

Single-event based TOF FBP image reconstruction in J-PET

Roman Shopa

National Centre for Nuclear Research,
Świerk Computing Centre, Otwock-Świerk, Poland

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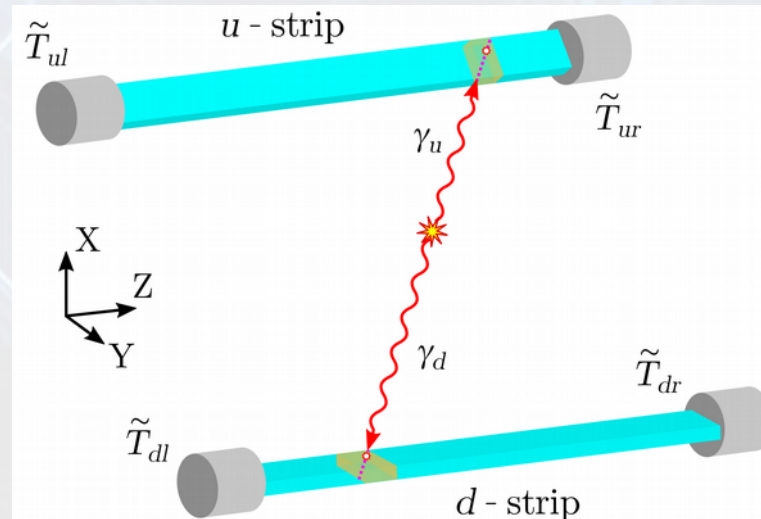
Outline

- **Motivation**
 - Time-of-flight resolution of J-PET
 - Real time image reconstruction
 - Continuous geometry (no bins)
- **Single-event TOF FBP**
 - Convolution vs sum: FBP in image space
 - Adding time-of-flight
 - 3D asymmetric kernel
 - Kernel optimisation
- **Results**
 - Spatial resolution of “big barrel”
 - Image quality for ideal J-PET
- **Summary and further plans**

Time-of-flight resolution of J-PET

In modern PET scanners, scintillation crystals (LSO:Ce, LYSO:Ce, LaBr₃:Ce) are capable of achieving **coincidence resolving time (CRT)** of ~ 100 ps. The lowest announced value is 249 ps, in fact **214 ps** (Biograph Vision scanner, Siemens) [ordineingegneripisa.it/obj/files/documenti/2018.2.23.10.46.26_834.pdf].

Plastic scintillators used in **Jagiellonian PET (J-PET)** are superior time-wise, despite the worse efficiency: CRT of **70 ps** – for 1-meter strips [[Moskal P et al. PMB 2016](#)]. The main factor is the readout – photomultipliers (PMs) attached at each end of the strips: silicon PM (SiPM) or tube PM (PMT).

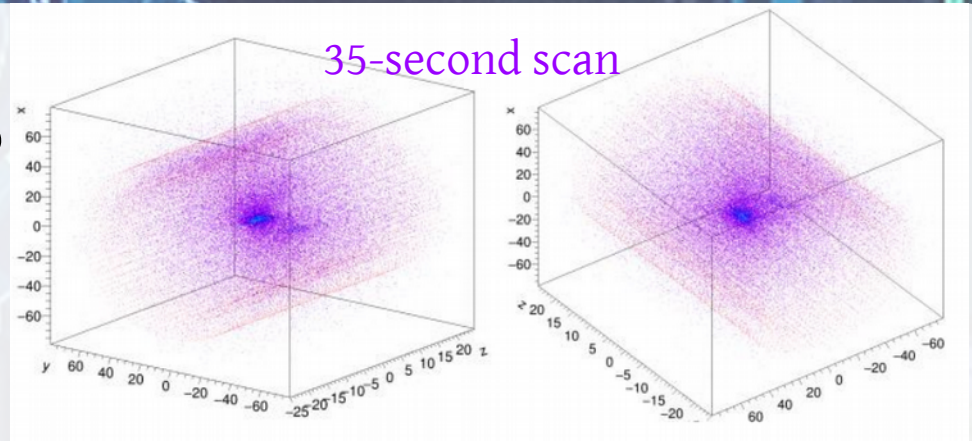


For **time-of-flight (TOF)** available and CRT below 100 ps analytical reconstruction methods may outperform iterative ones [[V Westerwoudt et al. IEEE Trans. Nucl. Sci. 2014](#)].

Real time image reconstruction

TOF reconstructions produce comparable results for much lower statistics compared to non-TOF methods. Matched with small CRT, it substantiates image reconstruction on the fly during real time scans.

A platform based on Field Programmable Gate Array (FPGA) System-on-Chip (SoC) **has already been implemented** for J-PET [G Korcyl et al. IEEE Trans. Med. Im. 2018]. It performs event building, filtering, coincidence search and so-called Region-Of-Response (ROR) reconstruction, yet without **filtered back projection (FBP)** – only coordinates of the reconstructed points in 3D space. *Newest solution (G.Korcyl presentation) – operate in projection space.*

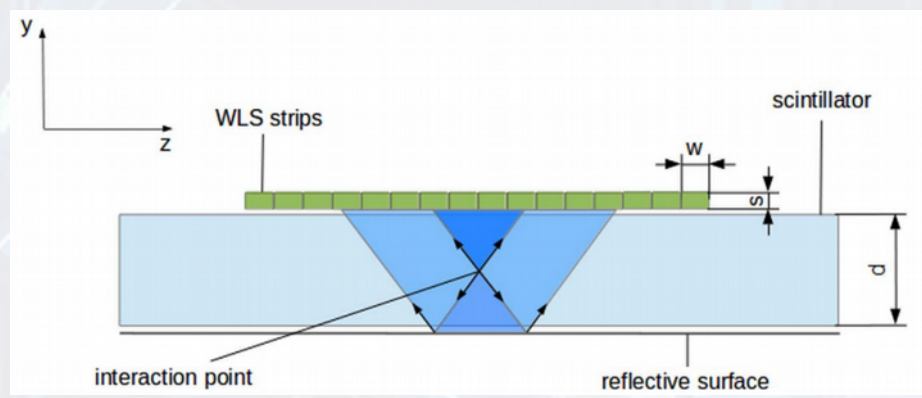


ROR implies that only small fraction of **field-of-view (FOV)** is processed for each event, hence it might be possible to add **Ramp/Hann/Hamming** filters.

Continuous geometry

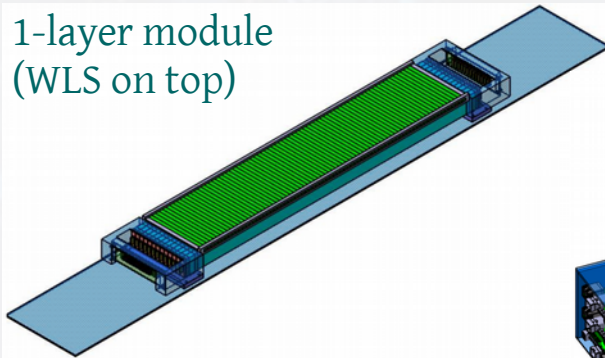
For non-TOF scanners, the measured 3D data of N detected emissions can be expressed as a set of projections: $\{\tilde{\mathbf{p}}_1, \tilde{\mathbf{p}}_2, \dots, \tilde{\mathbf{p}}_N\}$, $\tilde{\mathbf{p}}_k = (s, \phi, \zeta, \theta)_k$. The construction of a scanner defines discrete sets of s, ϕ, ζ, θ which form possible projection elements (bins). Adding TOF would expand it to \mathbb{R}^5 .

However, continuous strips in J-PET do not fix ζ and θ . Besides, there are prospects for partial **depth-of-interaction (DOI)** information be extracted with the array of **wavelength-shifting (WLS) strips** [J Smyrski et al., N. Instr. Meth. Phys. Res. A 2017].

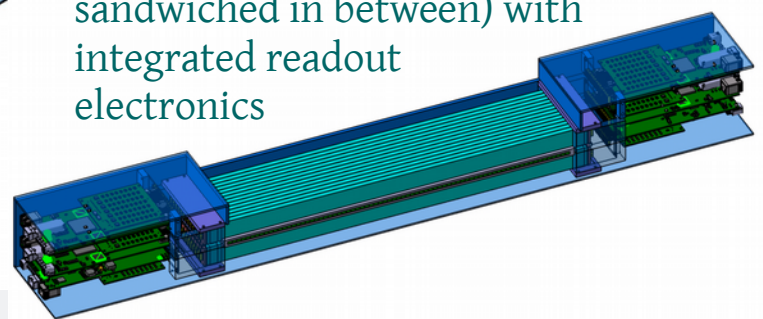


Using bins/projection space is unpractical!

1-layer module (WLS on top)



Prototype 2-layer module (WLS sandwiched in between) with integrated readout electronics



Single-event TOF FBP

Non-TOF FBP (2D): image $f(x, y) = (\mathcal{X}_{R_F}^* p^F)(x, y) = \int_0^\pi d\phi p^F(s = x \cos \phi + y \sin \phi, \phi)$

Projection (sinogram): $p^F(s, \phi) = \int_{-R_F}^0 ds' p(s', \phi) w(s - s')$

Convolution is used here only (w - Ram-Lak filter)

[DL Bailey et al., PET Basic Science, 2005]

3D TOF FBP (arbitrary voxel v): $f(v) = \int_\phi \int_\theta \int_\xi \sum_{i=1}^N \mathcal{F}^{-1}\{W(v_s) \mathcal{F}[p_i(s, \phi, \theta, \xi)]\} \cdot h(t - t_i)$
[Conti M et al. PMB 2005]

Forward and inverse Fourier transform, a filter in frequency domain $W(v_s)$, TOF kernel $h(t)$... **too cumbersome!**

The alternative: treat all lines-of-response (LORs) independently.

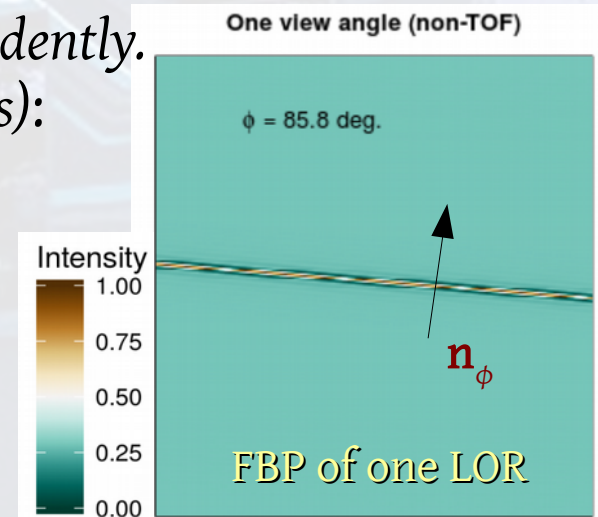
One LOR reflects **one point** on a sinogram, filtered by $w(s)$:

$$p_i(s, \phi) = \mathbf{1}|_{s=s_i, \phi=\phi_i} \rightarrow p_i^F(s, \phi) = w(s - s_i(\phi)),$$

All points for a fixed ϕ : $p^F(s, \phi) = \sum_i p_i^F(s, \phi)$,

$$f(x, y) = \sum_\phi \sum_i \mathbf{n}_\phi p_i^F(s = x \cos \phi + y \sin \phi, \phi),$$

\mathbf{n}_ϕ - vector that defines orientation.



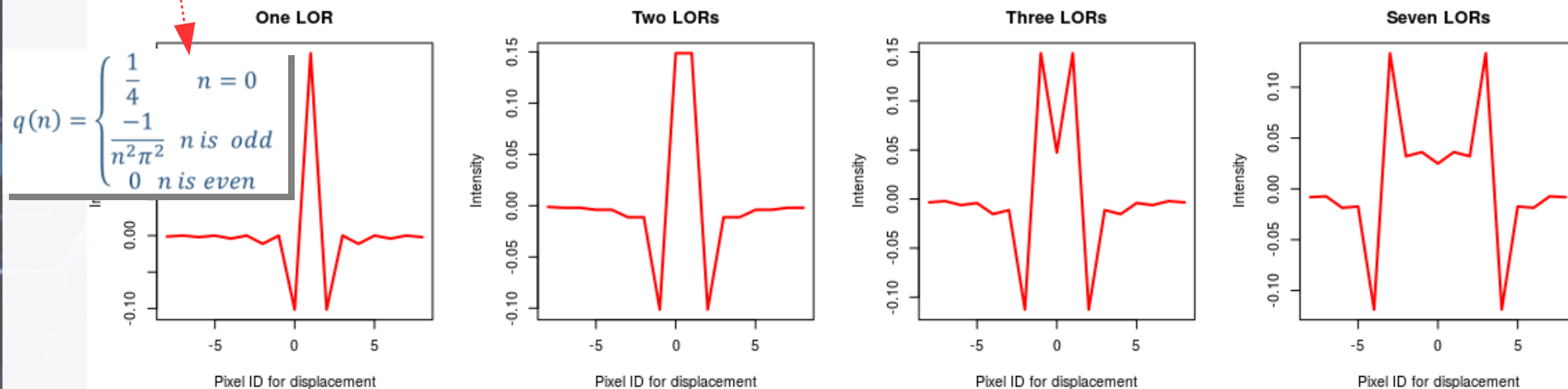
FBP in image space (2D)

Split the data into single-LOR backprojections first. We know how the reconstructed image looks like for one LOR. *The sum of such images for all LORs will reflect the Filtered Back Projection (FBP):*

$$f(x, y) = \sum_{\phi} \sum_i \mathbf{n}_{\phi} w(x \cos \phi + y \sin \phi - s_i(\phi))$$

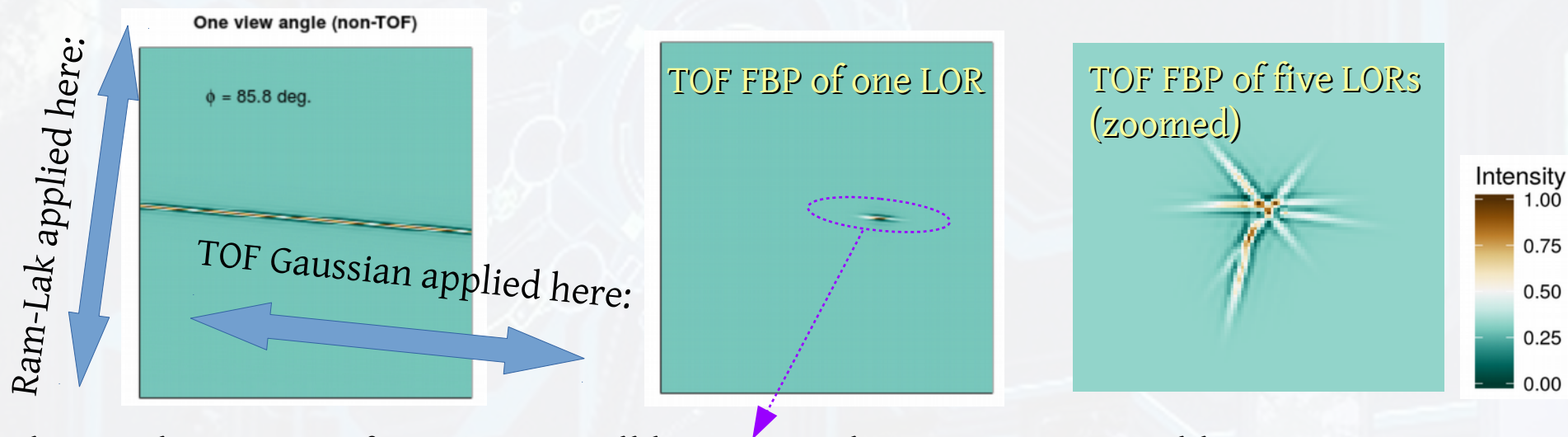


Options for filters: define analytically in image space (easy for Ram-Lak), use a table/polynome or create one image and move/rotate.



Adding time-of-flight (2D)

Apply Gaussian kernel (with standard deviation σ_{TOF} calculated from coincidence resolving time, CRT), centered at the point estimated from TOF:



The resulting map of intensities will be **sparse**, hence it is reasonable to restrict/truncate the known outcome – defined analytically or as an image.

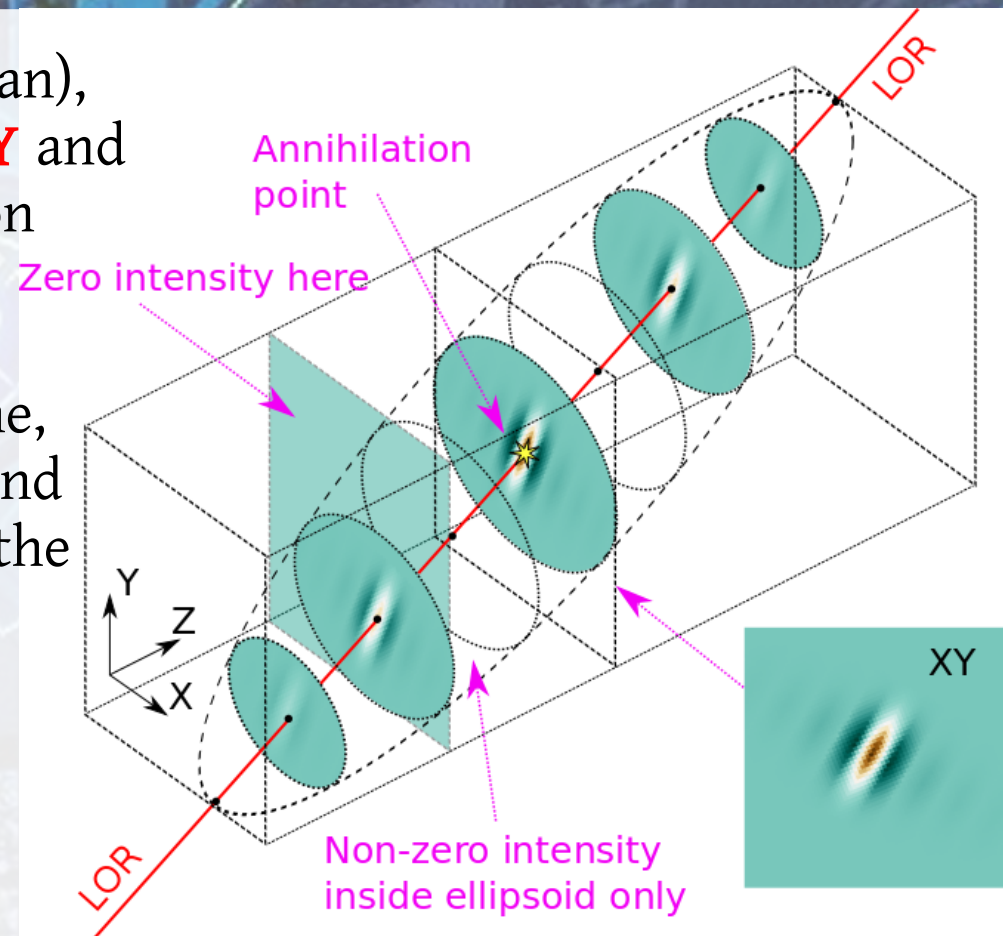
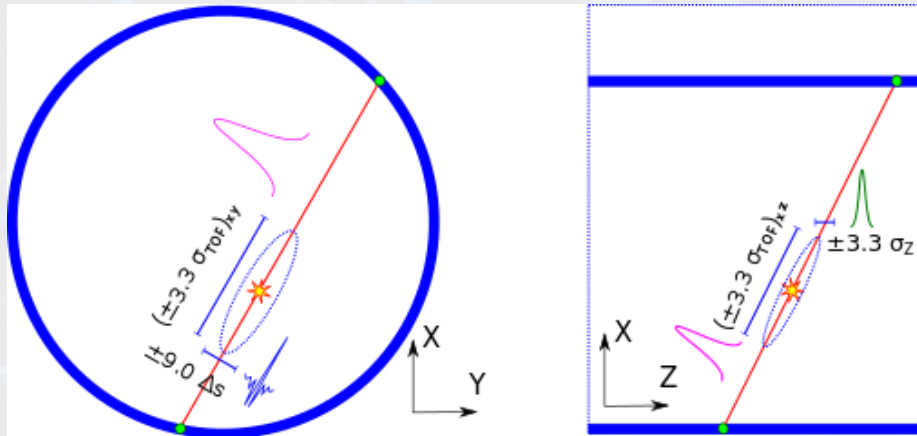
The narrower the TOF kernel is, the less memory is needed to update the result: the fastest will be the combination of SiPM/WLS, the slowest – PMT readout.

“Old” LORs *could be dropped* – **a time window for real time imaging is possible!**

3D asymmetric kernel

Apply **TOF kernel along LOR** (Gaussian),
Ram-Lak filter normal to LOR in XY and
 a **small Gaussian along Z** (depends on
 the distance between slices).

Update intensity within a small volume,
 limited by at least $\pm 3.3\sigma$ for Gaussian and
 $\pm 9.0\Delta s$ for Ram-Lak (Δs - sampling for the
 displacement s in projection space).



*The volume of the ellipsoid is
 much smaller than the whole FOV,
 resembles ROR in FPGA solution.*

Kernel optimisation

There are distinct similarities with multivariate **kernel density estimation (KDE)**, *applied to annihilation positions, estimated directly from TOF*.

For a d -dimensional dataset $\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_n$ of the size n , the KDE is:

$$\hat{f}_{nH}(\mathbf{x}) = n^{-1} \sum_{i=1}^n |\mathbf{H}|^{-1/2} K[\mathbf{H}^{-1/2}(\mathbf{x} - \mathbf{X}_i)]$$

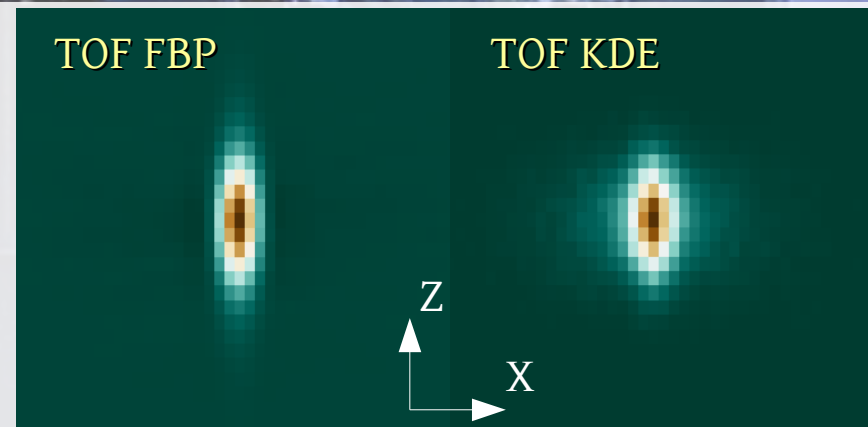
$\mathbf{x} = (x_1, x_2, \dots, x_d)$, $K(\cdot)$ – **spherically symmetric** multivariate kernel (e.g. Gaussian), \mathbf{H} – **bandwidth matrix**, symmetric and positive definite. Its choice is crucial!

There are lots of algorithms for bandwidth selection, e.g.:

- asymptotic approximation mean integrated squared error (AMISE)
- plug-in bandwidth selector (multistage) [Chacon JE et al. Test 2010]

Elements of matrix $\mathbf{H} \ll \sigma_{\text{TOF}} < \sigma_z!$

It is reasonable to optimise Gaussian kernels to smaller sigmas, otherwise it imposes additional smearing along Z.
(example for 1-mm source, ideal scanner, [Kowalski P et al. PMB 2018])



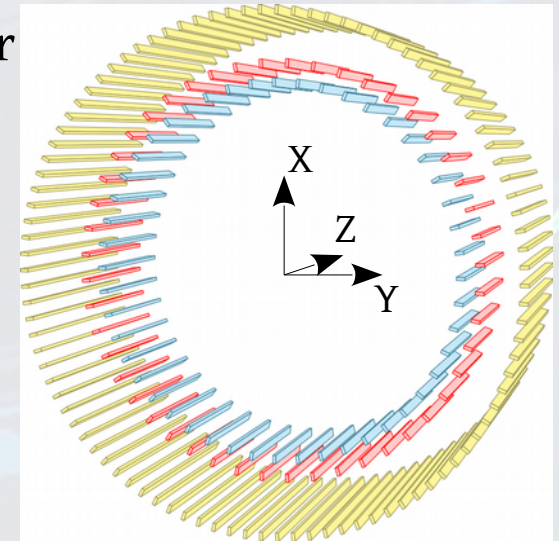
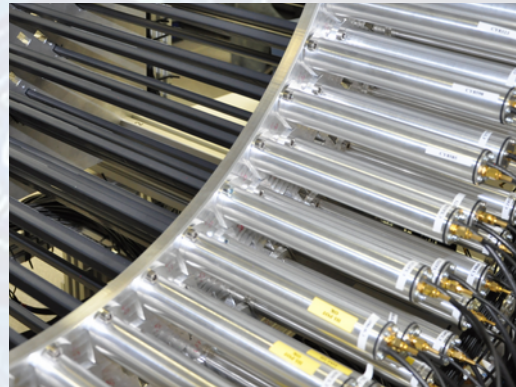
Results: spatial resolution

Jagiellonian PET (“big barrel”): 3 layers of plastic scintillator strips, 192 detector strips of the size: 7 mm × 19 mm × 500 mm

Radii:

- 425.0 mm (48 strips)
- 467.5 mm (48 strips)
- 575.0 mm (96 strips)

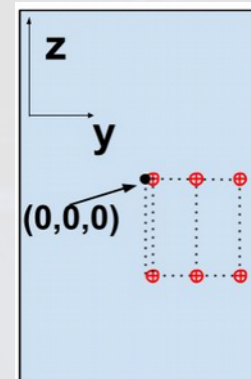
Gaps between strips are dictated by the size of PMT readouts.



The data:

- **simulations** made in GATE for 1-mm spherical NEMA source (370 kBq), at six positions ($x = 1 \text{ cm}/10 \text{ cm}/20 \text{ cm}$, $z = 0/18.75 \text{ cm}$), 100,000 coincidences per one set.
- **early experiment (Run-4)**, the source size may differ from NEMA + higher activity, placed at ($y = 1 \text{ cm}/10 \text{ cm}/20 \text{ cm}$, $z = 0/-18.75 \text{ cm}$), 150,000 events taken from each measurement.

Courtesy of **Monika Pawlik-Niedźwiecka**



Data: single ^{22}Na source placed in positions according to NEMA

Time of measurement for single position: 3 hours

Activity of source: $\sim 1134 \text{ kBq}$

Results: spatial resolution

Truncation problem and scanner sensitivity: sparse multilayer transverse geometry of “big barrel” implies non-uniform sensitivity, while “*total body*” size (50 cm) diminishes the truncation effects.

Re-projection was not used as in *STIR framework* [K Thielemans et al., PMB 2012], but sensitivity map was generated using hybrid 2D+2D approach: Monte Carlo simulation for XY plane and analytical estimation for XZ plane, based on the work [A Strzelecki Ph.D. dissertation PAN, Warsaw 2016].

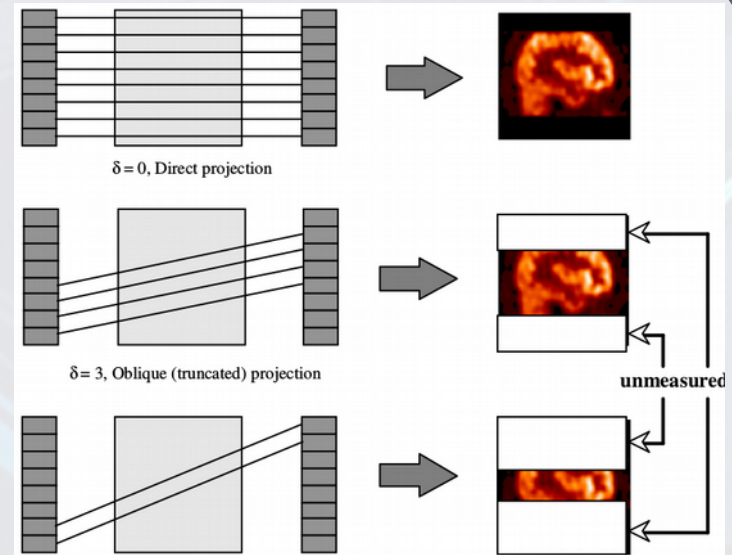
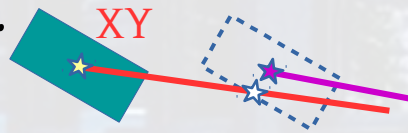
Reconstructed image (for a voxel v)

$$f_{\text{true}}(v) = f(v)/s(v),$$

$s(v)$ – sensitivity matrix.

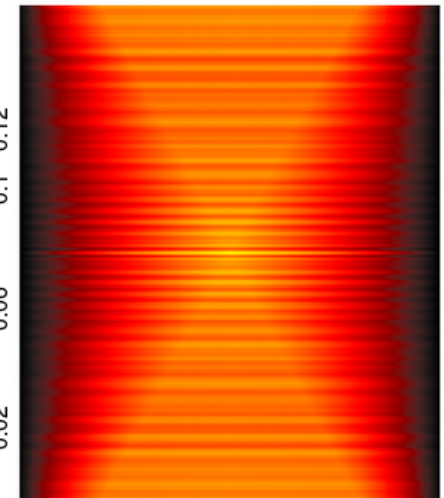
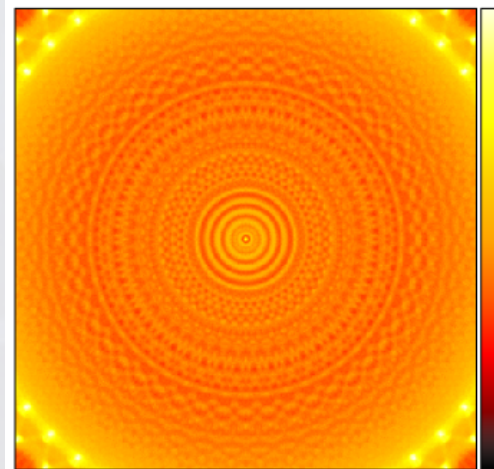
Reference images:

FBP 3DRP (STIR) – needs hit remapping onto a single layer ($R = 43.73$ cm, 384 strips)
TOF KDE – no filters, symmetric 3D kernel



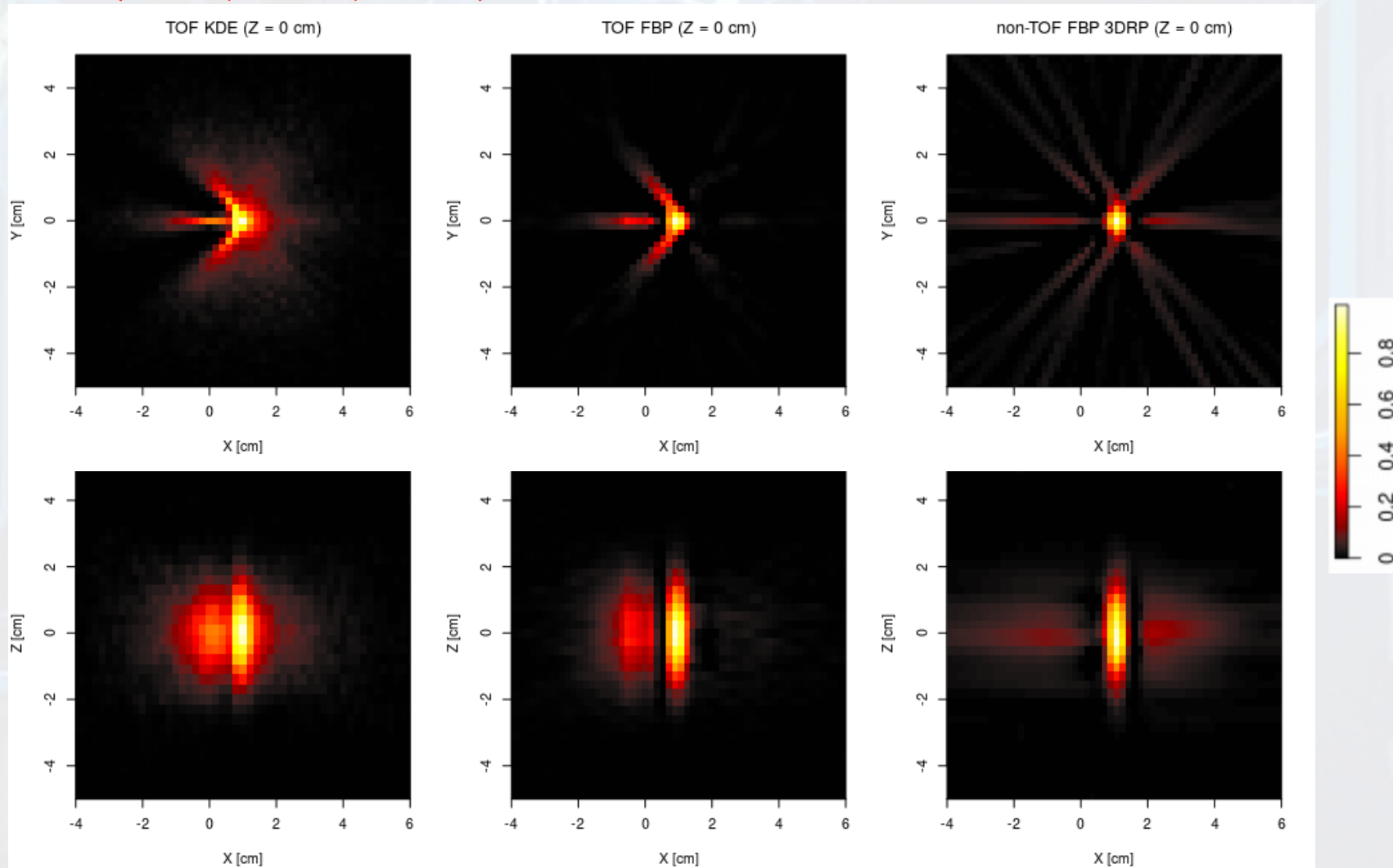
XY ranged [-35.2 cm, 35.2 cm]

XZ (whole FOV)



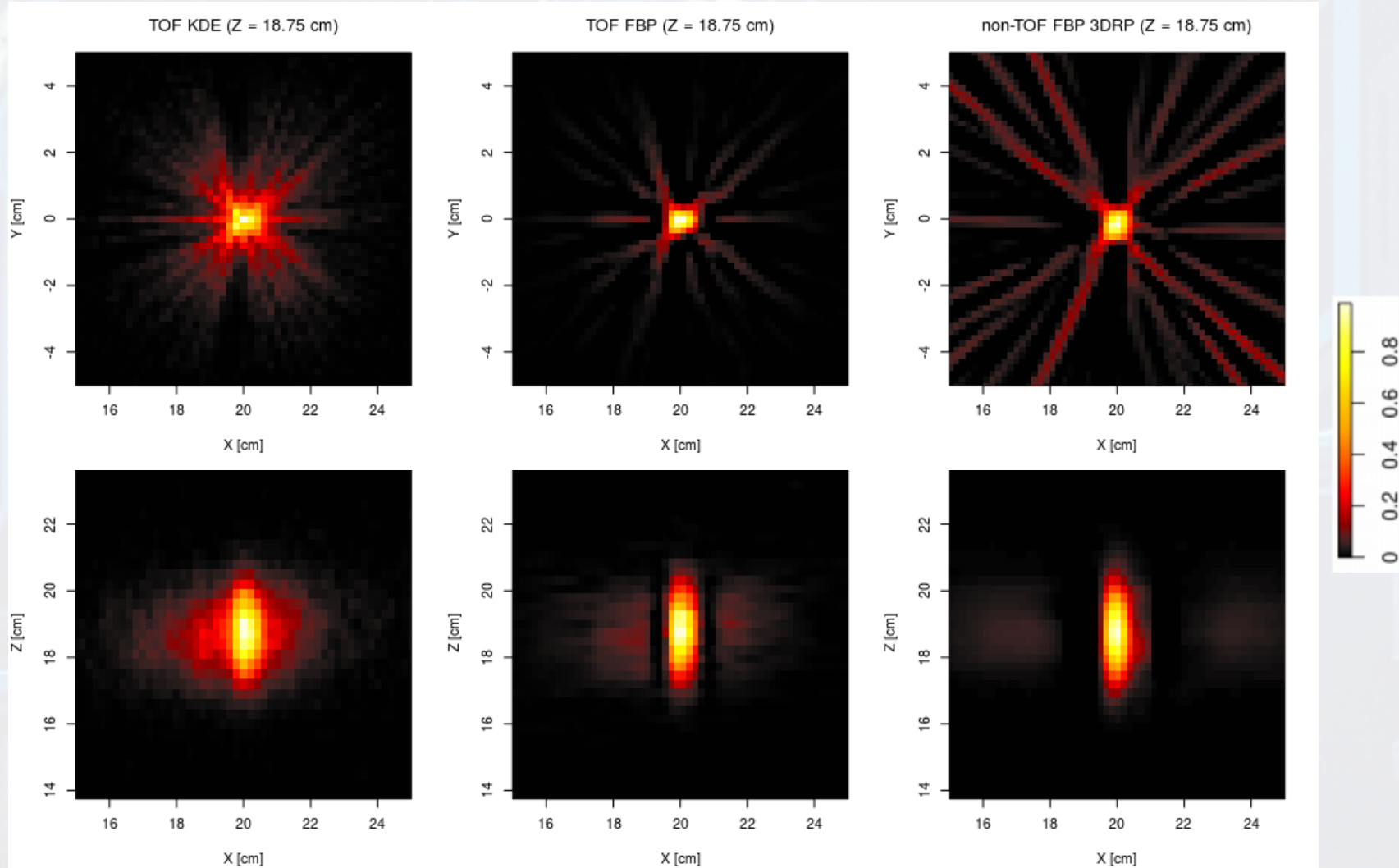
Results: spatial resolution

3D reconstructions by TOF KDE, TOF FBP and FBP 3DRP (STIR) for the *simulated* source at (1 cm, 0 cm, 0 cm), PMT readouts, ~100,000 events:



Results: spatial resolution

3D reconstructions by TOF KDE, TOF FBP and FBP 3DRP (STIR) for the *simulated* source at (20 cm, 0 cm, 18.75 cm), PMT readouts, ~100,000 events:



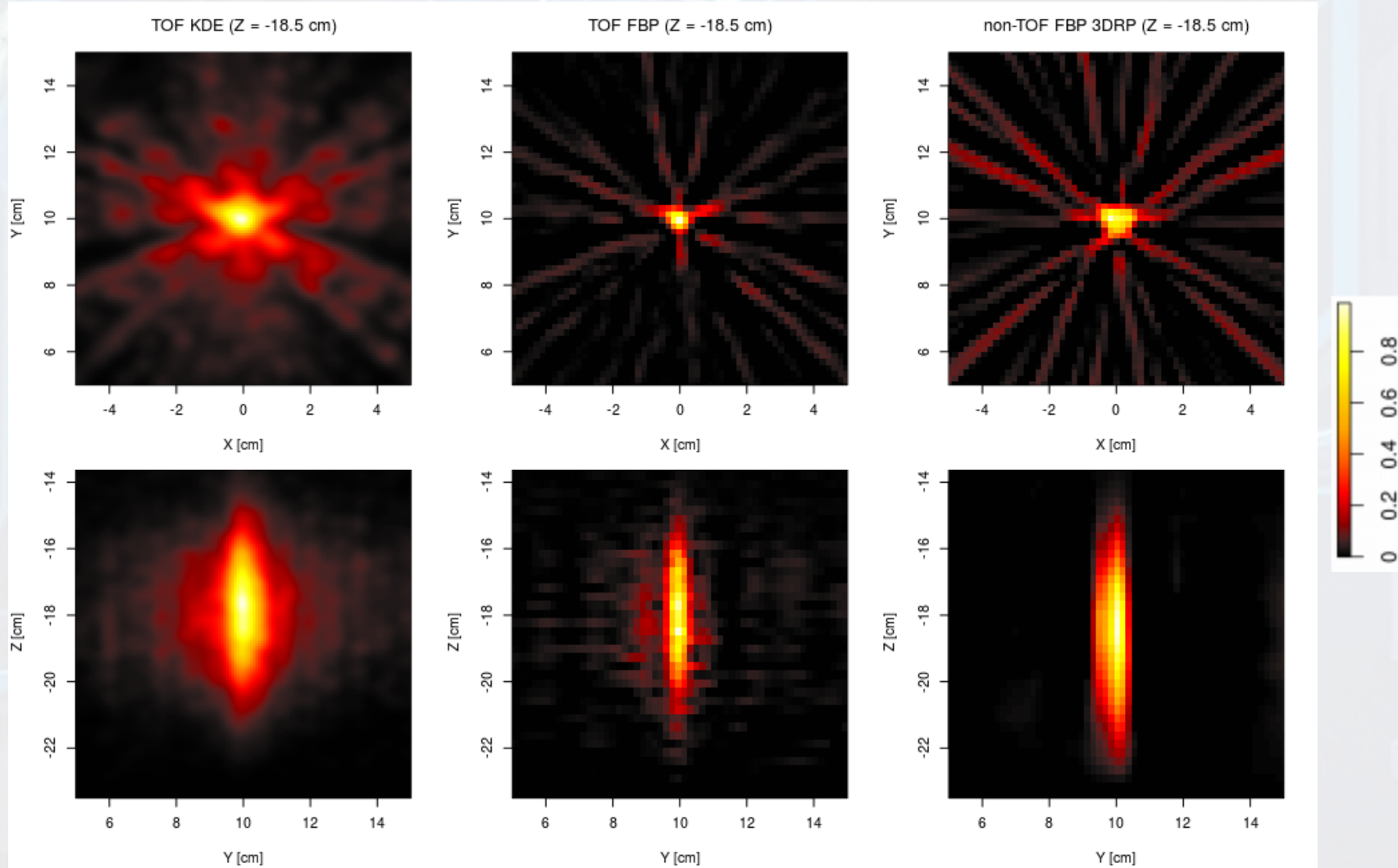
Results: spatial resolution

GATE simulations of 1-mm spherical source (370 kBq) – NEMA standard, PMT readout, for the accuracy of 0.5 mm. Voxel size 1.8 mm × 1.8 mm × 2.6 mm

Readout:		PMT		
Source at: ($y_{\text{src}} = 0$ cm)	Algorithm	FWHM (in mm) along axis		
		X	Y	Z
3-layer scanner ("big barrel") : $R = 42.5/46.75/57.5$ cm, 48/48/96 strips of the dimension $7 \times 19 \times 500$ mm, 100,000 events per simulation				
$x_{\text{src}} = 1$ cm $z_{\text{src}} = 0$ cm	FBP 3DRP	4.5	7.0	20.0
	TOF KDE	5.5	6.0	20.0
	TOF FBP	5.0	6.5	20.5
$x_{\text{src}} = 10$ cm $z_{\text{src}} = 0$ cm	FBP 3DRP	5.0	7.0	20.0
	TOF KDE	–	–	–
	TOF FBP	5.5	5.0	20.0
$x_{\text{src}} = 10$ cm $z_{\text{src}} = 18.75$ cm	FBP 3DRP	5.5	7.5	20.5
	TOF KDE	–	–	–
	TOF FBP	5.5	5.0	20.0
$x_{\text{src}} = 20$ cm $z_{\text{src}} = 18.75$ cm	FBP 3DRP	6.5	7.5	21.0
	TOF KDE	7.5	6.0	22.0
	TOF FBP	7.0	5.5	18.0

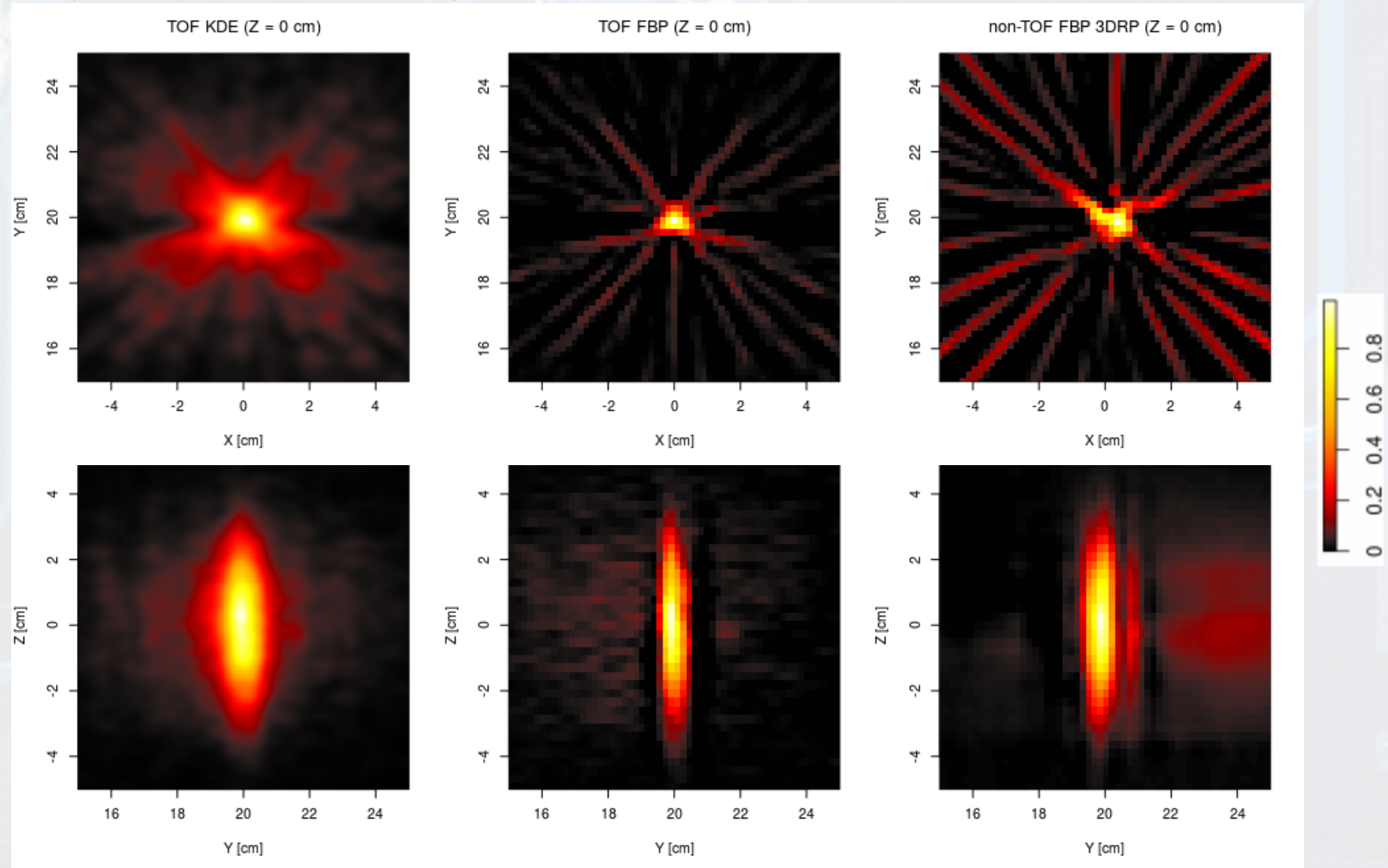
Results: spatial resolution

3D reconstructions by TOF KDE, TOF FBP and FBP 3DRP (STIR) for the *measured* source (0 cm, 10 cm, -18.5 cm?), PMT readouts, ~150,000 events:



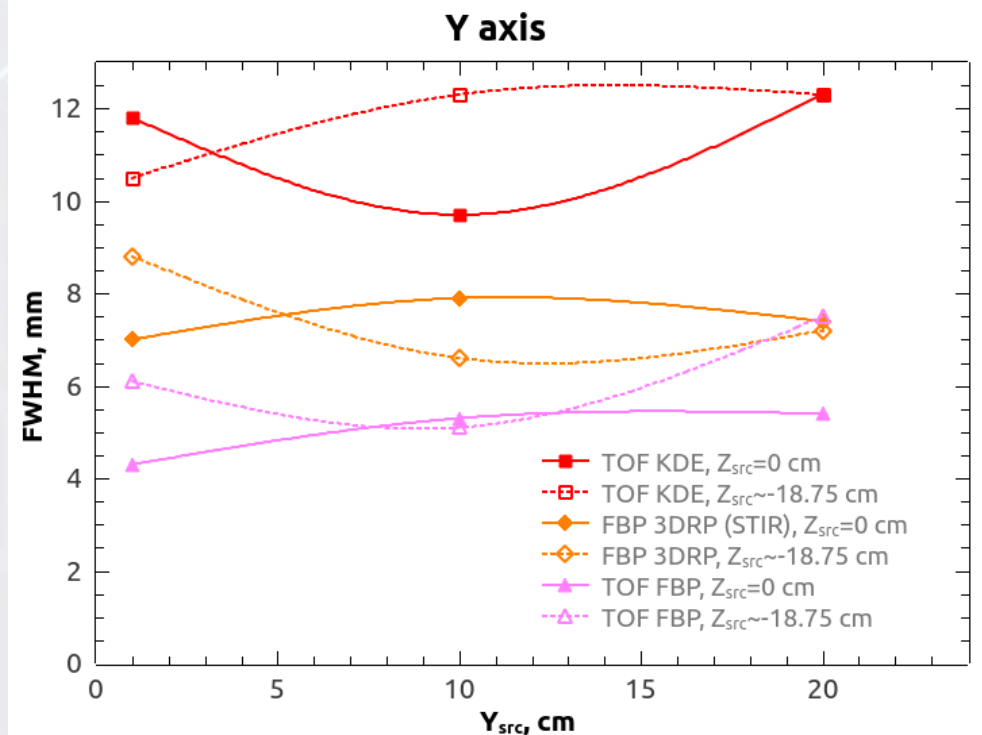
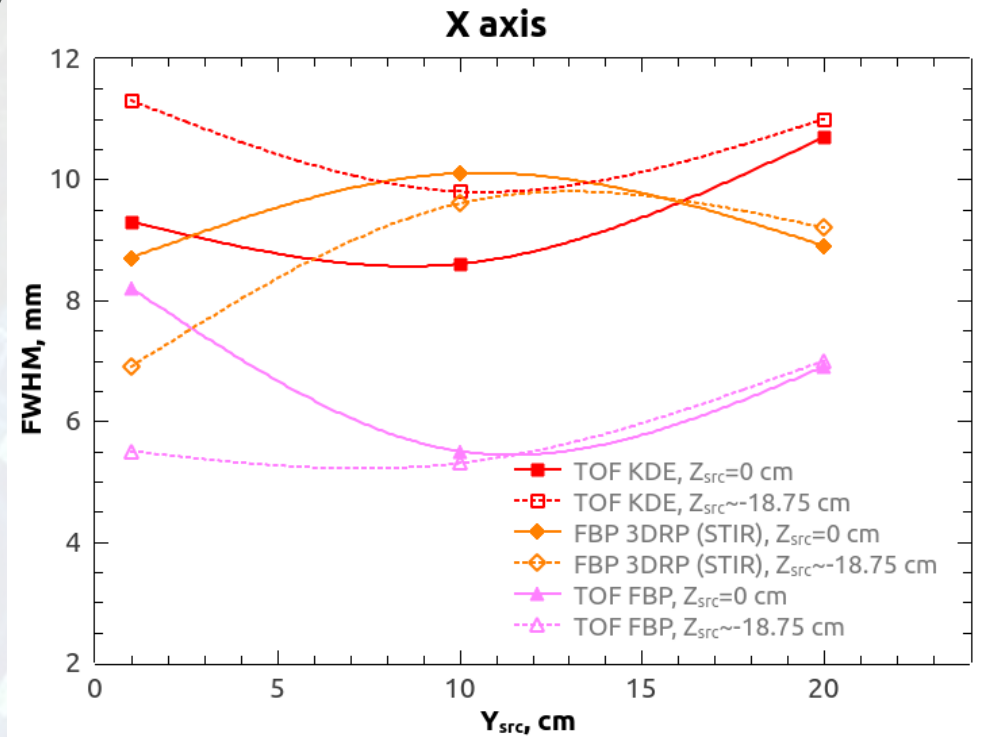
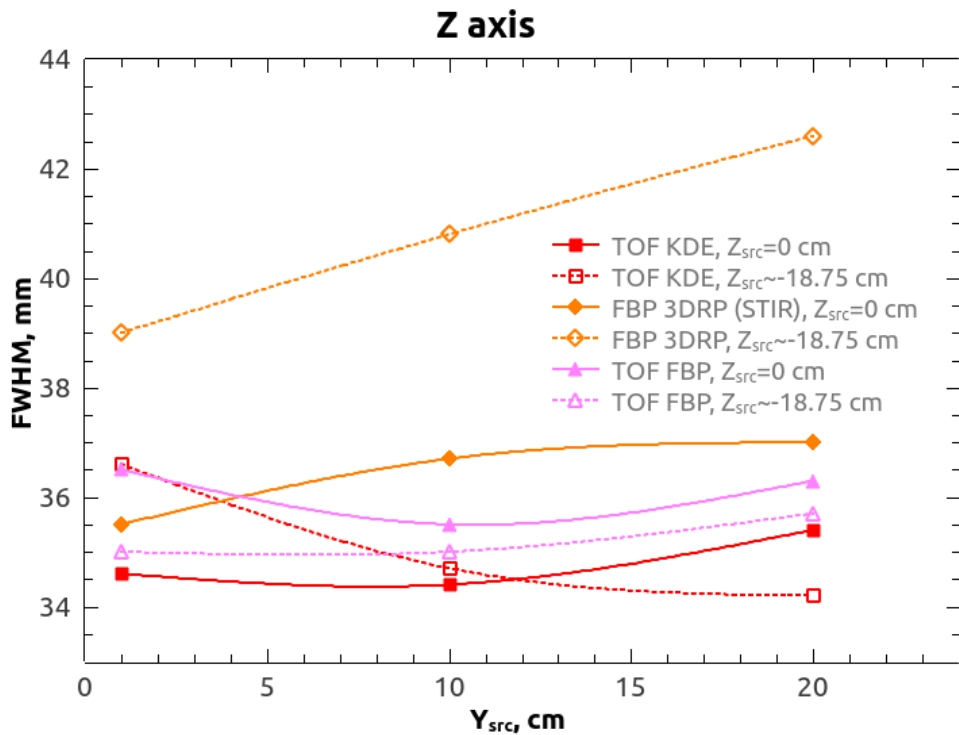
Results: spatial resolution

3D reconstructions by TOF KDE, TOF FBP and FBP 3DRP (STIR) for the *measured* source (0 cm, 20 cm, 0 cm), PMT readouts, ~150,000 events:



Results: spatial resolution

PSF (FWHM) values for Z axis are systematically lower if compared with the simulated data (*early experiment*)
TOF FBP is better than FBP 3DRP, but in order to use STIR, all hits were remapped onto 1-layer (384 strips)



Results: image quality

IEC NEMA phantom, simulated in GATE (at the centre of the scanner, one long measurement (3000 s), filtered by *true coincidences* only (data size 10-20 mln.)

Ideal geometry: 384 strips, R=43.73 cm, **SiPM** (CRT=235 ps)

Attenuation correction was added to TOF FBP: each LOR is treated as a projector, attenuation path is estimated based on Siddon algorithm (computing the intersecting length of a ray with each voxel) [R Li et al., Journ. Comp. Sci. 2010]:

Update intensity for each LOR as

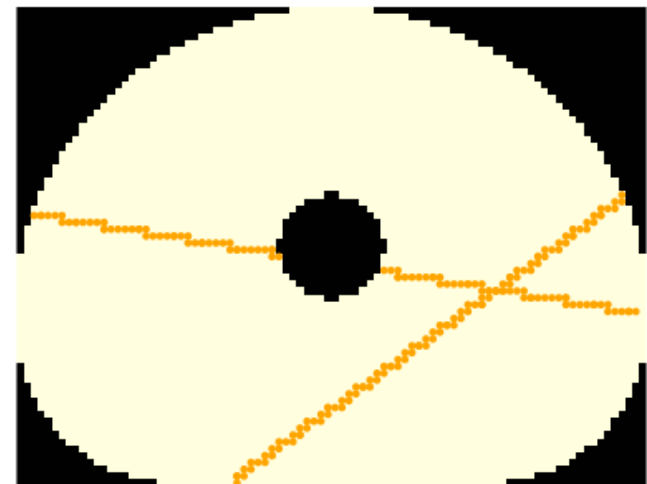
$$I = I_0 \exp(-\mu x), \text{ where } \mu^{\text{PET}}(\text{H}_2\text{O}) = 0.096 \text{ cm}^{-1}.$$

Attenuation map was created comprising all phantom volume filled with radioactive liquid, but without cold spheres and capillaries.

This upgrade extends reconstruction time by **less than 10%** (still possible for real time).



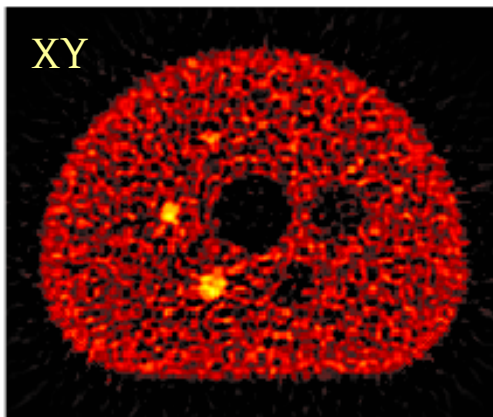
Attenuation map (XY)



Results: image quality

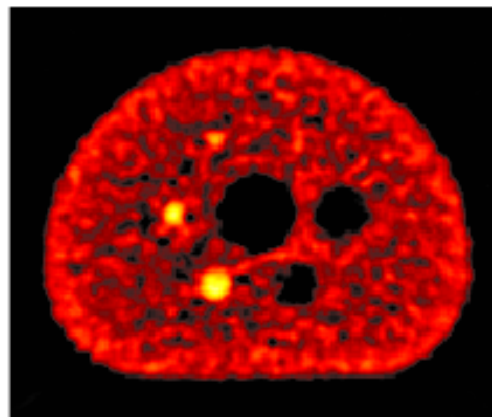
10 mln. true coincidences, results for FBP 3DRP are obtained by P. Kopka (see poster). **Contrast recovery coefficient (CRC)**, **background variation (BV)** and **signal-to-noise ratio (SNR)** were estimated for 13-mm and 22-mm spheres.

TOF FBP (Ram-Lak)



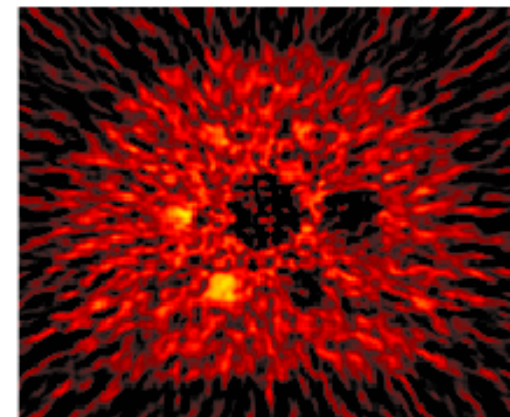
13 mm: CRC=0.49, BV=0.17, SNR=9.0
22 mm: CRC=0.82, BV=0.09, SNR=26.6

TOF FBP (Hamming)

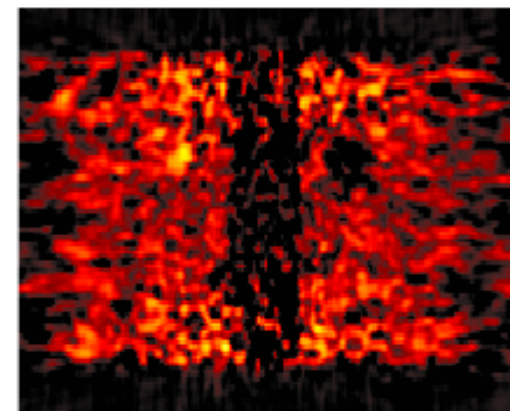
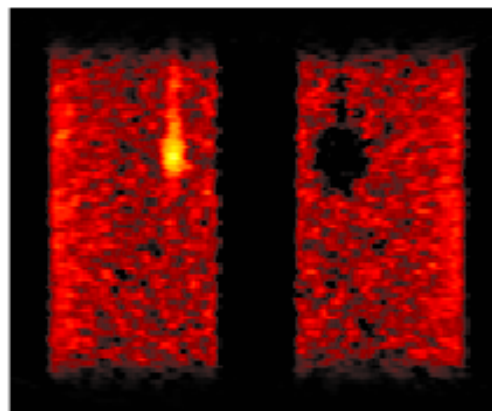
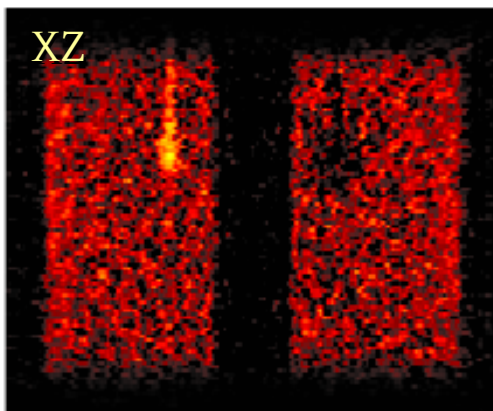


CRC=0.48, BV=0.16, SNR=9.1
CRC=0.94, BV=0.10, SNR=29.8

non-TOF FBP 3DRP



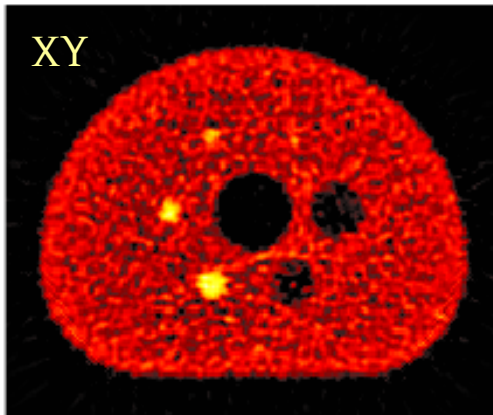
CRC=0.32, BV=0.27, SNR=3.5
CRC=0.77, BV=0.18, SNR=13.0



Results: image quality

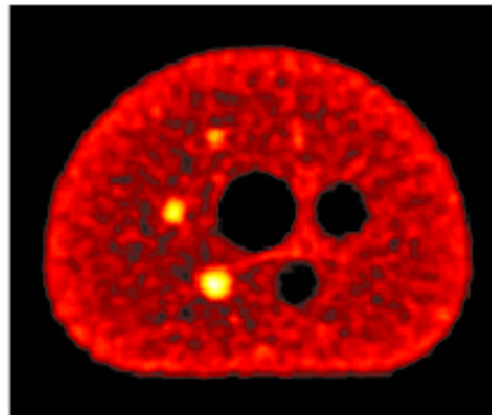
20 mln. true coincidences, results for OSEM (STIR) are obtained by P. Kopka.
Full-sized TOF kernel was used for all images.

TOF FBP (Ram-Lak)



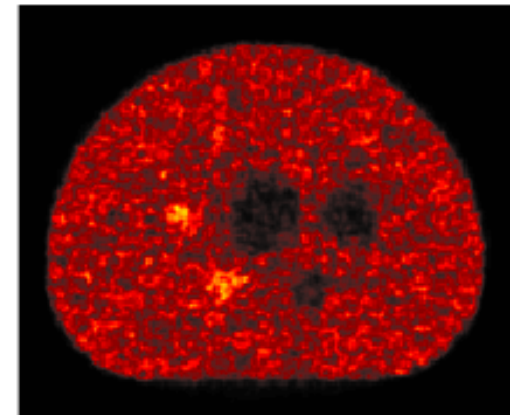
13 mm: CRC=0.43, BV=0.09, SNR=15.1
 22 mm: CRC=0.75, BV=0.05, SNR=41.2

TOF FBP (Hamming)

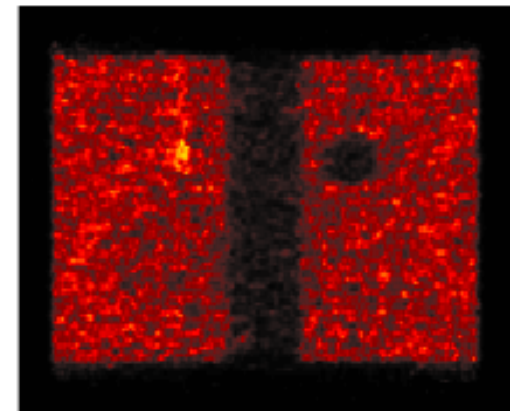
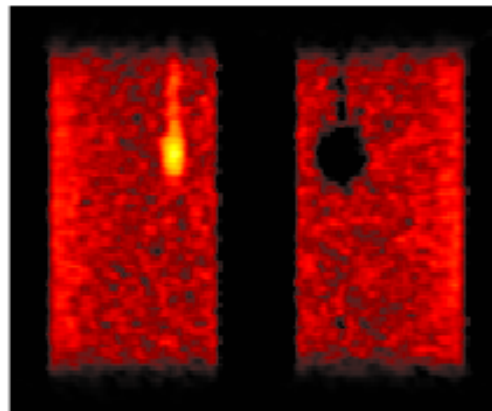
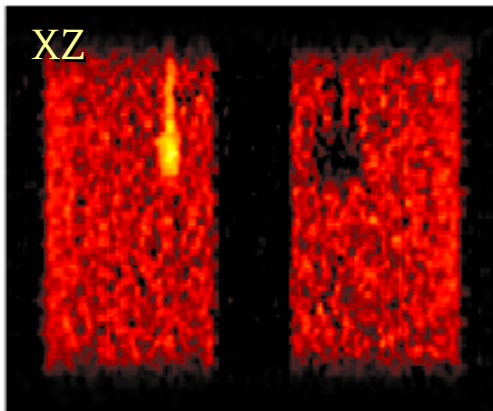


CRC=0.47, BV=0.10, SNR=14.5
 CRC=0.92, BV=0.06, SNR=44.2

non-TOF OSEM (25th Iter.)



CRC=0.14, BV=0.16, SNR=2.6
 CRC=0.43, BV=0.10, SNR=13.1




Summary and further plans

- TOF based reconstructions are promising solution for J-PET due to the excellent temporal resolution and shorter scan times if compared to non-TOF.
- There is a need for the specific algorithms due to the complex geometries of J-PET (sensitivity map is essential), the continuous character of strips (hence TOF), along with the eventual DOI information, estimated by WLS. Using bins in projection space is unpractical.
- TOF FBP could be employed using filters defined in image space, applied directly to each LOR as three separate kernels in event-by-event way. Scalability of this process opens up a possibility for real time imaging (already built and tested for non-filtered reconstructions using FPGA), since the intensity should be updated only for the small fraction of voxels.
- Imposing Gaussian kernels along LOR and Z-axis would blur the image thus reducing spatial resolution. The process of optimisation may differ from the bandwidth selection for TOF KDE, because 3D kernel is not symmetric.
- Single-event based TOF FBP achieve similar or better results for spatial resolution and image quality, compared to non-TOF reconstructions from STIR and non-filtered TOF KDE.

Yet to resolve:

- Explore the ways to optimise the parameters for asymmetric TOF FBP kernel, as well as the choice of optimal filter/cut-off frequency (apodisation).
- Compare the results for TOF FBP with other TOF based algorithms (MLEM, TV etc).
- Analyse performance benchmarks



Thank You for Your attention!