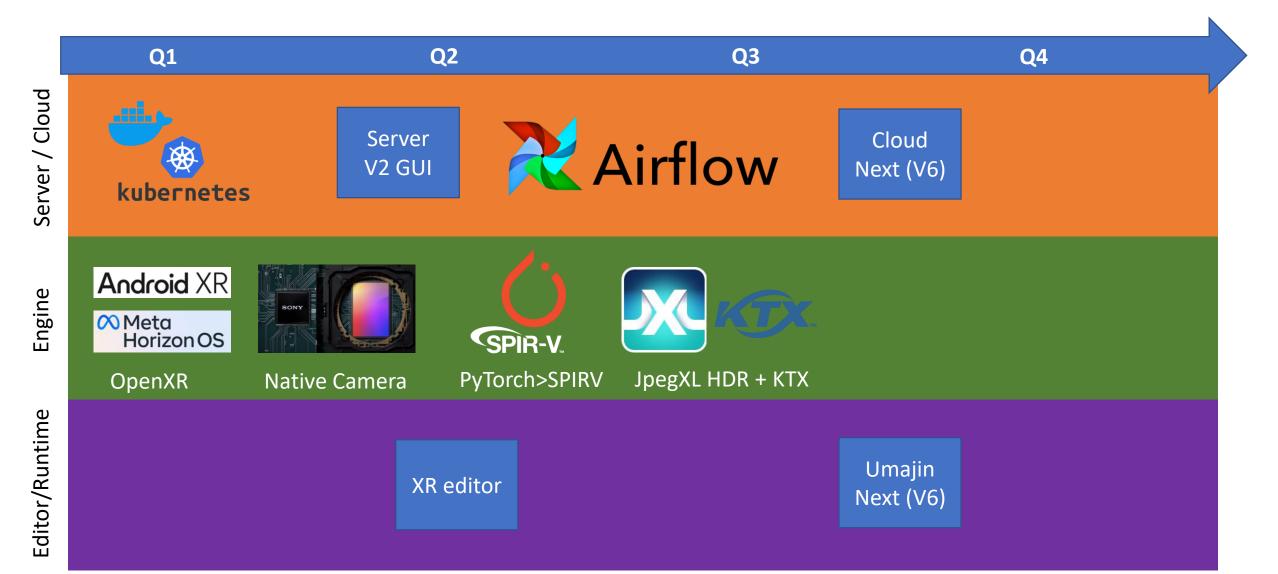


Multicore World 2025 Sparsification

David Brebner







Runtimes

- > Java > Python + QT
- > .NET + MAUI (Multi-platform App UI)

Browser + App frameworks

- > Chrome
- > Electron
- > Flutter / ReactNative / Ionic

3D engines

- > Unity
- > Godot
- > Unreal

Umajin

- > Better memory usage & safety
- > Improved compute (LLVM JIT & LLVM Compiled)
- > Modern capability framework; (rich media, 3D, VR, AR, geospatial, cloud first API's, AI, image processing, agents and more)

Umajin

- > An order of magnitude lower memory usage
- > Better graphical rendering & GPU usage
- > Better CPU performance
- > Better development & debugging

Umajin

- > For enterprise not game development
- > Optimised for enterprise devices
- > Long battery life, low power usage
- > Significantly faster development & debugging

Context for Sparsification

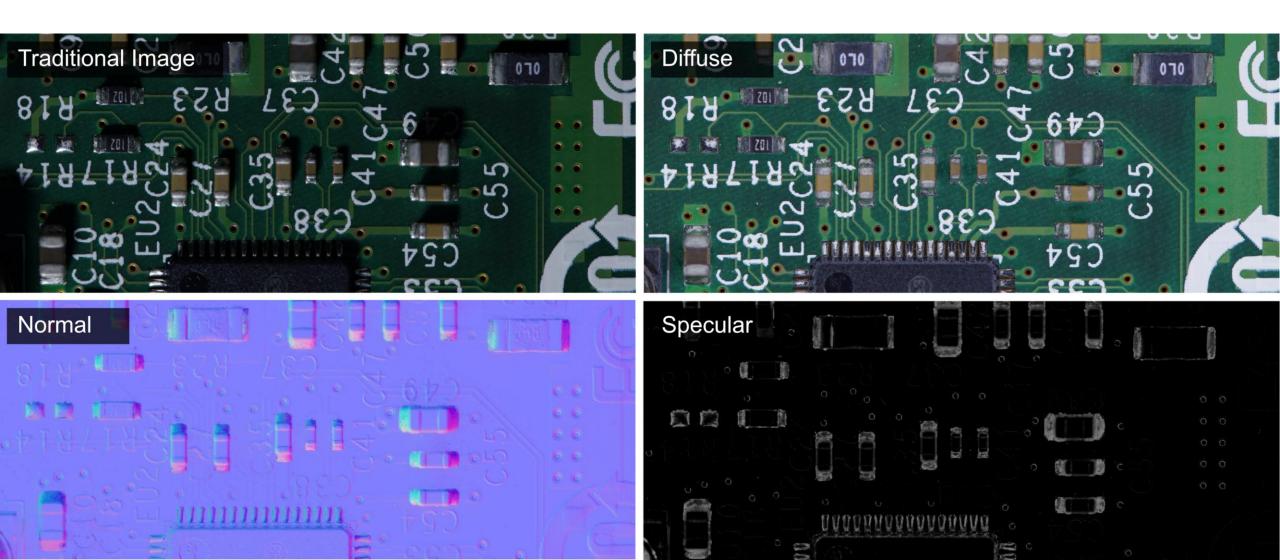


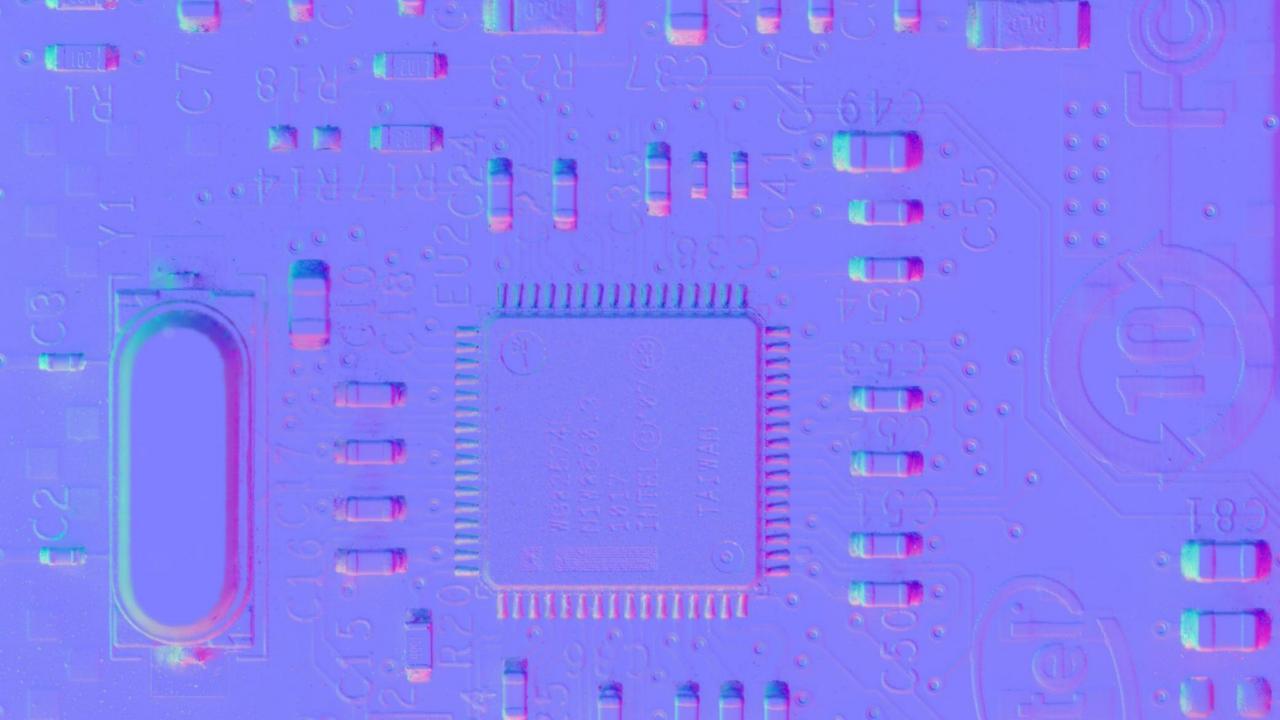
- 3D data
- Time series data
- Generated simulation or AI extrapolated metadata

>>> Exponential data growth



Generated Data

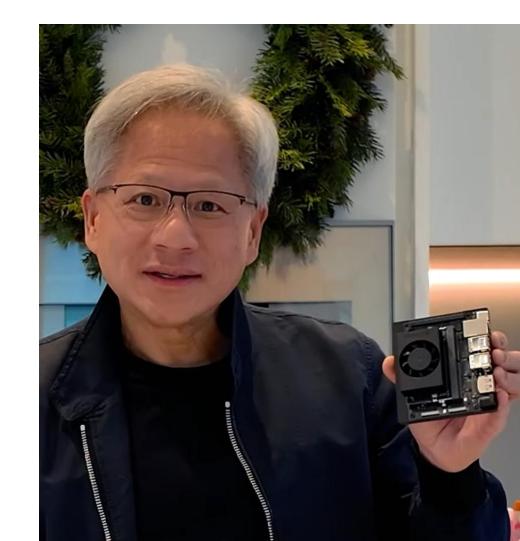




Sparsification

- Orin Nano Super
 - I showed the Orin Nano last year
 - This year same hardware has been rebranded as "super"
 - Where has Moore's law gone for GPU's?
- 3D time series data reduction
 - Reduce node count
 - 10,000x 1,000,000x reduction over voxels
 - Pre computed connectivity
 - Allows for much faster simulation and analysis
 - We don't want quantisation or compression





Sparsification



Voxel

For a 16384 x 16384 x 16384 voxel grid, assuming 4 floats per voxel Storage = $16384^3 \times 4$ floats $\approx 17 \times 10^{12}$ floats (70 TB assuming a 32bit float)

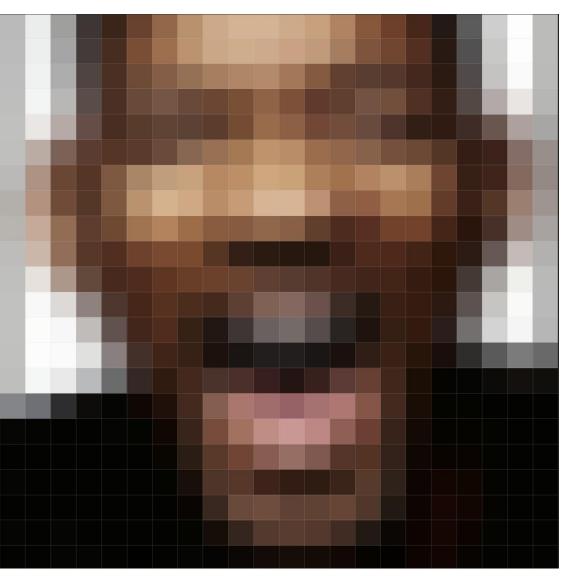
Sparse Scaleable Point Cloud

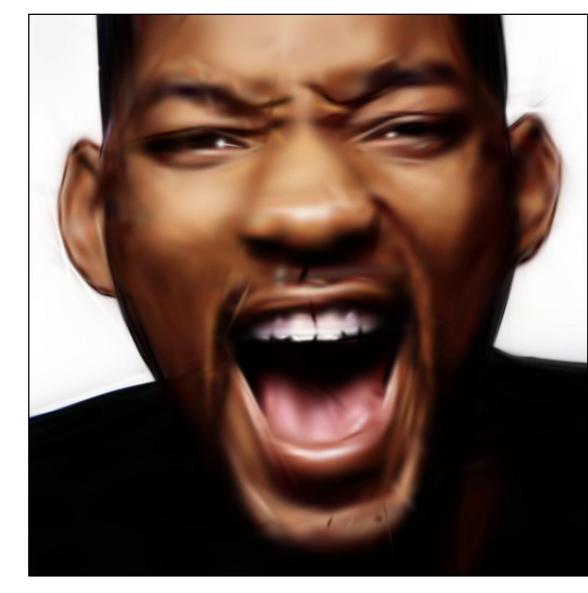
Size depends on complexity, but assuming biological samples

1 billion volumetric scaleable points
Storage = 1b × ~13 floats ≈ 13 billion floats
(52 GB assuming a 32bit float)

1 million volumetric scaleable points Storage = 1M × ~13 floats ≈ 13 million floats (52 MB assuming a 32bit float)

2D Intuition





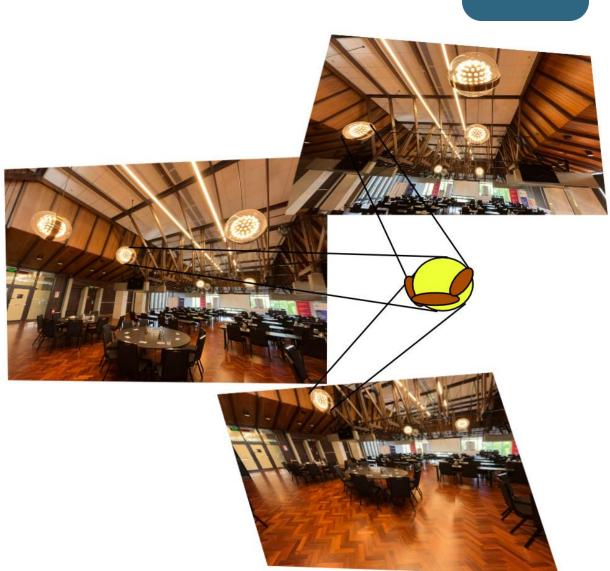
500 Gaussians

22x22 pixels



Sparse Fuzzy Oriented and Scaled Blobs

- 3DGS -Three dimensional Gaussian Splatting
- Rendering is a type of raycasting
- Generating blobs involves optimising to match the sample images from an estimated camera position



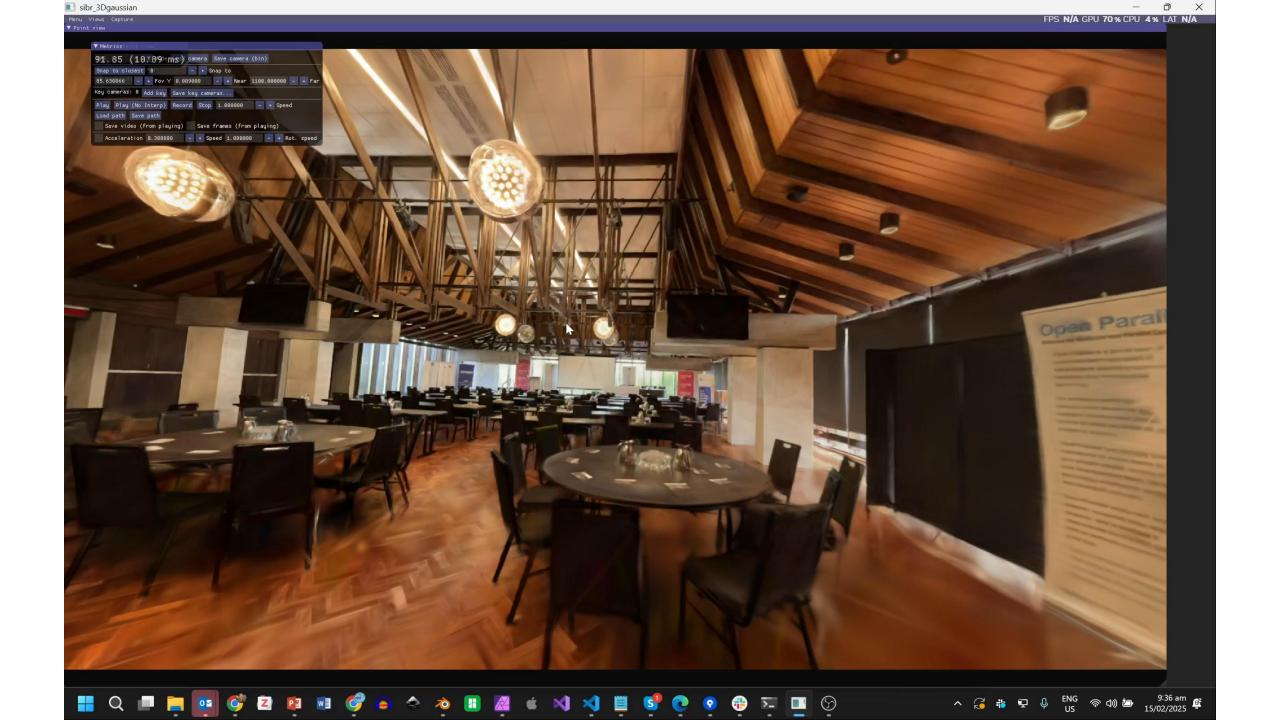
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Clear

Log

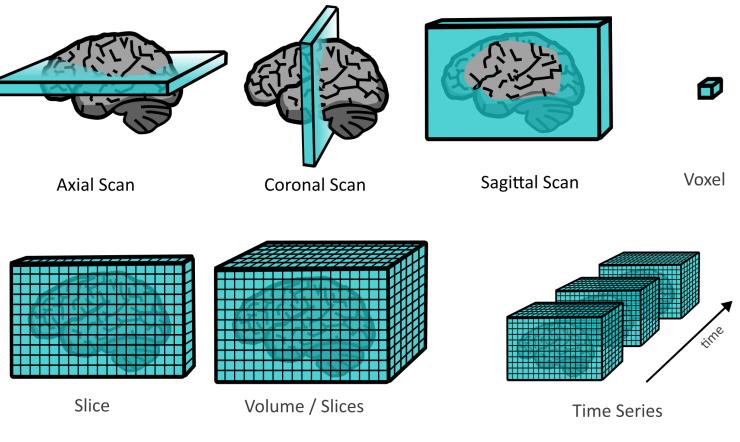
Fusing image [254/273] in 1.545s (5451723 poi Fusing image [255/273] in 1.592s (5451723 poi Fusing image [256/273] in 2.299s (5452901 poi Fusing image [257/273] in 2.265s (5453181 poi Fusing image [258/273] in 1.812s (5453220 poi Fusing image [259/273] in 1.627s (5453221 poi Fusing image [260/273] in 1.554s (5474954 poi Fusing image [261/273] in 1.640s (5497105 poi Fusing image [262/273] in 1.632s (5499145 poi Fusing image [263/273] in 1.562s (5500039 poi Fusing image [264/273] in 1.801s (5500136 poi Fusing image [265/273] in 2.216s (5500151 poi Fusing image [266/273] in 2.099s (5500152 poi Fusing image [267/273] in 1.555s (5500152 poi Fusing image [268/273] in 1.571s (5500152 poi Fusing image [269/273] in 1.573s (5500155 poi Fusing image [270/273] in 2.202s (5500155 poi Fusing image [271/273] in 2.258s (5500155 poi Fusing image [272/273] in 1.748s (5500155 poi Fusing image [273/273] in 1.654s (5500155 poi Number of fused points: Elapsed time: 10.728 [m Writing output: C:/User





Sparse Fuzzy Oriented and Scaled Blobs

- Known 'camera position'
- Volumes not surfaces
- Voxel Homogeneity
- Voxel Variation Frequency

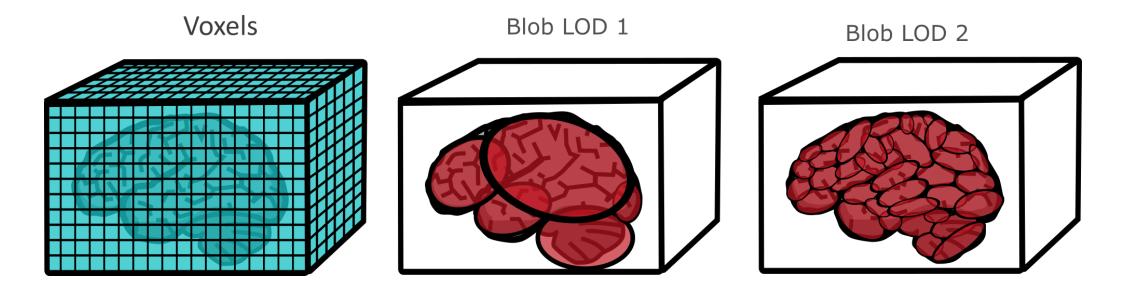


Conceptual MRI Data

Sparse Fuzzy Oriented and Scaled Blobs

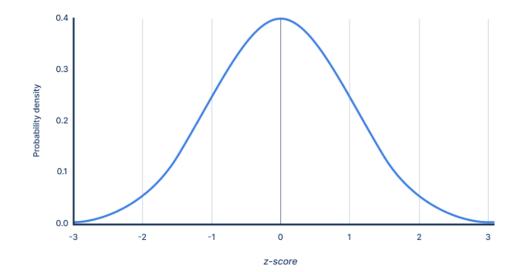


- Voxels converted into sparse, fuzzy, oriented and scaled blobs that represent the voxel values in 3D shape not just surface
- Different levels of detail of fidelity are able to be precomputed



Adding more complexity to blobs

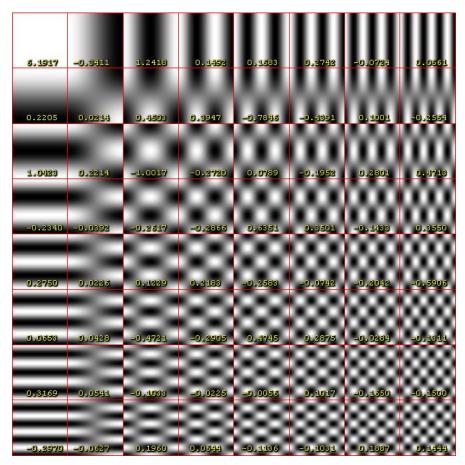
• DCT is the basis for JPG and VarDCT for JpegXL – we have added DCT rather than just a 'normal' Gaussian



Standard normal distribution

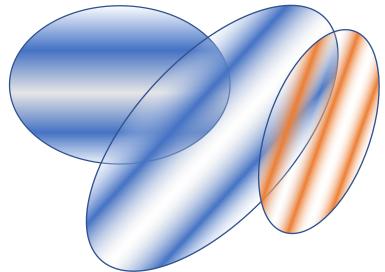
U

Discrete Cosine Transform



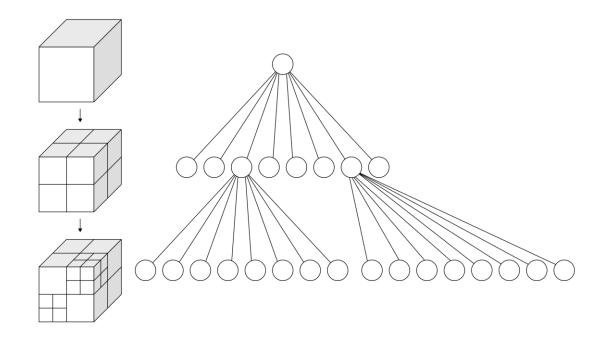
Adding more complexity to blobs

 Great compression – but requires convolving the blobs together for the final result so this is less useful for simulation where you would like your nodes to be independent



Managing variable complexity with spatial partitioning

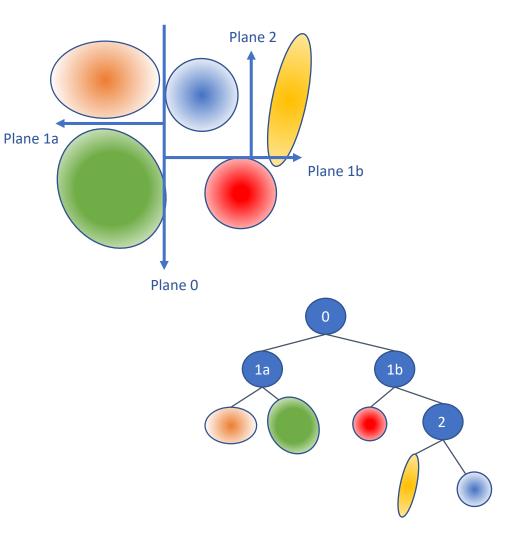
- OctTree allows for variable resolution and storing multiple levels of detail
- Deltas calculated per cube in the octree – allows for time series data to be variably spatially run length encoded for unchanging regions
- Spatial partitioning allows for parts of the data set to be processed in parallel





Simulation & processing using trees

- KD Trees for homogenous nodes
- KD Trees for connected nodes
- Replace per voxel style marching algorithms with significantly faster approaches using 10,000x 1,000,000x fewer nodes with precomputed connectivity
- Connectivity of different classifications (bone, tissue, material properties)
- Connectivity of variable density



Volume not surface optimisation

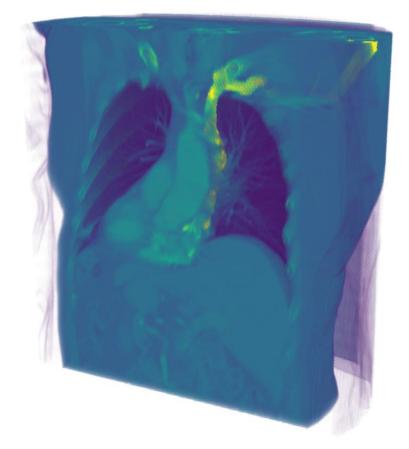


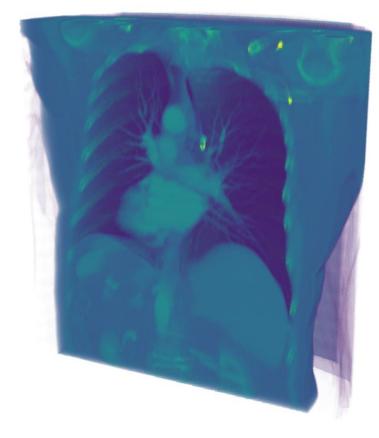
R²-Gaussian: Rectifying Radiative Gaussian Splatting for Tomographic Reconstruction

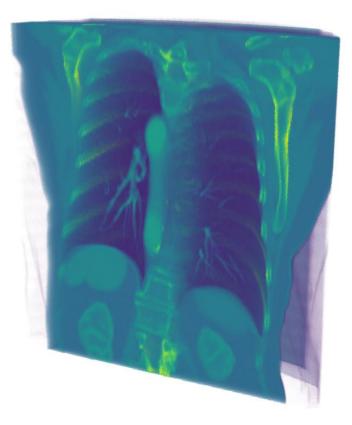
Ruyi Zha^{1,} Tao Jun Lin¹ Yuanhao Cai² Jiwen Cao¹ Yanhao Zhang³ Hongdong Li¹ ¹Australian National University ²Johns Hopkins University ³University of Technology Sydney

Volume not surface optimisation



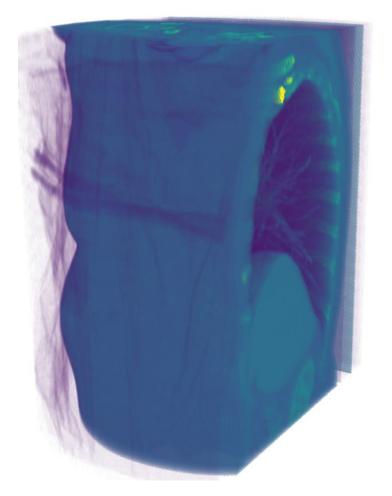


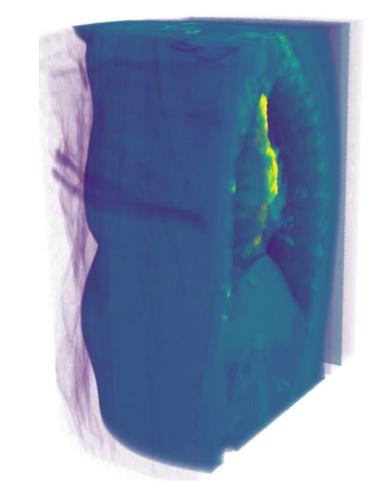


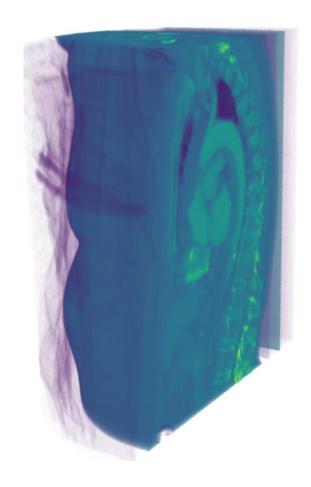


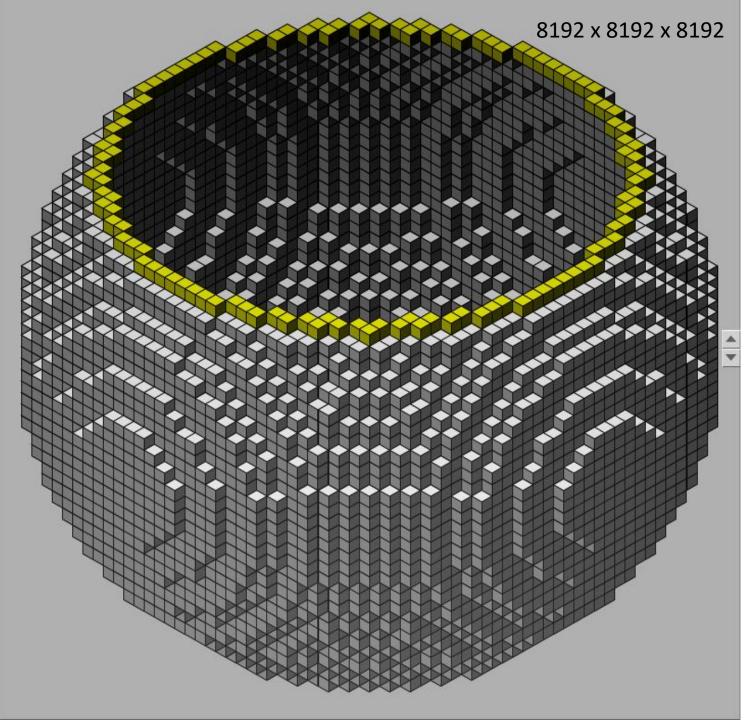
Volume not surface optimisation



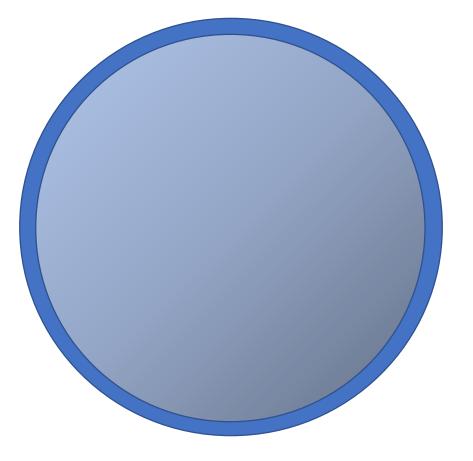








2 gaussian ellipsoids



Nanoscanner

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150μ

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