Pushing the limits of gravity field recovery from high-low satellite-to-satellite tracking – a combination of 10 years of data of the satellite pseudo-constellation CHAMP, GRACE and GOCE



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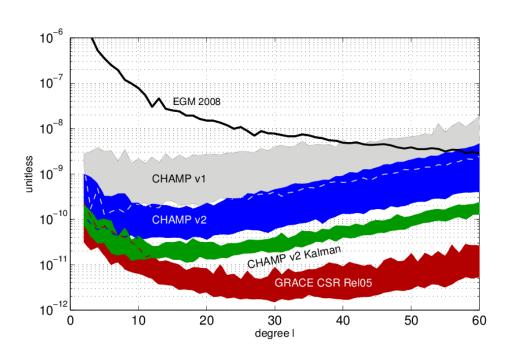


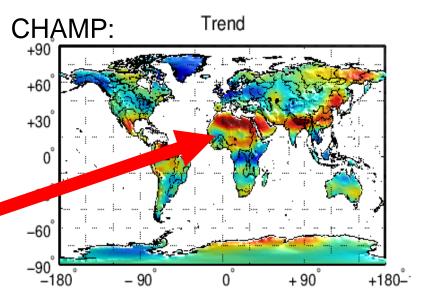
Mohamad J. Tourian, Nico Sneeuw

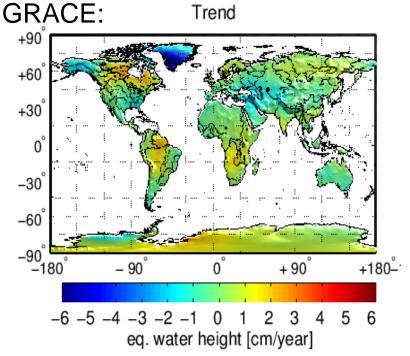


#### Recall van Dam et al. EGU 2013

- Long wavelength features can be recovered from CHAMP/hl-SST, e.g. the trend in Greenland
- Strong spatial error pattern,
  e,g. in Africa and Asia







# COMBINING CHAMP, **GRACE A/B AND GOCE**







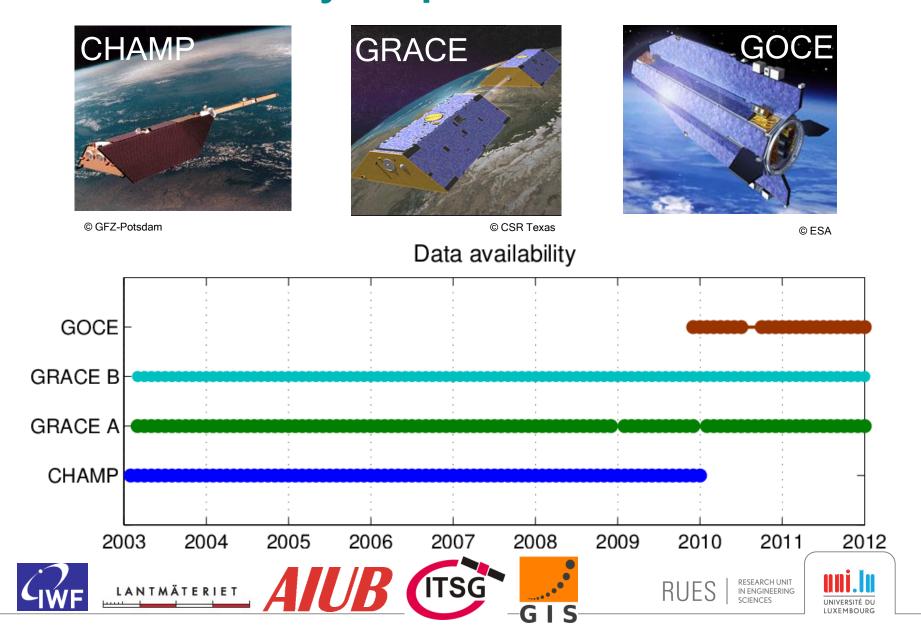








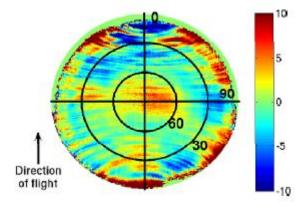
## Data availability for period 2003 to 2012



## **Data processing**

#### **GPS** positions for CHAMP:

- Prange 2010
- 10 s sampling
- empirical absolute antenna phase center model



#### Prange 2010

#### **GPS positions for GRACE A/B and GOCE:**

- Zehentner et al. 2014 (subsequent talk)
- 10 s sampling
- direct use of code and phase observations
- empirical absolute antenna phase center model

#### Approach:

- acceleration approach
- no accelerometer data used
- no regularization and no a priori model / information

Result: time series of monthly gravity field solutions for each satellite















# REFINED KALMAN-FILTER APPROACH









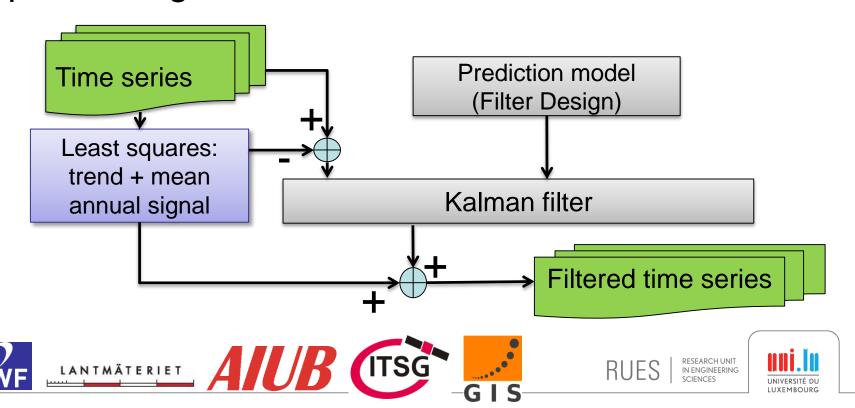






#### Kalman-Filter

- formerly using the approach of Davis et al. 2012
- changing to Kurtenbach et al. (2009)
- advantage: the process noise is implicitly defined
- processing scheme:



## Kalman-Filter: prediction model

- Kalman-Filter: concept of least-squares prediction
  - assuming a stochastic process
  - description by auto- and cross-correlation functions
    - → prediction model
- in Kurtenbach et al. (2009) correlation functions empirically derived from hydrological models
- Here: no usage of a priori information
- Instead: filter design can be converted to a correlation function
- Filter: only variations around the annual signal







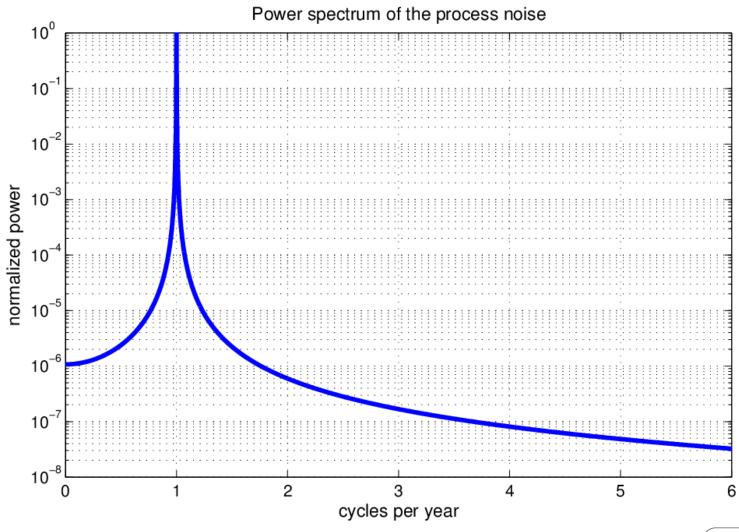








# Kalman-Filter: prediction model

















# **RESULTS**







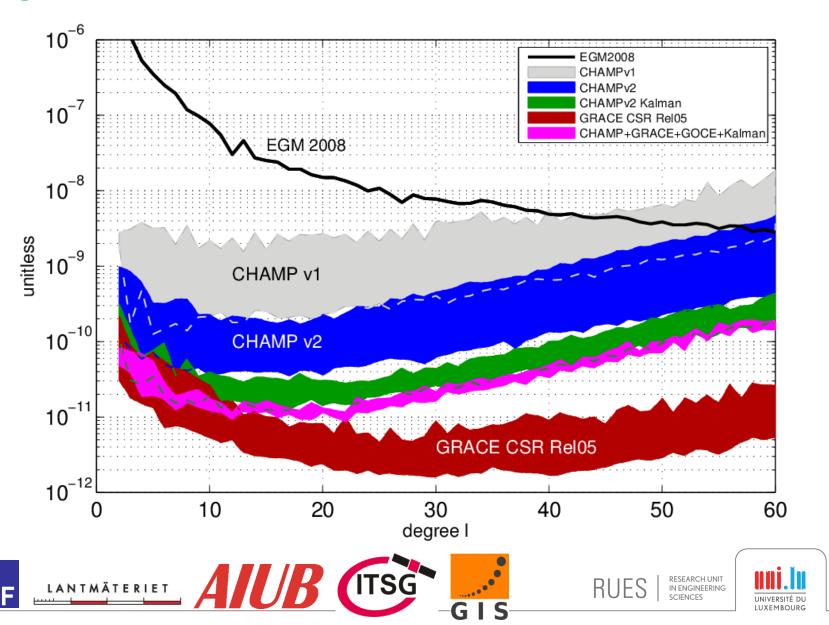




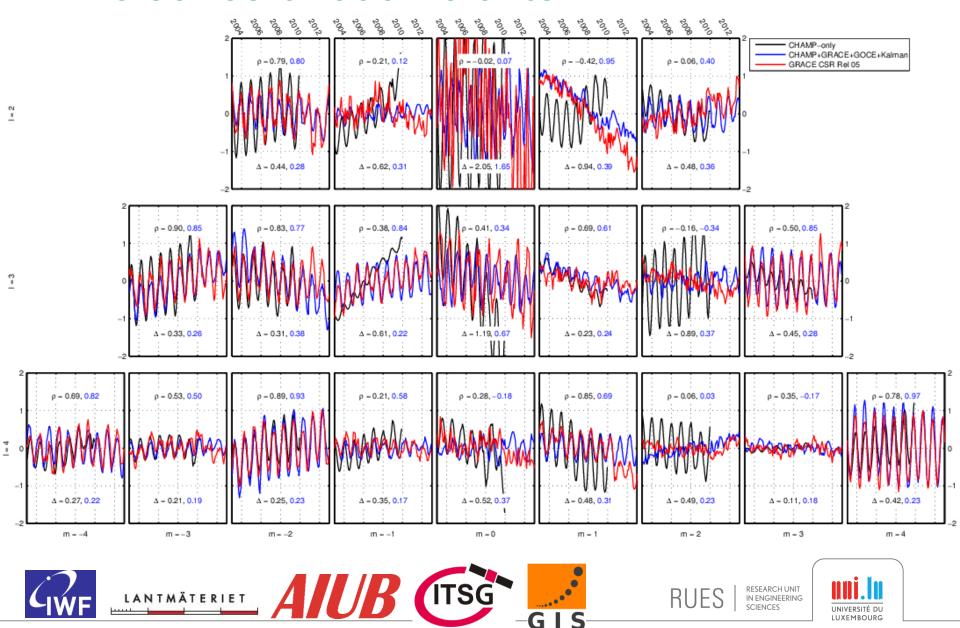




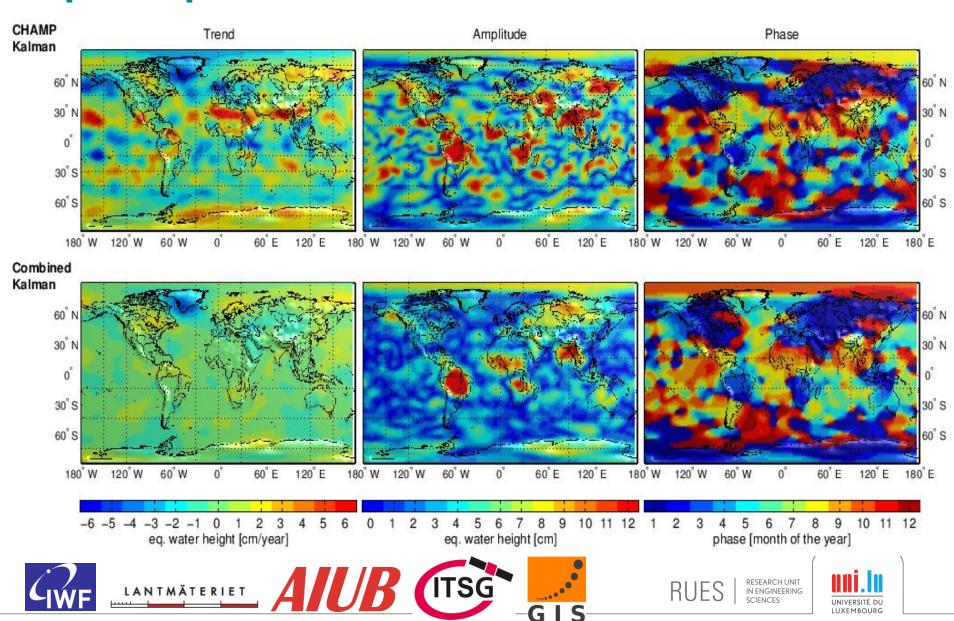
# **Degree RMS**



#### Time series of coefficients



## **Spatial pattern**



## VALIDATION AND APPLICATIONS









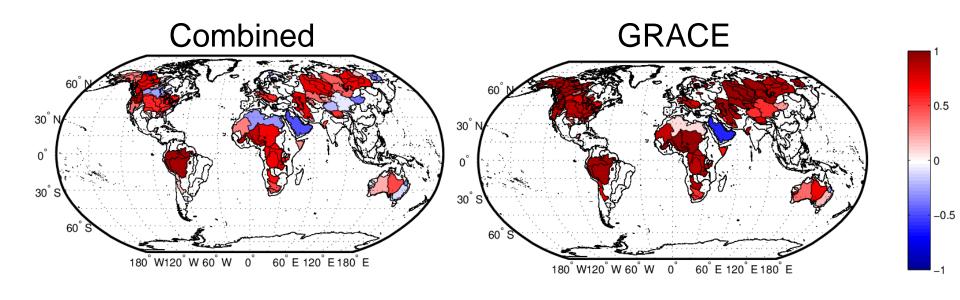






## Comparison with hydro-meteorological data

 Comparison with the difference of vertical integrated moisture flux divergences (ERA-INTERIM) and river discharge (GPCC)









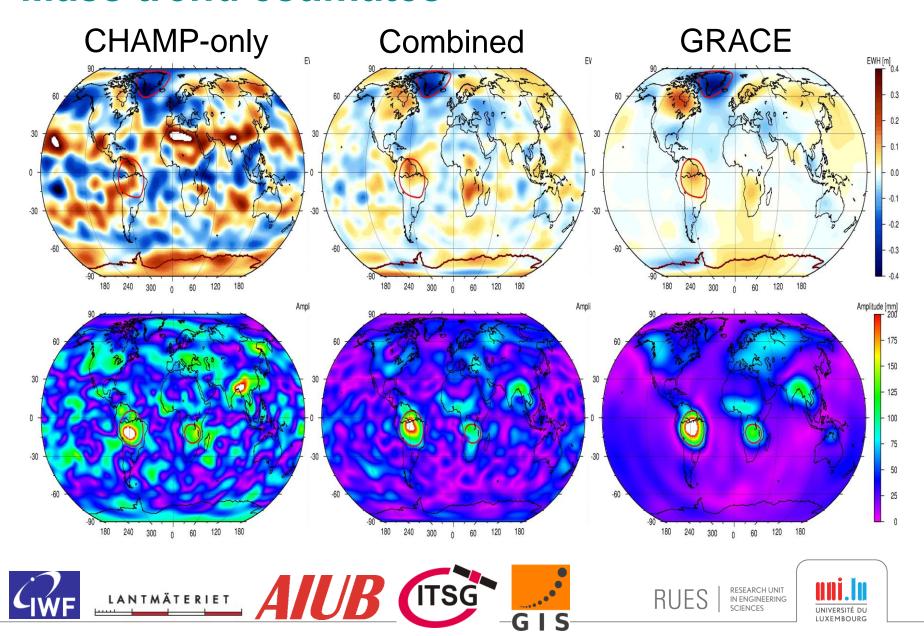








#### Mass trend estimates



## **Mass trend estimates**

Area	Filter radius	GRACE GT/yr	CHAMP- only GT/yr	∆ to GRACE in %	Combined GT/yr	∆ to GRACE in %
Greenland	1000 km	-239 ± 9	-261 ± 8	7	-208 ± 8	13
	750 km	-238 ± 7	-255 ± 7	9	-218 ± 7	8
Amazon	1000 km	90 ± 18	120 ± 9	33	95 ± 11	6
	750 km	92 ± 17	128 ± 9	39	96 ± 10	4
Antarctica	1000 km	52 ± 16	250 ± 21	481	42 ± 20	19
	750 km	50 ± 14	247 ± 20	494	39 ± 19	22









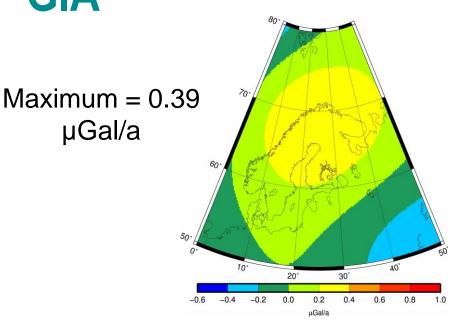


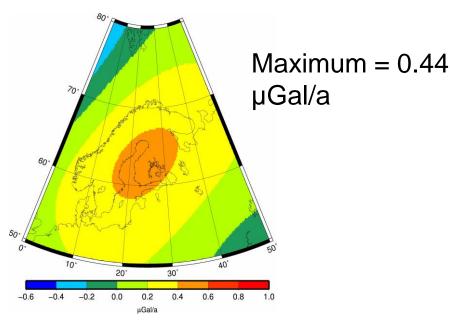




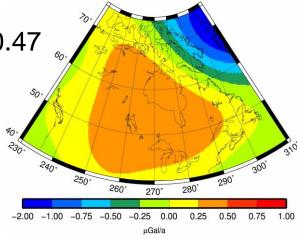
## **GIA**

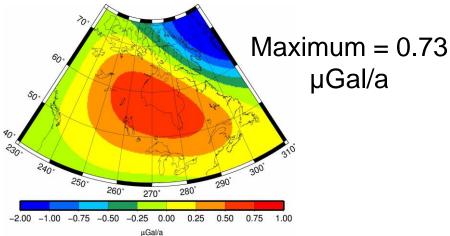
#### Combined hI-SST GRACE GFZ Rel05





Maximum = 0.47μGal/a



















#### **Conclusion:**

- Combination yields improved time-variable estimates from hl-SST
- Results agree well with GRACE, hydro-meterological data and loading from GNSS (not shown here).
- Spatial resolution improves from approximately degree 8 to 13.
- Mass estimates differ at most 22% to GRACE estimates.
- GIA estimates show first promising results but remain difficult.













