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Time-variable gravity field determination from GRACE Follow-On data using empirical noise modelling

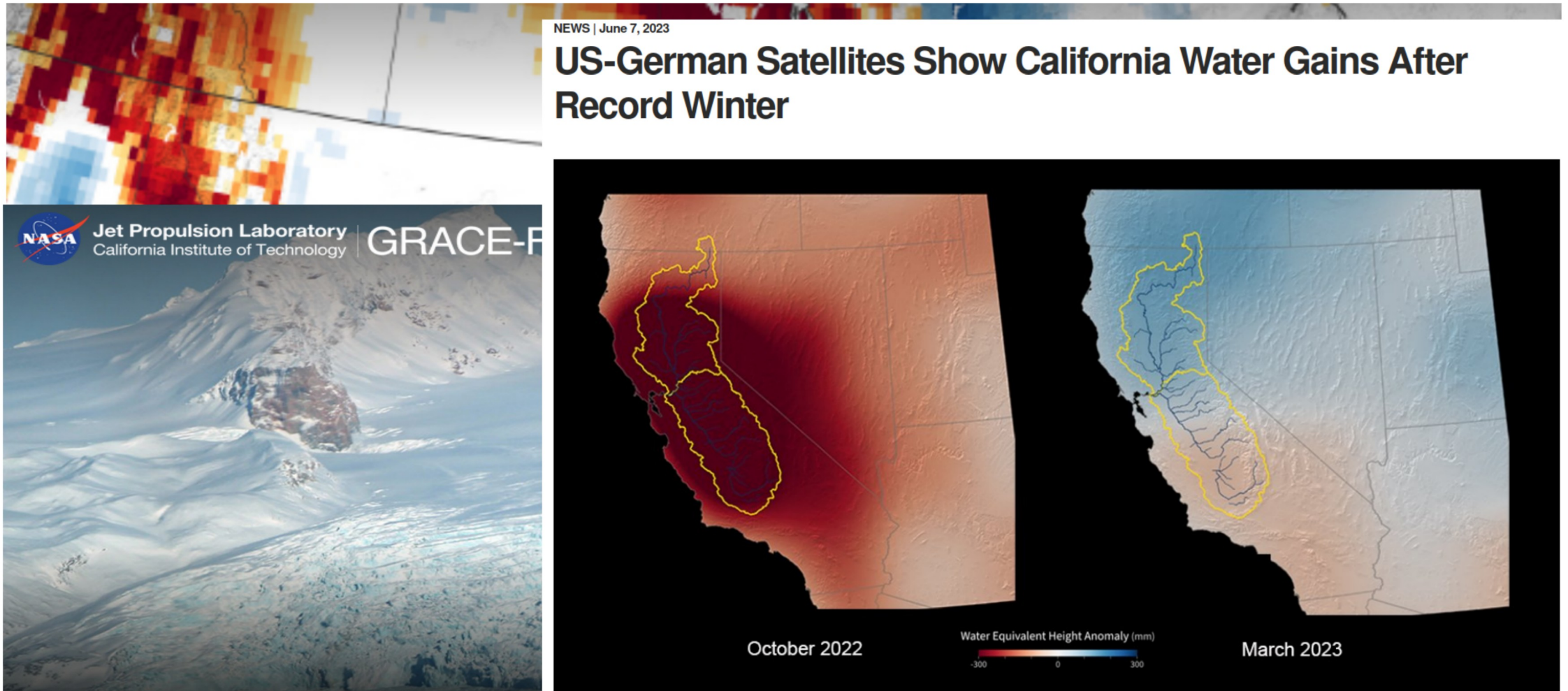
Martin Lasser, Ulrich Meyer, Daniel Arnold and Adrian Jäggi

21st Swiss Geoscience Meeting 2023, 17 – 18 November 2023, Mendrisio, Switzerland

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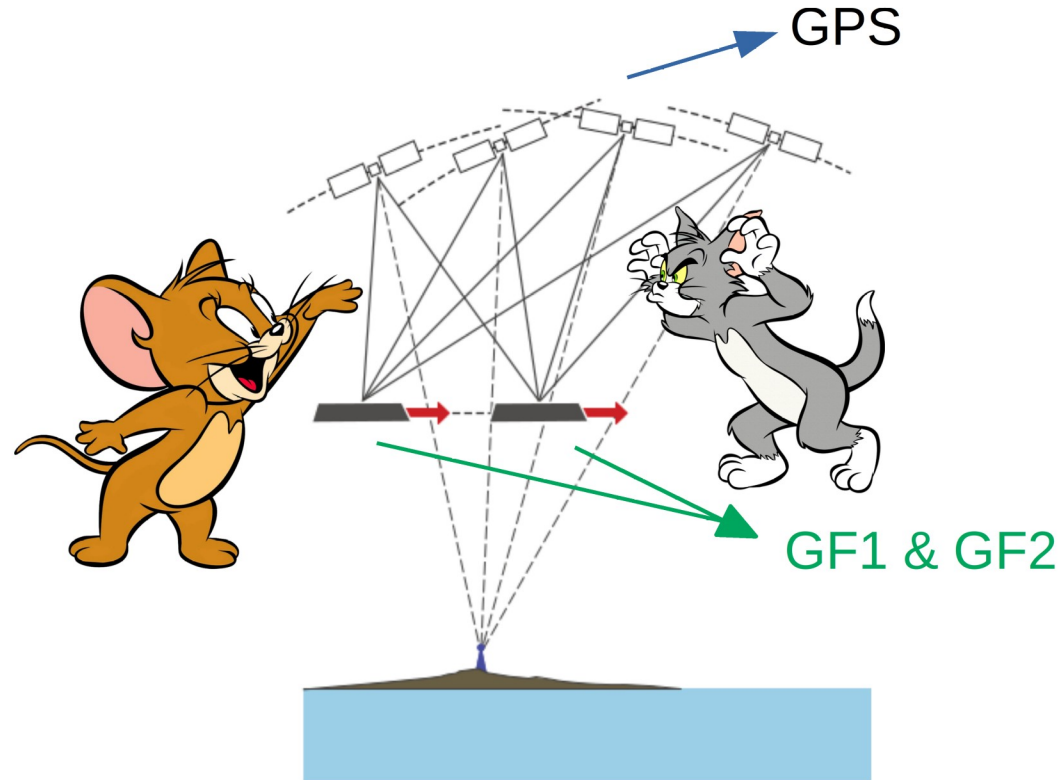
Satellite gravimetry with GRACE Follow-On

Recently in the news

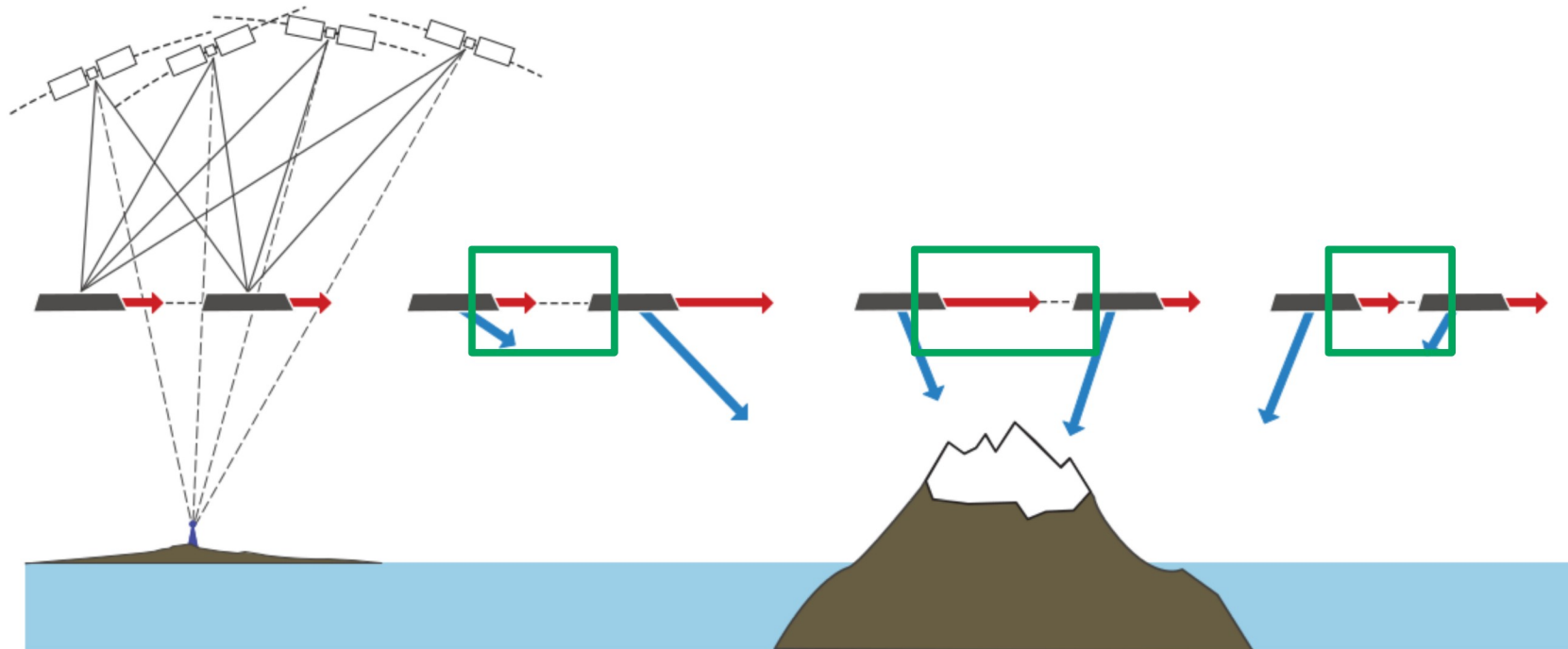


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GRACE Follow-On Observation concept



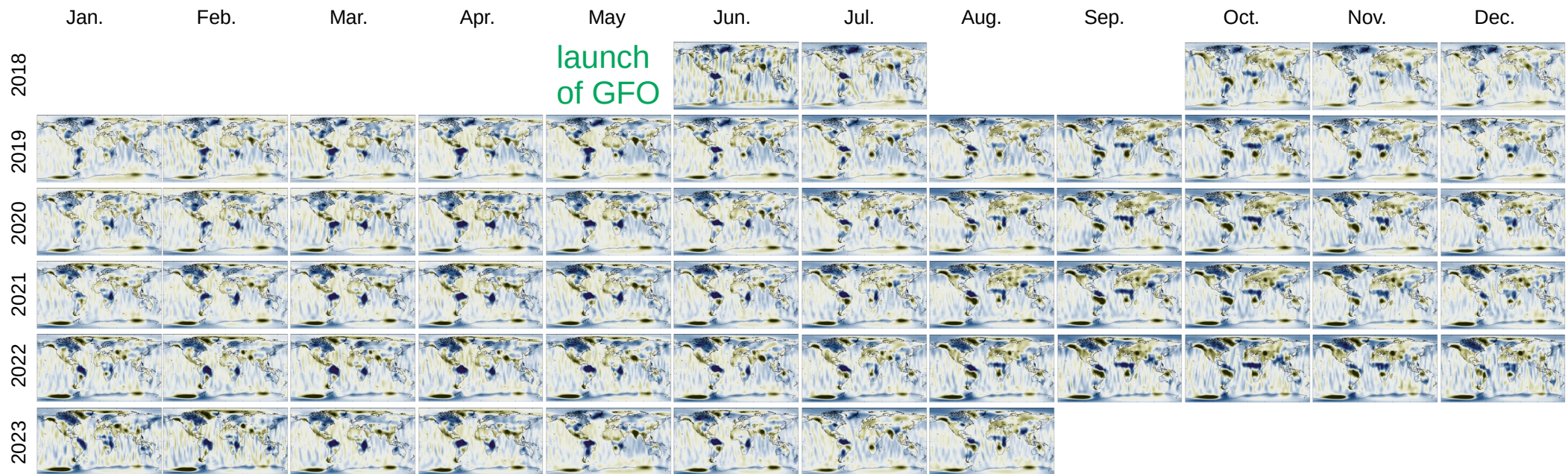
u^b GRACE Follow-On
Observation concept



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Operational GRACE Follow-On Solution

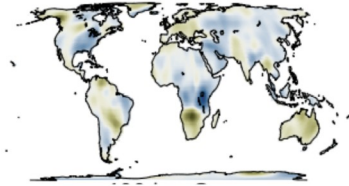
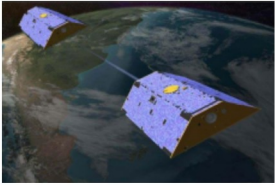
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Operational GRACE Follow-On Solution

Monthly gravity fields – parametrisation



Force models

Gravity field	AIUB-GRACE03S static
Astronomic bodies	JPL DE421 (all planets)
Mean pole	Linear
Solid Earth tides	IERS2010
Solid Earth pole tides	IERS2010
Ocean tides	FES2014b (+ admittances from TUG)
Ocean pole tides	Desai
Atmospheric tides	AOD RL06
Atmospheric & oceanic dealiasing	AOD RL06
Relativistic effects	IERS2010

Basic parametrisation

- initial conditions 2x[6]
- accelerometer bias 2x[3]
- accelerometer scaling 2x[3]

parameters per arc 24

Additional parameters

- 15 min PCA per satellite in
 - radial 2x[96]
 - along-track 2x[96]
 - cross-track 2x[96]

parameters per arc 576

in daily arcs (30 days):

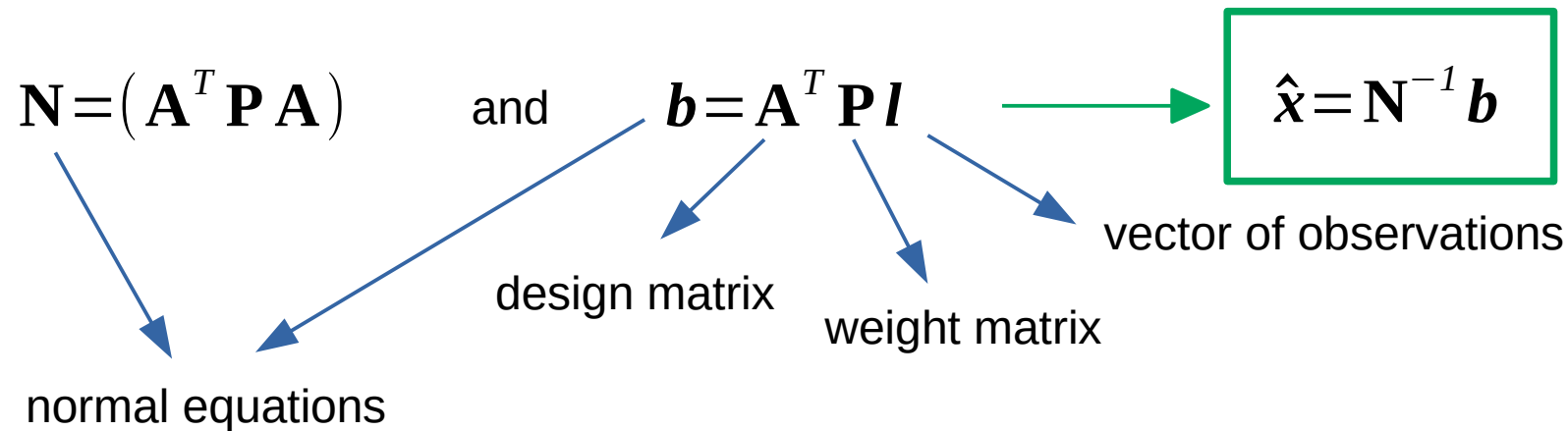
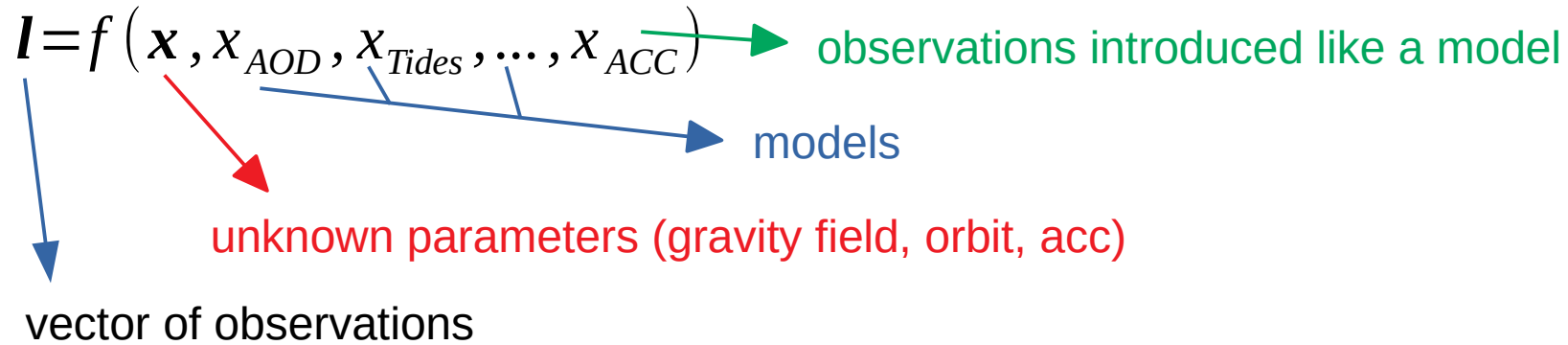
18000 <orbit> parameters

+ 9405 gravity field d/o=2..96

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Processing

Least-squares adjustment

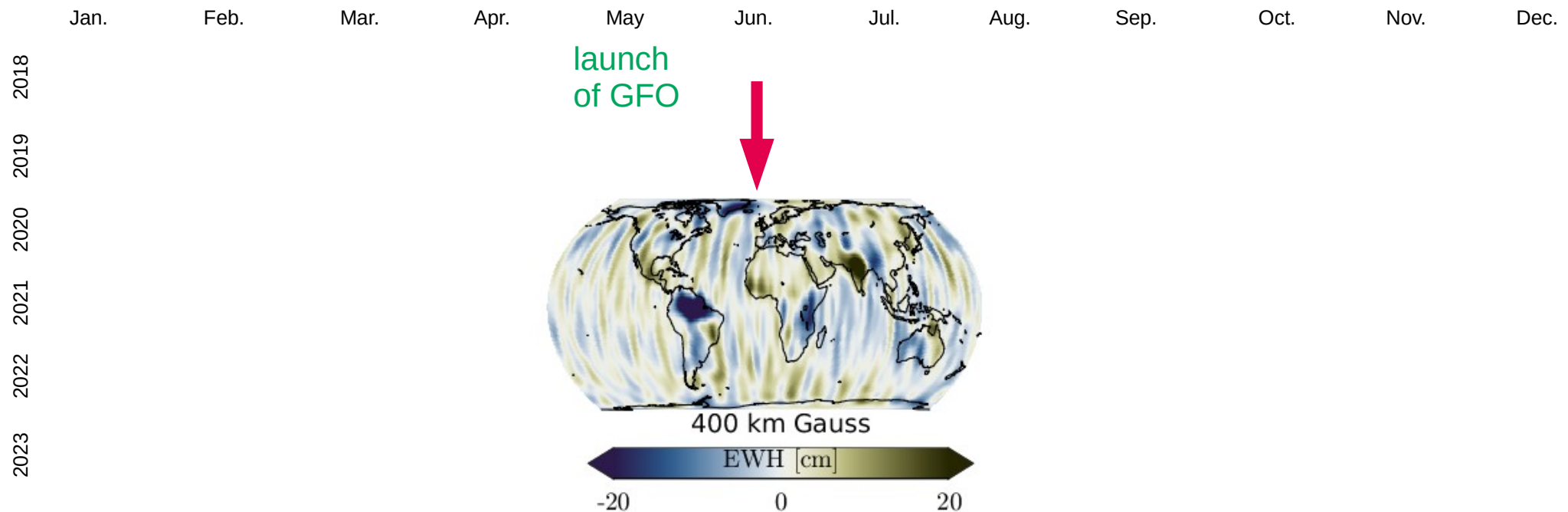


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Operational GRACE Follow-On Solution

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$$\mathbf{N} = (\mathbf{A}^T \mathbf{P} \mathbf{A}) \quad \longrightarrow \quad \hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$$
$$\mathbf{b} = \mathbf{A}^T \mathbf{P} \mathbf{l}$$

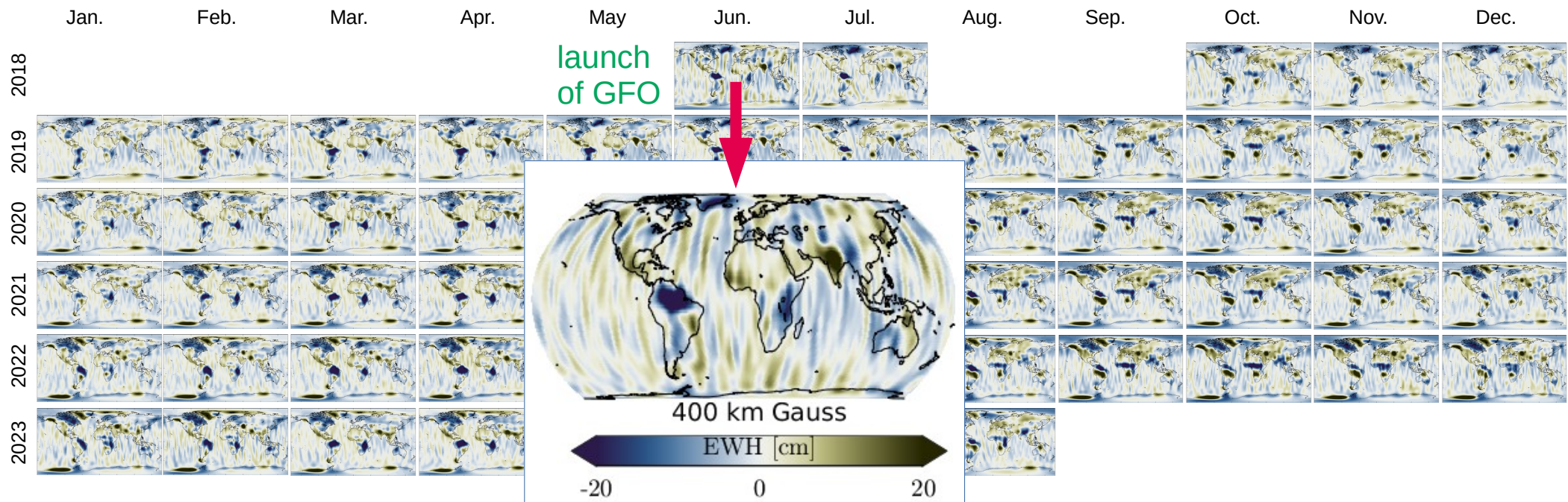


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Operational GRACE Follow-On Solution


Mosaic Jun 2018 – Aug 2023

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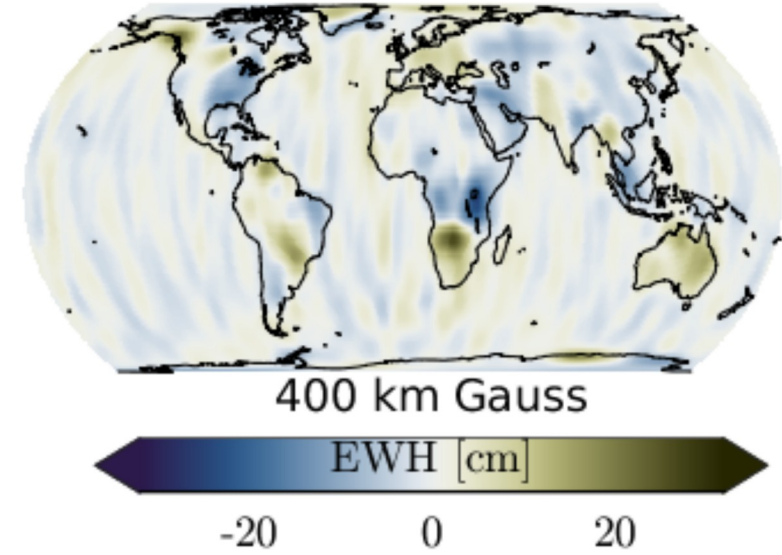
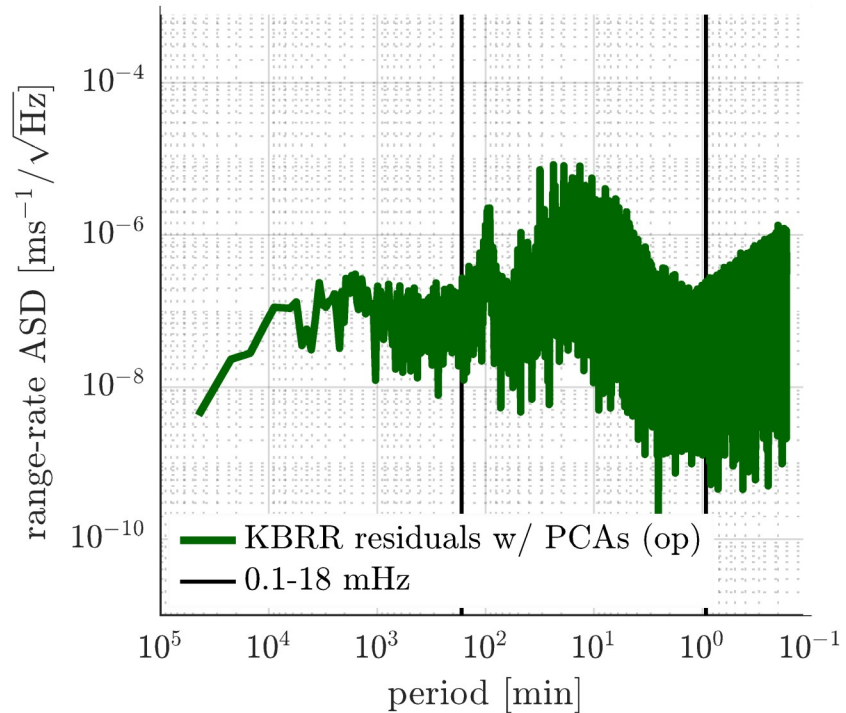
u^b Least-squares adjustment

What do we estimate?

Least-squares
 $\mathbf{N} = (\mathbf{A}^T \mathbf{P} \mathbf{A})$
 $\mathbf{b} = \mathbf{A}^T \mathbf{P} \mathbf{l}$

 $\hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$

$$\hat{\mathbf{e}} = \mathbf{l} - \mathbf{A} \hat{\mathbf{x}}$$

$$\hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$$




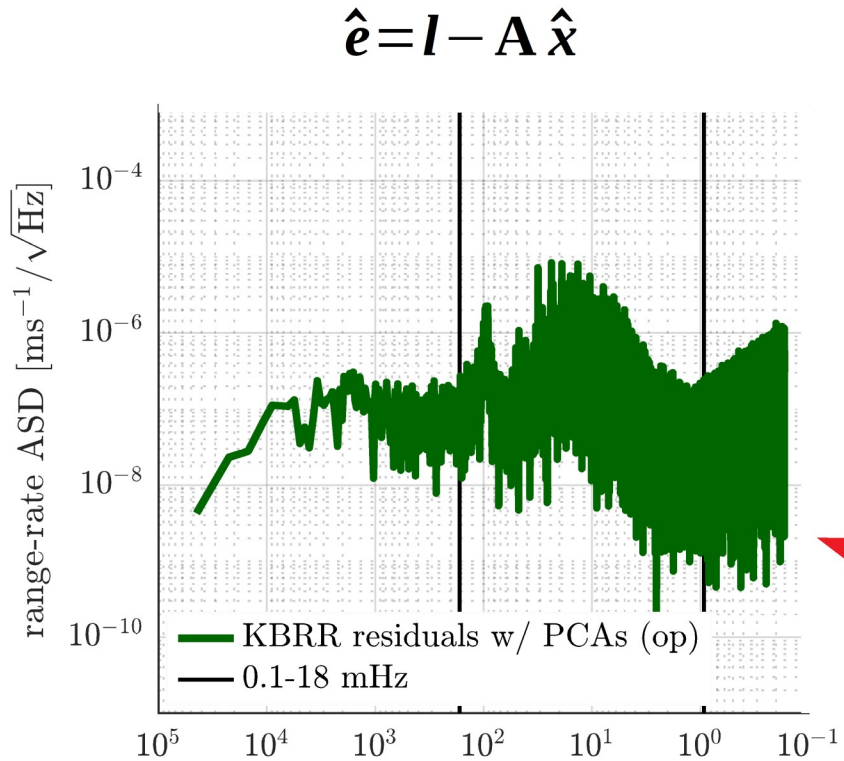
- information in the residuals (noise?)

- information in the parameters (signal?)

u^b Post-fit residuals

What do we expect?

Least-squares
 $\mathbf{N} = (\mathbf{A}^T \mathbf{P} \mathbf{A})$
 $\mathbf{b} = \mathbf{A}^T \mathbf{P} \mathbf{l}$

 $\hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$



The estimator is BLUE
 (best – linear – unbiased) if

- $E(\hat{\mathbf{x}}) = \mathbf{x}$
- $E(\mathbf{e}|\mathbf{x}) = E(\hat{\mathbf{e}}) = \mathbf{0}$
- $D(\mathbf{e}|\mathbf{x}, \sigma_0^2) = D(\mathbf{l}|\sigma_0^2) = \sigma_0^2 \mathbf{P}^{-1}$
 $= \sigma_0^2 \mathbf{Q}_{ee} = \sigma_0^2 \mathbf{Q}_{ll}$

$\hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$ unbiased

$\mathbf{C}_{\hat{\mathbf{x}}\hat{\mathbf{x}}} = \sigma_0^2 (\mathbf{A}^T \mathbf{P} \mathbf{A})^{-1}$ biased

$\mathbf{P} = \frac{1}{\sigma_0^2} \mathbf{I}$

u^b Post-fit residuals

A measure of noise

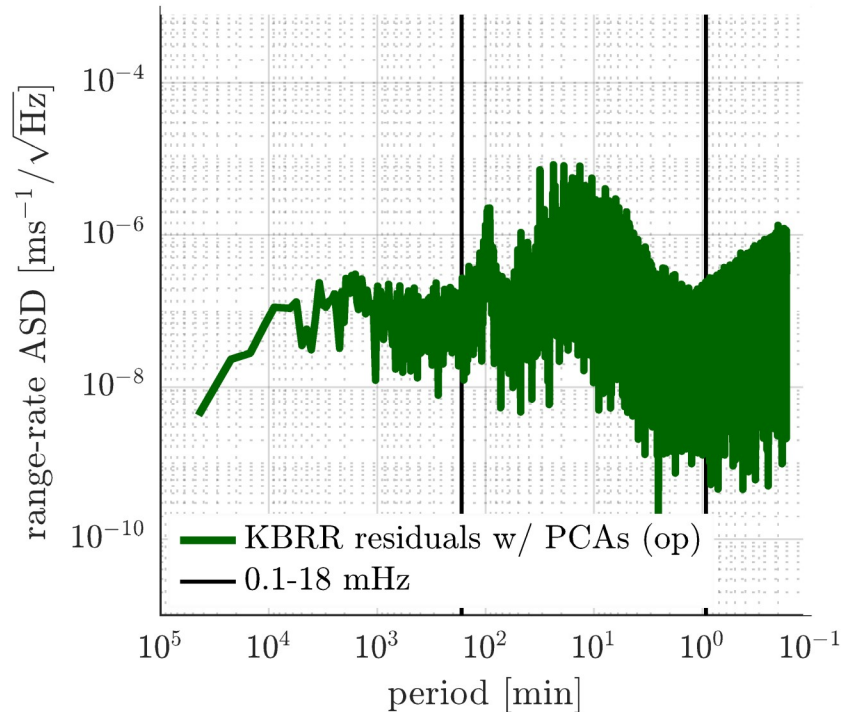
Least-squares

$$\mathbf{N} = (\mathbf{A}^T \mathbf{P} \mathbf{A})$$

$$\mathbf{b} = \mathbf{A}^T \mathbf{P} \mathbf{l}$$



$$\hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$$




$$\hat{\mathbf{e}} = \mathbf{l} - \mathbf{A} \hat{\mathbf{x}} \quad (\text{post-fit residuals})$$

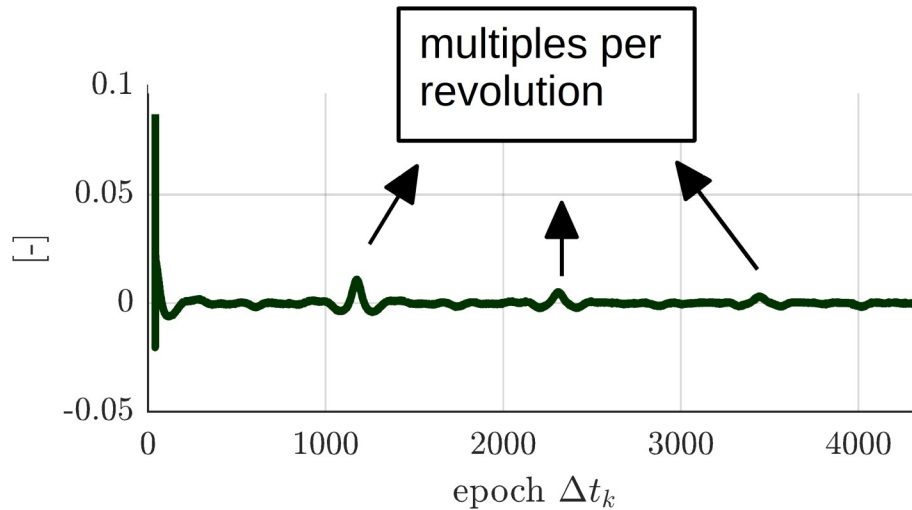
$$\text{cov}(\Delta t_k) = \frac{1}{N} \sum_{i=0}^N \hat{\mathbf{e}}(t_i) \hat{\mathbf{e}}(t_i + \Delta t_k)$$

- stationarity assumed
- biased estimation of auto-covariance
→ covariance matrix nondegenerate

u^b Post-fit residuals

Serial correlation

Least-squares
 $\mathbf{N} = (\mathbf{A}^T \mathbf{P} \mathbf{A})$
 $\mathbf{b} = \mathbf{A}^T \mathbf{P} \mathbf{l}$

 $\hat{\mathbf{x}} = \mathbf{N}^{-1} \mathbf{b}$



$$\hat{\mathbf{e}} = \mathbf{l} - \mathbf{A} \hat{\mathbf{x}} \quad (\text{post-fit residuals})$$

$$\text{cov}(\Delta t_k) = \frac{1}{N} \sum_{i=0}^N \hat{\mathbf{e}}(t_i) \hat{\mathbf{e}}(t_i + \Delta t_k)$$

- stationarity assumed
- biased estimation of auto-covariance
 → covariance matrix nondegenerate

block
Toeplitz
matrix

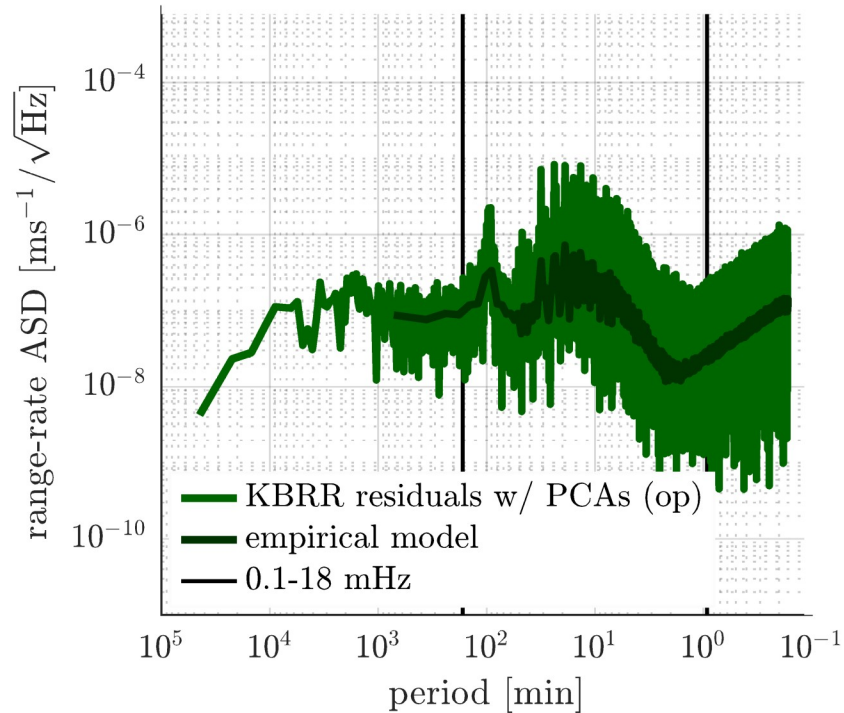


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Post-fit residuals

The estimated noise model



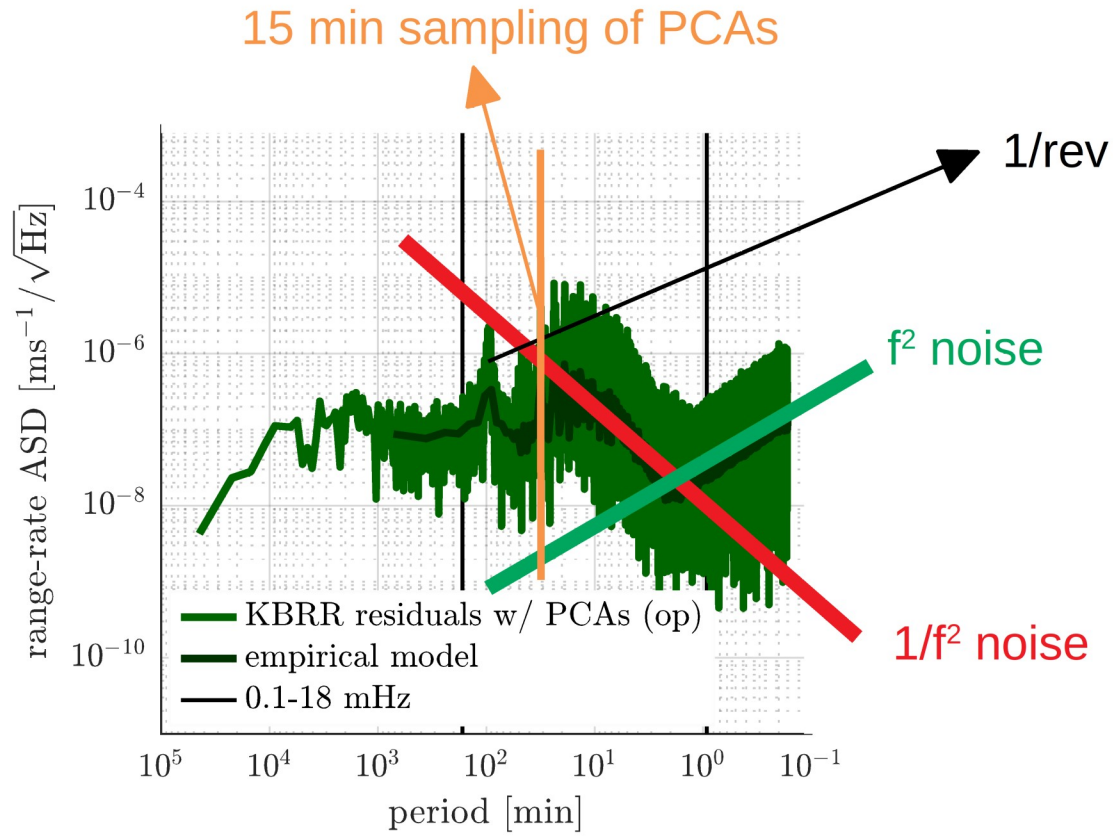
$$\hat{e} = l - A \hat{x}$$

$$\text{cov}(\Delta t_k) = \frac{1}{N} \cdot \sum_{i=0}^N \hat{e}(t_i) \hat{e}(t_i + \Delta t_k)$$

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Post-fit residuals

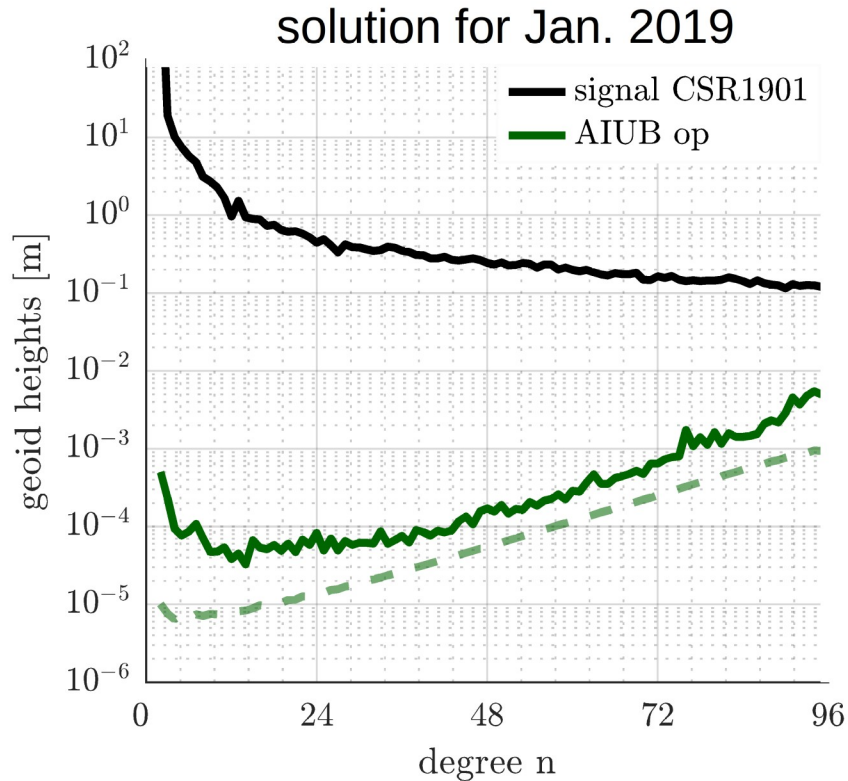
What we see



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Results

Difference degree amplitudes



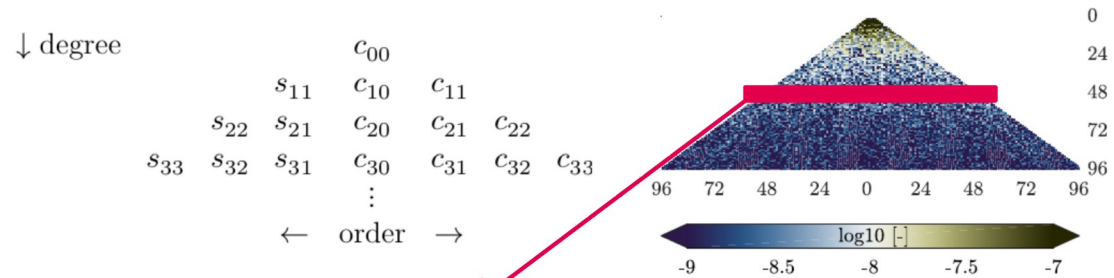
- formal errors too optimistic compared to the assessed noise

Gravity field representation in spherical harmonics

$$V = \frac{GM}{R} \sum_{n=0}^N \left(\frac{a_e}{r}\right)^n \sum_{m=0}^n P_{nm}(\sin(\varphi)) [c_{nm} \cos(m\lambda) + s_{nm} \sin(m\lambda)]$$

unknown coefficients

degree N point of evaluation (r, φ, λ)

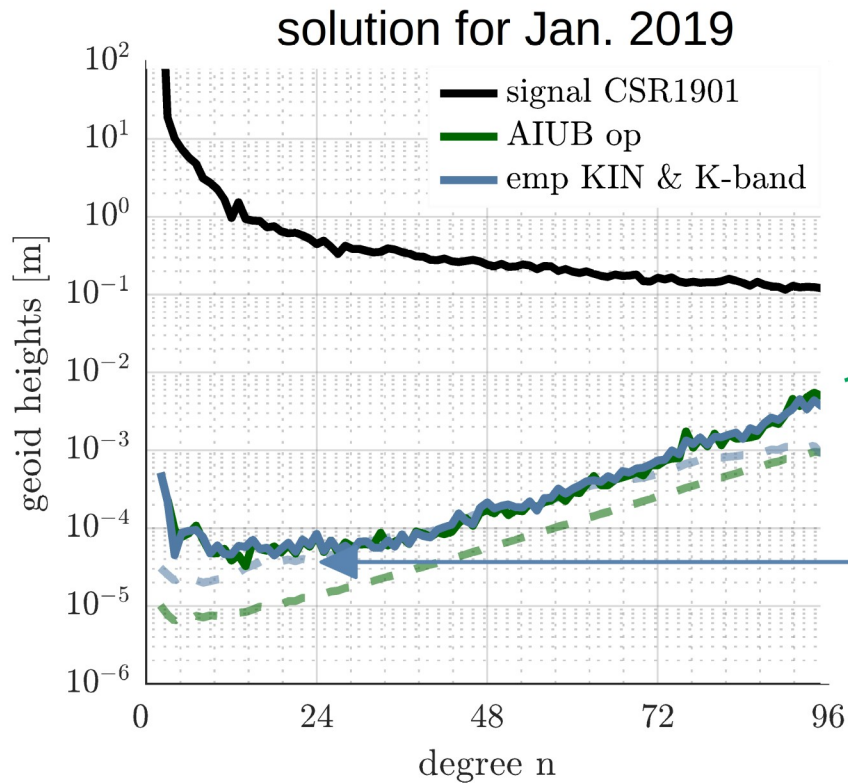


$$\sigma_n = \sqrt{\sum_{m=0}^n c_{nm}^2 + s_{nm}^2}$$

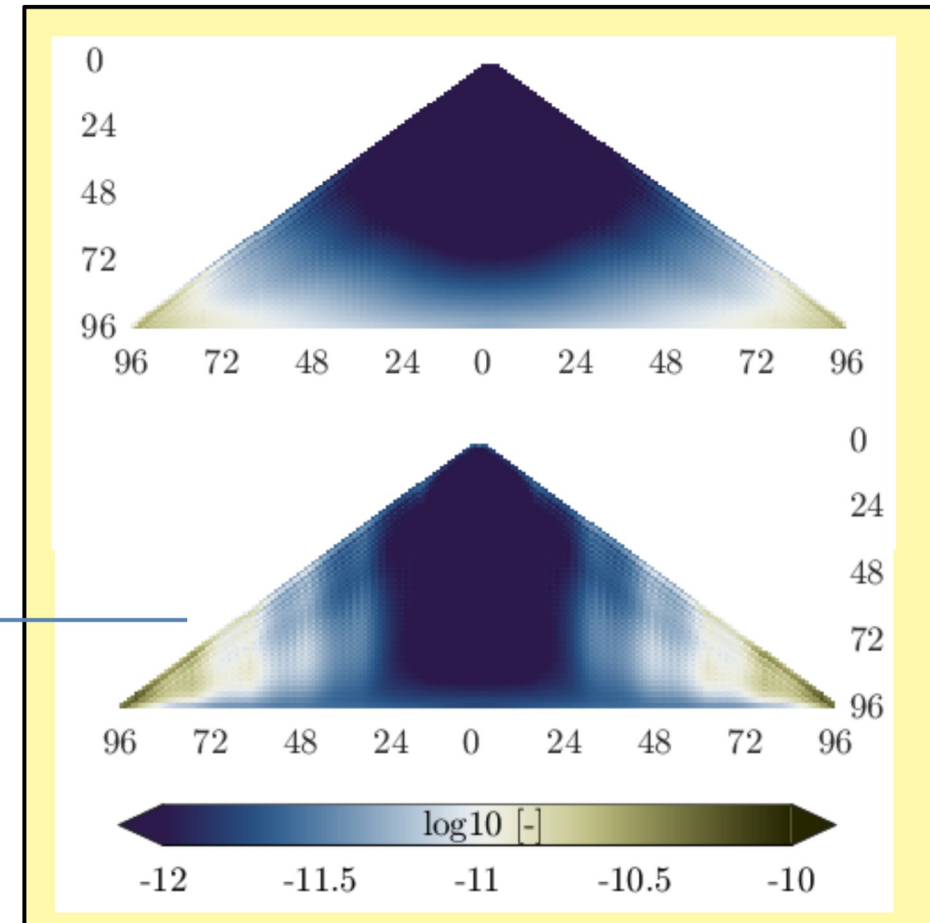
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Results

Spectral domain

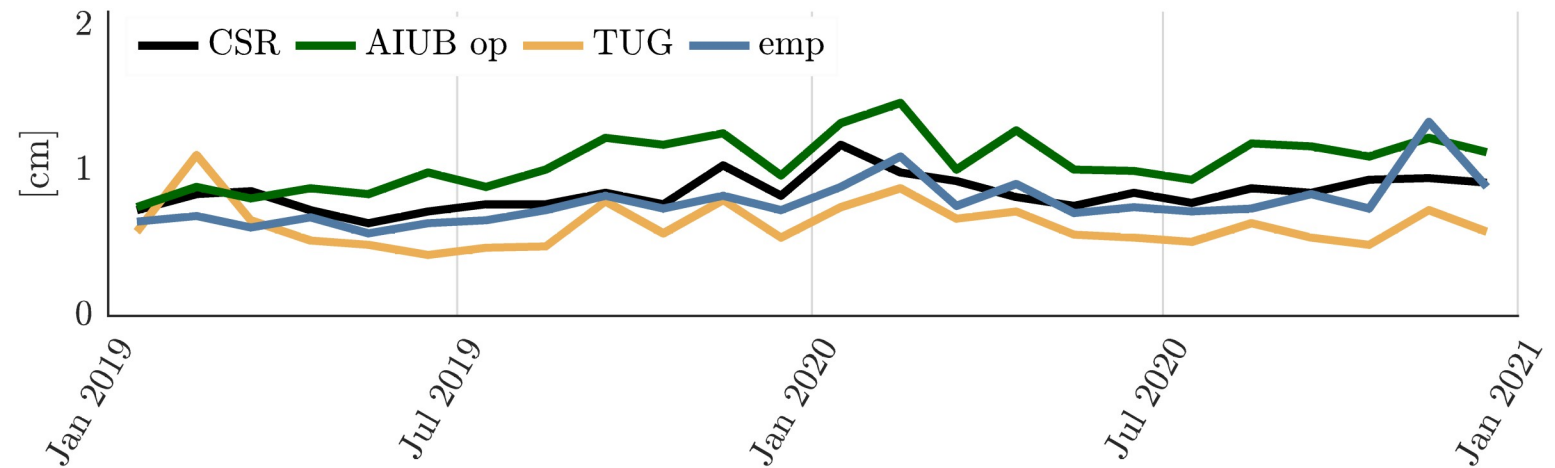
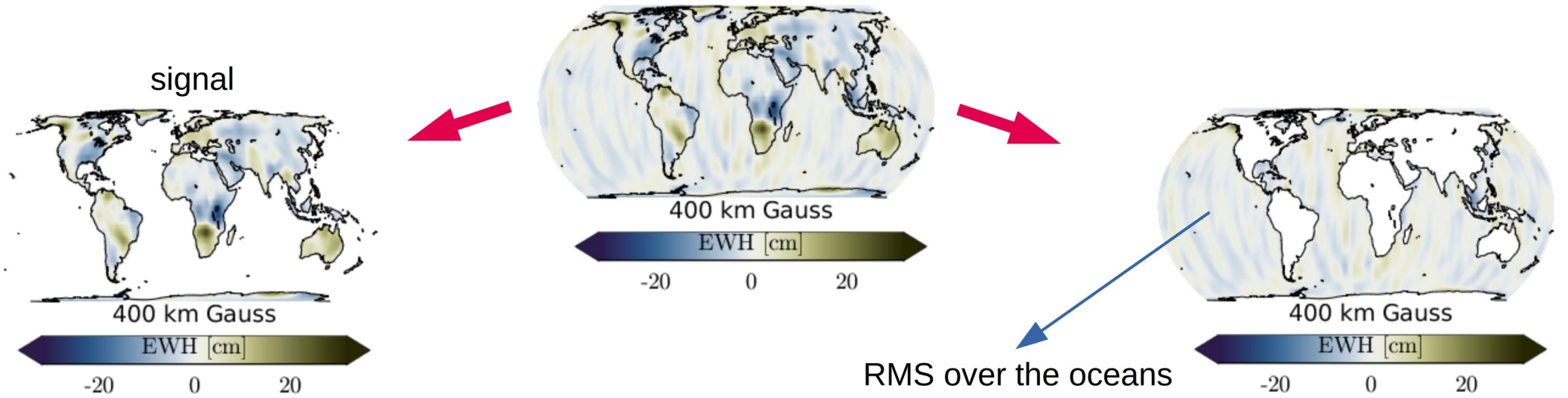


- formal errors reflect assessed noise very well
- including features of resonance orders



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Noise evaluation RMS over the oceans



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Thank you for your attention

Contact

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References

A

Beutler, G., Jäggi, A., Mervart, L. and Meyer, U. [2010]: The celestial mechanics approach: theoretical foundations. *Journal of Geodesy*, vol. 84(10), pp. 605-624. <https://doi.org/10.1007/s00190-010-0401-7>

Ellmer, M. [2018]: *Contributions to GRACE Gravity Field Recovery: Improvements in Dynamic Orbit Integration Stochastic Modelling of the Antenna Offset Correction, and Co-Estimation of Satellite Orientations*. PhD thesis, Graz University of Technology, In Monographic Series TU Graz, number 1, Verlag der Technischen Universität Graz, Graz, Austria. <https://doi.org/10.3217/978-3-85125-646-8>

Jäggi, A., Beutler, G., Meyer, U., Prange, L., Dach, R. and Mervart, L. [2012]: AIUB-GRACE02S - Status of GRACE Gravity Field Recovery using the Celestial Mechanics Approach. In S. Kenyon, M. C. Pacino, and U. Marti, editors, *International Association of Geodesy Symposia: Geodesy for Planet Earth*, volume 136, pages 161–170. Springer, Berlin-Heidelberg, Germany. ISBN 978-3-642-20337-4. https://doi.org/10.1007/978-3-642-20338-1_20

Kvas, A., Behzadpour, S., Ellmer, M., Klinger, B., Strasser, S., Zehentner, N. and Mayer-Gürr, T. [2019]: Overview and evaluation of a new GRACE-only gravity field time series. *Journal of Geophysical Research: Solid Earth*. ISSN 2169-9313. <https://doi.org/10.1029/2019JB017415>

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References

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Lasser, M. [2022]: *Noise Modelling for GRACE Follow-On Observables in the Celestial Mechanics Approach*. PhD thesis, University of Bern, Bern, Switzerland. <https://boristheses.unibe.ch/4127/>

Lasser, M., Meyer, U., Arnold, D. and Jäggi, A. [2020]: AIUB-GRACE-FO-operational - Operational GRACE Follow-On monthly gravity field solutions. <https://doi.org/10.5880/ICGEM.2020.001>

NASA Jet Propulsion Laboratory (JPL) [2019]: GRACE-FO Monthly Geopotential Spherical Harmonics CSR Release 6.0. <https://doi.org/10.5067/GFL20-MC060>