A Mountain-Induced Moist Baroclinic Wave Test Case for the Dynamical Cores of Atmospheric General Circulation Models

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Abstract.

Idealized test cases for the dynamical cores of Atmospheric General Circulation Models are informative tools to assess the accuracy of the numerical designs and to investigate the general characteristics of atmospheric motions. A new test case is introduced which is built upon a baroclinically-unstable base state with an added orographic barrier. The topography is analytically prescribed and acts as a trigger of both baroclinic Rossby waves and inertia-gravity waves on a rotating, regular-size planet. Both dry and idealized moist configurations are suggested. The latter utilizes the Kessler warm-rain precipitation scheme. The test case enhances the complexity of the existing test suite hierarchy and focuses on the impacts of two midlatitudinal mountain ridges on the circulation. Selected simulations examples from four dynamical cores are shown. These are the Spectral Element, Finite Volume, and Cubed-Sphere Finite Volume dynamical cores which are part of NCAR’s Community Earth System Model (CESM) version 2.2. In addition, the Model for Prediction Across Scales (MPAS) is tested. The overall flow patterns agree well in the four dynamical cores, but the details can vary greatly. The examples highlight the broad palette of use cases for the test case and also reveal physics-dynamics coupling issues.

1 Introduction

An important component of an Atmospheric General Circulation Model (AGCM) is the dynamical core which solves the fluid flow equations on a computational grid. The dynamical core thereby captures the resolved scales of the flow, defines the accuracy of the horizontal, vertical, and temporal numerical discretizations, determines the dissipation characteristics of the flow, and also selects the treatment of topography via the choice of the vertical coordinate. Testing the accuracy of a dynamical core is a paramount development step for weather and climate models. This is typically facilitated by performing dynamical core integrations of idealized test cases. These test cases have lower complexity than realistic weather forecasts or climate simulations and, for example, use only dry dynamical core configurations, dry or moist model setups with simplified physical processes or simplified lower-boundary conditions, and/or idealized initial conditions. This provides a controlled environment which captures selected atmospheric motions of interest. Such idealized model configurations serve two purposes. First, they allow assessments of the numerical schemes and serve as a standardized testing framework for model intercomparisons, thereby guiding the design and tuning decisions of developers. Second, idealized test cases are also used as atmospheric dynamics tools to understand physical phenomena, such as the dependence of orographic gravity waves on the Froude number, or to assess...
the impacts of mountains on midlatitudinal dynamics, precipitation, or the general circulation of the atmosphere. Our proposed test case serves both purposes.

The suite of test cases for dynamical core and idealized climate model validations spans a hierarchy of complexities. Test cases have, for example, been developed for the simpler shallow water equations (Williamson et al., 1992; Galewsky et al., 2004; Shamir et al., 2019) which serve as a 2D horizontal testbed for atmospheric motions. In addition, the hierarchy includes test cases for dry 3D dynamical cores (Held and Suarez, 1994; Jablonowski and Williamson, 2006; Wedi and Smolarkiewicz, 2009; Lauritzen et al., 2010; Kent et al., 2014a; Ullrich et al., 2014; Shamir and Paldor, 2016), idealized moist 3D dynamical cores (Thatcher and Jablonowski, 2016; Klemp et al., 2015), and aqua-planet models (Neale and Hoskins, 2000; Lee et al., 2008). Aqua-planet models use a full-complexity physical parameterization suite but a simplified lower boundary condition.

The latter is either built upon a flat, ocean-covered earth with analytically prescribed sea surface temperatures as in Neale and Hoskins (2000) or utilizes a slab ocean configuration with a constant mixed-layer depth as in Lee et al. (2008) or Kang et al. (2008).

One dynamical core design aspect that can be studied at various levels of complexity is the treatment of topography and the vertical coordinate. Often, the inclusion of topography in a dynamical core is first tested via simpler equation sets, for example, utilizing a hydrostatic, Boussinesq or anelastic approximation and setting the Coriolis parameter to zero. Typically, 2D Cartesian x-z configurations with smoothly-varying, e.g., bell-shaped, mountain profiles and idealized initial conditions with a constant background stratification and zonal flow are used. Examples are the 2D nonrotating test configurations by Klemp and Lilly (1978), Durrant and Klemp (1983) and Kurowski et al. (2013) that were designed for dry and moist orographic flows. Alternatively, Schär et al. (2002) and Guerra and Ullrich (2021) used dry, nonrotating, orographic gravity wave tests to assess their 2D x-z non-hydrostatic model designs. A portfolio of 2D hydrostatic and non-hydrostatic gravity waves as well as inertia-gravity waves with rotation on a fixed f−plane were assessed in Dudhia (1993) and Ullrich and Jablonowski (2012a). In addition, 3D Cartesian nonrotating mountain waves were analyzed in, e.g., Smolarkiewicz and Rotunno (1989) and Schär and Durrant (1997). For such idealized test scenarios, linear as well as nonlinear analytic steady-state gravity wave solutions can be computed as shown in Smith (1980) and Guerra and Ullrich (2021), respectively.

However, dynamical core test cases for orographic flows on the sphere are less abundant in the literature. In general, there are three aspects that get discussed. The first aspect addresses the accuracy of the vertical, often orography-following, coordinate and is sometimes called the “acid test”. This assesses whether a resting nonrotating atmosphere in hydrostatic equilibrium stays motionless in the presence of topography as, e.g., assessed in Lin (1997), Qian et al. (1998), or Zängl (2012). The second test principle mimics the Cartesian gravity wave configurations mentioned above. Idealized ridge mountains or mountains with circular shapes are then embedded in idealized flows with a solid body rotation and constant stratification on a nonrotating planet with either a full-size or reduced-size radius. Such configurations were suggested in Tomita and Satoh (2004) (case 3), Ullrich et al. (2012) (case 2) and Klemp et al. (2015). In particular, the Ullrich et al. (2012) test variant was specifically developed for the “Dynamical Core Model Intercomparison Project” (DCMIP) which conducts regular international dynamical core assessments (see also Jablonowski et al. (2008), Ullrich et al. (2016, 2017) and Zarzycki et al. (2019)). The third test principle uses a full-size earth with the earth’s rotation and focuses on the representation of orographically-induced Rossby
wave trains instead of gravity waves. Such test configurations with bell-shaped mountains were described in Tomita and Satoh (2004) (case 5), Jablonowski et al. (2008) (case 5), and Ullrich and Jablonowski (2012b). These are built upon highly idealized initial conditions, such as isothermal states, a constant stratification and solid body rotation. The induced 3D Rossby wave train thereby mimics the widely-used 2D shallow water “test case 5” as defined in Williamson et al. (1992). However, test cases for more complex, analytically-prescribed initial flows with topography have not been described yet for spherical geometries. Our proposed test case helps fill this gap, and, in particular, assesses the impact of mountains on baroclinic waves for both dry and idealized moist dynamical core configurations.

Previous work in spherical geometry highlighted the design and usefulness of baroclinic wave test cases for atmospheric flows without orographic obstacles (Polvani et al., 2004; Jablonowski and Williamson, 2006; Staniforth and White, 2011; Ullrich et al., 2014, 2016). In the absence of analytical solutions the evolution of a baroclinic wave is then typically computed over 10-20 days and intercompared to numerical solutions from other dynamical cores to gain insight into the flow characteristics. This sheds light on the diffusivity of the models and can even reveal dynamical core design flaws as, for example, demonstrated in Williamson et al. (2009). Adding 2D mountains to such test configurations is not necessarily straightforward since the initial steady-state background conditions are analytically balanced and zonally symmetric. These characteristics of the initial conditions get disrupted by 2D mountain shapes. Therefore, orographic effects on idealized baroclinic waves have only been assessed in 3D Cartesian model configurations so far. The initial conditions are easier to balance in Cartesian geometry and have, for example, been used for baroclinic wave studies and their interaction with a ridge mountain in Menchaca and Durran (2017, 2018).

The proposed test extends the test case hierarchy and describes the evolution of baroclinic waves on the rotating full-size planet which are triggered by idealized topography. The background flow field is based upon the ideas in Staniforth and White (2011) and Ullrich et al. (2014, 2016) who defined a family of steady-state initial conditions for baroclinic waves without topography in dry and moist environments. In particular, the moist steady-state from Ullrich et al. (2016) is mostly utilized here in conjunction with a Kessler warm rain scheme. The latter represents an idealized parameterization of moisture processes without a cloud phase (Kessler, 1969; Klemp et al., 2015) and was also used during DCMIP in 2016 (Ullrich et al., 2016).

The idealized precipitation triggered by the baroclinic wave then amplifies the wave in a highly nonlinear way. However, the moisture processes are optional, and both dry and moist dynamical core evaluations with topography are insightful use cases. In this paper we chose to add two mountain ridges in the northern midlatitudes which require adjustments of the initial state to recover the well-balanced background condition for baroclinic waves. A broad palette of topographic shapes, peak heights and locations are possible as long as the topographic profile has an analytic description. The latter informs the computation of the well-balanced, albeit not perfectly balanced, initial state. The mountains then act as triggers for baroclinic waves. They thereby replace the overlaid initial wind or temperature perturbations that are typically used in the absence of a topographic trigger.

In summary, this work introduces a test case which combines idealized moisture physics, topographic forcing, mountain-enhanced precipitation, and the evolution of baroclinic waves on a rotating full-size planet. The paper has three goals. First, we introduce the design of the mountain-induced baroclinic wave test case. Second, selected examples from the Spectral Element (SE, Lauritzen et al. (2018)) dynamical core of the Community Earth System Model (CESM) are used to illustrate the
characteristics of the test case and its orographically-induced flow. Third, snapshots of a brief model intercomparison are shown to gain insight into various dynamical core designs and the associated model spread. This intercomparison includes simulations with the Model for Prediction Across Scales (MPAS, Skamarock et al. (2012)), as well as the CESM Finite-Volume (FV, Lin (2004)) and CESM Cubed-Sphere Finite-Volume (FV3, Harris et al. (2021)) configurations. The latter two are part of the CESM version 2.2 release of the National Center for Atmospheric Research (NCAR). The expectation is that the test case will help diagnose numerical artifacts resulting from the inclusion of topography in dynamical cores and furthermore reveal physics-dynamics coupling issues. In addition, the test enables general assessments of the atmospheric circulation driven by mountain-generated gravity and Rossby waves, and thereby serves as a new generic tool in the atmospheric dynamics toolbox.

This article is structured as follows. Section 2 lays out the specifications of the test case and justifies the chosen parameters. Section 3 introduces the dynamical cores which are used for a brief model intercomparison. Section 4 analyzes the important characteristics of the orographically-induced baroclinic wave via the SE model. Section 5 highlights selected dynamical core intercomparisons and briefly surveys a physics-dynamics coupling aspect revealed by this test case. The appendices provide technical specifications for all four dynamical cores assessed here to make the results reproducible.

2 Test Case Design

Previously-designed 3D dynamical core test cases (Jablonowski et al., 2006; Ullrich et al., 2014, 2016) have demonstrated that baroclinic waves serve as an efficient tool to assess the characteristics of dry or moist flow fields. These test cases have two key components: a steady-state background state that is designed to be baroclinically unstable and an added perturbation that triggers the formation of a baroclinic wave. Our test case is designed with a moist and a dry variant. In moist runs, Kessler physics (Kessler, 1969; Klemp et al., 2015; Ullrich et al., 2016) is chosen as the precipitation mechanism. It is an idealized warm-rain scheme with three water species which are the dry mixing ratios of water vapor, liquid water, and rain water without ice. No other physical parameterizations are employed. This test setup thereby sheds light on the impact of the diabatic forcing from the precipitation on the evolution of the wave and the physics-dynamics coupling strategy. The dry, adiabatic variant of the test case is obtained by simply setting the initial humidity content to zero and avoiding the use of physical parameterizations.

The design of the test case is inspired by real-world phenomena and topographic shapes like the Andes or Rocky Mountains. In nature, extreme precipitation can result from topographic forcing, such as the interaction between atmospheric rivers and mountains in the Pacific northwest region of the United States. In our test case, the evolving precipitation bands that develop along with the topographically triggered baroclinic wave are qualitatively similar to atmospheric rivers. The test case is not designed to be complex enough to make direct comparisons with real-world atmospheric rivers. However, Hagos et al. (2015) highlighted the clear dependence of atmospheric river climatologies on the choice of the dynamical core in more complex aqua-planet simulations. Therefore, our idealized test configuration acts as a controlled setting to study the effects of the dynamical core design on such high-intensity precipitation scenarios.
Figure 1. Latitude-height profiles at $72^\circ$ E of the initial (a) zonal wind $u$, (b) temperature $T$, (c) specific humidity $q$, and (d) relative humidity. The idealized topographic profile is shown in white in the lower right of the plots.

2.1 Properties of the initial background state

The atmospheric base state for the baroclinic wave without an overlaid perturbation is taken from Ullrich et al. (2014, 2016). They describe an analytic steady-state solution to the dry and moist 3D fluid flow equations on a rotating sphere without topography. Both shallow-atmosphere and deep-atmosphere dynamical core designs are accommodated. All base-state prognostic variables are zonally symmetric in the absence of topography. Because the base state is drawn from previous work, most functional forms for the prognostic variables are relegated to Appendix A. In particular, the appendix lists the equations for the temperature $T$, zonal wind $u$, meridional wind $v$, pressure $p$, density $\rho$, and specific humidity $q$ in Eqs. (A1)-(A6). The latitude-height ($z$) cross sections of the initial conditions along the longitude $72^\circ$ E are shown in Fig. 1. This longitudinal location corresponds to the center position of the first mountain ridge which is depicted by the white area (see also Sect. 2.2). As
Figure 2. Latitude-longitude cross-sections of the (a) surface height \( z_s \) and (b) initial surface pressure \( p_s \).

outlined in Ullrich et al. (2014, 2016), models with a pressure-based vertical coordinate can be initialized by using a numerical root-finding technique to solve for the height \( z \) for any given pressure \( p \), and then substituting this height value into the provided equations. Figure 1a shows that the zonal wind is characterized by westerly jets in the midlatitudes. Their vertical wind shear profiles support the growth of baroclinic instability waves. The temperature distribution \( T \) (Fig. 1b) is in thermal wind balance with the zonal wind. The specific humidity \( q \) (Fig. 1c) is chosen to resemble the zonal-mean distribution of water vapor. Above the artificial tropopause level \( p_t = 150 \) hPa the \( q \) field is set to zero as listed in Eq. (A6) and Table A1. We note that this setting deviates slightly from Ullrich et al. (2016) who specified a minimum stratospheric specific humidity value of \( 10^{-12} \) kg kg\(^{-1} \) above 100 hPa. This change is irrelevant for the tropospheric baroclinic wave, but advised to prevent an initial supersaturation in the stratosphere. The moisture profile attains a maximum relative humidity of about 85% in the lower midlatitudes as shown in Fig. 1d. This calculation makes use of Tetens’ formula Eq. (A7) for the saturation condition as further explained in Appendix A.

2.2 Inclusion of topography: Surface height and surface pressure

The balanced background state is a steady-state solution in the absence of topography. The forcing by the added topography then triggers the baroclinic Rossby wave trains. The topographic profile and balanced surface pressure are shown in Fig. 2. These profiles utilize the mountain parameters and physical constants from Tables 1 and A1, and describe two non-overlapping ridges in the northern midlatitudes. The mountain shapes and peak heights impact the strength of the topographic forcing. They are chosen so that the baroclinic waves mature over the course of six days.

For the functional form of the topographic shape we define a modified longitude variable to make the description independent of the implemented longitudinal range of the model, such as \([0, 2\pi]\) or \([-\pi, \pi]\). Suppose that an AGCM parameterizes longitude over the interval \( \lambda \in [\lambda_{\text{min}}, \lambda_{\text{max}}] \) and \( \lambda_{\text{max}} - \lambda_{\text{min}} = 2\pi \). Then we define \( d_n(\lambda) = (\lambda - \lambda_{\text{min}}) - \lambda_n \), where \( n \in \{1, 2\} \) indexes each mountain. The corresponding longitudinal center locations \( \lambda_{1,2} \) are listed in Table 1. This leads to the modified longitude \( l_n(\lambda) = \min(d_n(\lambda), 2\pi - d_n(\lambda)) \) which ranges over the longitudinal distance \([-\pi, \pi]\) as measured from the longitudinal center.
Table 1. Parameters for the test case. The degrees are specified in radians as needed by the equations.

<table>
<thead>
<tr>
<th>Variable Name</th>
<th>Variable Description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$h_0$</td>
<td>Peak mountain height</td>
<td>$2 \times 10^3$ m</td>
</tr>
<tr>
<td>$\phi_{1,2}$</td>
<td>Latitude of mountain peak</td>
<td>$\pi/4$</td>
</tr>
<tr>
<td>$\lambda_{1,2}$</td>
<td>Longitude of mountain peak</td>
<td>$72\pi/180, 140\pi/180$</td>
</tr>
<tr>
<td>$\bar{\lambda}$</td>
<td>Nominal longitudinal width of mountain</td>
<td>$7\pi/180$</td>
</tr>
<tr>
<td>$\bar{\phi}$</td>
<td>Nominal latitudinal width of mountain</td>
<td>$40\pi/180$</td>
</tr>
<tr>
<td>$d$</td>
<td>Latitudinal scale parameter</td>
<td>$\frac{2}{3}(-\log(0.1))^{-1/6}$</td>
</tr>
<tr>
<td>$c$</td>
<td>Longitudinal scale parameter</td>
<td>$\frac{2}{3}(-\log(0.1))^{-1/2}$</td>
</tr>
</tbody>
</table>

The latitudes $\phi$ spans the interval $[-\pi/2, \pi/2]$. The mountain profile is then defined via the surface height

$$z_s(\lambda, \phi) = h_0 \sum_{n=1}^{2} \exp \left[ - \left( \frac{(\phi - \phi_n)}{d} \right)^6 + \left( \frac{I_n(\lambda)}{c} \right)^2 \right].$$

In general, any topographic profile is possible as long as it can be described by an analytical equation. The parameter $h_0$ represents the peak height of the topography. The functional form of each mountain, shown in Fig. 2a, is Gaussian in longitude. The exponent of the latitude term is increased to 6 from 2 in order to meridionally elongate the peak of the mountain. This elongation helps minimize any deviation of the maximum height of the discretized surface topography from the analytic maximum surface height $h_0$. The parameters $\phi_n$ and $\lambda_n$ represent the center latitude and longitude of the $n$th mountain, respectively. The parameter $\bar{\phi}$ specifies the distance along a line of constant longitude $\lambda = \lambda_n$ between the points where the surface topography is 10% of its maximum, that is, $z_s(\phi_n \pm \bar{\phi}/2, \lambda_n) = 0.1 \cdot h_0$. We treat this dimension as the nominal meridional extent of the mountain. The parameter $d$ transforms the specified $\bar{\phi}$ into the form required for the Gaussian-like functional form for the topography. Likewise, the parameter $\bar{\lambda}$ specifies the distance along a line of constant latitude $\phi = \phi_n$ such that $z_s(\phi_n, \lambda_n \pm \bar{\lambda}/2) = 0.1 \cdot h_0$, which is treated as the nominal zonal extent of the mountain. The parameter $c$ transforms the specified $\bar{\lambda}$ into the form required by the Gaussian functional form for the topography in the zonal direction.

The corresponding, balanced surface pressure can be calculated by evaluating the pressure profile from Eq. (A4) along the topographic profile:

$$p_s(\lambda, \phi) = p_0 \exp \left[ - \frac{q}{R_d} (\tau_{\text{int},1}(z_s(\lambda, \phi)) - \tau_{\text{int},2}(z_s(\lambda, \phi)) I_T(\phi)) \right].$$

Figure 2b shows that the surface pressure varies from $p_0 = 1000$ hPa at sea level to about 773 hPa near the northern tip of the ridges.

### 2.3 Vertical velocity

In hydrostatic dynamical cores with a pressure-based vertical coordinate the initial vertical pressure velocity $\omega$ does not need to be initialized. It will be computed diagnostically during the model integration. However, non-hydrostatic models need to take...
the initial vertical velocity into account which is induced by the non-zero zonal wind along the sloping topographic boundary. A non-zero $w$ can be added so that the vector $[u, v, w]^T$ runs parallel to vertically sloping model levels if a topography-following coordinate is used. This is achieved by setting $w = v_h \cdot \nabla_s z$ where $v_h$ symbolizes the horizontal wind vector at constant height $z$. The subscript $s^*$ denotes that the horizontal gradient needs to be computed along the transformed orography-following vertical coordinate, here symbolically represented as an $s^*$ surface.

For our background state with zero meridional wind $v$ the vertical velocity for non-hydrostatic models can be expressed as

$$w = \frac{u}{a \cos \phi} \left( \frac{\partial z}{\partial \lambda} \right)_{s^*}$$ \hspace{1cm} (3)

which utilizes a spherical notation for the derivative in the zonal direction. The exact functional form for $w$ depends on the choice of $s^*$. However, for illustration purposes a concrete example is displayed below. This closed form for $w$ is shown for the height-based orography-following Gal-Chen vertical coordinate $s^* = \bar{z}$ (Gal-Chen and Somerville, 1975; Kent et al., 2014a) which is often used in non-hydrostatic models. The relationship between the geometric height $z$ and the transformed Gal-Chen coordinate $\bar{z} = z_{\text{top}} (z - z_s) (z_{\text{top}} - z_s)^{-1}$ is then given by

$$z = \bar{z} + \frac{z - z_{\text{top}}}{z_{\text{top}} - z_s} z_s (\phi, \lambda) \quad \text{with} \quad \frac{\partial z}{\partial \lambda} = \frac{\partial z_s}{\partial \lambda} \left( 1 - \frac{\bar{z}}{z_{\text{top}}} \right)$$ \hspace{1cm} (4)

where $z_{\text{top}}$ symbolizes the constant height position of the model top and $\bar{z}$ is a constant along the sloping model levels. For the mountain profile shown in Eq. (1) the vertical velocity from Eq. (3) can then be expressed as

$$w(\lambda, \phi, z) = -\frac{u(\lambda, \phi, z)}{a \cos \phi} \left( 2 h_0 \left( 1 - \frac{z}{z_{\text{top}}} \right) \sum_{n=1}^{2} \left( \frac{\partial l_n}{\partial \lambda} \right) \left( \frac{l_n(\lambda)}{c^2} \right) \exp \left[ - \left( \frac{\phi - \phi_n}{d} \right)^6 + \left( \frac{l_n(\lambda)}{c} \right)^2 \right] \right)$$ \hspace{1cm} (5)

where

$$\frac{\partial l_n}{\partial \lambda} = \begin{cases} 
1 & \text{if } d_n(\lambda) < \pi \\
-1 & \text{if } d_n(\lambda) \geq \pi.
\end{cases}$$
For the example of a Gal-Chen coordinate with $z_{\text{top}} = 31$ km, the magnitude and spatial structure of $w$ given by Eq. (5) is plotted in Fig. 3. The vertical velocity in Fig. 3 is shown for both mountains. Figure 3a shows a latitude-longitude profile at a constant geometric height of $z = 10$ km above mean sea level. Updrafts are observed on the upwind side west of the mountain peaks, and downdrafts are present on the downwind side east of the mountain peaks. Figure 3b shows a longitude-height cross section at $45^\circ$N. The Gal-Chen coordinate exhibits vertically-sloped model levels which are also present near the zonal jet maxima. This causes $w$ to achieve its peak magnitudes at the approximate height of the zonal jet. Other choices for transformed vertical coordinates are also popular which let the terrain-following characteristics decay more rapidly from the surface, such as described in Klemp (2011) for MPAS. In this case the maximum initial magnitudes of $w$ are expected to be located at a lower position in the atmosphere. If non-Gal-Chen coordinates are used the expressions (4) and (5) need to be adjusted and might no longer have closed-form analytical descriptions.

For the mountain profiles described in Eq. (1) the initial vertical velocities are small. Therefore, we suggest that it is also acceptable to start these simulation with $w = 0$ m s$^{-1}$ if the initial imbalance is tolerated by the dynamical core and numerical scheme. This is the case for MPAS. When comparing an MPAS $w = 0$ simulation with a simulation that used a numerically computed $w$ in MPAS the evolutions of the baroclinic waves were almost indistinguishable (not shown). Therefore, the initialization of the non-zero $w$ profile can likely be omitted in most models for moderately steep mountain profiles with initial vertical velocities of order $10^{-1}$ m s$^{-1}$ or smaller. The particular initialization choice for $w$ needs to be documented when using the test case for non-hydrostatic configurations. For simplicity and to ease the comparison to other non-hydrostatic dynamical cores, all MPAS examples in this paper are shown for $w = 0$ m s$^{-1}$, which gets adjusted to the expected vertical updraft and downdraft patterns over one time step without triggering numerical noise. Our chosen other three dynamical cores are hydrostatic and compute the vertical velocity as a diagnostic quantity.

### 2.4 Design considerations

The moist test variant allows the examination of the interactions between subgrid-scale physical parameterizations and the dynamical evolution of baroclinic waves. In addition, the impact of topographic forcing on both dry and moist waves can be assessed. The parameters and functional forms of the initial conditions are guided by several design considerations. As displayed in Fig. 2a the two mountain ridges are centered at $45^\circ$ N, are separated by $68^\circ$ in longitude, and have a peak height of 2000 m. The shape of each mountain is chosen to broadly resemble the mean height of real mountain ranges such as the Andes or Rocky Mountains, and to have a comparable nominal zonal extent of around 7 degrees in longitude. By design, but unlike the real mountain ranges on earth, a second ridge with an identical shape is placed to the east of the first mountain. Although a single mountain is a sufficient perturbation to the steady state to trigger a baroclinic wave, the addition of a second mountain increases the utility of the test case in several ways. For notational convenience we refer to the mountain centered at $\lambda_1 = 72^\circ$ E as Mountain 1 (M1) and the mountain centered at $\lambda_2 = 140^\circ$ E as Mountain 2 (M2). The developing baroclinic wave downwind of M1 is referred to as Wave 1, and likewise the wave downwind of M2 is referred to as Wave 2. The evolution of Wave 1 is nearly identical to the wave downwind of Wave 2 until Wave 1 is forced over M2. The evolution of Wave 1 and Wave 2 can then be directly compared to determine the impact of the topographic lifting on the evolving Wave 1.
The longitudinal offset between the two mountains was chosen so that the band of large-scale precipitation along the leading frontal zone of Wave 1 has time to reach peak intensity and length before the precipitation band is forced over M2. This is shown in Figs. 4e-h that display the precipitation rates of the evolving waves at days 3, 4, 5, and 6, respectively. By the time Wave 1 reaches M2 around day 5, it approximately satisfies the detection criteria for an atmospheric river as, e.g., specified in Hagos et al. (2015). The topographic lifting of Wave 1 occurs before the wave breaking sets in slightly thereafter (around day 5.5-6). This destroys the coherent structure of the precipitation band. A full discussion of Fig. 4 and the flow characteristics is provided in Sect. 4.1.

In the dry configuration of this test case the missing diabatic forcing from the precipitation slows down the growth rate of the waves as shown later. This means that wave breaking has not occurred yet when the dry variant of Wave 1 reaches M2. This allows high-resolution model runs to be used as a reference solution. Although mathematical convergence cannot be expected when moist physics is added, the model intercomparisons presented in Sect. 5 show that model statistics still allow insightful comparisons between the dynamical cores for up to six days.

3 Description of the Dynamical Cores

Before discussing the simulation results, we briefly introduce the four dynamical cores used in this study. Three of these dynamical cores are available as options in the CESM model (Danabasoglu et al., 2020) version 2.2 (CESM 2.2). In particular, they are embedded in the CESM 2.2 atmospheric component which is called the Community Atmosphere Model version 6 (CAM6). CAM6 includes the “Spectral Element” dynamical core SE (Taylor and Fournier, 2010; Lauritzen et al., 2018), “Finite Volume” model FV on a latitude-longitude grid (Lin, 2004), and the GFDL Finite-Volume Cubed-Sphere model FV3 as described in Harris et al. (2021). In addition, we use the MPAS dynamical core (Skamarock et al., 2012) which is available as a development version in CAM6. However, for the comparisons here, the MPAS (version 7) simulations were performed with the stand-alone version of MPAS (Jacobson et al., 2019). All simulations are performed with 30 vertical levels. The hybrid pressure-based model level positions for SE, FV, and FV3 are listed in Reed and Jablonowski (2012) and are recommended to users of this test case. The model top lies near 2 hPa which corresponds to a model top height around 35 km. MPAS uses a height-based vertical coordinate with a model top around 31 km. This position corresponds to a top pressure around 8 hPa.

The relevant configuration details as well as the namelist settings for all four dynamical core including the portfolio of the dynamics, physics, tracer, remapping, or acoustic time steps are listed in Appendix B and the Tables B1-B4.

3.1 Spectral Element (SE)

The hydrostatic SE dynamical core in CESM is documented in Lauritzen et al. (2018) and was originally designed by Taylor et al. (1997) and Taylor and Fournier (2010). The spectral element method is also used in the Energy Exascale Earth System Model (E3SM) and supports non-hydrostatic extensions (Taylor et al., 2020). The spectral finite element method is formulated on an equiangular gnomonic cubed-sphere grid. Its horizontal discretization uses a mimetic A-grid with $4 \times 4$ continuous collocation points in each spectral element (the so-called “np4” configuration). This renders the numerical scheme fourth-
order accurate in the horizontal direction. The numerical method exactly satisfies several differential identities that provide desirable conservation properties, such as the conservation of the dry air mass to machine precision. The continuous equations of motion also conserve a measure of moist total energy which accounts for all prognostic water species (Taylor, 2011). The dry air mass is used for the formulation of the orography-following pressure-based vertical $\eta$ coordinate which utilizes the Lorenz vertical staggering. The vertical discretization utilizes a floating Lagrangian coordinate similar to FV. All prognostic variables are then periodically remapped to their reference positions during a physics time step. Fourth-order hyperviscosity terms are added to the prognostic equations to prevent the accumulation of numerical grid-scale noise. The time stepping for the prognostic variables is done using a five-stage, nonlinear, third-order Runge-Kutta method.

Various physics-dynamics coupling strategies are available which are controlled by a namelist parameter se_ftype. For se_ftype=0, the forcing due to physical parameterizations is distributed in equal increments (dribbled) and added to the prognostic variables during the integration of the sub-cycled dynamical core. For se_ftype=1, all physics adjustments are added as a lump adjustment after each physics time step. In case of se_ftype=2, the forcing of mass quantities like moisture tracers are added via the se_ftype=1 sudden adjustment strategy, while all other forcings for the, e.g., temperature or velocities are dribbled in (se_ftype=0). This option is considered the “hybrid” option. We use se_ftype=0 except in section 5.3 where the impact of the SE default se_ftype=2 is demonstrated.

### 3.2 Finite-Volume (FV)

The FV dynamical core solves the hydrostatic primitive equations on a latitude-longitude grid using a flux-form semi-Lagrangian scheme and a floating Lagrangian vertical coordinate (Lin, 2004). It utilizes the Piecewise Parabolic Method (Colella and Woodward, 1984) to represent sub-grid flux distributions and is horizontally third-order accurate. The horizontal discretization uses a combined C–D grid staggering. The vertical treatment allows several Lagrangian dynamics steps to be taken before remapping the vertical levels to a reference grid. Nonlinear limiters within the finite volume method introduce implicit diffusion. Explicit fourth-order horizontal divergence damping is added to the model to prevent the accumulation of energy at the grid scale (Whitehead et al., 2011). Second-order horizontal divergence damping is applied in the top layers to decrease the impact of wave reflection from the model top. The dynamics are integrated on a shorter sub-cycled time step than the physics time-step, and forcing due to microphysics is added to the prognostic variables as a lump adjustment after the physics time step (Neale et al., 2010).

### 3.3 Finite-Volume on a Cubed Sphere (FV3)

The FV3 dynamical core (Harris et al., 2021) has been originally developed by NOAA’s Geophysical Fluid Dynamics Laboratory and now serves as the fluid dynamics backbone of NOAA’s “Unified Forecast System” (UFS) for weather prediction applications in the U.S.. It shares many characteristics of the FV dynamical core. FV3 is a finite-volume model that can solve either the hydrostatic primitive equations or the non-hydrostatic shallow-atmosphere equations on an equiangular gnomonic cubed-sphere grid. Here, the hydrostatic version is chosen as implemented in CESM 2.2. Like FV, FV3 uses the Piecewise Parabolic Method on a C-D grid (Lin and Rood, 1997; Putman and Lin, 2007, 2009) and is horizontally third-order accurate.
A floating Langrangian vertical discretization is used. The cubed-sphere grid reduces the numerical difficulties posed by the pole point singularities in the FV latitude-longitude grid. To prevent the accumulation of noise at the grid scale, 6th-order horizontal divergence damping is activated. In addition, monotonicity constraints are used in the horizontal advection and vertical remapping algorithms which implicitly adds viscosity to the model. As in FV, a second-order divergence damping mechanism is utilized as a sponge layer near the model top. In addition, Rayleigh friction is applied to the horizontal wind velocities in the sponge layer if the model level pressure is less than 7.5 hPa. In our L30 configuration this only affects to topmost full model level. The maximum relaxation time is set to 10 days at the model top.

### 3.4 Model for Prediction Across Scales (MPAS)

MPAS (Skamarock et al., 2012) is a finite-volume model that solves the non-hydrostatic shallow-atmosphere equations. The horizontal discretization is built upon a centroidal Voronoi tesselation mesh with a staggered C-grid and is designed to use the mimetic so-called TRiSK discretization (Thuburn et al., 2009; Ringler et al., 2010). Horizontal advection is nominally third-to-fourth-order accurate. The vertical dimension is treated with a second-order finite volume method with a smoothed terrain-following geometric height coordinate as specified in Klemp (2011). Various smoothing options are available for the orography-following vertical coordinate called $\zeta$ which impact the accuracy of the numerical scheme. In our MPAS model simulations, we do not activate the smoothing and therefore convert to the Gal-Chen configuration shown in Eq. (4) with $\zeta = \zeta$. MPAS has several diffusion options to damp numerical noise, including a Smagorinsky-type eddy viscosity, fourth-order hyperdiffusion, and 3D divergence damping. Our MPAS model integrations are configured to use the Smagorinsky-type diffusion. A detailed discussion of MPAS’ treatment of the physics tendencies can be found in Klemp et al. (2007).

### 4 Characteristics of the Test Case

For demonstration purposes, the evolution of the baroclinic wave is first discussed for the Spectral Element dynamical core. Any other dynamical core could have been picked. The simulations were run with nominal grid spacings of $1^\circ$ (ne30), $0.5^\circ$ (ne60), $0.25^\circ$ (ne120), and $0.125^\circ$ (ne240) and 30 vertical levels where the “neXXX” notation refers to the number of supporting spectral elements in the horizontal direction per cubed-sphere face. For example, the ne30 setting has $30 \times 30$ supporting elements per cubed-sphere face. In addition, the chosen “np4” configuration provides $3 \times 3$ extra degrees of freedom per element. Therefore, the grids mentioned above have nominal geometric grid spacings of 100 km, 50 km, 25 km, and 12.5 km, respectively. As mentioned before, our SE simulations used the se_f_type=0 physics-dynamics strategy which deviates from the SE default se_f_type=2. The latter default setting is explored in Sect. 5.3.

#### 4.1 Baroclinic instability

Baroclinic instability is a crucial driver of weather systems in the midlatitudes. A systematic treatment of this phenomenon from the viewpoint of quasi-geostrophic theory can be found in, e.g., Holton (1992). Because the initial conditions in this test case are baroclinically unstable, each mountain triggers a synoptic-scale wave which develops downwind of the topographic forcing.
Figure 4. Latitude-longitude cross-sections of the baroclinic waves in the SE dynamical core on a 0.5° degree grid at days 3, 4, 5, and 6 (from left to right). Top row: mean sea level pressure, middle row: precipitation rate, bottom row: 850 hPa temperature. The location of the mountain ridges is indicated by the contour lines.

Each wave exhibits characteristics of baroclinic waves in the real atmosphere. For example, strong temperature gradients develop ahead of the synoptic-scale low pressure systems which trigger strong precipitation bands along these frontal zones.

Figure 4 illustrates the time evolution of the mean sea level pressure (MSLP), precipitation rate, and 850 hPa temperature for the moist baroclinic wave at days 3, 4, 5, and 6. In particular, Figs. 4a-d show the intensifying low and high MSLP systems that develop behind both mountains. At day 4 the two developing low pressure systems are nearly identical. At day 5 topographic forcing begins to impact Wave 1 as Wave 1 is forced over M2. By day 6 topographic forcing has caused significant deviation between the structure of the two waves. Figures 4e-h illustrate the development of the large-scale precipitation bands. These are fed by the high moisture that is transported from the tropical regions by the developing waves. At day 4 the bands begin to form to the east of the low pressure system. At day 5 the precipitation band associated with Wave 1 is forced up over the mountain. By day 6 the topographic forcing has significantly disrupted the structure of the precipitation band associated with Wave 1 compared to the precipitation associated with Wave 2. Figures 4i-l show the evolution of the synoptic-scale temperature fronts at 850 hPa. Note that this 850 hPa position represents an interpolated level that made use of extrapolations in the neighborhood of the mountain peaks. Nevertheless, we selected this low-lying level for the analysis as the wave signatures lose their sharpness with increasing altitude. These temperature fronts are driven by the transport of warm, moist, equatorial air into the midlatitudes which in turn drives updrafts and the development of the intense precipitation bands. The exponentially growing mode triggered...
by the addition of topography is well-resolved in horizontal grids with a 1° grid spacing. The agreement across resolutions breaks down in the moist case when wave breaking becomes dominant beyond day 6 (not shown).

In addition to the qualitative characteristics, several quantitative metrics are also assessed. Common quantities for assessing the development of baroclinic waves are the time evolution of the minimum MSLP and the Eddy Kinetic Energy (EKE) (Lorenz, 1955; Simmons and Hoskins, 1978; Pavan et al., 1999; Ullrich et al., 2014; Kurowski et al., 2015). Minimum MSLP measures the intensity of the most developed eddy and is calculated point-wise on an interpolated latitude-longitude grid. EKE measures the evolution of the kinetic wave energy relative to the background flow. It is computed by first subtracting the initial base state from the horizontal wind velocities $u$ and $v$ at each time slice. Then the point-wise kinetic energy of these eddy wind fields is calculated and integrated over the entire volume of the atmosphere. The calculation can be conducted in either height $z$ or pressure $p$ coordinates via

$$
EKE(t) = \frac{1}{4\pi a^2} \int_{z_s}^{z_{top}} \int_{A} \left( \frac{1}{2} \left( (u - \bar{u})^2 + (v - \bar{v})^2 \right) \rho \right) dA dz
$$

(6)

$$
= \frac{1}{4\pi a^2 g} \int_{p_{top}}^{p_s} \int_{A} \left( \frac{1}{2} \left( (u - \bar{u})^2 + (v - \bar{v})^2 \right) \right) dA dp
$$

(7)

where $A$ denotes the area of a grid cell and $\rho$ is the density of the air. The symbols $z_{top}$ and $p_{top}$ denote the height and pressure at the model top, respectively. The calculation only takes the horizontal velocities and their initial states $\bar{u}$ and $\bar{v}$ at each grid point into account, and measures the EKE in units of J m$^{-2}$.

In the dry adiabatic configuration point-wise convergence of EKE can be expected as the horizontal grid spacing is decreased. As was argued in Jablonowski and Williamson (2006) and Ullrich et al. (2014), this empirical point-wise convergence allows high-resolution model integrations to be used as reference solutions even when a closed-form solution for the evolution of the
Figure 6. Time series of baroclinic wave summary statistics from the moist SE model at nominal $1^\circ$ (ne30), $0.5^\circ$ (ne60), $0.25^\circ$ (ne120), $0.125^\circ$ (ne240) grid spacings. (a) Eddy Kinetic Energy, and (b) point-wise minimum MSLP which is a proxy for the amplification of the baroclinic wave.

In the moist variant of the test case, the thermodynamic forcing caused by large-scale precipitation intensifies the development of the wave. Figure 6 shows the calculated minimum MSLP and EKE for the moist SE model integrations for decreasing grid spacings. Unlike the dry case (Fig. 5), Fig. 6b illustrates that the minimum sea level pressure in the highest-resolution simulation diverges significantly from the lowest-resolution simulation once precipitation sets in between days 3 and 4. Figure 6a shows a time series of the integrated EKE. This demonstrates that the divergence of higher-resolution from lower-resolution model runs occurs over the whole structure of the wave, and the resolution dependence is not limited to the gridpoint at which MSLP is lowest. The EKE time series only illustrates the initial growth phase of the baroclinic wave. Saturation of the EKE...
values occurs later around day 10 with peak EKE values around $2 \times 10^5$ J m$^{-2}$ which compare well to the peak EKE values in Pavan et al. (1999).

The strength of the eddy moisture flux convergence, which drives large-scale precipitation, is well-correlated with the diabatic heating in idealized studies of baroclinic modes (Pavan et al., 1999). This diabatic heating speeds up the growth rate of the wave. The correlation can be seen by examining the temperature anomaly, which is defined as the difference between the temperature at a given time and the base state temperature. In addition, the vertical pressure velocity $\omega$ serves as a suitable proxy for the wave activity. The longitude-height cross sections of these fields are displayed at day 4 in Fig. 7 for both the dry and moist model integrations. Figure 7c shows that in the moist version, updrafts due to precipitation follow the progres-

![Longitude-height cross-sections](https://doi.org/10.5194/egusphere-2023-376)

**Figure 7.** Longitude-height cross-sections at 45° N of the (top) temperature perturbation and (bottom) vertical pressure velocity $\omega$ for the (a,c) moist and (b,d) dry atmosphere. SE ne60 (50 km) model integrations at day 4 are shown.
sion of the temperature front. The updrafts are significantly larger in the moist case than in the dry case as shown in Fig. 7c in the neighborhood of the frontal zone around 125° E. In addition, the ω patterns clearly highlight the hydrostatic, mostly upward-propagating inertia-gravity wave oscillations downwind of M2 near 140° E. These mountain wave patterns resemble the stationary hydrostatic inertia-gravity wave solutions from 2D x-z slice models on constant f–planes when tested with bell-shaped mountains (Dudhia, 1993; Ullrich and Jablonowski, 2012a). However, the spatial scale of the mountain-generated gravity waves in SE with full rotational effects is larger than that of the 2D models which is a result of the different model setups. Figures 7a-b display the distribution of the temperature perturbation at day 4. The moist configurations in Fig. 7a shows that the diabatic forcing, that is triggered by the precipitation combined with the induced updrafts, places the maximum positive temperature perturbation several kilometers into the atmosphere. The maximum positive temperature perturbation in the dry case (Fig. 7b) is located at the surface. The dependence of the wave intensification on the model resolution in Fig. 6 can be explained by noting that the maximum intensity of the extreme precipitation within the moisture bands increases as horizontal grid spacing decreases. We cannot expect point-wise convergence as grid spacing is decreased due to the nonlinearity of this forcing. However, it is reasonable to compare the statistics of the precipitation between models in the absence of point-wise convergence.

It was argued by Chen and Knutson (2008) that parameterizations of large-scale precipitation are best understood as an area-average of the precipitation within a grid cell. Under this interpretation, it is expected that different dynamical cores with comparable nominal grid spacings should have similar precipitation statistics, e.g. when assessing the accumulated precipitation. This holds even when point-wise convergence is not observed within a particular dynamical core as the grid spacing decreases.

When moisture is present, the evolving circulation around the low pressure systems induces moisture transport from the equatorial region. The circulation around the developing low pressure systems creates bands of extreme precipitation to the east of the low pressure centers. The spatial extent of these precipitation bands reaches as far as 60° N. In addition, the bands reach length scales of several thousand kilometers before wave breaking sets in (Fig. 4g). The precipitation bands are characterized by high values of the vertically integrated vapor transport (IVT) and the column-integrated precipitable water (CPW) which are standard detection criteria for atmospheric rivers (Hagos et al., 2015). In particular, the IVT and CPW values in the SE (ne60) simulation at day 5 exceed 1000 kg m\(^{-1}\) s\(^{-1}\) and 4 cm, respectively. These values lie well above the typical IVT and CPW detection thresholds of 250 kg m\(^{-1}\) s\(^{-1}\) and 2 cm for atmospheric rivers. Here, the quoted SE value for IVT represents the vapor transports at low levels (below 600 hPa) and the CPW was integrated over the entire vertical column (see also Hughes and Jablonowski (2021)). The precipitation bands therefore resemble atmospheric rivers as suggested earlier. In addition, the leading band of Wave 1 is orographically lifted over M2 at day 5, which mimics the impacts of the mountain ranges on atmospheric rivers along the U.S. west coast. Although the bands are narrow, the geographic distribution is well resolved even at the coarsest 1° horizontal grid spacing. Because any sources of moisture are omitted in our simulations, water exits the atmosphere when precipitation occurs. It is not replenished by surface fluxes of latent heat. Such a configuration with idealized surface fluxes represents a natural extension of the test case complexity as also described in Reed and Jablonowski (2012) and Thatcher and Jablonowski (2016), but is not considered here.
The presence of orography forces an upward motion of the precipitable water at day 5, thereby intensifying the precipitation rate as displayed in Fig. 4g. The comparison of the leading precipitation band triggered by M1 and the band triggered by M2 at day 6 (Fig. 4h) shows the reduction of the precipitation rate in Wave 1. This is caused by the orographic forcing of M2 which diminished the Wave 1 moisture pool in comparison to Wave 2. Furthermore, Fig. 4d illustrates that the interaction between the precipitation band and M2 slows down the intensification of the dominant Wave 1 low pressure system.

5 Selected Dynamical Core Intercomparisons

Besides SE we also tested the moist variant of the test case with FV, FV3, and MPAS to conduct a brief, non-exhaustive dynamical core intercomparison. Here, we provide selected snapshots of this intercomparison to highlight the capabilities of the test case. The simulations are conducted with 30 vertical levels and a nominal 0.5° grid spacing in all dynamical cores, which are labeled as “ne60” for SE, “FV05” for FV, “C192” for FV3, and “60km” for MPAS. The SE, FV3, and MPAS dynamical core analyses utilize model data on interpolated latitude-longitude grids which have uniform 0.5° × 0.5° horizontal grid spacings. The FV05 simulation uses the grid spacings 0.47° × 0.625° for its latitude-longitude grid.

5.1 Baroclinic wave metrics

Quantitative metrics can be used to compare the strength of an evolving baroclinic wave across dynamical cores. Although Fig. 6 shows a significant dependence of the wave intensification to the horizontal grid spacing in the SE model, comparisons can be done across dynamical cores if the horizontal grid spacings are comparable. Figure 8 shows a time series of the evolution of (a) Eddy Kinetic Energy integrated over the entire volume of the model domain, and (b) the point-wise minimum MSLP. The time series for the corresponding dry SE 0.5° model integration is shown in black.
Figure 9. Intercomparison of the precipitation bands at day 5 when Wave 1 is being orographically lifted. The most intense precipitation rate is observed on the upwind side of M2. However, the precipitation at the leading edge of Wave 1 on the downwind slope is still significantly more intense than the precipitation rate in the leading edge of Wave 1. The locations of the mountain ridges are marked by the light contours.
Wave 2. The differences in the flow patterns are amplified by the highly nonlinear forcing provided by the combination of the topographically induced vertical motion and the diabatic forcing resulting from the increase in precipitation. We observe that the dynamical cores differ in their ability to keep the long precipitation bands together as coherent structures before wave breaking processes break them up after day 5. For example, the precipitation bands in SE and MPAS in Fig. 9 already start developing small-scale but intense precipitation patches at day 5 that got separated from the main bands. These patches start to resemble so-called “grid-point storms” which are characterized by intense, truncation-scale storms with extreme updraft speeds and precipitation rates as analyzed in Williamson (2013). The coherent precipitation patches in FV and FV3 also break up as a result of wave breaking and stretching, but this happens slightly later. The reasons for these differences are complex and an in-depth analysis is beyond the scope of this paper. However, the differences are likely caused by a combination of the following factors: insufficient resolution to represent the thin bands, the simplicity of the precipitation scheme, and the choice of the physics and dynamics time steps. These factors are tightly coupled to the differences in the diffusion characteristics and the associated so-called “effective resolutions” of the dynamical cores (see also Jablonowski and Williamson (2011) and Kent et al. (2014b, c)), and the physics-dynamics coupling strategies (see also Gross et al. (2016, 2018)). The physics-dynamics coupling aspect will be briefly highlighted for the SE dynamical core in Sect. 5.3 which sheds light on additional application areas for the test case.
Figure 11. Time series of the accumulated precipitation integrated between $60^\circ - 300^\circ$ E and $0^\circ - 90^\circ$ N.

Figure 10 illustrates the MSLP at day 5 in each model as the wave is forced over M2. All models exhibit qualitative agreement in the overall structure of the low and high pressure systems. The most obvious difference is that the MPAS high pressure systems with MSLP values over 1010 hPa occupy visibly larger areas.

5.2 Precipitation and diabatic forcing

We treat the precipitation rate from the Kessler physics routine as an area-average over a grid cell. Using this interpretation the area-integrals of the precipitation rate over a selected region should be comparable even in the presence of large differences between the precipitation rates at individual grid points across the dynamical cores. Figure 11 shows a time series of the accumulated precipitation integrated between $60^\circ$E and $300^\circ$E in the northern hemisphere. The accumulation in the FV dynamical core is notably higher than the accumulation in the other dynamical cores. This holds even before day 3, when the precipitation bands along the developing frontal zones start to form. In FV, stationary orographic rain over the mountain tops is already present before hour 12. Other dynamical cores, like FV3, start the stationary orographic rain around hour 36. The reasons for these differences are not entirely clear. They are likely linked to the FV diffusion characteristics which also caused the time evolution of the integrated EKE and minimum MSLP to differ in Fig. 8.

Figure 12 intercompares the longitude-height cross sections at $45^\circ$ N of the temperature anomaly, the vertical pressure velocity, and the cloud liquid water mixing ratio as the Wave 1 precipitation band travels across the downwind slope of M2 at day 5. In particular, Figs. 12a-d show the temperature perturbation in the four dynamical cores. In spite of the grid-scale differences between models in Fig. 9, the temperature structure over the mountain is qualitatively similar across dynamical cores. The MPAS model exhibits the largest deviation from the base temperature profile. Figures 12e-h illustrate the vertical pressure velocities over the mountain at day 5. On purpose, the color range for $\omega$ slightly saturates to highlight to spatial patterns.
Figure 12. Intercomparison of longitude-height cross-sections of the (top) temperature perturbation and (middle) vertical pressure velocity $\omega$ and (bottom) cloud liquid water mixing ratio at day 5. The columns correspond to each dynamical core with a nominal 0.5° grid spacing as marked. Latitude is held constant at 45° N in SE, FV3, and MPAS, and at 44.88° N in FV (the closest grid point to 45° N). The outline of mountain M2 is shown near 140° E along the x-axis (longitudes).

and match the color scheme of Fig. 7 (at day 4) while not displaying the actual $\omega$ minima and maxima at day 5. Overall, the vertical patterns of $\omega$ are qualitatively similar in all dynamical cores but with differences in the peak magnitudes. The figure suggests that MPAS exhibits the most intense up- and downdrafts, closely followed by FV3 which mimics the relative strength of the temperature anomalies in Figs. 12c,d. The vertical velocity and temperature anomaly patterns and magnitudes are tightly connected to the precipitation rates of the baroclinic rainbands at 140° and 210° (= 150° W in Fig. 9). Intense updraft areas are co-located with the rain bands. Therefore, the varying magnitudes of the updrafts help explain the local differences in the precipitation rates.

Figure 12i-l displays the distributions of the cloud liquid water mixing ratios that serve as the reservoir for rain water and precipitation in the Kessler warm-rain physics scheme. Overall, the cloud liquid water patterns resemble each other, but the details vary greatly. For example, the maximum cloud liquid water mixing ratios in MPAS near 140° and 210° are located
at lower altitudes under 6 km whereas the peak cloud water regions in SE, FV, and FV3 are mostly found at heights around 9-10 km. However, this might not explain the precipitation differences as the majority of the precipitation forms below 6 km. The latter is indirectly depicted by the positive temperature anomaly patterns which also capture the diabatic heating effects due to precipitation. The positive temperature anomaly maxima near the rain bands at 140° and 210° are confined to regions under 6 km. MPAS exhibits the biggest heating signals among the four dynamical cores. It is also interesting to observe that FV3 develops two low-lying and small cloud water clusters near 125° and 195°. These are sensitive to the numerical diffusion settings in FV3 and do not appear in more diffusive FV3 configurations (not shown). As an aside, the FV simulations has difficulty keeping the cloud liquid water mixing ratio pattern concise near the 210° rain band.

5.3 Additional application aspects: Physics-dynamics coupling

The following brief discussion focuses on the physics-dynamics coupling strategy in SE and highlights an additional application area of the test case. The discussion refers back to the various physics-dynamics coupling choices for the SE model that are determined by the namelist input variable se_ftype in CESM 2.2. The relevant namelist settings for se_ftype were briefly explained in Sect. 3.1. Here, we shed light on the CESM 2.2 default setting se_ftype=2 which was not used for the other plots in this paper.

se_ftype=2 is the hybrid physics-dynamics coupling option in SE. It uses sudden adjustments of the moisture and mass fields after the physics time step (900 s in our case) and the dribbling strategy for all other physical forcings. In the chosen example at the nominal 0.5° resolution SE’s subcycled dynamics time step is 150 s. However, this strategy triggers spurious numerical noise (ringing) in SE which we analyze via the vertical pressure velocity. Figure 13 show snapshots of the 850 hPa vertical pressure velocity \( \omega \) at day 5 for all four dynamical cores. All dynamical cores show small-scale gravity wave activity which is caused by the mountains. These are physical waves and not the focus here. Note that we chose to saturate the color scale to draw attention to numerical artifacts. These are otherwise difficult to detect.

Figure 13a demonstrates the presence of the numerical ringing in SE which becomes more severe as the precipitation bands mature and the diabatic forcings get stronger. The ringing appears in concentric circles and likely originates from small hotspots with strong diabatic forcing, such as grid-point storms. The magnitude of the numerical noise is small in comparison to the vertical velocities caused by the baroclinic wave and the mountain-generated gravity waves. However, vertical velocities are tightly coupled to cloud and rainfall characteristics. Any numerical interference in this relationship is undesirable and could lead to artificial responses in the physical parameterization.

The grid-scale oscillations occur due to the sudden moisture adjustments present with SE’s se_ftype=2 strategy. The ringing is a characteristic of SE’s numerical approach which utilizes a continuous Galerkin technique for the horizontal discretization. This phenomenon in models with local or global spectral numerical schemes is also known as Runge’s phenomenon or Gibb’s ringing. It resembles a shock wave that appears when large and unbalanced physical forcings are added to the rather balanced motions in the dynamical core. The SE dynamical core is a highly accurate model with very low intrinsic dissipation. It becomes apparent that the explicitly-added diffusion mechanisms in the SE dynamical core are not effective enough to suppress these oscillations. Therefore, the occurrence of such noise is also tightly linked to the implicit-numerical and explicitly-added
diffusion characteristics of dynamical cores. When changing SE’s physics-dynamics coupling strategy to se_f_type=0, so that all physics tendencies are dribbled in with the subcycled 150 s dynamics time step, the spurious oscillations disappear. This was, for example, demonstrated in Hughes and Jablonowski (2021) who compared the default se_f_type=2 to an SE configuration with identical physics and dynamics time steps of 150 s. The reduction of the time step length in the physics reduces the strength of the physical forcings. This leads to more gentle adjustments of the dynamical core and avoids the numerical ringing.

Our analysis indicates that there is a downside when using the SE default physics-dynamics coupling strategy se_f_type=2, which was also reported for more complex, but still idealized, SE simulations in Thatcher and Jablonowski (2016). However, the dribbling strategy se_f_type=0 is also not free of numerical artifacts. When dividing the tracer tendencies for the mass quantities across dynamics substeps it can result in negative tracer values, such as negative moisture (Lauritzen, personal communication, 2022). Although this is a rare circumstance, negative moisture values are unphysical and lead to problems in the physical parameterization schemes if not filtered out beforehand via tracer mass fixers. More studies will be needed to diagnose the ringing phenomenon in more complex model configurations and determine the best way to mitigate it.

As an aside, Fig. 13a also demonstrates another unique behavior of the SE dynamical core. When comparing the vertical pressure velocity fields, SE shows very different patterns between 30° – 90° E just west of mountain M1. This is signature of a horizontally traveling hydrostatic acoustic mode (a “Lamb wave”) which gets initially excited in all dynamical cores due to the
slight imbalance of the initial fields near the topography. The FV, FV3 and MPAS dynamical cores damp out the Lamb wave rather efficiently after about 1-2 days. However, this is not the case in SE that propagates the Lamb wave rather persistently around the sphere with a phase speed around 330 m s$^{-1}$. This topic will be discussed in a future paper.

6 Conclusions

This paper enhances the suite of idealized test cases for the dynamical cores of AGCMS in spherical geometry. It is the first dynamical core test case that combines a complex initial flow field, such as the base condition for a baroclinically-unstable wave with varying stratification and vertical wind shear, with idealized topographic barriers on a rotating regular-size earth. Both dry and idealized moist test configurations are suggested for both pressure-based or height-based dynamical cores. The moist configuration utilizes a warm-rain Kessler physics parameterization that triggers precipitation and thereby provides diabatic forcing. The test accommodates the portfolio of hydrostatic and non-hydrostatic as well as shallow-atmosphere and deep-atmosphere dynamical core designs. In particular, we add an analytically-defined topography profile to an existing baroclinic wave base state. This necessitates the re-balancing of the initial conditions with a particular focus on the surface pressure and vertical velocity fields. The latter is only needed for non-hydrostatic models. The resulting initial conditions are well-balanced, but not a steady-state solution. The topography field acts as the trigger for rapidly growing baroclinic waves over the course of several days.

The test case provides a controlled environment which serves two purposes. First, it can be used as a model assessment, debugging, or tuning tool by model developers who have a need to assess the inclusion of topography and the chosen vertical coordinate in a dynamical core. This informs numerical design decisions for dynamical cores and their physics-dynamics coupling strategy, and contributes to dynamical core model intercomparisons. Second, the test case also serves the atmospheric dynamics science community. It can be used as a tool in the atmospheric dynamics toolbox, and sheds, for example, light on the impact of mountains on the general circulation. All mountain shapes can be accommodated as long as they are prescribed via analytic functions. An analytical solution does not exist. However, high-resolution reference solutions and dynamical core intercomparisons can be used to gain confidence in the model simulations. This is straightforward in dry configurations that converge with increasing resolution before non-linear wave breaking and mixing processes set in after day 6.5. However, moist configurations are impacted by the nonlinear forcing from the stationary orographic rain and, most importantly, the precipitation along the frontal zones from day 3 onwards. This leads to an increased spread in the model simulations that typically exhibit wave breaking after day 5 for the chosen topographic profile.

We illustrated the characteristics and capabilities of the test case via example simulations with various dynamical cores which are available at NCAR. These are the SE, FV, and FV3 dynamical cores of the CESM 2.2 model framework, and the standalone distribution of MPAS version 7. The dynamical characteristics of the topographically triggered baroclinic waves were studied and a model intercomparisons was performed. Real-world flow phenomena and mountain shapes were used to inspire the selection of the flow parameters, such as the shape of the two chosen ridge mountains. These triggered baroclinic waves that have similarities with atmospheric rivers. The chosen examples showcase the potential use cases of this test case.
Besides serving as a debugging tool, we briefly discussed the impact of diffusion on the flow and precipitation characteristics. Furthermore, the test revealed physics-dynamics coupling problems in the SE dynamical core. It also led to the discovery of an acoustic mode in the SE model that persistently propagates in the horizontal direction without much damping. Overall, the overall flow patterns in the dynamical core simulations resembled each other when evaluating selected quantitative metrics. However, the details can differ greatly at the local level. The results suggest that the evolution in FV was noticeably impacted by its diffusion characteristics.

There are several future directions for this research. First, the underlying atmospheric base state has both a shallow atmosphere and a deep atmosphere variant. So far, we tested the shallow-atmosphere variant and are intrigued to further apply the test to deep-atmosphere dynamical cores which become popular. Second, the test can also utilize a reduced-radius configuration to trigger non-hydrostatic model responses more easily. However, care will need to be taken, as the base state can develop statically-unstable regions depending on the radius reduction factor $X$. This was discussed in Skamarock et al. (2021) who suggested slight adjustments of the base flow to avoid unstable regions. Third, the model integrations with the SE dynamical core revealed that the dynamical core preserves a rapidly propagating acoustic-gravity (Lamb) mode. Although this mode is initially present in other dynamical cores due to slight imbalances of the initial conditions, it is rapidly damped in all tested dynamical cores except SE. A systematic analysis of this phenomenon is deferred to a future publication. In addition, it will be interesting to investigate the impact of implicit-numerical and explicitly-added diffusion on the evolution of the orographically-triggered baroclinic waves in a systematic way.

**Appendix A: Description of the Initial State**

This appendix presents selected equations for the moist initial state in a shallow-atmosphere configurations which were introduced in Ullrich et al. (2016). The equations that contain the adjustments for the topographic profile are discussed in Sect. 2.2. These topographic adjustments enter the equations via the height variable. Users of deep-atmosphere dynamical cores are advised to review the needed slight adjustments which are outlined in Ullrich et al. (2014). The equations below do not formally specify a dependence on the longitude. However, this dependence is implicit as the height $z$ and pressure $p$ along a model level are now functions of both horizontal directions over the topography. Table A1 lists all parameters and physical constants for the initial conditions, including an optional small-earth scaling factor $X$. It is set to an unscaled value of $X = 1$ here, but could be varied in the future to trigger non-hydrostatic model responses. Note that such scaling reduces the earth’s radius and speeds up the earth’s rotation at the same time to keep the Rossby number constant. Other changes are also needed for reduced-radius experiments as further explained in Ullrich et al. (2016). The implementation details for the initial conditions and the CESM 2.2 & MPAS simulations are provided in Appendix B.
Table A1. Parameters and physical constants for the initial conditions.

<table>
<thead>
<tr>
<th>Variable Name</th>
<th>Variable Description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$X$</td>
<td>Reduced-size planet scaling factor</td>
<td></td>
</tr>
<tr>
<td>$a$</td>
<td>Scaled radius of the earth</td>
<td>$6.37122 \times 10^6 \text{ m} \cdot X^{-1}$</td>
</tr>
<tr>
<td>$\Omega$</td>
<td>Scaled angular speed of the earth</td>
<td>$2 \pi \cdot X^{-1}$</td>
</tr>
<tr>
<td>$g$</td>
<td>Gravity</td>
<td>$9.80616 \text{ m s}^{-2}$</td>
</tr>
<tr>
<td>$R_d$</td>
<td>Dry air gas constant</td>
<td>$287.042311365 \text{ J kg}^{-1} \text{ K}^{-1}$</td>
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<tr>
<td>$p_0$</td>
<td>Reference pressure</td>
<td>$10^5 \text{ Pa}$</td>
</tr>
<tr>
<td>$b$</td>
<td>Jet half-width parameter</td>
<td>$2$</td>
</tr>
<tr>
<td>$K$</td>
<td>Power used for temperature field</td>
<td>$3$</td>
</tr>
<tr>
<td>$T_E$</td>
<td>Reference surface temperature at the equator</td>
<td>$310 \text{ K}$</td>
</tr>
<tr>
<td>$T_P$</td>
<td>Reference surface temperature at the poles</td>
<td>$240 \text{ K}$</td>
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<tr>
<td>$\Gamma$</td>
<td>Temperature lapse rate</td>
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</tr>
<tr>
<td>$\phi_w$</td>
<td>Specific humidity latitudinal width parameter</td>
<td>$40^\circ$</td>
</tr>
<tr>
<td>$p_w$</td>
<td>Specific humidity vertical pressure width parameter</td>
<td>$3.4 \times 10^4 \text{ Pa}$</td>
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<tr>
<td>$q_0$</td>
<td>Maximum specific humidity</td>
<td>$0.018 \text{ kg kg}^{-1}$</td>
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<tr>
<td>$q_t$</td>
<td>Specific humidity above artificial tropopause</td>
<td>$0 \text{ kg kg}^{-1}$</td>
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<tr>
<td>$p_t$</td>
<td>Pressure at artificial tropopause</td>
<td>$1.5 \times 10^4 \text{ Pa}$</td>
</tr>
</tbody>
</table>

A1 Temperature base state

The temperature equation is a particular form of the temperature family given in Staniforth and White (2011). This has a variation in the meridional direction which is determined by the parameter $K$:

$$I_T(\phi) = (\cos \phi)^K - \frac{K}{K+2} (\cos \phi)^{K+2}.$$

In addition, two height-dependent functions are needed. They are given by

$$\tau_1(z) = \frac{1}{T_0} \exp \left( \frac{\Gamma z}{T_0} \right) + \left( \frac{T_0 - T_P}{T_0 T_P} \right) \left[ 1 - 2 \left( \frac{z g}{b R_d T_0} \right)^2 \right] \exp \left[ - \left( \frac{z g}{b R_d T_0} \right)^2 \right]$$

and

$$\tau_2(z) = \frac{K+2}{2} \left( \frac{T_E - T_P}{T_E T_P} \right) \left[ 1 - 2 \left( \frac{z g}{b R_d T_0} \right)^2 \right] \exp \left[ - \left( \frac{z g}{b R_d T_0} \right)^2 \right]$$

with the vertical lapse rate $\Gamma$, and the meridional temperature gradient. The latter is expressed via the equatorial and polar temperature parameters $T_E$ and $T_P$, respectively. The parameter $T_0 = \frac{1}{2} (T_E + T_P)$ denotes the arithmetic mean.

In order to incorporate water vapor into our base state, we first specify the virtual temperature $T_v$ as follows

$$T_v(\phi, z) = \frac{1}{\tau_1(z) - \tau_2(z) I_T(\phi)}$$
which obeys the thermal wind balance. The prognostic temperature initialization $T$ is therefore

$$T = \frac{T_v}{1 + M_v q} \quad \text{(A1)}$$

with $M_v = 0.608$. The symbol $q$ denotes the specific humidity as explained below.

### A2 Zonal wind base state

The zonal wind and virtual temperature are connected via the thermal wind balance. The dependence on $T_v$ is sequestered in the auxiliary quantity:

$$U(\phi, z) = \frac{gK}{a} \tau_{\text{int},2}(z) \left[ (\cos \phi)^{K-1} - (\cos \phi)^{K+1} \right] T_v(\phi, z)$$

from which we can derive the prognostic zonal wind initialization as

$$u(\phi, z) = -\Omega a \cos(\phi) + \sqrt{(\Omega a \cos(\phi))^2 + a \cos(\phi) U(\phi, z)}. \quad \text{(A2)}$$

### A3 Meridional wind base state

The meridional wind is set to zero with

$$v \equiv 0 \text{ m s}^{-1}. \quad \text{(A3)}$$

### A4 Pressure and density base state

The pressure distribution is determined by

$$p(\phi, z) = p_0 \exp \left[ -\frac{g}{R_d} \right( \tau_{\text{int},1}(z) - \tau_{\text{int},2}(z) I_T(\phi) \right) \quad \text{(A4)}$$

where the integral of the height-dependent functional forms for temperature are given by

$$\tau_{\text{int},1}(z) = \frac{1}{\Gamma} \left[ \exp \left( \frac{z}{T_0} \right) - 1 \right] + z \left( \frac{T_0 - T_P}{T_0 T_P} \right) \exp \left[ -\left( \frac{zg}{b R_d T_0} \right)^2 \right]$$

and

$$\tau_{\text{int},2} = \frac{K + 2}{2} \left( \frac{T_E - T_P}{T_E T_P} \right) z \exp \left[ -\left( \frac{zg}{b R_d T_0} \right)^2 \right].$$

The surface pressure $p_s$ is then provided when plugging in the topographic height $z_s$ (Eq. (1)) into Eq. (A4) as shown in Eq. (2).

Using the virtual temperature equation, the density is determined by the ideal gas law

$$\rho = \frac{p}{R_d T_v}. \quad \text{(A5)}$$
A5 Specific humidity base state

With the help of the auxiliary quantity \( \eta \)

\[
\eta(\phi, z) = \frac{p(\phi, z)}{p_0}
\]

the initial distribution of the specific humidity is given by

\[
q(\phi, \eta) = \begin{cases} 
q_0 \exp \left[ - \left( \frac{\phi}{\phi_w} \right)^4 \right] \exp \left[ - \left( \frac{(\eta-1)p_0}{p_w} \right)^2 \right], & \text{if } \eta > \frac{p_t}{p_0} \\
q_t, & \text{otherwise.} 
\end{cases}
\] (A6)

A6 Relative humidity

The relative humidity distribution makes use of Tetens’ formula for the saturation mixing ratio as also shown in Klemp and Wilhelmson (1978) and Klemp et al. (2015). Note though that the Klemp et al. (2015) formulation (their Eq. (12)) contains a typographical error when stating that their pressure \( \bar{p}_{eq} \) has units of hPa. The correct unit for the pressure \( p \) in the denominator is Pa as shown below. Here, we use Tetens’ formula for the saturation specific humidity which is approximately equal to the saturation mixing ratio. The formula is

\[
q_{vs} = \frac{\epsilon}{p} e_0^* \exp \left( 17.27 \frac{T - 273 \text{ K}}{T - 36 \text{ K}} \right) = \frac{380 \text{ Pa}}{p} \exp \left( 17.27 \frac{T - 273 \text{ K}}{T - 36 \text{ K}} \right) 
\] (A7)

where the units of \( p \) and \( T \) are Pa and K, respectively. For illustration purposes, Eq. (A7) also lists the saturation vapor pressure \( e_0^* = 610.78 \text{ Pa} \) at the temperature triple point \( T_{00} = 273.16 \text{ K} \) and the symbol \( \epsilon = R_d R_v^{-1} \approx 0.622 \) which denotes the ratio of the gas constant for dry air \( R_d \) to that for water vapor with \( R_v = 461.5 \text{ J kg}^{-1} \text{ K}^{-1} \). The relative humidity can then be defined as

\[
\text{RH} = 100\% \cdot \frac{q}{q_{vs}}
\]

Appendix B: Implementation Details for the CESM2.2 Dynamical Cores and MPAS

We recommend using the default physical constants as implemented in a chosen dynamical core, and ideally in-lining the initialization routine in the codebase of the chosen model. This was the initialization strategy for the CESM 2.2 and MPAS dynamical cores, and all code modifications are provided in Hughes and Jablonowski (2022). In CESM 2.2 we made use of CESM’s “Simpler Models” framework which invokes the Kessler physics routine and the analytic initialization of the moist baroclinic wave (the Ullrich et al. (2016) default without topography) via the CESM compset “FKESSLER” and a CAM namelist entry for the variable \texttt{analytic_ic_type = 'moist_baroclinic_wave_dcmip2016'}. The CESM 2.2 code change then simply augments the existing initialization for the baroclinic wave and adds the topographic changes via a swap of the CAM routine \texttt{ic_baroclinic.F90}. Note that the routine \texttt{ic_baroclinic.F90} also accommodates
the dry variant of the baroclinic wave, which can be selected via the adiabatic compset “FADIAB”, the configure command

```
./xmlchange --append --file env_build.xml --id CAM_CONFIG_OPTS --val="--analytic_ic"
```

to activate the analytic in-lined initialization, and the alternative namelist option

```
analytic_ic_type = 'dry_baroclinic_wave_dcmip2016'.
```

These settings initialize the dry configuration with \( q = 0 \) and do not activate any physical parameterizations. All key namelist entries are provided below in Tables B1-B4. The CESM values for the physical constants are listed in Table A1. For the MPAS simulations the default physical constants of the MPAS stand-alone distribution were used (Jacobsen et al., 2019). MPAS provides the implementation of the Kessler warm-rain microphysics routine which can be activated via a namelist option as shown in Table B4. In addition, the initialization routine for the moist baroclinic wave with topography was added to MPAS’ existing framework for idealized test cases via a code change.

All dynamical core simulations are run with 30 model levels and use model tops near 2 hPa (SE, FV, FV3) and 8 hPa (MPAS). These model tops lie between 30-35 km for the provided temperature structure. The positions of the hybrid pressure-based model level used for SE, FV, and FV3 are listed in Reed and Jablonowski (2012) and are recommended to users of this test case. These are the default levels in CESM 2.2 once the compset FKESLER is invoked. For MPAS we use the 30 default levels for MPAS’ idealized testing framework. Most simulations presented in this study are run with a nominal 0.5° (about 50 km) grid spacing which corresponds to the grid resolution settings ne60 (SE), FV05 (FV), C192 (FV3), and 60 km (MPAS). These identifiers are used as labels in the figures, and also correspond to the time step and diffusion settings quoted below.

Table B1 contains the key namelist parameters to replicate our SE model integrations. The time steps used by the SE dynamical core are \( \Delta_{\text{phys}} = 900 \, \text{s} \), \( \Delta_{\text{vertical remap}} = \Delta_{\text{phys}}/2 = 450 \, \text{s} \), \( \Delta_{\text{dynamics}} = \Delta_{\text{vertical remap}}/3 = 150 \, \text{s} \), and \( \Delta_{\text{hyperviscosity}} = \Delta_{\text{dynamics}}/3 = 50 \, \text{s} \) as, for example, explained in Lauritzen et al. (2018). The se_nu_XX parameters denote diffusion coefficients which are resolution-dependent.

Table B2 contains the key namelist parameters to replicate the FV model integrations. The time steps used by the FV dynamical core are \( \Delta_{\text{phys}} = 900 \, \text{s} \), \( \Delta_{\text{vertical remap}} = \Delta_{\text{phys}}/2 = 450 \, \text{s} \), \( \Delta_{\text{tracer}} = \Delta_{\text{vertical remap}} = 450 \, \text{s} \), and \( \Delta_{\text{dynamics}} = \Delta_{\text{tracer}}/4 = 112.5 \, \text{s} \). The namelist entry \( \text{fv}_\text{div24del2flag} \) selects the 4th-order horizontal divergence damping mechanism. The monotonicity constraints for the horizontal advection and the vertical remap algorithm, denoted by the \( \text{fv}_\text{Xord} \) namelist entries, are called the “relaxed constraint” by Lin (2004) and denote the default settings.

Table B3 contains the key namelist parameters for the FV3 model integrations. The time steps used by the FV3 dynamical core are \( \Delta_{\text{phys}} = 900 \, \text{s} \), \( \Delta_{\text{vertical remap}} = \Delta_{\text{phys}}/2 = 450 \, \text{s} \), \( \Delta_{\text{tracer}} = \Delta_{\text{vertical remap}} = 450 \, \text{s} \), and \( \Delta_{\text{dynamics}} = \Delta_{\text{vertical remap}}/6 = 75 \, \text{s} \). The “fv3_nord = 2” setting activates the 6th-order horizontal divergence damping mechanism with the dimensionless resolution-independent coefficient \( \text{fv3}_d4_{bg} \). The optional vorticity damping is not activated. The choice of the monotonicity constraint for the horizontal advection, as determined by \( \text{fv3}_\text{hord}_\text{XX} \), picks the least diffusive option.

Table B4 contains the key namelist parameters for MPAS. The time steps used by the MPAS dynamical core are: \( \Delta_{\text{phys}} = 300 \, \text{s} \), \( \Delta_{\text{dynamics}} = \Delta_{\text{phys}}/3 = 100 \, \text{s} \), and \( \Delta_{\text{acoustic}} = \Delta_{\text{dynamics}}/2 = 50 \, \text{s} \). MPAS is a non-hydrostatic model, and so it ensures numerical stability in the presence of 3D acoustic waves by handling acoustic propagation with very short time steps.
Table B1. Key CESM 2.2 namelist parameters used in nominal 0.5° SE model integrations.

<table>
<thead>
<tr>
<th>Namelist Parameter</th>
<th>Value</th>
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<td>se_rsplit</td>
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Table B2. Key CESM 2.2 namelist parameters used in nominal 0.5° FV model integrations.

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<td>fv_div24de12flag</td>
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<td>fv_fft_flt</td>
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<td>fv_iord</td>
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Table B3. Key CESM 2.2 namelist parameters used in nominal 0.5° FV3 model integrations.

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Table B4. Key namelist parameters used in nominal 0.5° MPAS model integrations.

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**Code and data availability.** Model data as well as source code to run the test case in the CESM and MPAS models have been uploaded to a Zenodo dataset (Hughes and Jablonowski, 2022). The collection includes data analysis scripts which reproduce the figures from the article.

**Author contributions.** CJ suggested the study. OKH and CJ designed the specifications of the test case. OKH developed the code modifications, conducted the model experiments, and analyzed the model results. OKH and CJ jointly wrote the manuscript.

**Competing interests.** The authors declare that they have no conflict of interest.

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