



University
of Glasgow

MODEL-INDEPENDENT VELOCITY AND ACCELERATION OF HELIOSPHERIC DISTURBANCES

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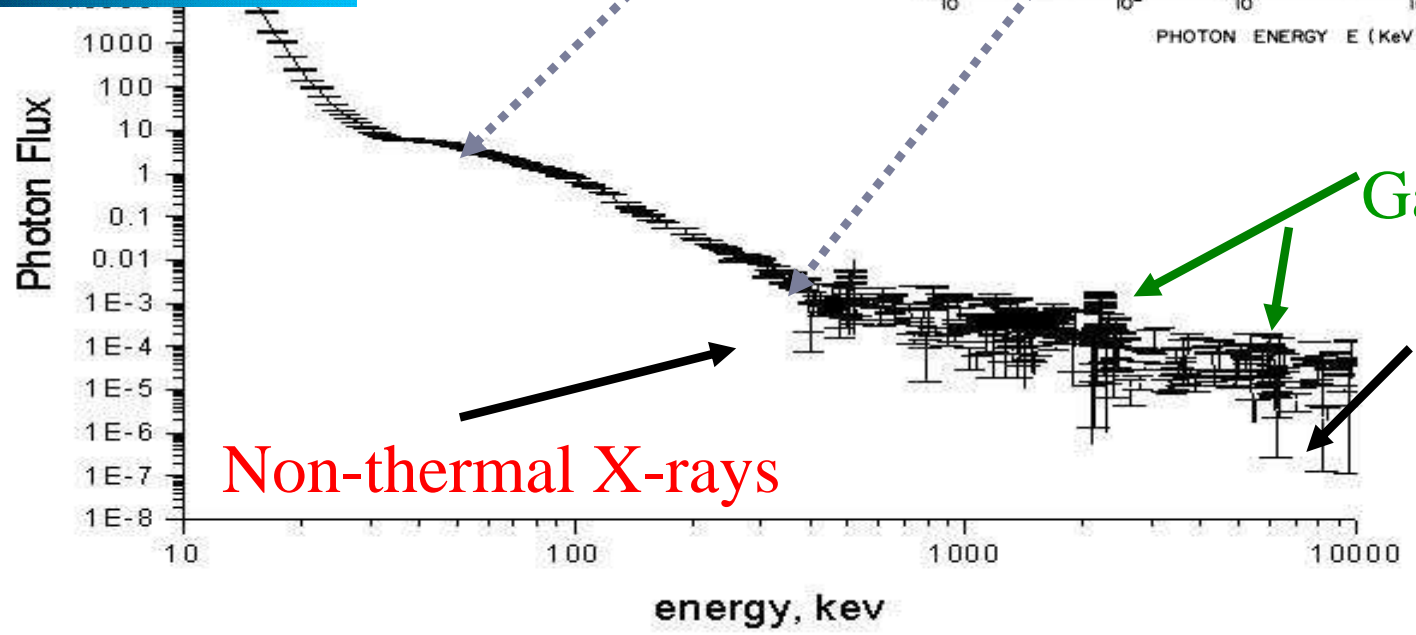
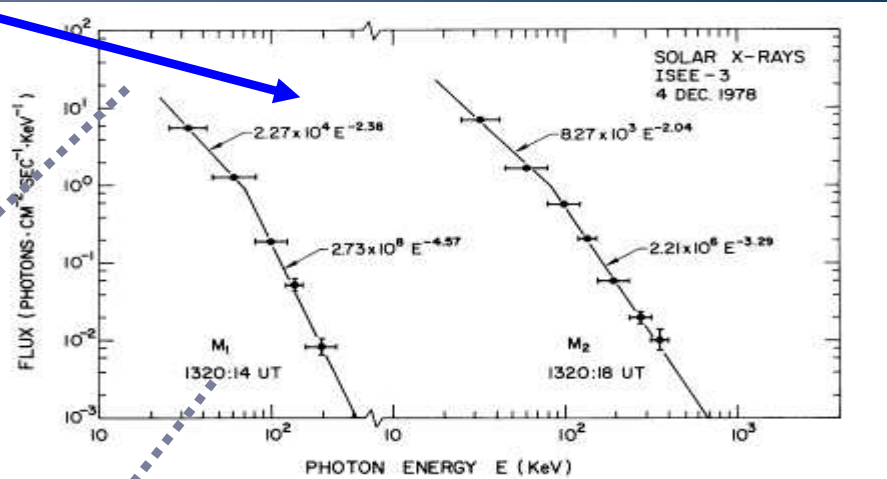
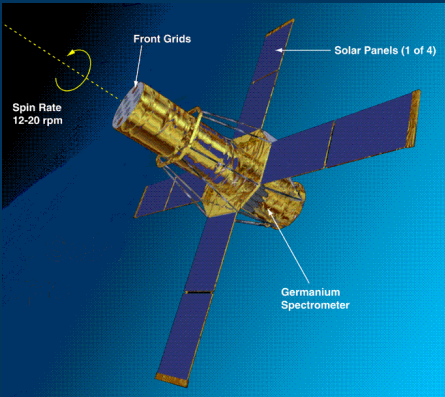
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“Acceleration errors are difficult...”

Angelos Vourlidas

Motivation – X-ray data

Pre-RHESSI X-ray measurements



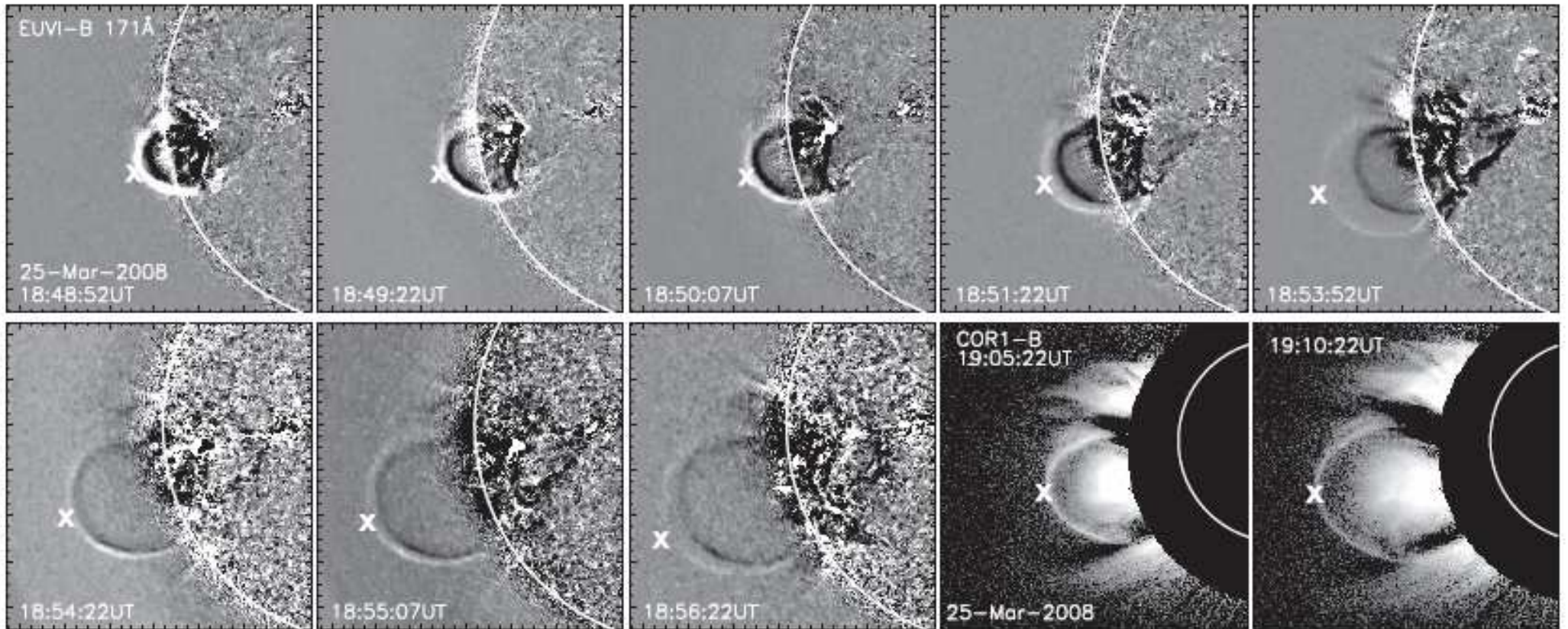
Gamma-ray lines

Non-thermal X-rays

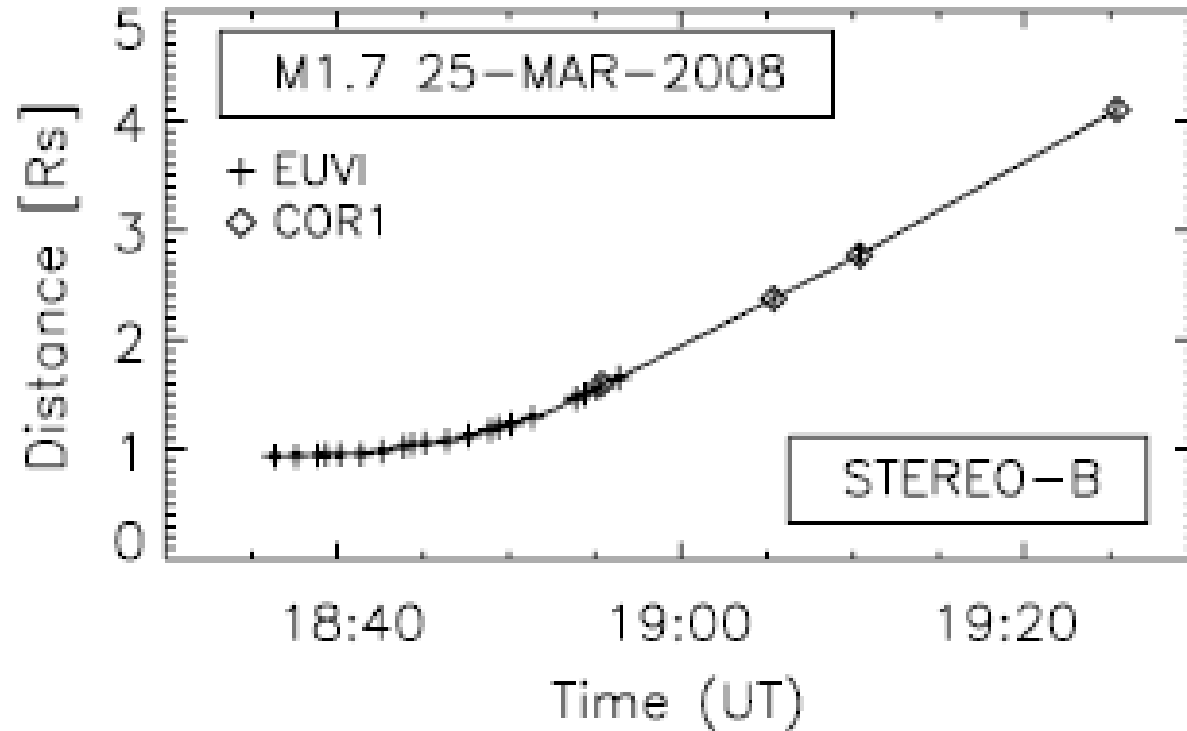
Ramaty High Energy Solar Spectroscopic Imager (RHESSI)

e.g. spectral index (Kontar and MacKinnon, 2005)

Let us consider CME propagation....



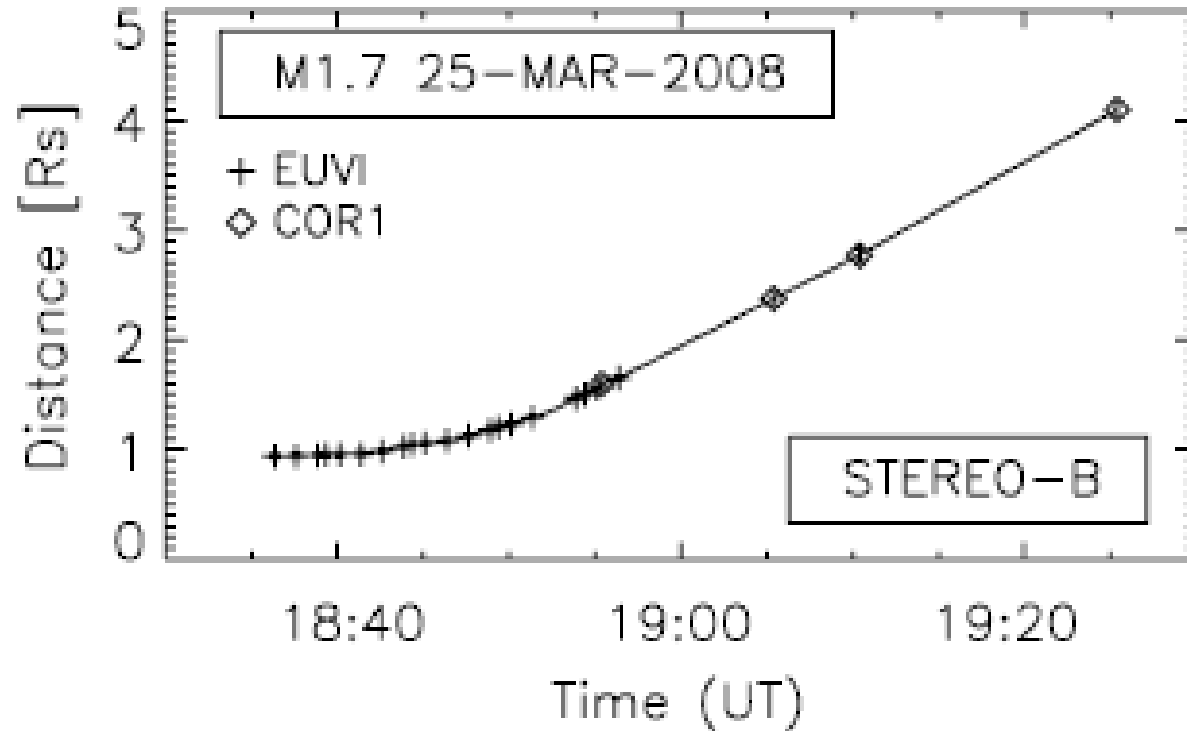
2008 March 25, M1.7 flare/CME event observed with *STEREO-B*
from (Temmer et al, *ApJ*, 2010)



Distance vs time for
2008 March 25, M1.7
flare/CME event
observed with
STEREO-B
from (Temmer et al,
[ApJ, 2010](#))

Two approaches to find velocity and acceleration of CMEs:

- 1) Forward fitting: to find the parameters of the model as a best fit to the original data
- 2) Model independent (**no model assumed**) inference of velocity and acceleration



Distance vs time for
2008 March 25, M1.7
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Two approaches to find velocity and acceleration of CMEs:

- ~~1) Forward fitting: to find the parameters of the model as a best fit to the original data~~
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Let us assume that we have a **analytical functions** $h(t)$ over the finite time interval (t_0, t_N) , while we are given a dataset of height measurements:

$$h_i, i = 1 \dots N$$

for a number of times

$$t_i, i = 1 \dots N.$$

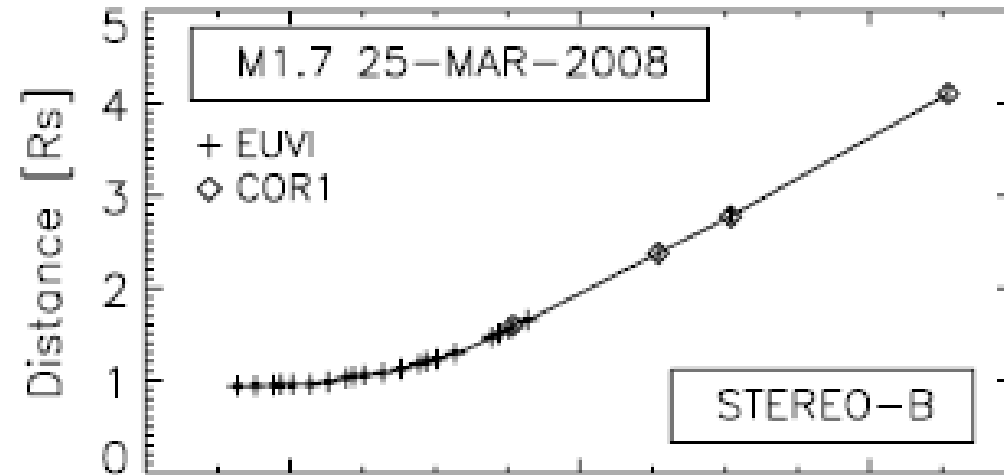
The dataset has a **finite uncertainty** of the measurements δh , so that

$$h_i - h(t_i) < \delta h$$

Now our problem is to find **the best smooth representations** of

$$v(t) = dh(t)/dt \quad (\text{velocity})$$

$$a(t) = d^2h(t)/dt^2 \quad (\text{acceleration})$$

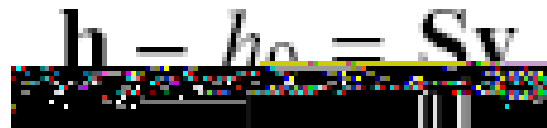


The problem of finding derivative can be written as the integral inversion problem (*Groetsch, C. W. 1984, Hanke, M., & Scherzer, O. 2001*)

Indeed, the height at a given time is given by an integral

$$h_i = h_0 + \int_{t_0}^{t_i} v(t') dt'$$

we can re-write this equation in the **matrix form**

The equation $\mathbf{h} - h_0 = \mathbf{S}\mathbf{v}$ is displayed with a colorful, pixelated background effect.

where **S** is the matrix representing our integral

h is the data-vector given $[h_1, \dots, h_N]$,

v is the velocity-vector to be found $[v_1, \dots, v_M]$,

In other words we are looking for a solution of the minimization problem

$$\|\mathbf{h} - h_0 - \mathbf{S}\mathbf{v}\|^2 = \min \quad [1]$$

where $\|\cdot\|^2$ is a norm defined as $\|\mathbf{h}\|^2 \equiv \mathbf{h}^T \mathbf{h}$.

This problem [1] **does not have a unique solution** and **additional constraints** are needed (e.g. Bertero et al, 1985).

$$\cancel{\mathbf{v} = \mathbf{S}^{-1} (\mathbf{h} - h_0)}$$

Using Tikhonov **regularization** technique (*Tikhonov, 1963*), our problem becomes:

$$\|\mathbf{h} - h_0 - \mathbf{S}\mathbf{v}\|^2 + \lambda \|\mathbf{L}\mathbf{v}\|^2 = \min \quad [2]$$

where \mathbf{L} is the matrix representation of constraint operator, and λ is a regularization constant.

Importantly, the problem [2] is well-behaved and has **a unique solution**.

The derivative error calculated from noisy data set:

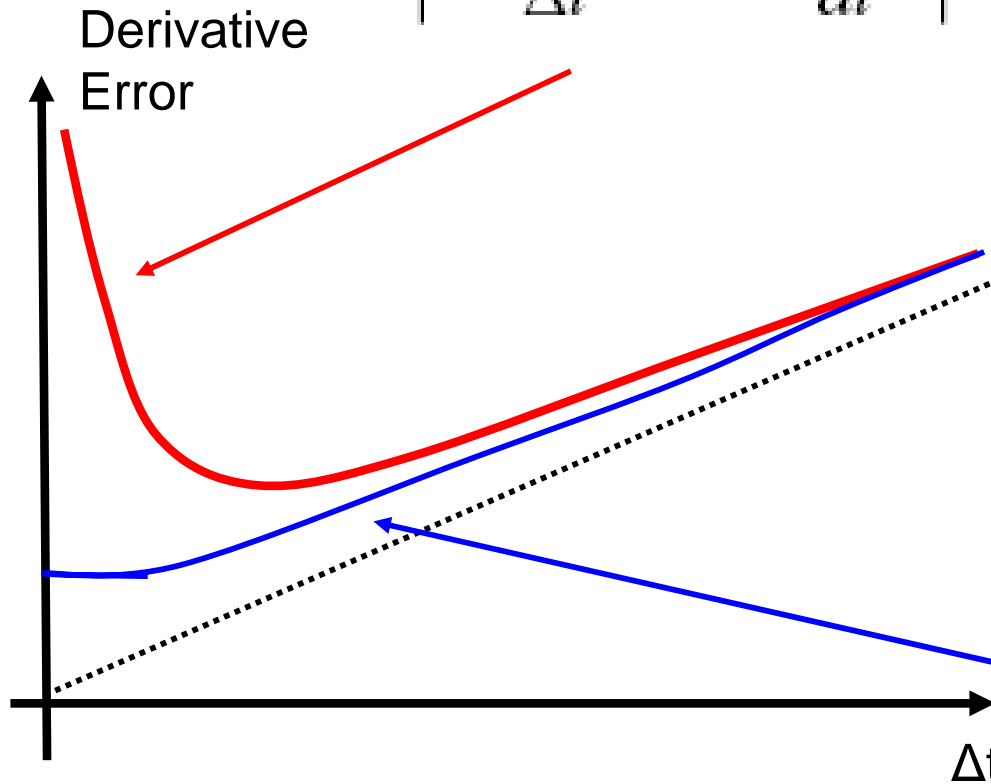
$$\left| \frac{h_{i+1} - h_i}{\Delta t} - \frac{dh(t_i)}{dt} \right| \leq O\left(\Delta t + \frac{\delta h}{\Delta t}\right)$$

We will look for a function $h(t)$ close to a given data set so that

$$\| \mathbf{h} - \mathbf{h}_0 - \mathbf{S}\mathbf{v} \|^2 = \| \delta \mathbf{h} \|^2$$

While the **second derivative** of $h(t)$ or **first derivative** of $v(t)$ has a minimum, the derivative error looks much better:

$$\left| \frac{h_{i+1} - h_i}{\Delta t} - \frac{dh(t_i)}{dt} \right| \leq O(\Delta t + \delta h)$$



Therefore, following *Hanke and Scherzer (2001)* we can choose $\mathbf{L}=\mathbf{D}_1$

Hence we can write an **explicit solution of minimization problem**, which minimizes the amplification of the errors in the resulting estimate for the derivative, i.e. velocity:

$$\mathbf{v}_\lambda = \mathbf{R}(\mathbf{h} - h_0), \quad \text{where} \quad \mathbf{R} = (\mathbf{S}^T \mathbf{S} + \lambda \mathbf{D}_1^T \mathbf{D}_1)^{-1} \mathbf{S}^T$$

The **only unknown parameter is λ** , which can be determined requiring the finite difference between our solution and the original dataset

$$\|\mathbf{h} - h_0 - \mathbf{S}\mathbf{v}_\lambda\|^2 = \alpha \|\delta\mathbf{h}\|^2$$


Parameter α tells us about the errors (should be around 1 in case of Gaussian errors)

Let assume that we know the **true solution** of our linear inverse problem \mathbf{v}_{true} , then we can write

$$\mathbf{h} - h_0 = \mathbf{S} \mathbf{v}_{true} + \delta\mathbf{h}$$

The regularized solution of our inverse problem is

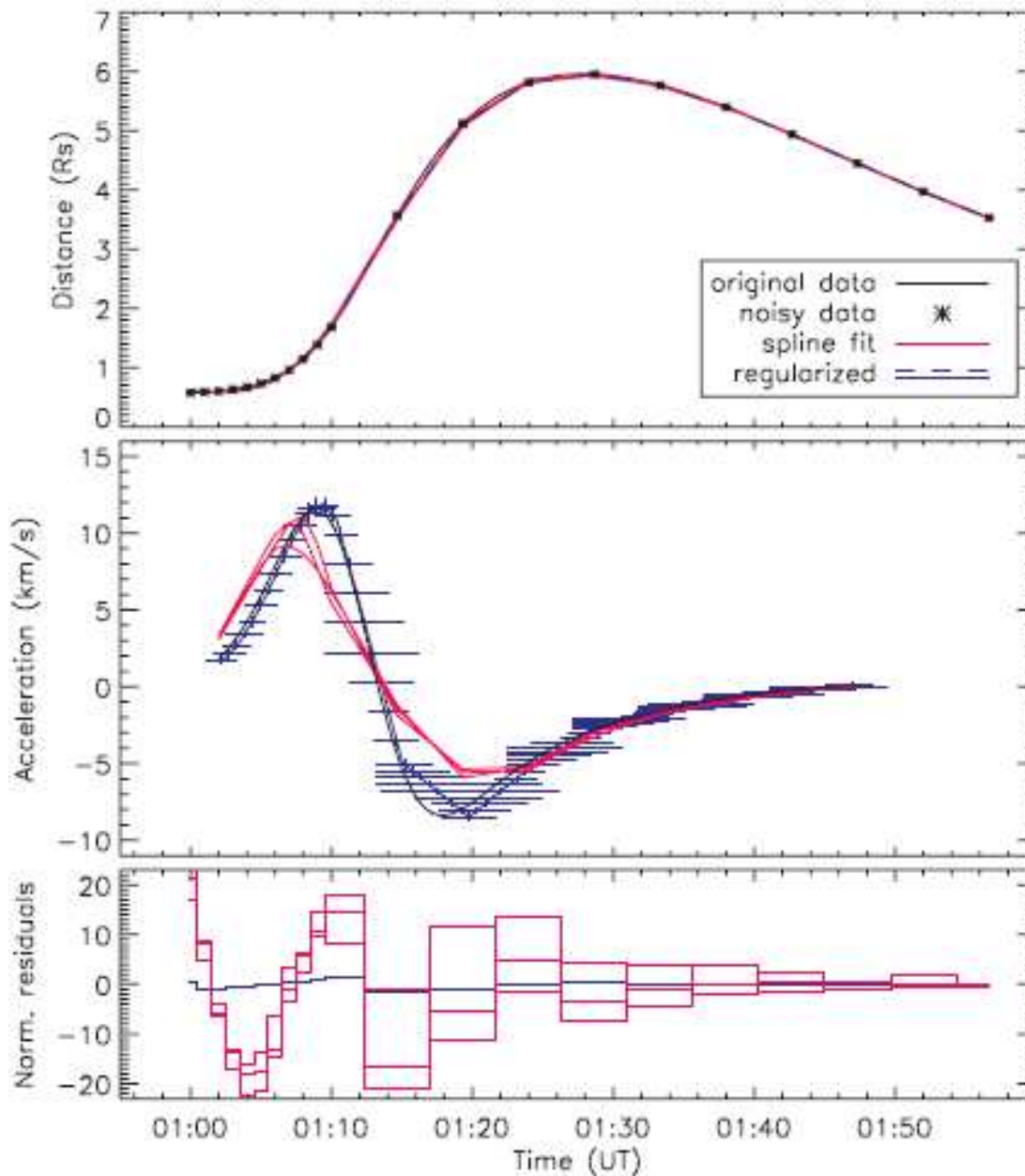
$$\mathbf{v} = \mathbf{R} (\mathbf{h} - h_0)$$

The difference between the true solution and our solution can be written as

$$\delta\mathbf{v} = \mathbf{v} - \mathbf{v}_{true} = (\mathbf{RS} - \mathbf{1})\mathbf{v}_{true} + \mathbf{R}\delta\mathbf{h}$$

Resolution
(horizontal error)

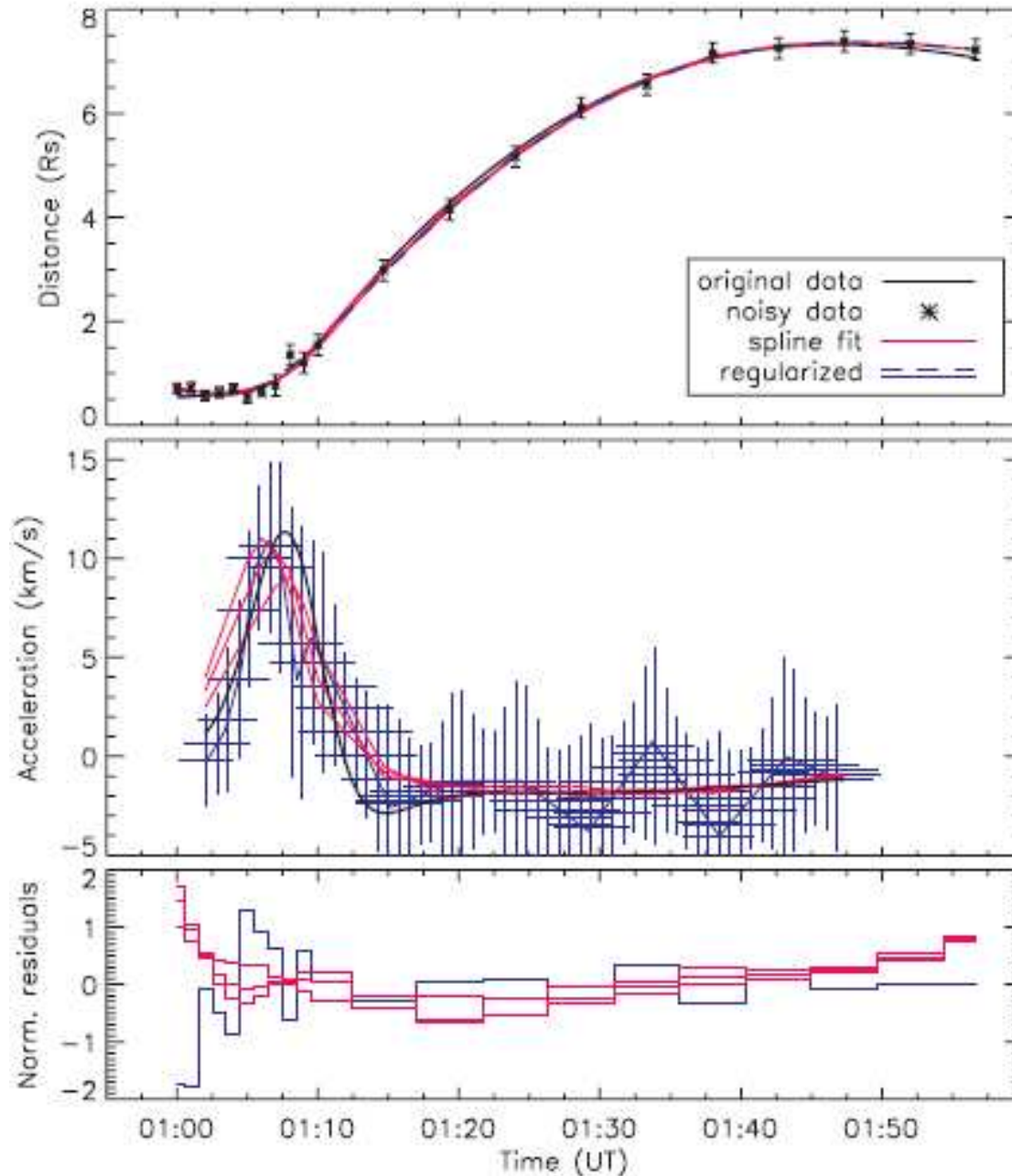
Data noise
propagation error
(vertical error)



Simulated data (no noise added but discrete data set)

Acceleration profile (error due to discrete data set is evident)

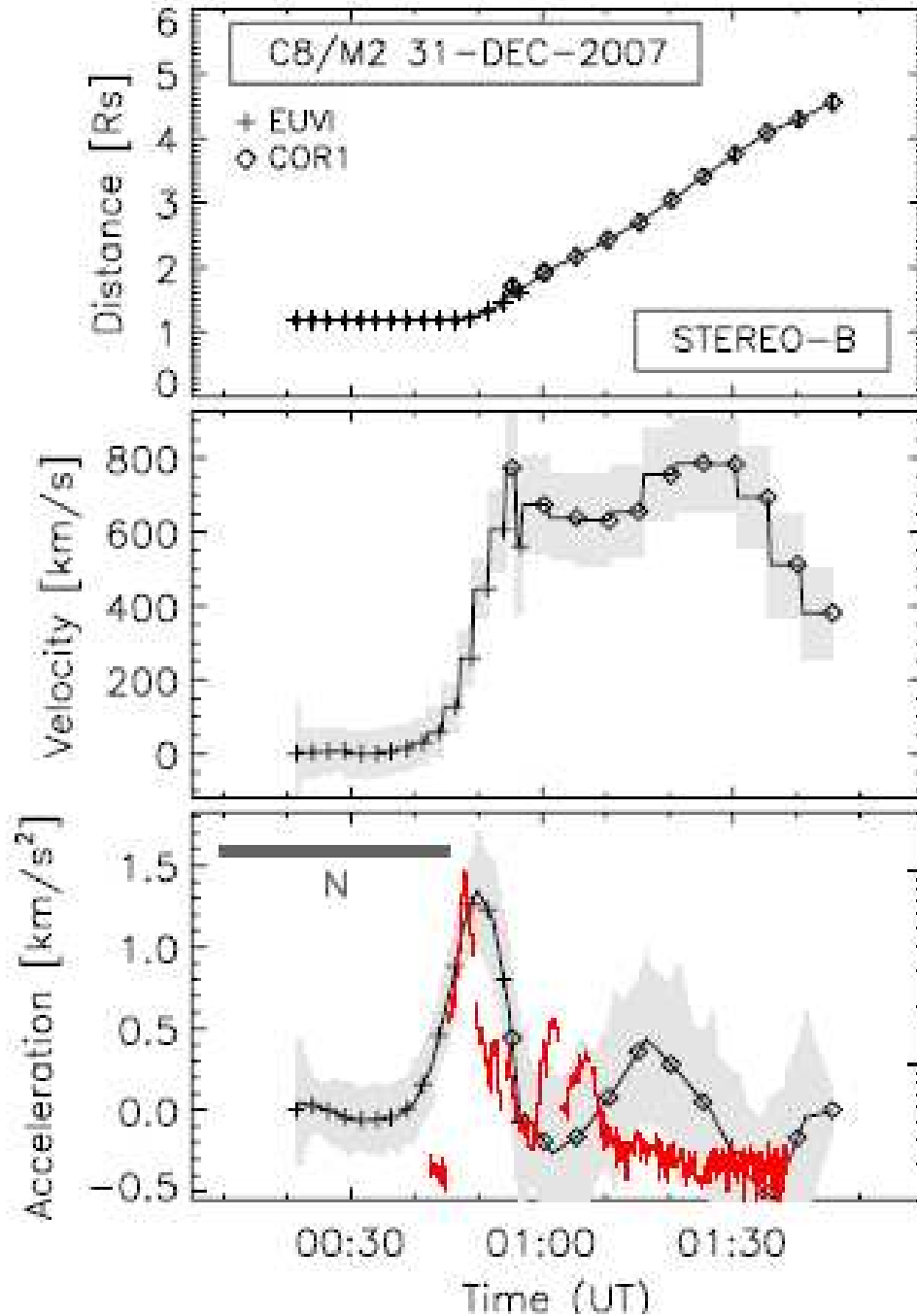
Normalised residuals



Simulated data (realistic noise added)

Acceleration profile

Normalised residuals



Height-time data (Temmer et al, ApJ, 2010)

Velocity profile

Acceleration profile (Note change in acceleration profile)

Regularized inversion gives us **model-independent** (**without assumptions on functional shape**) **velocity and acceleration as a function of time.**

=>provides us with horizontal and vertical error bars and hence gives us confidence range for **velocity** and **acceleration.**

⇒Regularized derivative is IDL based package and easy to use

⇒Can be applied not only to CME data but to EIT waves etc