Design and Optimisation of Ultra-Compact, High-Resolution X-Ray Imaging Systems

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3rd OMA Topical Workshop - Medical Accelerator Design and Diagnostics





QUASAR







UK Research and Innovation

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A MONTE CARLO APPROACH TO IMAGING AND DOSE SIMULATIONS IN REALISTIC PHANTOMS USING COMPACT X-RAY SOURCE

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absorbed dose in the head phantom, image reconstruction (figure 4). The fluence is given in photons/cm²/primary X-ray

Figure 4: X-ray radiography image reconstruction by visualisation of the X-ray fluence behind the phantom.

and the dose in GeV/g/primary X-ray.

Digital Tomosynthesis vs Computerised Tomography





Around double dose than X-ray but offers depth information.

Much higher dose than X-ray but offers great 3D detail.

The benefit of the emitter array





Contents



- 1. Electron generation and acceleration
- 2. X-ray production from deceleration of electrons in a high Z target.
- 3. Filtration of the X-ray beam.
- 4. Transport of X-rays in the phantom-patient.
- 5. Radiation collection in a pixelated detector and generation of an image.
- 6. 3D image reconstruction using generated images.
- 7. Discussion on the methods, approximations and next steps.

X-ray generation



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Compare the unfiltered spectrum with literature

XRF Spectrum for Pure (99.9%) Tantalum



The spectrum after the AI filter.



The emitter array

45 point X-ray sources7x7 configuration1cm pitchUniform cone beam each of40 degrees full cone angle

Spectrum is Gaussian approximation of the previous simulation.





The phantom human head

Voxel size: 0.488mmx0.488mmx1.25mm



Anonymous phantom in DICOM format taken from: Patient Contributed Image Repository www.pcir.org

Correlation between CT numbers and tissue parameters needed for Monte Carlo simulations of clinical dose distributions

Wilfried Schneider, Thomas Bortfeld and Wolfgang Schlegel <u>Physics in Medicine & Biology, Volume 45, Number 2</u>

The ideal X-ray detector



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The simulation geometry



X-Ray images



Beer-Lambert Law

Image reconstruction



3D Shepp-Logan phantom



Ellipsoid parameters taken from:

LI, Y.L., SUN, F.R., Liu, Z., QU, H.J. and LI, Q.N., 2005. The computation of three-dimension Shepp-Logan head phantom simulation projection data. *Journal of Shandong University* (Engineering Science), 1, p.013.

HU to material conversion by

Correlation between CT numbers and tissue parameters needed for Monte Carlo simulations of clinical dose distributions

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Shepp-Logan phantom digital tomosynthesis





Next steps: Use realistic description of the following

- Assumed monoenergetic e- pencil beam with zero emittance.
- Neglected electric and magnetic fields in the source geometry.
- Fitted a Gaussian instead of sampling energies from the actual spectrum.
- Assumed point sources.
- Assumed uniform cone beam of X-rays.
- Used an ideal detector.
- Did not include all the components of the system like collimators etc.

This is only a proof of principle that the image quality of DT can be assessed with Monte Carlo techniques and offer system optimisation potential.

Most important challenge: Phase space sampling



- Source simulations only generate a few 100.000 X-rays per day on cluster.
- Imaging with high resolution detector requires trillions of X-rays to reduce stat. uncert.

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A fast way to generate million times more particles must be found.

- Discrete multivariate distributions E(x,y,x`,y`)
- Multivariate Kernel Density Estimation
- Generative Adversarial Neural Networks
- Biasing techniques like particle splitting or even rare event.
- Apply some reasonable approximations.

Thank you

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