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#### Motivation

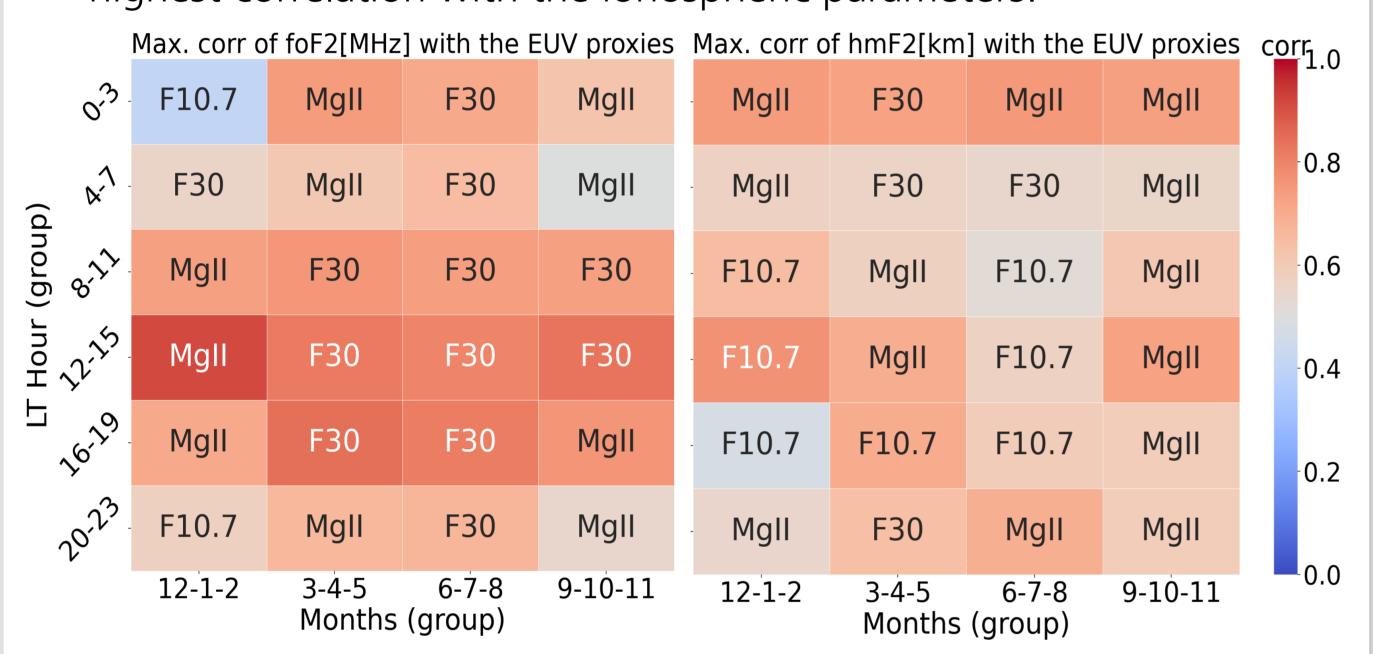
The ionospheric F2 layer plays a key role in radio wave propagation. It is characterized by the critical frequency (foF2) and the peak electron density height (hmF2), both of which are strongly influenced by solar activity. Accurately modeling their response to solar variability is essential for filtering out solar-driven effects and studying long-term trends. This work compares five models describing the response of foF2 and hmF2 to solar activity and evaluates different EUV proxies, selecting the most suitable one for subsequent long-term trend analysis.

## Objective

This work aims to perform a statistical comparison of different methods and EUV proxies in order to model the response of foF2 and hmF2 to solar activity

## **EUV** proxy comparison

An analysis of extreme ultraviolet (EUV) proxies, including MgII, F30, and F10.7, is carried out to identify the proxy with the highest correlation with the ionospheric parameters.



Heatmap of the maximum correlations between the EUV proxies and the ionospheric parameters (foF2 on the left and hmF2 on the right). The EUV proxy achieving the highest correlation is indicated in each panel,.

# Model comparison

A comparison of five methods for modeling the impact of solar activity on foF2 and hmF2 is conducted using hourly data from Juliusruh. The models considered are three polynomial regressions based on data clustered by solar cycle, month and hour of the day, and two Fourier Series applied to data clustered by hour of the day.

Models	Method (X=EUV PROXY)
Lineal Regression (1)	y = aX + b
Quadratic Regression (2)	$y = aX^2 + bX + c$
Cubic Regression (3)	$y = aX^3 + bX^2 + cX + d$
2 <sup>nd</sup> order Fourier Series (4)	$y = \sum_{n=0}^{2} \left[ (a_n X^3 + b_n X^2 + c_n X + d_n) \cos\left(\frac{2\pi n}{365} DoY\right) + (e_n X^3 + f_n X^2 + g_n X + h_n) \sin\left(\frac{2\pi n}{365} DoY\right) \right]$
3 <sup>rd</sup> order Fourier Series (5)	$y = \sum_{n=0}^{3} \left[ (a_n X^3 + b_n X^2 + c_n X + d_n) \cos\left(\frac{2\pi n}{365} DoY\right) \right] + (e_n X^3 + f_n X^2 + g_n X + h_n) \sin\left(\frac{2\pi n}{365} DoY\right) \right]$

The relation between two statistical tools such as the coefficient of determination (R<sup>2</sup>) and mean absolute percentage error (MAPE) is employed for a statistical comparison of the five methods.

Heatmap of the highest W values for each monthly-hourly group for hmF2. The model with the highest W value for each group is indicated, in each panel, using a number indicated in the table

Month

Also the Akaike Information Criterion (AIC), and Bayesian Information Criterion (BIC) are employed to study the balance between model complexity and explanatory power; and select the most appropriate method in these terms.

$$AIC = -2n * ln \left(\frac{SSE}{n}\right) + 2k$$

$$BIC = -2n * ln \left(\frac{SSE}{n}\right) + k * ln(n)$$

$$Nethod with the min AIC for hmF2[km]$$

$$Nethod with the min BIC for hmF2[km]$$

Colormaps for the comparison between polynomial regressions for AIC (left) and BIC (right) for hmF2. The color shown in each panel corresponds to the model with which the minimum value of AIC and BIC is achieved for that monthly-hourly group (red: linear regression, green: quadratic regression, blue: cubic regression)

#### Conclusion

The results indicate that F30 has the highest correlation with foF2 as an EUV proxy, while MgII shows the strongest correlation with hmF2. The third-degree polynomial regression is found to be the most effective method for modeling the ionospheric response to solar activity, based on a statistical analysis of the balance between model complexity and explanatory power.

#### References:

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