

Monitoring ion beam therapy with a Compton Camera: simulation studies of the clinical feasibility

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Abstract.

As more and more particle therapy centers are being built world-wide, there is increasing need in treatment monitoring methods, ideally in real-time. The present work investigates the clinical applicability of a Compton camera prototype design by means of Monte Carlo simulations. The Compton camera performance has been studied with the simulation of point-like source and beam irradiation of a PMMA phantom. The system absolute photon detection efficiency, measured via source irradiation, varies in the range $1\text{--}4 \times 10^{-4}$ with energy variations in the prompt gamma energy range and source position shifts with respect to the center of the camera. With proton and carbon beams impinging on a PMMA cylindrical phantom, the number of detected prompt-gamma detected related to various beam time structure has been studied. Finally, the accuracy of the camera in identifying the dose profile fall-off position has been estimated and two different event reconstruction methods have been compared for this purpose; one based on analytical calculation with a line-cone technique, the second relying on an iterative MLEM algorithm. Both methods showed the possibility to reconstruct the beam depth-dose profile and to retrieve the dose fall-off with millimeter precision on a spot basis.

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1. Introduction

Ion beam therapy is a cancer treatment technique which is rapidly gaining importance in the global tumor therapy panorama. In addition to the already operational 70 clinical facilities, with more than 175000 patients already treated by the end of 2017, several new centers have been designed and approved for construction worldwide [Particle Therapy Cooperative Group, 2017]. The favorable feature of this treatment technique is connected to the peculiar energy deposition profile of charged particles as a function of depth in matter. As first observed by Bragg [Bragg and Kleeman, 1904], the depth-dose profile of charged particles shows a maximum close to the end of their range in matter; in addition to this, a strong enhancement of the relative biological effectiveness (RBE) is observed for ions heavier than protons in the region of the Bragg peak [Elsässer et al., 2004, Weyrather et al., 1999], which further enhances the dose ratio between target and healthy tissues.

The maximum of the dose is deposited in the Bragg peak region in the patient and must be tuned to cover the target volume (defined via CT scan) and, at the same time, spare the surrounding healthy tissues. Treatment planning and delivery uncertainties limit the tumour targeting capabilities, like uncertainties in the material composition determination, CT units conversion to ion stopping power, patient mis-positioning, organ motion or morphological changes between treatment fractions. These uncertainties force the clinicians to fix relatively large safety margins around the planned treatment volume, up to $3.5\% + 3 \text{ mm}$ [Paganetti, 2012]. Ion-range verification is one of the conditions required for a broader usage of ion beam therapy and for its further development. With the goal of fully exploiting the ion beam therapy dosimetric potential, the monitoring should be in real-time and ideally in 3 dimensions, in order to be able to interrupt the treatment in case of severe issues, corresponding to important differences between the planned and delivered dose to the target volume or to surrounding organs, in particular in case of proximity to organs-at-risk (OAR).

Several range verification techniques have been considered worldwide for twenty years. Most of them rely on the detection of secondary radiation generated during the slowing down process of incident ions, in particular during nuclear reactions. Among these secondary radiations, positron emitters have been deeply studied in order to exploit positron emission tomography (PET) machines for treatment monitoring. The only available and functional range monitoring system in a clinical center is based on this technique [Enghardt et al., 2004], which is anyway affected by physical and technical limitations [Parodi, 2016].

In addition to positron annihilation products, the relaxation of excited nuclei also produces secondary photons in a wide energy range, between some hundreds of keV till about 8-10 MeV. After the first idea proposal published in 2003 [Stichelbaut and Jongen, 2003], these secondary products of particle treatment have been deeply investigated and the correlation of this gamma radiation to

the ion depth-dose profile has been confirmed by several research groups, starting from [Min et al., 2006] for protons and [Testa et al., 2008] for heavier ions (mainly carbon). The so-called prompt gamma-rays (PG) have the advantage to be emitted almost instantaneously after the beam interaction in the tissue, making them more adapted than PET 511 keV gammas for real-time monitoring. Consequently, different techniques have been proposed to exploit this signal for treatment monitoring purpose, with the related detection systems. Some methods are based on PG timing [Golnik et al., 2014, Krimmer et al., 2017a] or energy [Verburg and Seco, 2014] information and rely on non-collimated systems; more complex detection apparatus can achieve an actual PG imaging, by exploiting mechanical or electronic collimation (i.e. with collimated gamma camera or Compton camera, respectively) for the photon selection (see e.g. [Min et al., 2006, Bom et al., 2012, Smeets et al., 2012, Roellinghoff et al., 2014, Priegnitz et al., 2015, Frandes et al., 2010, Llosá et al., 2012, Kormoll et al., 2011, McCleskey et al., 2015, Matsuoka, 2015, Peterson et al., 2010, Solevi et al., 2016, Aldawood et al., 2017]). For a review on PG monitoring, see [Krimmer et al., 2017b]

Originally designed for astrophysics applications, the potential of Compton cameras for medical imaging has been soon recognized [Todd et al., 1974] and then directly translated to the ion beam therapy monitoring domain. Such a gamma detection system is generally composed of two sections: a scatterer and an absorber. The scatterer is dedicated to the gamma Compton scattering, and should be designed to optimize the Compton scattering probability in the prompt gamma energy range, while reducing the so-called Doppler broadening effect due to electron bounding and motion [Ordóñez et al., 1997]; in most of the cases, this leads to the choice of a light material (low Z), segmented in several subsections. On the other hand, heavier materials may be used to improve photon detection efficiency ([Solevi et al., 2016, Aldawood et al., 2017, Polf et al., 2015]). The absorber finally captures the Compton scattered photons via photoelectric effect and is often composed of segmented high- Z scintillating materials. A slightly different Compton camera configuration can also achieve Compton electron tracking in the scattering detector [Frandes et al., 2010, Yoshihara et al., 2017], which results in additional information for the further reconstruction algorithm. The collected interaction positions and energy depositions in the two detector sections are used to constrain the emission point to the surface of a cone (or to a segment in the cone if the Compton electron track information is retrieved), via Compton kinematics (for an electron initially at rest):

$$\cos \theta = 1 - \frac{m_e c^2 E_1}{E_2(E_1 + E_2)}, \quad (1)$$

where $m_e c^2 = 511$ keV, E_1 and E_2 are the energies, respectively, deposited in the scatterer and the absorber. Analytic or iterative algorithms use these cones to create the image of the prompt gamma emission distribution, with intrinsic 3 dimensional capability [McKisson et al., 1994, Kuchment and Terzioglu, 2016].

Several sources of uncertainty and signal background are connected to the above

described detection method. The reported Compton kinematics formula assumes valid the relation in equation 2:

$$E_0 = E_1 + E_2. \quad (2)$$

Since the initial photon energy (E_0) is not known a priori, a complete photon energy absorption is needed for the cone calculation, or at least three interactions are required in a single event. An underestimation of the total initial energy (caused by a photon non-complete absorption in the absorber section or by the Compton electron escape from the scatterer section), leads to an underestimation of the Compton angle, so to a Compton cone reconstruction uncertainty. If double scattered photons are selected, the initial photon energy can be calculated analytically so that a complete absorption is not mandatory. In addition to this, the Compton formula considers the Compton scattering electron initially at rest, and its energy configuration creates a blur in the Compton angle reconstruction, resulting in the already cited Doppler broadening effect [Ordonez et al., 1997]. Furthermore, the detection principle is based on time coincidences between the two detector sections, therefore the time structure of the incoming particles plays an important role. The final image accuracy suffers from false coincidences generated by two prompt gammas interacting within the same time window or by contamination of secondaries, mainly neutrons and protons. The effect of random coincidences can be reduced by high detector time resolution or background rejection methods [Draeger et al., 2017]. Energy selections can be applied to the collected coincidences [Polf et al., 2009, Hilaire et al., 2016] or the homogeneous neutron background can be reduced via time-of-flight information [Testa et al., 2010].

Ortega and colleagues [Ortega et al., 2015] presented a detailed analysis of the noise sources for Compton imaging in proton therapy monitoring, and the clinical application of this method for detecting range shifts is tested. The simulation study showed the relative expected rate of prompt gammas and neutrons, and the resulting rate of random coincidences ranging from 19 % to more than 60 % depending on the beam energy and the coincidence time window. This amount of fake events leads to complex reconstruction scenarios, where the identification of a 3 mm range shift is not clear for all cases.

Starting from these results, we propose in this paper to study with Monte Carlo simulations a Compton camera prototype based on semiconductor and scintillator detectors [Krimmer et al., 2015] developed by the ClaRyS collaboration (four French research institutions: IPNL and CREATIS in Lyon, LPSC in Grenoble, and CPPM in Marseille). The camera performance is studied with respect to the gamma energy in the prompt gamma energy range. Furthermore, the feasibility of its clinical application as depth-dose profile monitor during ion beam therapy clinical treatment is analyzed. After a preliminary study with point-like gamma sources irradiation focused on detector efficiency measurements as a function of the source position and gamma energy, clinical proton and carbon beams impinging on an homogeneous PMMA phantom are simulated to reproduce treatment conditions and analyze the prompt gamma detection resulting

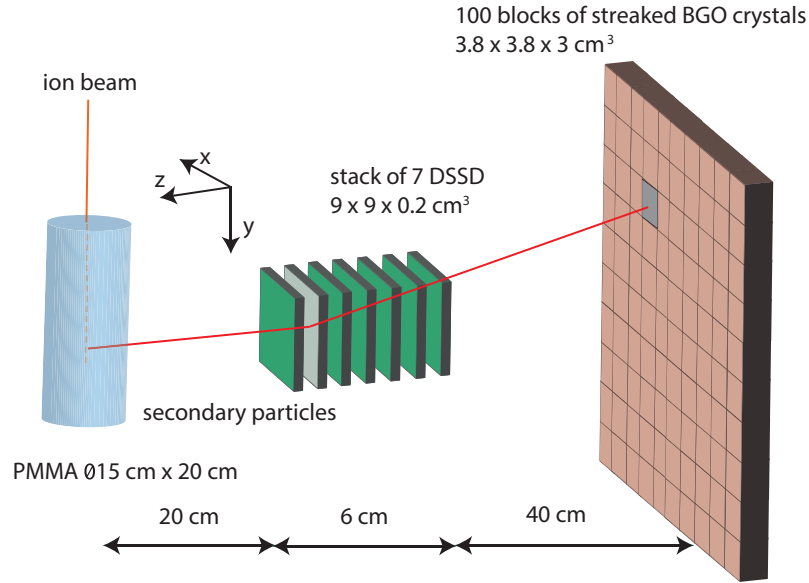


Figure 1: Scheme of the simulation setup (not at scale): a PMMA cylindrical phantom is set in front of the Compton camera prototype. The Compton camera is composed of a stack of 7 double sided silicon strip detectors (scatterer) and a plan of 100 single BGO blocks. The set distances are realistic for clinical conditions. This geometrical configuration has been used for all the simulations presented in this work.

scenario. The ratio between true and random coincidences is studied as a function of the beam intensity. Two kinds of reconstruction algorithms, a line-cone analytic method and a MLEM iterative one, are applied to the collected data in order to compare the imaging results. Finally, the precision with which the dose profile fall-off can be detected with the Compton camera is reported.

2. Material and methods

2.1. Simulation setup

The monitoring system modeled in this simulation work is a Compton camera prototype under development within the French collaboration CLaRyS. The detectors detailed characteristics can be found in [Krimmer et al., 2015]. A scheme of the simulation setup is given in figure 1. The scatterer consists of seven parallel planes of silicon detectors (double-sided silicon strip detectors, DSSDs), $9 \times 9 \times 0.2 \text{ cm}^3$, with 1 cm distance between the centers of two neighboring planes, while the absorber is composed of an array of 10×10 BGO (Bismuth Germanate - $\text{Bi}_{12}\text{GeO}_{20}$) blocks ($3.5 \times 3.5 \times 3.0 \text{ cm}^3$ each) placed behind the silicon layers at a distance which can be tuned according to the requirements.

The silicon detectors have a strip pitch of 1.4 mm, for a total of 64 strips per side (double-sided readout based on electron and hole pairs collection). Regarding

the BGO blocks, their entrance surface is streaked in a 8×8 matrix of pseudo-pixels, $4.4 \times 4.4 \text{ mm}^2$ size, and the readout is performed via 4 photo-multiplier tubes. The position reconstruction is achieved via Anger logic.

In the simulation setup, the ion beam interacts with a cylindrical PMMA (PolyMethylMethAcrylate) phantom (15 cm diameter and 20 cm length) placed in front of the Compton camera as target. It is placed 20 cm far from the first silicon plane (center-to-center distance) in order to exit to a patient in the treatment room. The distance between the last silicon layer and the absorber array (center-to-center) is set to 40 cm.

The silicon detector strips are not reproduced in the simulation code, and the transverse spatial resolution is set to 0.9 mm FWHM at the reference energy of 1 MeV, according to preliminary measurements performed on smaller detector prototypes. Concerning the transverse direction (perpendicular to the beam line), the interaction position is set to the center of the involved silicon plane. A mono-block crystal is simulated for the absorber for simplicity. The events are selected to be limited to a single block component based on the interaction localization, and the interaction position is reconstructed via center of gravity calculation if multiple interactions occur. An uncertainty contribution, randomly extracted by a Gaussian of 5 mm FWHM, is added to the reconstructed position to fit with the geometrical features. For what concerns the parallel direction, given the fact that the employed BGO blocks have not depth of interaction reconstruction capabilities, the interaction position is fixed to the center of the mono-block crystal.

The energy resolution of the BGO blocks was estimated in preliminary measurements and is accordingly set to 17% FWHM at the reference energy of 667 keV (a $^{137}\text{cesium}$ source has been used for the measurements). The energy resolution of the silicon detector is set to 2.3 keV (RMS) according to the design expectations; no characterization data are yet available for an instrumental estimate.

The time resolution has been set to 3.0 ns FWHM for the BGO blocks and to 15.0 ns FWHM for the silicon slabs, according to preliminary measurements performed on test detector modules at the GANIL center in France.

The detector resolutions play an important role in the Compton camera performances. The spatial resolution of the absorber influences the position of the apex of the Compton cone as well as its axis orientation. The energy resolution of the scatterer determines the Compton cone aperture angle. The time resolution impacts the coincidence window between the absorber and the scatterer, and therefore the detectors ability to distinguish between true and random coincidences.

The CLaRyS project also includes the development of a beam tagging hodoscope, composed of scintillating fibers read out by multi-channel photomultipliers. This detector is used to synchronize the beam time and space structure to the prompt gamma detection in order to tune the detection window reducing the background contamination. The hodoscope is not included in the simulation, but its time resolution has to be taken into account for the time-of-flight discrimination (see section 2.3). It is set to 1 ns

Table 1: Estimations of reachable resolutions with the detectors. Those resolutions are applied during the simulations.

Resolution (FWHM) at 1 MeV	Scatterer	Absorber	Hodoscope
spatial [mm]	0.9	5	1
energy	5.4 keV	17 %	/
timing [ns]	15	3	1

Table 2: Hadronic models used in the Geant4 simulations.

Process	Protons	Ions	Neutrons
Electromagnetic			
Inelastic	G4BinaryCascade	G4QMDReaction (G4IonsShenCrossSection)	standard _{option3} G4BinaryCascade + G4NeutronHPInelastic (<19 MeV)
Elastic	G4LElastic	G4LElastic	G4LElastic + G4NeutronHPElastic (<19 MeV)
Fission	/	/	G4LFission + G4NeutronHPFission(<19 MeV)
Capture	/	/	G4LCapture + G4NeutronHPCapture (<19 MeV)
Radioactivedecay	/	G4Radioactivedecay	/

FWHM. The detector’s spatial, energy, and time resolutions are summarized in table 1.

The Monte Carlo simulation is performed with the Geant4 toolkit, version 9.6 patch 02. The particle interactions in matter are described in this work by means of the models listed in table 2. Additionally, the Doppler broadening and the photon polarization effects are taken into account.

The two main beam particles used in clinics are considered: protons and carbon ions. The beam range of interest is 15.2 cm in the PMMA phantom, and the associated energy is 160 MeV for protons and 305 MeV/u for carbon ions.

The beam transverse dimension is modeled with a Gaussian distribution with a standard deviation of 5 mm for protons at 160 MeV and 3.5 mm for carbon ions at 305 MeV/u. The number of incident ions for a spot in pencil beam scanning (PBS) mode is 10^8 for protons (distal spot) and 10^5 for carbon ions. The beam time structure is applied at the data analysis stage and described in section 2.2.

2.2. Beam time structure

Two different beam time structures have been considered for this study, related to two kinds of accelerators used in clinical practice: the IBA C230 cyclotron for protons (used in 16 clinical centers worldwide) and the synchrotron installed in the Heidelberg Ion Therapy Center (HIT), Germany, for carbon ions. For protons at 160 MeV, the primary particles are grouped in bunches of 2 ns (this value may vary also according to the distance between the cyclotron and the treatment room, and energy spread selection) at a frequency of 106 MHz (9.42 ns) [Roellinghoff et al., 2014]. The clinical beam intensity is 3.2 nA which corresponds to about 200 protons per bunch. Concerning the carbon ion beam at 305 MeV/u, the estimated microstructure during the extraction phase is

composed of 30 ns duration bunches at a frequency of 5.9 MHz (170 ns period). The clinical beam intensity for carbon ions is 5×10^7 ions/s during extraction, corresponding to about 9 ions per bunch. This beam structure is extrapolated from measurements performed by our group in 2013 at HIT; the beam time structure was measured for 200 MeV/u and 400 MeV/u primary ion energy with a two-fiber hodoscope (basic prototype of the one at present under development) and the spill signal was given by the accelerator. Figure 2 shows the beam structure for carbon ions at 400 MeV/u. The pulses have a spill period of 150.2 ns and each bunch is approximately 21.5 ns. The mentioned measurements have shown that the spill phase changes during the extraction: this implies that the HF signal from the synchrotron can not be used to trigger the pulses, so that the use of an additional beam time stamp system like the hodoscope seems required for time-of-flight background rejection purposes.

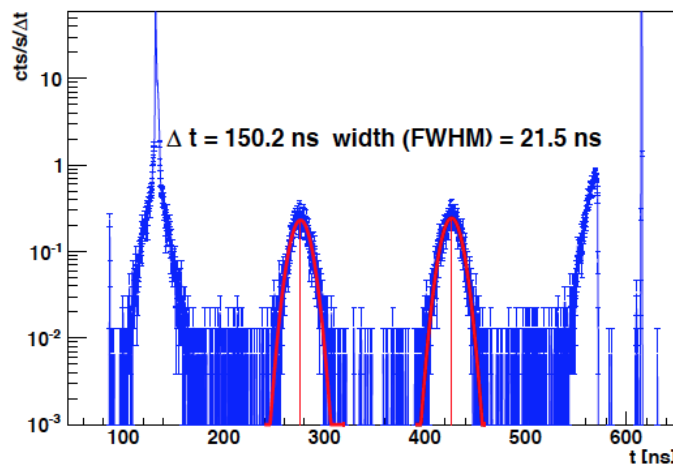


Figure 2: Time micro-structure measured from a carbon ion beam at 400 MeV/u delivered at HIT. The pulses have an extraction period of 150.2 ns and the bunches are 21.5 ns FWHM. The measurement was achieved with a two-scintillating-fiber hodoscope.

The coincidence window (between scatterer and absorber events) is set to 40 ns, centered on each absorber detected interaction. This value is adapted to the detectors time resolutions. Table 3 summarizes the presented beam time structures and coincidence reconstruction features.

2.3. TOF and energy based data selection

The Compton detection principle is based on a double interaction in the scatterer and absorber section, where an interaction is defined as an energy deposit in a detector module. As discussed in section 2.2, the coincidence reconstruction relies on a defined time window, fixed according to the detector resolution. In a simulation environment, different kinds of coincidence events can be distinguished and studied:

- real coincidences: created by a single photon first interacting in a single scatterer plane and then in a single absorber block;

Table 3: Description of the two beam structures studied: the IBA cyclotron C230 for protons and the synchrotron installed at the Heidelberg Ion Therapy Center (HIT) in Germany for carbon ions. The macro-structure of the synchrotron, at the second time scale, is not considered here. The beam structures are applied to the simulation data.

		Protons	Carbon ions
Clinical features	Facility	IBA Cyclotron C230	Synchrotron at HIT
	Clinical intensity	2×10^{10} p/s	5×10^7 ions/s
	Energy	160 MeV	305 MeV/u
Beam structure	Bunch time [ns]	3.2	30
	Period [ns]	9.4	170
	Primaries/bunch	217	9
Detectors	Coincidence window [ns]	40	40
	Time resolution (FWHM) [ns]	Si: 15 and BGO: 3	

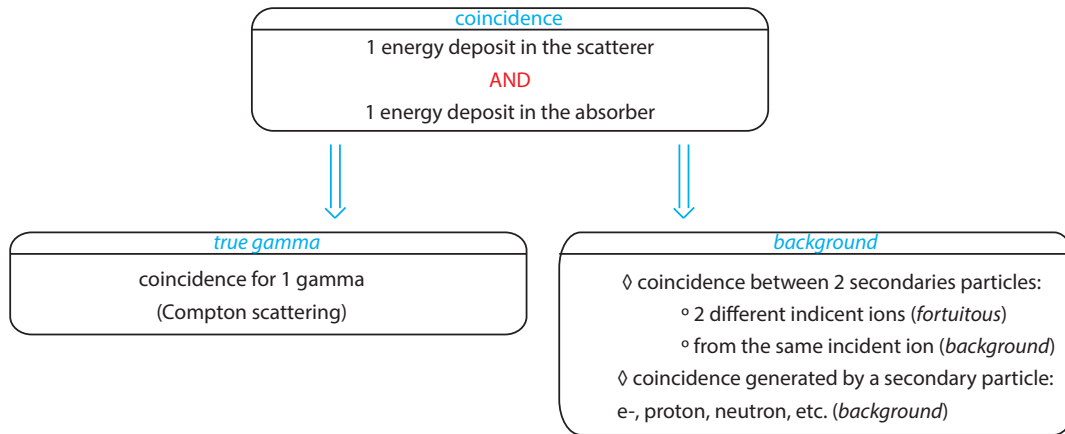


Figure 3: Diagram showing the different definitions of coincidences in the Compton camera.

- quasi-simultaneous interaction of two secondary particles;
- double interaction of the same particle, not a photon (protons, neutrons).

In an experiment the collected data are affected by a certain number of the so-called random coincidences, which cannot be experimentally distinguished from true coincidences, from the timing coincidence point of view. The amount of random coincidences depends on the detector time resolutions, on the fixed time coincidence window and on the beam time structure, on the phantom composition and camera prototype setup. In figure 3, a schematic view of the different kinds of coincidences is presented.

In addition to this, the prompt gamma measurement is contaminated by other secondary particles (mainly massive and charged particles like protons and neutrons), produced by the interaction of the primary particles with the patient/phantom. Ad-hoc filtering methods are applied to reduce the above described contamination.

- Time-Of-Flight: it has been demonstrated by our group that a time-of-flight

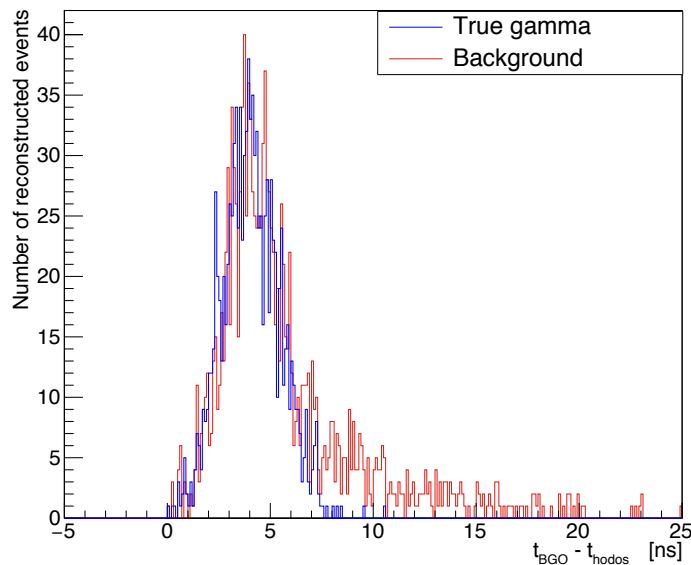


Figure 4: Time of flight spectra of true gamma coincidences (blue) and background events (red) obtained with 10^8 160 MeV incident protons. The detection time for the absorber is given by the first deposit energy.

discrimination is possible and effective in reducing the background generated by massive particles interactions [Testa et al., 2010]. The massive particles approach the detector at a lower speed with respect to photons. The time information provided by the hodoscope and the absorber can be combined to fix a detection time window and reject all the events outside the window. The time elapsed between the incident particle creation and the secondary particle detection in the absorber is considered as the time-of-flight. The hodoscope time resolution $u_{\text{hodoscope}}$ (1 ns FWHM) is applied to the primary particle creation time, with a contribution randomly extracted from a Gaussian with $\sigma = 1/2.35$ ns. The detection time in the absorber is affected by the absorber time resolution.

$$TOF = t_{\text{absorber}} - t_{\text{hodoscope}} \quad (3)$$

$$TOF_{\text{simulation}} = t_{\text{absorber}} - (t_{\text{creation}} + u_{\text{hodoscope}}) \quad (4)$$

The time-of-flight spectrum resulting from the simulation shows that the coincidences of interest (produced by prompt-gamma rays) are included in a window between 0 and 8 ns (figure 4). Therefore, all the coincidences with a TOF higher than 8 ns have been rejected.

- **Energy selection:** energy thresholds are defined for the event detection. 50 keV and 100 keV are set as lower threshold for the energy deposited in a single silicon layer and absorber, respectively. For a complete event, a total absorbed energy lower limit is set to 1 MeV. In addition to the effect of background rejection, this selection also reduces the impact of partially absorbed photons and events with

Compton electron escape.

Further energy selections, assuming for instance $E_1 + E_2$ equal to one of the strong gamma lines, have been applied by other authors. Also, filters checking the possibility of reconstructing a Compton cone could be used. At this stage we did not consider such approaches, in order to cope with simple considerations on signal to noise on raw data.

2.4. Reconstruction algorithms

Once the coincidences are defined and selected according to the fixed physical cuts, the prompt-gamma emission point has to be reconstructed for each event. This can be done via analytic or iterative algorithms based on the Compton kinematics. Both are presented in the following sections.

2.4.1. Line-cone algorithm The reconstruction via line-cone algorithm exploits the energy deposit and position information collected by the camera in addition to the beam planned path. Thanks to the deposited energies in the detectors and the interaction positions, a cone surface is analytically defined via the Compton equation 1. Figure 5 shows a sketch of the reconstruction principle. The interaction position in the scatterer gives the cone apex and the line connecting the interaction positions in scatterer and absorber gives the cone axis. We assume that the initial energy of the gamma ray is fully absorbed in the absorber. In order to constrain the reconstruction, the beam direction is used to limit the possible solutions (lying on the reconstructed cone surface) to two points (intersection of the beam direction and the reconstructed cone). The set of all the reconstructed points gives the emission source distribution. The final image is the mono-dimensional projection of the prompt gamma emission spectrum.

2.4.2. LM-MLEM algorithm The iterative methods allow to get a 3D image reconstruction, potentially by taking into account the spatial resolution and the energy resolution of the detectors. Few iterative algorithms have been developed for Compton event reconstruction [Schone et al., 2010, Zoglauer et al., 2011, Gillam et al., 2011, Mackin et al., 2012, Lojacono et al., 2013].

The *List-Mode Maximum Likelihood Expectation Maximization* (LM-MLEM) algorithm is an MLEM version which allows to reconstruct the image directly from the list of detected events. The LM-MLEM algorithm used for this study is detailed in [Maxim, 2014, Hilaire et al., 2014].

The first step is to define the volume which includes the origin of the prompt gamma ray detected. This volume is divided into equal voxels and the emission intensity is assumed homogeneous for each voxel j , with a Poisson distribution of parameter λ_j (a vector of the emissions intensities of all the voxels). The algorithm is based on a system matrix T composed of the coefficients t_{ij} which represent the probability that a photon produced in the voxel j is detected in coincidence by the Compton camera as an event

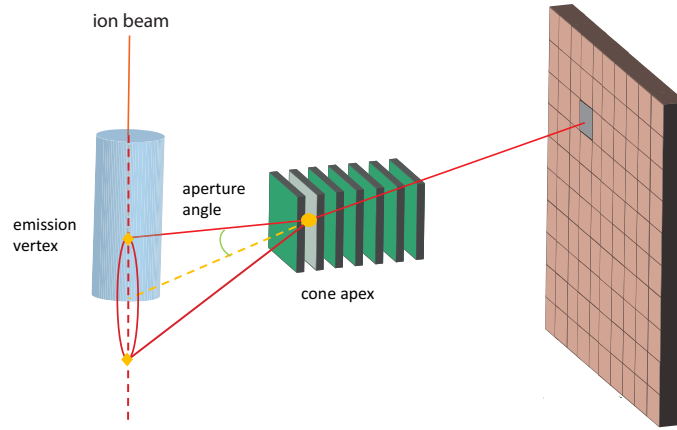


Figure 5: Scheme of the reconstruction principle for Compton events. For line-cone reconstruction methods, two points are extracted for each event (diamonds in the figure), provided by the intersection between the reconstructed Compton cone and the beam line. The reconstructed image is strictly 1D. For the iterative MLEM algorithm, each algorithm iteration adds constraints to the reconstructed cone surfaces, leading to a 3D image.

i. The probability for a gamma detected in coincidence to be emitted from the voxel j is denoted as s_j . The LM-MLEM algorithm starts with an initial value $\lambda_j^{(0)}$, which can be the simple back-projection reconstruction. The iterations rely on the following recurrence relation:

$$\lambda_j^{(l+1)} = \frac{\lambda_j^{(l)}}{s_j} \sum_{i=1}^{N_\gamma} t_{ij} \frac{1}{P_i^{(l)}}, \quad \text{with} \quad P_i^{(l)} = \sum_{k=1}^{N_v} t_{ik} \lambda_k^{(l)}, \quad (5)$$

where N_γ is the number of detected events and N_v is the number of voxels in the image.

For each photon detected, the coefficients in column i are calculated by taking into account the uncertainties on the angle between the source and the involved scatterer plane and the angle between the scatterer plane and the absorber involved module. The matrix elements t_{ij} are calculated as:

$$t_{ij} = K(\beta_i, E_{tot}) \frac{|\cos(\theta_{V_2 V_1})|}{V_2 V_1^2} \int_{M \in v_j} \frac{|\cos(\theta_{V_1 M})|}{V_1 M^2} h_i(M) dv, \quad (6)$$

where β_i is the Compton scattering angle, V_1 the interaction position in the scatterer, V_2 the interaction position in the absorber, h_i the spatial kernel which models the uncertainties on the Compton angle for each voxel M , $K(\beta_i, E_{tot})$ the differential cross section and v_j the reconstructed volume.

In order to simplify and speed up the calculation of the t_{ij} matrix, the voxels located far from the reconstructed cone are set to 0. The distance between the cone and the

voxel is calculated by taking the voxel center as reference point. For each iteration, the matrix T is stored and the reconstructed image can be produced.

2.5. Performance study

In this section the parameters studied for the camera performance evaluation are explained. The study is mainly divided into three sections:

- Absolute efficiency: the camera absolute efficiency for gamma detection is studied by means of mono-energetic irradiation with point-like sources;
- Rate of random coincidences: with proton and carbon beams interacting with the PMMA phantom, the rate of random and true detected coincidences is studied as a function of the beam intensity;
- Camera precision: the camera capability of identifying the fall-off of the prompt-gamma emission profile is tested and the two reconstruction methods presented above are compared.

2.5.1. Absolute detection efficiency The absolute efficiency is crucial for the Compton camera performances and for its possible application in treatment monitoring. An efficient monitoring system should be ideally in real time, in order to allow for a treatment adaptation or interruption in case of severe issues detected in the delivered dose profile with respect to the planned treatment. In order to achieve an online detection of such deviations, given the reduced prompt gamma emission rate per incident ion [Ortega et al., 2015], a high detection efficiency is required to perform a monitoring on, ideally, a beam spot basis. In addition to this, the absolute detection efficiency directly affects the image reconstruction quality, which is in general increased for increased statistics.

In order to well define the expected camera efficiency, it has been studied with the irradiation from point-like mono-energetic gamma sources, set in different positions with respect to the center of the camera on the transverse plane.

The absolute efficiency ϵ is defined as:

$$\epsilon = \frac{N_{\gamma_{\text{recons}}}}{N_{\gamma_{\text{total}}}}, \quad (7)$$

with $N_{\gamma_{\text{recons}}}$ the number of gamma events corresponding to true coincidences, $N_{\gamma_{\text{total}}}$ the total number of gamma emitted.

The setup is the same as figure 1, with the exception of the PMMA phantom which is removed to leave the gamma source in air. The point-like source is set in the range -300 mm to $+300$ mm (with the center of the camera transverse section set in the position 0) with a step of 20 mm close to the center of the camera, then increasing up to 1 cm for the most peripheral point. The movement followed the transverse axis of the camera. Different energies have been tested to mimic different prompt gamma spectroscopic lines: 300 keV, 500 keV, 1 MeV, 2 MeV, 4 MeV, 6 MeV. No time structure is reproduced for this part for the study.

2.5.2. Rate of random coincidences As the Compton detection principle relies on time coincidences, in addition to the main importance played by the detectors energy resolutions, the beam intensity and time structure are important parameters to be studied in order to assess the possible clinical implementation of a Compton detection based monitoring of ion beam treatment. The ability of the detection system to distinguish between true and random coincidences, i.e. the resulting signal over noise ratio, strongly depends on the beam time structure. The number of true and random coincidences (where the two cases listed in section 2.3 are included in the random coincidences) is studied as a function of the beam intensity, before and after data reconstruction via line-cone algorithm (see section 2.4).

In the simulation, the beam intensity is modeled by an average number of particles per bunch. The exact number of particles in each bunch is given by a random extraction from a Poisson distribution, where the mean value is the selected beam intensity. The range of intensities is defined in order to cover a wide range of operation: from a very low beam intensity to a realistic clinical particle rate. Therefore, for proton and carbon ions, the lowest beam intensity is set to 0.1 particles per bunch on average, while the upper limit is set to 217 protons or 70 carbon ions per bunch. All the simulations are performed with a total of 10^8 primary protons and 2×10^5 primary carbon ions. Those value correspond to realistic clinical values used for the treatment of a single spot [Grevillot et al., 2011]. For the analysis of the results, the coincidence yields are scaled to the number of incident ions and the beam intensity to the average number of ions per bunch.

2.5.3. Camera precision The camera precision is defined as the difference between the predicted PG fall-off (according to the treatment planning) and the detected one. In this study, a reference profile has been defined as the reconstructed emission vertex profile at high statistics (10^{10} incident protons). This Bragg peak position is used as reference and compared to the ones at lower statistics, deduced as described in the following.

For the LM-MLEM reconstruction, the reconstruction volume is set to $20 \times 40 \times 1 \text{ cm}^3$ around the Bragg peak expected position, with a $100 \times 200 \times 5$ voxel matrix. To be noticed that the two reconstruction methods produce different results: the line-cone method is based on the beam direction information, so that it naturally returns a mono-dimensional image as result, while the LM-MLEM method is able to reconstruct the prompt-gamma emission distribution in three dimensions. A mono-dimensional projection along the beam direction is used for the falloff identification and for a direct comparison of the line-cone method. The *SmoothKern* method, with the Nadaraya-Watson regression [Nadaraya, 1964, Watson, 1964], is used to smooth the reference profiles in order to reduce relative statistical fluctuations.

A region of interest (ROI) ranging from $y = 0 \text{ mm}$ to $y = +100 \text{ mm}$ is defined around the expected Bragg peak position, located at $y = +50 \text{ mm}$ in the phantom. The reference reconstructed profiles are modeled in the ROI by a linear combination of Non-Uniform Rational Basis Splines (NURBS) [Rogers, 2001].

The reference data set is normalized in order to obtain lower statistics profiles, in the range 1×10^8 to 5×10^9 . Statistical fluctuations, randomly extracted following a Poisson law, are added to the obtained subsets to mimic realistic reconstructed profiles. This method is used to reduce the analysis time.

For each chosen statistics, 1000 data subsets are generated and a custom minimization method is applied to deduce the minimal shift between the reference profile and the one obtained from the data subset. The minimization algorithm shifts the reference profile in a 60 mm range, between -30 mm and $+30$ mm with respect to the initial position, with a step of 0.1 mm, and for each position it compares its position to the low statistics profile by calculating the χ^2 as follows:

$$\chi^2 = \sum_{i=1}^{N_{bin}} (y_{\text{sample},i} - y_{\text{NURBS},i})^2, \quad (8)$$

where for every bin i y_{sample} is the number of events in the low statistics profile, y_{NURBS} is the number of events for the reference profile NURBS (scaled at the same low statistic). Several minimization algorithms are available, but in order to avoid artifacts at low statistic, a robust custom defined approach has been preferred.

The global minimum of all the calculated χ^2 is then retrieved and the shift associated to this value is added to the shift distributions. The standard deviation of the distribution resulting of the thousand results gives the precision of the camera for a given number of incident protons.

3. Results

The possible implementation of the ClaRyS Compton camera as a monitoring system for ion beam therapy has been tested. The detection efficiency of the camera has been measured with the irradiation with point-like gamma sources set in different positions with respect to the center of the camera. A PMMA cylindrical phantom has been simulated and exposed to proton and carbon beams at increasing intensities for an analysis of the prompt gamma detection environment (background, random coincidence contamination). The camera precision in the identification of the fall-off of the prompt gamma emission profile has been investigated. For this purpose, the comparison of two different reconstruction methods (line-cone analytical reconstruction and MLEM iterative algorithm) is presented. In the following sections we show the obtained results.

3.1. Absolute detection efficiency

Figure 6 shows the absolute gamma detection efficiency as a function of the gamma source position with respect to the center of the camera in the transverse plane. On the left side, we show the results achieved with an ideal detector, while on the right side realistic energy cuts are applied on each detector section: the lower energy limit has been set to 50 keV for the scatterer and 100 keV for the absorber.

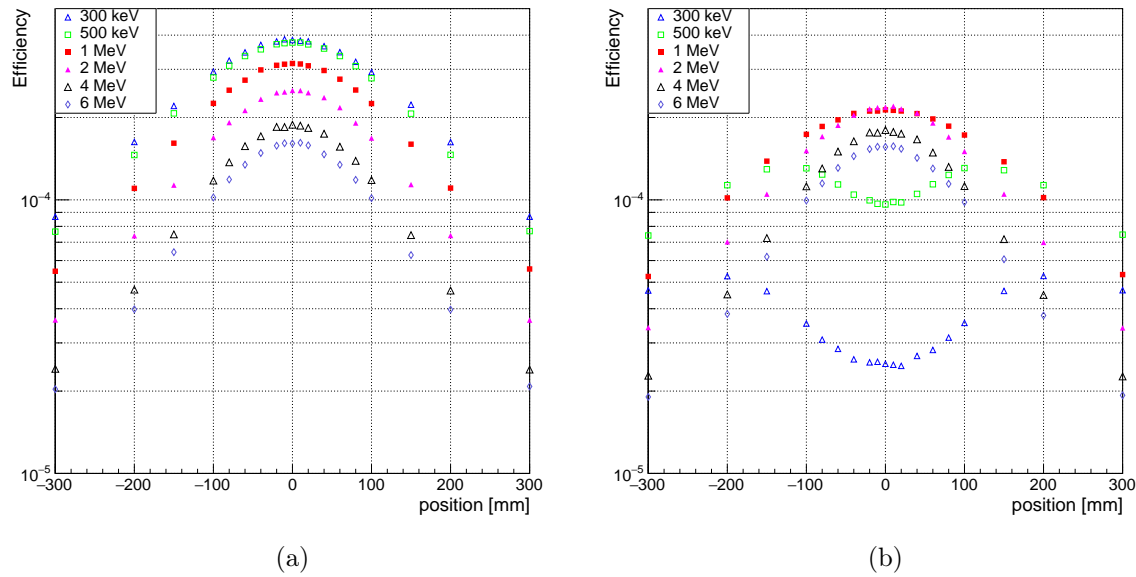


Figure 6: Absolute Compton camera efficiency as a function of the gamma source position for different gamma energies, in the range between 300 keV to 6 MeV. The left side shows the camera efficiency with no data selection. In the right side, detection energy thresholds are applied: the lower energy thresholds are set to 50 keV for the scatterer and 100 keV for the absorber, to reproduce a realistic scenario. These values can change for the final configuration, according to the detector energy resolutions achieved.

As expected according to the interaction probability energy dependency, the efficiency is higher for low gamma energies, and it lies in the range 4×10^{-4} at 300 keV and 1.5×10^{-4} at 6 MeV (figure 6). Moreover, it can be noticed how the efficiency quickly drops as the point source is shifted away from the camera center: this effect is more important for high energies, for which the incident gamma is less deflected in the scatterer for the same energy deposited compared to a low energy gamma (see equation 1).

Figure 6(b) shows the effect of realistic camera detection thresholds as opposed to ideal detection. The gamma detection efficiency drops of a factor ranging from about 1.25 to more than an order of magnitude for the central detection area for energies in the range 300 keV - 2 MeV respectively. The effect is reduced by the distance of the source from the center of the camera. Negligible effects are detected for positions with a distance greater than 200 mm from the center of the camera, and for any distance at energies above 2 MeV, while the efficiency is reduced for low energy gammas in the central area of the camera.

3.2. Rate of random coincidences

In figure 7, the different components of the signal resulting from the PMMA exposure to proton and carbon beams are shown as a function of the beam intensity. The true coincidences represent scatterer-absorber time coincidences generated by the same gamma ray. All the other coincidence types compose the background. The collected data sets are reported with and without the applied time-of-flight discrimination, mainly employed for neutron rejection, as mentioned in section 2.3.

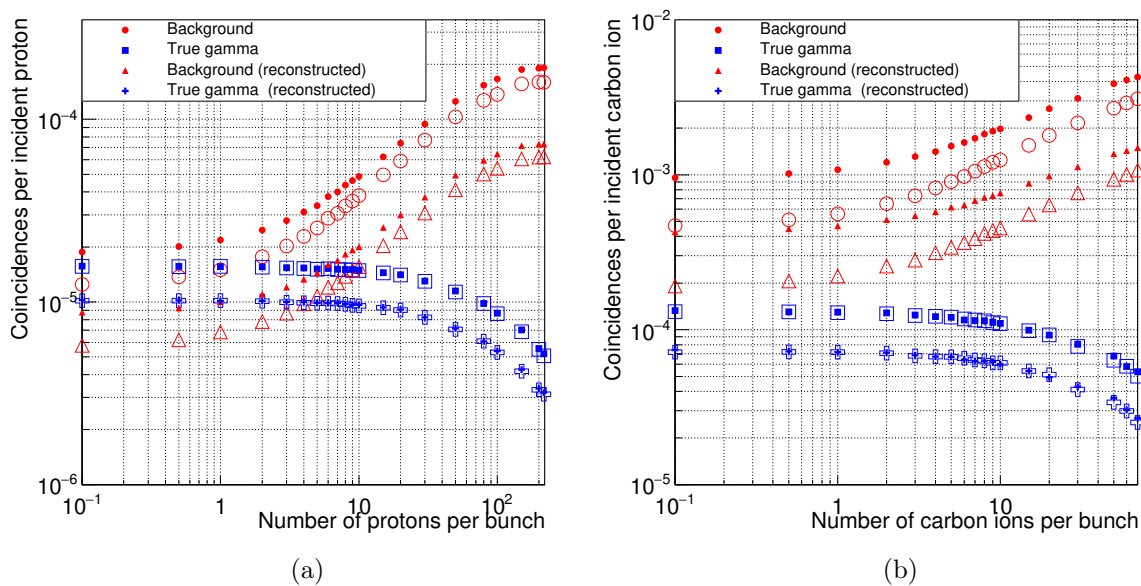


Figure 7: Coincidences yield for protons (left) and carbon ions (right) as a function of the beam intensity. The intensity is reported as number of incident particles per bunch. The filled markers correspond to the collected data without time-of-flight discrimination, while this cut is applied to the data reported with empty markers. Moreover, the yields are given before and after the profile reconstruction with the line-cone algorithm.

In figure 7(a) and (b) the amount of true gamma coincidences and background events are reported before and after reconstruction via line-cone algorithm as a function of the beam intensity for proton (a) and carbon ion (b) beams. In addition to this, for each curve realized with the complete collected data set, the related one realized after time-of-flight selection of events is sketched. All the curves have been normalized to the number of incident ions.

The amount of background events (mainly random coincidences) increases with the increasing beam intensity: a factor of about 30 with respect to true gamma events is obtained for proton beams at the intensity of 200 protons per bunch with no event selection, while a factor more than two times higher is reported for carbon ions in the same conditions. The time-of-flight selection can slightly improve the signal-over-noise

ratio by reducing the amount of background events. The amount of true gamma events and background events becomes similar at the intensity of about 1 proton per bunch.

3.3. Camera precision

Given the results presented in the previous section, the camera precision in the fall-off identification is investigated with proton beams at reduced intensities. A single data set has been collected, corresponding to the irradiation of the PMMA phantom with a proton 160 MeV monoenergetic beam, with a reduced intensity of 1 proton per bunch. A total of 10^{10} protons has been simulated to define the reference Bragg peak, and then different sub-data sets have been extracted for the precision estimate at different statistics, as explained in section 2.5.3.

The high statistics profile reconstructed via line-cone algorithm is shown in figure 8a and via the LM-MLEM reconstruction method in figure 8b. A NURBS model of a low statistics subset sample (10^8 incident protons) is shown in figures 8c for the line cone and 8d for the MLEM. Figures 9a and 9b show the distribution of χ^2 calculated for a low statistic profile at 10^8 incident protons. The retrieved optimal shift distribution is shown in figure 9c and 9d for the line-cone and LM-MLEM algorithm respectively, for 10^8 incident protons as well. Figure 10 shows the results of the reconstruction of the data set obtained for the selection of 10^8 primary protons.

The analysis method described in section 2.5.3 is applied to the different data set to retrieve the camera precision in the dose profile falloff identification. The results are shown in figure 11, where the two reconstruction methods are represented by different markers.

The iterative MLEM reconstruction method enables one to achieve a better precision, with a reduction of about 3 mm of the falloff retrieval accuracy in the whole range of statistics explored. A linear behavior, highlighted by the performed linear fit of the two data sets, is verified with increasing number of primary protons, starting from the single spot scale of about 10^8 primaries, till 5×10^9 protons, which can correspond to the monitoring of a group of spot with the same planned range.

4. Discussion

We studied in this simulation work the performance of the CLaRyS Compton camera prototype and its possible implementation as prompt gamma detector for ion beam therapy monitoring. The proposed analysis is focused on three main points: absolute gamma detection efficiency, true and random coincidence rate and camera precision in the identification of the prompt gamma emission profile fall-off.

The absolute gamma detection efficiency has been measured with the detector exposed to gamma sources at six different energies, and the efficiency variation as a function of the source position has been reported. The whole study has been performed with a fixed camera geometry, but different geometrical configuration can

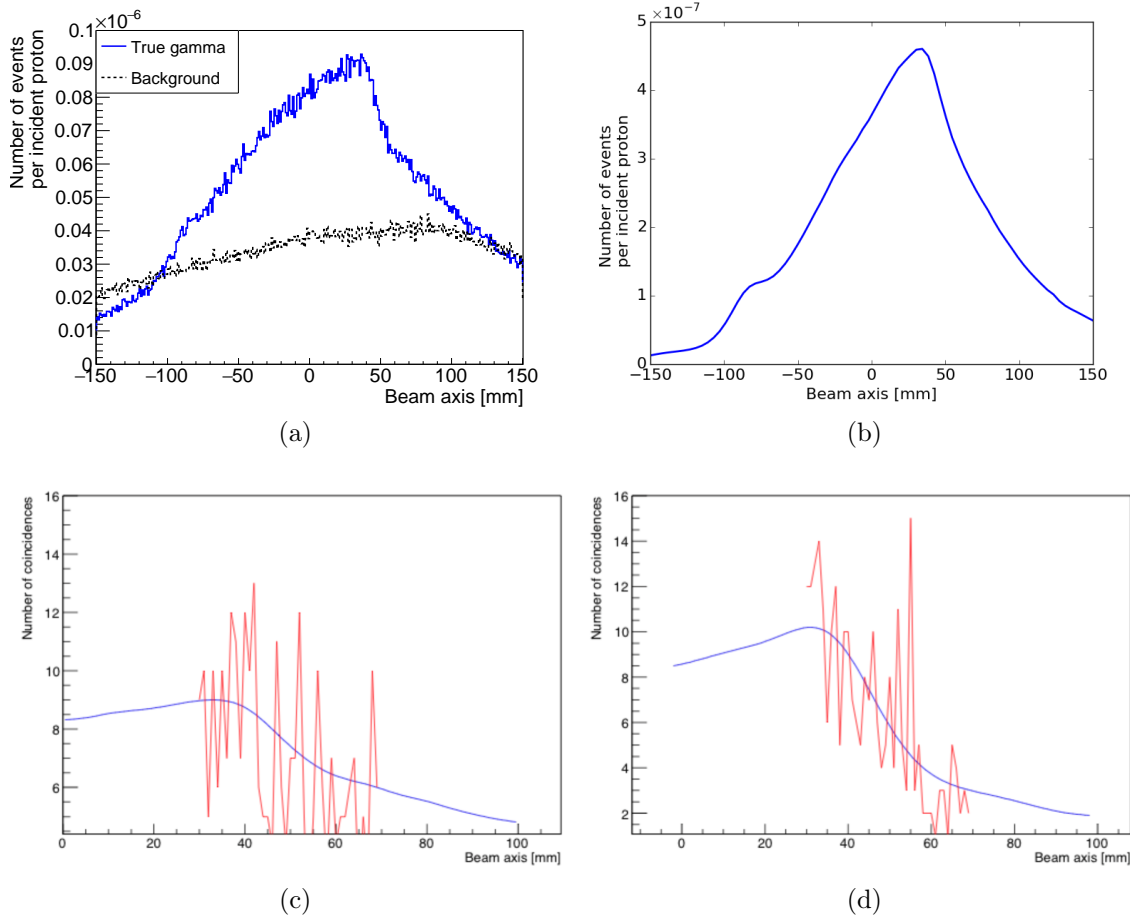


Figure 8: Data processing comparison for the same proton simulation with the line cone algorithm (left column) and the LM-MLEM algorithm (right column). The first row gives the reconstructed profile for 10^{10} incident protons. The second row shows the reference curve (blue) and the curve obtained with a 10^8 incident protons subset.

change the obtained results. In [Fontana et al., 2017] we tested the same prototype with two absorber configurations, showing an absolute efficiency reduction of a factor approximately two when reducing the absorber from a 8×6 block matrix to a 3×3 one. To this efficiency reduction corresponds an increased spatial resolution, probably given by the selection of gamma scattered with small angles. This hypothesis is confirmed by the efficiency variation generated by the applied energy thresholds. Indeed, for low energies, below 2 MeV, the increased efficiency in the central section of the camera active surface shown in figure 6(a) is linked to the increased relative number of photons approaching the camera with small angles. These photons are more likely undergoing Compton scattering with a reduced energy deposition, which is recorded by an ideal detector and rejected by the energy threshold. The effect is all the more important as the primary gamma energy is limited, creating the peculiar energy dependence of the efficiency reduction observed in the results in figure 6(b). Therefore, in case the

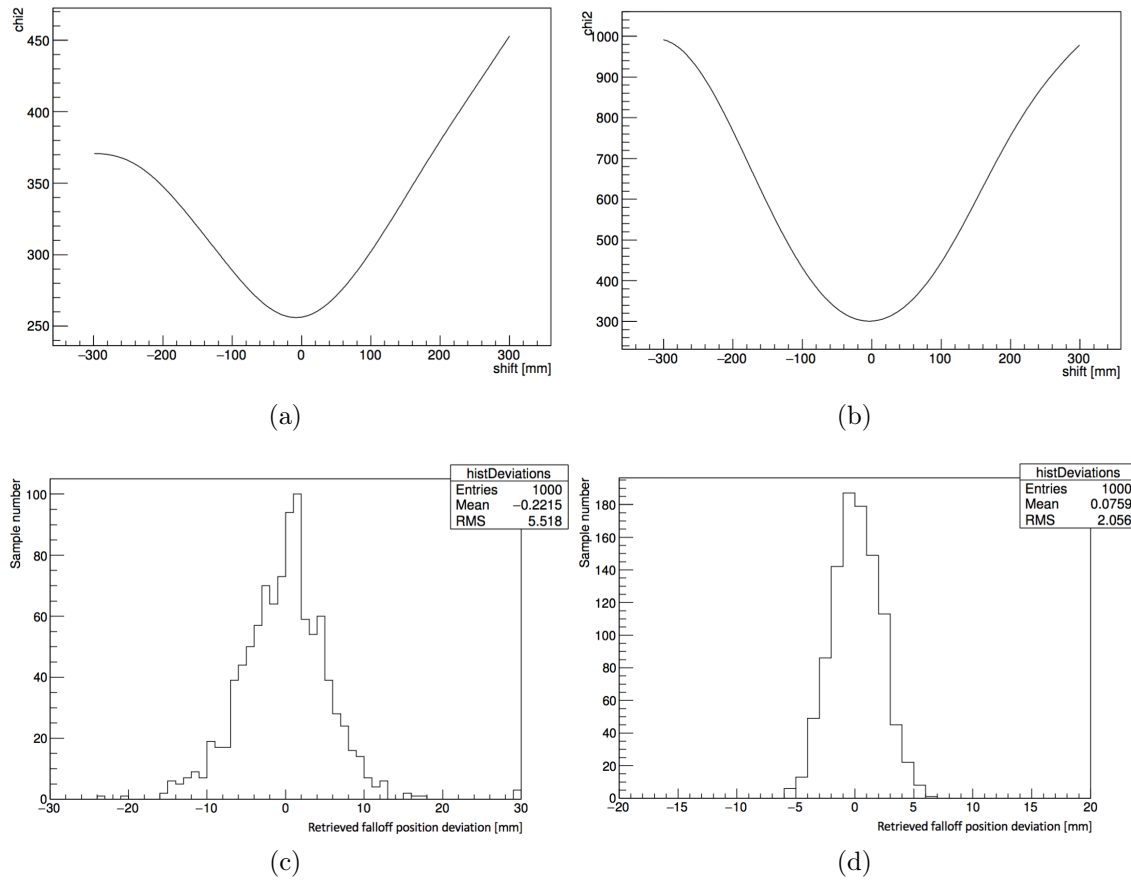


Figure 9: Data processing comparison for the same proton simulation with the line cone algorithm (left column) and the LM-MLEM algorithm (right column). The first row shows the χ^2 distribution for one data subset. The last row represents the distribution of the minimal calculated shifts for 1000 such subsets.

compactness of the detector must be privileged, the efficiency reduction is not dramatic for the Bragg peak monitoring purpose. Moreover, such a compact configuration allows for the implementation of several detector heads, which can be set at different angles and provide additional spatial information. In case the highest possible efficiency is required, the distance between scatterer and absorber can be modified, knowing that for reduced distances the efficiency is increased because a minimal amount of scattered photons escapes the absorber field of view. On the other hand, a reduction of the inter-detector distance is not recommended for accurate TOF measurements (for fixed target-scatterer distance), given the uncertainty added by the detectors time resolutions. In addition to this, also the spatial resolution is affected by a reduction of the scatterer-absorber distance, given the increased amount of photons scattered with large Compton angles included in the collected events. Indeed, large angle scatterings tend to degrade the spatial resolution. Regarding the dependence of the efficiency on the Compton camera position, an accurate setup with respect to the expected Bragg peak position appears

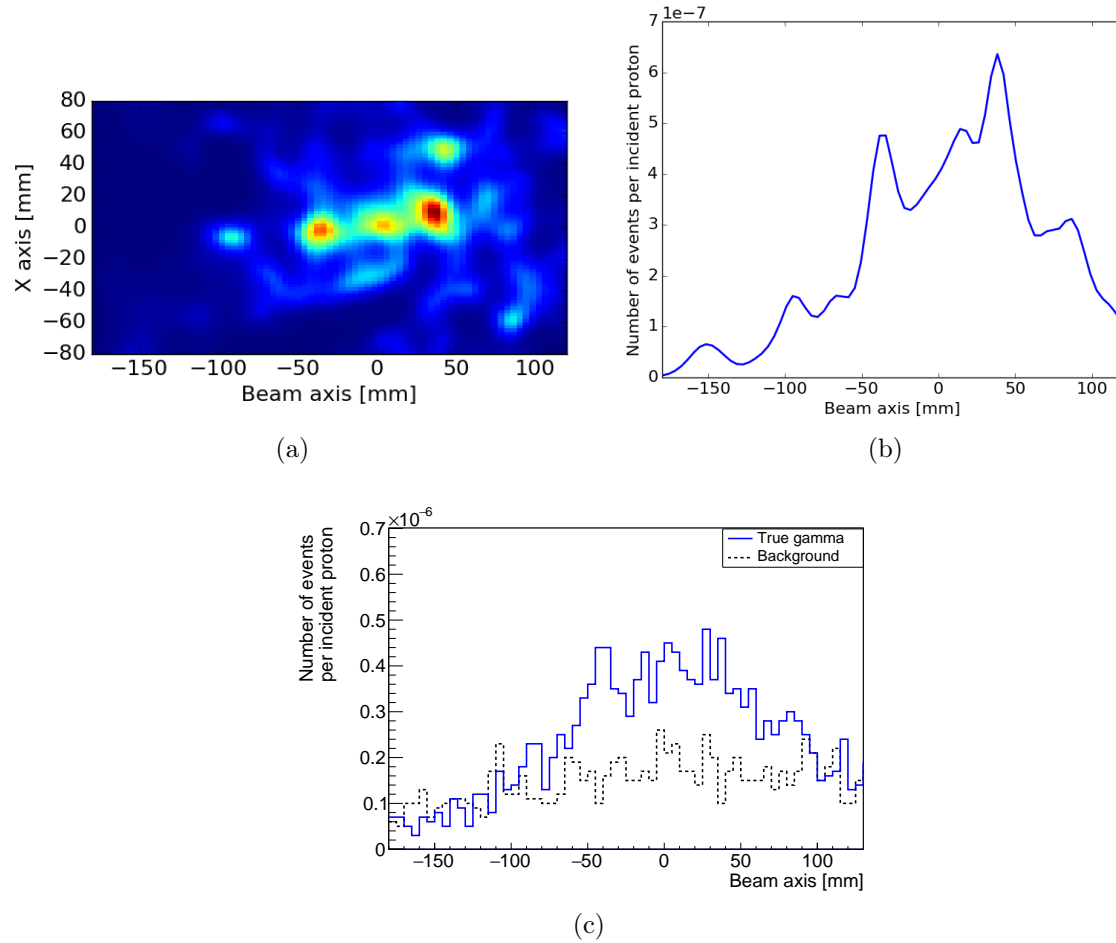


Figure 10: Line-cone and LM-MLEM reconstruction for a 160 MeV proton beam, 10^8 total incident protons. The beam intensity is 1 proton per bunch. The Compton camera is centered at the expected Bragg peak position, $y = +50$ mm. The time-of-flight event selection is applied on the collected data set. 20 iterations are performed for the MLEM reconstruction. Figure (a) represents the MLEM reconstructed 2D image in the plane (x, y) , parallel to the camera entrance surface. The position $x = 0$ mm corresponds to the center of the PMMA phantom and the y direction corresponds to the beam axis, with the target entrance at $y = -100$ mm and the target end at $y = +100$ mm. In figure (b) the mono-dimensional profile along the y axis is sketched. The profile fall-off is located at $y = +50$ mm. Figure (c) shows the profile obtained by means of the line-cone algorithm for the same time-of-flight selected data.

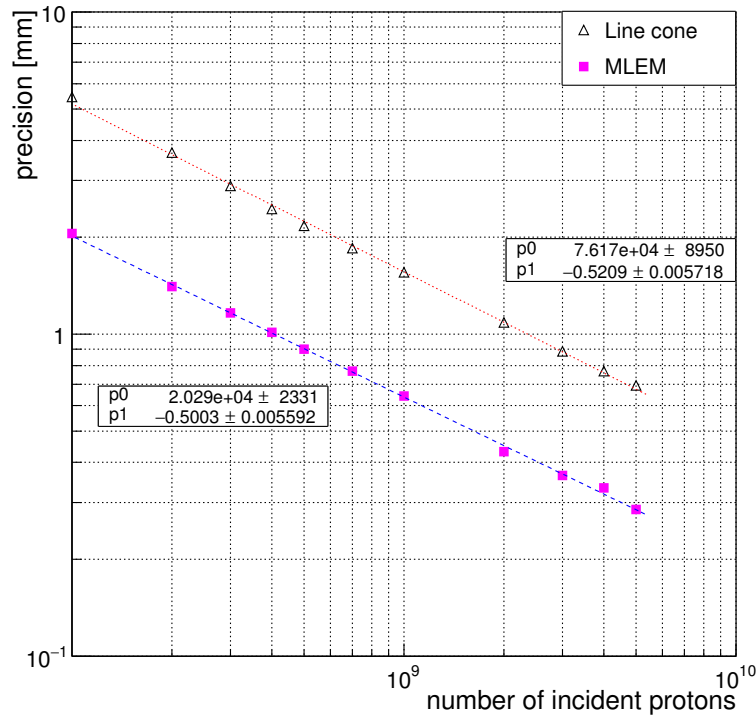


Figure 11: Compton camera precision for two different reconstruction algorithms: line-cone and LM-MLEM. The precision is shown as a function of the total number of incident protons, in the range 1×10^8 to 5×10^9 . A linear fit is realized on the results in order to obtain the slope of the results (p1 parameter in legend). A logarithmic scale is used on both axes.

mandatory for the detection optimization.

A further study showed how this effect is mainly due to the cut applied on the scatterer detector. The study of the signal to noise ratio as a function of beam intensity was performed (see figure 7). The results show that, at clinical intensities, this ratio is very low. The TOF selection is effective in the case of carbon beams, where a significant proton and neutron contamination is expected. If we consider the case of proton beams, the amount of random and true coincidences is comparable at an intensity of about 1 proton per bunch, so that a clinical intensity reduction should be necessary to fit this configuration. Even if a possible image reconstruction is not excluded by the low signal-over-noise ratio detected at realistic beam intensity, given the fact that the random coincidence are distributed in an homogeneous way in the reconstructed volume, an intensity reduction can be an option in order to obtain more significant data sets. It must be noticed that the need for Bragg peak position online check is all the more necessary for distal spots, which are in general the first to be treated, so that a beam intensity reduction at the beginning of the treatment can be foreseen in case an accurate monitoring is strictly needed. This would not affect the treatment delivery, nor the

planned patient rate in the clinical work-flow. In the case of carbon ions, the larger amount of secondary neutron produced during the patient treatment seems to require other background rejection methods in order to lead to an advantageous signal-over-background ratio. Note also that online filtering strategies may be used to improve the quality of the data: already in figure 7, one can notice that the amount of reconstructed events is about half that of true gamma, which means that partial absorption in the absorber leads to events which cannot be reconstructed via the line-cone method. More refined pre-analysis could be used and have been proposed in [Draeger et al., 2017].

The camera precision has been estimated starting from a reference prompt gamma emission profile obtained at high primary particle statistics (10^{10}), with the random extraction of 1000 data subsets per intensity and applying a minimization robust algorithm to define the shift of the identified profile fall-off with respect to the reference one. The camera precision in the falloff identification rapidly increases for increasing primary particle statistics. A sufficiently good precision is achieved on a spot basis for proton beams, where the precision is about 2 mm with a MLEM reconstruction: a qualitative monitoring of each spot seems then possible. In order to achieve millimeter or sub-millimeter precision, some spot grouping methods must be considered. As a general result, the LM-MLEM iterative algorithm, which is now the standard for this kind of image reconstruction, guarantees a better fall-off identification precision over the whole explored intensity range. However, MLEM does not exploit the additional information provided by the knowledge of the beam position in the transverse plane. In future studies, such a priori information should be considered in order to improve the reconstruction rapidity and image quality. At present, given the long calculation time required by the iterative algorithm, the line-cone reconstruction method can still be an option for on-line treatment check, when safety limit can be fixed in order to exclude severe deviation from the treatment planning and an interruption of the dose delivery in real time can be foreseen. Moreover, the TOF information can be included in the line-cone reconstruction method in order to constrain the emission on a single point, and improve its accuracy.

The Compton detection principle has already proven its potential in detecting prompt gammas for ion beam therapy monitoring purpose, and the CLaRyS Compton camera prototype shows promising results for this application. The detector is now at the final development stage, its components are being tested on beam in clinical facilities and a first beam test with a complete system is foreseen for the next year. New simulation studies are to be carried out to benchmark the Compton camera device to other detection systems, like PET machines or collimated detectors, already used or tested in clinics for ion beam therapy monitoring.

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References

- [Aldawood et al., 2017] Aldawood, S., Thirolf, P. G., Miani, A., Böhmer, M., Dedes, G., Gernhäuser, R., Lang, C., Liprandi, S., Maier, L., Marinšek, T., Mayerhofer, M., Schaart, D. R., Valencia Lozano, I., and Parodi, K. (2017). Development of a Compton camera for prompt-gamma medical imaging. *Radiation Physics and Chemistry*, 140:190 – 197. 2nd International Conference on Dosimetry and its Applications (ICDA-2) University of Surrey, Guildford, United Kingdom, 3-8 July 2016.
- [Bom et al., 2012] Bom, V., Joulaeizadeh, J., and Beekman, F. (2012). Real-time prompt gamma monitoring in spot-scanning proton therapy using imaging through a knife-edge-shaped slit. *Physics in Medicine and Biology*, 57(2):297.
- [Bragg and Kleeman, 1904] Bragg, W. H. M. A. and Kleeman, R. (1904). On the ionization curves of radium. *The London, Edinburgh, and Dublin Philosophical Magazine and Journal of Science*, 8(48):726–738.
- [Draeger et al., 2017] Draeger, E., Peterson, S., Mackin, D., Chen, H., Beddar, S., and Polf, J. C. (2017). Feasibility studies of a new event selection method to improve spatial resolution of Compton imaging for medical applications. *IEEE Trans Radiat Plasma Med Sci*, 1(4):358–367.
- [Elsässer et al., 2004] Elsässer, T., Weyrather, W. K., Friedrich, T., Durante, M., Iancu, G., Krämer, M., Kragl, G., Brons, S., Winter, M., Weber, K.-J., and Scholz, M. (2004). Quantification of the relative biological effectiveness for ion beam radiotherapy: Direct experimental comparison of proton and carbon ion beams and a novel approach for treatment planning. *International Journal of Radiation Oncology Biology Physics*, 78(4):1177–1183.
- [Enghardt et al., 2004] Enghardt, W., Crespo, P., Fiedler, F., Hinz, R., Parodi, K., Pawelke, J., and Pönisch, F. (2004). Charged hadron tumour therapy monitoring by means of PET. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 525(1):284 – 288. Proceedings of the International Conference on Imaging Techniques in Subatomic Physics, Astrophysics, Medicine, Biology and Industry.
- [Fontana et al., 2017] Fontana, M., Dauvergne, D., Létang, J., Ley, J.-L., Testa, E., and Maxim, V. (2017). Versatile Compton camera for high-energy gamma rays: Monte carlo comparison with anger camera for medical imaging. *Acta Physica Polonica B*, 48:1639–1646.
- [Frandes et al., 2010] Frandes, M., Zoglauer, A., Maxim, V., and Prost, R. (2010). A tracking Compton-scattering imaging system for hadron therapy monitoring. *IEEE Transactions on Nuclear Science*, 57(1):144–150.
- [Gillam et al., 2011] Gillam, J. E., Lacasta, C., Torres-Espallardo, I., Candela Juan, C., Llosá, G., Solevi, P., Barrio, J., and Rafecas, M. (2011). A Compton imaging algorithm for on-line monitoring in hadron therapy. volume 7961, pages 79611O–79611O–8.
- [Golnik et al., 2014] Golnik, C., Hueso-Gonzalez, F., Muller, A., Dendooven, P., Enghardt, W., Fiedler, F., Kormoll, T., Roemer, K., Petzoldt, J., Wagner, A., and Pausch, G. (2014). Range assessment in particle therapy based on prompt gamma-ray timing measurements. *Phys Med Biol*, 59(18):5399–5422.
- [Grevillot et al., 2011] Grevillot, L., Bertrand, D., Dessy, F., Freud, N., and Sarrut, D. (2011). A Monte Carlo pencil beam scanning model for proton treatment plan simulation using GATE/GEANT4. *Phys Med Biol*, 56(16):5203–5219.
- [Hilaire et al., 2014] Hilaire, E., Robert, C., Lojacono, X., Sarrut, D., Buvat, I., Peyrin, F., and Maxim, V. (2014). Compton imaging in proton therapy: reconstructed image of the simulated prompt- γ distribution. In *ICTR-PHE 2014*, page S43, Geneva, Switzerland.
- [Hilaire et al., 2016] Hilaire, E., Sarrut, D., Peyrin, F., and Maxim, V. (2016). Proton therapy

- monitoring by Compton imaging: influence of the large energy spectrum of the prompt-gamma radiation. *Phys Med Biol*, 61(8):3127–3146.
- [Kormoll et al., 2011] Kormoll, T., Fiedler, F., Schöne, S., Wüstemann, J., Zuber, K., and Enghardt, W. (2011). A Compton imager for in-vivo dosimetry of proton beams - A design study. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 626-627:114 – 119.
- [Krimmer et al., 2017a] Krimmer, J., Angellier, G., Balleyguier, L., Dauvergne, D., Freud, N., Hérault, J., Létang, J. M., Mathez, H., Pinto, M., Testa, E., and Zoccarato, Y. (2017a). A cost-effective monitoring technique in particle therapy via uncollimated prompt-gamma peak integration. *Applied Physics Letters*, 110(15):154102.
- [Krimmer et al., 2017b] Krimmer, J., Dauvergne, D., Létang, J., and Testa, E. (2017b). Prompt-gamma monitoring in hadrontherapy: A review. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*.
- [Krimmer et al., 2015] Krimmer, J., Ley, J.-L., Abellan, C., Cachemiche, J.-P., Caponetto, L., Chen, X., Dahoumane, M., Dauvergne, D., Freud, N., Joly, B., Lambert, D., Lestand, L., Magne, M., Mathez, H., Maxim, V., Montarou, G., Morel, C., Pinto, M., Ray, C., Reithinger, V., Testa, E., and Zoccarato, Y. (2015). Development of a Compton camera for medical applications based on silicon strip and scintillation detectors. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 787:98–101.
- [Kuchment and Terzioglu, 2016] Kuchment, P. and Terzioglu, F. (2016). 3D Image Reconstruction from Compton camera data.
- [Llosá et al., 2012] Llosá, G., Barrio, J., Cabello, J., Crespo, A., Lacasta, C., Rafecas, M., Callier, S., de La Taille, C., and Raux, L. (2012). Detector characterization and first coincidence tests of a Compton telescope based on LaBr3 crystals and SiPMs. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 695:105 – 108. New Developments in Photodetection NDIP11.
- [Lojacono et al., 2013] Lojacono, X., Richard, M.-H., Ley, J.-L., Testa, E., Ray, C., Freud, N., Létang, J., Dauvergne, D., Maxim, V., and Prost, R. (2013). Low statistics reconstruction of the Compton camera point spread function in 3D prompt- γ imaging of ion beam therapy. *Nuclear Science, IEEE Transactions on*, 60(5):3355–3363.
- [Mackin et al., 2012] Mackin, D., Peterson, S., Beddar, S., and Polf, J. (2012). Evaluation of a stochastic reconstruction algorithm for use in Compton camera imaging and beam range verification from secondary gamma emission during proton therapy. *Physics in Medicine and Biology*, 57(11):3537–3553.
- [Matsuoka, 2015] Matsuoka, Y. (2015). Performance of a new electron-tracking Compton camera under intense radiations from a water target irradiated with a proton beam. *JINST*, 10(01):C01053.
- [Maxim, 2014] Maxim, V. (2014). Filtered backprojection reconstruction and redundancy in Compton camera imaging. *IEEE Transactions on Image Processing*, 23(1):332–341.
- [McCleskey et al., 2015] McCleskey, M., Kaye, W., Mackin, D. S., Beddar, S., He, Z., and Polf, J. C. (2015). Evaluation of a multistage CdZnTe Compton camera for prompt- γ imaging for proton therapy. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 785:163 – 169.
- [McKisson et al., 1994] McKisson, J. E., Haskins, P. S., Phillips, G. W., King, S. E., August, R. A., Piercey, R. B., and Mania, R. C. (1994). Demonstration of three-dimensional imaging with a germanium Compton camera. *IEEE Transactions on Nuclear Science*, 41(4):1221–1224.
- [Min et al., 2006] Min, C.-H., Kim, C. H., Youn, M.-Y., and Kim, J.-W. (2006). Prompt-gamma measurements for locating the dose falloff region in the proton therapy. *Applied Physics Letters*, 89(18):183517.
- [Nadaraya, 1964] Nadaraya, E. A. (1964). On estimating regression. *Theory of Probability & Its Applications*, 9(1):141–142.
- [Ordóñez et al., 1997] Ordóñez, C. E., Bolozdynya, A., and Chang, W. (1997). Doppler broadening

- and energy uncertainties in Compton cameras. *IEEE Nucl. Sci. Symp. Conf. Rec. 2* 1361–65.
- [Ortega et al., 2015] Ortega, P. G., Torres-Espallardo, I., Cerutti, F., Ferrari, A., Gillam, J. E., Lacasta, C., Llosá, G., Oliver, J. F., Sala, P. R., Solevi, P., and Rafecas, M. (2015). Noise evaluation of Compton camera imaging for proton therapy. *Phys Med Biol*, 60(5):1845–1863.
- [Paganetti, 2012] Paganetti, H. (2012). Range uncertainties in proton therapy and the role of Monte Carlo simulations. *Physics in Medicine and Biology*, 57(11):R99–R117.
- [Parodi, 2016] Parodi, K. (2016). On- and off-line monitoring of ion beam treatment. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 809:113 – 119. Advances in detectors and applications for medicine.
- [Particle Therapy Cooperative Group, 2017] Particle Therapy Cooperative Group, P. (2017). Ptcog stats. www.ptcog.ch.
- [Peterson et al., 2010] Peterson, S. W., Robertson, D., and Polf, J. (2010). Optimizing a three-stage Compton camera for measuring prompt-gamma rays emitted during proton radiotherapy. *Phys Med Biol*, 55(22):6841–6856.
- [Polf et al., 2015] Polf, J. C., Avery, S., Mackin, D. S., and Beddar, S. (2015). Imaging of prompt-gamma rays emitted during delivery of clinical proton beams with a Compton camera: feasibility studies for range verification. *Physics in Medicine and Biology*, 60(18):7085.
- [Polf et al., 2009] Polf, J. C., Peterson, S., Ciangaru, G., Gillin, M., and Beddar, S. (2009). Prompt-gamma-ray emission from biological tissues during proton irradiation: a preliminary study. *Phys Med Biol*, 54(3):731–743.
- [Priegnitz et al., 2015] Priegnitz, M., Helmbrecht, S., Janssens, G., Perali, I., Smeets, J., Vander Stappen, F., Sterpin, E., and Fiedler, F. (2015). Measurement of prompt gamma profiles in inhomogeneous targets with a knife-edge slit camera during proton irradiation. *Phys Med Biol*, 60(12):4849–4871.
- [Roellinghoff et al., 2014] Roellinghoff, F., Benilov, A., Dauvergne, D., Dedes, G., Freud, N., Janssens, G., Krimmer, J., Létang, J. M., Pinto, M., Prieels, D., Ray, C., Smeets, J., Stichelbaut, F., and Testa, E. (2014). Real-time proton beam range monitoring by means of prompt-gamma detection with a collimated camera. *Physics in Medicine and Biology*, 59(5):1327.
- [Rogers, 2001] Rogers, D. F. (2001). *An Introduction to NURBS: With Historical Perspective*. Morgan Kaufmann Publishers Inc., San Francisco, CA, USA.
- [Schone et al., 2010] Schone, S., Shakirin, G., Kormoll, T., Herbach, C., Pausch, G., and Enghardt, W. (2010). A common approach to image reconstruction for different applications of Compton cameras. In *Nuclear Science Symposium Conference Record (NSS/MIC), 2010 IEEE*, pages 2292–2293.
- [Smeets et al., 2012] Smeets, J., Roellinghoff, F., Prieels, D., Stichelbaut, F., Benilov, A., Busca, P., Fiorini, C., Peloso, R., Basilavecchia, M., Frizzi, T., Dehaes, J. C., and Dubus, A. (2012). Prompt gamma imaging with a slit camera for real-time range control in proton therapy. *Phys Med Biol*, 57(11):3371–3405.
- [Solevi et al., 2016] Solevi, P., Munoz, E., Solaz, C., Trovato, M., Dendooven, P., Gillam, J. E., Lacasta, C., Oliver, J. F., Rafecas, M., Torres-Espallardo, I., and Llosá, G. (2016). Performance of MACACO Compton telescope for ion-beam therapy monitoring: first test with proton beams. *Phys Med Biol*, 61(14):5149–5165.
- [Stichelbaut and Jongen, 2003] Stichelbaut, F. and Jongen, Y. (2003). Verification of the proton beam position in the patient by the detection of prompt gamma-rays emission. *Proceeding of the 39th PTCOG meeting, San Francisco*.
- [Testa et al., 2008] Testa, E., Bajard, M., Chevallier, M., Dauvergne, D., Le Foulher, F., Freud, N., Létang, J.-M., Poizat, J.-C., Ray, C., and Testa, M. (2008). Monitoring the bragg peak location of 73 MeV ^{12}C carbon ions by means of prompt γ -ray measurements. *Applied Physics Letters*, 93(9):093506.
- [Testa et al., 2010] Testa, M., Bajard, M., Chevallier, M., Dauvergne, D., Freud, N., Henriquet, P., Karkar, S., Le Foulher, F., Létang, J. M., Plescak, R., Ray, C., Richard, M.-H., Schardt, D., and

- Testa, E. (2010). Real-time monitoring of the Bragg-peak position in ion therapy by means of single photon detection. *Radiat Environ Biophys*, 49(3):337–343.
- [Todd et al., 1974] Todd, R. W., Nightingale, J. M., and Everett, D. B. (1974). A proposed γ camera. *Nature*, 251:132 EP –.
- [Verburg and Seco, 2014] Verburg, J. M. and Seco, J. (2014). Proton range verification through prompt gamma-ray spectroscopy. *Phys Med Biol*, 59(23):7089–7106.
- [Watson, 1964] Watson, G. S. (1964). Smooth regression analysis. *Sankhyā: The Indian Journal of Statistics, Series A (1961-2002)*, 26(4):359–372.
- [Weyrather et al., 1999] Weyrather, W. K., Ritter, S., Scholz, M., and Kraft, G. (1999). RBE for carbon track-segment irradiation in cell lines of differing repair capacity. *Int J Radiat Biol*, 75(11):1357–1364.
- [Yoshihara et al., 2017] Yoshihara, Y., Shimazoe, K., Mizumachi, Y., Takahashi, H., Kamada, K., Takeda, A., Tsuru, T., and Arai, Y. (2017). Development of electron-tracking Compton imaging system with 30- μm SOI pixel sensor. *Journal of Instrumentation*, 12(01):C01045.
- [Zoglauer et al., 2011] Zoglauer, A., Boggs, S. E., Galloway, M., Amman, M., Luke, P. N., and Marc Kippen, R. (2011). Design, implementation, and optimization of MEGAlib’s image reconstruction tool Mimrec. *Nuclear Instruments and Methods in Physics Research Section A: Accelerators, Spectrometers, Detectors and Associated Equipment*, 652(1):568–571.