

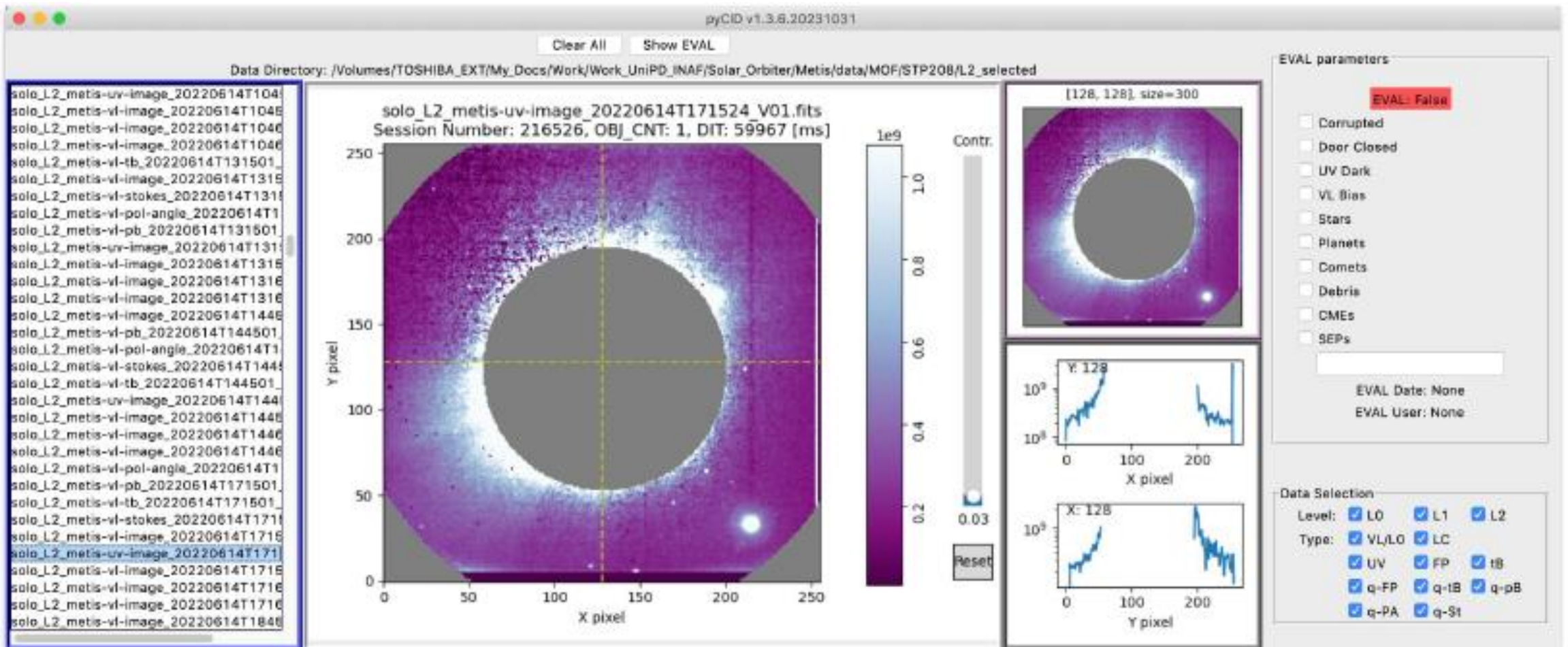
Contribution to the pyCID GUI: a semi-automatic CME detection tool



Gianluca Napoletano, Università di Firenze

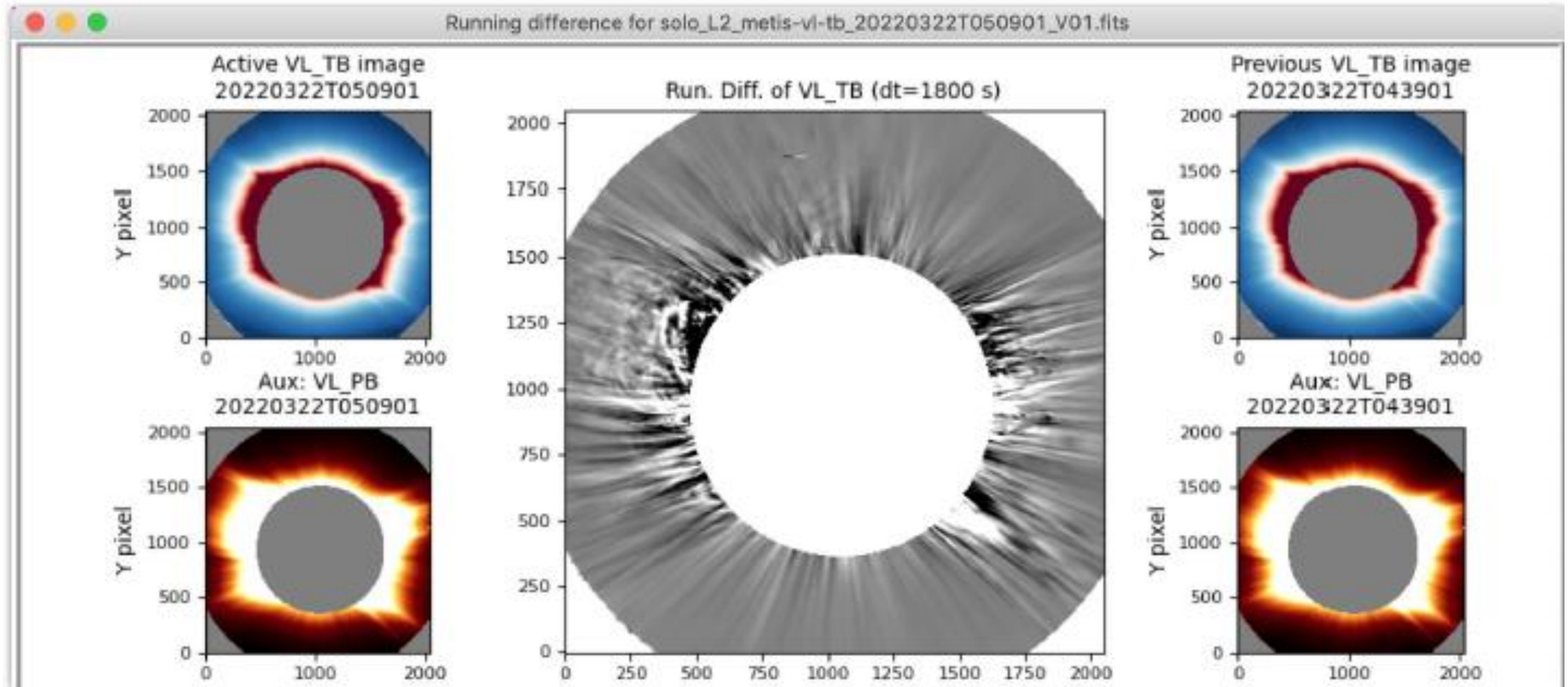
What is pyCID?

pyCID is a quick-look GUI tool for visualization and validation of Metis data of different types (images, light curves) and levels (L0, L1 and L2). It was inspired by the more sophisticated IDL tool iCID, it is written in Python and was presented during the previous METIS meeting by A. Burtovoi



What is pyCID?

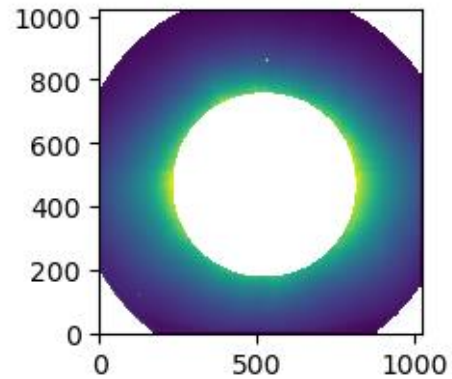
It makes use of the SunPy affiliated package “sunkit-image” to implement additional tools like coordinate transformation and filters for image enhancement and processing



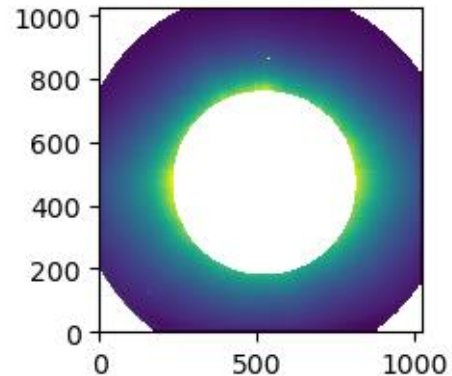
Running differences

- Running differences enhance moving features in the FOV such as Coronal Mass Ejections

L2_metis-v1-tb_20231214T193001

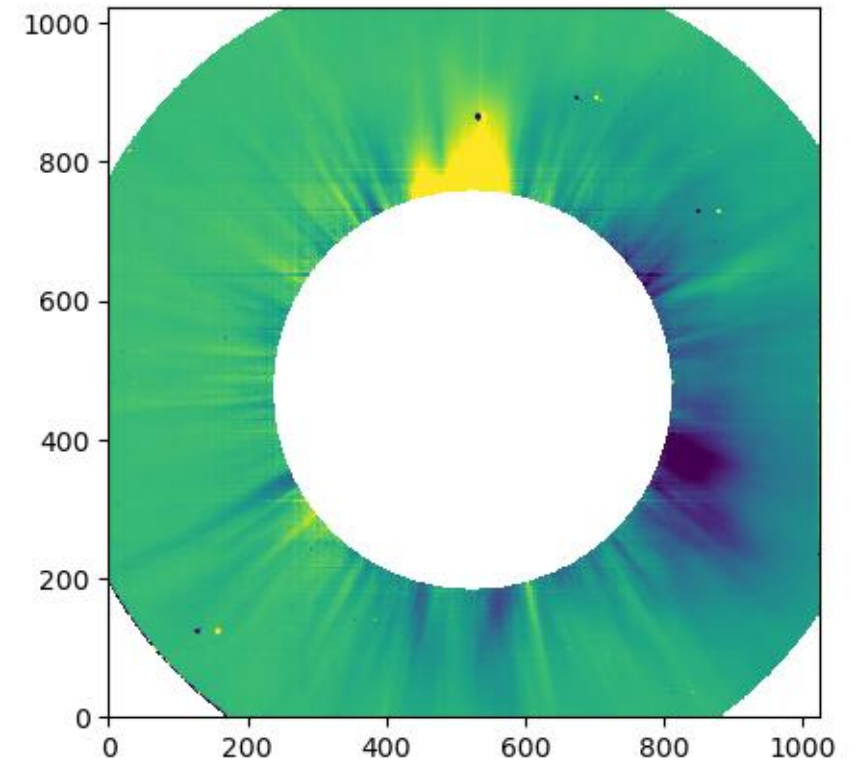


L2_metis-v1-tb_20231215T002957



zero-mean running difference

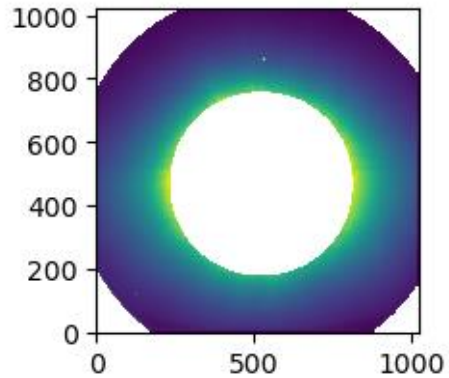
$$u = n_i - n_{i-1} \left(\frac{\bar{n}_i}{\bar{n}_{i-1}} \right)$$



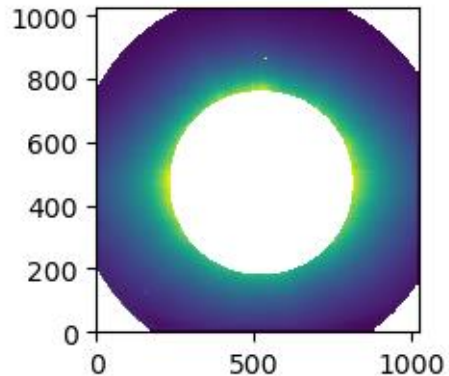
Running differences

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L2_metis-v1-tb_20231214T193001



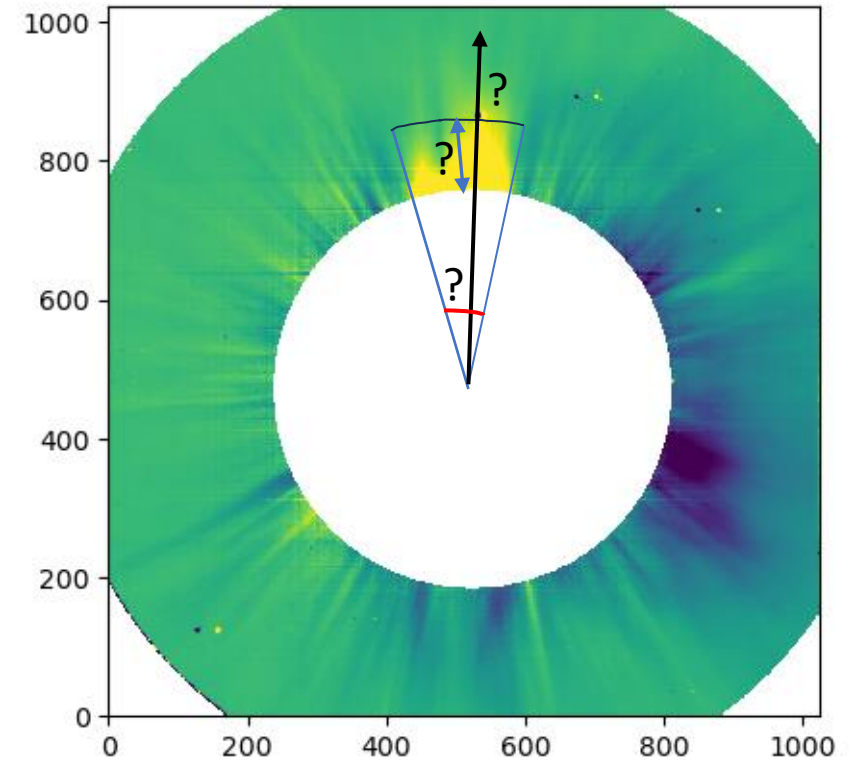
L2_metis-v1-tb_20231215T002957



zero-mean running difference

$$u = n_i - n_{i-1} \left(\frac{\bar{n}_i}{\bar{n}_{i-1}} \right)$$

Goal: provide an additional function to the pyCYD GUI to assist the CME hunters in assessing additional information, such as CME width, position angle, main direction and height;



Algorithm steps

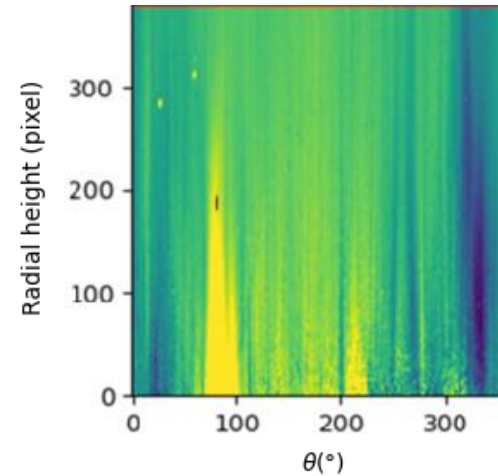
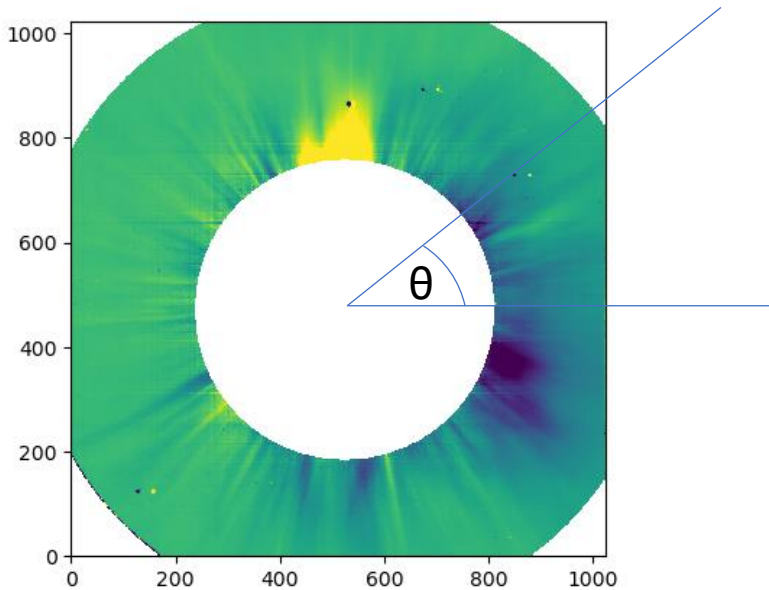
- The main idea is adapted from the CACTUS fully automatic CME detection algorithm employed to image sequences from LASCO (see Robbrecht et al. 2004, Olmedo et al. 2007, Robbrecht et al. 2009).
- We opted for a semi-automatic algorithm where two parameters are manually adjusted;

1. Conversion to polar coordinates

Algorithm steps

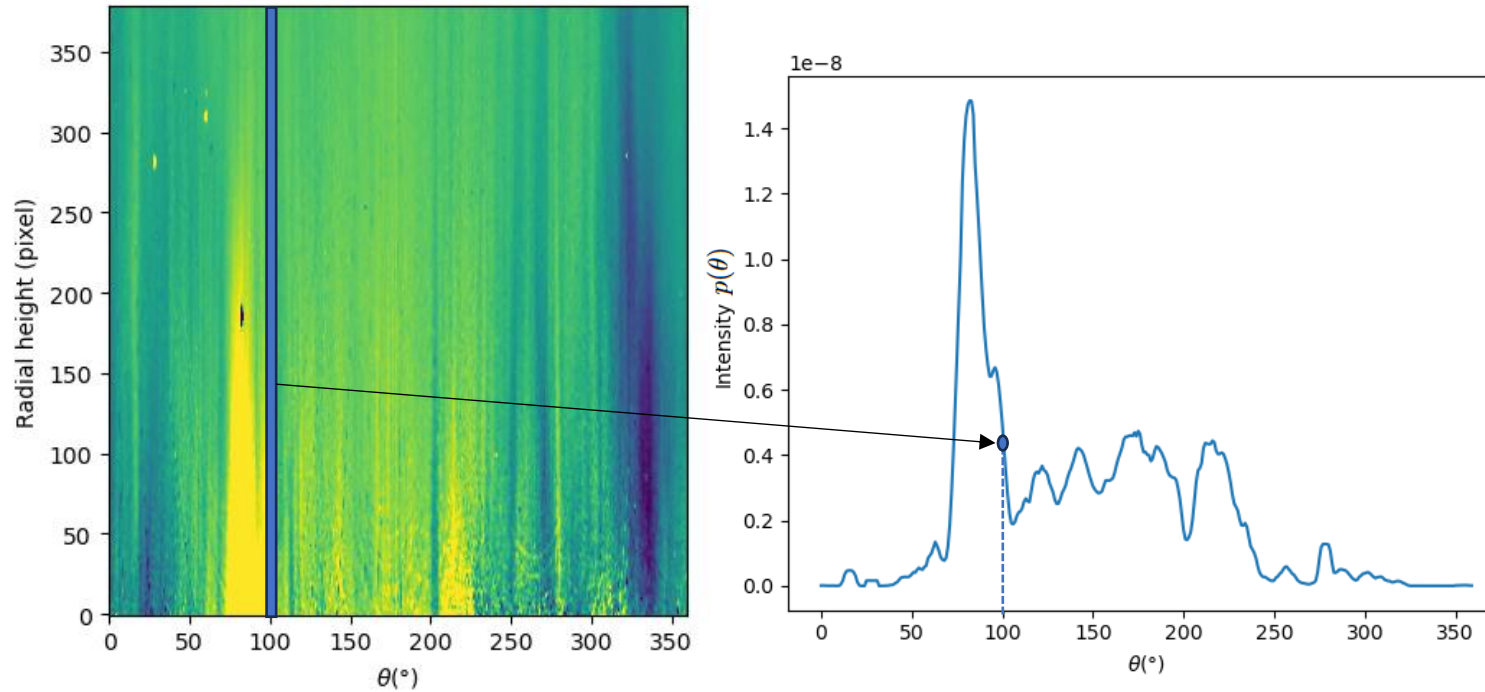
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1. Conversion to polar coordinates



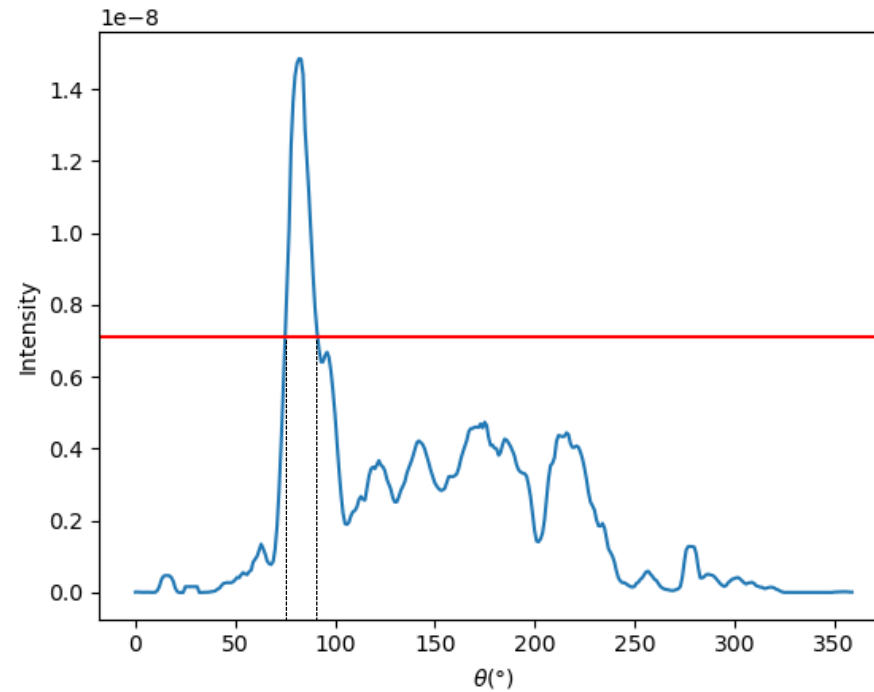
Algorithm steps

2. Compute angular intensity: for each angle (image column) positive pixels are added to construct an angular intensity map $p(\theta)$



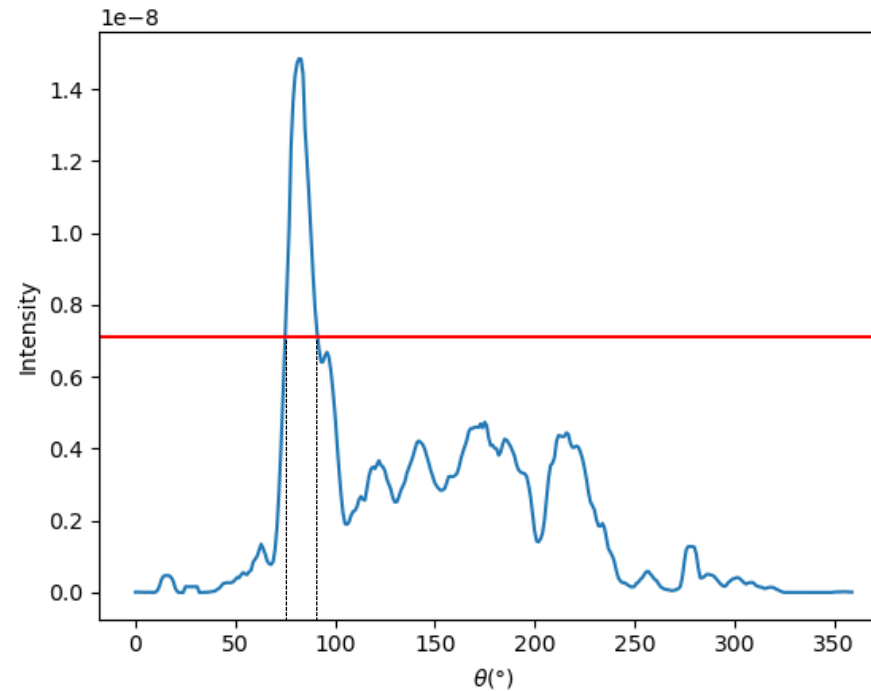
Algorithm steps

3. Define the threshold $T_1 = \bar{p} + N_1\sigma_p$ and look for intervals where $p(\theta)$ is above T1. This defines the first angular window detection where the CME is located (core region):



Algorithm steps

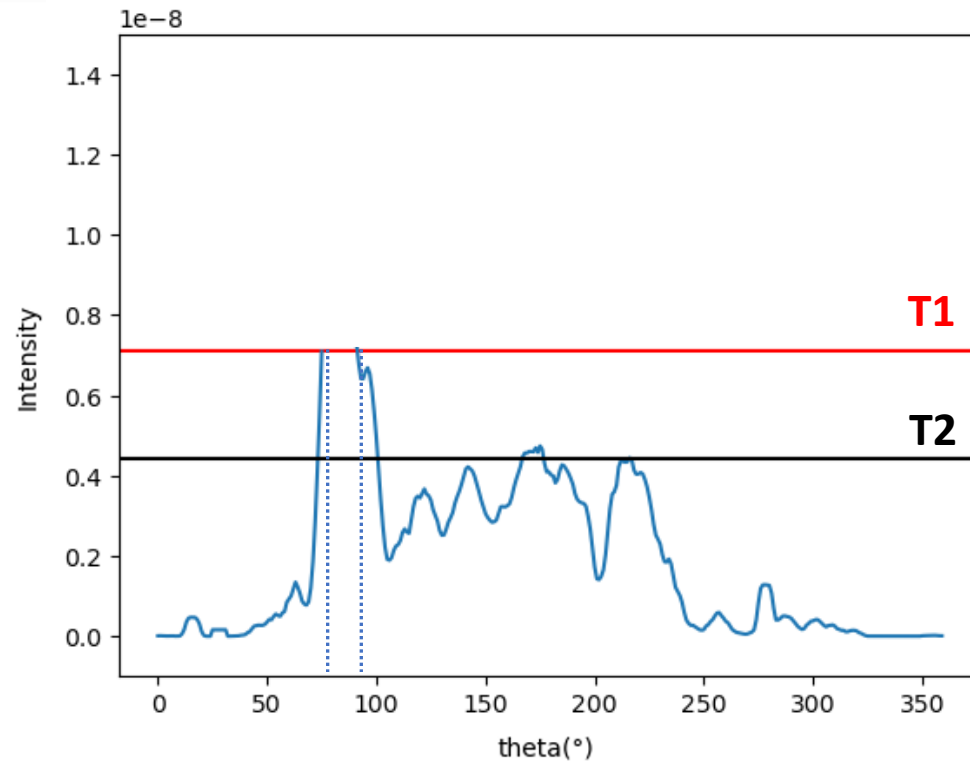
3. Define the threshold $T_1 = \bar{p} + N_1\sigma_p$ and look for intervals where $p(\theta)$ is above T1. This defines the first angular window detection where the CME is located (core region):



N_1 is the first user controlled parameter, defining where the brightest feature of the image will be. A lower threshold will result in a wider core region.

Algorithm steps

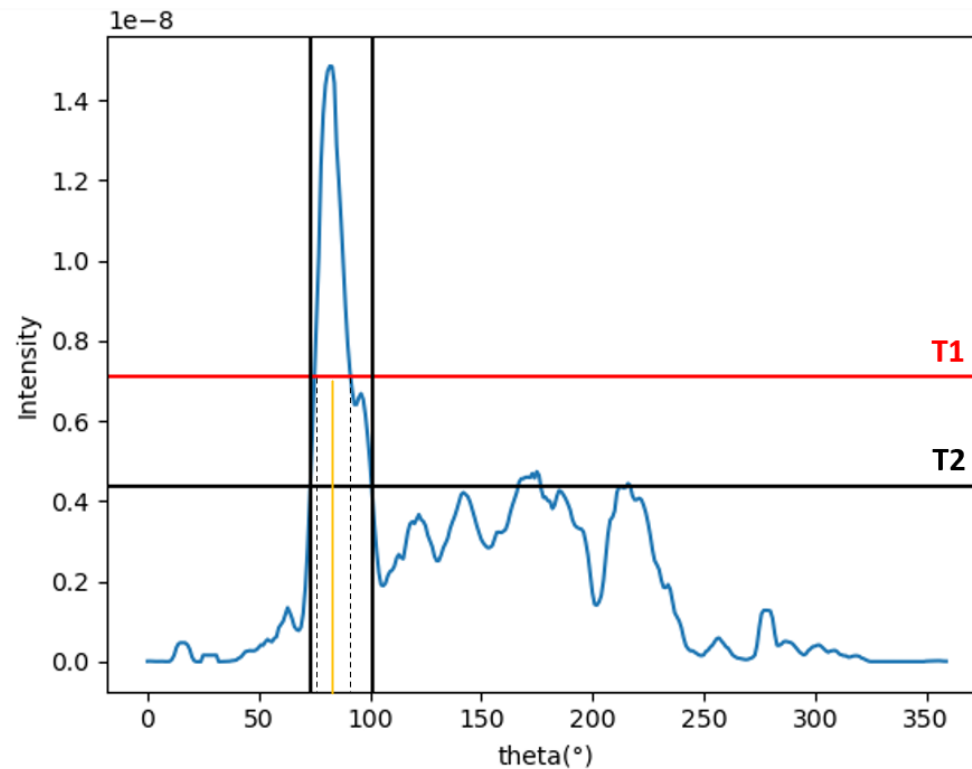
4. Increase the core region by setting a new threshold $T_2 = \overline{p^*} + N_2 \sigma_{p^*}$ where p^* is a reduced distribution which omits values of $p(\theta)$ within the core region.



Algorithm steps

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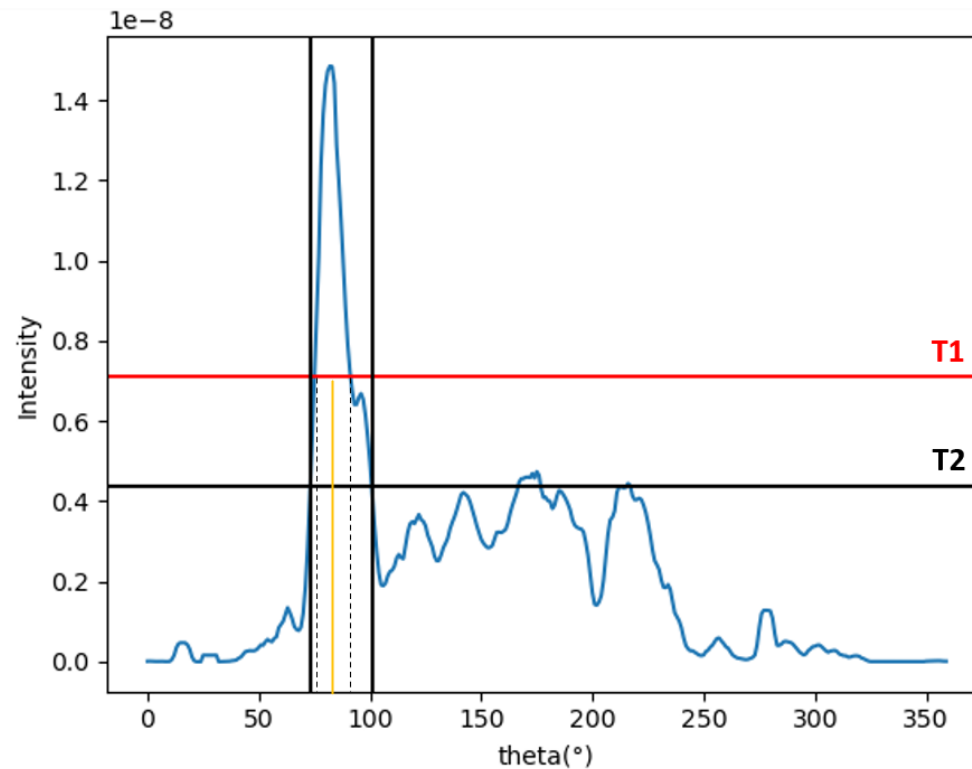
The core angle region is widened up to the angular positions where intensity falls below the new threshold T2.



Algorithm steps

4. Increase the core region by setting a new threshold $T_2 = \bar{p}^* + N_2 \sigma_{p^*}$ where p^* is a reduced distribution which omits values of $p(\theta)$ within the core region.

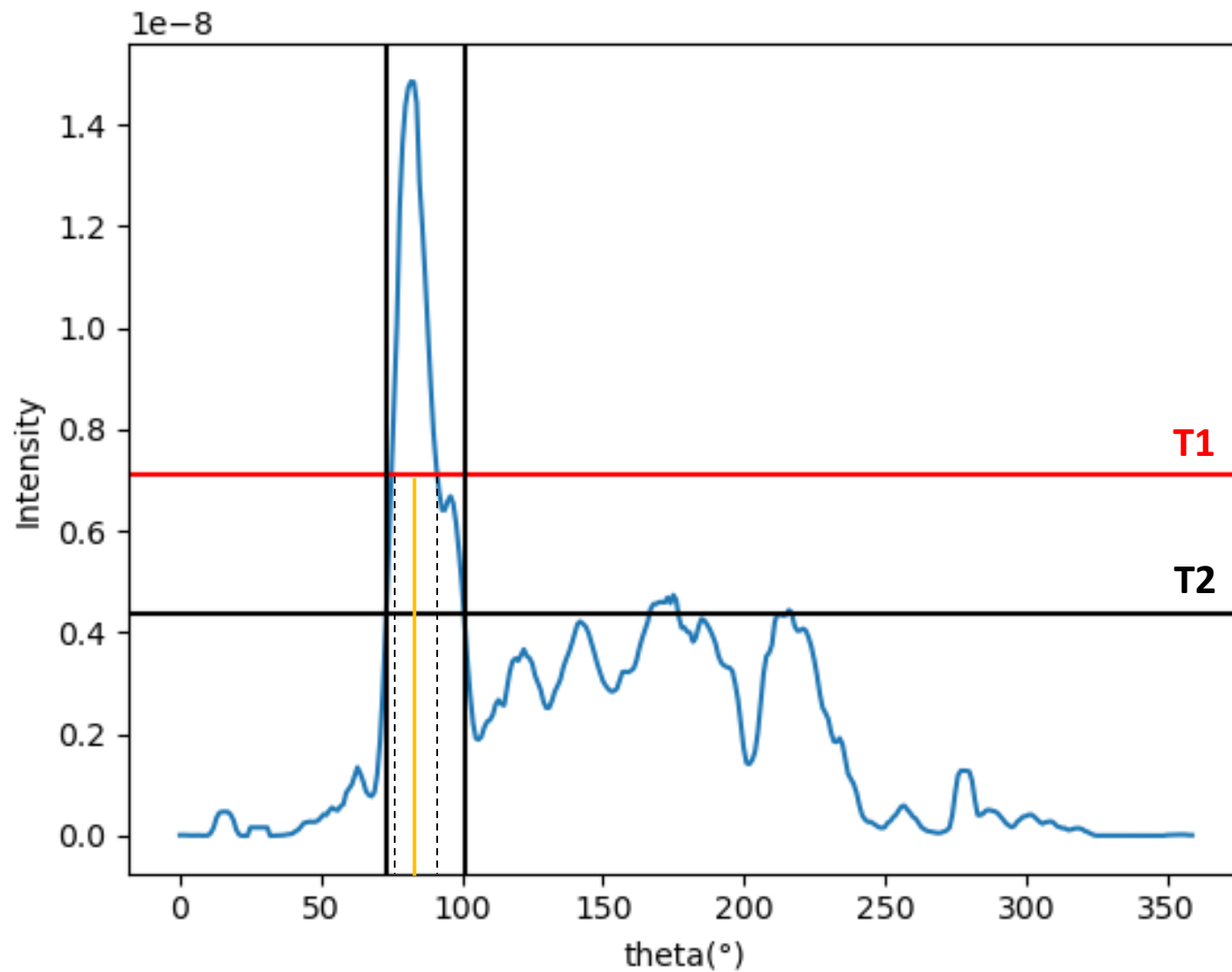
The core angle region is widened up to the angular positions where intensity falls below the new threshold T2.



N_2 is the second user controlled parameter, with the effect of widening the core region to establish the final CME angular window.

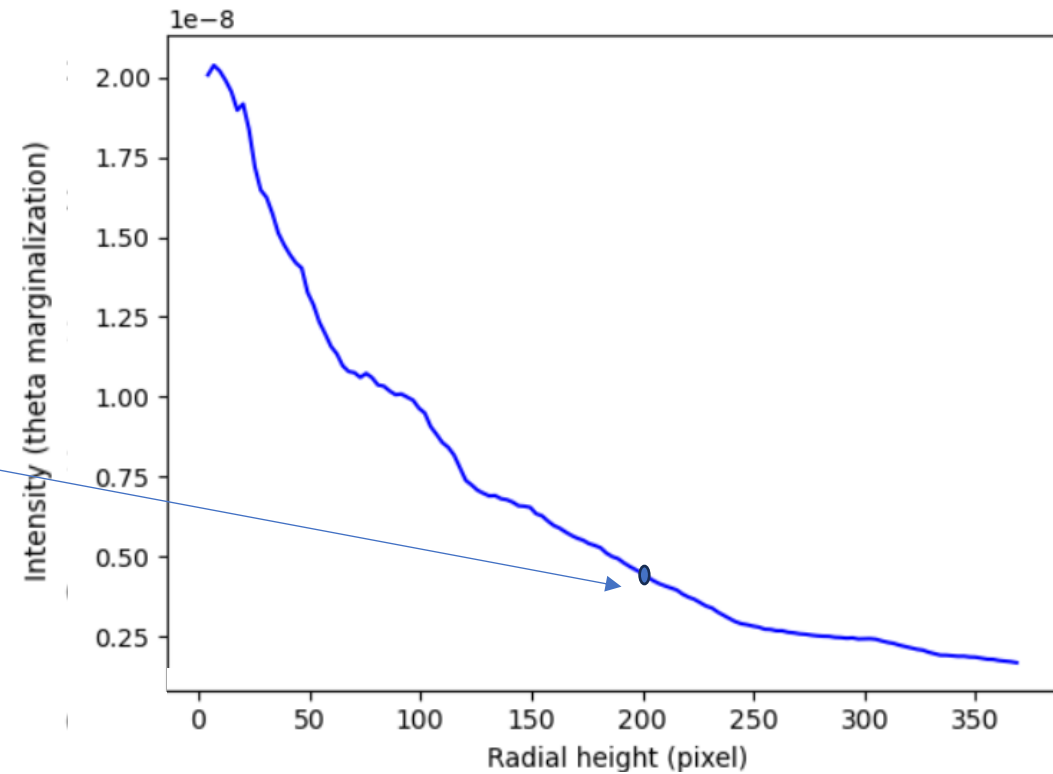
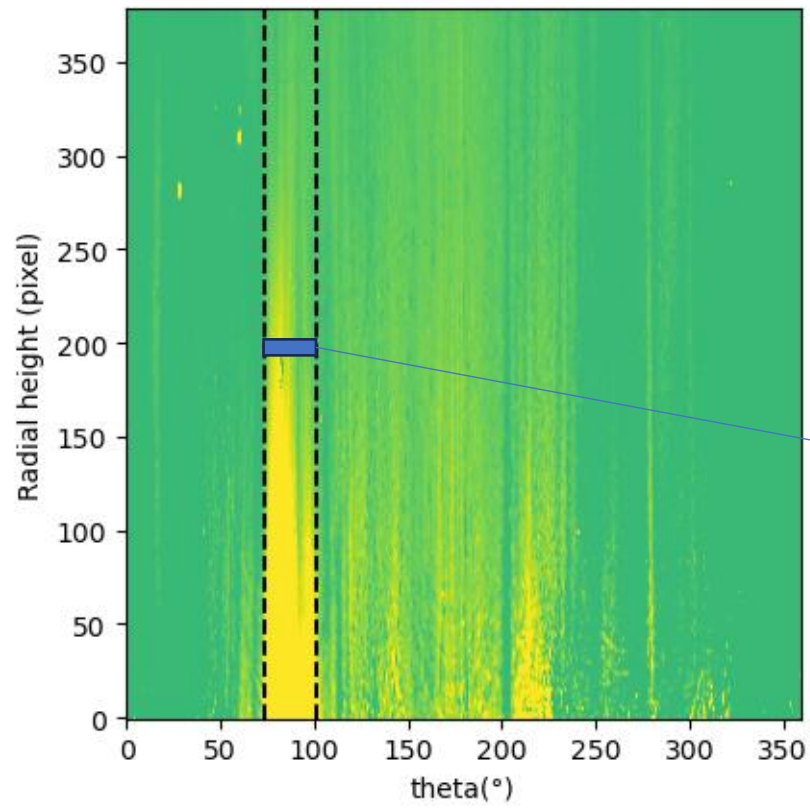
Algorithm steps

Core region center



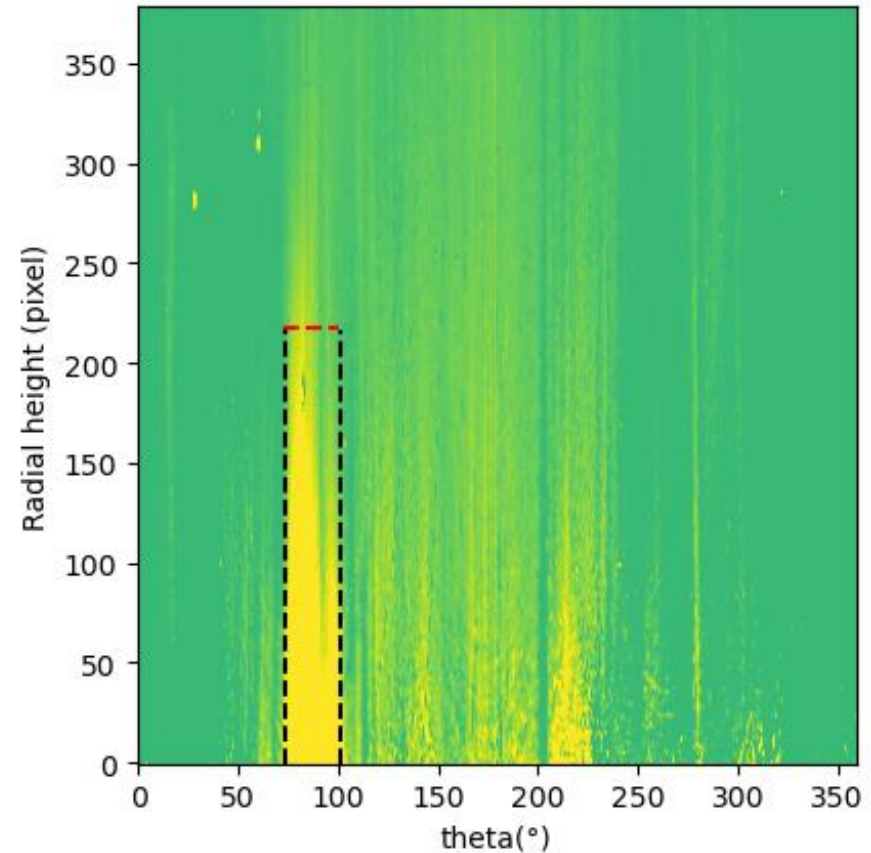
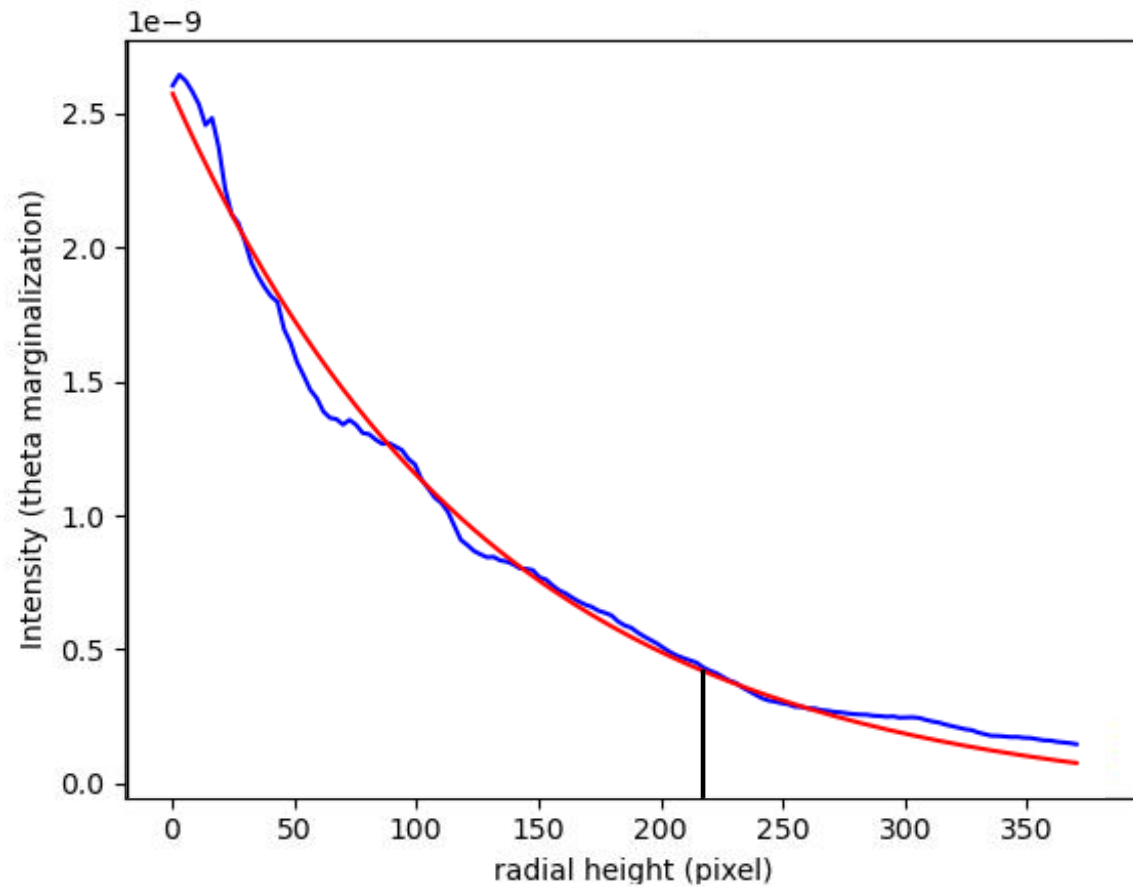
Algorithm steps

5. CME leading edge detection: within the CME core region only, fix radial height and sum the intensity values at all angles to define the intensity distribution



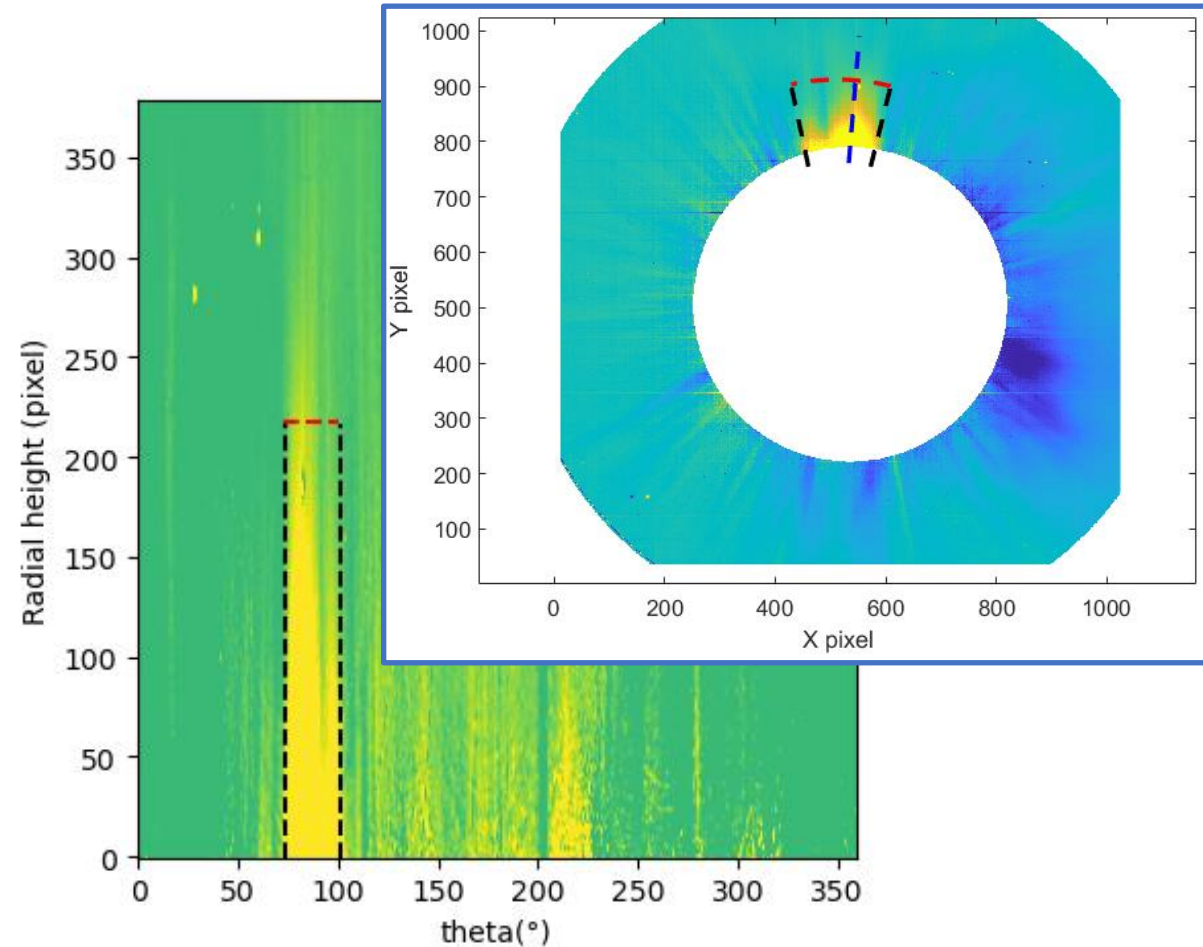
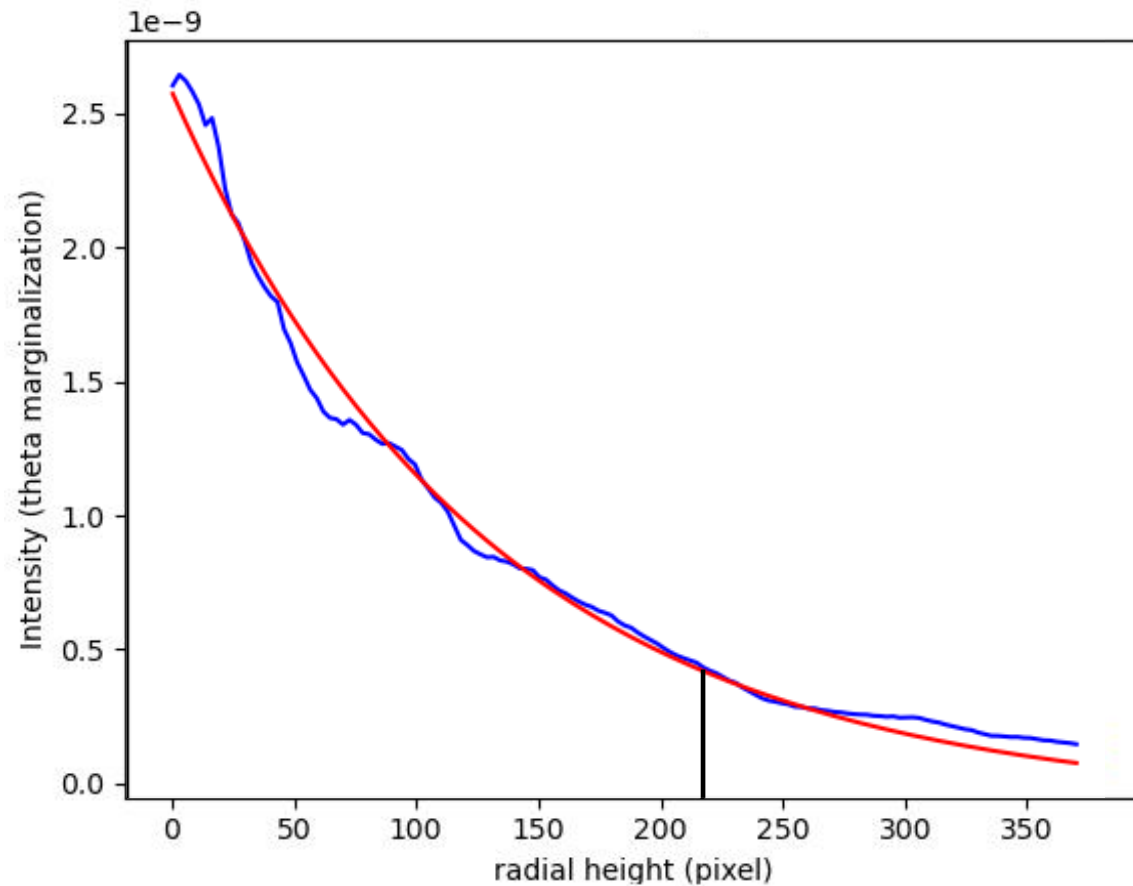
Algorithm steps

6. We then perform an exponential fit and set the CME height as 2 times the exponential scale length



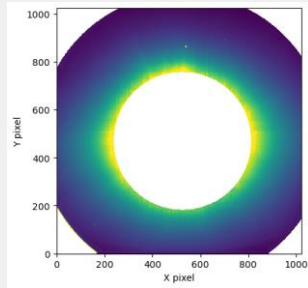
Algorithm steps

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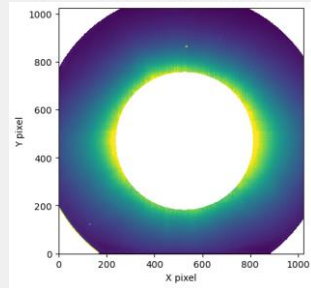


CME detector tool

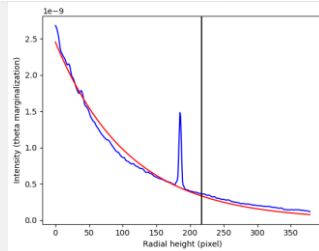
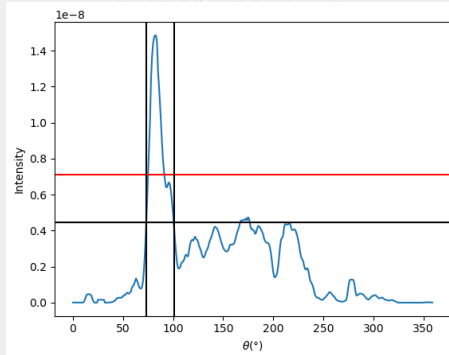
Active VL_TB image
20231215T003013



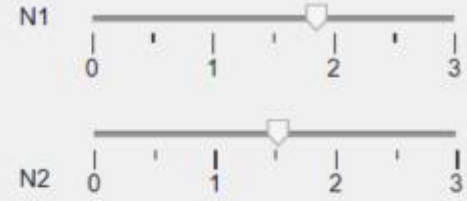
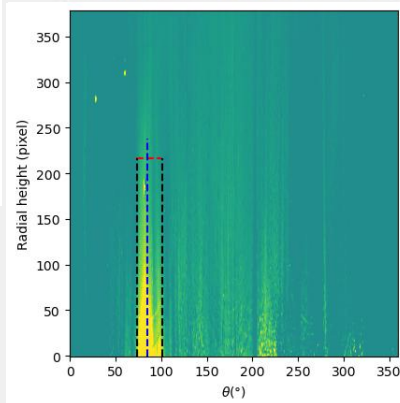
Previous VL_TB image
20231214T193001



Intensity distributions



CME detection on polar plane

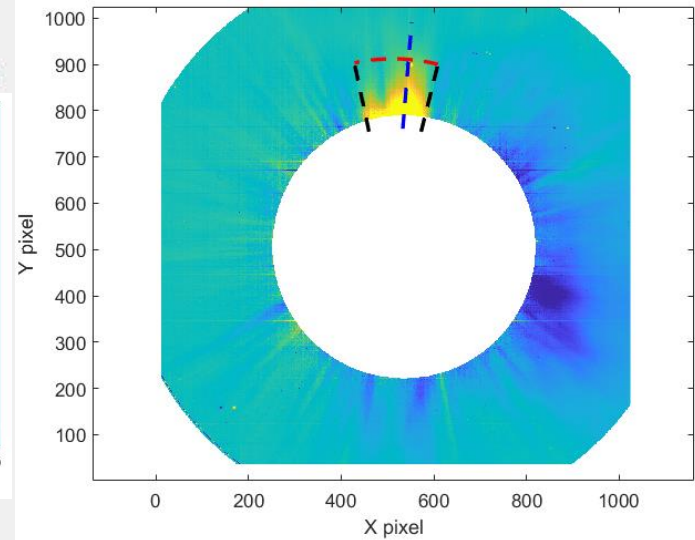


detected CME height (pixel)

detected CME angle ($^{\circ}$) CME width ($^{\circ}$)

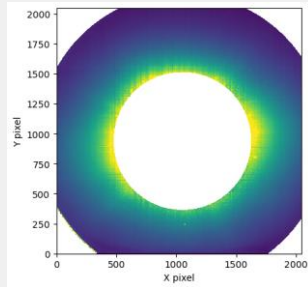
CME direction ($^{\circ}$)

Running difference ($\Delta t = 18012$ s)

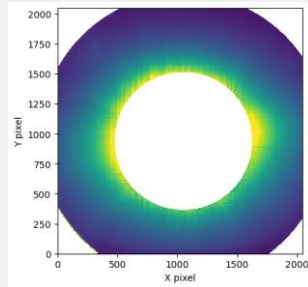


CME detector tool

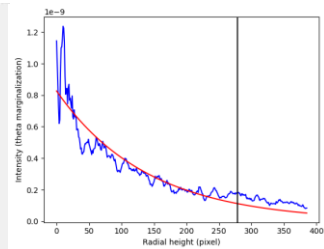
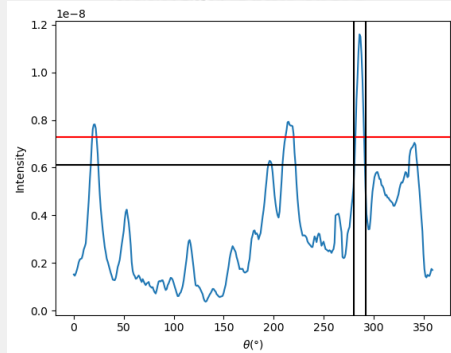
Active VL_TB image
20230620T213001



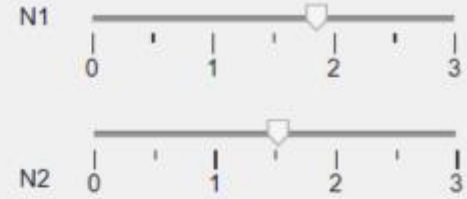
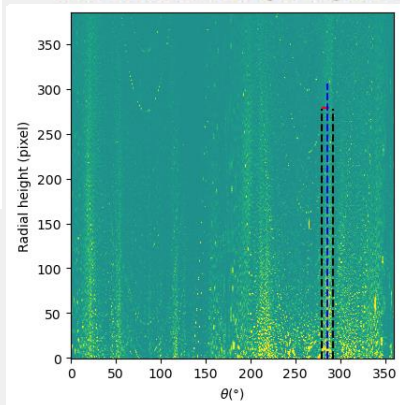
Previous VL_TB image
20230620T153001



Intensity distributions



CME detection on polar plane

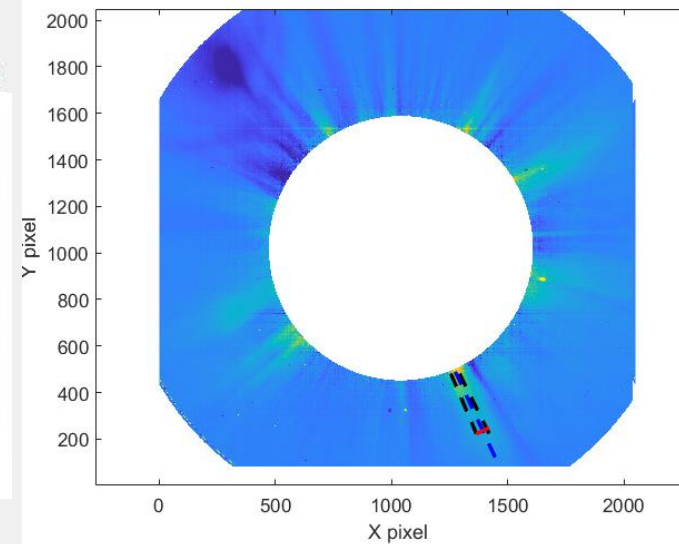


detected CME height (pixel)

detected CME angle (°) CME width (°)

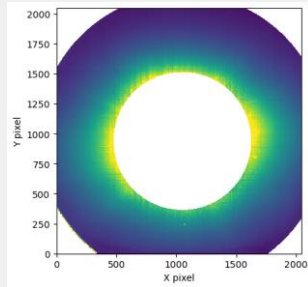
CME direction (°)

Running difference ($\Delta t = 21600$ s)

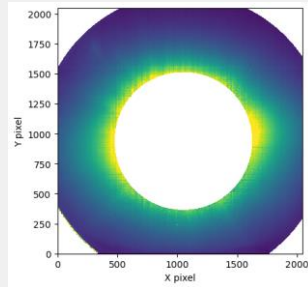


CME detector tool

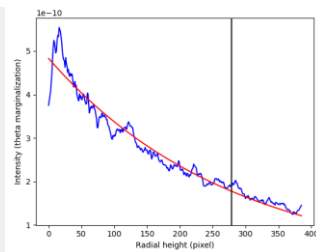
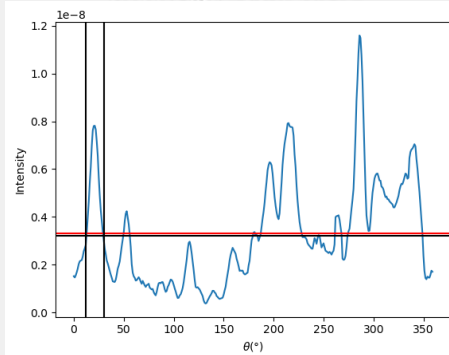
Active VL_TB image
20230620T213001



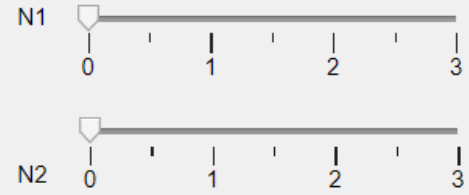
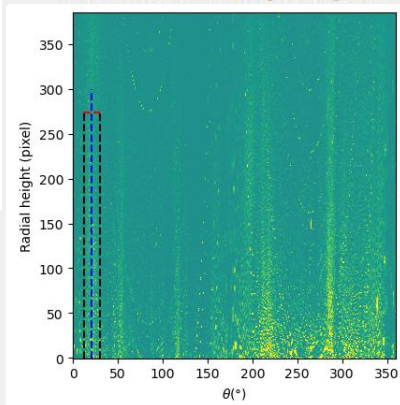
Previous VL_TB image
20230620T153001



Intensity distributions



CME detection on polar plane

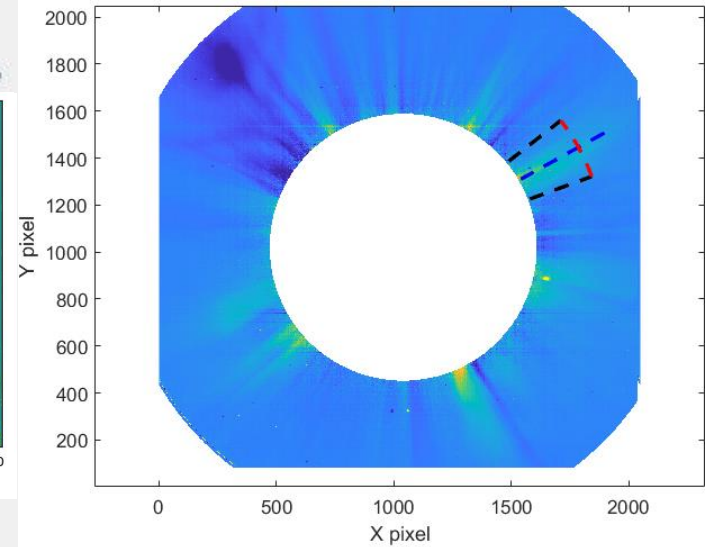


detected CME height (pixel)

detected CME angle (°) CME width (°)

CME direction (°)

Running difference ($\Delta t = 21600$ s)



To do & ongoing developement

- Regularly employ the CMEdetector to report bugs and improve it
- Adapt the algorithm behaviour to peculiar cases
- Improve the choice of the parameters (eventually make it fully automatic)

Possible applications

- CME kinematics, initial conditions for forecasting models
- CME deprojection from multiple Points of View

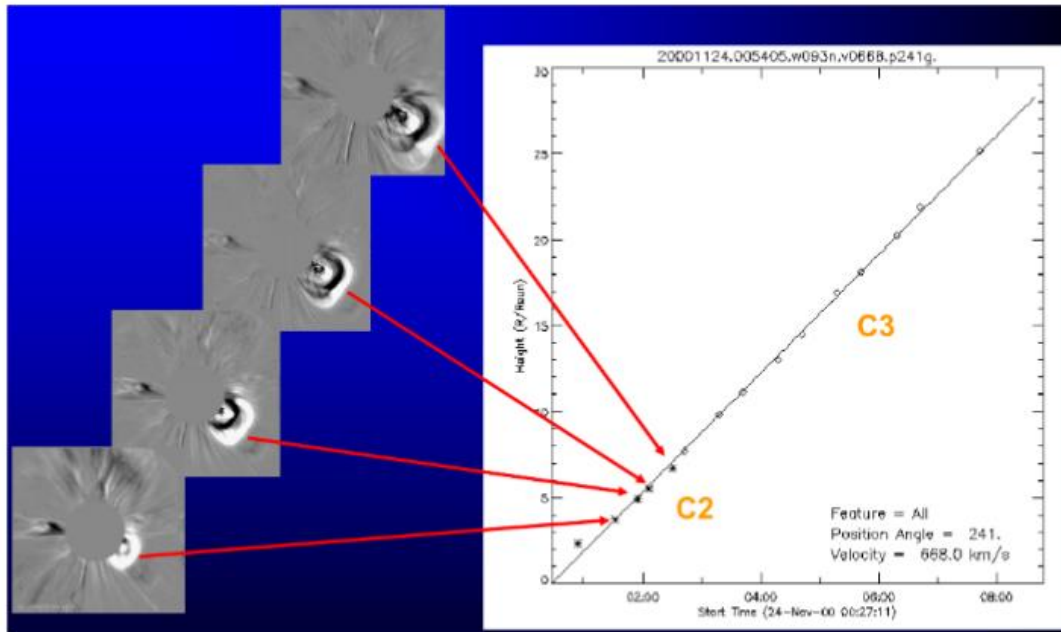
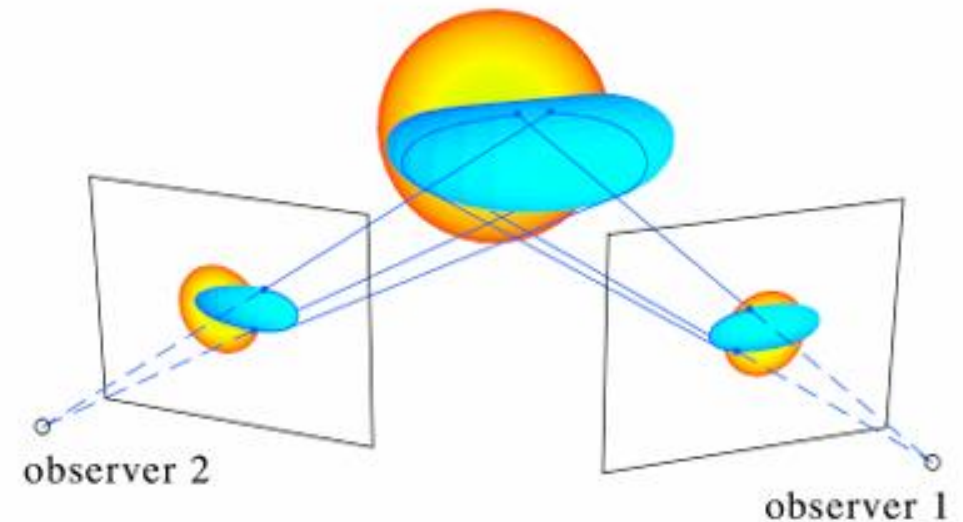


Figure 3.2: Height time measurements related to coronagraph observations and



CME detection in coronagraph FOV

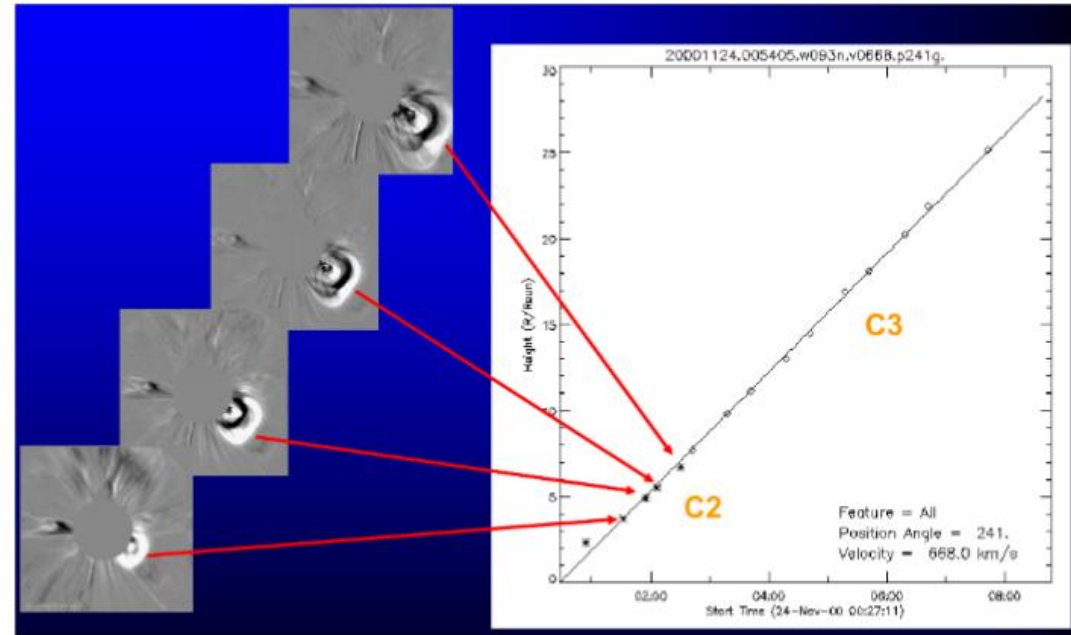
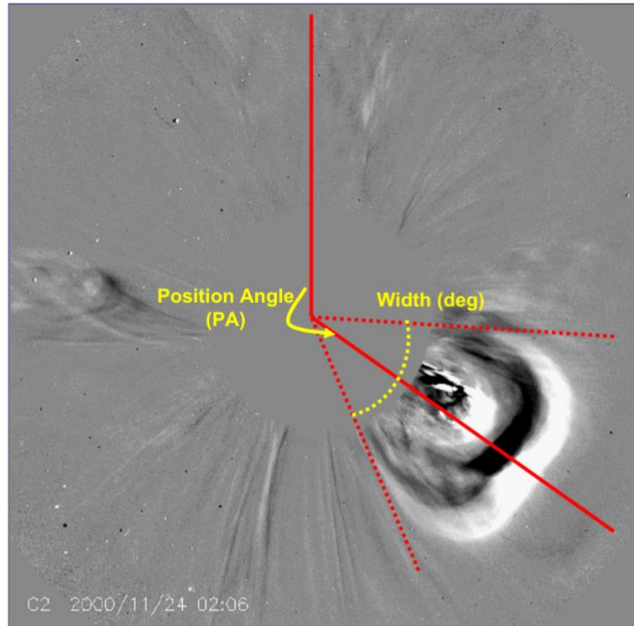


Figure 3.2: Height time measurements related to coronagraph observations and

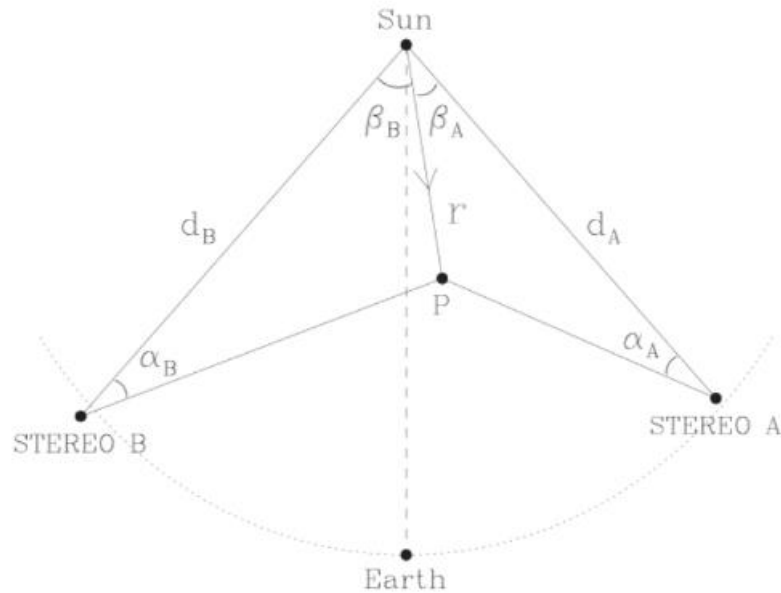
Features like CME height and width at each time are essential for:

- Studying erupting CME evolution in time
- Providing initial conditions to forecasting models
- Assess CME kinematics and morphology

The easiest deprojection model

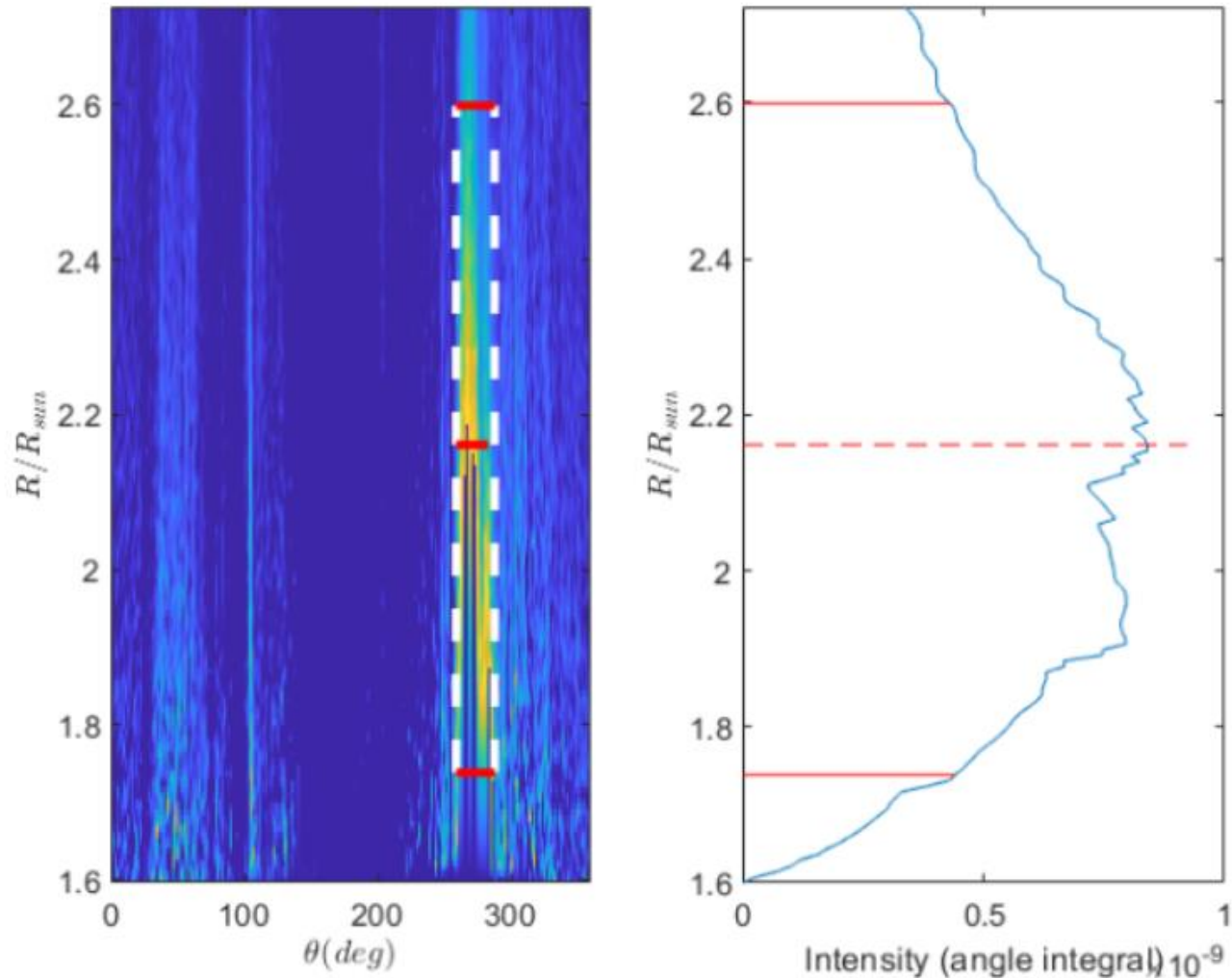
- We only take into account the farthest visible element of the detected CME
- We assume it is the same point (discuss error with angular separation between observers)

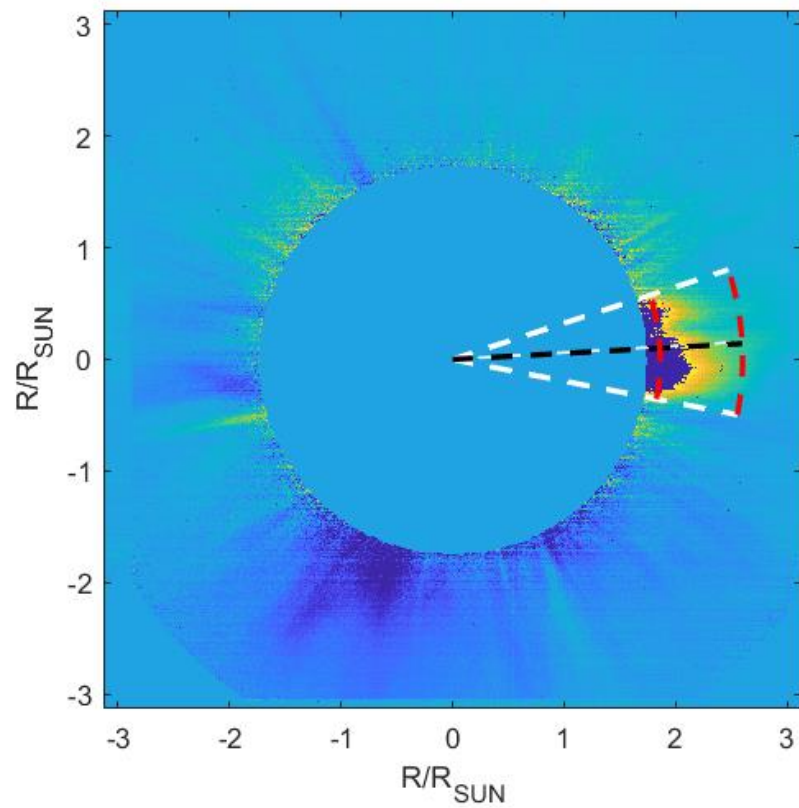
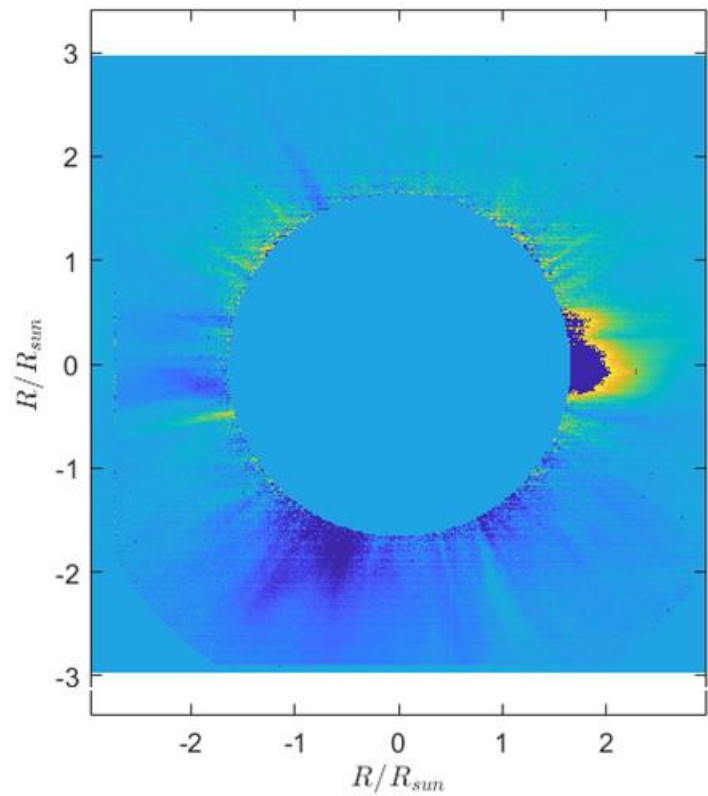
$$\begin{aligned}r \sin(\alpha_A + \beta_A) &= d_A \sin \alpha_A \\r \sin(\alpha_B + \beta_B) &= d_A \sin \alpha_B \\ \beta_A + \beta_B &= \gamma\end{aligned}$$



Algorithm steps

6. In the case of a maximum or bell-shape intensity distribution, take points at half maximum and define a leading and trailing edge





CME angular width
CME propagation direction (weighted average)