

Introduction to Visualization and Computer Graphics DH2320
Prof. Dr. Tino Weinkauf

Introduction to Visualization and Computer Graphics

Grids and Interpolation

 In general, the content of all slides can be asked in the exam.

Some content is very likely to be asked in the exam.
 Marked with this sign:

very important

• Some content will not be asked. Marked with this sign:

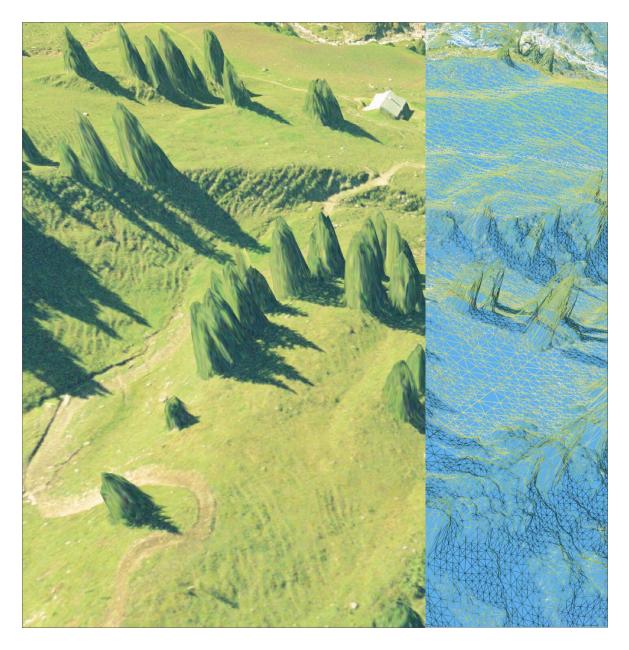




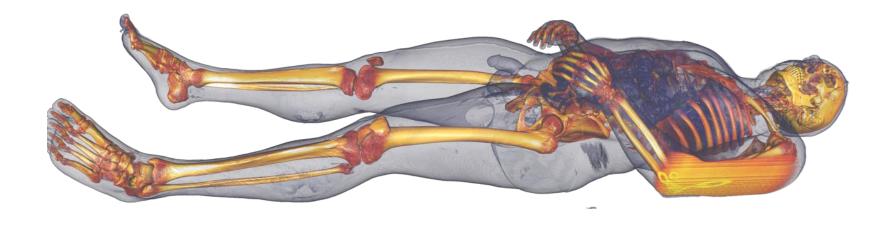
Introduction to Visualization and Computer Graphics DH2320
Prof. Dr. Tino Weinkauf

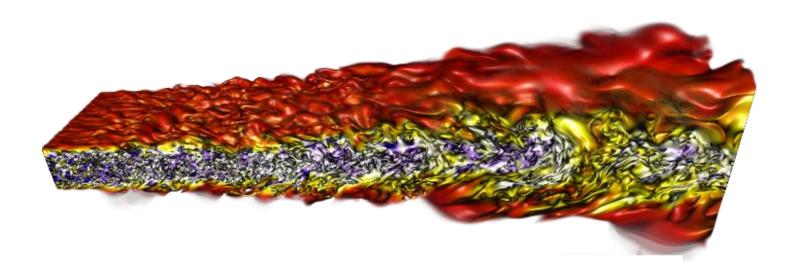
Grids and Interpolation

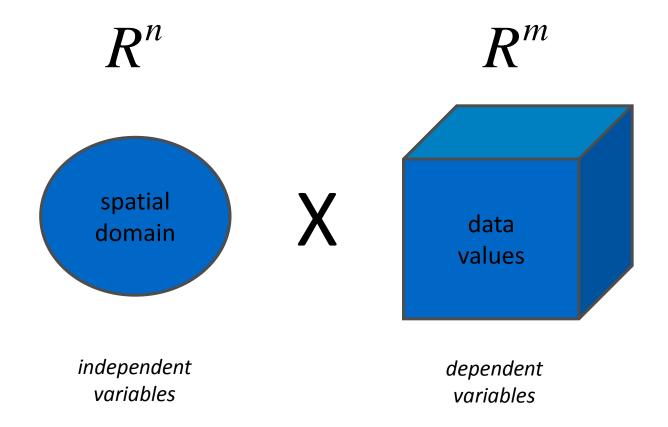
Structured Grids Unstructured Grids



Introduction to Visualization and Computer Graphics, Tino Weinkauf, KTH Stockholm

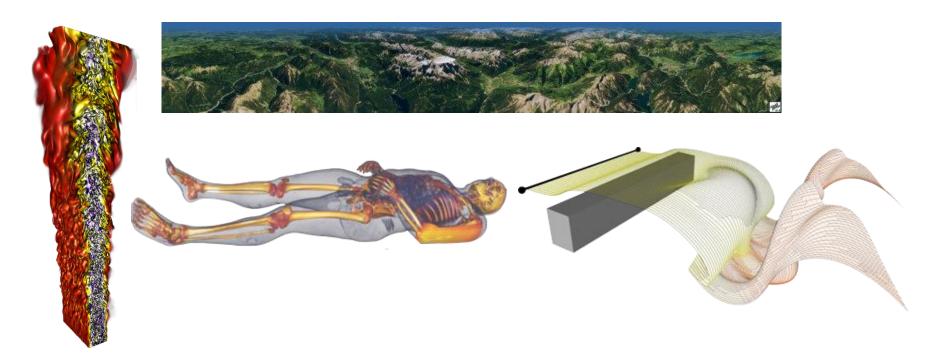






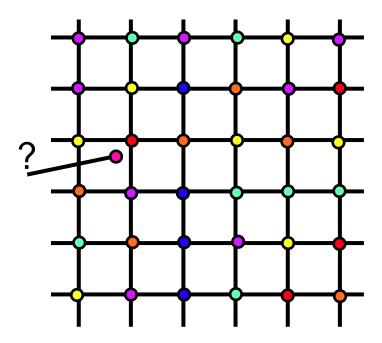
- In most cases, the visualization data represent a continuous real object, e.g., an oscillating membrane, a velocity field around a body, an organ, human tissue, etc.
 - This object lives in an n-dimensional space the domain
- Usually, the data is only given at a finite set of locations, or samples, in space and/or time
 - Remember imaging processes like numerical simulation and CTscanning, note similarity to pixel images
- We call this a discrete structure, or a discrete representation of a continuous object

- Discrete representations
 - We usually deal with the reconstruction of a continuous real object from a given discrete representation



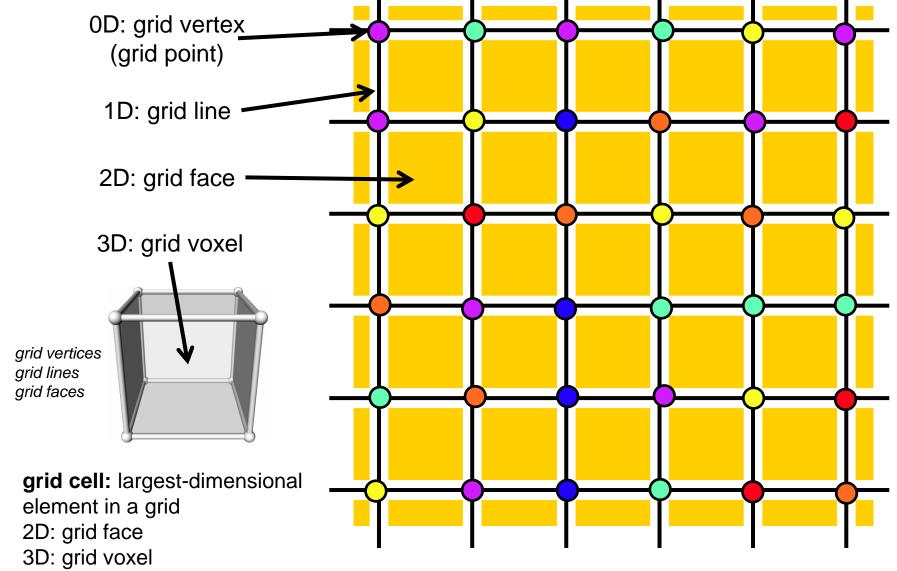
- Discrete structures consist of point samples
- Often, we build grids/meshes that connect neighboring samples

- Discrete representations
 - We usually deal with the reconstruction of a continuous real object from a given discrete representation



- Discrete structures consist of point samples
- Often, we build grids/meshes that connect neighboring samples

Grid terminology

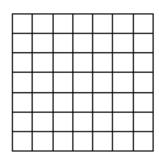


Data Connectivity

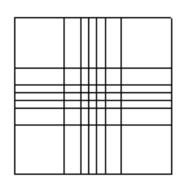
- There are different types of grids:
- Structured grids connectivity is implicitly given.
 - Block-structured grids
 combination of several structured grids
- Unstructured grids connectivity is explicitly given.
- Hybrid grids combination of different grid types

Structured grids

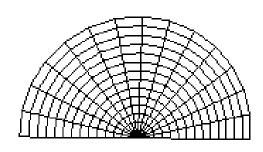
- "Structured" refers to the implicitly given connectivity between the grid vertices
- We distinguish different types of structured grids regarding the implicitly or explicitly given coordinate positions of the grid vertices



uniform grid implicitly given coordinates



rectilinear grid semi-implicitly given coordinates



curvilinear grid explicitly given coordinates

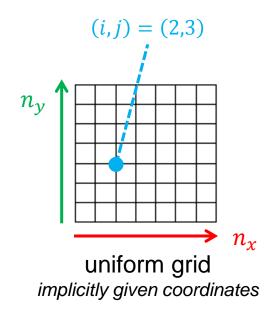
Structured grids

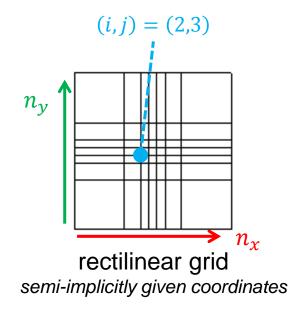
- Number of grid vertices: n_x , n_y , n_z
- We can address every grid vertex with an index tuple (i, j, k)

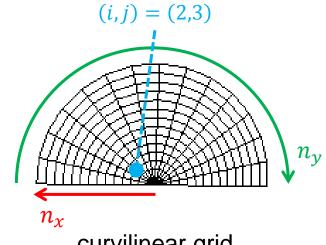
•
$$0 \le i < n_x$$

$$0 \le j < n_y$$

$$0 \le k < n_z$$







curvilinear grid explicitly given coordinates

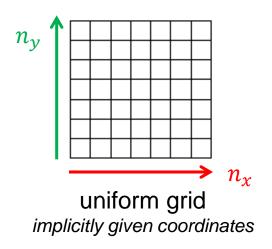
Structured grids

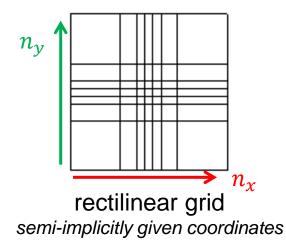
- Number of grid vertices: n_x , n_y , n_z
- We can address every grid cell with an index tuple (i, j, k)
 - $0 \le i < n_x 1$

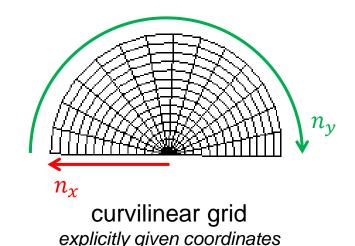
$$0 \le j < n_{v} - 1$$

$$0 \le k < n_z - 1$$

• Number of cells: $(n_x - 1) \times (n_y - 1) \times (n_z - 1)$

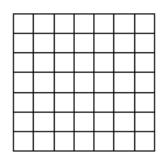


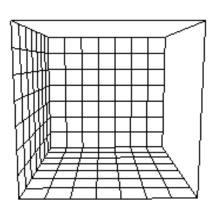




• Regular or uniform grids

- Cells are rectangles or rectangular cuboids of the same size
- All grid lines are parallel to the axes



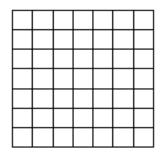


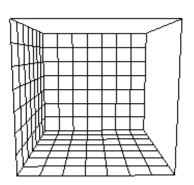
- To define a uniform grid, we need the following:
 - Bounding box: $(x_{min}, y_{min}, z_{min}) (x_{max}, y_{max}, z_{max})$
 - Number of grid vertices in each dimension: n_x , n_y , n_z
 - \rightarrow Cell size: d_x , d_y , d_z

Regular or uniform grids

- Well suited for image data (medical applications)
- Coordinate → cell is very simple and cheap
 - Global search is good enough; local search not required
- Coordinate of a grid vertex:

$$(i \cdot d_x, j \cdot d_y, k \cdot d_z)$$



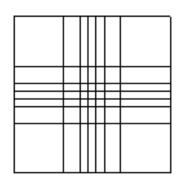


• Cartesian grid

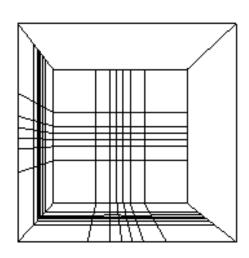
- Special case of a uniform grid: $d_x = d_y = d_z$
- Consists of squares (2D), cubes (3D)

Rectilinear grids

- Cells are rectangles of *different* sizes
- All grid lines are parallel to the axes

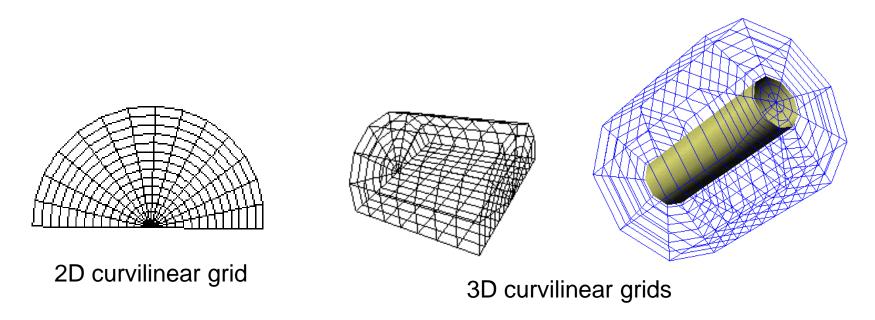


- Vertex locations are inferred from positions of grid lines for each dimension:
 - $XLoc = \{0.0, 1.5, 2.0, 5.0, ...\}$
 - YLoc = {-1.0, 0.3, 1.0, 2.0, ...}
 - ZLoc = {3.0, 3.5, 3.6, 4.1, ...}
- Coordinate → cell still quite simple



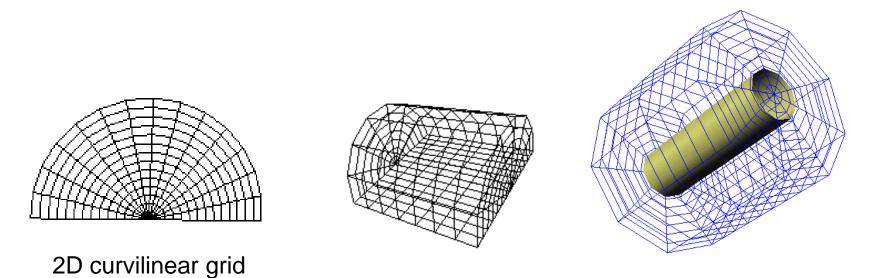
Curvilinear grids

- Vertex locations are explicitly given
 - $XYZLoc = \{(0.0, -1.0, 3.0), (1.5, 0.3, 3.5), (2.0, 1.0, 3.6), \ldots\}$
- Cells are quadrilaterals or cuboids
- Grid lines are not (necessarily) parallel to the axes



Curvilinear grids

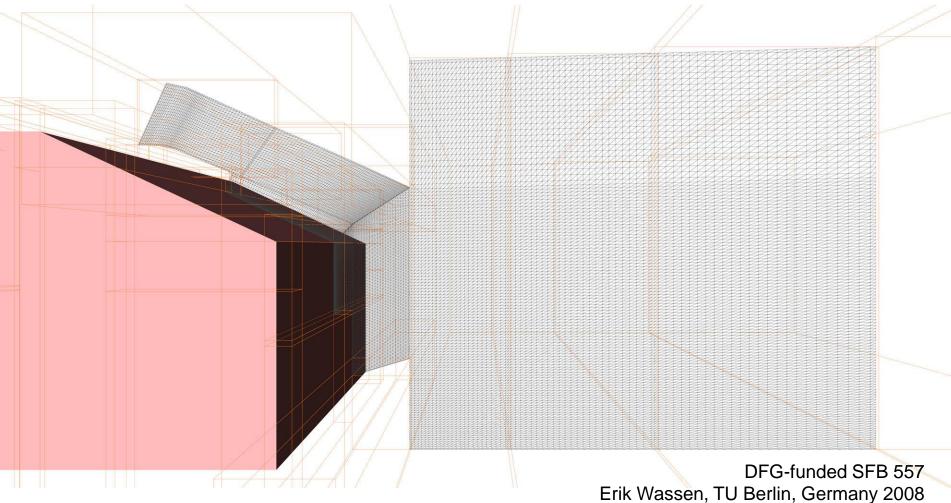
- Coordinate → cell:
 - Local search within last cell or its immediate neighbors
 - Global search via quadtree/octree



3D curvilinear grids

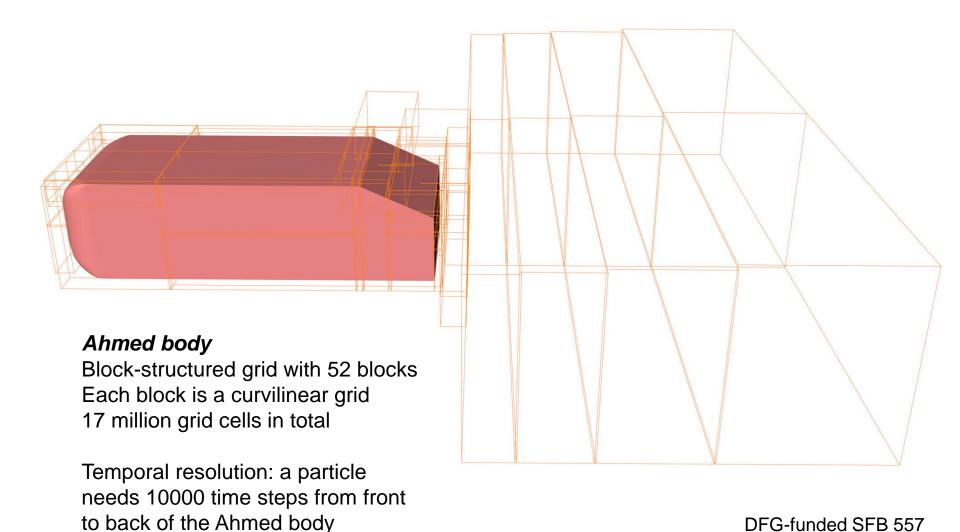
• Block-structured grids

combination of several structured grids

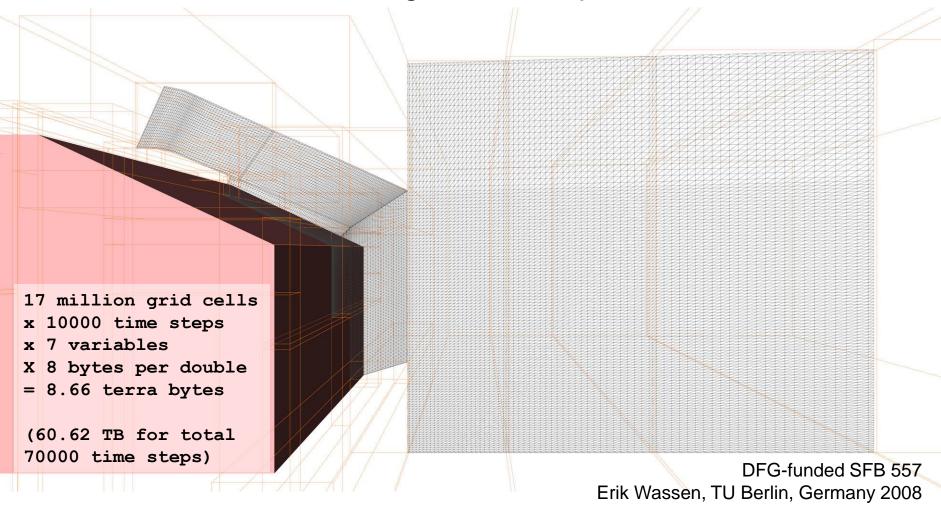


Erik Wassen, TU Berlin, Germany 2008

Demands on data storage, an example:

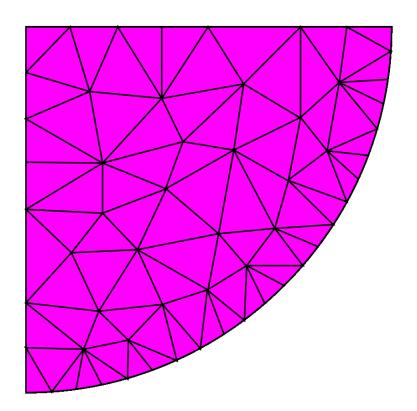


Demands on data storage, an example:

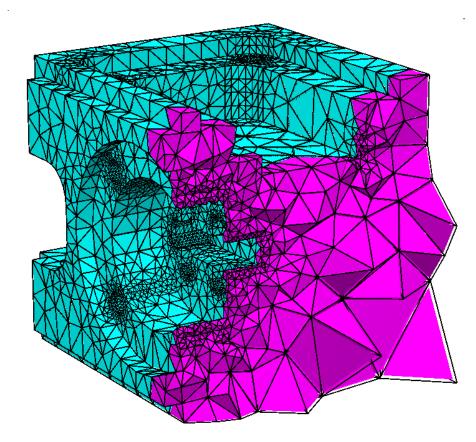


→ Do not save every time step, not every variable, and not every block.

• Unstructured grids



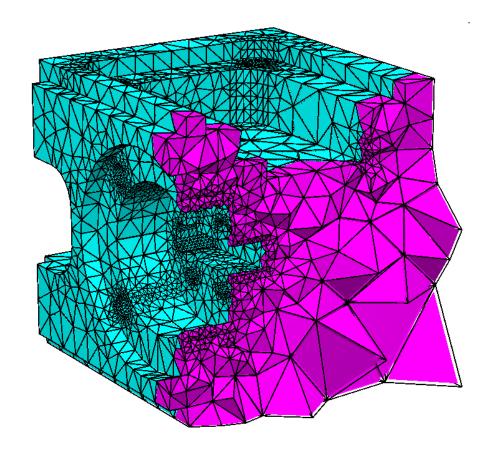
2D unstructured grid consisting of triangles



3D unstructured grid consisting of tetrahedra (from TetGen user manual)

Unstructured grids

- Vertex locations and connectivity explicitly given
- Linear interpolation within a triangle/tetrahedron using barycentric coordinates
- Coordinate triangle/tetra:
 - Local search within last triangle/tetra or its immediate neighbors
 - Global search via quadtree/octree



3D unstructured grid consisting of tetrahedra (from TetGen user manual)

How to store unstructured grids? Different requirements:

- Efficient storage
 - bytes per face / bytes per vertex
- Efficient access
 - of face / vertex properties (e.g., position)
- Efficient traversal
 - e.g., neighboring face, 1-ring of a vertex,...
- Requirements are competing

Face set

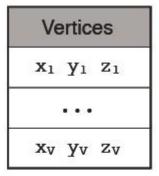
- Store faces
 - 3 positions
 - no connectivity ("match positions")
- Example: STL
 - very simple structure (too simple, unpractical!)
 - easily portable

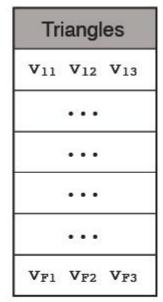
Triangles								
X11	y 11	Z11	X12	y 12	Z12	X13	у 13	Z13
X ₂₁	У 21	Z ₂₁	X22	У 22	Z ₂₂	X23	У 23	Z ₂₃
			• • •					
X _{F1}	y _{F1}	z_{F1}	XF2	YF2	z_{F2}	X _{F3}	Угз	z_{F3}

36 B/f = 72 B/vno connectivity!

Shared vertex

- vertex table stores positions
- triangle table stores indices into vertices
- No explicit connectivity
- Examples: OFF, OBJ, PLY
 - Quite simple and efficient
 - Enables efficient operations on static meshes

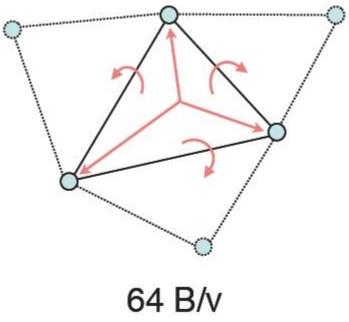




12 B/v + 12 B/f = 36 B/vno neighborhood info

Face-based connectivity

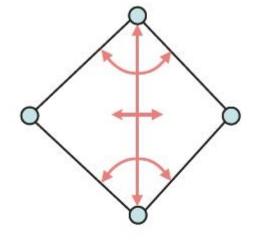
- vertices store
 - position
 - face reference
- faces store
 - 3 vertex references
 - references to 3 neighboring faces



64 B/v no edges!

Edge-based connectivity

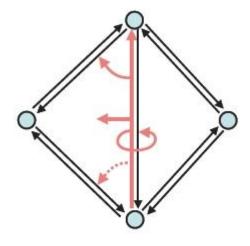
- vertex stores
 - position
 - reference to 1 edge
- edge stores references to
 - 2 vertices
 - 2 faces
 - 4 edges
- face stores
 - reference to 1 edge



120 B/v edge orientation?

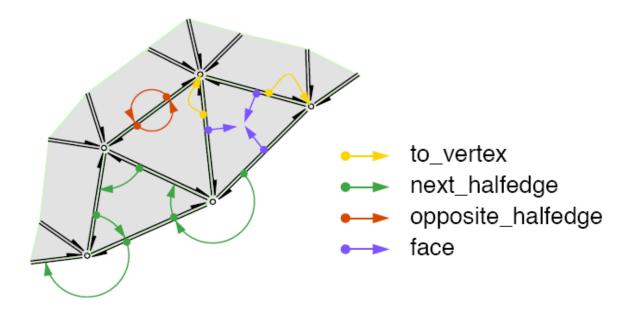
Half-edge based connectivity

- vertex stores
 - position
 - reference to 1 half-edge
- half-edge stored references to
 - 1 vertex
 - 1 face
 - 1, 2, or 3 half-edges
- face stores
 - reference to 1 half-edge



96 to 144 B/v no case distinctions during traversal

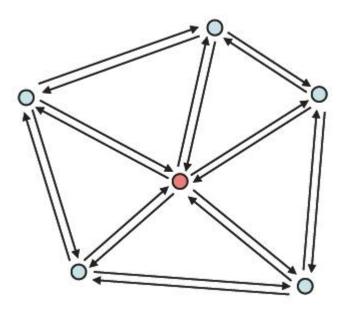
Half-edge based connectivity



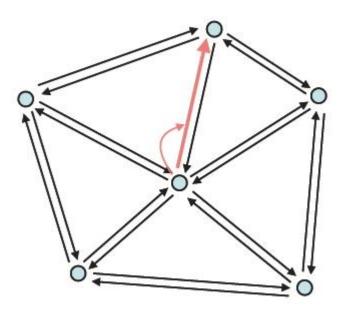
Half-edge based connectivity: Traversal

- Building blocks
 - Vertex to (outgoing) halfedge
 - half-edge to next (previous) halfedge
 - half-edge to neighboring half-edge
 - half-edge to face
 - half-edge to start (end) vertex
- Example: Traverse around vertex (1-ring)
 - enumerate vertices/faces/half-edges

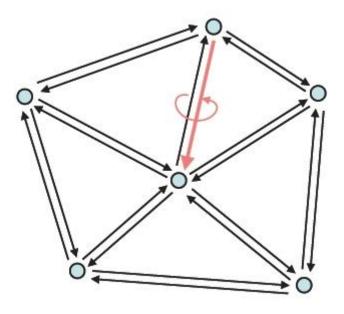
Start at vertex



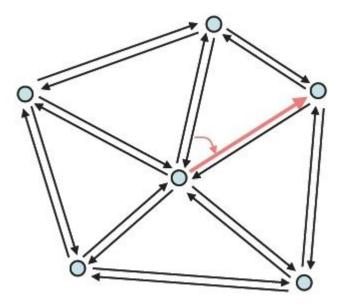
- Start at vertex
- Outgoing halfedge



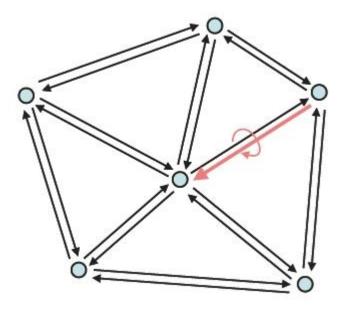
- Start at vertex
- Outgoing halfedge
- Opposite halfedge



- Start at vertex
- Outgoing halfedge
- Opposite halfedge
- Next half-egde

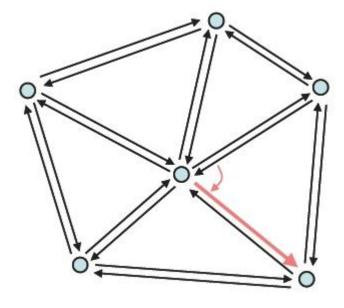


- Start at vertex
- Outgoing halfedge
- Opposite halfedge
- Next half-egde
- Opposite ...



- Start at vertex
- Outgoing halfedge
- Opposite halfedge
- Next half-egde
- Opposite ...
- Next ...

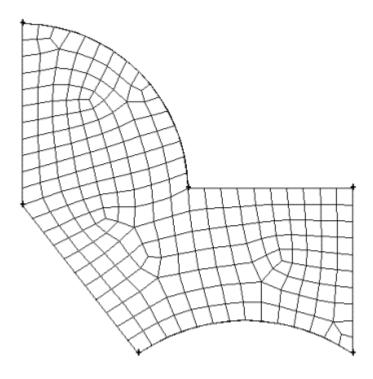
...



CGAL

- www.cgal.org
- Computational geometry
- Free for non-commercial use
- Open Mesh
 - www.openmesh.org
 - Mesh processing
 - Free, LGPL license
- gmu (gmu-lite)
 - proprietary, directed edges

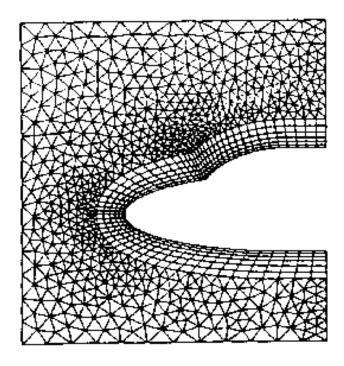
• Unstructured grids



2D unstructured grid consisting of quads

Source: https://www.sharcnet.ca/Software/Gambit/html/modeling_guide/mg0303.htm

- Hybrid grids
- combination of different grid types



2D hybrid grid

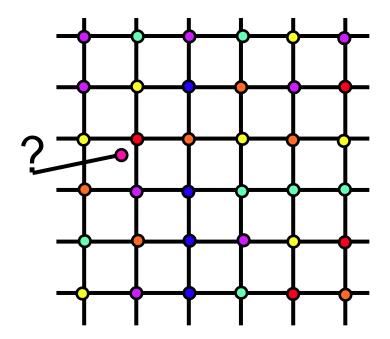


Introduction to Visualization and Computer Graphics DH2320
Prof. Dr. Tino Weinkauf

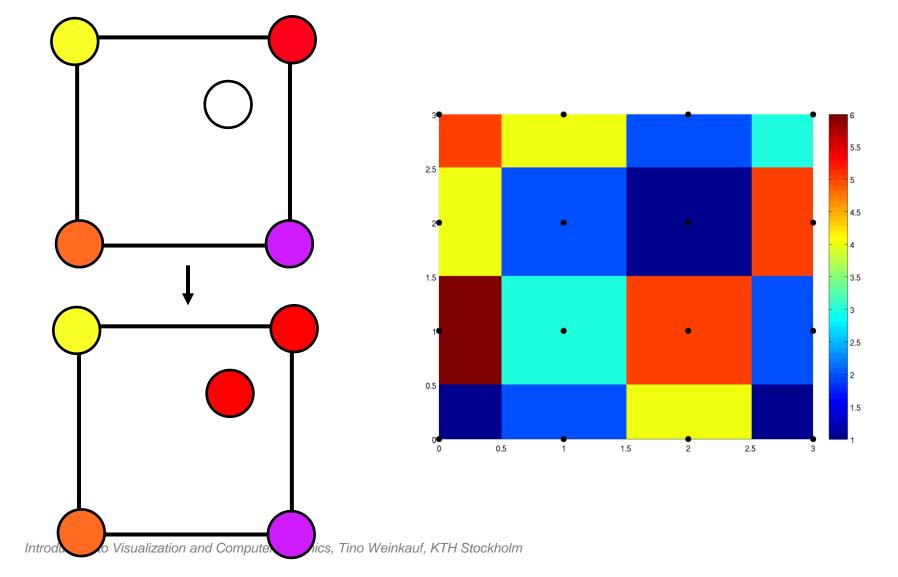
Grids and Interpolation

Linear, Bilinear, Trilinear Interpolation in Structured Grids
Gradients
Linear Interpolation in Unstructured Grids

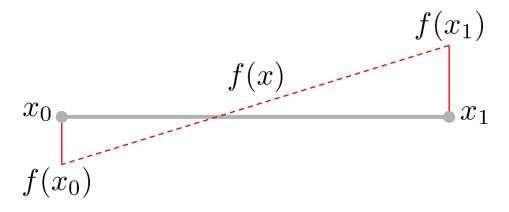
- A grid consists of a finite number of samples
 - The continuous signal is known only at a few points (data points)
 - In general, data is needed in between these points
- By interpolation we obtain a representation that matches the function at the data points
 - Reconstruction at any other point possible



- Simplest approach: Nearest-Neighbor Interpolation
 - Assign the value of the nearest grid point to the sample.



- Linear Interpolation (in 1D domain)
 - Domain points x, scalar function f(x)



General:

$$f(x) = \frac{x_1 - x_0}{x_1 - x_0} f(x_0) + \frac{x - x_0}{x_1 - x_0} f(x_1)$$
 $x \in [x_0, x_1]$

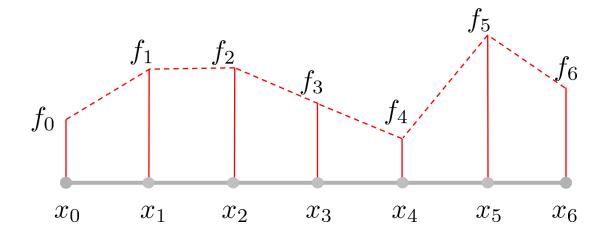
Special Case:

$$f(x) = (1 - x) f(0) + x f(1) \qquad x \in [0, 1]$$

$$= \begin{bmatrix} (1 - x) & x \end{bmatrix} \begin{pmatrix} f(0) \\ f(1) \end{pmatrix} = \begin{bmatrix} 1 & x \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -1 & 1 \end{bmatrix} \begin{pmatrix} f(0) \\ f(1) \end{pmatrix}$$

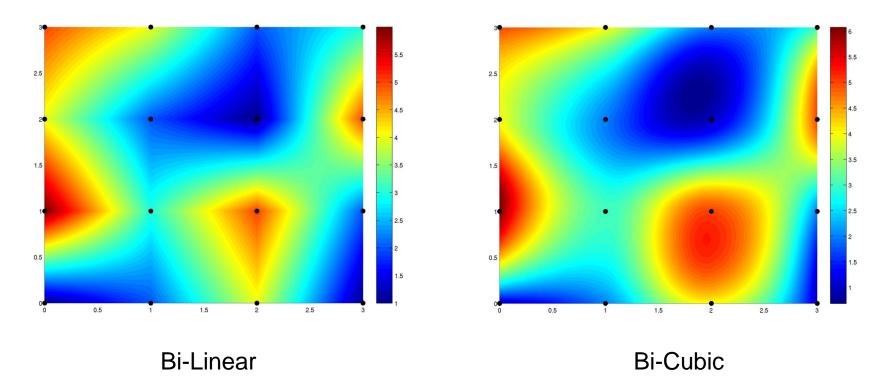
Basis Coefficients

- Linear Interpolation (in 1D domain)
 - Sample values $f_i := f(x_i)$



- C⁰ Continuity (discontinuous first derivative)
 - Use higher order interpolation for smoother transition, e.g.,
 cubic interpolation

Interpolation in 2D, 3D, 4D, ...

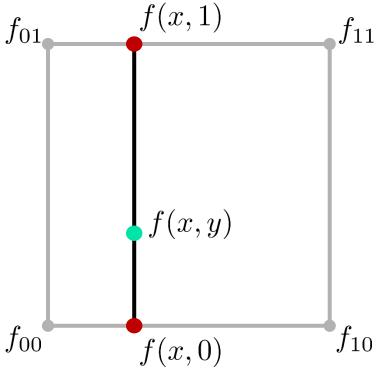


Tensor Product Interpolation

 Perform linear / cubic ... interpolation in each x,y,z ... direction separately

Bilinear Interpolation

2D, "bi-linear"



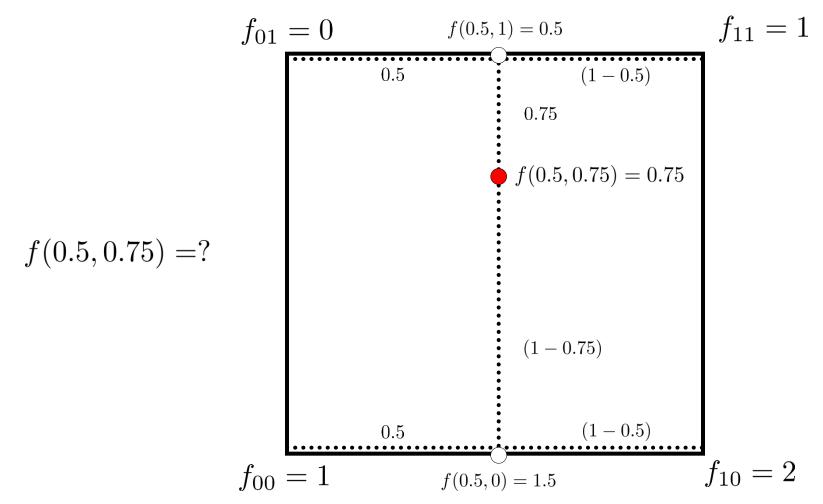
$$f(x,y) = (1-x)(1-y) f_{00} + x (1-y) f_{10} + (1-x) y f_{01} + x y f_{11}$$

$$= (1 - y) ((1 - x) f_{00} + x f_{10}) + y ((1 - x) f_{01} + x f_{11})$$

"interpolate twice in x direction and then once in y direction"

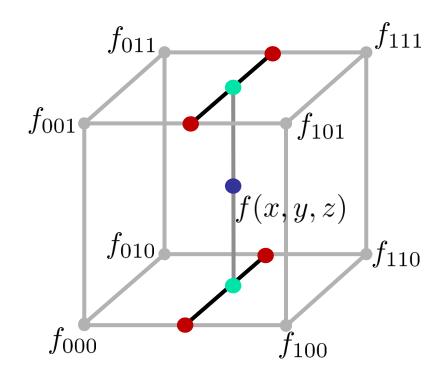
• Example: Bi-linear interpolation in a 2D cell

Repeated linear interpolation



Trilinear Interpolation

$$f(x, y, z) = \sum_{k=0}^{p} \sum_{j=0}^{m} \sum_{i=0}^{n} b_i(x)b_j(y)b_k(z) f_{ijk}$$



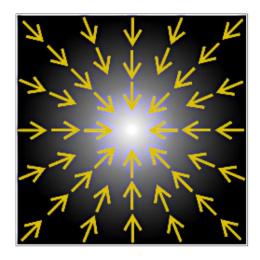
"interpolate four times in x direction, twice in y direction, and once in z direction"

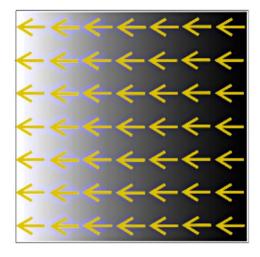
Function Derivative Estimation

- Called Gradients for multidimensional functions
- Have a lot of important applications (e.g., normal for volume rendering, critical point classification for vector field topology ...)

$$\nabla f(x,y,z) = \begin{pmatrix} \frac{\partial}{\partial x} \\ \frac{\partial}{\partial y} \\ \frac{\partial}{\partial z} \end{pmatrix} \ f(x,y,z) = \begin{pmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \\ \frac{\partial f}{\partial z} \end{pmatrix} \qquad \text{``vector of partial derivatives''}$$

Describes direction of steepest ascend





- Two ways to estimate gradients:
 - Direct derivation of interpolation formula
 - Finite differences schemes

• Field Function Derivatives, Bi-Linear

$$f(x,y) = \begin{bmatrix} (1-x) & x \end{bmatrix} \begin{bmatrix} f_{00} & f_{01} \\ f_{10} & f_{11} \end{bmatrix} \begin{bmatrix} (1-y) \\ y \end{bmatrix} \longrightarrow$$

derive this interpolation formula

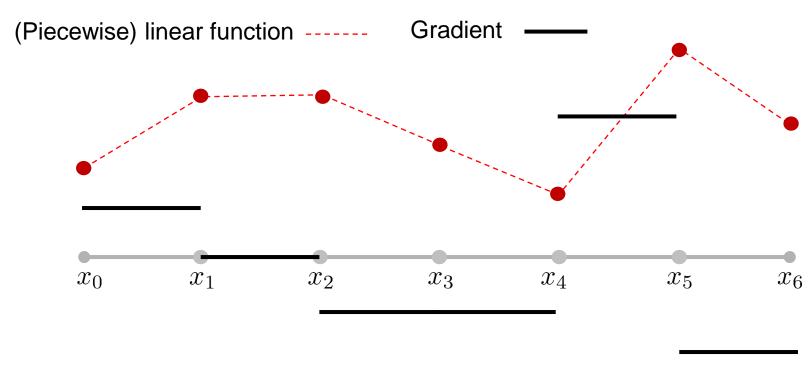
$$\frac{\partial f(x,y)}{\partial x} = \begin{bmatrix} -1 & 1 \end{bmatrix} \begin{bmatrix} f_{00} & f_{01} \\ f_{10} & f_{11} \end{bmatrix} \begin{bmatrix} (1-y) \\ y \end{bmatrix}$$
$$= (f_{10} - f_{00}) (1-y) + (f_{11} - f_{01}) y$$

"constant in x direction"

$$\frac{\partial f(x,y)}{\partial y} = \left[(1-x) \ x \right] \left[\begin{array}{c} f_{00} \ f_{01} \\ f_{10} \ f_{11} \end{array} \right] \left[\begin{array}{c} -1 \\ 1 \end{array} \right]$$
$$= (f_{01} - f_{00}) (1-x) + (f_{11} - f_{10}) x$$

"constant in y direction"

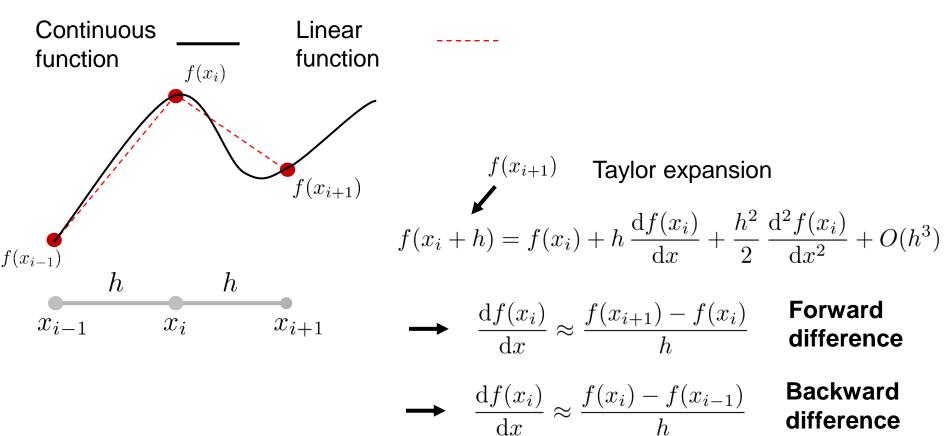
 Problem of exact linear function differentiation: discontinuous gradients



- Solution:
 - Use higher order interpolation scheme (cubic)
 - Use finite difference estimation

Finite Differences Schemes

Apply Taylor series expansion around samples



Finite Differences Schemes

$$f(x_{i+1}) = f(x_i) + h \frac{\mathrm{d}f(x_i)}{\mathrm{d}x} + \frac{h^2}{2} \frac{\mathrm{d}^2 f(x_i)}{\mathrm{d}x^2} + O(h^3)$$
$$f(x_{i-1}) = f(x_i) - h \frac{\mathrm{d}f(x_i)}{\mathrm{d}x} + \frac{h^2}{2} \frac{\mathrm{d}^2 f(x_i)}{\mathrm{d}x^2} + O(h^3)$$

Difference

$$(f(x_{i+1}) - f(x_i)) - (f(x_{i-1}) - f(x_i)) = 2h \frac{\mathrm{d}f(x_i)}{\mathrm{d}x} + O(h^3)$$

$$ightharpoonup rac{\mathrm{d}f(x_i)}{\mathrm{d}x} pprox rac{f(x_{i+1}) - f(x_{i-1})}{2h}$$
 Central difference

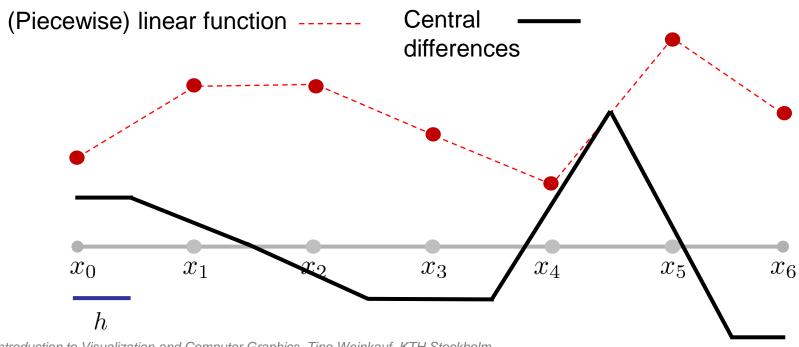
 Central differences have higher approximation order than forward / backward differences

Finite Differences Schemes, Higher order derivatives

$$\frac{d^2 f(x_i)}{dx^2} \approx \frac{f(x_{i+1}) - 2f(x_i) + f(x_{i-1})}{h^2}$$

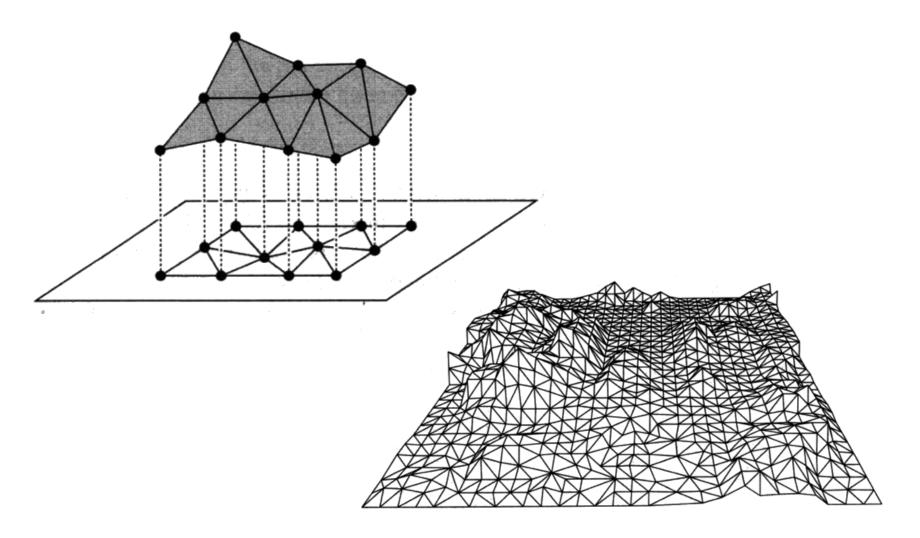
$$\frac{\partial^2 f(x_i, y_j)}{\partial xy} \approx \frac{f(x_{i+1}, y_{j+1}) - f(x_{i+1}, y_{j-1}) - f(x_{i-1}, y_{j+1}) + f(x_{i-1}, y_{j-1})}{4 h_x h_y}$$

1D Example, linear interpolation



Introduction to Visualization and Computer Graphics, Tino Weinkauf, KTH Stockholm

• Piecewise Linear Interpolation in Triangle Meshes



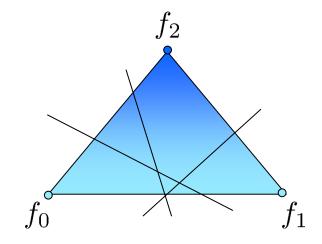
• Linear Interpolation in a Triangle

- There is exactly one linear function that satisfies the interpolation constraint
- A linear function can be written as

$$f(x,y) = a + bx + cy$$

 Polynomial can be obtained by solving the linear system

$$\begin{bmatrix} 1 & x_0 & y_0 \\ 1 & x_1 & y_1 \\ 1 & x_2 & y_2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} f_0 \\ f_1 \\ f_2 \end{bmatrix}$$

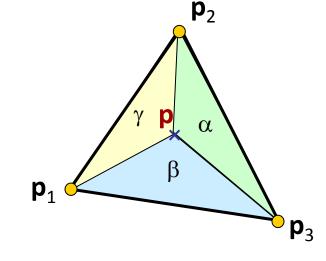


- ullet Linear in x and y
 - Interpolated values along any ray in the plane spanned by the triangle are linear along that ray

Barycentric Coordinates:

Planar case:
 Barycentric combinations of 3 points

$$\mathbf{p} = \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \gamma \mathbf{p}_3$$
, with $: \alpha + \beta + \gamma = 1$
 $\gamma = 1 - \alpha - \beta$



Area formulation:

$$\alpha = \frac{area(\Delta(\mathbf{p}_{1}, \mathbf{p}_{3}, \mathbf{p}))}{area(\Delta(\mathbf{p}_{1}, \mathbf{p}_{2}, \mathbf{p}_{3}))}, \beta = \frac{area(\Delta(\mathbf{p}_{1}, \mathbf{p}_{3}, \mathbf{p}))}{area(\Delta(\mathbf{p}_{1}, \mathbf{p}_{2}, \mathbf{p}_{3}))}, \gamma = \frac{area(\Delta(\mathbf{p}_{1}, \mathbf{p}_{2}, \mathbf{p}))}{area(\Delta(\mathbf{p}_{1}, \mathbf{p}_{2}, \mathbf{p}_{3}))}$$

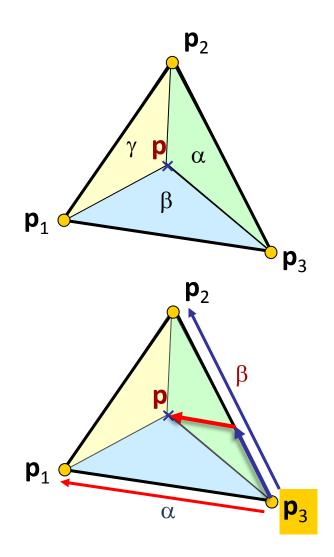
- Barycentric Coordinates:
 - Linear formulation:

$$\mathbf{p} = \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \gamma \mathbf{p}_3$$

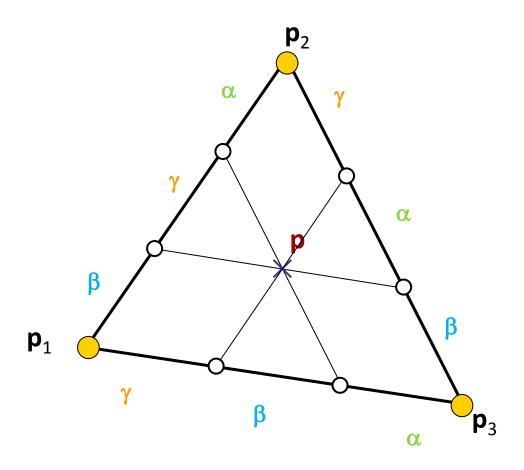
$$= \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + (1 - \alpha - \beta) \mathbf{p}_3$$

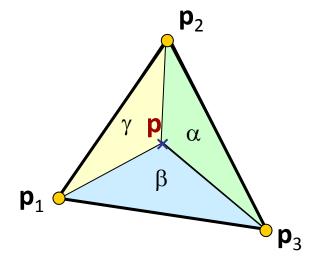
$$= \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \mathbf{p}_3 - \alpha \mathbf{p}_3 - \beta \mathbf{p}_3$$

$$= \mathbf{p}_3 + \alpha (\mathbf{p}_1 - \mathbf{p}_3) + \beta (\mathbf{p}_2 - \mathbf{p}_3)$$



$$\mathbf{p} = \alpha \mathbf{p}_1 + \beta \mathbf{p}_2 + \gamma \mathbf{p}_3$$
, with : $\alpha + \beta + \gamma = 1$





• Barycentric Interpolation in a Triangle

very important

The linear function of a triangle can be computed at any point as

$$f(x,y)=\alpha_0(x,y)f_0+\alpha_1(x,y)f_1+\alpha_2(x,y)f_2$$
 with $\alpha_0+\alpha_1+\alpha_2=1$ (Barycentric Coordinates)

- This also holds for the coordinate $\mathbf{x} = \begin{pmatrix} x \\ y \end{pmatrix}$ of the triangle: $\mathbf{x} = \alpha_0 \, \mathbf{x}_0 + \alpha_1 \, \mathbf{x}_1 + \alpha_2 \, \mathbf{x}_2$
 - \rightarrow Can be used to solve for unknown coefficients α_i :

$$\begin{bmatrix} x_0 & x_1 & x_2 \\ y_0 & y_1 & y_2 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \alpha_0 \\ \alpha_1 \\ \alpha_2 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Barycentric Interpolation in a Triangle

• Solution of
$$\begin{bmatrix} x_0 & x_1 & x_2 \\ y_0 & y_1 & y_2 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \alpha_0 \\ \alpha_1 \\ \alpha_2 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$
 (e.g. Cramer's rule):

$$\alpha_0 = \frac{1}{2A} \det \left(\begin{bmatrix} x & x_1 & x_2 \\ y & y_1 & y_2 \\ 1 & 1 & 1 \end{bmatrix} \right) \qquad \alpha_0 = \frac{\operatorname{Area}([\mathbf{x}, \mathbf{x}_1, \mathbf{x}_2])}{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2])}$$

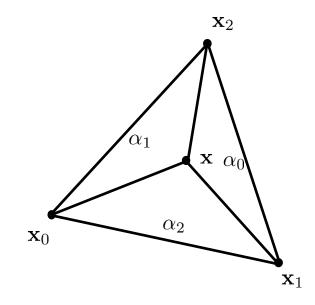
$$\alpha_0 = \frac{\operatorname{Area}([\mathbf{x}, \mathbf{x}_1, \mathbf{x}_2])}{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2])}$$

$$\alpha_1 = \frac{1}{2A} \det \left(\begin{bmatrix} x_0 & x & x_2 \\ y_0 & y & y_2 \\ 1 & 1 & 1 \end{bmatrix} \right) \qquad \alpha_1 = \frac{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}, \mathbf{x}_2])}{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2])}$$

$$\alpha_1 = \frac{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}, \mathbf{x}_2])}{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2])}$$

$$\alpha_2 = \frac{1}{2A} \det \left(\begin{bmatrix} x_0 & x_1 & x \\ y_0 & y_1 & y \\ 1 & 1 & 1 \end{bmatrix} \right) \qquad \alpha_2 = \frac{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}])}{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2])}$$

$$\alpha_2 = \frac{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}])}{\operatorname{Area}([\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2])}$$



$$A=rac{1}{2}\det\left(egin{bmatrix}x_0&x_1&x_2\y_0&y_1&y_2\1&1&1\end{bmatrix}
ight)$$

Inside triangle criteria

$$0 \le \alpha_0, \, \alpha_1, \, \alpha_2 \le 1$$

- Barycentric Interpolation in a Tetrahedron
- Analogous to the triangle case

Gradient of a linearly interpolated function in a triangle/tetrahedron

Constant!

