

PET Image Reconstruction Cluster at Turku PET Centre

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1 Background

- Image Reconstruction from Projections
- High Resolution Research Tomograph
- Reconstruction Problem on the HRRT

2 Method

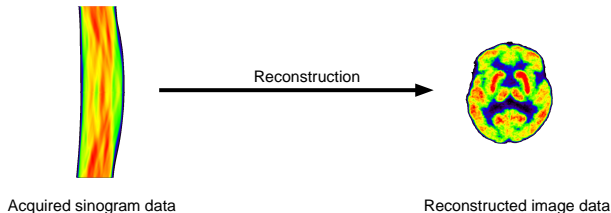
- Parallel Computing
- Parallel 3D-OSEM Reconstruction
- Reconstruction Cluster at Turku PET Centre

3 Results

4 Discussion

Image Reconstruction from Projections.

- General reconstruction problem: transform the acquired sinogram data to human readable image data



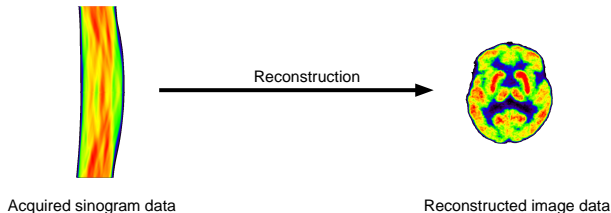
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$$g(s, \theta) \triangleq \mathcal{R}f = \iint_{\mathbb{R} \times \mathbb{R}} f(x, y) \delta(x \cos \theta + y \sin \theta - s) dx dy.$$

- The inverse Radon transform is estimated with analytical (e.g. 2D/3D FBP) and statistical (e.g. 2D/3D OSEM) methods

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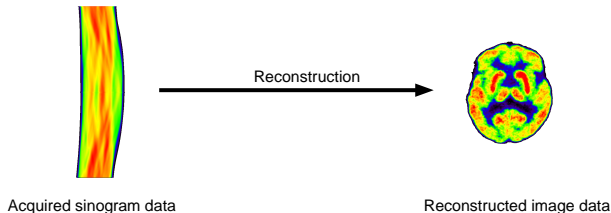
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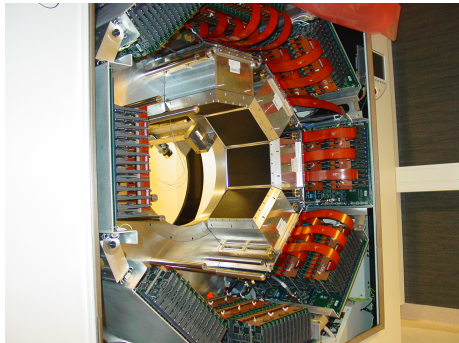


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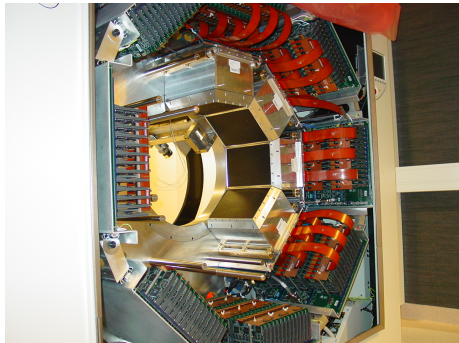
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The High Resolution Research Tomograph (HRRT).



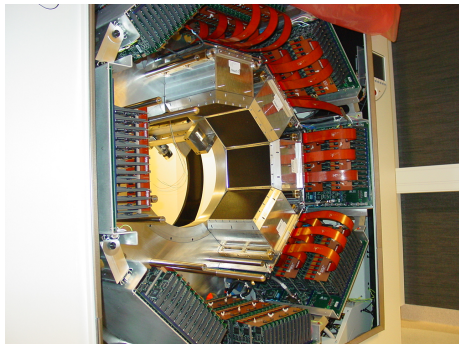
- 3D acquisition (no septa)
- Octagonal design
 - ▶ 8 flat panels
 - ▶ 9×13 blocks in each panel
 - ▶ 8×8 crystals in each block
 - \Rightarrow 119808 detectors
 - $\Rightarrow 4.5 \times 10^9$ LORs
- FOV
 - ▶ length 25.2 cm
 - ▶ head to head distance 46.9 cm

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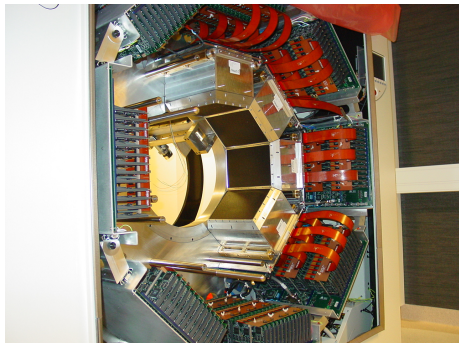
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Reconstruction Problem on the HRRT.

Challenges:

- Data size (in normal mode i.e. RD 67 and span 9)
 - ▶ 2209 sinograms with 256 radial elements and 288 views
 $\approx 50\times$ the data size of the GE Advance in 2D mode
 - ▶ 311MB/frame of emission data
 - ▶ 52MB/frame of image data
- Reconstruction times of span 9 data on a standard processor PC

Per frame calculations	SCF calculation	FORE + 2D-OSEM	3D-OSEM
Time	$\sim 8\text{min}$	$\sim 15\text{min}$	$\sim 3\text{hrs } 30\text{min}$

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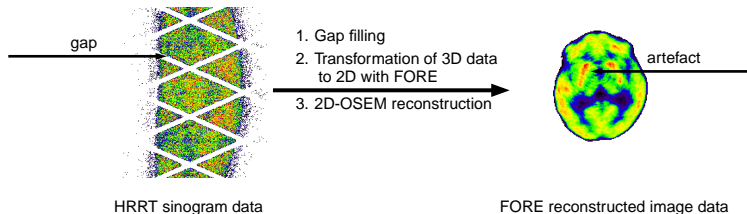
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Reconstruction Problem on the HRRT: Conclusion.

Challenges (continued):

- Octagonal geometry introduces gaps in the sinogram data



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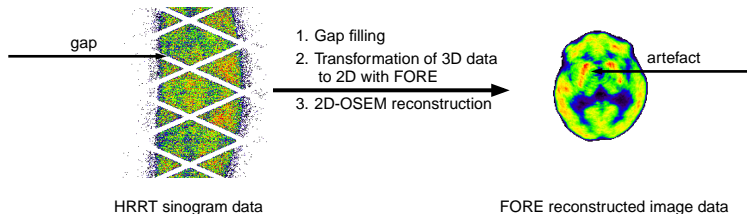
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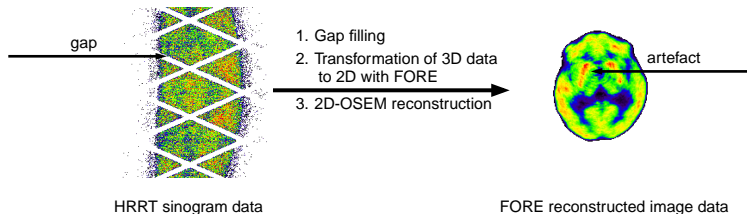
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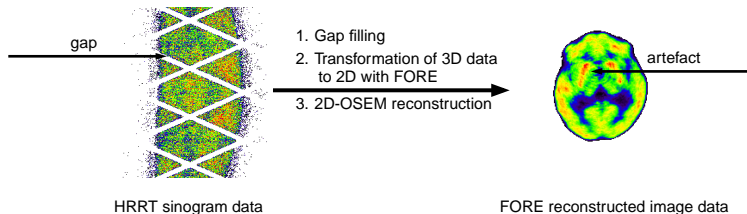
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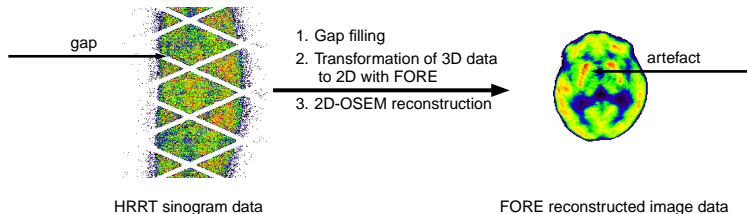
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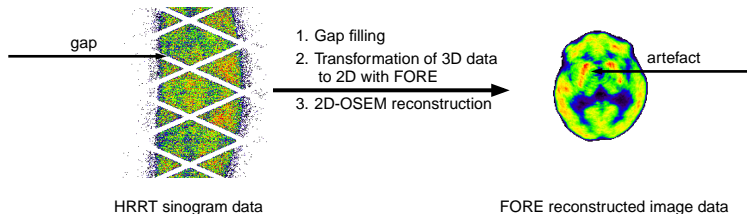
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Description of Parallel Computing.

A computer system in which interconnected processors perform concurrent or simultaneous execution of two or more processes.

Methodical design of parallel programs

- Partitioning
 - ▶ decomposing the computation and the data describing the problem into small tasks
- Communication
 - ▶ defining the communication that is necessary between the tasks in order to solve the problem
- Agglomeration
 - ▶ grouping the tasks together to larger units to achieve sufficient locality in the tasks
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 - ▶ deciding how tasks are placed onto physical processors

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Parallelization of the 3D-OSEM Reconstruction.

CTI Solution: Two designs.

1. Projection Space Decomposition (PSD)

● Partitioning and agglomeration

- ▶ each node has a complete local copy of the image volume
- ▶ each node has $1/N^{th}$ of the projection space data
- ▶ computation is decomposed in the projection space
→ computations require no interprocessor communication

● Communication

- ▶ local copy of the image volume is synchronised with **all** the other nodes in the end of every subset; i.e. $52\text{MB} \times (N - 1)$ packets to send and receive

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Parallelization of the 3D-OSEM Reconstruction (cont.).

2. Image Space Decomposition (ISD)

- Partitioning and agglomeration

- ▶ a round-robin method assigns parts of size P of the image and projection space data to each node
- ▶ computation is decomposed in the image space
→ part of the computation (back projection) is done only for the assigned part of data

- Communication

- ▶ interprocessor communication with $G \ll N$ nearest nodes is required after each computation step (forward projection)
- ▶ local copy of the image volume is synchronised with G nodes in the end of every subset

Comparison: the performances of the two models are indistinguishable.

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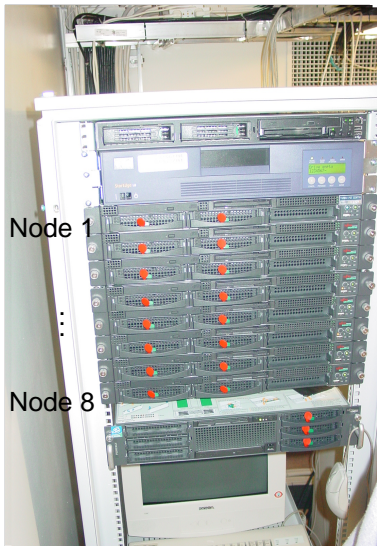
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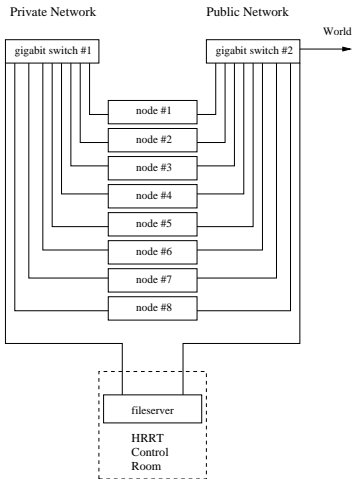
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Reconstruction Cluster at Turku PET Centre (Hardware).



- 8 Fujitsu-Siemens RX200 Primergy computing nodes
 - ▶ 2 Intel Xeon 3.06GHz CPUs
 - ▶ 2GB of RAM
 - ▶ 2 mirrored 36GB SCSI disks
 - ▶ 2 gigabit ethernet interfaces
- 1 Sun Microsystems SE3510 700GB disk array
- 2 gigabit ethernet switches

Reconstruction Cluster at Turku PET Centre (Software).



● Cluster

- ▶ Operating system Windows XP Pro
- ▶ Parallel 3D-OSEM reconstruction software by CTI; implements ANW and ordered poisson (OP) weighting

● Fileserver

- ▶ Operating system Windows XP Pro
- ▶ Reconstruction toolset with GUI (CTI)

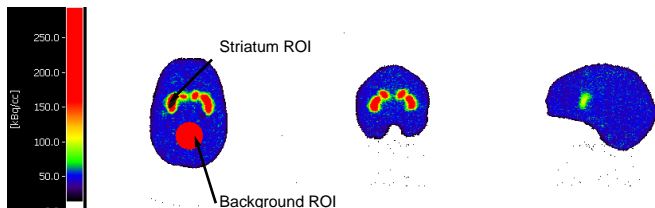
Assignment of reconstruction jobs and queuing to the cluster is embedded in the CTI reconstruction tool.

Results: speedup.

Reconstruction time is reduced to ~ 10 minutes per core iteration, i.e.
 ~ 30 minutes per frame.

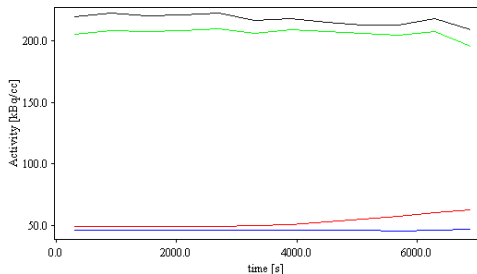
- The measured speedup of 7 correlates nicely to the number of cluster nodes 8
- A 16 node cluster has been tested at CTI, and shown to be capable to almost linear speedup

Results: images.

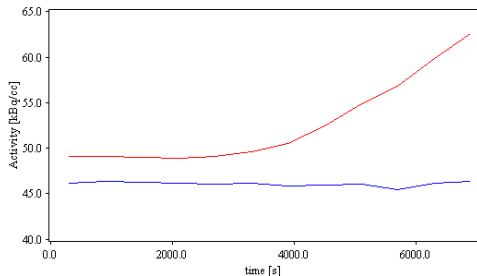


- First calibrated cluster images
- Striatum phantom filled with ^{11}C , 21 MBq in the background and 147 MBq in the structures
→ 47.5 (kBq/cc) in the background and 218.7 (kBq/cc) in the structures measured with Wizard
- Two hour scan histogrammed to 12, 10 minute frames
- Reconstructed with parallel 3D-OSEM with ANW and OP weighting schemes, both with 2 iterations and 16 subsets

- Time Activity Curve: Striatum (ANW 2ITR 16SS)
- Time Activity Curve: Background (ANW 2ITR 16SS)
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- Activity, both in the striatum and background are slightly overestimated
- OP-OSEM uses smoothing prior to randoms correction → performs better in low statistics

[Discussion here.]