# 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 ClaRyS French collaboration is developing a Compton camera prototype based on silicon detectors and BGO scintillator blocks. The present work investigates the clinical applicability of this design by means of Monte Carlo simulations. The Compton camera performance has been studied with the simulation of point-like source irradiation and beam irradiation of a PMMA phantom. The system absolute photon detection efficiency, measured via source irradiation, varies in the range  $1-4 \times 10^{-4}$  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 prompt gamma detected coincidences 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 addition to the already operational in the global tumor therapy panorama. 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 - capability of inducing DNA damages at fixed dose) is observed for ions heavier than protons in the region of the Bragg peak [Elsässer et al., Weyrather et al., 1999], which further enhance 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 mispositioning, 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 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 theses 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, also the relaxation of excited nuclei 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. 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 [Ordonez 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 [Frances 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):

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

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:

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

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 incertitude. 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 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].

A collaboration of 4 research institutions (Institut de Physique Nucleaire and Centre de Recherche en Acquisition et Traitement de l'Image pour la Santé in Lyon, Laboratoire de Physique Subatomique et Cosmologie in Grenoble, Centre de Physique des Particules in Marseille) is focused on the design and development of gamma detectors for online ion beam therapy monitoring. The project includes the development and clinical test of a Compton camera prototype based on semiconductor and scintillator detectors [Krimmer et al., 2015]; the feasibility of its clinical application is studied in this work with Monte Carlo simulations.

In a previous study by Ortega and colleagues [Ortega et al., 2015] a detailed analysis of the noise sources for Compton imaging in proton therapy monitoring is presented, 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, the ClaRyS camera performance is studied as a function of the gamma energy in the prompt gamma energy range. Further, the possible

application of this detector 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 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.

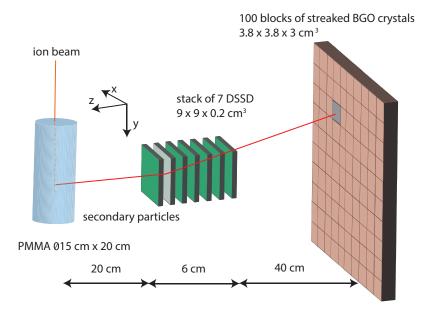


Figure 1: Scheme of the simulation setup: a PMMA cylindrical phantom is set in front of the Compton camera prototype. The Compton camera is composed of a stack of 7 double sided silicon strip detectors (scatterer) and a plan of 100 single BGO blocks. The set distances are realistic for clinical conditions. This geometrical configuration has been used for all the simulations presented in this work.

# 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.

Like most of the Compton camera devices, the CLaRyS prototype includes a scatterer and an absorber. The scatterer consists of seven parallel planes of silicon detectors (double-sided silicon strip detectors, DSSDs),  $9 \times 9 \times 0.2$  cm<sup>3</sup>, with 1 cm distance between the centers of two neighboring planes, while the absorber is composed of an array of  $10 \times 10$  BGO (Bismuth Germanate - Bi<sub>12</sub>GeO<sub>20</sub>) blocks ( $3.5 \times 3.5 \times 3.0$  cm<sup>3</sup> 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, 4.4×4.4 mm<sup>2</sup> 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 interaction occur. An incertitude 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-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 2.2.2). It is set to 1 ns FWHM.

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

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	$2.3~\mathrm{keV}$	17~%	/
${f timing}$ $[{f ns}]$	15	3	1

Table 2: Hadronic models used in the Geant4 simulations.

Process	Protons	Ions	Neutrons	
Electromagnetic	$\operatorname{standard}_{\operatorname{option}3}$			
Inelastic	G4BinaryCascade	G4QMDReaction	G4BinaryCascade	
		(G4IonsShenCrossSection)	+ G4NeutronHPInelastic (<19 MeV)	
Elastic	G4LElastic	G4LElastic	G4LElastic + G4NeutronHPElastic (<19 MeV)	
Fission	/	/	G4LFission + G4NeutronHPFission(<19 MeV)	
$\mathbf{Capture}$	/	/	G4LCapture + G4NeutronHPCapture (<19 MeV)	
Radioactivedecay	/	G4Radioactivedecay	/	

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/n 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/n. The beam intensity for a spot in pencil beam scanning (PBS) mode is  $10^8$  for protons and  $10^5$  for carbon ions. The beam time structure is applied at the data analysis stage and described in section 2.2.1.

#### 2.2. Data processing

2.2.1. Time structure and processing 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) [F Roellinghoff, 2014]. The clinical beam intensity is 3.2 nA which corre-

sponds to about 200 protons per bunch. Concerning the carbon ion beam at 305 MeV/u, the estimated microstructure 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  $5 \times 10^7$  ions/s, 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.

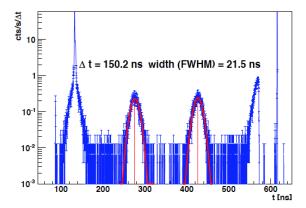


Figure 2: Time micro-structure measured from a carbon ion beam at 400 MeV/u delivered at HIT. The pulses have an extraction period of 150.2 ns and the bunches are 21.5 ns FWHM. The measurement was achieved with a two-scintillating-fiber hodoscope.

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.

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	$2 \times 10^{10} \; { m p/s}$	$5 \times 10^7 \text{ ions/s}$
	Energy	$160~{ m MeV}$	$305~\mathrm{MeV/u}$
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 [ns]	Si: 15 and BGO: 3	

2.2.2. 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 2.2.1, the coincidence reconstruction relies on a defined time window, fixed according to the detector resolution. In a simulation environment, different kinds of coincidences 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;
- 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 socalled random coincidences, which cannot be experimentally distinguished form true coincidences. 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 discrimination is possible and effective in reducing the background generated by massive particles interactions [Testa et al., 2010]. The photons are moving at approximately the speed of light while the massive particles approach the detector at a lower speed. 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 between the incident particle creation and the secondary particle

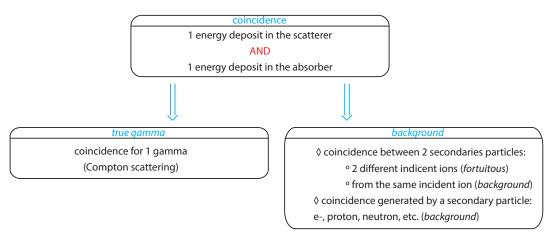


Figure 3: Diagram showing the different definitions of coincidences in the Compton camera.

detection in the absorber is considered as the time-of-flight. The hodoscope time resolution  $u_{\rm hodoscope}$  (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.

$$TOF_{\text{theoretica}} = t_{\text{absorber}} - t_{\text{hodoscope}}$$
 (3)

$$TOF_{\text{simulation}} = t_{\text{absorber}} + t_{\text{creation}} + u_{\text{hodoscope}}$$
 (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.

#### 2.3. Reconstruction algorithm

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.

2.3.1. 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

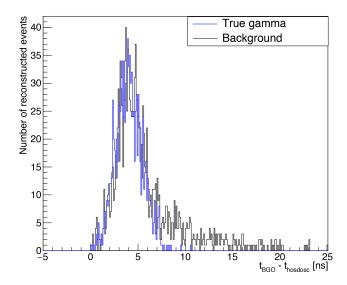


Figure 4: Time of flight spectrum obtained by means of the simulation for a proton beam at 160 MeV and  $1 \times 10^8$  incident protons. The detection time for the absorber is given by the tag for the first deposit energy. The blue curve represents the time of flight for true events and the black one represents the background.

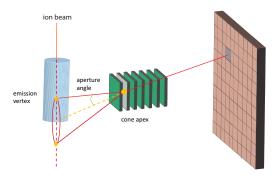


Figure 5: Scheme of the reconstruction principle for Compton events.

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.

We assume that the initial energy of the gamma ray is fully absorbed in the absorber. The interaction position in the scatter gives the cone apex and the line connecting the interaction position in scatterer and absorber gives the cone axis. 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.

2.3.2. 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 a MLEM version which allows to free from sinograms and to reconstruct the image directly from the list of detected events. 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  $\lambda_j$  (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 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  $\lambda^{(0)}$ , which can be the simple back-projection reconstruction. The iterations rely on the following recurrence relation:

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

where  $N_{\gamma}$  is the number of detected events and  $N_v$  is the number of voxels in the image.

The LM-MLEM algorithm used for this study is the one developed by the CREATIS research group in Lyon [Maxim, 2014, Hilaire et al., 2014].

For each photon detected, the system matrix T is 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:

$$t_{ij} = K(\beta_i, E_{tot}) \frac{|\cos(\theta_{\overline{V2V1}})|}{V_2 V_1^2} \int_{M \in v_i} \frac{|\cos(\theta_{\overline{V_1M}})|}{V_1 M^2} h_i(M) dv, \tag{6}$$

where  $\beta_i$  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 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. The spatial resolutions are not included in this version of the algorithm.

For each iteration, the matrix T is stored and the reconstructed image can be produced via a Matlab simple interface.

### 2.4. Precision estimation

The camera precision is defined as the difference between the expected Bragg peak position (according to the treatment planning) and the Bragg peak position reconstructed through the data. This is a critical parameter which will be used to fix a decision threshold during a ion beam treatment.

In this study a reference profile has been defined as the reconstructed emission vertex profile at high statistics, via analytic (line cone) and iterative (LM-MLEM) algorithms; this will be used to mimic the treatment planning. This Bragg peak position is compared to the ones reconstructed at lower statistics (close to clinical conditions) via analytic and iterative algorithms.

The high statistic profile has been obtained with  $2 \times 10^{10}$  incident protons (figure 6a) for the line-cone reconstruction and with  $1 \times 10^{10}$  incident protons (figure 6b) for the LM-MLEM reconstruction method. The SmoothKern method, with the Nadaraya-Watson regression, is used to smooth the reference profiles in order to reduce relative statistic fluctuations.

A region of interest (ROI) ranging from y = 0 mm to y = +100 mm is defined around the expected Bragg peak position, located at y = +50 mm in the phantom. The reference profiles are modelled in the ROI by a linear combination named Non-Uniform Rational Basis Splines (NURBS).

A new independent profile should be simulated at low statistics for the precision calculation. However, with the aim to speed up the analysis, a random extraction from the NURBS profile (6c and 6d) is performed following the Poisson law, in order to select reduced data sets. The desired low statistics ranges from  $10^8$  to  $5 \times 10^9$  incident protons.

The minimal distance between the NURBS reference profile and the low statistic reconstructed profiles is estimated thanks to the  $\chi^2$  method. For the  $\chi^2$  estimate, the low statistics profile is defined around the fall-off position from y = +30 mm to y = +70 mm. The minimization process tests the profiles at low statistics on a 60 mm variation range, between -30 mm and +30 mm with respect to the initial position, with a step of 0.1 mm. The  $\chi^2$  is calculated as follows:

$$\chi^2 = \sum_{i=1} \left( y_{sample,i} - y_{NURBS,i} \right)^2, \tag{7}$$

where  $y_{sample}$  is the number of coincidences for the low statistics profile,  $y_{NURBS}$  is the number of coincidences for the reference profile NURBS (scaled at the same low statistic) and i the step number.

The global minimum of all the reconstructed profiles is then retrieved. The figures 6e and 6f show the distribution of  $\chi^2$  calculated for a low statistic profile at  $10^8$  incident protons.

A total of a thousand profiles at the low statistic are generated (named realizations) and the  $\chi^2$  minimization is applied to each of them. The standard deviation of the

distribution resulting of the thousand results gives the precision of the camera for a given number of incident protons. The figure 6g and 6h show the distributions at  $10^8$  incident protons for the line-cone and LM-MLEM algorithms.

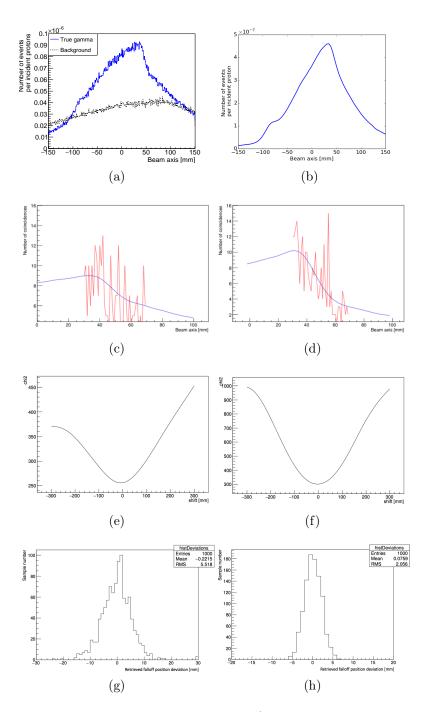


Figure 6: Treatment comparison for the same proton simulation with the line cone algorithm (left column: figures 6a, 6c, 6e, 6g) and the LM-MLEM algorithm (right column: figures 6b, 6d, 6f, 6h). The first row gives the reconstructed profile for  $2 \times 10^{10}$  incident protons (figure 6a) and for  $1 \times 10^{10}$  incident protons (figure 6b) respectively. The figures 6c and 6d are showing the reference curve (blue) and the estimated curve with Poisson's law at  $1 \times 10^8$  incident protons. The figures 6e and 6f are the  $\chi^2$  minimization results for one realization. The minimum of the curve gives the best fit between the reference curve and the low statistics one. Finally, the figures 6g and 6h represents the results for  $1000 \chi^2$  minimizations. The Compton camera precision is estimated thanks to the standard deviation of the distribution.

#### 3. Results

In the following paragraphs the results of the simulation study are presented and discussed with constant reference to the possible clinical application of the CLaRyS Compton camera system as particle therapy monitor. The clinical implementation of such a monitoring system has been tested with the simulation of clinical proton and carbon ion beams, with realistic spatial and time structure reproducing two accelerators at present employed for treatments 2.2.1. As preliminary study, the detection efficiency of the camera has been tested with the simulation of the irradiation with point-like gamma sources in different position with respect to the center of the camera. After that, the beam exposure has been simulated with different beam intensities for an analysis of the clinical-like detection environment (background, random coincidence contamination), in order to focus in the end on the event reconstruction with the comparison of two different methods (line-cone analytical reconstruction and MLEM iterative algorithm) and on the estimate of the camera precision in the identification of the fall-off of the dose profile.

## 3.1. Compton camera absolute efficiency

The absolute efficiency is crucial for the Compton camera performances and for its possible application in treatment monitoring. As already mentioned in section 1, 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], an 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 monoenergetic gamma sources, set in different positions with respect to the center of the camera on the transverse plane.

The absolute efficiency  $\epsilon$  is defined as:

$$\epsilon = \frac{N\gamma_{recons}}{N\gamma_{total}},\tag{8}$$

with  $N\gamma_{recons}$  the number of gamma events in coincidences,

 $N\gamma_{total}$  the total number of gamma emitted: 10<sup>8</sup>.

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 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 has been tested to mimic different prompt gamma spectroscopic lines: 300 keV, 500 keV, 1 MeV, 2 MeV, 4 MeV, 6 MeV.

The obtained results are shown in figure 7. 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.

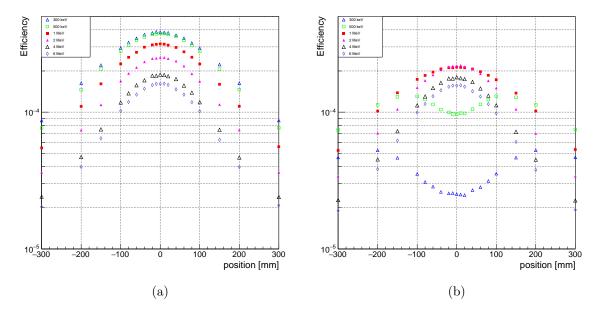


Figure 7: Absolute Compton camera efficiency as a function of the gamma source position for different gamma energies, in the range between 300 keV to 6 MeV. The left side shows the camera efficiency without any applied energy cut. It models an ideal Compton camera. In the right side, detector energy cuts are applied: the lower energy thresholds are set to 50 keV for the scatterer and 100 keV for the absorber, to reproduce realistic scenario. This values can change for the final detector 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  $4 \times 10^{-4}$  at 300 keV and  $1.5 \times 10^{-4}$  at 6 MeV (figure 7). Moreover, it can be noticed how the efficiency quickly drops quickly as the point source drifts 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). Therefore, considering 1 MeV as the average and reference gamma energy for ion beam therapy monitoring, the Compton camera position optimization with respect to the expected Bragg peak position appears mandatory for an efficient fall-off reconstruction.

Regarding the applied energy cuts, figure 7(b) clearly shows an important effect on low energy gammas in the central area of the camera. The effect is negligible for positions with a distance greater than 200 mm from the center of the camera, and for any distance at energies above 2 MeV. A further study showed how this effect is mainly due to the cut applied on the scatterer detector. Indeed, for low energies, below 2 MeV, the increased efficiency in the central section of the camera active surface shown in figure 7(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 cut. The effect is all the more important as the primary gamma energy is limited, creating the peculiar energy dependence of the efficiency reduction emerged in the results in figure 7(b).

# 3.2. Beam intensity

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 asses the possible clinical implementation of a Compton detection based monitoring of ion beam treatment. 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. 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.

In figure 8 the different components of the resulting secondary signal are shown as a function of the beam intensity. The true coincidences represent detector time coincidences generated by the same gamma ray. All the other coincidence types compose the background. All the simulations are performed with  $10^8$  primary protons and  $2 \times 10^5$  primary carbon ions. Those value correspond to realistic clinical values used for the treatment of a single spot. The collected data sets are reported with and without the applied time-of-flight discrimination, mainly employed for neutron rejection, as mentioned in section 2.2.2.

In figure 8(a) and (b) the amount of true gamma coincidences and background events are reported before and after reconstruction (the reconstruction algorithms select the events according to the chosen reconstruction parameters) 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

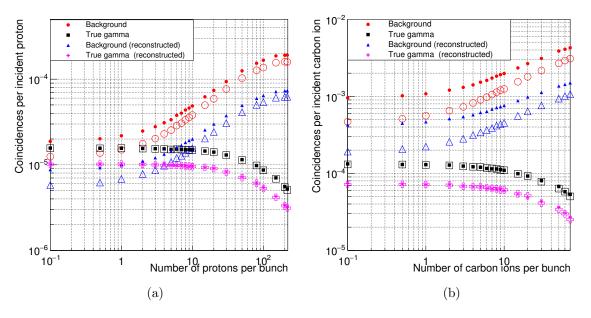


Figure 8: Coincidences yield for protons (left) and carbon ions (right) as a function of the beam intensity. The intensity is reported as number of incident particles per bunch. The filled markers correspond to the collected data without time-of-flight discrimination, while this cut is applied to the data reported with empty markers. Moreover, the yields are given before and after the profile reconstruction with the line cone algorithm.

#### number of incident ions.

As expected, 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 clinical intensity (200 proton/bunch) with no event selection, while a factor more than two times higher is reported for carbon ions in the same conditions. Even if the time-of-flight selection can slightly improve this result by reducing the amount of background events, it is evident that such a beam intensity is not optimal for a proper monitoring. Furthermore, considering the rate of single interactions per detector, the results obtained by a preliminary analysis showed, as an example, a single rate of about 300 MHz on the absorber and 20 MHz on the first scatterer planes. These rates are not compatible with the detection rate capabilities of the detector read-out chains. As a result, it appears not possible to perform a valuable treatment monitoring with the CLaRyS Compton camera at a clinical beam intensity. Nevertheless, the monitoring objective becomes feasible in a reduced beam intensity scenario for proton irradiation. Figure 8(a) shows how the amount of true gamma events and background events becomes compatible at the intensity of about 1 proton per bunch. 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.

Given the results presented in this paragraph, in the following the attention will be focused on proton beams at reduced intensities, with the aim of comparing the line-cone reconstruction and the MLEM reconstruction algorithm and to estimate the camera fall-off identification precision.

# 3.3. Compton camera precision: comparaison LM-MLEM vs Line cone reconstruction

The simulation setup shown in figure 1 has been implemented to test the line-cone analytic reconstruction method and compare it to the iterative LM-MLEM algorithm developed in Lyon by the CREATIS group [Maxim, 2014, Hilaire et al., 2014]. A single data set has been collected, corresponding to the irradiation of the PMMA phatom with a proton 160 MeV monoenergetic beam, with a reduced intensity of 1 proton per bunch. A total of 10<sup>10</sup> 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 2.4. Figure 9 shows the results of the reconstruction of the data set obtained for the selection of 10<sup>8</sup> primary protons of 160 MeV, via the two reconstruction method already detailed before. 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 produce a mono-dimensional image as result, while the MLEM method is able to reconstruct the prompt-gamma emission distribution in 3 dimensions. A projection along the beam direction is then extracted from the MLEM reconstructed image in order to allow a direct comparison.

Starting from the reference data set with  $10^{10}$  primary protons, 11 subsets are extracted to mimic the dose profile falloff reconstruction for different primary particle statistics, in the range  $1\times10^8$  to  $5\times10^9$ . To be noticed that the delivered beam intensity is always of 1 proton per bunch, as a consequence of the results presented in section 3.2. The analysis method described in section 2.4 is applied to the different data set to retrieve the camera precision in the dose profile falloff identification. The results are shown in figure 10, where the two reconstruction methods are represented by different markers.

At the expense of an increased calculation time, the iterative MLEM reconstruction method allows to achieve a better precision, with a reduction of about 3 mm of the falloff retrieval precision 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  $5\times10^9$  protons, which can correspond to the monitoring of a group of spot with the same planned range. If we consider the important increase in the precision of the falloff identification given by the increased statistics, the spot grouping method seems to be promising for monitoring purpose when high accuracy is required, probably obtained in the post-treatment due to the long required reconstruction time. A sufficiently good

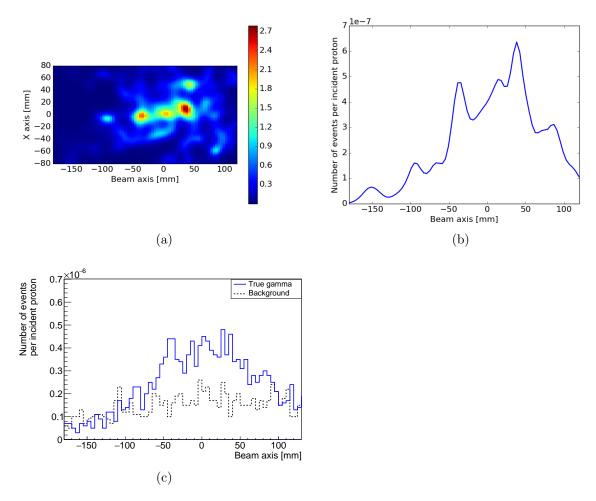
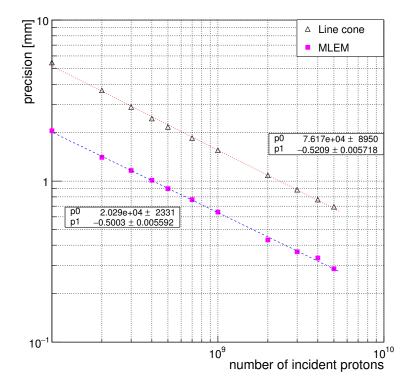


Figure 9: Line cone and LM-MLEM reconstruction for a 160 MeV proton beam,  $10^8$  total incident protons. The beam intensity is 1 proton per bunch. The Compton camera is centered at the expected Bragg peak position, y = +50mm. The time-of-flight event selection is applied on the collected data set. 20 iterations are performed for the MLEM reconstruction.

Figure (a) represents the MLEM reconstructed 2D image in the plan (x,y), parallel to the camera entrance surface. The position x=0mm corresponds to the center of the PMMA phantom and the y direction corresponds to the beam axis. In figure (b) the mono-dimensional profile along the y axis is sketched. The profile peak is located at y=+50mm. Figure (c) shows the profile obtained by means of line-cone algorithm for the same time-of-flight selected data.

precision is achieved on a spot basis, where the precision is about 2 mm with a MLEM reconstruction: a qualitative monitoring of each spot seems then possible. Given the long calculation time required by the MLEM algorithm, the line-cone reconstruction method, despite the reduced precision, 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

Figure 10: 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  $1 \times 10^8$  to  $5 \times 10^9$ . 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.



planning and an interruption of the dose delivery in real time can be foreseen.

# 4. Discussion

DISCUSSION

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