CS 247 – Scientific Visualization
Lecture 12: Scalar Field Visualization, Pt. 5

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Reading Assignment #7 (until Mar 16)

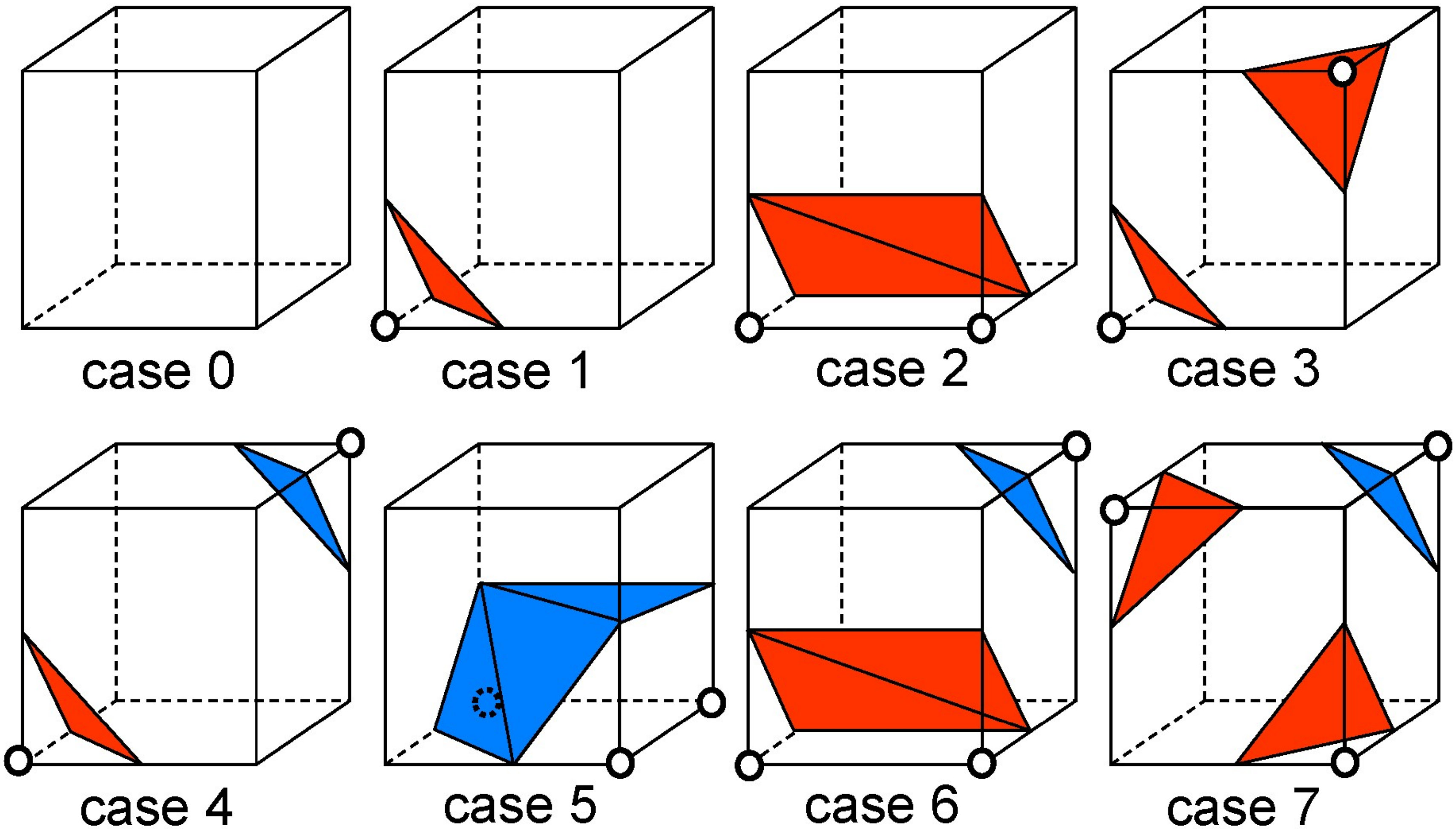
Read (required):

- Real-Time Volume Graphics, Chapters 5.5 and 5.6 (you already had to read - 5.4)
  (*Local Volume Illumination*)

Look at (optional):

- Riemannian Geometry for Scientific Visualization (notes and videos [part 1])
  `https://vccvisualization.org/RiemannianGeometryTutorial/`

The marching cubes algorithm



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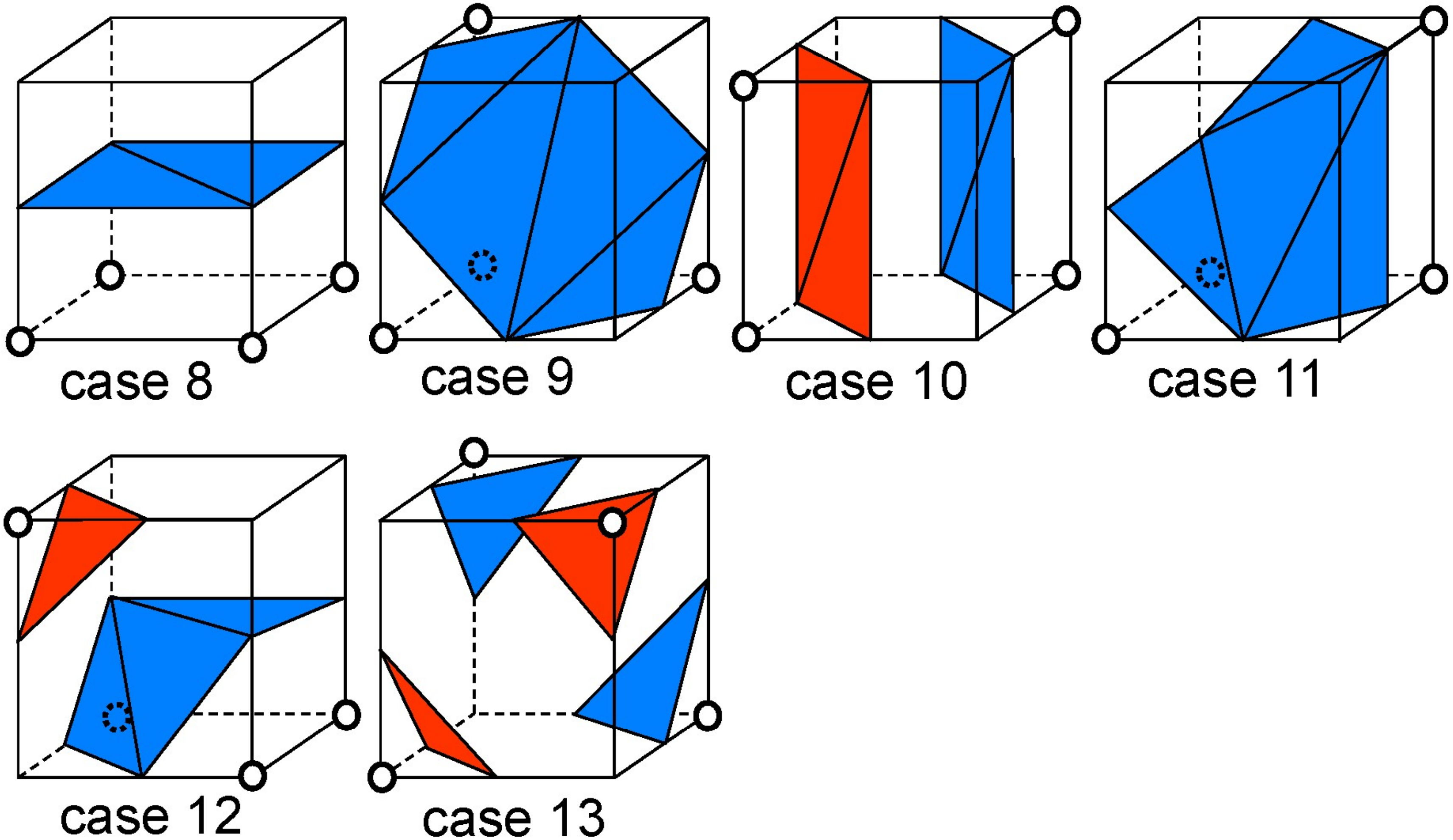
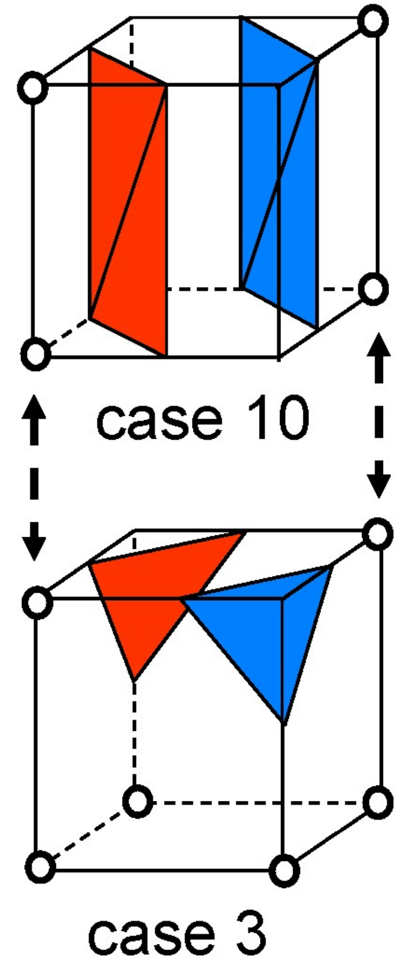


Do the pieces fit together?

- 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.

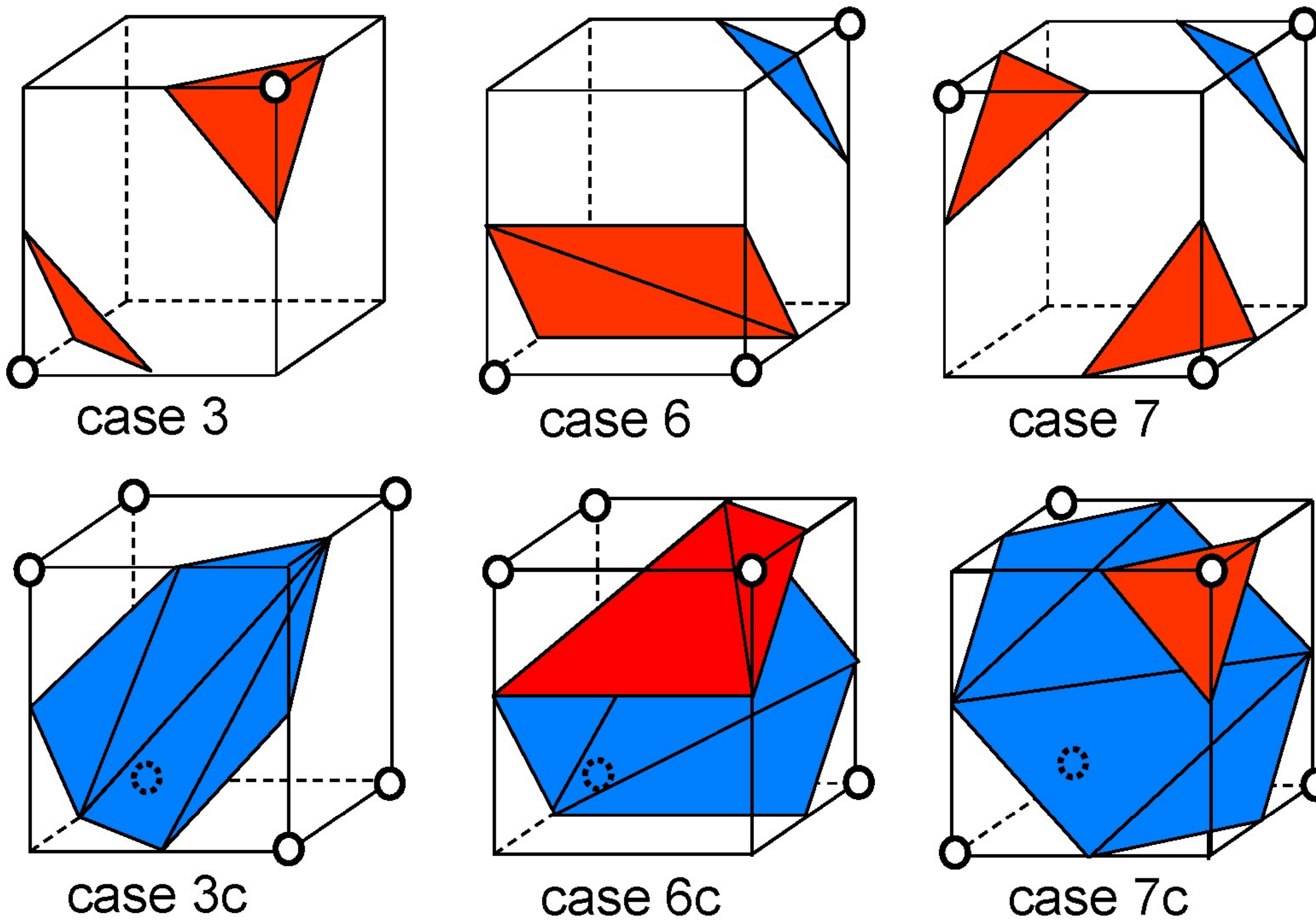


Example

- case 10, on top of
- case 3 (rotated, signs changed)

have matching signs at nodes but polygons don't fit.

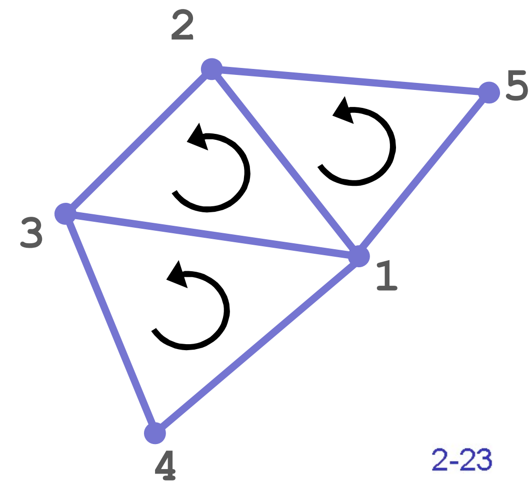
The marching cubes algorithm



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).



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)

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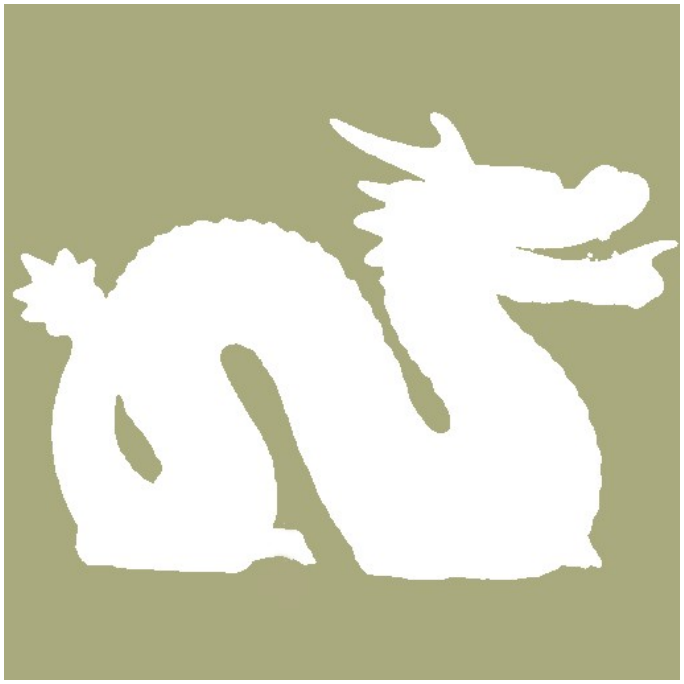
Iso-Surface / Volume Illumination

What About Volume Illumination?

Crucial for perceiving shape and depth relationships



this is a scalar volume (3D distance field)!

Local Illumination in Volumes

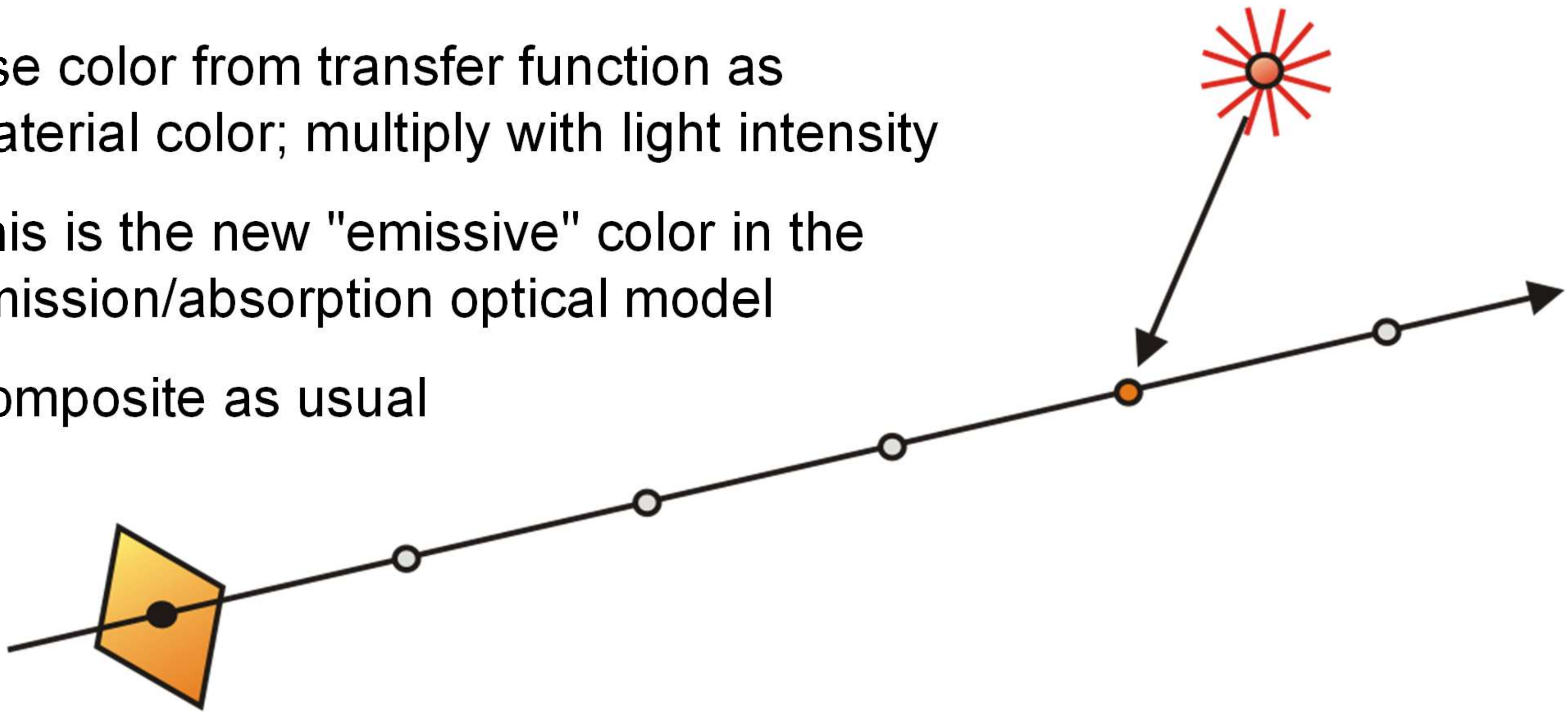
Interaction between light source and point in the volume

Local shading equation; evaluate at each point along a ray

Use color from transfer function as material color; multiply with light intensity

This is the new "emissive" color in the emission/absorption optical model

Composite as usual



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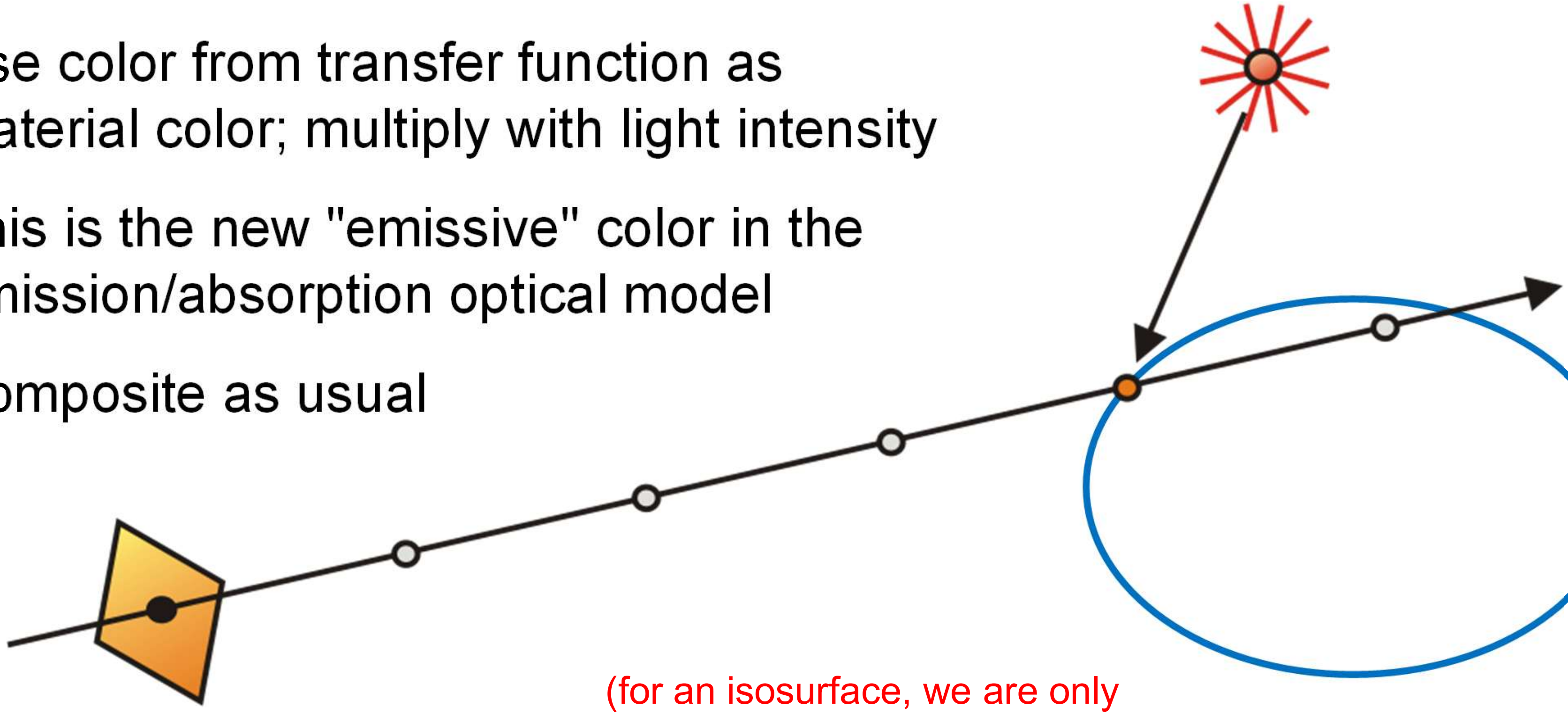
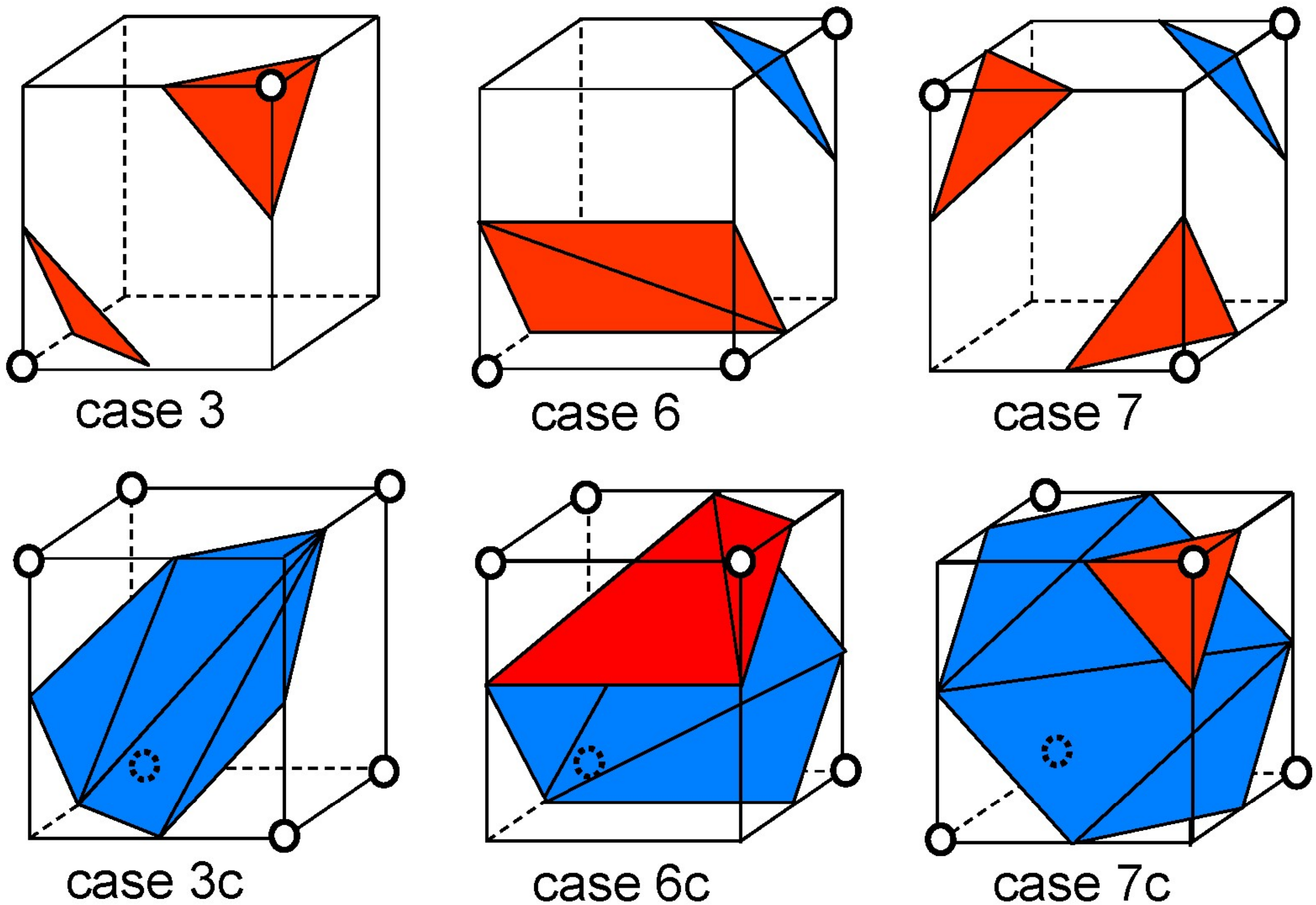


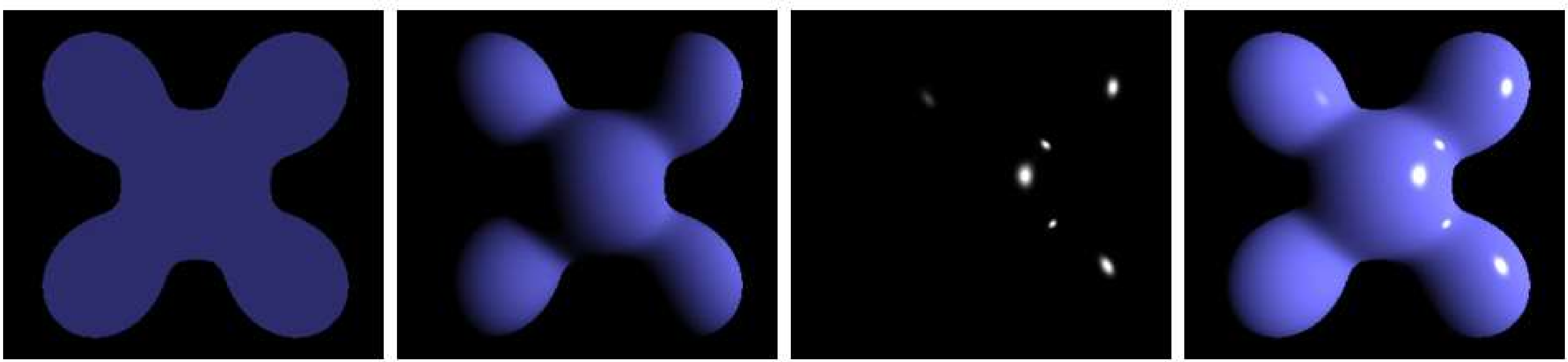
(for an isosurface, we are only interested in points *on* the surface; in marching cubes: the vertices)

The marching cubes algorithm



Local Illumination Model: Phong Lighting Model

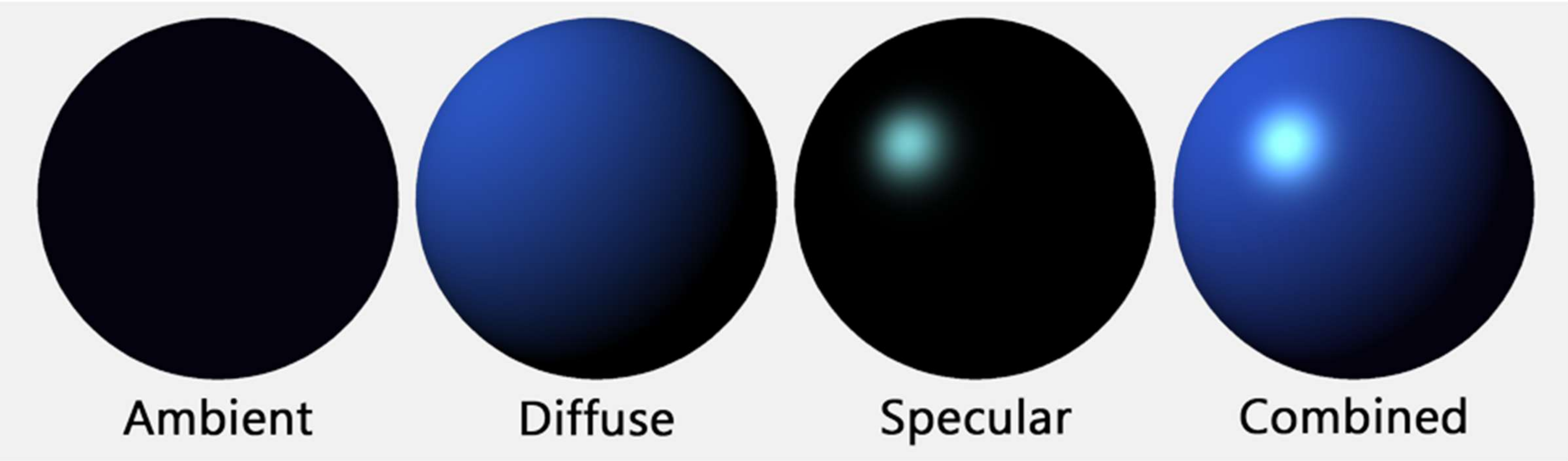
$$I_{\text{Phong}} = I_{\text{ambient}} + I_{\text{diffuse}} + I_{\text{specular}}$$



Ambient + Diffuse + Specular = Phong Reflection

Local Illumination Model: Phong Lighting Model

$$\mathbf{I}_{\text{Phong}} = \mathbf{I}_{\text{ambient}} + \mathbf{I}_{\text{diffuse}} + \mathbf{I}_{\text{specular}}$$

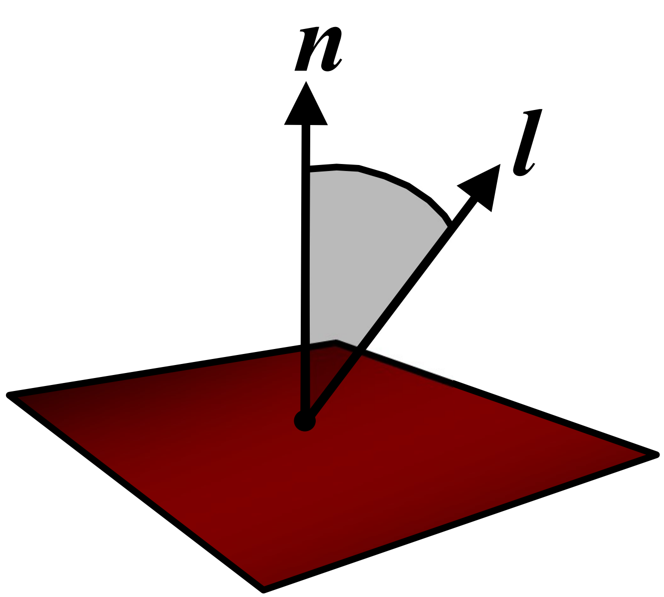


Local Shading Equations

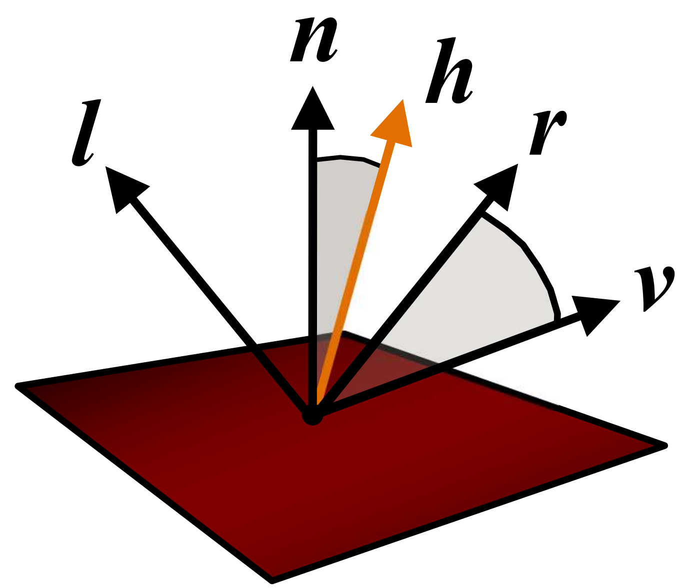
Standard volume shading adapts surface shading

Most commonly Blinn/Phong model

But what about the "surface" normal vector?



diffuse reflection



specular reflection

Local Illumination Model: Phong Lighting Model

$$\mathbf{I}_{\text{Phong}} = \mathbf{I}_{\text{ambient}} + \mathbf{I}_{\text{diffuse}} + \mathbf{I}_{\text{specular}}$$

$$\mathbf{I}_{\text{ambient}} = k_a \, \mathbf{M}_a \, \mathbf{I}_a$$

Local Illumination Model: Phong Lighting Model

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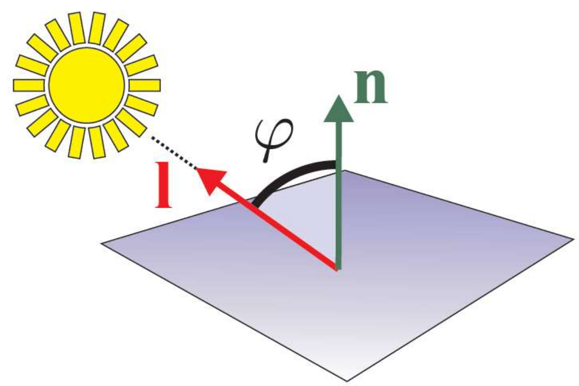


$$\begin{aligned}\mathbf{I}_{\text{diffuse}} &= k_d\,\mathbf{M}_d\,\mathbf{I}_d \cos\varphi \qquad \text{if } \varphi \leq \frac{\pi}{2} \\ &= k_d\,\mathbf{M}_d\,\mathbf{I}_d \max((\mathbf{n}\cdot\mathbf{l}), 0)\end{aligned}$$

The Dot Product (Scalar / Inner Product)

Cosine of angle between two vectors times their lengths

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

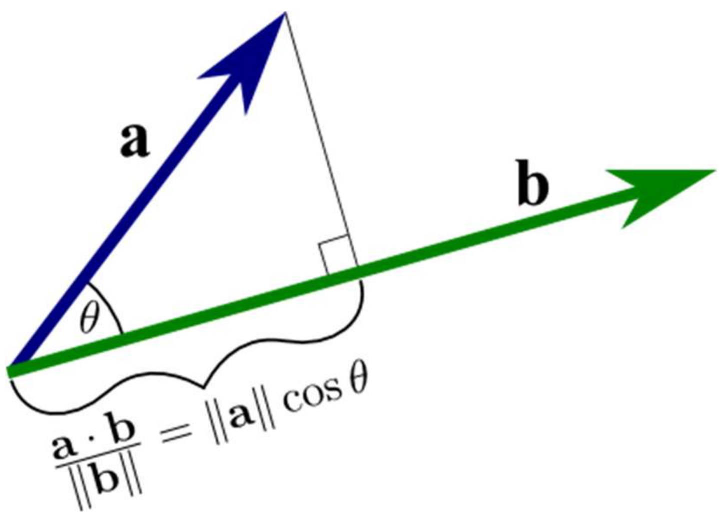
(geometric definition,
independent of coordinates)

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

(standard inner product
in Cartesian coordinates)

Many uses:

- Project vector onto another vector
- Project into basis (using the dual basis, see later)
- Project into tangent plane



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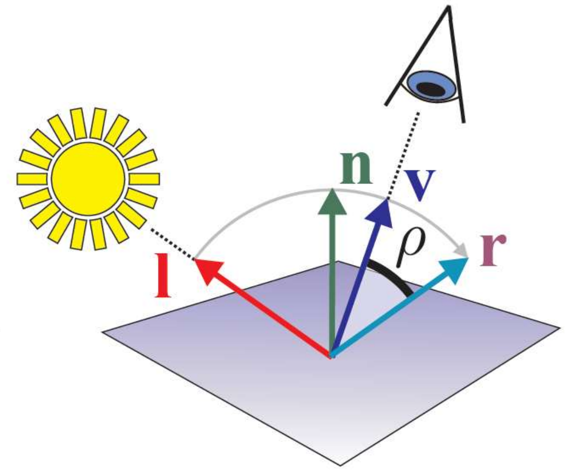

$$\begin{aligned}\mathbf{I}_{\text{specular}} &= k_s\, \mathbf{M}_s\, \mathbf{I}_s \cos^n \rho\,, \quad \text{if } \rho \leq \frac{\pi}{2} \\ &= k_s\, \mathbf{M}_s\, \mathbf{I}_s\, (\mathbf{r} \cdot \mathbf{v})^n\end{aligned}$$

must also clamp!

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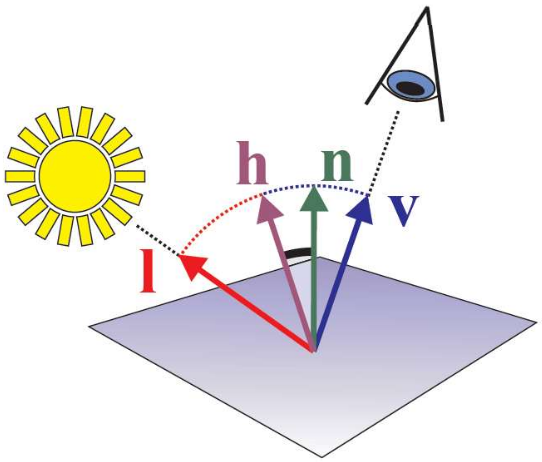


$$\mathbf{I}_{\text{specular}} \approx k_s\, \mathbf{M}_s\, \mathbf{I}_s\, (\mathbf{h} \cdot \mathbf{n})^n$$

must also clamp!

$$\mathbf{h} = \frac{\mathbf{v} + \mathbf{l}}{\|\mathbf{v} + \mathbf{l}\|}$$

half-way vector

The Gradient as Normal Vector

Gradient of the scalar field gives direction+magnitude of fastest change

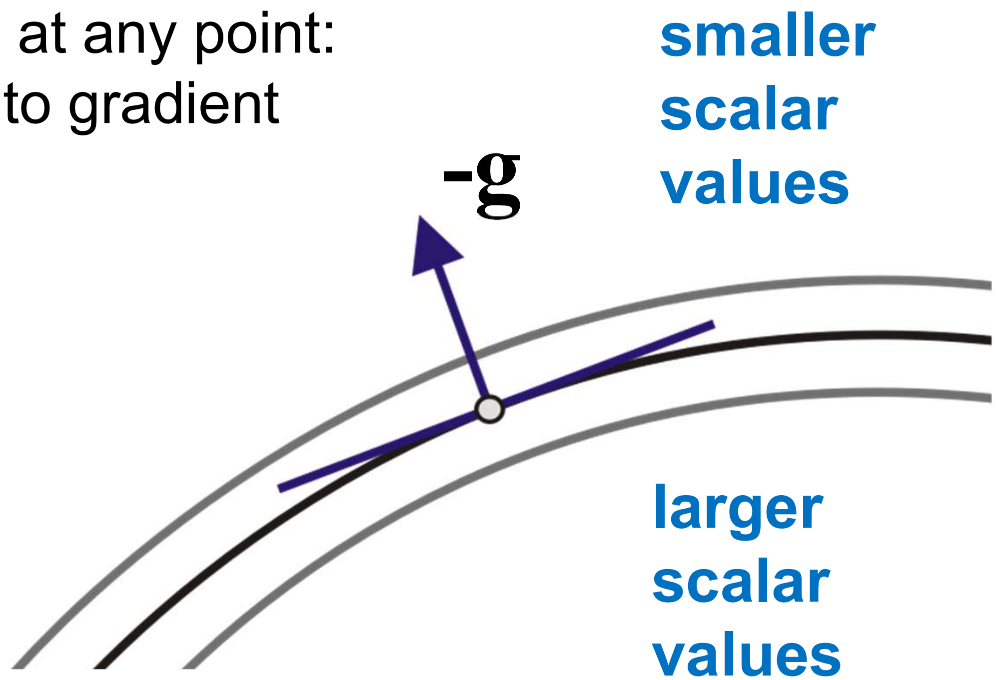
$$\mathbf{g} = \nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z} \right)^{\mathbf{T}}$$

(only correct in Cartesian coordinates: see later)

Local approximation to isosurface at any point:
tangent plane = plane orthogonal to gradient

Normal of this isosurface:
normalized gradient vector
(negation is common convention)

$$\mathbf{n} = -\mathbf{g}/|\mathbf{g}|$$



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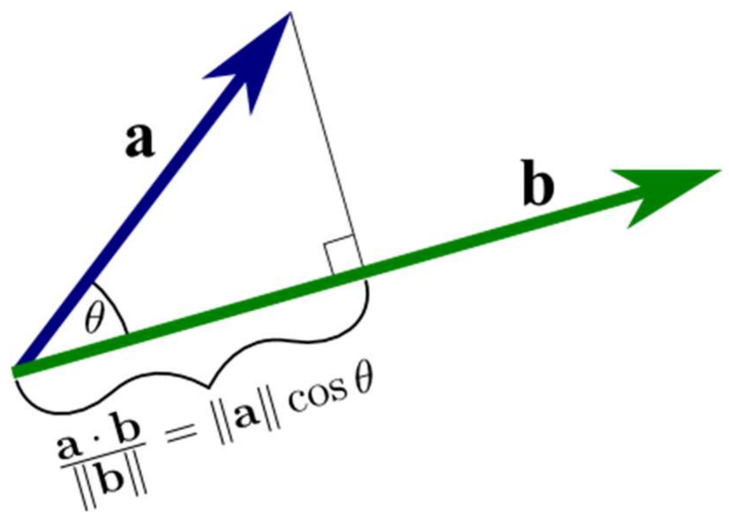

Gradient and Directional Derivative

Gradient $\nabla f(x,y,z)$ of scalar function $f(x,y,z)$:

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

(only correct in Cartesian coordinates: see later)

Directional derivative in direction $\mathbf{u}$:

$$D_{\mathbf{u}} f(x,y,z) = \nabla f(x,y,z) \cdot \mathbf{u}$$

And therefore also:

$$D_{\mathbf{u}} f(x,y,z) = ||\nabla f|| \; ||\mathbf{u}|| \; \cos\theta$$

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(only correct in Cartesian coordinates: see later)

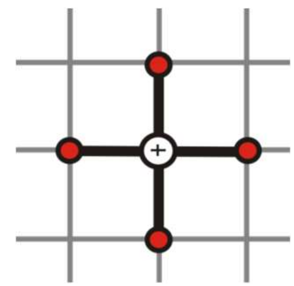
(Cartesian vector components; basis vectors not shown)

But: always need **basis vectors**! With Cartesian basis:

$$\nabla f(x,y,z) = \frac{\partial f(x,y,z)}{\partial x}\,\mathbf{i} + \frac{\partial f(x,y,z)}{\partial y}\,\mathbf{j} + \frac{\partial f(x,y,z)}{\partial z}\,\mathbf{k}$$

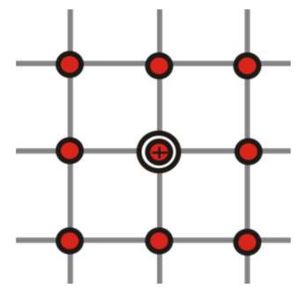
(Numerical) Gradient Reconstruction

We need to reconstruct the derivatives of a continuous function given as discrete samples



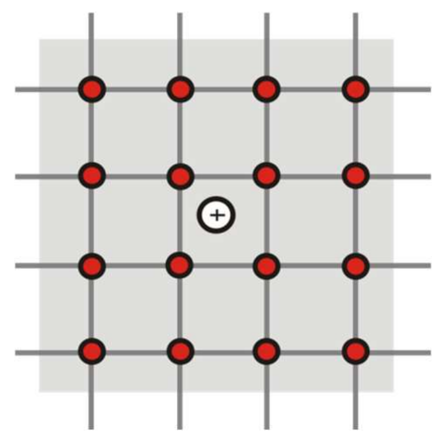
Central differences

- Cheap and quality often sufficient (2*3 neighbors in 3D)



Discrete convolution filters on grid

- Image processing filters; e.g. Sobel (3^3 neighbors in 3D)



Continuous convolution filters

- Derived continuous reconstruction filters
- E.g., the cubic B-spline and its derivatives (4^3 neighbors)

Finite Differences

Obtain first derivative from Taylor expansion

$$\begin{aligned} f(x_0+h) &= f(x_0) + \frac{f'(x_0)}{1!}h + \frac{f''(x_0)}{2!}h^2 + \dots \\ &= \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!}h^n . \end{aligned}$$

Forward differences / backward differences

$$f(x_0)' = \frac{f(x_0+h) - f(x_0)}{h} + o(h)$$

$$f(x_0)' = \frac{f(x_0) - f(x_0-h)}{h} + o(h)$$

Finite Differences

Central differences

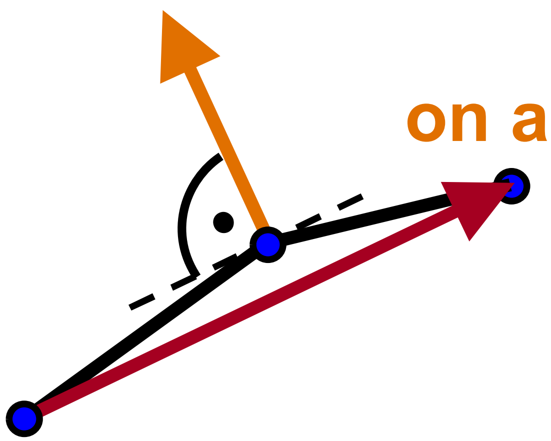
$$
\begin{aligned}
f(x_0 + h) &= f(x_0) + \frac{f'(x_0)}{1!} h + \frac{f''(x_0)}{2!} h^2 + o(h^3) \\
f(x_0 - h) &= f(x_0) - \frac{f'(x_0)}{1!} h + \frac{f''(x_0)}{2!} h^2 + o(h^3)
\end{aligned}
$$

$$
f'(x_0) = \frac{f(x_0 + h) - f(x_0 - h)}{2h} + o(h^2)
$$

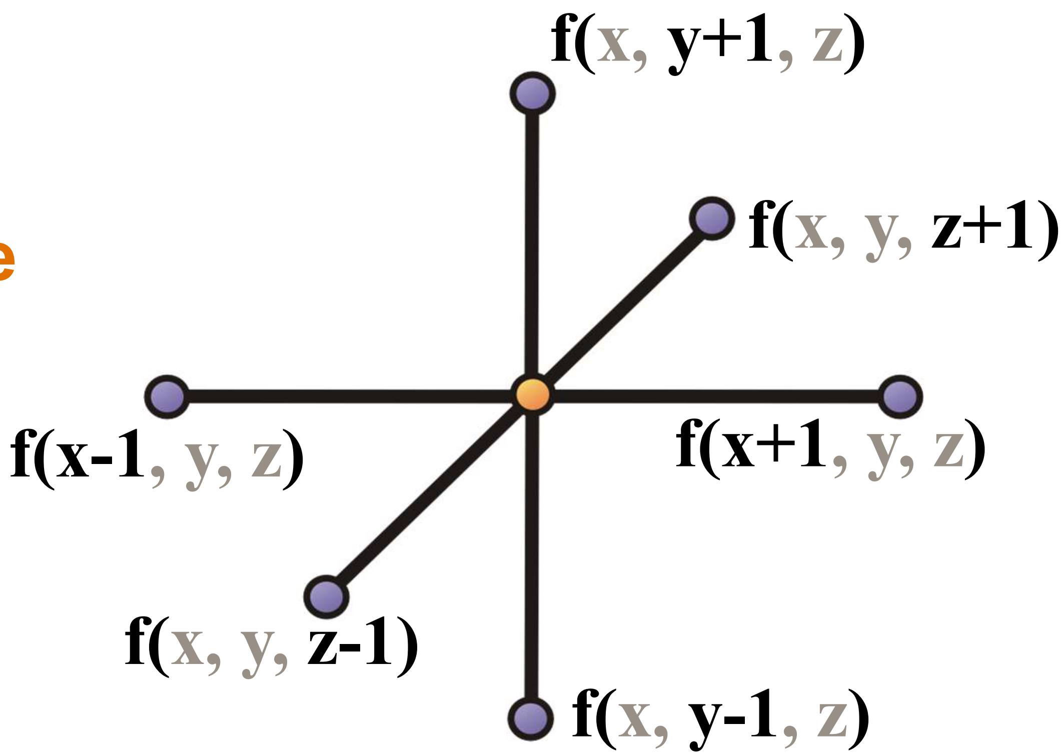
Central Differences

Need only two neighboring voxels per derivative

Most common method



on a curve




```
gx = 0.5( f(x+1, y, z) - f(x-1, y, z) )
gy = 0.5( f(x, y+1, z) - f(x, y-1, z) )
gz = 0.5( f(x, y, z+1) - f(x, y, z-1) )
```

in a volume

Thank you.

Thanks for material

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