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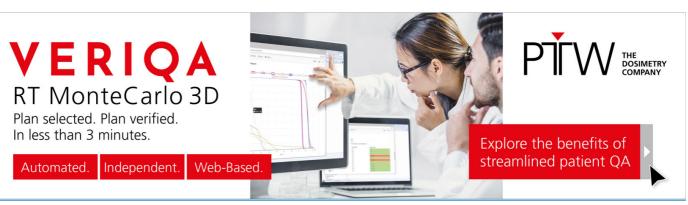
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ProTheRaMon—a GATE simulation framework for proton therapy range monitoring using PET imaging

Damian Borys^{1,2,*}, Jakub Baran^{3,4}, Karol Brzeziński^{5,6}, Jan Gajewski⁵, Neha Chug^{3,4}, Aurelien Coussat^{3,4}, Eryk Czerwiński^{3,4}, Meysam Dadgar^{3,4}, Kamil Dulski^{3,4}, Kavya V Eliyan^{3,4}, Aleksander Gajos^{3,4}, Krzysztof Kacprzak^{3,4}, Łukasz Kapłon^{3,4}, Konrad Klimaszewski⁷, Paweł Konieczka⁷, Renata Kopeć⁵, Grzegorz Korcyl^{3,4}, Tomasz Kozik^{3,4}, Wojciech Krzemień⁸, Deepak Kumar^{3,4}, Antony J Lomax^{9,10}, Keegan McNamara^{9,10}, Szymon Niedźwiecki^{3,4}, Paweł Olko⁵, Dominik Panek^{3,4}, Szymon Parzych^{3,4}, Elena Perez del Rio^{3,4}, Lech Raczyński⁷, Sushil Sharma^{3,4}, Shivani^{3,4}, Roman Y Shopa⁷, Tomasz Skóra¹¹, Magdalena Skurzok^{3,4}, Paulina Stasica⁵, Ewa Ł Stępień^{3,4,14}, Keyvan Tayefi^{3,4}, Faranak Tayefi^{3,4}, Damien C Weber^{9,12,13}, Carla Winterhalter^{9,10}, Wojciech Wiślicki⁷, Paweł Moskal^{3,4,14}, and Antoni Rucinski⁵,

¹ Silesian University of Technology, Department of Systems Biology and Engineering, Gliwice, Poland

- Biotechnology Centre, Silesian University of Technology, Gliwice, Poland
- Faculty of Physics, Astronomy, and Applied Computer Science, Jagiellonian University, Łojasiewicza 11, 30-348 Kraków, Poland
- Total-Body Jagiellonian-PET Laboratory, Jagiellonian University, Kraków, Poland
- ⁵ Institute of Nuclear Physics PAN, Kraków, Poland
- Instituto de Física Corpuscular (IFIC), CSIC-UV, Valencia, Spain
- Department of Complex Systems, National Centre for Nuclear Research, Otwock-Świerk, Poland
- High Energy Physics Division, National Centre for Nuclear Research, Otwock-Świerk, Poland
- Centre for Proton Therapy, Paul Scherrer Institute, Villigen, Switzerland
- ¹⁰ Physics Department, ETH Zürich, Zürich, Switzerland
- 11 National Oncology Institute, National Research Institute, Krakow Branch, Krakow, Poland
- ¹² Radiation Oncology Department, University Hospital of Bern, University of Bern, Switzerland
- ¹³ Radiation Oncology Department of the University Hospital of Zürich, Switzerland
- ¹⁴ Center for Theranostics, Jagiellonian University, Kraków, Poland
- * $\,$ Author to whom any correspondence should be addressed.

E-mail: damian.borys@polsl.pl

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Abstract

Objective. This paper reports on the implementation and shows examples of the use of the ProTheRaMon framework for simulating the delivery of proton therapy treatment plans and range monitoring using positron emission tomography (PET). ProTheRaMon offers complete processing of proton therapy treatment plans, patient CT geometries, and intra-treatment PET imaging, taking into account therapy and imaging coordinate systems and activity decay during the PET imaging protocol specific to a given proton therapy facility. We present the ProTheRaMon framework and illustrate its potential use case and data processing steps for a patient treated at the Cyclotron Centre Bronowice (CCB) proton therapy center in Krakow, Poland. Approach. The ProTheRaMon framework is based on GATE Monte Carlo software, the CASTOR reconstruction package and inhouse developed Python and bash scripts. The framework consists of five separated simulation and data processing steps, that can be further optimized according to the user's needs and specific settings of a given proton therapy facility and PET scanner design. Main results. ProTheRaMon is presented using example data from a patient treated at CCB and the J-PET scanner to demonstrate the application of the framework for proton therapy range monitoring. The output of each simulation and data processing stage is described and visualized. Significance. We demonstrate that the ProTheRaMon simulation platform is a high-performance tool, capable of running on a computational cluster and suitable for multi-parameter studies, with databases consisting of large number of patients, as well as different PET scanner geometries and settings for range monitoring in a clinical environment. Due to its modular structure, the ProTheRaMon framework can be adjusted

for different proton therapy centers and/or different PET detector geometries. It is available to the community via github (Borys *et al* 2022).

1. Introduction

The main advantage of proton radiotherapy, compared to conventional radiotherapy using x-rays or electrons, is the ability to precisely deliver the therapeutic dose to the tumor, while sparing surrounding healthy tissues in direct vicinity of the target volume. This characteristic property is due to the depth dose distribution of proton beams, which deposit the maximum dose at the end of their range (in the Bragg peak; BP), followed by a steep dose fall-off (Wilson 1946). Although the steep dose gradient could be beneficial for sparing critical structures located distally to the tumor, it is not fully exploited in clinical practice because of the uncertainty of the BP position in depth, commonly named *range uncertainty* (Knopf and Lomax 2013). The range uncertainty may occur due to uncertainties in CT calibration, imaging artifacts and variations in patient anatomy during the course of the fractionated treatment, and may lead to differences between the planned dose distribution and the dose actually administered. Therefore, a major subject in proton radiotherapy research is the development of detector techniques for proton range monitoring, which would allow treatment planning and delivery to fully benefit from the physical properties of therapeutic proton beams (Parodi and Enghardt 2000, Shakirin *et al* 2011, Knopf and Lomax 2013, Parodi 2020).

The most widely explored approaches for range monitoring are based on measurements of secondary particles produced in nuclear reactions of the proton beam with nuclei in the patient's body. One approach concentrates on measurements of prompt gammas emitted from the excited nuclei (Kurosawa et al 2012, Krimmer et al 2018). Another focuses on the measurements of 511 keV electron-positron annihilation gammas, where the positrons are the decay product of β^+ -emitting isotopes generated by the proton beam during irradiation. These 511 keV gammas can be measured with positron emission tomography (PET) scanners (Hishikawa et al 2002, Enghardt et al 2004, Parodi et al 2005, 2007b, Nishio et al 2010). Different approaches to PET range monitoring have been investigated to overcome the limitation of incorporating the scanner in the treatment room (Kraan 2015). Commercially available diagnostic PET systems installed outside the treatment room were used for imaging patients offline, however the obtained PET images suffered low signal and biological washout (Bauer et al 2013), which limited their clinical use for range monitoring. Early developments of on-line PET range monitoring systems date to 1990s at Gesellschaft für Schwerionenforschung GSI, Darmstadt, Germany (Jäkel et al 2022) and in 2000s at the proton therapy facility of the National Cancer Center, Kashiwa, Japan (Nishio et al 2010). The application of the commercial PET scanners tested at Heidelberg Ion Beam Therapy Center (HIT), Germany, and Massachusets General Hospital (MGH), Boston, USA (Parodi et al 2008) showed insufficient efficiency of the off-line range monitoring. The mid 2010s saw the development of new era in situ PET systems capable of acquiring treatment verification images during and/or just after the irradiation (Ferrero et al 2018).

Monte Carlo (MC) simulations are a useful clinical and research tool in medical physics, enabling study of the feasibility of new treatment and imaging techniques, often supplementing complex and demanding measurements (Muraro et al 2020). In proton therapy, MC techniques are extensively used to optimize the design of new detectors and for preliminary characterization of their performance (Kraan 2015). The design and optimization of new PET scanner systems for range monitoring is particularly challenging. It requires the simulation of proton beam delivery and production of β^+ -emitting isotopes in the patient geometry, taking into account the isotope decay during patient handling in the treatment room, the simulation of 511 keV gamma transport to the PET detectors of particular design, the coincidence events scoring and PET image reconstruction. The full MC simulation in a one-step approach is not optimal for this purpose because it requires time-consuming tracking of the history of several particles and their interactions and a complex data analysis (Muraro et al 2020). Further challenges relate to the implementation of the software environment and the time performance of simulations. The simulation of the production of β^+ -emitters requires relatively large statistics of primary particles compared to simulations of dose or linear energy transfer distributions. Moreover, the effectiveness of range monitoring depends not only on the detector system performance but is also specific to treatment indication, tumor volume and location. Therefore, new PET designs need to be optimized and verified by extensive simulation studies with multiple treatment scenarios and patient treatment plans. Given the high complexity of the proton treatment and PET imaging simulations, there is need for a simulation framework which is flexible, automated, and has good time performance.

In this paper, we present a simulation framework, PROton THErapy RAnge MONitoring— ProTheRaMon, based on the Geant4/GATE Monte Carlo toolkit, the Customizable and Advanced Software for Tomographic Reconstruction (CASTOR) PET image reconstruction toolkit and in-house implemented scripts. The ProTheRaMon framework was designed with the aim of characterizing and optimizing the performance of a PET scanner for proton therapy range monitoring. Our goal was to implement an automated and time-efficient simulation platform to compute the expected activity of β^+ emitters in phantoms and patients undergoing proton therapy and to predict PET images that may be acquired after the irradiation of separate fields or after the whole treatment.

The requirement of time-efficient simulations has been addressed here by a multi-stage simulation protocol and its deployment on a computational cluster. In the methods section of this paper, we present the implementation and functionality of the framework. In the results section, we show an example application of the framework to the simulation of proton treatment and computation of the β^+ production in a patient treated at the Cyclotron Centre Bronowice (CCB) in Kraków, Poland, as well as PET acquisition and image reconstruction using the J-PET, a novel multi-photon portable J-PET scanner based on plastic scintillator (Rucinski *et al* 2018, Jakub *et al* 2019, Moskal *et al* 2021a, 2021b).

So far, every institute who wants to perform PET simulations has to set up their own simulation workflow. ProTheRaMon is available to the community, and as it is modular, it can easily be adapted and expanded for the specific institution.

2. Materials and methods

The ProTheRaMon framework addresses the challenge of combining software of different types and data in various formats. The clinical data include a CT scan of the phantom or patient, description of clinical structures and a treatment plan consisting of one or more irradiation fields. Accurate simulations of dose and activity produced in a phantom or a patient require a dosimetrically validated proton beam model from a specific proton facility that needs to be provided. Due to the time gap between the end of the irradiation and beginning of PET image acquisition, the activity decay is modeled for each β^+ -emitting radioisotope individually using in-house developed scripts. The simulation of the emission of the 511 keV annihilation photons towards the PET detector is followed by scoring of coincidence events in a predefined PET scanner geometry. The PET scanner geometry definition is also used during image reconstruction.

For the validity of simulations, it is important to implement a geometrical transformation between the coordinate systems specific for the irradiation and for PET scan acquisition. See figure 1 for details on relationship between those coordinate systems.

For the simulation of the proton treatment and PET imaging, the following software packages have been used. The ProTheRaMon framework combines the GATE software package (Jan *et al* 2004, 2011), which is based on the general purpose Geant4 Monte Carlo simulation platform (Agostinelli *et al* 2003, Allison *et al* 2016), CASTOR image reconstruction toolkit (Merlin *et al* 2018), GPU-accelerated Monte Carlo code FRED (Schiavi *et al* 2017, Gajewski *et al* 2021), and several in-house implemented Python and bash scripts enabling parallelization of the tasks on a computational cluster, conversion between data formats, data processing, analysis and visualization. We have created, tested and evaluated our framework using GATE version 9.0 (with Geant4 10.7 patch 1), CASTOR version 3.1, FRED version 3.60, FREDtools version 0.6.54. The software framework design allows its easy adaptation to a specific proton therapy facility and PET scanner geometry as well as the needs of any research group conducting simulation-based investigations into using PET for proton therapy range monitoring approaches and development of PET detector technologies.

For MC simulations, the GATE software was selected due to its unique capability of combining proton therapy simulations in the patient, the scoring of the production of β^+ emitters, β^+ decay and annihilation, the propagation of the 511 keV photons, as well as detection by the PET scanner and processing into coincidence data stored in *root* files. For the PET image reconstruction, among several available image reconstruction software tools, e.g. STIR (Thielemans *et al* 2012), PRESTO (Scheins *et al* 2011), OMEGA (Wettenhovi *et al* 2021) and others, we have selected CASTOR (Merlin *et al* 2018). The choice of CASTOR is motivated by its design intrinsically compatible with the GATE output data format and the ability to import the PET geometry directly from GATE macros. The advanced PET image reconstruction options offered by CASTOR include time-offlight (TOF) processing as well as handling of multi-layer and non-cylindrical PET scanner designs often used for PET-based range monitoring, which are additional requirements for the PET detector used in our investigations. CASTOR offers tools for automatic conversion of the GATE geometry macro file to the CASTOR format when using the *cylindricalPET* system. CASTOR is very flexible, users could potentially create any PET scanner geometry and describe it in a *.lut file which does not create any limitations for GATE simulations. In that case, the user would have to create a proper converter of the coincidences saved in ROOT format to the CASTOR compatible format.

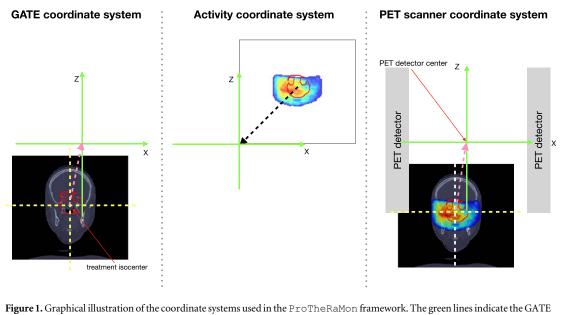


Figure 1. Graphical illustration of the coordinate systems used in the ProTheRAMON framework. The green lines indicate the GATE coordinate system associated with the GATE *World* volume, the yellow dashed lines show the treatment isocenter, and the white dashed line corresponds to the CT coordinate system. The left panel shows he translation of the treatment isocenter to the center of the GATE coordinate system. The middle panel shows how the activity distribution is aligned with the CT and GATE coordinate systems. The right panel shows how the CT and activity images are positioned in the PET detector coordinate system. Note that for the positioning of the patient for PET imaging (right panel), the X and Y coordinates belong to the CT coordinate system, while the Z coordinate is taken from the treatment isocenter. This approach guarantees that the patient fits in the cylindrical PET scanner in the same way as in the CT scanner (X and Y coordinates), concurrently optimizing the PET signal in the detector (Z coordinate).

STAGE 1	STAGE 2	STAGE 3	STAGE 4	STAGE 5
IN-HOUSE (Python / Bash)	GATE	IN-HOUSE (Python / Bash)	GATE	CASToR
GATE coordinate system	GATE coordinate system	Activity coordinate system	PET scanner coordinate system	GATE coordinate system
СТ			ст	
Treatment Plan			Treatment Plan	
Beam model				
CT callibration			CT callibration	
		Production maps	Activity maps	Coincidences list
			PET scanner geometry	

Figure 2. Graphical illustration of the five stages of the ProTheRaMon framework, software tools used by each of the stages, as well as data required for each of the stages. Blue refers to the input data for each stage, red refers to data that are treated as input data for one stage but was produced by previous stages.

2.1. Design and implementation of the ProTheRaMon framework

In this section we present the concept, design and implementation of the ProTheRaMon framework. The framework is executed in the following five stages that are also illustrated in figure 2:

- Stage 1 : Data pre-processing.
- Stage 2 : GATE Monte Carlo simulation of β^+ -emitting isotope production.
- Stage 3 : Activity map modeling.

- Stage 4 : GATE Monte Carlo simulation of β^+ emission and PET acquisition.
- Stage 5 : CASTOR image reconstruction.

The motivation for the design of the framework, consisting of separated data processing, simulation and reconstruction stages is the optimization of the simulation time (Stage 2 and 4), and flexibility to study simulation and reconstruction parameters (Stage 2, 4 and 5). Dividing the framework execution into stages allows also for time efficient, parallelised use of available computing resources. Moreover, the full simulations of primary and secondary particles are overall more time consuming with respect to simulations of a limited number of physics interactions, transport of given particle type, and scoring of selected quantities. Therefore, the proposed framework has a modular construction where each stage has a separate settings, for example a separate physics list in MC simulations to optimize the processing time.

For instance, the modification of the PET scanner geometry requires repeating of the Stage 4 and 5 only, while modification of the reconstruction parameters only Stage 5. Also, Stage 2 can be replaced by other tools or user defined isotope distributions, when the β^+ -emitters production map is obtained with a Monte Carlo simulation code different than GATE, and, once data format compatibility is warranted, the new map can be used as an input of Stage 4.

2.1.1. Stage 1: data pre-processing

The goal of Stage 1 is to prepare all the necessary input data for the subsequent stages. The data required for the simulations in Stage 2 are patient/phantom geometry, treatment plan as well as facility-specific beam model data, exported from the treatments planning system (TPS) and stored in DICOM format (Pianykh 2010). The phantom/patient files include: CT scan of the phantom/patient (DICOMCT), treatment plan (DICOM RT) and structures used during treatment planning (DICOM RS). Furthermore, the proton beam model and CT calibration must be converted from the clinical TPS format to the GATE-specific format and validated as, for example, was performed in previous work (Grevillot *et al* 2011, Winterhalter *et al* 2018, 2020a, Gajewski *et al* 2021).

The CT image, converted from the DICOM CT images, and provided to the MC is required by GATE in one of the acceptable formats, from which we have chosen the MetaImage format (extension *.mhd or *.mha) (Johnson *et al* 2015). Also, it is good practice to reduce the CT image boundaries to the object of interest only (phantom or patient), by cropping the unnecessary space covered by air in the CT image. This allows greater flexibility in detector positioning while avoiding overlapping volumes of the phantom (or patient) and the scanner in GATE. For the image cropping, we used the *external* structure saved in a DICOM RS file, which contains the patient body and immobilizing elements. Moreover, to perform the simulations more efficiently, it is recommended to resample the CT image to the reconstucted PET resolution. In our study we used a $2.5 \times 2.5 \times 2.5 \text{ mm}^3$ voxel size. This voxel size is conditioned by the simulation statistics and PET scanner properties.

The DICOM RT file contains all the information concerning the treatment plan, including definition of one or more irradiation fields, irradiation direction, gantry and couch rotations, isocenter position, etc. A dedicated Python script getPlan reads the required data from the DICOM file and converts it into an ASCII macro file in a format that can be read by GATE. Additionally, another ASCII file (CSV format) is prepared containing the information about each field of the treatment plan and consisting of the following columns: DeliveryNo— (integer) the order of delivery of the fields; RS—(boolean) determine the usage of the range shifter in the field; GantryAngle_deg—(floating-point number) angle of the gantry rotation for the field; CouchAngle_deg —(floating-point number) angle of the field; PencilBeams—(integer) number of individual pencil beams in the field; ProtonNo—(integer) number of protons to be delivered in the field.

All the necessary conversions and manipulations of the DICOM information are performed using the FREDtools package described in section 2.4. Some exemplary files are available at the project github webpage ProTheRaMon (Borys *et al* 2022).

Additional files required for the simulation are the beam model and the CT calibration. The beam model describes energy-dependent parameters of pencil beams in terms of polynomials. The parameters required are: mean energy and energy spread, parameters describing lateral propagation, separately for the X and Y directions, and a scaling factor converting the monitor units (MU) retrieved from the DICOM RT plan to the number of protons. In our study we defined the polynominals for the entire available nominal energy range from 70 to 220 MeV. Additionally, we have implemented a range shifter (RS) used clinically, which is used in the simulations if requested in the DICOM RT plan. In our study, the beam model was prepared and validated for two gantry rooms available at CCB by converting the experimentally validated beam model implemented in our previous work for the FRED MC code (Gajewski *et al* 2021) to the ProTheRaMon framework. The CT calibration files contain the information about composition and density for each Hounsfield unit (HU) present

isotope	half-time[s]
¹⁵ O	122.2
¹⁴ O	70.6
¹³ N	597.9
¹¹ C	1223.4
¹⁰ C	19.2
³⁰ P	149.9
³⁸ K	458.2

Table 1. Isotopes scored in the form of	
production maps.	

in the CT image. This allows GATE to convert each voxel of the CT image to specific material. The facilityspecific clinical CT calibration curve was prepared by means of stoichiometric calibration proposed by Schneider *et al* (1996). The CT calibration curve contains information on the composition and density of 421 materials.

The description of the PET scanner geometry must be provided in the form of a macro file in a format specific to GATE. This file will be used in Stage 4, for PET acquisition simulation, and in the Stage 5 for definition of the reconstruction geometry specific to CASTOR.

2.1.2. Stage 2: GATE simulation of β^+ -emitting isotope production

The main objective of Stage 2 is to perform the Monte Carlo simulation from which we obtain production maps of β^+ -emitting radioisotopes. In our framework, we score production maps for 7 isotopes with the longest half-life and the largest contribution to positron emission using the ProductionAndStopping actor, which are listed in the table 1. The generated isotope production maps are used in further steps for PET imaging simulation.

The set of required input files for this step includes: CT image in MetaImage format and field statistics in CSV format, as well as the treatment plan, CT calibration, beam model, all in GATE macro format. The simulation results of β^+ -emitter production maps are stored in ASCII files, with a simplified header containing basic information about the matrix size, resolution, voxel size and the number of voxels. The production maps have the same resolution as the input CT and output reconstructed PET image, which was 2.5 × 2.5 × 2.5 mm³ in our study, corresponding to the reconstructed resolution of the J-PET scanner (Moskal *et al* 2021b).

Simulations in Stage 2 exploit the QGSP BIC HP EMY physics list, that includes the Binary Cascade (QGSP_BIC) to model hadronic interactions, high precision library of neutron interactions (HP) below 20 MeV and electromagnetic physics component in option 3 (EMY) (Zacharatou Jarlskog and Paganetti 2008, Winterhalter et al 2020b). Use of the QGSP_BIC model does not directly calculate the cross sections, but instead samples the possible products of each nuclear interaction from the binary cascade. It is possible to instead use user defined cross sections, e.g. experimentally measured cross sections, through use of look-up tables as applied by Parodi et al (2007a), and McNamara et al (2022). Additional limiters and cuts are used to speed up the calculations while not affecting β^+ production accuracy in a patient or phantom. Because the *world* region, in which all the simulation is 'immersed', is filled with vacuum, the step limiter and production cuts for electrons, positron and gammas are set to 1 cm. The particles in the patient or phantom CT volumes are tracked with the step limiter (10 mm), larger than the voxel size, hence the maximum step size was limited by the voxel size and physics interactions. Because the simulations are devoted to production of β^+ -emitting isotopes, the production of secondary electrons, positrons and gammas is irrelevant and their production cuts are set to large value (5 m in our study). On the other hand, the production cut of secondary protons is set to 10 μ m. Additionally, in order to simulate properly a pencil beam scattering on the RS, the step limiter and the production cuts were set to 1 mm in the RS region.

Because of the relatively long time required to simulate the production maps, it is good practice to parallelize the calculations and use a computing cluster. In our study, all GATE simulations were performed on the Ziemowit cluster located at the Laboratory of Bioinformatics and Computational Biology at the Silesian University of Technology in Gliwice, Poland.

The GATE package includes tools, namely the GATE job splitter (gjs), to split a simulation into separate processes. Using the previously prepared macro files, files for individual processes are generated. It is important to notice that the gjs script does not change the number of protons in each sub-simulation but just divides the simulation time by the number of threads or machines. Therefore, it is necessary to know the number of threads or machines used for the simulation when generating the master script by the getPlan, before the simulation is split into many sub-simulations, in order to achieve the desired total number of primary protons.

Given the relatively long simulation time and limited available computational resources, it is possible to simulate only a fraction of the total number of protons defined for a field and stored in the field statistics CSV file. In our study, we chose to simulate 10% of protons after determining that this level of statistics resulted in acceptable variance of the isotope production in simulated production maps, while maintaining an acceptable simulation time. This approach requires multiplying the activity values during the simulation of β^+ emission by an appropriate value—in our case 10.

Within this work, Monte Carlo simulation of the production of β^+ emitters has been performed using GATE. However, the modular workflow allows other Monte Carlo codes and scoring techniques to be implemented, such as FLUKA or TOPAS (Augusto *et al* 2018, Onecha *et al* 2022). The GPU accelerated Monte Carlo code FRED (Gajewski *et al* 2021) included the implementation of the scoring of the production of isotopes within a patient during proton therapy (McNamara *et al* 2022). The work validated the predicted isotope production scored in FRED against the isotope productions predicted in this work by GATE. No statistically significant deviations were found, providing confidence that the isotope production from FRED may be reliably used. The GPU accelerated implementation allows for simulation of the isotope production maps for the isotopes in table 1 within a fraction of the time necessary for the GATE simulations, with an average time for simulation of a field being 2.4 min. The integration of FRED into the ProTheRaMon workflow reduces the computational requirements of future simulation studies. However, the development of this was concurrent to the development of the ProTheRaMon framework.

Additional to the production maps of the β^+ -emitting radioisotopes, it is possible to score dose and LET maps at this stage, depending on the settings for the getPlan script, with the appropriate physics settings adjusted to these needs.

2.1.3. Stage 3: activity map modeling

The aim of this stage is to aggregate and post-process the production maps generated in the previous step into a single activity map, taking into account the imaging protocol.

The imaging protocol takes into consideration the times necessary for the irradiation and patient setup in the PET detector. Depending on the specific needs of the user, different time structures may be implemented for the treatment as well as in-beam, in-room or off-line imaging protocols. For the purpose of our studies, we chose 90 s delay before the start of irradiation of each field, except the first one. This estimates the duration of exposure (plan delivery) as taking 60 and 30 s is estimated as the time necessary to set up the gantry and beam for the next field. After the last field, the imaging was done immediately, without any additional delay. The time structure was chosen to reflect a clinically relevant protocol and is necessary to take into account the half-lives of individual isotopes when final production maps are generated. More realistic beam times, using log files from the real treatment planning, can easily be integrated if needed.

All activity maps, produced by each sub-simulation in Stage 2, are aggregated for each isotope separately and saved as single files in an Interfile format (extension *.h33/*.i33), ready to be used in the next stage of the workflow.

2.1.4. Stage 4: GATE Monte Carlo simulation of β^+ emission and PET acquisition

The aim of Stage 4 is to perform simulations of the β^+ emission in a patient or phantom, propagation of the 511 keV annihilation photons and signal acquisition in a predefined PET geometry.

The input data needed to prepare the macro files for GATE are: a CT file in MetaImage format, a treatment plan in DICOM RT format, with the location of the isocenter, and the activity maps for the seven isotopes listed in section 2.1.2 in the Interfile format. Also, the description of the PET scanner geometry, in the GATE macro file format, has to be prepared. Based on this data, a Python script called petPlan generates GATE macro files needed for simulations. As in the (isotope production simulation) Stage 2, this is the basis for using the GATE *gjs* script to split the simulation into multiple, shorter, simulations that can be executed in parallel on a multi-core computer or computer. To allow different scanner geometries to be tested, the geometry name is one of the script parameters.

Simulations in Stage 4 are performed using the *emlivermore_polar* list of physics processes (Kowalski *et al* 2018) with all the default production cuts, but the user can substitute any other physics model. The list includes the interactions of electrons and photons with matter down to 10 eV, interpolated based on the Livermore library, which is sufficient for the PET response investigation. Physical processes include the photoelectric effect, Compton scattering, gamma conversion, Rayleigh scattering, ionisation and bremsstrahlung.

The simulation results are saved at this stage in ROOT (extension *.root) files (Antcheva *et al* 2009), where information about the individual events occurring in the detector's sensitive volume is stored. The coincidence data are saved for image reconstruction and, optionally, singles and hits may be stored for further analysis.

It must be kept in mind that the voxelised activity source has a different origin, with the image corner placed starting at [0, 0, 0] mm (x, y, z), than the voxelised phantom image (CT or phantom), which are placed in the

center of their geometrical center. In order to align the activity image with a phantom volume, a proper translation is needed. Additionally, to keep the planning target volume (PTV) center (isocenter) in the center of the field of view (FOV) of the scanner, we perform an extra translation of the voxelised source in the *z* axis.

2.1.5. Stage 5: image reconstruction

The last stage of the workflow collects the data from the second simulation step and aims to produce the final PET images. For the image reconstruction step, the CASTOR framework (version 3.1) (Merlin *et al* 2018) was adapted. However, any software can be used here, depending on the needs. We have used a modified version of CASTOR that allows to produce sensitivity maps that include the effect of patient (or phantom) attenuation. This non-standard, for CASTOR, combination of sensitivity map with attenuation map in one set is due to the non-standard (multi-layer and non-cylindrical) geometry of the scanner used in our research. In CASTOR by default this information is separated and delivered as separate files to the reconstructor.

The geometrical sensitivity is specific for a given PET scanner, while the effect of attenuation is computed for an individual phantom/patient based on the information from a CT scan. CT images are converted to the attenuation map with the bi-linear function described in Carney *et al* (2006). The sensitivity map was calculated by performing a back-projection of counts from simulated data, weighted by the attenuation of each line of response (LOR) due to the patient. The simulations for the sensitivity map are based on the acquisition of a cylindrical air phantom uniformly distributed with the β^+ activity (back-to-back sources) covering the whole field-of-view (FOV), resulting in high-statistics list-mode data.

The input files for this stage are: the ROOT files with the coincidence data, the attenuation map stored in the Interfile standard, and the geometry description file (GATE macro). One of the reasons to choose the CASTOR image reconstruction software was the fact that it can easily use, with a simple conversion, the GATE geometry description macro file. The first two preprocessing steps at this stage are: the GATE geometry to CASTOR format conversion and the simulation to generate the coincidence data from a homogeneous source for each PET scanner geometry used.

The CASTOR package uses the maximum likelihood expectation maximization (ML-EM) iterative reconstruction algorithm and can include information about time-of-flight (TOF), which can be set up at the ROOT to CASTOR conversion phase.

The reconstructed images are saved separately for each reconstruction iteration in Interfile format, without any information about the coordinate system, but the resulting PET images are in the frame of reference, voxel size and image size as the patient/phantom CT volume.

2.2. Implementation of the framework on a computational cluster

Calculations were performed on the Ziemowit computer cluster in the Laboratory of Bioinformatics and Computational Biology at the Silesian University of Technology. One of the crucial components of each computer cluster is the queue management system that allows for the management of the users' jobs under a multicomputer and multiuser environment, and allows to assign proper resources for the users' tasks. One of the most popular is simple linux utility for resource management (SLURM) (Yoo *et al* 2003). SLURM ensures that the framework built using this queueing system will have the widest range of usability. For example, our cluster contains two types of computational resources: 100 servers based on older x5650 processors (2 CPUs in Westmere architecture, 6 cores each) with 36–120 GB RAM memory and 28 servers based on newer E5-2660v3 CPUs (Haswell architecture with AVX2 instructions set, 10 cores each) with 256 GB RAM. The choice of the computational node type is done by the selection of resources in the queue system.

From the cluster user's perspective, there is a need to create an additional shell script with the commands needed for SLURM that describe the most important parameters of the job, which are the number of threads, the partition name, execution time etc.

2.3. Example application of the framework for CCB Krakow proton facility and J-PET scanner

An example patient was used to demonstrate how the framework works. In this example we present the individual results for Stage 2, in the form of maps of the production of β^+ -emitters for the entire volume, as well as divided into individual fields and isotopes listed in table 1. Moreover, we present the effect of the applied imaging protocol and its impact on the modeled activity. The final result of the framework is a reconstructed PET image, which may be used to monitor the beam range in proton therapy.

Patient example. To present the effect of each stage, one exemplary treatment plan of a patient treated at CCB was used. The patient, a 49 years old woman, had been diagnosed with a skull-base chondrosarcoma and the treatment plan was prepared in the Eclipse 13.6 (Varian) TPS with 70 Gy (RBE) treatment dose delivered uniformly to the PTV with four irradiation directions (fields) in 35 fractions of 2 Gy (RBE) each. The

DICOM RT treatment plan, DICOM CT images and DICOM RS structure set were exported from the TPS and used as the input to the ProTheRaMon workflow.

J-PET scanner. In this work we are presenting results for one of the J-PET scanner designs, which was designed at the Jagiellonian University (Krakow, Poland) (Moskal *et al* 2014, 2016, Niedźwiecki *et al* 2017). The J-PET detector uses plastic strips (Kapłon 2020, Kapłon and Moskal 2021) with photomultipliers at their ends to optimize production cost (Moskal and Stępień 2020, Alavi *et al* 2022). For this specific work, a dual-layer cylindrical scanner design was used consisting of 48 modules. The design with the double number of detector layers aims to improve the PET signal quality (Moskal *et al* 2019, 2021b).

2.4. FREDtools for data analysis and visualization

FREDtools (Gajewski 2022) is an open-source collection of Python functions and classes for manipulation and analysis of scalar or vector images of dimension up to 5D. The images read by FREDtools are SimpleITK (Lowekamp *et al* 2013, Yaniv *et al* 2018) objects and, additionally to new implementations, all the functionalities of the Insight Toolkit (ITK) (Yoo *et al* 2002, Mccormick *et al* 2014, Johnson *et al* 2015) are available. The basic methods have been developed for analysis of the images in MetaImage format (*.mha or *.mhd) produced by most of the Monte Carlo frameworks, but can also be applied for images in other formats, e.g. DICOM.

The FREDtools methods provide the functionalities of image manipulation, such as: resampling, affine transformations (including beam-eye view rotation), mapping DICOM structures to 3D image masks, as well as image analysis, such as: dose-volume histogram (DVH) analysis, multithread gamma index analysis, Bragg peak analysis (including fitting methods of curves proposed by Bortfeld (Bortfeld 1997)) and retrieving any lower-dimensional image (e.g. a profile or a slice from a 3D image) with interpolation. All the functionalities implemented in FREDtools have been validated against the corresponding results evaluated in a commercial TPS (Eclipse v. 13.6 (Varian)).

FREDtools has been used in the ProTheRaMon framework in multiple stages. Primarily, it was used to prepare the input data in Stage 1. Among others tasks, the CT resampling, preparing the final directory structure and treatment plan processing were done using the tools. FREDtools was used in each processing step where it was necessary to use the treatment plan information or retrieve patient data to prepare the visualization of the results.

3. Results

In this section we will present the most important results from all stages of the ProTheRaMon framework. Stage 1 focuses mainly on the preparation of input data for the other stages. It results in appropriate CT image format, appropriate treatment plan format for use with GATE, beam model, CT calibration and scanner geometry description. These are all key data for the next Stages. However, they are not suitable for presentation in graphical form or any kind of summary or plot drawing, therefore the presentation of results starts from Stage 2.

Stage 2 output. The main output of interest from Stage 2 are the production maps, but other output, depending on the settings, are available, such as linear energy transfer (LET) and dose. An example of the β^+ -emitters production accompanied with the dose (left) and with modeled activity (right) maps is shown in figure 3. The dose map can be considered here as a reference distribution. The two next figures give more detailed information about the β^+ -emitter production. Each field that is planned for the therapy is simulated separately and the cumulative map (for all isotopes) for each field is shown in figure 4 (top row). Even more detailed information, presented for the last field, is shown in figure 5 for each isotope separately. Under each of the subplots the colorbar presents the level of the intensity in terms of radionuclei per voxel, representing the amount of radionuclei for each field or for each isotope produced and scored during the simulation.

On all images presented here and further, the PTV and Brain Stem regions are displayed, the second represents an organ at risk.

Simulations of the presented in the paper patient, that has four fields in his treatment plan, took an average of 160 min (209; 130; 118 and 181 [min] each field) using newer nodes with Haswell 400 CPU cores.

Stage 3 output. The activity modeling stage takes into account the imaging protocol, its type (in-beam, inroom etc) and the delay times between the irradiated fields.

In our simulations we have tested a scenario in which imaging was performed after the last irradiated field. This last field had the greatest impact on the production of β^+ -emitters and indirectly on the activity distribution, as it decays the least compared to the previously deposited fields.

This effect can be seen in figure 3—where the β^+ -emitters production map (image in the middle) is compared with the modeled activity from all fields (image on the right). The activity distribution in figure 3 largely corresponds to that of Field 4 shown in figure 4.

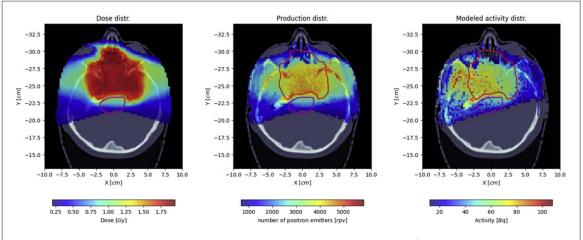
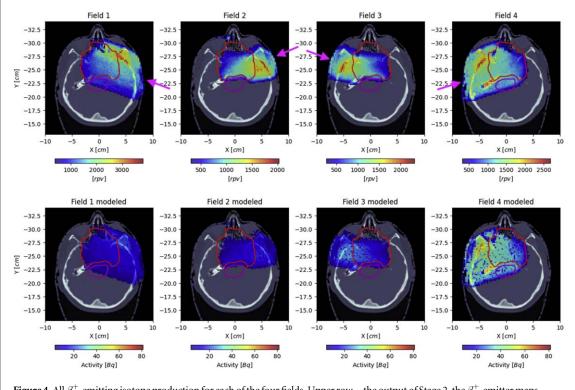
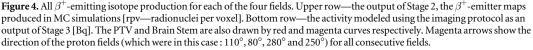


Figure 3. Left image: dose image acquired at Stage 2 [Gy]. Middle image: production map for all β^+ -emitting isotopes and all fields direct output of Stage 2 [rpv—radionuclei per voxel]. Right image: activity map after activity modeling at Stage 3 for all isotopes and all fields [Bq]. PTV and Brain Stem are drawn by red and magenta curves respectively.





All Fields separately are presented in figure 4 as β^+ -emitter production maps (upper row) and as a modeled activity distribution after Stage 3 (bottom row). Here we can compare and clearly see what is the contribution of the β^+ -emitter production from each Field to the final activity. In the bottom row, the level of activity increases for the consecutive fields, with the greatest values for the Field 4.

Stage 4 output. The second MC simulation of β^+ emission and PET acquisition gives us the hits in the detector volume and coincidences lists which are stored in a *ROOT* file format that is not feasible to present in a graphical way. Both types of computing nodes were used for this simulations and it took about 90 [min] using 200 CPU cores with older Westmere nodes or about 30 [min] with Haswell nodes.

Stage 5 output. The last figure in this section, figure 6, presents the output of the CASTOR image reconstruction software as a final reconstructed image in three planes (axial, coronal and sagittal). The images

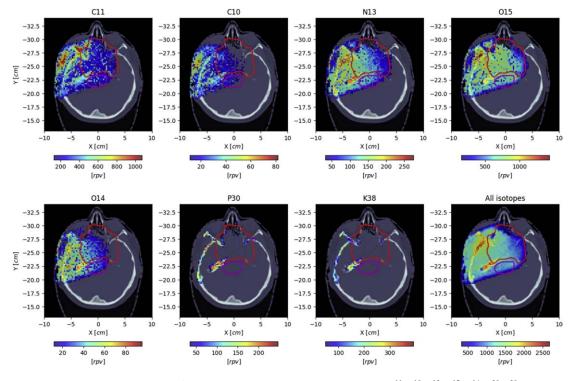
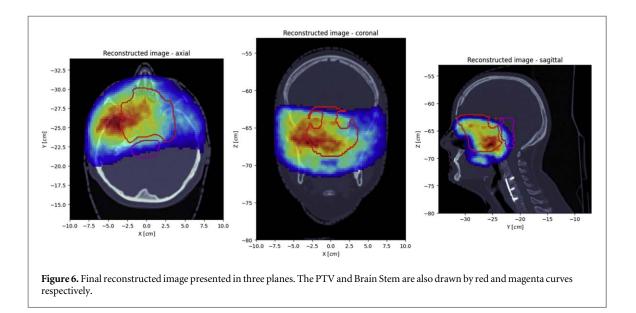


Figure 5. The last applied field (Field 4) β^+ -emitter production for all isotopes separately (for ¹¹C, ¹⁰C, ¹³N, ¹⁵O, ¹⁴O, ³⁰P, ³⁸K) and all isotopes as one map (All). The PTV and Brain Stem are also drawn by red and magenta curves respectively.



presented were postprocessed by using a Gaussian smoothing filter with sigma = 2 voxels, in order to remove statistical noise.

4. Discussion

The ProTheRaMon framework provides the user with the ability to perform MC simulations of proton therapy treatment and secondary radiation detection for range monitoring using PET. However, instead of creating a single-step full simulation that provides a unique output for a given set of input parameters, we have decided to design our framework as a set of separate and independent stages. This has allowed us to prepare a more flexible and versatile tool. This design also allows the user to decide on a range of relevant parameters at each step. In the second stage, we have decided to score the seven radioisotopes listed in table 1 and also shown as an example of production maps in figure 5. However, if any of these are not of interest in the specific study, they can be

switched off or other isotopes can be added. Also, a variety of different PET imaging protocols that differ depending on the capabilities of specific proton centers and specific scanner designs can be proposed. Off-line, in-room or in-beam (if time information is incorporated) imaging protocols may be implemented to model the PET signal decay in Stage 3. The protocol we have used is presented in figure 4, where the weakening (by decay) of activity from earlier fields is apparent. In our study we have not implemented a biological washout model, only radioactive decay with the assumption that all β^+ -emitter radionuclides are produced at the beginning of the field irradiation. The effect of biological washout can be added by the user in Stage 3 according their needs. An example of the method that can be additionally and easily implemented by the user into the proposed framework can be found in (Mizuno *et al* 2003).

Two stages of the ProTheRaMon workflow, i.e. Stage 2 and Stage 4, exploit MC simulations that are used for different purposes. In Stage 2 (see 2.1.2), simulations of a treatment plan are performed in order to obtain distributions of β^+ -emitting isotopes produced during the irradiation. The transport of photons, resulting from positron annihilations, out of the patient body and the photon energy deposition in the PET detectors is simulated in Stage 4. Therefore, the two stages require different setting of physics options which can be set individually for each stage of MC simulation. The production of β^+ -emitting isotopes during the proton irradiation requires physics models incorporating hadronic and neutron interactions, while the detection of γ particles in the PET scanner detector can be done with electromagnetic models only. The exclusion of unnecessary physics processes has the effect of decreasing the simulation time.

Stages 4 and 5 require the definition of a PET scanner geometry, which can be optimized and changed independently from the β^+ activity simulation performed in Stage 2. By changing one macro file with the geometry description the user can easily implement many different detector designs.

The treatment and PET imaging parameters may be varied and investigated in a more time-efficient manner by using a staged architecture implemented in ProTheRaMon. For example, at the activity production simulation stage (Stage 2), we simulated 10% out of the total number of particles used in the treatment. This was motivated by the need to optimize the simulation time with limited computational resources. For our purposes, this level of statistics represented a satisfactory compromise between the variance of the isotope production and the simulation time. The other settings that will significantly affect calculations times are the resolution of the voxelized phantom and source as well as the reconstructed image resolution. Such parameters should be adjusted to the available computational resources and the performance of the PET detector.

The uncertainty of β^+ activity modeling is essential when validating the performance of a new PET detector for proton beam range monitoring. The uncertainty of Monte Carlo simulations is intrinsically related to the uncertainty of the nuclear model used (Sihver *et al* 2012). The experimental validation of Geant4 and other MC codes in terms of the yields of β^+ -emitters during proton therapy has been the subject of extensive research (for a review of the subject see (Kraan 2015)) and has resulted in the tuning of the relevant cross section data. In simulation of PET activity maps in patients, further inaccuracies may result from modeling of the clinical proton beam, which for our example has been recently addressed in Gajewski *et al* (2021), while the uncertainty related to the modeling of patient tissue composition based on CT scans has been addressed, e.g. by Paganetti (2012). Overall, these uncertainties have a minor impact on the application of the ProTheRaMon framework to simulation studies aiming at PET detector and protocol designs because these studies are based on a relative comparison of the effectiveness of different approaches.

5. Conclusions

 $\label{eq:proTheRaMon} ProTheRaMon was developed to simplify simulation studies needed to design, test and validate PET scanner geometries and protocols dedicated to proton therapy range monitoring in a clinical environment.$ $<math display="block">ProTheRaMon is an automated multi-stage protocol designed for the following tasks: to simulate the production of β^+-emitting radioisotopes originating from nuclear reactions of the primary proton beam with the patient during proton therapy, to simulate the emission and transport of the gamma rays originating from electron-positron annihilation to the PET detector, and to reconstruct the PET image. ProTheRaMon offers complete processing of proton therapy treatment plans and patient CT geometries in the coordinate system of the treatment room and PET scanner, taking into account the activity decay related to the PET imaging protocol specific to the proton therapy center.$

So far, any institute wishing to perform Monte Carlo simulations of range monitoring in proton therapy using PET must set up their own simulation workflow. The ProTheRaMon framework can simplify this task as it is available to the community via the github repository ProTheRaMon (Borys *et al* 2022). Moreover, the modular design of the ProTheRaMon framework allows for easy adjustment and expansion of its functionality for other proton therapy range monitoring applications, where two basic MC simulation steps, i.e. β^+ -emitters production and PET imaging, have to be performed. ProTheRaMon was designed to execute the most computationally intensive stages on a computational cluster. Due to the time requirements for performing the simulations yielding β^+ -emitter distributions and β^+ emission simulations, the use of a computational cluster is essential to perform the study on a large group of patient treatment plans.

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ORCID iDs

Damian Borys https://orcid.org/0000-0003-0229-2601 Jakub Baran (1) https://orcid.org/0000-0002-4946-3837 Karol Brzeziński i https://orcid.org/0000-0002-9795-5158 Jan Gajewski @ https://orcid.org/0000-0002-7416-5145 Eryk Czerwiński D https://orcid.org/0000-0002-5076-3175 Kamil Dulski () https://orcid.org/0000-0002-4093-8162 Kavya V Eliyan (b) https://orcid.org/0000-0003-1636-3706 Aleksander Gajos lttps://orcid.org/0000-0002-7536-6459 Łukasz Kapłon () https://orcid.org/0000-0002-2203-0425 Konrad Klimaszewski D https://orcid.org/0000-0003-0741-5922 Paweł Konieczka https://orcid.org/0000-0001-6859-0180 Renata Kopeć (1) https://orcid.org/0000-0002-0919-9859 Wojciech Krzemień b https://orcid.org/0000-0002-9546-358X Paweł Olko (10 https://orcid.org/0000-0001-5554-8178 Elena Perez del Rio D https://orcid.org/0000-0001-7761-4084 Roman Y Shopa ^(b) https://orcid.org/0000-0002-1089-5050 Tomasz Skóra ⁽¹⁾ https://orcid.org/0000-0001-6322-0615 Magdalena Skurzok https://orcid.org/0000-0002-4794-5154 Paulina Stasica https://orcid.org/0000-0002-7109-1136 Ewa Ł Stępień () https://orcid.org/0000-0003-3589-1715 Keyvan Tayefi https://orcid.org/0000-0001-8764-1588 Damien C Weber https://orcid.org/0000-0003-1166-8236 Wojciech Wiślicki D https://orcid.org/0000-0001-5765-6308 Paweł Moskal D https://orcid.org/0000-0002-4229-3548 Antoni Rucinski l https://orcid.org/0000-0002-5815-4606

References

Agostinelli S et al 2003 Geant4-a simulation toolkit Nucl. Instrum. Methods Phys. Res. A 506 250–303

Alavi A, Werner T J, Stępień E and Moskal P 2022 Unparalleled and revolutionary impact of PET imaging on research and day to day practice of medicine *Bio-Algorithms Med-Syst.* 17 203–12

Allison J et al 2016 Recent developments in geant4 Nucl. Instrum. Methods Phys. Res. A 835 186–225

Antcheva I *et al* 2009 ROOT—a C++ framework for petabyte data storage, statistical analysis and visualization *Comput. Phys. Commun.* 180 2499–512

Augusto R et al 2018 An overview of recent developments in fluka pet tools Phys. Med. 54 189-99

Bauer J, Unholtz D, Sommerer F, Kurz C, Haberer T, Herfarth K, Welzel T, Combs S E, Debus J and Parodi K 2013 Implementation and initial clinical experience of offline PET/CT-based verification of scanned carbon ion treatment *Radiath. Oncol.* **107** 218–26

Bortfeld T 1997 An analytical approximation of the Bragg curve for therapeutic proton beams. *Med. Phys.* **24** 2024–33 Borys D, Baran J, Brzezinski K, Gajewski J and Rucinski A 2022 ProTheRaMon repository (https://github.com/borysd/ProTheRaMon) Carney J P, Townsend D W, Rappoport V and Bendriem B 2006 Method for transforming ct images for attenuation correction in pet/ct imaging *Med. Phys.* 33 976–83

Enghardt W, Parodi K, Crespo P, Fiedler F, Pawelke J, Pönisch F and Ponisch F 2004 Dose quantification from in-beam positron emission tomography *Radiother. Oncol.* **73** S96–8

Ferrero V *et al* 2018 Online proton therapy monitoring: clinical test of a Silicon-photodetector-based in-beam PET *Sci. Rep.* **8** 4100 Gajewski J 2022 FREDtools—radiotherapy image analysis in python. (www.fredtools.ifj.edu.pl) (https://fredtools.ifj.edu.pl/)

Gajewski J *et al* 2021 Commissioning of GPU-accelerated monte carlo code fred for clinical applications in proton therapy *Front. Phys.* 8 1–18

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 5203–19

Hishikawa Y, Kagawa K, Murakami M, Sakai H, Akagi T and Abe M 2002 Usefulness of positron-emission tomographic images after proton therapy *Int. J. Radiat. Oncol. Biol. Phys.* 53 1388–91

Jakub B, Jan G, Monika P N, Pawel M and Antonion R 2019 Studies of j-PET detector to monitor range uncertainty in proton therapy *IEEE* Nuclear Science Symposium and Medical Imaging Conference (NSS/MIC) (Manchester, UK, 26 October–02 November 2019) (IEEE) pp 1–4

Jan S *et al* 2011 GATE V6: a major enhancement of the GATE simulation platform enabling modelling of CT and radiotherapy *Phys. Med. Biol.* 56 881–901

Jan S et al 2004 GATE: a simulation toolkit for PET and SPECT Phys. Med. Biol. 49 4543-61

Johnson H J, McCormick M M, Ibáñez L and Consortium I S 2015 The ITK software guide: design and functionality *ITK Software Guide: ITK* 4.7 Kitware (https://books.google.pl/books?id=SMwdrgEACAAJ)

Jäkel O, Kraft G and Karger C P 2022 The history of ion beam therapy in germany Z. Med. Phys. 32 6–22

Kapłon L 2020 Technical attenuation length measurement of plastic scintillator strips for the total-body J-PET scanner *IEEE Trans. Nucl. Sci.* 67 2286–9

Kapłon L and Moskal G 2021 Blue-emitting polystyrene scintillators for plastic scintillation dosimetry *Bio-Algorithms Med-Syst.* 17 191–7 Knopf A C and Lomax A 2013 *In vivo* proton range verification: a review *Phys. Med. Biol.* 58 131–60

Kowalski P *et al* 2018 Estimating the NEMA characteristics of the J-PET tomograph using the GATE package *Phys. Med. Biol.* **63** 165008 Kraan A C 2015 Range verification methods in particle therapy: underlying physics and monte carlo modeling *Front. Oncol.* **5** 150 Krimmer J, Dauvergne D and Létang Testa J M 2018 Prompt-gamma monitoring in hadrontherapy: a review *Nucl. Inst. and Meth. in Phys.*

Res. A 878 58–73

Kurosawa S *et al* 2012 Prompt gamma detection for range verification in proton therapy *Curr. Appl. Phys.* **12** 364–8 Lowekamp B C, Chen D T, Ibáñez L and Blezek D 2013 The design of simpleITK *Front. Neuroinform.* **7** 1–14 Mccormick M, Liu X, Jomier J, Marion C and Ibanez L 2014 Itk: enabling reproducible research and open science *Front. Neuroinform.* **8** 1–11 McNamara K *et al* 2022 GPU accelerated Monte Carlo scoring of positron emitting isotopes produced during proton therapy for PET

verification Submitted Phys. Med. Biol. Merlin T, Stute S, Benoit D, Bert J, Carlier T, Comtat C, Filipovic M, Lamare F and Visvikis D 2018 CASTOR: a generic data organization and

Merlin T, Stute S, Benoit D, Bert J, Carlier T, Comtat C, Filipovic M, Lamare F and Visvikis D 2018 CASToR: a generic data organization and processing code framework for multi-modal and multi-dimensional tomographic reconstruction *Phys. Med. Biol.* **63** 185005

Mizuno H *et al* 2003 Washout measurement of radioisotope implanted by radioactive beams in the rabbit *Phys. Med. Biol.* **48** 2269–81 Moskal P *et al* 2021a Positronium imaging with the novel multiphoton PET scanner *Sci. Adv.* 7 eabh4394

Moskal P et al 2019 Feasibility study of the positronium imaging with the j-PET tomograph Phys. Med. Biol. 64 055017

Moskal P et al 2021b Simulating NEMA characteristics of the modular total-body J-PET scanneran economic total-body PET from plastic scintillators *Phys. Med. Biol.* 66 175015

Moskal P *et al* 2014 Test of a single module of the j-PET scanner based on plastic scintillators *Nucl. Inst. Methods Phys. Res.* **764** 317–21 Moskal P *et al* 2016 Time resolution of the plastic scintillator strips with matrix photomultiplier readout for j-PET tomograph *Phys. Med. Biol.* **61** 2025–47

Moskal P and Stępień E 2020 Prospects and clinical perspectives of total-body PET imaging using plastic scintillators *PET Clin.* **15** 439–52 Muraro S, Battistoni G and Kraan A C 2020 Challenges in Monte Carlo simulations as clinical and research tool in particle therapy: a review *Front. Phys.* **8** 567800

Niedźwiecki S et al 2017 J-PET: a new technology for the whole-body PET imaging Acta Phys. Pol. B 48 1567-76

Nishio T, Miyatake A, Ogino T, Nakagawa K, Saijo N and Esumi H 2010 The development and clinical use of a beam ON-LINE PET system mounted on a rotating gantry port in proton therapy *Int. J. Radiat. Oncol. Biol. Phys.* **76** 277–86

Onecha V V, Galve P, Ibáñez P, Freijo C, Arias-Valcayo F, Sanchez-Parcerisa D, España S, Fraile L M and Udías J M 2022 Dictionary-based software for proton dose reconstruction and submilimetric range verification *Phys. Med. Biol.* 67 045002

Paganetti H 2012 Range uncertainties in proton therapy and the role of Monte Carlo simulations *Phys. Med. Biol.* 57 99–117 Parodi K 2020 Latest developments in *in vivo* imaging for proton therapy *Br. J. Radiol.* 93 20190787

Parodi K, Bortfeld T and Haberer T 2008 Comparison between in-beam and offline positron emission tomography imaging of proton and carbon ion therapeutic irradiation at synchrotron- and cyclotron-based facilities *Int. J. Radiat. Oncol. Biol. Phys.* 71 945–56
Parodi K and Enghardt W 2000 Potential application of PET in quality assurance of proton therapy *Phys. Med. Biol.* 45 N151–56

Parodi K, Ferrari A, Sommerer F and Paganetti H 2007a Clinical CT-based calculations of dose and positron emitter distributions in proton therapy using the FLUKA monte carlo code *Phys. Med. Biol.* **52** 3369–87

Parodi K *et al* 2007b Patient study of *in vivo* verification of beam delivery and range, using positron emission tomography and computed tomography imaging after proton therapy *Int. J. Radiat. Oncol., Biol., Phys.* **68** 920–34

Parodi K, Ponisch F and Enghardt W 2005 Experimental study on the feasibility of in-beam PET for accurate monitoring of proton therapy IEEE Trans. Nucl. Sci. 52 778–86

Pianykh O S 2010 Digital Imaging and Communications in Medicine (DICOM): A Practical Introduction and Survival Guide 2nd edn (Berlin, Heidelberg: Springer) (https://doi.org/10.1007/978-3-642-10850-1)

Rucinski A, Baran J, Garbacz M, Pawlik-Niedzwiecka M and Moskal P 2018 2018 IEEE Nuclear Science Symp. and Medical Imaging Conf., NSS/MIC 2018—Proc. pp 24–7

- Scheins J, Herzog H and Shah N 2011 Fully-3d pet image reconstruction using scanner-independent, adaptive projection data and highly rotation-symmetric voxel assemblies *IEEE Trans. Med. Imaging* **30** 879–92
- Schiavi A, Senzacqua M, Pioli S, Mairani A, Magro G, Molinelli S, Ciocca M, Battistoni G and Patera V 2017 Fred: a GPU-accelerated fast-Monte Carlo code for rapid treatment plan recalculation in ion beam therapy *Phys. Med. Biol.* 62 7482–504

Schneider U, Pedroni E and Lomax A 1996 The calibration of CT hounsfield units for radiotherapy treatment planning *Phys. Med. Biol.* 41 111–24

- Shakirin G, Braess H, Fiedler F, Kunath D, Laube K and Parodi K 2011 Implementation and workflow for PET monitoring of therapeutic ion irradiation: a comparison of in-beam, in-room, and off-line techniques *Phys. Med. Biol.* **56** 1281–98
- Sihver L, Lantz M, Bohlen T T, Mairani A, Cerutti A F and Ferrari A 2012 A comparison of total reaction cross section models used in FLUKA GEANT4 and PHITS IEEE Aerospace Conf. (Big Sky, MT, USA, 03–10 March 2012) (Picastaway, NJ: IEEE) pp 1–10
- Thielemans K, Tsoumpas C, Mustafovic S, Beisel T, Aguiar P, Dikaios N and Jacobson M W 2012 STIR: software for tomographic image reconstruction release 2 *Phys. Med. Biol.* 57 867–83

Wettenhovi V, Vauhkonen M and Kolehmainen V 2021 Omegaopen-source emission tomography software *Phys. Med. Biology* 66 065010 Wilson R R 1946 Radiological use of fast protons *Radiology* 47 487–91

Winterhalter C, Aitkenhead A, Oxley D, Richardson J, Weber D C, MacKay R I, Lomax A J and Safai S 2020a Pitfalls in the beam modelling process of monte carlo calculations for proton pencil beam scanning The *Br. J. Radiol.* **93** 20190919

- Winterhalter C *et al* 2018 Validating a Monte Carlo approach to absolute dose quality assurance for proton pencil beam scanning *Phys. Med. Biol.* 63 175001
- Winterhalter C et al 2020b Evaluation of GATE-RTion (GATE/Geant4) Monte Carlo simulation settings for proton pencil beam scanning quality assurance Med. Phys. 47 5817–28

Yaniv Z, Lowekamp B C, Johnson H J and Beare R 2018 SimpleITK image-analysis notebooks: a collaborative environment for education and reproducible research J. Digit. Imaging 31 290–303

Yoo A B, Jette M A and Grondona M 2003 Job Scheduling Strategies for Parallel Processing ed D Feitelson et al vol 2862 (Berlin, Heidelberg: Springer) pp 44–60

Yoo T S, Ackerman M J, Lorensen W E, Schroeder W, Chalana V, Aylward S, Metaxas D and Whitaker R 2002 Engineering and algorithm design for an image processing Api: a technical report on ITK-the Insight Toolkit *Stud. Health Technol. Inform.* **85** 586–92

Zacharatou Jarlskog C and Paganetti H 2008 Physics Settings for using the Geant4 toolkit in proton therapy *IEEE Trans. Nucl. Sci.* 55 1018–25