

Black sea annual and inter-annual water mass variations from space

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Abstract This study evaluates the performance of two widely used GRACE solutions (CNES/GRGS RL02 and CSR RL04) in deriving annual and inter-annual water mass variations in the Black Sea for the period 2003–2007. It is demonstrated that the GRACE derived water mass variations in the Black Sea are heavily influenced by the leakage of hydrological signals from the surrounding land. After applying the corresponding correction, we found a good agreement with water mass variations derived from steric-corrected satellite altimetry observations. Both GRACE and altimetry show significant annual water mass variations of roughly 7 cm amplitude peaking in May and a semi-annual signal of roughly 3 cm peaking in June and in December. The amplitude of the annual water mass signal varies significantly from year to year and is significantly larger during 2004–2006 than in 2003 and 2007. This is also in agreement with the steric corrected altimetry.

Keywords GRACE · Jason-1 · Water mass variations · Black sea

1 Introduction

Since the launch of the GRACE twin satellites, results obtained from the analyses of GRACE-based models have improved the understanding of mass variations and mass transports in the Earth system, which includes processes in the oceans, atmosphere, hydrosphere, cryosphere and geosphere. In spite of numerous results published in recent years (<http://www.csr.utexas.edu/grace/publications/citation.html>), the quality of GRACE-based models is still the subject of investigation, due to the fact that the ability of GRACE to recover mass variations in a region is of major importance for application to studies such as terrestrial water storage (e.g., Tapley et al. 2004; Wahr et al. 2004) and non-steric sea level change (Chambers et al. 2004; Fenoglio-Marc et al. 2006; Swenson and Wahr 2007).

In this study, we evaluate the ability of two recent GRACE-based models to recover the annual and inter-annual water mass (WM) variations in the Black Sea including the Azov Sea. GRACE-based models from the Center for Space Research at the University of Texas (CSR) and from the Centre National d'Études Spatiales and Groupe de Recherche en Géodesie Spatiale (CNES/GRGS) are used. These two GRACE solutions were selected because they are widely used and produced using different techniques. The CSR solutions require destriping, such as with the decorrelation filter used by Swenson and Wahr (2006), and Gaussian smoothing to remove the residual spatial noise after decorrelation filtering (Chen et al. 2008a). However, any smoothing attenuates the signal and introduces a leakage effect from surrounding regions. Consequently, an investigation of the optimal choice of the radius for the Gaussian smoothing for the region was carried out. CNES/GRGS constrained GRACE solutions, on the other hand, do not suffer from the striping effect. This is because the CNES/GRGS coefficients have been computed

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Fig. 1 The Black Sea with its surrounding countries



with a constraint towards a mean gravity field that optimally reduces the short-wavelength striping of the solutions (Lemoine et al. 2007). Steric corrected Jason-1 satellite altimetry data are used as an independent method of assessing the performance of these two GRACE-based models.

The Black Sea is an interesting study area because it has relatively strong temporal WM variations. Furthermore, altimetry data in the Black Sea can be used as an independent tool for the evaluation of GRACE-based models. The wide drainage area of the Black Sea (Fig. 1) covers a large part of Europe and Asia and provides a total fresh water supply of about $350 \text{ km}^3 \text{ yr}^{-1}$ (Ozsoy and Unluata 1998). River runoff affects the physical characteristics of the sea, and is strongly dependent on the hydrological cycle over continental Europe (Stanev et al. 2002). The large river runoff into the Black Sea changes from year to year and has also a semi-annual component. Therefore, not only the annual but also the semi-annual and the inter-annual components of WM variability of the Black Sea are analyzed.

The leakage effect of land hydrology on the WM variations of the Black Sea is initially quantified using a land hydrology model. Subsequently, the leakage effects on the two GRACE solutions are estimated and used to correct them before the WM variations are derived. Finally, the annual, semi-annual and inter-annual variability is quantified and compared with Jason-1 based estimates.

2 Data and processing

2.1 GRACE level-2 products

Our GRACE time series are from the CNES/GRGS Release-02 (RL02) and CSR Release-04 (RL04) solutions.

CNES/GRGS solutions include one hundred and seventy-one 10-day gravity field solutions expressed in normalized Spherical Harmonic (SH) geopotential coefficients from degree 2 up to degree and order 50, covering the period from 03 March 2003 to 29 December 2007. CSR solutions include fifty-seven monthly gravity field solutions for the period March 2003 to December 2007, supplied as normalized SH geopotential coefficients from degree 2 up to degree and order 60. CNES/GRGS solutions are stabilized towards the EIGEN-GRGS.RL02.MEAN-FIELD mean gravity field at each given epoch, with a constraint law that depends on the degree and order of each coefficient (<http://bgi.cnes.fr:8110/geoid-variations/README.html>). Monthly means from the 10-day solutions are derived by averaging to be consistent with the temporal resolution of the CSR solutions.

The SH coefficients of CSR level-2 monthly solutions from degree 2 up to degree and order 50 are used to be consistent with the CNES/GRGS GRACE solutions. The data are decorrelated by applying a modified version of the Swenson and Wahr (2006) decorrelation filter called P4M6 (Chen et al. 2007; Chen 2008b). For a given SH of order 6 and above, this filter fits a polynomial of order 4 and removes this from even and odd pairs of coefficients. Finally, Gaussian smoothing (Jekeli (1981)) is applied. We test various radii for the Gaussian function, from 500 to 0 km in order to evaluate which represents the best agreement with steric-corrected altimetry in the Black Sea.

Since the degree-one coefficients are not part of these GRACE solutions, an estimate of geocenter motion (Swenson et al. 2008) is added to both GRACE solutions to account for the degree-one components of the gravity field. For both GRACE solutions, atmospheric pressure variations over land and ocean tides have been removed using

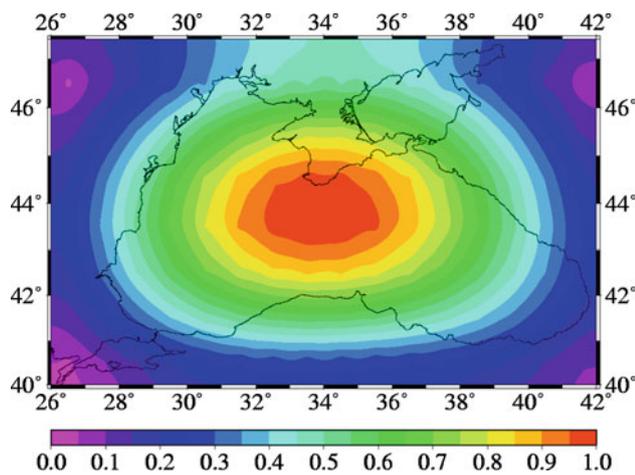


Fig. 2 The averaging kernel used to obtain basin-average GRACE derived WM variations from the CNES/GRGS and the CSR GRACE solutions

the European Centre for Meteorological Weather Forecasting (ECMWF) model (<http://www.ecmwf.int>) and the Finite Element Solution 2004 (FES2004) (Lyard et al. 2006) model respectively. Barotropic ocean signals have been removed from CNES/GRGS solutions using the MOG2D-G barotropic ocean model (Carrère and Lyard 2003). The monthly averages of the MOG2D-G barotropic ocean model output should in principle be restored in the CNES/GRGS monthly solutions since we are interested in the total ocean mass signal (Lombard et al. 2007). However, we do not do that because the applied version of the MOG2D-G model does not cover the Black Sea (Lemoine, personal communication, 2009).

The Ocean Model for Circulation and Tides (OMCT) baroclinic model (Bettadpur 2007; Flechtner 2007) have been used to remove non-tidal short-term oceanic mass variations from the CSR RL04 solutions (Bettadpur 2007). In order to make the CSR solutions consistent with CNES/GRGS solutions, the monthly averages of the short-term non-tidal oceanic contribution are restored to CSR solutions using the CSR RL04 GAD products (Bettadpur 2007).

2.2 Scaling factors for GRACE derived WM variations

The Black Sea basin-averaged WM variations from the two GRACE solutions are derived using averaging kernels (Swenson and Wahr 2002). The averaging kernel used is constructed by expanding the Black Sea land-sea mask into SH up to degree and order 50 (Fig. 2). The mask is defined as 1 at points inside the Black Sea and 0 outside of it.

Various processing steps, such as the truncation of the spherical harmonic expansion, decorrelation filtering and smoothing will tend to reduce the amplitude of the real signal. In order to investigate this, scaling factors were estimated by simulating a uniform 1 cm synthetic mass signal

Table 1 Scaling factors used for CSR solutions that are decorrelated and smoothed with Gaussian filters of varying half-widths

Half-width of Gaussian filter (km)	Scaling factor
0	1.78 (1/0.56)
100	1.89 (1/0.53)
200	2.22 (1/0.45)
300	2.79 (1/0.36)
500	4.59 (1/0.22)

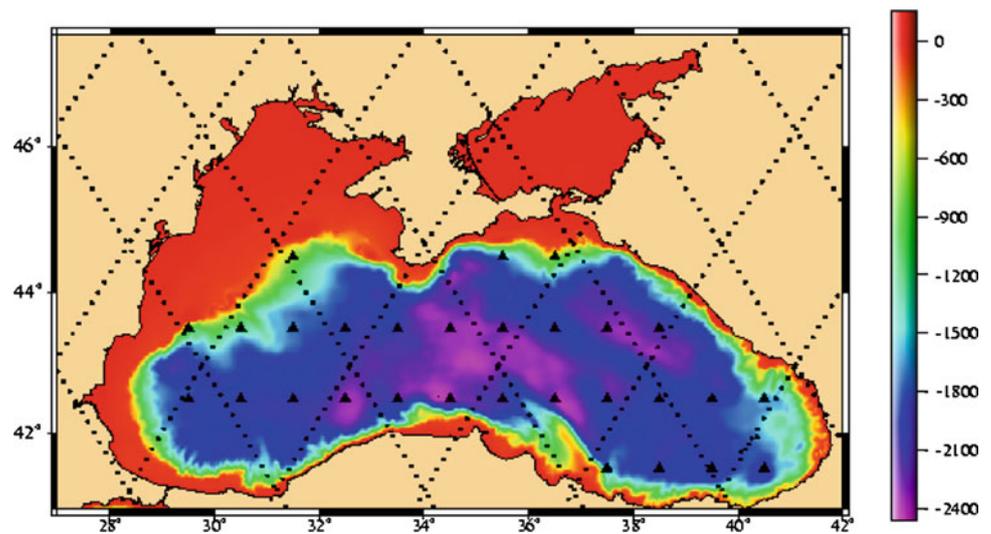
of the Black Sea, processing it in the same manner as the GRACE solutions and comparing the retrieved signal with the original one. After expanding the synthetic signal into SH up to degree and order 50, we obtained 0.65 cm. Therefore, a factor of 1.53 (1/0.65) is used to scale basin-average WM time series of the CNES/GRGS solutions, as no further processing is applied. For the CSR solutions, we decorrelate the synthetic signal using the P4M6 filter and apply a Gaussian smoothing. The decorrelation reduces the basin average to 0.56 cm, and the Gaussian smoothing reduces it further to 0.22–0.53 cm, depending on the filter half-width (Table 1). Therefore, scaling factors given in Table 1 are used for the various filtered CSR solutions. A test of the scaling factor for the Caspian Sea revealed slightly different result than the one of Swenson and Wahr (2007) (2.7 versus 2.4). We believe that the small difference is due to a different land–sea mask and/or a difference in the P4M6 decorrelation filter.

2.3 Terrestrial water mass variations

Terrestrial water mass (TWM) variability around the Black Sea exceeds the WM variability in the Black Sea, and any TWM will leak into the sea areas due to the limited degree and order of the SH expansion and the spatial filtering applied. In order to estimate this leakage, monthly TWM variability at one degree resolution is derived from NASA's Global Land Data Assimilation System (GLDAS) (Rodell et al. 2004 <http://csr.utexas.edu/research/ggfc/dataresources.html>). The Noah land surface model version 2.7.1 (Ek et al. 2003) with observed precipitation and solar radiation included was used. Also, the Climate Prediction Center (NOAA/CPC) land hydrology model was investigated (Fan and van den Dool 2005), but problems with the land–sea mask rendered this model unrealistic for the Black Sea region.

For consistency, GLDAS data were processed like the GRACE-based models (Swenson and Wahr 2007): GLDAS data were expanded into SH to degree and order 50 for CNES/GRGS solutions and additionally decorrelated using P4M6 filter and smoothed by a Gaussian filter in the context of CSR solutions.

Fig. 3 Water depths of the Black Sea in meters derived from the DNSCO8 bathymetry data (Andersen et al. 2010). The $1^\circ \times 1^\circ$ grid points used for the computation of steric sea level at 300 m depth are represented by the *black triangles*. The Jason-1 altimetry tracks covering the Black Sea are depicted with *black dots*



2.4 Sea level data

Monthly altimetry data from the Jason-1 satellite for the period from March 2003 to December 2007 were extracted from the Version 3.1 of the Radar Altimeter Database System, applying the standard corrections (Scharroo 2009). The monthly averages of along-track Jason-1 sea level data were spatially averaged over the Black Sea and the Azov Sea. The Jason-1 altimetry tracks covering the Black Sea are depicted with black dots in Fig. 3. Note that the inverse barometer (IB) correction was not applied to the altimetry-based sea level data, in order to be consistent with both GRACE solutions that observe the total sea mass signal (Lombard et al. 2007).

2.5 Steric heights from WOA05 seasonal climatology data

Steric sea level variations are not associated with WM variations, and in order to determine and remove the steric contribution to the observed altimetric sea level variations, the monthly temperature and salinity data on one degree resolution grids from the World Ocean Atlas 2005 (WOA2005) database are used (Locarnini et al. 2006; Antonov 2006). The steric sea level (SSL_{WOA05}) was estimated by integrating the specific volume anomaly from the surface to a depth of 300 m, as Tsimplis et al. (2004) suggested that there is no significant contribution to the specific volume anomaly from water below 300 m in the Black Sea. Depth contours and grid points used for the SSL_{WOA05} computation are depicted in Fig. 3. Due to the 300 m depth constraint, the averaged steric sea level variation could only be computed from 29 out of the 51 grid points in the Black Sea (Fig. 3).

3 Results

The leakage effects from the land surrounding the Black Sea are subtracted from the GRACE derived signals. Then, WM variations at annual and semi-annual scales are quantified by fitting a bias, annual and semi-annual terms to the monthly basin-average time series and compared with Jason-1 based estimates. A realistic estimate of the magnitude of errors for the WM estimates cannot be provided as we were unable to obtain realistic estimates of the magnitudes of the errors for the GLDAS model and the steric effect.

3.1 Leakage effect from the surrounding land

Annual TWM variations from the GLDAS model surrounding the Black and Caspian Seas are presented in Fig. 4 corresponding to different GRACE processing strategies, in order to demonstrate the importance of accounting for leakage from nearby TWM especially due to Gaussian smoothing. The figure shows that the TWM signal is considerably larger around the Black Sea than around the Caspian Sea where Swenson and Wahr (2007) obtained the best result using a Gaussian smoothing of 300-km. Consequently, the leakage will increase in the Black Sea with larger smoothing. Figure 4a shows the annual amplitude of TWM variations derived from the original GLDAS data, which ranges up to 10–20 cm at the southeastern border of the Black Sea. The truncated, decorrelated and smoothed GLDAS data in Fig. 4 are not scaled. In order to quantify the leakage effect, the annual amplitudes of the scaled basin-average hydrological leakage are presented in Tables 2 and 3, corresponding to CNES/GRGS and CSR GRACE solutions, respectively.

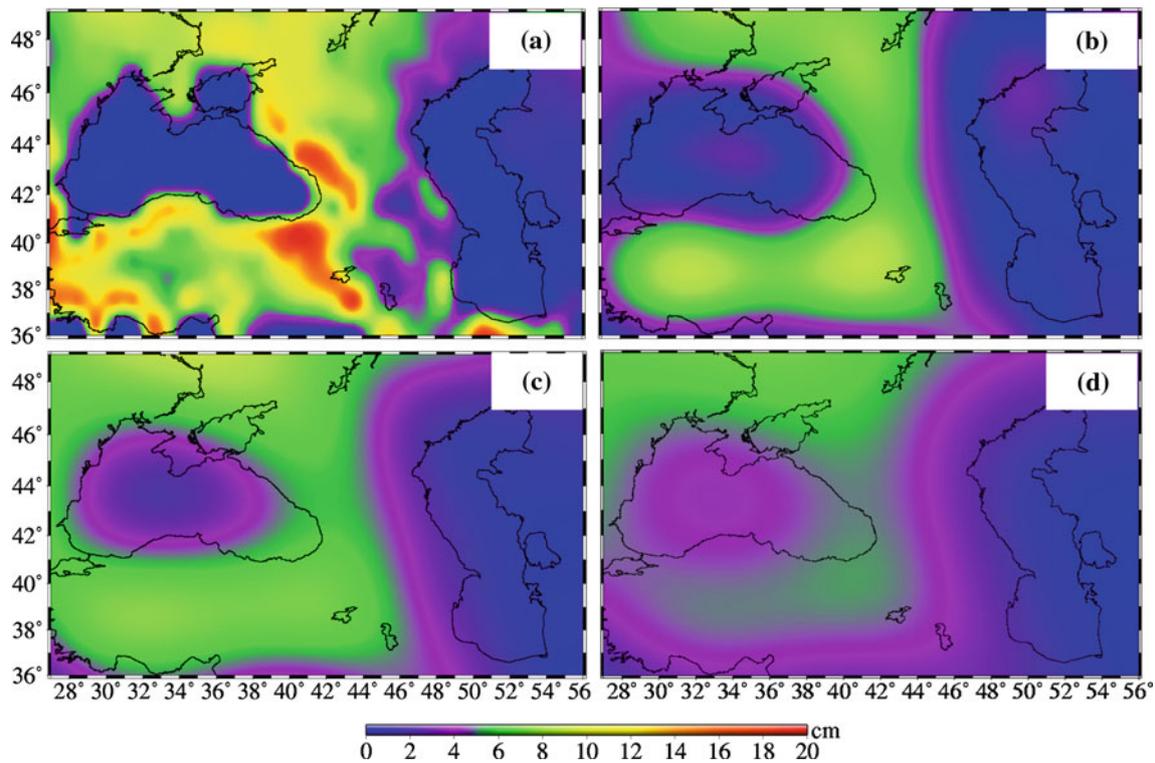


Fig. 4 Annual amplitude (cm) of the TWM from the GLDAS model **a** original GLDAS data averaged into one degree resolution **b** GLDAS data expanded into spherical harmonics (*SH*) up to degree and order 50 **c** GLDAS data in Fig. 4b, decorrelated using the P4M6 filter and

smoothed using the 300-km Gaussian filter **d** GLDAS data in Fig. 4b, decorrelated using the P4M6 filter and smoothed using the 500-km Gaussian filter. Note that the GLDAS data presented in this figure are not scaled

Table 2 Annual and semi-annual amplitudes and phases of the CNES/GRGS GRACE mass time series, GLDAS WM time series and the GRACE derived WM time series (CNES/GRGS-GLDAS) obtained

Basin-average time series	Annual signal		Semi-annual signal	
	Amplitude (cm)	Phase (days)	Amplitude (cm)	Phase (days)
CNES/GRGS	8.2 ± 1.2	99 ± 9	2.8 ± 1.2	162 ± 13
GLDAS	3.2 ± 0.3	61 ± 5	0.2 ± 0.3	89 ± 38
CNES/GRGS-GLDAS	6.4 ± 1.3	127 ± 12	2.9 ± 1.3	162 ± 13

after correcting the basin-average of CNES/GRGS GRACE mass time series for the leakage signal

The results presented in this table are obtained from the scaled basin-average time series

Table 2 shows that the effect of smoothing from the truncation of the SH at degree and order 50 (Fig. 4b) generates a small hydrological leakage into the sea. Table 3 shows that the leakage is larger, when decorrelation and the 300-km Gaussian smoothing are applied (Fig. 4c). It becomes even larger after the 500-km smoothing (Fig. 4d). Therefore, we investigated the optimal radius of Gaussian smoothing for the CSR solutions. Several half-widths of Gaussian filters (0, 100, 200, 300 and 500 km) were tested. Table 3 compares the annual and semi-annual components of the CSR derived WM estimates and Table 4 evaluates them with steric corrected altimetry in terms of RMS difference and temporal correlation. We found that the larger the half-width of the

Gaussian filter, the larger the amplitude of the WM variations, and hence the larger the amplitude difference between CSR GRACE and Jason-1 derived WM variations.

The best agreement with Jason derived WM variations was obtained for unsmoothed (but decorrelated) CSR solutions in terms of amplitude, phase of the annuals signals (Table 3) as well as rms difference whereas the temporal correlation was equally high for various filtered CSR solutions (Table 4). Consequently we used the unsmoothed solution for further analyses.

If it was possible to correct for leakage perfectly, the results would not be dependent on the choice of the smoothing radius for the CSR solutions. However our attempts were

Table 3 Annual and semi-annual amplitudes and phases of the sinusoids best fitting to the monthly basin-averages of CSR GRACE, GLDAS and CSR GRACE derived WM (CSR–GLDAS) time series

Basin-average time series	Gaussian filter half-width (km)	Annual signal		Semi-annual signal	
		Amplitude (cm)	Phase (days)	Amplitude (cm)	Phase (days)
CSR	0	10.7 ± 1.8	105 ± 10	2.5 ± 1.8	159 ± 21
	100	11.7 ± 1.9	104 ± 10	2.7 ± 1.9	160 ± 21
	200	14.6 ± 2.0	101 ± 8	3.0 ± 2.1	161 ± 20
	300	19.3 ± 2.4	98 ± 7	3.3 ± 2.4	163 ± 21
	500	32.7 ± 3.3	96 ± 6	3.7 ± 3.4	165 ± 27
GLDAS	0	4.8 ± 0.4	67 ± 5	0.4 ± 0.4	124 ± 29
	100	5.7 ± 0.5	68 ± 5	0.5 ± 0.5	122 ± 28
	200	8.3 ± 0.6	69 ± 4	0.7 ± 0.6	118 ± 25
	300	12.7 ± 0.8	70 ± 4	1.1 ± 0.8	115 ± 23
	500	25.2 ± 1.4	71 ± 3	2.1 ± 1.4	112 ± 20
CSR -GLDAS	0	7.4 ± 1.9	129 ± 15	2.4 ± 1.9	164 ± 23
	100	7.7 ± 1.9	129 ± 15	2.6 ± 2.0	165 ± 22
	200	8.5 ± 2.2	131 ± 15	3.0 ± 2.2	168 ± 21
	300	10.0 ± 2.6	135 ± 15	3.5 ± 2.6	172 ± 21
	500	14.3 ± 3.7	143 ± 15	4.6 ± 3.7	178 ± 23

The results presented in this table are obtained from the scaled basin-average time series

Table 4 Temporal correlations and RMS differences between JASON-1 derived basin-average WM and CSR GRACE derived scaled basin-average WM time series (CSR–GLDAS) obtained after applying P4M6 decorrelation filter and Gaussian filters of varying half-widths

Half-width of Gaussian filter (km)	Temporal correlation	RMS (cm)
0	0.87	4.79
100	0.87	4.86
200	0.88	5.26
300	0.88	6.34
500	0.86	10.51

not very successful, which was partly due to the fact that the GLDAS model does not model groundwater variations whereas GRACE observes the integrated water storage.

3.2 Water mass variations on annual and semi-annual scales

The Jason-1 shows an annual amplitude of 6.8 ± 1.1 cm peaking in June. The steric sea level from WOA05 has an annual signal with an amplitude of 2.2 ± 0.9 cm peaking in August–September, which agrees well with values of [Stanev et al. \(2000\)](#) who observed an annual amplitude of 2.5 cm peaking in mid-August calculated from climatic monthly mean heat fluxes. The steric correction reduces the amplitude of the Jason-1 time series slightly and advances the phase by

obtained after correcting the basin-average of CSR GRACE time series for the leakage signals

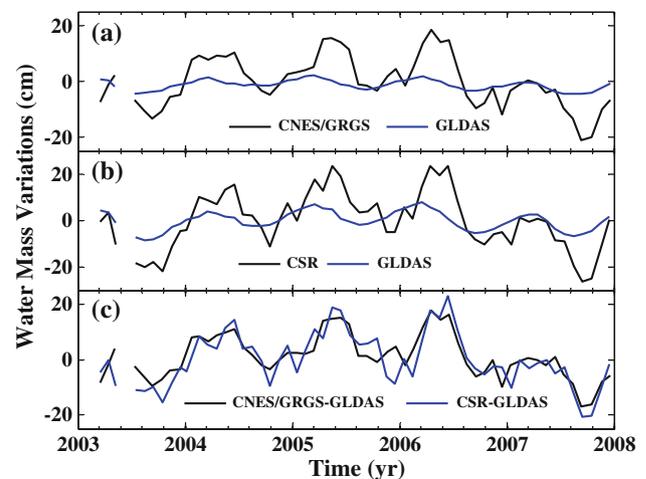


Fig. 5 Scaled basin-average time series of the mass variations from GRACE and the leakage effects from GLDAS for **a** CNES/GRGS solutions **b** CSR GRACE solutions after the P4M6 decorrelation filter is applied **c** GRACE derived WM time series obtained after correcting for the leakage effects

one month. The resulting Jason-1 derived WM signal has an annual amplitude of 6.5 ± 1.4 cm peaking in May–June.

CNES/GRGS WM estimates shows an annual amplitude of 8.2 ± 1.2 cm (Table 2) peaking in April (Fig. 5a). Isolating the Black Sea WM by correcting for the leakage signal (Fig. 5a) from CNES/GRGS yields an annual amplitude of 6.4 ± 1.3 cm peaking in May (Table 2 and Fig. 5c).

Table 5 Annual and semi-annual amplitudes and phases of the sinusoids best fitting to the monthly GRACE derived scaled basin-average WM time series and to the other fields (JASON-1 and SSL_{WOA05}) used to obtain Jason-1 derived basin-average WM time series

Basin-average WM time series	Annual signal		Semi-annual signal	
	Amplitude (cm)	Phase (days)	Amplitude (cm)	Phase (days)
JASON-1	6.8 ± 1.1	171 ± 9	2.2 ± 1.1	179 ± 14
SSL_{WOA05}	2.2 ± 0.9	246 ± 23	0.7 ± 0.9	69 ± 35
JASON-1 derived WM	6.5 ± 1.4	152 ± 12	2.8 ± 1.4	174 ± 14
CNES/GRGS–GLDAS	6.4 ± 1.3	127 ± 12	2.9 ± 1.3	162 ± 13
CSR–GLDAS ^a	7.4 ± 1.9	129 ± 15	2.4 ± 1.9	164 ± 23

^a CSR solutions and the corresponding GLDAS model outputs are truncated to degree and order 50, decorrelated by P4M6 filter and no Gaussian smoothing is applied

CSR WM estimates have annual amplitudes of 10.7 ± 1.8 cm (Table 3) peaking in April (Fig. 5b). Removing the leakage signal from the CSR solutions yields an annual amplitude of 7.4 ± 1.9 cm peaking in May (Table 3 and Fig. 5c).

Table 5 compares CNES/GRGS and CSR derived WM estimates with the steric corrected altimetry. There is a fairly good agreement between the both GRACE derived WM estimates and the Jason-1 derived WM signal in terms of both phase and amplitude and furthermore the results are consistent with Stanev et al. (2000) who suggested that river runoff into the Black Sea peaks in May.

The presence of the semi-annual signal is a consequence of a small secondary maximum in December–January observed in both GRACE derived and Jason-1 derived WM time series (Fig. 6). This maximum is likely explained by the increased precipitation in fall (Stanev et al. 2000). The Jason-1 derived WM signal has a semi-annual signal with an amplitude of 2.8 ± 1.4 cm peaking in December–January and June–July. CNES/GRGS and CSR GRACE solutions have semi-annual signals with amplitudes of 2.9 ± 1.3 cm and 2.4 ± 1.9 cm, respectively, both estimates peaking in December–January and June–July.

3.3 Inter-annual water mass variations

The two GRACE derived WM time series are shown in Fig. 6 in comparison with the WM estimated from steric corrected Jason-1 altimetry. The scaled WM time series from the two GRACE based estimates have a temporal correlation of 0.84 with an rms difference of 5.4 cm with each other. The WM time series from CNES/GRGS and CSR solutions show a good agreement with Jason-1 derived WM (Fig. 6) and the temporal correlations are 0.83 and 0.87 with rms differences of 4.6 and 4.80 cm, respectively. Figure 6 also shows that the annual WM varies significantly from year to year. The annual variation is largest in 2004, 2005 and 2006, and significantly reduced in 2003 and 2007. Also the data shows a 10–15 cm drop in WM during the second half of 2007. This drop can potentially be caused by increasing

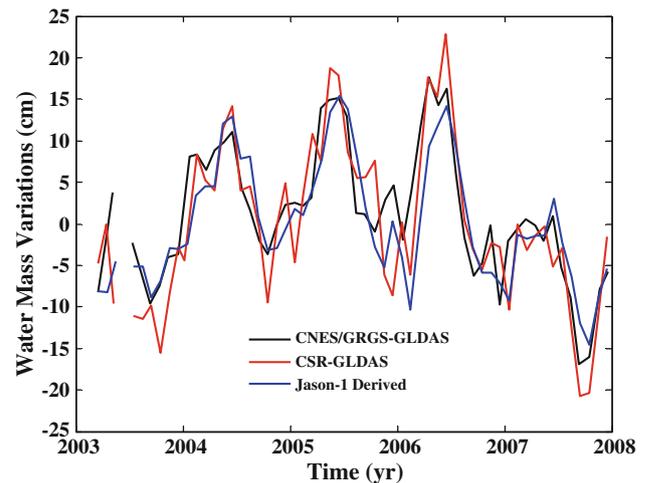


Fig. 6 Comparison of scaled GRACE derived and Jason-1 derived basin-average WM time series. *Black curve* and *red curve* represent CNES/GRGS based and CSR WM based estimates corrected for leakage effects, respectively. Jason-1 derived WM variations corrected for steric effect are shown in the *blue curve*

evapotranspiration or reduced precipitation over the river drainage basin during the summer of 2007.

4 Conclusion

Two widely used GRACE solutions (CNES/GRGS RL02 and CSR RL04) have been compared with steric corrected Jason-1 altimetry for the period 2003–2007 in order to assess their performance in deriving regionally averaged oceanic WM variations at annual and inter-annual scales in the Black Sea.

The results show that GRACE estimated WM variations in the Black Sea are heavily influenced by the leakage of hydrological signals from the surrounding land. Consequently, the CSR solutions which normally require filtering are more affected by leakage than the CNES/GRGS solutions. GLDAS was applied to study the leakage, but as GLDAS only models the soil moisture signal in the upper 2 meters of soil and

ignores WM variations from ground water and surrounding open bodies (e.g. the Mediterranean Sea, etc.) it was impossible to precisely quantify and correct for TWM leakage for the Black Sea. Comparison with WM from satellite altimetry led us to prefer the unsmoothed (no Gaussian filtering) decorrelated CSR solutions, along with and CNES/GRGS GRACE solutions, for further investigation.

Good agreement between both GRACE solutions and the steric corrected altimetry was found in terms of both amplitude and phase. This showed that GRACE can be used to derive water mass variations in the Black Sea and other regions of comparable size, provided that the signal leakage from surrounding regions is corrected for. Both GRACE WM estimates show an annual signal of roughly 7 cm peaking in May and a semi-annual signal of roughly 3 cm peaking in June and December. Semi-annual signals in Black Sea WM variations are small but significant, as they modify the shape of the annual signal, creating a secondary maximum in December.

Similarly, the amplitude of the annual signals varies significantly from year to year and is significantly larger during 2004–2006 than in 2003 and 2007. These year-to-year variations in the annual water mass mostly reflect the variations in inflow and outflow of water mass in the Black Sea through nearby rivers and straits as also suggested by [Stanev et al. \(2000\)](#).

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