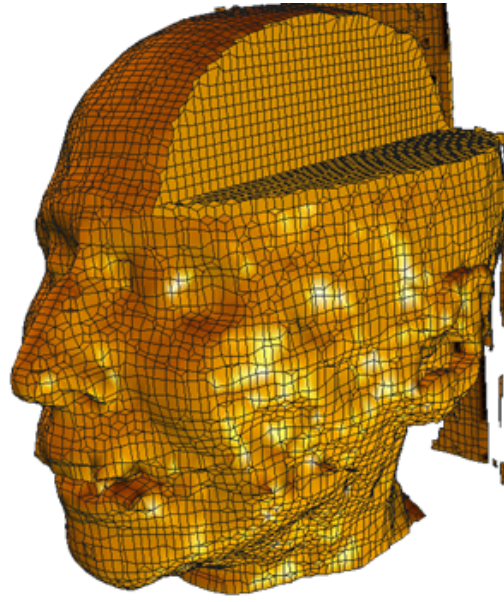
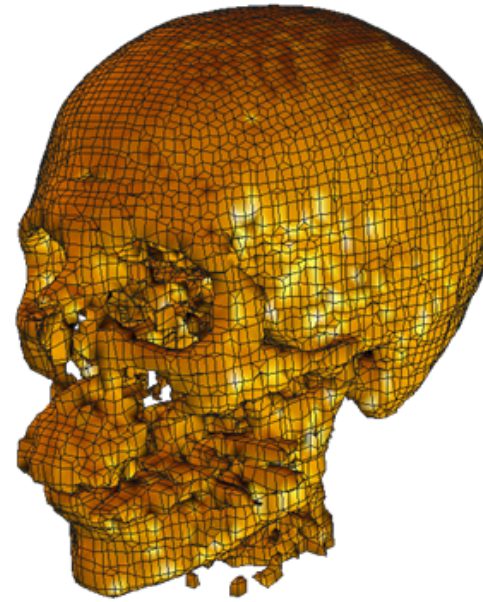


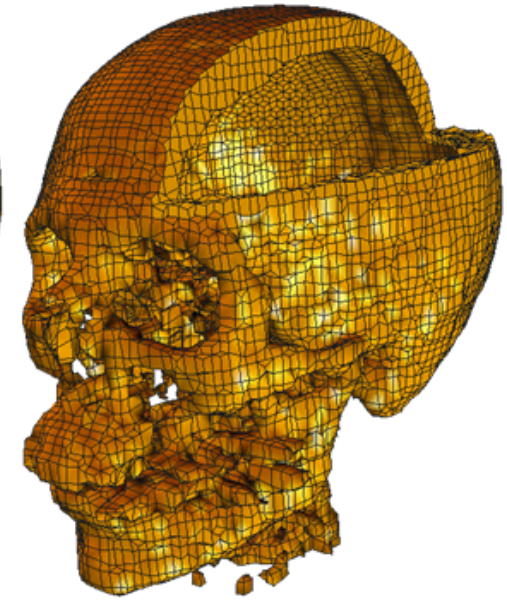
(a)



(b)



(c)



(d)

Zhang et al. 2004

Geometry From Imaging Data

Part II: Geometry Reconstruction

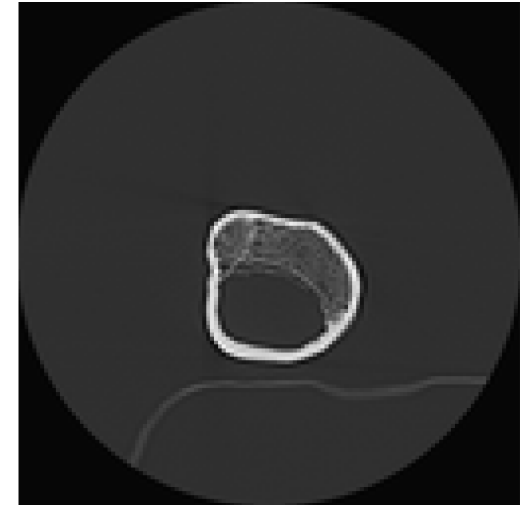
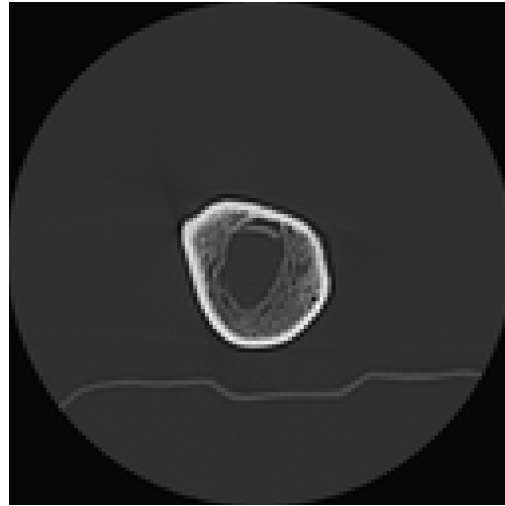
Computational Biomechanics

Summer Term 2016

Lecture 5/12

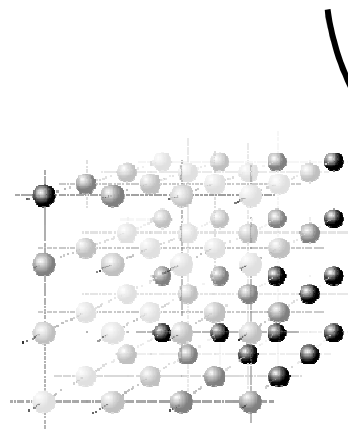
Frank Niemeyer

Starting Point



Note: intentionally reduced resolution

Images made of pixels (DICOM, PNG, ...)



wikimedia.org

Stack

Volume defined by
voxels $I(x, y, z)$

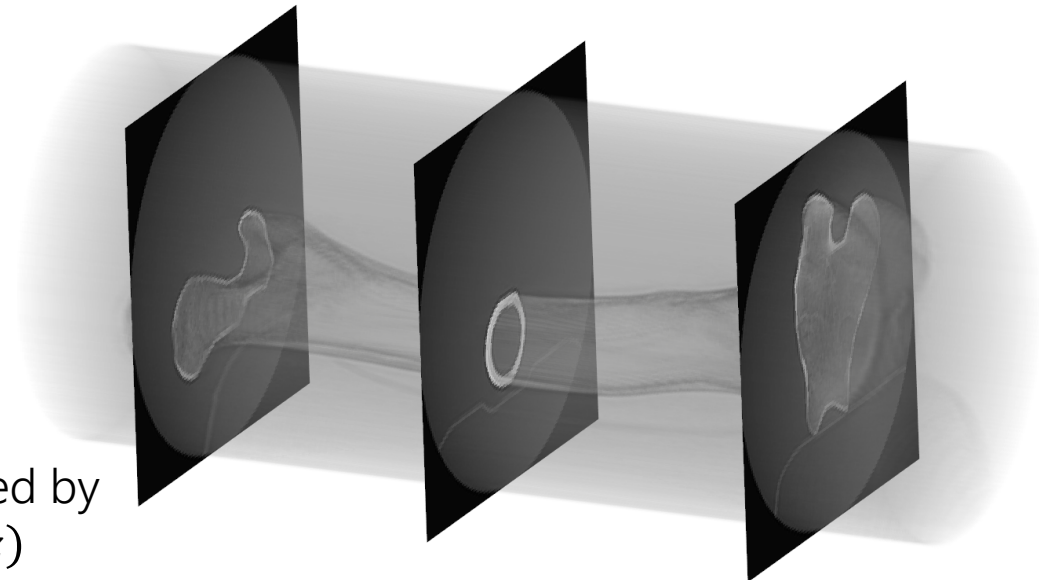
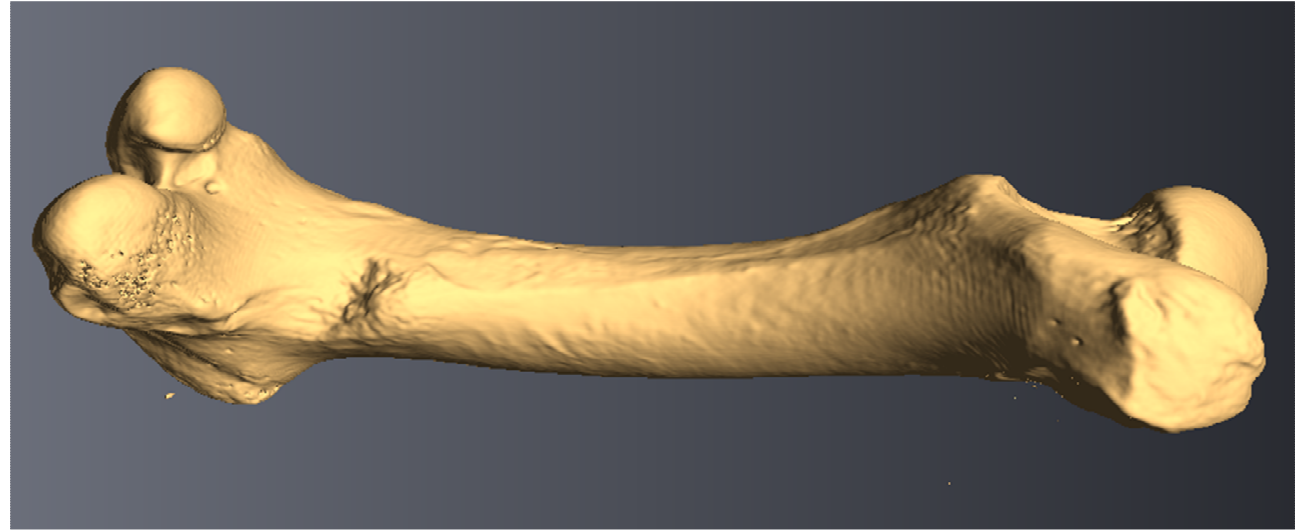
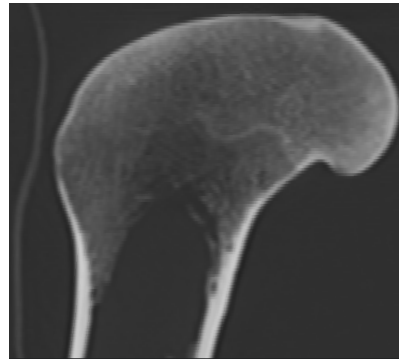


Image Processing

- Visualization
 - Multi-planar
 - Volume rendering
 - Iso-surface
- Cropping (ROI)
- Windowing
- De-noising
- Artifact removal
- Segmentation, labelling, classification:
 - Identify connected components
 - Locate object boundaries, interfaces
 - Manually or (semi-)automated, machine learning



Iso-surface rendering of sheep femur



Original



Windowed, equalized

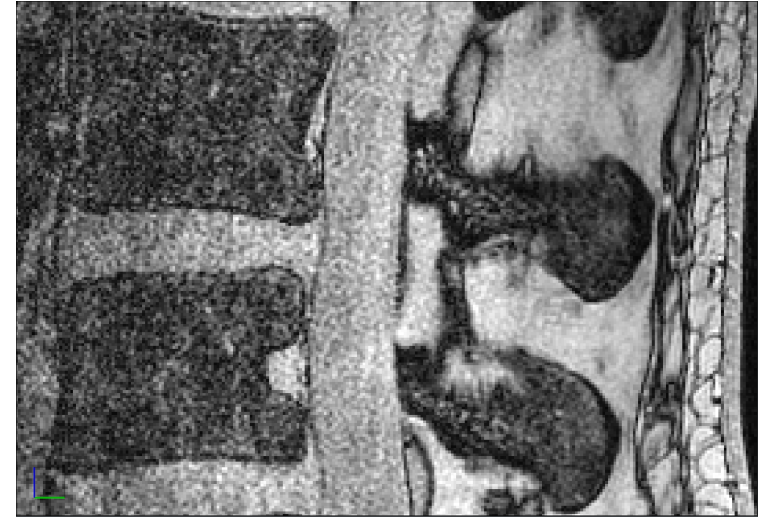


Segmented

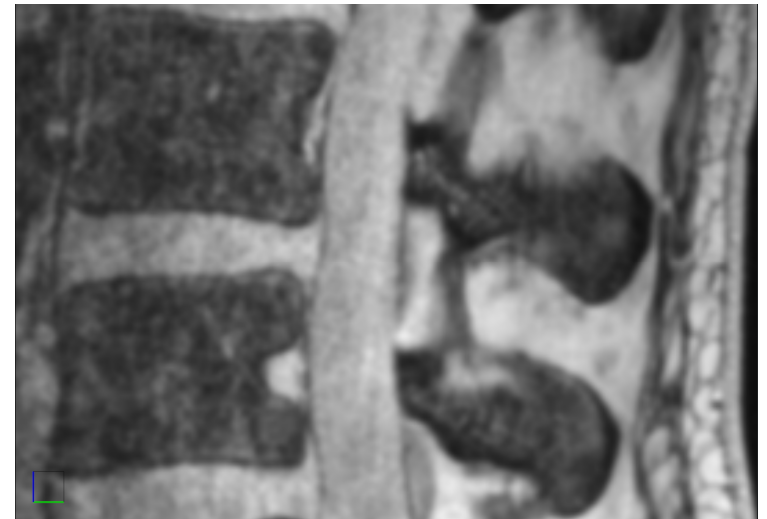
Linear (Low-Pass) Filtering

- Replace each pixel with weighted average of surrounding pixels/voxels (2D or 3D)
- Equivalent to suppressing high-frequency components as $I * K = \mathcal{F}(I) \cdot \mathcal{F}(K)$
- K : kernel (weights), e.g. Gaussian, Lanczos...
- Simple & fast
- Removes noise ...
- ... but also blurs edges

Original



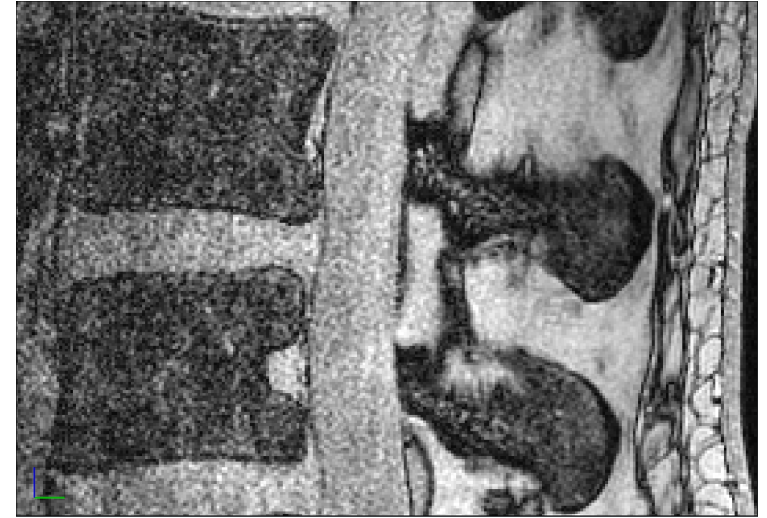
Gaussian filter



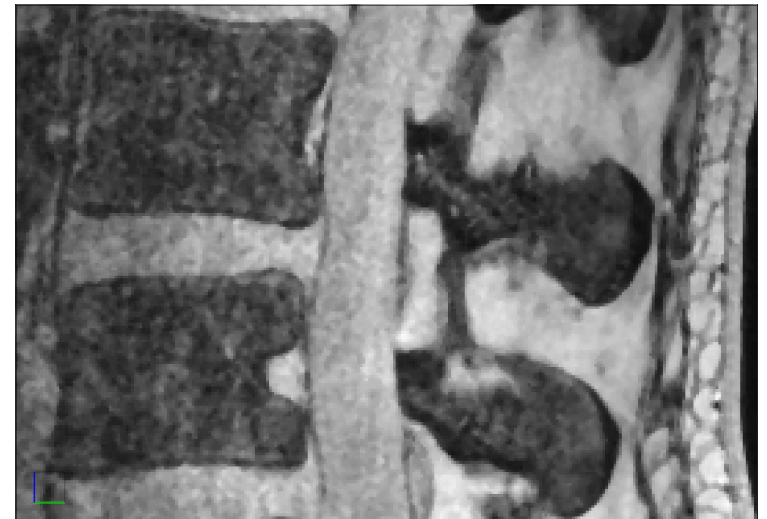
Rank Filters

- Replace each pixel with min/max/median (or any other “rank”) of adjacent pixels/voxels
- Arbitrary adjacency (box, diamond, cross, ...)
- Median: “robust” estimator → particularly useful against impulse-like noise (speckle, salt & pepper)
- Better at preserving edges than Gaussian
- Relatively expensive (sorting!)

Original



Median filter



Bilateral Filter

- Replace each pixel with weighted average of surrounding pixels/voxels (2D or 3D)
 - Weights depend on distance (c.f. linear filter)
 - ... but also on *similarity* to current pixel

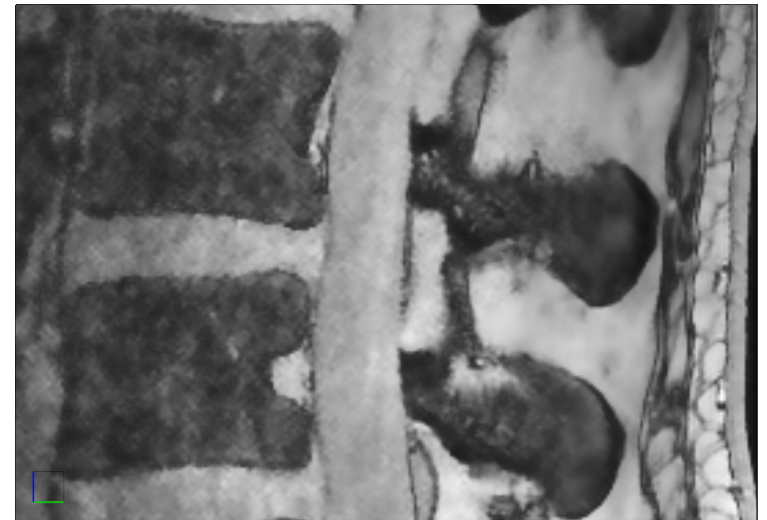
$$\mathcal{B}(I)(\mathbf{x}) \propto \sum_{\mathbf{x}_i \in \mathcal{N}(\mathbf{x})} I(\mathbf{x}_i) K_r(|I(\mathbf{x}_i) - I(\mathbf{x})|) K_s(\|\mathbf{x}_i - \mathbf{x}\|)$$

- K_r : range kernel, K_s : spatial kernel, $\mathcal{N}(\mathbf{x})$: neighborhood
- Preserves edges ...
- ... but not gradients (“gradient reversal”)
- Introduces “staircase effect” (cartoon-look)

Original



Bilateral filter



Anisotropic Diffusion

- Observe: Convolution with Gaussian kernel G
= solution of an isotropic, homogeneous diffusion equation

$$\partial_t I = D \nabla^2 I \Rightarrow I(t + \Delta t) = I(t) * G(D, \Delta t)$$

- Preserve edges by making the diffusion tensor D depend on the image

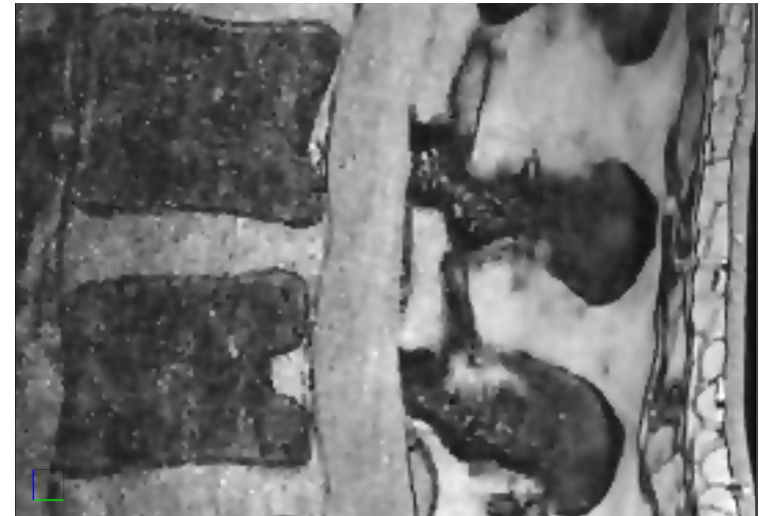
$$\partial_t I = \nabla \cdot (D(I) \nabla I)$$

- Expensive (non-linear PDE)
- Can even be edge-*enhancing* (a.k.a. coherence enhancing diffusion)

Original



Anisotropic diffusion
(actually probably Perona-Malik,
i.e. inhomogeneous but isotropic)



Non-Local Means

- Idea: Exploit self-similarity of images
- Replace each pixel with average of *all other pixels*, weighted by similarity of their neighborhoods

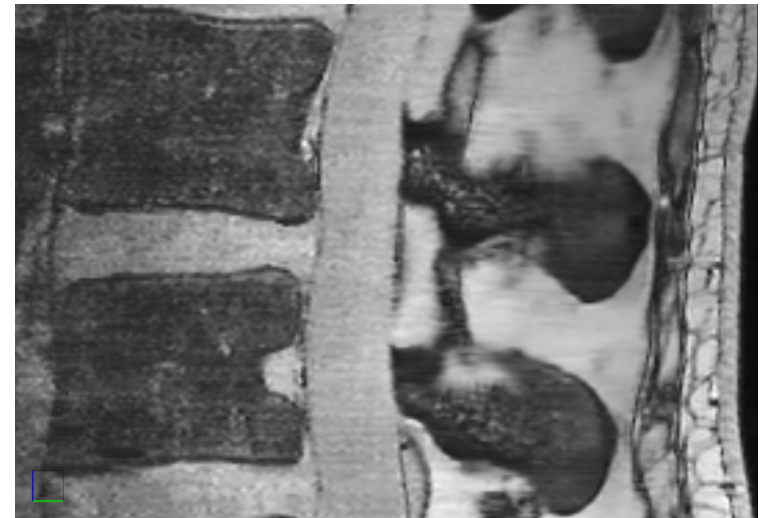
$$\mathcal{M}(I)(i) \propto \sum_j G(d_{ij})I(j)$$

- with $d_{ij} = \|\mathbf{n}_i - \mathbf{n}_j\|^2$
- \mathbf{n}_i : gray values of neighborhood of pixel i
- G : Gaussian (e.g.)
- Less loss of detail than local methods
- Incredibly expensive \rightarrow GPGPU
- Assumes white noise

Original



Windowed
non-local means
(only in transversal plane)

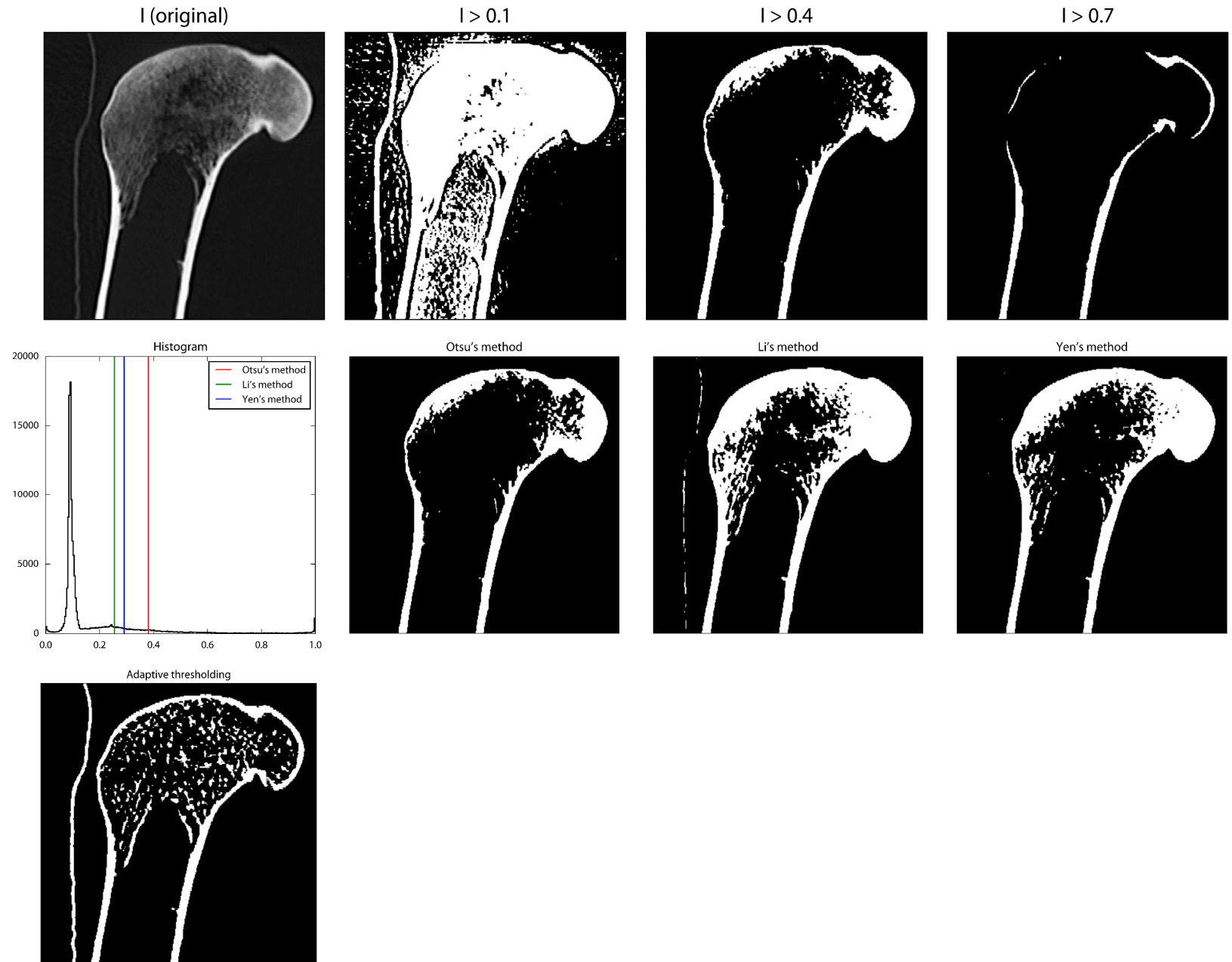


Segmentation Techniques

- Goal: Partition image according to some *similarity measure* into connected subsets of pixels (segments, regions) I_R . Examples for such measures:
 - $\max I_R - \min I_R$
 - $\max(|\bar{I}_R - \max I_R|, |\bar{I}_R - \min I_R|)$
 - $\max\|\nabla I_R\| - \min\|\nabla I_R\|$
 - Others: Texture variance, entropy, energy, statistics of derivatives ...
- Pixel-based: Thresholding, clustering
- Region-based: Region growing, split & merge, watershed transform, texture segmentation
- Edge-based: Edge detection and linking
- Model-based: Template matching, active contours, level set, ANN ...

Thresholding

- Manually pick a global gray value threshold
- Histogram-based, automatic (Otsu's, Li's, Yen's methods)
- Adaptive (local) thresholding based on local neighborhood

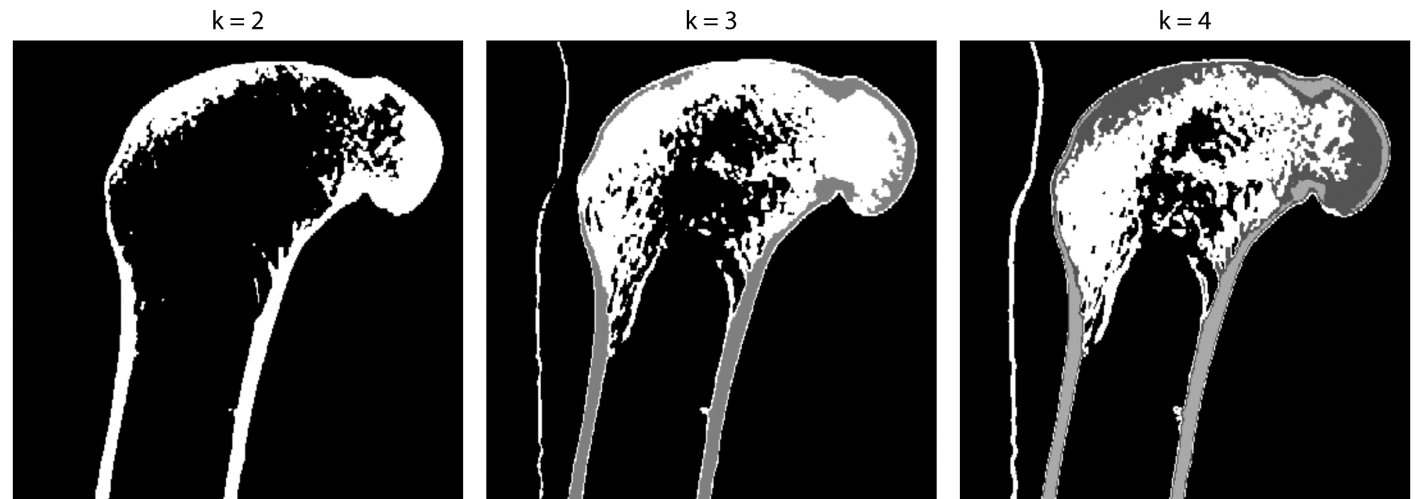
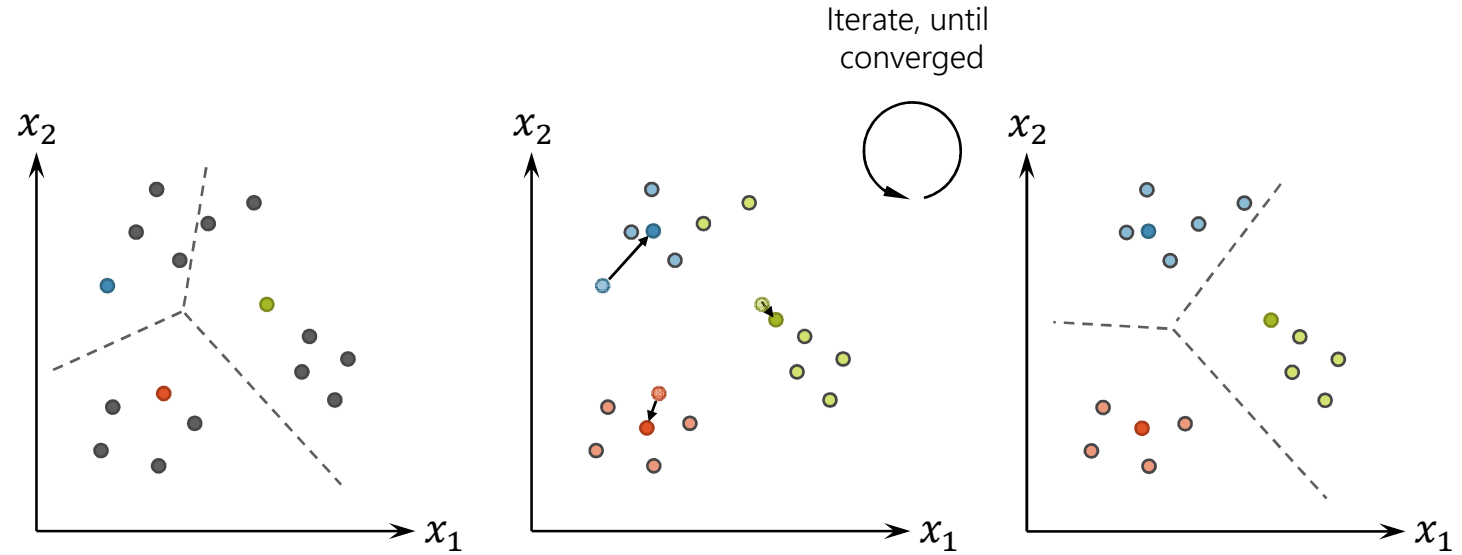


Clustering

- Assumptions: Segments share similar features and are thus clustered in "feature space" → cluster analysis
- k -Means: partition pixels into a set of k Clusters \mathcal{C} with means μ_i , such that

$$\sum_{i=1}^k \sum_{x \in C_i} \|x - \mu_i\|^2 \rightarrow \min$$

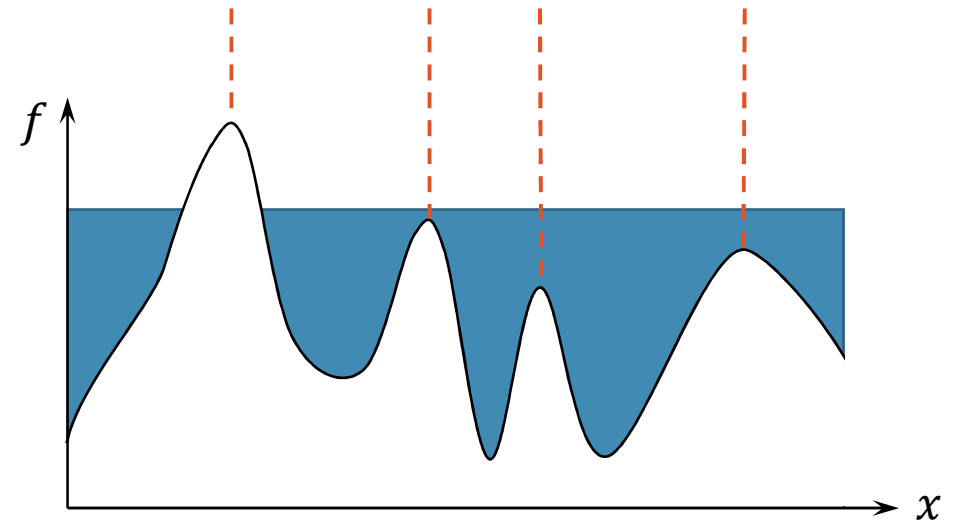
- Assumes convex & isotropic clusters of similar size
- Number of clusters?
- Sensible metric?



Using scikit-learn, `sklearn.cluster.KMeans`, 1D feature space (gray value)

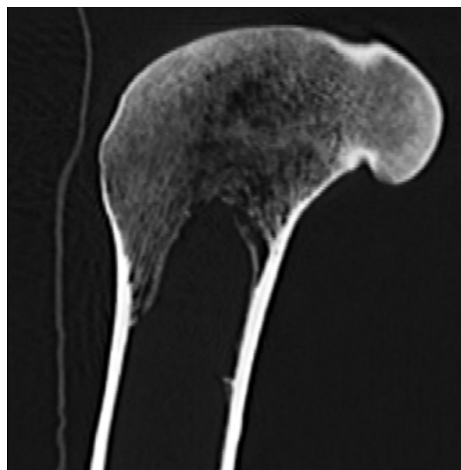
Watershed Transform

- Intuition:
 - Interpret f (some function of the image I , e.g. $\|\nabla I\|_2$) as height map
 - "Hole" at each local minimum of f
 - Immerse landscape in water → "water reservoirs" (regions)
 - To avoid merging of adjacent regions → build "dam" (contours)
- f must be constructed such that "valleys" = objects to segment
- Pros: intuitive, always creates closed contours
- Cons: susceptible to noise, over-segmentation



Watershed Transform

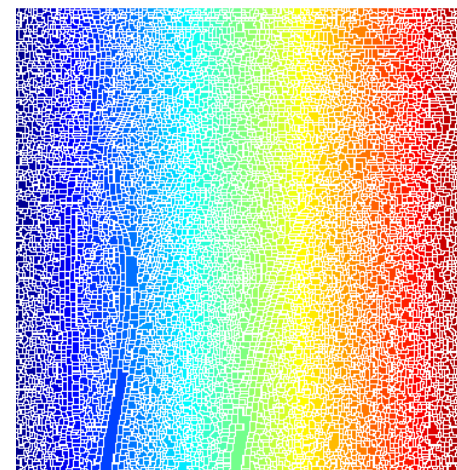
Classic



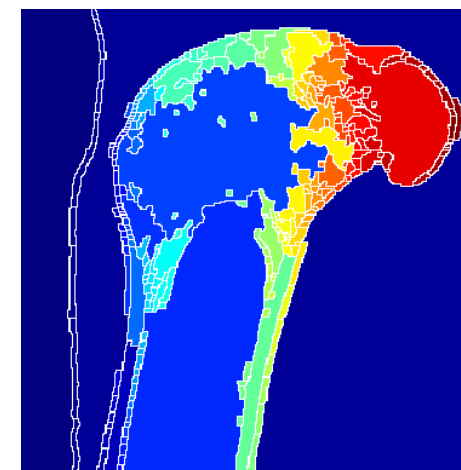
Original image



Gradient magnitude
(Sobel)

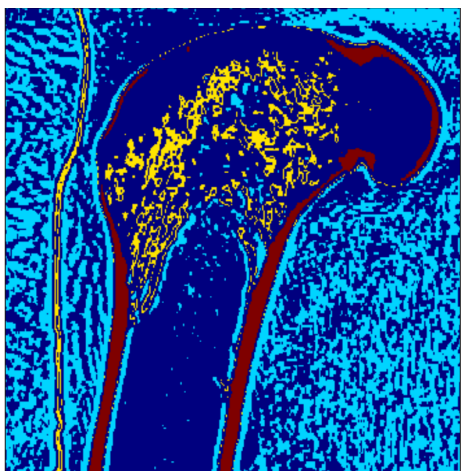


Over-segmented
image



H-min transformed
segmented image

With markers



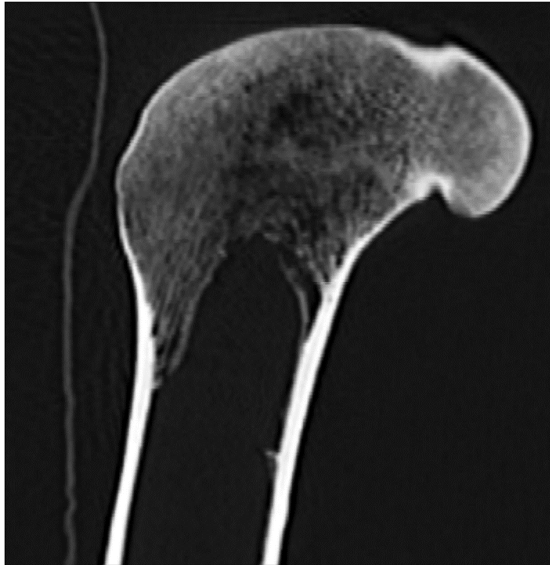
Marker pixels



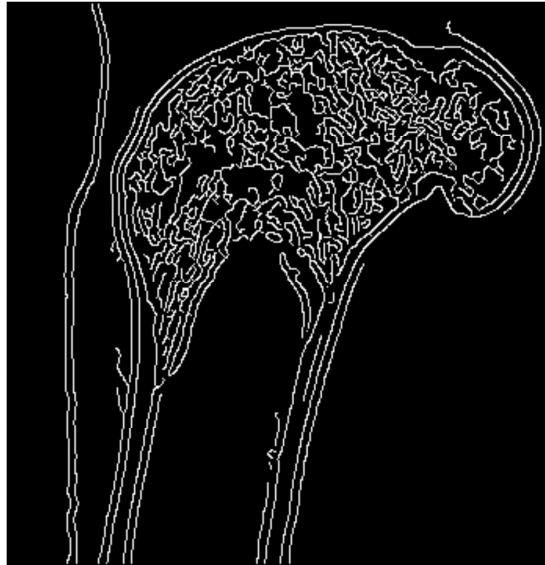
Segmented image

Canny Edge Detection

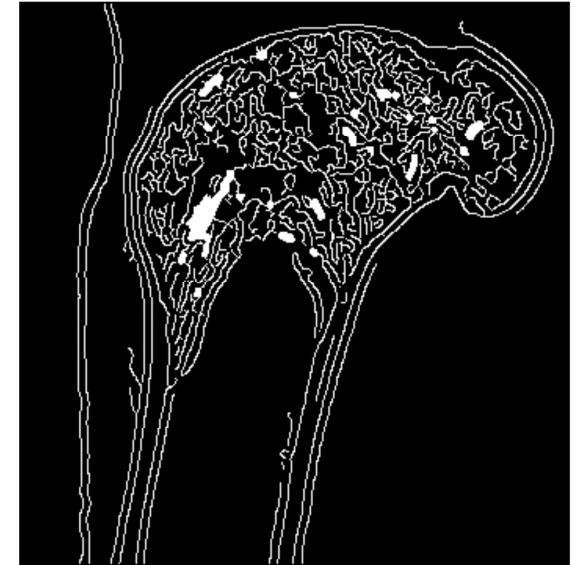
- Basic algorithm:
 - Smooth image with Gaussian
 - Compute image gradient (Sobel)
 - Edge thinning via non-maximum suppression
 - Hysteresis thresholding to suppress weak/unconnected edge pixels



Original image



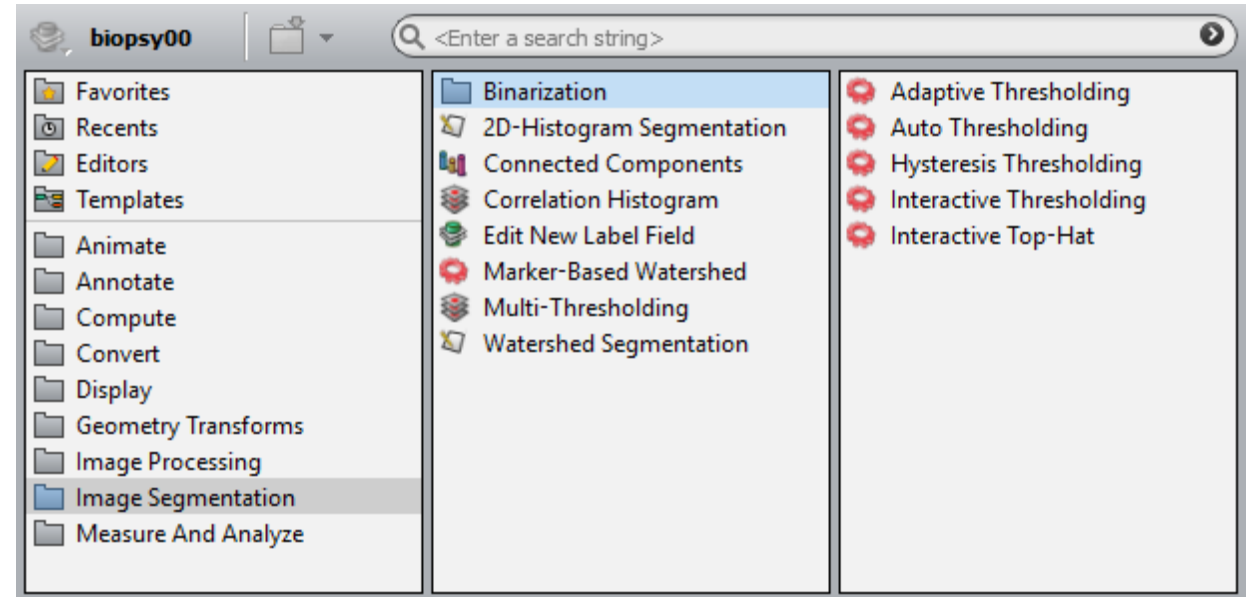
Detected edges



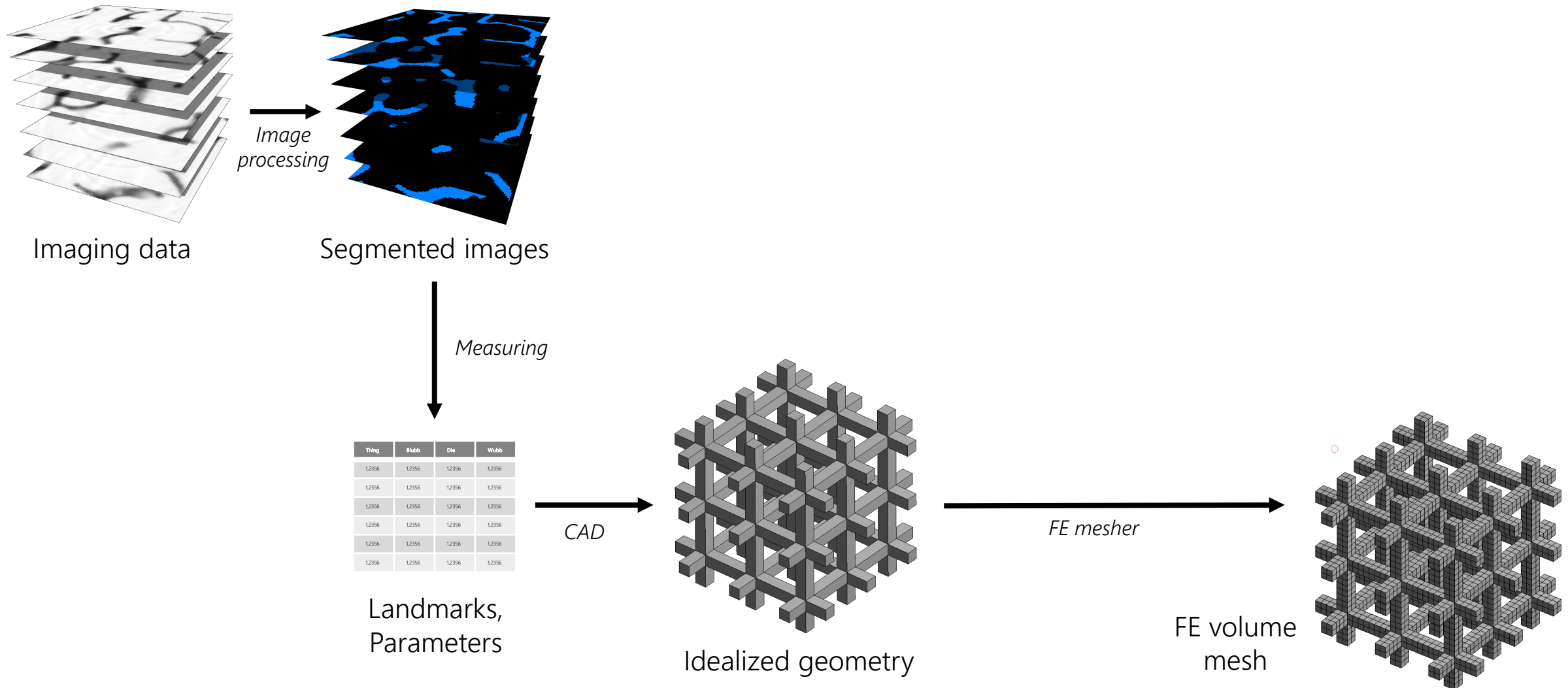
Filled closed regions

Avizo Segmentation Tools

- Automatic, e.g.
 - (Multi-)Thresholding
 - Joint histogram
 - (Marker-based) watershed
- Interactive, semi-automatic (“Segmentation Editor”)
 - Brush
 - Magic Wand (region growing)
 - Propagating Contour (active contour)
 - Watershed
- “Note that even with the advanced tools provided in Avizo, image segmentation can be a time-consuming process! Due to limited main-memory and for performance reasons, there is only a limited undo space for 2D and 3D interaction. Therefore it is highly recommended to *frequently save the label field* during the process of segmentation.” (Avizo Manual)

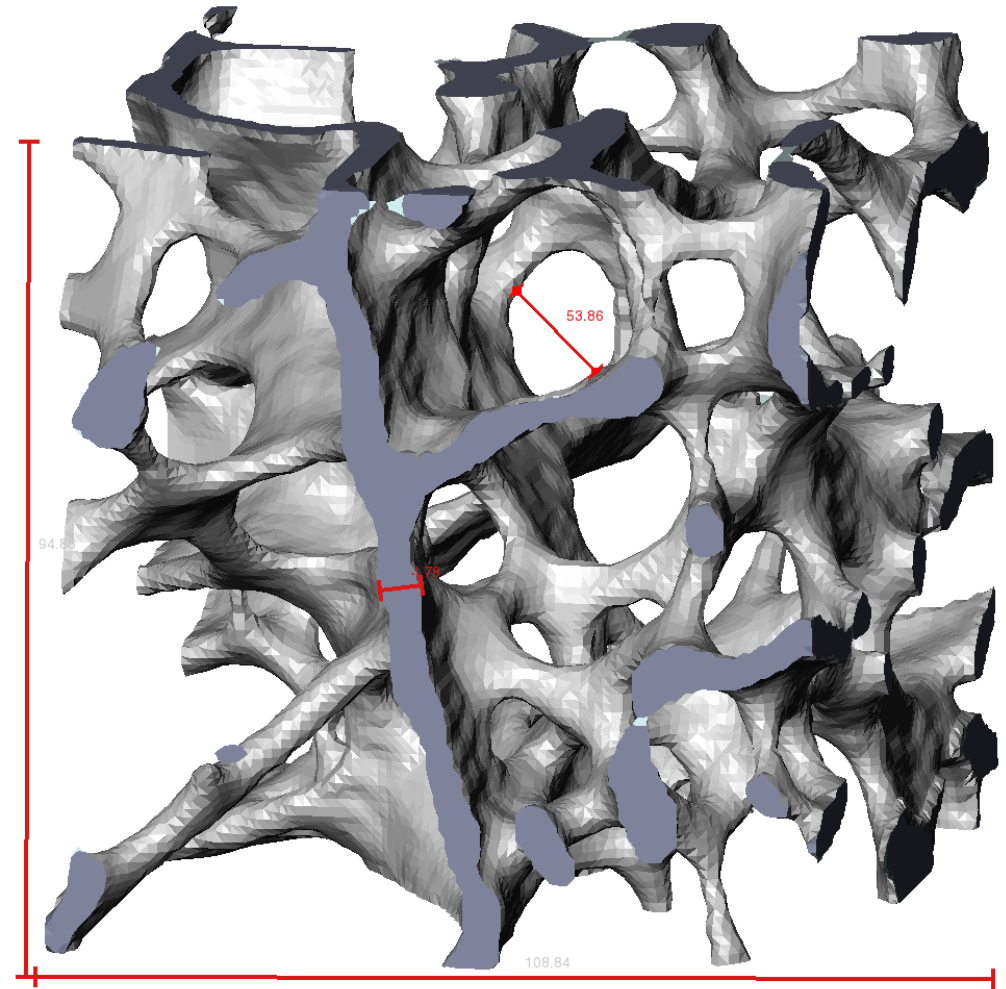


Geometry Reconstruction Approaches



“Bottom-Up” (Solid Modeling)

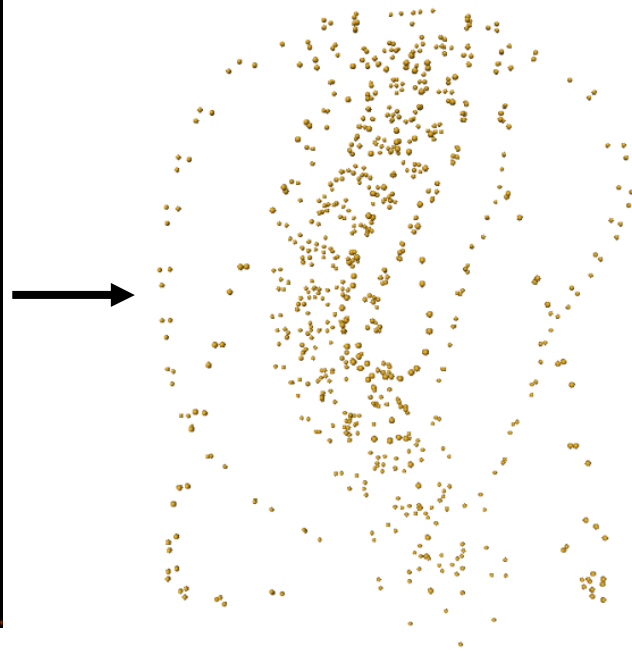
1. Identify landmarks
2. Derive geometry parameters
3. Generate solid geometry
 - Points → Lines → Faces → Volumes
 - Boolean operations (CSG)
4. Mesh geometry
 - Pros & Cons
 - Flexible: parametric, generic
 - Doesn't necessarily require full volumetric data (but prior knowledge)
 - Simplified, idealized anatomy
 - May be tedious to create (depending on level of detail)



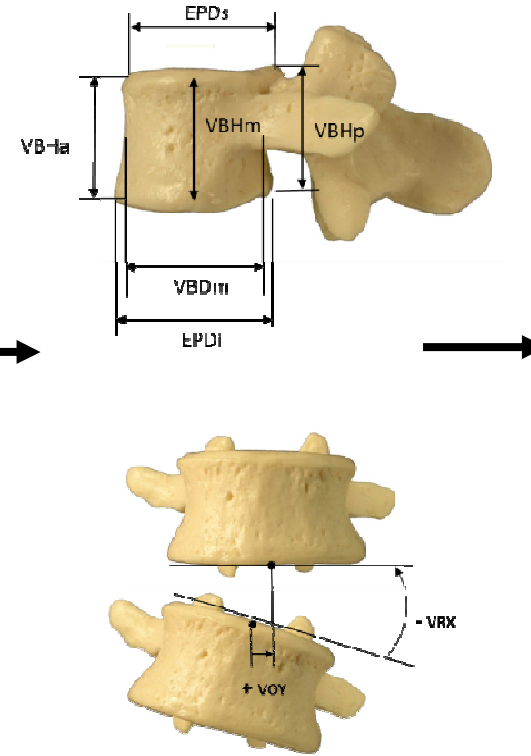
(Semi-Automated) Bottom-Up



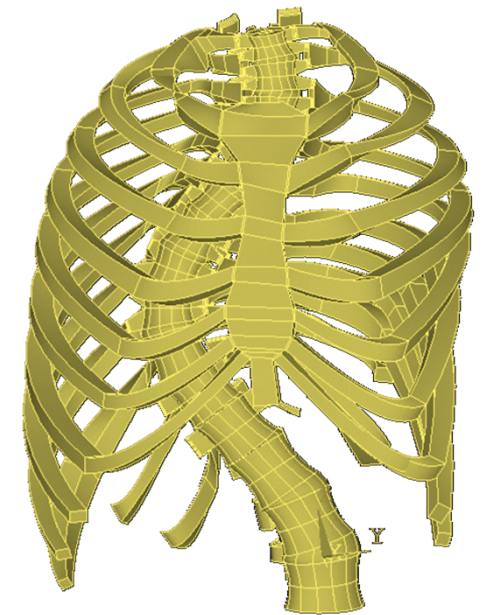
Imaging data



Landmarks

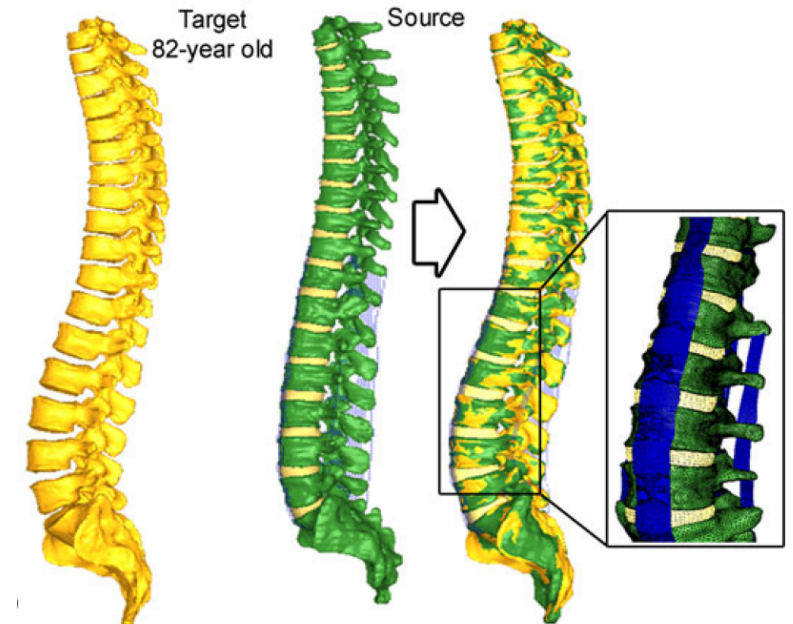
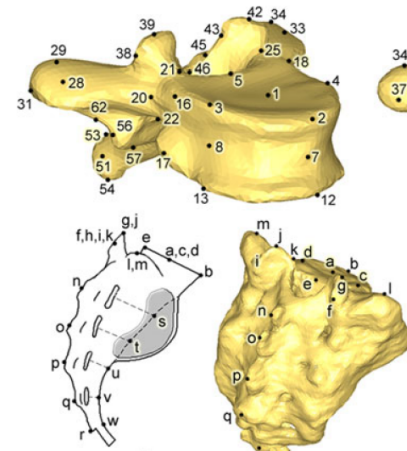
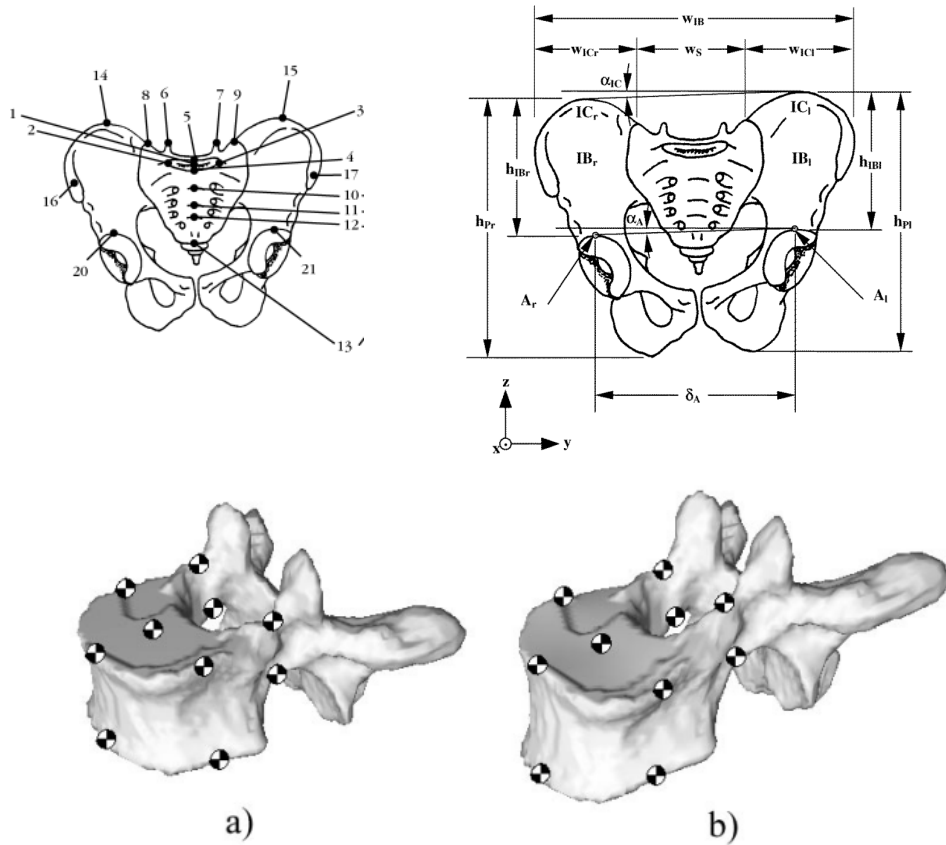


Parameters



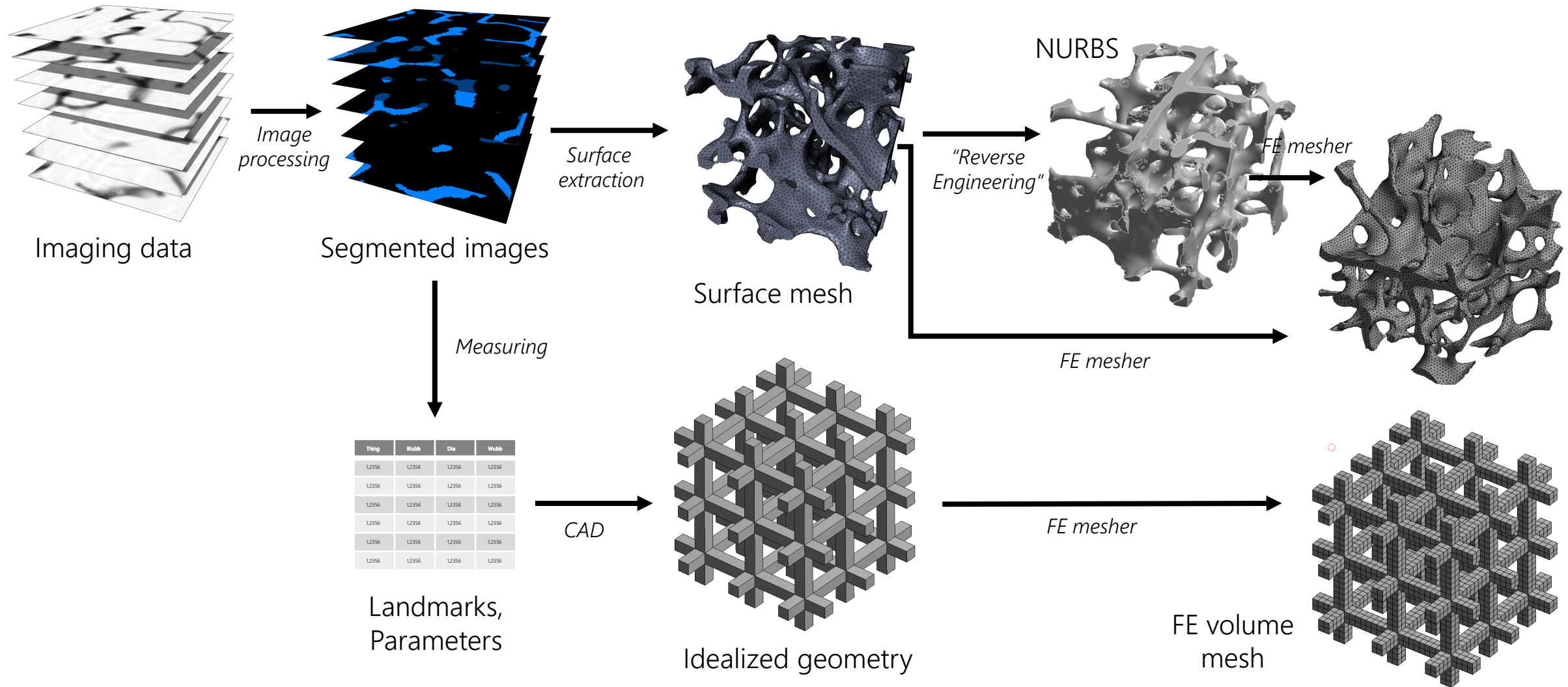
Solid geometry

Template Morphing



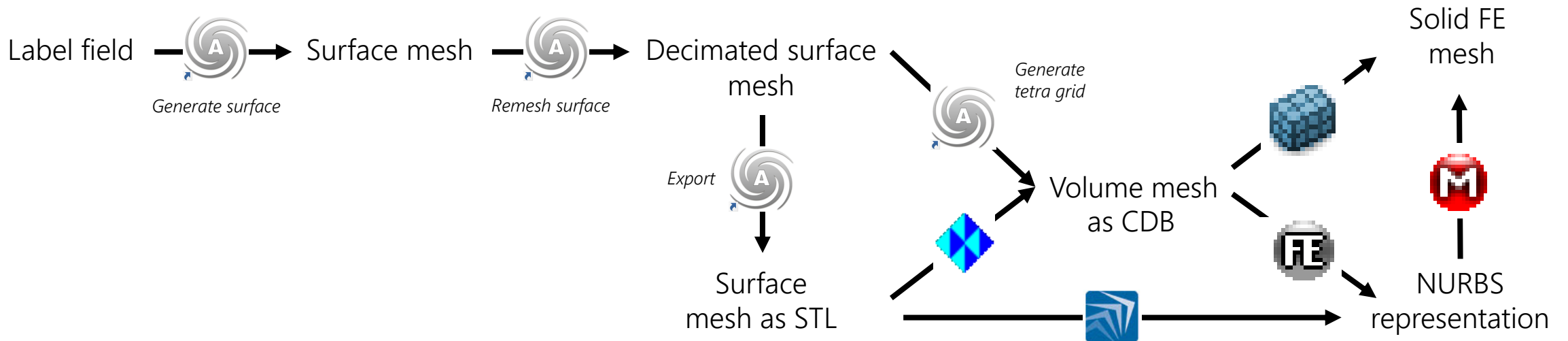
Geometry Reconstruction

Approaches



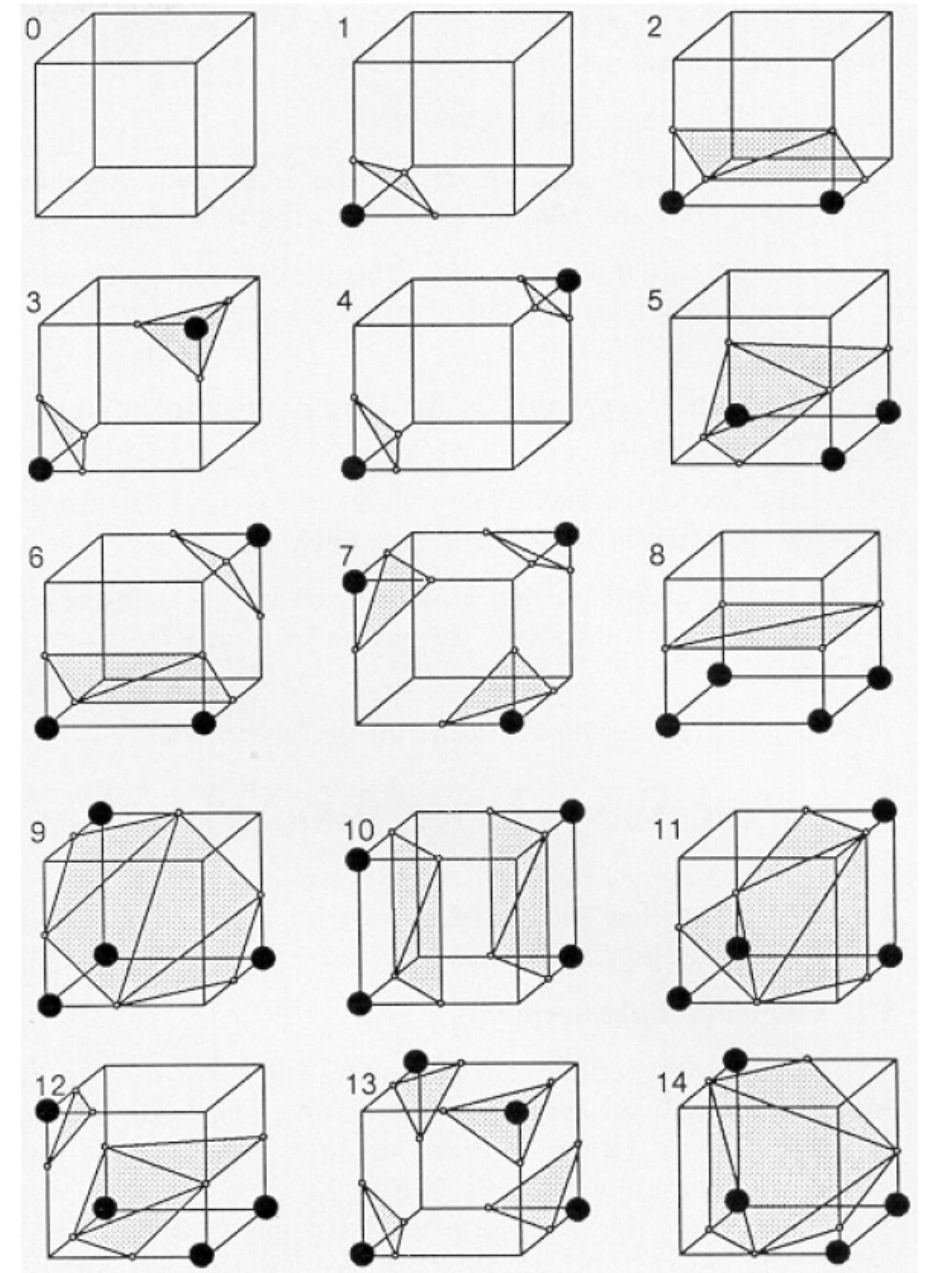
"Avizo-to-ANSYS" Workflow(s)

- Segmentation → label field
- Generate surface mesh → triangulated surface
- Remesh surface → decimated mesh
- Create volume mesh → tetrahedral grid
- Either use that directly ("External model"), or
- Try to reverse engineer NURBS representation

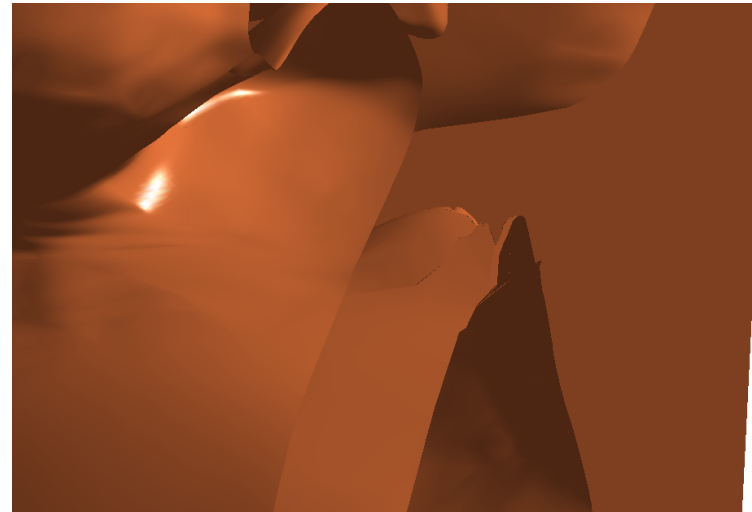
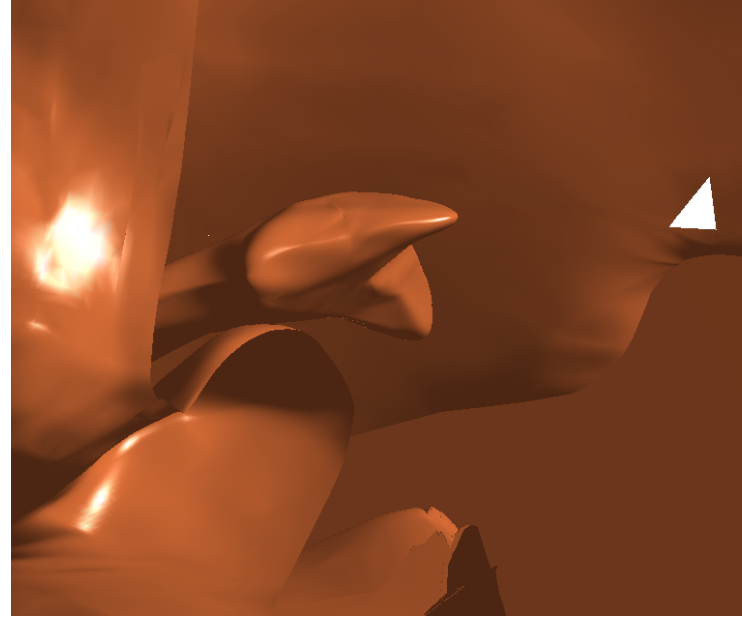
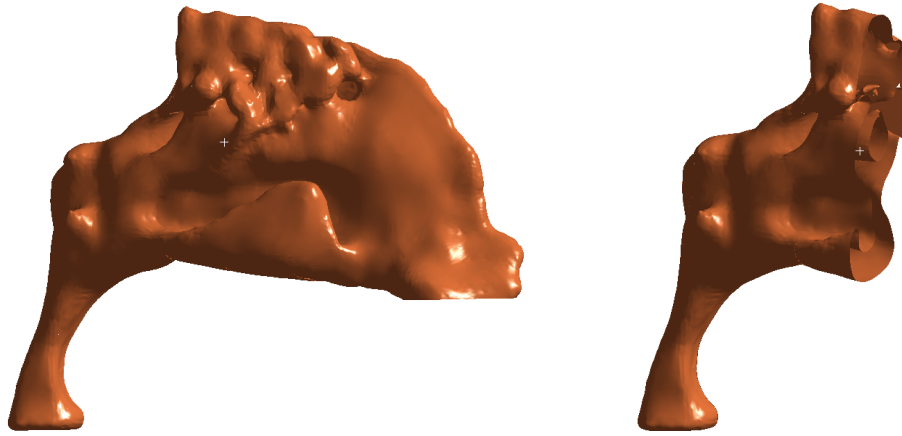


Marching Cubes

- Iterate over voxel grid à cubes of 8 voxels each
- Choose triangulation depending on voxel values:
 - All values above/below threshold: no surface
 - Otherwise: look-up table → triangulation
- Interpolate between voxels to estimate exact surface-edge-intersection
- Results in triangle mesh
 - Facetted, piecewise-linear, non-smooth (C^0 cont.)
- Sensitive to noise
- Often requires repairing if used in non-visualization contexts



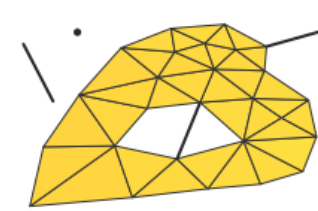
Marching Cubes



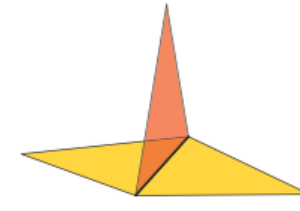
Thanks to Lucas Engelhardt

Mesh Repairing

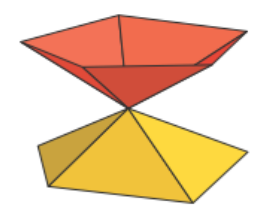
- Fill holes and self-intersections
- Fix singular edges, vertices
- Remove non-manifold edges
- Remove unconnected components
- Fix inconsistent normal direction
- Smoothing
- Software: e.g. Avizo's surface editor, MeshLab, Meshmixer, netfabb, Blender, SpaceClaim...
 - c.f. <http://meshrepair.org>



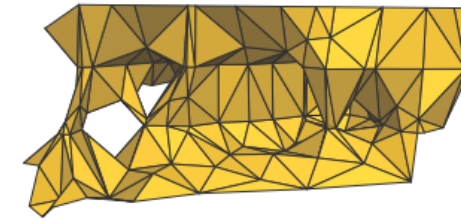
Isolated & Dangling Elements



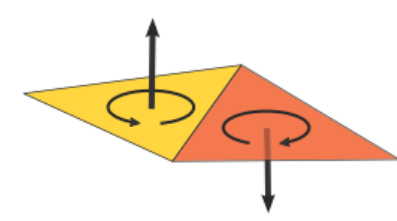
Singular Edge



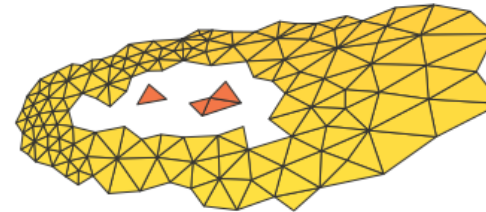
Singular Vertex



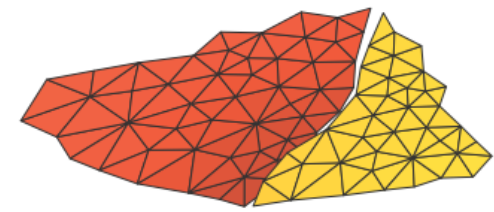
Topological Noise



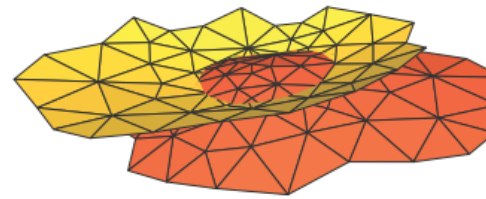
Inconsistent Orientation



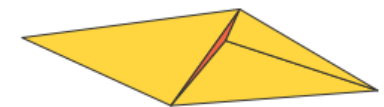
Hole (with Islands)



Gap (with partial Overlap)



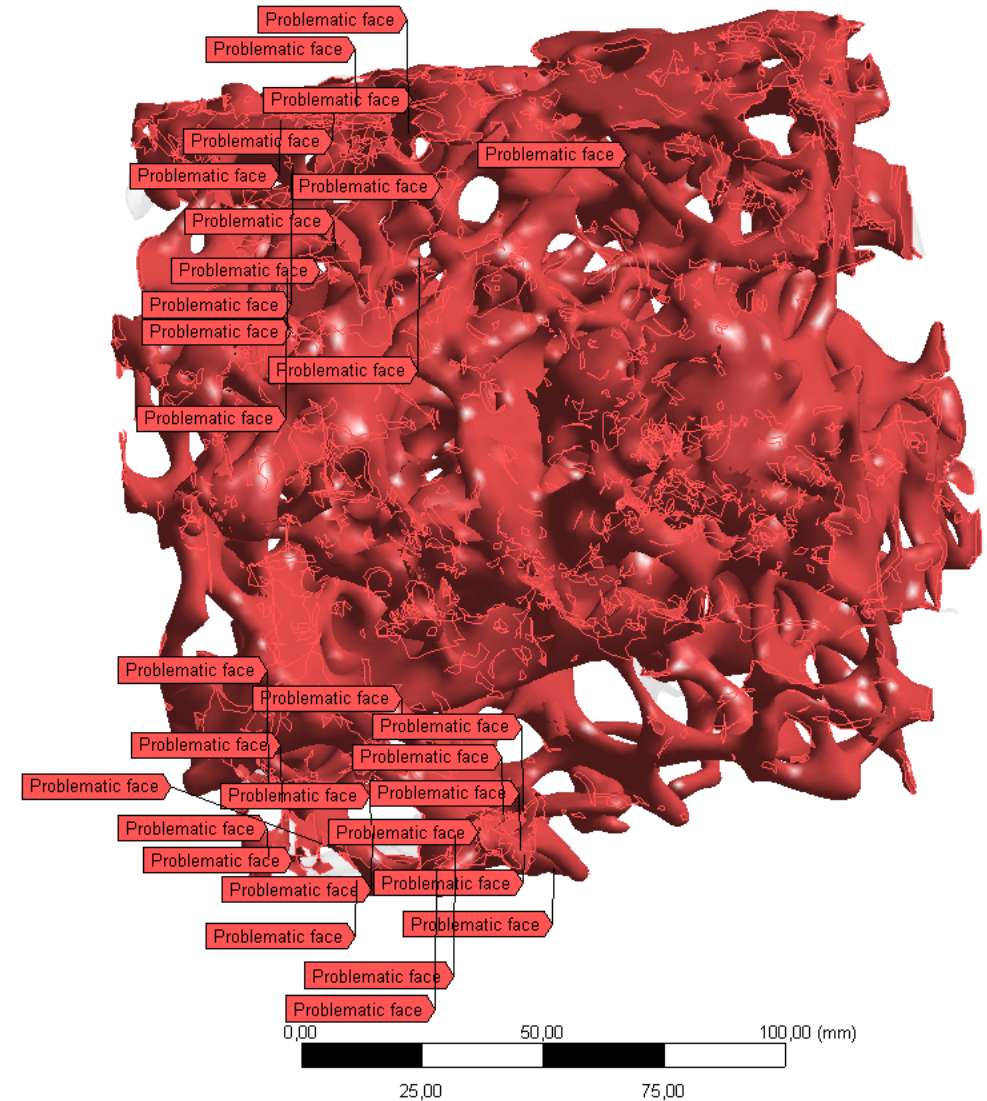
(Self-)Intersection



(Near) Degeneracy

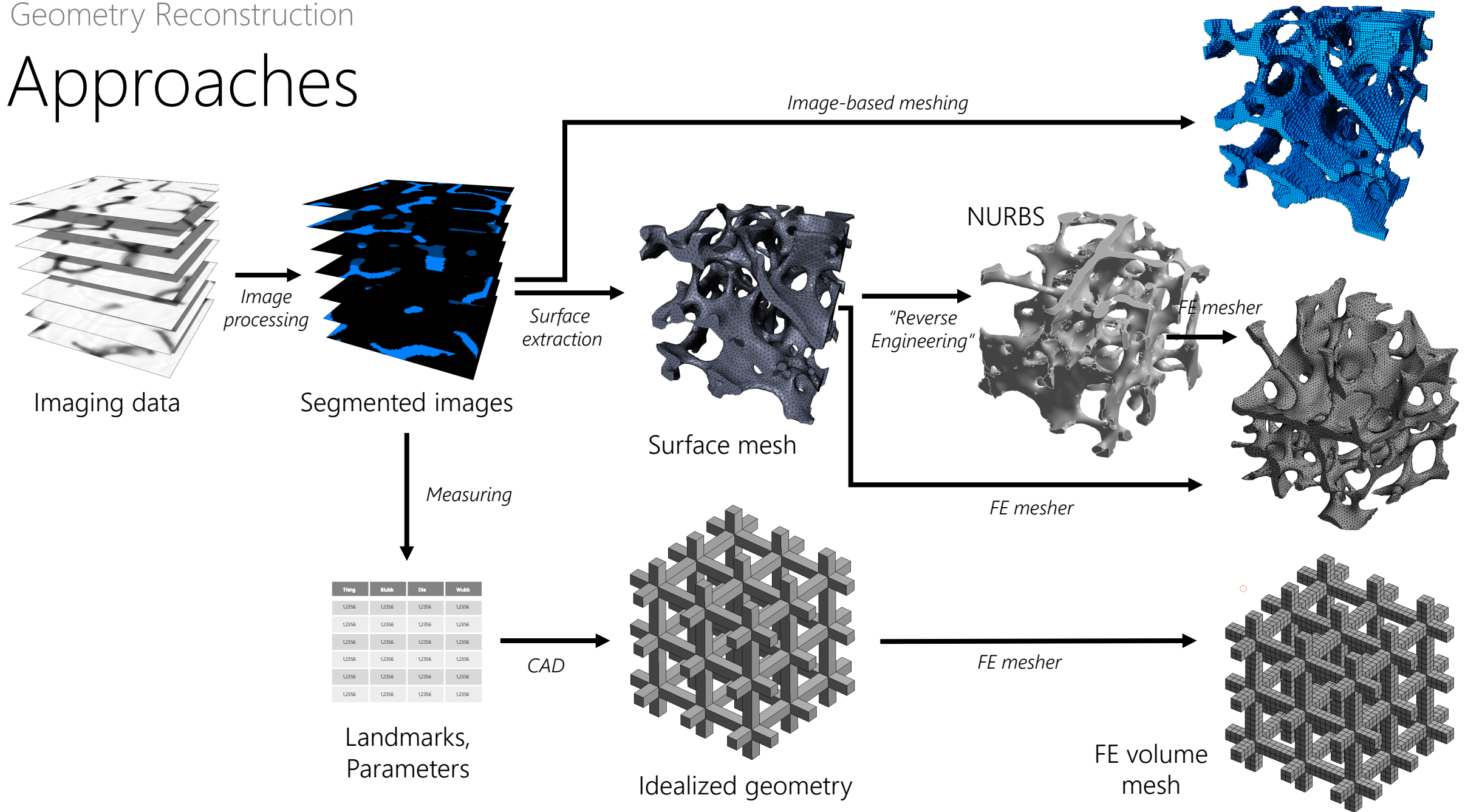
NURBS Reconstruction

- Further processing often requires “CAD” geometry (analytical, compact description)
- *Either:* Convert faceted surface mesh to NURBS representation, e.g.
 - Finite Element Modeler (ANSYS Workbench)
 - Rhino3D
 - SpaceClaim
 - SolidWorks (treat STL as complicated CAD geometry)
- *Or:* Use special software to work directly with STL, e.g.
 - Materialise Mimics + 3-matic
 - MeshLab, Meshmixer, SpaceClaim ...



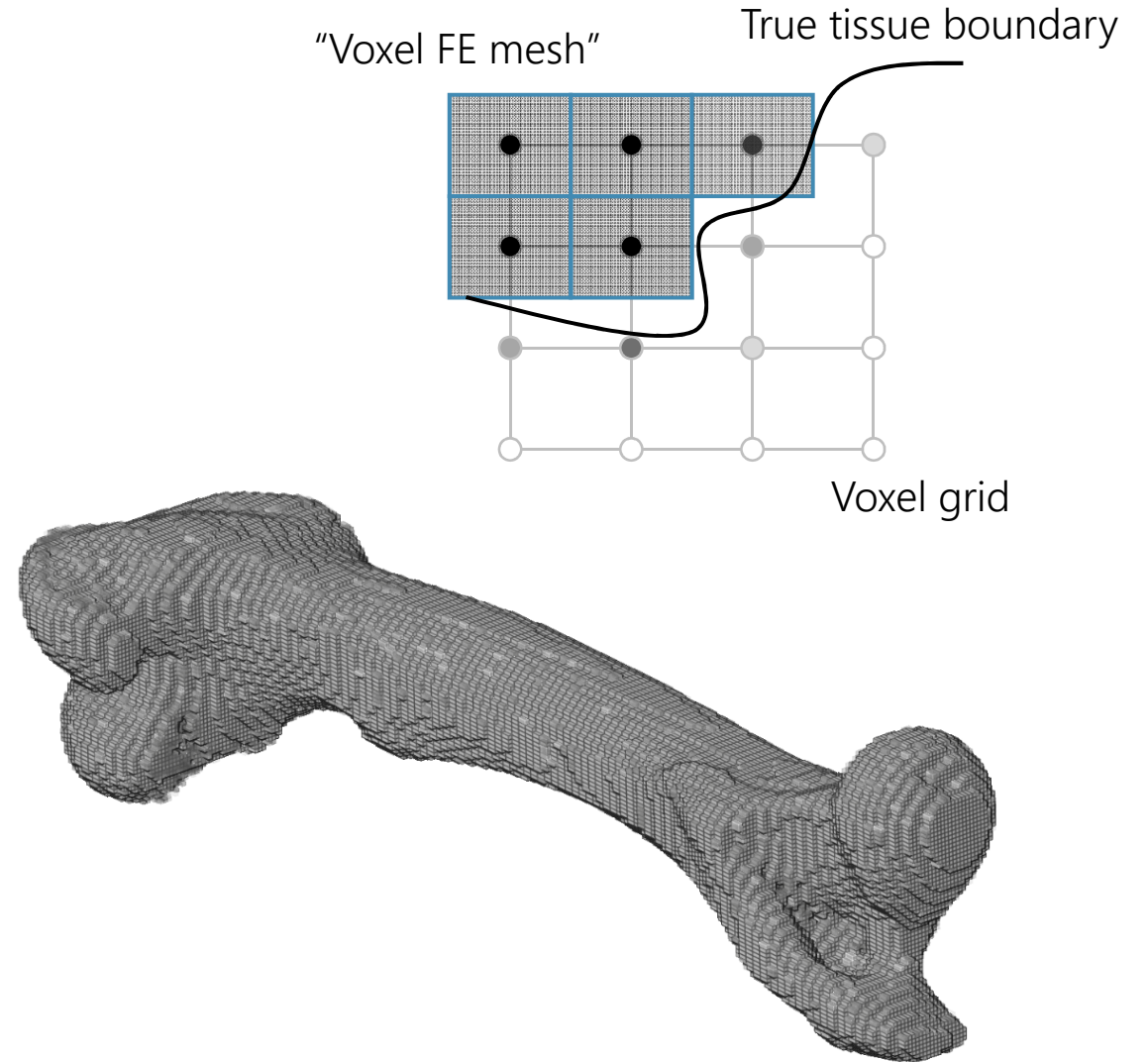
FE Modeler in action

Geometry Reconstruction Approaches



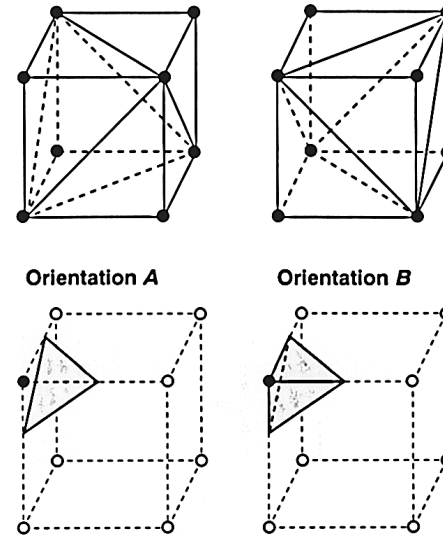
"Voxel Method"

- Trivial approach" (Keyak et al. 1990)
 - Directly convert voxel grid to FE mesh
 - 1 voxel \rightarrow 1 hexahedron (i.e. nearest neighbor interpolation, uniform spatial discretization)
- No prior (explicit) surface reconstruction
- Robust, arbitrarily complex topologies
- Automatically conforming interfaces of parts
- Optimal element quality
- Many DOFs
- Poor surface reconstruction, singularities
- Neither surface nor volume preserving

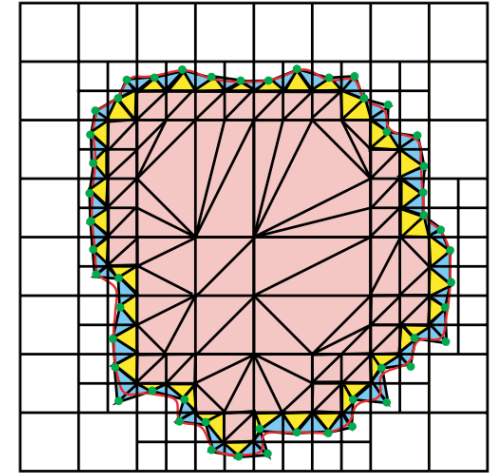


Advanced Algorithms

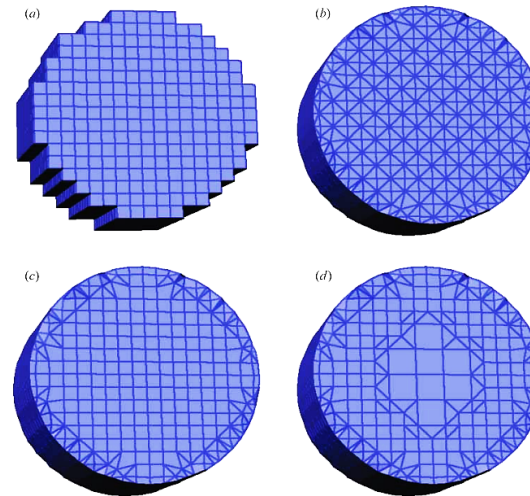
- Try to detect material boundaries and align generated element edges accordingly
- VoMaC (Müller & Rüegsegger 1995):
 - Create tetrahedrons instead of triangles
 - Inner "cubes" → five tetrahedra
- Zhang et al. 2005:
 - Adaptive meshing of inner cells
 - Hexahedral mesh generation
- EVoMaC (Young et al. 2008):
 - Multi-part meshing with conforming interfaces
 - Octree-based mesh decimation
- Software: SimpleWare + ScanFE



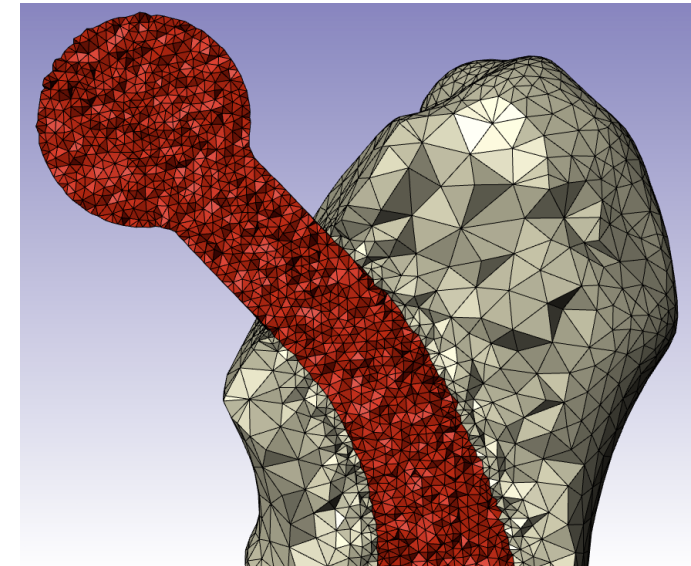
Müller & Rüegsegger, 1995



Zhang et al., 2005



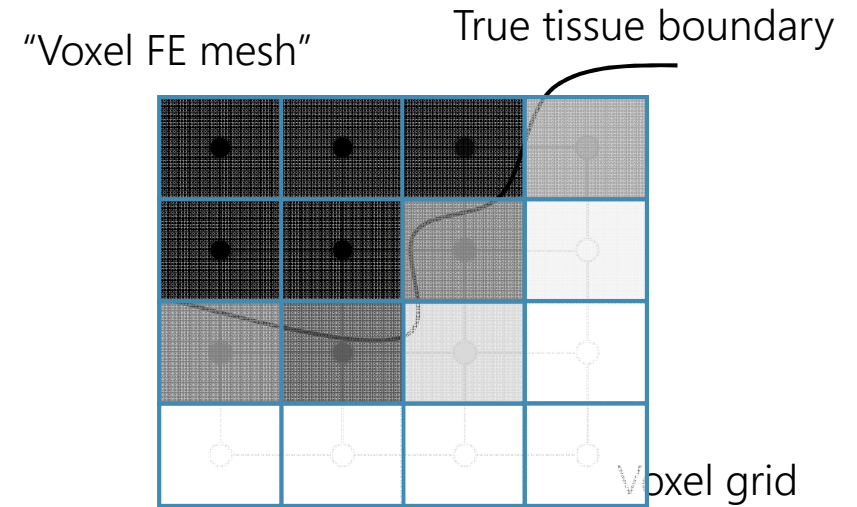
Young et al., 2008



© Simpleware Ltd.

Material Properties from Gray Values

- Image-based meshing: 1-to-1 relation between elements and voxels
- Idea: element-wise (apparent) material properties depending on image intensity
- $I \propto \mu \propto \rho = f^{-1}(E)$
- f : density-stiffness relation for some material
- f depends on material type and scale
- E.g. Carter and Hayes 1977 (homogenized trabecular bone)
 - $E = f(\rho) \propto \rho^3$



Geometry Reconstruction Approaches

