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Practical 4: Regional analysis of gridded total water storage data

Purpose of the practical: In this practical a GRACE time series of gridded total water storage anomalies (TWSA) will be analyzed for regional (geophysical) signals. First, hydrological signals in selected river catchments will be evaluated. Then, ice mass changes of the Greenland ice sheet will be computed followed by an investigation of an unusual signal in West Malaysia. Furthermore, the impact of the leakage effect will assessed. Finally, GRACE data over South Africa will be analyzed with respect to droughts.

Directories: The data sets needed for this practical can be found in the directory *data_Lab34*. The provided matlab functions can be found in the directory *functions_Lab34*. The provided python functions can be imported from the file SpringSchoolLib34.py.

Exercise 1: Investigation of TWSA in selected hydrological catchments

- 1. Load the gridded GRACE TWSA time series from practical 3 (grace_tws.mat) and the TWSA data from the GLDAS-NOAH model (noah_tws.mat). Both time series are centered with respect to the temporal mean 2003 to 2016.
- 2. Load polygons from individual hydrological catchments (i.e. area of land where precipitation collects and drains off into a common outlet): amazon.mat, brahmaputra.mat, danube.mat, volta.mat. Each file contains a matrix with two columns, the first column consisting of the longitude values of the polygon points, the second column containing the latitude values.
- 3. Compute catchment averaged time series for GRACE and GLDAS-NOAH data using the function catchmentAverage for each of the four catchments and visualize the two time series within one plot for each catchment using the function showTimeseries.
- 4. Compute trend, annual- and semiannual amplitudes of the catchment-averaged time series for GRACE and GLDAS-NOAH from 2003 to 2023 using the functions time2decimalYears and fitTrend and compare. The amplitude of the annual (semiannual) signal can be computed from corresponding sine and cosine parameters according to $A = \sqrt{p_{sin}^2 + p_{cos}^2}$.
- 5. Compute the residual signal of the Danube time series (2003 to 2023) by removing trend and seasonal signals using the approximated data provided by the function fitTrend. Visualize the residual signal using the function showTimeseries.

- 6. Interpret your results considering also the following aspects:
 - Discuss possible reasons for the different trends of GRACE and GLDAS-NOAH in the Brahmaputra catchment.
 - Discuss possible reasons for the large inter-annual variability indicated by the GRACE time series in the Volta catchment.
 - Analyze the residuals time series of the Danube catchment with respect to hydrological extremes.
 - Discuss limitations of hydrological models and GRACE observations in representing TWSA.

Exercise 2: Investigation of ice mass changes

- 1. Load polygon of Greenland (greenland.mat).
- 2. Compute a GRACE-based time series of total ice mass changes of the Greenland ice sheet in gigatons (Gt) using the function catchmentAverage and visualize it using the function showTimeseries. Identify periods with pronounced mass balance anomalies.
- 3. Compute the trend of ice mass changes in Gt/year using the function fitTrend. Why does your computed trend differ from trends published in recent papers?

Exercise 3: Investigation of a mysterious signal

- 1. Load the polygon of West Malaysia (westMalaysia.mat).
- 2. Compute a catchment averaged time series of GRACE TWSA using the function catchmentAverage and visualize it using the function (showTimeseries).
- 3. Can you explain the signal in winter 2004/2005?

Exercise 4: Dealing with the leakage effect

- 1. Load unfiltered GLDAS-NOAH TWSA (noah_unfiltered_tws.mat) additionally to DDK filtered GRACE and GLDAS-NOAH data.
- 2. Compute TWSA time series of the Rhine catchment for the three data sets (catchmentAverage) and visualize them within one plot (showTimeseries).
- 3. Compute annual amplitudes for the two GLDAS-NOAH data sets. Why do they differ?
- 4. How can we take into account the leakage effect? Implement a (simple) approach for correcting the filtered GRACE time series.
- 5. (optional) Investigate the leakage effect for other catchments. On which conditions does the magnitude of the leakage effect depend?

Exercise 5 (optional): Computation of drought severity indices

- 1. Load the polygon of South Africa (southAfrica.mat).
- 2. Compute catchment averaged time series of TWSA from GRACE and GLDAS-NOAH using the function catchmentAverage and visualize them using the function showTimeseries.
- 3. Compute the drought severity index (DSI) according to $DSI_{i,j} = \frac{TWSA_{i,j} \overline{TWSA_j}}{\sigma_j}$ from GRACE and GLDAS-NOAH, respectively. $\overline{TWSA_j}$ is the climatology mean of TWSA of month j over all given years and σ_j the corresponding climatology standard deviation *i...* year, *j...* month).
- 4. Visualize the computed DSI using the function showTimeseries.
- 5. A DSI of less than -2.0 is an exceptional drought, from -1.99 to -1.60 is an extreme drought, from -1.59 to -1.30 is a severe drought, from -1.29 to -0.80 is a moderate drought, from -0.79 to -0.50 is abnormally dry, from -0.49 to 0.49 is near normal, from 0.50 to 0.79 is slightly wet, from 0.80 to 1.29 is moderately wet, from 1.30 to 1.59 is very wet, from 1.60 to 1.99 is extremely wet, and higher than 2.0 is exceptionally wet.
 - In which years can you identify extreme droughts?
 - Do GRACE and GLDAS-NOAH agree regarding the drought classification?
 - Which advantage has a drought severity index based on GRACE and based on hydrological models, respectively?

Functions:

function [time] = time2 decimalYears (years, months)		
Convert time provided as vectors of years and months to decimal years		
Input	• years: $p \times 1$ vector of years	
	• months: $p \times 1$ vector of months	
Output	• time: $p \times 1$ vector of time in decimal years	

function [twsa] = catchmentAverage (data, lon, lat, polygon)			
Computation of catchment averaged TWSA.			
Input	• data: $n \times m \times p$ matrix containing the gridded data set (p # time steps)		
	• lon : $n \times 1$ vector containing the longitudes		
	• lat : $m \times 1$ vector containing the latitudes		
	• polygon: $b \times 2$ matrix which contains the polygon of a selected region (e.g. Amazon). The first column consists of the longitude values of the polygon points, the second column contains the latitude values.		
Output	• twsa: $p \times 1$ vector containing a time series catchment averaged TWSA [mm]		

function [param, dataApprox] = fitTrend (time, data, mode)			
Compute trend, annual and semi-annual signal			
Input	• time: $p \times 1$ vector of time in decimal years		
	• data: $p \times 1$ time series to be approximated by the estimated parameters, with p number of time steps		
	• mode: string, defining if the semiannual signal should be included in the esti- mation process (choose 'annual' or 'semiannual')		
Output	 param: #param × 1 vector of estimated parameters (bias, trend, sine and cosine component of the annual signal, sine and cosine component of the semiannual signal) dataApprox: p × 1 vector of data approximated by the estimated parameters 		

function showTimeseries (data, ylim, legendStrings, titleString, ylabelStrings)		
Visualization of an arbitrary number of time series also with different number of time steps.		
Input	• data: struct containing the time and data values in the format	
	[years1,months1,data1],[years2,months2,data2],, this allows to provide se-	
	veral time series with different number of time steps	
	• ylim: lower and upper limit of the y-axis, e.g. [-150 150]	
	• legendStrings: legend entries provided as a struct of strings, e.g. {'GRACE',	
	'NOAH'}	
	• titleString: A string describing the Catchment (e.g. "Amazon" or "Rhi-	
	ne"), used for the title of the plot.	
	• ylabelString: string with label of the y-axis, e.g. "mm"	
Output	• Plot of the time series.	
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