

Application of a Statistically Based, Iterative Image Reconstruction Algorithm for Proton Computed Tomography

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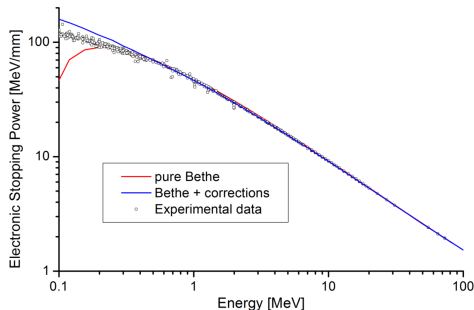
² Budapest University of Technology and Economics

on behalf of Bergen proton CT collaboration
(full collaboration list)

The Bethe Formula – How to Apply?

The Bethe formula describes the energy loss of charged particle travelling through material. The non relativistic version of formula was introduced by Bethe in 1930, and the relativistic in 1932.

$$-\left\langle \frac{dE}{dx} \right\rangle = K z^2 \frac{Z}{A} \frac{1}{\beta^2} \left[\frac{1}{2} \ln \frac{2m_e c^2 \beta^2 \gamma^2 T_{max}}{I^2} - \beta^2 - \frac{\delta(\beta\gamma)}{2} \right],$$



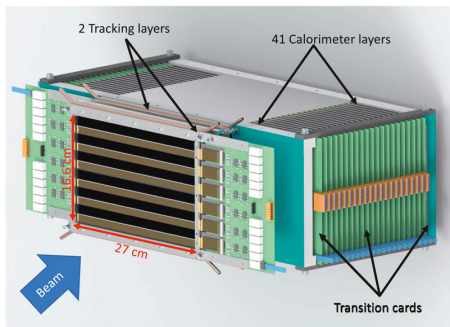
The Importance of Proton CT Imaging

- A CT measures the distribution of a material property
 - Relative stopping power (RSP) in case of pCT
- Nowadays RSP distribution is converted from X-ray CT
 - Large uncertainty \Rightarrow significant reduction with pCT
 \Rightarrow reduced safety zone around the tumour
- The number of proton facilities is rapidly increasing
 \Rightarrow Increasing importance to develop this technique



Bergen pCT Collaboration

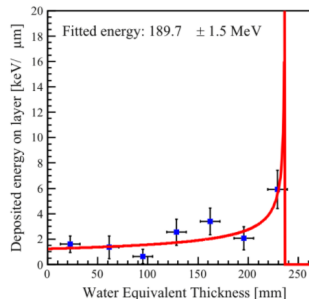
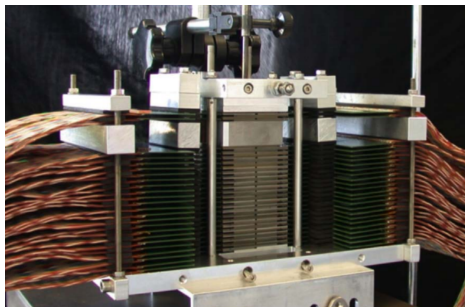
- Goal: reach the clinical testing with a prototype pCT detector
- Monolithic active pixel sensor (MAPS)
- Pencil beam (~ 7 mm)
- Individual measurement of 10^7 proton per second



Bergen pCT Collaboration – Previous Results

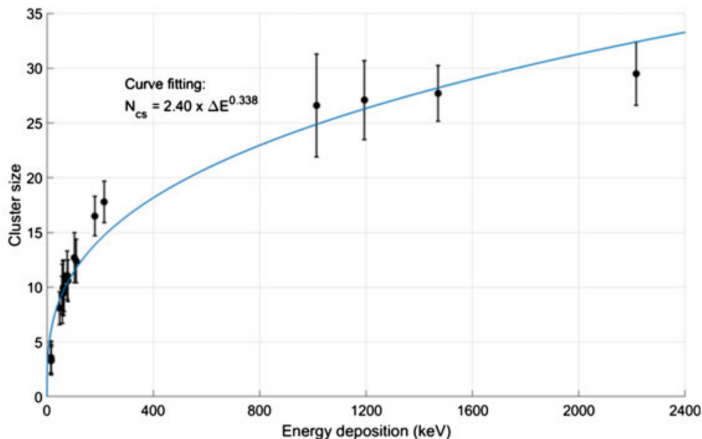
Proof of concept detector:

- Prototype of the high-granularity digital tracking calorimeter for pCT application
- MC simulation 10^6 proton per second
- 4 % water equivalent thickness (WET) range resolution for individual protons



Bergen pCT Collaboration – Previous Results

- Cluster size – deposited energy
- Measured in Heidelberg Ion Beam Therapy Center (HIT)
- Proton, helium and carbon; energy: 50-230 MeV/u



Bergen pCT Collaboration – Wigner Contribution

- **Data analysis of test beam measurements: ALPIDE characterization:**

G. Tambave *et al* (included Á. Sudár), Characterization of monolithic CMOS pixel sensor chip with ion beams for application in particle computed tomography, Nuclear Instruments and Methods in Physics Research Section A, **958**, 2020, doi: 10.1016/j.nima.2019.162626.

- **Numerical analysis of the cooling system:**

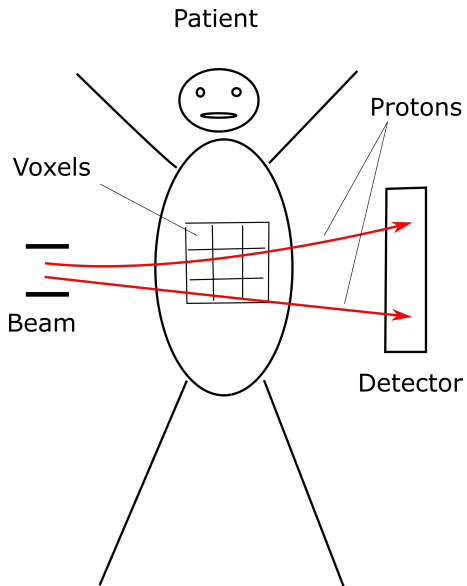
J. Alme *et al* (included Á. Sudár), A High-Granularity Digital Tracking Calorimeter Optimized for Proton CT, Frontiers in Physics, **8**, 2020, doi: 10.3389/fphy.2020.568243.

Bergen pCT Collaboration – Recent Studies

Image reconstruction

Application of the Richardson–Lucy algorithm

Imaging with protons



Relative Stopping Power

Stopping power in units of stopping power of water

⇒ Advantage: almost energy independent (50 - 250 MeV/u)

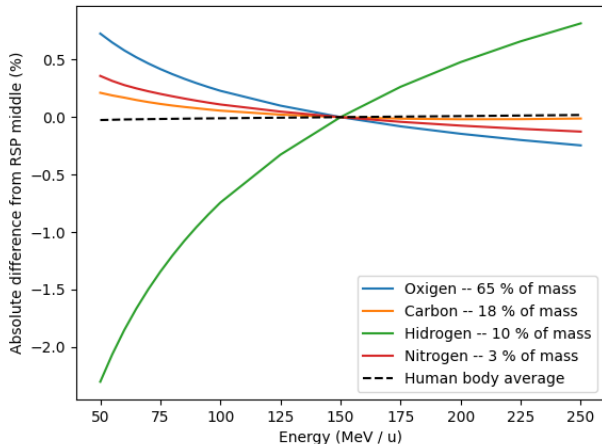


Image Reconstruction – a Huge Linear Problem

The image reconstruction is a huge linear problem:

$$\mathbf{y} = \mathbf{A} \mathbf{x} ,$$

where:

- \mathbf{y} is the energy loss of protons \Leftrightarrow track integral of RSP
- \mathbf{x} RSP value of voxels
- \mathbf{A} system matrix: proton – voxel interaction coefficients

Goal: Solve the linear problem

$$\mathbf{x} = \mathbf{f} (\mathbf{y}, \mathbf{A}) .$$

Image Reconstruction – the Richardson – Lucy algorithm

- First application in the field of proton CT imaging
- Originally developed for astrophysics image reconstruction
- It is a fixed point iteration for sparse systems
- Initialization: arbitrary positive vector
Usually unit vector or approximate solution

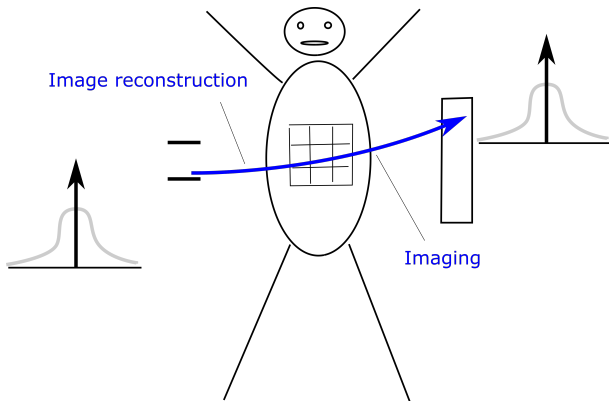
Approximation of the i^{th} voxel of the next iteration:

$$x_i^{k+1} = x_i^k \frac{1}{\sum_j A_{i,j}} \sum_j \frac{y_j}{\sum_l A_{l,j} x_l^k} A_{i,j} ,$$

where k is the iteration number. Typically takes 20-300 iterations.

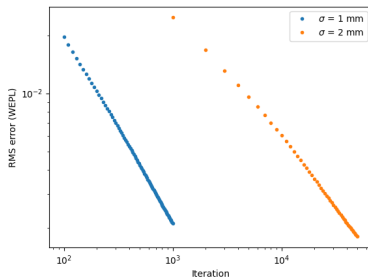
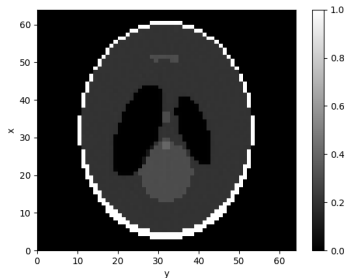
Ideal Imaging

No angular end spatial error in the endpoints



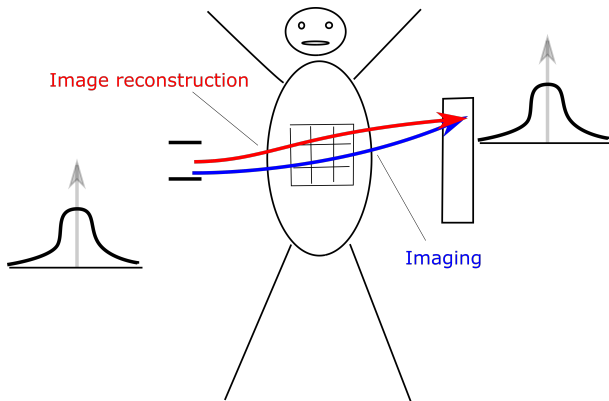
Ideal Imaging – Shepp–Logan Phantom

- Reconstructed RSP distribution and convergence



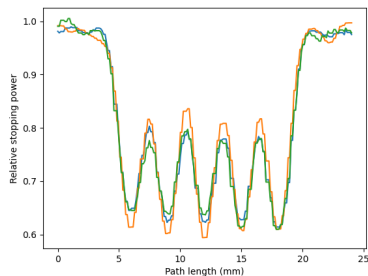
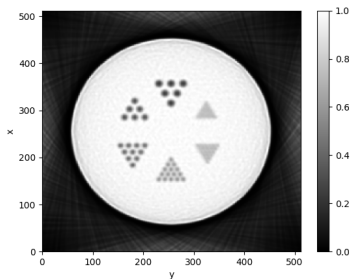
Realistic Imaging – Taking into Account Errors

Analytically calculated angular end spatial error in the endpoints



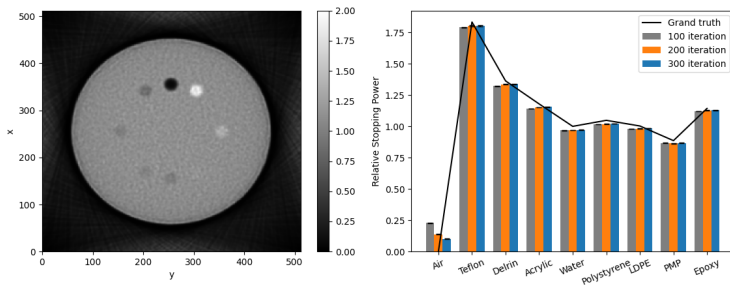
Derenzo Phantom – Spatial Resolution

- Reconstructed RSP distribution and valley-to-peak distribution
- Spatial resolution is the FWHM of the point spread function
- Proton CT literature: 3.1 mm < my algorithm: 4.3 mm



CTP404 Phantom – RSP Accuracy

- Reconstructed RSP distribution and avg. RSP of the inserts
- RSP accuracy: pCT literature: $0.4\% < \text{my algorithm: } 3\%$



Summary

Technique:

- Application of Richardson – Lucy algorithm for pCT

Results:

- Works well, promising results
- Further investigations is required

Bergen pCT Collaboration:

- Develop a pCT detector for clinical testing

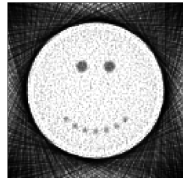
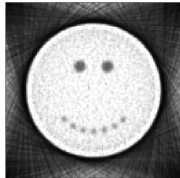
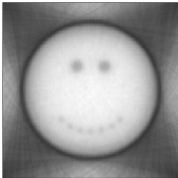
Reached results:

- Working proof of concept detector system
- Measurements for low energy ALPIDE characteristics
- A detailed engineering design \Rightarrow under construction

The Bergen pCT Collaboration

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Thank you for your attention!



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