

CS 247 – Scientific Visualization

Lecture 6: Scalar Fields, Pt. 2 [preview]

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Reading Assignment #3 (until Feb 14)



Read (required):

- Data Visualization book, finish Chapter 3 (read starting with 3.6)
- Data Visualization book, Chapter 5 until 5.3 (inclusive)

Scalar Fields

Contours



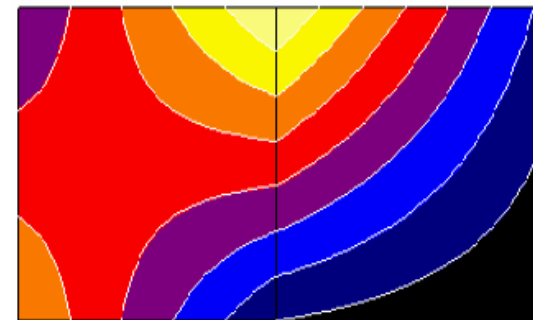
Set of points where the scalar field $f(x)$ has a given value c :

$$S(c) := f^{-1}(c) \quad S(c) := \{x \in \mathbb{R}^n : f(x) = c\}$$

Common contouring algorithms

- 2D: marching squares, marching triangles
- 3D: marching cubes, marching tetrahedra

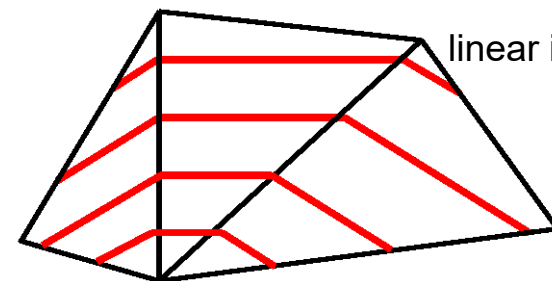
bilinear interpolation



Implicit methods

- Point-on-contour test
- Isosurface ray-casting

linear interpolation



Contours



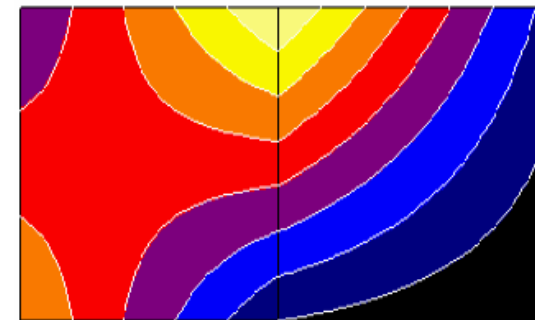
Set of points where the scalar field $f(x)$ has a given value c :

$$S(c) := f^{-1}(c) \quad S(c) := \{x \in \mathbb{R}^2 : f(x) = c\}$$

Common contouring algorithms

- 2D: marching squares, marching triangles
- 3D: marching cubes, marching tetrahedra

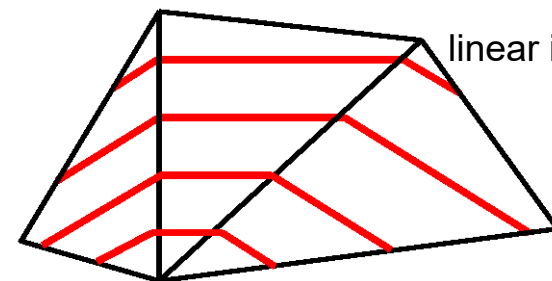
bilinear interpolation



Implicit methods

- Point-on-contour test
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linear interpolation



Contours



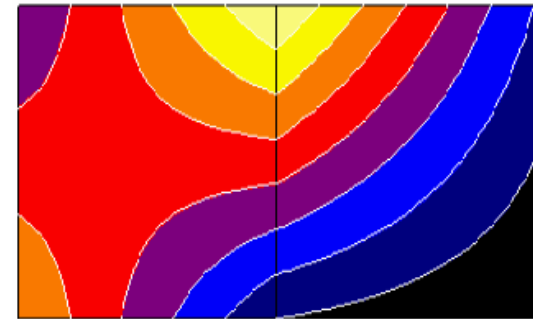
Set of points where the scalar field $f(x)$ has a given value c :

$$S(c) := f^{-1}(c) \quad S(c) := \{x \in \mathbb{R}^3 : f(x) = c\}$$

Common contouring algorithms

- 2D: marching squares, marching triangles
- 3D: marching cubes, marching tetrahedra

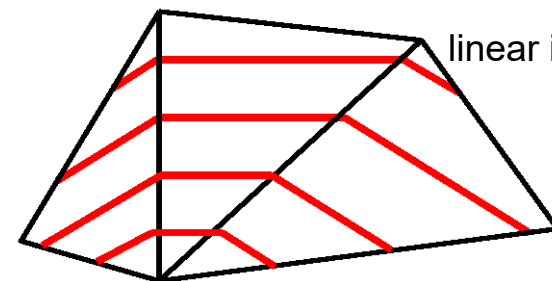
bilinear interpolation



Implicit methods

- Point-on-contour test
- Isosurface ray-casting

linear interpolation



What are contours?

Set of points where the scalar field s has a given value c :

$$S(c) := \{x \in \mathbb{R}^n : f(x) = c\}$$

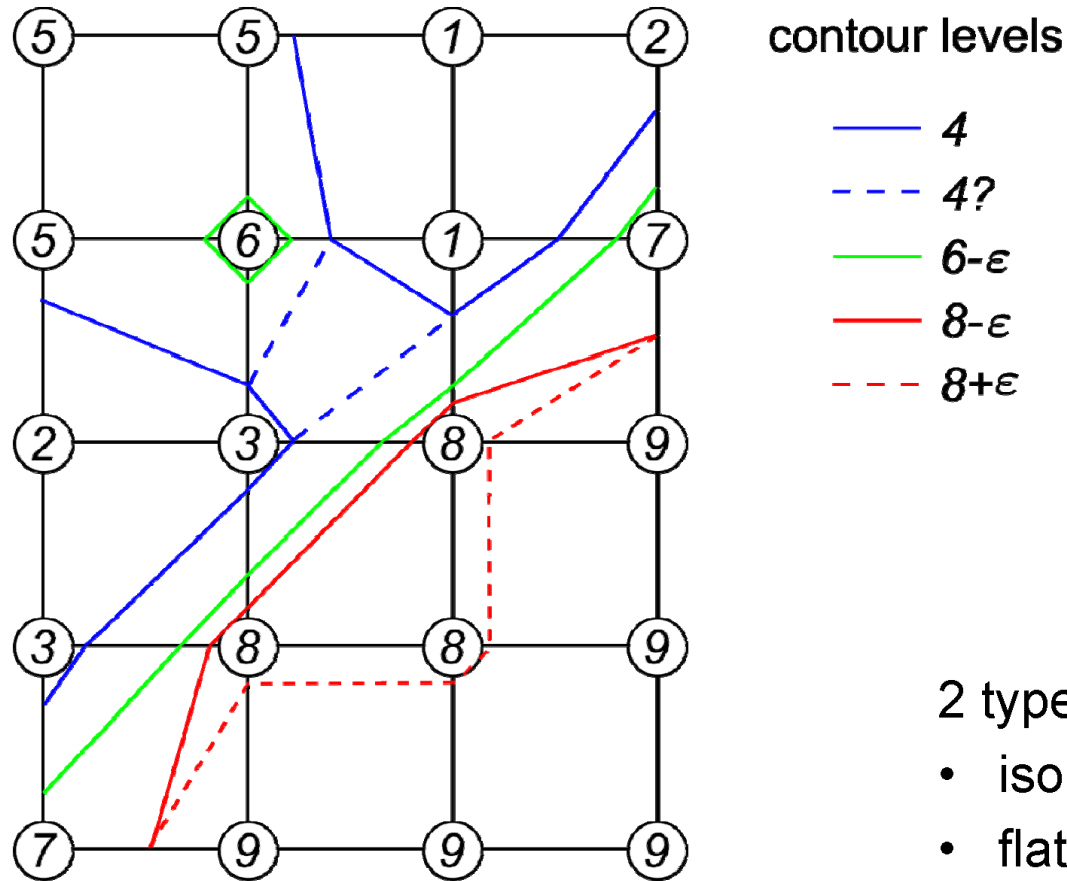
Examples in 2D:

- height contours on maps
- isobars on weather maps

Contouring algorithm:

- find intersection with grid edges
- connect points in each cell

Example



Contours in a quadrangle cell

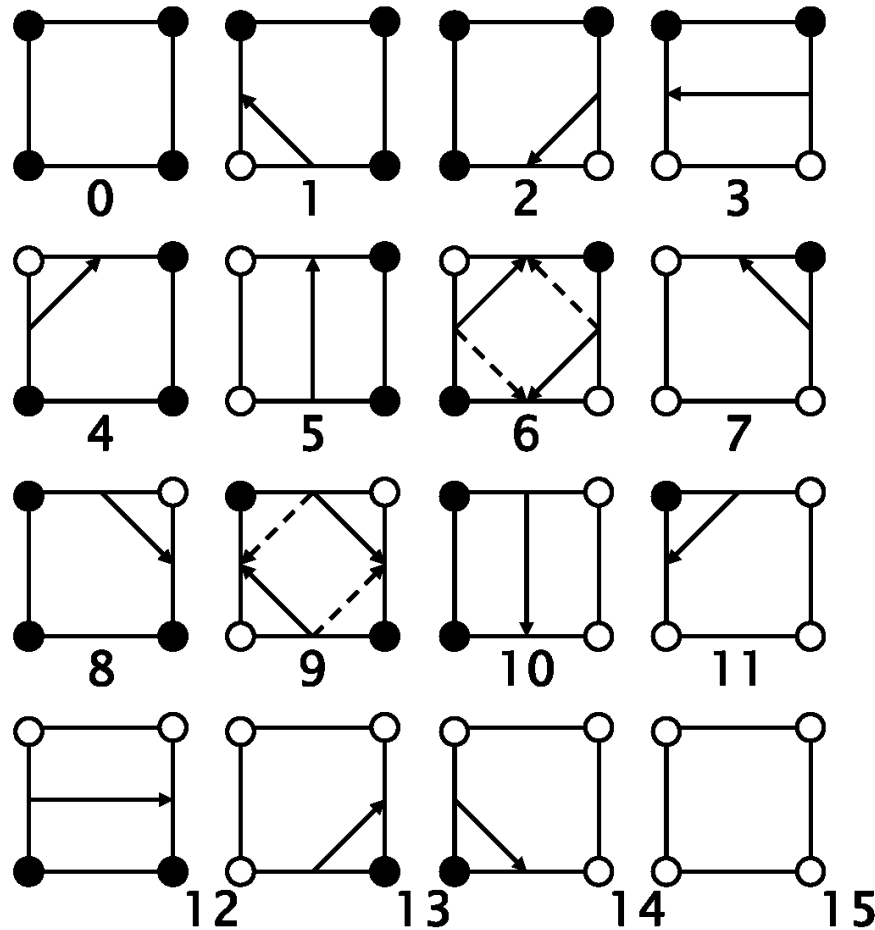
Basic contouring algorithms:

- **cell-by-cell** algorithms: simple structure, but generate disconnected segments, require post-processing
- **contour propagation** methods: more complicated, but generate connected contours

"Marching squares" algorithm (systematic cell-by-cell):

- process nodes in ccw order, denoted here as x_0, x_1, x_2, x_3
- compute at each node \mathbf{x}_i the reduced field $\tilde{f}(x_i) = f(x_i) - (c - \varepsilon)$ (which is forced to be nonzero)
- take its sign as the i^{th} bit of a 4-bit integer
- use this as an index for lookup table containing the connectivity information:

Contours in a quadrangle cell



- $\tilde{f}(x_i) < 0$
- $\tilde{f}(x_i) > 0$

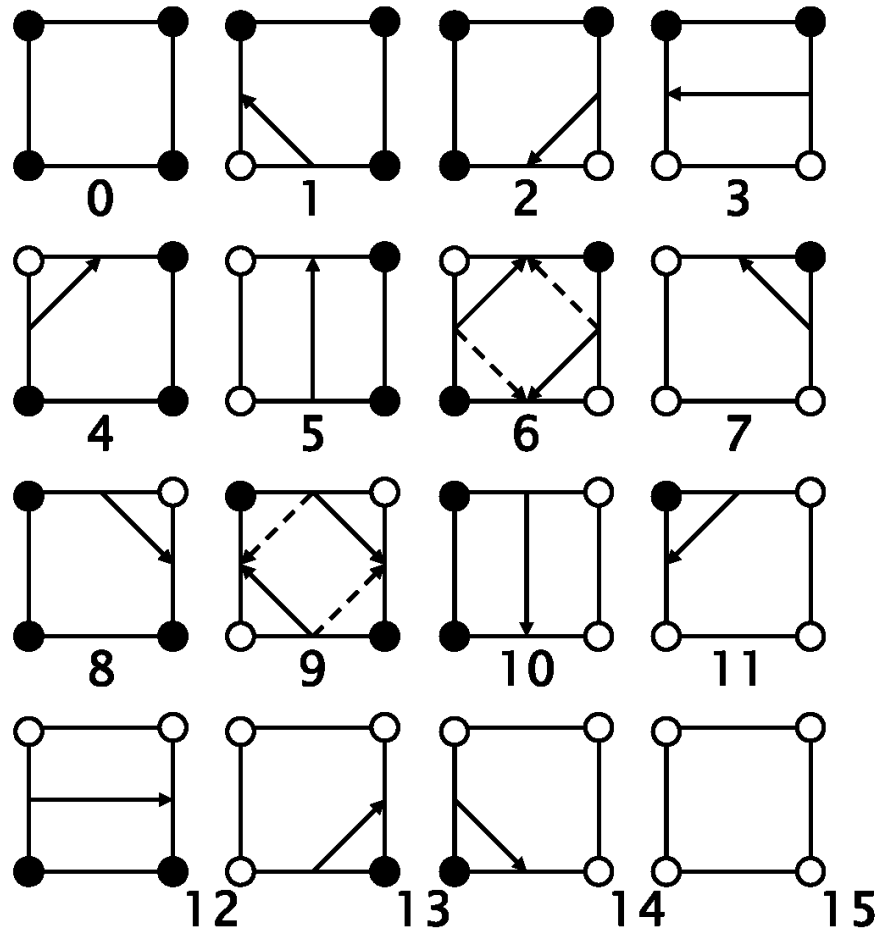
Alternating signs exist
in cases 6 and 9.

Choose the solid or
dashed line?

Both are possible for
topological
consistency.

This allows to have a
fixed table of 16
cases.

Contours in a quadrangle cell



● $f(x_i) < c$

○ $f(x_i) \geq c$

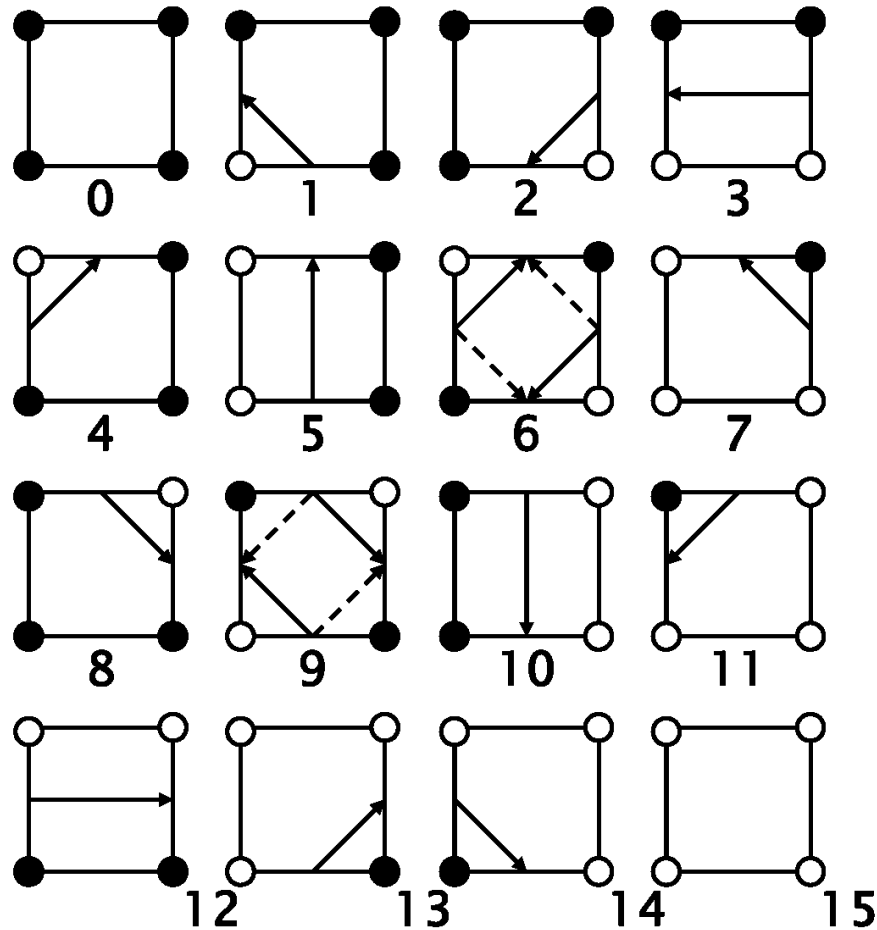
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Choose the solid or
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Both are possible for
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Contours in a quadrangle cell



- $f(x_i) \leq c$
- $f(x_i) > c$

Alternating signs exist
in cases 6 and 9.

Choose the solid or
dashed line?

Both are possible for
topological
consistency.

This allows to have a
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cases.

Orientability (1-manifold embedded in 2D)

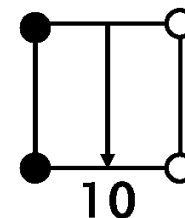
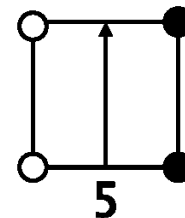


Orientability of 1-manifold:

Possible to assign consistent left/right orientation

Iso-contours

- Consistent side for scalar values...
 - greater than iso-value (e.g, *left* side)
 - less than iso-value (e.g., *right* side)
- Use consistent ordering of vertices (e.g., larger vertex index is “tip” of arrow; if (0,1) points “up”, “left” is left, ...)



not orientable



Möbius strip
(only one side!)

$$\bullet \tilde{f}(x_i) < 0$$

$$\circ \tilde{f}(x_i) > 0$$

Orientability (2-manifold embedded in 3D)



Orientability of 2-manifold:

Possible to assign consistent normal vector orientation

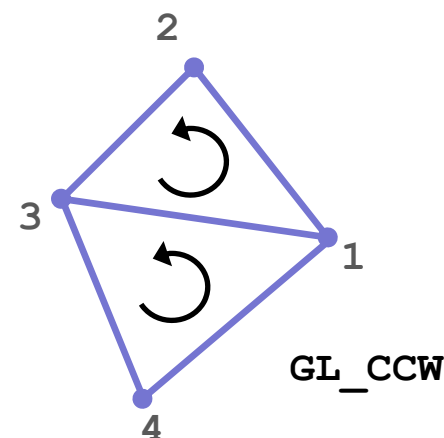
Triangle meshes

- Edges
 - Consistent ordering of vertices: CCW (counter-clockwise) or CW (clockwise) (e.g., (3,1,2) on one side of edge, (1,3,4) on the other side)
- Triangles
 - Consistent front side vs. back side
 - Normal vector; or ordering of vertices (CCW/CW)
 - See also: “right-hand rule”

not orientable



Möbius strip
(only one side!)



Topological consistency

To avoid degeneracies, use **symbolic perturbations**:

If level c is found as a node value, set the level to $c - \varepsilon$ where ε is a symbolic infinitesimal.

Then:

- contours intersect edges at some (possibly infinitesimal) distance from end points
- flat regions can be visualized by pair of contours at $c - \varepsilon$ and $c + \varepsilon$
- contours are **topologically consistent**, meaning:

Contours are **closed, orientable, nonintersecting lines**.

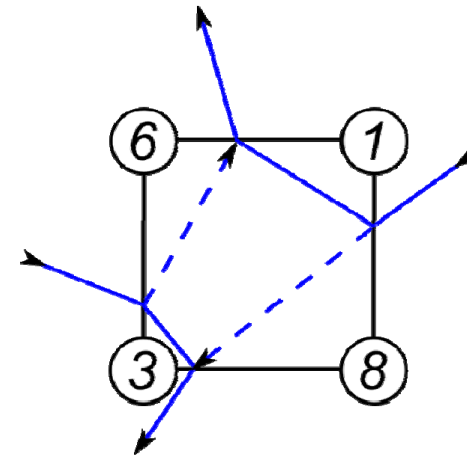
(except where the
boundary is hit)

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 $f(x)$.

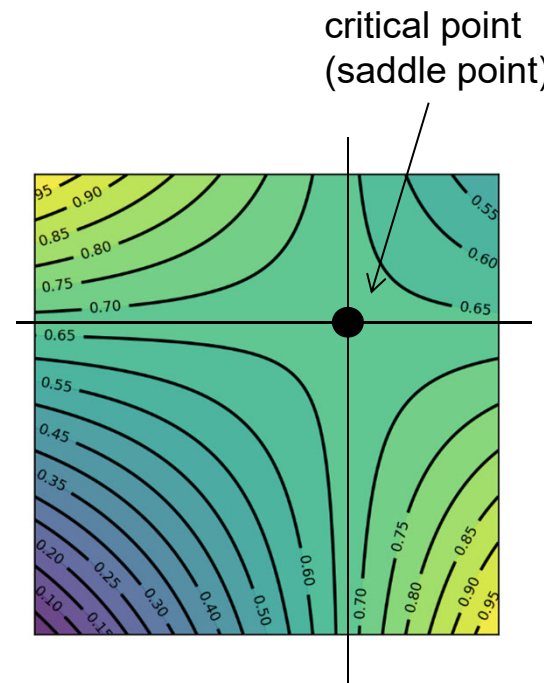
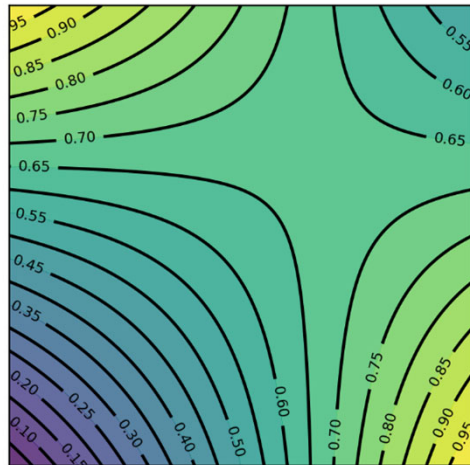
But: different interpolation schemes are possible.

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

Bi-Linear Interpolation: Critical Points



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



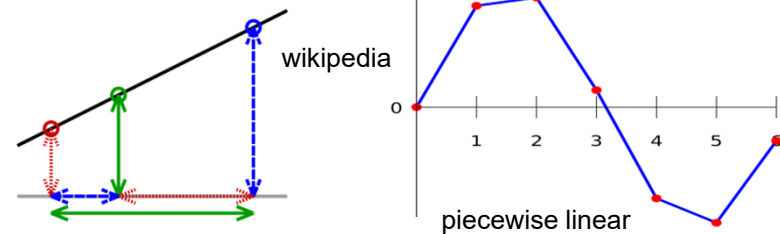
“Asymptotic decider”: resolve ambiguous configurations (6 and 9) by comparing specific iso-value with critical value (scalar value at critical point)

Linear Interpolation / Convex Combinations



Linear interpolation in 1D:

$$f(\alpha) = (1 - \alpha)v_1 + \alpha v_2$$



Line embedded in 2D (linear interpolation of vertex coordinates/attributes):

$$f(\alpha_1, \alpha_2) = \alpha_1 v_1 + \alpha_2 v_2$$

$$\alpha_1 + \alpha_2 = 1$$

$$f(\alpha) = v_1 + \alpha(v_2 - v_1)$$

$$\alpha = \alpha_2$$

Line segment: $\alpha_1, \alpha_2 \geq 0$ (\rightarrow convex combination)

Compare to line parameterization
with parameter t :

$$v(t) = v_1 + t(v_2 - v_1)$$

Linear Interpolation / Convex Combinations



Linear combination (n -dim. space):

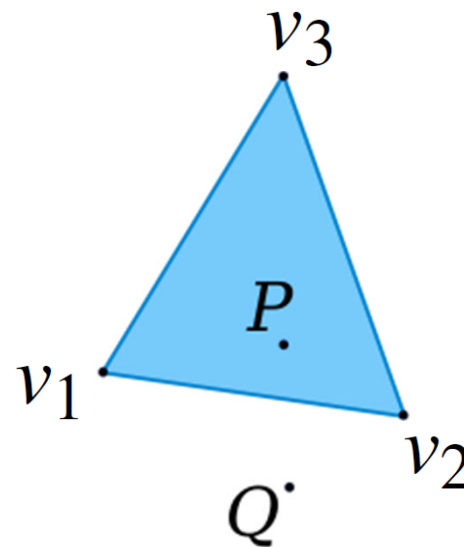
$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = \sum_{i=1}^n \alpha_i v_i$$

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

$$\alpha_1 + \alpha_2 + \dots + \alpha_n = \sum_{i=1}^n \alpha_i = 1$$

Convex combination: $\alpha_i \geq 0$

(restrict to simplex in subspace)



Linear Interpolation / Convex Combinations



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

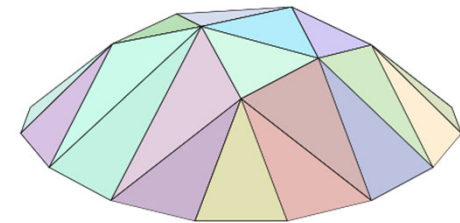
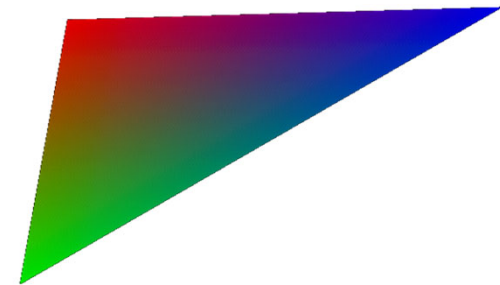
→ linear attribute interpolation in simplex

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = \sum_{i=1}^n \alpha_i v_i$$

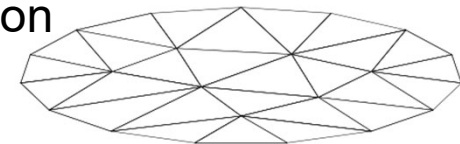
$$\alpha_1 + \alpha_2 + \dots + \alpha_n = \sum_{i=1}^n \alpha_i = 1$$

$$\alpha_i \geq 0$$

attribute interpolation



spatial position
interpolation



wikipedia

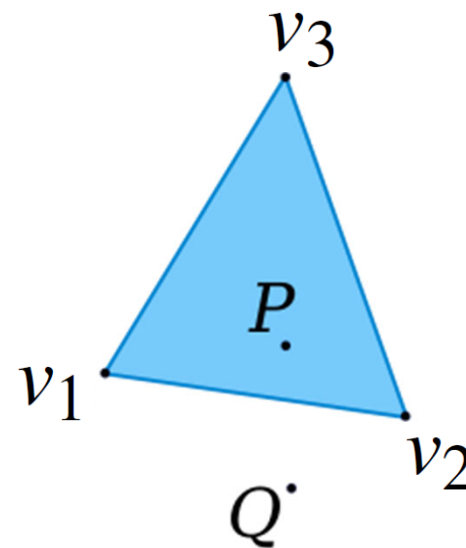
Linear Interpolation / Convex Combinations



$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = \sum_{i=1}^n \alpha_i v_i$$
$$\alpha_1 + \alpha_2 + \dots + \alpha_n = \sum_{i=1}^n \alpha_i = 1$$

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

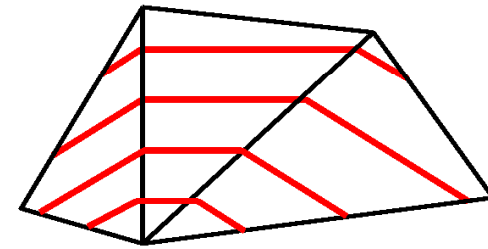
$$\begin{aligned}\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3 &= \\ \tilde{\alpha}_1 (v_2 - v_1) + \tilde{\alpha}_2 (v_3 - v_1) + v_1 & \\ \tilde{\alpha}_1 &= \alpha_2 \\ \tilde{\alpha}_2 &= \alpha_3\end{aligned}$$



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".

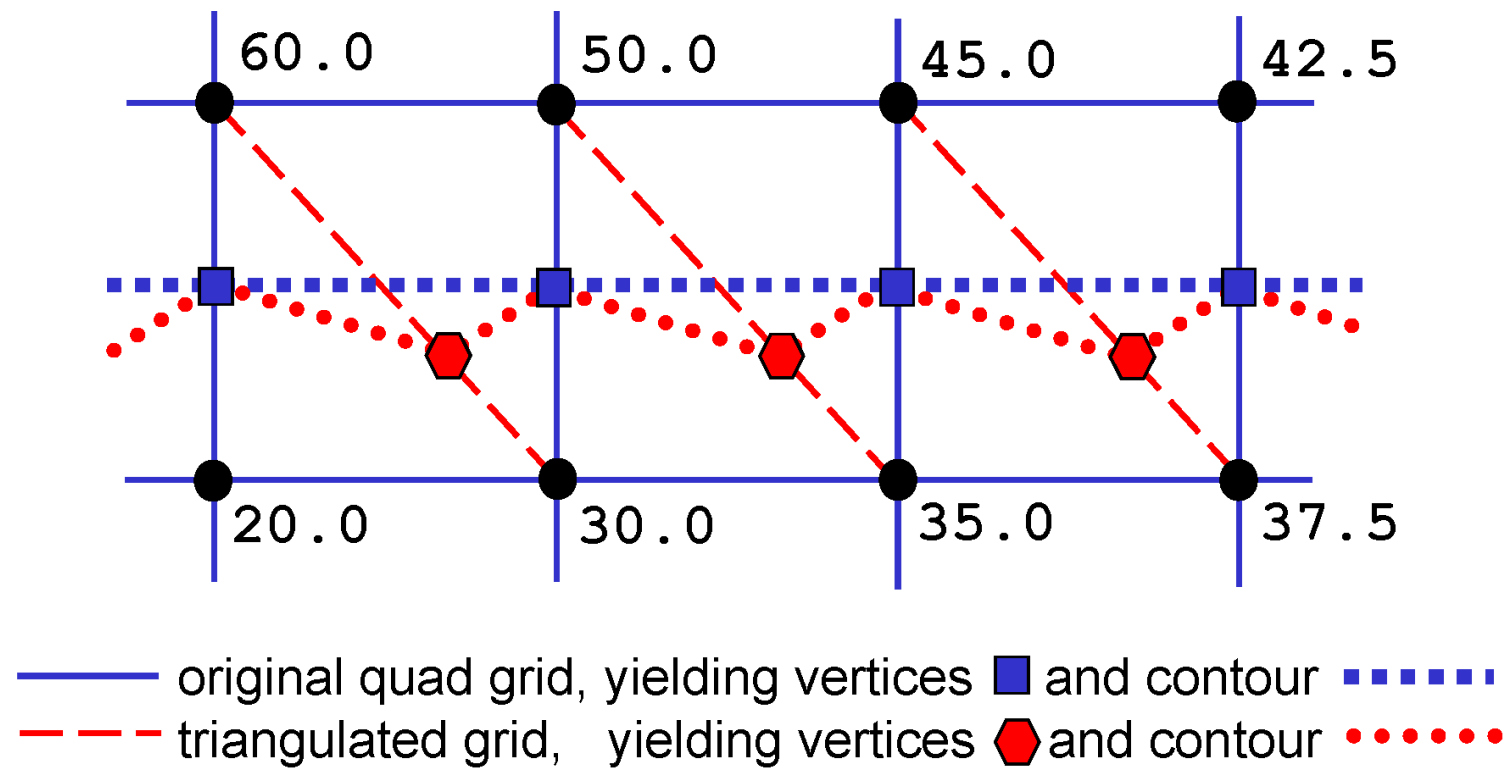


Question: Why not split quadrangles into two triangles (and hexahedra into five or six tetrahedra) and use marching triangles (tetrahedra)?

Answer: This can introduce periodic artifacts!

Contours in triangle/tetrahedral cells

Illustrative example: Find contour at level $c=40.0$!

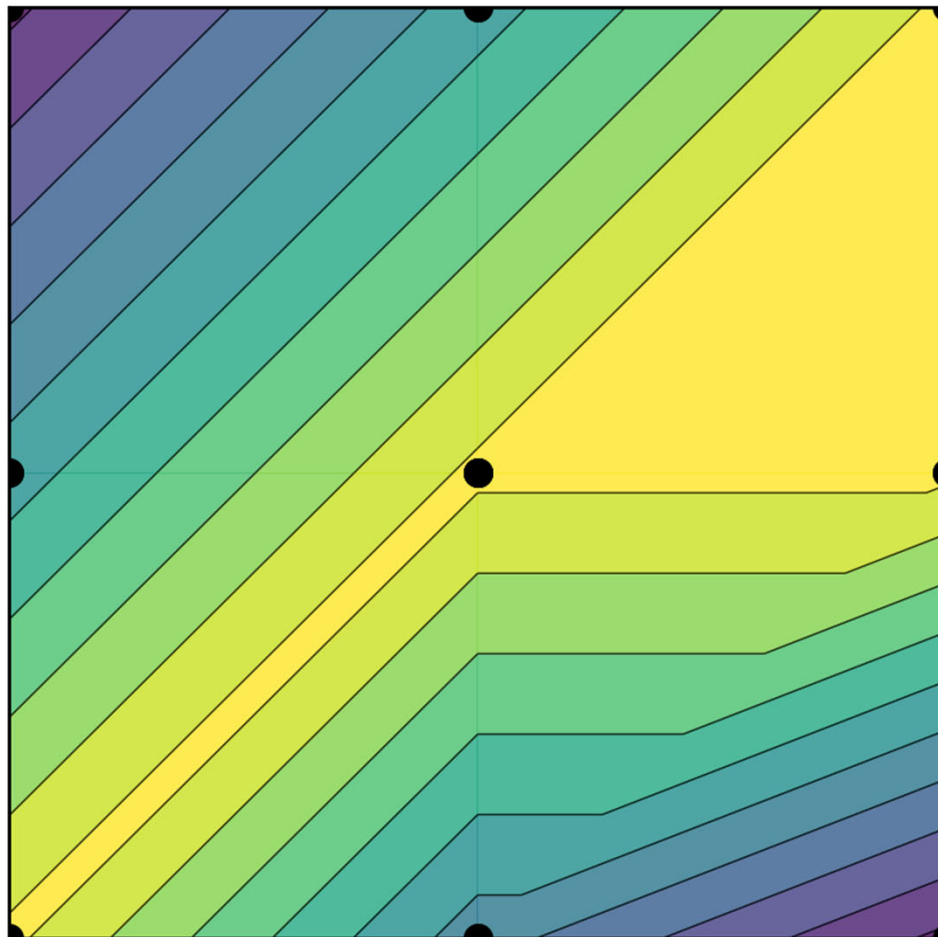


Bi-Linear Interpolation: Comparisons



linear

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

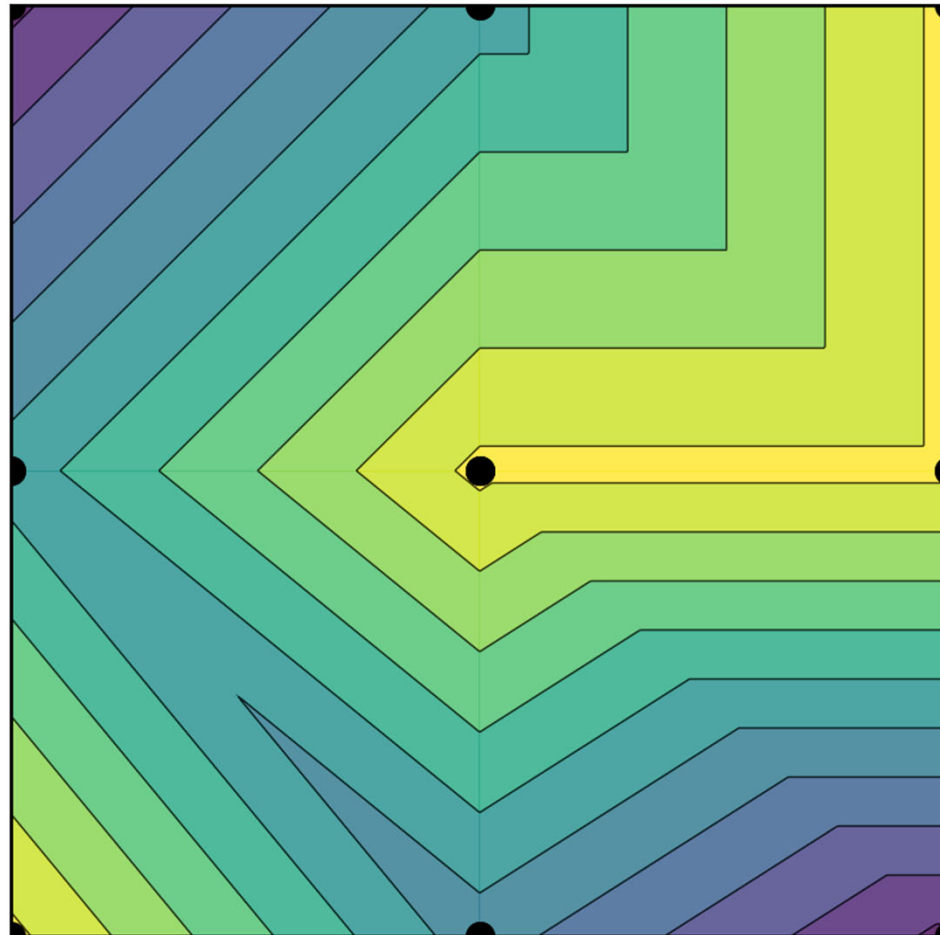


Bi-Linear Interpolation: Comparisons



linear

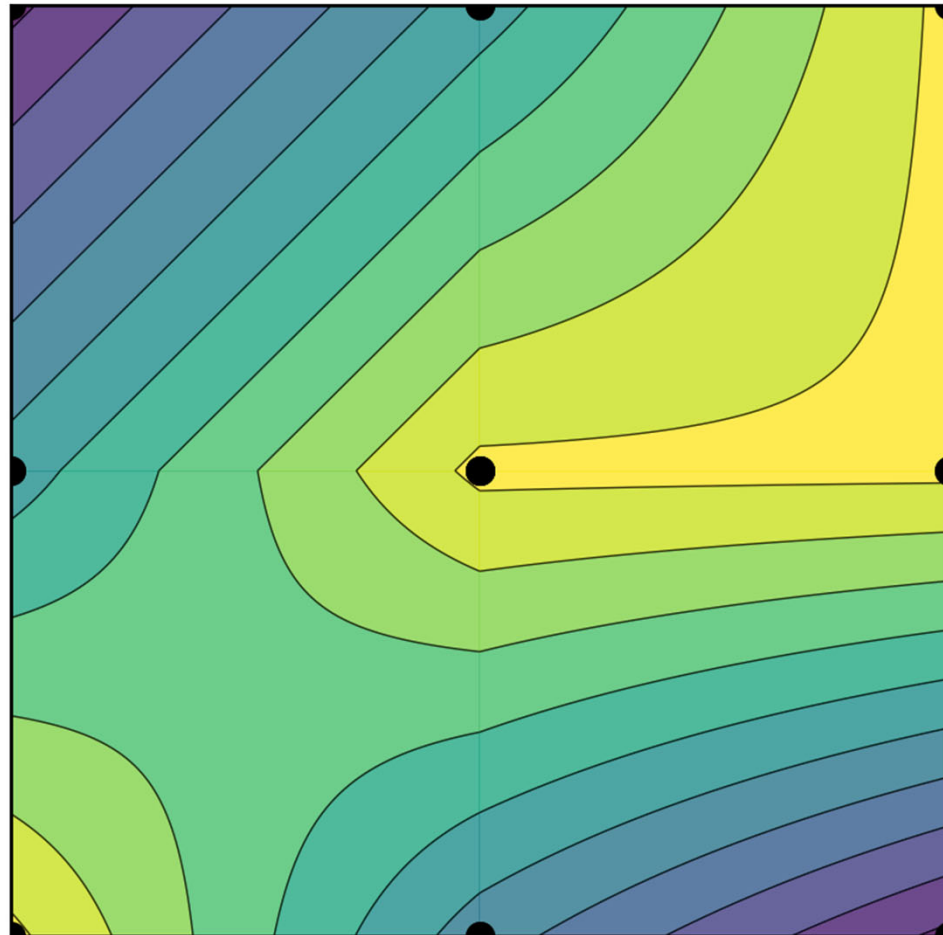
(2 triangles per quad;
diagonal:
top-left,
bottom-right)



Bi-Linear Interpolation: Comparisons



bi-linear



From 2D to 3D (Domain)



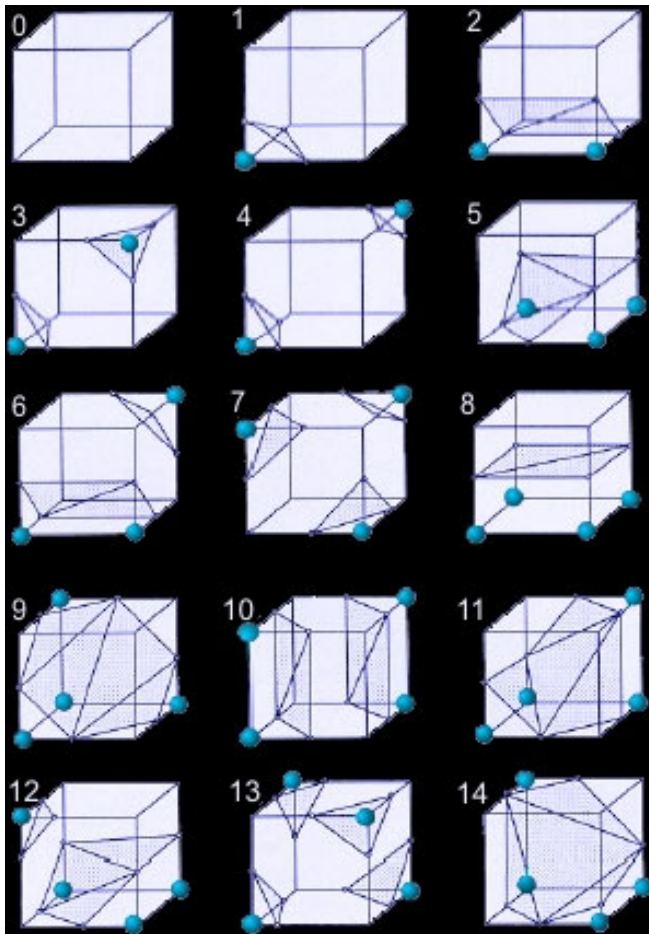
2D - Marching Squares Algorithm:

1. Locate the contour corresponding to a user-specified iso value
2. Create lines

3D - Marching Cubes Algorithm:

1. Locate the surface corresponding to a user-specified iso value
2. Create triangles
3. Calculate normals to the surface at each vertex
4. Draw shaded triangles

Marching Cubes



- For each cell, we have 8 vertices with 2 possible states each (inside or outside).
- This gives us 2^8 possible patterns = 256 cases.
- Enumerate cases to create a LUT
- Use symmetries to reduce problem from 256 to 15 cases.

Explanations

- Data Visualization book, 5.3.2
- Marching Cubes: A high resolution 3D surface construction algorithm, Lorensen & Cline, ACM SIGGRAPH 1987

The marching cubes algorithm

Contours of 3D scalar fields are known as **isosurfaces**.

Before 1987, isosurfaces were computed as

- contours on planar **slices**, followed by
- "contour stitching".

The **marching cubes** algorithm computes contours **directly in 3D**.

- Pieces of the isosurfaces are generated on a cell-by-cell basis.
- Similar to marching squares, a 8-bit number is computed from the 8 signs of $\tilde{f}(x_i)$ on the corners of a hexahedral cell.
- The isosurface piece is looked up in a table with 256 entries.

The marching cubes algorithm

How to build up the table of 256 cases?

Lorensen and Cline (1987) exploited 3 types of symmetries:

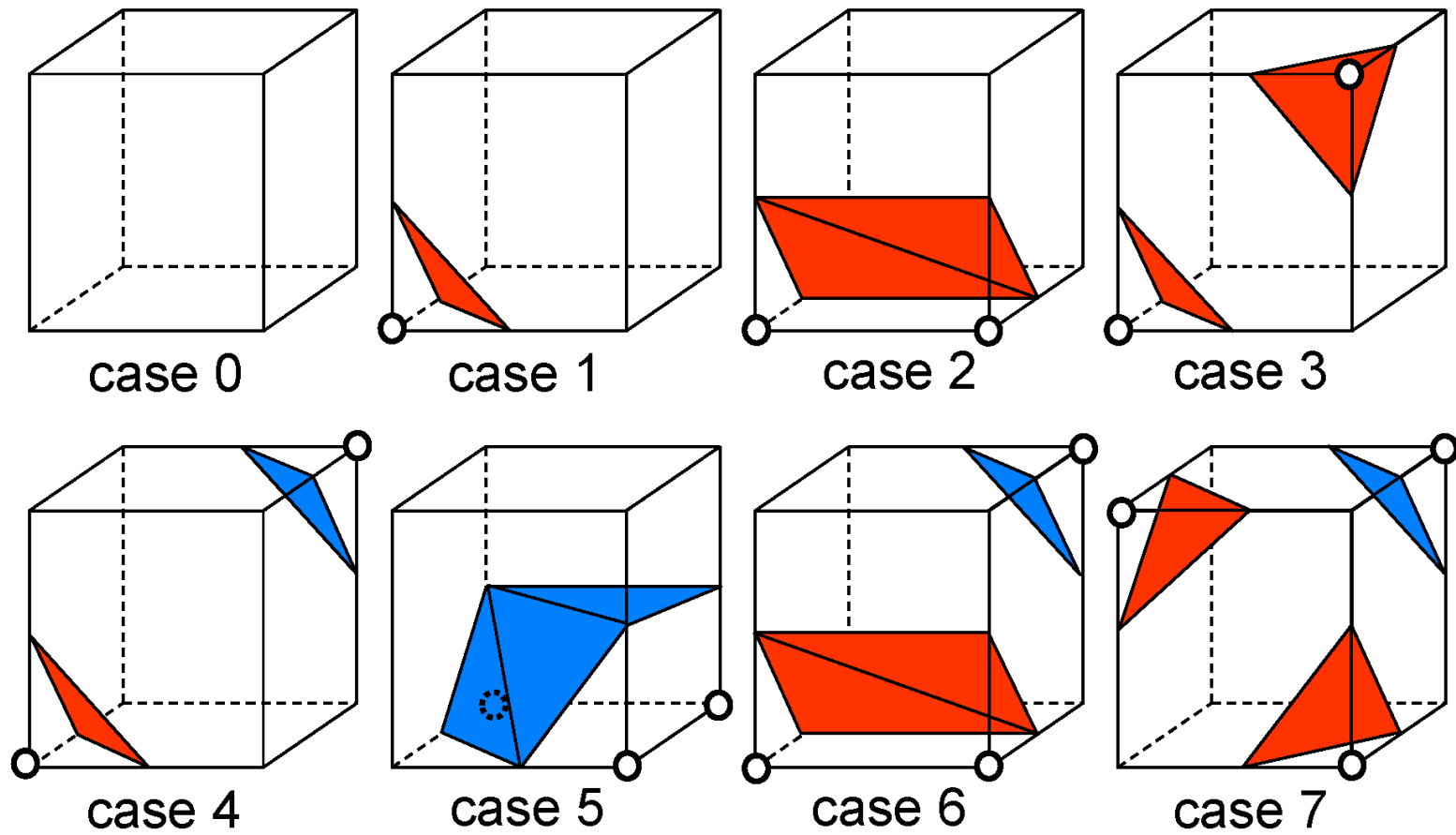
- rotational symmetries of the cube
- reflective symmetries of the cube
- sign changes of $\tilde{f}(x_i)$

They published a reduced set of 14^{*)} cases shown on the next slides where

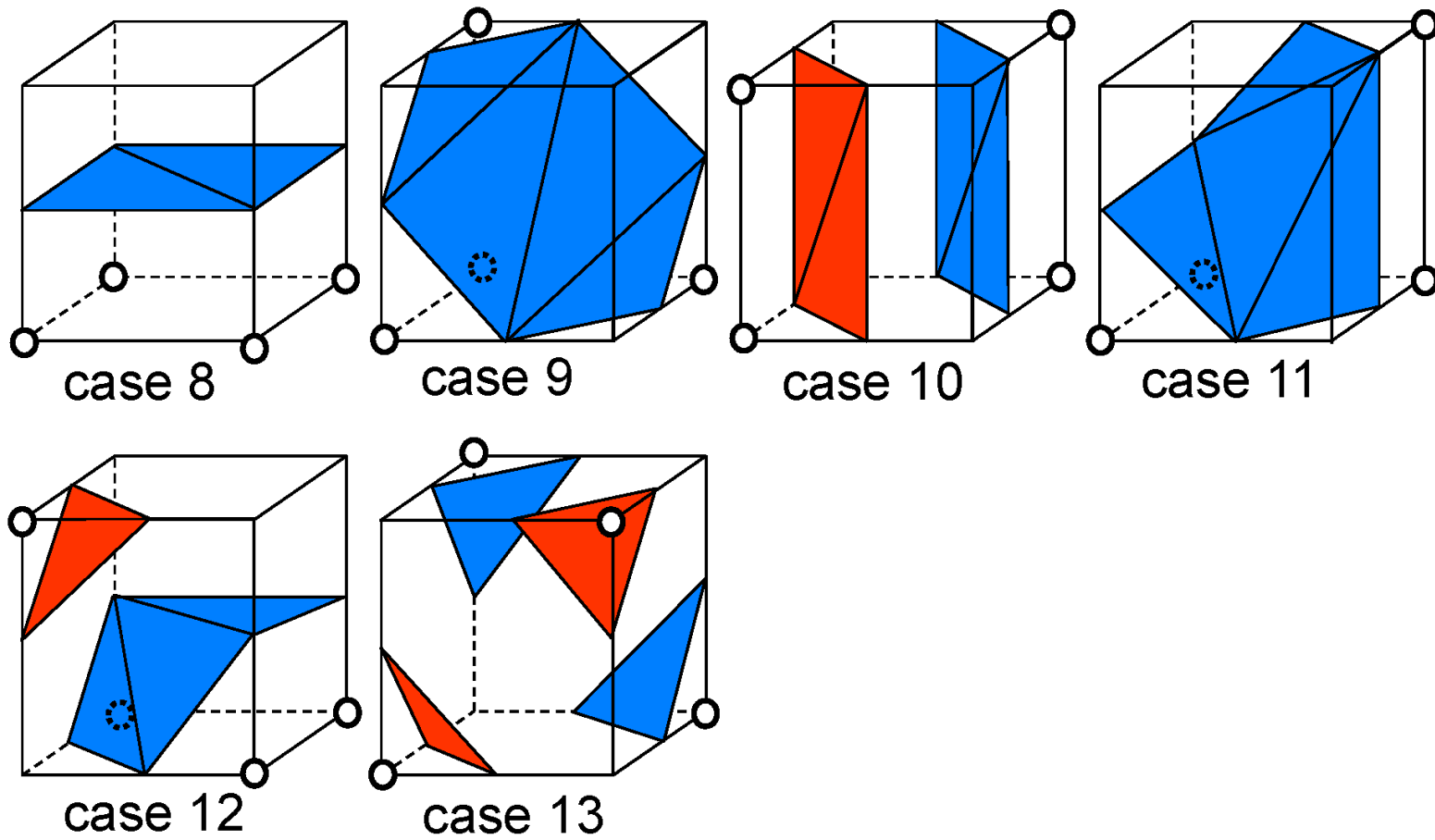
- white circles indicate positive signs of $\tilde{f}(x_i)$
- the positive side of the isosurface is drawn in red, the negative side in blue.

*) plus an unnecessary "case 14" which is a symmetric image of case 11.

The marching cubes algorithm



The marching cubes algorithm



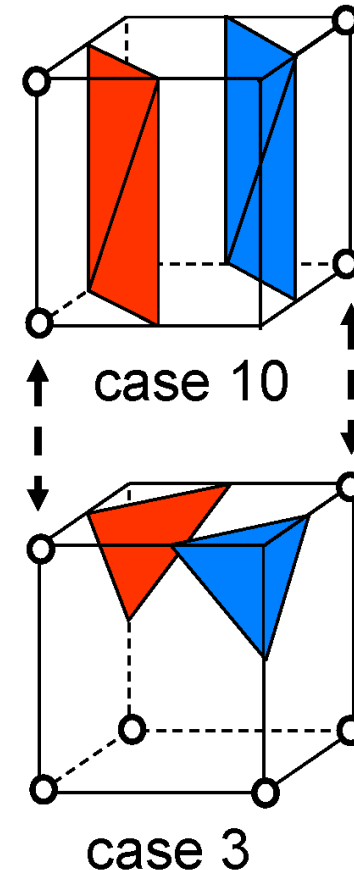
The marching cubes algorithm

Do the pieces fit together?

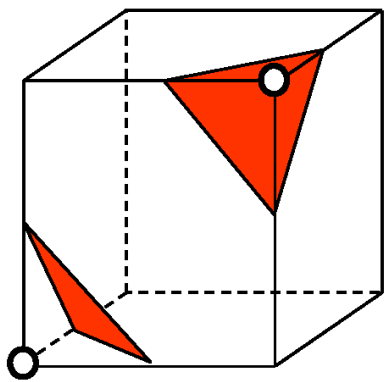
- The correct isosurfaces of the **trilinear interpolant** would fit (trilinear reduces to bilinear on the cell interfaces)
- but the marching cubes polygons don't necessarily fit.

Example

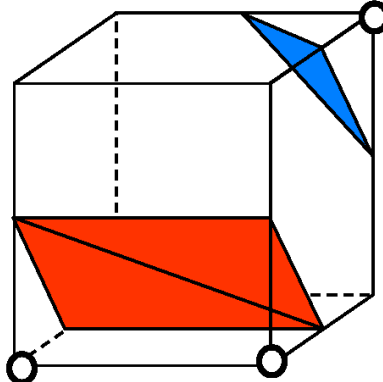
- case 10, on top of
 - case 3 (rotated, signs changed)
- have matching signs at nodes but polygons don't fit.



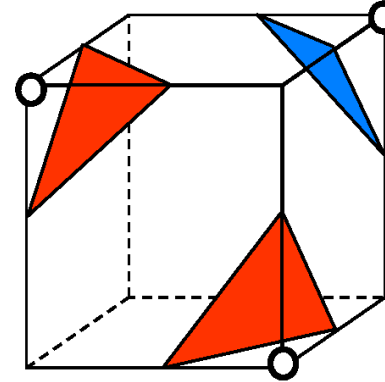
The marching cubes algorithm



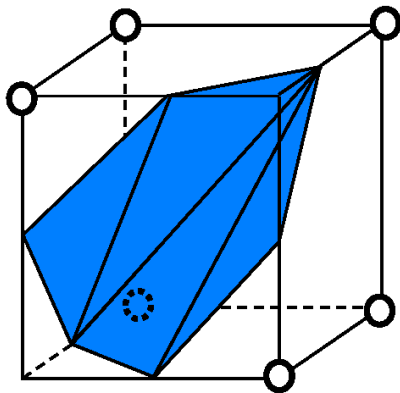
case 3



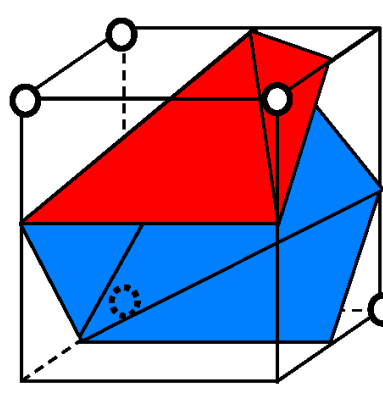
case 6



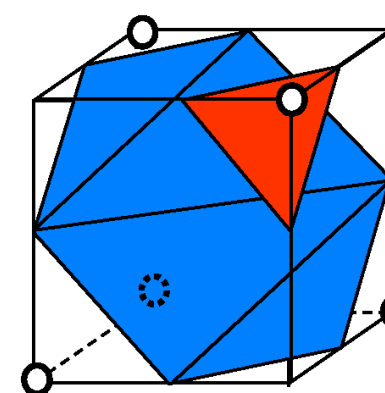
case 7



case 3c



case 6c



case 7c

The marching cubes algorithm

Summary of marching cubes algorithm:

Pre-processing steps:

- build a table of the 28 cases
- derive a table of the 256 cases, containing info on
 - intersected cell edges, e.g. for case 3/256 (see case 2/28):
 $(0,2), (0,4), (1,3), (1,5)$
 - triangles based on these points, e.g. for case 3/256:
 $(0,2,1), (1,3,2)$.

The marching cubes algorithm

Loop over cells:

- find sign of $\tilde{f}(x_i)$ for the 8 corner nodes, giving 8-bit integer
- use as index into (256 case) table
- find intersection points on edges listed in table, using linear interpolation
- generate triangles according to table

Post-processing steps:

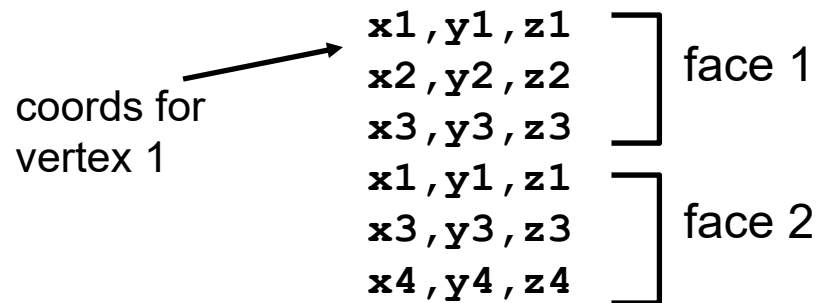
- connect triangles (share vertices)
- compute normal vectors
 - by averaging triangle normals (problem: thin triangles!)
 - by estimating the gradient of the field $f(x_i)$ (better)

Triangle Mesh Data Structure (1)

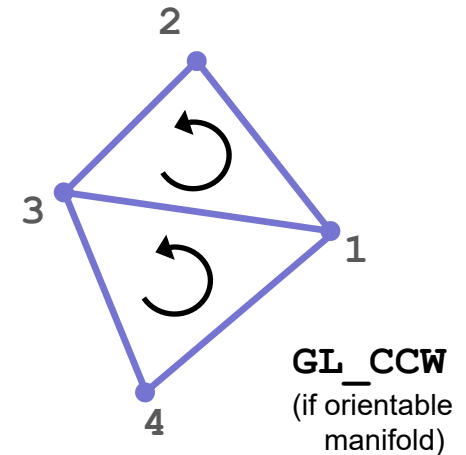


Store list of vertices; vertices shared by triangles are replicated

Render, e.g., with OpenGL immediate mode, ...



```
struct face
  float verts[3][3]
  DataType val;
```



...

Redundant, large storage size, cannot modify shared vertices easily

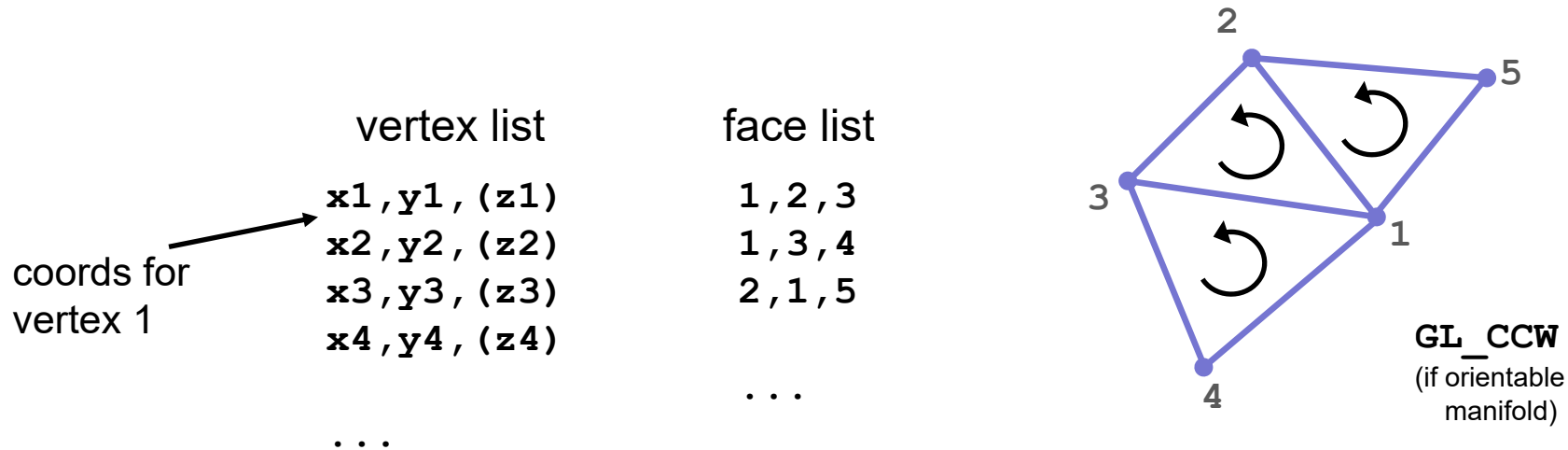
Store data values per face, or separately

Triangle Mesh Data Structure (2)



Indexed face set: store list of vertices; store triangles as indexes

Render using separate vertex and index arrays / buffers



Less redundancy, more efficient in terms of memory

Easy to change vertex positions; still have to do (global) search for shared edges (local information)

Orientability (2-manifold embedded in 3D)



Orientability of 2-manifold:

Possible to assign consistent normal vector orientation

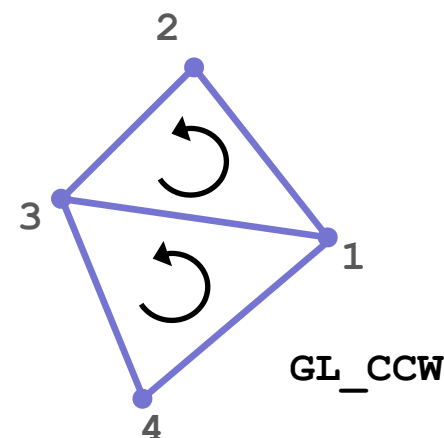
Triangle meshes

- Edges
 - Consistent ordering of vertices: CCW (counter-clockwise) or CW (clockwise) (e.g., (3,1,2) on one side of edge, (1,3,4) on the other side)
- Triangles
 - Consistent front side vs. back side
 - Normal vector; or ordering of vertices (CCW/CW)
 - See also: “right-hand rule”

not orientable



Möbius strip
(only one side!)



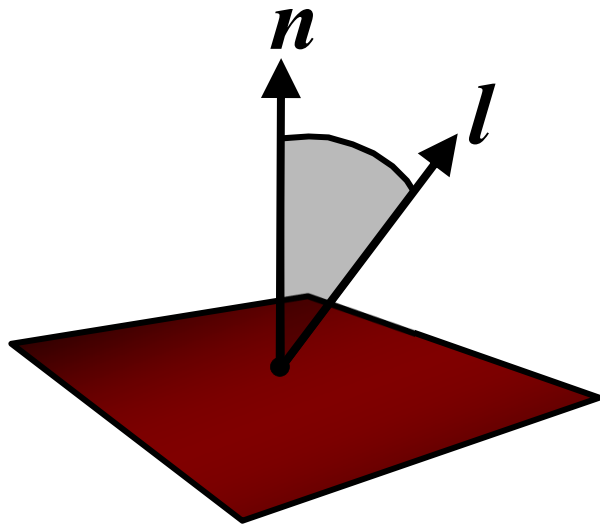
Local Shading Equations



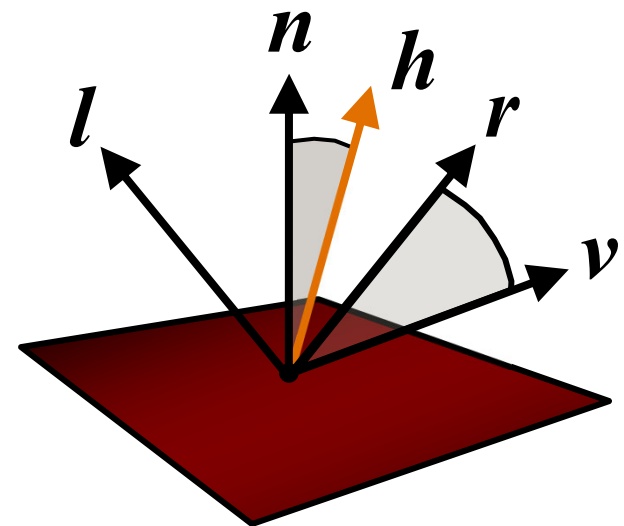
Standard volume shading adapts surface shading

Most commonly Blinn/Phong model

But what about the "surface" normal vector?



diffuse reflection



specular reflection

The Dot Product (Scalar / Inner Product)



Cosine of angle between two vectors times their lengths

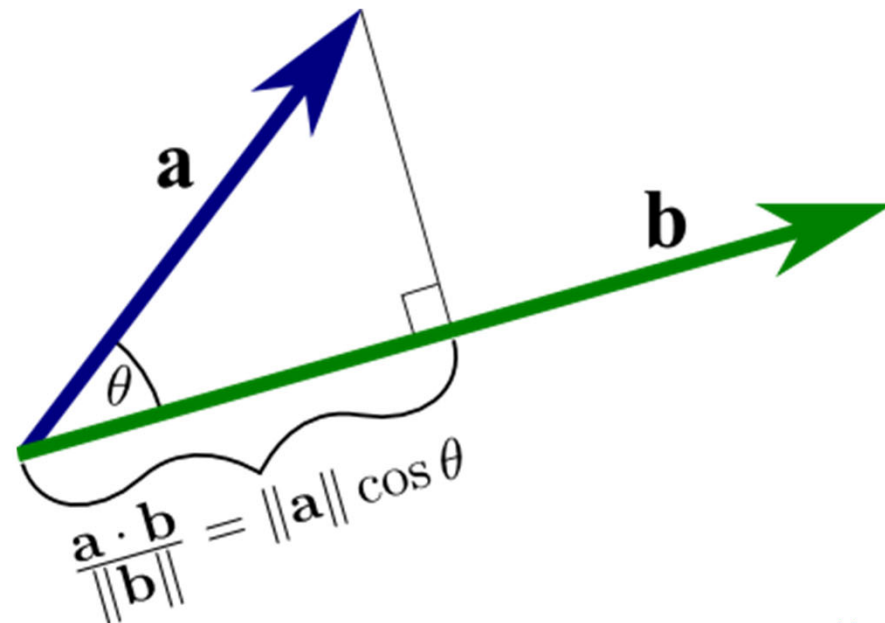
$$\mathbf{a} \cdot \mathbf{b} = \sum_{i=1}^n a_i b_i$$

$$\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos \theta$$

(standard inner product in Cartesian coordinates)

Many uses:

- Project vector onto another vector,
project into basis,
project into tangent plane,
...



Thank you.

Thanks for material

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