Topography based local spherical Voronoi grid refinement on classical and moist shallow-water finite volume models

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Abstract. Locally refined grids for global atmospheric models are attractive since they are expected to provide an alternative to solve local phenomena without the requirement of a global high-resolution uniform grid, whose computational cost may be prohibitive. The Spherical Centroidal Voronoi Tesselations (SCVT), as used in the atmospheric Model for Prediction Across Scales (MPAS), allows a flexible way to build and work with local refinement. Alongside, the Andes Range plays a key role in the South American weather, but it is hard to capture its fine structure dynamics in global models. This paper describes how to generate SCVT grids that are locally refined in South America and that also capture the sharp topography of the Andes Range by defining a density function based on topography and smoothing techniques. We investigate the use of the mimetic finite volume scheme employed in the MPAS dynamical core on this grid considering the non-linear classic and moist shallow-water equations on the sphere. We show that the local refinement, even with very smooth transitions from different resolutions, generates spurious numerical inertia-gravity waves that may even numerically de-stabilize the model. In the moist shallow-water model, where physical processes such as precipitation and cloud formation are included, our results show that the local refinement may generate spurious rain that is not observed in uniform resolution SCVT grids. Fortunately, the spurious waves originate from small-scale grid-related numerical errors and therefore can be mitigated using small amounts of numerical diffusion. We show that, in some cases, the clouds are better represented in a variable resolution grid when compared to a respective uniform resolution grid with the same number of cells, while in other cases, grid effects can deteriorate the cloud and rain representation.

1 Introduction

The Andes, the world's longest mountain range, located in entire South America's western coast, acts as an obstacle to the atmospheric circulation and plays a key role in the weather of South America. For instance, the South American low-level jet, formed in the Andes eastern slope, has an important influence in the Brazilian southeast, among other regions, precipitation transporting moisture from the Amazon basin (Garreaud, 2009). Furthermore, the Andes is essential for the formation of the

South American low-level jet (Insel et al., 2010) and Junquas et al. (2015) shows that the remotion of the Andes mountain leads to a reduction of the precipitation in the Brazilian southeast. Despite its length, the Andes has a narrow width of less than 200 km almost everywhere in the mountain and its elevation is 4000 m on average but the elevation may suddenly change in a few kilometers.

The Andes complex topography is hard to be captured in a global model due to its abrupt changes in elevation and resolution limitations imposed by computational costs. Furthermore, Figueroa et al. (2016) shows that both Brazilian global Atmospheric Model (BAM) and the Global Forecast System (GFS) produce dry or wet biases in the Amazon basin, over the Andes, and in the Brazilian Southeast. This bias is also observed in other models (Silva et al., 2011; Chou et al., 2020). Therefore, aiming to improve the precipitation forecast in those regions, it is desirable to better represent the Andes Range shape in the employed model. One alternative to overcome the problem of resolution limitations is to employ regional models (e.g. Freitas et al. (2017)). A second alternative is to develop a locally refined global grid that captures well the Andes Range. The second alternative has the advantage of not requiring any artificial boundary conditions as it is usual for regional models and has been developed for some recent global models (Ullrich et al., 2017).

Locally refined grids for global models are built aiming to solve local phenomena without requiring a global high-resolution grid, in which computational cost may be prohibitive. For instance, Barros and Garcia (2004) have proposed a locally refined latitude-longitude grid with a coarse global uniform grid successively refined to a region of interest. Local refinements also have been developed for the cubed-sphere grid. For instance, in Harris et al. (2016) a local refinement for the cubed-sphere was developed using stretching transformations, where the region of interest has a resolution 7 times that in the coarse grid region. On other hand, Lean et al. (2008) propose to nest models with resolutions of 12, 4, and 1 km using the UK Met Office operational model at that time, where the 4 and 1 km models use two squares centered in England as the domain. Those works propose feasible ways to have a higher resolution in areas of interest. However, they are usually constrained to a rectangular geometry in the region of interest and impose interpolation constraints between different resolutions, which may affect the dynamics of in/out propagating waves.

Aiming to run at very high resolutions, many recent global atmospheric models are employing quasi-uniform geodesic spherical grids (Staniforth and Thuburn, 2012). The geodesic grids usually have more complex cell geometry, but they can allow more flexibility in local grid refinement. The added geometry complexity on these grids, such as those using Voronoi (hexagonal-pentagonal) cells, makes the development of numerical schemes on these grids more difficult than in regular quadrilateral grids. Nevertheless, the finite volume method for the shallow-water equations proposed by Thuburn et al. (2009) and Ringler et al. (2010), hereafter named TRSK, was designed to work on arbitrary orthogonal Voronoi C-grids. This scheme has many desirable mimetic problems, such as mass and total energy conservation, preservation of stationary geostrophic modes on the f-sphere, among others. It is used in the dynamical core of some atmospheric models such as in the Model for Prediction Across Scale (MPAS) (Skamarock et al., 2012) developed jointly by the National Center for Atmospheric Research (NCAR) and the Los Alamos Laboratory, and the model Icosahedral Dynamical Core (DYNAMICO) (Dubos et al., 2015).

TRSK applies to a variety of grids, in particular, to the Spherical Centroidal Voronoi Tesselation (SCVT) grids which are adopted in MPAS and may be thought of as an optimized version of the hexagonal-pentagonal grid generated from an

icosahedral grid. The SCVT grids are orthogonal grids where each grid cell is a Voronoi diagram and the center of mass with

respect to a given density function is the corresponding Voronoi region generator (Ju et al., 2011). This grid has the flexibility

60 of allowing local refinements by carefully defining a density function and employing Lloyd's method (Du et al., 1999, 2003;

Ringler et al., 2011). Using a circular refinement region, Park et al. (2014) shows that SCVT Variable Resolution (VR) grids

have no significant wave reflection detected in the transition zone between high- and low-resolution regions, which occurs in

the nested model Weather Research and Forecasting (WRF) model analyzed in this work. The work of Kramer et al. (2018)

investigates three different meteorological events in Europe and concludes that the VR grid used in MPAS global model yields

65 comparable results to the WRF model. Furthermore, the VR grid performed better than the uniform resolution grid in the

organized convection test case, being computationally cheaper. VR grids are also feasible in the ocean component of MPAS.

For instance, Hoch et al. (2020), considers grids with local refinement in the North American coast and concludes that even

when the grid quality is degraded, no spurious effects are observed in the MPAS-Ocean model simulation results. On the other

hand, some studies show that the local refinements in the MPAS grid may impact negatively the results of the model. For

70 instance, Rauscher and Ringler (2014) shows that the VR grid generates an eddy kinetic energy maximum in the refined region

for the aquaplanet and Held-Suarez test cases. This maximum is caused by the use of the VR grid and is not observed in the

uniform resolution grid. While MPAS allows a model that is suitable to work with grids with local refinement in arbitrary

shapes by only defining a density function, wave representation and interaction in these grids is still not well understood.

In this work, we propose to generate SCVT grids with local refinement based on Andes topography. Our locally refined grids

75 will be designed aiming to capture well the Andes mountain and the South American continent with smooths transitions to the

coarse grid regions, focusing on obtaining a grid that may allow better precipitation forecasts in South America. Also, we will

perform a detailed analysis of the intrinsic advantages/problems of using TRSK scheme in such grids in two frameworks that

capture many important structures of the atmosphere.

In the first framework, we will consider the non-linear shallow-water equations on the sphere and some standard tests

80 proposed in the literature (Williamson et al., 1992; Galewsky et al., 2004) and a more recent test proposed by Shamir et al.

(2019). In the second framework, the moist shallow-water model proposed by Zerroukat and Allen (2015) is investigated.

This model includes physical processes in the shallow-water model and allows us to simulate rain and clouds. The moist

shallow-water model will allow us to study the impact of the local refinement on cloud and rain formation.

This paper is outlined as follows. In Sect. 2, we describe the development of the SCVT grids with local refinement on

85 the Andes mountains. In Sect. 3, the classical and moist shallow-water models are presented, along with their discretization

schemes. In Sect. 4 we present the results obtained using TRSK on the developed grids considering the classical and moist

shallow-water model. Conclusions are drawn in Sect. 5.

2 Grids

2.1 Spherical Centroidal Voronoi Tesselations

90 We start by defining the Spherical Voronoi concepts that will be important for the local refinement technique discussed in this work.

Given a set of distinct points on the sphere, that we call generators, $\{\boldsymbol{x_i}\}_{i=1}^n \subset \mathcal{S}^2$, we define the i-th Voronoi region as,

$$\Omega_i = \{\boldsymbol{x} \in \mathcal{S}^2 : d(\boldsymbol{x}, \boldsymbol{x_i}) < d(\boldsymbol{x}, \boldsymbol{x_j}), \forall i = 1, \cdots, n, i \neq j\}, \tag{1}$$

where $d(\boldsymbol{x}, \boldsymbol{y})$ is the geodesic distance on the sphere, $\forall \boldsymbol{x}, \boldsymbol{y} \in \mathcal{S}^2$. The sets $\{\Omega\}_{i=1}^n$ are called spherical Voronoi tesselation

95 (SVT) and decompose the sphere into disjoint convex spherical polygons and generates a spherical grid (Ju et al., 2011).

With the Voronoi tesselation, we can associate a dual grid by connecting nearest neighbor generators. The geodesic segment connecting neighbors is orthogonal to the sharing edge between these two generators. This orthogonality is important for the numerical schemes that we shall present later. Under the assumption that no 4 near neighbors are co-circular with empty interior, these dual cells all consist of triangles and form a Delaunay triangulation (Okabe et al., 2000). In this duality relation,

100 each Voronoi region corresponds to a unique dual cell vertex, each Voronoi vertex is associated with a unique dual-cell and each Voronoi edge is associated with a unique dual-edge.

An example of SVT is the primal grid of the icosahedral grid, the pentagonal/hexagonal grid. The icosahedral grid is obtained by gnomonically projecting the edges of an icosahedral inscribed within a sphere. This process generates 12 vertices and 20 triangles. Each of these triangles can be divided into 4 new triangles and we can proceed in this manner recursively. Given a

105 level of recursion $l = 0, 1, 2, \cdots$, it can be shown that the icosahedral grid obtained by recursion has $N_l = 10 \cdot 2^{2l} + 2$ vertices, and consequently, the dual pentagonal/hexagonal grid has N_l cells. The number l is also called grid level.

A particular case of SVT, that we are interested in, is the spherical centroidal Voronoi tesselation (SCVT). Given a density function $\rho : \mathcal{S}^2 \rightarrow]0, \infty[$, for each Voronoi region Ω, the center of mass with respect to ρ is given by:

$$\boldsymbol{x}^* = \frac{\int_\Omega \boldsymbol{x} \rho(\boldsymbol{x}) d\Omega(\boldsymbol{x})}{\int_\Omega \rho(\boldsymbol{x}) d\Omega(\boldsymbol{x})} \tag{2}$$

110 For this definition of the center of mass, does not necessarily hold that $\boldsymbol{x}^* \in \mathcal{S}^2$. Therefore, following Du et al. (2003), we introduce the concept of spherical center of mass, which may be calculated using radial projection,

$$\boldsymbol{x}^c = \frac{\boldsymbol{x}^*}{\|\boldsymbol{x}^*\|}. \tag{3}$$

A SCVT is a SVT such that each generator is the center of mass of the corresponding Voronoi region with respect to a given density function ρ. As we can notice, SCVT generates spherical grids, that we call SCVT grids. An algorithm that creates

115 SCVT from initial random generators is the iterative Lloyd's method (Du et al., 1999, 2003), which iteratively pushes the Voronoi generators to the Voronoi mass centroid position, and is the method employed in this work.

(a) (b) (c)
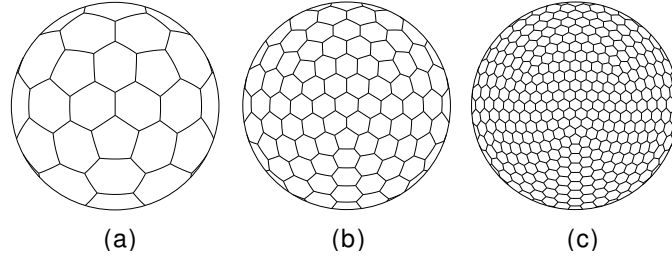
Figure 1. SCVT grids using a uniform density function for 12 (a), 42 (b) and 162 (c) generators respectively.

(a) (b) (c)
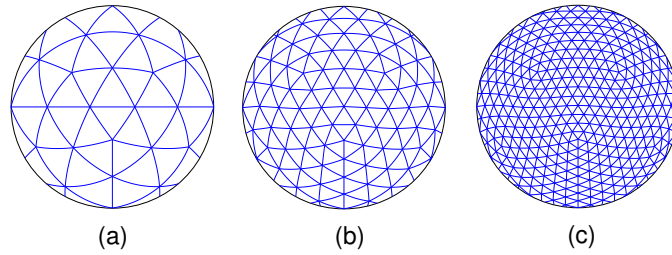
Figure 2. Dual (triangular) grids of the grids shown in Fig. 1 for 12 (a), 42 (b) and 162 (c) generators respectively.

Given two Voronoi cells Ω_i and Ω_j, we denote their diameter by h_i and h_j, respectively. It is possible to show that (Du et al., 2003; Ju et al., 2011):

$$\frac{h_i}{h_j} \approx \left(\frac{\rho(\boldsymbol{x_j})}{\rho(\boldsymbol{x_i})} \right)^{\frac{1}{4}}. \tag{4}$$

120   This estimation plays a key role in this work because it allows us to build grids with local refinement.

We show examples of SCVT's for $\rho = 1$ generated using Lloyd's method in Fig. 1 where we use the pentagonal/hexagonal grid as initial guess in Lloyd's method. The corresponding dual grid is showed in Fig. 2. We use N_l generators, for $l = 2, 3, 4$. The dual grid of SCVT grid may be seen as an optimized version of the icosahedral grid (Miura and Kimoto, 2005).

Next, we present how to build a grid with local refinement on Andes mountain based in estimative given in Eq. (4), in the
125   density function presented in Ju et al. (2011) and in the topography data from Amante and Eakins (2009).

2.2   Locally refined SCVT

We represent the $\mathbb{R}^3$ coordinates of a point $\boldsymbol{x} \in \mathcal{S}^2$ in spherical coordinates as,

$$\boldsymbol{x} = (\sin\varphi\cos\theta, \sin\phi\sin\theta, \cos\varphi), \quad \varphi \in \left[-\frac{\pi}{2}, \frac{\pi}{2} \right], \theta \in [-\pi, \pi], \tag{5}$$

and define the parameters $(\varphi_c, \theta_c) = \left(-\frac{\pi}{9}, -\frac{\pi}{3} \right)$, and $\boldsymbol{x_c} = (\sin\varphi_c\cos\theta_c, \sin\phi_c\sin\theta_c, \cos\varphi_c)$. We also consider real values γ,
130   α and ε that will be set later. Denoting the Euclidian norm as $\|\cdot\|$, we define the distance from $\boldsymbol{x_c}$ by $d(\boldsymbol{x}, \boldsymbol{x_c}) = \|\boldsymbol{x} - \boldsymbol{x_c}\|$, for

$\boldsymbol{x} \in \mathcal{S}^2$, and introduce the following auxiliary function:

$$s(\boldsymbol{x}) = \begin{cases} 1 & \text{if } d(\boldsymbol{x}, \boldsymbol{x_c}) \leq \alpha, \\ \frac{\alpha + \varepsilon - d(\boldsymbol{x}, \boldsymbol{x_c})}{\varepsilon} & \text{if } \alpha \leq d(\boldsymbol{x}, \boldsymbol{x_c}) \leq \alpha + \varepsilon, \\ 0 & \text{otherwise.} \end{cases} \tag{6}$$

A density function that will allow us to refine based on two criterias, in our case the topography (Andes Mountains) and the continent (South America) is given by:

$$\rho(\boldsymbol{x}) = \frac{1}{\gamma^4} + \left(1 - \frac{1}{\gamma^4}\right)(\lambda s(\boldsymbol{x}) + (1-\lambda)b(\boldsymbol{x})), \quad x \in \mathcal{S}^2 \tag{7}$$

where b is represents the topography (in our case, the Andes region), in this work retrieved from Amante and Eakins (2009) and normalized for the interval $[0,1]$, $\gamma \geq 1$ and $\lambda \in [0,1]$. Note that for $\lambda = 1$, the density function will refine only South America without capturing the Andes mountain sharp topography. For $\lambda = 0$, the density function will refine only the Andes mountain sharp topography. Therefore, for a value $\lambda \in]0,1[$, the generated grid will capture both Andes mountain sharp topography and the South American continent and will have a transition to the coarse global grid. The closer the parameter λ is to 1, the less the topography will be represented in the refinement. And the closer the parameter λ is to 0, the more the topography sharp will be represented and less refinement will be in the rest of the South American continent. The parameter γ represents approximately the ratio between coarse and fine grid resolution areas, which follows from the estimative given in Eq. (4).

The ETOPO data (Amante and Eakins, 2009) used in the function b is defined in a latitude-longitude grid with 720×1440 points, which is enough resolution to represent well the Andes' topography for our experiments. We employed bilinear interpolation to compute the function b in SCVT grid points that are not in the latitude-longitude grid. In order to guarantee grids with smooth transitions between the high and low-resolution regions, we applied a smoothing technique on ETOPO data based on Jacobi method. For each index (i,j) representing a point on the latitude-longitude grid, $0 \leq i \leq 720$, $0 \leq j \leq 1440$, we apply the classic Jacobi iteration process:

$$b_{ij}^{k+1} = \frac{1}{4}(b_{i-1,j}^k + b_{i+1,j}^k + b_{i,j-1}^k + b_{i,j+1}^k), \tag{8}$$

where $k = 0,1,2,\cdots$ and b_{ij}^0 is the initial Andes topography data from ETOPO. The number of iteration was empirically defined as 500 steps, ensuring that the basic shape of the Andes Range is still well represented in the topography dataset. More importantly, this smoothing later ensures that the generated SCVT primal grids have all circumcenters inside the triangles.

We also applied a smoothing filter in the function s, considering it defined in the latitude-longitude grid with 720×1440 points. We used a moving average method by replacing the value of s at a point in the latitude-longitude grid considering the mean of s in a box with 55×55 points centered at the considered point. This box size showed to guarantee that the generated grids have circumcenters inside the triangles.

We set the parameters to $\gamma = 3$, $\alpha = \frac{7\pi}{45}$, $\varepsilon = \frac{\pi}{12}$ and $\lambda = 0.6$. In Fig. 3, we show the generated grid for 10242 generators after applying Lloyd's method. The diameters of cells are also shown in Fig. 3. The diameter of a Voronoi cell is estimated as the average distance from its generator to its neighbors. As we can notice in the ratio of the diameters of the cells on the Andes

(a)   (b)
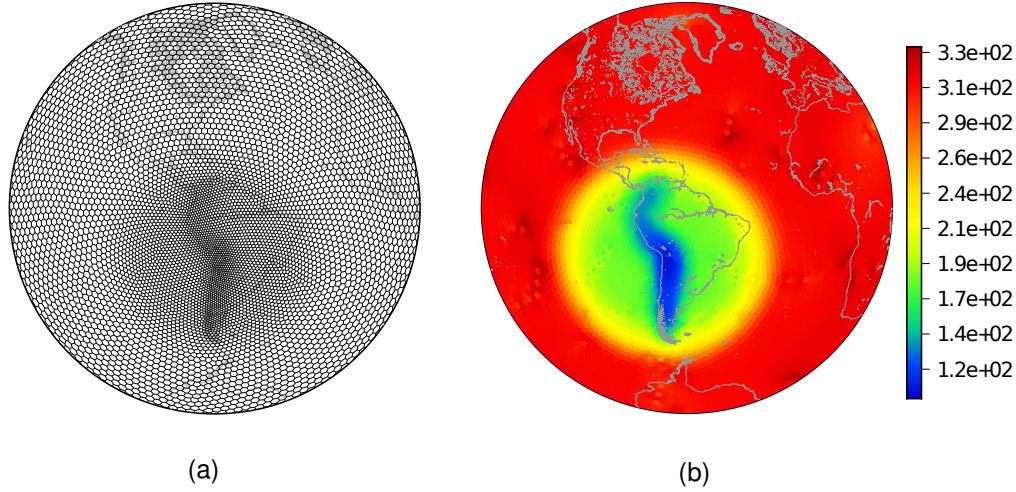
Figure 3. (a) SCVT grid with local refinement on Andes mountain. (b) the cell diameters in kilometers for 10242 generators considering $\gamma = 3$, $\alpha = \frac{7\pi}{45}$, $\varepsilon = \frac{\pi}{12}$ and $\lambda = 0.6$.

and the cells in the coarser grid region is approximately 3. We generate the variable resolution SCVT grids using the density function defined in Eq. (6) using $N_l = 10 \cdot 2^{2l} + 2$ generators, for $l = 1, 2, \cdots, 7$ and we refer to these grids as VR1, VR2, $\cdots$, VR7, respectively. These generated grids and the uniform SCVT grids will be used in our numerical experiments in Sect. 4. The uniform resolution SCVT grids were generated from grid level 1 up to 9 and we refer to these grids as UR1, UR2, $\cdots$, UR9. We show the minimum and maximum diameters of the cells for the uniform and variable resolution SCVT grids in Table 1.

3 Model description

3.1 Shallow-water model

As it is usual in the development of numerical methods for global atmospheric models, we first evaluate it considering the shallow-water equations on the sphere. These equations represent important aspects of the atmosphere, such as Rossby waves, Coriolis effect and geostrophic adjustment, and they have the advantage of being two-dimensional, therefore computationally cheaper to run at high-resolutions.

We denote a point on the sphere by $\boldsymbol{x}$ and a time instant by $t \geq 0$. Then, we consider the non-linear shallow-water equations on the sphere written in the vector invariant form (Ringler et al., 2010):

$$\frac{\partial \boldsymbol{u}}{\partial t} = -qh\boldsymbol{u}^{\perp} - \nabla B, \tag{9}$$

$$\frac{\partial h}{\partial t} = -\nabla \cdot (h\boldsymbol{u}), \tag{10}$$

Grid level	Number of Voronoi cells	Voronoi cells diameters (km) of SCVT grids					
		Uniform resolution			Variable resolution		
		Grid name	Min	Max	Grid name	Min	Max
1	42	UR1	4084	4649	VR1	2479	5189
2	162	UR2	2035	2283	VR2	858	2672
3	642	UR3	996	1128	VR3	425	1320
4	2562	UR4	484	562	VR4	198	673
5	10242	UR5	235	281	VR5	98	336
6	40962	UR6	114	140	VR6	47	172
7	163842	UR7	55	70	VR7	24	87
8	655362	UR8	27	35	-	-	-
9	2621442	UR9	14	17	-	-	-

Table 1. Minimum and maximum diameters of the Voronoi cells for the uniform SCVT grid (UR) and for the variable resolution (VR) grid that captures well the Andes mountain.

Variable description	Simbol	Definition
Bottom topography	b	-
Earth gravity	g	9.8 m s^{-2}
Earth rotation velocity	Ω	7.292×10^{-5} rad s^{-1}
Latitude	ϕ	-
Local unit vector pointing in vertical such that $\boldsymbol{u} \cdot \boldsymbol{k} = 0$	$\boldsymbol{k}$	-
Perpendicular velocity vector	$\boldsymbol{u}^{\perp}$	$\boldsymbol{k} \times \boldsymbol{u}$
Coriolis parameter	f	$2\Omega \sin \phi$
Kinetic energy	K	$|\boldsymbol{u}|^2/2$
Bernoulli potential	B	$g(h+b) + K$
Relative Vorticity	ζ	$\boldsymbol{k} \cdot \nabla \times \boldsymbol{u}$
Absolute vorticity	η	$\zeta + f$
Potential vorticity	q	$\frac{\eta}{h}$

Table 2. Variables and parameters for the shallow-water model

where $h = h(\boldsymbol{x}, t)$ denotes the total fluid depth, $\boldsymbol{u} = \boldsymbol{u}(\boldsymbol{x}, t)$ is the velocity vector tangent to the sphere for all $t \geq 0$, and all other terms are defined in Table 2. The operators ∇ and $\nabla \cdot$ are assumed to be the horizontal gradient and divergence on the sphere, respectively. These equations define an initial value problem that describes the evolution of a thin layer of a two-dimensional fluid on the sphere.

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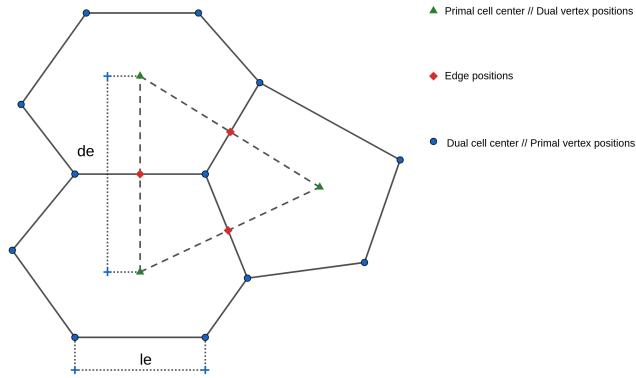


Figure 4. Variable positions for a C-grid staggering.

Description	Simbol
Primal cell	i
Dual cell	v
Primal/dual edge	e
Center of a primal cell i	$\boldsymbol{x_i}$
Edge point of an edge e	$\boldsymbol{x_e}$
Center of a dual cell v	$\boldsymbol{x_v}$
Primal edge length	l_e
Dual edge length	d_e

Table 3. Grid notation following Ringler et al. (2010).

The finite volume scheme adopted here was proposed in Thuburn et al. (2009) and Ringler et al. (2010), also denoted as TRSK, and considers a grid with an orthogonal dual grid and a C-grid staggering of the variables. We consider three positions in the C-grid: primal cell centers (or dual-cell vertices), primal cell vertices (or dual-cell centers), and the intersection point between a primal edge and dual-edge that we call edge points (Fig. 4).

185   We summarize the notation needed for TRSK in the Table 3. In TRSK, the height h is stored at primal centers and is denoted by $h_i(t) = h(\boldsymbol{x_i}, t)$. Given an edge e, we define a normal vector $\boldsymbol{n_e}$ to the edge pointing to the direction such that the normal component of the velocity given by $u_e(t) = \boldsymbol{u}(\boldsymbol{x_e}, t) \cdot \boldsymbol{n_e}$ is positive. The normal component of the velocity is stored at edge points. Our prognostic variables in TRSK are h_i and u_e for each primal center i and edge e. Taking the dot product of Eq. (9)

with $\boldsymbol{n}_e$ and get:

$$\frac{\partial u_e}{\partial t} = -[qh\boldsymbol{u}^\perp]_e - [\nabla B]_e, \tag{11}$$

$$\frac{\partial h_i}{\partial t} = -[\nabla \cdot (h\boldsymbol{u})]_i, \tag{12}$$

where $[\boldsymbol{F}]_e = \boldsymbol{F}(\boldsymbol{x_e}, t) \cdot \boldsymbol{n_e}$ and $[G]_i = G(\boldsymbol{x_i}, t)$, for any vector field $\boldsymbol{F}$ and scalar field G. Each term on the right hand side of
Eq. (11) and Eq. (12) are approximated using the spatial discrete operators from Ringler et al. (2010). After discretizing each
term for all primal cells and edges, we get an ordinary differential equation system. In this work, we employed a fourth-order
Runge-Kutta scheme to solve the ordinary differential equation system since we are aiming to analyze only the impact of the
local refinement on the horizontal discretization. At last, TRSK has been known for its mimetic properties, i.e, the discrete
equations mimics some of the continuous equations, such as:

1. Mass conservation;

2. Coriolis term is not an energy source/sink ($u_e u_e^\perp = 0$);

3. Total energy is conserved within time error truncation;

4. Preservation of stationary geostrophic modes on the f-sphere.

The TRSK discretization is also known to be very low order accurate, and may even lack consistency on nonregular grids
(Peixoto and Barros, 2013; Peixoto, 2016). However, due to its mimetic properties and flexibility, it is still adopted in some
modern global atmospheric models such as MPAS (Skamarock et al., 2012) and DYNAMICO (Dubos et al., 2015).

3.2 Moist shallow-water model

Even though the shallow-water model retains keys properties of the atmosphere, they lack the representation of sub-scale
physical processes. Therefore, we consider here an intermediate step of a two-dimensional moist shallow water model, avoiding
the computational dispendure of a three-dimensional primitive equation model while being able to account for water-related
sub-grid processes.

We follow the moist shallow-water model proposed by Zerroukat and Allen (2015), where moist dynamics are included
in the classic shallow water-model. This model is two-dimensional, which is computationally cheaper than three-dimensional
models at high resolutions, and is derived from the Boussinesq approximation to the three-dimensional Euler equations. The
model can be summarized in the following equations:

$$\frac{\partial \boldsymbol{u}}{\partial t} = -qh\boldsymbol{u}^\perp - \nabla B + S_u, \tag{13}$$

$$\frac{\partial h}{\partial t} = -\nabla \cdot (h\boldsymbol{u}), \tag{14}$$

$$\frac{\partial h\theta}{\partial t} = -\nabla \cdot (h\theta\boldsymbol{u}) + hS_\theta, \tag{15}$$

$$\frac{\partial hq^k}{\partial t} = -\nabla \cdot (hq^k\boldsymbol{u}) + hS_q^k \tag{16}$$

where θ is the temperature, $k = 1, 2, 3$, $q^1 = q_v$ is the water vapor state, $q^2 = q_c$ is the cloud state, $q^3 = q_r$ is the rain state. We are using the flux form for the advection equations for numerical purposes that we shall see later. The moisture variables are

220 tracers that are advected in this system. The other variables are the same as in the shallow-water model (Table 2). The source terms of the advection equations define a three-state moist physics, where vapor may be converted to clouds, clouds are allowed to evaporate and clouds may be converted to rain. This three-state moist physics depends on a saturation function, which is given by Zerroukat and Allen (2015):

$$q_{sat}(\theta) = \frac{q_0}{g(h + b)} \exp(20\theta), \tag{17}$$

225 and the initial vapor state is given by $q_v(\lambda, \phi, 0) = G(\lambda, \phi)q_{sat}(\theta)$, where λ denotes the longitude and q_0 is a constant chosen in order to guarantee that the initial maximum value of q_v is 0.02 and G is a function between 0 and 1. The additional variables and parameters to the moist shallow water model are shown in Table 4. Note that $S_{q_v} + S_{q_c} + S_{q_r} = 0$, therefore the integral of $h(q_v + q_c + q_r)$ is conserved in this model.

Variable description	Symbol	Definition
Precipitation threshold	q_{precip}	10^{-4}
Pseudo latent heat constant	L	10
Vapor-cloud conversion rate	γ_v	$(1 + L\frac{\partial q_{sat}}{\partial \theta})^{-1}$
Cloud-rain conversion rate	γ_r	10^{-3}
Time step used in the numerical model	Δt	-
Vapor-cloud conversion	Δq_v	$\max\{0, \gamma_v(q_v - q_{sat})\}/\Delta t$
Cloud evaporation	Δq_c	$\min\{q_c, \max\{0, \gamma_v(q_v - q_{sat})\}\}/\Delta t$
Cloud-rain conversion	Δq_r	$\max\{(0, \gamma_r(q_c - q_{precip})\}/\Delta t$
-	Π	$\frac{1}{2}h^2\theta$
Momentum equation source	S_u	$g\theta\nabla b + \frac{1}{h}\nabla\Pi$
Temperature equation source	S_θ	$L(\Delta q_v - \Delta q_c)$
Vapor equation source	S_q^v	$\Delta q_c - \Delta q_v$
Cloud equation source	S_q^c	$\Delta q_v - \Delta q_c - \Delta q_r$
Rain equation source	S_q^r	Δq_r

Table 4. Variables and parameters for the moist shallow-water model (Zerroukat and Allen, 2015)

In this model, we consider the prognostic variables are $\boldsymbol{u}$, h, $h\theta$ and hq^k, $k = 1, 2, 3$. Similarly to the shallow water model,

230 we store h, $h\theta$ and hq^k at Voronoi centers and the normal component of $\boldsymbol{u}$ is stored at the edges. We write similar equations to Eq. (11) and Eq. (12) for the moist shallow-water equations and discretize the right hand side using TRSK operators.

The divergence term in Eq. (15) and Eq. (16) for $h\theta$ and hq^k may be computed using the divergence discretization of TRSK, and this explains why we used the advection equation in flux form. The source terms for Eq. (15)-(16) are straightforward to

calculate once they depend only on θ and q^k, thus we only need to divide these variables by h. Equation (13) has a source term

235 and it needs to be evaluated at edge points. This source may be calculated using TRSK interpolation and gradient operators. We point out that MPAS solves the tracer equations such as Eq. (15)-(16) using a high-order advection scheme (Skamarock and Gassmann, 2011). Here, we consider the TRSK scheme to solve tracer equations aiming to analyze how TRSK simulates the formation of cloud and rain. Furthermore, as we shall see in Sec. 4.2, a monotonic filter is employed in our simulations.

4   Numerical Results

240 This section is dedicated to present the results of the TRSK method on the variable resolution grids considering two frameworks: the classical shallow-water model and the moist shallow-water model developed by Zerroukat and Allen (2015). We will consider the variable resolution grids developed in Sect. 2.2 and the uniform SCVT grids to compare the impact of introducing a local refinement. We also consider the addition of numerical diffusion on the locally refined grids tests. It will be clear whether we are adding diffusion or not. The numerical diffusion was included in the momentum equation (Eq. (11)), following

245 Skamarock et al. (2012), based on the following vector identity for the vector Laplacian,

$$\Delta\boldsymbol{u} = \nabla(\nabla\cdot\boldsymbol{u}) - \nabla\times(\nabla\times\boldsymbol{u}). \tag{18}$$

Evaluating the above identity on an edge e and taking the dot product with corresponding normal vector $\boldsymbol{n}_e$, we have,

$$[\Delta\boldsymbol{u}]_e = [\nabla(\nabla\cdot\boldsymbol{u})]_e - [\nabla\times(\nabla\times\boldsymbol{u})]_e. \tag{19}$$

The first term on the right hand side of Eq. (19) is calculated using the gradient at edges in normal direction using the divergence

250 given at Voronoi centers. The second term of the right hand side of Eq. (19) is calculated as $\nabla\zeta(\boldsymbol{x}_e)\cdot\boldsymbol{t}_e$ (Skamarock et al., 2012), where $\boldsymbol{t}_e = \boldsymbol{k}\times\boldsymbol{n}_e$, and is obtained as the gradient at edges in tangent direction using the relative vorticity ζ given at primal vertex positions. In the simulations where diffusion was employed, we considered a constant diffusion coefficient, which is equal to $8225.13\,m^2 s^{-1}$ and is approximately what is used in spectral models with truncation level T85 (Jablonowski and Williamson, 2011).

255   The L_2 error will be refereed as root mean square error (RMS, hereafter) and the L_∞ error as the maximum error.

4.1   Shallow-water model

4.1.1   Steady geostrophic flow

This test case is proposed in Williamson et al. (1992) and it is known as test case 2 (hereafter, TC2). The initial conditions are given by:

260   $$h(\lambda,\phi,0) = h_0 - \frac{1}{g}\left(a\Omega u_0 + \frac{u_0^2}{2}\right)\sin^2\phi, \tag{20}$$

$$u(\lambda,\phi,0) = u_0\cos\phi, \tag{21}$$

$$v(\lambda,\phi,0) = 0. \tag{22}$$

The bottom topography is equal to zero. The functions defined above satisfy the shallow water equations and represent a balanced geostrophic flow. Thus, the initial condition should remain constant. We define the parameters $h_0 = 3 \times 10^3$ and

265    $u_0 = \frac{2\pi a}{12 days}$, where a is the Earth radius.

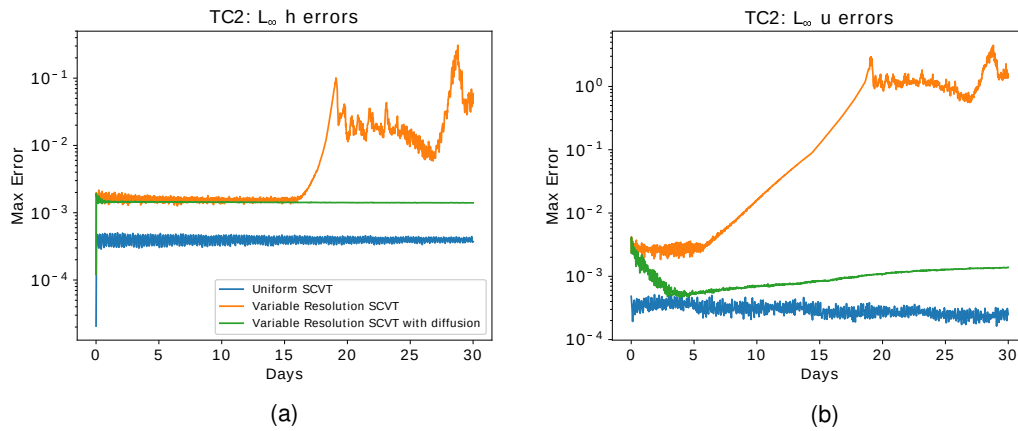



Figure 5. Steady geostrophic flow test case: Time evolution of relative error in the L_∞ norm for the height field (a) and the normal component of the velocity field (b). The blue lines indicate the errors for the UR7 grid, while the orange lines show the error for the VR7 grid with refinement based on topography and the green lines show the error for the same VR7 grid but adding numerical diffusion in the shallow-water model.

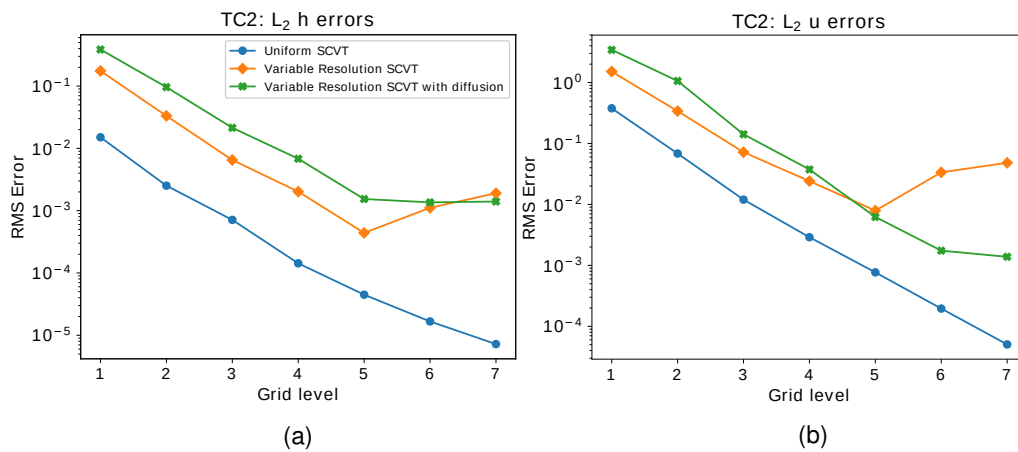



Figure 6. Steady geostrophic flow test case: Convergence of relative error in the L_2 norm for the height field (a) and the normal component of the velocity field (b). The blue lines indicate the errors for the uniform SCVT grids, while the orange lines show the error for the variable resolution grids with refinement based on topography and the green lines show the error for the same variable resolution grids but adding numerical diffusion in the shallow-water model.

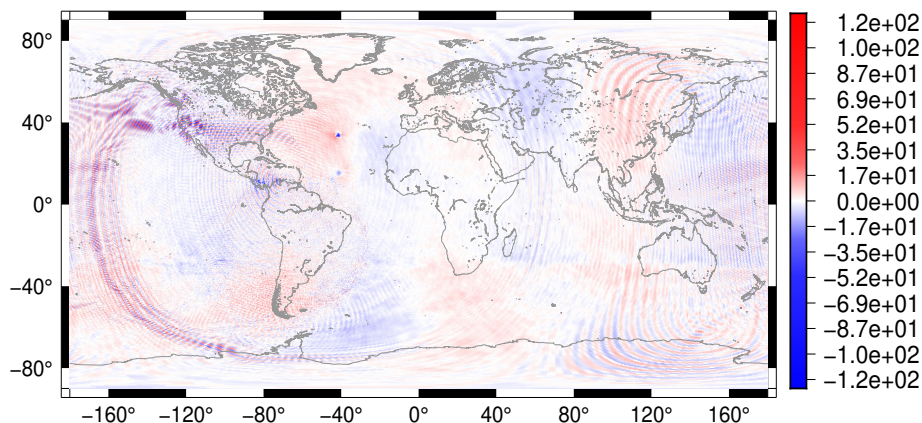
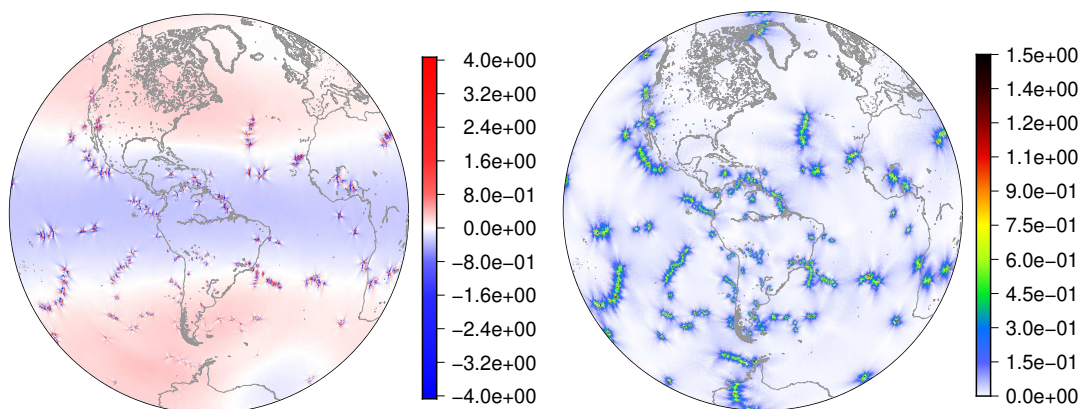



Figure 7. Steady geostrophic flow test case: Plots for the VR7 grid. (a): height field error for each grid cell after 30 days. (b): height field error for each grid cell after 30 days using numerical diffusion in the shallow-water model. (c): alignment index.

In Fig. 5 we show the evolution of the relative error in norm L_∞ for the grids of level 7 (UR7 and VR7), where we included a simulation on the variable resolution grid adding numerical diffusion. This result shows that the error for the normal component of the velocity field grows since the beginning of the integration for the VR7 grid without numerical diffusion. This error triggers errors in the height field that start to grow significantly after day 15. This behavior does not occur in the UR7

270 grid. However, this error is mitigated when we add numerical diffusion. Thus, the error evolution in the variable resolution with numerical diffusion has similar behavior to the UR7 grid. The total energy is conserved with precision 10^{-9} without diffusion and 10^{-4} with diffusion.

Furthermore, Fig. 6 shows that adding diffusion in the variable resolution SCVT grid simulations is needed to achieve better convergence in the L_2 norm. We can see that the error does not converge to zero for both height and velocity fields, even when

we add diffusion. Nevertheless, lack of convergence is also observed in the uniform resolution SCVT grids, although in the maximum norm, as it is shown in Peixoto (2016).

At last, Fig. 7 shows the error after 30 days for the height field in the VR7 grid without (Fig. 7a) and with diffusion (Fig. 7b). We also show the alignment index defined by Peixoto and Barros (2013), which is a measure of how aligned a cell is and may be used to detect problematic grid cells that can impose larger errors. The higher is this index, the more ill aligned is the cell, and a worse accuracy is expected in the cell. From Fig. 7a, we can see that the error for the height field propagates similarly to gravity waves, although spurious, when no numerical diffusion is employed. From Fig. 7b and 7c, it is clear that the error obtained for the height field is greater in the cells that are ill aligned. This result is in agreement with the results from Peixoto and Barros (2013), where they show that the divergence discretization accuracy used in TRSK is related to the alignment index.

4.1.2 Flow over Andes mountain

This test case is an adaption of the flow over a mountain test case proposed by Williamson et al. (1992), where we replace the bottom topography b with the smooth Andes topography described in Sect. 2.2. The velocity field is initialized as in the previous test case of the steady geostrophic flow. The initial height field is then given by:

$$h(\lambda, \phi, 0) = h_0 - \frac{1}{g}\left(a\Omega u_0 + \frac{u_0^2}{2}\right)\sin^2\phi - b(\lambda, \phi). \tag{23}$$

We adopted $h_0 = 5400$ and normalized the bottom topography to lie in the interval $[0, 2000]$. Since the exact solution for this test case is not available, we use as a reference solution the solution yielded by ENDGame, a semi-Lagrangian dynamical core model developed by Thuburn et al. (2010), in a latitude-longitude grid of 720×1440 points. We interpolate the values from the latitude-longitude grid to the grid points needed in SCVT grids using cubic interpolation.

We ran this test case for 30 days on the uniform and variable resolution SCVT grids of level 7 (UR7 and VR7, respectively). In Fig. 8, we show the error of the height field considering the ENDGame solution as a reference. We can notice that the error is initially concentrated on the Andes mountain (Fig. 8a) and it is propagated to the other regions of the grid (Fig. 8b-c) and finally on the whole grid (Fig. 8 (d)). A similar result is obtained in Tian (2020), where the author considers this test case using a cosine bell topography instead of the smoothed Andes topography and it is shown that this test case generates errors that propagate as gravity waves and these errors can be larger in the high-resolution grid region.

Our simulation shows that, even though the Andes is represented more accurately in the VR7 grid, the height field errors are larger in the Andes mountain in the first days of integration. Despite that, the magnitude of the does not increase as in the steady geostrophic flow test case. However, Fig. 9a illustrates how the L_∞ error for the velocity field increases with time. Even though they increase, they don't trigger larger errors as in the steady geostrophic flow test case. Figure 9b shows that the normal component of the velocity error is greater in the Africa continent, which is a coarse grid region. This velocity field error in the VR7 grid is mitigated when we add diffusion in the shallow-water model, as Fig. 9a is showing. Again, the addition of diffusion allowed us to have a similar temporal error evolution in both variable and uniform resolution SCVT grids, breaking the total energy conservation which is conserved with precision 10^{-9} without diffusion and 10^{-5} with diffusion.

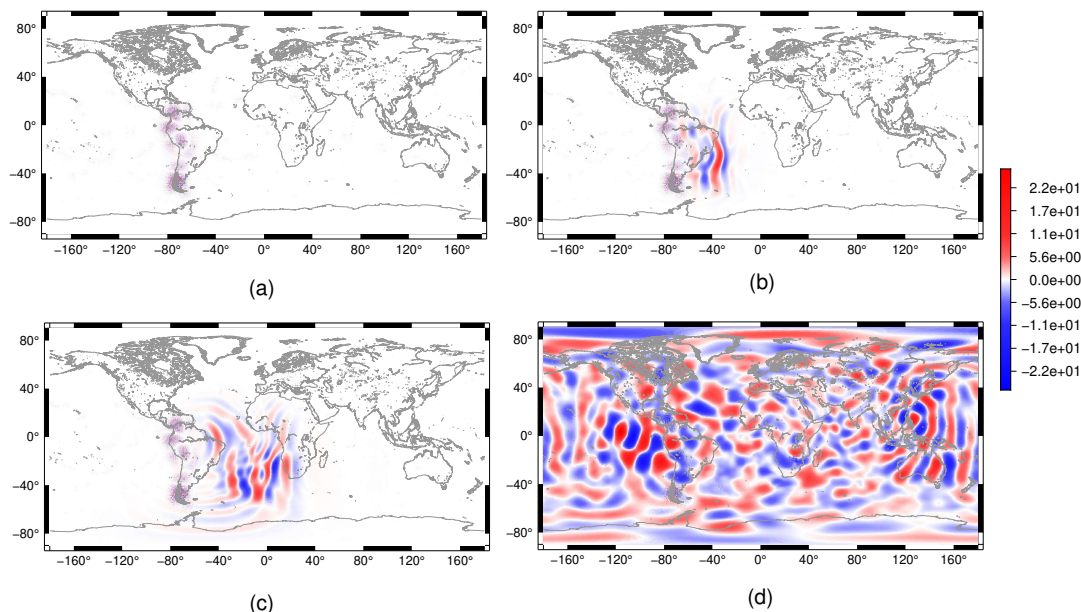



Figure 8. Flow over Andes mountain test case for the shallow-water model: Height field cellwise error plots for the VR7 grid after 1 hour (a), 2 days (b), 3 days (c), and 30 days (d). We are considering the ENDGame solution as a reference solution.

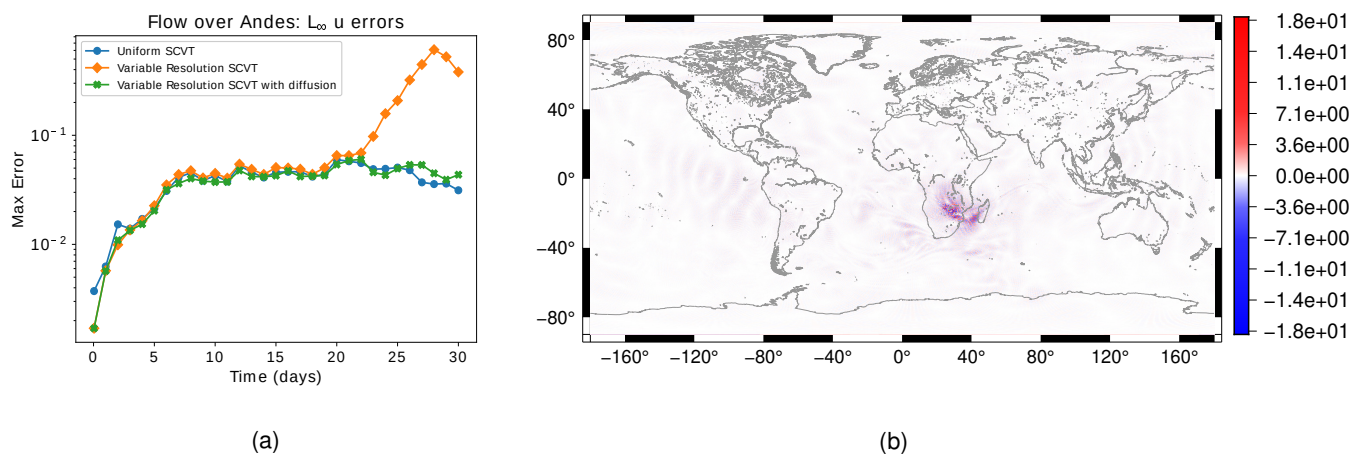



Figure 9. Flow over Andes mountain test case for the shallow-water model: (a) Time evolution of relative error in L_∞ norm for the normal component of the velocity field for the UR7 grid (blue line), for the VR7 with (green line) and without (orange line) numerical diffusion. (b) Cellwise error for the normal component of the velocity field at day 30 for the VR7 grid. We are considering the ENDGame solution as a reference solution.

4.1.3 Barotropic unstable zonal jet

Now, we shall analyze the test case proposed by Galewsky et al. (2004). The initial fields are given by:

$$u(\lambda,\phi,0) = \begin{cases} 0 & \text{if } \phi \leq \phi_0 \\ \frac{u_{\max}}{e_n} \exp \frac{1}{(\phi-\phi_0)(\phi-\phi_1)} & \text{if } \phi_0 < \phi \leq \phi_1, \\ 0 & \phi \geq \phi_1, \end{cases} \tag{24}$$

310    $$v(\lambda,\phi,0) = 0, \tag{25}$$

$$h(\lambda,\phi,0) = h_0 - \frac{1}{g}\int^{\phi} au(\lambda,\phi',0)\left(f + \frac{\tan\phi'}{a}u(\lambda,\phi',0)\right)d\phi', \tag{26}$$

where we define $\phi_0 = -5°$, $\phi_1 = -45°$, $u_{\max} = 80$ and $e_n = \exp-\frac{4}{(\phi_0-\phi_1)^2}$. These parameters define the jet in the Southern hemisphere. At last, we add the following perturbation in the height field in order to trigger the instability.

$$\hat{h}(\lambda,\phi) = \hat{h}\cos(\phi)e^{-\left(\frac{\lambda}{\alpha}\right)^2}e^{-\left(\frac{\phi-\phi_2}{\beta}\right)^2} \tag{27}$$

315   where $\hat{h} = 120$, $\alpha = \frac{1}{3}$, $\beta = \frac{1}{15}$ and $\phi_2 = -25°$. The bottom topography is set to zero.

We ran this test for 7 days for the VR7 grid and the UR7 and UR8 grids. In Fig. 10 we illustrate the potential vorticity. On day 5, in the VR7 grid, numerical noise is generated in a few grid cells on the topography-based refined region, as it is shown in Fig. 10a. This noise propagates to other grid cells at day 6 (Fig. 10b) and its magnitude reduces. These results show that the performance of the VR7 grid is problematic for this test case. For this reason, we ran again this test employing numerical

320   diffusion in the shallow-water model. As Fig. 10c shows, the numerical noise in the VR7 grid is mitigated when we add numerical diffusion. In Fig. 10d and Fig. 10e, we depict the potential vorticity at day 6 for UR7 and UR8 grids, respectively, where no numerical diffusion was employed. When we compare the potential vorticity at day 6 for the grids VR7 and UR7, both of level 7, with the UR8 grid, we can observe that the potential vorticity in the VR7 grid (Fig. 10c), considering numerical diffusion, represents better the vortex formed near to Andes region than in the uniform resolution grid with the same number

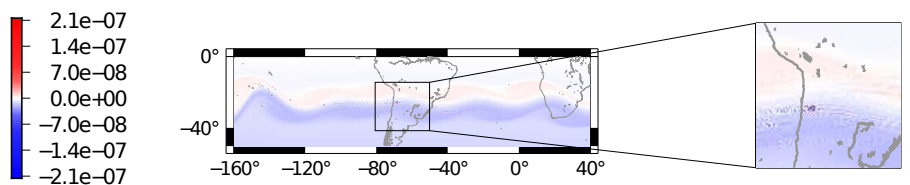
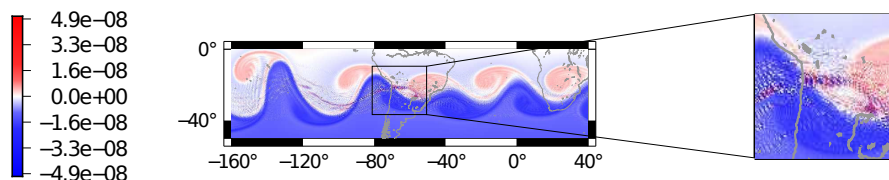
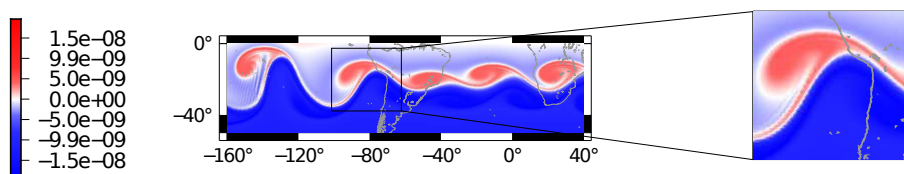
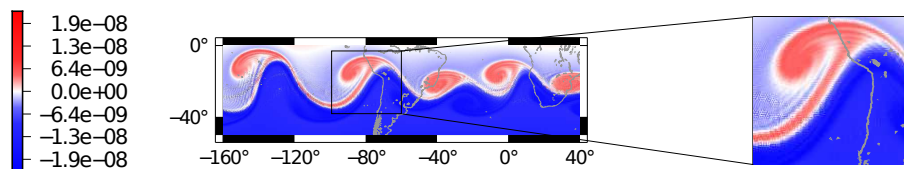
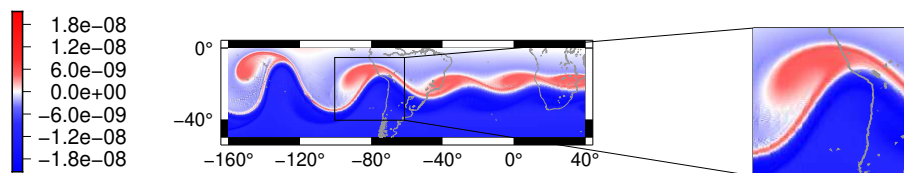
325   of cells (Fig. 10d), i.e., the vortex formed near to the Andes region in the VR7 grid with numerical diffusion is more similar to the vortex formed in the same area in the UR8 grid (Fig. 10e) compared to the vortex generated in the UR7 grid. This simulation shows that the variable resolution grid has a good performance in this test case when we add numerical diffusion in the shallow-water model compared to the uniform grid with the same number of grid cells, showing a computational advantage of the variable resolution grid.

(a) VR7 at day 5

(b) VR7 at day 6

(c) VR7 at day 6 - with diffusion

(d) UR7 at day 6

(e) UR8 at day 6

Figure 10. Barotropic unstable zonal jet for the shallow-water model: Potential vorticity for the VR7 grid at (a) day 5 and (b) day 6 without diffusion. (c) shows the potential vorticity at day 6 with diffusion in the shallow-water model for the VR7 grid. In (d) and (e) we depict the potential vorticity at day 6 for the UR7 and UR8 grids, respectively, as references.

4.1.4 Matsuno baroclinic wave

This test case has been proposed by Shamir et al. (2019) and considers analytical wave solutions to the linearised shallow-water equations on the beta plane that approximate analytical solutions to the shallow-water equations on the sphere for low gravity waves speed and are taken as initial and reference solutions for this test case. It is recommended to analyse two types of waves: eastward inertia-gravity wave (EIGW, hereafter) and Rossby wave (RW, hereafter). We follow Shamir et al. (2019) analyzing both waves in this work. The mean atmosphere depth is set to $H = 30$m, the wavenumber is defined as $k = 5$ and the mode number is defined as $n = 1$. The EIGW period is $T = 1.9$ days and the RW period is $T = 18.5$ days. We recall that the time-frequency is defined by $\omega = \frac{2\pi}{T}$. For these parameters, the analytical solutions are expected to be waves that propagate in the zonal direction, where the EIGW propagates in the eastward direction and the RW propagates in the westward direction. In order to guarantee that waves pass through the whole Andes topography based locally refined region, we applied a three-dimensional rotation of $21°$ in the $x - z$ plane in the VR7 grid. We integrate the shallow-water equations for 100 wave periods for each EIGW and RW, as it is suggested in Shamir et al. (2019), for both rotated VR7 and UR7 grids to investigate any local refinement impact on the solution.

We define the variable $h' = h - H$, in order to analyze the mean atmosphere depth perturbation and for plot reasons. In Fig. 11 we depict the h' field obtained for the EIGW (Fig. 11a) and the analytical h' (Fig. 11b) in the VR7 grid. The same results for the RW are shown Fig. 11c and Fig. 11d. For both EIGW and RW, the error caused is due to phase error, which is a common error for many numerical schemes and similar results are obtained for the UR7 grid. Despite the waves propagating along the locally refined region, the obtained solutions preserved the wave shape even for a long time of integration.

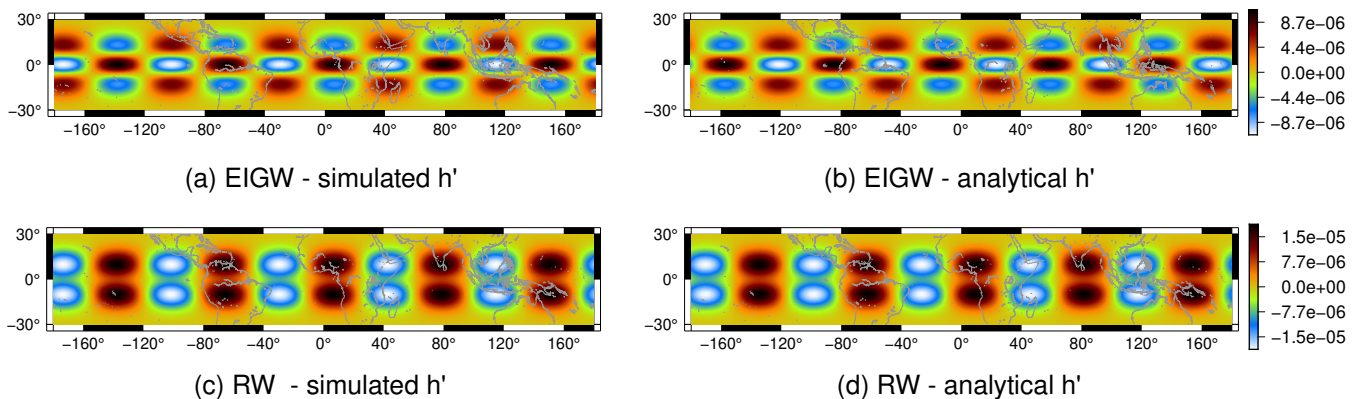



Figure 11. Matsuno baroclinic wave test case: Simulated h' for the EIGW in the rotated VR7 grid (a) and the analytical h' (b) after 100 EIGW periods (190 days). In (c) and (d) we depict the simulated and analytical h' for the Rossby wave in the rotated VR7 grid after 100 RW periods (1850 days).

We also depict the Hovmöller diagrams for h' considering the EIGW and RW in Fig. 12 and Fig. 13, respectively, for both rotated VR7 and UR7 grids. The latitude-time Hovmöller diagram is obtained intersecting h' at $\lambda = -65°$ and the longitude-time Hovmöller diagram is obtained intersecting h' at $\phi = 0°$. As Fig. 12a and (b) show, the VR7 and UR7 grids have the same

phase error for h' in the EIGW test case. From Fig. 12c-d, comparing the black lines with the white line, we can conclude that the obtained EIGW is propagating in agreement with the analytical speed for both grids since the analytical solutions propagate zonally, the longitude-time Hovmöller diagrams are expected to be straight lines with slope $\frac{k}{\omega}$ (Shamir et al., 2019). The same analysis holds for the RW Hovmöller diagrams depicted in Fig. 13. Therefore, the local refinement introduces no impact in the

355   obtained solution for this test case as we observed in the previous test cases.

(a) Rotated VR7   (b) UR7   (c) Rotated VR7   (d) UR7

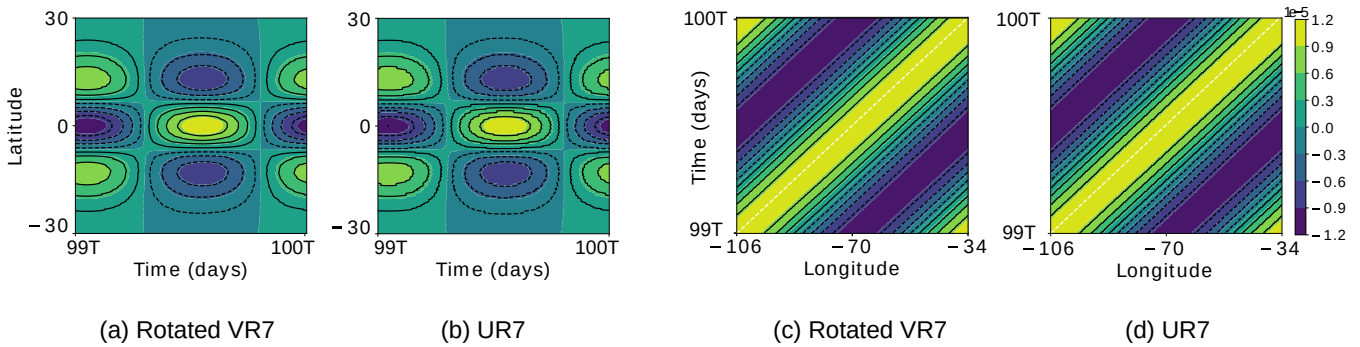



Figure 12. EIGW latitude-time Hovmöller diagrams for the simulated h' in the rotated VR7 (a) and UR7 (b) grids, obtained fixing the longitude at $\lambda = -65°$, and EIGW longitude-time Hovmöller diagrams for the simulated h' in the rotated VR7 (c) and UR7 (d) grids obtained fixing the latitude at $\phi = 0°$. T denotes the period of the EIGW (1.9 days). The white dashed line has slope $\frac{k}{\omega}$, where ω and k are the EIGW time-frequency and wavenumber, respectively.

(a) Rotated VR7   (b) UR7   (c) Rotated VR7   (d) UR7

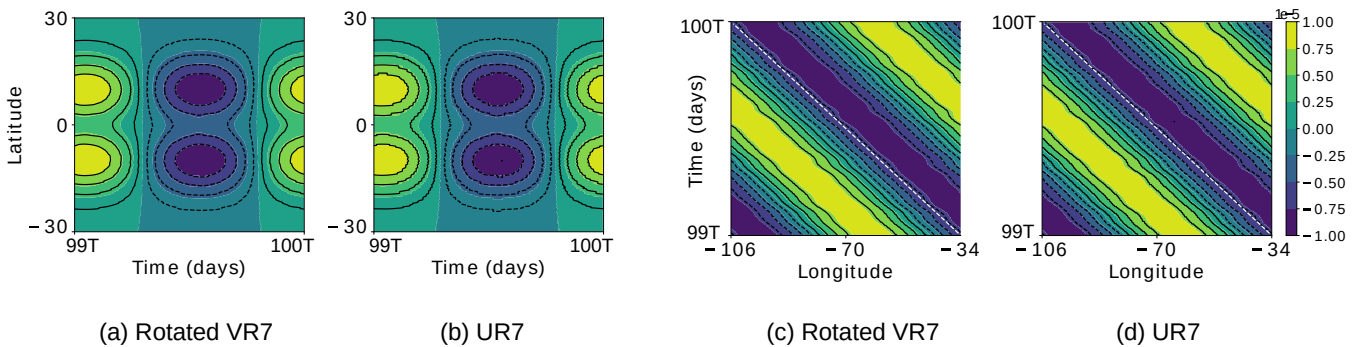



Figure 13. Same as Fig. 12 for the RW.

    We remark that these test cases proposed by Shamir et al. (2019), while providing valuable insight in wave propagation properties, seem to be little sensitive to different numerical methods. We expect that most numerical methods would perform well on these tests. This was the case for our experiments on both uniform and locally refined grids and was also noticed in Brachet and Croisille (2021) for a cubed-sphere based model.

4.2 Moist shallow-water model

We now focus our analyses of the variable resolution grids on the moist shallow-water model framework (Zerroukat and Allen, 2015), described in Sect. 3.2. We consider two test cases. The first test case is suggested by Zerroukat and Allen (2015) and is similar to flow over a mountain test case (Williamson et al., 1992). The second test case is similar to the barotropic unstable jet test case (Galewsky et al., 2004). This test case was analyzed in Ferguson et al. (2019), however, in their work, the rain formed is considered as precipitated and is removed from the model. In our analysis, we consider the rain to be advected as a tracer following Zerroukat and Allen (2015).

The vapor, cloud, and rain variables in the moist shallow-water model must assume only non-negative values. For this purpose, we employed a monotonic filter to ensure that the moisture variables assume only non-negative values. At each time step of the model integration, we apply the monotonic filter for each tracer (vapor, cloud and rain), which is implemented as an iterative process. At each iteration of the monotonic filter, the tracer mass is computed for each grid cell. The grid cells with negative tracer mass are set to zero mass and its negative mass is distributed equally in the neighbor's cells. This iterative process is repeated until the maximum magnitude of the negative values is small enough. After that, all the remaining mass negative values are set to zero and the tracer value for a cell is computed as the value of its mass divided by its grid cell area. This process, even though does not conserve the tracer mass, showed to avoid negative mass and the total mass of $h(q_v + q_c + q_r)$ relative variation is less than 10^{-5} for the simulations that we shall see in the remaining of this section.

4.2.1 Flow over a mountain

This test case is proposed in Zerroukat and Allen (2015). We only slightly changed the initial conditions by translating the initial fields in 30 degrees in the longitude in order to ensure cloud and rain formation in the South American continent. The initial height and velocity are defined as in the flow over a mountain test case for the classical shallow-water model described in Sect. 4.1.2. The initial temperature and vapor, respectively, are given by:

$$\theta(\lambda,\phi,0) = F(\theta^{SP}, (1-\mu_1)\theta^{EQ}, \theta^{NP}, \phi) + \mu_1 \theta^{EQ} \cos\phi \sin\left(\lambda + \frac{\pi}{6}\right), \tag{28}$$

$$q_v(\lambda,\phi,0) = \mu_2 q_{sat}(b, h(\lambda,\phi,0), \theta(\lambda,\phi,0)), \tag{29}$$

where F is the function given by:

$$F(f_1, f_2, f_3, \phi) = \frac{2}{\pi^2}\left[\phi\left(\phi - \frac{\pi}{2}\right)f_1 - 2\phi\left(\phi + \frac{\pi}{2}\right)\left(\phi - \frac{\pi}{2}\right)f_2 + \phi\left(\phi + \frac{\pi}{2}\right)f_3\right]. \tag{30}$$

The parameters are set as $\theta^{NP} = -40\varepsilon$, $\theta^{EQ} = 40\varepsilon$, $\theta^{SP} = -20\varepsilon$, $\varepsilon = \frac{1}{300}$, $\mu_1 = 0.05$, $\mu_2 = 0.98$. The function q_{sat} is described in Sect. 3.2, Eq. (17). The topography is given by:

$$b(\lambda,\phi) = \begin{cases} 2000\left(1 - \frac{r}{r_{\max}}\right) & \text{if } r \leq r_{\max} \\ 0 & \text{otherwise.} \end{cases} \tag{31}$$

where $r = \sqrt{(\lambda - \lambda_0)^2 + (\phi - \phi_0)^2}$, $r_{max} = \frac{\pi}{9}$, $\lambda_0 = \frac{2\pi}{3}$ and $\phi_0 = -\frac{\pi}{6}$. The initial fields are shown in Fig. 14. The initial cloud and rain are set as zero.

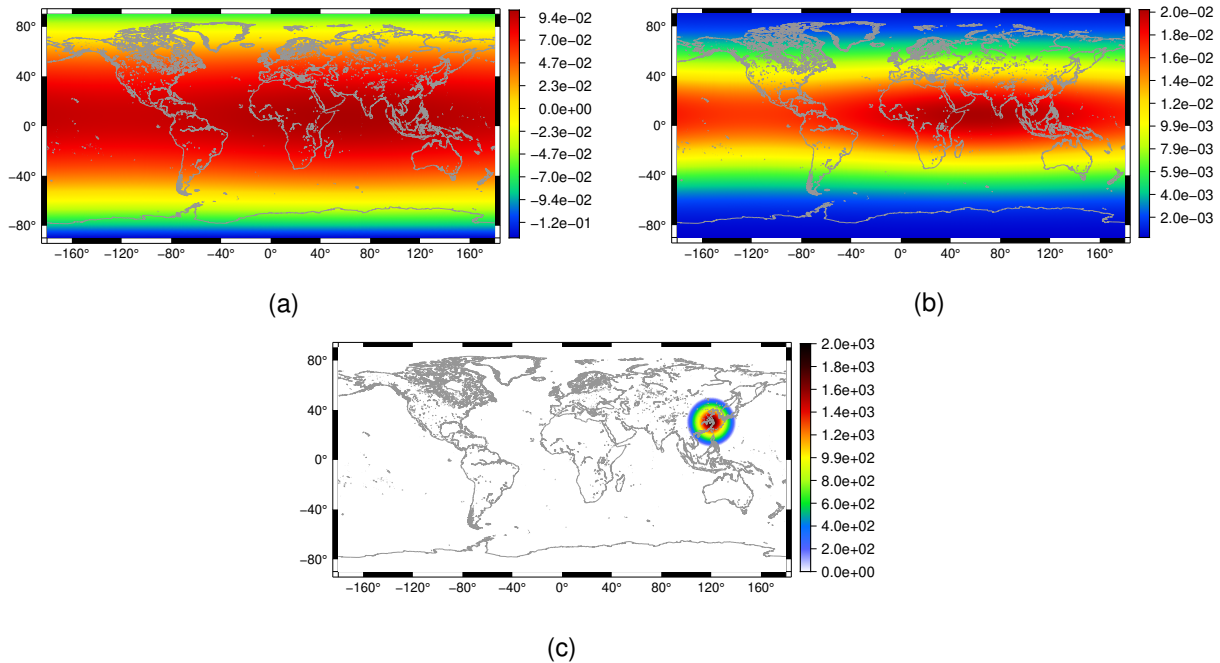



Figure 14. Flow over a mountain test case for the moist shallow-water model: initial conditions for the vapor (a) and temperature (b) fields, and the considered topography (c).

We ran this test for 30 days on the uniform resolution SCVT grids, using the grids UR6, $\cdots$, UR9 and the VR7 grid. In Fig. 15 we present the results for the cloud field in the South of the continent of South America at day 30 for the grids considered in our simulations. Figure 16 illustrates the same for the rain field. We employed numerical diffusion in the VR7 grid simulation.

From Fig. 15a-d and Fig. 16a-d we can notice that both cloud and rain fields are converging in the uniform resolution SCVT grids. For the variable resolution SCVT grid with local refinement on the Andes mountain, the cloud field is better represented in the refined region, when we compare Fig. 15e and Fig. 15b with Fig. 15d. This illustrates a potential benefit of the locally refined grid, as the resulting clouds are well represented in the refined region while having lower computation cost with respect to the globally high-resolution uniform grid. However, when we look at the rain field in Fig. 16, we can see that the rain in the VR7 grid (Fig. 16e) is more similar to the rain in the UR7 (Fig. 16b) than in the UR8 (Fig. 16c) or UR9 (Fig. 16d), when we look in the topography based refined region. Notice that the formed rain in the topography-based refined region that is not formed in the UR8 and UR9 grids is present in the UR6 and UR7 grids. This indicates that the low-resolution region of the variable resolution grid is affecting the observed rain in the high-resolution region.

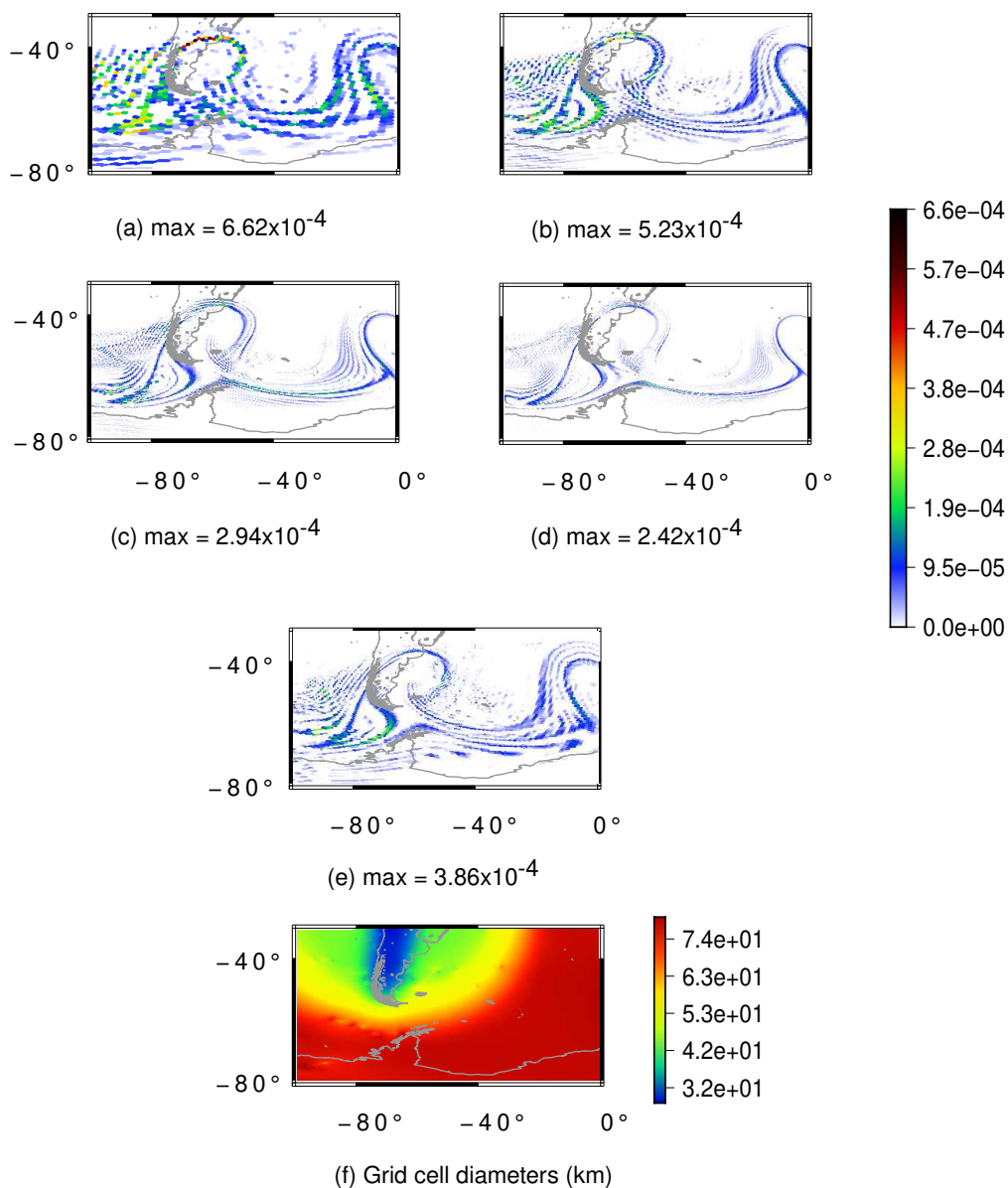



Figure 15. Flow over a mountain test case for the moist shallow-water model: cloud field at day 30 for the grids UR6 (a), UR7 (b), UR8 (c), UR9 (d), VR7 (e) and the cell diameters in km for the VR7 grid (f). The UR6, UR7, UR8 and UR9 grids have an average resolution of 120 km, 60 km, 30 km and 15 km, respectively.

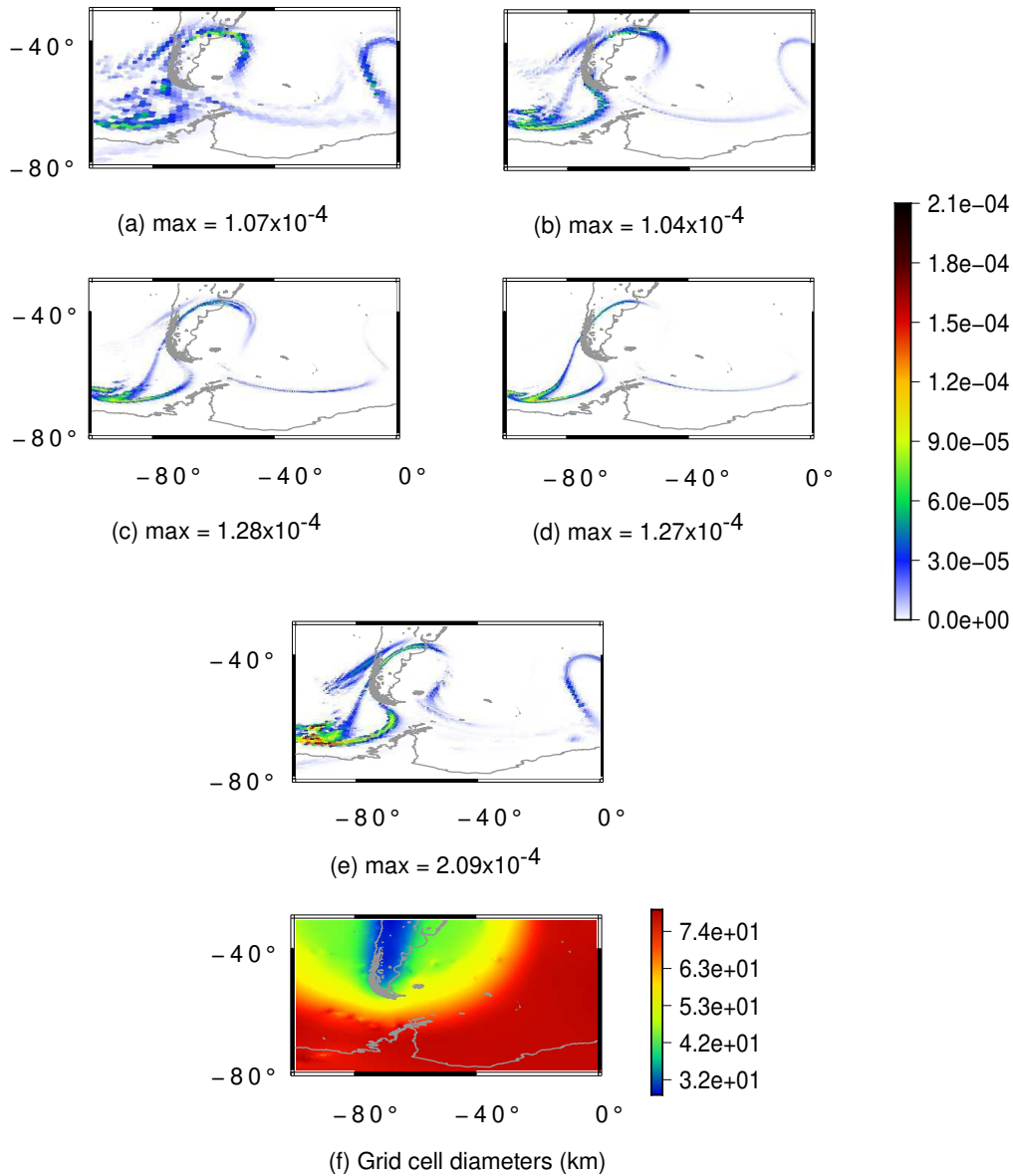



Figure 16. Flow over a mountain test case for the moist shallow-water model: rain field at day 30 for the grids UR6 (a), UR7 (b), UR8 (c), UR9 (d), VR7 (e) and the cell diameters in km for the VR7 grid (f). The UR6, UR7, UR8 and UR9 grids have an average resolution of 120 km, 60 km, 30 km and 15 km, respectively.

4.2.2 Barotropic unstable zonal jet

In this test case, we initialize the height and velocity fields as in the unstable barotropic jet for the classical shallow-water model (Sect. 4.1.3). The temperature and vapor are set as in the previous test (Eq. (28) and Eq. (29)) with $\mu_1 = 0$. This test

405 considers the bottom topography zero on the whole sphere. A perturbation in the height field is also added and we consider the

longitude coordinated translated using the transformation $\lambda' = \lambda - 0.93\pi$ only for the perturbation since the other fields do not depend on λ.

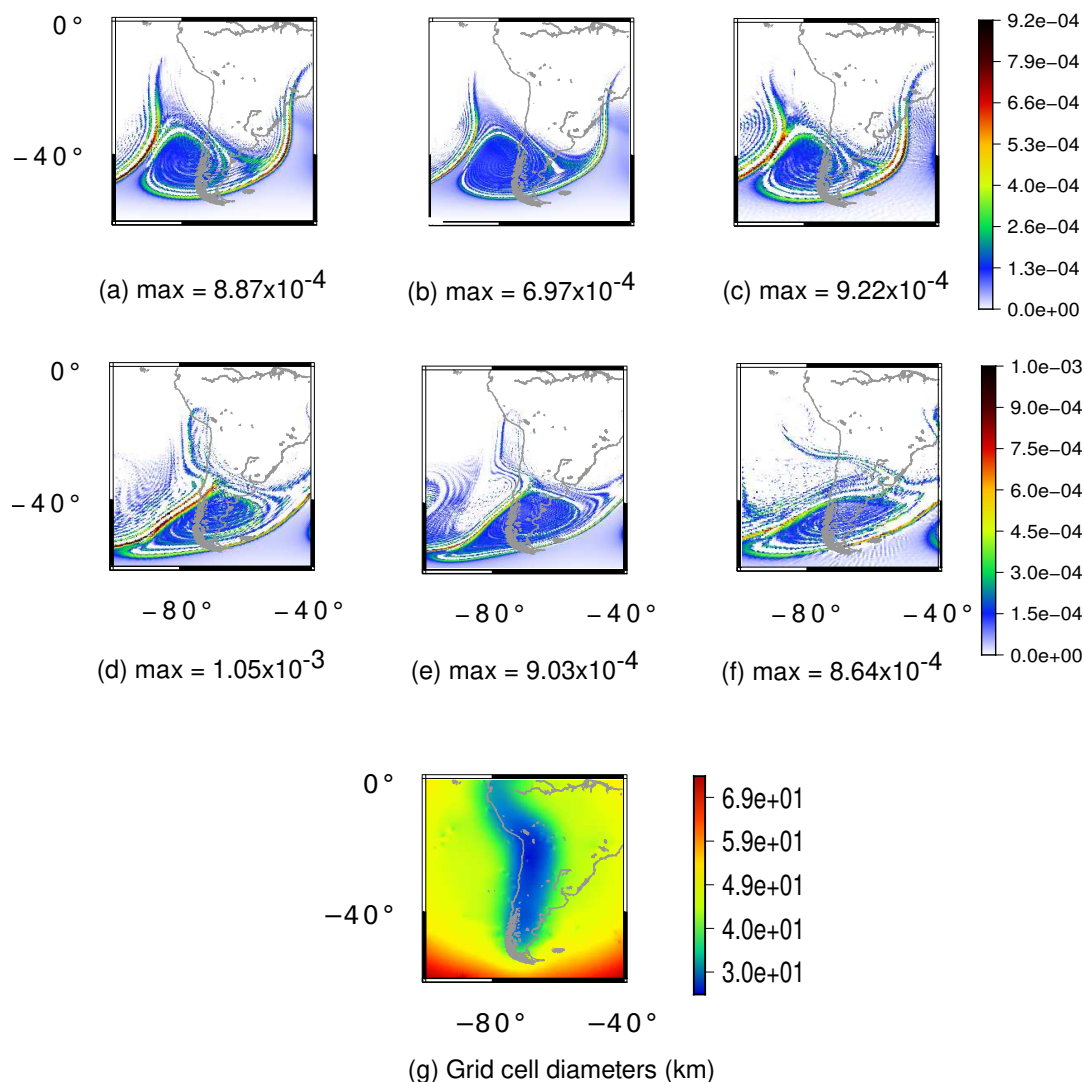



Figure 17. Unstable barotropic jet test case for the moist shallow-water model. (a)-(c): cloud field at day 7 for the UR8 and UR9 grids and VR7 grid, respectively. (d)-(f): cloud field at day 8 for the UR8 and UR9 grids and VR7 grid, respectively. (g): grid cell diameters in km for the VR7 grid. The UR8 and UR9 grid have an average resolution of 30 km and 15 km, respectively.

We ran this test for 8 days in the UR8, UR9 and VR7 grids, where we consider numerical diffusion only for the VR7 grid just as the previous test case. Figures 17 and 18 show cloud and rain formation for these grids at days 7 and 8. The rain at day 7 in the variable resolution grid (Fig. 18c) has a higher magnitude when compared to the UR8 (Fig. 18a) and UR9 grids (Fig.

18b). Nevertheless, this is expected since the region of the maximum magnitude of the rain is in the transition zone between the low and high-resolution grid. A similar analysis applies for the cloud field at day 7.

At day 8, the rain is generated on Andes mountain in the UR8 (Fig. 18d) and UR9 grids (Fig. 18e). However, this rain formation is not properly generated on the Andes region (Fig. 18f), being misplaced westward of the Andes, even though the
415 Andes is better represented in our variable resolution grid. This illustrates how the local refinement was not able to solve the rain field in this case, and influenced the large-scale dynamics. Once again, a similar influence is observed for the cloud field on day 8.

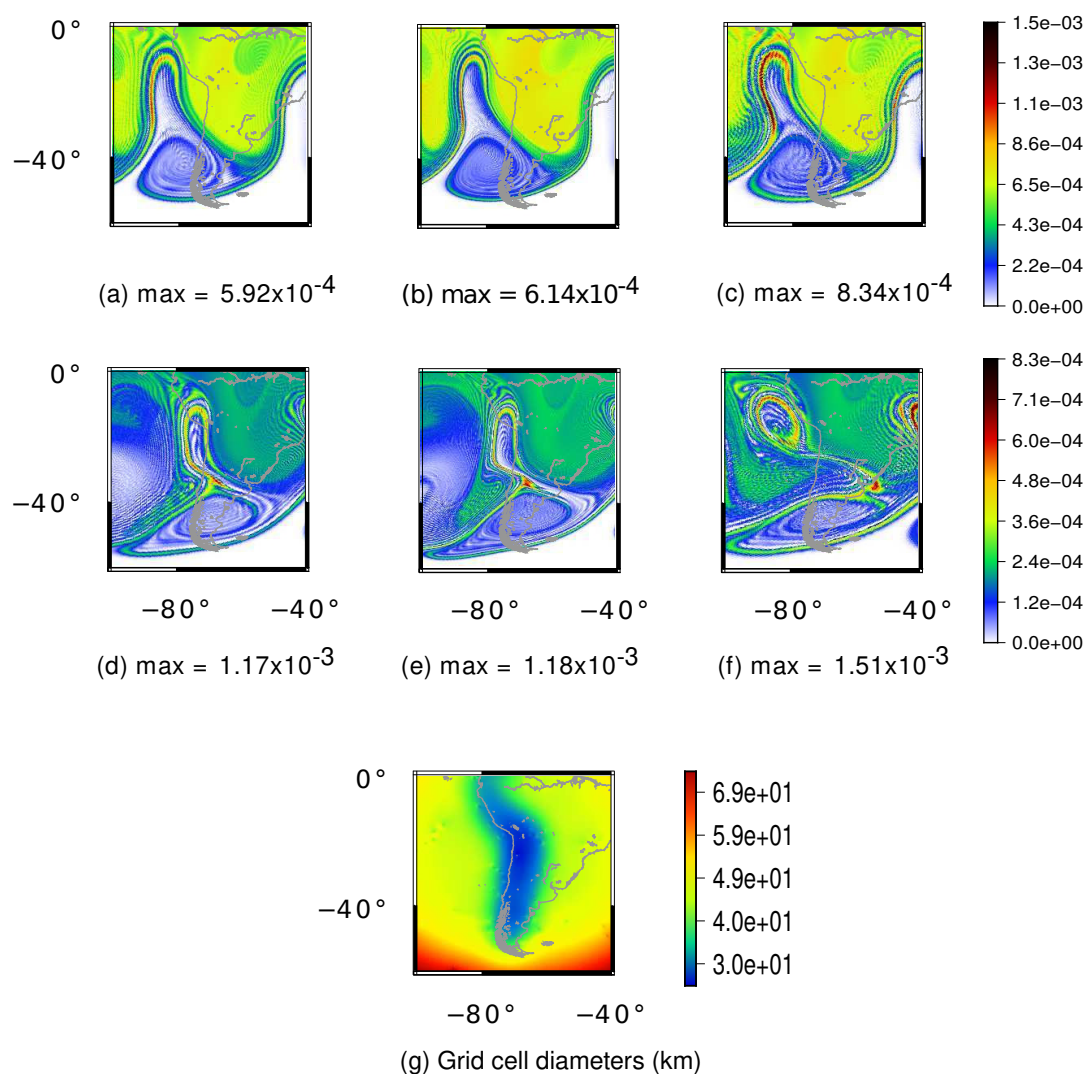



Figure 18. Unstable barotropic jet test case for the moist shallow-water model. (a)-(c): rain field at day 7 for the UR8 and UR9 grids and VR7 grid, respectively. (d)-(f): rain field at day 8 for the UR8 and UR9 grids and VR7 grid, respectively. (g): grid cell diameters in km for the VR7 grid. The UR8 and UR9 grid have an average resolution of 30 km and 15 km, respectively.

5   Conclusions

In this work, we developed SCVT grids with local refinement based on topography, that refines and captures well the Andes
range and is smoothly transitioned to a circular grid region centered in South America. This circular region is again smoothly
transitioned to a coarse global uniform grid. We used smoothing techniques in the Andes topography data in order to generate
these grids using Lloyd's method. These grids were developed in the hope to provide a better basis for weather forecasting in
South American, aiming at reduced computational cost. We evaluate the TRSK method on these grids, considering the classical
shallow-water model and a moist shallow-water model proposed by Zerroukat and Allen (2015), discussing the pros and cons
of the developed grids in this work.

We began our analysis with the classical shallow-water model. Although the TRSK method is designed for arbitrary or-
thogonal C-grids, in particular SCVT grids, we showed some of its deficiencies when considering the SCVT grid with local
refinement based on topography. For instance, in test case 2 from Williamson et al. (1992), the error in the normal velocity
component grows since the beginning of the model integration, triggering spurious gravity waves in the height field. This test
case is known for being stable for a long time of integration in the uniform resolution SCVT grids. Therefore, our results show
a negative impact on the local refinement. However, we could mitigate the error evolution of the velocity field by adding a small
diffusion. After adding diffusion, gravity waves were not generated in the height field and we obtained a similar evolution of the
errors with time to the ones obtained in the uniform SCVT grids. For test case 5 from Williamson et al. (1992), we replace their
mountain considering a smooth Andes topography data. This test case showed that the height errors were concentrated over the
Andes and generated gravity waves. This result is undesirable since the height errors were greater in the high-resolution region.
We could notice that the error in the normal component of the velocity started to grow significantly after 23 days, but this was
mitigated with small diffusion. For the unstable barotropic jet test case from Galewsky et al. (2004), we observed that a lot of
numerical noise was observed in the potential vorticity on the Andes refined region. Again, diffusion mitigated the problem,
and a vortex near to Andes formation was better represented in the variable-resolution grid in comparison with an uniform grid
considering the same number of cells. From these results, we can conclude that the local refinement introduced more errors
in the velocity field that may generate spurious gravity waves visible in the height field. A small diffusion is necessary to be
employed to stabilize the velocity field error.

The moist-shallow water model developed by Zerroukat and Allen (2015) includes moist variables in the classical shallow-
water model, where conversion of vapor to cloud, cloud to vapor and cloud to rain is possible. We analyzed the TRSK method
on the variable-resolution grids considering two test cases. In the first test, similar to the test case 5 from Williamson et al.
(1992), we observed that the cloud field after 30 days was better represented in the variable-resolution grid of level 7 than in
the uniform grid of the same level. However, for the cloud field, we notice spurious rain in the transition zone. Furthermore,
spurious rain was also detectable over the Andes refined region, which we address as an interference of the coarse grid. For the
unstable barotropic test case with moist, we notice that both cloud and rain fields were not properly represented in the Andes
refined region after 8 days in comparison with a high-resolution solution, where we observed a misposition of the rain and
cloud on the Andes region, despite the employment of numerical diffusion.

Our work shows that, despite of its flexibility, working with general local refinements in SCVT grids using TRSK, which is also possible in the MPAS global model, requires some care. The main issues are related to spurious numerical waves that may trigger numerical instabilities, requiring numerical diffusion to solve this problem. This directly impacts the total energy

455 conservation of TRSK. Furthermore, the grid may influence rain and cloud formation in the refined region. Overall, our results indicate that locally refined grids can be effective for short periods of time (few days), but require additional caution for longer periods.

We remark that our tracer equations were discretized here considering the same scheme as used for the continuity equation for the TRSK scheme, which is very low order. MPAS employs a high-order advection scheme for the tracer equations, which

460 could reduce some of the issues encountered here. Finally, the grids developed in this work can be applied directly in the MPAS model and its impact may be investigated considering the full three-dimensional model.

Code availability. All codes used in this work are openly provided in the GitHub repository https://github.com/pedrospeixoto/imodel and the release version relative to results of this paper may be accessed as the Zenodo file http://doi.org/10.5281/zenodo.4614571.

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465 LS did the implementation of the numerical methods. Both authors wrote and revised the manuscript.

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