

Towards total-body J-PET: overview of data correction techniques for image reconstruction

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with contributions from Wojciech Krzemień, Szymon Parzych, Jakub Baran, and the J-PET collaboration

June 5, 2024

Outline

Introduction

Scatter correction

Random correction

Normalization correction

Resolution modeling

Conclusion

Outline

Introduction

Scatter correction

Random correction

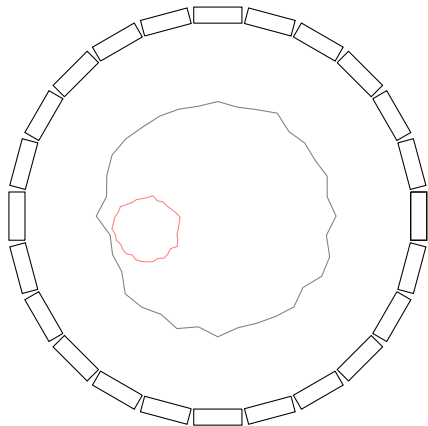
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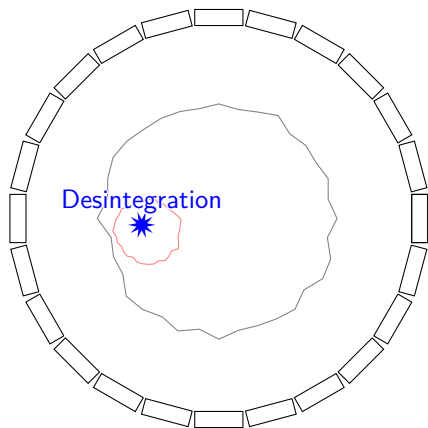
Positron Emission Tomography principles

- ▶ The radioisotope is injected into the patient



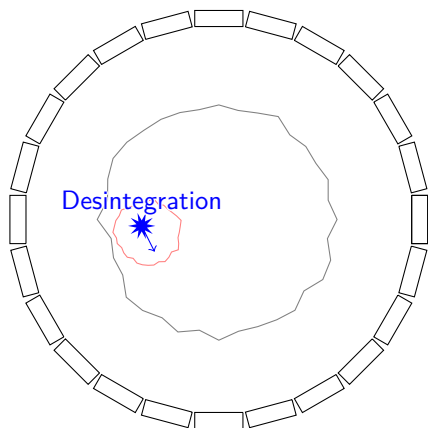
Positron Emission Tomography principles

- ▶ The radioisotope is injected into the patient
- ▶ The radioisotope undergoes positron emission decay (β^+)



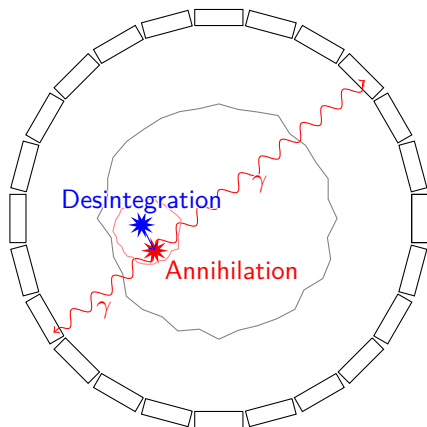
Positron Emission Tomography principles

- ▶ The radioisotope is injected into the patient
- ▶ The radioisotope undergoes positron emission decay (β^+)
- ▶ The positron travels for a short distance (about 1 mm)



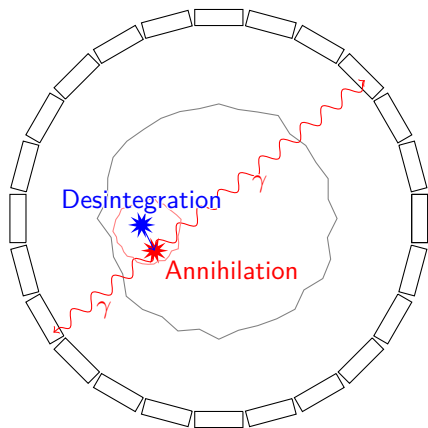
Positron Emission Tomography principles

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- ▶ The positron interacts with an electron, forming a pair of two back-to-back gamma photons



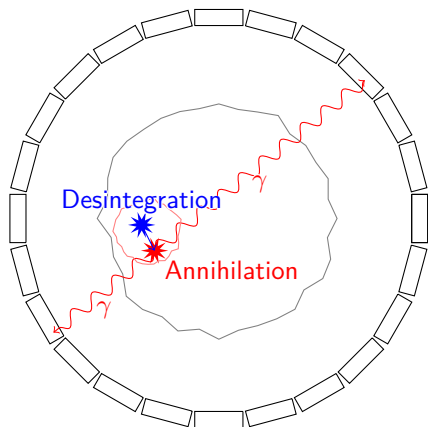
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- ▶ A list of coincidence events is constructed based on time window



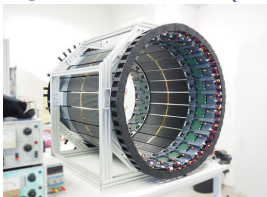
Positron Emission Tomography principles

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- ▶ A list of coincidence events is constructed based on time window
- ▶ Coincidences form LORs that hint at where the initial annihilation happened



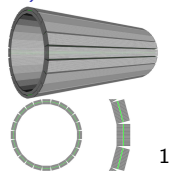
J-PET

Modular Jagiellonian PET (J-PET)



- ▶ 24 arrays of 13 plastic scintillator strips
- ▶ Length of 50 cm
- ▶ Radius of 40 cm

Total-body (TB) J-PET



- ▶ 7 rings composed of 2 layers of scintillator strips
- ▶ Layers separated by wavelength shifters
- ▶ Total length of more than 2 m

¹P Moskal et al. "Simulating NEMA characteristics of the modular total-body J-PET scanner—an economic total-body PET from plastic scintillators". In: *Physics in Medicine & Biology* 66.17 (Sept. 2021), p. 175015.

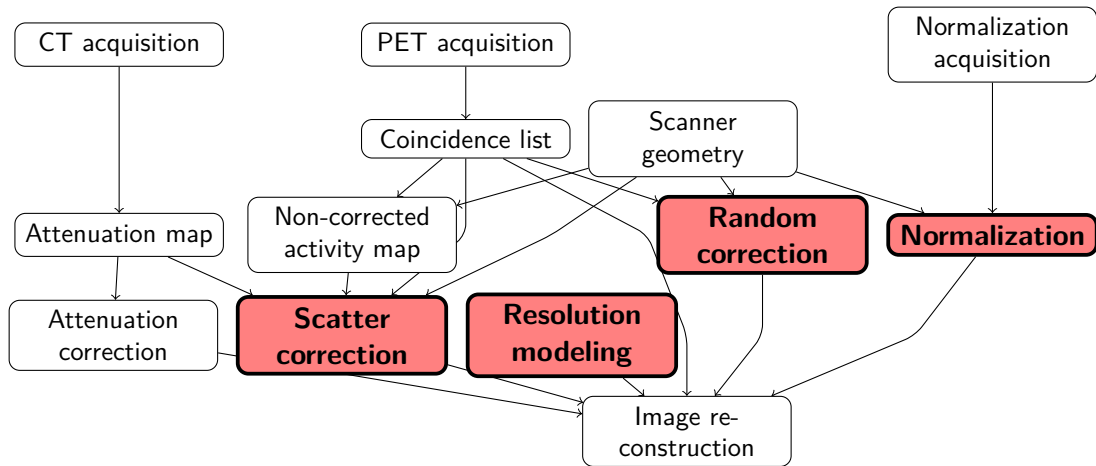
PET image correction

- ▶ PET images are degraded due to several effects
- ▶ Those effects can be compensated using different techniques:
 - ▶ Attenuation correction
 - ▶ Scatter correction
 - ▶ Random correction
 - ▶ Normalization correction
 - ▶ Resolution modeling

PET image correction

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Overview of PET corrections for image reconstruction



Maximum-Likelihood Expectation-Maximization

Update equation of Maximum-Likelihood Expectation-Maximization (MLEM):

$$x_q^{(k+1)} = \frac{x_q^{(k)}}{S_q} \sum_{p=1}^P A_{pq} \frac{y_p}{\eta_p \left(\sum_{q'=1}^Q A_{pq'} x_{q'}^{(k)} \right) + s_p + r_p} \quad (1)$$

Outline

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Scatter correction

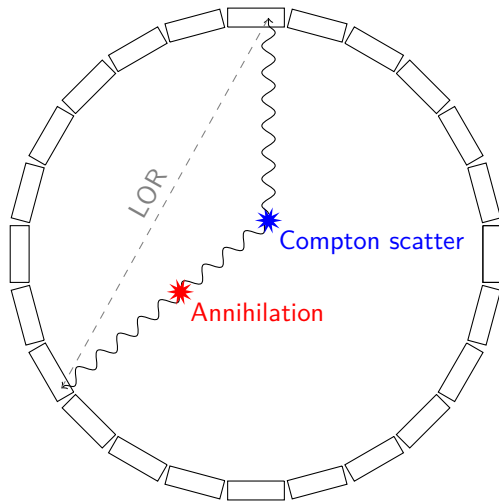
Random correction

Normalization correction

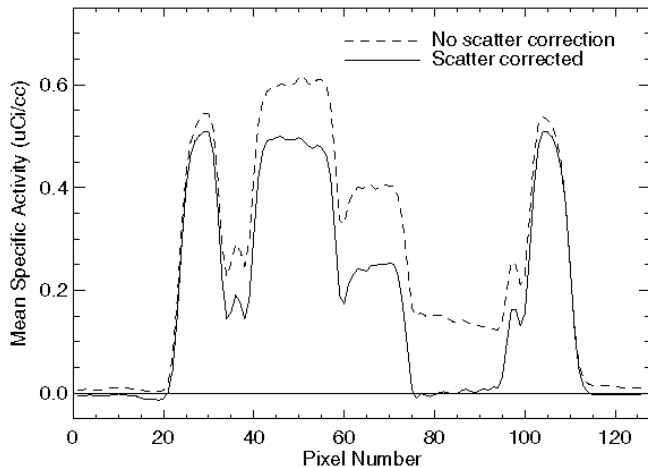
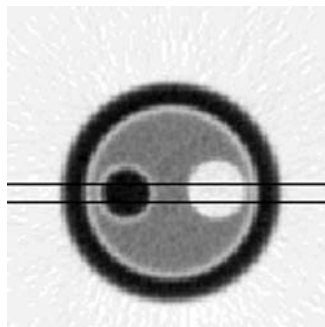
Resolution modeling

Conclusion

The problem with scatter coincidences



Importance of scatter correction



After simulation based scatter correction.

²Dale L. Bailey, ed. *Positron emission tomography: basic sciences*. New York: Springer, 2005.

Scatter correction

- ▶ Scattered coincidences correspond from 20 % to 50 % of all coincidences³
- ▶ Various corrections exist:
 - ▶ Empirical approaches (tail fitting)
 - ▶ Two (or more) energy windows
 - ▶ Convolution/deconvolution
 - ▶ Modeling of the scatter distribution during forward projection
 - ▶ Analytic methods (single scatter simulation (SSS))
 - ▶ Monte Carlo (MC) methods
 - ▶ Machine-learning-based approaches

³Bailey, *Positron emission tomography*.

Single scatter simulation

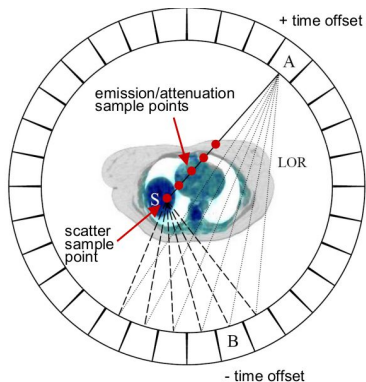


Fig. 1. The single scatter simulation model.

Single-scatter coincidence rate along LOR (A, B) is estimated as the volume integral of a scattering kernel over the scattering medium⁴:

$$S^{AB} = \int_{V_s} dS \left(\frac{\sigma_{AS}\sigma_{BS}}{4\pi R_{AS}^2 R_{BS}^2} \right) \frac{\mu}{\sigma_c} \frac{d\sigma_c}{d\Omega} [I^A + I^B] \quad (2)$$

⁴C C Watson. "New, Faster, Image-Based Scatter Correction for 3D PET". In: *IEEE Transactions on Nuclear Science* (2000).

Reconstructions (courtesy of Jakub Baran)

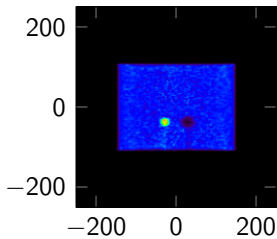
- ▶ Use of STIR⁵ SSS implementation
- ▶ Reconstructions of standard NEMA IEC phantom
- ▶ Reconstruction by MLEM (from Customizable and Advanced Software for Tomographic Reconstruction⁶)

⁵Kris Thielemans et al. “STIR: software for tomographic image reconstruction release 2”. In: *Physics in Medicine and Biology* 57.4 (Feb. 2012), pp. 867–883.

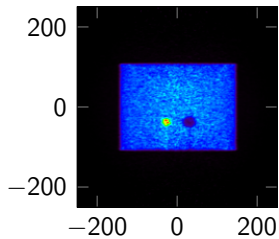
⁶Thibaut Merlin et al. “CASToR: a generic data organization and processing code framework for multi-modal and multi-dimensional tomographic reconstruction”. In: *Physics in Medicine & Biology* 63.18 (Sept. 2018), p. 185005.

Reconstruction results (x profiles)

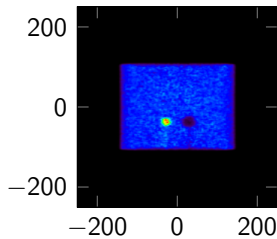
True



True + scatter

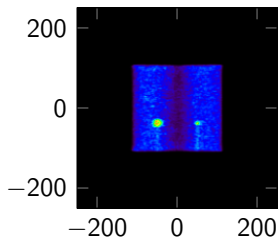


True + scatter + SSS

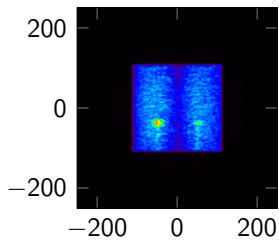


Reconstruction results (y profiles)

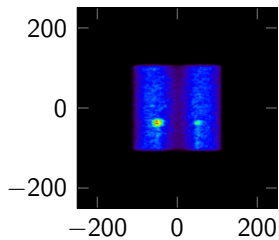
True



True + scatter

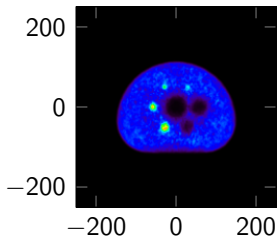


True + scatter + SSS

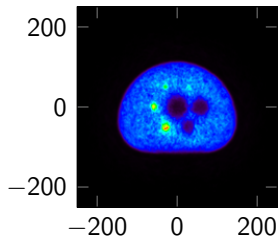


Reconstruction results (z profiles)

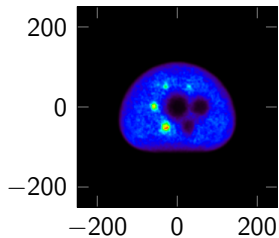
True



True + scatter



True + scatter + SSS



Scatter correction conclusions

- ▶ Scatter correction is an important step in PET image reconstruction
- ▶ For J-PET: extend SSS to take into account time-of-flight information⁷

⁷Charles C Watson. "Extension of Single Scatter Simulation to Scatter Correction of Time-of-Flight PET". In: *IEEE Nuclear Science Symposium (2005)*.

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Introduction

Scatter correction

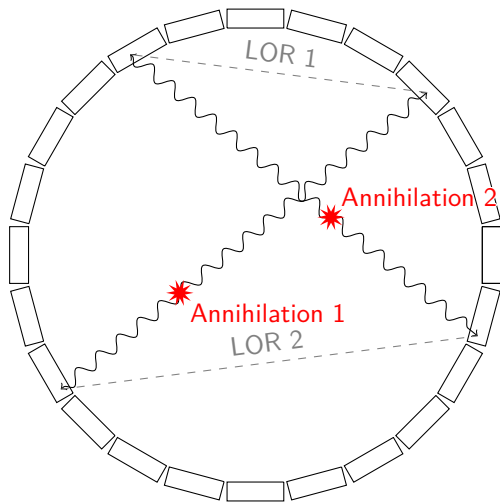
Random correction

Normalization correction

Resolution modeling

Conclusion

The problem with random coincidences



Random correction

- ▶ Unlike true coincidences, random coincidences tend to be somewhat uniformly distributed across the field-of-view⁸
- ▶ Various corrections exist:
 - ▶ Tail-fitting methods
 - ▶ Single rate
 - ▶ Singles-prompts
 - ▶ Delayed time window

⁸Bailey, *Positron emission tomography*.

Overview of selected random correction techniques

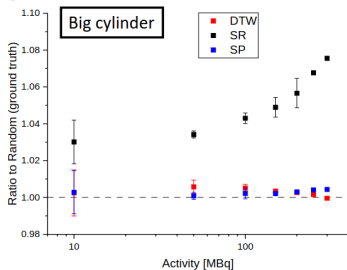
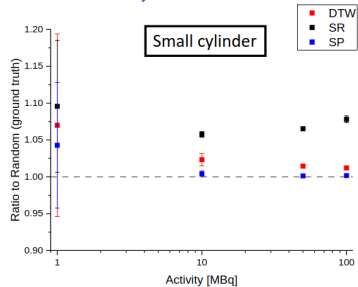
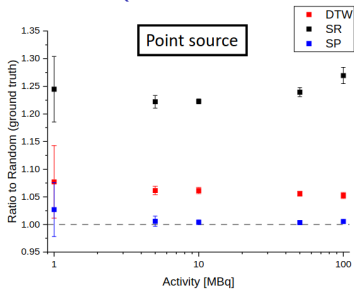
- ▶ Single rate: $R_{ij}^{SR} = 2\tau R_i R_j$
- ▶ Singles-prompts: extension to the conventional single rate approach using information from singles and prompts rate
- ▶ Delayed time window
 - ▶ Timing signals from one detector are delayed by a time significantly greater than the time window (τ)
 - ▶ Number of coincidences found estimate the number of random coincidences
 - ▶ This estimate is then subtracted to the total number of coincidences

Simulations set-up (courtesy of Szymon Parzych)

- ▶ MC simulations conducted with GATE 9.0⁹
- ▶ Phantoms:
 - ▶ Point source at the center
 - ▶ Small water-filled cylinder (radius of 15 cm, length of 22 cm)
 - ▶ Large water-filled cylinder (radius of 10.555 cm, length of 168 cm)
 - ▶ NEMA IEC
- ▶ Coincidence time window: $\tau = 3$ ns

⁹David Sarrut et al. "Advanced Monte Carlo simulations of emission tomography imaging systems with GATE". In: *Physics in Medicine & Biology* 66.10 (May 2021), 10TR03.

Preliminary results (courtesy of Szymon Parzych)



Conclusions for random correction (courtesy of Szymon Parzych)

- ▶ Singles-prompts method provides the best estimation of total random coincidences
- ▶ However, the delayed time window method provides the best distribution of random coincidences
- ▶ Delayed time window seems adapted to J-PET due to its triggerless acquisition
- ▶ More investigations to be done, especially for TB J-PET

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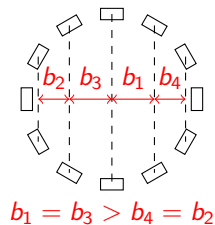
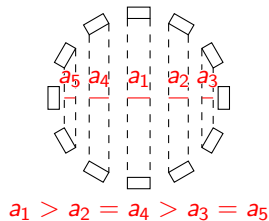
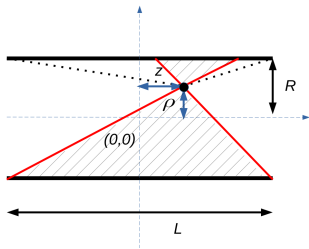
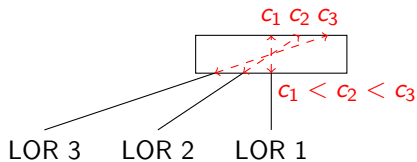
Normalization correction

Resolution modeling

Conclusion

Why is normalization needed?

Several effects can affect LOR sensitivity



The problem of normalization

- ▶ Assuming a “perfect” source of activity A , we have

$$C_{\text{LOR}} = A \quad (3)$$

- ▶ In practice, a number of effects affect the count rate:

$$C_{\text{LOR}} = F_{\text{LOR}} \times A \quad (4)$$

- ▶ Goal of normalization: find η_{LOR} such that

$$C_{\text{LOR}} \times \eta_{\text{LOR}} = A \quad (5)$$

- ▶ Two approaches:
 - ▶ Direct normalization
 - ▶ Component-based normalization

Notations taken from Theodorakis et al.¹⁰.

¹⁰Lampros Theodorakis et al. “A review of PET normalization: striving for count rate uniformity”. In: *Nuclear Medicine Communications* 34.11 (Nov. 2013), pp. 1033–1045.

Direct normalization

► We want

$$C_{\text{LOR}} \times \eta_{\text{LOR}} = A \quad (5)$$

► Therefore,

$$\eta_{\text{LOR}} = \frac{A}{C_{\text{LOR}}} \quad (6)$$

Direct normalization

- ▶ We want

$$C_{\text{LOR}} \times \eta_{\text{LOR}} = A \quad (5)$$

- ▶ Therefore,

$$\eta_{\text{LOR}} = \frac{A}{C_{\text{LOR}}} \quad (6)$$

- ▶ Problem: statistics

- ▶ Modular J-PET has $24 \times 13 \times 25 = 7800$ “detector pixels”
- ▶ ...hence $\frac{7800 \times (7800 - 1)}{2} = 30\,416\,100$ possible LORs
- ▶ 1% error \rightarrow 10 000 counts per LOR
- ▶ Thus we need about 304 161 000 000 coincidences!
- ▶ TB J-PET requires 49 times more coincidences!

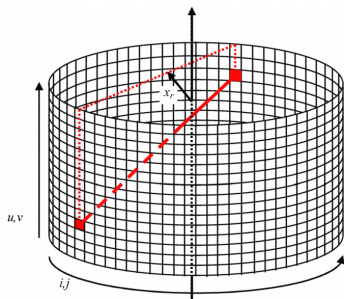
Component-based normalization

General idea

Improve statistics and decrease variance by considering several LORs for normalization computation¹¹:

$$\eta_{uivj} = g_{uv}^{\text{ax}} \times g_{ij}^{\text{tr}} \times \epsilon_{ui} \times \epsilon_{vj} \quad (7)$$

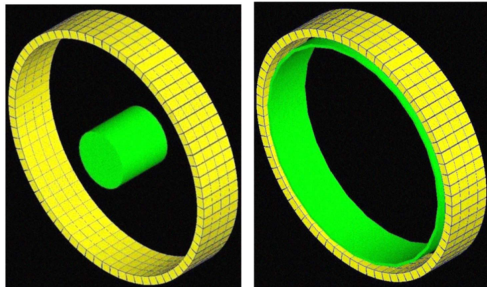
- g^{ax} Axial geometric factors
- g^{tr} Transverse geometric factors
- ϵ Intrinsic detector efficiency



¹¹Audrey Pépin et al. "Normalization of Monte Carlo PET data using GATE". In: *2011 IEEE Nuclear Science Symposium Conference Record*. Valencia, Spain: IEEE, Oct. 2011, pp. 4196–4200.

Simulations

Normalization scans



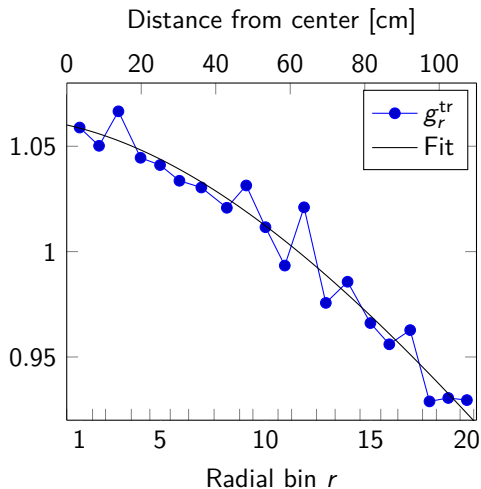
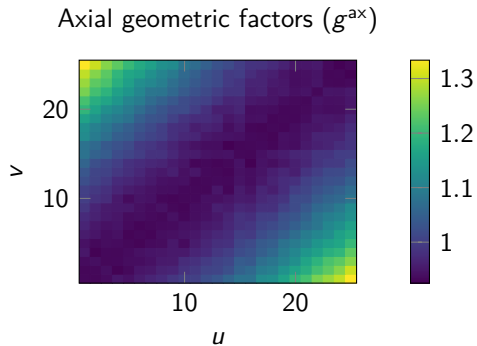
Courtesy of Pépin et al.¹²

Reconstruction phantom

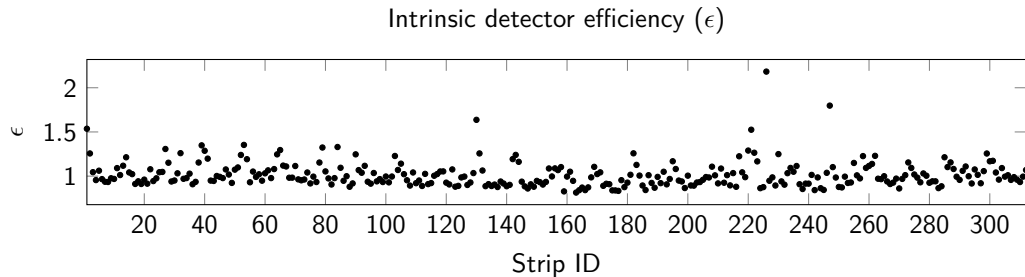
- ▶ Uniform cylinder
- ▶ Length: 40 cm (Modular J-PET: 50 cm)
- ▶ Radius: 10 cm (Modular J-PET: 40 cm)

¹²Pépin et al., “Normalization of Monte Carlo PET data using GATE”.

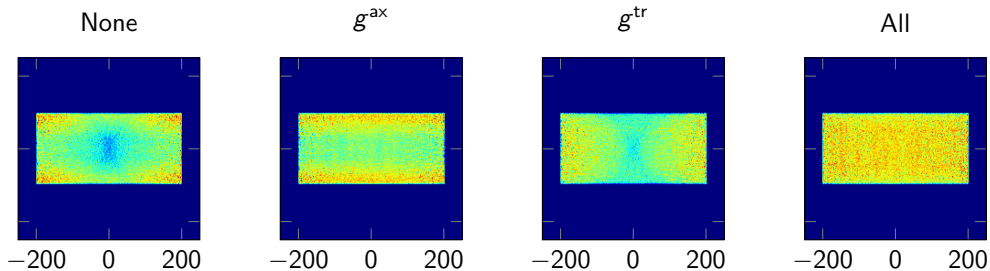
Computed normalization factors



Computed normalization factors (cont.)

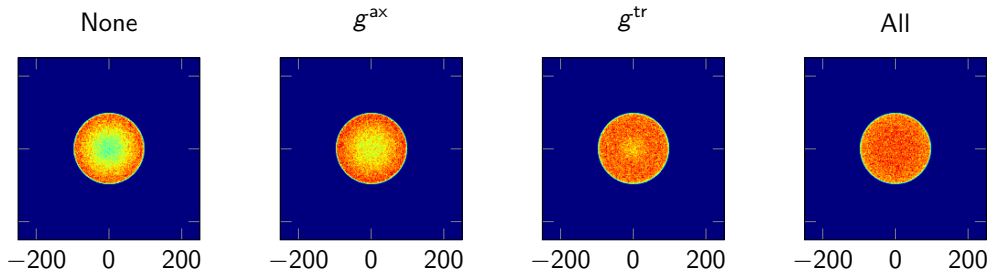


Reconstructions (x profiles)



	Axial uniformity	Radial uniformity
None	9%	24%
All	1%	12%

Reconstructions (z profiles)



	Axial uniformity	Radial uniformity
None	9%	24%
All	1%	12%

Normalization conclusion

- ▶ Normalization favorably compensates for several effects, including some geometrical effects or intrinsic detector efficiencies
- ▶ Normalization implemented for Modular J-PET
- ▶ Must be investigated for TB J-PET, especially due to its length

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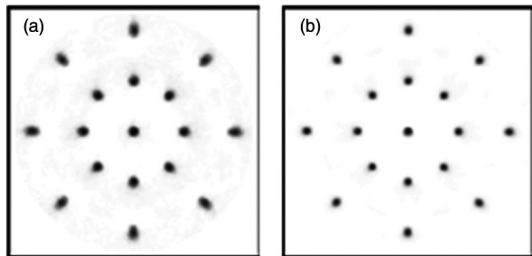
Normalization correction

Resolution modeling

Conclusion

The problem

- ▶ Resolution degrading factors translate to undesired cross-contamination between adjacent functional regions
 - ▶ Positron range
 - ▶ Photon noncollinearity
 - ▶ Detector-related effects
 - ▶ Intercrystal scattering
 - ▶ Intercrystal penetration



Kisung Lee et al. "Pragmatic fully 3D image reconstruction for the MiCES mouse imaging PET scanner". In: *Physics in medicine & biology* 49.19 (2004), p. 4563

Solutions to resolution degradation

- ▶ Post-processing techniques
 - ▶ ROI-based techniques (from segmented MRI images)
 - ▶ Voxel-based techniques
- ▶ Incorporation of anatomical information within the reconstruction algorithm
 - ▶ Typically superior to post-processing techniques
 - ▶ Drawback: simplifying assumptions
- ▶ Resolution modeling

Resolution modeling techniques

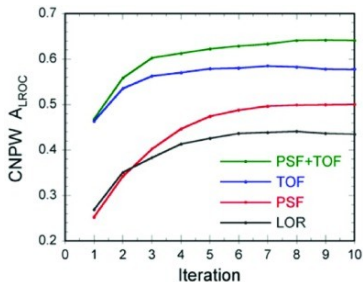
- ▶ Idea: incorporate the resolution modeling directly within the system matrix
- ▶ The system matrix is modeled as \mathbf{A} when a_{ij} is the probability that an event generated in voxel j is detected along a LOR i
- ▶ Example of system matrix decomposition:

$$\mathbf{A} = \mathbf{A}_{\text{det.sens}} \mathbf{A}_{\text{det.blur}} \mathbf{A}_{\text{attn}} \mathbf{A}_{\text{geom}} \mathbf{A}_{\text{positron}} \quad (8)$$

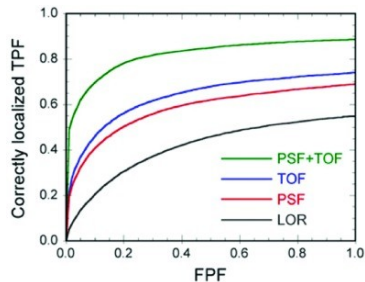
Some results from the literature



Alain P Pecking, Dominique Bellet, and Jean Louis Alberini. "Immuno-SPET/CT and immuno-PET/CT: a step ahead to translational imaging". In: *Clinical & experimental metastasis* 29 (2012), pp. 847–852

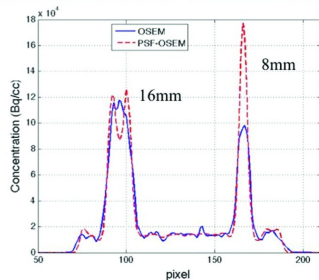
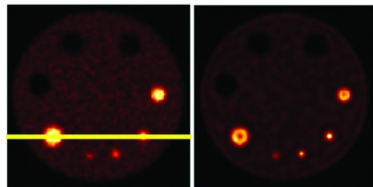


Dan J Kadrmas et al. "Impact of time-of-flight on PET tumor detection". In: *Journal of Nuclear Medicine* 50.8 (2009), pp. 1315–1323



Limitations of resolution modeling

- ▶ Resolution modeling can lead to notable edge artifacts, reminiscent of the Gibbs phenomenon
- ▶ Some solutions exist
 - ▶ Use a reconstruction filter that underestimates the true resolution
 - ▶ Amplify a frequency band in the Fourier domain



Bing Bai and Peter D Esser. "The effect of edge artifacts on quantification of positron emission tomography". In: *IEEE Nuclear Science Symposium & Medical Imaging Conference*. IEEE, 2010, pp. 2263–2266

Resolution modeling conclusions

- ▶ Resolution modeling results in significant improvements in image resolution and contrast
- ▶ Effects on noise is less straightforward to assess
- ▶ Main drawback is the edge artifacts, that are not yet fully understood
- ▶ Still an open topic in the context of J-PET!

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Corrections matter

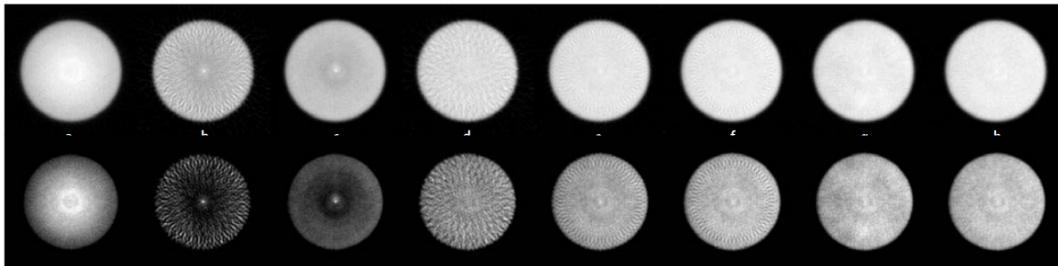


Figure 5.10. Effects of normalisation on image uniformity. Images (summed over all axial planes) from a low-variance 20 cm cylinder acquisition, performed in 3D mode on a Siemens/CTI ECAT 951. (From [15], with permission.)

(Upper row) linear grey scale covering entire dynamic range.

(Lower row) linear grey scale, zero-point set to 70% of image maximum.

(a) no scatter correction;

(e) no transaxial block profile correction;

(b) no normalisation;

(f) no crystal interference correction;

(c) no correction for the radial profile;

(g) no time alignment correction;

(d) no crystal efficiency correction;

(h) fully normalised and scatter corrected.

Dale L. Bailey, ed. *Positron emission tomography: basic sciences*. New York: Springer, 2005

General conclusions

- ▶ PET imaging requires several corrections to become quantitative
- ▶ A final calibration is required to convert reconstructed values to physical units
- ▶ Currently only partially implemented in the context of J-PET
 - ▶ Current study focus on Modular J-PET
 - ▶ ...but TB J-PET is kept as a goal
- ▶ Moving from MC simulations to real data
- ▶ Work is ongoing!

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Thanks for your attention!