



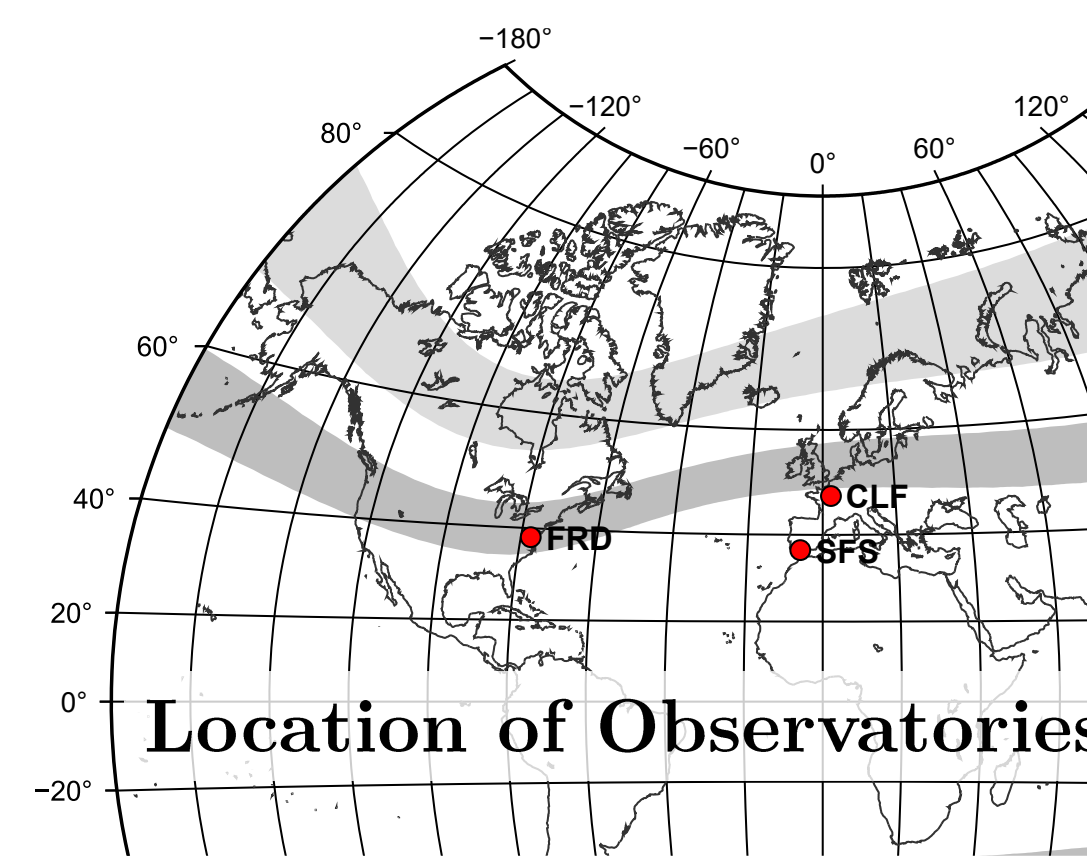
# Extraction of Solar Forcing Signatures in Ground Magnetometer Data from Sub-Auroral Regions

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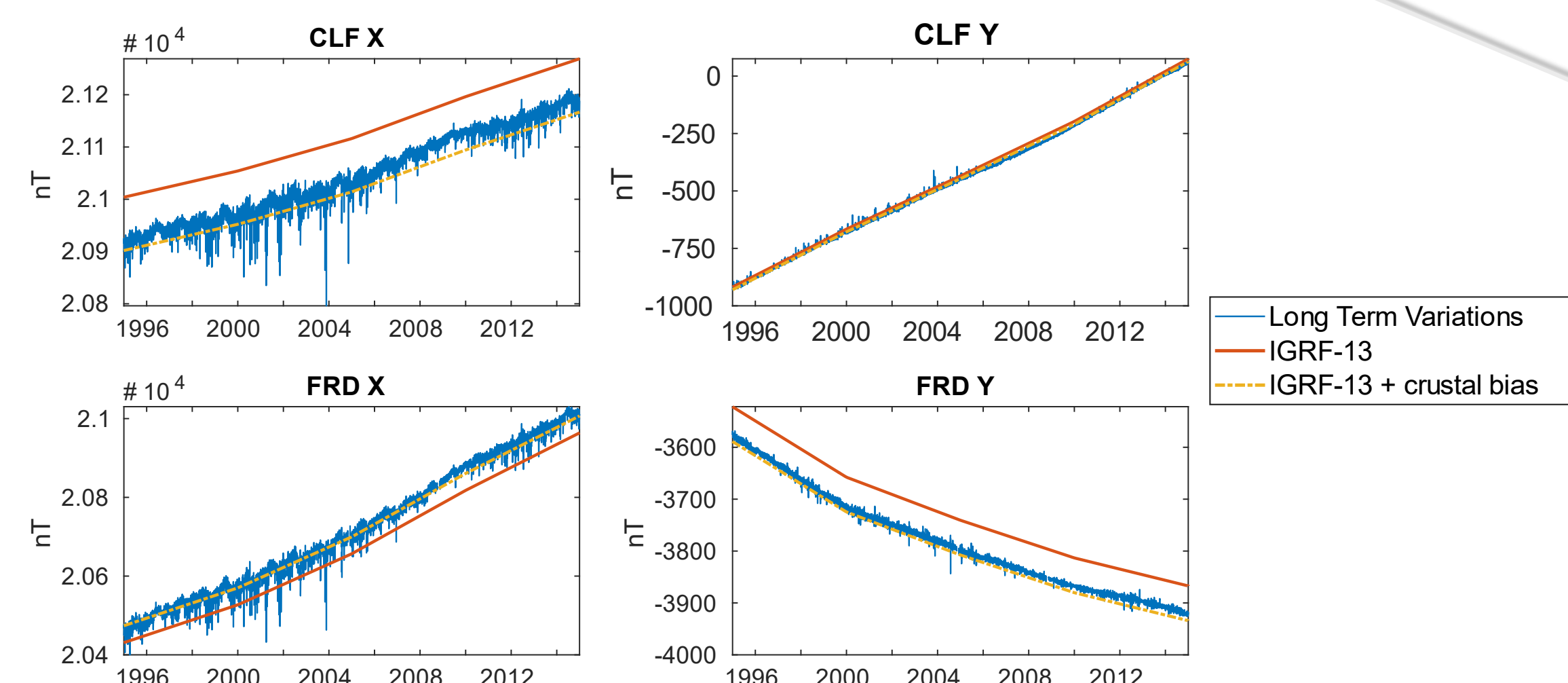


Location of Observatories

## Introduction

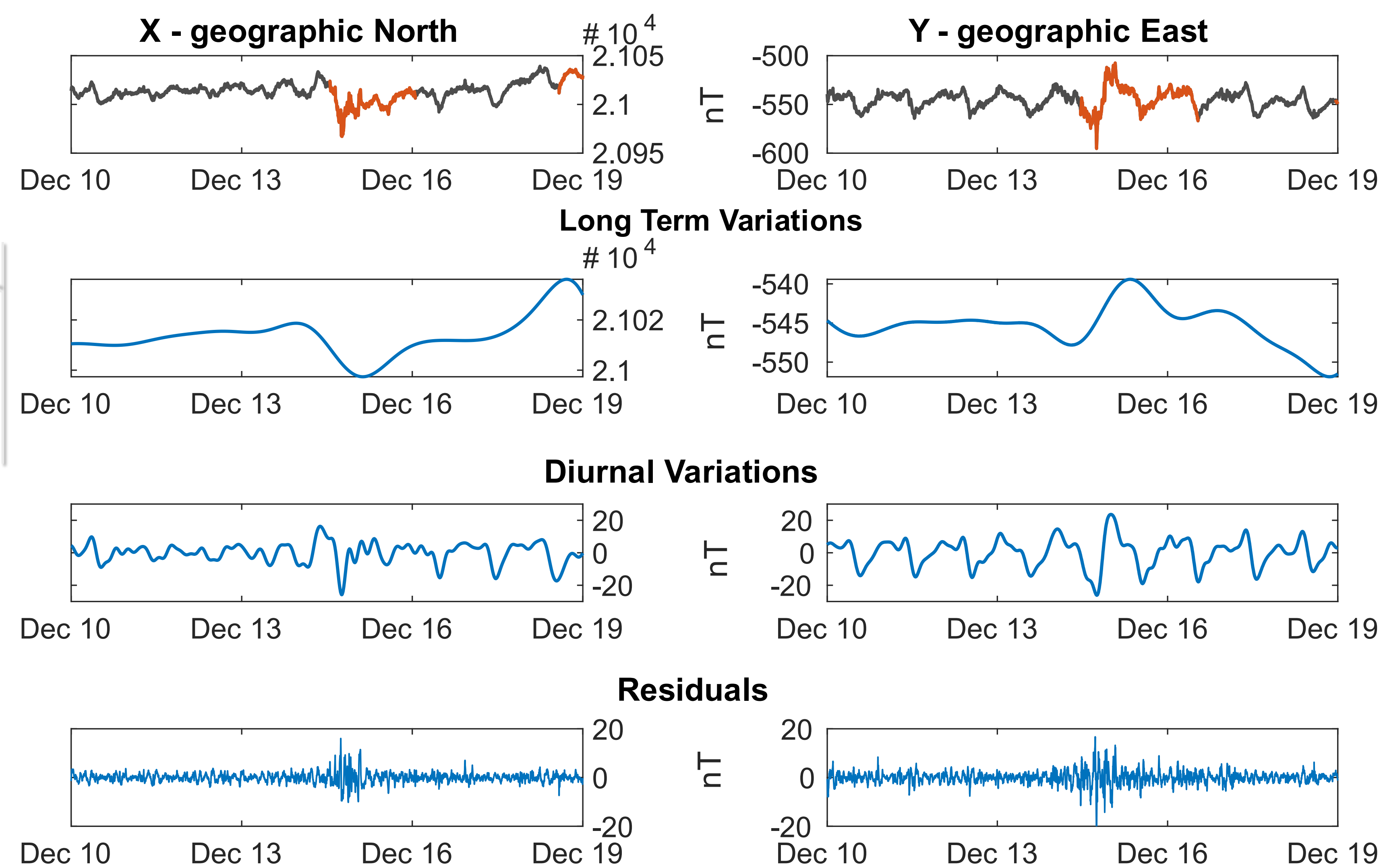
Strong solar events that reach Earth can significantly disturb the geomagnetic field. As the recorded magnetic signal is a superposition of sources that act on a broad amplitude-frequency spectrum, **the isolation of storm signatures within geomagnetic field measurements** becomes an important and nontrivial task. In order to determine the strength in real-time, **the identification of storms** needs to be automated.

### Long Term Variations

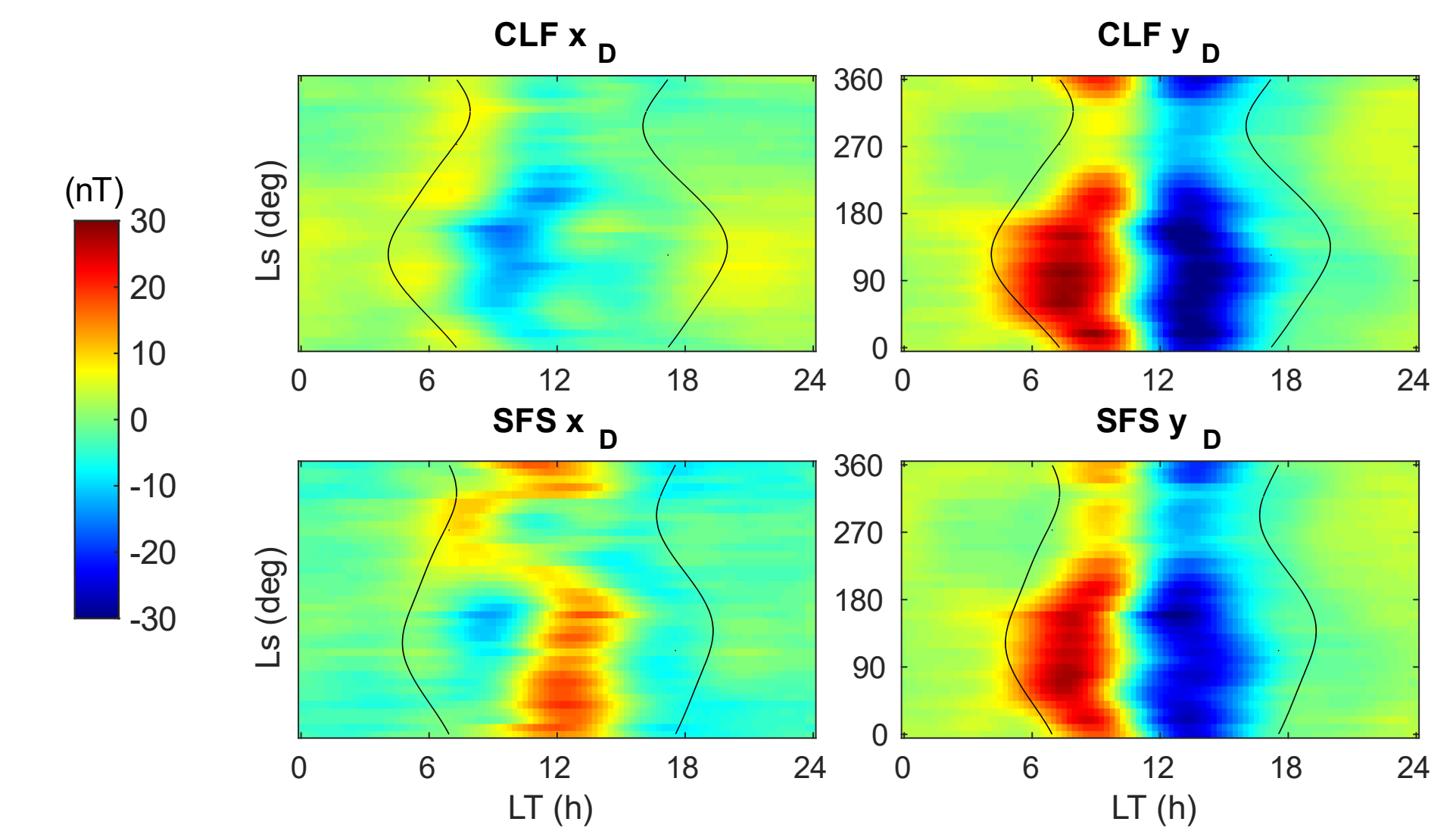


The overall trend of the extracted long term variations is in **good agreement with the IGRF model** (Alken et al. 2021), indicating that the filter catches **the secular variation**. Contributions of other internal and external sources are also present.

— **Geomagnetic Field Measurement**  
— **Storm Signature**  
— **Source Extraction**



### Diurnal Variations

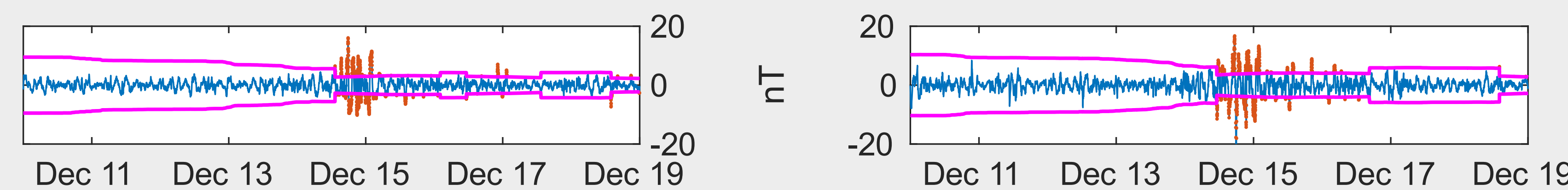


The superposed epoch analysis of diurnal variations in dependence of solar longitude and local time is generated over geomagnetically quiet periods. The characteristics are in good agreement with the **solar quiet current system** (Haberle et al. 2022).

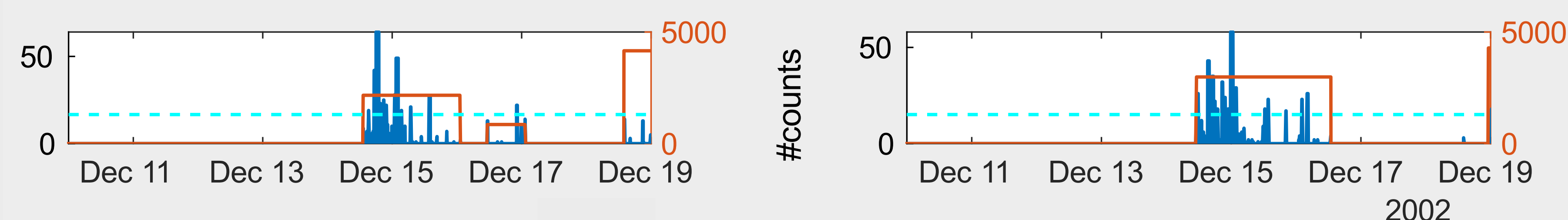
## Extraction of Storm Signatures

Removing long-term and diurnal variations from geomagnetic field measurements results in **the residuals containing predominantly solar forcing contributions**. These can then be used to characterise disturbances within the geomagnetic field. The automatic identification of storm signatures is done in two steps. First, typical quiet levels are determined, followed by the counting of non-quiet residuals per time interval.

### Step 1:



### Step 2:



### Step 1: Determining quiet levels

A **quiet-time expectation boundary** is defined by taking into account the previous three months. Residuals outside this boundary are considered **storm candidates**.

### Step 2: Defining storm intervals

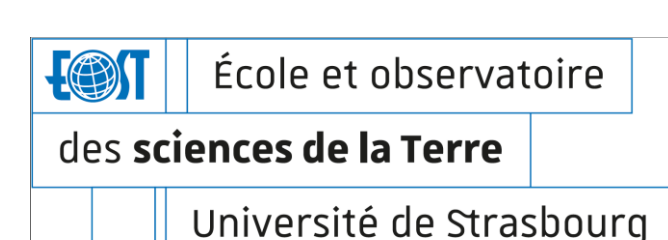
**Zero-crossings** for storm candidates between upper and lower boundaries are taken care of. A **minimum storm time duration threshold** defines storm signatures.

## References

Alken, P. et al. Earth Planets Space 73, 49 (2021).  
<https://doi.org/10.1186/s40623-020-01288-x>  
Haberle, V. et al. (2022) Journal of Geophysical Research: Space Physics, 127. <https://doi.org/10.1029/2022JA030407>

## Conclusion and Perspective

With these traditional methods stronger storms are identified correctly including recovery phases. Sometimes shorter, quiet periods are mistakenly marked as a storm but tuning of parameters is time-consuming.  
Can artificial intelligence help in finding optimal parameters and improve storm identification?



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More questions? Reach out!  
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