

# Assessment of experimental EGSIEM series

### Martin Horwath, TUD (presented by Uli)

### **EGSIEM Progress Meeting # 2**

University of Luxembourg January 18 – 19, 2016

















Horizon2020

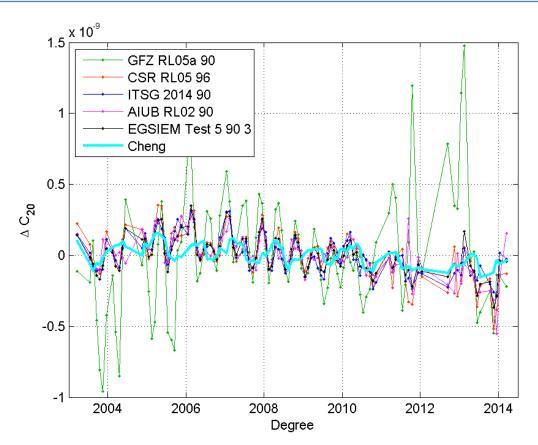
### **Comparison of the following time series:**

- CSR (96), truncated to degree 90
- GFZ RL05a (90)
- ITSG 2014 (90)
- AIUB RL2 (90)
- EGSIEM (90), test combination
  Only months that are common to all series were used.
- **1. In the Spherical harmonic domain**
- 2. In the Spatial Domain

### **3. Mass Changes in Antarctic Drainage Basins**







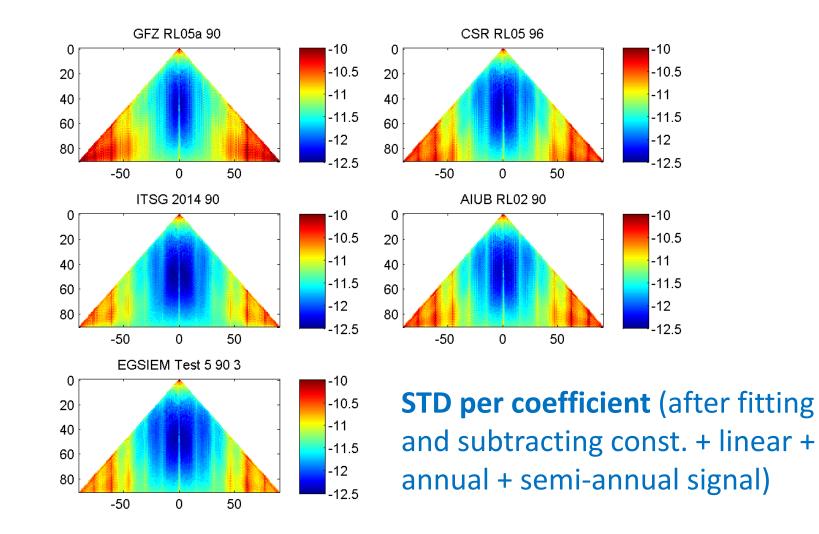
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C<sub>20</sub> is replaced by values from SLR (Cheng et al. CSR series) and degree-1 terms are added (Swenson, Chamber, Wahr series) to make the Antarctic mass balance results consistent.





# Analyses in the Spherical harmonic domain (1/3).

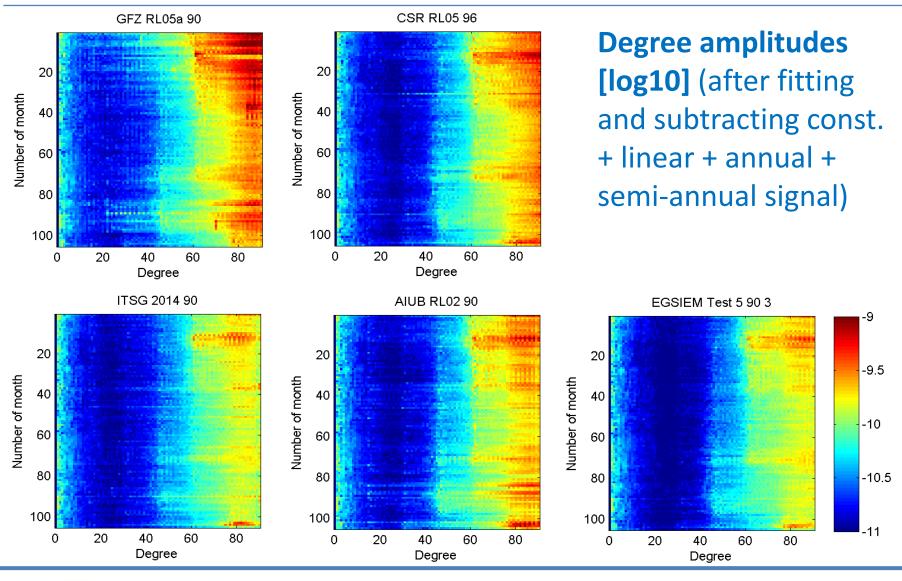


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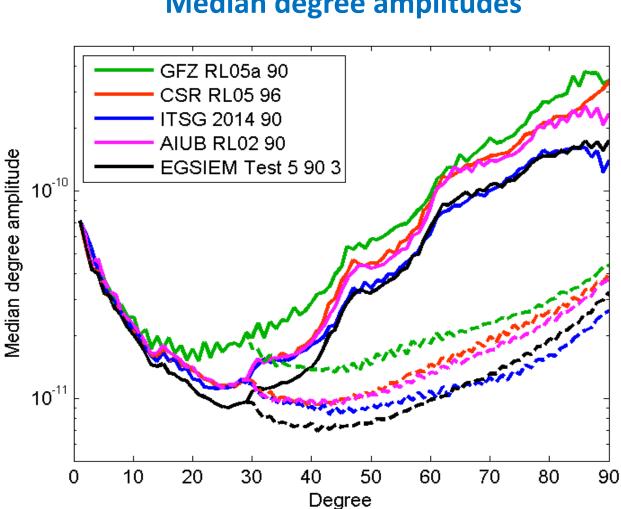
# Analyses in the Spherical harmonic domain (2/3).



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# Analyses in the Spherical harmonic domain (3/3).



Median degree amplitudes

Dashed: Median degree amplitudes calculated for orders  $m = 0 \dots 29$  only (these orders are most important for polar signals).





# Analyses in the Spherical harmonic domain: conclusion

- From looking at the ascending (error-dominated) part of the degree amplitude curves, ITSG and EGSIEM show the lowest noise levels.
- For n>60, ITSG noise level is lower than EGSIEM noise level. This is particularly pronounced for the **near-zonals**, which are most important for polar signals.
- For n<60, EGSIEM has the lowest level of variability. This is visible even in the very low degrees. Later we will see that this is not related to signal attenuation.



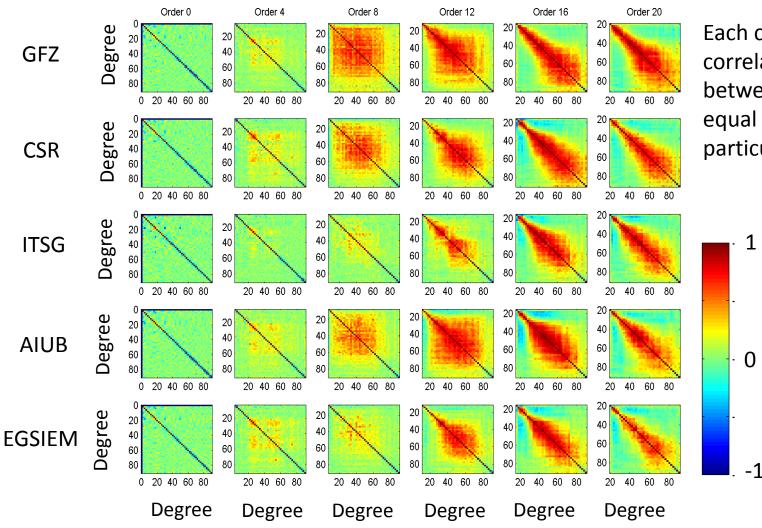


- **EWH anomalies** = EWH coefficients model (bias + trend + annual + semi-annual variation)
- **Empirical correlation matrices** between EWH anomaly coefficients of the same order and even (odd) degrees.
- Starting from order 8 or so we see the typical "**striping**" correlations.
- **ITSG** series shows weaker correlations than the other series.





# **Empirical correlations (1/2)**

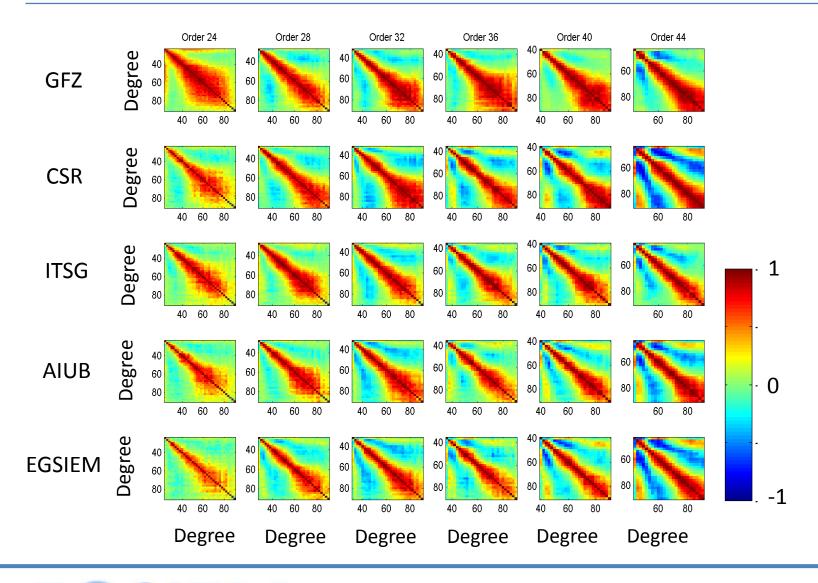


Each column shows correlation matrices between degrees of equal parity for a particular order.

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# **Empirical correlations (2/2)**



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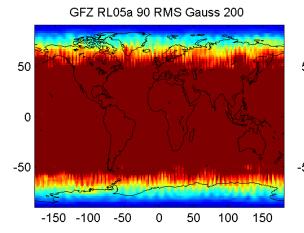
- 1. Fit and removal of Bias + Trend + annual + semi-annual signal,
- 2. filtering in the spectral domain: destriping, 200/400 km Gauss,
- 3. monthly maps of EWH-anomalies,
- 4. standard deviation or median of absolute temporal variability.

Noise is assessed over regions of low signal: oceans, inner Antarctica.

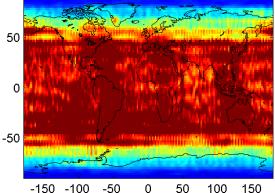




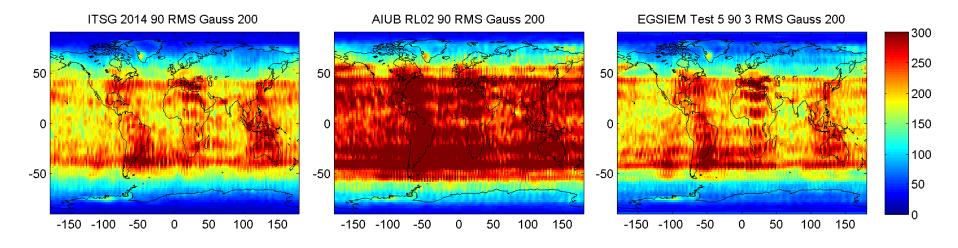
# Analyses in the spatial domain (1/12)



CSR RL05 96 RMS Gauss 200



### RMS of EWH variability: 200 km Gaussian filtering



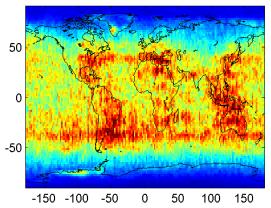
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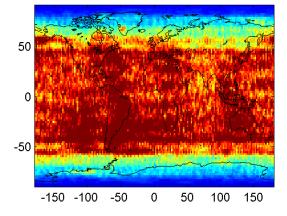
# Analyses in the spatial domain (2/12)

GFZ RL05a 90 Median Gauss 200

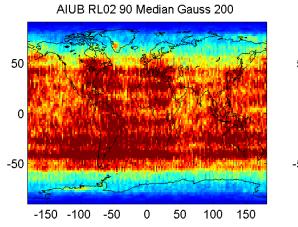
ITSG 2014 90 Median Gauss 200

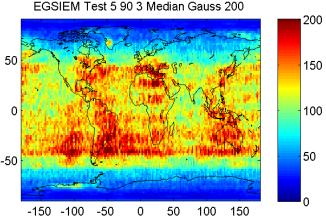


CSR RL05 96 Median Gauss 200



### Median of EWH variability: 200 km Gaussian filtering



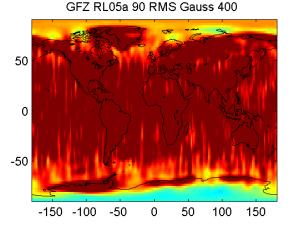


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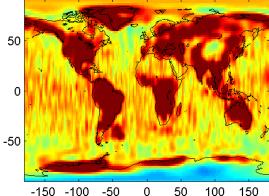
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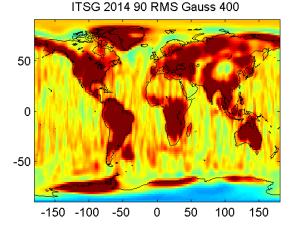
# Analyses in the spatial domain (3/12)



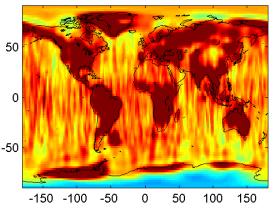
CSR RL05 96 RMS Gauss 400



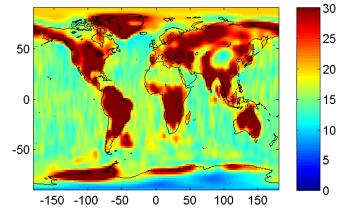
### RMS of EWH variability: 400 km Gaussian filtering



AIUB RL02 90 RMS Gauss 400



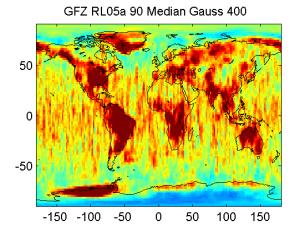
EGSIEM Test 5 90 3 RMS Gauss 400



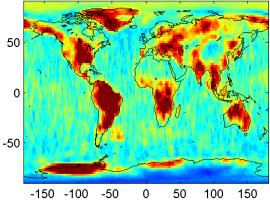
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# Analyses in the spatial domain (4/12)

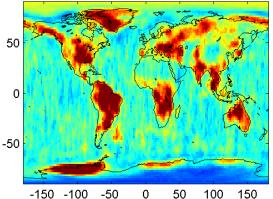


CSR RL05 96 Median Gauss 400

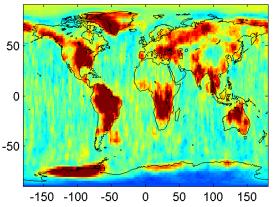


### Median of EWH variability: 400 km Gaussian filtering

ITSG 2014 90 Median Gauss 400



AIUB RL02 90 Median Gauss 400

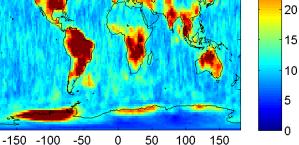


EGSIEM Test 5 90 3 Median Gauss 400

50

n

-50



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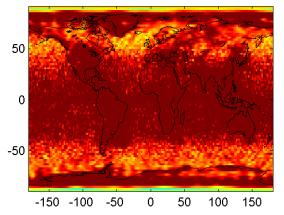
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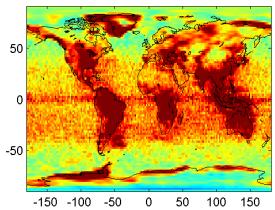
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## Analyses in the spatial domain (5/12)

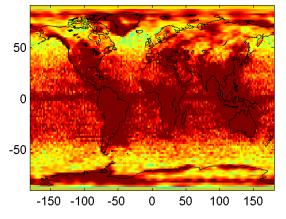
#### GFZ RL05a 90 RMS Swenson+Gauss 200



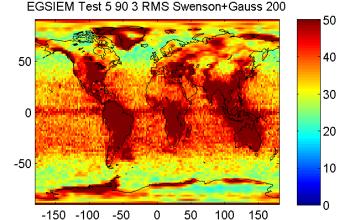
ITSG 2014 90 RMS Swenson+Gauss 200



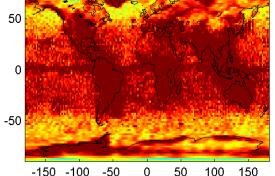
CSR RL05 96 RMS Swenson+Gauss 200



RMS of EWH variability: destriping + 200 km Gaussian filtering



AIUB RL02 90 RMS Swenson+Gauss 200



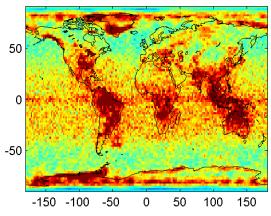


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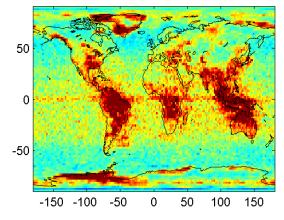
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# Analyses in the spatial domain (6/12)

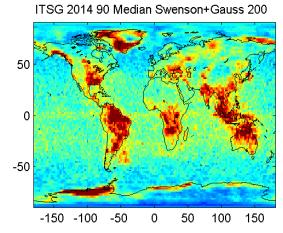
GFZ RL05a 90 Median Swenson+Gauss 200



CSR RL05 96 Median Swenson+Gauss 200

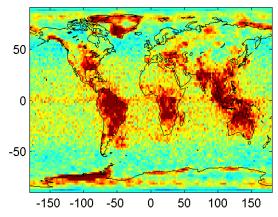


Median of EWH variability: destriping + 200 km Gaussian filtering



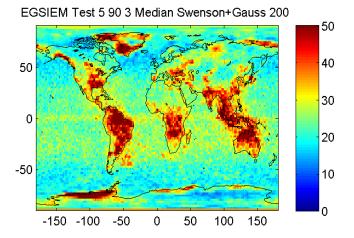
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AIUB RL02 90 Median Swenson+Gauss 200



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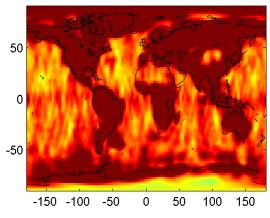
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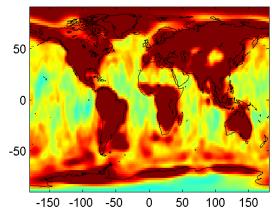
### \*\*\*\* \*\*\*\* Horizon2020

# Analyses in the spatial domain (7/12)

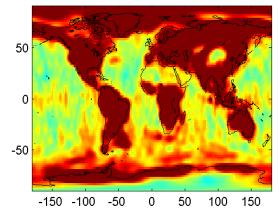
### GFZ RL05a 90 RMS Swenson+Gauss 400



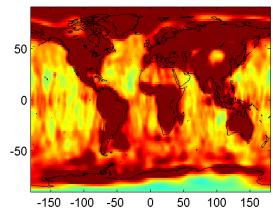
ITSG 2014 90 RMS Swenson+Gauss 400



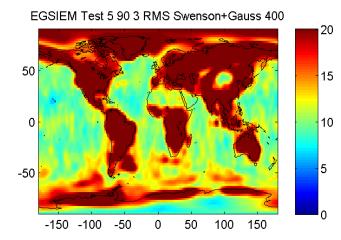
CSR RL05 96 RMS Swenson+Gauss 400



AIUB RL02 90 RMS Swenson+Gauss 400



RMS of EWH variability: destriping + 400 km Gaussian filtering



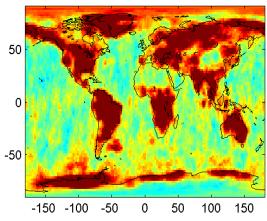


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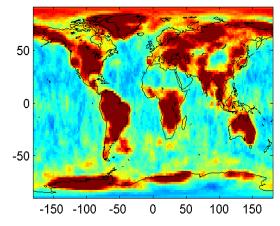
# Analyses in the spatial domain (8/12)

### GFZ RL05a 90 Median Swenson+Gauss 400

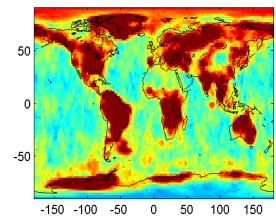


ITSG 2014 90 Median Swenson+Gauss 400

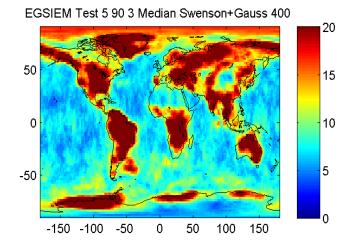
50 0 -50 -150 -100 -50 0 50 100 150 CSR RL05 96 Median Swenson+Gauss 400



AIUB RL02 90 Median Swenson+Gauss 400



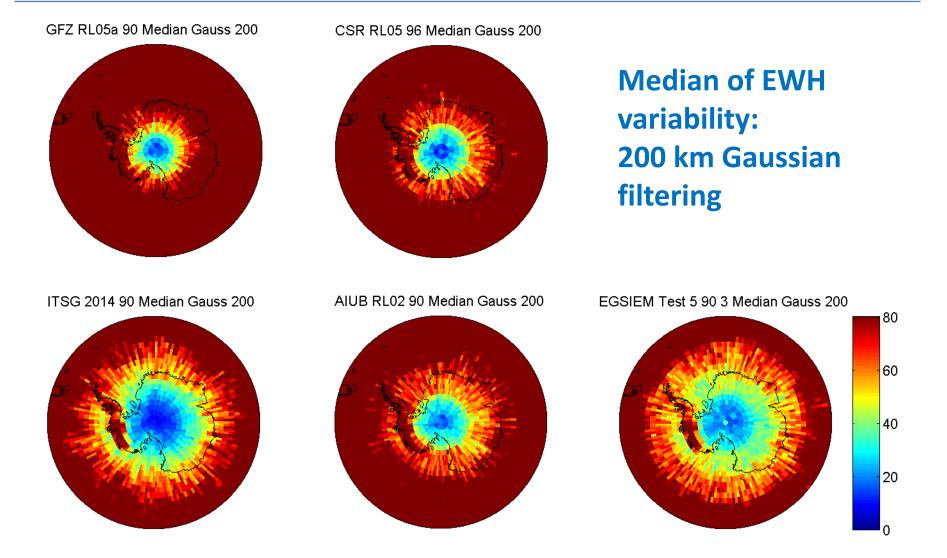
### Median of EWH variability: destriping + 400 km Gaussian filtering







# Analyses in the spatial domain (9/12)

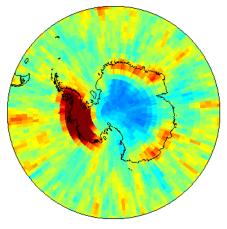


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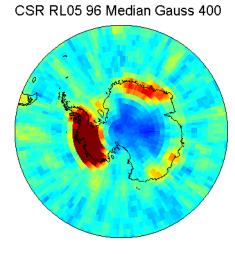


# Analyses in the spatial domain (10/12)

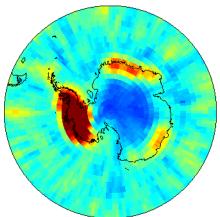
GFZ RL05a 90 Median Gauss 400

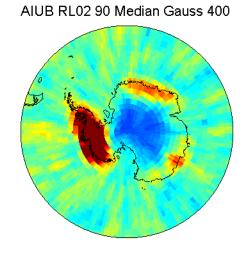


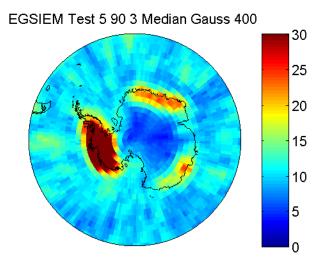
ITSG 2014 90 Median Gauss 400



Median of EWH variability: 400 km Gaussian filtering







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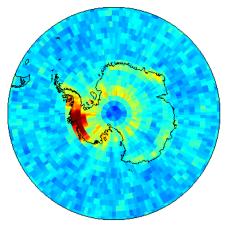
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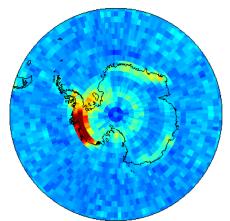


# Analyses in the spatial domain (11/12)

GFZ RL05a 90 Median Swenson+Gauss 20

CSR RL05 96 Median Swenson+Gauss 200

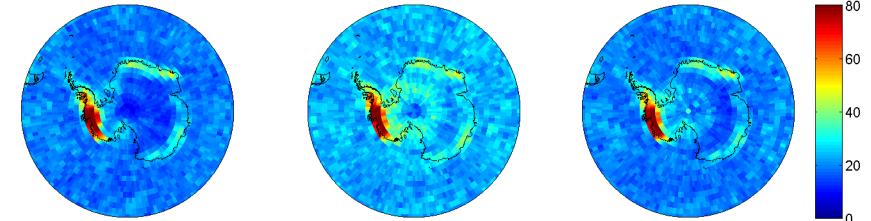




Median of EWH variability: destriping + 200 km Gaussian filtering

ITSG 2014 90 Median Swenson+Gauss 20 A

AIUB RL02 90 Median Swenson+Gauss EGSIEM Test 5 90 3 Median Swenson+Gauss 200



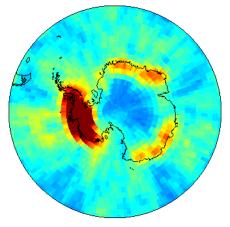


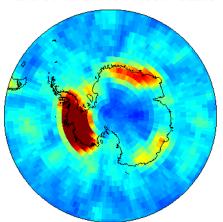


# Analyses in the spatial domain (12/12)

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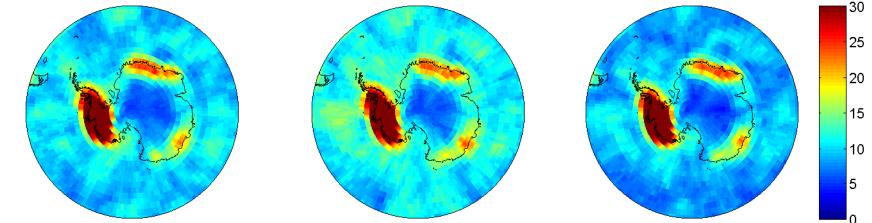
CSR RL05 96 Median Swenson+Gauss 400





Median of EWH variability: destriping + 400 km Gaussian filtering

ITSG 2014 90 Median Swenson+Gauss 40 AIUB RL02 90 Median Swenson+Gauss EGSIEM Test 5 90 3 Median Swenson+Gauss 400







# Analyses in the spatial domain: conclusion

### 200 km Gaussian filter:

- **ITSG** and **EGSIEM** show the lowest noise levels.
- **ITSG** noise level < **EGSIEM** noise level.

This is **consistent with the assessment in the spectral domain**, where **ITSG** has the lowest noise level in the high degrees

### 400 km Gaussian filter:

- EGSIEM, ITSG and CSR show the lowest noise levels.
- EGSIEM noise level < ITSG, CSR noise level.

The relative differences between releases remain the same irrespective of STD or Median and destriping or not.





**Basin masses** are computed using an **integration kernel** (in the spectral domain) which is designed to attain a compromise between leakage errors and propagated GRACE errors.

The kernel design depends on empirical GRACE error covariances that are specific for each series. To achieve comparable results kernels were derived for all series and then averaged.

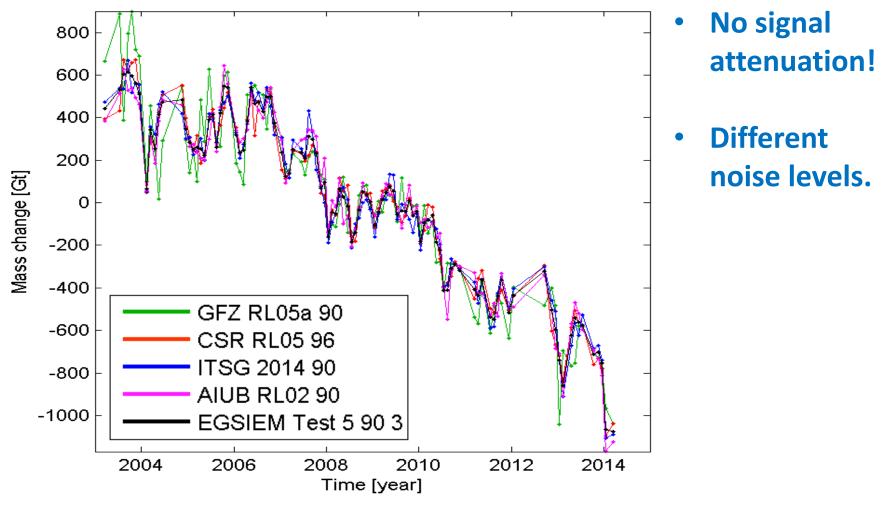
GIA models as in Shepherd et al. (2012).





# Mass changes of Antarctic drainage basins (1/4)

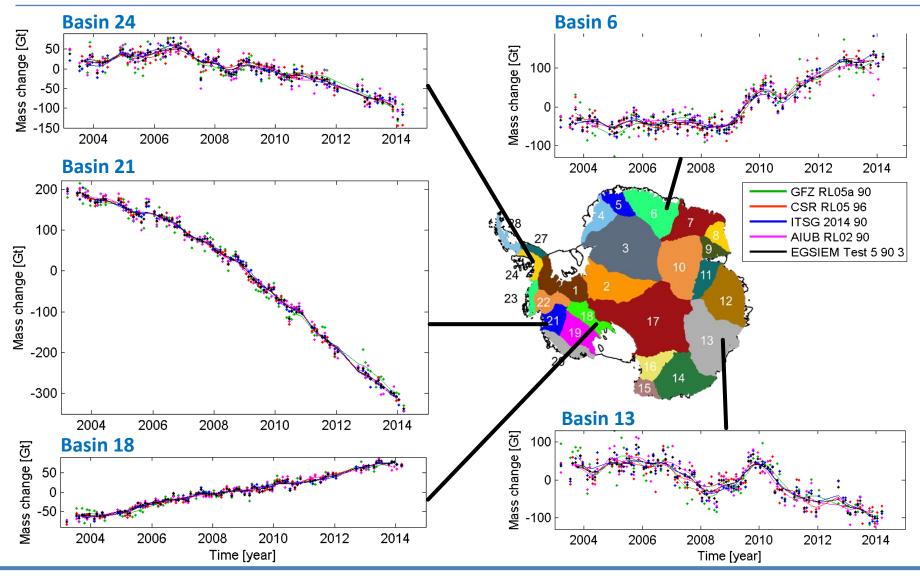
### **Entire Antarctica**



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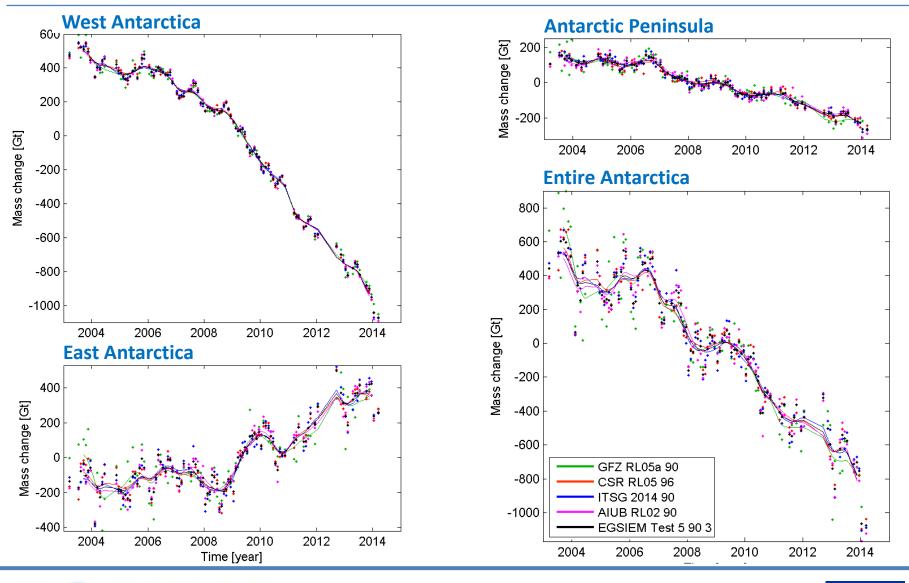
### Mass changes of Antarctic drainage basins (2/4)



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### Mass changes of Antarctic drainage basins (3/4)



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# Mass changes of Antarctic drainage basins (4/4)

The uncorrelated (white) noise content was assessed based on the STD of high-pass filtered time series.

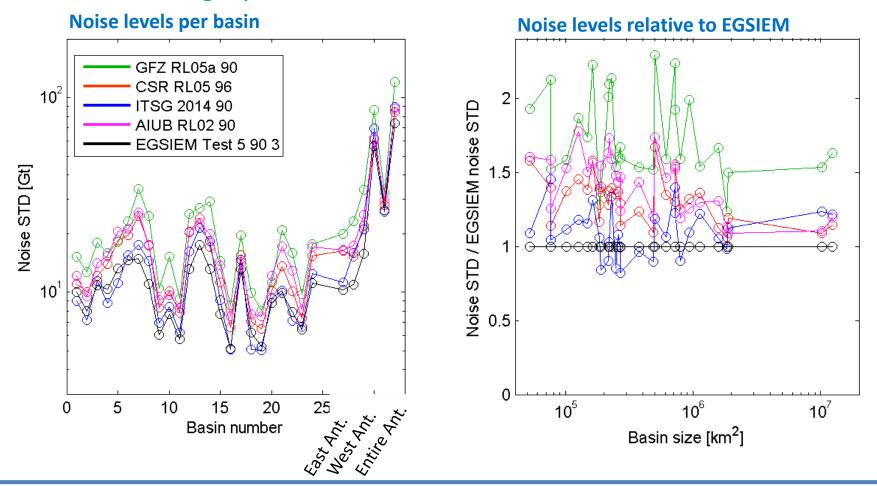


Image: Signal stateEGSIEM Progress Meeting # 2University of Luxembourg, January 18 – 19, 2016



### Mass changes of Antarctic drainage basins: conclusion

- Differences in noise STD between the series are on the order of 10% to 50%
- For most basins (including the large aggregations), EGSIEM has the lowest noise level (dependent on attenuation of high-degree noise by integration kernel; ITSG may benefit from less aggressive dampening).



