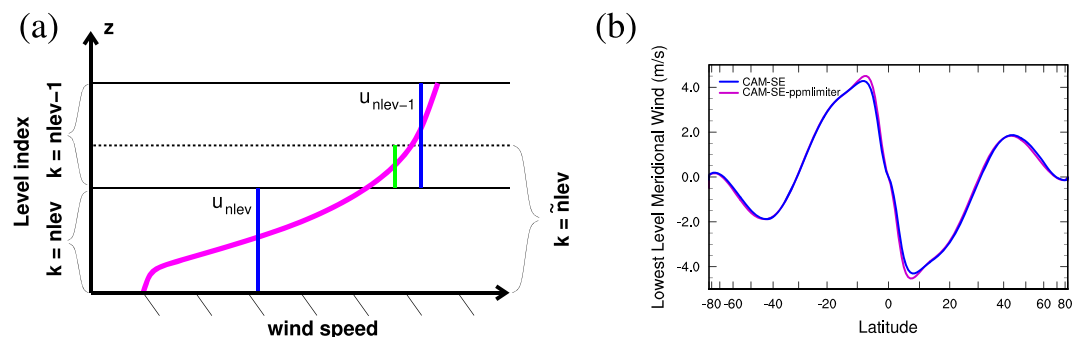


Figure 9. (a) A schematic illustration of the influence of the limiter on the vertical remapping of the winds of the lowest model level, in a convergent flow regime. Solid black lines show the layer interface, and the dashed line shows the top of the lowest model level in the following time step. The discretized model winds are denoted as blue bars, while the PPM reconstruction is shown as the pink curve. The green bar is the wind computed by integrating the reconstruction from the top of level $nlev$ to the top of level $nlev-1$. Notations are provided in the text. (b) Time-mean, zonally averaged lowest model level meridional wind in the CAM-SE and CAM-SE-ppmlimiter simulations, computed from the entirety of a 4.5 year simulation. CAM = Community Atmosphere Model; SE = spectral element.

Dynamical core Axial Angular Momentum (AAM) diagnostics for CAM-SE

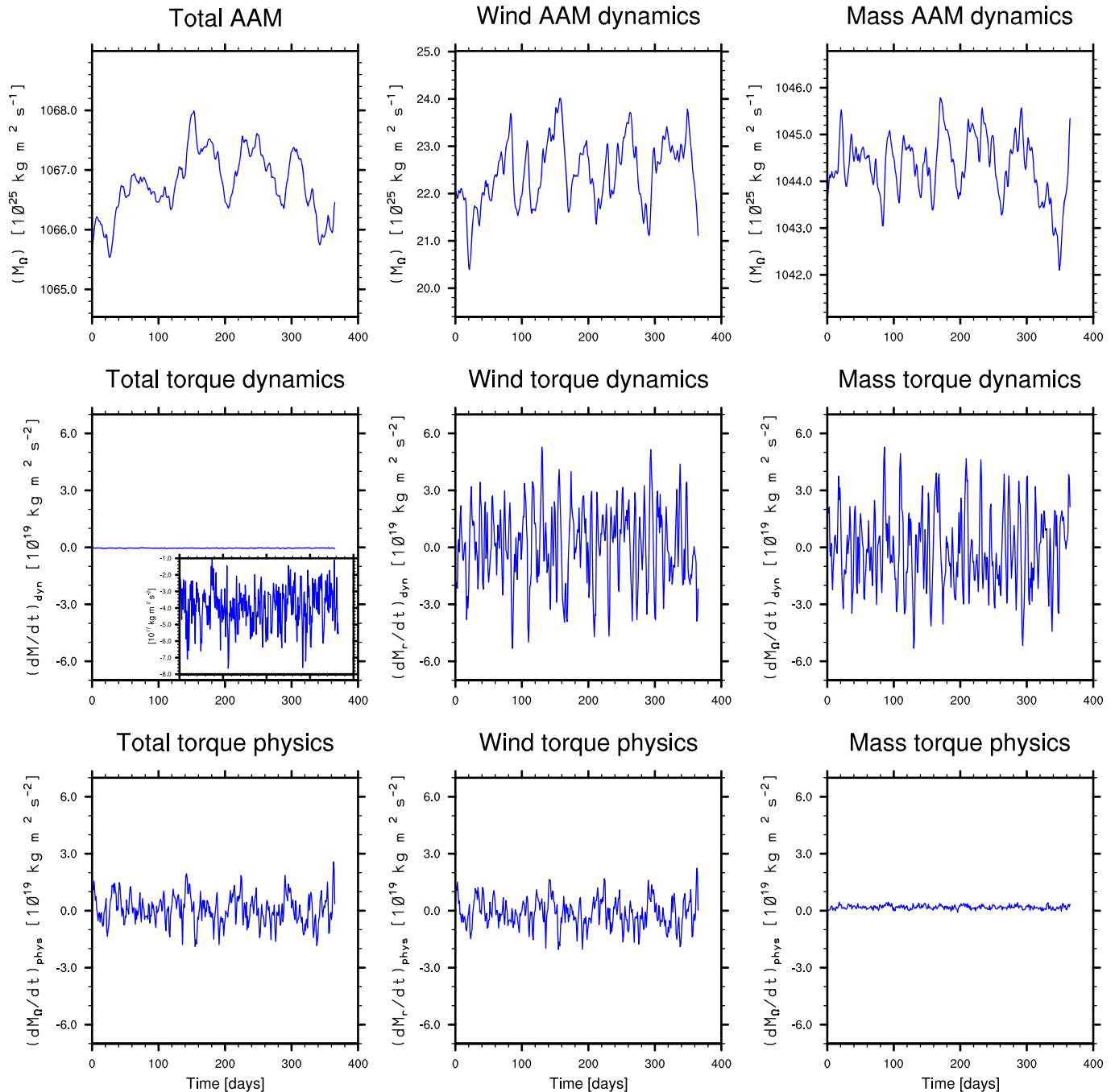


Figure 10. The first row shows total (column 1), wind (column 2), and mass (column 3) AAM as a function of time for the first year of the CAM6 aqua-planet simulation. The remaining plots show total, wind, and mass AAM torques as a function of time (columns 1, 2, and 3, respectively) for the dynamical core (row 2) and parameterizations (row 3). Note that it is necessary to enlarge the y axis of the dynamical core torque by a factor of approximately 100 to visualize the dynamical core torque. Hence, the spurious torque from the dynamical core is small compared to the physical torques from the parameterizations. CAM = Community Atmosphere Model; SE = spectral element.

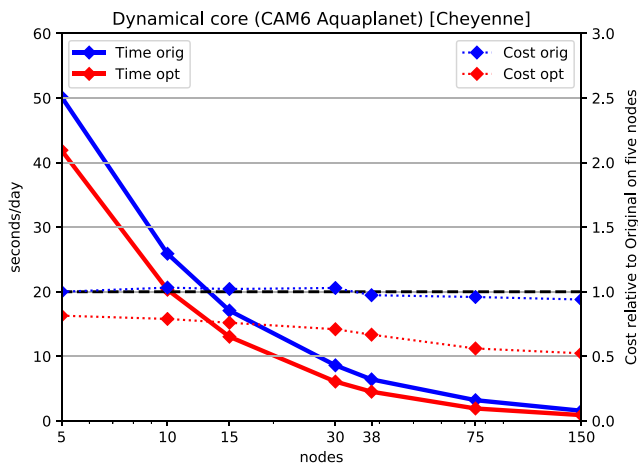


Figure 11. The execution time of the SE dynamical core in an aqua-planet configuration using CAM6 physics at 1° horizontal resolution on Cheyenne. Both the original (CAM-HOMME) and optimized code (CAM-SE) are indicated by the solid blue and red lines, while the dotted lines represent the computational cost. CAM = Community Atmosphere Model; HOMME = High-Order Methods Modeling environment; SE = spectral element.

core were also made which both reduce the computational cost at both modest and large processor count. In particular, new communication operators were developed which reduces the amount of data movement between MPI ranks as well as through the memory hierarchy. Derivative operators were optimized to increase code vectorization, and the limiter operator was rewritten to reduce cost. The CAM-HOMME configuration is referred to as the original codebase, and CAM-SE is referred to as the optimized codebase in this section. So when comparing computational cost of CAM-HOMME and CAM-SE, it reflects not only the code optimizations but also the numerous reformulations of the dynamical core in CAM-SE that require extra computational work (e.g., the heat capacity is a function of water loading tracers and therefore not constant; (moist) pressure is not a prognostic variable and must be diagnosed; the reference pressure computation is not in CAM-HOMME).

We provide the execution time and computational cost on Cheyenne for the dynamical core component for the 1° CAM6 aqua-planet configuration used in section 4.2 ($N_e = 30$, $N_p = 4$, $n_{lev} = 32$). CAM6 has 33 advected species. Cheyenne is an SGI ICE Cluster with 4,032 dual-socket Intel Xeon-based nodes with 36 cores/nodes.

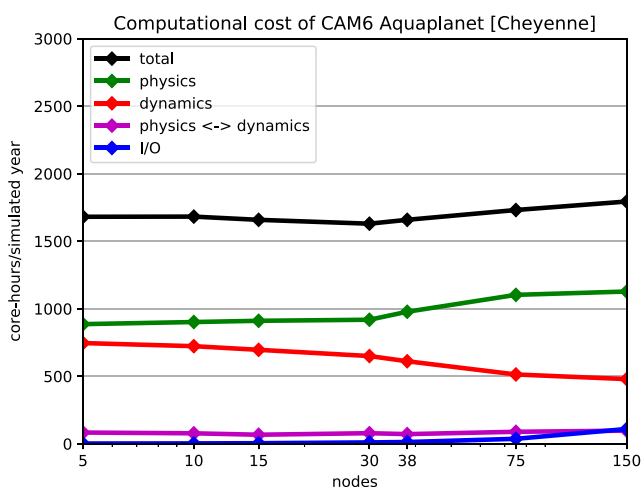


Figure 12. The cost in core hours per simulated year is provided for several different subcomponents of CAM for the 1° horizontal resolution aqua-planet simulation on Cheyenne (see text for more details). CAM = Community Atmosphere Model.

inherently AAM conserving. That said, Figure 10 shows that CAM-SE conserves AAM very well: The spurious torques from the dynamical core are approximately a factor of 100 smaller than the physical torques from the parameterizations. These results are very similar to the dynamical core torques found in the dry Held-Suarez setup used in Lauritzen et al. (2014).

Similarly to the AAM diagnostics, the total column-integrated moist energy is output at various places in the dynamical core and physics parameterizations. The dynamical core conserves the moist total energy, equation (63), to about 0.1 W/m^2 . The frictional heating term described in section 2.6.2 is approximately 0.4 W/m^2 and hence an important term for total moist energy conservation. As mentioned in section 2.7 the CAM physics energy fixer enforces a different energy than the comprehensive moist energy. This inconsistency should be removed in future CAM versions; however, this is not a trivial modification to the CAM physics package. The discrepancy between the two definitions of energy is approximately 0.5 W/m^2 (similar to what Taylor, 2011, found when just including the correct heat capacity for water vapor in the total energy equation). A detailed energy analysis will be the subject of a future publication.

4.3. Performance

In addition to the numerical changes described in the previous section, a number of changes to the computational structure of the SE dynamical

core were also made which both reduce the computational cost at both modest and large processor count. In particular, new communication operators were developed which reduces the amount of data movement between MPI ranks as well as through the memory hierarchy. Derivative operators were optimized to increase code vectorization, and the limiter operator was rewritten to reduce cost. The CAM-HOMME configuration is referred to as the original codebase, and CAM-SE is referred to as the optimized codebase in this section. So when comparing computational cost of CAM-HOMME and CAM-SE, it reflects not only the code optimizations but also the numerous reformulations of the dynamical core in CAM-SE that require extra computational work (e.g., the heat capacity is a function of water loading tracers and therefore not constant; (moist) pressure is not a prognostic variable and must be diagnosed; the reference pressure computation is not in CAM-HOMME).

We provide the execution time and computational cost on Cheyenne for the dynamical core component for the 1° CAM6 aqua-planet configuration used in section 4.2 ($N_e = 30$, $N_p = 4$, $n_{lev} = 32$). CAM6 has 33 advected species. Cheyenne is an SGI ICE Cluster with 4,032 dual-socket Intel Xeon-based nodes with 36 cores/nodes.

The x axis is the number of nodes, while the left y axis corresponds to execution time in seconds/day, and the right y axis corresponds to the relative computational cost normalized to the original code run on 5 nodes or 180 cores of Cheyenne. Execution times and computational costs are also provided for up to 150 nodes or 5,400 cores where a single spectral element in the horizontal is allocated to each core.

It is clear from Figure 11 that the execution time for the dynamical core for the optimized version (CAM-SE) is significantly less than the original codebase (CAM-HOMME) for all core counts. The reduction in execution time for the optimized versus the original code varies from approximately 20% at small core counts to slightly more than 50% at larger core counts. The percentage reduction in execution time for the optimized versus original code is readily apparent by looking at the relative computational cost curves indicated by the dotted lines in Figure 11. A value greater than 1 indicates that it is more expensive to run a particular configuration than the original code on five nodes, while a value less than 1 indicates that it is cheaper to run a particular configuration. This approach allows for the comparison of both the impact of the optimizations have on a particular node code as well as the impact of optimizations to code scalability. Interestingly, while the original nearly fixed cost regardless of core count,

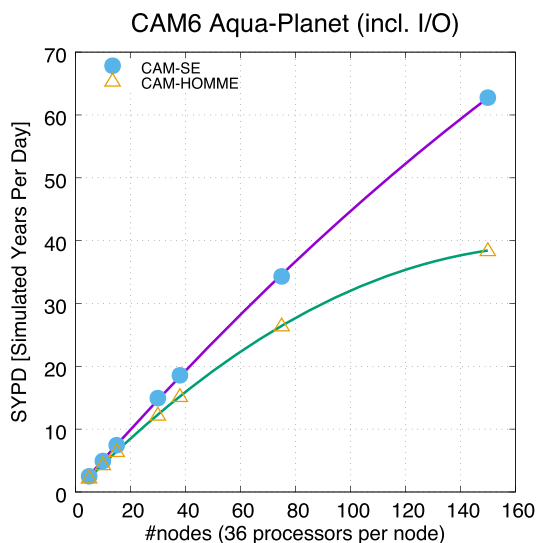


Figure 13. Throughput in terms of simulated years per day for CAM6 aqua-planet including I/O for CAM-SE and CAM-HOMME as a function of number of nodes. The curved line is a parabolic least squares fit to the data points. Note that for the rightmost data point, there is only one element in the horizontal per processor (150 nodes is 5,400 processors and there are $6 \times N_e^2 = 6 \times 30^2 = 5,400$ elements in the horizontal). CAM = Community Atmosphere Model; SE = spectral element; HOMME = High-Order Methods Modeling environment; SYPD = simulated years per day.

the optimized code actually becomes cheaper. In particular, the greatest reduction of 40% in computational cost occurs on 150 nodes. We suspect that the decrease in computational cost illustrated in Figure 11 is likely due to the fact that the calculations performed in the dynamical core (in particular tracer advection code) now fit into the Level 3 (L3) cache on 75 nodes. The decreased execution time due to the calculations being L3 cache resident is sufficiently large as to overcome any increase in execution time increase due to message passing.

The cost of the dynamical core for the optimized code base that is compared to other pieces of the CAM is illustrated in Figure 12. We categorize four different pieces of CAM for timing purposes: physics, dynamics, I/O, and the remapping of data structures necessary between the physics and dynamics. The fraction of time that CAM spends performing the dynamics drops from a maximum of 44% on five nodes, to a minimum of 27% on 150 nodes. While there is an increase in the relative cost for both the I/O and physics to dynamics interface, the largest relative increase in cost is seen in the physics calculations after 30 nodes. We suspect that this very modest increase in the cost of the physics is a result of a minor unavoidable load imbalance of various physics processes. Figure 12 illustrates that for CAM-SE the number of core hours per simulated year is quite constant and only varies from a low of 1,630 to 1,794 over a very large core count range.

Note that for this comparison we use the same aqua-planet CAM6 configuration, which achieves simulation rates that range from a low of 2.5 simulated years per day (SYPD) on 5 nodes to 62.8 SYPD on 150 nodes and

run for 1 month (see Figure 13). This length of simulation includes the default monthly history I/O output as well as the generation of a restart file. A rule of thumb for CESM is that the model should produce a minimum of 20 SYPD to be fast enough for doing *standard* climate science with the model. CAM-SE achieves this threshold with fewer than 2,000 processors. At this processor count CAM-FV exhibits similar throughput (not shown) and with larger processor counts CAM-SE continues to scale well.

5. Conclusions

The NCAR release of CAM-SE in CESM2.0 is presented. This version uses a dry-mass vertical coordinate and has a comprehensive treatment of condensates both in the thermodynamic equation and momentum equations. The conservation properties in terms of total energy and AAM for this system of equations were derived. The discretization of the equations of motion is explained in detail and is intended to serve as a comprehensive documentation for CAM-SE. Idealized simulations demonstrate the accuracy and conservation property of the numerical model. In particular, we show that the reduction in viscosity parameters (made possible by changes to the details of how hyperviscosity is applied) greatly improved the total kinetic energy spectrum of CAM-SE and that the comprehensive treatment of moist thermodynamics and condensate loading significantly changes precipitation rates in aqua-planet simulations. These changes are not due to the dry-mass vertical coordinate per se, but we argue that the comprehensive treatment of energy in the Earth system model as a whole is much simpler in a dry-mass vertical coordinate (e.g., a dry-mass vertical coordinate does not change during parameterization updates). Last but not least the CAM-SE model has been sped up significantly (from 20% to 50% depending on core count) compared to its predecessor CAM-HOMME.

Appendix A: Dissipation

A.1. Discretization

In the CAM-SE model explicit fourth-order hyperviscosity ($\nu \nabla^4 \phi$) is used as the main stabilization mechanism. For scalar fields such as T , p , q , etc., a scalar viscosity is applied, while for the horizontal momentum equations vector viscosity ($\nu \nabla^4 \mathbf{v}$) is employed. The coefficient of viscosity ν may be constant or a spatially varying quantity depending on the application. High-order viscosity operator $\nu \nabla^{2m} \phi$, $m = 2, 3, \dots$, for an arbitrary variable ϕ can be constructed by successively applying the basic Laplacian-type viscosity operator $\nabla^2(\cdot)$. In order to

describe the SE discretization, we consider the basic Laplacian without the coefficient of viscosity ν

$$L(\phi) = \nabla^2 \phi. \quad (\text{A1})$$

Integrating (A1) over an element Ω_e with boundary Γ_e using Greens method results in

$$\int_{\Omega_e} L \psi \, d\Omega_e = \int_{\Gamma_e} \phi \nabla \psi \, d\Gamma_e - \int_{\Omega_e} \nabla \psi \cdot \nabla \phi \, d\Omega_e. \quad (\text{A2})$$

For the continuous Galerkin (SE) method the boundary integral vanishes ($\psi = 0$ at the element boundaries) and the *rhs* simplifies to a surface integral. The tensor gradients in the integrand can be expressed in terms of its contravariant components ($\tilde{F}^1, \tilde{F}^2$), using (75) such that

$$\nabla \psi \cdot \nabla \phi = \tilde{F}^1 \frac{\partial \psi}{\partial x^1} + \tilde{F}^2 \frac{\partial \psi}{\partial x^2}, \quad \begin{bmatrix} \tilde{F}^1 \\ \tilde{F}^2 \end{bmatrix} = \mathbf{D}^{-1} \mathbf{D}^{-T} \begin{bmatrix} \partial \phi / \partial x^1 \\ \partial \phi / \partial x^2 \end{bmatrix}. \quad (\text{A3})$$

Thus, the discretization of the Laplacian for SE method can be obtained by simplifying the integral

$$\int_{\Omega_e} L \psi \, d\Omega_e = - \int_{\Omega_e} \nabla \psi \cdot \nabla \phi \, d\Omega_e = - \int_{\Omega_e} \left[\tilde{F}^1 \frac{\partial \psi}{\partial x^1} + \tilde{F}^2 \frac{\partial \psi}{\partial x^2} \right] d\Omega_e. \quad (\text{A4})$$

As in the case of (83), the weak formulation (A4) can be evaluated on the standard element using the polynomial approximations (82) for ϕ, ψ on Ω_e and GLL quadrature rule. Further simplification of (A4) leads to

$$L_{kl}^e = -(\tilde{M}_{kl}^e)^{-1} \left[\sum_{l=0}^N J_e^{(1)}(i, l) \tilde{F}_{il}^1 D_{ik}^{(1)} w_l w_i + \sum_{l=0}^N J_e^{(2)}(k, l) \tilde{F}_{kl}^2 D_{li}^{(2)} w_k w_l \right], \quad (\text{A5})$$

where L_{kl}^e is the value of Laplacian term for a grid point (k, l) on Ω_e .

Note that in practice, the contravariant gradient terms in (A3) are first computed using collocation differentiation then the weak divergence of the gradients is computed; this is followed by a DSS operation which yields the discrete Laplacian ($L(\phi^e) \Rightarrow \text{div}(\text{grad}(\phi^e))$).

In CAM-SE the vector viscosity is handled using the vector Laplacian $\nu \nabla^2 \mathbf{v}$, which is consistent with the curvilinear formulation of the momentum equation. By using the vector identity $\nabla^2 \mathbf{v} = \nabla(\nabla \cdot \mathbf{v}) - \nabla \times (\nabla \times \mathbf{v})$, the coefficient of viscosity ν may be split into the viscosity corresponding to the divergence part ν_{div} and that for the vorticity part ν_{vor} such that

$$\nu \nabla^2 \mathbf{v} = \nu_{\text{div}} \nabla(\nabla \cdot \mathbf{v}) - \nu_{\text{vor}} \nabla \times (\nabla \times \mathbf{v}). \quad (\text{A6})$$

The SE discretization of (A6) is handled separately for each component using a vector test function $\vec{\psi}$ and the weak formulation

$$\nu_{\text{div}} \int_{\Omega_e} \vec{\psi} \cdot \nabla(\nabla \cdot \mathbf{v}) \, d\Omega_e - \nu_{\text{div}} \int_{\Omega_e} (\nabla \cdot \vec{\psi})(\nabla \cdot \mathbf{v}) \, d\Omega_e, \quad (\text{A7})$$

$$- \nu_{\text{vor}} \int_{\Omega_e} \vec{\psi} \cdot \nabla \times (\nabla \times \mathbf{v}) \, d\Omega_e = - \nu_{\text{vor}} \int_{\Omega_e} (\nabla \times \vec{\psi}) \cdot (\nabla \times \mathbf{v}) \, d\Omega_e, \quad (\text{A8})$$

where the line integrals along the boundaries associated with weak formulation vanish for SE discretization. The *rhs* of the above equations are converted into equivalent tensor form for the divergence and curl terms and discretized as in the case of (A5).

A.2. Reference Pressure for Dry-Mass Layer Thickness Damping

Applying ∇^4 damping to $\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right)$ values directly leads to nonphysical model behavior in that physically meaningful gradients in level thicknesses along model surfaces are suppressed, particularly in the vicinity of topography. As an alternative to computing and applying the necessary correction terms to the damping operator, we construct smooth reference values which convey these gradients and apply the damping only to the deviations from these values. Since they are immune from damping processes and can have a significant effect on model behavior, these values must be as accurate as possible and should not impose any predetermined structure. The area weighted averages for each element are least affected by small-scale errors and thereby contain the most accurate information available. However, directly computed element averages

of $\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}}\right)$ cannot be used to determine the desired reference values. Averages of this variable or any other variable that cannot be adequately represented as a local linear function of height will contain a systematic bias in the vicinity of topography. To avoid this bias, which varies with $|\nabla \Phi|$, the reference values must be constructed from other available information. In the following, let $\langle \cdot \rangle$ denote element average values or derived values consistent with computed element averages, and let $M^{(d)}$ denote the weight of dry air per unit area.

To obtain $\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}}\right)^{(\text{ref})}$, we begin by smoothly interpolating $\langle \Phi \rangle$, $\langle \Phi_s \rangle$ and $\langle T \rangle$ values onto the model grid points. From these, the corresponding average pressure-related values are constructed in two steps using the hydrostatic equation for dry air (46). First, define the function

$$\mu_k \equiv \frac{1}{\langle M^{(d)} \rangle} \left(\frac{\partial \langle M^{(d)} \rangle}{\partial \eta^{(d)}} \right)_k \quad (\text{A9})$$

and iterate from the surface upward to determine its values for each model layer

$$\mu_k = \frac{2 \left(\langle \Phi \rangle_k - \langle \Phi \rangle_{k+1/2} \right)}{R^{(d)} \langle T \rangle_k}, \quad (\text{A10})$$

where $\langle \Phi \rangle_{nlev+1/2} = \langle \Phi_s \rangle$. Then iterate downward from the model top to determine the pressure values,

$$\left(\frac{\partial \langle M^{(d)} \rangle}{\partial \eta^{(d)}} \right)_k = \langle M^{(d)} \rangle_{k-1/2} \left(\frac{2\mu_k}{2 - \mu_k} \right), \quad (\text{A11})$$

$$\langle M^{(d)} \rangle_k = \langle M^{(d)} \rangle_{k-1/2} \left(\frac{2}{2 - \mu_k} \right), \quad (\text{A12})$$

$$\langle M^{(d)} \rangle_{k+1/2} = \langle M^{(d)} \rangle_{k-1/2} \left(\frac{2 + \mu_k}{2 - \mu_k} \right), \quad (\text{A13})$$

where $\langle M^{(d)} \rangle_{1/2} = M_t^{(d)}$ and $\langle M^{(d)} \rangle_{nlev+1/2} = \langle M_s^{(d)} \rangle$. At this point, we have a set of self-consistent element average values which satisfy the hydrostatic equation

$$\frac{\partial \langle \Phi \rangle}{\partial \eta^{(d)}} = - \frac{R^{(d)} \langle T \rangle}{\langle M^{(d)} \rangle} \left(\frac{\partial \langle M^{(d)} \rangle}{\partial \eta^{(d)}} \right). \quad (\text{A14})$$

These averages are not suitable for direct use as reference surface values because at the lower boundary they are consistent with the element average of surface geopotential $\langle \Phi_s \rangle$, not with the actual Φ_s values. To obtain the appropriate reference values corresponding to Φ_s , first adjust the surface pressures

$$(M_s^{(d)})^{(\text{ref})} = \langle M_s^{(d)} \rangle \left[1 + \frac{1}{\langle T \rangle} \frac{\partial \langle T \rangle}{\partial \langle \Phi \rangle} (\Phi_s - \langle \Phi_s \rangle) \right]^{-\frac{1}{R^{(d)} \frac{\partial \langle T \rangle}{\partial \langle \Phi \rangle}}} \text{ for } \left(\frac{\partial \langle T \rangle}{\partial \langle \Phi \rangle} \right)_{nlev} \neq 0, \quad (\text{A15})$$

and

$$(M_s^{(d)})^{(\text{ref})} = \langle M_s^{(d)} \rangle e^{-\left(\frac{\Phi_s - \langle \Phi_s \rangle}{R^{(d)} \langle T \rangle} \right)} \text{ for } \left(\frac{\partial \langle T \rangle}{\partial \langle \Phi \rangle} \right)_{nlev} = 0. \quad (\text{A16})$$

Then use the resulting change in weight of dry air per unit area at the surface to determine the reference values for dry-mass thicknesses

$$\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right)_k^{(\text{ref})} = \left(\frac{\partial \langle M^{(d)} \rangle}{\partial \eta^{(d)}} \right)_k + \left(\frac{\partial B}{\partial \eta^{(d)}} \right)_k \left[(M_s^{(d)})^{(\text{ref})} - \langle M_s^{(d)} \rangle \right], \quad (\text{A17})$$

where B is the hybrid coefficient.

A.3. Hyperviscosity Coefficients

The following hyperviscosity coefficients are used in CAM-SE:

$$\nu_T = \nu_{\text{vor}} = 0.150 \times \left(\frac{30}{N_e} 1.1 \times 10^5 \right)^3 \frac{m^4}{s}, \quad (\text{A18})$$

$$\nu_q = \nu_p = \nu_{\text{div}} = 0.751 \times \left(\frac{30}{N_e} 1.1 \times 10^5 \right)^3 \frac{m^4}{s}, \quad (\text{A19})$$

where $N_e = 30$ and $N_e = 120$ for the 1° and $1/4^\circ$ horizontal resolution configurations. Note that mass-wind consistency may be violated if $v_p \neq v_q$ where v_q is the viscosity coefficient for tracers. The term inside the parenthesis is the average grid spacing in kilometers, and the scaling with resolution is what is used with MPAS (Model for Prediction Across Scales; Skamarock et al., 2014). The damping of temperature and vorticity in CAM-SE is similar to the damping in MPAS (i.e., MPAS uses a coefficient of 0.05 in front of the $(\cdot)^3$ term and CAM-SE uses 0.751) except that MPAS also applies second-order Smagorinsky diffusion.

In the top three levels (sponge layer) the prognostic variables are damped with a Laplacian damping (second-order viscosity)

$$v_T = v_{vor} = v_{div} = v_p = \begin{cases} 4 \times 2.5 \frac{m^2}{s}, & \text{for } k = 1, \\ 2 \times 2.5 \frac{m^2}{s}, & \text{for } k = 2, \\ 1 \times 2.5 \frac{m^2}{s}, & \text{for } k = 3, \end{cases} \quad (\text{A20})$$

for resolutions up to (and including) $N_e = 120$.

Appendix B: Global Conservation

Below it is shown that the continuous equations of motion, that include the effect of condensates, conserve AAM and total energy. For the derivations first note that adding the continuity equations $\ell \in \mathcal{L}_{\text{all}}$ (51), using hydrostatic balance (52) and ideal gas law (17), yields a moist air continuity

$$\frac{\partial}{\partial t} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \right] + \nabla_{\eta^{(d)}} \cdot \left[\mathbf{v} \rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \right] = 0, \quad (\text{B1})$$

or, equivalently, in Lagrangian form

$$\frac{d}{dT} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \delta A \right] = 0, \quad (\text{B2})$$

where δA is a horizontal area of a Lagrangian air parcel so that $\nabla_{\eta^{(d)}} \cdot \mathbf{v} = \frac{1}{\delta A} \frac{d}{dT} (\delta A)$, and $\frac{d}{dT} = \frac{\partial}{\partial t} + \mathbf{v} \cdot \nabla_{\eta^{(d)}}$ is the material/total derivative.

B.1. Axial Angular Momentum (AAM)

The conservation law for angular momentum is derived in spherical coordinates for which the zonal momentum equation takes the form

$$\frac{du}{dT} = \frac{uv \tan \varphi}{r} + 2\Omega v \sin \varphi - \frac{1}{\rho r \cos \varphi} \frac{\partial p}{\partial \lambda}, \quad (\text{B3})$$

where u and v are the zonal and meridional velocity components, respectively, φ is latitude and λ longitude, Ω rotation rate of Earth ($f = 2\Omega \sin \varphi$), and r is the mean radius of Earth.

The conservation law for AAM, $\mathcal{M} = (u + \Omega r \cos \varphi) r \cos \varphi$, can be conveniently derived using the Lagrangian form. Consider the material derivative of $\mathcal{M}$ multiplied by the volume of a Lagrangian air parcel, $\rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \delta A$

$$\frac{d}{dT} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \delta A (u + \Omega r \cos \varphi) r \cos \varphi \right]. \quad (\text{B4})$$

Using the chain rule, continuity equation on the form (B2), the equality $v = r \frac{d\varphi}{dT}$ and substituting (B3), one can obtain an evolution equation for AAM

$$\frac{d}{dT} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \delta A \mathcal{M} \right] = - \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \frac{\partial p}{\partial \lambda}, \quad (\text{B5})$$

or, equivalently, in Eulerian form

$$\frac{\partial}{\partial t} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \mathcal{M} \right] + \nabla_{\eta^{(d)}} \cdot \left[\mathbf{v} \rho \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \mathcal{M} \right] = - \left(\frac{\partial z}{\partial \eta^{(d)}} \right) \frac{\partial p}{\partial \lambda}. \quad (\text{B6})$$

Repeatedly using the chain rule, one can show that the right-hand side of (B5) and (B6) can be written as

$$-\left(\frac{\partial z}{\partial \eta^{(d)}}\right) \frac{\partial p}{\partial \lambda} = -\frac{\partial}{\partial \lambda} \left[\left(\frac{\partial z}{\partial \eta^{(d)}}\right) p \right] + p \frac{\partial}{\partial \lambda} \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \quad (\text{B7})$$

$$= -\frac{\partial}{\partial \eta^{(d)}} \left(\frac{\partial z}{\partial \lambda} p\right) + \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \frac{\partial p}{\partial \lambda} + p \frac{\partial}{\partial \lambda} \left(\frac{\partial z}{\partial \eta^{(d)}}\right), \quad (\text{B8})$$

$$= -\frac{\partial}{\partial \eta^{(d)}} \left(\frac{\partial z}{\partial \lambda} p\right) + \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \frac{\partial p}{\partial \lambda} - \frac{\partial}{\partial \lambda} \left[p \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \right] - \frac{\partial p}{\partial \lambda} \left(\frac{\partial z}{\partial \eta^{(d)}}\right). \quad (\text{B9})$$

$$= -\frac{\partial}{\partial \eta^{(d)}} \left(\frac{\partial z}{\partial \lambda} p\right) - \frac{\partial}{\partial \lambda} \left[p \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \right]. \quad (\text{B10})$$

Substituting (B10) on the right-hand side of (B6) and integrating (B6) in the horizontal and vertical result in the final AAM equation (61) in the main text.

B.2. Total Energy

The derivation of the total energy equation closely follows Kasahara (1974) but for a dry mass vertical coordinate and inclusion of condensates in the equations of motion. The equation for the horizontal kinetic energy per unit mass, $K = \frac{1}{2} \mathbf{v} \cdot \mathbf{v}$, is derived by multiplying the momentum equation (49) with $\rho \mathbf{v} \left(\frac{\partial z}{\partial \eta^{(d)}}\right)$ and the continuity equation (B1) with K , adding the two resulting equations and simplifying using the chain rule yields

$$\frac{\partial}{\partial t} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) K \right] + \nabla_{\eta^{(d)}} \cdot \left[\rho \mathbf{v} \left(\frac{\partial z}{\partial \eta^{(d)}}\right) K \right] = - \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \mathbf{v} \nabla_{\eta^{(d)}} p - g \rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \mathbf{v} \cdot \nabla_{\eta^{(d)}} \mathbf{z}. \quad (\text{B11})$$

For the derivation of a flux-form version of the total energy equation (that includes a geopotential term) we substitute

$$g \rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \mathbf{v} \cdot \nabla_{\eta^{(d)}} \mathbf{z} = \nabla_{\eta^{(d)}} \cdot \left[g z \rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \right] - g z \nabla_{\eta^{(d)}} \cdot \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \right], \quad (\text{B12})$$

$$= \nabla_{\eta^{(d)}} \cdot \left[g z \rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \right] - z \nabla_{\eta^{(d)}} \cdot \left[\rho \left(\frac{\partial p}{\partial \eta^{(d)}}\right) \right], \quad (\text{B13})$$

(where the hydrostatic balance equation (52) has been used on the right-hand side of (B13) on the right-hand side of (B11) and rearrange terms

$$\frac{\partial}{\partial t} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) K \right] + \nabla_{\eta^{(d)}} \cdot \left[\rho \mathbf{v} \left(\frac{\partial z}{\partial \eta^{(d)}}\right) (K + g z) \right] = - \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \mathbf{v} \nabla_{\eta^{(d)}} p - z \nabla_{\eta^{(d)}} \cdot \left[\rho \left(\frac{\partial p}{\partial \eta^{(d)}}\right) \right]. \quad (\text{B14})$$

Now, by multiplying the thermodynamic equation (50) with $\rho \frac{\partial z}{\partial \eta^{(d)}} c_p$ and the continuity equation (B1) with $c_p T$, adding the resulting equations and simplifying using the chain rule yields

$$\frac{\partial}{\partial t} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) c_p T \right] + \nabla_{\eta^{(d)}} \cdot \left[\rho \mathbf{v} \left(\frac{\partial z}{\partial \eta^{(d)}}\right) c_p T \right] = \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \omega. \quad (\text{B15})$$

By using that $\omega \equiv \frac{\partial p}{\partial t} + \mathbf{v} \cdot \nabla_{\eta^{(d)}} p$, using the chain rule and the continuity equation (B1), one can show that

$$\left(\frac{\partial z}{\partial \eta^{(d)}}\right) \omega = \frac{\partial}{\partial \eta^{(d)}} \left(z \frac{\partial p}{\partial t} \right) + \left(\frac{\partial z}{\partial \eta^{(d)}}\right) \mathbf{v} \nabla_{\eta^{(d)}} p + z \nabla_{\eta^{(d)}} \cdot \left[\rho \left(\frac{\partial p}{\partial \eta^{(d)}}\right) \right]. \quad (\text{B16})$$

Substituting (B16) on the right-hand side of (B15) and adding the resulting equation to (B14), the two terms on the right-hand side of (B14) cancel the two last terms on the right-hand side of (B16) and we get the total energy equation

$$\frac{\partial}{\partial t} \left[\rho \left(\frac{\partial z}{\partial \eta^{(d)}}\right) (K + c_p T) \right] + \nabla_{\eta^{(d)}} \cdot \left[\rho \mathbf{v} \left(\frac{\partial z}{\partial \eta^{(d)}}\right) (K + g z + c_p T) \right] = \frac{\partial}{\partial \eta^{(d)}} \left(z \frac{\partial p}{\partial t} \right). \quad (\text{B17})$$

This equation is in the same form as (5.7) in Kasahara (1974), but ρ , c_p , and p include the effect of condensates.

For the global integral of the total energy equation in a dry-mass vertical coordinate, we substitute the hydrostatic relation on the form (52) into (B17), integrate in the vertical, and use that the pressure at the model top is constant,

$$\begin{aligned} \frac{\partial}{\partial t} \int_{\eta=0}^{\eta=1} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right) \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) (K + c_p T) d\eta^{(d)} \\ + \nabla_{\eta^{(d)}} \cdot \int_{\eta=0}^{\eta=1} \mathbf{v} \left(\sum_{\ell \in \mathcal{L}_{\text{all}}} m^{(\ell)} \right) \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) (K + gz + c_p T) d\eta^{(d)} = -z_s \frac{\partial p_s}{\partial t}, \end{aligned} \quad (\text{B18})$$

which can also be written as

$$\begin{aligned} \frac{\partial}{\partial t} \int_{\eta=0}^{\eta=1} \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) \sum_{\ell \in \mathcal{L}_{\text{all}}} \left[m^{(\ell)} (K + c_p^{(\ell)} T + \Phi_s) \right] d\eta^{(d)} \\ + \nabla_{\eta^{(d)}} \cdot \int_{\eta=0}^{\eta=1} \mathbf{v} \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) \sum_{\ell \in \mathcal{L}_{\text{all}}} \left[m^{(\ell)} (K + c_p^{(\ell)} T + gz) \right] d\eta^{(d)} = 0 \end{aligned} \quad (\text{B19})$$

by expanding c_p using (38), rearranging terms, and using that $\Phi_s = gz_s$ is time independent and that

$$p_s = g \int_{\eta=0}^{\eta=1} \sum_{\ell \in \mathcal{L}_{\text{all}}} \left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) m^{(\ell)} d\eta^{(\eta)}. \quad (\text{B20})$$

Note that the energy terms (inside square brackets) in (B19) separate into all the components of moist air

$$\left(\frac{\partial M^{(d)}}{\partial \eta^{(d)}} \right) \sum_{\ell \in \mathcal{L}_{\text{all}}} \left[m^{(\ell)} (K + c_p^{(\ell)} T + \Phi_s) \right]. \quad (\text{B21})$$

Similarly for the flux term in the second square brackets on the left-hand side of (B19), integrating (B19) over the entire sphere results in (63) given in the main text.

Appendix C: The Kinnmark-Gray Five-Stage Third-Order RK Scheme

The thermodynamic equation, momentum equations, and dry air mass continuity equation in CAM-SE are evolved using the five-stage third-order RK scheme described in Guerra and Ullrich (2016; see their equation (56)). The stability of this class of time-stepping schemes is discussed in Dubos et al., 2015 (2015; see their section 3.4). For a given initial state vector at time step n , $\vec{\Lambda}^{(0)} = \vec{\Lambda}^n$, the updated state vector $\vec{\Lambda}^{(5)} = \vec{\Lambda}^{n+1}$ is computed as follows:

$$\begin{aligned} \vec{\Lambda}^{(1)} &= \vec{\Lambda}^{(0)} + \frac{\Delta t}{5} \vec{Y}(\vec{\Lambda}^{(0)}), \\ \vec{\Lambda}^{(2)} &= \vec{\Lambda}^{(0)} + \frac{\Delta t}{5} \vec{Y}(\vec{\Lambda}^{(1)}), \\ \vec{\Lambda}^{(3)} &= \vec{\Lambda}^{(0)} + \frac{\Delta t}{3} \vec{Y}(\vec{\Lambda}^{(2)}), \\ \vec{\Lambda}^{(4)} &= \vec{\Lambda}^{(0)} + \frac{2\Delta t}{3} \vec{Y}(\vec{\Lambda}^{(3)}), \\ \vec{\Lambda}^{(5)} &= -\frac{1}{4} \vec{\Lambda}^{(0)} + \frac{5}{4} \vec{\Lambda}^{(1)} + \frac{3\Delta t}{4} \vec{Y}(\vec{\Lambda}^{(4)}), \end{aligned} \quad (\text{C1})$$

where $\vec{Y}(\vec{\Lambda})$ denotes the discrete right-hand side terms of the equations of motion. The resulting method is linearly and nonlinearly third-order accurate. The scheme possesses a stability region which is provably optimal in terms of its extent along the imaginary axis among all five-stage third-order RK schemes, $[-i\sqrt{15}, i\sqrt{15}]$. Since the largest eigenvalue of the 1-D fourth-order spectral element spatial discretization ($N_p = 4$) is $i\sqrt{10/3}$, the resulting scheme satisfies a Courant-Friedrichs-Lewy condition given by

$$\frac{c\Delta t}{\Delta x} \leq \frac{3}{\sqrt{2}}, \quad (1\text{-D condition}), \quad (\text{C2})$$

where Δx denotes the average distance between degrees of freedom (equal to $3 \times \Delta x_e$, the width of a spectral element), c is the gravity wave speed, and Δt is the time step size. As dimension splitting is not employed in CAM-SE, in 2-D, and on a uniformly spaced grid this condition is restricted by a further factor of $1/\sqrt{2}$ to

$$\frac{c\Delta t}{\Delta x} \leq \frac{3}{2}, \quad (2\text{-D condition}). \quad (\text{C3})$$

Appendix D: Analytical Initial Condition Functions

In this section the analytical expressions for the moist baroclinic wave are given. The moist surface pressure is constant $p_s = 1,000$ hPa, meridional wind component is 0, $v(\varphi, z) = 0$ m/s, and the surface geopotential is 0, $\Phi_s(\varphi, z) = 0$ m²/s². The reference virtual temperature is given by

$$T_v(\varphi, z) = \left\{ F_1(z) - F_2(z) \left[(\cos \varphi)^{\mathcal{K}} - \frac{\mathcal{K}}{\mathcal{K}+2} (\cos \varphi)^{\mathcal{K}+2} \right] \right\}^{-1}, \quad (D1)$$

where

$$F_1(z) = \frac{1}{T_0} \exp\left(\frac{\Gamma z}{T_0}\right) + \left(\frac{T_0 - T_p}{T_0 T_p}\right) \left[1 - 2 \left(\frac{zg}{bR^{(d)}T_0} \right)^2 \right] \exp\left[- \left(\frac{zg}{bR^{(d)}T_0} \right)^2 \right], \quad (D2)$$

$$F_2(z) = \frac{(\mathcal{K}+2)}{2} \left(\frac{T_E - T_p}{T_E T_p} \right) \left[1 - 2 \left(\frac{zg}{bR^{(d)}T_0} \right)^2 \right] \exp\left[- \left(\frac{zg}{bR^{(d)}T_0} \right)^2 \right], \quad (D3)$$

with $T_0 = \frac{1}{2}(T_E + T_p)$. Parameter $T_E = 310$ K is the temperature at the equatorial surface, $T_p = 240$ K is the polar surface temperature, $\mathcal{K} = 3$ is the jet width parameter, $b = 2$ is the jet half-width parameter, and $\Gamma = 0.005$ K/m is the lapse rate.

To maintain hydrostatic balance, the pressure is given by

$$p(\varphi, z) = p_0 \exp\left[- \frac{g}{R^{(d)}} (F_3(z) - F_4(z) I_T(\varphi)) \right], \quad (D4)$$

where

$$F_3(z) = \frac{1}{\Gamma} \left[\exp\left(\frac{\Gamma z}{T_0}\right) - 1 \right] + z \left(\frac{T_0 - T_p}{T_0 T_p} \right) \exp\left[- \left(\frac{zg}{bR^{(d)}T_0} \right)^2 \right], \quad (D5)$$

$$F_4(z) = \frac{(\mathcal{K}+2)}{2} \left(\frac{T_E - T_p}{T_E T_p} \right) z \exp\left[- \left(\frac{zg}{bR^{(d)}T_0} \right)^2 \right]. \quad (D6)$$

To define the zonal velocity component, define the great circle distance between (λ, φ) and (λ_p, φ_p) :

$$R(\lambda, \varphi; \lambda_p, \varphi_p) = r \arccos(\sin \varphi \sin \varphi_p + \cos \varphi \cos \varphi_p \cos(\lambda - \lambda_p)). \quad (D7)$$

The zonal velocity component is

$$u(\varphi, z) = -\Omega r \cos(\varphi) + \sqrt{(\Omega r \cos(\varphi))^2 + r \cos(\varphi) U(\varphi, z)} + u'(\lambda, \varphi, z), \quad (D8)$$

where the zonally symmetric part of the velocity field is given by

$$U(\varphi, z) = \frac{g\mathcal{K}}{r} F_4(z) [(\cos \varphi)^{\mathcal{K}-1} - (\cos \varphi)^{\mathcal{K}+1}] T_v(\varphi, z). \quad (D9)$$

$r = 6371.22$ m is the mean radius of Earth and angular velocity is $\Omega = \frac{2\pi}{86164} \frac{1}{s}$ (denominator is length of day in seconds), and $u'(\lambda, \varphi, z)$ is the exponential bell-shaped perturbation to the zonally balanced velocity field

$$u'(\lambda, \varphi, z) = \begin{cases} U_p Z(z) \exp\left[- \left(\frac{R(\lambda, \varphi; \lambda_p, \varphi_p)}{r_p} \right)^2 \right], & \text{if } R(\lambda, \varphi; \lambda_p, \varphi_p) < r_p, \\ 0, & \text{otherwise,} \end{cases} \quad (D10)$$

where perturbation velocity is $U_p = 1$ m/s, longitude/latitude of the zonal wind perturbation center point is $(\lambda_p, \varphi_p) = (\pi/9, 2\pi/9) = (20^\circ \text{E}, 40^\circ \text{N})$ and

$$Z(z) = \begin{cases} 1 - 3 \left(\frac{z}{z_p} \right)^2 + 2 \left(\frac{z}{z_p} \right)^3, & \text{if } z \leq z_p, \\ 0, & \text{otherwise,} \end{cases} \quad (D11)$$

where $z_p = 15,000$ m is the maximum height of the zonal wind perturbation. The specific humidity (moist mixing ratio for water vapor) is specified in terms of (moist) pressure (as the vertical variable)

$$q^{(wv)}(\lambda, \varphi, p) = \begin{cases} q_0 \exp \left[- \left(\frac{\varphi}{\varphi_w} \right)^4 \right] \exp \left[- \left(\frac{p-p_0}{p_w} \right)^2 \right], & \text{if } p > p_0/10, \\ q_t, & \text{otherwise,} \end{cases} \quad (D12)$$

where $p_w = 340$ hPa is a pressure width parameter, $q_0 = 0.018$ kg/kg is maximum specific humidity, $q_t = 1.0 \times 10^{-12}$ kg/kg is specific humidity above artificial tropopause, and $\varphi_w = 2\pi/9$ is the specific humidity latitudinal width parameter. In addition to water vapor the Kessler microphysics water species are cloud liquid, $m^{(cl)}$, and rain water, $m^{(rw)}$. Both are initialized to 0 kg/kg.

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References

- Abdi, D. S., Giraldo, F. X., Constantinescu, E. M., Carr, L. E., Wilcox, L. C., & Warburton, T. C. (2017). Acceleration of the IMplicit-EXplicit nonhydrostatic unified model of the atmosphere on manycore processors. *The International Journal of High Performance Computing Applications*, 0(0), 1094342017732395. <https://doi.org/10.1177/1094342017732395>
- Abdi, D. S., Wilcox, L. C., Warburton, T. C., & Giraldo, F. X. (2017). A GPU-accelerated continuous and discontinuous Galerkin non-hydrostatic atmospheric model. *The International Journal of High Performance Computing Applications*, 0(0), 1094342017694427. <https://doi.org/10.1177/1094342017694427>
- Ainsworth, M. (2014). Dispersive behaviour of high order finite element schemes for the one-way wave equation. *Journal of Computational Physics*, 259, 1–10. <https://doi.org/10.1016/j.jcp.2013.11.003>
- Ainsworth, M., & Wajid, H. A. (2009). Dispersive and dissipative behavior of the spectral element method. *SIAM Journal on Numerical Analysis*, 47(5), 3910–3937. <https://doi.org/10.1137/080724976>
- Bacmeister, J. T., Lauritzen, P. H., Dai, A., & Truesdale, J. E. (2012). Assessing possible dynamical effects of condensate in high resolution climate simulations. *Geophysical Research Letters*, 39, L04806. <https://doi.org/10.1029/2011GL050533>
- Bacmeister, J. T., Reed, K. A., Hannay, C., Lawrence, P., Bates, S., Truesdale, J. E., et al. (2016). Projected changes in tropical cyclone activity under future warming scenarios using a high-resolution climate model. *Climatic Change*, 39, L04806. <https://doi.org/10.1007/s10584-016-1750-x>
- Baer, F., Wang, H., Tribbia, J. J., & Fournier, A. (2006). Climate modeling with spectral elements. *Monthly Weather Review*, 134(12), 3610–3624. <https://doi.org/10.1175/MWR3360.1>
- Bannon, P. R. (2003). Hamiltonian description of idealized binary geophysical fluids. *Journal of the Atmospheric Sciences*, 60(22), 2809–2819. [https://doi.org/10.1175/1520-0469\(2003\)060<2809:HDOIBG>2.0.CO;2](https://doi.org/10.1175/1520-0469(2003)060<2809:HDOIBG>2.0.CO;2)
- Bellucci, A., Gualdi, S., & Navarra, A. (2010). The double-ITCZ syndrome in coupled general circulation models: The role of large-scale vertical circulation regimes. *Journal of Climate*, 23, 1127. <https://doi.org/10.1175/2009JCLI3002.1>
- Canuto, C., Hussaini, M. Y., Quarteroni, A., & Zang, T. (2007). *Spectral methods: Evolution to complex geometries and applications to fluid dynamics* (1st ed.). Berlin Heidelberg: Springer.
- Charney, J. (1971). Geostrophic turbulence. *Journal of the Atmospheric Sciences*, 28, 1087–1095.
- Choi, S.-J., & Hong, S.-Y. (2016). A global non-hydrostatic dynamical core using the spectral element method on a cubed-sphere grid. *Asia-Pacific Journal of Atmospheric Sciences*, 52(3), 291–307. <https://doi.org/10.1007/s13143-016-0005-0>
- Colella, P., & Woodward, P. R. (1984). The piecewise parabolic method (PPM) for gas-dynamical simulations. *Journal of Computational Physics*, 54, 174–201.
- Dennis, J. M., Edwards, J., Evans, K. J., Guba, O., Lauritzen, P. H., Mirin, A. A., et al. (2012). CAM-SE: A scalable spectral element dynamical core for the Community Atmosphere Model. *International Journal of High Performance Computing*, 26(1), 74–89. <https://doi.org/10.1177/1094342011428142>
- Dubos, T., Dubey, S., Tort, M., Mittal, R., Meurdesoif, Y., & Hourdin, F. (2015). Dynamico-1.0, an icosahedral hydrostatic dynamical core designed for consistency and versatility. *Geoscientific Model Development*, 8, 3131–3150. <https://doi.org/10.5194/gmd-8-3131-2015>
- Emanuel, K. (1994). *Atmospheric convection*. New York: Oxford University Press.
- Erath, C., Lauritzen, P. H., Garcia, J. H., & Tufo, H. M. (2012). Integrating a scalable and efficient semi-Lagrangian multi-tracer transport scheme in HOMME. *Procedia Computer Science*, 9, 994–1003.
- Evans, K., Lauritzen, P. H., Mishra, S., Neale, R., Taylor, M. A., & Tribbia, J. J. (2012). AMIP simulations with the CAM4 spectral element dynamical core. *Journal of Climate*, 26, 689–709.
- Fournier, A., Taylor, M. A., & Tribbia, J. J. (2004). The spectral element atmosphere model (SEAM): High-resolution parallel computation and localized resolution of regional dynamics. *Monthly Weather Review*, 132(3), 726–748.
- Gettelman, A., Callaghan, P., Larson, V. E., Zarzycki, C. M., Bacmeister, J. T., Lauritzen, P. H., et al. (2018). Regional climate simulations with the Community Earth System Model. *Journal of Advances in Modeling Earth Systems*, 10. <https://doi.org/10.1002/2017MS001227>
- Giraldo, F. X., Kelly, J. F., & Constantinescu, E. M. (2013). Implicit-explicit formulations of a three-dimensional nonhydrostatic unified model of the atmosphere (NUMA). *SIAM Journal on Scientific Computing*, 35(5), B1162–B1194. <https://doi.org/10.1137/120876034>
- Guba, O., Taylor, M., & St-Cyr, A. (2014). Optimization-based limiters for the spectral element method. *Journal of Computational Physics*, 267(0), 176–195. <https://doi.org/10.1016/j.jcp.2014.02.029>
- Guba, O., Taylor, M. A., Ullrich, P. A., Overfelt, J. R., & Levy, M. N. (2014). The spectral element method (SEM) on variable-resolution grids: Evaluating grid sensitivity and resolution-aware numerical viscosity. *Geoscientific Model Development*, 7(6), 2803–2816. <https://doi.org/10.5194/gmd-7-2803-2014>
- Guerra, J. E., & Ullrich, P. A. (2016). A high-order staggered finite-element vertical discretization for non-hydrostatic atmospheric models. *Geoscientific Model Development*, 9(5), 2007–2029. <https://doi.org/10.5194/gmd-9-2007-2016>
- Herrington, A. R., & Reed, K. A. (2017). An explanation for the sensitivity of the mean state of the community atmosphere model to horizontal resolution on aquaplanets. *Journal of Climate*, 30(13), 4781–4797. <https://doi.org/10.1175/jcli-d-16-0069.1>
- Hurrell, J. W., Holland, M. M., Gent, P. R., Ghan, S., Kay, J. E., Kushner, P. J., et al. (2013). The community Earth system model: A framework for collaborative research. *Bulletin of the American Meteorological Society*, 94(9), 1339–1360. <https://doi.org/10.1175/BAMS-D-12-00121.1>

- Jablonowski, C., & Williamson, D. L. (2006). A baroclinic instability test case for atmospheric model dynamical cores. *Quarterly Journal of the Royal Meteorological Society*, 132, 2943–2975.
- Karniadakis, G., & Sherwin, S. (2013). *Spectral/hp element methods for computational fluid dynamics* (Vol. 2, pp. 1–686). New York: Oxford University Press.
- Kasahara, A. (1974). Various vertical coordinate systems used for numerical weather prediction. *Monthly Weather Review*, 102(7), 509–522.
- Kelly, J. F., & Giraldo, F. X. (2012). Continuous and discontinuous Galerkin methods for a scalable three-dimensional nonhydrostatic atmospheric model: Limited-area mode. *Journal of Computational Physics*, 231(24), 7988–8008. <https://doi.org/10.1016/j.jcp.2012.04.042>
- Kessler, E. (1969). On the distribution and continuity of water substance in atmospheric circulations. *Meteorological Monographs*, 10(32), 88.
- Kinnmark, I. P., & Gray, W. G. (1984). One step integration methods of third-fourth order accuracy with large hyperbolic stability limits. *Mathematics and Computers in Simulation*, 26(3), 181–188. [https://doi.org/10.1016/0378-4754\(84\)90056-9](https://doi.org/10.1016/0378-4754(84)90056-9)
- Lauritzen, P. H., Bacmeister, J. T., Dubos, T., Lebonnois, S., & Taylor, M. A. (2014). Held-Suarez simulations with the Community Atmosphere Model Spectral Element (CAM-SE) dynamical core: A global axial angular momentum analysis using Eulerian and floating Lagrangian vertical coordinates. *Journal of Advances in Modeling Earth Systems*, 6, 129–140. <https://doi.org/10.1002/2013MS000268>
- Lauritzen, P. H., & Goldhaber, S. (2017). CESM simpler models: Moist baroclinic wave with Kessler microphysics. Retrieved from <http://www.cesm.ucar.edu/models/simpler-models-index/fkessler/index.html>
- Lauritzen, P. H., Mirin, A., Truesdale, J., Raeder, K., Anderson, J., Bacmeister, J., & Neale, R. B. (2011). Implementation of new diffusion/ filtering operators in the CAM-FV dynamical core. *The International Journal of High Performance Computing Applications*, 26, 63–73. <https://doi.org/10.1177/1094342011410088>
- Lauritzen, P. H., Nair, R. D., & Ullrich, P. A. (2010). A conservative semi-Lagrangian multi-tracer transport scheme (CSLAM) on the cubed-sphere grid. *Journal of Computational Physics*, 229, 1401–1424. <https://doi.org/10.1016/j.jcp.2009.10.036>
- Lauritzen, P. H., Taylor, M. A., Overfelt, J., Ullrich, P. A., Nair, R. D., Goldhaber, S., & Kelly, R. (2017). CAM-SE-CSLAM: Consistent coupling of a conservative semi-lagrangian finite-volume method with spectral element dynamics. *Monthly Weather Review*, 145(3), 833–855. <https://doi.org/10.1175/MWR-D-16-0258.1>
- Lauritzen, P., & Thuburn, J. (2012). Evaluating advection/transport schemes using interrelated tracers, scatter plots and numerical mixing diagnostics. *Quarterly Journal of the Royal Meteorological Society*, 138(665), 906–918. <https://doi.org/10.1002/qj.986>
- Lebonnois, S., Covey, C., Grossman, A., Parish, H., Schubert, G., Walterscheid, R., et al. (2012). Angular momentum budget in general circulation models of superrotating atmospheres: A critical diagnostic. *Journal of Geophysical Research*, 117, E12004. <https://doi.org/10.1029/2012JE004223>
- Lebonnois, S., Hourdin, F., Eymet, V., Crespin, A., Fournier, R., & Forget, F. (2010). Superrotation of Venus's atmosphere analysed with a full general circulation model. *Journal of Geophysical Research*, 115, E06006. <https://doi.org/10.1029/2009JE003458>
- Lin, S.-J. (2004). A 'vertically Lagrangian' finite-volume dynamical core for global models. *Monthly Weather Review*, 132, 2293–2307.
- Maday, Y., & Patera, A. T. (1987). Spectral element methods for the incompressible Navier Stokes equations. In A. K. Noor, & J. T. Oden (Eds.), *State of the Art Surveys on Computational Mechanics* (pp. 71–143). New York: American Society of Mechanical Engineers.
- Medeiros, B., Williamson, D. L., & Olson, J. G. (2016). Reference aquaplanet climate in the community atmosphere model, version 5. *Journal of Advances in Modeling Earth Systems*, 8, 406–424. <https://doi.org/10.1002/2015MS000593>
- Muller, C. J., & O'Gorman, P. A. (2011). An energetic perspective on the regional response of precipitation to climate change. *Nature Climate Change*, 1(5), 266–271. <https://doi.org/10.1038/nclimate1169>
- Nair, R., Choi, H.-W., & Tufo, H. (2009). Computational aspects of a scalable high-order discontinuous Galerkin atmospheric dynamical core. *Computers & Fluids*, 38(2), 309–319. <https://doi.org/10.1016/j.compfluid.2008.04.006>
- Nair, R. D., Thomas, S. J., & Loft, R. D. (2005). A discontinuous Galerkin global shallow water model. *Monthly Weather Review*, 133(4), 876–888. <https://doi.org/10.1175/MWR2903.1>
- Nastrom, G. D., & Gage, K. S. (1985). A climatology of atmospheric wavenumber spectra of wind and temperature observed by commercial aircraft. *Journal of the Atmospheric Sciences*, 42, 950–960.
- Neale, R. B., Chen, C.-C., Gettelman, A., Lauritzen, P. H., Park, S., Williamson, D. L., et al. (2012). Description of the NCAR Community Atmosphere Model (CAM 5.0) (NCAR Technical Note NCAR/TN-486+STR). National Center of Atmospheric Research.
- Neale, R. B., & Hoskins, B. J. (2000). A standard test for AGCMs including their physical parametrizations: I: The proposal. *Atmospheric Science Letters*, 1(2), 101–107. <https://doi.org/10.1006/asle.2000.0022>
- O'Brien, T. A., Collins, W. D., Kashinath, K., Rübel, O., Byna, S., Gu, J., et al. (2016). Resolution dependence of precipitation statistical fidelity in hindcast simulations. *Journal of Advances in Modeling Earth Systems*, 8, 976–990. <https://doi.org/10.1002/2016ms000671>
- Ooyama, K. V. (1990). A thermodynamic foundation for modeling the moist atmosphere. *Journal of the Atmospheric Sciences*, 47(21), 2580–2593. [https://doi.org/10.1175/1520-0469\(1990\)047<2580:ATFFMT>2.0.CO;2](https://doi.org/10.1175/1520-0469(1990)047<2580:ATFFMT>2.0.CO;2)
- Ooyama, K. V. (2001). A dynamic and thermodynamic foundation for modeling the moist atmosphere with parameterized microphysics. *Journal of the Atmospheric Sciences*, 58(15), 2073–2102. [https://doi.org/10.1175/1520-0469\(2001\)058<2073:ADATFF>2.0.CO;2](https://doi.org/10.1175/1520-0469(2001)058<2073:ADATFF>2.0.CO;2)
- Polvani, L. M., Clement, A. C., Medeiros, B., Benedict, J. J., & Simpson, I. R. (2017). When less is more: Opening the door to simpler climate models. *Eos*, 98. <https://doi.org/10.1029/2017EO079417>
- Rančić, M., Purser, J., & Mesinger, F. (1996). A global shallow water model using an expanded spherical cube. *Quarterly Journal of the Royal Meteorological Society*, 122, 959–982.
- Reed, K. A., Bacmeister, J. T., Rosenbloom, N. A., Wehner, M. F., Bates, S. C., Lauritzen, P. H., et al. (2015). Impact of the dynamical core on the direct simulation of tropical cyclones in a high-resolution global model. *Geophysical Research Letters*, 42, 3603–3608. <https://doi.org/10.1002/2015GL063974>
- Rhoades, A. M., Huang, X., Ullrich, P. A., & Zarzycki, C. M. (2016). Characterizing sierra nevada snowpack using variable-resolution CESM. *Journal of Applied Meteorology and Climatology*, 55(1), 173–196. <https://doi.org/10.1175/JAMC-D-15-0156.1>
- Richter, J. H., Solomon, A., & Bacmeister, J. T. (2014). On the simulation of the quasi-biennial oscillation in the Community Atmosphere Model, version 5. *Journal of Geophysical Research: Atmospheres*, 119, 3045–3062. <https://doi.org/10.1002/2013JD021122>
- Sadourny, R. (1972). Conservative finite-difference approximations of the primitive equations on quasi-uniform spherical grids. *Monthly Weather Review*, 100, 136–144.
- Satoh, M. (2003). Conservative scheme for a compressible nonhydrostatic model with moist processes. *Monthly Weather Review*, 131(6), 1033–1050. [https://doi.org/10.1175/1520-0493\(2003\)131<1033:CSFACN>2.0.CO;2](https://doi.org/10.1175/1520-0493(2003)131<1033:CSFACN>2.0.CO;2)
- Simmons, A. J., & Burridge, D. M. (1981). An energy and angular-momentum conserving vertical finite-difference scheme and hybrid vertical coordinates. *Monthly Weather Review*, 109(4), 758–766.
- Skamarock, W. C., Park, S.-H., Klemp, J. B., & Snyder, C. (2014). Atmospheric kinetic energy spectra from global high-resolution nonhydrostatic simulations. *Journal of the Atmospheric Sciences*, 71(11), 4369–4381. <https://doi.org/10.1175/JAS-D-14-0114.1>

- Spiteri, R., & Ruuth, S. (2002). A new class of optimal high-order strong-stability-preserving time discretization methods. *SIAM Journal on Numerical Analysis*, 40(2), 469–491. <https://doi.org/10.1137/S0036142901389025>
- Staniforth, A., White, A., Wood, N., Thuburn, J., Zerroukat, M., Cordero, E., et al. (2006). Joy of u.m. 6.3—Model formulation (UK Met Office Technical Note 15). Exeter, England: UK Meteorological.
- Starr, V. P. (1945). A quasi-Lagrangian system of hydrodynamical equations. *Journal of the Atmospheric Sciences*, 2, 227–237.
- Taylor, M. A. (2011). Conservation of mass and energy for the moist atmospheric primitive equations on unstructured grids. In P. H. Lauritzen, et al. (Eds.), *Numerical techniques for global atmospheric models* (Vol. 80, pp. 357–380). Lecture Notes in Computational Science and Engineering. Berlin, Heidelberg: Springer. https://doi.org/10.1007/978-3-642-11640-7_12
- Taylor, M., Edwards, J., & St-Cyr, A. (2008). Petascale atmospheric models for the community climate system model: New developments and evaluation of scalable dynamical cores. *Journal of Physics: Conference Series*, 125, 12023. <https://doi.org/10.1088/1742-6596/125/1/012023>
- Taylor, M. A., & Fournier, A. (2010). A compatible and conservative spectral element method on unstructured grids. *Journal of Computational Physics*, 229(17), 5879–5895. <https://doi.org/10.1016/j.jcp.2010.04.008>
- Thatcher, D. R., & Jablonowski, C. (2016). A moist aquaplanet variant of the Held-Suarez test for atmospheric model dynamical cores. *Geoscientific Model Development*, 9(4), 1263–1292. <https://doi.org/10.5194/gmd-9-1263-2016>
- Thomas, S., & Loft, R. (2000). Parallel semi-implicit spectral element methods for atmospheric general circulation models. *Journal of Scientific Computing*, 15, 499–518.
- Thuburn, J. (2008). Some conservation issues for the dynamical cores of NWP and climate models. *Journal of Computational Physics*, 227, 3715–3730.
- Ullrich, P. A. (2014). A global finite-element shallow-water model supporting continuous and discontinuous elements. *Geoscientific Model Development*, 7(6), 3017–3035. <https://doi.org/10.5194/gmd-7-3017-2014>
- Ullrich, P. A., Jablonowski, C., Kent, J., Lauritzen, P. H., Nair, R., Reed, K. A., et al. (2017). Dcmip2016: A review of non-hydrostatic dynamical core design and intercomparison of participating models. *Geoscientific Model Development*, 10, 4477–4509. <https://doi.org/10.5194/gmd-10-4477-2017>
- Ullrich, P. A., Melvin, T., Jablonowski, C., & Staniforth, A. (2014). A proposed baroclinic wave test case for deep and shallow-atmosphere dynamical cores. *Quarterly Journal of the Royal Meteorological Society*, 140(682), 1590–1602.
- Williamson, D. L., Blackburn, M., Hoskins, B., Nakajima, K., Ohfuchi, W., Takahashi, Y., et al. (2012). The ape atlas (NCAR Technical Note, NCAR/TN-484+STR). Boulder, CO: National Center for Atmospheric Research. <https://doi.org/10.5065/D6FF3QBR>
- Williamson, D. L., Olson, J. G., Hannay, C., Toniazzo, T., Taylor, M., & Yudin, V. (2015). Energy considerations in the community atmosphere model (CAM). *Journal of Advances in Modeling Earth Systems*, 7, 1178–1188. <https://doi.org/10.1002/2015MS000448>
- Zarzycki, C. M., & Jablonowski, C. (2014). A multidecadal simulation of Atlantic tropical cyclones using a variable-resolution global atmospheric general circulation model. *Journal of Advances in Modeling Earth Systems*, 6, 805–828. <https://doi.org/10.1002/2014MS000352>
- Zarzycki, C. M., Jablonowski, C., & Taylor, M. A. (2014). Using variable-resolution meshes to model tropical cyclones in the community atmosphere model. *Monthly Weather Review*, 142(3), 1221–1239. <https://doi.org/10.1175/MWR-D-13-00179.1>
- Zarzycki, C. M., Levy, M. N., Jablonowski, C., Overfelt, J. R., Taylor, M. A., & Ullrich, P. A. (2014). Aquaplanet experiments using CAM's variable-resolution dynamical core. *Journal of Climate*, 27(14), 5481–5503. <https://doi.org/10.1175/JCLI-D-14-00004.1>
- Zhang, K., Rasch, P. J., Taylor, M. A., Wan, H., Leung, L.-Y. R., Ma, P.-L., et al. (2017). Impact of numerical choices on water conservation in the E3SM atmosphere model version 1 (EAM v1). *Geoscientific Model Development Discussions*, 2017, 1–26. <https://doi.org/10.5194/gmd-2017-293>
- Zhao, I., Held, I. M., & Lin, S.-J. (2012). Some counterintuitive dependencies of tropical cyclone frequency on parameters in a gcm. *Journal of the Atmospheric Sciences*, 69, 2272–2283. <https://doi.org/10.1175/JAS-D-11-0238.1>