



CASTOR^[1]: A Generic Data Organization and Processing Code Framework for Multi-Modal and Multi-Dimensional Tomographic Reconstruction

Thibaut Merlin^{1,*}, Simon Stute^{2,*}, Didier Benoit¹, Julien Bert¹, Thomas Carlier³, Claude Comtat², Frédéric Lamare⁴, and Dimitris Visvikis¹





¹LaTIM - U1101 INSERM, *Brest, FRANCE*

Equally contributed

Motivations for an unified tomographic image reconstruction platform

Natural differences in the dataset acquisition / organization ...

... lead to algorithms optimized for specific applications, with potential drawbacks:

- Restricted use of methodologies however compatible with other sets of conditions
- Difference in duplication of implementations
- Possibly hardly tractable code development

• Modality (PET / SPECT / CT)

System configuration file

· Generic geometry description

Scanner elements geometric

·Scanner element dimension

CT

Rigid/Elastic

"Specific" classes

cartesian coordinates

Modality

System Geometry

Datafile

Event

Projector

- Purpose (tracer dynamics, gated dataset)
- Data format (list-mode/histogram)

Additional drives for an unified platform:

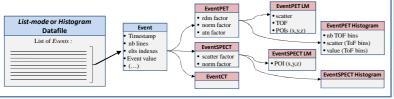
- Analogous components (projector, optimizer, ...) for PET/SPECT/CT tomographic reconstruction
- Re-emergence of iterative reconstruction in CT^[2]

Proposed solution

- · Unified and generic data organization and processing code framework for multi-modal and multidimensional tomographic reconstruction.
- Analogous data organization for list-mode/histogram
- Compromise between genericity and efficiency
- Focus on the modularity and extensibility of the

General data file Description

- Generic Event structure for all type of data (modality/data mode)
- Reconstruction framework is Event-based
- PET/SPECT Event can be of "list-mode" type (i.e. single detected event) or histogram type (i.e. content of a histogram bin).
- Events must contain mandatory fields: timestamp, event value, geo localizer indices and optionally: TOF, scatter/random rate, norm factors, etc... (optional fields)



Implementation

- ullet C++ open-source platform (CASToR^[1])
- 3 levels of temporal dimension :
 - Dynamic frames
 - Respiratory gating
 - Cardiac gating
- Events loop step equal to the number of subsets (Fig. 1)
- Subsets balance respected for both list-mode / histogrammed data.

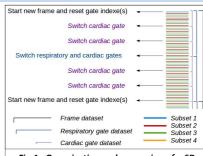


Fig.1: Organization and processing of a 6D dataset in an OSEM framework

CASToR^[1] Iterative framework



- Image dimensions Number of iterations/subsets
- Selected reconstruction

Geometry generation

Components Initializations

Sensitivity (list-mode) / Blank scan (CT) generation

Main loops of the iterative core algorithm

// Parallel loop for $e \leftarrow index$ start to nb events step index step

Get Event corresponding to e;
Check Image (forward) Deformation for e;
Compute/load the system matrix elements
associated to this event;

index step← number of subsets; Image Convolution (forward step);

- components mponents configuration
- parameters/files
- PET/SPECT/CT datafile • Modality • Mode (histogram/list)
- Number of events System name
- · Acquisition metadata
- Enabled/Disabled corrections Raw data

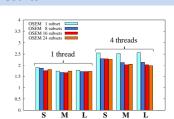


Fig.2: Speed-up ratio of CASTOR OSEM reconstructions compared to standard "2-calls to projector" implementation:

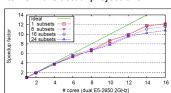
- List-mode (10M prompts): 88x88x63 (S), 175x175x125 (M), 350x350x250 (L) 4,2,1mm3 voxels - Histogram (45M events): 128x128x50 (S), 256x256x100 (M), 512x512x200 (L) 4,2,1mm³ voxels

Results

> Improved computational efficiency (ratios from 1.5 to 2.75) due to the separation of the computation of the system matrix elements from the actual projections

Fig. 3: Speed-up factor with respect to the number of threads for a 3D reconstruction

> Computing efficiency presented limited dependence on the number of subsets



Conclusions & Perspectives

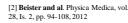
- > Proposed architecture handles histogram/list-mode multi-modal and multi-dimensional iterative reconstruction
- Simplified integration of new reconstruction features / methodologies, with limited duplication of implementations
- > Good computing performances in its parallel execution (Unique call to the projector by event allowed a 1.5 to 2.75 increase speedup ratios on the test platform, as well as moderate cost of
- ➤ Proposed implementation will be soon available through an open-source software^[1]

Compute the forward projection; // Up to Compute the values to be back-projected: Compute the backward projection; // Up Spatial regularization Synchronize all data Image (backward) Convolution Image (backward) Deformation Image deformation Update image according to the optimization algorithm Dynamic Model Generic classes Ex: List-mode 4D reconstruction Ex: Histogram reconstruction using different without (left) and with (right) optimization algorithms temporal regularization Landweber[4

> Separation of Generic and Specific classes favors limited duplication of implementations

> Generic implementation of each reconstruction component allows easy integration of new methodologies

References



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