# Image reconstruction for the Walk-Through PET system

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# Belgium

## Costa Rica









• Engineering Physics at Ghent University

- Engineering Physics at Ghent University
- Master thesis at Kyushu University
  - Molecular dynamics and (S)TEM to study nanoparticles







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  - Fraud detection with deep learning and graph theory
- PhD at Ghent University
  - Time of flight in monolithic PET detectors
  - Image reconstruction for the WT-PET





# Cerenkov in monolithic BGO

A simulation study in GATE

### Cerenkov in monolithic BGO Cerenkov emission





#### Cerenkov in monolithic BGO

#### Cerenkov emission



	# Cerenkov photons
Compton scattered	10
Purely photoelectric	17
Average	13

#### **Compton scattered** Photoelectric effect 511 keV e ionization 420 keV $v_{e^{-}} > c$ 50 keV $v_{e^-} < c$

0 keV

#### Purely photoelectric

9



#### Cerenkov in monolithic BGO Cerenkov emission



### Cerenkov in monolithic BGO Cerenkov detection

lateral surfaces 1.0 1.0 Broadcom NUV-MT PDE Cerenkov BGO Cerenkov LYSO scintillation BGO 0.8 0.8 ---- scintillation LYSO reflectance 0.6 0.6 PDE 0.4 rough black (BGO) rough black (LYSO) 0.4 0.2 rough bare (BGO) rough bare (LYSO) polished grease ESR (BGO) polished grease ESR (LYSO) 0.2 back surfaces 20 40 60 80 0 angle of incidence (°) 1.00 polished grease (BGO) polished grease (LYSO) rough grease (BGO) 0.0 rough grease (LYSO) 500 600 700 800 900 300 400 0.75 wavelength (nm) reflectance 0.50 0.25 0.00

60

80

40

angle of incidence (°)

20

0

л 1.0

0.8

<sup>0.6</sup> ), vield (a.u.)

0.2

0.0

Cerenkov in monolithic BGO Cerenkov detection



Spread out over 8x8 (64) SiPMs

- $\rightarrow$  Unlikely to detect > 1 Cerenkov photons per SiPM
- $\rightarrow$  Requires triggering below the single photoelectron level

# The Walk-Through PET

A flat panel LAFOV PET design based on monolithic detectors

# The Walk-Through PET The team in Belgium



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Meysam Dadgar



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**Boris Vervenne** 

# The Walk-Through PET Outline

- Design and current state
- NEMA system characterization
- Image reconstruction
- Remaining challenges
  - Limited angle artefacts
  - CT-less attenuation correction
- XCAT lesion detectability

# Design and current state Total-Body PET



#### **Increased sensitivity**

- Better image quality
- Lower dose scans
- Shorter scans

#### Single bed position scanning

• Dynamic imaging

#### Increased system cost

- Fixed costs: CT, data/reconstruction server, ...
- Scales linearly with AFOV: scintillators, SiPMs, electronics

#### Patient throughput remains limited

Sub one-minute scans become possible
→ patient preparation becomes limiting factor

# Design and current state Walk-Through PET



Dimensions based on CT-scans from PET patients 106 cm AFOV: simultaneous head + torso imaging

#### Increased patient throughput

• Aiming for 30 s acquisitions

## Design and current state Monolithic detectors



#### \*

#### **Spatial resolution:**

- 2D: 1.3 mm
- DOI: 2 mm

#### **TOF resolution:**

• 327 / 400 ps

#### **Energy resolution:**

• 15 %

#### Maximum countrates:

- 1 Mcps
- 370 ns deadtime

#### Current state



- Fundamental system/detector design
- Simulations for system characterization
- Software development



- System integration (electronics, mechanical)
- Cooling system
- End-user usability



• Provides the electronic readout: TOFPET2 ASIC

# Sparse LYSO configuration



# Sparse LYSO configuration



	Centered	Off center at X = 10cm	Off center at Y = 10cm
Sensitivity (kcps/MBq) Full detector coverage	150	121	117
Sensitivity (kcps/MBq) <b>Sparse 70% coverage</b>	77	63	60

A comparison with Quadra based on GATE simulations

Quadra (70 cm source)

-20

0

axial position (cm)

20

40

60

- 0 cm (85 MRD)

## Sensitivity



0 -20 0 20 40 60 axial position (cm)	70 cm source (center)	Ser (sin cps
Quadra (106 cm source)	WT-PET	
(322 MRD) — 0 cm (85 MRD) n (322 MRD) — 10 cm (85 MRD)	Quadra (MRD 85)	
	Quadra (MRD 322)	

70 cm source (center)	Sensitivity (simulated) cps/kBq	Sensitivity (experimental) cps/kBq		
WT-PET	154.0	-		
Quadra (MRD 85)	87.0	82.6		
Quadra (MRD 322)	179.7	175.3		

### Scatter fraction

	Simulated	Experimental		
WT-PET (offset along panels)	30.72 %			
WT-PET (offset towards panels)	29.58 %			
Quadra (MRD 85)	34.80 %	36 %		
Quadra (MRD 322)	36.18 %	37 %		
Includes patient bed				

#### Count rates



Prenosil et al.

#### NEMA system characterization Spatial resolution



#### Spatial resolution: detector smearing







### NEMA system characterization Spatial resolution



#### Spatial resolution: NEMA point sources

Scannor	Source Position		FWHM (mm)			FWTM (mm)		
Scallier	(cm)		х	У	Z	х	У	Z
		(1, 0, 0)	1.20	1.62	1.30	2.87	4.14	3.06
	Center	(10, 0, 0)	1.16	1.90	1.12	2.81	4.83	2.89
		🔻 (20, 0, 0)	1.17	1.94	1.19	2.60	4.36	2.81
		(1, 0, 39.75)	1.14	2.13	1.24	2.86	5.22	3.20
	3/8 of AFOV	(10, 0, 39.75)	1.21	2.48	1.29	3.19	7.11	3.06
		\star (20, 0, 39.75)	1.26	2.52	1.05	2.08	5.58	2.08
W/T_PET								
		(0, 1, 0)	1.18	1.75	1.34	2.43	3.79	2.54
	Center	(0, 10, 0)	1.22	1.92	1.33	2.54	4.20	2.54
		\dagger (0, 20, 0)	1.32	2.25	1.41	2.71	7.08	2.71
		(0, 1, 39.75)	1.29	2.24	1.16	2.58	5.71	2.59
	3/8 of AFOV	(0, 10, 39.75)	1.52	2.71	1.46	2.91	6.54	3.35
		🕈 (0, 20, 39.75)	1.65	3.36	1.88	3.76	8.18	3.96
		. (1 0 0)	0.55	0.00	0.05	F (0	C 45	7 00
	Carta	(1, 0, 0)	2.55	2.62	2.85	5.08	0.45	7.03
	Center	(10, 0, 0)	3.70	3.27	2.95	1.24	7.98	1.18
Quadra 322 MRD		▼ (20, 0, 0)	5.24	3.97	3.62	10.39	9.39	8.03
		(1, 0, 39.75)	2.09	2.35	2.24	4.70	5.70	4.68
	3/8 of AFOV	(10, 0, 39.75)	2.78	3.14	2.89	5.85	7.64	5.03
		<b>(20, 0, 39.75)</b>	4.63	3.31	3.20	8.61	6.81	5.70
		• (1 0 0)	2.62	2 70	2 72	F 02	5 66	1 82
Quadra 85 MRD	Contor	(1, 0, 0)	2.02	2.70	2.12	5.02	5.00 6.43	4.02 5.86
	Center	(10, 0, 0)	2.00	2.12	2.10	0.00	0.45	5.00
		(20, 0, 0)	2.24	5.40 2.44	2.94	9.21	5.71	J.19 A QA
	$2/9$ of $\Lambda EOV$	(1, 0, 39.75)	2.22	∠.44 3.02	2.40	4.00	5.71 6.67	4.04 5.05
	S/O ULAFUV	(10, 0, 39.75)	3.31	3.02	2.71	0.33	10.07	5.95 E 00
		<b>V</b> (20, 0, 39.75)	4.29	2.00	2.23	0.00	10.78	5.ŏ2



#### Spatial resolution: additional point sources





### Spatial resolution: additional point sources





### Spatial resolution: additional point sources





### NEMA system characterization Image quality: visual



- 4:1 activity ratio
- 30 s acquisition
- trues only

#### Image quality: contrast recovery



#### Image quality: additional lesions


# PETRecon

- Iterative listmode reconstruction on the GPU
- Written in Julia
  - High level but fast (similar speed to C++)
  - Good support for GPU programming with CUDA
  - Good support for deep learning
  - Supports auto differentiation
- Implementations for:
  - Emission tomography (MLEM)
  - Transmission tomography (MLTR)
  - Simultaneous estimation of activity and attenuation (MLAA)

# PETRecon

### Siddon's algorithm



- Thread 1
- Thread 2
- Thread 3
- Thread 4

#### **Slice-based raytracing**



GPU





### PETRecon





- Male XCAT phantom (BMI=22.7)
- 30 s acquisition
- trues only







# Limited angle artefacts

# Limited angle artefacts Angular coverage



224°/360° = **62%** 

# Limited angle artefacts Angular coverage





# Limited angle artefacts Impact of TOF



### Limited angle artefacts

# Rotating configuration





# Limited angle artefacts Rotating configuration



### Limited angle artefacts

# Rotating configuration





- Remove limited angle artefacts
- Reduced sensitivity (about 33% lower)
- Increased system complexity
- Decreased patient throughput
- $\rightarrow$  Other solution may be preferable (e.g. AI based)



# Limited angle artefacts Deep learning artefact correction



# Limited angle artefacts Deep learning artefact correction

Artefact correction as a **regularization** step

$$\hat{x} = argmin_{x}[f(Hx + e, y) + \lambda R(x)]$$
data consistency regularized

regularization

$$R(x) = \|x - F(x)\|_2$$

Artefact correcting neural network

# Limited angle artefacts Deep learning artefact correction



#### Network needs to be trained on a wide variety of inputs!

### Limited angle artefacts

# Deep learning artefact correction





- Randomize angular range
  - Add Poisson noise (randomized magnitude)
  - Add Gaussian blur (randomized sigma)







CT-less attenuation correction Available methods

- Transmission based MLTR ←
- Deep learned (Florence)

### Transmission based



#### 8x 3MBq Ge-68 transmission sources

#### $3MBq \rightarrow 10\%$ of maximum detector count rate

CT-less attenuation correction Transmission based



Can be done simultaneously with emission scan

But for the purpose of simulation done separately

# Transmission based: stationary WT-PET



fixed configuration





x (mm)

# Transmission based: rotating WT-PET



70 cm

# CT-less attenuation correction Emission based



Maximum likelihood activity and attenuation (MLAA):



For TOF PET: attenuation / activity determined up to a constant → Additional constraint required (e.g. injected activity)

# Emission based: stationary WT-PET



fixed configuration



# Emission based: rotating WT-PET



rotating configuration

# Comparison

### • MLTR

- Better accuracy of activity map
- Faster reconstruction
- MLAA
  - Reduced system complexity
  - Better reconstruction of attenuation map (for fixed configuration)

# CT-less attenuation correction Comparison

MLTR



MLEM

$$\lambda_j^{(k+1)} = rac{\lambda_j^{(k)}}{\displaystyle{\sum_{i \in I}} H_{ij}} \displaystyle{\sum_{i \in E}} rac{H_{ijt}}{\displaystyle{\sum_{j' \in J} H_{ij't} \lambda_{j'} + r_i + s_i}}$$

- B = blank scan
- T = transmission scan
- E = emission scan
- I = space of possible LORs

MLAA

$$a_i^{(k)} = \exp\left(-\sum_{j\in J} l_{ij} \mu_j^{(k)}
ight)$$

$$\lambda_j^{(k+1)} = rac{\lambda_j^{(k)}}{\displaystyle{\sum_{i\in I}}a_i^{(k)}c_{ij}}\displaystyle{\sum_{i\in E}}rac{c_{ijt}}{\displaystyle{\sum_{j'\in J}c_{ij't}\lambda_{j'}^{(k)}}}$$

$$ar{y}_i^{(k+1)} = a_i^{(k)}\sum_{j\in J}c_{ij}\lambda_j^{(k+1)}$$

$$\mu_{j}^{(k+1)} = \mu_{j}^{(k)} - \frac{\sum_{i \in E} l_{ij} - \sum_{i \in I} l_{ij} \bar{y}_{i}^{(k+1)}}{\sum_{i \in I} \left( l_{ij} \bar{y}_{i}^{(k+1)} \sum_{j' \in J} l_{ij'} \right)}$$

# XCAT lesion detectability

XCAT lesion detectability Lesion parameters

- TBR: 8:1, 4:1, 2:1
  - Only 8:1 shown here
- Lesion size (diameter): 10 mm, 7 mm, 5 mm
- Location: liver, lung, breast

# XCAT lesion detectability Transverse slices



# XCAT lesion detectability Coronal slices



# XCAT lesion detectability Sagittal slices


#### **XCAT** lesion detectability

## TBR values, female

BMI = **18.64** 

BMI = 22.47



## XCAT lesion detectability TBR values, male



# Motion study

Motion study

#### V1: optical cameras



### Motion study V2: Infrared camera













#### Motion study

### Motion artefact reduction

**1.** Reduce motion as much as possible

#### 2. Correct for motion during reconstruction



