



Medium-term predictions of F10.7 and F30 cm solar radio flux with the adaptive Kalman filter

1-24 months ahead



Tatiana Podladchikova¹, Elena Petrova^{1,2}, Astrid Veronig^{3,4},
Stijn Lemmens⁵, Benjamin Bastida Virgili⁶, Tim Flohrer⁵



Radio Emissions from the Sun: **ON**line **AN**alytical Computer-aided Estimator



9 September 2021, ESPM-16
t.podladchikova@skoltech.ru

1. Skolkovo Institute of Science and Technology, Moscow, Russia
2. Centre for Mathematical Plasma Astrophysics, KU Leuven, Belgium
3. Institute of Physics, University of Graz, Austria
4. Kanzelhöhe Observatory for Solar and Environmental Research, Austria
5. ESA/ESOC, Germany
6. IMS GmbH@ESA, Germany

Relevance

F10.7 and F30 cm radio flux are required by most models characterizing the state of the upper Earth's atmosphere, such as the thermosphere and ionosphere in order to specify **satellite orbits, re-entry services, collision avoidance maneuvers** and modeling of **space debris evolution**.

Solar EUV irradiance

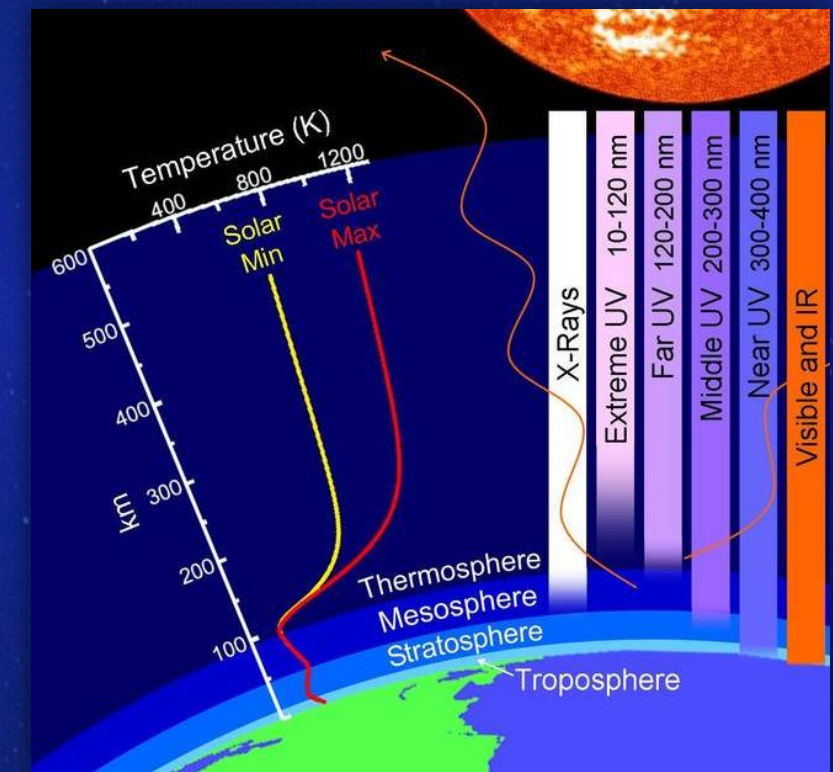
Major source of thermosphere density variations

Atmosphere models in use:
NRLMSIS-00, DTM-2013, GOST-2004
Bastida Virgili, 2017

Change of density

Atmospheric drag change

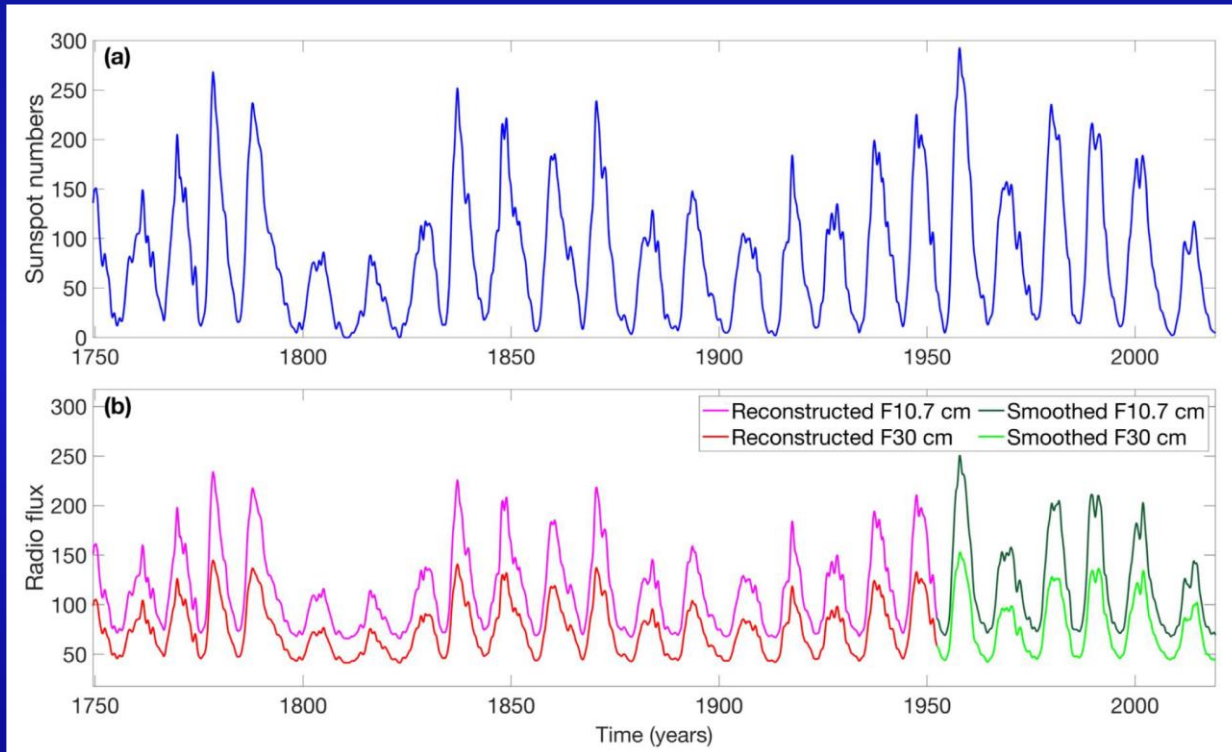
Decay and following re-entry



Properties of Earth's atmosphere, NASA



Data Preparation



✓ **F10.7/F30 cm** is a measurement of the **total solar emission** at a wavelength of **10.7/30 cm** from all sources present on the solar disk averaged over an hour.

✓ **Smoothing** of monthly mean data using optimized smoothing technique (Podladchikova et. al, 2017)

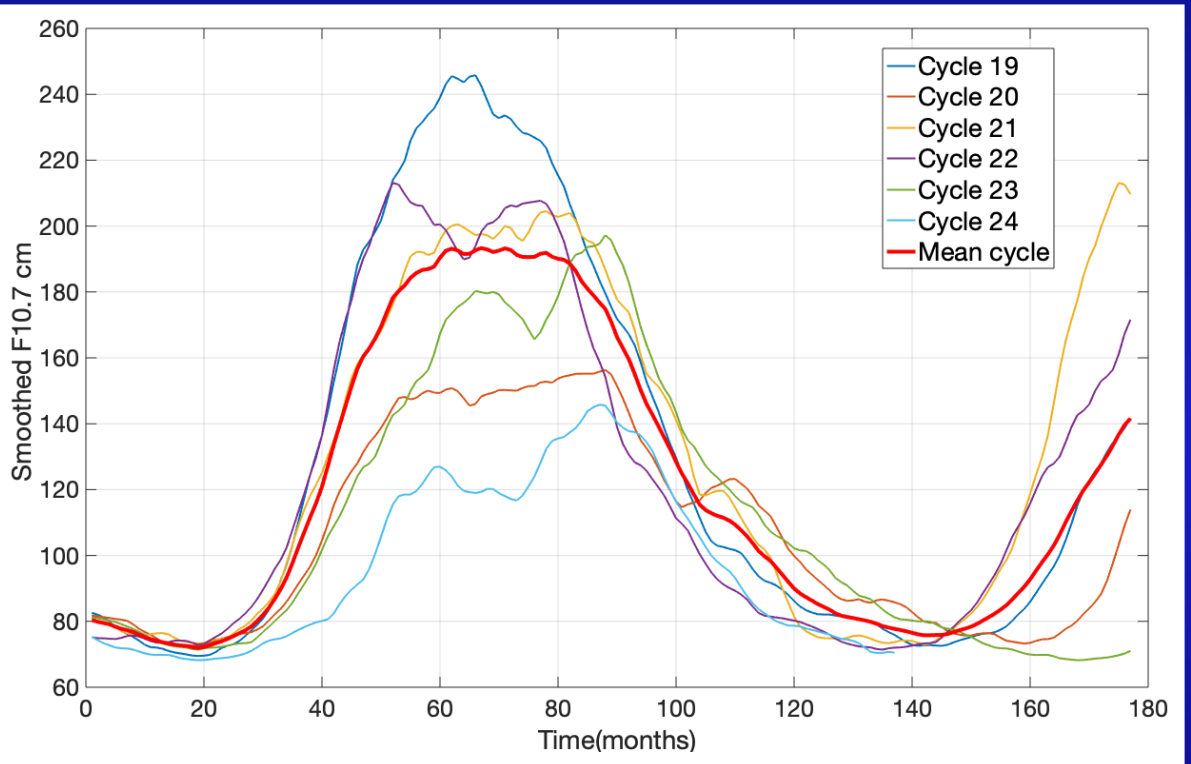
✓ Reconstruction of F10.7 and F30 using **sunspot numbers** and 3d order linear regression

F30 cm radio flux: Toyokawa & Nobeyama Observatories, Japan
F10.7 cm radio flux: Ottawa and Pentincton Observatory, Canada

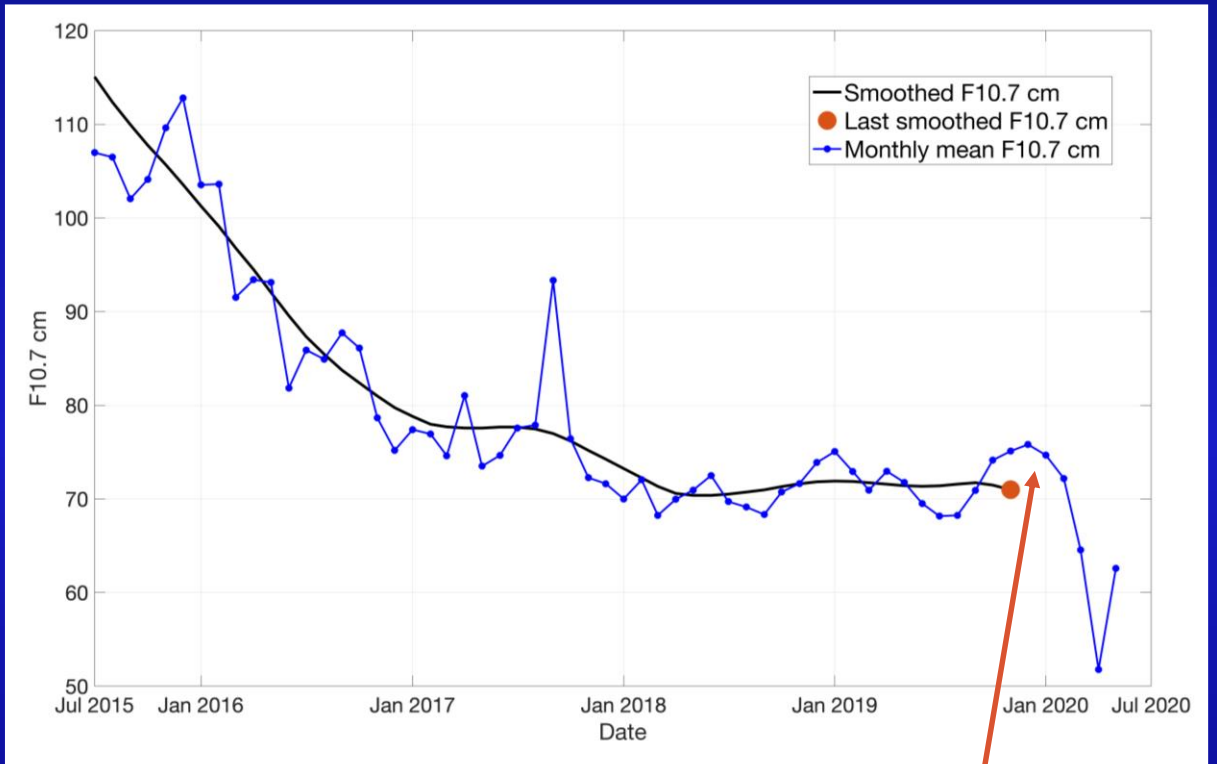
$$1 \text{ sfu} = 10^{-22} \text{ Wm}^{-2}\text{Hz}^{-1}$$

Tapping, 2013; Dudok de Wit and Bruinsma, 2017

Radio flux predictions with McNish-Lincoln method



Mean cycle – average of all cycles smoothed solar radio flux aligned on the month of the minimum



Prediction for month m :

$$F_m = \bar{F}_m + k_{mi}(F_i - \bar{F}_i)$$

mean value (month m)

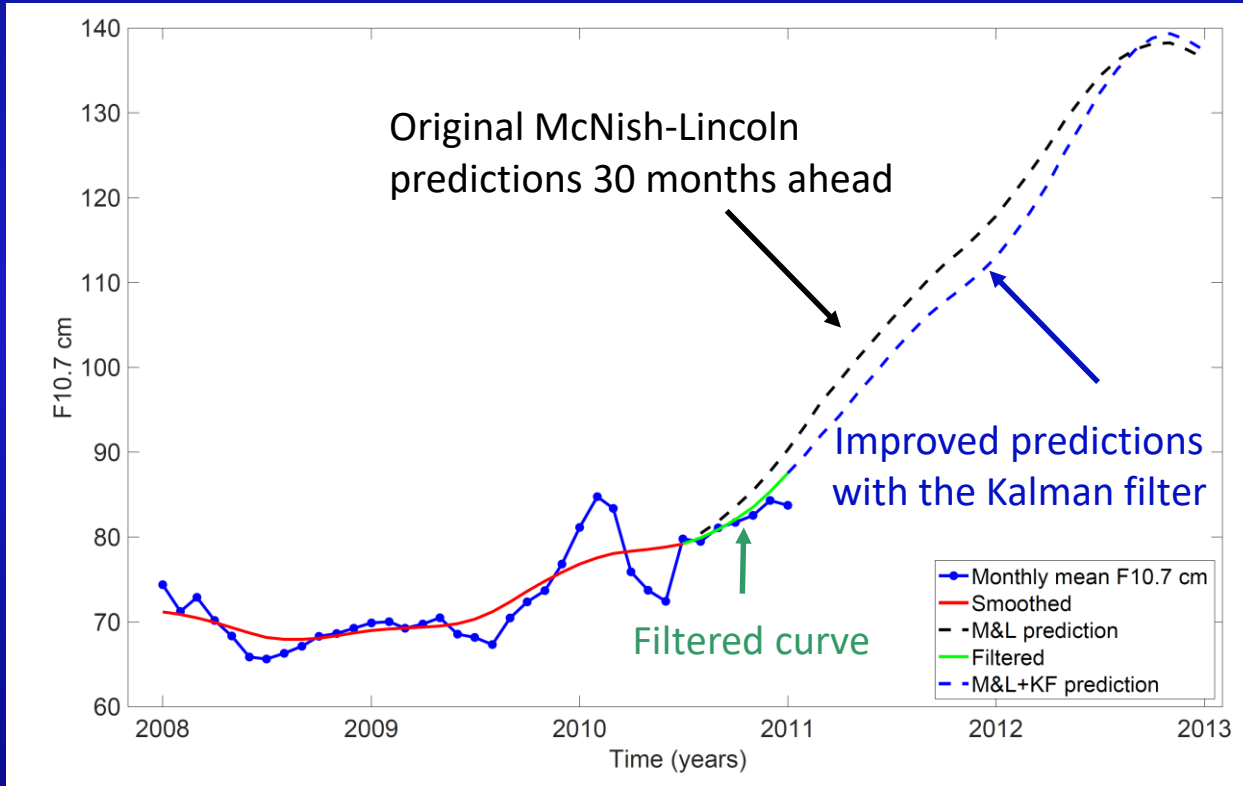
mean value (month i)

last smoothed value (month i)

Drawback

Available with the **6-month delay** with respect to the current time.

Improvement of initial McNish-Lincoln predictions with the adaptive Kalman filter



Main idea

To **assimilate** the monthly mean radio flux data available over the last 6-month period and estimate **the smoothed radio flux at the current time**.

Apply McNish-Lincoln method again, but use the **estimated smoothed radio flux at the current time** as the starting point for the predictions.

The improvements in prediction accuracy with application of the McNish-Lincoln + Kalman filter reach **36% for F10.7** and **39% for F30**.

The **RMS error of predictions** with lead times from 1 to 24 months is **5–27 solar flux units (sfu)** for the F10.7 and **3–16 sfu** for F30.

We remove the drawback of a 6-month delay

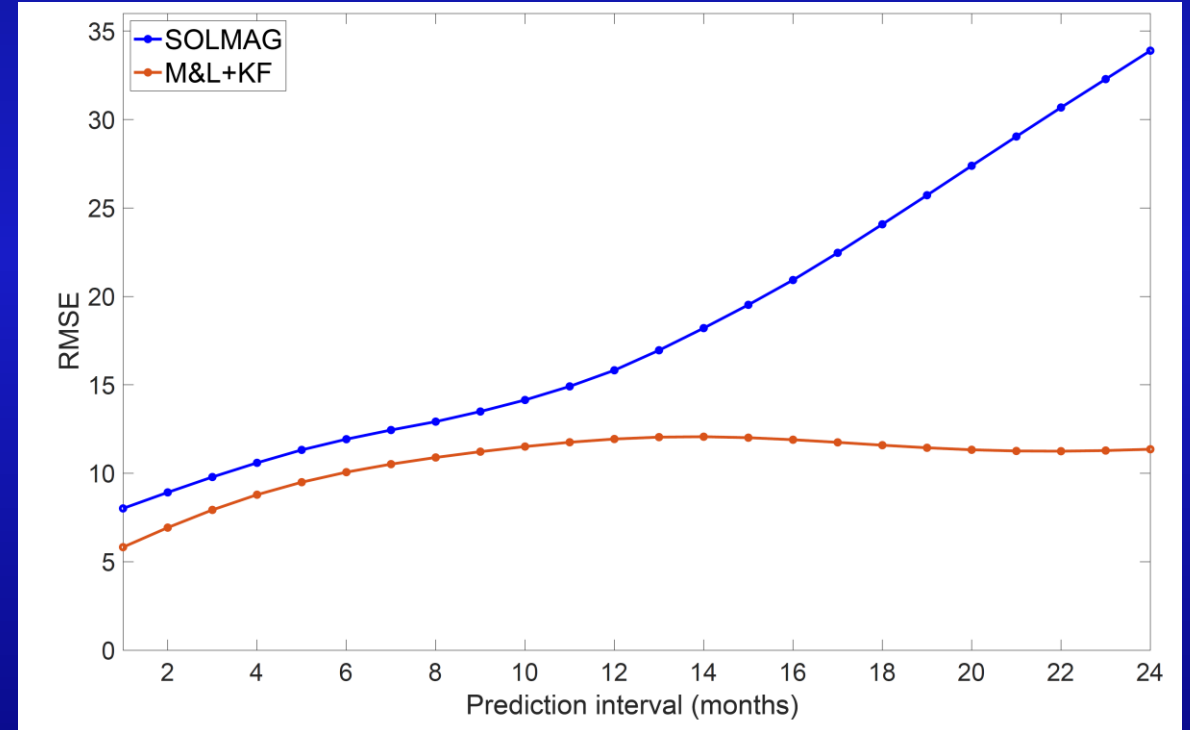
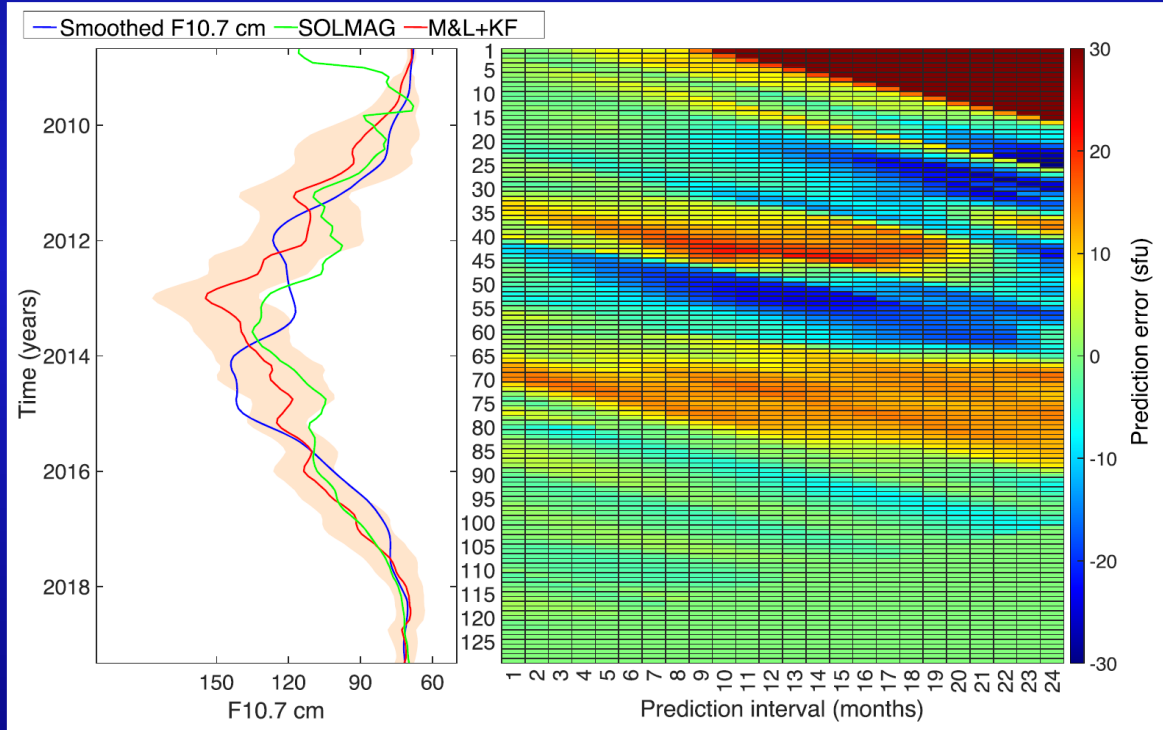
Performance results for solar cycle 24

F10.7 predictions
12 months ahead

Heat map:

>0: RESONANCE outperforms
<0: SOLMAG outperforms

Root-mean square error of prediction
1-24 month ahead (in sfu)



SOLMAG – short and long-term solar and geomagnetic activity prediction developed and currently employed by ESA

RESONANCE (M&L+KF) statistically outperforms the ESA SOLMAG method by 15.5-66.5%.

Re-entry predictions with past ESA re-entry campaigns

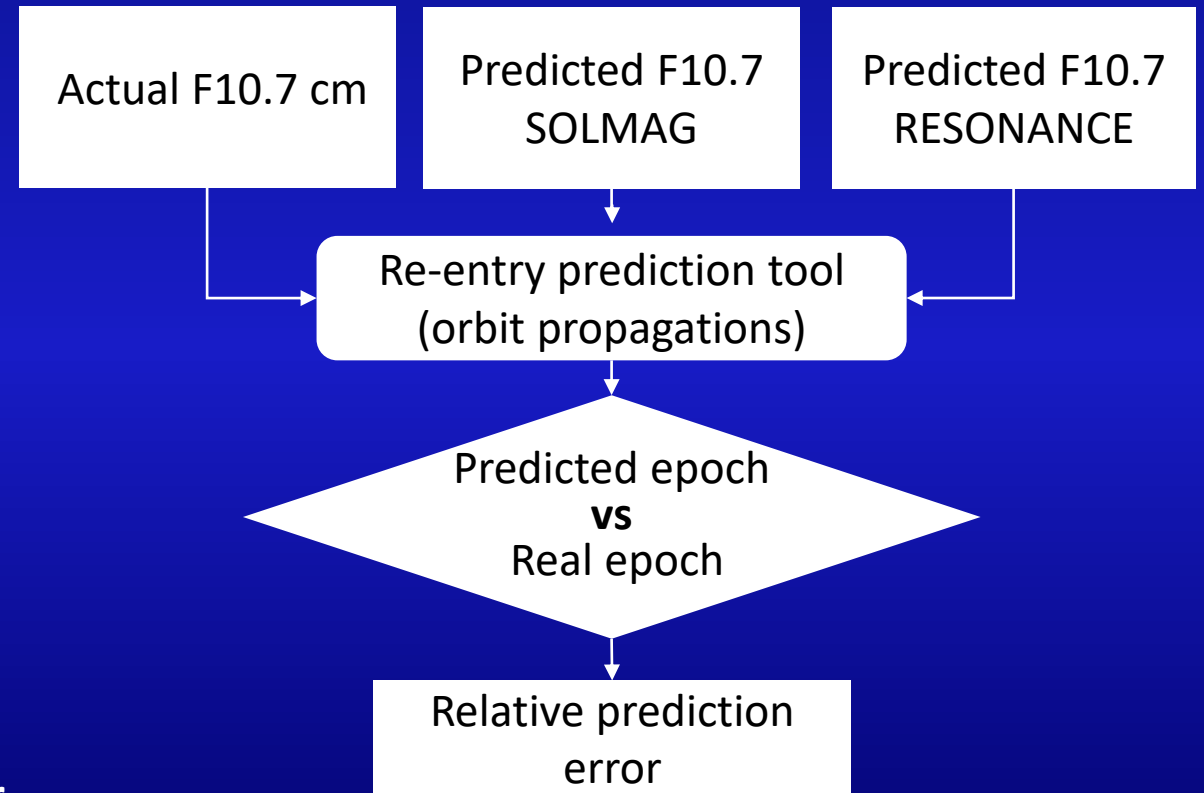
The objects re-entered over 2006-2019, covering the period over **the full solar cycle**.

- ✓ 602 payloads and rocket bodies (1635 predictions)
- ✓ 2344 objects of space debris (9584 predictions)

Recovered space debris, ESA

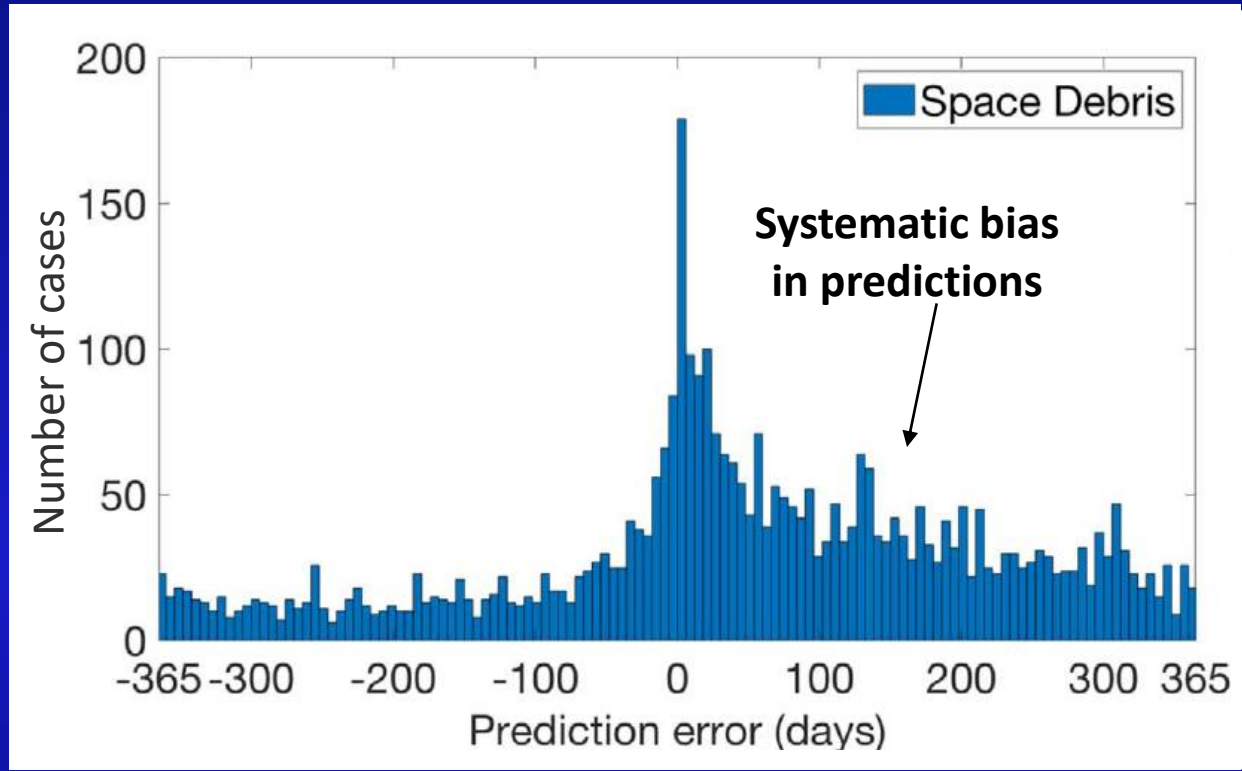
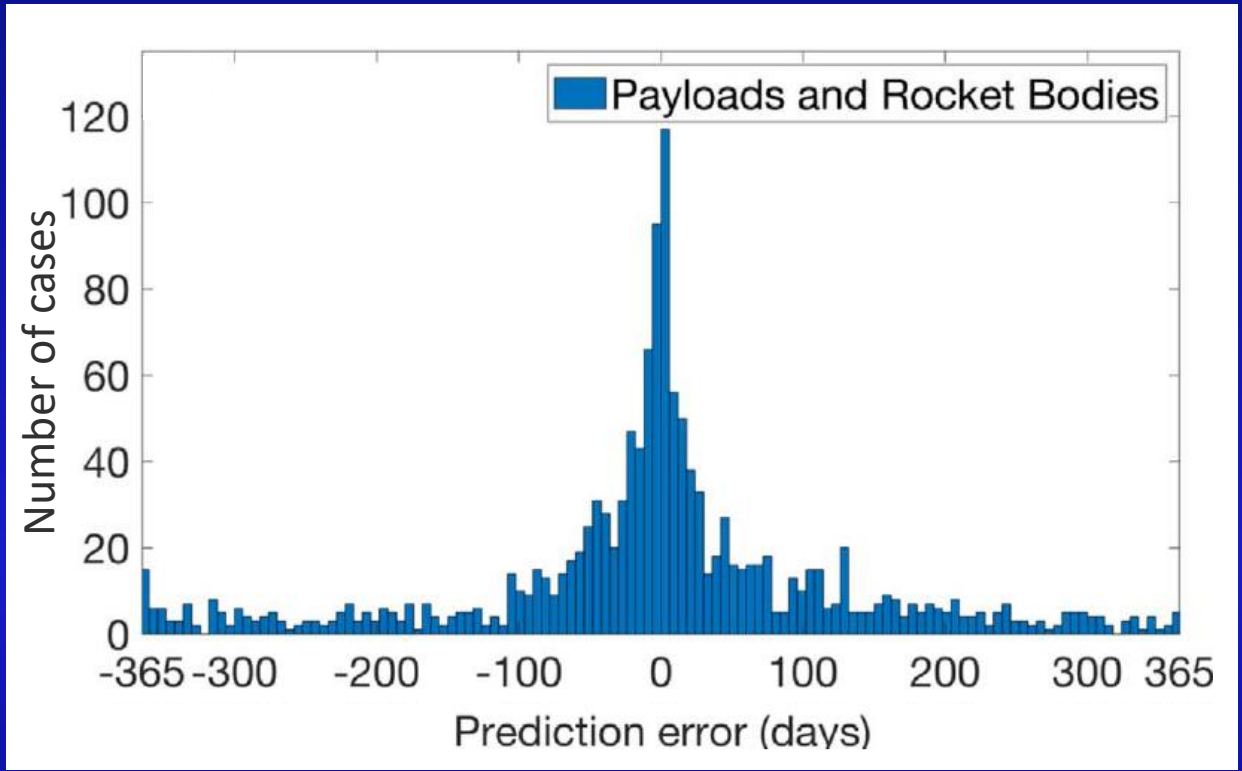


- ✓ On average **2 small tracked objects** (< 1 m) re-enter the Earth atmosphere and burn up **every day**.
- ✓ Objects of **moderate size** (> 1 m) – **once a week**.
- ✓ **Large objects** (heavy science satellites) **few times in a year**.



Systematic evaluation of re-entry predictions using actual F10.7 as an input

Positive/negative errors show cases of the re-entry observed earlier (later) than its prediction.



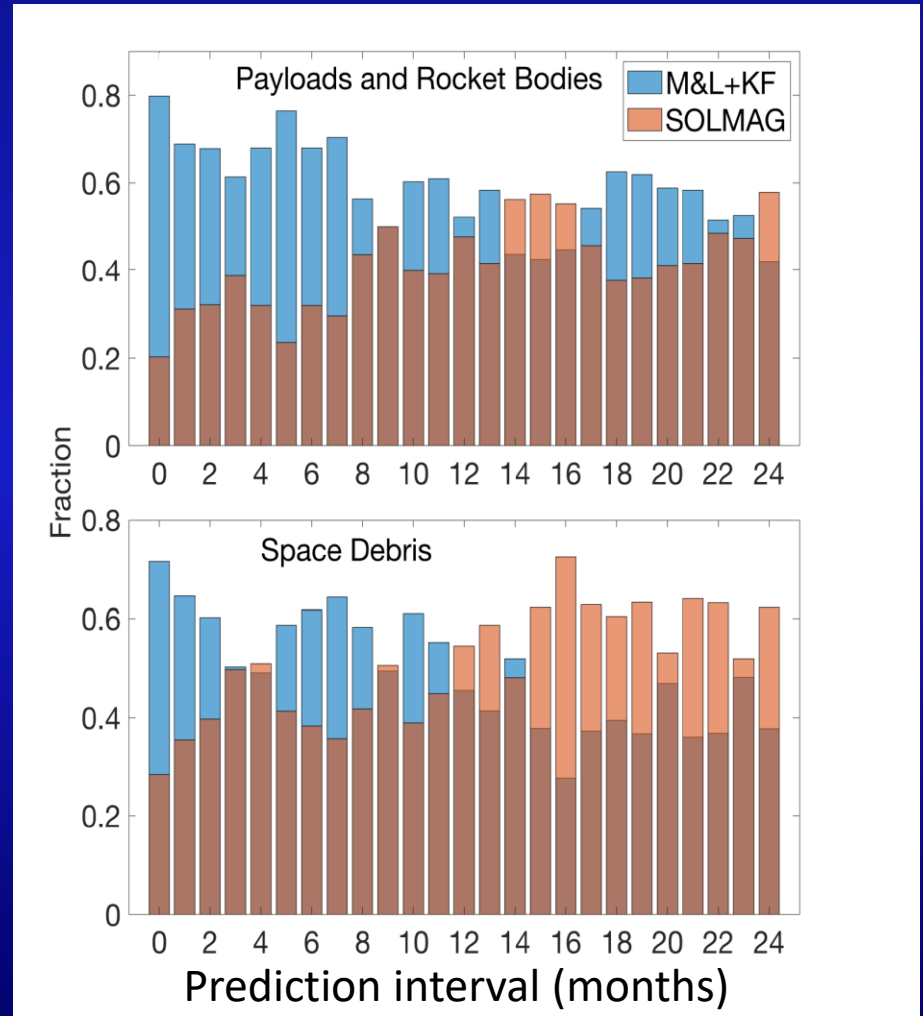
16% of cases have the absolute error of re-entry prediction more than **365 days**

61% of cases have the absolute error of re-entry prediction more than **365 days**

Comparison of re-entry predictions using predicted F10.7 as an input

Fraction of all the forecasts, where

- RESONANCE (M&L+KF) outperforms
- SOLMAG outperforms
- non-advantageous



RESONANCE (M&L+KF) provides a larger fraction of more accurate re-entry forecast compared to the SOLMAG method.

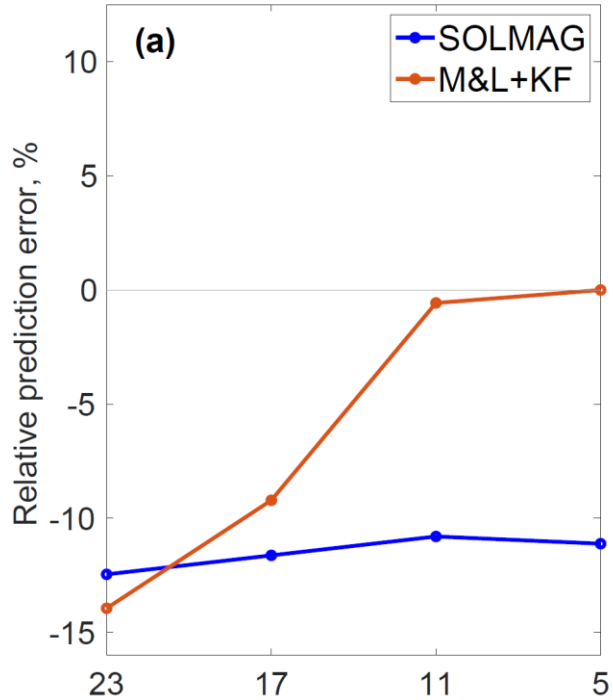
Majority of predicted re-entries are observed at the second half of the cycle

Our approach provides a larger fraction of more accurate re-entry forecast compared to the SOLMAG method for the prediction lead times 1-11 months ahead.

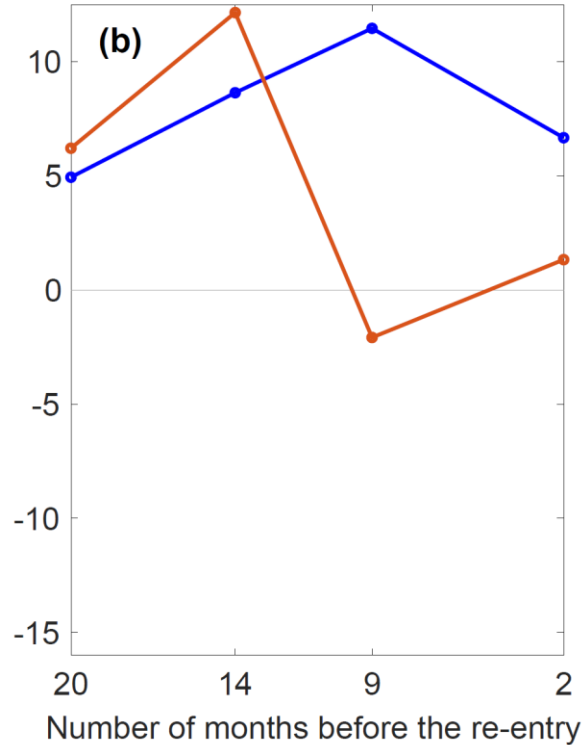
Majority of predicted re-entries are observed at the first half of the cycle

Example of re-entry predictions for three rocket bodies

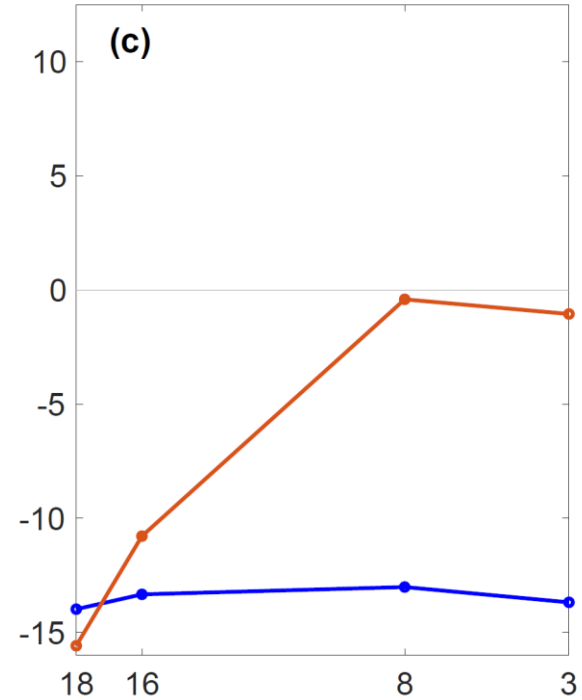
Second stage of a Delta 3 rocket



Centaur
the second stage of an Atlas 1



Third stage of a N-1 rocket



A negative error - predictions are earlier than the re-entry.



5-month lead forecast

Re-entry time: 29-Dec-2008

SOLMAG: 10-Dec-2008

RESONANCE (M&L+KF): 28-Dec-2008

Improvement – 18 days

9-month lead forecast

Re-entry time: 20-Oct-2014

SOLMAG: 22-Nov-2014

RESONANCE (M&L+KF): 14-Oct-2014

Improvement – 27 days

8-month-lead forecast

Re-entry time: 9-Mar-2009

SOLMAG: 5-Feb-2009

RESONANCE (M&L+KF): 8-Mar-2009

Improvement – 31 days



- ✓ **RESONANCE** statistically **outperforms** the currently used ESA SOLMAG method by 15.5-66.5% and in general leads to improvements of re-entry predictions, and thus can be recommended for the real-time implementation.
- ✓ The proposed approach based on **the Kalman filter** is **universal** and can be applied to improve the initial predictions of a process under study provided **by any other forecasting method**.
- ✓ **Re-entry predictions** with **F30** as an input could be more accurate, as **F30** may be a **better proxy** for **the atmosphere modeling** (Dudok de Wit et al. 2014), and it is **better predictable** than F10.7.
- ✓ **The further improvement** of the re-entry forecast should include both refinements of the atmospheric models and solar activity predictions.

Petrova E., T. Podladchikova T., A. Veronig, S. Lemmens, B. Bastida Virgili, T. Flohrer (2021), Medium-term predictions of F10.7 and F30 cm solar radio flux with the adaptive Kalman filter, The Astrophysical Journal Supplement Series, 254,9, [10.3847/1538-4365/abef6d](https://doi.org/10.3847/1538-4365/abef6d).

THANK YOU!