

KAUST

CS 247 – Scientific Visualization Lecture 9: Scalar Field Visualization, Pt. 3

Reading Assignment #5 (until Mar 5)



Read (required):

· Gradients of scalar-valued functions

https://en.wikipedia.org/wiki/Gradient

• Critical points

https://en.wikipedia.org/wiki/Critical_point_(mathematics)

• Multivariable derivatives and differentials

https://en.wikipedia.org/wiki/Total_derivative

```
https://en.wikipedia.org/wiki/Differential_of_a_function#
Differentials_in_several_variables
```

https://en.wikipedia.org/wiki/Hessian_matrix

• Dot product, inner product (more general)

https://en.wikipedia.org/wiki/Dot_product

https://en.wikipedia.org/wiki/Inner_product_space

Quiz #1: Mar 1



Organization

- First 30 min of lecture
- No material (book, notes, ...) allowed

Content of questions

- Lectures (both actual lectures and slides)
- Reading assignments (except optional ones)
- Programming assignments (algorithms, methods)
- Solve short practical examples

Linear Interpolation / Convex Combinations



Linear combination (*n*-dim. space):

$$\alpha_1v_1 + \alpha_2v_2 + \ldots + \alpha_nv_n = \sum_{i=1}^n \alpha_iv_i$$

Affine combination: Restrict to (n-1)-dim. subspace:

$$lpha_1+lpha_2+\ldots+lpha_n=\sum_{i=1}^nlpha_i=1$$

 v_3 v_1 Q v_2

Convex combination:

 $\alpha_i \ge 0$

(restrict to simplex in subspace)

Linear Interpolation / Convex Combinations

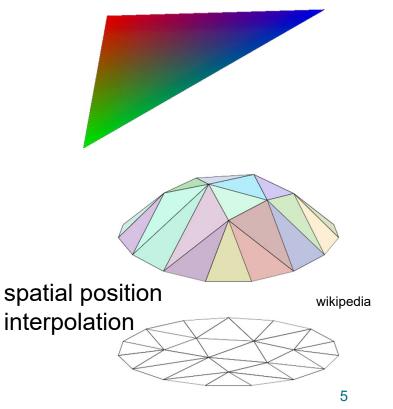


The weights α_i are the *n* normalized **barycentric** coordinates

 \rightarrow linear attribute interpolation in simplex

$$lpha_1 v_1 + lpha_2 v_2 + \ldots + lpha_n v_n = \sum_{i=1}^n lpha_i v_i$$
 $lpha_1 + lpha_2 + \ldots + lpha_n = \sum_{i=1}^n lpha_i = 1$
 $lpha_i \ge 0$

attribute interpolation



Linear Interpolation / Convex Combinations



$$lpha_1 v_1 + lpha_2 v_2 + \ldots + lpha_n v_n = \sum_{i=1}^n lpha_i v_i$$
 $lpha_1 + lpha_2 + \ldots + lpha_n = \sum_{i=1}^n lpha_i = 1$

Can re-parameterize to get (n-1) *affine* coordinates:

$$\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3 =$$

$$\tilde{\alpha}_1 (v_2 - v_1) + \tilde{\alpha}_2 (v_3 - v_1) + v_1$$

$$\tilde{\alpha}_1 = \alpha_2$$

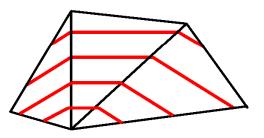
$$\tilde{\alpha}_2 = \alpha_3$$

 v_3 v_1 Q v_2

Markus Hadwiger

Contours in triangle/tetrahedral cells

Linear interpolation of cells implies piece-wise linear contours.



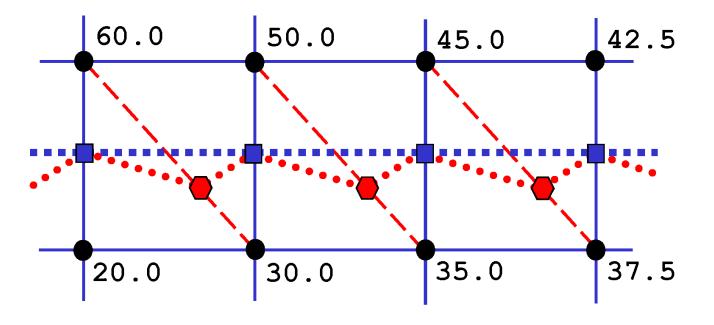
Contours are unambiguous, making "marching triangles" even simpler than "marching squares".

Question: Why not split quadrangles into two triangles (and hexahedra into five or six tetrahedra) and use marching triangles (tetrahedra)?

Answer: This can introduce periodic artifacts!

Contours in triangle/tetrahedral cells

Illustrative example: Find contour at level c=40.0 !



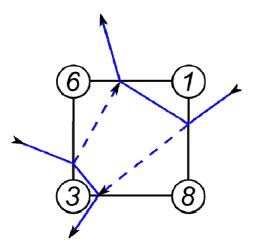
original quad grid, yielding vertices and contour
 triangulated grid, yielding vertices and contour

Ambiguities of contours

What is the **correct** contour of *c*=4?

Two possibilities, both are orientable:

- connect high values _____
- connect low values



Answer: correctness depends on interior values of f(x).

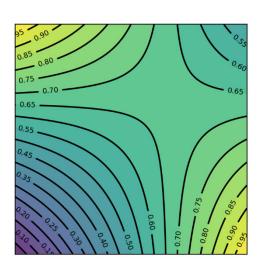
But: different interpolation schemes are possible.

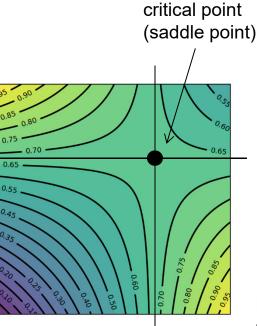
Better question: What is the correct contour with respect to bilinear interpolation?

Ronald Peikert



Critical points are where the gradient vanishes (i.e., is the zero vector)



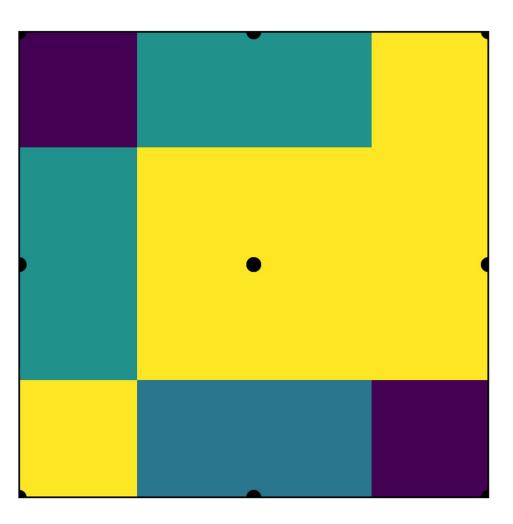


here, the critical value is 2/3=0.666...

"Asymptotic decider": resolve ambiguous configurations (6 and 9) by comparing specific iso-value with critical value (scalar value at critical point)



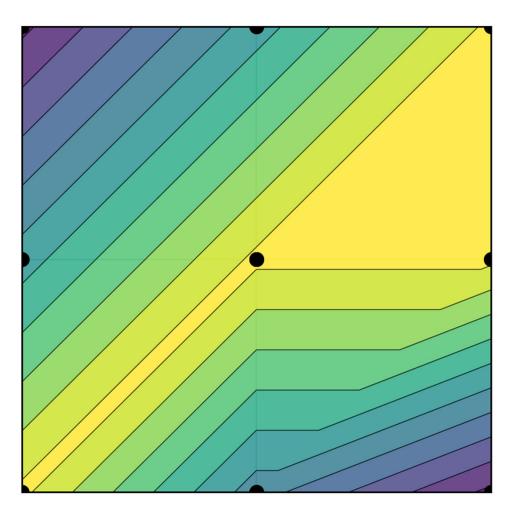
nearest-neighbor





piecewise linear

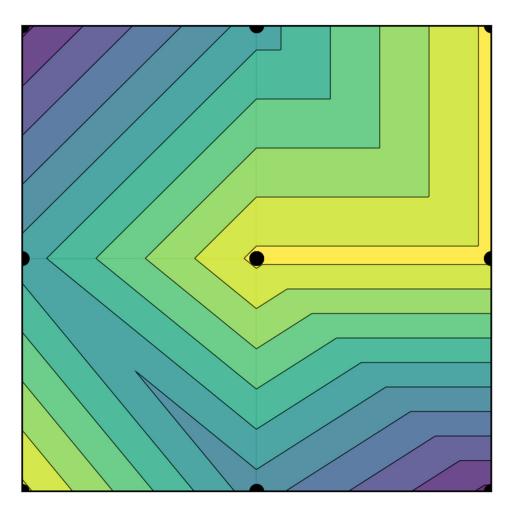
(2 triangles per quad; diagonal: bottom-left, top-right)





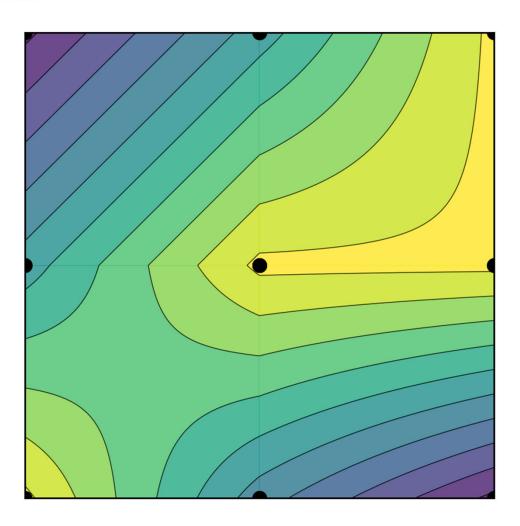
piecewise linear

(2 triangles per quad; diagonal: top-left, bottom-right)





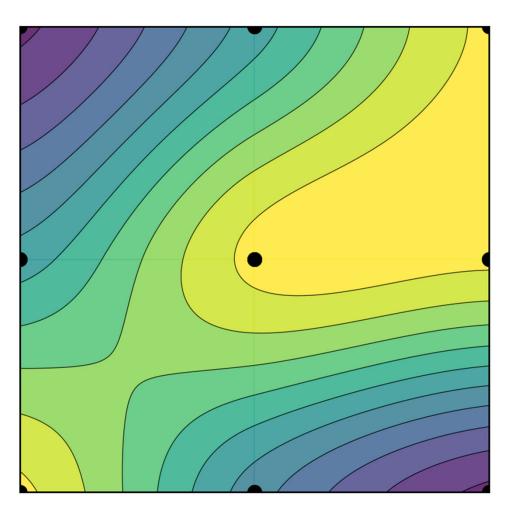
bi-linear





bi-cubic

(Catmull-Rom spline)



Texture Filtering Example (Magnification)





Original texture image



Nearest neighbor



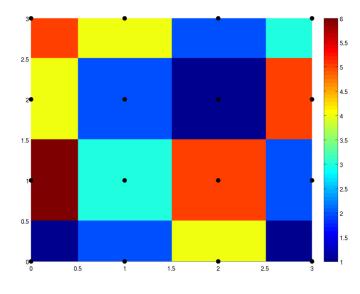
Bi-linear filtering



Eduard Gröller, Stefan Jeschke

Bi-Linear Interpolation vs. Nearest Neighbor



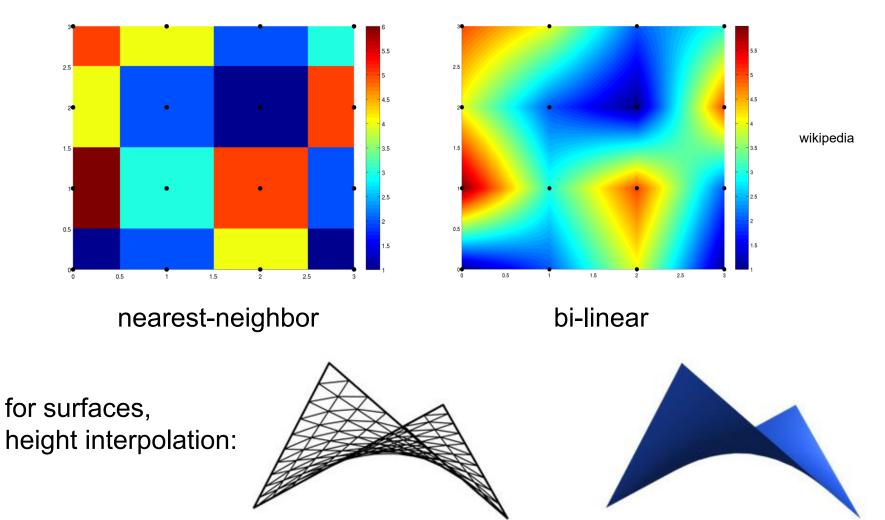


nearest-neighbor

bi-linear

Bi-Linear Interpolation vs. Nearest Neighbor

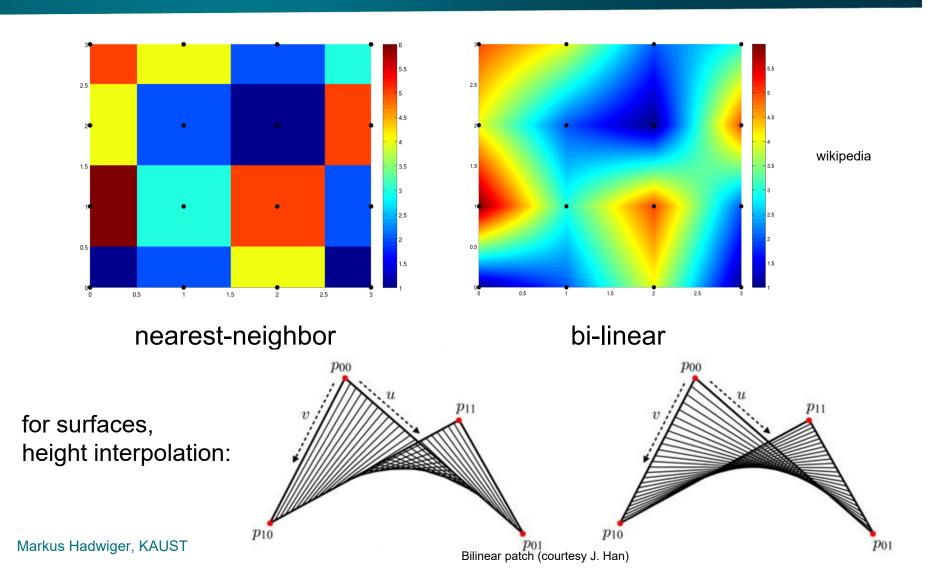




Bilinear patch (courtesy J. Han)

Bi-Linear Interpolation vs. Nearest Neighbor

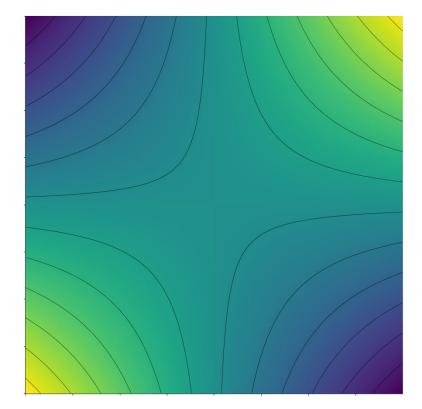


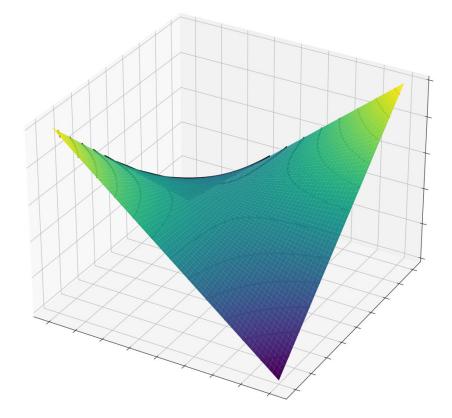




Consider area between 2x2 adjacent samples (e.g., pixel centers)

Example #1: 1 at bottom-left and top-right, 0 at top-left and bottom-right

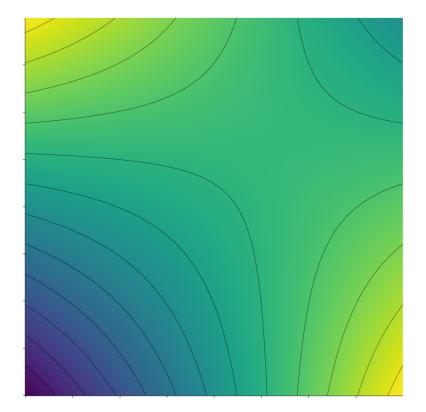


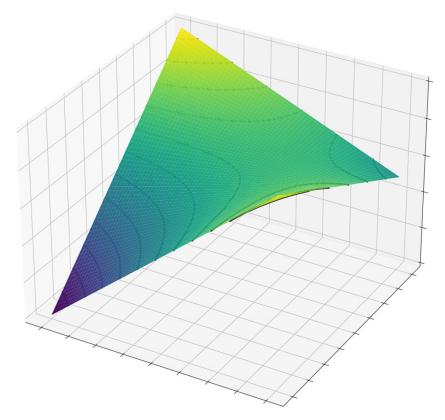




Consider area between 2x2 adjacent samples (e.g., pixel centers)

Example #2: 1 at top-left and bottom-right, 0 at bottom-left, 0.5 at top-right







Consider area between 2x2 adjacent samples (e.g., pixel centers):

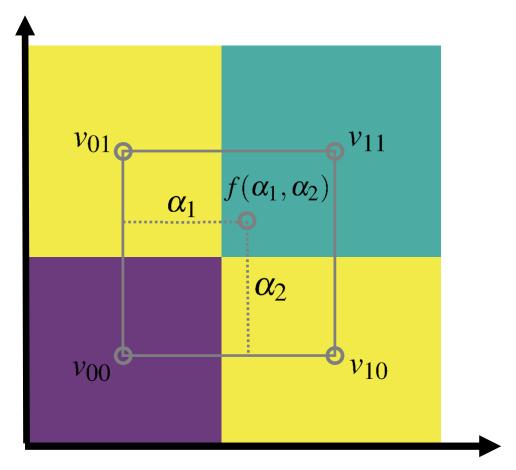
Given any (fractional) position

$\alpha_1 := x_1 - \lfloor x_1 \rfloor$	$lpha_1 \in [0.0, 1.0)$
$\alpha_2 := x_2 - \lfloor x_2 \rfloor$	$lpha_2 \in [0.0, 1.0)$

and 2x2 sample values

$$\begin{bmatrix} v_{01} & v_{11} \\ v_{00} & v_{10} \end{bmatrix}$$

Compute: $f(\alpha_1, \alpha_2)$





Consider area between 2x2 adjacent samples (e.g., pixel centers):

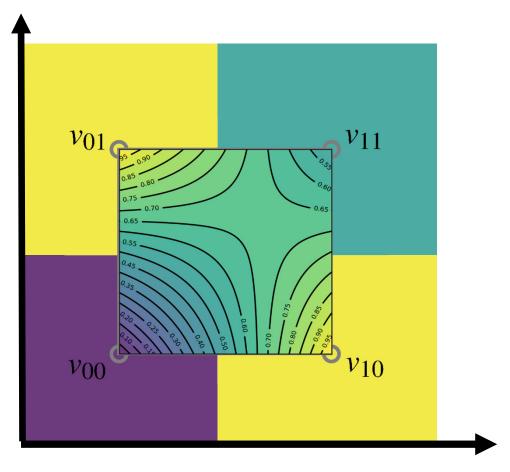
Given any (fractional) position

$\alpha_1 := x_1 - \lfloor x_1 \rfloor$	$\alpha_1 \in [0.0, 1.0)$
$\alpha_2 := x_2 - \lfloor x_2 \rfloor$	$lpha_2 \in [0.0, 1.0)$

and 2x2 sample values

$$\begin{bmatrix} v_{01} & v_{11} \\ v_{00} & v_{10} \end{bmatrix}$$

Compute: $f(\alpha_1, \alpha_2)$





Interpolate function at (fractional) position (α_1, α_2):

$$f(\boldsymbol{\alpha}_1, \boldsymbol{\alpha}_2) = \begin{bmatrix} \boldsymbol{\alpha}_2 & (1 - \boldsymbol{\alpha}_2) \end{bmatrix} \begin{bmatrix} v_{01} & v_{11} \\ v_{00} & v_{10} \end{bmatrix} \begin{bmatrix} (1 - \boldsymbol{\alpha}_1) \\ \boldsymbol{\alpha}_1 \end{bmatrix}$$

$$= \begin{bmatrix} \alpha_2 & (1 - \alpha_2) \end{bmatrix} \begin{bmatrix} (1 - \alpha_1)v_{01} + \alpha_1v_{11} \\ (1 - \alpha_1)v_{00} + \alpha_1v_{10} \end{bmatrix}$$

$$= \begin{bmatrix} \alpha_2 v_{01} + (1 - \alpha_2) v_{00} & \alpha_2 v_{11} + (1 - \alpha_2) v_{10} \end{bmatrix} \begin{bmatrix} (1 - \alpha_1) \\ \alpha_1 \end{bmatrix}$$



Interpolate function at (fractional) position (α_1, α_2):

$$f(\boldsymbol{\alpha}_1, \boldsymbol{\alpha}_2) = \begin{bmatrix} \boldsymbol{\alpha}_2 & (1 - \boldsymbol{\alpha}_2) \end{bmatrix} \begin{bmatrix} v_{01} & v_{11} \\ v_{00} & v_{10} \end{bmatrix} \begin{bmatrix} (1 - \boldsymbol{\alpha}_1) \\ \boldsymbol{\alpha}_1 \end{bmatrix}$$

$$= (1 - \alpha_1)(1 - \alpha_2)v_{00} + \alpha_1(1 - \alpha_2)v_{10} + (1 - \alpha_1)\alpha_2v_{01} + \alpha_1\alpha_2v_{11}$$

$$= v_{00} + \alpha_1(v_{10} - v_{00}) + \alpha_2(v_{01} - v_{00}) + \alpha_1\alpha_2(v_{00} + v_{11} - v_{10} - v_{01})$$

Bi-Linear Interpolation: Contours

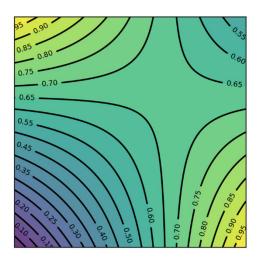


Find one specific iso-contour (can of course do this for any/all isovalues):

$$f(\boldsymbol{\alpha}_1, \boldsymbol{\alpha}_2) = c$$

Find all (α_1, α_2) where:

 $v_{00} + \alpha_1(v_{10} - v_{00}) + \alpha_2(v_{01} - v_{00}) + \alpha_1\alpha_2(v_{00} + v_{11} - v_{10} - v_{01}) = c$



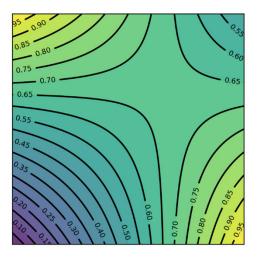


Compute gradient (critical points are where gradient is zero vector):

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = (v_{10} - v_{00}) + \alpha_2(v_{00} + v_{11} - v_{10} - v_{01})$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = (v_{01} - v_{00}) + \alpha_1(v_{00} + v_{11} - v_{10} - v_{01})$$

Where are lines of constant value / critical points?

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = 0: \qquad \alpha_2 = \frac{v_{00} - v_{10}}{v_{00} + v_{11} - v_{10} - v_{01}}$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = 0: \qquad \alpha_1 = \frac{v_{00} - v_{01}}{v_{00} + v_{11} - v_{10} - v_{01}}$$





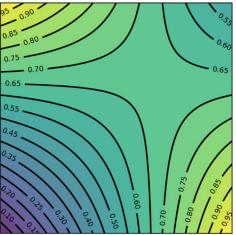
Compute gradient (critical points are where gradient is zero vector):

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = (v_{10} - v_{00}) + \alpha_2(v_{00} + v_{11} - v_{10} - v_{01})$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = (v_{01} - v_{00}) + \alpha_1(v_{00} + v_{11} - v_{10} - v_{01})$$

Where are lines of constant value / critical points?

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = 0: \qquad \alpha_2 = \frac{v_{00} - v_{10}}{v_{00} + v_{11} - v_{10} - v_{01}}$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = 0: \qquad \alpha_2 = \frac{v_{00} - v_{10}}{v_{00} - v_{01}}$$

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = 0: \qquad \alpha_1 = \frac{v_{00} - v_{01}}{v_{00} + v_{11} - v_{10} - v_{01}}$$



if denominator is zero, bi-linear interpolation has degenerated to linear interpolation (or const)! (also means: no isolated critical points!)

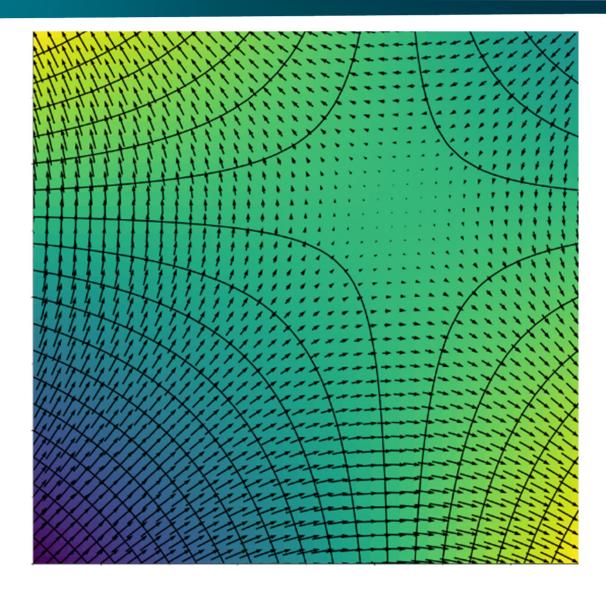


Compute gradient

Note that isolines are farther apart where gradient is smaller

Note the horizontal and vertical lines where gradient becomes vertical/horizontal

Note the critical point



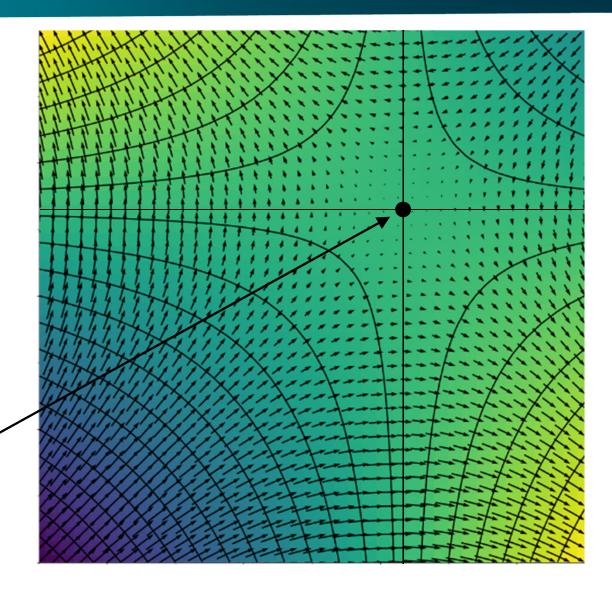


Compute gradient

Note that isolines are farther apart where gradient is smaller

Note the horizontal and vertical lines where gradient becomes vertical/horizontal

Note the critical point



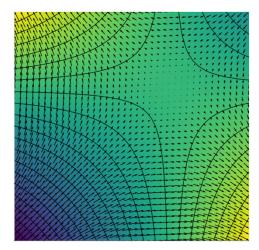


Compute gradient (critical points are where gradient is zero vector):

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = (v_{10} - v_{00}) + \alpha_2(v_{00} + v_{11} - v_{10} - v_{01})$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = (v_{01} - v_{00}) + \alpha_1(v_{00} + v_{11} - v_{10} - v_{01})$$

Where are lines of constant value / critical points?

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = 0: \qquad \alpha_2 = \frac{v_{00} - v_{10}}{v_{00} + v_{11} - v_{10} - v_{01}}$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = 0: \qquad \alpha_1 = \frac{v_{00} - v_{01}}{v_{00} + v_{11} - v_{10} - v_{01}}$$





Compute gradient (critical points are where gradient is zero vector):

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = (v_{10} - v_{00}) + \alpha_2(v_{00} + v_{11} - v_{10} - v_{01})$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = (v_{01} - v_{00}) + \alpha_1(v_{00} + v_{11} - v_{10} - v_{01})$$

Where are lines of constant value / critical points?

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = 0: \qquad \alpha_2 = \frac{v_{00} - v_{10}}{v_{00} + v_{11} - v_{10} - v_{01}}$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = 0: \qquad \alpha_1 = \frac{v_{00} - v_{01}}{v_{00} + v_{11} - v_{10} - v_{01}}$$

critical point (saddle point)



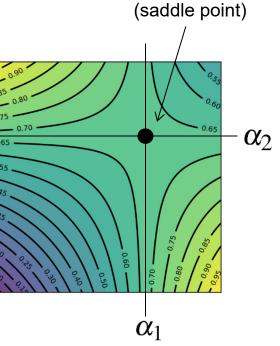
Compute gradient (critical points are where gradient is zero vector):

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = (v_{10} - v_{00}) + \alpha_2(v_{00} + v_{11} - v_{10} - v_{01})$$
$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = (v_{01} - v_{00}) + \alpha_1(v_{00} + v_{11} - v_{10} - v_{01})$$

Where are lines of constant value / critical points?

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_1} = 0: \qquad \alpha_2 = \frac{v_{00} - v_{10}}{v_{00} + v_{11} - v_{10} - v_{01}}$$

$$\frac{\partial f(\alpha_1, \alpha_2)}{\partial \alpha_2} = 0: \qquad \alpha_1 = \frac{v_{00} - v_{01}}{v_{00} + v_{11} - v_{10} - v_{01}}$$



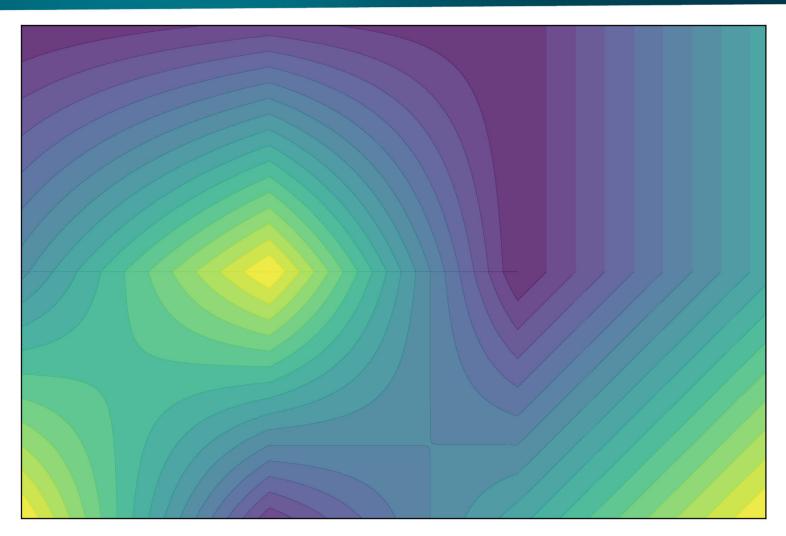
critical point



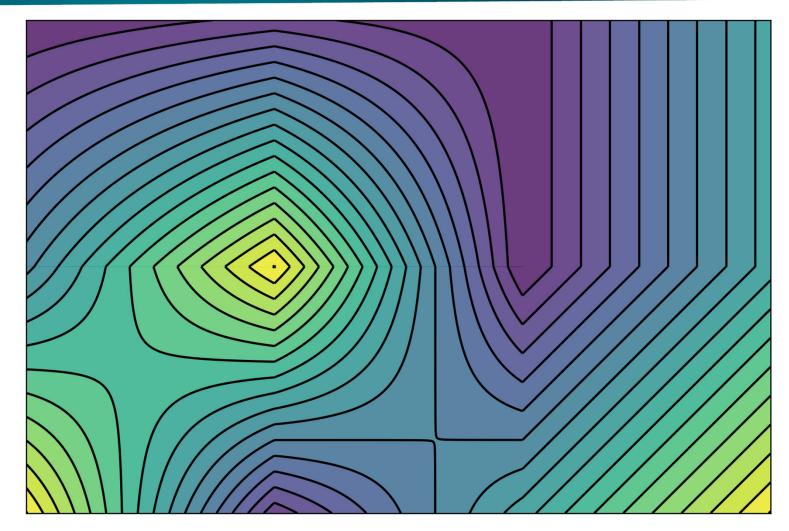
More examples

Piecewise Bi-Linear (Example: 3x2 Cells)

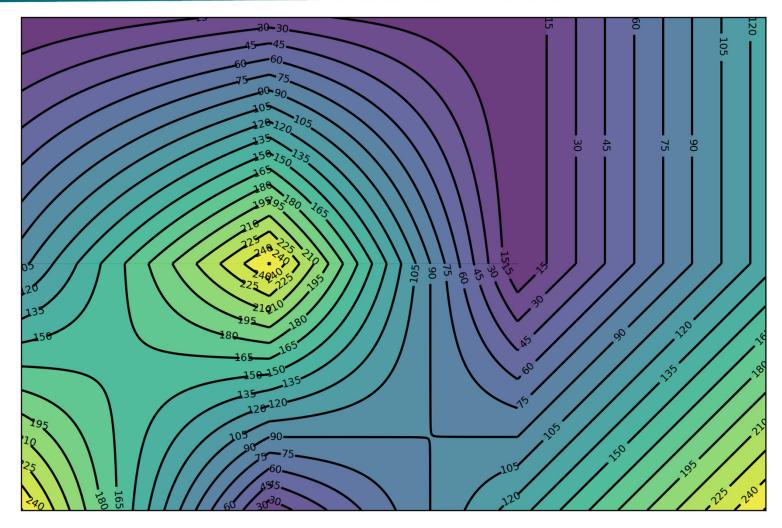




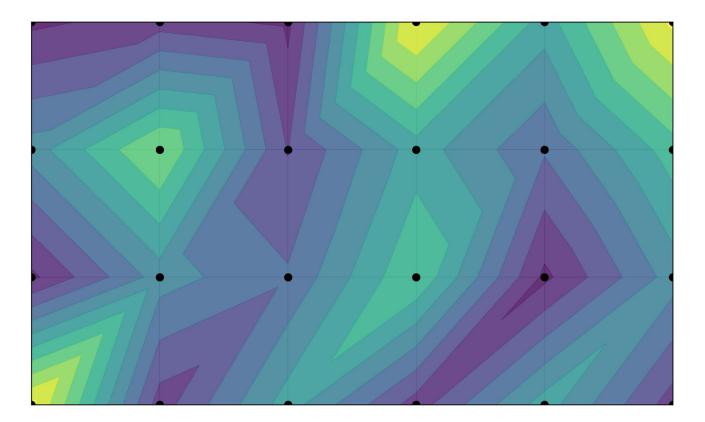
Piecewise Bi-Linear (Example: 3x2 Cells)



Piecewise Bi-Linear (Example: 3x2 Cells)

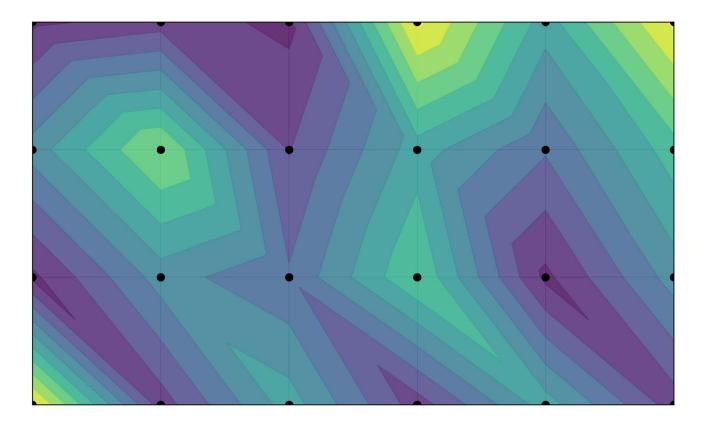






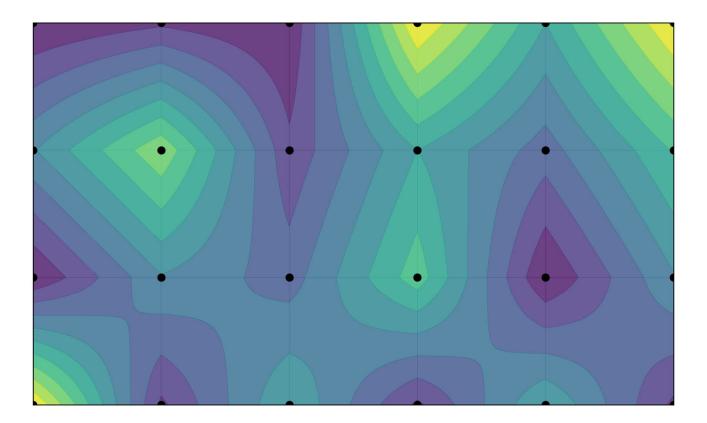
linear (diagonal 1)





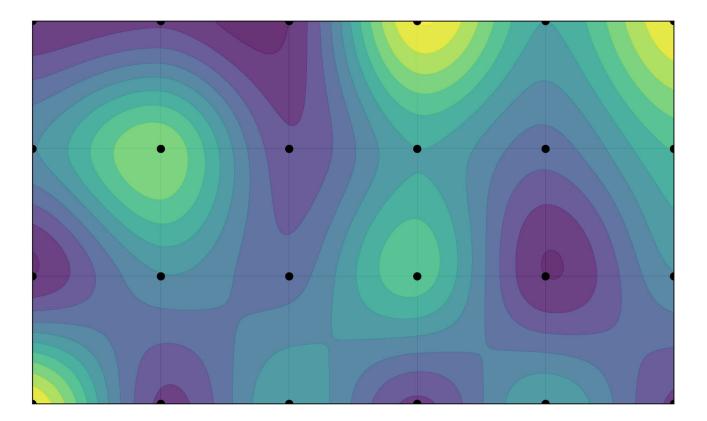
linear (diagonal 2)





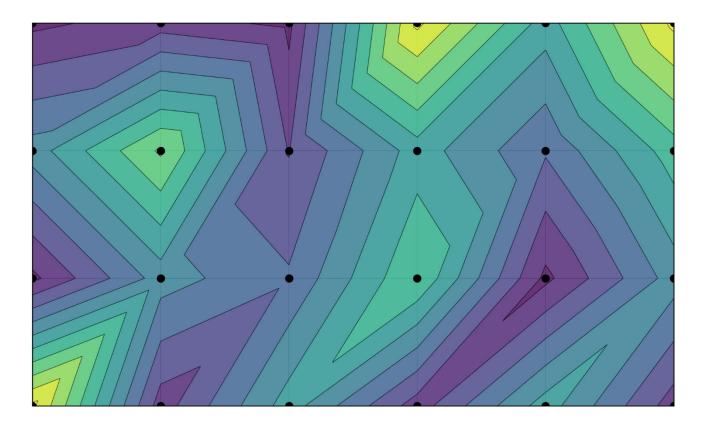
bi-linear (in 3D: tri-linear)





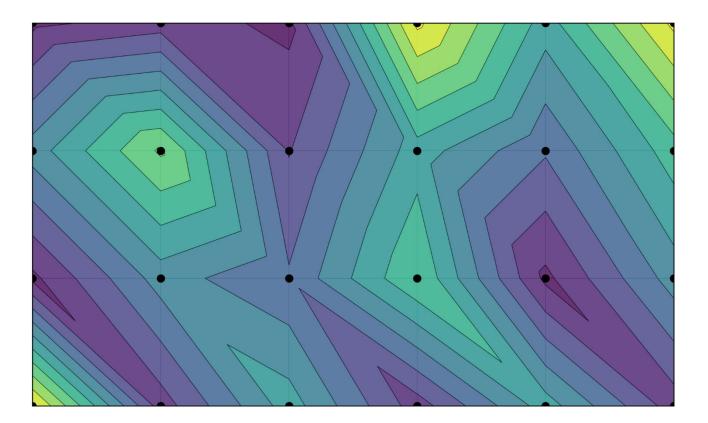
bi-cubic (in 3D: tri-cubic)





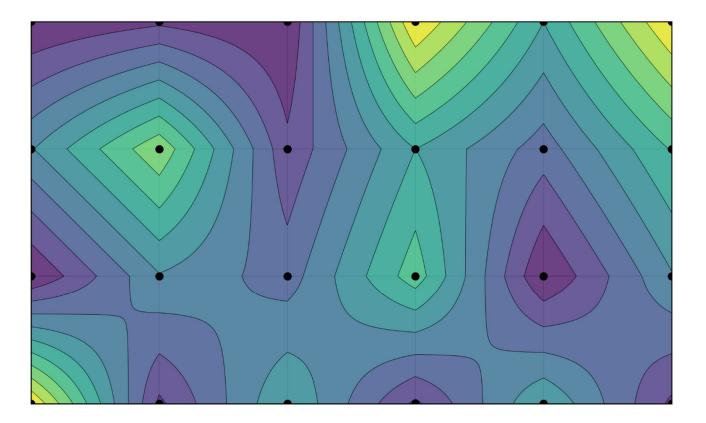
linear (diagonal 1)





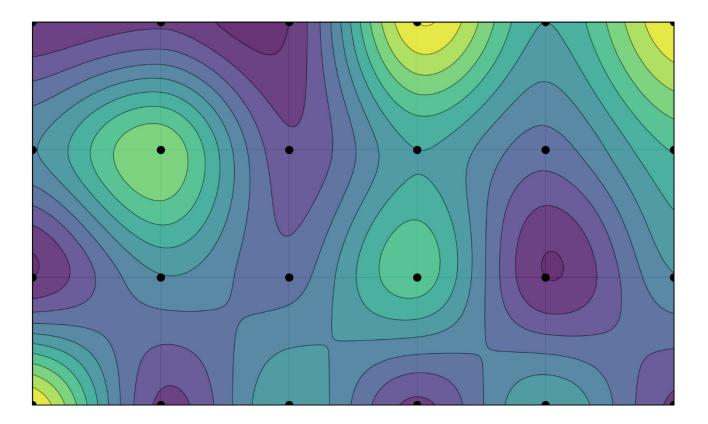
linear (diagonal 2)





bi-linear (in 3D: tri-linear)





bi-cubic (in 3D: tri-cubic)

Thank you.

Thanks for material

- Helwig Hauser
- Eduard Gröller
- Daniel Weiskopf
- Torsten Möller
- Ronny Peikert
- Philipp Muigg
- Christof Rezk-Salama