# Lecture 6: Introduction to Geometry

#### Computer Graphics: Rendering, Geometry, and Image Manipulation Stanford CS248A, Winter 2025

# Increasing the complexity of our modelsTransformationsGeometryMaterials, lighting, ...









#### What is geometry?

"Earth" "measure"

ge•om•et•ry/jē'ämətrē/ n. 1. The study of shapes, sizes, patterns, and positions. angles, etc.) can be measured.





Plato: "...the earth is in appearance like one of those balls which have leather coverings in twelve pieces..."

# 2. The study of spaces where some quantity (lengths,







Photo of original Utah teapot (now sitting in Computer History Museum in Mountain View)

Martin Newell's early teapot renderings (Martin created teapot model in 1975 using Bezier curves)





Cornell Box: Originally created in 1984 (This image was rendered in 1985 by Cohen and Greenberg)





**The Stanford Bunny** (Mesh created by reconstruction from laser scans)



#### Photograph of scanned statue (Statue purchased by Greg Turk at a store on University Ave in 1994)





#### Laser scan of Michelangelo's David (Stanford's Digital Michelangelo project, 1999)







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Photo credit:Natasha Moustache/TAS23









1.01







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#### What's the best way to encode geometry on a computer?



# Many ways to digitally encode geometry

#### **EXPLICIT**

- point cloud
- polygon mesh
- subdivision, NURBS
- •••

#### IMPLICIT

- level set
- algebraic surface
- L-systems
- $\bullet \bullet \bullet$

Each choice best suited to a different task/type of geometry 





### "Implicit" representations of geometry

- Points aren't known directly, but satisfy some relationship
- E.g., unit sphere is all points such that  $x^2+y^2+z^2=1$
- More generally, f(x,y,z) = 0



#### of geometry some relationship +v<sup>2</sup>+z<sup>2</sup>=1



# Many implicit representations in graphics

- algebraic surfaces
- constructive solid geometry
- level set methods
- blobby surfaces
- fractals



#### (Will see some of these a bit later.)





### But first, let's play a game:

# I'm thinking of an implicit surface f(x,y,z)=0

Find any point on it.



### Give up?

#### My function was f(x,y,z) = x - 1.5 (a plane):



#### Implicit surfaces make some tasks hard (like sampling).





#### Let's play another game.

#### I have a new surface $f(x,y,z) = x^2 + y^2 + z^2 - 1$

#### I want to see if a point is *inside* it.



# **Check if this point is inside the unit sphere** How about the point (3/4, 1/2, 1/4)?

- 9/16 + 4/16 + 1/16 = 7/8
- 7/8 < 1
- YES.

#### Implicit surfaces make other tasks easy (like inside/outside tests).





### **Recall: implicit form of a line**

#### Easy to test if a point is on the "positive" or negative side of the line





# "Explicit" representations of geometry

- All points are given directly
- **E.g.**, points on sphere are
- More generally:



(Might have a bunch of these maps, e.g., one per triangle!)

 $(\cos(u)\sin(v),\sin(u)\sin(v),\cos(v)),$ for  $0 \le u \le 2\pi$  and  $0 \le v \le \pi$ 

 $f: \mathbb{R}^2 \to \mathbb{R}^3; (u, v) \mapsto (x, y, z)$ 



# "Explicit" representations of geometry

- More generally:
- Example: a triangle



 $f: \mathbb{R}^2 \to \mathbb{R}^3; (u, v) \mapsto (x, y, z)$ 

#### $\mathbf{x} = f(\beta, \gamma) = \mathbf{a} + \beta(\mathbf{b} - \mathbf{a}) + \gamma(\mathbf{c} - \mathbf{a})$



# Many explicit representations in graphics

- triangle meshes
- polygon meshes
- subdivision surfaces
- NURBS
- point clouds





#### (Will see some of these a bit later.)



#### But first, let's play another game:

# I'll give you an explicit surface.

### You give me some points on it.



#### Sampling an explicit surface My surface is f(u, v) = (1.5, u, v). Just plug in any values (u,v)!









#### Let's play another game.

#### I have a new surface f(u,v).

#### I want to see if a point is *inside* it.



#### **Check if this point is inside the torus** My surface is $f(u,v) = (2+\cos(u))\cos(v), 2+\cos(u))\sin(v), \sin(u)$ How about the point $(1,\sqrt{3},5/4)$ ?

...NO!



#### Explicit surfaces make other tasks hard (like inside/outside tests).



# **CONCLUSION:** Some representations work better than others depending on the task!



# **Different representations will be better suited to** different types of geometry.

# Let's take a look at some common representations used in computer graphics.



### **Algebraic surfaces (implicit)**

- Surface is zero set of a polynomial in x, y, z ("algebraic variety")
- **Examples:**

What about more complicated shapes?



**Very hard to come up with polynomials for complex shapes!** 











# **Blobby surfaces (implicit)**

Instead of booleans, gradually blend surfaces together: 



**Easier to understand in 2D:** 

 $\phi_{p}(x) := e^{-|x-p|^{2}} \quad \text{(Gaussian centered at p)}$ 

$$f := \phi_p + \phi_q$$



# Level set methods (implicit)

- Implicit surfaces have some nice features (e.g., merging/splitting)
- But, hard to describe complex shapes in closed form



Surface is determined by where the *interpolated* value equals zero **Provides much more explicit control over shape** Often demands sophisticated *filtering* (trilinear, tricubic...)

Alternative: store a grid of values approximating a continuous function (samples of the function)



# Level sets encode, e.g., constant tissue density




#### Level sets in physical simulation Level set encodes distance to air-liquid boundary





#### See <a href="http://physbam.stanford.edu">http://physbam.stanford.edu</a>



### Sparse level set storage

- **Drawback: storage for 2D surface is now O(n<sup>3</sup>)**
- Can reduce cost by storing only a narrow band of distances around surface:



In this figure: red = clearly within water blue = clearly outside water

#### green = regions where we store level set values to encode surface



#### Neural representations for compressing distance functions

#### Simple solution:

- Train a DNN to evaluate distance\_to\_surface(x,y,z)
- Use conventional dense or sparse grid representation to create training data pairs
- Good: massive compression (surface represented by weights of DNN)
- Bad: very high evaluation cost (must evaluate large DNN to determining distance from surface

#### More recent "hybrid" approaches:

- Use neural code to represent local surface structure
- Store neural "code" at cells of traditional uniform grid, or sparse grid
  - e.g., code[x,y,z]
- Train a "tiny" DNN to produce distance(x,y,z) = DNN(x,y,z, code[x,y,z])
- Idea: DNN only has to translate code into a surface location = much cheaper to evaluate



y weights of DNN) e DNN to

e rid, or sparse grid

(x,y,z, code[x,y,z]) location = much



903.63 KB

### Fractals (implicit)

- No precise definition; exhibit self-similarity, detail at all scales
- New "language" for describing natural phenomena
- Hard to control shape!



#### rity, detail at all scales henomena



#### Mandelbrot set - zooming in



#### (Colored according to how quickly each point diverges/converges.)



### Mandelbrot set - definition

- For each point c in the plane:
  - double the angle
  - square the magnitude
  - add the original point c
  - repeat

#### If the point remains bounded (never goes to $\infty$ ), it's in the set.





#### Mandelbrot set - examples

#### <u>starting point</u>

(0, 1/2) (converges) (0, 1) (periodic) (1/3, 1/2) (diverges)





### Implicit representations - pros and cons

- Pros:
  - Description can be very compact (e.g., a polynomial)
  - Easy to determine if a point is in our shape (just plug it in!)
  - Other queries may also be easy (e.g., distance to surface)
  - For simple shapes, exact description/no sampling error
  - Easy to handle changes in topology (e.g., fluid)
- **Cons:** 
  - Expensive to find all points in the shape (e.g., for drawing)
  - Very difficult to model complex shapes



#### What about explicit representations?



### **Point cloud (explicit)**

- **Easiest representation: list of points (x,y,z)**
- Often augmented with *normals*
- Easily represent any kind of geometry
- **Useful for LARGE datasets (>>1 point/pixel)**
- Hard to interpolate undersampled regions
- Hard to do processing / simulation / ...





### Point cloud via laser scanning



Image Credit: 3Dling

#### Another example: Microsoft XBox 360 Kinect





Image credit: iFixIt





#### **Structured light**

System: one light source emitting known beam + one camera measuring scene appearance If the scene is at reference plane, image that will be recorded by camera is known (correspondence between pixel in recorded image and scene point is known)



Single spot illuminant is inefficient! (must "scan" scene with spot to get depth, so high latency to retrieve a single depth image)



### Infrared image of Kinect illuminant output



Credit: www.futurepicture.org



### Infrared image of Kinect illuminant output



Credit: www.futurepicture.org



### **3D volumes**

#### Consider storage requirements: 1024<sup>3</sup> cells (let's assume boolean per cell for occupancy)



Typical challenge: limited resolution



Credit: Voxel Ville NFT (<u>voxelville.io</u>)



### Polygon mesh (explicit)

- Store vertices *and* polygons (most often triangles or quads)
- Easier to do processing/simulation, adaptive sampling
- More complicated data structures
- Perhaps most common representation in graphics



(Much more about polygon meshes in upcoming lectures!)





### **Triangle mesh (explicit)**

- Store vertices as triples of coordinates (x,y,z)
- Store triangles as triples of indices (i,j,k)
- E.g., tetrahedron:

#### VERTICES

0: 2: **3:** -1

#### Use linear interpolation to define points inside triangles:





f(u,v) = a + u(b-a) + v(c-a)



### Linear interpolation of samples (in 1D)

# $f(x_0) = f_0$

 $x_0$ 





#### Can think of linear interpolation as linear combination of two functions

Weights are given by the two values ( $f_0$  and  $f_1$ ) being interpolated





#### Note: this is the idea of representing a function in a new basis, again!

#### My function *f* is represented as a superposition (weighted sum) of a set of basis functions





$$0.1 \times$$







## Problem with piecewise linear interpolation: derivates not continuous

 $f(x_0) = f_0$ 

 $x_0$ 





#### Smooth interpolation?







### Bernstein basis

- Why limit ourselves to just linear interpolation?
- More flexibility by using higher-order polynomials
- Instead of usual basis (1, x, x<sup>2</sup>, x<sup>3</sup>, ...), use Bernstein basis:



#### olation? olynomials a Bernstein basis:



## **Bézier curves (explicit)**

A Bézier curve is a curve expressed in the Bernstein basis:

- For n=1, just get a line segment!
- For n=3, get "cubic Bézier":
- **Important features:** 
  - 1. interpolates endpoints
  - 2. tangent to end segments
  - 3. contained in convex hull (*nice for rasterization*)





### Piecewise Bézier curves (explicit)

- More interesting shapes: piece together many Bézier curves
- Widely-used technique (Illustrator, fonts, SVG, etc.)



#### Formally, piecewise Bézier curve:

**piecewise Bézier**   $\gamma(u) := \gamma_i \left( \frac{u - u_i}{u_{i+1} - u_i} \right),$  $\int u_{i+1} - u_i /$ single Bézier





 $u_i \le u < u_{i+1}$ 



#### **Vector fonts**

### The Quick Brown Fox Jumps Over The Lazy Dog

ABCDEFGHIJKLMNOPQRSTUVWXYZ abcdefghijklmnopqrstuvwxyz 0123456789



#### **Baskerville font - represented as cubic Bézier splines**



#### **Bézier curves** — tangent continuity To get "seamless" curves, want *points* and *tangents* to line up:



- **Ok**, but how?
- Each curve is cubic:  $au^3 + bu^2 + cu + d$
- Q: How many constraints vs. degrees of freedom?

## Q: Could you do this with quadratic Bézier? Linear Bézier?



#### **Tensor product**

Can use a pair of curves to get a surface the *"tensor product"*):



#### Value at any point (u,v) given by product of a curve f(u) and a curve g(v) (sometimes called







### **Bézier patches**

*Bézier patch* is sum of (tensor) products of Bernstein bases







### Bézier surface

#### Just as we connected Bézier curves, can connect Bézier patches to get a surface:

#### Very easy to draw: just dice each patch into regular (u,v) grid!





### Many different spline patch schemes

- NURBS, Gregory, Pm, polar...
- Tradeoffs:
  - degrees of freedom
  - continuity
  - difficulty of editing
  - cost of evaluation
  - generality
  - •••

As usual: pick the right tool for the job!





### Subdivision (explicit or implicit?)

- Alternative starting point for curves/surfaces: subdivision
- **Start with control curve**
- Insert new vertex at each edge midpoint
- Update vertex positions according to fixed rule
- For careful choice of averaging rule, yields smooth curve
  - Some subdivision schemes correspond to well-known spline schemes!



Slide cribbed from Don Fussell.



### Subdivision surfaces (explicit) Start with coarse polygon mesh ("control cage")

- Subdivide each element
- Update vertices via local averaging
- Many possible rule:
  - Catmull-Clark (quads)
  - Loop (triangles)
- **Common issues:** 
  - interpolating or approximating?
  - continuity at vertices?
- Easier than splines for modeling; harder to evaluate pointwise







#### Subdivision in action (Pixar's "Geri's Game")





#### **Measurements of surfaces**


### Surface tangent







### Surface normal (N) is orthogonal to all tangents





# A common visualization of normals

### Encode normal direction as RGB color as difference from gray

R = 0.5 + 0.5 N.xG = 0.5 + 0.5 N.yB = 0.5 + 0.5 N.z

Notice: scale and bias normal values so we can represent negative components of normal as valid colors



### Image credit: https://www.3dgep.com/forward-plus/



### Curvature is change in normal





### **Radius of curvature**





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