

# INVERSE PROBLEM FOR LIGHT SPREADING APPLIED TO MEDICAL TOMOGRAPHY

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For the teaching unit: "Projet"

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## CONTEXT

- Proposed by the MOCO Team at IRMA
- Inverse problems encountered in:
  - > medical imaging,
  - > computer vision,
  - > radar, etc.

Hard to solve!!!

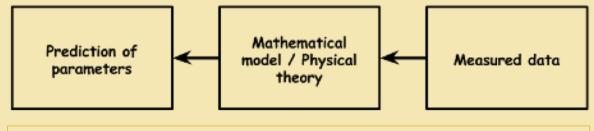


Figure 1: Inverse problem (Subramaniyam, 2018)

#### DESCRIPTION

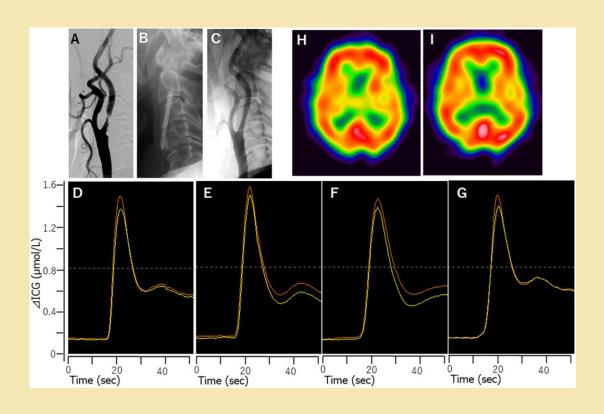


Figure 2: Near-infrared spectroscopy (Nakagawa et al., 2017)

#### **Objectives**

- Solve the forward problem for light spreading
- Get ready for AI prediction using neural networks

#### Tasks

- Read about the mathematical model to be used
- Implement the numerical scheme using the finite volumes method
- Run test cases
- Analyze the data
- Write the reports
- Present the work that was done

#### ROADMAP

#### Mathematical basis

The phenomenon:

- Emission with opacity  $\sigma_e$
- Absorption with opacity  $\sigma_a$
- Scattering with opacity  $\sigma_c$

With major simplifications:

- Local thermodynamic equilibrium
- Radiative equilibrium
- Etc...

Give us the radiative transfer equation (RTE)

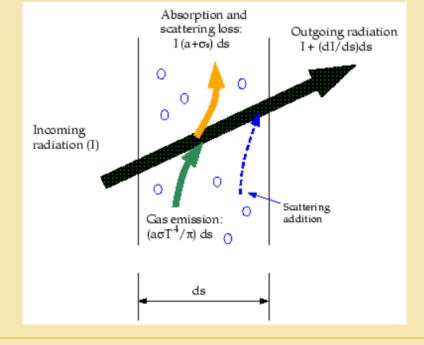


Figure 3: Radiative transfer phenomenon in a gas (Ansys Fluent, n.d)

$$\frac{1}{c} \frac{\partial}{\partial t} I(t, x, \Omega, v) + \Omega \cdot \nabla_{x} I(t, x, \Omega, v) 
= \sigma_{a}(\rho, \Omega, v) (B(v, T) - I(t, x, \Omega, v)) 
+ \frac{1}{4\pi} \int_{0}^{\infty} \int_{c^{2}} \sigma_{c}(\rho, \Omega, v) p(\Omega' \to \Omega) (I(t, x, \Omega', v) - I(t, x, \Omega, v)) d\Omega' dv$$
(RTE)

#### ROADMAP

#### Mathematical basis

Other simplifications lead to the  $P_1$  model:

$$\begin{cases} \partial_t E + c \partial_x F = c \sigma_a (aT^4 - E) \\ \partial_t F + \frac{c}{3} \partial_x E = -c \sigma_c F \\ \rho C_v \partial_t T = c \sigma_a (E - aT^4) \end{cases}$$

#### Where:

- $a = \frac{4\sigma}{c}$  with  $\sigma$  the Stefan–Boltzmann constant
- $C_v$  is the thermal capacity of the medium
- c is the speed of light

- T(t,x) > 0 is the temperature of the medium
- $E(t,x) \in \mathbb{R}$  is the energy of the photons
- $F(t,x) \in \mathbb{R}$  is the flux of the photons
- $\rho(x) > 0$  is the density of the medium
- $\sigma_a(\rho,T) > 0$  is the absorption opacity
- $\sigma_c(\rho, T) > 0$  is the scattering opacity

## ROADMAP Solving the problem

- Using the finite volumes method
- Using a "splitting" scheme
- CFL condition for stability:  $\Delta t < \frac{\Delta x}{c}$

#### Step 1 (the coupling part)

- With  $\Theta = aT^4$
- On each independent cell *j*:
- At each time iteration step n (the big loop)
- We iterate on q (the small loop)

$$\begin{cases} \frac{E_{j}^{q+1} - E_{j}^{n}}{\Delta t} = c\sigma_{a} \left( \Theta_{j}^{q+1} - E_{j}^{q+1} \right) \\ \frac{F_{j}^{q+1} - F_{j}^{n}}{\Delta t} = 0 \\ \rho_{j} C_{v} \mu_{q} \frac{\Theta_{j}^{q+1} - \Theta_{j}^{n}}{\Delta t} = c\sigma_{a} \left( E_{j}^{q+1} - \Theta_{j}^{q+1} \right) \end{cases}$$

Where: 
$$\mu_q = \frac{1}{T^{3,n} + T^n T^{2,q} + T^q T^{2,n} + T^{3,q}}$$

• Until we get to the fixed point  $(E_i^*, F_i^*, \theta_i^*)$ 

#### Step 2 (the hyperbolic part)

• We iterate on n (the big loop)

$$\begin{cases} \frac{E_{j}^{n+1} - E_{j}^{*}}{\Delta t} + c \frac{F_{j+\frac{1}{2}} - F_{j-\frac{1}{2}}}{\Delta t} = 0\\ \frac{F_{j}^{n+1} - F_{j}^{*}}{\Delta t} + c \frac{E_{j+\frac{1}{2}} - E_{j-\frac{1}{2}}}{\Delta t} = cS_{j}\\ \rho_{j} C_{v} \mu_{q} \frac{T_{j}^{n+1} - T_{j}^{*}}{\Delta t} = 0 \end{cases}$$

Where:

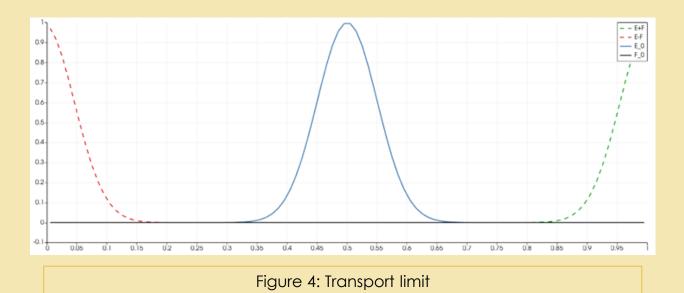
$$\begin{split} F_{j+\frac{1}{2}} &= M_{j+\frac{1}{2}} \left( \frac{F_{j+1}^n + F_j^n}{2} - \frac{E_{j+1}^n - E_j^n}{2} \right), \quad F_{j-\frac{1}{2}} &= M_{j-\frac{1}{2}} \left( \frac{F_j^n + F_{j-1}^n}{2} - \frac{E_j^n - E_{j-1}^n}{2} \right) \\ E_{j+\frac{1}{2}} &= M_{j+\frac{1}{2}} \left( \frac{E_{j+1}^n + E_j^n}{2} - \frac{F_{j+1}^n - F_j^n}{2} \right), \quad E_{j-\frac{1}{2}} &= M_{j-\frac{1}{2}} \left( \frac{E + E_{j-1}^n}{2} - \frac{F_j^n - F_{j-1}^n}{2} \right) \\ S_j &= -\frac{1}{2} \left( M_{j+\frac{1}{2}} \sigma_{j+\frac{1}{2}} + M_{j-\frac{1}{2}} \sigma_{j-\frac{1}{2}} \right) F_j^{n+1} \\ M_{j+\frac{1}{2}} &= \frac{2}{2 + \Delta x \sigma_{j+\frac{1}{2}}}, \quad M_{j-\frac{1}{2}} &= \frac{2}{2 + \Delta x \sigma_{j-\frac{1}{2}}} \\ \sigma_{j+\frac{1}{2}} &= \frac{1}{2} \left( \sigma_c \left( \rho_j, T_j^n \right) + \sigma_c \left( \rho_{j+1}, T_{j+1}^n \right) \right), \quad \sigma_{j-\frac{1}{2}} &= \frac{1}{2} \left( \sigma_c \left( \rho_{j-1}, T_{j-1}^n \right) + \sigma_c \left( \rho_j, T_j^n \right) \right) \end{split}$$

Until we reach the final time



#### Test case 1: The transport limit

We take  $\sigma_a = \sigma_c = 0$  and we place a Gaussian function at the center of the domain. E + F is transported at the speed c while E - F is transported at the speed -c.



#### Test case 2: The diffusion limit

We take  $\sigma_a = 0$  and  $\sigma_c = c = 1000$  and  $t_0 = 0.002$ .

$$E(t,x) = \frac{1}{\sqrt{4\pi(t+t_0)}} e^{-\frac{\left(x-\frac{1}{2}\right)^2}{4(t+t_0)}}$$

is solution to the diffusion equation

$$\begin{cases} \frac{\partial}{\partial t} E(t, x) - \frac{\partial^2}{\partial x^2} E(t, x) = 0, & t > 0, \ x \in [0, 1] \\ E_0(x) = E(0, x) = \frac{1}{\sqrt{4\pi t_0}} e^{-\frac{\left(x - \frac{1}{2}\right)^2}{4t_0}} \end{cases}$$

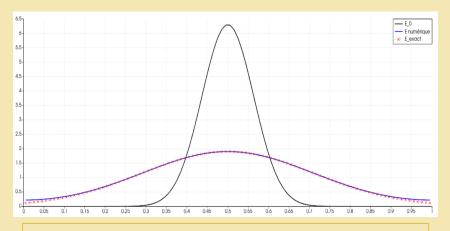


Figure 5: Spatial visualization for the diffusion limit

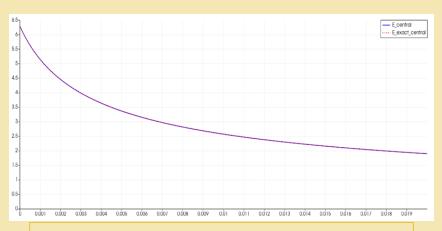


Figure 6: Temporal decrease in energy at the center of the domain for the diffusion limit

#### Test case 3: Olson-Auer-Hall

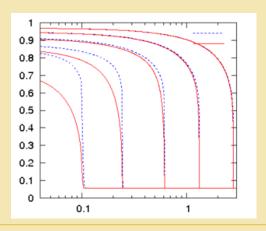


Figure 7: Expected results for t=1/c, 3/c, 10/c, 30/c, and 100/c (Franck, 2012, p.171)

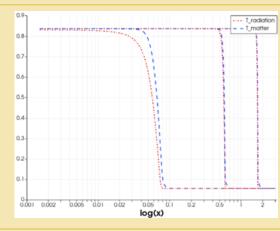


Figure 8: Obtained results for t=0.1/c, 1/c, and 3/c

The domain [0,3] is heated by a blackbody to the left (with  $T_r = 0.056234 \ keV$ ). As the body gets hotter, its absorption opacity increases, causing thermal equilibrium between the radiation and the matter (Franck, 2012, p.171)



Influence of the density

It seems that as the density increases:

- The radiative energy E increases
- It doesn't change the flux F
- The temperature varies less.
   The equilibrium temperature seems to be higher.

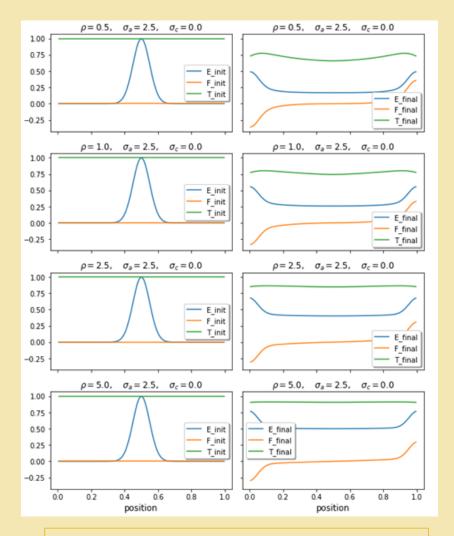


Figure 9: : Influence of the density on E, F, and T for  $\sigma_a=2.5$  and  $\sigma_c=0$ 



Influence of the absorption opacity

It seems that as the opacity  $\sigma_a$  increases:

- The diffusion phenomenon is faster, but the energy E keeps the same pic value
- The flux F is attenuated during transport
- The temperature change is more important, and the thermal equilibrium is faster

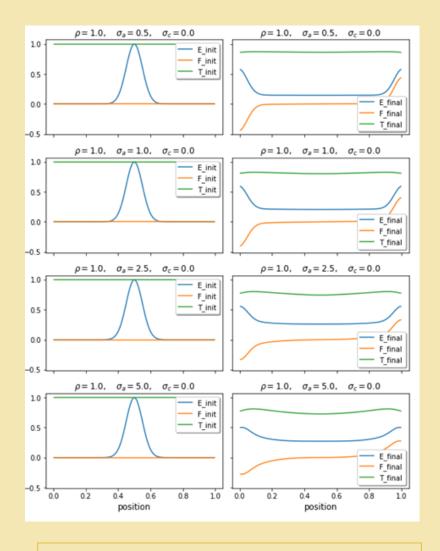


Figure 10: Influence of  $\sigma_a$  for  $\rho=1$  and  $\sigma_c=0$ 

Influence of the scattering opacity

It seems obvious that an increase in  $\sigma_c$ :

- Decreases the values of both the energy and the flux.
- However, to the naked eye, it is less obvious to notice any change in temperature.

These are just conjectures. With the help of a neural network, we could confirm or invalidates these trends.

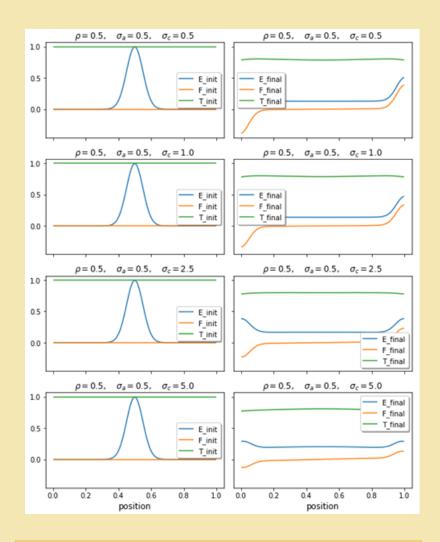
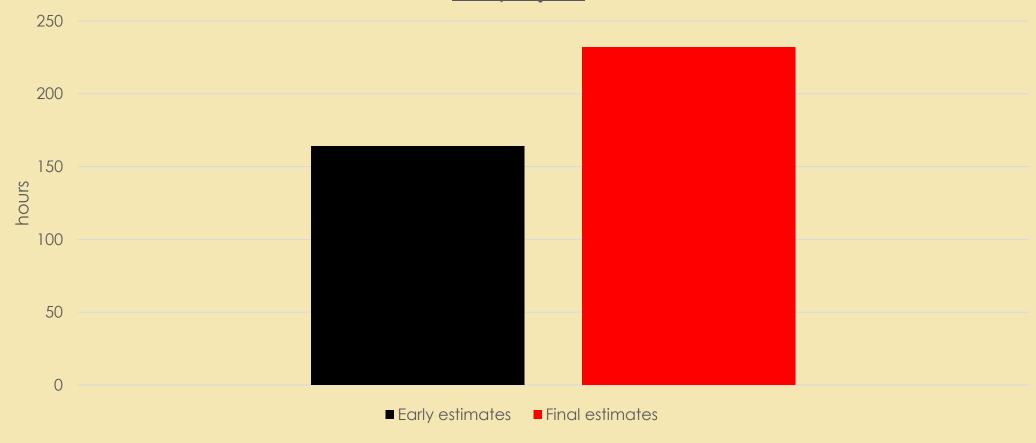


Figure 11: Influence of  $\sigma_c$  for  $\rho=0.5$  and  $\sigma_a=0.5$ 

## MILESTONES

Comparison between initial and final **estimates** for the time spent on the project



# CHALLENGES

- Notations and nomenclature (stellar atmosphere in astrophysics, nuclear fusion by inertial confinement, etc...)
- Performance in the C++ implementation of the scheme



#### PERSPECTIVES

- Since the  $P_1$  model only works well for  $f = \frac{\|F\|}{cE} \le 0.57$  "(Turpault, 2003, p.22)", use a more precise model for the radiative transfer equation  $(P_n, M_1, \text{ etc..})$ .
- Train a neural network that will then predict the optical properties of the medium.

#### THANK YOU!



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