

GDR 2026

AN ALTERNATING APPROACH TO RECONSTRUCT PROMPT GAMMA DISTRIBUTION AND HADRON VELOCITY FROM TIME-OF-FLIGHT MEASUREMENTS

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PGTI
PROMPT GAMMA TIME IMAGING

Outline

1. Context

2. Setting of the problem

3. An alternating minimisation approach

4. Numerical results

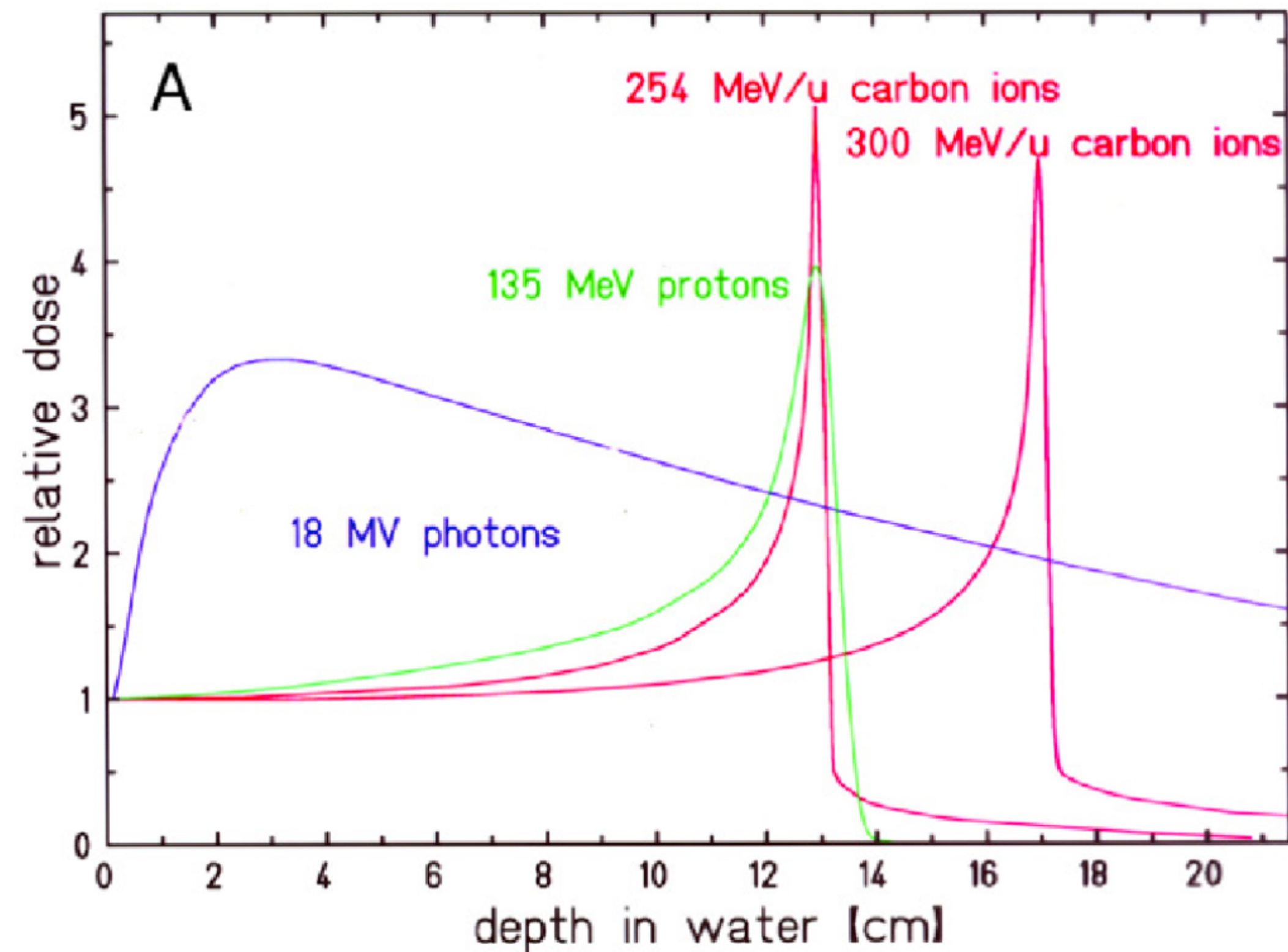
5. Perspectives

Hadrontherapy

Advantage: Protons deposit a maximum dose at a specific depth (Bragg Peak), sparing healthy tissues behind tumor

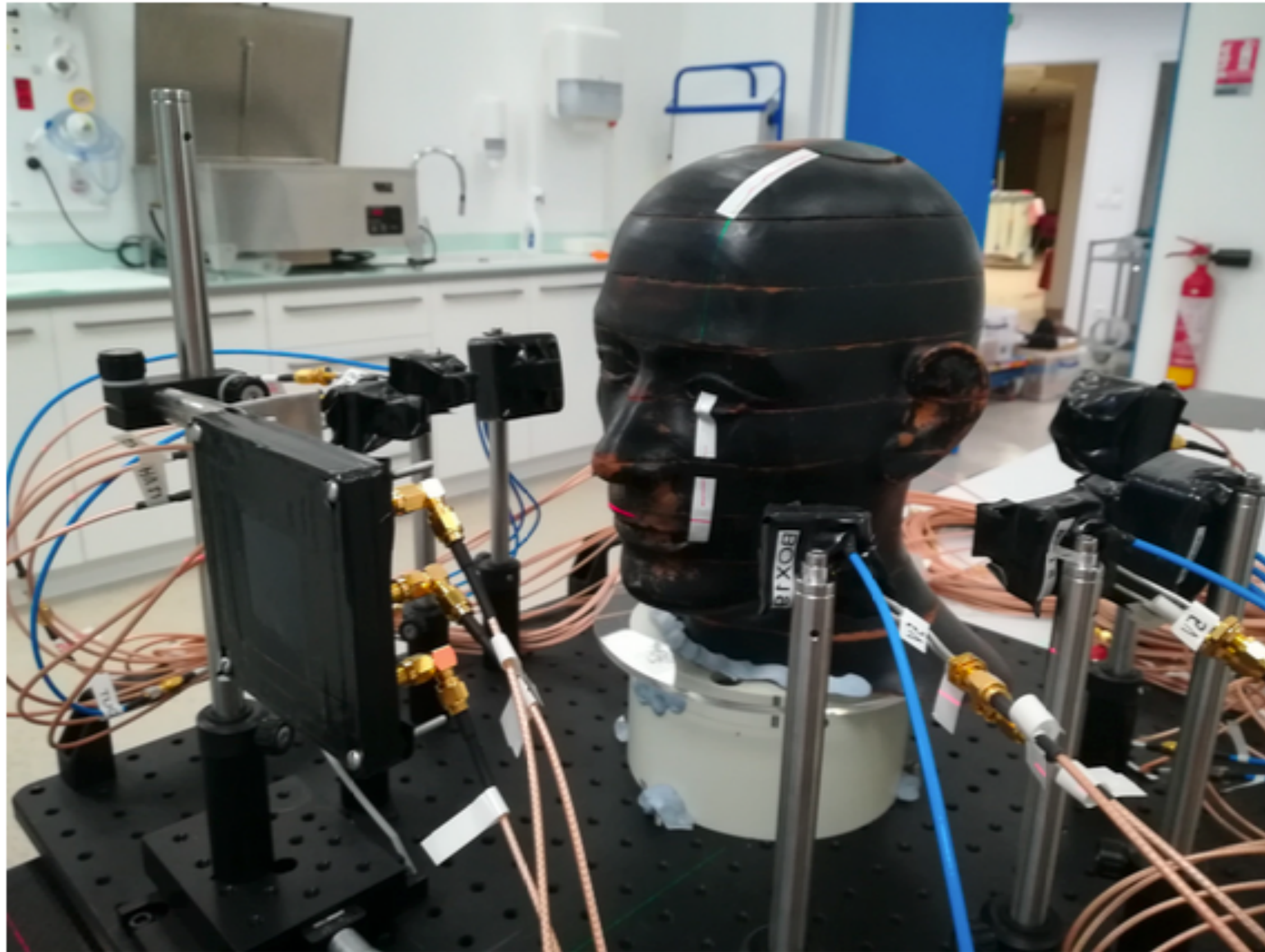
Range sensitivity: Small changes in patient anatomy significantly shift the Bragg Peak position

Challenge: Develop real-time range verification methods to ensure accurate dose administration



Comparison of the energy loss of protons, carbons and X-rays in water. The Bragg peak is clearly visible for protons and carbons, while X-rays show a continuous energy loss (Durante et al. [2016](#)).

The Prompt Gamma Time Imaging (PGTI) project

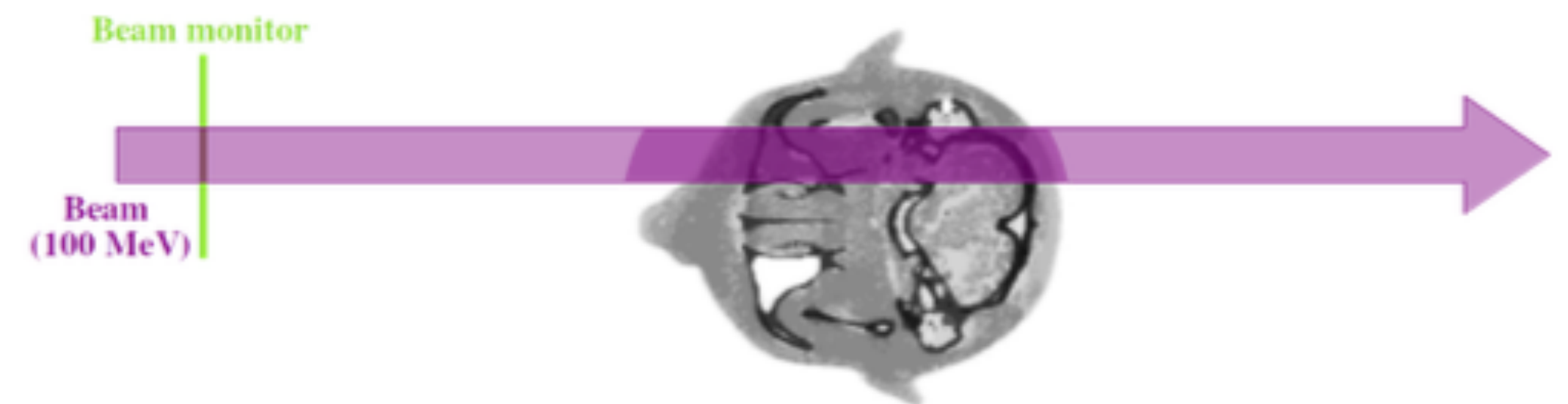


TIARA detector: Beam monitor + Prompt gamma detectors

Physics Principle: Protons interact with patient nuclei, emitting **Prompt Gammas (PG)** along their path

Mechanism: PGTI project develops the TIARA detector to measure Time-Of-Flights (TOFs)

Encoded Information: TOF depends on two unknowns: the **emission position** and the **proton velocity**



Outline

1. Context

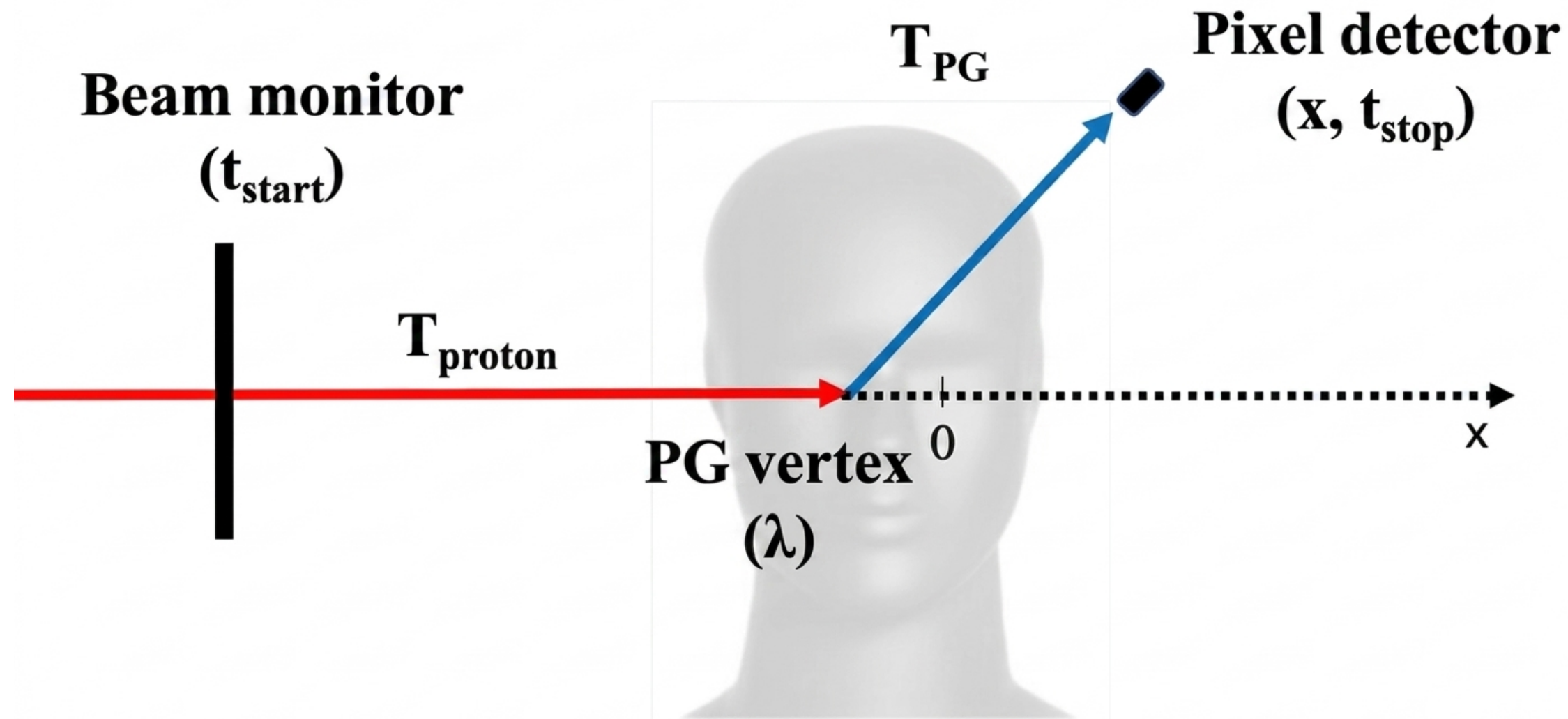
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Setting of the problem



For **one** prompt Gamma emitted λ , the total Time of flight reads:

$$\begin{cases} T(\lambda, \mathbf{v}) = T_{proton}(\lambda, \mathbf{v}) + T_{PG}(\lambda) \\ = \int_0^\lambda \frac{1}{\mathbf{v}(s)} ds + \frac{\|\lambda - x\|}{c} \end{cases}$$

Where:

c : speed of light

$\mathbf{v}$: the velocity of the proton

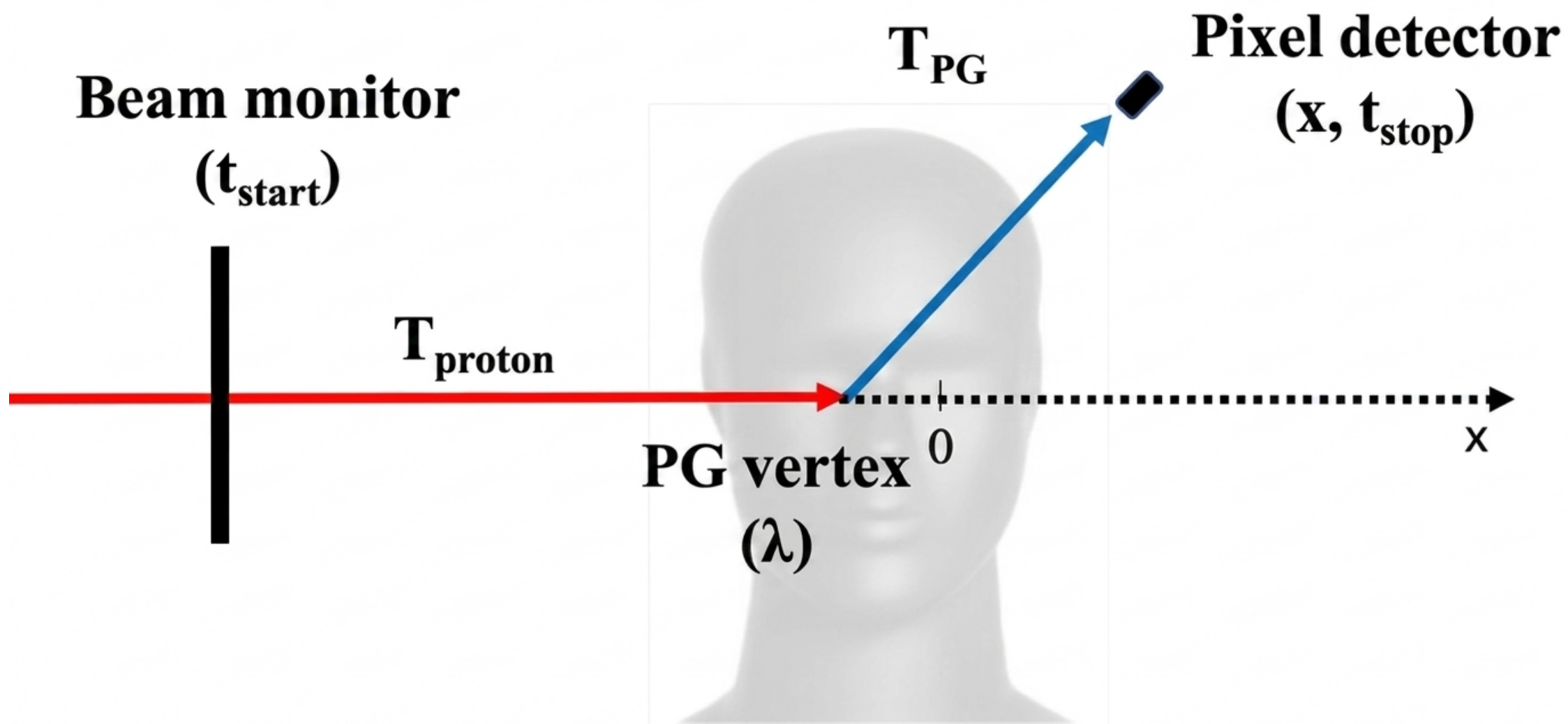
For **n** prompt gammas, the emission positions are denoted by $\lambda = [\lambda_1, \dots, \lambda_n] \in \mathbb{R}^n$

$$T_i(\lambda_i, \mathbf{v}) = \int_0^{\lambda_i} \frac{1}{\mathbf{v}(s)} ds + \frac{\|\lambda_i - x_i\|}{c}, \quad \forall 0 < i < n$$

The proton **mean velocity**, discretized on spatial dimension, is given by $\mathbf{v} = [v_1, \dots, v_d] \in \mathbb{R}^d$. Let's take $\boldsymbol{\gamma} = [1/v_1, \dots, 1/v_d] \in \mathbb{R}^d$

$$T(\lambda, \boldsymbol{\gamma}) = \begin{pmatrix} T_1(\lambda_1, \boldsymbol{\gamma}) \\ \vdots \\ T_n(\lambda_n, \boldsymbol{\gamma}) \end{pmatrix} = A(\lambda) \boldsymbol{\gamma} + \frac{1}{c} \begin{pmatrix} \|\lambda_1 - x_1\| \\ \vdots \\ \|\lambda_n - x_n\| \end{pmatrix} \quad \text{where} \quad A(\lambda) = \begin{pmatrix} a_1(\lambda_1)^t \\ a_2(\lambda_2)^t \\ \vdots \\ a_n(\lambda_n)^t \end{pmatrix}$$

Setting of the problem



For n prompt Gammas emitted $\lambda = [\lambda_1, \dots, \lambda_n] \in \mathbb{R}^n$, the total Time of flight reads:

$$\begin{cases} T(\lambda, \gamma) = T_{proton}(\lambda, \gamma) + T_{PG}(\lambda) \\ = A(\lambda)\gamma + \frac{\|\lambda - x\|}{c} \end{cases}$$

Where:

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$x = [x_1, \dots, x_n] \in \mathbb{R}^n$ the detected position of prompt gammas

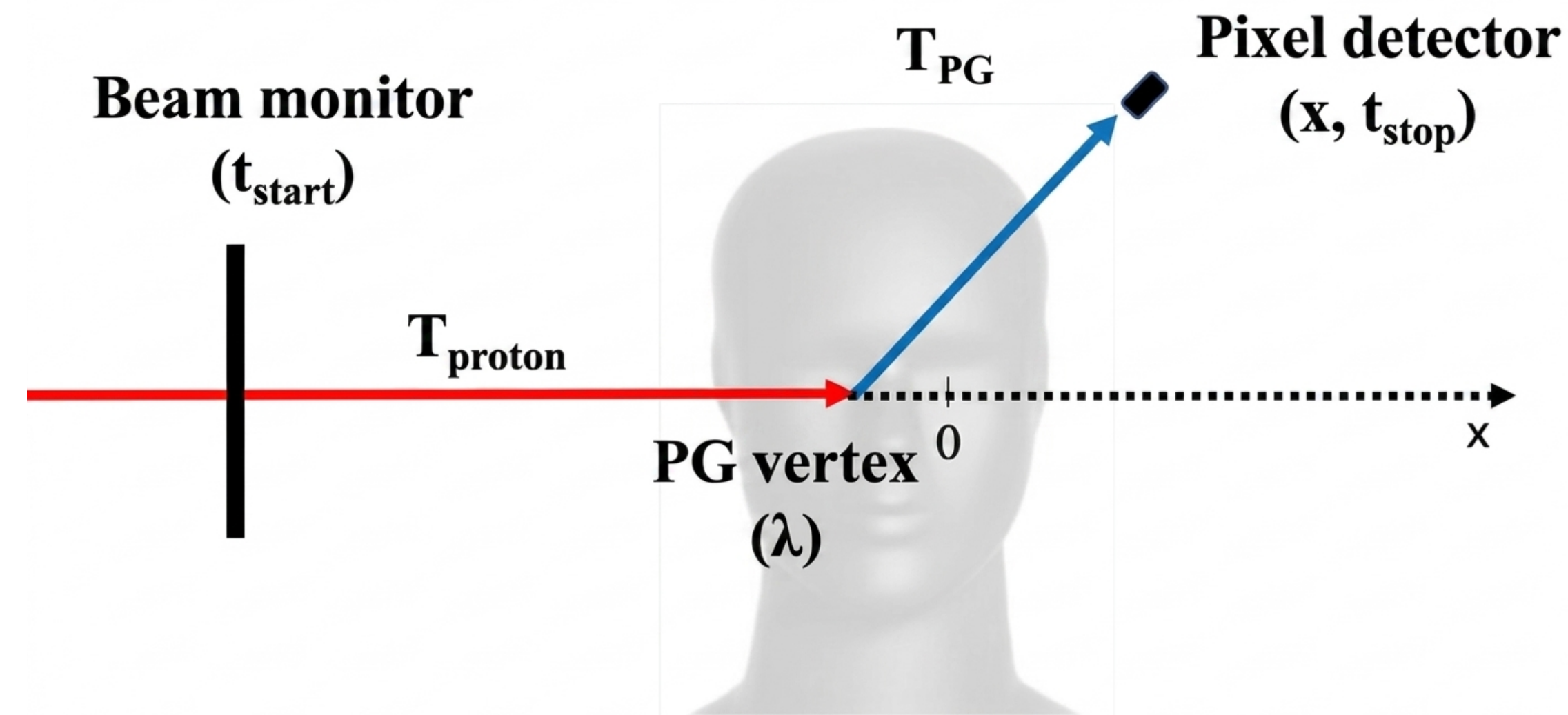
$\gamma = [\gamma_1, \dots, \gamma_d] \in \mathbb{R}^d$ the velocity of the proton

Objective: Find $\lambda \in \mathbb{R}^n, \gamma \in \mathbb{R}^d$

$$\min_{\lambda, \gamma} J(\lambda, \gamma) = \frac{1}{2\sigma^2} \|t - T(\lambda, \gamma)\|^2$$

Where: $t = t_{stop} - t_{start}$

Setting of the problem



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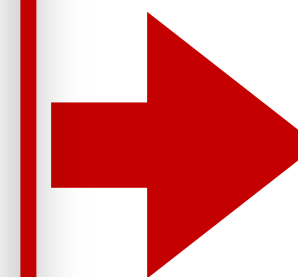
$x = [x_1, \dots, x_n] \in \mathbb{R}^n$ the detected position of prompt gammas

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ill-posed problem !!!

NON-CONVEXITY

Outline

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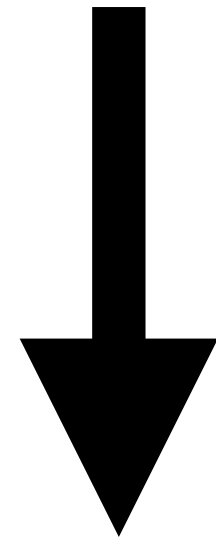
2. Setting of the problem

3. An alternating minimization approach

4. Numerical results

5. Perspectives

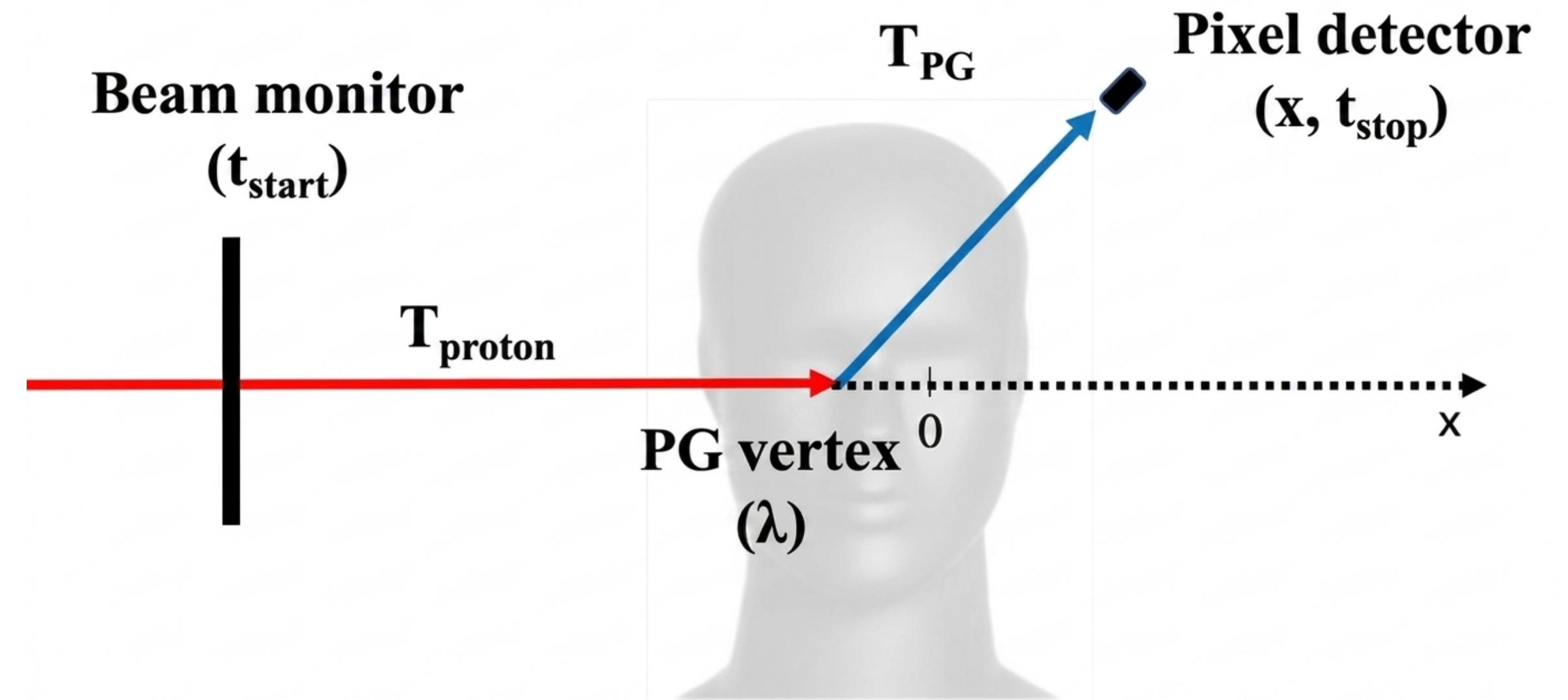
An alternating minimization approach ?



Requires Bi-Convexity!

$$\min_{\lambda, \gamma} J(\lambda, \gamma) = \frac{1}{2\sigma^2} \| t - T(\lambda, \gamma) \|^2$$

$$t = t_{stop} - t_{start}$$



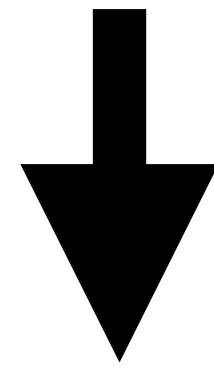
$$\begin{cases} T(\lambda, \gamma) = T_{proton}(\lambda, \gamma) + T_{PG}(\lambda) \\ = A(\lambda)\gamma + \frac{\|\lambda - x\|}{c} \end{cases}$$

$$t = [t_1, \dots, t_n] \in \mathbb{R}^n$$

$$\lambda = [\lambda_1, \dots, \lambda_n] \in \mathbb{R}^n$$

$$\gamma = [\gamma_1, \dots, \gamma_d] \in \mathbb{R}^d$$

An alternating minimization approach ?



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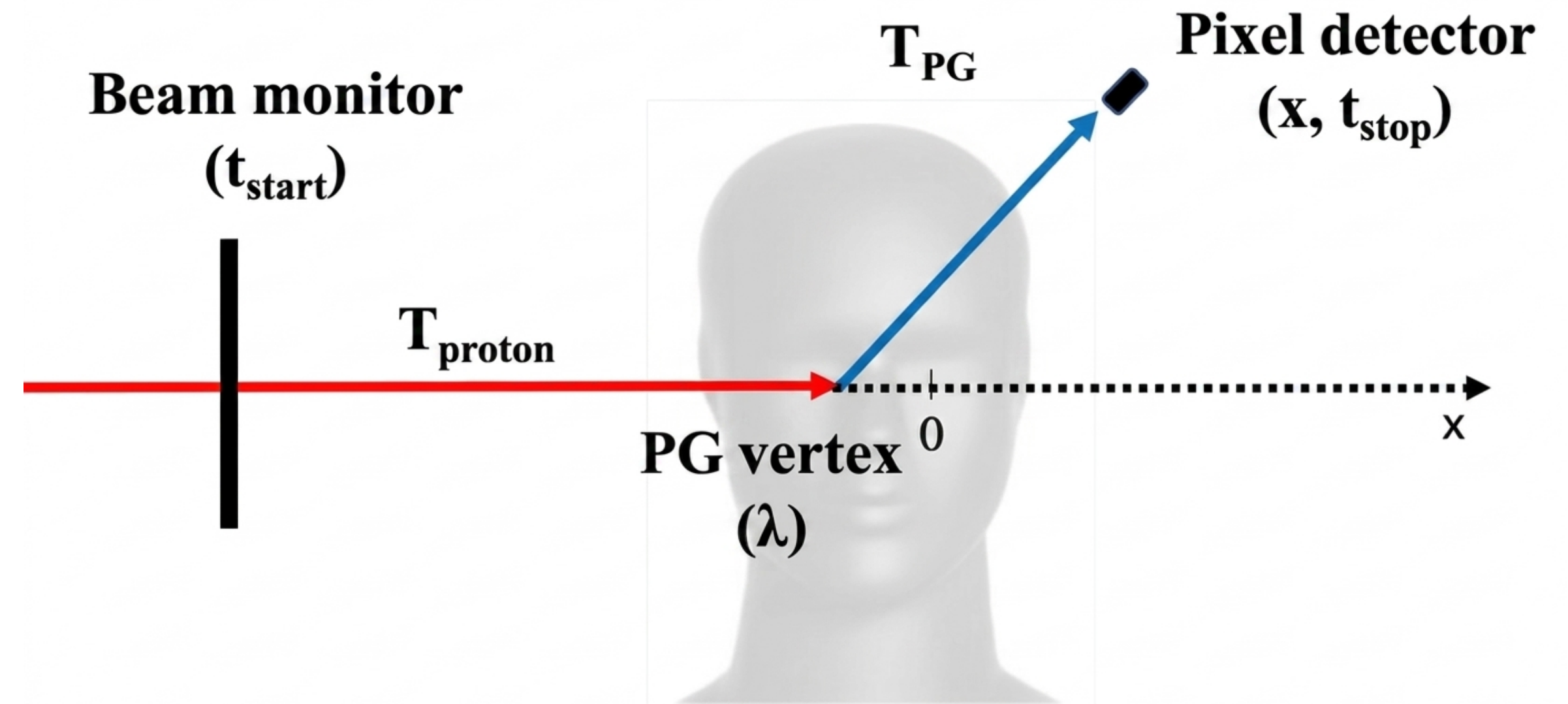
$$t = t_{\text{stop}} - t_{\text{start}}$$

Step 1: Fix γ

$$\min_{\lambda} J(\lambda, \cdot) = \frac{1}{2\sigma^2} \| t - T(\lambda, \cdot) \|^2$$

Step 2: Fix λ

$$\min_{\gamma} J(\cdot, \gamma) = \frac{1}{2\sigma^2} \| t - T(\cdot, \gamma) \|^2$$



$$\begin{cases} T(\lambda, \gamma) = T_{\text{proton}}(\lambda, \gamma) + T_{PG}(\lambda) \\ \quad \quad \quad = A(\lambda)\gamma + \frac{\|\lambda - x\|}{c} \end{cases}$$

$$t = [t_1, \dots, t_n] \in \mathbb{R}^n$$

$$\lambda = [\lambda_1, \dots, \lambda_n] \in \mathbb{R}^n$$

$$\gamma = [\gamma_1, \dots, \gamma_d] \in \mathbb{R}^d$$

An alternating minimization approach ?



Requires Bi-Convexity!

$$\min_{\lambda, \gamma} J(\lambda, \gamma) = \frac{1}{2\sigma^2} \| t - T(\lambda, \gamma) \|^2$$



- Assume that the proton **mean velocity** is known

$$\min_{\lambda \in \mathbb{R}^n} J_1(\lambda, \gamma) = \frac{1}{2\sigma^2} \| t - T(\lambda, \gamma) \|^2$$

Proposition: $\lambda \rightarrow J_1(\lambda, \cdot)$ is **quasi-convex** on $[\lambda_{min}, \lambda_{max}]$

- $J'_1(\lambda) = -\frac{1}{\sigma^2} T'(\lambda)(t - T(\lambda))$
- $J''_1(\lambda) = \frac{1}{\sigma^2} (T'(\lambda)^2 - T''(\lambda))(t - T(\lambda))$

J_1 has global minimum at $\lambda^* = T^{-1}(t)$

$J_1 \searrow$ on $[\lambda_{min}, \lambda^*]$ $J_1 \nearrow$ on $[\lambda^*, \lambda_{max}]$

An alternating minimization approach ?



- Assume that the proton **mean velocity** is known

$$\min_{\lambda \in \mathbb{R}^n} J_1(\lambda, \gamma) = \frac{1}{2\sigma^2} \| t - T(\lambda, \gamma) \|^2$$

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Requires Bi-Convexity!

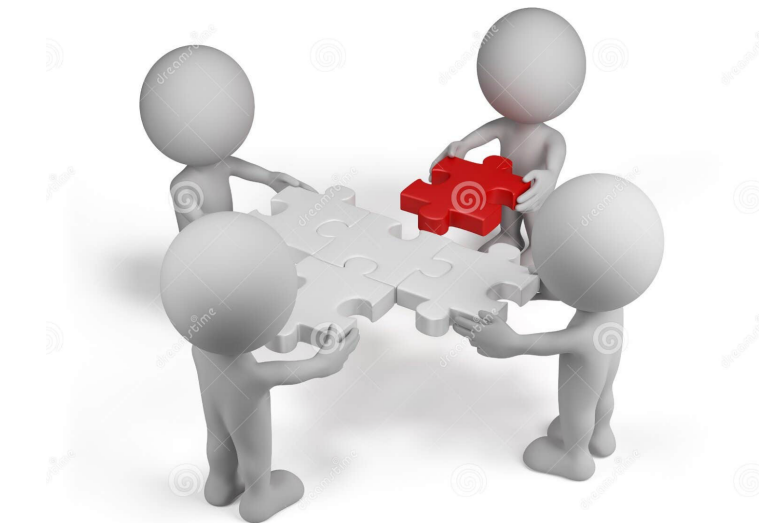
$$\min_{\lambda, \gamma} J(\lambda, \gamma) = \frac{1}{2\sigma^2} \| t - T(\lambda, \gamma) \|^2$$

- Assume that the distribution of **prompt gammas** is known

$$\min_{\gamma \in \mathbb{R}^d} J_2(\lambda, \gamma) = \frac{1}{2\sigma^2} \| b - A\gamma \|^2$$

Proposition: $\gamma \rightarrow J_2(\cdot, \gamma)$ is **convex** $\forall \gamma$ 

- $J'_2(\gamma) = A^t(b - A\gamma)$
- $J''_2(\gamma) = A^t A$
- $J''_2(\gamma) \geq 0$



An alternating minimization approach ?

- Alternating algorithm

$$\left\{ \begin{array}{l} \text{Initialisation} \\ \lambda_0, \gamma_0 \\ \text{For } i = 0, \dots, N_{max} \\ \left\{ \begin{array}{l} \lambda_{i+1} = \arg \min_{\lambda \in \mathbb{R}^n} J_1(\lambda_i, \gamma_i) \\ \gamma_{i+1} = \arg \min_{\gamma \in \mathbb{R}^d} \tilde{J}_2(\lambda_{i+1}, \gamma_i) \end{array} \right. \\ \text{Update} \\ \gamma_i = \gamma_{i+1} \\ \lambda_i = \lambda_{i+1} \end{array} \right.$$

$$J_1(\lambda, \gamma) = \frac{1}{2\sigma^2} \| t - T(\lambda, \gamma) \|^2$$

$$\tilde{J}_2(\lambda, \gamma) = \frac{1}{2\sigma^2} \| A^t \mathbf{b} - A^t A \gamma \|^2 + \underbrace{\chi \| I_{adp} \gamma \|^2}_{\text{Regularization term}}$$

- $I_{adp} = \text{diag}(A^t A)$
- Accounts for local sensitivity
- Improves **stability** of the inverse problem

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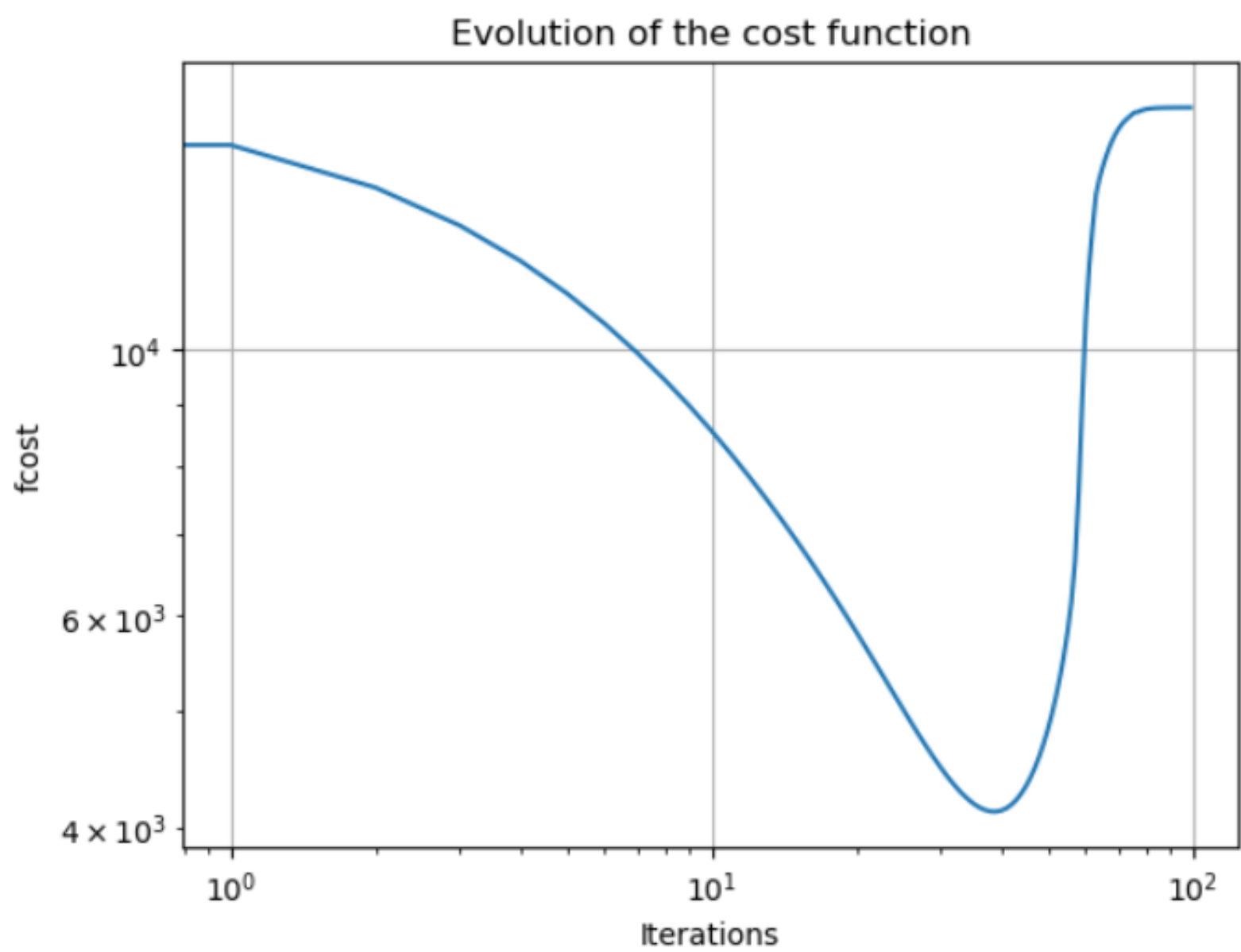
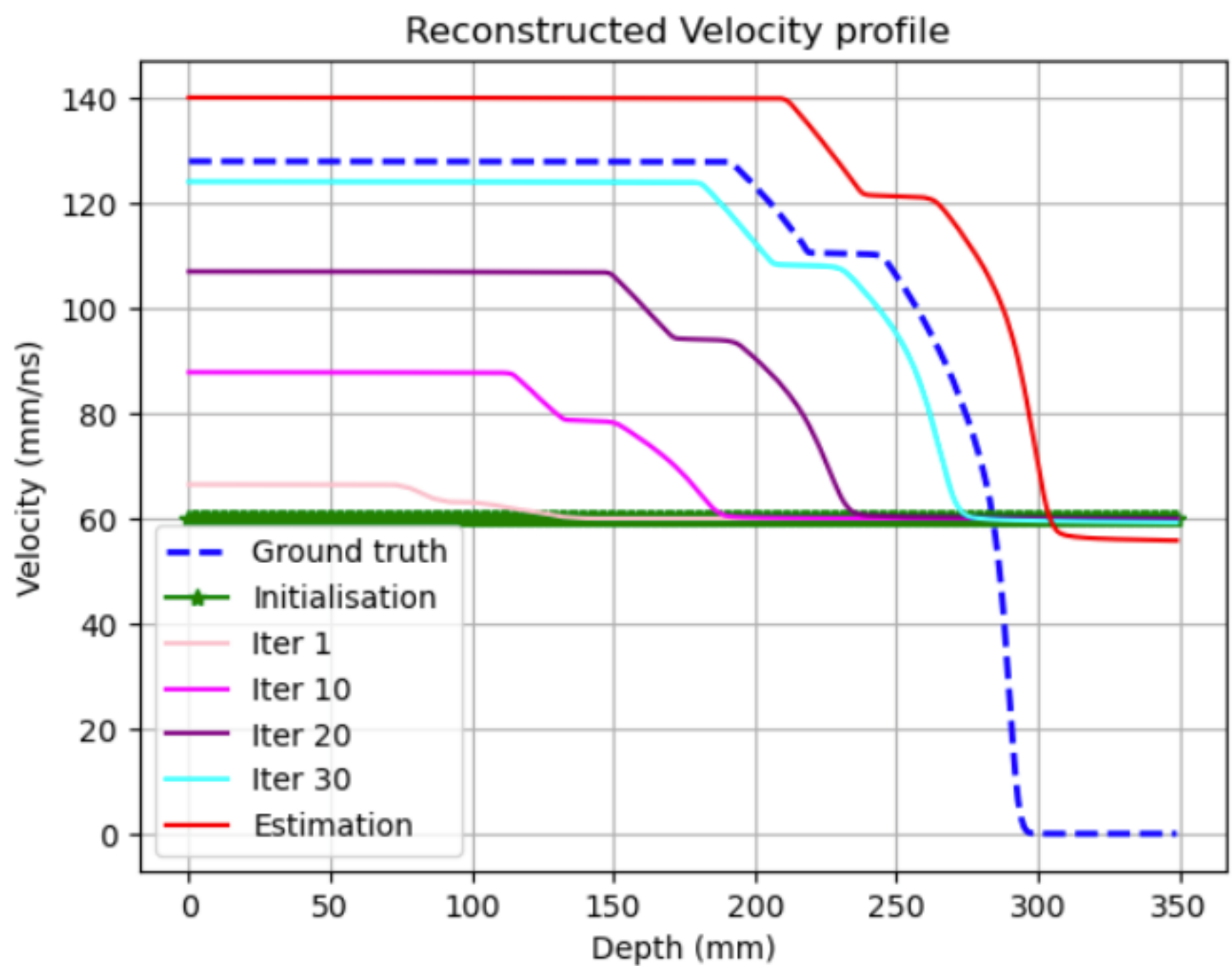
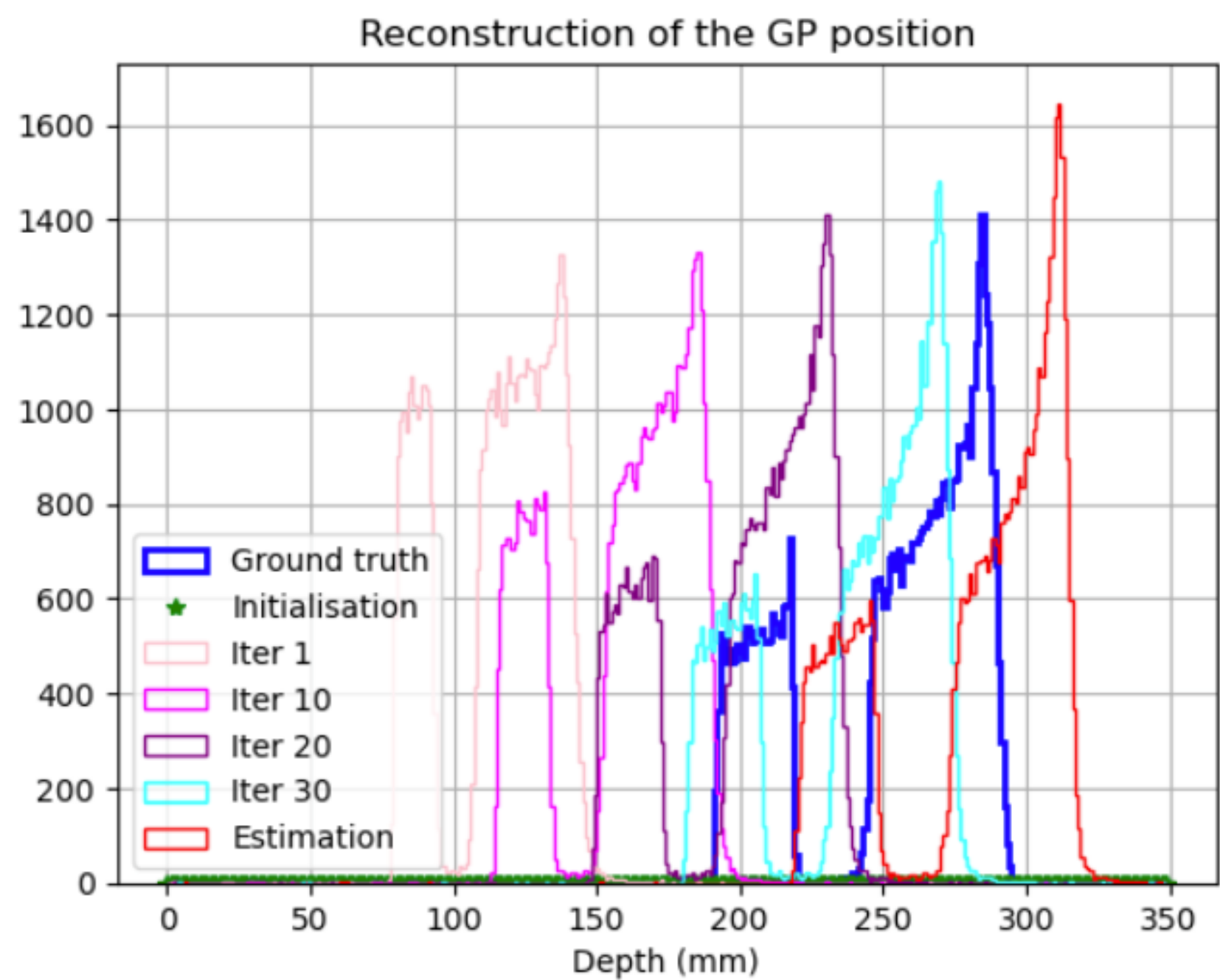
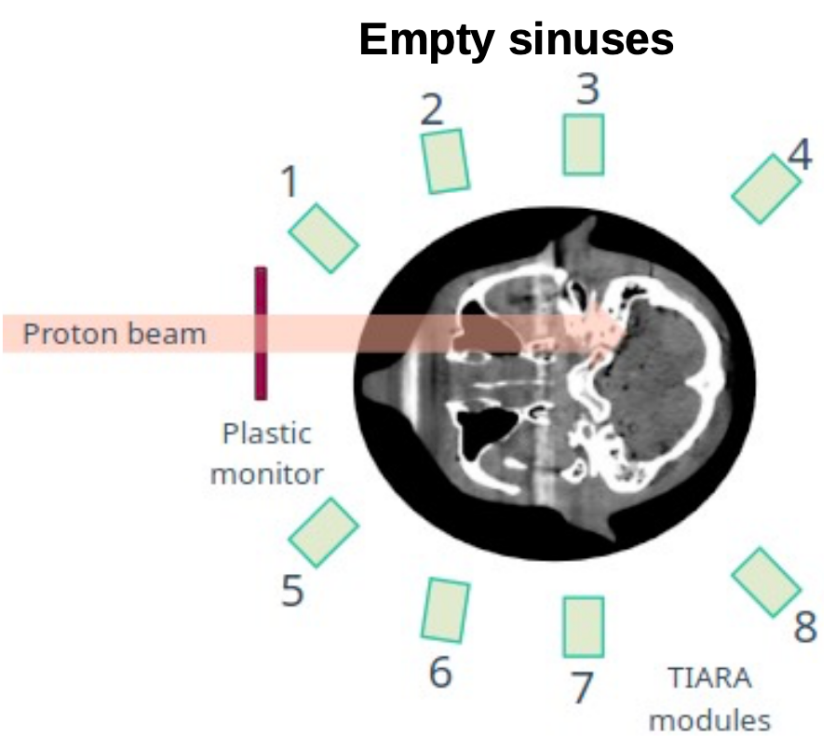
3. An alternating minimization approach

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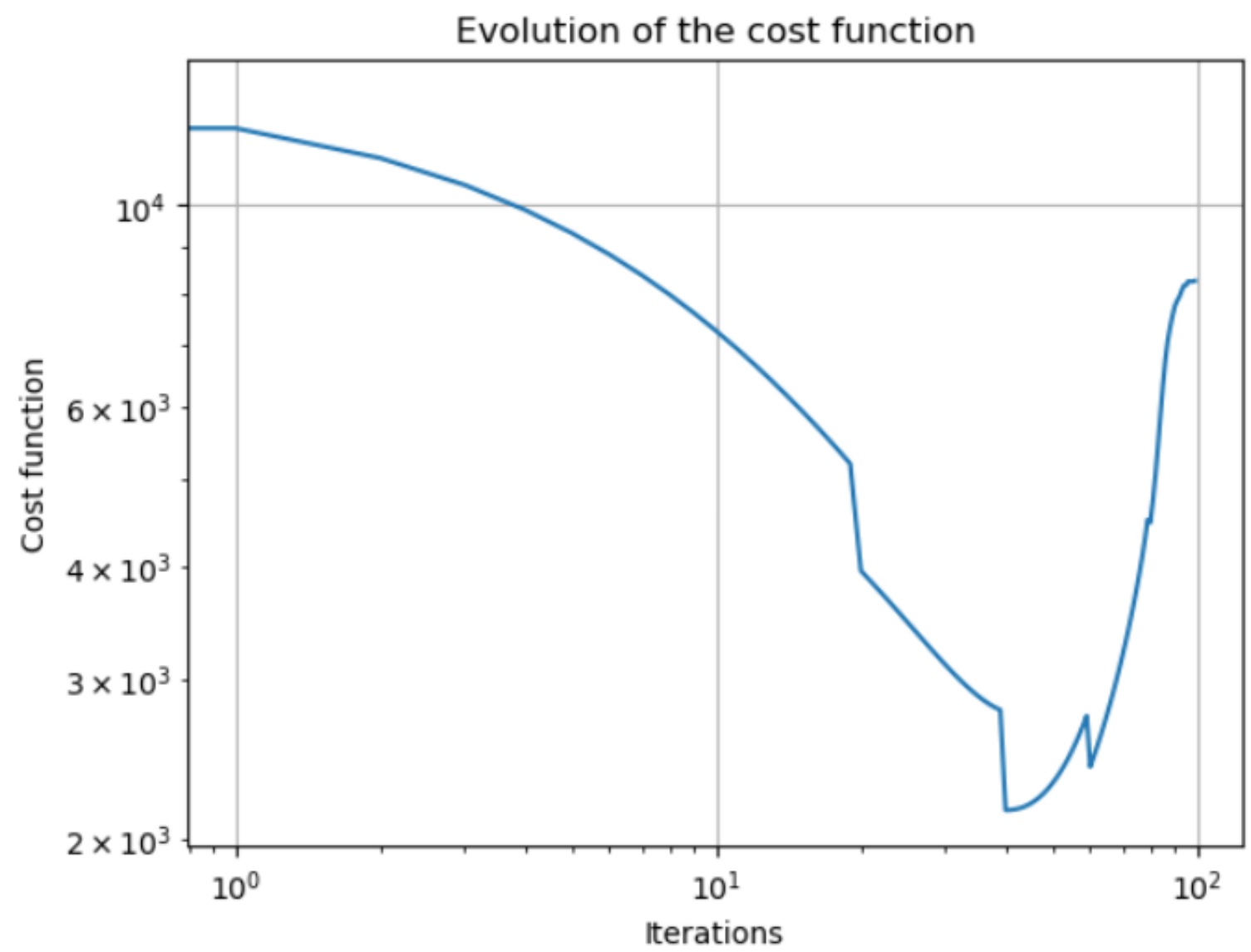
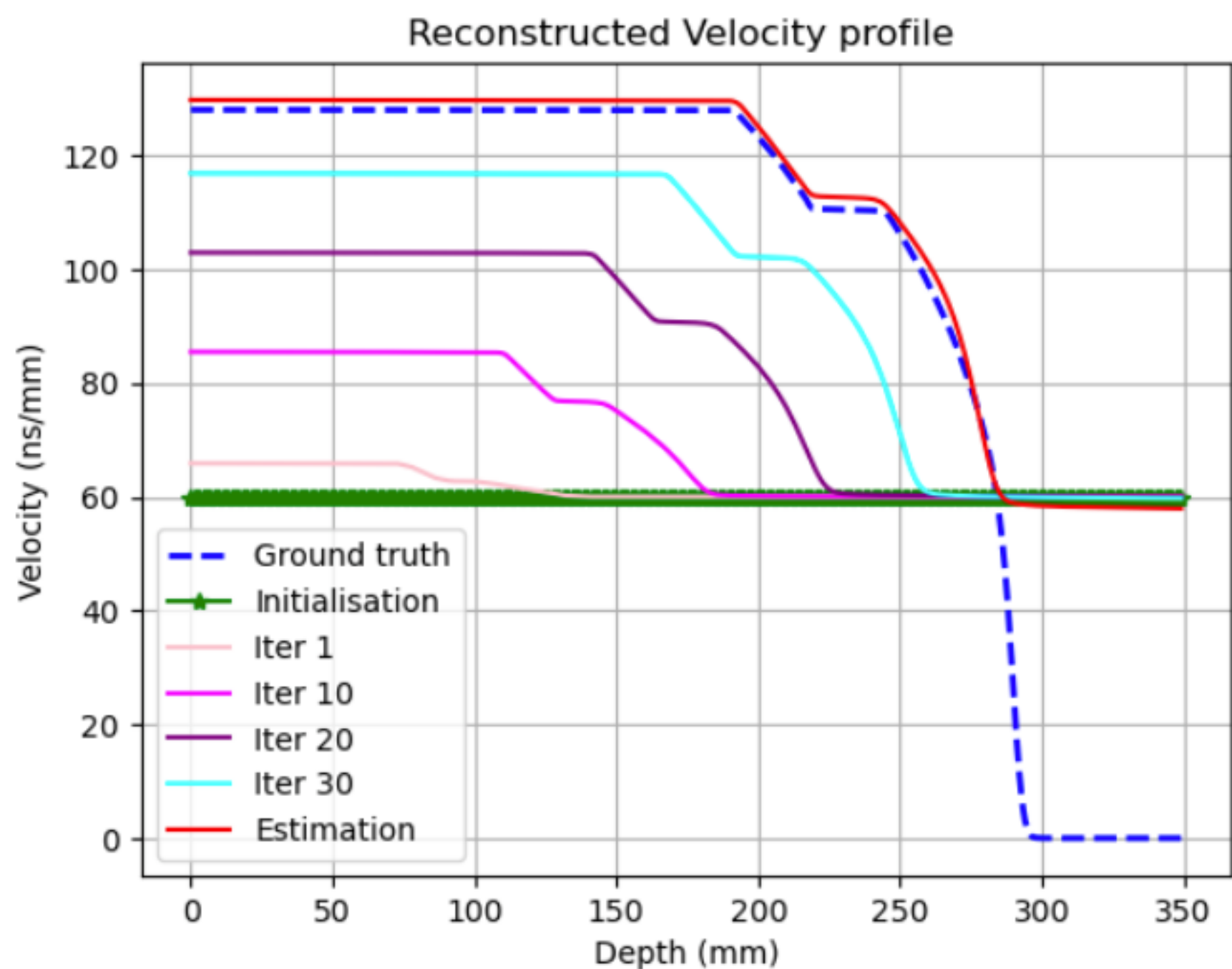
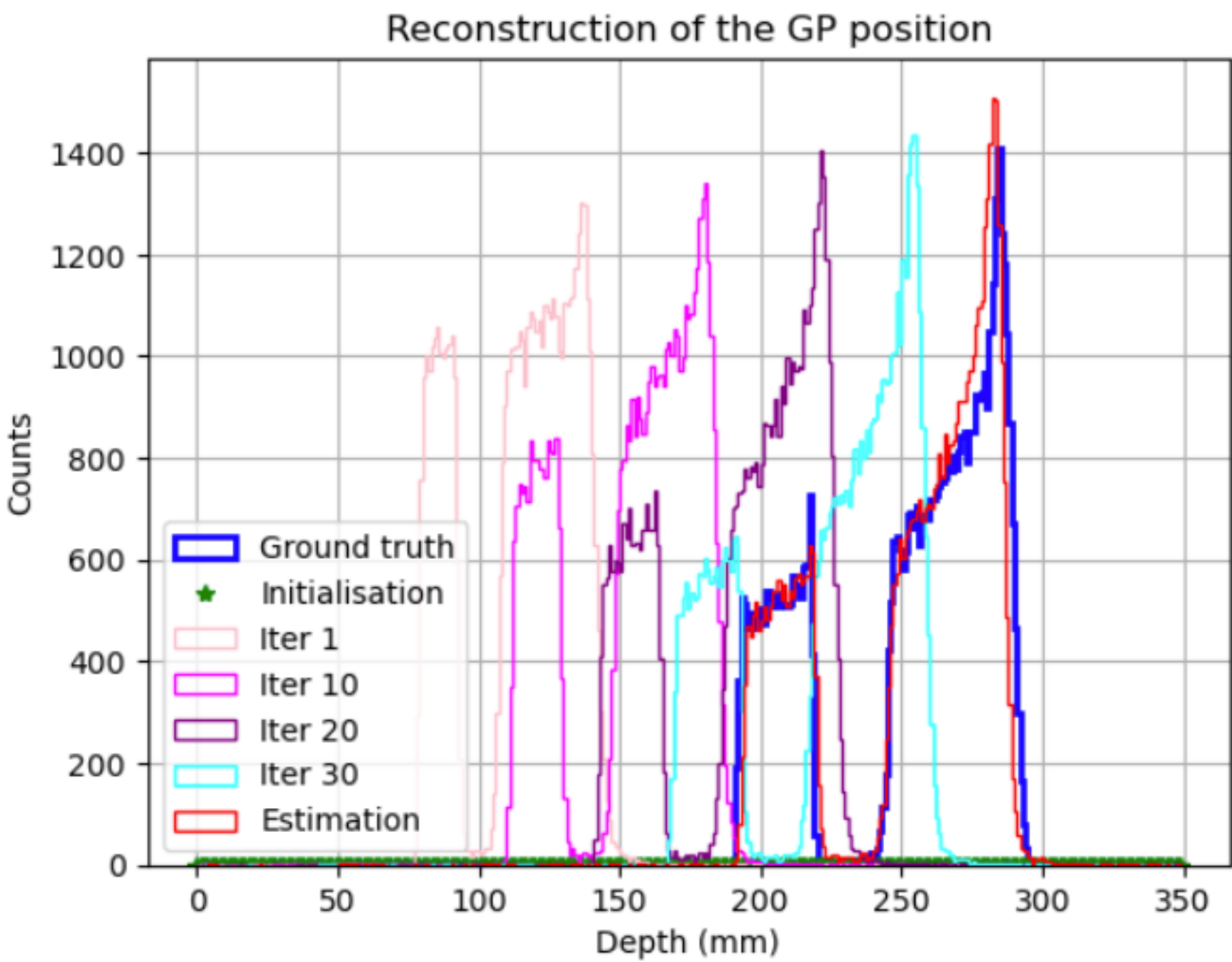
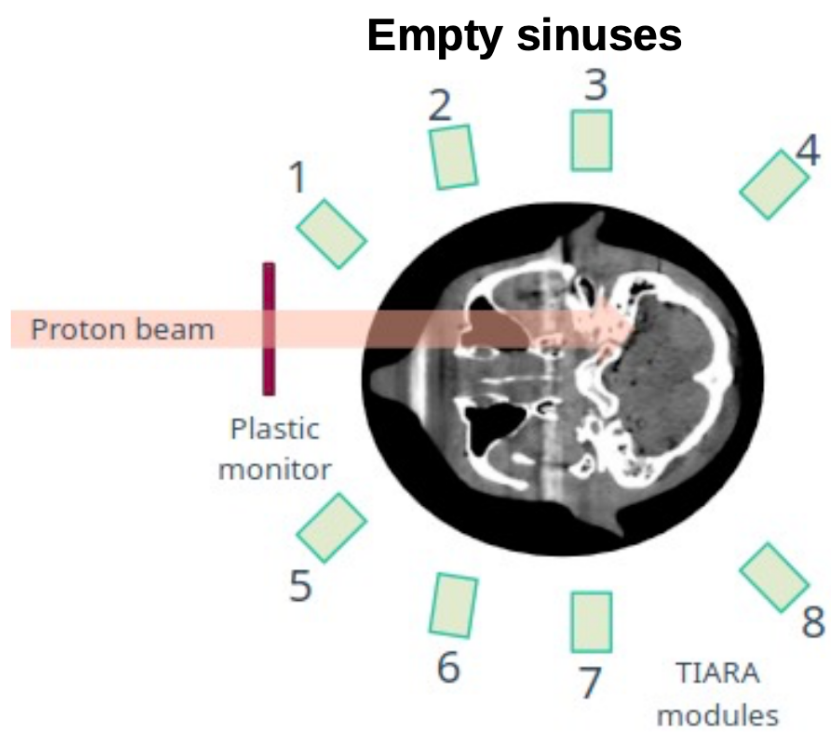
Estimating Both Unknowns via an Alternating Algorithm

Empty sinus case



Estimating Both Unknowns via an Alternating Algorithm - Empty sinus

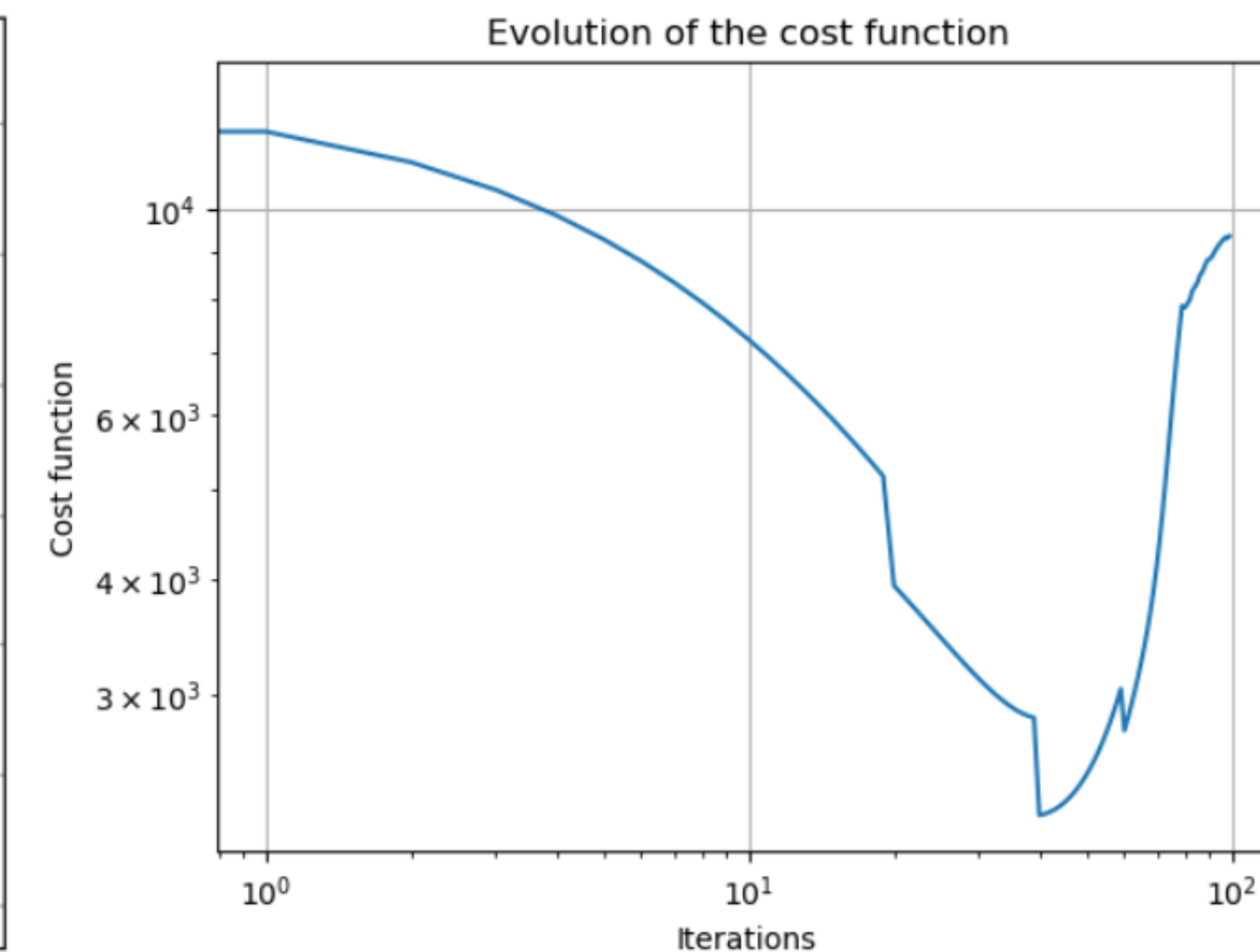
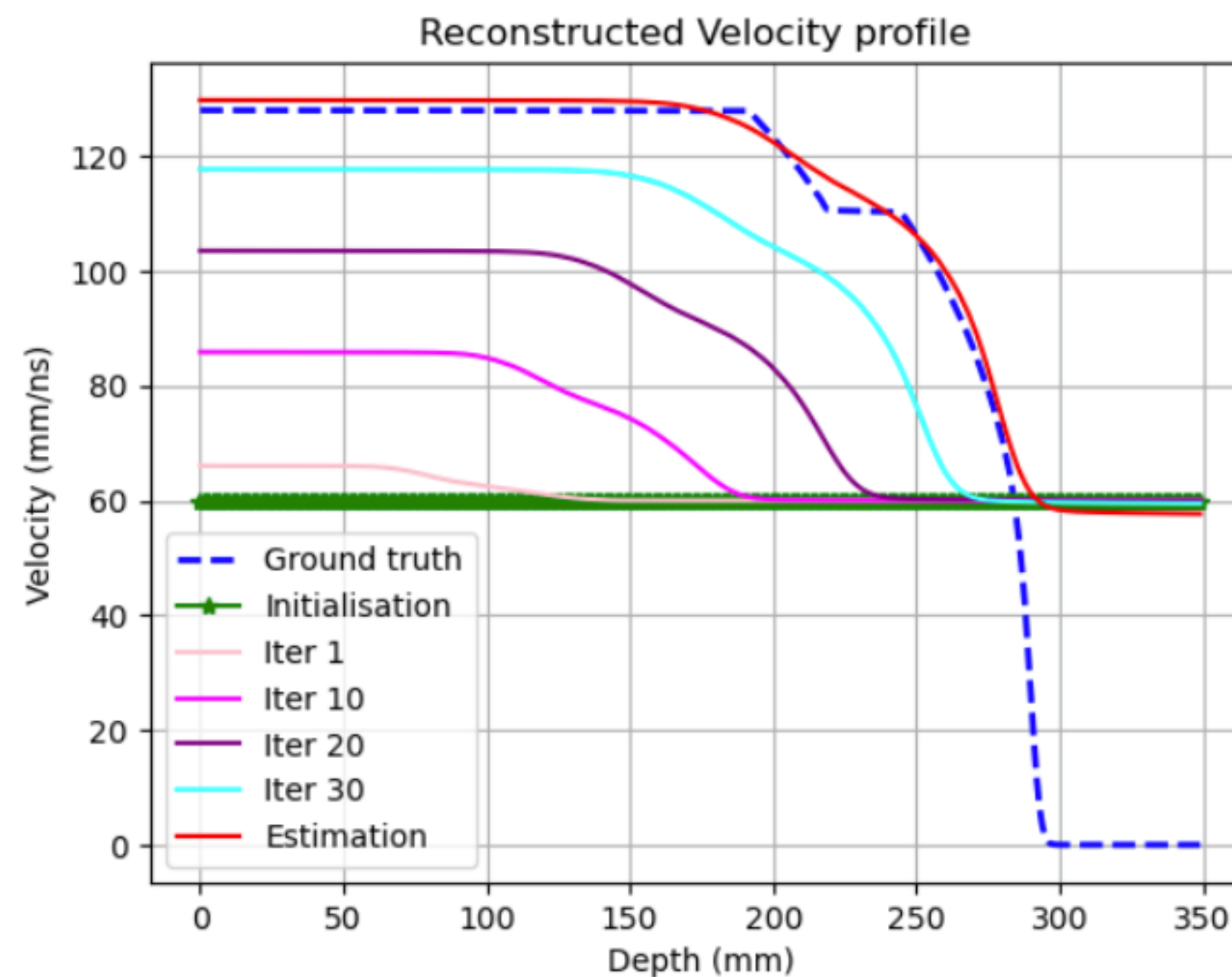
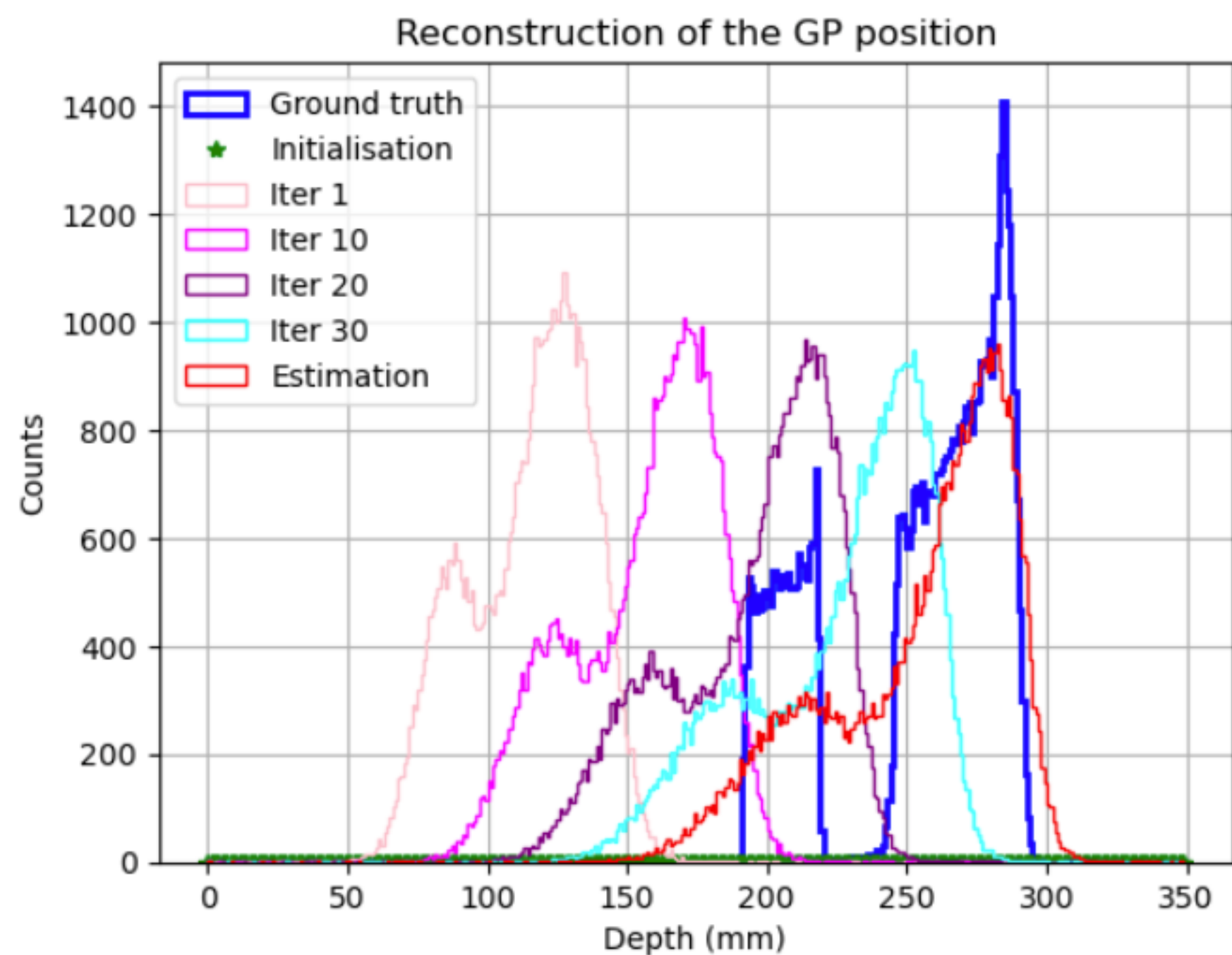
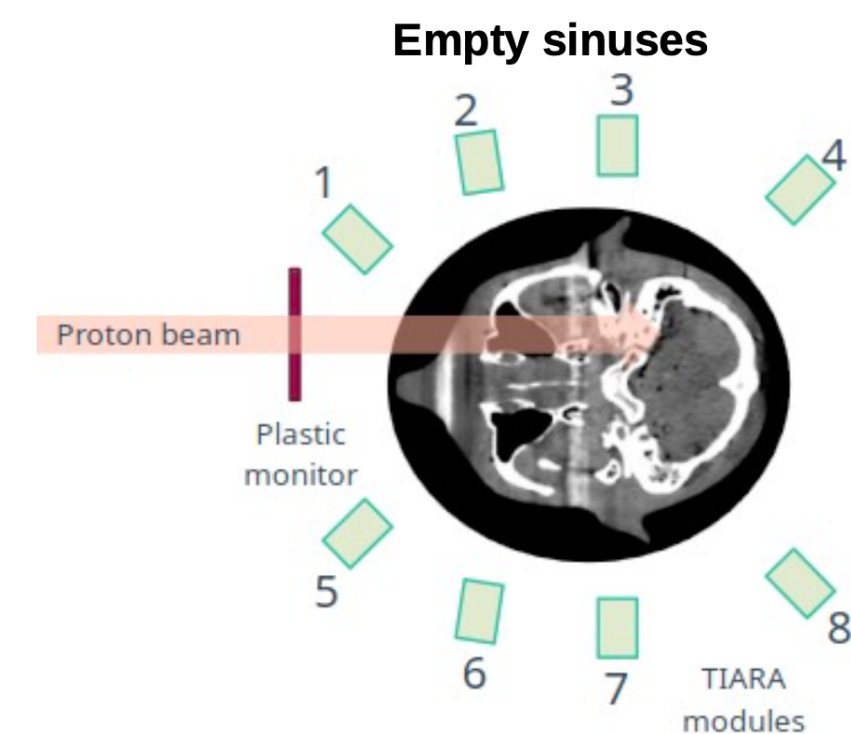
Decaying strategy



Estimating Both Unknowns via an Alternating Algorithm - Empty sinus

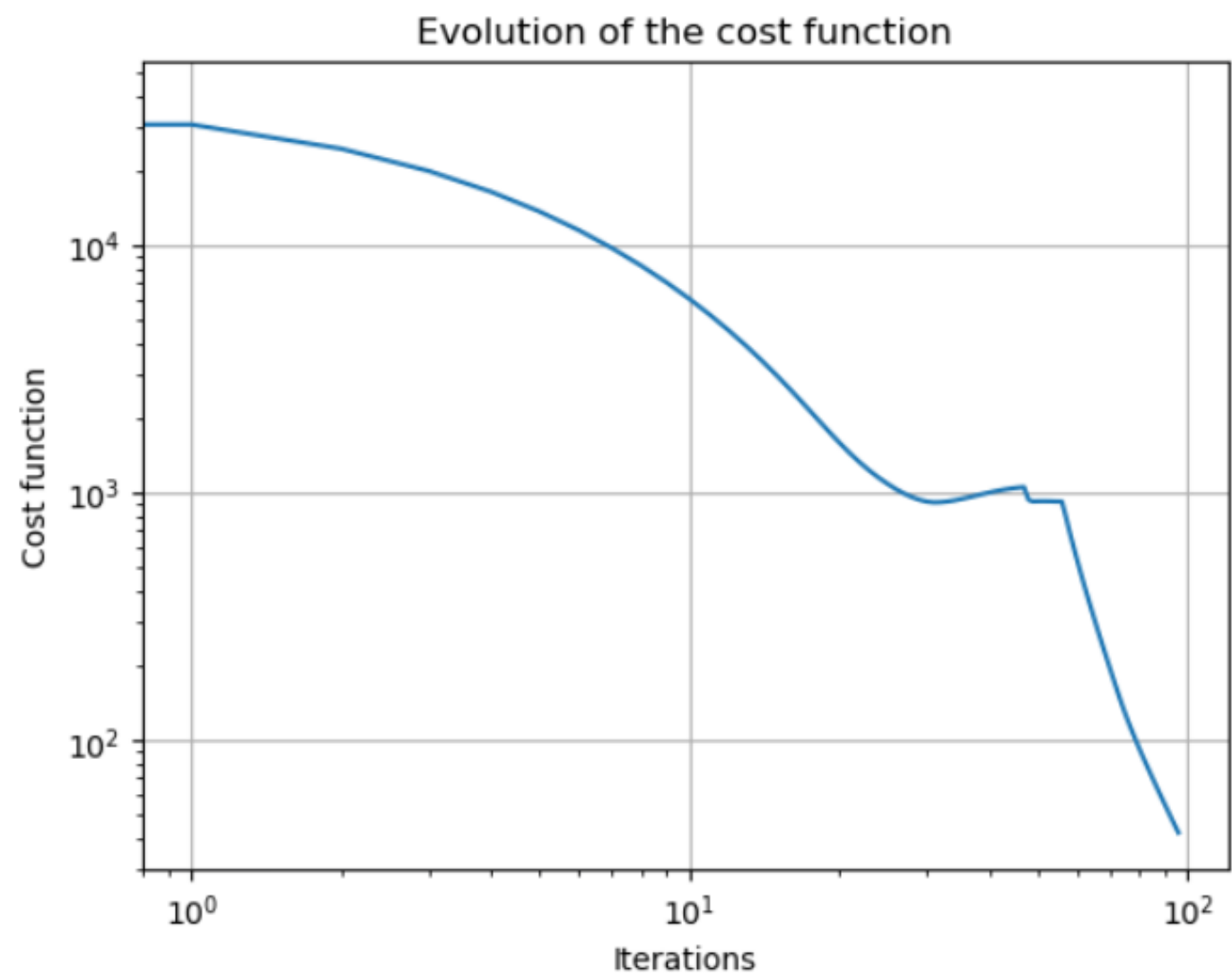
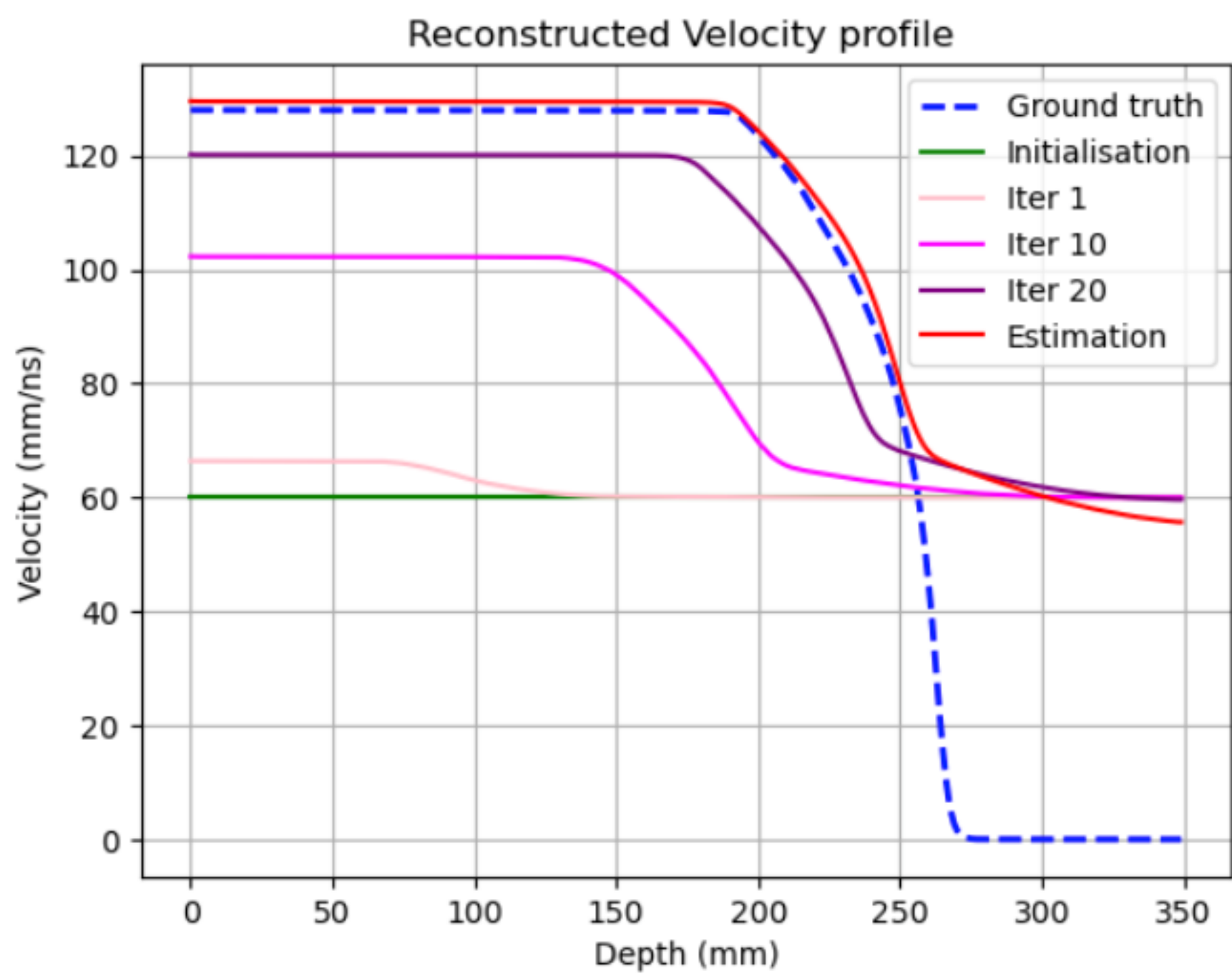
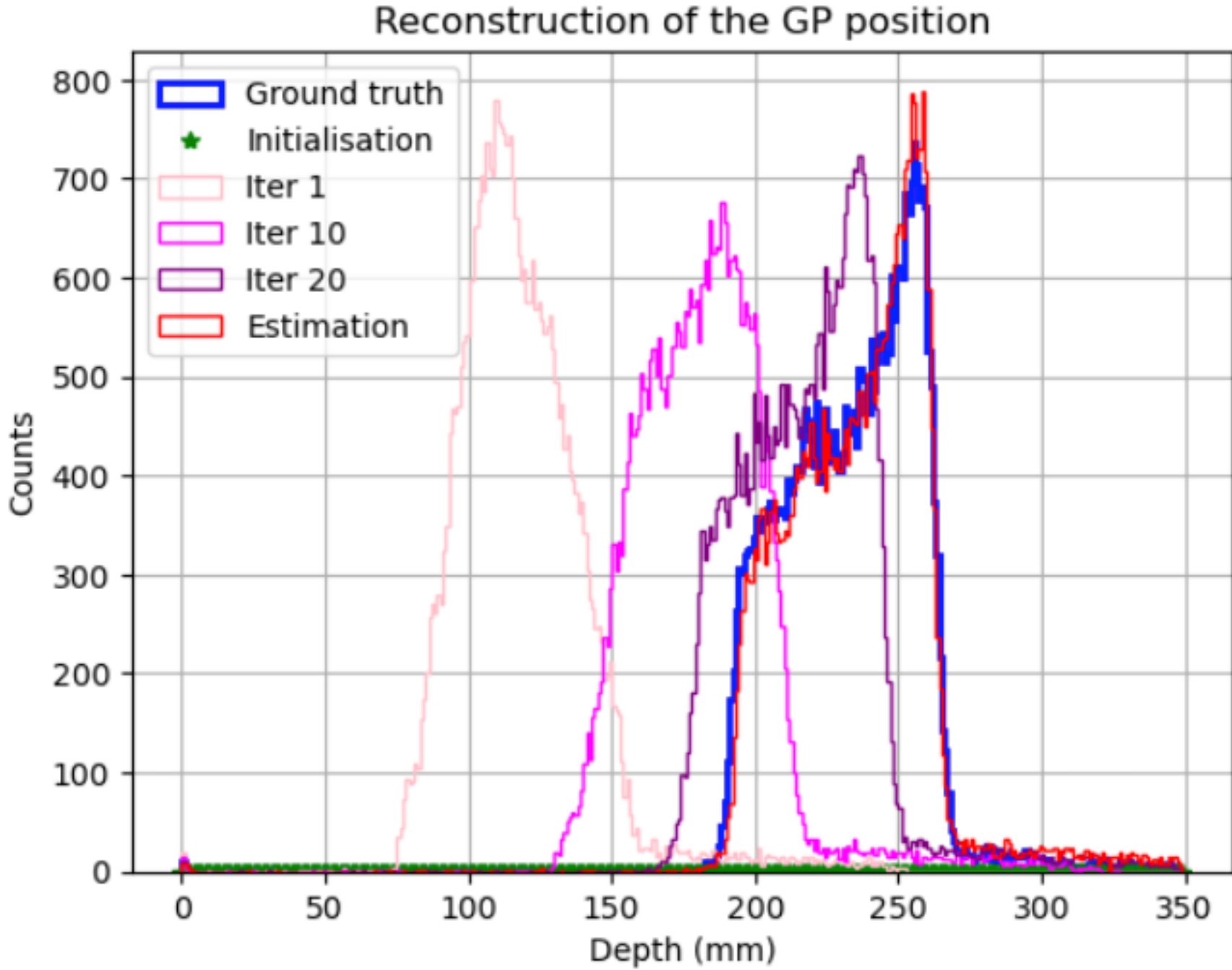
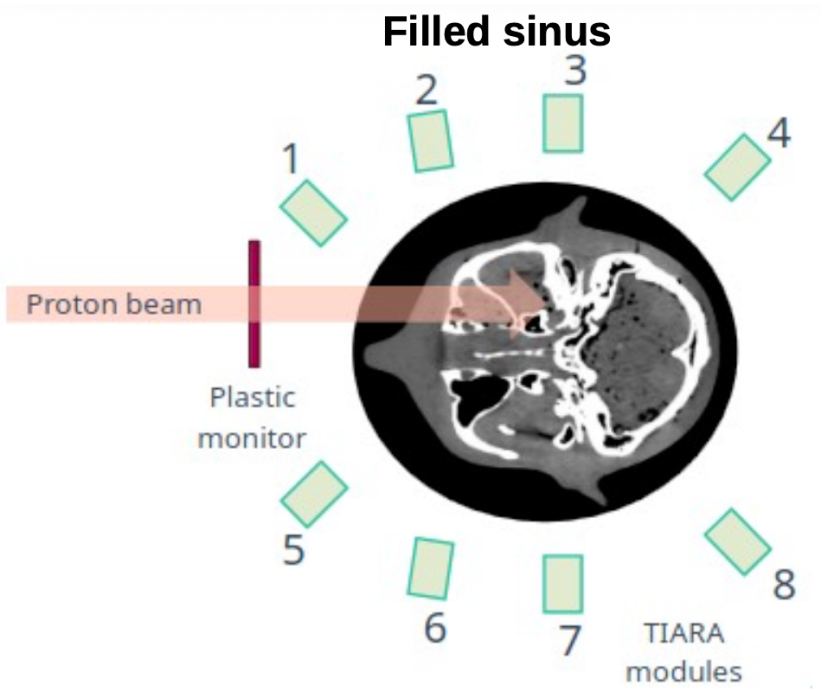
Noisy Data

Noise: $\sigma = 100$ ps



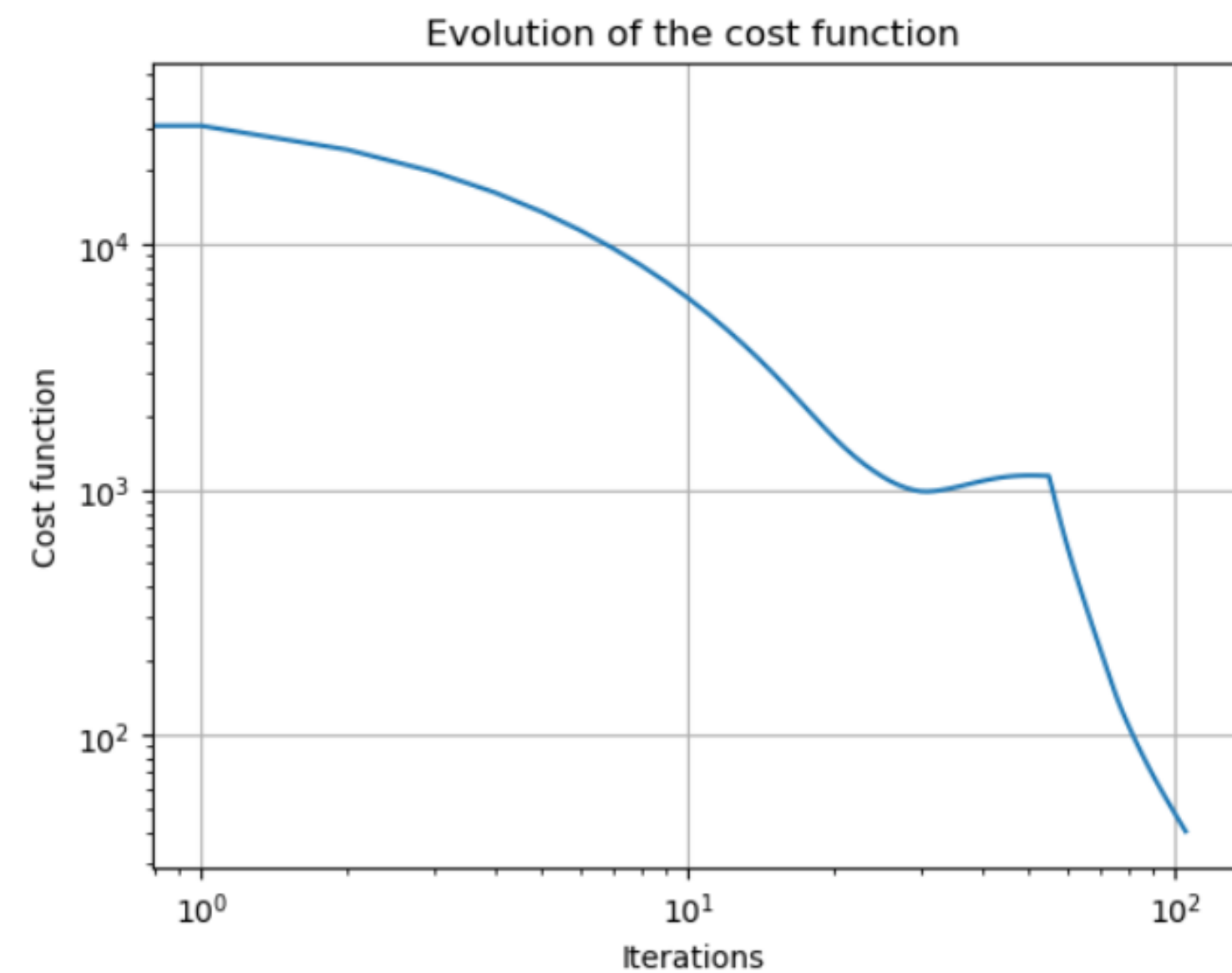
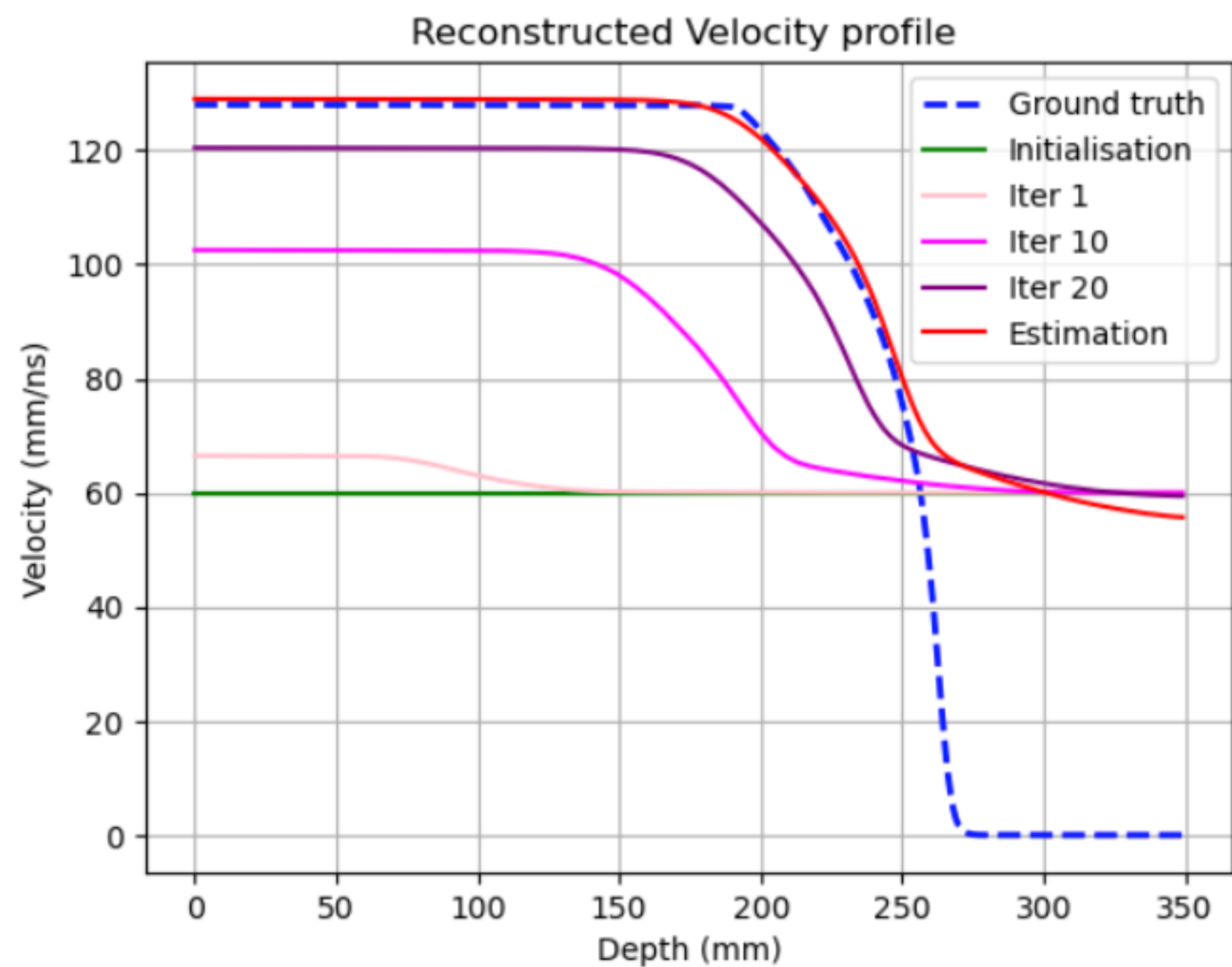
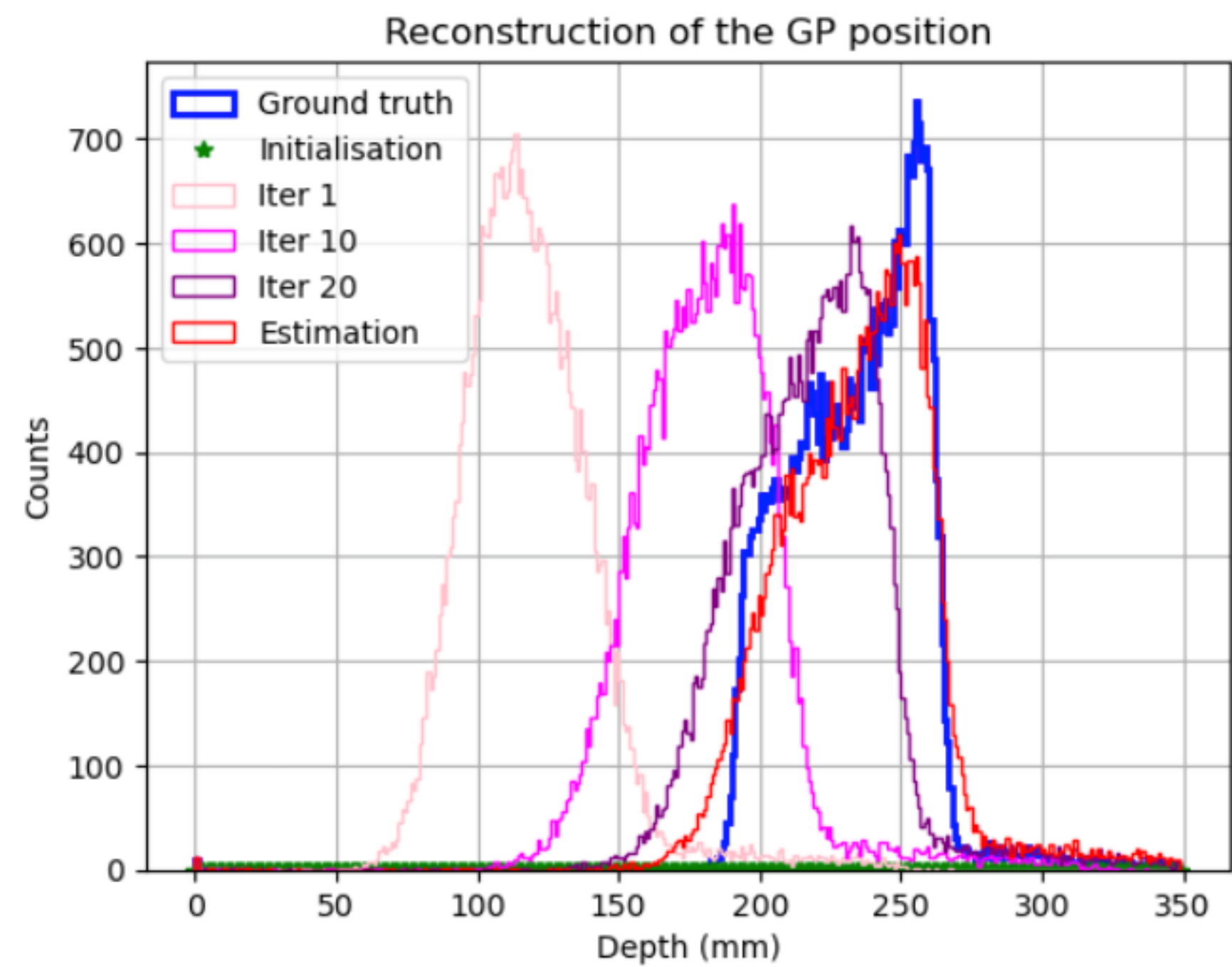
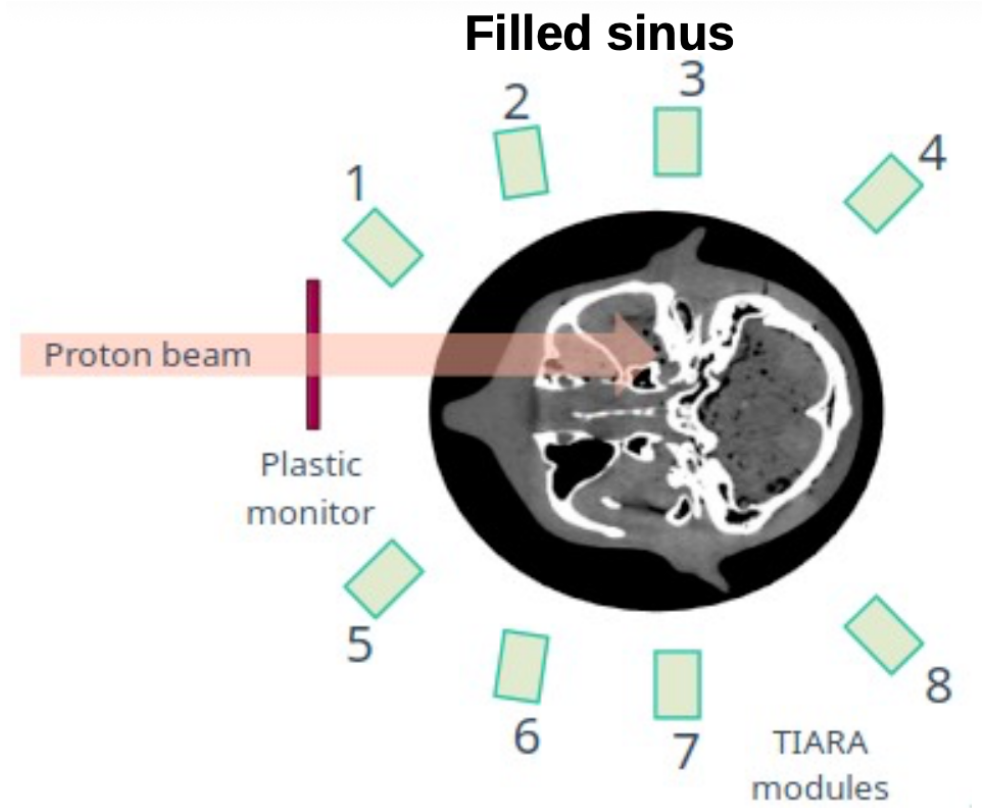
Estimating Both Unknowns via an Alternating Algorithm

Full sinus case



Estimating Both Unknowns via an Alternating Algorithm - Full sinus

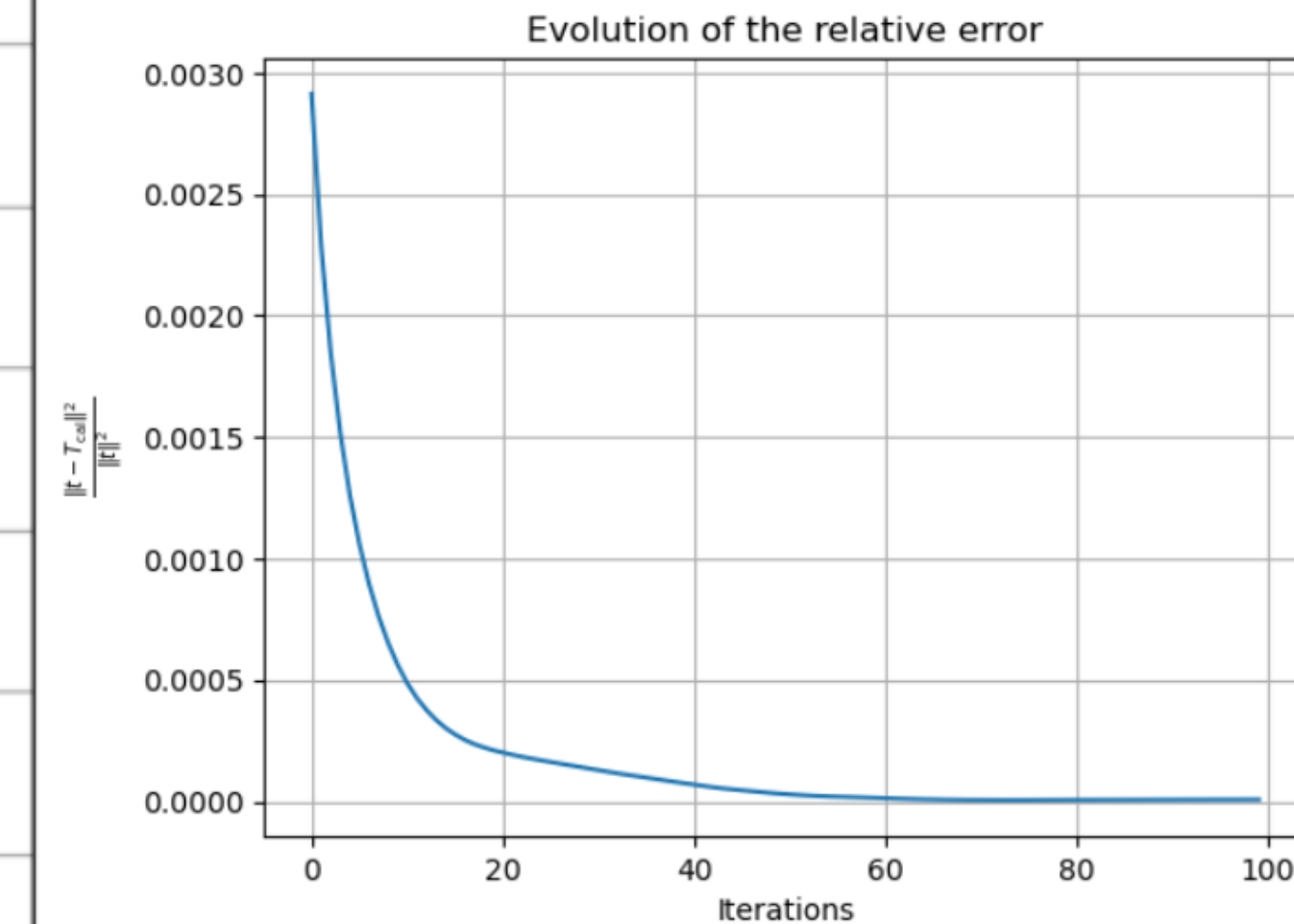
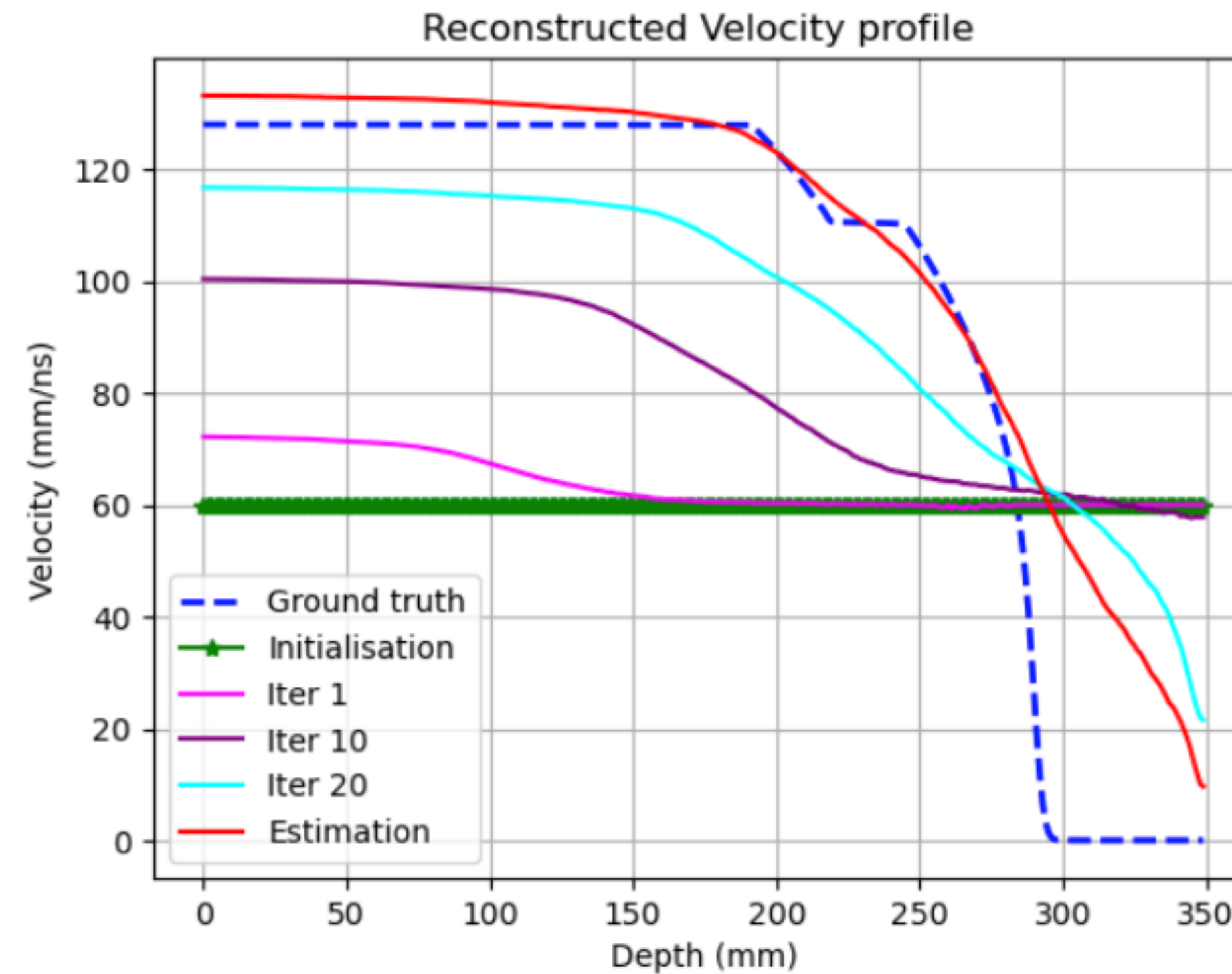
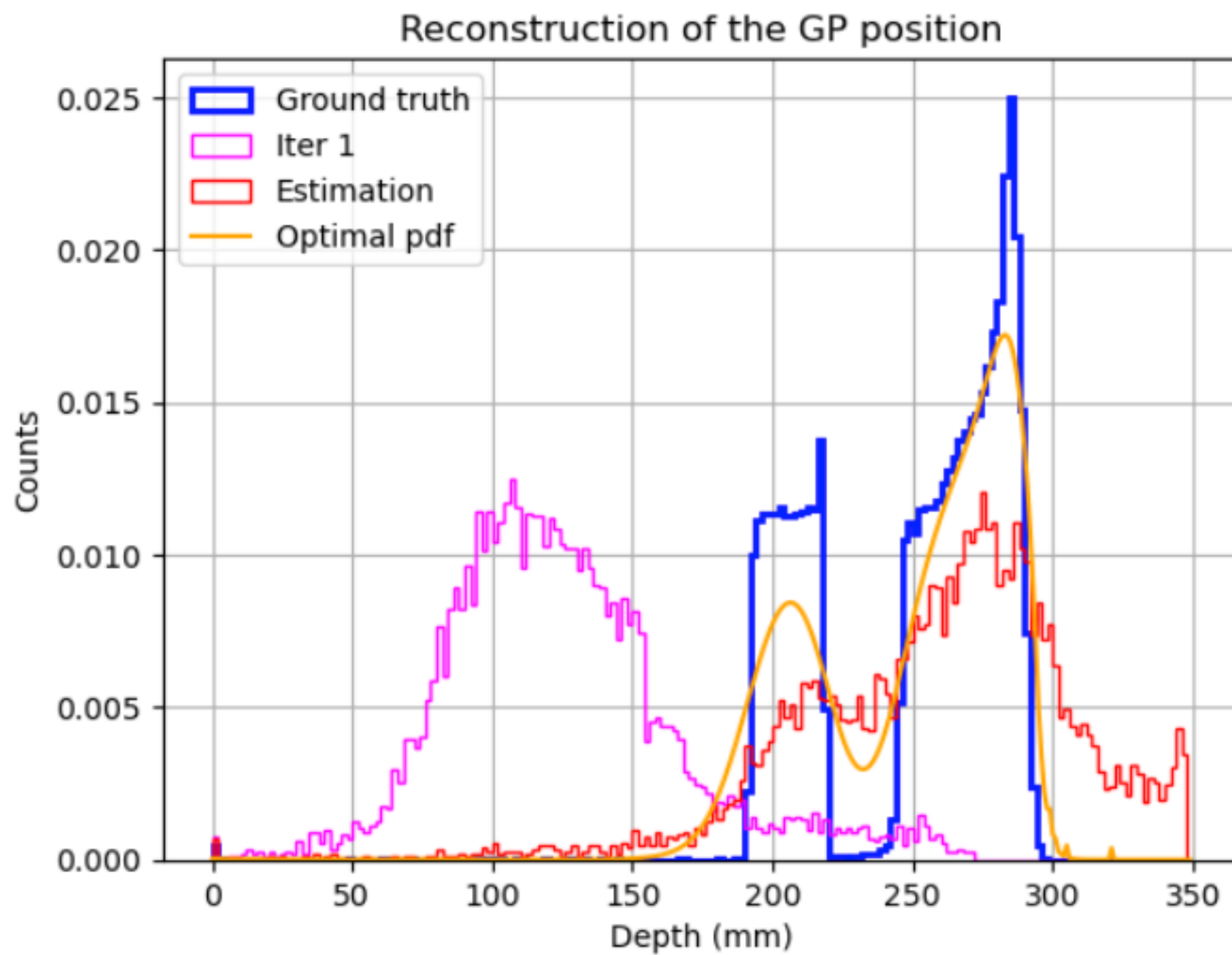
Noisy Data



Preliminary Results

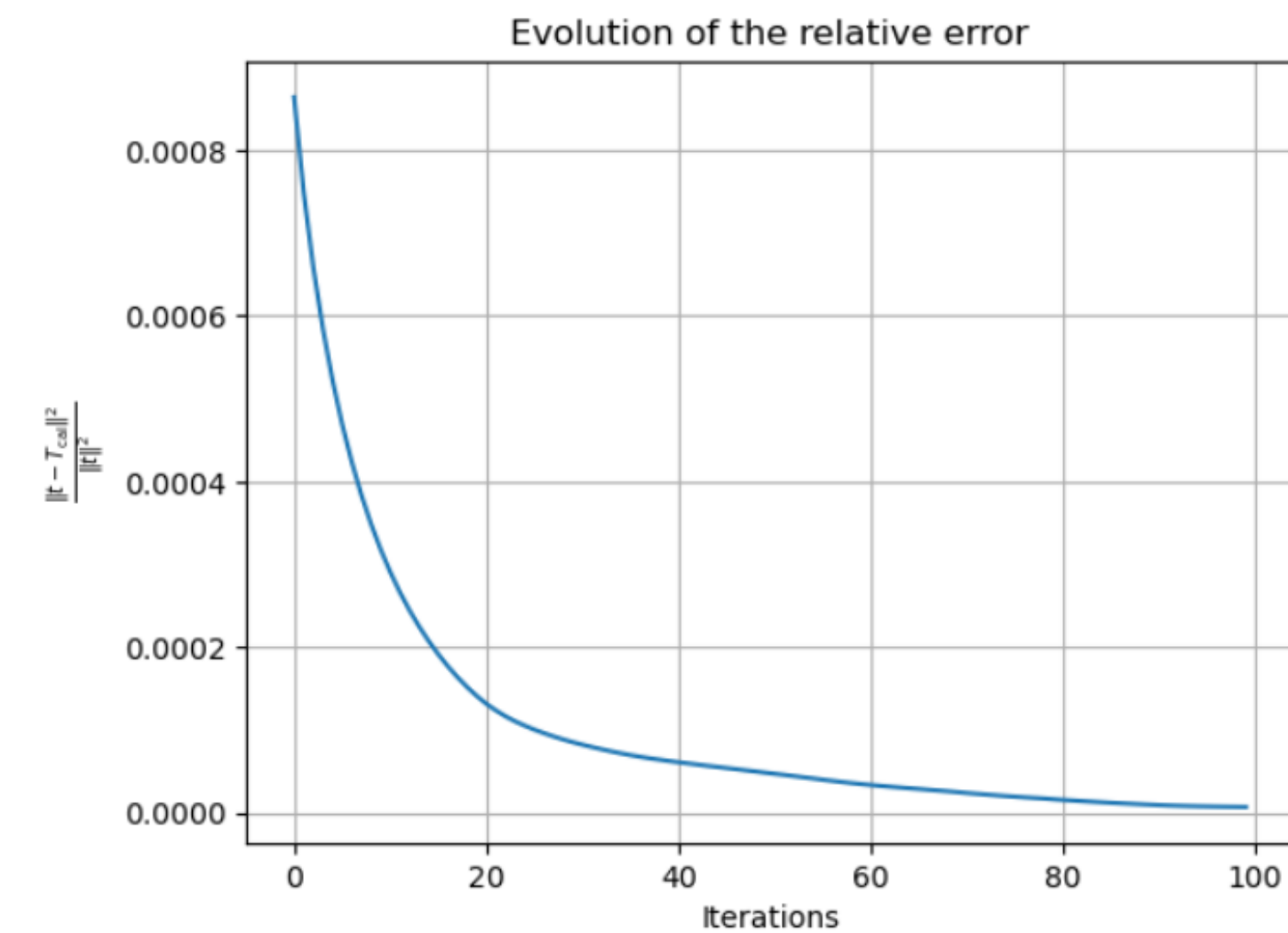
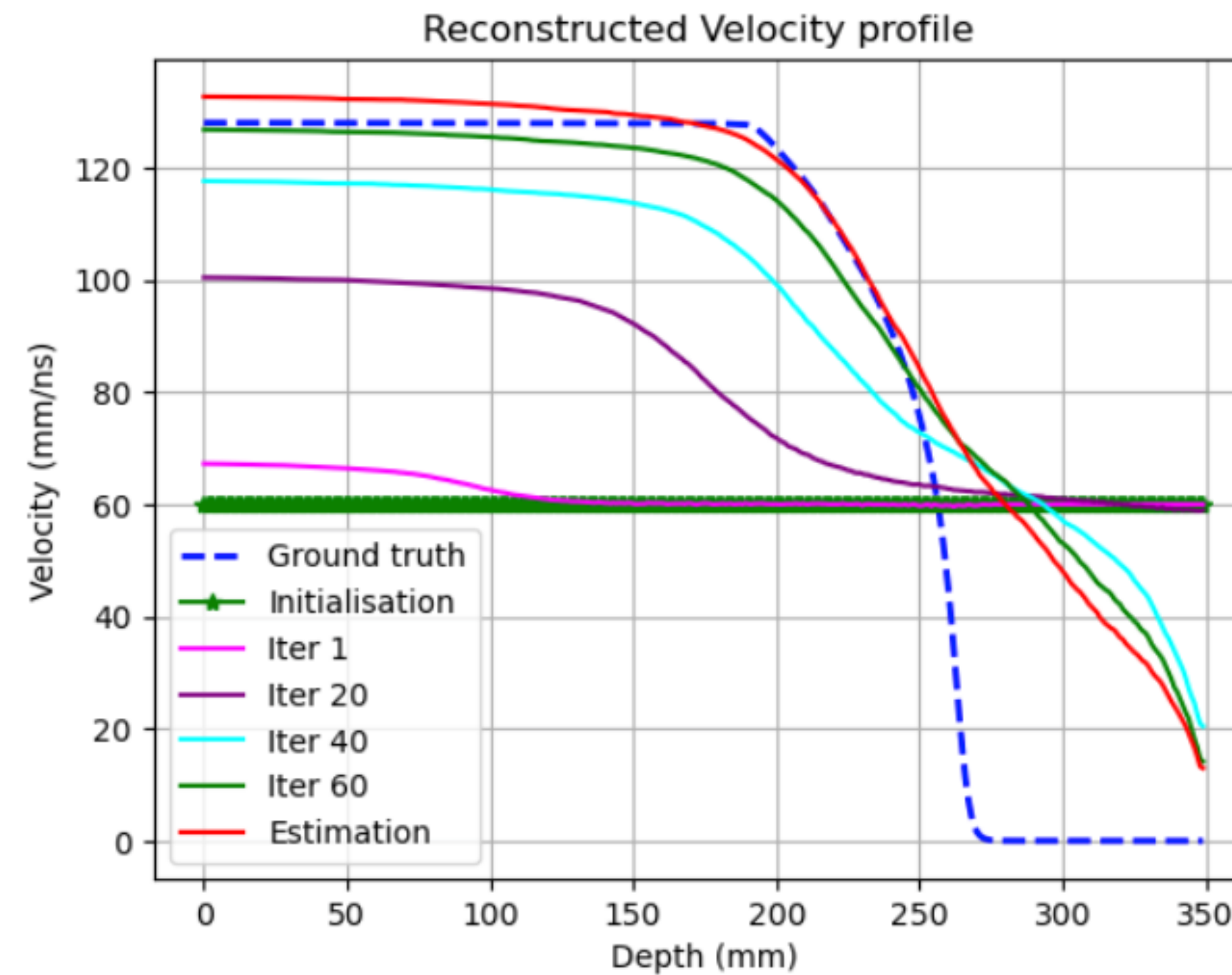
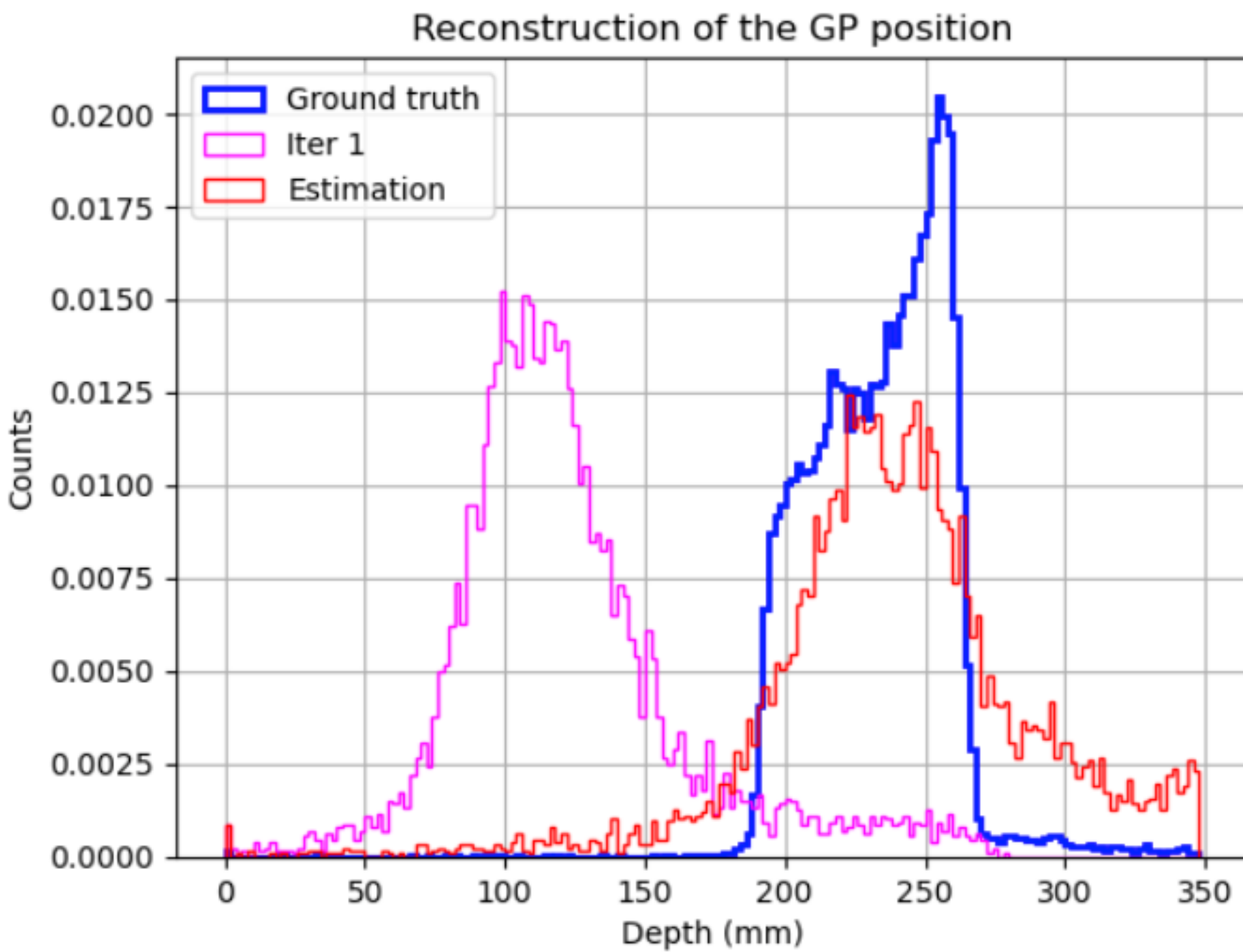
Estimating Both Unknowns via an Alternating Algorithm

(Experimental Data- ES case)



Estimating Both Unknowns via an Alternating Algorithm

(Experimental Data- FS case)



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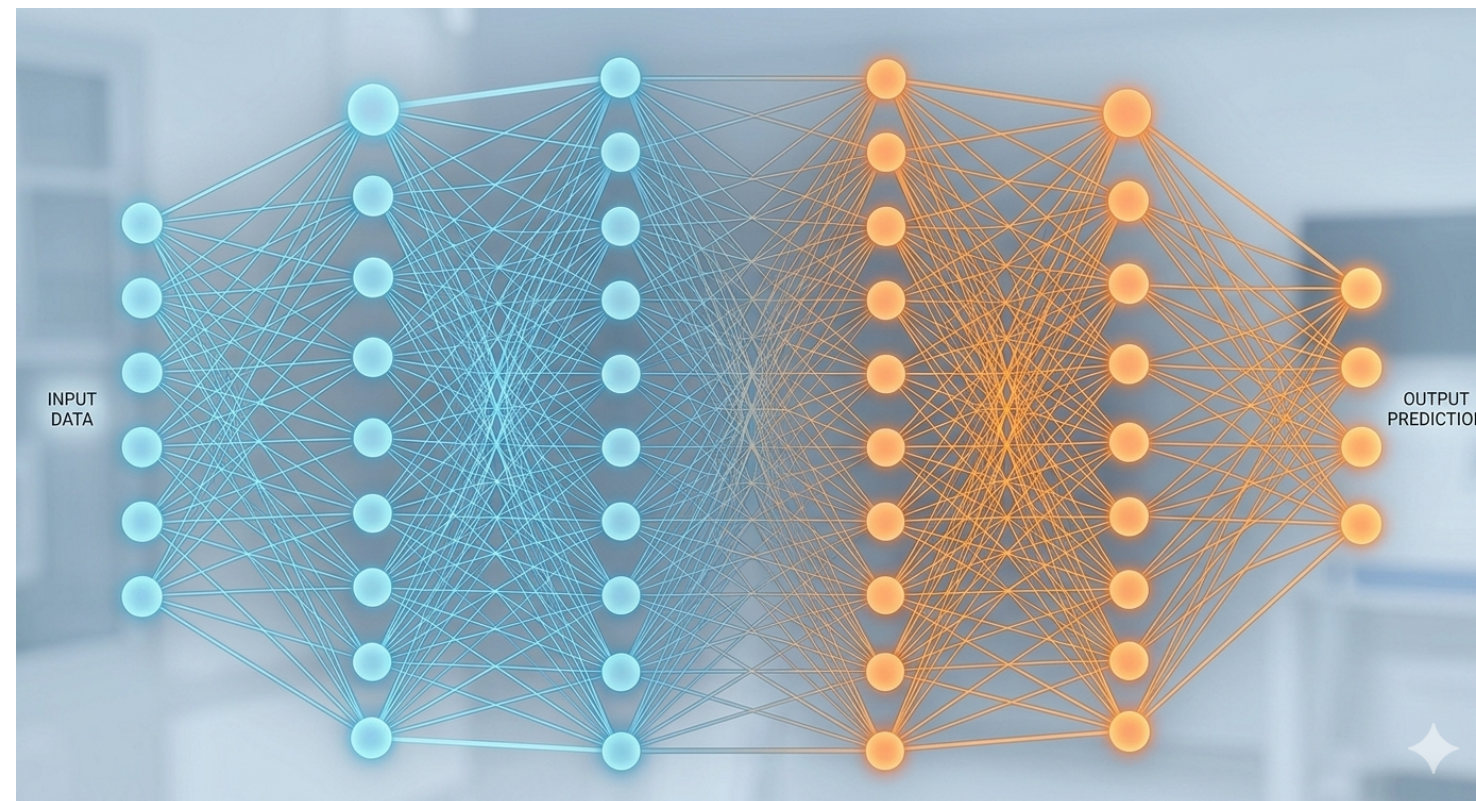
3. An alternating minimization approach

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5. Perspectives

Perspectives

- Using AI in combination with the alternating algorithm to improve result accuracy
- Extending the reconstruction approach to full 3D geometries



Thank you for your attention

The PGTI Collaboration



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European Research Council

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