

# Visual Data Analytics Data Reconstruction and Interpolation II

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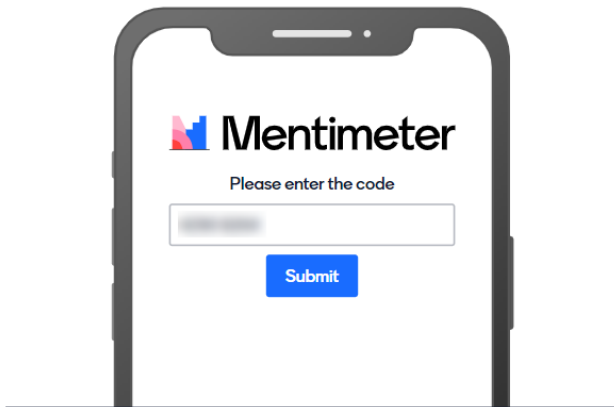
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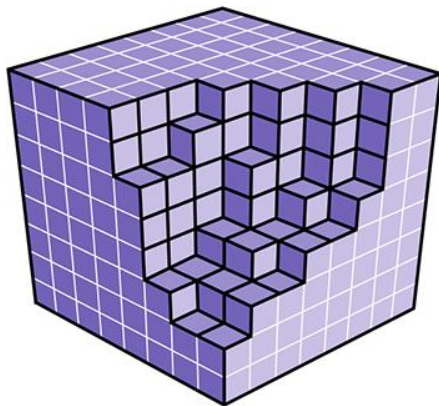
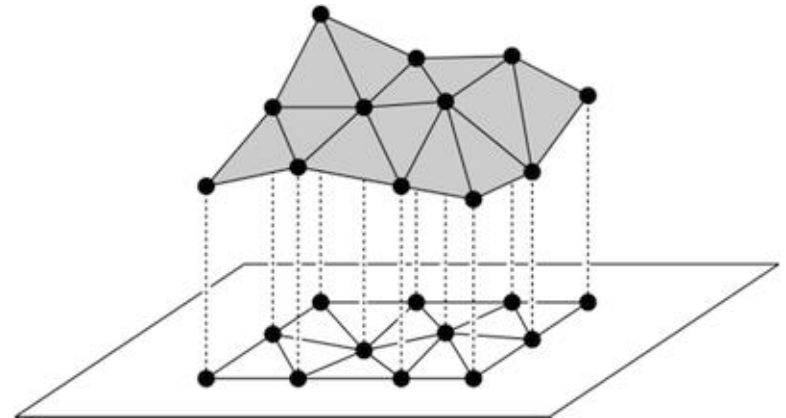
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Or use QR code

# Overview

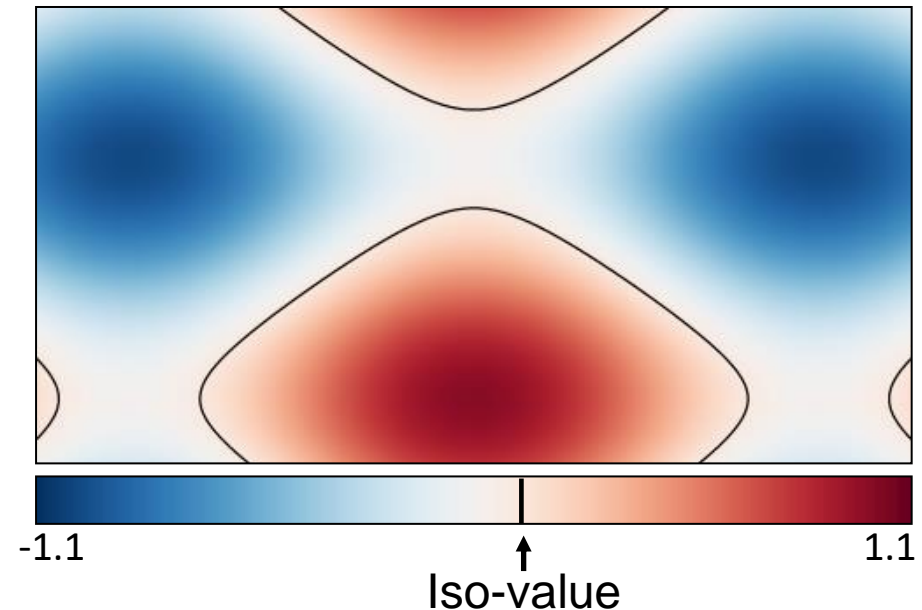
- Interpolation on grids / meshes
  - Barycentric interpolation
  - Bi-/trilinear interpolation
  - High-quality reconstruction



Tricubic vs. trilinear interpolation

# Data interpolation

- Why do we need a **continuous representation** of data given on grids / meshes
  - Better communication of spatial data distribution
  - Some techniques require a continuous representation (e.g., color mapping, iso-lines)

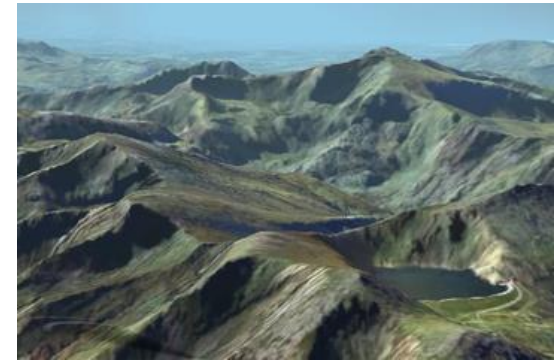
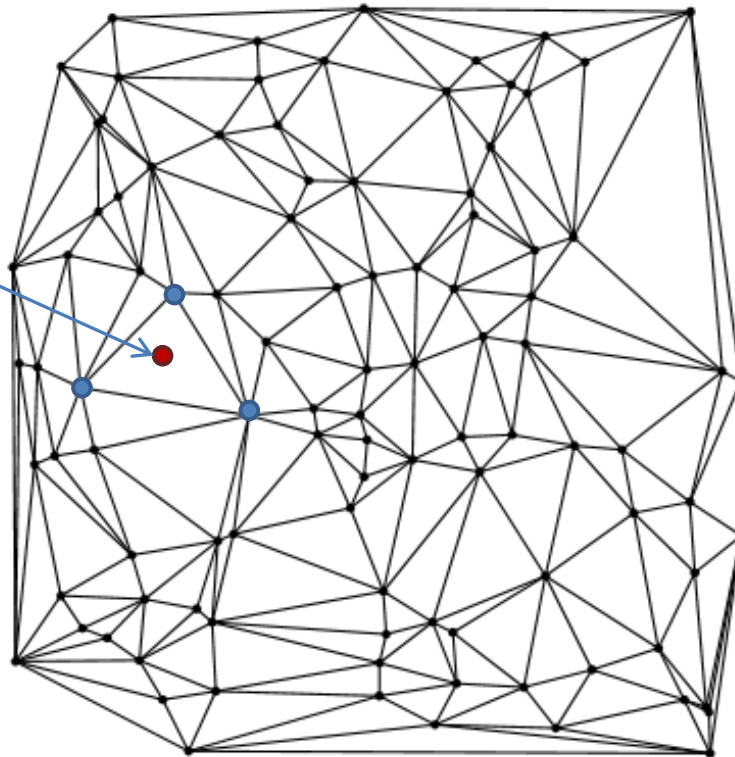


**Isoline** = all points having  
a certain value

# Data interpolation

- Try finding a piecewise (local) reconstruction function
  - Connect the points so that a **triangulation** is obtained
  - Interpolate locally within the triangles

Value obtained by  
only considering  
values at triangle  
corners

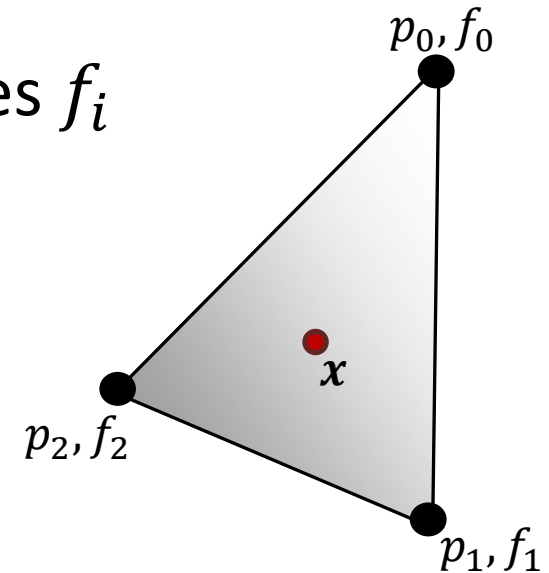


# Data interpolation

- How to interpolate inside a triangle
  - The triangle lives in a  $(N = 2)$ D plane; it has  $N + 1$  points  $(x_i, y_i)$  with values  $f_i$
  - Can we find a function  $f$  that interpolates  $f_i$  at the points  $p_i$ , i.e.,

$$f(p_i) = f_i, \quad i = 0, \dots, N$$

(interpolation constraint)



- If so, then the value at any point  $x$  can be interpolated by evaluating  $f(x)$

# Data interpolation

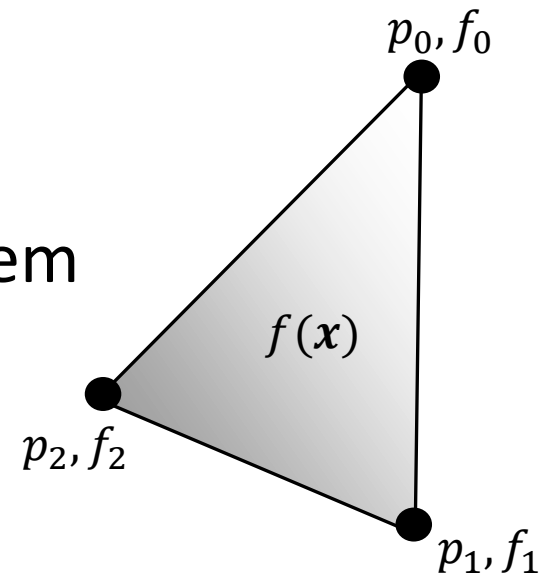
- There is a unique linear function that satisfies the interpolation constraint

- A linear function can be written as

$$f(\mathbf{x}) = a + bx + cy$$

- The unknown coefficients  $a, b, c$  can be obtained by solving the system

$$\begin{array}{l} p_0 \rightarrow \\ p_1 \rightarrow \\ p_2 \rightarrow \end{array} \begin{bmatrix} 1 & x_0 & y_0 \\ 1 & x_1 & y_1 \\ 1 & x_2 & y_2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} f_0 \\ f_1 \\ f_2 \end{bmatrix}$$





# Data interpolation

- Example

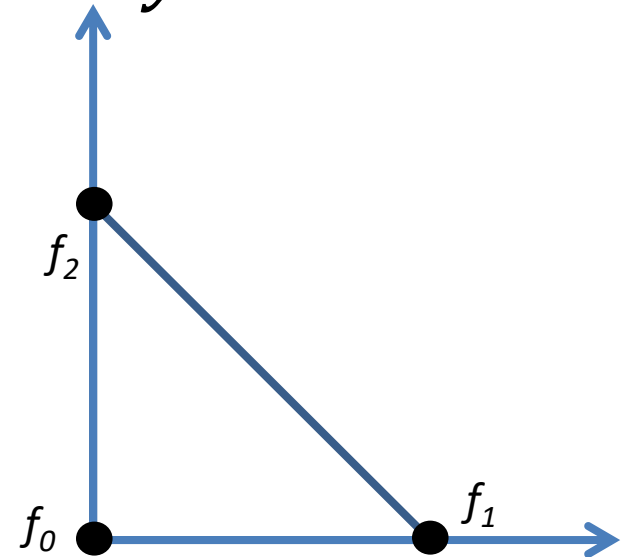
$$p_0 = (0, 0), \quad p_1 = (1, 0), \quad p_2 = (0, 1) \quad \text{with} \\ f_0 = 1, \quad f_1 = 8, \quad f_2 = 2$$

- Obtain  $a, b, c$  in interpolation function

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

$$\begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 1 \\ 8 \\ 2 \end{bmatrix}$$

$$\begin{aligned} a &= 1 \\ b &= 7 \\ c &= 1 \end{aligned}$$



# Data interpolation

- Example

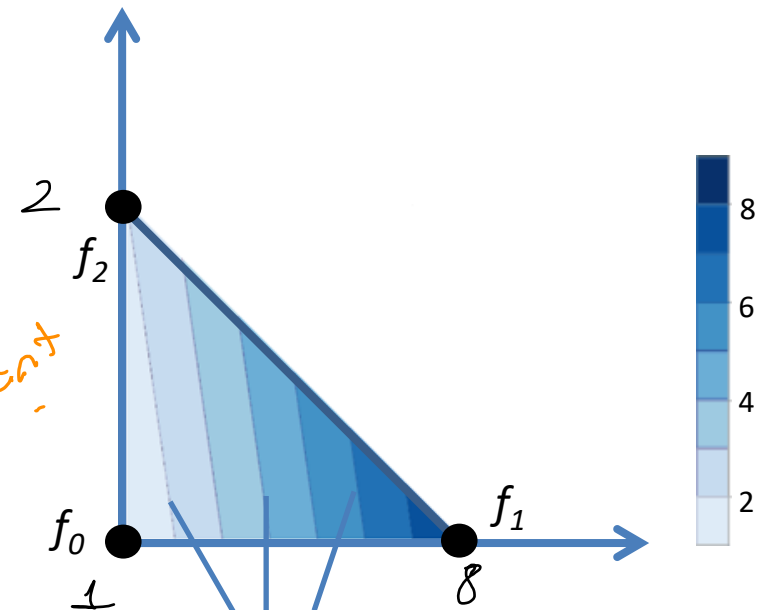
$$p_0 = (0, 0), p_1 = (1, 0), p_2 = (0, 1) \text{ with} \\ f_0 = 1, f_1 = 8, f_2 = 2$$

- Obtain  $a, b, c$  by solving the system

$$\begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 1 \\ 8 \\ 2 \end{bmatrix}$$

$$a = 1, b = 7, c = 1 \\ f(x, y) = 1 + 7x + 1y$$

*important*

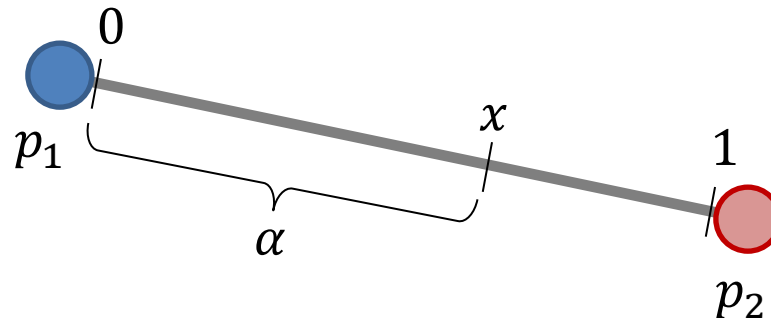


Notice that the isolines are straight lines

# Data interpolation

- Barycentric interpolation

- Another way to interpolate inside a triangle, which yields the same linear interpolation as before
- But let's solve a simpler problem first

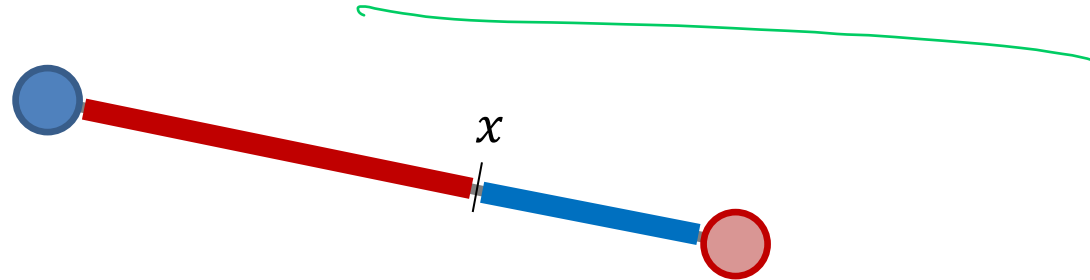


- We want to define a color for every  $\alpha \in [0, 1]$



# Data interpolation

- How do we come up with an equation?



The closer  $x$  is to the red point, the more red we want

The closer  $x$  is to the blue point, the more blue we want

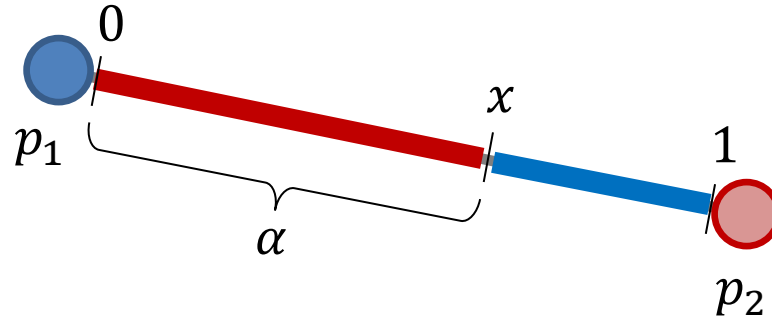


Percentage red = (length of red segment) / (total length)

Percentage blue = (length of blue segment) / (total length)

# Data interpolation

- How do we come up with an equation?



The closer  $x$  is to the red point, the more red we want

The closer  $x$  is to the blue point, the more blue we want



Percentage red =  $\alpha$

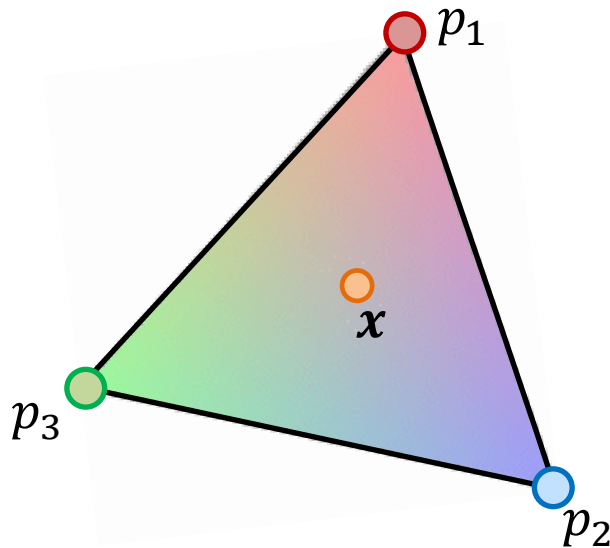
 Normalized

Percentage blue =  $1 - \alpha$

$$f(\alpha) = (1 - \alpha) \cdot p_1 + \alpha \cdot p_2$$

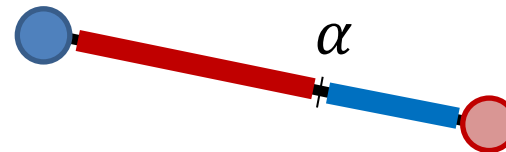
# Data interpolation

- Barycentric interpolation
  - Now what about triangles?



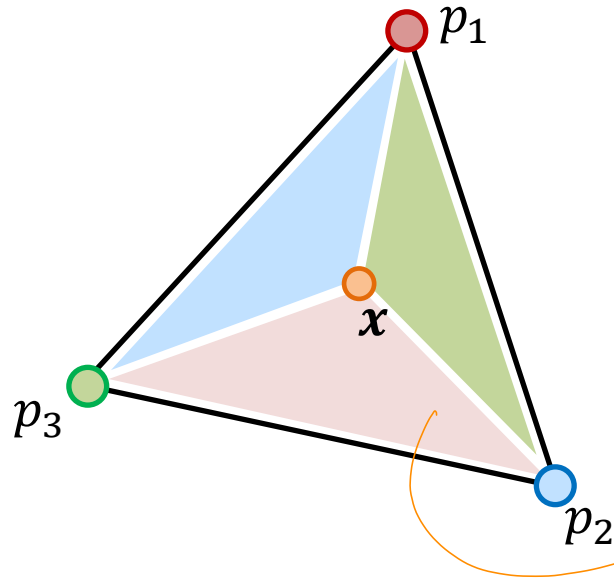
What's the interpolated value at the point  $x$ ?

In 1D we used ratios of lengths



# Data interpolation

- Barycentric interpolation
  - Now what about triangles?

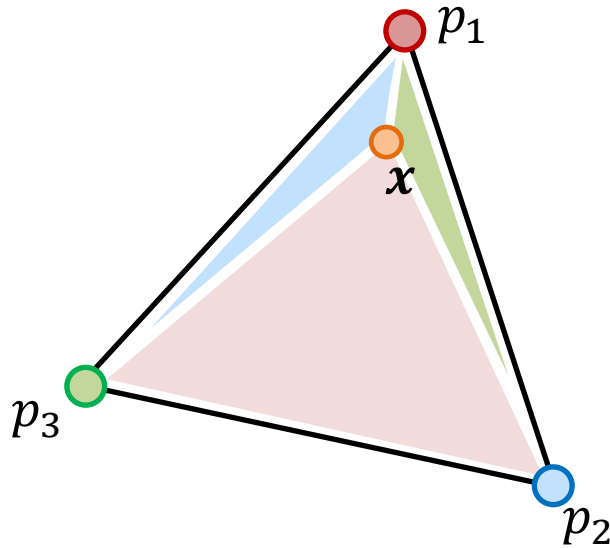


What about ratios of 2D areas?

*Sizes of this triangle represent  
how much red I want.  
percentage of red.*

# Data interpolation

- Barycentric interpolation
  - Now what about triangles?

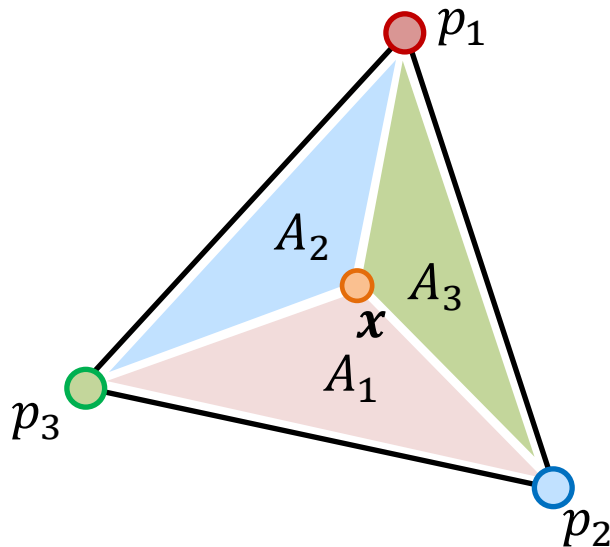


As  $x$  approaches the red point, the red area (for example) covers more of the triangle



# Data interpolation

- Barycentric interpolation
  - Just like before:



$$\alpha_1 = A_1 / A$$

percentage red

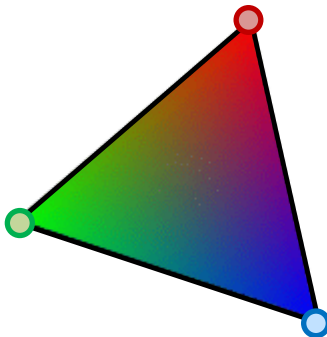
$$\alpha_2 = A_2 / A$$

percentage blue

$$\alpha_3 = A_3 / A$$

percentage green

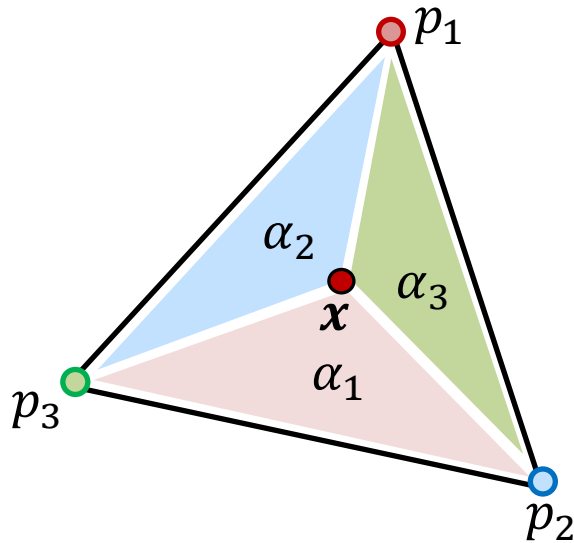
$A$  ... area of whole triangle



$$x = \alpha_1 p_1 + \alpha_2 p_2 + \alpha_3 p_3$$

$$\alpha_1 + \alpha_2 + \alpha_3 = 1$$

# Data interpolation



$$\alpha_1 = \frac{\text{area}(\Delta xp_2p_3)}{\text{area}(\nabla p_1p_2p_3)}$$

$$\alpha_2 = \frac{\text{area}(\Delta p_1xp_3)}{\text{area}(\nabla p_1p_2p_3)}$$

$$\alpha_3 = \frac{\text{area}(\Delta p_1p_2x)}{\text{area}(\nabla p_1p_2p_3)}$$

## Barycentric interpolation

$$\mathbf{x} = \alpha_1 p_1 + \alpha_2 p_2 + \alpha_3 p_3$$

$$\alpha_1 + \alpha_2 + \alpha_3 = 1$$

Inside triangle criteria

$$0 \leq \alpha_1 \leq 1$$

$$0 \leq \alpha_2 \leq 1$$

$$0 \leq \alpha_3 \leq 1$$

# Data interpolation

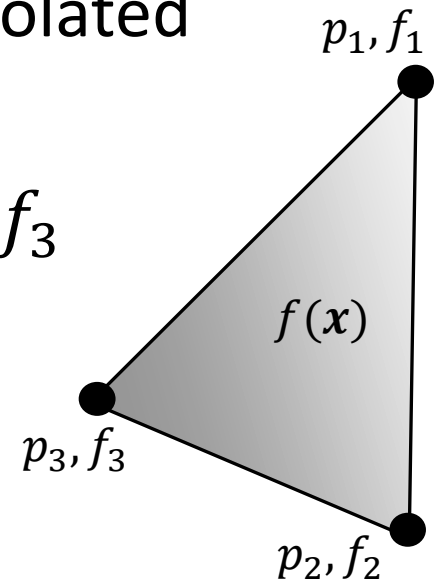
- Barycentric interpolation
  - Every point  $x$  in a triangle can be written as a barycentric combination of the vertices  $p_i$ :

$$x = \sum_i \alpha_i p_i \quad \text{with} \quad \sum_i \alpha_i = 1$$

( $\alpha_i$  ... barycentric coordinates)

- If  $\alpha_i$  are known, then  $f(x)$  can be interpolated from values  $f_i$  at the vertices via

$$f(x) = \alpha_1 f_1 + \alpha_2 f_2 + \underbrace{(1 - \alpha_1 - \alpha_2)}_{\alpha_3} f_3$$



# Data interpolation

**Example:** Given a triangle with vertices  $p_1 = (0.5, 2.5)$ ,  $p_2 = (1.5, 4.5)$  and  $p_3 = (2.5, 2.5)$ . Compute the barycentric coordinates of the points  $P = (1.5, 2.5)$  and  $Q = (1.5, 0.5)$  with respect to the triangle.

The area of  $p_2$  is 0 because  $P$  is in the bottom of the triangle,  $a_2 = 0$   
 $a_1 = 0.5$   $a_3 = 0.5$

So for the  $Q$  value we need to calculate areas

Firstly, we put  $x$  values

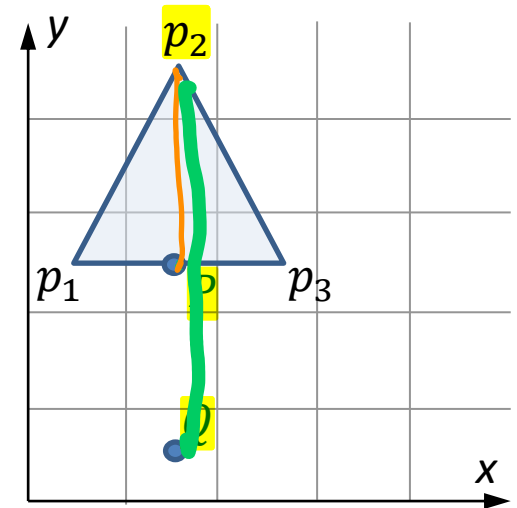
$$1.5 = 0.5 a_1 + 1.5 a_2 + 2.5 a_3 \quad (0.5)$$

$$0.5 = 2.5 a_1 + 4.5 a_2 + 2.5 a_3$$

$$1 = a_1 + a_2 + a_3 \quad (2.5)$$

$$-12 \quad 2 a_1 + 3 a_2$$

$$2 a_2 = -2 \quad a_2 = \boxed{-1} \quad a_1 = 1 \quad a_3 = 1$$



# Data interpolation

**Example:** Given a triangle with vertices  $p_1 = (0.5, 2.5)$ ,  $p_2 = (1.5, 4.5)$  and  $p_3 = (2.5, 2.5)$ . Compute the barycentric coordinates of the points  $P = (1.5, 2.5)$  and  $Q = (1.5, 0.5)$  with respect to the triangle.

Point  $P$ :

$$\alpha_2 = 0 \rightarrow \alpha_1 = \alpha_3 = 0.5$$

Point  $Q$ :

$$I: 1.5 = 0.5 \alpha_1 + 1.5 \alpha_2 + 2.5 \alpha_3 \quad \leftarrow \text{x coordinates}$$

$$II: 0.5 = 2.5 \alpha_1 + 4.5 \alpha_2 + 2.5 \alpha_3 \quad \leftarrow \text{y coordinates}$$

$$III: 1 = \alpha_1 + \alpha_2 + \alpha_3$$

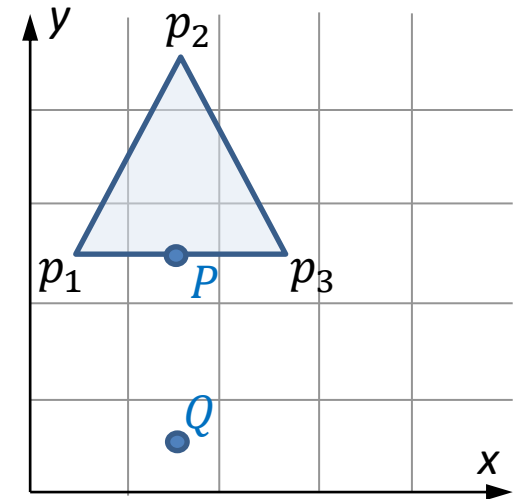
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$$I - II: 1 = -2\alpha_1 - 3\alpha_2$$

$$II - 2.5 III: -2 = 2\alpha_2 \rightarrow \alpha_2 = -1$$

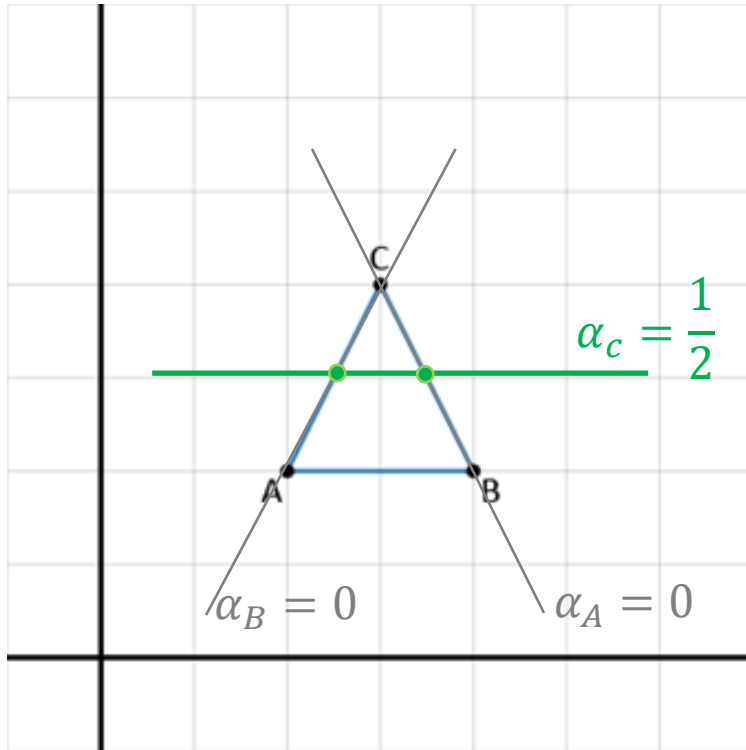
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$$\alpha_2 \rightarrow I': -2 = -2\alpha_1 \rightarrow \alpha_1 = 1 \rightarrow \alpha_3 = 1$$



# Barycentric coordinates

**Example:** Given a triangle with vertices  $(2, 2)$ ,  $(4, 2)$ , and  $(3, 4)$ . Draw the iso-contours along which the barycentric coordinate  $\alpha_C$  corresponding to point  $C$  has the value  $\frac{1}{2}$ ,  $-\frac{1}{2}$ ,  $\frac{3}{2}$ , respectively.



$$\alpha_A + \alpha_B + \alpha_C = 1$$

$$\alpha_C = \frac{1}{2}, \quad \alpha_A = 0 \rightarrow \alpha_B = \frac{1}{2}$$

$$\alpha_B = 0 \rightarrow \alpha_A = \frac{1}{2}$$

So we assume that

$$\alpha_A = 0, \text{ then } \alpha_B = 1/2$$

For finding exact coordinate,

$$x = 4 \times \frac{1}{2} + 3 \cdot \frac{1}{2} = 3.5$$

$$y = 2 \cdot \frac{1}{2} + 4 \cdot \frac{1}{2} = 3$$

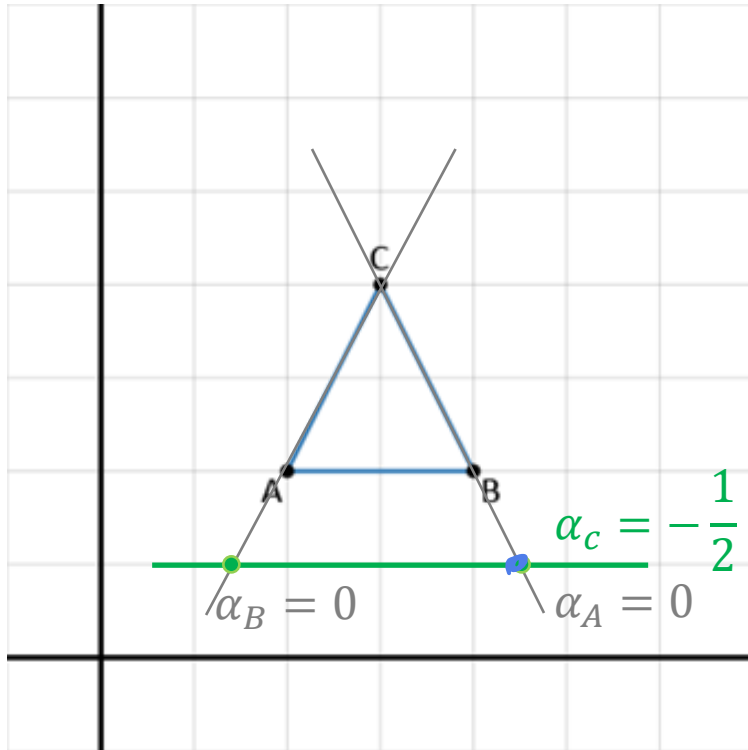
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$$\alpha_A + \alpha_B + \alpha_C = 1$$

$$\alpha_C = -\frac{1}{2}, \quad \alpha_A = 0 \rightarrow \alpha_B = \frac{3}{2}$$

$$\alpha_B = 0 \rightarrow \alpha_A = \frac{3}{2}$$

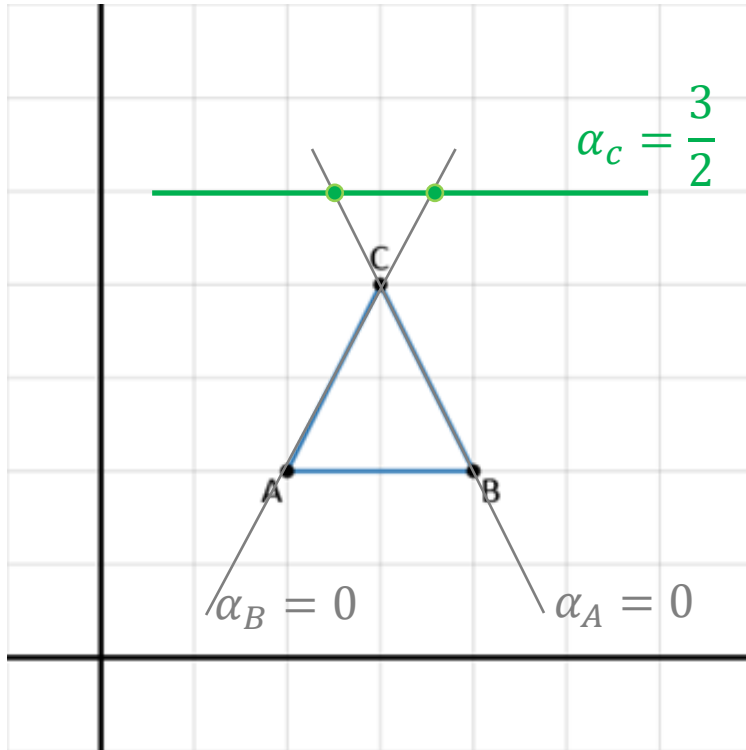


# Barycentric coordinates

**Example:** Given a triangle with vertices  $(2, 2)$ ,  $(4, 2)$ , and  $(3, 4)$ . Draw the iso-contours along which the barycentric coordinate  $\alpha_c$  corresponding to point  $C$  has the value  $\frac{1}{2}$ ,  $-\frac{1}{2}$ ,  $\frac{3}{2}$ , respectively.

$$\alpha_A + \alpha_B + \alpha_C = 1$$

$$\alpha_C = \frac{3}{2}, \quad \alpha_A = 0 \rightarrow \alpha_B = -\frac{1}{2}$$
$$\alpha_B = 0 \rightarrow \alpha_A = -\frac{1}{2}$$

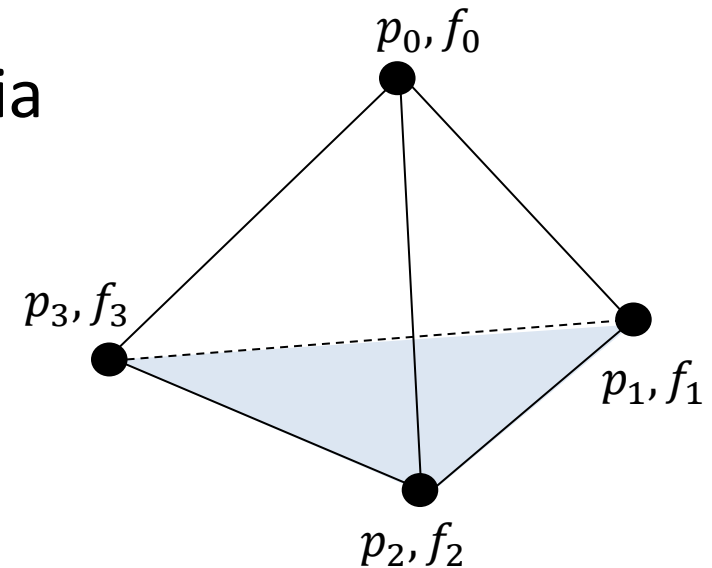




# Data interpolation

- Interpolation of scalar values in a tetrahedron
  - A unique linear interpolation function  $f(\mathbf{x}) = a + bx + cy + dz$  exists which interpolates the scalar values at the vertices
  - Solve for coefficients  $a, b, c, d$  via

$$\begin{bmatrix} 1 & x_0 & y_0 & z_0 \\ 1 & x_1 & y_1 & z_1 \\ 1 & x_2 & y_2 & z_2 \\ 1 & x_3 & y_3 & z_3 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} f_0 \\ f_1 \\ f_2 \\ f_3 \end{bmatrix}$$



# Data interpolation

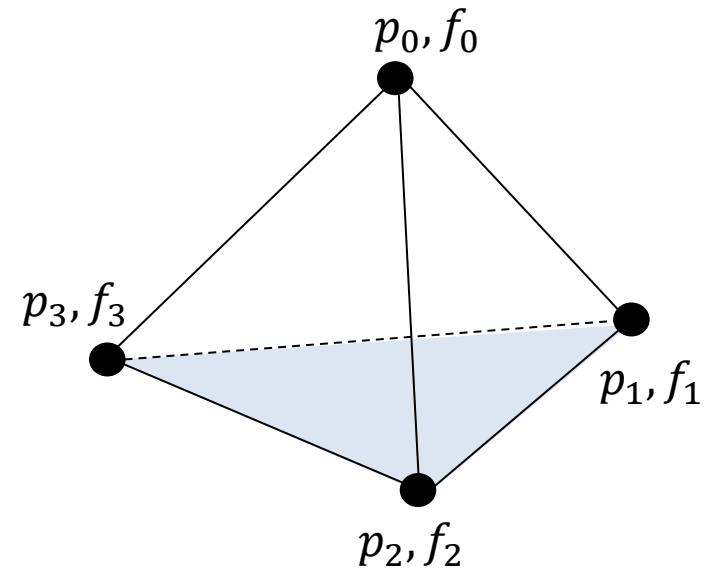
- How to get the **gradient** inside the tetrahedron?
  - Given the linear interpolation function

$$f(\mathbf{x}) = a + bx + cy + dz$$

- The gradient of the interpolated scalar field can be obtained by

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \\ \frac{\partial f}{\partial z} \end{pmatrix} = \begin{pmatrix} b \\ c \\ d \end{pmatrix}$$

- The **gradient is constant** within the tetrahedron



# Data interpolation

**Example:** For the tetrahedron with vertices  $A = (0,0,0)$ ,  $B = (1,0,0)$ ,  $C = (0,1,0)$ ,  $D = (0,0,1)$ , compute the linear interpolation function  $f(x, y, z) = a + b \cdot x + c \cdot y + d \cdot z$  which interpolates the scalar values  $f_A = 1, f_B = 0, f_C = 0, f_D = 1$  at the corresponding vertices.

$$f_A = 1 = a$$

$$f_B = 0 = a + b \rightarrow -1$$

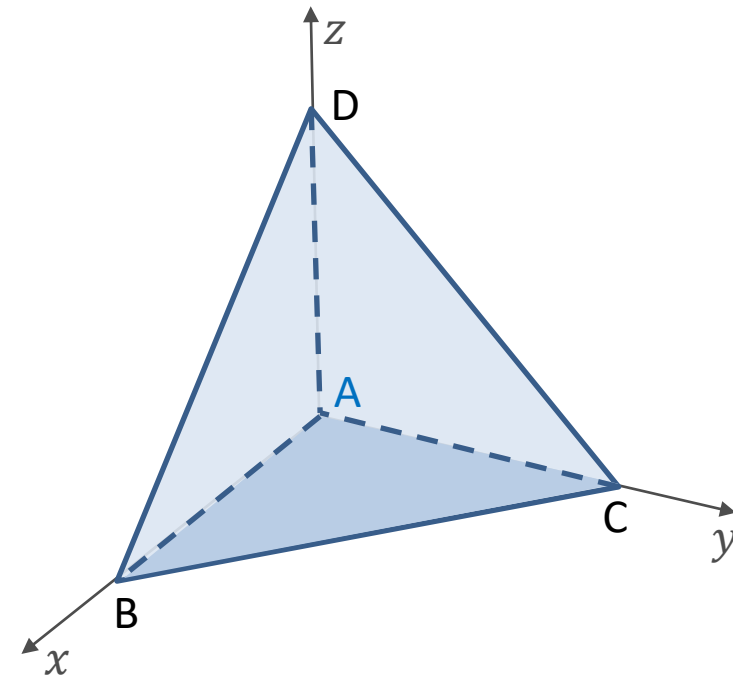
$$f_C = 0 = a + c \rightarrow -1$$

$$f_D = 1 = a + d \rightarrow 0$$

$$f(x, y, z) = 1 - x - y$$

$$\nabla f = (-1, -1, 0)^T$$

Compute the gradient of the interpolated scalar field at the center of the tetrahedron.

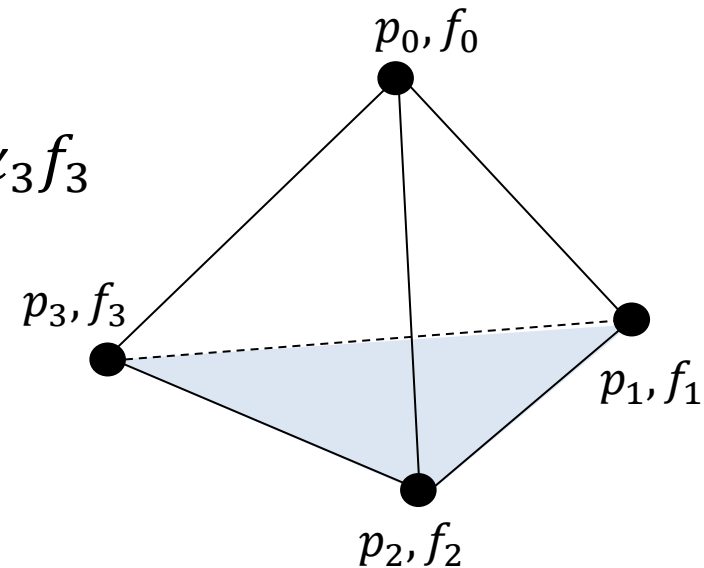


# Data interpolation

- Barycentric interpolation in 3D
  - Scalar values can be interpolated by means of barycentric coordinates:

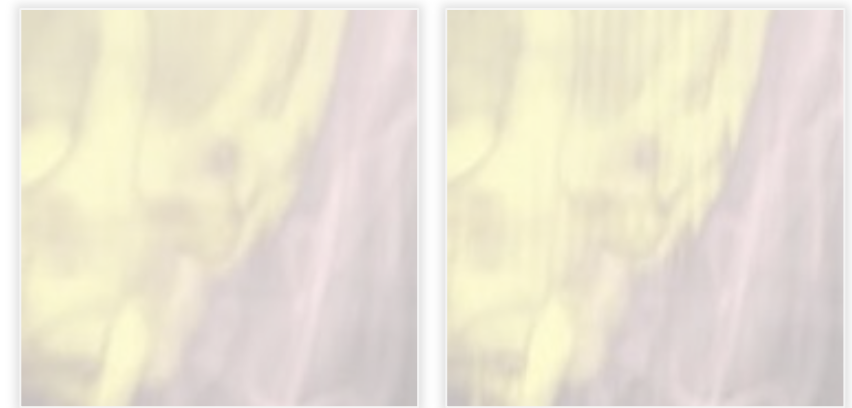
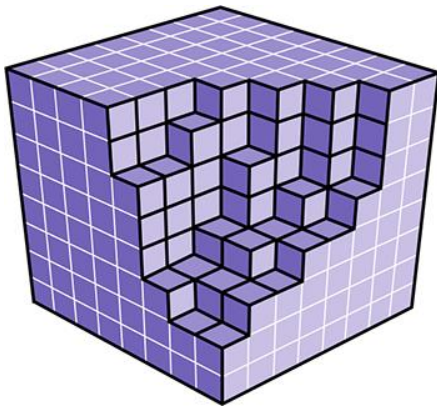
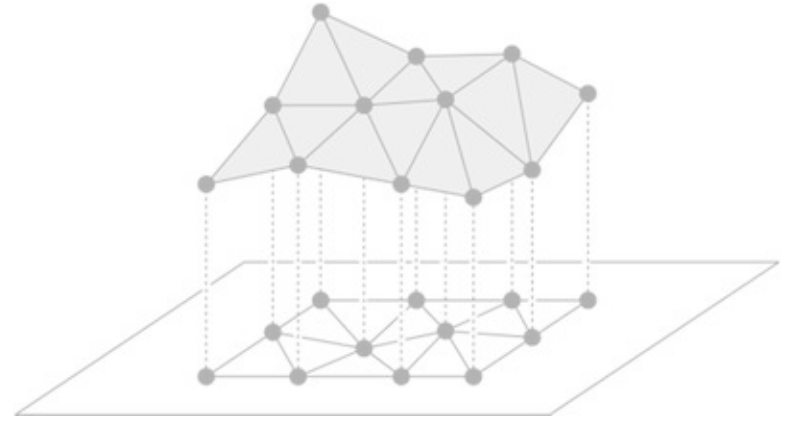
$$\mathbf{x} = \alpha_0 \mathbf{p}_0 + \alpha_1 \mathbf{p}_1 + \alpha_2 \mathbf{p}_2 + \alpha_3 \mathbf{p}_3$$
$$\alpha_0 + \alpha_1 + \alpha_2 + \alpha_3 = 1$$

$$\rightarrow f(x, y, z) = \alpha_0 f_0 + \alpha_1 f_1 + \alpha_2 f_2 + \alpha_3 f_3$$



# Overview

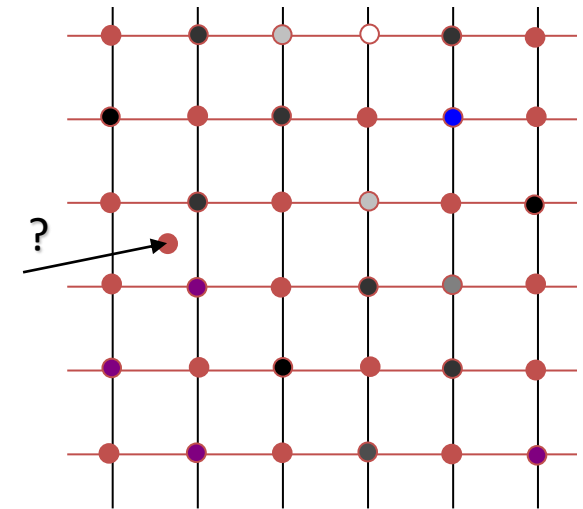
- Interpolation on grids / meshes
  - Barycentric interpolation
  - Bi-/trilinear interpolation
  - High-quality reconstruction



Tricubic vs. trilinear interpolation

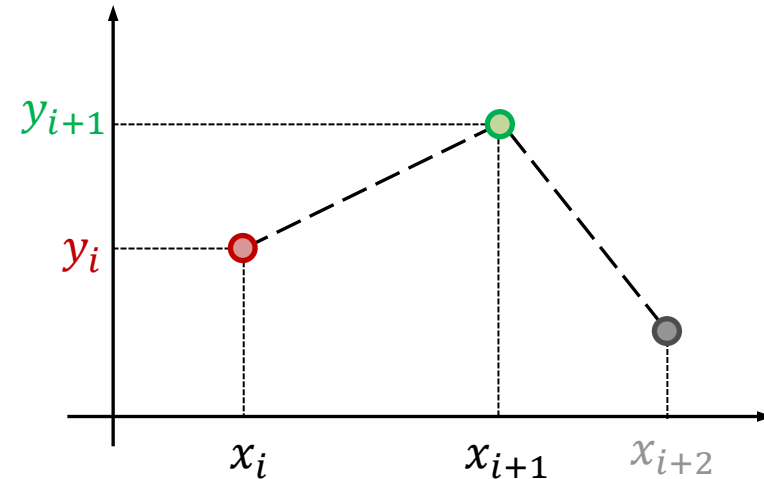
# Interpolation on grids

- **Problem**: assume data values are given only at vertices of a Cartesian grid
- How can we hide the underlying grid structure, i.e., how can we get a **continuous** data distribution over the spatial domain?
- This can be done via **interpolation**!



# Interpolation on grids

- Piecewise linear interpolation
  - Simplest approach (except for piece-wise constant interpolation)
  - Data points:  $(x_1, y_1), \dots, (x_N, y_N)$



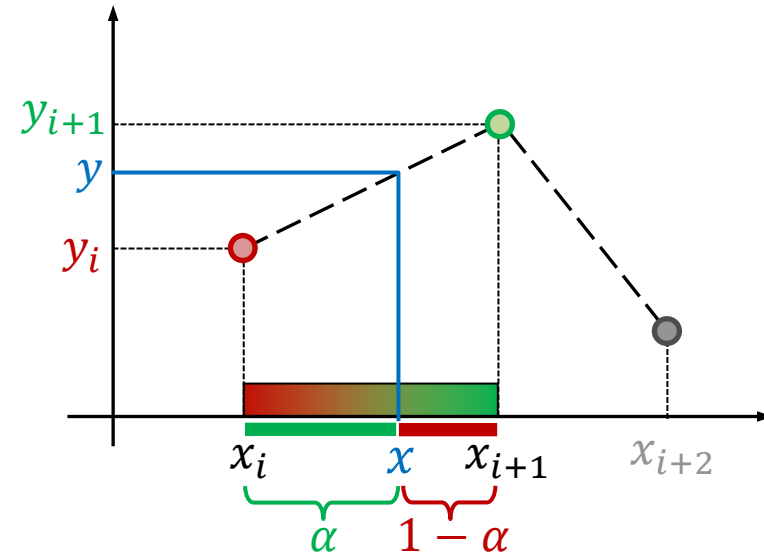
# Interpolation on grids

- Piecewise linear interpolation
  - Simplest approach (except for piece-wise constant interpolation)
  - Data points:  $(x_1, y_1), \dots, (x_N, y_N)$
  - For any point  $x$  with

$$x_i \leq x \leq x_{i+1}$$

evaluate  $f(x) = (1 - \alpha)y_i + \alpha y_{i+1}$

where  $\alpha = \frac{x - x_i}{x_{i+1} - x_i} \in [0, 1]$

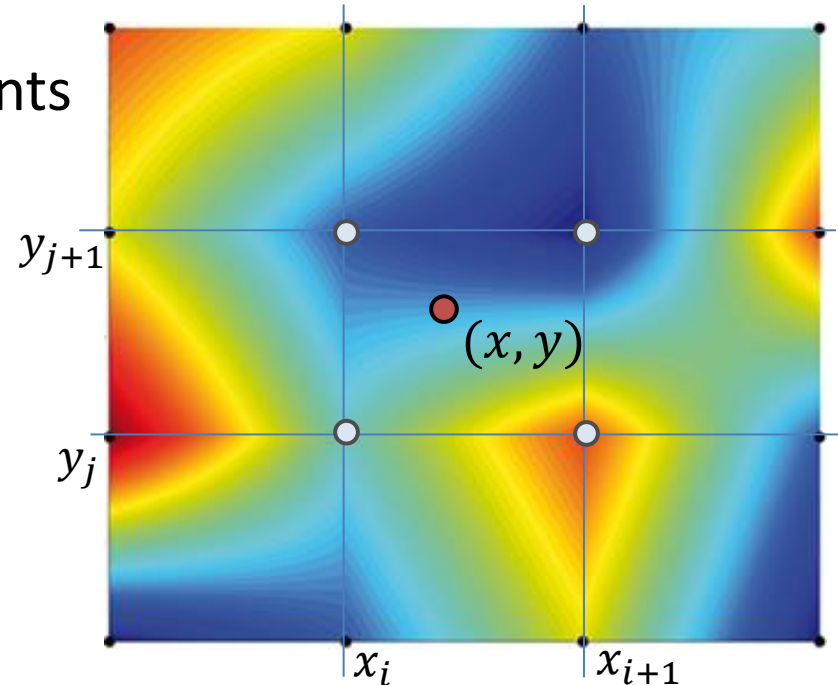


The closer  $x$  is to the green point, the more green we want



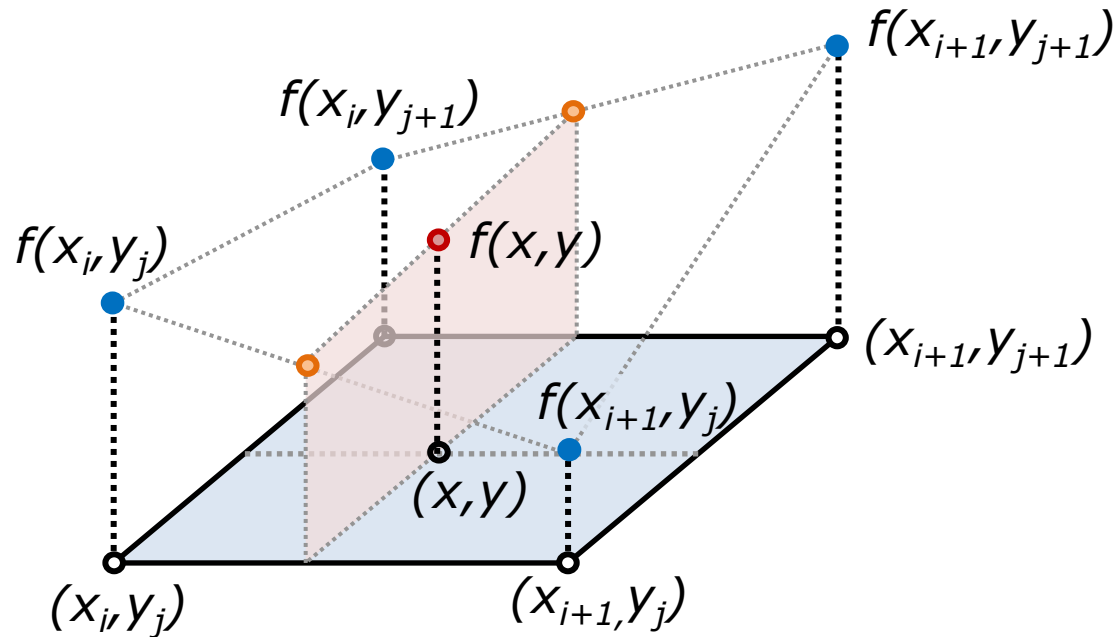
# Interpolation on grids

- Linear interpolation
  - $C^0$  continuity at segment boundaries
    - Tangents don't match at segment transition
  - Easily extendible to 2D
    - 2D cell consisting of 4 data points  $(x_i, y_j), \dots, (x_{i+1}, y_{j+1})$  with scalar values  $f_{k,l} = f(x_k, y_l)$
    - Bilinear interpolation of points  $(x, y)$  with  $x_i \leq x \leq x_{i+1}$  and  $y_j \leq y \leq y_{j+1}$



# Interpolation on grids

- Bilinear interpolation on a rectangle



# Interpolation on grids

- Bilinear interpolation on a rectangle

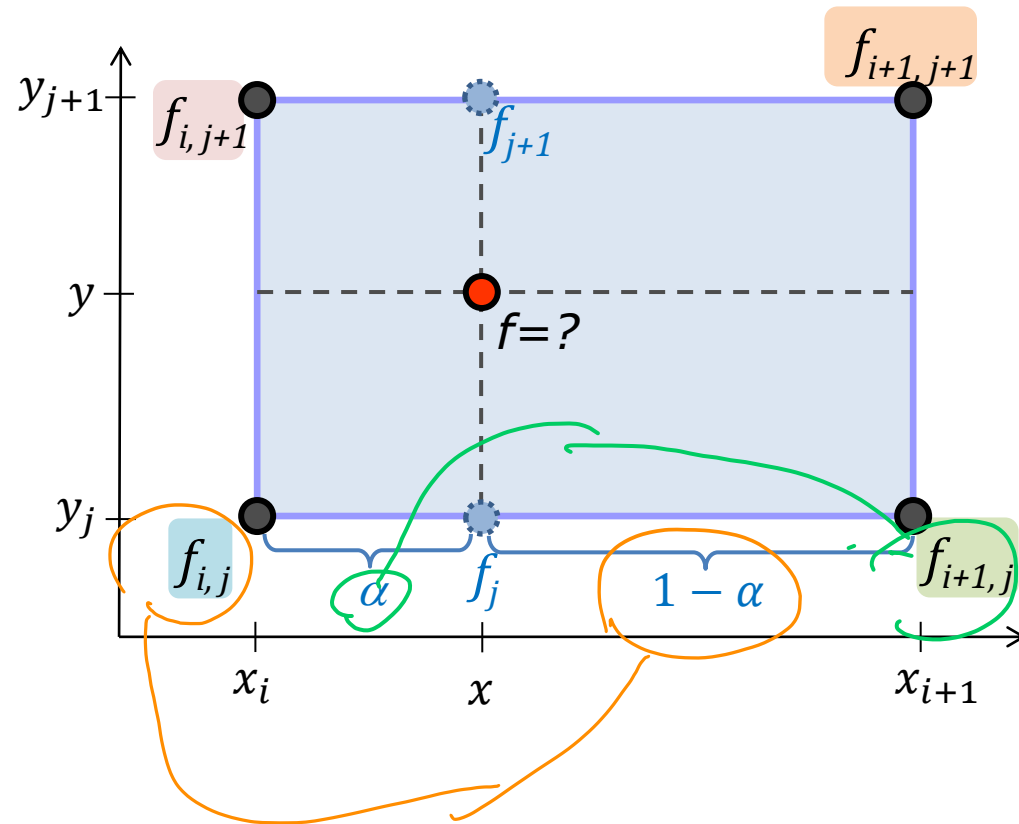
$$f(\alpha, \beta) = ?$$

Interpolate horizontally

$$f_j = (1 - \alpha) f_{i,j} + \alpha f_{i+1,j}$$

$$f_{j+1} = (1 - \alpha) f_{i,j+1} + \alpha f_{i+1,j+1}$$

$$\alpha = \frac{x - x_i}{x_{i+1} - x_i}$$



# Interpolation on grids

- Bilinear interpolation on a rectangle

$$\begin{aligned} f(\alpha, \beta) &= (1 - \beta)[(1 - \alpha)f_{i,j} + \alpha f_{i+1,j}] + \beta[(1 - \alpha)f_{i,j+1} + \alpha f_{i+1,j+1}] \\ &= (1 - \beta)f_j + \beta f_{j+1} \end{aligned}$$

with

$$f_j = (1 - \alpha)f_{i,j} + \alpha f_{i+1,j}$$

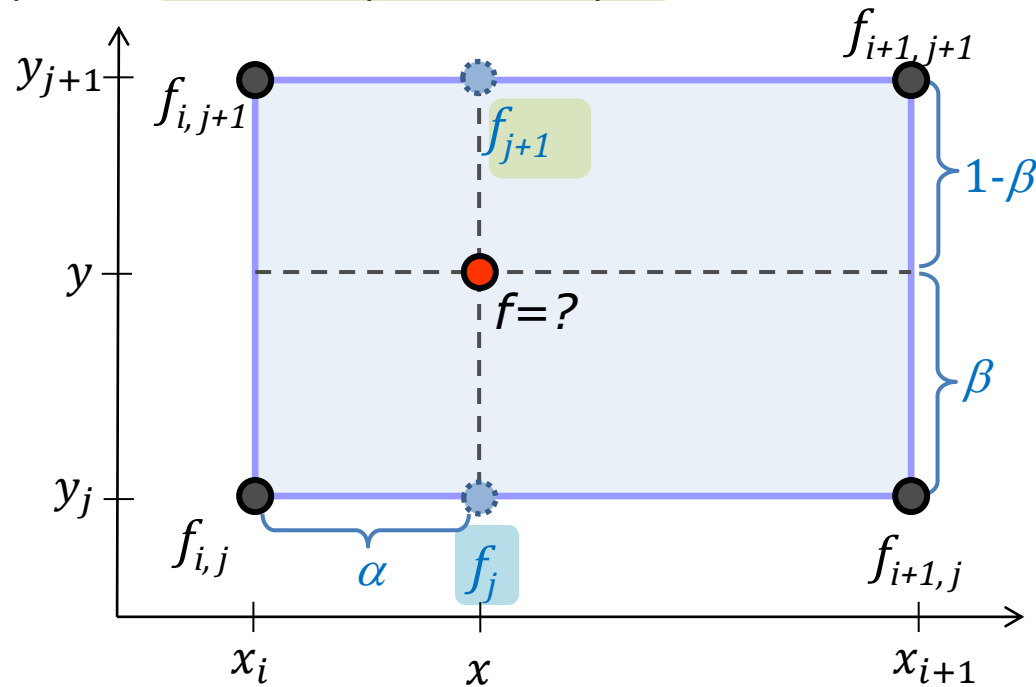
$$f_{j+1} = (1 - \alpha)f_{i,j+1} + \alpha f_{i+1,j+1}$$

and local coordinates

$$\alpha = \frac{x - x_i}{x_{i+1} - x_i},$$

$$\beta = \frac{y - y_j}{y_{j+1} - y_j},$$

$$\alpha, \beta \in [0, 1]$$



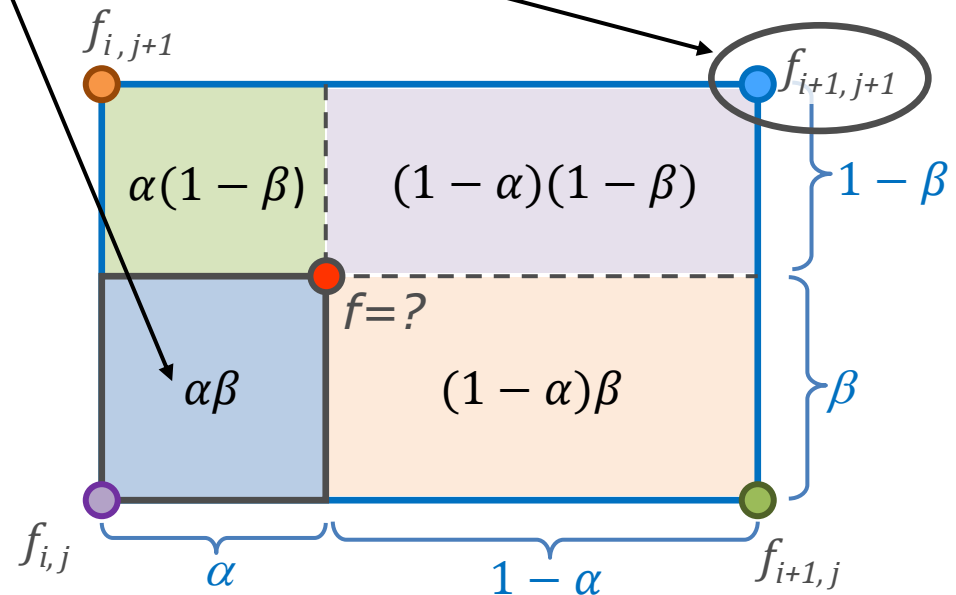
# Interpolation on grids

- Geometric interpretation of bilinear interpolation

$$f(\alpha, \beta) = (1 - \beta)[(1 - \alpha)f_{i,j} + \alpha f_{i+1,j}] + \beta[(1 - \alpha)f_{i,j+1} + \alpha f_{i+1,j+1}]$$

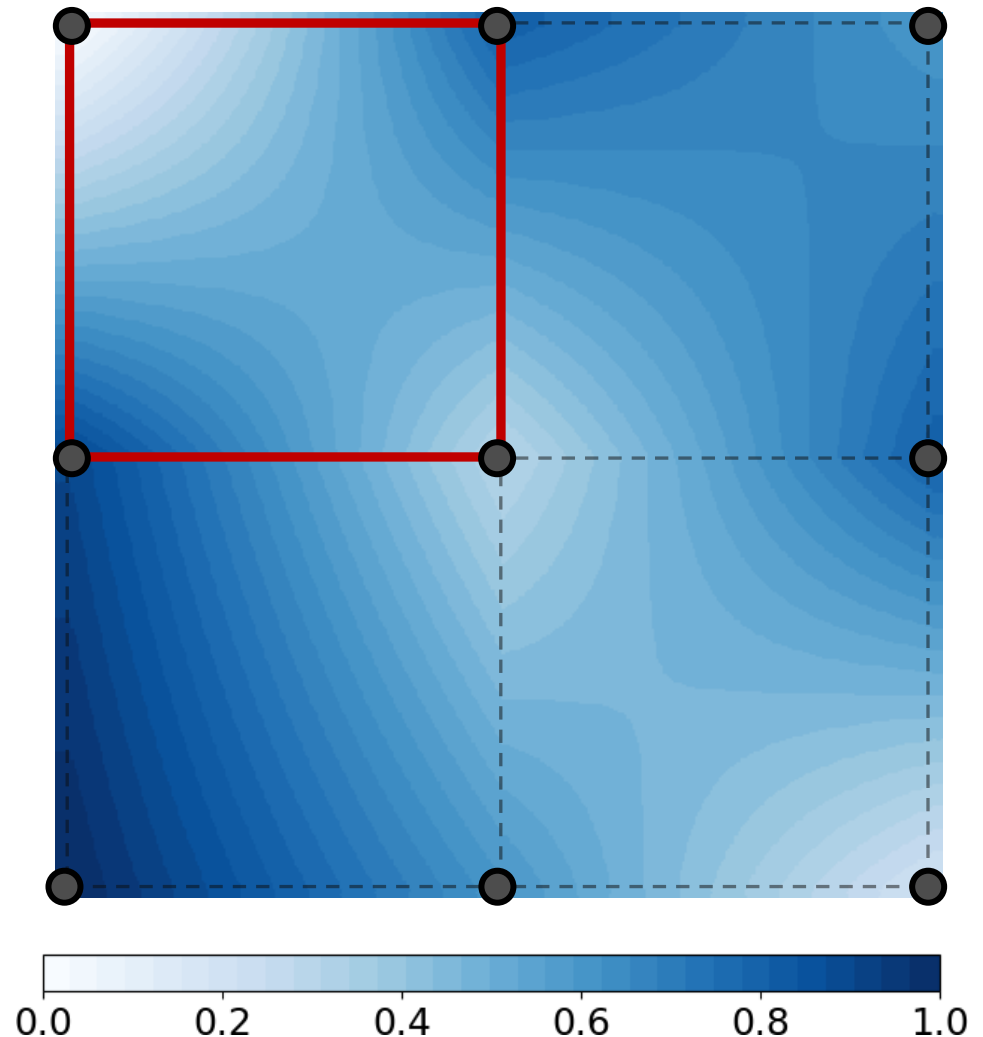
$$= (1 - \alpha)(1 - \beta)f_{i,j} + \alpha(1 - \beta)f_{i+1,j} \\ + (1 - \alpha)\beta f_{i,j+1} + \alpha\beta f_{i+1,j+1}$$

- Opposite points are weighted by local areas



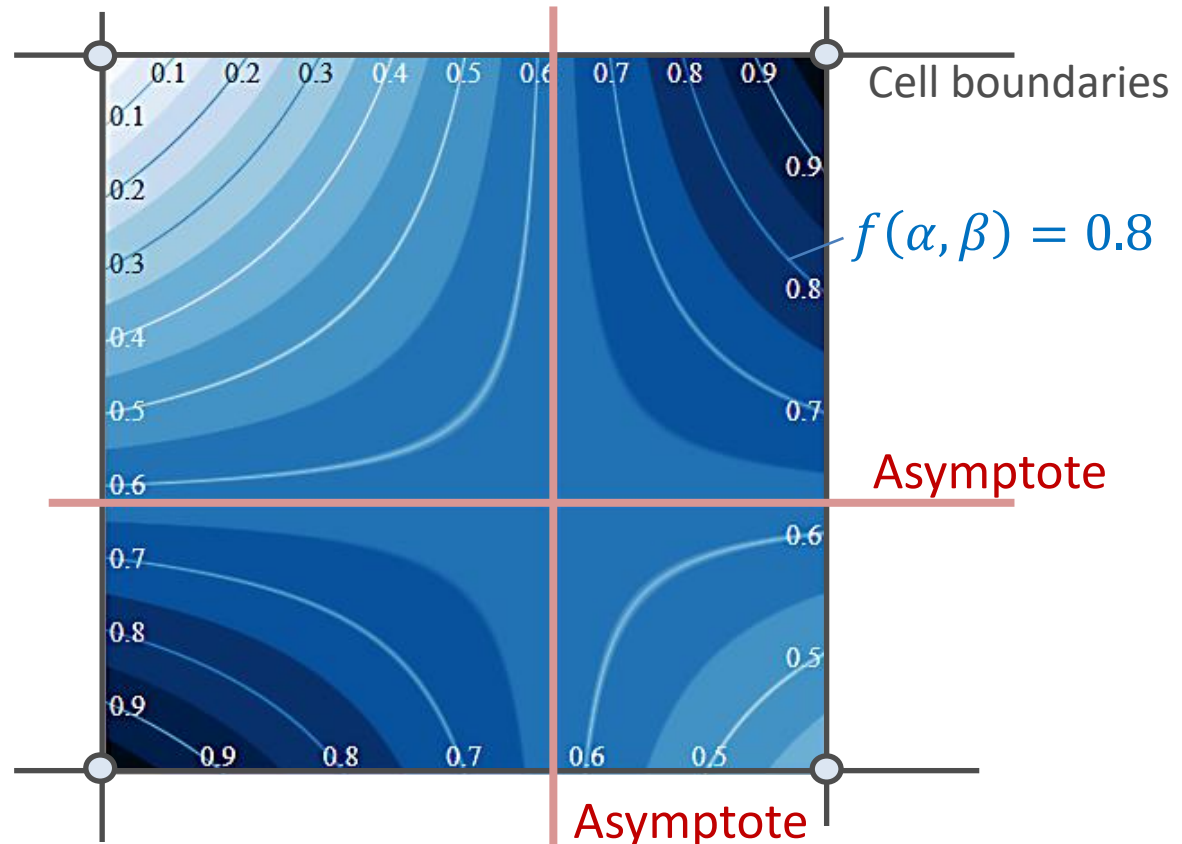
# Interpolation on grids

- Example
  - Data values given at 9 grid vertices
  - Bilinear interpolation used within grid cells



# Interpolation on grids

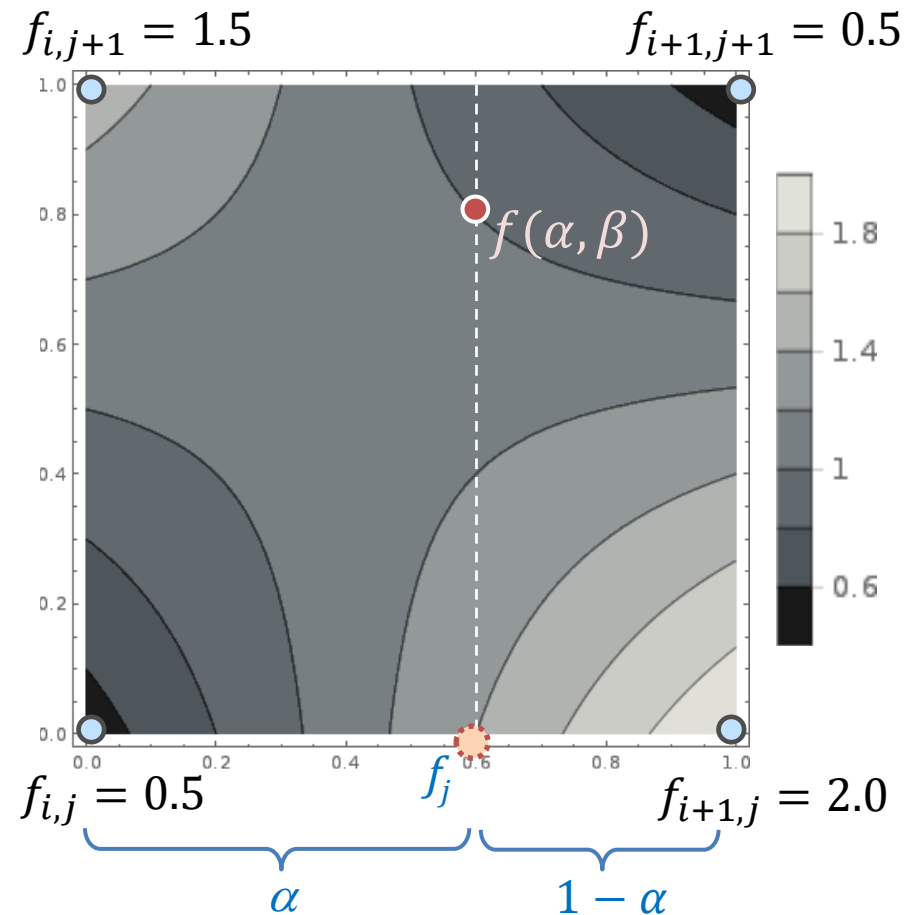
- When bilinear interpolation is used, isolines within a cell are **hyperbolas**
  - **Isoline:** curve on which all points have a certain value



# Interpolation on grids

- How to evaluate the isolines?

$$\begin{aligned}f_j &= \alpha f_{i+1,j} + (1 - \alpha)f_{i,j} \\&= f_{i,j} + (f_{i+1,j} - f_{i,j})\alpha \\&= 0.5 + 1.5\alpha\end{aligned}$$



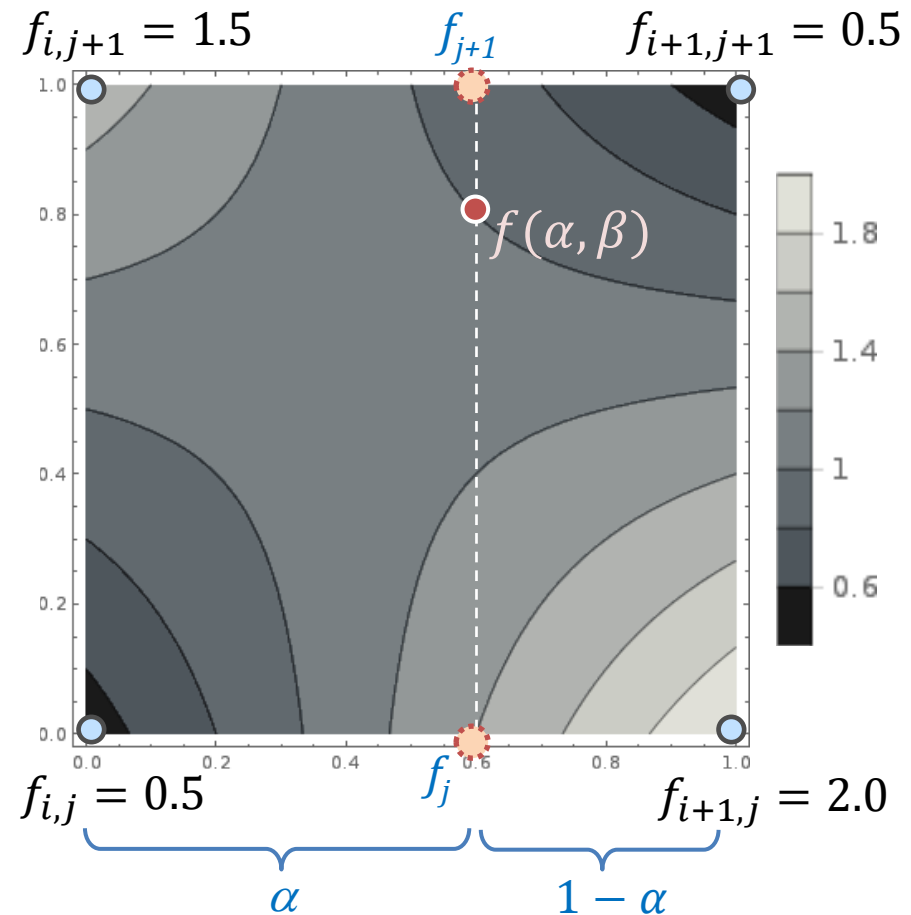


# Interpolation on grids

- How to evaluate the isolines?

$$\begin{aligned}f_j &= \alpha f_{i+1,j} + (1 - \alpha)f_{i,j} \\&= f_{i,j} + (f_{i+1,j} - f_{i,j})\alpha \\&= 0.5 + 1.5\alpha\end{aligned}$$

$$\begin{aligned}f_{j+1} &= f_{i,j+1} + (f_{i+1,j+1} - f_{i,j+1})\alpha \\&= 1.5 - \alpha\end{aligned}$$



# Interpolation on grids

- How to evaluate the isolines?

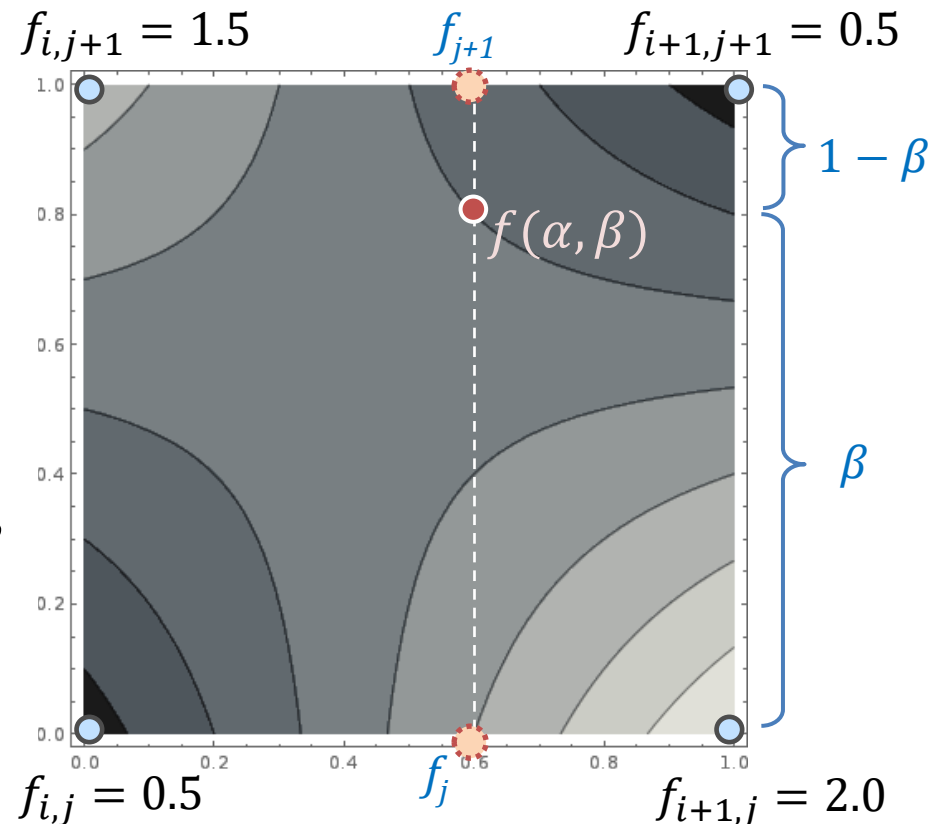
$$\begin{aligned}f_j &= \alpha f_{i+1,j} + (1 - \alpha)f_{i,j} \\&= f_{i,j} + (f_{i+1,j} - f_{i,j})\alpha \\&= 0.5 + 1.5\alpha\end{aligned}$$

$$\begin{aligned}f_{j+1} &= f_{i,j+1} + (f_{i+1,j+1} - f_{i,j+1})\alpha \\&= 1.5 - \alpha\end{aligned}$$

$$\begin{aligned}f(\alpha, \beta) &= f_j + (f_{j+1} - f_j)\beta \\&= (0.5 + 1.5\alpha) + \\&\quad (1.5 - \alpha - (0.5 + 1.5\alpha))\beta\end{aligned}$$

$$f(\alpha, \beta) = 0.5 + 1.5\alpha + \beta - 2.5\alpha\beta$$

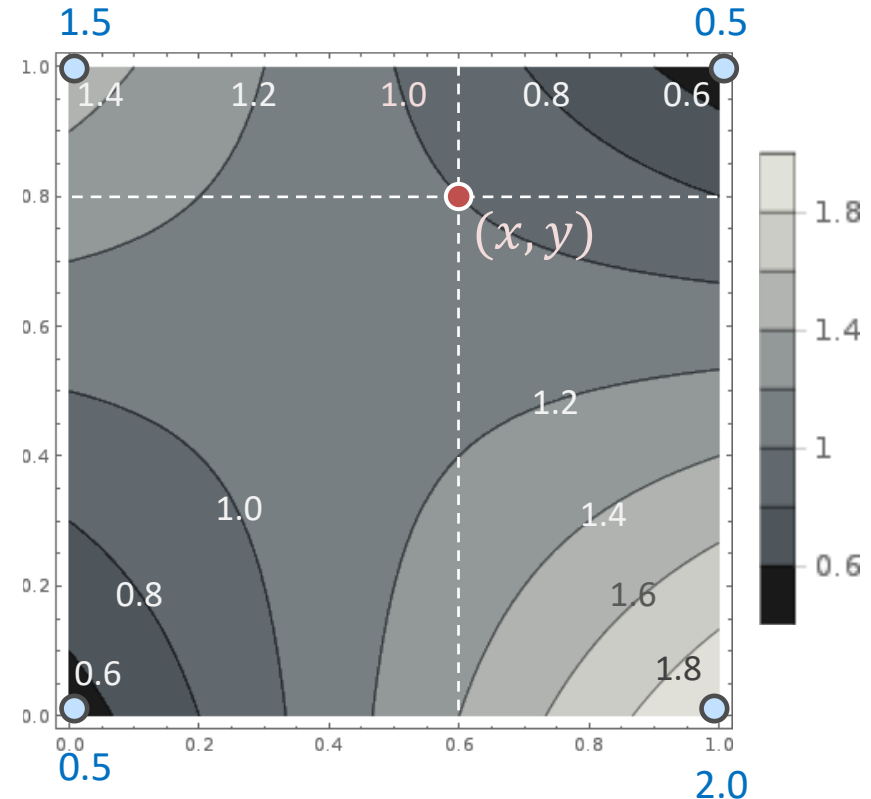
Bi-linear interpolation function  
defining the scalar value at  
each point  $(\alpha, \beta)$  within the cell



# Interpolation on grids

- How to evaluate the isolines?
  - Compute y-coordinate of a point  $(x, y)$  with  $x = 0.6$  which is on the iso-contour  $f(x, y) = 1$

$$f(x, y) = 0.5 + 1.5x + y - 2.5xy$$



# Interpolation on grids

- How to evaluate the isolines?

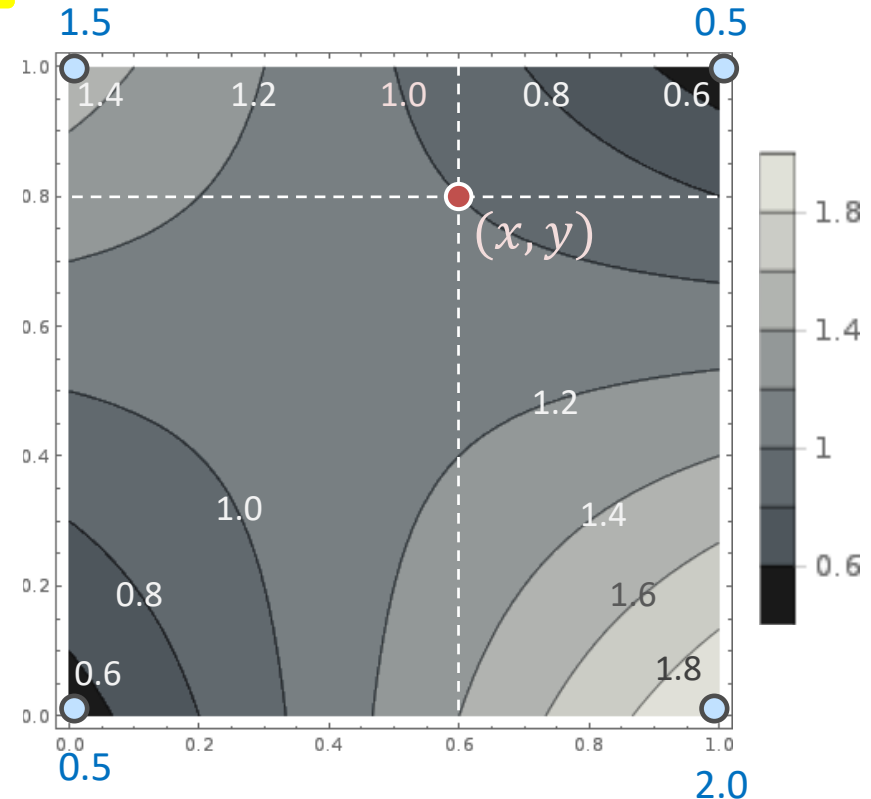
- Compute y-coordinate of a point  $(x, y)$  with  $x = 0.6$  which is on the iso-contour  $f(x, y) = 1$

$$f(x, y) = 0.5 + 1.5x + y - 2.5xy$$

$$1 = 0.5 + 1.5 \cdot 0.6 + y -$$

$$2.5 \cdot 0.6 \cdot y$$

$$1 = 1.4 - 0.5y$$

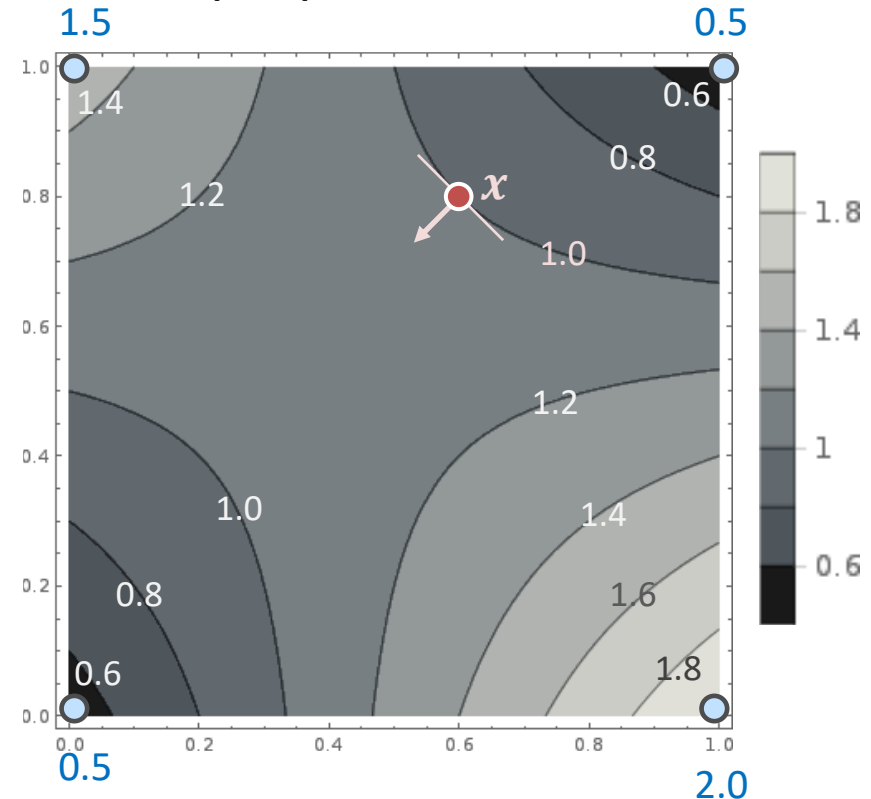


The coordinates of the point are  $(0.6, 0.8)$

# Interpolation on grids

- What is the **normal** at a point  $\mathbf{x}$  on an iso-surface?
  - It is the **gradient** at this point, which is perpendicular to the tangent of the iso-surface
  - Gradient points into direction of steepest ascent of  $f$

$$\nabla f(\mathbf{x}) = \left( \frac{\partial}{\partial x} f(\mathbf{x}), \frac{\partial}{\partial y} f(\mathbf{x}) \right)$$



# Interpolation on grids

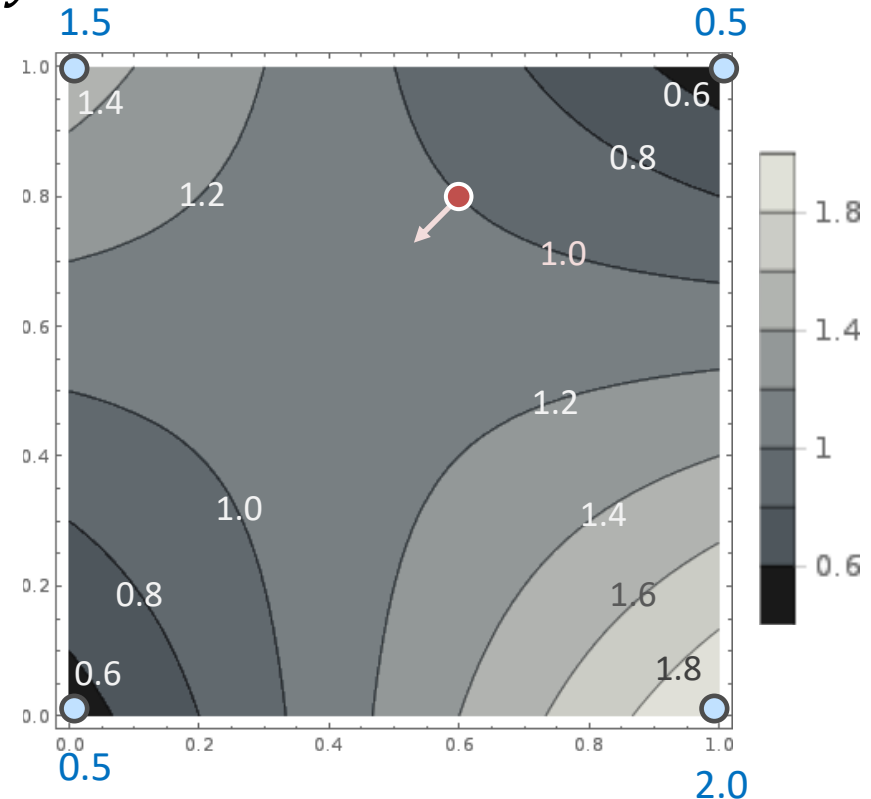
- What is the normal at point (0.6, 0.8) on the iso-surface?

$$f(x, y) = 0.5 + 1.5x + y - 2.5xy$$

$$\frac{\partial f}{\partial x} = 1.5 - 2.5y$$

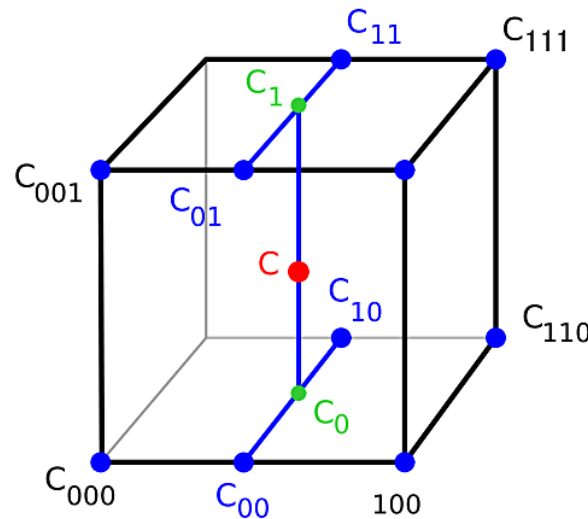
$$\frac{\partial f}{\partial y} = 1 - 2.5x$$

$$\text{Gradient at } (0.6, 0.8) : \begin{pmatrix} -0.5 \\ -0.5 \end{pmatrix}$$



# Interpolation on grids

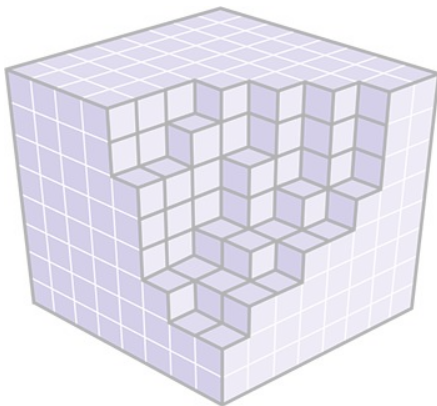
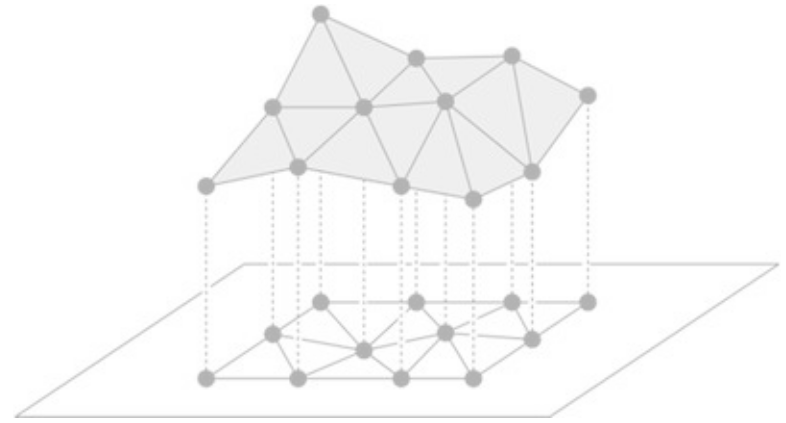
- In 3D we use **trilinear interpolation**:



- Apply linear interpolation of the initial data along the edges to obtain  $C_{00}$ ,  $C_{01}$ ,  $C_{10}$ ,  $C_{11}$
- Interpolate linearly between  $C_{00}$  and  $C_{01}$ , and between  $C_{10}$  and  $C_{11}$  to obtain  $C_0$  and  $C_1$
- Finally, interpolate between  $C_1$  and  $C_0$  to obtain  $C$ .

# Overview

- Interpolation on grids / meshes
  - Barycentric interpolation
  - Bi-/trilinear interpolation
  - High-quality reconstruction

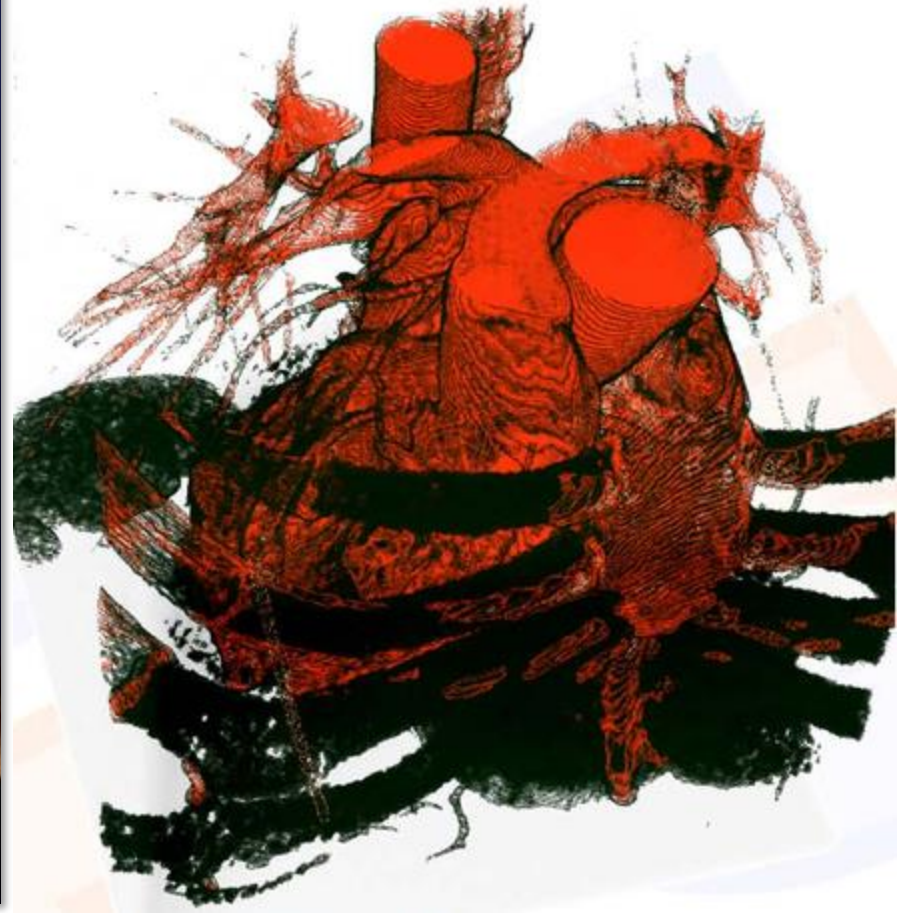


Tricubic vs. trilinear interpolation



# Interpolation on grids

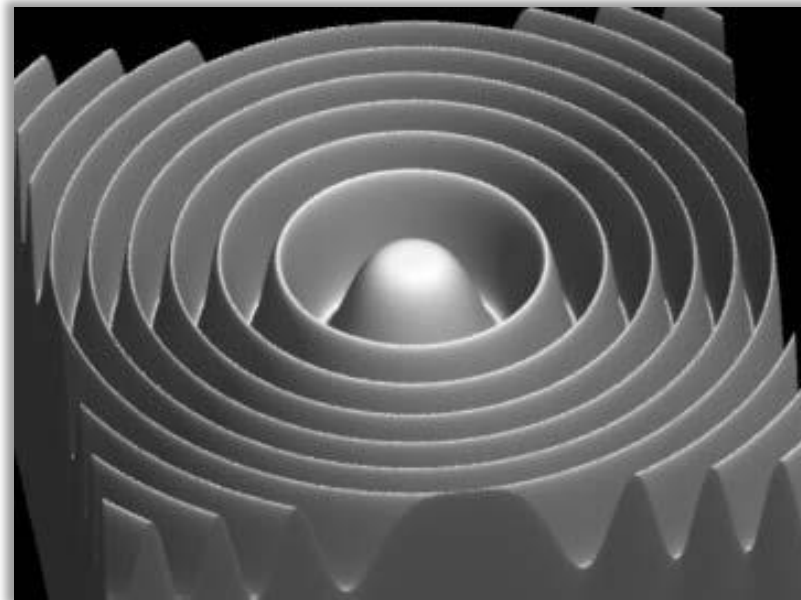
- This is what we want...



...but sometimes this is what we get

# Interpolation on grids

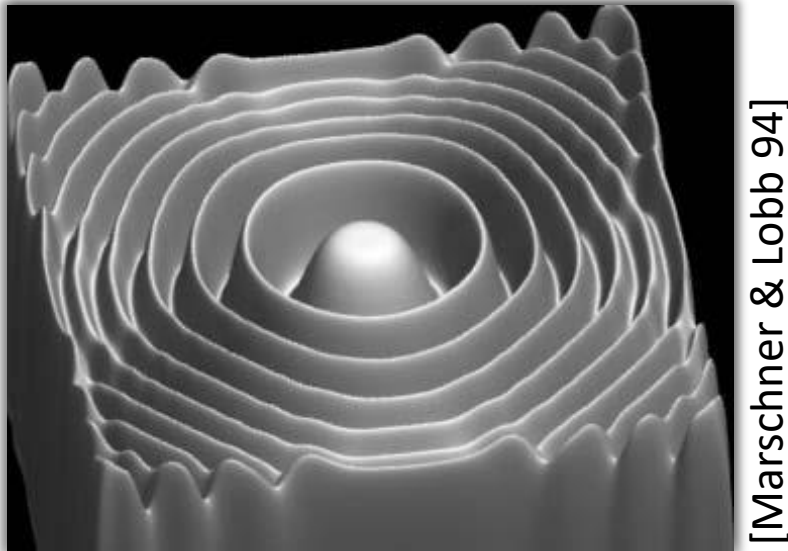
- Higher-order reconstruction/interpolation
  - Required if very high quality is needed
  - Usually tested on Marschner-Lobb function
    - High amount of its energy is near its Nyquist frequency
    - Very demanding test for accurate reconstruction



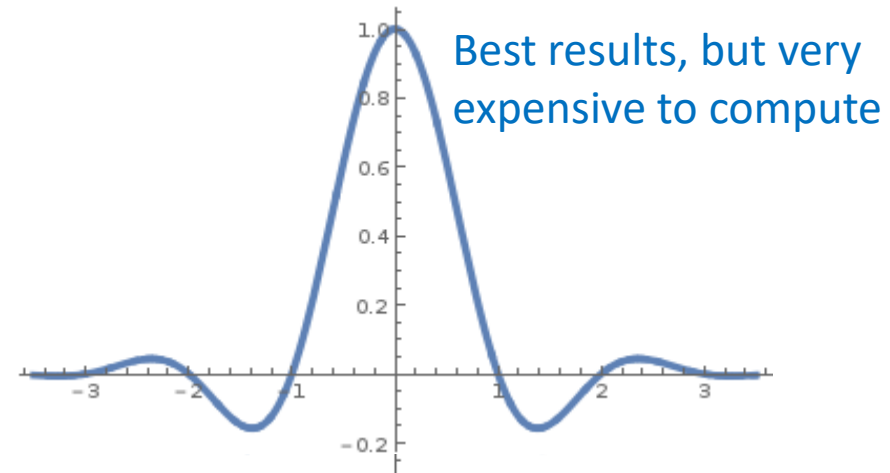
[Marschner & Lobb 94]

# High-quality Reconstruction

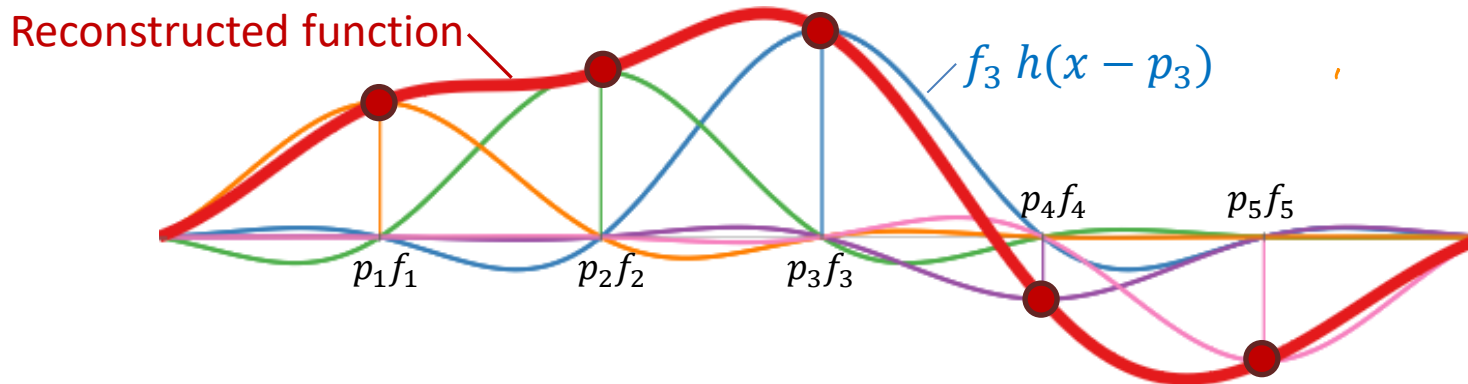
- Sinc function



Windowed sinc (Hann window)

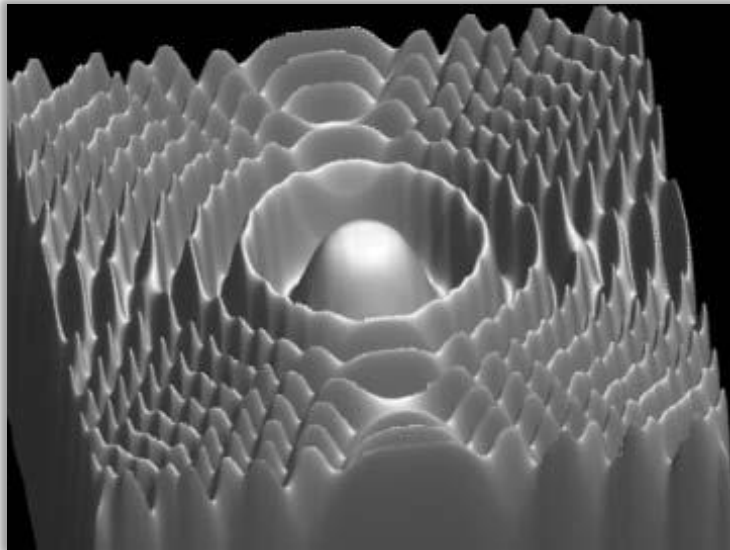


**“Optimal”** reconstruction filter

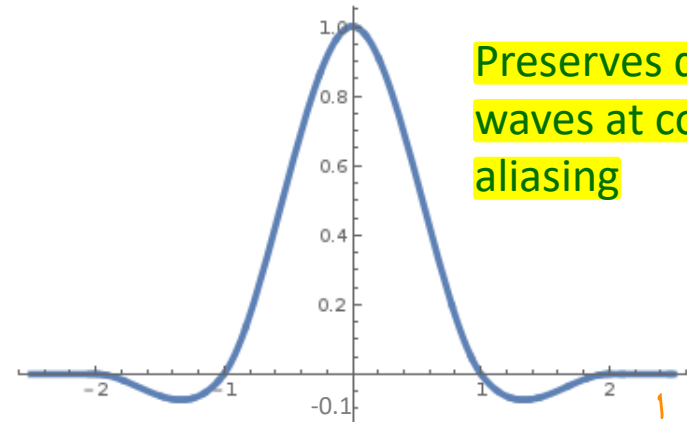


# High-quality Reconstruction

- Bicubic interpolation (Catmull-Rom spline)



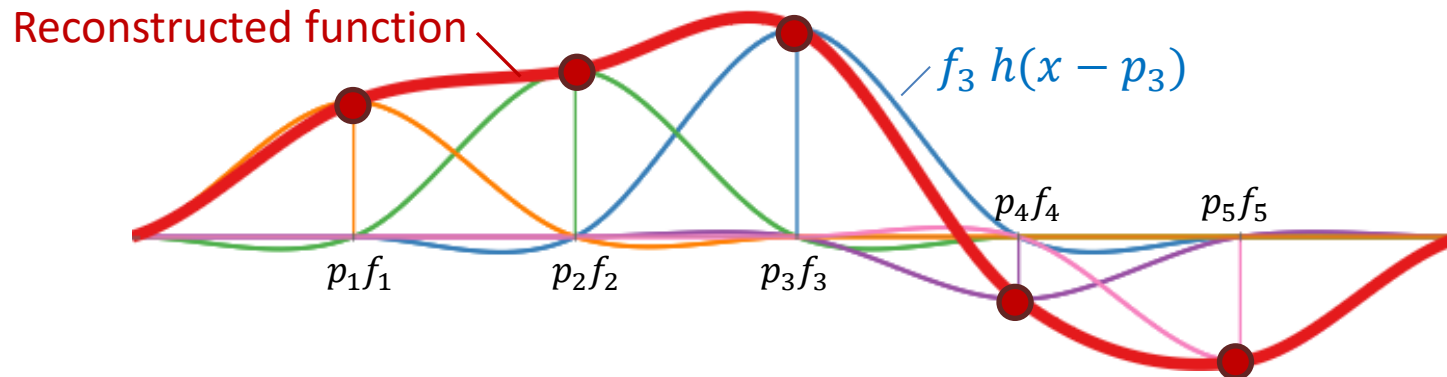
[Marschner & Lobb 94]



Preserves depth of  
waves at cost of  
aliasing

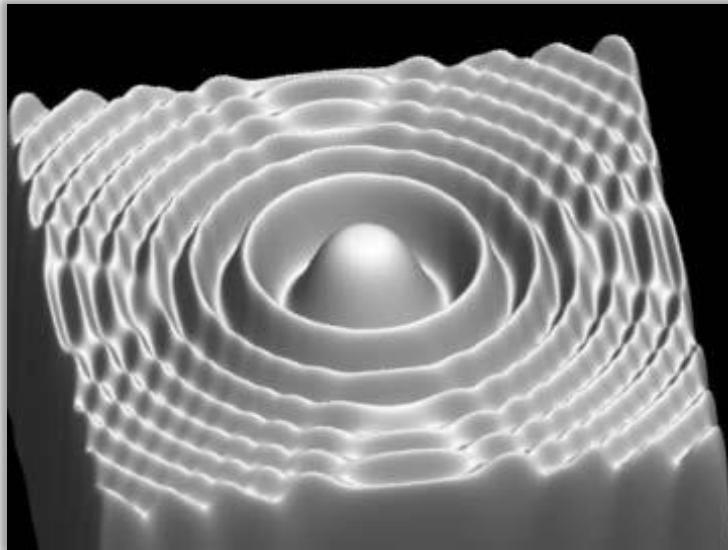
Reconstructed Marschner-Lobb function

Interpolation: 1 at center and 0 at integers



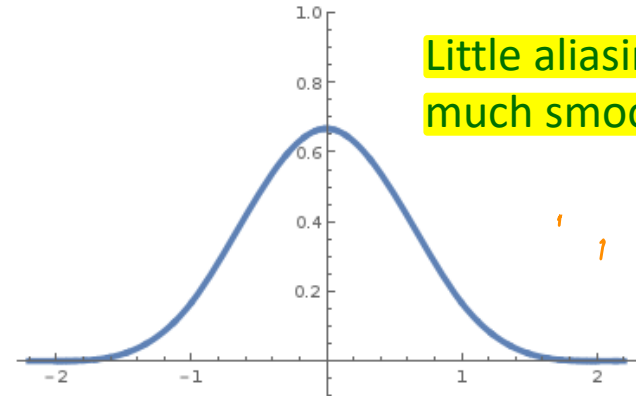
# High-quality Reconstruction

- Cubic B-spline (with smoothing)



[Marschner & Lobb 94]

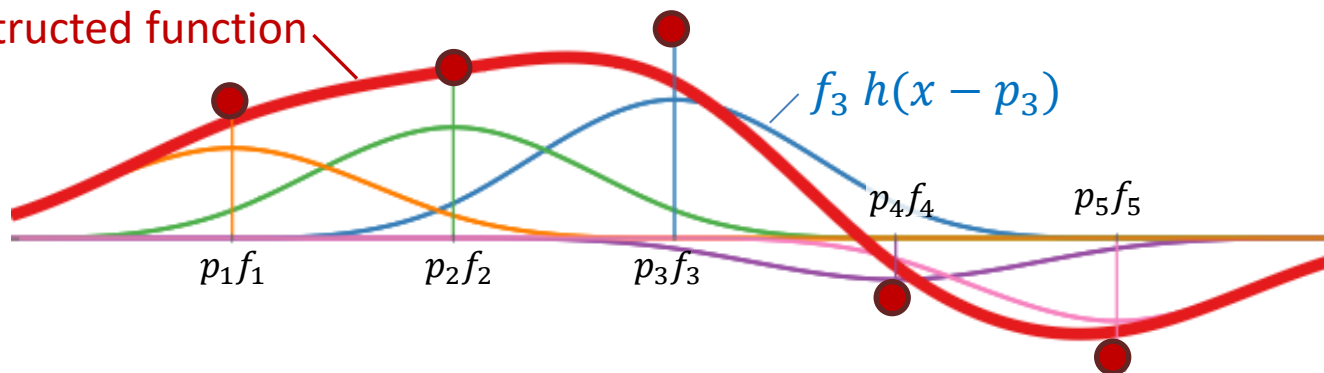
Reconstructed Marschner-Lobb function



Little aliasing, but  
much smoothing

Smoothing:  $\frac{2}{3}$  at center and  $\frac{1}{6}$  at  $\pm 1$

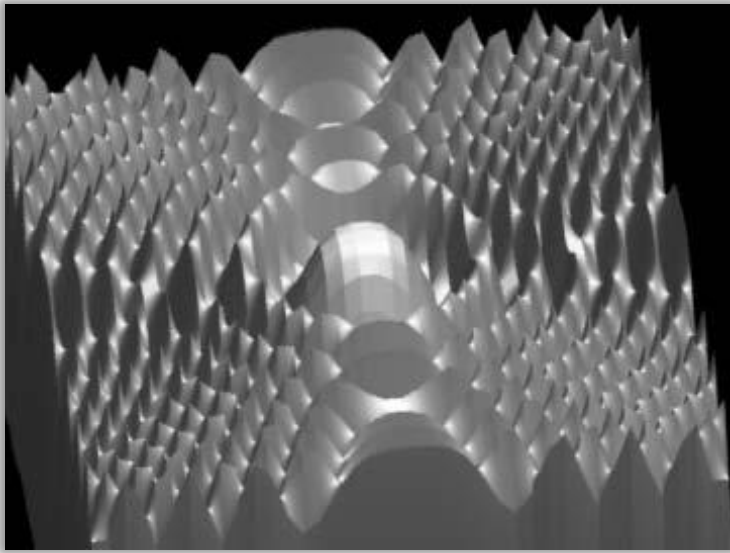
Reconstructed function



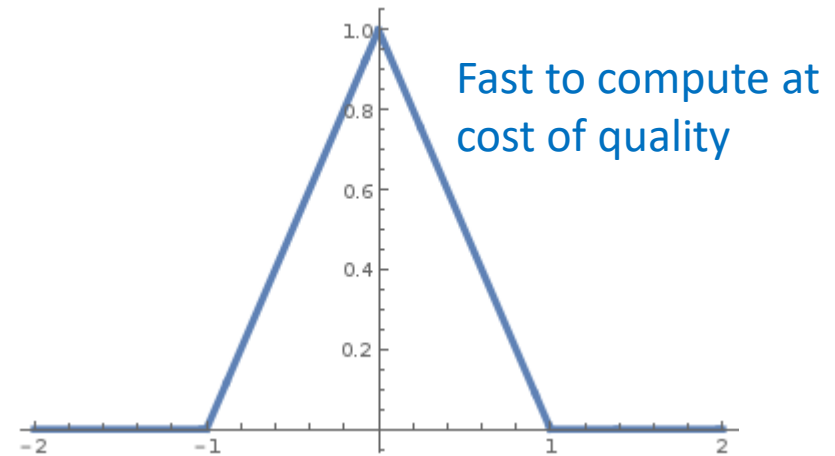


# High-quality Reconstruction

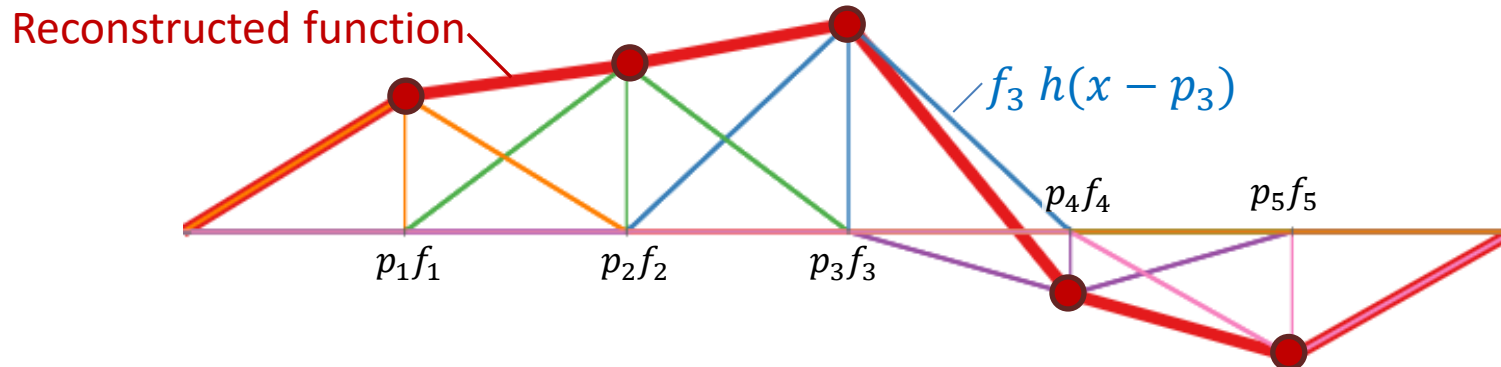
- Trilinear interpolation (Tent)



Reconstructed Marschner-Lobb function

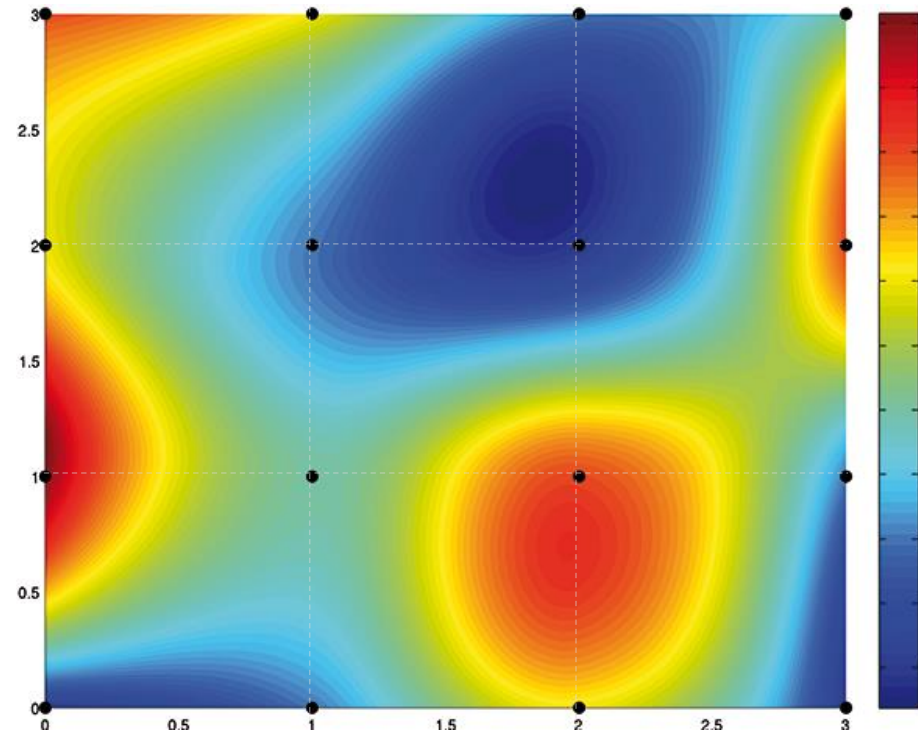


Interpolation: 1 at center and 0 at integers



# Interpolation on grids

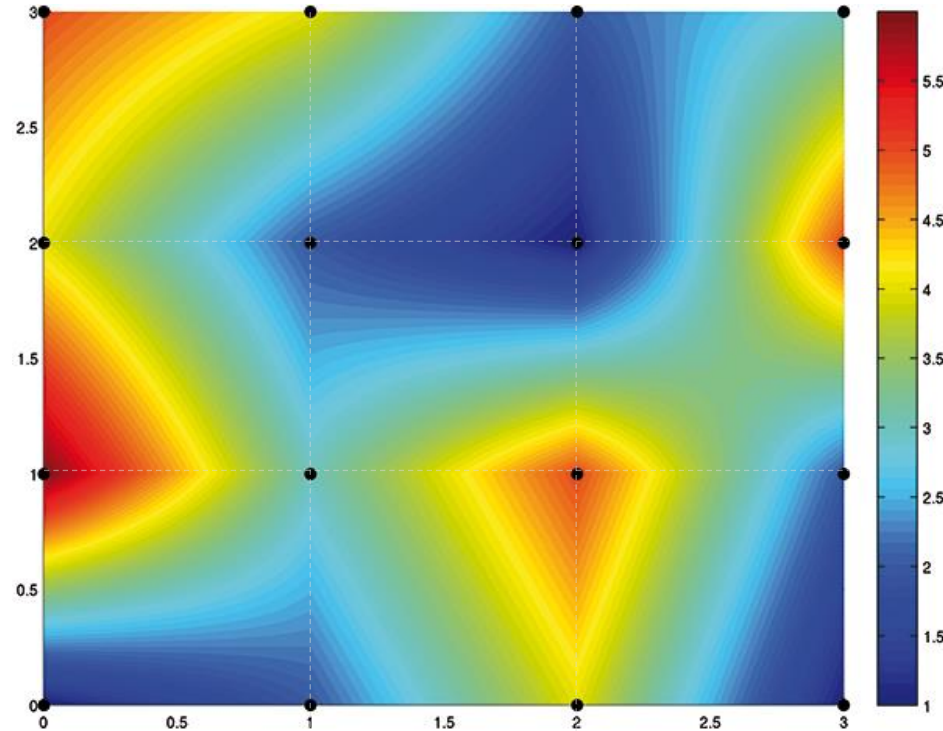
- 2D interpolation with different reconstruction filters



Bicubic interpolation (default in Photoshop)

$C^1$  continuous  $\rightarrow$  tangents match  
at segment transition

Cubic B-spline ( $C^2$  continuous  
 $\rightarrow$  also curvature matches)



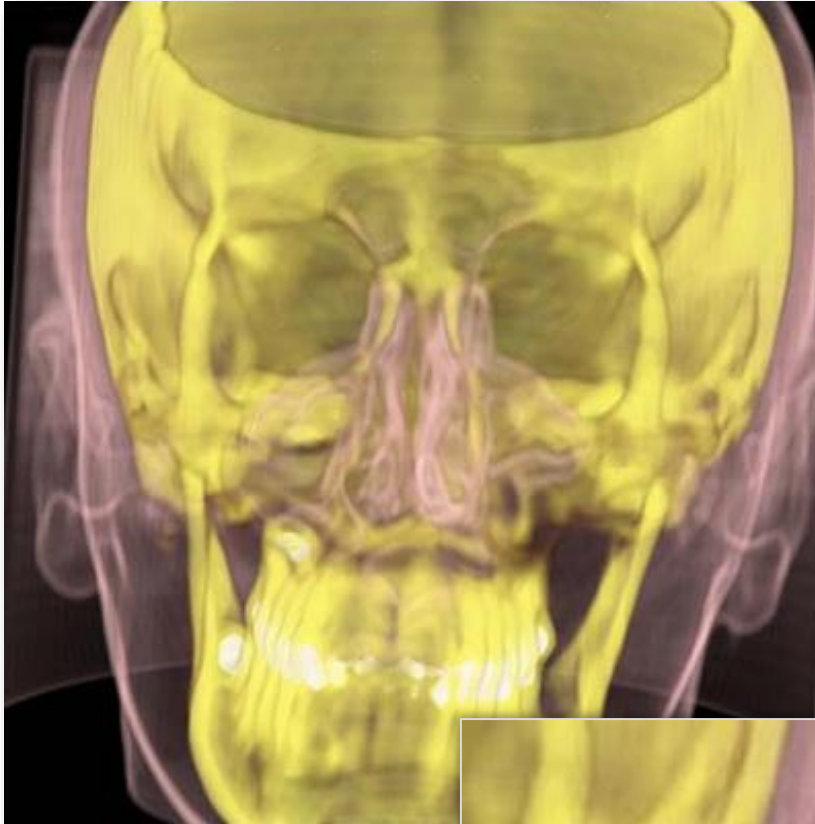
Bilinear interpolation

( $C^0$ ) continuous  $\rightarrow$  values match  
at segment transition)

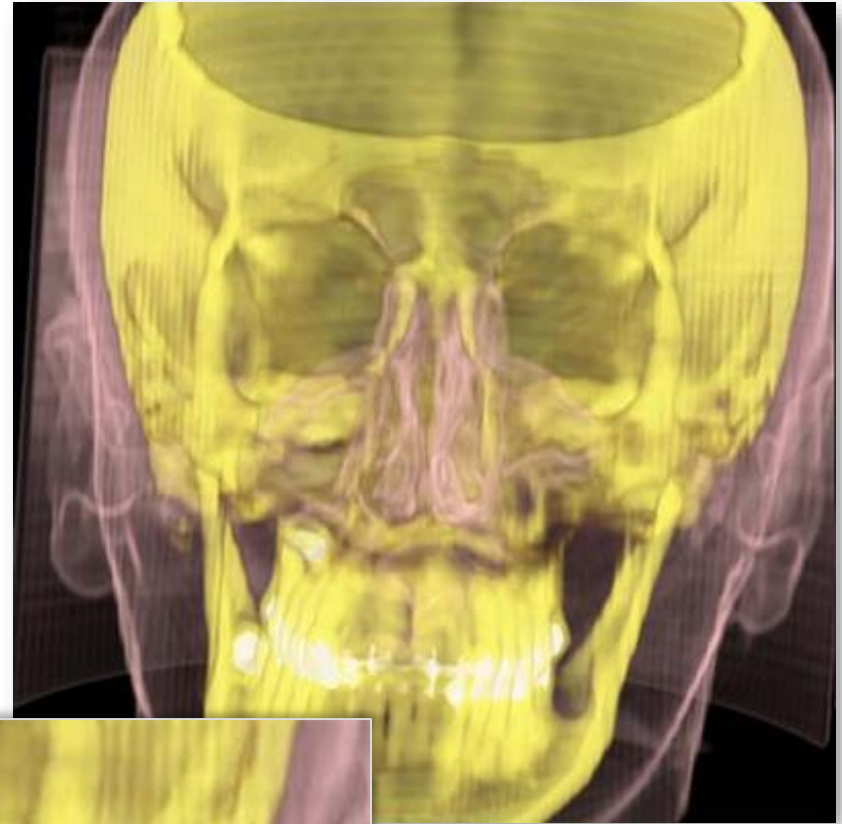
No derivative

# Interpolation on grids

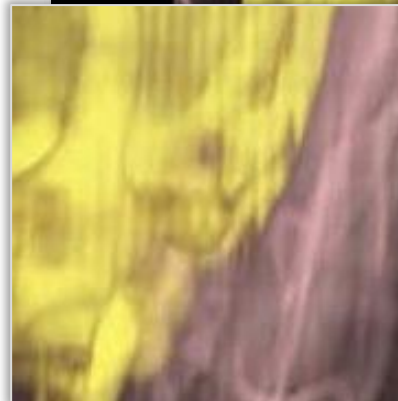
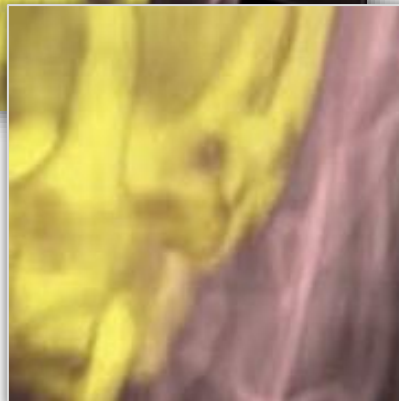
- Volume rendering with different reconstruction filters



Tricubic



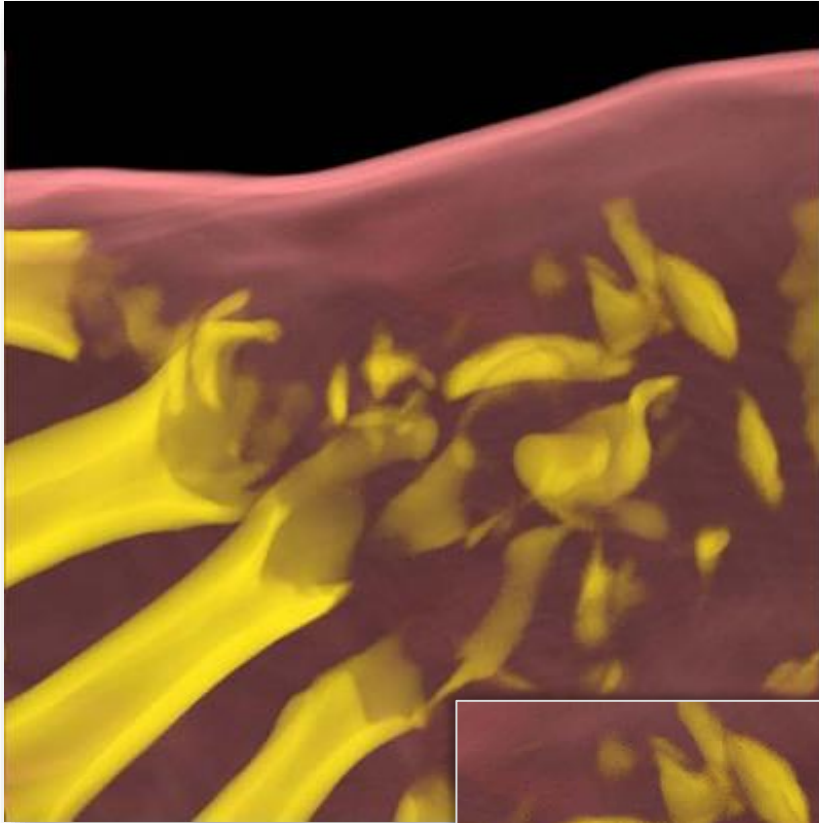
Trilinear



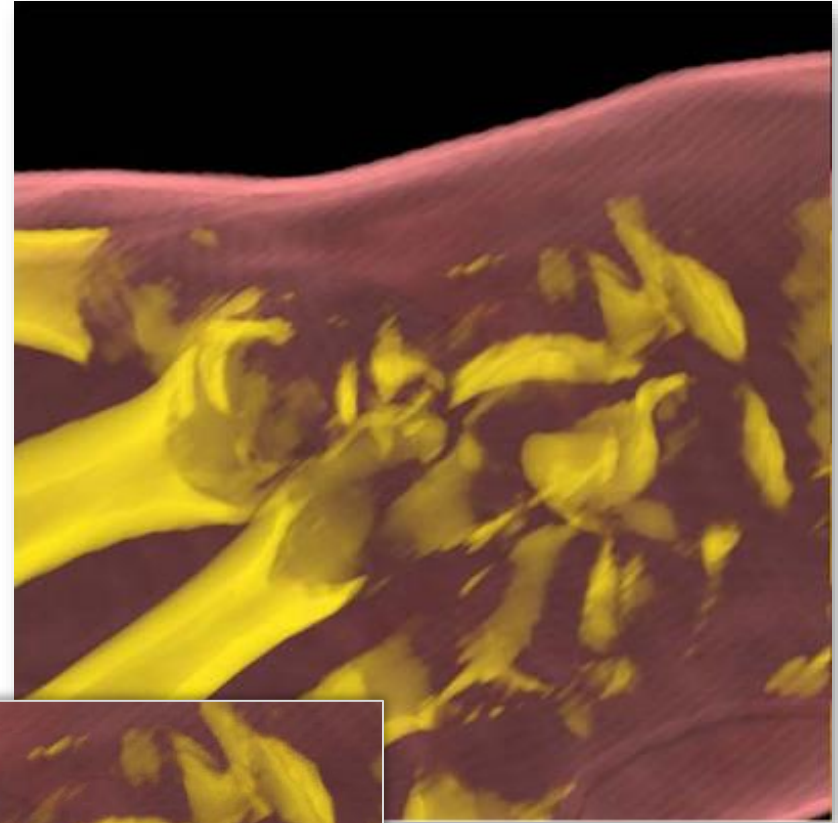


# Interpolation on grids

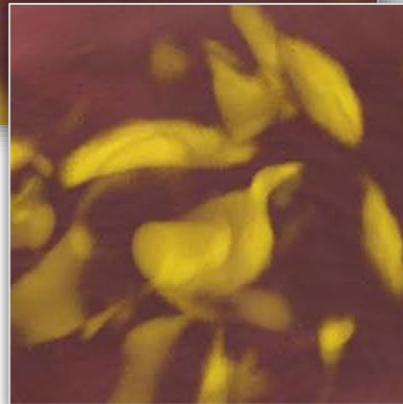
- Volume rendering with different reconstruction filters



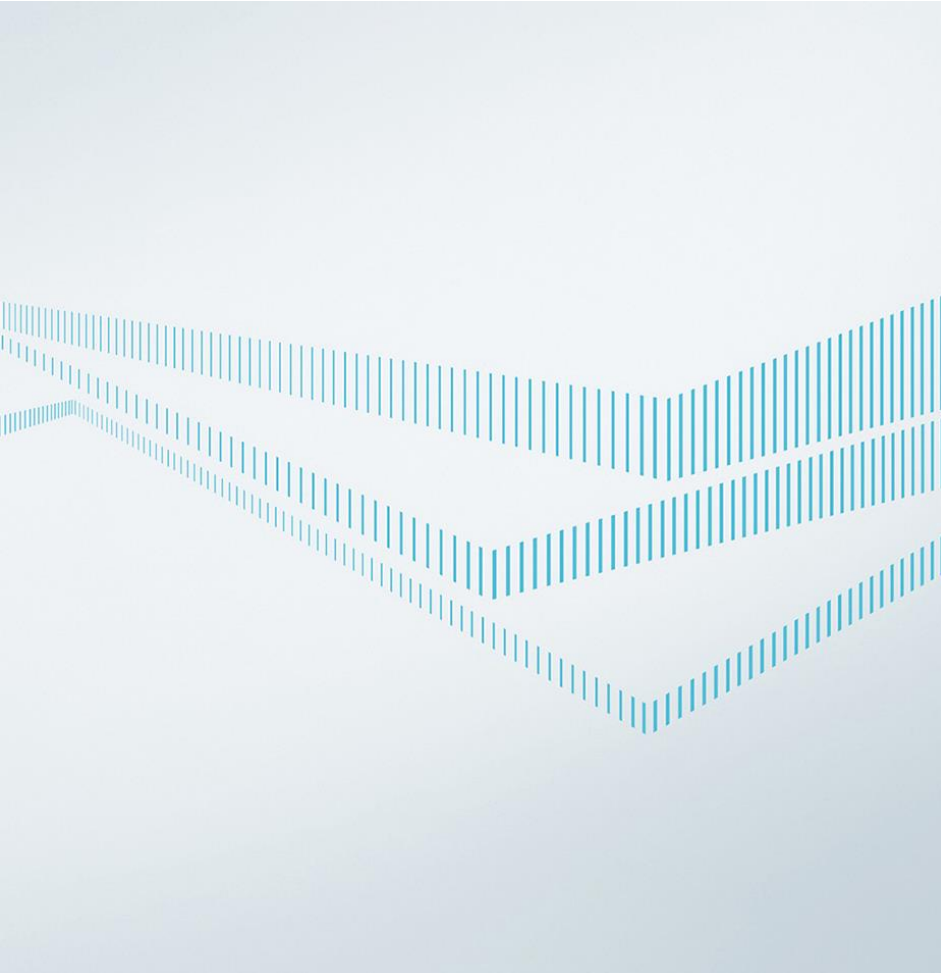
Tricubic



Trilinear



# Contact information



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