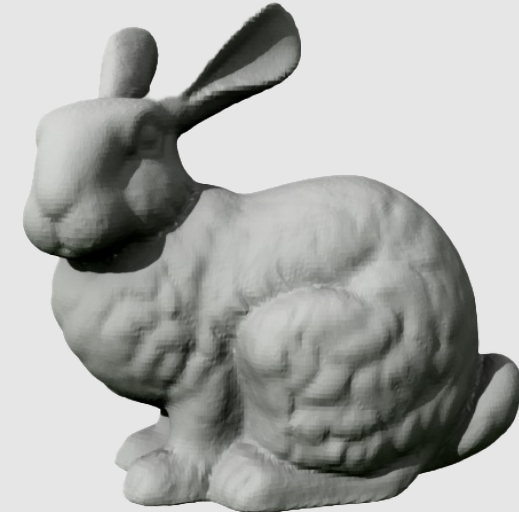
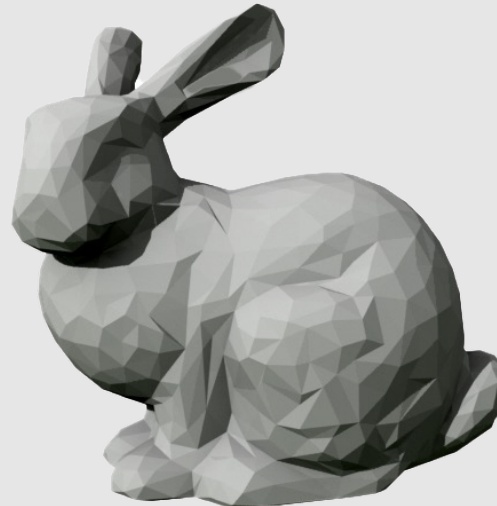
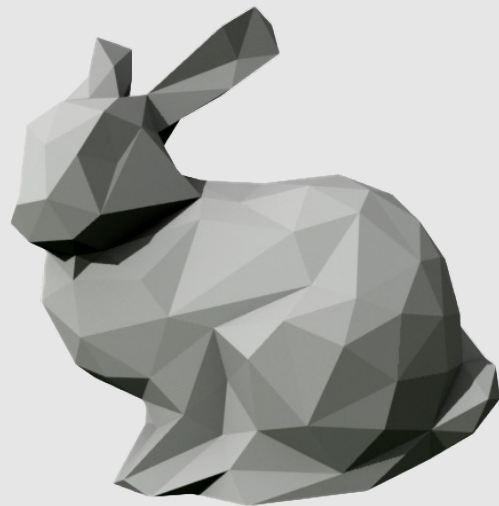
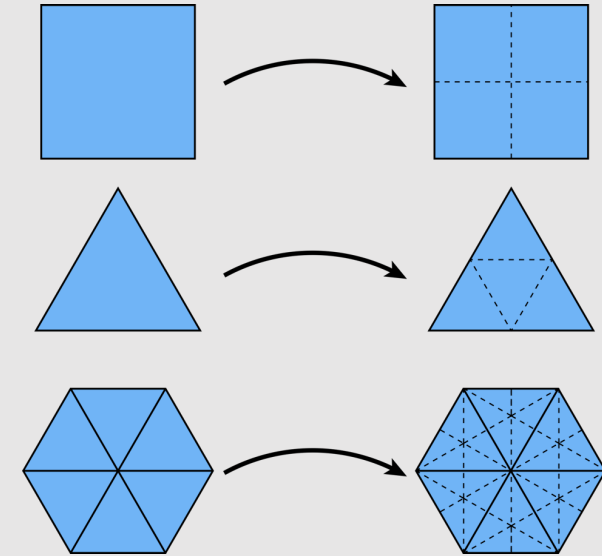


# More Digital Geometric Processing

- Digital Geometric Processing
  - Geometric Subdivision
  - Geometric Simplification
  - Geometric Remeshing
  - Geometric Queries

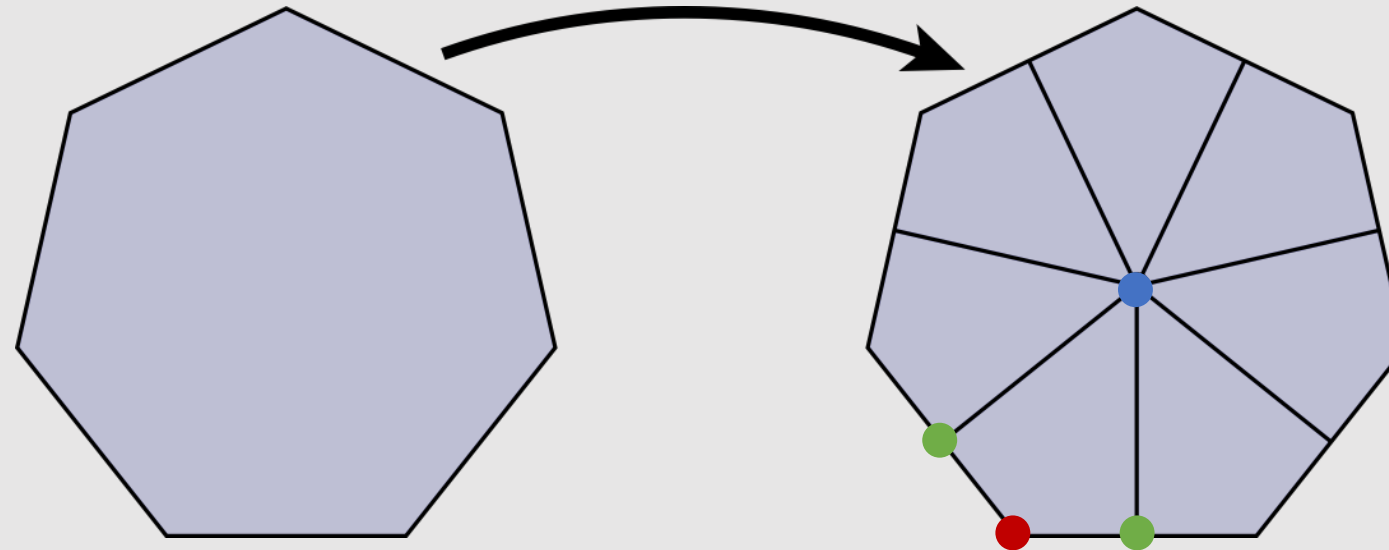
# Subdivision

- Subdivision is the process of **upsampling** a mesh
- General formula:
  - **Split Step**: split faces into smaller faces
  - **Move Step**: replace vertex positions/properties with weighted average of neighbors



# Linear Subdivision [Split Step]

- Split every polygon (any # of sides) into quadrilaterals

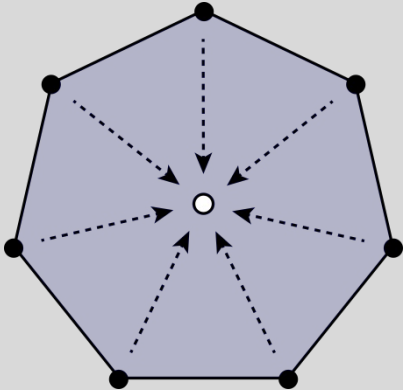


- Each new quadrilateral now has:
  - **[face coords]** : 1 new vertex from the mesh face center
  - **[edge coords]** : 2 new vertices from the new edges
  - **[vertex coords]** : 1 new vertex from the original mesh face



# Linear Subdivision [Move Step]

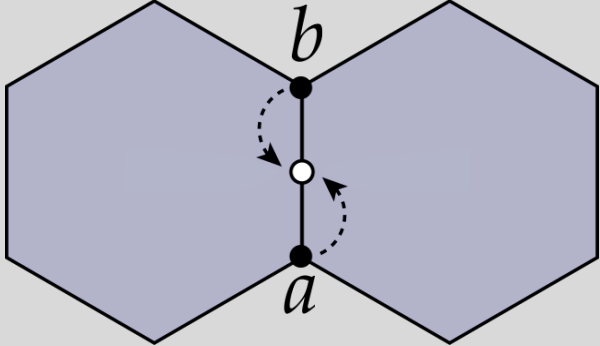
Step 1: Face Coords



$\frac{1}{n} \sum_i p_i$

The diagram shows a light blue shaded polygon with a central white point. Dashed lines with arrows point from the central point to each of the polygon's vertices, representing the calculation of the face's centroid.

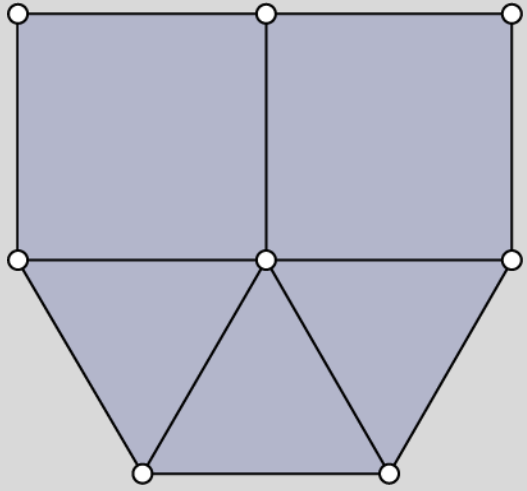
Step 2: Edge Coords



$(a + b) / 2$

The diagram shows two light blue shaded polygons sharing a vertical edge. On this edge, there are two points labeled 'a' (bottom) and 'b' (top). A white dot is positioned on the edge between 'a' and 'b', with dashed curved arrows indicating its movement towards the midpoint of the segment 'ab'.

Step 3: Vertex Coords

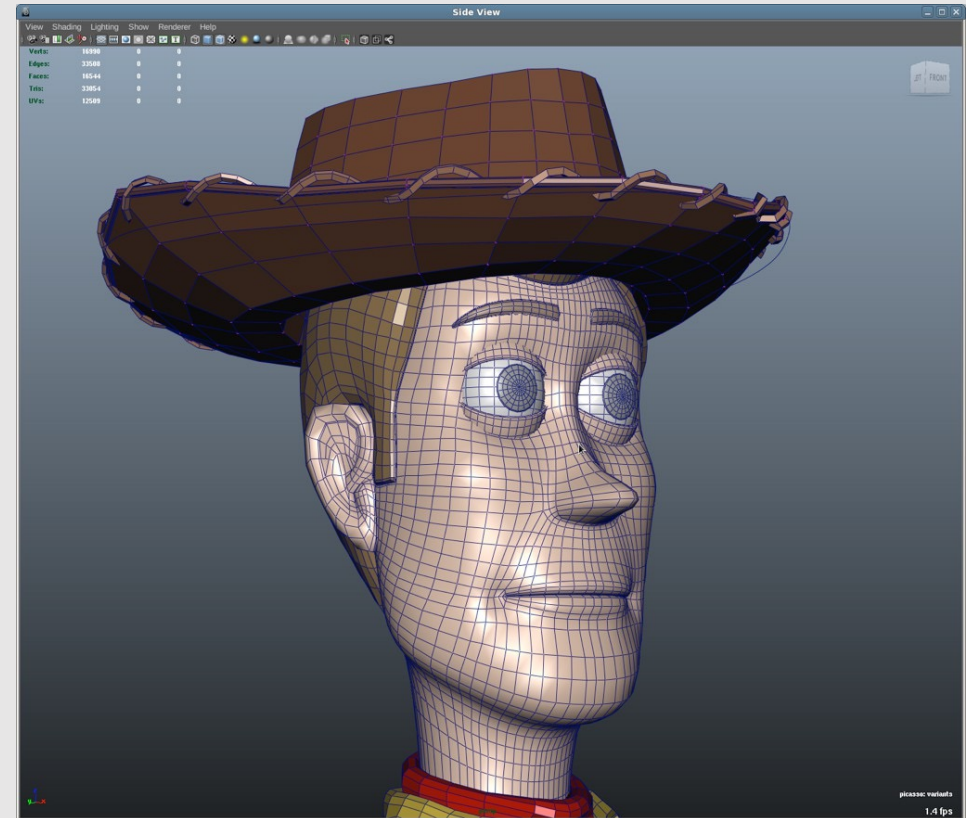


$v_i = v_i$

The diagram shows a light blue shaded polygon that has been subdivided into several smaller polygons. The vertices are marked with white circles. The label  $v_i$  is placed next to one of the vertices, and the equation  $v_i = v_i$  is shown to the right, indicating that the vertex coordinates remain unchanged during this step.

# Catmull Clark Subdivision

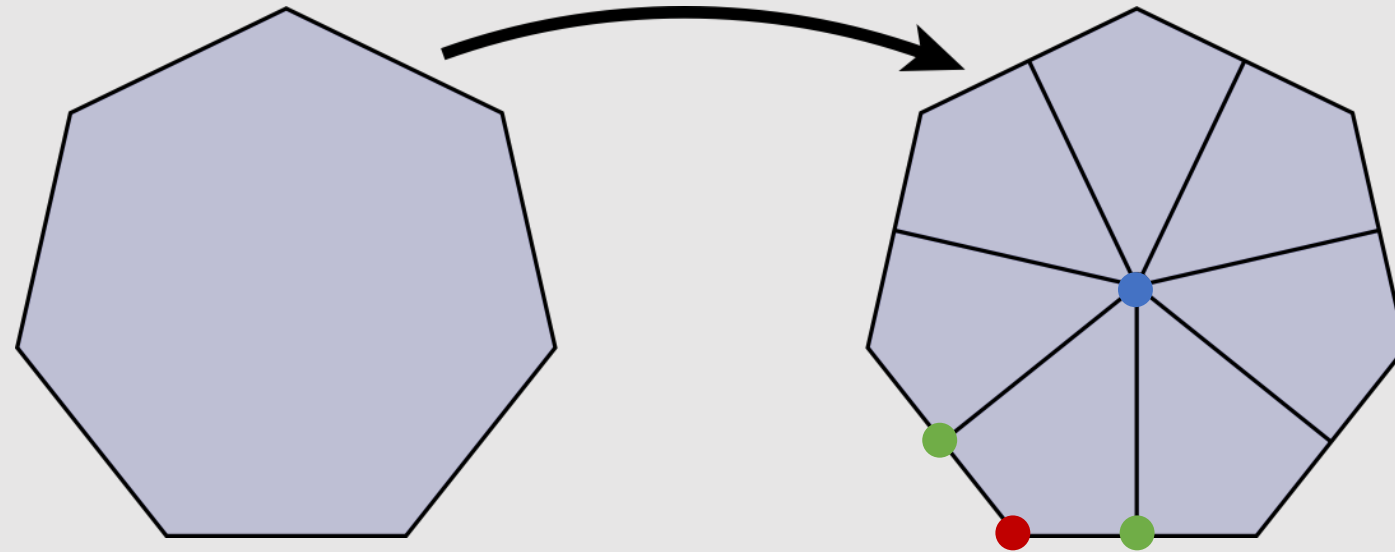
- In 1978, Edwin Catmull (Pixar co-founder) and Jim Clark wanted to create a generalization of **uniform bi-cubic b-splines** for 3D meshes
  - We will cover what this means in a future lecture : )
- Became ubiquitous in graphics
  - Helped Catmull win an Academy Award for Technical Achievement in 2005



OpenSubdiv V2 (2018) Pixar

# Catmull-Clark Subdivision [Split Step]

- Split every polygon (any # of sides) into quadrilaterals

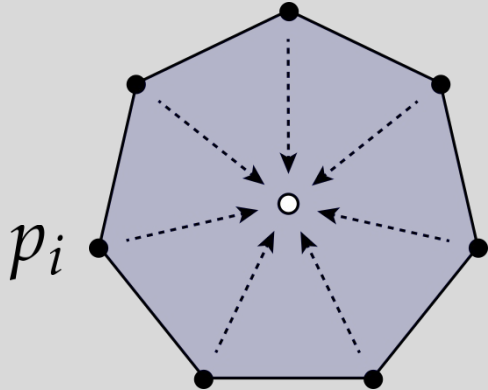


- Each new quadrilateral now has:
  - **[face coords]** : 1 new vertex from the mesh face center
  - **[edge coords]** : 2 new vertices from the new edges
  - **[vertex coords]** : 1 new vertex from the original mesh face

**No different than  
Linear Subdivision!**

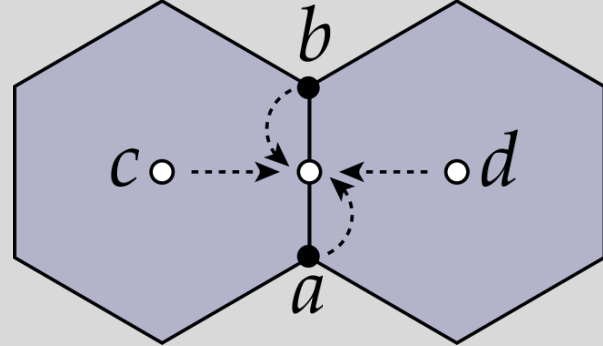
# Catmull-Clark Subdivision [Move Step]

Step 1: Face Coords



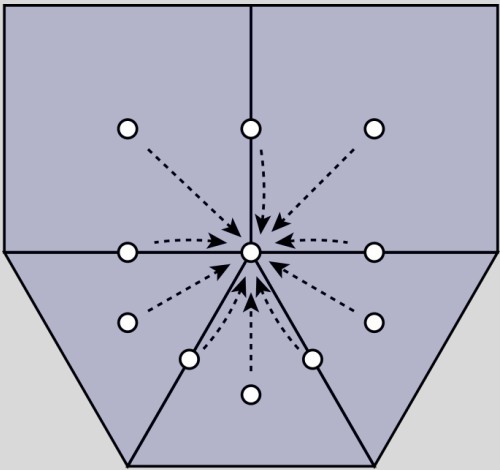
$\frac{1}{n} \sum_i p_i$

Step 2: Edge Coords



$(a+b+c+d)/4$

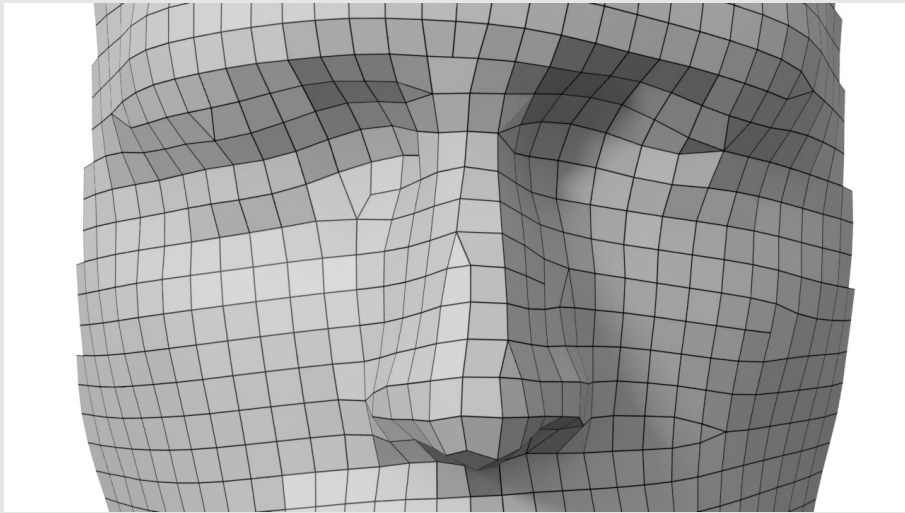
Step 3: Vertex Coords



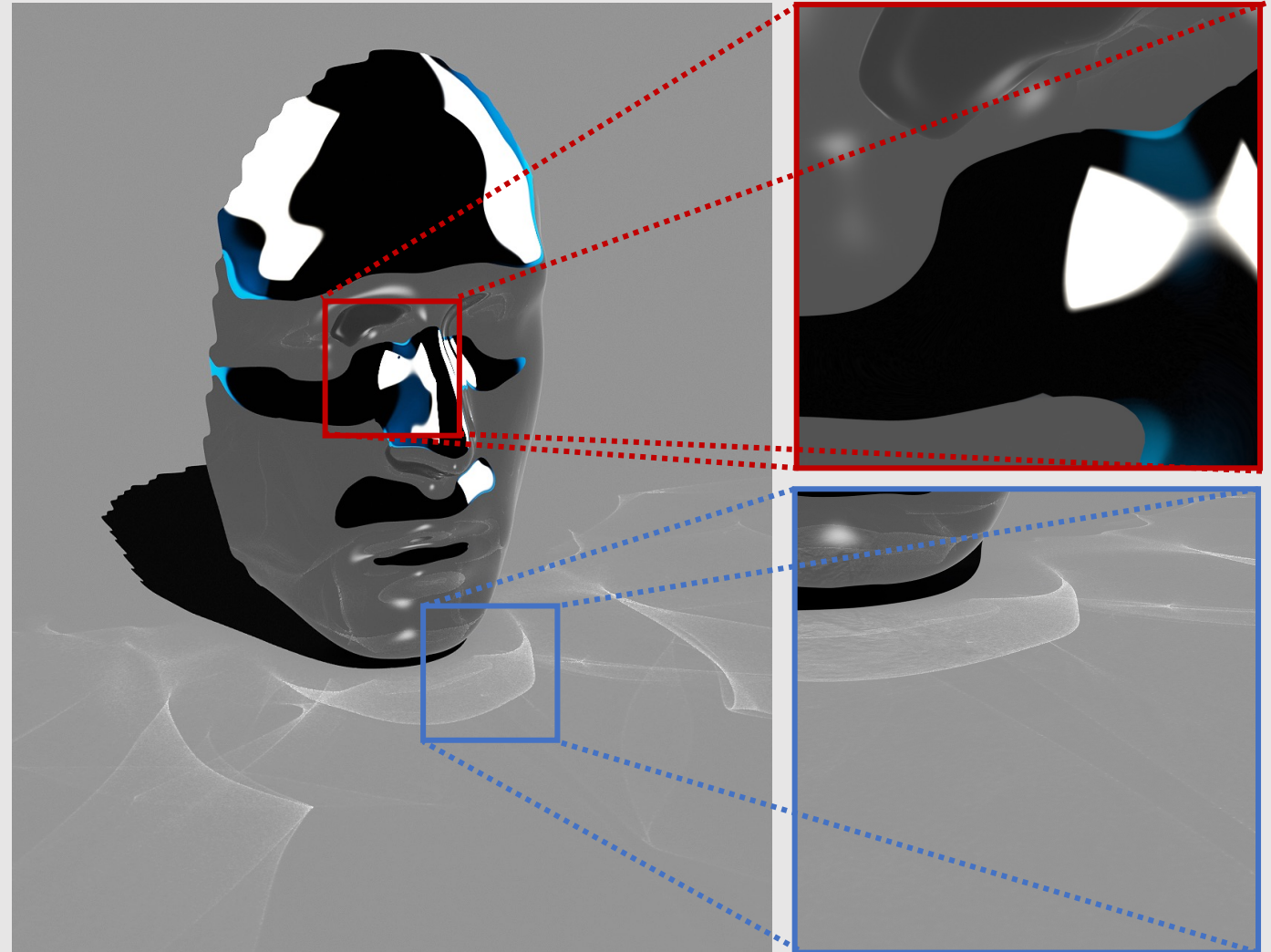
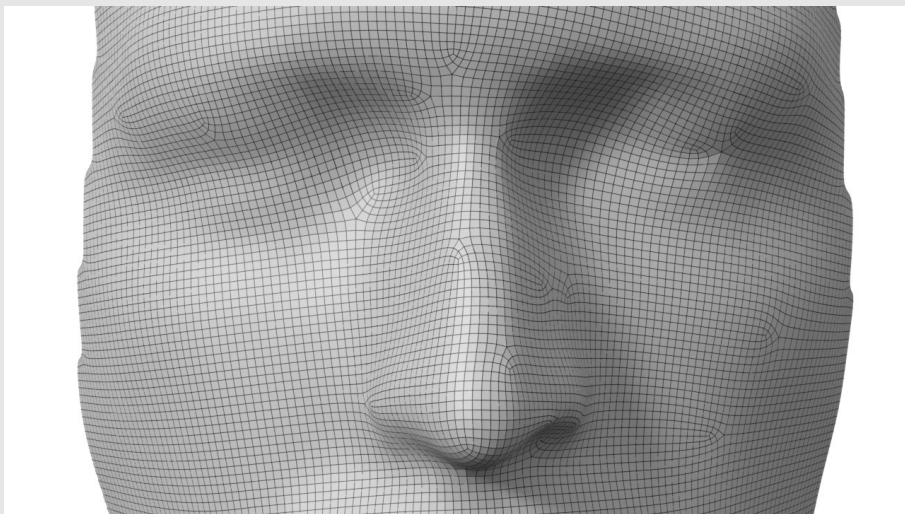
$$\frac{Q + 2R + (n - 3)S}{n}$$

- $n$  - vertex degree
- $Q$  - average of face coords around vertex
- $R$  - average of edge coords around vertex
- $S$  - original vertex position

# Catmull-Clark Subdivision [Quads]



Few irregular vertices

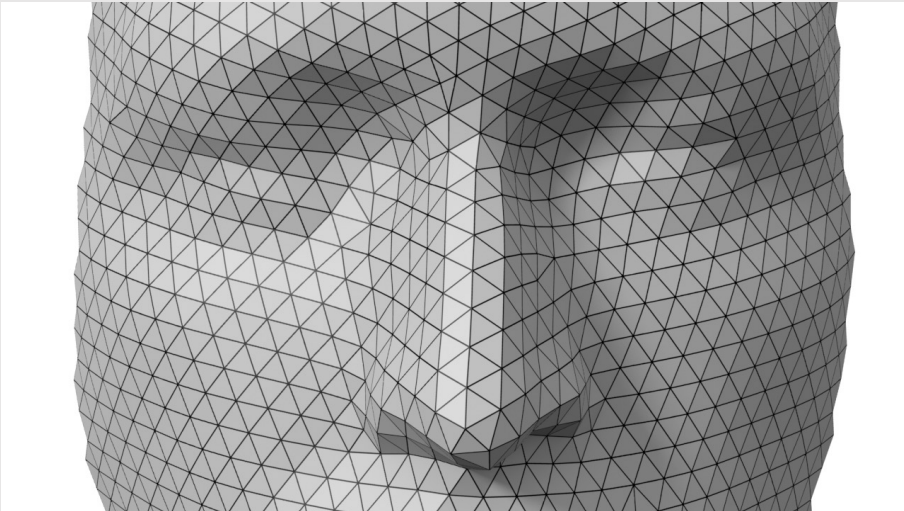


Smoothly-varying surface normals

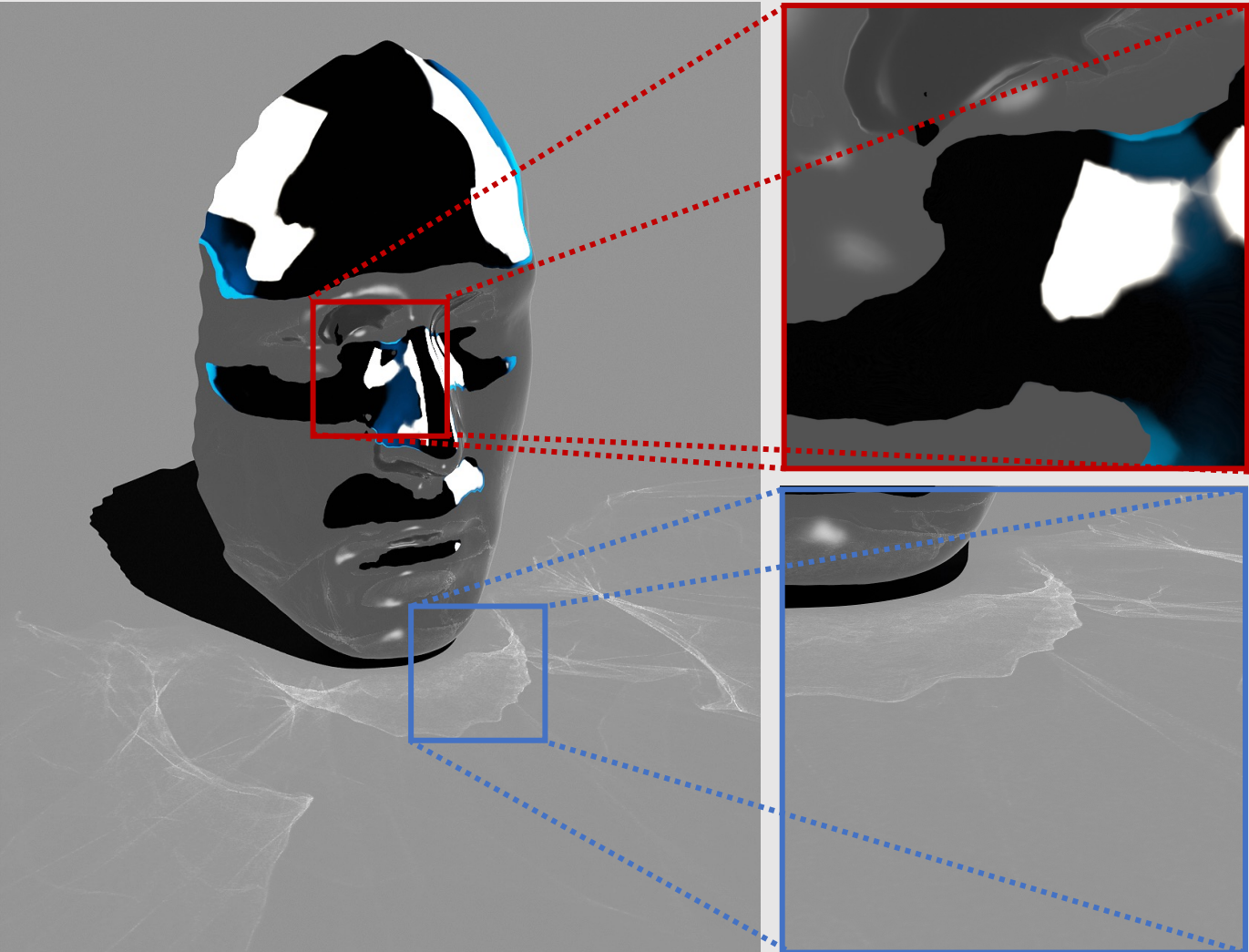
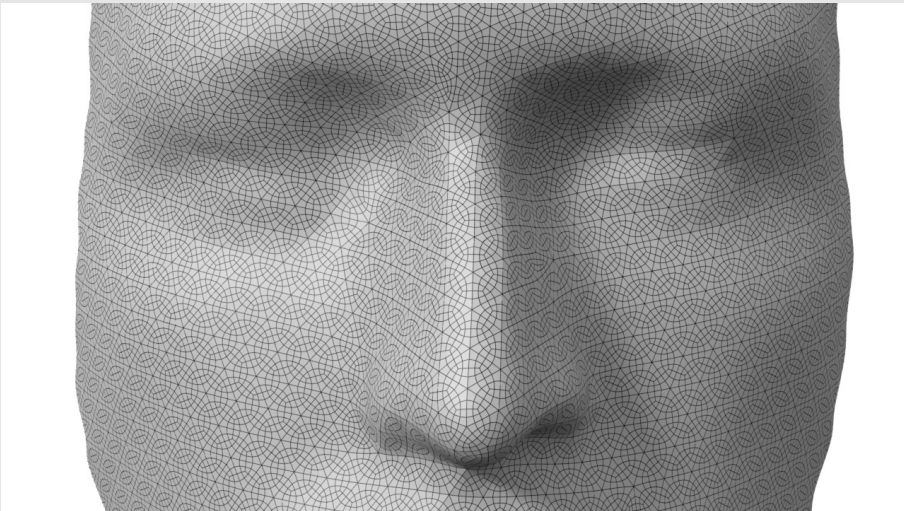
Smooth reflections/caustics



# Catmull-Clark Subdivision [Triangles]



Many irregular vertices



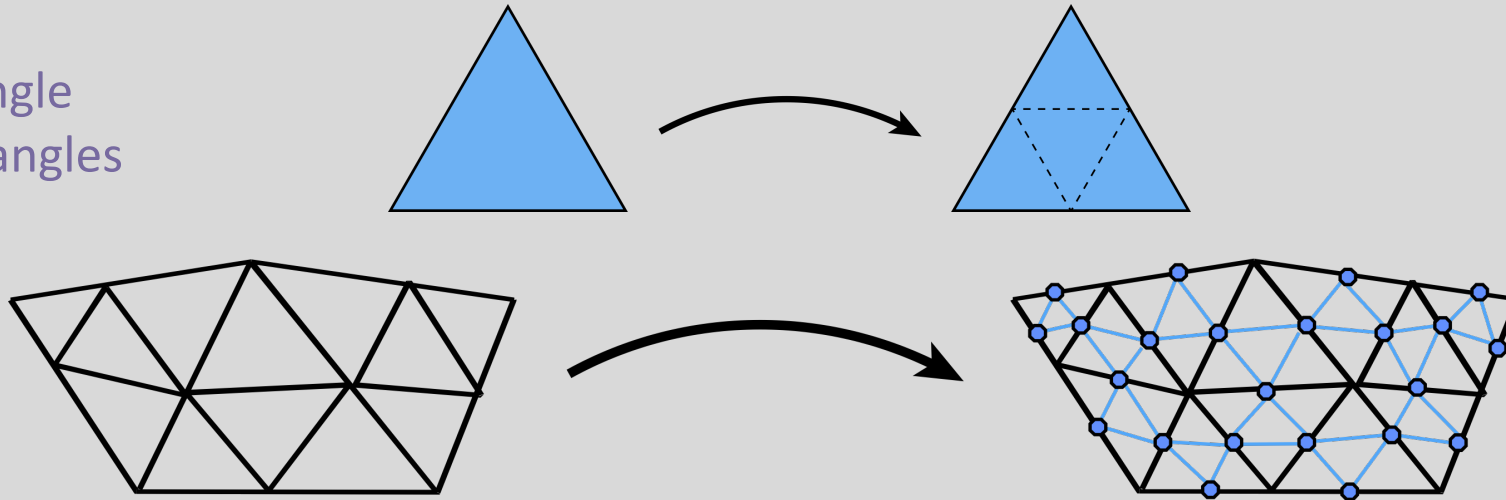
Erratic surface normals

Jagged reflections/caustics

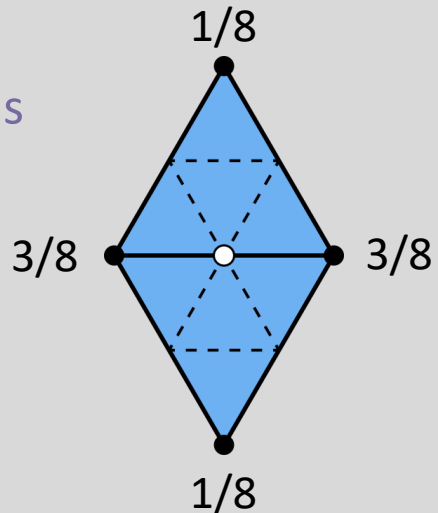
Is there a better subdivision scheme we can use for triangulated meshes?

# Loop Subdivision

Step 1:  
Split triangle  
into 4 triangles

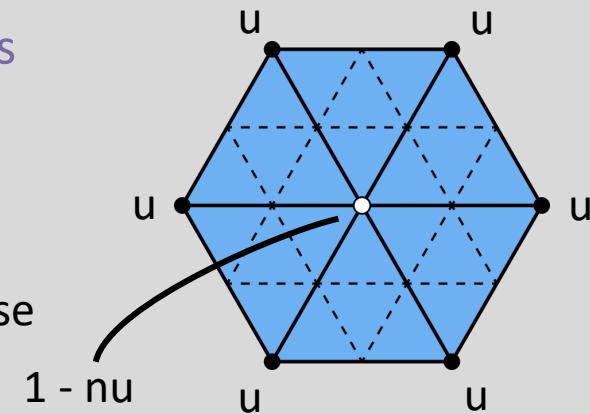


Step 2:  
Assign new coords



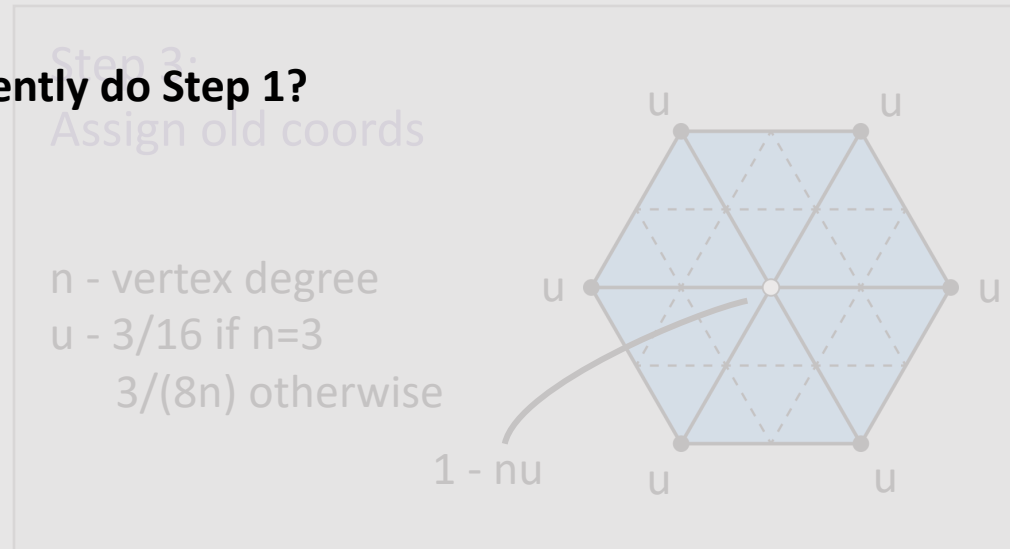
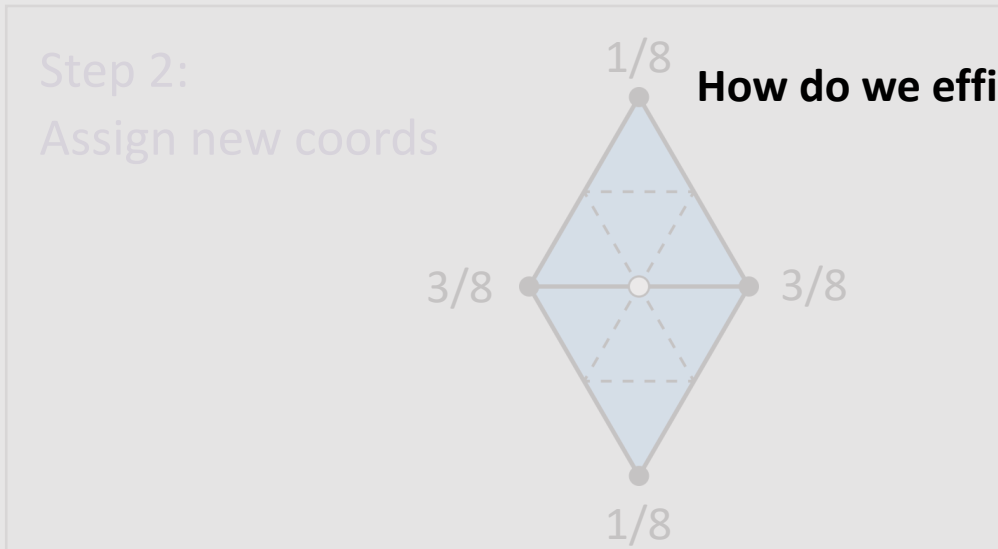
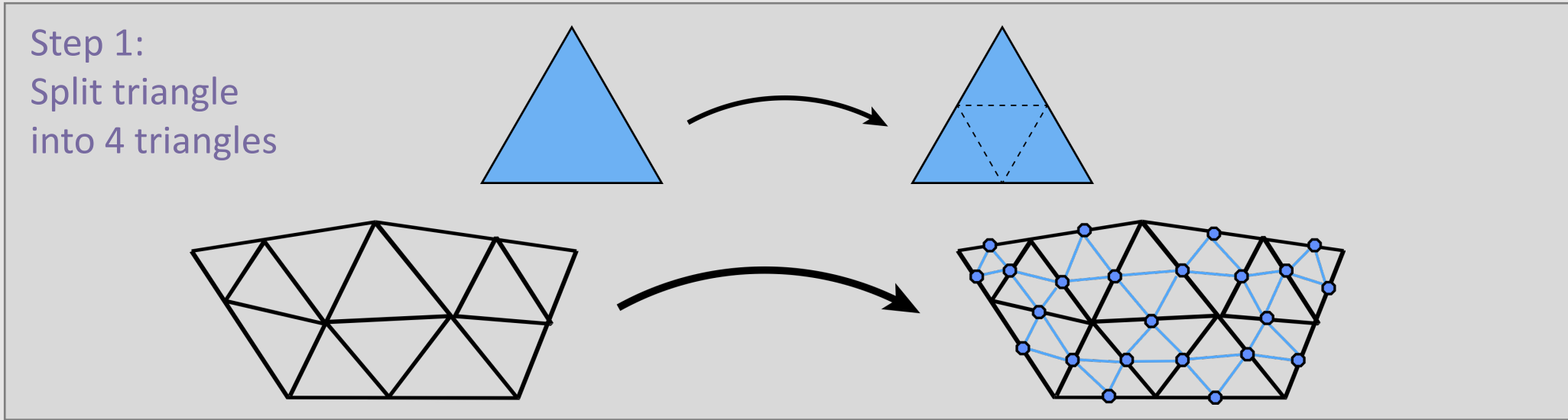
Step 3:  
Assign old coords

$n$  - vertex degree  
 $u$  -  $3/16$  if  $n=3$   
 $3/(8n)$  otherwise



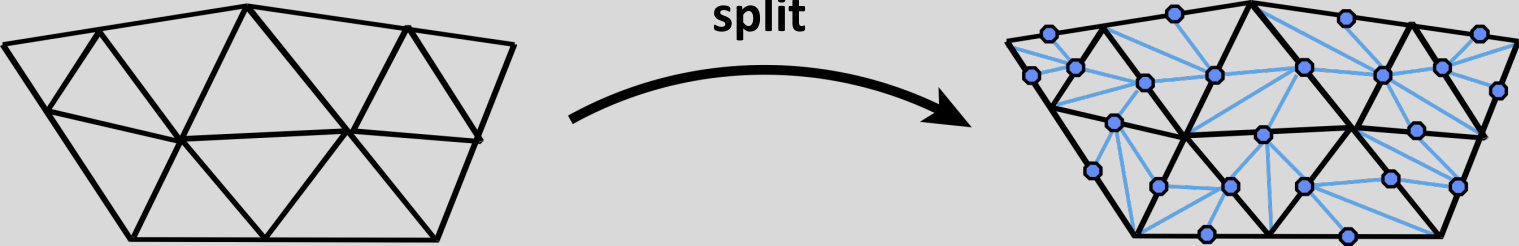


# Loop Subdivision

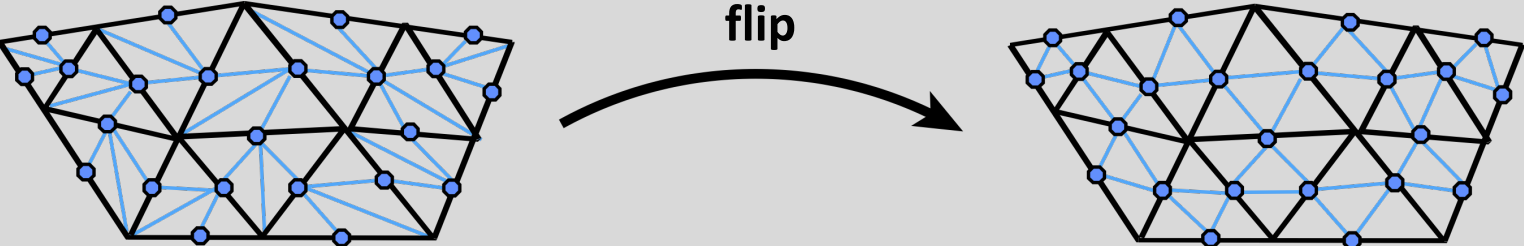


# Loop Subdivision Using Local Ops

Step 1:  
Split all edges in any order

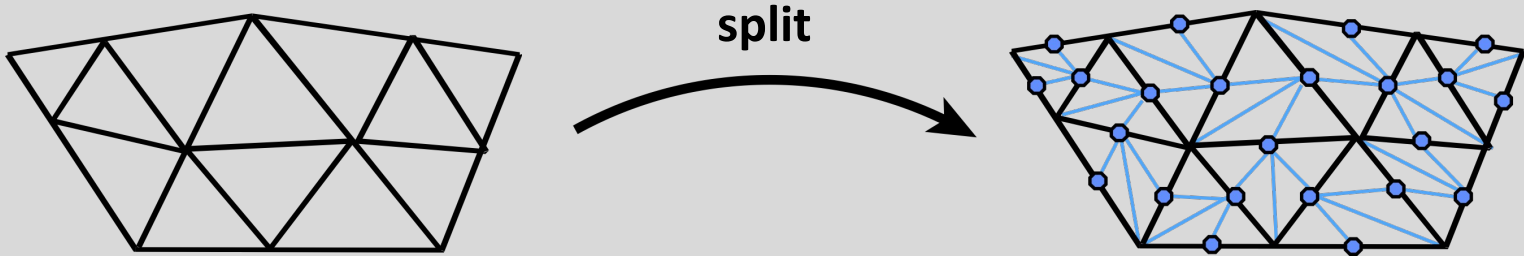


Step 2:  
Flip new edges until they touch two new vertices

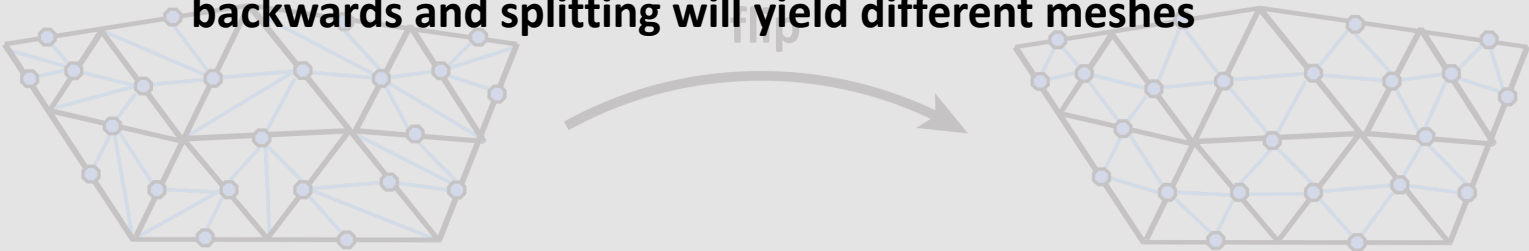


# Loop Subdivision Using Local Ops

Step 1:  
Split all edges in any order

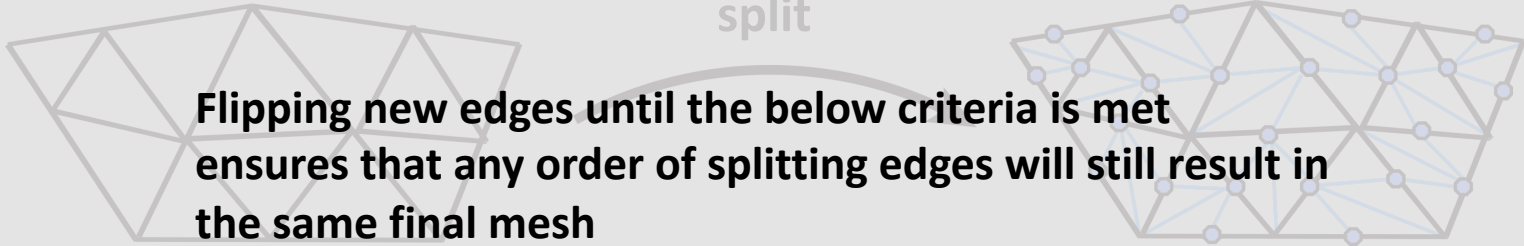


Step 2:  
**The order we traverse the edges and split them matter!**  
Flip new edges until they touch two new vertices  
**Traversing edges forward and splitting vs traversing them backwards and splitting will yield different meshes**

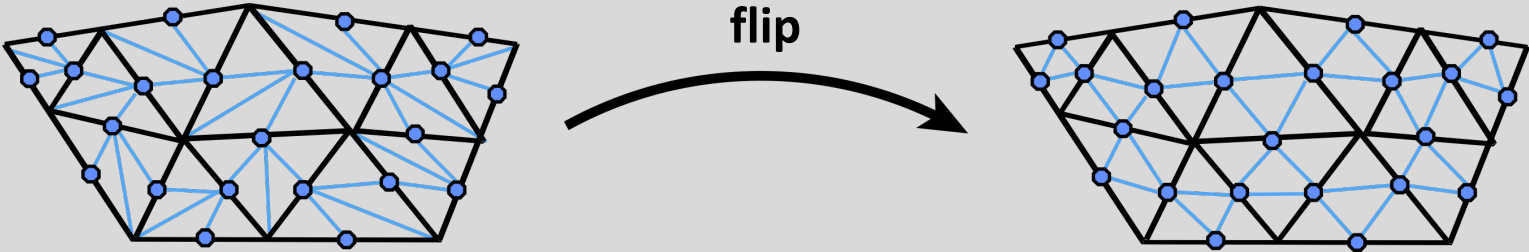


# Loop Subdivision Using Local Ops

Step 1:  
Split all edges in any order



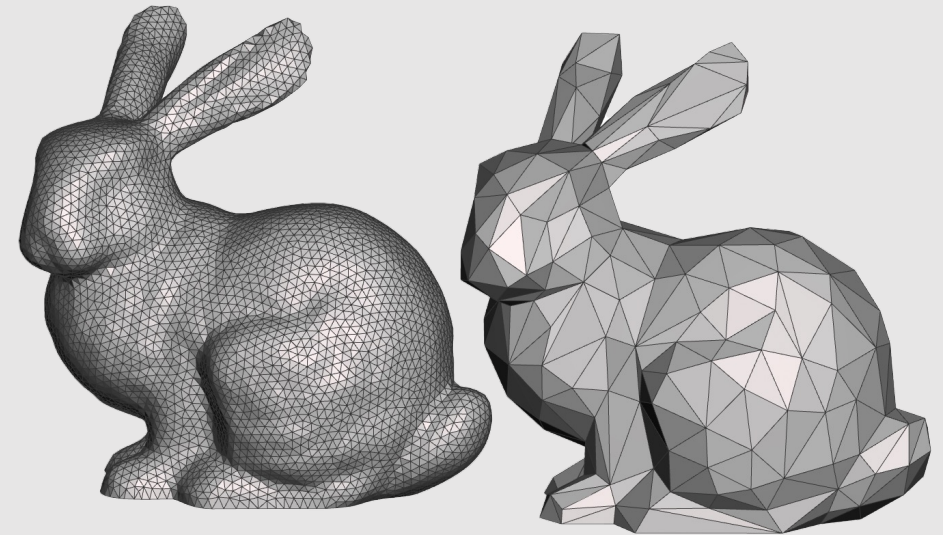
Step 2:  
Flip new edges until they touch two new vertices



- Digital Geometric Processing
  - ~~Geometric Subdivision~~
  - Geometric Simplification
  - Geometric Remeshing
  - Geometric Queries

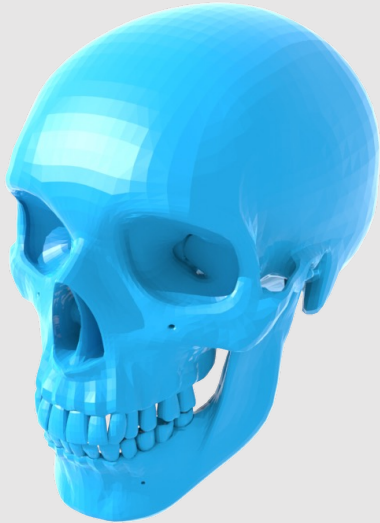
# Simplification

- Simplification is the process of **downsampling** a mesh
  - Less Storage overhead
    - Smaller file sizes
  - Less Processing overhead
    - Less elements to iterate over
  - Larger mesh modifications
    - Instead of moving tens of smaller mesh elements, move one larger mesh element

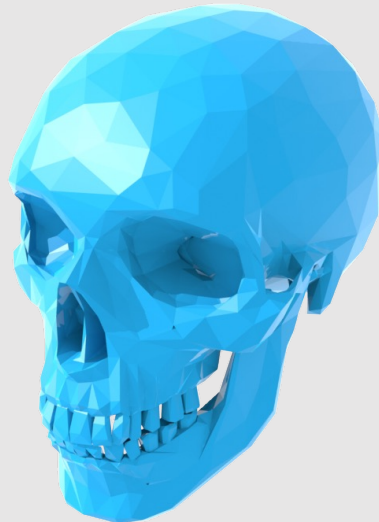


# Simplification Algorithm Basics

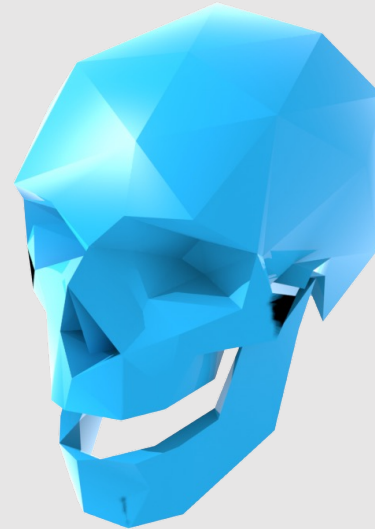
- Greedy Algorithm:
  - Assign each edge a cost
  - Collapse edge with least cost
  - Repeat until target number of elements is reached
- Particularly effective cost function: **quadratic error metric**\*\*



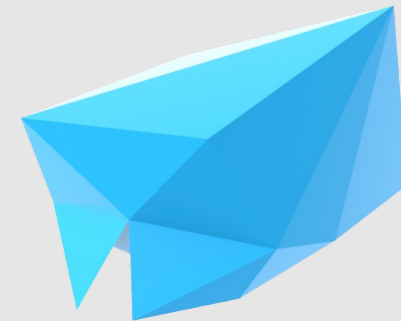
[ 30,000 triangles ]



[ 3,000 triangles ]



[ 300 triangles ]



[ 30 triangles ]

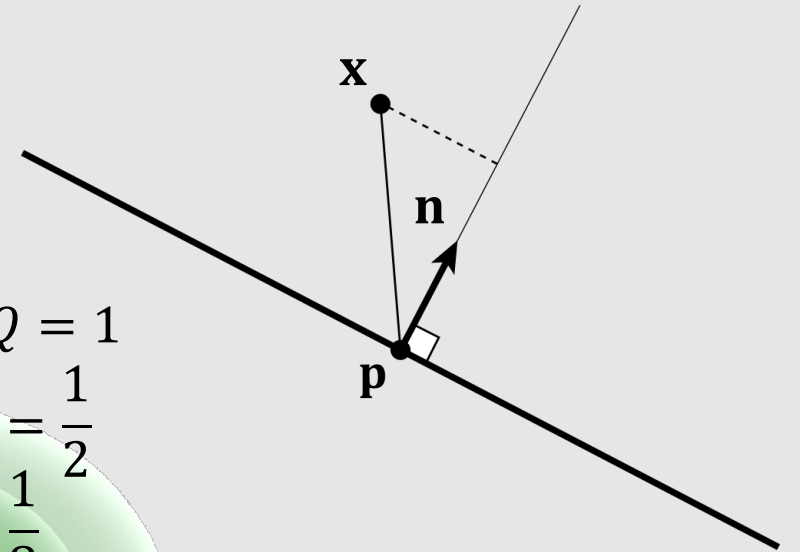
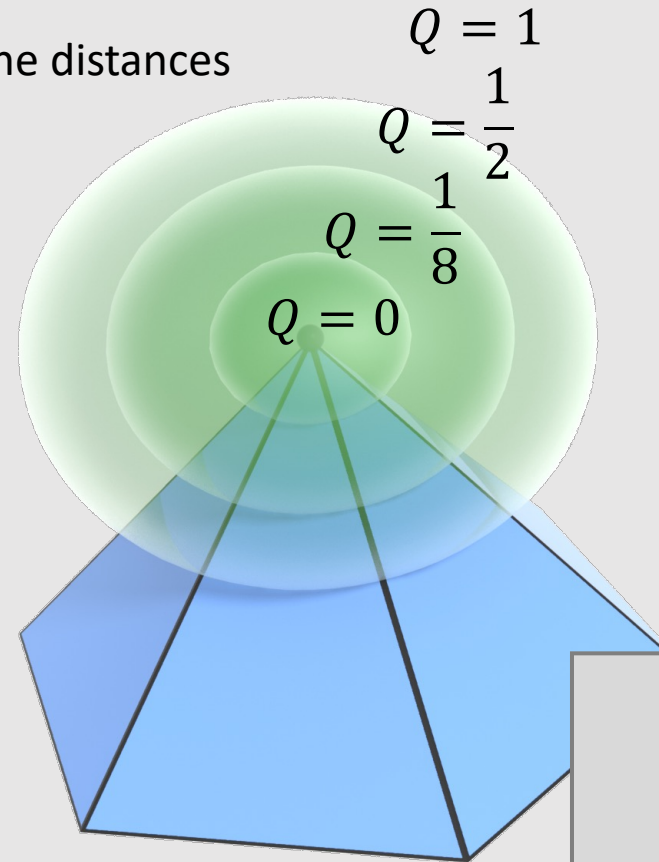
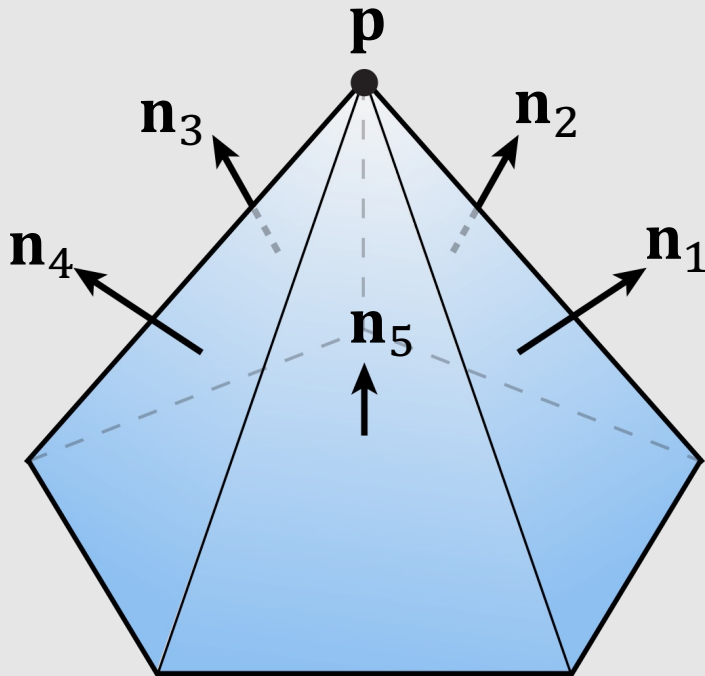
\*\*invented at CMU (Garland & Heckbert 1997)

# Quadric Error Metric

- **Goal:** approximate a point's distance from a collection of triangles
  - **Review:** what is the distance of a point  $\mathbf{x}$  from a plane  $\mathbf{p}$  with normal  $\mathbf{n}$ ?

$$\text{dist}(\mathbf{x}) = \langle \mathbf{n}, \mathbf{x} \rangle - \langle \mathbf{n}, \mathbf{p} \rangle = \langle \mathbf{n}, \mathbf{x} - \mathbf{p} \rangle$$

- Quadric error is the sum of squared point-to-plane distances



$$Q(\mathbf{x}) := \sum_{i=1}^k \langle \mathbf{n}_i, \mathbf{x} - \mathbf{p} \rangle^2$$

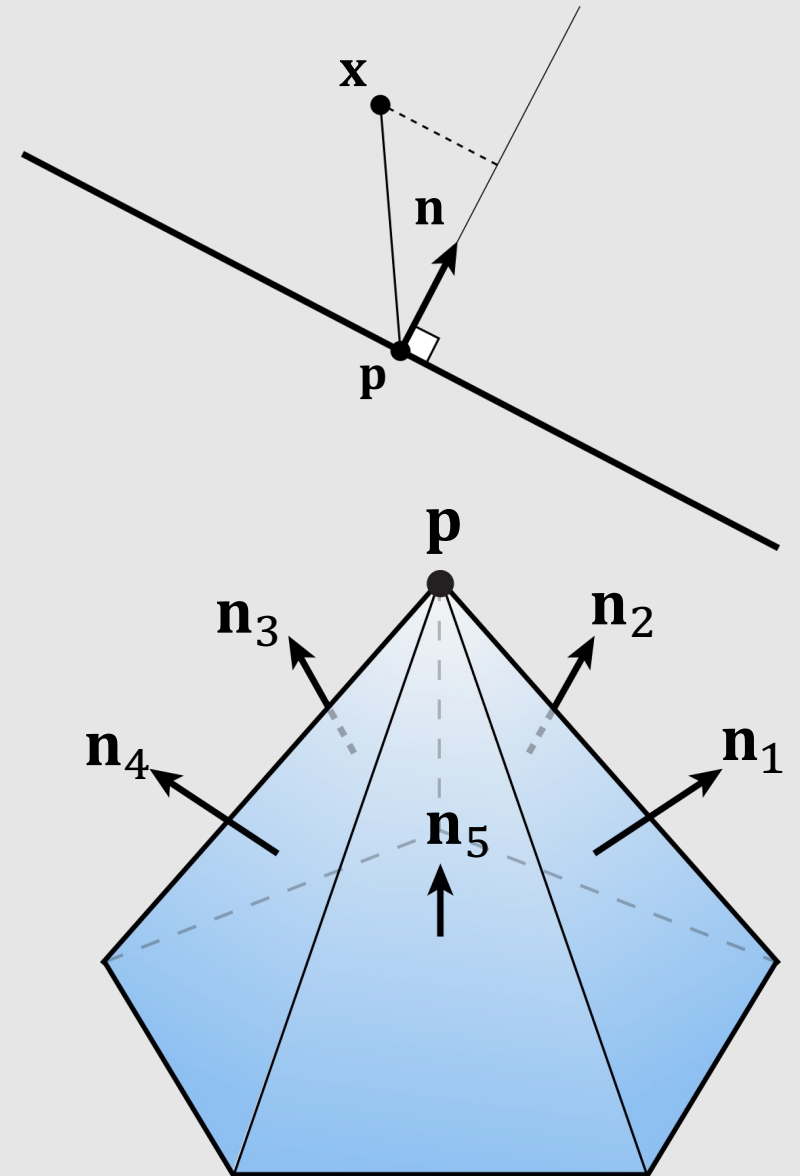


# Quadric Error Metric

- Given:
  - Query point  $\mathbf{x} = (x, y, z)$
  - Normal  $\mathbf{n} = (a, b, c)$
  - Offset from origin  $e = \langle \mathbf{n}, \mathbf{p} - 0 \rangle = \langle \mathbf{n}, \mathbf{p} \rangle$
  - We want the negative of this value to make a plane equation
    - $d = -e = -\langle \mathbf{n}, \mathbf{p} \rangle$
- We can rewrite in homogeneous coordinates:
  - $\mathbf{u} = (x, y, z, 1)$
  - $\mathbf{v} = (a, b, c, d)$
- Signed distance to plane is then just  $\langle \mathbf{u}, \mathbf{v} \rangle = ax + by + cz + d$ 
  - Note that it is zero in the plane!
- Squared distance is  $\langle \mathbf{u}, \mathbf{v} \rangle^2 = \mathbf{u}^T (\mathbf{v}\mathbf{v}^T) \mathbf{u} =: \mathbf{u}^T K \mathbf{u}$ 
  - Matrix  $K = \mathbf{v}\mathbf{v}^T$  encodes squared distance to plane

$$K = \begin{bmatrix} a^2 & ab & ac & ad \\ ab & b^2 & bc & bd \\ ac & bc & c^2 & cd \\ ad & bd & cd & d^2 \end{bmatrix}$$

$$\mathbf{u}^T K_1 \mathbf{u} + \mathbf{u}^T K_2 \mathbf{u} = \mathbf{u}^T (K_1 + K_2) \mathbf{u}$$



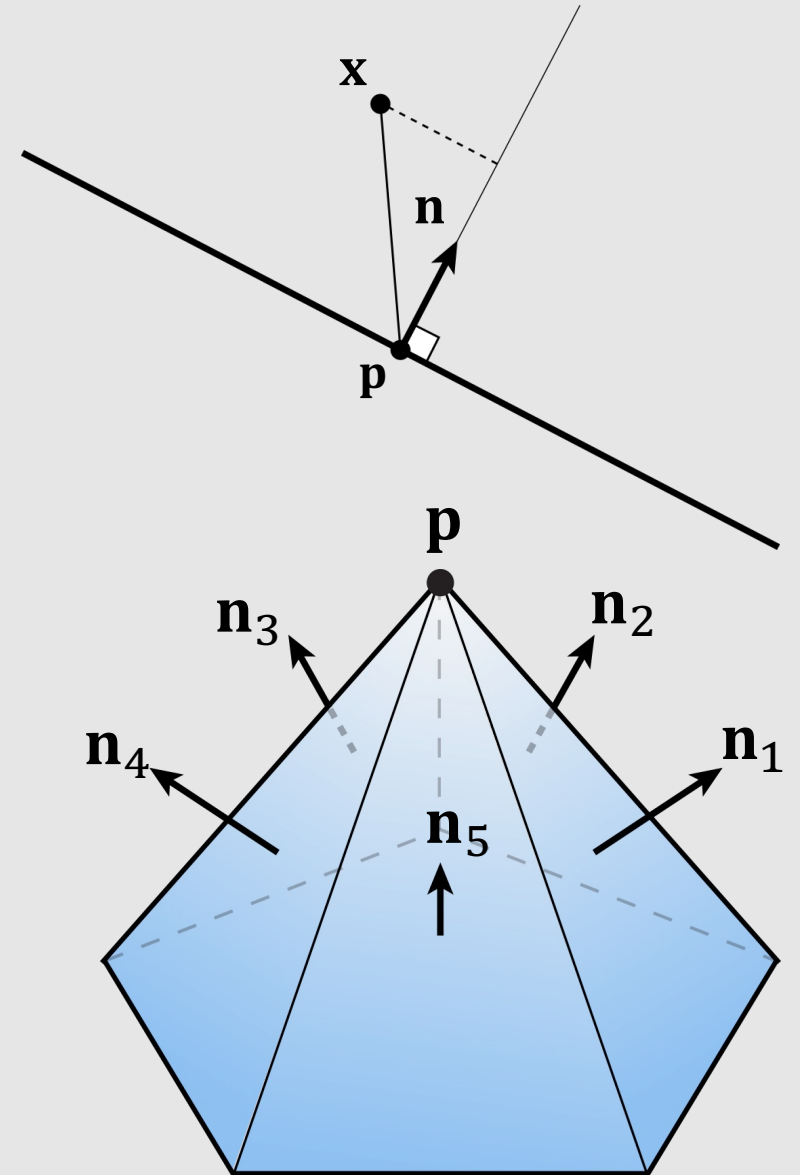
# Quadric Error Metric

- Given:
  - Query point  $\mathbf{x} = (x, y, z)$
  - Normal  $\mathbf{n} = (a, b, c)$
  - $d = -\langle \mathbf{n}, \mathbf{p} \rangle$
  - $\mathbf{u} = (x, y, z, 1)$
  - $\mathbf{v} = (a, b, c, d)$
- Signed distance to plane is  $\langle \mathbf{u}, \mathbf{v} \rangle = ax + by + cz + d$
- Squared distance is  $\langle \mathbf{u}, \mathbf{v} \rangle^2 = \mathbf{u}^\top (\mathbf{v}\mathbf{v}^\top) \mathbf{u} =: \mathbf{u}^\top K \mathbf{u}$ 
  - Matrix  $K = \mathbf{v}\mathbf{v}^\top$  encodes squared distance to plane

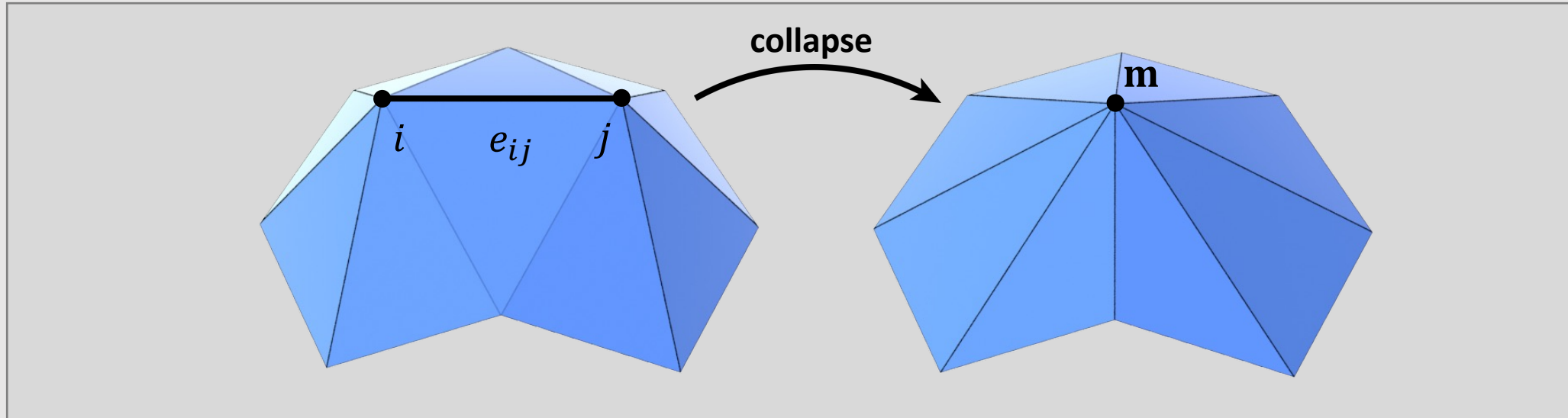
$$K = \begin{bmatrix} a^2 & ab & ac & ad \\ ab & b^2 & bc & bd \\ ac & bc & c^2 & cd \\ ad & bd & cd & d^2 \end{bmatrix}$$

- Key Idea:** sum of matrices  $K$  represents distance to a union of planes

$$\mathbf{u}^\top K_1 \mathbf{u} + \mathbf{u}^\top K_2 \mathbf{u} = \mathbf{u}^\top (K_1 + K_2) \mathbf{u}$$



# Quadric Error of Edge Collapse

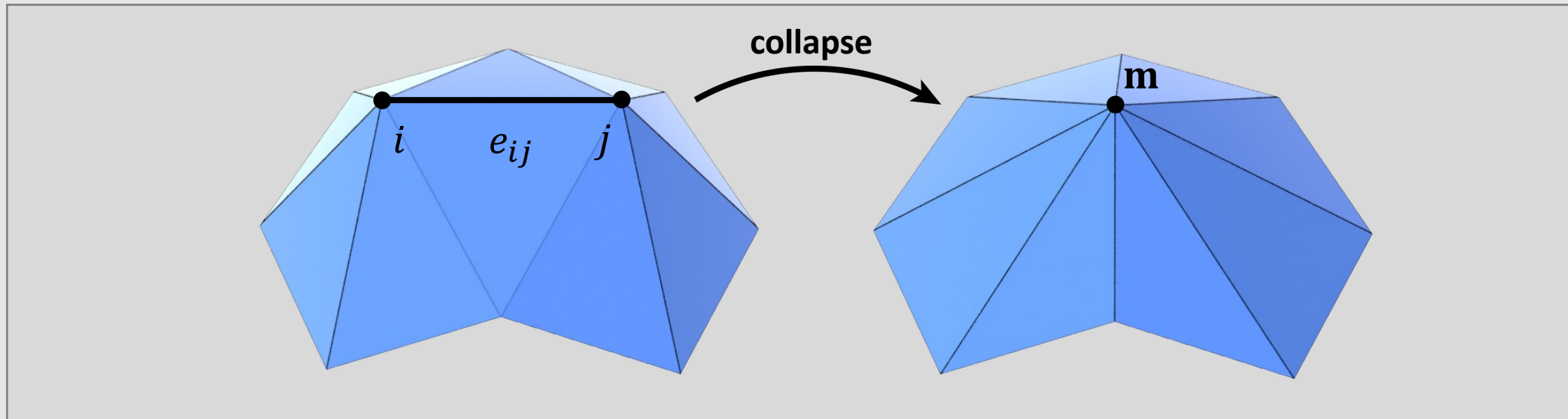


- How much does it cost to collapse an edge  $e_{ij}$ ?
  - Compute midpoint  $\mathbf{m}$ , measure error as

$$Q(\mathbf{m}) = \mathbf{m}^T(K_i + K_j)\mathbf{m}$$

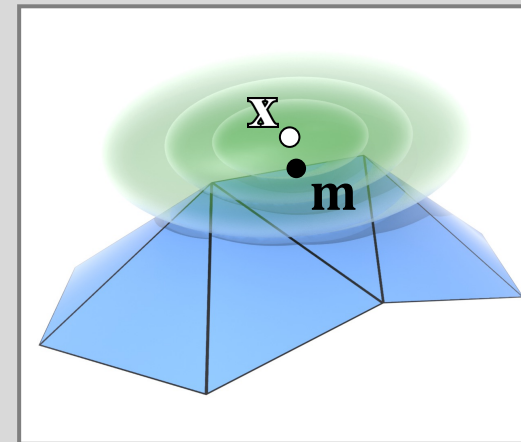
- Error becomes “score” for  $e_{ij}$ , determining priority
  - Q: where to put  $\mathbf{m}$ ?

# Quadric Error of Edge Collapse



$$Q(\mathbf{m}) = \mathbf{m}^T (K_i + K_j) \mathbf{m}$$

- Find point  $\mathbf{x}$  that minimizes error
  - Take derivatives!



How to take a derivative of a function involving matrices?

# Minimizing a Quadratic Function

To find the min of a function  $f(x)$

$$f(x) = ax^2 + bx + c$$

take derivative  $f'(x)$  and set equal to 0

$$f'(x) = 2ax + b = 0$$

$$x = -b/2a$$

same structure

can also write any quadratic function of n variables as a symmetric matrix A  
consider the multivariable function

$$f(x, y) = ax^2 + bxy + cy^2 + dx + ey + g$$

we can rewrite it as:

$$\mathbf{x} = \begin{bmatrix} x \\ y \end{bmatrix} \quad A = \begin{bmatrix} a & b/2 \\ b/2 & c \end{bmatrix} \quad \mathbf{u} = \begin{bmatrix} d \\ e \end{bmatrix}$$

$$f(x, y) = \mathbf{x}^T A \mathbf{x} + \mathbf{u}^T \mathbf{x} + g$$

take derivative  $f'(x)$  and set equal to 0

$$f'(x, y) = 2A\mathbf{x} + \mathbf{u} = 0$$

$$\mathbf{x} = -\frac{1}{2}A^{-1}\mathbf{u}$$

same structure

# Positive Definite Quadratic Form

How do we know if our solution minimizes quadratic error?

$$\mathbf{x} = -\frac{1}{2}A^{-1}\mathbf{u}$$

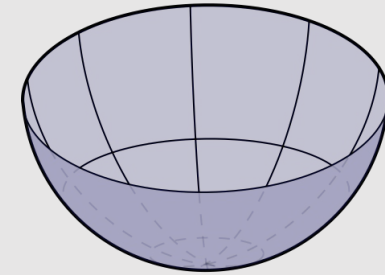
In the 1D case, we minimize the function if

$$\begin{aligned}xax &= ax^2 > 0 \\ a &> 0\end{aligned}$$

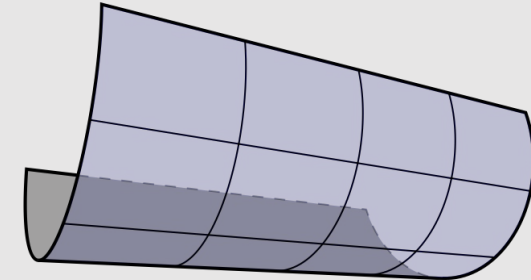
In the ND case, we minimize the function if

$$\mathbf{x}^T A \mathbf{x} > 0 \quad \forall \mathbf{x}$$

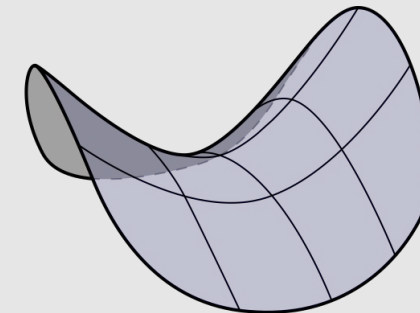
This is known as the function being positive semidefinite



[ positive definite ]



[ positive semidefinite ]



[ indefinite ]

# Minimizing Quadric Error

Find “best” point for edge collapse by minimizing quadratic form

$$\min_{\mathbf{u} \in \mathbb{R}^4} \mathbf{u}^T K \mathbf{u}$$

Already know fourth (homogeneous) coordinate for a point is 1

Break up our quadratic function into two pieces

$$\begin{aligned} & \begin{bmatrix} \mathbf{x}^T & 1 \end{bmatrix} \begin{bmatrix} B & \mathbf{w} \\ \mathbf{w}^T & d^2 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ 1 \end{bmatrix} \\ &= \mathbf{x}^T B \mathbf{x} + 2\mathbf{w}^T \mathbf{x} + d^2 \end{aligned}$$

Can minimize as before

$$\begin{aligned} 2B\mathbf{x} + 2\mathbf{w} &= 0 \\ \mathbf{x} &= -B^{-1}\mathbf{w} \end{aligned}$$



# Quadric Error Simplification Algorithm

```
// compute K for each face
for(v : vertices) {
    for(f : faces) {
        Vec4 ve(N, d);
        f->K = outer(ve, ve);
    }
}
```

```
// compute K for each vertex
for(v : vertices)
    for(f : v->faces())
        v->K += f->K;
```

```
// compute K for each edge
// place into priority queue
PriorityQueue pq;
for(e : edge) {
    for(v : e->vertices())
        e->K += v->K;
    pq.push(e->K, e);
}
```

```
// iterate until mesh is a target size
while(faces.length() > target_size) {

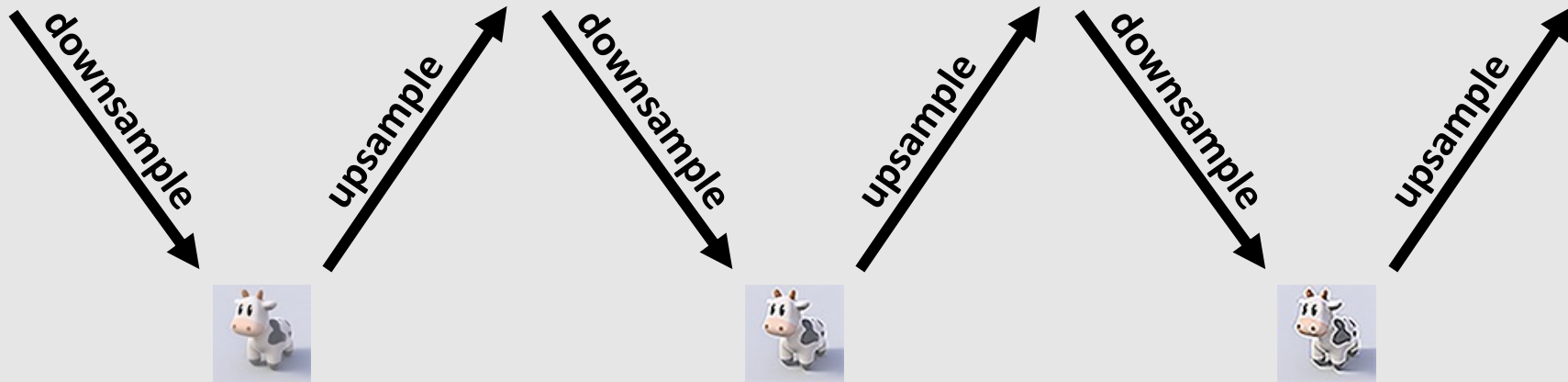
    // collapse edge with smallest cost
    e = pq.pop();
    K = e->K;
    v = collapse(e);

    // position new vertex to optimal pos
    v->pos = -B.inv() * w

    // update K for vertex
    // update K for edges touching vertex
    v->K = K;
    for(e2 : v->edges()) {
        e2->K = 0
        for(v2 : e2->vertices())
            e2->K += v2->K;
    }
}
```

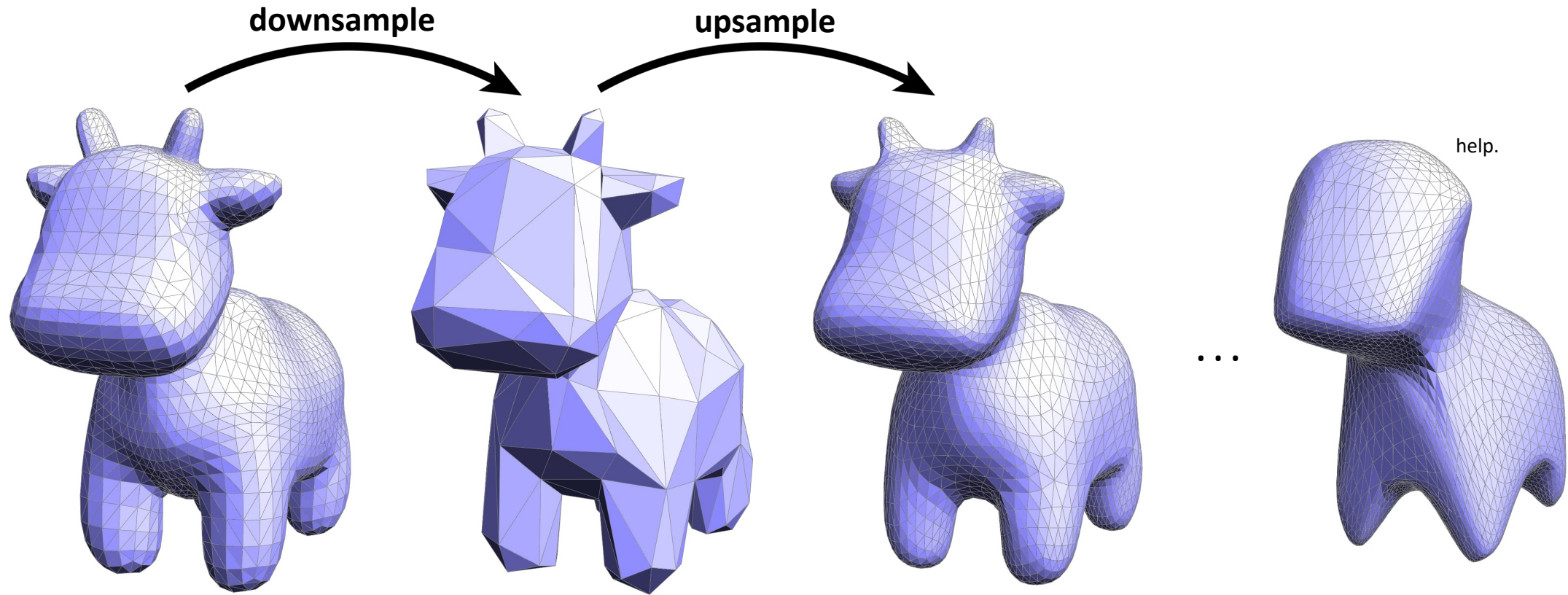
Is simplification the inverse operation of subdivision?

# Dangers of Resampling



**Repeatedly resampling an image degrades signal quality!**

# Dangers of Resampling



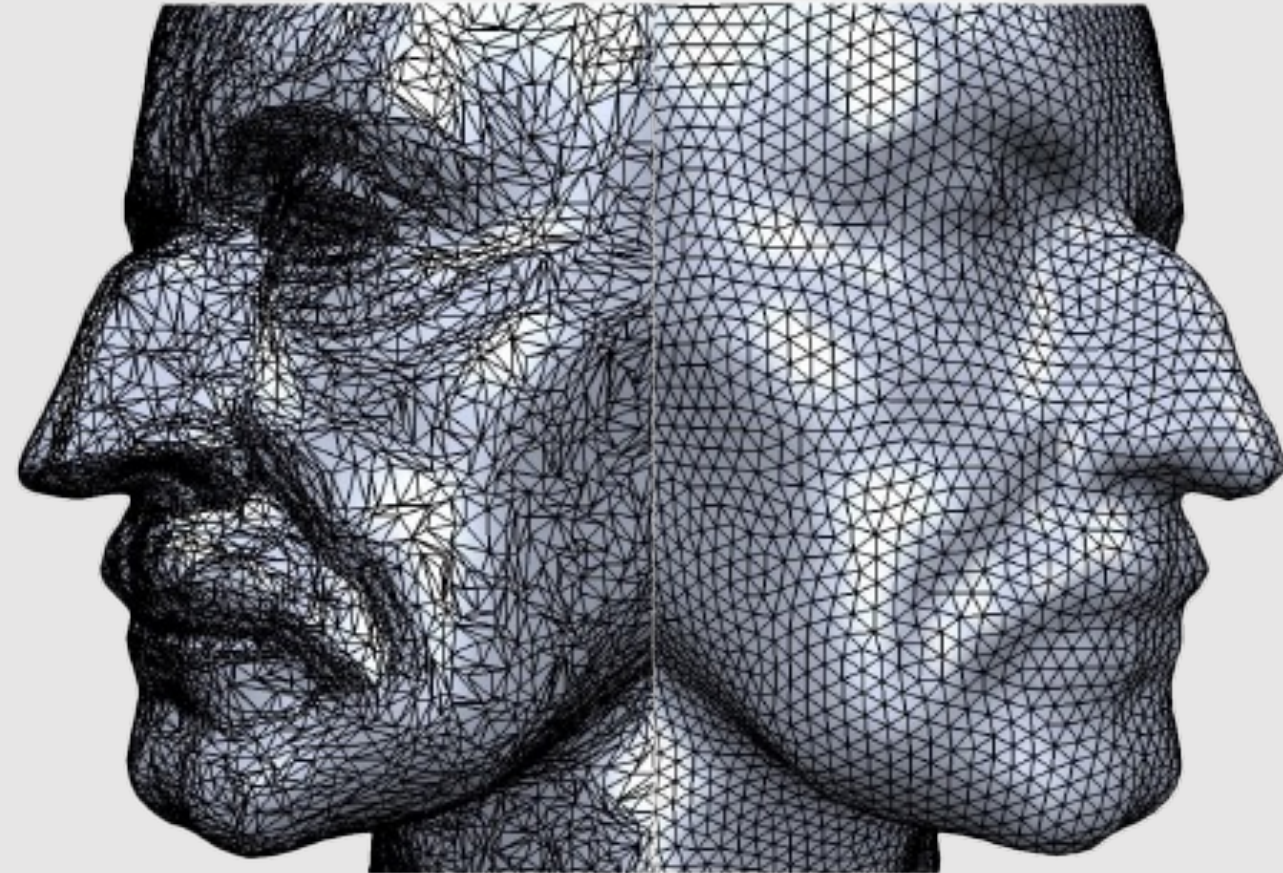
**Repeatedly resampling a mesh also degrades signal quality!**

- Digital Geometric Processing
  - ~~Geometric Subdivision~~
  - ~~Geometric Simplification~~
  - Geometric Remeshing
  - Geometric Queries



# Isotropic Remeshing

- **Isotropic:** same value when measured in any direction
- **Remeshing:** a change in the mesh
  - **Goal:** change the mesh to make triangles more uniform shape and size
- Helps achieve good mesh properties:
  - Good approximation of original shape
  - Vertex degrees close to 6
  - Angles close to 60deg
  - Delaunay triangles

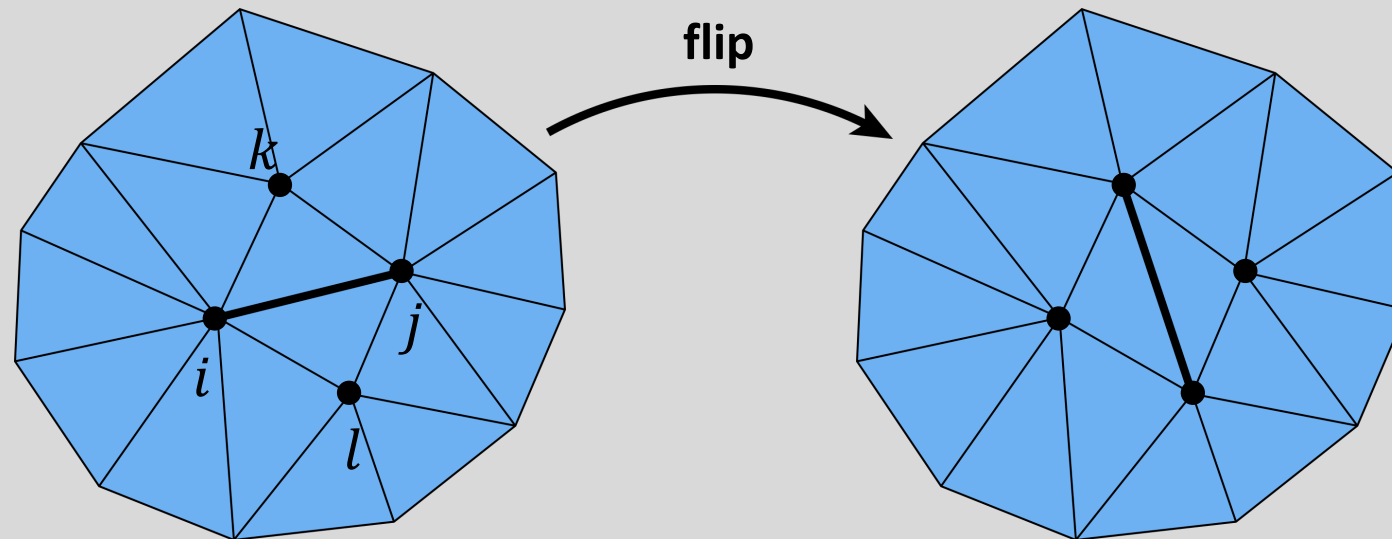


# Improving Degree

Vertices with degree 6 makes triangles more regular

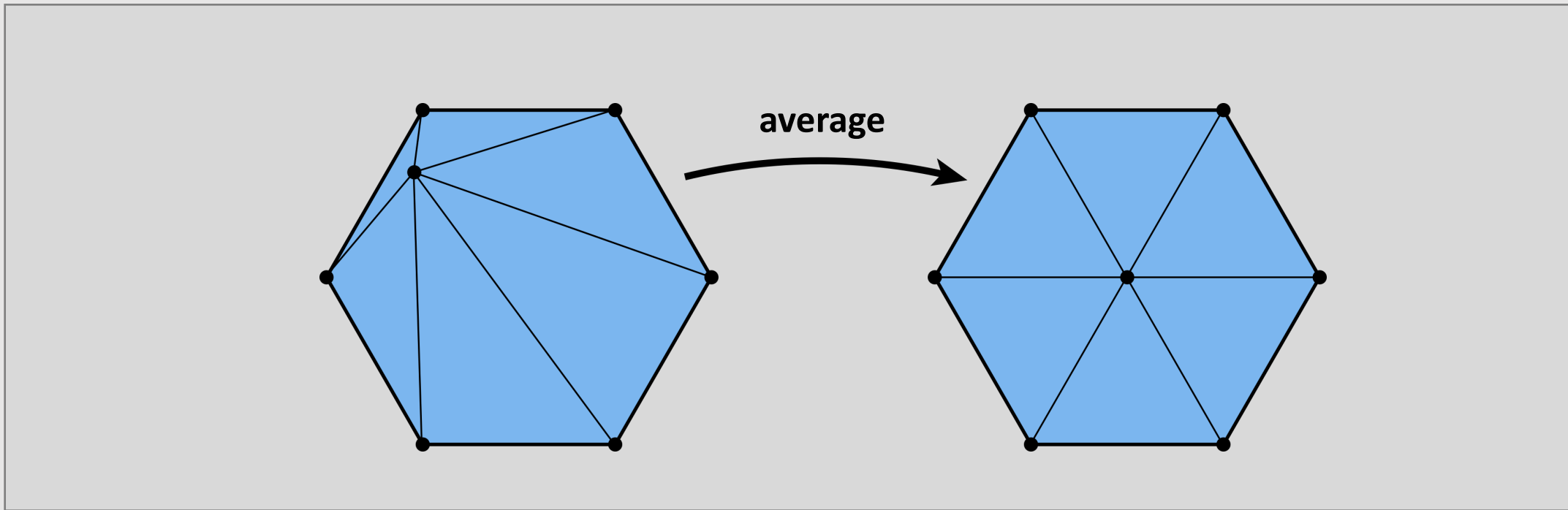
**Deviation function:**  $|d_i - 6| + |d_j - 6| + |d_k - 6| + |d_l - 6|$

If flipping an edge reduces deviation function, flip edge



# Improving Vertex Positioning

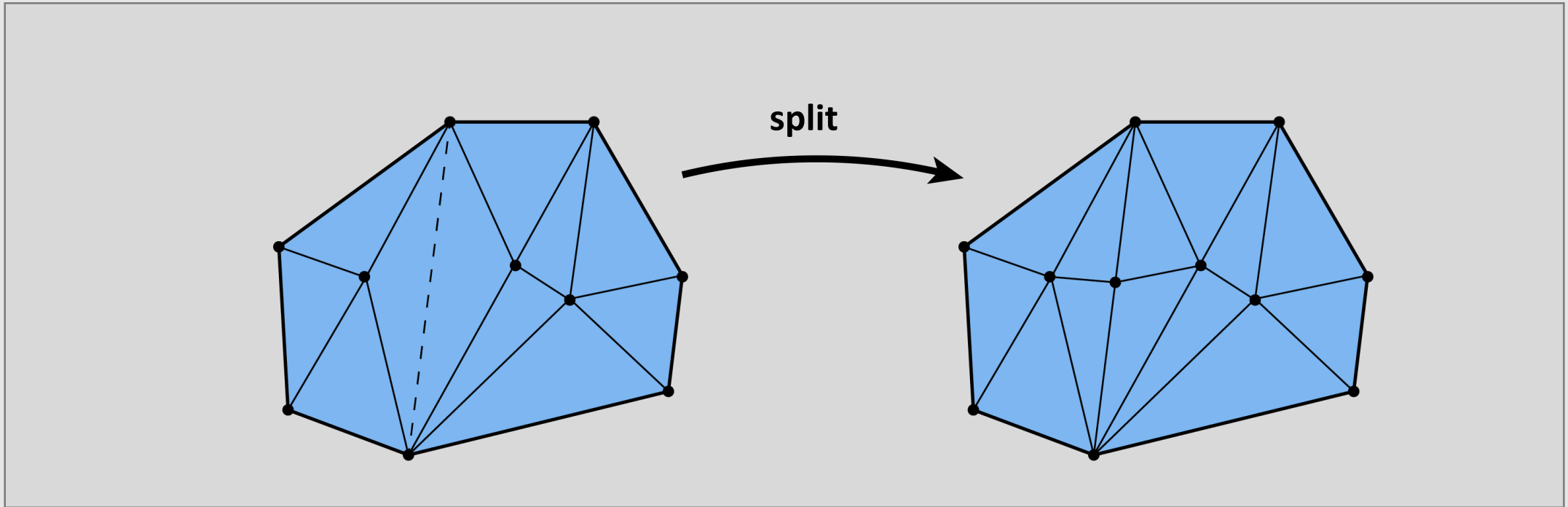
Center vertices to make triangles more even in size





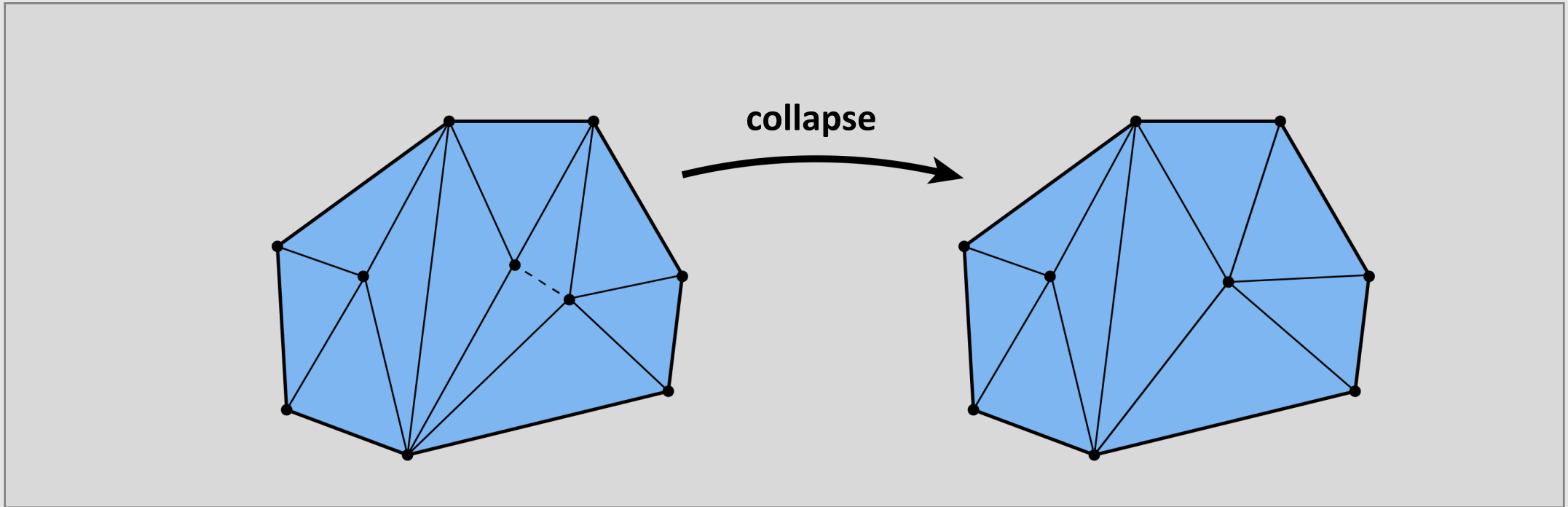
# Improving Edge Length

If an edge is longer than  $(4/3 * \text{mean})$  length, split it

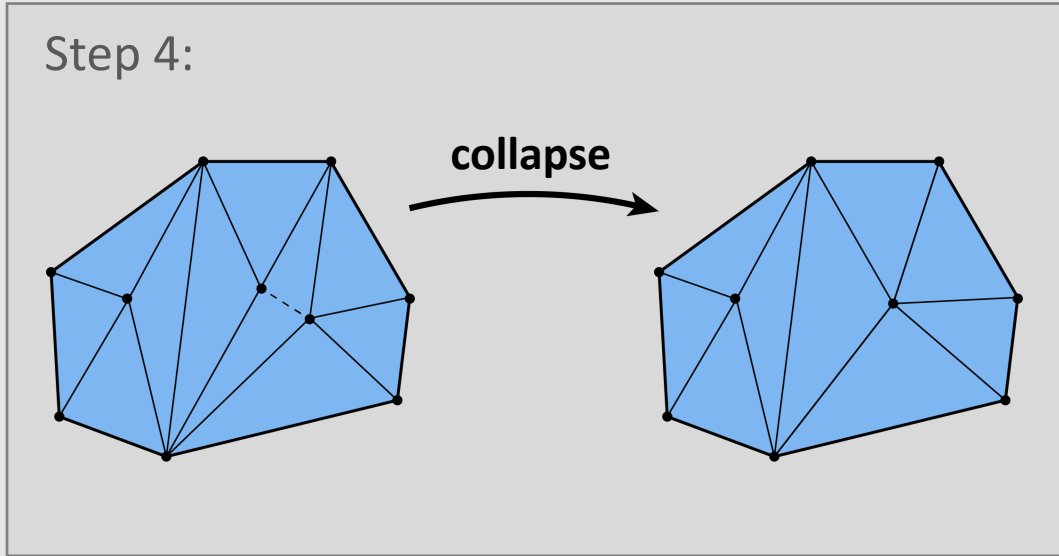
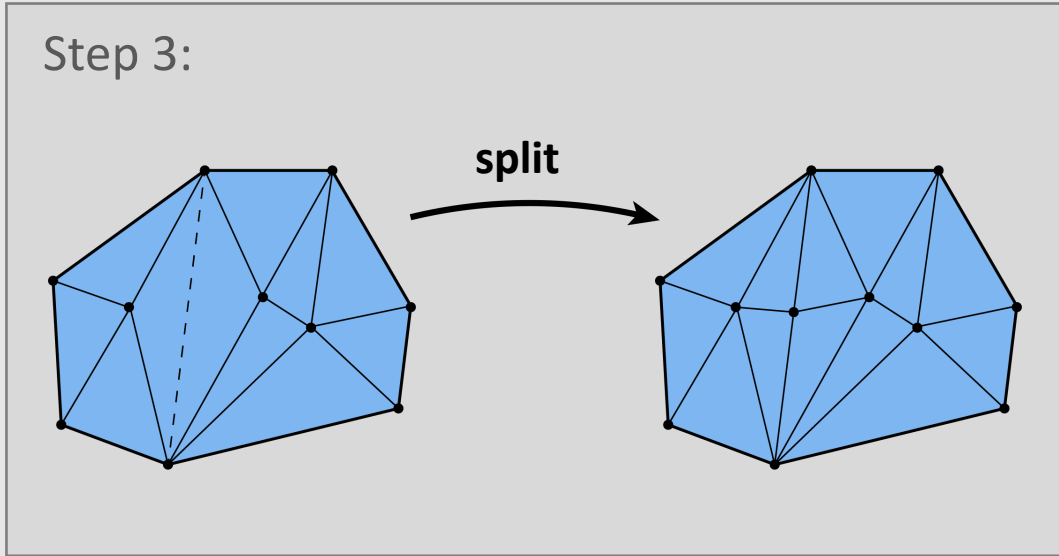
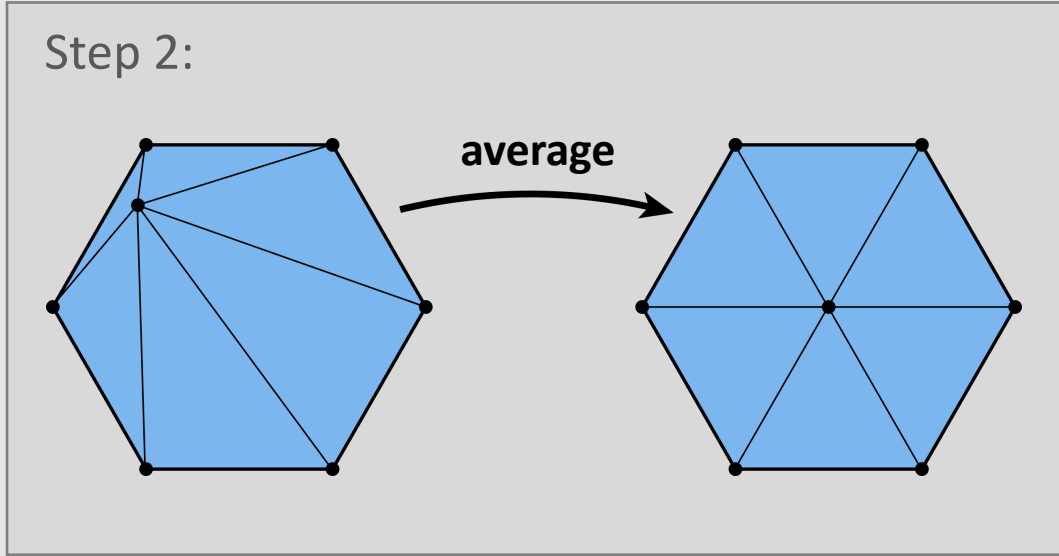
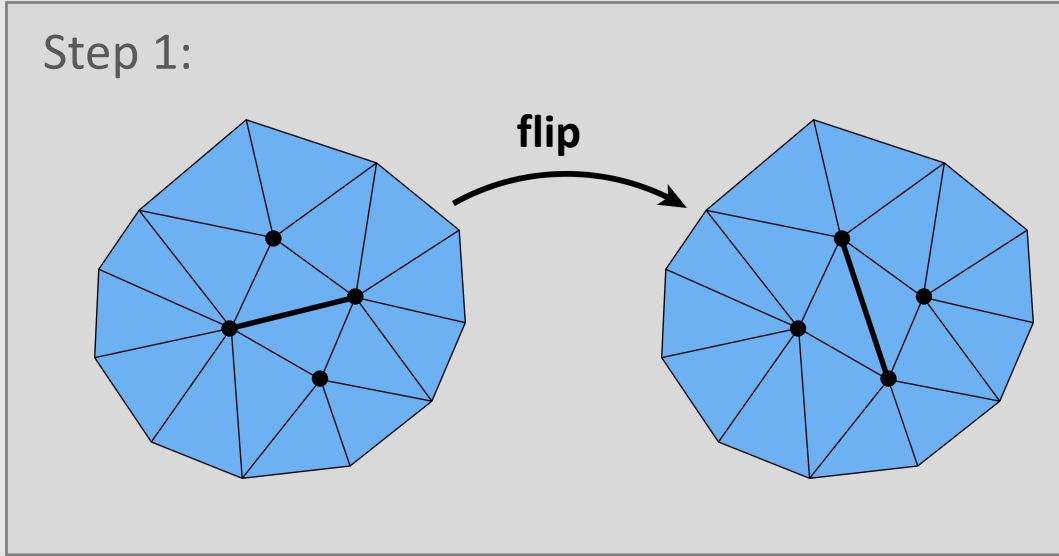


# Improving Edge Length

If an edge is shorter than  $(4/5 * \text{mean})$  length, collapse it



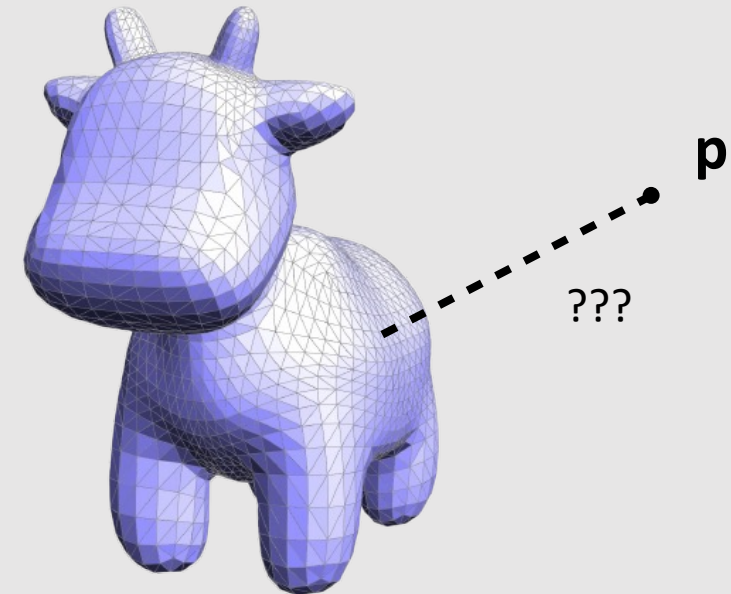
# Isotropic Remeshing



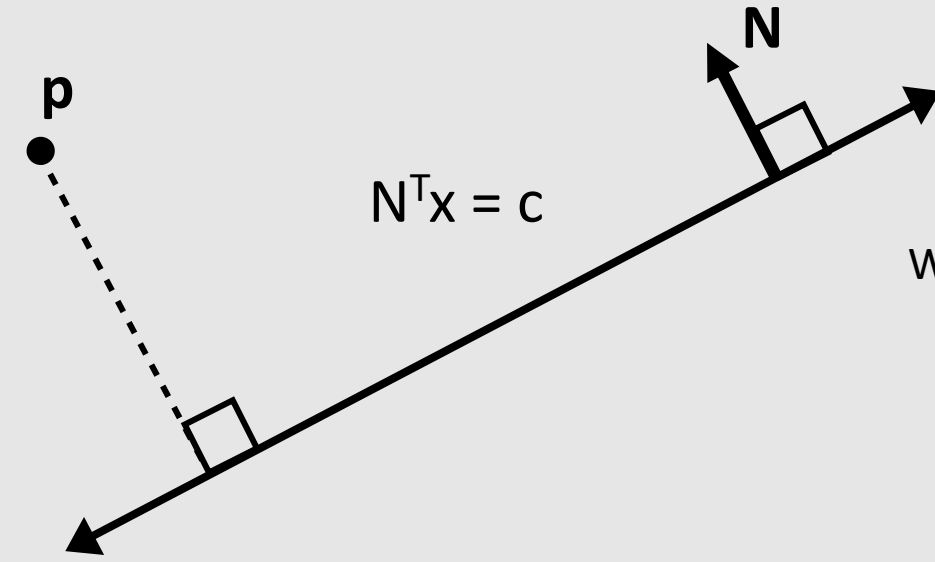
- Digital Geometric Processing
  - ~~Geometric Subdivision~~
  - ~~Geometric Simplification~~
  - ~~Geometric Remeshing~~
  - Geometric Queries

# Closest Point Queries

- **Problem:** given a point, in how do we find the closest point on a given surface?
- Several use cases:
  - Ray/mesh intersection in pathtracing
  - Kinematics/animation
  - GUI/user selection
    - When I click on a mesh, what point am I actually clicking on?



# Closest Point on a Line



To find the closest point to  $p$  along  $N^T x = c$   
We can have  $p$  travel along  $N$  for some time  $t$

$$N^T(p + tN) = c$$

Multiplying the terms out

$$N^T p + tN^T N = c$$

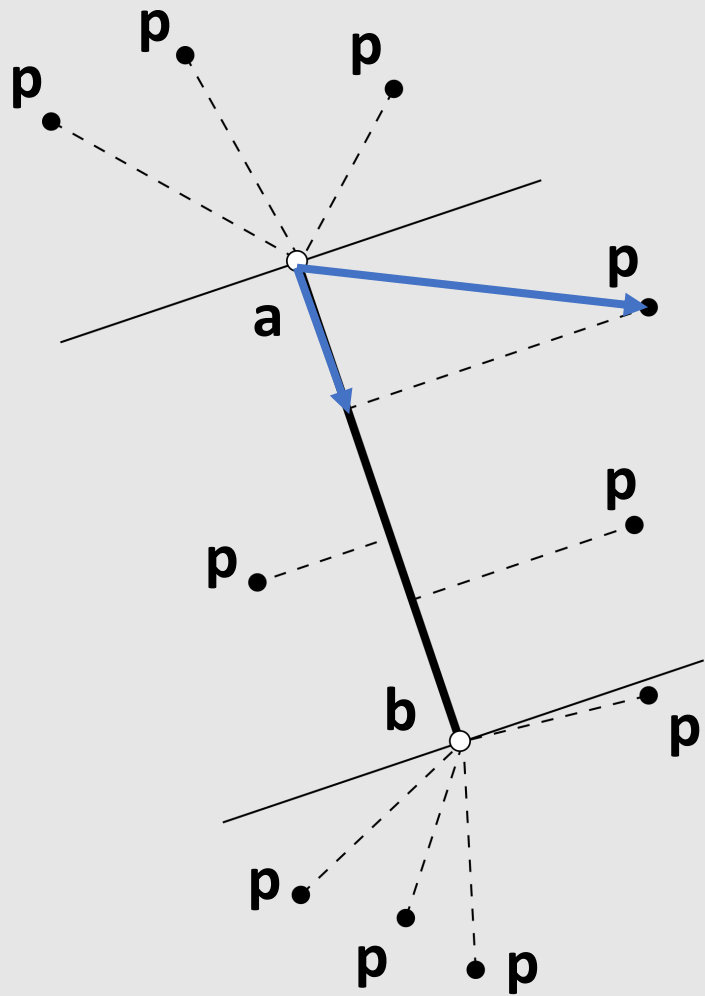
The unit norm multiplied by itself is 1  
Solve for  $t$

$$t = c - N^T p$$

Propagate  $p$  along  $N$  for time  $t$

$$p + tN$$
$$p + (c - N^T p)N$$

# Closest Point on a Line Segment



Compute the vector  $\mathbf{p}$  from the line base  $\mathbf{a}$  along the line

$$\langle \mathbf{p} - \mathbf{a}, \mathbf{b} - \mathbf{a} \rangle$$

Normalize to get a time

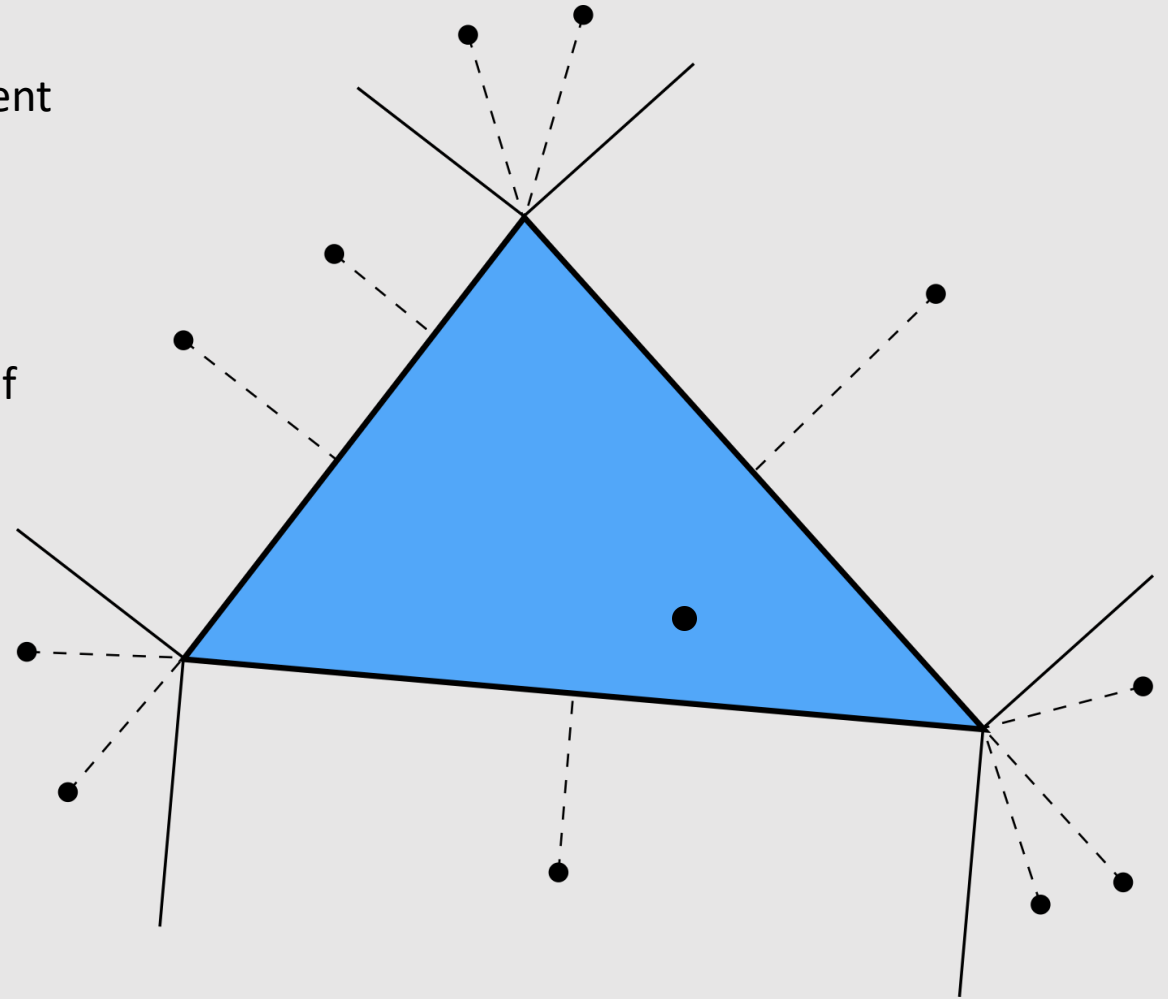
$$t = \frac{\langle \mathbf{p} - \mathbf{a}, \mathbf{b} - \mathbf{a} \rangle}{\langle \mathbf{b} - \mathbf{a}, \mathbf{b} - \mathbf{a} \rangle}$$

Clip time to range  $[0,1]$  and interpolate

$$\mathbf{a} + (\mathbf{b} - \mathbf{a})t$$

# Closest Point on a 2D Triangle

- Easy! Just compute closest point to each line segment
  - For each point, compute distance
  - Point with smallest distance wins
- What if the point is inside the triangle?
  - Even easier! The closest point is the point itself
  - Recall point-in-triangle tests





# Closest Point on a 3D Triangle

- **Method #1: Projection\*\***
  - Construct a plane that passes through the triangle
    - Can be done using cross product of edges
  - Project the point to the closest point on the plane
    - Same expression as with a line:  $p + (c - N^T p)N$
    - Check if point is in triangle using half-plane test
  - Else, compute distance from each line segment in 3D
    - Same expression as with a 2D line segment
- **Method #2: Rotation\*\***
  - Translate point + triangle so that triangle vertex v1 is at the origin
  - Rotate point + triangle so that triangle vertex v2 sits on the z-axis
  - Rotate point + triangle so that triangle vertex v3 sits on the yz-axis
  - Disregard x-coordinate of point
    - Problem reduces to closest point on 2D triangle

\*\*<https://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.104.4264&rep=rep1&type=pdf>

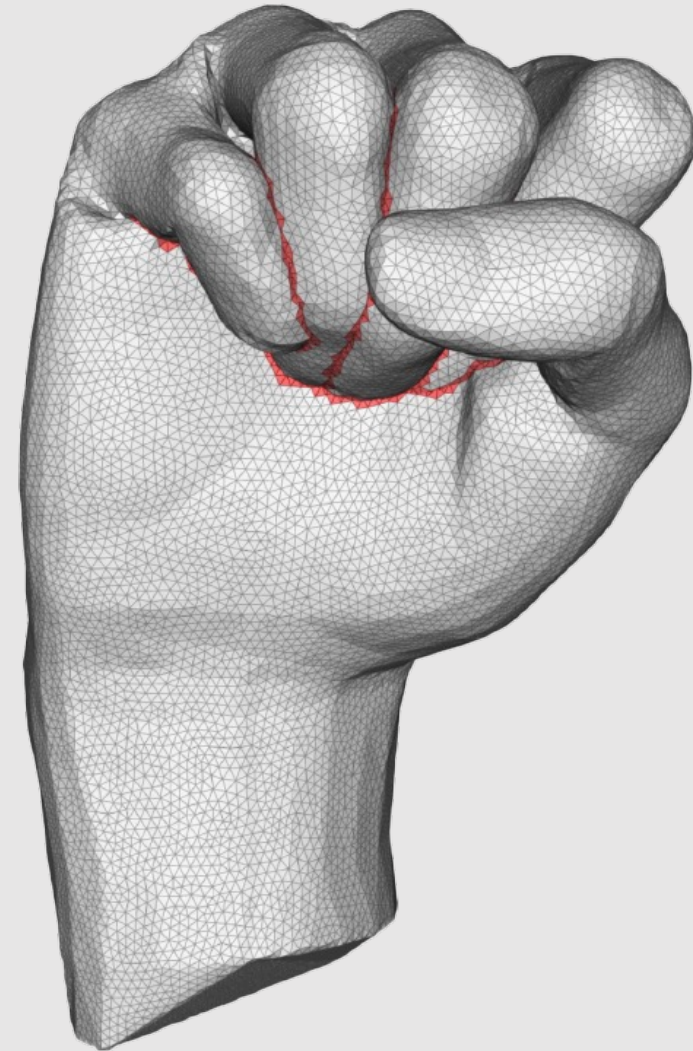
# Closest Point on a 3D Triangle Mesh

- Conceptually easy!
  - Loop over every triangle
  - Compute closest point to current triangle
  - Keep track of globally closest point
- Not practical in real world
  - Meshes have billions of triangles
  - Programs make thousands of geometric queries a second
- Will look at better solutions next time

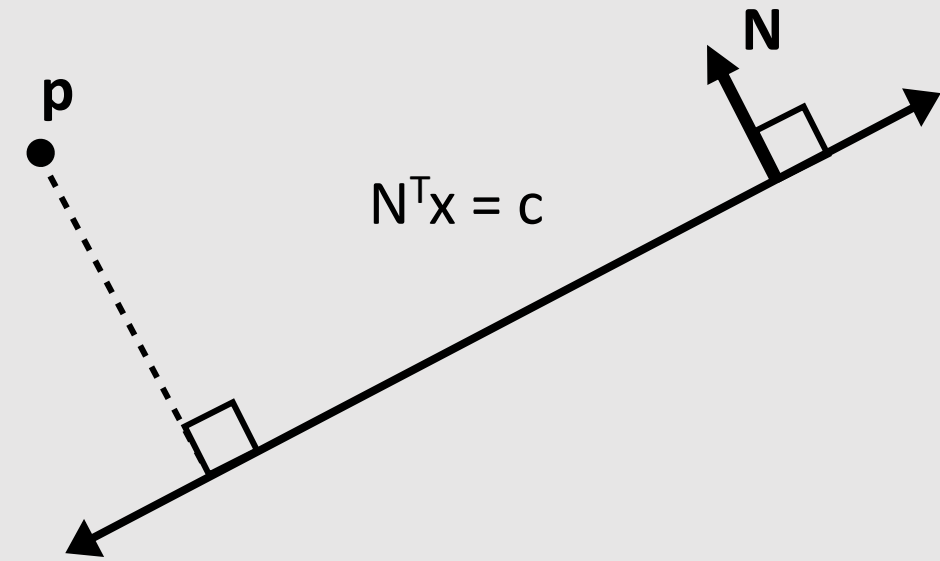


# Mesh-Mesh Intersections

- Sometimes when editing geometry, a mesh will intersect with itself
- Likewise, sometimes when animating geometry, meshes will collide
- How do we check for/prevent collisions?



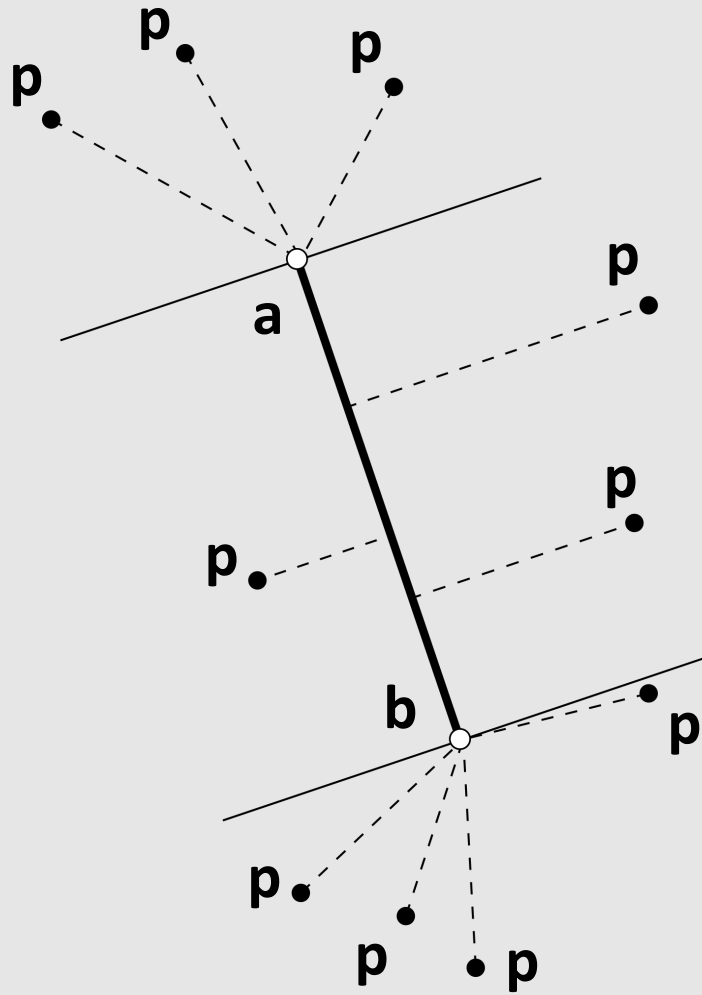
# Point-Line Intersection



Just plug point in

$$N^T p = c?$$

# Point-Line Segment Intersection



Check if adding distances equals net distance\*\*

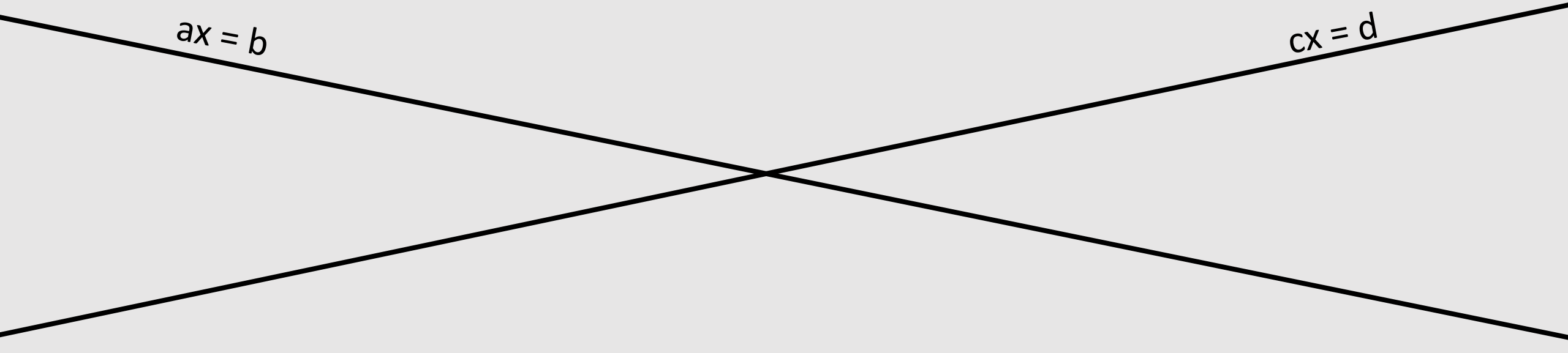
$$\text{dist}(a, p) + \text{dist}(p, b) = \text{dist}(a, b)$$

\*\*Potential numeric stability issues

# Line-Line Intersection

Two equations, two unknowns  
Solve a linear system

$$\begin{bmatrix} a_1 & a_2 \\ c_1 & c_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b \\ d \end{bmatrix}$$

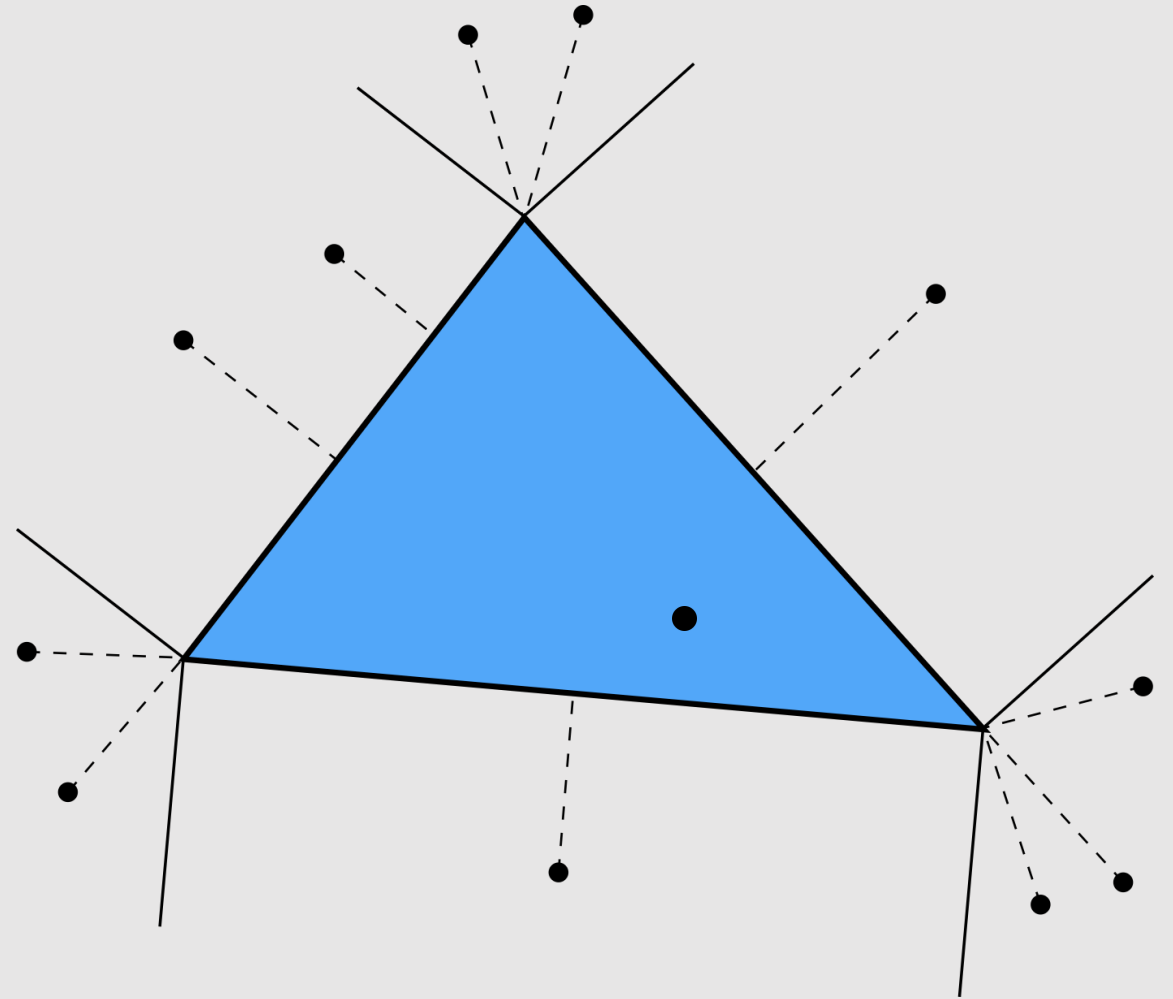


$ax = b$

$cx = d$

# Point-Triangle Intersection

You know this : )



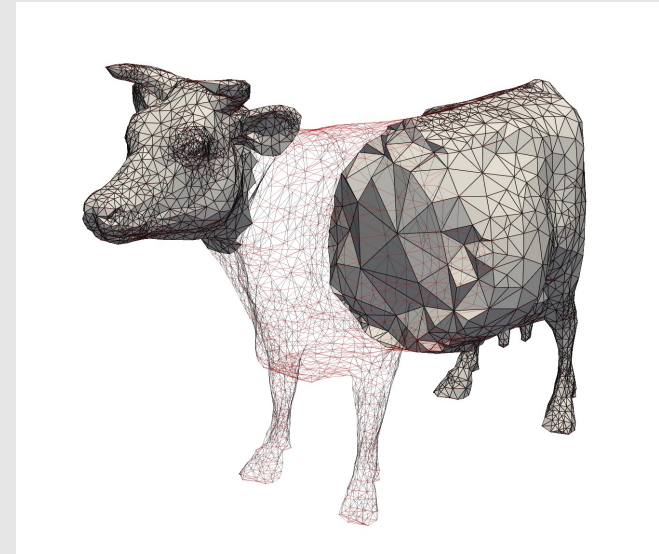
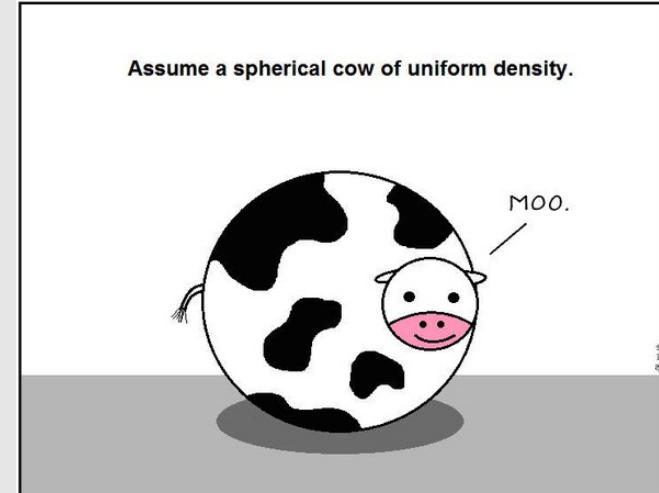
# Special Topics in A2: Geometric Representations



- **Marching Cubes**
- Signed Distance Fields
- NERFs

# Explicit vs Implicit

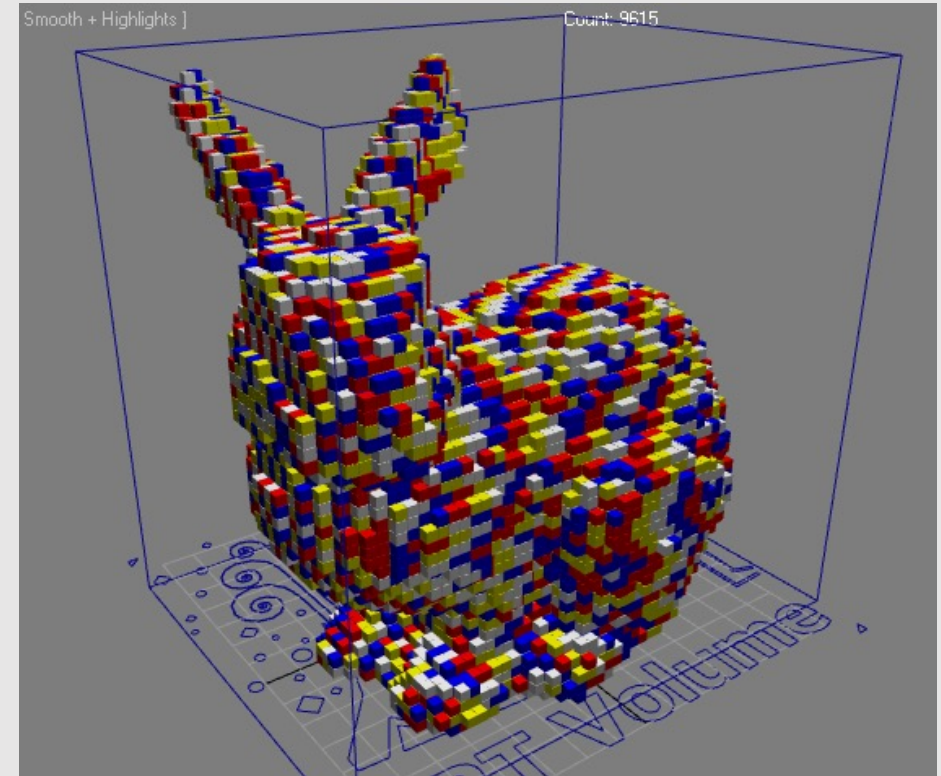
- Not one ideal geometry
- **Explicit:**
  - [+] finding any point on the surface
  - [-] finding if a given point lies on the surface
- **Implicit:**
  - [-] finding any point on the surface
  - [+] finding if a given point lies on the surface
- Pick the geometry best for the task at hand!



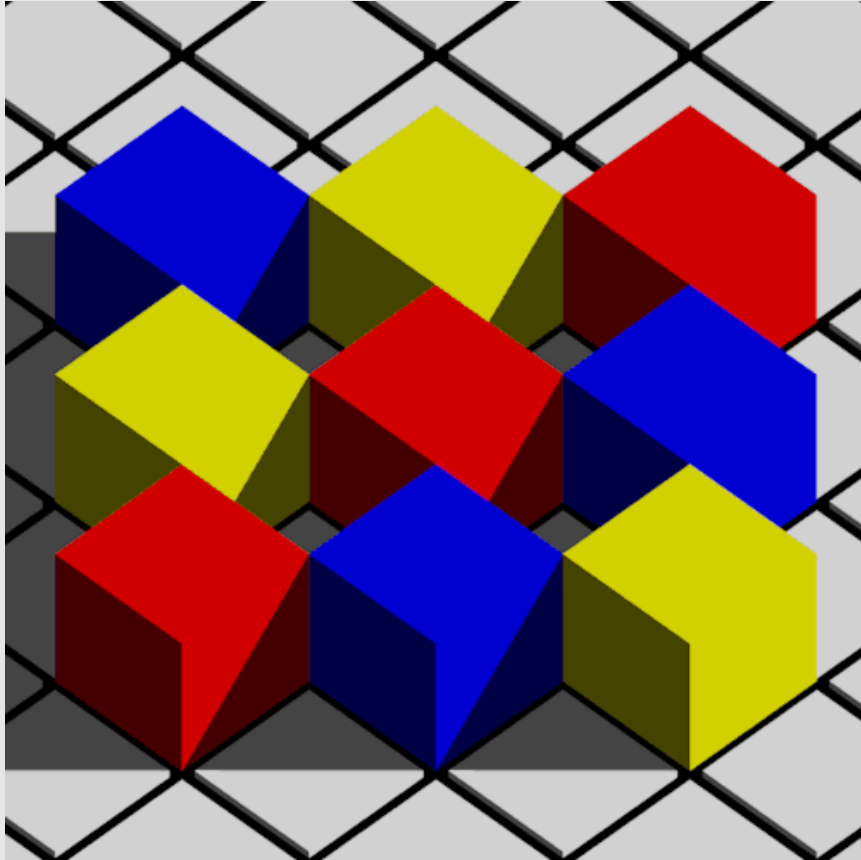
How do we convert implicit geometry to explicit geometry?

# Voxel Grid

- **Idea:** for an implicit function  $f(x, y, z)$ , sample points uniformly along the function's domain
  - Plot points where  $f(x, y, z) = 0$
  - Results in point cloud
- **Issue:** how many samples to take
  - More samples lead to higher precision, but are more expensive to compute
- **Issue:** does not include any info on connectivity
  - Difficult to interpolate data



# Marching Cubes

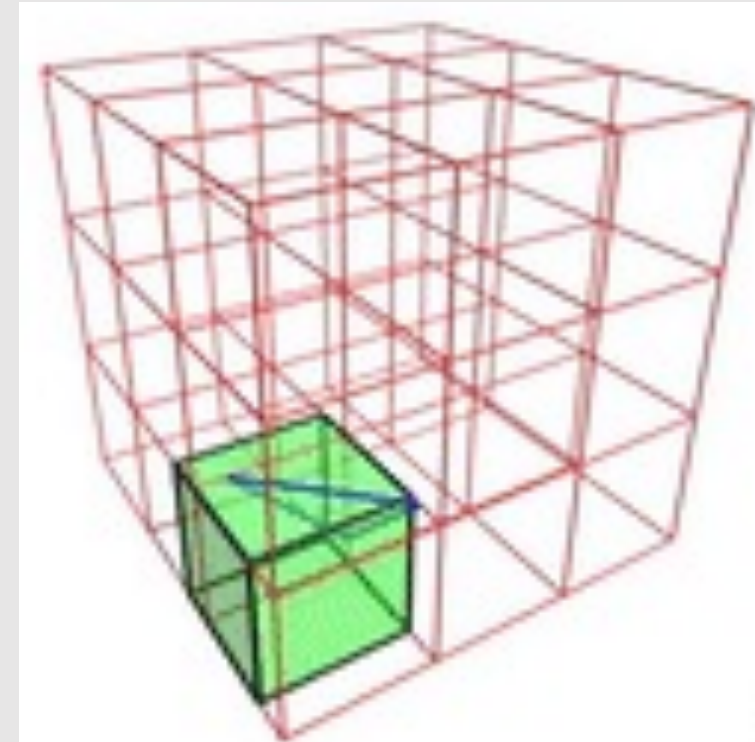


This is not the marching cubes algorithm.  
This is literally cubes marching.

- Marching cubes is an algorithm for converting implicit geometry to explicit
  - Adds both positional (vertices) and connectivity (edges)

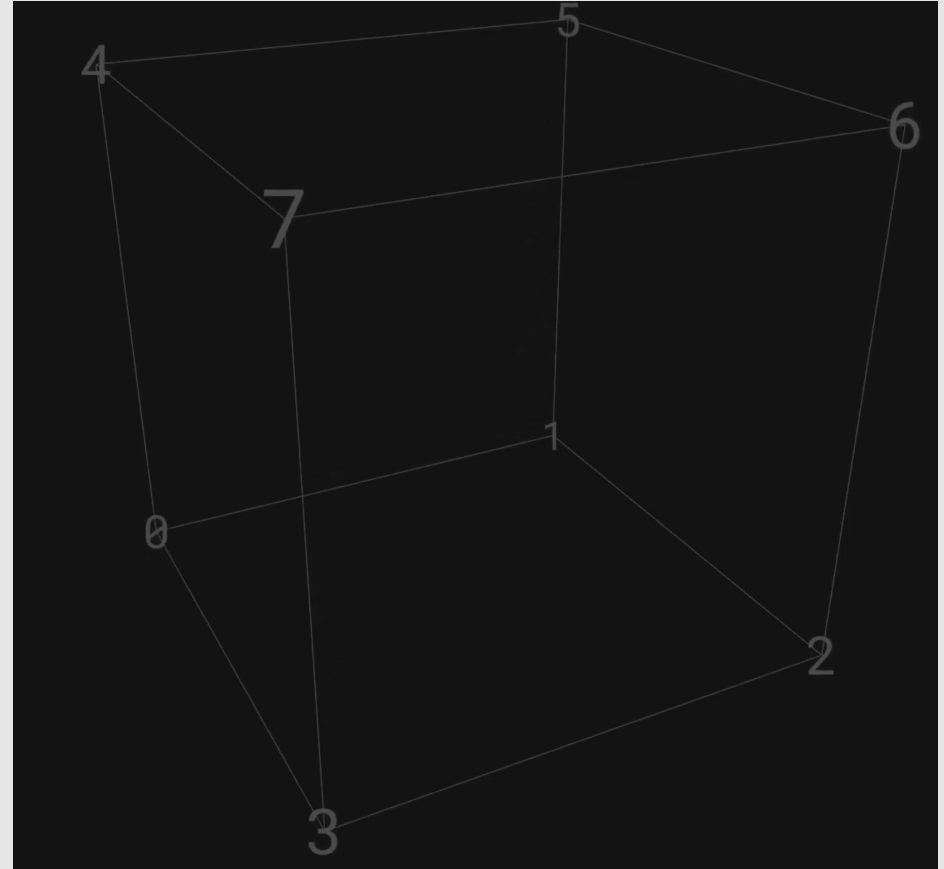
# Marching Cubes

- **Idea:** march a cube through the scene, checking if each of the vertices in the cube lie inside or outside the implicit function  $f(x, y, z)$ 
  - 8 vertices, 8 checks
  - Can encode as an 8-bit number
  - Generate geometry that makes sure inside vertices are enclosed by the geometry, and outside geometry are kept out
- **Issue:** how big of a cube to use
  - A smaller cube leads to finer details
  - A smaller cube also requires more samples



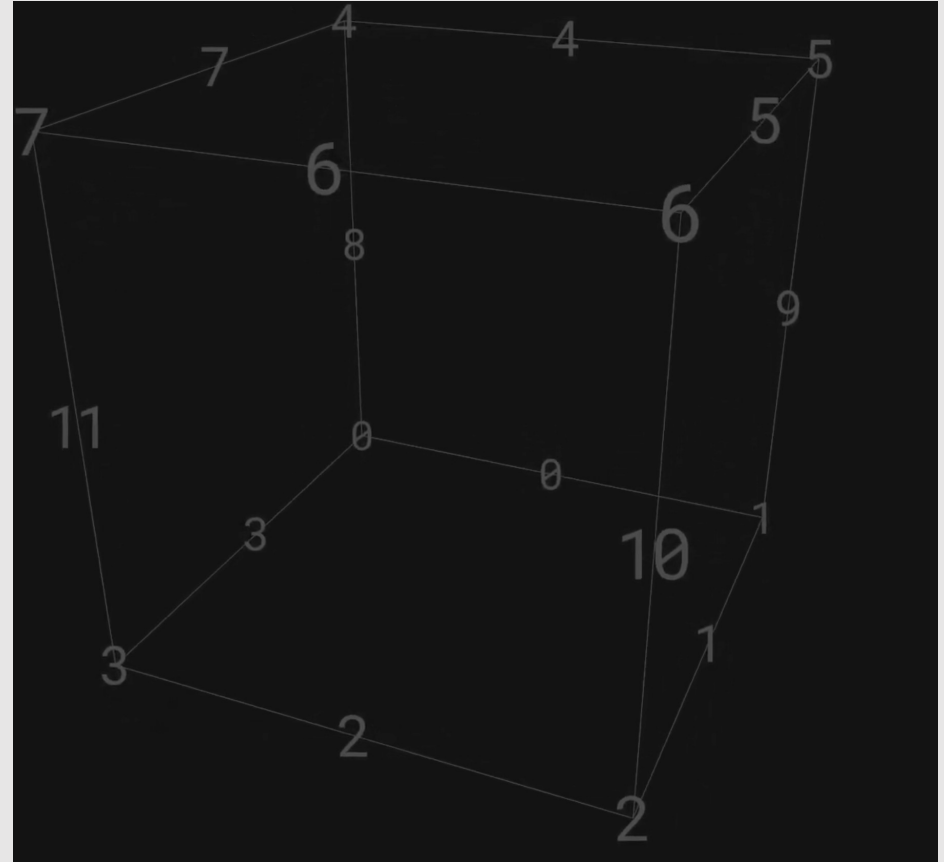
# Marching Cubes Vertices

- Each cube has 8 vertices
- Check if each vertex lies inside or outside the implicit function  $f(x, y, z)$ 
  - Can be encoded as an 8-bit number
    - 1 – inside
    - 0 - outside



# Marching Cubes Edges

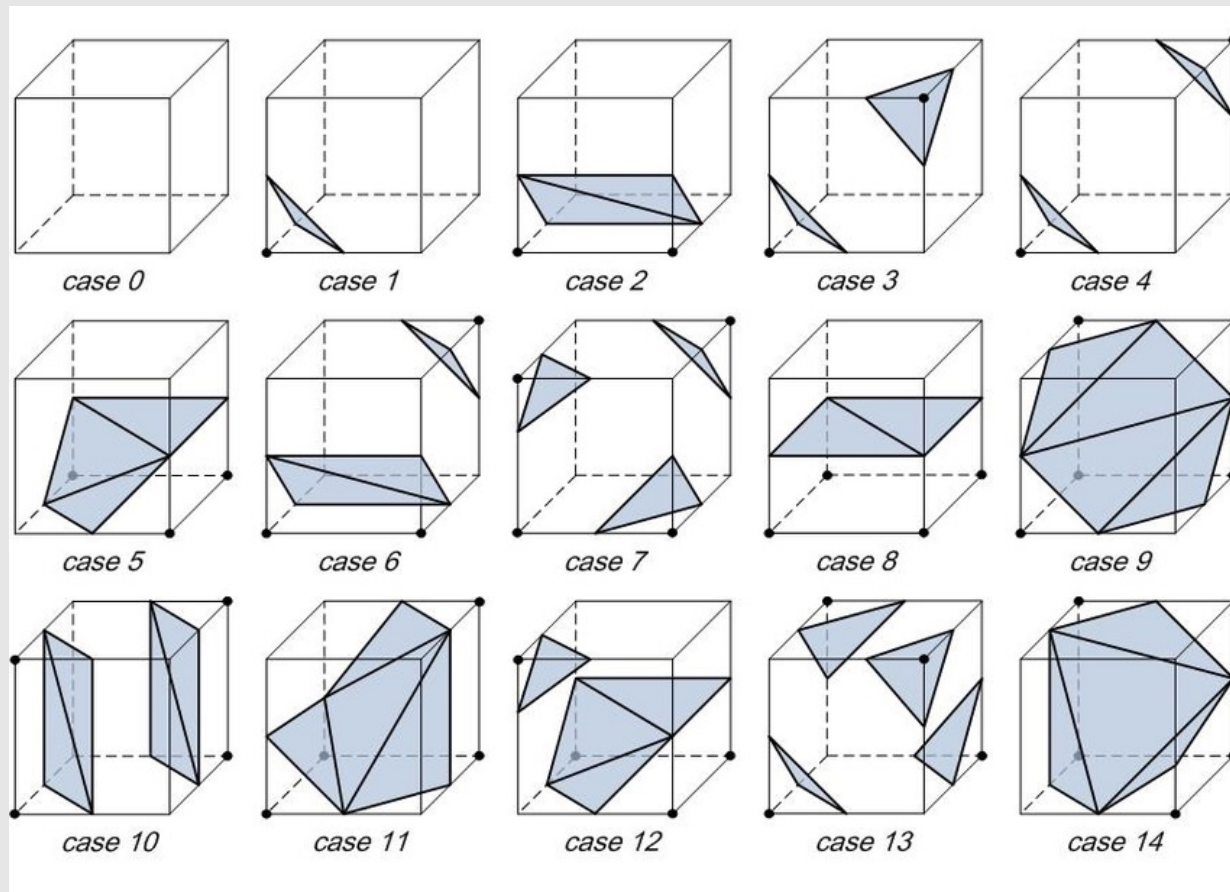
- Each cube has 12 edges
- Goal is to create geometry with vertices along the edges of the cube that enclose inside vertices and excludes outside vertices





# Marching Cubes Geometry

- $2^8 = 256$  possible configurations
- How do we know which one to use?



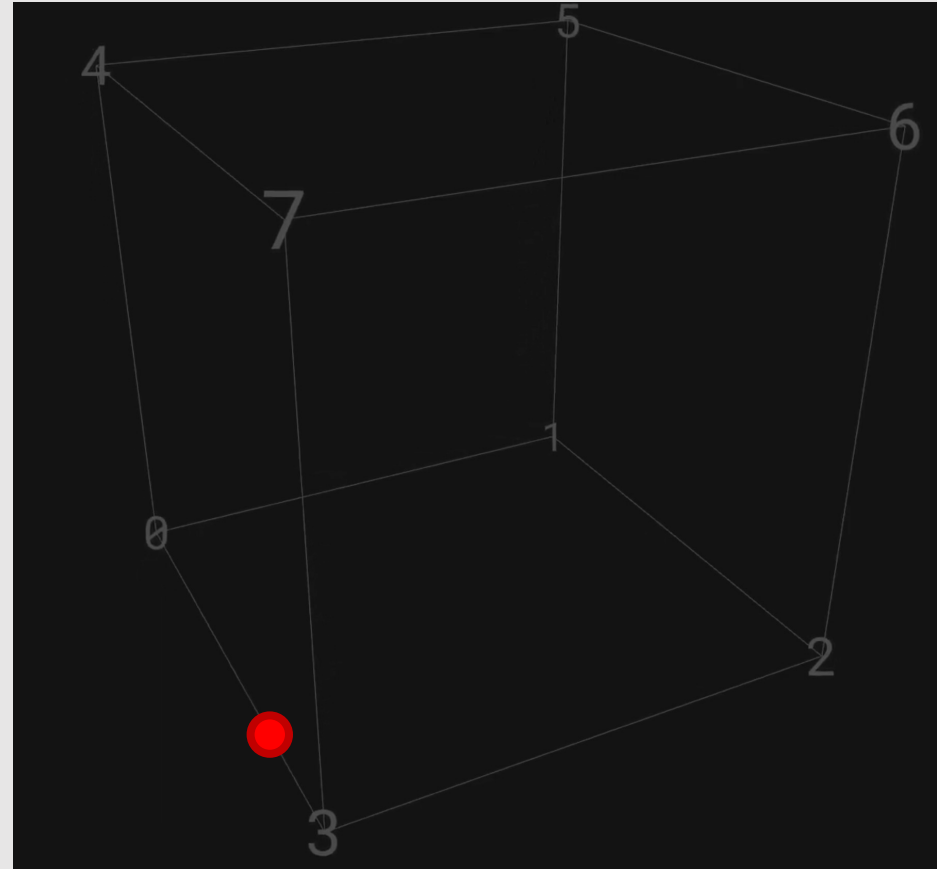
# Marching Cubes Lookup Table

```
// For each MC case, a list of triangles, specified as triples of edge indices, terminated by -1
const TriangleTable = [
  [ -1 ],
  [ 0, 3, 8, -1 ],
  [ 0, 9, 1, -1 ],
  [ 3, 8, 1, 1, 8, 9, -1 ],
  [ 2, 11, 3, -1 ],
  [ 8, 0, 11, 11, 0, 2, -1 ],
  [ 3, 2, 11, 1, 0, 9, -1 ],
  [ 11, 1, 2, 11, 9, 1, 11, 8, 9, -1 ],
  [ 1, 10, 2, -1 ],
  [ 0, 3, 8, 2, 1, 10, -1 ],
  [ 10, 2, 9, 9, 2, 0, -1 ],
  [ 8, 2, 3, 8, 10, 2, 8, 9, 10, -1 ],
  [ 11, 3, 10, 10, 3, 1, -1 ],
  [ 10, 0, 1, 10, 8, 0, 10, 11, 8, -1 ],
  [ 9, 3, 0, 9, 11, 3, 9, 10, 11, -1 ],
  [ 8, 9, 11, 11, 9, 10, -1 ],
  [ 4, 8, 7, -1 ],
  [ 7, 4, 3, 3, 4, 0, -1 ],
  [ 4, 8, 7, 0, 9, 1, -1 ],
  [ 1, 4, 9, 1, 7, 4, 1, 3, 7, -1 ],
  [ 8, 7, 4, 11, 3, 2, -1 ],
  [ 4, 11, 7, 4, 2, 11, 4, 0, 2, -1 ],
  [ 0, 9, 1, 8, 7, 4, 11, 3, 2, -1 ],
  [ 7, 4, 11, 11, 4, 2, 2, 4, 9, 2, 9, 1, -1 ],
  [ 4, 8, 7, 2, 1, 10, -1 ],
  [ 7, 4, 3, 3, 4, 0, 10, 2, 1, -1 ],
  [ 10, 2, 9, 9, 2, 0, 7, 4, 8, -1 ],
  [ 10, 2, 3, 10, 3, 4, 3, 7, 4, 9, 10, 4, -1 ],
  [ 1, 10, 3, 3, 10, 11, 4, 8, 7, -1 ],
  [ 10, 11, 1, 11, 7, 4, 1, 11, 4, 1, 4, 0, -1 ],
  [ 7, 4, 8, 9, 3, 0, 9, 11, 3, 9, 10, 11, -1 ],
  [ 7, 4, 11, 4, 9, 11, 9, 10, 11, -1 ],
  [ 9, 4, 5, -1 ],
  [ 9, 4, 5, 8, 0, 3, -1 ],
  [ 4, 5, 0, 0, 5, 1, -1 ],
  [ 5, 8, 4, 5, 3, 8, 5, 1, 3, -1 ],
```

- Just use a lookup table!

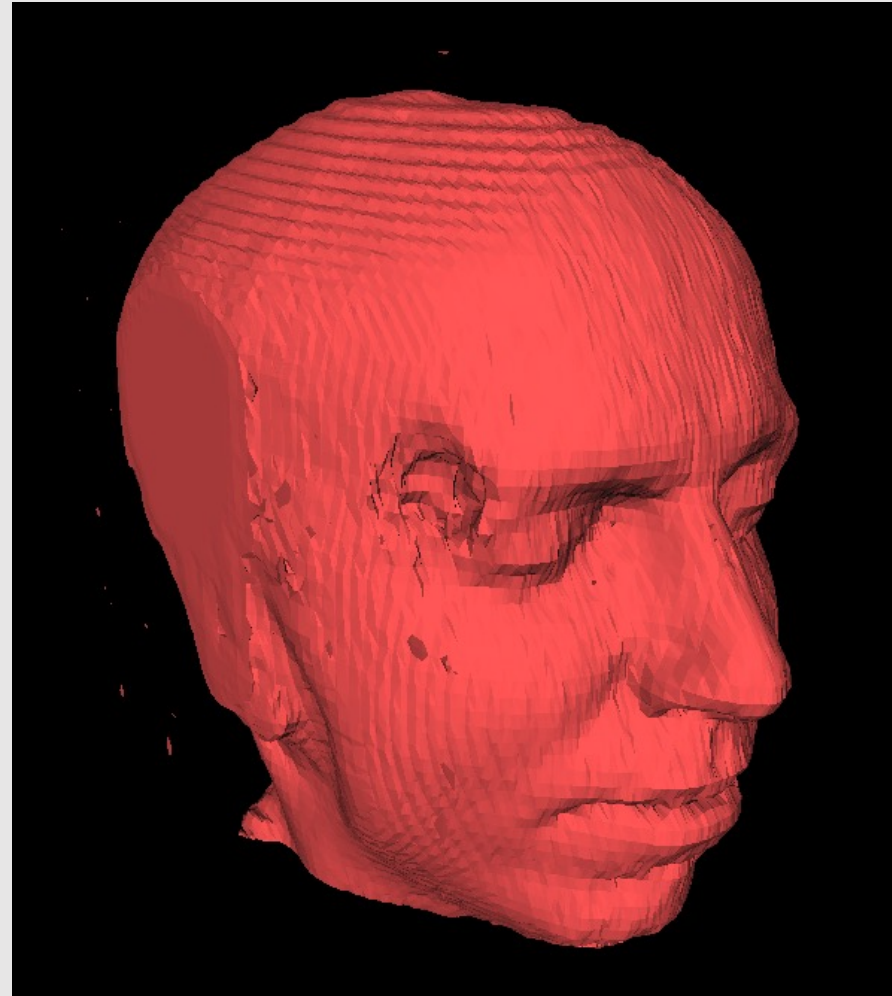
# Marching Cubes Linear Interpolation

- **Issue:** lookup table only tells us on what edges to place vertices and how to connect them
  - Does not tell us the specific location of vertices
- When placing vertices, can linearly interpolate them on the edges depending on the evaluated values on the cube vertices
- Example:
  - $f(x_0, y_0, z_0) = -0.75$
  - $f(x_3, y_3, z_3) = +0.25$ 
    - Vertex is placed  $\frac{1}{4}$  distance away from corner 3,  $\frac{3}{4}$  distance from corner 0



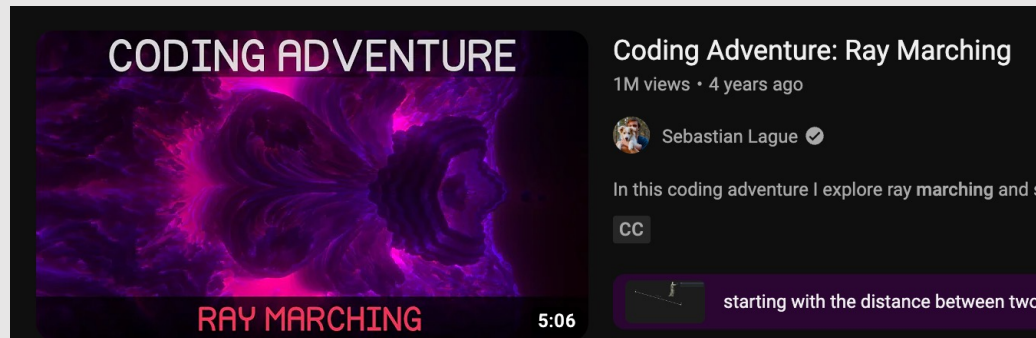
# Marching Cubes Examples

- **Issue:** very cube-like
  - Easy to see cube artifacts
- How to fix?
  - Run refinement
  - Run denoising
  - Run remeshing



# Marching Cubes Application

- Terrain generation
- Implicitly generate terrain with algebraic surfaces and noise
  - Convert to explicit mesh for easy rendering



- ~~Marching Cubes~~

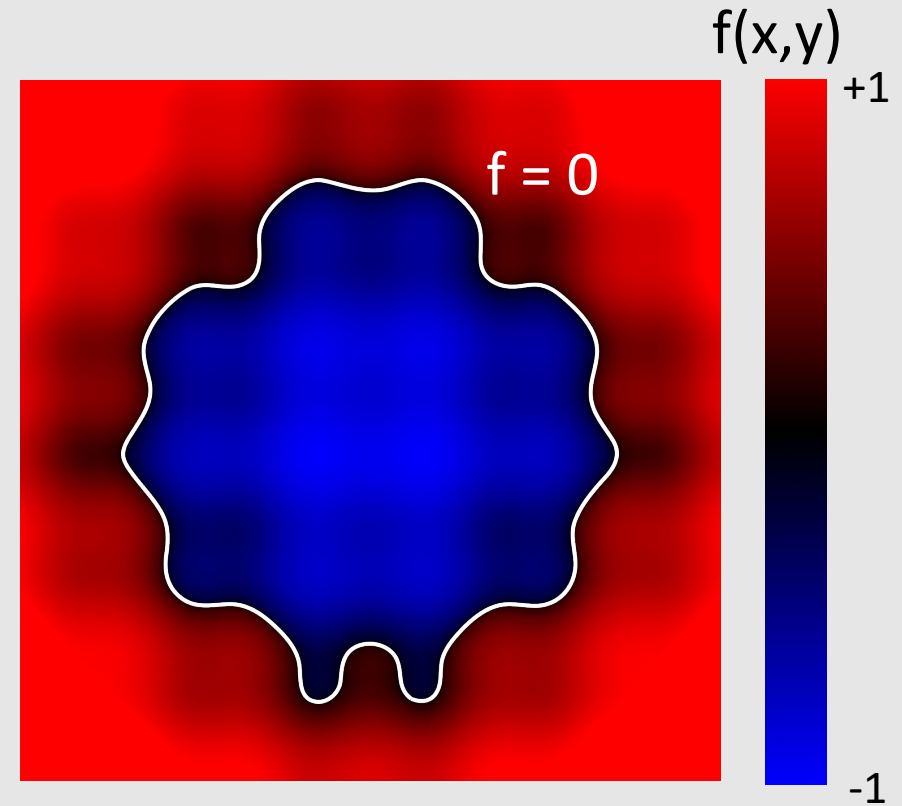
- Signed Distance Fields

- NERFs

How do we convert explicit geometry to implicit geometry?

# Signed Distance Fields

- Signed distance fields are implicit functions  $f(x, y, z)$  that tell us the sign (inside/outside) and the distance away from the boundary
  - Gradient  $\nabla f(x, y, z)$  makes finding the boundary easier
- SDFs make it easy to check where and how far a point is from a surface





# Converting Mesh To SDF

- **Idea:** SDF of a mesh should be proportional to the closest point  $p$  on a mesh to some query point  $q$
- **Issue:** how to accelerate computing the closest point on the mesh
  - Accelerated geometric queries

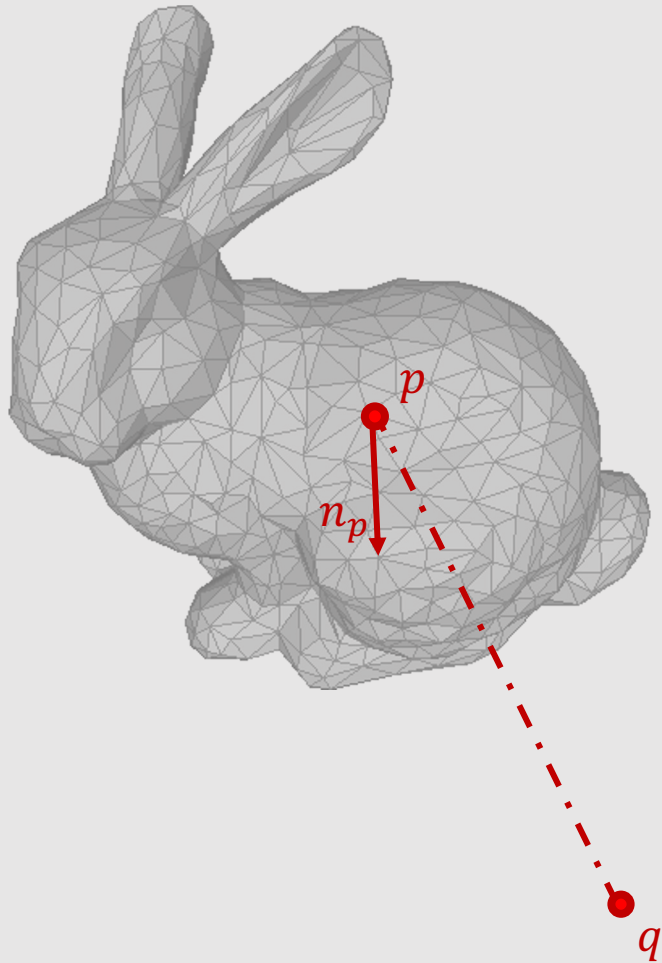
For a given query point  $q$ :

Compute the closest point on the mesh  $p$

Compute the normal  $n_p$  for  $p$

Project the vector  $(q-p)$  onto  $n_p$

# Converting Mesh To SDF



For a given query point  $q$ :

Compute the closest point on the mesh  $p$

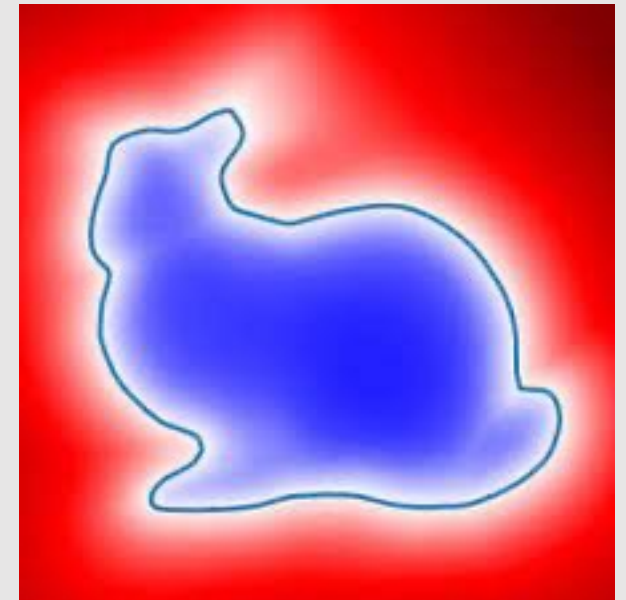
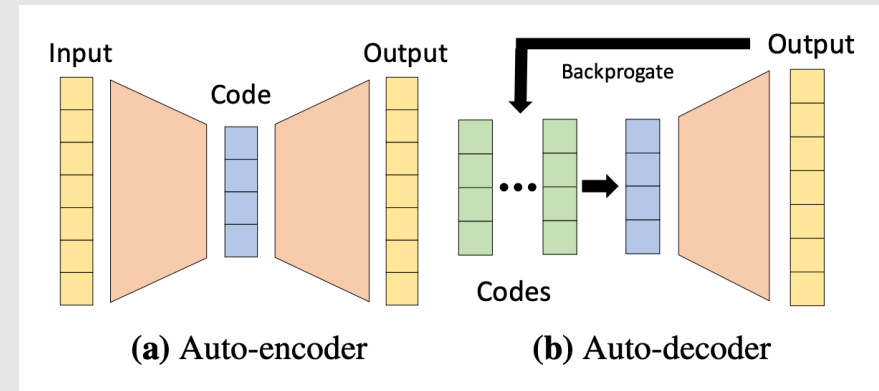
Compute the normal  $n_p$  for  $p$

Project the vector  $(q-p)$  onto  $n_p$

- **Distance** encoded by  $|q - p|$
- **Sign** encoded by  $(q - p) \cdot n_p$

# Neural SDFs

- Constructing a SDF can be difficult/expensive
- Throw a bunch of evaluated samples into an autoencoder
  - Learn a SDF representation of the data
- Neural net maps  $(x,y,z)$  to a signed distance
  - Can be used same way as an SDF



DeepSDF (2019) Park et al.

- ~~Marching Cubes~~

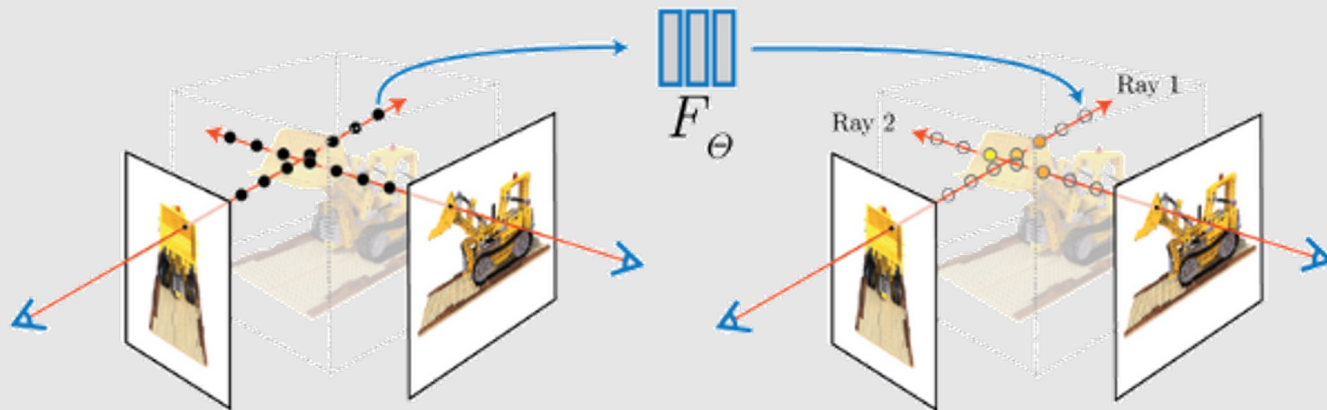
- ~~Signed Distance Fields~~

- **NERFs**

# Neural Radiance Fields

$$(x, y, z, \theta, \phi) \rightarrow \begin{array}{|c|} \hline \text{Neural Network} \\ \hline \end{array} \rightarrow (RGB\sigma)$$

$F_{\Theta}$

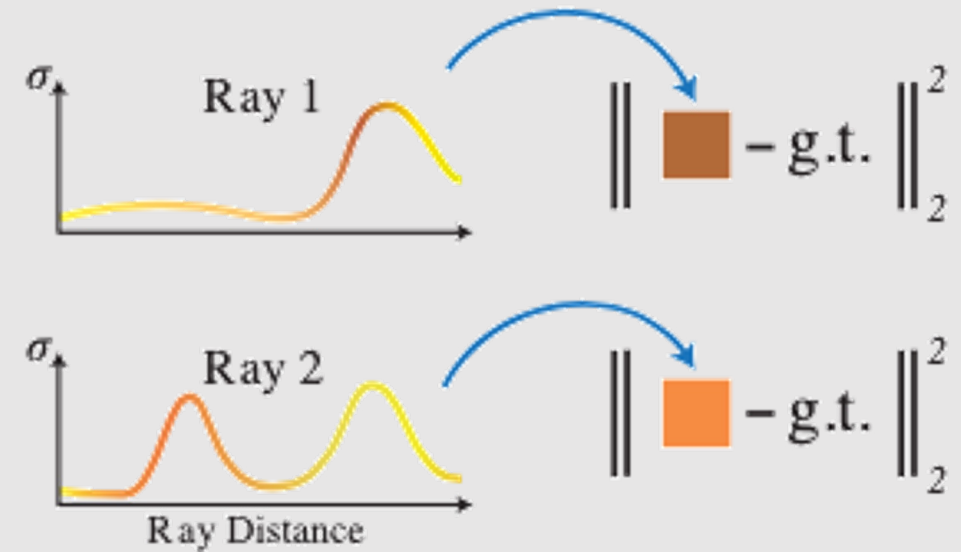


B. Mildenhall (2020)

- Train neural network  $F$  on multiple images
  - Training data:  $(x, y, z)$  of pixel + view angle + RGB
    - Depth of pixel must be known
  - Test data:  $(x, y, z)$  of requested pixel + view angle
    - Outputs RGB
    - What if we don't know the  $z$  we want?
      - Just grab the nearest  $z$
      - No different than 0-depth ray tracing
- To train properly, need multiple images each from a different view angle
  - **Key Assumption:** images must rotate about a fixed origin
    - Hence only 2 d.o.f with view angle

# Neural Radiance Fields

- We are building a field of radiance values
  - In case that wasn't clear : )
- Output RGB depends on depth
  - Think of it as a slice in a ray's direction
  - Optimize known (depth, RGB) pairs along a ray
    - Model learns to interpolate to unseen RGB values
- **Key Assumption:** lighting should remain constant between scenes
  - **Q: What would the distribution look like if not?**

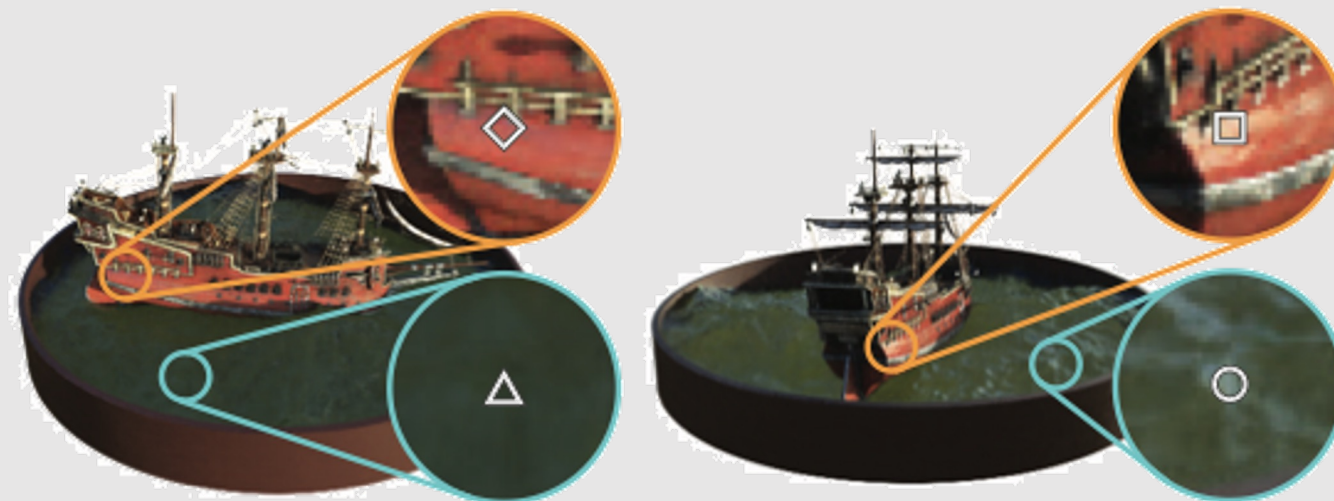


B. Mildenhall (2020)



# View-Dependent Appearances

- Some changes in lighting are inevitable
  - Mirror reflections
  - Glass refractions
  - Anisotropic materials
- **Idea:** treat these as normal pixels
  - View-dependent lighting will get baked into the model
    - Recall, when evaluating our model, we pass in the view direction
      - View-dependent lighting will only be present if same view angle passed into neural network



B. Mildenhall (2020)



# Extra Features



B. Mildenhall (2020)



- What can we use NERFs for?
  - Depth Maps
  - Image Relighting
  - Object insertion

<https://www.matthewtancik.com/nerf>

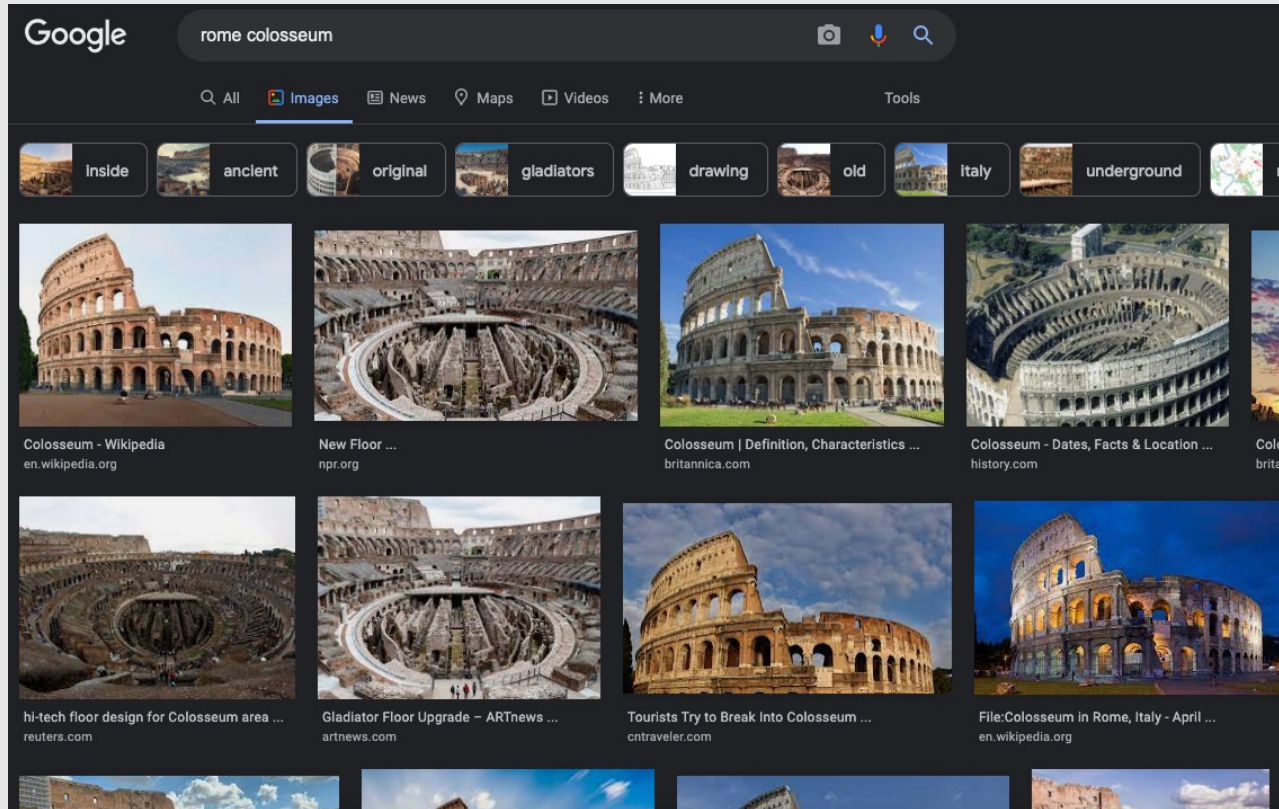


# NeRF In The Wild



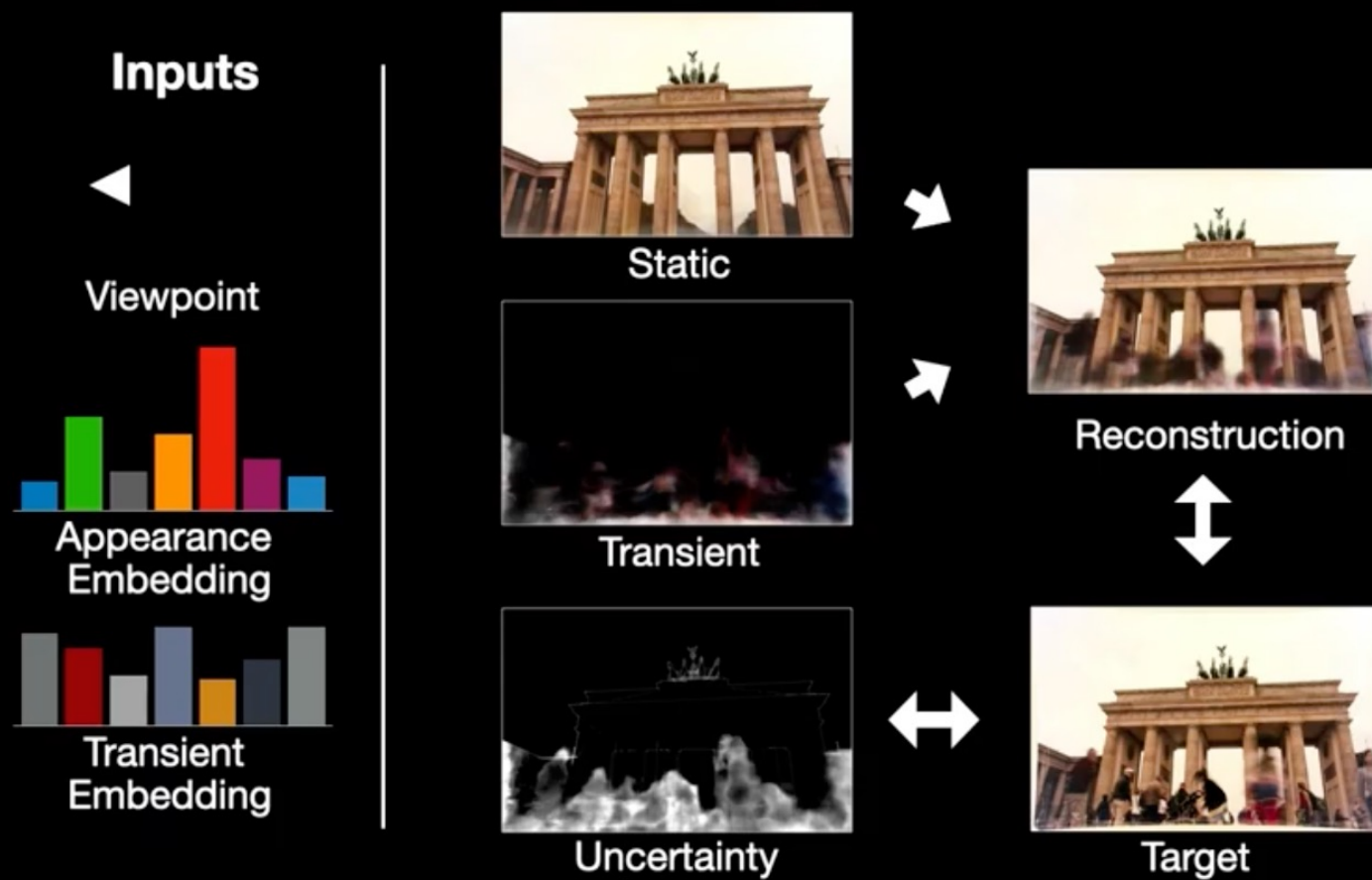
R. Martin-Brualla (2021) Google

# Image Databases



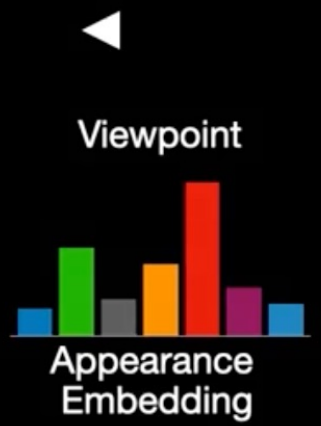
- **Key Idea:** photos for scene reconstruction are not always taken by the same user in the same conditions
  - Dealing with large database of image requires understanding of static properties and transient properties
    - Transient properties include:
      - People and object occlusions
      - Weather
      - Time of day

# Removing Transient Features



- Learn Appearance and transient embedding that minimizes reconstruction error

# Removing Transient Features



- Modify appearance embedding to change weather and lighting
  - Allows multiple photos to share same global properties
- Now we have an image set to perform NeRF on
  - Recall viewport given