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6 **Analysis of Terrestrial Water Storage Changes from GRACE**
7 **and GLDAS**
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9 **Tajdarul H. Syed¹, James S. Famiglietti¹, Matthew Rodell², Jianli Chen³, Clark R. Wilson⁴**
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11 ¹Department of Earth System Science, University of California, Irvine

12 ²Hydrological Sciences Branch, NASA Goddard Space Flight Center, Greenbelt, MD

13 ³Center for Space Research, University of Texas at Austin

14 ⁴Department of Geological Sciences, University of Texas at Austin
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17 Corresponding author
18 James S. Famiglietti
19 jfamigli@uci.edu
20 Phone: 949-824-9434
21 Fax: 949-824-3874
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1 **Abstract**

2 Since March 2002, the Gravity Recovery and Climate Experiment (GRACE) has provided
3 first estimates of land water storage variations by monitoring the time-variable component of
4 Earth's gravity field. Here we characterize spatial-temporal variations in terrestrial water storage
5 changes (TWSC) from GRACE and compare them to those simulated with the Global Land Data
6 Assimilation System (GLDAS). Additionally, we use GLDAS simulations to infer how TWSC is
7 partitioned into snow, canopy water and soil water components, and to understand how
8 variations in the hydrologic fluxes act to enhance or dissipate the stores. Results quantify the
9 range of GRACE-derived storage changes during the studied period and place them in the
10 context of seasonal variations in global climate and hydrologic extremes including drought and
11 flood, by impacting land memory processes. The role of the largest continental river basins as
12 major locations for freshwater redistribution is highlighted. GRACE-based storage changes are
13 in good agreement with those obtained from GLDAS simulations. Analysis of GLDAS-
14 simulated TWSC illustrates several key characteristics of spatial and temporal land water storage
15 variations. Global averages of TWSC were partitioned nearly equally between soil moisture and
16 snow water equivalent, while zonal averages of TWSC revealed the importance of soil moisture
17 storage at low latitudes and snow storage at high latitudes. Evapotranspiration plays a key role in
18 dissipating globally-averaged terrestrial water storage. Latitudinal averages showed how
19 precipitation dominates TWSC variations in the tropics, evapotranspiration is most effective in
20 the mid-latitudes, and snowmelt runoff is a key dissipating flux at high latitudes. Results have
21 implications for monitoring water storage response to climate variability and change, and for
22 constraining land model hydrology simulations.

- 1 **KEYWORDS:** GRACE, GLDAS, Terrestrial water storage changes, Hydroclimatology, Large
- 2 scale terrestrial water balance.

2 **1. Introduction**

3 Terrestrial water storage (TWS) is defined as all forms of water stored above and underneath
4 the surface of the Earth. TWS is a key component of the terrestrial and global hydrological
5 cycles, exerting important control over the water, energy and biogeochemical fluxes, thereby
6 playing a major role in Earth's climate system [*Famiglietti, 2004*]. For example, soil water
7 storage affects the partitioning of water and energy fluxes at the land surface, with implications
8 for precipitation recycling, hydrologic extremes including drought and flood and by impacting
9 land memory processes [*Shukla and Mintz, 1982; Eltahir and Bras, 1996*]. Surface water storage
10 impacts rates of freshwater, sediment and nutrient transport, and plays an important role in
11 greenhouse gas emissions to the atmosphere [*Richey et al., 2002*]. TWS is a key unknown in the
12 calculation of current rates of global mean sea level rise [*Church et al., 2001*]; and it impacts
13 Earth rotation variations such as length of day [*Chao and O'Connor, 1988*]. As an integrated
14 measure of surface and groundwater availability, TWS has significant implications for water
15 resources management.

16 In spite of its manifold importance, until recently, TWS has not been adequately measured at
17 the continental scale [*Lettenmaier and Famiglietti, 2006*]. This is primarily due to the lack of a
18 comprehensive global network for routine TWS monitoring. While ground and satellite based
19 techniques can measure some individual components such as soil moisture [*Njoku et al., 2003*]
20 and surface water [*Alsdorf and Lettenmaier, 2003*], there has been no integrated measurement of
21 TWS.

22 The dearth of direct observations of large scale TWS estimates was resolved by the launch
23 of Gravity Recovery and Climate Experiment (GRACE) twin satellite mission in March, 2002.
24 Although primarily aimed at accurately mapping time variations in Earth's gravity field at ~ 30

2 day intervals, GRACE has shown remarkable prospects for inferring water mass changes over
3 the globe [Tapley *et al.*, 2004a; Wahr *et al.*, 2004].

4 Most GRACE hydrology studies to date, including those described above and in Section 2
5 below, have dealt with either comparison of derived terrestrial water storage anomalies (TWSA,
6 i.e. TWS deviations from the mean rather than month to month changes) to models and to limited
7 observations; methods of data processing and error analyses; and to new applications for
8 monitoring TWS components and fluxes. While the capability for GRACE to monitor
9 continental scale anomalies and changes in monthly water storage is now well documented, little
10 if any work has addressed fundamental issues such as the characterization of its space-time
11 variability and its role in terrestrial hydroclimatology, namely how observed TWSC is
12 distributed amongst the terrestrial subsurface and surface stores, and how the fluxes of
13 precipitation, evapotranspiration and runoff act to enhance or dissipate the storages.

14 Here we present a detailed analysis of continental scale water storage changes using
15 GRACE and output from a high quality global land hydrological modeling system. Unlike the
16 other studies described here, the emphasis of this work is towards understanding the spatial-
17 temporal variability in the role of different hydrologic fluxes and storages influencing the
18 magnitude and distribution of TWSC over the globe. Note that the lack of global-scale
19 observations of TWSC necessitates a model-based approach to the analyses. In the first part of
20 our work we focus on the characterization of spatial-temporal variability in observed storage
21 changes over land. Total water storage changes are quantified over the different continents and in
22 some of its largest river basins. In the second part we compare estimates of TWSC from GRACE
23 and the Global Land Data Assimilation System (GLDAS; Rodell *et al.*, 2004b). Having
24 demonstrated good agreement between the two, in the third part of the analysis, we discuss the

2 zonal and global patterns of variability in TWSC and how these patterns are controlled by the
3 various hydrologic and climatologic factors, using GLDAS-based states and fluxes.

4

5 **2. Background**

6 GRACE is a joint venture between NASA and the DLR launched in March 2002. The
7 mission objective is to accurately measure the mean and time varying component of Earth's
8 gravity field at monthly time scales for a period of at least 5 years. The mission consists of twin
9 satellites spaced ~ 220 km apart in a near circular polar orbit at an altitude of ~500km. Spatial-
10 temporal variations in Earth's gravity field affect the distance between the two satellites: a
11 continuous and accurate measurement of changes of this distance (inter-satellite range) by the
12 onboard K-Band microwave ranging system [Tapley *et al.*, 2004a], combined with other
13 ancillary data, enables precise maps of Earth's time-variable gravity field to be produced. Over
14 land, time variations of these global gravity fields are primarily due to water mass variations
15 [Wahr *et al.*, 1998; Tapley *et al.*, 2004b]. This has allowed for the first time, observations of
16 variations in TWS at large river basin [Swenson *et al.*, 2003; Chen *et al.*, 2005; Seo *et al.*, 2006;
17 Winsemius *et al.*, 2006] to continental scales [Wahr *et al.*, 2004; Ramillien *et al.*, 2005; Klees *et*
18 *al.*, 2007]. Extraction of these hydrologic signals over land by the removal of effects from other
19 time-varying geophysical factors is one of the prime motivations behind the GRACE satellite
20 mission and is an active area of research.

21 *Rodell and Famiglietti* [1999, 2001] showed promising results in pre-launch assessment of
22 some of the key aspects of GRACE, such as the potential detectability and accuracy of
23 measuring TWSC, using estimated GRACE errors and modeled and observed water storage data.
24 Since the mission began, several studies have described GRACE's ability to detect water storage

2 changes at varied spatial scales over different parts of the globe [*Wahr et al.*, 2004; *Ramillien et*
3 *al.*, 2004], to monitor the mass balance of the ice sheets [*Velicogna and Wahr*, 2006a, 2006b], to
4 quantify fluxes [*Rodell et al.*, 2004a; *Syed et al.*, 2005; *Ramillien et al.*, 2006; *Swenson and*
5 *Wahr*, 2006a] and storages [*Rodell et al.*, 2007; *Yeh et al.*, 2006; *Frappart et al.*, 2006a; *Schmidt*
6 *et al.*, 2006] in land surface hydrology and for the validation and improvement of the terrestrial
7 water balance in global land surface models [*Niu and Yang*, 2006; *Swenson and Milly*, 2006]. In
8 addition there are numerous studies addressing different filtering techniques to retrieve water
9 storage change signals and its associated error structures [*Seo and Wilson*, 2005; *Swenson and*
10 *Wahr*, 2002; *Chen et al.*, 2005; *Ramillien et al.*, 2005, *Swenson and Wahr*, 2006b].

11 Herein we present a study, complementary to earlier studies, but primarily aimed at
12 characterization and understanding of the role of TWSC in terrestrial hydroclimatology. Our
13 overall emphasis is on the analysis of process controls and partitioning of continental water
14 storage changes at varied spatial and temporal scales using state of the art assimilated
15 hydrological model data. This will help us in trying to understand how the terrestrial storage is
16 partitioned at different spatial and temporal scales and how these estimates are affected by the
17 hydrologic fluxes at similar scales.

18

19 **3. Methods**

20 In order to investigate the water storage changes, corrected GRACE Stokes coefficients
21 (Level 2 Gravity Field Product User Handbook, *Bettadpur, S.*, 2003) provided by the Center for
22 Space Research (CSR) at the University of Texas at Austin were expanded to degree and order
23 60 and smoothed with a 1000 km half-width Gaussian averaging kernel to produce the time
24 varying gravity estimates. The coefficients of the lowest degree zonal harmonics, the degree two

2 and order zero term was not taken into consideration, mainly due to large unquantifiable errors
3 associated with this term. Subsequently these smoothed spherical harmonic coefficients were
4 transformed into 1x1 degree gridded data that reflect vertically-integrated water mass changes
5 averaged over a few hundred kilometers with an accuracy of ~1.5 cm of equivalent water
6 thickness [Wahr *et al.*, 2004]. Since we average gridded land water storage changes over much
7 larger spatial domains for this analysis, the error at these spatial scales is smaller than 0.1
8 cm/month [Ramillien *et al.*, 2006]. Errors in GRACE data estimated by the above mentioned
9 studies represent a combination of measurement and processing errors, see Wahr *et al.*, [2006]
10 for additional details. The reader is referred to Wahr *et al.* [1998] and Tapley *et al.* [2004a] for a
11 more detailed description of the processing of GRACE data. Note that techniques for processing
12 GRACE data continue to evolve and improve [Han *et al.*, 2005; Seo and Wilson, 2005; Swenson
13 and Wahr, 2006a]. The GRACE data set used in this study is CSR RL01 which spans from April
14 2002 to July 2004 excluding some months in 2002 (May, June and July) and June in 2003.
15 Longer time periods of GRACE data are becoming available and will allow for studies of
16 interannual variations. The impact of the length of the smoothing radius has also been addressed
17 [Chen *et al.*, 2006]. However, for the purposes of this work, the above-described dataset
18 sufficiently captures the key features of terrestrial hydroclimatology.

19 The primary land surface flux and storage component data were obtained from NASA's
20 Global Land Data Assimilation System (GLDAS) [Rodell *et al.*, 2004b]. GLDAS parameterizes,
21 forces, and constrains multiple land surface models with ground and satellite observation based
22 datasets, towards the goal of accurate simulation of water and energy cycle states and fluxes. For
23 this study we used 1-degree, 3-hourly output from a 1979-present run of the Noah land surface
24 model [Ek *et al.*, 2003] driven by GLDAS. Due to the model's inability to represent ice sheet

2 flow and mass balance, Antarctic was not simulated and output from Greenland was excluded
3 from the analysis. For this investigation we extracted the relevant hydrological fluxes and
4 storages from January, 2002 to December, 2004, and aggregated them to monthly averages or
5 accumulations as appropriate (Table 1).

6 GRACE-derived TWSC estimates were obtained by differencing the monthly TWS
7 anomalies, which themselves were obtained by removing the mean gravity field from each of the
8 monthly GRACE solutions. These estimates of TWSC can be interpreted as average changes in
9 TWS from one month to the other.

10 A comparable replication of GRACE observations from GLDAS land surface output is
11 based on the following equation

$$12 \quad \left[\frac{\Delta S}{\Delta t} \right]_N = \left[\frac{\bar{S}_{i,N} - \bar{S}_{i,N-1}}{\Delta t} \right] \quad (1)$$

13 where \bar{S} represents the average TWS for the indexed day (i), the subscripts i and N represent
14 day of month and month respectively, and t is time. TWS considered here constitutes total
15 column soil moisture (TSM), Snow Water Equivalent (SWE) and Canopy Water Storage (CWS).
16 Neither surface water storage in inland water bodies nor groundwater storage is represented in
17 the model simulations. Both can be important components of TWS in certain regions of the globe
18 [Rodell and Famiglietti, 2001; Frappart et al., 2006b]. Our analysis of storage partitioning is
19 therefore limited to TSM, SWE and CWS and cannot give a complete description of the lateral
20 and vertical distribution of water storage until surface and groundwater components are added to
21 land model used here. Such work is ongoing in our research team. Hence following equation (1),
22 estimates of TWSC from GLDAS that closely approximate GRACE were calculated as follows

$$\left[\frac{\Delta S}{\Delta t} \right]_N = \left[\frac{\left\{ \bar{S}_{soil(15,N)} + \bar{S}_{snow(15,N)} + \bar{S}_{canopy(15,N)} \right\} - \left\{ \bar{S}_{soil(15,N-1)} + \bar{S}_{snow(15,N-1)} + \bar{S}_{canopy(15,N-1)} \right\}}{\Delta t} \right] \quad (2)$$

The terms on the right hand side are 15th day averages of each calendar month of the year. We assume that an averaged estimate of the 15th day can be considered representative of the ~30 day average. The method showed promising results in an earlier study [Syed *et al.*, 2005] and also compared well with other published methods of aggregation of monthly fluxes [Swenson and Wahr, 2006a; Rodell *et al.*, 2004a]. Additionally, TWSC can be computed using a monthly basin-scale terrestrial water balance which can be approximated as follows

$$\left[\frac{\Delta S}{\Delta t} \right]_N = \sum_{N-1}^N P - \sum_{N-1}^N E - \sum_{N-1}^N R \quad (3)$$

where P is precipitation, R is runoff and E is evapotranspiration.

11

12 **4. Results and discussion**

13 **4.1. Water Storage Changes from GRACE**

14 In this section we characterize the spatial-temporal variability in the observed water storage
15 change signals from GRACE. The underlying causes of these variations are discussed in more
16 detail in subsequent sections.

17 Figure 1a shows that the time series of globally-averaged TWSC peaks during NH Winter
18 (DJF) with an amplitude of roughly 0.6 centimeters/month. Figure 1b shows the latitudinal
19 distribution of seasonally-averaged TWSC. A clear dominance of the strongest water storage
20 change signals in a Southern Hemisphere (SH) 0° to 30° S latitudinal band is apparent for all the
21 seasons, with lesser peaks in the NH subtropics and at 60°N. In the tropics, Summer (Winter) is
22 dominated by increases (decreases) in TWSC, due to increases (decreases) in precipitation in

2 response to seasonal migration of the ITCZ. In contrast, mid-latitudes during JJA (DJF) are
3 dominated by decreases (increases) in TWSC, due to increases (decreases) in evapotranspiration.
4 The polar regions are similar to the tropics, but with slight JJA (DJF) increases (decreases) in
5 TWSC, particularly in the NH. The amplitude of the seasonal cycle in the zonally-averaged
6 absolute value of TWSC (Figure 1c) has associated peaks in the corresponding regions.

7 The TWSC variations in the tropics shown in Figure 1b can be readily explained by the
8 migration and strength of the Inter Tropical Convergence Zone (ITCZ), with maxima associated
9 with enhanced precipitation. The hemispheric differences in the amplitudes of TWSC (± 4 - 5
10 cm/month in SH; ± 2 cm/month in NH) are manifestations of greater land precipitation in the SH
11 in comparison to the NH, especially in equatorial South East Asia, South America and Africa
12 [Adler *et al.*, 2003]. Minima correspond to shifts in the subtropical depressions where
13 evapotranspiration increases. In addition to the large fluctuations in the tropics, there is a NH
14 mid-latitude zone of much lower yet prominent variability in the range of ± 1 centimeter/month.
15 Positive storage changes in DJF result from mid-latitude polar frontal precipitation and snow
16 storage. Snowmelt and evapotranspiration account for the decreasing (MAM) and negative (JJA,
17 SON) peaks in this zone. Note that the TWSC variations during SON and MAM can be viewed
18 as intermediate stages of the stronger end members prevalent during JJA and DJF.

19 The amplitude of seasonal cycle in zonally-averaged value of TWSC (Figure 1c) provides
20 perspective on the magnitude of the storage changes, both positive and negative, across the
21 continents. The greatest variation in storage changes occur in the SH Tropics with an amplitude
22 greater than 7 cm/month, followed by the NH Tropics (~ 3.2 cm/month), the NH mid-latitudes
23 (~ 2.4 cm/month) and the SH mid-latitudes (almost 2 cm/month). Figure 1c further highlights
24 where the principal zones for mass exchange between the land and the atmosphere or ocean

2 occur, and that they are consistent with the major features of the atmospheric general circulation
3 and global patterns of precipitation and evaporation [*Hartmann,1994; Peixoto and Oort, 1992*].
4 This also includes the desert regions with zero or low TWSC (near 30° N and S).

5 Figures 1b and 1c have two important implications for terrestrial hydroclimatology. The first
6 is that global scale measurements of TWSC, available for the first time with GRACE, have
7 identified significant regions of dynamic change, and that they are consistent with global patterns
8 of weather and climate. The second, more subtle implication is that the GRACE mission has
9 shown that terrestrial water storage responds in predictable ways to precipitation and evaporation
10 processes, hence providing important “memory” of past atmospheric phenomena.

11 Table 2 lists the annual means, amplitudes of fitted annual cycles and seasonal means of
12 GRACE-based TWSC, averaged for each continent and the river basins shown in Figure 2.
13 Although insignificant compared to the amplitude of the cycles, annual mean values over
14 Europe, South America and Asia show a net accumulation of water mass with values of 0.32
15 cm/month, 0.30 cm/month and 0.08 cm/month respectively for the period of the GRACE data
16 used here. On the other hand, even lesser depletion of total water storage is noted in Australia (-
17 0.13 cm/month), North America (-0.06 cm/month) and Africa (-0.02 cm/month). The seasonal
18 means again point to the influence of ITCZ migration on the distribution of land water storage,
19 similar to what we have noted in Figure 1. While tropical basins in the NH gain water (e.g.
20 Yangtze (2.44 cm/month), Ganges/Brahmaputra (4.65 cm/month), Orinoco (2.80 cm/month) and
21 Niger (2.03 cm/month)) during JJA from enhanced precipitation, basins in the SH tropics and
22 those in NH mid-to-high latitudes tend to lose water (e.g. Zambezi (-3.40 cm/month), Amazon (-
23 2.80 cm/month), Congo (-2.74 cm/month), Ob (-2.80 cm/month) and Lena (-1.93 cm/month))
24 due to lack of precipitation and increased evapotranspiration. On the contrary, basins in the SH

2 tropics tends to gain water during DJF while those in the NH tropics experience a net loss in
3 storage, underscoring the dominant role of climate in defining the spatio-temporal heterogeneity
4 of observed storage change. Furthermore, the amplitude of the annual cycles in South America
5 (4.10 cm/month) stands out from those for the rest of the continents, including the Amazon basin
6 (7.60 cm/month) which has the largest amplitude amongst the river basins. Amplitudes of
7 variability secondary to those in the Amazon are found in Ganges/Brahmaputra (5.80 cm/month),
8 Dniepr (5.28 cm/month) and Zambezi (5.18 cm/month) river basins.

9 It is important to note here that, while it is necessary to smooth the Stokes coefficients from
10 GRACE to reduce the noise in derived mass change fields, the process also suppresses the
11 variability of the storage change signal. The length scale used for smoothing further affects the
12 derived storage change estimates. While a large averaging radius can decrease the strength in the
13 storage change signal [*Chen et al.*, 2006], a smaller radius can produce spurious north-south
14 stripes [*Swenson and Wahr*, 2006b]. Hence, our estimates of mean (annual and seasonal) and
15 amplitude of seasonal cycles based on the use of 1000 km half-width Gaussian averaging kernel
16 are conservative characterizations of basin-to-continental storage changes observed by GRACE.

17 To understand the relative contributions from the large river basins in Figure 2 towards the
18 TWSC for an entire continent, ratios of the sum of absolute value of TWSC in a basin to that of
19 the continent were computed for North America, South America, Africa and Asia. Figure 3
20 illustrates the relative contributions of some of the largest river basins towards the total storage
21 change in North America, South America, Africa and Asia. Also shown in the figure is the
22 percentage of continental area residing within each of the river basins. The results show that just
23 a few of these river basins can account for a notable portion of the total storage change over the
24 entire continent in which the basins are located. This is particularly noteworthy in continental

2 South America, where the change in the Amazon basin is on average of about 45% of the
3 continental storage change, and the aggregate (Amazon, Parana and Orinoco) contributes about
4 70% while the contributing area is ~44% of the area of South America. To a lesser degree,
5 similar results are also seen in Africa and Asia, where aggregated storage changes in the basins
6 shown account for 50%, 35% of the continental water storage changes respectively.

7

8 **4.2. GRACE – GLDAS Comparisons**

9 In this section, we compare seasonal estimates of TWSC from GRACE to those from
10 GLDAS [Rodell *et al.* 2004b]. For consistency with the GRACE data, TWSC from GLDAS was
11 computed using equations (1) – (2). Although not a perfect reproduction of observations, global
12 model output such as that from GLDAS captures the magnitude and variability of terrestrial
13 hydrology sufficiently enough, so that in the absence of any similar, global observational
14 datasets, it provides a reasonable opportunity for evaluation and understanding of the GRACE
15 hydrology signal [Syed *et al.*, 2004]. For comparison with GRACE, GLDAS-based TWSC
16 estimates were converted into spherical harmonic coefficients, smoothed with a 1000 km half-
17 width Gaussian averaging kernel and transformed into 1x1 degree gridded data.

18 Global plots of seasonal storage change estimates obtained from GRACE and GLDAS are
19 shown in Figure 4. GLDAS results used here are for the same period as the GRACE
20 measurements. There is very good overall agreement between the two estimates with Root Mean
21 Square Errors (RMSE) ranging between ~1 cm/month in JJA and ~0.7 cm/month in DJF. Some
22 of biggest storage change signals, consistent with Figure 3, are occurring in the Amazon,
23 Ganges/Brahmaputra, Congo river basins and over large regions of Northern Europe and
24 Western North America. While there are some small differences in magnitude of the TWSC

2 estimates, GLDAS performs reasonably in capturing the global spatial patterns of observed
3 storage changes at seasonal time scales.

4 Time series of TWSC from GRACE and GLDAS for four of the major river basins in
5 continental North and South America are shown in Figure 5. Also included in the plots for
6 Mississippi and Amazon basins are independent estimates of TWSC from a combined land-
7 atmosphere water balance (LAWB) [Syed *et al.*, 2005]. GLDAS estimates agree very well with
8 GRACE, with RMSE values of ~1.5 cm/month in Mississippi and Mackenzie and ~2.5
9 cm/month in Amazon and Parana river basins. Estimates of storage change from GLDAS and
10 LAWB also track each other fairly well in both the Amazon (RMSE = 4.5 cm/month) and
11 Mississippi (RMSE = 1.6 cm/month) basins except for the periods of September-October in 2002
12 and late JJA 2003. Discrepancies between TWSC estimates from GLDAS and LAWB are
13 attributed to errors in the horizontal divergence of water vapor (DivQ) and are discussed in detail
14 by Syed *et al.* [2005]. Furthermore, model estimates of storage change are less variable than
15 GRACE-derived storage changes, primarily due to the absence of contributions from surface and
16 groundwater in the simulations.

17 Overall, Figures 4 and 5 show good agreement in the spatial-temporal variability of TWSC
18 estimates from GRACE and GLDAS. The differences in magnitude between the two estimates
19 can either be due to model deficiencies, such as inadequate snow or missing surface or
20 groundwater components in the models, or due to uncertainties in the GRACE data (e.g. due to
21 processing, aliasing, instrument error, etc.). One consequence of the GRACE errors is that true
22 water storage change signals may be enhanced or dampened in both regional and global scales
23 [Swenson and Wahr, 2006b; Chen *et al.*, 2006a; Seo and Wilson, 2005]. Nevertheless, we believe

2 that the agreement between GRACE and GLDAS is sufficient, so that GLDAS output fields can
3 be studied to better understand the processes contributing to terrestrial water storage variations.

4

5 **4.3. Analysis of process controls in TWSC**

6 **4.3.1 Time Series Analysis**

7 Globally-averaged TWSC, along with its storage (TSM, SWE and CWS, following equation
8 2) and flux (P, E and R) components, obtained from GLDAS, are shown in Figure 6. The annual
9 cycles (solid line) of TWSC and its storage components (Figure 6a) show distinctive seasonal
10 variations with the lowest values during the months of June-July and highs around December of
11 each year. The insignificant role played by CWS in TWSC variations is also clear from the
12 figure so that it is excluded from further discussion.

13 Changes in TSM and SWE contribute nearly equally towards temporal variability in
14 globally-averaged TWSC estimates in terms of both amplitude and phase. SWE estimates, while
15 limited in geographic extent, contribute significantly towards globally-averaged TWSC estimates
16 by the virtue of the larger magnitude of snow water storage. On the other hand TSM estimates
17 have a contrasting feature, i.e. wider spatial extent but smaller magnitudes. Furthermore, peak
18 TSM storage lags that of SWE, indicative of the contribution of snowmelt to soil moisture
19 recharge.

20 Variations in SWE are driven by NH snow storage, which peaks in DJF. Total soil moisture
21 peaks in DJF and reaches a minimum in JJA. The wetting sequence starts in SON and ends in the
22 following MAM, and is an integrated effect of the monsoons in both the hemispheres and snow
23 melt episodes. Global soil moisture begins to dry considerably during late-MAM and reaches its

2 lowest in mid-JJA, when evaporation depletes the water stored from the previous season's rain
3 and snow melt events.

4 Time series of TWSC flux components (Figure 6b) also show distinct annual periodicity and
5 significantly greater amplitudes in the variability of E and TWSC in comparison to P and R. The
6 role of E as a major influence on the variability of globally-averaged TWSC estimates becomes
7 quite apparent from the similarity in amplitude and cross-varying nature of the estimates (TWSC
8 and E are almost mirror images of each other).

9 The dominance of NH climatology is reflected in the high and low E values during JJA and
10 DJF, primarily driven by high and low NH insolation respectively. P estimates, although higher
11 in magnitude, have the smallest amplitude amongst the other fluxes, with highs and lows in late
12 JJA and early MAM respectively. This is also in part due to the greater percentage of land in the
13 Northern Hemisphere. The amplitude of the annual cycle of globally-averaged R is also small
14 relative to E, and closely follows the annual cycles of P and E. Hence we observe that as P, the
15 sole source of water into the system, increases, so too do E and R, so that globally-averaged
16 TWSC and P are actually out-of-phase. The larger amplitude of E dominates annual variations in
17 TWSC.

18

19 **4.3.2. Spatial Analysis**

20 **Zonal Variability.** In this section we explore how latitudinal variations in GLDAS estimates
21 of TSM and SWE contribute to TWSC, and we characterize how P, E, and R fluxes enhance or
22 dissipate TWSC. Here we focus on seasonal time scales instead of monthly.

23 Throughout the seasons, changes in TSM contribute the most towards TWSC in the tropics,
24 while SWE is critical in the NH high latitudes (Figure 7). The highest and the lowest values of

2 TWSC and TSM are mainly centered on 15° north and south of the equator. In addition, the plots
3 also reveal seasonal variations in the latitudinal extent of snow dominance in the estimates of
4 storage change. During DJF, SWE begins contributing to TWSC near 40°N latitude, while in JJA
5 the two estimates become closely related nearer to 60°N. As a result we see increased TWSC in
6 the mid-latitudes (40°N - 60°N) during DJF from snow storage, and a subsequent drop in TWSC
7 values during MAM due to snow pack melting. The role of SWE in high latitudes becomes more
8 evident in MAM, when the values of TWSC near 60°N drop along with SWE, even with the
9 coinciding increase in total column soil moisture estimates due to soil water recharge by
10 snowmelt.

11 During SON and MAM, the overall variability of TWSC and its storage components is
12 intermediary in nature when compared to JJA and DJF. In section 4.1 we discussed similar
13 behavior.

14 The latitudinal variability of TWSC and its component fluxes is shown in Figure 8 for each
15 season. High amplitudes of variability in TWSC estimates along with P, R and E are noted in the
16 ITCZ. As in Figure 1, these high values shift across the equator towards the south during DJF
17 and towards the north during JJA, following the natural variability of the ITCZ.

18 The principal role of P in controlling the zonal averages of TWSC in the tropics is evident
19 from the zonal profiles of all the seasons. While there is little difference in the estimates of E in
20 the tropics through the seasons, significant seasonal variations are noted in the storage changes
21 of the region, mostly related to P. Storage changes in the mid-latitudes are more closely
22 controlled by the E since P decreases away from the tropics.

23 In addition to the maximum TWSC values found in the tropics, a secondary maxima is
24 located around 60°N and 60°S. Secondary maxima in P in this region result from polar frontal

2 convergence. In DJF, low values of E and R contribute to increased TWSC in NH while a
3 significantly increased E in the SH leads to a decrease in TWSC. In MAM and JJA, increases in
4 snowmelt-derived R lead to decreases in TWSC. In fact the MAM and JJA TWSC at the NH
5 mid-to-high latitudes is nearly a mirror image of R due to snow. In summary, we find that P is a
6 dominant control on TWSC variations in the tropics, E plays a critical role in the mid-latitudes,
7 while snow accumulation and snowmelt-driven R is significant in at high latitudes.

8 ***Global Variability.*** The global distribution of seasonally-averaged TWSC and its component
9 fluxes are shown in Figures 9 & 10 for DJF and JJA respectively. Note that different scales are
10 used to portray the spatial heterogeneity over the globe.

11 Figures 9 and 10 support the ideas outlined above regarding how the various fluxes act to
12 increase or decrease TWSC. However, as discussed in Section 4.1, a distinctive feature
13 discernible in all the spatial plots is that the greatest storage changes occur in major river basins
14 over the globe. Some of the key changes in the SH tropics are associated with the Amazon,
15 Parana and Congo river basins and those in the NH with Ganges/Brahmaputra basins in
16 India/Bangladesh along with Mekong in south East Asia and some major African river basins
17 such as the Niger and Volta.

18 In response to the shifting ITCZ, during DJF, South American river basins north of the
19 equator (Orinoco and Magdalena) are seen to lose water, whereas those located south of the
20 equator (Amazon) gain. Similarly, the river basins above the equator in Africa and South East
21 Asia (Niger, Volta and Ganges/Brahmaputra) tend to lose water during this season and the basins
22 below the equator (Congo and Zambezi) gain water.

23

24 **4.4 Correlation Analysis**

2 Figure 11 shows the global and latitudinal distribution of the correlation coefficients
3 between monthly GLDAS-based TWSC and the hydrologic fluxes for the entire length of the
4 simulation. P acts as a positive flux in terrestrial water balance; hence areas with positive
5 correlations are interpreted as areas where the values of TWSC are largely impacted by P. On the
6 contrary, evapotranspiration and runoff are variables that deplete water storage; hence negative
7 correlations are indicative of the regions where these processes are most effective in controlling
8 magnitude and variability of the continental water storage changes.

9 A comparison of the three global correlation plots suggests that positive correlations
10 between precipitation and water storage changes (Figure 11a, first column) have the maximum
11 spatial coverage over the globe followed by the negative correlations between E and R (Figure
12 11b and c, first column). The latitudinal dependence of the controlling processes discussed in the
13 previous section (Figure 9&10) is also evident here. The tropics are consistently dominated by
14 the high positive correlations between P and TWSC while the TWSC estimates in the NH mid-
15 latitudes are correlated more with E than with P. Figure 11 shows a significant increase in
16 correlation between E and TWSC (Figure 11b, second column) in the region between 30°-70°
17 N/S and the concomitant decrease in correlation between P and TWSC (Figure 11a, second
18 column). In addition, the dominance of snowmelt-derived runoff in the NH high latitudes is also
19 distinctly discernible from the considerably higher absolute values of correlation between R and
20 TWSC (Figure 11c, second column).

21

22 **5. Summary and Conclusions**

23 In this study we characterize TWSC variations using GRACE and GLDAS. The results
24 discussed here illustrate spatial-temporal variability of water storage changes over land, with

2 implications for a better understanding of terrestrial water balance and its role in the global
3 hydrologic cycle.

4 Global, zonal and basin-scale estimates of GRACE-based storage changes showed a wide
5 range in variability and magnitude, emphasizing the space-time heterogeneity in TWSC
6 response. Manifestations of hemispheric differences in precipitation were noted in seasonal
7 TWSC. In the SH tropics seasonally-averaged TWSC had higher amplitudes ($\pm 4\text{-}5$ cm/month)
8 of latitudinal variability in comparison to those in the NH (± 2 cm/month). Zonally-averaged
9 TWSC was found to have the greatest amplitude in the SH tropics (~ 7 cm/month), and the spatial
10 distribution showed major TWSC signals coincident with some of the largest river basins.
11 Comparisons between GLDAS and GRACE-based estimates of TWSC at river basin scales
12 compared well with RMSE of ~ 1.5 cm/month in Mississippi and Mackenzie and ~ 2.5 cm/month
13 in Amazon and Parana.

14 Analysis of the hydrologic components in the terrestrial water balance from GLDAS
15 revealed the partitioning and process controls of TWSC, both globally and varying with latitude.
16 The Noah land model used in the GLDAS simulations did not include surface and groundwater
17 stores, so that we were unable to quantify their potentially considerable contributions to storage
18 changes in some regions. Global averages of TWSC were found to be partitioned nearly equally
19 between TSM and SWE. Analysis of zonally-averaged TWSC showed how storage varies by
20 latitude, with changes in soil moisture accounting for most of the storage change at low and mid-
21 latitudes, whereas at high latitudes, TWSC was more closely associated with changes in SWE.
22 Globally-averaged estimates of fluxes showed that E plays a key role in dissipating P-driven
23 storage anomalies. Zonal analysis highlighted variations in the role of these fluxes with respect
24 to TWSC. The P flux dominates TWSC variations in the tropics, and E plays a critical role in the

2 mid-latitudes. In addition, MAM-snowmelt-runoff played a particularly important role in space-
3 time variability of TWSC in the NH high latitudes. The results were further reconfirmed by the
4 correlation analysis, which showed P as the leading flux over major portions of the globe,
5 followed by E in the mid-latitudes and R in the NH high latitudes.

6 Comparison of GRACE-based TWSC with GLDAS model simulations also underscores the
7 potential for validating and improving global land surface models [*Swenson and Milly, 2006; Niu*
8 *and Yang, 2006; Lettenmaier and Famiglietti, 2006*] using GRACE data. Some of the noted
9 differences between GRACE-based TWSC and GLDAS can in part be attributed to the missing
10 surface and groundwater components, or snow parameterization deficiencies. For example, *Niu*
11 *and Yang* [2006] and *Niu et al.*, [2007] showed better agreement with GRACE storage anomalies
12 after including a groundwater component in their land surface parameterization. Future work will
13 be directed toward analysis and comparison of GRACE observations with multiple hydrological
14 models with and without explicit representation of surface and groundwater components in order
15 to fully characterize their role in TWS variations. Results also suggest that with longer time
16 series, GRACE will contribute to improved understanding of how terrestrial water storage
17 responds to climate change and variability.

18

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22 Terrestrial Hydrology and Solid Earth and Natural Hazards Programs.

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18

19

2 **List of Tables**

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4 Table 2. Estimates of annual mean, amplitude of fitted seasonal cycle and seasonal
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6

2 **Figure Captions**

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4 estimates (black dots) are shown along with the fitted seasonal cycle (black solid line).
5 (b) Zonally-averaged TWSC estimates from GRACE for each season, JJA, SON, DJF
6 and MAM. (c) Amplitudes of seasonal cycles fitted to the zonally-averaged absolute
7 value of TWSC estimates from GRACE.

8 Figure 2. River basins referred to in this study: (1) Mackenzie; (2) Mississippi; (3)
9 Magdalena; (4) Orinoco; (5) Amazon; (6) Parana; (7) Volta; (8) Niger; (9) Congo; (10)
10 Zambezi; (11) Nile; (12) Danube; (13) Dniepr; (14) Don; (15) Volga; (16) Ob; (17)
11 Yenisei; (18) Lena; (19) Amur; (20) Ganges/Brahmaputra; (21) Yangtze; (22) Mekong;
12 (23) Murray.

13 Figure 3. Ratio of GRACE-derived TWSC in the listed river basin to that of the
14 entire continent. Total represents the sum of ratios for the river basins considered in each
15 continent. Also shown are (above each bar) the percentage of continental area occupied
16 by each of the river basins.

17

18 Figure 4. Spatial patterns of seasonally-averaged TWSC (cm/month) from GRACE
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21

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4 Mississippi and Amazon basins are TWSC from a Land-Atmosphere Water Balance
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6

7 Figure 6. (a) Time series of globally-averaged TWSC and changes in its storage
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9 storage (CWS). (b) Time series of globally-averaged TWSC and the hydrologic fluxes,
10 precipitation (P), runoff (R) and evapotranspiration (E). All the variables shown here are
11 based on GLDAS outputs. Shown are the monthly estimates (symbols) and fitted
12 seasonal cycles (solid lines) of each variable in their respective colors.

13

14 Figure 7. Latitudinal profile of zonally-averaged TWSC and changes in its storage
15 components (TSM, SWE and CWS) obtained from GLDAS for the four seasons, DJF,
16 MAM, JJA and SON.

17

18 Figure 8. Latitudinal profile of zonally-averaged TWSC and terrestrial hydrologic
19 fluxes (P, R and E) obtained from GLDAS for the four seasons, DJF, MAM, JJA and
20 SON..

21 Figure 9. DJF average of TWSC and fluxes from GLDAS in cm/month: (a) TWSC;
22 (b) precipitation; (c) runoff; and (d) evapotranspiration.

23

2 Figure 10. JJA average of TWSC and fluxes from GLDAS in cm/month: (a) TWSC;
3 (b) precipitation; (c) runoff; and (d) evapotranspiration.

4 Figure 11. Spatial and latitudinal distribution of the correlation coefficients between
5 GLDAS-based TWSC and (a) P; (b) E; and (c) R.

6

2

3 Table 1. GLDAS variables used in this study.

| Parameters | Spatial resolution | Temporal resolution | Time span | Spatial extent |
|--|---------------------------|----------------------------|------------------|---------------------------|
| Precipitation (Includes both solid and liquid rainfall) (P) | 1° x 1° | Monthly sum | Jan'02 -Dec'04 | 180°W-180°E 90°N- 60°S |
| Total soil moisture (4 layers from 0-200 cm depth) (TSM) | 1° x 1° | Monthly average | Jan'02-Dec'04 | 180°W-180°E 90°N-60°S |
| Evapotranspiration (E) | 1° x 1° | Monthly sum | Jan'02-Dec'04 | 180°W-180°E 90°N-60°S |
| Runoff (Includes both surface and subsurface flow)(R) | 1° x 1° | Monthly sum | Jan'02-Dec'04 | 180°W-180°E 90°N-60°S |
| Canopy water storage (CWS) | 1° x 1° | Monthly average | Jan'02 -Dec'04 | 180°W-180°E 90°N-60°S |
| Snow Water Equivalent (SWE) | 1° x 1° | Monthly average | Jan'02-Dec'04 | 180°W-180°E 90°N-60°S |

4

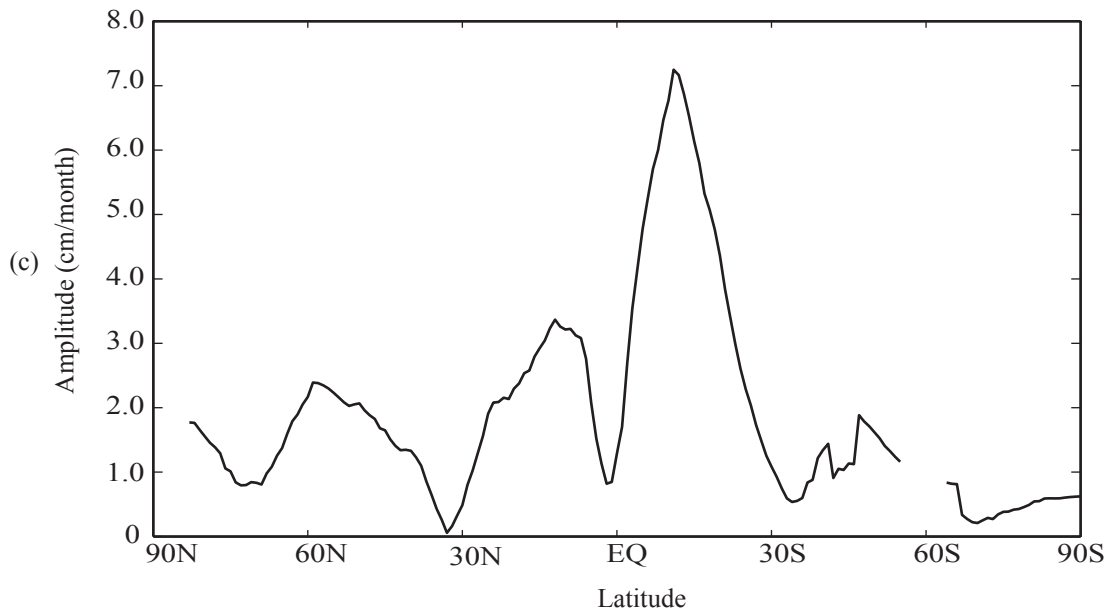
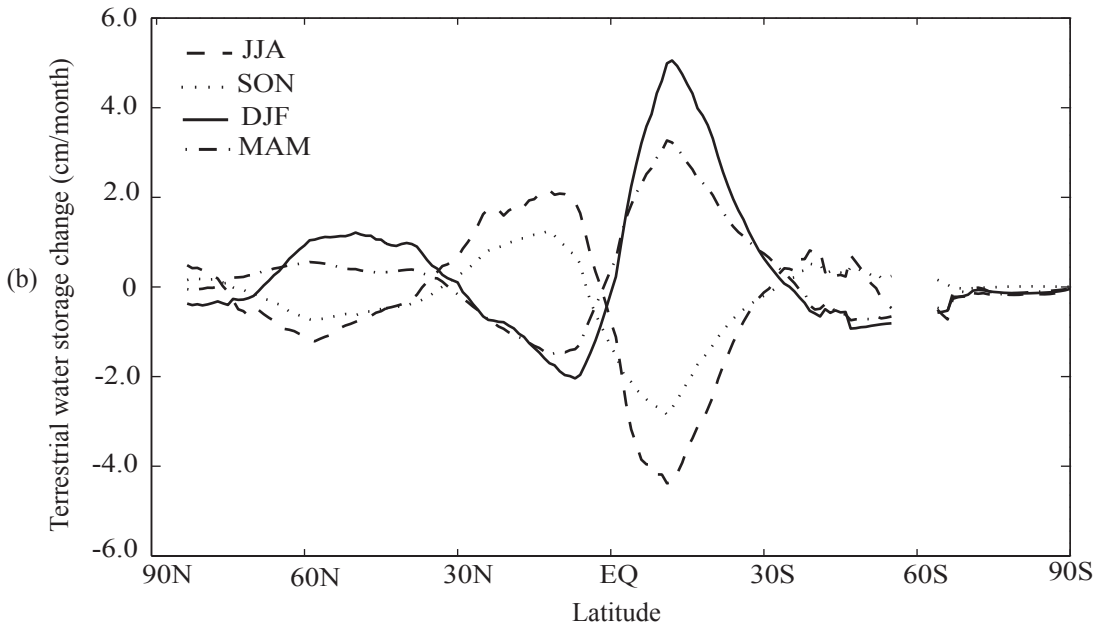
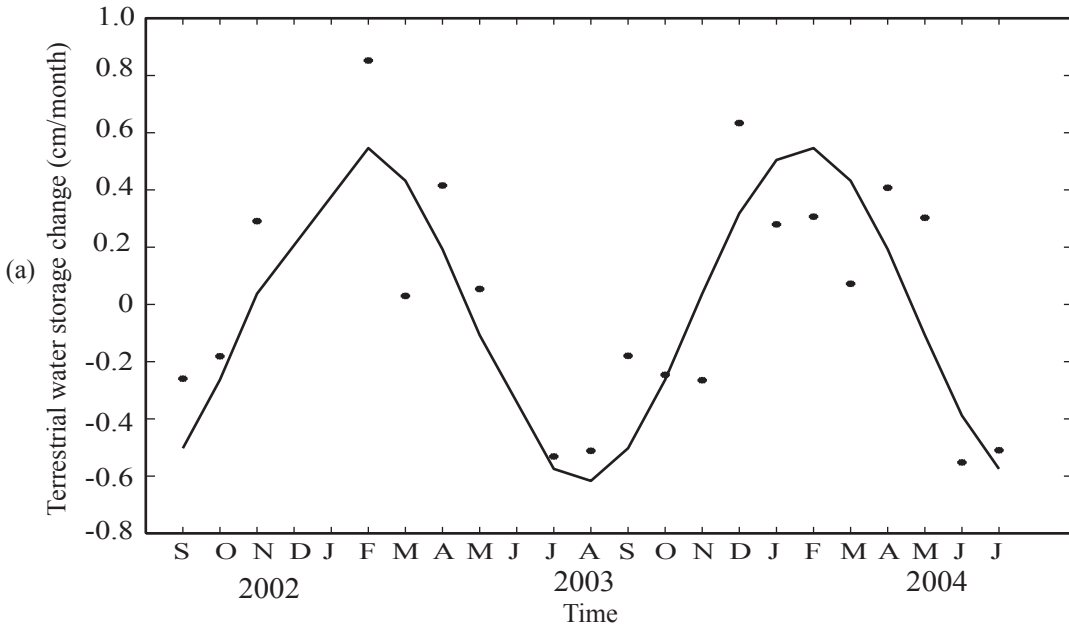
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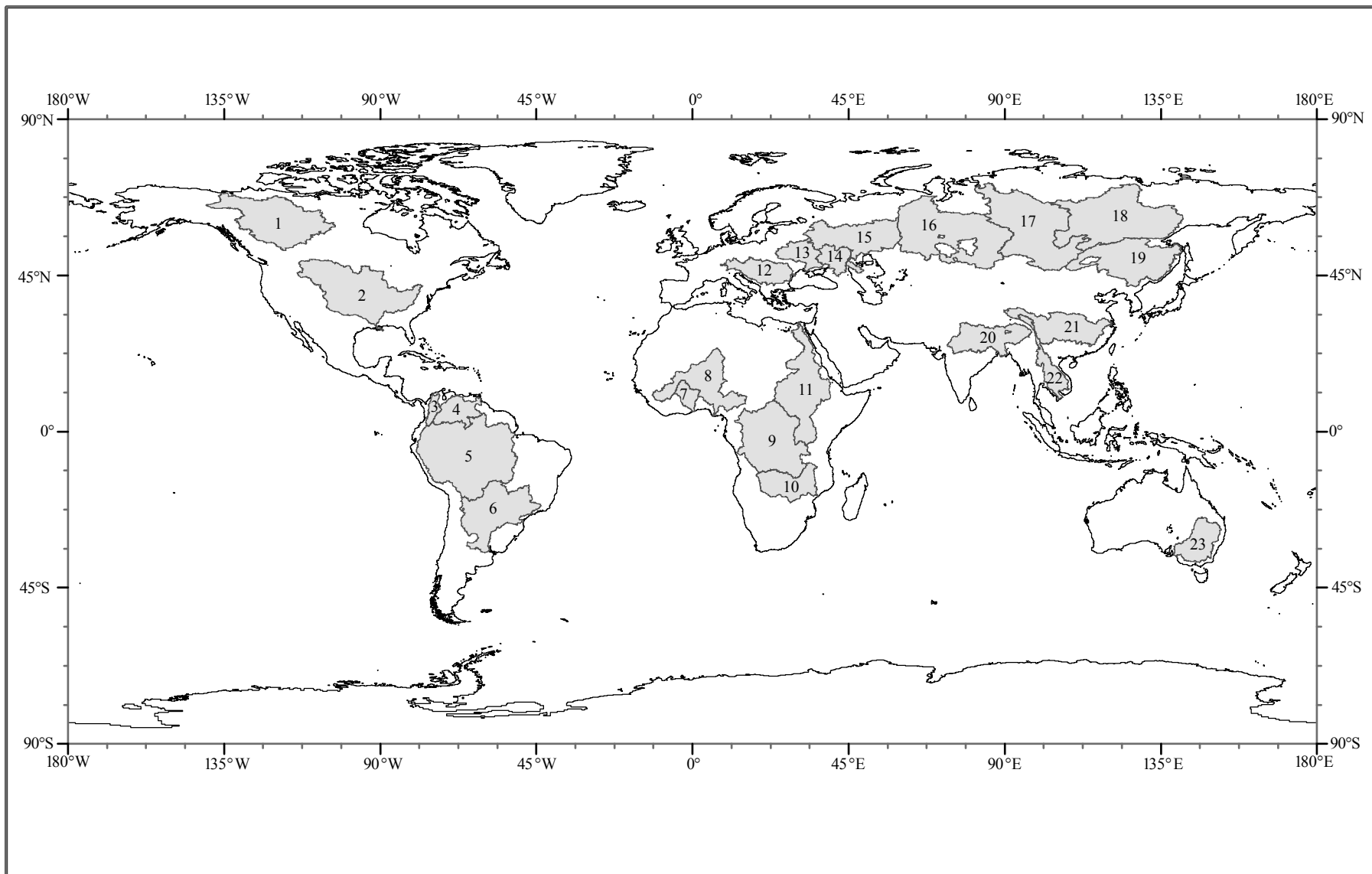
2 Table 2. Estimates of annual mean, amplitude of fitted annual cycle and seasonal mean
 3 for the continents and the largest river basins.

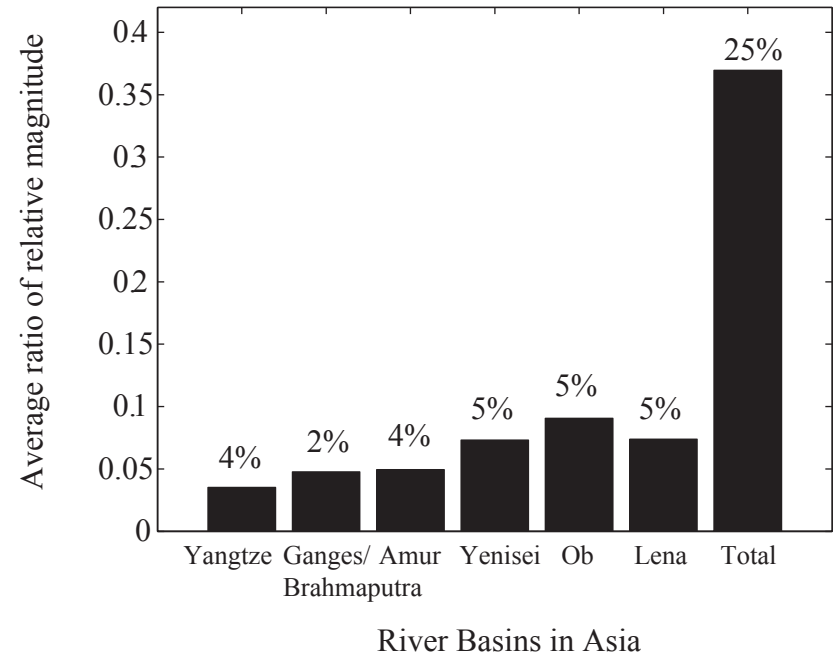
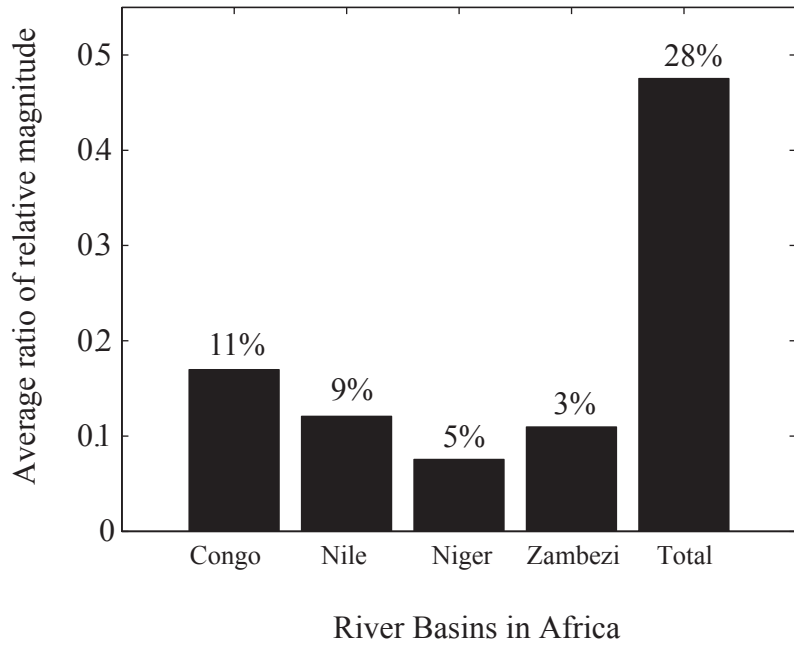
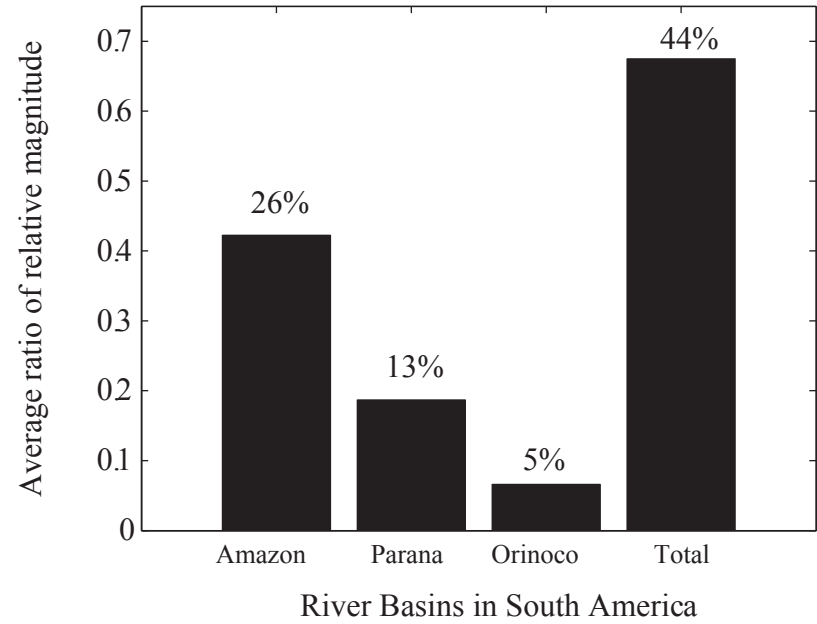
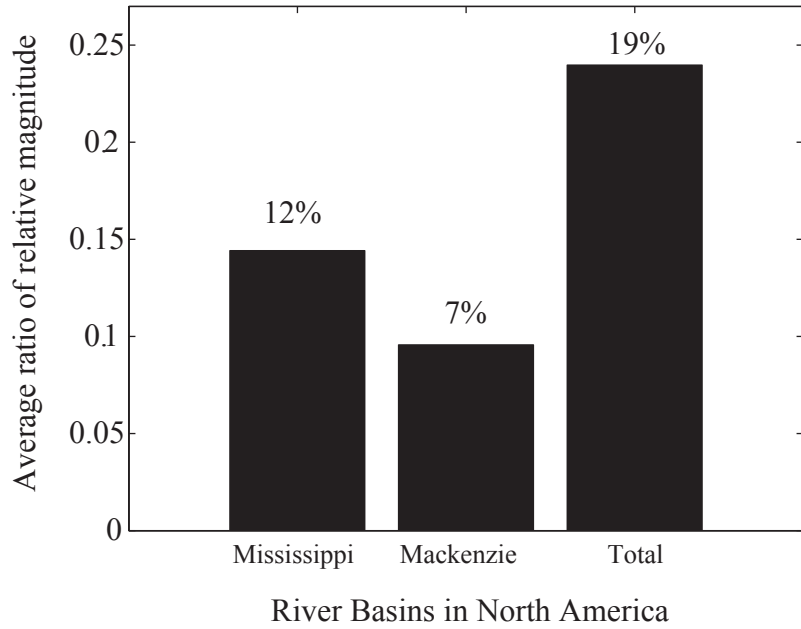
| Region | Annual Mean (cm/month) | Amplitude (cm/month) | Seasonal Mean (cm/month) | | | |
|------------------------|------------------------|----------------------|--------------------------|-------|-------|-------|
| | | | DJF | MAM | JJA | SON |
| North America | -0.06 | 0.50 | 0.73 | -0.18 | -0.34 | -0.18 |
| Mississippi | 0.30 | 1.33 | 1.33 | -0.57 | -0.06 | 0.28 |
| Mackenzie | -0.23 | 1.32 | 1.87 | -0.46 | -1.17 | -0.89 |
| South America | 0.30 | 4.10 | 2.44 | 0.87 | -1.45 | -1.43 |
| Amazon | 0.55 | 7.60 | 3.85 | 1.86 | -2.78 | -2.69 |
| Parana | 0.11 | 3.41 | 2.75 | -1.03 | -0.26 | -0.83 |
| Orinoco | 0.66 | 3.27 | -2.10 | 3.23 | 2.80 | -2.73 |
| Asia | 0.08 | 0.60 | 0.16 | 0.31 | -0.16 | -0.10 |
| Yangtze | 0.20 | 2.69 | -1.44 | 0.34 | 2.44 | -1.45 |
| Ganges/ Brahmaputra | 0.13 | 5.80 | -1.85 | -1.30 | 4.65 | -1.02 |
| Amur | 0.26 | 0.46 | 0.28 | 0.39 | 0.05 | 0.07 |
| Yenisei | -0.11 | 3.06 | 1.31 | 0.62 | -2.50 | 0.42 |
| Ob | 0.06 | 4.21 | 1.45 | 1.41 | -2.81 | 0.04 |
| Lena | 0.07 | 1.87 | 0.66 | 0.86 | -1.93 | 0.75 |
| Africa | -0.02 | 0.60 | 0.05 | -0.21 | -0.33 | 0.22 |
| Congo | -0.28 | 1.95 | 0.36 | -0.72 | -2.57 | 0.96 |
| Nile | -0.12 | 1.90 | -0.32 | -1.03 | 0.99 | -0.26 |
| Niger | 0.05 | 4.04 | -1.92 | -0.18 | 2.04 | 0.64 |
| Zambezi | -0.01 | 5.18 | 3.20 | 0.42 | -3.40 | -0.49 |
| Europe | 0.32 | 3.67 | 1.81 | -0.17 | -1.20 | 0.59 |
| Volga | 0.50 | 4.93 | 2.27 | -0.11 | -1.42 | 0.69 |
| Danube | 0.31 | 4.34 | 2.87 | -0.37 | -2.01 | 0.79 |
| Dniepr | 0.58 | 5.28 | 3.17 | -0.76 | -1.16 | 0.64 |
| Don | 0.46 | 4.84 | 2.90 | -0.32 | -1.36 | 0.56 |
| Australia | -0.14 | 2.50 | 1.67 | 0.13 | -1.89 | -0.41 |
| Murray | -0.11 | 1.90 | 1.01 | 0.14 | -1.34 | -0.26 |

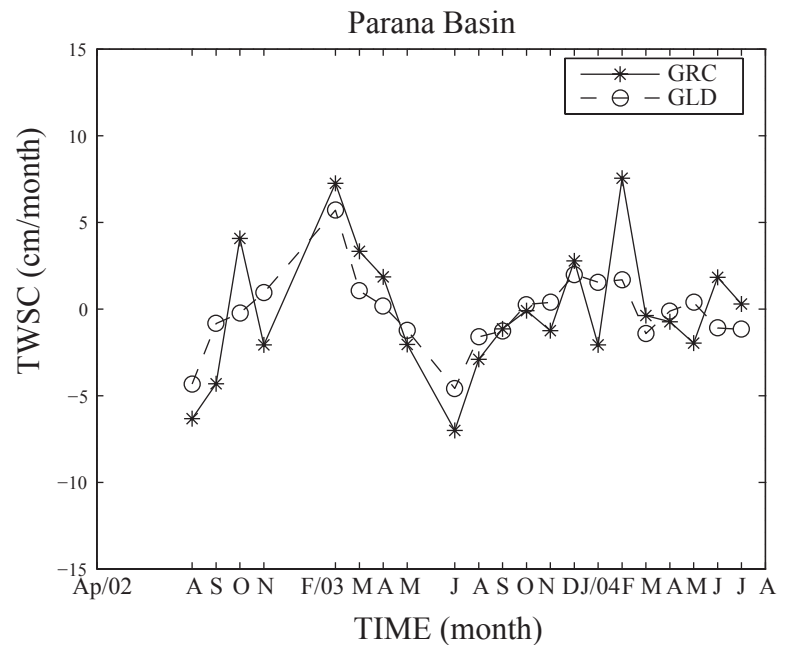
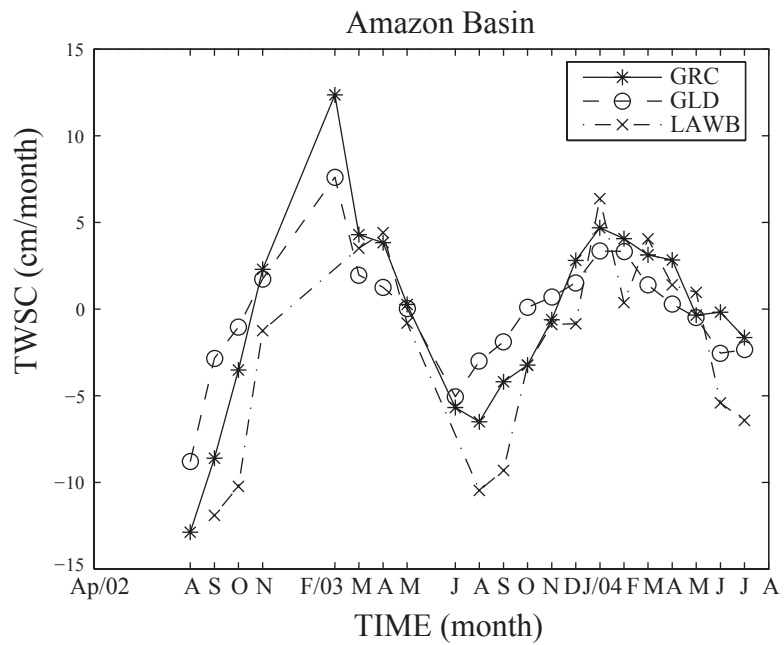
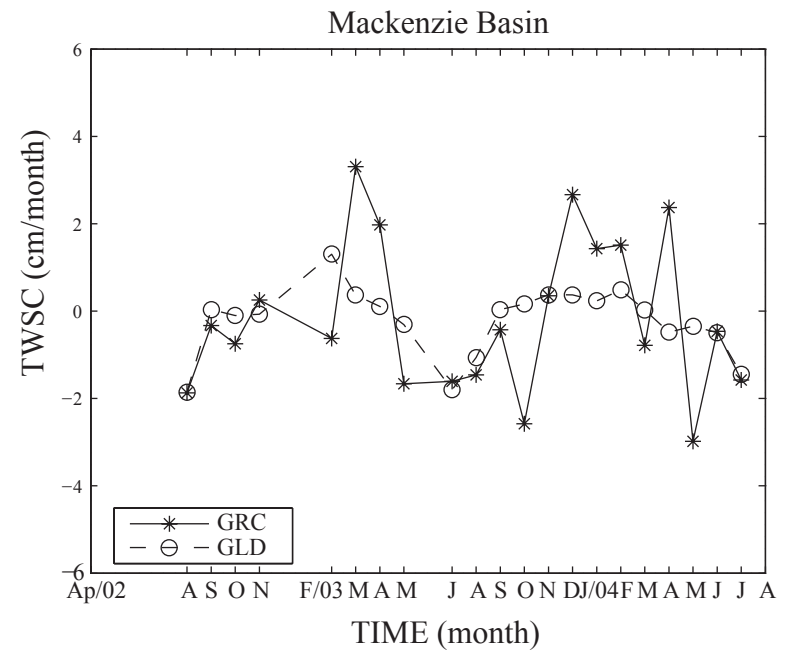
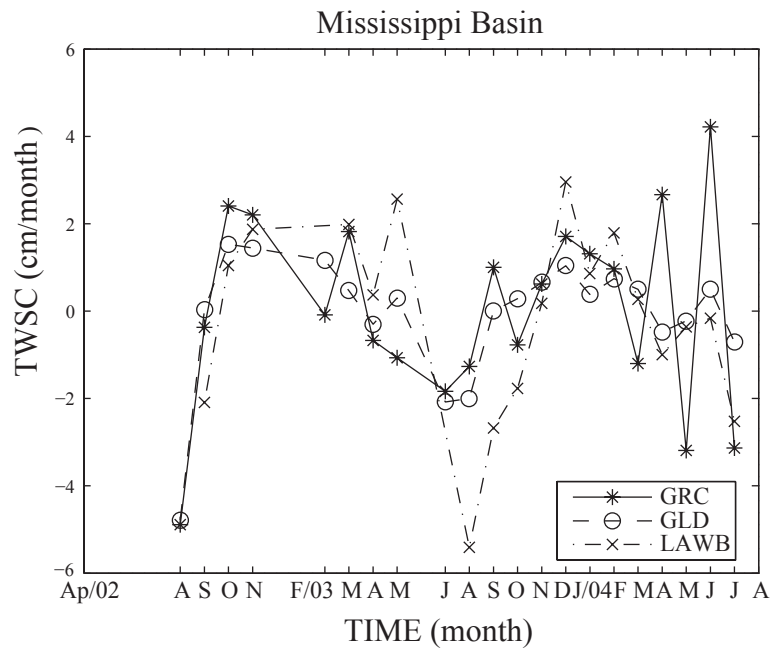
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5

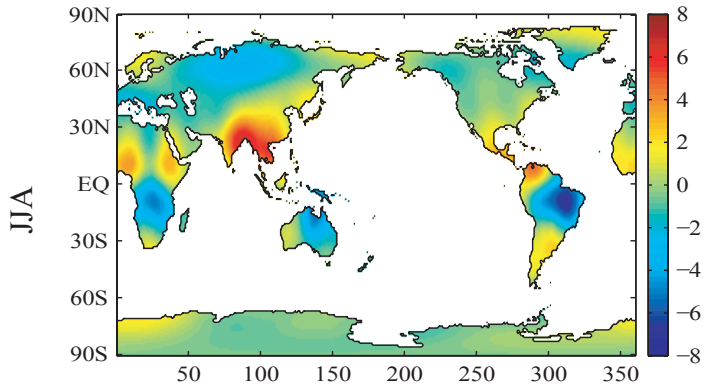




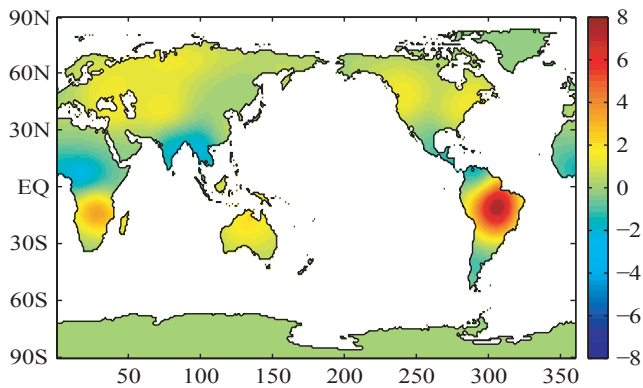
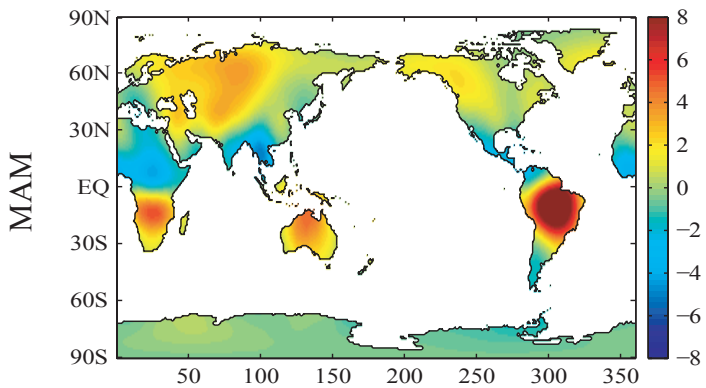
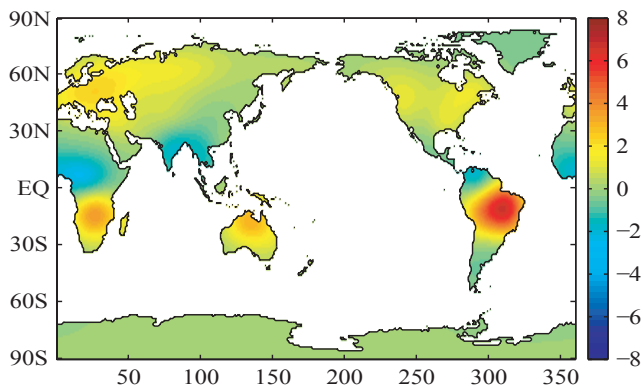
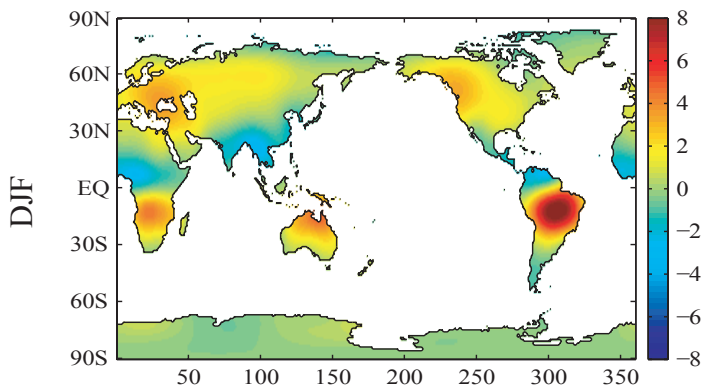
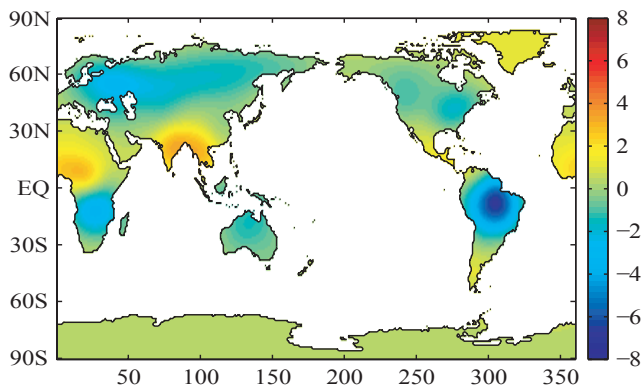
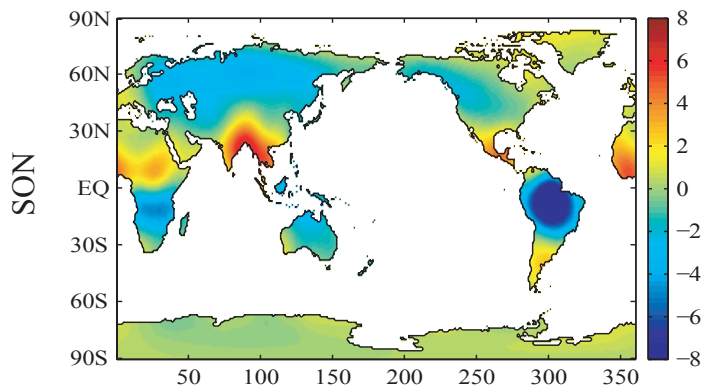
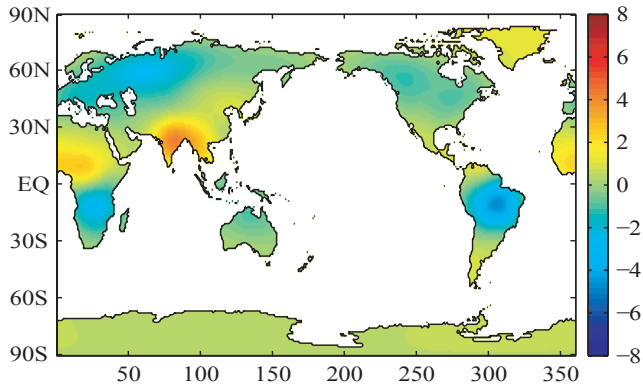


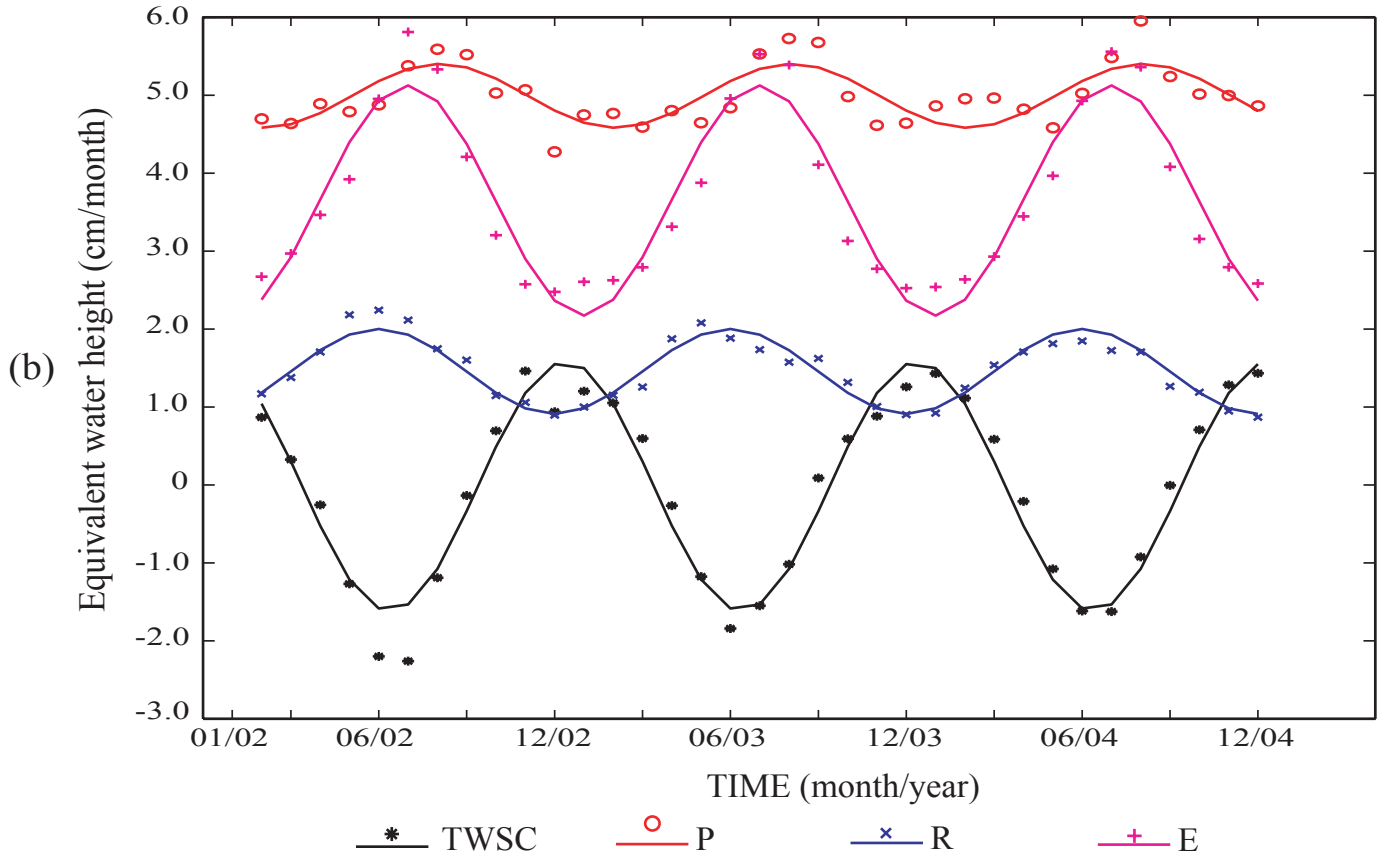
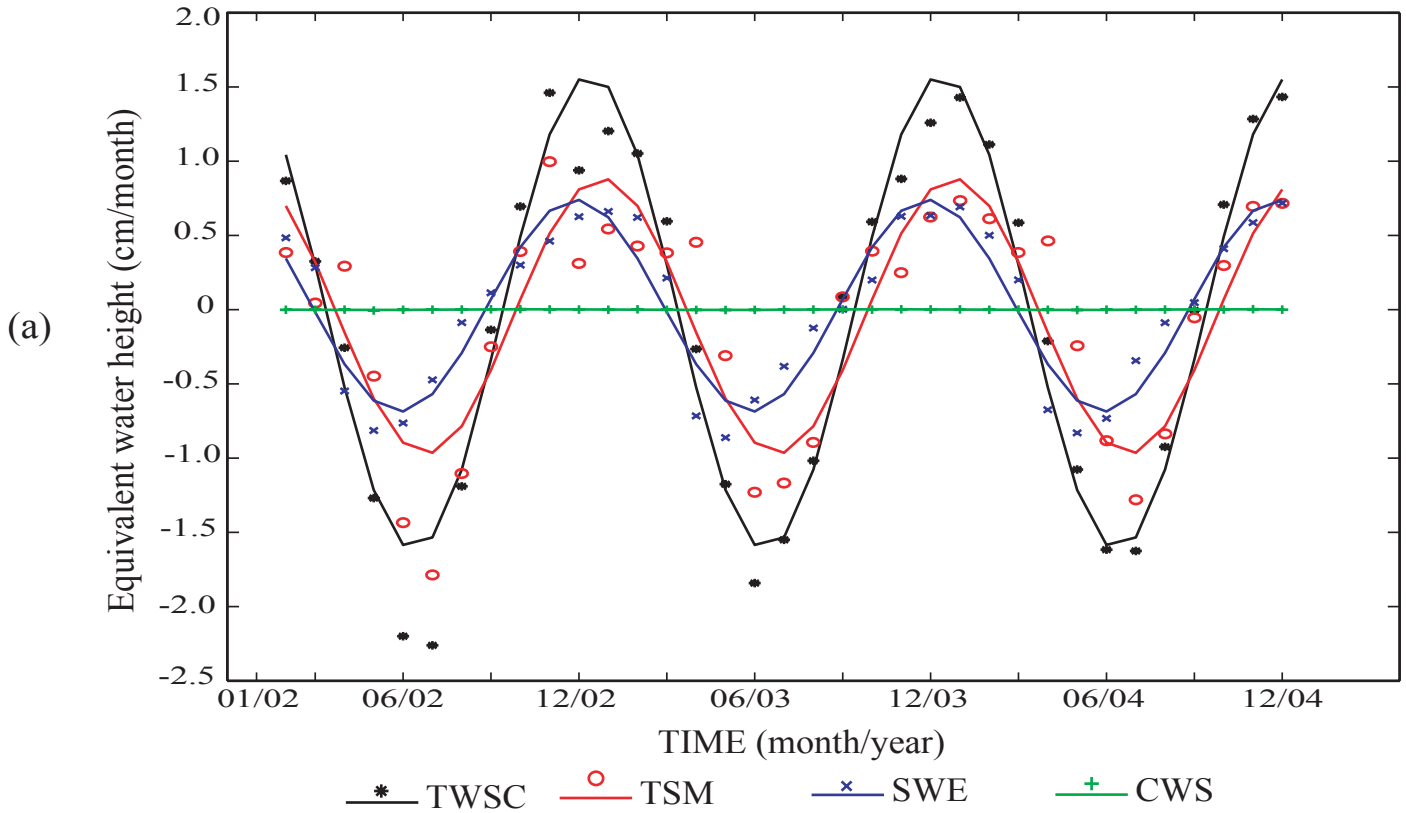


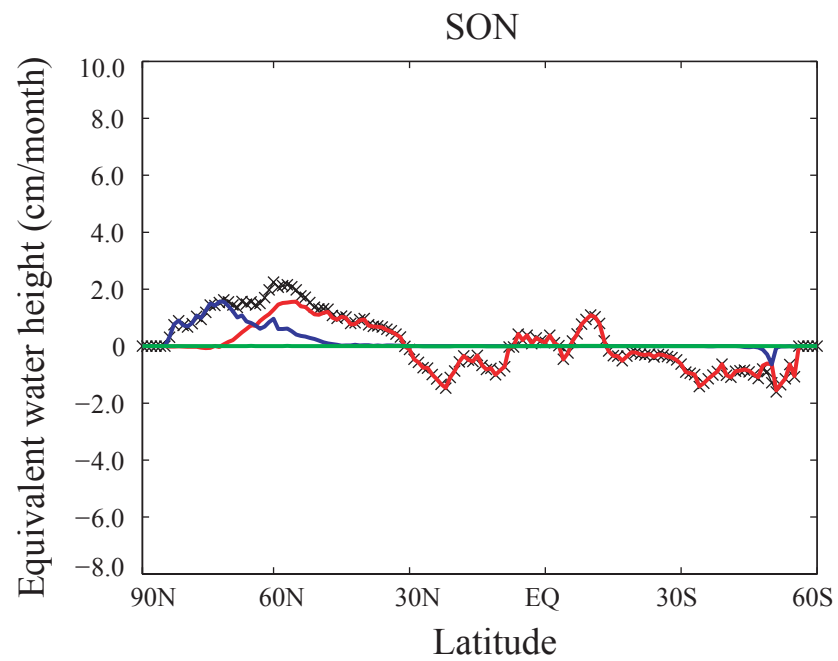
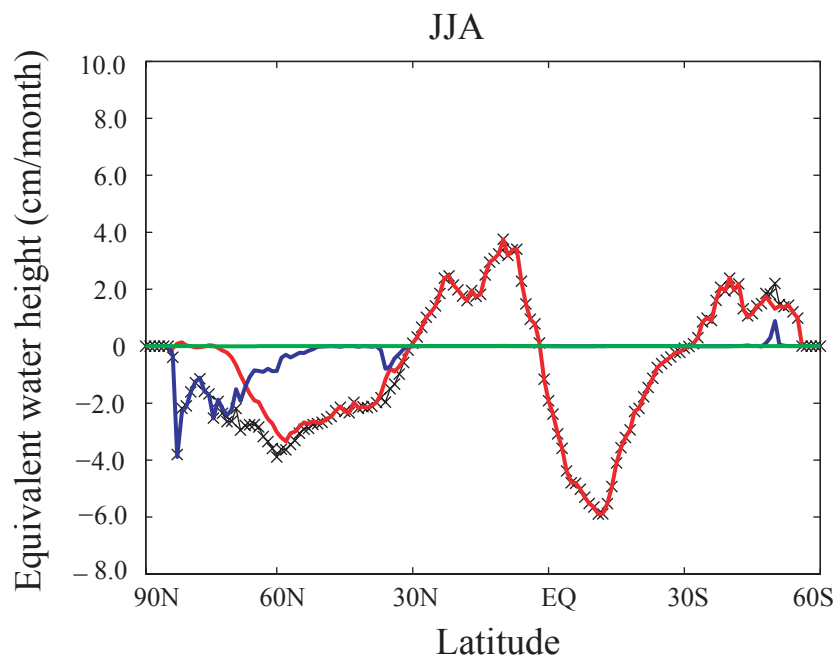
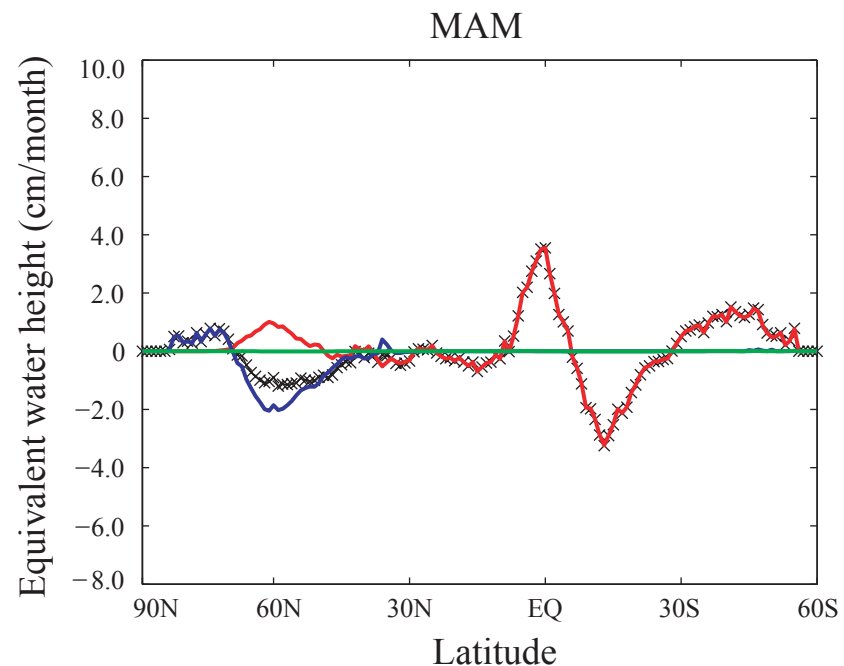
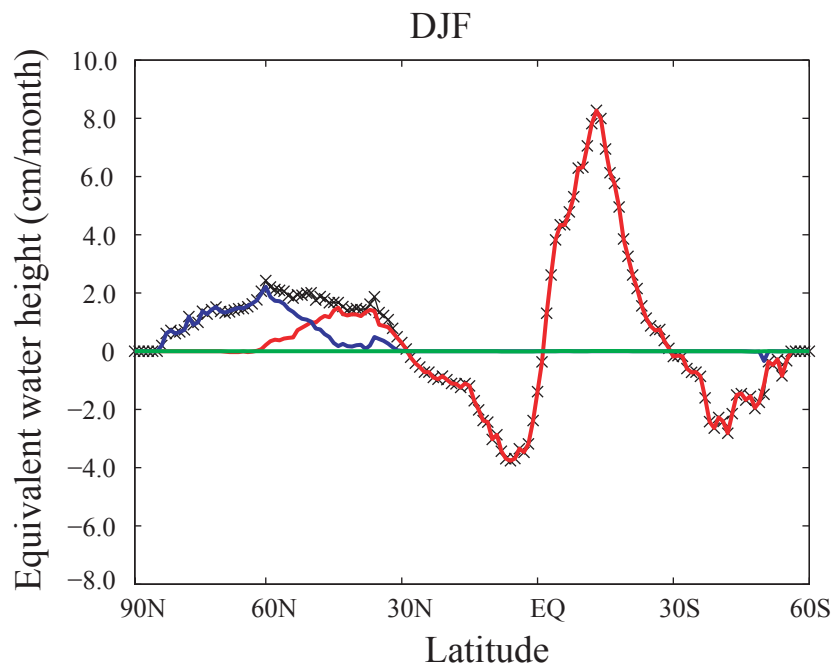
GRACE



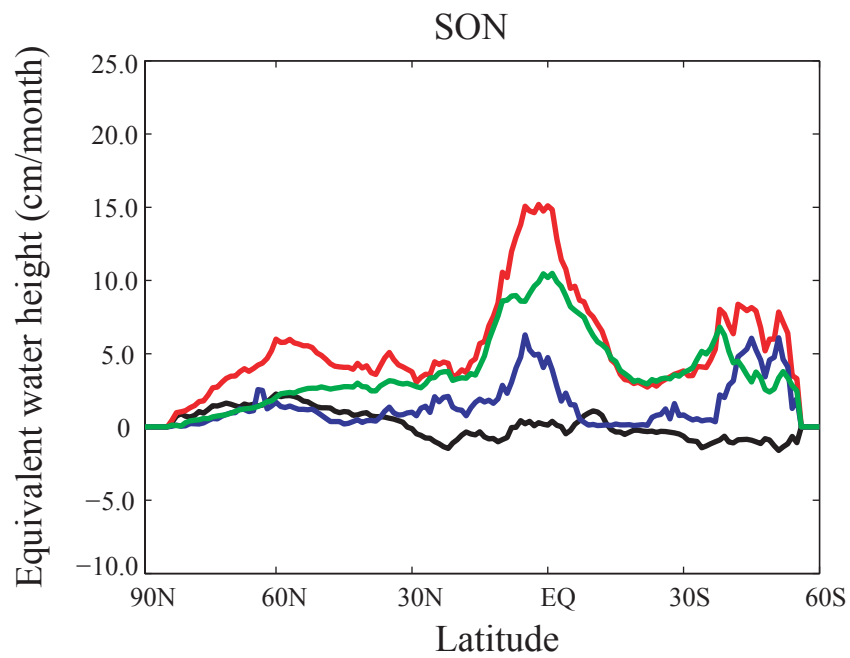
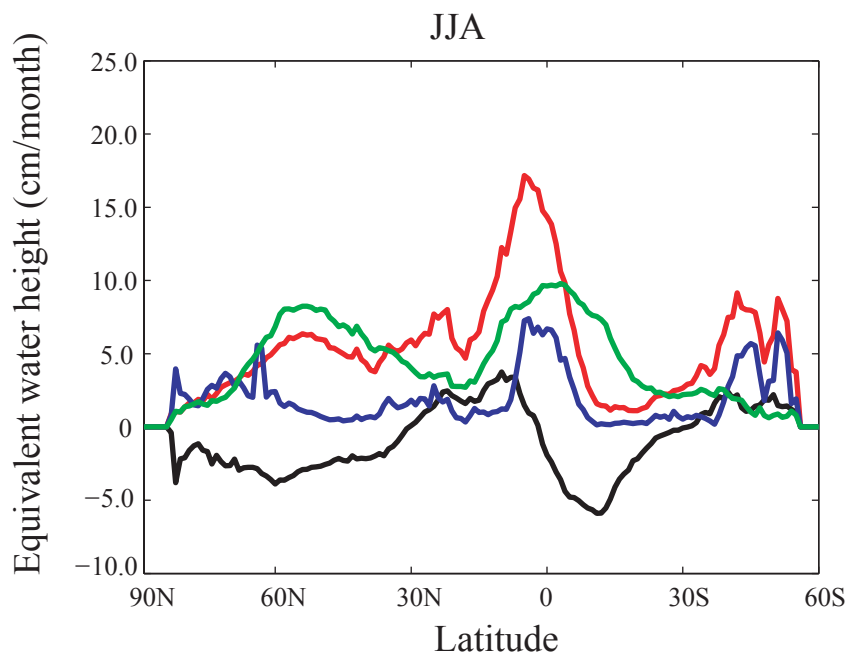
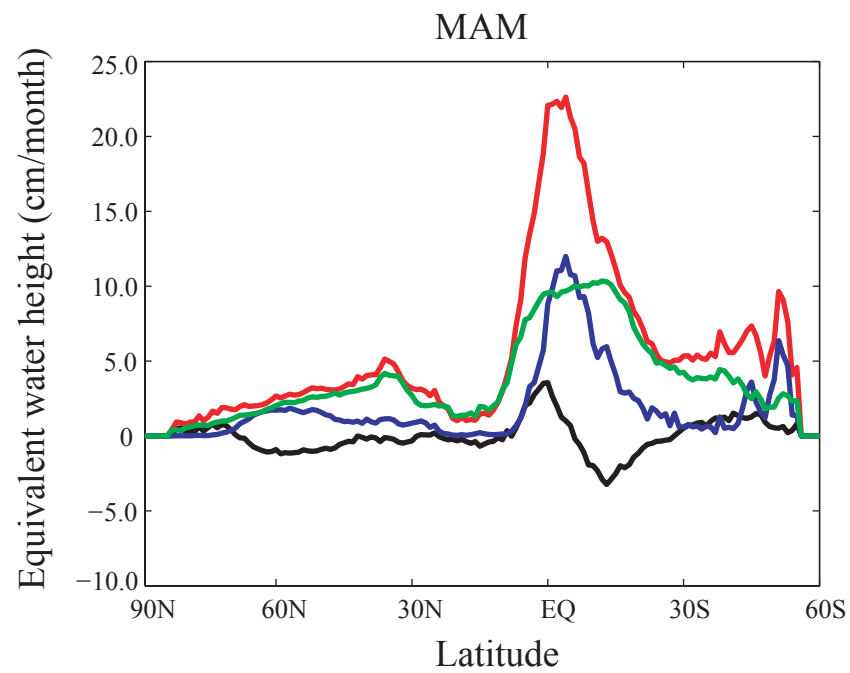
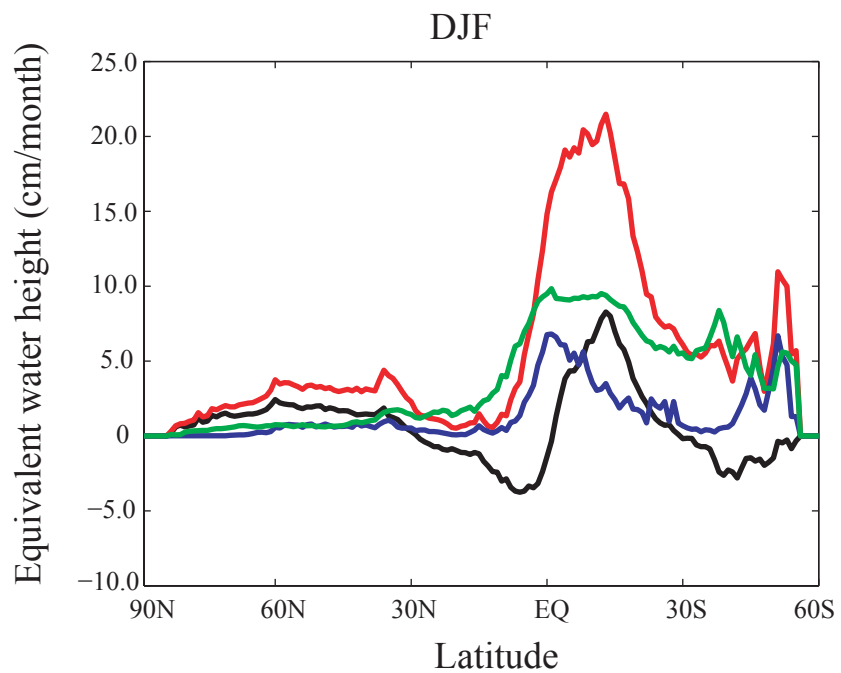
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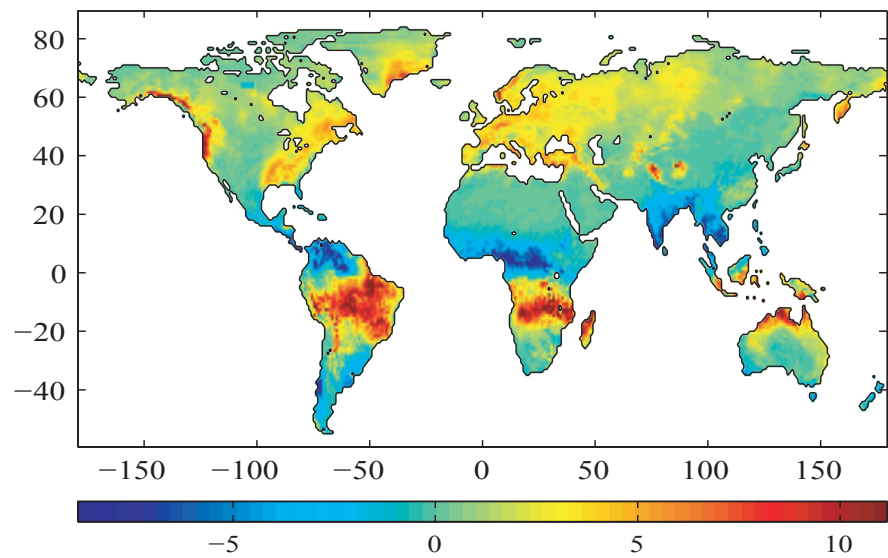


TWSC
 TSM
 SWE
 CWS

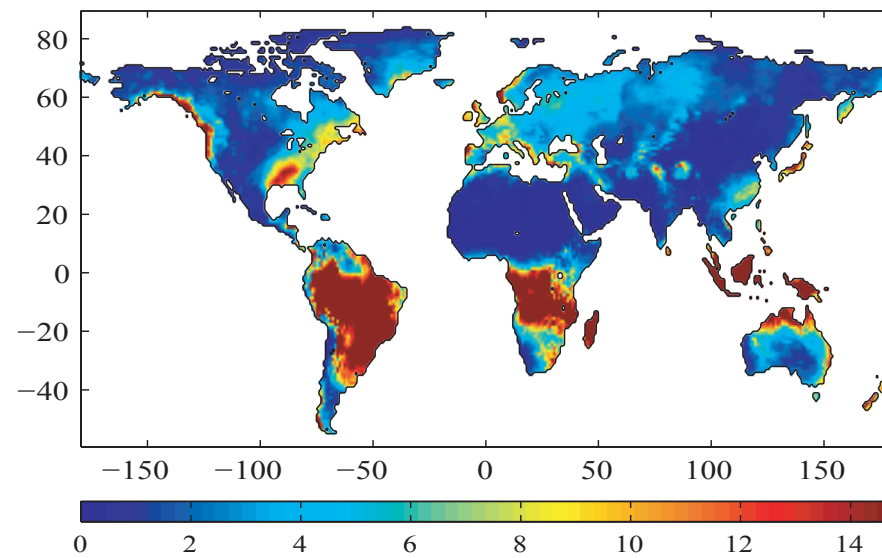


— TWSC — P — R — E

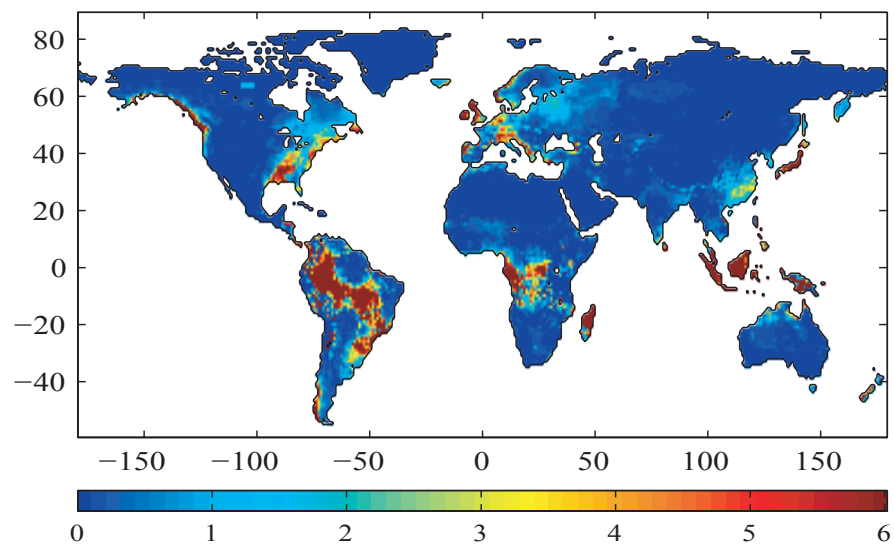
(a)



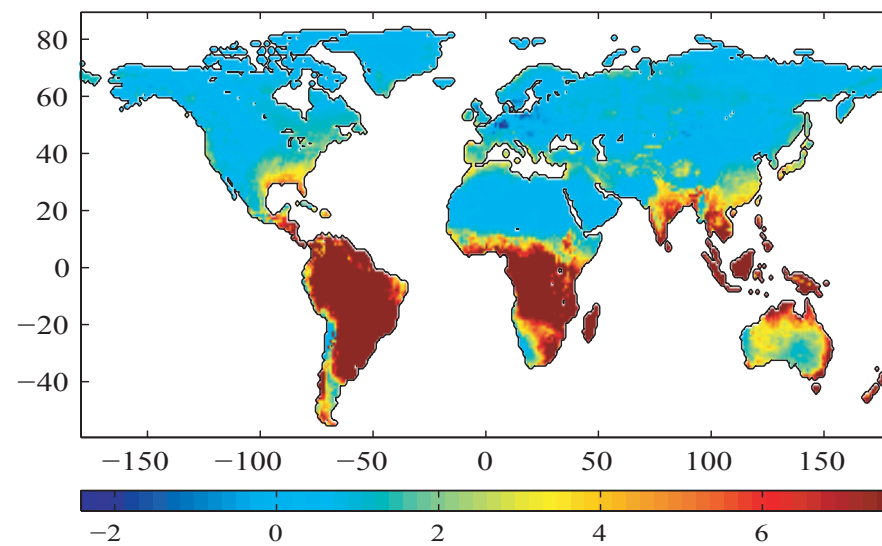
(b)



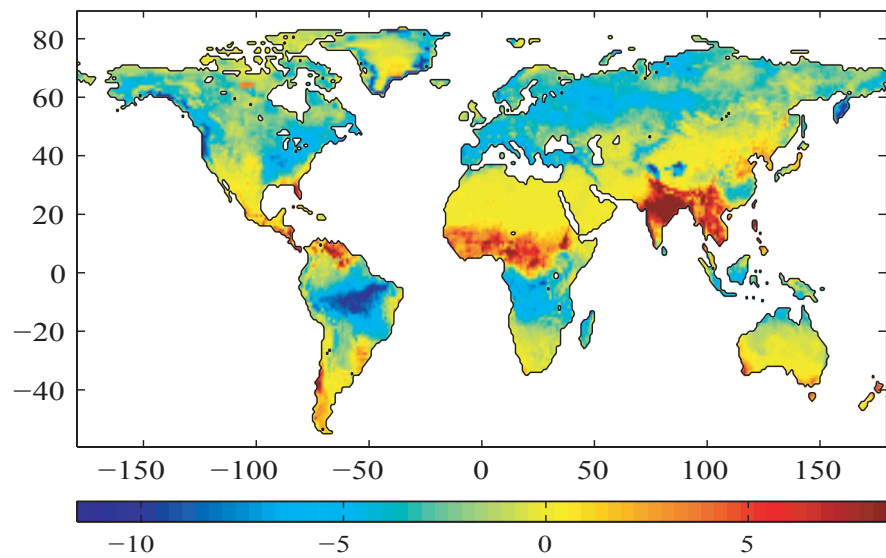
(c)



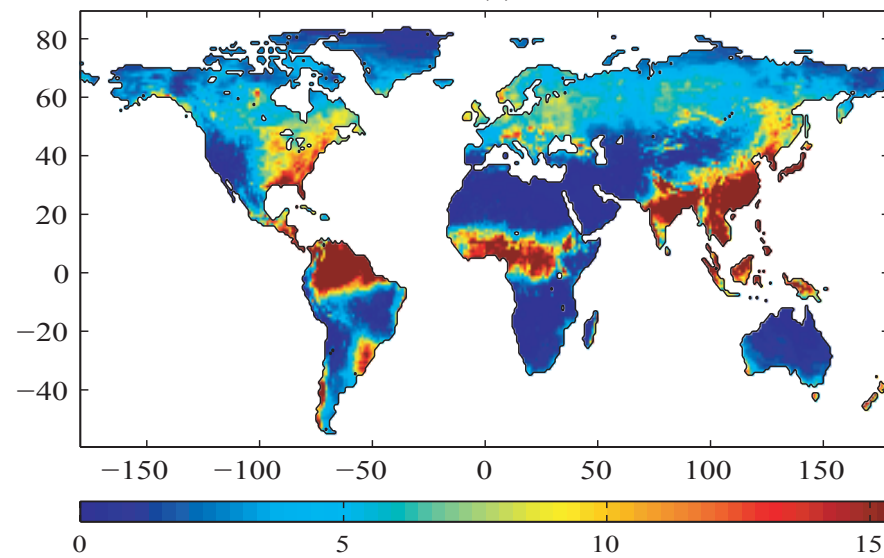
(d)



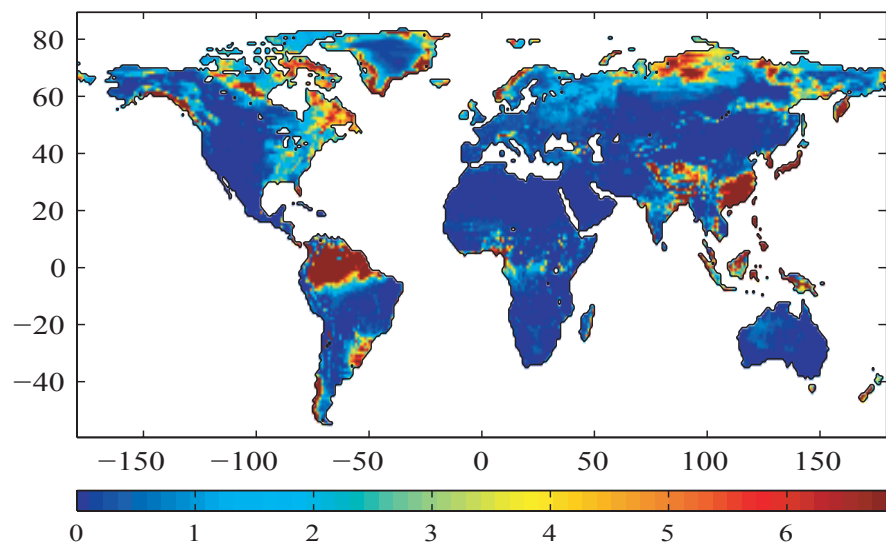
(a)



(b)



(c)



(d)

