



A Fast-Converging Maximum A Posteriori Algorithm for 3D Image Reconstruction Using Multi-view Compton Data

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Outline

- ① Compton Camera as a 3D imager
- ② Temporal imaging CeBr₃ Compton camera
- ③ 3D image reconstruction: Maximum A Posteriori (MAP) algorithm
- ④ Numerical results
- ⑤ Application on Dream Scanner project
- ⑥ Conclusions and perspectives



3D imaging: A promising concept for waste characterization

To optimize waste storage packaging and costs a 3D map of major isotopes contamination is the best solution

But today this is not realistic economically as 3D scanning systems, based on collimated Germanium are both expensive and slow.

The cost of a 3D scan is a function of:

- The number of views needed for 3D reconstruction. This number depends on:
 - Camera Field of View FOV (nb of views to 2D image an object)
 - How many photons are needed by voxel/by view
- The sensitivity of the camera (how long should we wait to reach a given detection limit?)
- How fast the reconstruction algorithms converges with a limited number of photons & views

Why are Compton cameras uniquely suited for 3D scans?



Compton camera has a wide Field of View ($90^\circ \times 90^\circ$ here)

→ A limited number of views should fully cover an object

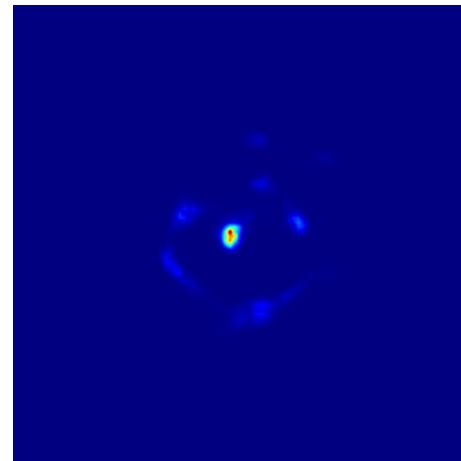
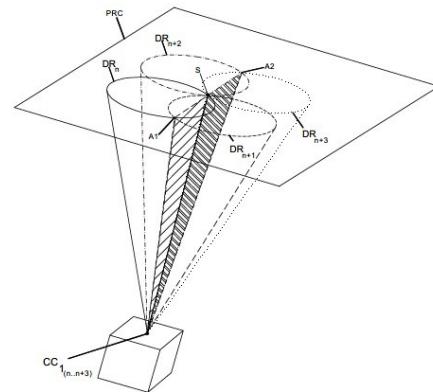
Compton events generate large thick cones : « fuzzy objects »

→ In 2D reconstructions High statistics are needed to avoid false positive

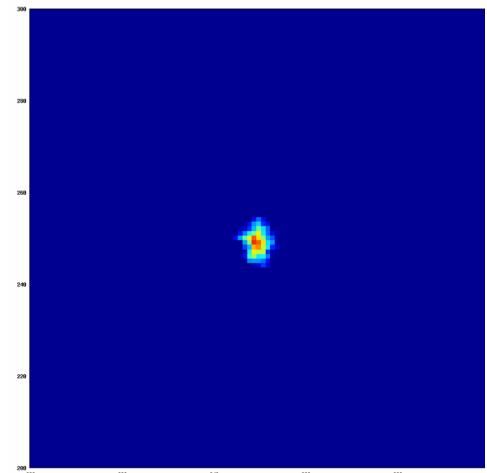
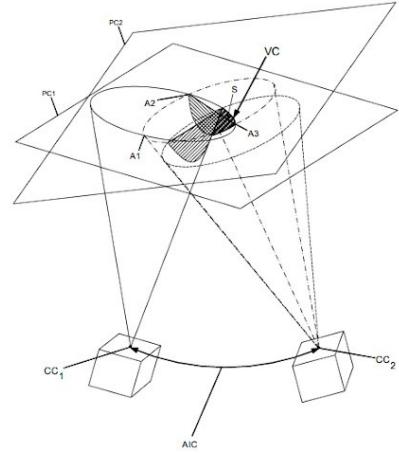
→ In 3D the volume of intersection between cone shrinks by an order of magnitude

Compton image reconstruction works much better in 3D than in 2D

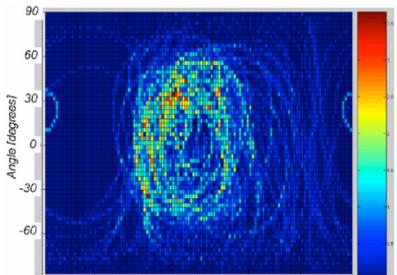
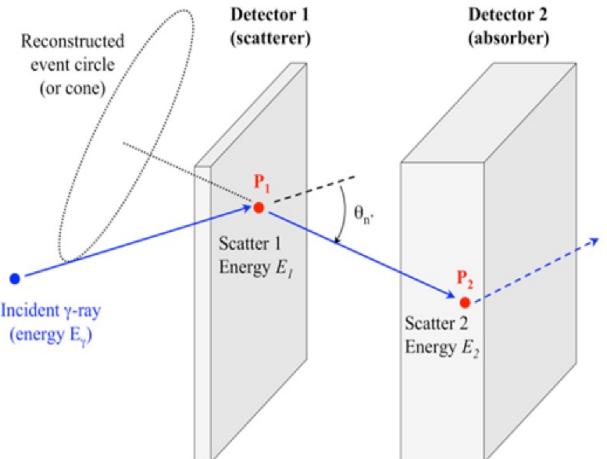
One view



2 or more views



Temporal imaging Compton camera CeBr₃: A technology well suited for 3D imaging



Low noise level

- Two fast CeBr₃ scintillating crystals plates
- Low natural background
- <500 ps Coincidence veto on Compton events

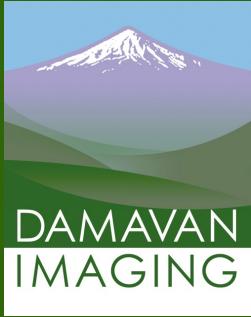
High angular resolution

- <8° vs < 20° for CZT cameras

Large Field of View

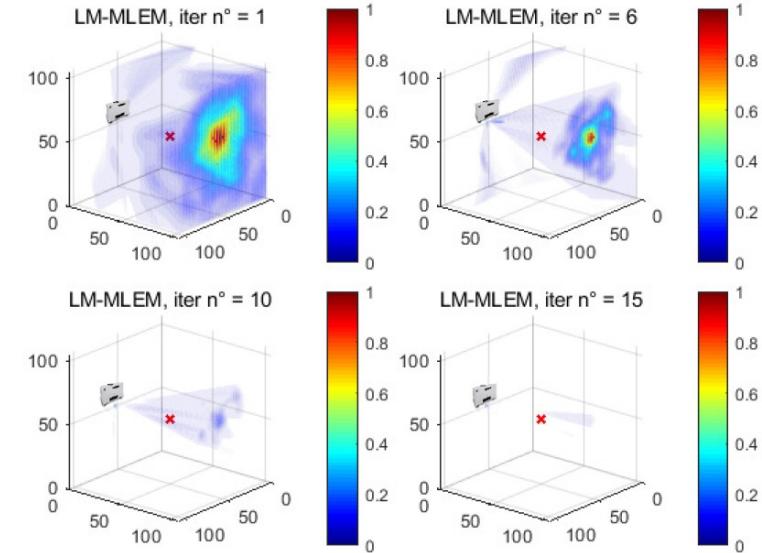
- 90° x 90°

3D Compton image reconstruction: *Problem statement*



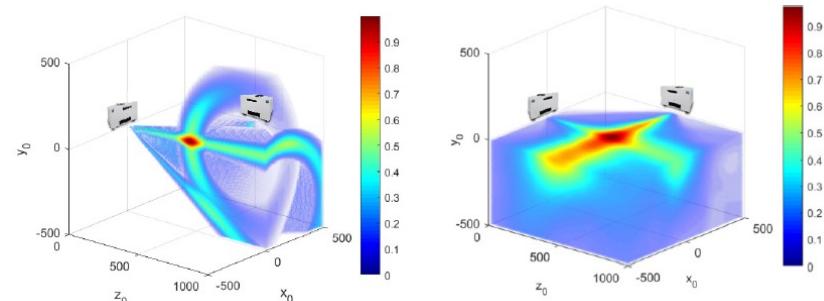
Single-view reconstruction

- 2 approaches: analytical, iterative
- 2 classes of models: deterministic, probabilistic
- Reconstruction is marginal if parallax is sufficient (large detector, near field)
- **3D reconstruction Fails** if parallax is low (compact camera, far field)

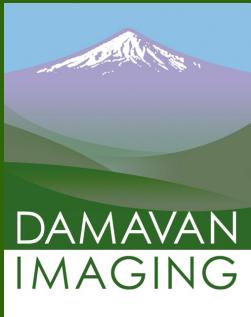


Multi-view reconstruction

- Improve the parallax of detected events
- Methods to be developed



Reconstruction 3D – Extended LMMLEM algorithm

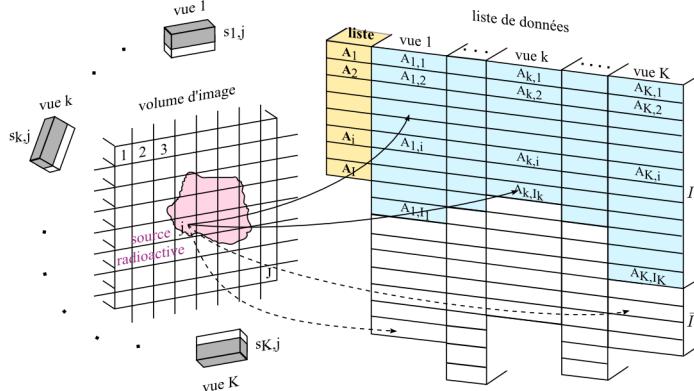


Ordinary LMMLEM

$$\hat{f}_j^{(t+1)} = \frac{\hat{f}_j^{(t)}}{s_j} \sum_{i=1}^I \frac{t_{ij}}{\sum_{l=1}^J t_{il} \hat{f}_l^{(t)}}$$

(failed due to lack of parallax)

new list-mode data

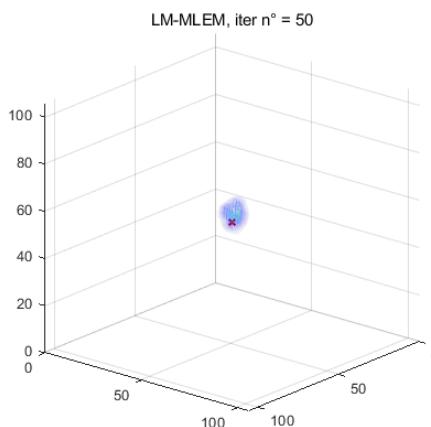
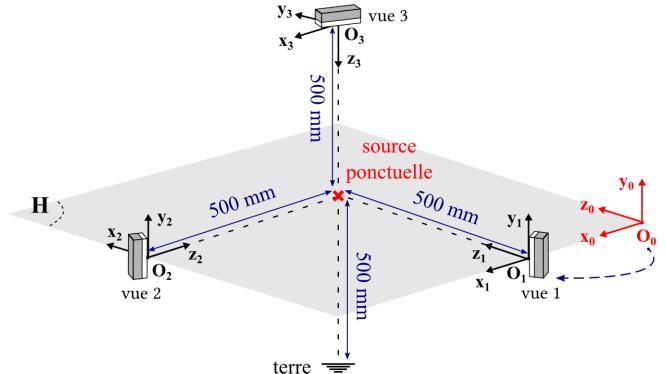


extended LMMLEM

$$\hat{f}_j^{(t+1)} = \frac{\hat{f}_j^{(t)}}{\sum_{k=1}^K s_{jk}} \sum_{i=1}^I \frac{\sum_{k=1}^K t_{ijk} \hat{f}_i^{(t)}}{\sum_{l=1}^J t_{il} \hat{f}_l^{(t)}}$$

(improved parallax)

Result given by extended LMMLEM algorithm



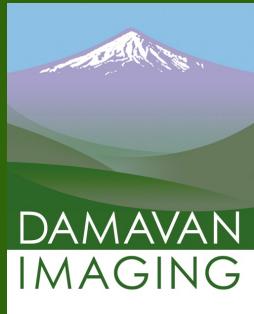
Problem

- Convergence is slow
- Large number of data is needed

Solutions

- Using a priori knowledge of source
- Bayésian reconstruction
- Optimisation algorithm

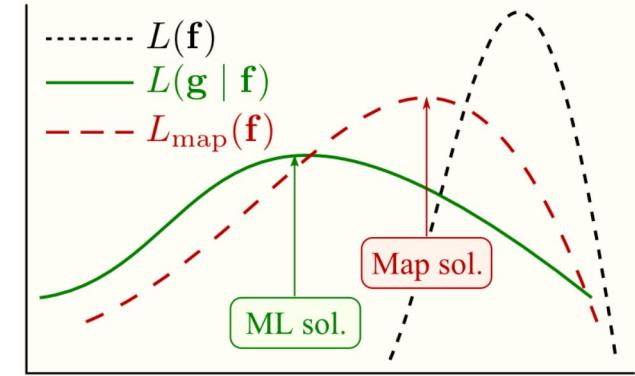
Markov Random Field MAP algorithm (LM-MRFMaP)



Bayesian approach: maximum a posteriori (MAP) estimate

$$\hat{\mathbf{f}} = \arg \max_{\mathbf{f} \geq 0} \{p(\mathbf{f} | \mathbf{g})\} = \arg \max_{\mathbf{f} \geq 0} \{L(\mathbf{g} | \mathbf{f}) + L(\mathbf{f})\}$$

MAP with Markov random field (MRF) prior



$$\hat{\mathbf{f}} = \arg \max_{\mathbf{f} \geq 0} \left\{ \sum_{i=1}^I \sum_{j=1}^J \left(\frac{t_{ij} f'_j}{\sum_{s=1}^J t_{is} f'_s} - t_{ij} f_j \right) \underbrace{- \sum_{j=1}^J \sum_{\{j,l\} \in C} \beta_{jl} \rho(f_j - f_l)}_{\text{energy of MRF prior}} \right\} = \arg \max_{\mathbf{f} \geq 0} \{L_{\text{map}}(\mathbf{f})\}$$

surrogate of Poisson log-likelihood

Iterative maximization scheme for a concave $L_{\text{map}}(\mathbf{f})$

$$\hat{\mathbf{f}}^{(k+1)} \leftarrow \hat{\mathbf{f}}^{(k)} + \sum_{j=1}^J a_j^{(k)} \mathbf{e}_j^{(k)} \quad \text{with} \quad a_j^{(k)} \leftarrow \arg \max_{a_j \geq -f_j^{(k)}} \{L_{\text{map}}(\hat{\mathbf{f}}^{(k)} + a_j \mathbf{e}_j^{(k)})\}$$

Numerical experiments on real data



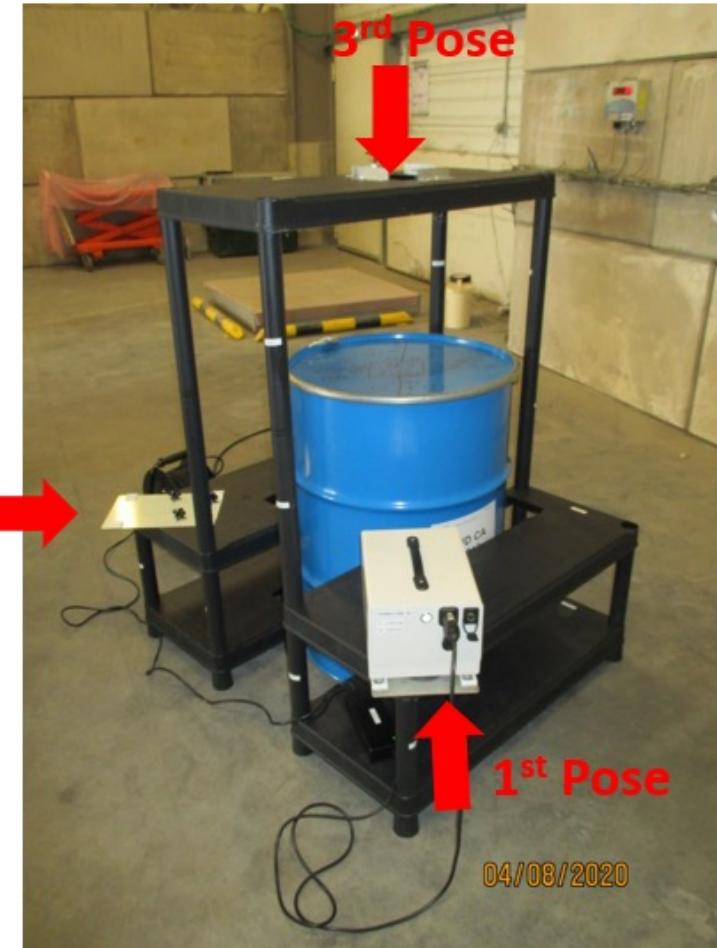
Temporal Imaging Compton camera V3 developped by Damavan Imaging



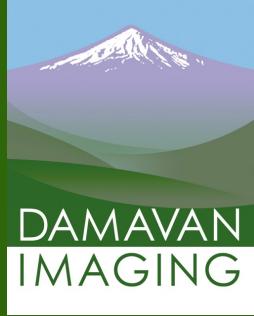
Radioactive sources :

- Sodium ^{22}Na
- 0,2 MBq of total activity
- Acquisition times: 20mn/view

Real experiment setting

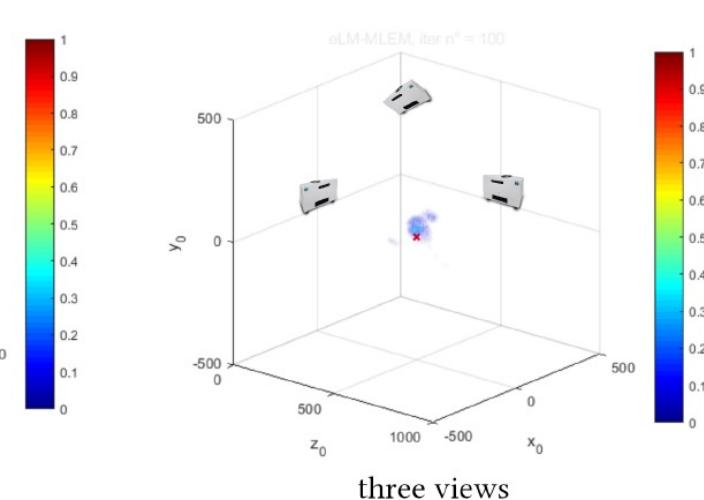
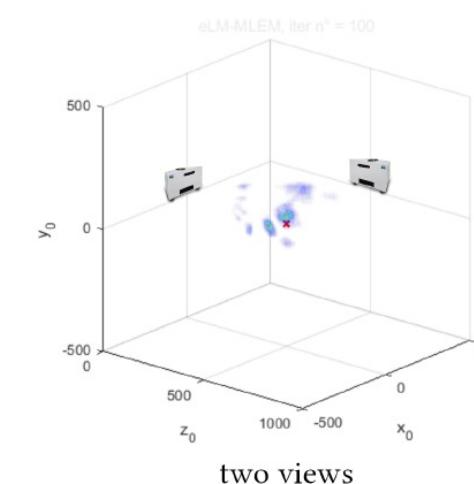
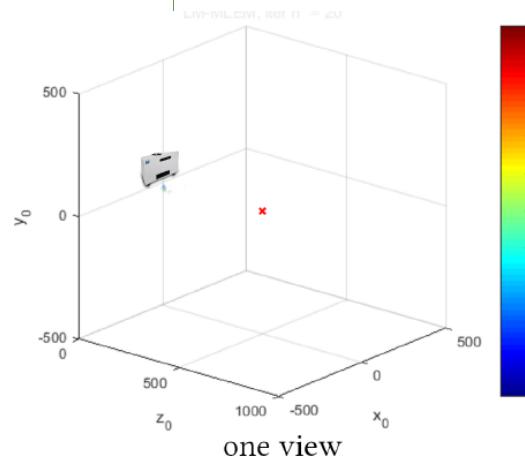


3D Reconstruction results – eLMMLEM algorithm



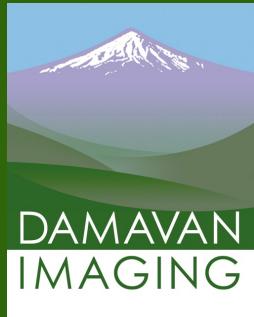
Reconstruction information

- Source volume: 1000 x 1000 x 1000 mm
- Energy range: 1.1 MeV – 1.3 MeV
- Iteration number: 100

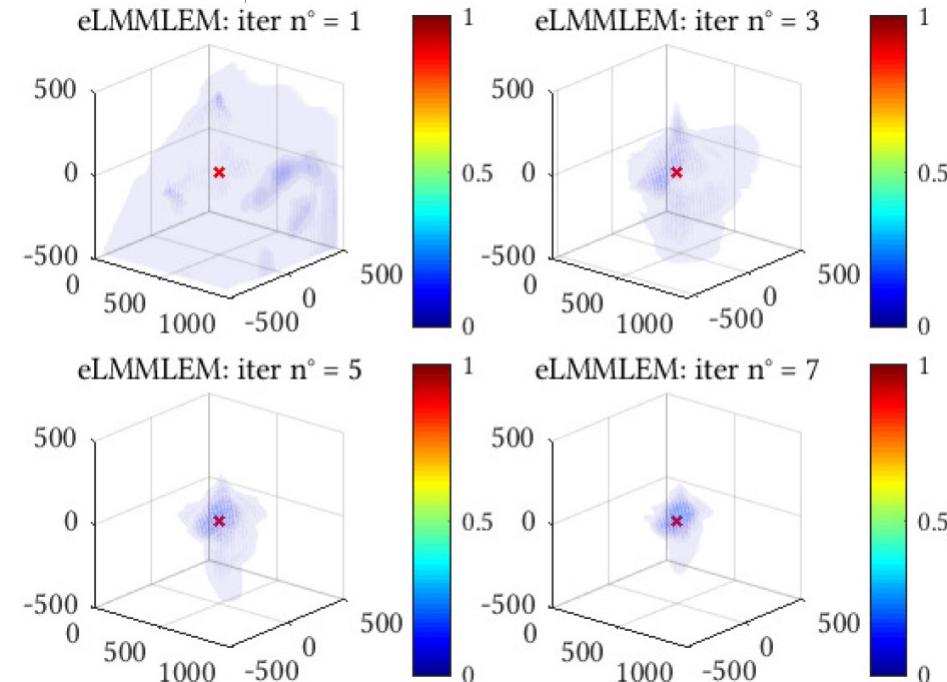


The higher the view number, the more the reconstruction result given by the eLMMLEM algorithm is better

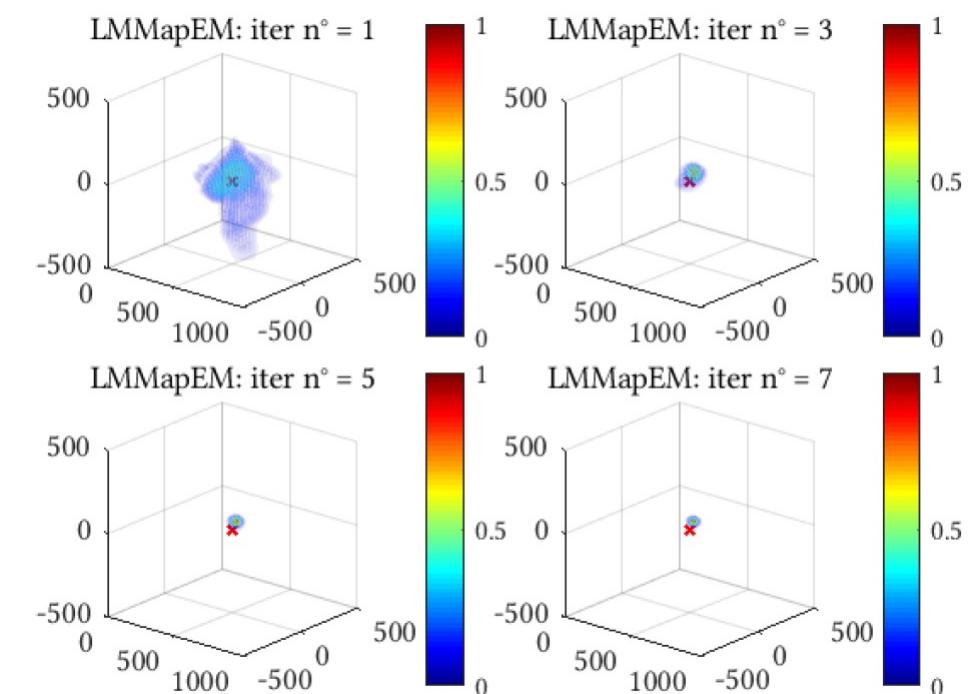
Reconstruction results - LM-MRFMaP algorithm



eLMMLEM algorithm



LM-MRFMaP algorithm



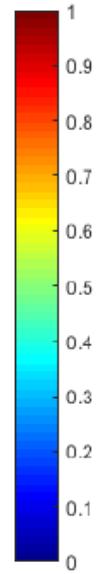
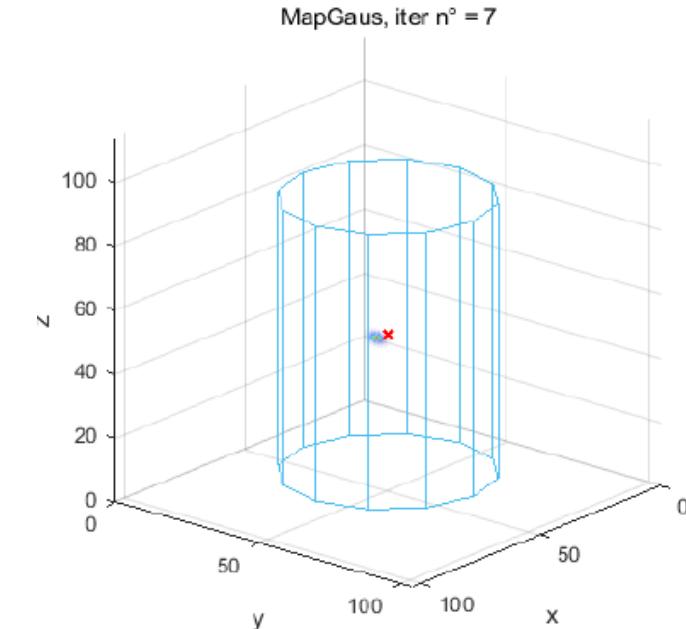
LM-MRFMaP algorithm converges **much faster** than the extended eLMMLEM algorithm

Application to the Dream Scanner project

This reconstruction technology will be implemented on Dream scanner prototype

This scanner is a joint project with Orano DS

- 3 Compton camera heads
- The scanner moves, not the drum
- Will Implements the new algorithm
- Our target is to reach 10 minutes for a 3D scan of a low density 225 l drum



Conclusions & perspectives



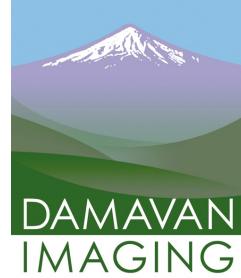
Conclusions

Building a new list-mode data space, adapted system response function & sensitivity
eLM-MLEM 3D reconstruction algorithm extended to multi-view Compton data
Proposing a new maximization scheme for MRF MAP algorithm
Numerical tests on real datasets show the outperformance
More advanced tests on reconstruction of extended sources underway

Perspectives

Extending the proposed algorithm for other kinds of MRF-based priors

Experiment on other radioactive sources (multi-point source, extended source, ...)



Thank you for your attention!

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