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Uncertainty of Low-Degree Space Gravimetry Observations: Surface Processes Versus Earth's Core Signal

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Key Points:

- Deep Earth's processes occur at large spatial and inter-annual temporal scales
- Time-lapse gravity satellite data are compared with geophysical models at scales of interest
- Large uncertainties on satellite data and geophysical models conceal the gravity signals originated from the Earth's core

Supporting Information:

Supporting Information may be found in the online version of this article.

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Abstract Space gravity measurements have been mainly used to study the temporal mass variations at the Earth's surface and within the mantle. Nevertheless, mass variations due to the Earth's core might be observable in the gravity field variations as measured by Gravity Recovery and Climate Experiment (GRACE) and GRACE Follow-On satellites. Earth's core dynamical processes inferred from geomagnetic field measurements are characterized by large-scale patterns associated with low spherical harmonic degrees of the potential fields. To study these processes, the use of large spatial and inter-annual temporal filters is needed. To access gravity variations related to the Earth's core, surface effects must be corrected, including hydrological, oceanic or atmospheric loading (Newtonian attraction and mass redistribution). However, these corrections for surface processes add errors to the estimates of the residual gravity field variations enclosing deep Earth's signals. As our goal is to evaluate the possibility to detect signals of core origin embedded in the residual gravity field variations, a quantification of the uncertainty associated with gravity field products and geophysical models used to minimize the surface process signatures is necessary. Here, we estimate the dispersion for GRACE solutions as about 0.34 cm of equivalent water height (EWH) or 20% of the total signal. Uncertainty for hydrological models is as large as 0.89–2.10 cm of EWH. We provide estimates of Earth's core signals whose amplitudes are compared with GRACE gravity field residuals and uncertainties. The results presented here underline how challenging is to get new information about the dynamics of the Earth's core via high-resolution, high-accuracy gravity data.

Plain Language Summary The motions of the Earth's fluid core are deduced from ground and satellite measurements of the geomagnetic field variations. Because the long-term variations of the Earth's gravity field might be correlated to the Earth's magnetic field, new information about the Earth's fluid core and its density changes could be accessed with gravimetry. The observation of the core processes must be done at very large spatial scales, in which case it is necessary to use gravity data from satellites. However, variations in the Earth's gravity field are also created by heterogeneous superficial sources such as ocean and atmospheric currents, variations in water storage, etc. To recover a signature of the Earth's fluid core, we need first to remove all other known effects of larger amplitudes from satellite observations of the gravity field. Our study compares models of gravity variations for different sources in order to estimate their uncertainty. Such uncertainties are discussed in view of the expected amplitudes of signals originated from the core.

1. Introduction

Gravity field variations measured by the Gravity Recovery and Climate Experiment (GRACE) and GRACE Follow-On (GRACE-FO) missions are sensitive to the redistribution of masses located above, at or below the Earth's surface (Chen et al., 2022). GRACE and GRACE-FO (referred to as GRACE) satellite data are used to estimate the Earth's mass variations from regional to global scales since 2002 (Tapley et al., 2004; Landerer et al., 2020). For example, GRACE satellite data became essential to monitor the evolution of terrestrial water storage (TWS), ice sheets, glaciers and sea level in a worldwide changing climate (Tapley et al., 2019). GRACE satellite data are, by nature, integrative, so that it may be difficult to separate the sources of change in the gravity field. Each process has a specific spatial and temporal signature that can go from global to local and from the secular to the sub-daily scales (Figure 1). We refer to certain surface processes with the term "loading" defined here as the Newtonian attraction and mass redistribution associated with elastic deformation. By approximate order of magnitude, the processes include in GRACE records are tidal effects from extraterrestrial bodies, post-glacial rebound (e.g., Purcell et al., 2011), hydrological (e.g., Rodell et al., 2018), atmospheric (e.g., Kusche & Schrama, 2005) and oceanic (e.g., Dobslaw et al., 2017)

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loading, water mass displacement across ocean, hydrosphere and cryosphere (e.g., Pfeffer, Cazenave, & Barnoud, 2022), pre-seismic (e.g., Bouih et al., 2022), co-seismic and post seismic (e.g., Deggim et al., 2021) mass re-distributions, sea level changes (e.g., Adhikari et al., 2019; Horwath et al., 2022; Peltier, 2009) and finally core processes.

In addition to its primary purposes, some new applications of the GRACE measurements were proposed to study the deep Earth's interior. Panet et al. (2018) gave an example of possible seismic precursor in the mantle before Tohoku earthquake in 2011; this kind of signature was also observed before the Maule-Chile event (Bouih et al., 2022). Other authors have proposed to improve the knowledge of the dynamical processes of the Earth's core. Dumberry (2010b) and Dumberry and Mandea (2021) predicted a gravity perturbation generated by various core processes that might be observable on the low degrees of the gravity field. No signature of these perturbations has yet been observed in the gravity variations. However, Mandea et al. (2012) showed a correlation between the variations of the geomagnetic field and the gravity field. Processes of dissolution and crystallization at the core-mantle boundary (CMB) were advocated to explain this correlation (Mandea et al., 2015).

Established methods of seismic tomography, Earth's rotation, gravity and geomagnetic data analysis and geodynamic modeling constrain distributions of seismic velocity, density, electrical conductivity, and viscosity at depth, all depending on the internal structure of the Earth. Global Earth's interior models based on different observables often lead to rather different images. For example, the analysis of the time-variable magnetic field allows to focus on the dynamical features of the core field (Gillet et al., 2010, 2022). On the other hand, gaining information about the Earth's core from the analysis of the gravity field is difficult, because it requires to separate the different sources of signal with independent observations and/or models. In this context, gravimetry has the potential to bring new constraint about the density anomalies in the core and at its boundaries in a complimentary way to seismology (Koelemeijer, 2021).

One way to extract the Earth's core signal from gravity observations is to use independent information from models of shallower sources (i.e., water mass redistribution in the hydrosphere, ocean, atmosphere, cryosphere and solid Earth's processes associated with earthquakes and glacial isostatic adjustment (GIA)) to remove such larger amplitude contributions and to study the remaining signal. In this paper, we propose different models of post-glacial rebound, hydrological, atmospheric and oceanic mass redistribution for this purpose. The main objective of this work is to estimate the uncertainty associated with each category of models at large spatial scales over 1,200 km and inter-annual time scales to compare with the expected gravitational signature of some core processes. This estimation can not be done for the earthquakes and for the cryosphere because the existing models are not independent from GRACE observations (Adhikari et al., 2016; Deggim et al., 2021).

To our knowledge, there was no published study evaluating gravity field products and models at these scales. A first paper in this direction has assessed the accuracy of satellite laser ranging (SLR) and hydrological loading products at inter-annual time-scales and for degree-2 as compared with surface deformation from global navigation satellite system (GNSS) (Rosat et al., 2021). They showed that the gravity and surface deformation signatures of inter-annual degree-2 pressure flows at the CMB are much lower than the observed uncertainties.

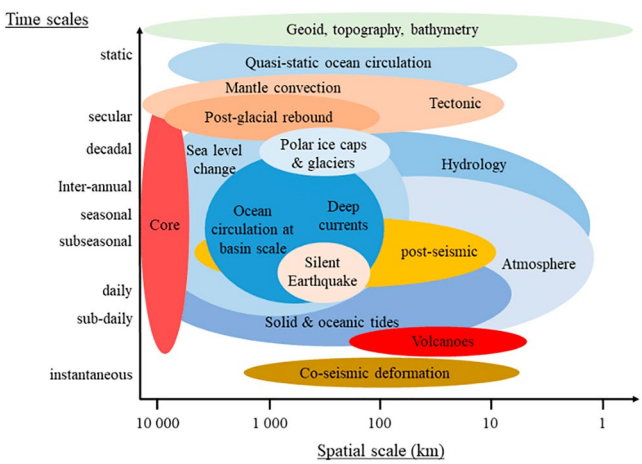


Figure 1. Spatial and temporal scales of the physical processes causing mass variations in the Earth system adapted from Ilk et al. (2004).

Here we focus on the gravitational signature induced by various core processes that are first presented (Section 2). We then present the spherical harmonics (SH) products and geophysical models used to estimate gravity variations (Section 3). A minimum threshold of uncertainty is provided for each category of products and models (Section 4). These uncertainties are finally discussed and compared with expected amplitudes of the presented core processes (Section 5).

2. Expected Gravitational Signals From the Earth's Core

Dynamical core processes disturb the time-varying gravity field through the direct Newtonian effect of mass anomalies in the fluid core. Dynamical core processes also have indirect effects, such as pressure changes at the CMB

induced by varying core flows or changes in the rotation vector of the solid Earth. Dumberry and Mandea (2021) provided a review of the surface deformation and gravity variations induced by core dynamics, as well as a quantification of the expected amplitudes. In this part, we aim to provide a brief summary of these effects and an estimation of the amplitude in equivalent water height (EWH) at the temporal scales observable with GRACE.

2.1. Spherical Harmonics (SH) Representation

In the following, we note $C_{l,m}$ and $S_{l,m}$ the degree-l, order-m fully normalized Stokes coefficients of the SH representation of the Earth's gravitational potential. With $\hat{C}_{l,m}$ and $\hat{S}_{l,m}$ the unnormalized coefficients and $\delta_{m,0}$ the Kronecker delta, the normalization is given by

$$C/S_{l,m} = \sqrt{\frac{(n+m)!}{(2-\delta_{m,0})(2n+1)(n-m)!}}\hat{C}/\hat{S}_{l,m}, \tag{1}$$

where $C/S_{l,m}$ is either $C_{l,m}$ or $S_{l,m}$.

The amplitude of the Stokes coefficient can be represented as EWH. An EWH amplitude, $\Delta\sigma(\lambda, \phi)$ is function of the longitude λ and the latitude ϕ (Wahr et al., 1998):

$$\Delta\sigma(\lambda,\phi) = \frac{R\bar{\rho}}{3\rho_w}\sum_{l=0}^{\infty}\sum_{m=0}^{l}\frac{2l+1}{1+k_l}\left[\Delta C_{l,m}\cos(m\lambda) + \Delta S_{l,m}\sin(m\lambda)\right]\bar{P}_l^m(\cos\phi), \tag{2}$$

where $\bar{P}_l^m(\cos\phi)$ are the associated fully normalized Legendre polynomials (4π normalization). R is the Earth's radius (6.371×10^6 m), $\bar{\rho}$ is the mean density of the Earth (5,515 kg.m^{-3}), ρ_w is the density of water (1,000 kg.m^{-3}) and k_l is the load Love number of degree l.

2.2. Newtonian Effect of Mass Anomalies in the Fluid Core

Core flows create redistribution of density anomalies (Dumberry, 2010b). This first perturbation leads to an adjustment in the internal stress field. A secondary density perturbation is then created because of a global elastic deformation, due to this stress field.

A density perturbation, $\Delta\rho(r, \lambda, \phi)$ is function of the radius r, the longitude λ and the latitude ϕ. There is an expansion in SH for each radius r with the associated coefficients $\rho^c_{l,m}(r)$ and $\rho^s_{l,m}(r)$:

$$\Delta\rho(r,\lambda,\phi) = \sum_{l=0}^{\infty}\sum_{m=0}^{l}\left[\rho^c_{l,m}(r)\cos(m\lambda) + \rho^s_{l,m}(r)\sin(m\lambda)\right]\bar{P}_l^m(\cos\phi) \tag{3}$$

The gravity variation created by this density perturbation can be expressed as a SH coefficient variation of the gravity field, $\Delta C_{l,m}$ or $\Delta S_{l,m}$, by integrating the density perturbation for each radius in the fluid core between the Inner Core Boundary (ICB) and the CMB (Dumberry, 2010b).

$$\Delta C/S_{l,m} = \frac{4\pi}{2l+1}\frac{1}{MR^l}\int_{r_{ICB}}^{r_{CMB}}\rho^{c/s}_{l,m}(r)[1+\kappa_l(r)]r^{l+2}\mathrm{d}r, \tag{4}$$

where M corresponds to the mass of the Earth (5.972×10^{24} kg) and $\kappa_l(r)$ characterize the additional contribution due to global elastic deformation at degree l and radius r. $\kappa_l(r)$ values comes from Dumberry (2010b) and for degree $l > 2$, they fall within the range of approximately 0.2 and −0.2.

To have an order of amplitude of the gravitational effect created by density anomalies, we can take upper bound values for the density variations. The amplitude of density variations within the Earth's core increases with the time scale of the analysis. This is because longer time scales allow for the observation of larger and more gradual changes in the density of the core, such as those caused by large-scale convection patterns (Dehant et al., 2022). At decadal and inter-annual time scales (maximal time-length achievable, yet, with GRACE observations), the upper bound of the density variation is $\Delta\rho = 1 \times 10^{-5}$ kg.m^{-3} (Dumberry & Mandea, 2021). For an annual period, this amplitude is smaller by one order of magnitude.

Supposing as an upper bound a variation with an amplitude of $\Delta\rho = 1 \times 10^{-5}$ kg.m^{-3} at each radius of the fluid core, we compute the effect for degree $l = 2$, 6 and 10. At inter-annual and decadal time scales, this gives

respective Stokes coefficient variations of 2×10^{-11}, 1×10^{-13}, and 4×10^{-15}. This values can be estimated in cm EWH and for degree 2, 6 and 10, we respectively obtain as upper-bound values 0.1, 0.006 and 0.0005 cm EWH, over a decadal period.

2.3. Pressure Flows Effect

Besides the direct Newtonian effect, core flows create a tension on the CMB. This tension results in elastic deformations of the boundary and therefore, density perturbation (Dumberry, 2010b; Dumberry & Bloxham, 2004a). The same process occurs at the ICB.

In the same way as for density perturbation, we can describe the pressure anomalies $\Delta p(\lambda, \phi)$ with an expansion in SH at the CMB with the associated coefficients $p^c_{l,m}$ and $p^s_{l,m}$:

$$\Delta p(\lambda, \phi) = \sum_{l=0}^{\infty} \sum_{m=0}^{l} \left[p^c_{l,m} \cos(m\lambda) + p^s_{l,m} \sin(m\lambda) \right] \bar{P}^m_l(\cos \phi) \tag{5}$$

The gravity variations created by these pressure anomalies can be expressed as a SH coefficient variation of the gravity field $\Delta C_{l,m}$ or $\Delta S_{l,m}$ (Dumberry, 2010b):

$$\Delta C/S_{l,m} = \bar{k}_l \frac{R}{GM\bar{\rho}} p^{c/s}_{l,m}, \tag{6}$$

where G is the gravitational constant (6.674×10^{-11} $m^3.kg^{-1}.s^{-2}$) and $\bar{k}_l$ are potential Love numbers corresponding to degree l. For degree 2, 6 and 10, $\bar{k}_l$ values are respectively 1.116×10^{-1}, 1.957×10^{-3} and 9.856×10^{-5} (Dumberry & Mandea, 2021).

To have an order of amplitude of the gravitational effect created by pressure anomalies, we can use typical pressure variations. As for the density, the pressure amplitude is dependent on the period. As the time scale of the analysis increases, the amplitude of the pressure variations also increases (Gillet et al., 2020). At decadal and inter-annual time scales, the typical pressure variations at the CMB should be $\Delta p = 100$ Pa (Dumberry & Mandea, 2021). For annual period, this amplitude is one order of magnitude smaller.

Supposing as an upper bound a variation with an amplitude of $\Delta p = 100$ Pa at the CMB, we compute the effect for degree $l = 2$, 6 and 10. At inter-annual and decadal time scales, this gives Stokes coefficient variations of 3×10^{-11}, 6×10^{-13}, and 3×10^{-14} and corresponding EWH of 0.5, 0.04, and 0.004 cm EWH, over a decadal period.

2.4. Specific Effects on Degree 2 of the Gravity Field

We have previously discussed two mechanisms responsible for generating mass variations at different length scales. However, there are processes like alteration of the rotation vector and inner core reorientation that also lead to degree 2 variations.

2.4.1. Rotation Effects of the Core

Core dynamics can cause variations in the gravitational field through the alteration of the rotation vector of the solid Earth. For example, the exchange of angular momentum between the core and mantle produces changes in the angular velocity of the Earth, also express as Length of Day (LOD) variations. Pressure flows are responsible for decadal LOD variations (Jault & Finlay, 2015). Because Earth's angular momentum must be conserved, a change in the Earth's oblateness $\left(J_2 = -\sqrt{5} C_{2,0} \right)$ is associated with a change in rotation. A 50 Pa change in $p_{2,0}$ at decadal periods result in $J_2 \approx 8 \times 10^{-12}$ (Gillet et al., 2020). This corresponds to $C_{2,0} \approx 4 \times 10^{-12}$ and 0.06 cm EWH.

A similar computation for the inner core rotation creates a variation of $C_{2,0}$ term that is five orders of magnitude lower (Dumberry & Bloxham, 2004b). It can then be ignored.

Table 1
Decadal Amplitude of Mass Variations Due To Core Processes at Different Degree in cm Equivalent Water Height

	EWH (cm)		
Gravitational effect	Degree 2	Degree 6	Degree 10
Amplitude observed by GRACE	5	20	15
Density anomalies	0.1	0.006	0.0005
CMB pressure anomalies	0.5	0.04	0.004
Inner core rotation	0.2	X	X

2.4.2. Inner Core Reorientation

The inner core is supposed to have a topography at degree 2 and order 2, $h_{2,2}$. When the inner core is tilted by an angle α, it creates a variation on the coefficient $S_{2,2}$. This variation can be approximated by:

$$\Delta S_{2,2} \approx 10^{-10} h_{2,2} \alpha \tag{7}$$

under the hypothesis of a non-convecting inner core and with a density almost uniform at hydrostatic equilibrium (Dumberry, 2010a).

Dumberry and Mandea (2021) estimated the amplitude of the inner core reorientation supposing $\alpha = 0.4^o$ and $h_{2,2} = 18$ m on decadal time period. It gives $\Delta S_{2,2} = 10^{-11}$ and 0.2 cm EWH.

2.5. Summary of the Gravitational Signals From the Earth's Core

Table 1 presents the amplitude of mass variations due to various core processes at different degree in EWH. The amplitude observed by GRACE is at least one order of magnitude larger than the predicted effects. Density anomalies have the lowest amplitude at degree 2 (0.1 cm EWH) and strongly decrease as the degree increases. These results suggest that mass variations due to core processes are most prominent at small degrees, and strongly decrease at higher degrees.

This observation is consistent with Rosat et al. (2021), which reports that at spherical harmonic degree 2, the contribution of core processes to gravity variations and ground deformations is approximately 10 times smaller than the observed fluctuations caused by dynamical processes within the fluid layers at the Earth' surface.

This section points out that the study of the Earth's core trough gravity field variations can, yet, only be done at large spatial scales and inter-annual/decadal time scales. Consequently, identifying signals of core origin poses a significant challenge and requires the accurate removal of all surface effects that are larger (Figure 2).

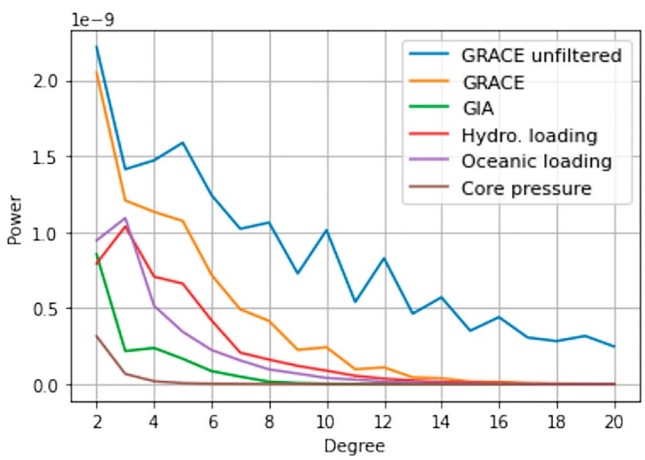


Figure 2. Power of spherical harmonics degree with a Gaussian spatial filtering of radius 1,200 km up to degree 20 (detailed in Section 3.1). Showed for GRACE (COST-G) unfiltered and with filtering, a glacial isostatic adjustment model (ICE-6G_D), a hydrological loading model (ISBA-CTRIP), an oceanic loading model (MPIOM) and synthetic effect of the core pressure with $\Delta p = 100$ Pa. All the solutions are presented in this Section 3.

3. Data Presentation

Solutions for the time-variable gravity field are obtained using GRACE measurements with SLR measurements for low degrees. Geophysical models representing hydrological, oceanic and glacial isostatic adjustment (GIA) processes are obtained from independent models and not from GRACE inputs. These surface signals have more power in SH representation than the theoretical signal from Earth's core pressure anomalies (Figure 2) and other Earth's interior signals after spatial and temporal filtering as detailed in the following.

3.1. Mathematical Approach

Models and solutions are provided in either SH or grid representation (Swenson & Wahr, 2002). Since we are interested in large spatial scales, we primarily use SH processing and representation. We only use the grid format to represent our results in a geographically interpretable way.

To study hypothetical gravity variations originating from the Earth's core, we filter the products and models considered in this study at appropriate spatial and temporal scales (Section 2.5). The spatial filtering is done with a Gaussian filter (Jekeli, 1981) of radius 1,200 km to access large spatial scales and avoid Gibbs aliasing. We do not use the usual isotropic spatial filter (Kusche, 2007) that allows to recover high resolution signals. Post-filtered SH are increasingly reduced to degree 12 because of the Gaussian filter

Table 2
Characteristics of the GRACE Gravity-Field Models

Model	Mean gravity field model	Ocean tides	Atmospheric mass variations	Oceanic non-tidal mass variations	Data sources	References
CSR RL06	GGM05C	GOT4.8	AOD1B RL06 GAA	AOD1B RL06 GAB	https://podaac-tools.jpl.nasa.gov/drive/	CSR RL6.0 (2018)
GFZ RL06	GGM05C	FES2014b	AOD1B RL06 GAA	AOD1B RL06 GAB	https://podaac-tools.jpl.nasa.gov/drive/	Dahle et al. (2019)
JPL RL06	EIGEN-6C4	FES2014	AOD1B RL06 GAA	AOD1B RL06 GAB	https://podaac-tools.jpl.nasa.gov/drive/	JPL RL6.0 (2018)
ITSG-Grace2018	ITSG-GraceGoce2017	FES2014b + GRACE estimates	AOD1B RL06 GAA and LSDM for sub-monthly hydrology de-aliasing	AOD1B RL06 GAB	https://icgem.gfz-potsdam.de/	Mayer-Gürr et al. (2018)
CNES RL05	EIGEN-GRGS.RL04.MEAN-FIELD	FES2014b	3-D ECMWF ERA-Interim + AOD1B RL06 GAA	TUGO + AOD1B RL06 GAB	https://grace.obs-mip.fr/	Lemoine et al. (2019)
COST-G RL01	X	X	X	X	https://icgem.gfz-potsdam.de/	Meyer et al. (2020)

(Figure 2). The temporal filtering is done with a Butterworth low-pass filter with a cut-off period at 2 years. This removes high-amplitude signals with annual and semi-annual periods in the products and models.

3.2. GRACE

GRACE gravity-field SH solutions are distributed by several analysis centers, providing GRACE Satellite-only Model (GSM) coefficients of the geopotential (Bettadpur, 2018). In this study, we considered six GSM solutions (see Section 2 for details) from the three Science Data System centers (Center for Space Research (CSR) (CSR RL6.0, 2018), German Research Centre for Geosciences (GFZ) (Dahle et al., 2019), and Jet Propulsion Laboratory (JPL) (JPL RL6.0, 2018)) and three non-official centers (International Combination Service for Time-variable Gravity Fields (COST-G) (Meyer et al., 2020), Institute of Geodesy at Graz University of Technology (IFG-TU GRAZ) (Mayer-Gürr et al., 2018) and Centre national d'études spatiales (CNES) (Lemoine et al., 2019)). GRAZ and CNES centers propose different approaches: sub-monthly hydrological de-aliasing for GRAZ, addition of SLR inputs for low degree determination for CNES. COST-G is a combination of the solutions from the other 5 centers used in this paper with the addition of Astronomical Institute University Bern solution. Detailed information about considered solutions are given in Table 2.

The 6 GRACE solutions considered in this study have a quasi-monthly time resolution. Time series span from the start of the GRACE mission, April 2002, to April 2021. There is a gap of one year between mid-2017 and mid-2018 between the GRACE and the GRACE-FO missions. As we are interested in the low degrees of the gravity field variations, we use only SH models and not MASCON products. SH solutions are global whereas MASCON products are designed to access higher spatial resolution with pre-established grid that are an a priori of the mass distribution (Scanlon et al., 2016). Others institutes propose GRACE solutions, but they are not considered here.

The $C_{2,0}$ estimation with GRACE data is affected by a disturbing 161-day periodic signal (Chen et al., 2005; Cheng & Ries, 2017) without a consensual explanation for this issue. It has then become a standard to replace the GRACE determination of $C_{2,0}$ by the SLR one. We use the Technical notes TN14 solution based on SLR data and recommended in Loomis et al. (2019a). The GRACE $C_{3,0}$ is also poorly observed when the satellites pair is operating without two fully functional accelerometers (Loomis et al., 2020). The TN14 solution also provides a $C_{3,0}$ estimation that we include after October 2016 (GRACE month > 178). These two problematic estimations are suspected to also affect other coefficients such as $C_{4,0}$, $C_{5,0}$, and $C_{6,0}$ (Cheng & Ries, 2017; Sośnica et al., 2015; Loomis et al., 2020). However, the quality of these GRACE coefficients is comparable with the quality of the SLR coefficient estimation (Cheng & Ries, 2017; Velicogna et al., 2020). It seems then not relevant to replace these coefficients. Dahle et al. (2019) suggested to have a special attention to $C_{2,1}$ and $S_{2,1}$ coefficients that contain an anomaly correlated with a failure of the accelerometers. We choose to replace these two coefficients with the SLR solution from Cheng et al. (2011) after October 2016. These replacements are not included in the CNES solution because it already includes SLR data at low degrees. Geocenter coefficients $C_{1,0}$, $C_{1,1}$, and $S_{1,1}$ are not included in our data and are set to 0 for the CNES solution where they come from SLR.

Previous studies provided estimates of the uncertainty of GRACE products from different centers, but not at large spatial and inter-annual time scales. For example, Kvas et al. (2019) compared the GRAZ solution with those from CSR, GFZ and JPL in terms of temporal Root Mean Square (RMS) over a grid, quiet RMS time series and 161-day signal. Wang et al. (2021) and Dobslaw et al. (2020) compared the estimations of global mean ocean mass and mean barystatic sea level with solutions from different centers. Blazquez et al. (2018) compared the trends of the global water budget components from 5 GRACE centers. It also estimated the uncertainties associated with the processing parameters, namely, the

Table 3
Characteristics of the Hydrological Models

Acronym	Precipitation model	Sampling period	Space resolution
ERA5	Simultaneously generate	1 hr	0.25°
GLDAS	GPCP	3 hr	0.25°
ISBA	GPCC	3 hr	1°
WGHM	GPCC	Monthly average	0.5°
LSDM	ECMWF	Daily	1°

geocenter motions, $C_{2,0}$, filtering, leakage and GIA. Another estimation of the GRACE products uncertainty can be given by the RMS value over ocean but it has not been proposed, yet, for inter-annual time scales (Chen et al., 2021). It is also worth noting that MASCON products can be useful in error assessment (Loomis, Luthcke, & Sabaka, 2019). In the following, we compare GIA, hydrology and non-tidal oceanic models.

3.3. Glacial Isostatic Adjustment (GIA)

The GIA signal induces linear trends in the gravity field variations. Effects of the post-glacial rebound are apparent in Antarctica, Northern America and Scandinavia. This signal rectification uses GIA models based on global ice-loading history and mantle viscosity. We do not consider regional GIA models since they would give spurious estimates of the GIA signal out of the specific regions for which they have been designed (Whitehouse et al., 2012). Present-day ice melting is not taken into account in the post-glacial rebound models, it hence constitutes another source of uncertainty.

We compare three different global GIA models, namely, A13 (Geruo et al., 2013), ICE-6G_D (VM5a) (Peltier et al., 2015, 2018), and Caron18 (Caron et al., 2018).

A13 is based on the ICE5G ice-loading history model (Peltier, 2004) and on the multilayered viscosity profile VM2 (Peltier, 2004). A13 is computed via a 3-D finite element method that creates a 3-D viscosity structure. ICE-6G_D (VM5a) uses an update of ICE5G ice-load history with the addition of GNSS vertical rates constraints and Antarctica ice height change data (Argus et al., 2014). ICE-6G_D (VM5a) includes a more recent viscosity profile VM5a. Caron18 represents the mean of an ensemble of 128,000 forward models calculated in a Bayesian framework. For each run model, the viscosity structure and the scaling coefficients for the ice-load history of the Australian National University model (Lambeck et al., 2010, 2014) vary. As a result, some of the models included may not fit the uplift rate measured by GNSS. The final Caron18 GIA is a weighting of each model inferred by the probabilistic information and contains an estimate of the uncertainty from the dispersion between the models.

Comparisons between these three GIA models already exist, mainly with regard to the uplift rates as measured by GNSS and the viscosity profiles. Argus et al. (2014) and Peltier et al. (2015) compare ICE-6G_D with A13 respectively on Antarctica and North America. Caron et al. (2018) and Argus et al. (2021) compare ICE-6G_D with Caron18 on North America. It is worth noting that the closest model to the measured GNSS uplift rate is ICE-6G_D.

Global GIA models are not associated with any uncertainty except for Caron18 and studies rarely discuss that point (Caron et al., 2018; Melini & Spada, 2019; Peltier et al., 2022). Li et al. (2020) also discussed the uncertainty associated of ICE-6G_D (VM5a) in North America. A way of estimating the impact of the uncertainty of those models is by comparing some of them for a specific application. Śliwińska et al. (2021) used two different GIA models to estimate polar motion while Blazquez et al. (2018) compared three GIA models for the determination of global ocean mass change and sea level budget. In the case of regional applications, Kappelsberger et al. (2021) compared three global and two regional models with the uplift estimation from GNSS on the northeast of Greenland. However, to the best of our knowledge, there is no comparative study of GIA models based on the SH approach that was published, and more specifically, on low SH degrees.

3.4. Hydrology

We compare five global hydrological models, namely the Global Land Data Assimilation System Noah 2.1 (GLDAS) (Rodell et al., 2004), ERA5 (Hersbach et al., 2020), WaterGAP Global Hydrology Model version 2.2d (WGHM) (Döll et al., 2003), Interaction Sol-Biosphère-Atmosphère CNRM version of TRIP (ISBA-CTRIP, further referred to as ISBA) (Decharme et al., 2019) and Hydrological Land Surface Discharge Model (LSDM) (Dill, 2008). Hydrological models contain mainly annual and semi-annual signals. With the temporal and spatial filtering to access the core-like scales, the residuals studied are small compared to the original signals. For example, the RMS value of ISBA over continent is 3.64 cm in EWH and 1.47 cm EWH after temporal filtering. These residuals contain climatic modes like El Niño-Southern Oscillation.

Table 4
Characteristics of the Ocean Models

Acronym	Sampling period	Spatial resolution	Inverted barometer
OMCT	90 min	1°	No
MPIOM	20 min	1°	Yes
T-UGOm	3 hr	Unstructured grid	No

The five hydrological models considered (see Table 3) solve the vertical water mass balance but only three of them also solve the lateral fluxes. The water mass balance is expressed as the TWS anomaly.

For GLDAS, the permanently ice-covered areas have been masked out. GLDAS has a spatial resolution of 0.25° per 0.25° and a temporal resolution of 3 hr. ERA5 has the same temporal and spatial resolutions. ERA5 is the new global model from Copernicus Climate Change Service that replaces the ERA-Interim reanalysis (Dee et al., 2011). GLDAS uses Global Precipitation Climatology Project (GPCP) V1.3 Daily Analysis (Adler et al., 2003) as precipitation model. GPCP is a family of precipitation models based on in situ raingauge data to estimate monthly precipitation. For GLDAS and ERA5, gravitational potential changes induced by hydrological mass redistribution and loading (Newtonian attraction and mass redistribution associated with elastic deformation) are computed as detailed in Petrov and Boy (2004) and Gégout et al. (2010).

WGHM, ISBA, and LSDM are also supplemented with lateral fluxes solving. We use the variant IRR100 of WGHM forced with Global Precipitation Climatology Centre (GPCC) monthly V7.0 precipitation (Schneider et al., 2016). The output of the WGHM that we use in this study was already at a monthly averaged temporal scale and the spatial resolution is 0.5°. ISBA-CTRIP is the combination of a water balance model (ISBA) with a runoff model (CTRIP). ISBA has a temporal resolution of 3 hr and a spatial resolution of 1° and it also uses GPCC V6 as a precipitation model. LSDM has a daily temporal frequency and a spatial resolution of 1°. LSDM has been designed for large spatial scale geodetic applications such as the study of Earth's polar motion (Dill et al., 2010; Jin et al., 2012). Among the three models, only WGHM includes human-induced effects of freshwater resources. This contribution is extremely important when accounting for the contribution of freshwater fluxes to the global ocean (Schmied et al., 2020).

Each models has been resampled to a monthly time scale with an average over the month. The time coverage of comparison goes from 2002 to the end of 2016, this corresponds to the end of the WGHM model provided to us.

Previous studies compared hydrological models with GRACE gravity field variations but not with this diversity of models and not at these inter-annual and large spatial scales (Jin & Feng, 2013; Lenczuk et al., 2020; Liu et al., 2019). At inter-annual and decadal scales, hydrological models compared with GRACE solution are underestimating the hydrological signal on river basins and regarding climate modes (Pfeffer, Cazenave, & Barnoud, 2022; Pfeffer, Cazenave, Blazquez, et al., 2022; Scanlon et al., 2018).

3.5. Non-Tidal Oceanic

We compare three oceanic loading models, namely Ocean Model for Circulation and Tides (OMCT) (Dobslaw et al., 2013), Max-Planck-Institute for Meteorology Ocean Model (MPIOM) (Jungclaus et al., 2013) and Toulouse Unstructured Grid Ocean model (T-UGOm) (Carrere & Lyard, 2003). These models are used in GRACE solutions to correct for oceanic loading effects. For official centers, these models correspond to the GAB solution that contains the contribution of the dynamic ocean to ocean bottom pressure. OMCT has been used by official GRACE centers between Releases 1 and 5. MPIOM is used for the Release 6. T-UGOm is used by the CNES for the correction of the GRACE data (and not for GRACE-FO).

OMCT and MPIOM are baroclinic ocean models with a spatial resolution of 1^{o}. They are adjustments from another model, the climatological Hamburg Ocean Primitive Equation model. They are forced by external information from the operational analyses of the European Centre for Medium-Range Weather Forecast (ECMWF). They compute water elevations, three-dimensional horizontal velocities, potential temperature and salinity. Both MPIOM and OMCT are forced by surface winds, pressure, atmospheric freshwater fluxes and surface temperature. MPIOM is using river runoff, sea-ice and corrects for the inverted barometer response of the oceans as opposed to OMCT. The T-UGOm barotropic ocean model is based on an unstructured grid with a higher resolution on coastal area. It does not represent variations of temperature and salinity but only displacement of the barotropic fluid. T-UGOm is using wind and atmospheric pressure forcing from ERA-interim and does not correct the inverted barometer response. Temporal and spatial resolutions of each model are detailed in Table 4.

To compare these three models we cannot use the GAB solutions from GRACE releases because of the difference in the correction of the inverted barometer effect. The GAB solution for AOD1B RL06 with MPIOM uses the correction of the inverted barometer effect. It implies that the AOD1B RL06 GAA solution, which corresponds to

Table 5
Root Mean Square (RMS) Differences in cm Equivalent Water Height Between Different GRACE Solutions and RMS Value of Each Model After Spatial and Temporal Filtering

	JPL	GFZ	GRAZ	CNES	COST-G	RMS
CSR	0.17	0.35	0.32	0.45	0.16	1.82
JPL		0.32	0.31	0.45	0.16	1.81
GFZ			0.45	0.53	0.30	1.82
GRAZ				0.45	0.27	1.87
CNES					0.42	1.86
COST-G						1.81

Note. Small values are highlighted in yellow and large values in blue.

the atmospheric loading effect, is equal to a constant value over oceanic area. For OMCT and T-UGOm, the GAB solution contains the inverted barometer effect and the GAA solution does not contain the inverted barometer effect. Regarding this, we compare the GAC solutions which are in fact the sum of the GAB (ocean loading) and the GAA (atmospheric loading) solutions over the ocean. This sum over oceanic areas corresponds to the oceanic bottom pressure and is given by the GAD solution in GRACE releases. To compare these oceanic loading models, the best way is to use the related GAD solutions.

Previous studies compared these models but at sub-monthly time scales (Bonin & Save, 2019; Dobslaw et al., 2016). To our knowledge, there are no comparative studies of ocean loading models on inter-annual and decadal temporal scales. Schindelegger et al. (2021) also compared some other oceanic models with MPIOM at sub-monthly time scales. We did not include these other models because some are in-house products and other are GRACE-dependent.

4. Comparison of Gravity Field Solutions and Models

In our approach, we cannot directly estimate the accuracy of solutions and models. We use an ensemble approach where the dispersion between solutions and models provides an estimate of the uncertainty. This estimate is a first lower bound that does not take into account any bias. This approach is similar to Blazquez et al. (2018) or Marti et al. (2022).

Comparisons between solutions and models are quantified as the RMS difference between both objects weighted by latitude. In order to compute the weighted RMS, solutions and models are projected on a grid of 0.5° × 0.5° degree and we compute the difference between the grids.

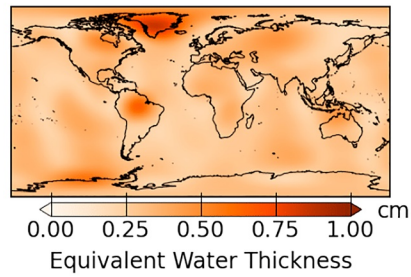
All the presented comparisons between GRACE solutions and between geophysical models in the Section 4.1 are provided after spatial and temporal filtering. This filtering allows providing a lower bound of the uncertainty between products of the same family at core-like scales.

4.1. Differences Between GRACE Solutions

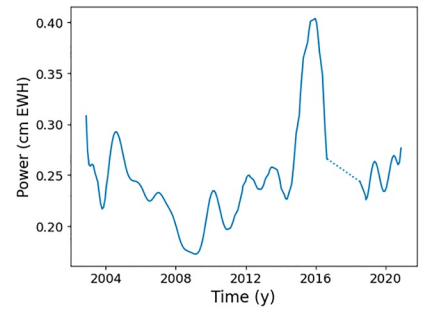
4.1.1. GRACE Analysis Centers

Comparison between GRACE solutions requires to minimize side effects due to the temporal filtering. We hence remove the first and last 3 months of the solutions.

Table 5 contains the RMS differences in cm EWH between the spatially and temporally filtered GRACE solutions from the different analysis centers (compared in this study). For reference, the RMS value of the CSR solution is 1.82 cm EWH. The first group, CSR, JPL, and COST-G solutions, is the most similar with an RMS difference of 0.16–0.17 cm EWH or 9% of the original RMS value for one solution. There is an increase of the difference to 0.22 cm EWH in 2016 at the end of GRACE lifespan corresponding to the accelerometer failure of one of the two satellites. Then comes a second group with GFZ and GRAZ which have an RMS difference of 0.3 cm EWH with the first group or 17% of the original RMS value for one solution. But the difference of these two solutions with the first group is different according to the considered epoch. GFZ has a peak going up to 0.7 cm EWH at the end of the GRACE lifespan. For GRAZ, in this temporal period, the difference goes up to 0.5 cm EWH but then it goes to 0.7 cm EWH at the end of the GRACE-FO time series. For the GFZ, the spatial distribution of differences corresponds to a global noise without any specific pattern. But for the GRAZ solutions, differences are located in areas of large signals, in the Amazon basin and Greenland. The CNES solution has a RMS difference of 0.45 cm EWH (25% of the original RMS value) with respect to other solutions with a temporal difference of 1 cm EWH at the beginning of the GRACE mission and at the end of the GRACE life span. The spatial localization of these differences are located in areas of strong hydrological signal like the Amazon basin and India. Figures to illustrate these analyses are available in Appendix A in Supporting Information S1. To continue the analysis of the differences between the GRACE solutions, it is important to consider the average of the RMS differences between all the solutions over time and in different spatial areas (Figure 3). The highest values over Greenland, Antarctica and Amazonia correspond



(a) Average of RMS differences in cm EWH spatially represented



(b) Average of RMS differences in cm EWH represented trough time, the year between GRACE and GRACE-FO is represented by the dotted line

Figure 3. Average of Root Mean Square differences between all the solutions (compared in this study) in cm equivalent water height after spatial and temporal filtering. Small values are highlighted in yellow and large values in blue.

spatially to areas with strong inter-annual signals. Thus, the stronger the signal, the larger the differences between the solutions. For the temporal variations of the RMS differences between solutions, the difference are nearly twice as large at the end of the GRACE mission (before the dotted line that represents the gap). The degradation of the quality of GRACE solutions is well known and has already been documented (Dahle et al., 2019; Kvas et al., 2019). This degradation is due to the failure of the accelerometer after November 2016 and is smoothed trough time in Figure 3b because of the temporal filtering. Otherwise, the RMS values over time are about 0.25 cm in EWH.

4.1.2. GIA Models

Figure 4 represents the difference in rate of EWH in mm per year between the models with a spatial resolution of 2,400 km after a truncation at degree 60 and the application of a Gaussian filter. In Appendix B in Supporting Information S1, the same figure without spatial filtering is available. The models are similar in Scandinavia. The Caron18 model differs from the others in North America and the A13 model differs from the others in Antarctica. These two statements correspond to previous observations (Argus et al., 2014, 2021). There are small differences between A13 and the ICE-6G_D model in North America (±1 mm/y in EWH) compared to those in Antarctica (±3 mm/y). Peltier et al. (2015) pointed out a larger difference on the western and eastern sides of Hudson Bay in Canada that we recovered without the spatial filtering (Appendix B in Supporting Information S1). However,

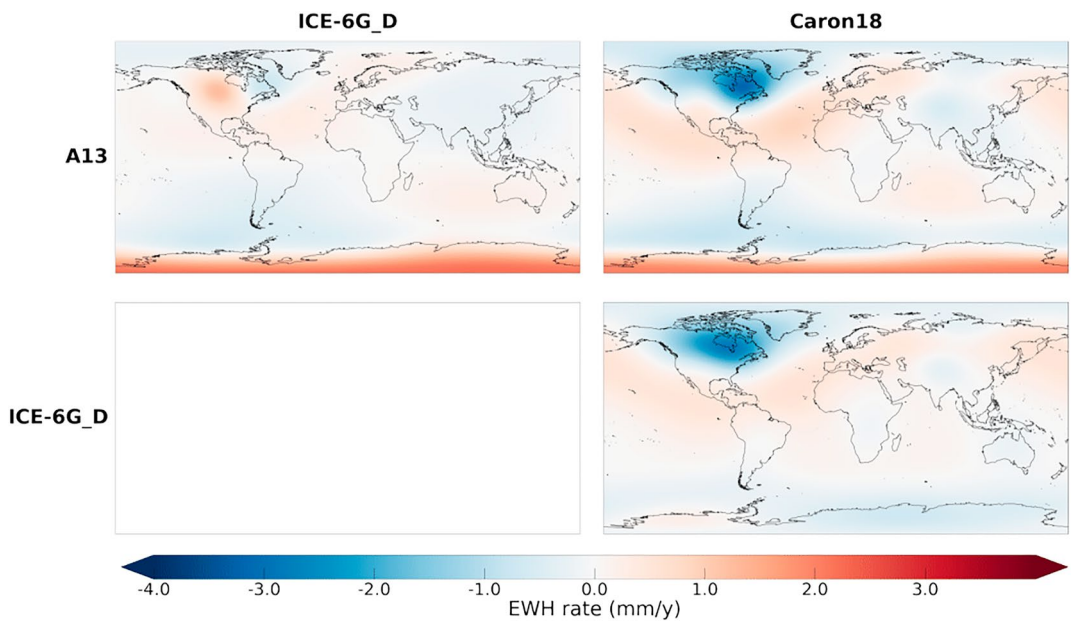


Figure 4. Difference between glacial isostatic adjustment models spatially filtered in equivalent water height rate (mm/y).

Table 6
Root Mean Square (RMS) Difference in cm Equivalent Water Height Between Hydrological Models and RMS Value of Each Model After Spatial and Temporal Filtering Over the Continents

	GLDAS	ISBA	WGHM	LSDM	RMS
ERA5	0.89	0.89	1.36	1.50	0.91
GLDAS		0.89	1.20	1.74	1.26
ISBA			1.13	1.56	1.00
WGHM				2.10	1.36
LSDM					1.66

Note. Small values are highlighted in yellow and large values in blue.

in Figure 4, the spatial filtering reduces these differences, one being negative and the other positive, they counterbalance each other.

In North America, the disagreement between models goes up to 6 mm in EWH per year. In Antarctica, the differences between models are up to 10 mm in EWH per year. These differences in velocity are currently accumulated over 20 years and at the time of publication of this article, they lead to a potential error of 12 cm in EWH per year over North America and of 20 cm in EWH per year over Antarctica.

4.1.3. Hydrological Models

Table 6 contains the RMS differences in cm EWH between spatially and temporally filtered hydrological loading models over continents without Greenland and Antarctica. The RMS difference ranges from 0.89 to 2.10 cm EWH or 100%–155% of the original RMS value for one model. For example, the RMS values of ISBA and LSDM are respectively 1.00 and 1.66 cm EWH.

Because hydrological models take into account different processes, they yield very different TWS anomalies, leading to large differences in the predicted gravity variations at large spatial and temporal scales. At inter-annual and large spatial scales, ERA5, GLDAS, and ISBA display relatively similar signals (Figure 5a). Probably because it takes into account anthropogenic use of freshwater, WGHM exhibits larger differences, with larger TWS changes at inter-annual signals located in India and in the northern hemisphere than the other models (Figure 5c).

LSDM shows the largest difference with other models. It has a very strong signal over the Nile area in North Africa (Figure 5b). The difference between LSDM and other hydrological models like GLDAS has been documented and explained by the particular river channels redistribution of water (Dill & Dobslaw, 2013; Dill et al., 2018).

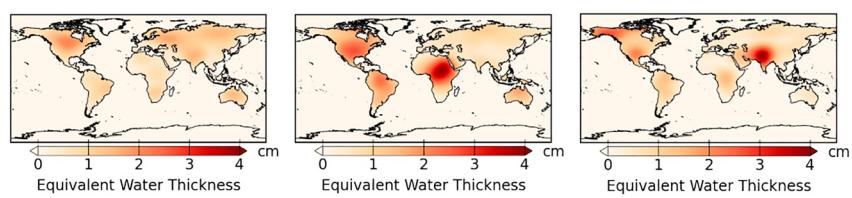
The same analysis has been done on hydrological loading model without spatial filtering in Appendix C in Supporting Information S1.

The quality of hydrological loading models is uneven. To evaluate this quality we look at the percentage of RMS explained by the models in the variation of the gravity field. We compare, over the continents, the RMS of the GRACE time series (COST-G) with the RMS of GRACE minus a hydrological model. The variation of the RMS value gives the percentage of RMS explained by the model in the GRACE time series (Table 7) over non-glaciated continents (Greenland and Antarctica are not included).

At inter-annual and large spatial scales, ISBA and WGHM reduce the variance of GRACE solutions by more than 20%. According to this criteria they have the best quality among the five models considered. ERA5 and GLDAS are close to 0% and LSDM is negative with −16%. It does not modelize gravity field variations in GRACE time-series and contains other signals. Global hydrological models struggle to explain GRACE data, likely due to inaccurate meteorological forcing, unresolved groundwater processes, anthropogenic influences, changing vegetation cover, limited calibration and validation datasets (Pfeffer, Cazenave, Blazquez, et al., 2022).

4.1.4. Non-Tidal Oceanic Models

Table 8 contains the RMS differences in cm EWH between spatially and temporally filtered oceanic and atmospheric loading products over the oceans. The RMS difference ranges from 0.33 to 0.45 cm EWH between models



(a) ERA5 minus GLDAS (b) ISBA minus LSDM (c) ISBA minus WGHM

Figure 5. Maps of Root Mean Square difference between hydrological models over the continents after spatial and temporal filtering.

Table 7
Percentage of Root Mean Square Explained by Hydrological Models in the GRACE Time Series at Inter-Annual Scales With a Spatial Filtering Over Non-Glaciated Continents

	ERA5	GLDAS	ISBA	WGHM	LSDM
Percentage (%)	7	0	24	21	−16

or 79%–107% of the original RMS value for one model. For comparison, the RMS value for OMCT is 0.42 cm EWH. Because oceanic loading models come from different climate and fluid mechanics models, they have a very different spatial and temporal content, leading to large differences. Differences are mostly located in Arctic and Antarctic areas, coastal regions and in the Antarctic Circumpolar Current area (Figure 6). OMCT has more signal in the Arctic while MPIOM and T-UGOm have more signal near Antarctica in the Ross Sea (Figure 6). There is another difference between these models: they are monthly products with potential missing days each month. These missing days correspond to low quality data but may vary between models and releases. This is the case for months at the beginning and at the end of the GRACE mission in 2002 and between 2012 and 2017. For example, for the month of August 2016, the MPIOM products from official centers contain measurements from days of year 221–247 while the T-UGOm products from the CNES contain measurements from days of year 214–244.

The same analysis has been done for oceanic loading models without spatial filtering (Appendix D in Supporting Information S1).

4.2. Impact of Geophysical Corrections on Stokes Coefficients

We have quantified the uncertainties of GRACE solutions and correction models in terms of RMS of the differences over grids. Another interesting approach is to look at SH coefficients. Core processes signal might be present from degree 2 onward to higher degrees with decreasing amplitudes.

To estimate the impact of an error in a model on specific SH coefficients, we have performed some synthetic tests. In the following, we choose synthetic signals with regard to the observed errors in the GIA and hydrological loading models. An artificial synthetic signal is added to the GRACE gravity data on a bounded area. We study the effects of this synthetic signal on the retrieved Stokes coefficients in terms of RMS value. To compare with the time-variable gravity measured by GRACE, we normalized each SH coefficients by the standard deviation $\sigma_{l,m}^{GRACE}$ of the degree-l, order-m Stokes coefficient from the COST-G solution. We note $I_{l,m}$ the normalized RMS value of the coefficient of degree l and order m given by:

$$I_{l,m} = \frac{\sqrt{\frac{1}{n}\sum_t \Delta C_{l,m}(t)^2}}{\sigma_{l,m}^{GRACE}} \tag{8}$$

With t the index of the time vector and n the number of samples of the time vector. This representation gives an estimate of the contamination by an error on the correction model with respect to the corrected GRACE signal.

4.2.1. Impact of an Error in the GIA Model

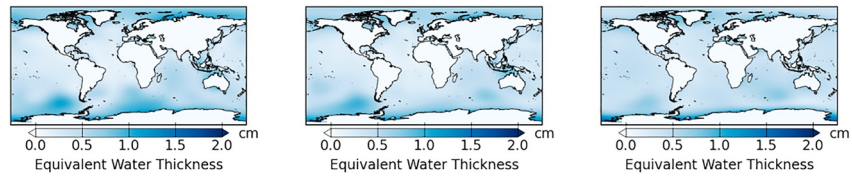
To study the effect of adding a fiducial GIA rectification, we create three synthetic signals corresponding to errors seen in Section 4.1.2.

- A linear signal of 10 mm/y in EWH located in North America with latitude between 50° and 70° and longitude between −95° and −65°.
- A linear signal of 6 mm/y in EWH located in Antarctica with latitude under −80°.
- A linear signal of 3 mm/y in EWH located in Antarctica with latitude under −70° and longitude between −160° and −30°.

Table 8
Root Mean Square (RMS) Difference in cm Equivalent Water Height Between Oceanic Loading Products and RMS Value of Each Model After Spatial and Temporal Filtering Over the Oceans

	MPIOM	T-UGOm	RMS
OMCT	0.33	0.45	0.42
MPIOM		0.42	0.39
T-UGOm			0.44

Introducing a 10 mm/y trend in North America alters the SH coefficients (Figure 7). The error created on the GRACE $S_{4,1}$ coefficient by this fiducial reduction might be up to 30%. The other two synthetics experiments, with a trend at lower latitudes, affect the coefficients of orders 0 and 1 (Appendix E in Supporting Information S1). The largest effect for a trend of 6 mm/y over Antarctica center is on $C_{8,0}$ with a trended bias of 50% of the GRACE RMS value. For a 3 mm/y trend in Antarctica between −160° and −30° in longitude, the effects are smaller with 15% of the GRACE RMS value on $S_{6,1}$ and $S_{8,1}$ (Appendix E in Supporting Information S1).



(a) OMCT (b) MPIOM (c) T-UGOm

Figure 6. Maps of Root Mean Square for oceanic loading products after spatial and temporal filtering over the oceans.

4.2.2. Hydrology

Three cases have been simulated with a sinusoidal signal of period 3 years. They correspond to the difference between hydrological models established in Figure 5 over large hydrological basins.

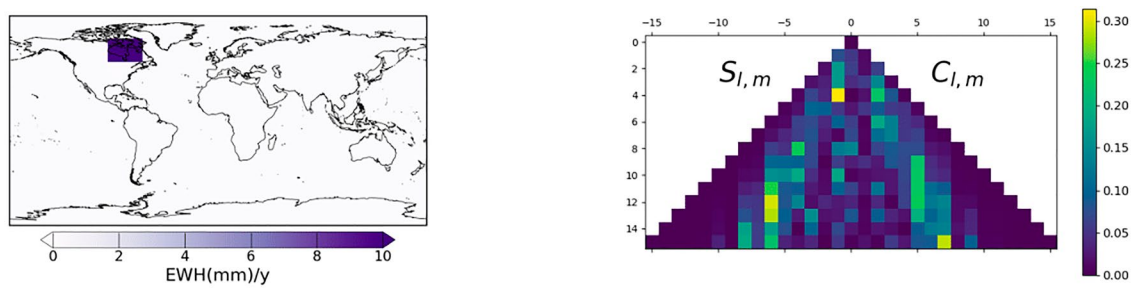
- A sinusoidal signal of 4 cm in EWH over Africa (latitude between −10° and 10°, longitude between 10° and 35°).
- A sinusoidal signal of 3 cm in EWH over Amazonia (latitude between 0° and 20°, longitude between −70° and −40°).
- A sinusoidal signal of 4 cm in EWH over India (latitude between 20° and 30°, longitude between 70° and 90°).

The 3-year period was chosen arbitrarily and represents a residual hydrological signal. A 4-cm sinusoidal signal over Africa affects $C_{5,1}$ and $S_{8,4}$ by an amount of 25% of the GRACE RMS value (Figure 8). A 3 cm sinusoidal signal over Amazonia affects $C_{4,3}$ and $S_{2,2}$ by an amount of 20%, while a 4 cm signal over India affects $C_{8,7}$ and $S_{8,6}$ by an amount of 10% (Appendix F in Supporting Information S1).

5. Discussions and Conclusions

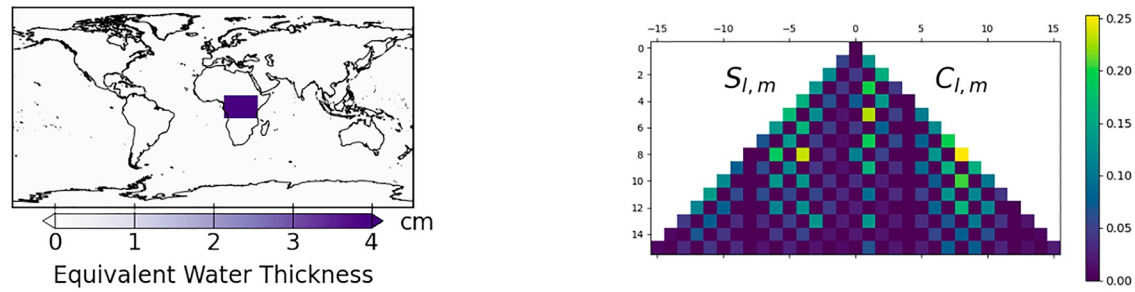
In this paper, we first addressed different core processes that can create gravity variations and estimated their amplitudes. Then, we presented different GRACE SH solutions, GIA, and loading models. We compared each family of products with respect to the differences in RMS or trend at large spatial and inter-annual time scales. From this, we estimated their uncertainties and the associated SH uncertainties.

A summary of the orders of magnitude of predicted core signals and of the dispersion between the different solutions and models obtained in this article is given in Table 9. It contains the amplitude of the RMS difference at degrees 2, 6, and 10 expressed in cm EWH after a spatial and temporal filtering. The largest core signals amplitude with regard to the uncertainty is found at degree 2. At degrees 6 and 10, the amplitude estimated from core signals is an order of magnitude smaller than the estimated uncertainty of the GRACE solutions. To summarize the information on amplitude from this table.



(a) Synthetic signal in North America in EWH (b) SH power normalized by GRACE standard deviation up to degree 15

Figure 7. Effect of a 10 mm/y trend in North America in the glacial isostatic adjustment model (a) on GRACE spherical harmonic coefficients (b).



(a) Synthetic 3-yr signal over Africa with an amplitude of 4 cm EWH

(b) SH power normalized by GRACE standard deviation up to degree 15

Figure 8. Effect of a sinusoidal signal over Africa (a) on GRACE spherical harmonics coefficients (b).

- Mass variations from the core are characterized by their low degree signature and by an inter-annual/decadal time scales. The maximal amplitude of core effects is evaluated at 0.5 cm EWH which is slightly larger than the estimated GRACE uncertainty at inter-annual and large spatial scales.
- GRACE solutions are in good agreement with a dispersion that represents some 10%–20% of the total signal, however, the agreement is not the same over the time span covered by the two missions, with difference mainly at the beginning and end of each.
- For hydrological loading models, the agreement is uneven (see also Figure 5 and Tables 6 and 7). The dispersion between models is as large as the RMS value of models themselves. However, ISBA and WGHM are closer to GRACE solutions.
- For the oceanic loading models, the agreement is generally poor (see also Figure 6). For each model, high-intensity signals are spatially located in different areas at inter-annual time scales. For example, T-UGOm is the only model to report large oceanic mass variations in the south of Africa.
- The GIA effects are not included in this recapitulating table as they are localized in specific areas: North America, Greenland and Antartica. To remind, GIA-mismodeled linear effects are as large as a 20 cm EWH after 20 years over North America. GIA errors will impact the trend but not the inter-annual signals.

When surface processes models are considered to minimize their signature in the gravity data, they might indeed create some spurious signals on some areas. This would also create a spurious signal on specific SH (Figures 7 and 8) up to 50% of the total signal on inter-annual time scales.

The estimated maximum amplitude of core signals is 5 mm in EWH at the Earth's surface. Core mass variations are most significant on decadal time scales and at low degrees. In this context, it is relevant to analyze the Earth's gravity products from GRACE and loading models trough these specific scales. The RMS difference between GRACE solutions of 3.4 mm in EWH shows how difficult is to detect potential core signals. This difficulty is somehow reinforced when considering the uncertainties of loading models used to reduce the gravity signal.

Table 9
Amplitude of Estimated Core Signals Compared to Root Mean Square Difference Between Products Expressed in cm Equivalent Water Height at Inter-Annual and Large Spatial Scales and at Degrees 2, 6 and 10 After Spatial and Temporal Filtering

	(cm EWH)	Filtered difference (cm EWH)		
Type of data	Mean RMS difference	Degree 2	Degree 6	Degree 10
Maximum of the estimated core signals	0.5	0.5	0.04	0.004
GRACE solutions	0.34	0.2[a]	0.1	0.04
Hydrological loading models	1.32	0.37	0.38	0.41
Oceanic loading models	0.40	0.16	0.08	0.03

[a]The RMS difference between GRACE solutions at degree 2 is biased. The $C_{2,0}$ coefficient is replaced by the product (TN14) for each solutions except CNES. $C_{2,1}$ and $S_{2,1}$ coefficients are replaced with an SLR solutions after August 2016.

A careful analysis of the time-variable gravity field data needs to be done to detect and to extract core signals. First, the data-gap between GRACE and GRACE-FO should be filled to ensure continuity and to improve the products quality (Richter et al., 2021). The largest signals in GRACE solutions are due to the Earth's surface processes. The inter-annual variability analysis produced by climate modes (Pfeffer, Cazenave, & Barnoud, 2022) needs also to be considered. In order to detect tiny signals related to the core more sophisticated methods are needed such as empirical orthogonal function analysis (Schmeer et al., 2012) or independent component analysis (Frappart et al., 2011). Recently, Saraswati et al. (2023) applied Singular Value Decomposition, Principal Component Analysis, and Multivariate Singular Spectrum Analysis to separate distinct spatio-temporal patterns in magnetic and gravity field. Moreover, synthetic tests based on modeled processes have to be performed to evaluate the sensitivity of these methods with respect to Earth's core signals.

Acronyms

AIUB	Astronomical Institute University Bern
CMB	Core-Mantle Boundary
CNES	Centre national d'études spatiales
CSR	Center for Space Research
EWH	Equivalent Water Height
GFZ	German Research Centre for Geosciences
GIA	Glacial isostatic adjustment
GLDAS	Global Land Data Assimilation System
GRACE	Gravity Recovery and Climate Experiment
GRACE-FO	Gravity Recovery and Climate Experiment Follow-On
GSM GRACE	Satellite-only Model
IFG TU	Graz Institute of Geodesy at Graz University of Technology
ISBA	Interaction Sol-Biosphère-Atmosphère
ISBA-CTRIP	Interaction Sol-Biosphère-Atmosphère CNRM version of TRIP
JPL	Jet Propulsion Laboratory
MPIOM	Max-Planck-Institute for Meteorology Ocean Model
OMCT	Ocean Model for Circulation and Tides
RMS	Root Mean Square
SH	Spherical harmonics
SLR	Satellite laser ranging
T-UGOm	Toulouse Unstructured Grid Ocean model
TWS	Total water storage
WGHM	WaterGAP Global Hydrology Model

Data Availability Statement

GRACE and GRACE-FO missions are sponsored by the National Aeronautics and Space Administration and the Deutsches Zentrum für Luft-und Raumfahrt. GRACE and GRACE-FO Level-2 temporal solutions were obtained from the PO.DAAC Drive for CSR, GFZ and JPL centers, from https://icgem.gfz-potsdam.de/ for ITSG center, from https://grace.obs-mip.fr/ for CNES center and from https://icgem.gfz-potsdam.de/ for the COST-G combination. The GIA models were obtained from the PO.DAAC Drive (http://grace.jpl.nasa.gov). Time-variable gravity field coefficients due to hydrological loading can be downloaded from EOST loading service (http://loading.u-strasbg.fr/) for ERA5 and GLDAS models. Land Surface Discharge Model model is available on the ESMGFZ Product repository (http://rz-vm115.gfz-potsdam.de:8080) and is produced by IERS Associated Product Centre Deutsches GeoForschungsZentrum GFZ Potsdam. The ISBA-CTRIP model made by the "Centre National de Recherches Météorologiques" (CNRM) of Météo-France has been provided by Bertrand Decharmes. The WGHM model (http://www.watergap.de/) has been provided by Denise Caceres from Frankfurt University. MPIOM and OMCT are available on the PO.DAAC Drive and T-UGOm model has been provided by the CNES. Finally, the Python 3.8 code used for this publication is based on a Github project by Tyler Tsutterley (https://github.com/tsutterley/read-GRACE-harmonics). The adapted version can

be found on https://github.com/hulecom/read-GRACE-harmonics repository. Love numbers and $\kappa(r)$ values to estimate Earth's core signals have been kindly provided by Mathieu Dumberry, we thank him for these inputs.

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