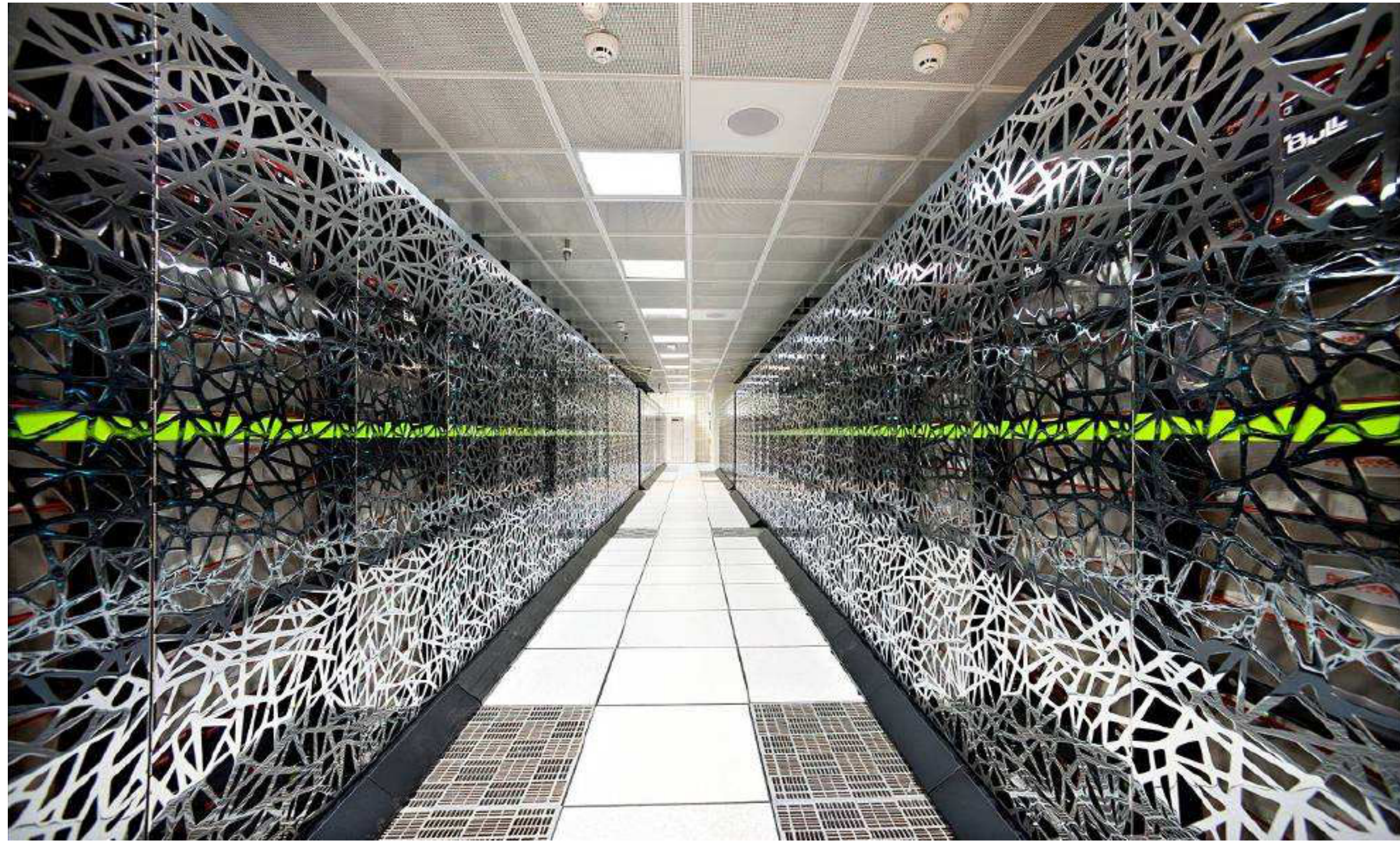


Domain Representations

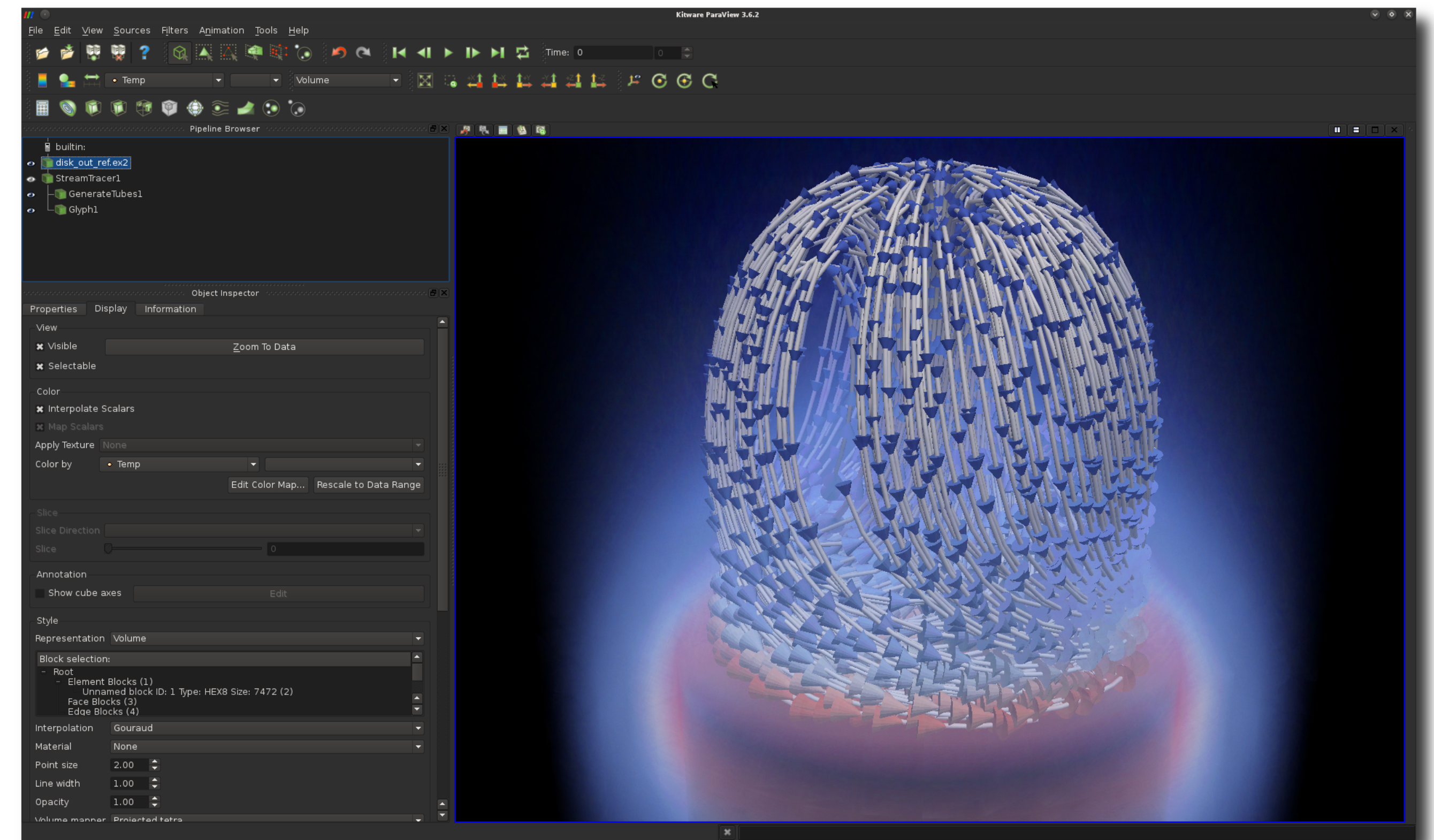
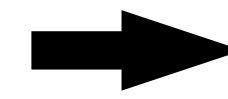
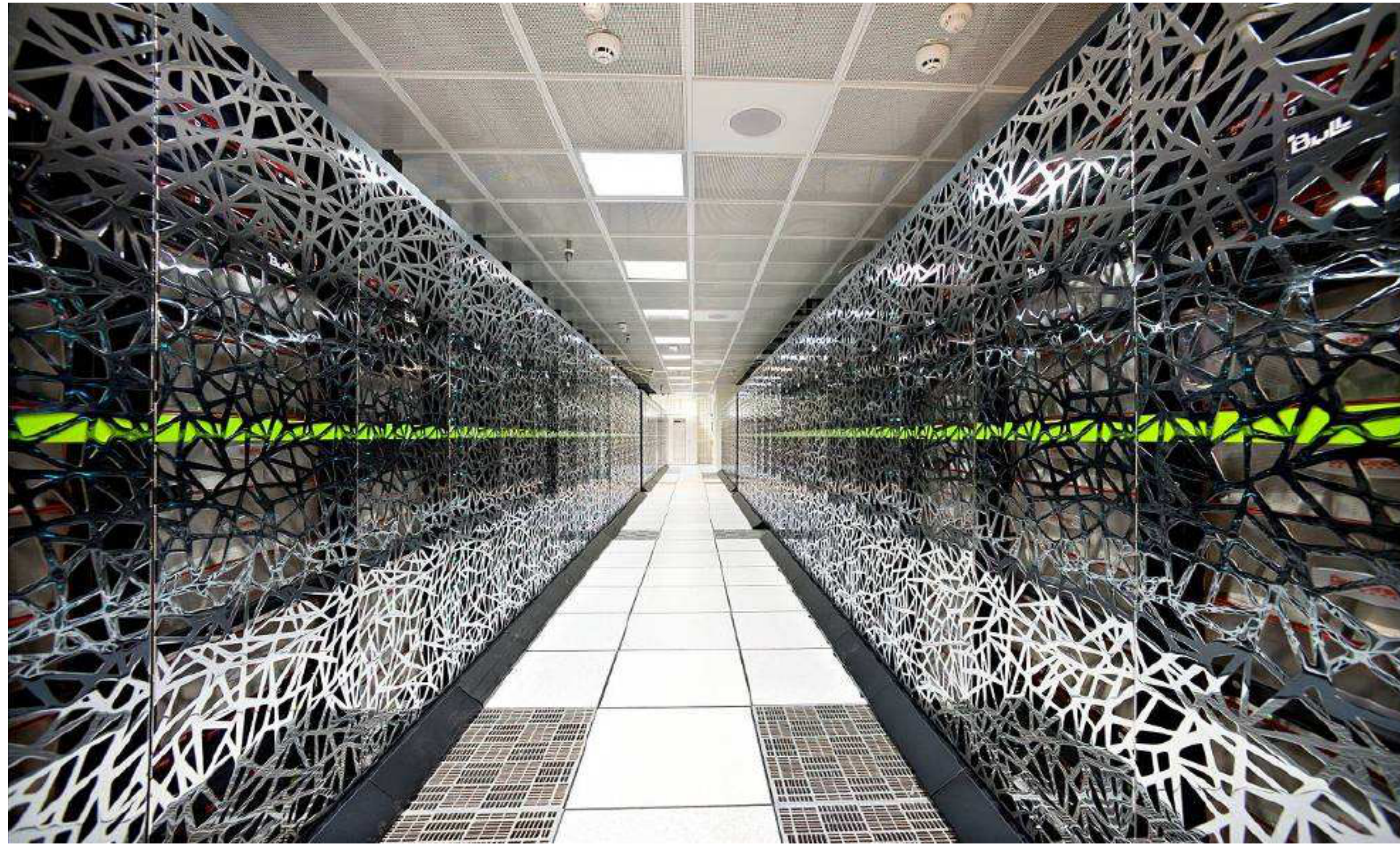
M2S

What?

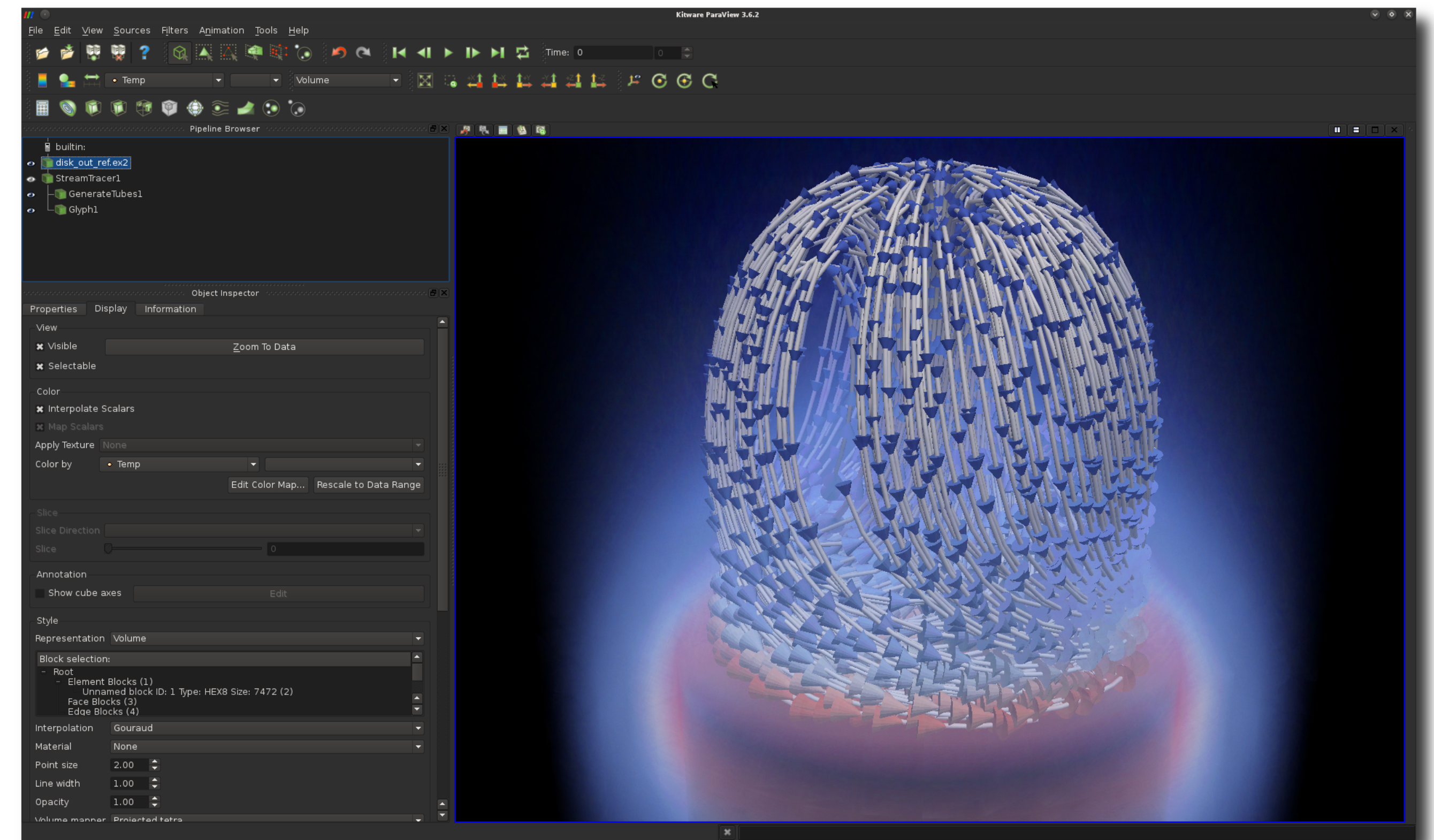
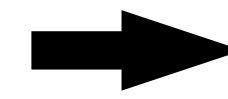
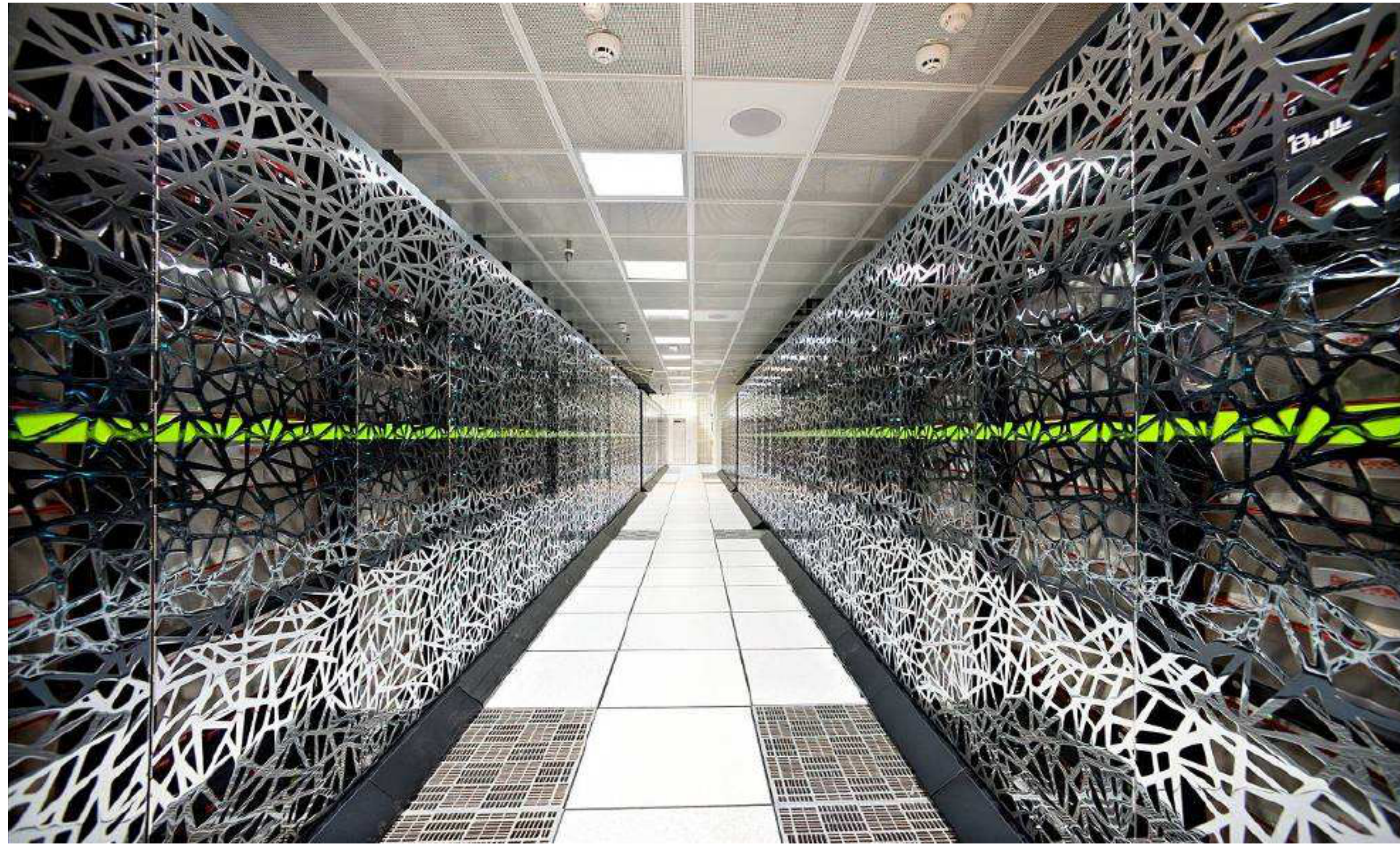
What?



What?

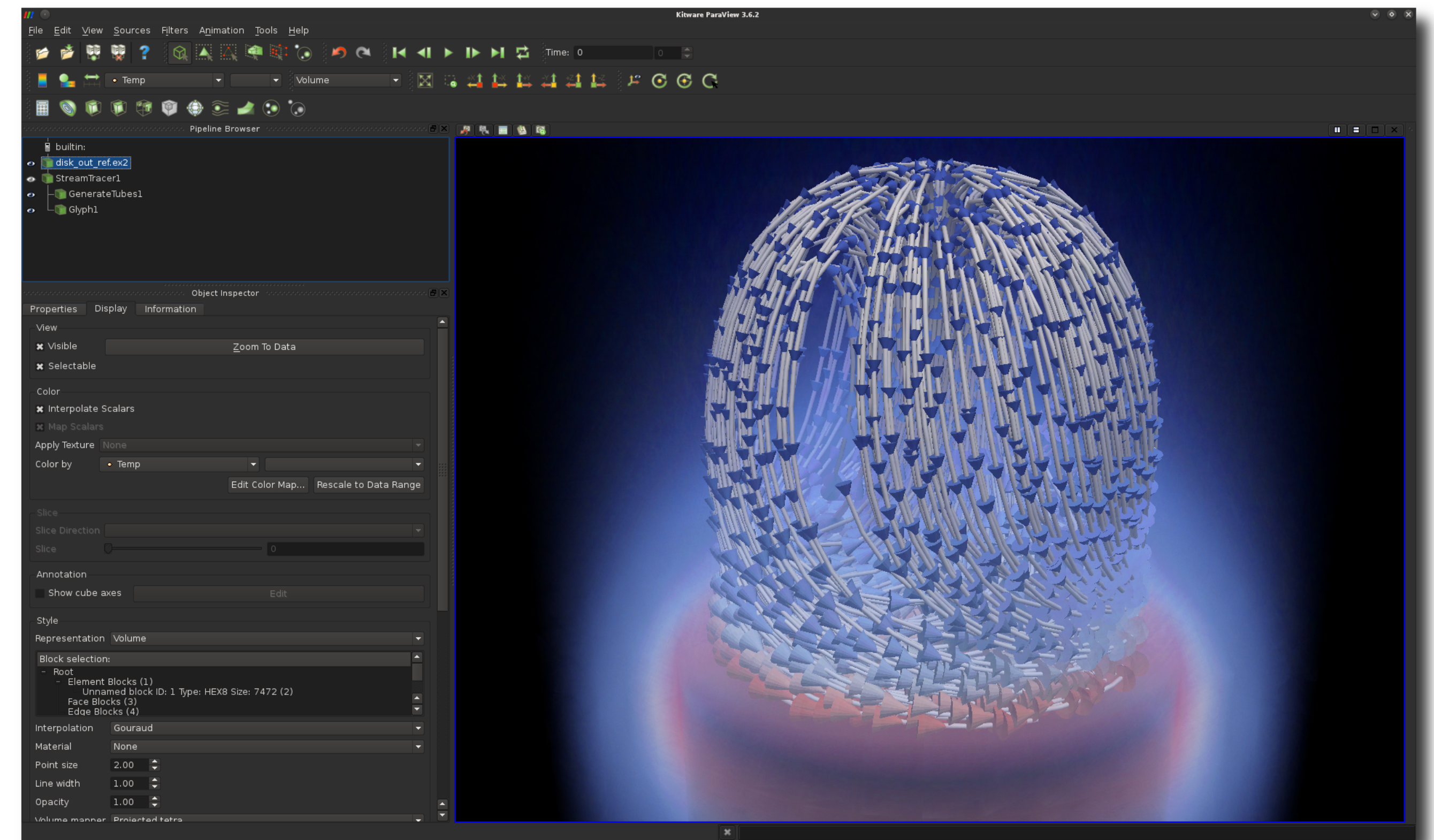
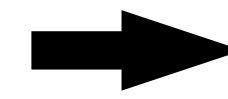
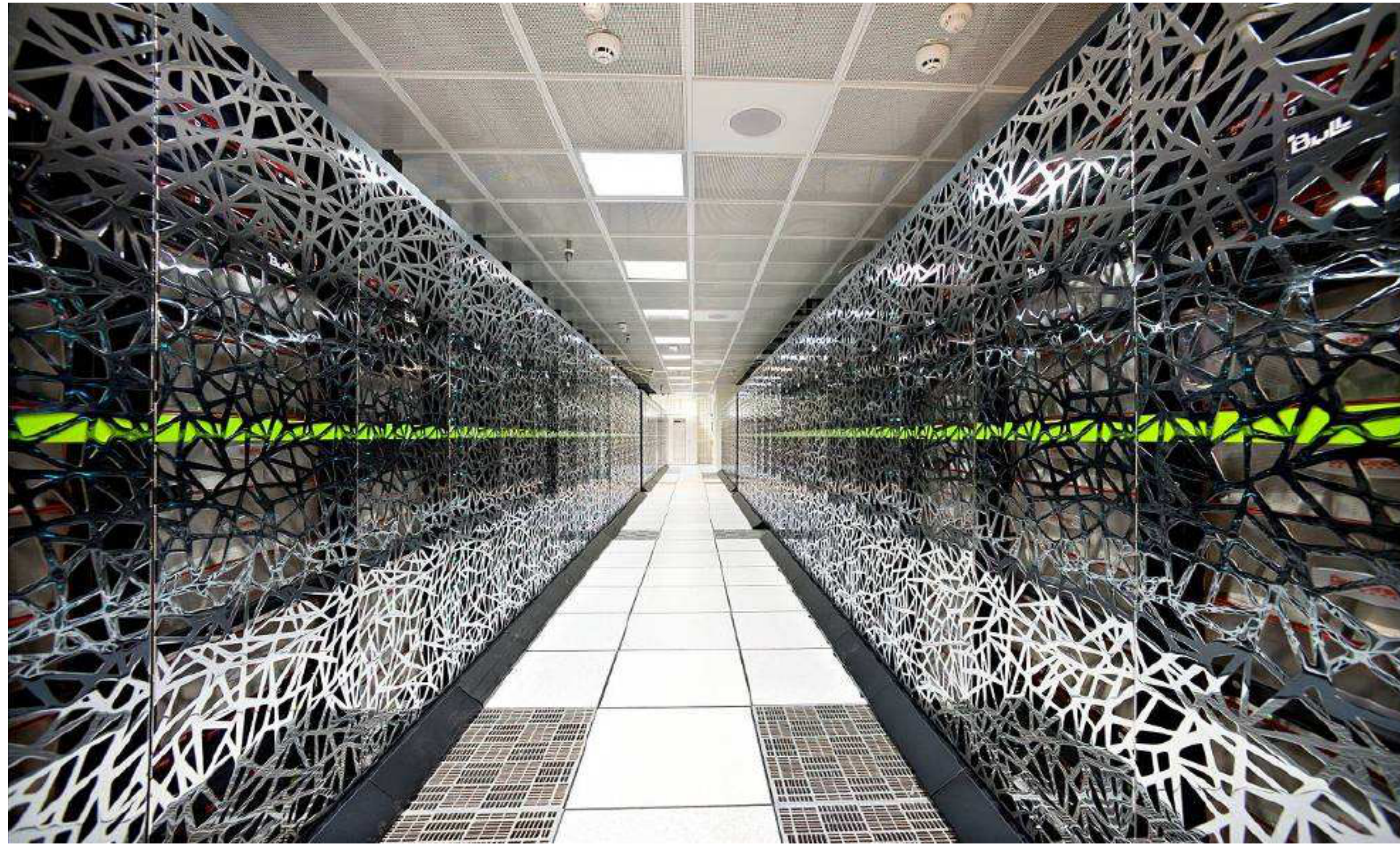


What?



- What have we computed?

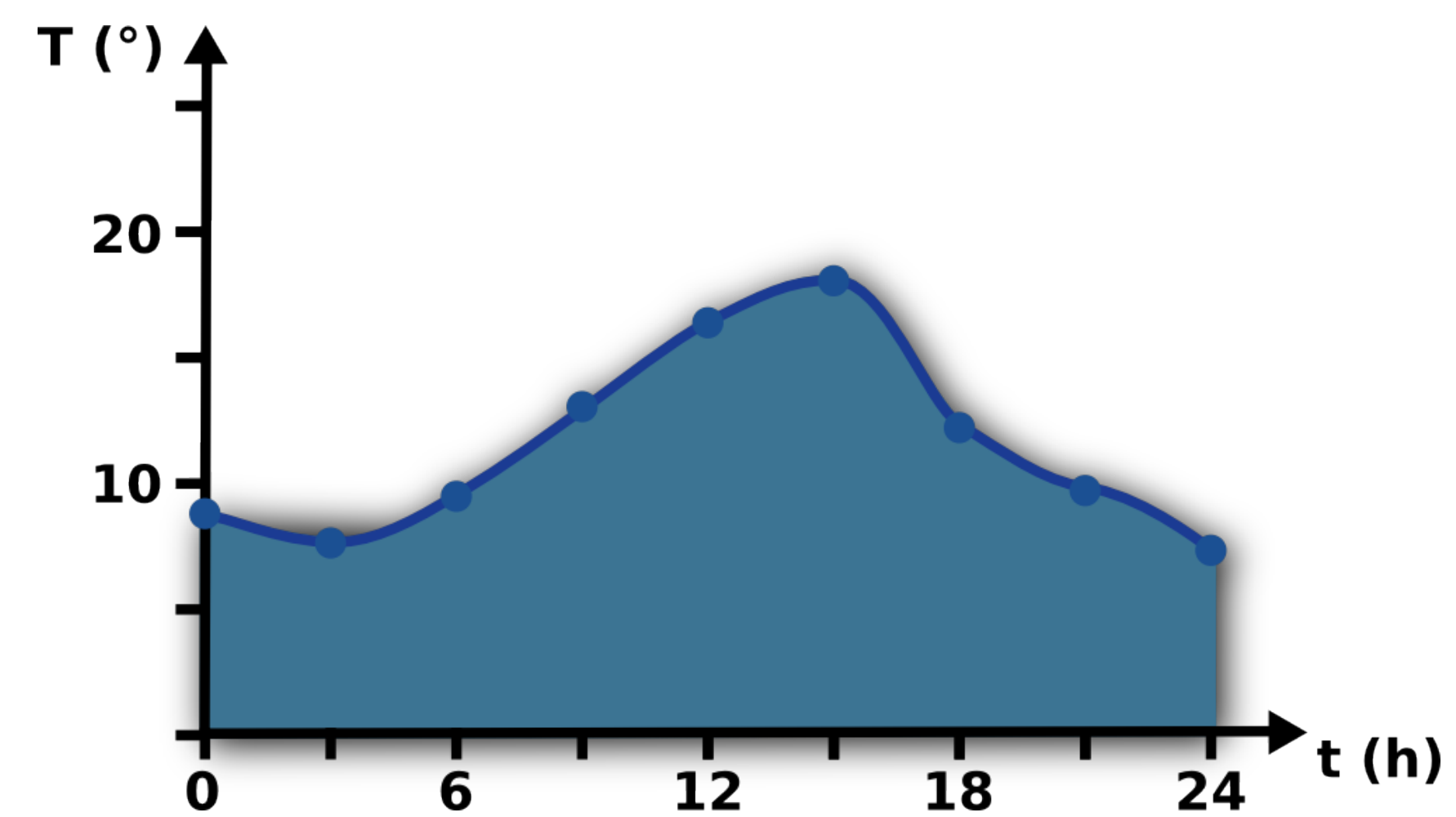
What?



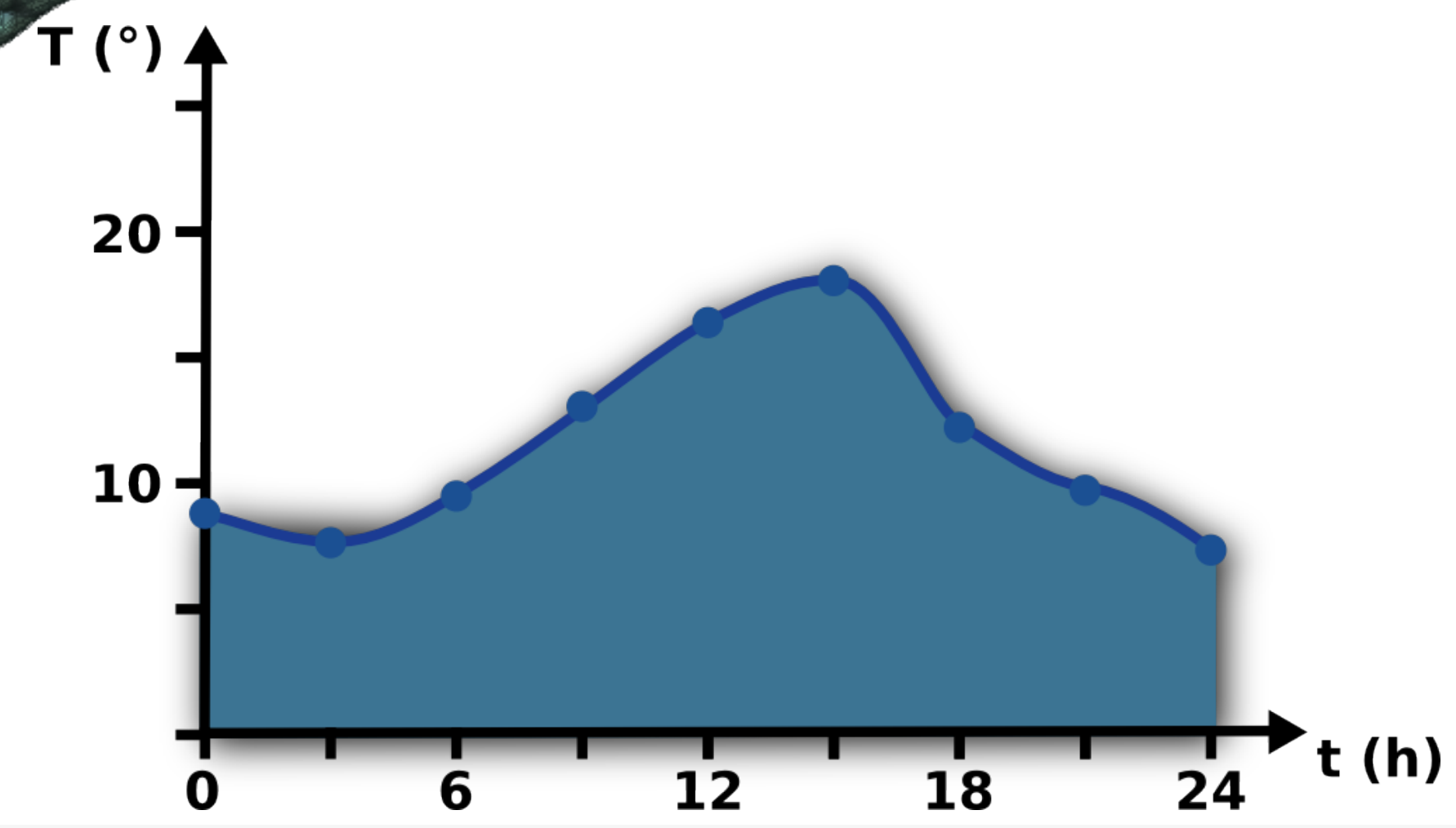
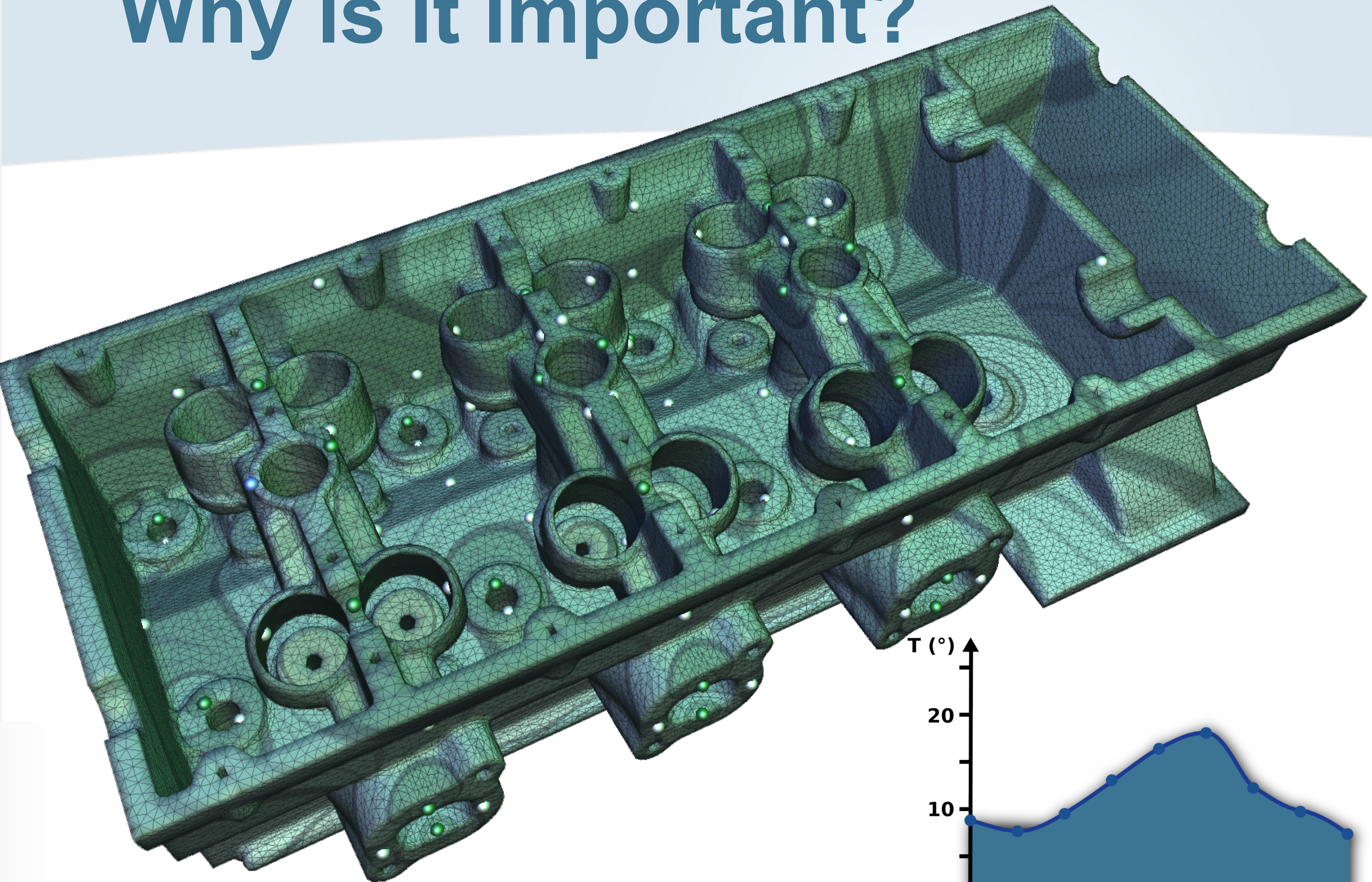
- What have we computed?
- On what kind of *discretized-space*?

Why is it important?

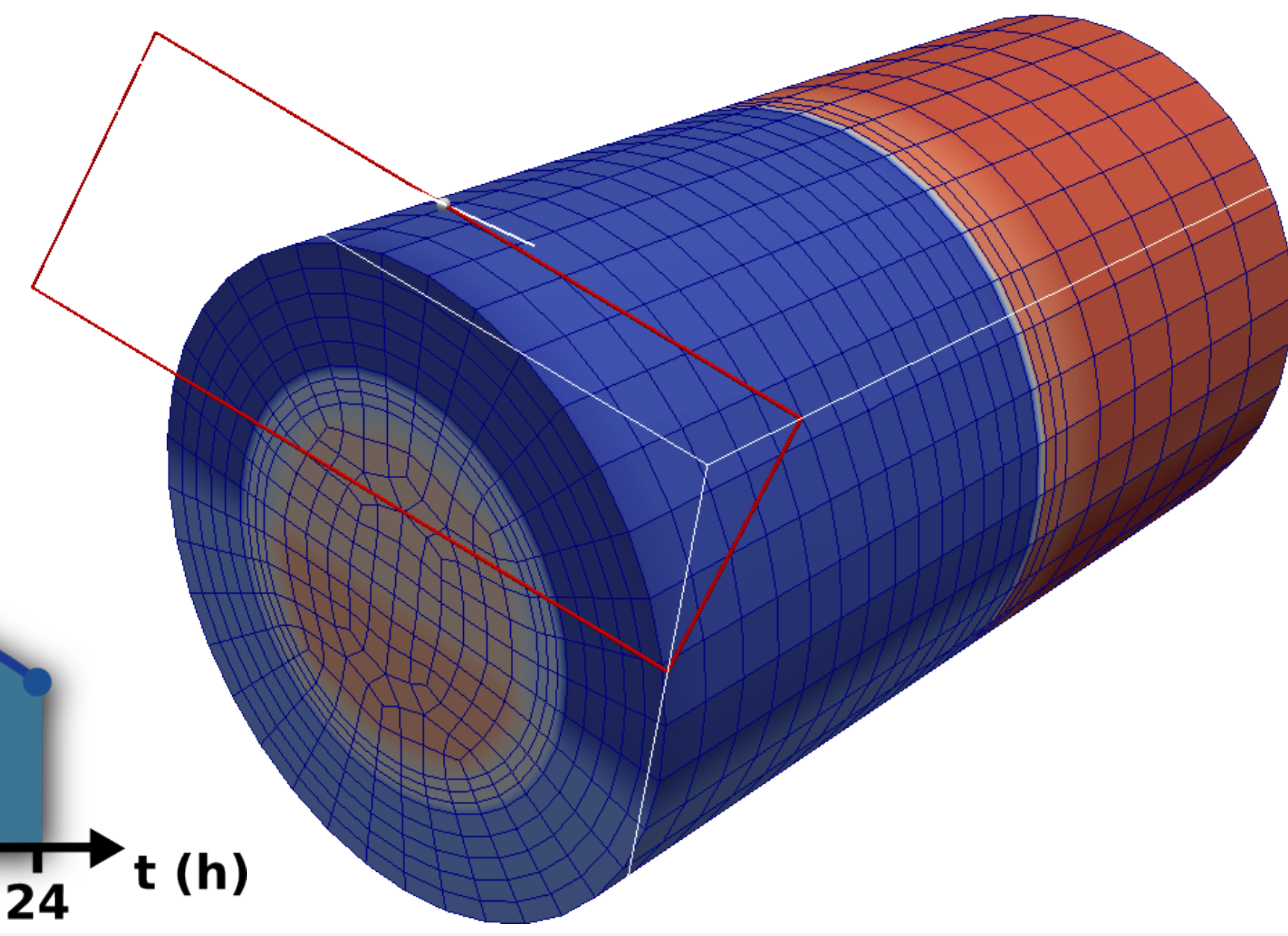
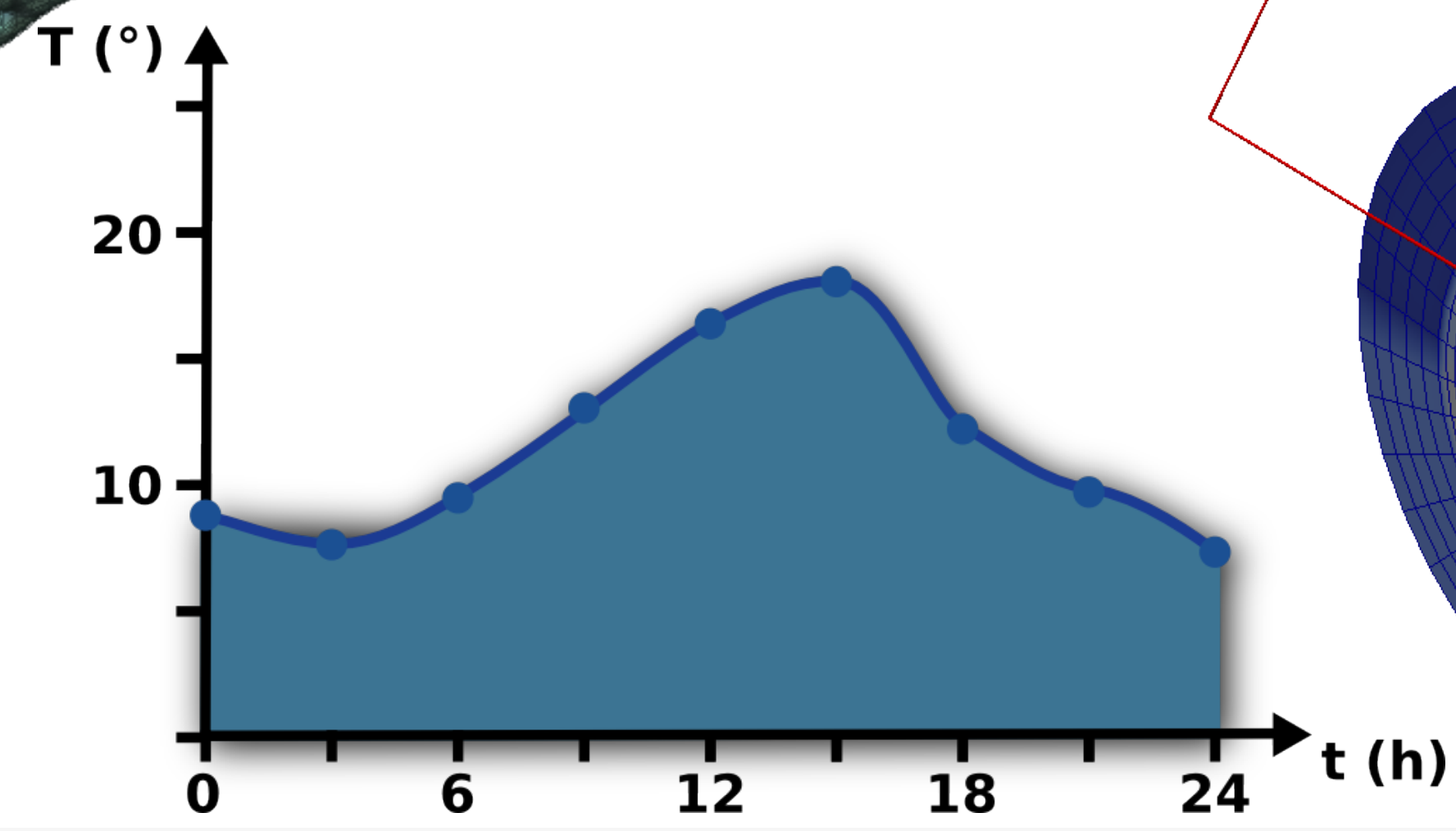
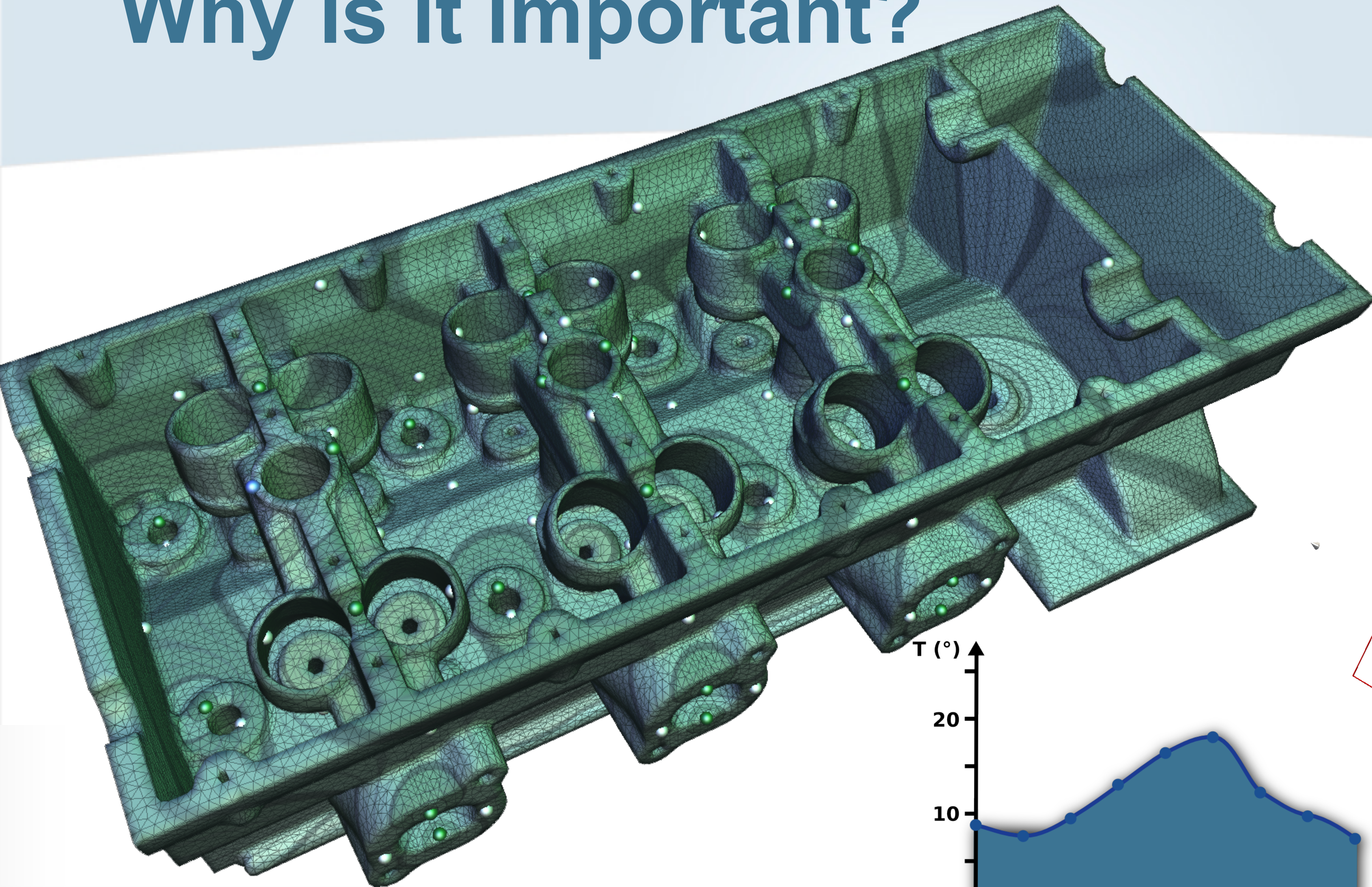
Why is it important?



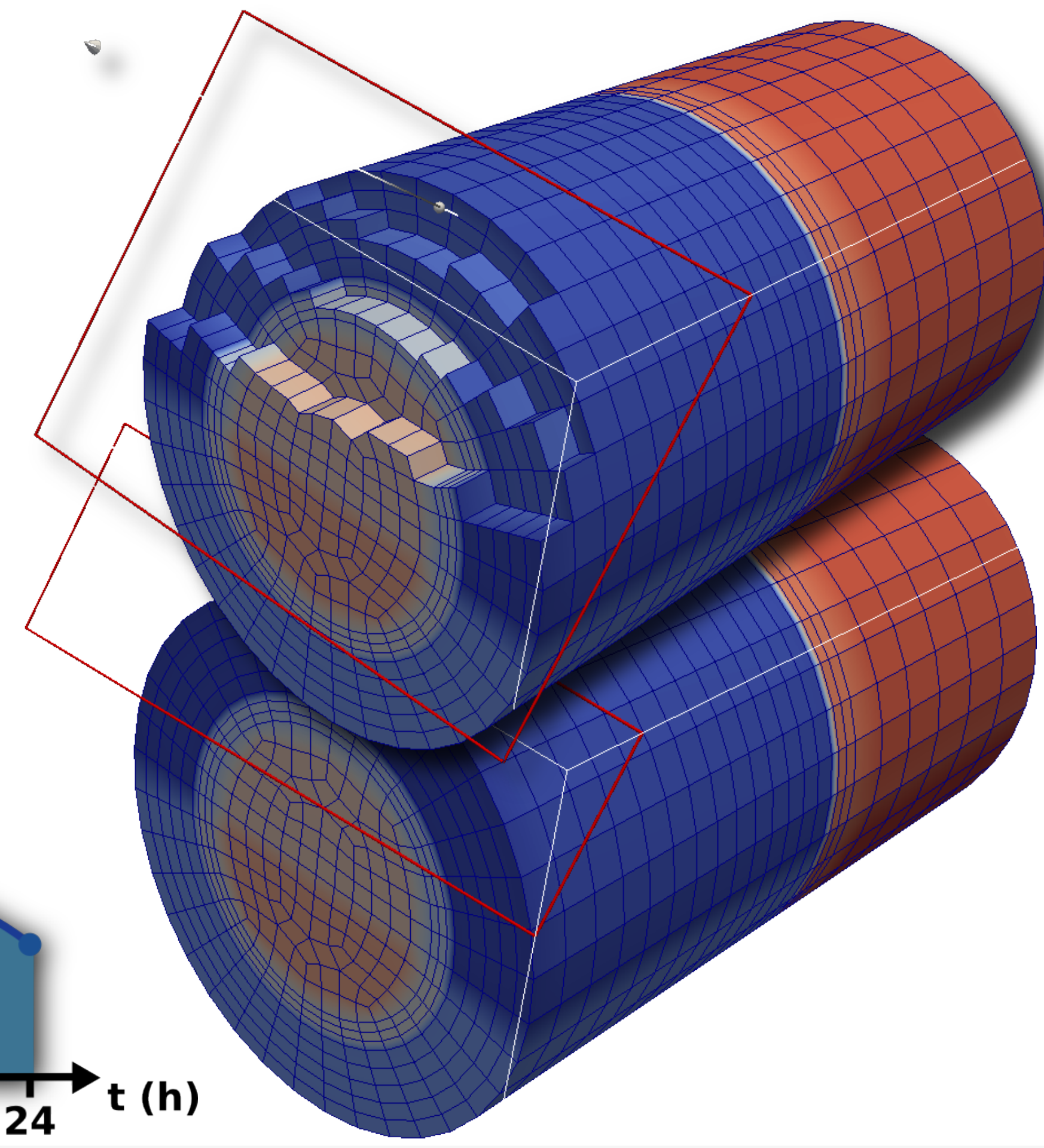
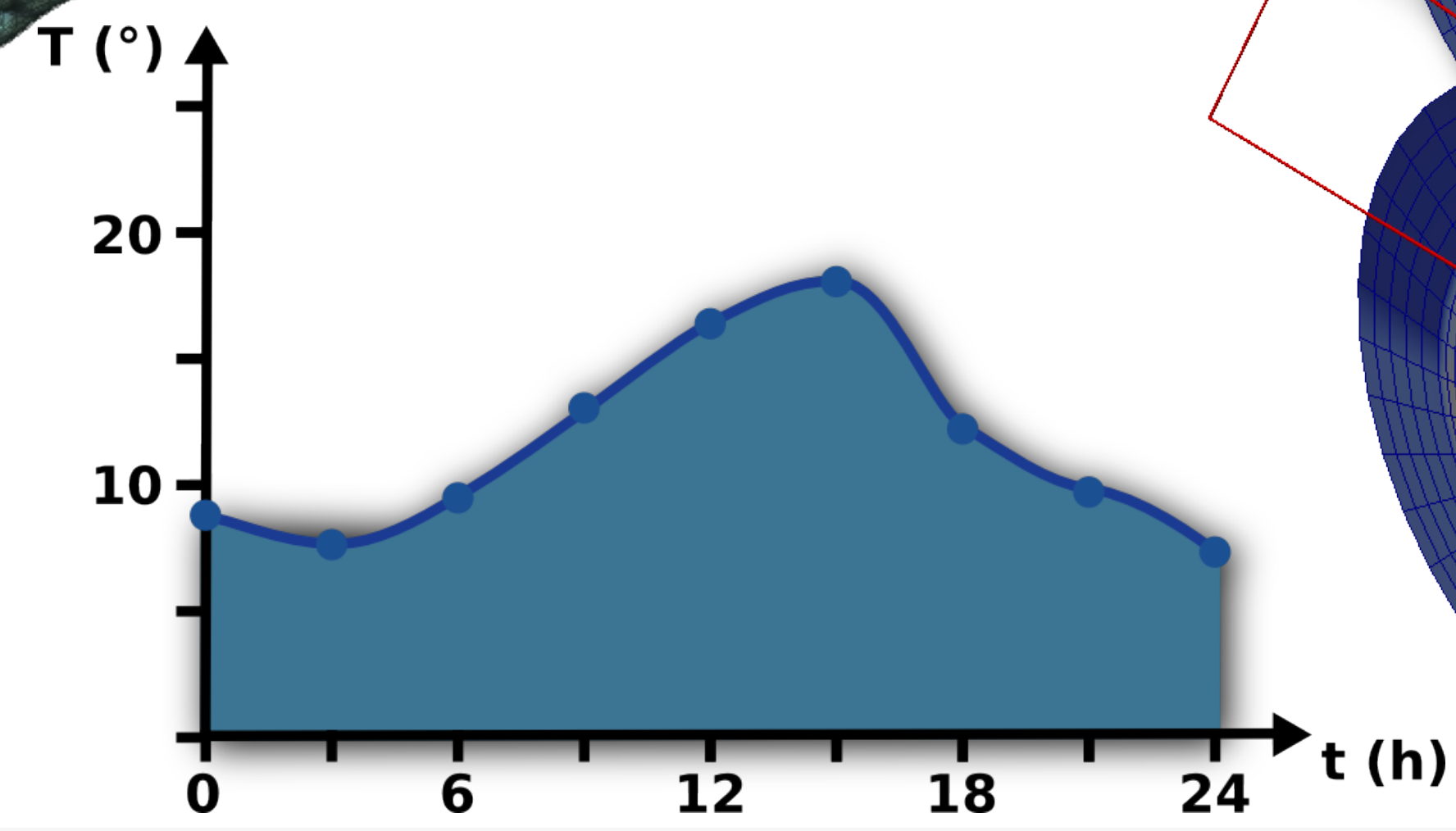
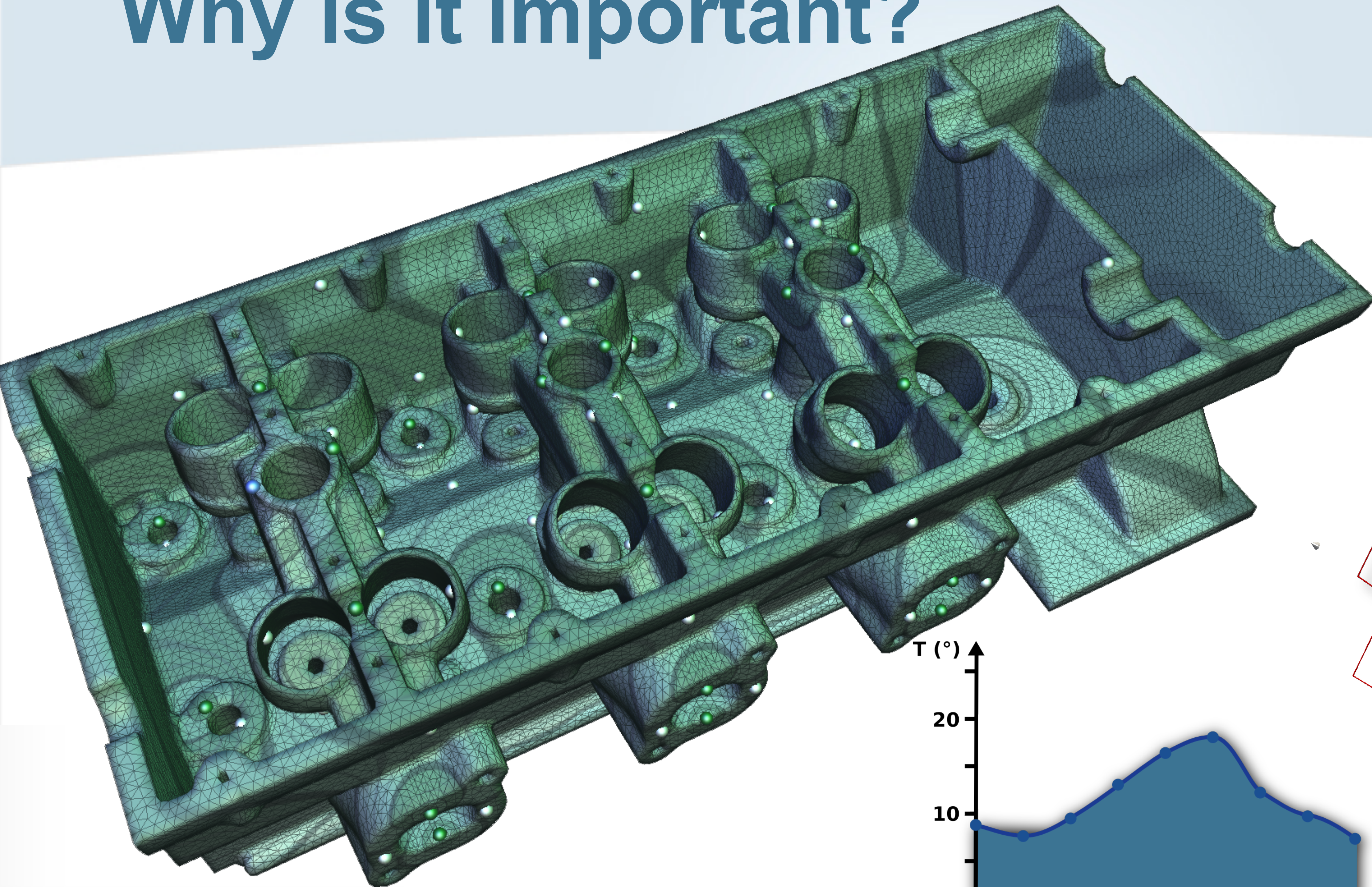
Why is it important?



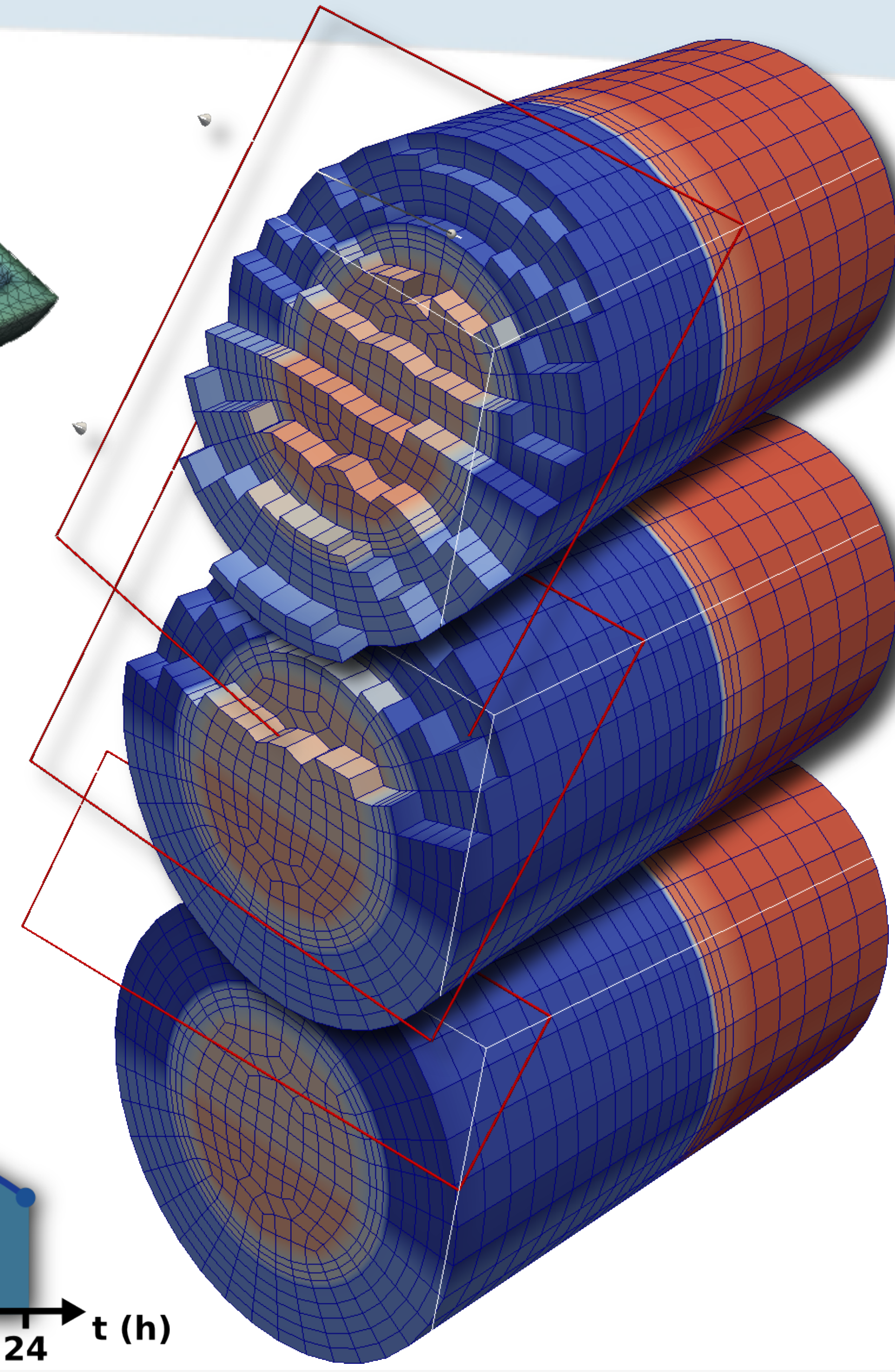
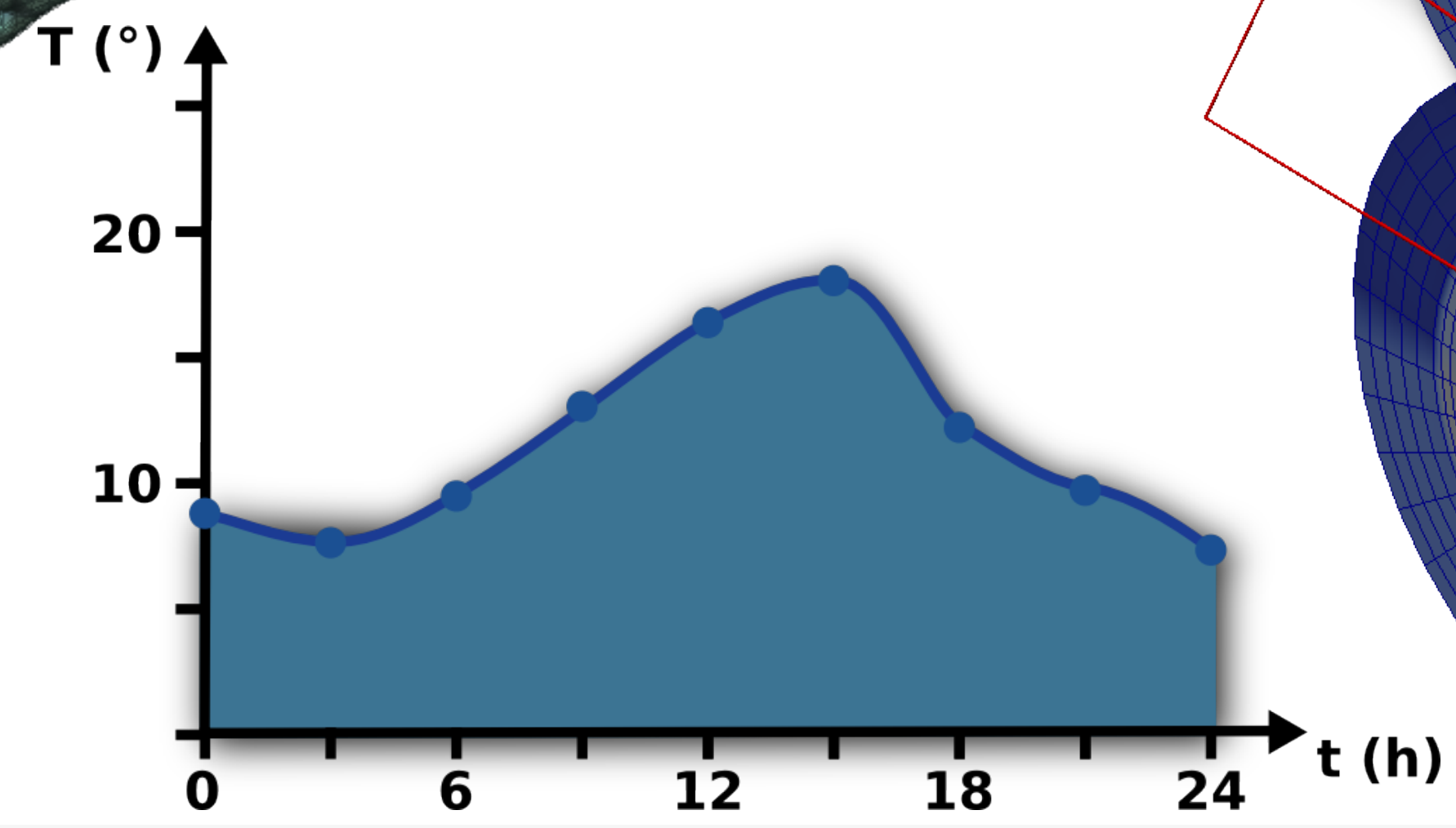
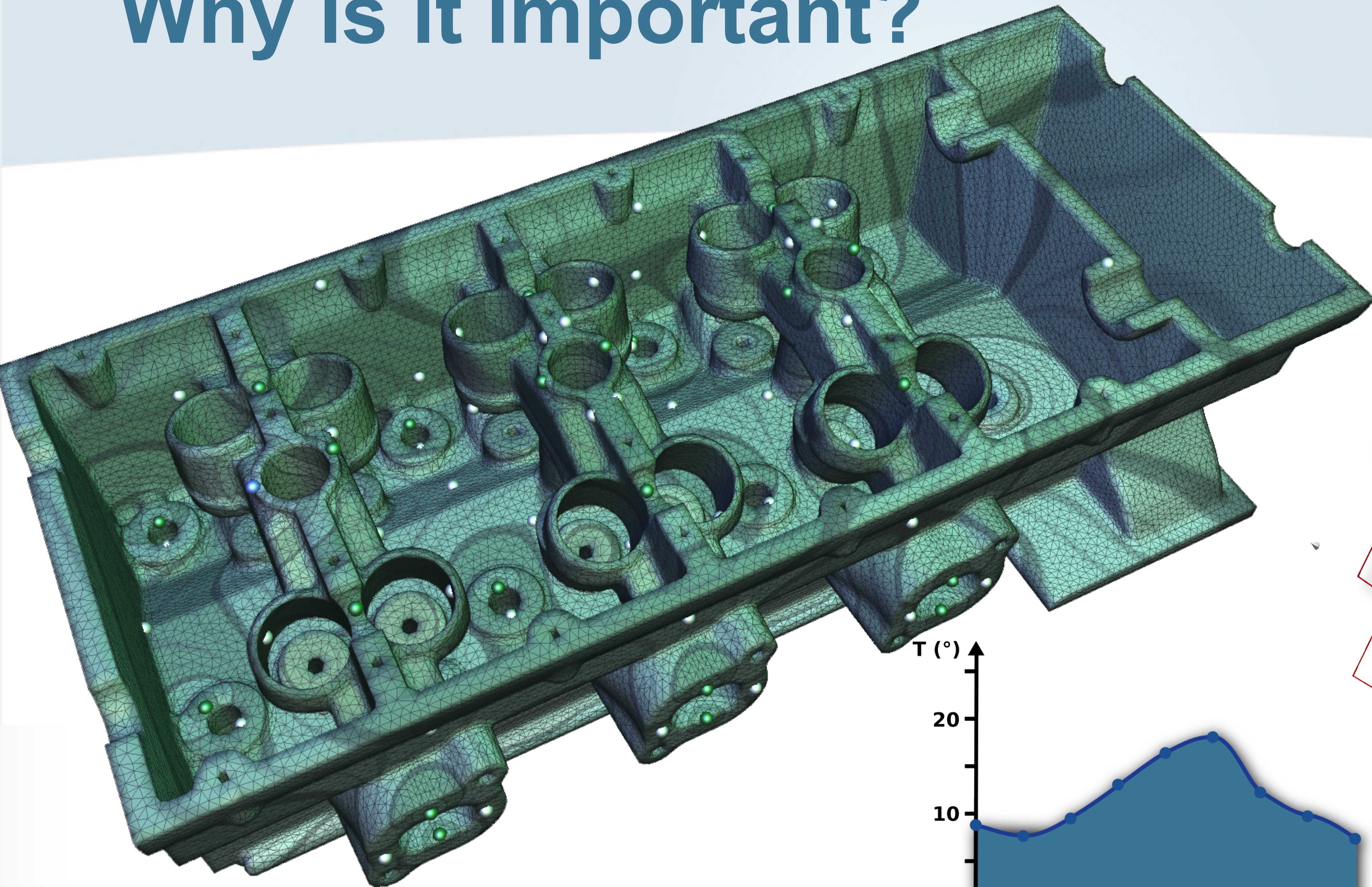
Why is it important?



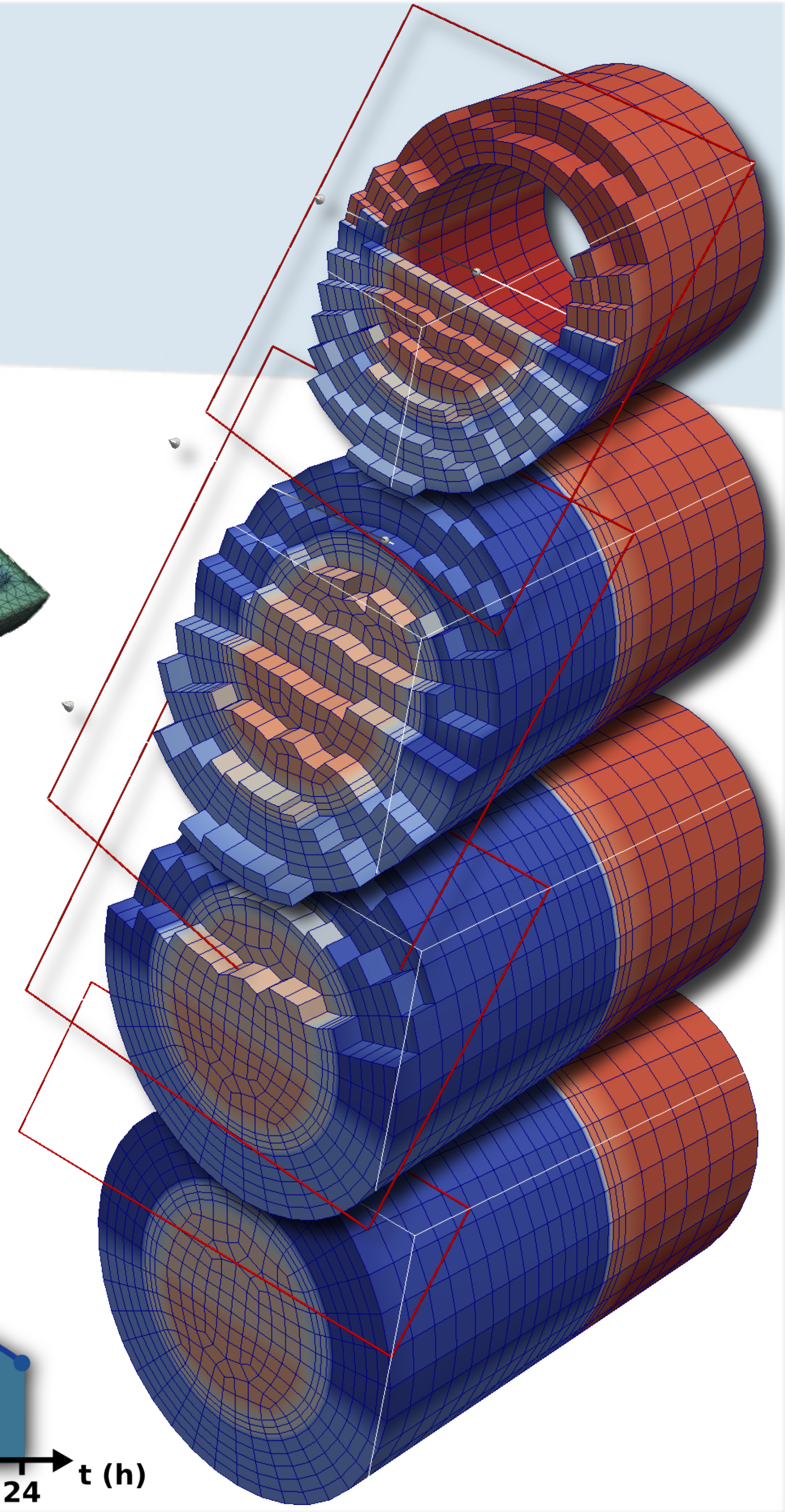
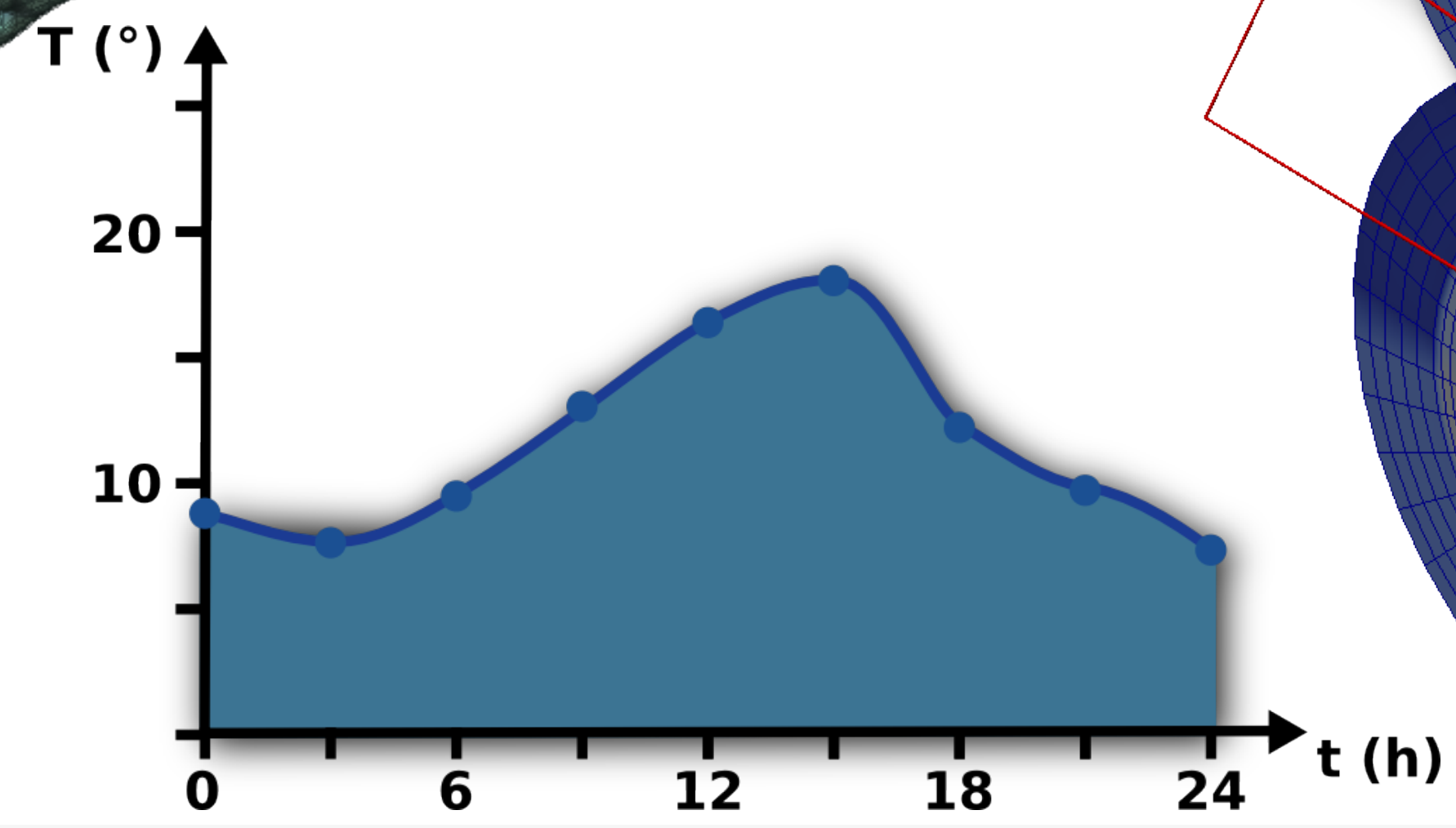
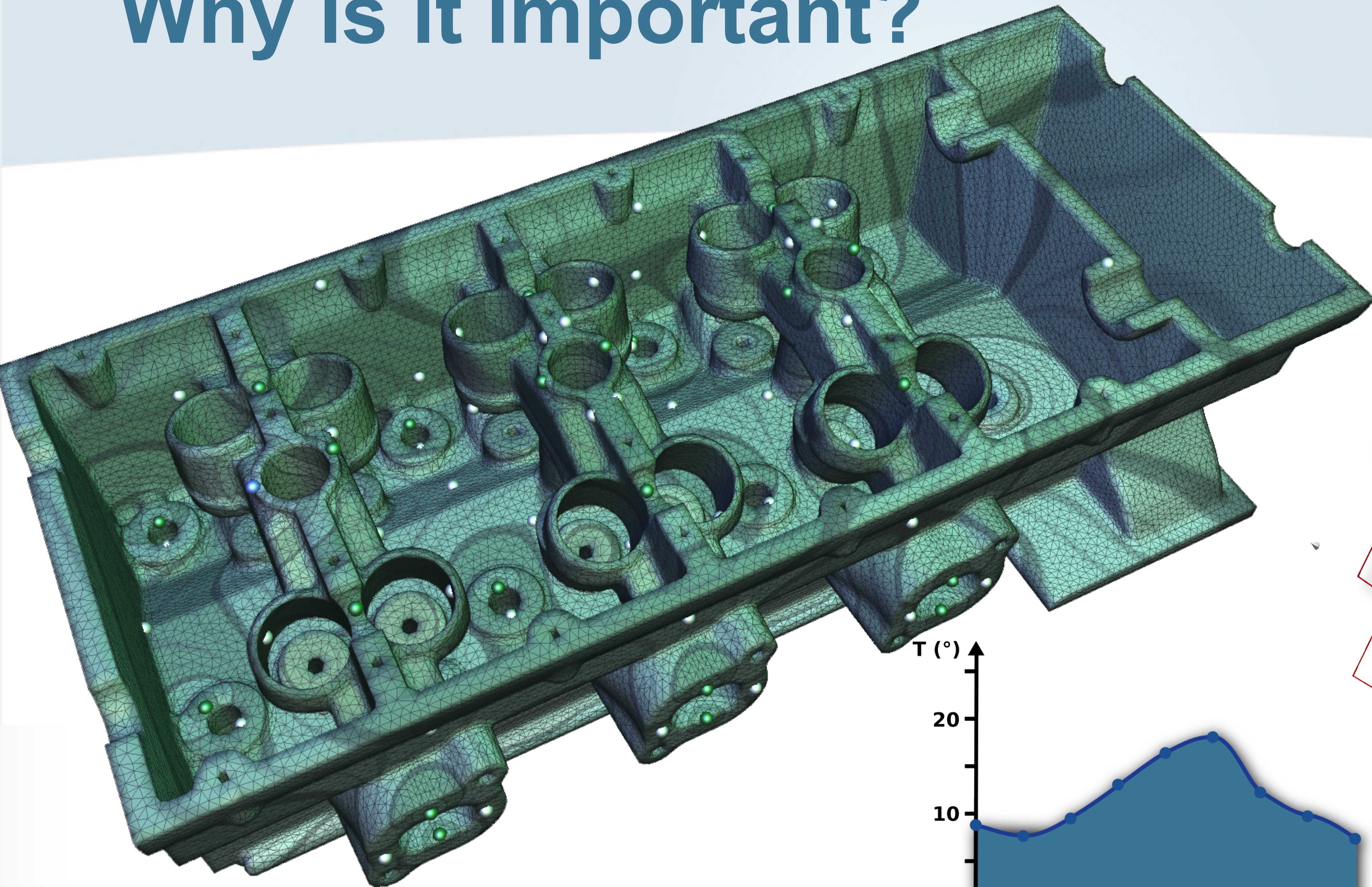
Why is it important?



Why is it important?

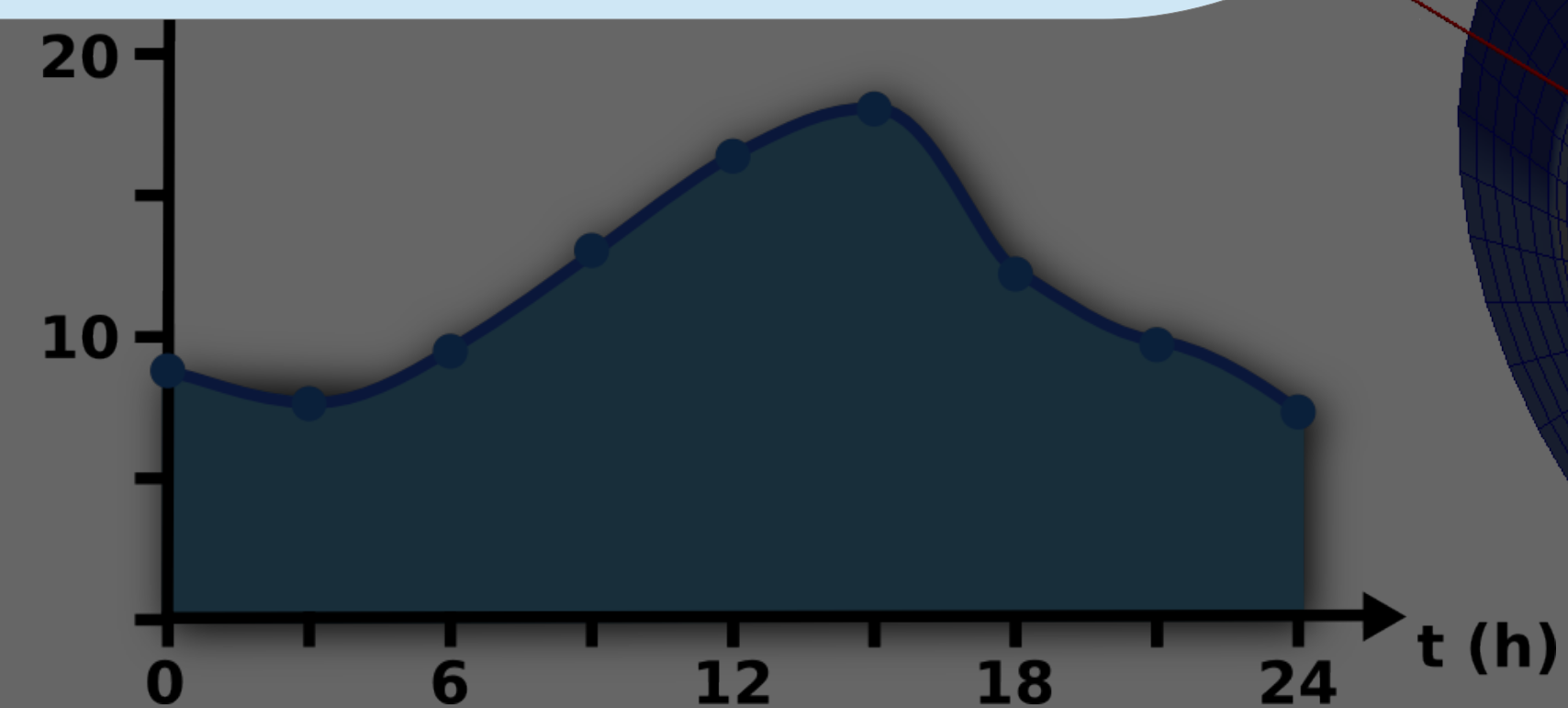


Why is it important?



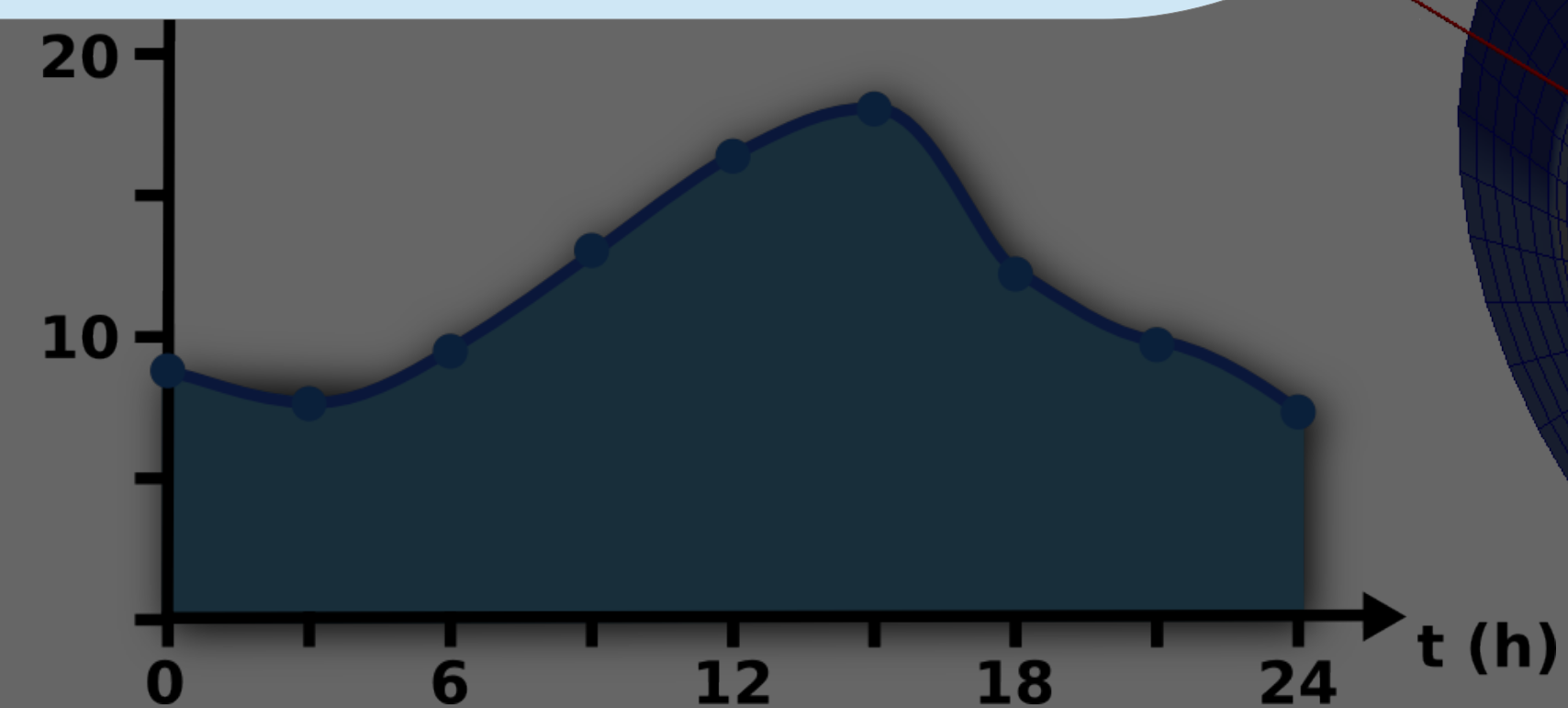
Why is it important?

- Dimension of the domain
- Topology of the domain
- Local topology of the discretization
- ...



Why is it important?

- Dimension of the domain
 - Topology of the domain
 - Local topology of the discretization
 - ...
- **Adapted algorithms**



What do we mean by domain?

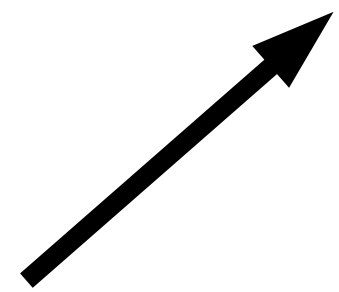
What do we mean by domain?

$$f : \mathcal{D} \rightarrow \mathcal{A}$$

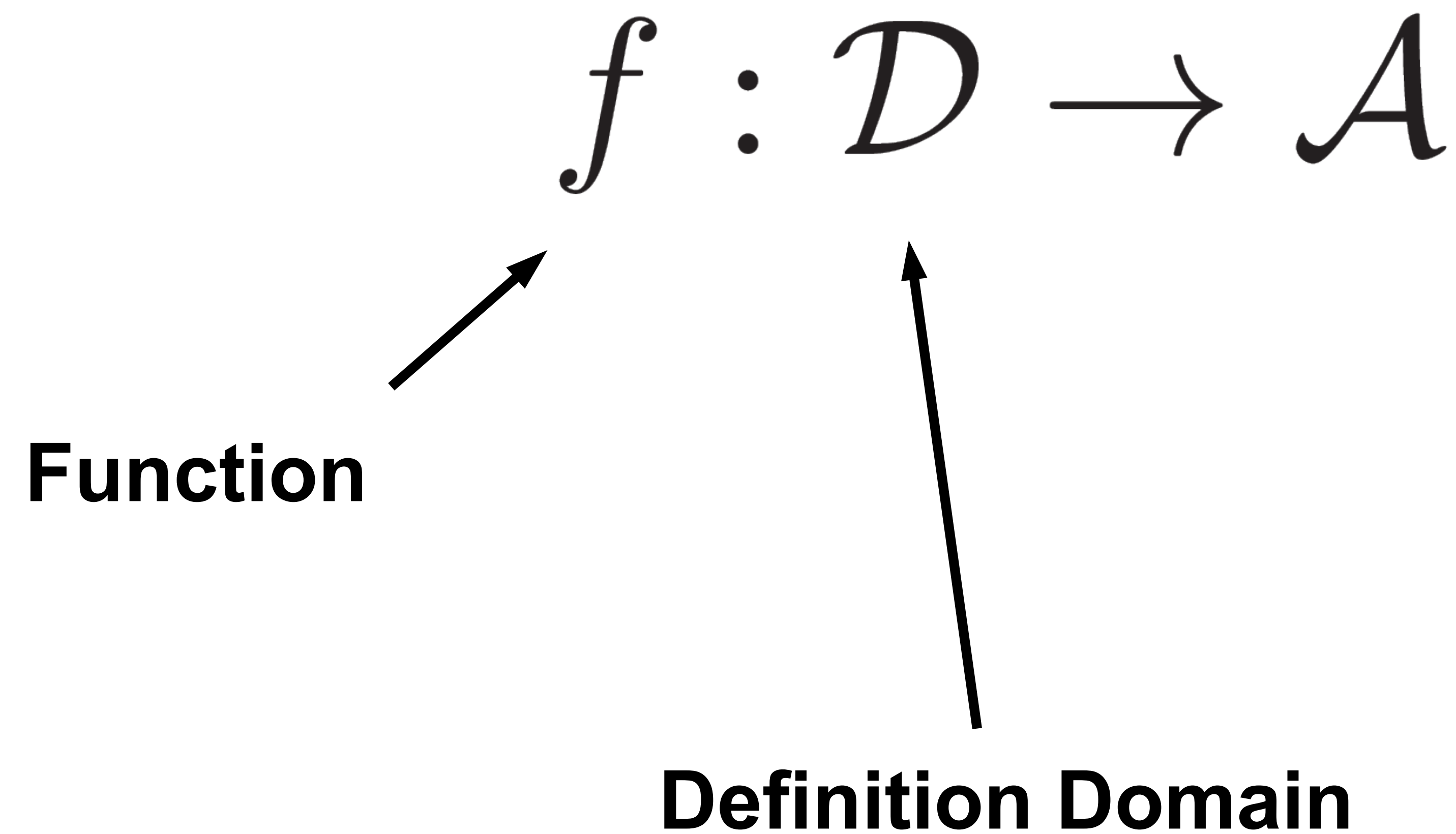
What do we mean by domain?

$$f : \mathcal{D} \rightarrow \mathcal{A}$$

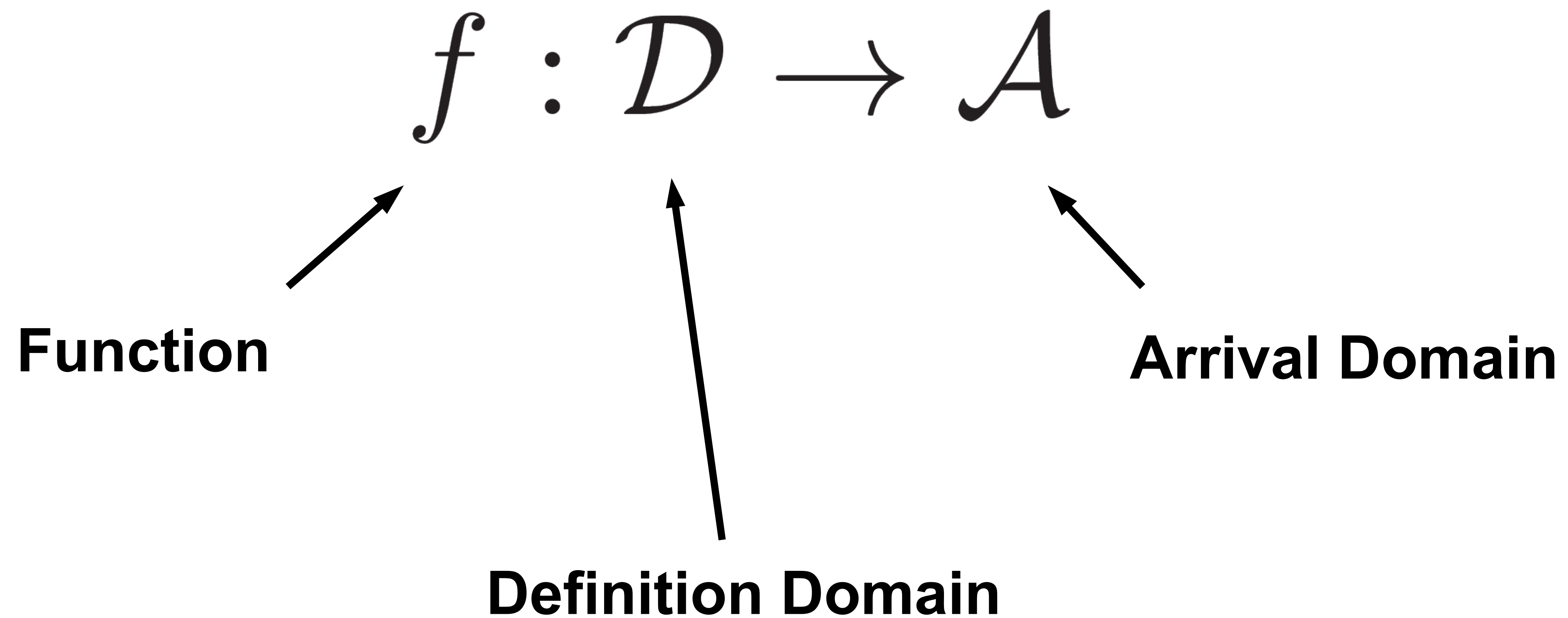
Function



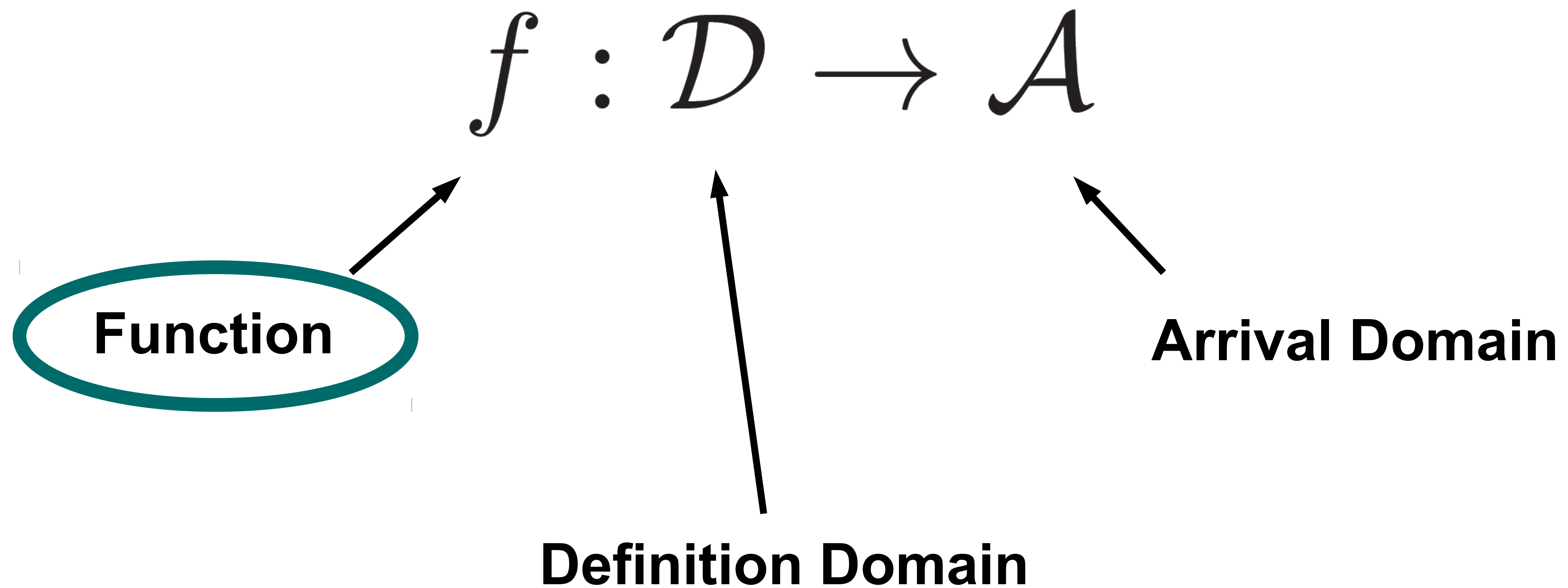
What do we mean by domain?



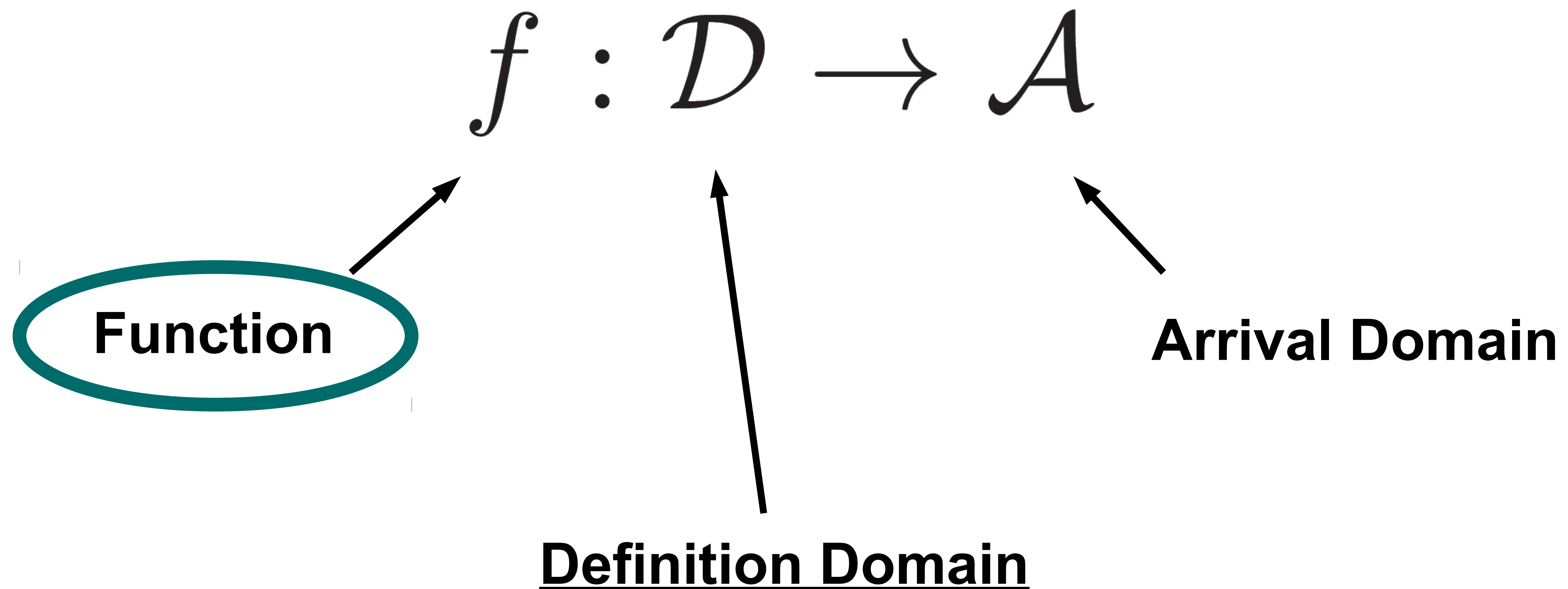
What do we mean by domain?



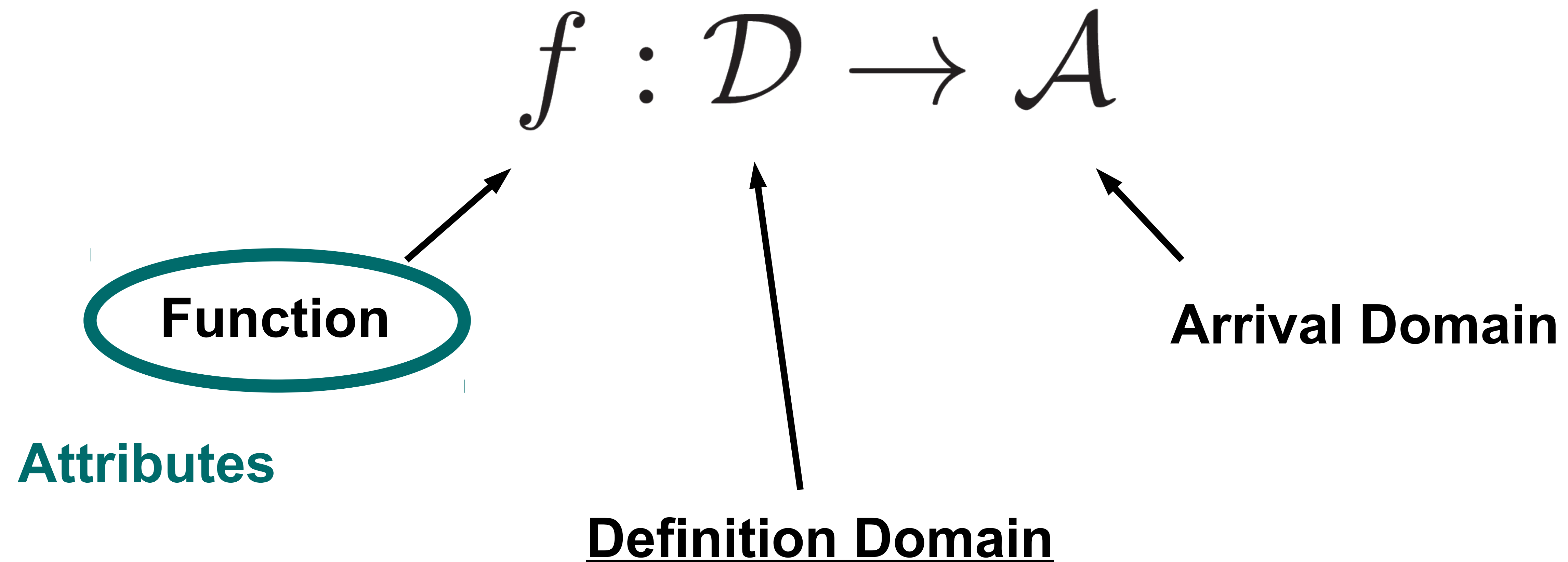
What do we mean by domain?



What do we mean by domain?



What do we mean by domain?



What do we mean by domain?

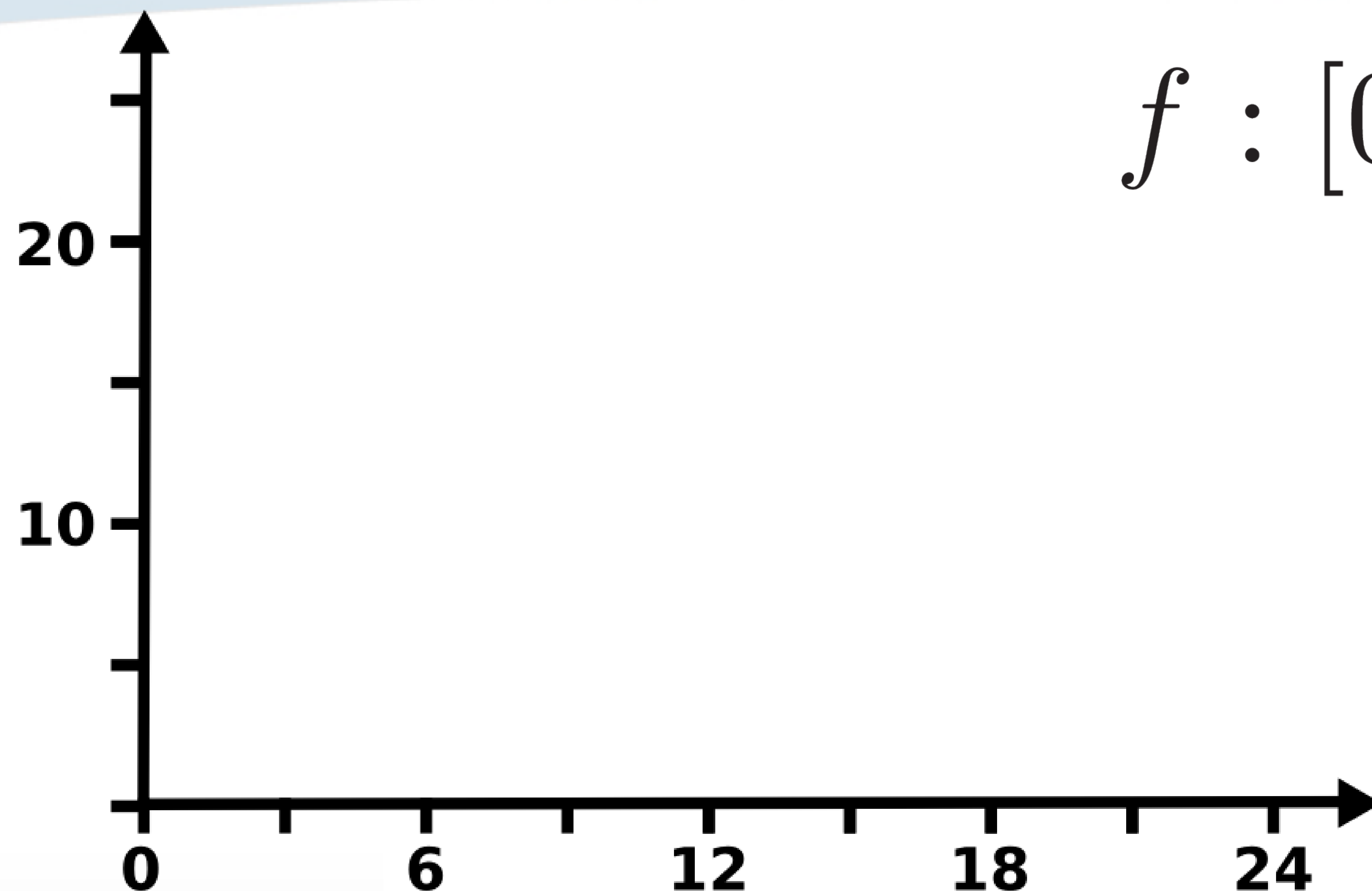
$$f : [0, 24] \rightarrow \mathbb{R}$$

What do we mean by domain?

$$f : [0, 24] \rightarrow \mathbb{R}$$

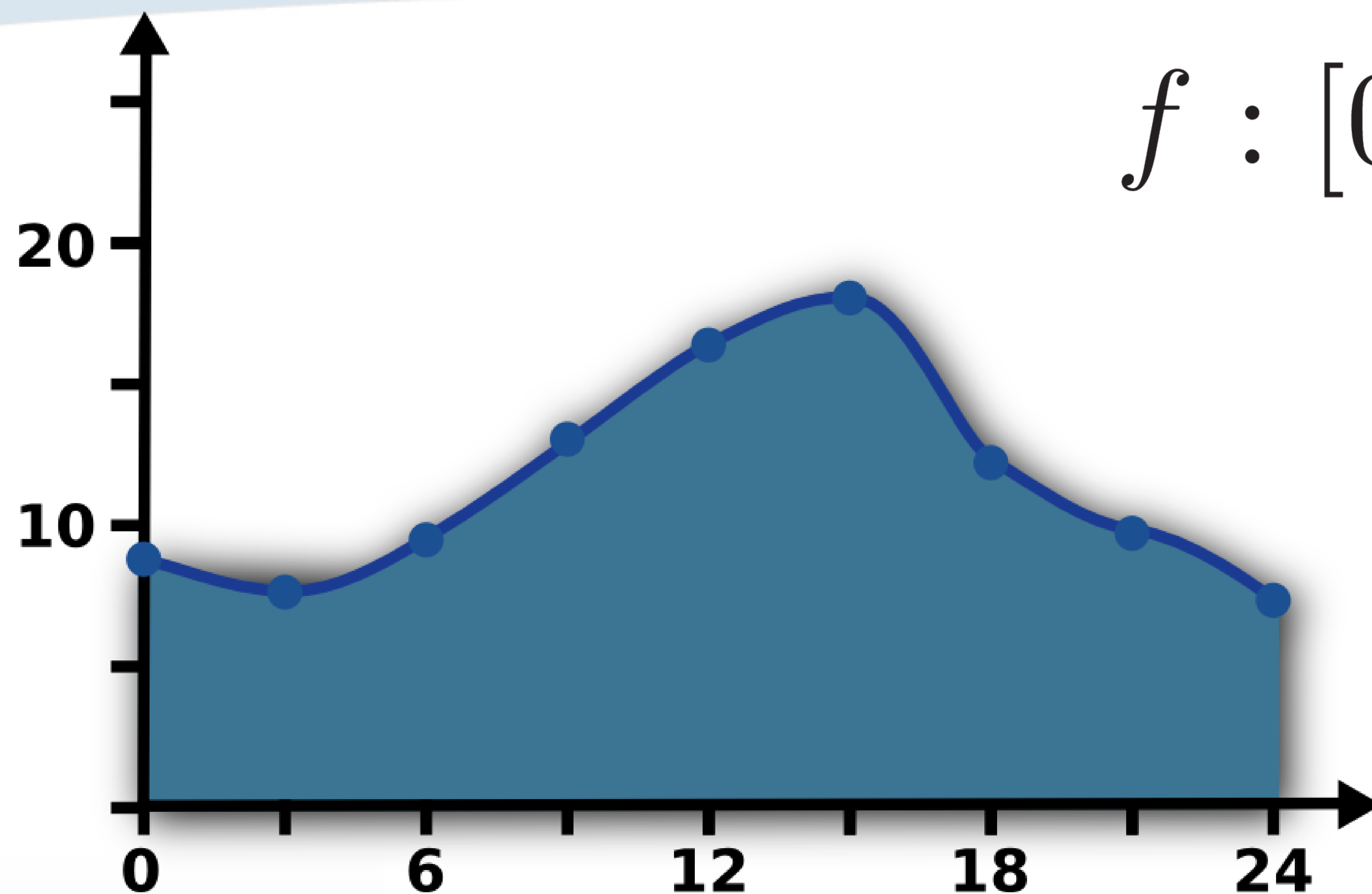


What do we mean by domain?



$$f : [0, 24] \rightarrow \mathbb{R}$$

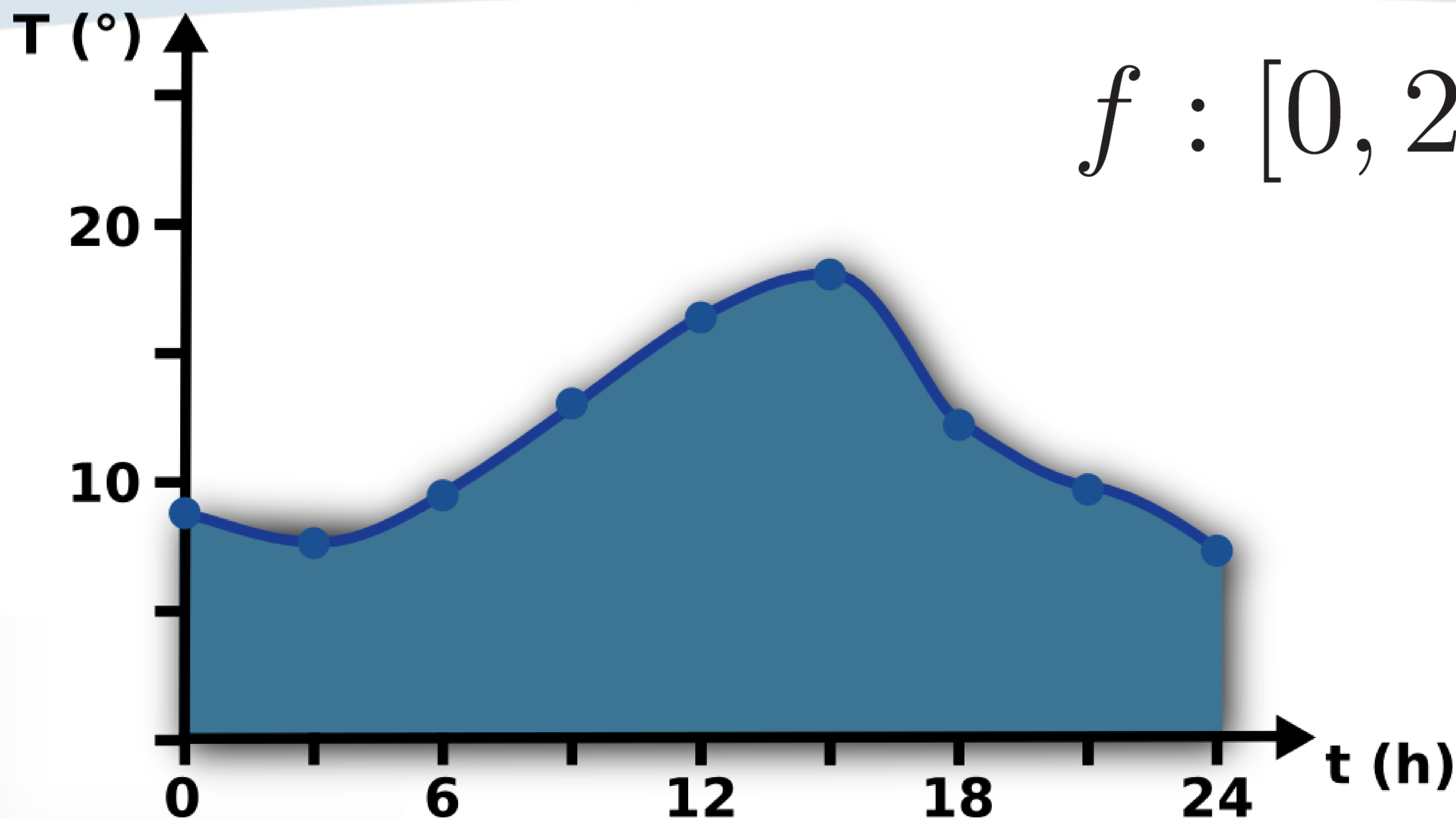
What do we mean by domain?



$$f : [0, 24] \rightarrow \mathbb{R}$$

What do we mean by domain?

$$f : [0, 24] \rightarrow \mathbb{R}$$



What do we mean by domain?

What do we mean by domain?

- Continuous discrete representation of a space \mathbb{X}

What do we mean by domain?

- *Continuous discrete* representation of a space \mathbb{X}

What do we mean by domain?

- *Continuous discrete* representation of a space X
 - Finite set of samples

What do we mean by domain?

- *Continuous discrete* representation of a space \mathbb{X}
 - Finite set of samples
 - Positional information in an embedding space \mathbb{E}

What do we mean by domain?

- *Continuous discrete* representation of a space \mathbb{X}
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 - Attributes

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What do we mean by domain?

- *Continuous discrete* representation of a space \mathbb{X}
 - Finite set of samples
 - Positional information in an embedding space \mathbb{E}
 - Attributes
- Connectivity of the samples

What do we mean by domain?

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 - Attributes
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What do we mean by domain?

- *Continuous discrete* representation of a space X
 - Finite set of samples
 - Positional information in an embedding space \mathbb{E}
 - Attributes
- Connectivity of the samples
 - Continuity of the domain
 - Cellular elements (dimension of X)

What do we mean by domain?

- *Continuous discrete* representation of a space \mathbb{X}
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 - Attributes
- Connectivity of the samples
 - Continuity of the domain
 - Cellular elements (dimension of \mathbb{X})
- Cell interpolation scheme

What do we mean by domain?

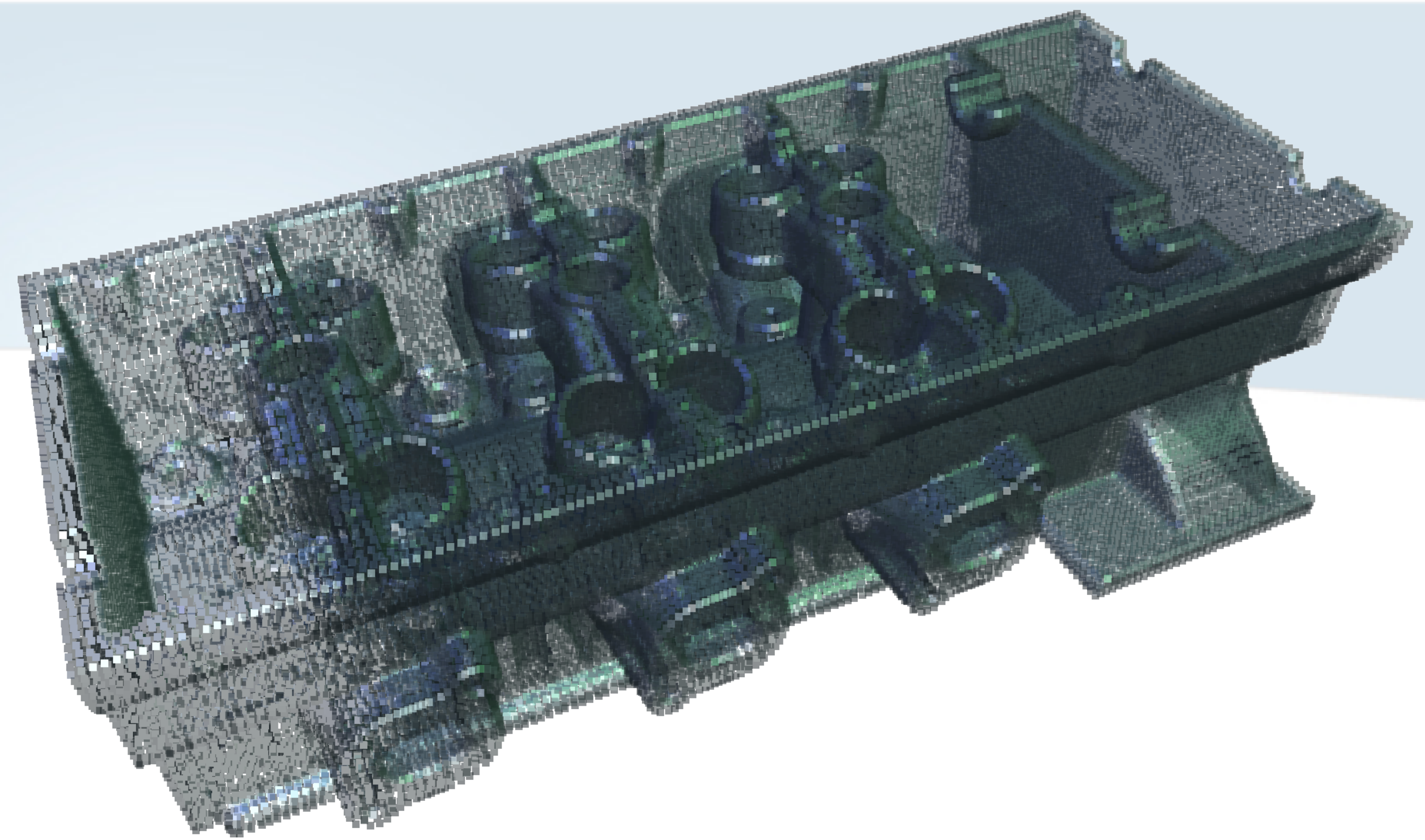
- *Continuous discrete* representation of a space \mathbb{X}
 - Finite set of samples
 - Positional information in an embedding space \mathbb{E}
 - Attributes
- Connectivity of the samples
 - Continuity of the domain
 - Cellular elements (dimension of \mathbb{X})
- Cell interpolation scheme
 - Continuity of the attributes

What do we mean by domain?

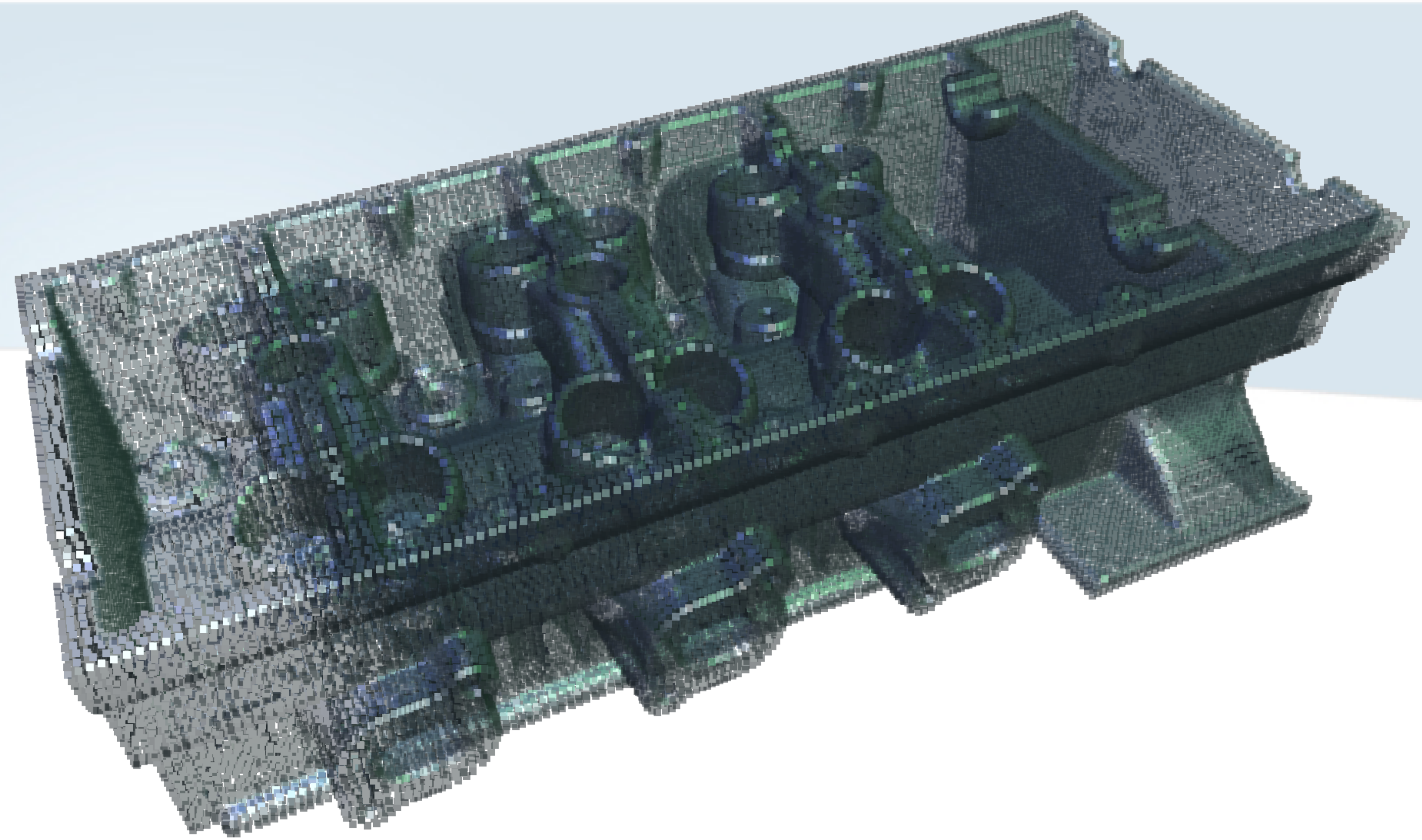
- *Continuous discrete* representation of a space \mathbb{X}
 - **Finite set of samples**
 - Positional information in an embedding space \mathbb{E}
 - Attributes
 - **Connectivity of the samples**
 - Continuity of the domain
 - Cellular elements (dimension of \mathbb{X})
 - **Cell interpolation scheme**
 - Continuity of the attributes

Example

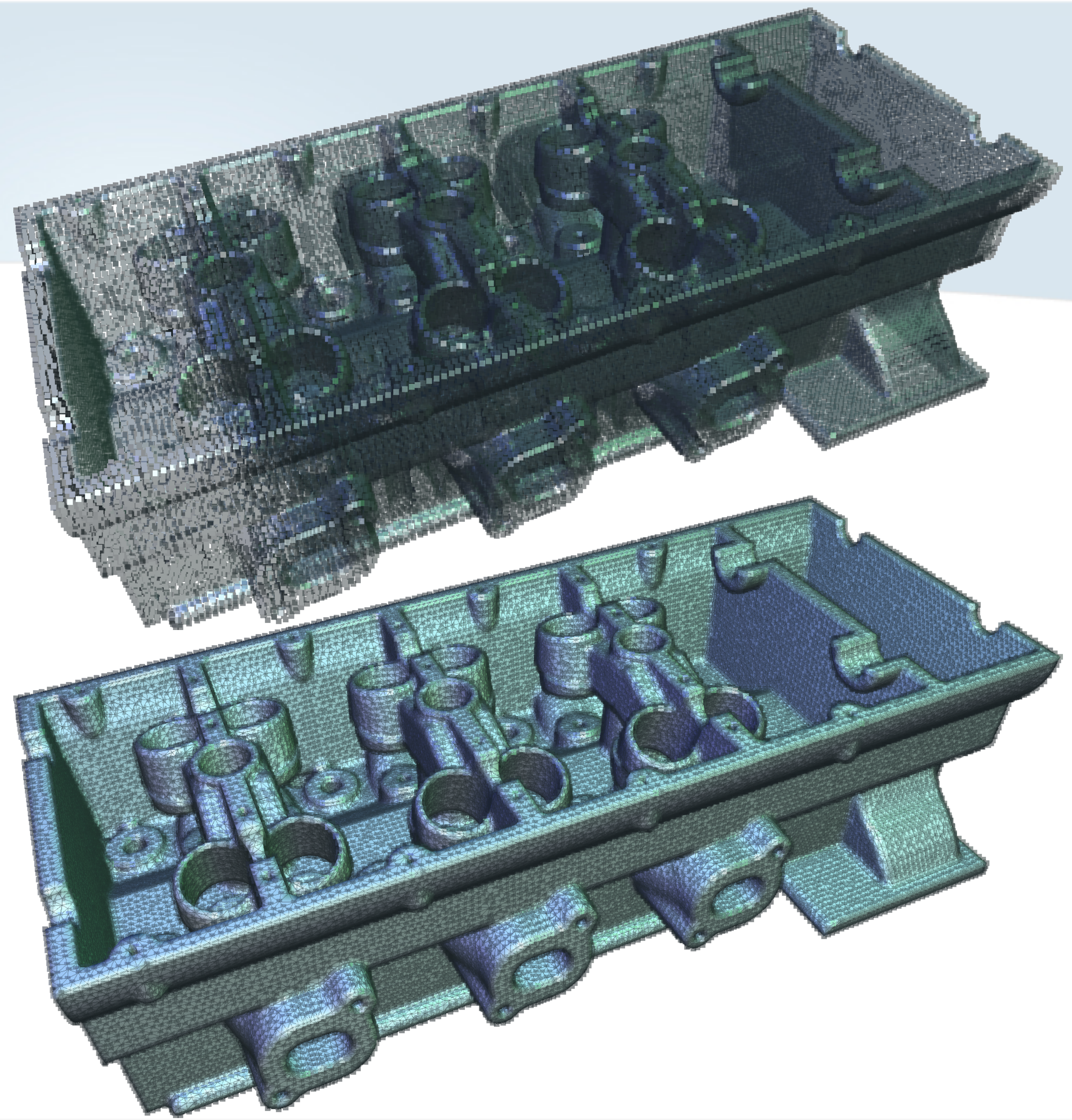
Example



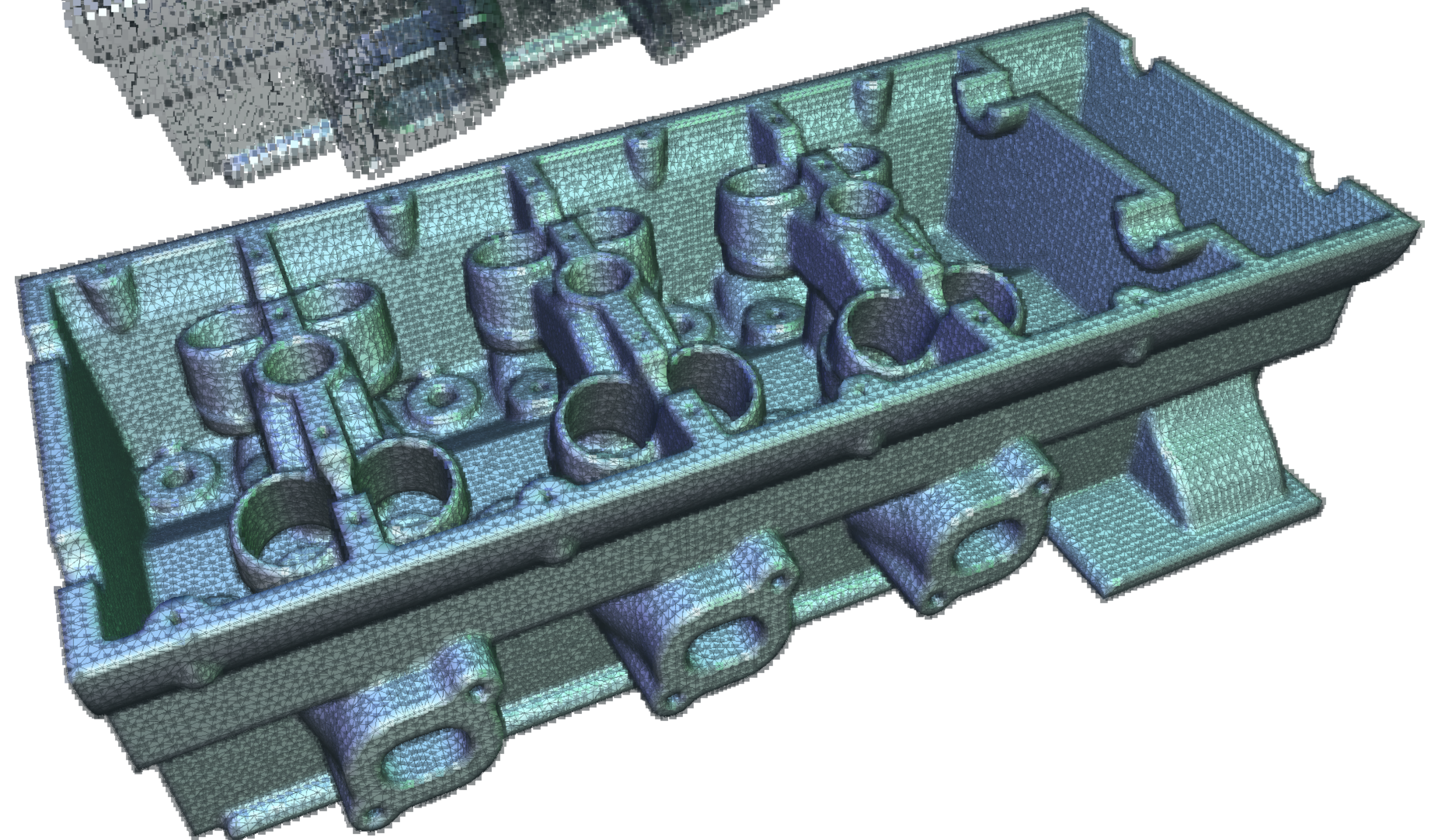
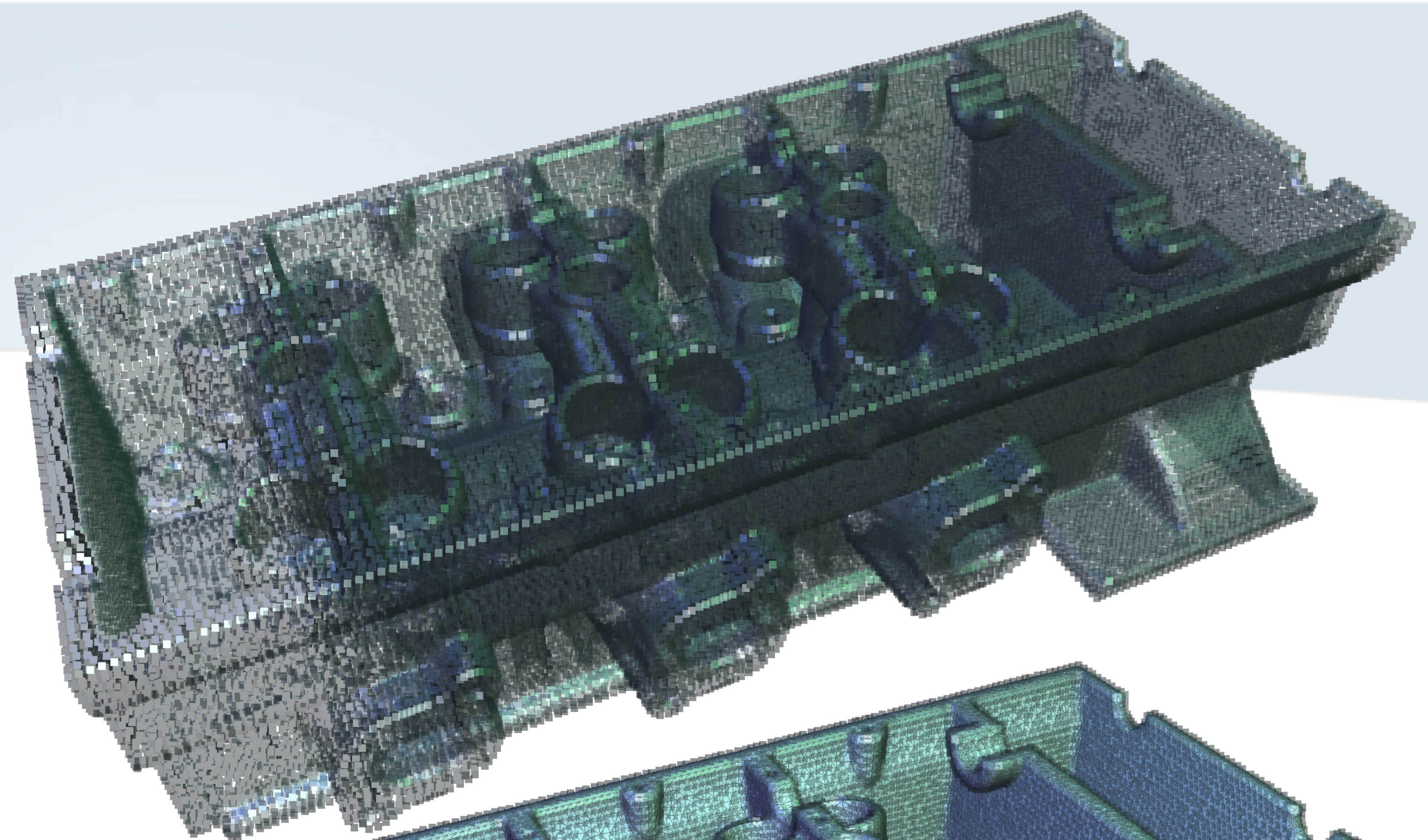
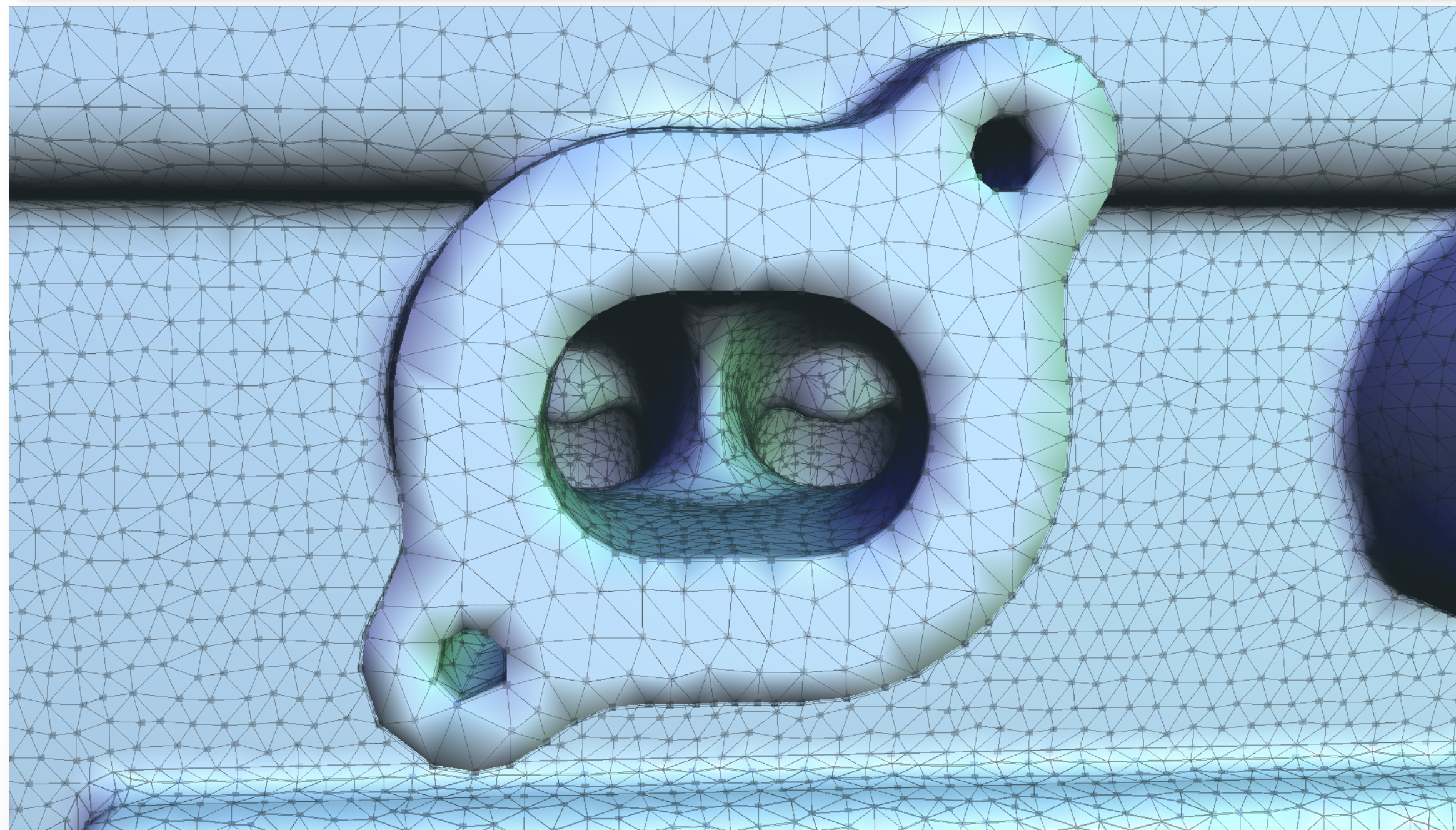
Example



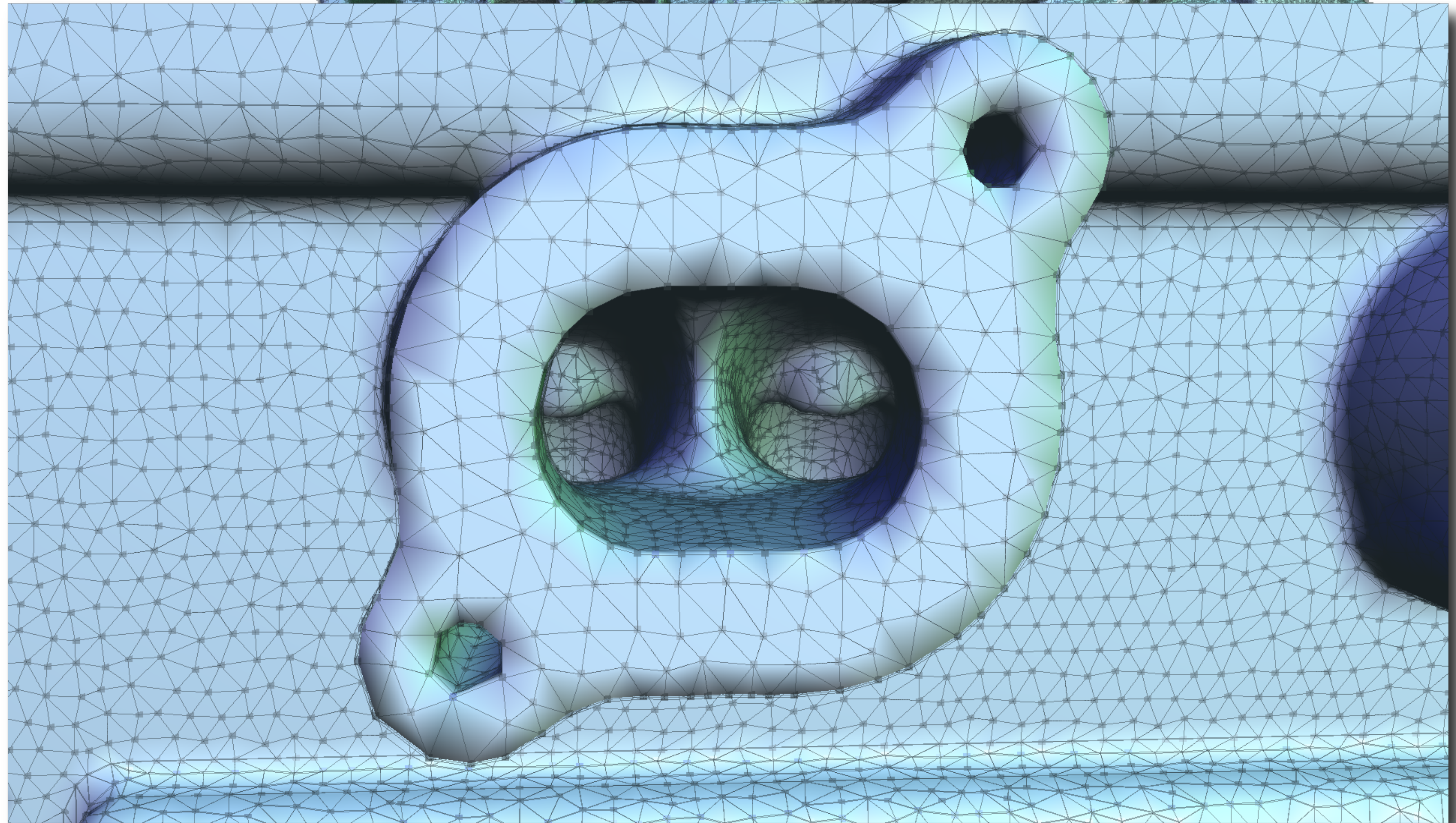
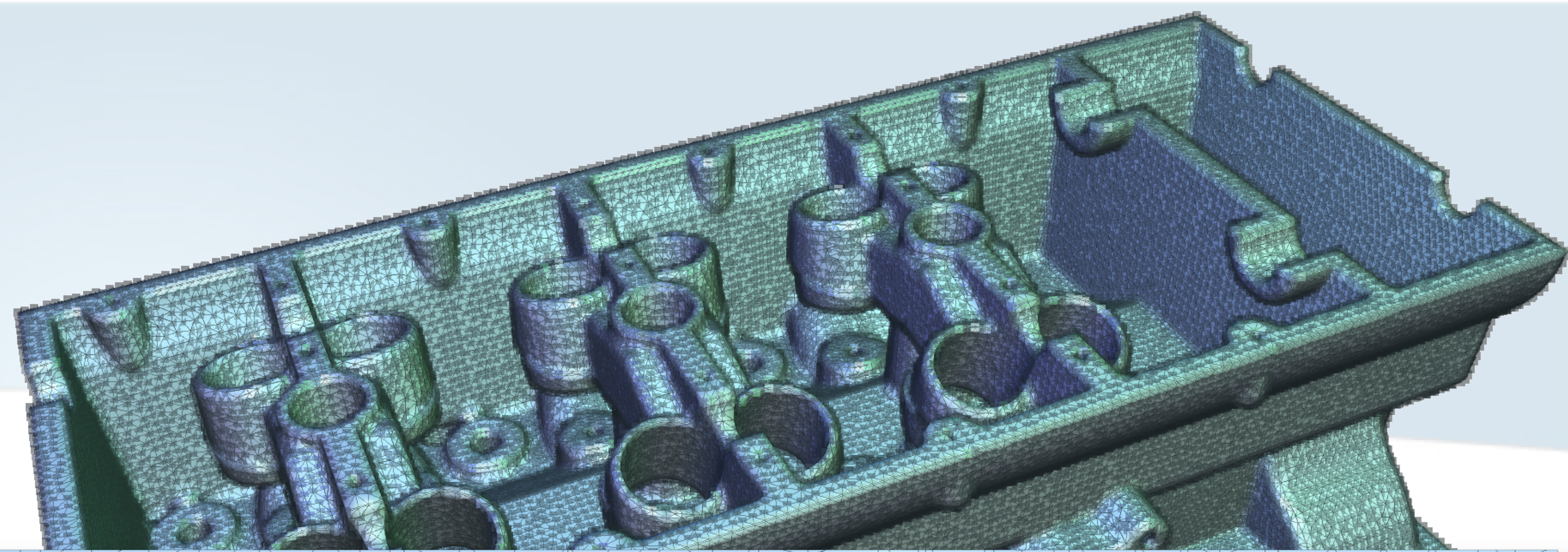
Example



Example

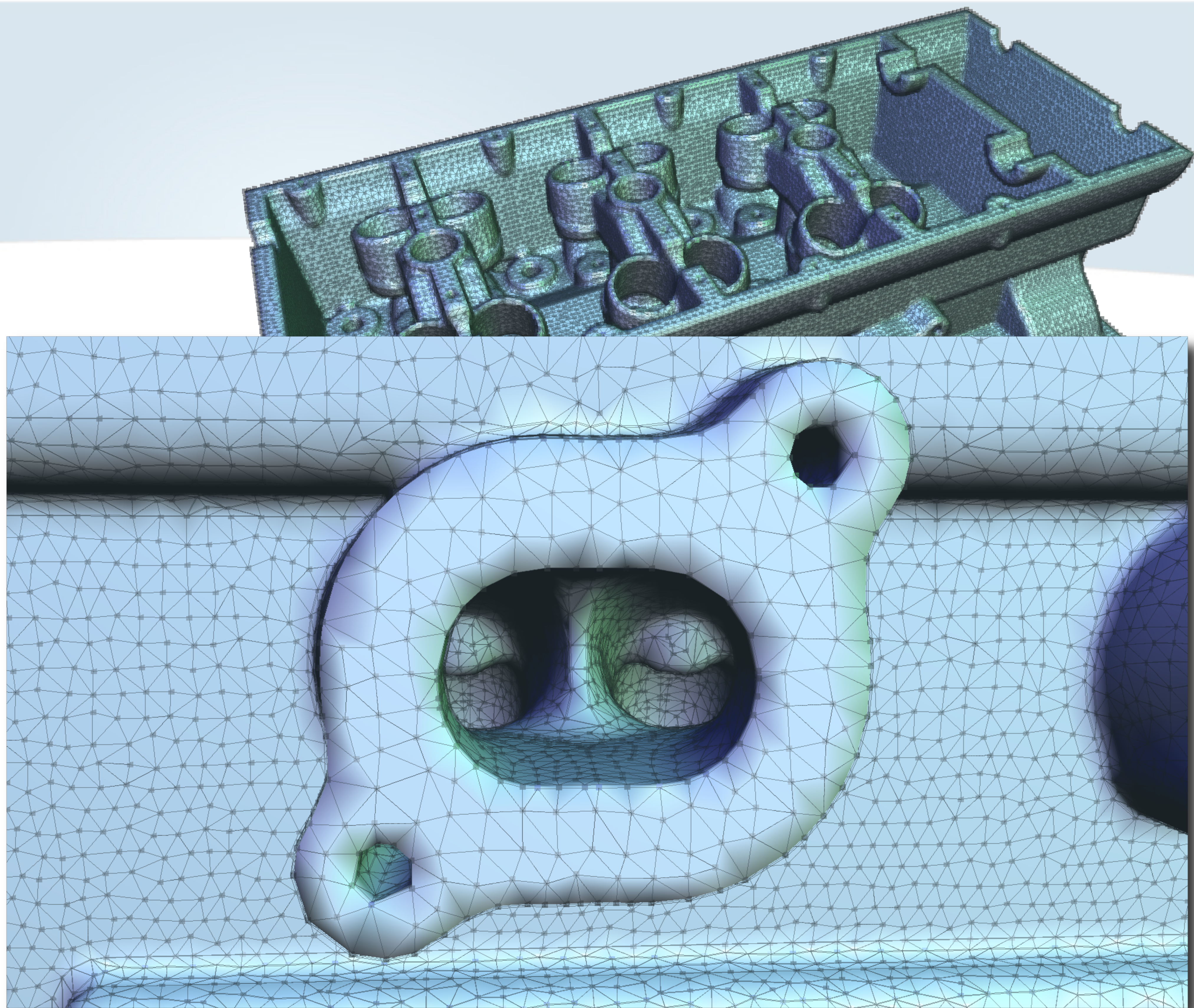


Example



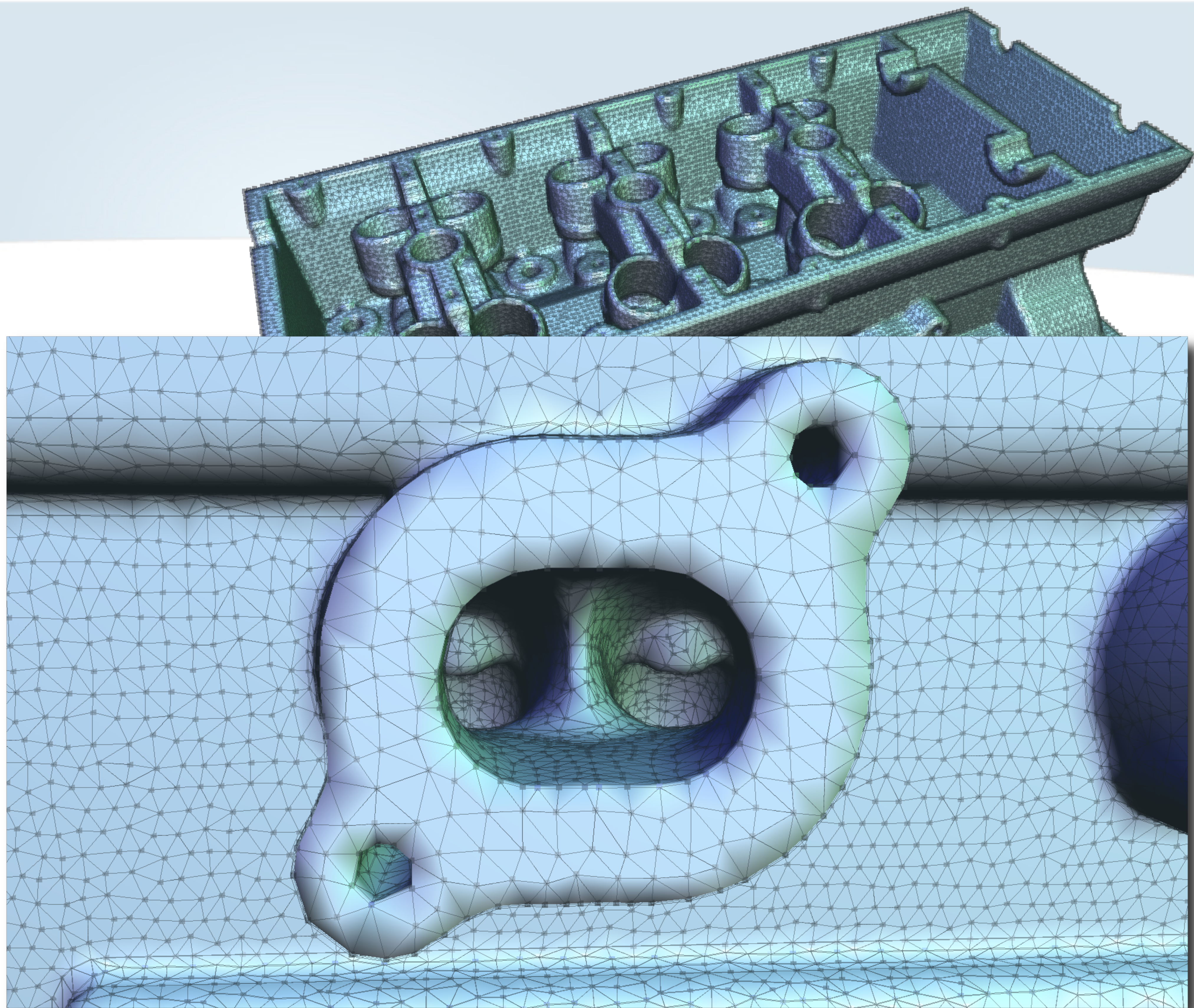
Example

- Domain
- Embedding space
- Cellular elements



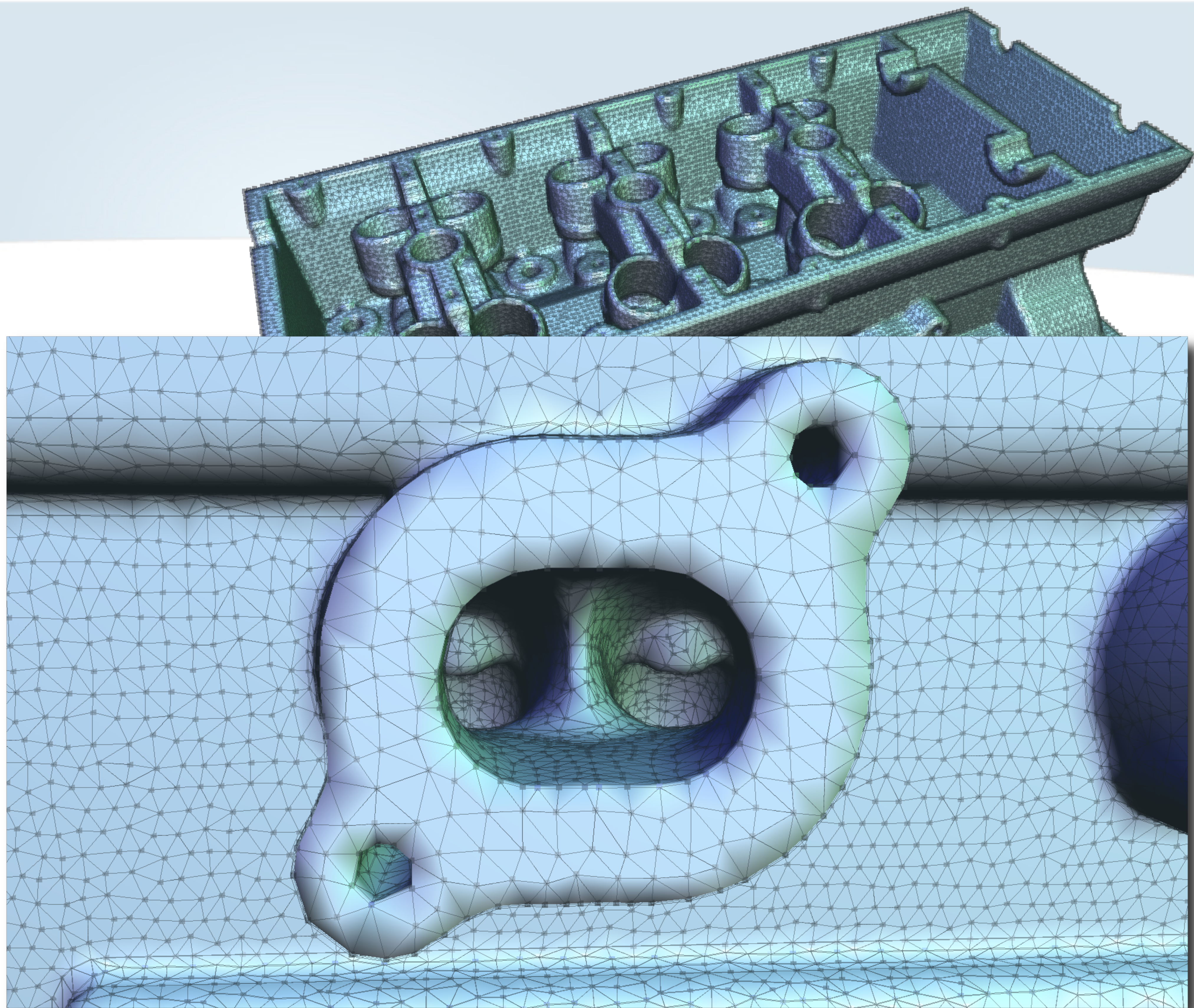
Example

- Domain
 - 2-manifold \mathcal{S}
- Embedding space
- Cellular elements



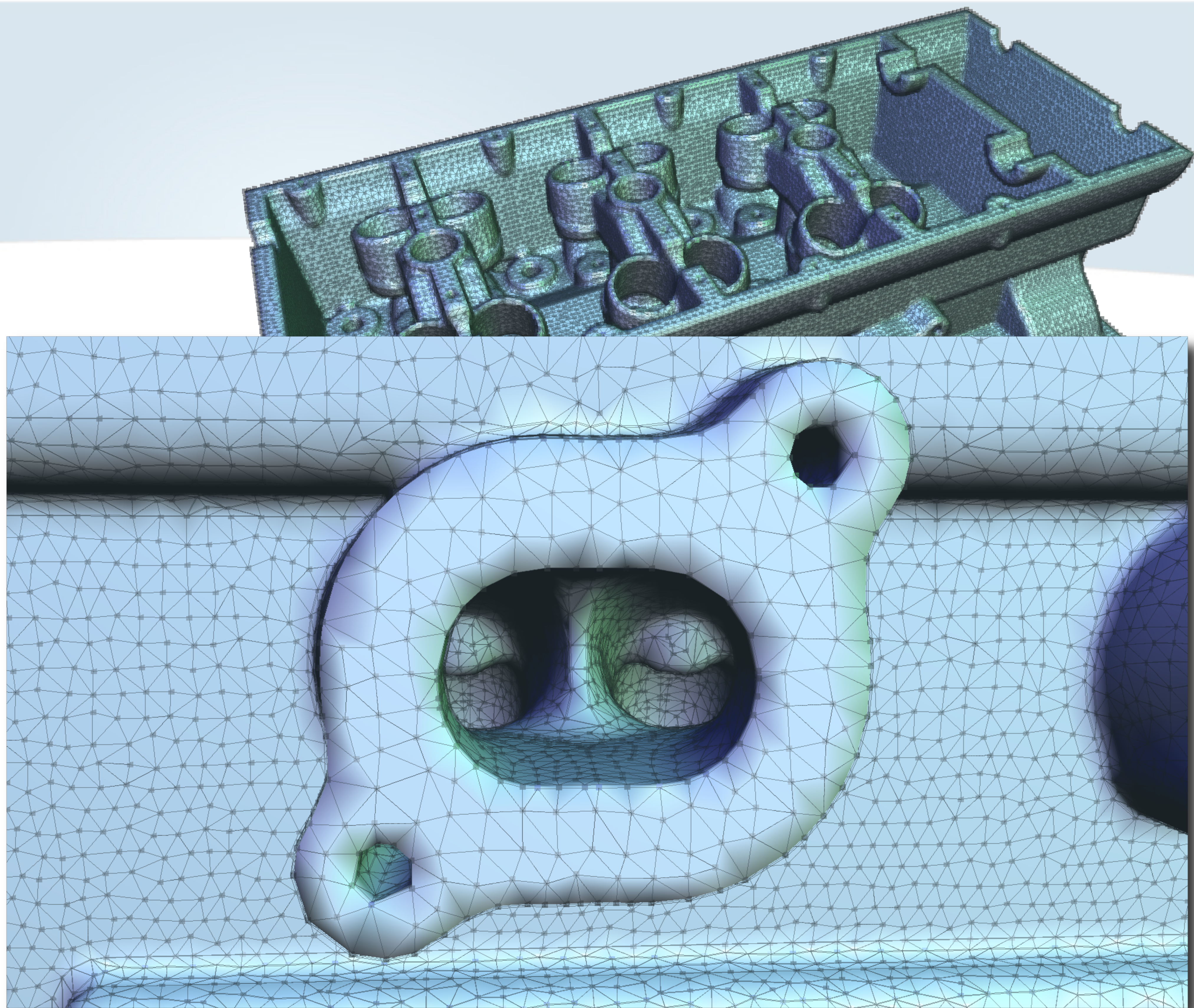
Example

- Domain
 - 2-manifold \mathcal{S}
- Embedding space
 - \mathbb{R}^3
- Cellular elements

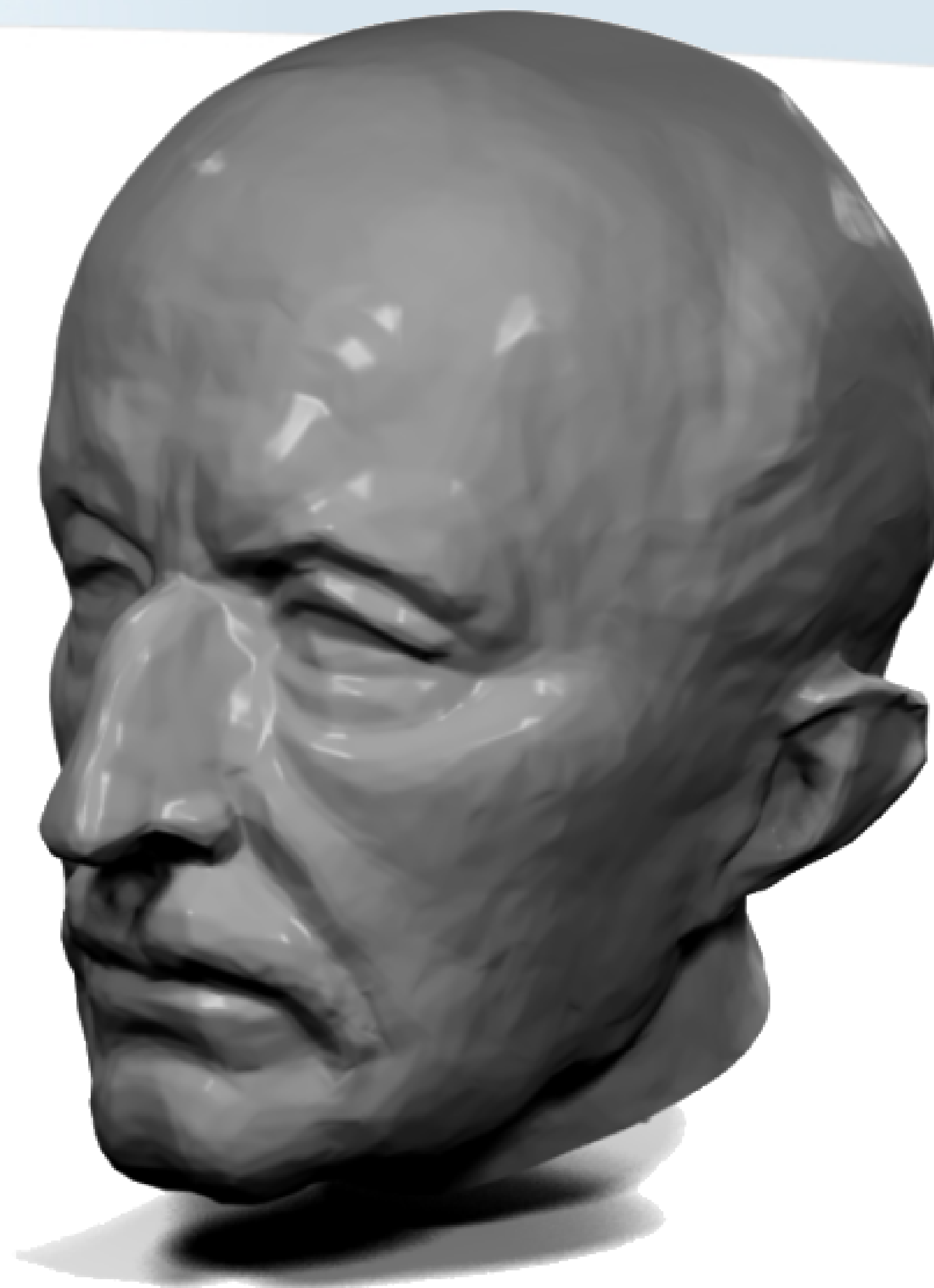


Example

- Domain
 - 2-manifold \mathcal{S}
- Embedding space
 - \mathbb{R}^3
- Cellular elements
 - Triangles



Domains of interest: Manifolds



Domains of interest: Manifolds

- d-manifold



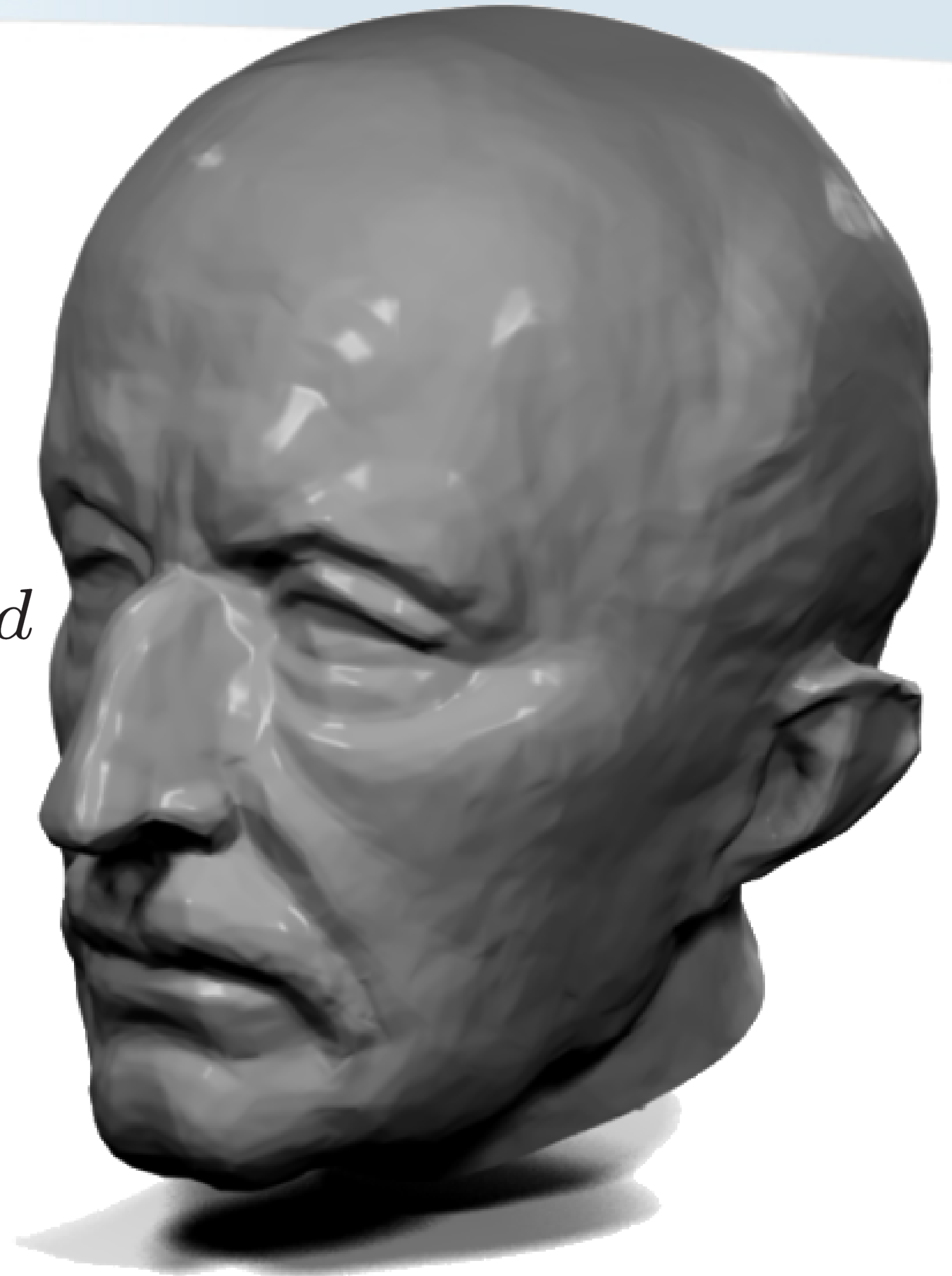
Domains of interest: Manifolds

- d-manifold
 - Topological space such that:



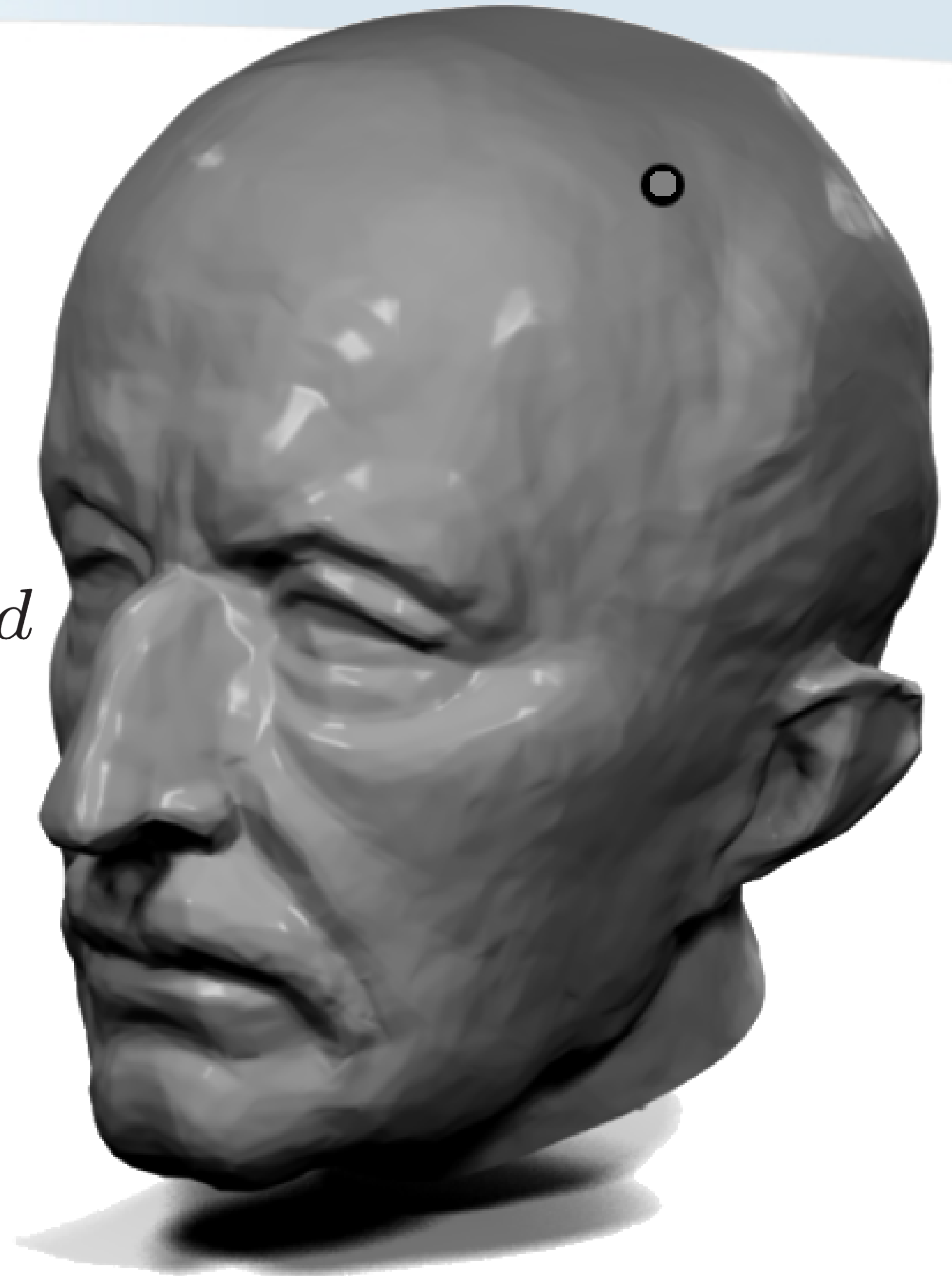
Domains of interest: Manifolds

- d-manifold
 - Topological space such that:
 - Any point has a neighborhood homeomorphic to a neighborhood of \mathbb{R}^d



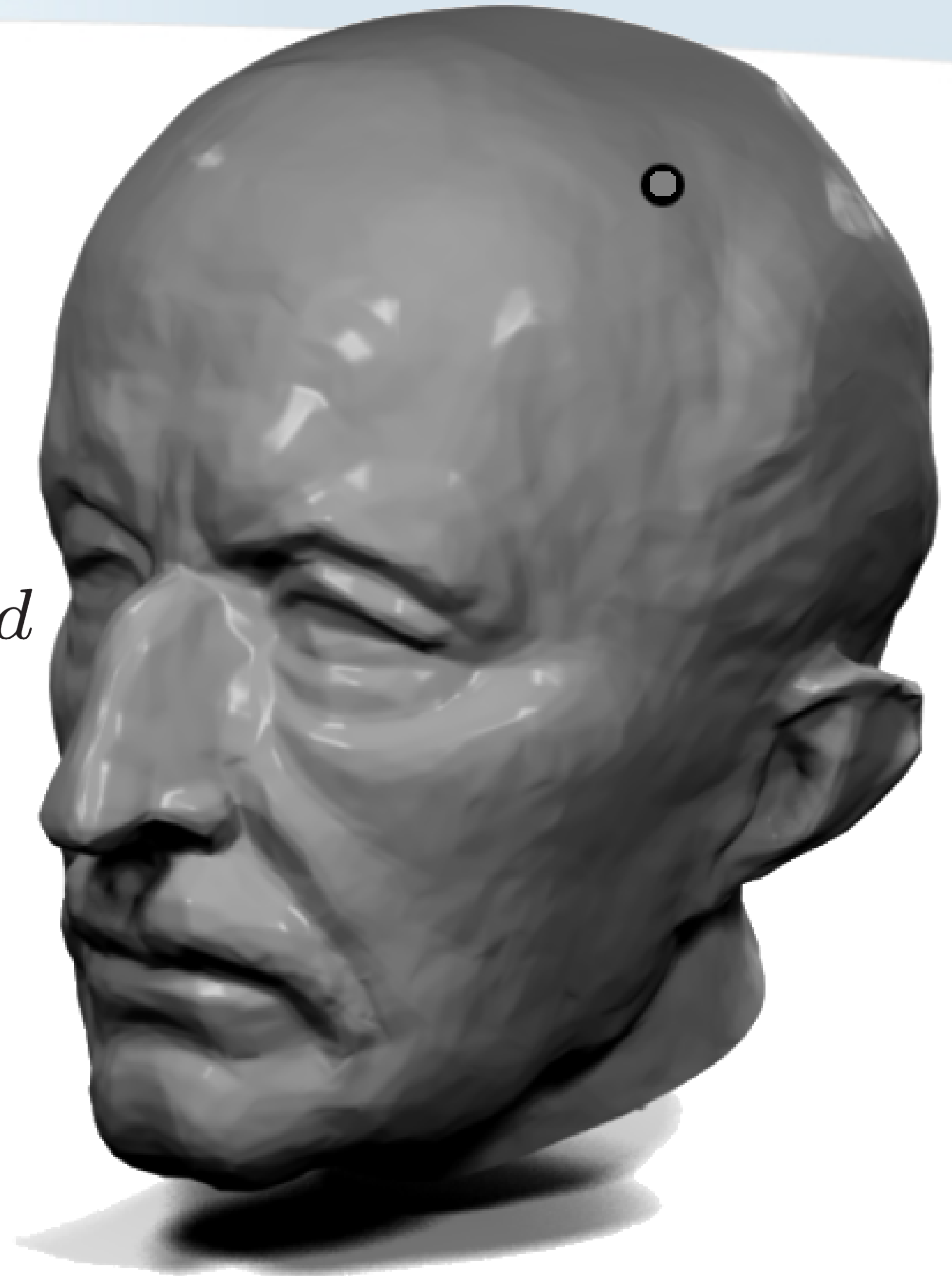
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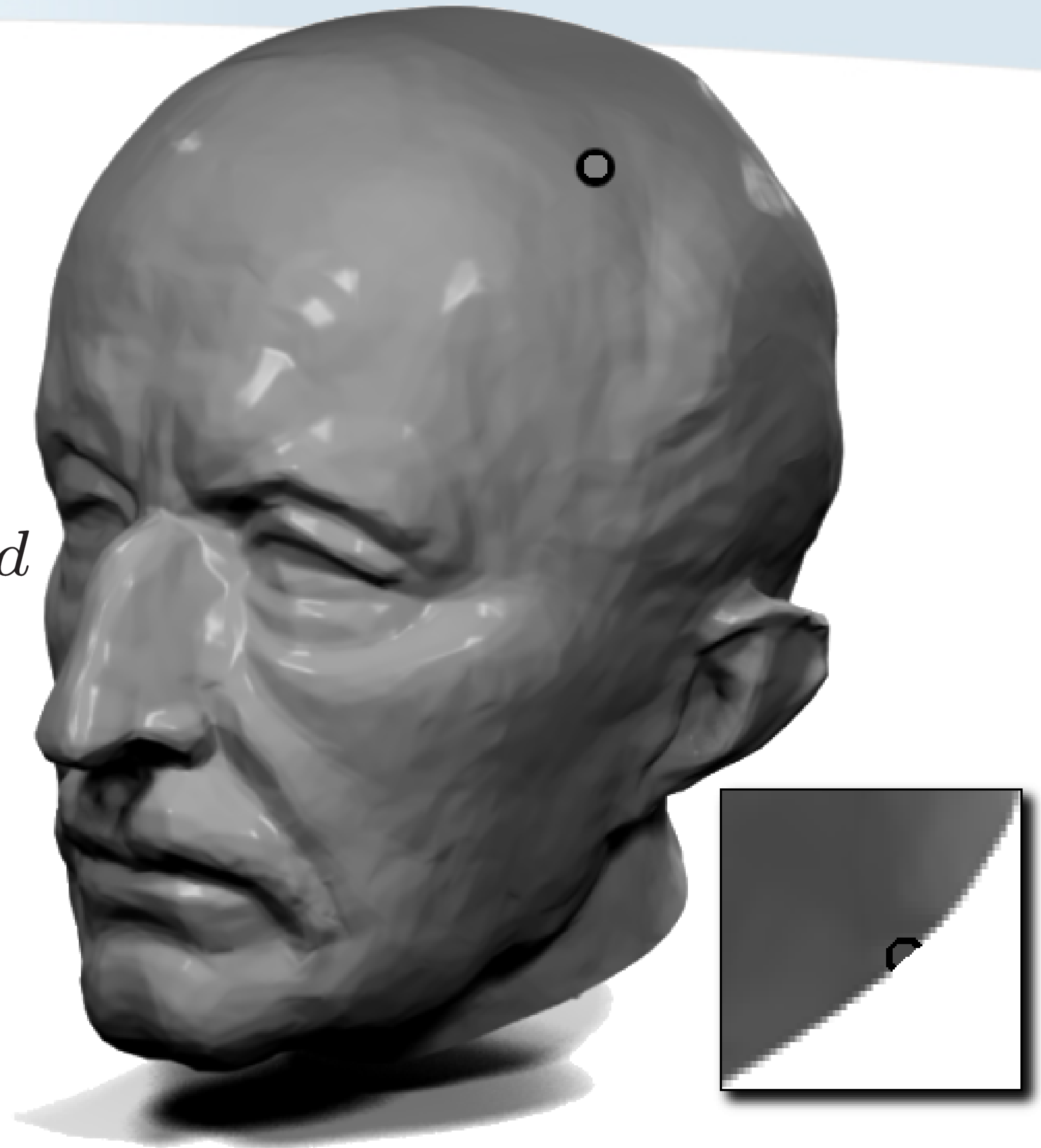
Domains of interest: Manifolds

- d-manifold with boundary
 - Topological space such that:
 - Any point has a neighborhood homeomorphic to a neighborhood of \mathbb{R}^d or its half space



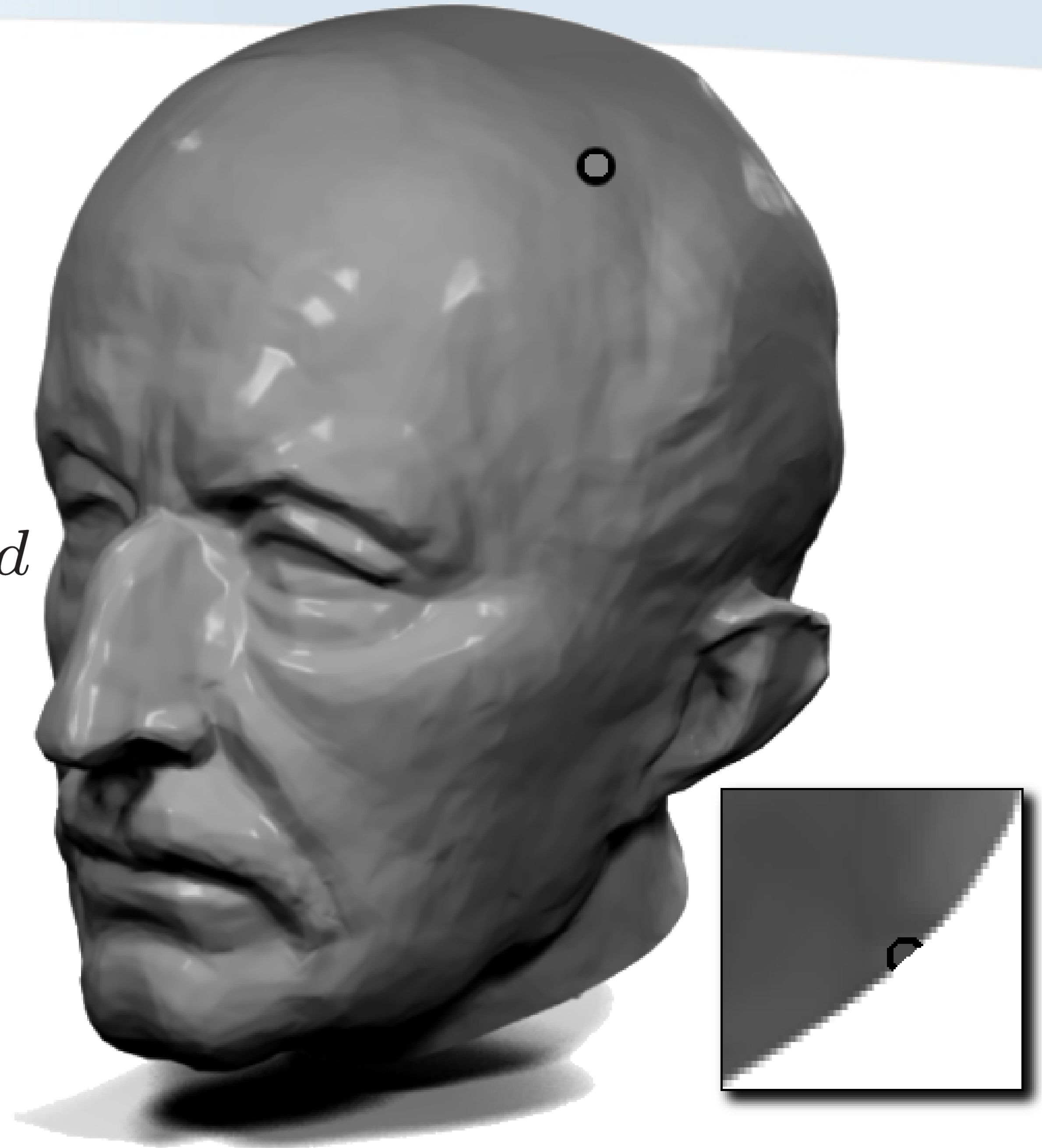
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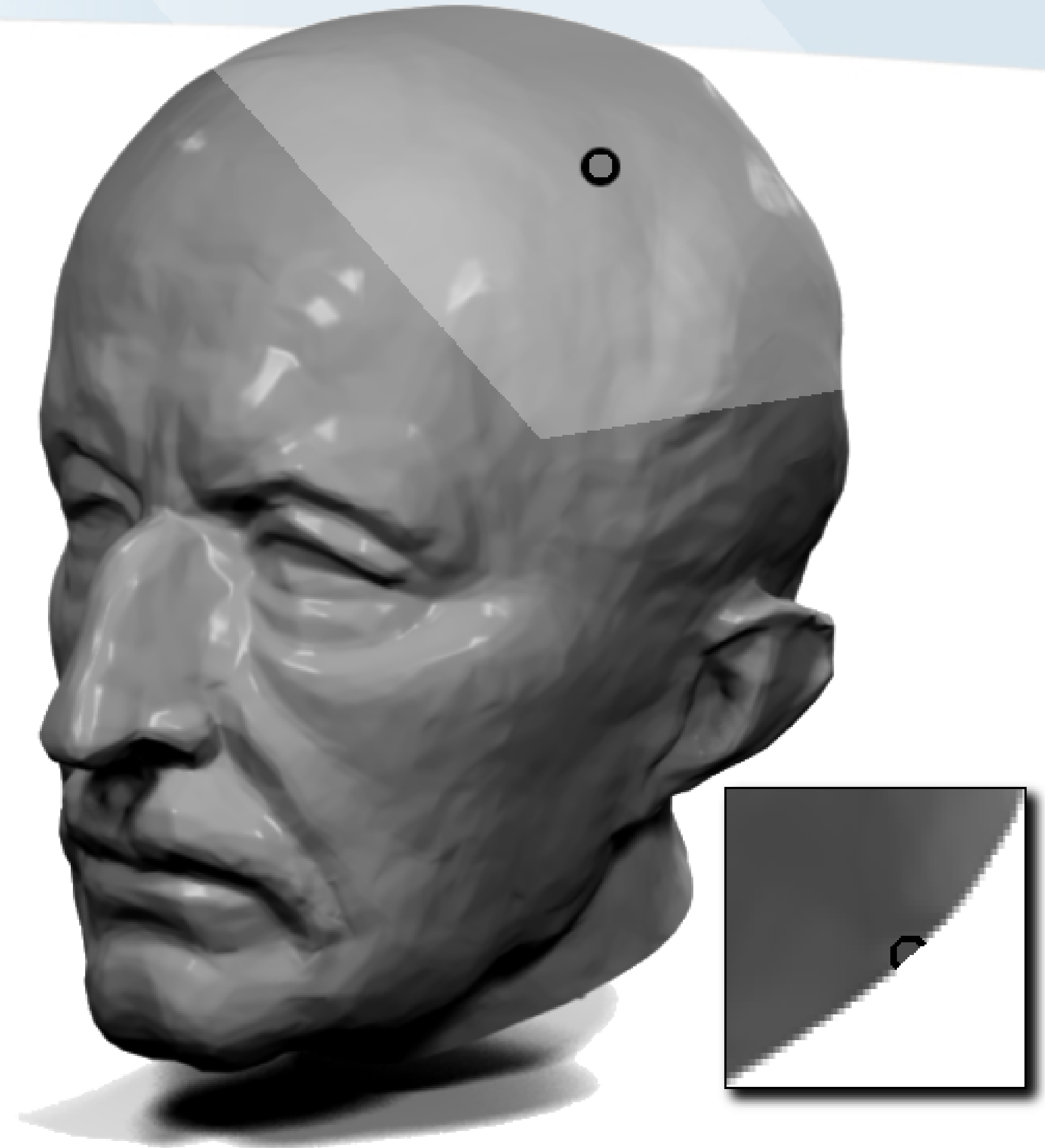
Domains of interest: Manifolds

- d-manifold with boundary
 - Topological space such that:
 - Any point has a neighborhood homeomorphic to a neighborhood of \mathbb{R}^d or its half space
- Boundary: closed (d-1)-manifold



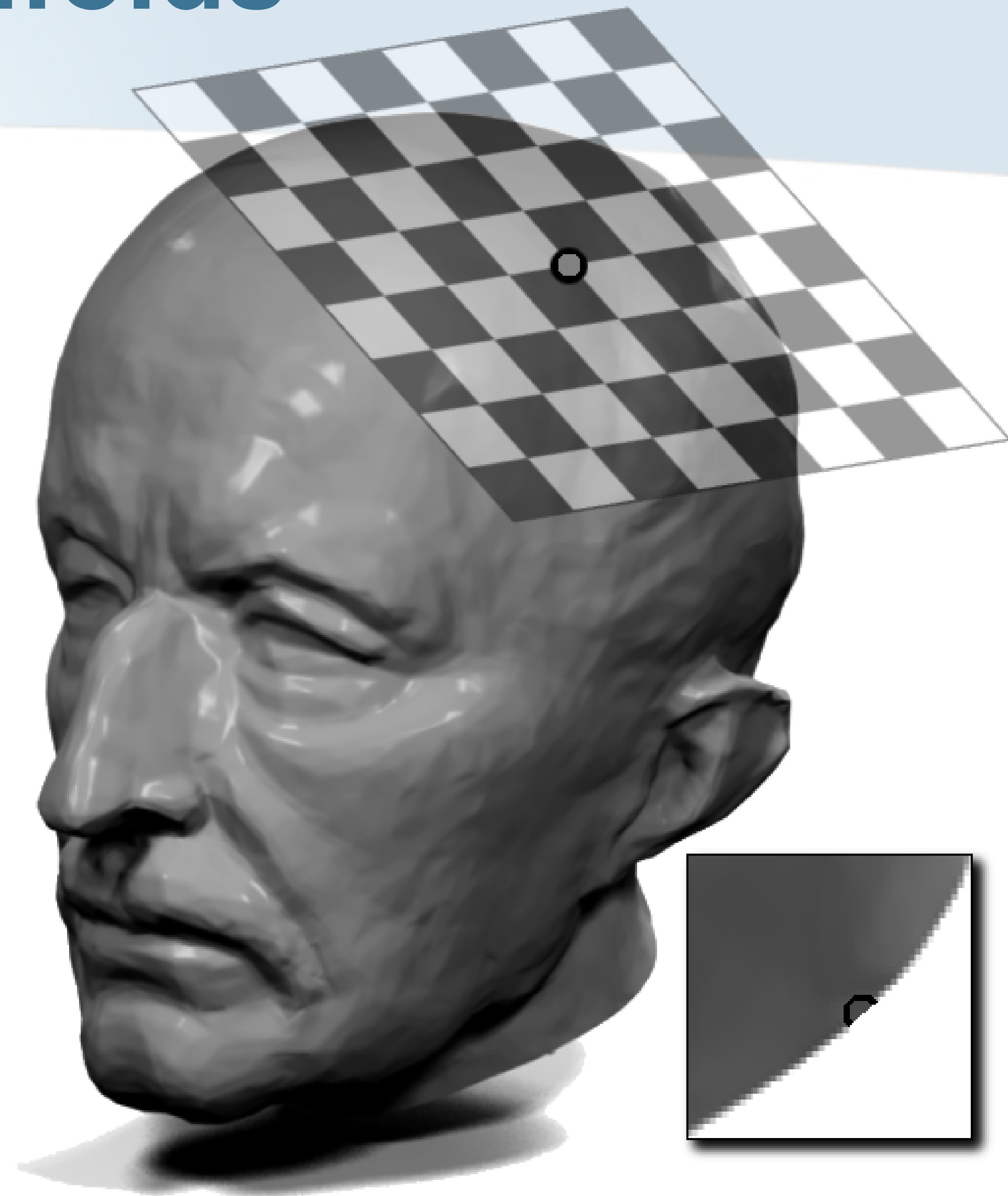
Domains of interest: Manifolds

- Tangent space
 - Localized Euclidean approximation



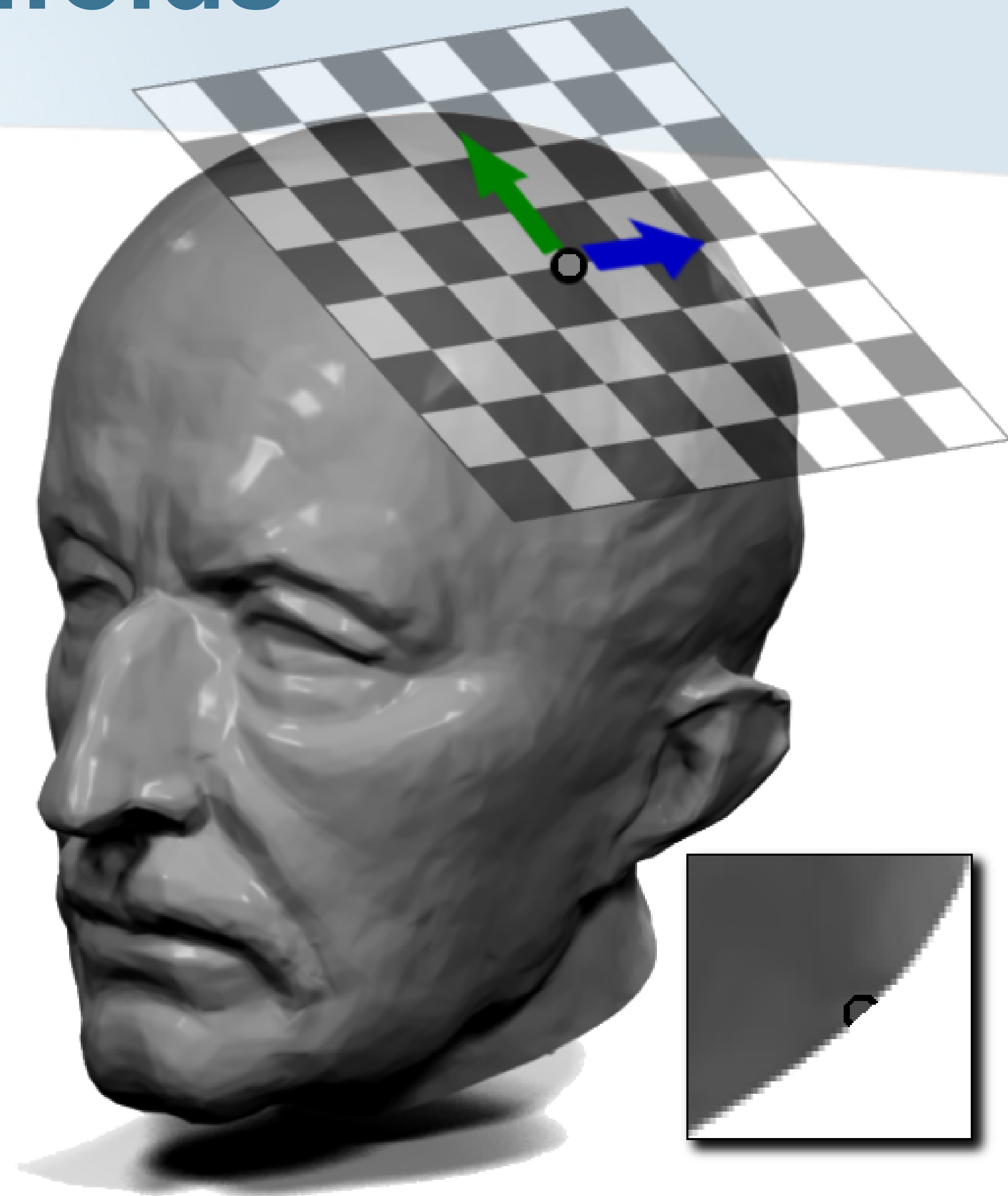
Domains of interest: Manifolds

- Tangent space
 - Localized Euclidean approximation
 - Localized Euclidean geometry



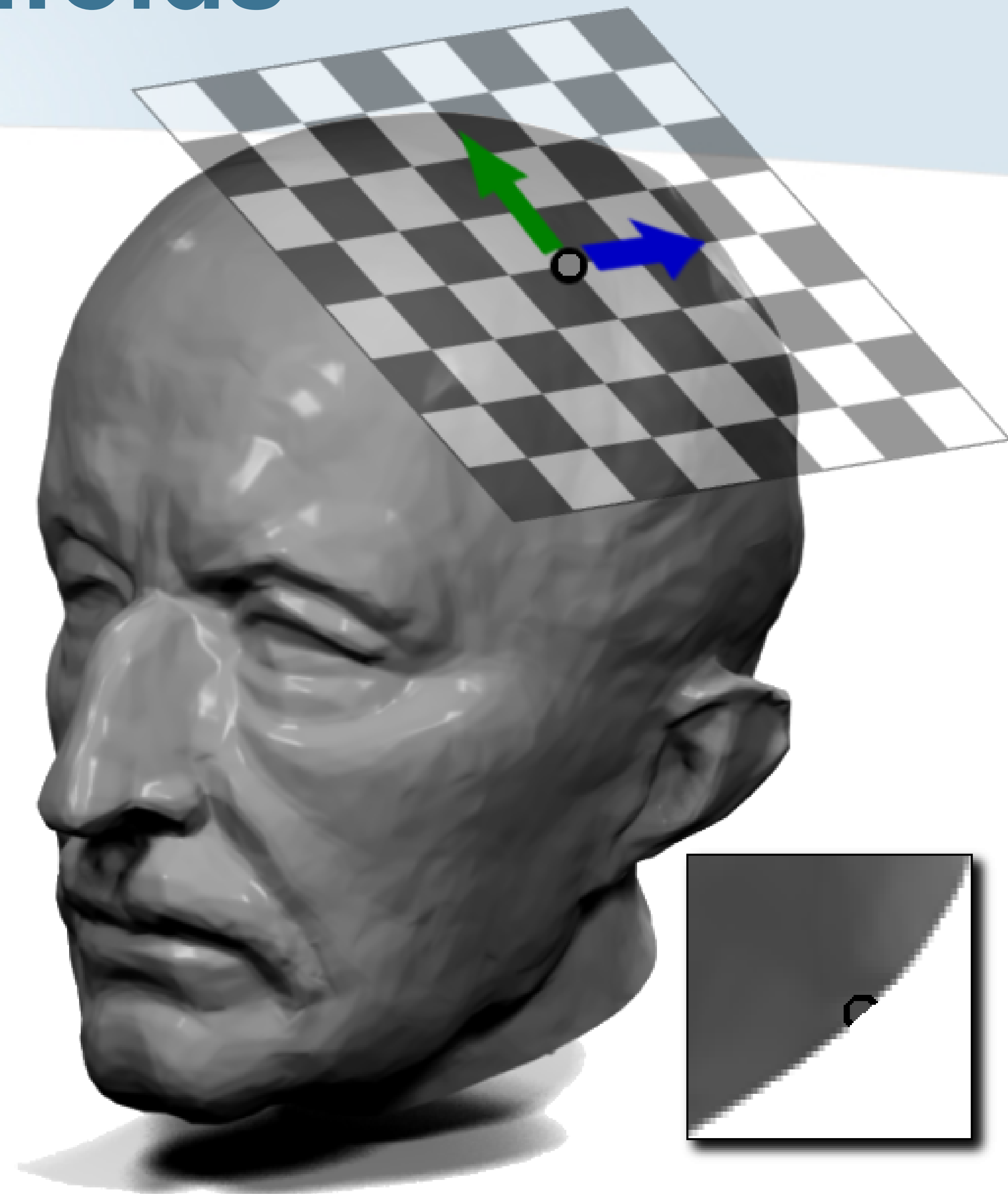
Domains of interest: Manifolds

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 - Inner products



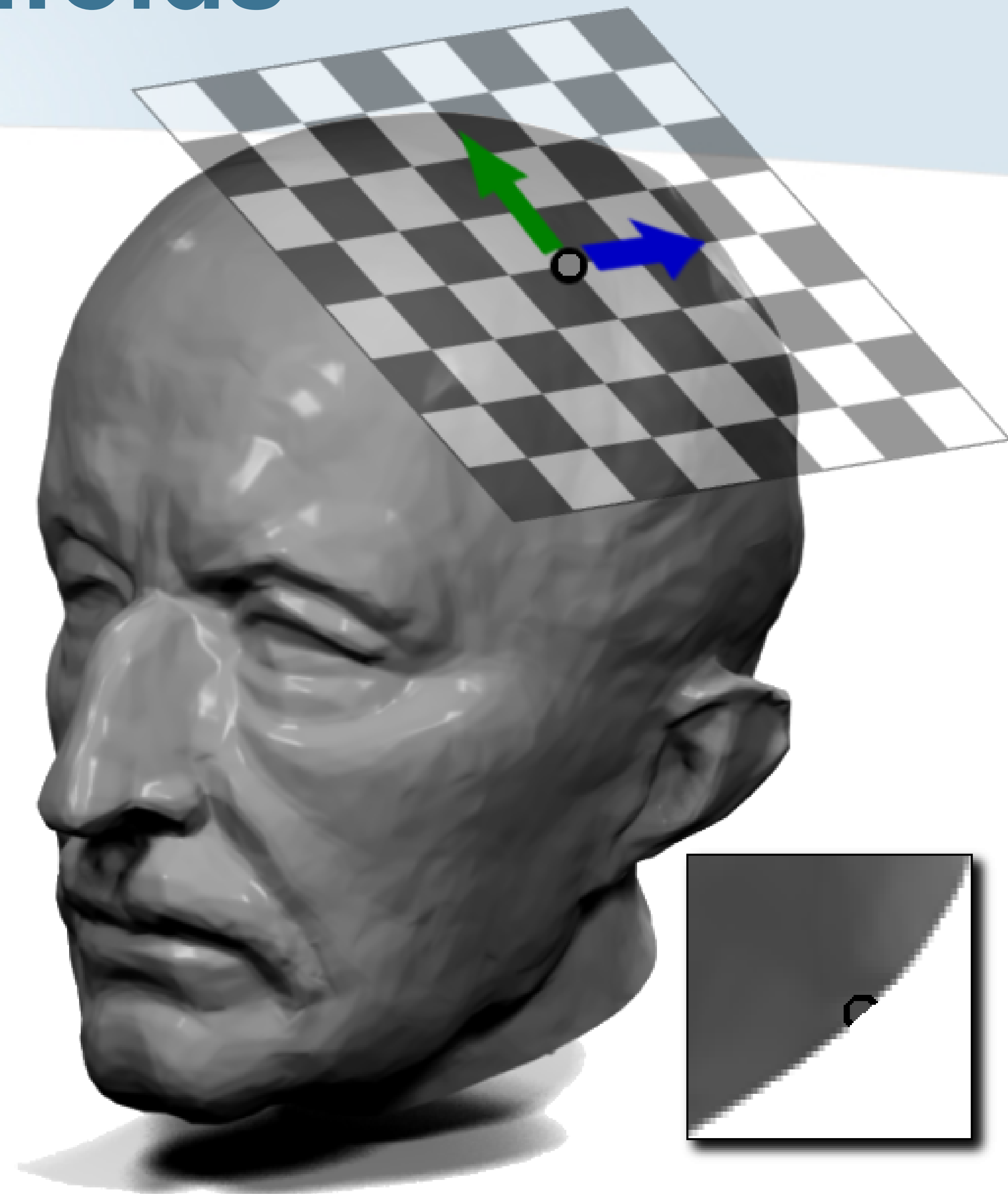
Domains of interest: Manifolds

- Tangent space
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 - Distances, angles, areas, ...



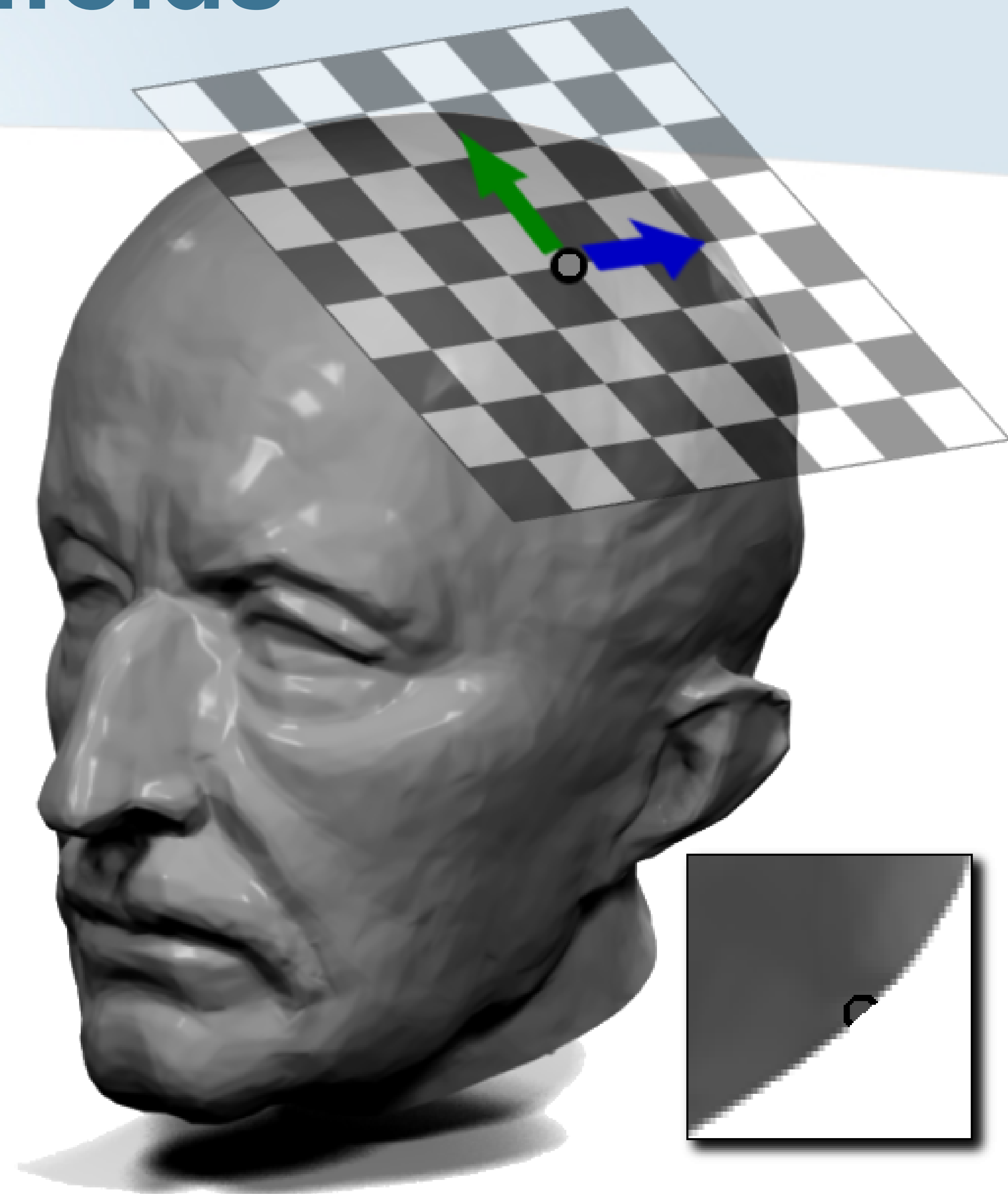
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Domains of interest: Manifolds

- Tangent space
 - Localized Euclidean approximation
 - Localized Euclidean geometry
 - Inner products
 - Distances, angles, areas, ...
 - Gradient, Laplace operator, etc.
- Discrete surface in \mathbb{R}^3
 - Triangular (planar) cells



Manifold examples

Manifold examples

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

\mathbb{R}

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

$[0, 1]$

\mathbb{R}

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

\mathbb{R}

$[0, 1]$

- Dimension?
- Embedding space?
- Boundary?

$$\{p \in \mathbb{R}^2 \mid \sqrt{p_x^2 + p_y^2} = r\}$$

Manifold examples

$$\mathbb{R}^2$$

$$[0, 1]$$

- Dimension?
- Embedding space?
- Boundary?

$$\{p \in \mathbb{R}^2 \mid \sqrt{p_x^2 + p_y^2} = r\}$$

Manifold examples

$$\mathbb{R}^2$$

$$[0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^2 \mid \sqrt{p_x^2 + p_y^2} = r\}$$

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

$$\mathbb{R}^2$$

$$[0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^2 \mid \sqrt{p_x^2 + p_y^2} \leq r\}$$

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

$$\mathbb{R}^3$$

$$[0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^2 \mid \sqrt{p_x^2 + p_y^2} \leq r\}$$

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

$$\mathbb{R}^3$$

$$[0, 1] \times [0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^2 \mid \sqrt{p_x^2 + p_y^2} \leq r\}$$

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

$$\mathbb{R}^3$$

$$[0, 1] \times [0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^3 \mid \sqrt{p_x^2 + p_y^2 + p_z^2} = r\}$$

- Dimension?
- Embedding space?
- Boundary?

Manifold examples

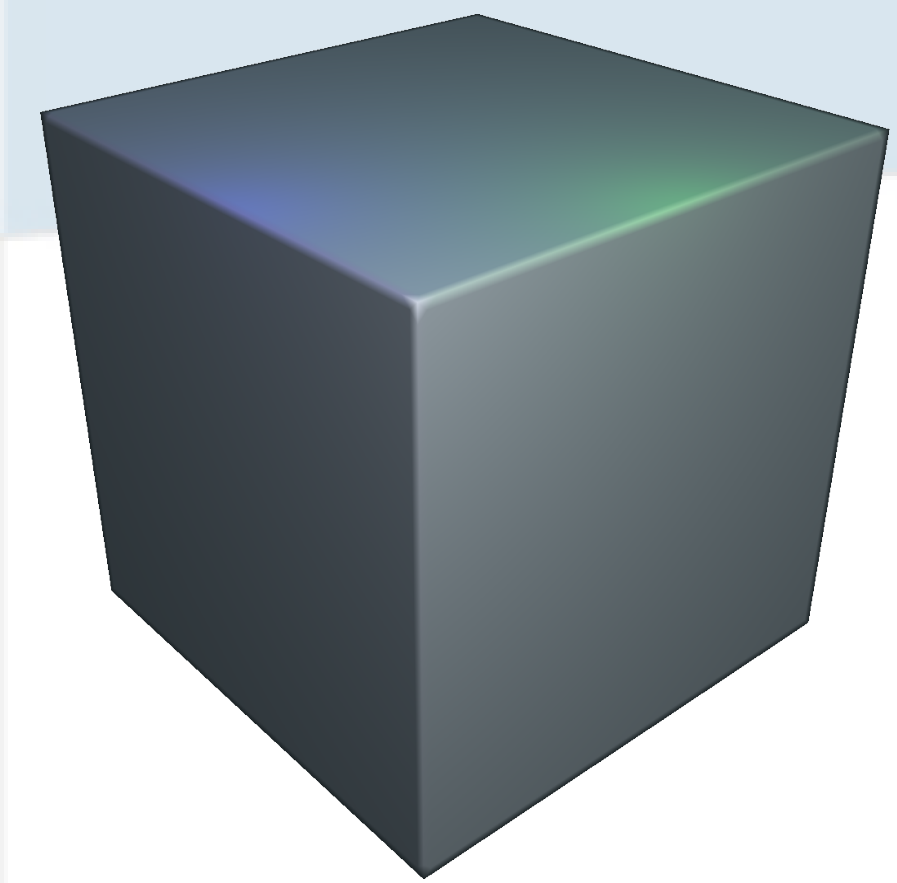
$$\mathbb{R}^3$$

$$[0, 1] \times [0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^3 \mid \sqrt{p_x^2 + p_y^2 + p_z^2} \leq r\}$$

- Dimension?
- Embedding space?
- Boundary?

Manifold examples



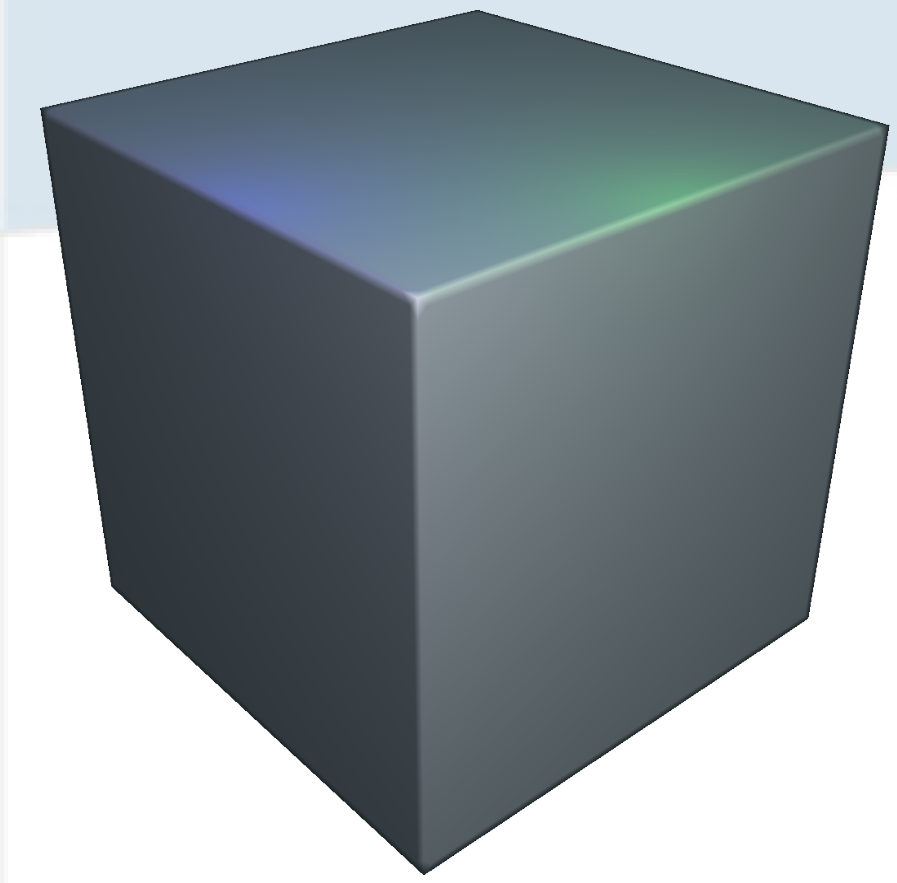
$$\mathbb{R}^3$$

$$[0, 1] \times [0, 1] \times [0, 1]$$

$$\{p \in \mathbb{R}^3 \mid \sqrt{p_x^2 + p_y^2 + p_z^2} \leq r\}$$

- Dimension?
- Embedding space?
- Boundary?

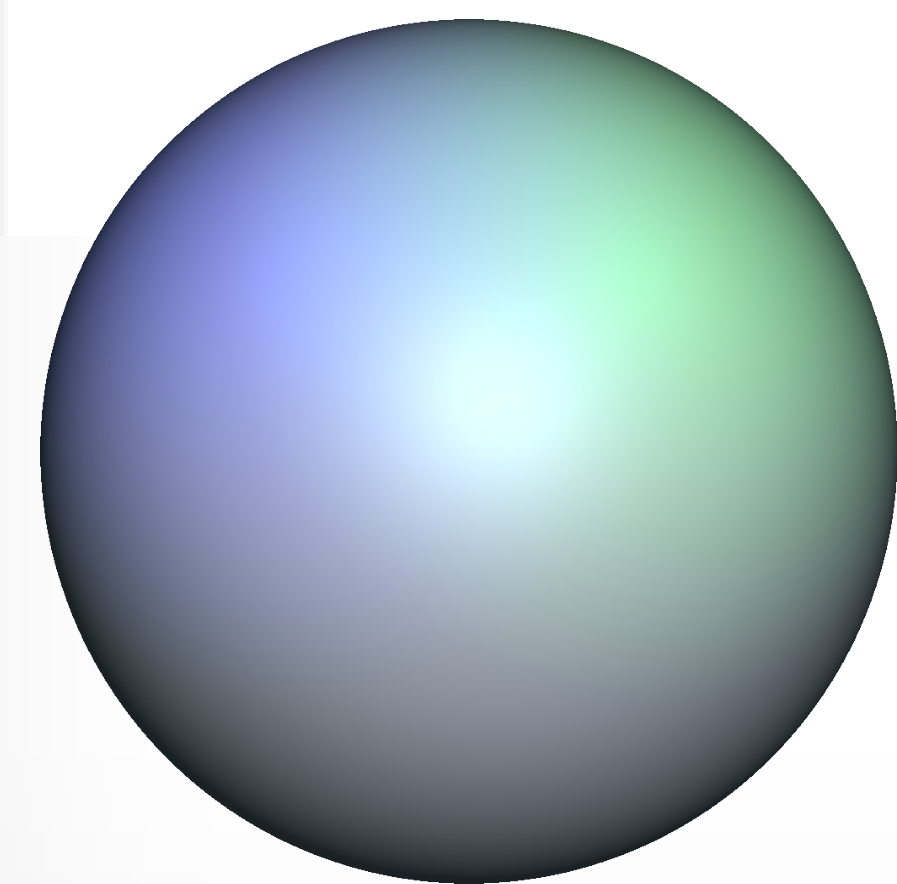
Manifold examples



$$\mathbb{R}^3$$

- Dimension?
- Embedding space?
- Boundary?

$$[0, 1] \times [0, 1] \times [0, 1]$$



$$\{p \in \mathbb{R}^3 \mid \sqrt{p_x^2 + p_y^2 + p_z^2} \leq r\}$$

Summary

Summary

- Euclidean representations

Summary

- Euclidean representations
- Non-euclidean representations

Summary

- Euclidean representations
- Non-euclidean representations
- Dimension-specific data-structures

Summary

- Euclidean representations
- Non-euclidean representations
- Dimension-specific data-structures
- Dimension-specific interpolants

Summary

- Euclidean representations
- Non-euclidean representations
- Dimension-specific data-structures
- Dimension-specific interpolants
- Implementation examples

Euclidean spaces

Euclidean spaces

- Named after Greek mathematician Euclid
 - 325 BC, 265 BC

Euclidean spaces

- Named after Greek mathematician Euclid
 - 325 BC, 265 BC



Euclidean spaces

- Named after Greek mathematician Euclid
 - 325 BC, 265 BC
- *“Elements”* (300 BC)



Euclidean spaces

- Named after Greek mathematician Euclid
 - 325 BC, 265 BC
- “*Elements*” (300 BC)
 - First *systematic* math book



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Euclidean spaces

- Named after Greek mathematician Euclid
 - 325 BC, 265 BC
- “*Elements*” (300 BC)
 - First *systematic* math book
 - Definitions, axioms, demonstrations
 - 5 axioms (first tome)



Euclidean spaces on a computer



Euclidean spaces on a computer

- Given an origin



Euclidean spaces on a computer

- Given an origin
 - $o = (0, 0, \dots, 0) \in \mathbb{R}^n$



Euclidean spaces on a computer

- Given an origin
 - $o = (0, 0, \dots, 0) \in \mathbb{R}^n$
- And an orthonormal basis



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- Points can be identified uniquely in that space



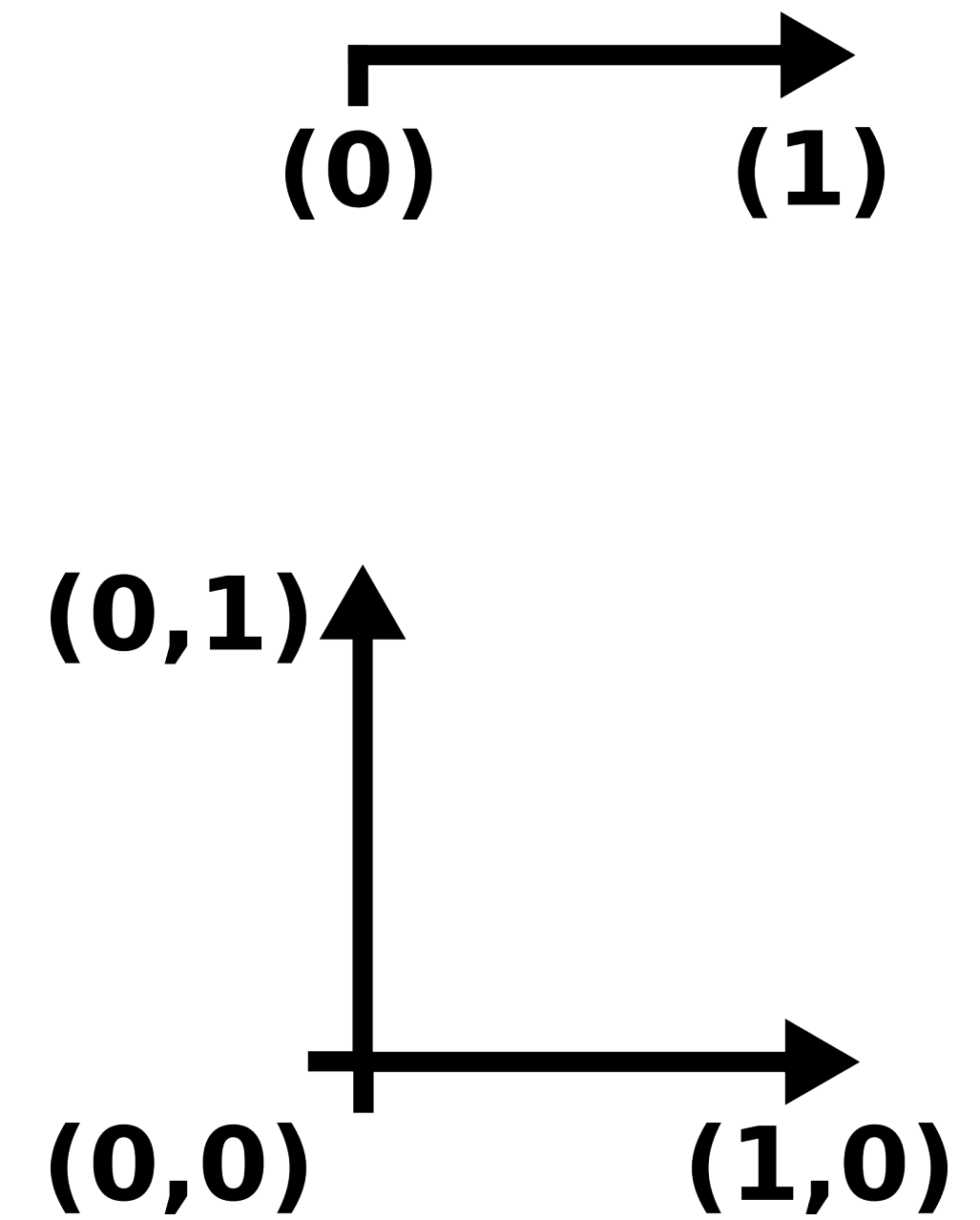
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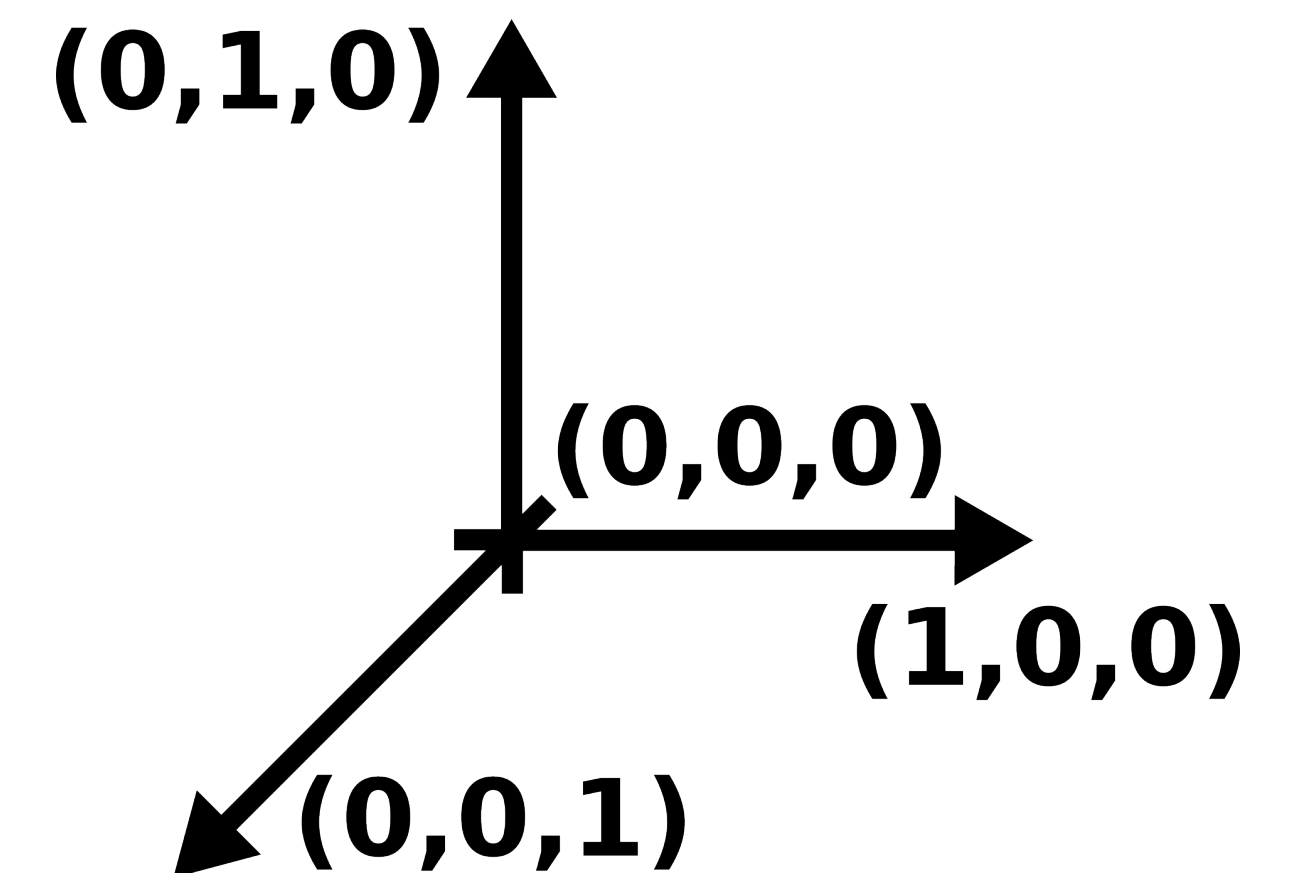
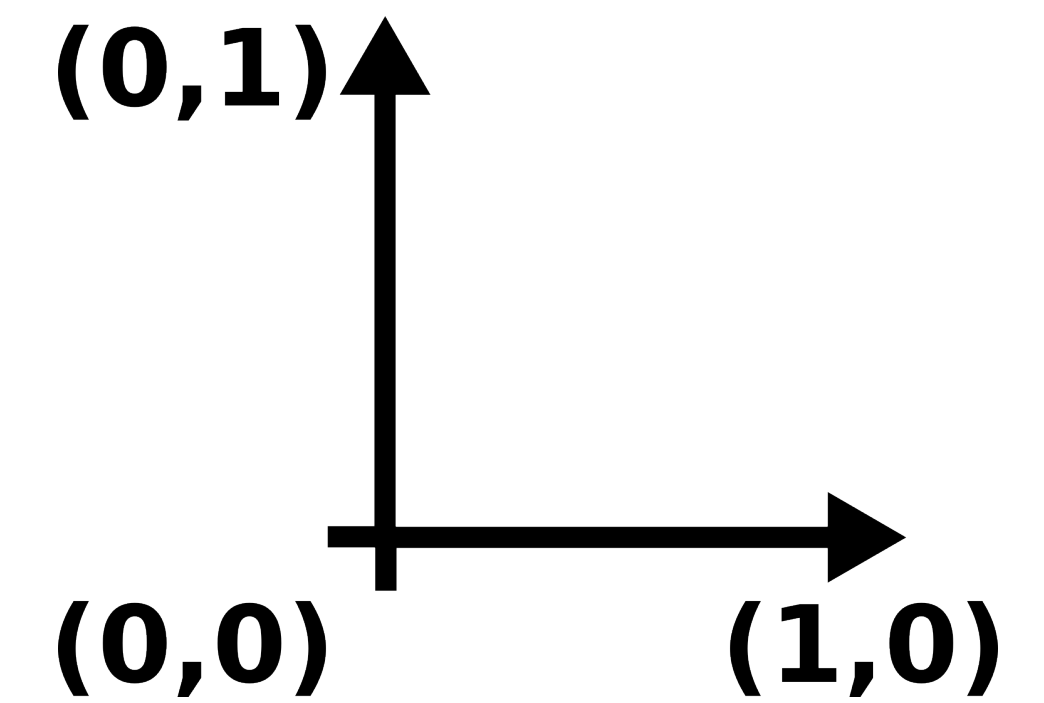
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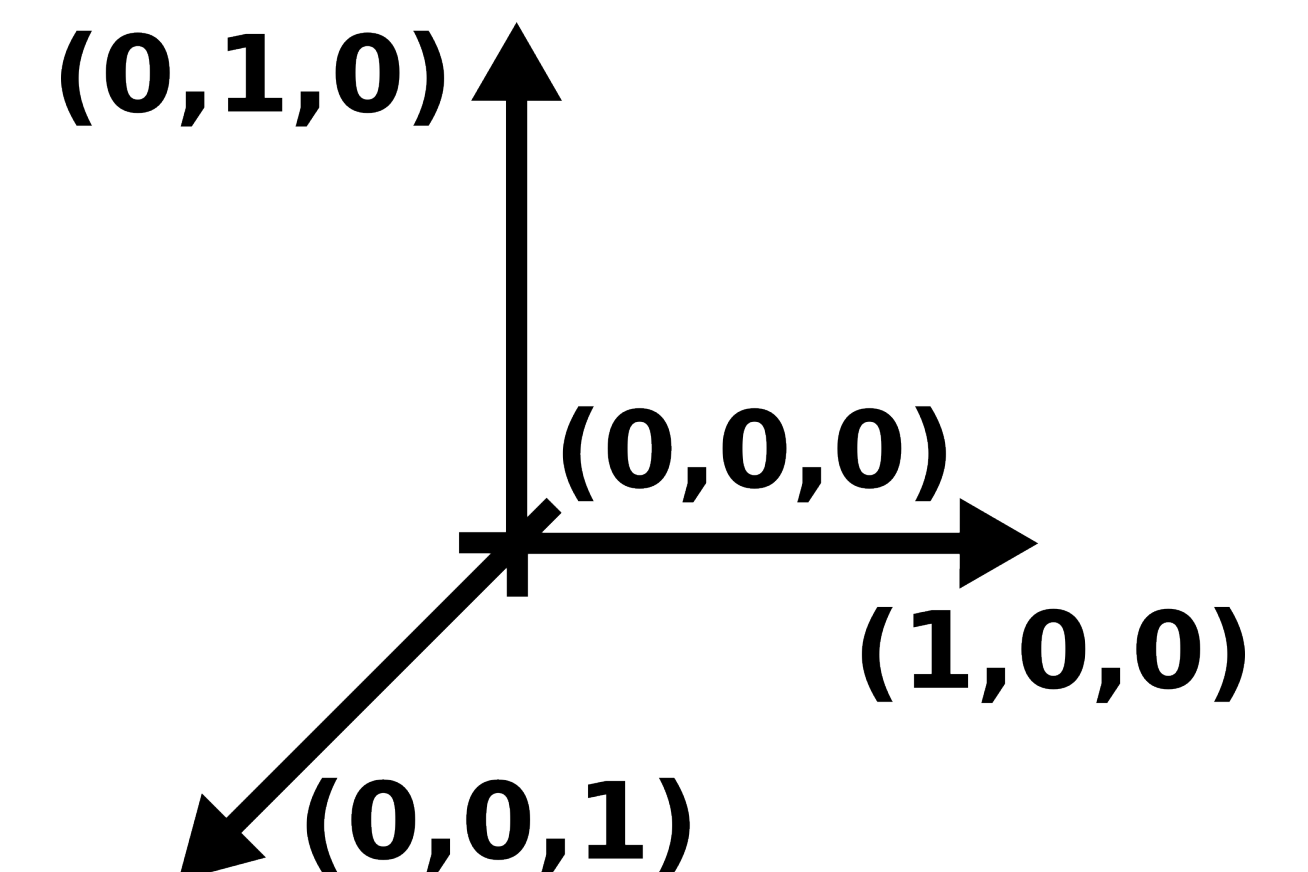
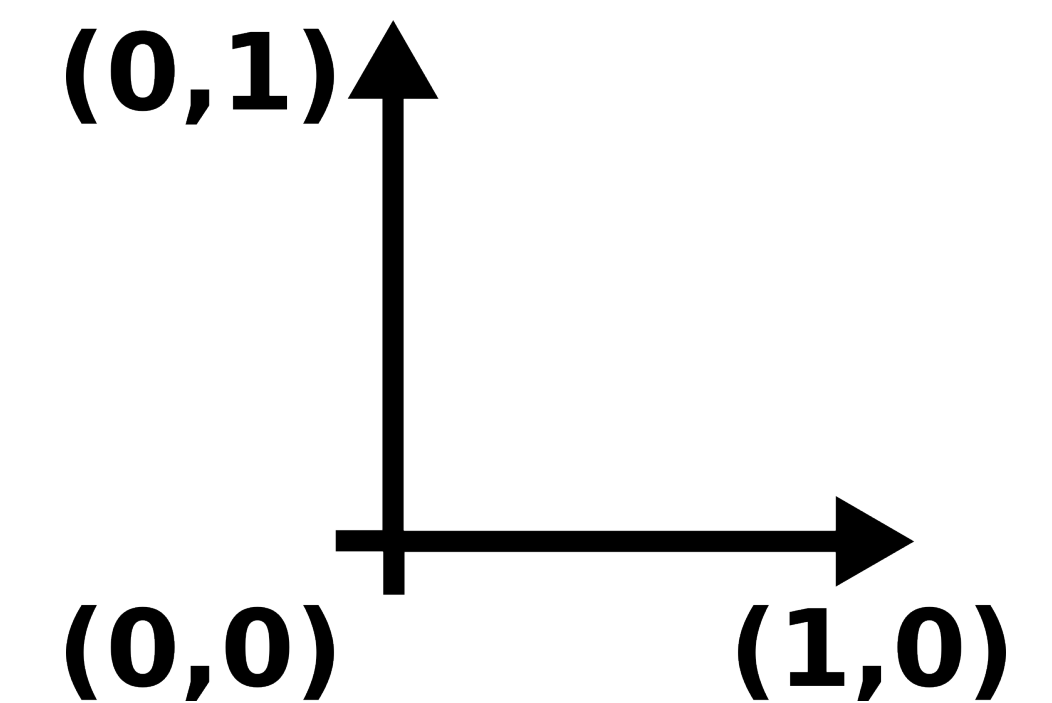
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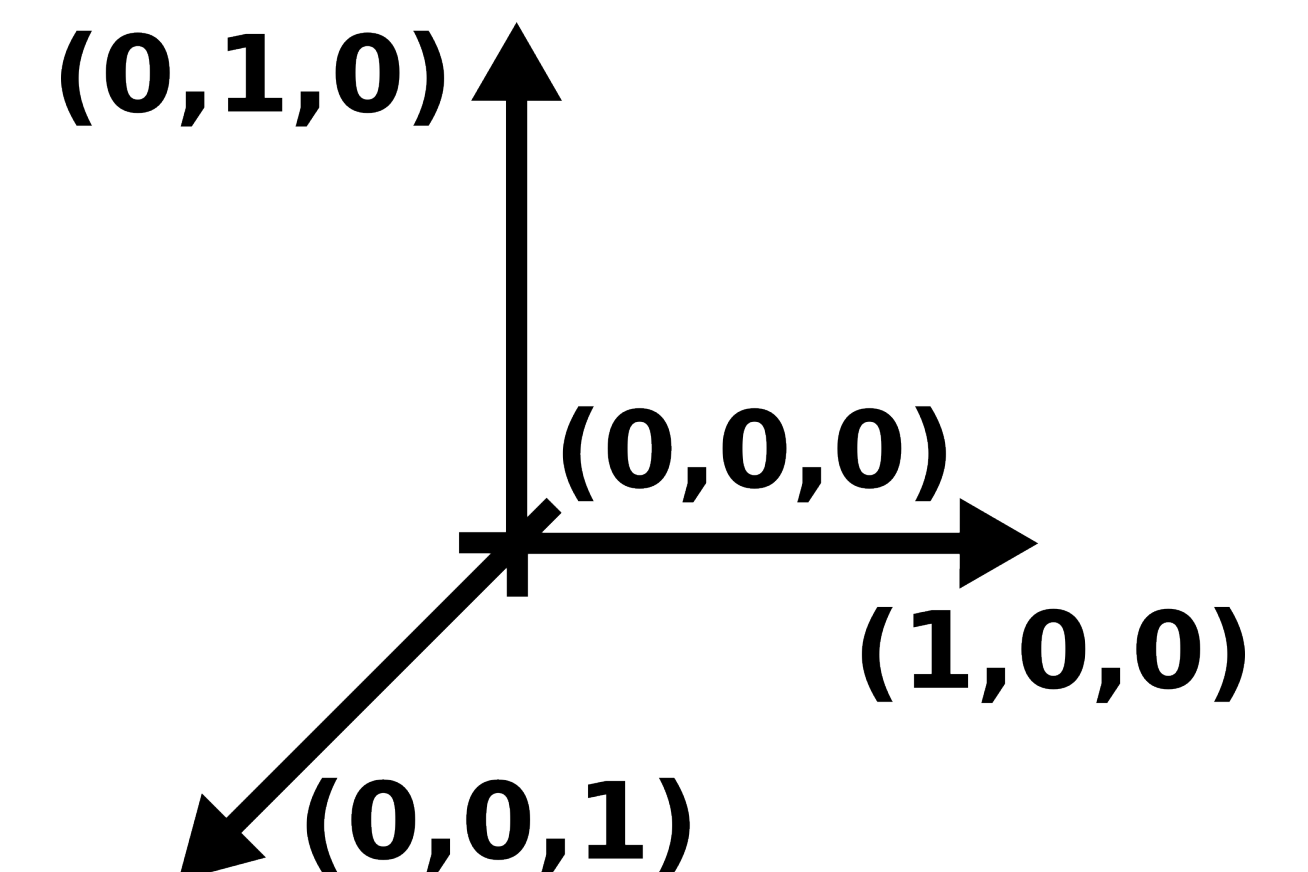
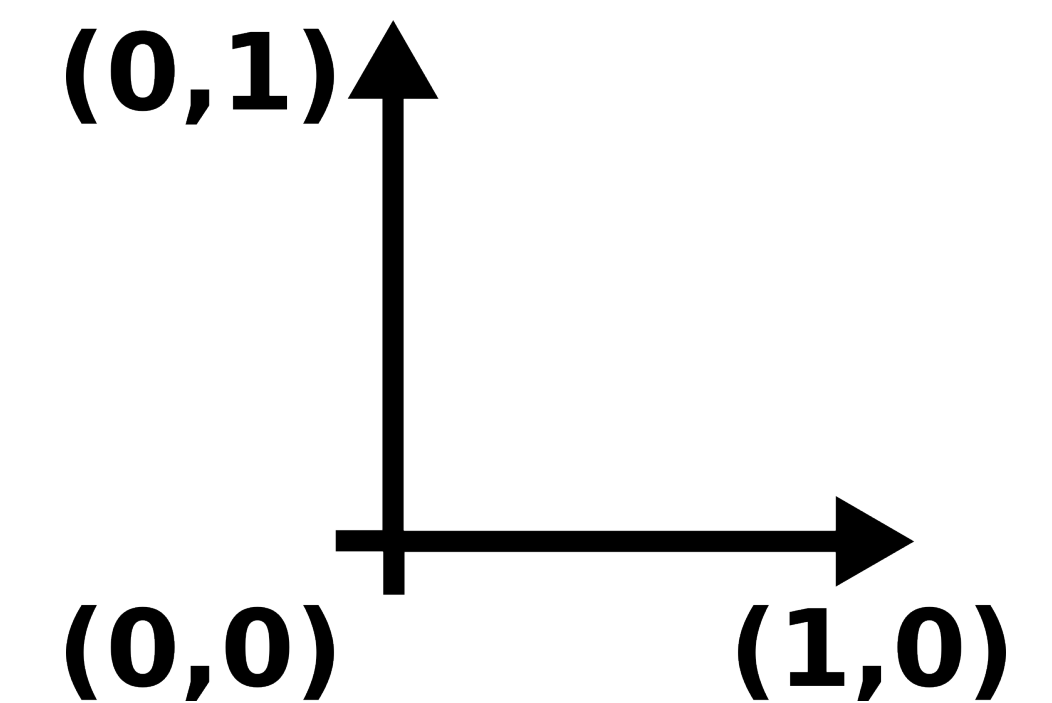
Euclidean spaces on a computer

- Direct product of closed unit intervals



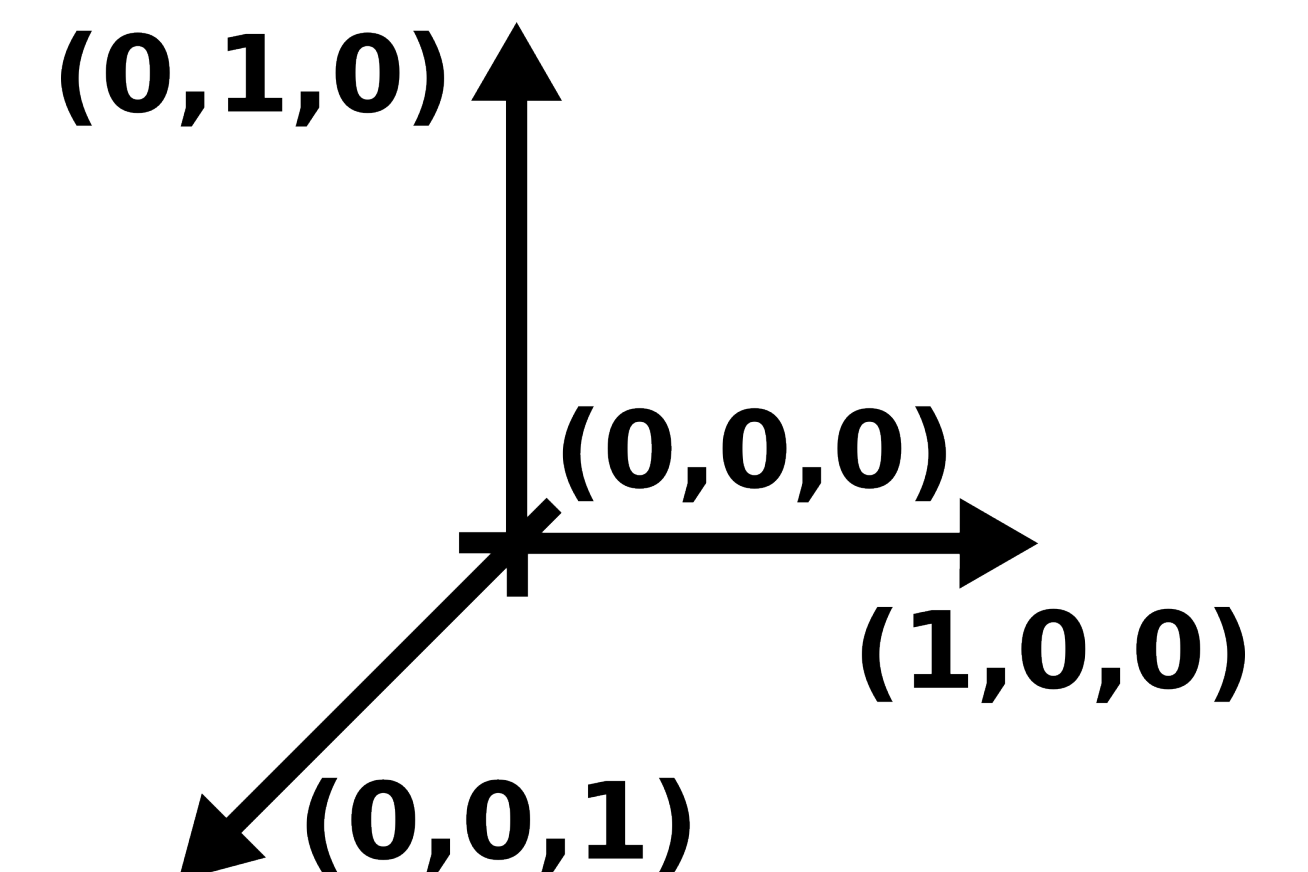
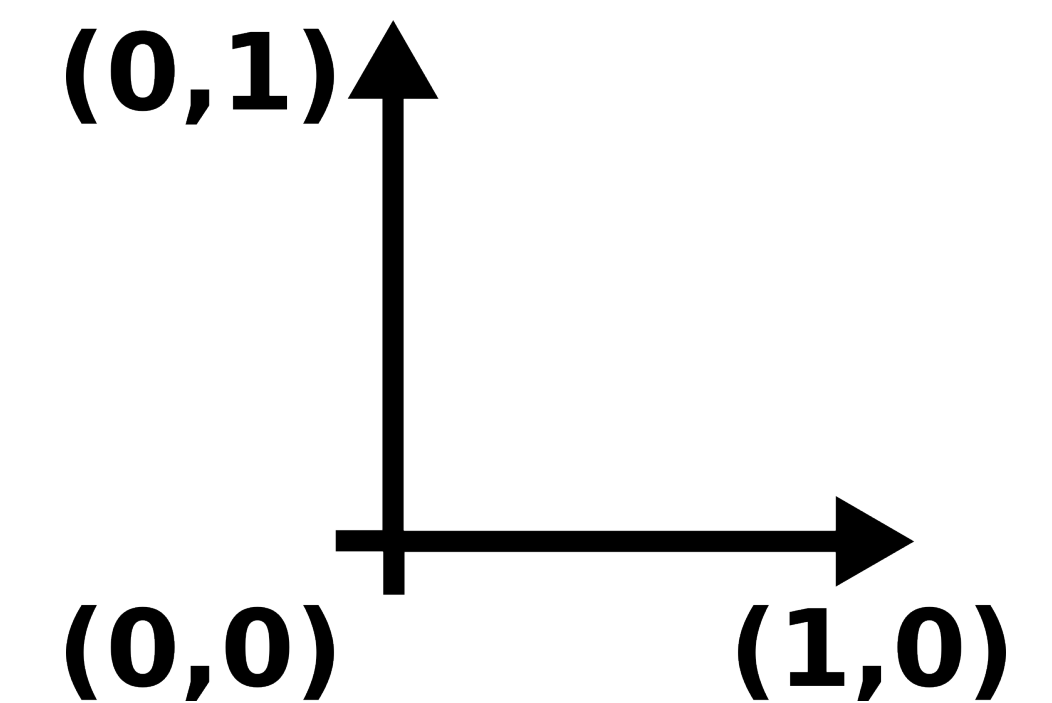
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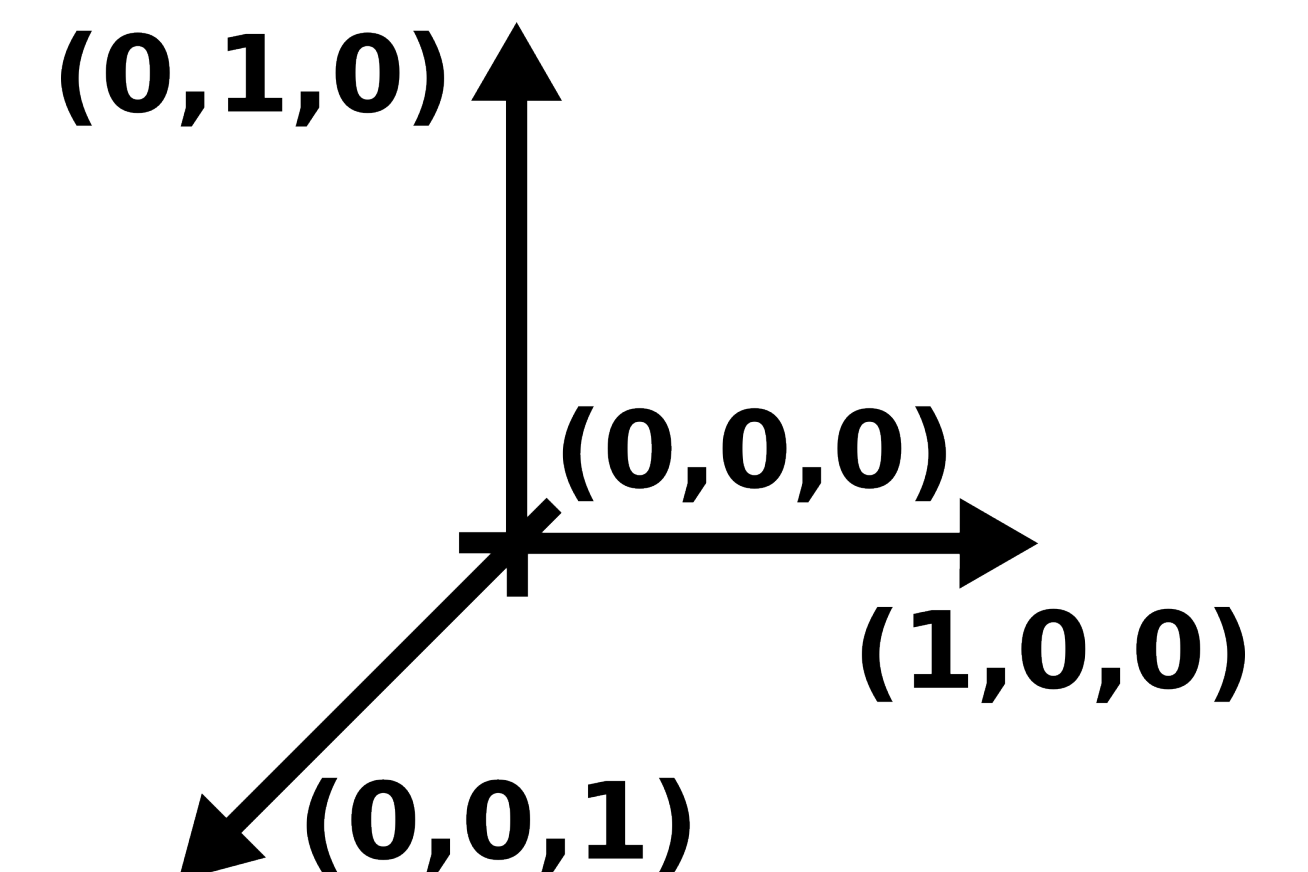
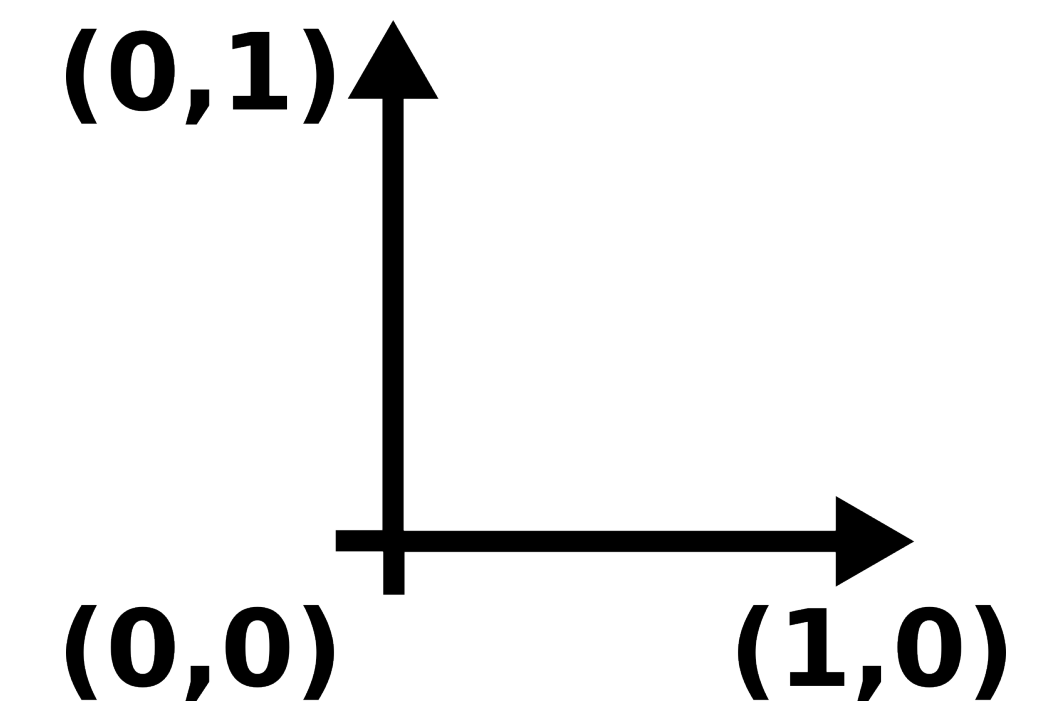
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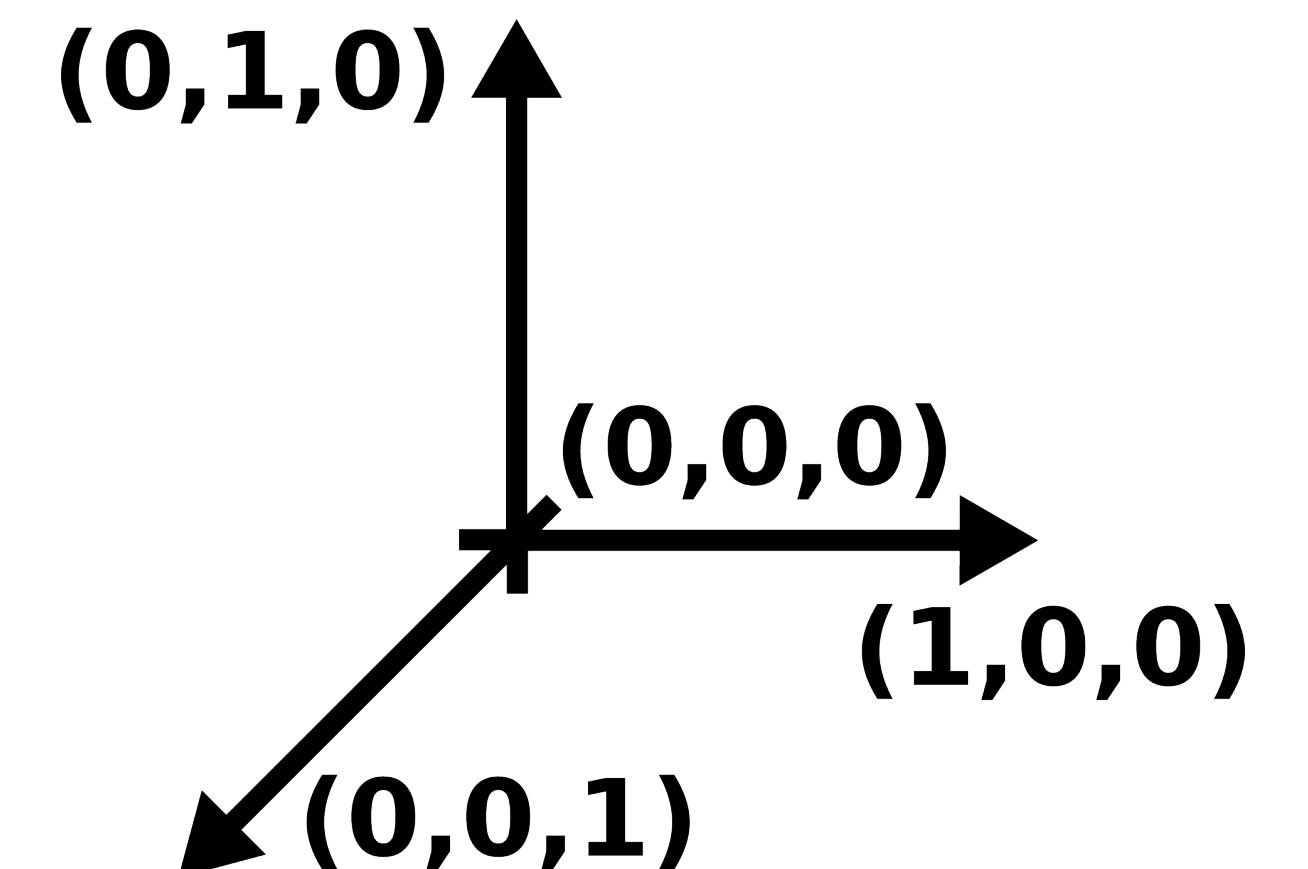
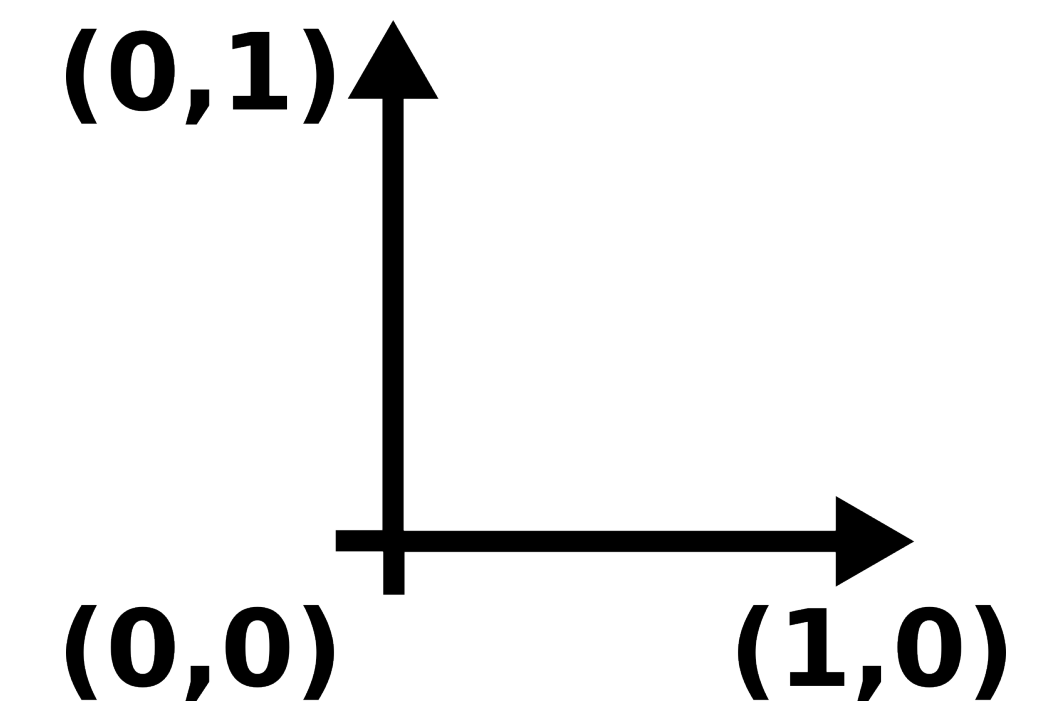
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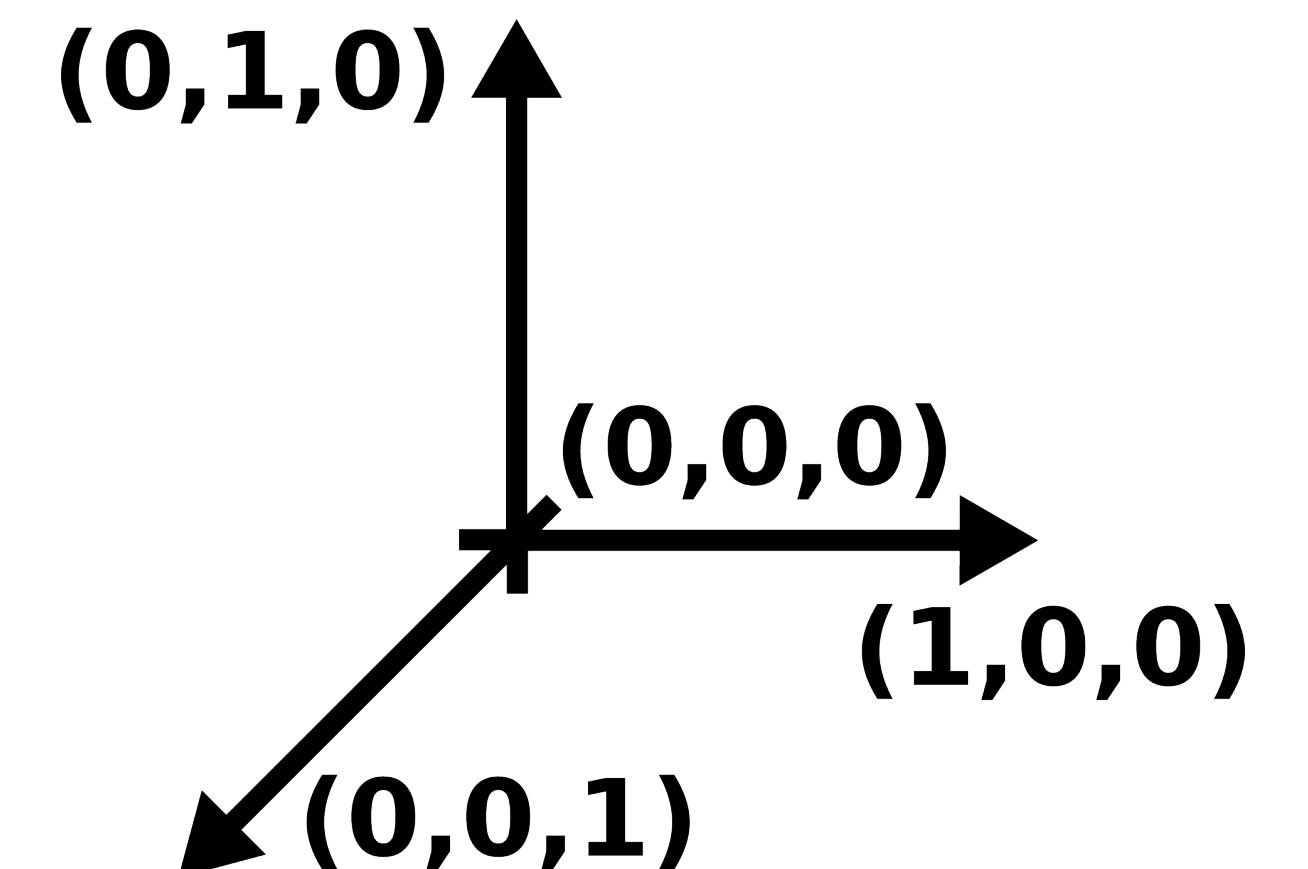
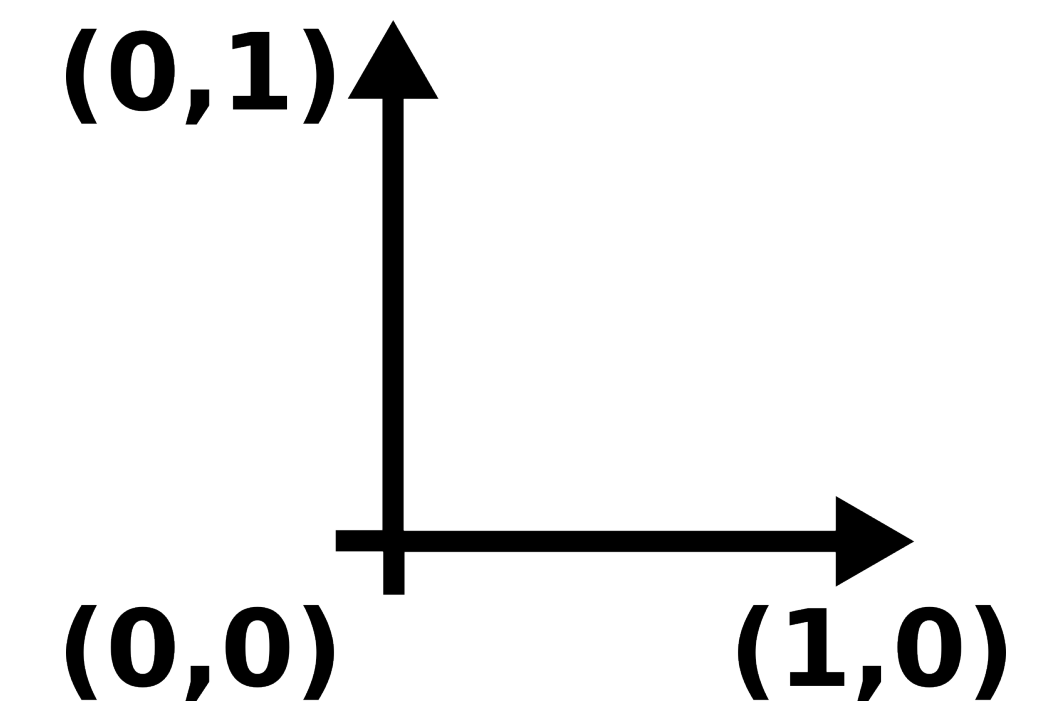
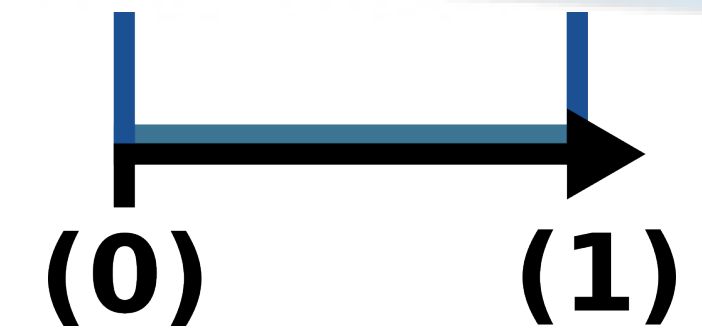
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- Direct product of closed unit intervals
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 - Covering a bounded, compact region of \mathbb{R}^n



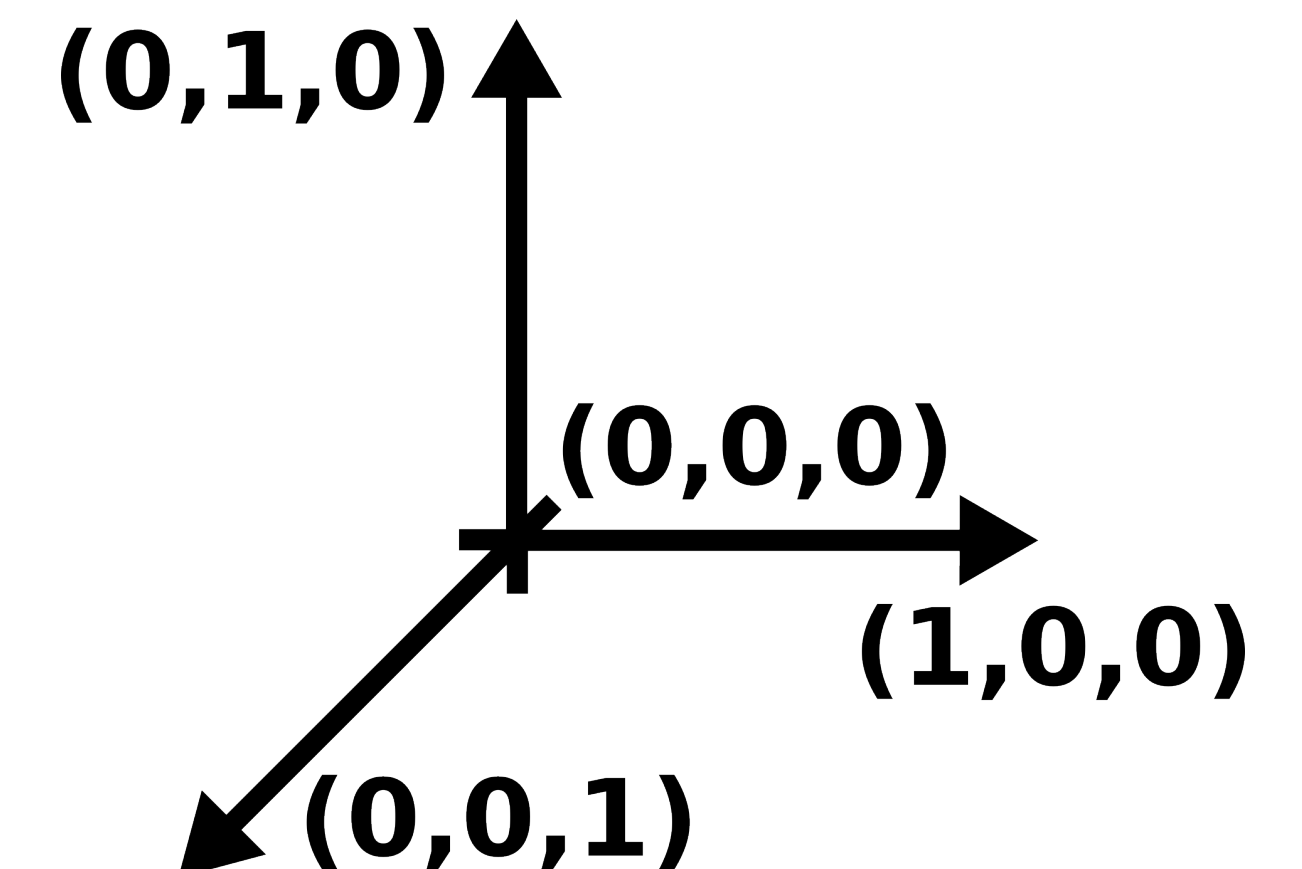
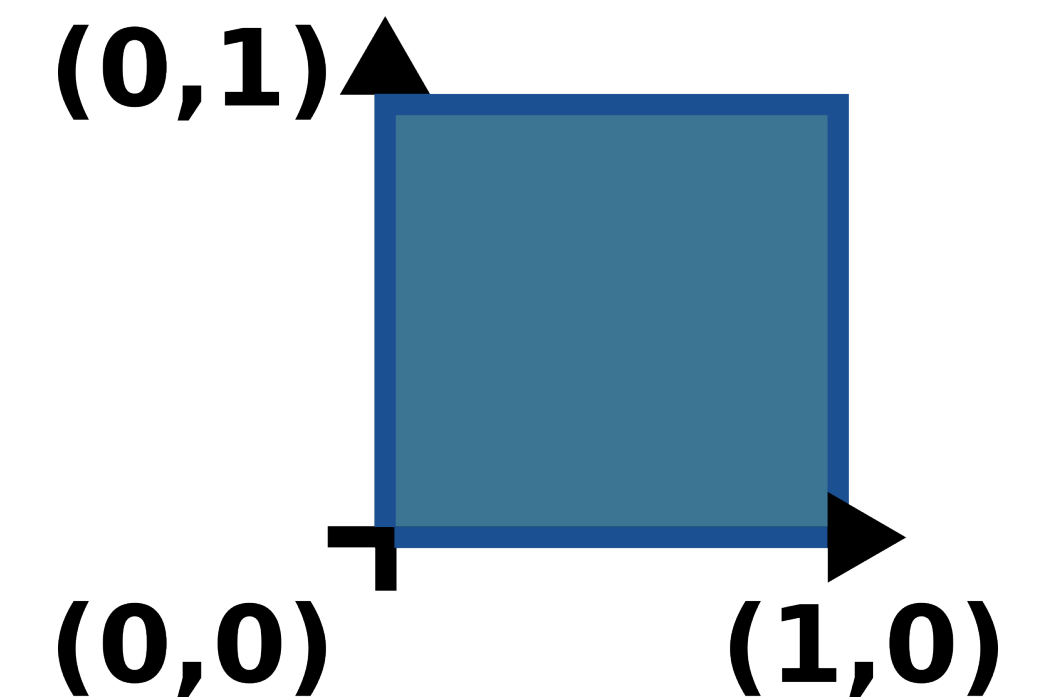
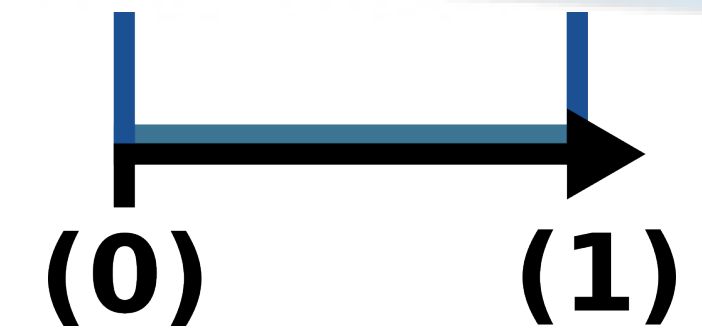
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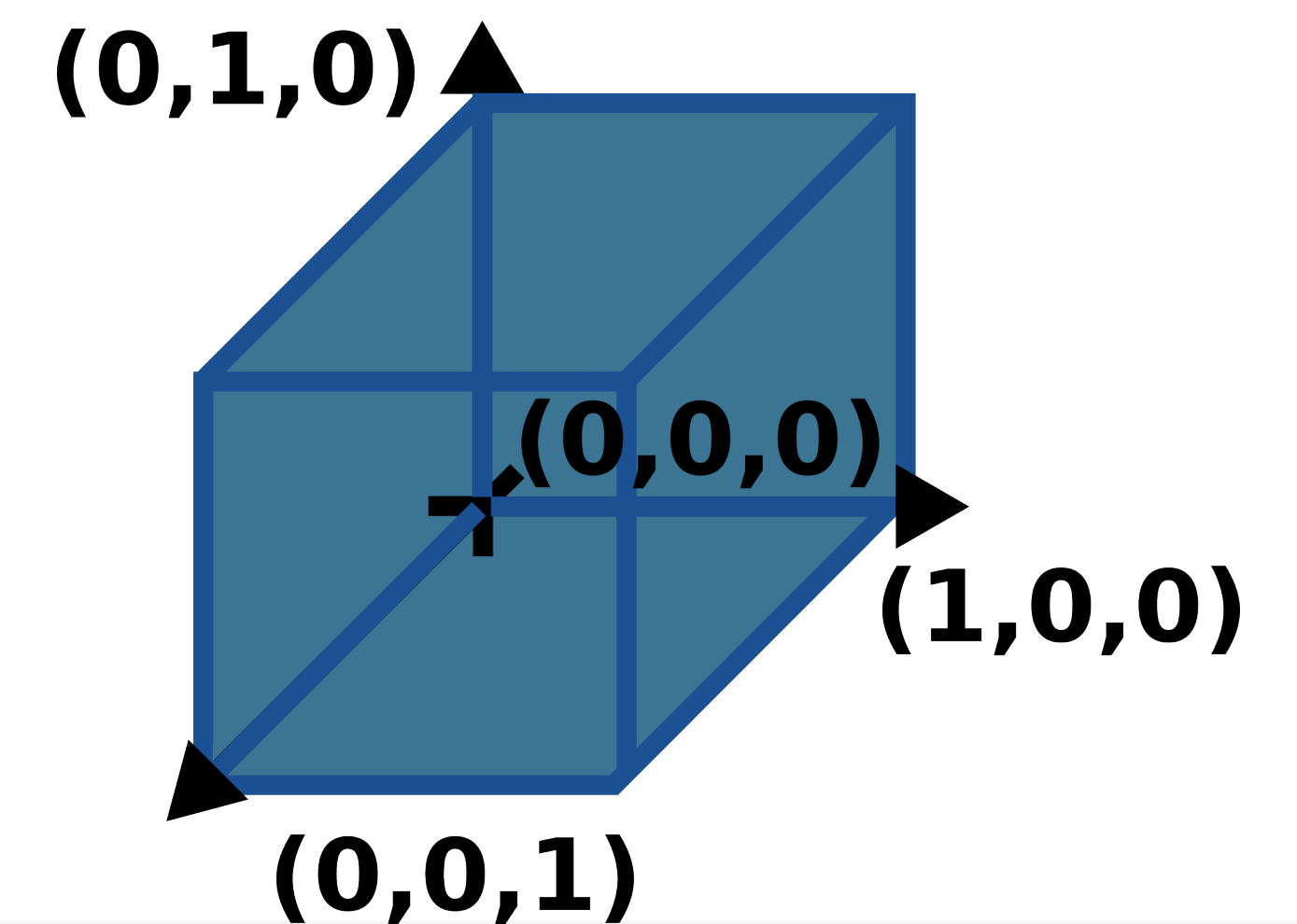
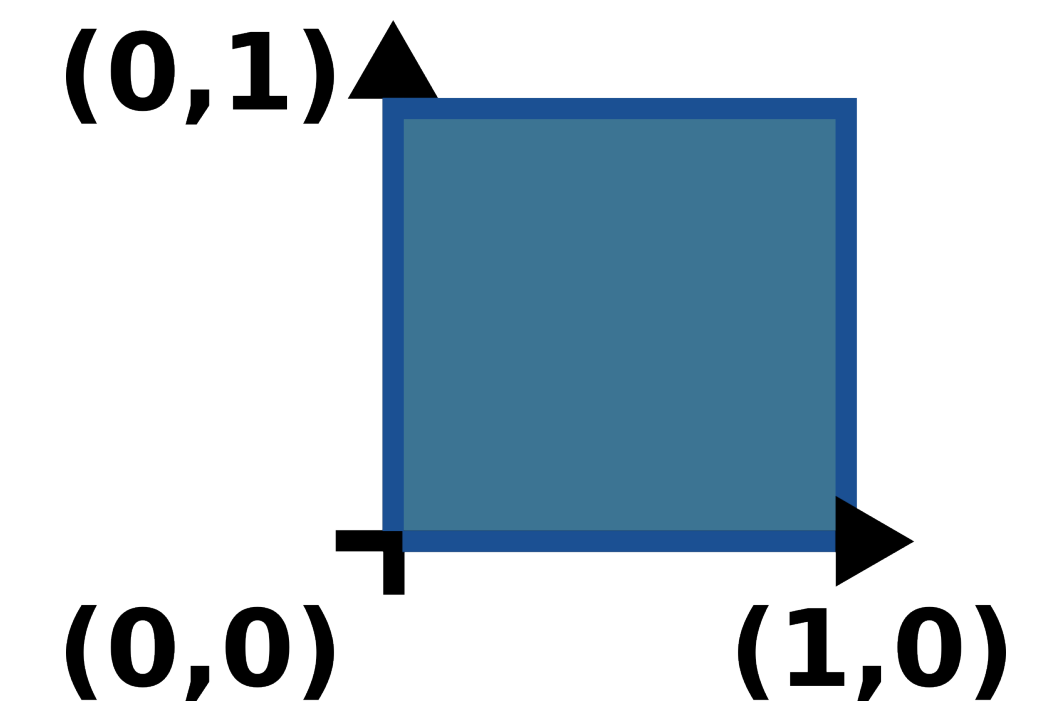
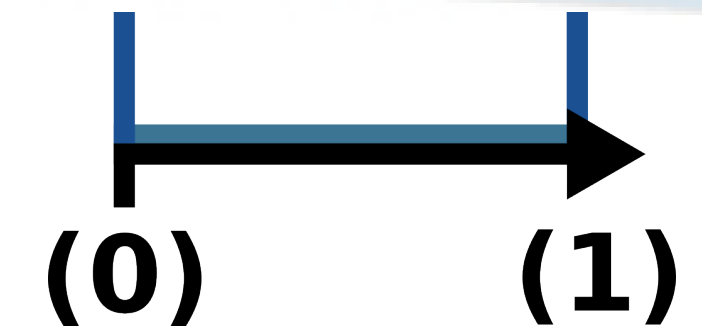
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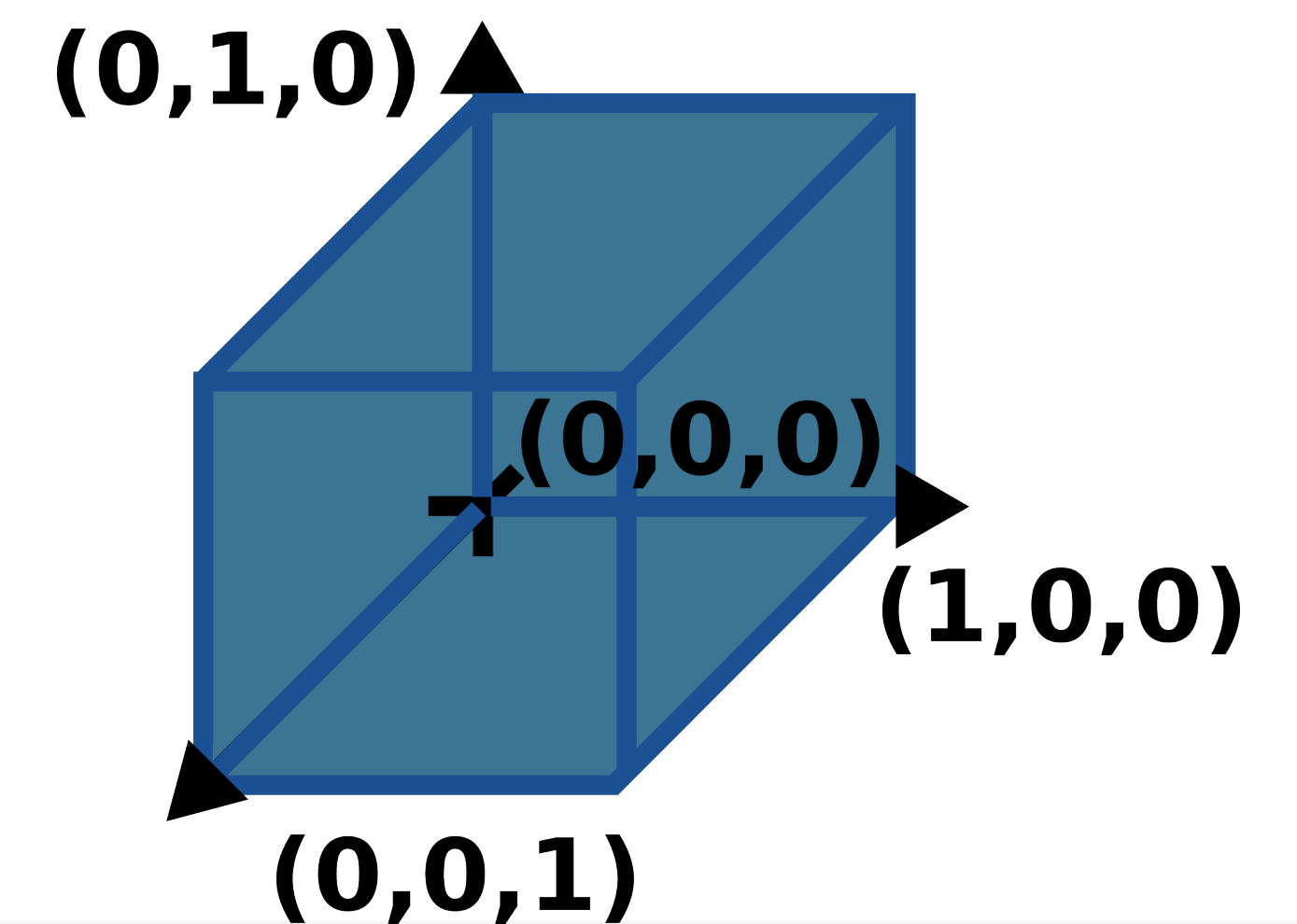
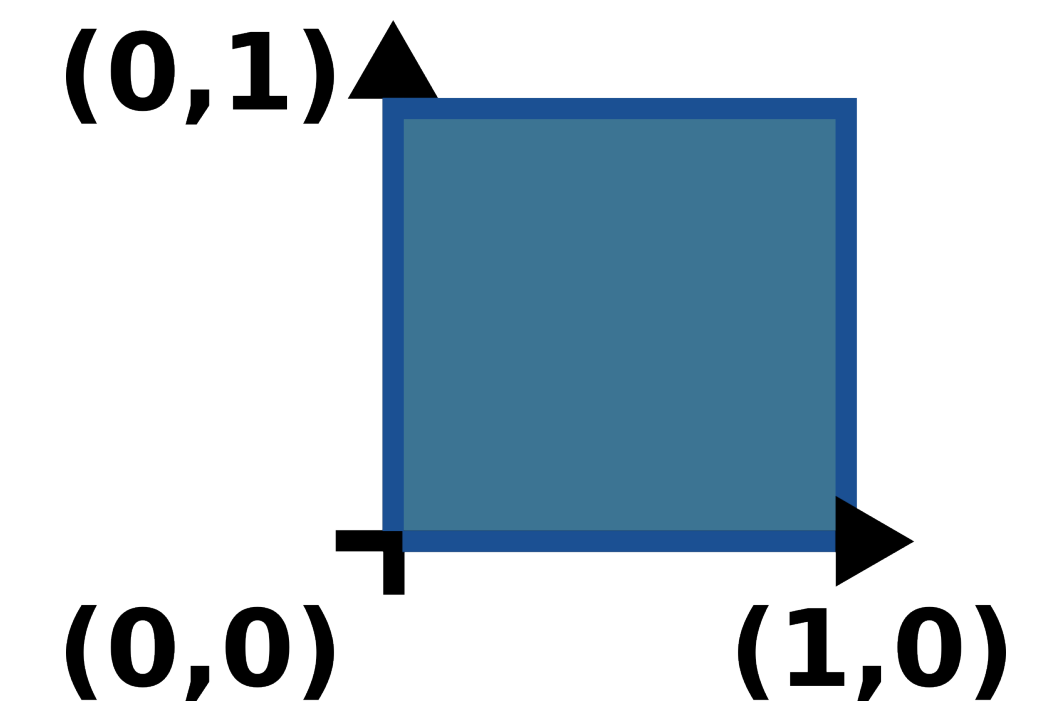
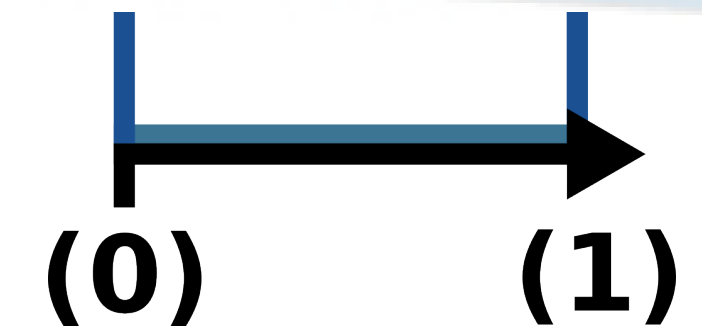
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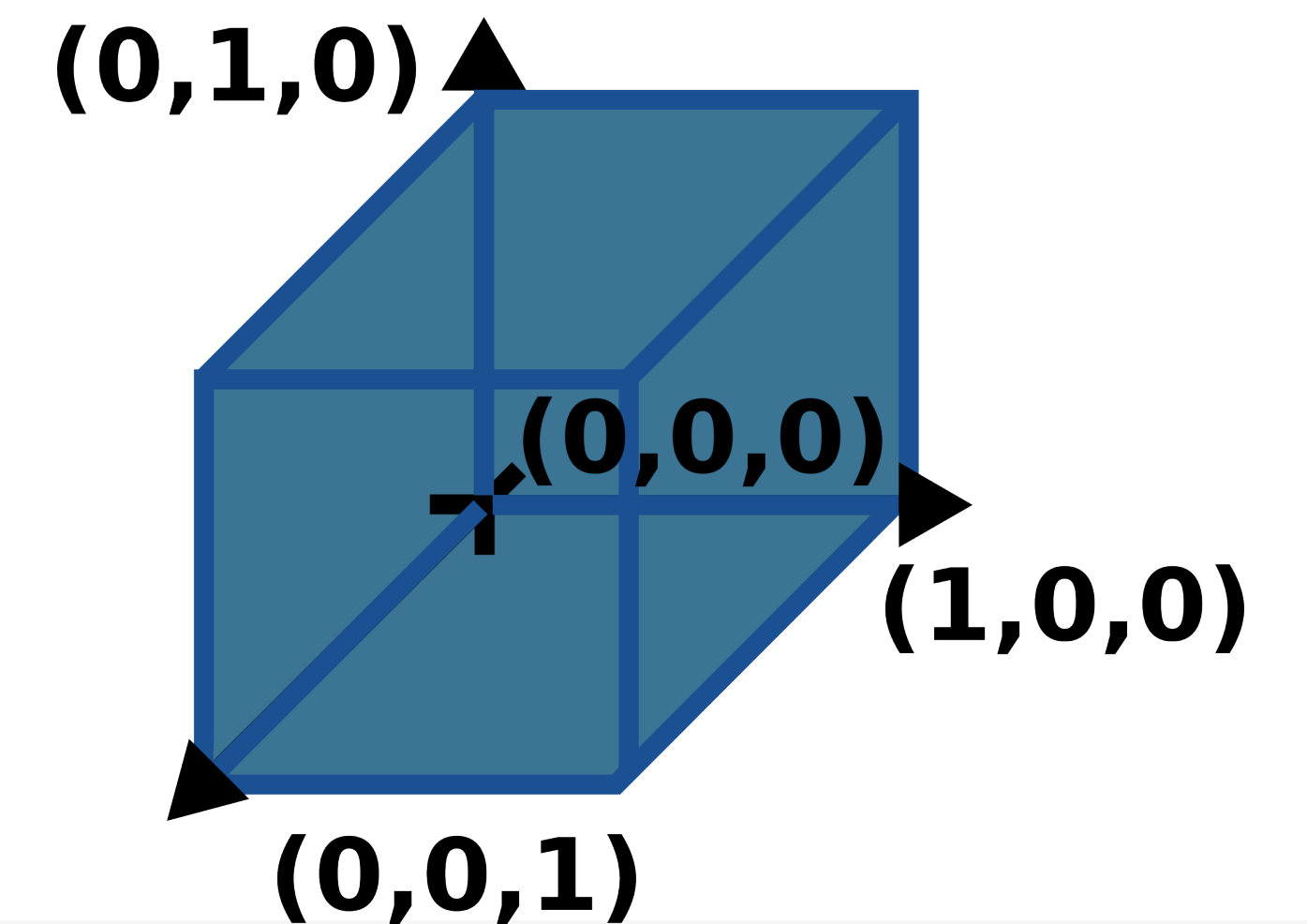
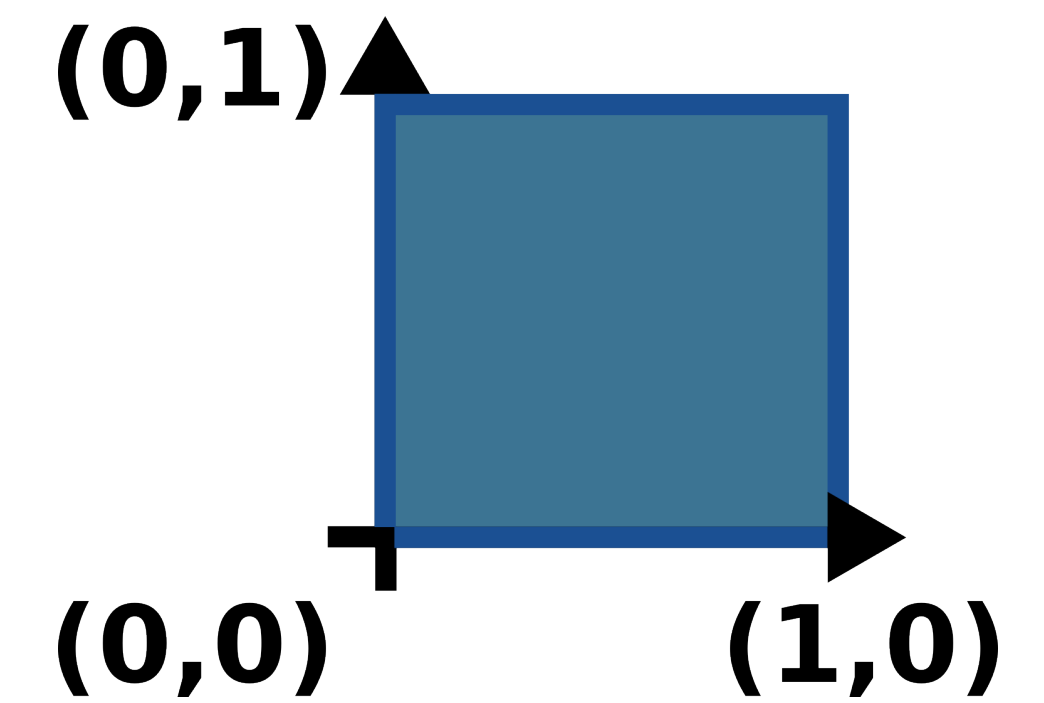
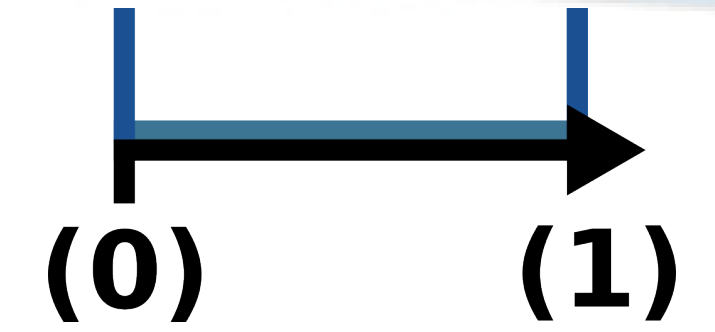
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 - Unit translations along the vectors of the orthonormal basis
 - Covering a bounded, compact region of \mathbb{R}^n
- Unit cell



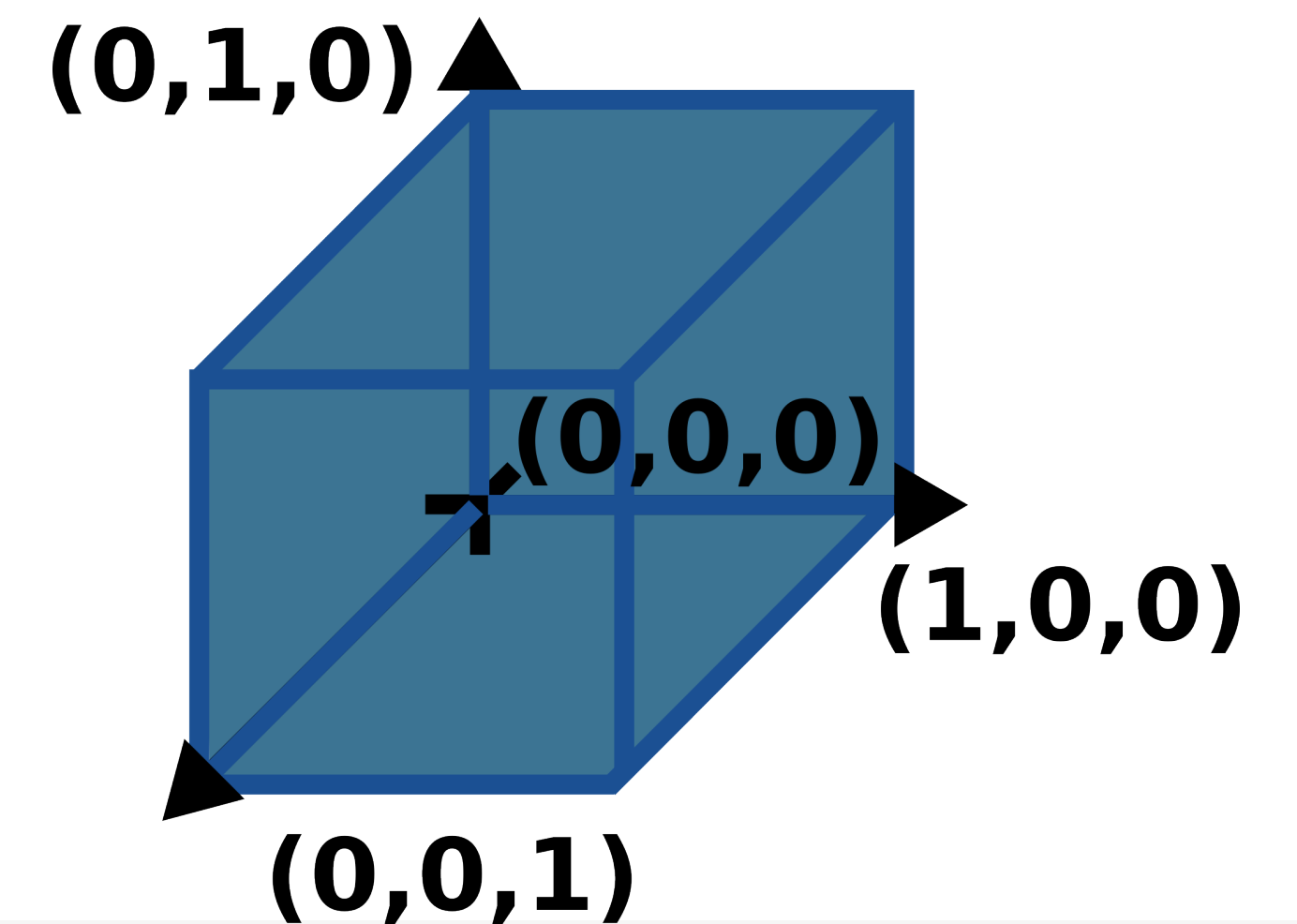
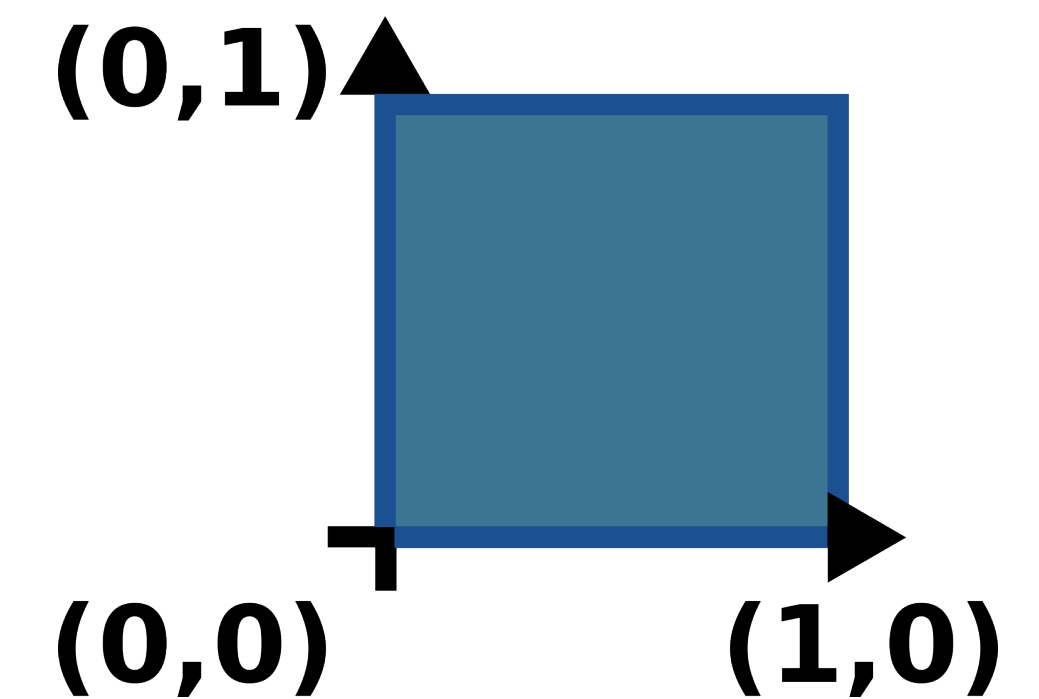
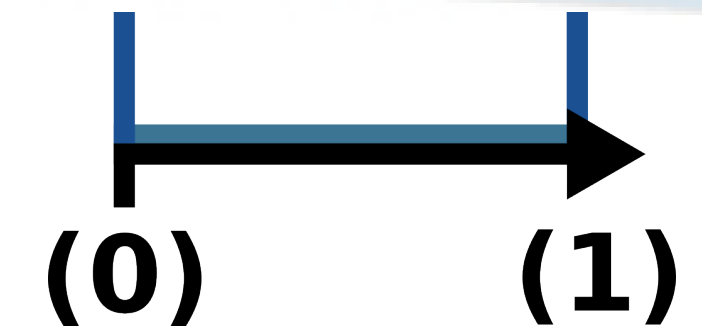
Euclidean spaces on a computer

- Given a unit cell of \mathbb{R}^n



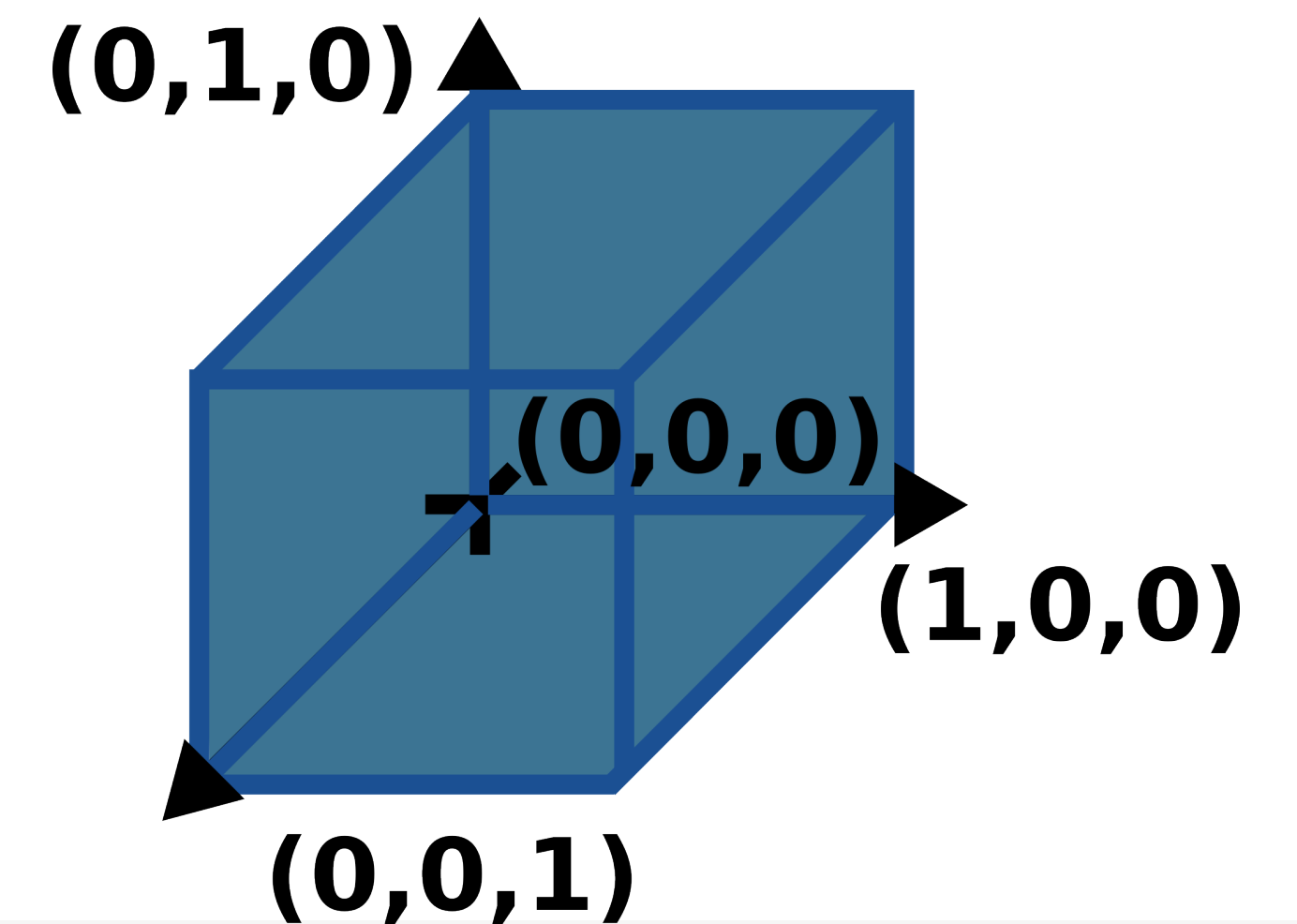
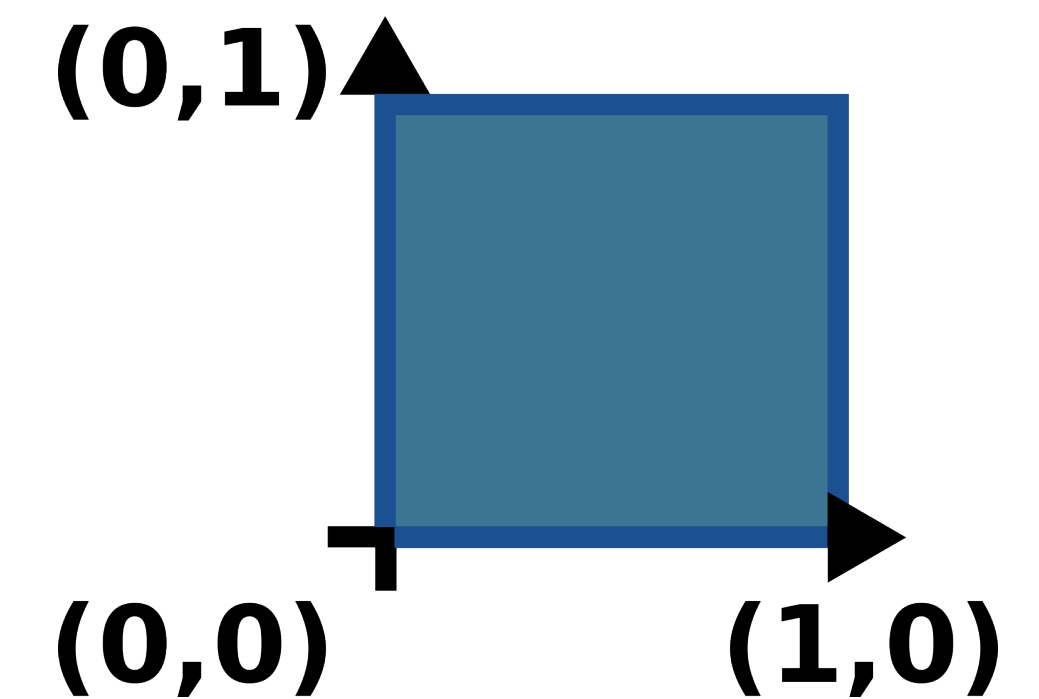
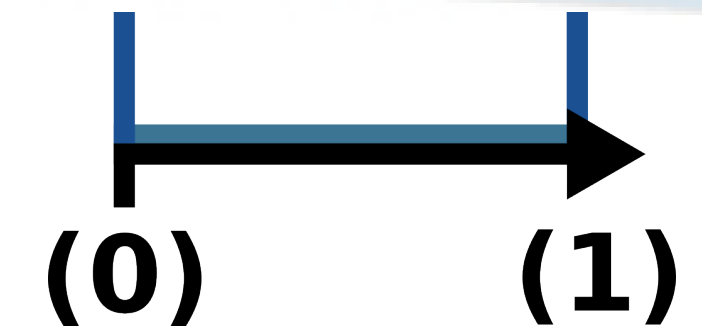
Euclidean spaces on a computer

- Given a unit cell of \mathbb{R}^n
 - What's the minimum number of samples necessary to describe its convex hull?



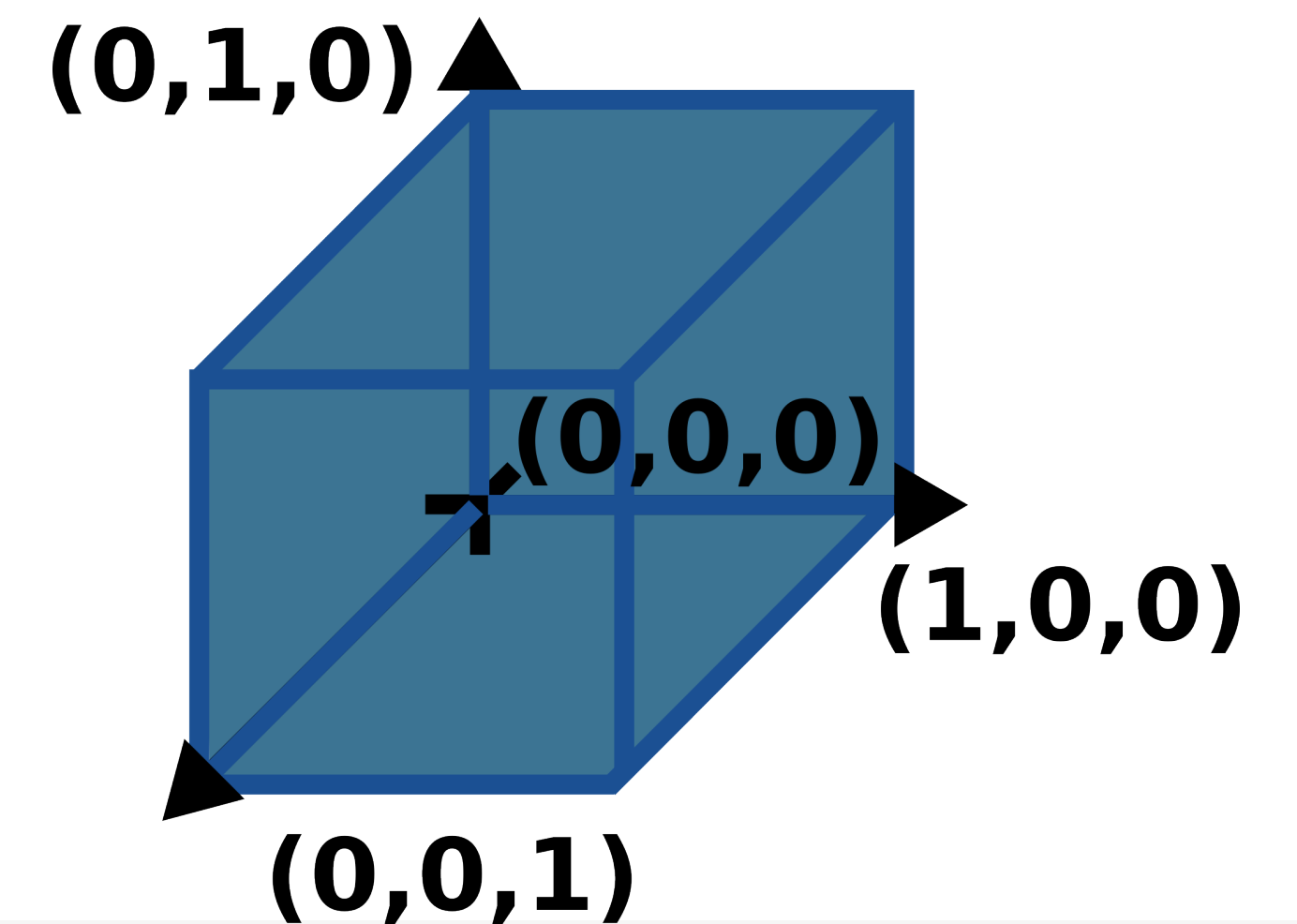
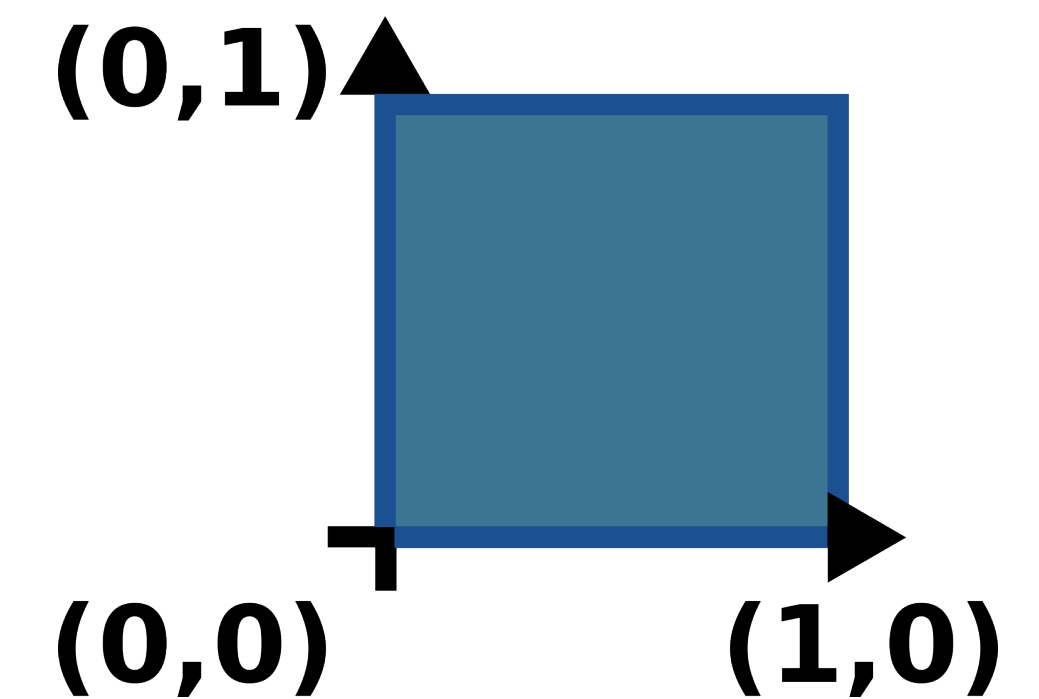
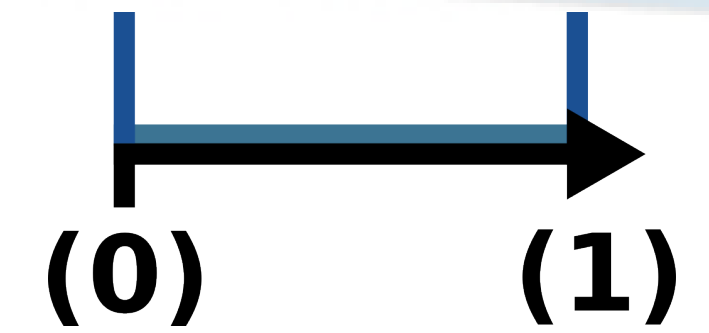
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- Given a unit cell of \mathbb{R}^n
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 - 2^n



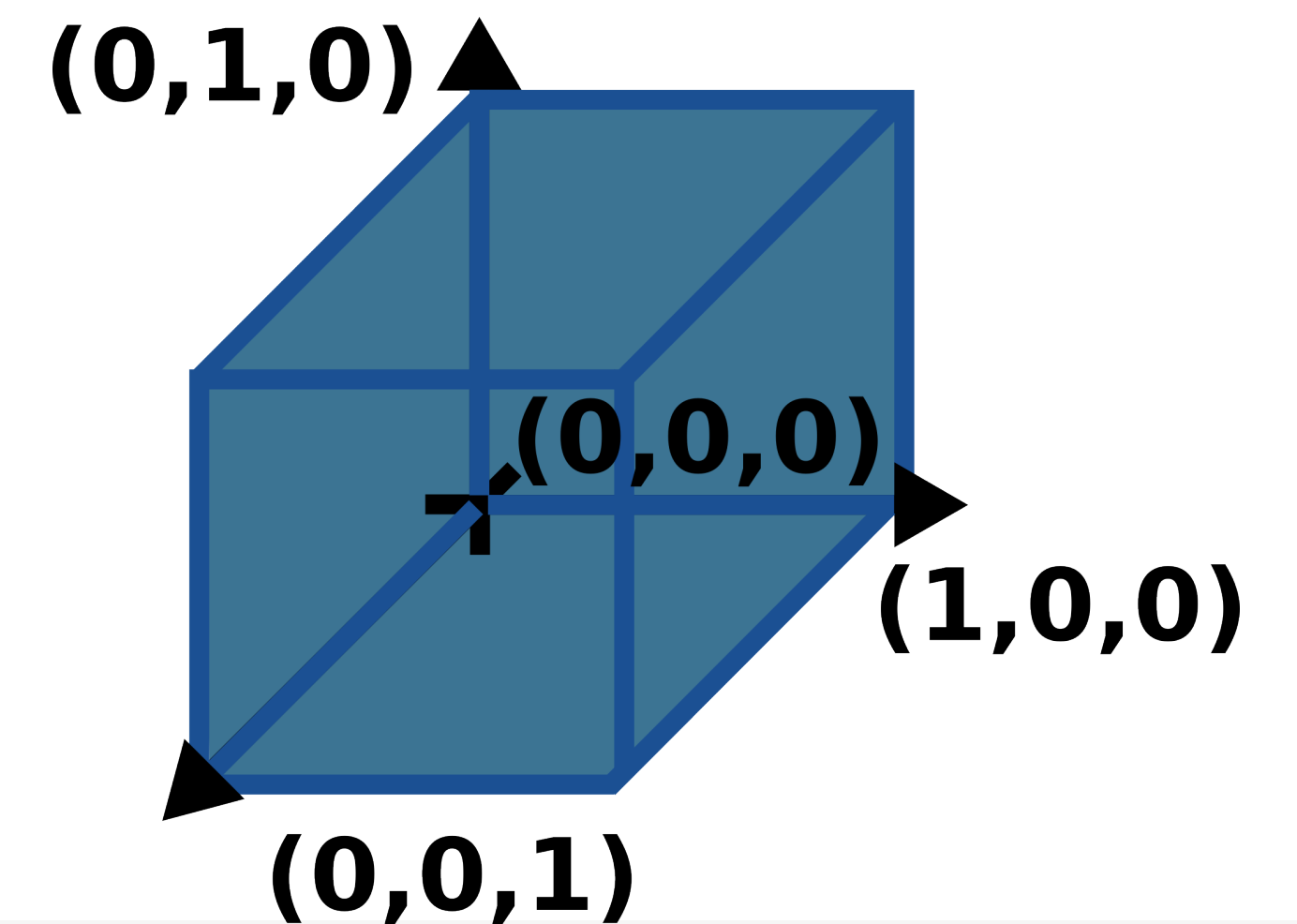
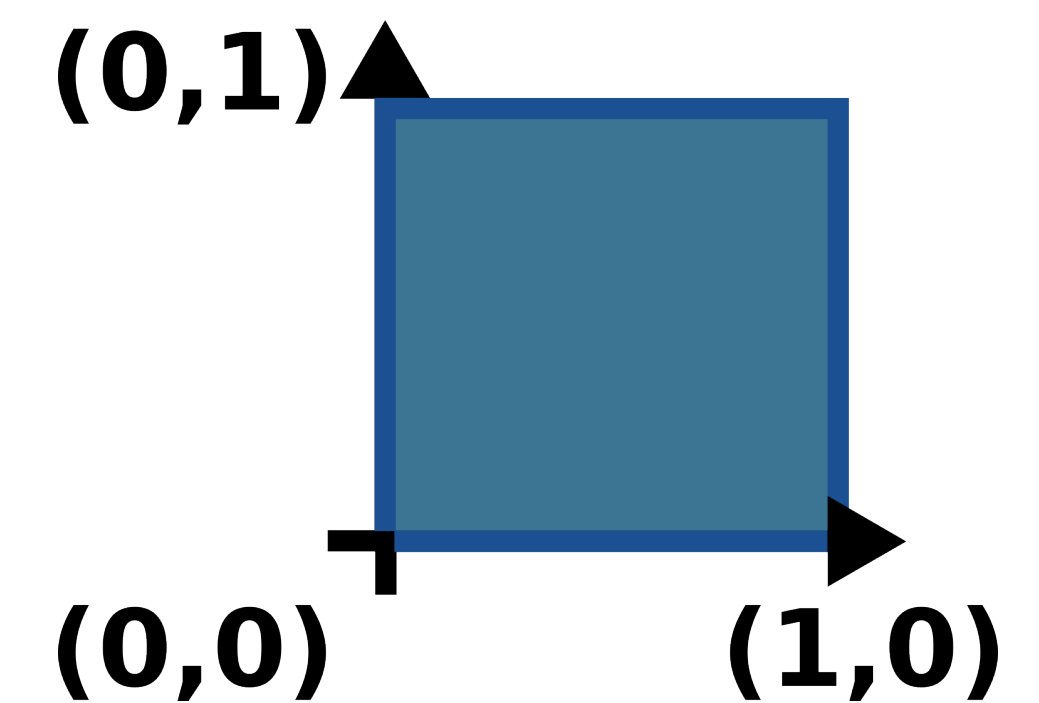
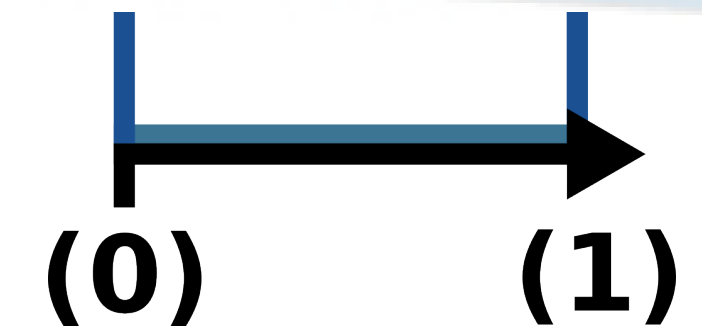
Euclidean spaces on a computer

- Given a unit cell of \mathbb{R}^n
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- Arbitrary closed and bounded subsets of \mathbb{R}^n



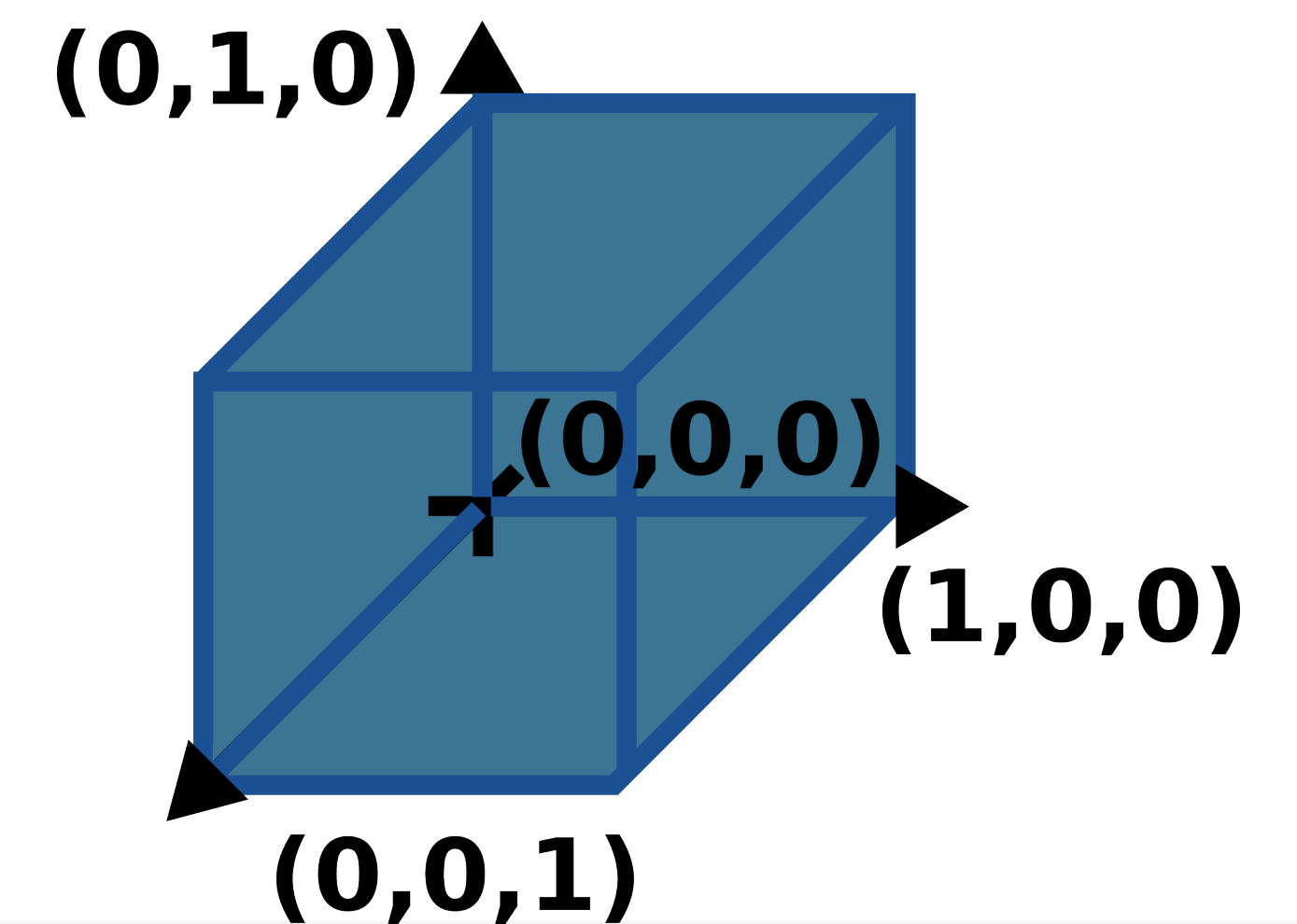
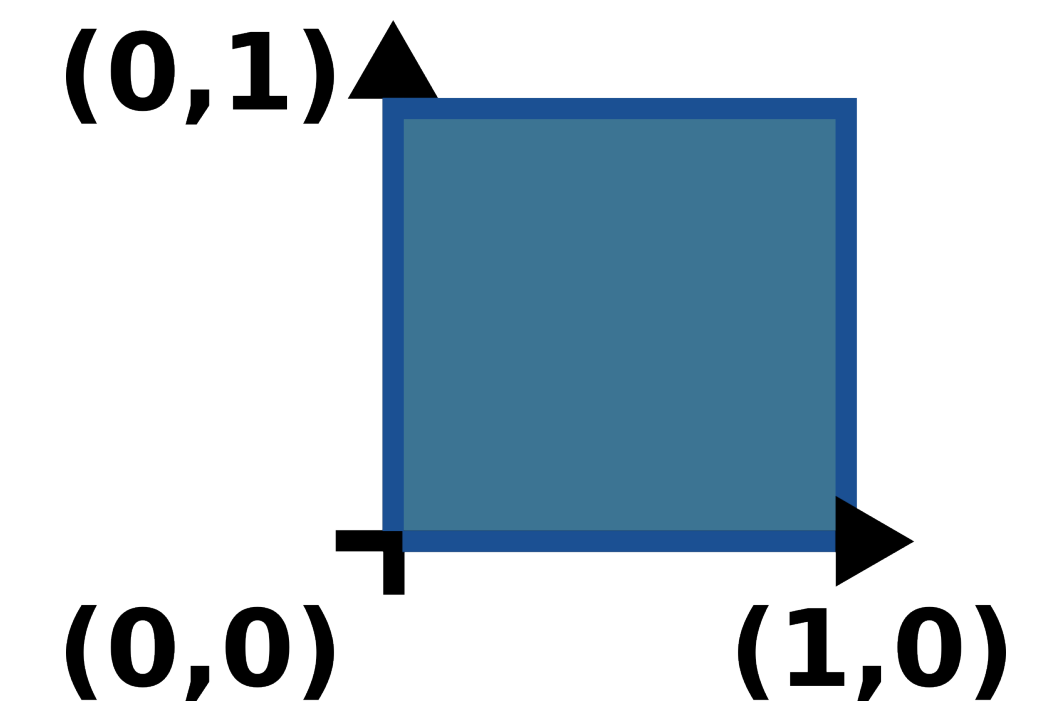
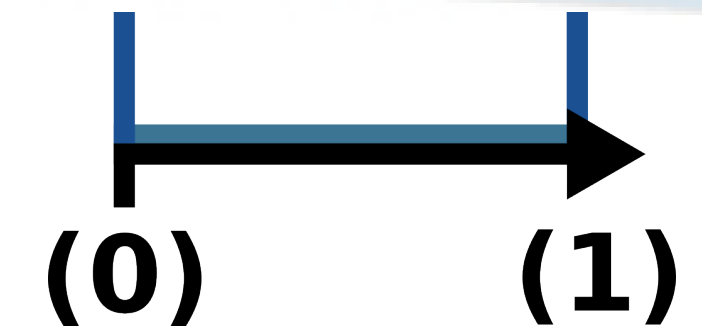
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 - 2^n
- Arbitrary closed and bounded subsets of \mathbb{R}^n
 - Collection of unit cells



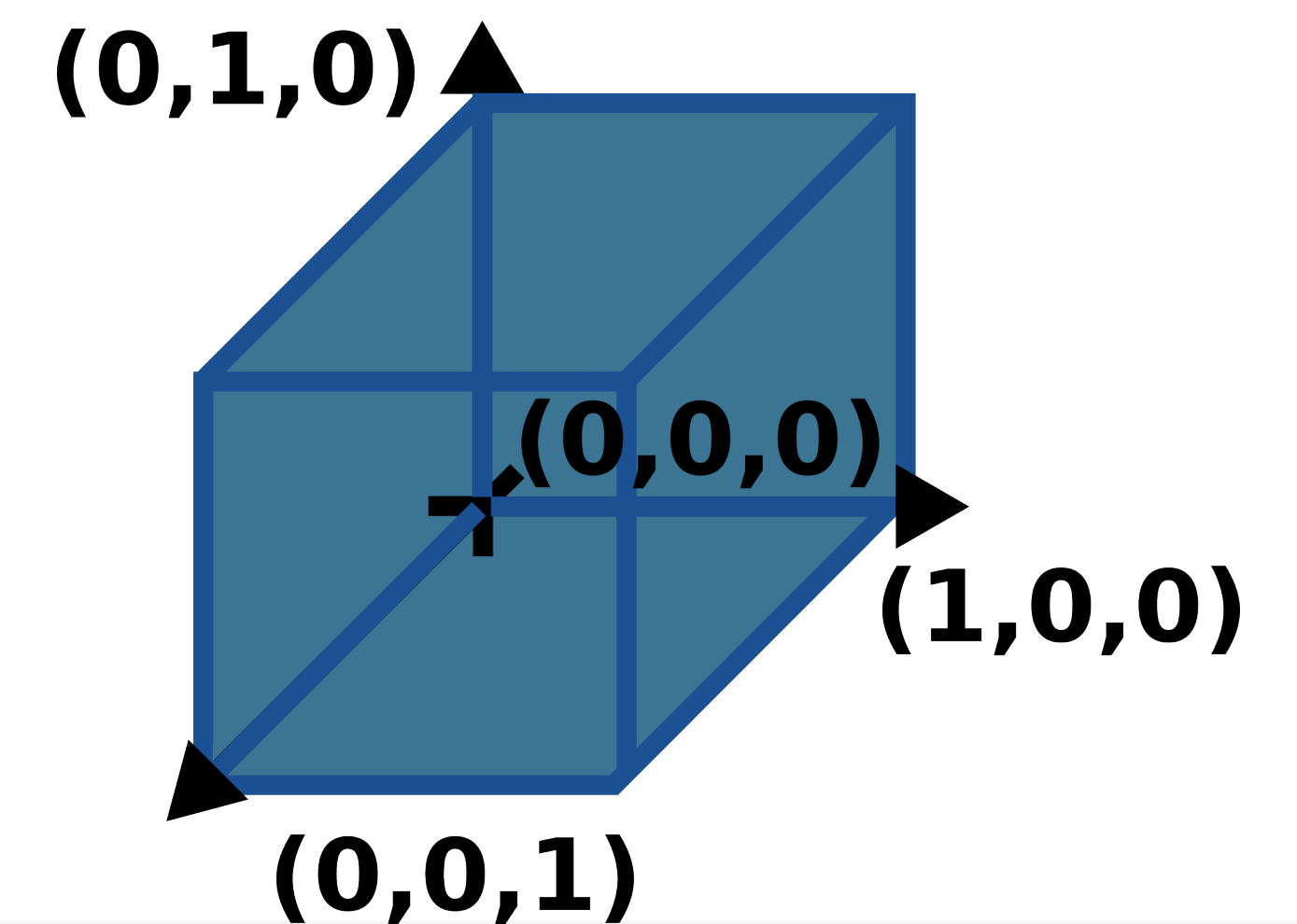
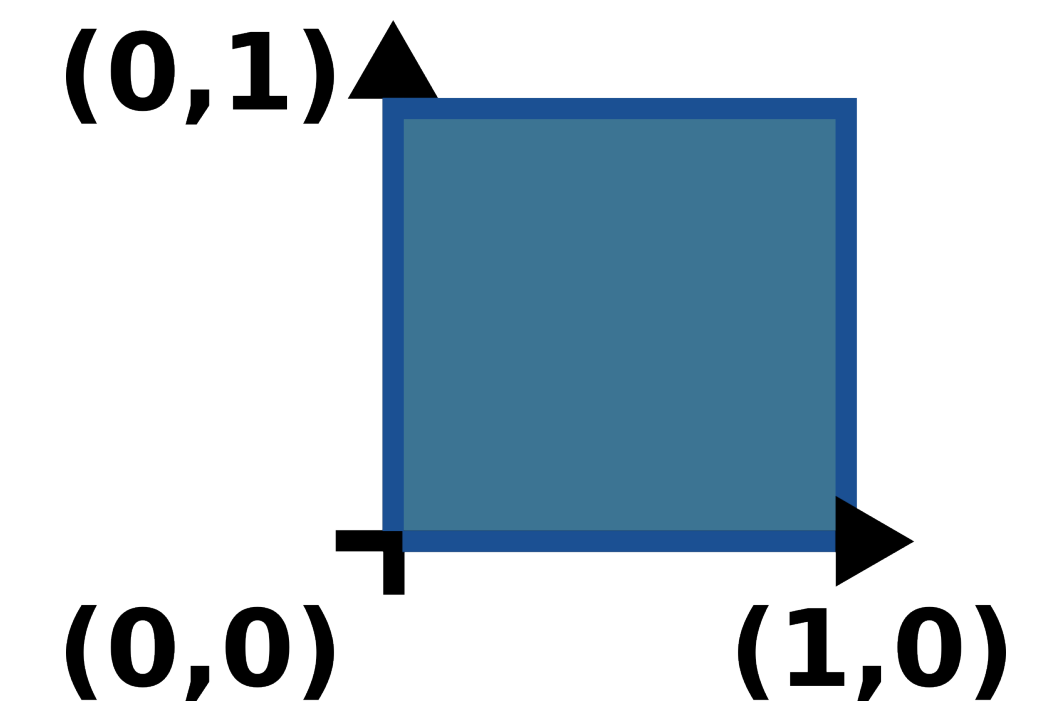
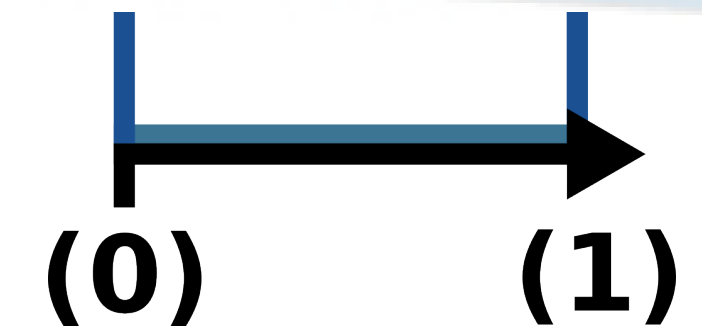
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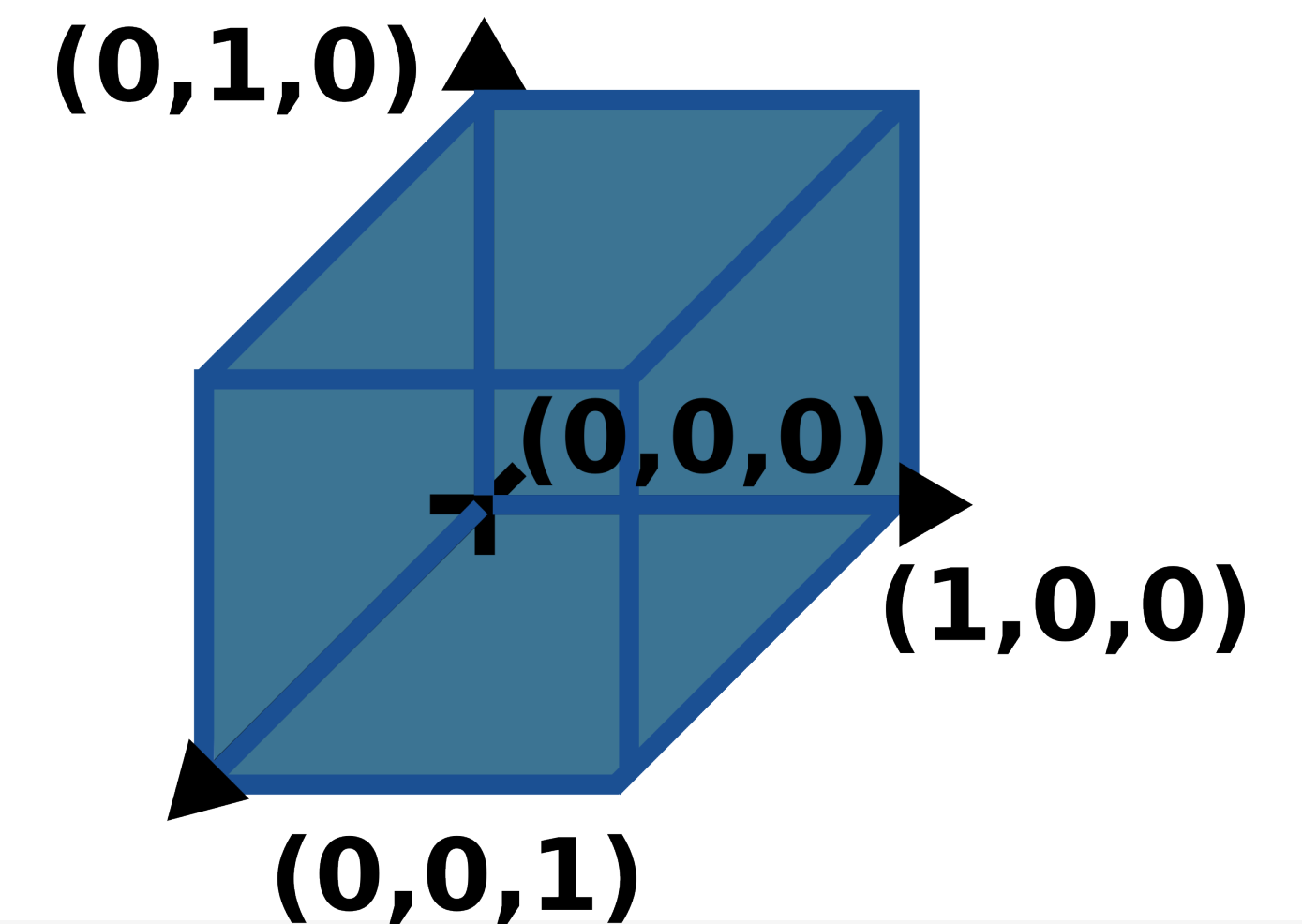
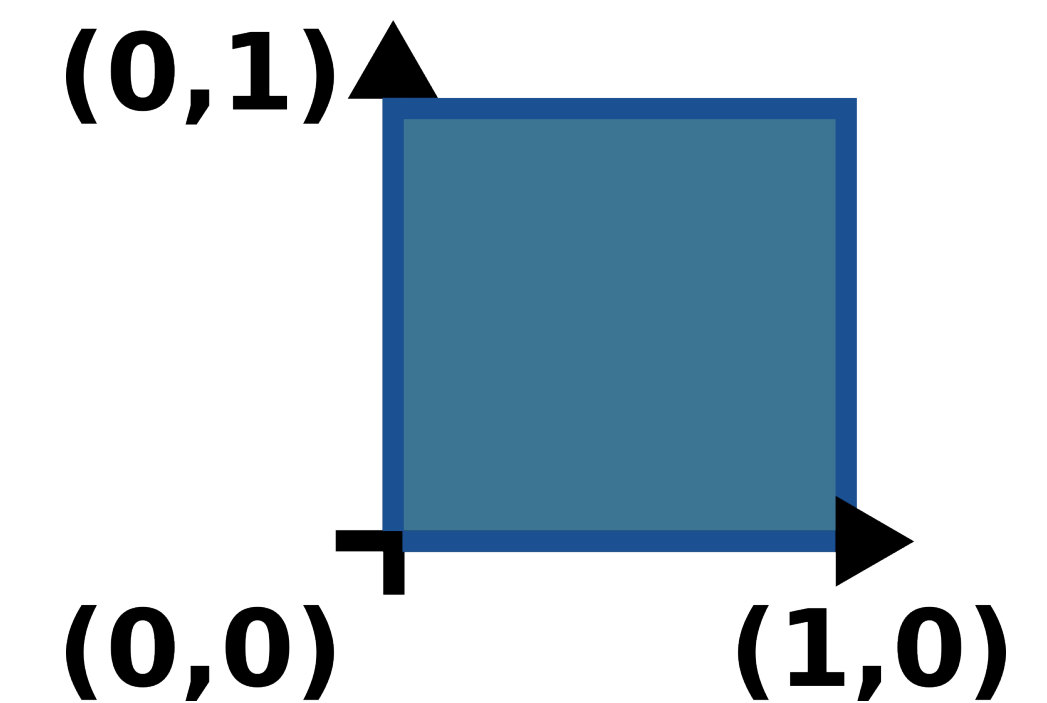
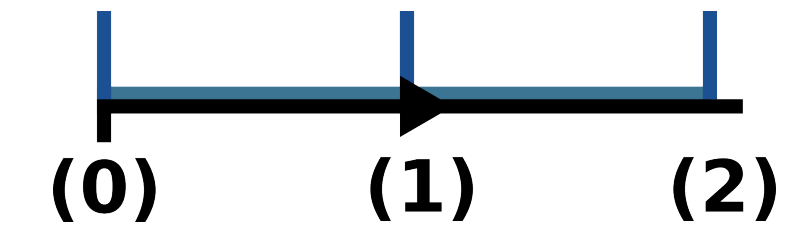
Euclidean spaces on a computer

- Notion of **regular grid** \mathcal{G}
 - Finite collection of unit cells
 - Entirely covering the direct product of closed intervals, such that
 - $[0, k_1] \times [0, k_2] \times \cdots \times [0, k_n]$
 - $k_i \in \mathbb{N}$



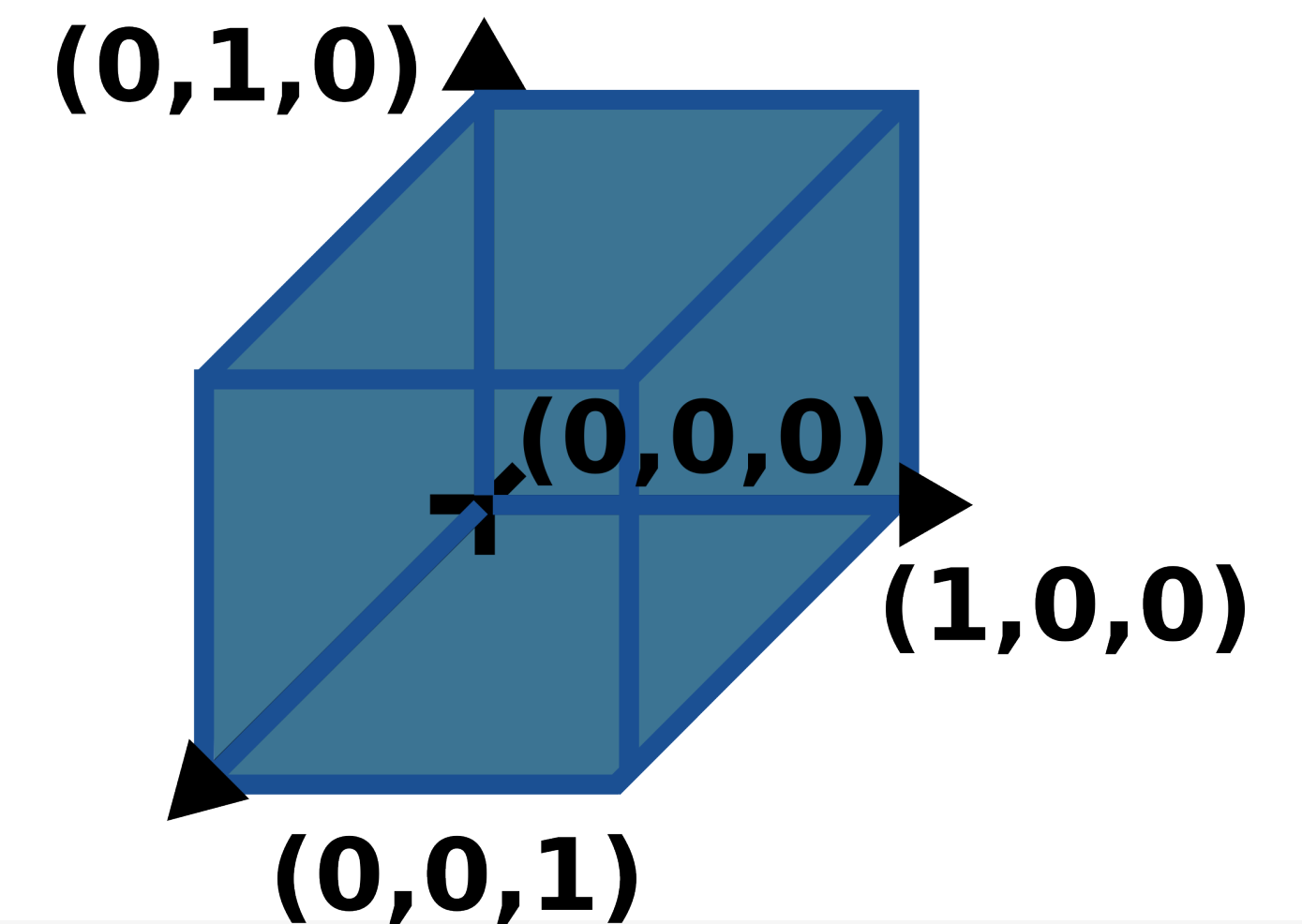
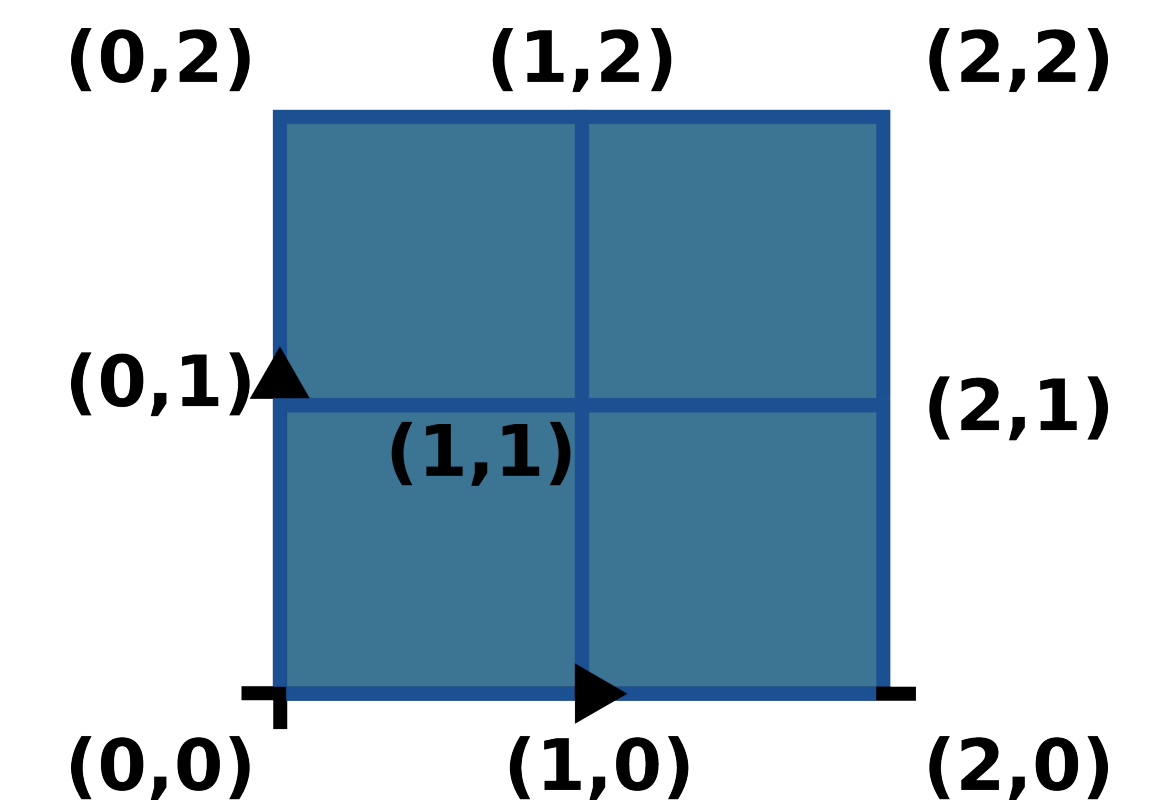
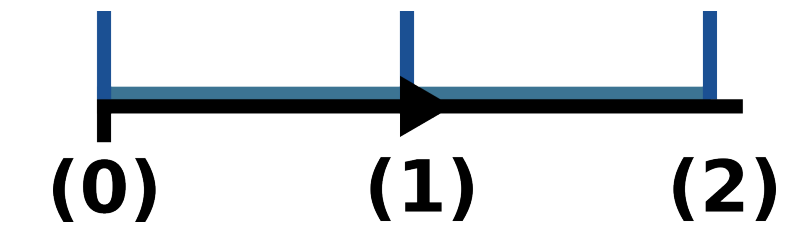
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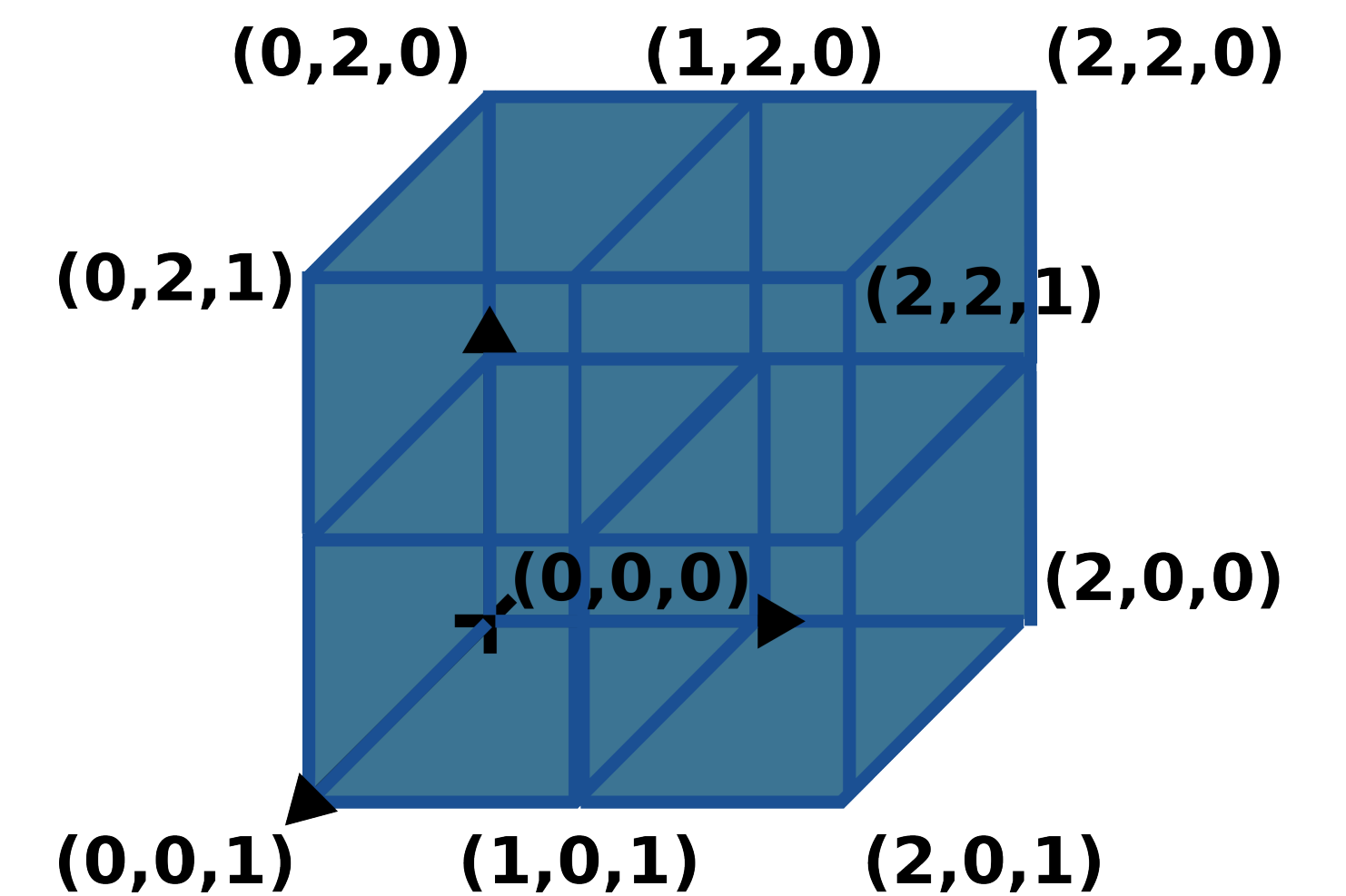
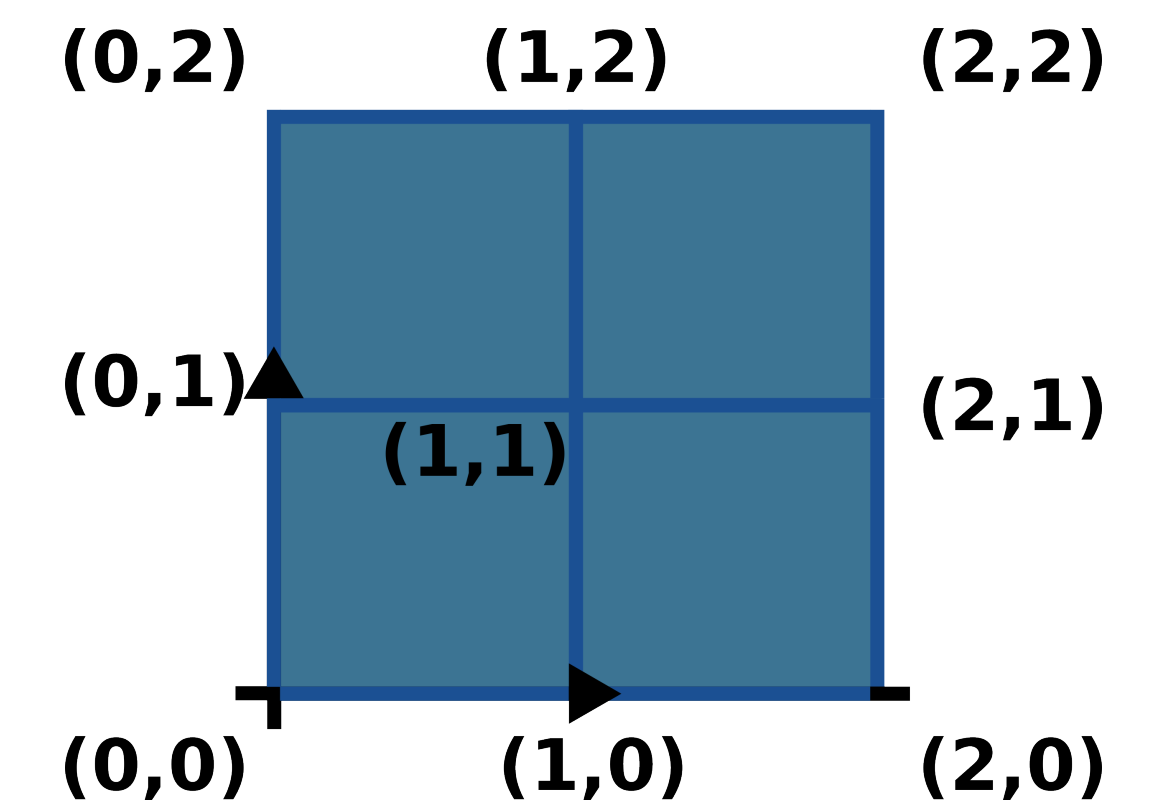
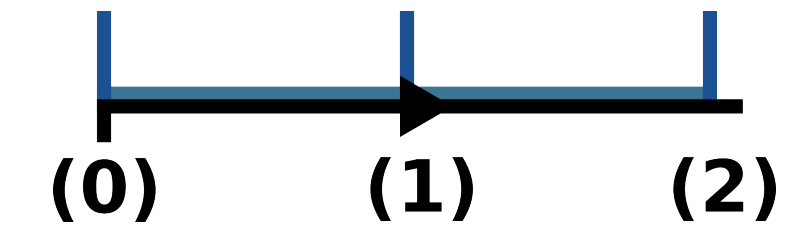
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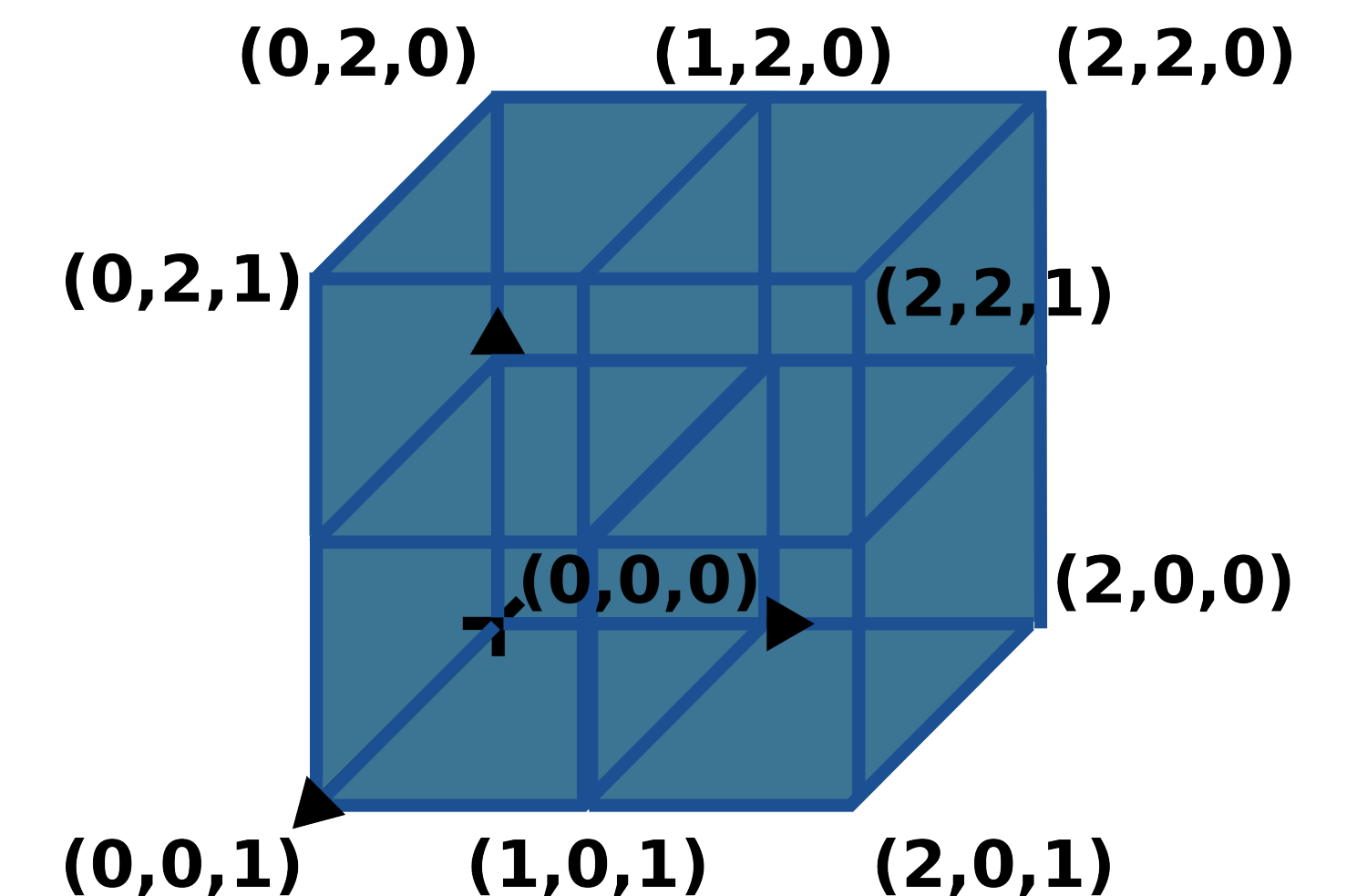
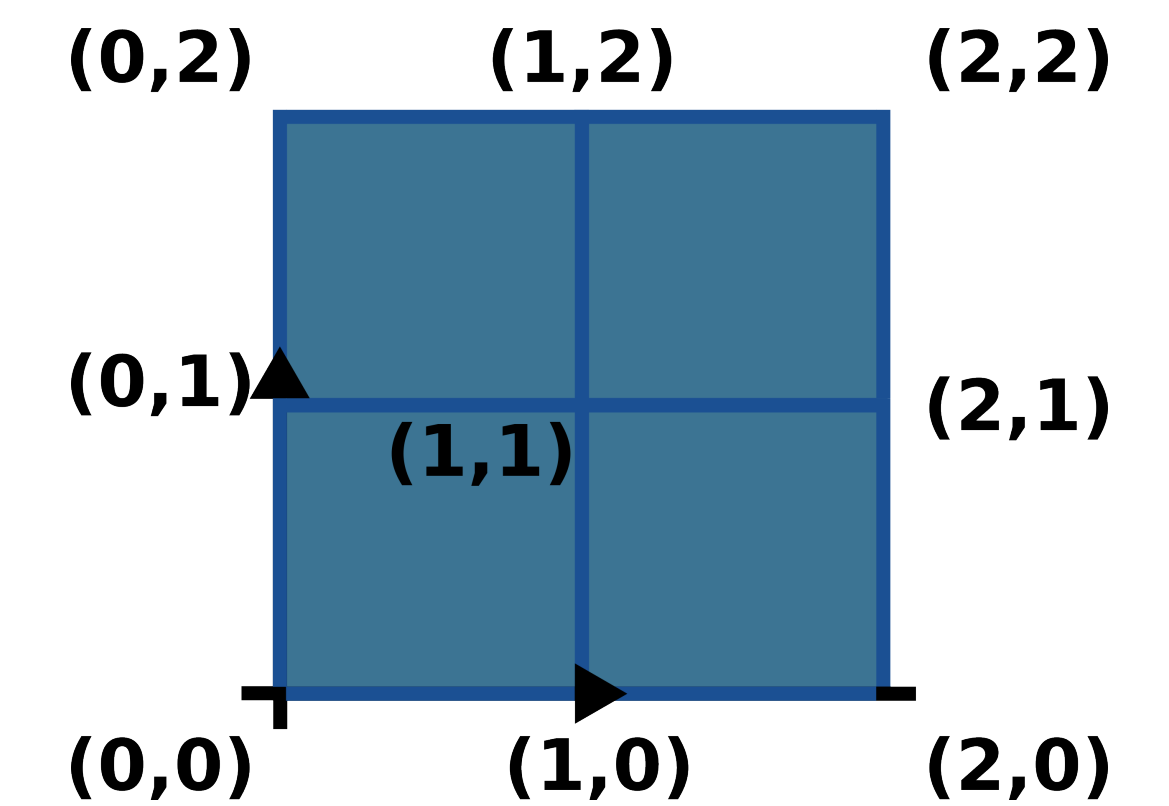
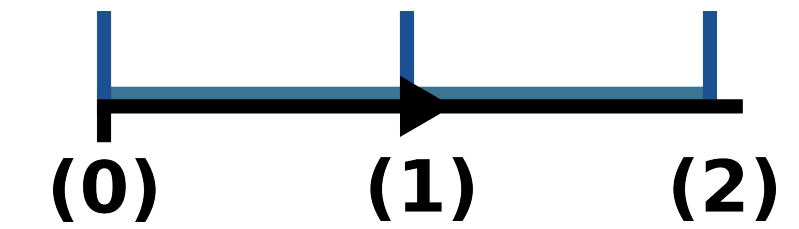
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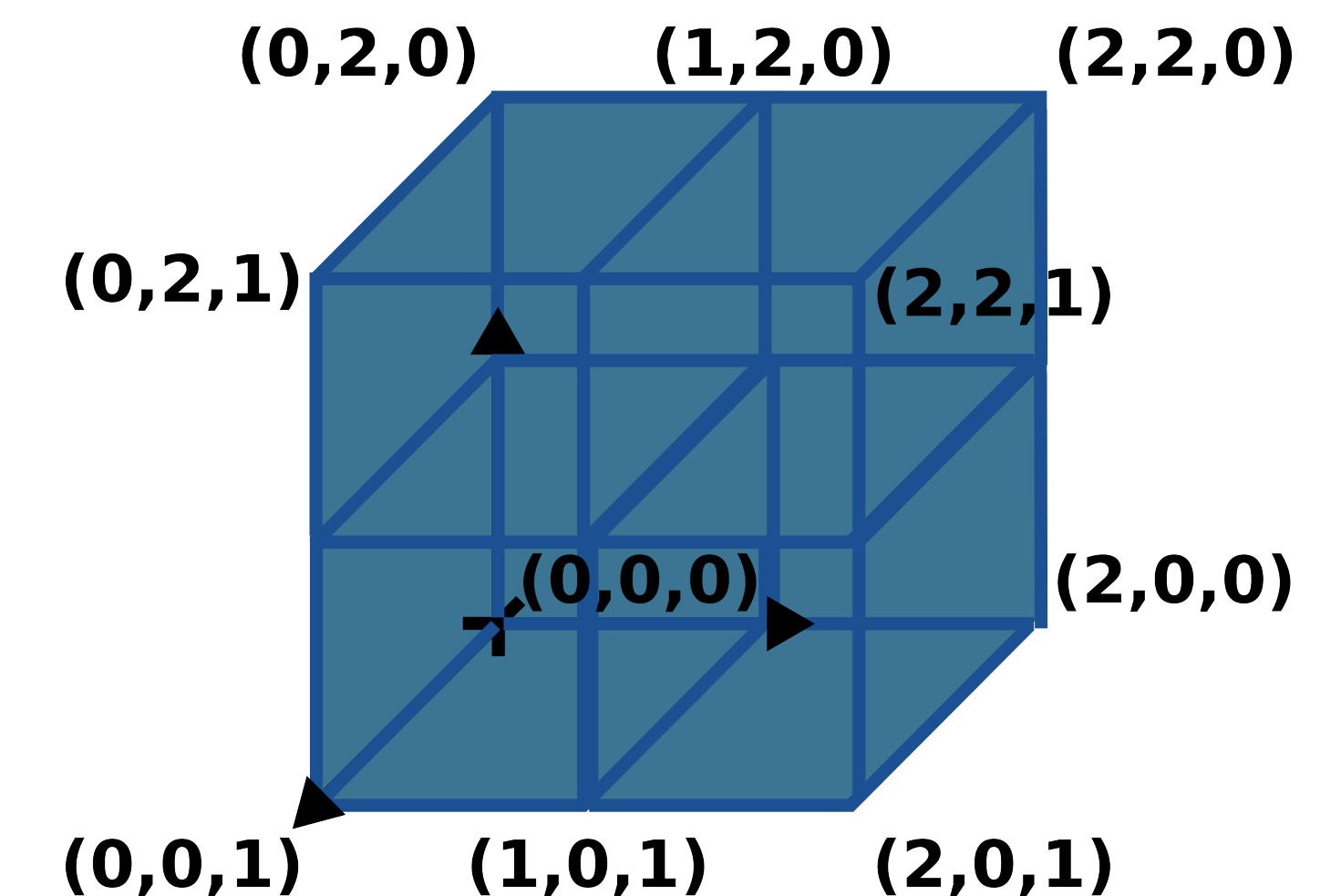
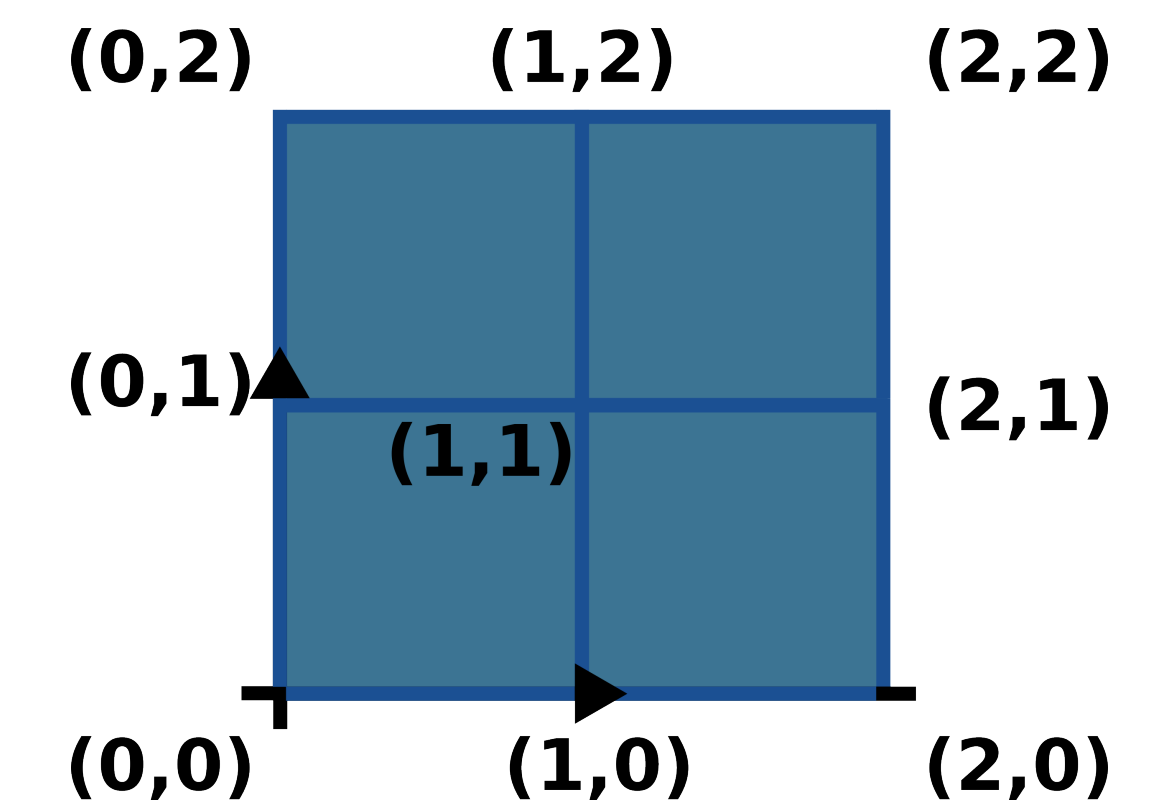
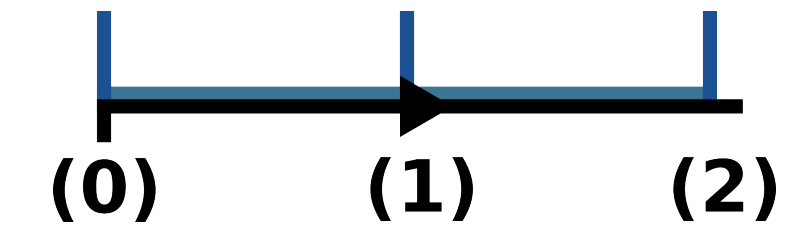
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 - $k_i \in \mathbb{N}$
- How many unit cells in $[0, k_1] \times [0, k_2] \times [0, k_3]$?



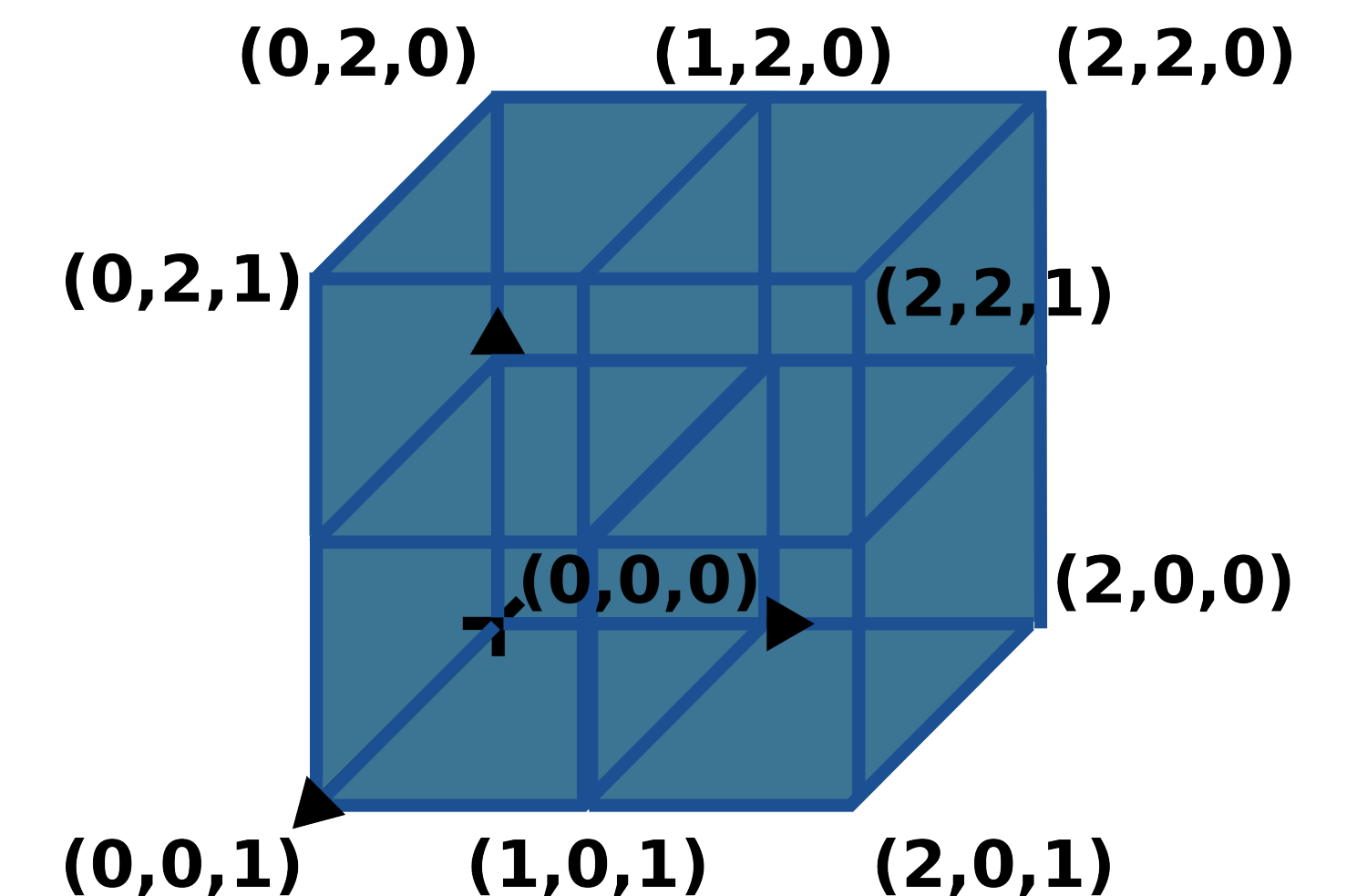
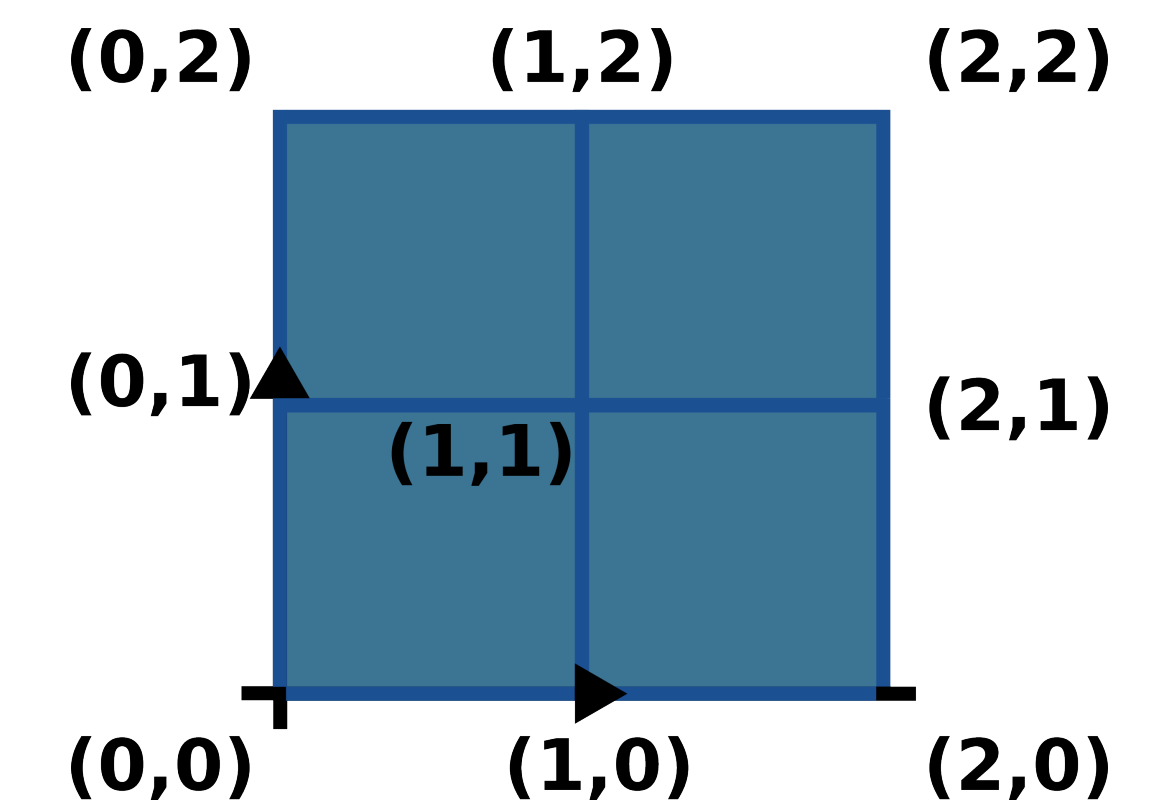
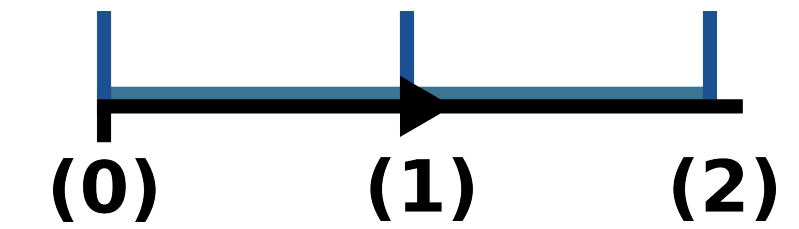
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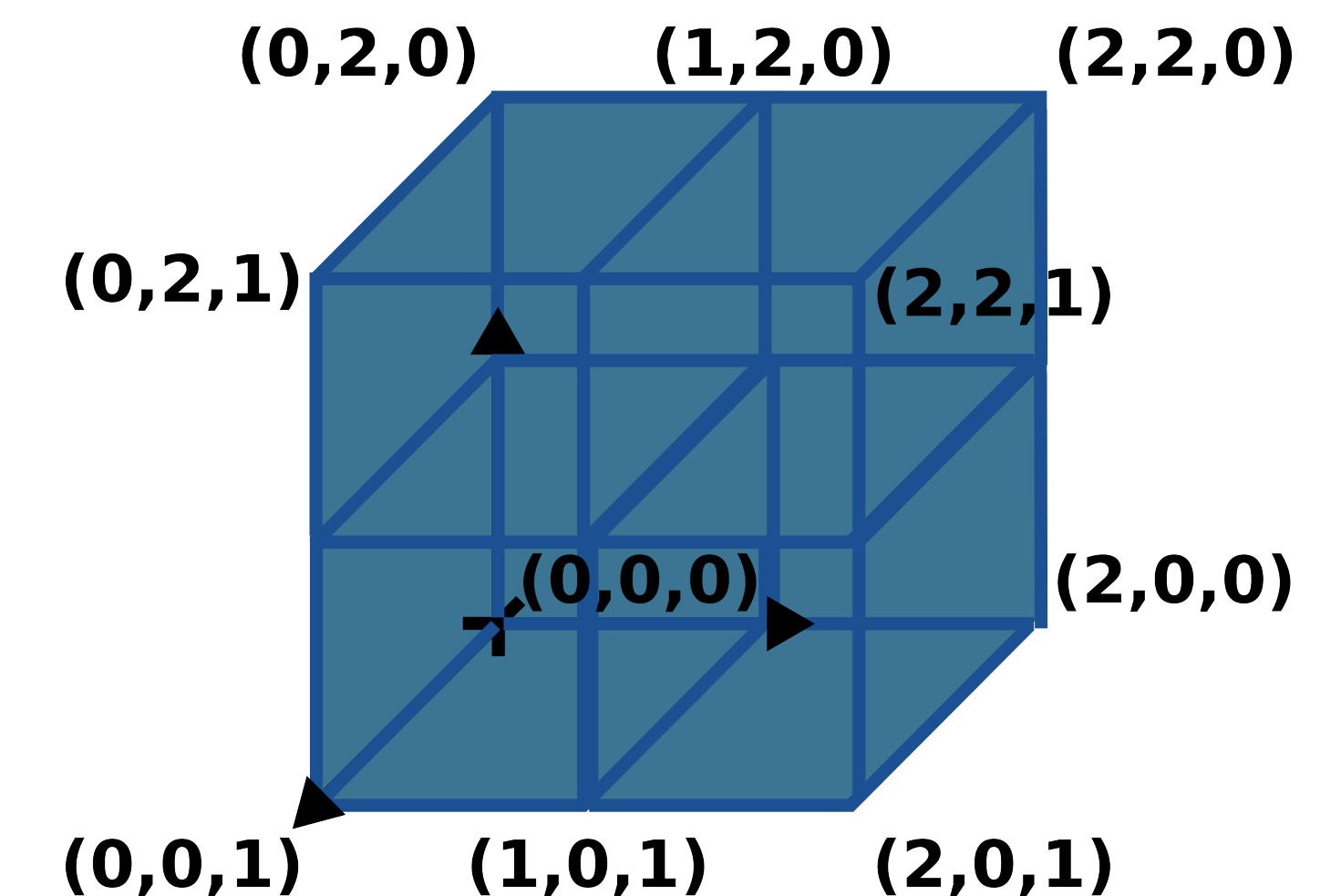
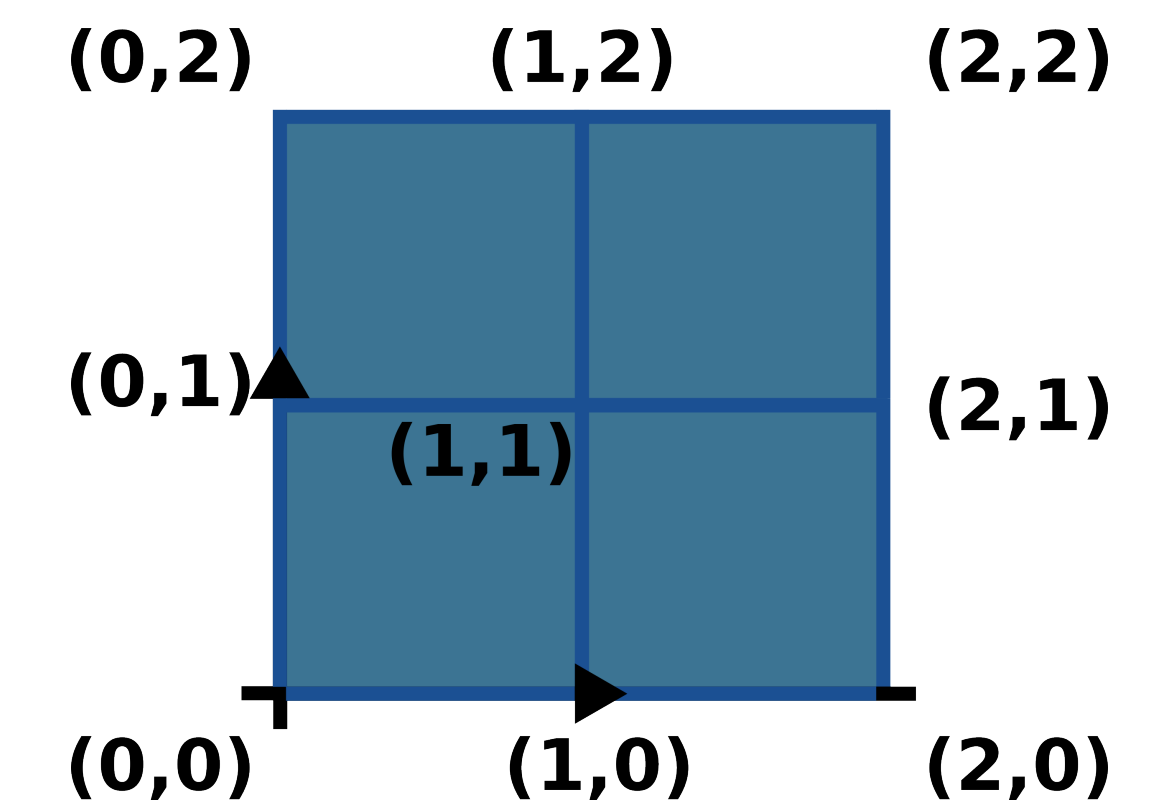
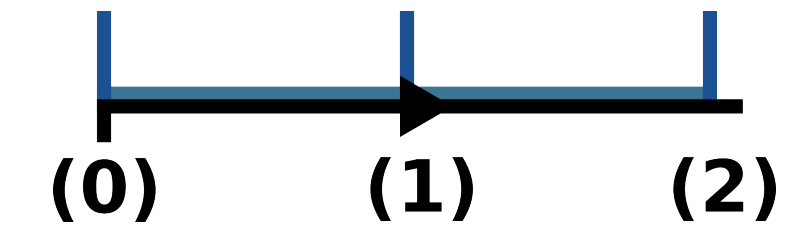
Euclidean spaces on a computer

- Notion of **regular grid** \mathcal{G}
 - Finite collection of unit cells
 - Entirely covering the direct product of closed intervals, such that
 - $[0, k_1] \times [0, k_2] \times \cdots \times [0, k_n]$
 - $k_i \in \mathbb{N}$
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 - $k_1 \cdot k_2 \cdot k_3$
- How many samples?



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 - $k_1 \cdot k_2 \cdot k_3$
- How many samples? $(k_1 + 1) \cdot (k_2 + 1) \cdot (k_3 + 1)$



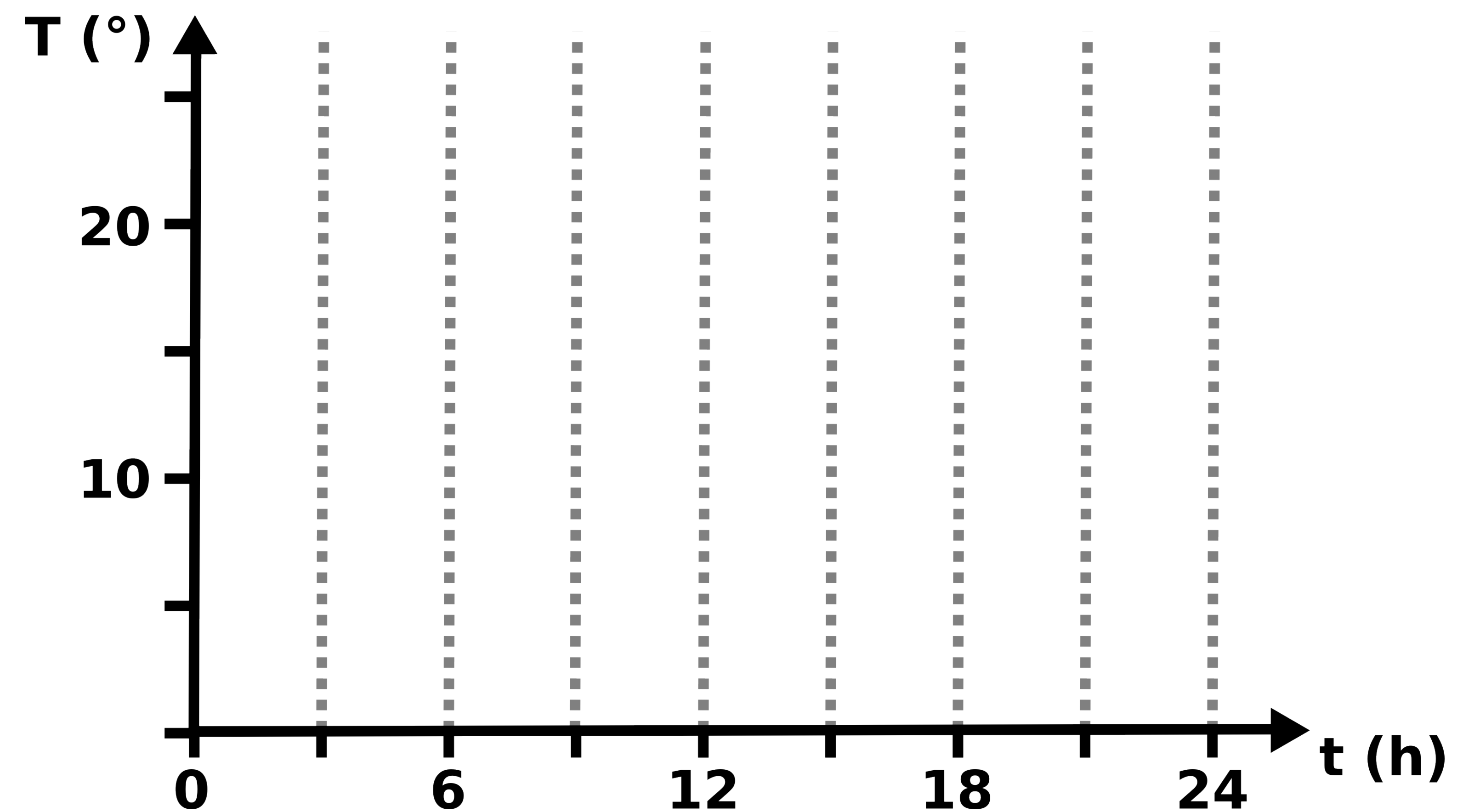
Regular grids of \mathbb{R}

Regular grids of \mathbb{R}

$$f : [0, 24] \rightarrow \mathbb{R}$$

Regular grids of \mathbb{R}

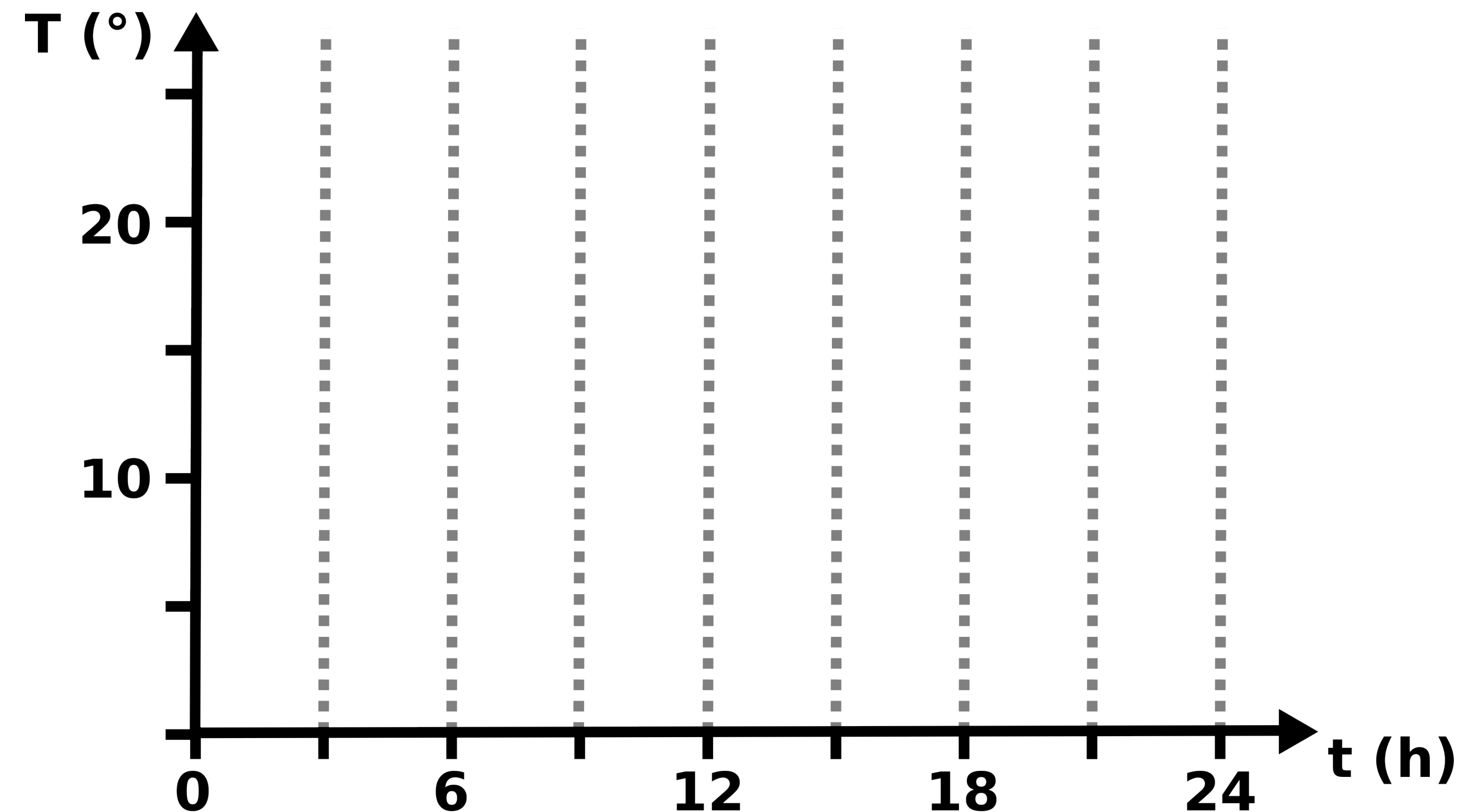
$$f : [0, 24] \rightarrow \mathbb{R}$$



Regular grids of \mathbb{R}

- We need
 - A one-dimensional structure

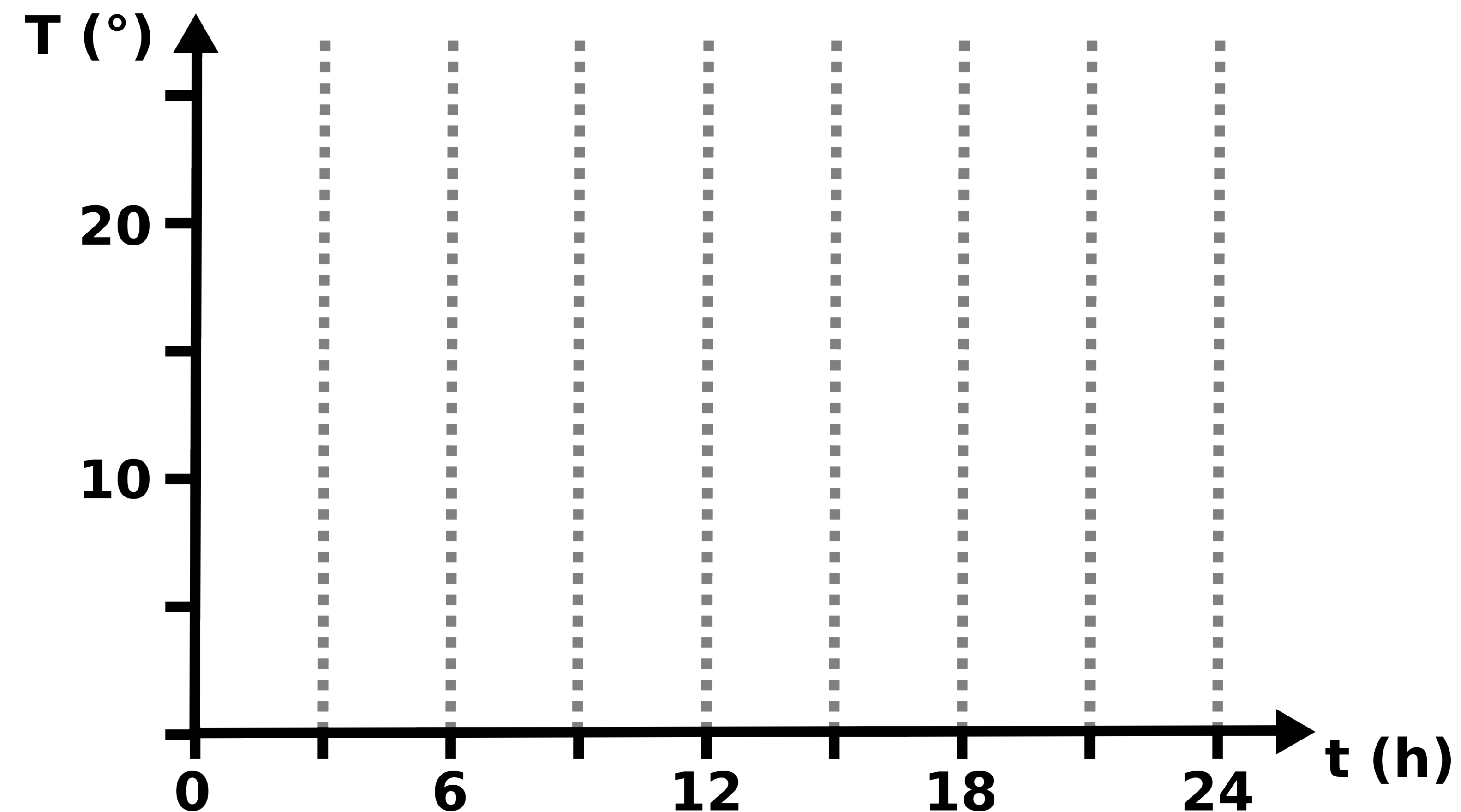
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Regular grids of \mathbb{R}

- We need
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 - To store values
 - Scalars, vectors, tensors...

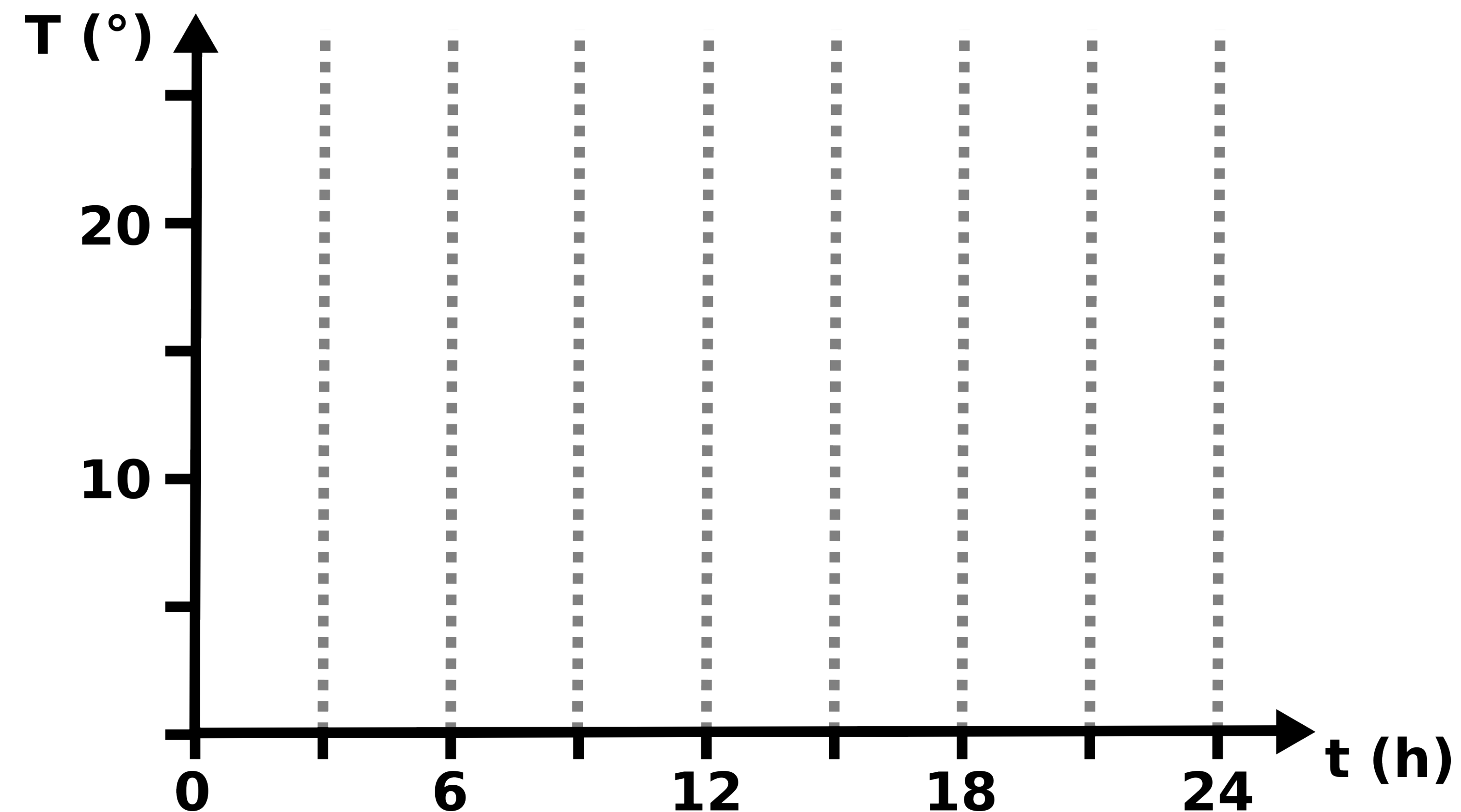
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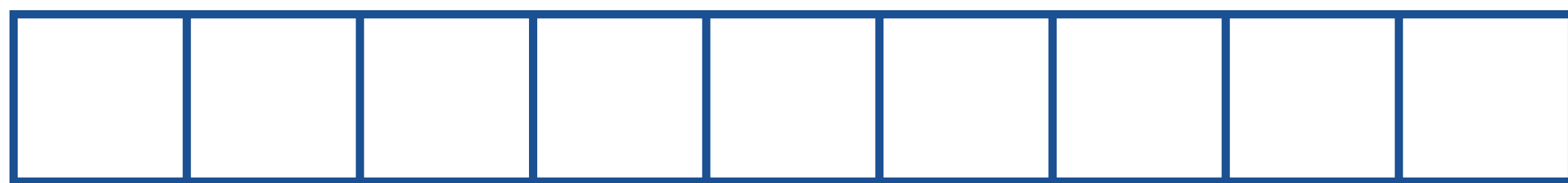
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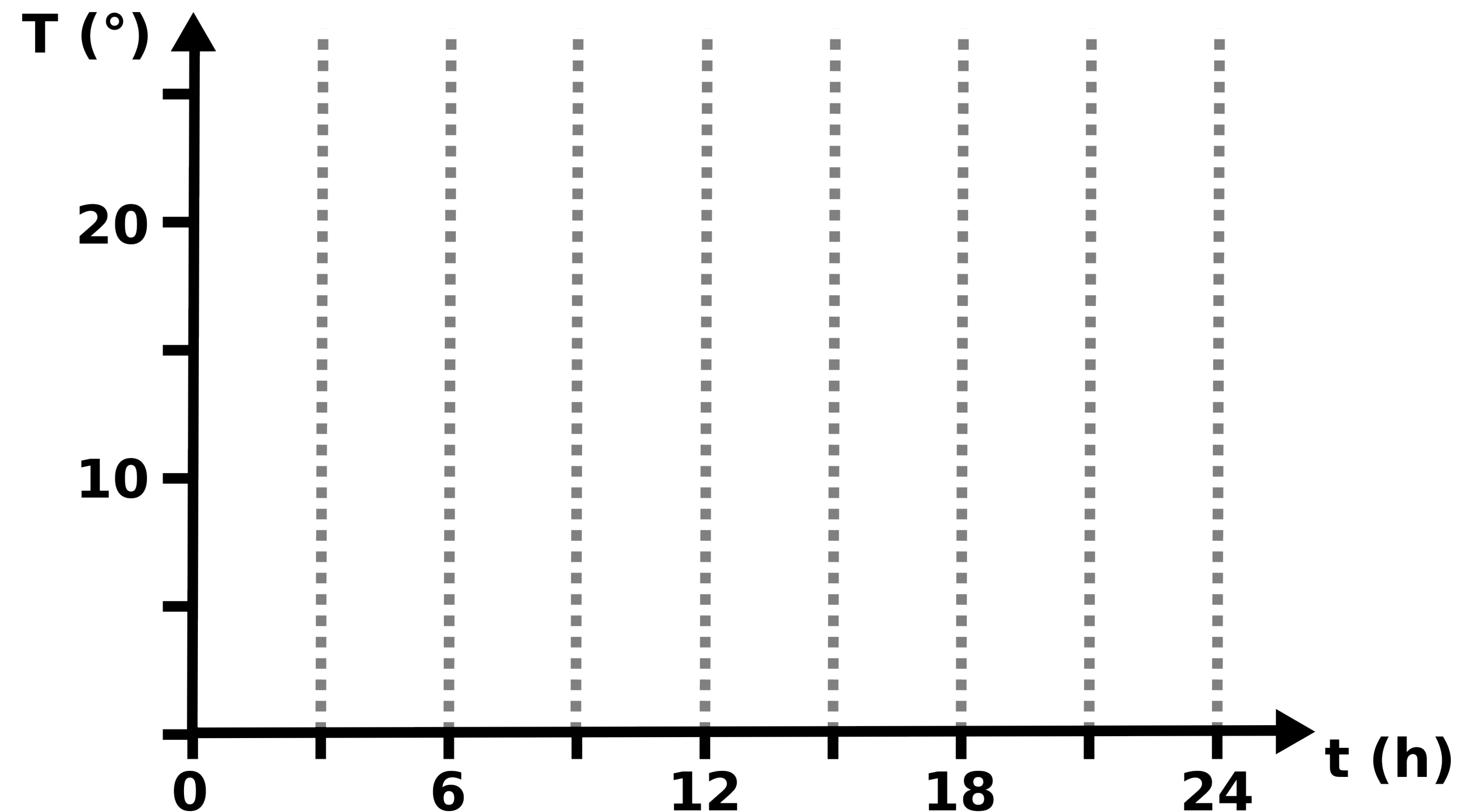


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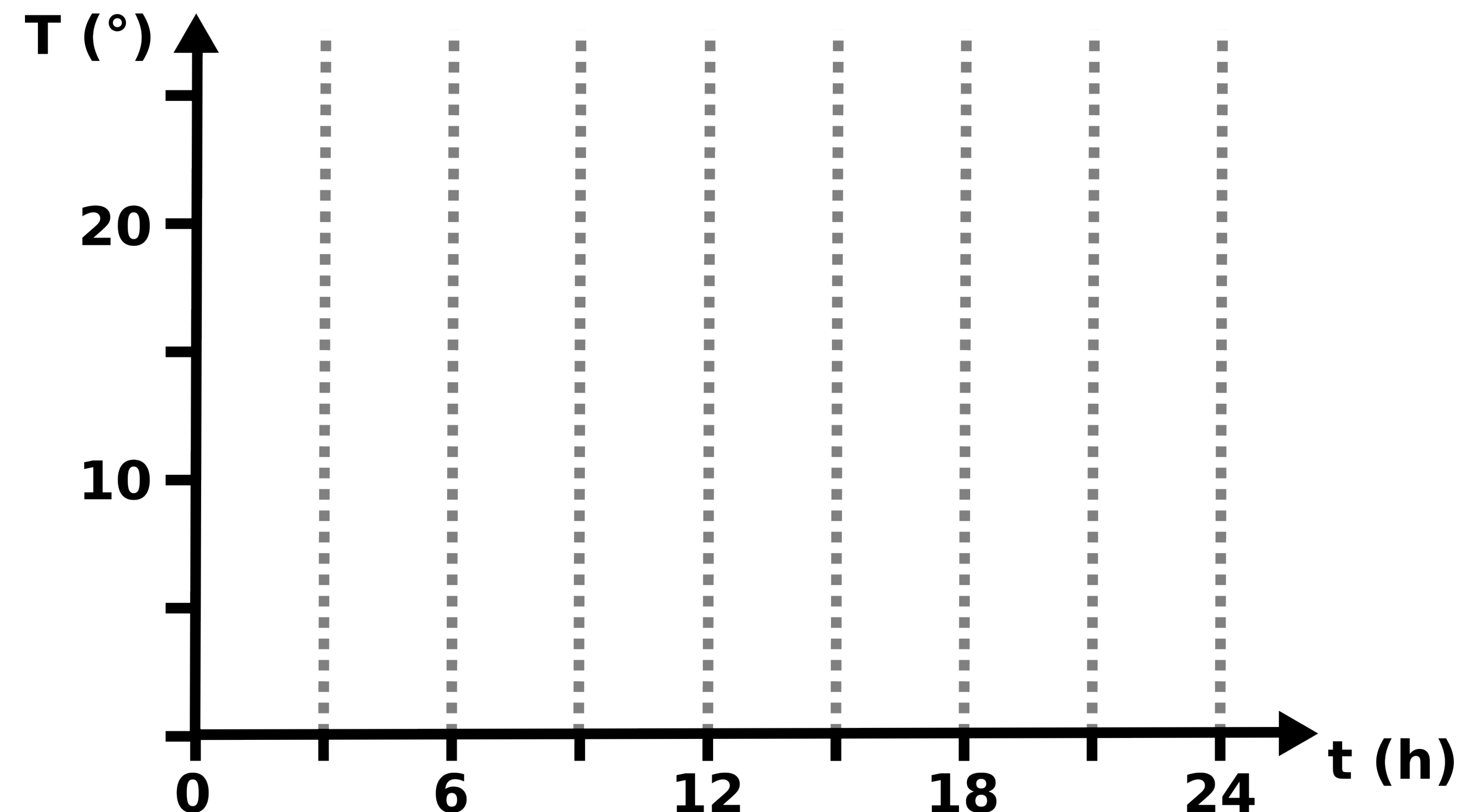


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9	7	10	12	16	18	11	10	7
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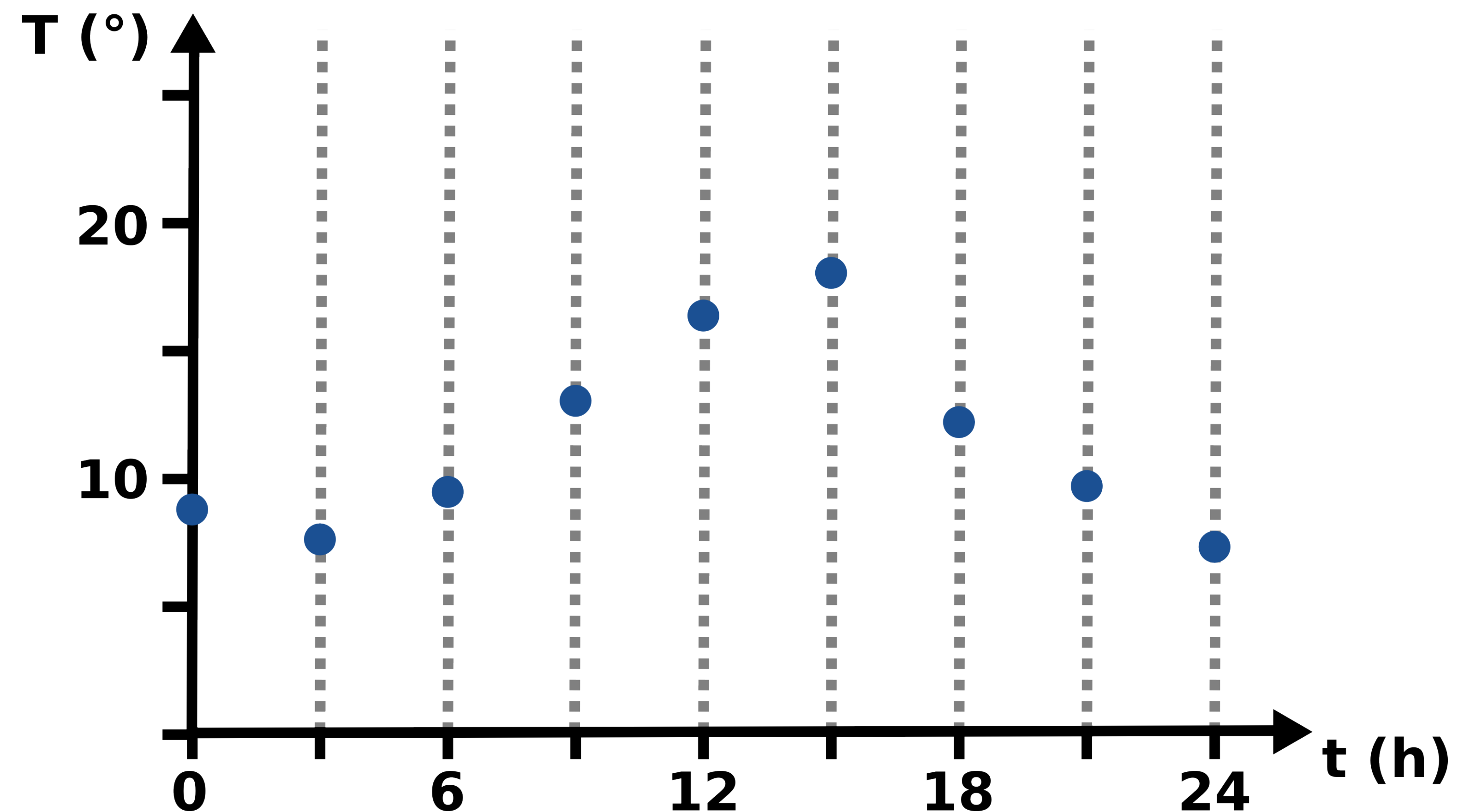


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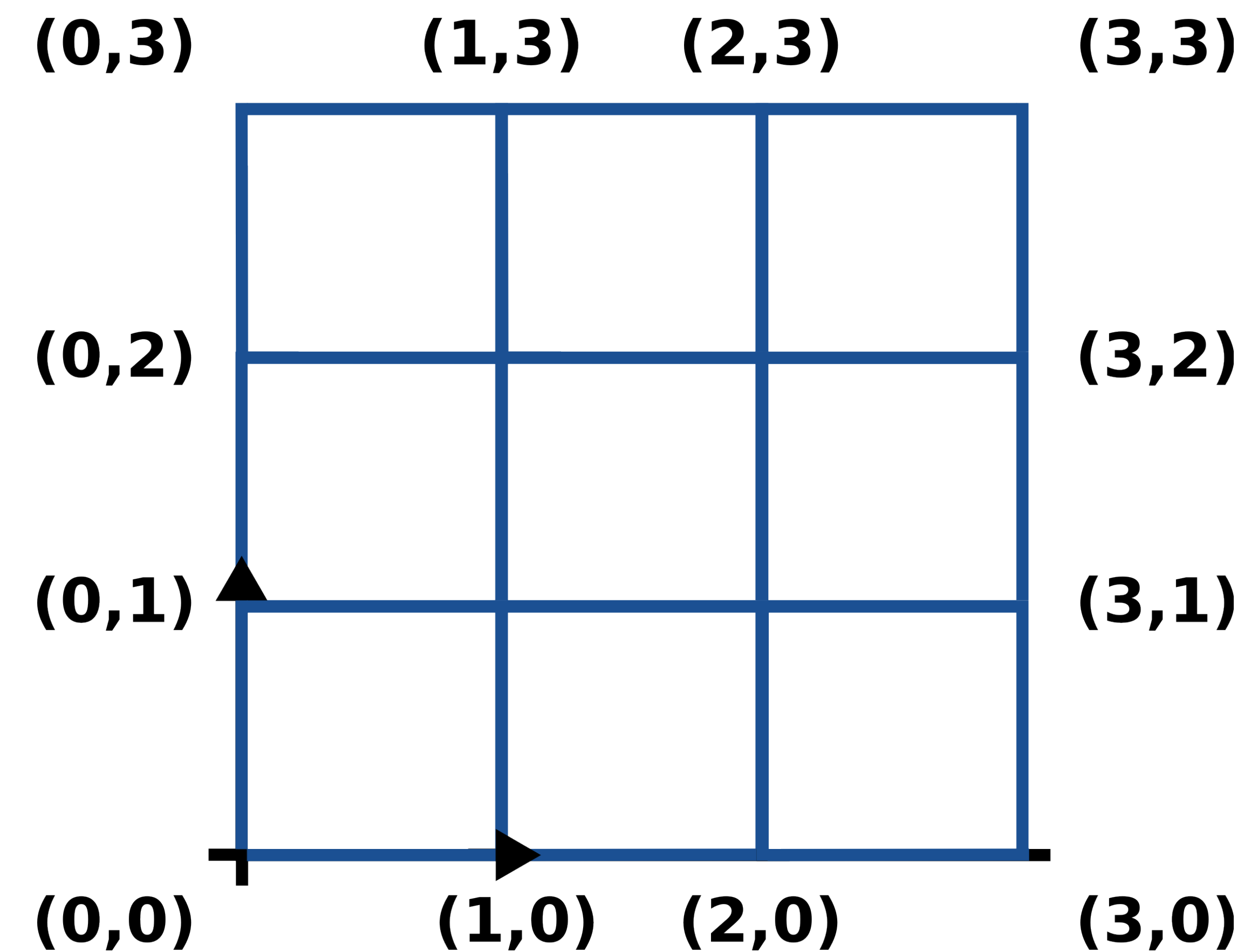
Regular grids of \mathbb{R}^2

Regular grids of \mathbb{R}^2

$$[0, 3] \times [0, 3]$$

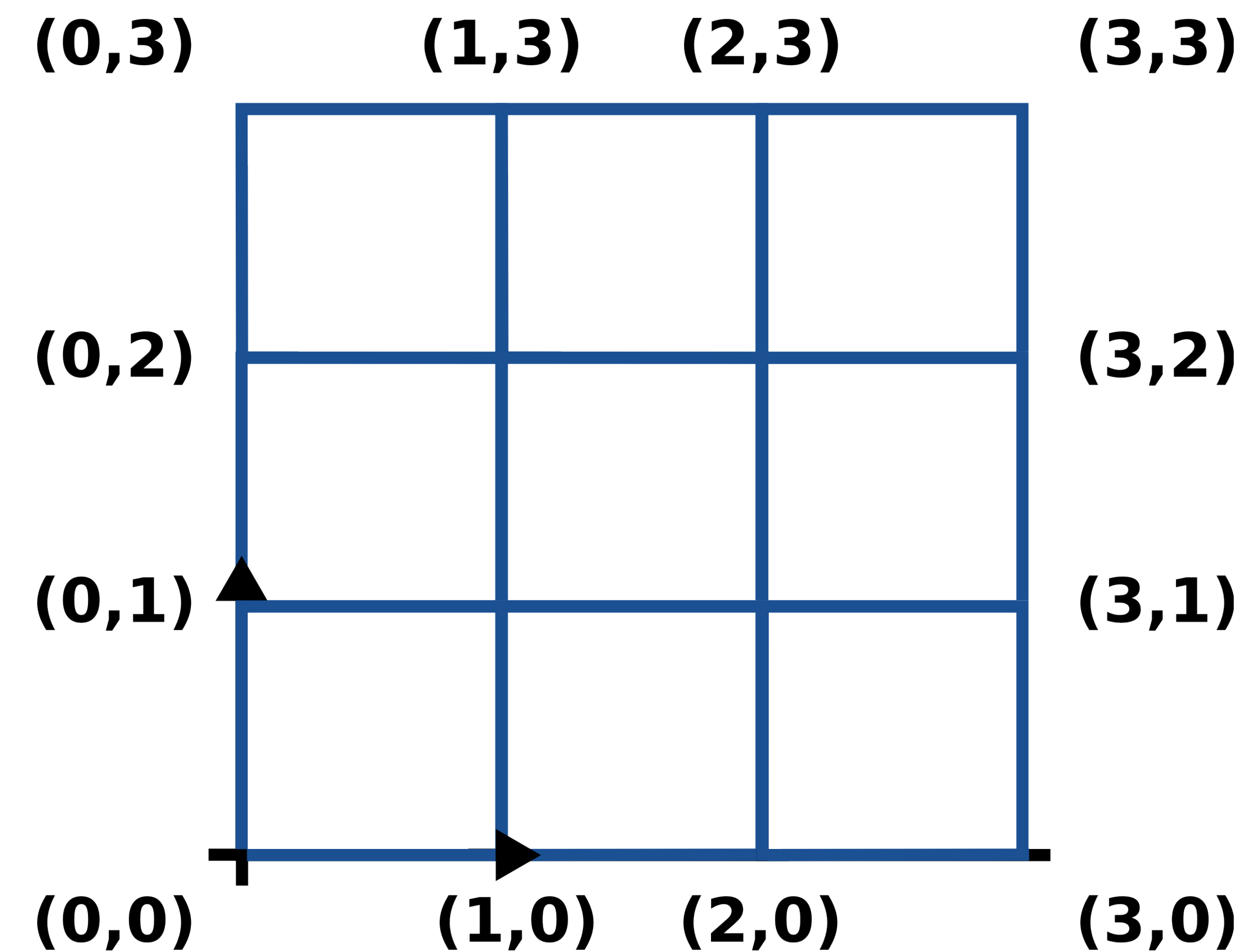
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Regular grids of \mathbb{R}^2

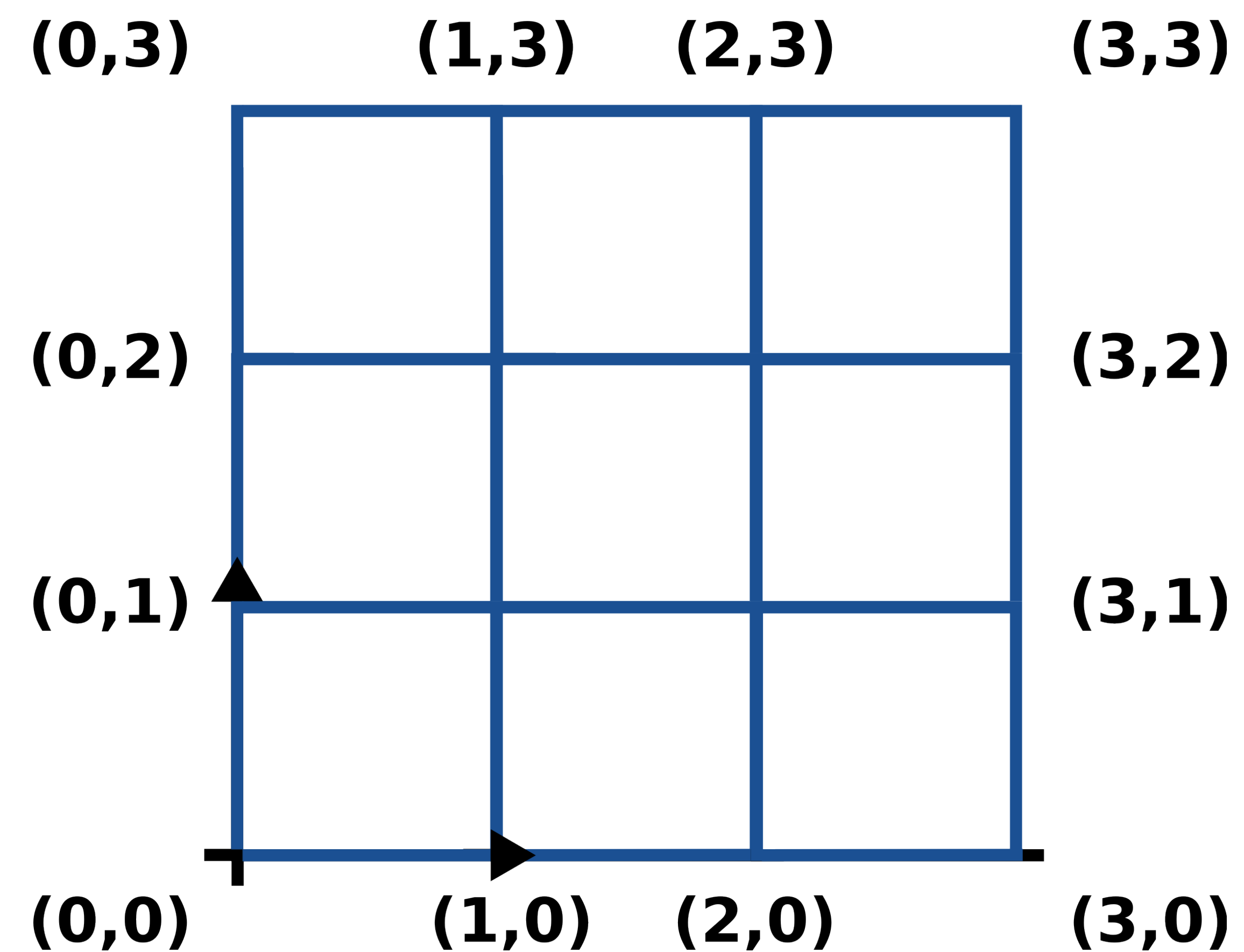
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Regular grids of \mathbb{R}^2

$$f(p) = \sqrt{p_x^2 + p_y^2}$$

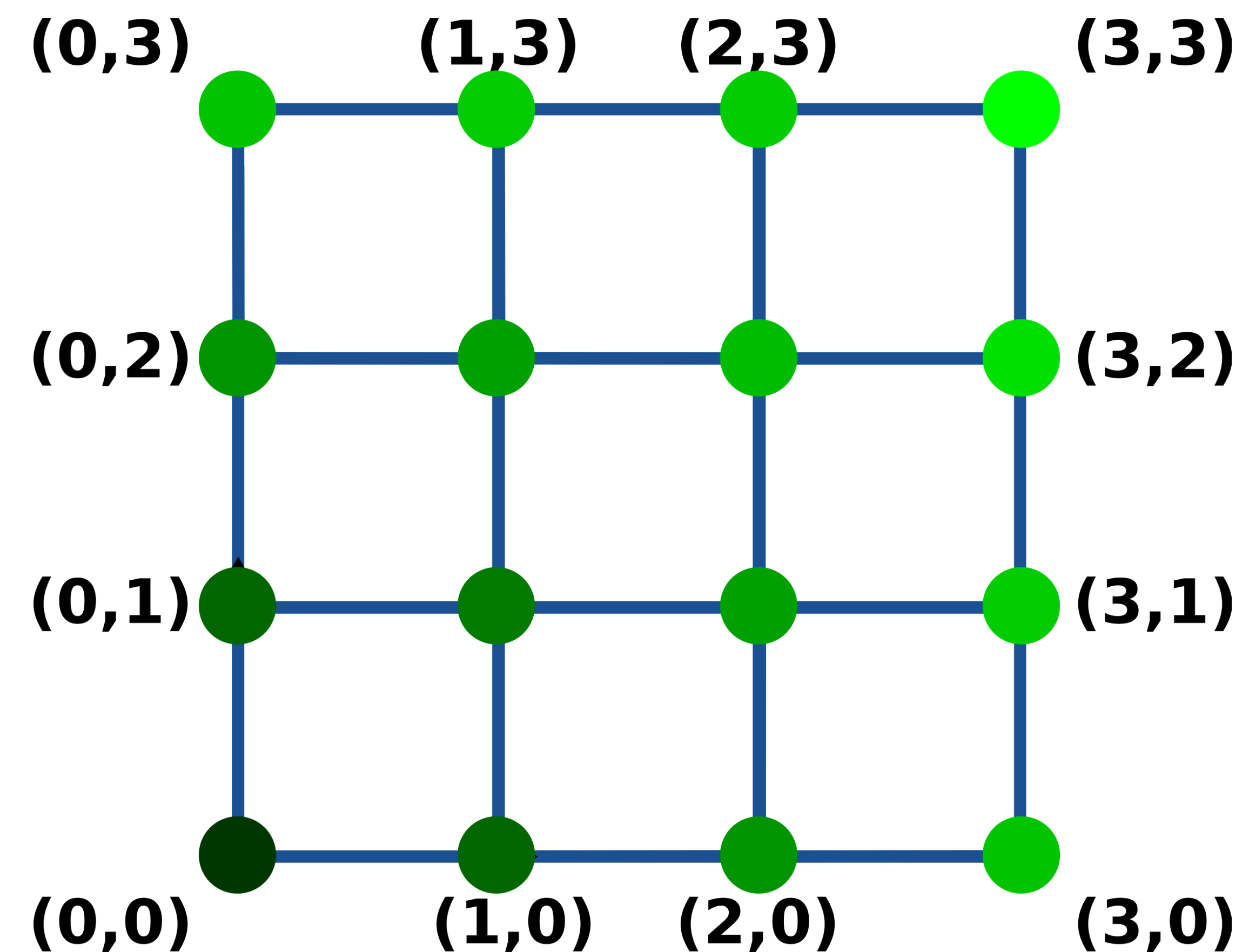
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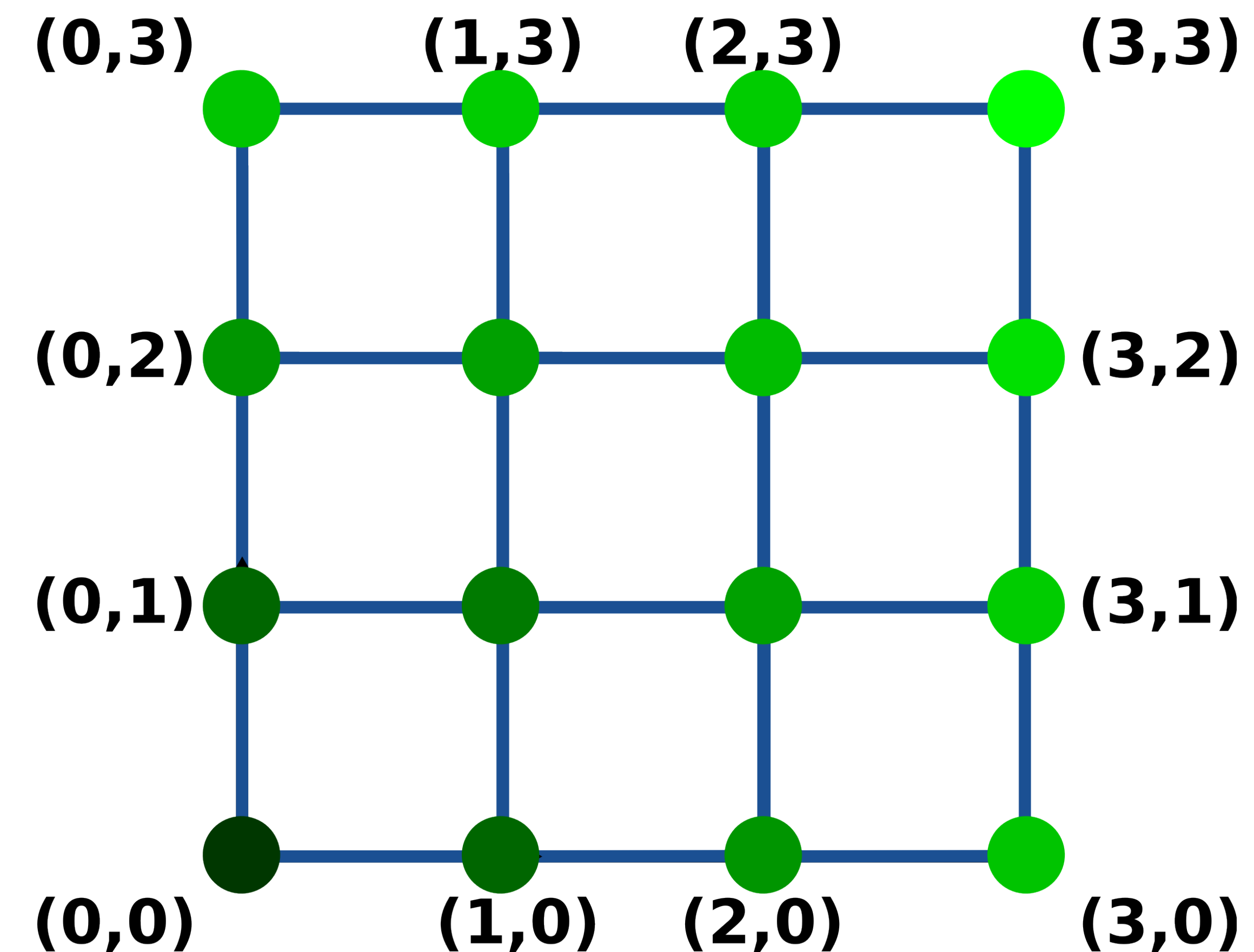


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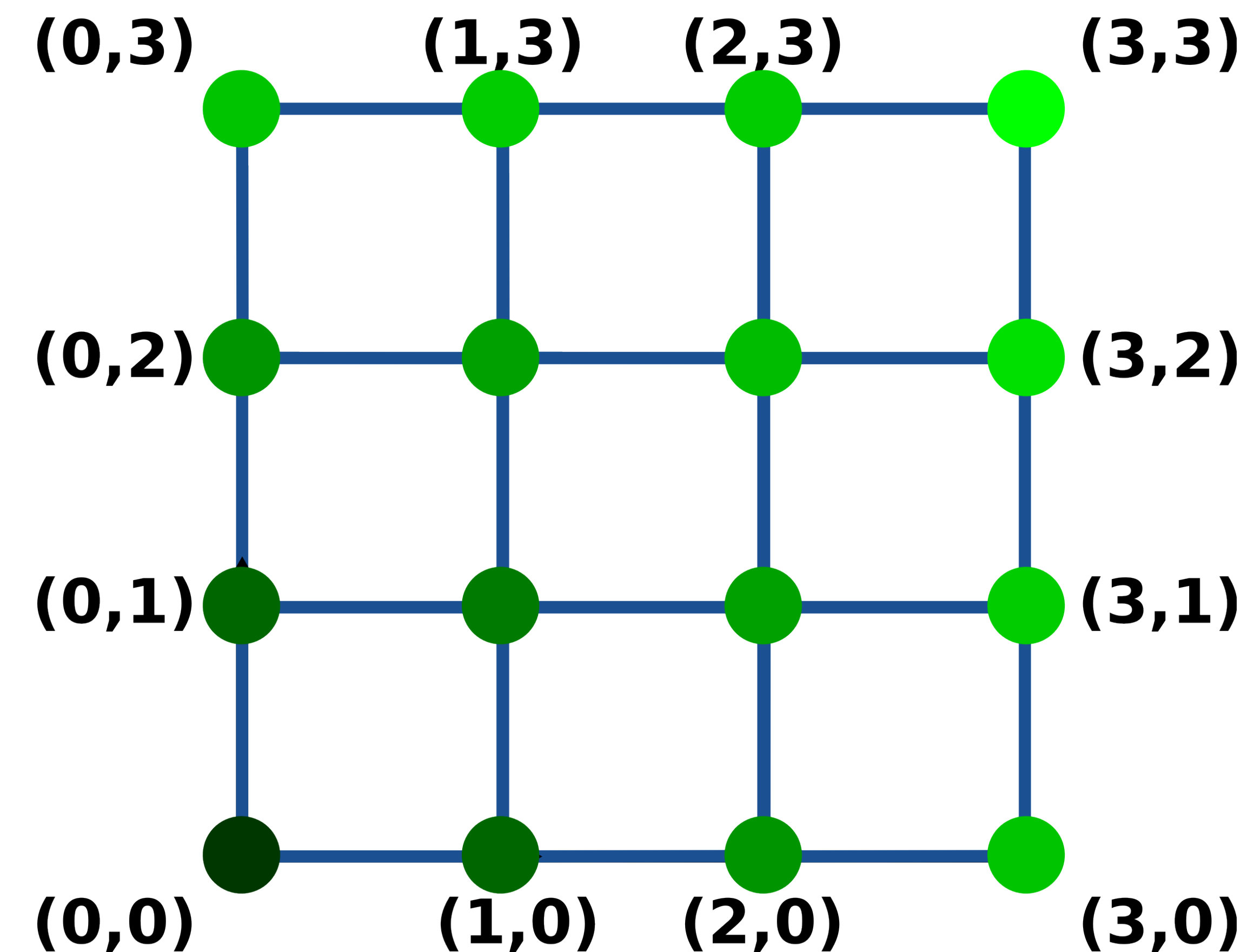
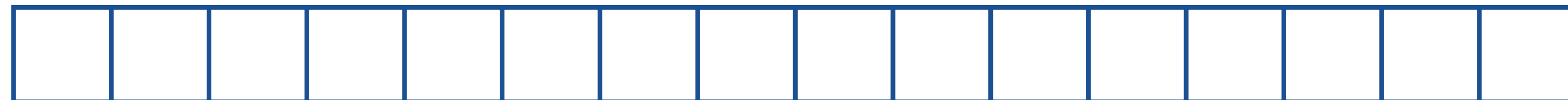


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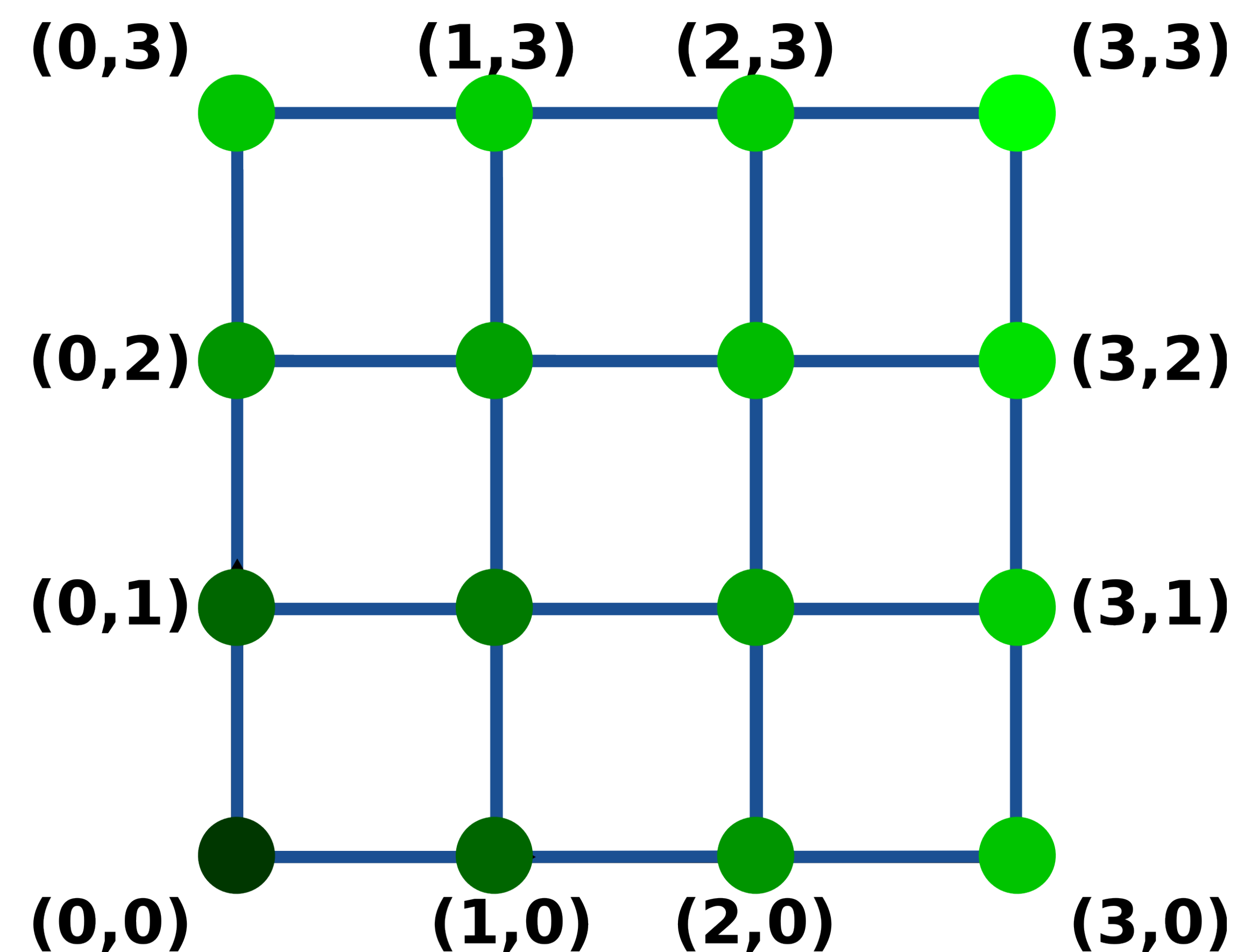


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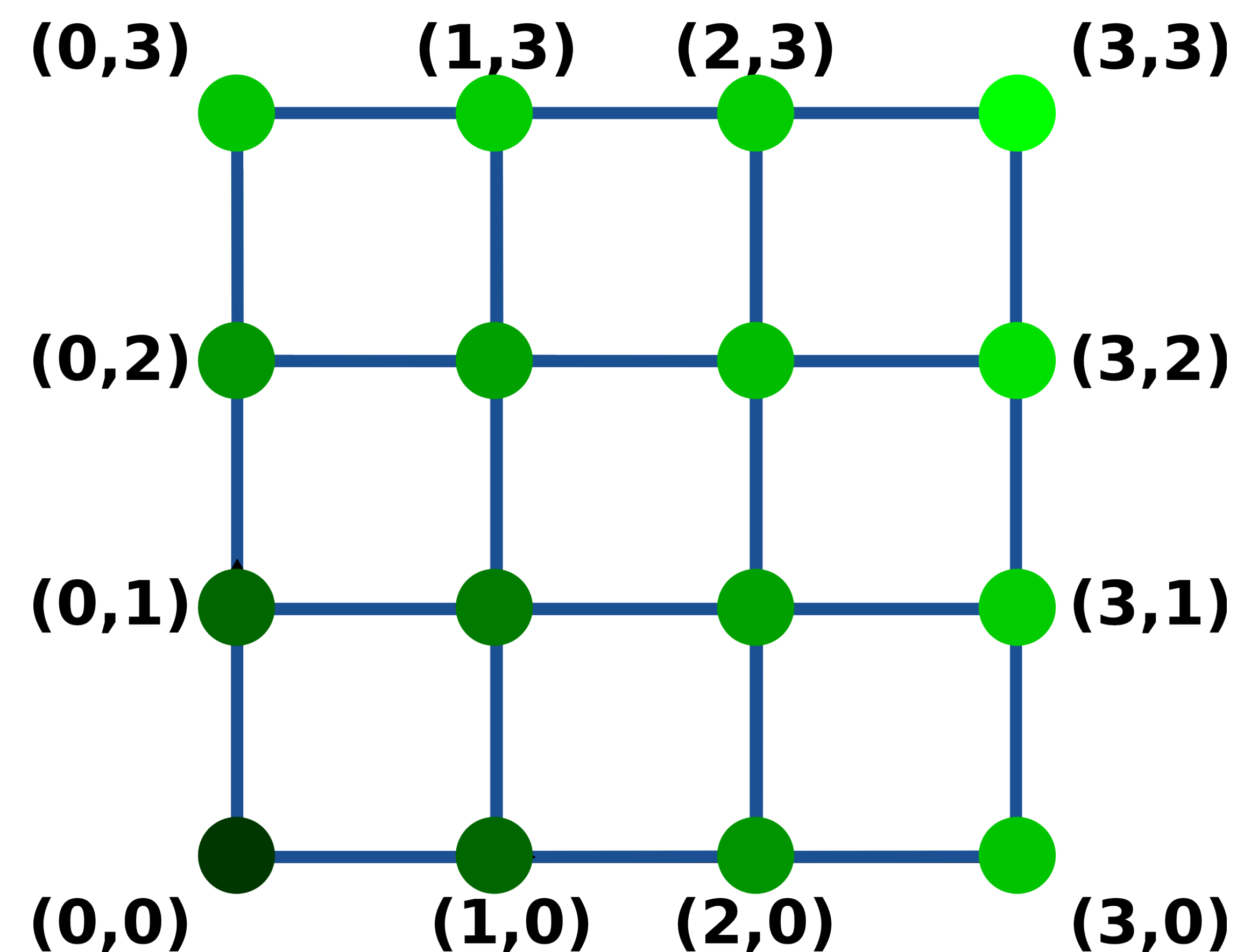


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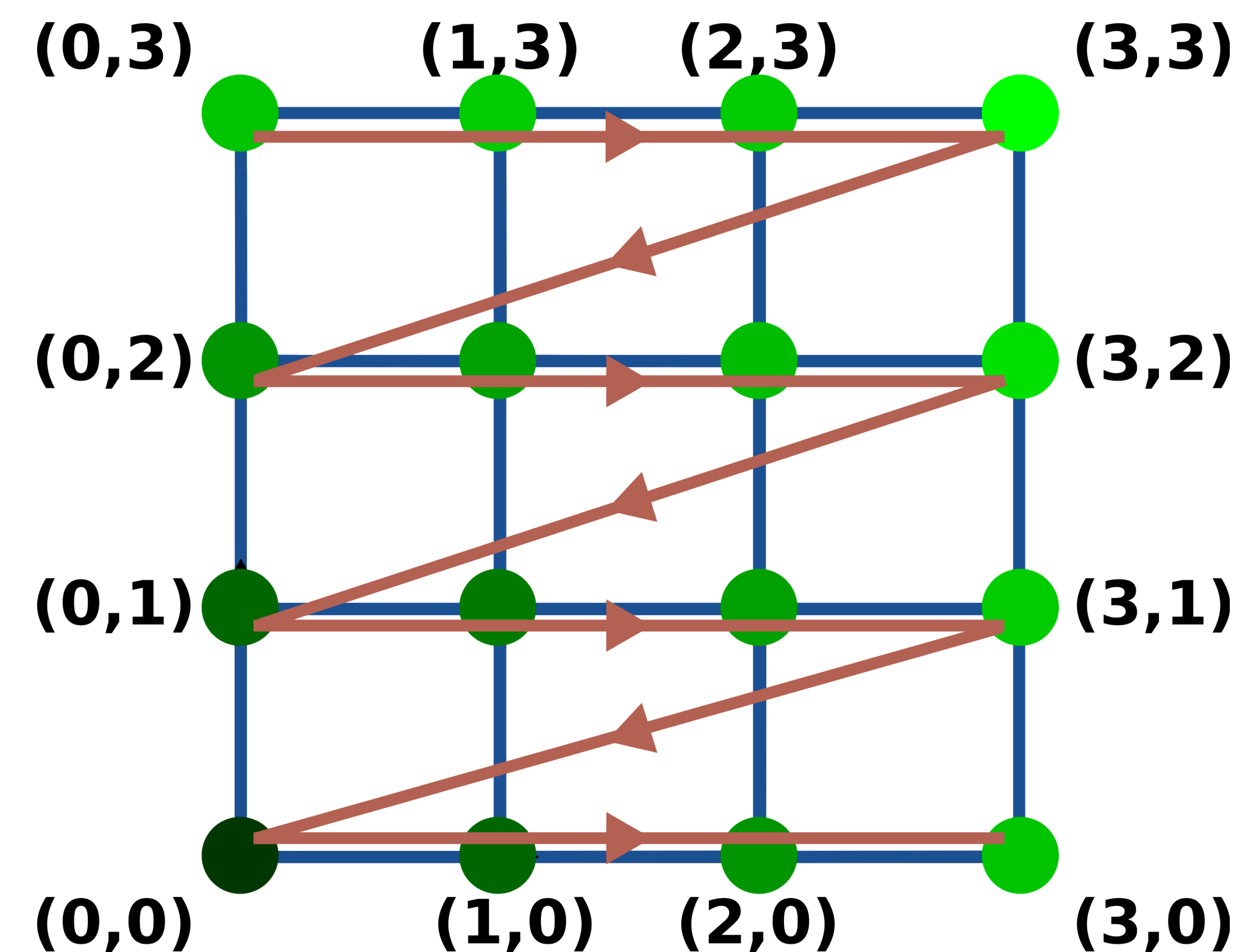
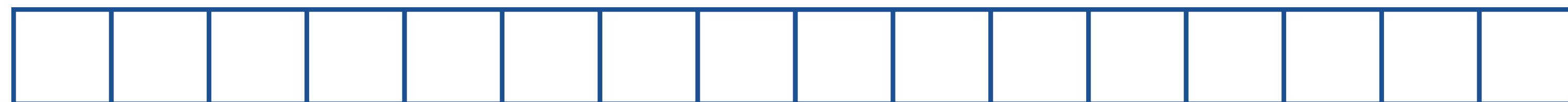


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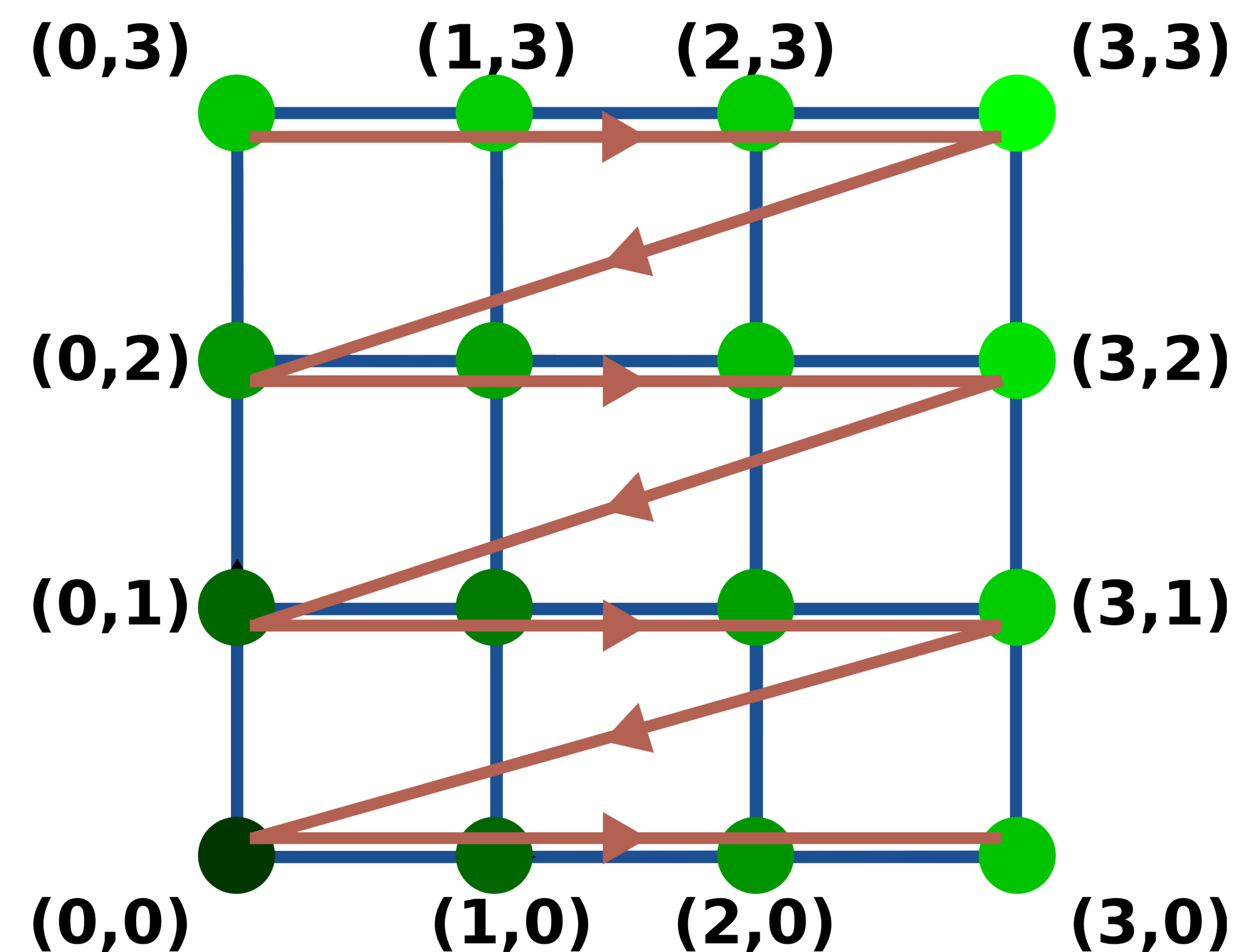
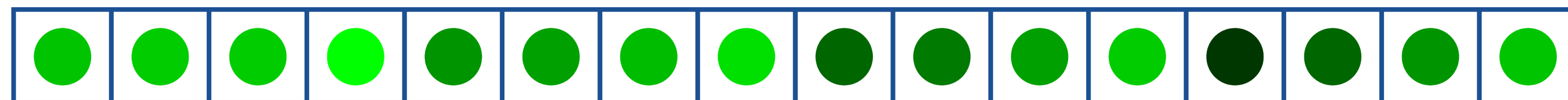


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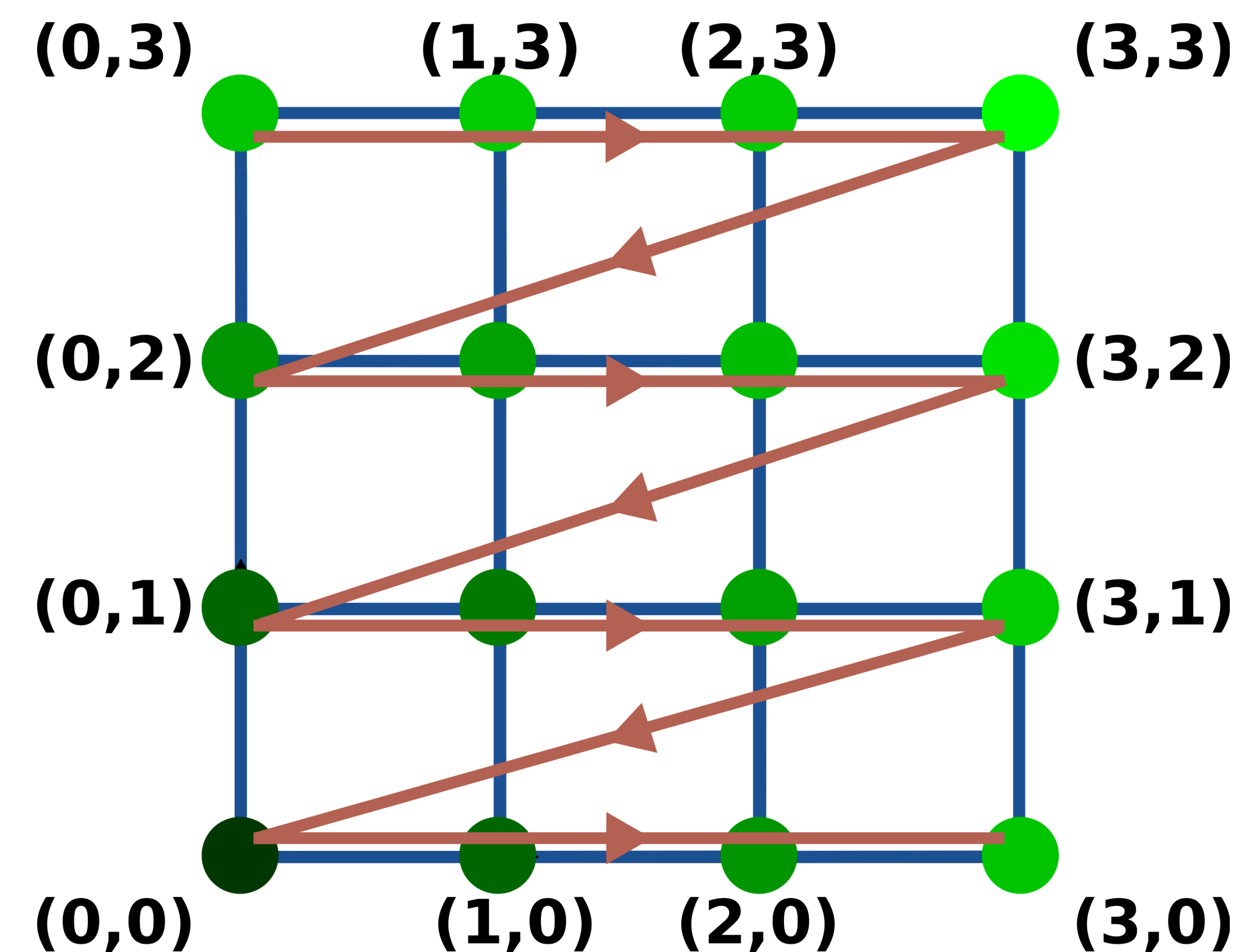
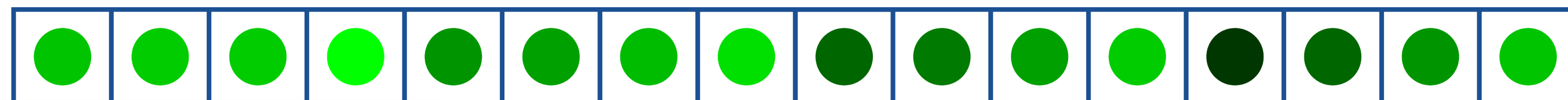


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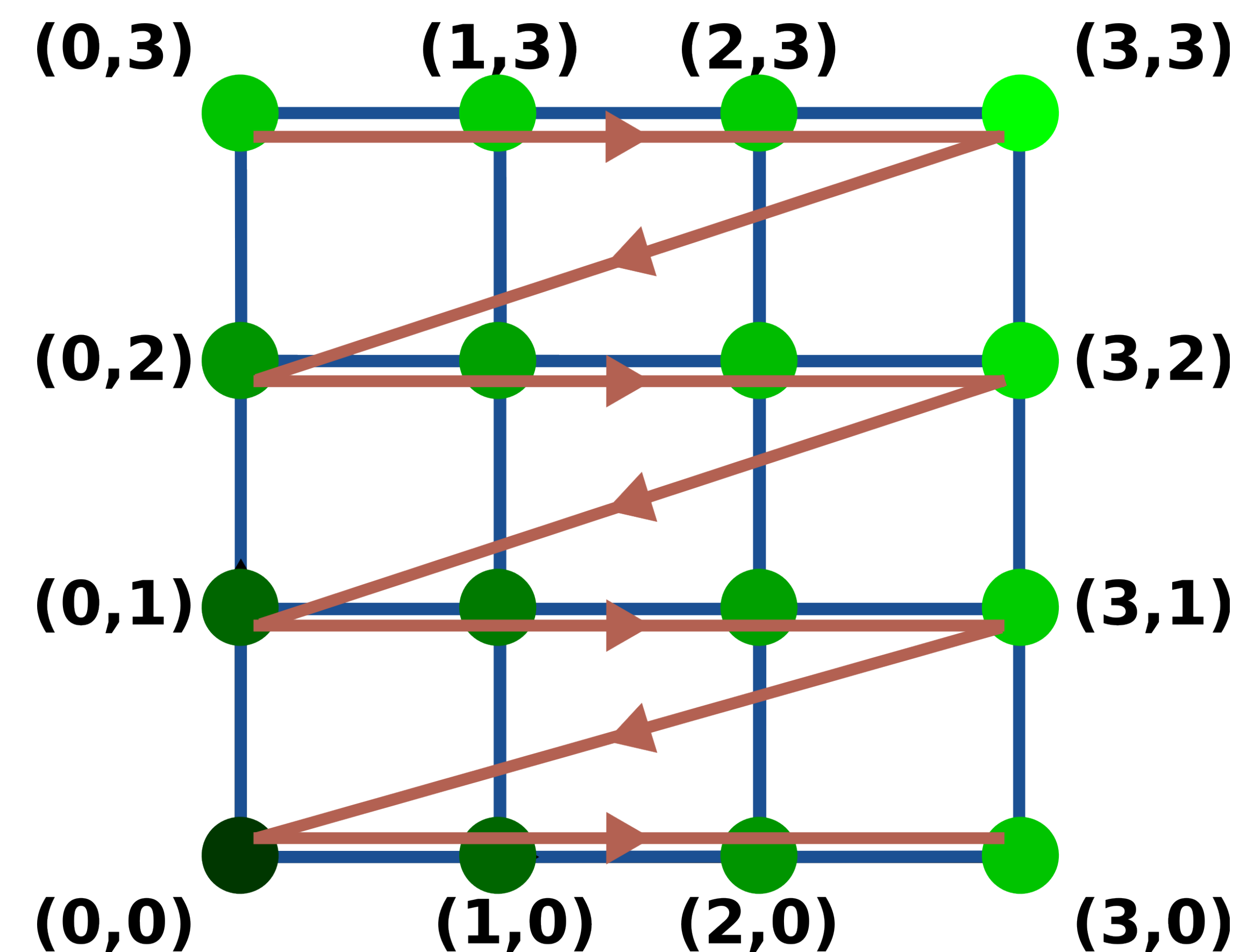
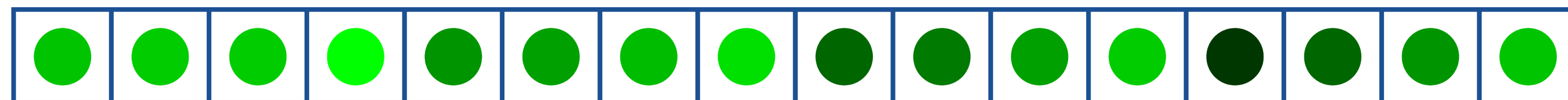


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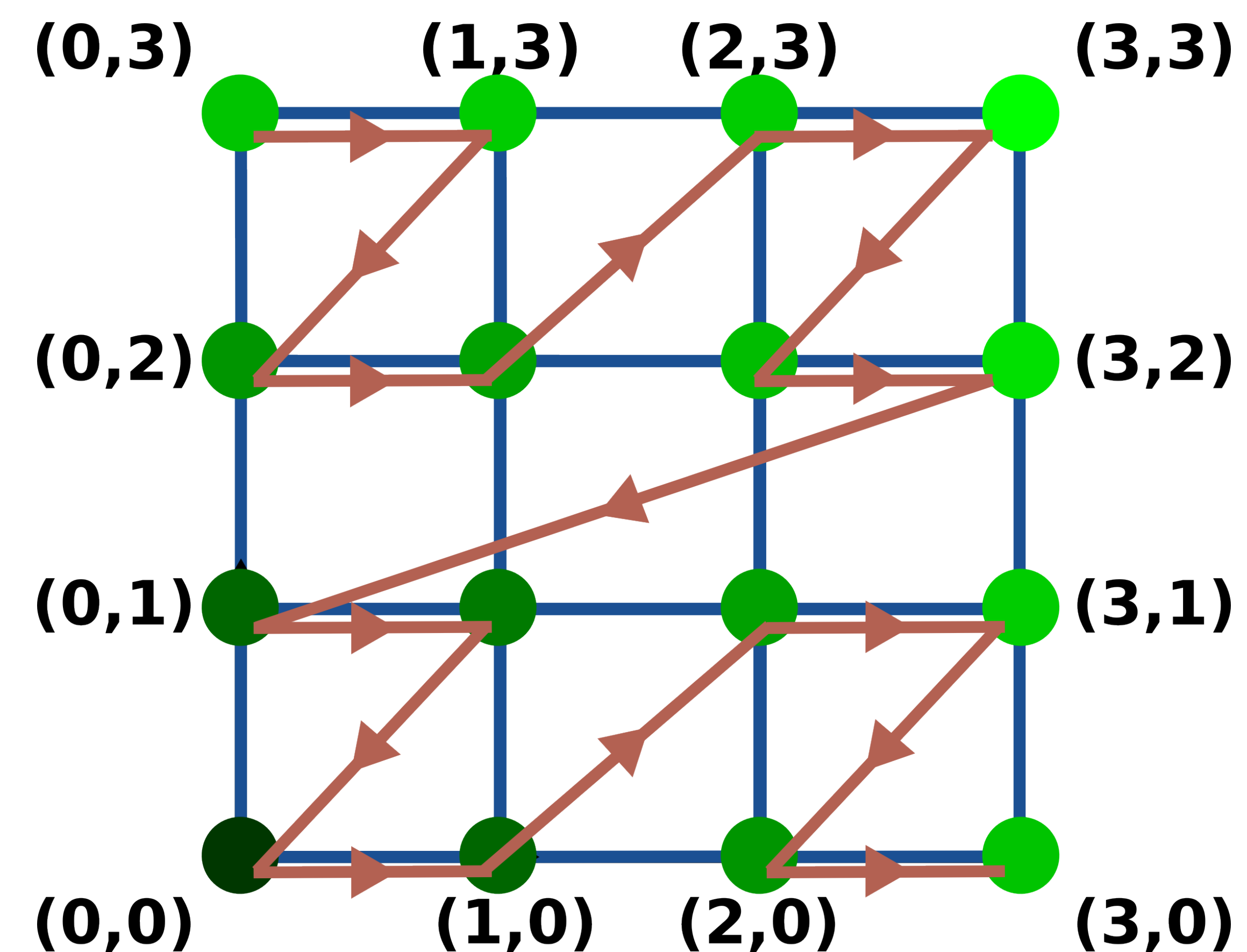
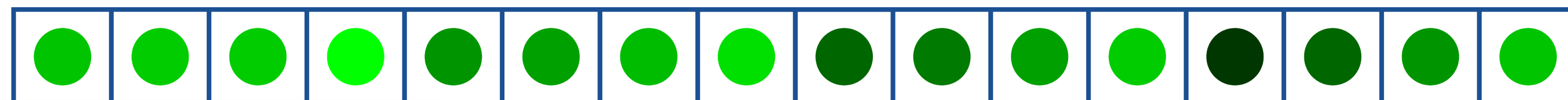


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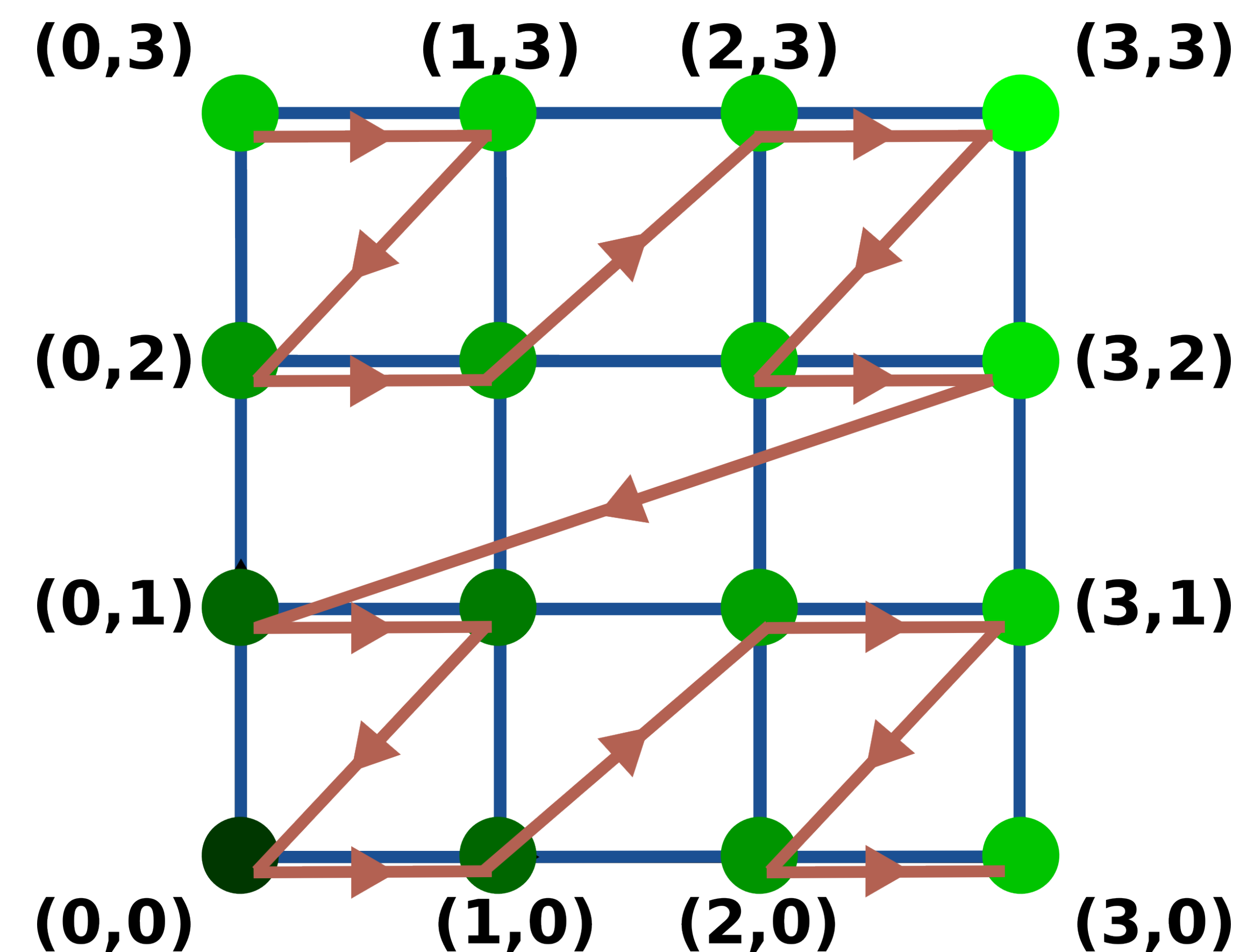
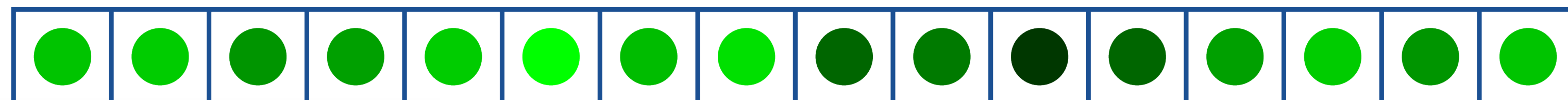


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Regular grids of \mathbb{R}^3

Regular grids of \mathbb{R}^3

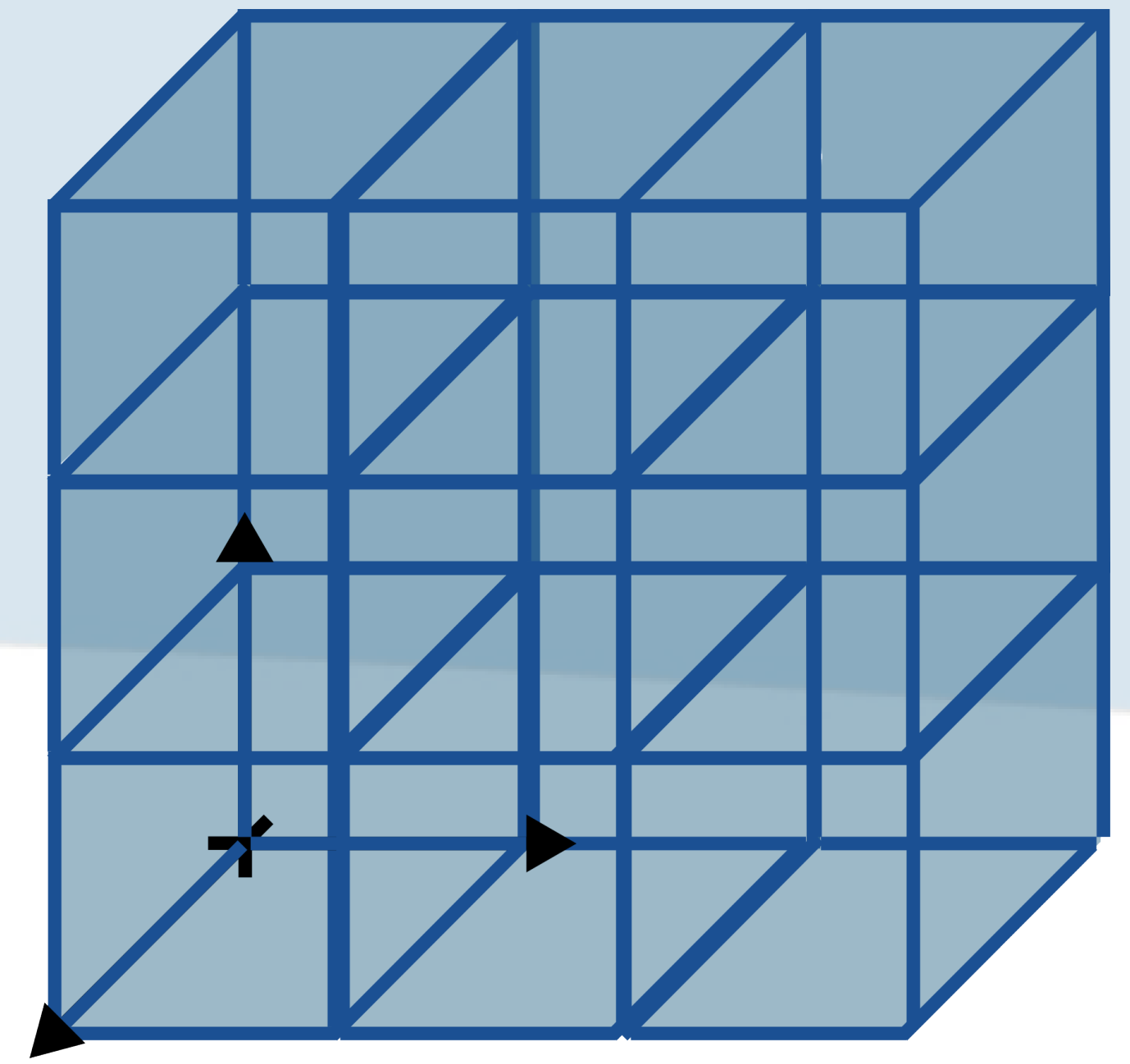
- Same problem
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Regular grids of \mathbb{R}^3

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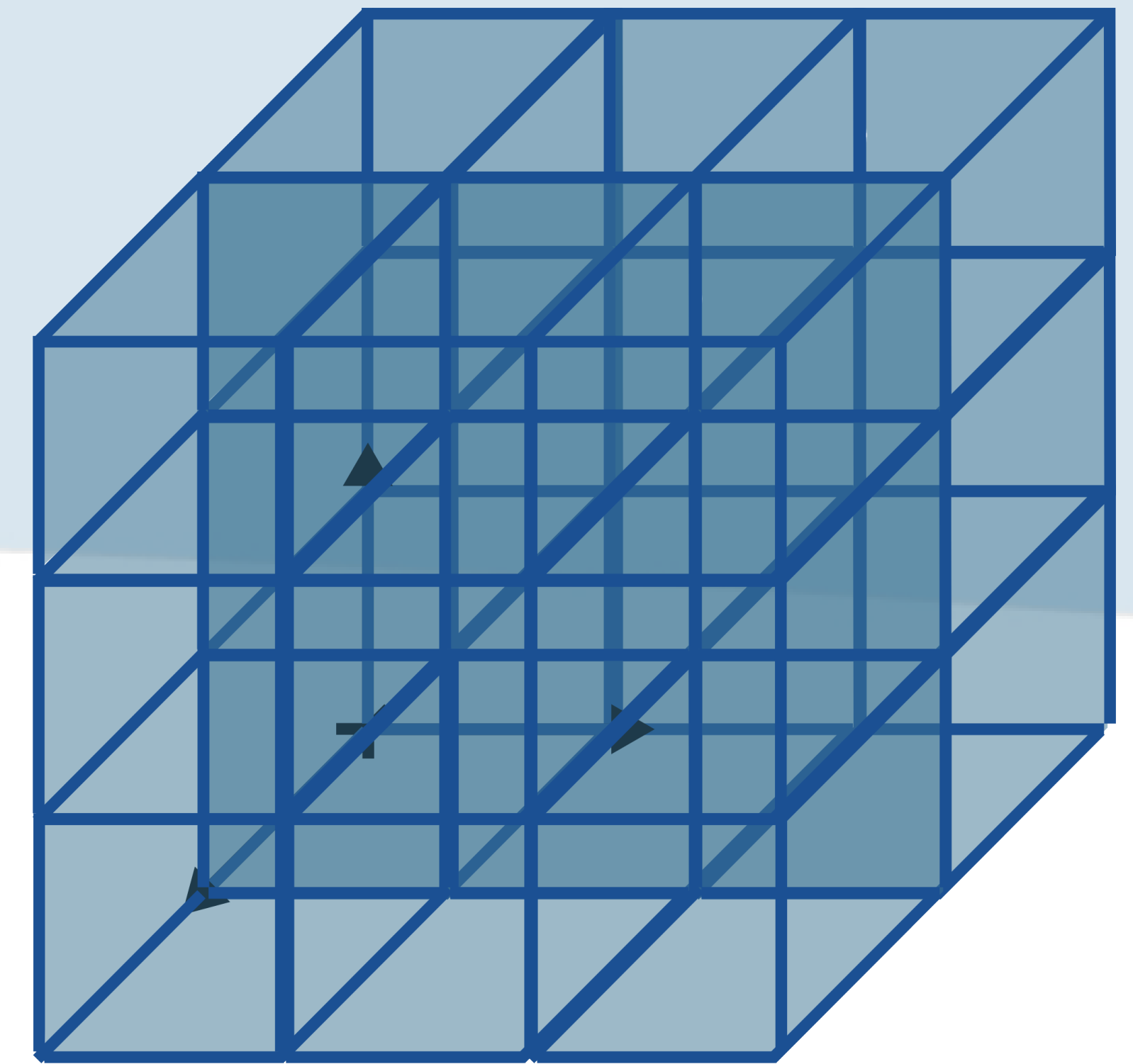
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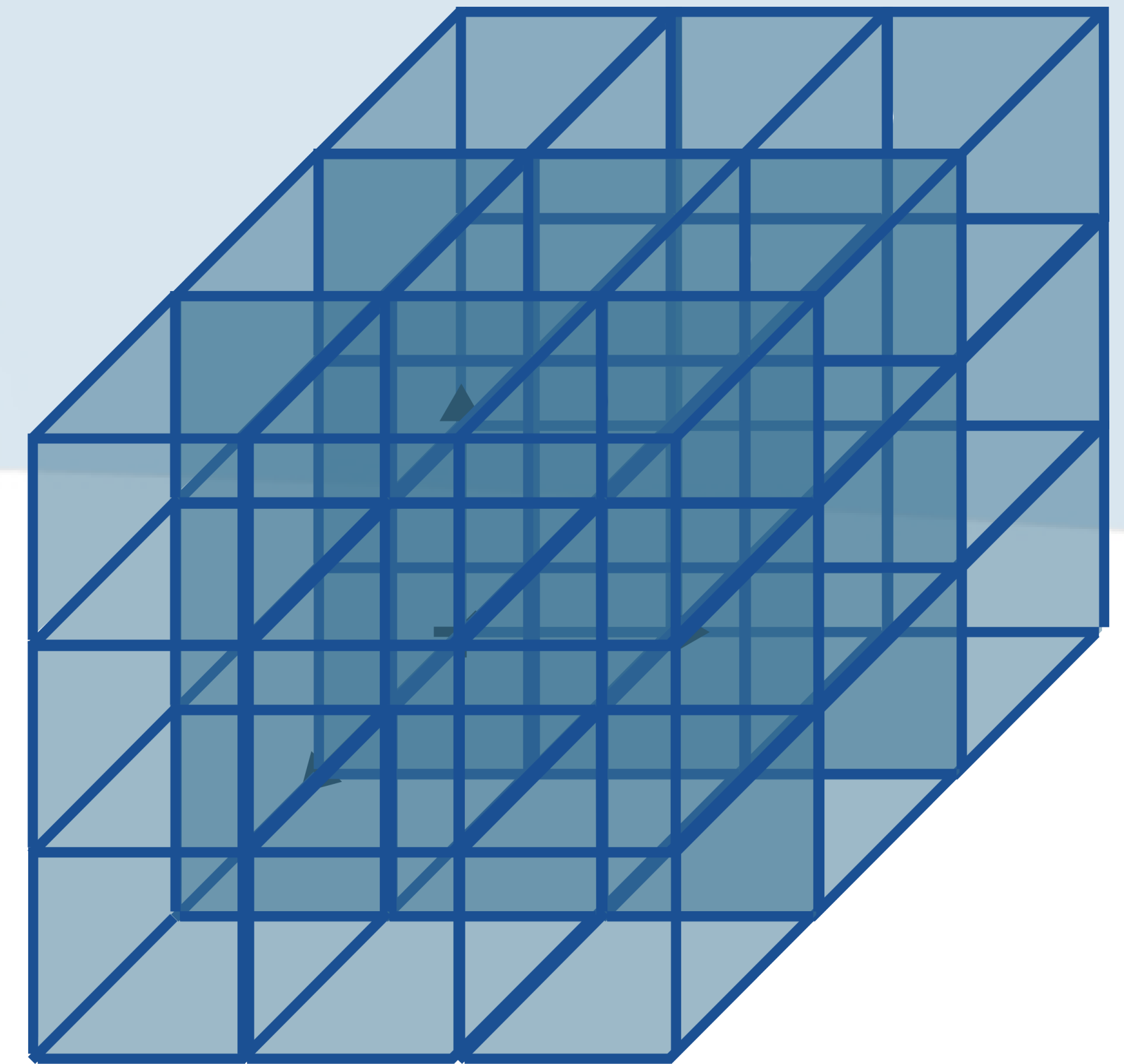
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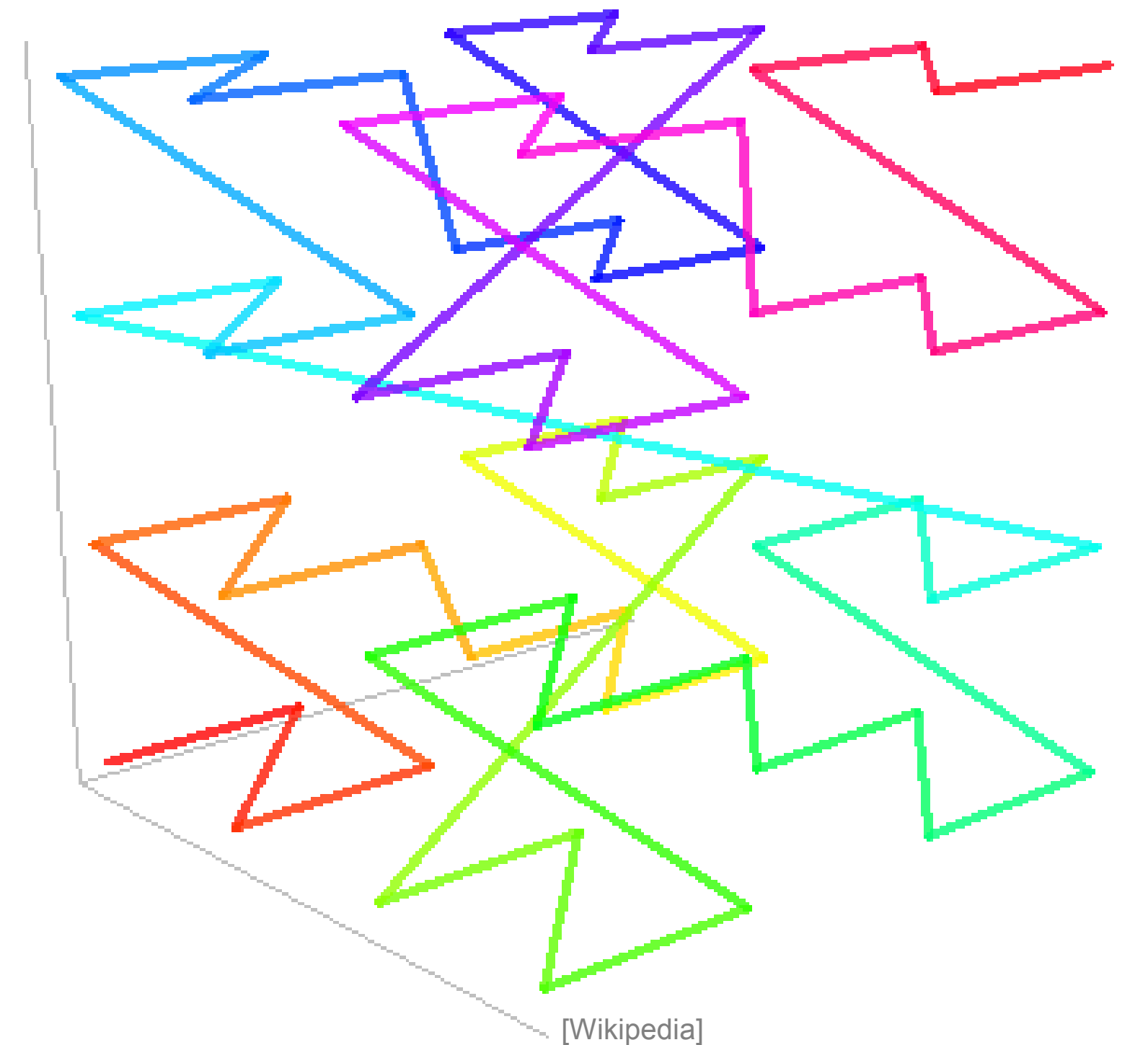
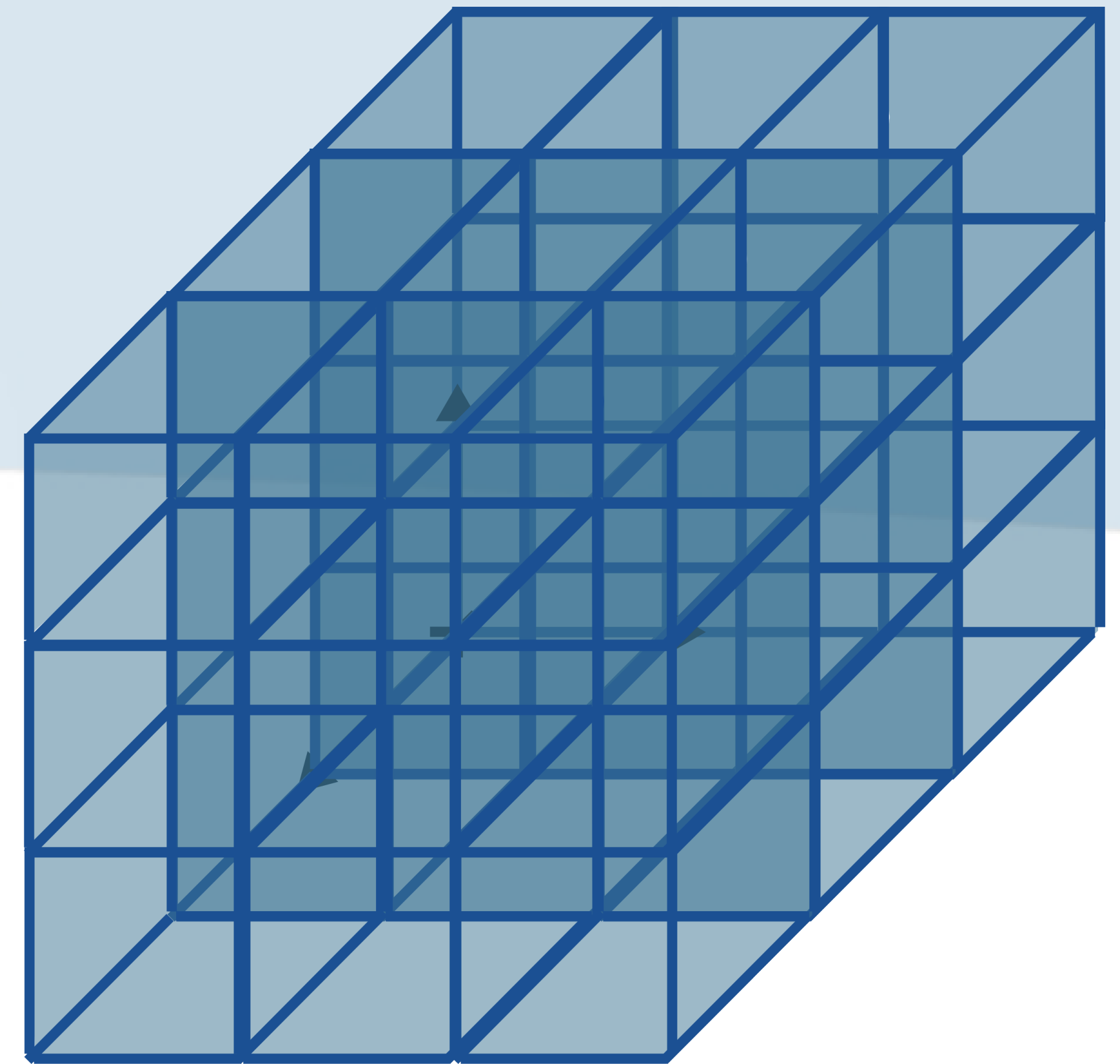
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Regular grids of \mathbb{R}^3

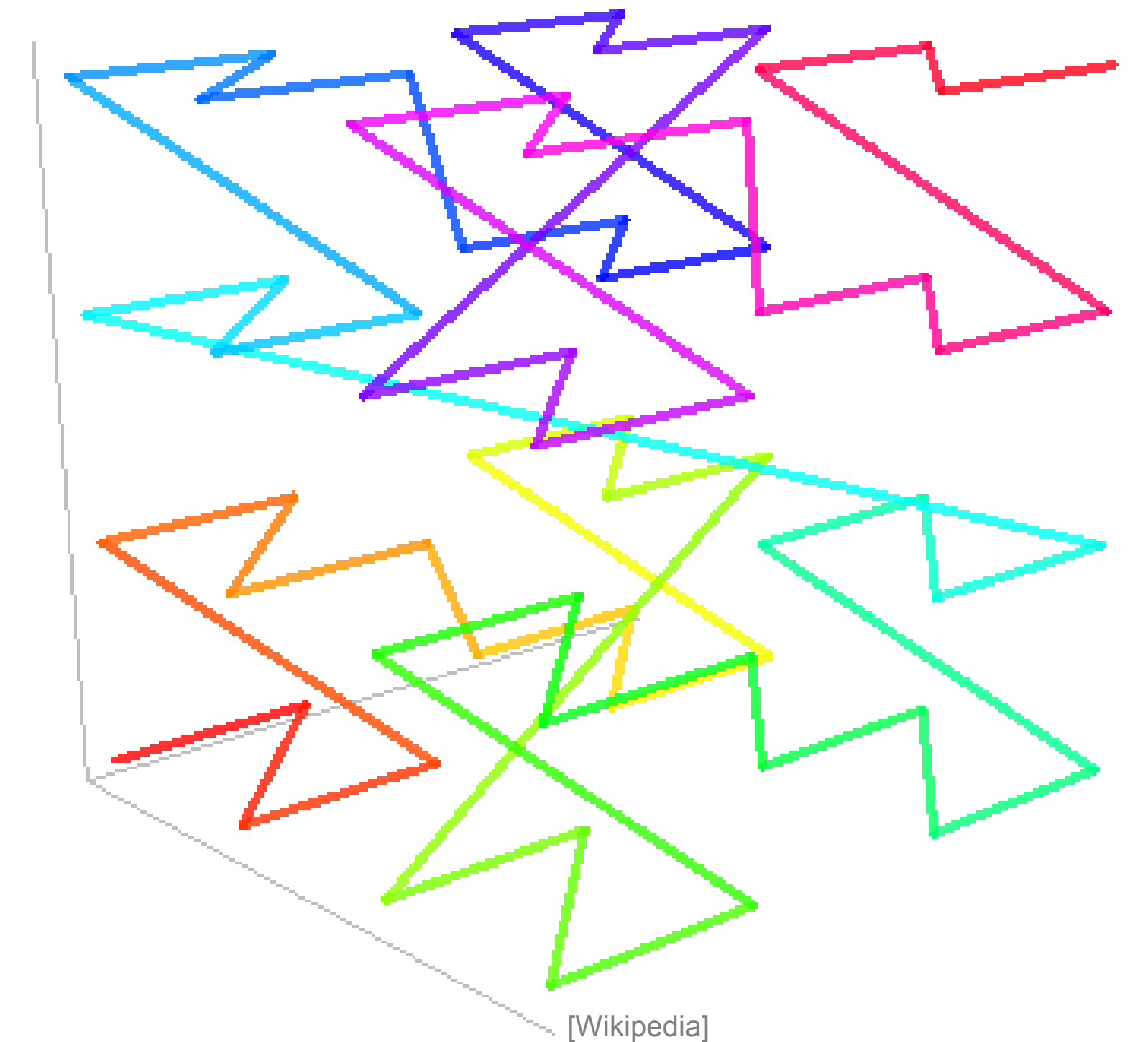
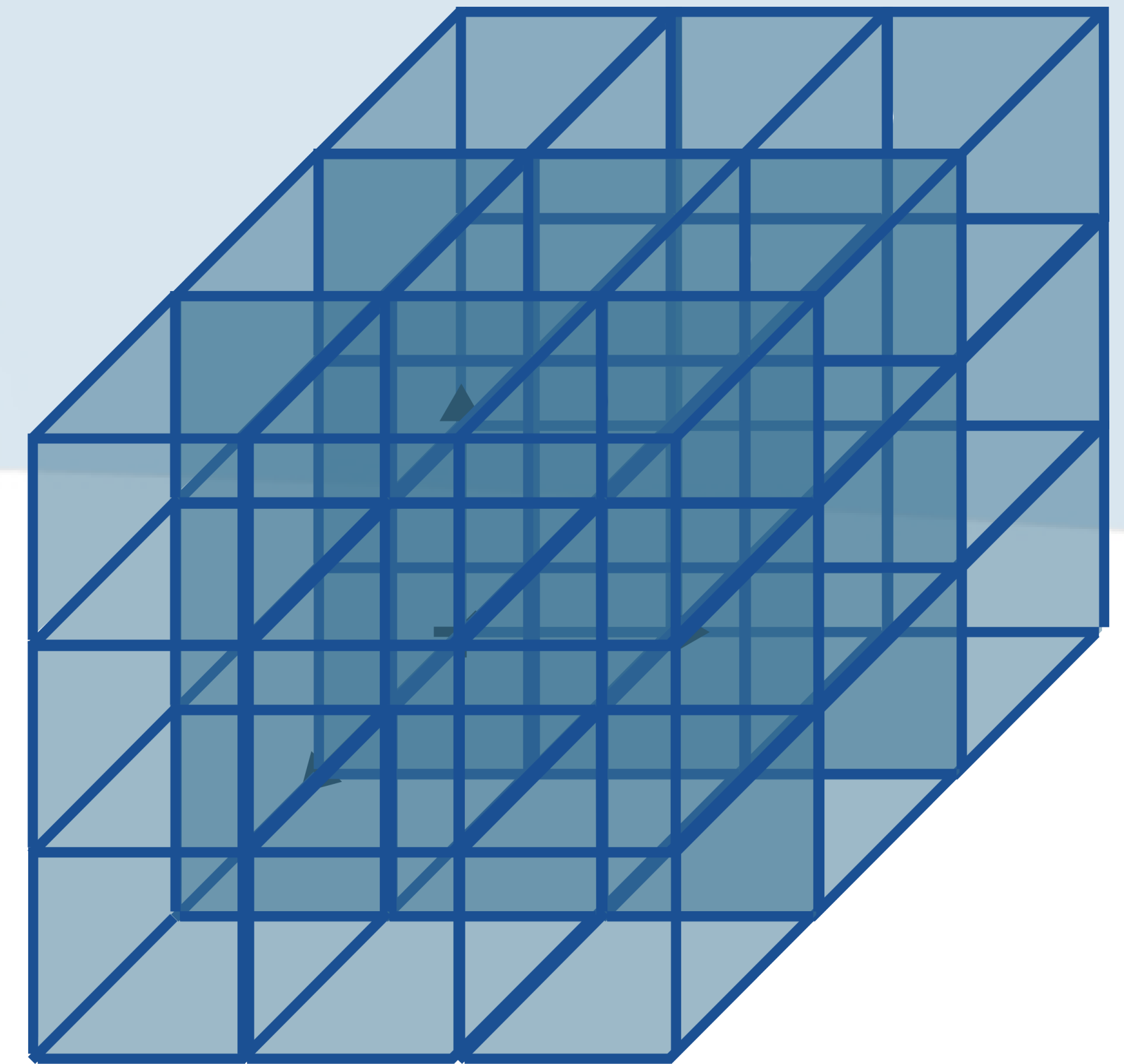
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[Wikipedia]

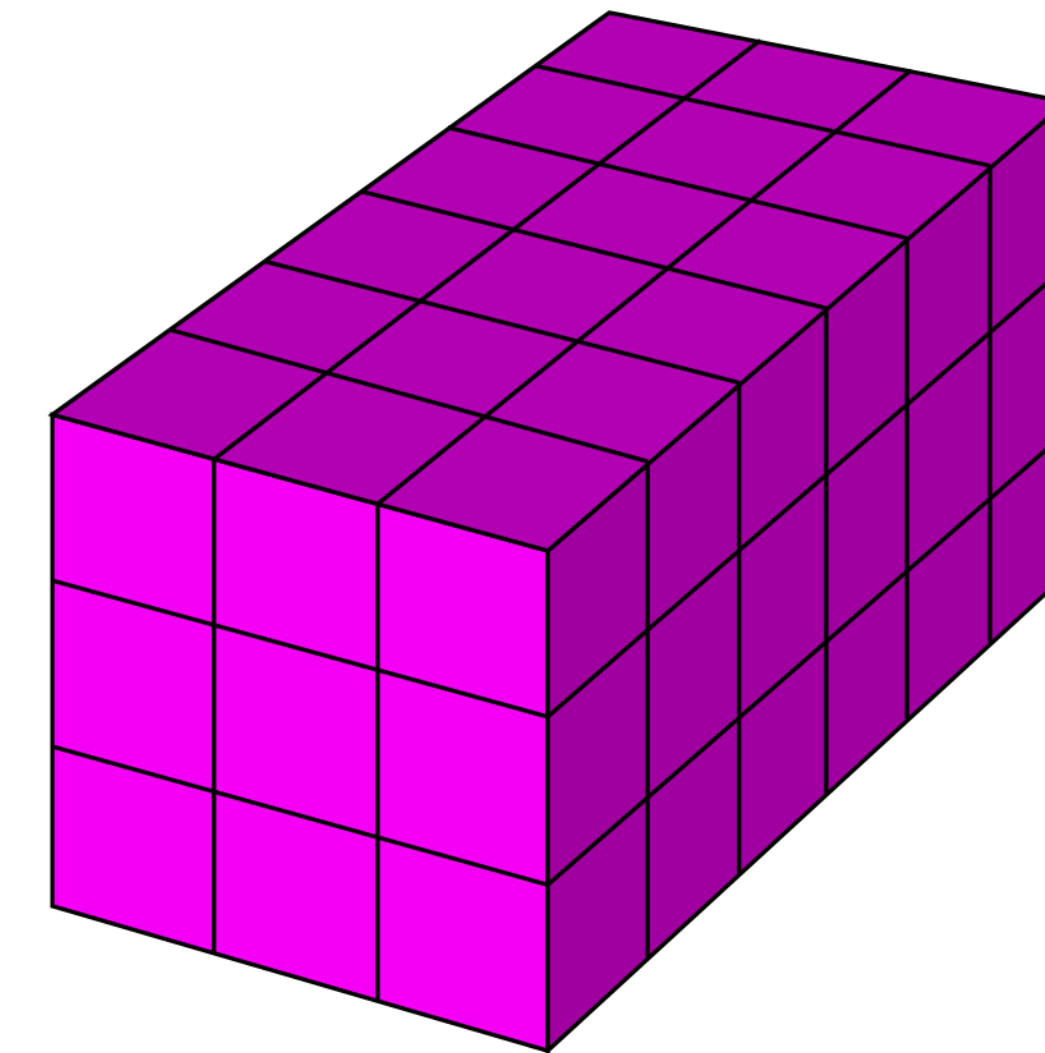
Regular grids of \mathbb{R}^n

- Same problem
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 - Same solutions
 - One dimension at a time
 - Peano's curves
 - Hilbert's curves, ...
 - Lebesgue's curves (z-order)
- Generalizes to higher dimensions



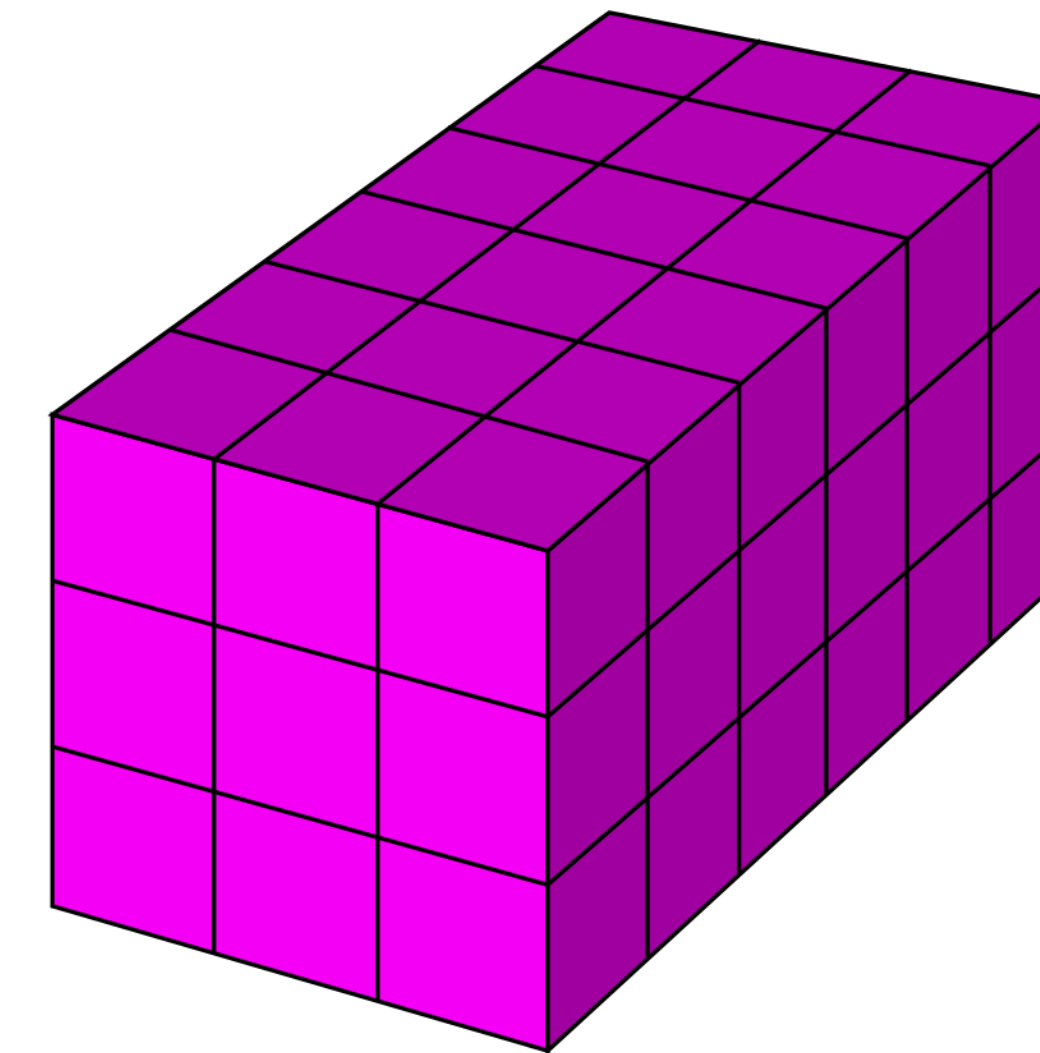
[Wikipedia]

Embedding functions



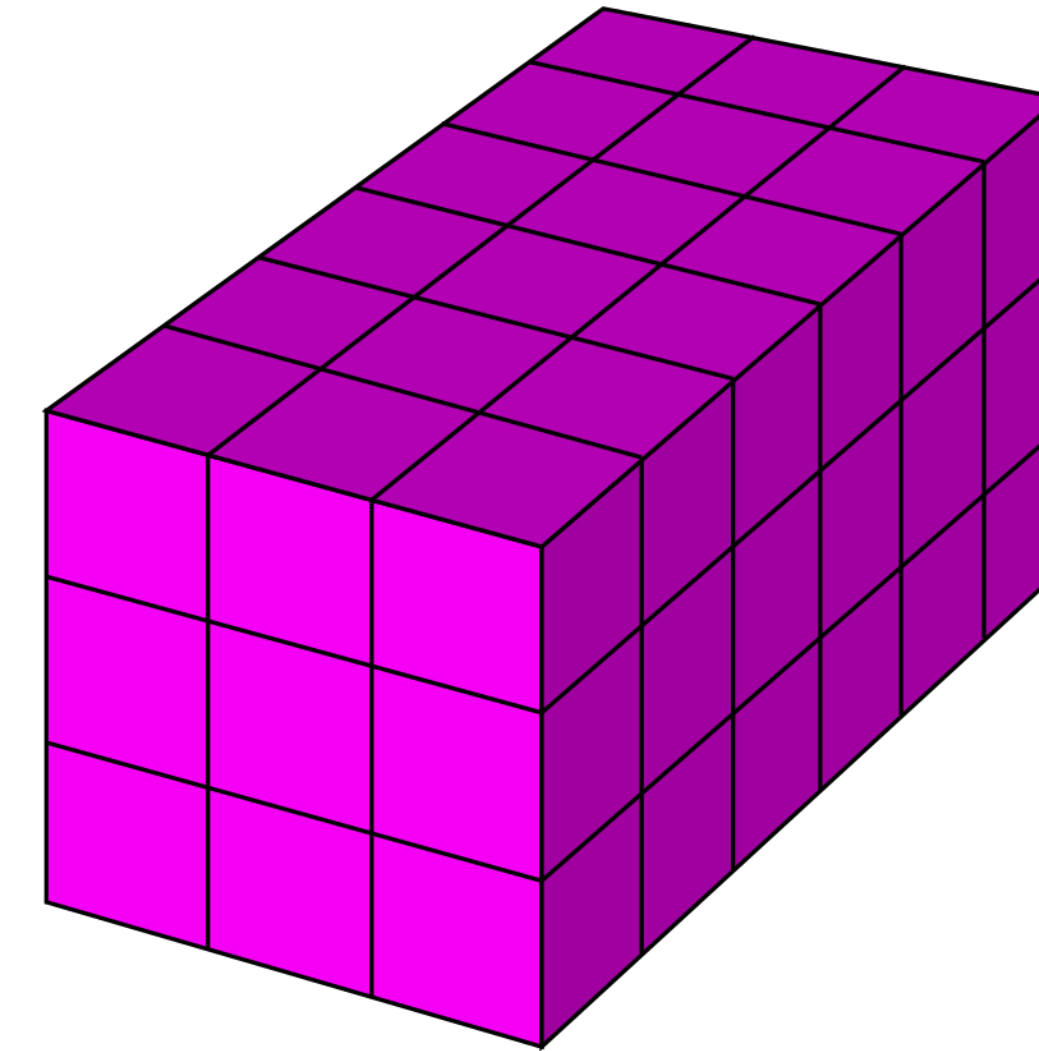
Embedding functions

- Problem
 - What if we want to adapt the sampling to the geometry of the data?



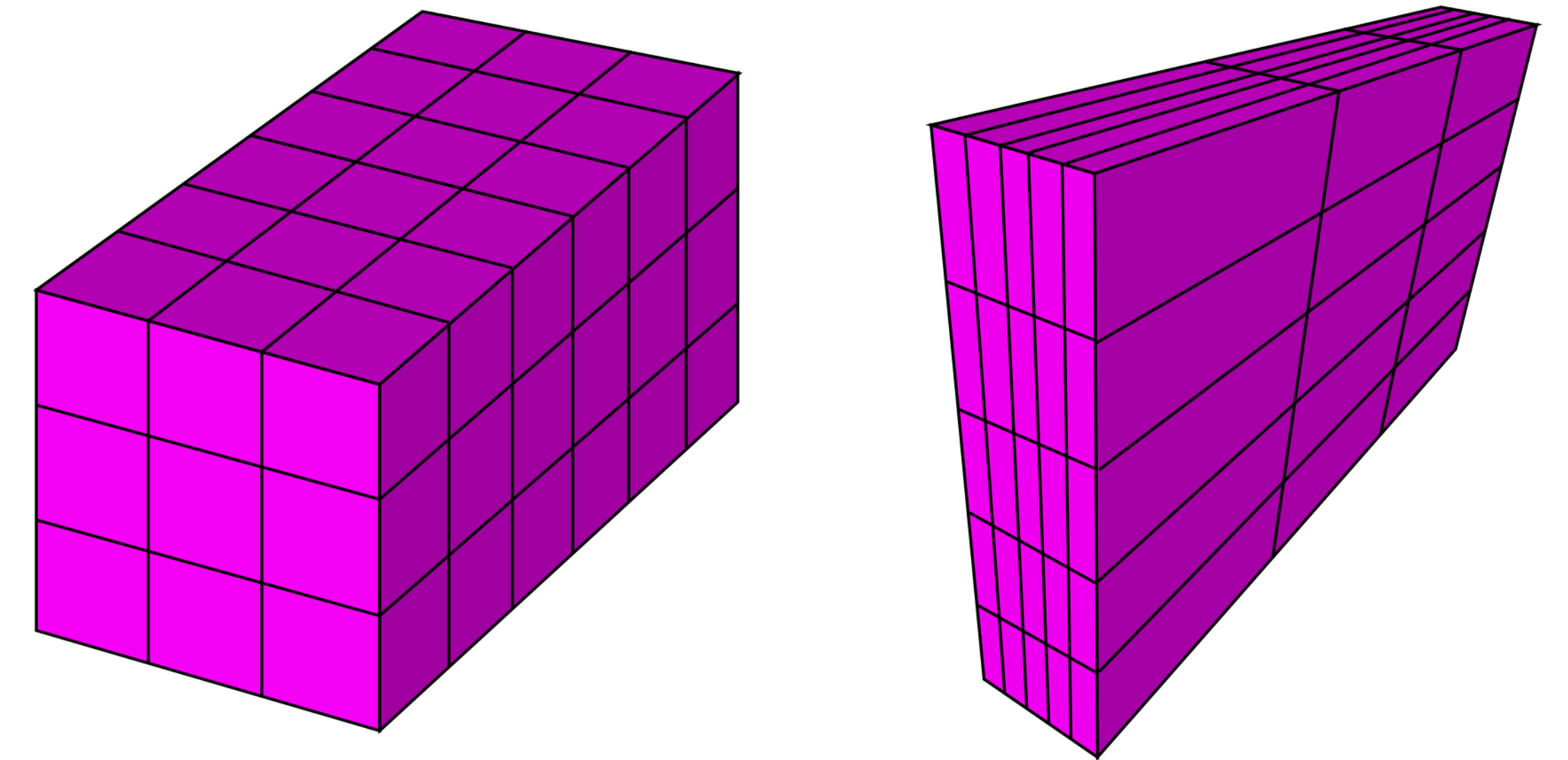
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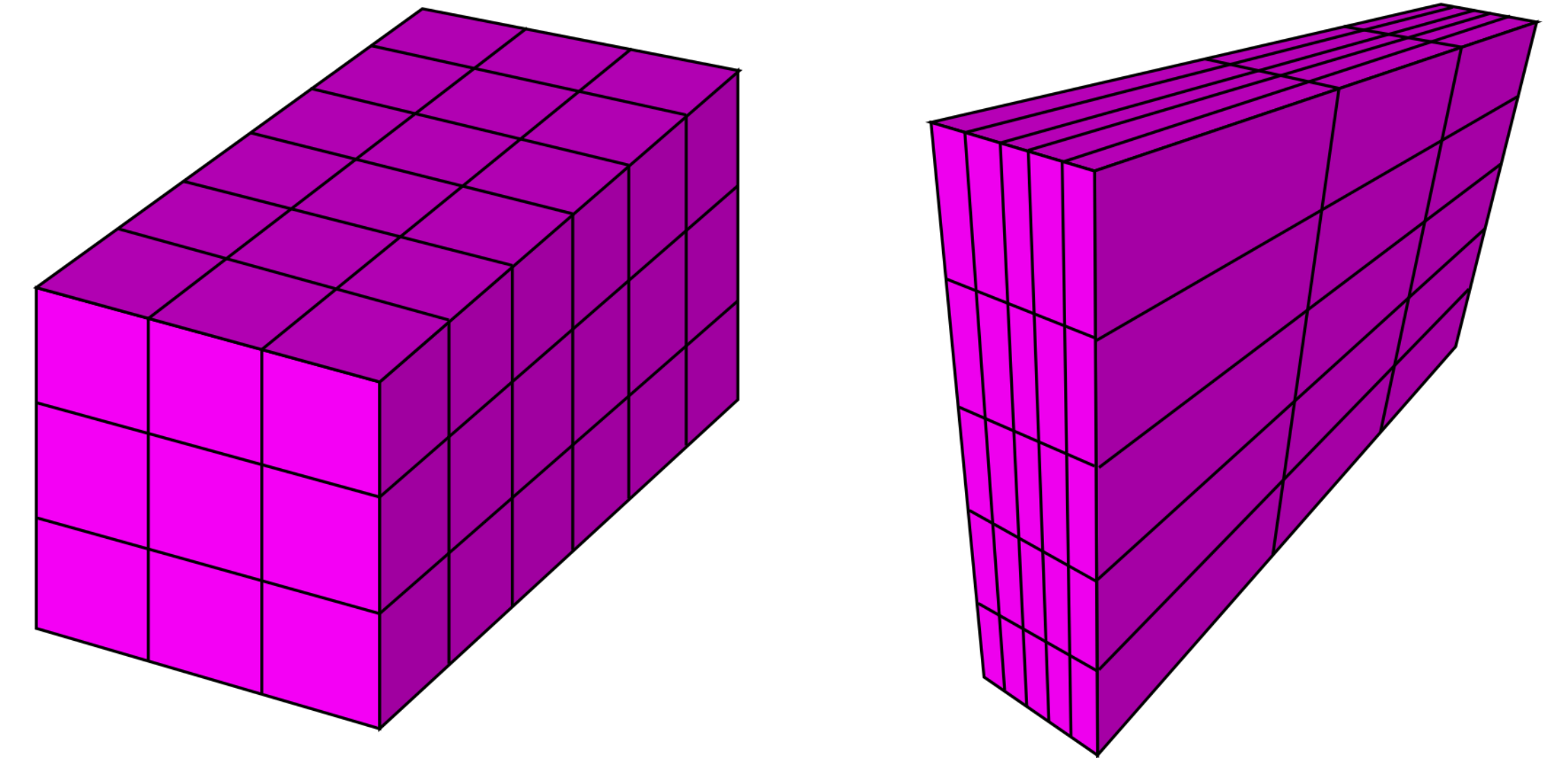
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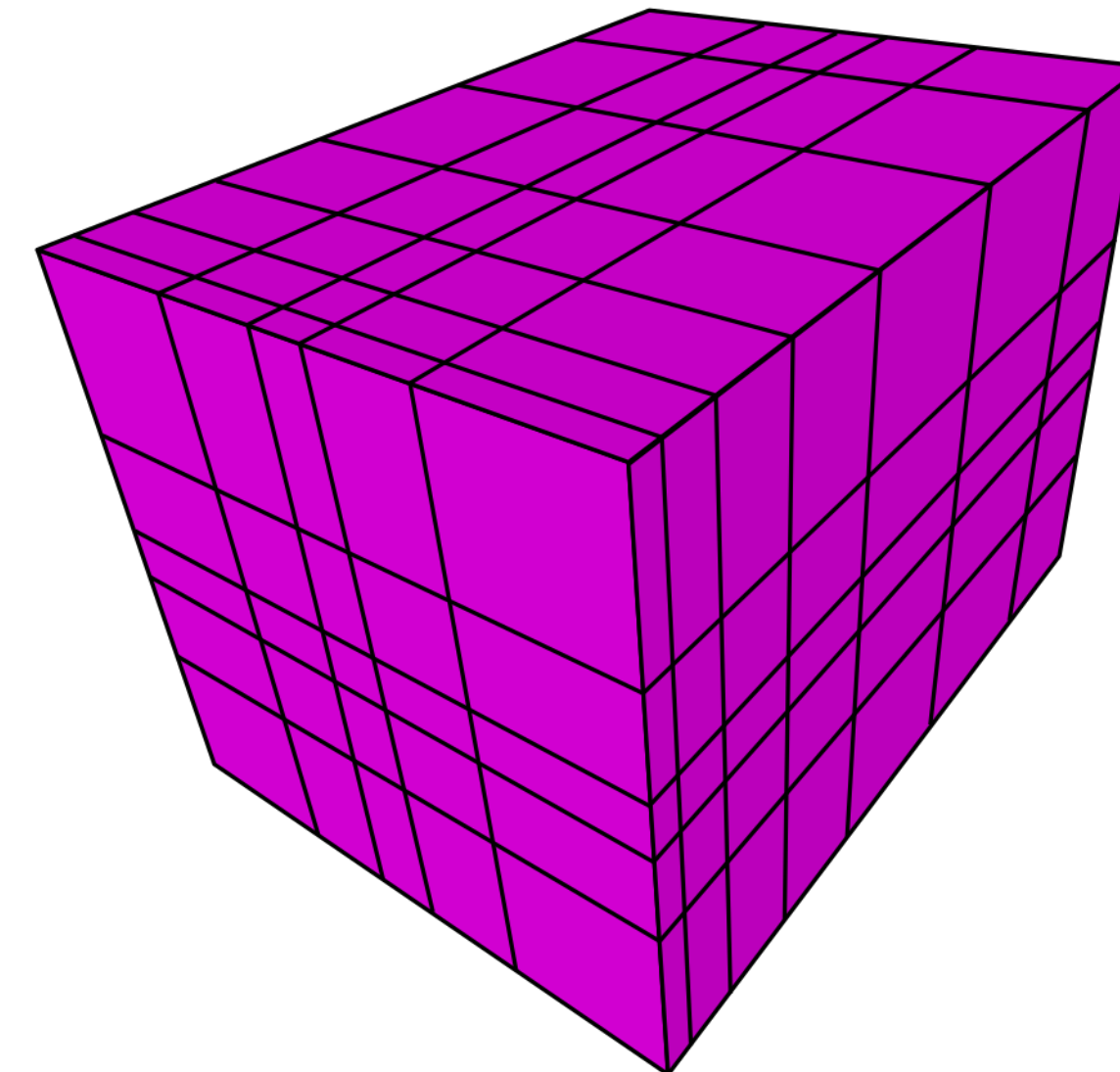
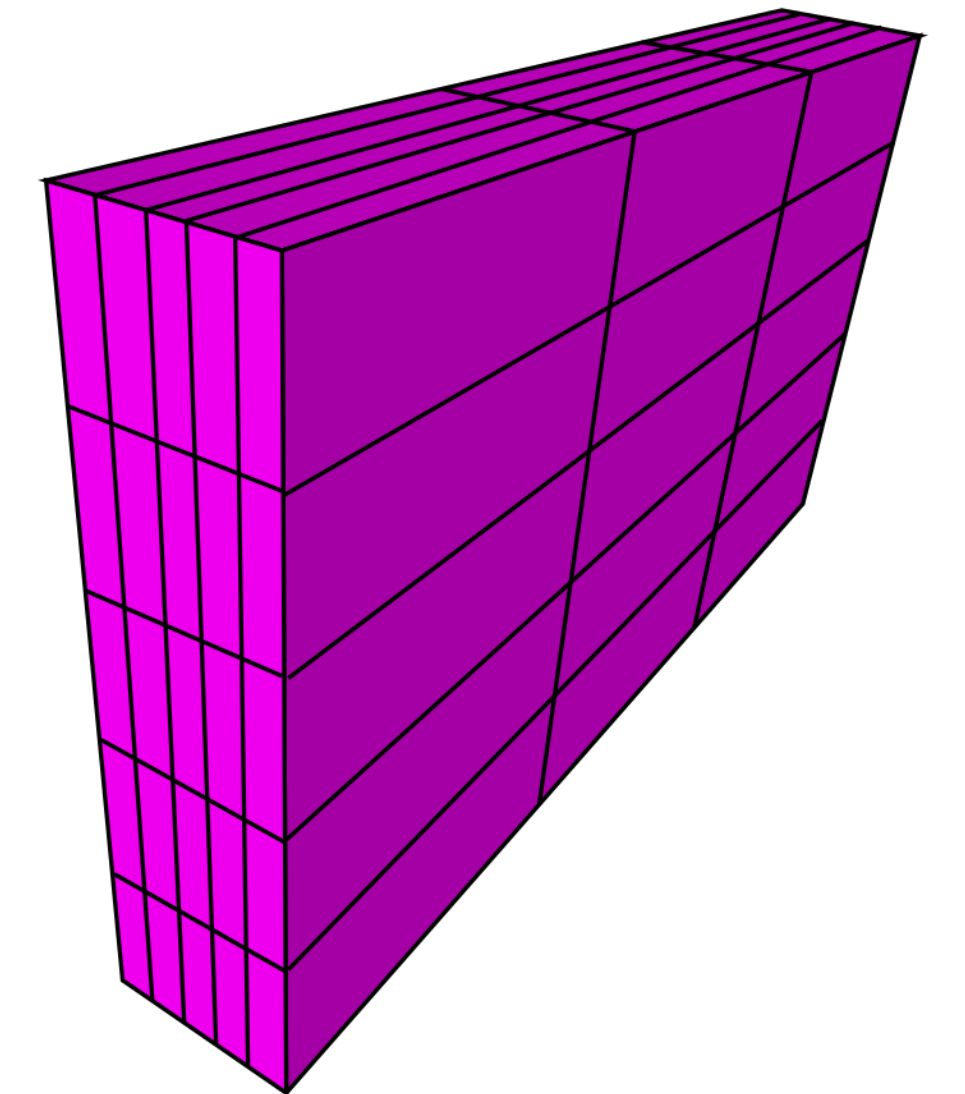
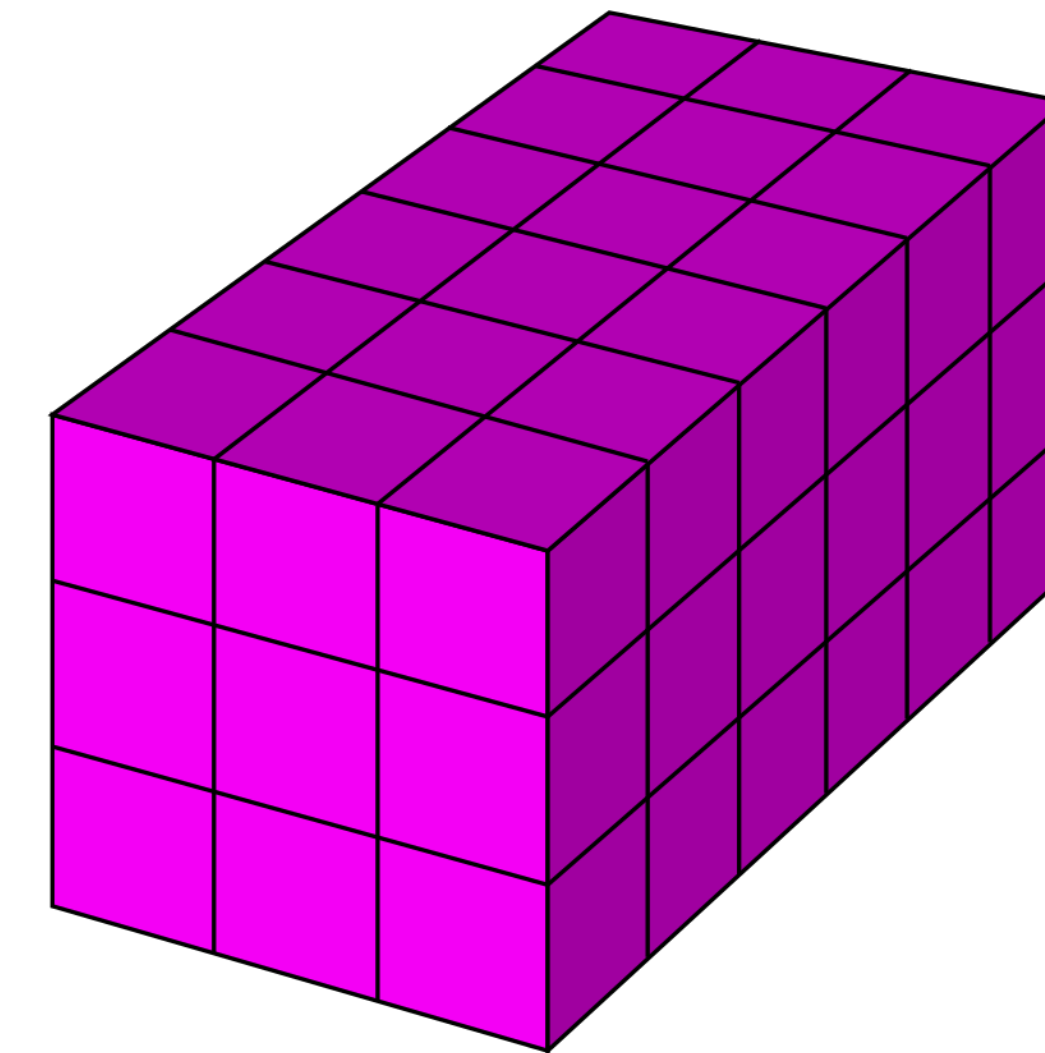
Embedding functions

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 - $e : \mathcal{G} \rightarrow \mathbb{R}^n$
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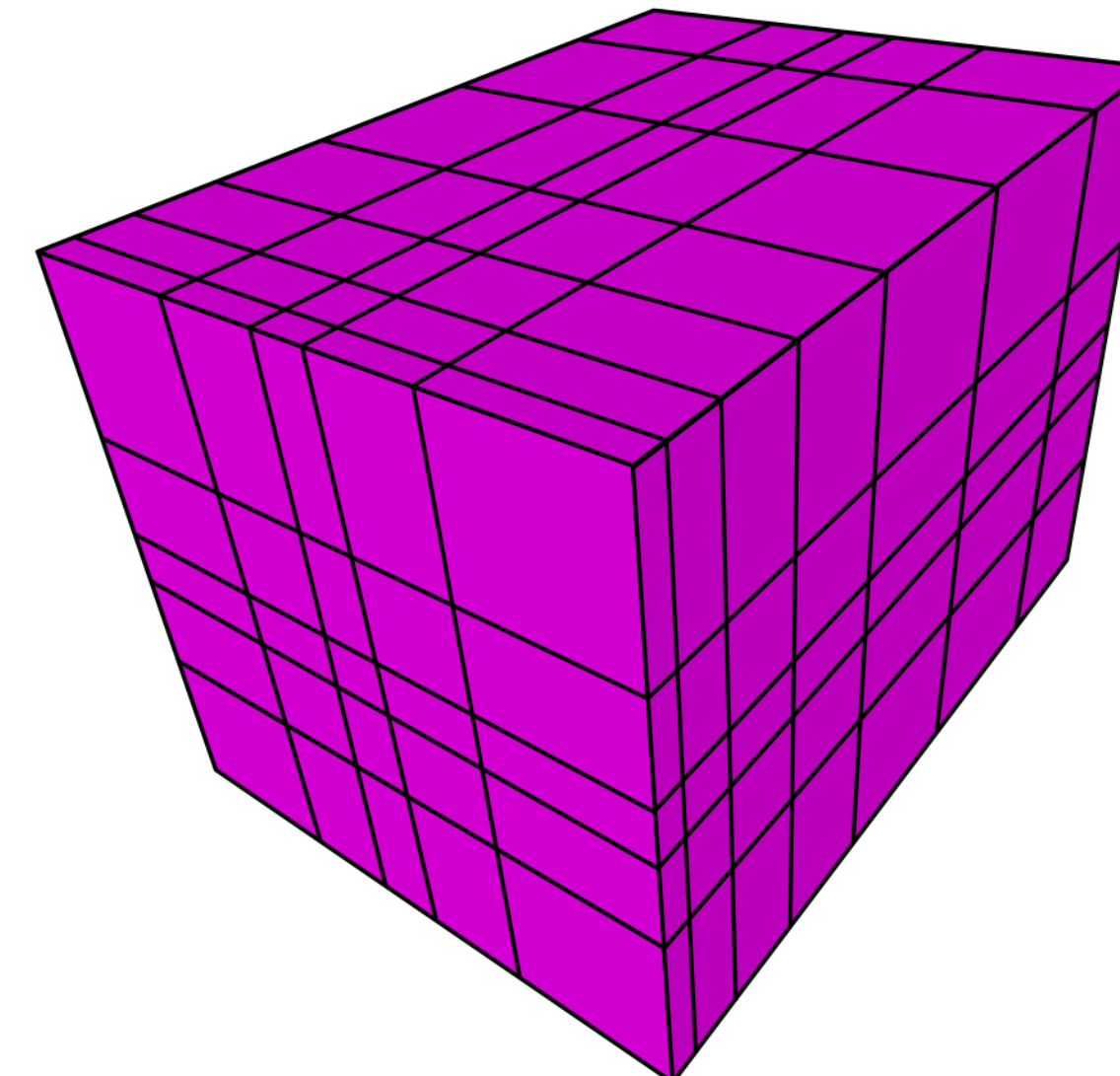
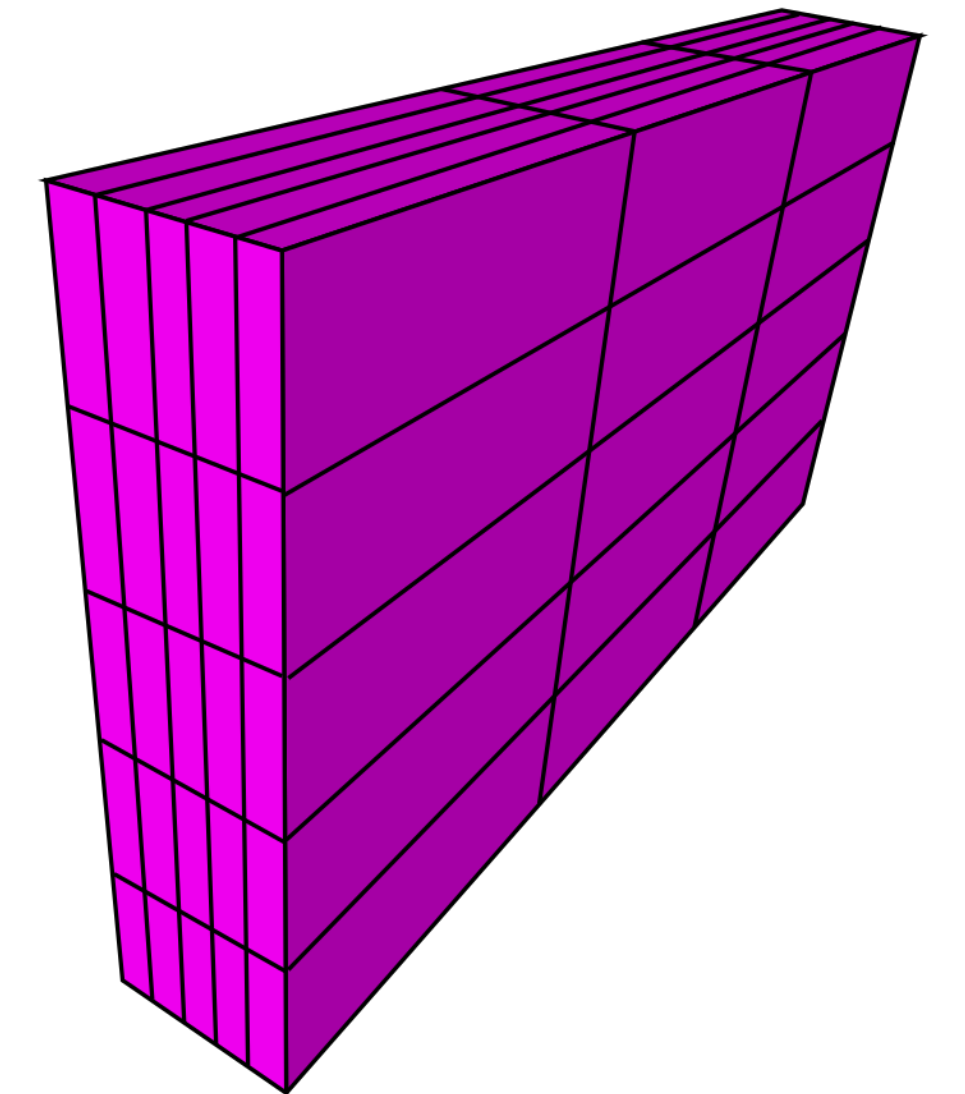
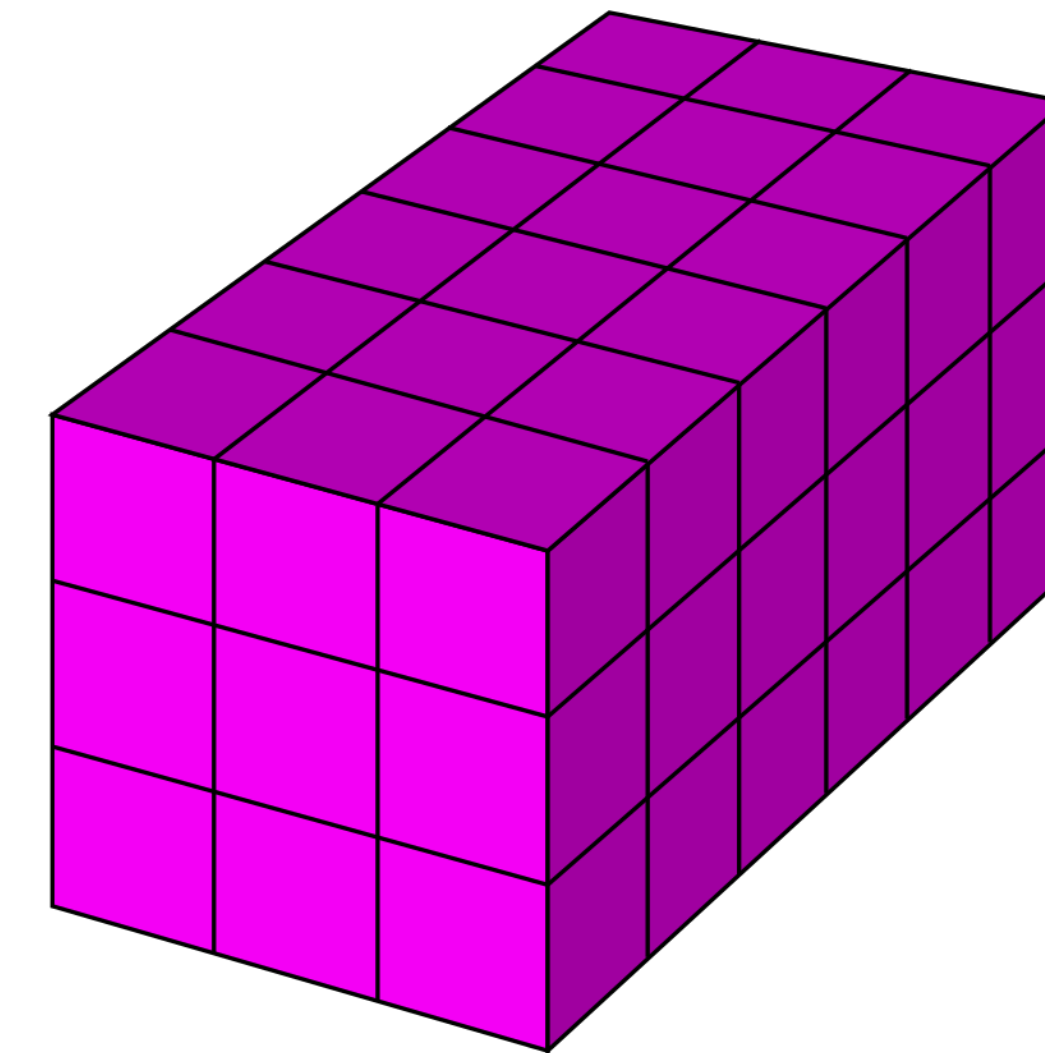
Embedding functions

- Problem
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 - $e : \mathcal{G} \rightarrow \mathbb{R}^n$
 - Attributes attached to the grid
- Non uniform sampling



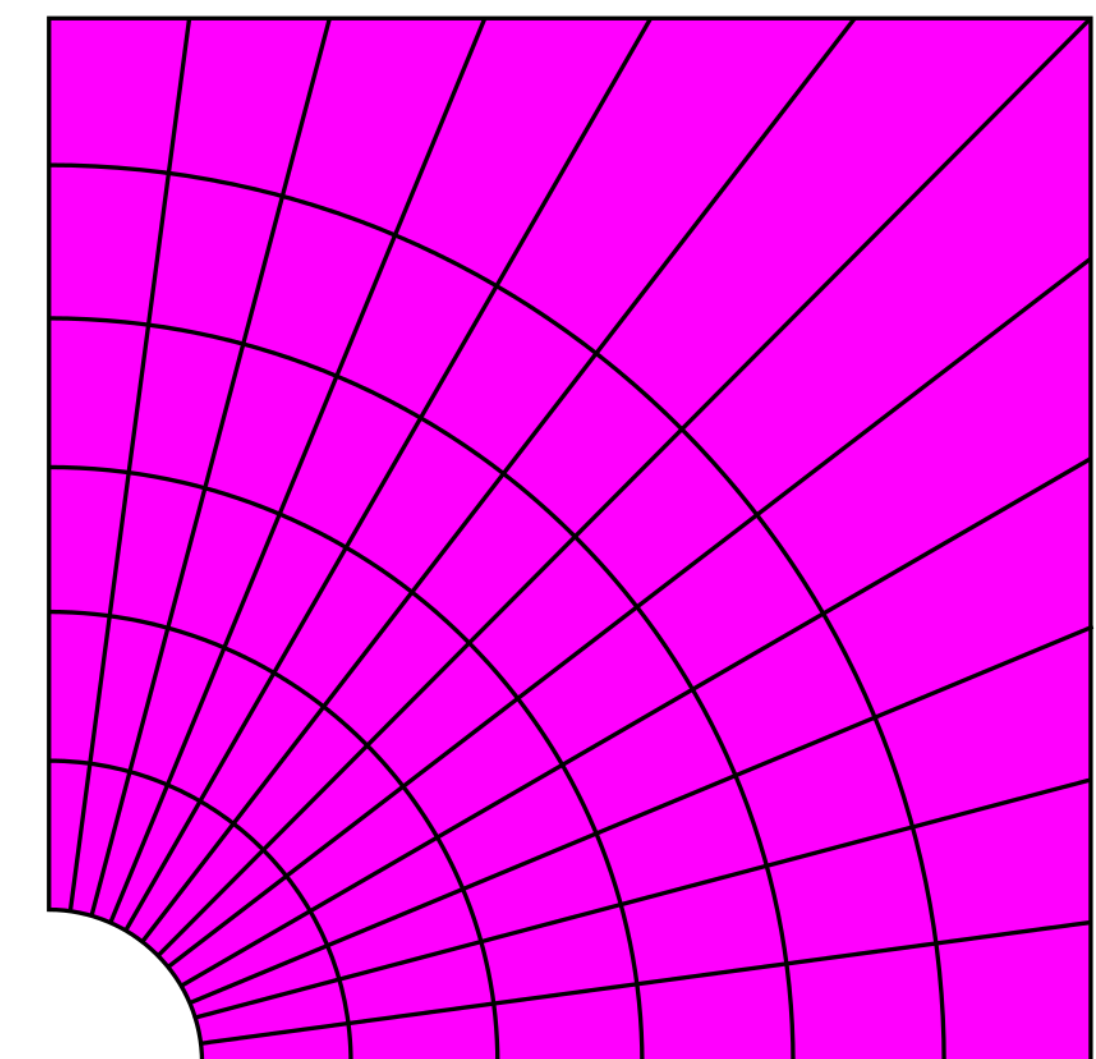
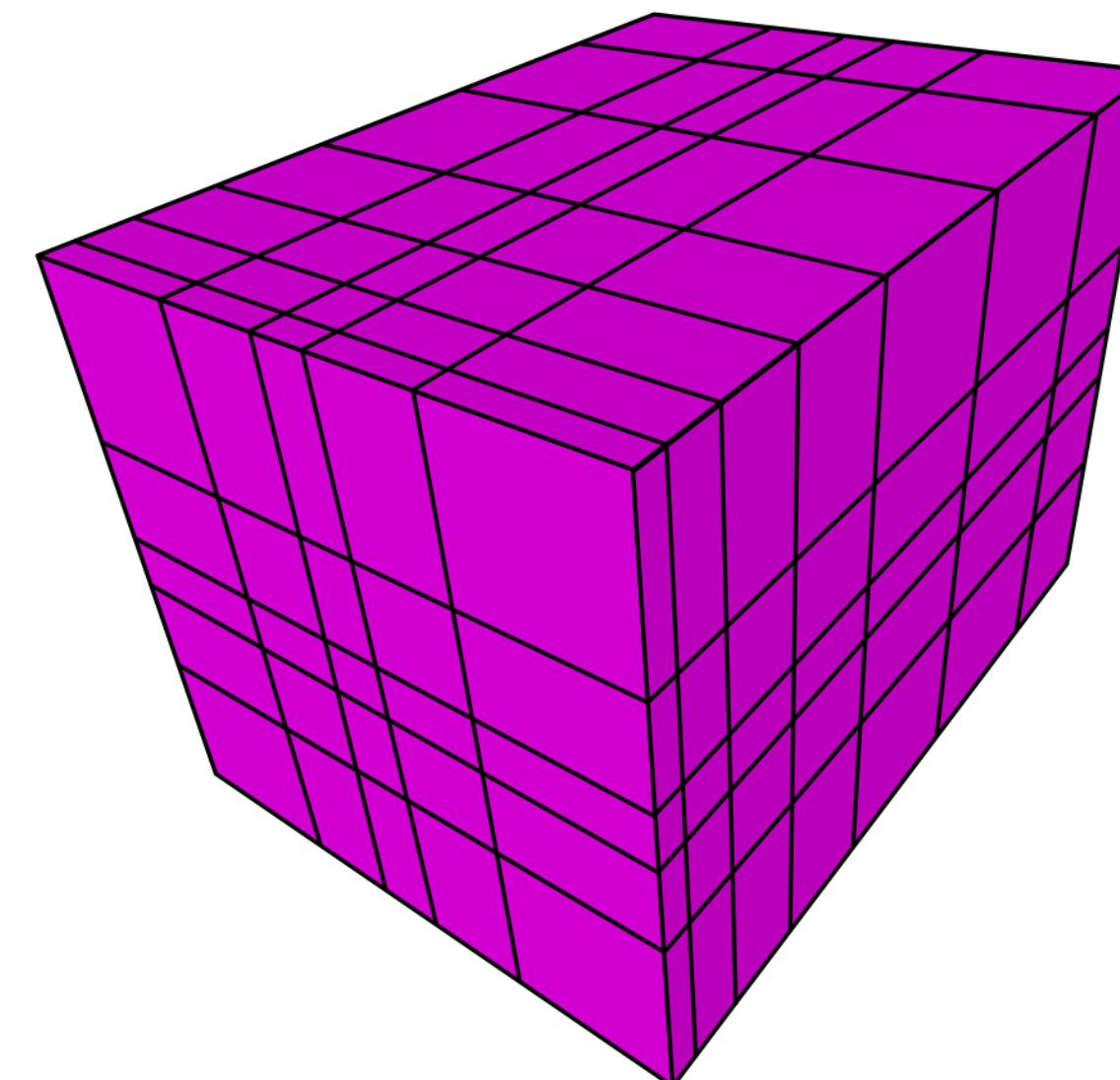
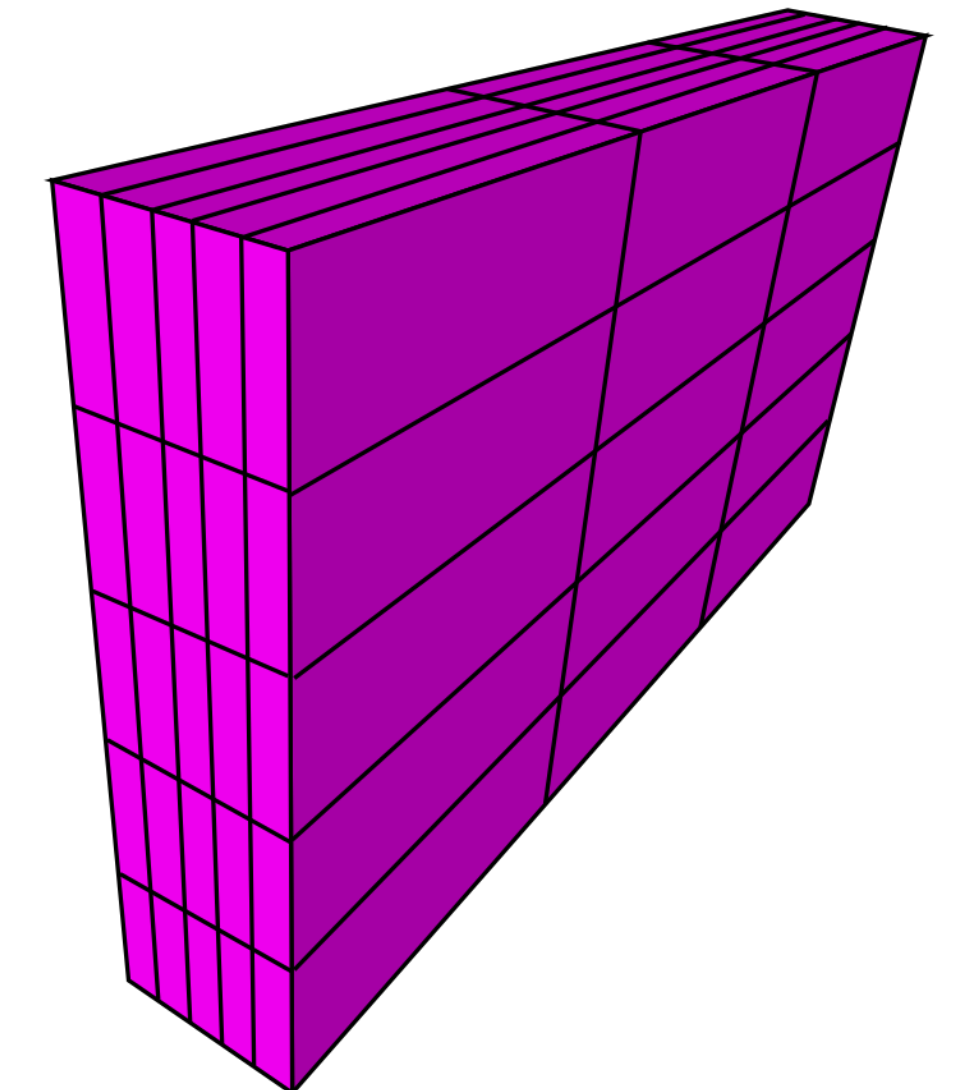
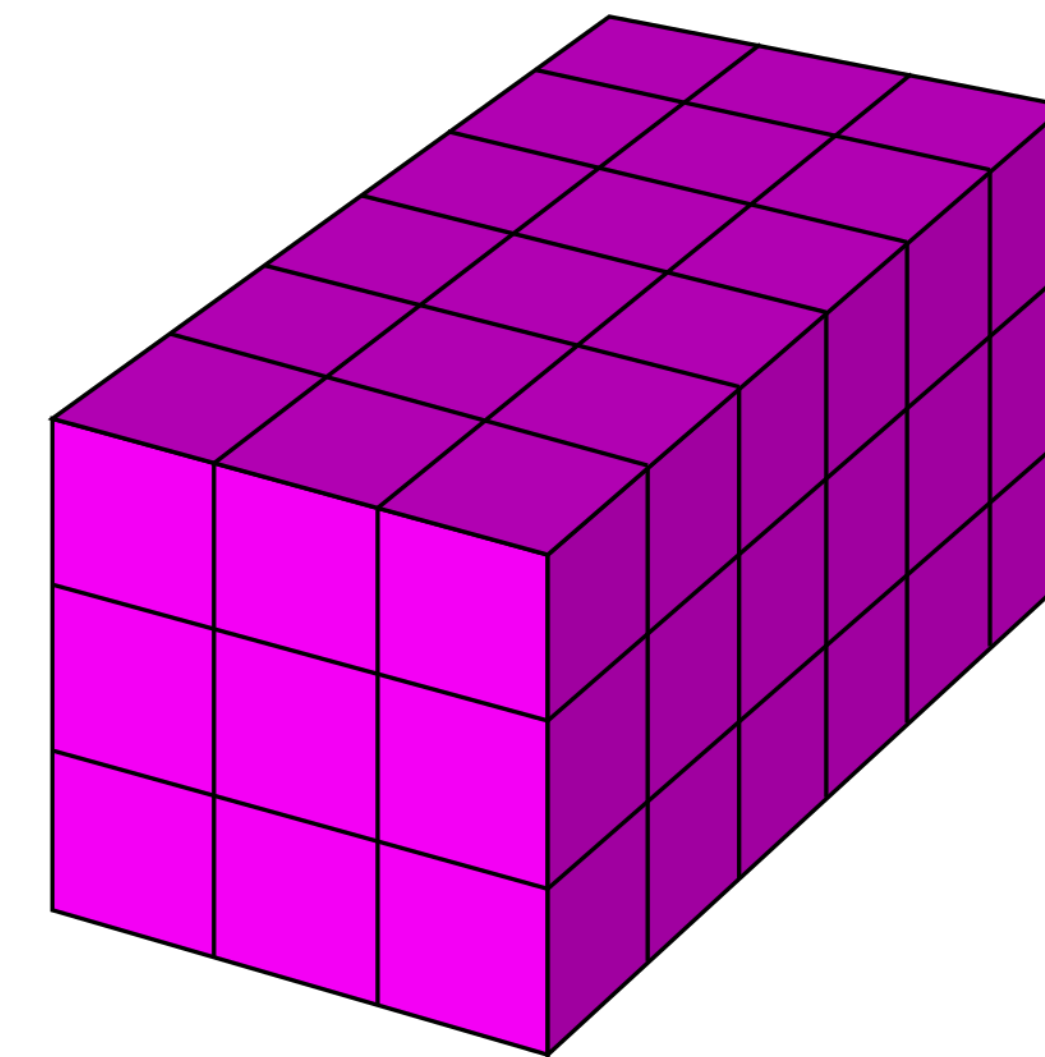
Embedding functions

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 - $e : \mathcal{G} \rightarrow \mathbb{R}^n$
 - Attributes attached to the grid
- Non uniform sampling
 - *Rectilinear* grid



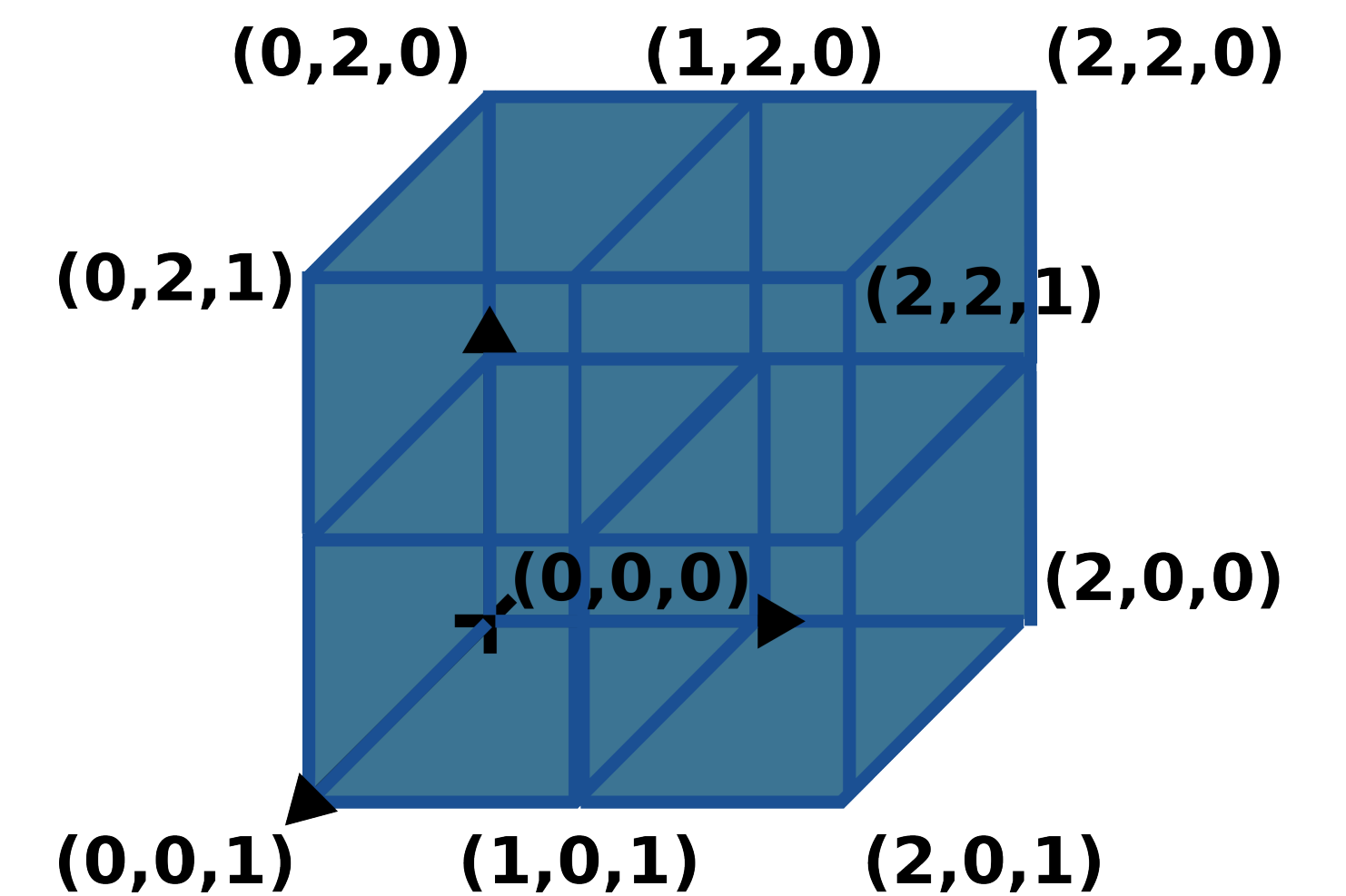
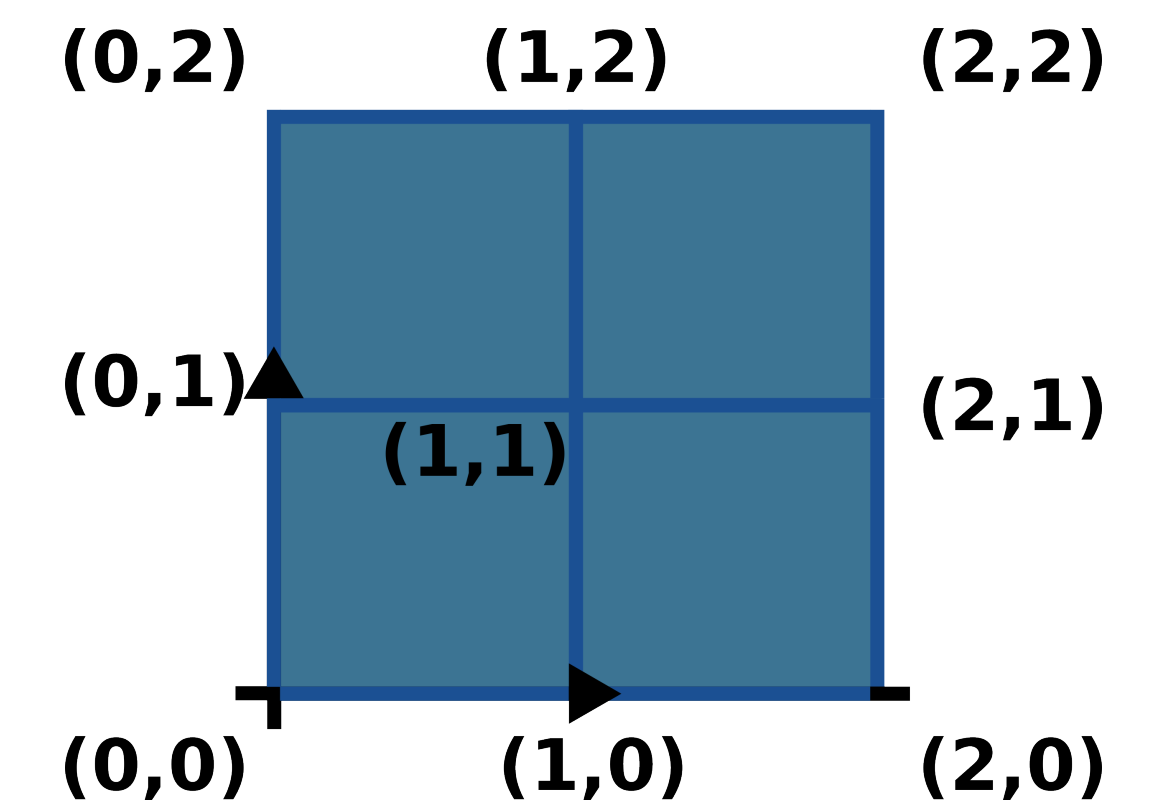
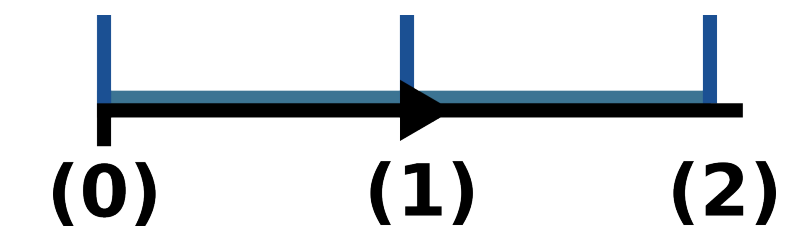
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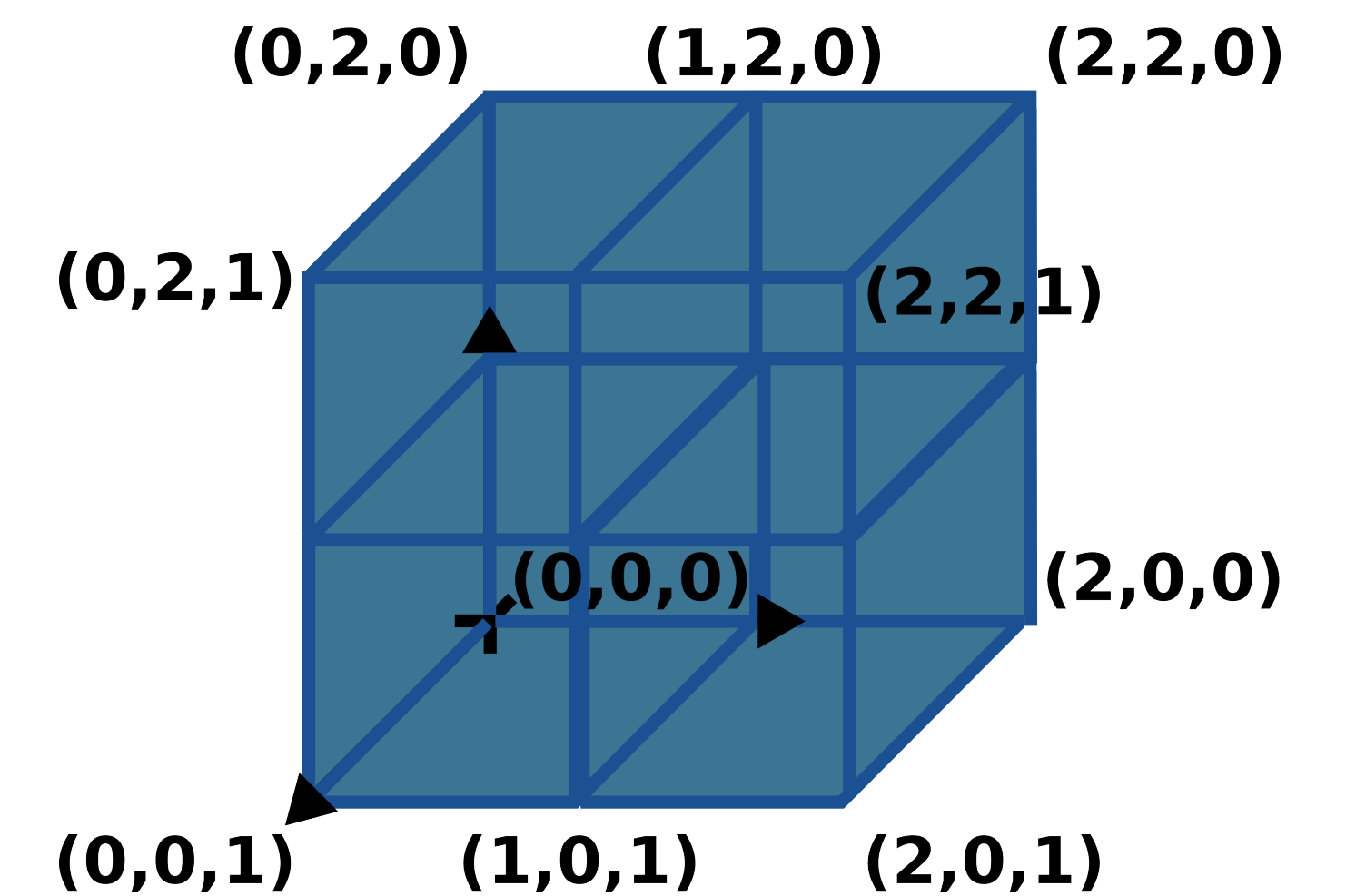
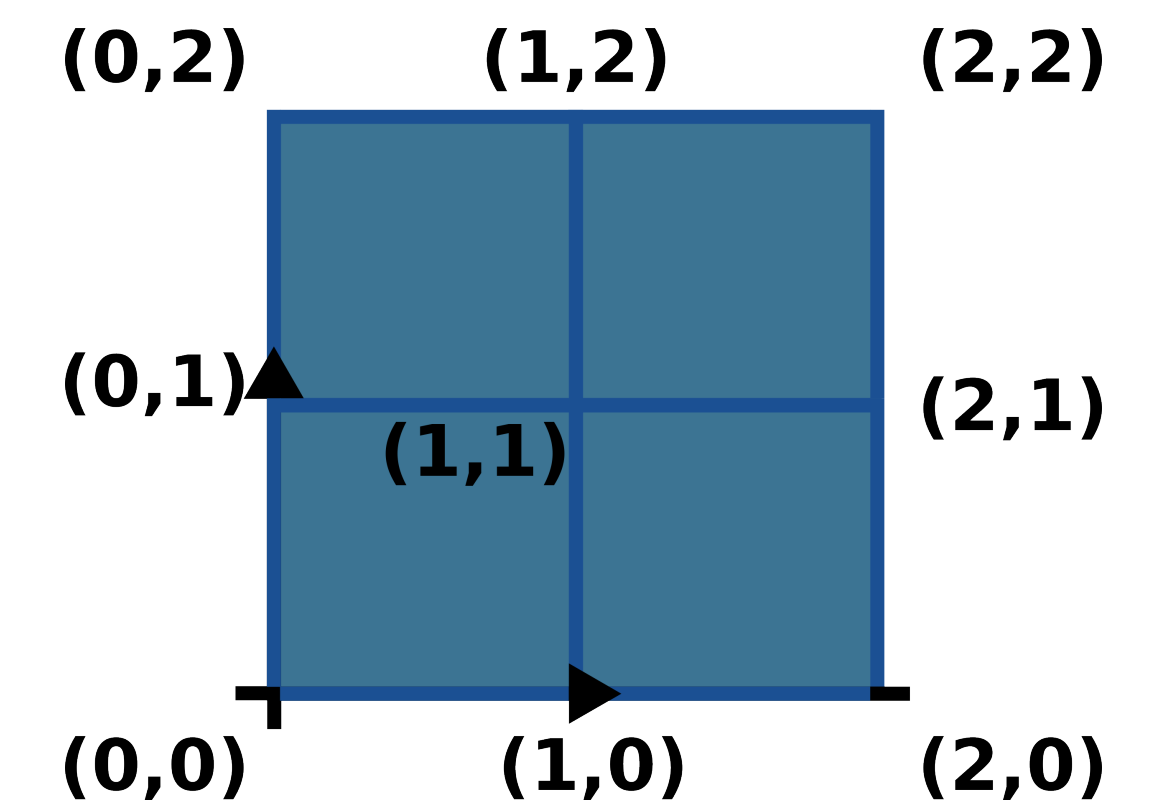
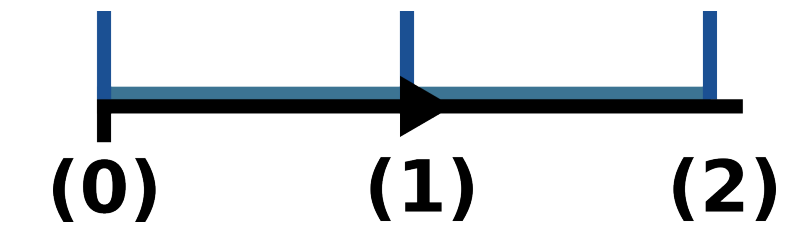
Euclidean spaces on a computer

- Notion of regular grid \mathcal{G}



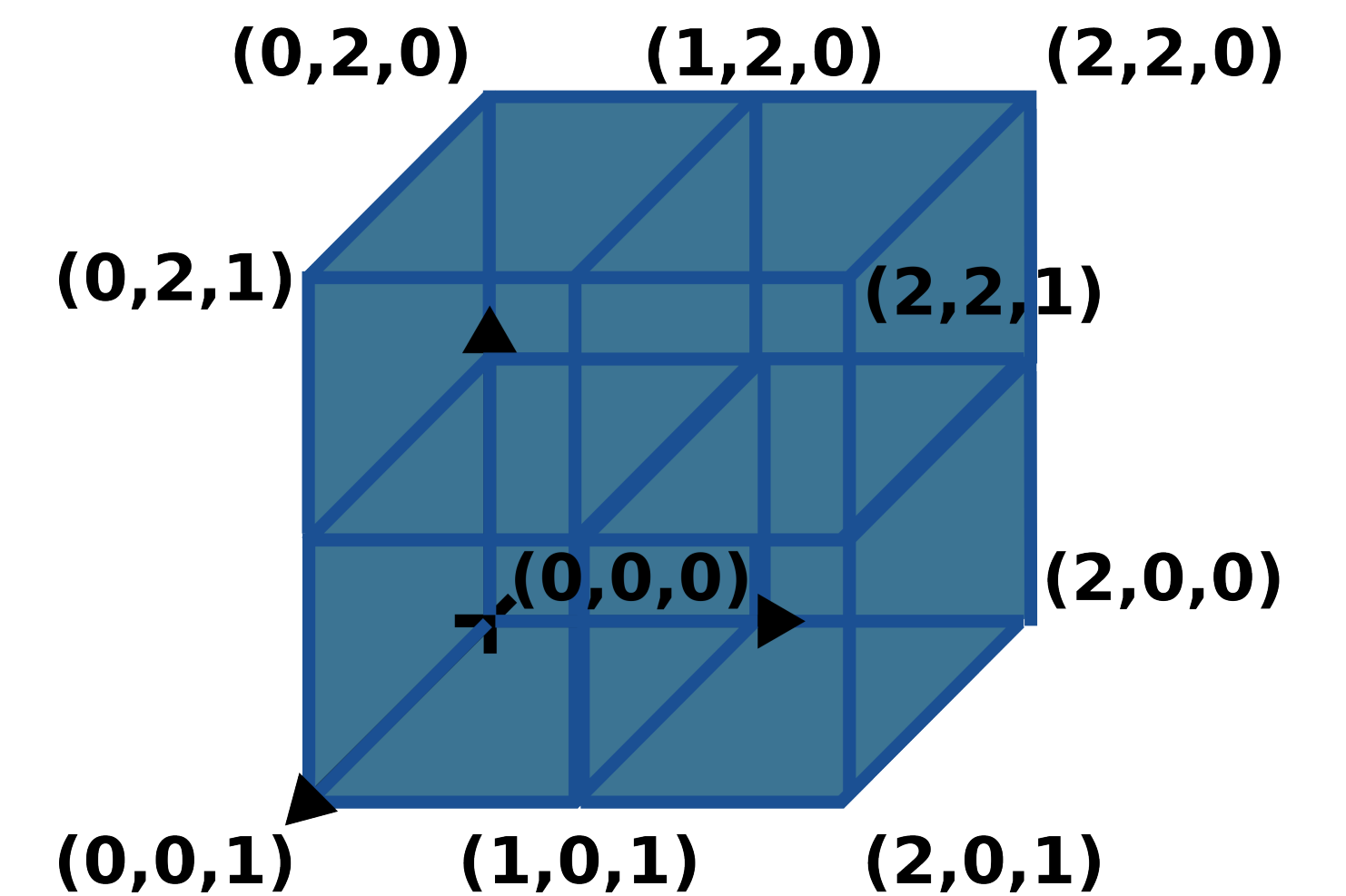
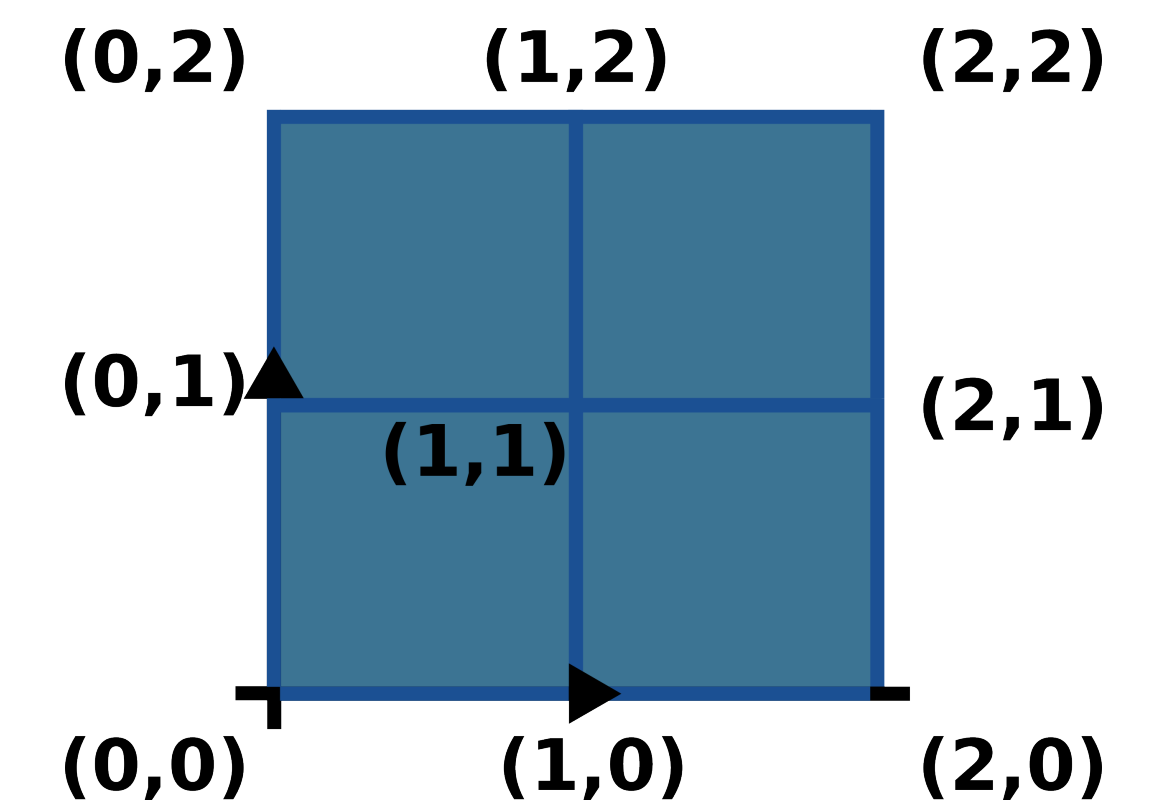
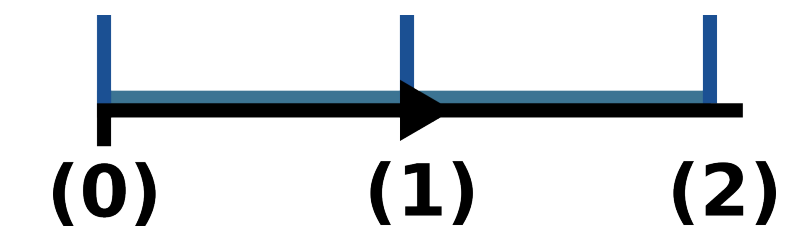
Euclidean spaces on a computer

- Notion of **regular grid** \mathcal{G}
 - Set of **vertices** (samples) in \mathbb{R}^n
 - With embedding functions to \mathbb{R}^n



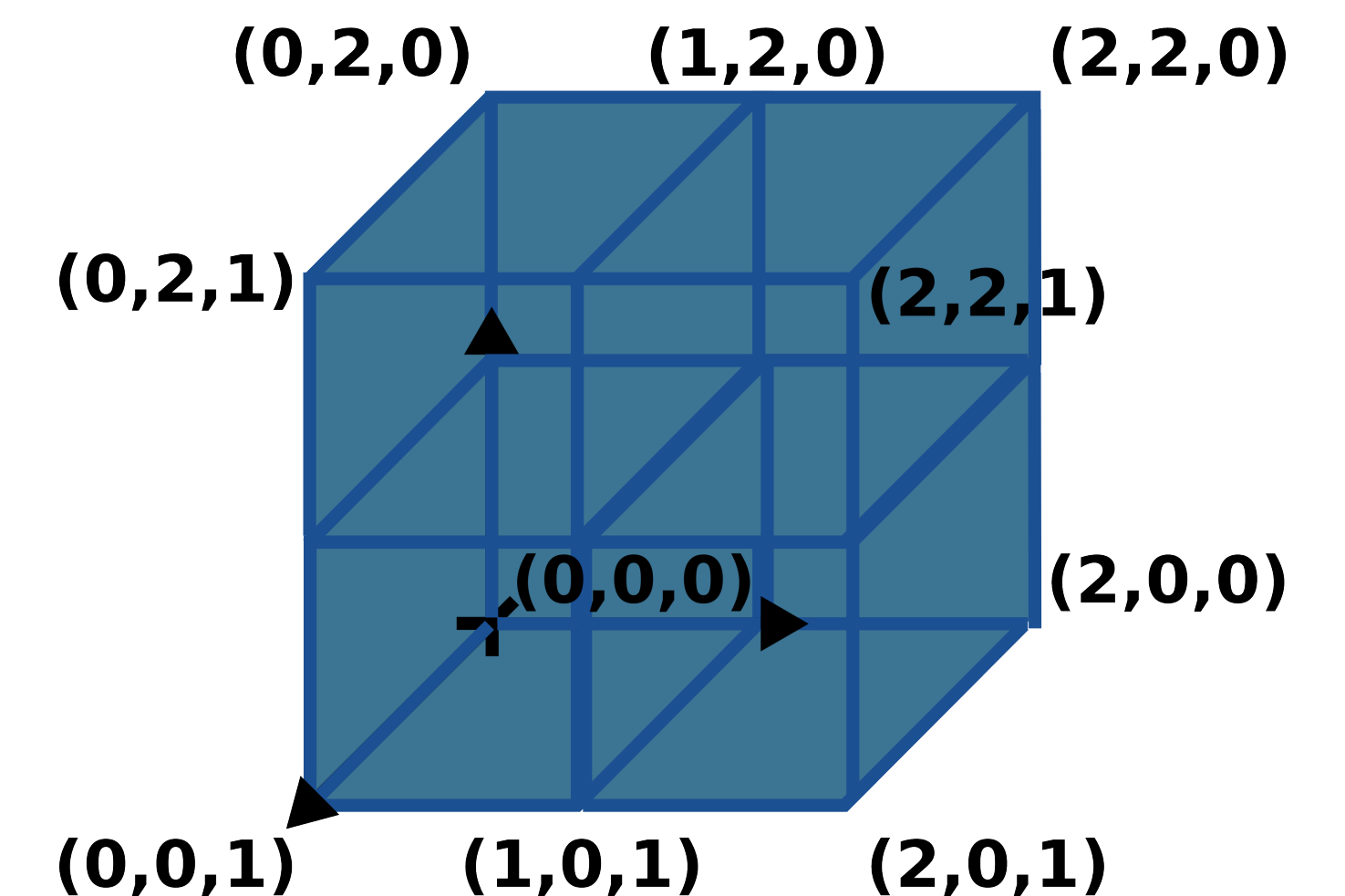
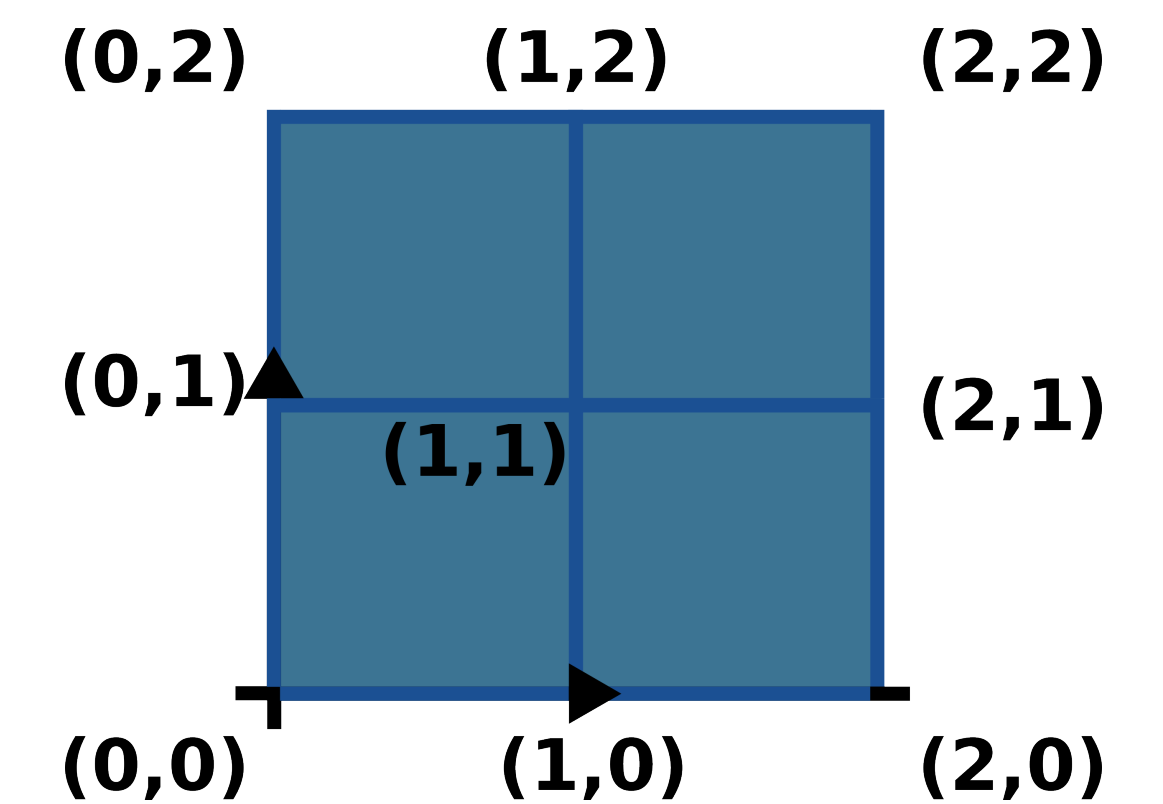
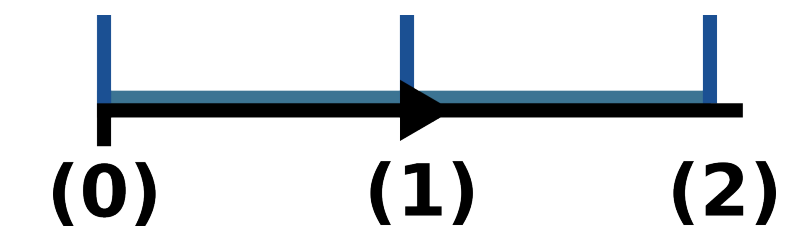
Euclidean spaces on a computer

- Notion of **regular grid** \mathcal{G}
 - Set of **vertices** (samples) in \mathbb{R}^n
 - With embedding functions to \mathbb{R}^n
 - **Connectivity**



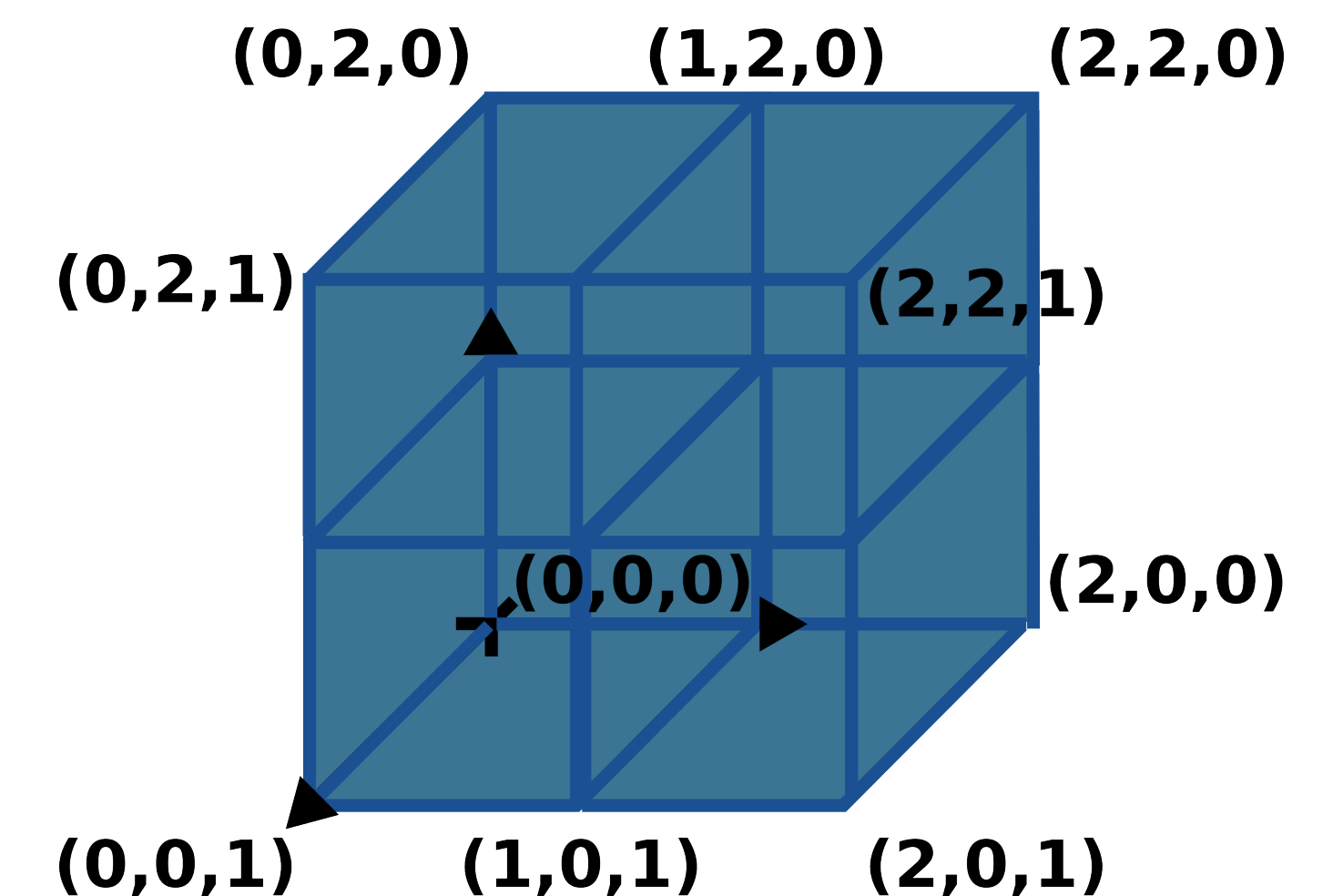
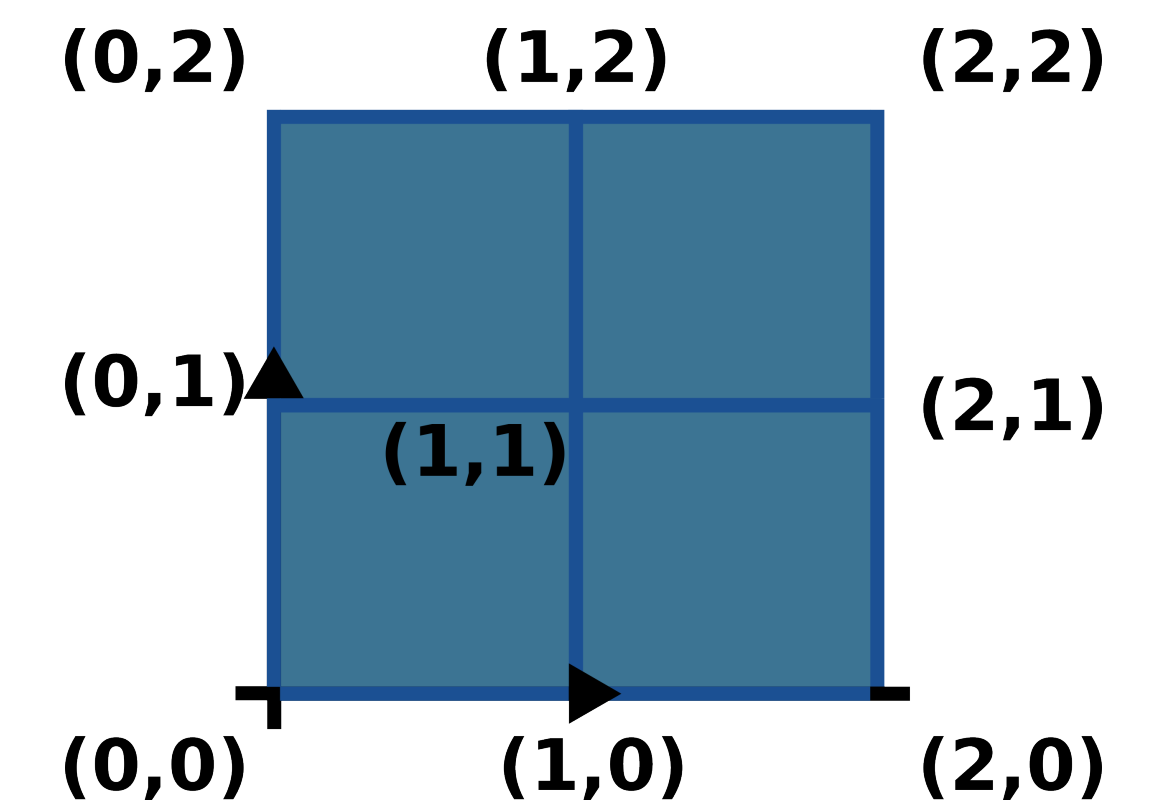
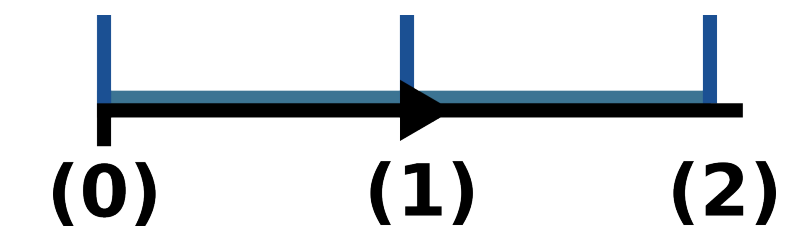
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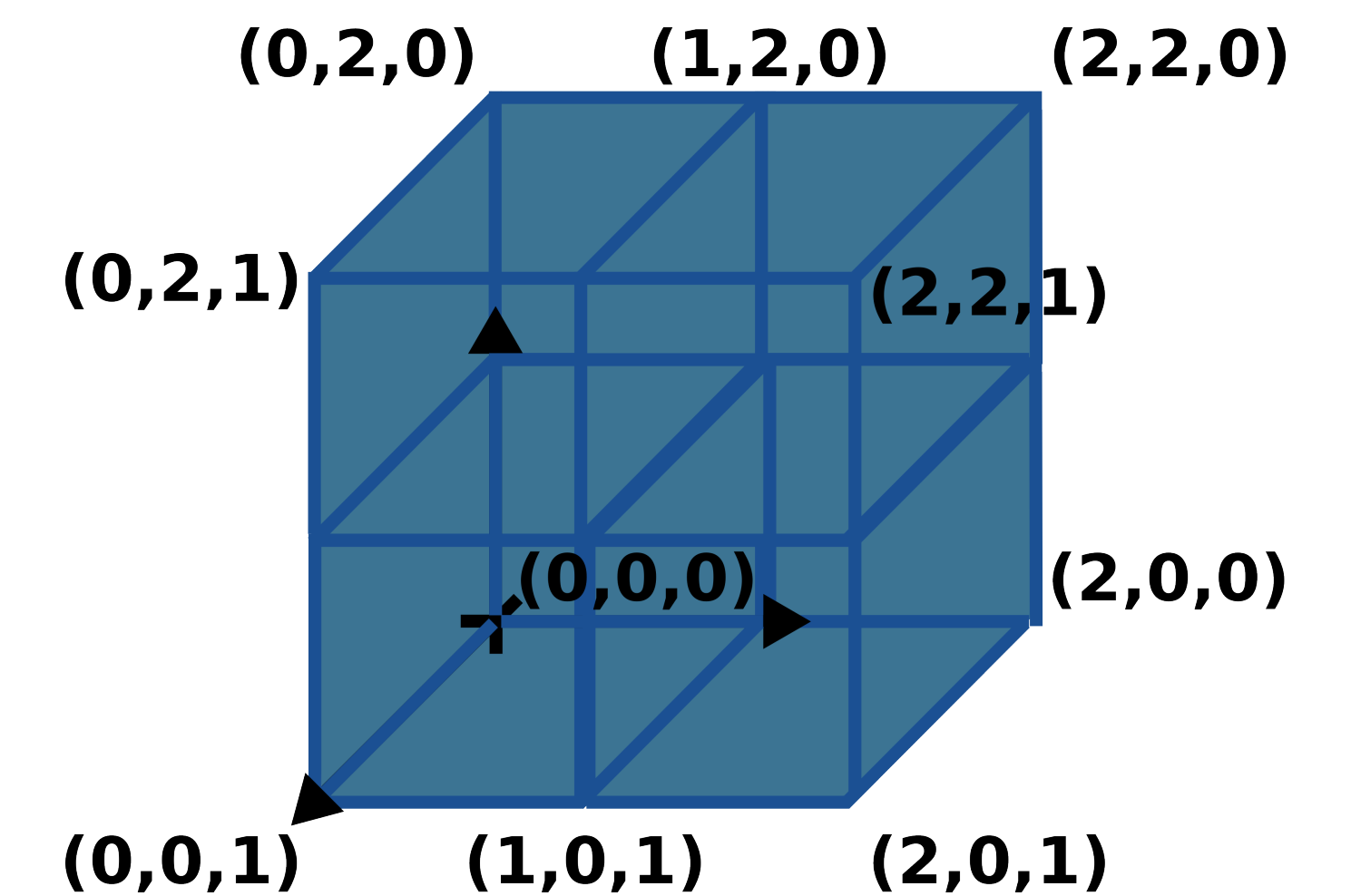
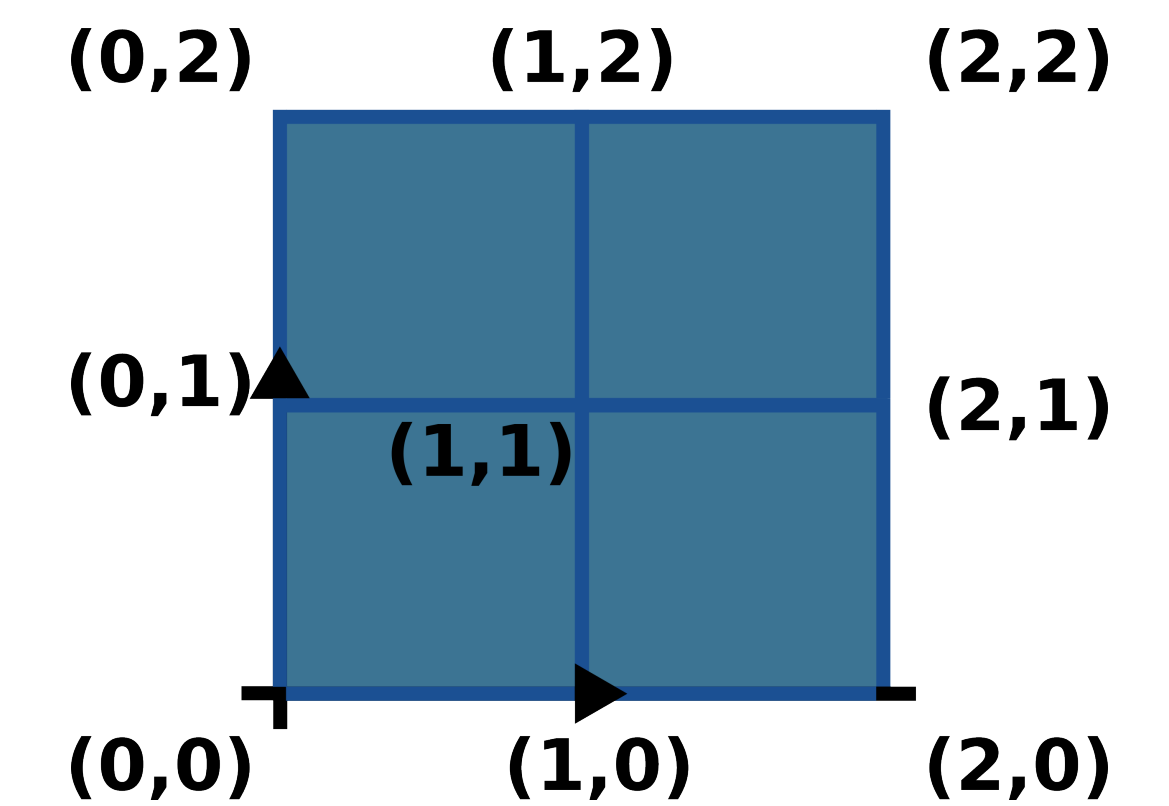
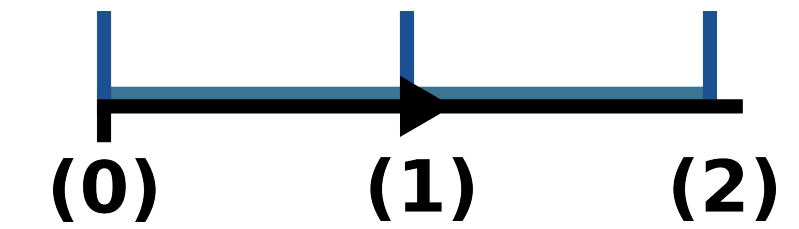
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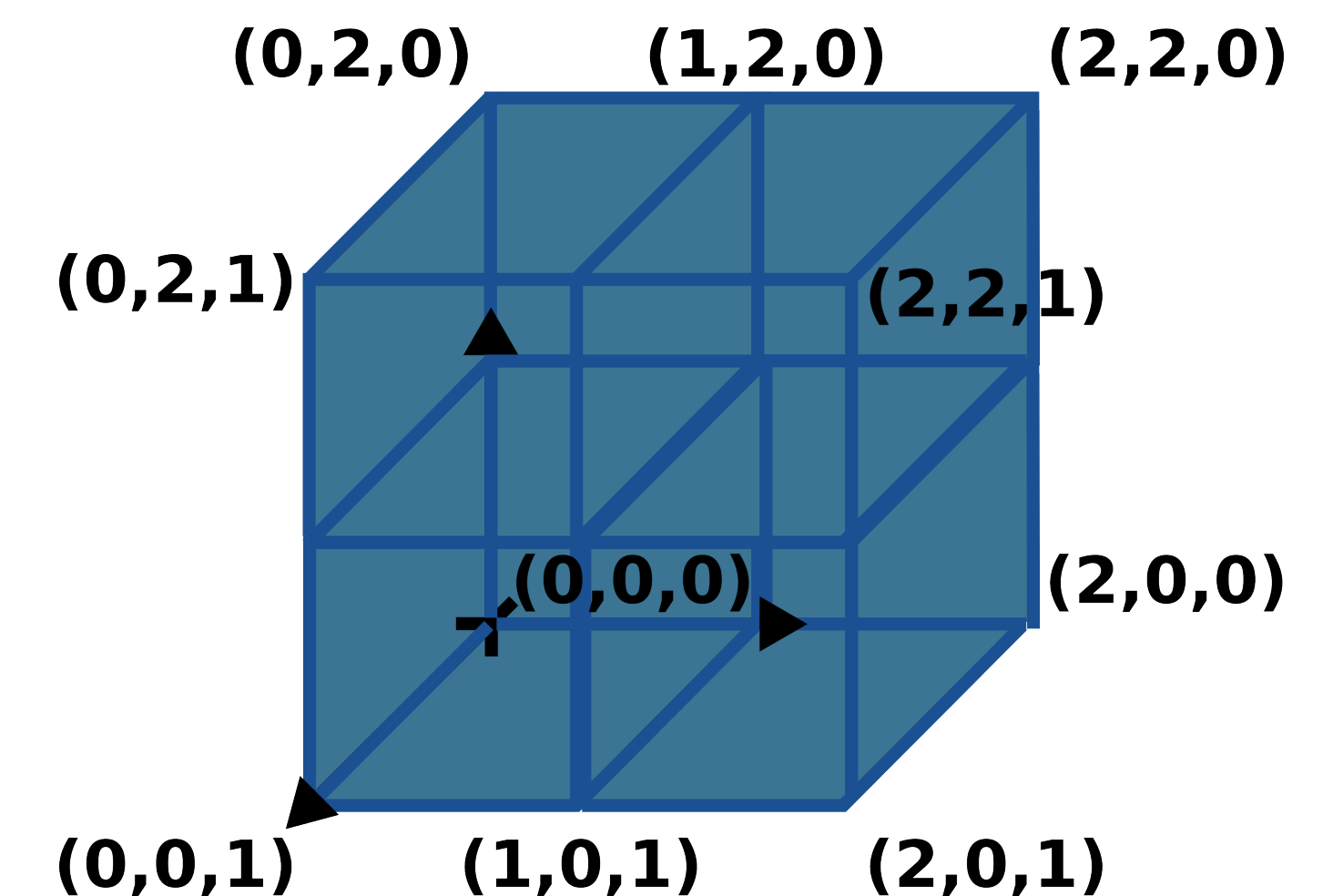
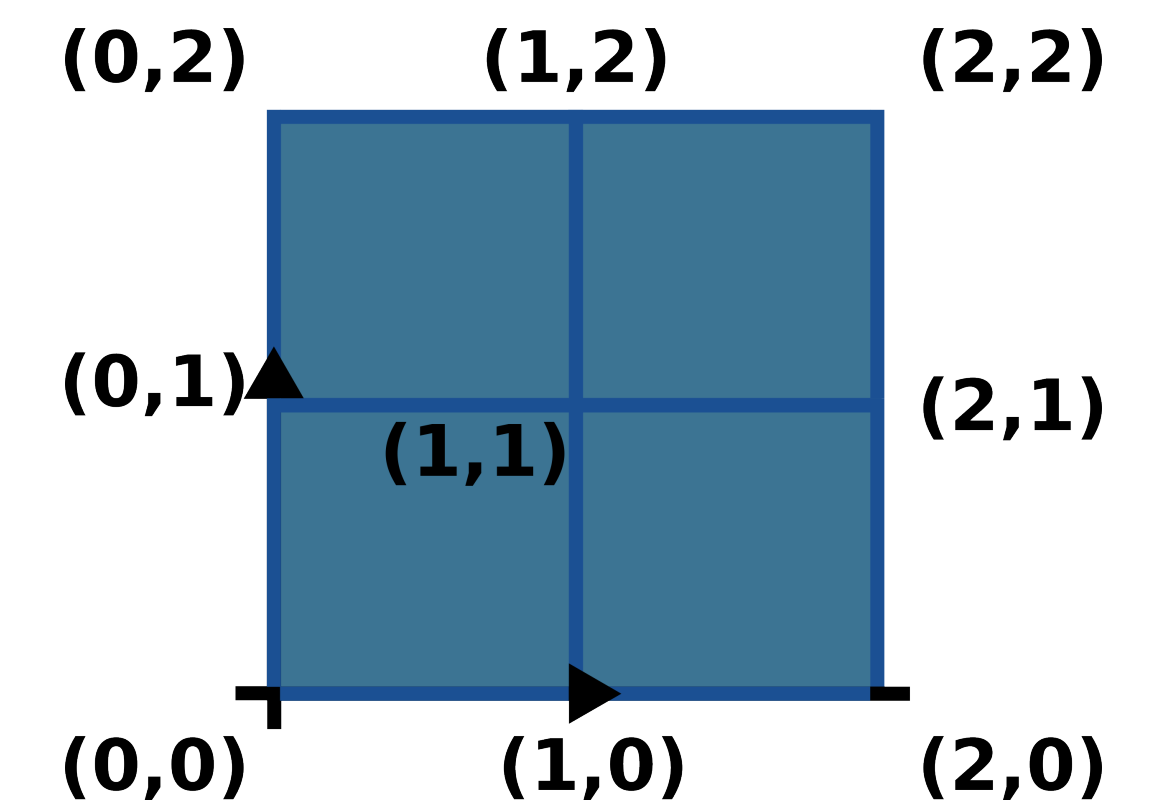
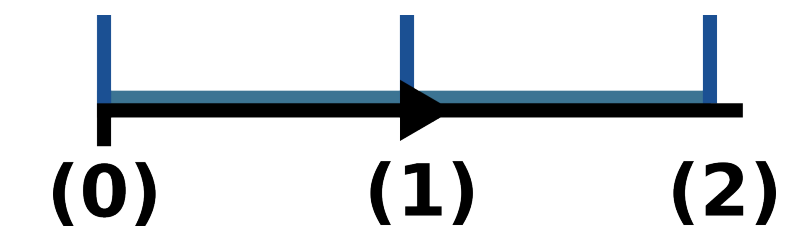
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 - What's missing?



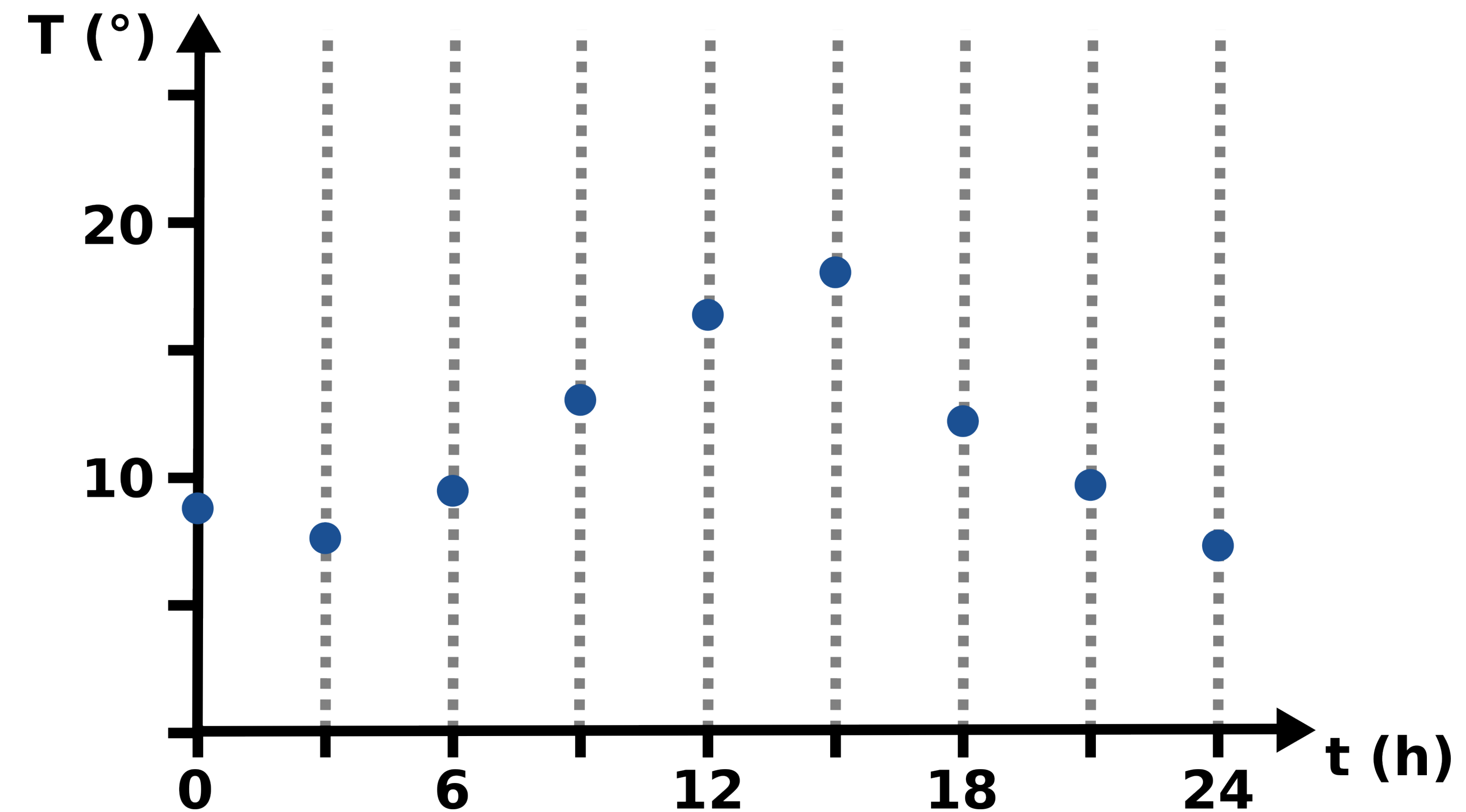
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- What's missing?
 - The **interpolation** scheme



Interpolants for regular grids

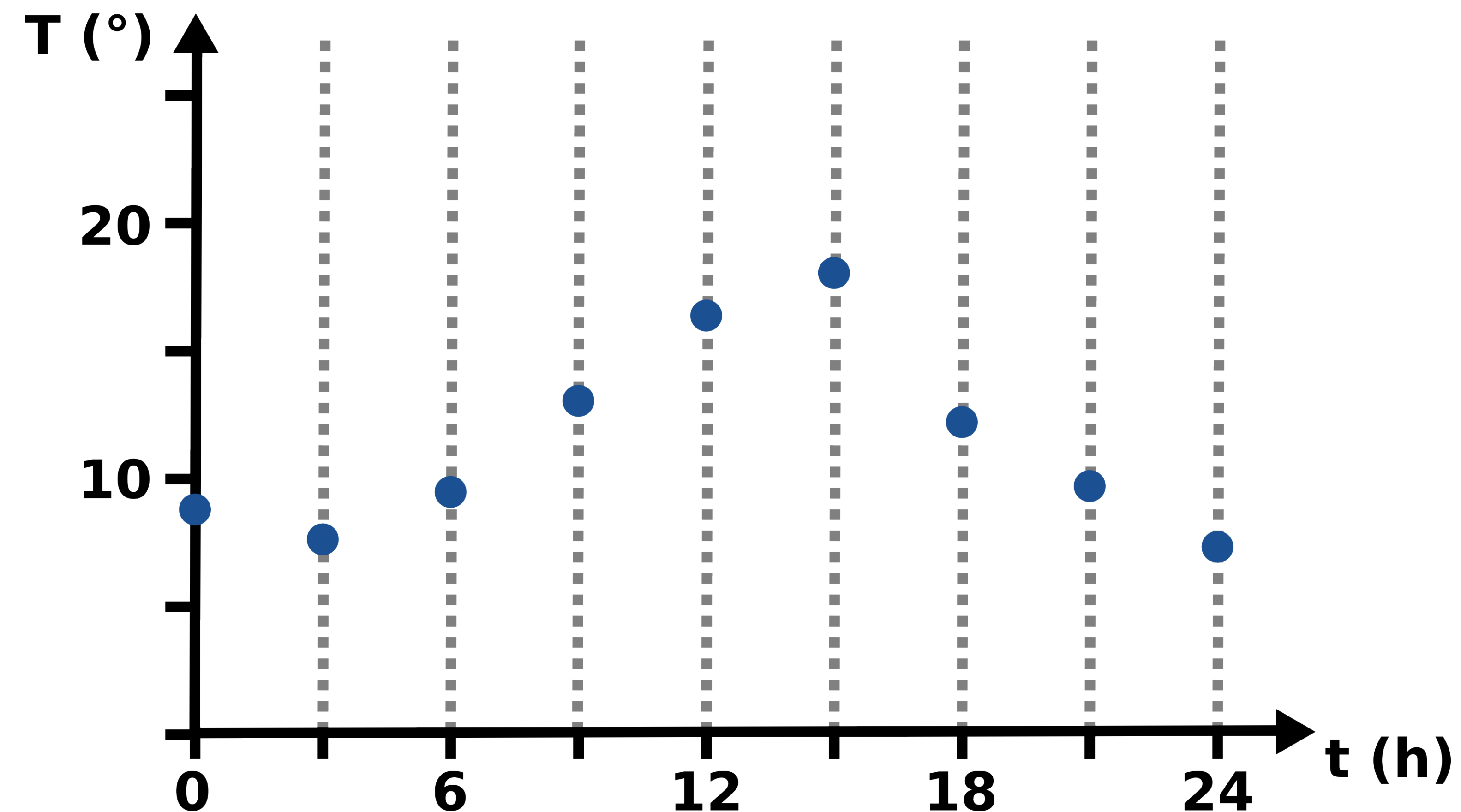
$$f : [0, 24] \rightarrow \mathbb{R}$$



Interpolants for regular grids

- Given values on the vertices

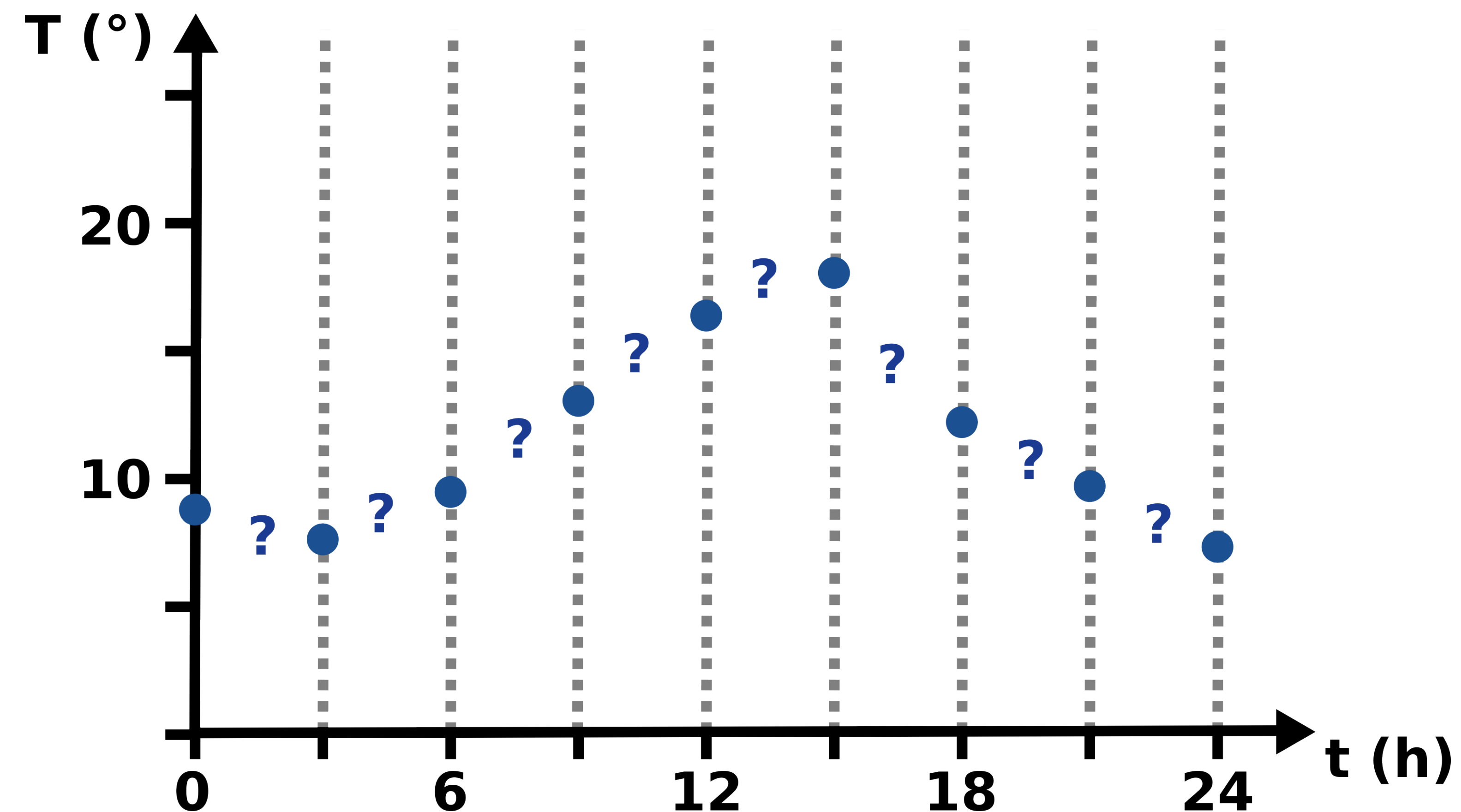
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Interpolants for regular grids

- Given values on the vertices
 - How can we deduce the values within the unit cells?

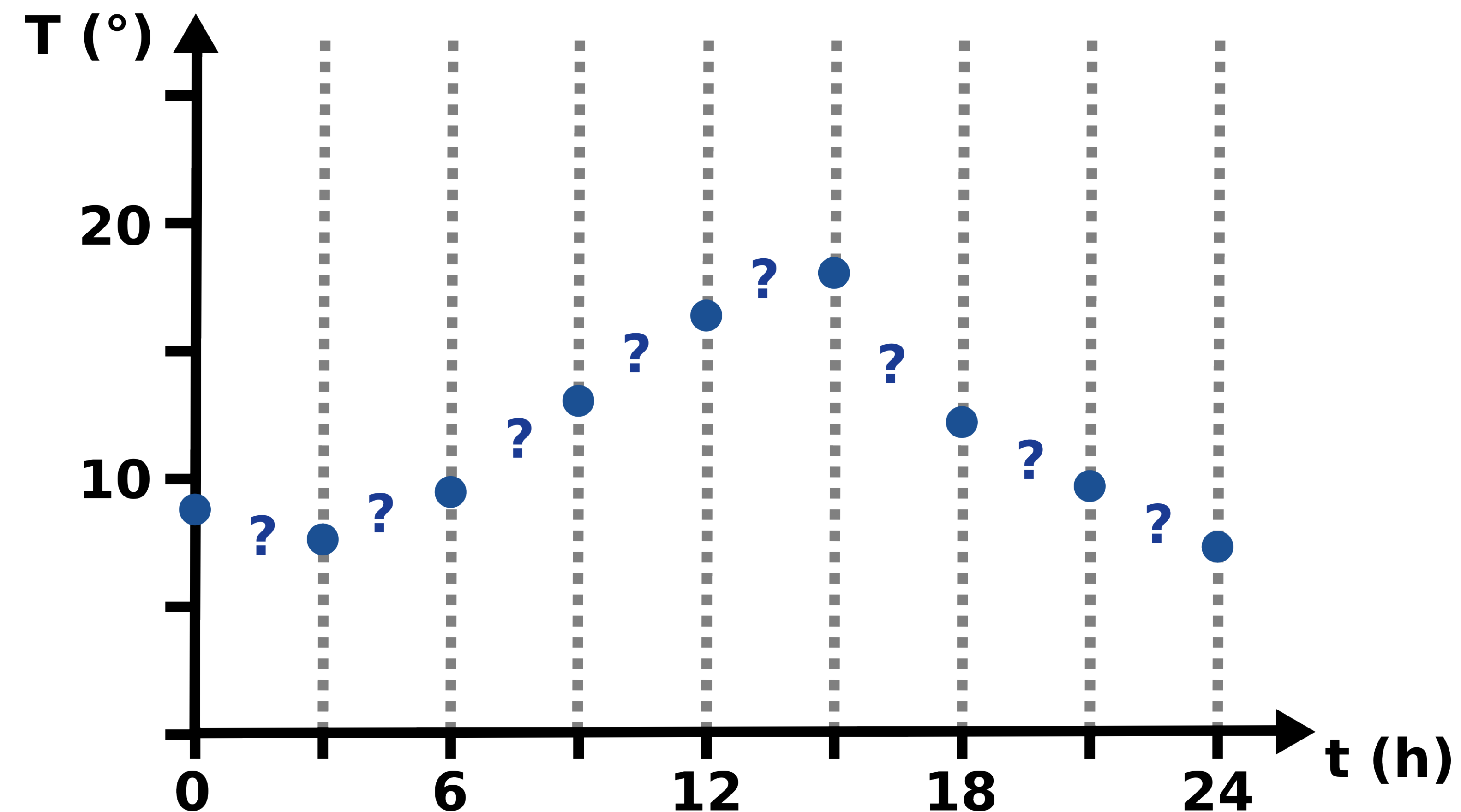
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Interpolants for regular grids

- Given values on the vertices
 - How can we deduce the values within the unit cells?
- For regular grids of \mathbb{R}
 - $\mathbb{I} :]0, 1[\rightarrow]0, 1[$

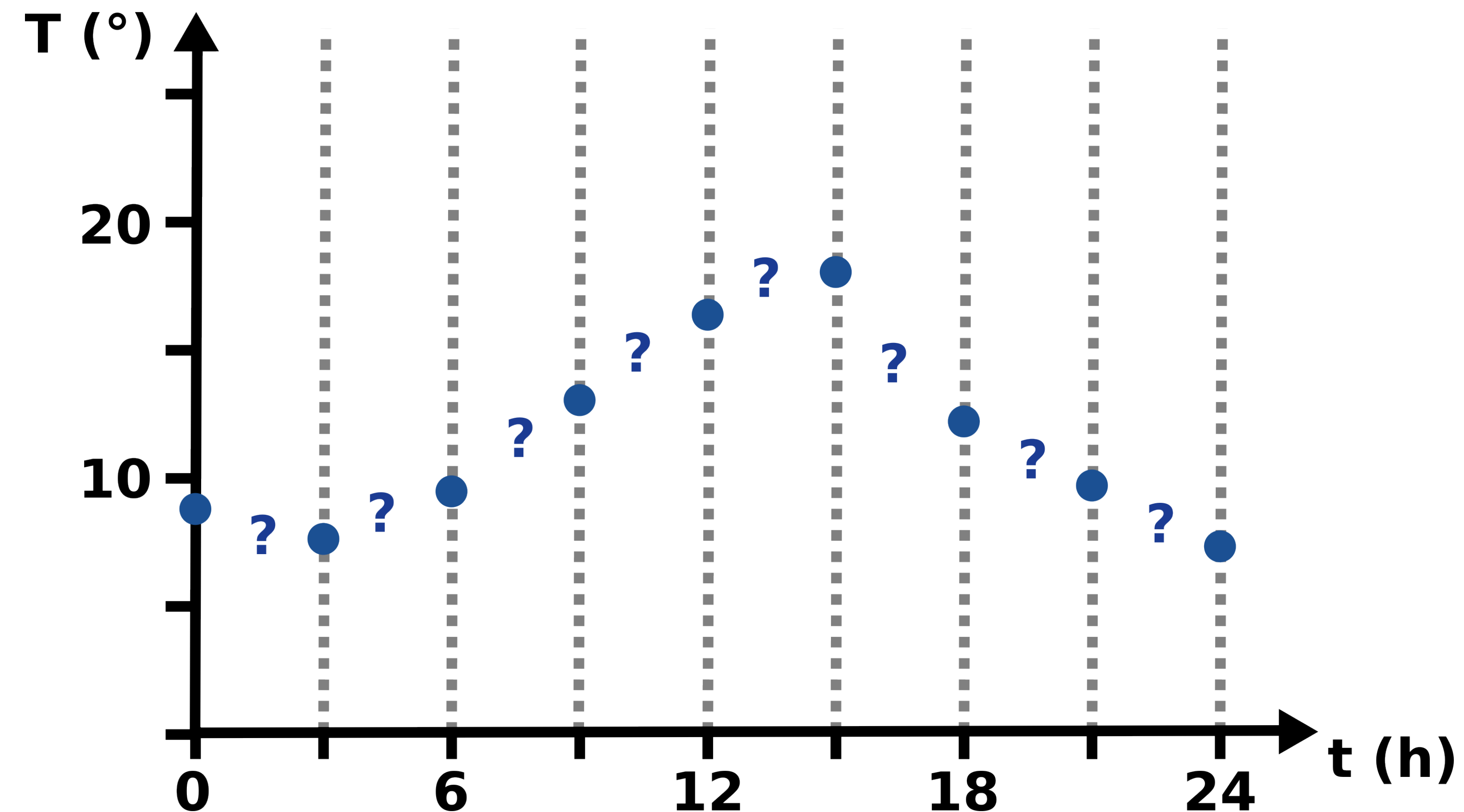
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Interpolants for regular grids

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$$f : [0, 24] \rightarrow \mathbb{R}$$

- For regular grids of \mathbb{R}

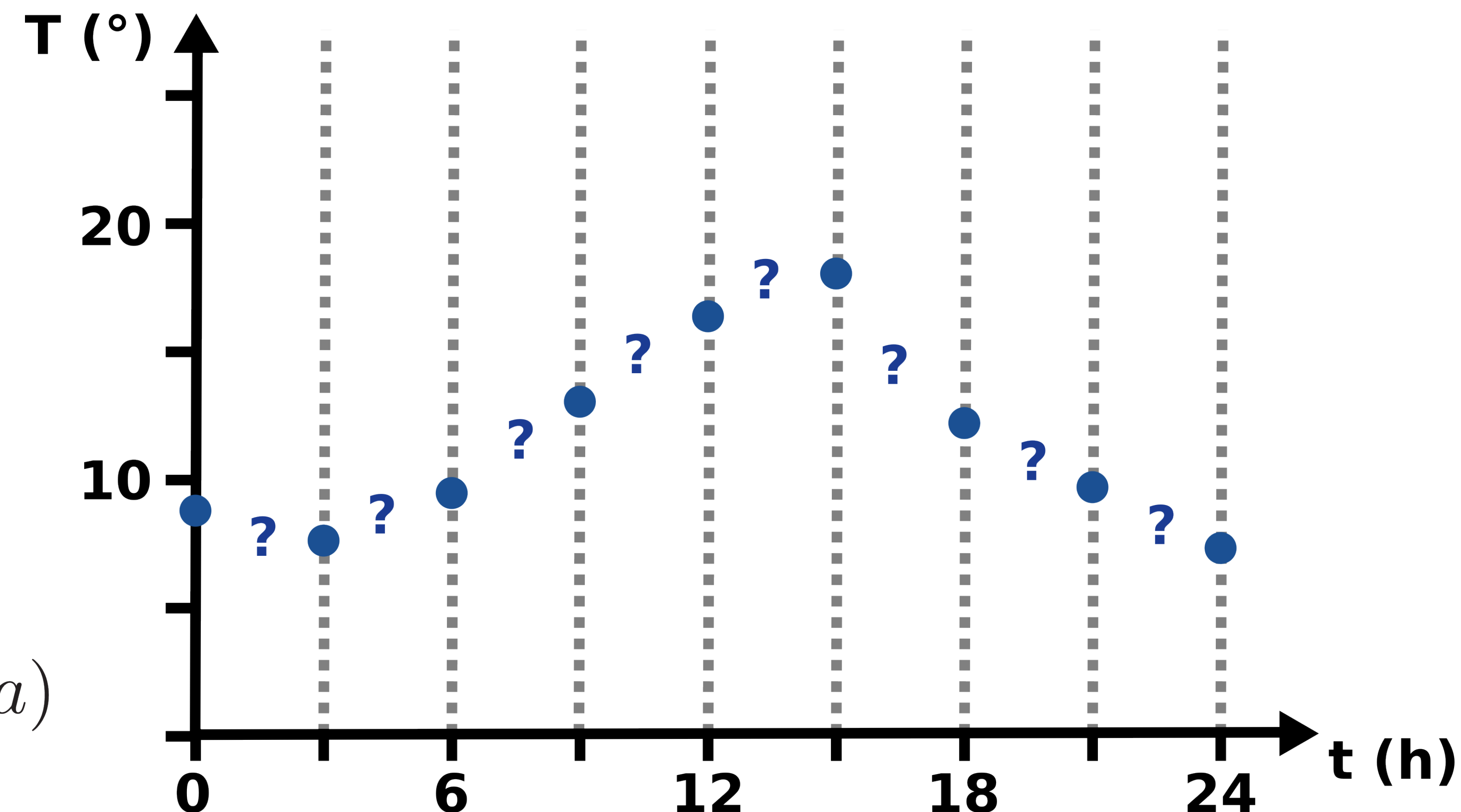
- $\mathbb{I} :]0, 1[\rightarrow]0, 1[$

- $\lim_{x \rightarrow 0} \mathbb{I}(x) = 0$

- $\lim_{x \rightarrow 1} \mathbb{I}(x) = 1$

- $\forall x \in [a, b]$

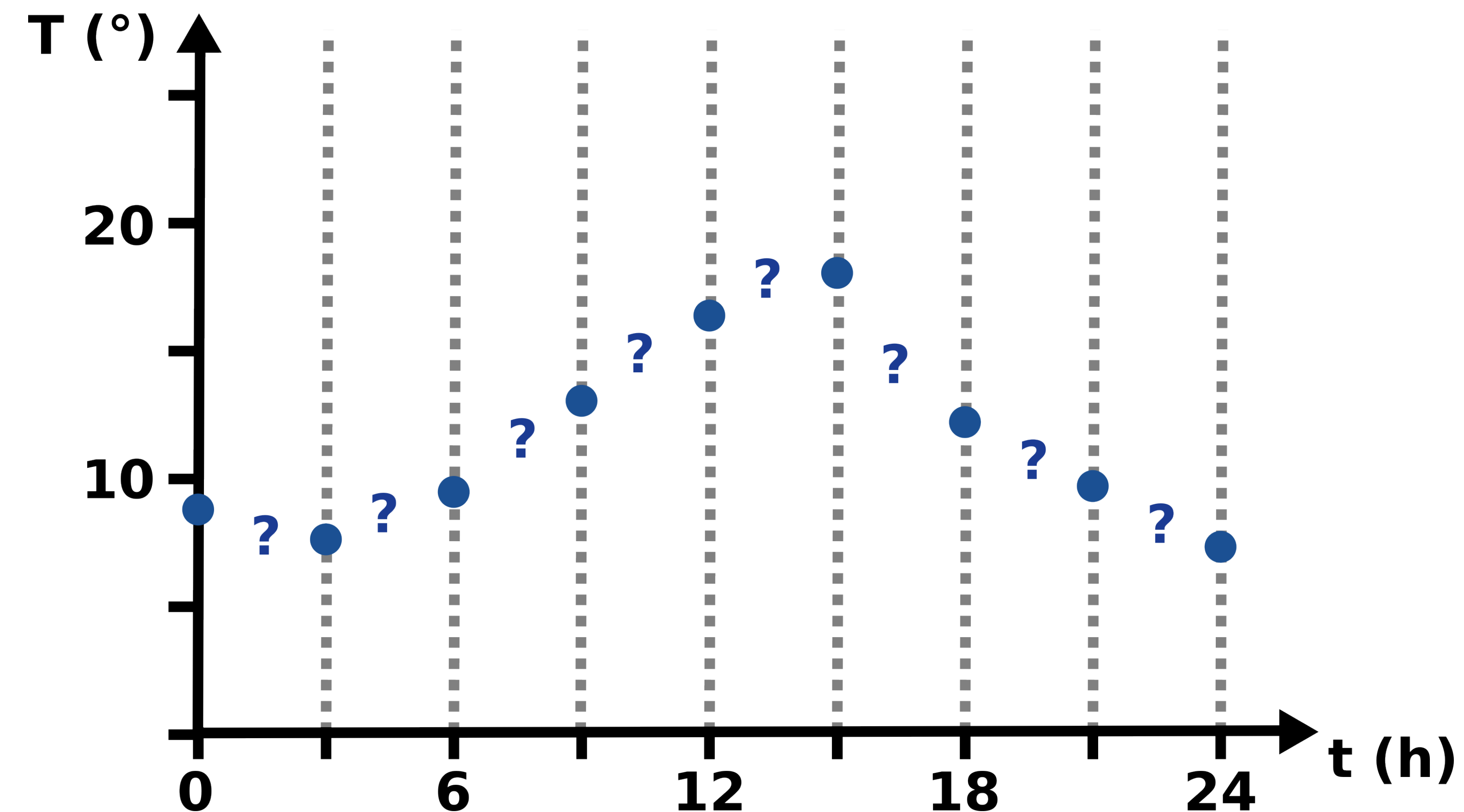
- $f(x) = \mathbb{I}\left(\frac{x - a}{b - a}\right) (f(b) - f(a)) + f(a)$



Interpolants for regular grids

- Interpolation function

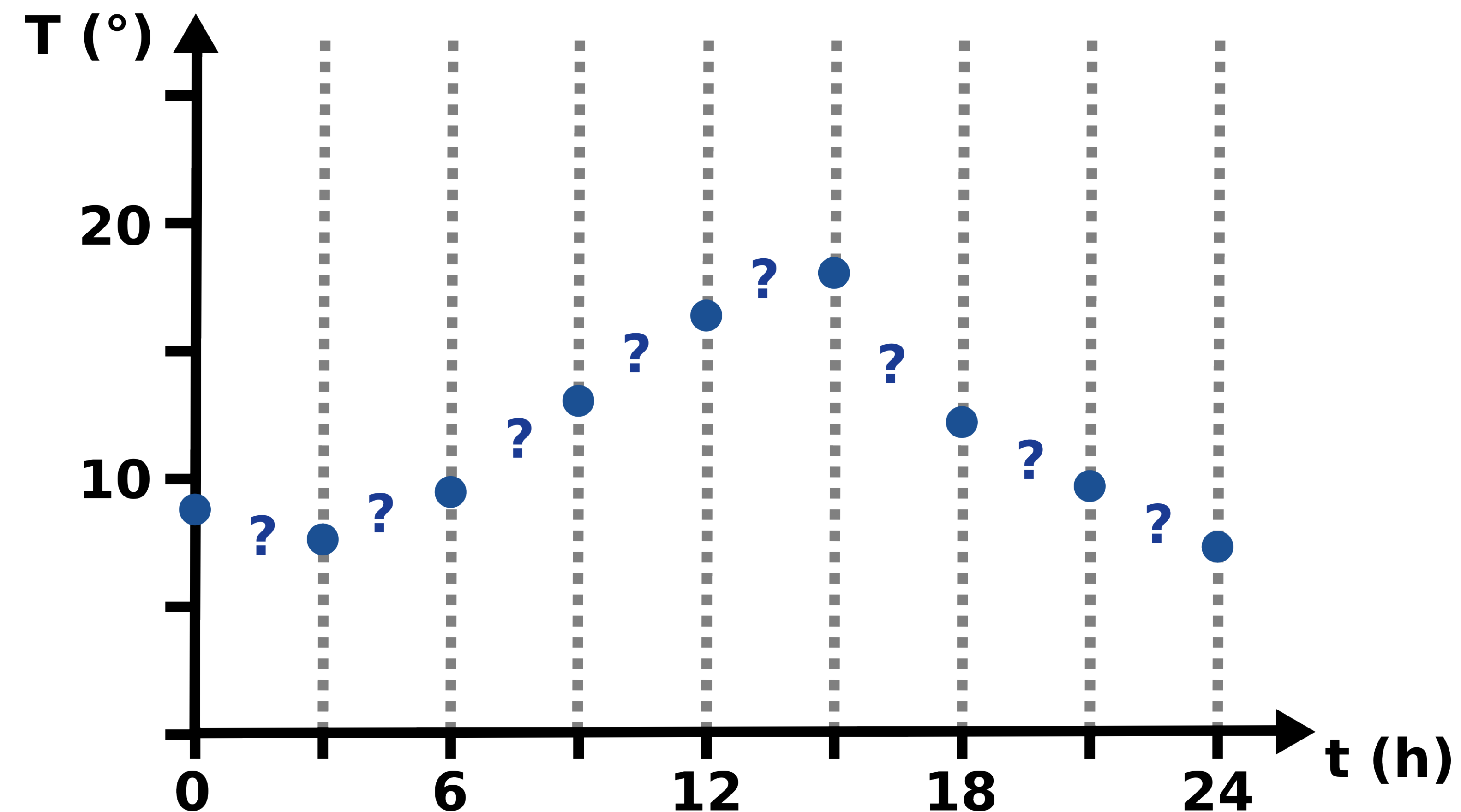
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Interpolants for regular grids

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 - Given with the data

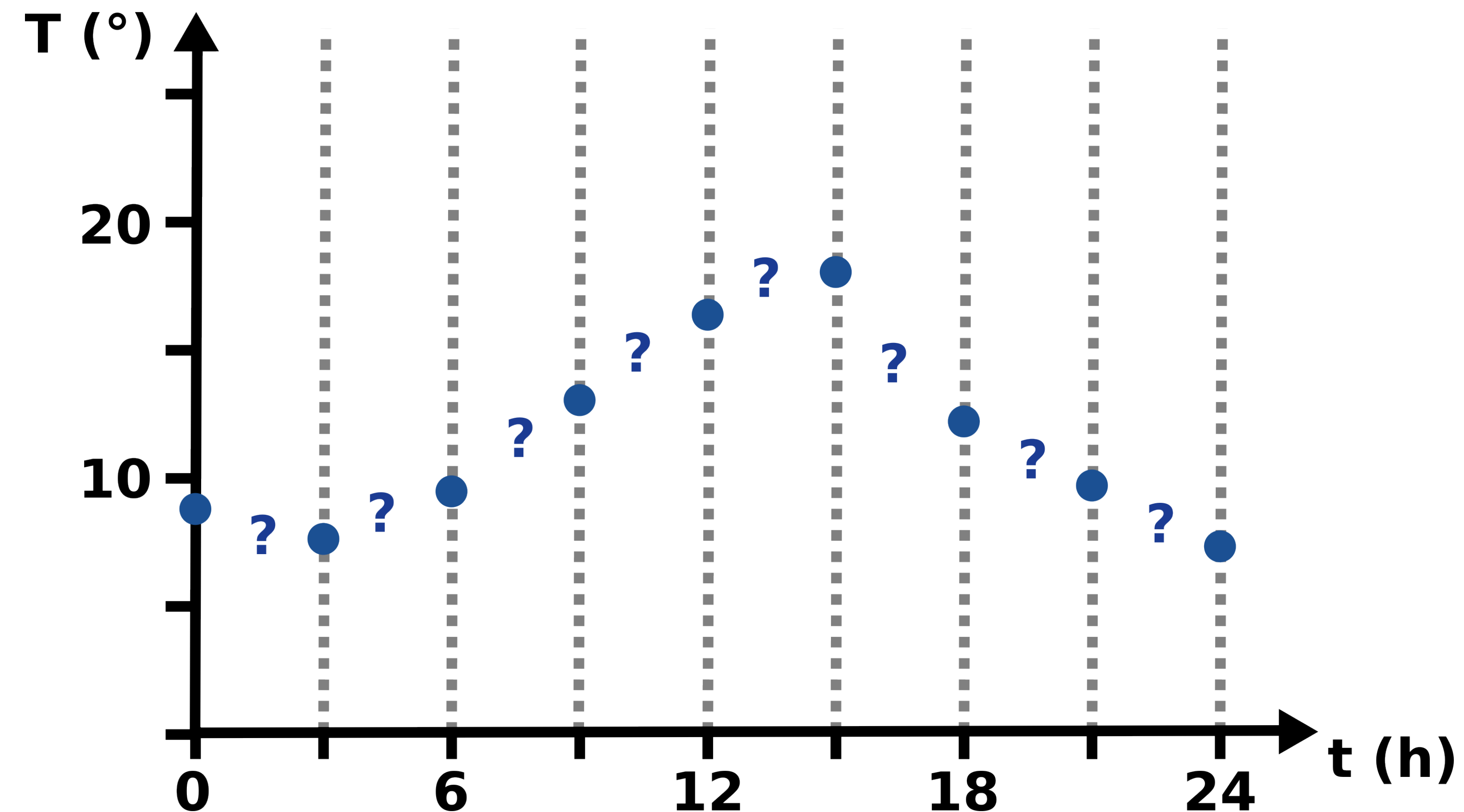
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Interpolants for regular grids

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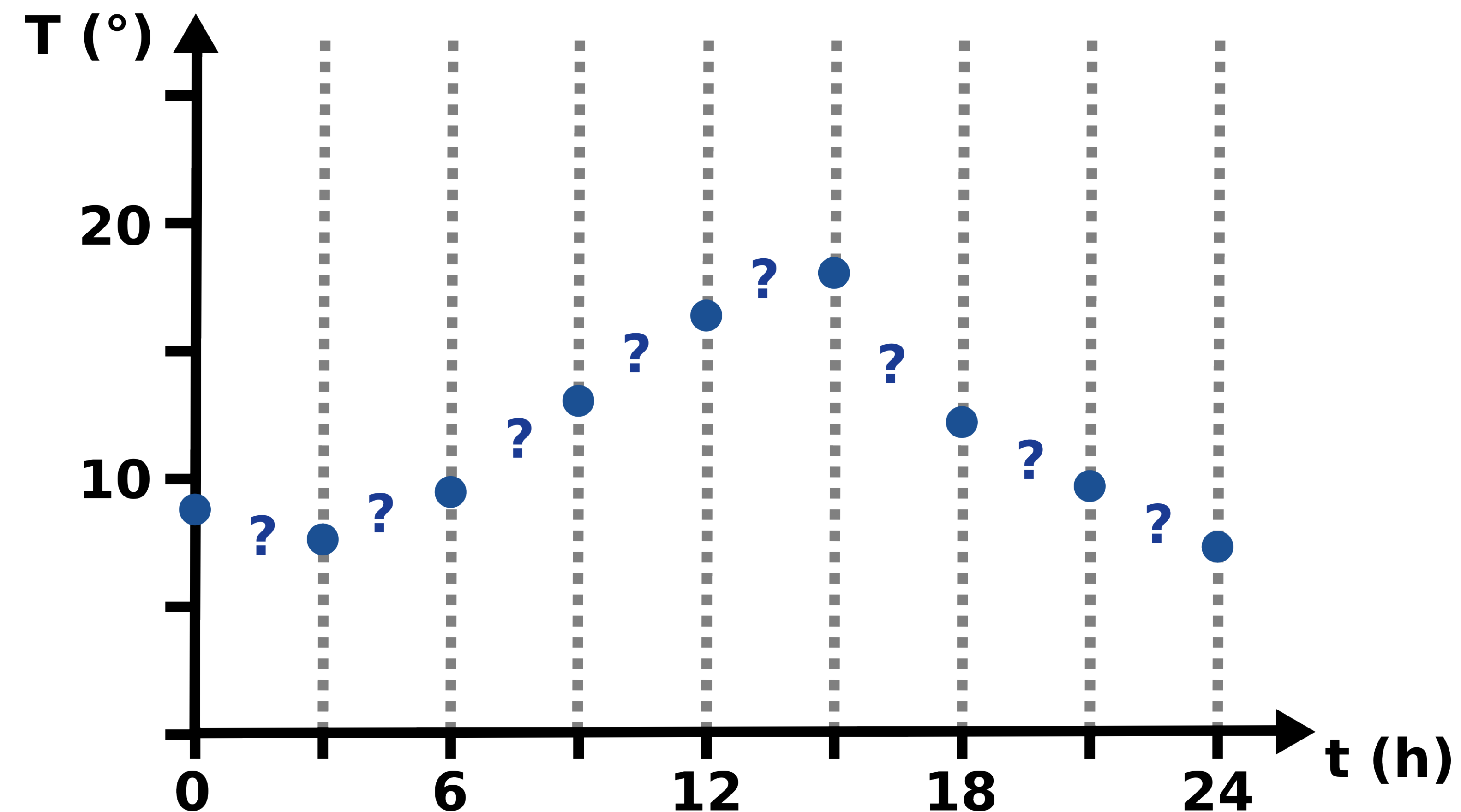
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Interpolants for regular grids

- Interpolation function
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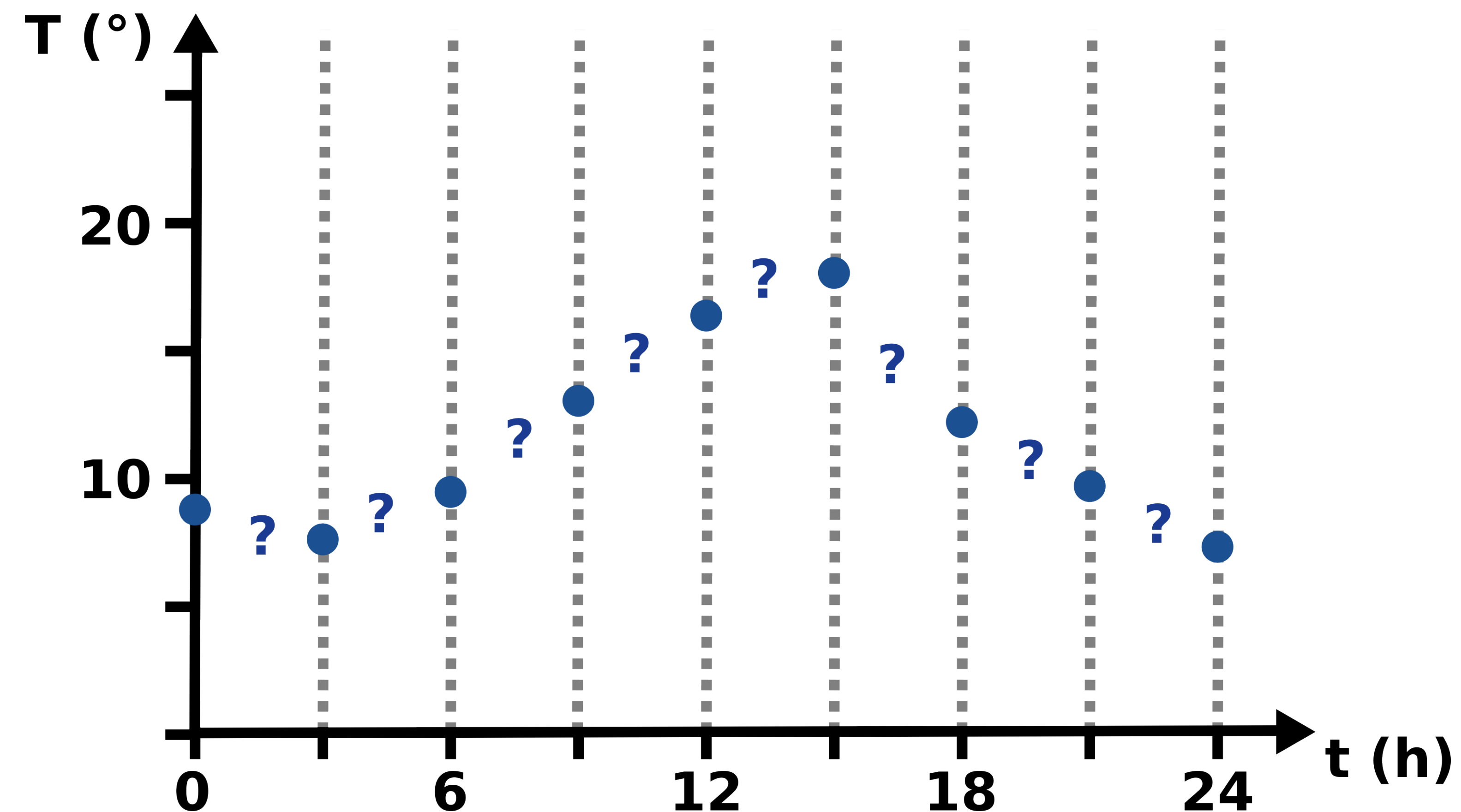
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Interpolants for regular grids

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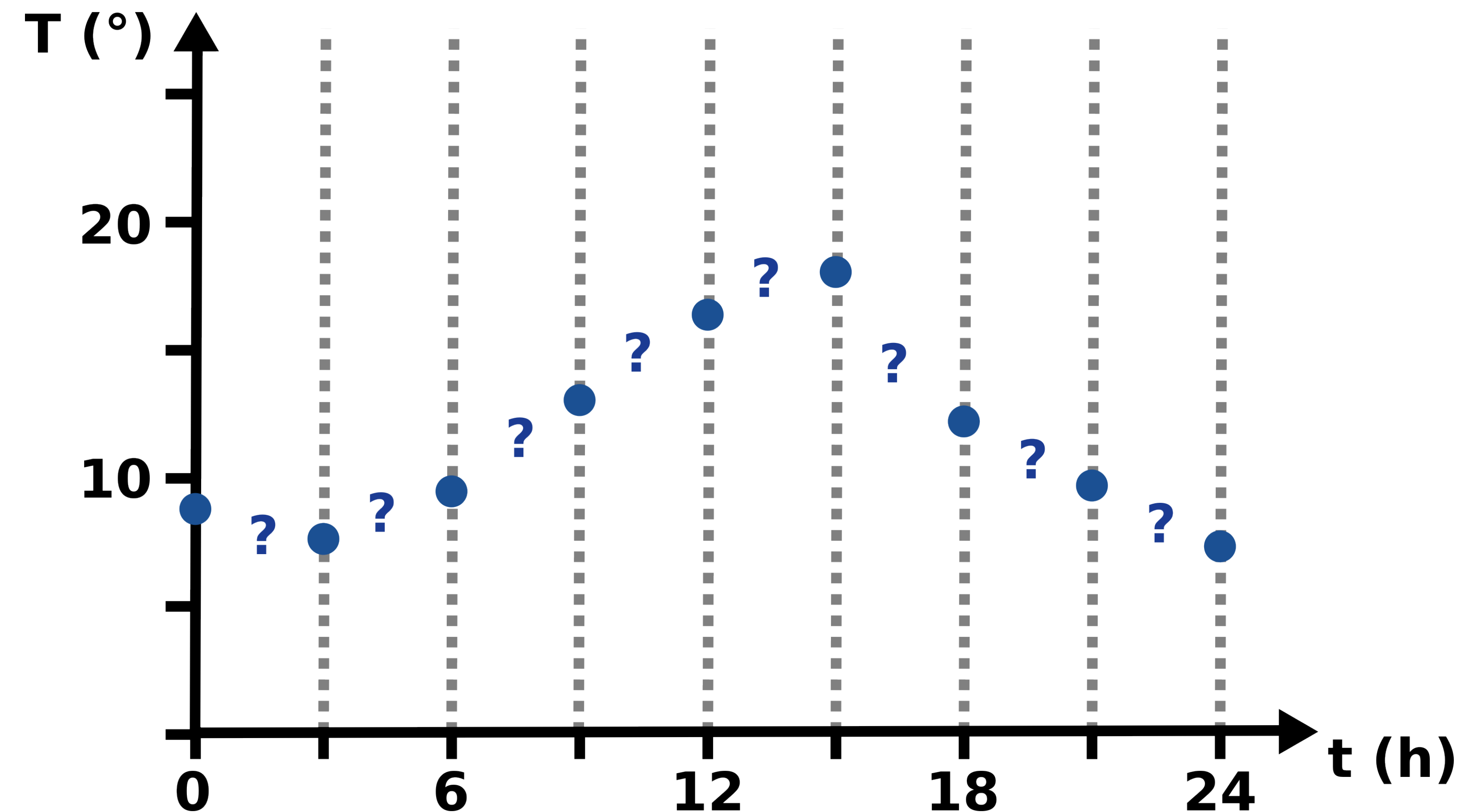
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Interpolants for regular grids

- Interpolation function
 - Given with the data
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- Examples $\mathbb{I} :]0, 1[\rightarrow]0, 1[$
 - Piecewise constant
 - $\mathbb{I}(x) = \delta(x - 1)$

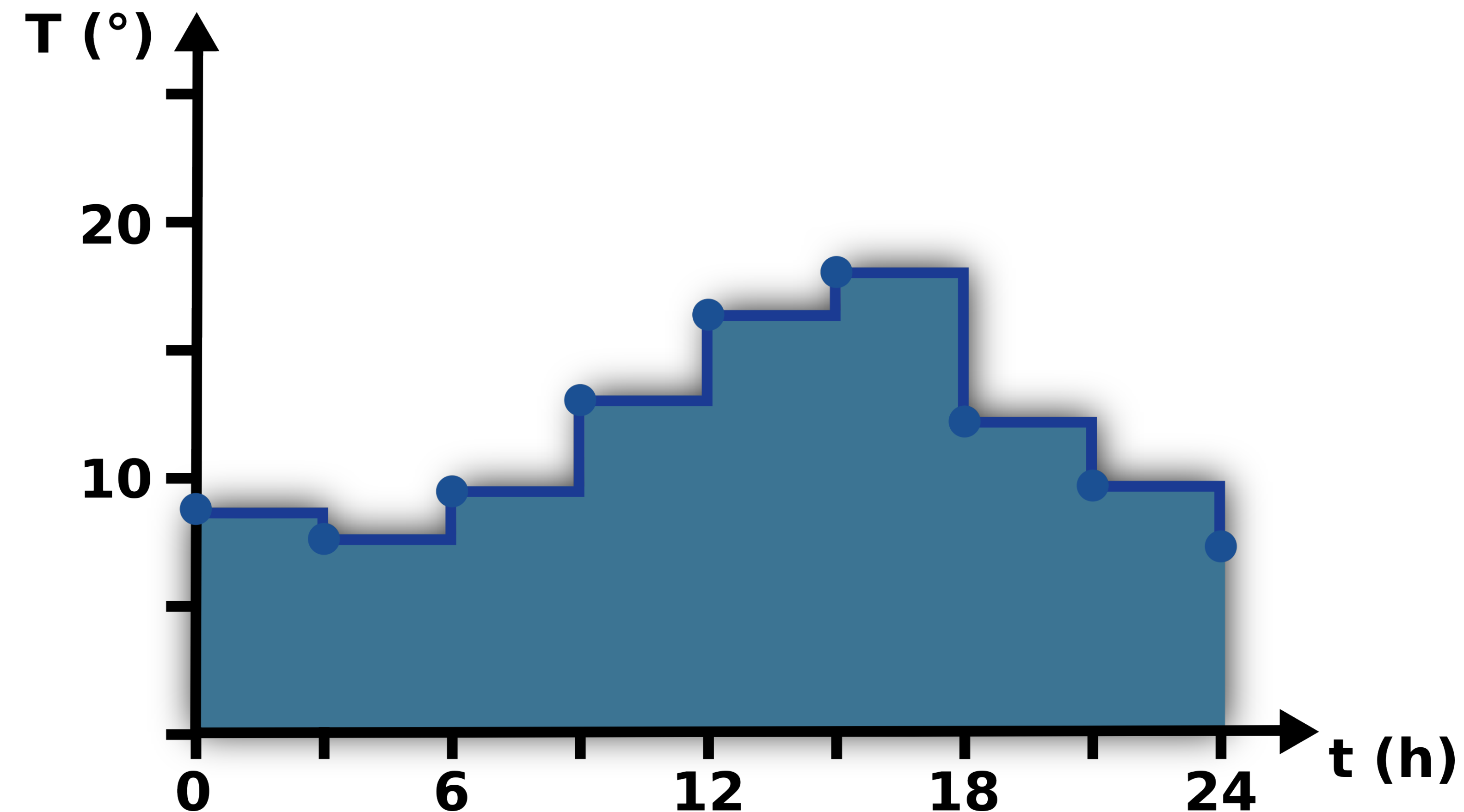
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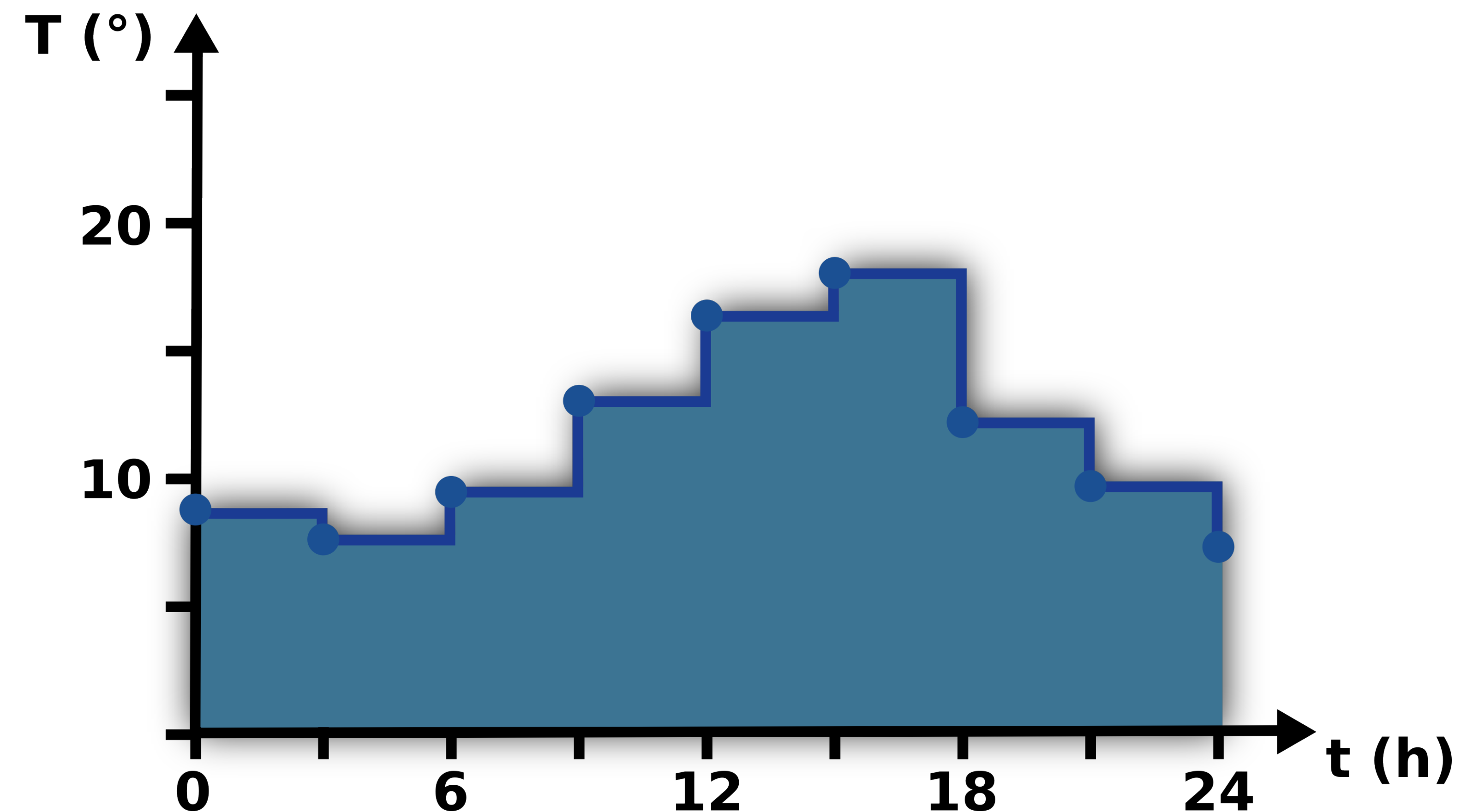
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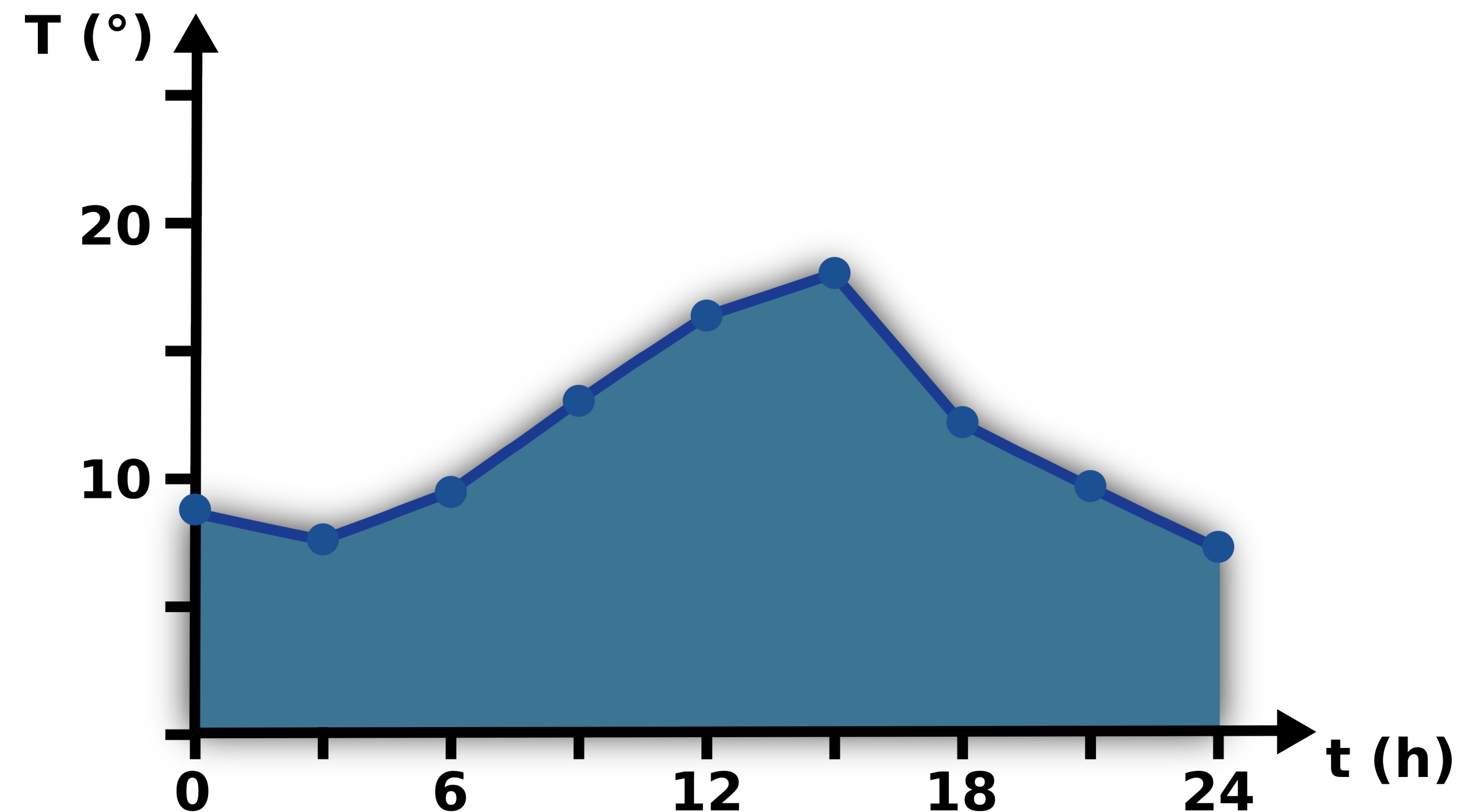
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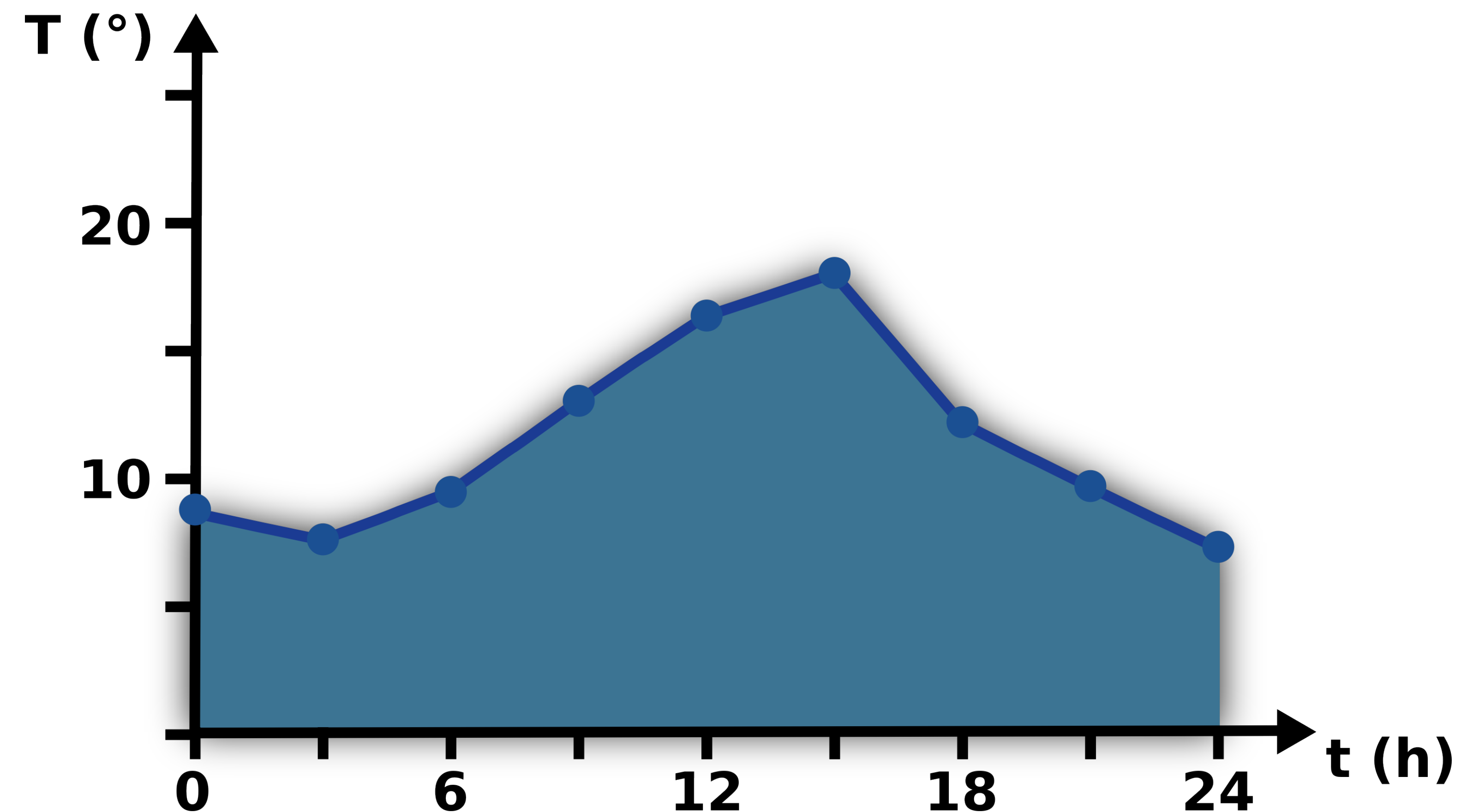
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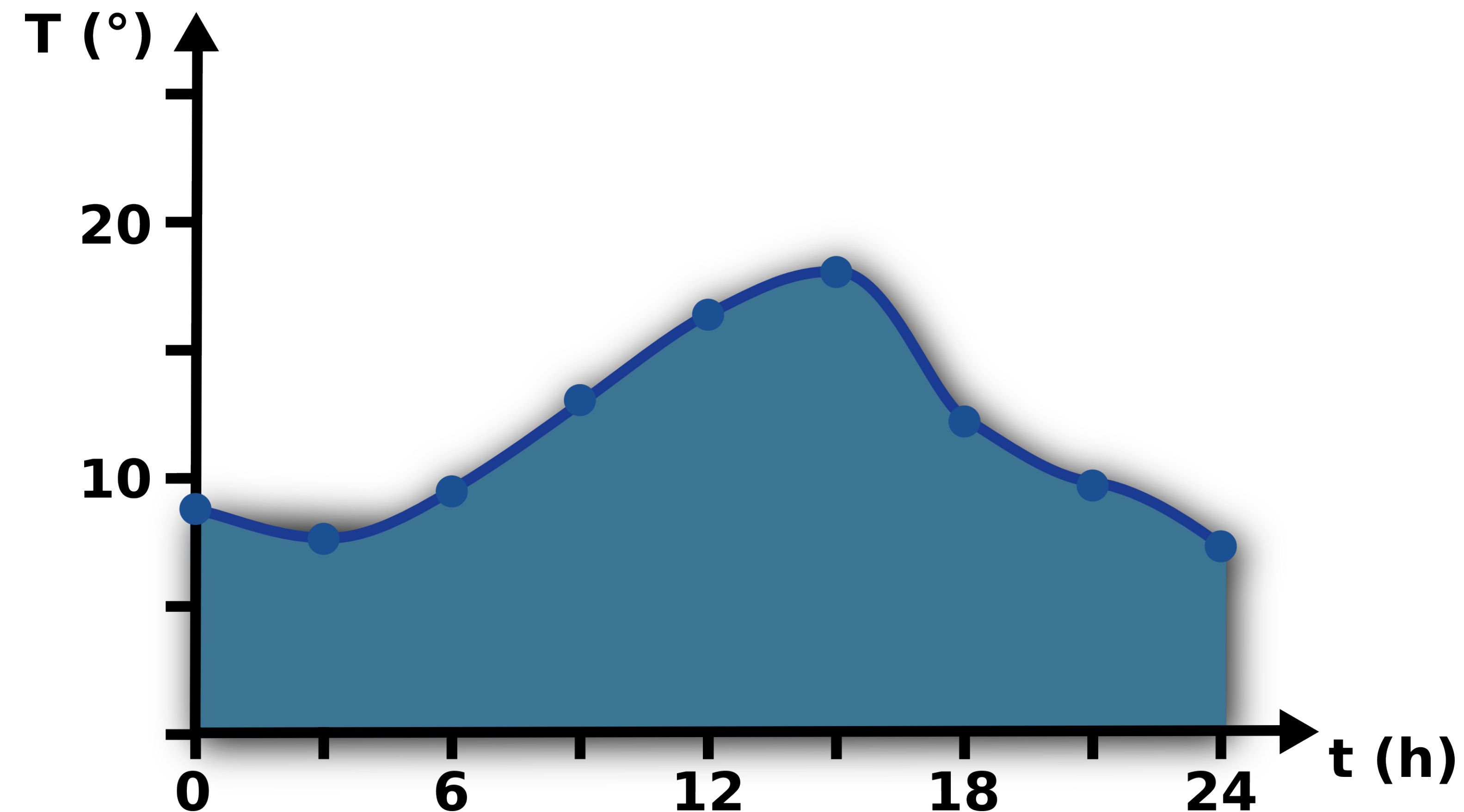
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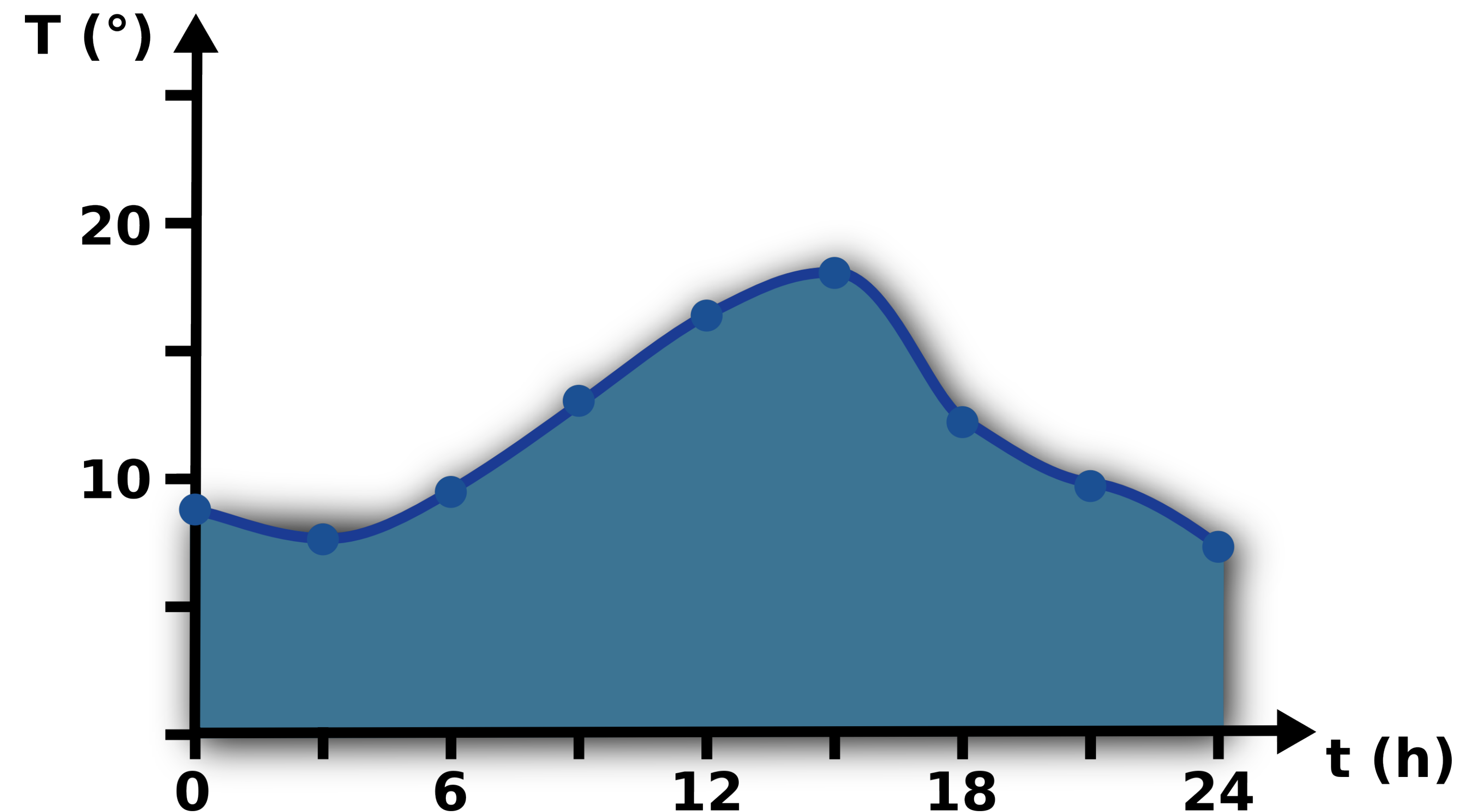
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$$f : [0, 24] \rightarrow \mathbb{R}$$



Interpolants for regular grids

- For regular grids of \mathbb{R}^2

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Interpolants for regular grids

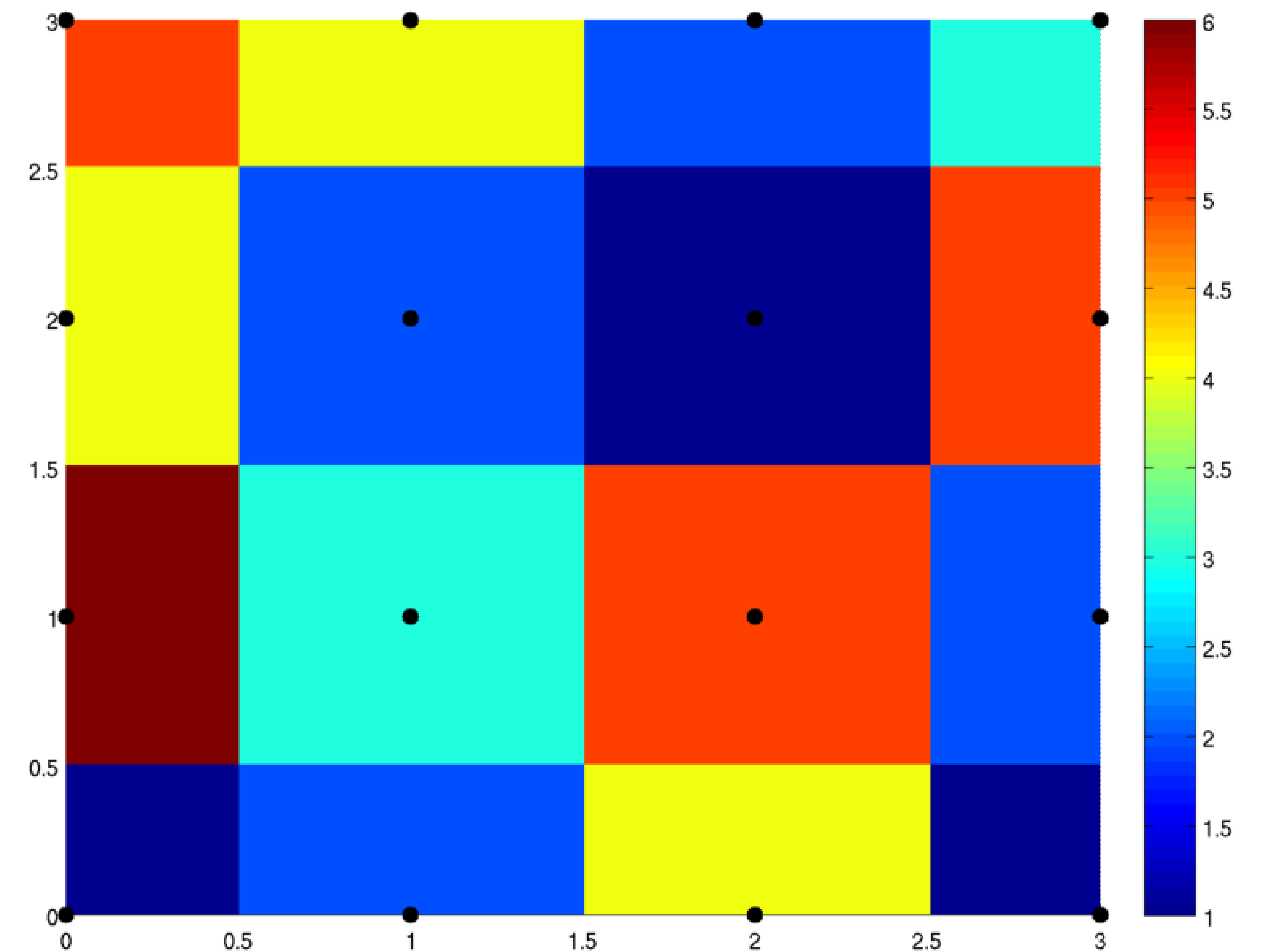
- For regular grids of \mathbb{R}^2
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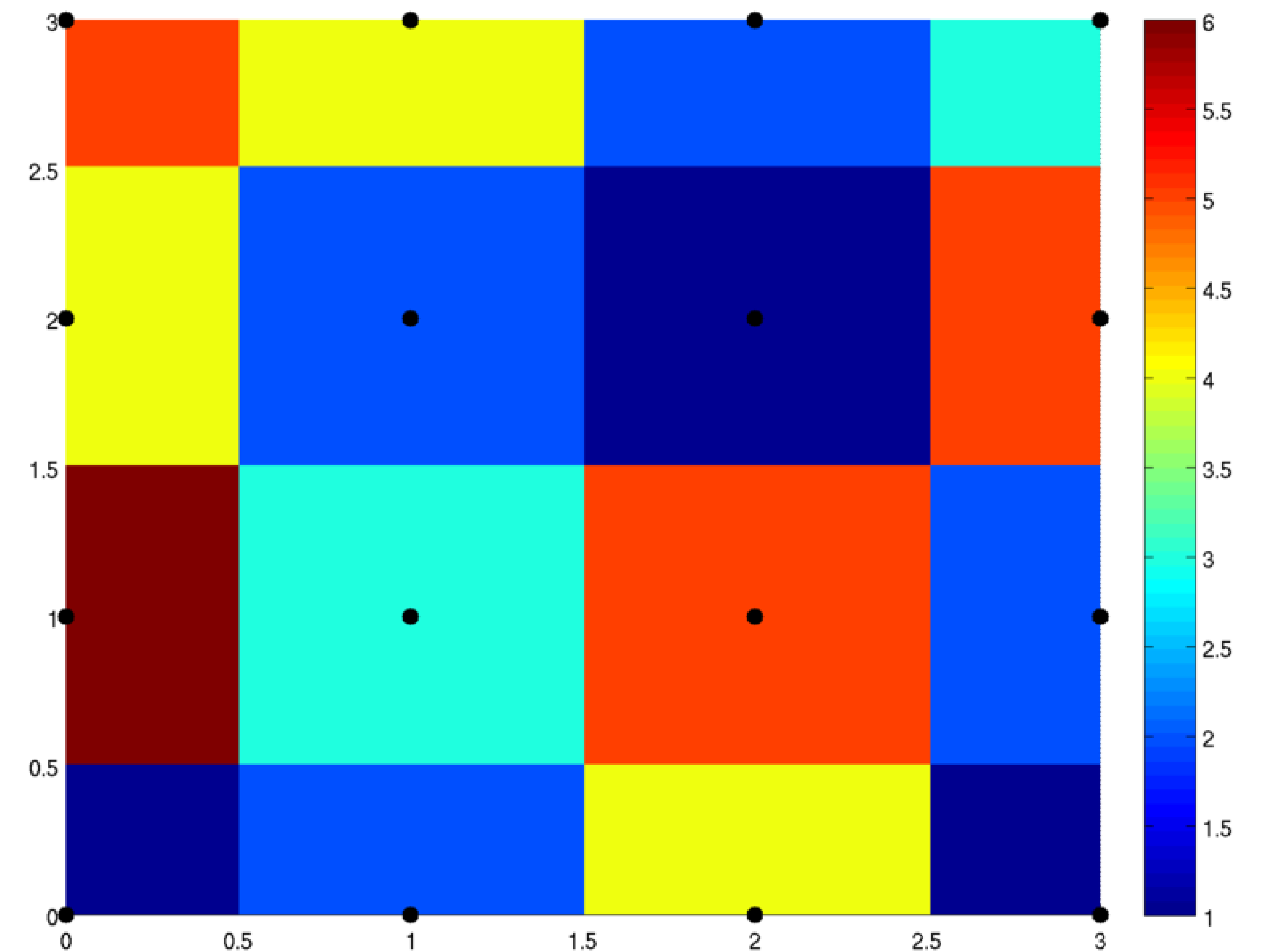
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Interpolants for regular grids

- For regular grids of \mathbb{R}^2
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- Examples
 - Bilinear interpolation
 - One dimension at a time



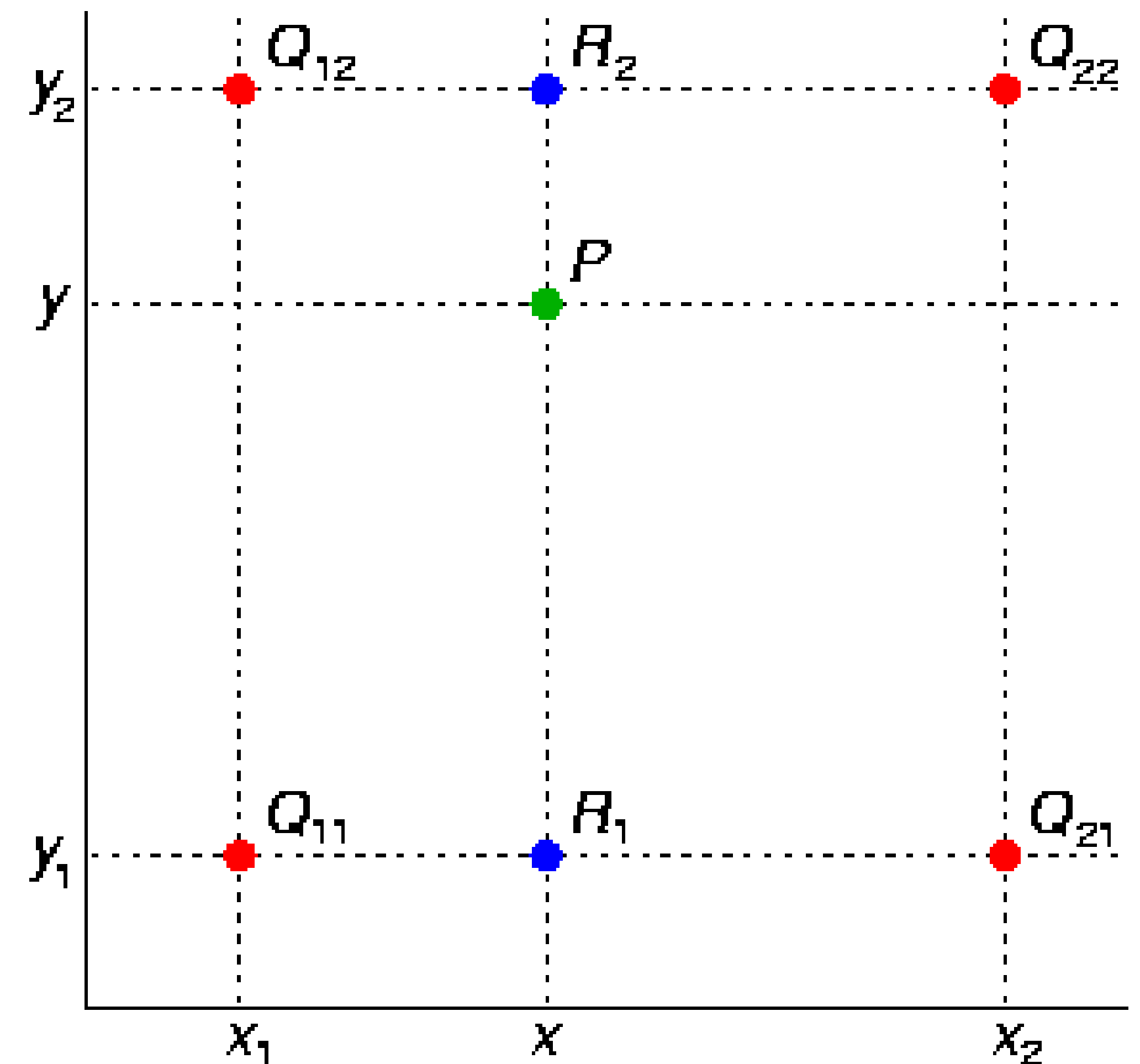
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$$f_{\mathbb{I}}(x, y_1) = \frac{x - x_1}{x_2 - x_1} (f(x_2, y_1) - f(x_1, y_1)) + f(x_1, y_1)$$

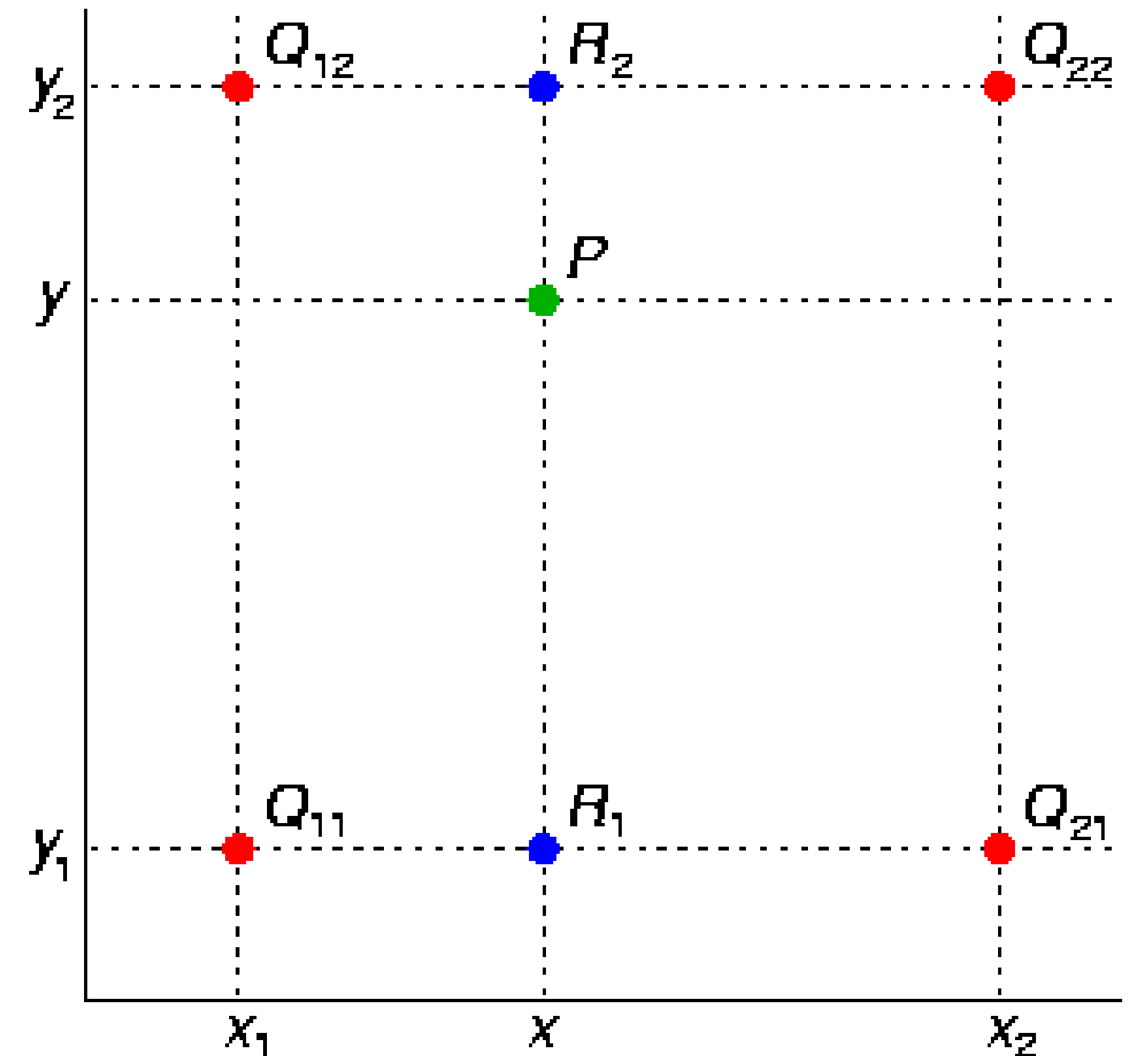


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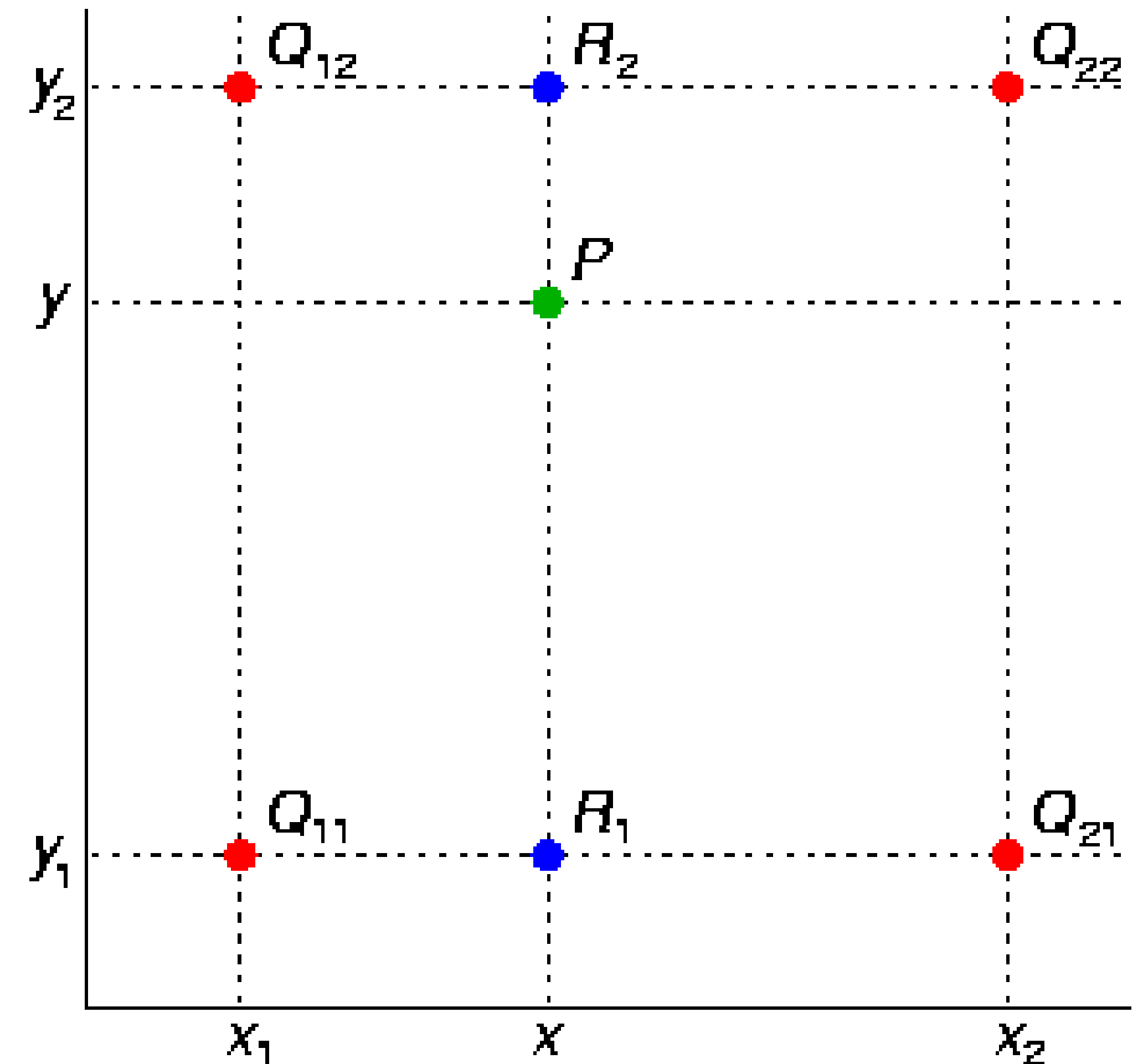
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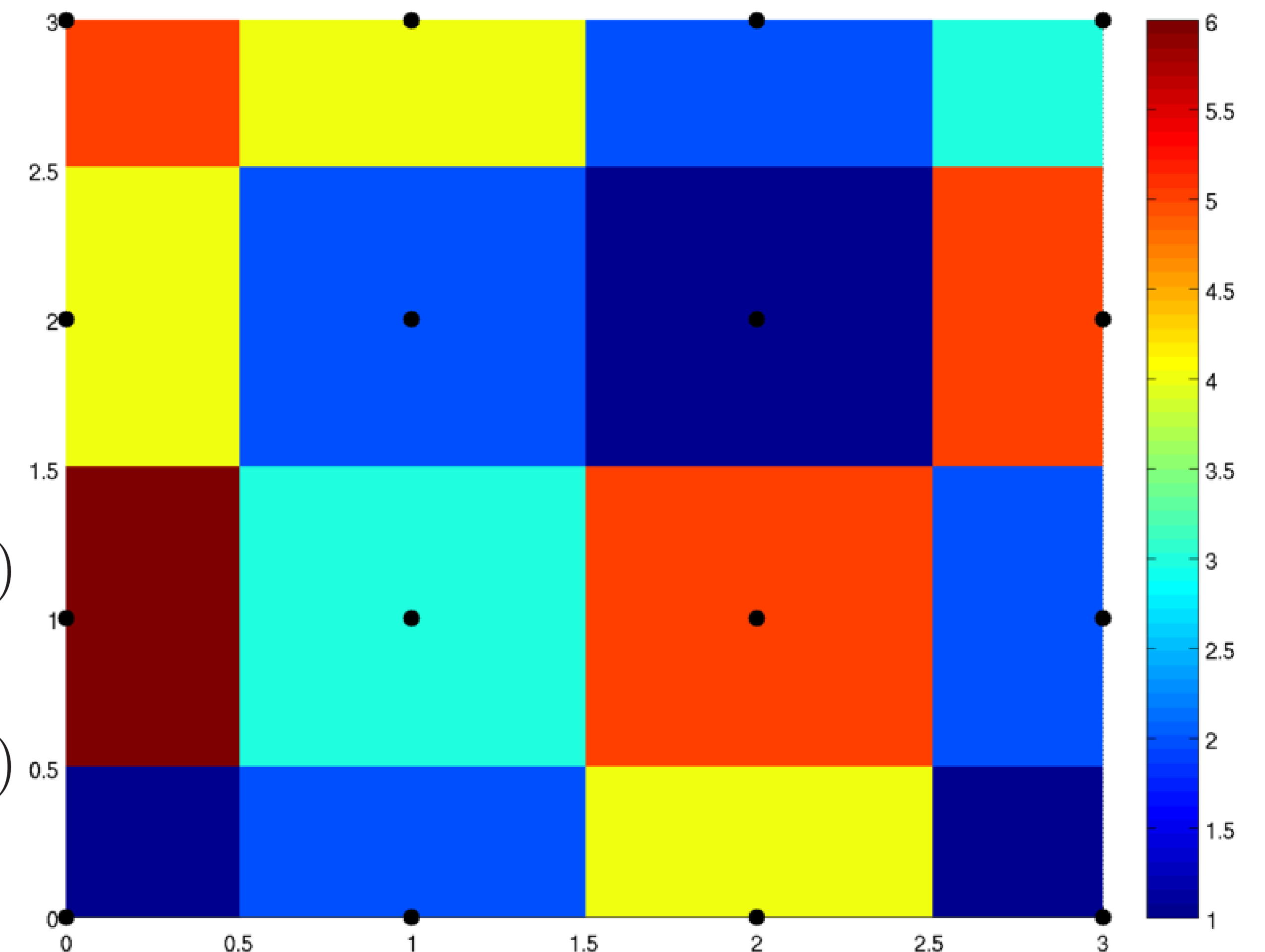
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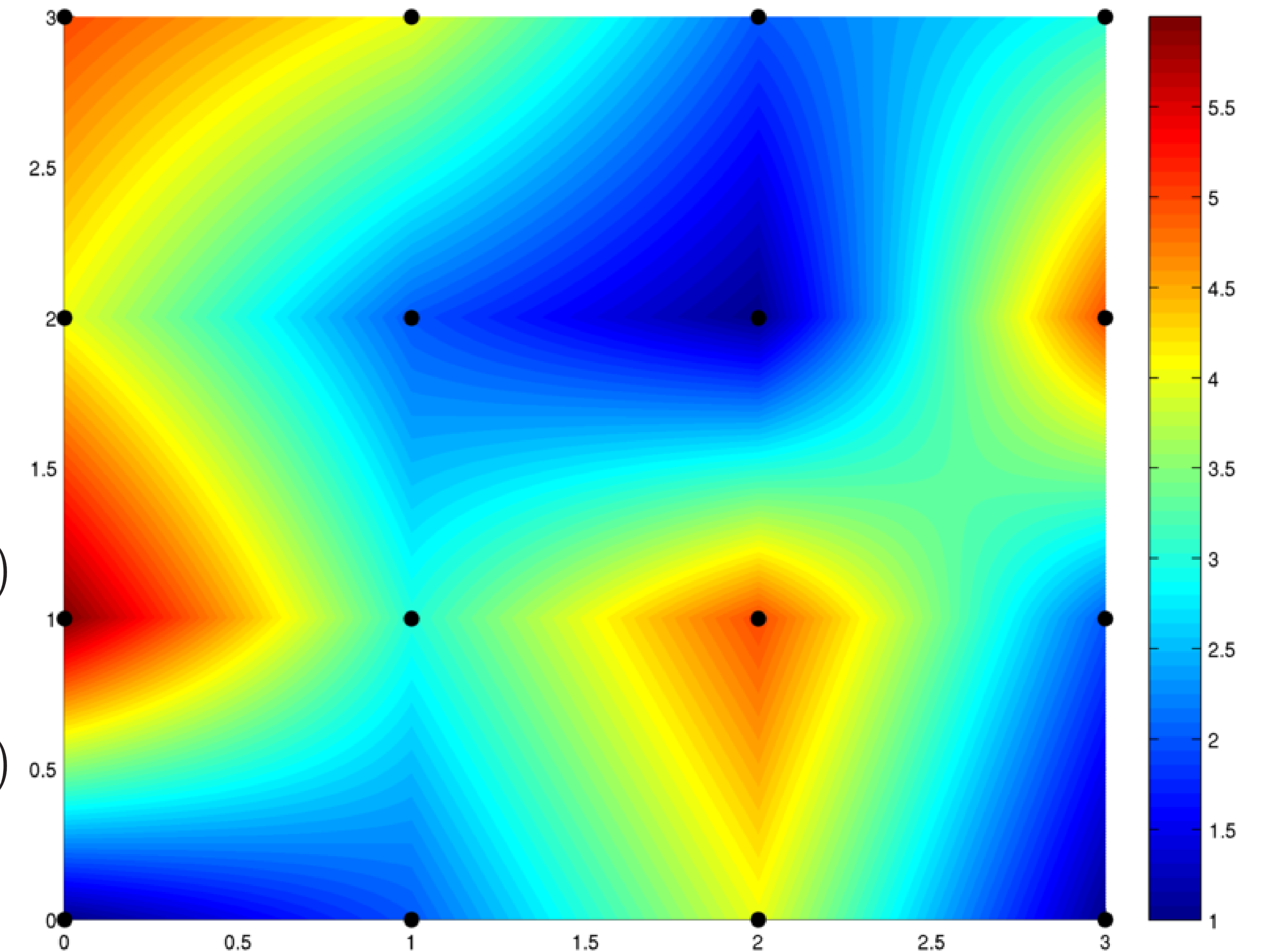
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[Wikipedia]

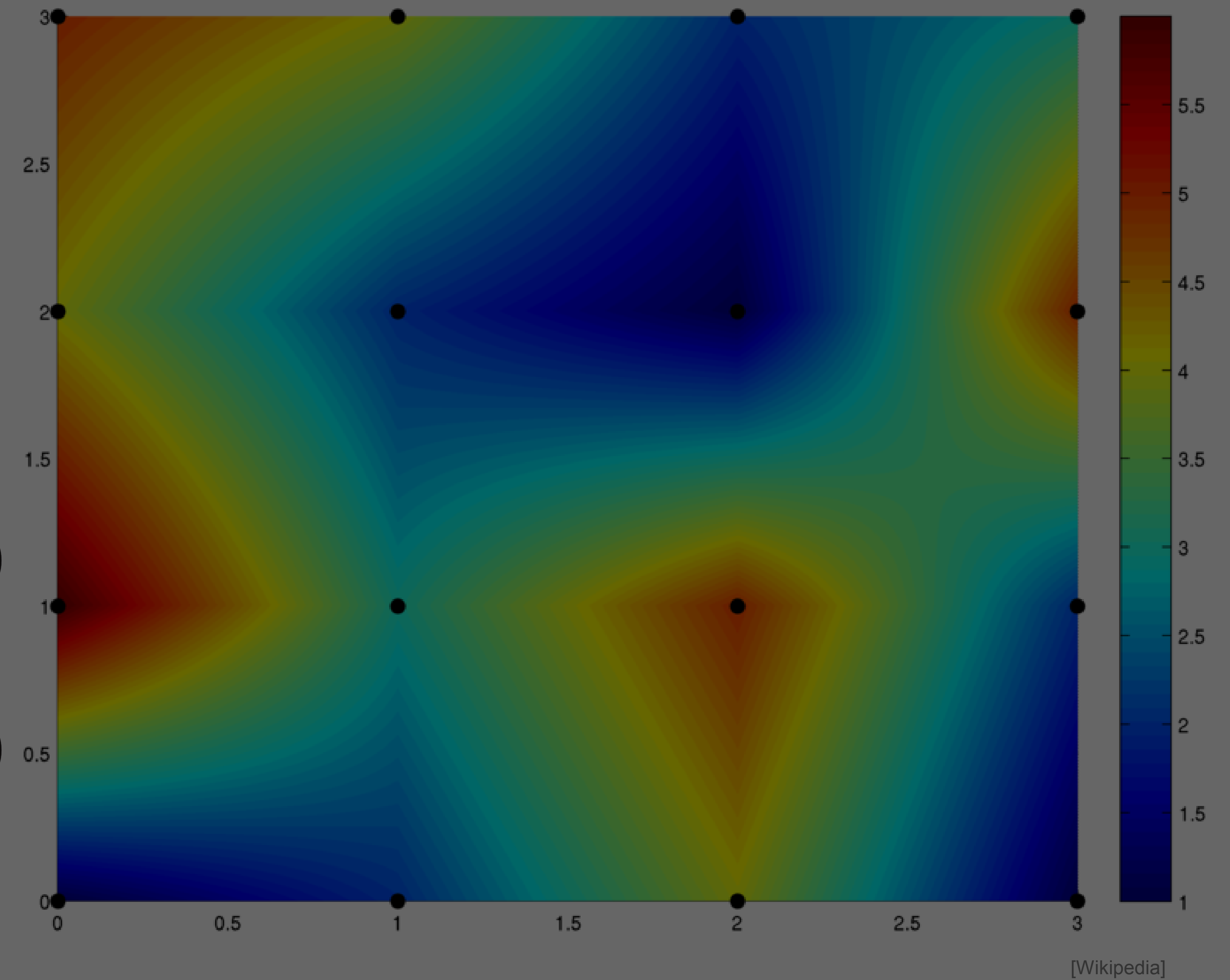
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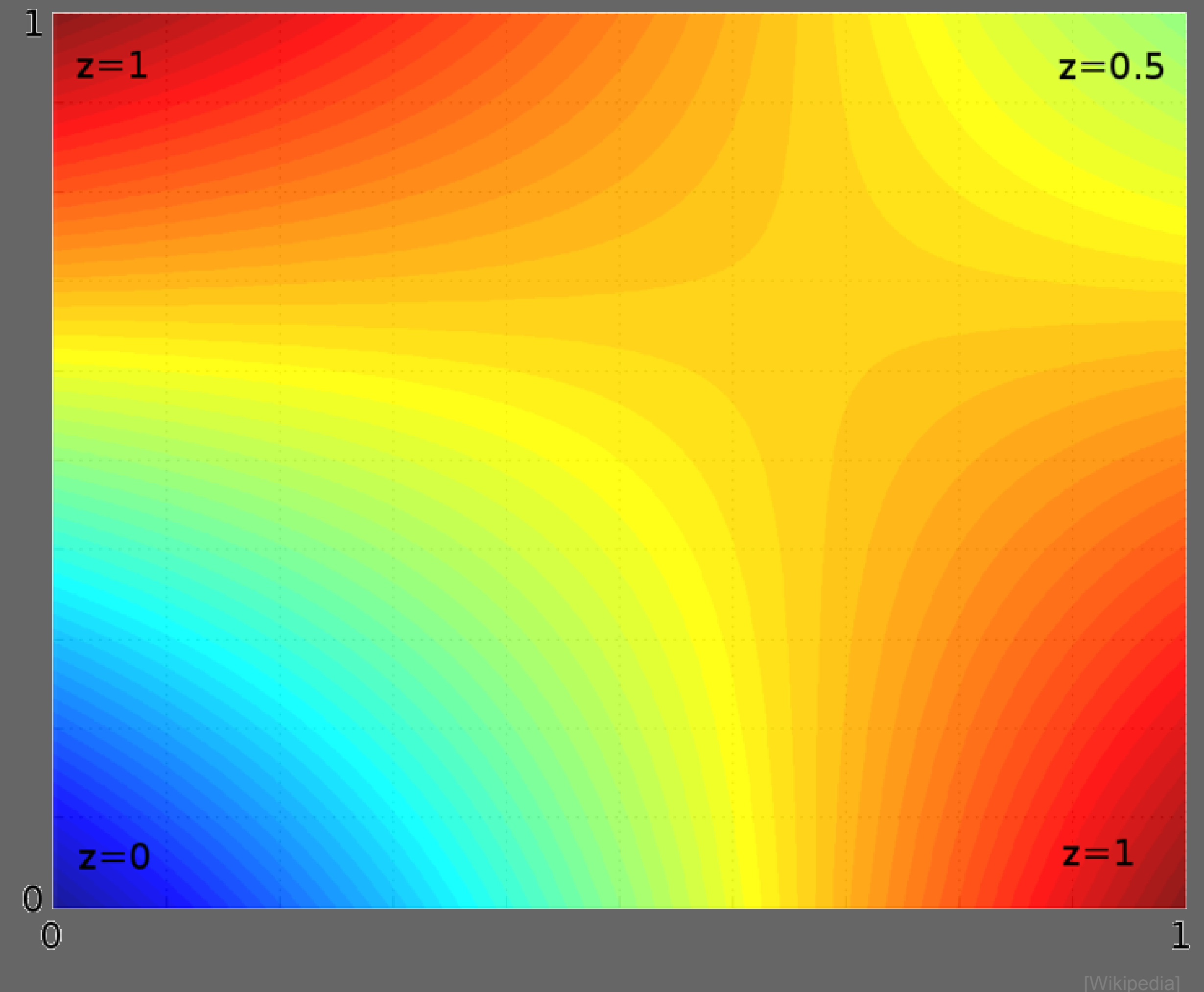
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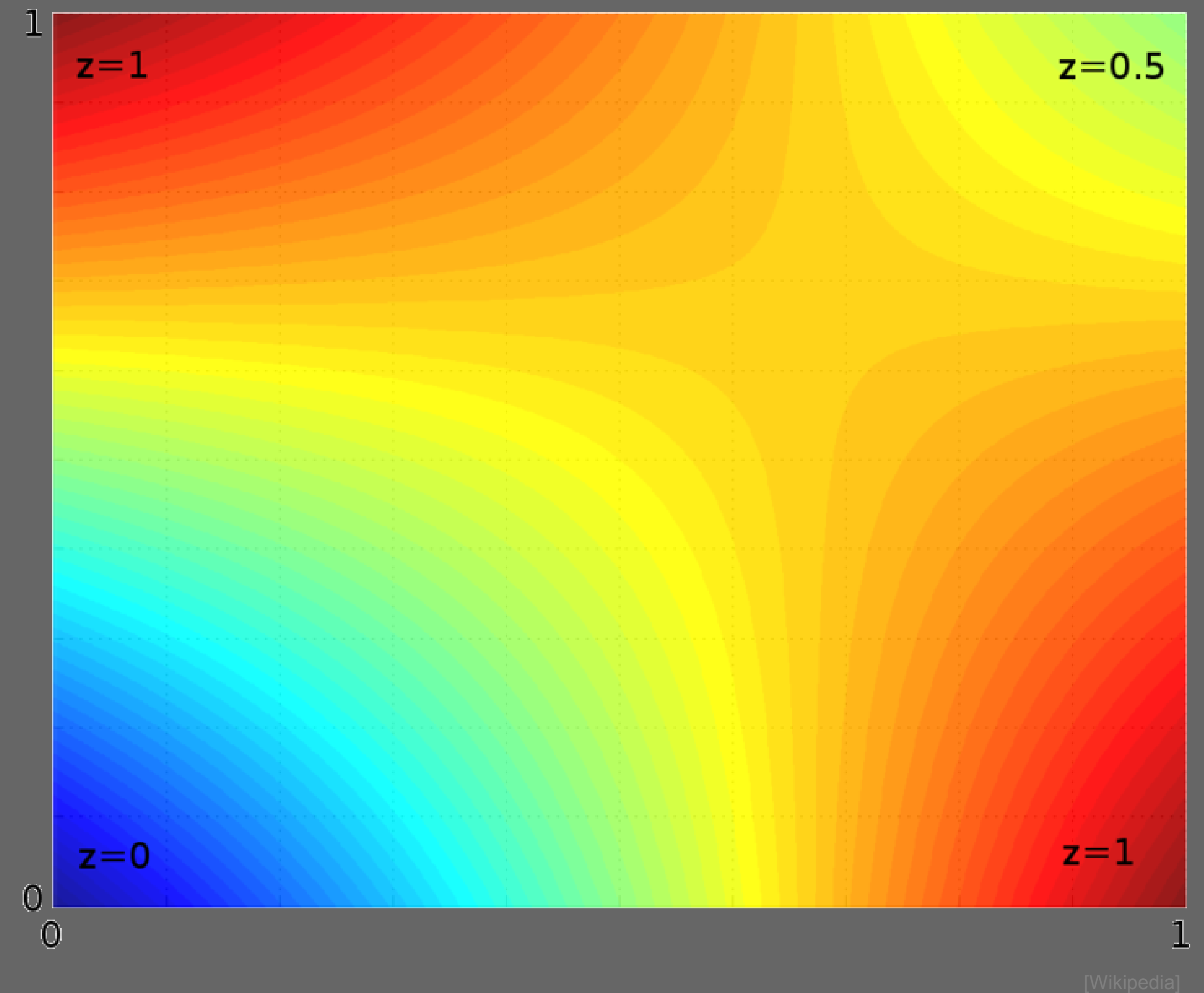


Interpolants for regular grids

- For regular grids of \mathbb{R}^2

- Problem

$$f_{\mathbb{I}}(x, y_2) = \frac{x - x_1}{x_2 - x_1} (f(x_2, y_2) - f(x_1, y_2)) + f(x_1, y_2)$$
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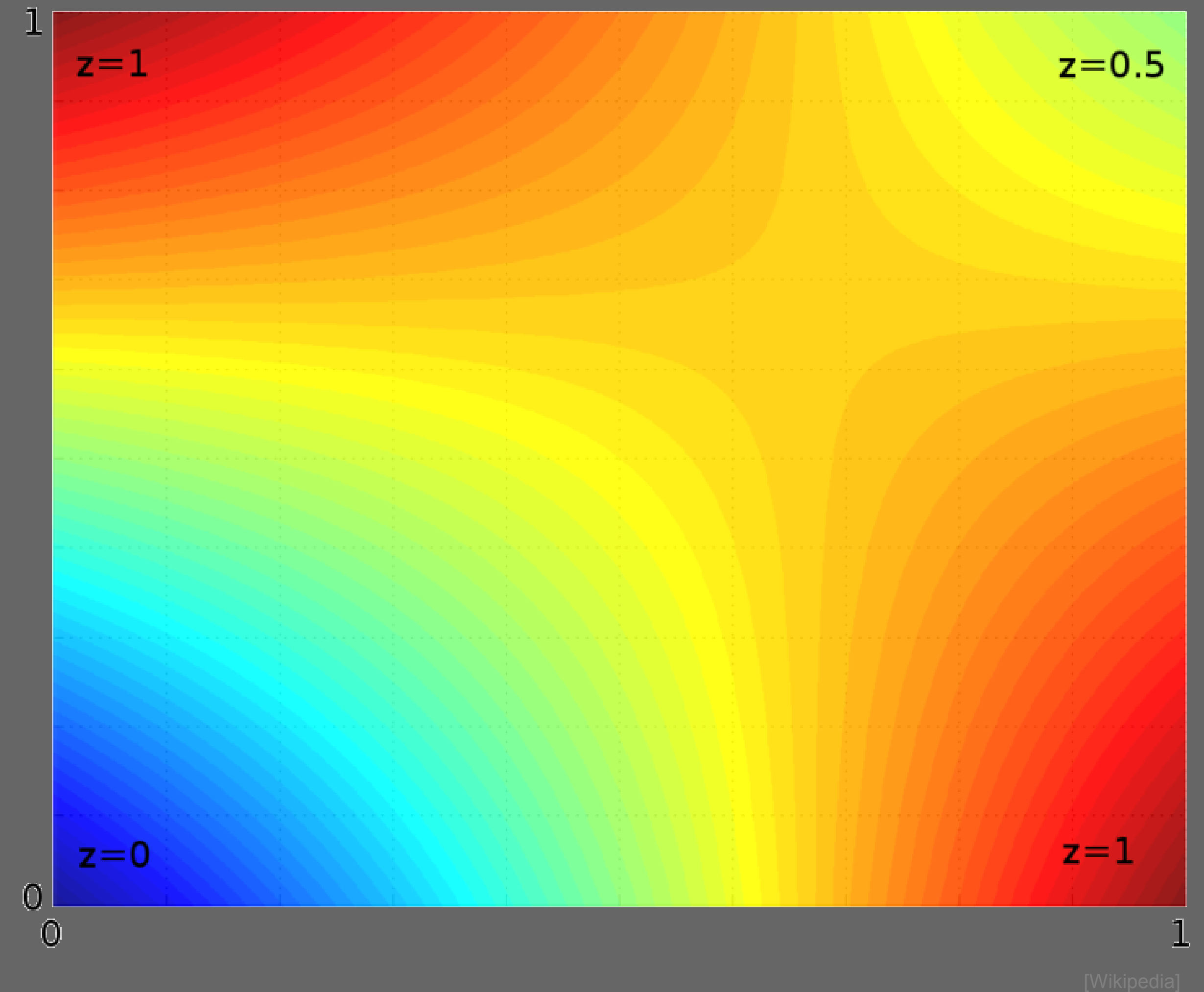
Interpolants for regular grids

- For regular grids of \mathbb{R}^2

- Problem
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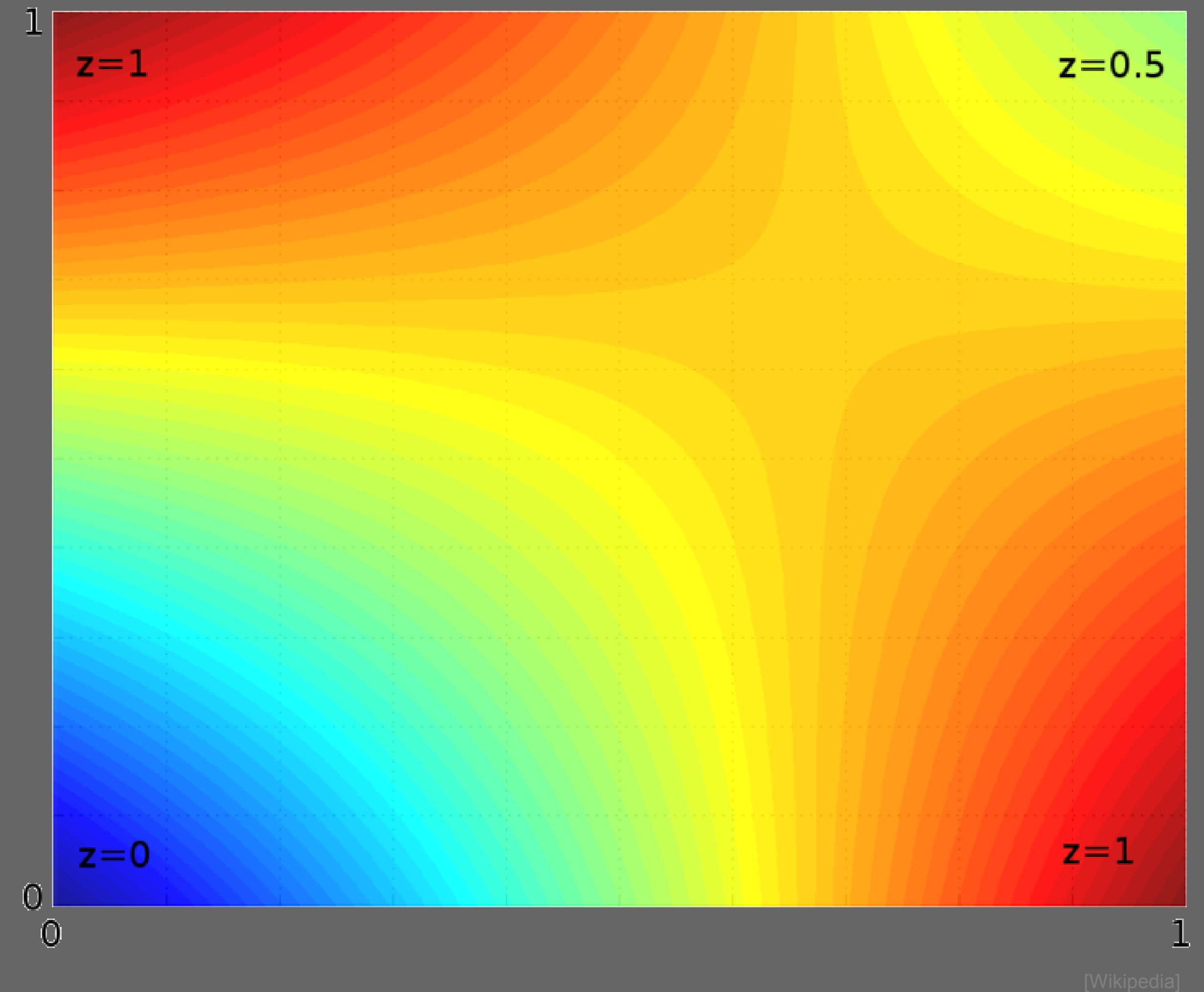
Interpolants for regular grids

- For regular grids of \mathbb{R}^2

- Problem
 - Critical points in the interior
 - Source of many ambiguities in visualization

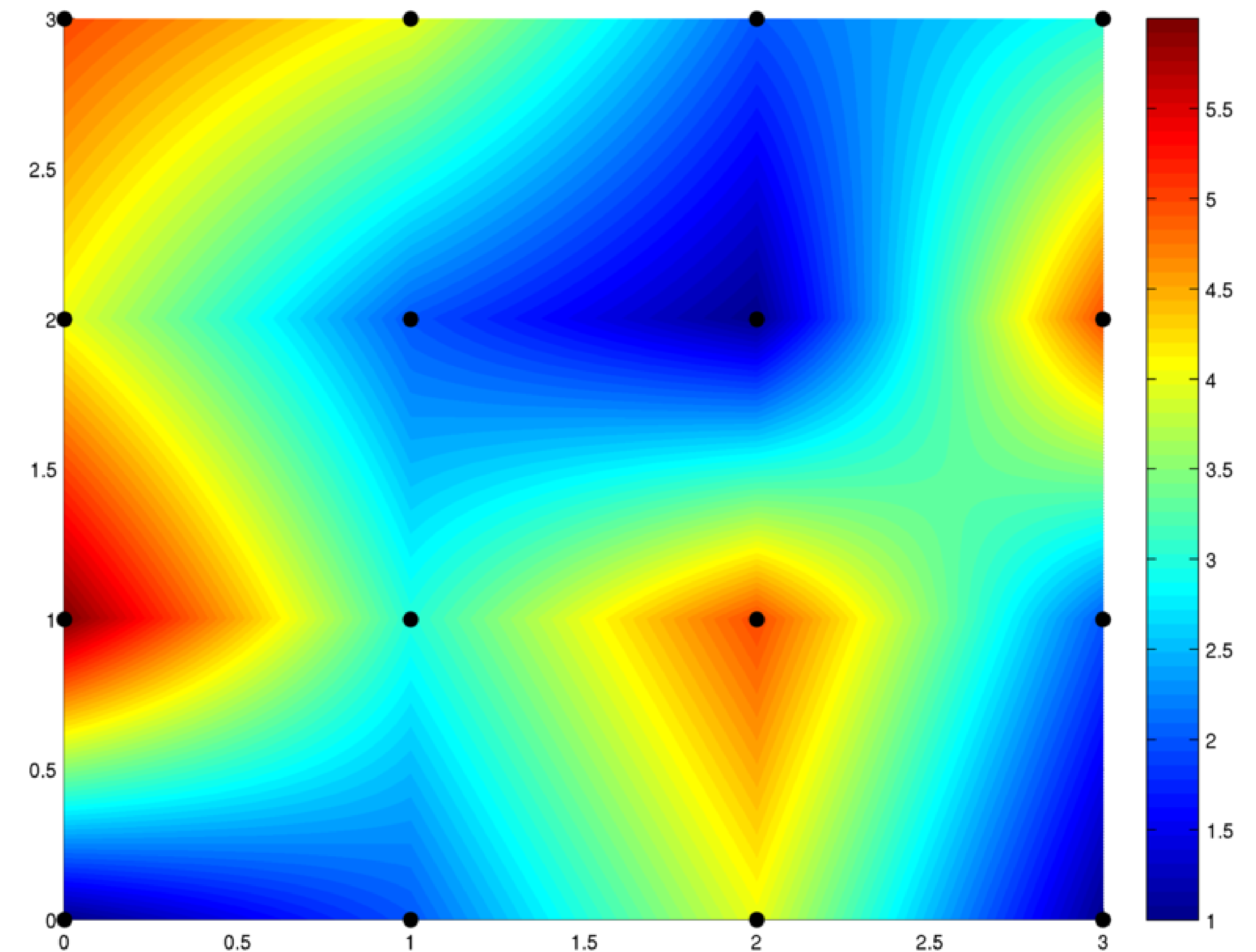
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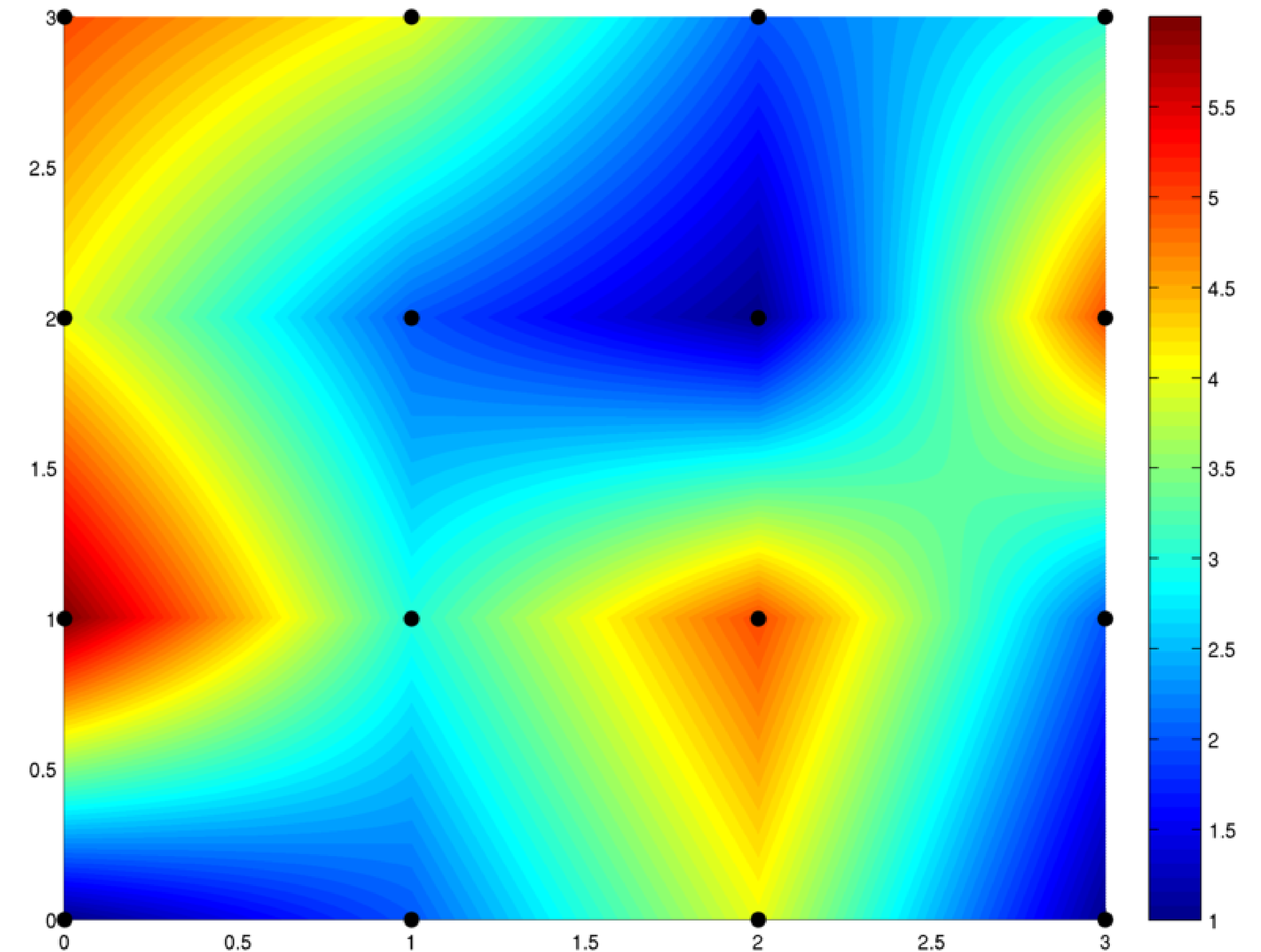
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 - Examples
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[Wikipedia]

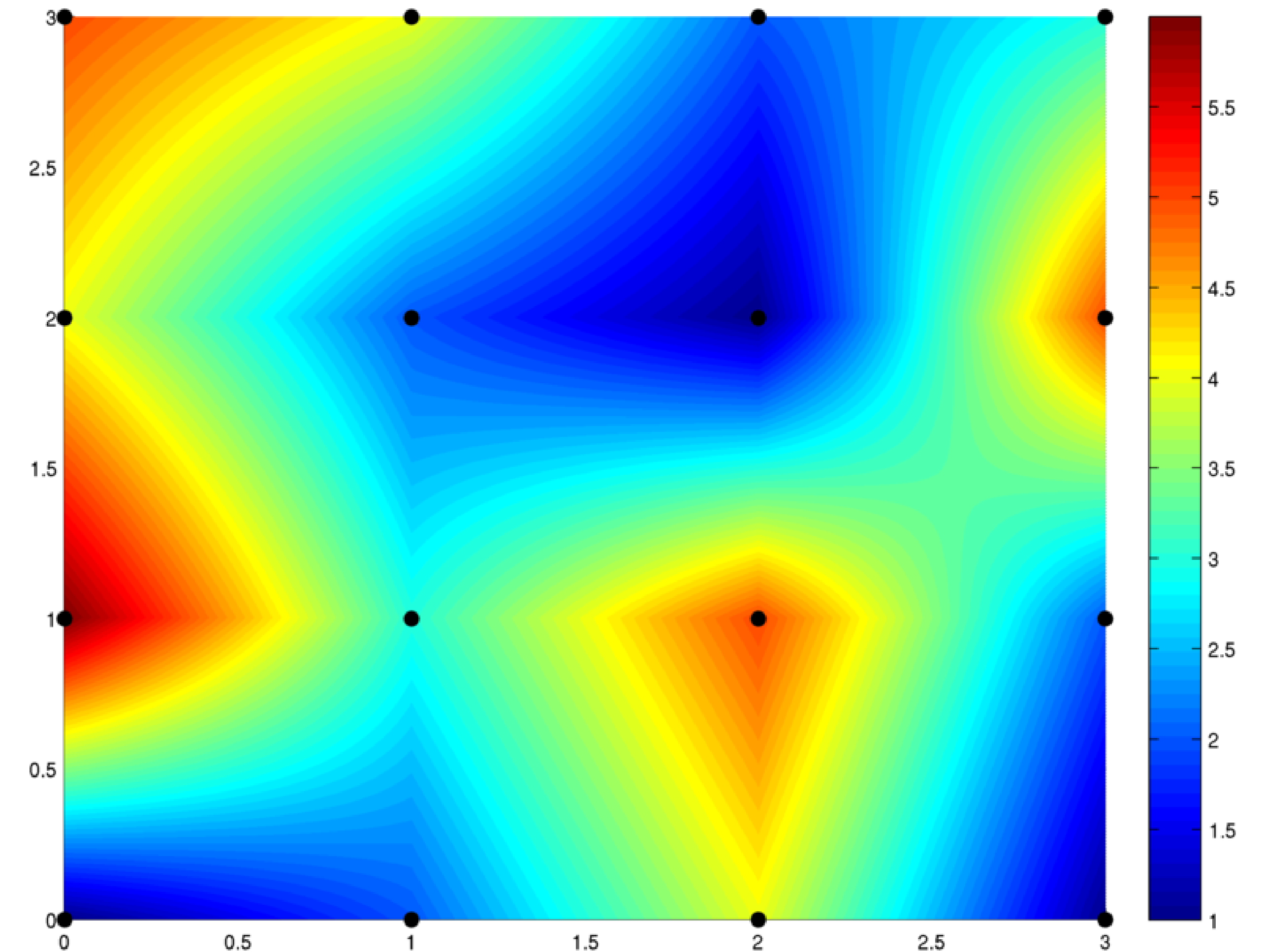
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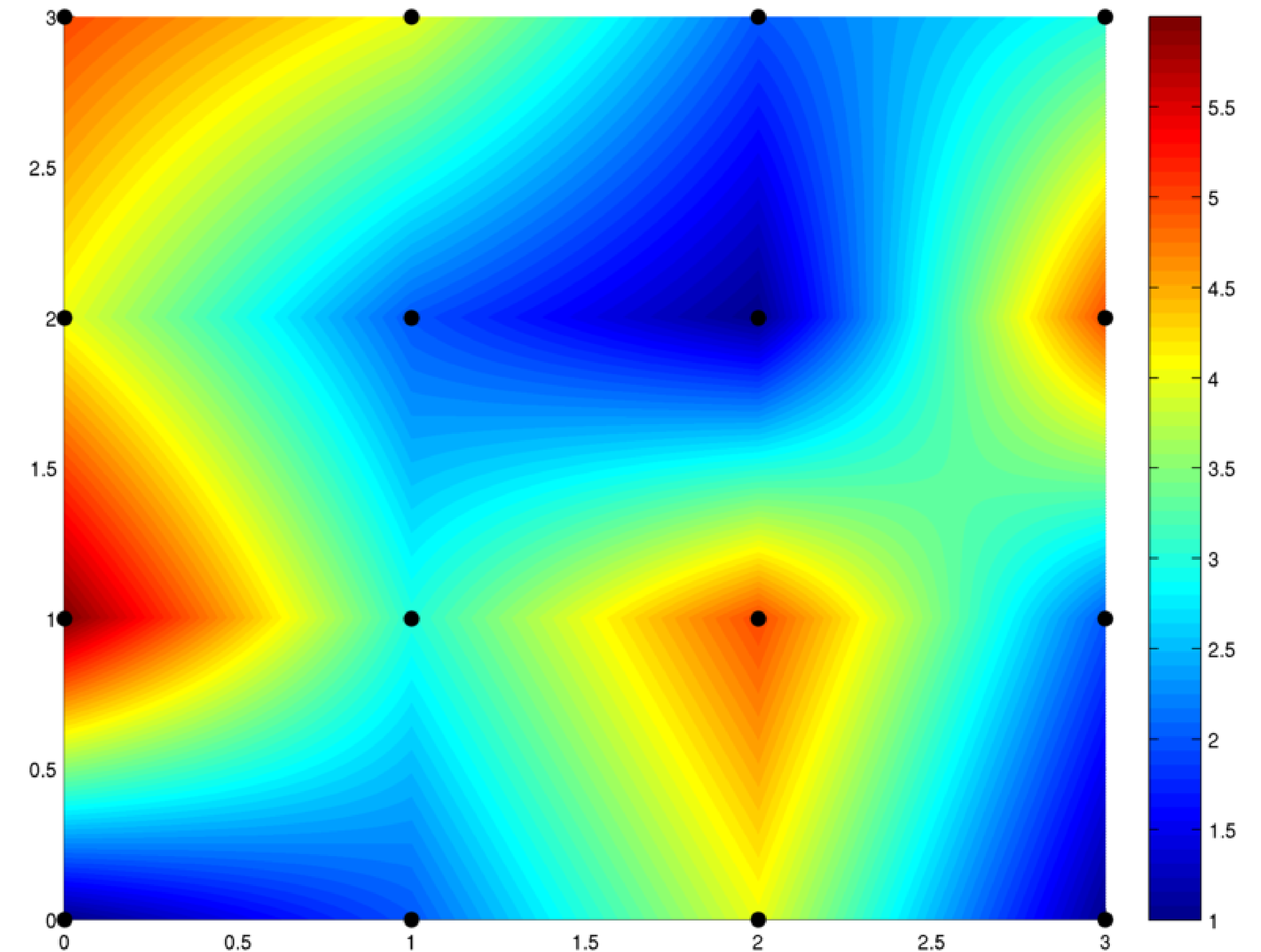
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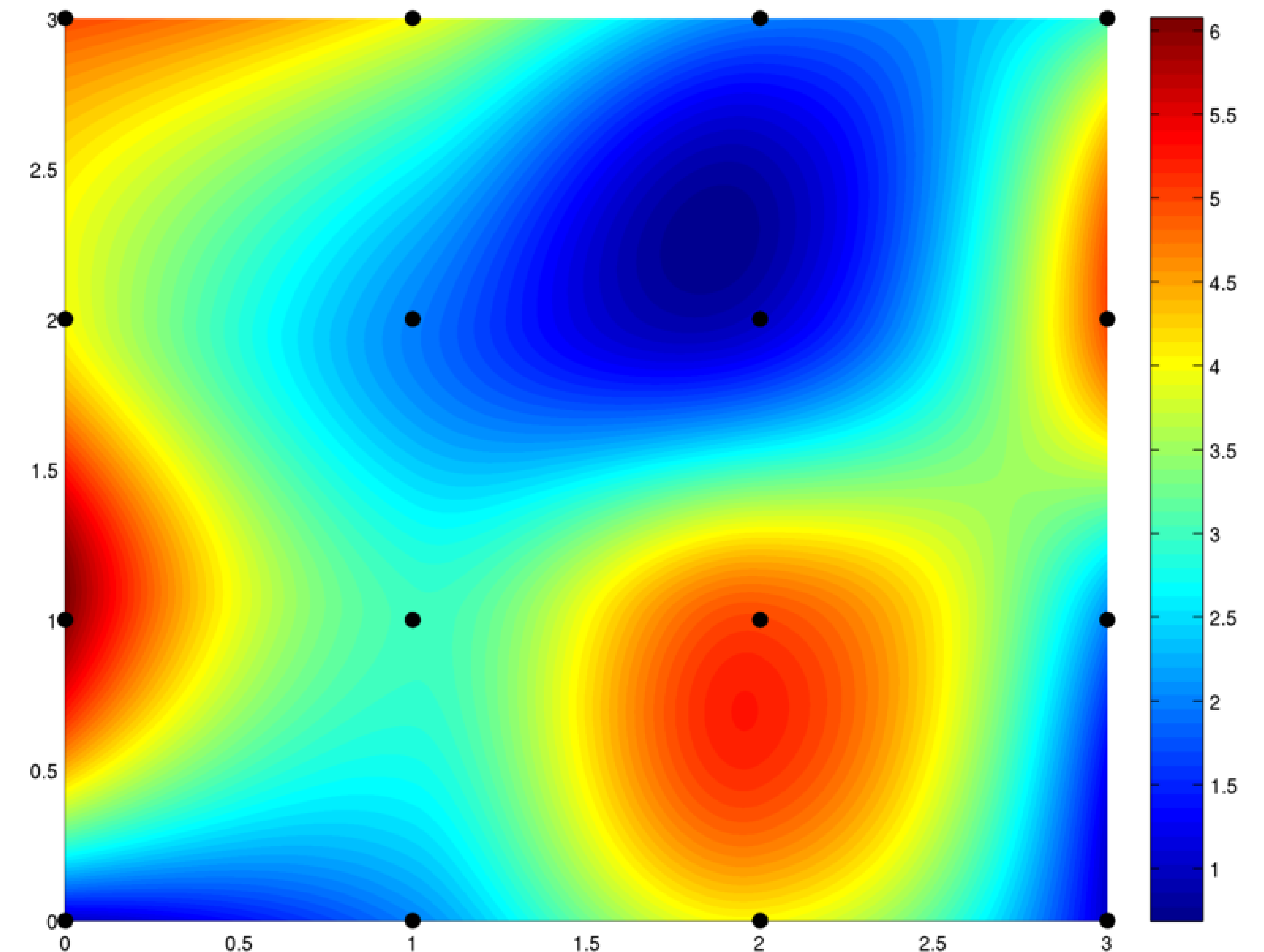
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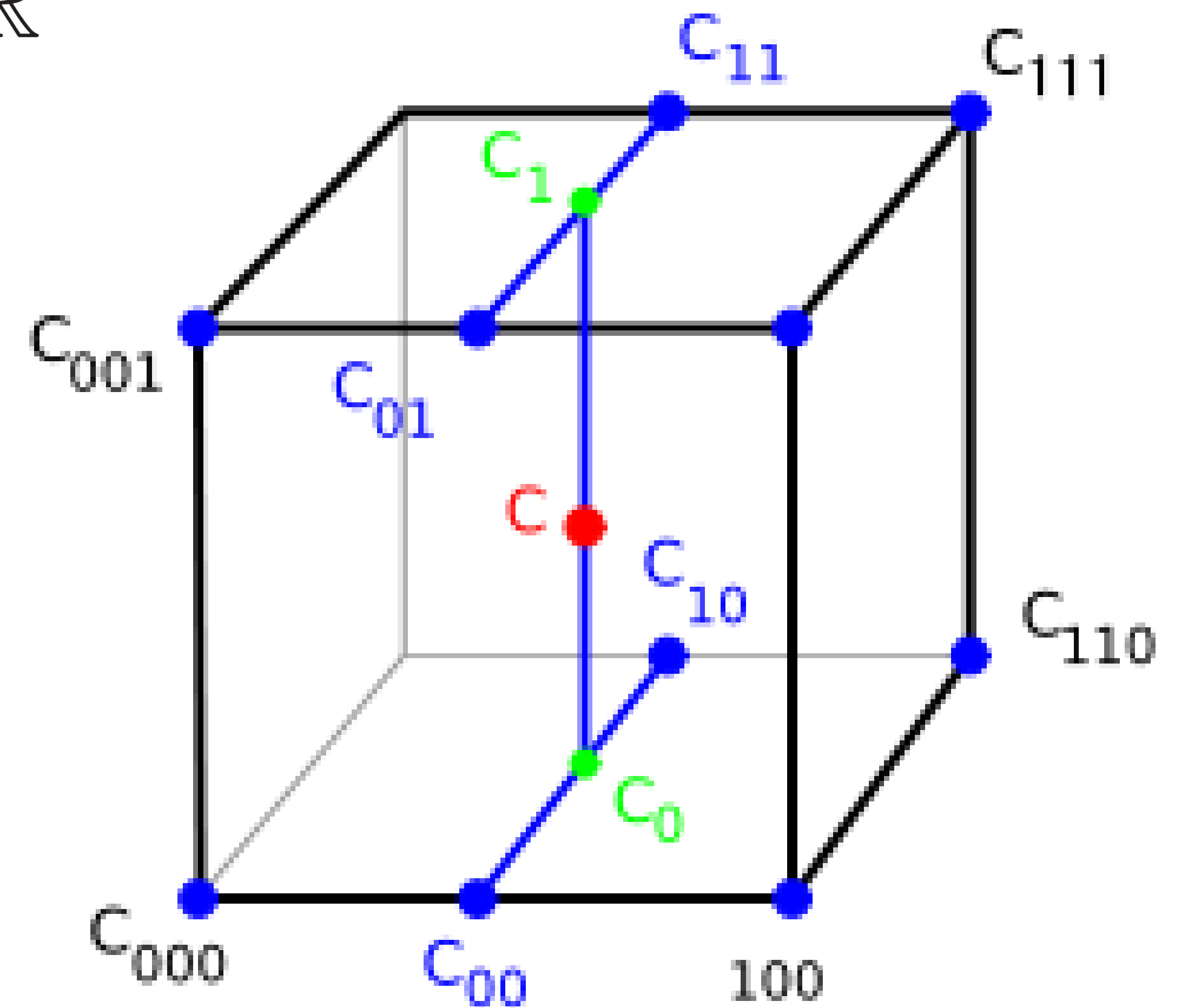
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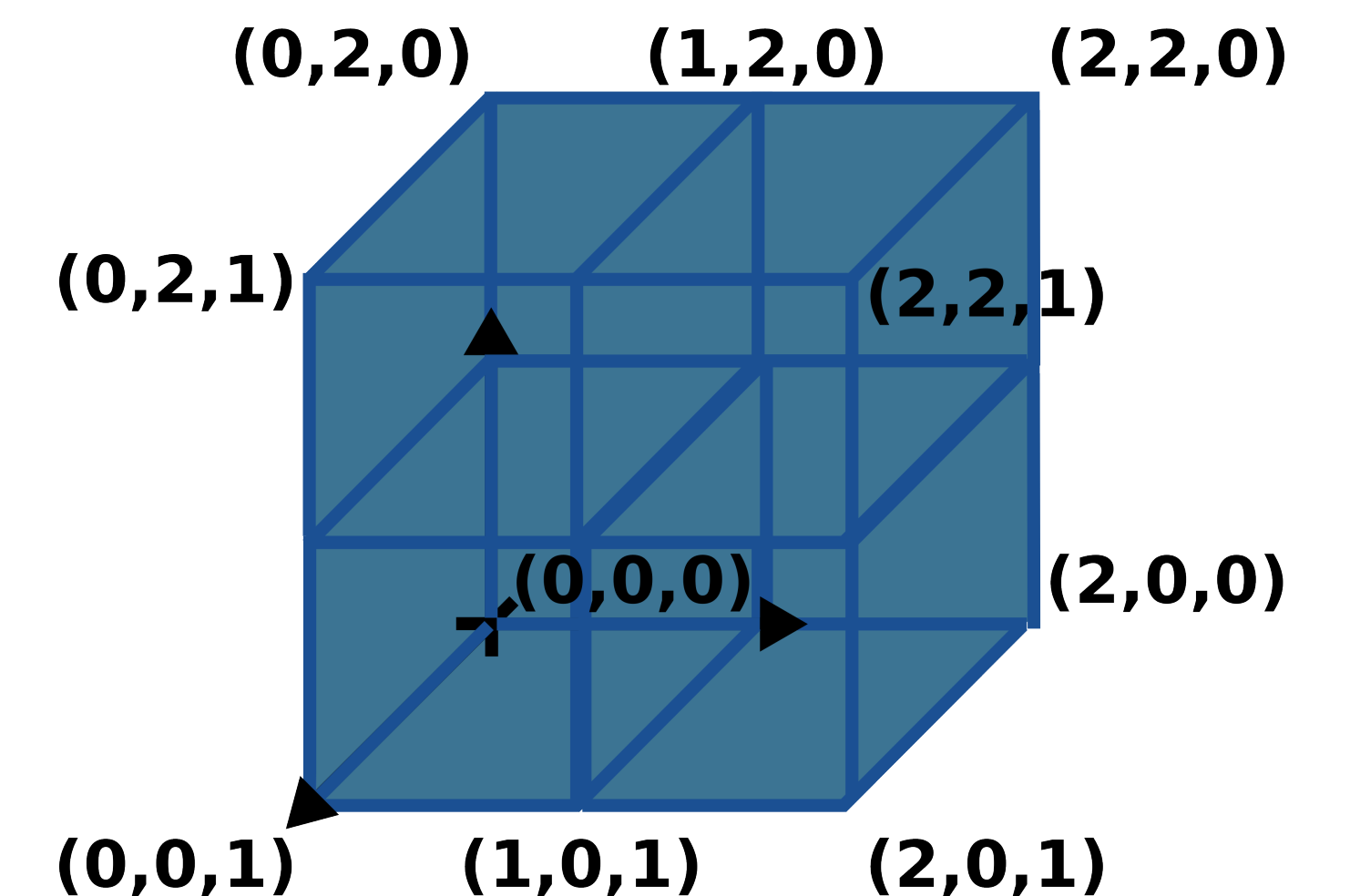
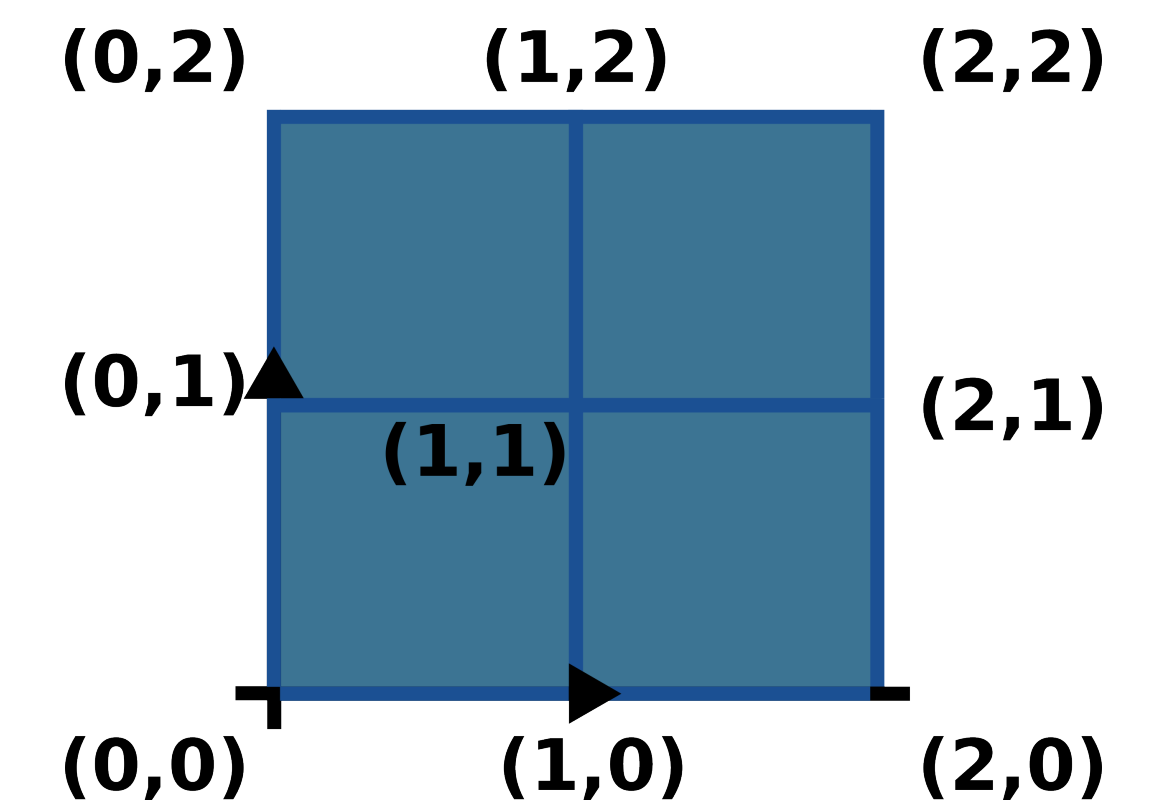
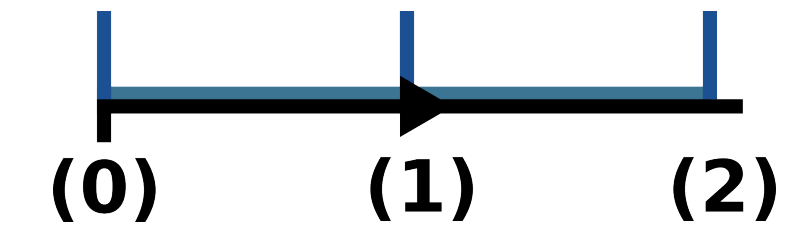
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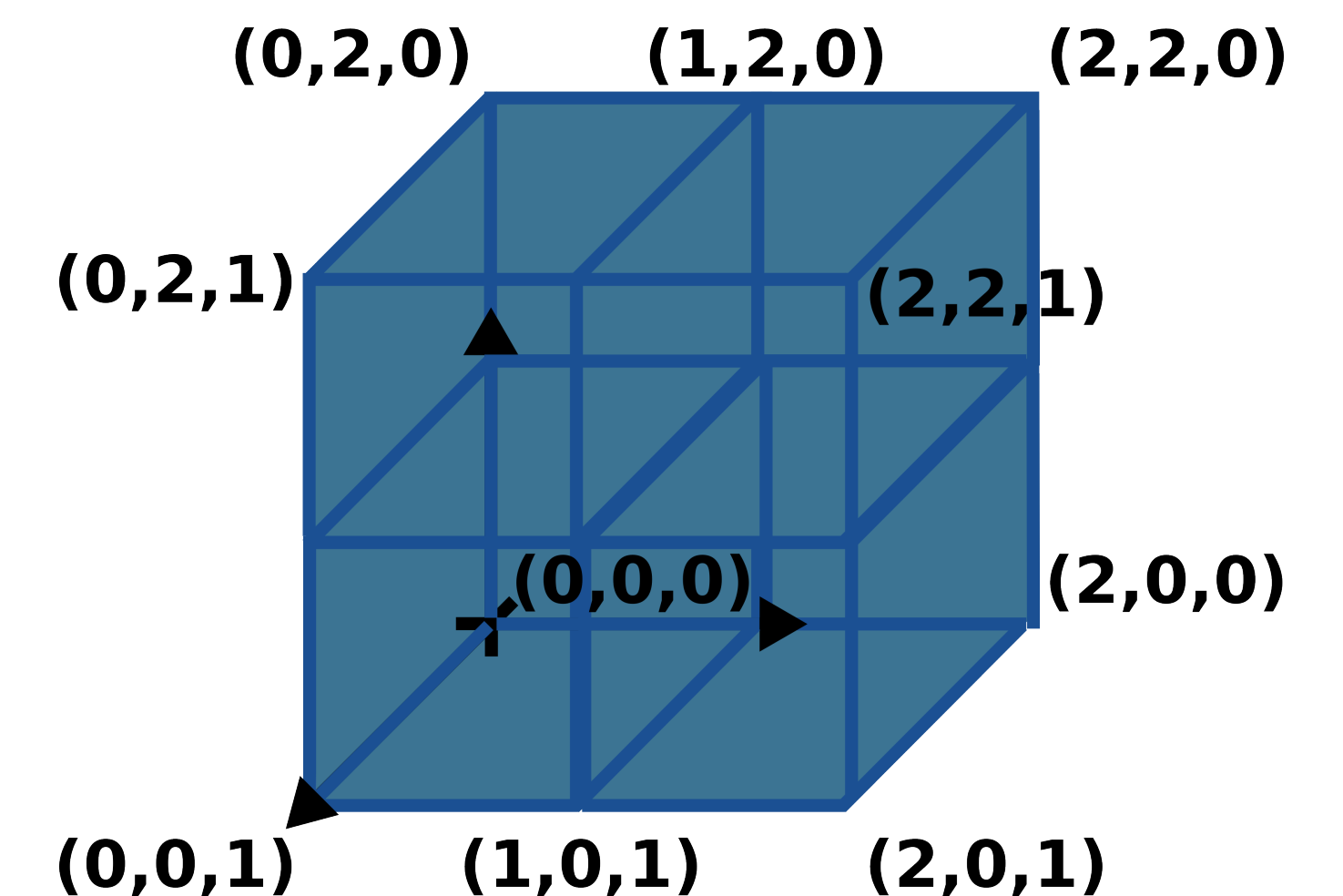
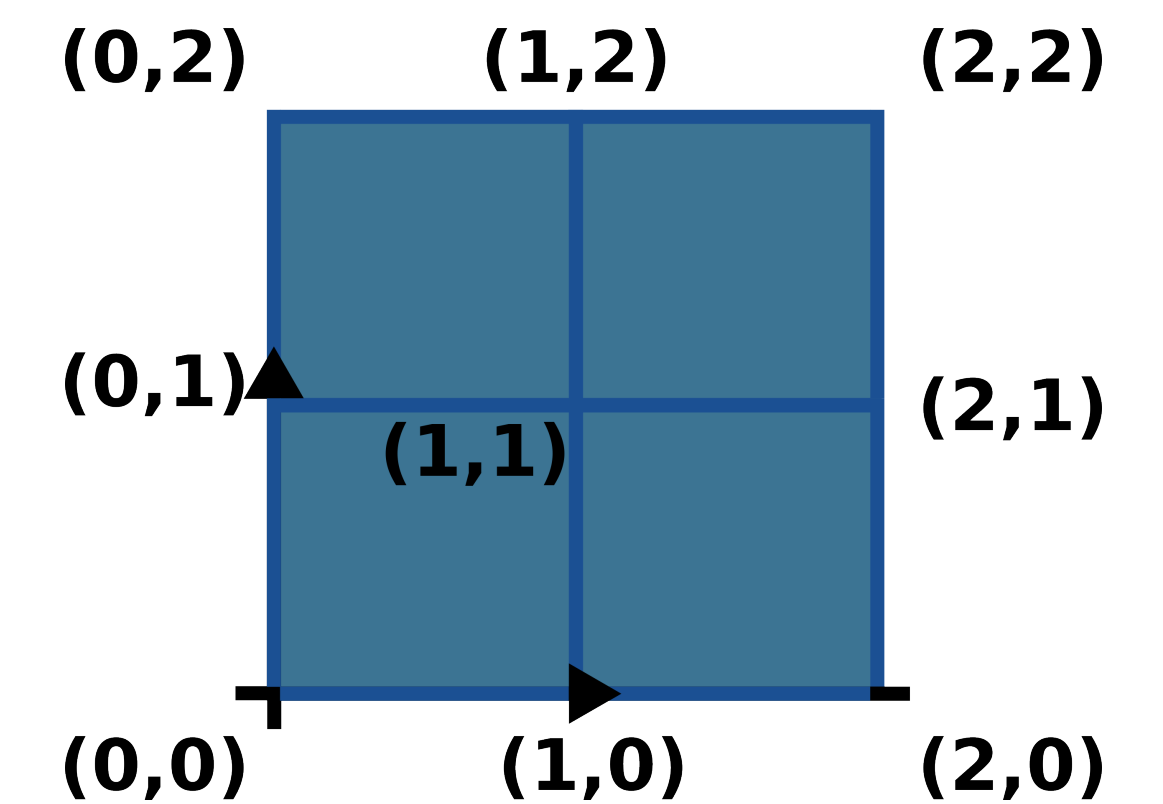
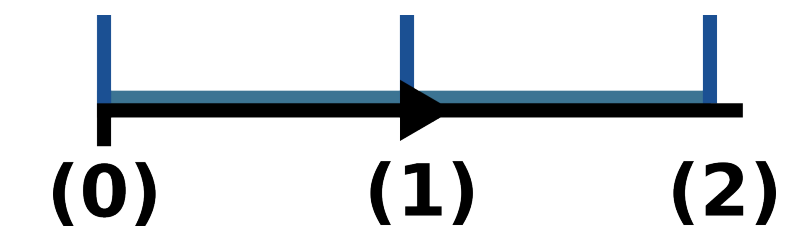
Euclidean spaces on a computer

- Notion of **regular grid** \mathcal{G}
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 - With embedding functions to \mathbb{R}^n
 - **Connectivity**
 - Unit cells (pixels, voxels)
 - Implicitly encoded by the samples (space fill)
 - **Interpolant**



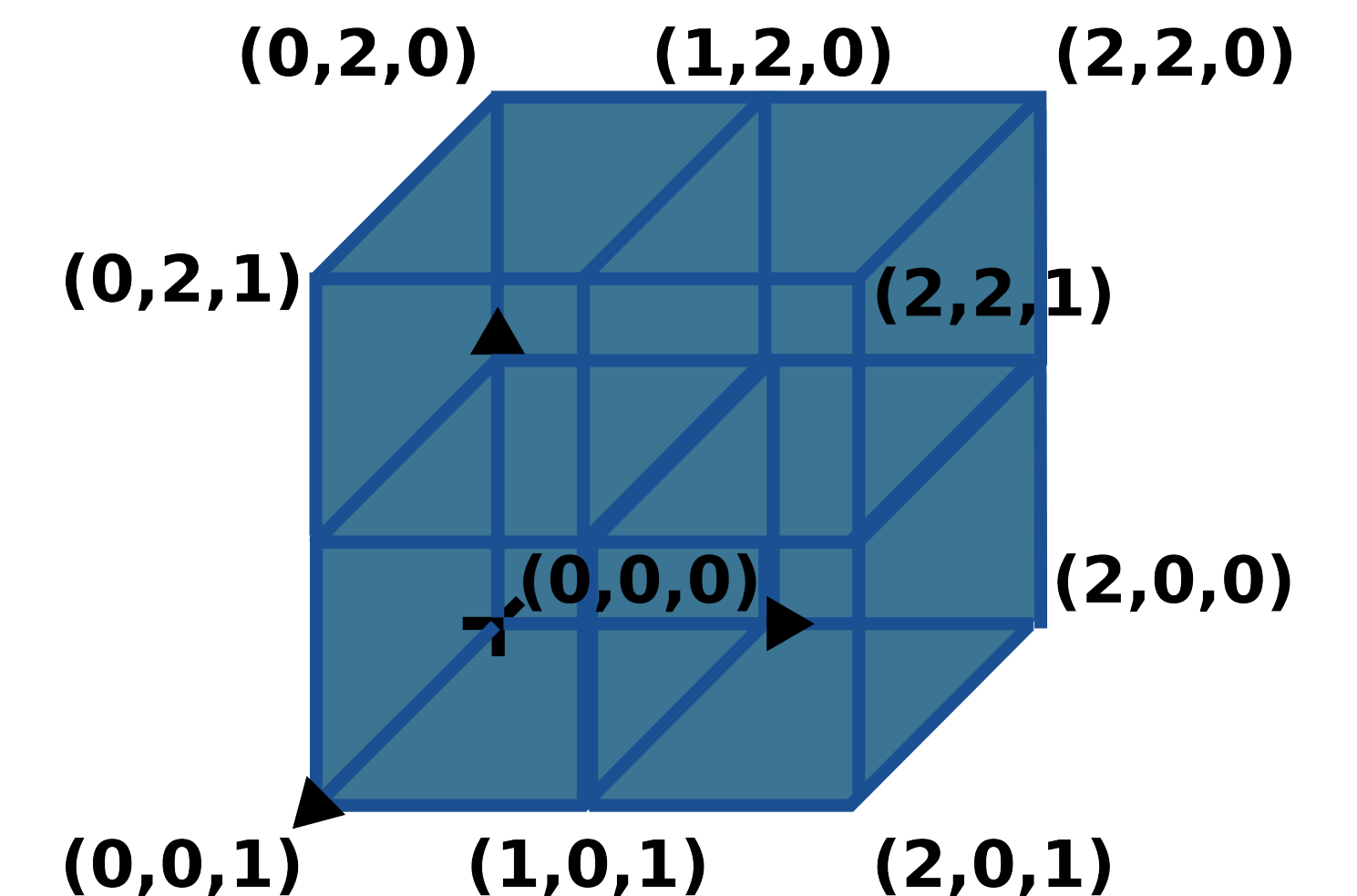
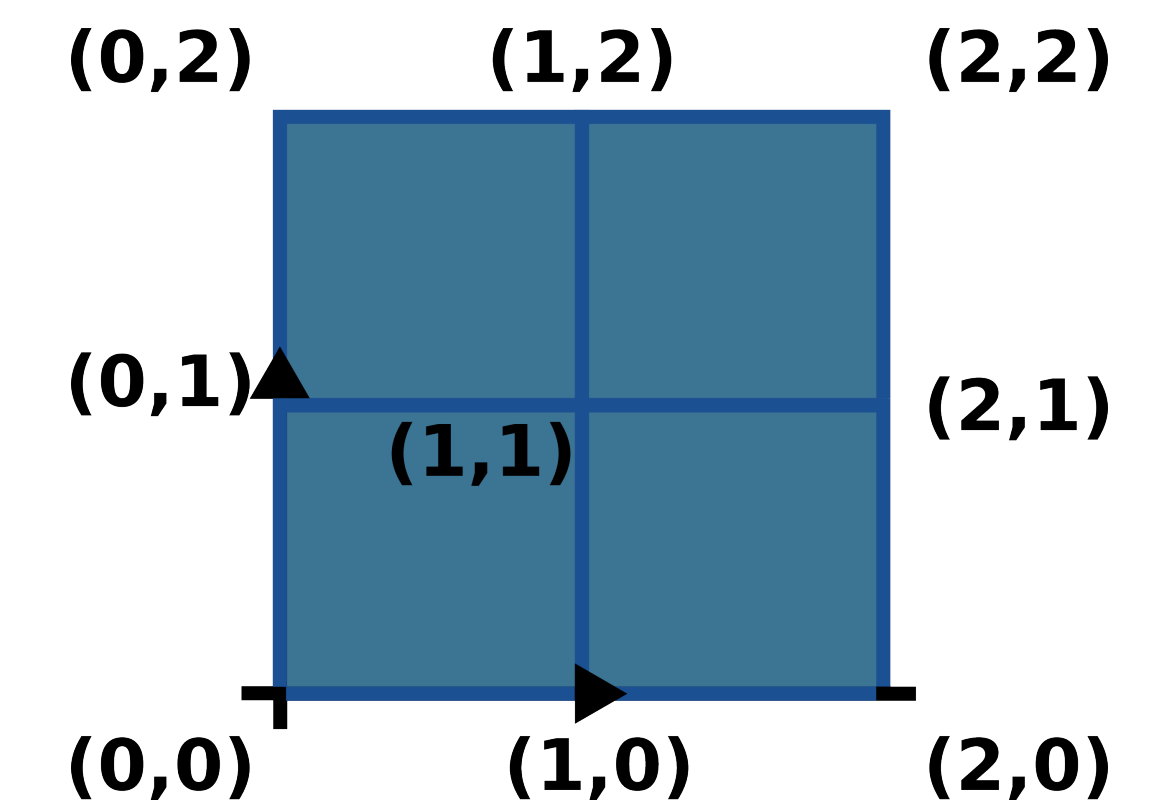
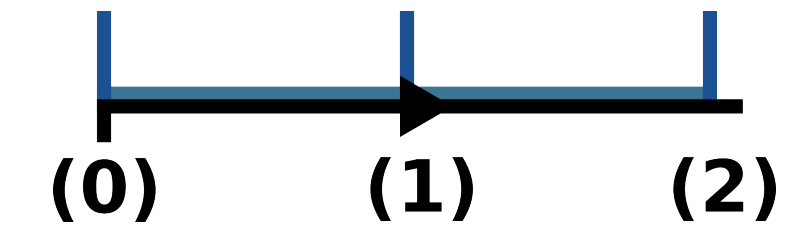
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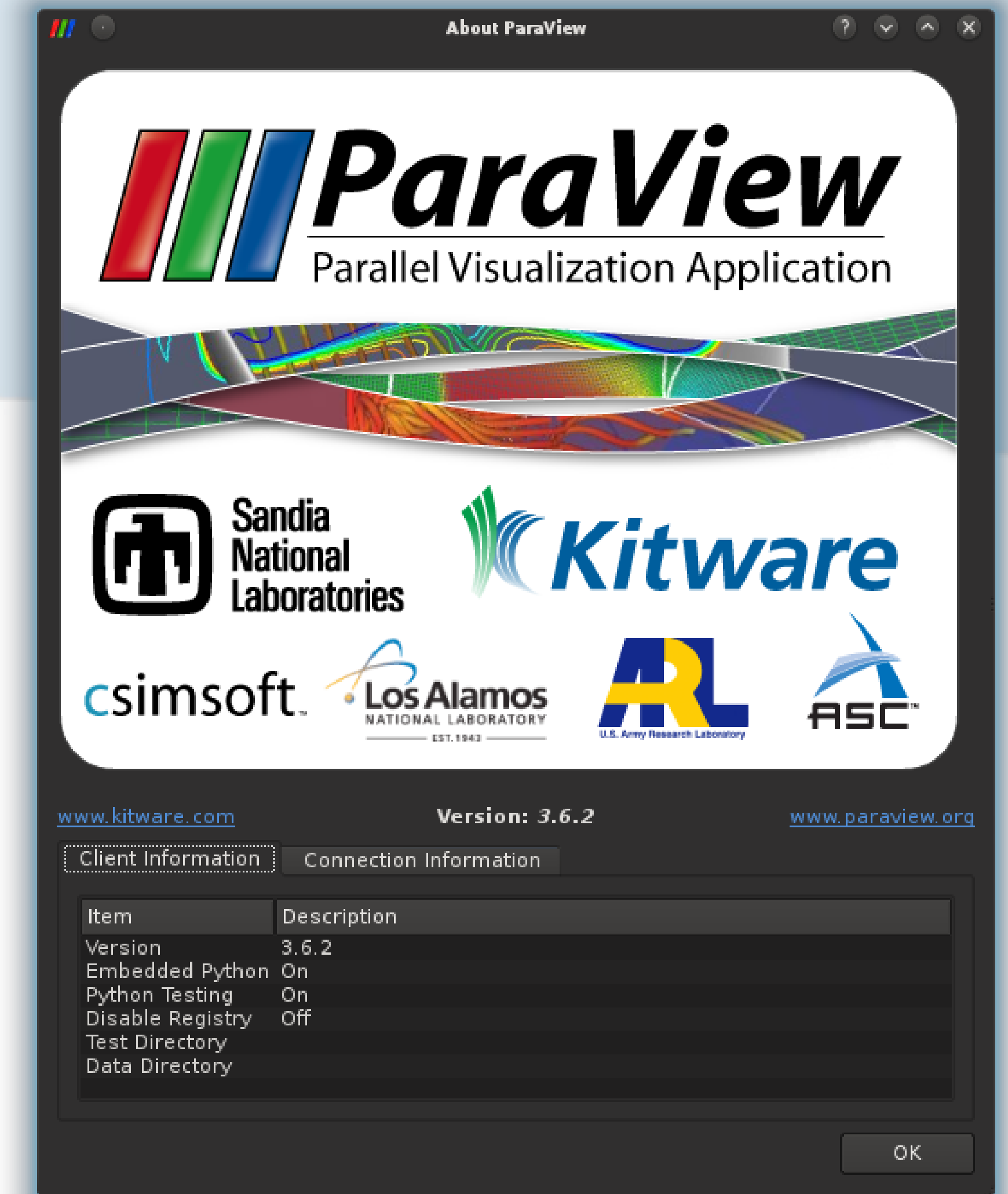
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 - Usually linear functions



In practice

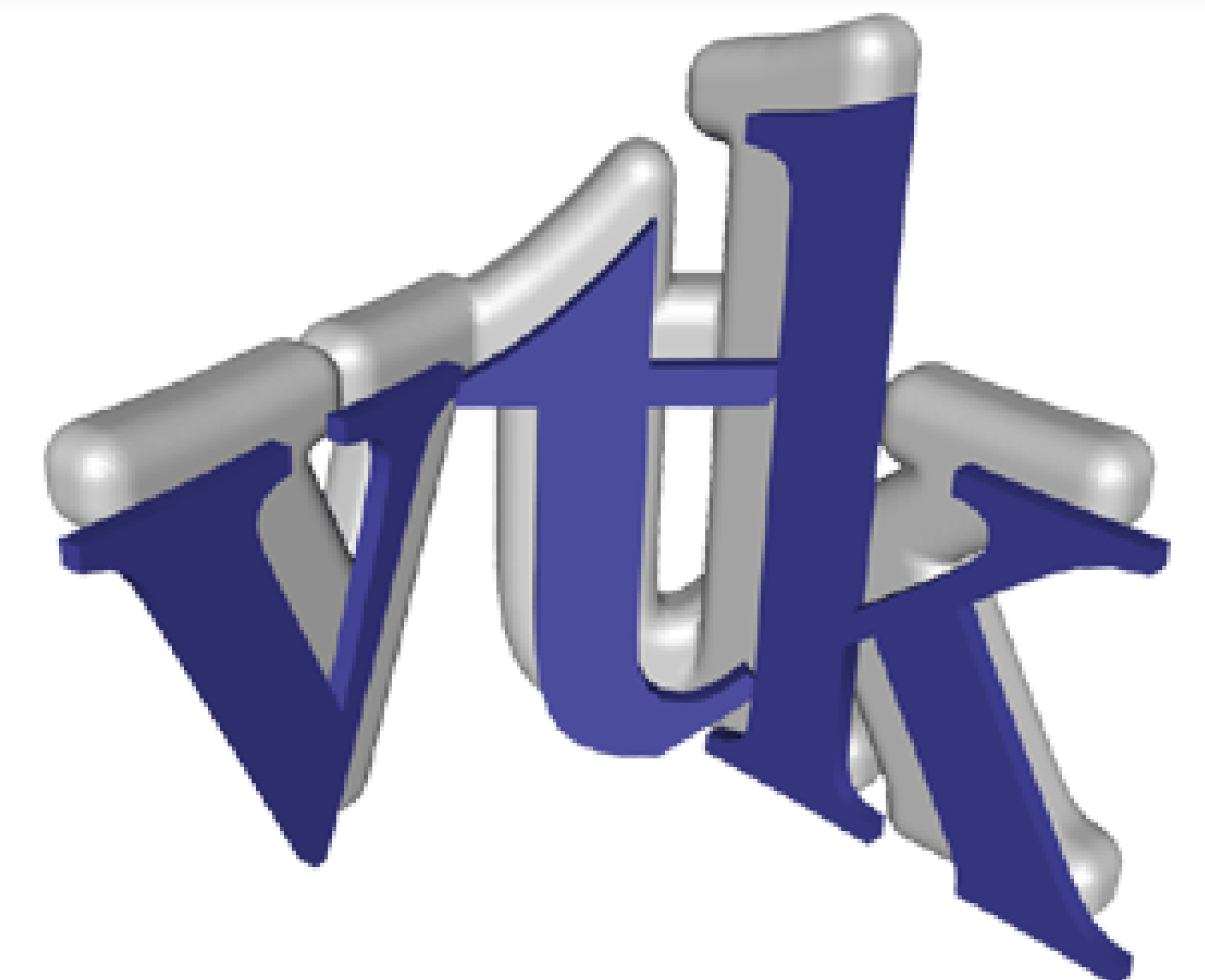
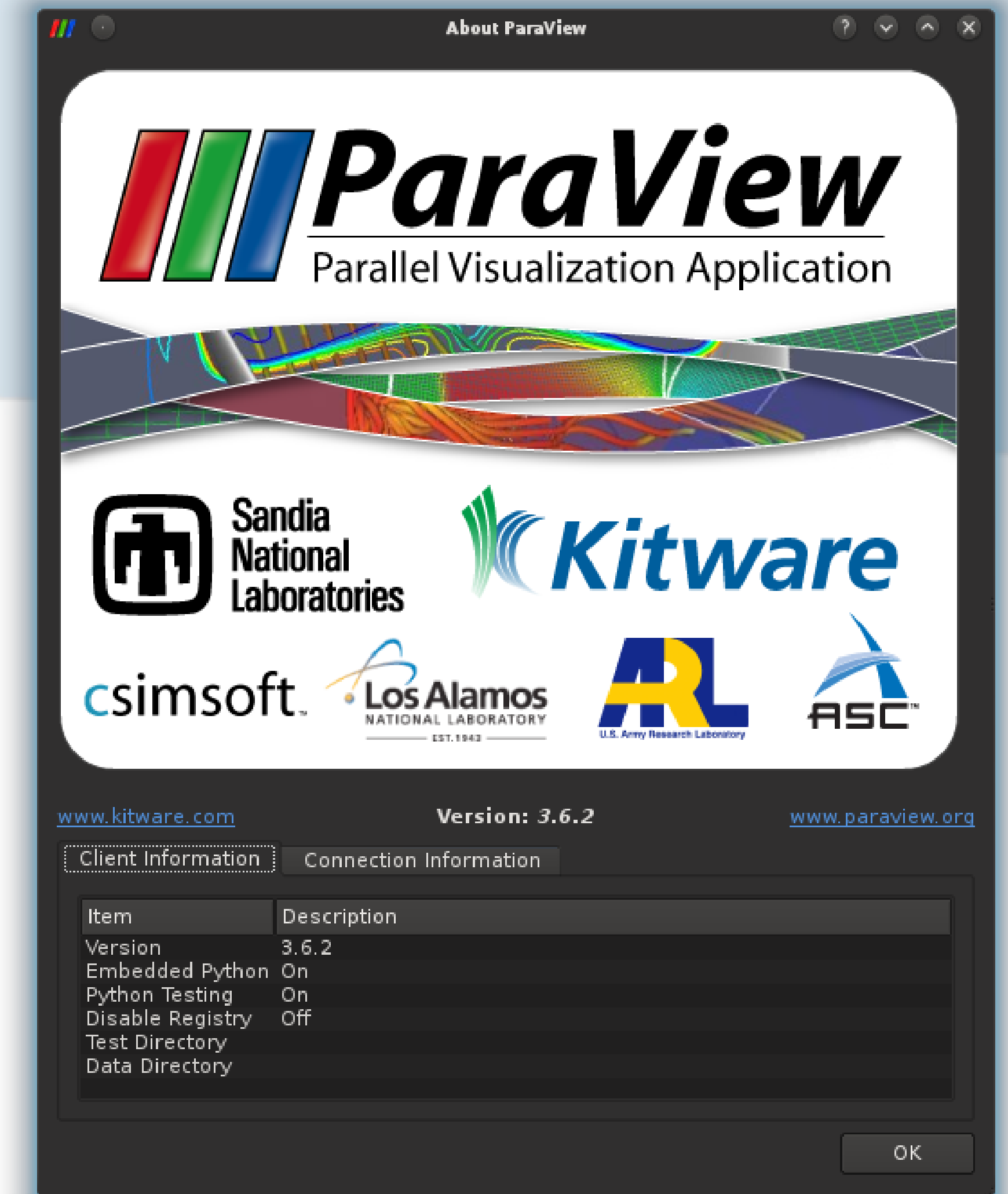
In practice

- Paraview
 - Open-source, interactive visualization platform



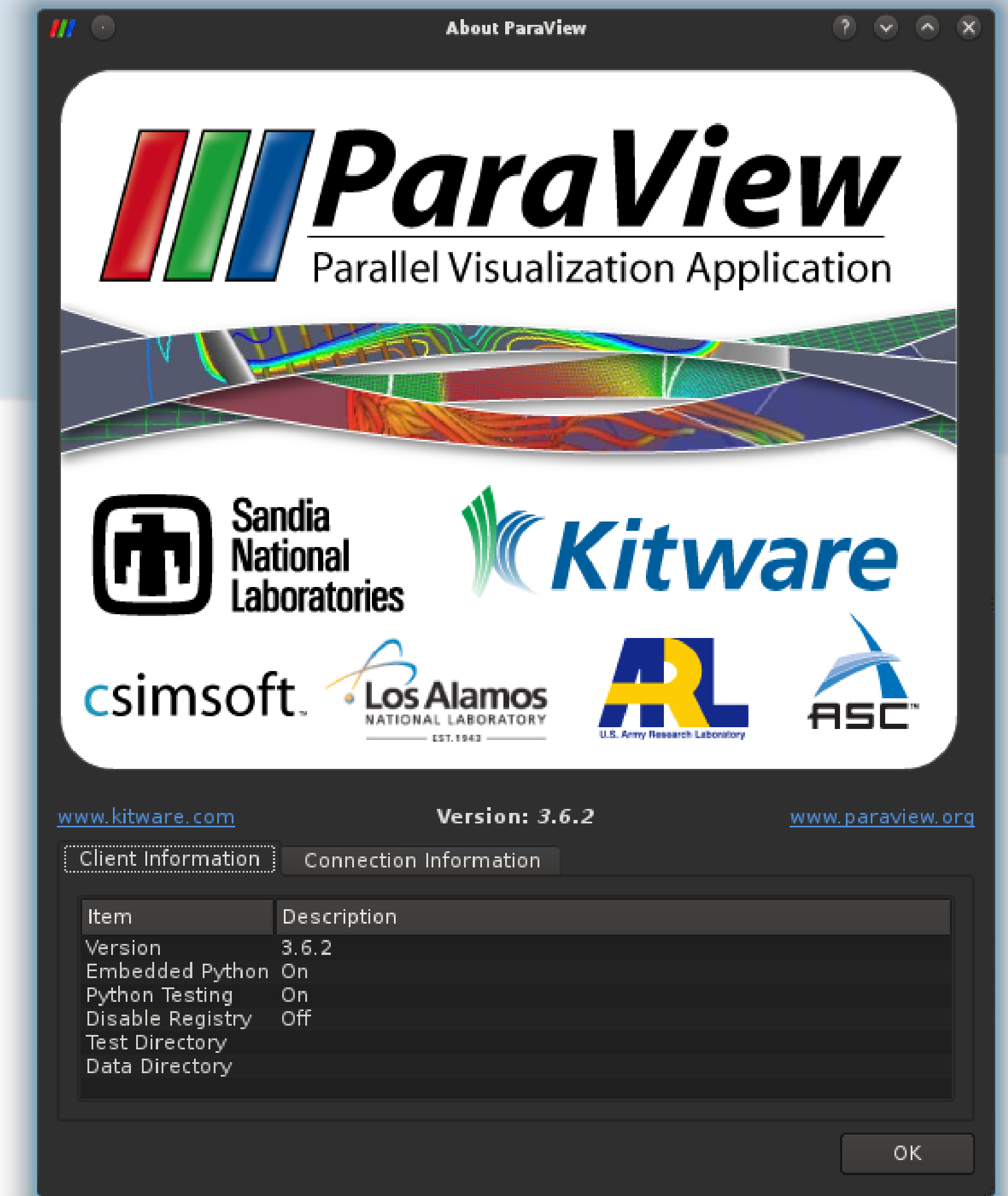
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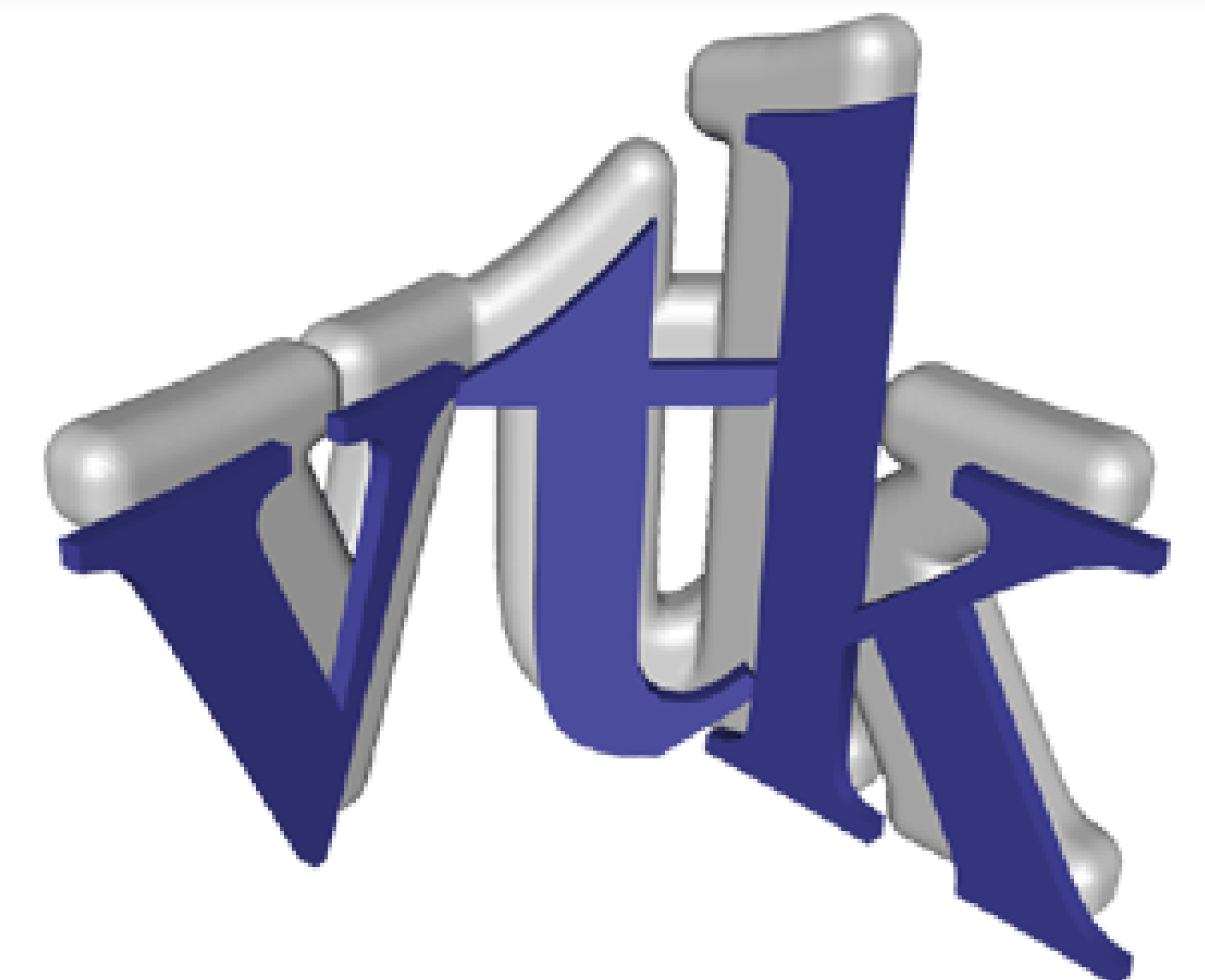
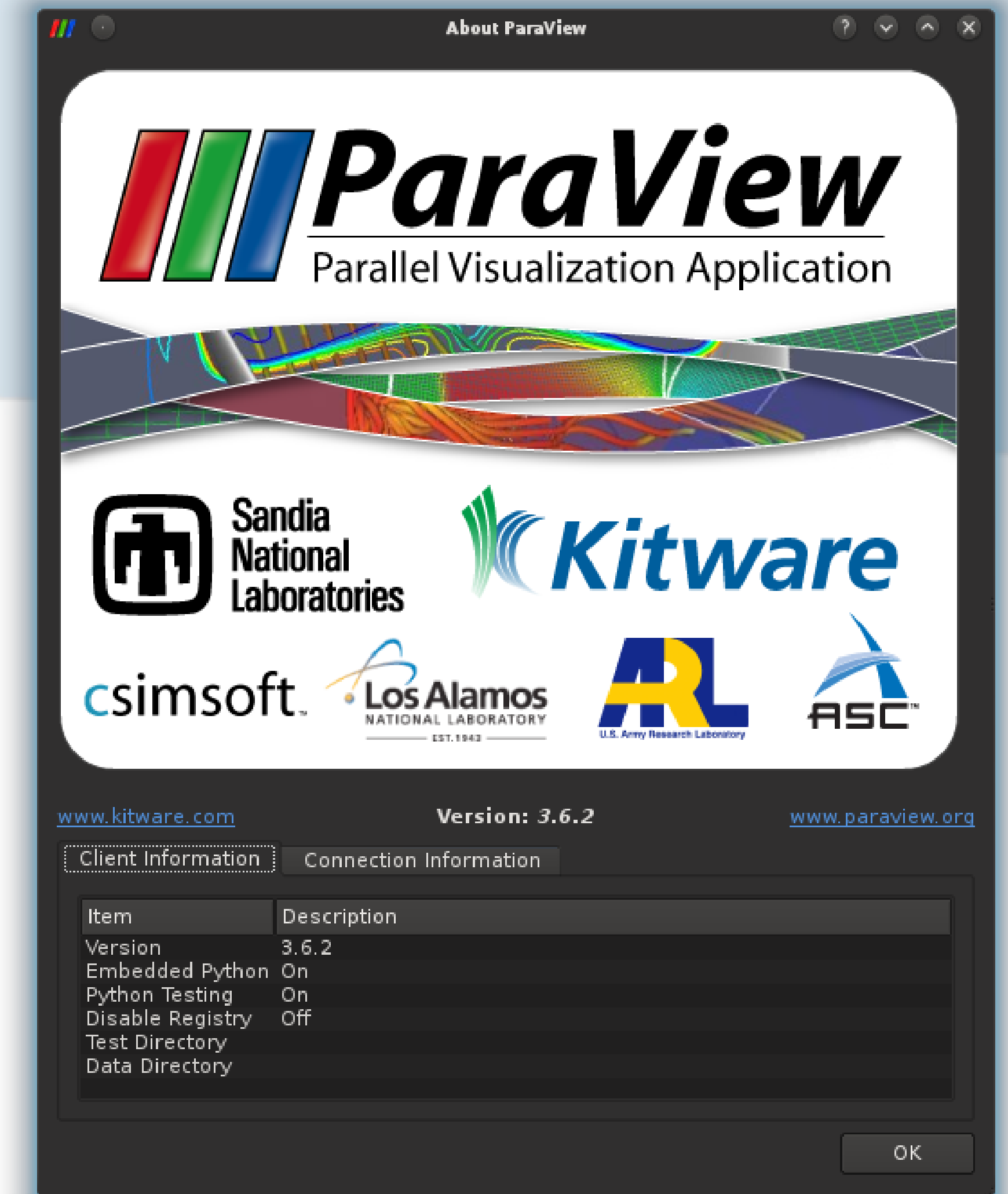
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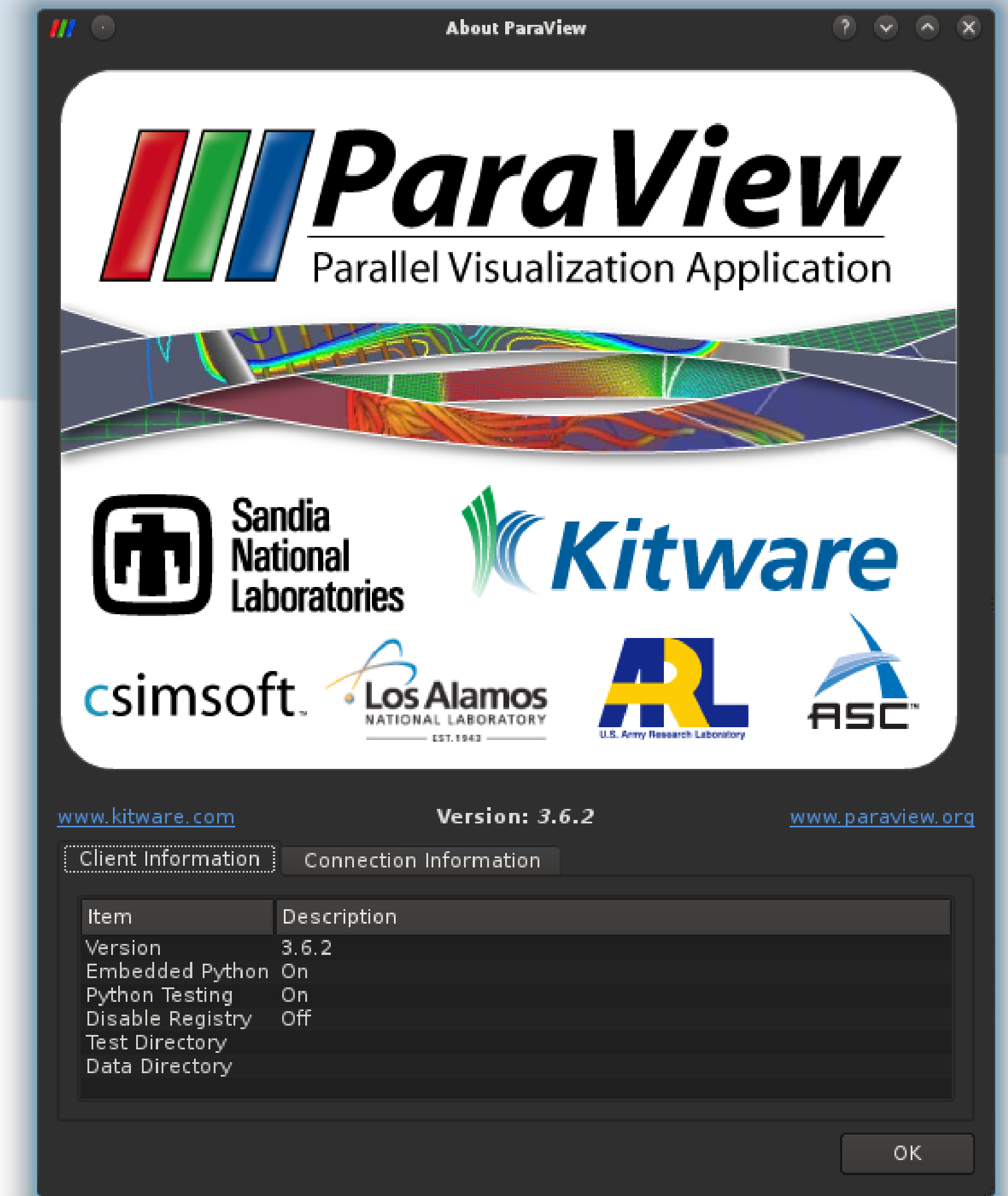
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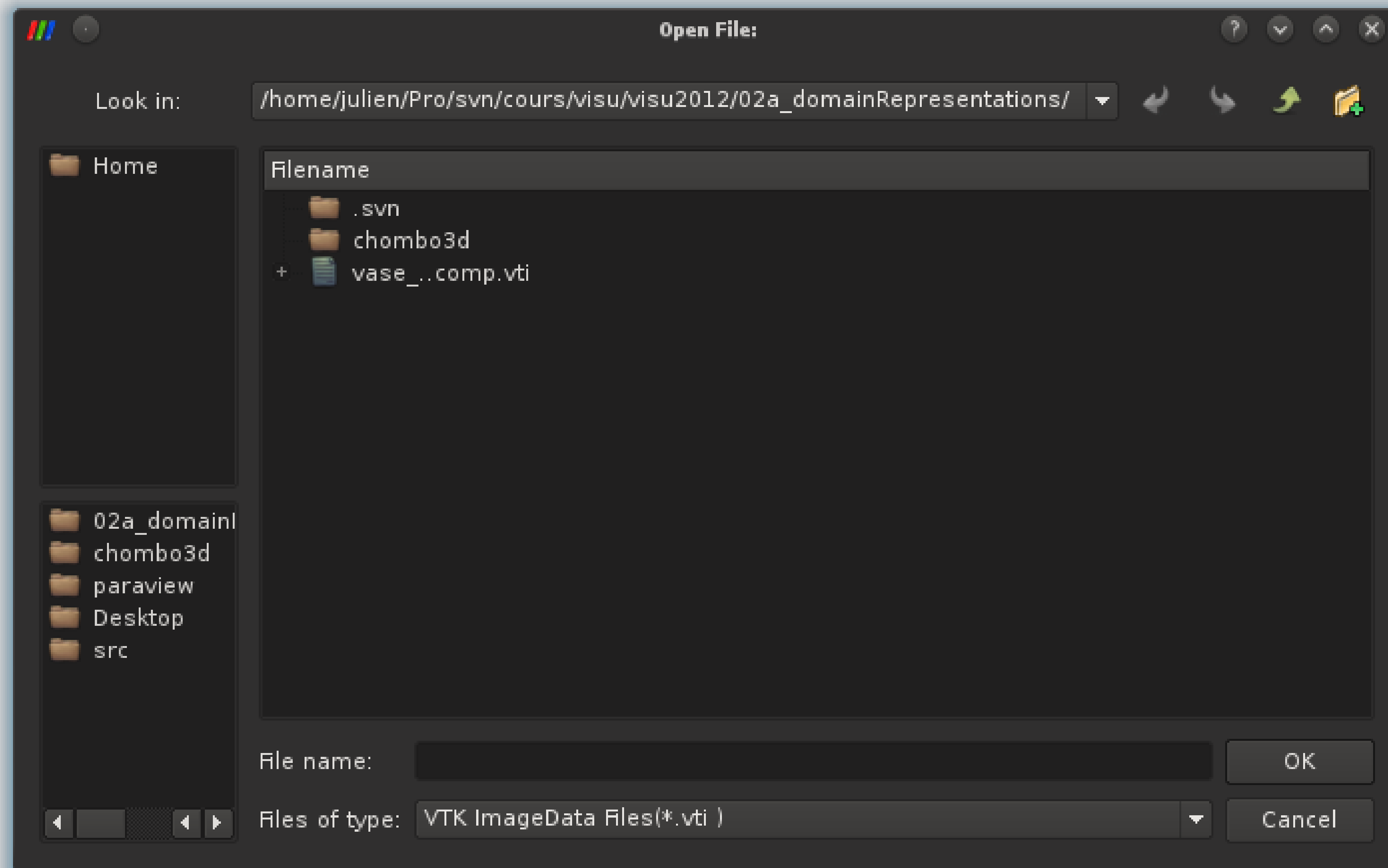


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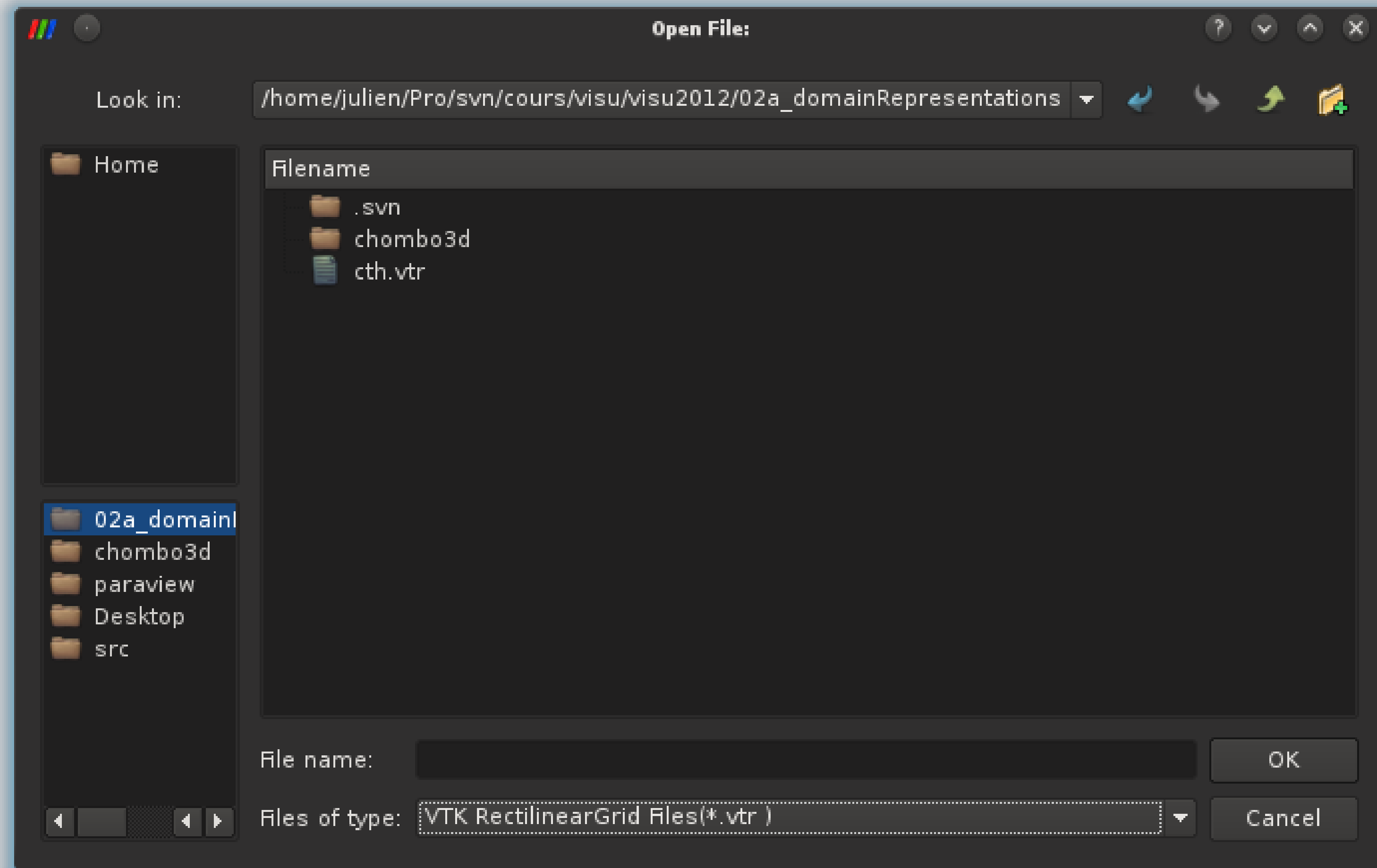
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 - Tens of thousands of users worldwide



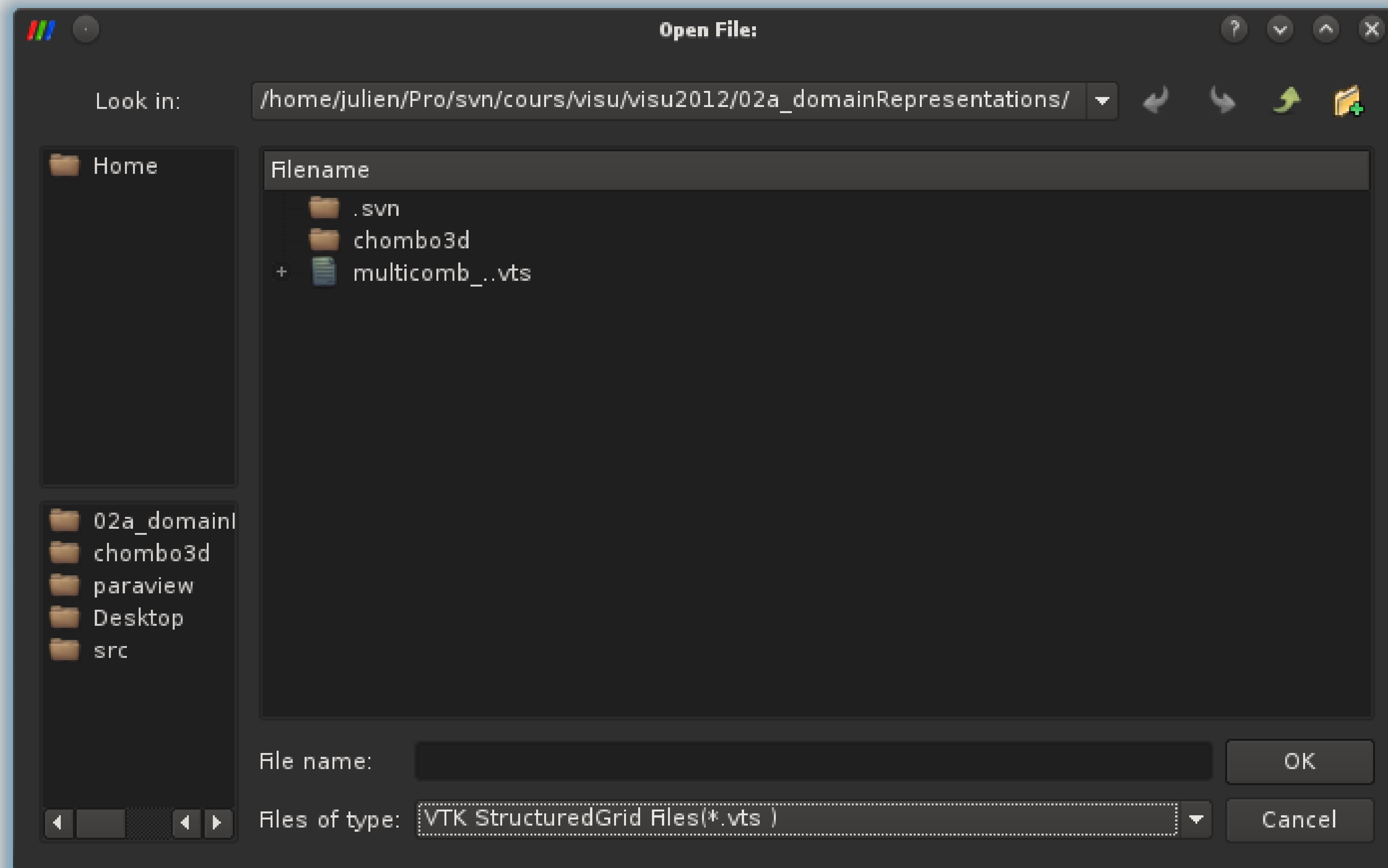
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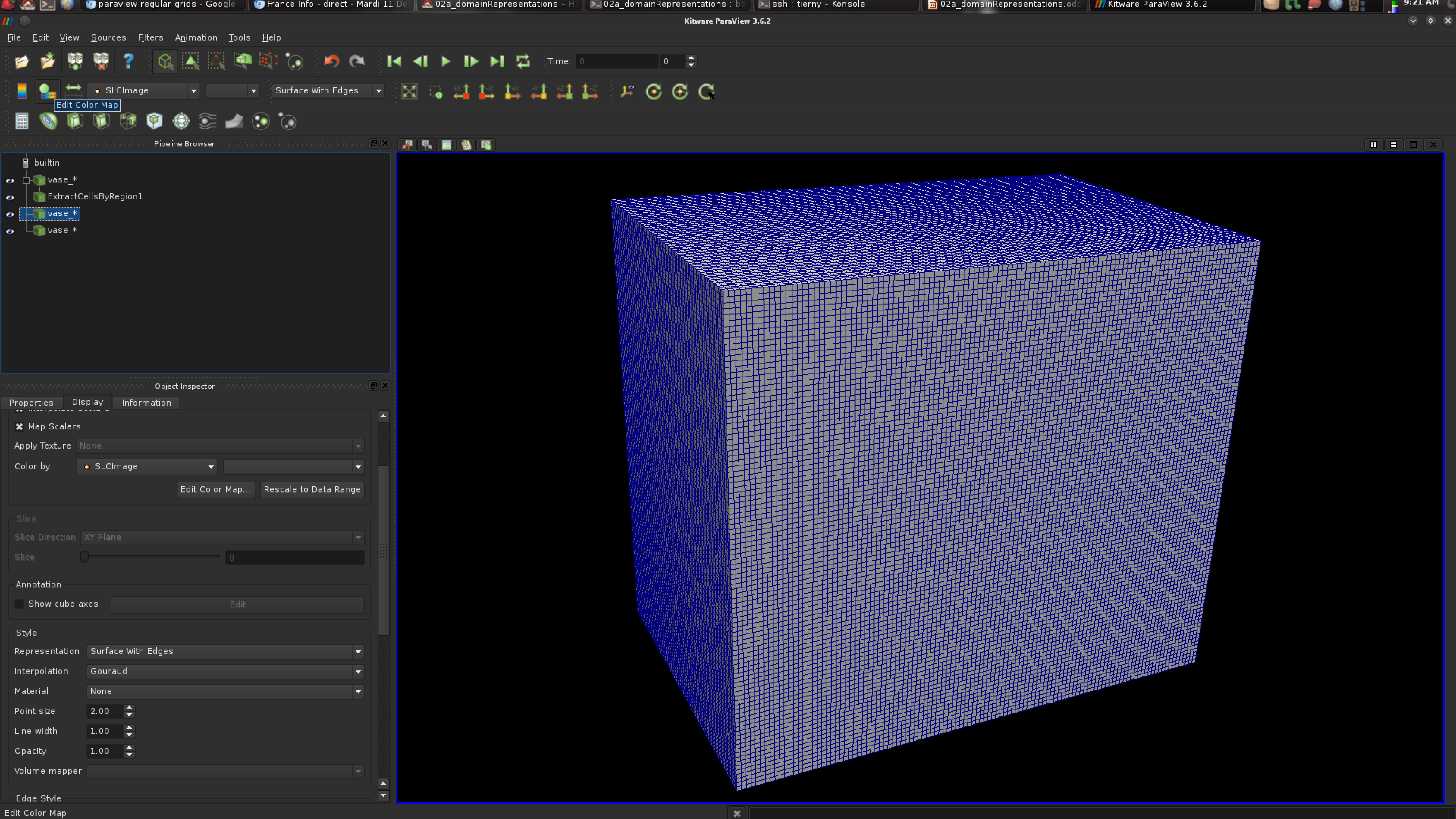


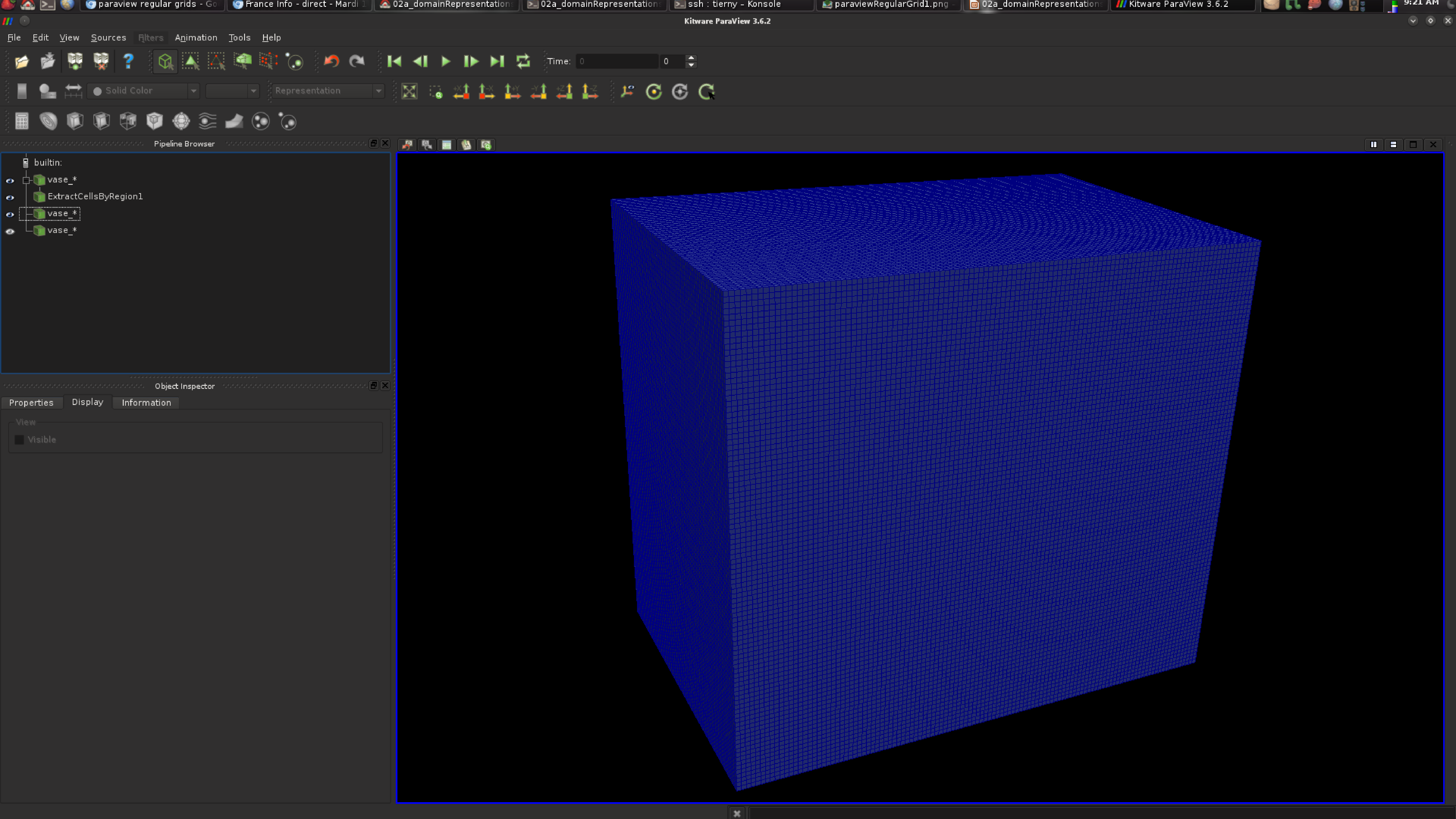
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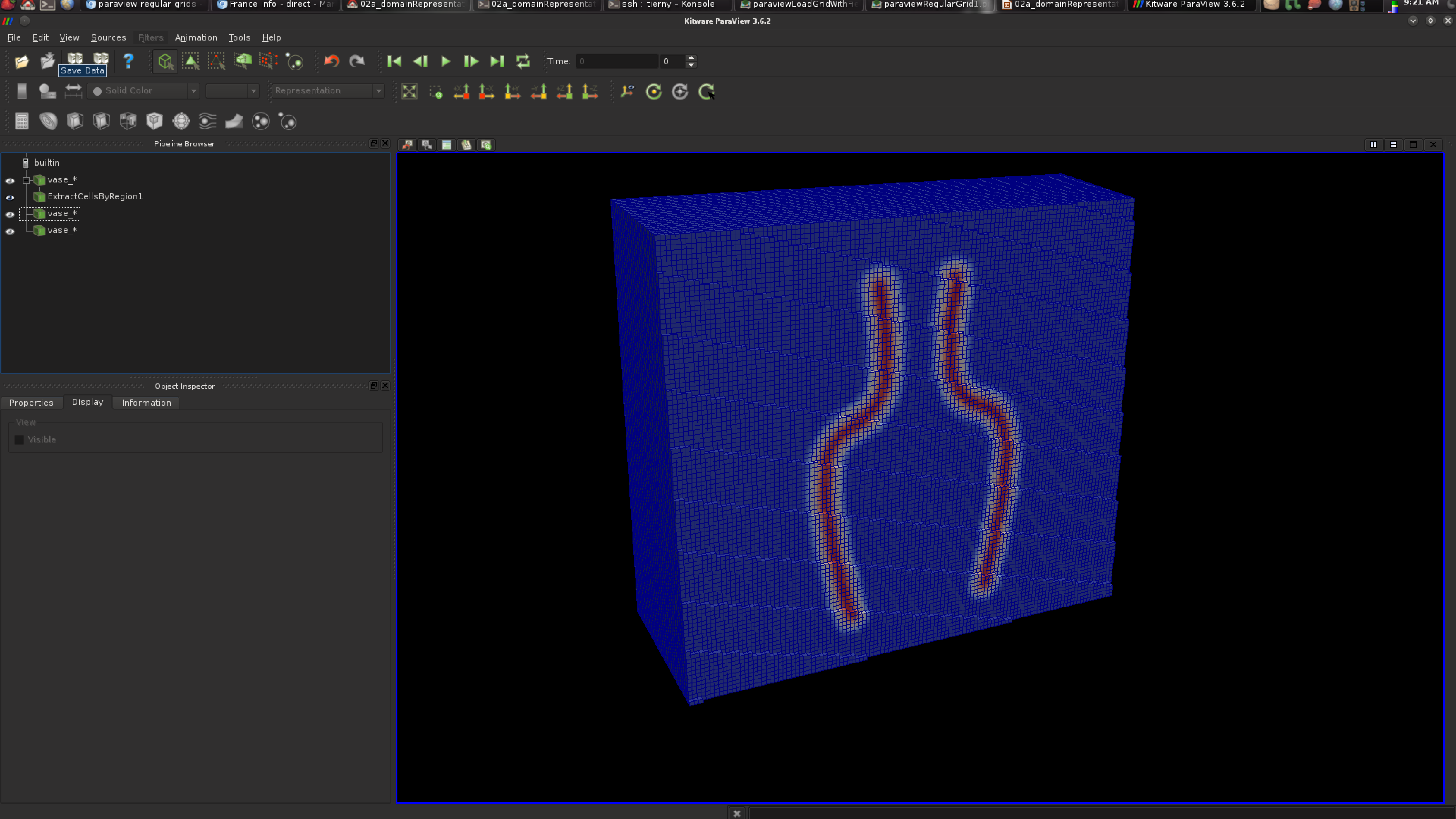


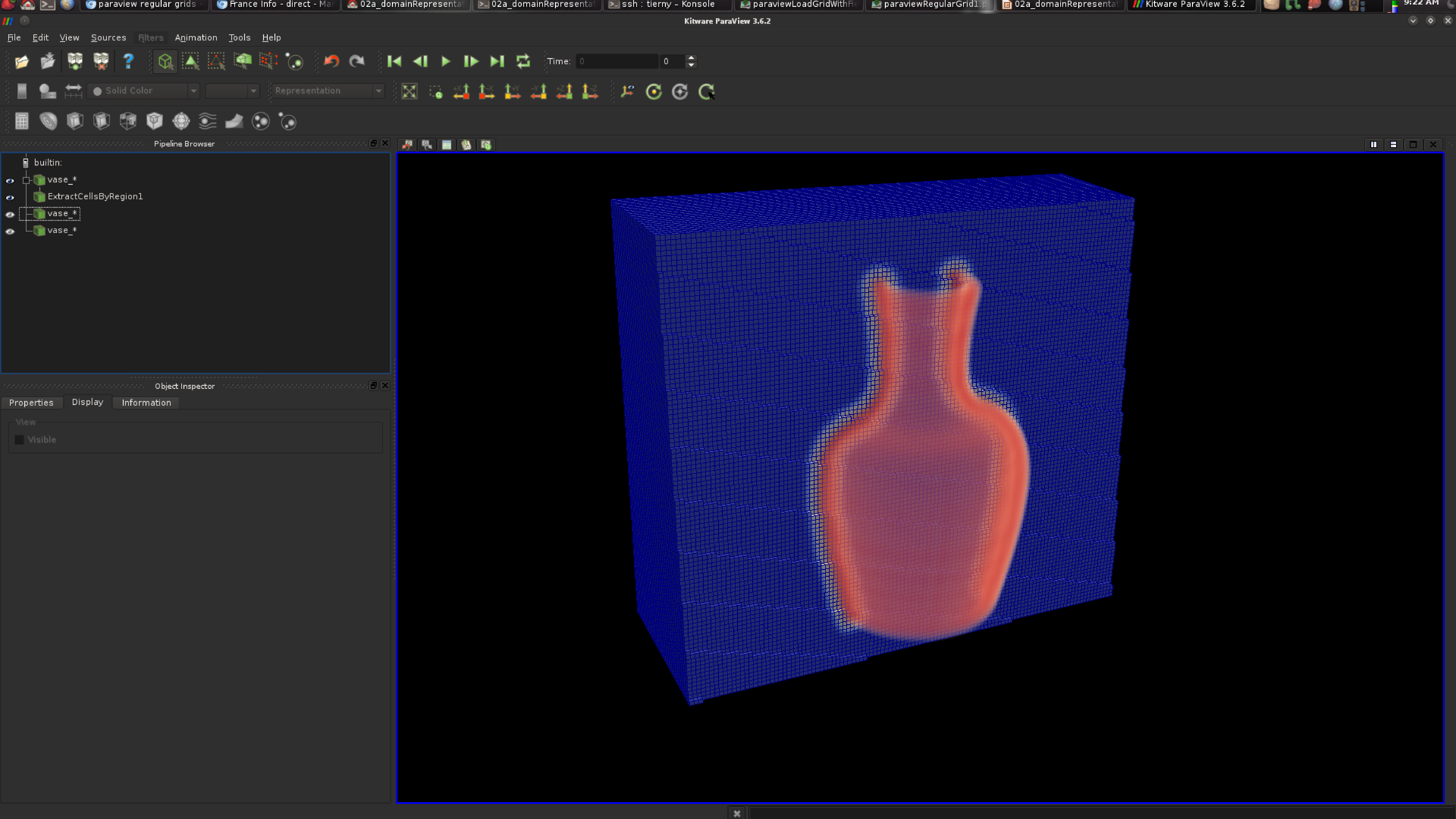
In practice











VTk

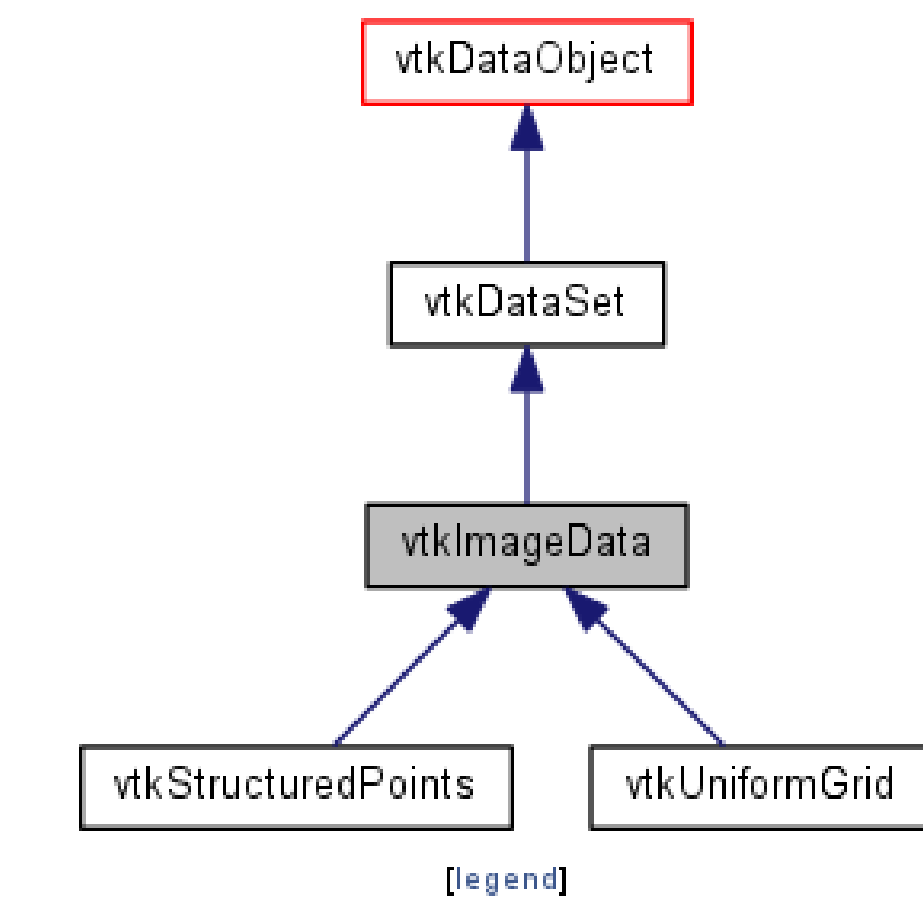
Main Page	Related Pages	Namespaces	Classes	Files
Class List	Class Index	Class Hierarchy	Class Members	

vtkImageData Class Reference

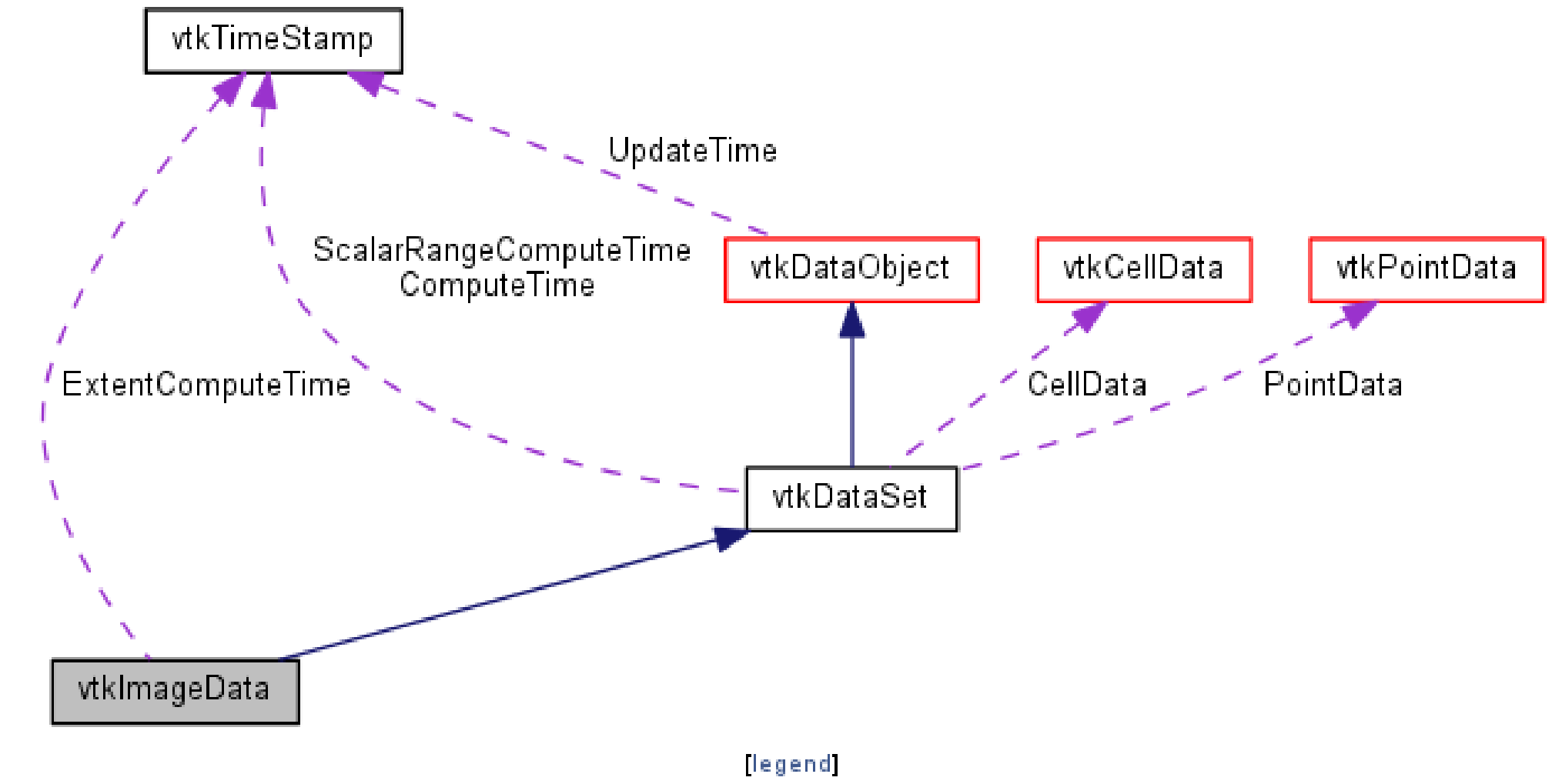
topologically and geometrically regular array of data [More...](#)

#include `<vtkImageData.h>`

Inheritance diagram for vtkImageData:



Collaboration diagram for vtkImageData:



Public Types

typedef `vtkDataSet` [Superclass](#)

Public Member Functions

virtual int	IsA (const char *type)
vtkImageData *	NewInstance () const
void	PrintSelf (ostream &os, vtkIndent indent)
virtual void	CopyStructure (vtkDataSet *ds)
virtual int	GetDataObjectType ()
virtual void	Initialize ()
virtual void	SetDimensions (int i, int j, int k)
virtual void	SetDimensions (const int dims[3])
virtual int	GetDataDimension ()
virtual void	AllocateScalars (int dataType, int numComponents)
virtual void	AllocateScalars (vtkInformation *pipeline_info)
virtual void	Crop (const int *updateExtent)
virtual unsigned long	GetActualMemorySize ()
int	GetScalarType ()
const char *	GetScalarTypeAsString ()
virtual void	CopyInformationFromPipeline (vtkInformation *information)
virtual void	PrepareForNewData ()
void	GetArrayIncrements (vtkDataArray *array, vtkIdType increments[3])
void	ComputeInternalExtent (int *intExt, int *tgtExt, int *bnds)
virtual int	GetExtentType ()
virtual vtkIdType	GetNumberOfCells ()
virtual vtkIdType	GetNumberOfPoints ()
virtual double *	GetPoint (vtkIdType ptId)
virtual void	GetPoint (vtkIdType id, double x[3])
virtual vtkCell *	GetCell (vtkIdType cellId)
virtual void	GetCell (vtkIdType cellId, vtkGenericCell *cell)
virtual void	GetCellBounds (vtkIdType cellId, double bounds[6])
virtual vtkIdType	FindPoint (double x, double y, double z)
virtual vtkIdType	FindPoint (double x[3])
virtual vtkIdType	FindCell (double x[3], vtkCell *cell, vtkIdType cellId, double tol2, int &subId, double pcoords[3], double *weights)
virtual vtkIdType	FindCell (double x[3], vtkCell *cell, vtkGenericCell *gencell, vtkIdType cellId, double tol2, int &subId, double pcoords[3], double *weights)
virtual vtkCell *	FindAndGetCell (double x[3], vtkCell *cell, vtkIdType cellId, double tol2, int &subId, double pcoords[3], double *weights)
virtual int	GetCellType (vtkIdType cellId)
virtual void	GetCellPoints (vtkIdType cellId, vtkIdList *ptIds)
virtual void	GetPointCells (vtkIdType ptId, vtkIdList *cellIds)
virtual void	ComputeBounds ()
virtual int	GetMaxCellSize ()
virtual int *	GetDimensions ()
virtual void	GetDimensions (int dims[3])
virtual int	ComputeStructuredCoordinates (const double x[3], int ijk[3], double pcoords[3])

Recap

- Euclidean domains



Recap

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 - Easily represented with regular grids



Recap

- Euclidean domains
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- What about other spaces?



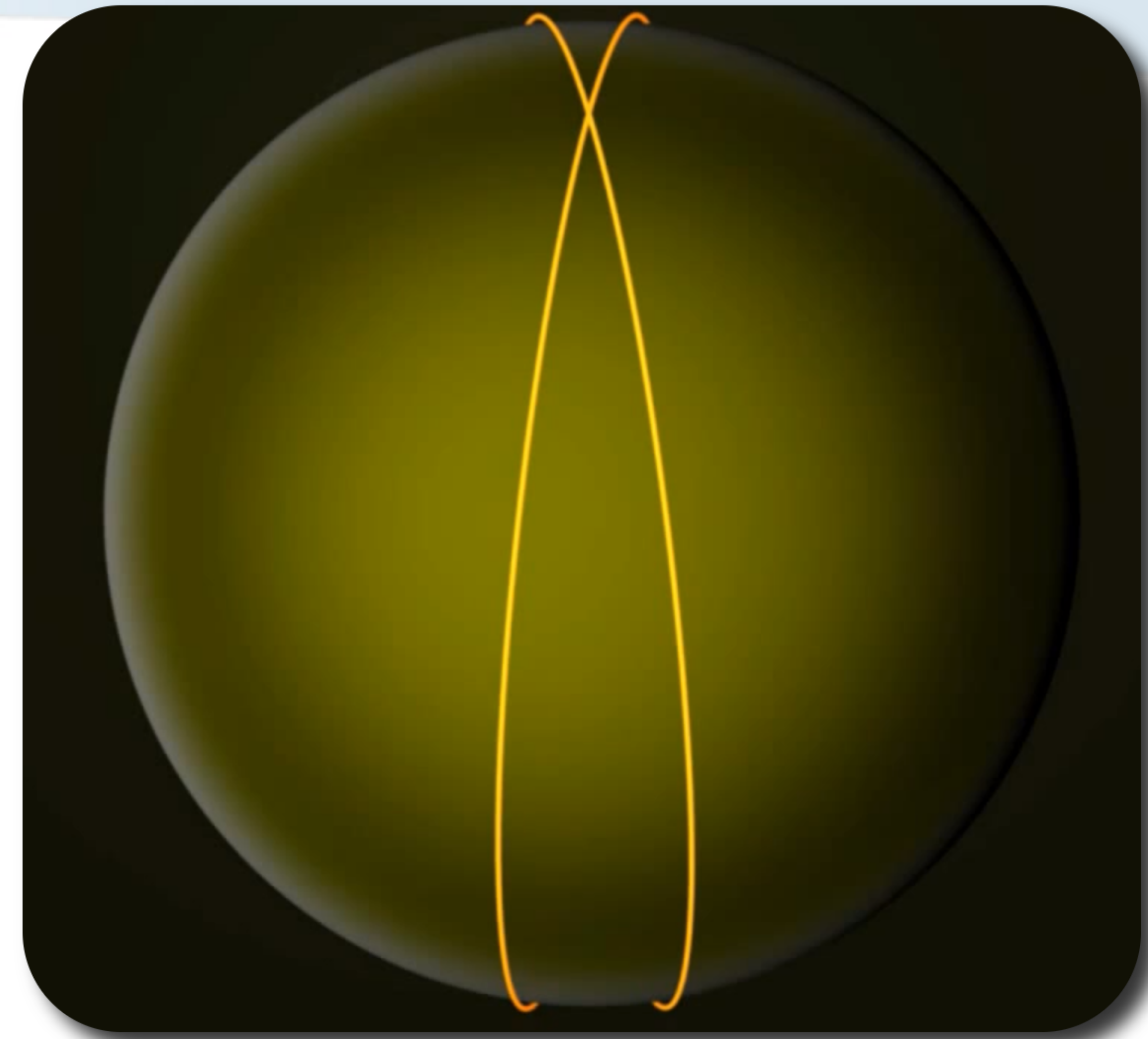
Recap

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 - *Through a point not on a given straight line, at most one line can be drawn that never meets the given line*



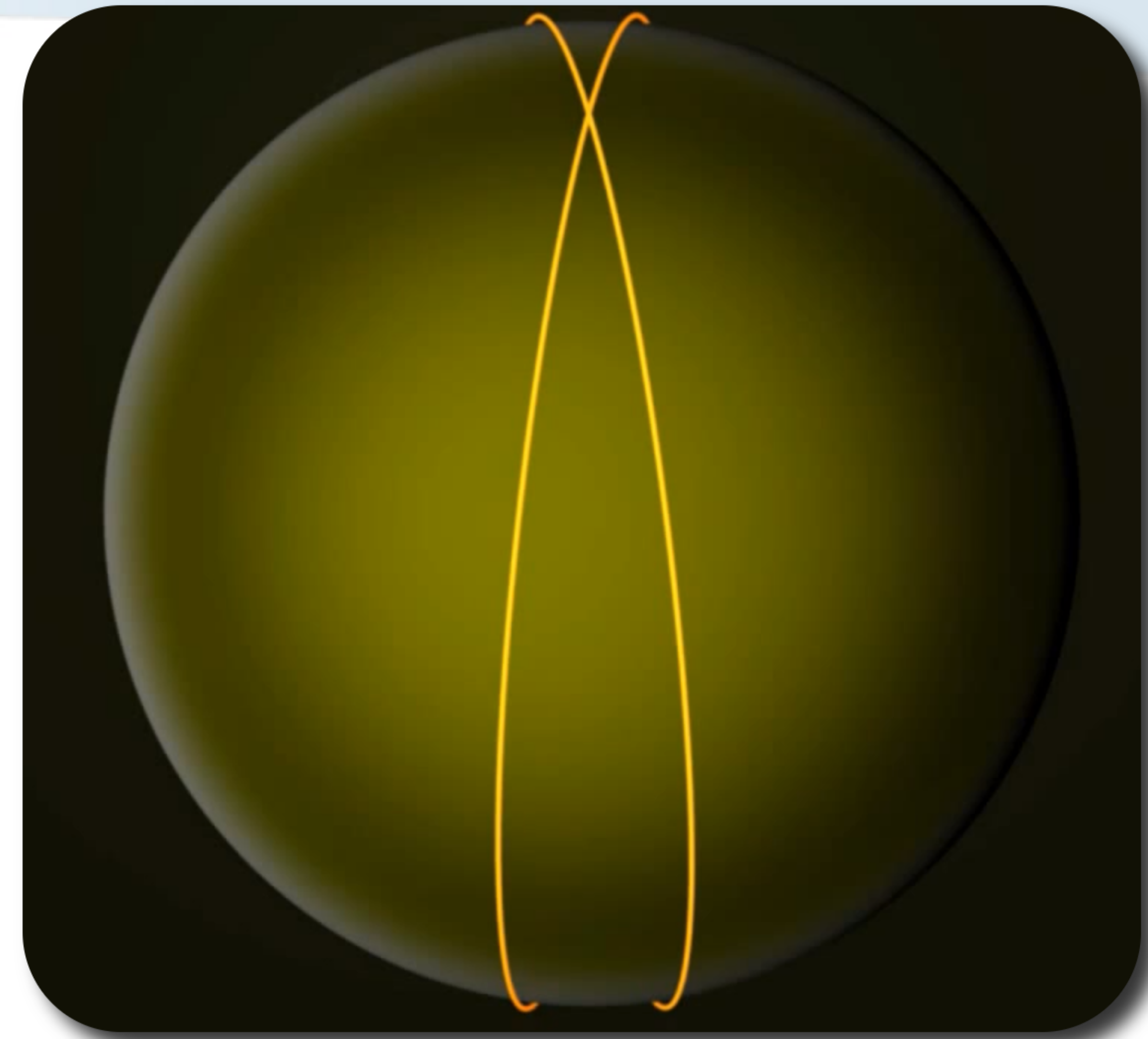
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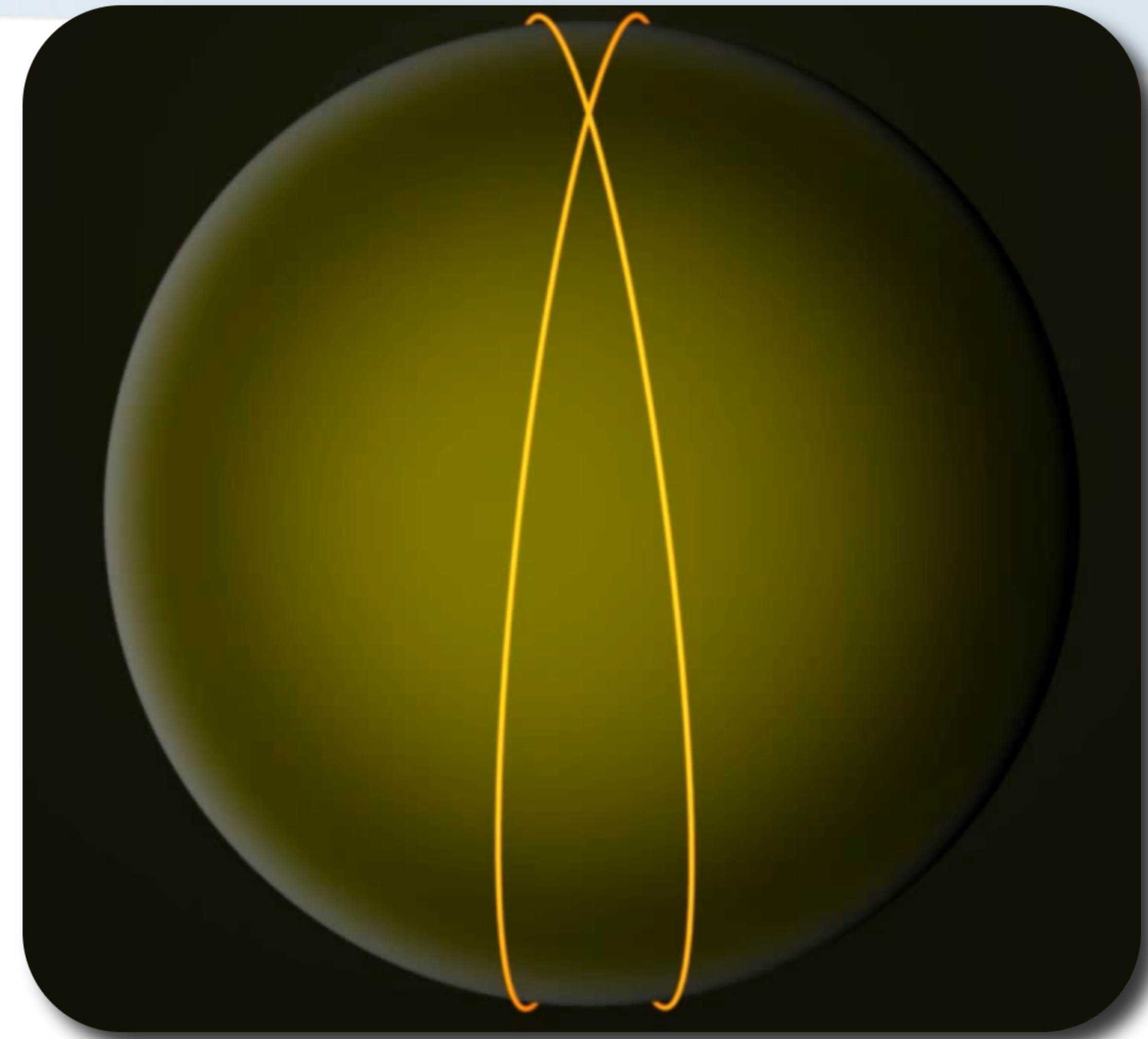
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 - 5th axiom of the *Elements*:
 - *Through a point not on a given straight line, at most one line can be drawn that never meets the given line*
 - What if this axiom was not true?
 - What would geometry look like?



Recap

- Euclidean domains
 - Easily represented with regular grids
- What about other spaces?
 - Riemannian geometry



Recap

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 - 19th century



Recap

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Recap

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 - General notion of manifold



Manifolds on a computer

- Notion of simplex
 - A d -simplex σ is the convex hull of $(d + 1)$ affinely independent points of an euclidean space \mathbb{R}^n
 - $0 \leq d \leq n$
 - d is the dimension of σ



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 - A d -simplex σ is the convex hull of $(d + 1)$ affinely independent points of an euclidean space \mathbb{R}^n
 - $0 \leq d \leq n$
 - d is the dimension of σ
 - Smallest combinatorial construction to represent a cell of dimension d



Manifolds on a computer

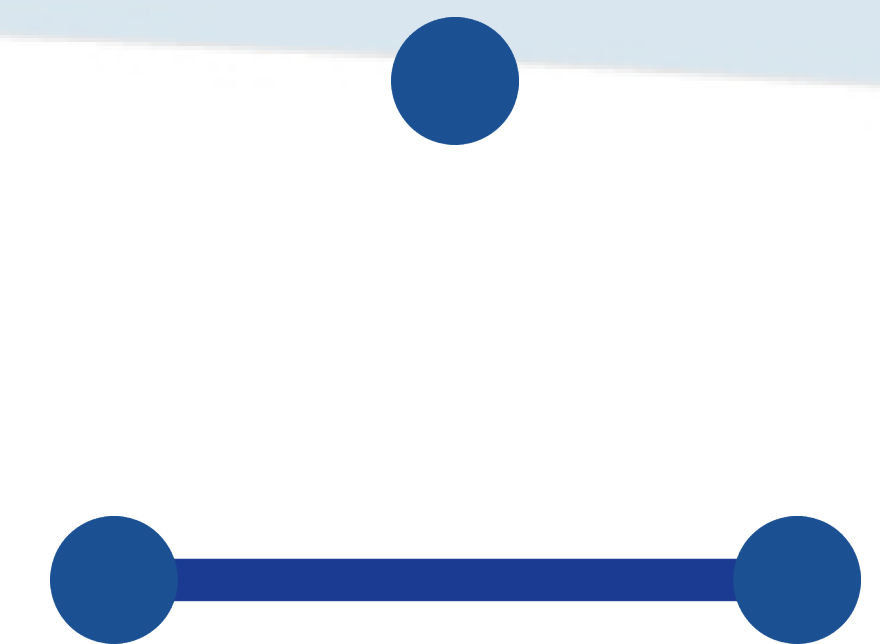
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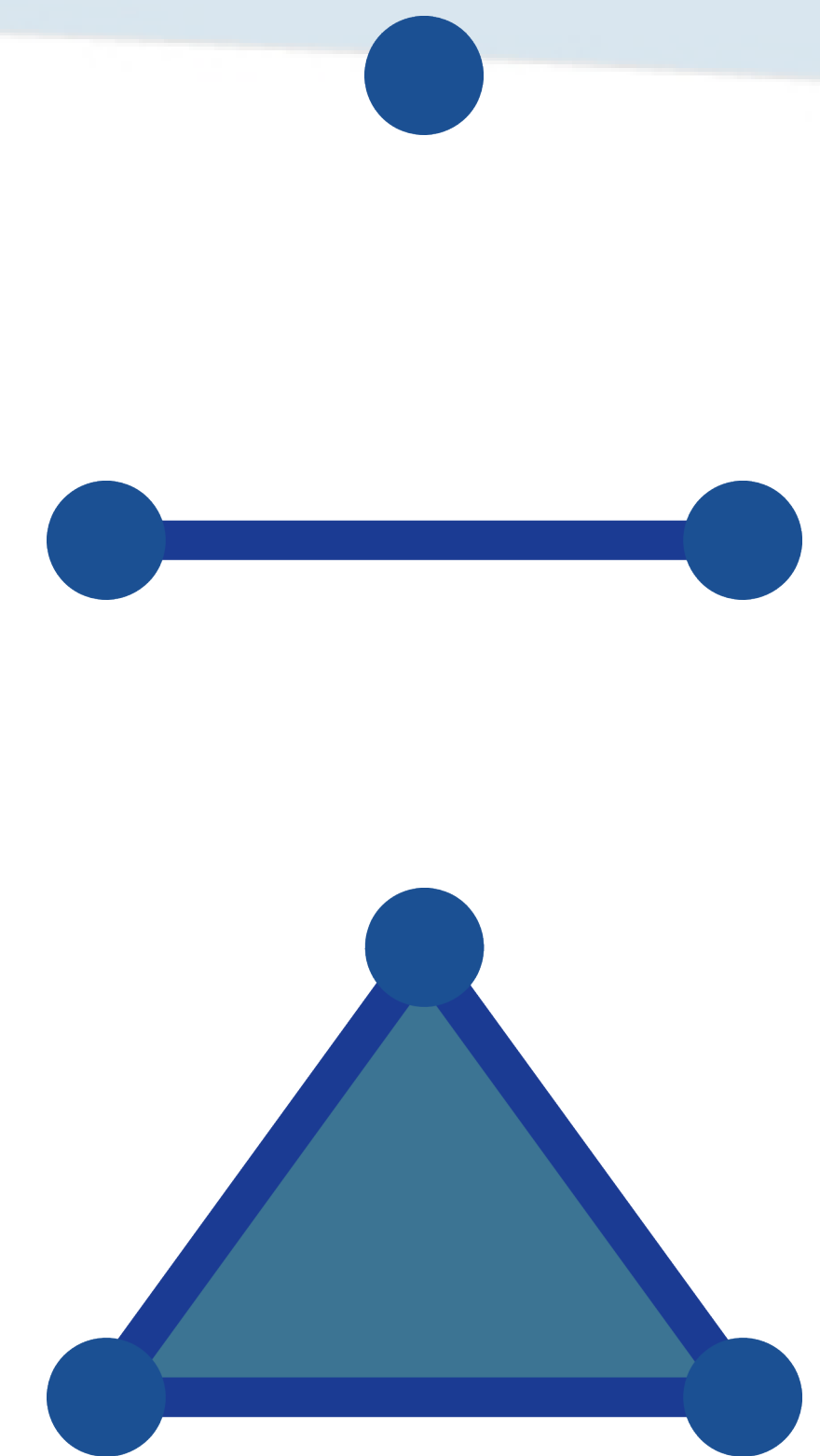
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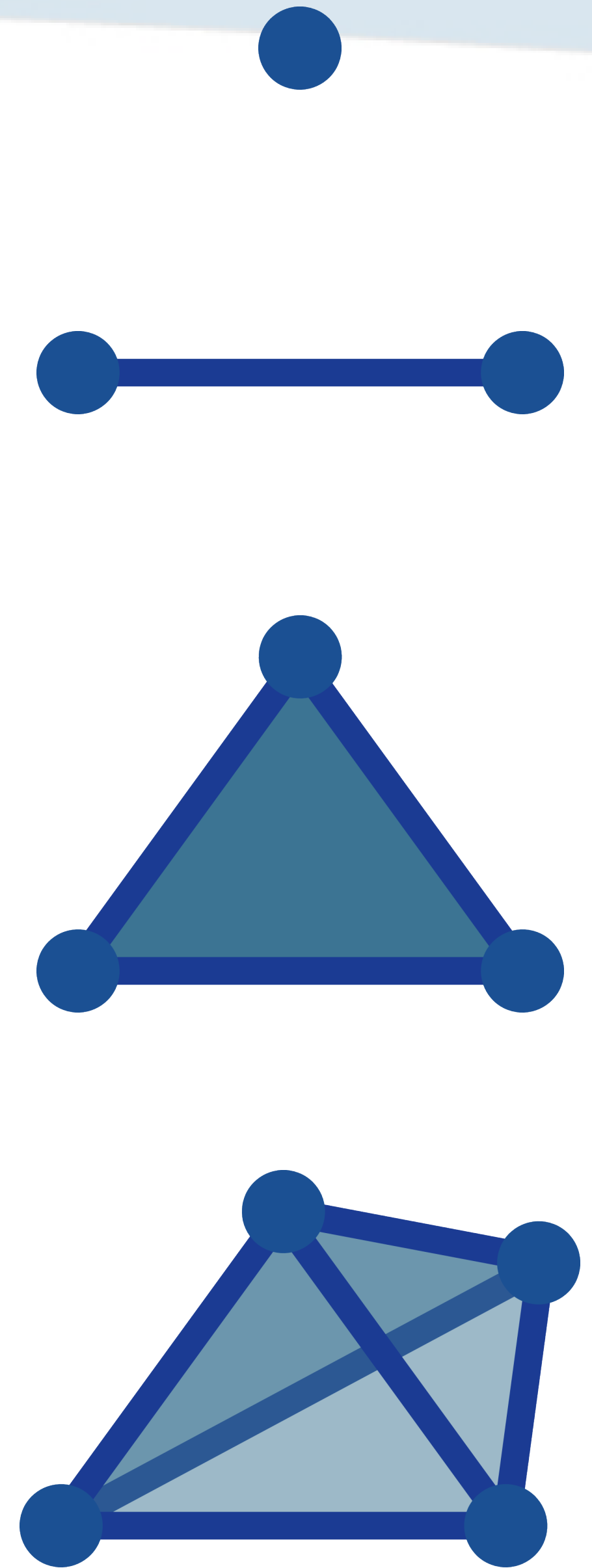
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 - 3-simplex: *tetrahedron*



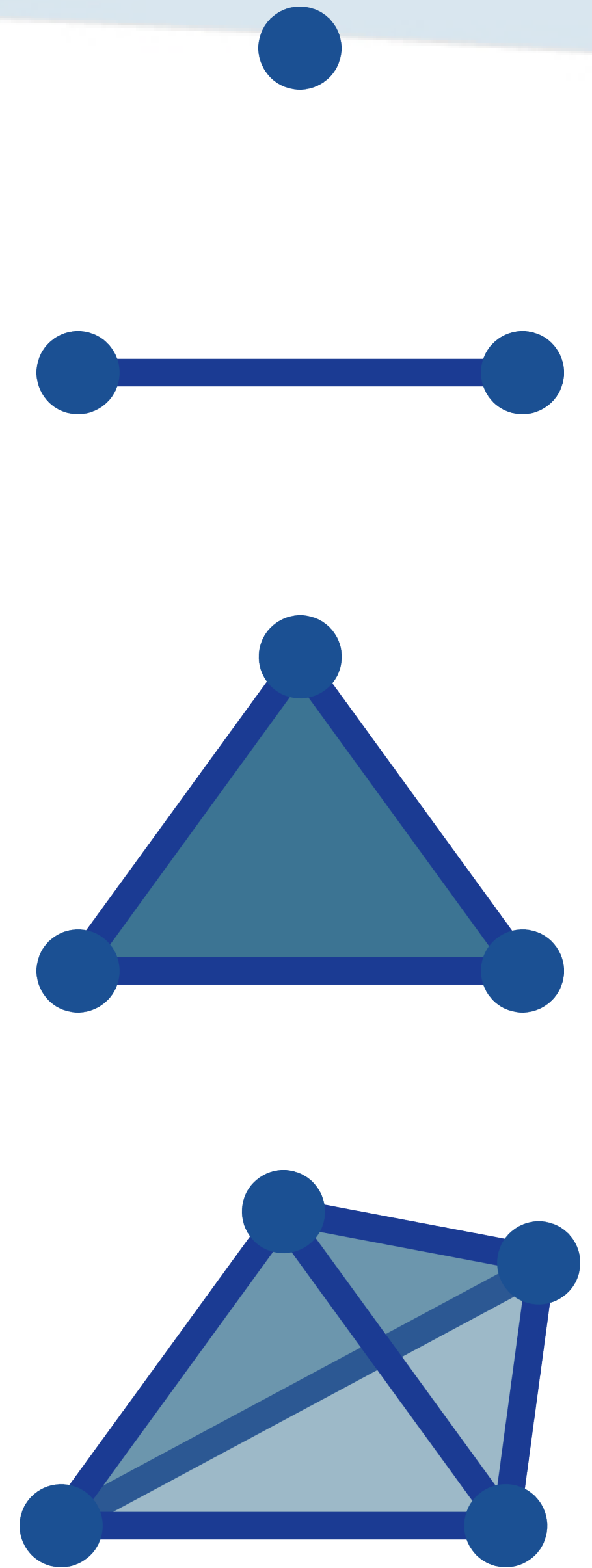
Manifolds on a computer

- Notion of face



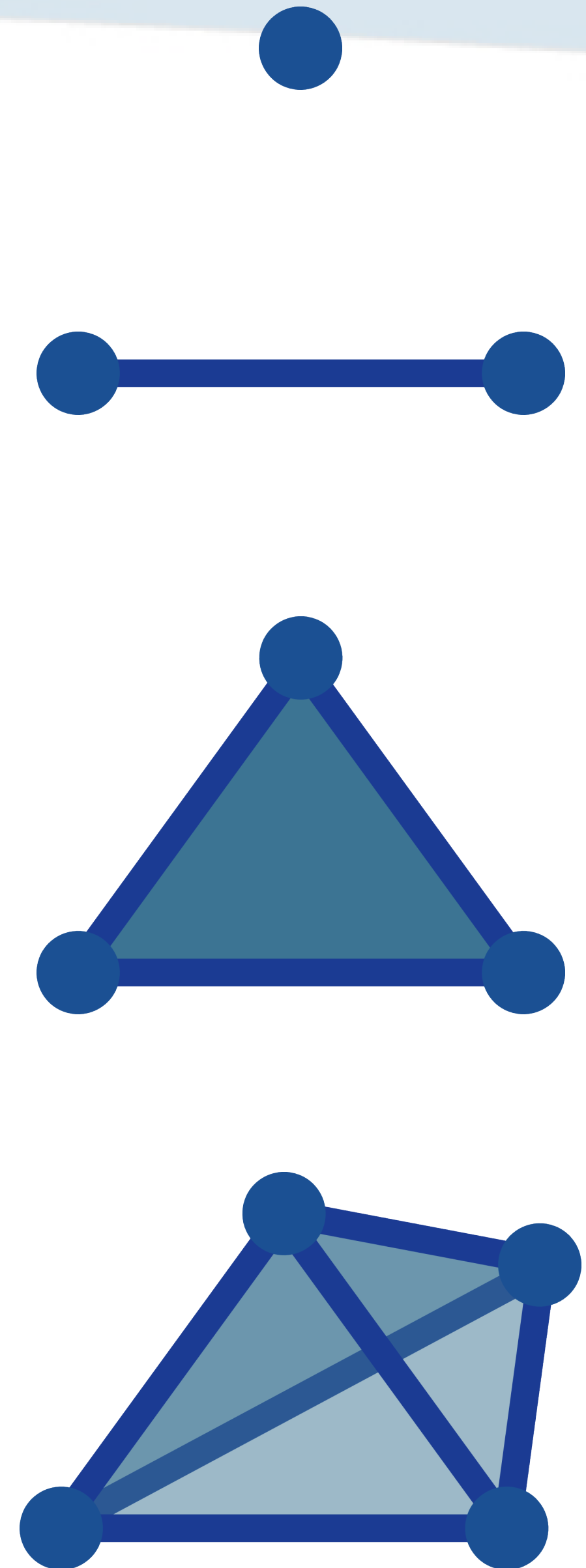
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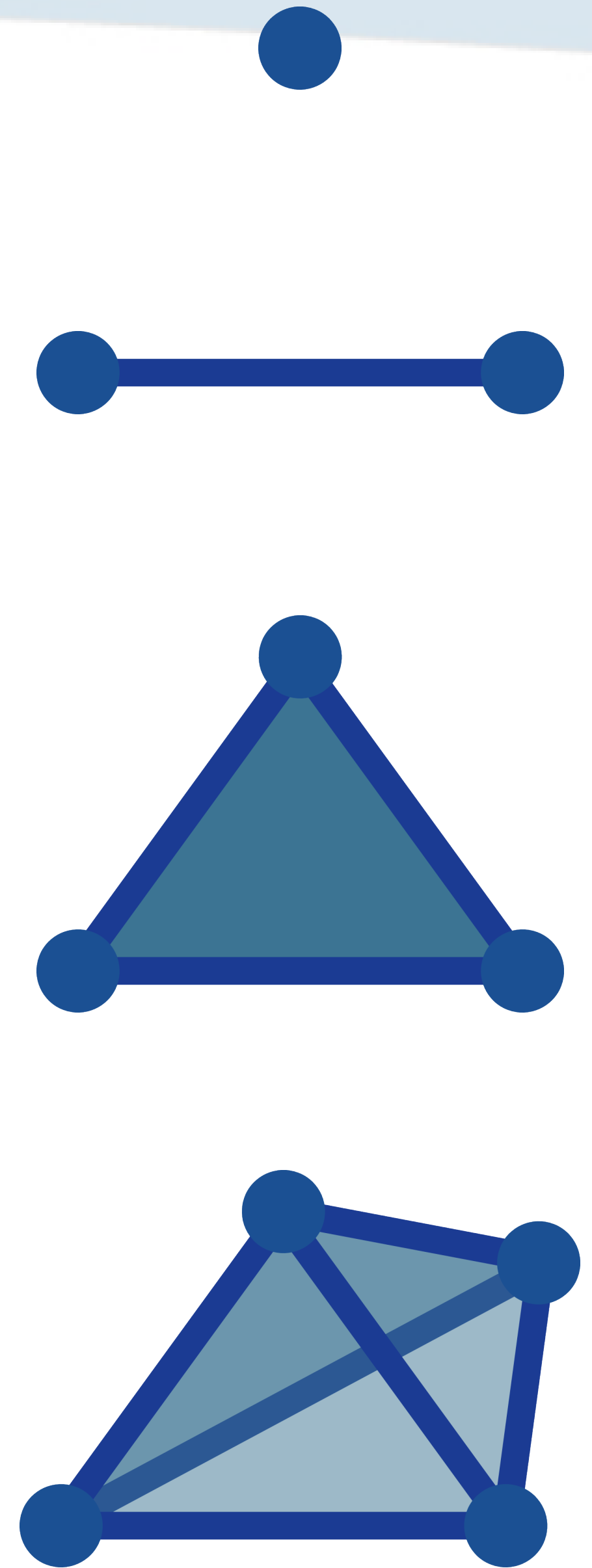
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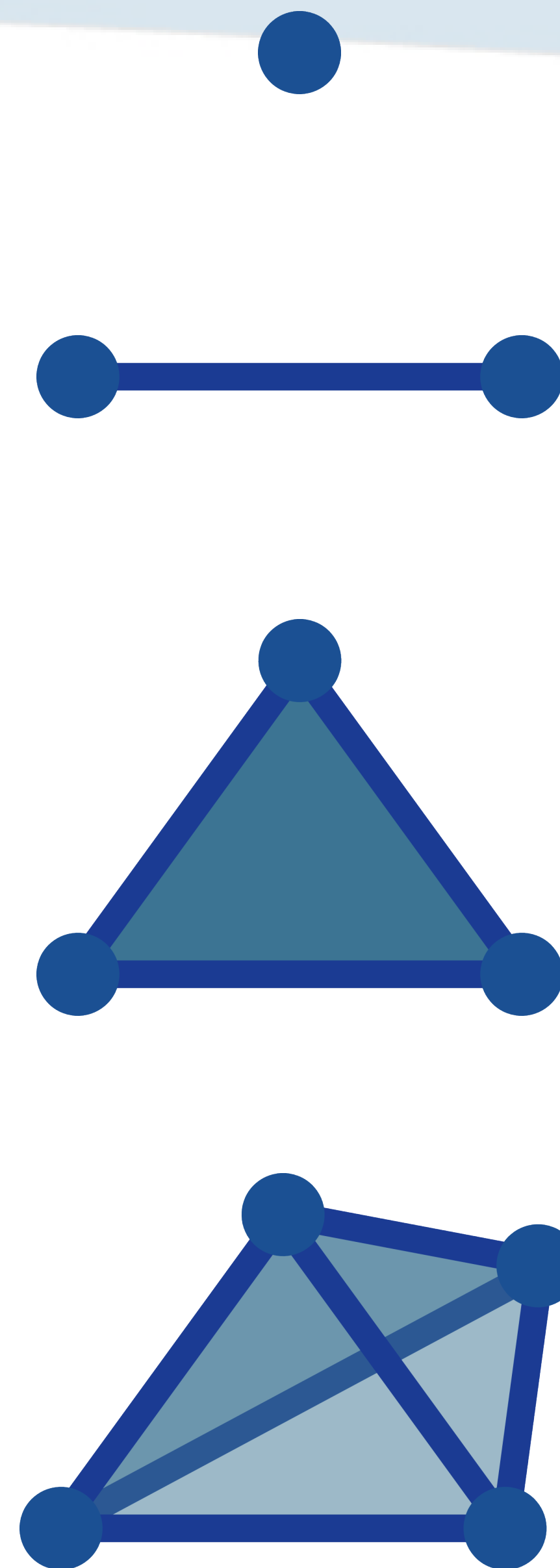
Manifolds on a computer

- Notion of **simplicial complex**



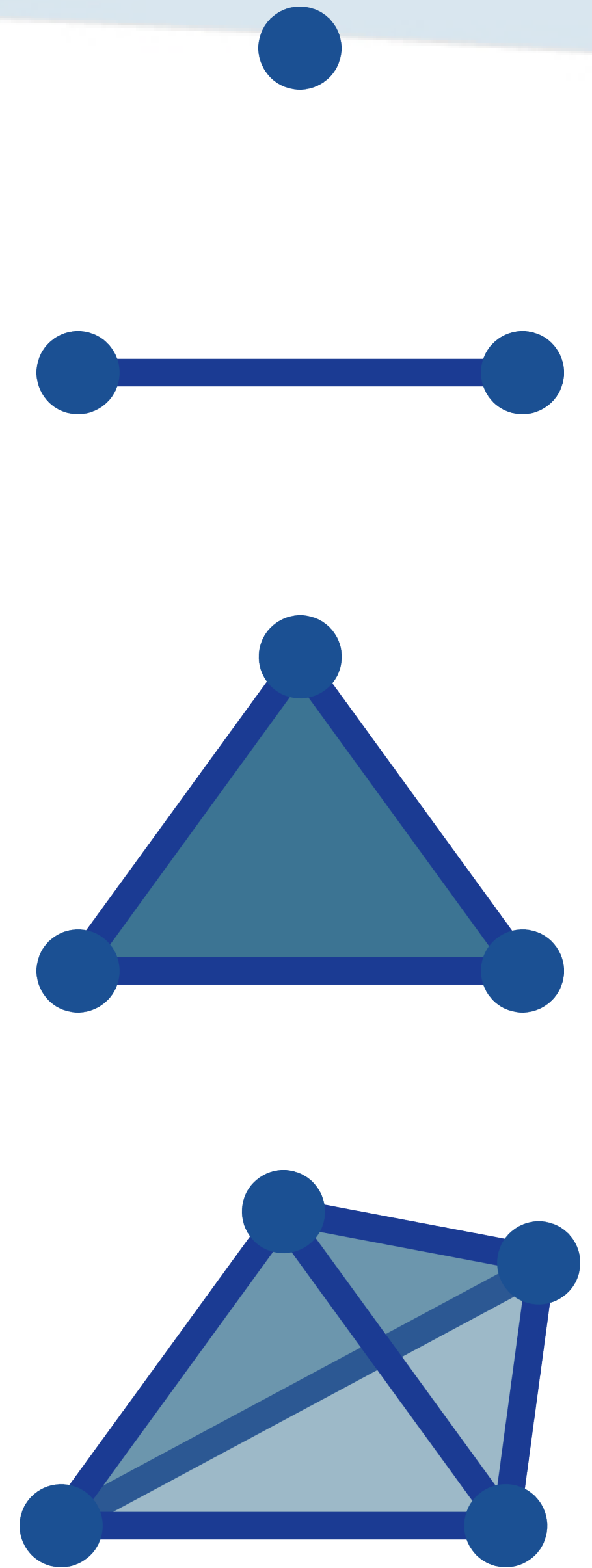
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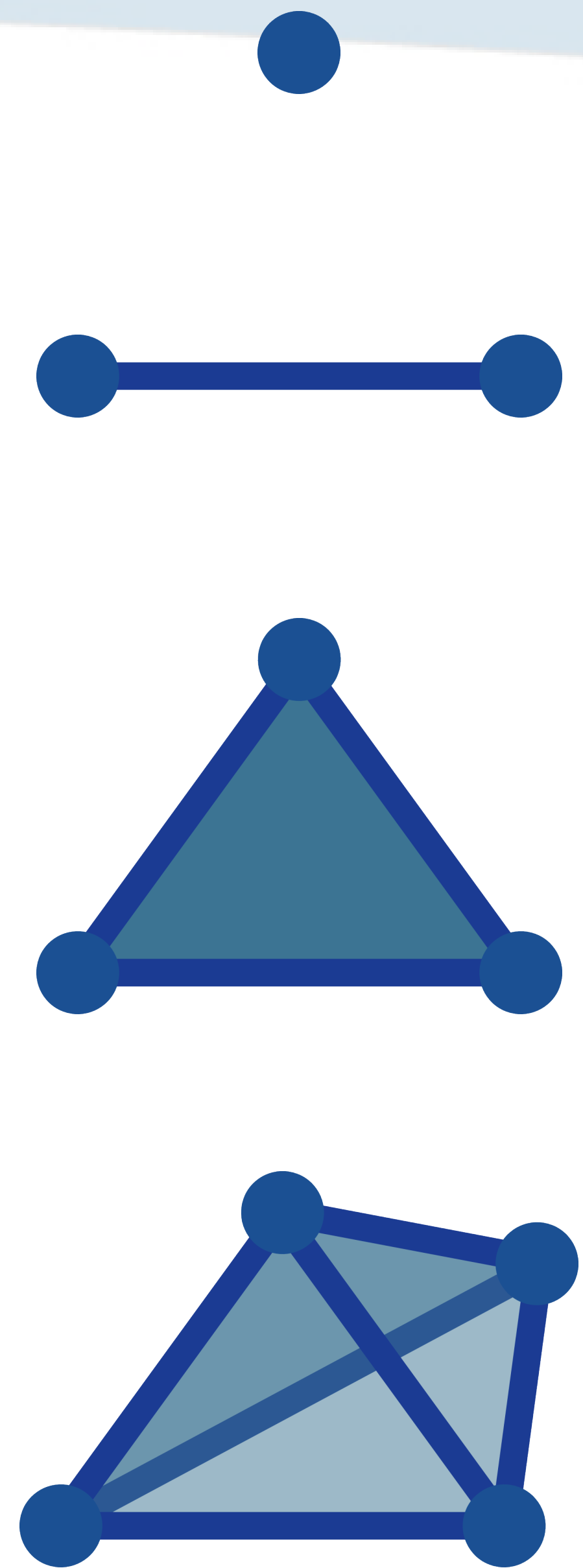
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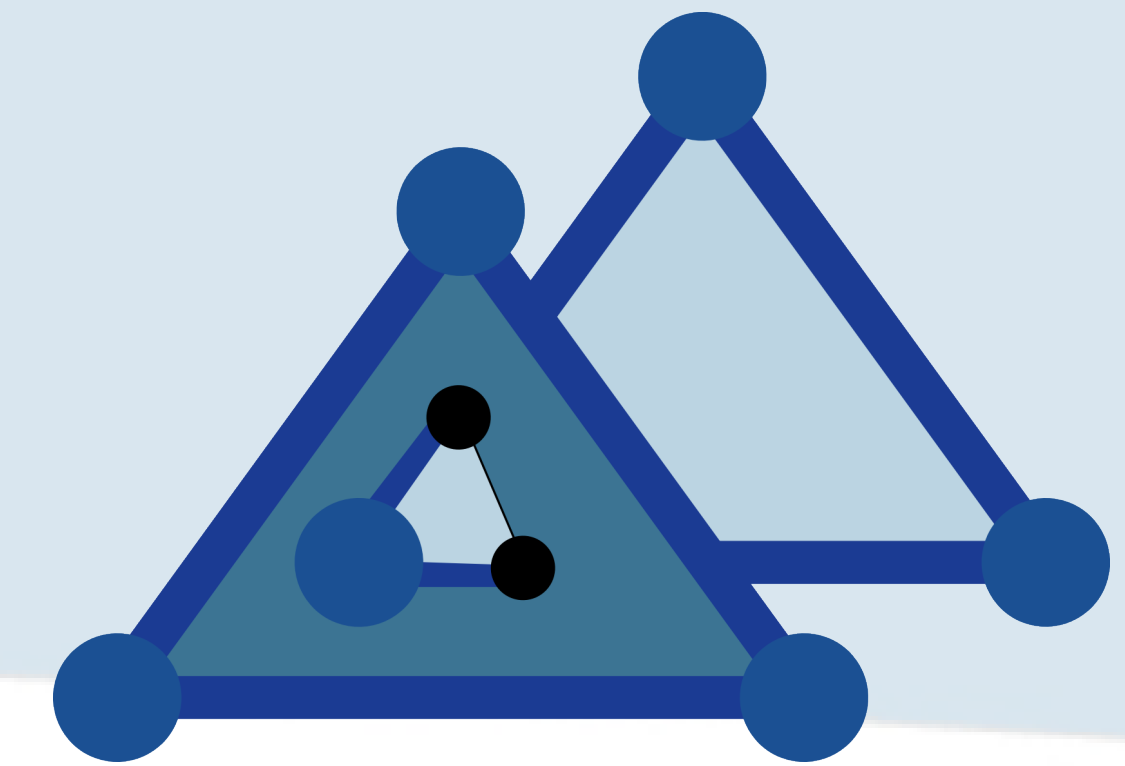


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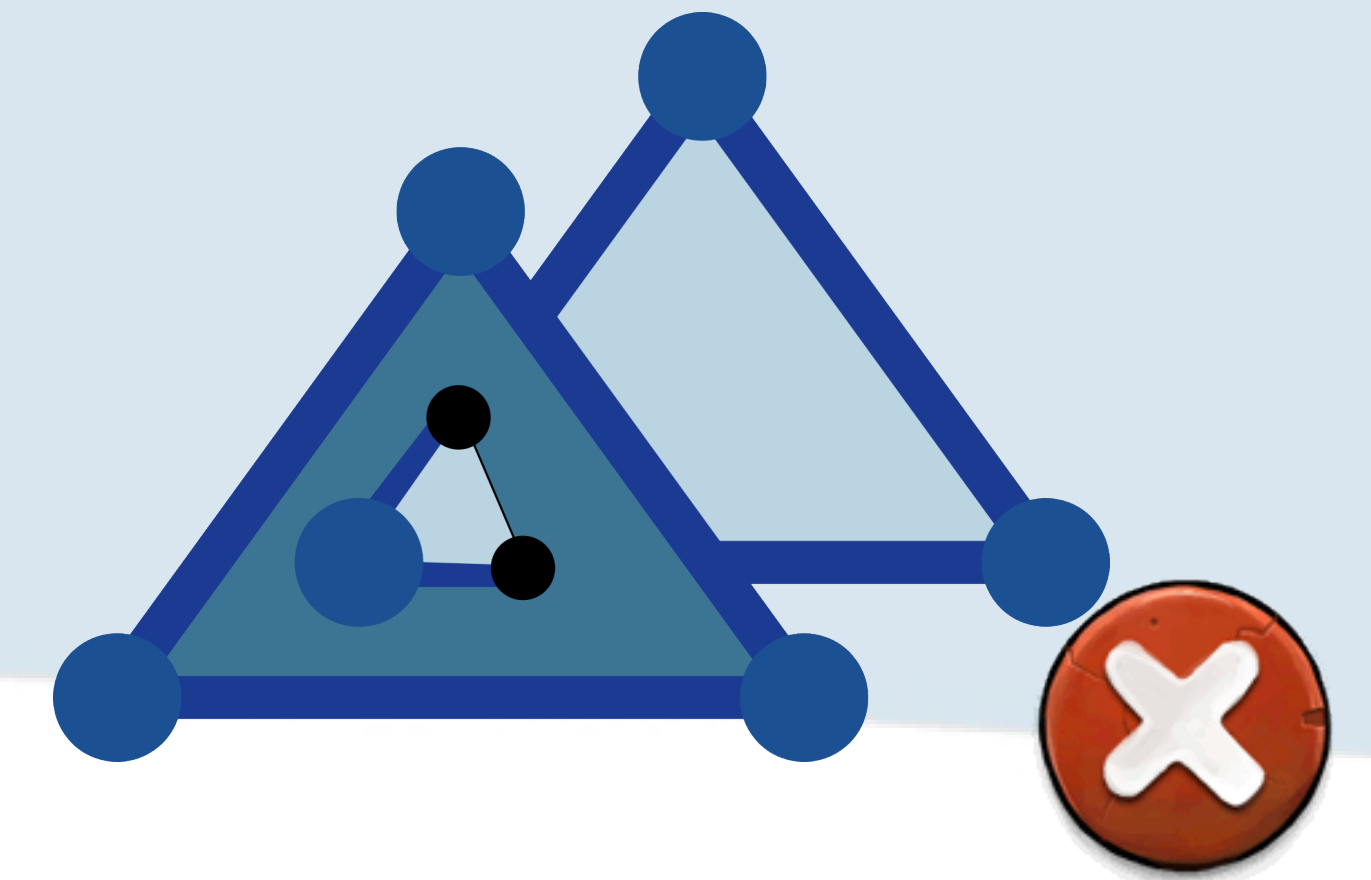


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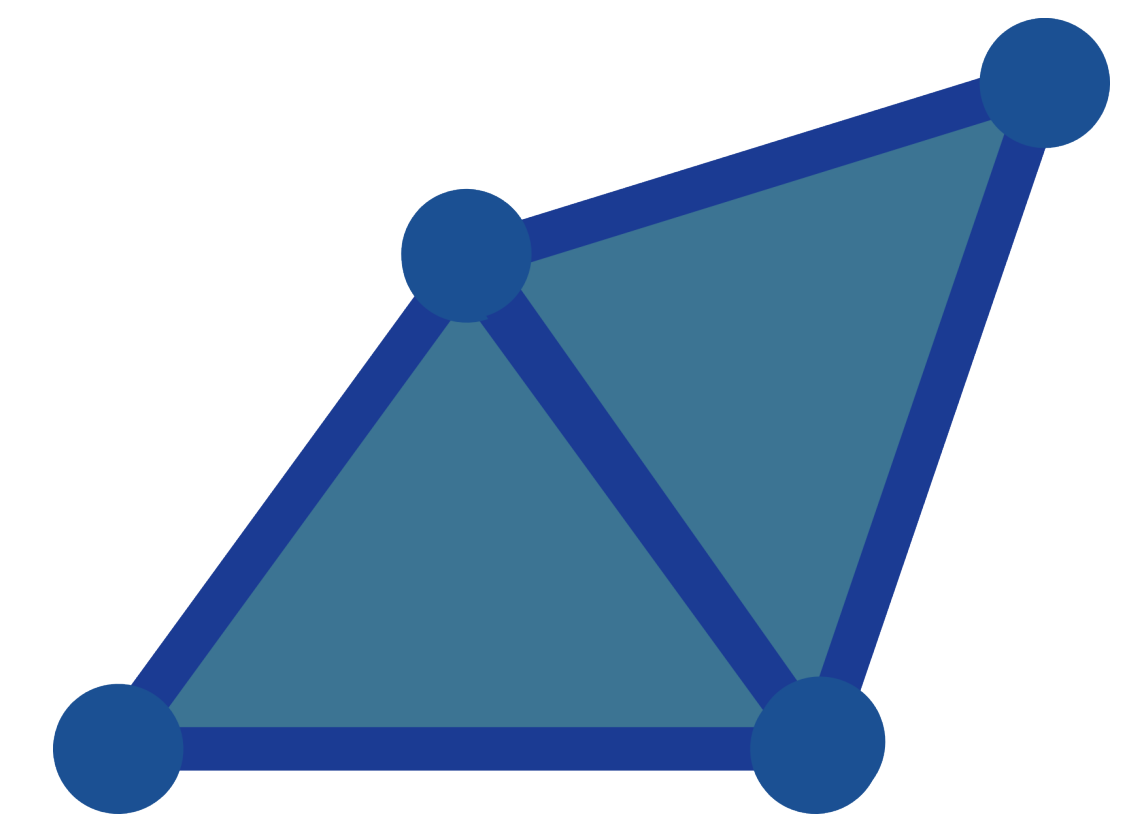
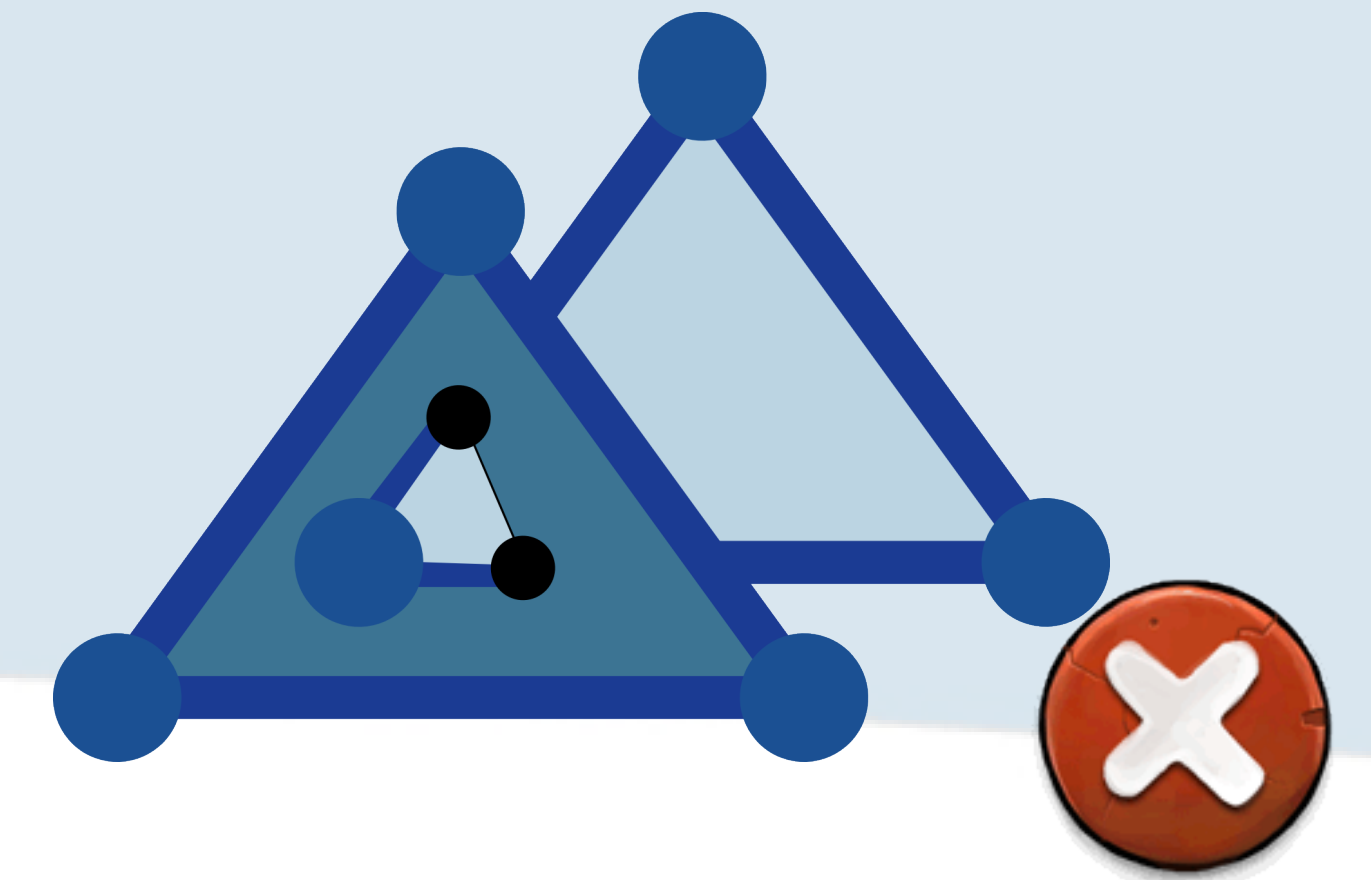
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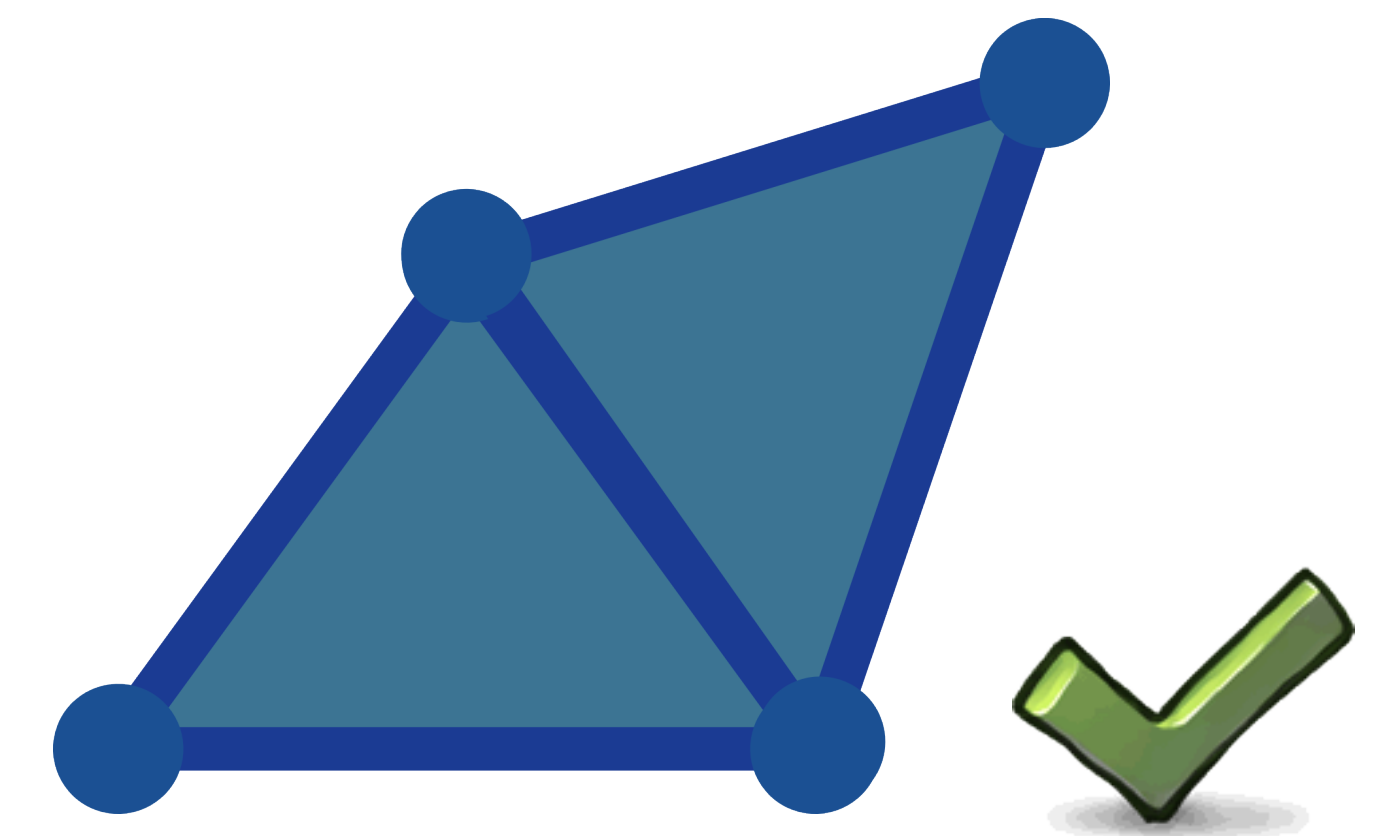
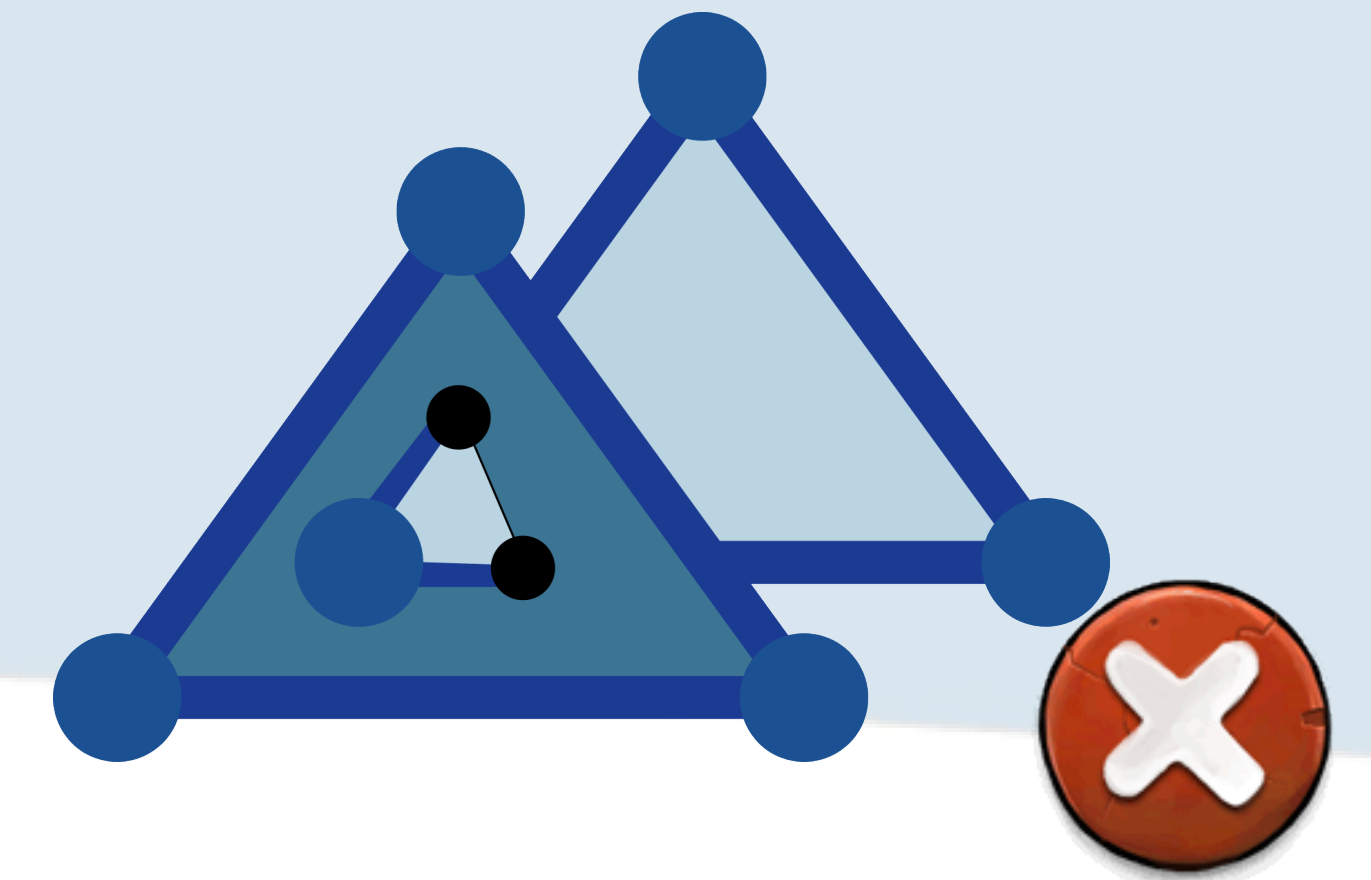
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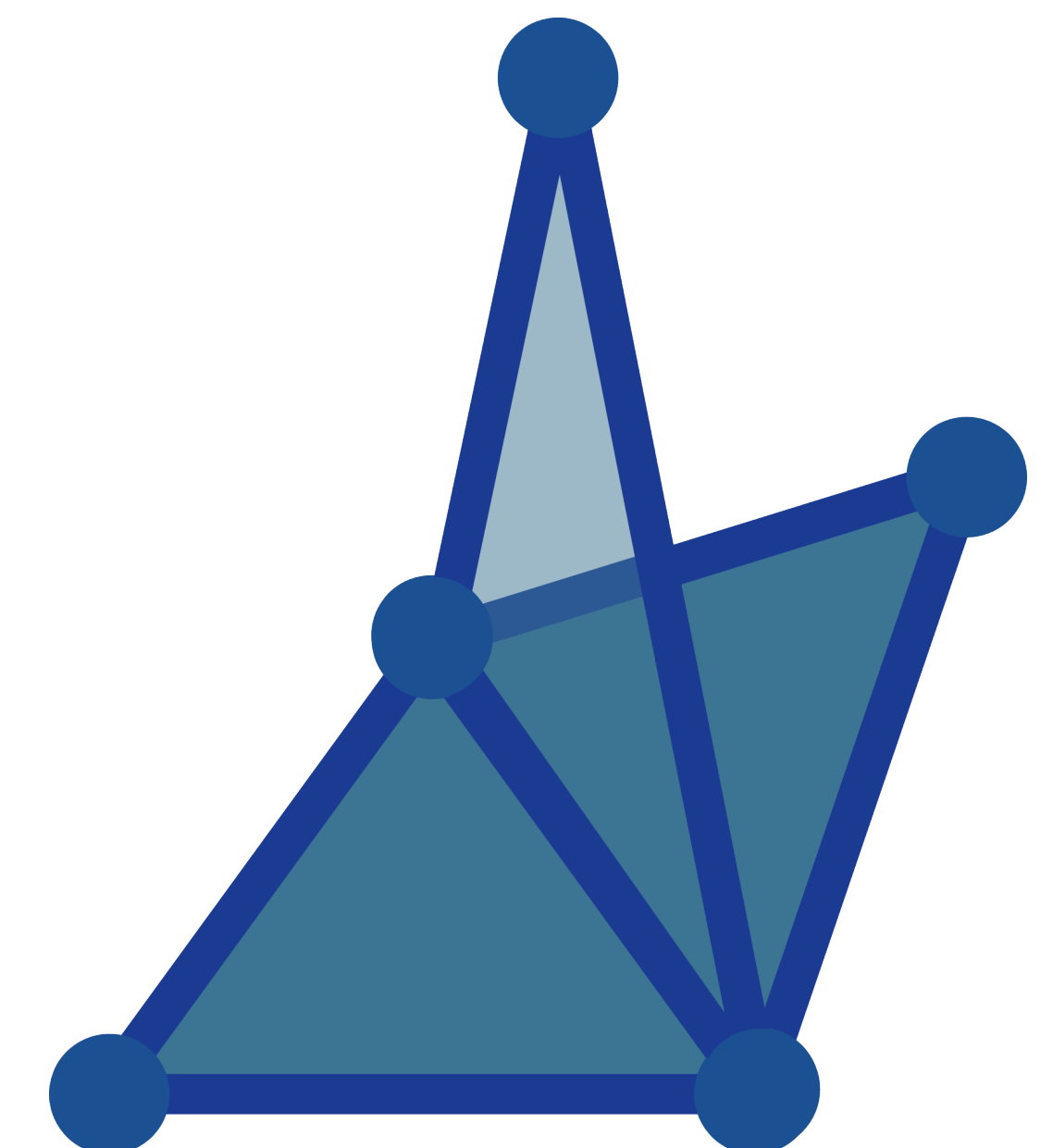
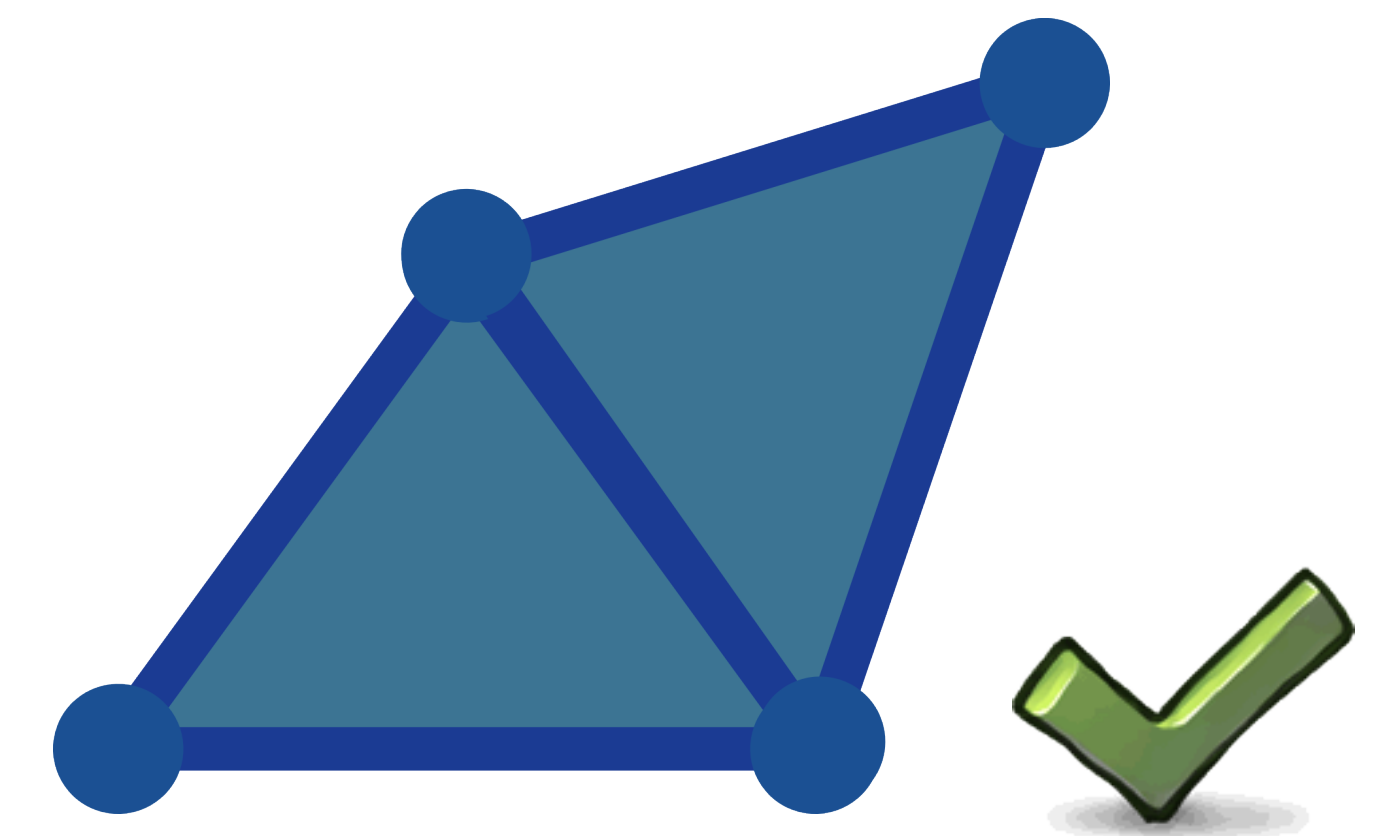
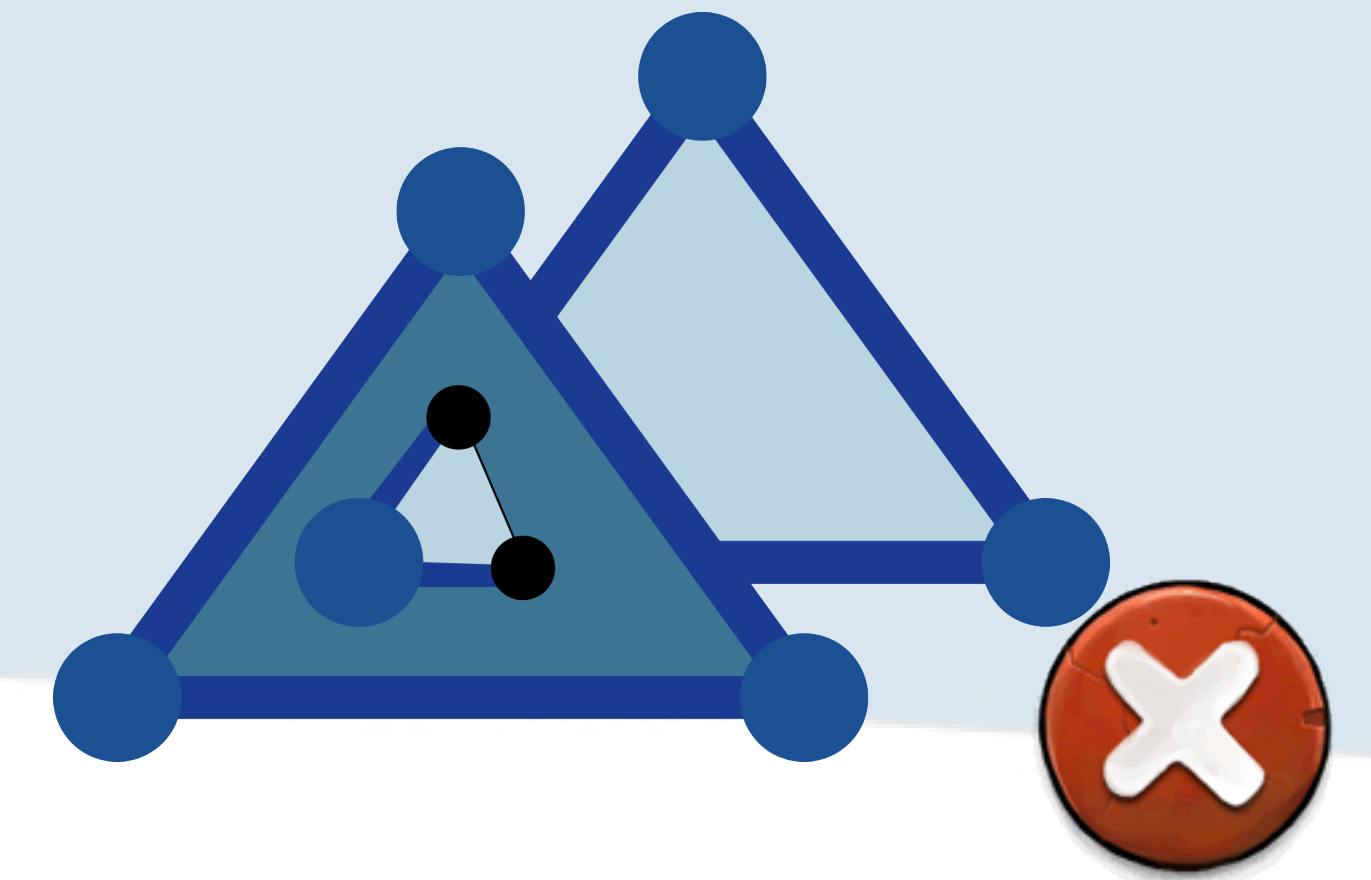
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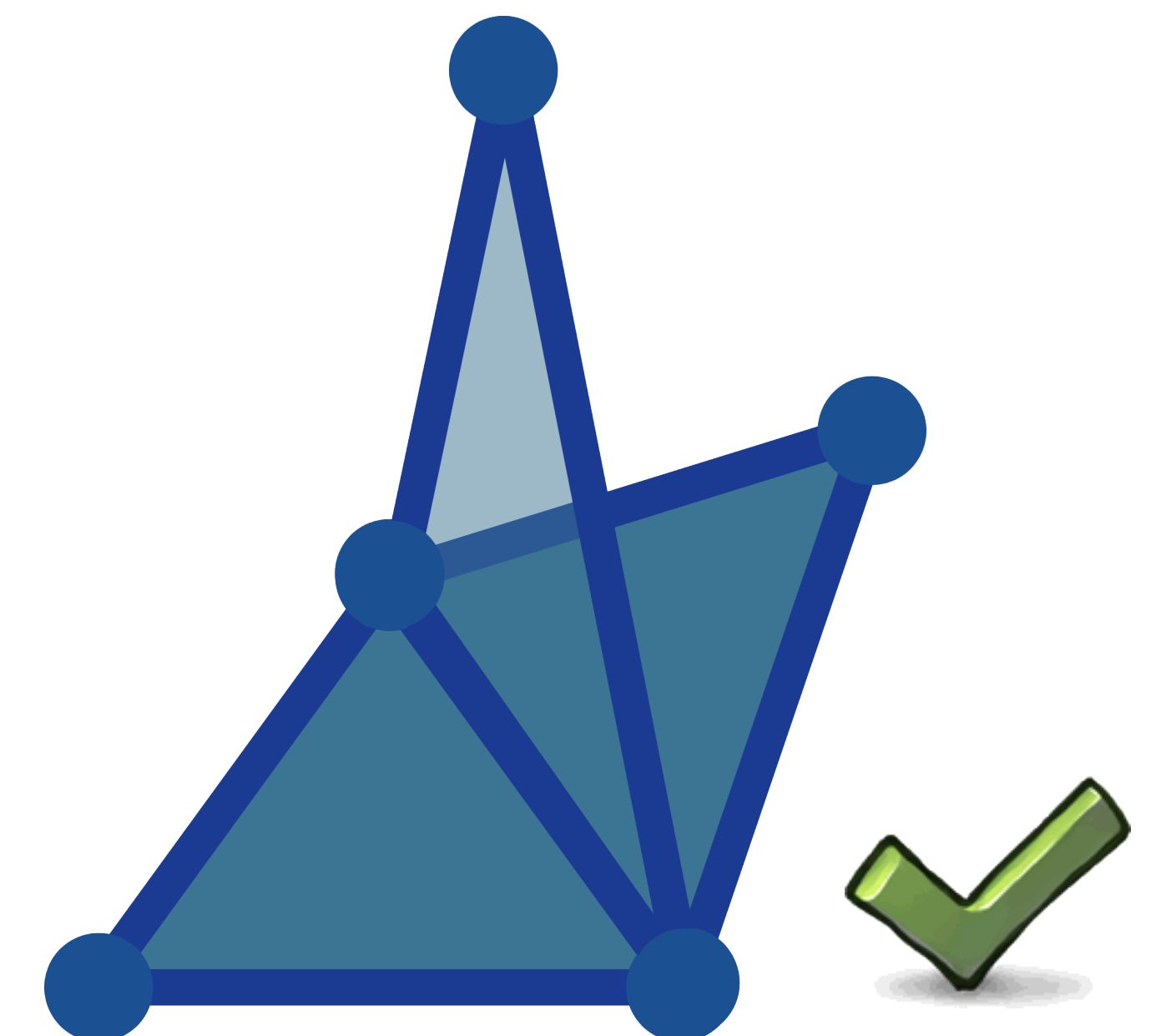
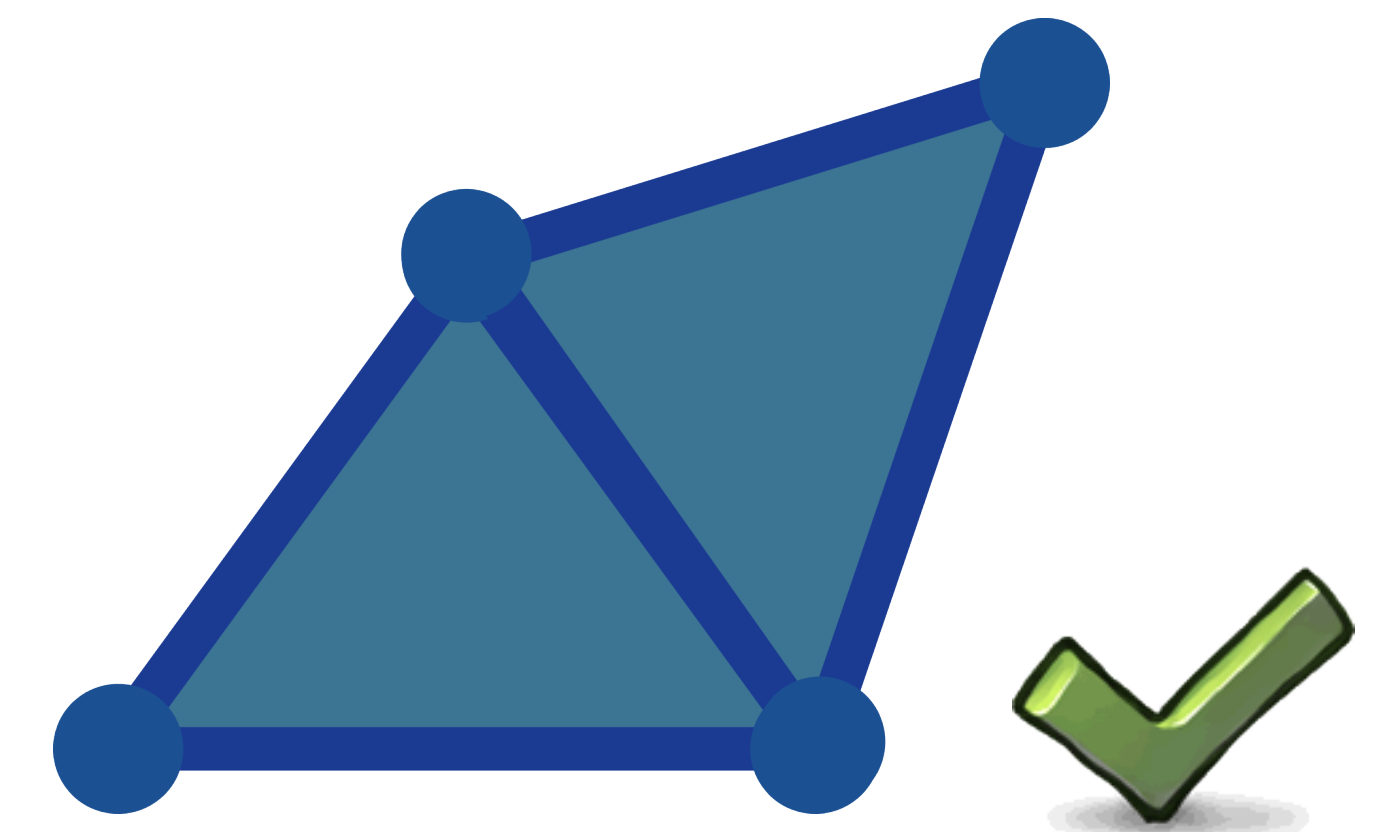
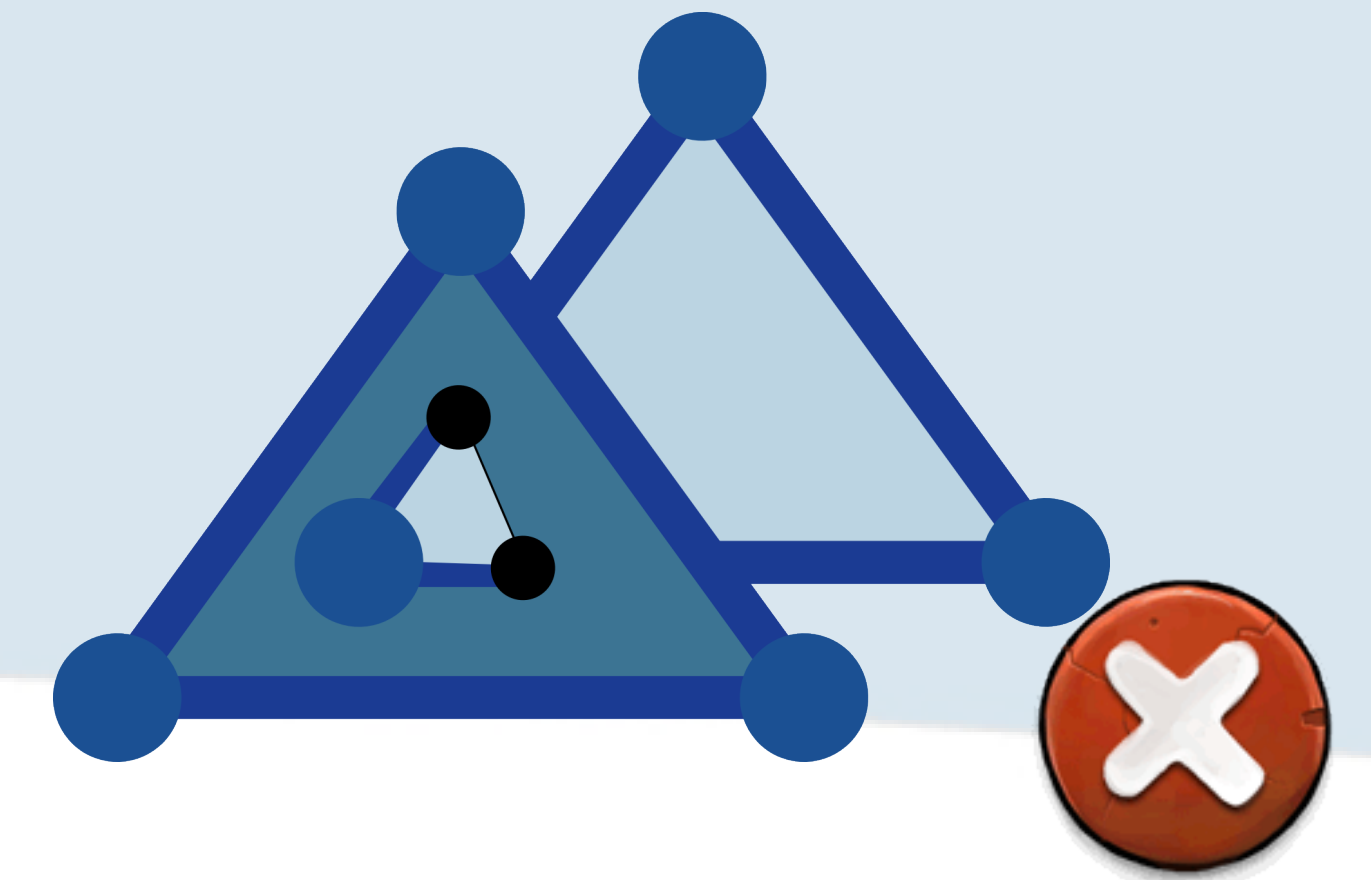
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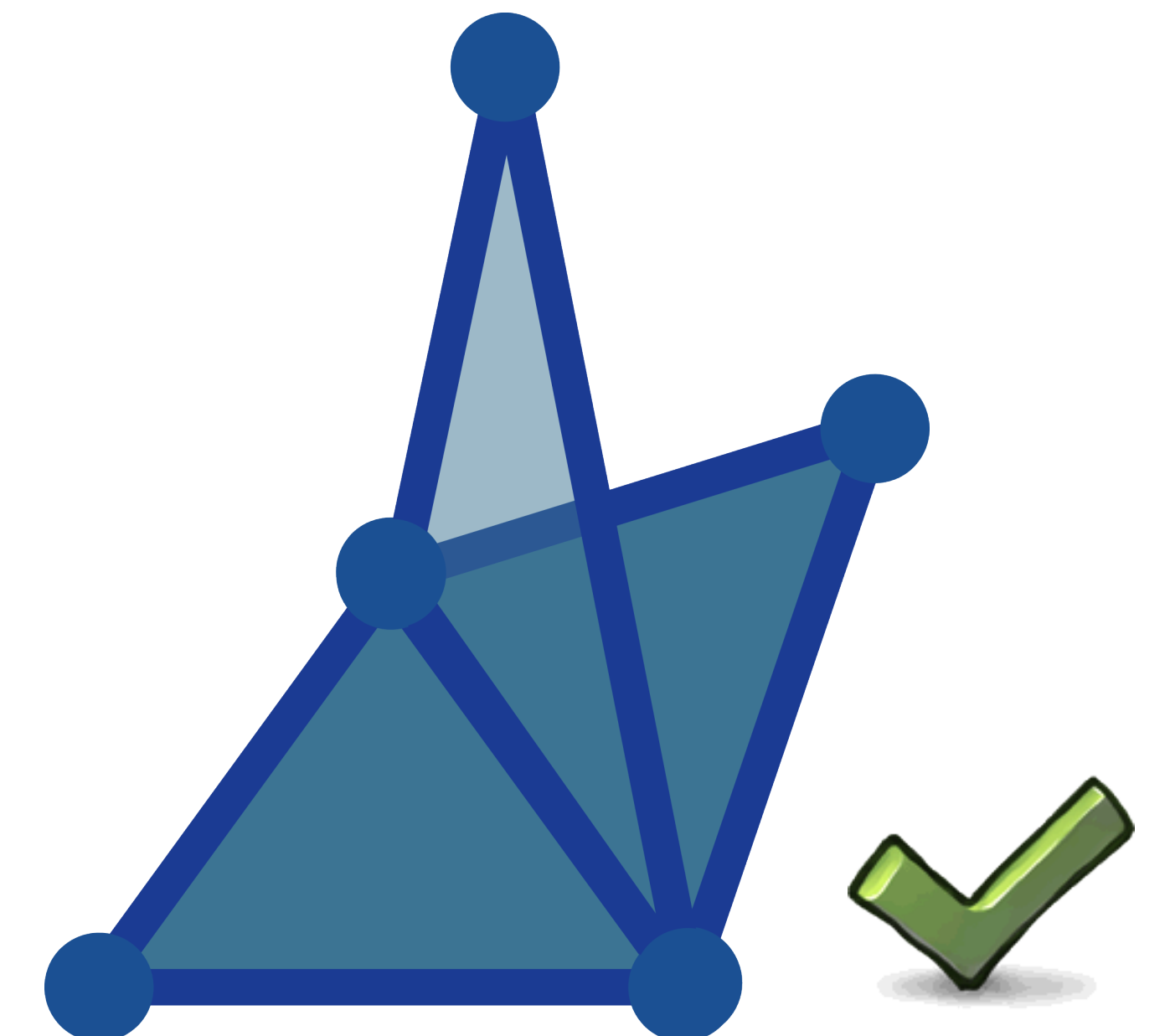
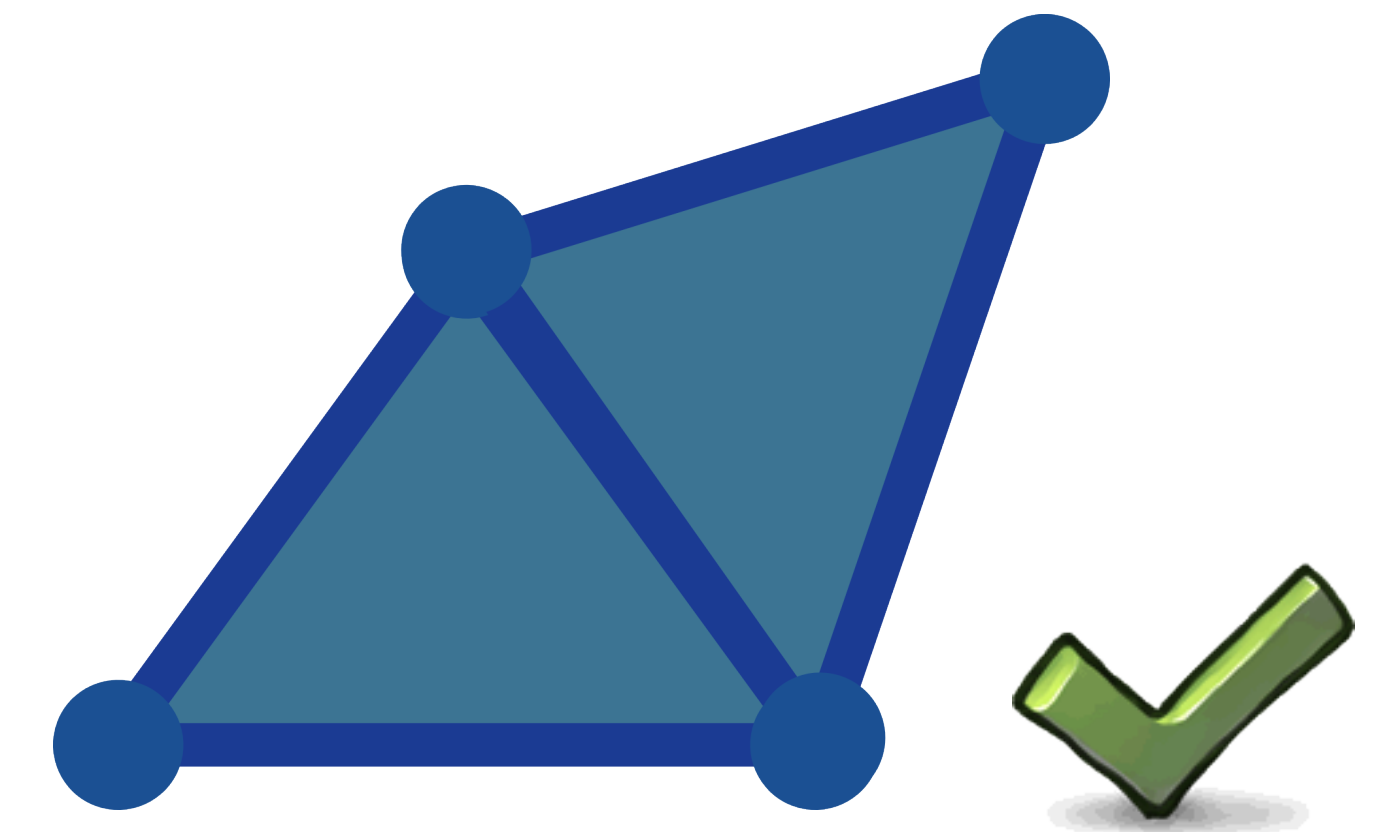
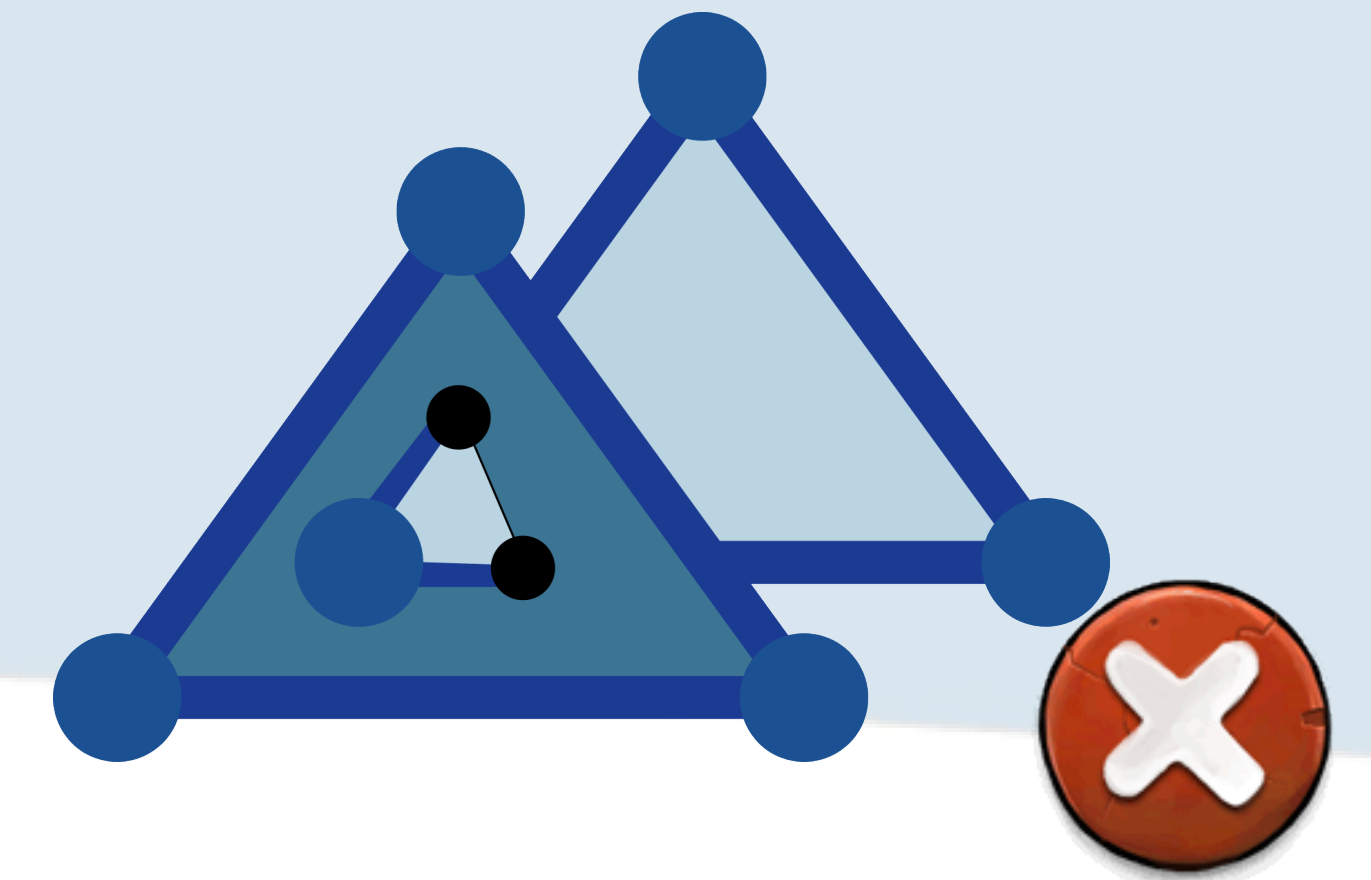
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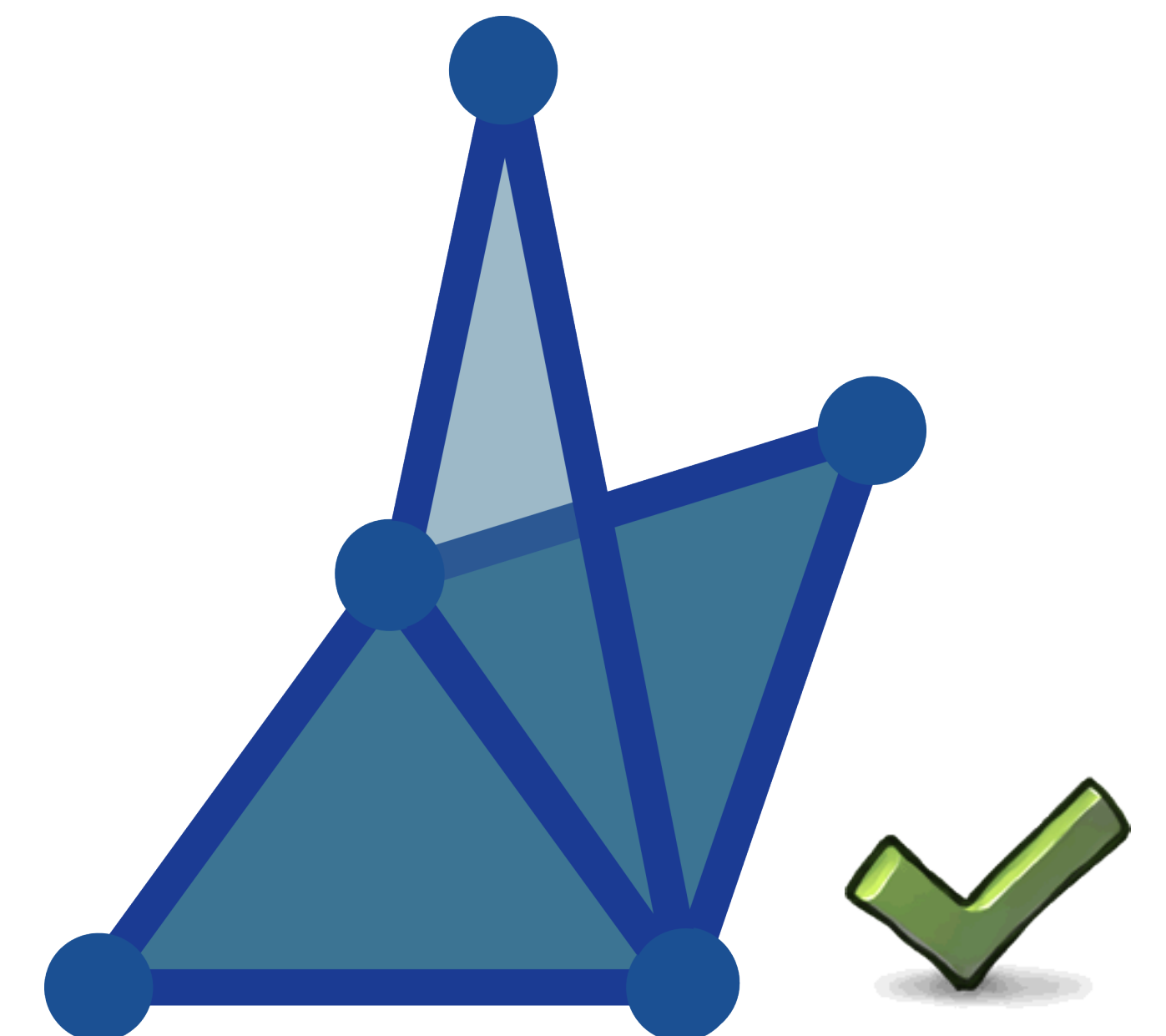
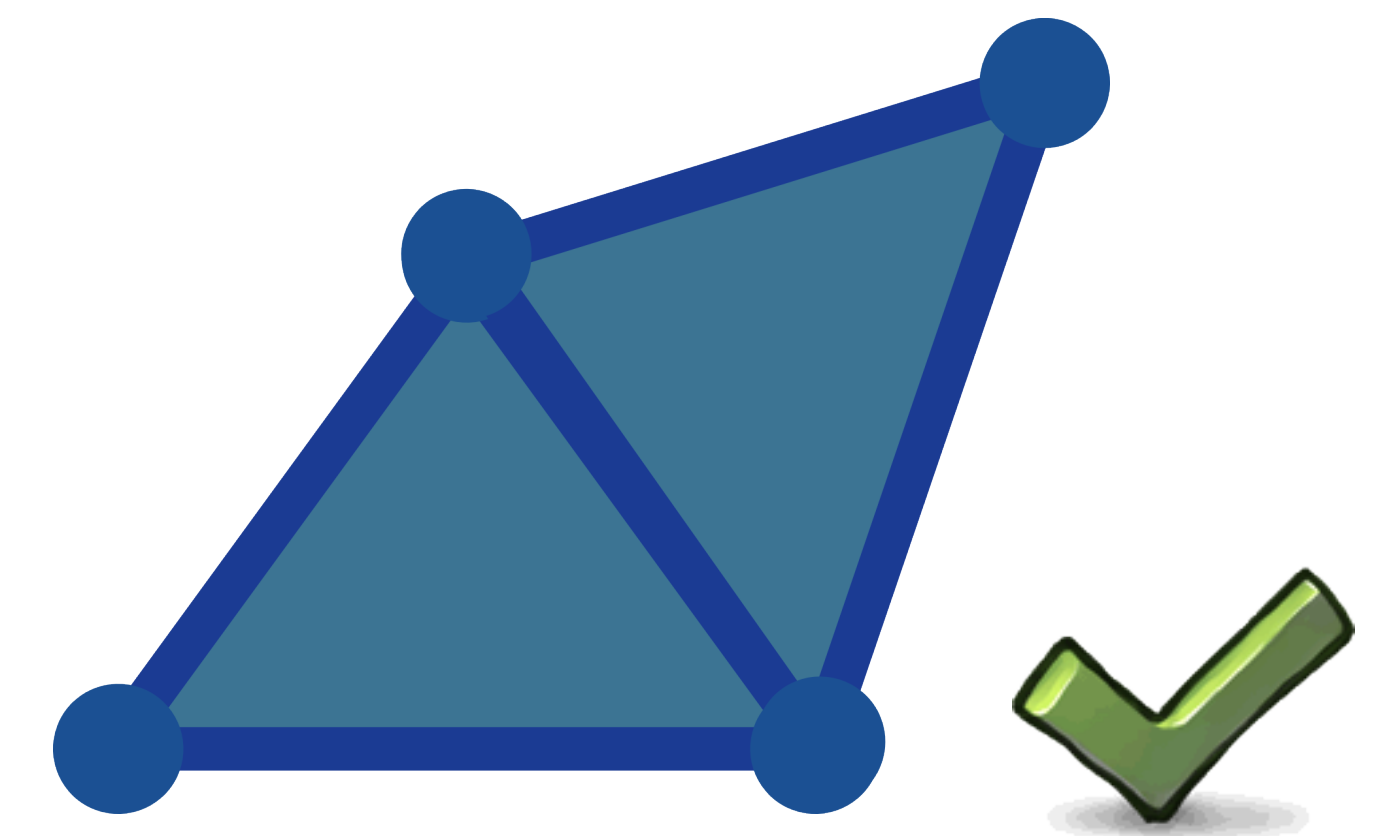
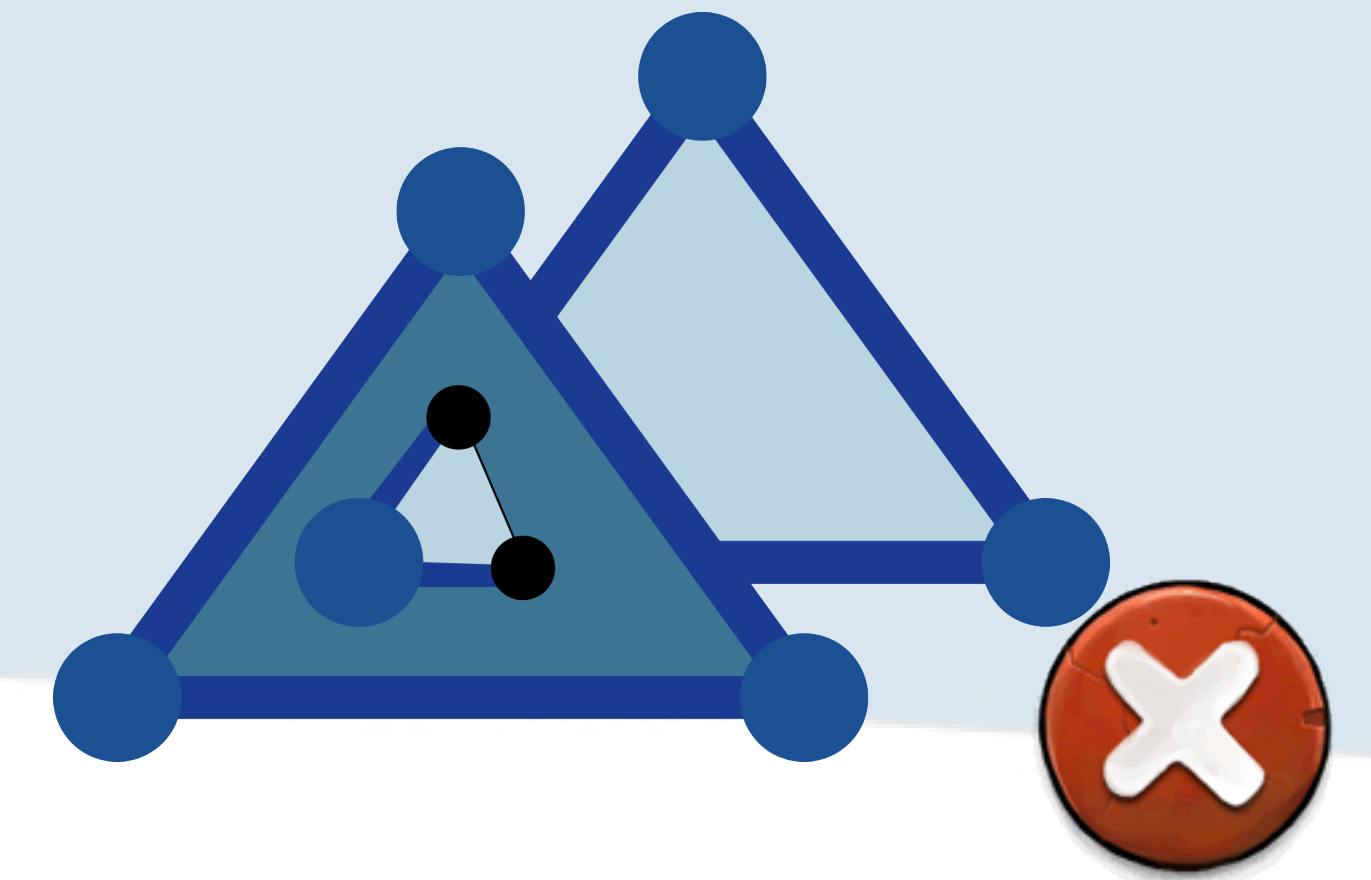
Manifolds on a computer

- Notion of **triangulation**



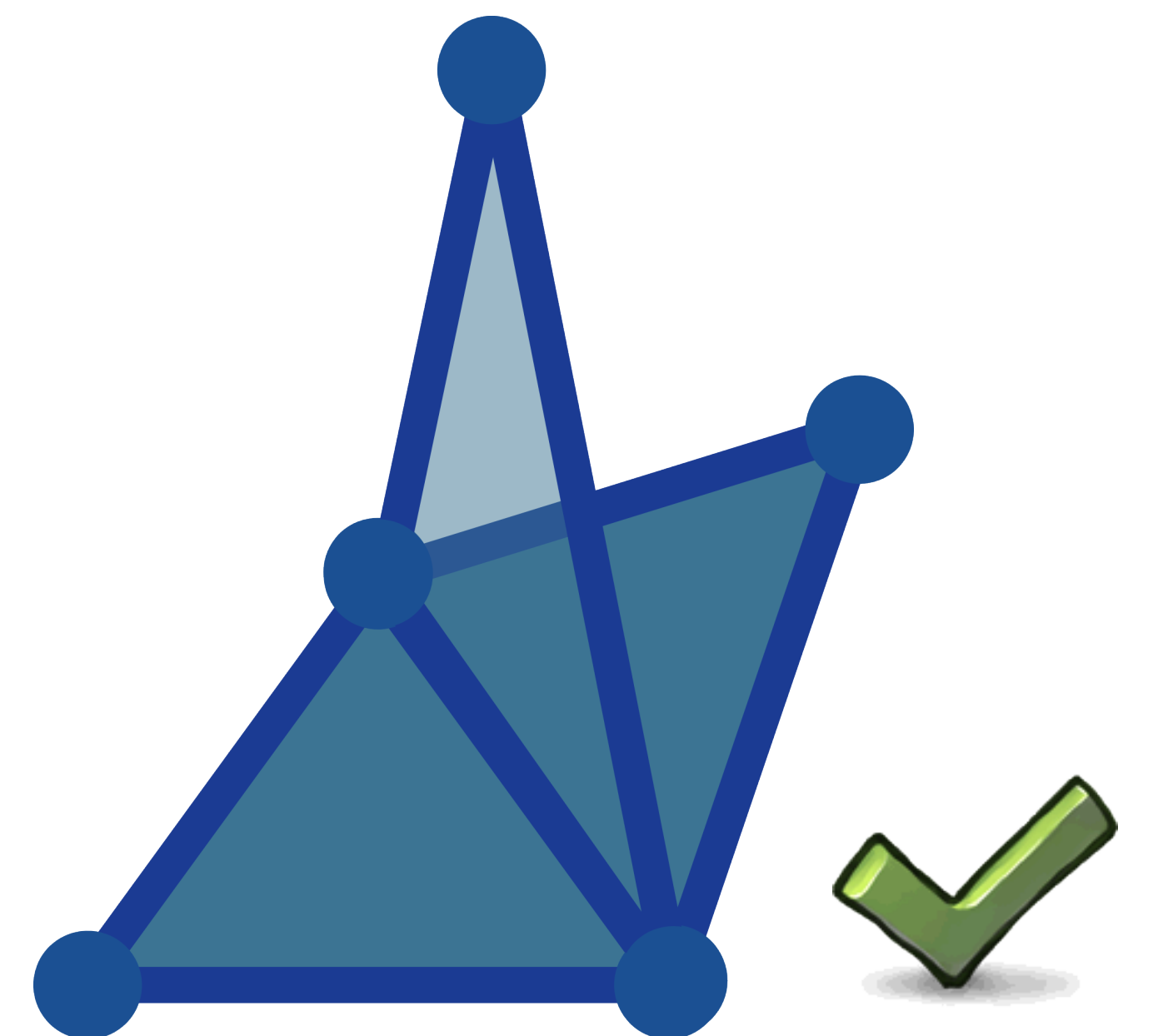
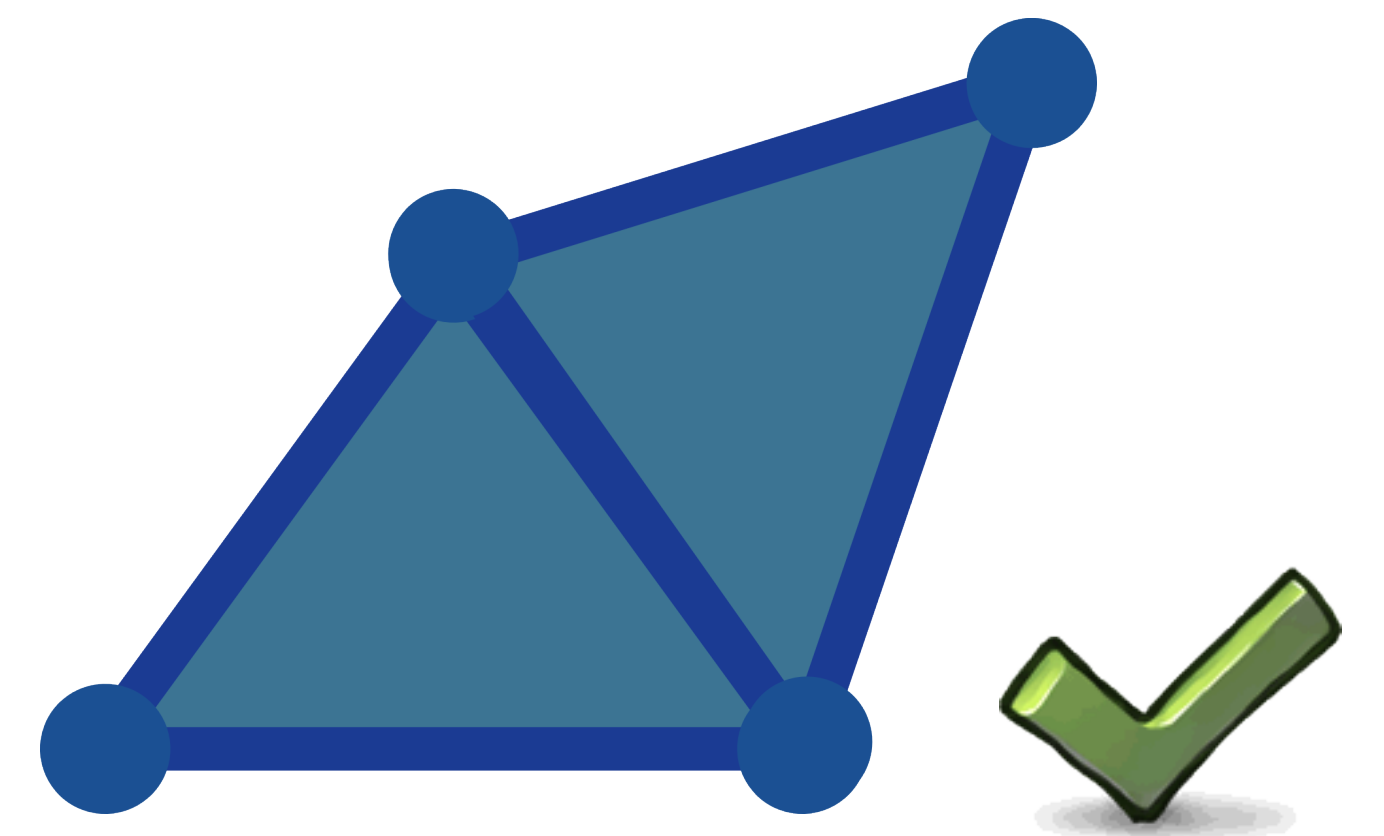
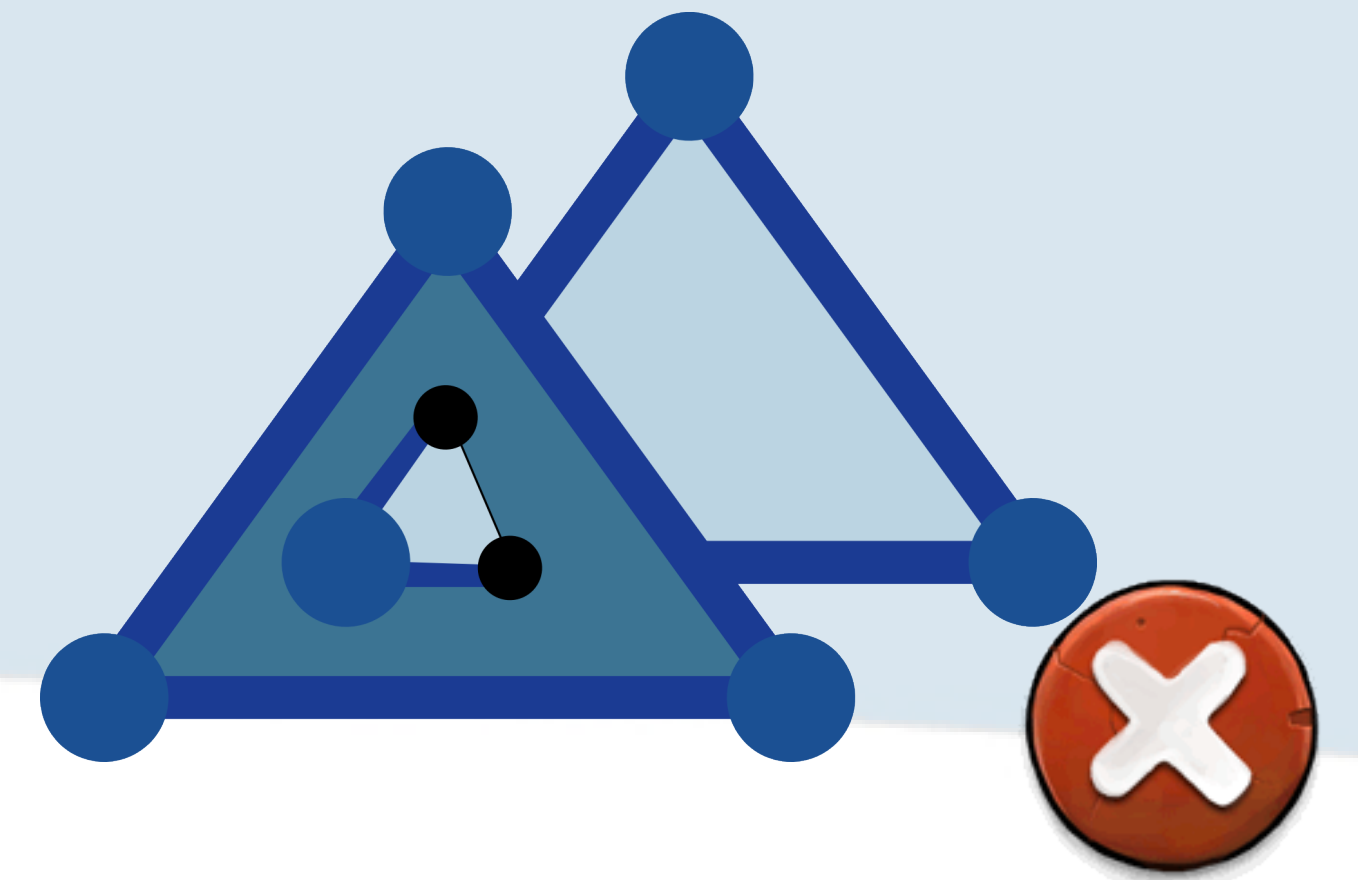
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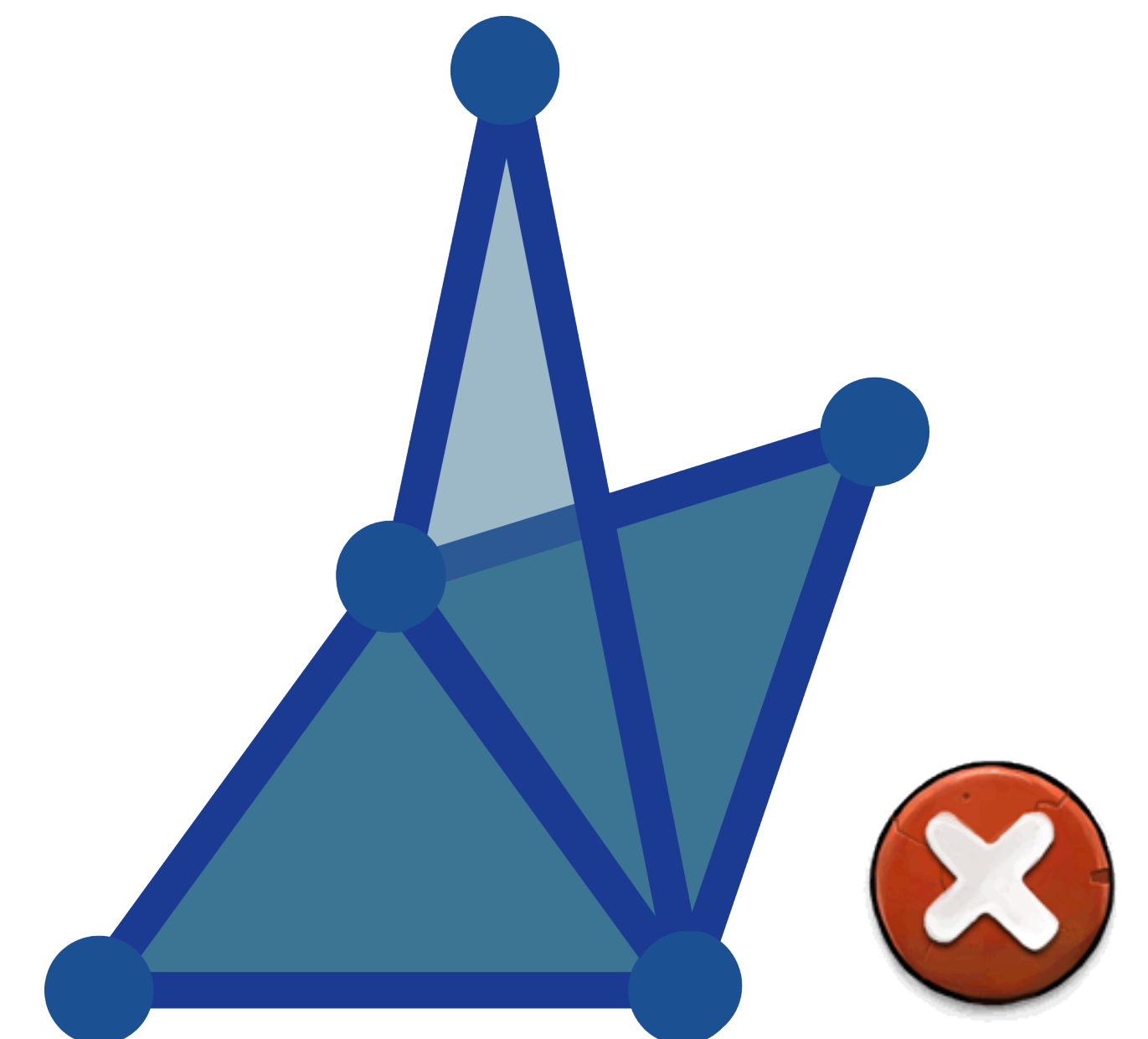
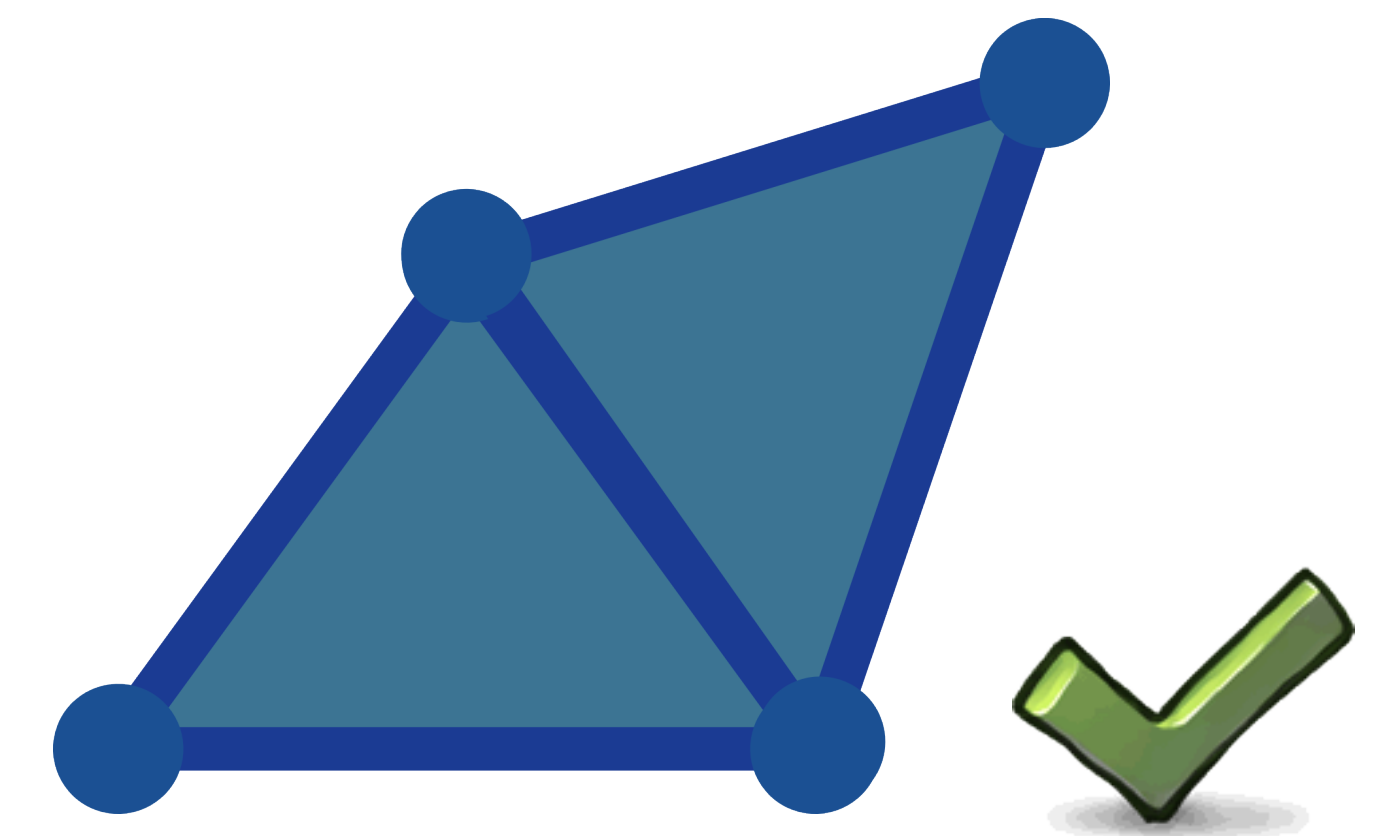
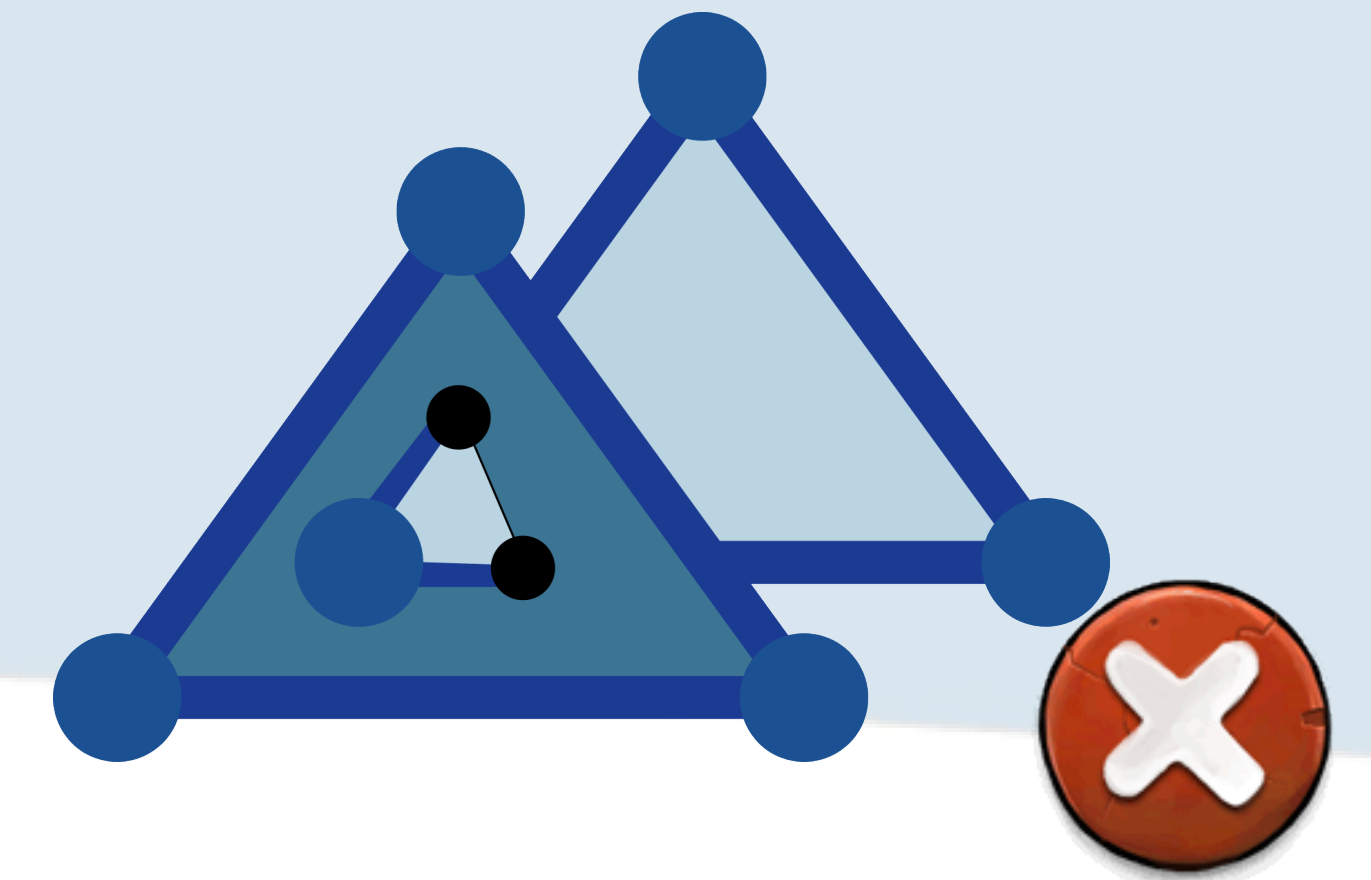
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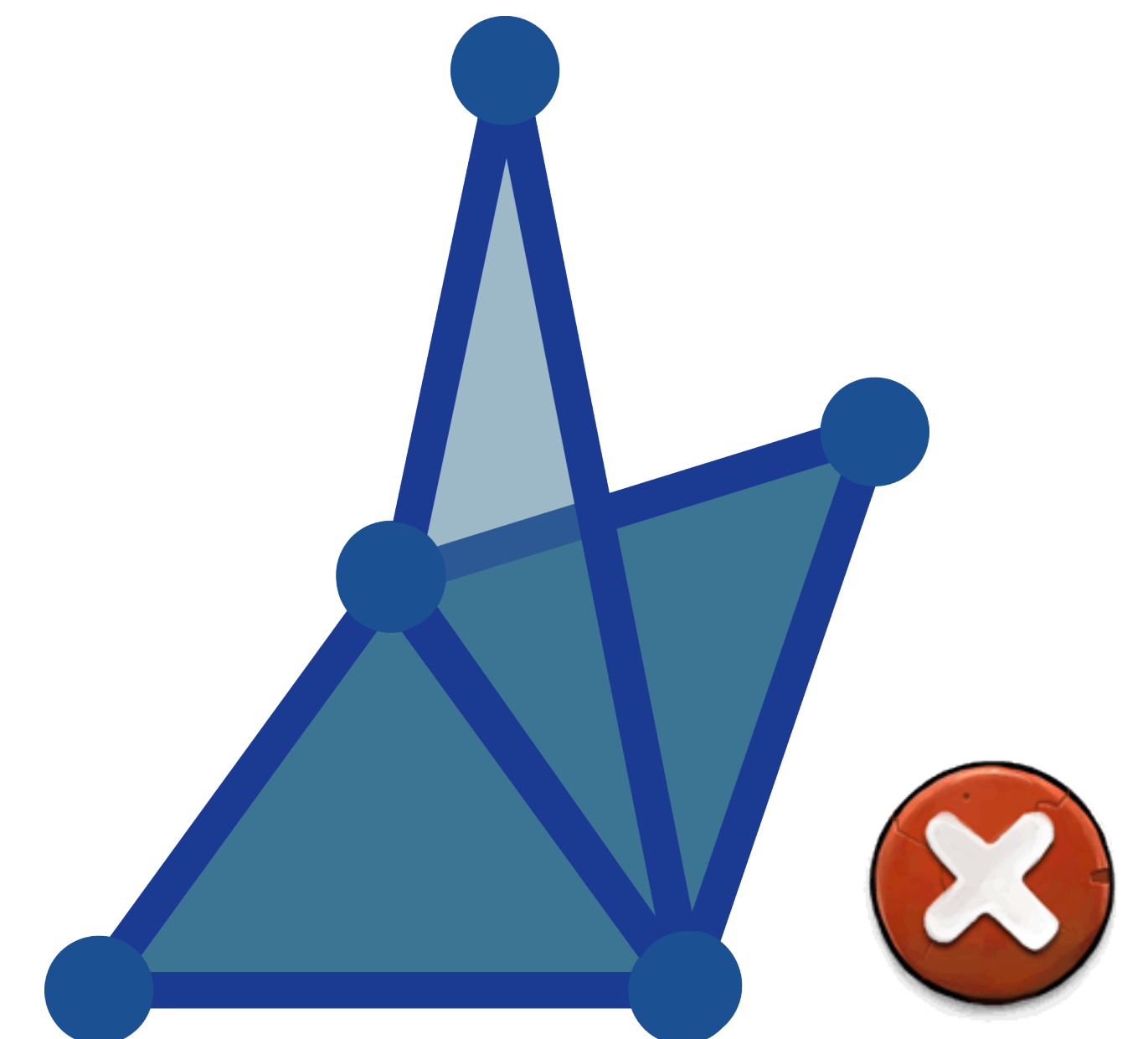
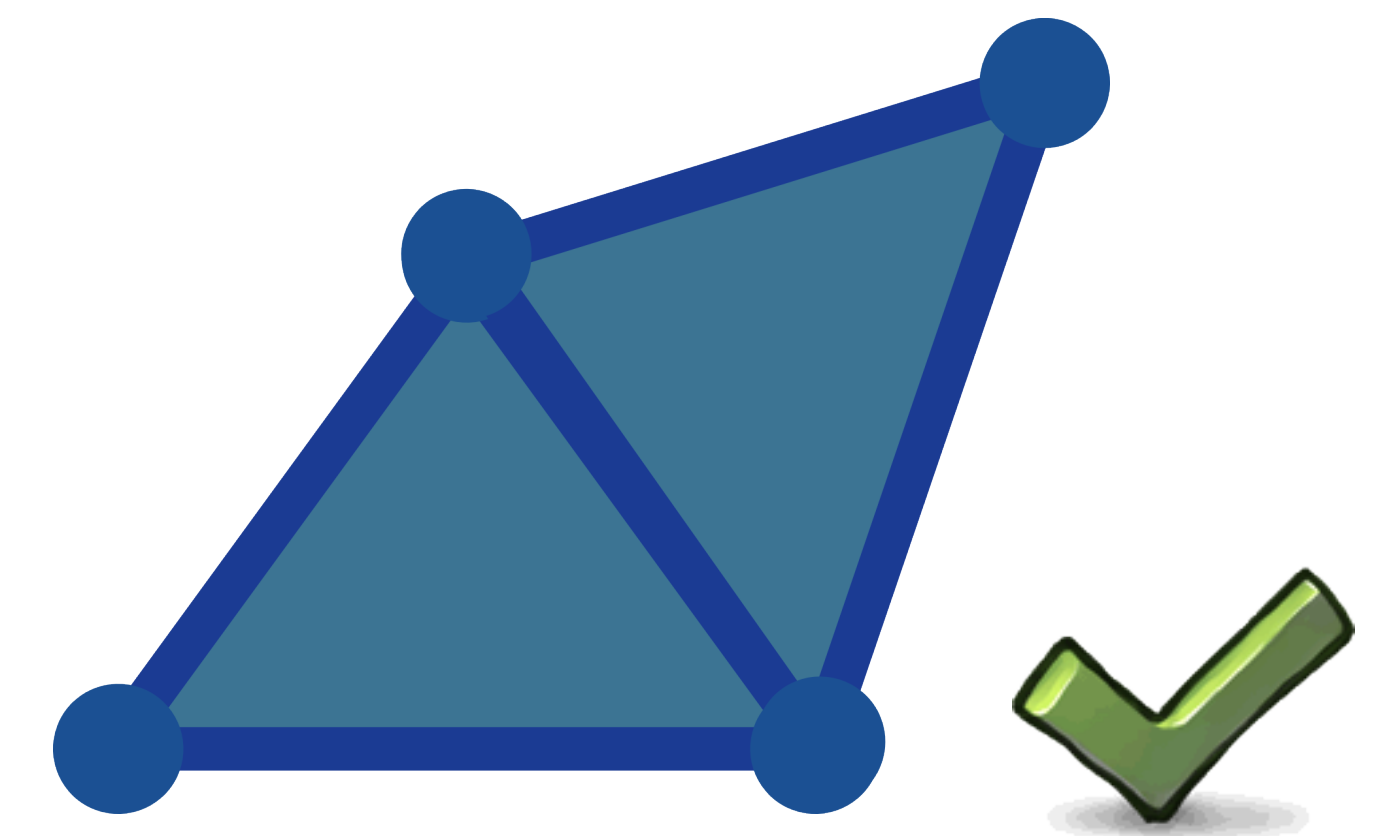
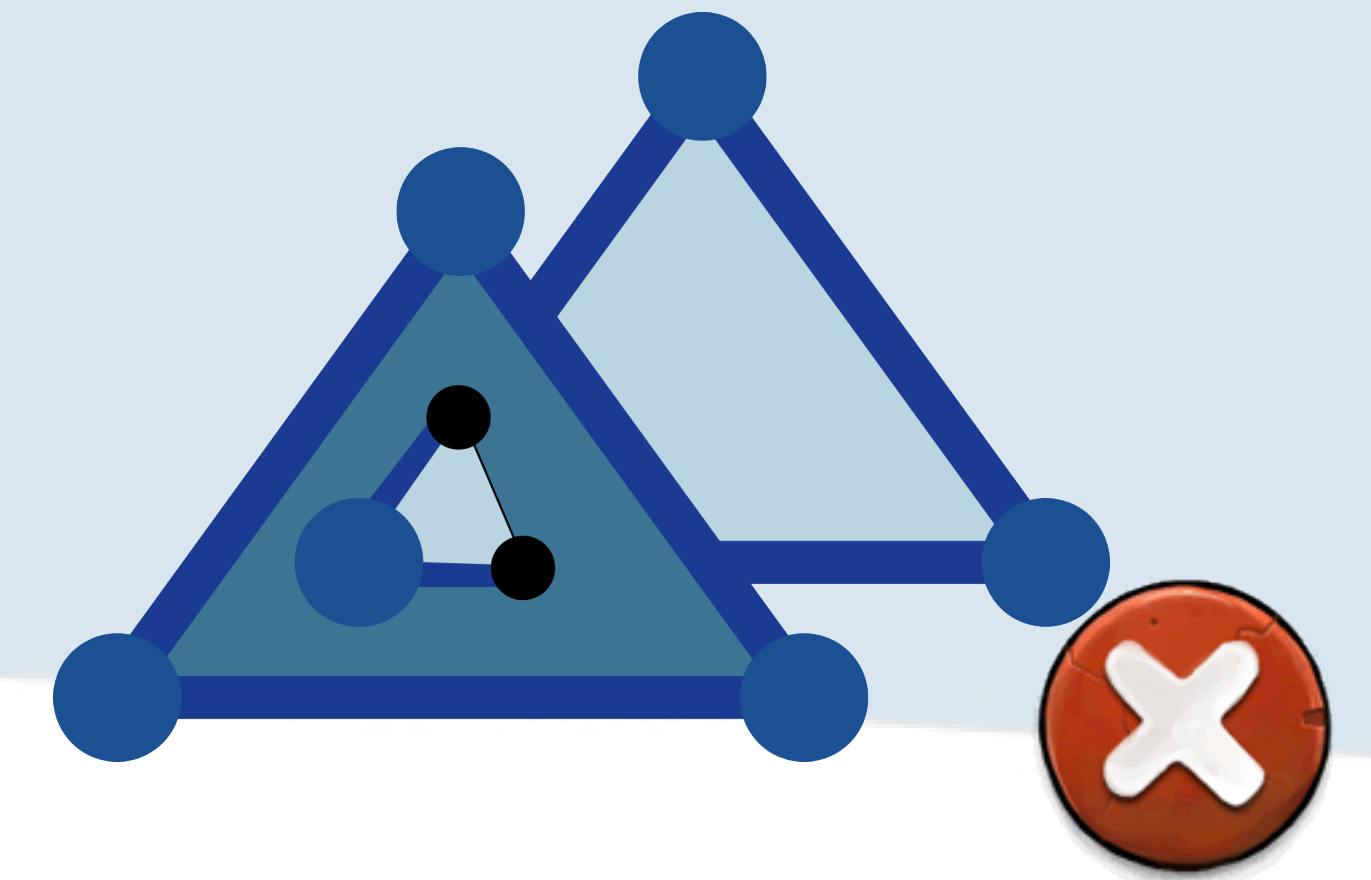
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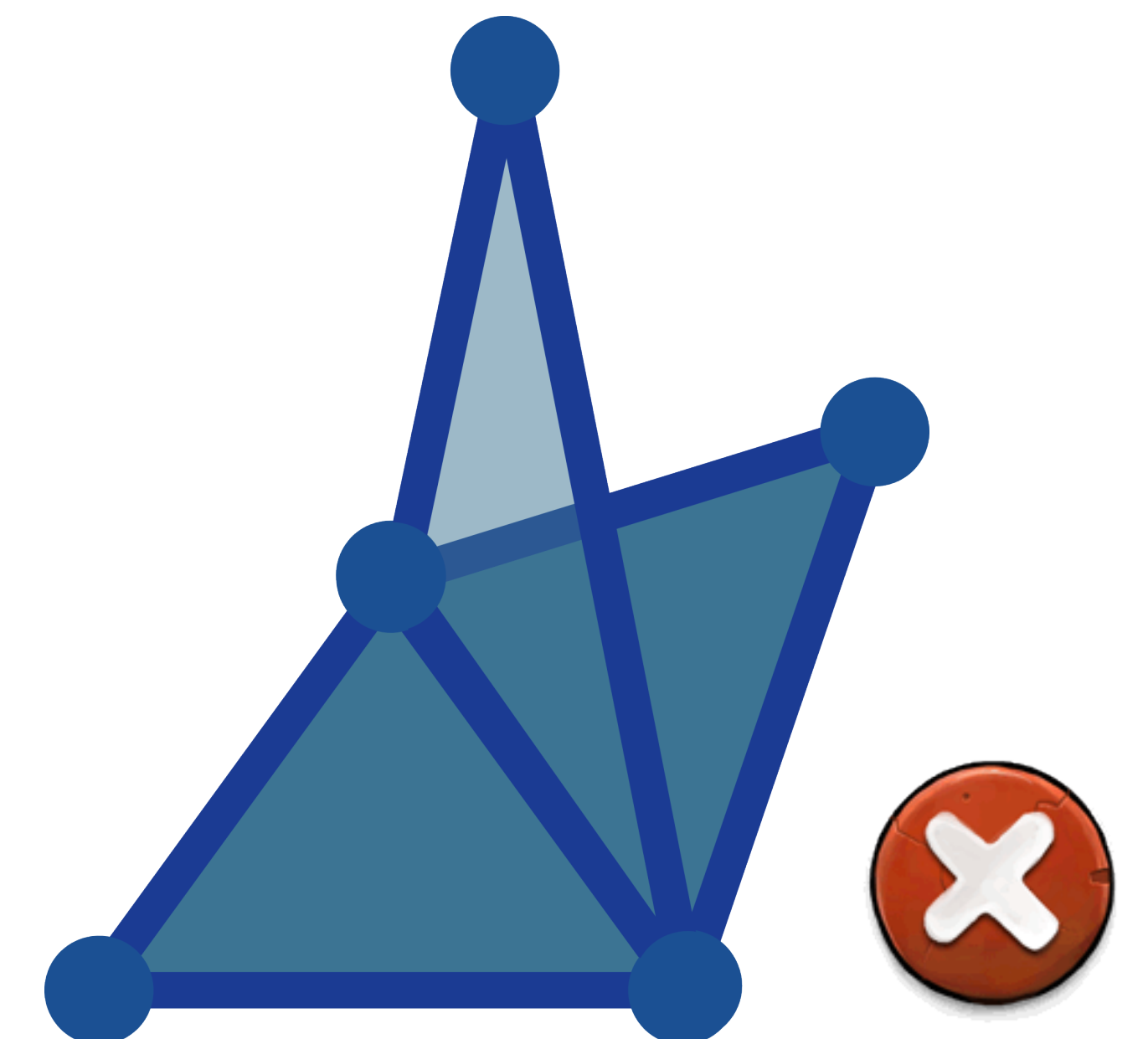
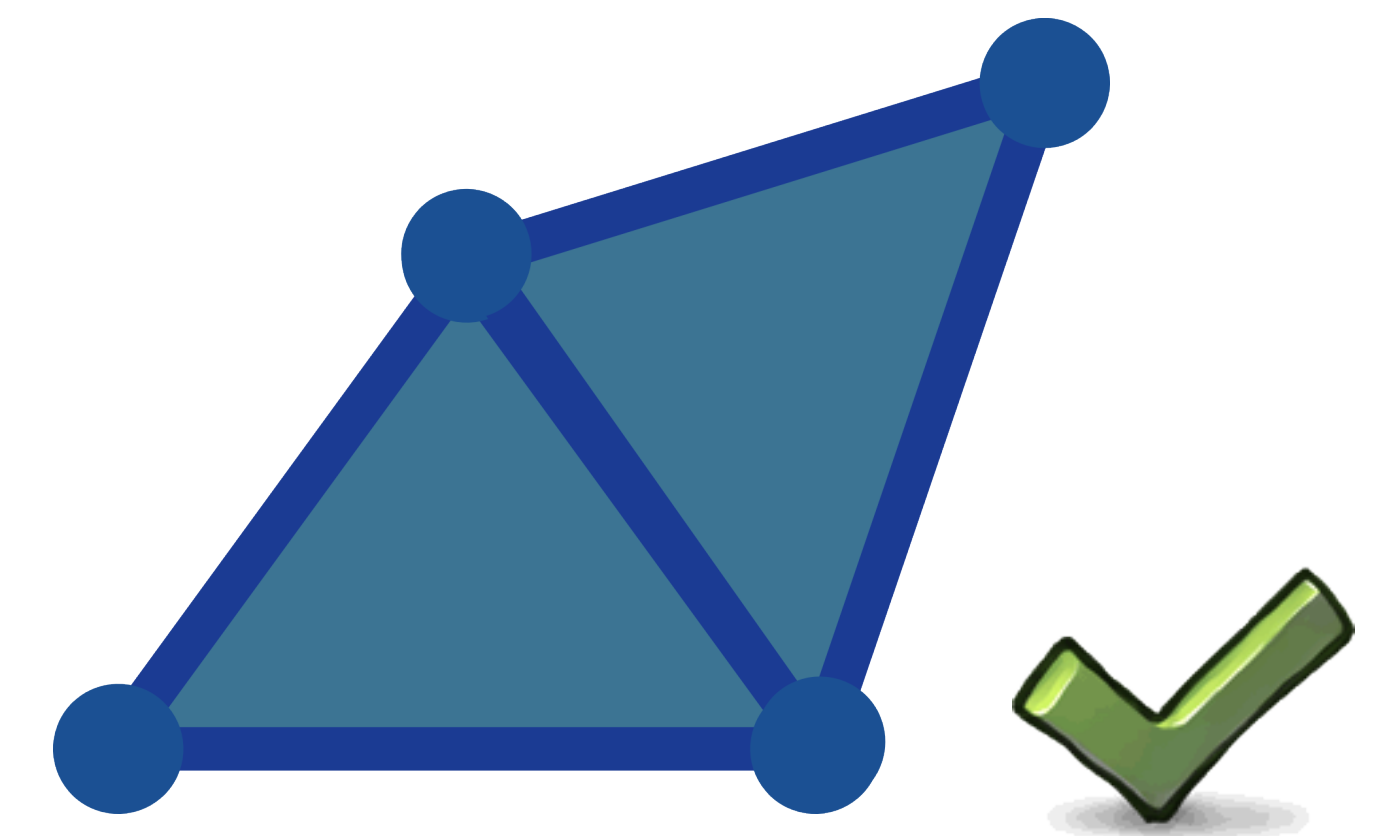
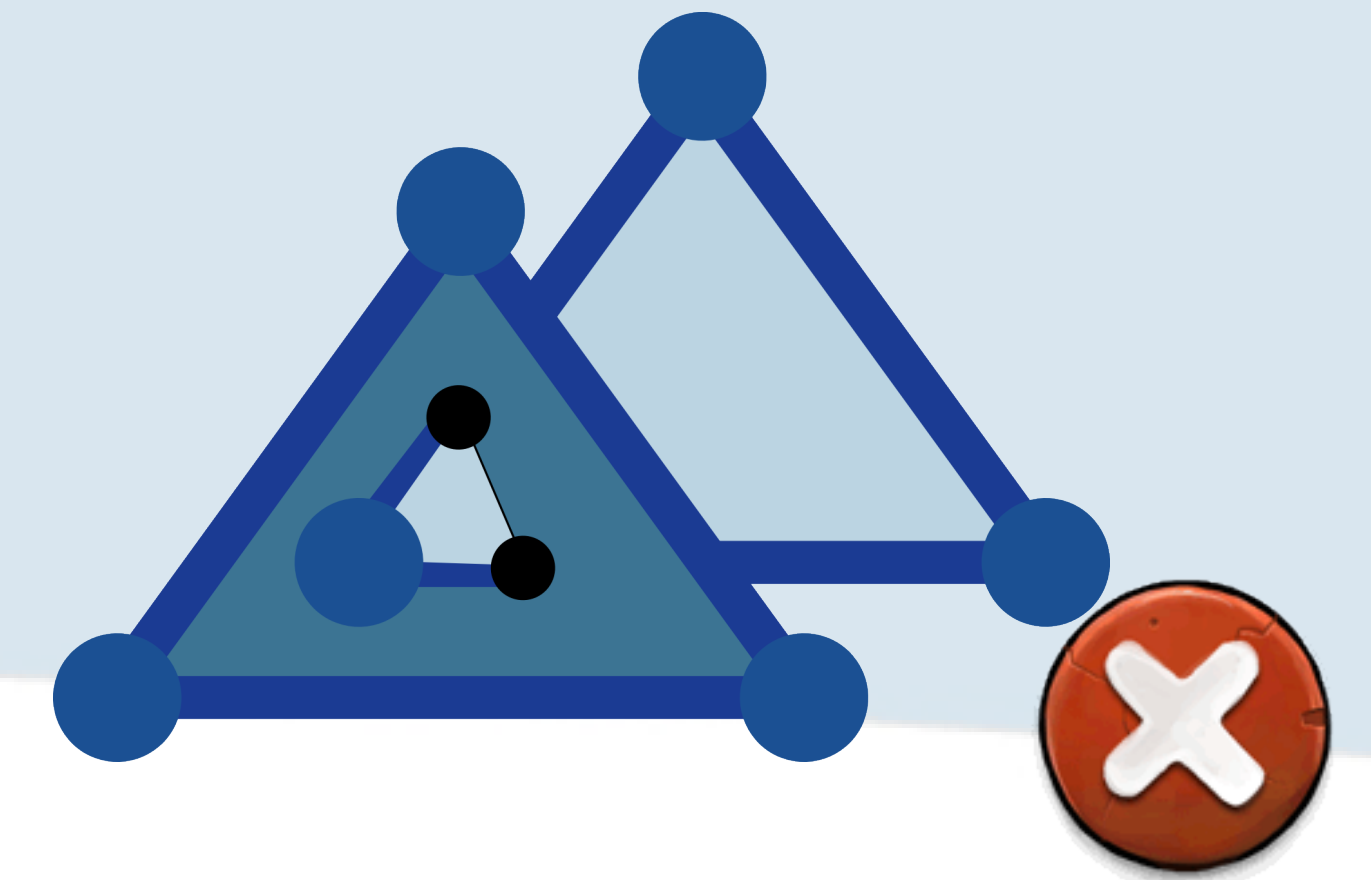
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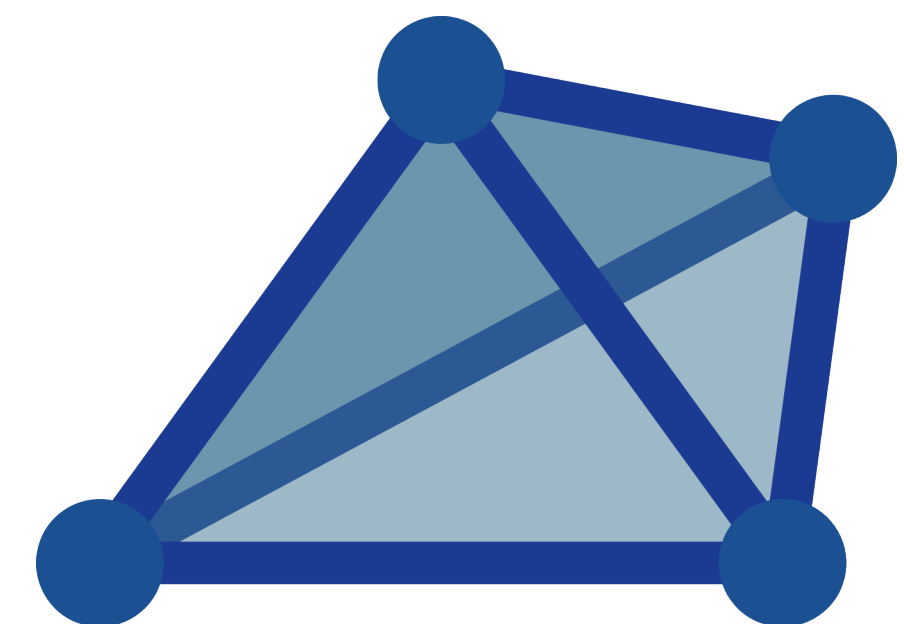
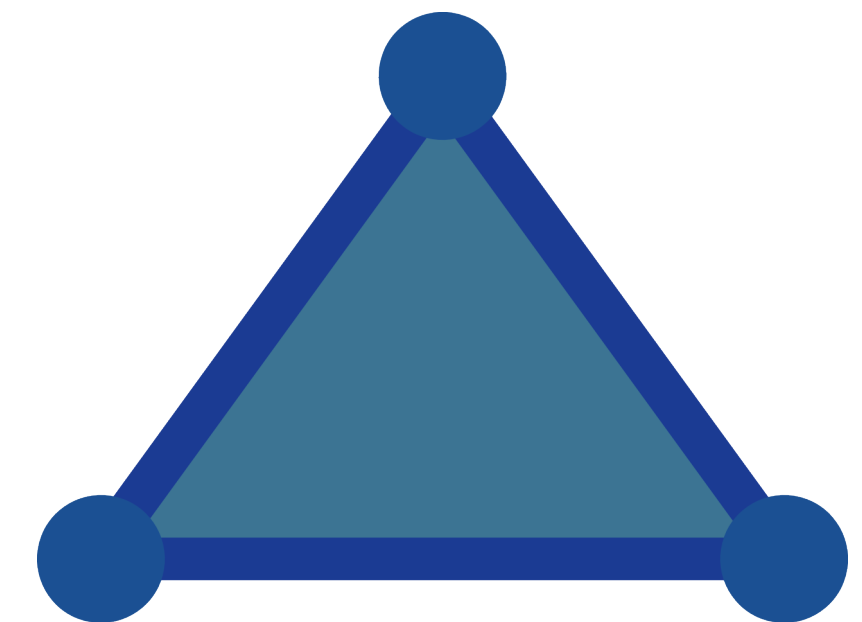
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- 2-triangulation: *triangle mesh*
- 3-triangulation: *tetrahedral mesh*



Manifolds on a computer

- How to represent that in memory?



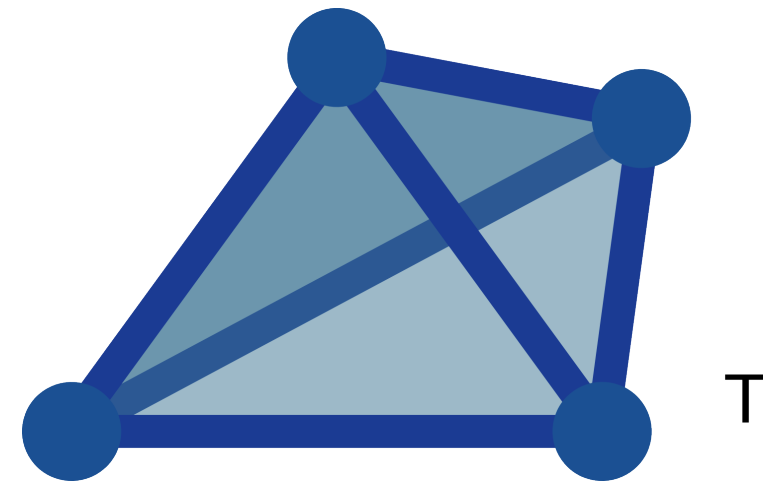
Manifolds on a computer

- How to represent that in memory?
 - Much trickier (arbitrary valence of the vertices)
 - The connectivity cannot be inferred by the geometry
 - It has to be explicitly stored



Manifolds on a computer

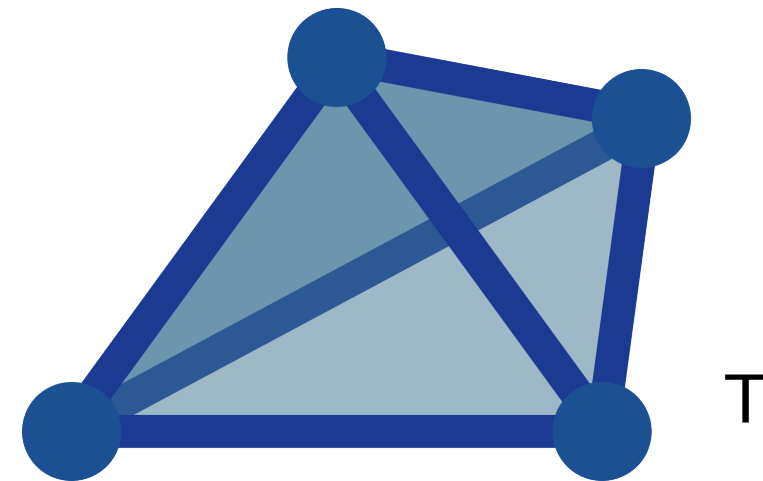
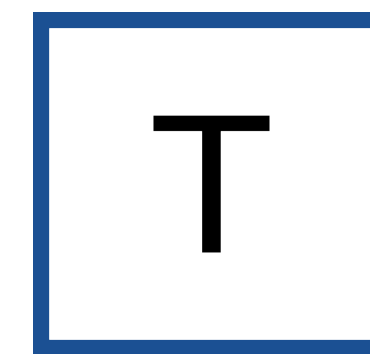
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T

Manifolds on a computer

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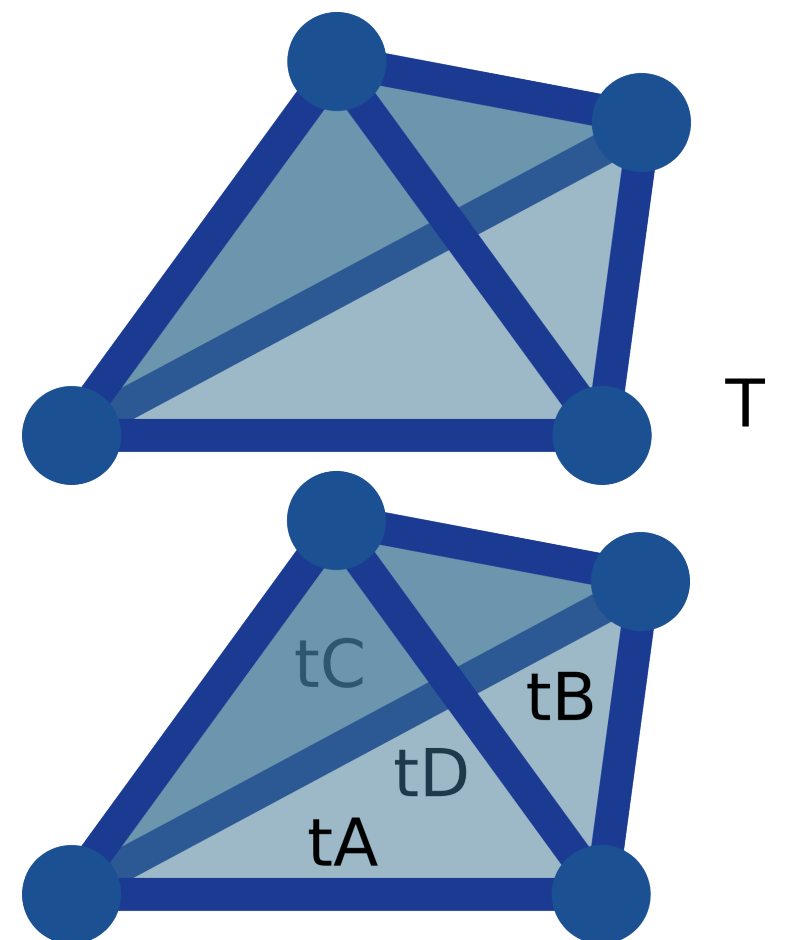


Manifolds on a computer

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T

tA	tB	tC	tD
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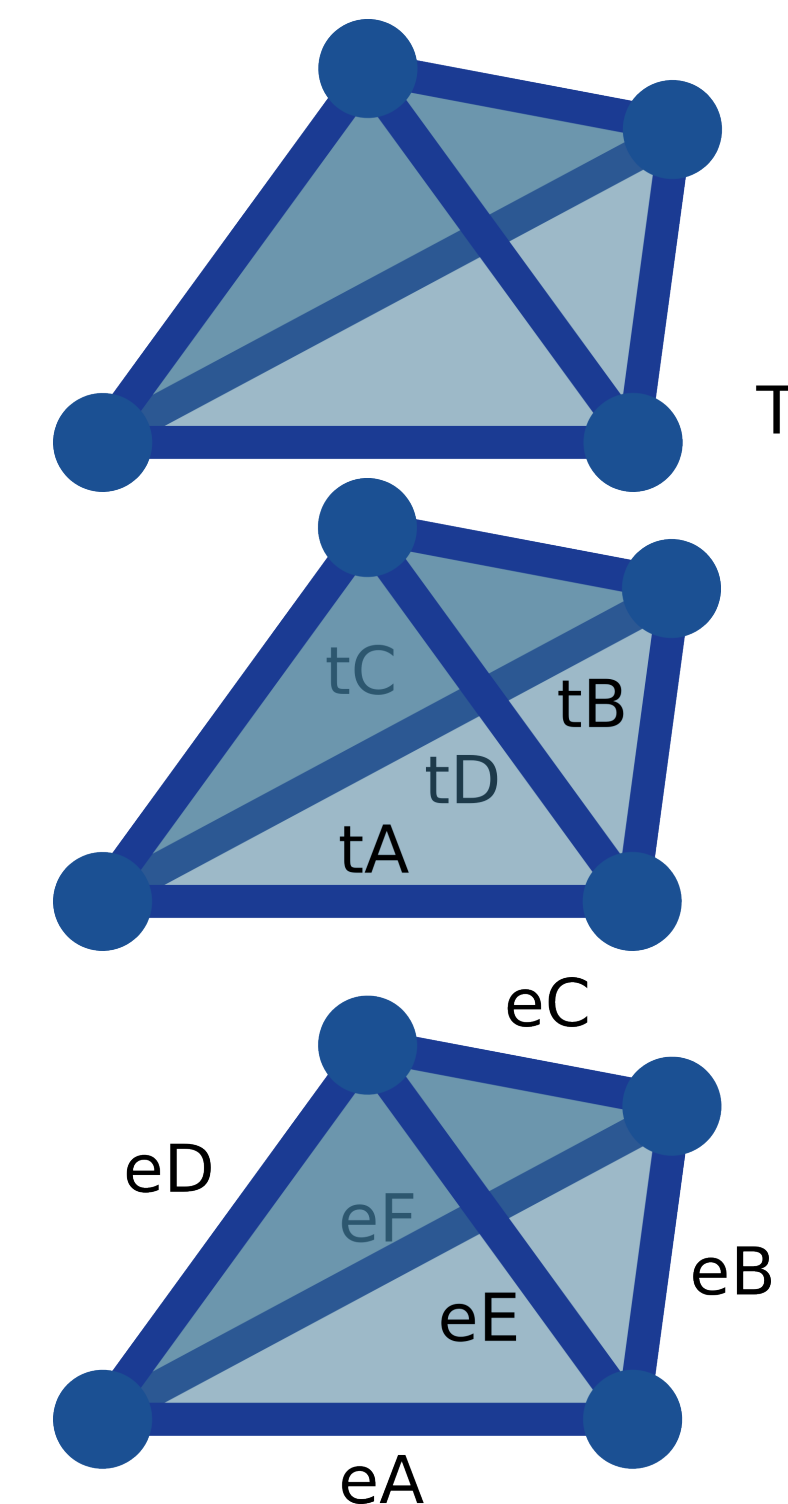
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T

tA tB tC tD

eA eB eC eD eE eF



Manifolds on a computer

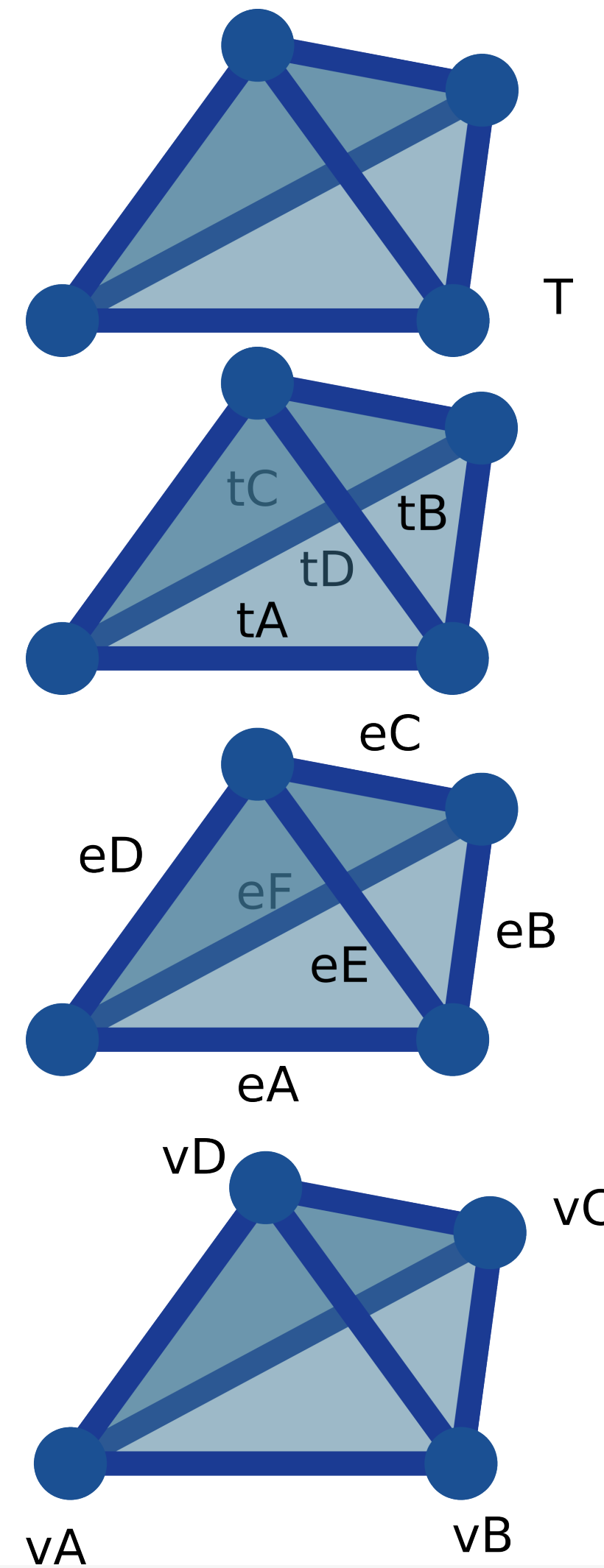
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T

tA tB tC tD

eA eB eC eD eE eF

vA vB vC vD



Manifolds on a computer

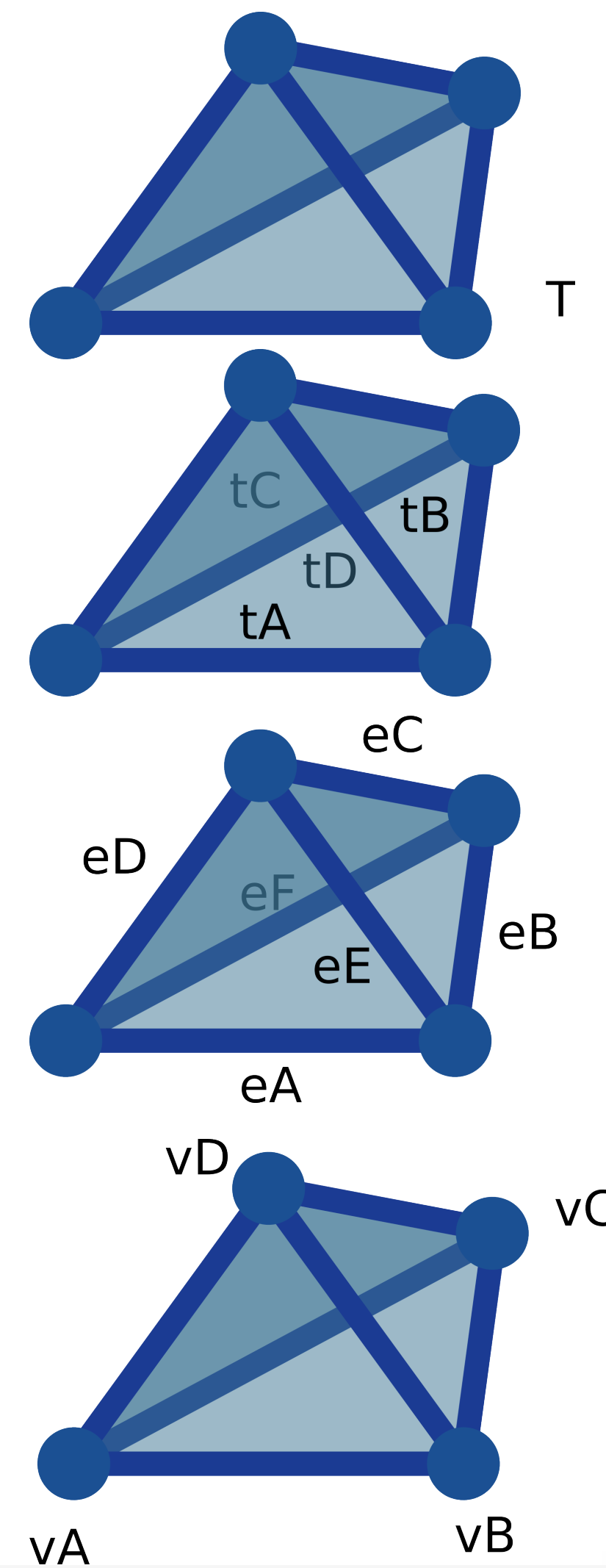
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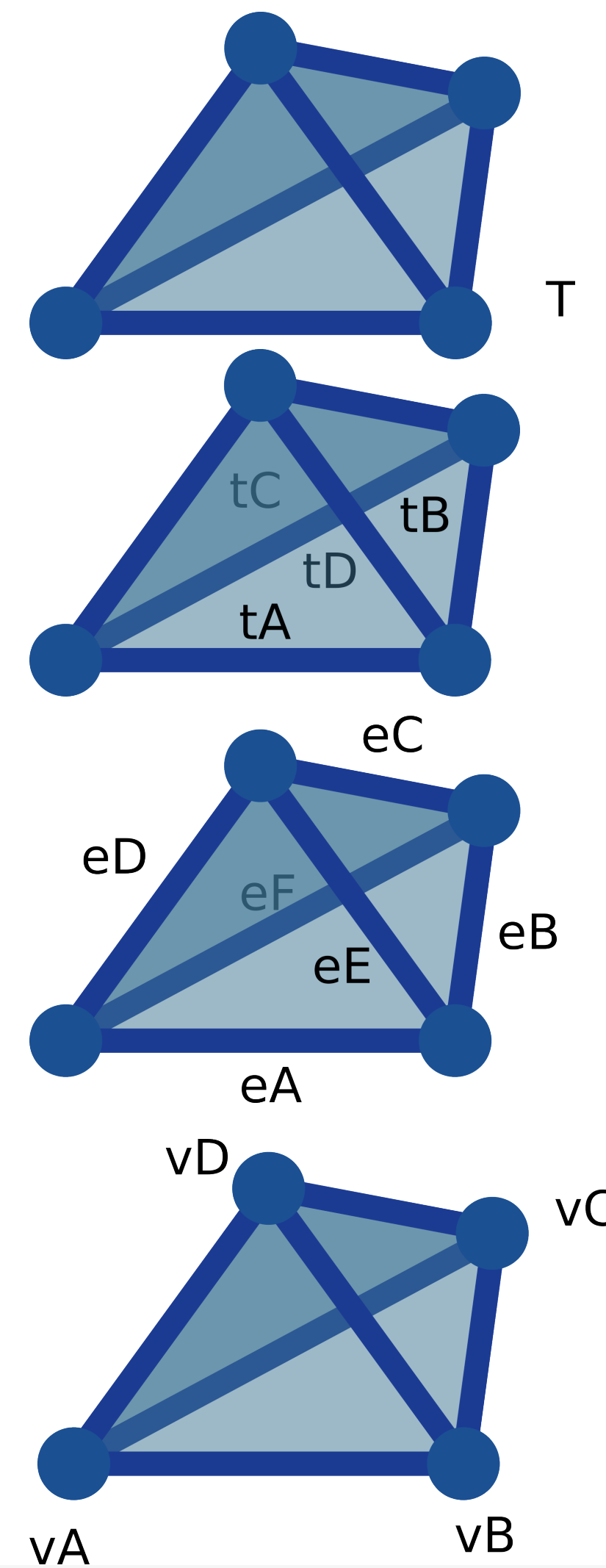
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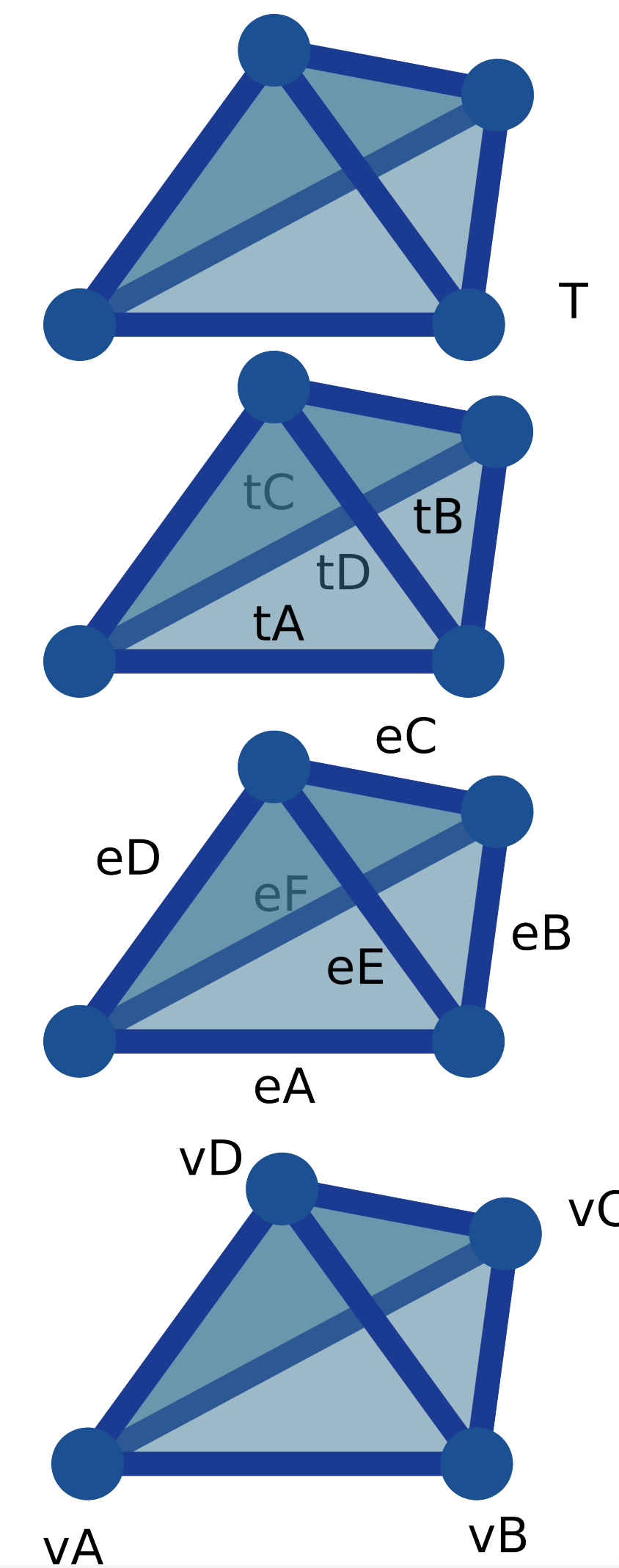
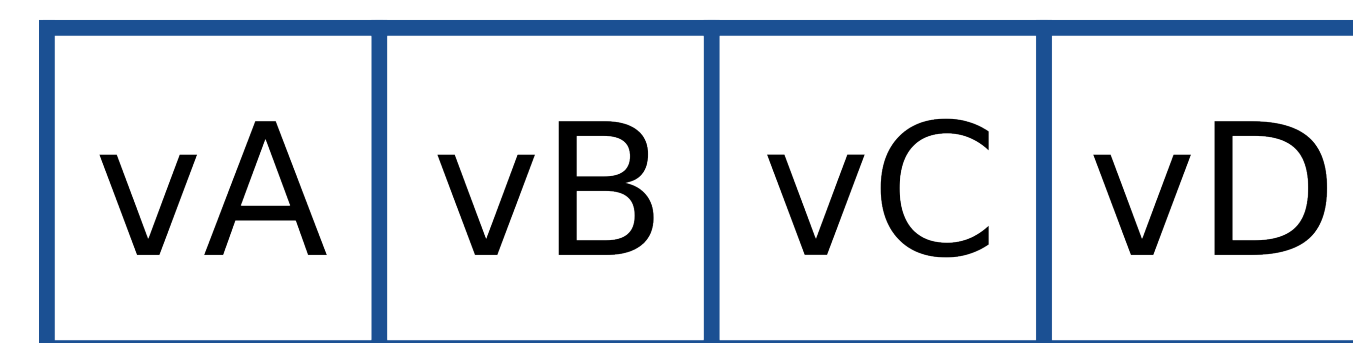
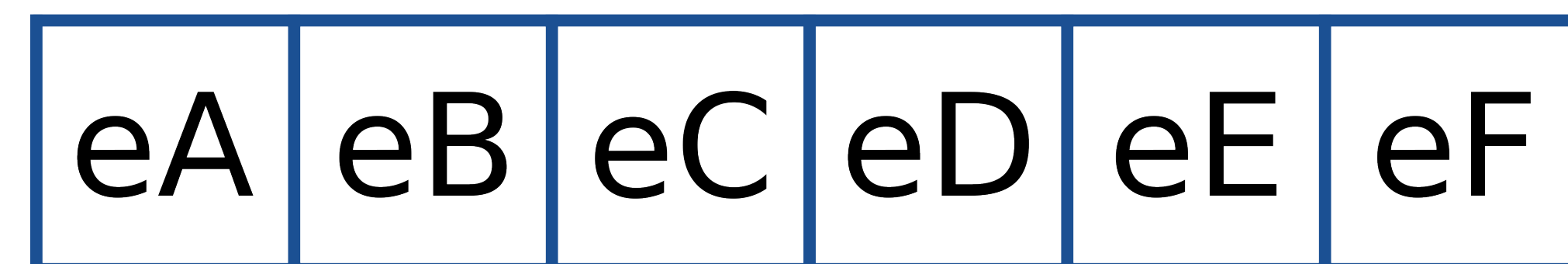
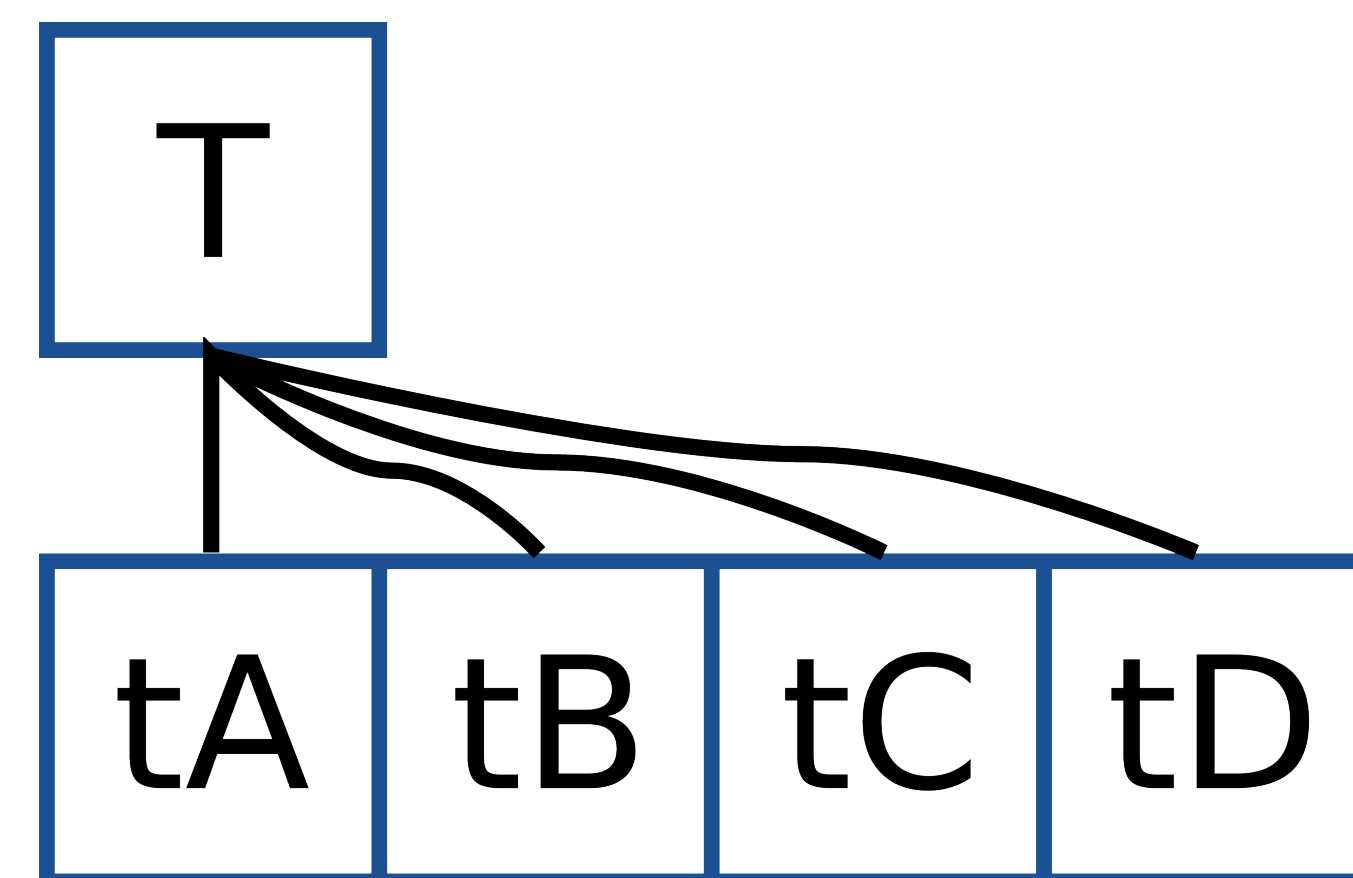
eA eB eC eD eE eF

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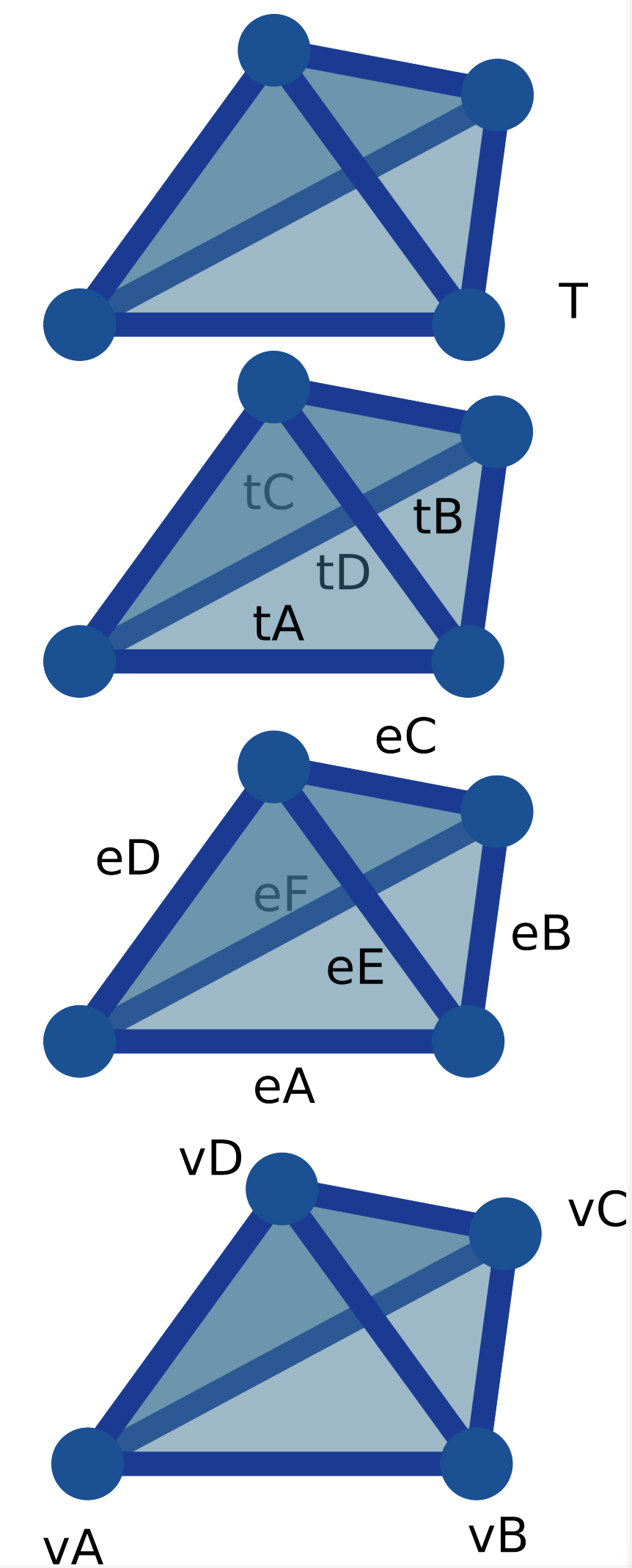
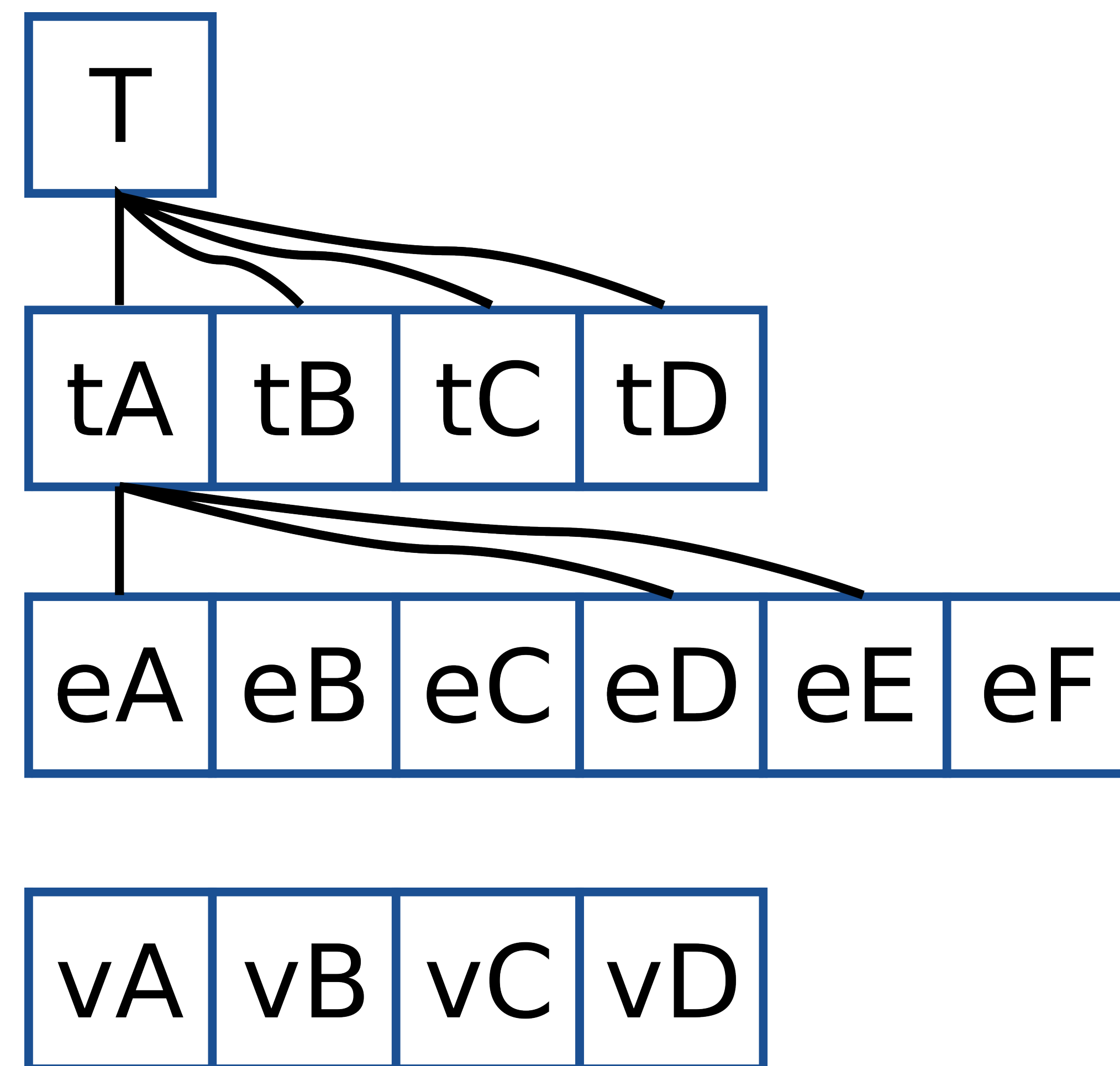
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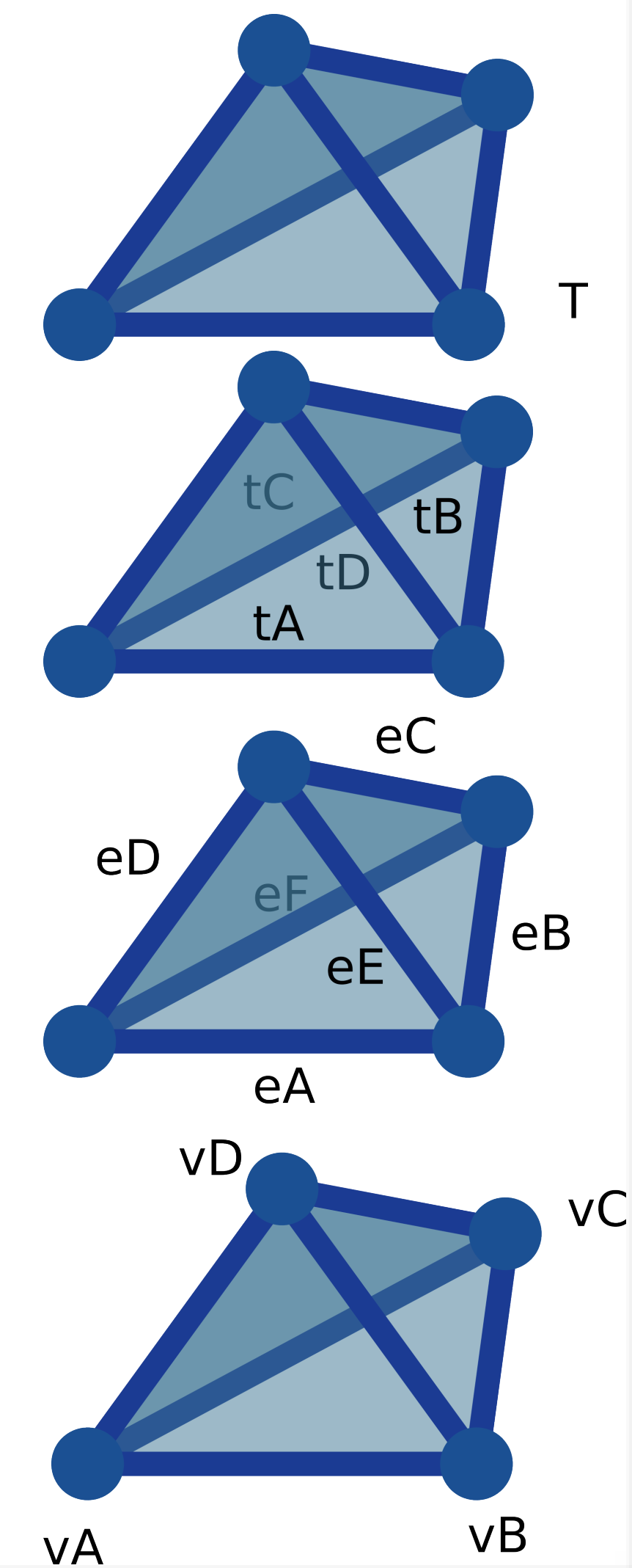
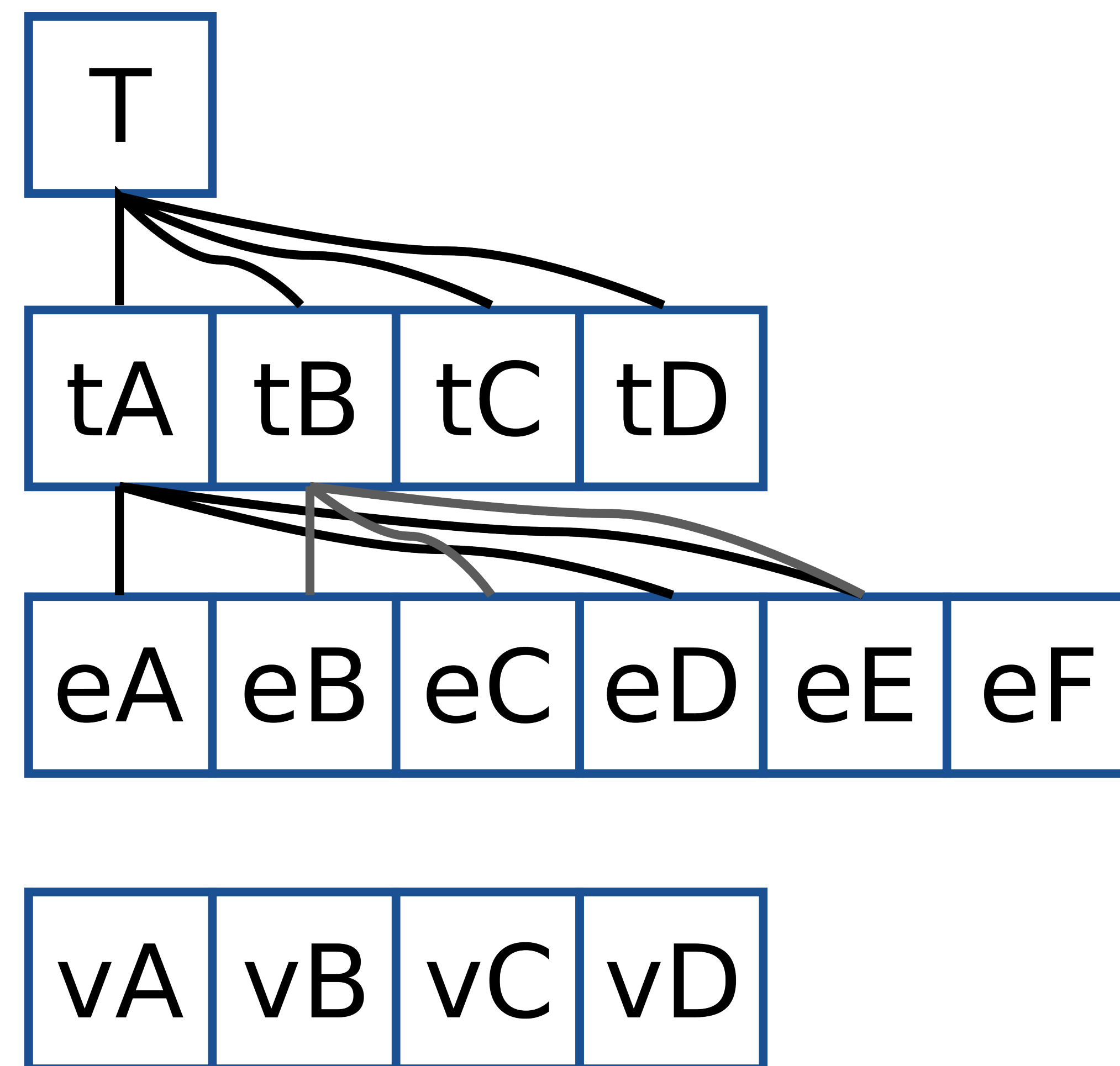
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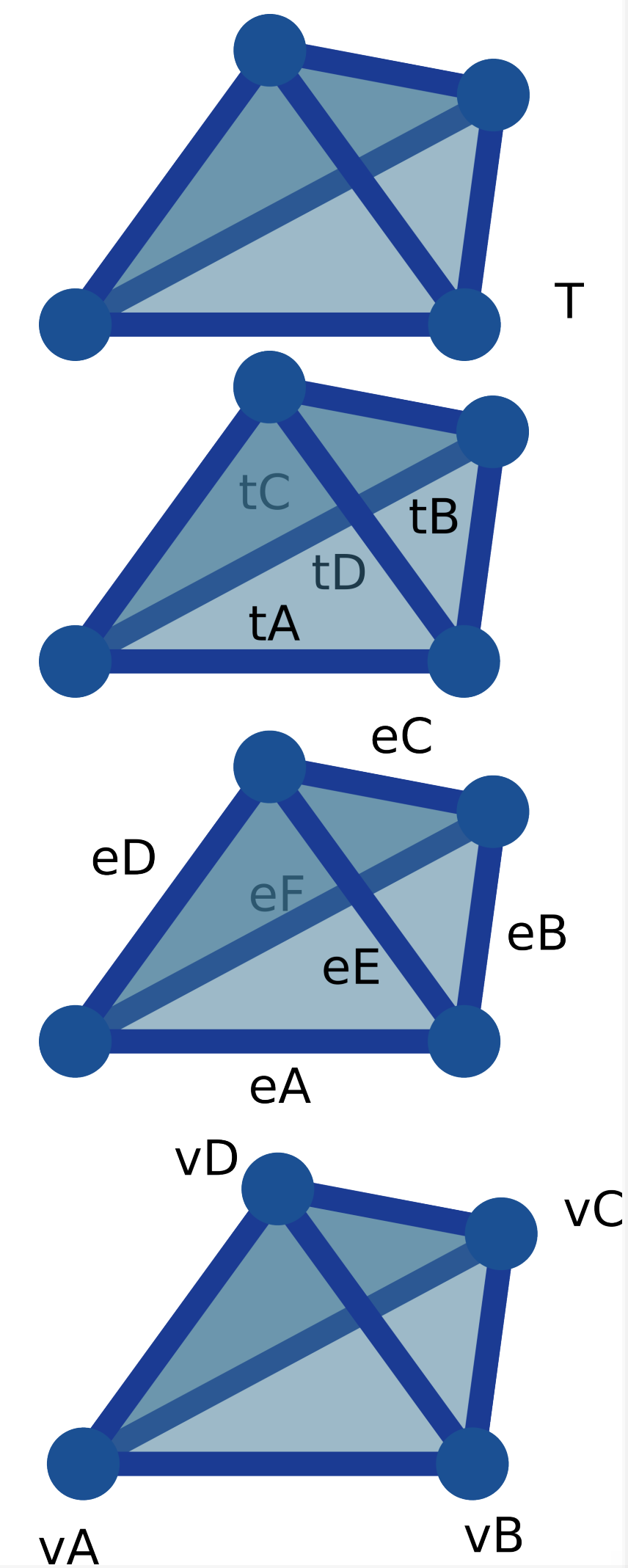
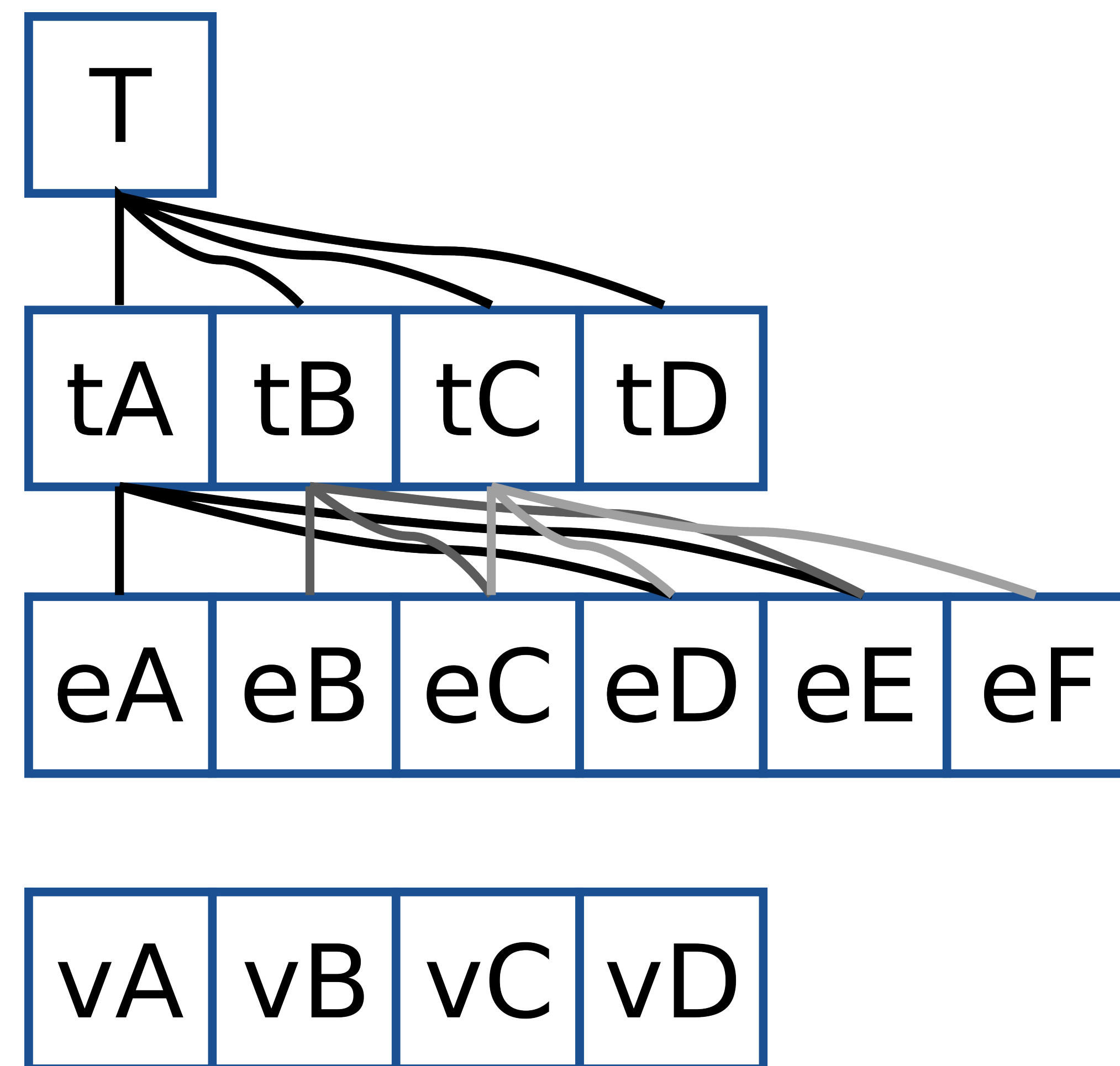
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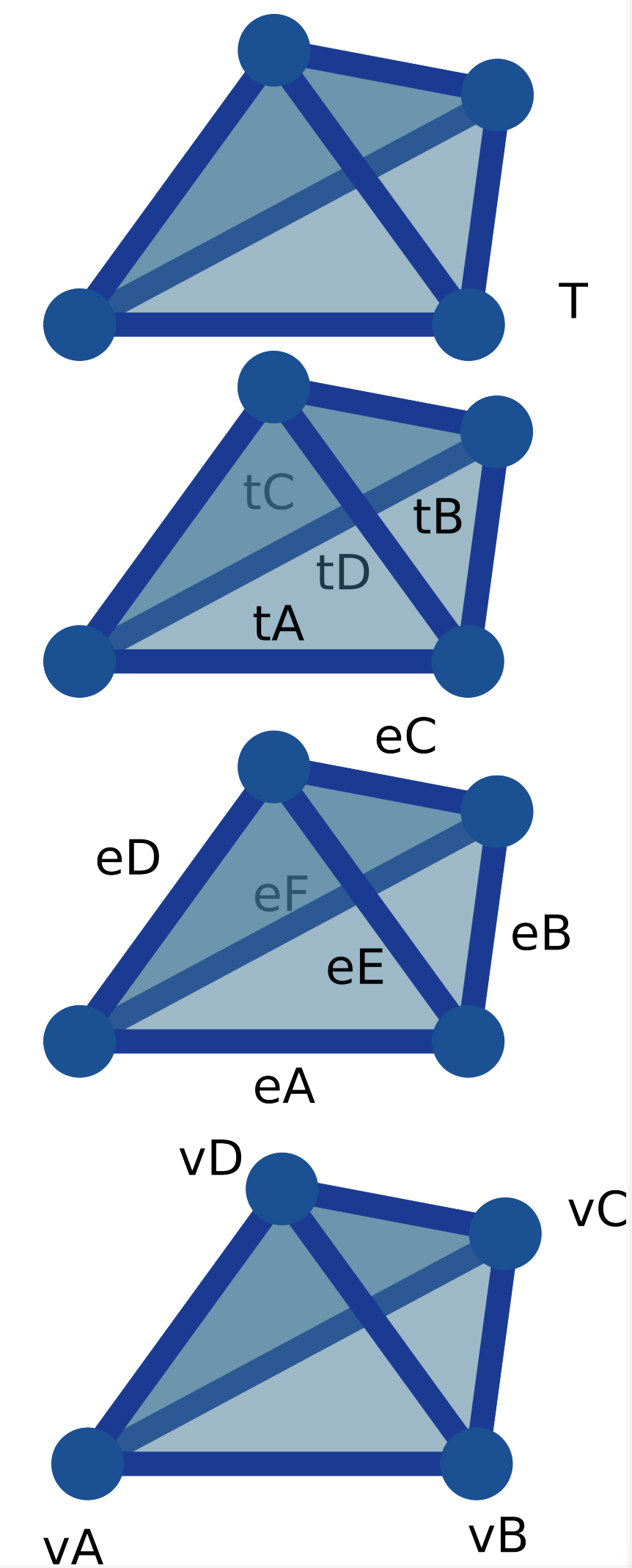
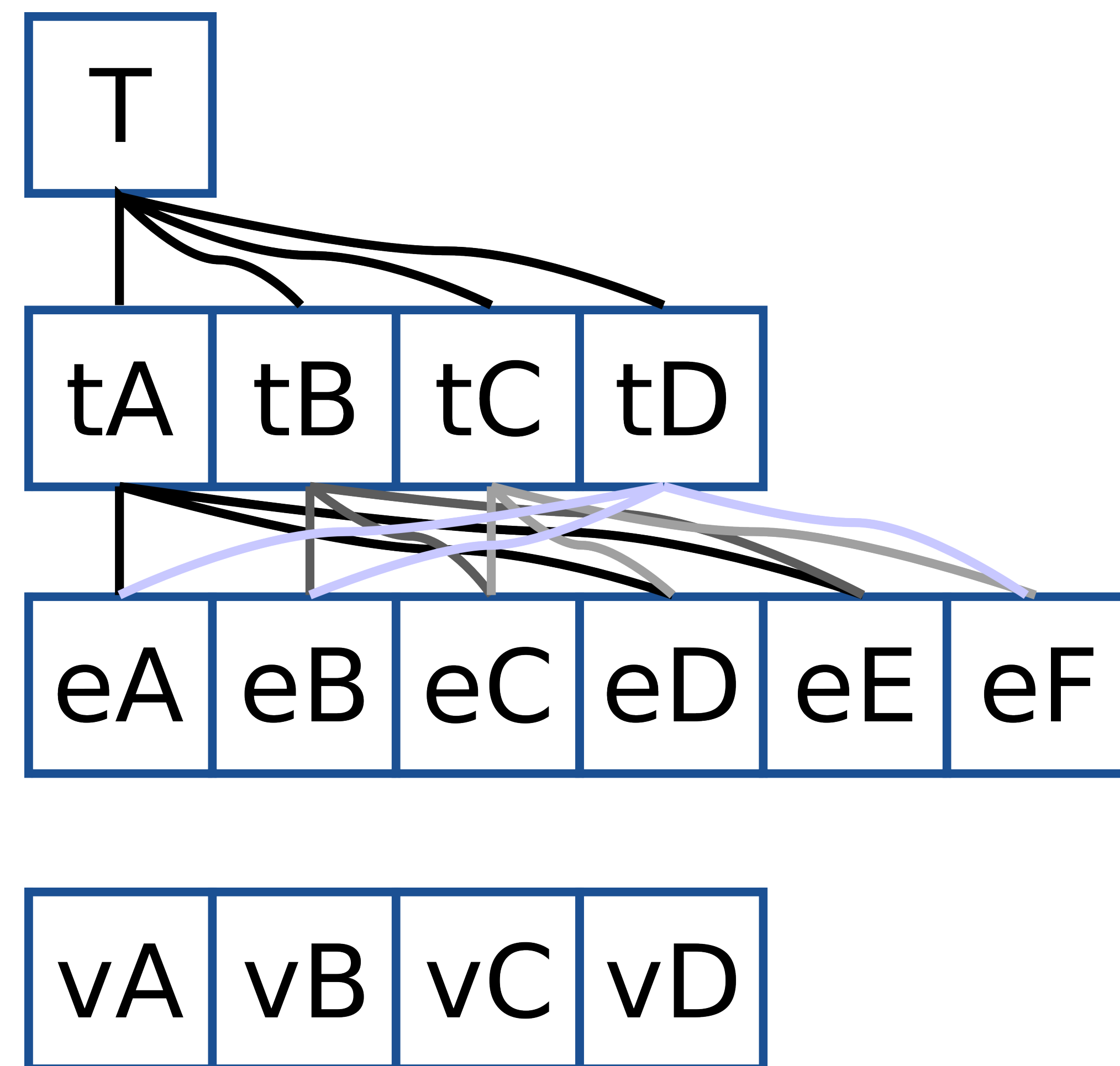
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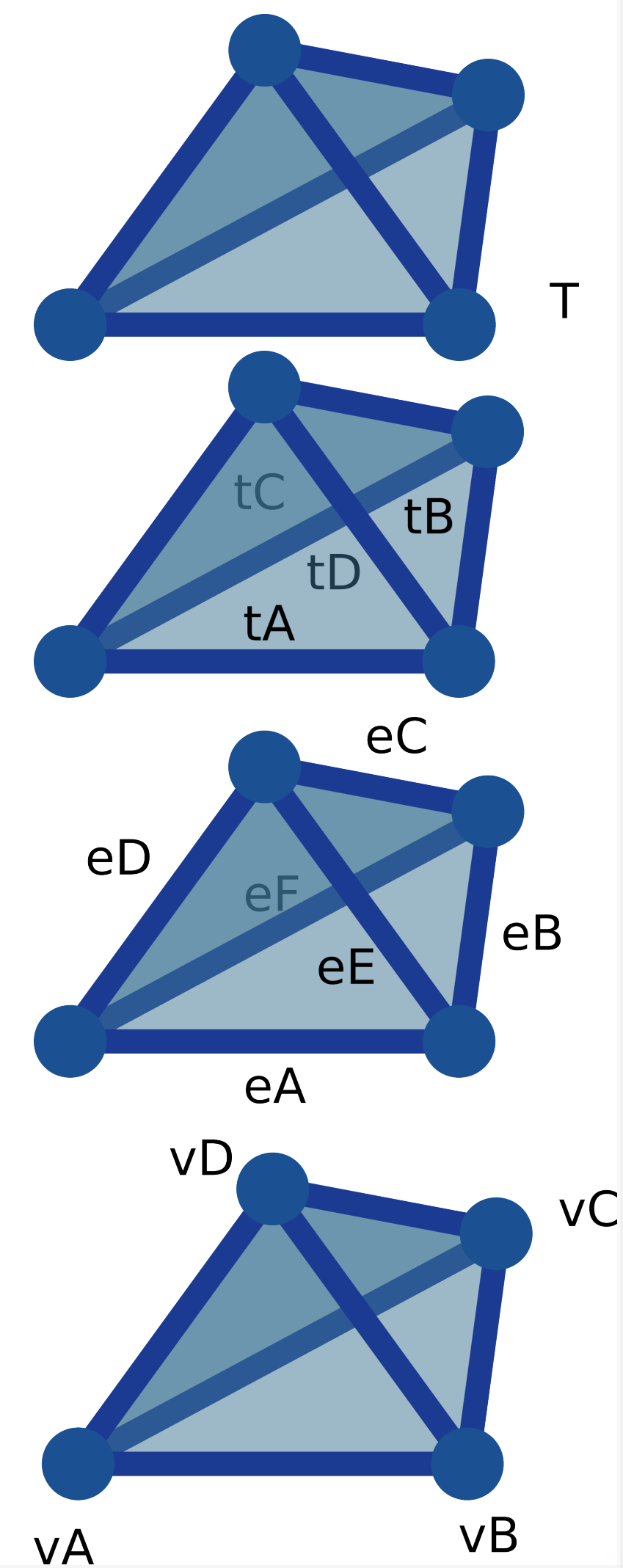
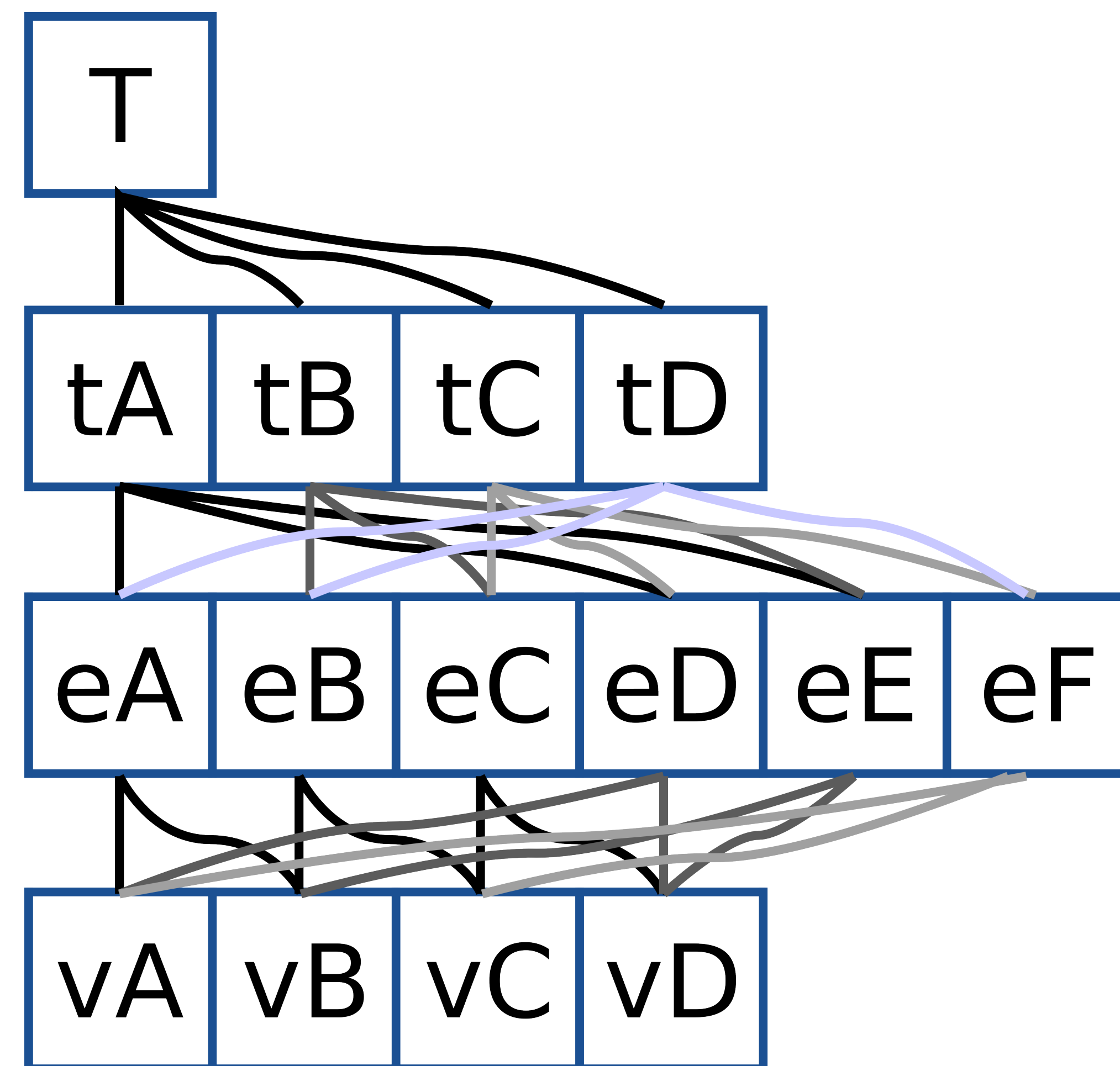
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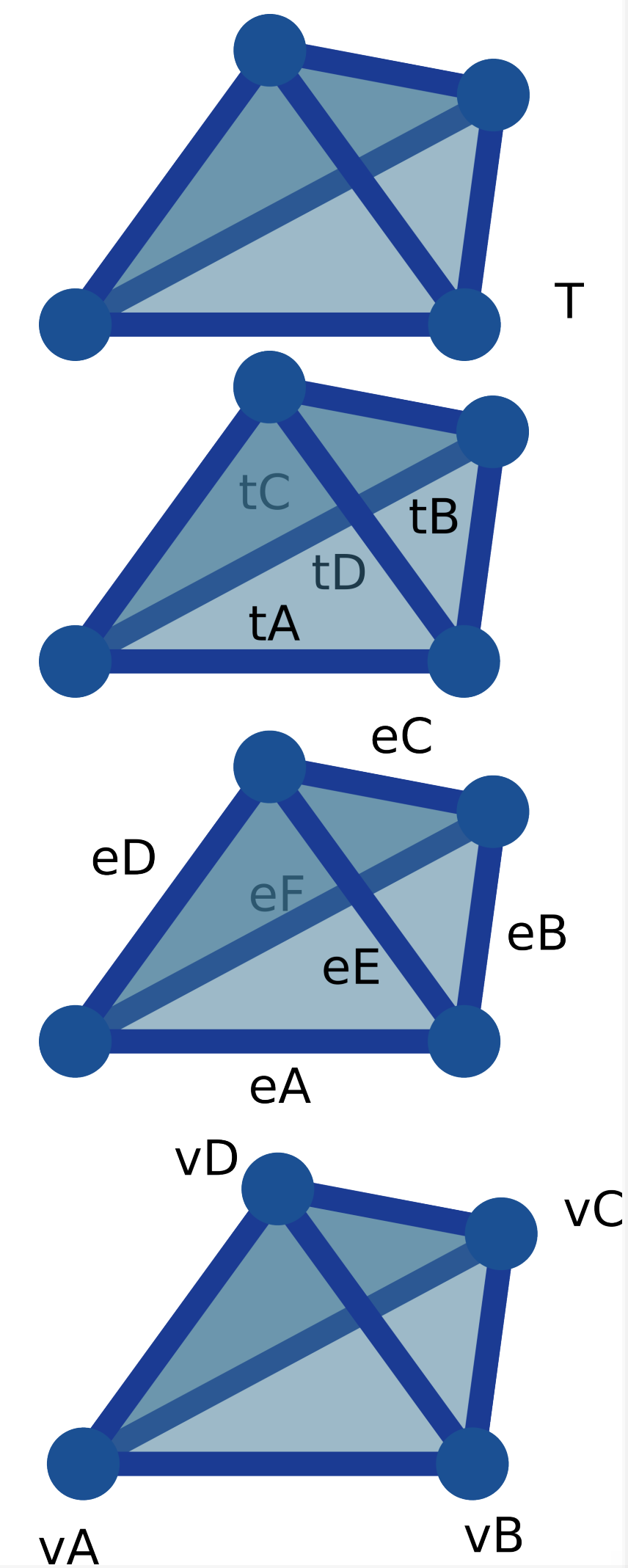
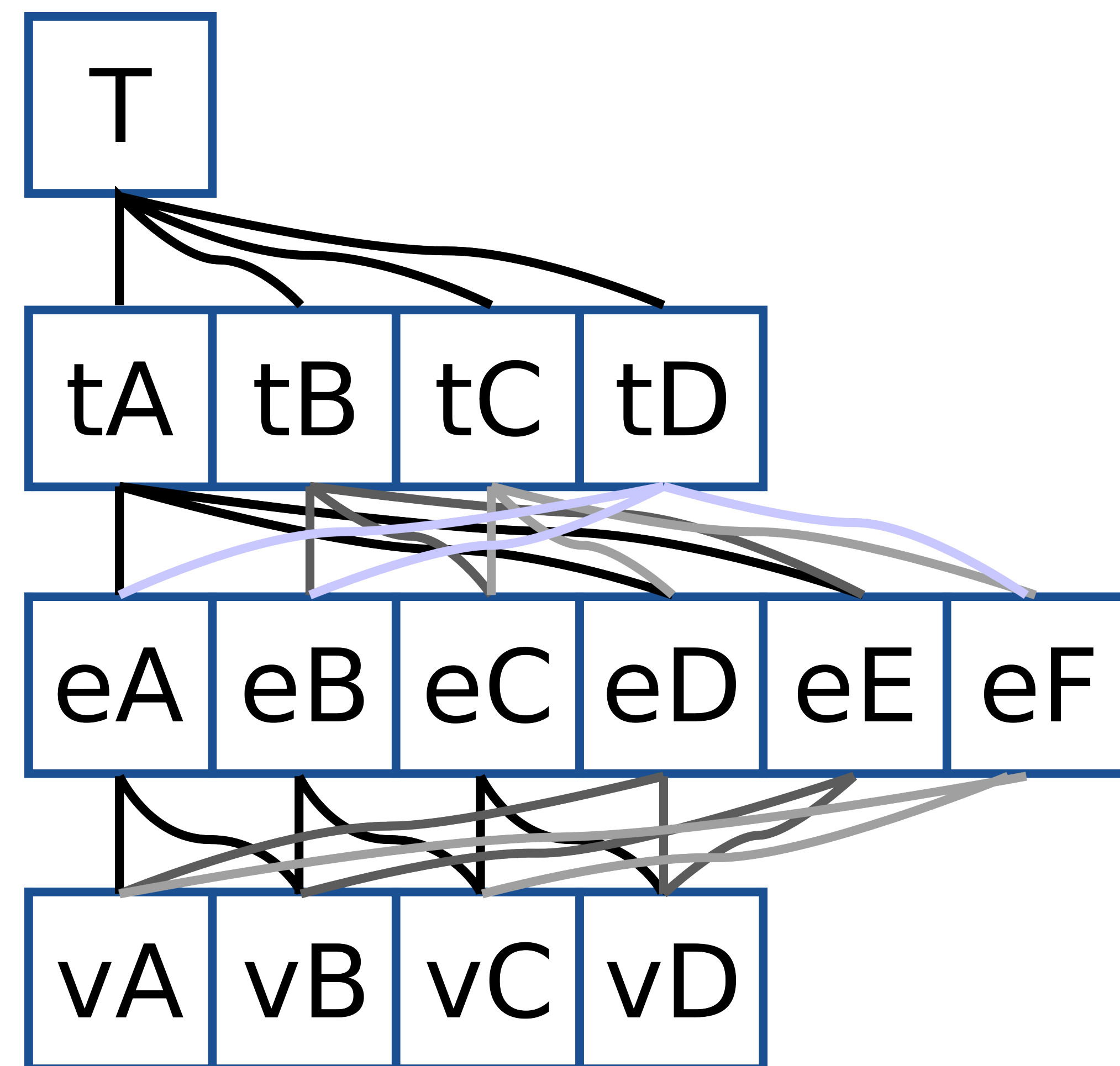
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 - Navigation with bi-directional pointers



Consistency check

- There's a lot of room for bugs....

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$$\chi = \sum_{i=0}^d (-1)^i |\{\sigma^i\}|$$

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$$\chi = \#_{vertices} - \#_{edges} + \#_{triangles} - \#_{tetrahedra} = \sum_{i=0}^d (-1)^i \beta_i$$

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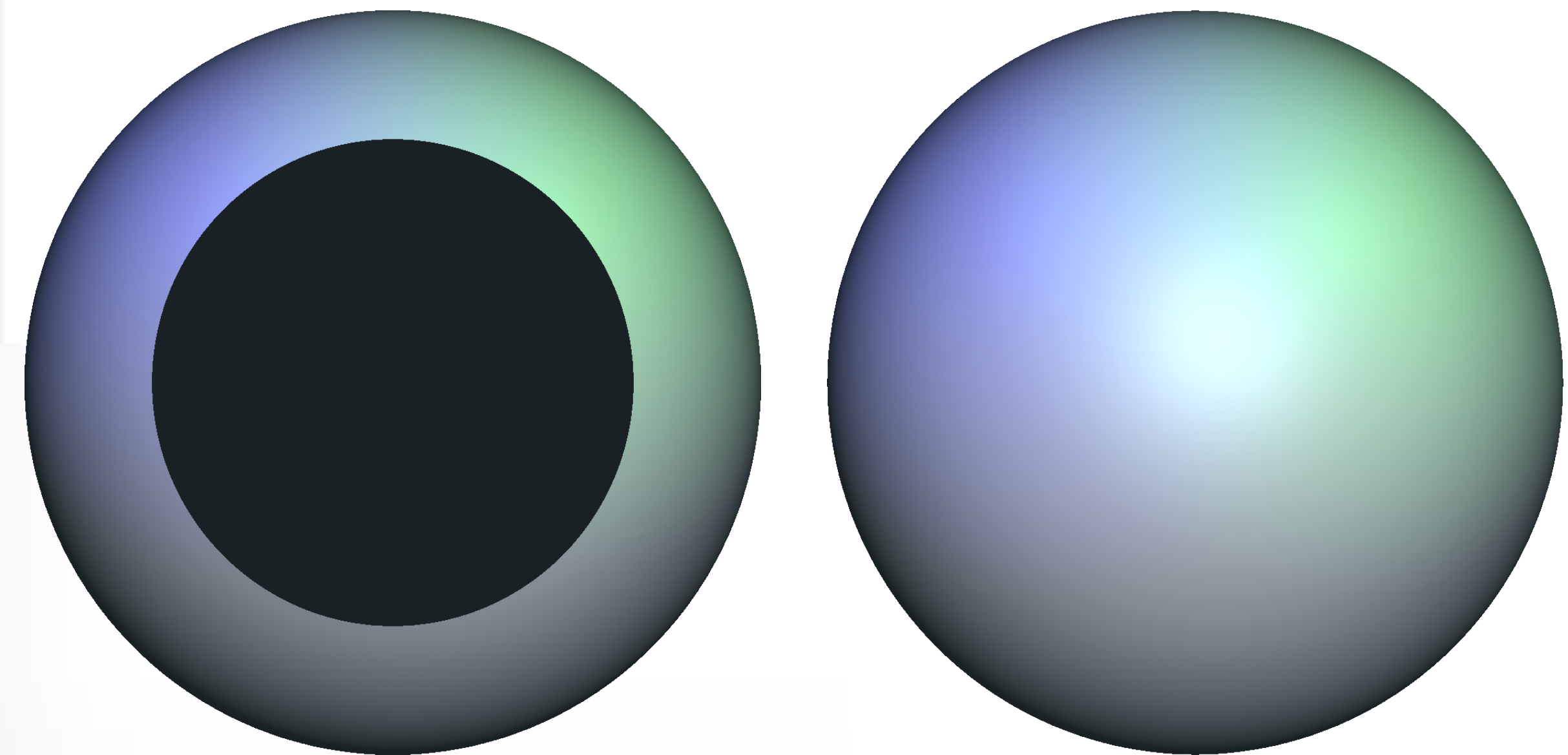
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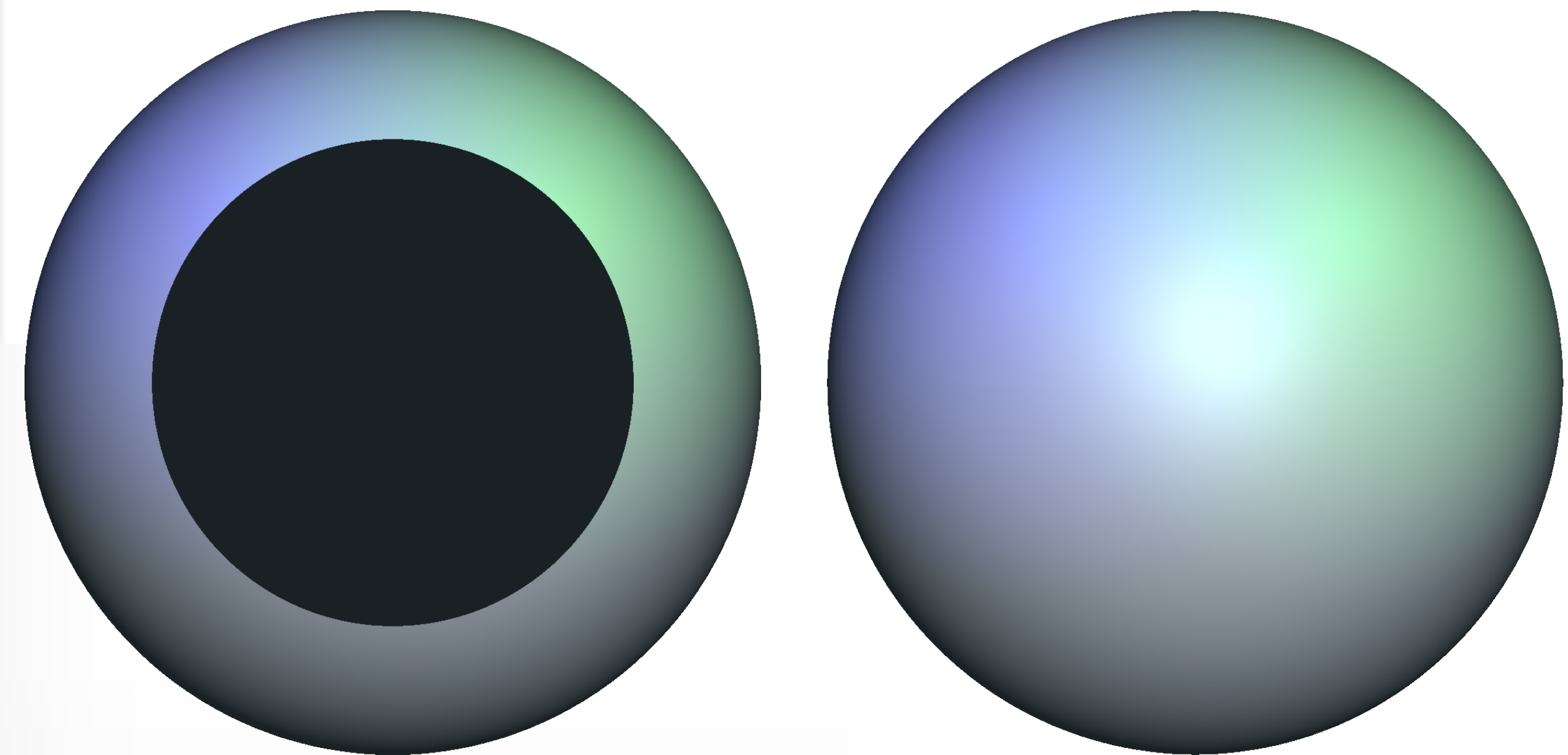
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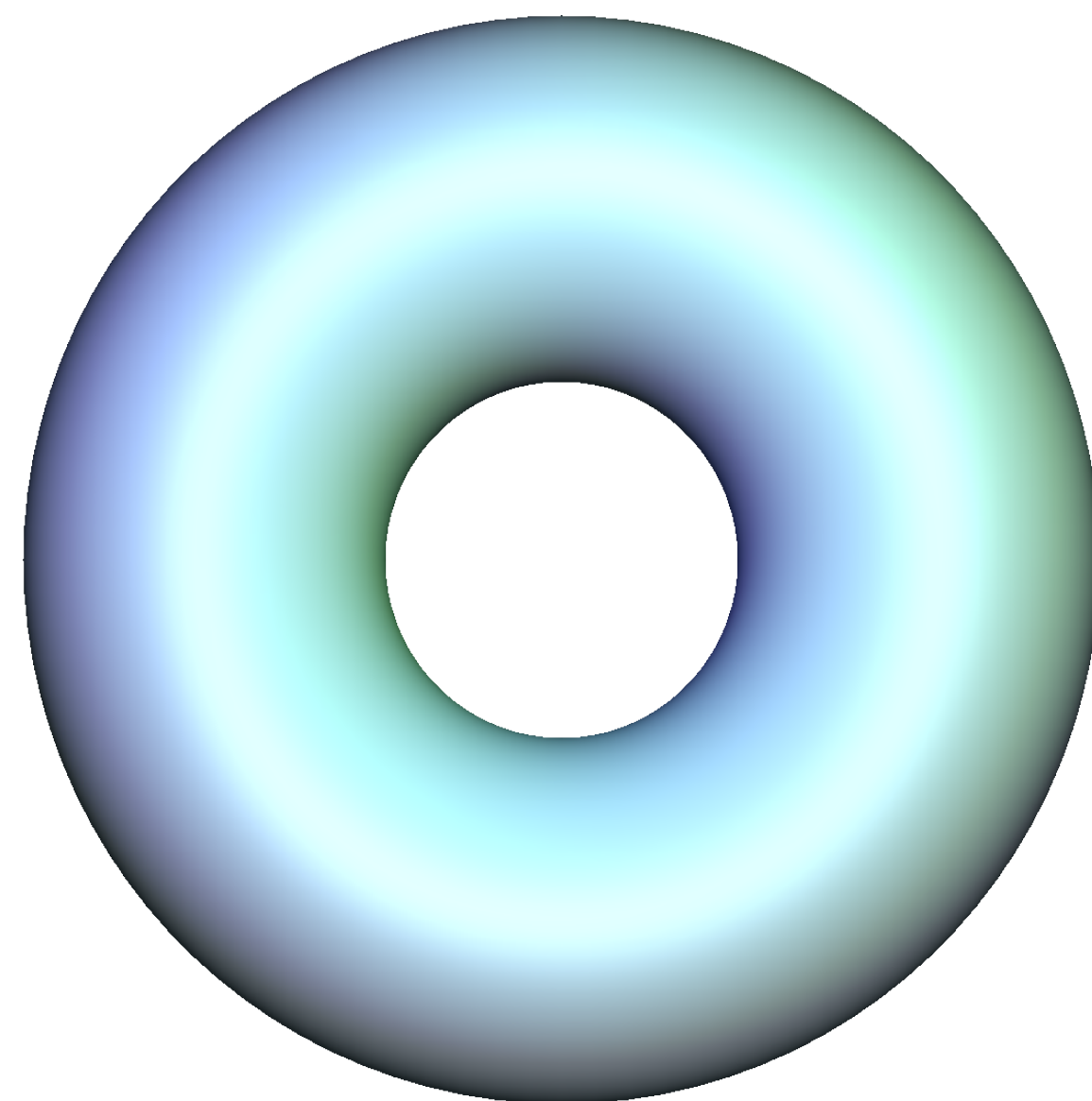
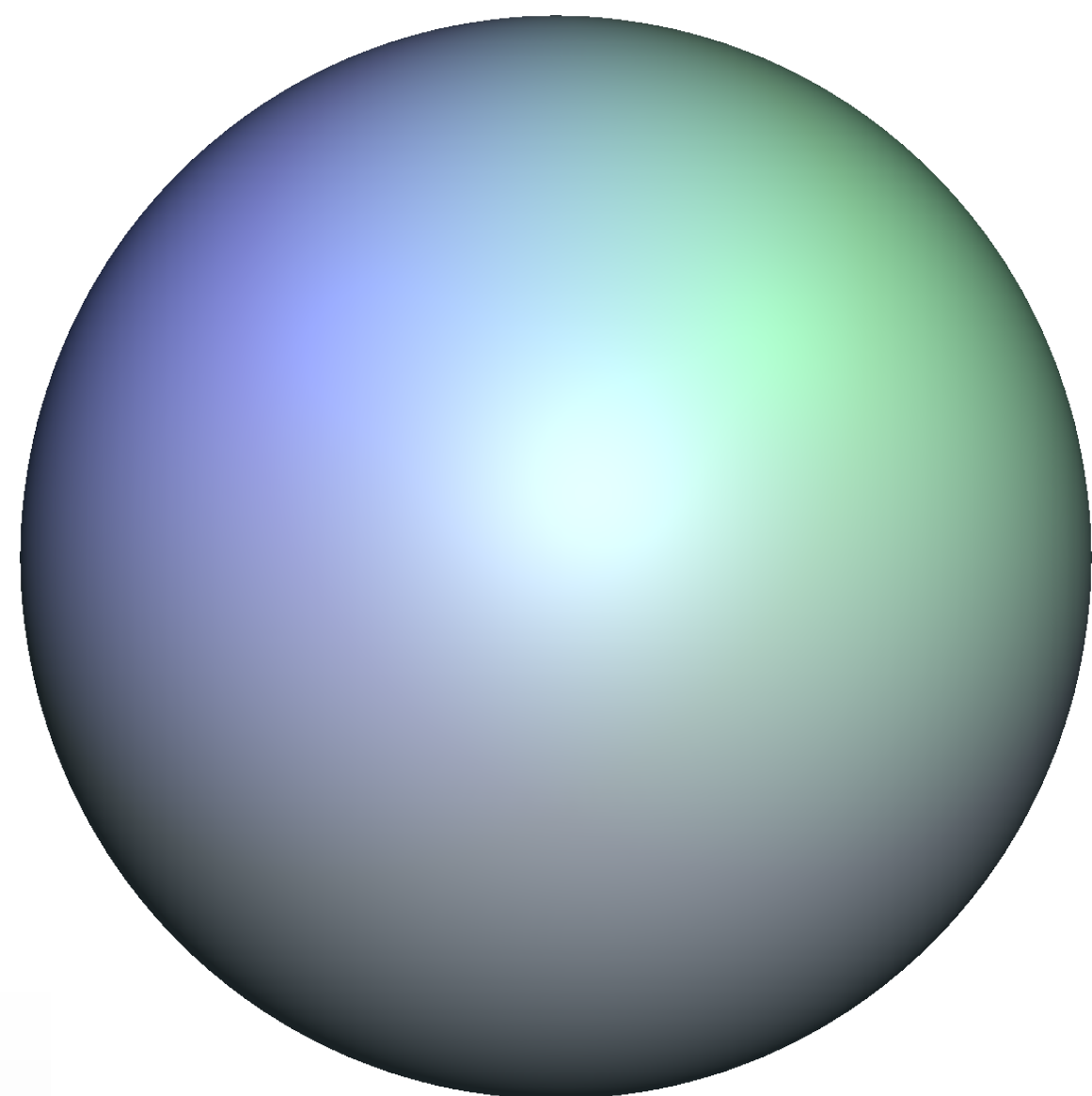
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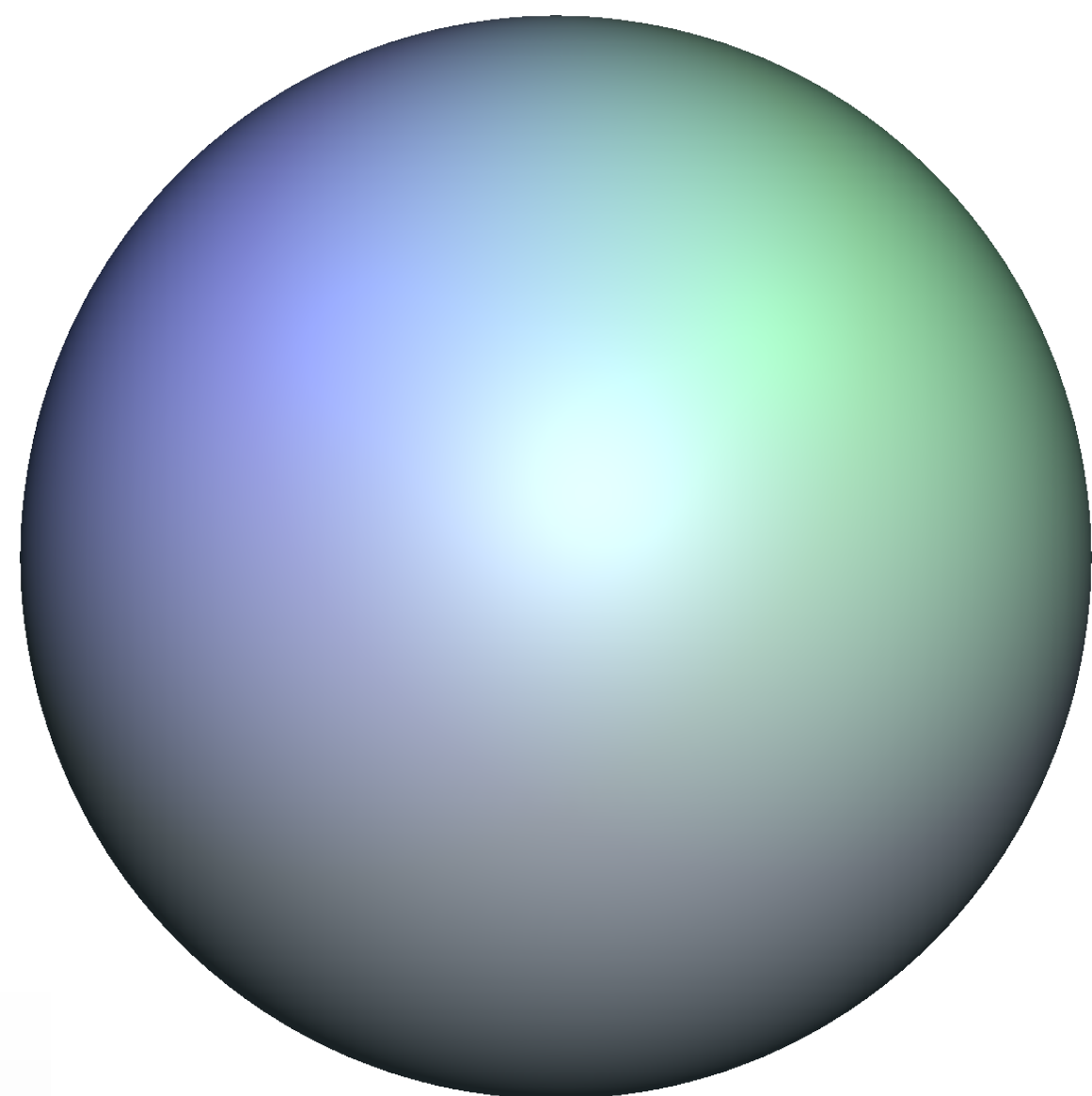
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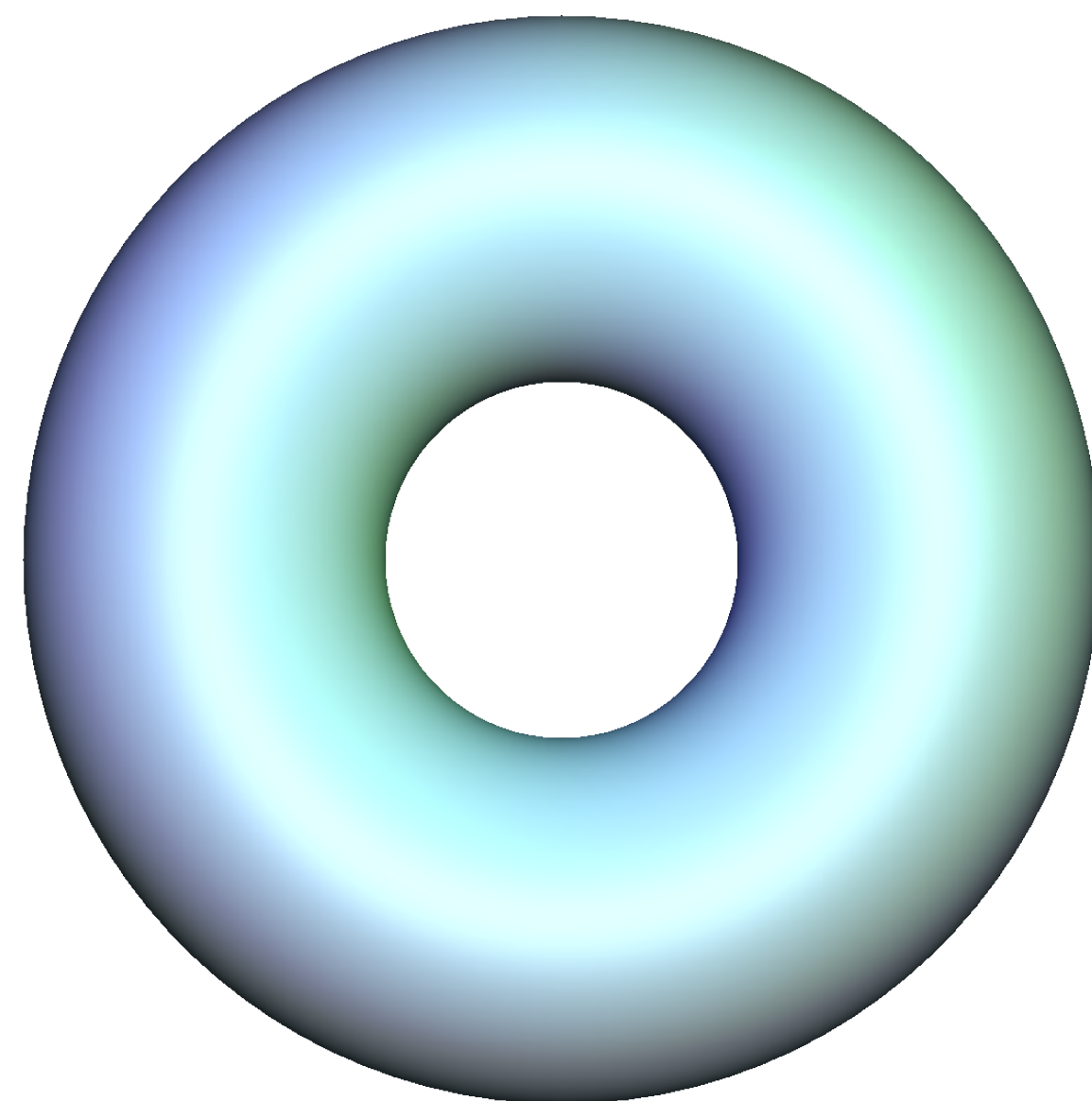
$$1, 0, 0, \chi = 1$$



$$1, 0, 1, \chi = 2$$



$$1, 2, 1, \chi = 0$$



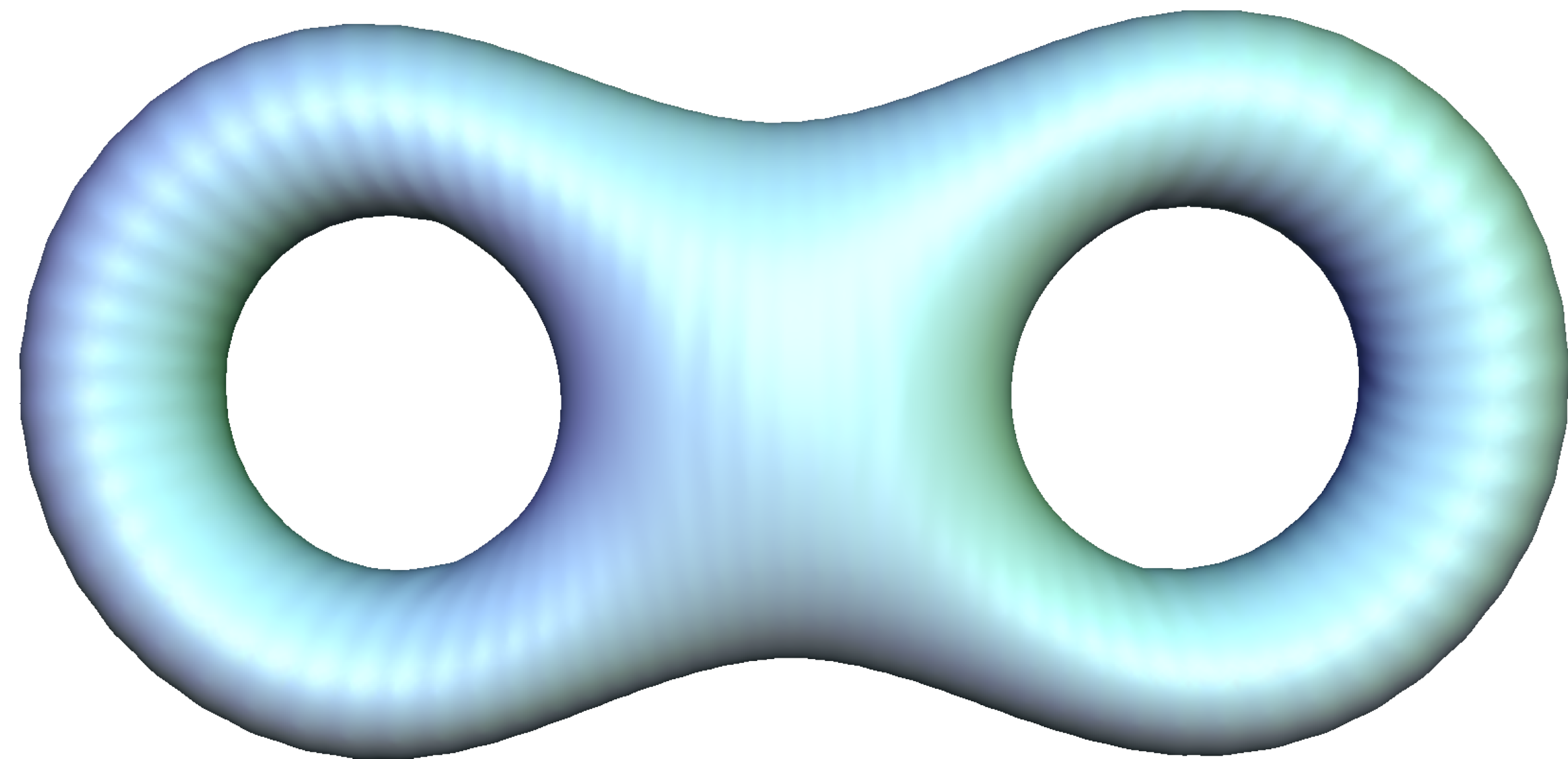
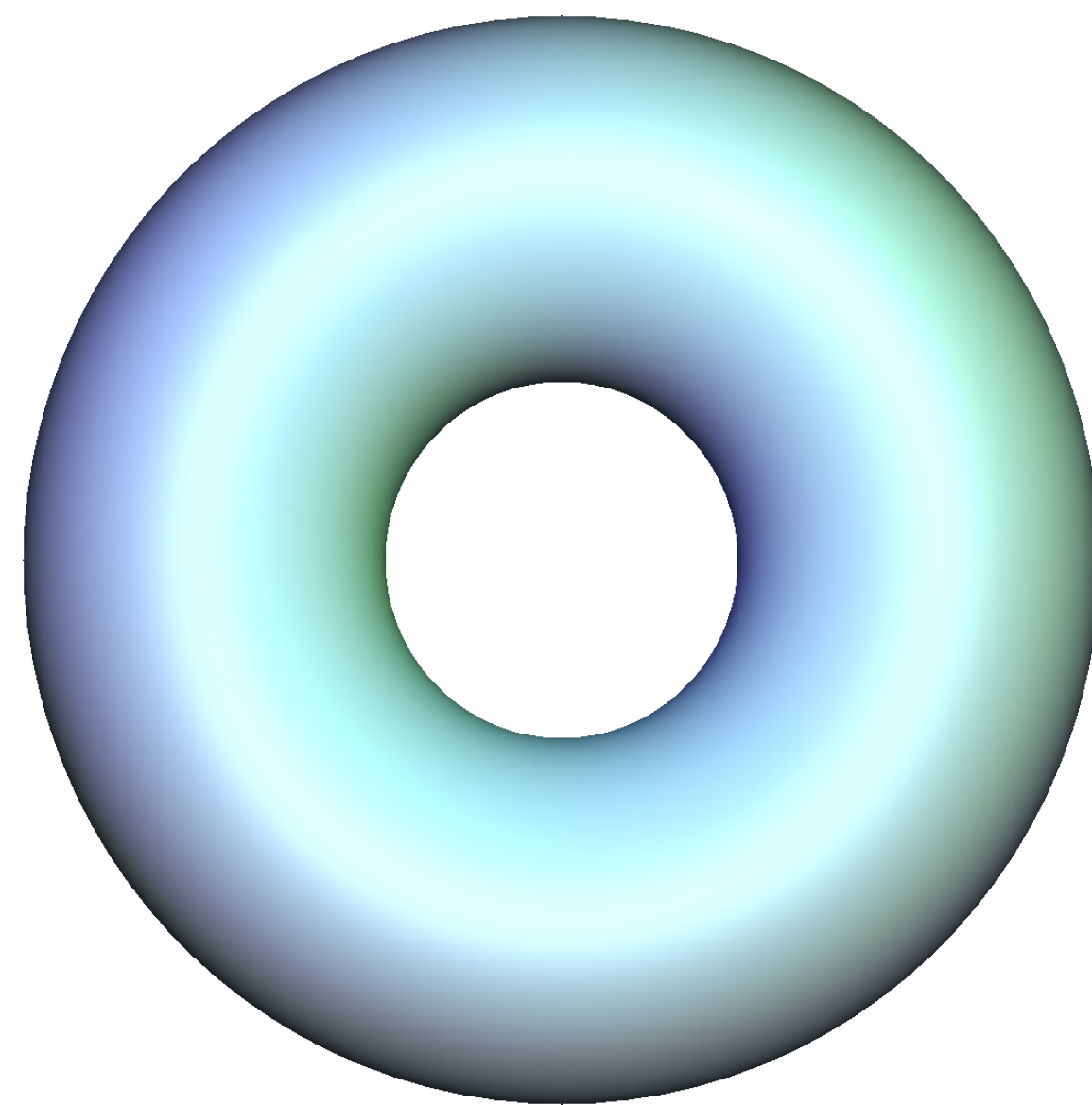
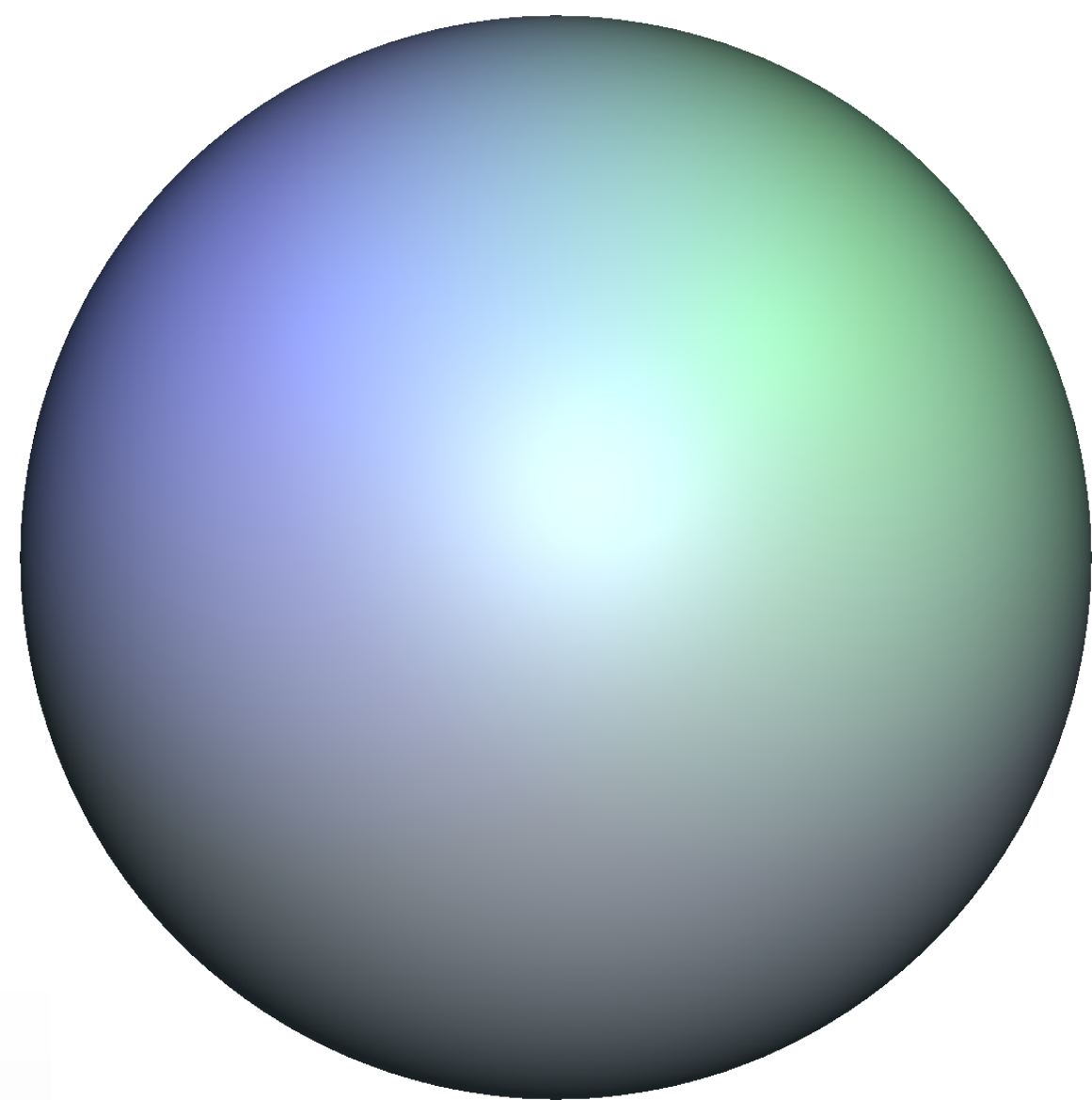
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$$\chi = \#_{vertices} - \#_{edges} + \#_{triangles} - \#_{tetrahedra} = \sum_{i=0}^d (-1)^i \beta_i$$

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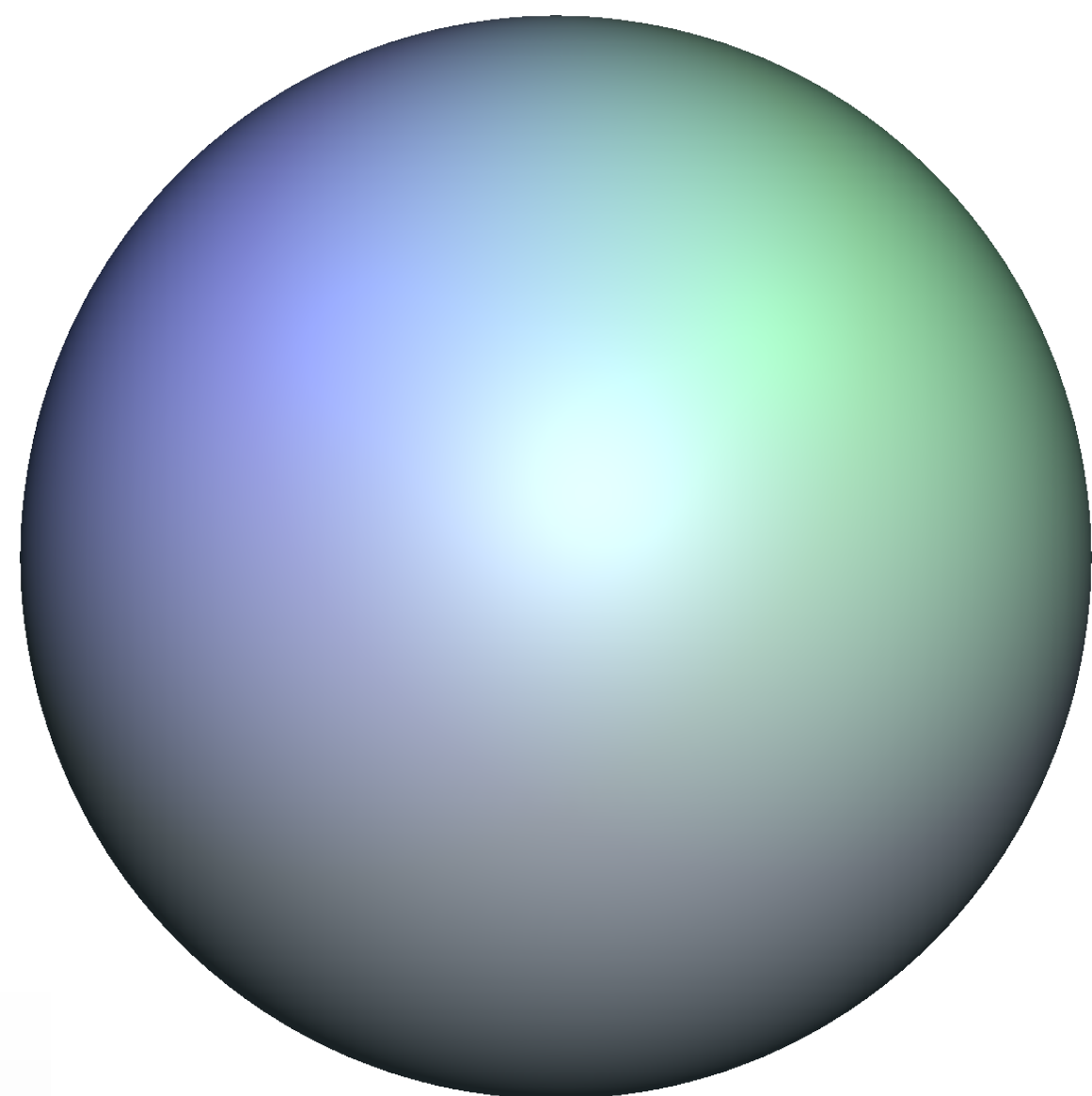
$$\chi = \#vertices - \#edges + \#triangles - \#tetrahedra = \sum_{i=0}^d (-1)^i \beta_i$$

$$\chi = \#connected_components - \#cycles + \#voids$$

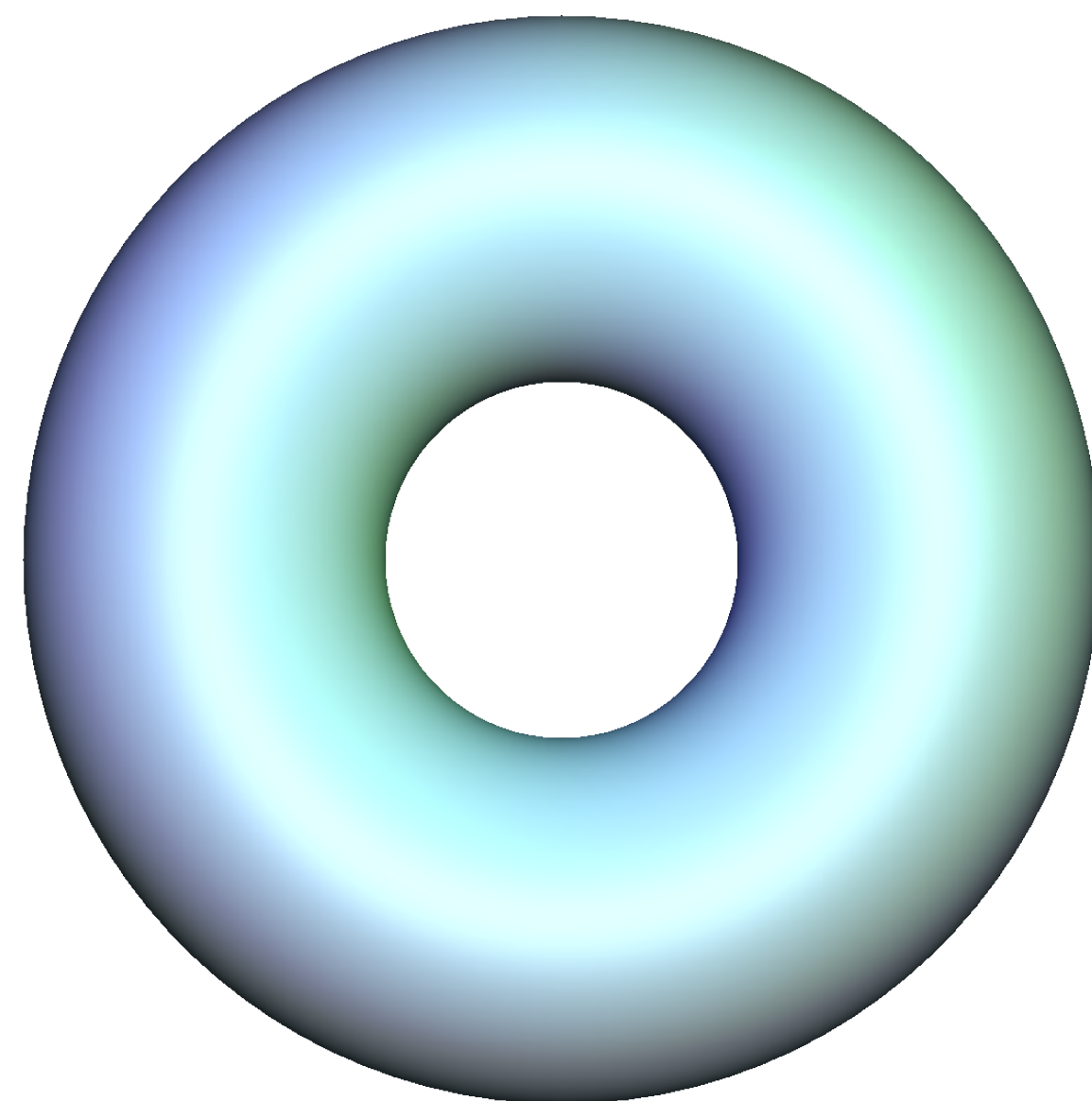
$$1, 0, 0, \chi = 1$$



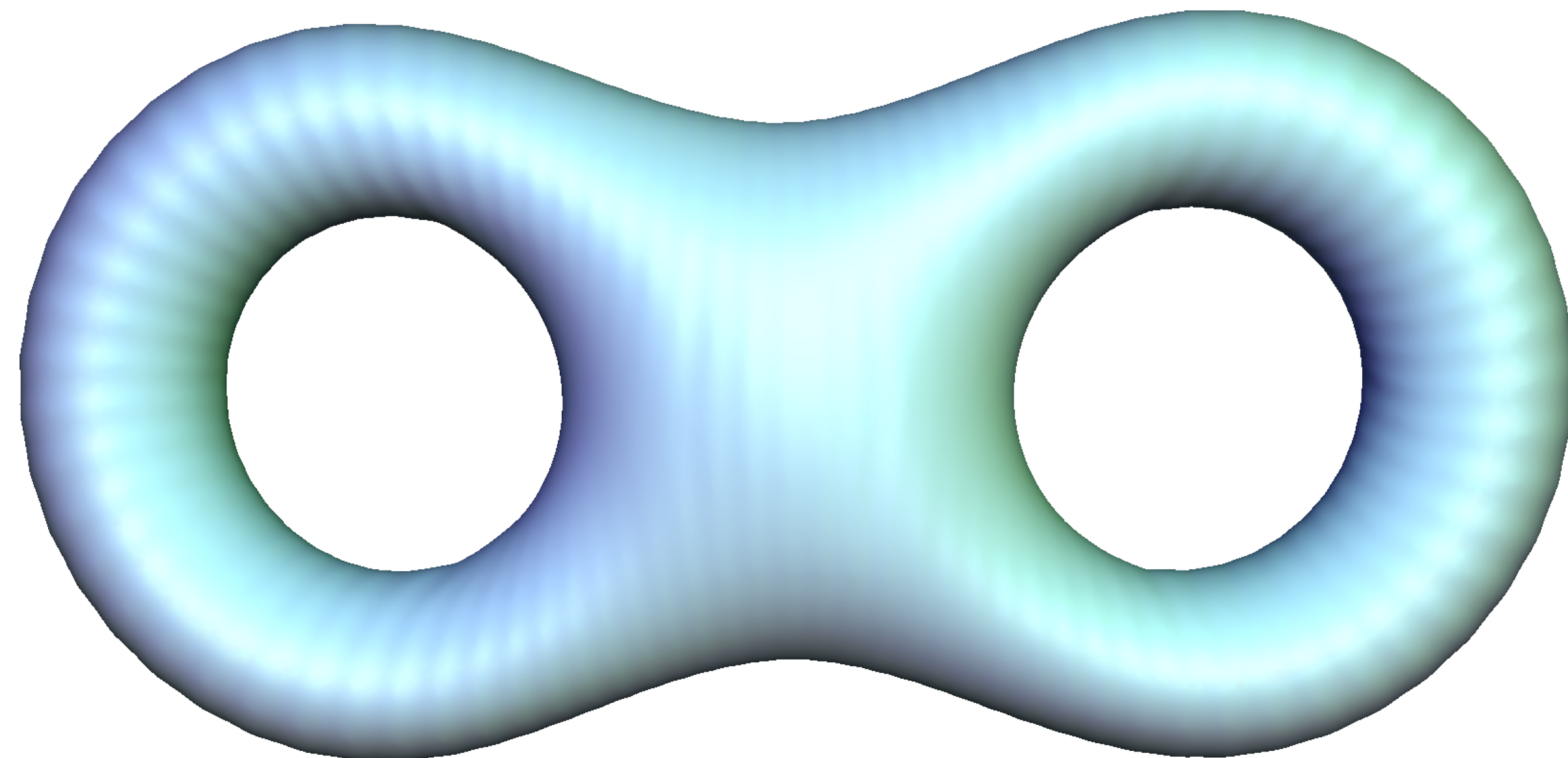
$$1, 0, 1, \chi = 2$$



$$1, 2, 1, \chi = 0$$

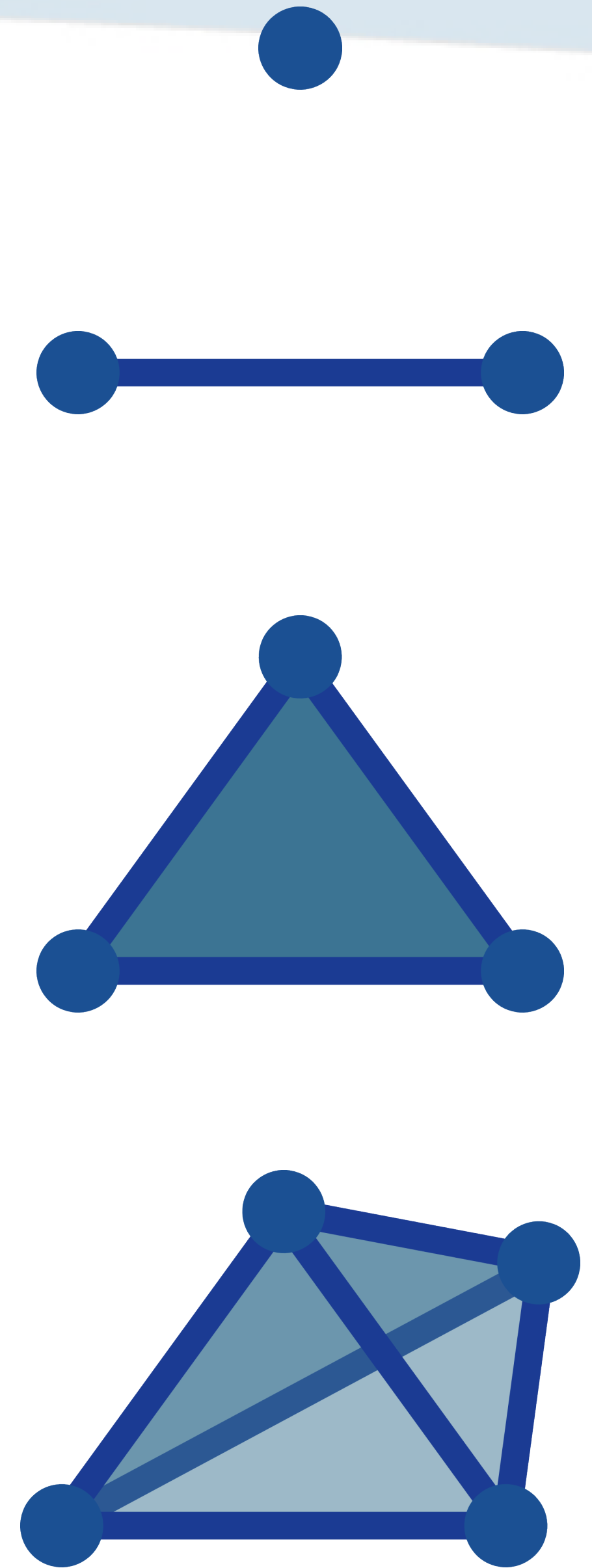


$$1, 4, 1, \chi = -2$$



Manifold spaces on a computer

- Notion of triangulation \mathcal{T}



Manifold spaces on a computer

- Notion of **triangulation** \mathcal{T}
 - Set of **vertices** (samples)
 - With embedding functions to \mathbb{R}^n



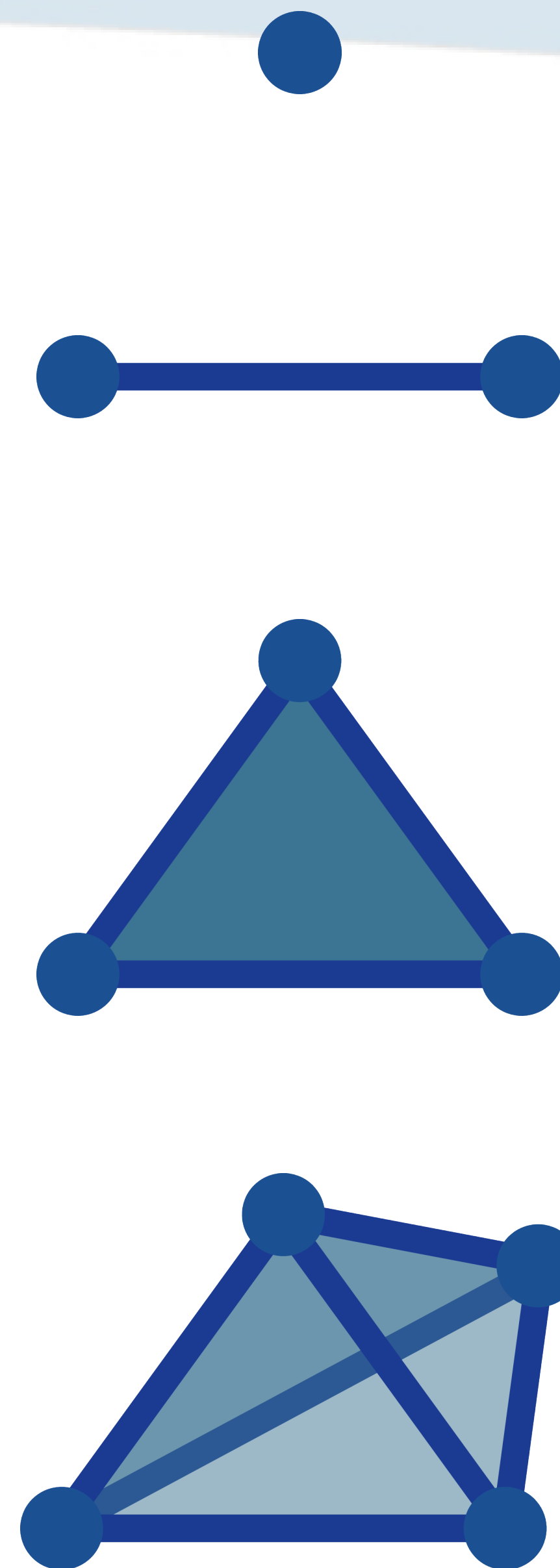
Manifold spaces on a computer

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 - **Connectivity**
 - **Explicitly** represented
 - Enumeration of simplices and their faces



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- What's missing?



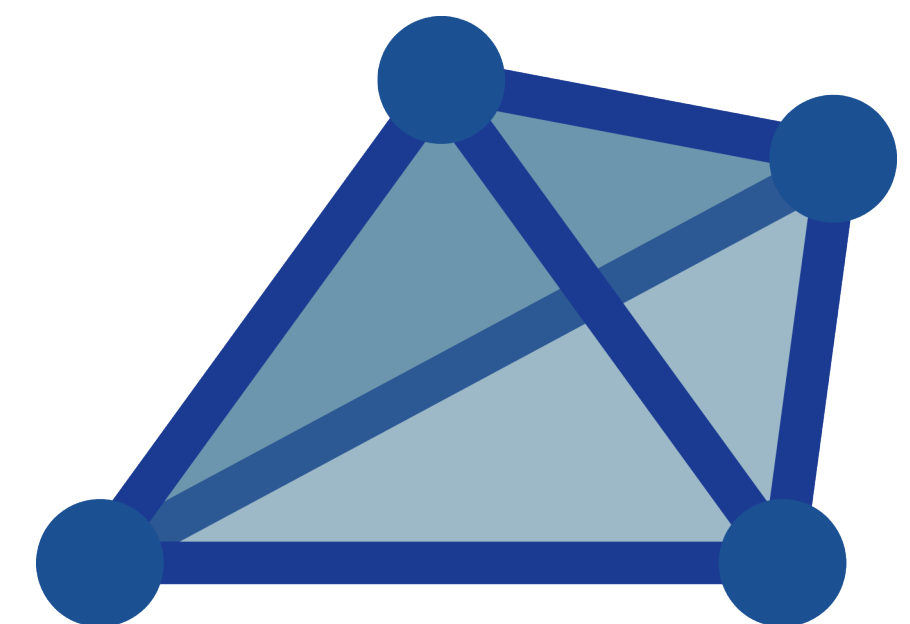
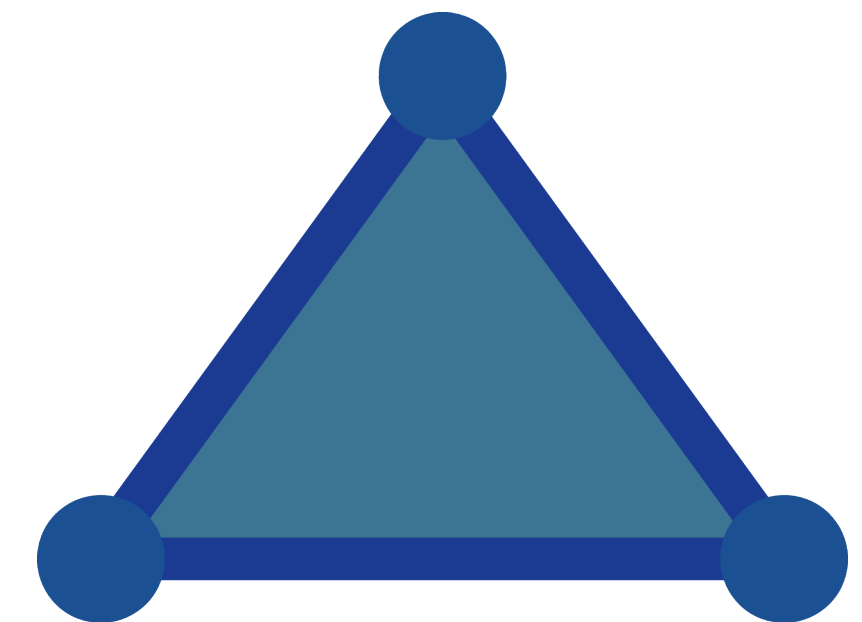
Manifold spaces on a computer

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- What's missing?
 - The **interpolation** scheme



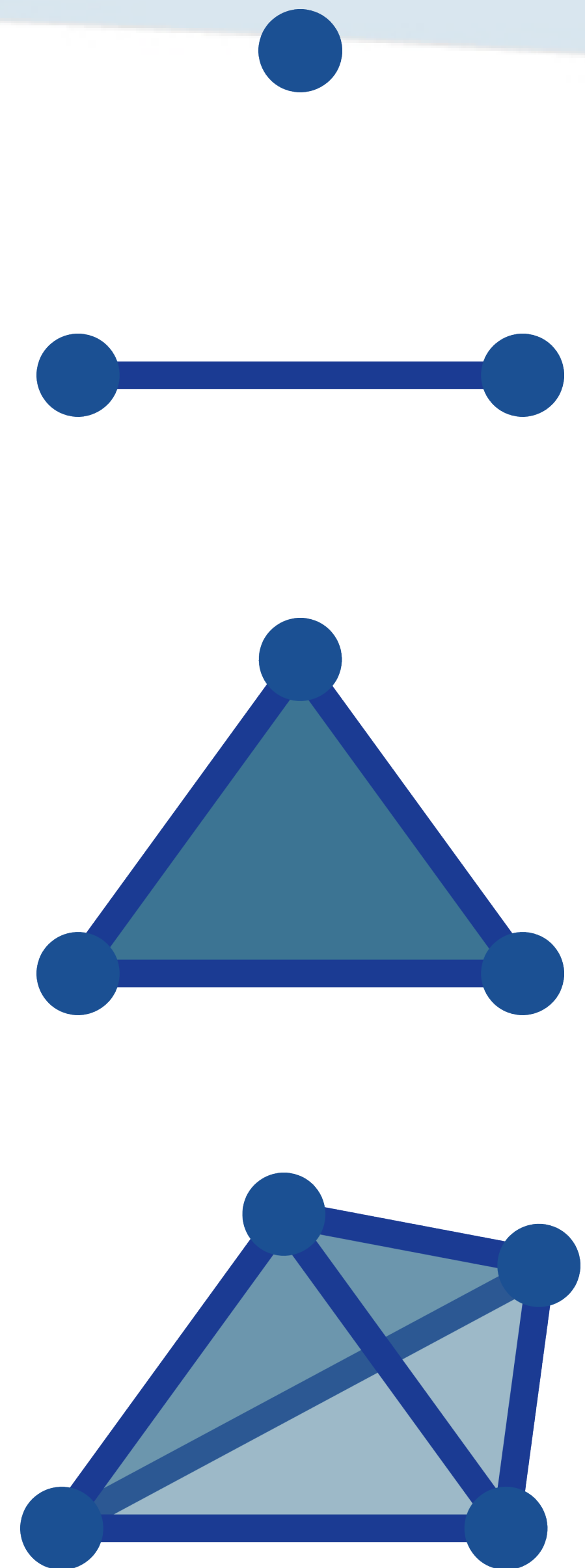
Interpolants for triangulations

- Like regular grids, several options



Interpolants for triangulations

- Like regular grids, several options
 - Piecewise constant
 - Piecewise linear
 - Piecewise polynomials,
 - etc.



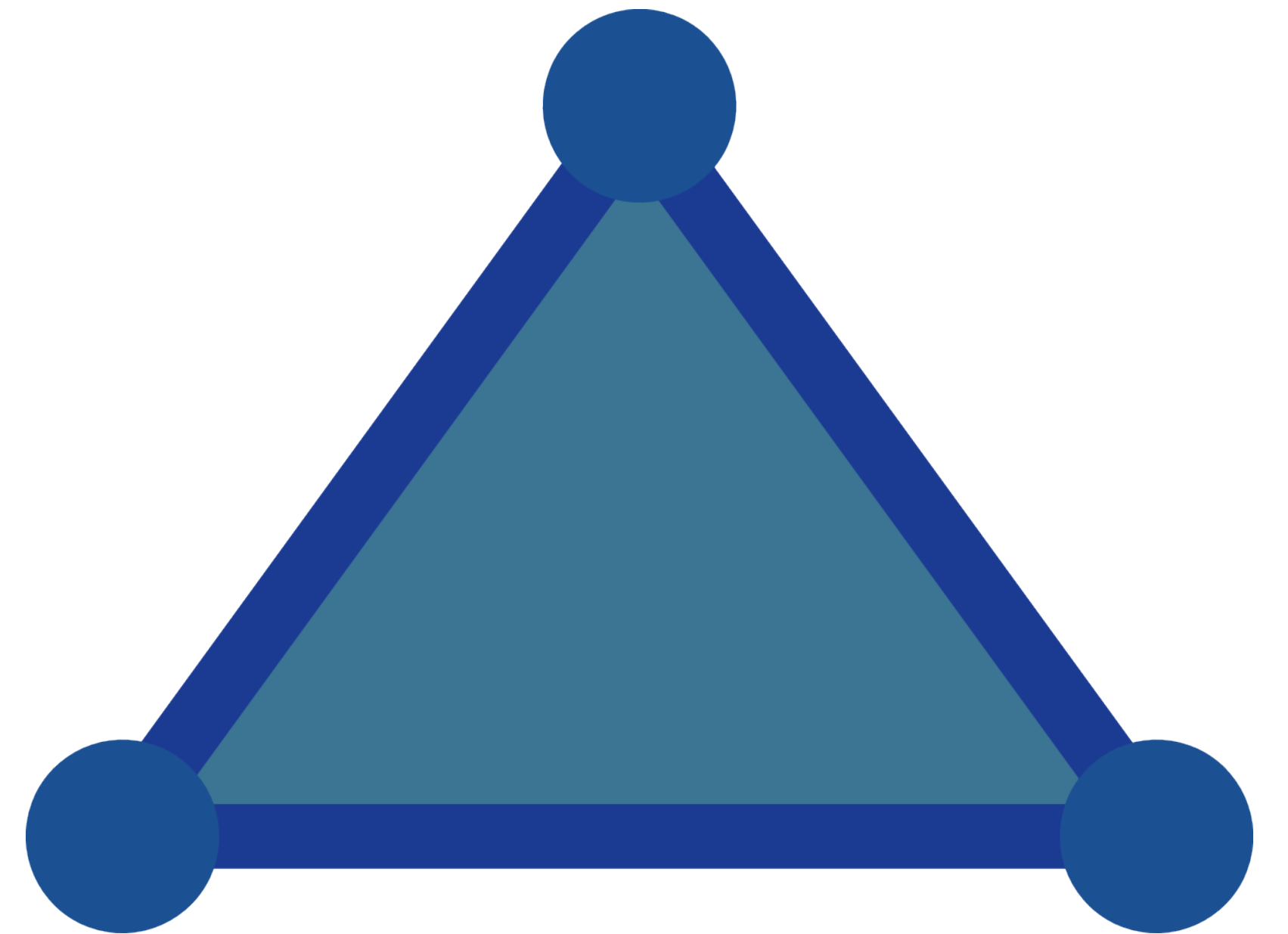
Interpolants for triangulations

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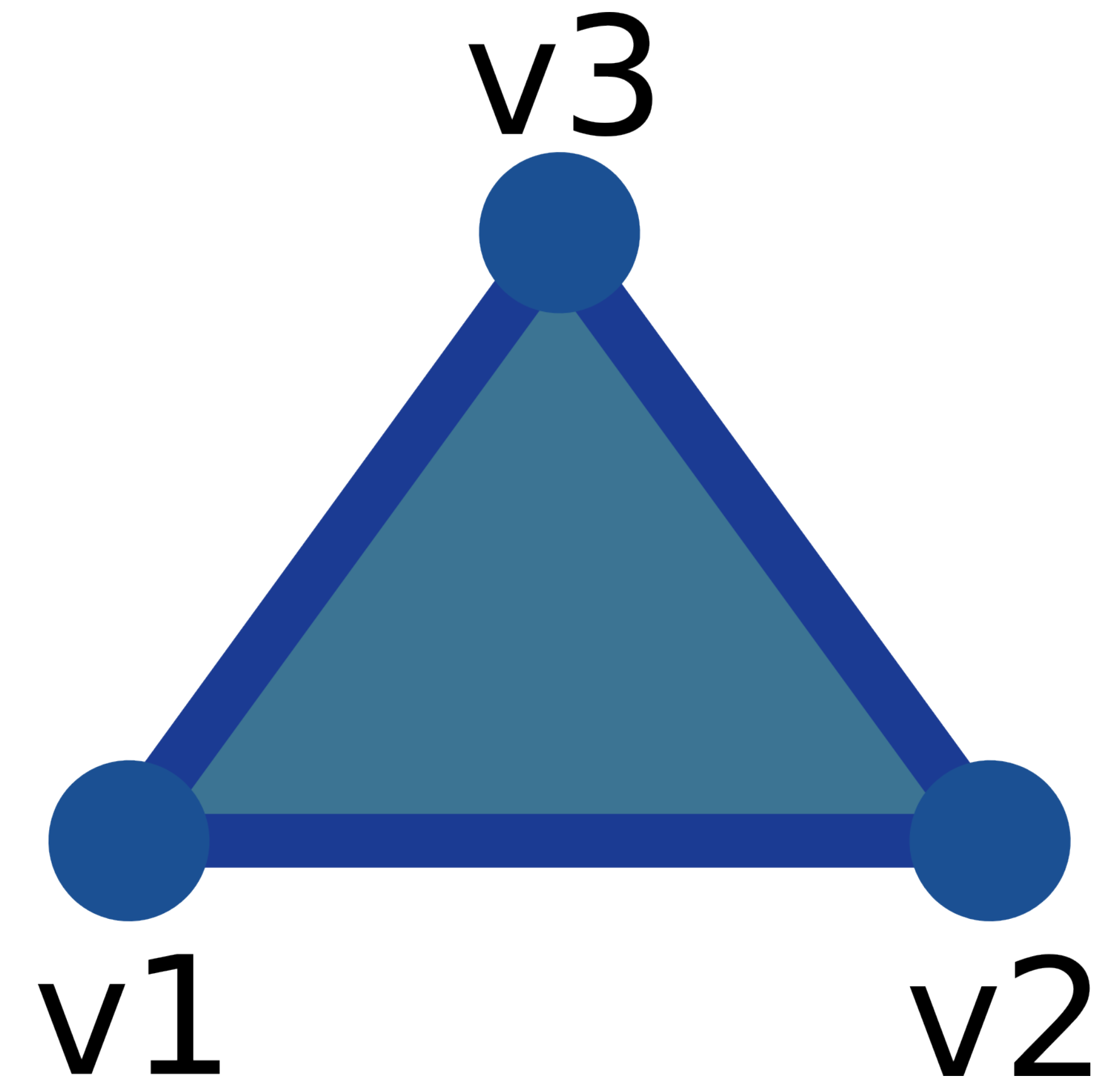
Piecewise linear interpolant on triangulations

- Interpolation as a boundary value problem



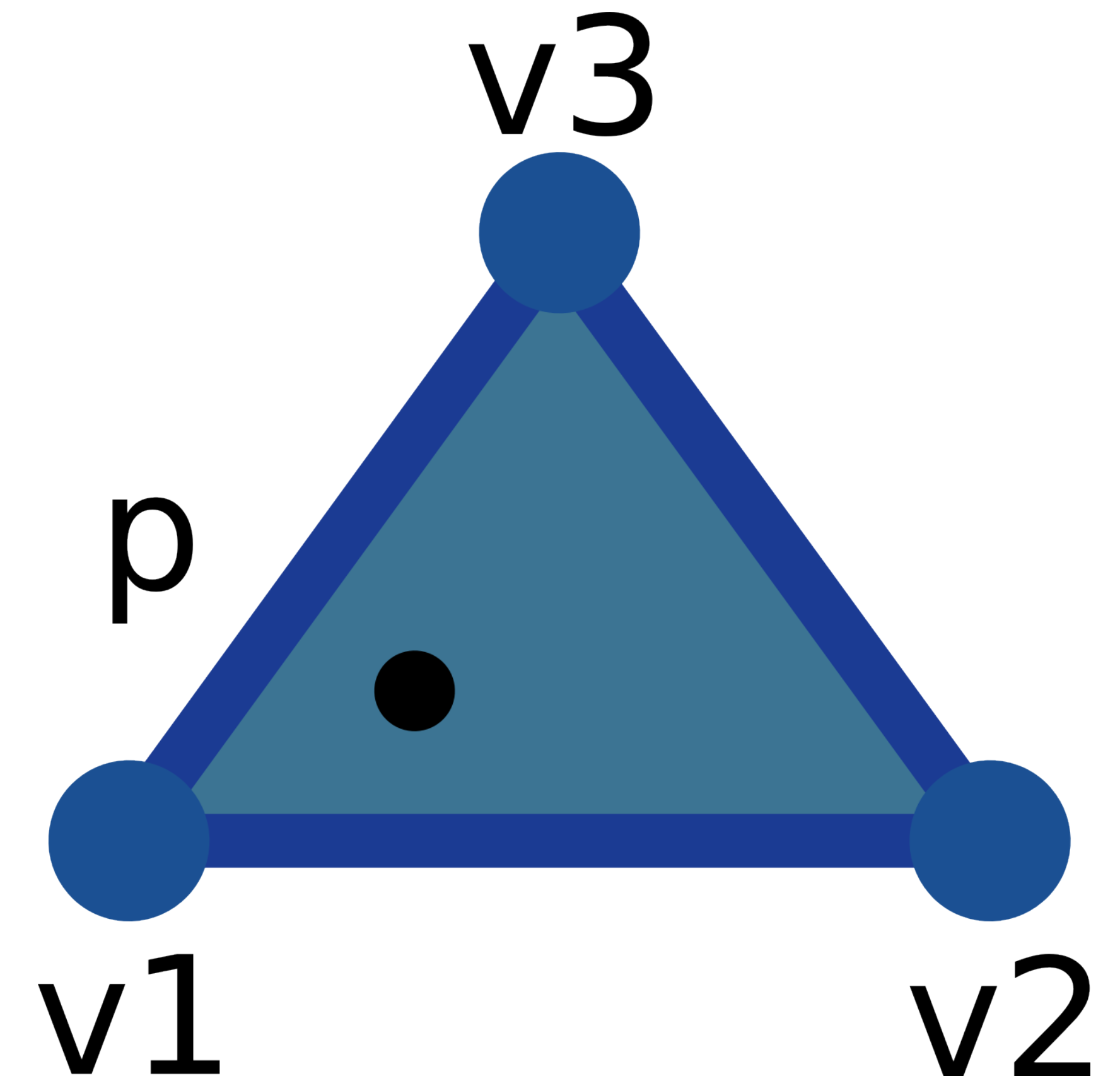
Piecewise linear interpolant on triangulations

- Interpolation as a boundary value problem
 - Given a simplex and its vertices



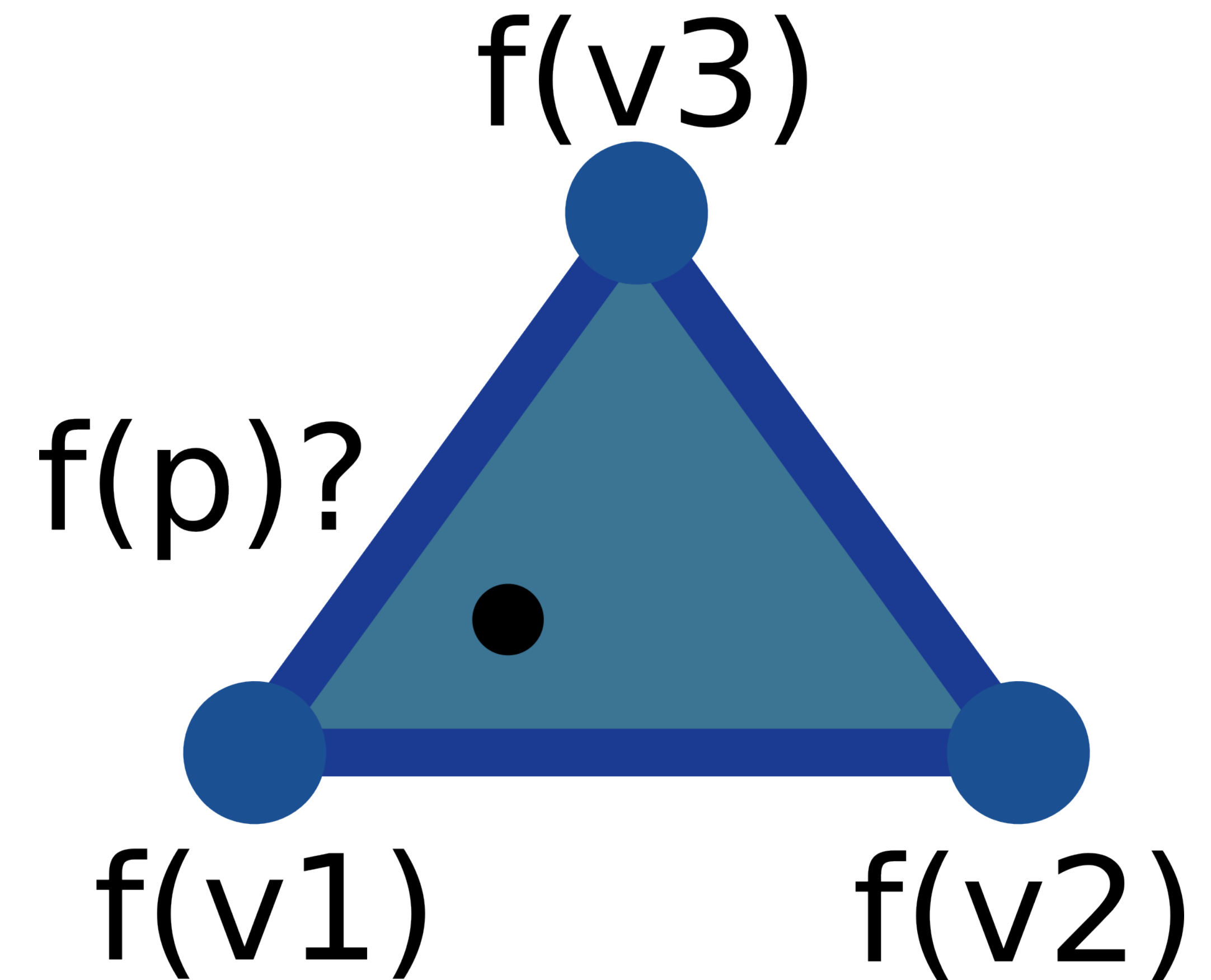
Piecewise linear interpolant on triangulations

- Interpolation as a boundary value problem
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 - Express the value of any point



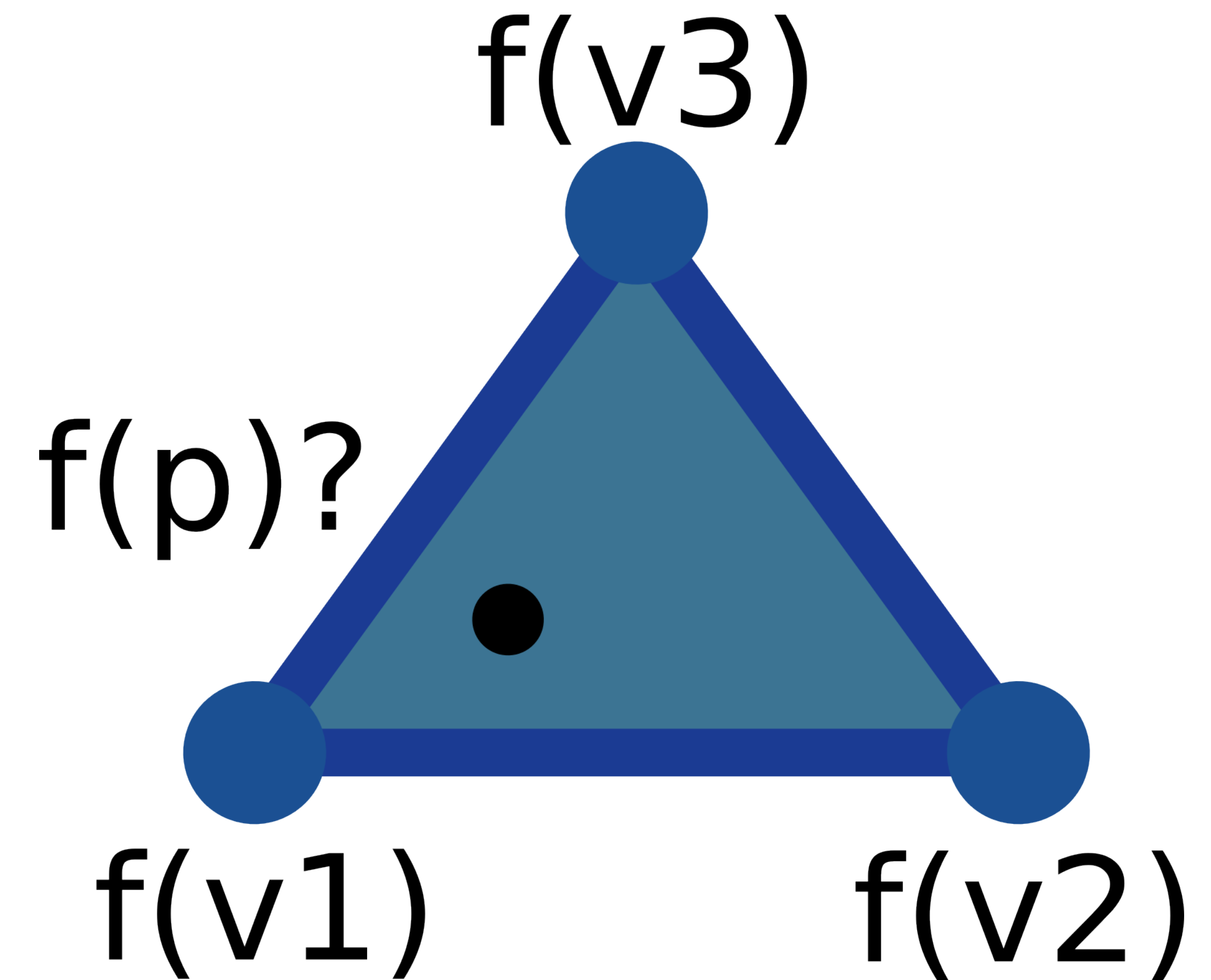
Piecewise linear interpolant on triangulations

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 - As a linear combination of those of the vertices



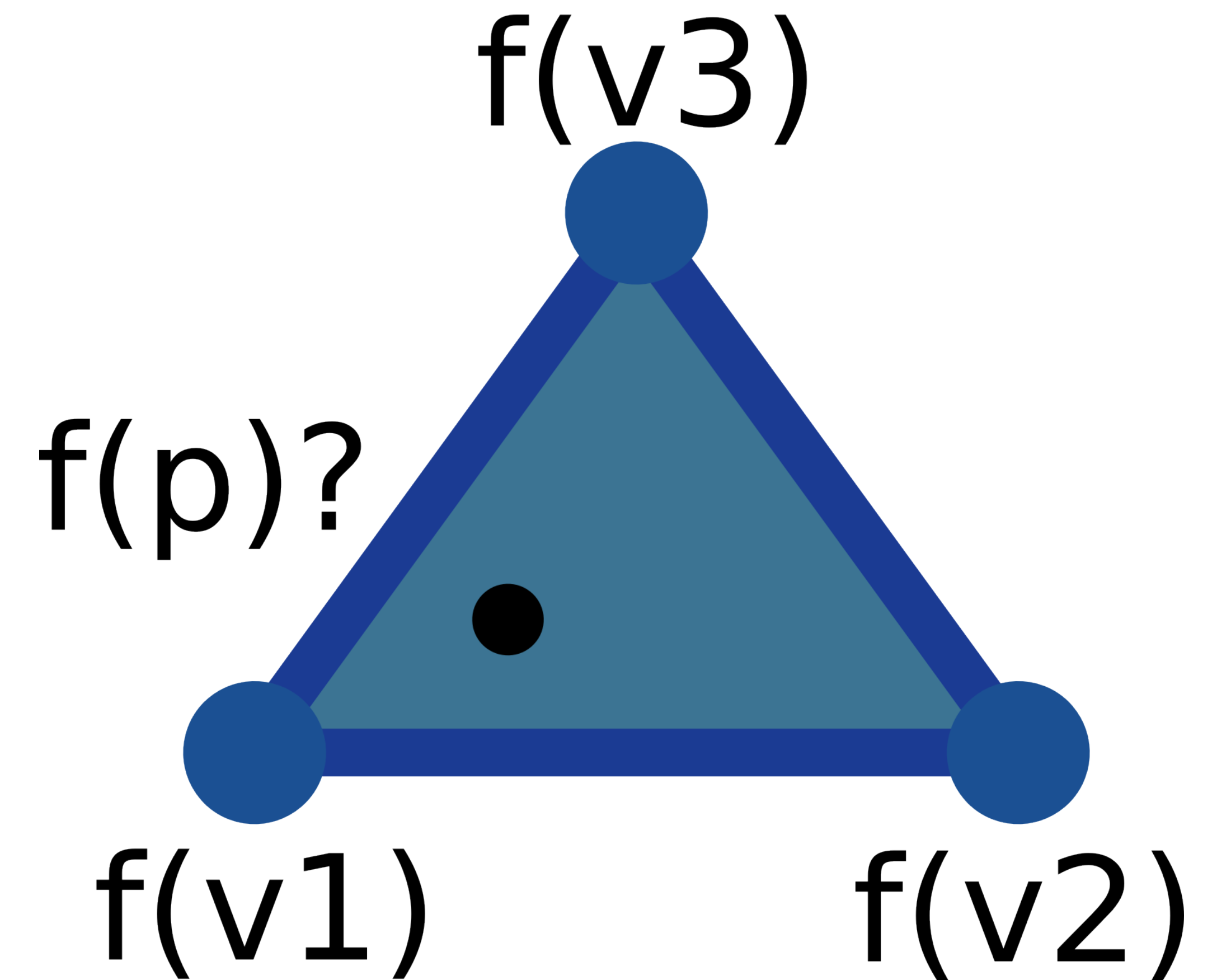
Piecewise linear interpolant on triangulations

- Interpolation as a boundary value problem
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- Notion of barycentric coordinates:
 - $f(p) = \alpha_1 f(v_1) + \alpha_2 f(v_2) + \alpha_3 f(v_3)$



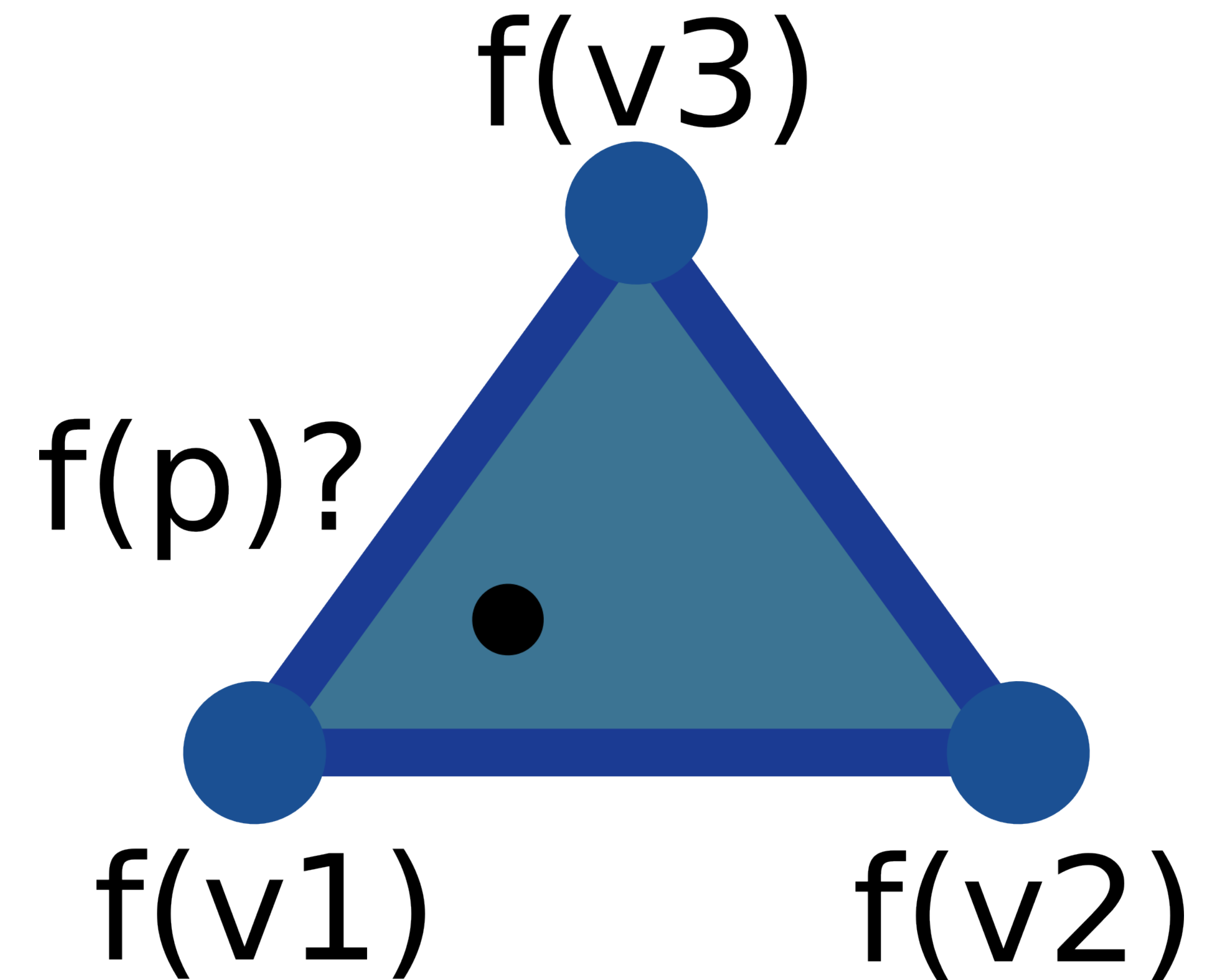
Barycentric coordinates

- There exists many forms
 - With specific properties



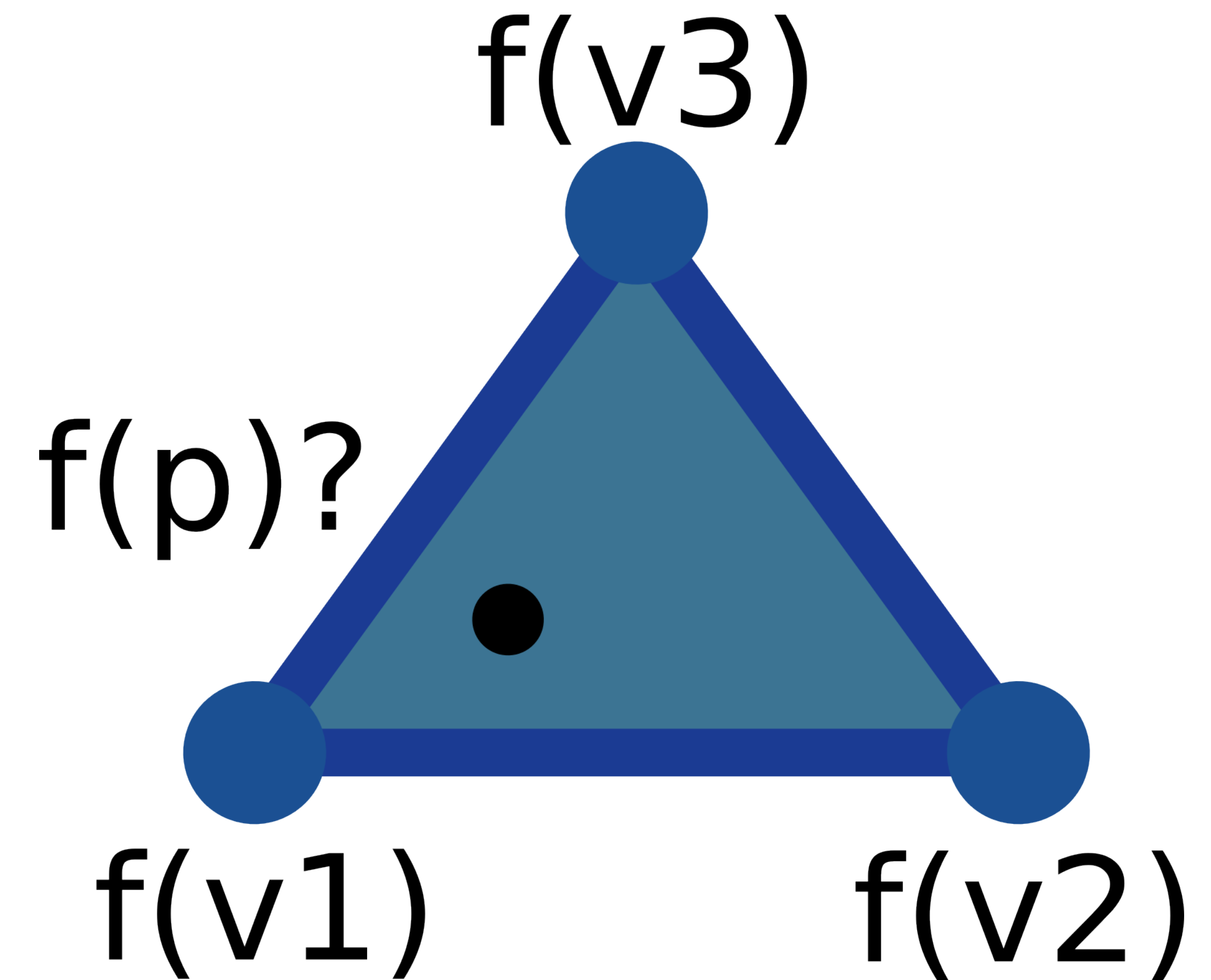
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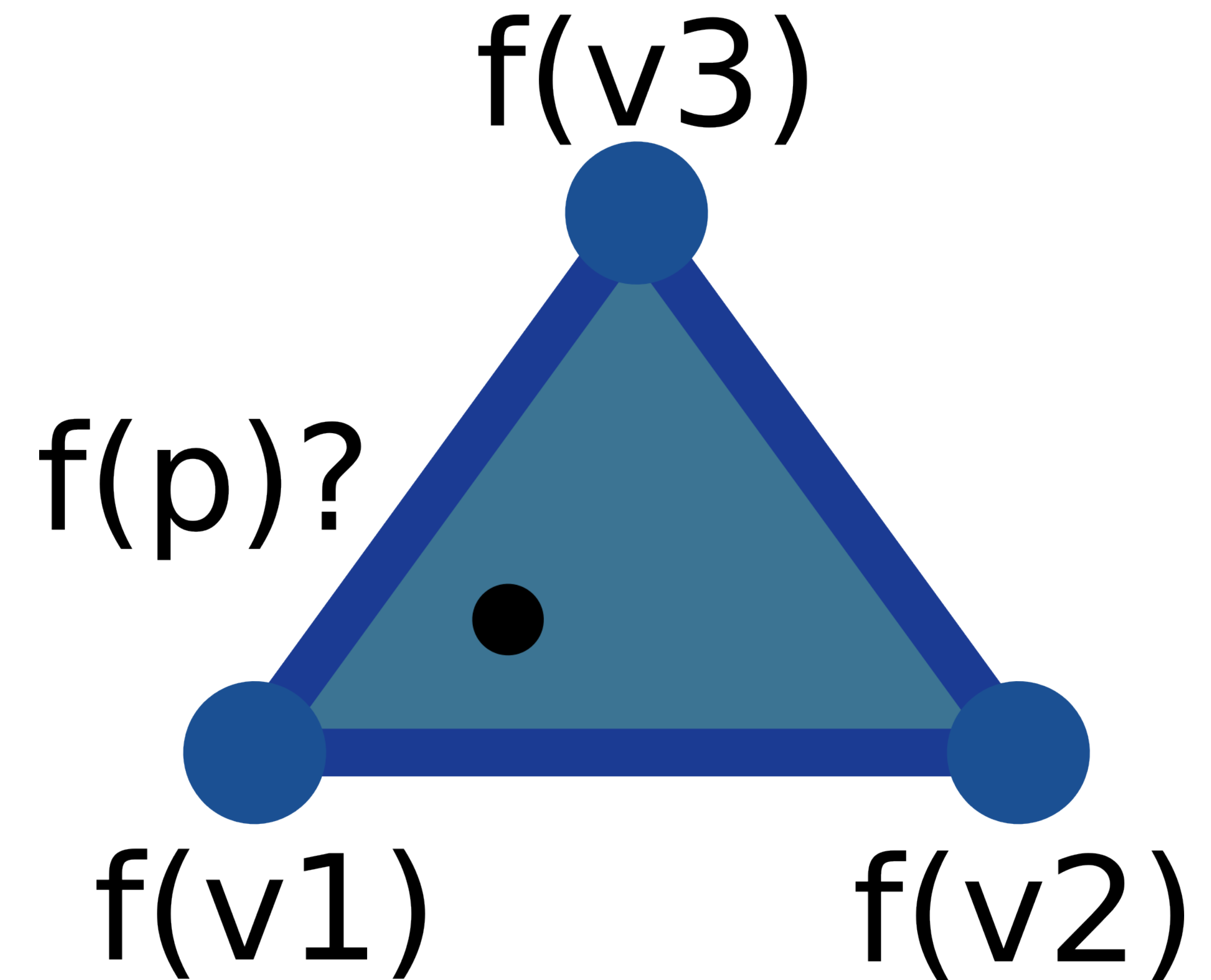
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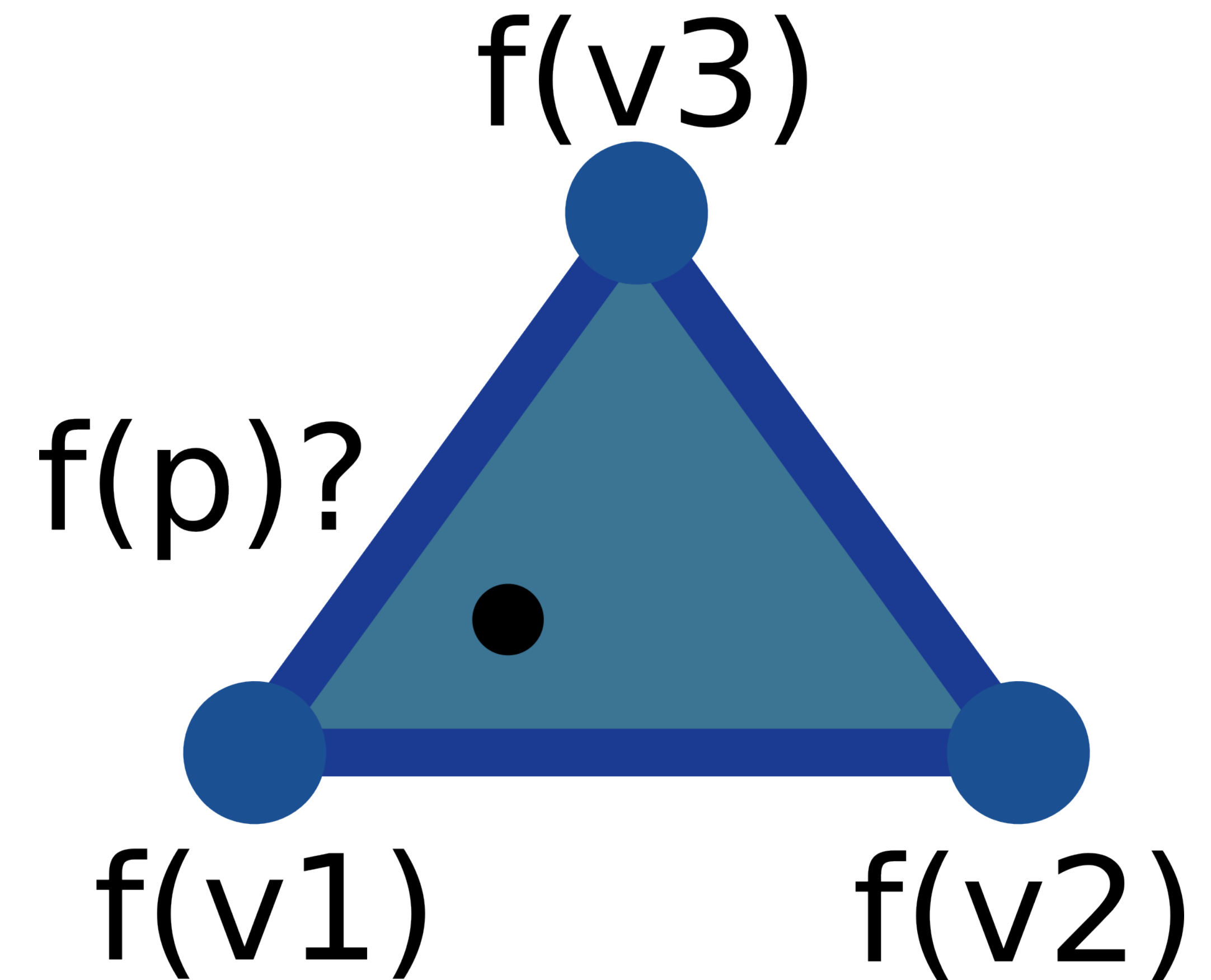
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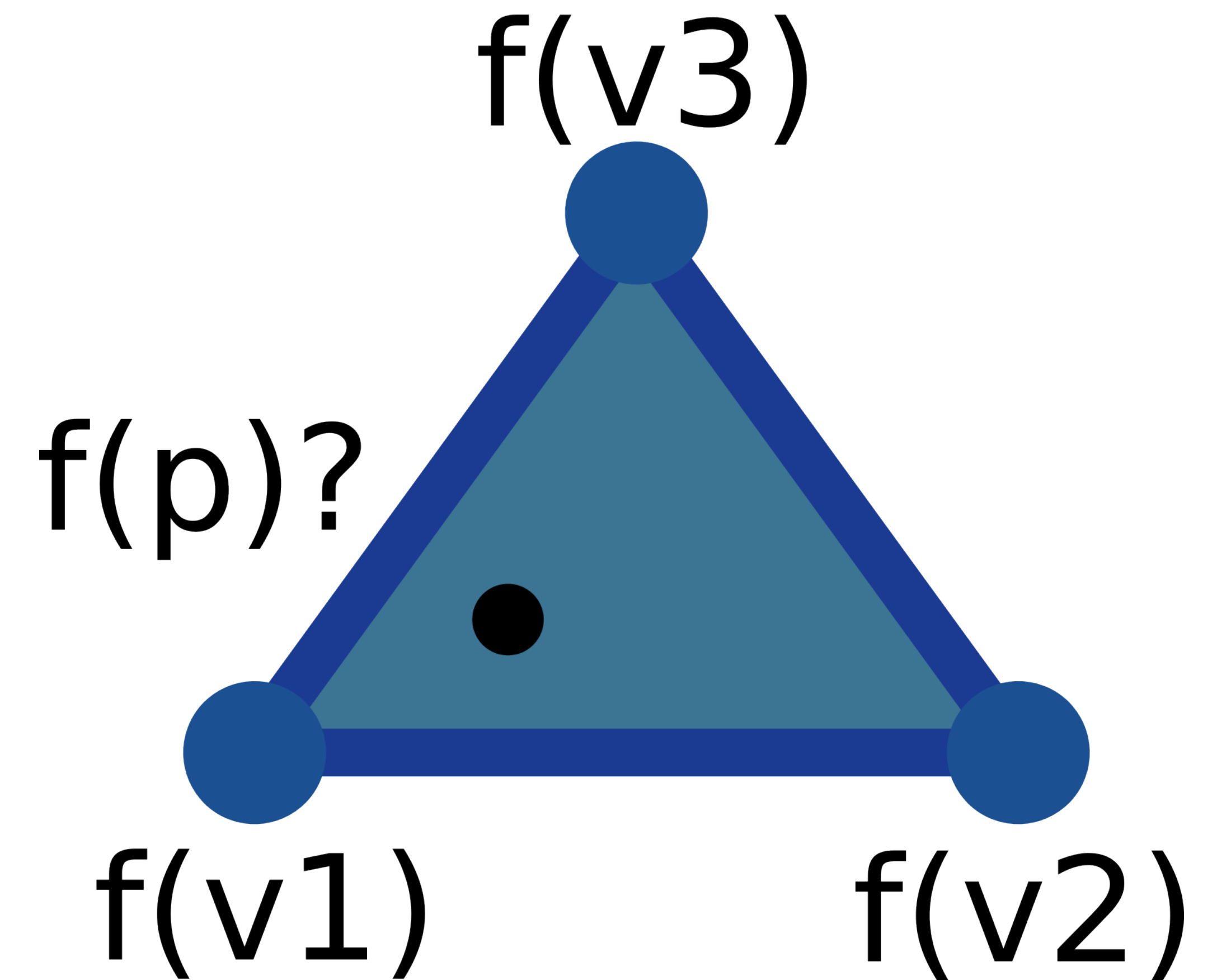
Barycentric coordinates

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- Notion of barycentric coordinates:
 - $f(p) = \alpha_1 f(v_1) + \alpha_2 f(v_2) + \alpha_3 f(v_3)$
 - Must hold for any $f : \mathcal{T} \rightarrow \mathbb{R}$



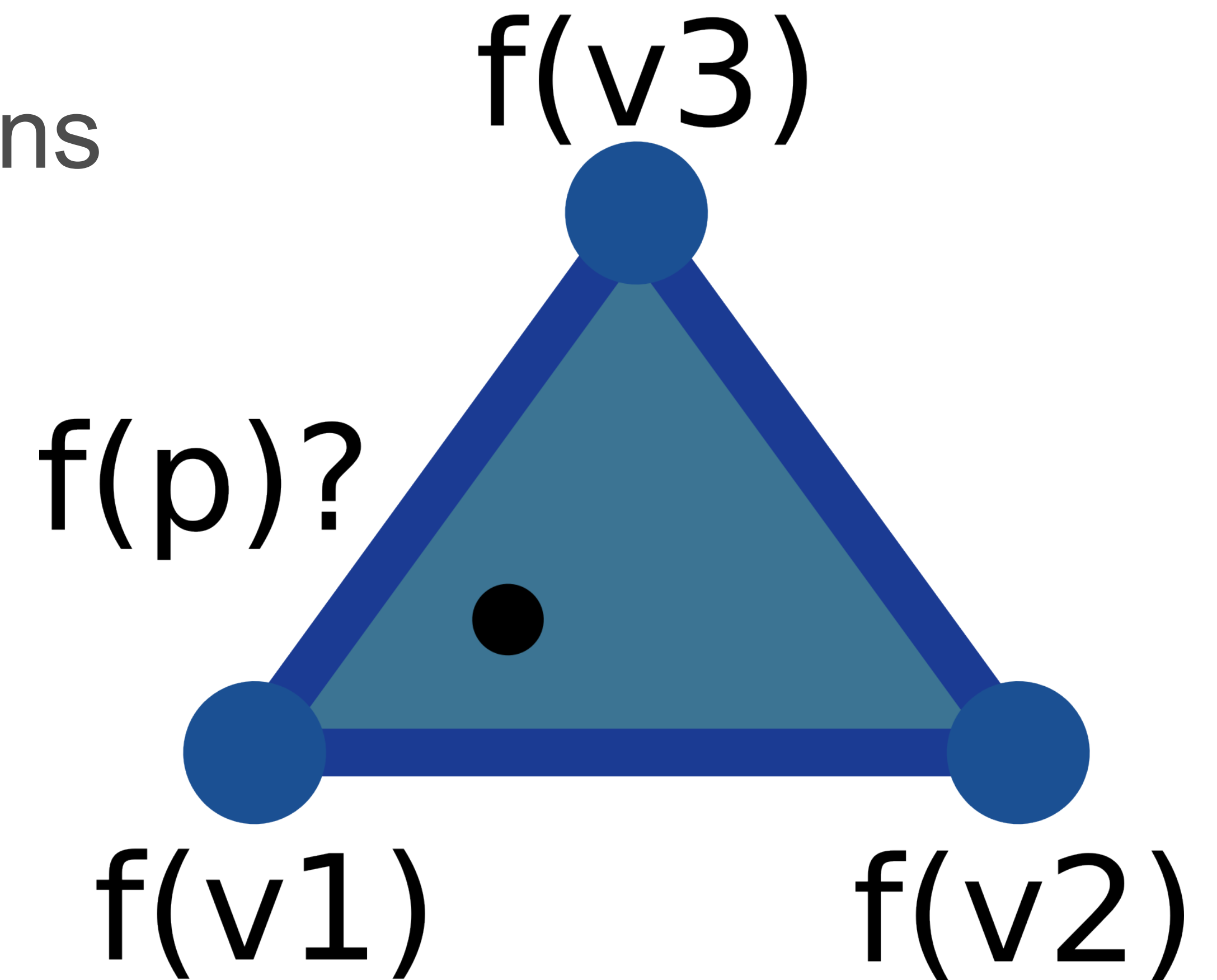
Barycentric coordinates

- In particular



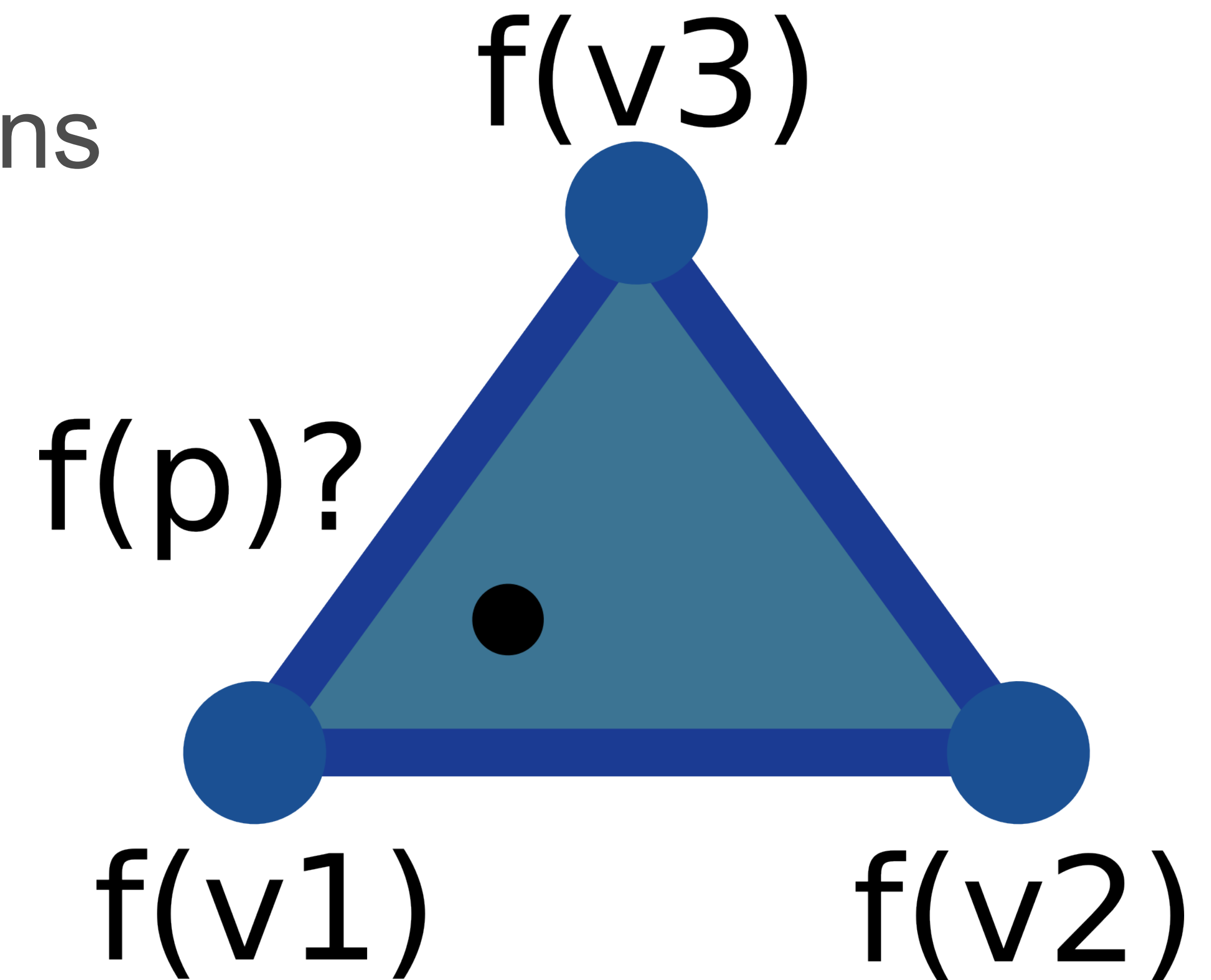
Barycentric coordinates

- In particular
 - It must also hold for the embedding functions



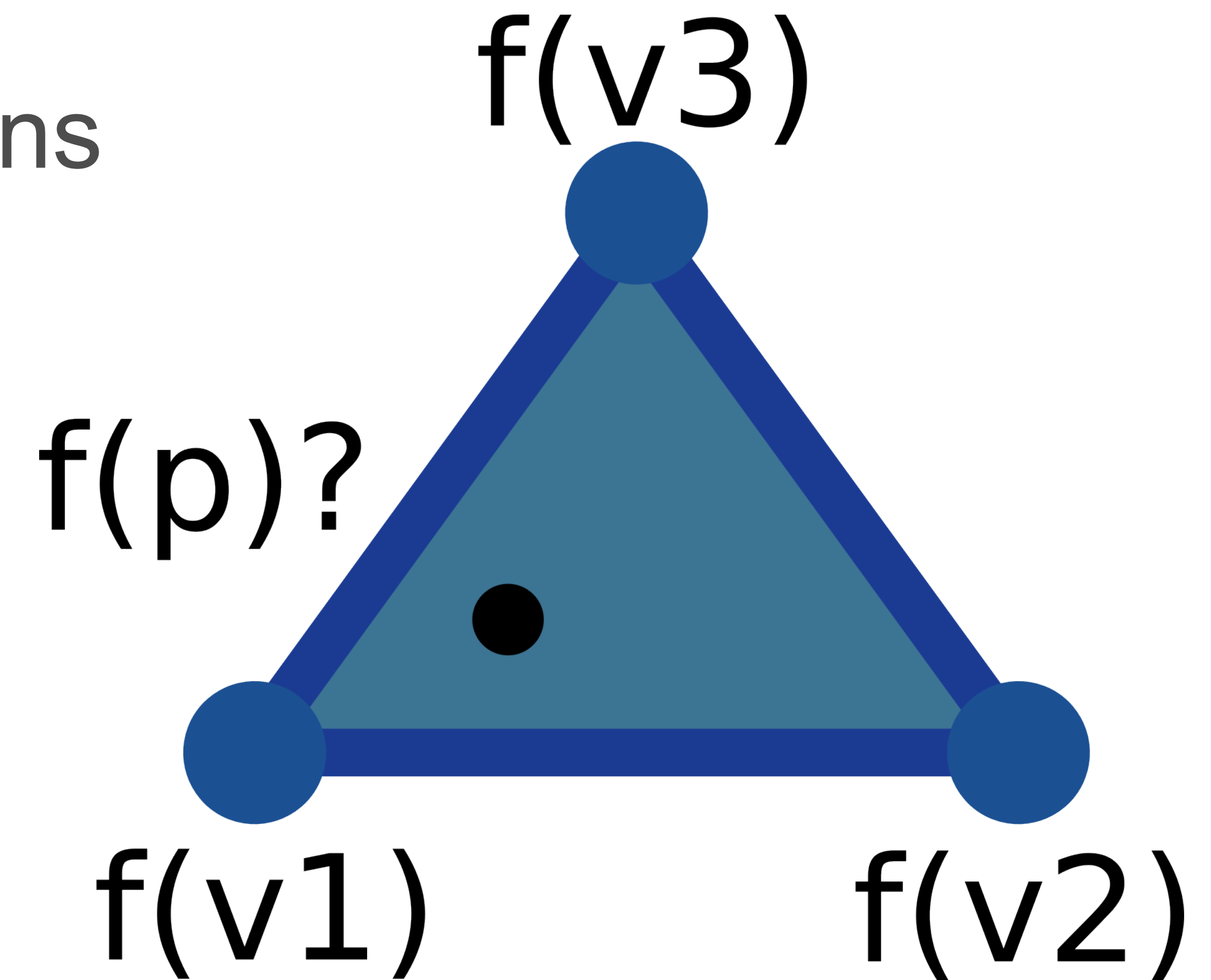
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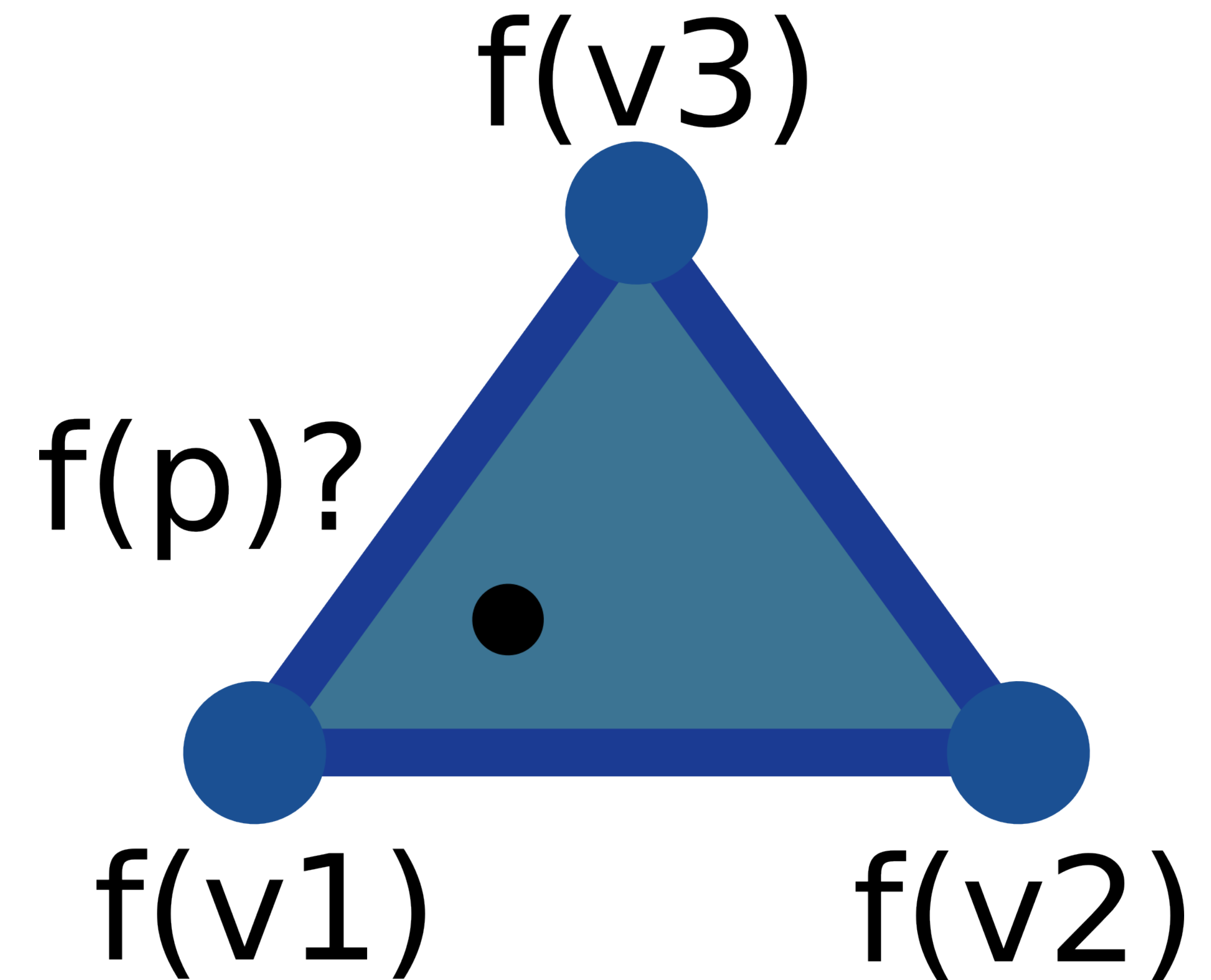
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- 3 linear equations with 3 unknowns



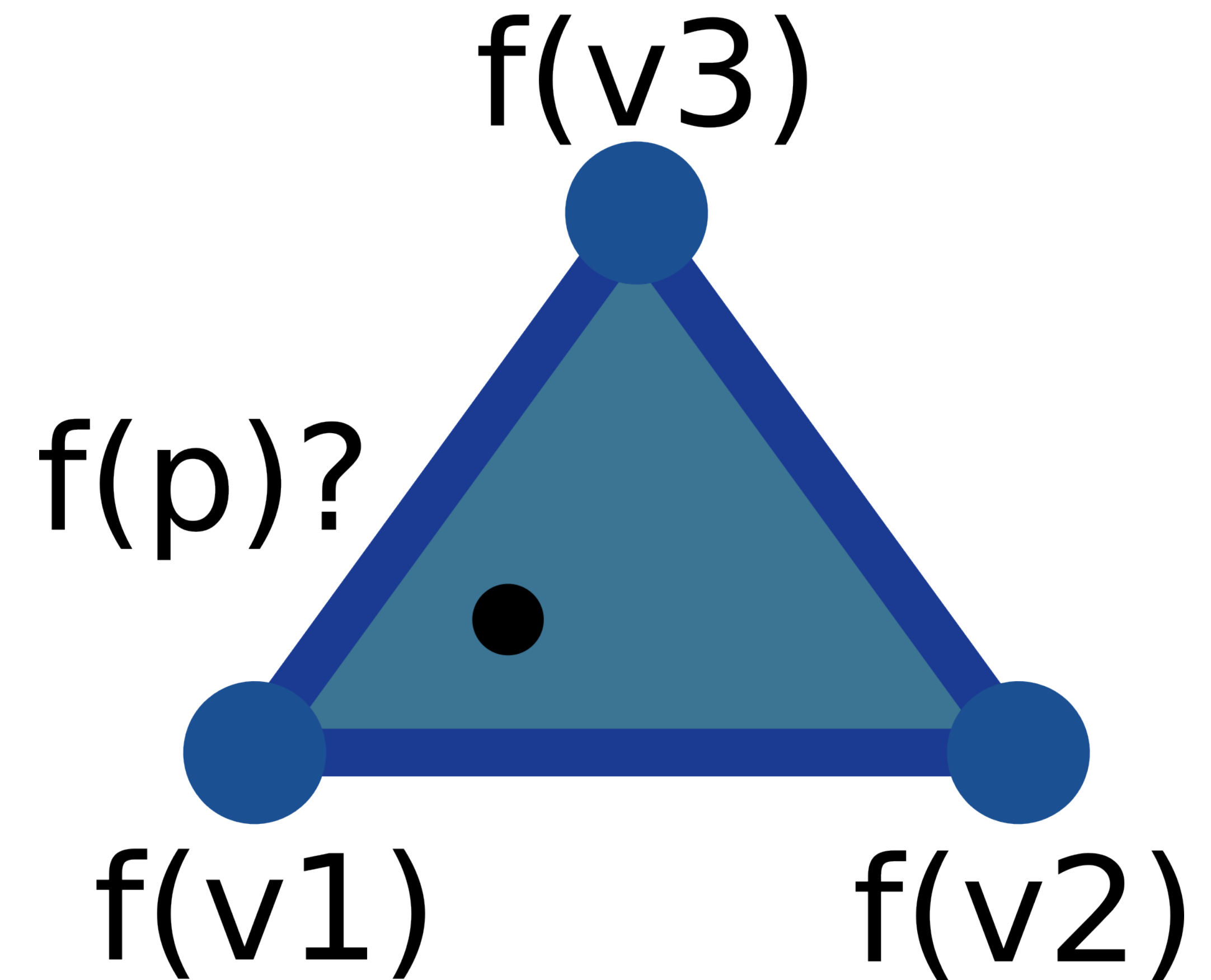
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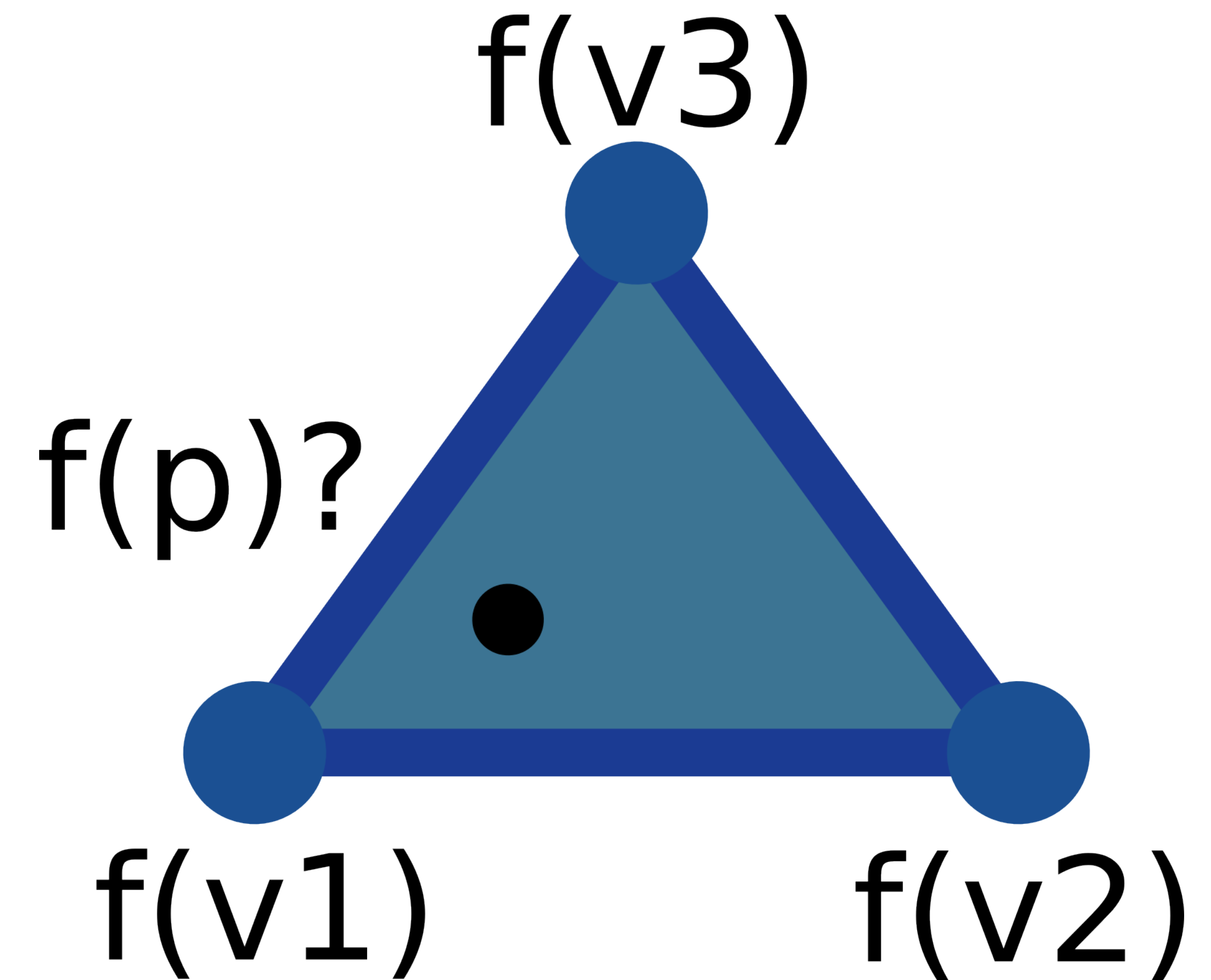
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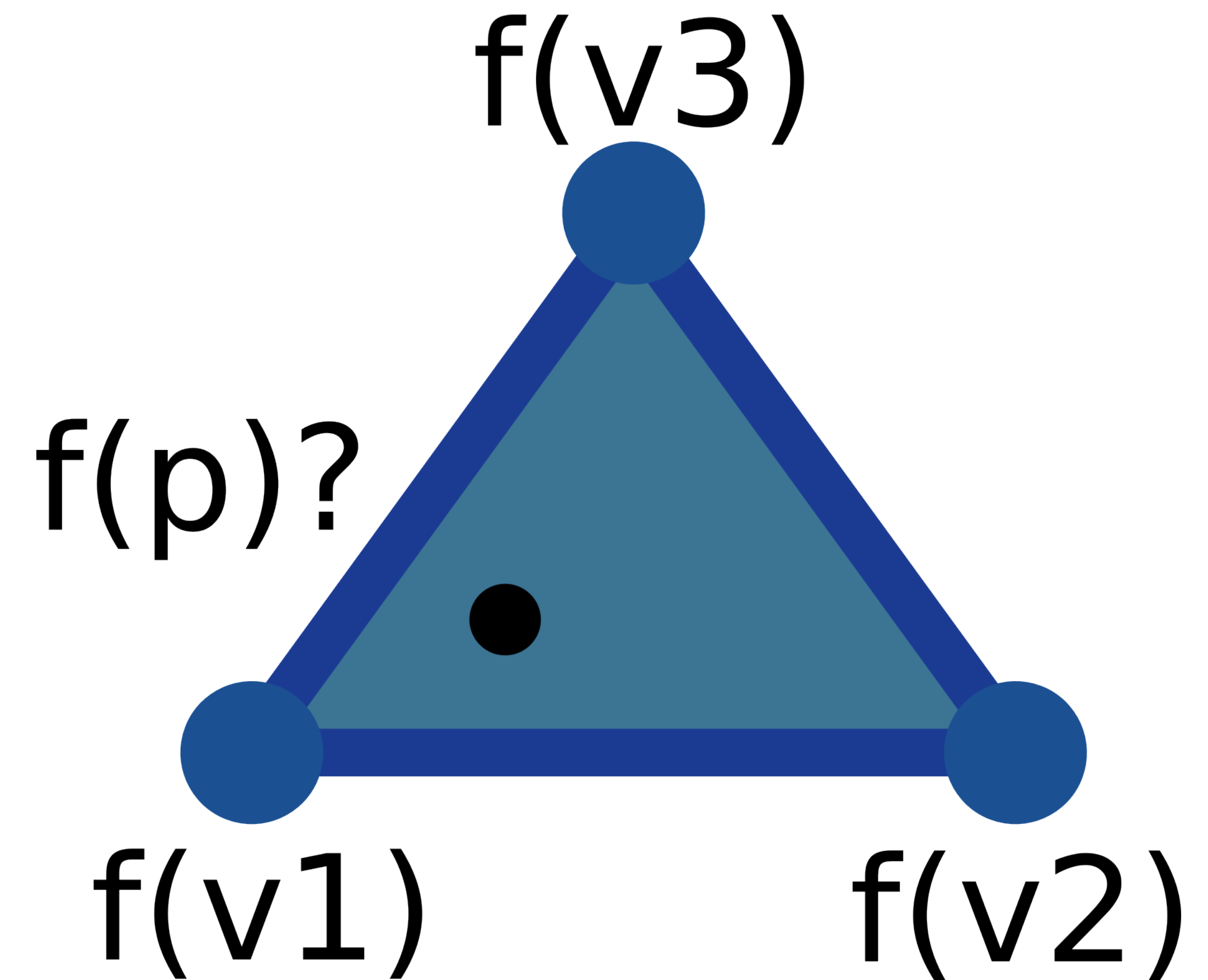
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- Similar reasoning for d-simplex



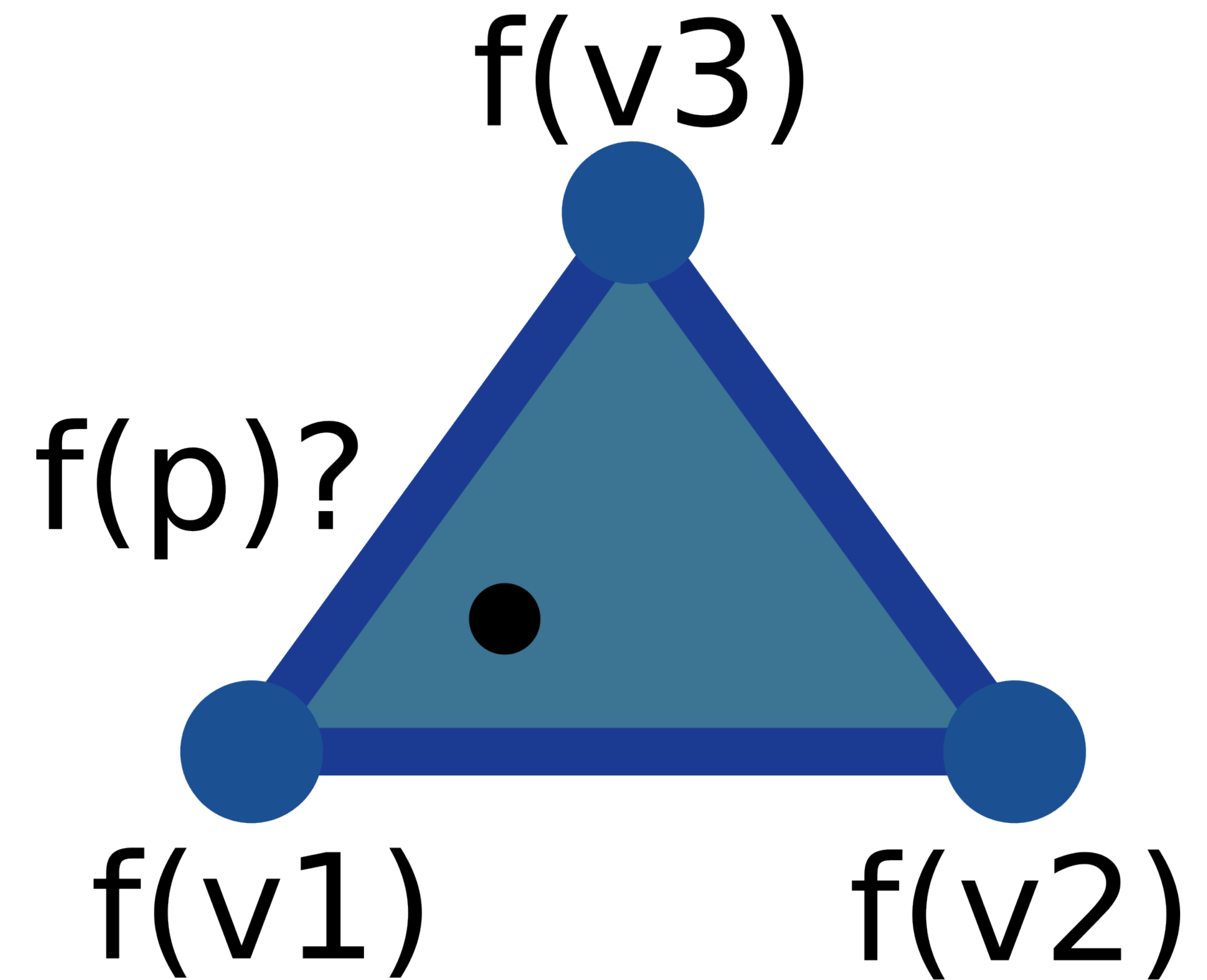
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- Similar reasoning for d-simplex
- Can be used to determine if a point lies within a simplex



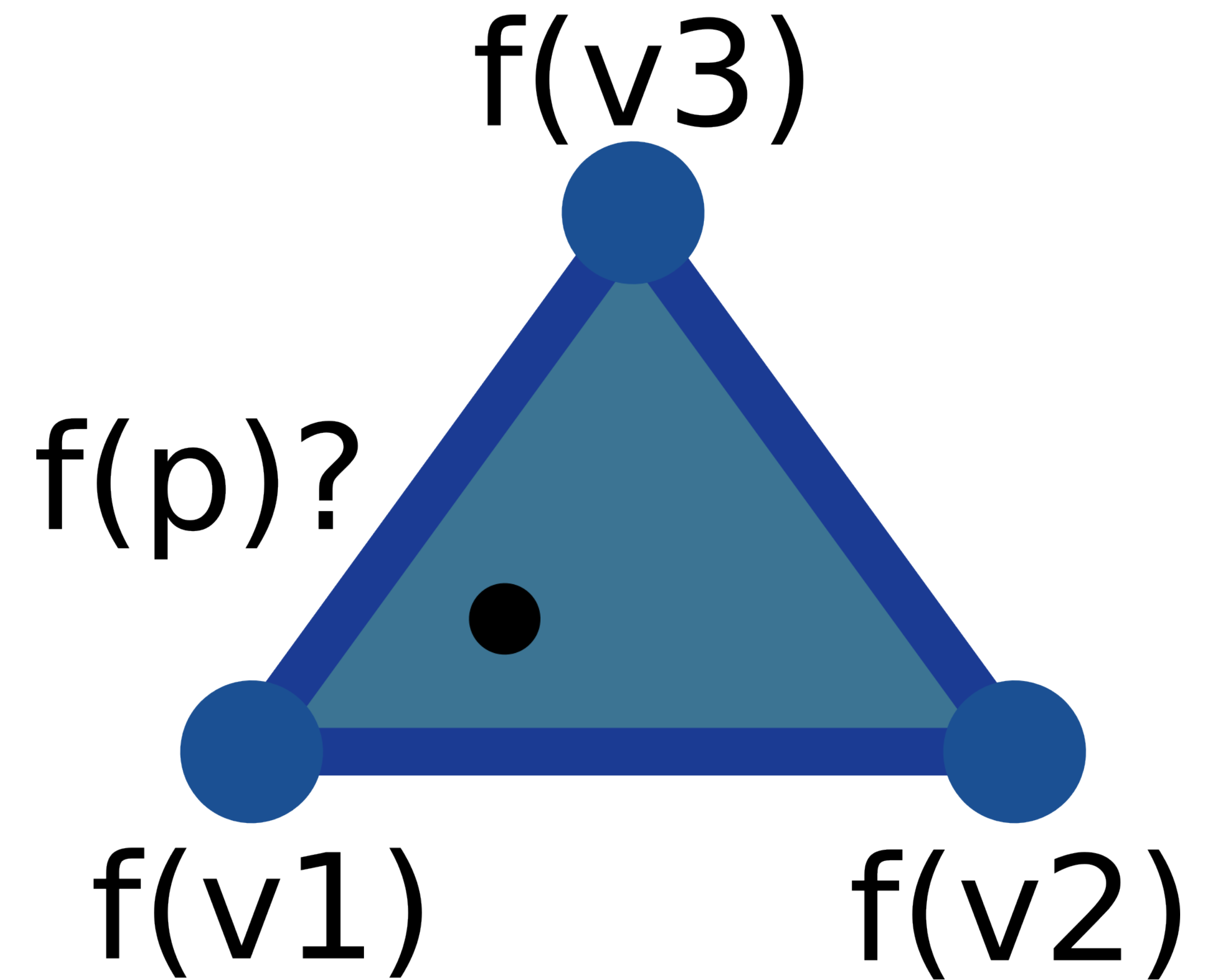
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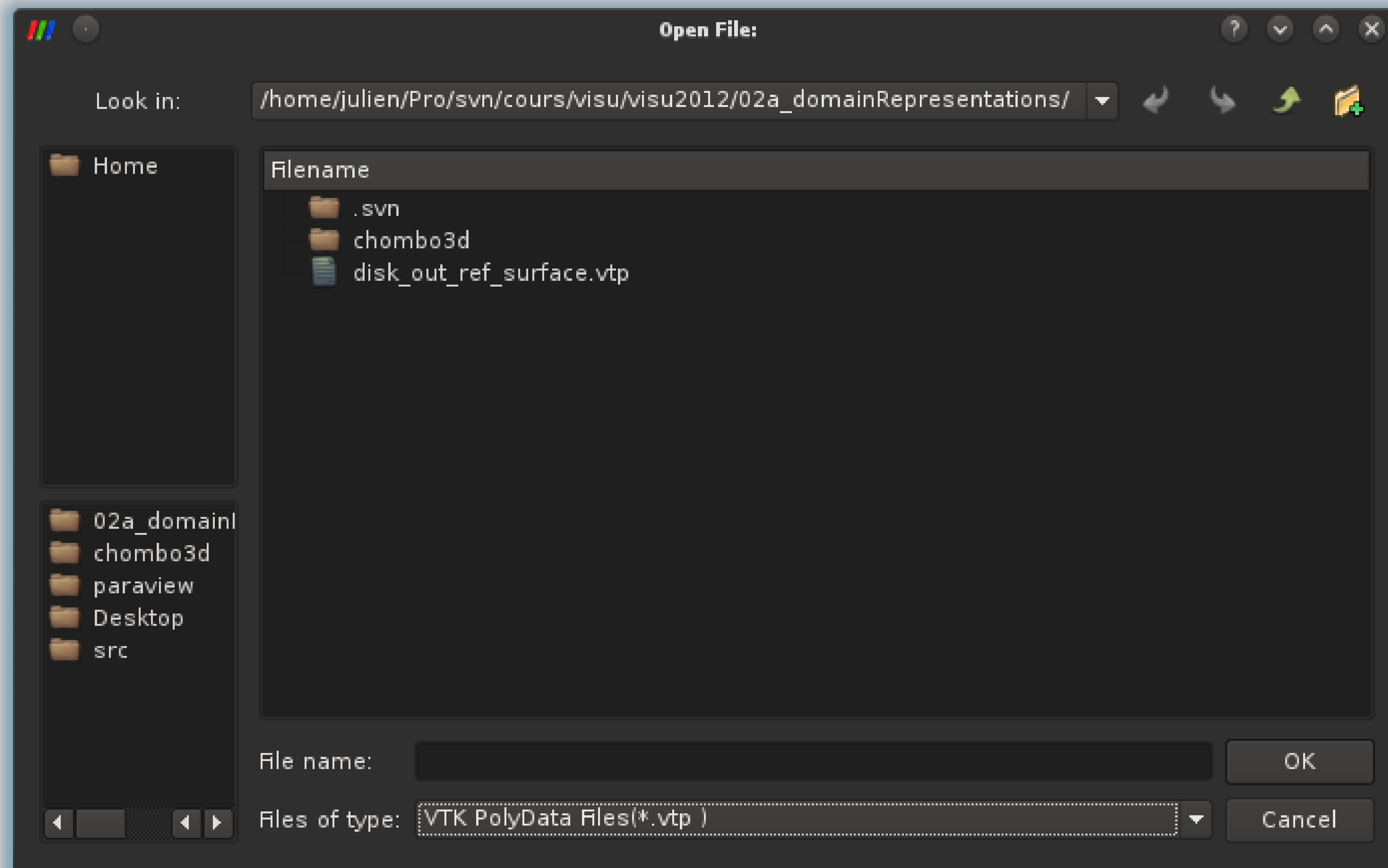


Barycentric coordinates

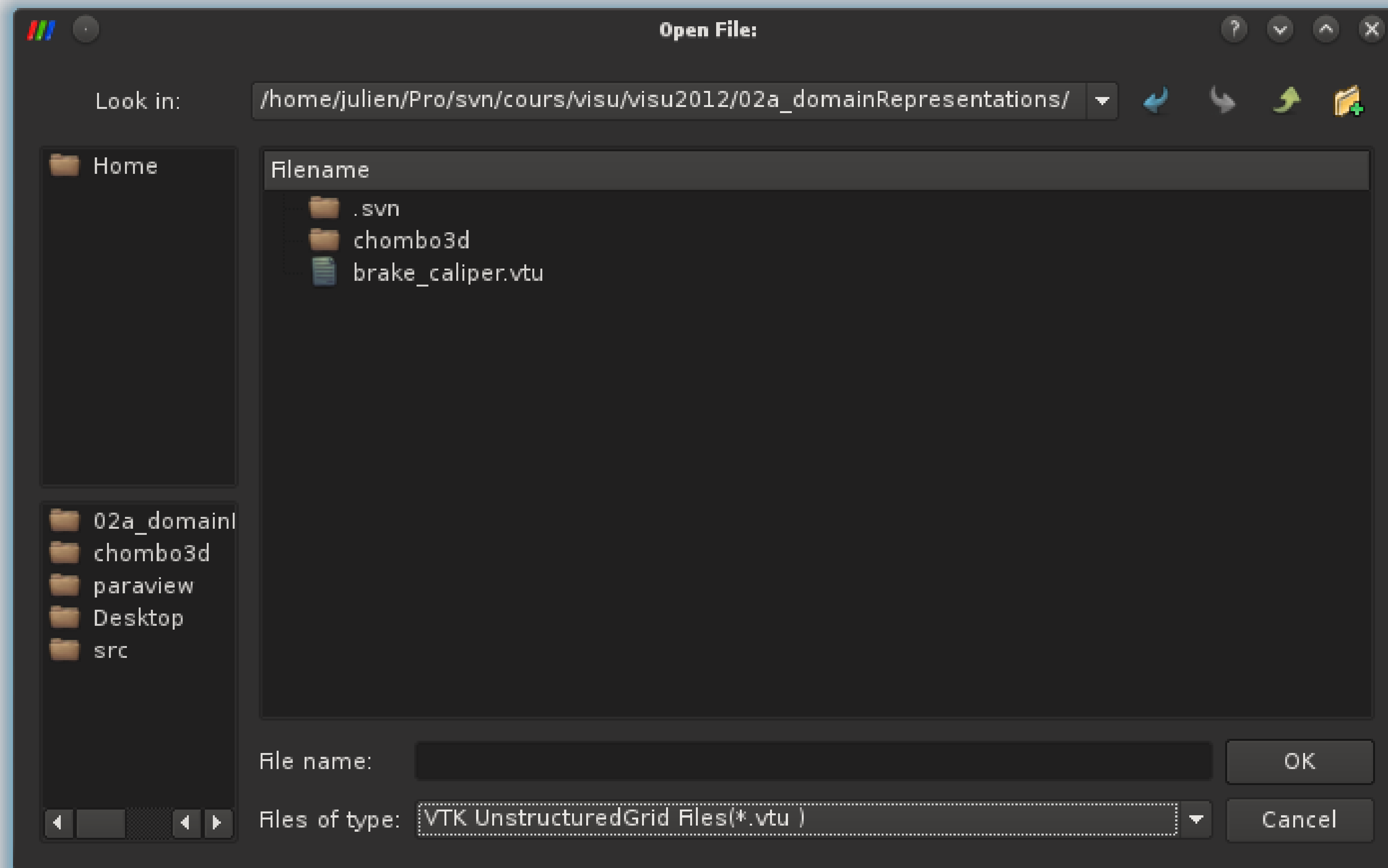
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- Can be used to determine if a point lies within a simplex $(\alpha_1, \alpha_2, \alpha_3) \in [0, 1] \times [0, 1] \times [0, 1]$
- No critical point in the interior!

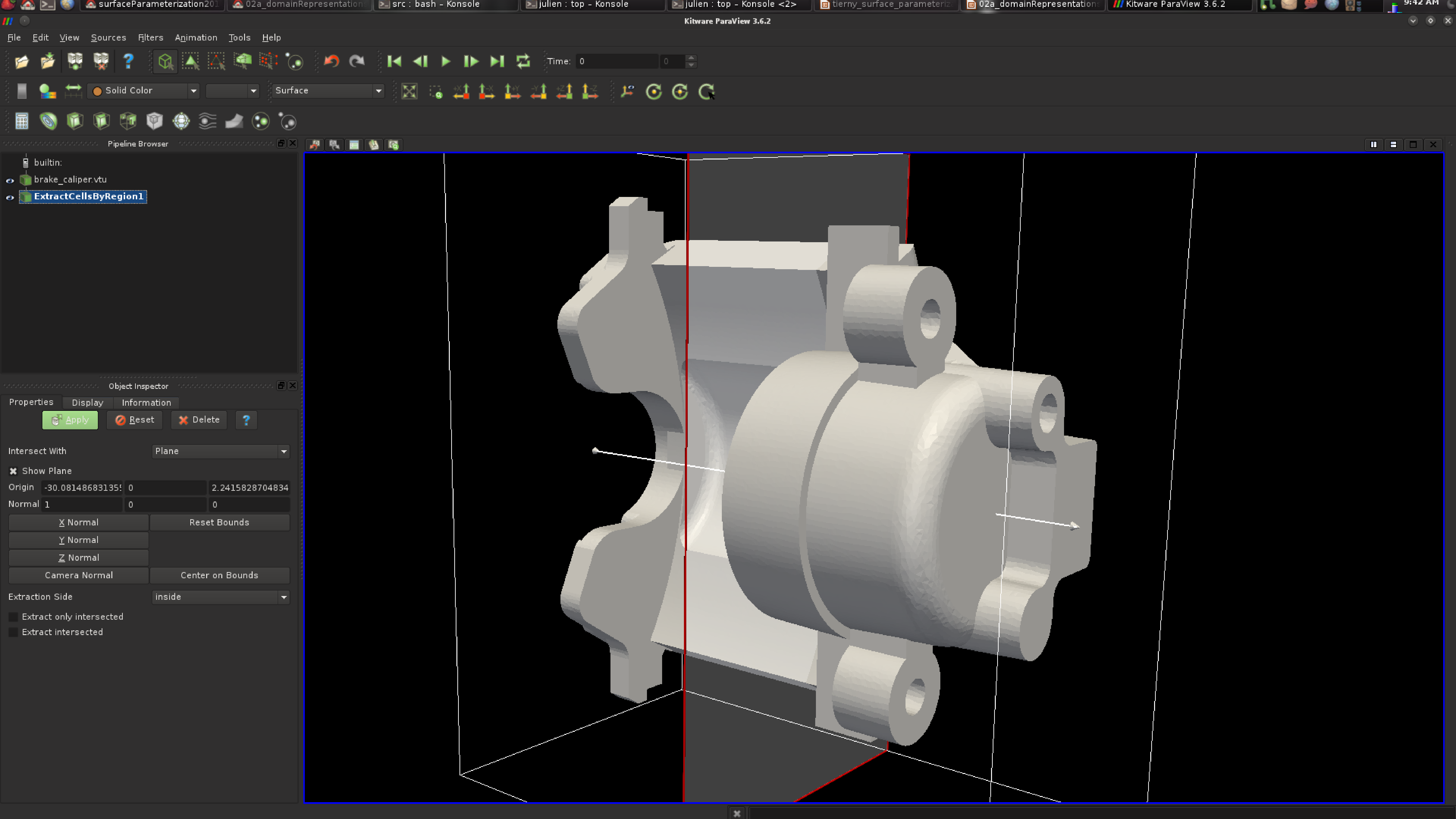


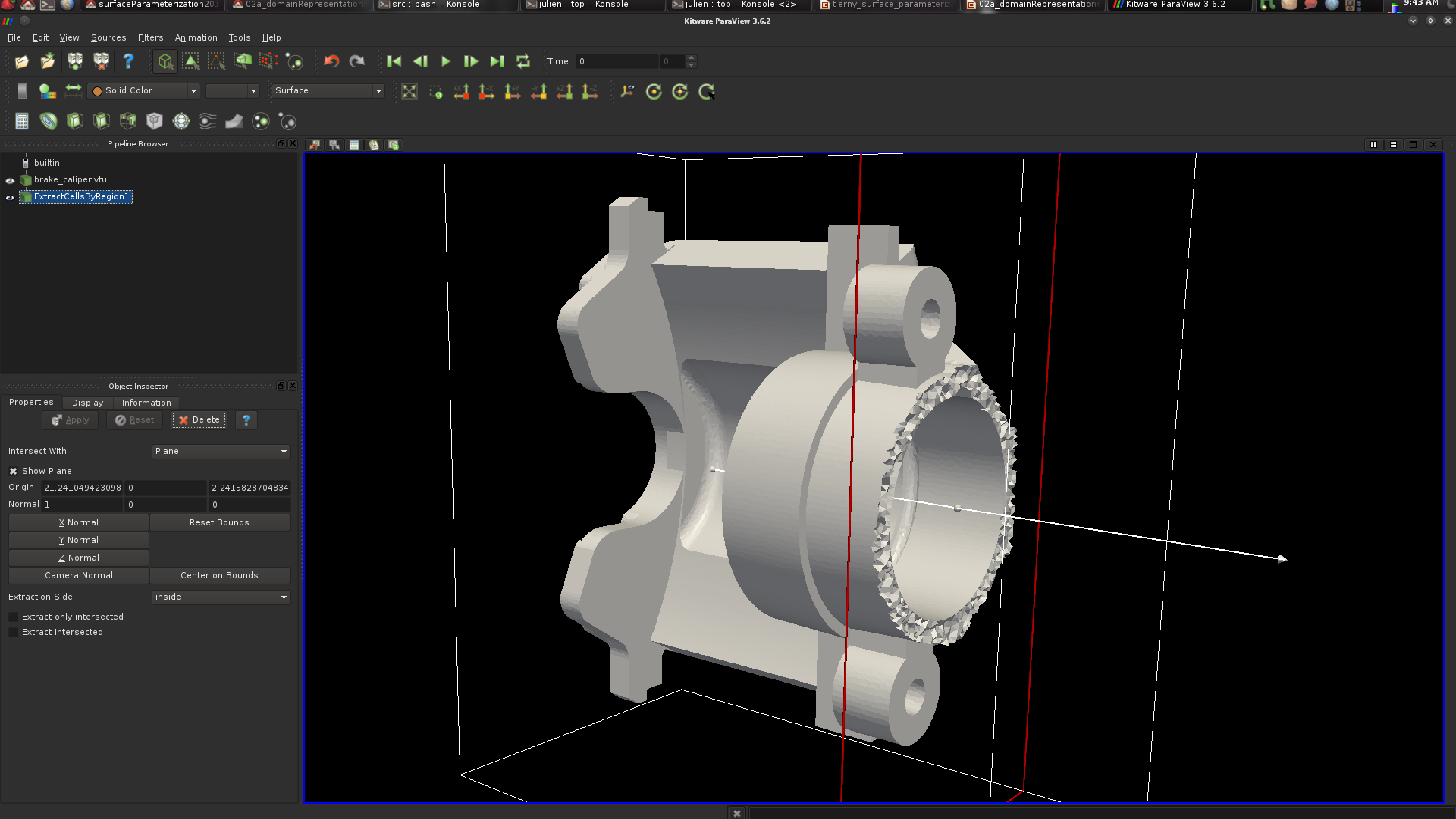
In practice

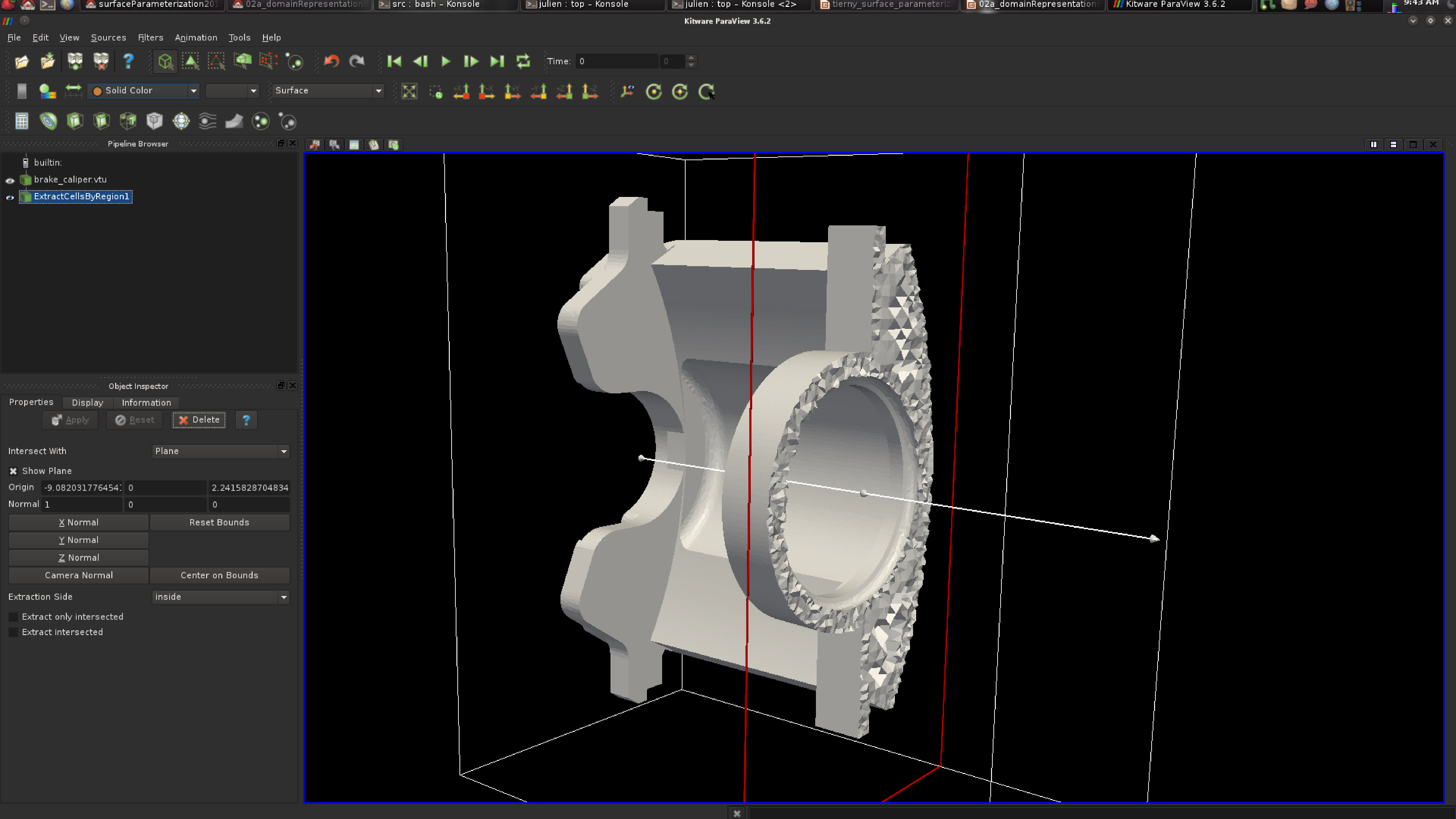


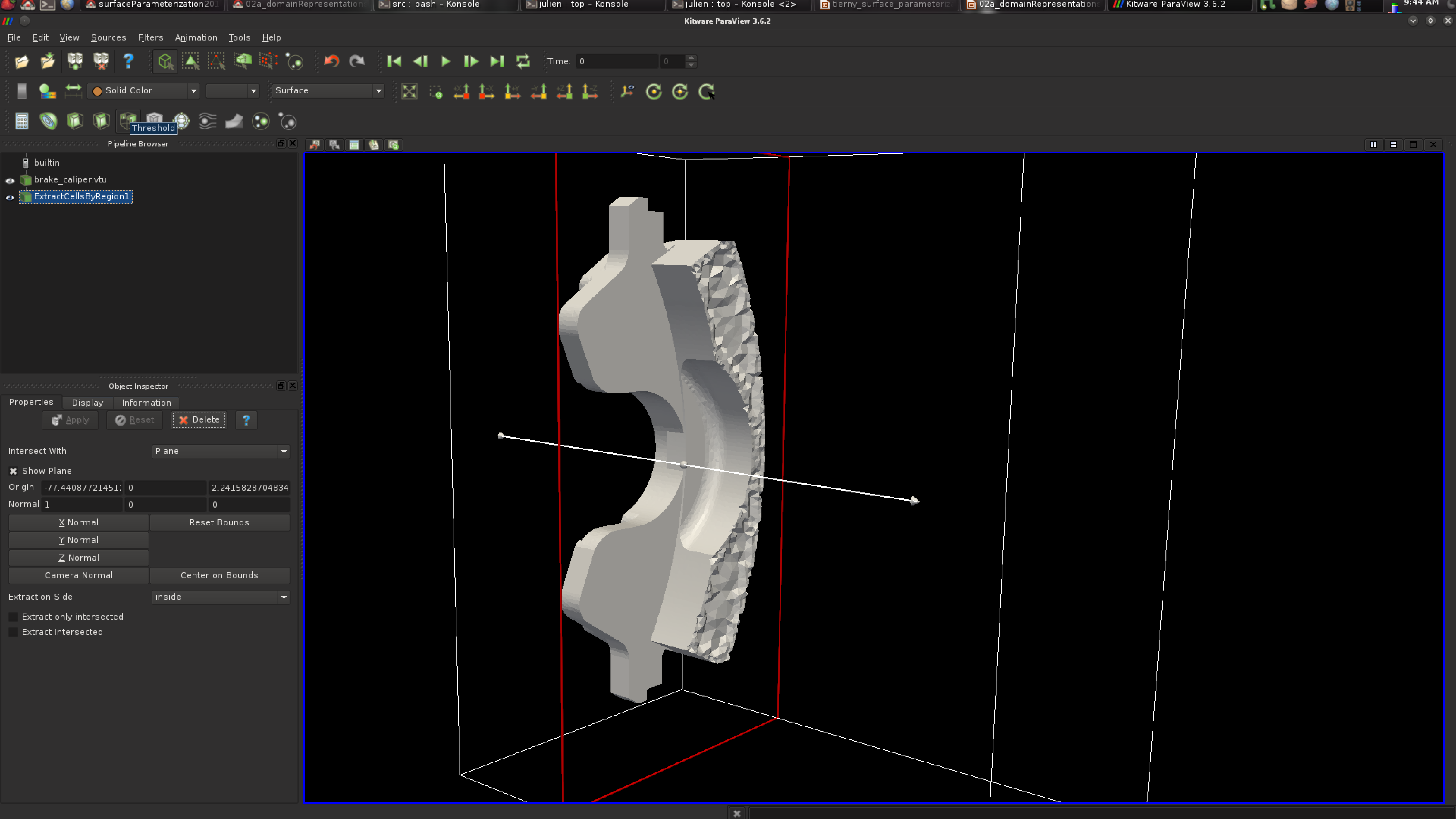
In practice











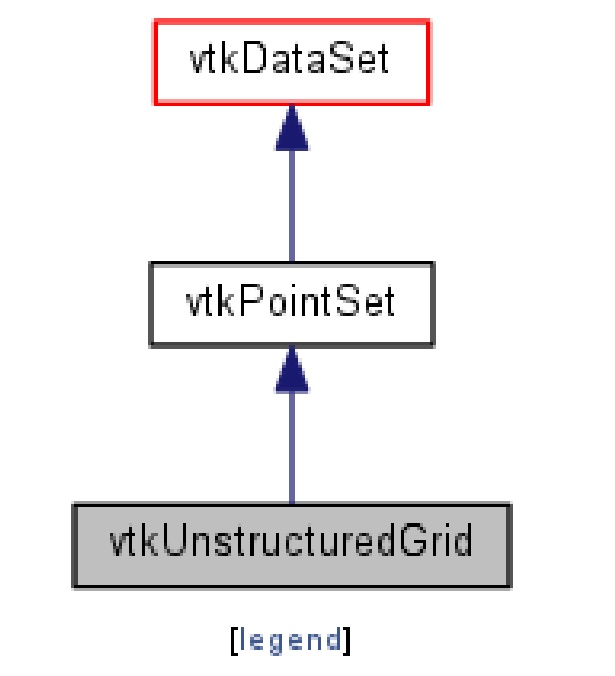
Public Types | Public Member Functions | Static Public Member Functions | Protected Member Functions | Protected Attributes | List of all members

vtkUnstructuredGrid Class Reference

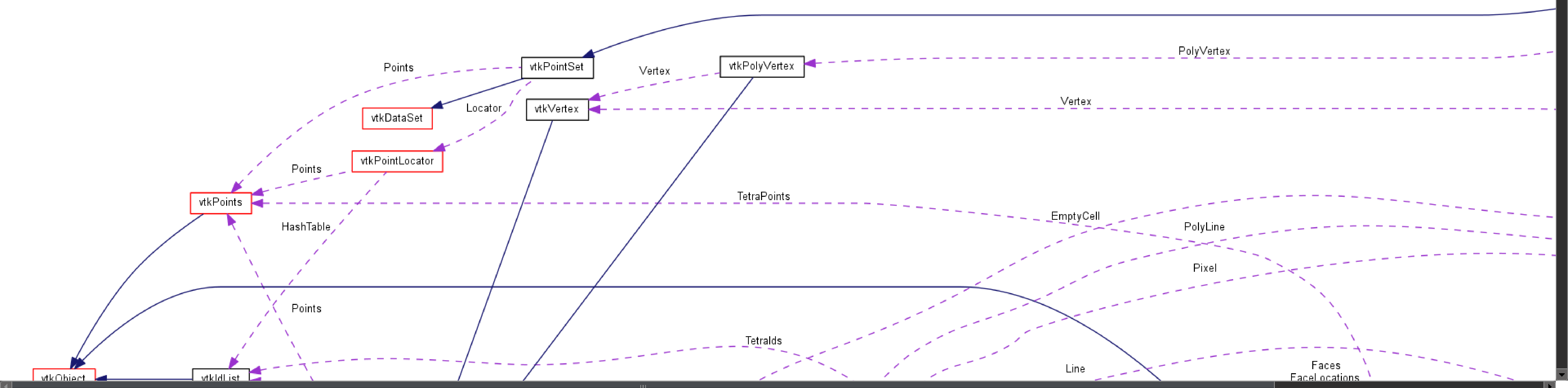
dataset represents arbitrary combinations of all possible cell types [More...](#)

```
#include <vtkUnstructuredGrid.h>
```

Inheritance diagram for vtkUnstructuredGrid:



Collaboration diagram for vtkUnstructuredGrid:



Public Member Functions

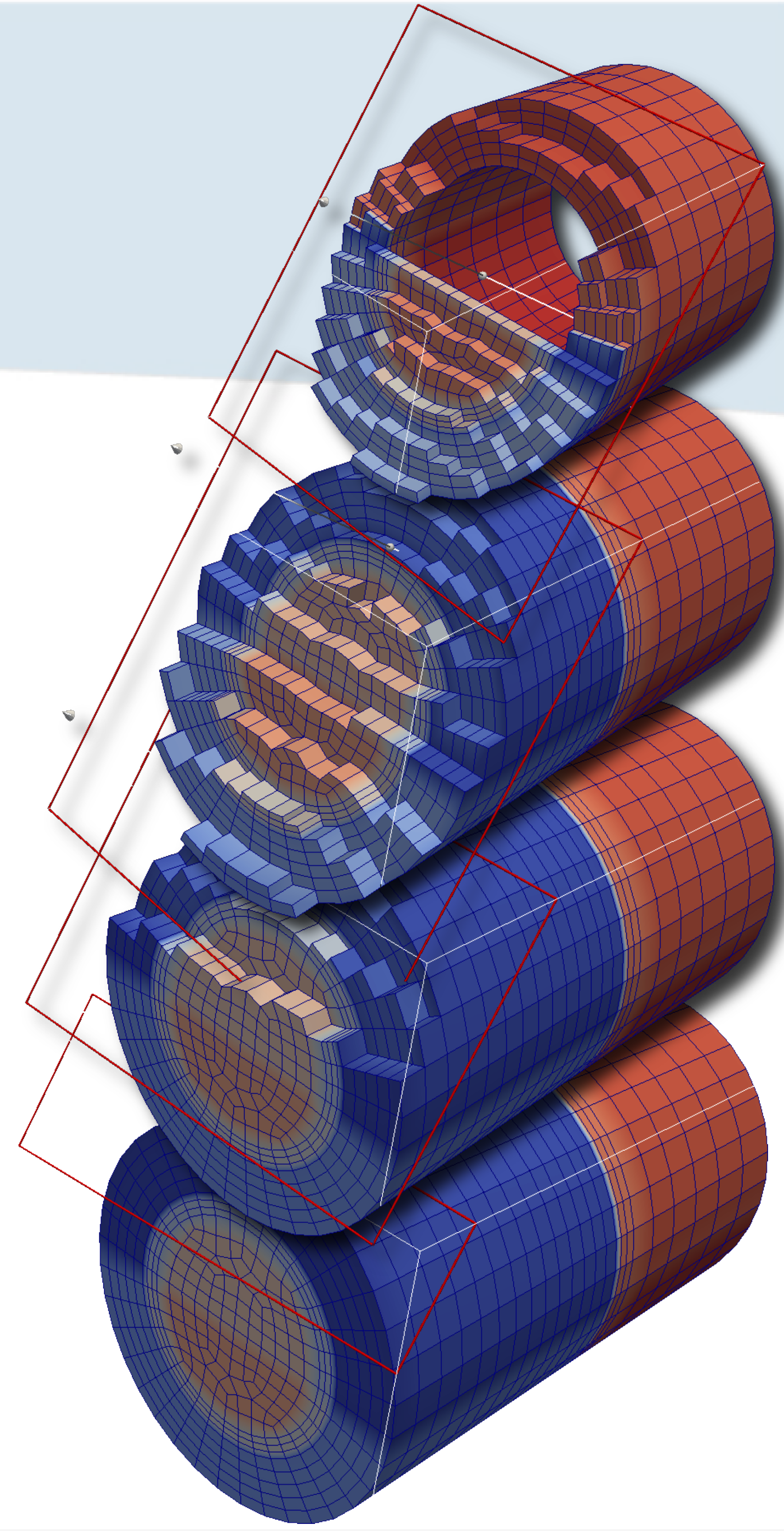
virtual int	IsA (const char *type)
vtkUnstructuredGrid *	NewInstance () const
void	PrintSelf (ostream &os, vtkIndent indent)
vtkIdType	InsertNextCell (int type, vtkIdType npts, vtkIdType *ptIds)
vtkIdType	InsertNextCell (int type, vtkIdList *ptIds)
vtkIdType	InsertNextCell (int type, vtkIdType npts, vtkIdType *ptIds, vtkIdType nfaces, vtkIdType *faces)
int	GetCellType (vtkIdType cellId)
vtkUnsignedCharArray *	GetCellTypesArray ()
vtkIdTypeArray *	GetCellLocationsArray ()
void	Squeeze ()
void	Initialize ()
int	GetMaxCellSize ()
void	BuildLinks ()
vtkCellLinks *	GetCellLinks ()
virtual void	GetCellPoints (vtkIdType cellId, vtkIdType &npts, vtkIdType *pts)
void	GetFaceStream (vtkIdType cellId, vtkIdList *ptIds)
void	GetFaceStream (vtkIdType cellId, vtkIdType &nfaces, vtkIdType *ptIds)
vtkCellArray *	GetCells ()
void	ReplaceCell (vtkIdType cellId, int npts, vtkIdType *pts)
vtkIdType	InsertNextLinkedCell (int type, int npts, vtkIdType *pts)
void	RemoveReferenceToCell (vtkIdType ptId, vtkIdType cellId)
void	AddReferenceToCell (vtkIdType ptId, vtkIdType cellId)
void	ResizeCellList (vtkIdType ptId, int size)
virtual int	GetGhostLevel ()
unsigned long	GetActualMemorySize ()
void	GetIdsOfCellsOfType (int type, vtkIdTypeArray *array)
int	IsHomogeneous ()
void	RemoveGhostCells (int level)
vtkIdType *	GetFaces (vtkIdType cellId)
int	InitializeFacesRepresentation (vtkIdType numPrevCells)
int	GetDataObjectType ()
virtual void	Allocate (vtkIdType numCells=1000, int extSize=1000)
void	Reset ()
virtual void	CopyStructure (vtkDataSet *ds)
vtkIdType	GetNumberOfCells ()
virtual vtkCell *	GetCell (vtkIdType cellId)
virtual void	GetCell (vtkIdType cellId, vtkGenericCell *cell)
virtual void	GetCellBounds (vtkIdType cellId, double bounds[6])
virtual void	GetCellPoints (vtkIdType cellId, vtkIdList *ptIds)
void	GetPointCells (vtkIdType ptId, vtkIdList *cellIds)
void	SetCells (int type, vtkCellArray *cells)

Are we done?

- Nearly

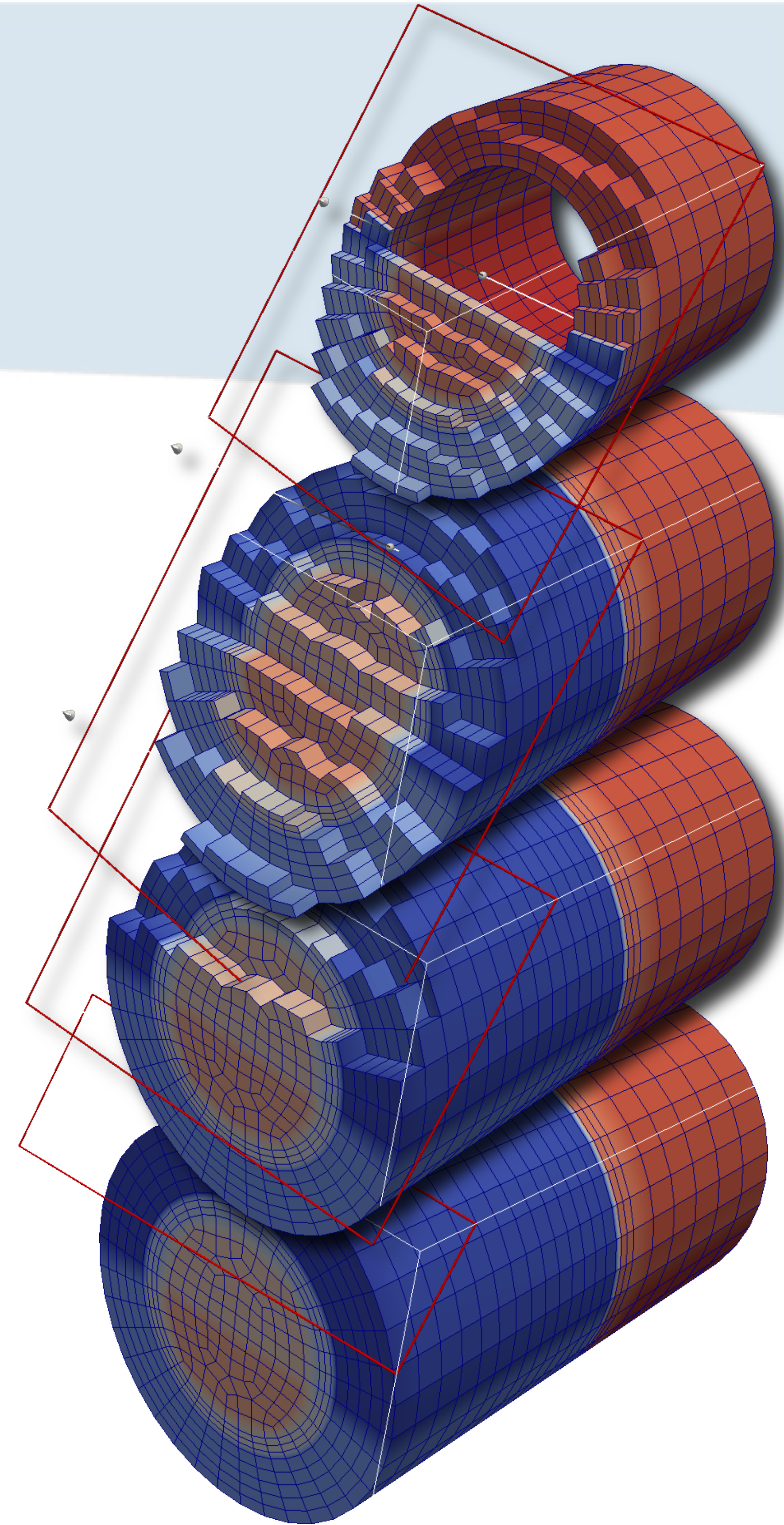
Are we done?

- Nearly
 - What about these objects?



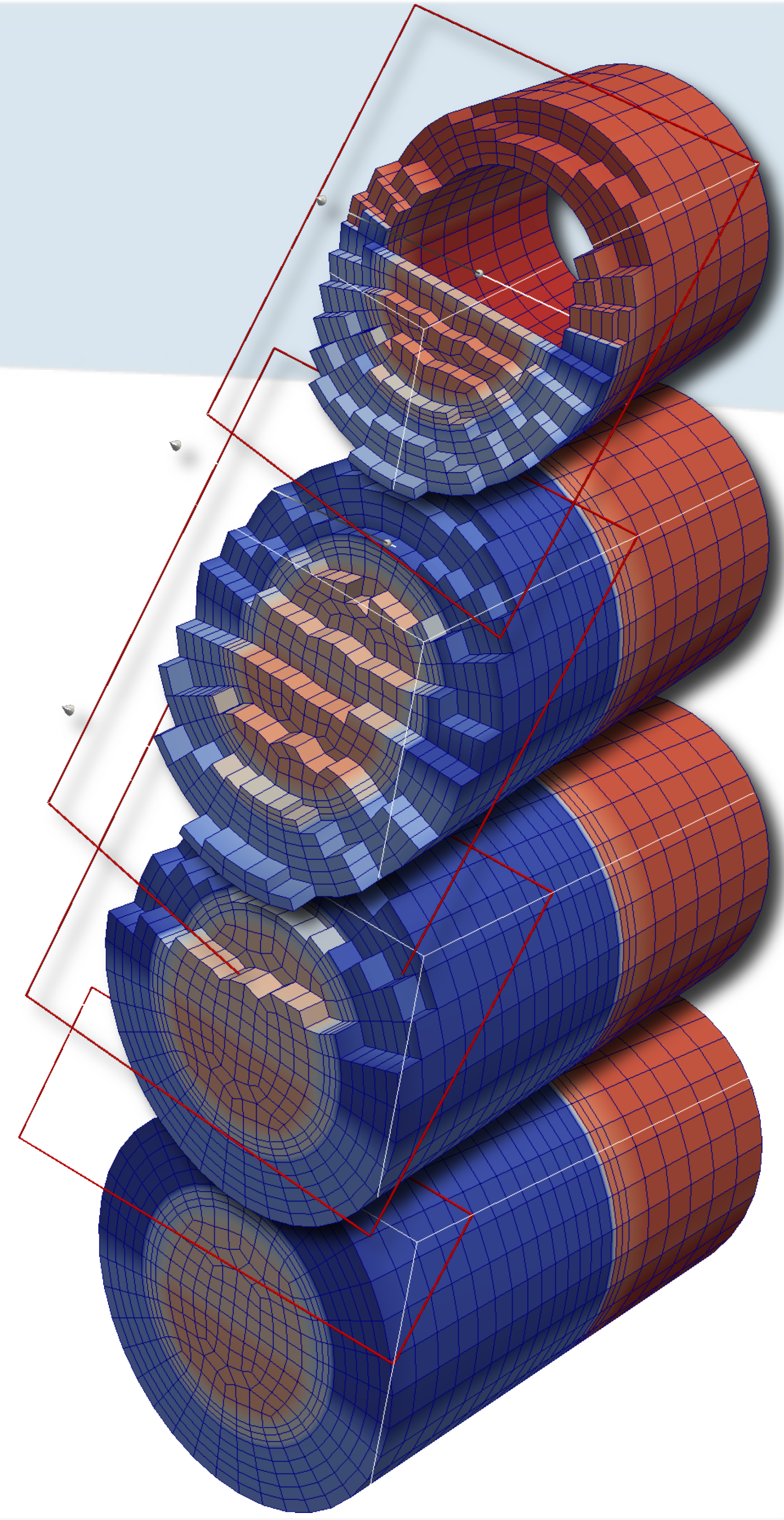
Are we done?

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- Other manifold representations



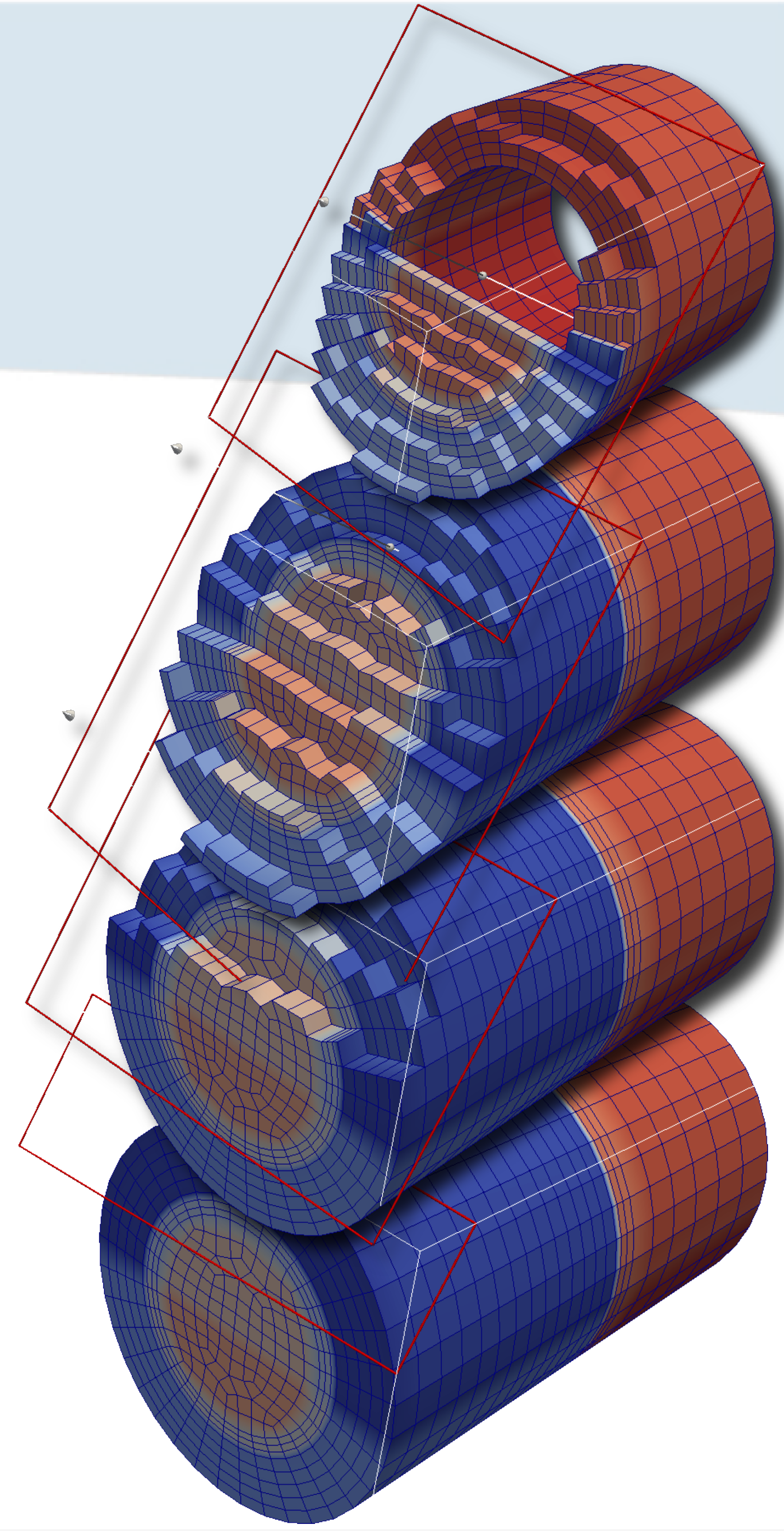
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 - Closure-finite Weak-topology (CW) complex



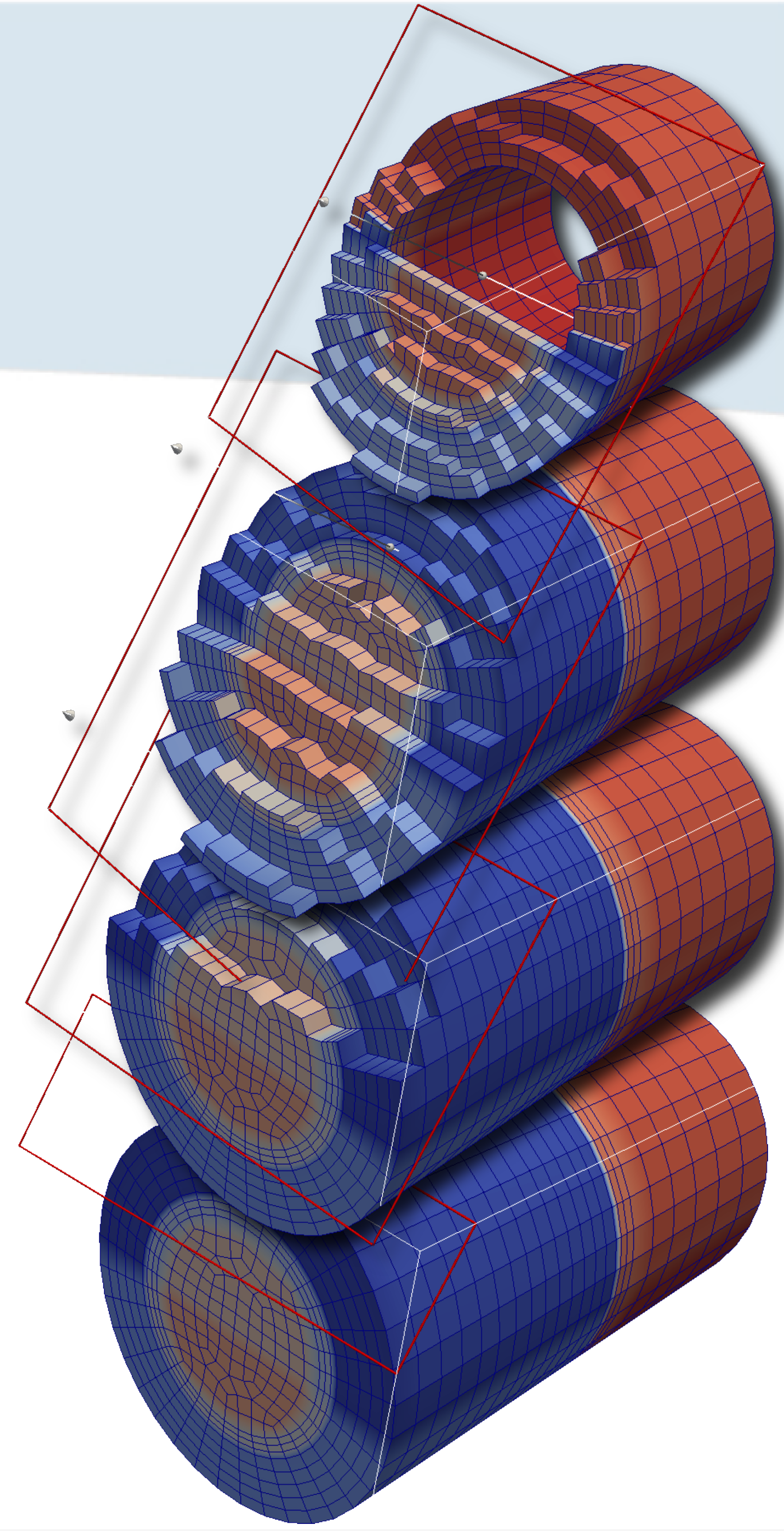
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 - Arbitrary number of samples per cell



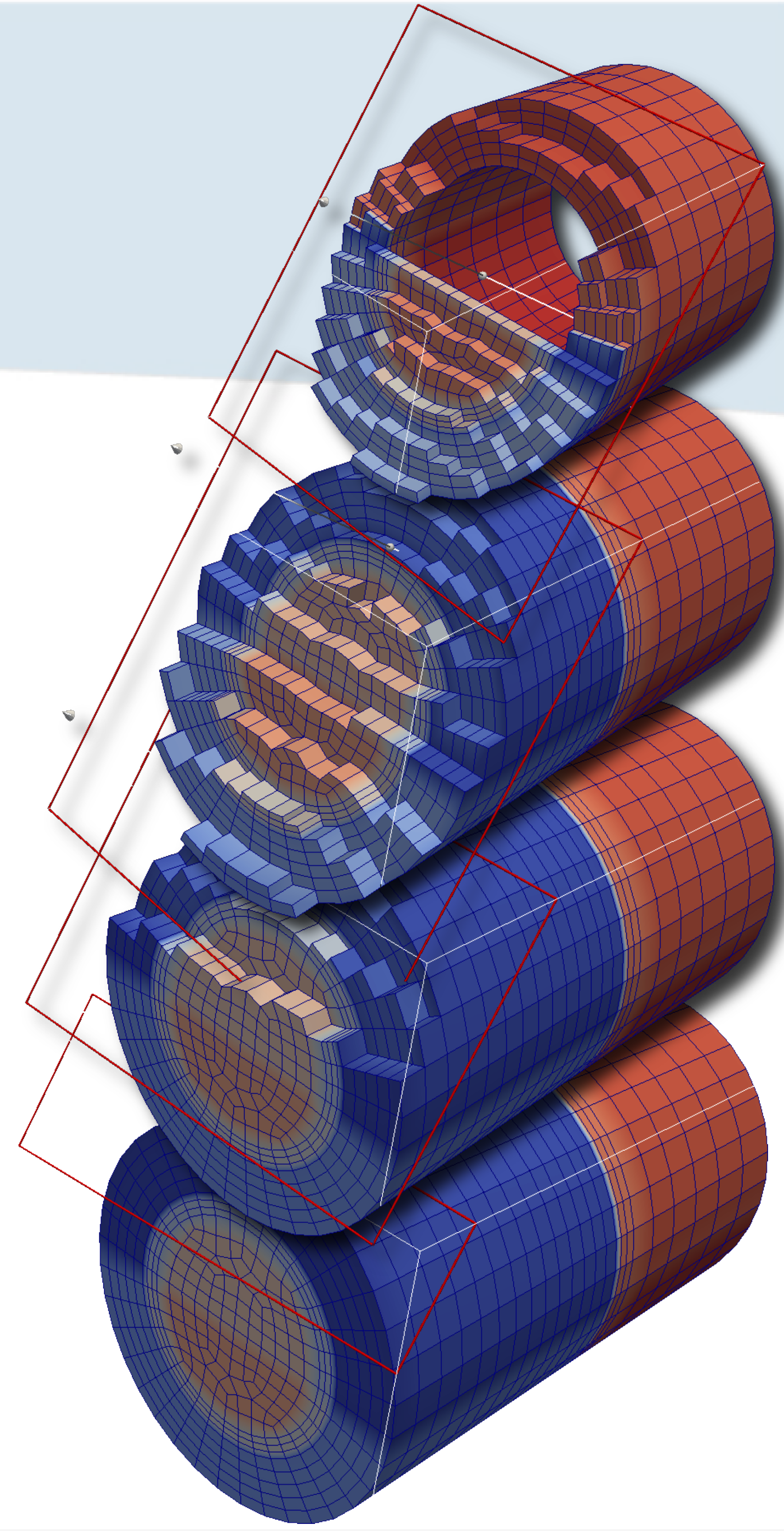
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 - Closure-finite Weak-topology (CW) complex
 - Arbitrary number of samples per cell
 - Similar implementation



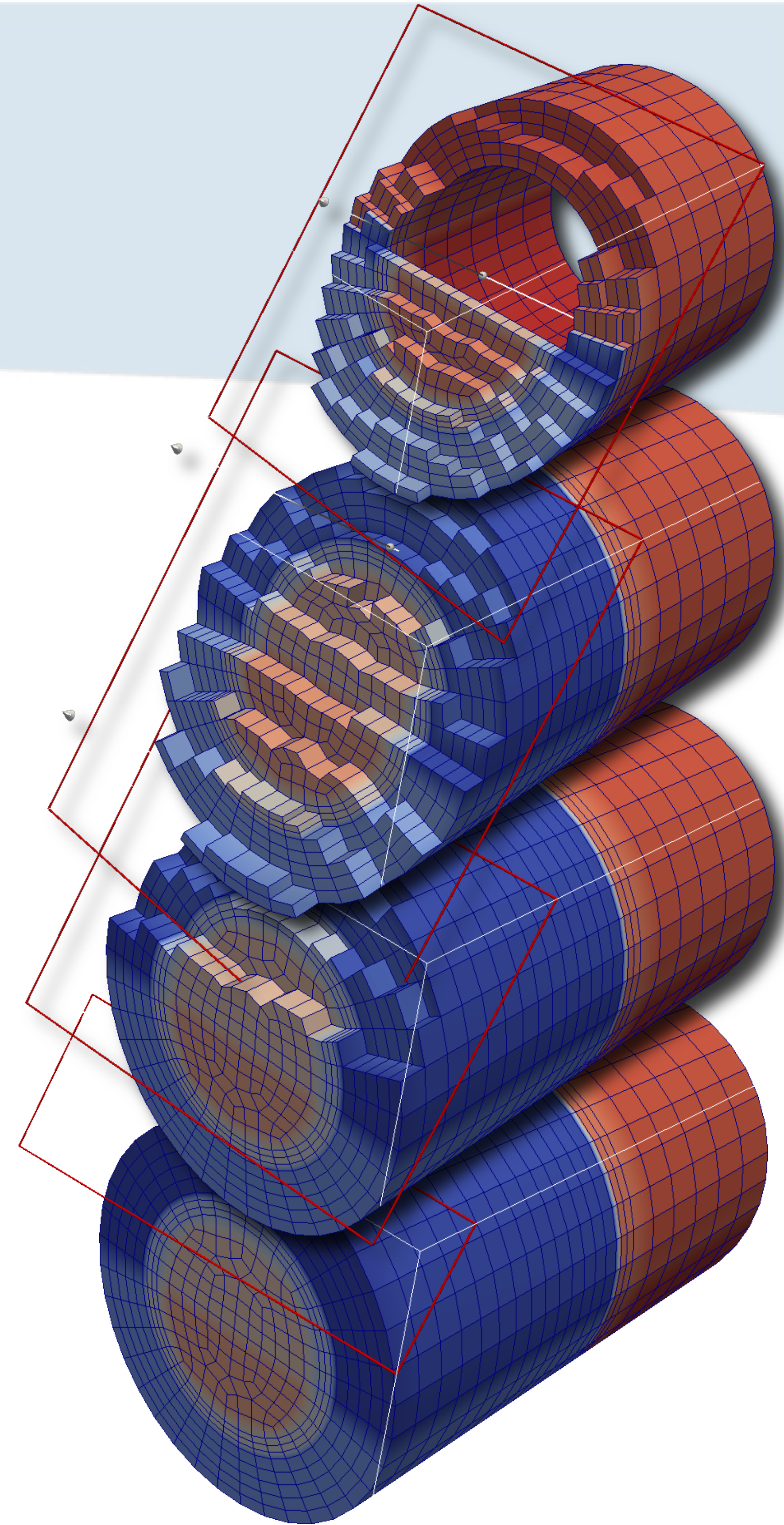
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 - Similar implementation (vtkUnstructuredGrid)



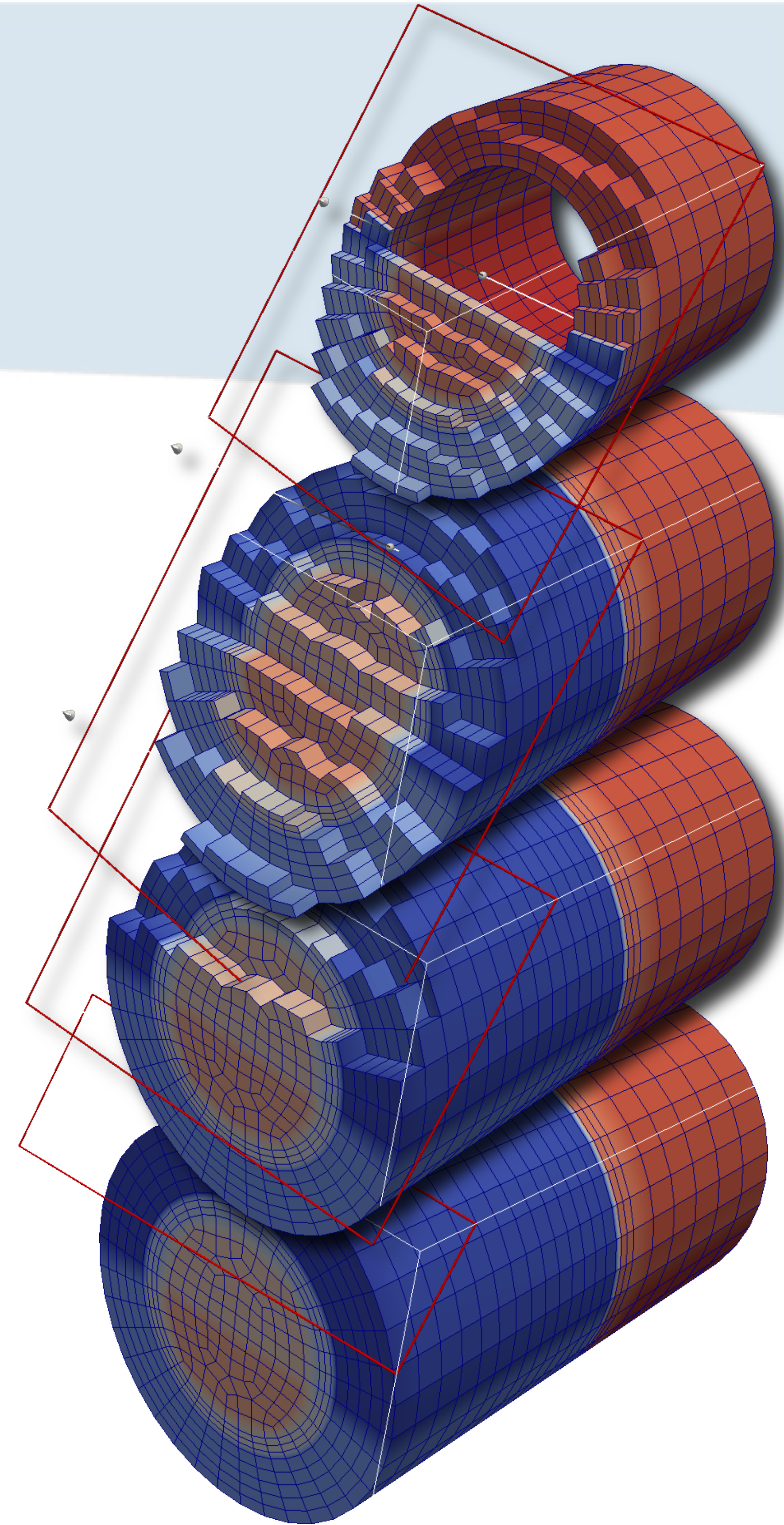
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 - Special cases with 2^d samples per cell



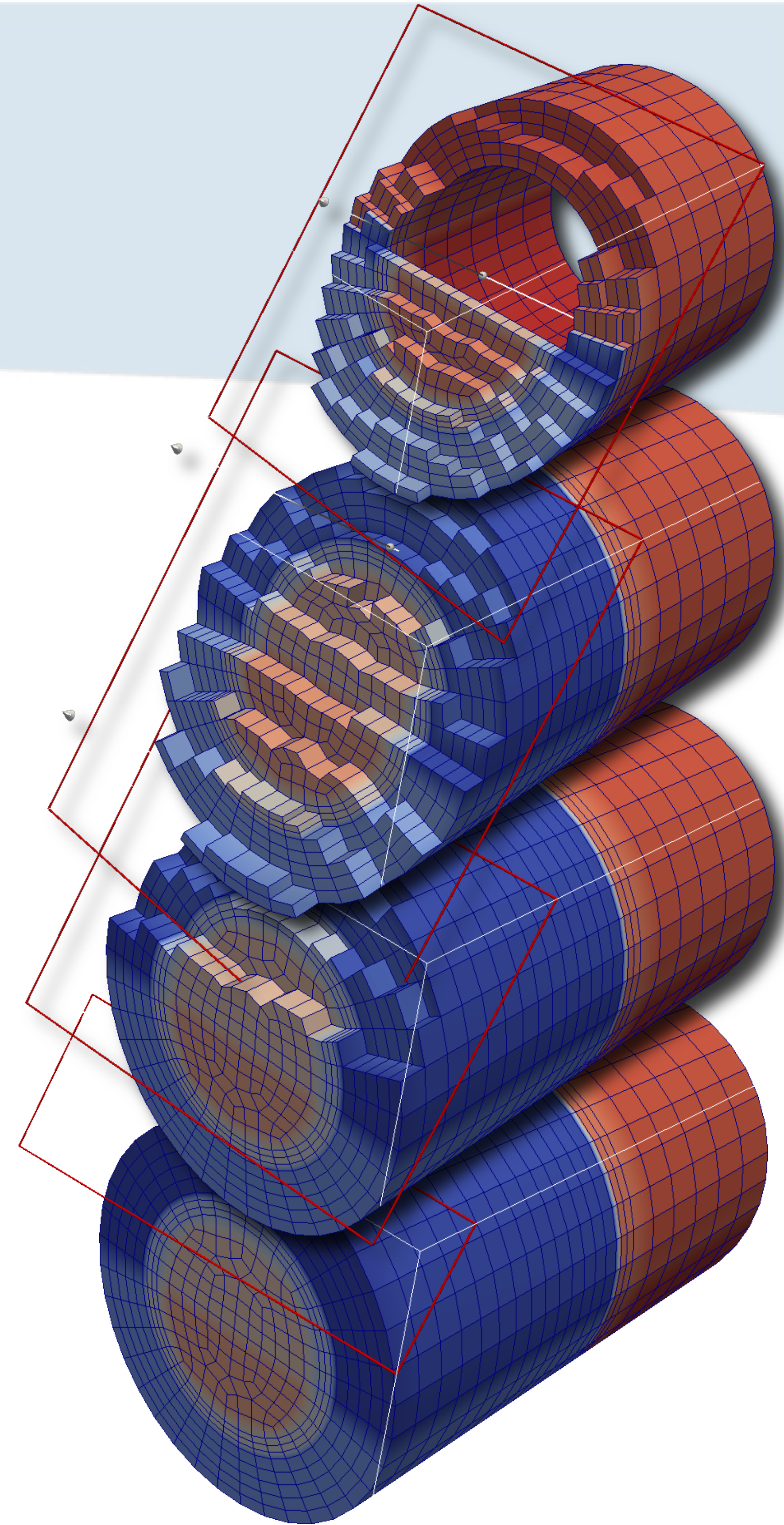
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 - Special cases with 2^d samples per cell
 - Quad mesh
 - Hexahedral mesh



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 - Similar implementation (vtkUnstructuredGrid)
 - Special cases with 2^d samples per cell
 - Quad mesh
 - Hexahedral mesh ... intriguing properties



In conclusion

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 - How to implement representations for manifold spaces
 - How to interpolate numerical values all over these domains
 - Example of available implementations