## Surface Modeling and Texture

3D Vision
University of Illinois

Derek Hoiem

### This class: Surface and Texture

Meshes and Implicit Surface Functions

- Surface reconstruction methods
  - Poisson and Screened Poisson Reconstruction
  - Floating Scale Surface Reconstruction
  - Delaunay Graph Cuts (Labatut'09)

- Texturing
  - Texrecon (Let there be color!)

### Surfaces vs. points

#### Surfaces

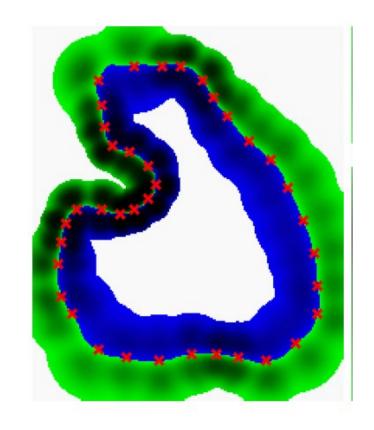
- Structured, encodes connectivity
- More precise measurement
- Render complete images
- May introduce artifacts if incorrectly estimated

#### **Points**

- Unstructured, unordered set
- Easy to stream, combine, subsample, manipulate
- More difficult to use for measurement/rendering

### Important concepts

- Point cloud
- Octrees
- Oriented points
- Visibility graph
- Mesh
- Implicit surface
  - Surface defined by F(x,y,z)=0
- Explicit surface



## Goals of surface generation

Approximate the point positions and normals

Smooth noise while preserving detail

Incorporate visibility constraints and discard outliers

Compact and regular mesh

### Poisson Reconstruction (Kazhdan et al. 2006)

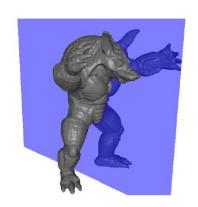
Input: oriented points (xyz, N)



 Approach: Solve for what volume is interior/exterior (indicator) and extract isosurface

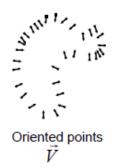


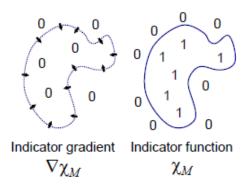
Output: watertight triangulated mesh

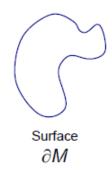


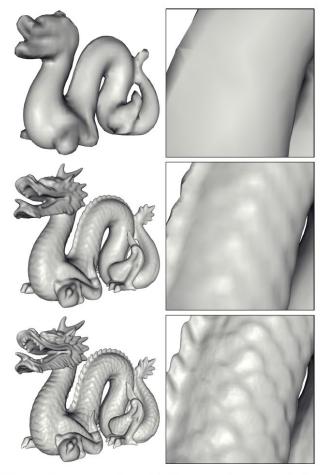
#### Poisson Reconstruction

- Create vector field V
  - Populate octree of depth D with oriented points
  - Compute vector field of each point as weighted sum of normals of points in voxel neighborhood
  - Choice of basis functions is a key design decision
  - Points in lower density areas get more weight and larger neighborhoods
- Poisson solution: solve for indicator  $\chi$  so that its gradient approximates V
  - Solve sparse linear system over octree nodes
- Extract isosurface using Marching Cubes variant
  - Fit local isosurfaces to 2x2x2 blocks and fuse









**Figure 3:** Reconstructions of the dragon model at octree depths 6 (top), 8 (middle), and 10 (bottom).

Tree Depth	Time	Peak Memory	# of Tris.
7	6	19	21,000
8	26	75	90,244
9	126	155	374,868
10	633	699	1,516,806



Figure 6: Reconstructions of the "Happy Buddha" model using VRIP (left) and Poisson reconstruction (right).

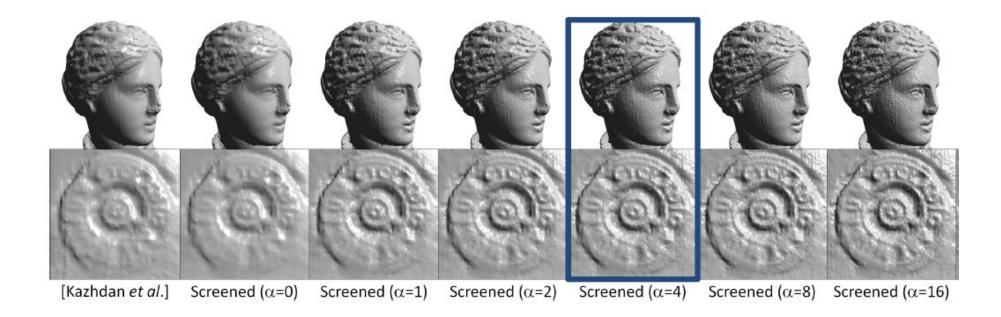
#### Poisson Reconstruction Pros and Cons

- Well suited to dense, accurate point clouds of objects, e.g. produced by laser scanners
  - Produces watertight surface
  - Completes gaps
  - Fits points while smoothing over noise, controlled by octree depth
- Does not account for visibility constraints or known holes
- Watertight surface can produce artifacts when entire scene is not captured (e.g. outdoors)
  - Requires "trimming" of parts of surface corresponding to shallower octree nodes
- Sensitive to outliers and can oversmooth the data

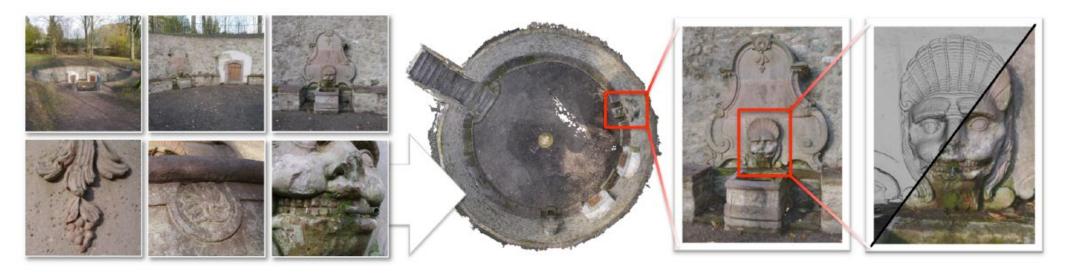


### Screened Poisson Reconstruction (Kazhdan Hoppe 2013)

- Adds term to optimization that penalizes non-zero isovalues at point locations
- Resulting surface fits points more closely
- Improved optimization (faster but uses a little more memory)



### Floating Scale Surface Reconstruction (Fuhrmann Goesele 2014)

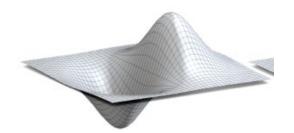


**Figure 1:** Floating Scale Surface Reconstruction example. 6 out of 384 input images of a multi-scale dataset (left). Registered images are processed with multi-view stereo which yields depth maps with drastically different sampling rates of the surface. Our algorithm is able to accurately reconstruct every caputured detail of the dataset using a novel multi-scale reconstruction approach (right).

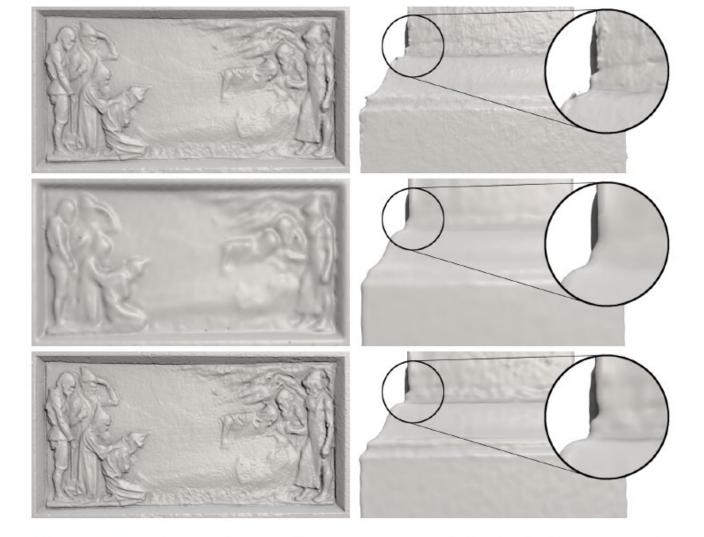
- Reconstructs surface only where supported by points
- Accounts for scale and confidence of points (scale is different than density here)
- Avoids need for global solution (but still slow)

### Floating Scale Surface Reconstruction: Algorithm

- Generate octree
- Sample implicit function at corners of leaf nodes
  - Weighted sum of signed basis functions
  - Basis functions depend on scale, confidence, position, normal of each point
  - Efficient due to octree and limited influence range of basis function
- Marching cubes to extract isosurface







Dataset Name	Number of Samples	Recon. Time	Peak Memory	Output Vertices
Bunny	362 K	30s + 9s	320 MB	277 K
Dragon	2.3 M	83s + 17s	603  MB	455 K
Armadillo	2.4 M	63s + 13s	553 MB	293 K
David	472 M	247m + 38m	114 GB	81.9 M
Temple	22.8 M	5m + 5s	1.96 GB	176 K
Elisabeth	39.3 M	19m + 1m	$4.39  \mathrm{GB}$	2.3 M
Fountain	196 M	178m + 6m	19.9 GB	10.2 M

**Figure 15:** Comparison with PSR on the Elisabeth dataset. Top row: Reconstruction with PSR at level 11, which reconstructs details (left) but produces noise in low resolution regions (right). Middle row: PSR at level 9 smoothly reconstructs low resolution regions but fails on the details. Bottom row: Our method reproduces both high- and low resolution regions appropriately.

Delaunay + Graph cuts (Labatut et al. 2009)
"Robust and efficient surface reconstruction from range data"

- Perform Delaunay triangulation on points
  - Results in good and bad faces

$$E(S) = E_{vis}(S) + \lambda_{qual} E_{qual}(S)$$

- Label tetrahedra as inside or outside based on visibility and quality
  - Roughly, each ray from visibility graph votes for which tetrahedra are inside and outside and which facets are on the surface
  - "Soft visibility" to allow for noise

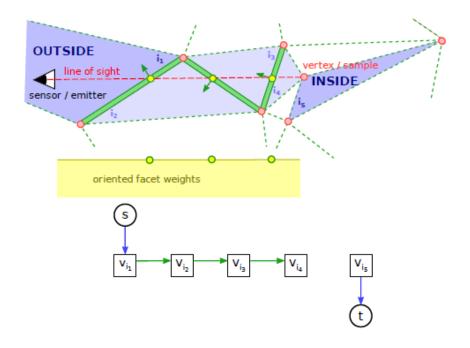








Figure 15: *Reconstruction details of rue Soufflot*. Acquired images (top), corresponding reconstruction results of Poisson (middle) and our method (bottom).

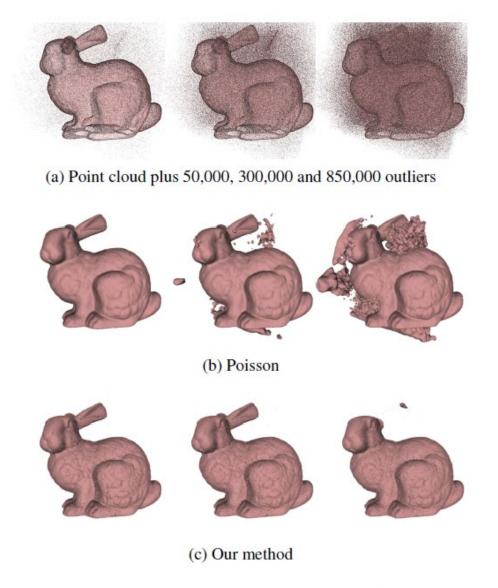
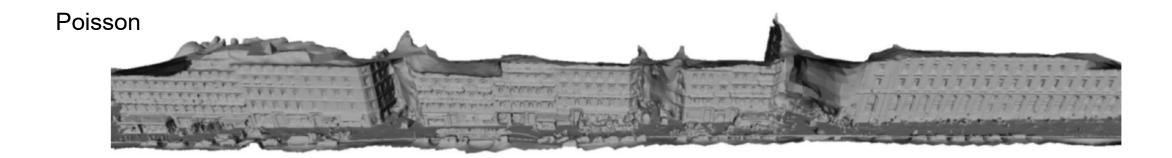
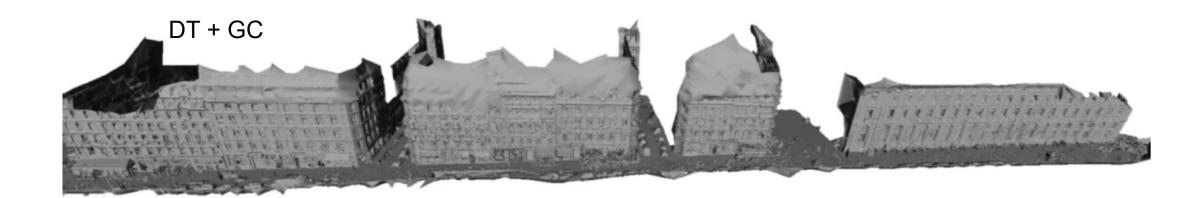
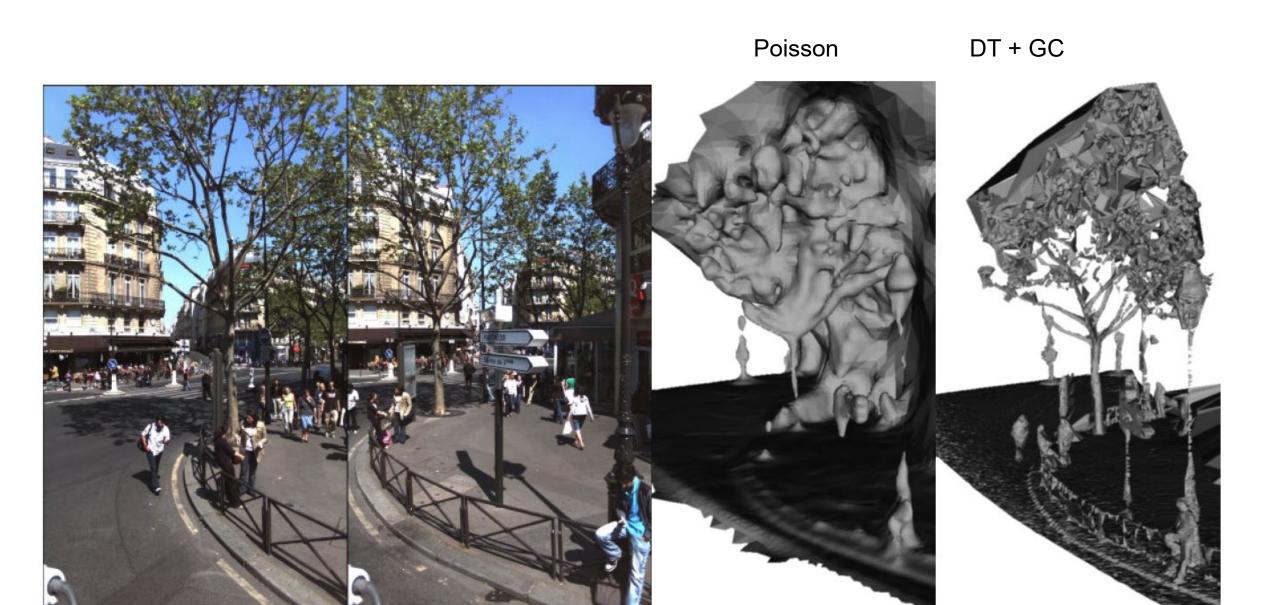


Figure 13: Robustness to large amounts of outliers.







### Delaunay + Graph Cuts pros and cons

- Preserves most original points
- Removes outliers that violate visibility constraints
- Completes surfaces where not contradicted by visibility graph
- Very detailed (but noisy) mesh
  - Can be smoothed using laplacian smoothing
- Not easy to parallelize, high memory usage
  - Similar running time to Poisson on single thread

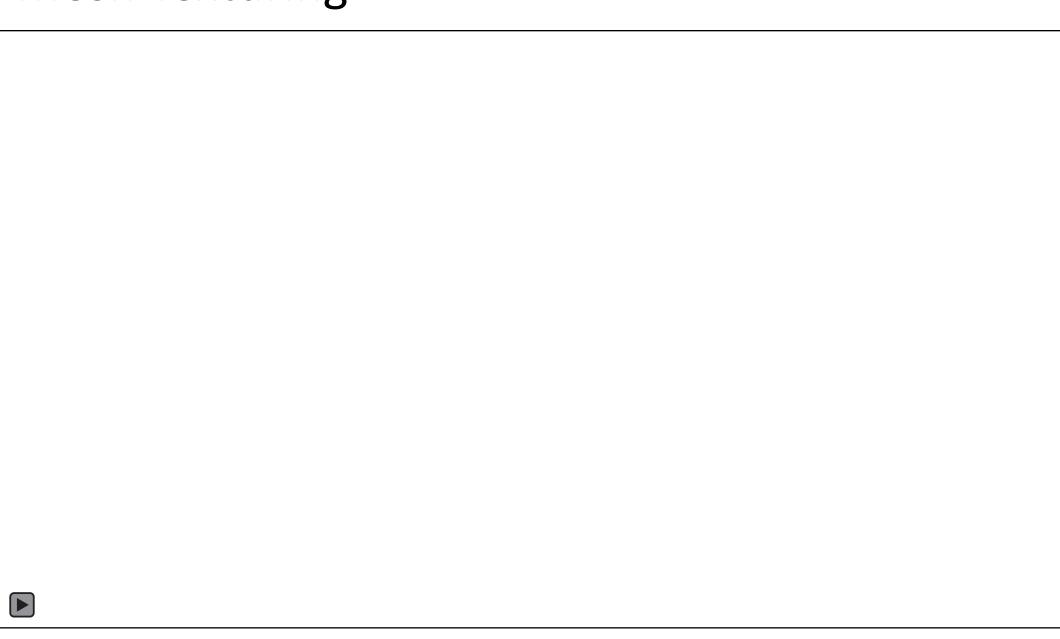
### Surface reconstruction, bottom line

 Poisson Reconstruction is well used and has continued to be extended; good code available: <a href="https://github.com/mkazhdan/PoissonRecon">https://github.com/mkazhdan/PoissonRecon</a>

Poisson has annoying bubbly artifacts that are avoided by FSSR and DT+GC

DT+GC effectively removes outliers (especially helpful for MVS inputs) but is more affected by noisy estimates

# Mesh Texturing



### How to texture meshes (texrecon)

- 1. Select which view to use to obtain texture for each face
  - Face must be visible to the selected view
  - Prefer close, frontal views, and camera axis passes near face
  - Prefer to use same image for neighboring faces (solve MRF)
  - Apply photo consistency check
- 2. Adjust colors and blend seams
  - Where neighboring faces are textured by different images, seams will occur
  - Adjust colors to minimize difference at boundary
  - Apply Poisson image editing on strip around boundary to force gradient closer to zero

### Open problems

- Not aware of surface reconstruction algorithms that handle large numbers of outliers and smooth away noise (without oversmoothing) well
- Surface reconstruction does not account for learned priors or appearance,
   e.g.
  - Shape priors, such as planes and cones
  - Repeated structures within and across scenes, e.g. window ledges and columns
  - Geometric smoothness based on image texture
- Texturing is still challenging when there are frequent occlusions (e.g. 360 capture) and relies on quality mesh

#### Research idea

 Goal: Given an MVS point cloud, produce a mesh that when resampled maximizes accuracy/completion

#### Evaluation:

- Accuracy/completeness on MVS benchmarks, compare raw point cloud to resampled from mesh
- Speed of mesh construction, compactness of mesh (# faces)
- Rough ideas for approach:
  - Encode points with position, normal, view rays, image features
  - Fill octree and predict occupancy, position, and connectivity for leaf nodes
  - Potentially use graph CNNs or PointNet variant

## Summary

- Surface reconstruction is extensively studied for reconstructing laser scanned objects, often fitting isosurface
  - A lot of room for improvement for turning MVS point clouds + images into nice surfaces

 Texture mapping involves finding close, direct, unobstructed views and blending