A Survey of GPU-Based Large-Scale Volume Visualization

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Overview

- Part 1: More tutorial material (Markus)

  - Motivation and scope

  - Fundamentals, basic scalability issues and techniques

    - Data representation, work/data partitioning, work/data reduction

- Part 2: More state of the art material (Johanna)

  - Scalable volume rendering categorization and examples

    - Working set determination

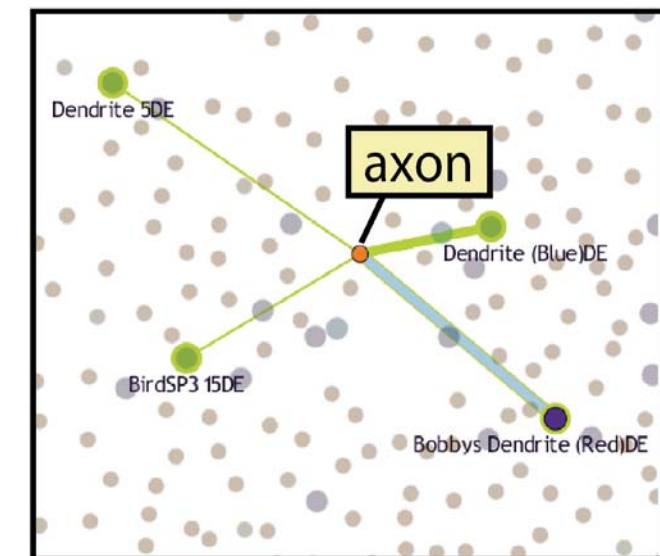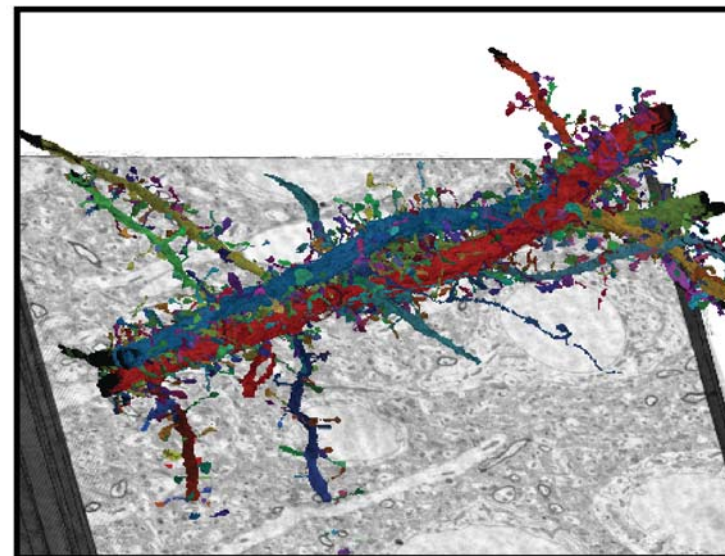   - Working set storage and access

    - Rendering (ray traversal)

Motivation and Scope

Big Data

"In information technology, big data is a collection of data sets so large and complex that it becomes difficult to process using on-hand database management tools or traditional data processing applications.
The challenges include capture, curation, storage, search, sharing, analysis, and visualization."

'Big Data' on wikipedia.org

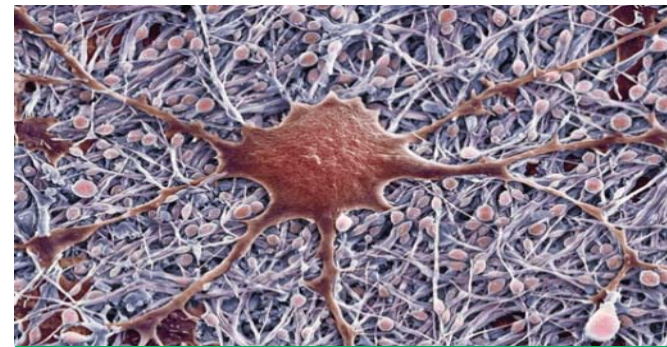Our interest:
Very large 3D volume data

Example: Connectomics (neuroscience)
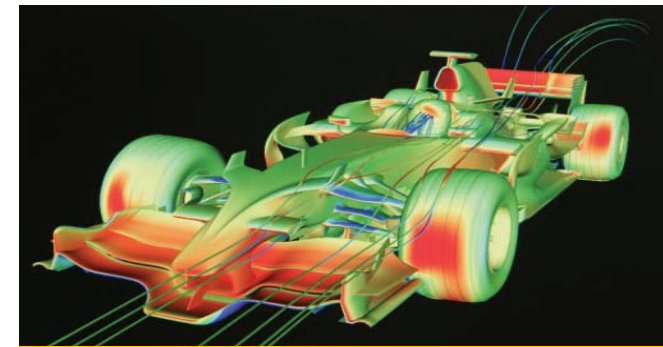
Data-Driven Science (eScience)
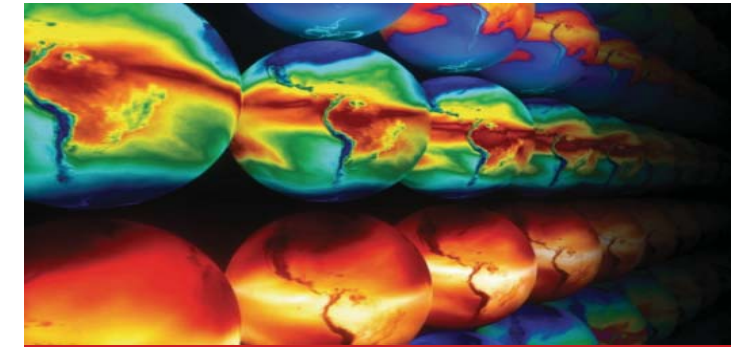

MEDICINE
Digital Health Records

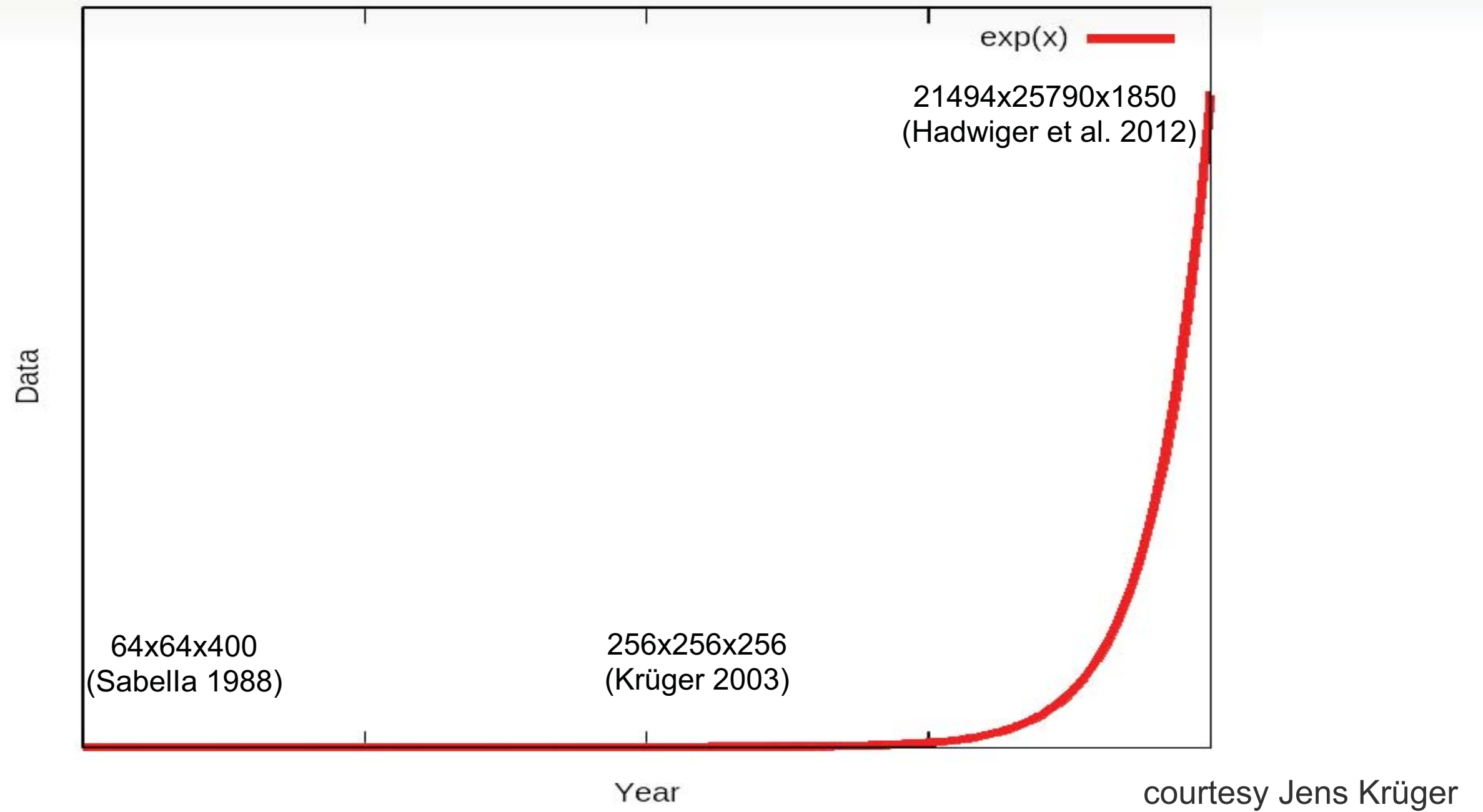**BIOLOGY**
Connectomics

ENGINEERING
Large CFD Simulations

EARTH SCIENCES
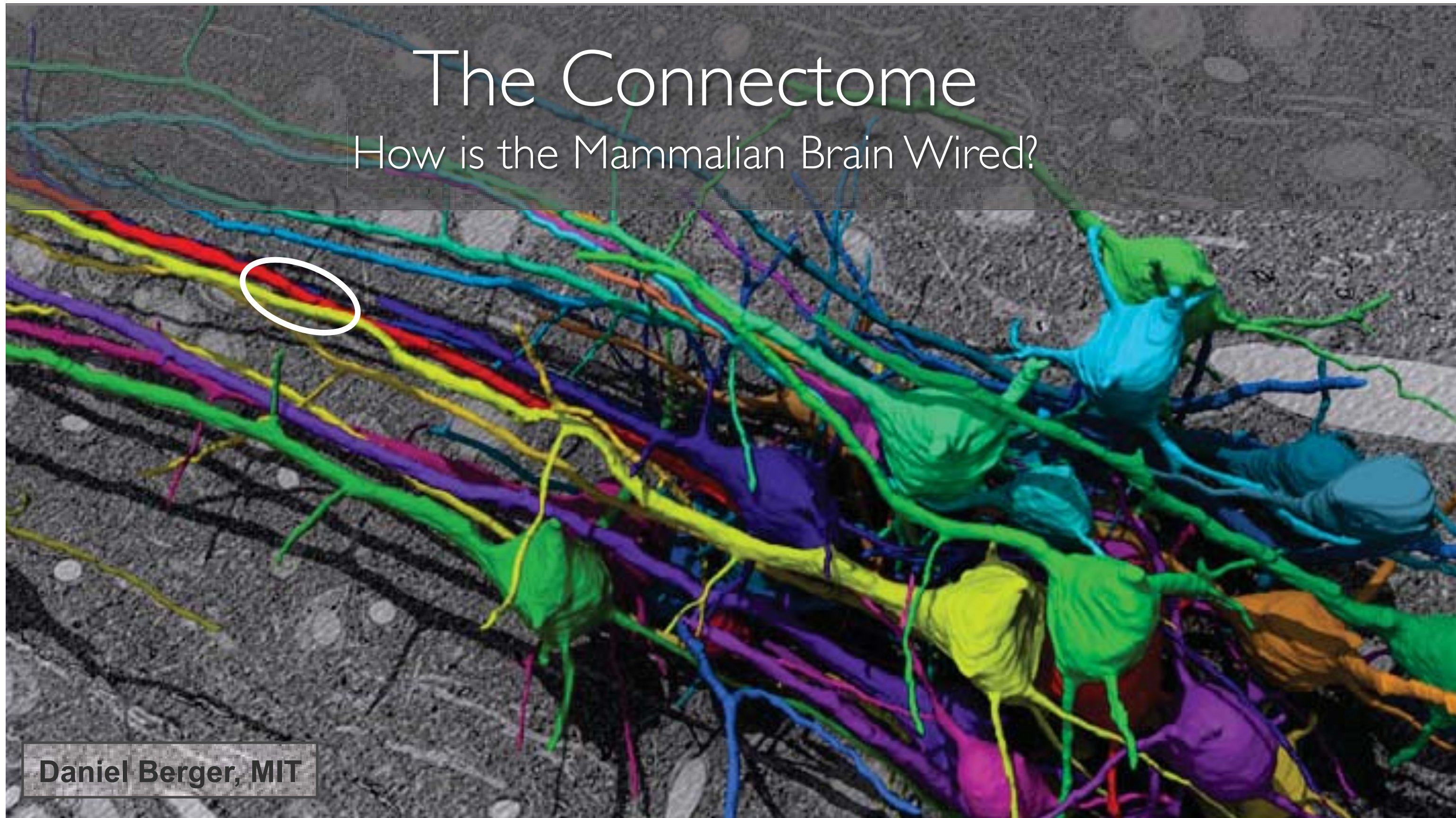Global Climate Models

courtesy Stefan Bruckner

Volume Data Growth

exp(x) ━━━

21494x25790x1850
(Hadwiger et al. 2012)

64x64x400
(Sabella 1988)

256x256x256
(Krüger 2003)

Data

Year

courtesy Jens Krüger

Data Size Examples

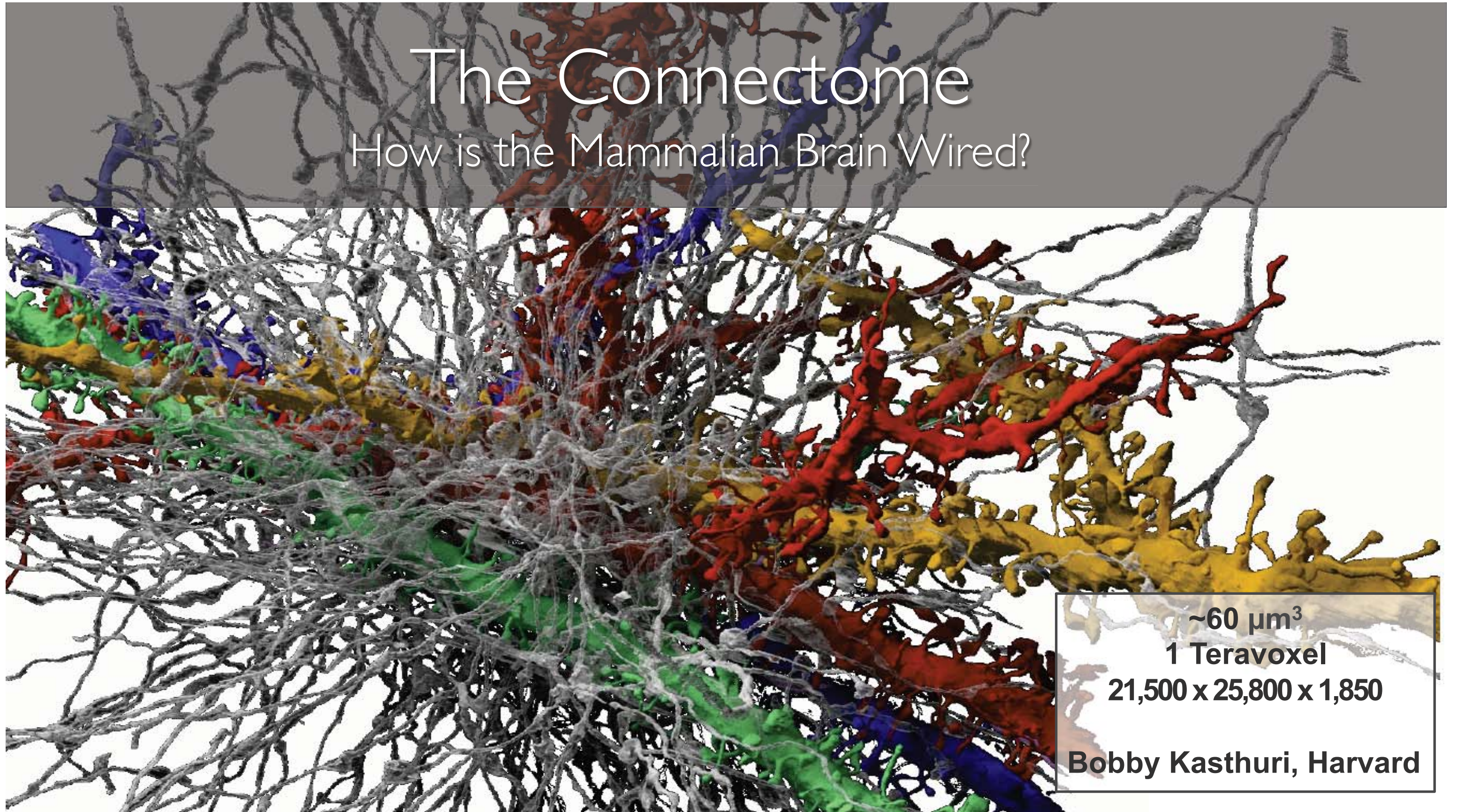| year	paper	data set size	comments
2002	Guthe et al.	512 x 512 x 999 (500 MB) 2,048 x 1,216 x 1,877 (4.4 GB)	multi-pass, wavelet compression, streaming from disk
2003	Krüger & Westermann	256 x 256 x 256 (32 MB)	single-pass ray-casting
2005	Hadwiger et al.	576 x 352 x 1,536 (594 MB)	single-pass ray-casting (bricked)
2006	Ljung	512 x 512 x 628 (314 MB) 512 x 512 x 3396 (1.7 GB)	single-pass ray-casting, multi-resolution
2008	Gobbetti et al.	2,048 x 1,024 x 1,080 (4.2 GB)	'ray-guided' ray-casting with occlusion queries
2009	Crassin et al.	8,192 x 8,192 x 8,192 (512 GB)	ray-guided ray-casting
2011	Engel	8,192 x 8,192 x 16,384 (1 TB)	ray-guided ray-casting
2012	Hadwiger et al.	18,000 x 18,000 x 304 (92 GB) 21,494 x 25,790 x 1,850 (955 GB)	ray-guided ray-casting visualization-driven system
2013	Fogal et al.	1,728 x 1,008 x 1,878 (12.2 GB) 8,192 x 8,192 x 8,192 (512 GB)	ray-guided ray-casting

The Connectome
How is the Mammalian Brain Wired?

Daniel Berger, MIT

The Connectome
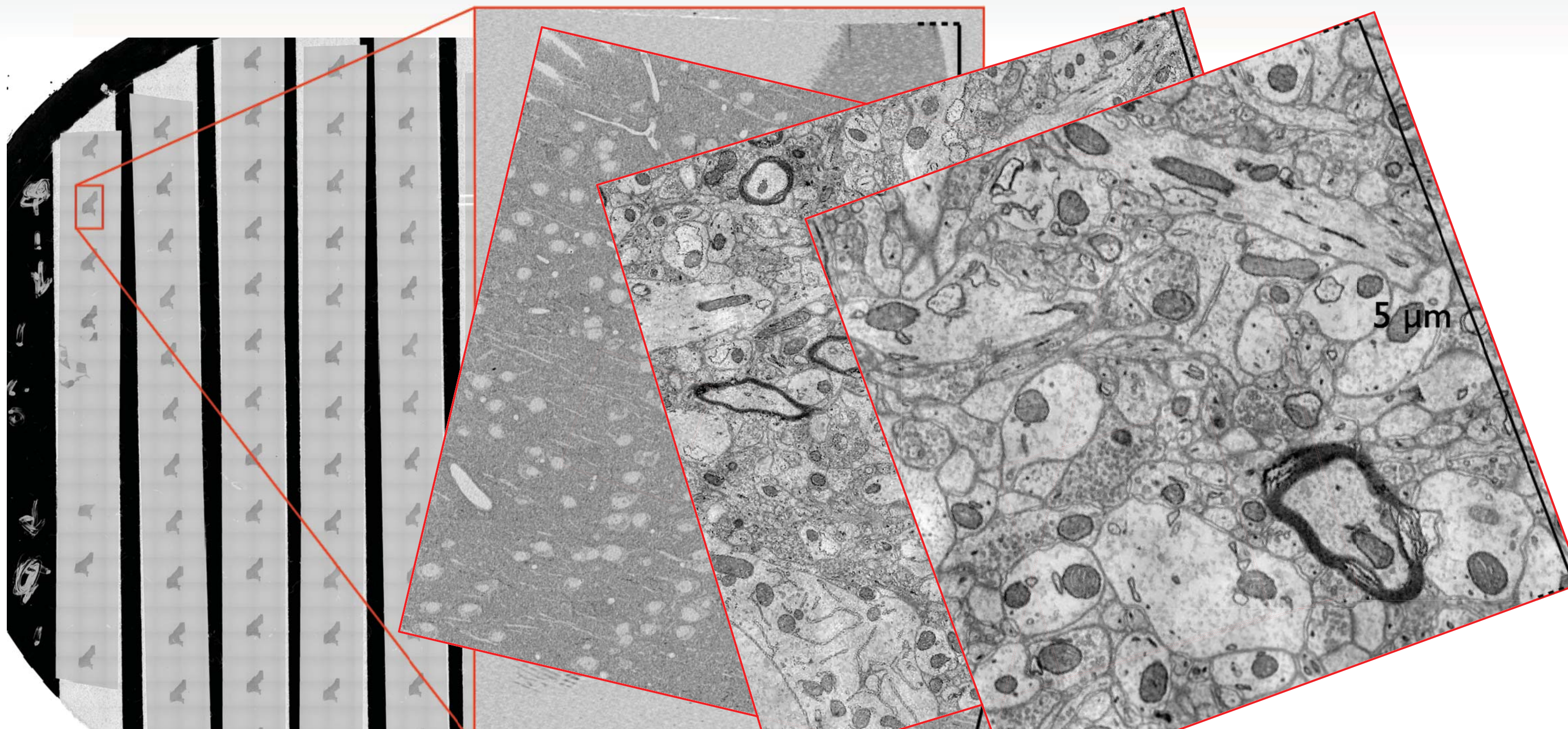## How is the Mammalian Brain Wired?

~60 μm³
1 Teravoxel
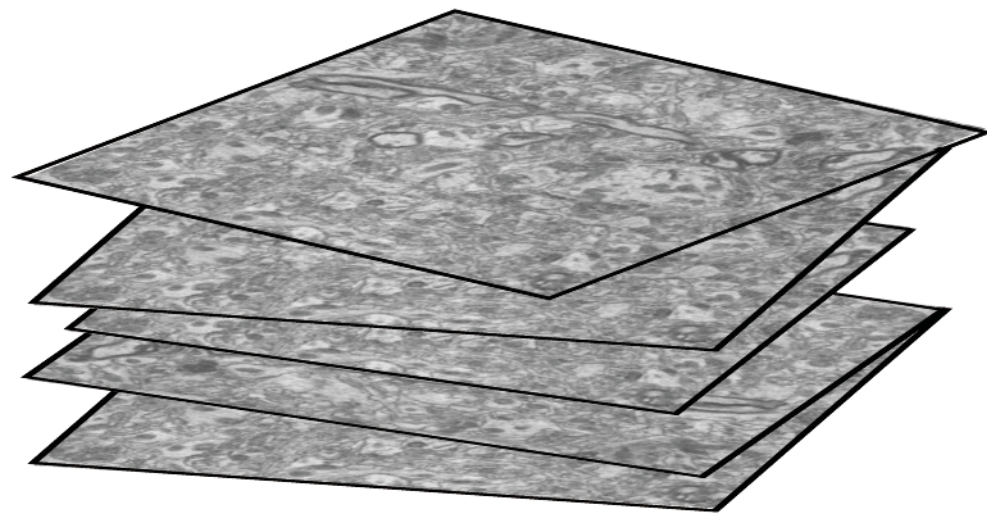**21,500 x 25,800 x 1,850**

Bobby Kasthuri, Harvard
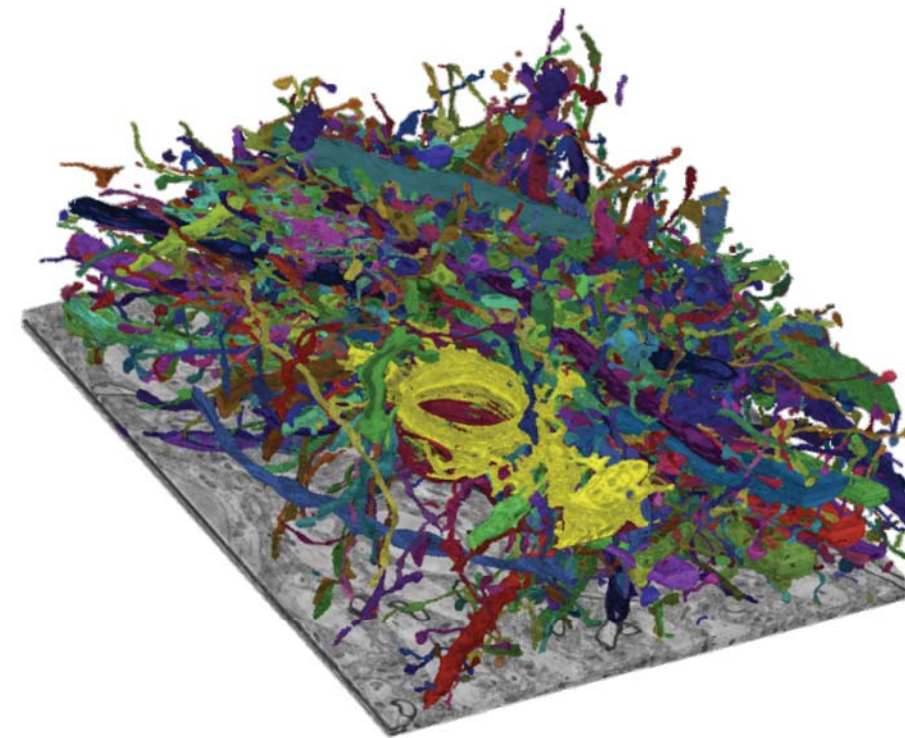
EM Slice Stacks (1)

5 µm

EM Slice Stacks (2)

- Huge amount of data (terabytes to petabytes)

- Scanning and segmentation take months

1 mm^3 at 5 nm x 50 nm
- 200k x 200k x 20,000
- 40 gigapixels x 20k = **800 teravoxels**

High-throughput microscopy
- 40 megapixels / second
- 800 teravoxels = **8 months**

Survey Scope

- Focus
  - (Single) GPUs in standard workstations
  - Scalar volume data; single time step
  - But a lot applies to more general settings…

- Orthogonal techniques (won't cover details)
  - Parallel and distributed rendering, clusters, supercomputers, …
  - Compression

Related Books and Surveys

- Books
  - Real-Time Volume Graphics, Engel et al., 2006
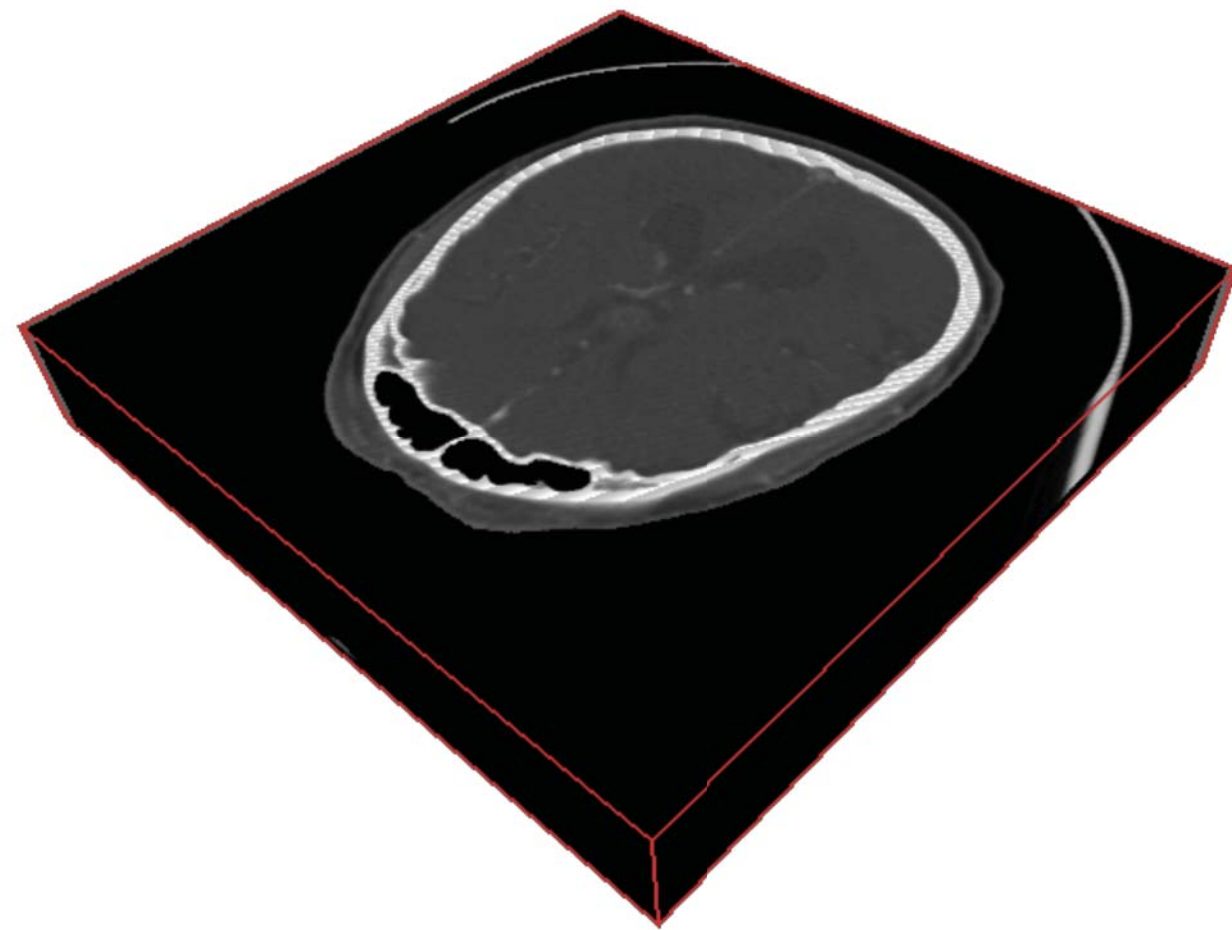 - High-Performance Visualization, Bethel et al., 2012

- Surveys
  - Parallel Visualization: Wittenbrink '98, Bartz et al. '00, Zhang et al. '05
  - Real Time Interactive Massive Model Visualization: Kasik et al. '06
  - Vis and Visual Analysis of Multifaceted Scientific Data: Kehrer and Hauser '13
  - Compressed GPU-Based Volume Rendering: Rodriguez et al. '13

Fundamentals

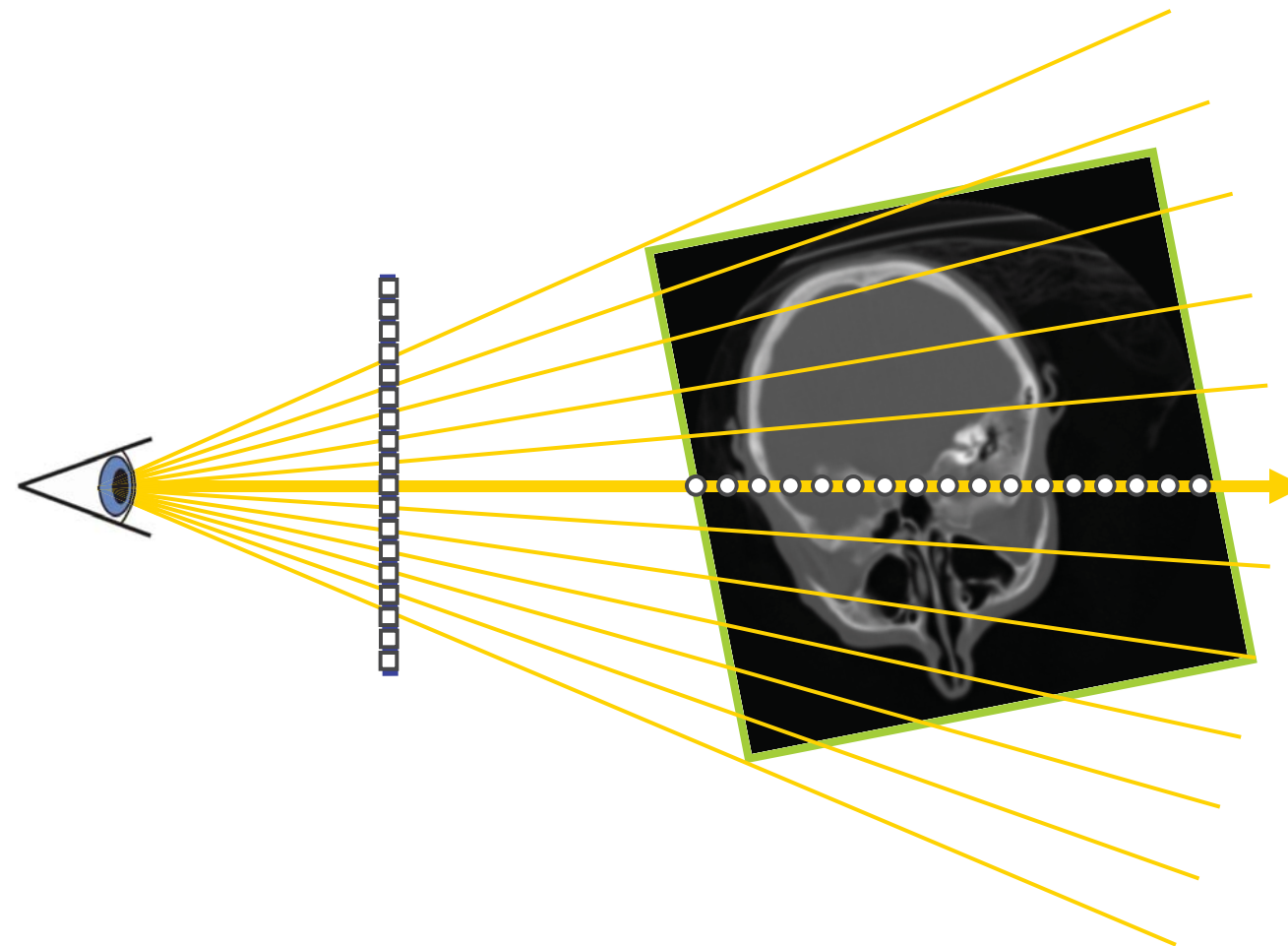# Volume Rendering (1)

- Assign optical properties (color, opacity) via *transfer function*

courtesy Christof Rezk-Salama

Volume Rendering (2)
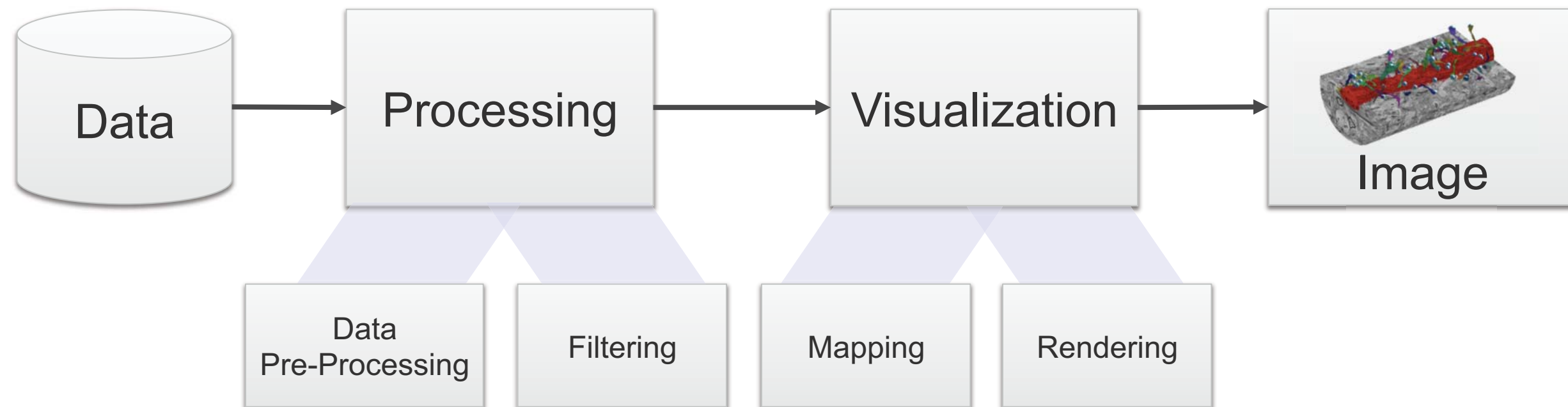
- Ray-casting

courtesy Christof Rezk-Salama

Scalability

- Traditional HPC, parallel rendering definitions

  - Strong scaling ("more nodes are faster for same data")

  - Weak scaling ("more nodes allow larger data")

- Our interest/definition: output sensitivity

  - Running time/storage proportional to size of output instead of input

    - Computational effort scales with visible data and screen resolution

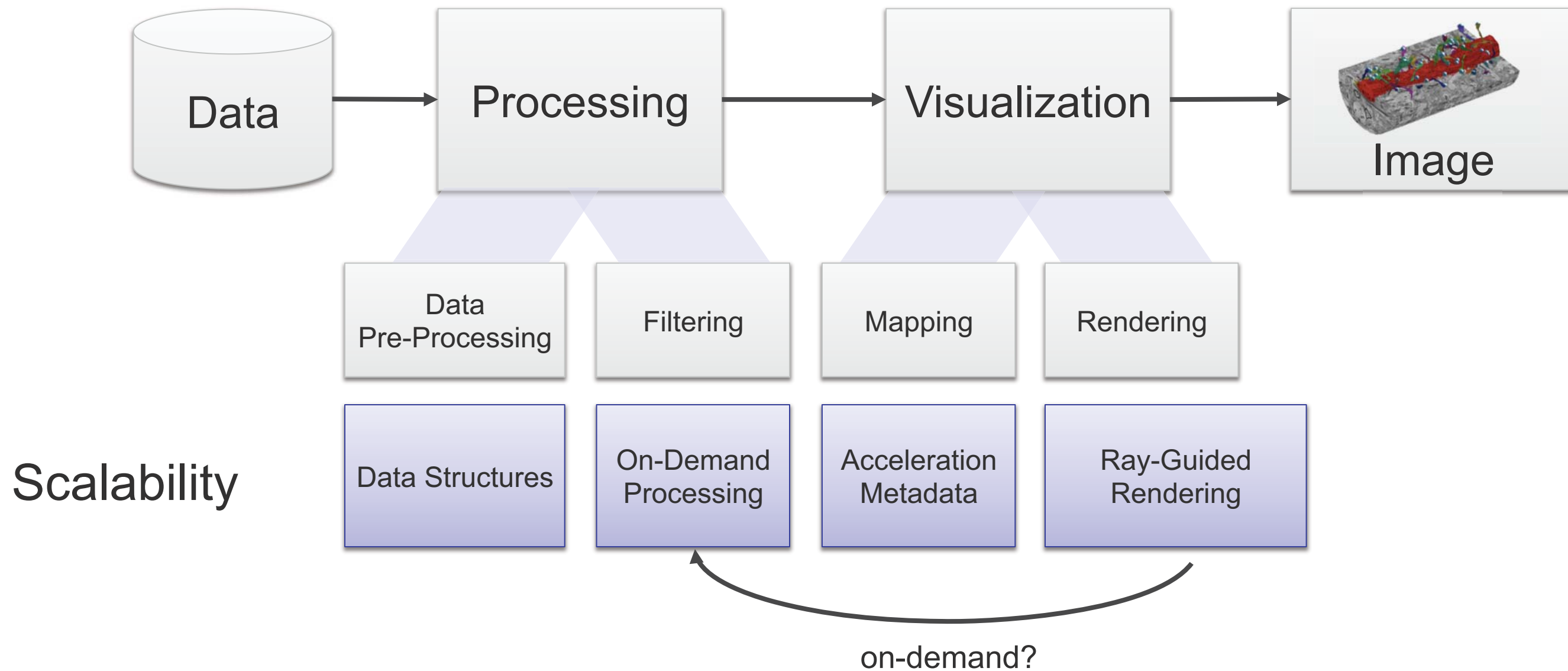   - Working set independent of original data size

Some Terminology

- Output-sensitive algorithms

  - Standard term in (geometric) occlusion culling

- Ray-guided volume rendering

  - Determine working set via ray-casting

  - Actual visibility; not approximate as in traditional occlusion culling

- Visualization-driven pipeline

  - Drive entire visualization pipeline by actual on-screen visibility

- Display-aware techniques

  - Image processing, … for current on-screen resolution

Large-Scale Visualization Pipeline

Large-Scale Visualization Pipeline

Basic Scalability Issues

Scalability Issues

Scalability issues	Scalable method
Data representation and storage	Multi-resolution data structures
	Data layout, compression
Work/data partitioning	In-core/out-of-core
	Parallel, distributed
Work/data reduction	Pre-processing
	On-demand processing
	Streaming
	In-situ visualization
	Query-based visualization

Scalability Issues

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Data Representations

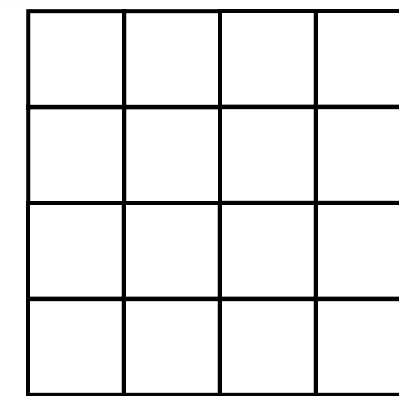| Data structure	Acceleration	Out-of-Core	Multi-Resolution
Mipmaps	-	Clipmaps	Yes
Uniform bricking	Cull bricks (linear)	Working set (bricks)	No
Hierarch. bricking	Cull bricks (hierarch.)	Working set (bricks)	Bricked mipmap
Octrees	Hierarchical traversal	Working set (subtree)	Yes (interior nodes)

- Additional issues
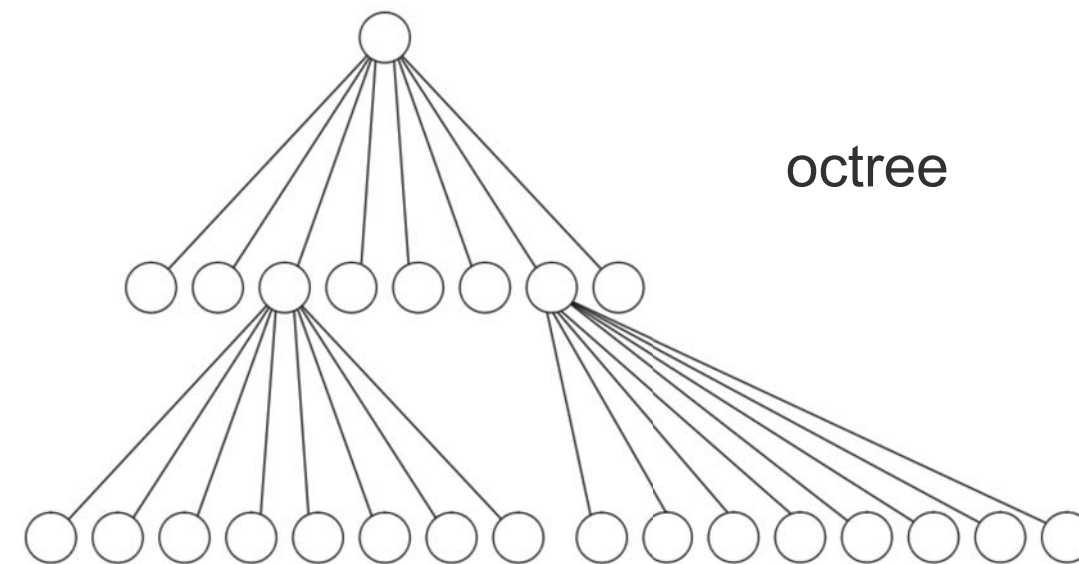 - Data layout (linear order, Z order, …)
  - Compression

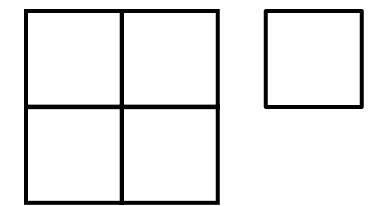# Uniform vs. Hierarchical Decomposition

- Grids
  - Uniform or non-uniform
- Hierarchical data structures
  - Pyramid of uniform grids
    - Bricked 2D/3D mipmaps
  - Tree structures
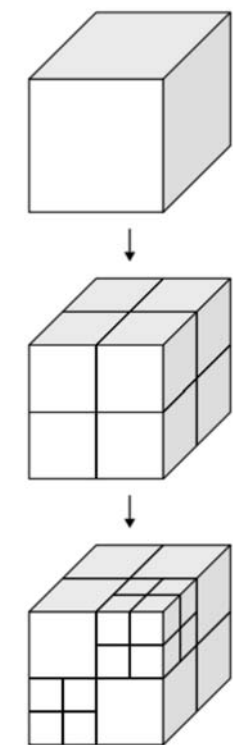   - kd-tree, quadtree, octree
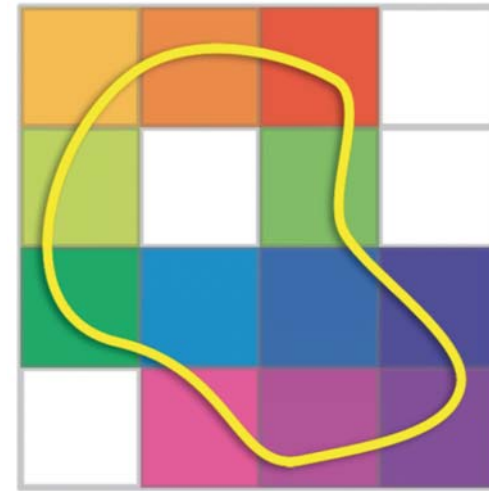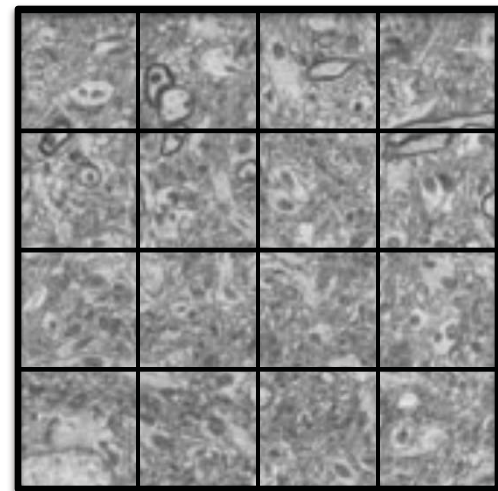
uniform grid

bricked mipmap

octree

wikipedia.org

Bricking (1)

- Object space (data) decomposition

  - Subdivide data domain into small bricks

  - Re-orders data for spatial locality

  - Each brick is now one unit (culling, paging, loading, …)
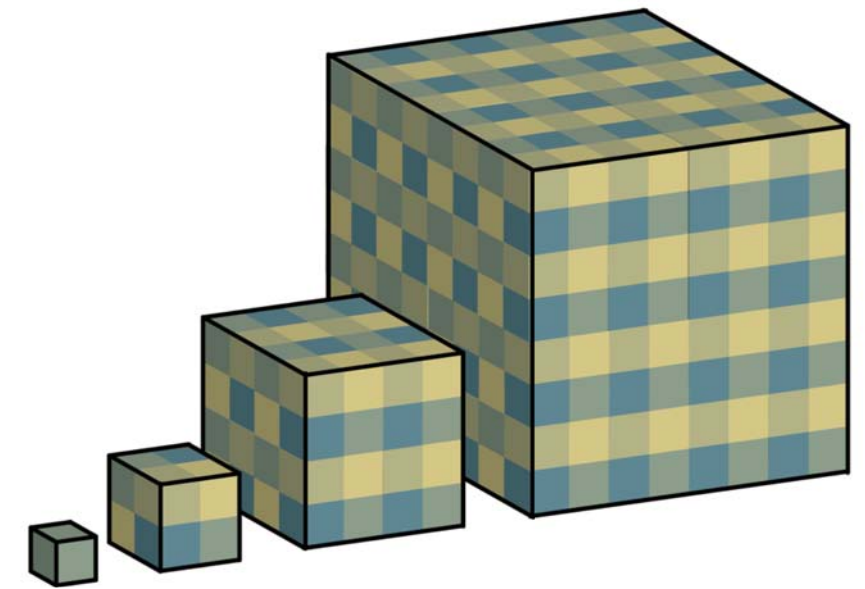
Bricking (2)

- What brick size to use?

  - Small bricks

    + Good granularity
    (better culling efficiency, tighter working set, …)

    - More bricks to cull, more overhead for ghost voxels,
    one rendering pass per brick is infeasible

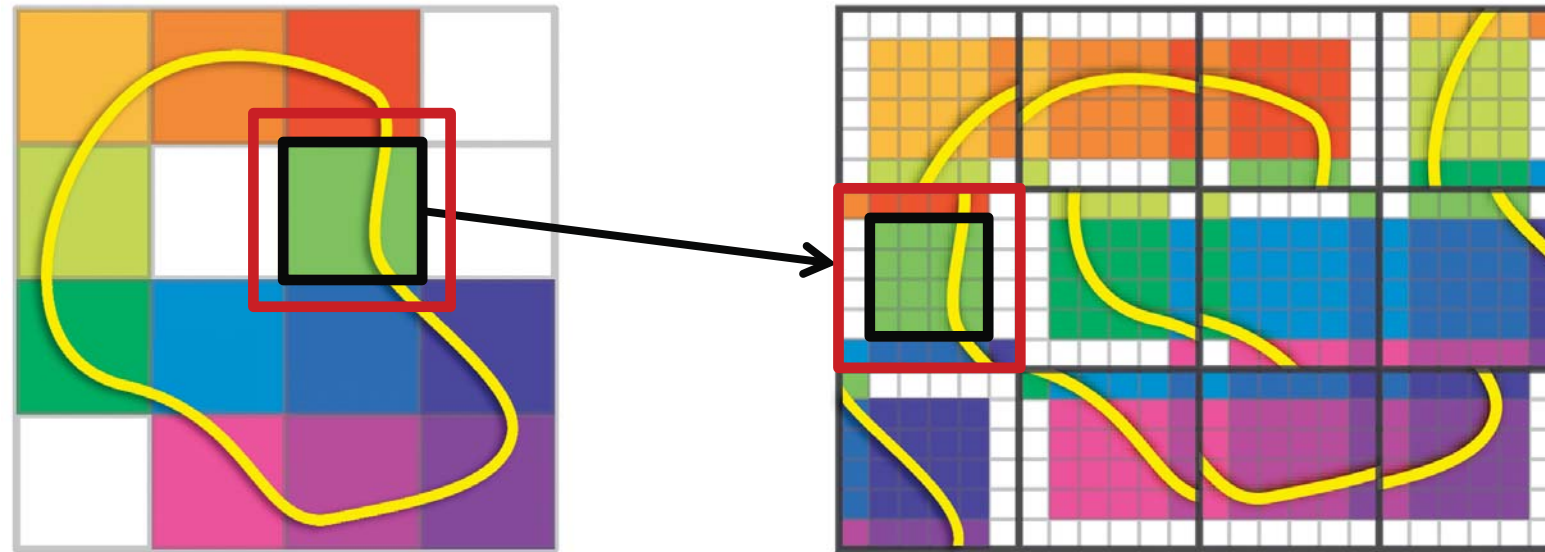 - Traditional out-of-core volume rendering: **large** bricks (e.g., 256^3)

  - Modern out-of-core volume rendering: **small** bricks (e.g., 32^3)

    - Task-dependent brick sizes (small for rendering, large for disk/network storage)

Analysis of different brick sizes: [Fogal et al. 2013]

Filtering at Brick Boundaries

- Duplicate voxels at border (ghost voxels)

  - Need at least one voxel overlap

  - Large overhead for small bricks

- Otherwise costly filtering at brick boundary

  - Except with new hardware support: *sparse textures*
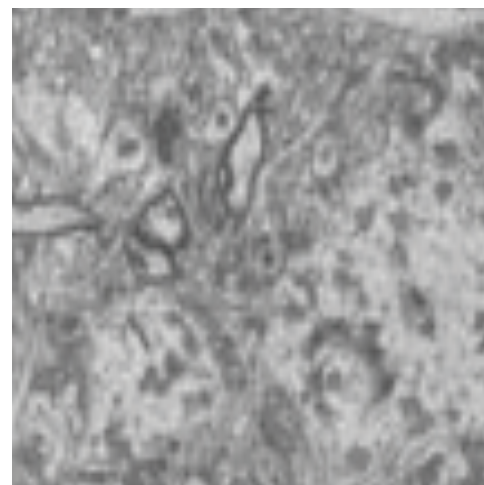
Pre-Compute All Bricks?

- Pre-computation might take very long

  - Brick on demand? Brick in streaming fashion (e.g., during scanning)?

- Different brick sizes for different tasks (storage, rendering)?

  - Re-brick to different size on demand?

  - Dynamically fix up ghost voxels?

- Can also mix 2D and 3D

  - E.g., 2D tiling pre-computed, but compute 3D bricks on demand
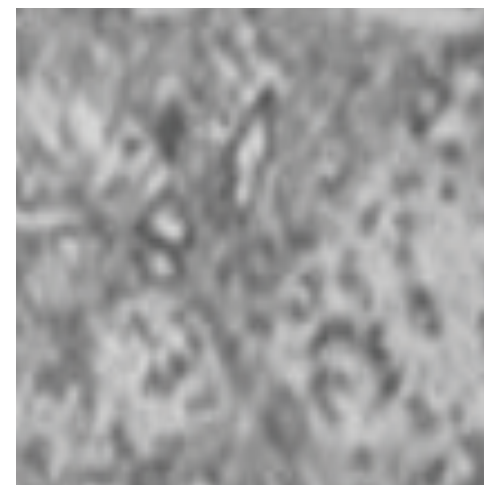
Multi-Resolution Pyramids (1)

- Collection of different resolution levels
  - Standard: dyadic pyramids (2:1 resolution reduction)
  - Can manually implement arbitrary reduction ratios

- Mipmaps
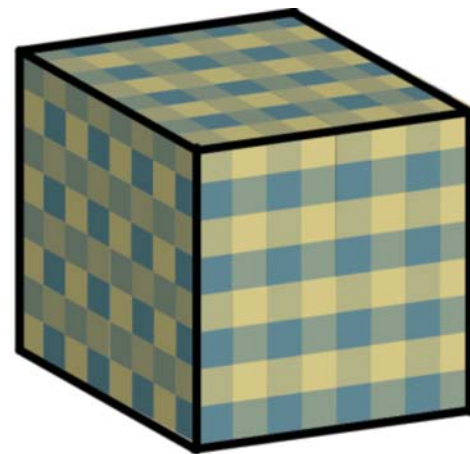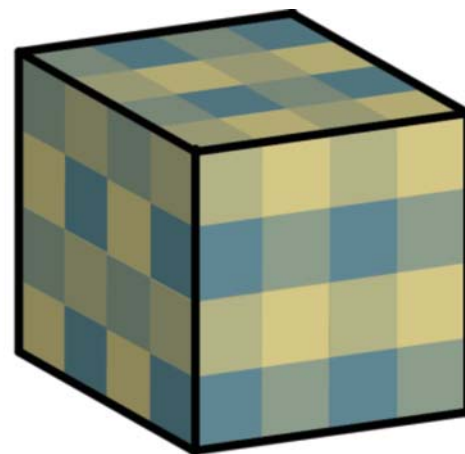 - Isotropic

level 0   level 1   level 2   level 3
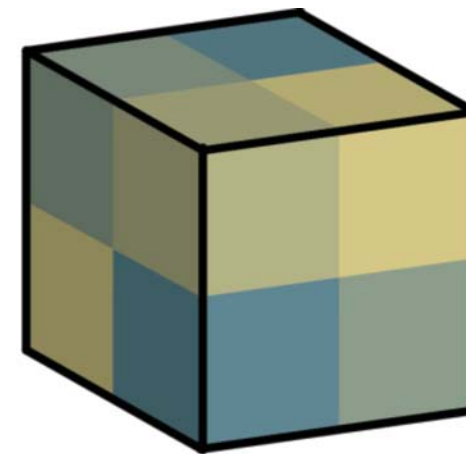
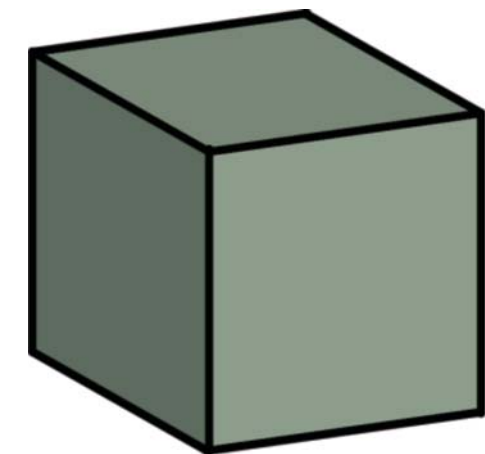# Multi-Resolution Pyramids (2)

- 3D mipmaps
  - Isotropic

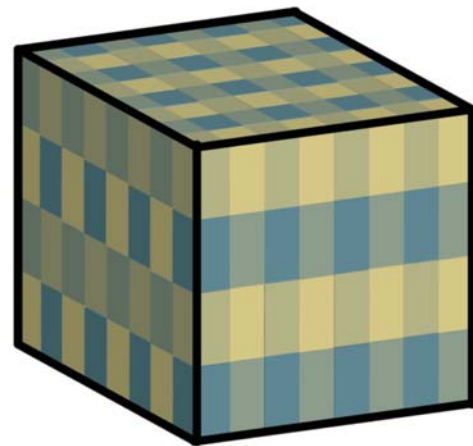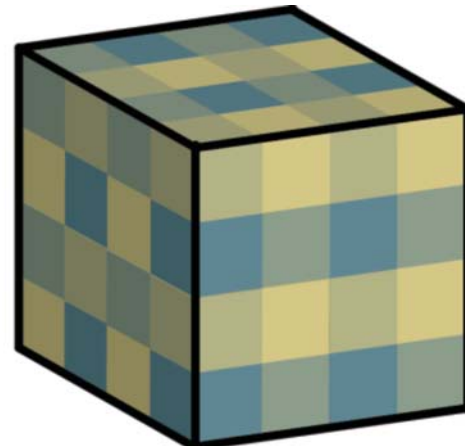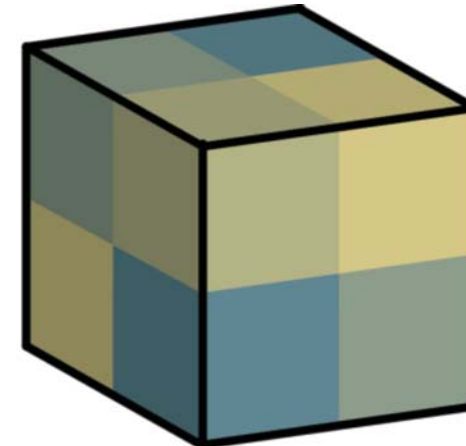| level 0	level 1	level 2	level 3
(8x8x8)	(4x4x4)	(2x2x2)	(1x1x1)

Multi-Resolution Pyramids (3)

- Scanned volume data are often anisotropic
  - Reduce resolution anisotropically to reach isotropy
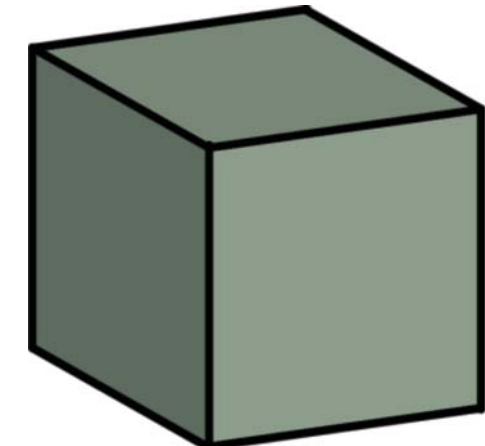

level 0
(8x8x4)

level 1
(4x4x4)

level 2
(2x2x2)

level 3
(1x1x1)

Bricking Multi-Resolution Pyramids (1)
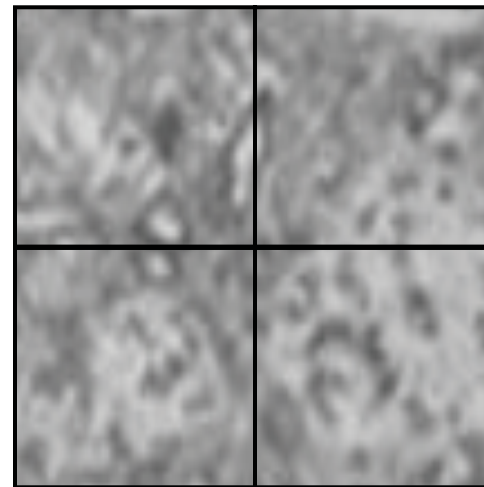
- Each level is bricked individually
  - Use same brick resolution (# voxels) in each level

level 0 level 1 level 2

spatial extent

Bricking Multi-Resolution Pyramids (2)

- Virtual memory: Each brick will be a "page"

  - "Multi-resolution virtual memory": every page lives in some resolution level

memory extent

4x4 pages   2x2 pages   1 page

Octrees for Volume Rendering (1)

- Multi-resolution
  - Adapt resolution of data to screen resolution
    - Reduce aliasing
    - Limit amount of data needed
- Acceleration
  - Hierarchical empty space skipping
  - Start traversal at root
    (but different optimized traversal algorithms:
    kd-restart, kd-shortstack, etc.)

Octrees for Volume Rendering (2)

- Representation
  - Full octree
    - Every octant in every resolution level
  - Sparse octree
    - Do not store voxel data of empty nodes
- Data structure
  - Pointer-based
    - Parent node stores pointer(s) to children
  - Pointerless
    - Array to index full octree directly

wikipedia.org

Scalability Issues

Scalability issues	Scalable method
Data representation and storage	Multi-resolution data structures
	Data layout, compression
Work/data partitioning	In-core/out-of-core
	Parallel, distributed
Work/data reduction	Pre-processing
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	Streaming
	In-situ visualization
	Query-based visualization

Work/Data Partitioning

- Out-of-core techniques

- Domain decomposition

- Parallel and distributed rendering

Out-of-Core Techniques (1)

- Data too large for GPU memory
  - Stream volume bricks from CPU to GPU on demand

- Data too large for CPU memory
  - Stream volume bricks from disk on demand

- Data too large for local disk storage
  - Stream volume bricks from network storage

Out-of-Core Techniques (2)

- Preparation
  - Subdivide spatial domain
    - May also be done "virtually", i.e., data re-ordering may be delayed
  - Allocate cache memory (e.g., large 3D cache texture)
- Run-Time
  - Determine **working set**
  - Page working set into cache memory
  - Render from cache memory

Domain Decomposition (1)

- Subdivide image domain (image space)
  - "Sort-first rendering" [Molnar, 1994]
  - View-dependent

Domain Decomposition (2)

- Subdivide data domain (object space)
  - "Sort-last rendering" [Molnar, 1994]
  - View-independent

Sort-First vs. Sort-Last

sort-first
(image domain)

sort-last
(data domain)

Scalability Issues

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On-Demand Processing

- First determine what is visible / needed

- Then process only this working set

  - Basic processing

    - Noise removal and edge detection

    - Registration and alignment

    - Segmentation, …

  - Basic data structure building

    - Construct pages/bricks/octree nodes only on demand?

Example: 3D Brick Construction from 2D EM Streams

[Hadwiger et al., IEEE Vis 2012]

Example: Denoising & Edge Enhancement

- Edge enhancement for EM data

- Caching scheme

  - Process only currently visible bricks

  - Cache result for re-use

- GPU Implementation

  - CUDA and shared memory for fast computation

- Different noise removal and filtering algorithms

[Jeong et al., IEEE Vis 2009]
Scalable and Interactive Segmentation and
Visualization of Neural Processes in EM Datasets

Example: Registration & Alignment

• Registration at screen/brick resolution

[Beyer et al., CG&A 2013]
Exploring the Connectome – Petascale Volume
Visualization of Microscopy Data Streams

Questions for Part 1?

Next:

(More) Scalable Volume Rendering

THANKS

Webpage:

http://people.seas.harvard.edu/~jbeyer/star.html

HARVARD
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and Applied Sciences

KAUST

Part 2 -
Scalable Volume Rendering

Part 2 - Scalable Volume Rendering

- History

- Categorization

  - Working Set Determination

  - Working Set Storage & Access

  - Rendering (Ray Traversal)

- Ray-Guided Volume Rendering Examples

- Conclusion

History (1)

- Texture slicing [Cullip and Neumann '93, Cabral et al. '94, Rezk-Salama et al. '00]

  + Minimal hardware requirements (can run on WebGL)

  - Visual artifacts, less flexibility

History (2)

- GPU ray-casting [Röttger et al. '03, Krüger and Westermann '03]

  + standard image order approach, embarrassingly parallel

  + supports many performance and quality enhancements

History (3)

- Large data volume rendering

  - Octree rendering based on texture-slicing
    [LaMar et al. '99, Weiler et al. '00, Guthe et al. '02]

  - Bricked single-pass ray-casting
    [Hadwiger et al. '05, Beyer et al. '07]

  - Bricked multi-resolution single-pass ray-casting
    [Ljung et al. '06, Beyer et al. '08, Jeong et al. '09]

  - Optimized CPU ray-casting [Knoll et al. '11]

Examples

Octree Rendering and Texture Slicing

- GPU 3D texture mapping with arbitrary levels of detail

- Consistent interpolation between adjacent resolution levels

- Adapting slice distance with respect to desired LOD (needs opacity correction)

- LOD based on user-defined focus point

[Weiler et al., IEEE Symp. Vol Vis 2000]
Level-Of-Detail Volume Rendering via 3D Textures

Working set determination:	View frustum
Volume representation:	Octree
Rendering:	CPU octree traversal, texture slicing

Bricked Single-Pass Ray-Casting

- 3D brick cache for out-of-core volume rendering

- Object space culling and empty space skipping in ray setup step

- Correct tri-linear interpolation between bricks

[Hadwiger et al., Eurographics 2005]
Real-Time Ray-Casting and Advanced
Shading of Discrete Isosurfaces

Working set determination:	Global, view frustum
Volume representation:	Single-resolution grid (page table)
Rendering:	Bricked single-pass ray-casting

Bricked Multi-Resolution Ray-Casting

- Adaptive object- and image-space sampling

  - Adaptive sampling density along ray

  - Adaptive image-space sampling, based on statistics for screen tiles

- Single-pass fragment program

  - Correct neighborhood samples for interpolation fetched in shader

- Transfer function-based LOD selection

[Ljung, Volume Graphics 2006]
Adaptive Sampling in Single Pass, GPU-based Raycasting of Multiresolution Volumes

Working set determination:	Global, view frustum
Volume representation:	Multi-resolution grid
Rendering:	Bricked single-pass ray-casting

Categorization

- Main questions

  - Q1: How is the working set determined?

  - Q2: How is the working set stored?

  - Q3: How is the rendering done?

Huge difference between 'traditional' and 'modern' ray-guided approaches!

Categorization

Working set determination	Full volume	Basic culling (global attributes, view frustum)		Ray-guided / visualization-driven
Volume data representation	- Linear (non-bricked)	- Single-resolution grid - Grid with octree per brick	- Octree - Kd-tree - Multi-resolution grid	- Octree - Multi-resolution grid
Rendering (ray traversal)	- Texture slicing - Non-bricked ray-casting	- CPU octree traversal (multi-pass) - CPU kd-tree traversal (multi-pass) - Bricked/virtual texture ray-casting (single-pass)		- GPU octree traversal (single-pass) - Multi-level virtual texture ray-casting (single-pass)
Scalability	Low	Medium		High

Q1: Working Set Determination - Traditional

- Global attribute-based culling (view-independent)

  - Cull against transfer function, iso value, enabled objects, etc.

- View frustum culling (view-dependent)

  - Cull bricks outside the view frustum

- Occlusion culling?

Global Attribute-Based Culling

- Cull bricks based on attributes; view-independent
  - Transfer function
  - Iso value
  - Enabled segmented objects

- Often based on min/max bricks
  - Empty space skipping
  - Skip loading of 'empty' bricks
  - Speed up on-demand spatial queries

View Frustum, Occlusion Culling

- Cull all bricks against view frustum

- Cull all occluded bricks

Q1: Working Set Determination – Modern (1)

- Visibility determined during ray traversal
  - Implicit view frustum culling (no extra step required)
  - Implicit occlusion culling (no extra steps or occlusion buffers)

Q1: Working Set Determination – Modern (2)

- Rays determine working set directly
  - Each ray writes out list of bricks it requires (intersects) front-to-back
  - Use modern OpenGL extensions
    (`GL_ARB_shader_storage_buffer_object`, …)

Q2: Working Set Storage - Traditional

- Different possibilities:

  - Individual texture for each brick

    - OpenGL-managed 3D textures (paging done by OpenGL)

    - Pool of brick textures (paging done manually)

  - Multiple bricks combined into single texture

    - Need to adjust texture coordinates for each brick

Q2: Working Set Storage – Modern (1)

- Shared cache texture for all bricks ("brick pool")

Q2: Working Set Storage – Modern (2)

- Caching Strategies
  - LRU, MRU
- Handling missing bricks
  - Skip or substitute lower resolution
- Strategies if the working set is too large
  - Switch from single-pass to multi-pass rendering
  - Interrupt rendering on cache miss ("page fault handling")

Q3: Rendering - Traditional

- Traverse bricks in front-to-back visibility order
  - Order determined on CPU
  - Easy to do for grids and trees (recursive)
- Render each brick individually
  - One rendering pass per brick
- Traditional problems
  - When to stop? (early ray termination vs. occlusion culling)
  - Occlusion culling of each brick usually too conservative

Q3: Rendering - Modern

- Preferably single-pass rendering

- All rays traversed in front-to-back order

- Rays perform dynamic address translation (virtual to physical)

- Rays dynamically write out brick usage information

  - Missing bricks ("cache misses")

  - Bricks in use (for replacement strategy: LRU/MRU)

- Rays dynamically determine required resolution

  - Per-sample or per-brick

Virtual Texturing

- Similar to CPU virtual memory but in 2D/3D texture space

  - Domain decomposition of virtual texture space: pages

  - Page table maps from virtual pages to physical pages

  - Working set of physical pages stored in cache texture

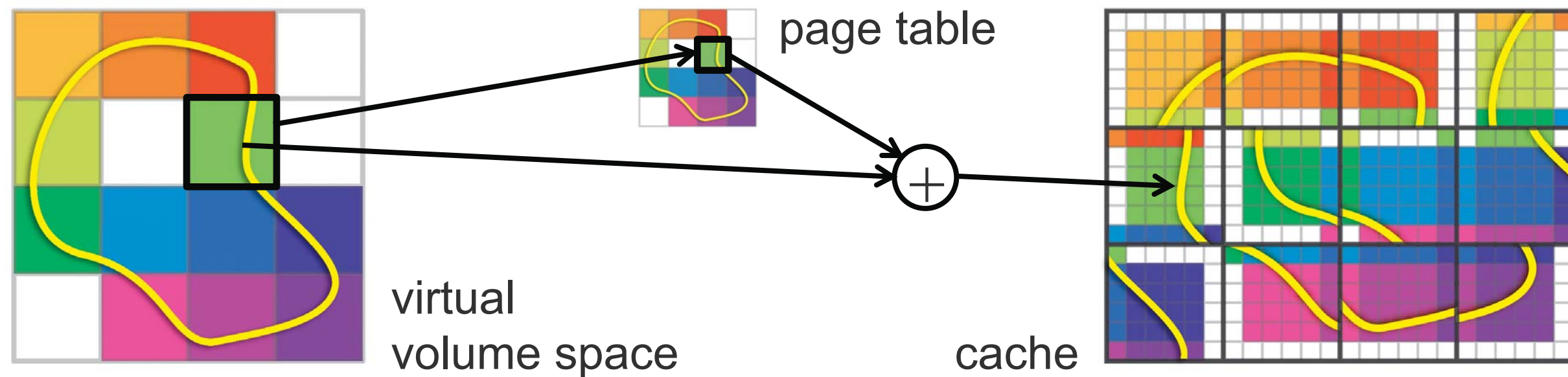virtual image or
volume space

cache

[Kraus and Ertl, Graphics Hardware '02]
Adaptive Texture Maps

Address Translation

- Map virtual to physical address
  - `pt_entry = pageTable[ virtAddx / brickSize ];`
  - `physAddx = pt_entry.physAddx + virtAddx % brickSize;`

page table

virtual
volume space

cache

If cache is a texture, need to transform coordinates to texture domain (scale factor)!

Address Translation Variants

- Tree (quadtree/octree)
  - Linked nodes; dynamic traversal
- Uniform page tables
  - Can do page table mipmap; uniform in each level
- Multi-level page tables
  - Recursive page structure decoupled from multi-resolution hierarchy
- Spatial hashing
  - Needs collision handling; hashing function must minimize collisions
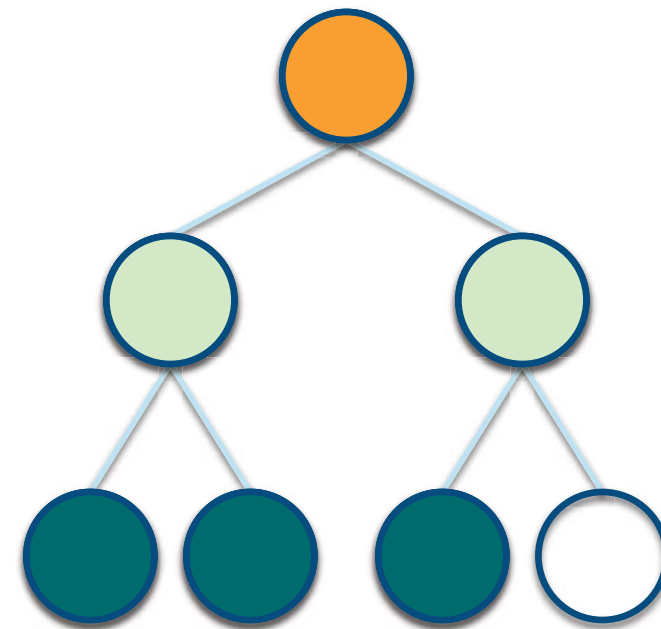
Tree Traversal

- Adapt tree traversal from ray tracing
  - Standard traversal: recursive with stack
  - GPU algorithms without or with limited stack
    - Use "ropes" between nodes [Havran et al. '98, Gobbetti et al. '08]
    - kd-restart, kd-shortstack [Foley and Sugerman '05]

courtesy Foley and Sugerman

Variant 1: Tree Traversal

- Tree can be seen as a 'page table'
  - Linked nodes; dynamic traversal
  - Nodes contain page table entries

"page table hierarchy" (tree) coupled to resolution hierarchy!

Variant 1: Tree Traversal

- Tree can be seen as a 'page table'
  - Linked nodes; dynamic traversal
  - Nodes contain page table entries

does not require full tree!

Variant 2: Uniform Page Tables

- Only feasible when page table is not too large (depends on brick size)
  - For "medium-sized" volumes or "large" page/brick sizes

requires full-size page table!

Variant 3: Multi-Level Page Tables

- Virtualize page tables recursively
  - Same idea as in CPU multi-level page tables
  - Pages of page table entries like pages of voxels
- Recursive page table hierarchy
  - Decoupled from data resolution levels !
  - \# page table levels << \# data resolution levels

page directory
(top-level page table)

page table
(virtual)

data
(virtual)

Multi-Level Page Tables: Multi-Resolution

multi-resolution
page directory

[Hadwiger et al., 2012]

Variant 4: Spatial Hashing (1)

- Instead of virtualizing page table, put entries into hash table
  - Hashing function maps virtual brick to page table entry
  - Hash table size is maximum working set size

working set

Ray-guided Volume Rendering (1)

- Working set determination on GPU

  - Ray-guided / visualization-driven approaches

- Prefer single-pass rendering

  - Entire traversal on GPU

  - Use small brick sizes

  - Multi-pass only when working set too large for single pass

- Virtual texturing

  - Powerful paradigm with very good scalability

Ray-Guided Volume Rendering (2)

- With octree traversal (kd-restart)
  - Gigavoxels [Crassin et al., 2009]
    - Gigavoxel isosurface and volume rendering
  - Tera-CVR [Engel, 2011]
    - Teravoxel volume rendering with dynamic transfer functions
- Virtual texturing instead of tree traversal
  - Petascale volume exploration of microscopy streams [Hadwiger et al., 2012]
    - *Visualization-driven* pipeline, including data construction
  - ImageVis3D [Fogal et al., 2013]
    - Analysis of different settings (brick size, …)

Examples

Early 'Ray-Guided' Octree Ray-Casting (1)

Data structure:

- Octree with ropes

  - Pointers to 8 children, 6 neighbors and volume data

  - Active subtree stored in spatial index structure and texture pool on GPU

[Gobbetti et al., The Visual Computer, 2008]
A single-pass GPU ray casting framework for interactive out-of-core rendering of massive volumetric datasets

Working set determination:	Interleaved occlusion queries
Volume representation:	Octree
Rendering:	GPU octree traversal

Early 'Ray-Guided' Octree Ray-Casting (2)

Rendering:

• Stackless GPU octree traversal (rope tree)

Culling:

• Culling on CPU (global transfer function, iso-value, view frustum)

  • Only nodes that were marked as visible in previous rendering pass refined

  • Occlusion queries to check bounding box of node against depth of last sample during raycasting
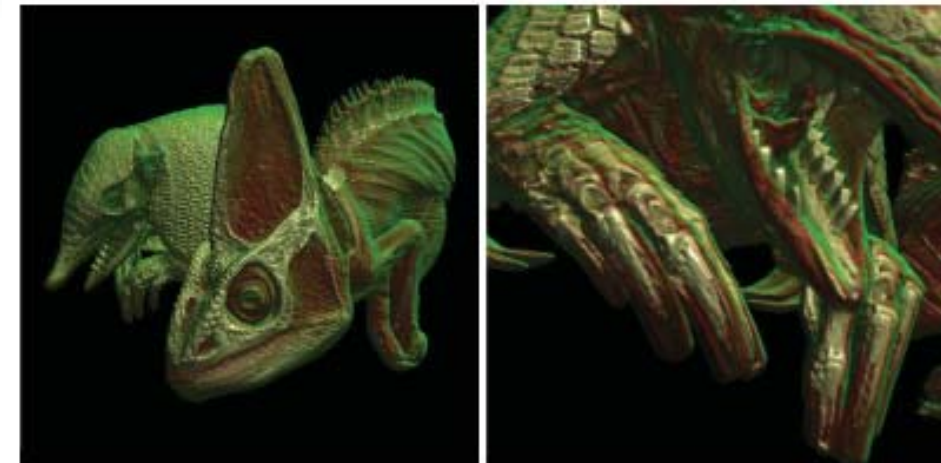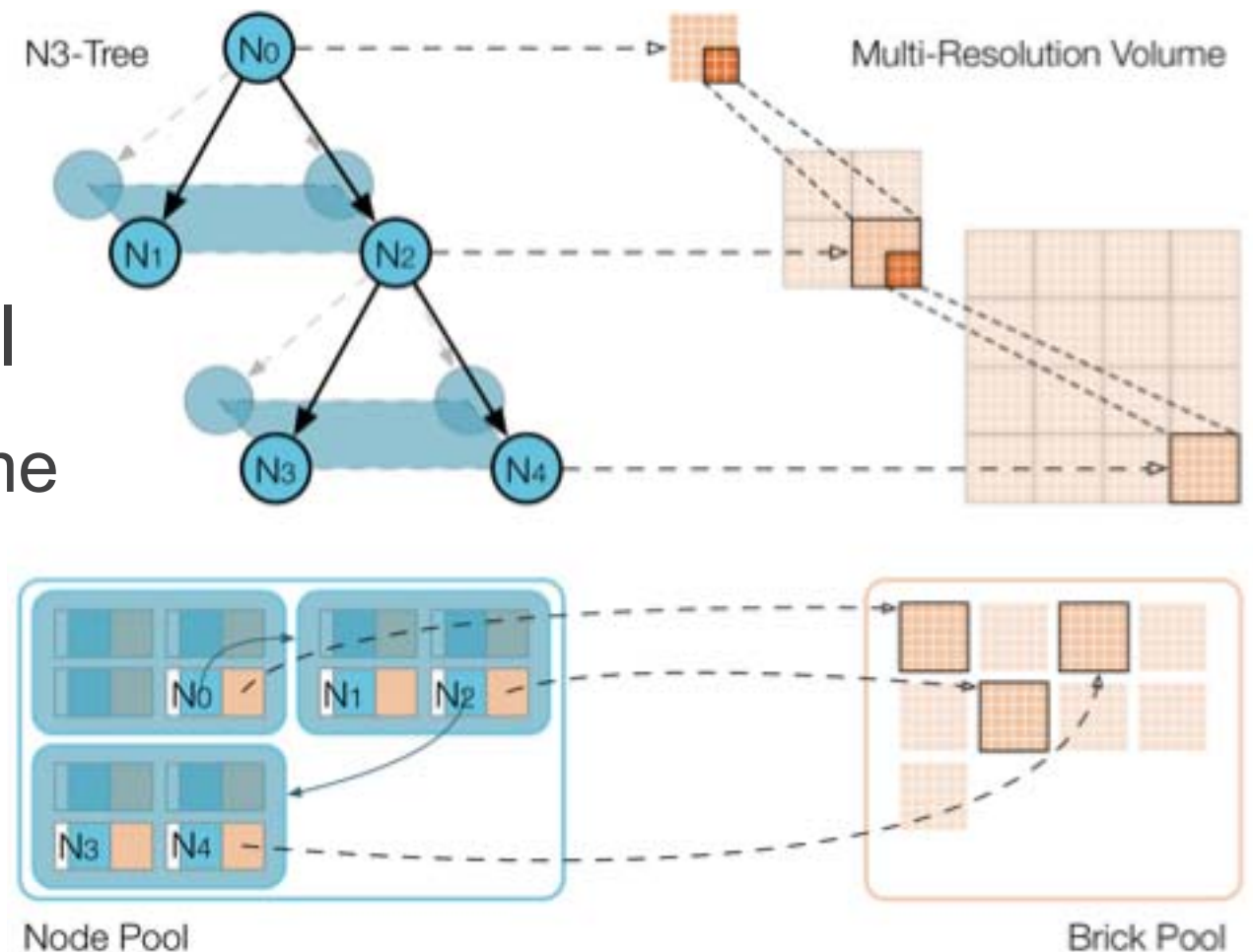
[Gobbetti et al., The Visual Computer, 2008]
A single-pass GPU ray casting framework for interactive out-of-core rendering of massive volumetric datasets

Working set determination:	Interleaved occlusion queries
Volume representation:	Octree
Rendering:	GPU octree traversal

Ray-Guided Octree Ray-Casting (1)

Data structure:

- N^3 tree + multi-resolution volume

- Subtree stored on GPU in node/brick pool

  - Node: 1 pointer to children, 1 pointer to volume brick

  - Children stored together in node pool

[Crassin et al., ACM SIGGRAPH i3D, 2009]
GigaVoxels: Ray-Guided Streaming for Efficient and Detailed Voxel Rendering

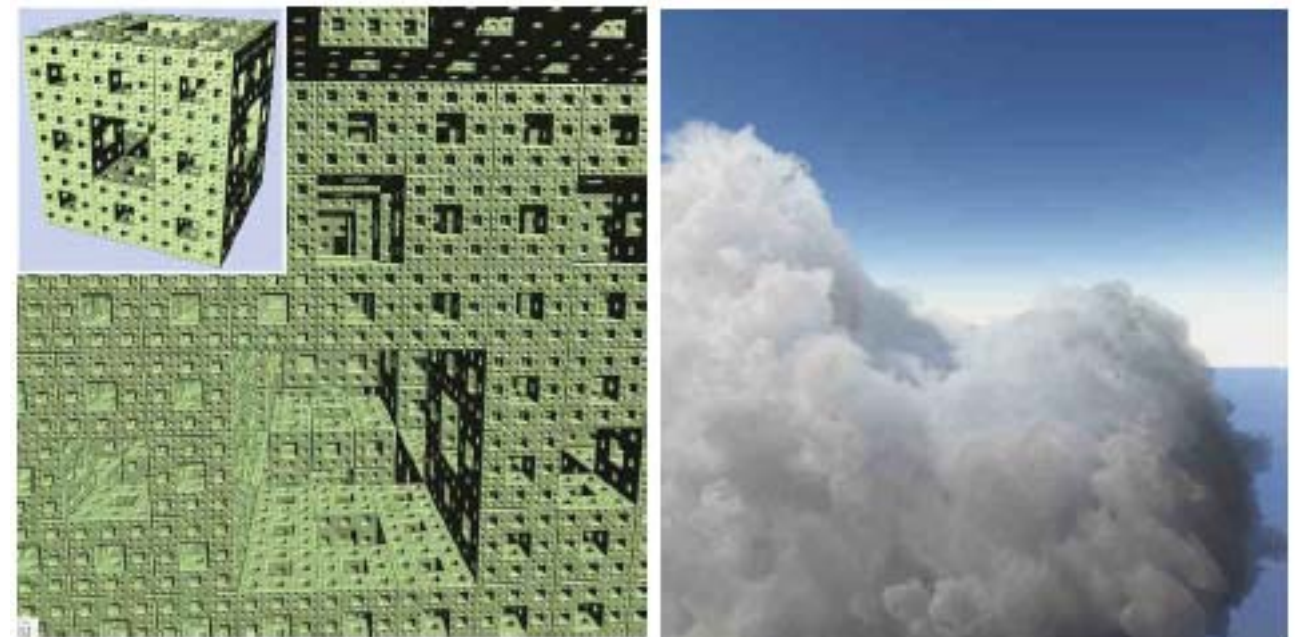| Working set determination:	Ray-guided
Volume representation:	Octree
Rendering:	GPU octree traversal

Ray-Guided Octree Ray-Casting (2)

Rendering:

- Stackless GPU octree traversal (Kd-restart)

- 3 mipmap levels for correct filtering

- Missing data substituted by lower-res data

Culling:

- Multiple render targets write out data usage

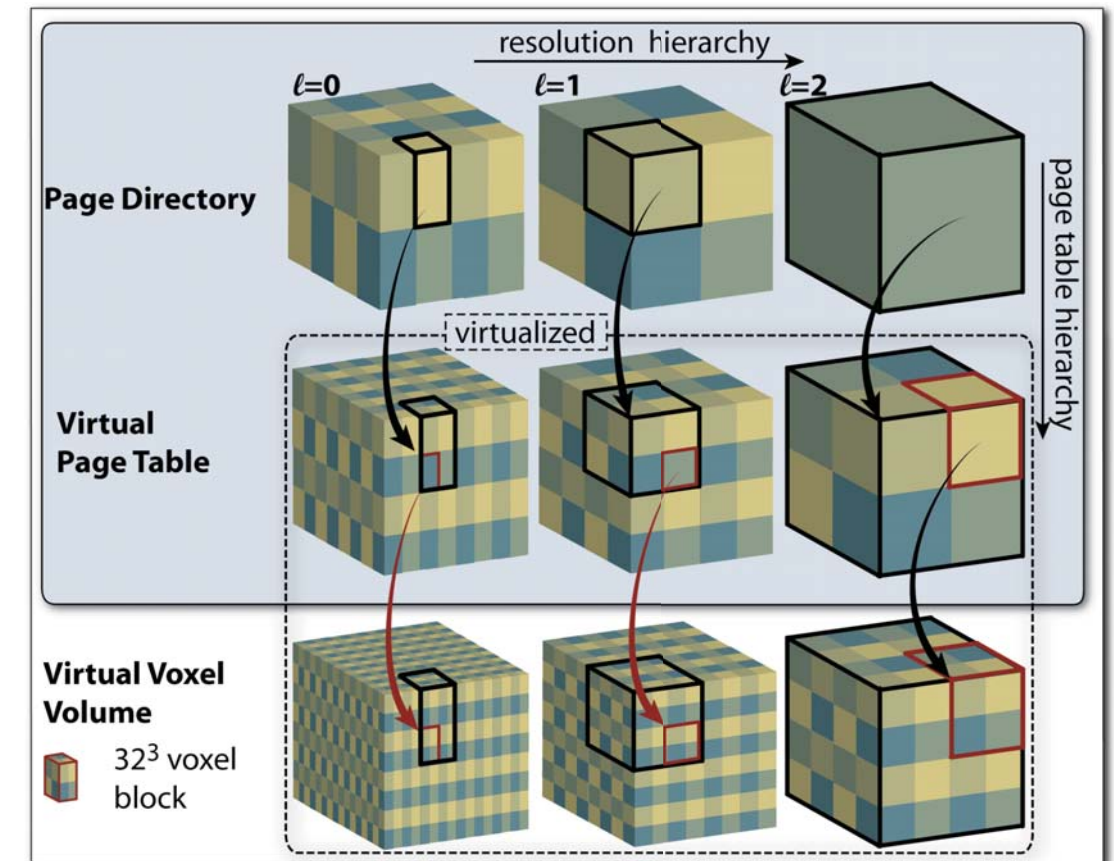 - Exploits temporal and spatial coherence

[Crassin et al., ACM SIGGRAPH i3D, 2009]
GigaVoxels: Ray-Guided Streaming for
Efficient and Detailed Voxel Rendering

Working set determination:	Ray-guided
Volume representation:	Octree
Rendering:	GPU octree traversal

Ray-Guided Multi-Level Pagetable Ray-Casting (1)

Data structure:

- On-the-fly reconstruction of bricks

- Stored on disk in 2D multi-resolution grid (supports highly anisotropic data)

- Multi-level multi-resolution page table on GPU

- Larger bricks for disk access, smaller bricks for rendering

[Hadwiger et al., IEEE SciVis 2009]
Interactive Volume Exploration of Petascale Micro-scopy Data Streams Using a Visualization-Driven Virtual Memory Approach

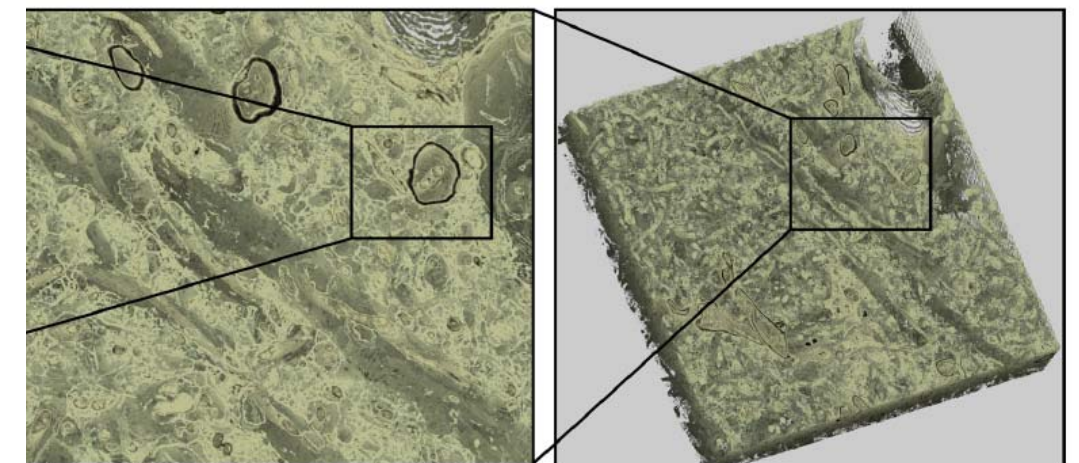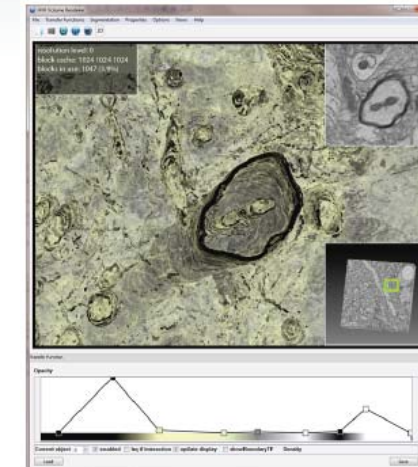| Working set determination:	Ray-guided
Volume representation:	Multi-resolution grid
Rendering:	Multi-level virtual texture ray-casting

Ray-Guided Multi-Level Pagetable Ray-Casting (2)

Rendering:

- Multi-level virtual texture ray-casting

- LOD chosen per individual sample

- Data reconstruction triggered by ray-caster

Culling:

- GPU hash table to report missing blocks

  - Exploits temporal and spatial coherence

[Hadwiger et al., IEEE SciVis 2009]
Interactive Volume Exploration of Petascale Micro-
scopy Data Streams Using a Visualization-Driven
Virtual Memory Approach

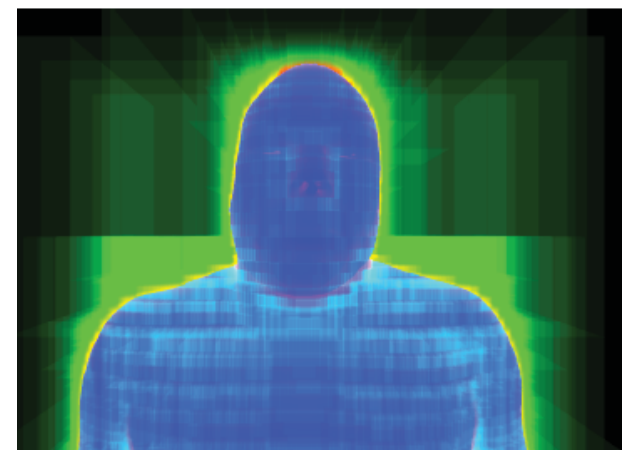| Working set determination:	Ray-guided
Volume representation:	Multi-resolution grid
Rendering:	Multi-level virtual texture ray-casting

Ray-Guided Multi-Level Pagetable Ray-Casting - Analysis

Implementation differences:

- Lock-free hash table, pagetable lookup only per brick

- Fallback for multi-pass rendering

Analysis:

- Many detailed performance numbers (see paper)

- Working set size: typically lower than GPU memory

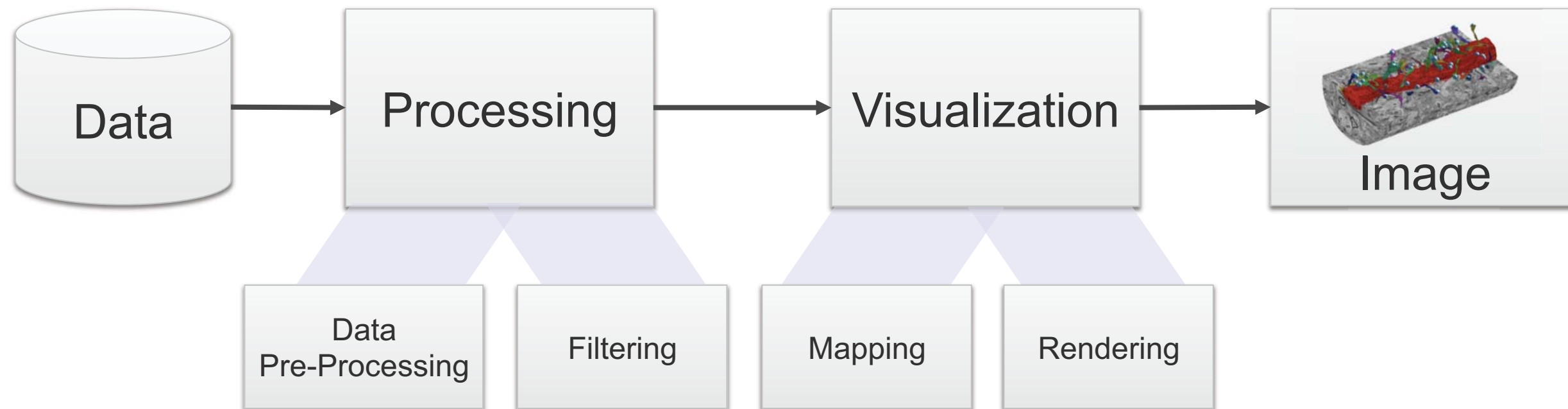- Brick size: larger on disk (>= 64^3), smaller for rendering (16^3, 32^3)

[Fogal et al., IEEE LDAV 2013]
An Analysis of Scalable GPU-Based
Ray-Guided Volume Rendering

Working set determination:	Ray-guided
Volume representation:	Multi-resolution grid
Rendering:	(Multi-level) virtual texture ray-casting

Conclusion

Conclusion (1)

- Many volumes larger than GPU memory
  - Determine, manage, and render working set of visible bricks efficiently

Conclusion (2)

- Traditional approaches

  - Limited scalability

  - Visibility determination on CPU

  - Often had to use multi-pass approaches

- Modern approaches

  - High scalability (output sensitive)

  - Visibility determination (working set) on GPU

  - Dynamic traversal of multi-resolution structures on GPU

Conclusion (3)

- Orthogonal approaches
  - Parallel and distributed visualization
  - Clusters, in-situ setups, client/server systems

- Future challenges
  - Web-based visualization
  - Raw data storage

THANKS

Webpage:

http://people.seas.harvard.edu/~jbeyer/star.html