

# Finite Difference and Finite Element Methods

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Fraunhofer Institute for Medical Image Computing MEVIS, Bremen

Based on a presentation by Tobias Preusser

EMBC 2015 Workshop on  
Practical Computer Modeling for Medical Device Development

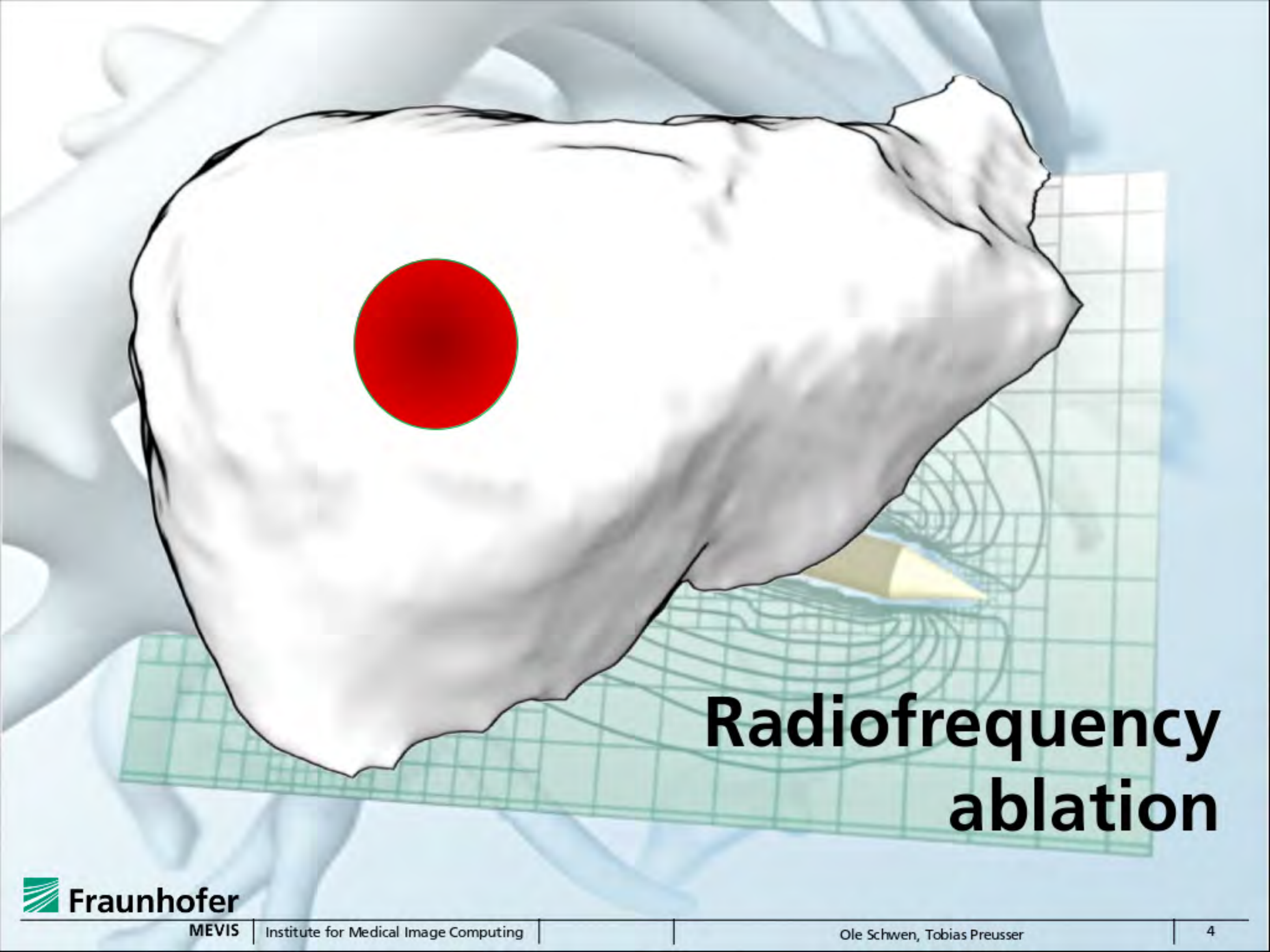
Milan, 2015-08-25

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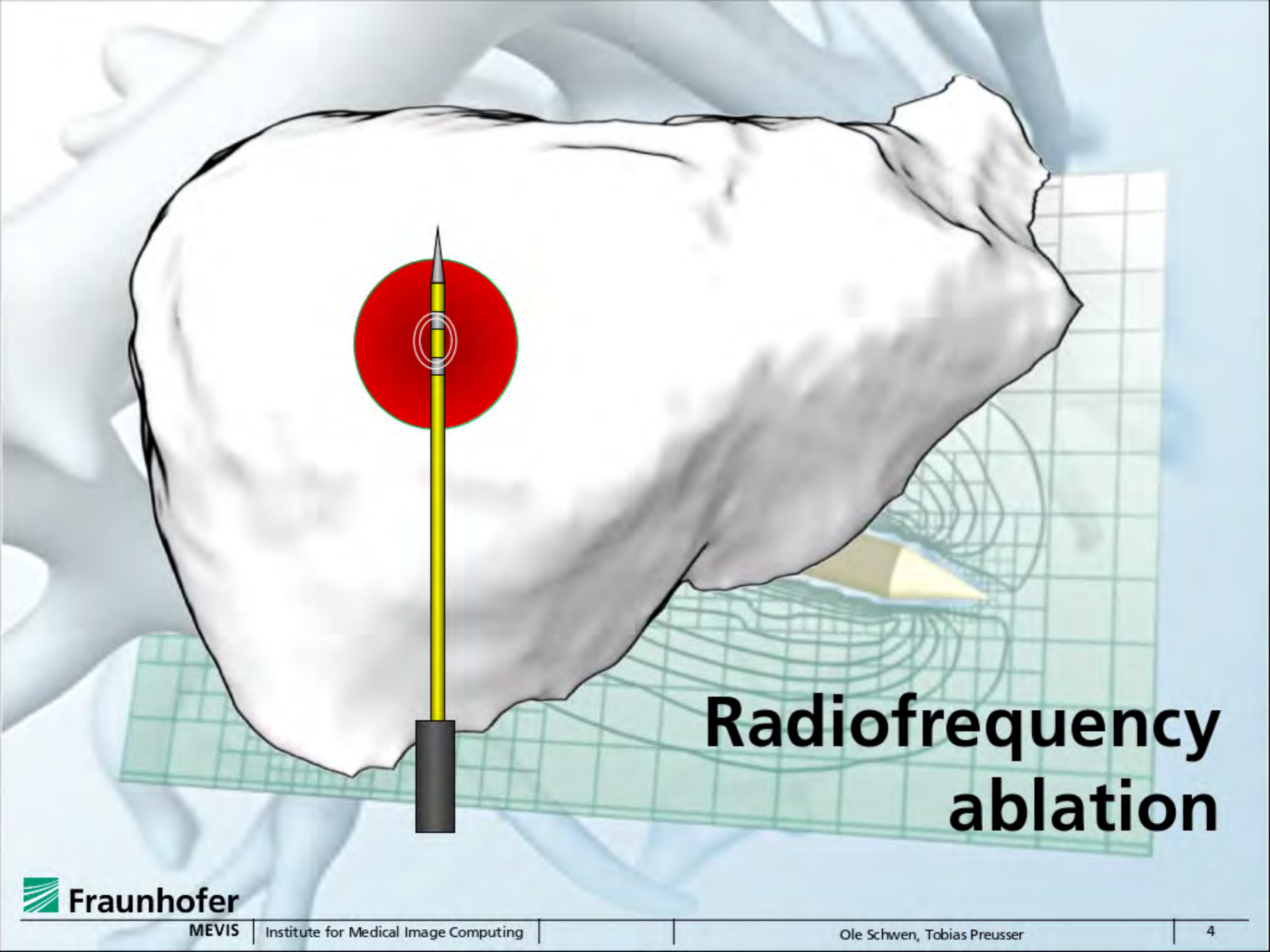
- ① Modeling Thermal Ablation
- ② Discretizing the Electrostatic Equation
  - » Finite Difference Method (FDM)
  - » Finite Element Method (FEM)
- ③ Discretizing the Bioheat Transfer Equation
- ④ Pros and Cons: FDM vs. FEM
- ⑤ Examples

A 3D rendering of several hands holding a sheet of graph paper. A pencil is lying on the paper, and a large green arrow points upwards from the top right corner. The background is a light blue gradient.

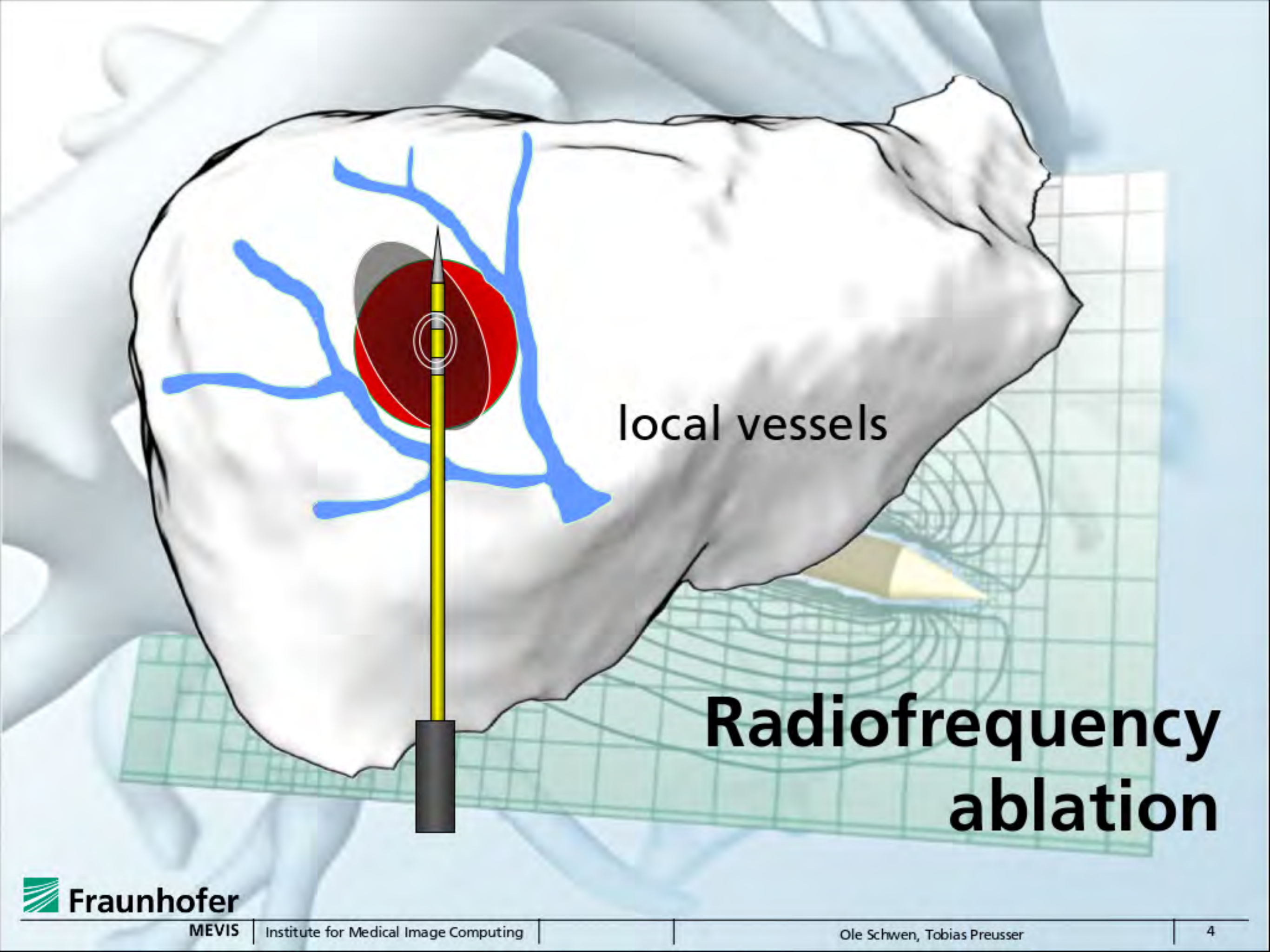
# Modeling Thermal Ablation



# Radiofrequency ablation



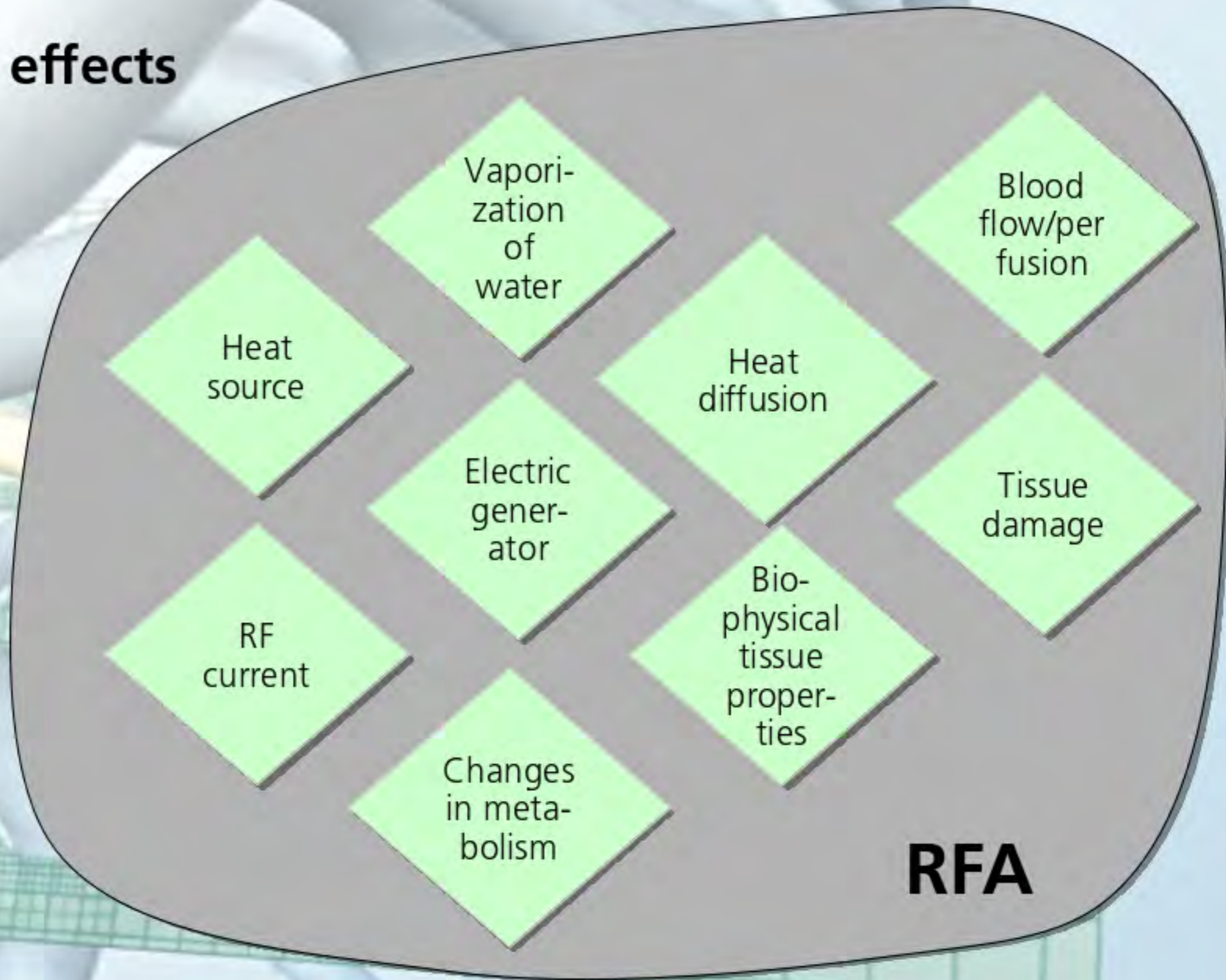
# Radiofrequency ablation



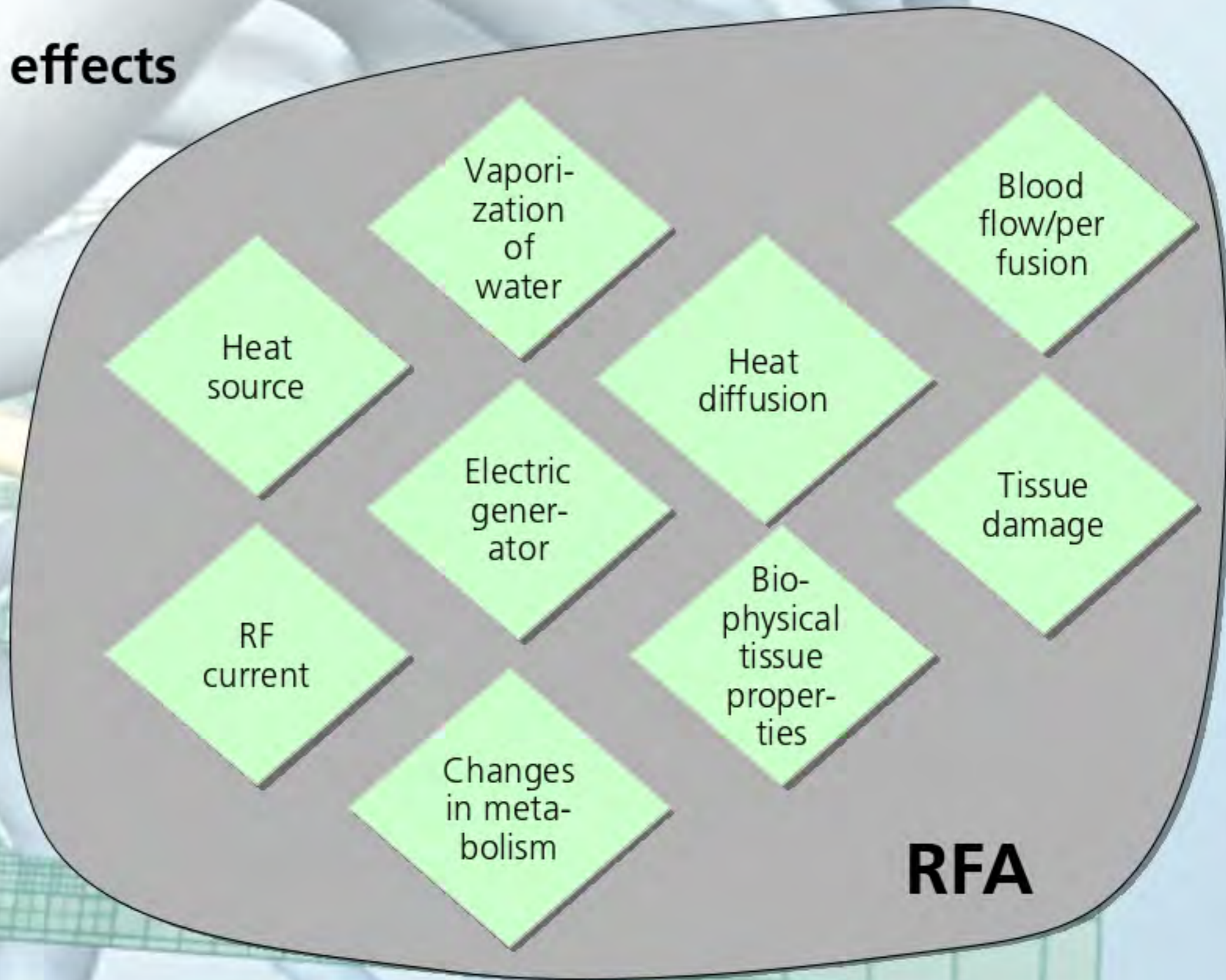
local vessels

# Radiofrequency ablation

# Bio-physical effects during RFA

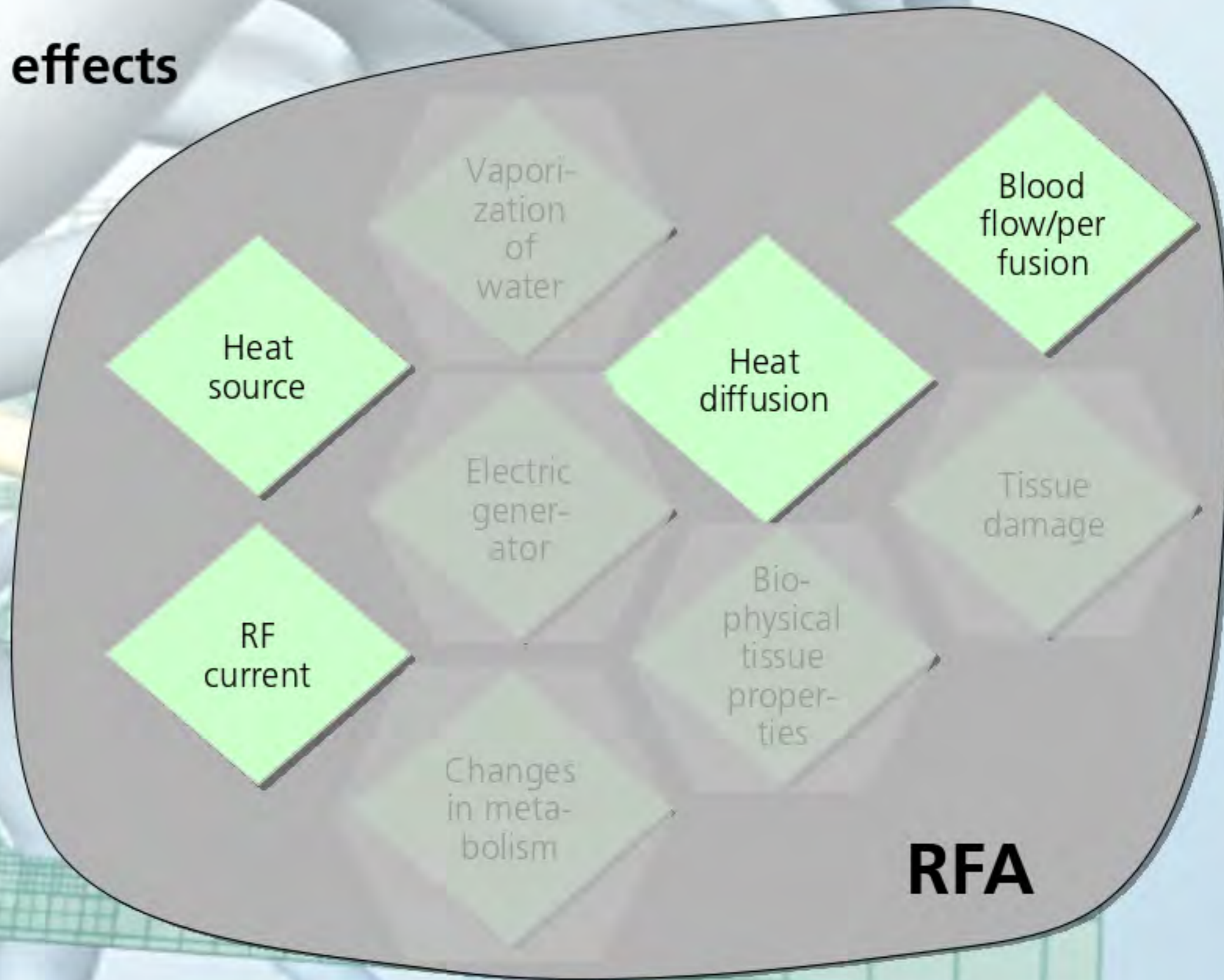


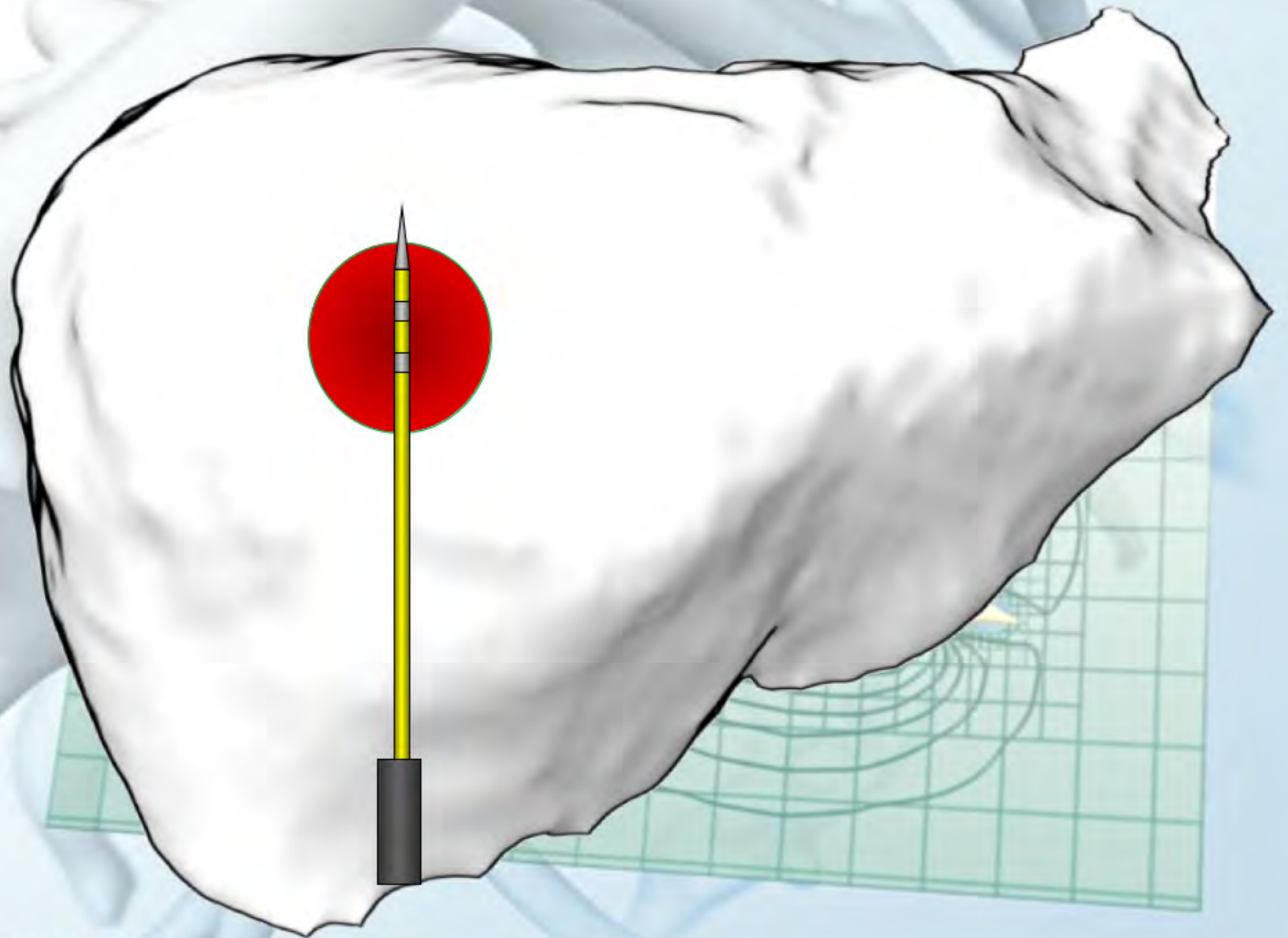
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# Modeling RF Ablation

Electrostatic equation

$$-\operatorname{div}(\sigma(x)\nabla\phi(x)) = 0$$

# Modeling RF Ablation

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boundary condition at electrodes  $\phi(x) = \pm 1$

+ boundary condition at outer boundary of computational domain

# Modeling RF Ablation

Electrostatic equation

$$-\text{div}(\sigma(x) \nabla \phi(x)) = 0$$

Electric potential  
[V]

boundary condition at electrodes

$$\phi(x) =$$

+ boundary condition at outer boundary of computational domain

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Electric conductivity  
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$$\rho(x)c(x)\partial_t T(t,x) - \text{div}(\lambda(x)\nabla T(t,x)) = \sigma(x)|\nabla \phi(x)|^2 - \nu(x)(T(t,x) - T_b)$$

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[J/s m<sup>3</sup> K]

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Heat capacity [J/kg K]

Relative perfusion rate [J/s m<sup>3</sup> K]

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Density  
[kg/m<sup>3</sup>]

Body temperature  
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[W/K m]

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Electrostatic equation

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**Partial derivative** = directional derivative in a coordinate direction

$$u(x_1, x_2, x_3) \quad \frac{\partial u}{\partial x_1} = \partial_{x_1} u = \text{slope of } u \text{ in direction } x_1$$

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**Gradient** = vector of all partial derivatives

$$\nabla u = (\partial_{x_1} u, \partial_{x_2} u, \partial_{x_3} u)^t$$

- ▶ Points in direction of steepest ascent
- ▶ Perpendicular to level lines

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**Electrostatic equation in 1D**

$$-(\sigma \phi')' = 0$$

**Electrostatic equation in 2D**

$$-\partial_{x_1}(\sigma \partial_{x_1} \phi) - \partial_{x_2}(\sigma \partial_{x_2} \phi) = 0$$

# 2

## Discretizing The Electrostatic Equation

# Solving Partial Differential Equations With a Computer

1

- **Discretize the computational domain**

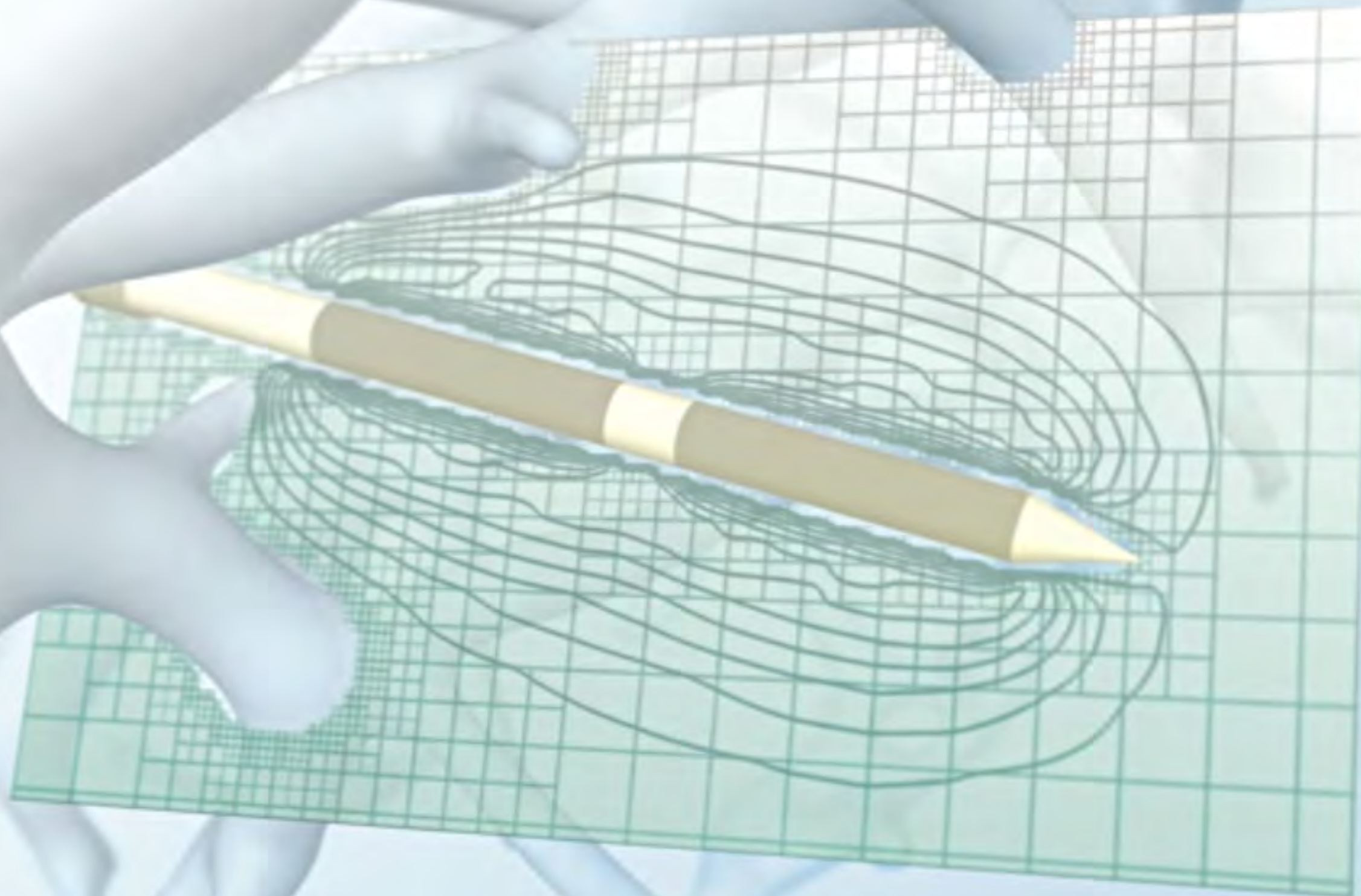
2

- **Discretize the differential operators**

3

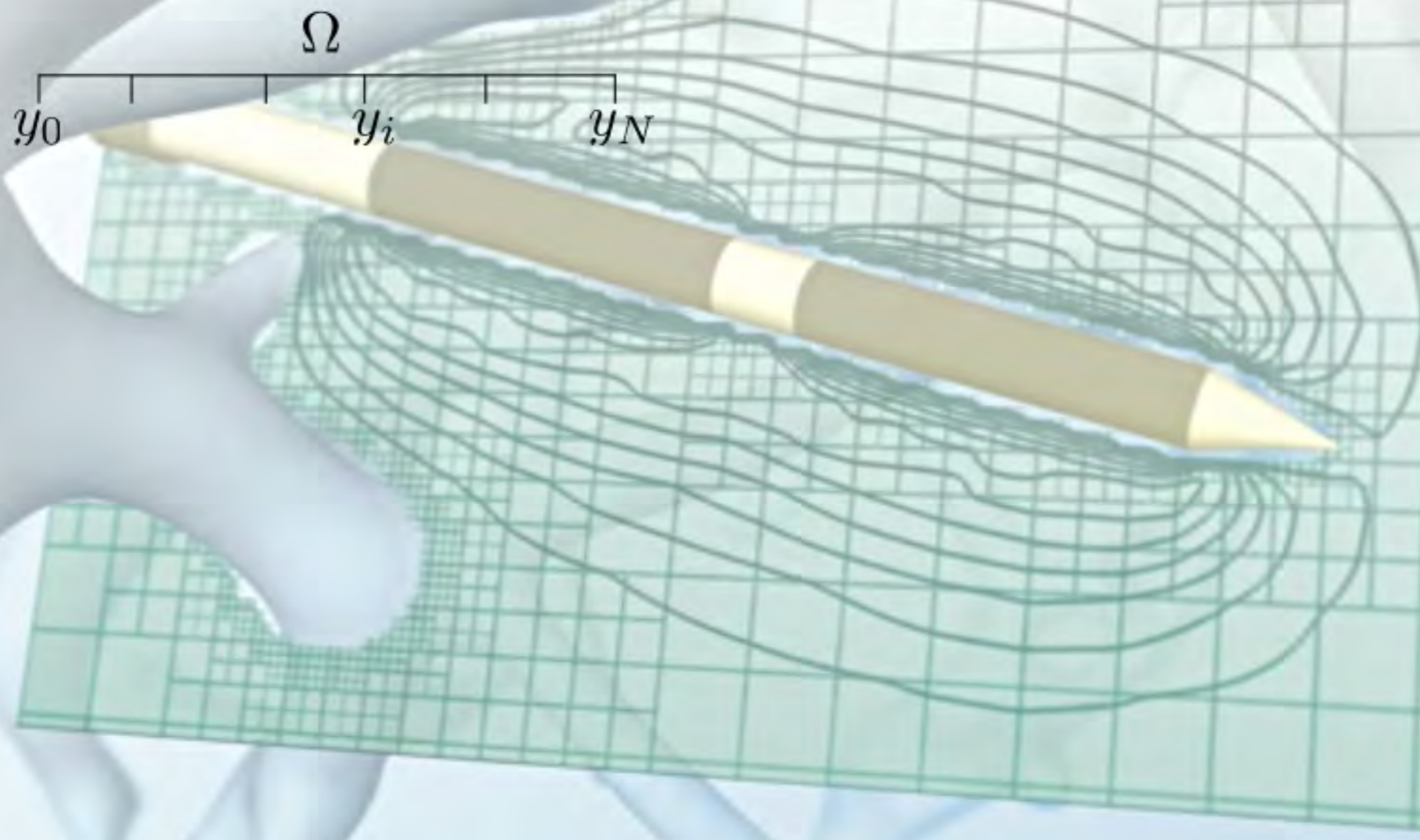
- **Solve a system of algebraic equations**

# Discretizing Computational Domains



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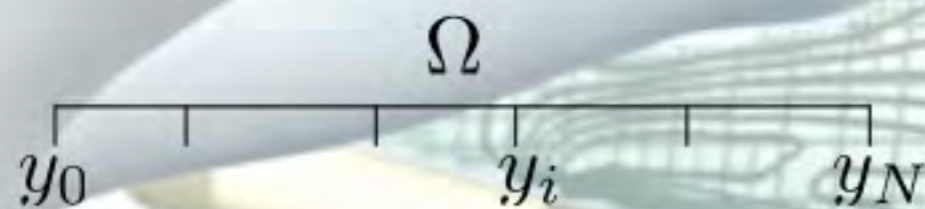
- › One space dimension



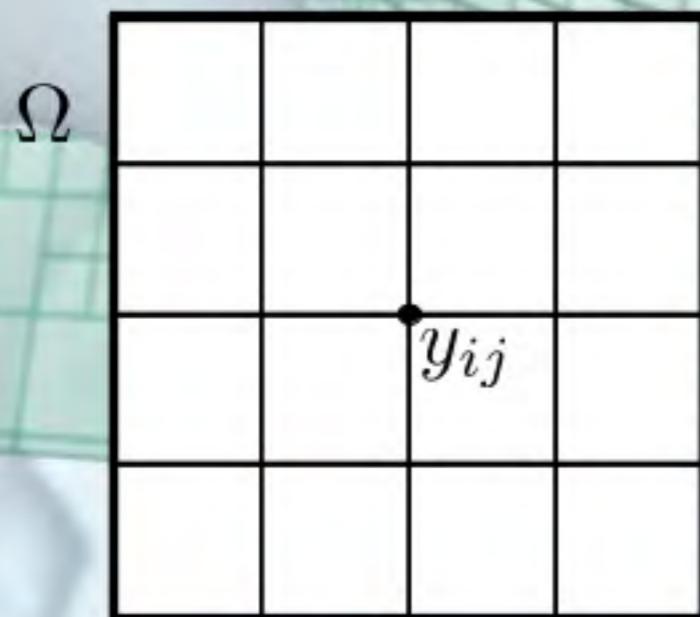


# Discretizing Computational Domains

- › One space dimension

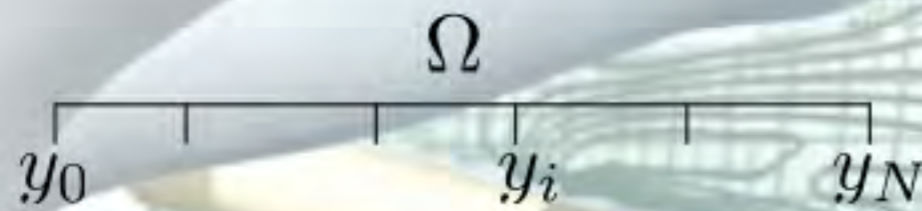


- › Two space dimensions

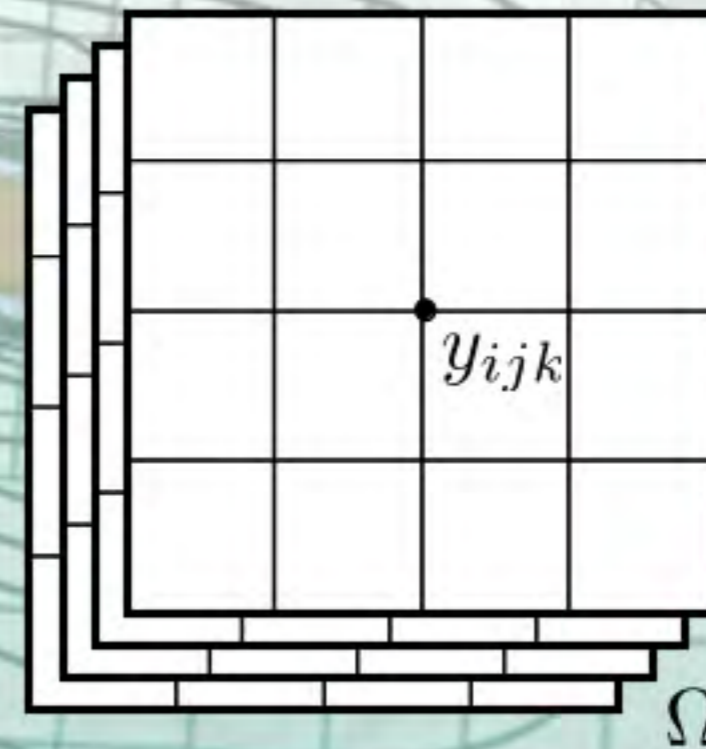


# Discretizing Computational Domains

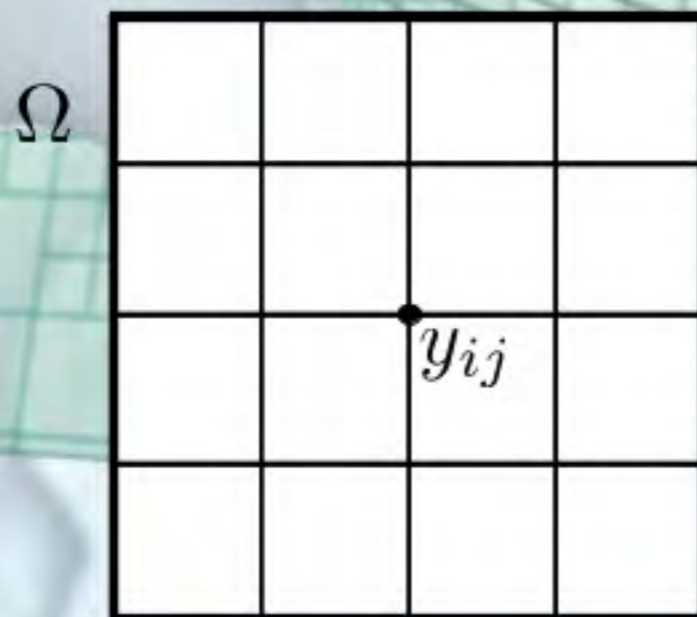
› One space dimension



› Three space dimensions



› Two space dimensions



# Discretizing Differential Operators

Finite Difference Method (FDM)

Finite Element Method (FEM)

# Discretizing Differential Operators

## Finite Difference Method (FDM)

- › Difference quotients

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- › Not very flexible w.r.t geometry of domain

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- › Shape functions of variable smoothness

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- › Shape functions of variable smoothness
- › Difficult to implement
- › Very flexible w.r.t geometry of domain
- › Rich mathematical framework

# The Finite Difference Method

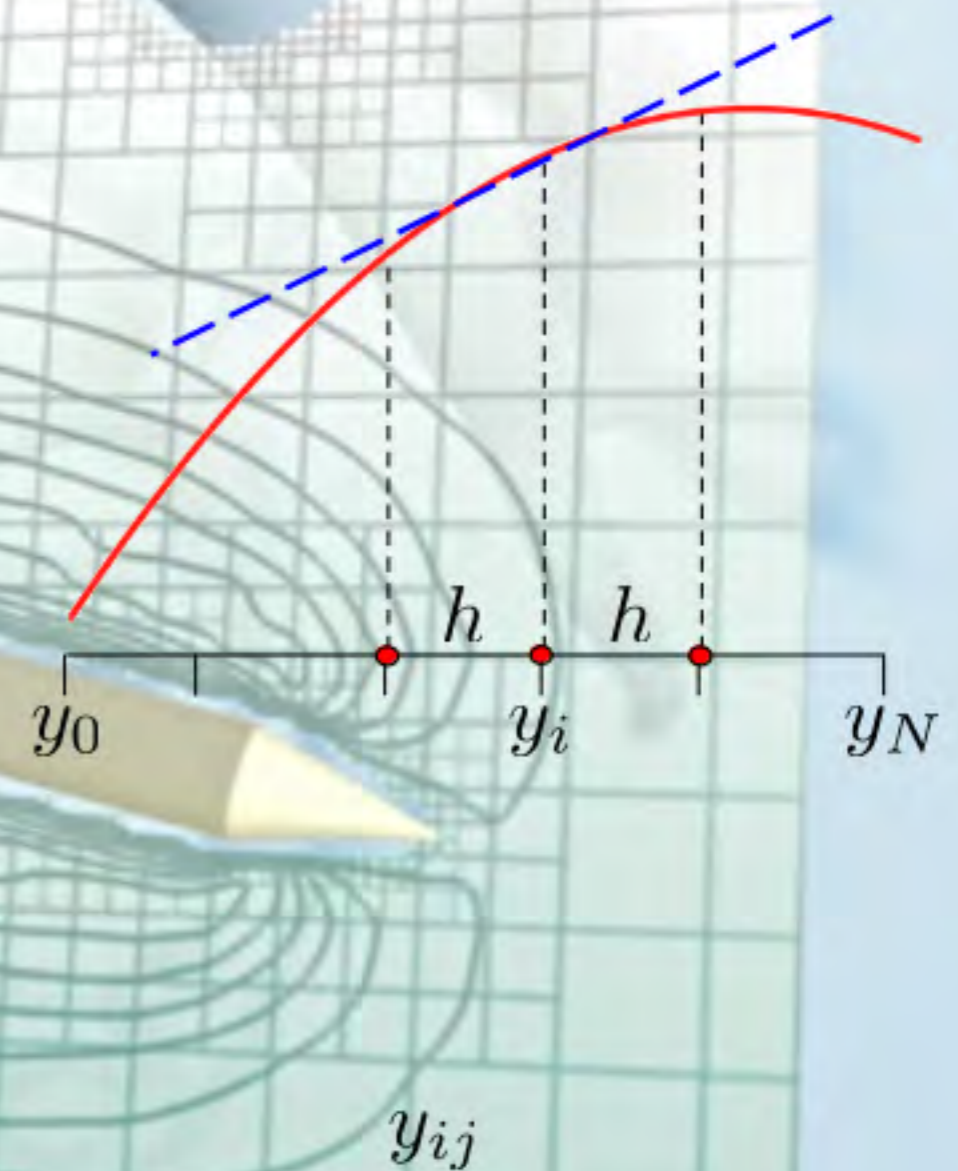
# FDM Discretization of Derivatives

- › Approximate slope of function by secant

 $y_0$  $y_i$  $y_N$  $y_{ij}$

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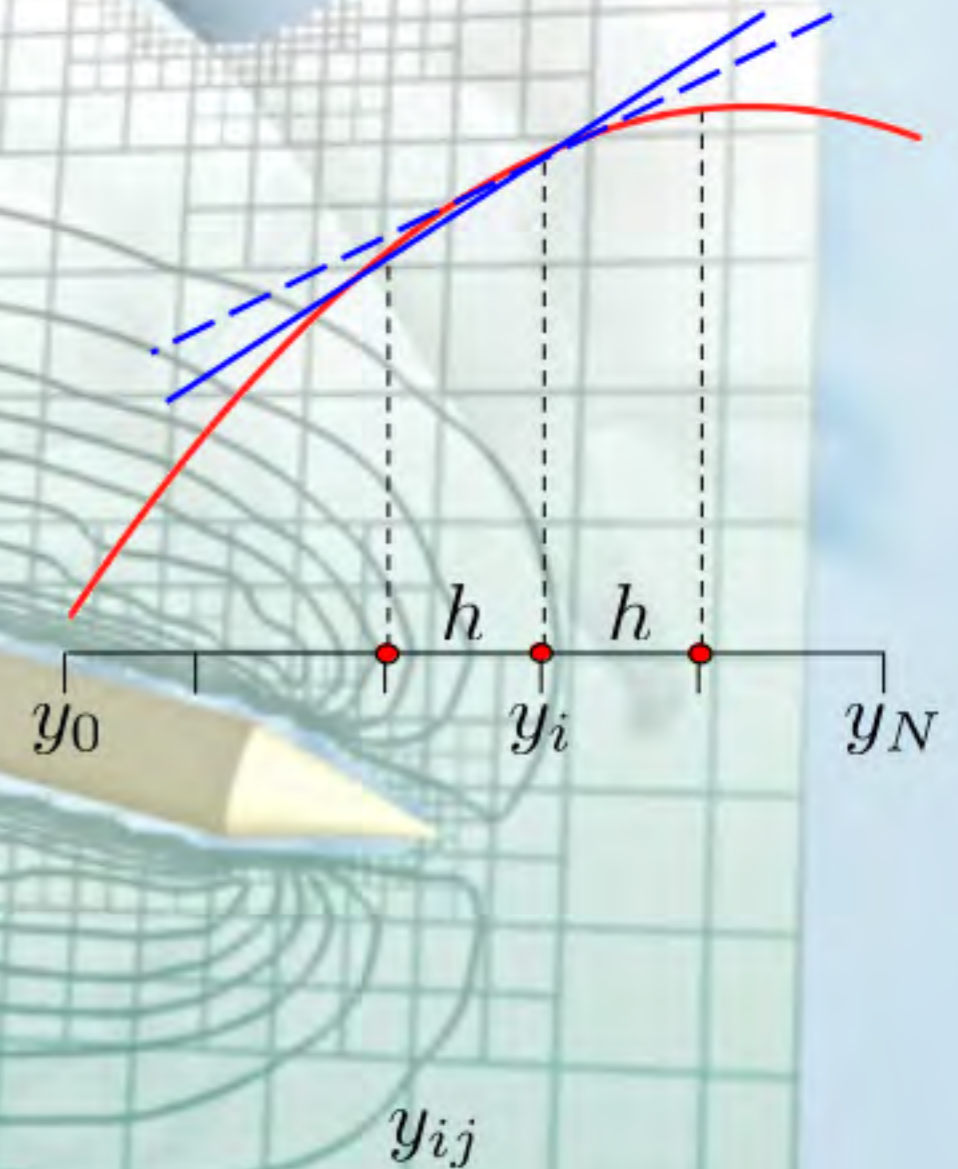


# FDM Discretization of Derivatives

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Backward difference

$$u'(y_i) \approx \frac{u_i - u_{i-1}}{h}$$



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**Central difference**  $u'(y_i) \approx \frac{u_{i+1} - u_{i-1}}{2h}$

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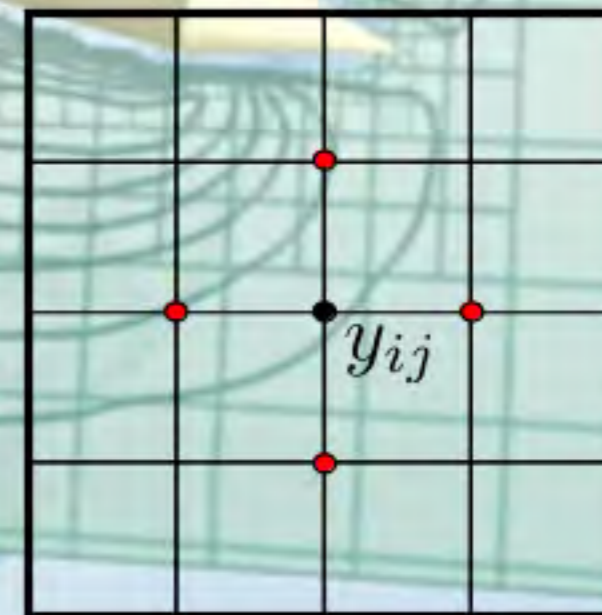
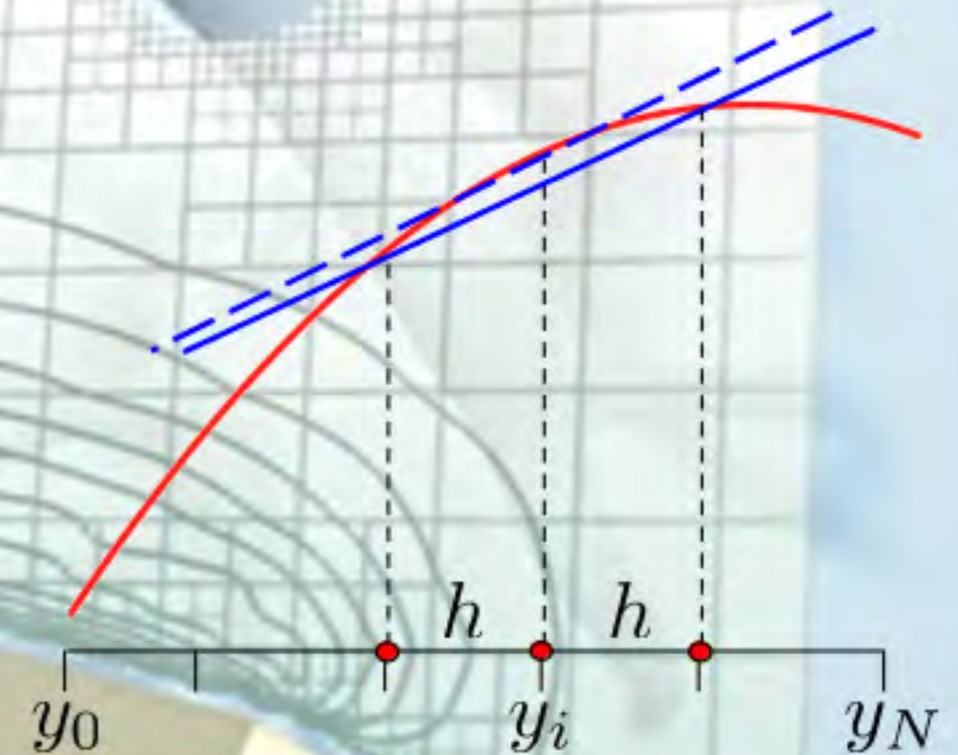
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**Central difference**  $u'(y_i) \approx \frac{u_{i+1} - u_{i-1}}{2h}$

- › Higher dimensions: partial derivatives

$$\partial_{x_1} u(y_{ij}) \approx \frac{u_{i,j} - u_{i-1,j}}{h}$$

$$\partial_{x_2} u(y_{ij}) \approx \frac{u_{i,j+1} - u_{i,j}}{h}$$



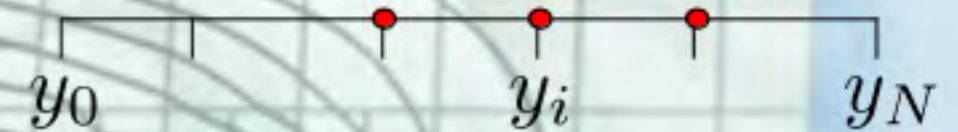
# Discrete Differential Operators

# Discrete Differential Operators

- › Discrete Laplacian in 1D

$$\Delta u(y_i) = u''(y_i) \approx \frac{u_{i-1} - 2u_i + u_{i+1}}{h^2}$$

3 point stencil

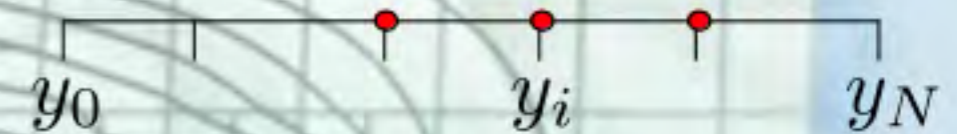


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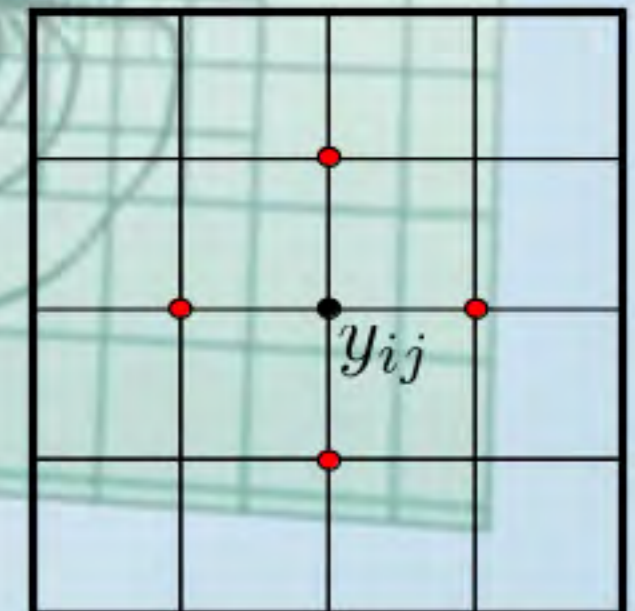


- › Discrete Laplacian in 2D

$$\Delta u(y_{ij}) = \partial_{x_1 x_1} u(y_{ij}) + \partial_{x_2 x_2} u(y_{ij})$$

$$\approx \frac{u_{i-1,j} + u_{i+1,j} + u_{i,j-1} + u_{i,j+1} - 4u_{ij}}{h^2}$$

5 point stencil

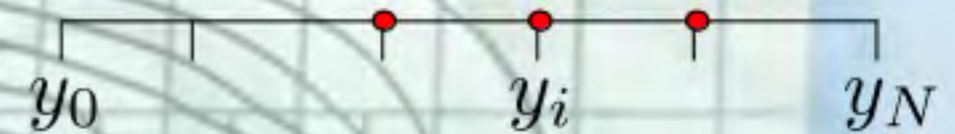


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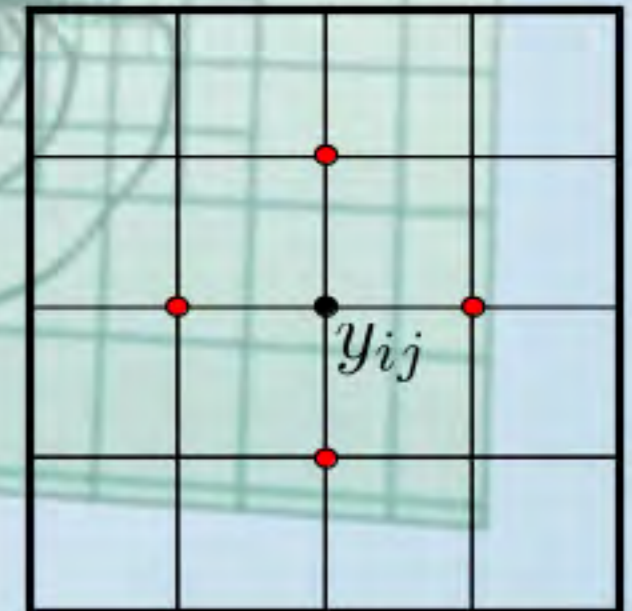
3 point stencil



- › Discrete Laplacian in 2D

$$\begin{aligned} \Delta u(y_{ij}) &= \partial_{x_1 x_1} u(y_{ij}) + \partial_{x_2 x_2} u(y_{ij}) \\ &\approx \frac{u_{i-1,j} + u_{i+1,j} + u_{i,j-1} + u_{i,j+1} - 4u_{ij}}{h^2} \end{aligned}$$

5 point stencil



- › Discrete Laplacian in 3D

7 point stencil



# Linear Systems of Equations

- › Continuous equation

$$-\operatorname{div}(\sigma(x)\nabla\phi(x)) = 0 \quad \text{for every point } x \in \Omega$$

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- › Linear system of equations

$$A\Phi = F$$

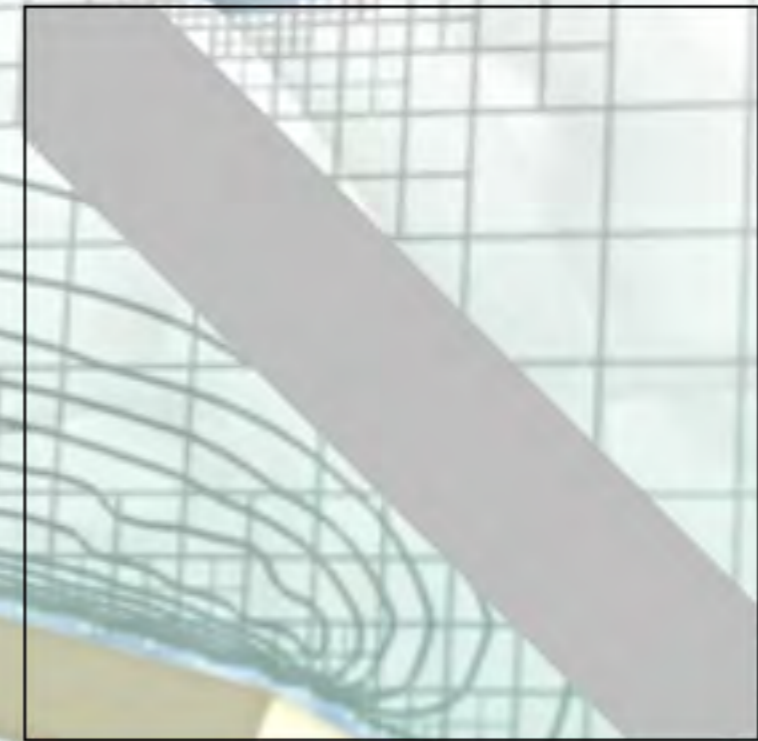
$$A \in \mathbb{R}^{(N+1) \times (N+1)}$$

$$\Phi, F \in \mathbb{R}^{(N+1)}$$



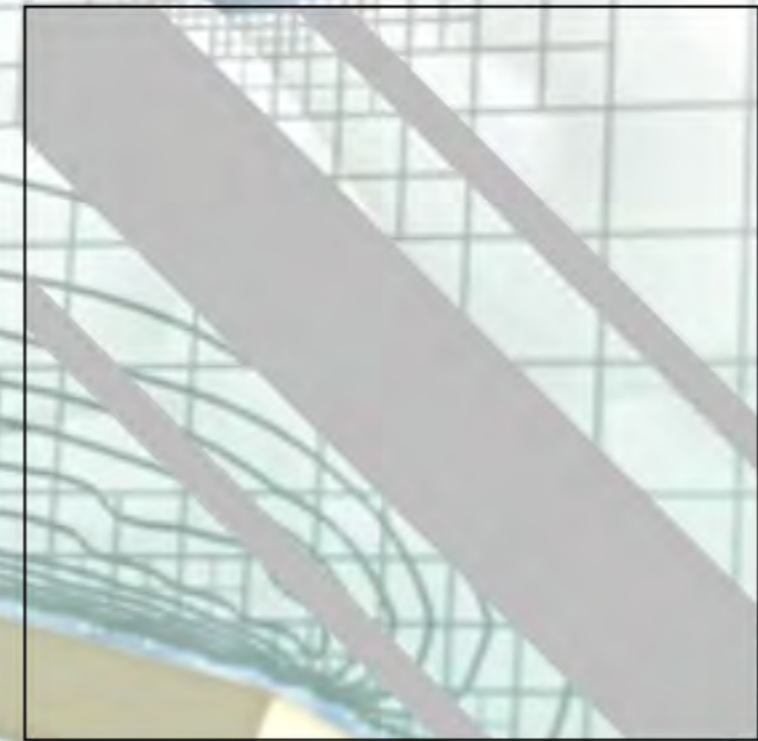
# Solving Linear System of Equations

- › 3-Band matrix in 1D



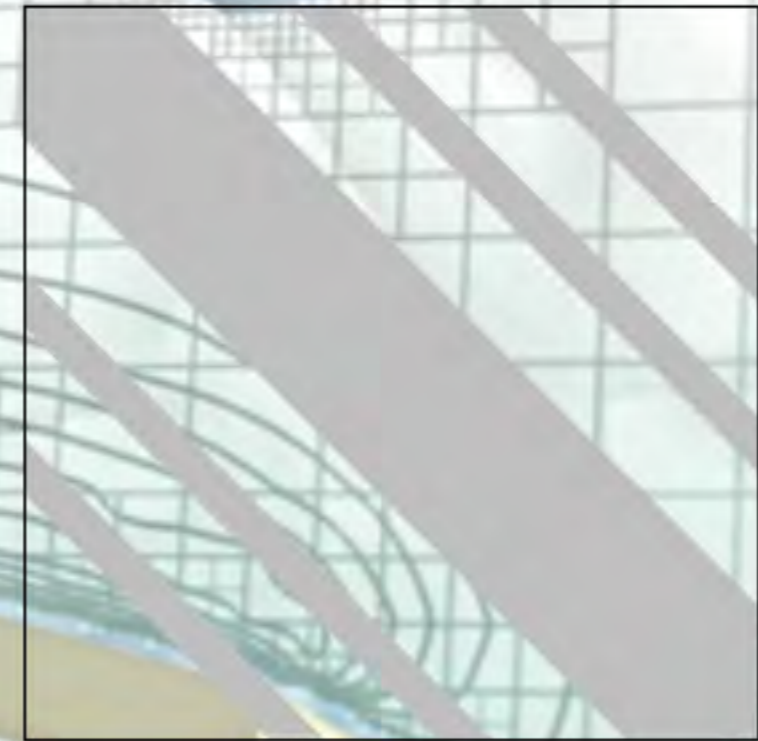
# Solving Linear System of Equations

- › 3-Band matrix in 1D
- › 5-Band matrix in 2D



# Solving Linear System of Equations

- › 3-Band matrix in 1D
- › 5-Band matrix in 2D
- › 7-Band matrix in 3D



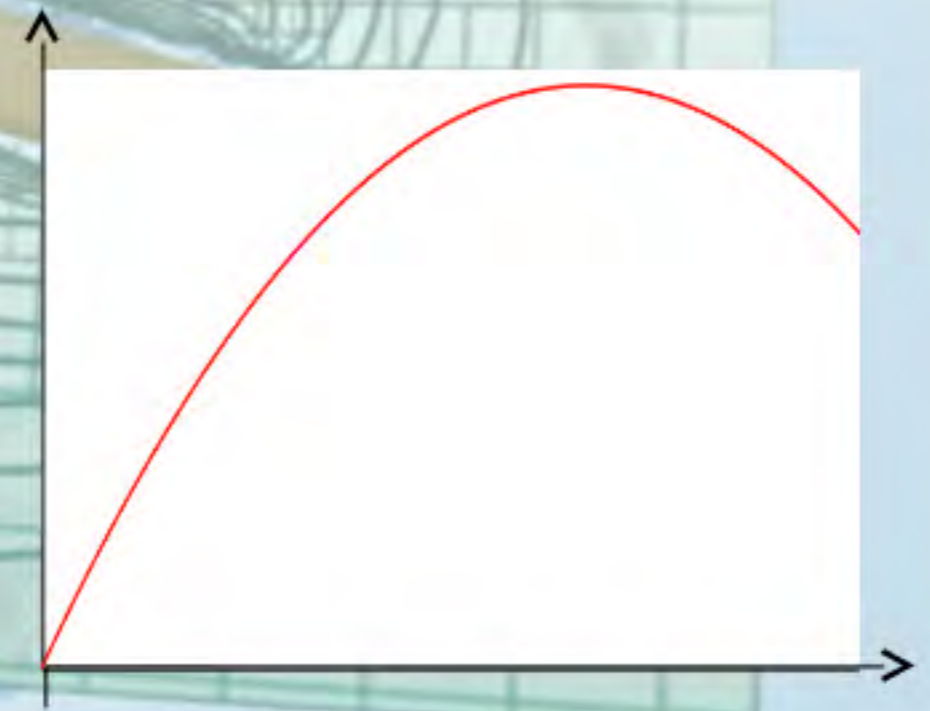
# Solving Linear System of Equations

- › 3-Band matrix in 1D
- › 5-Band matrix in 2D
- › 7-Band matrix in 3D
  
- › Matrix inversion:
  - » Direct method, e.g. Thomas algorithm
  - » Iterative method: e.g. Gauss-Seidel method, conjugate gradients



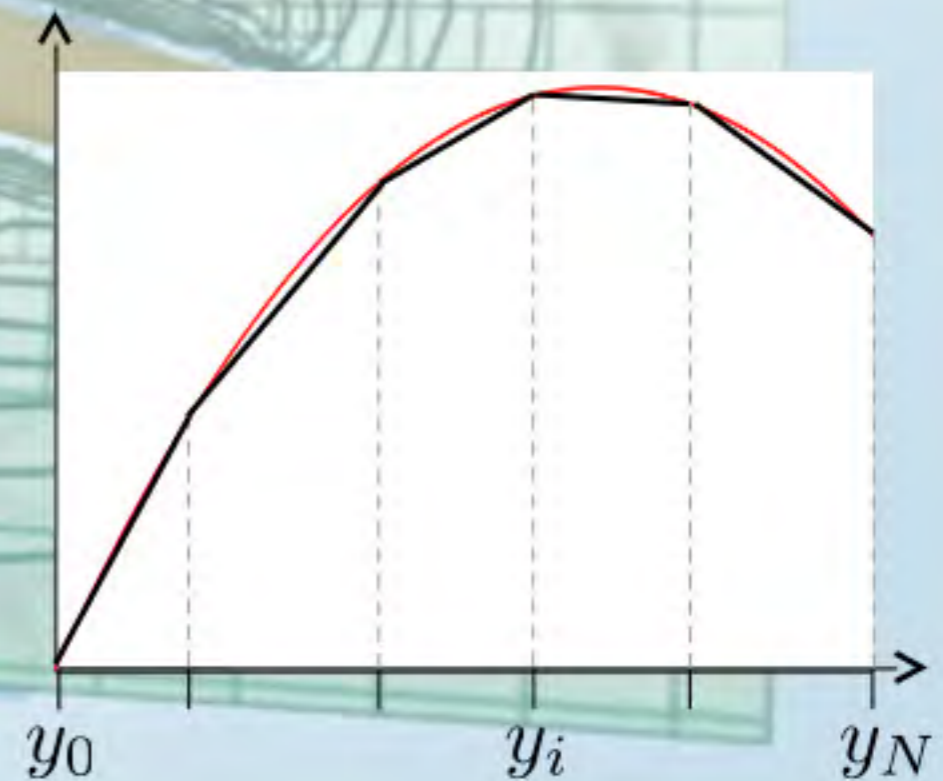
## Summary FDM

- › Discretize the domain by a regular grid
- › Approximate differential operators by difference quotients
- › System of linear equations
- › Solve by iterative or direct method
- › Obtain a numerical solution  $\phi_h$  of the PDE



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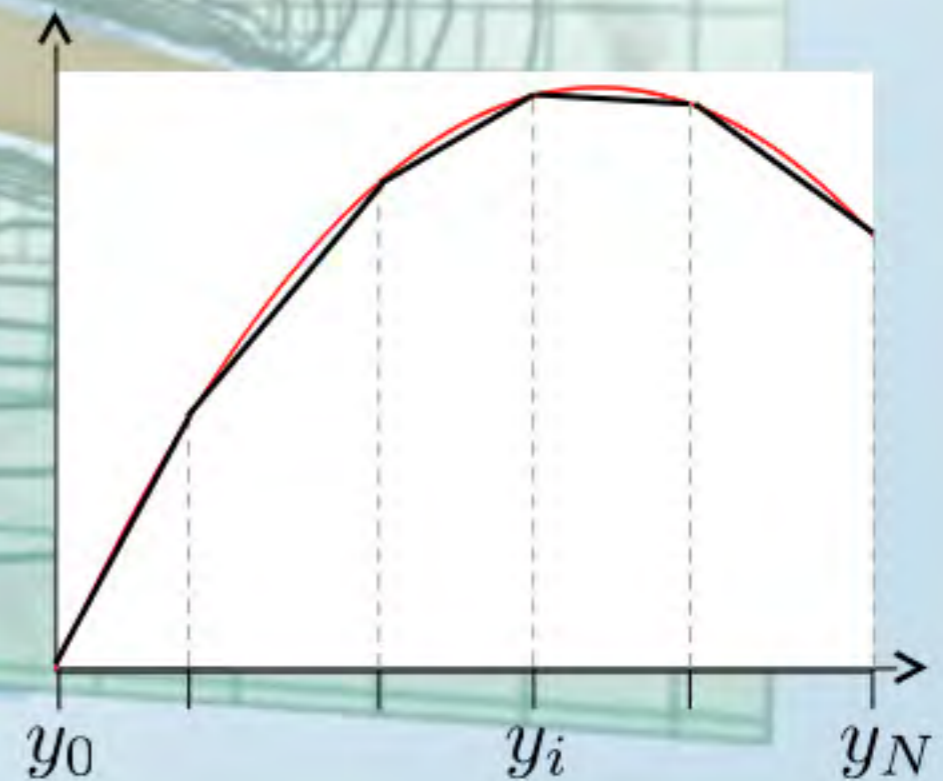
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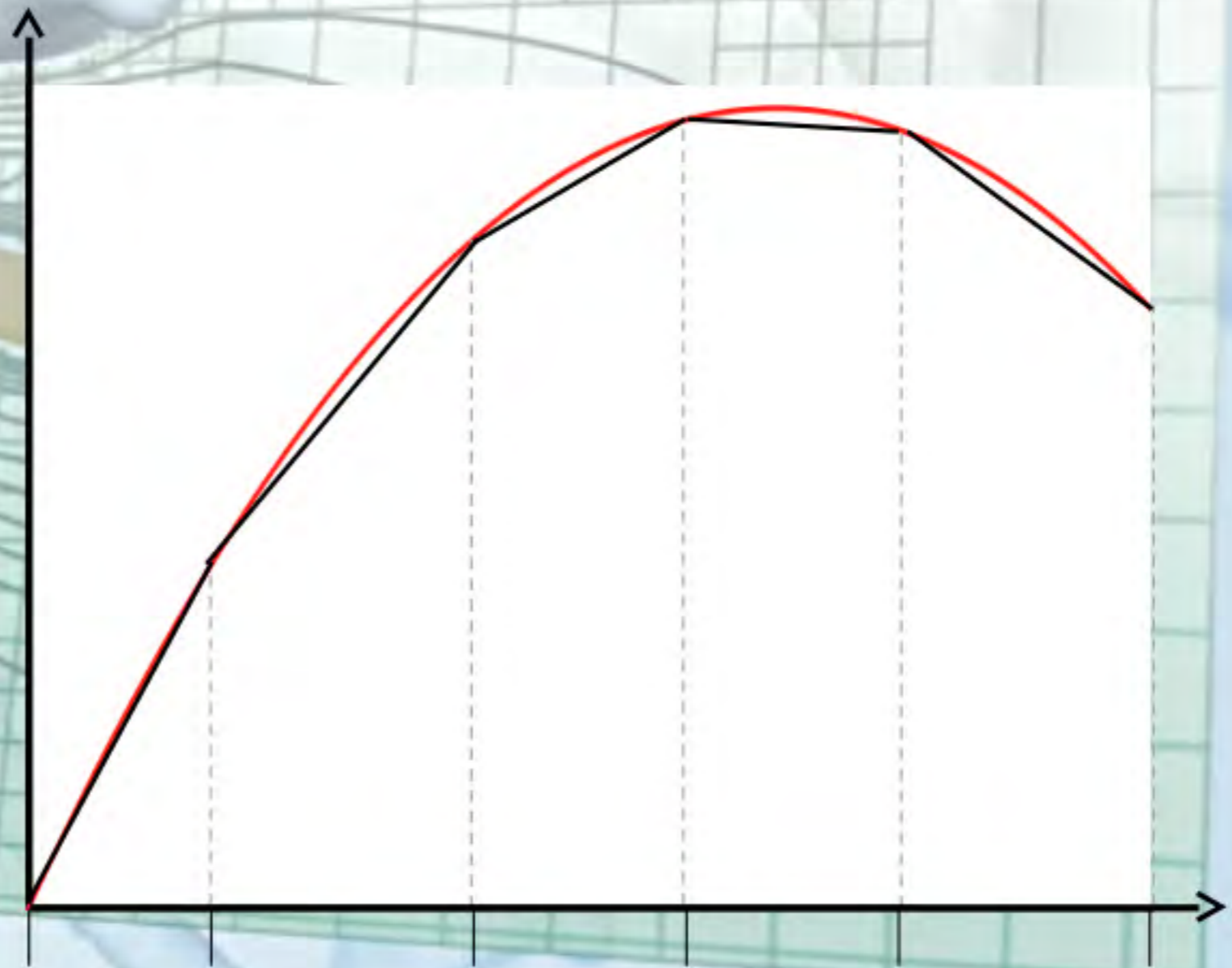
- › Accuracy  $\|\phi - \phi_h\|_{L^2} = Ch^2$



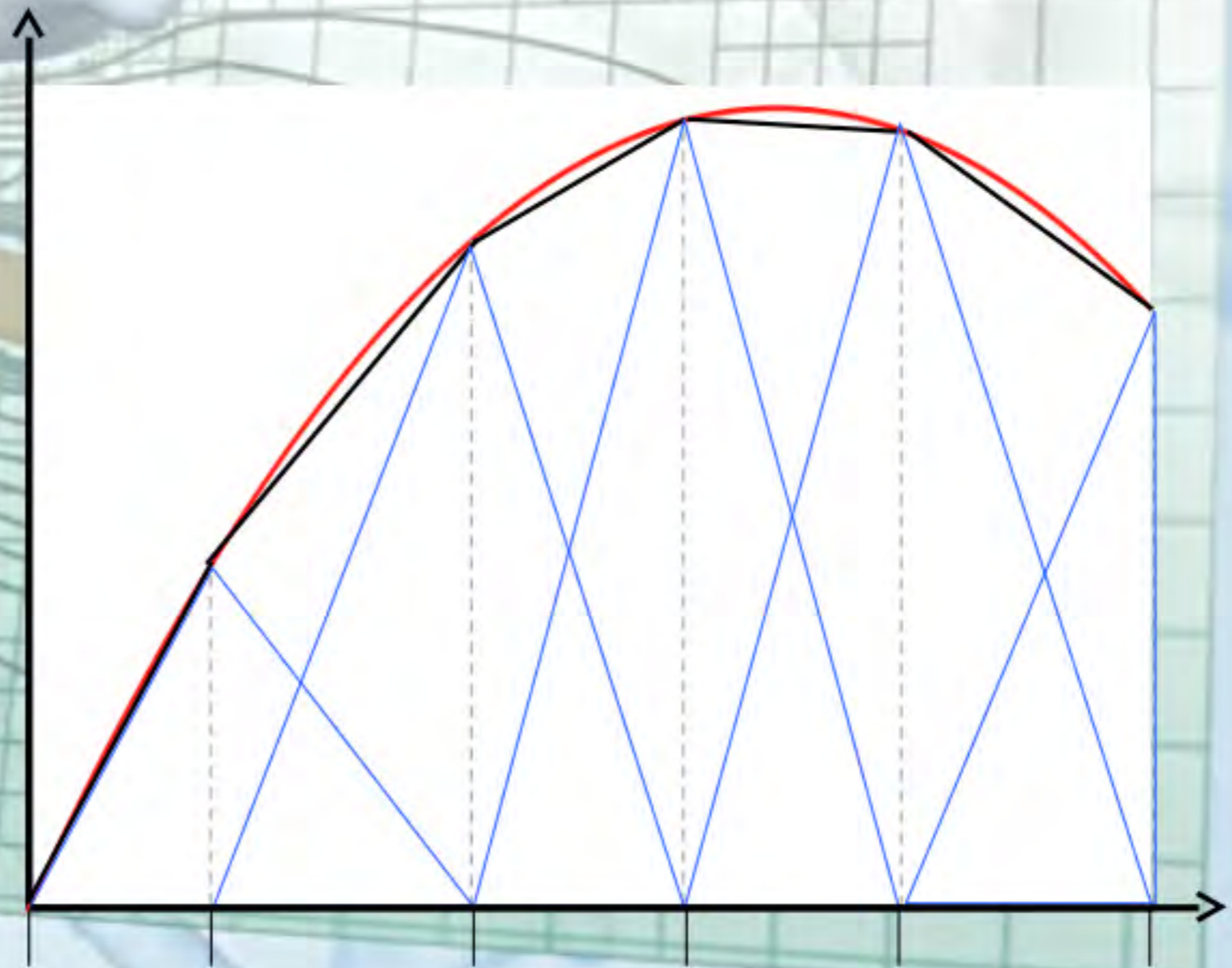
# The Finite Element Method



# Finite Element Shape Functions

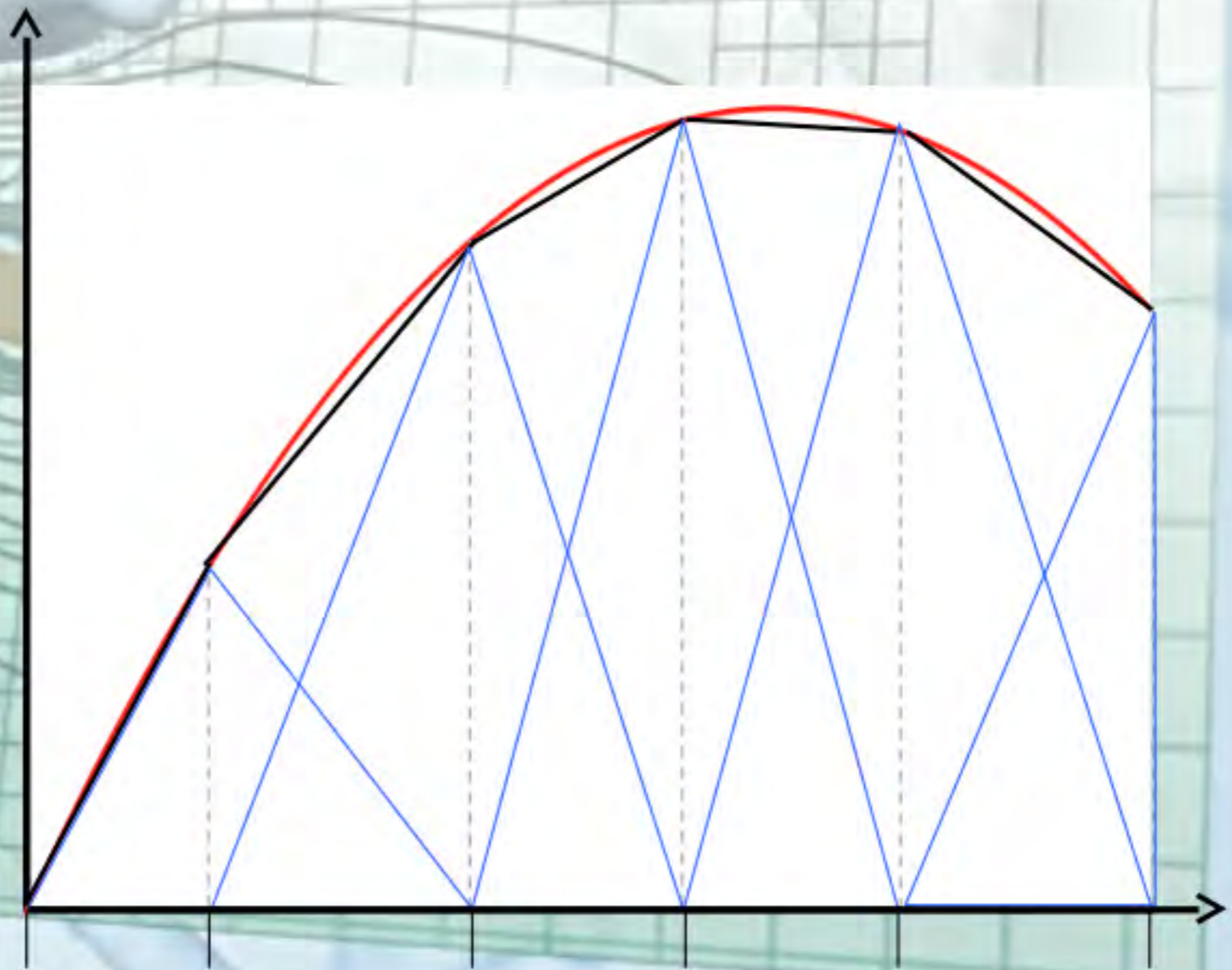


# Finite Element Shape Functions

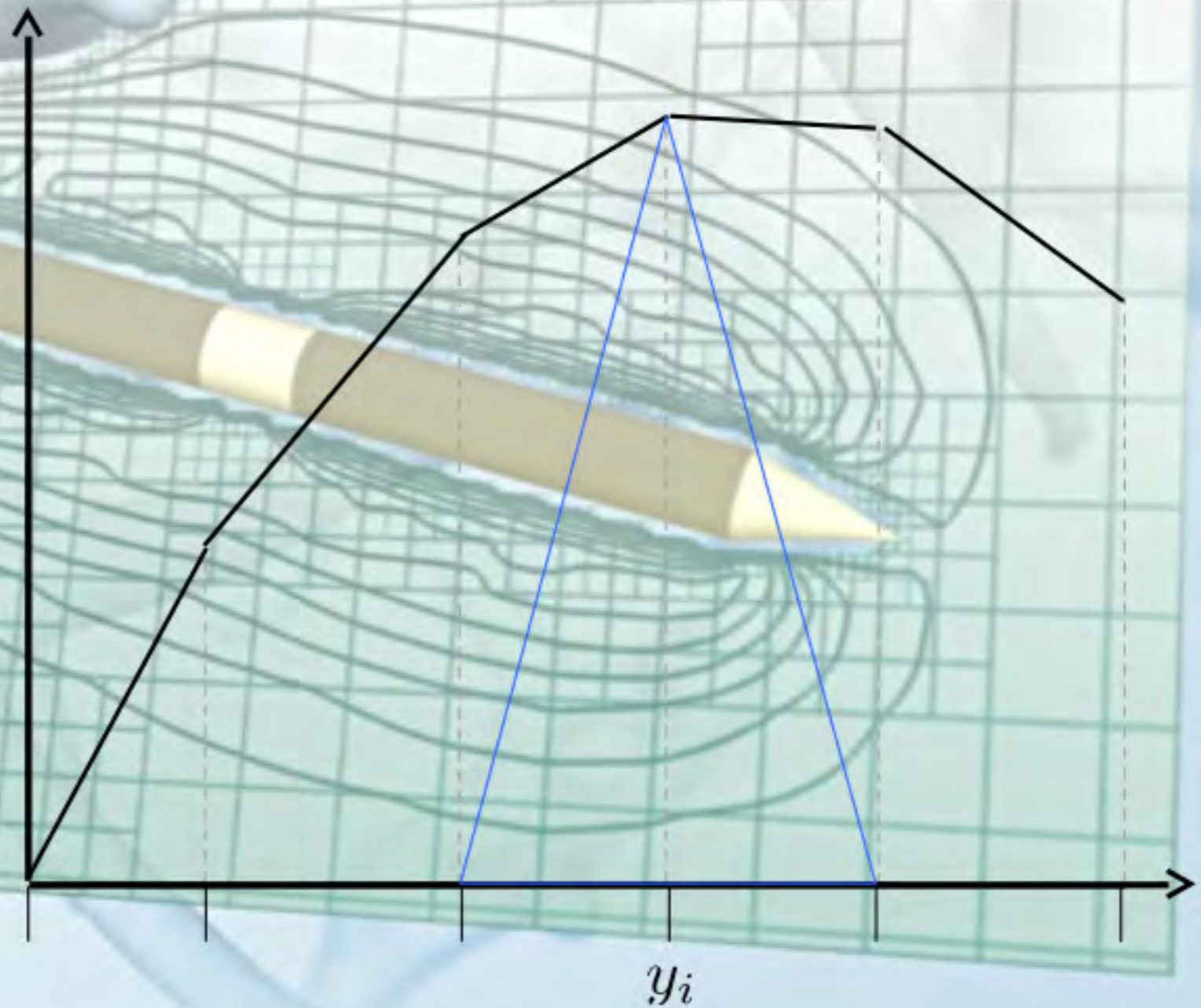


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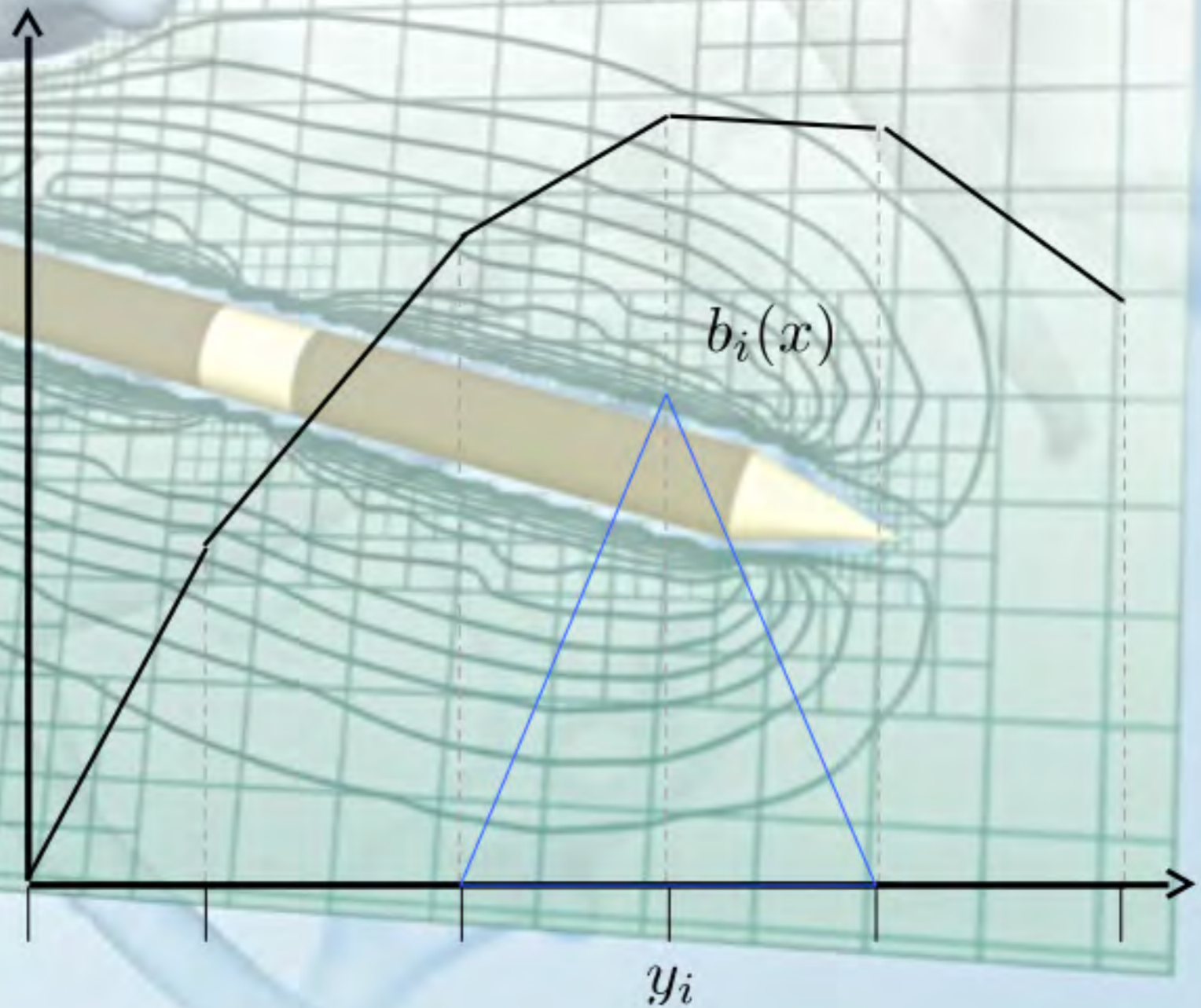
- › Compose the solution from predefined shape functions



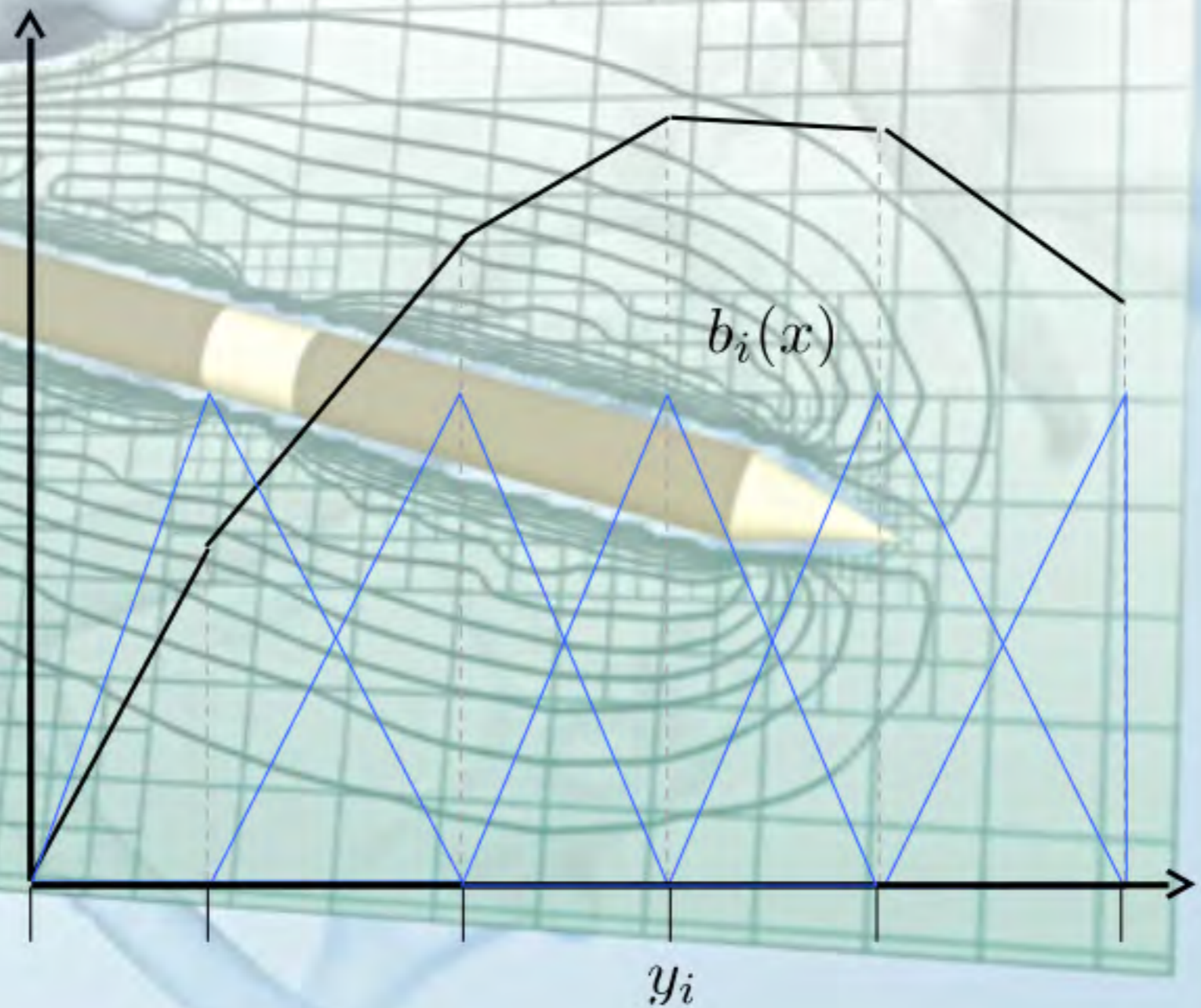
# Finite Element Shape Functions



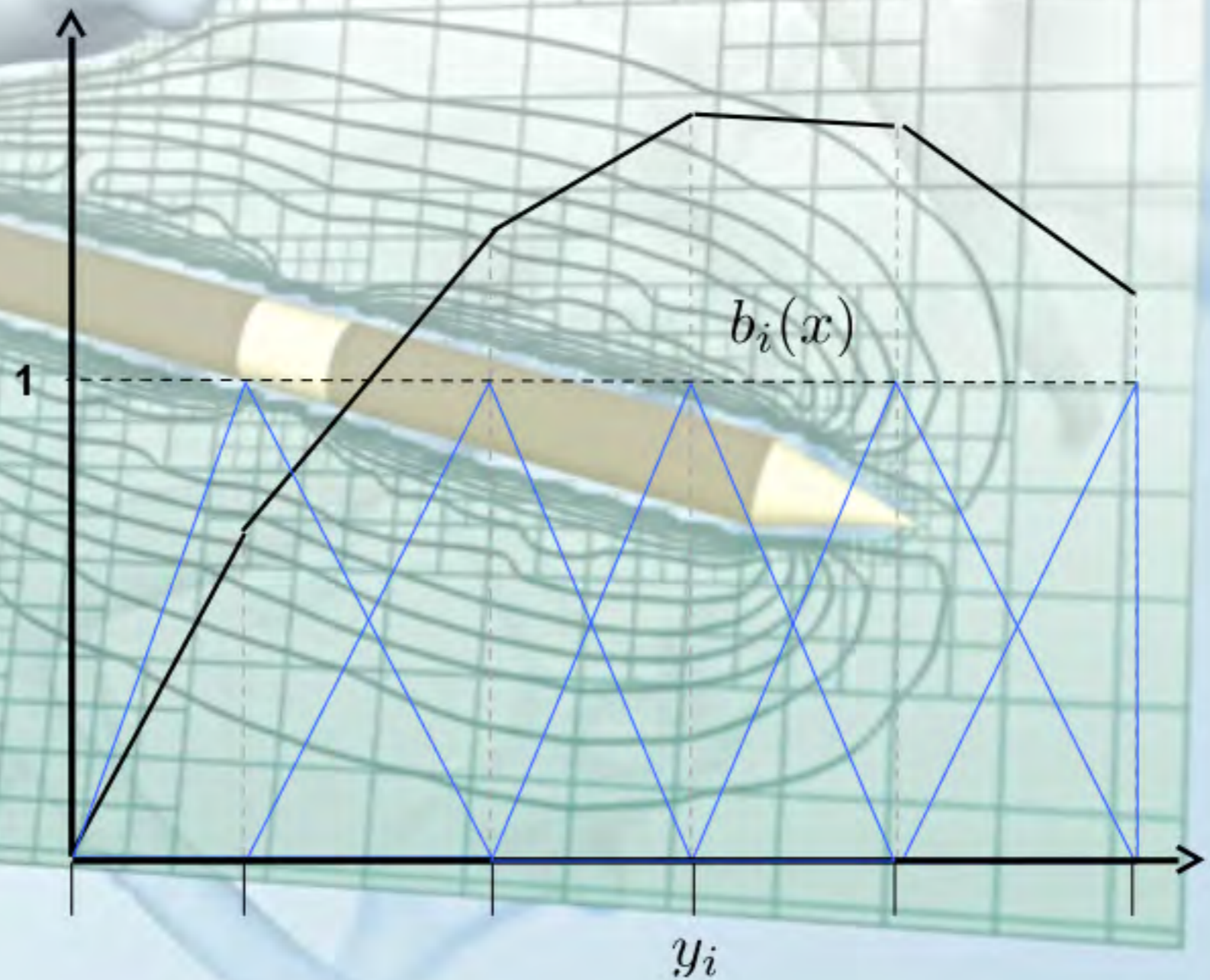
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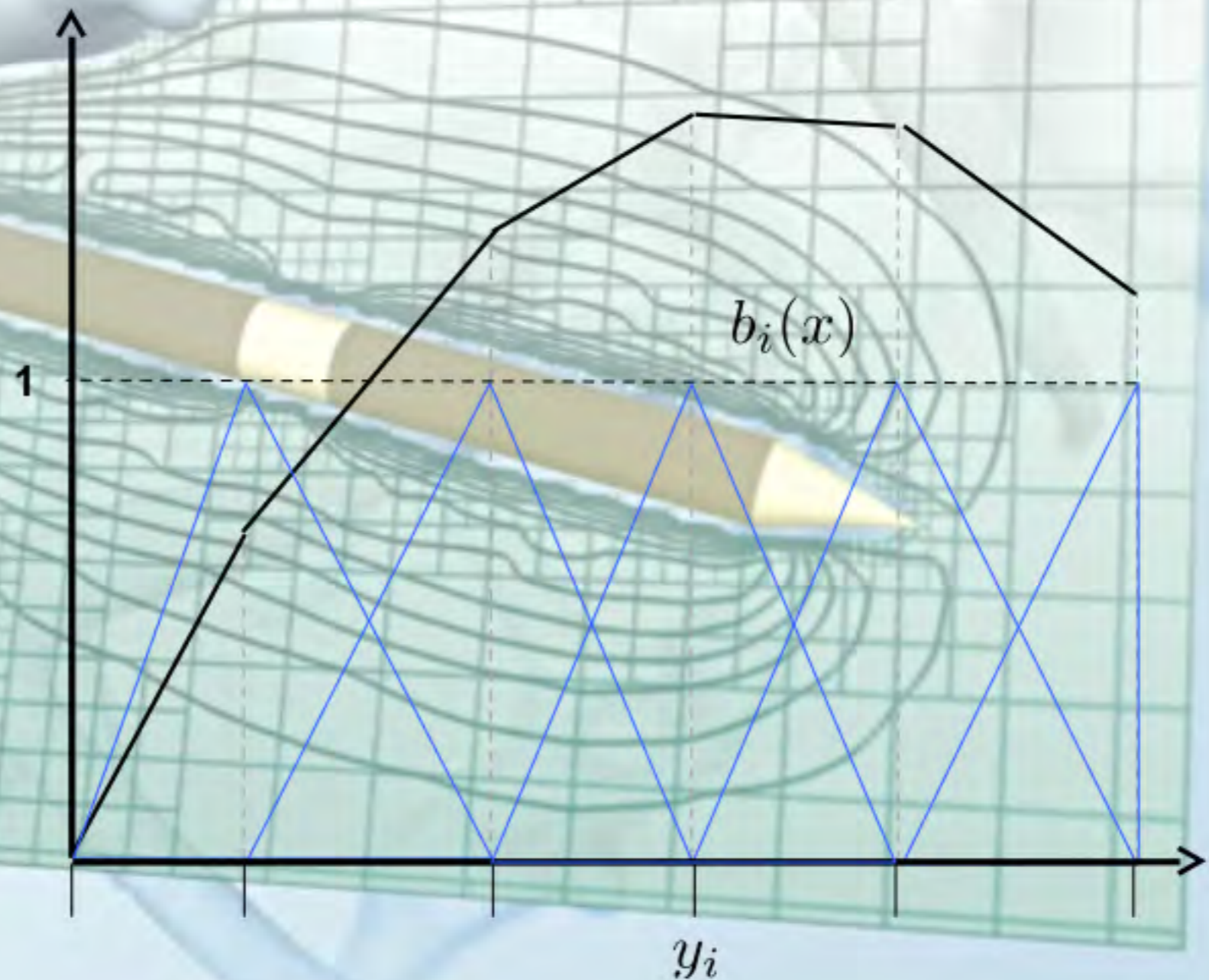
# Finite Element Shape Functions



# Finite Element Shape Functions

$$\phi(x) = \phi_0 b_0(x) + \phi_1 b_1(x) + \dots + \phi_N b_N(x)$$

$$= \sum_{i=0}^N \phi_i b_i(x)$$



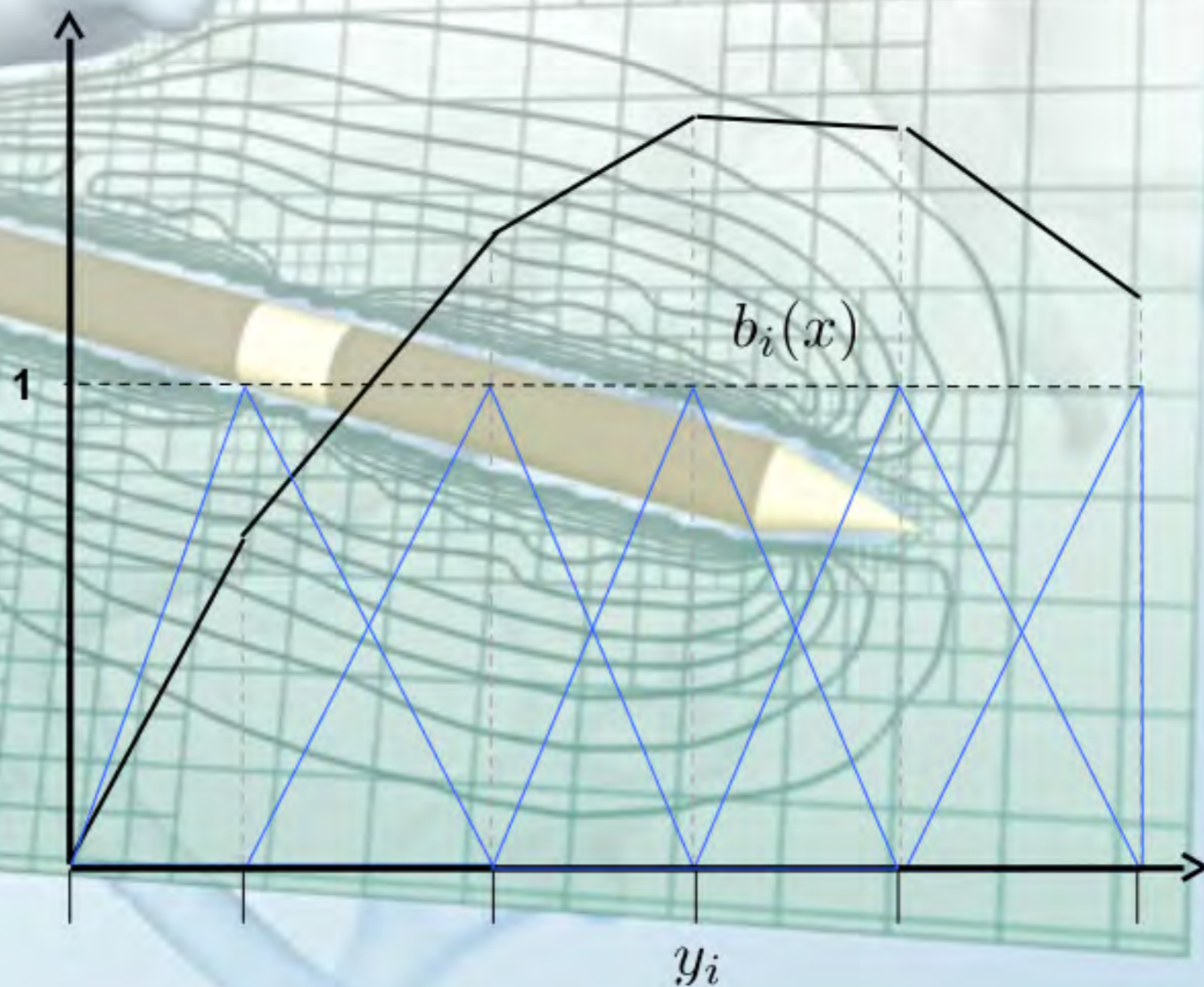


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Piecewise linear function

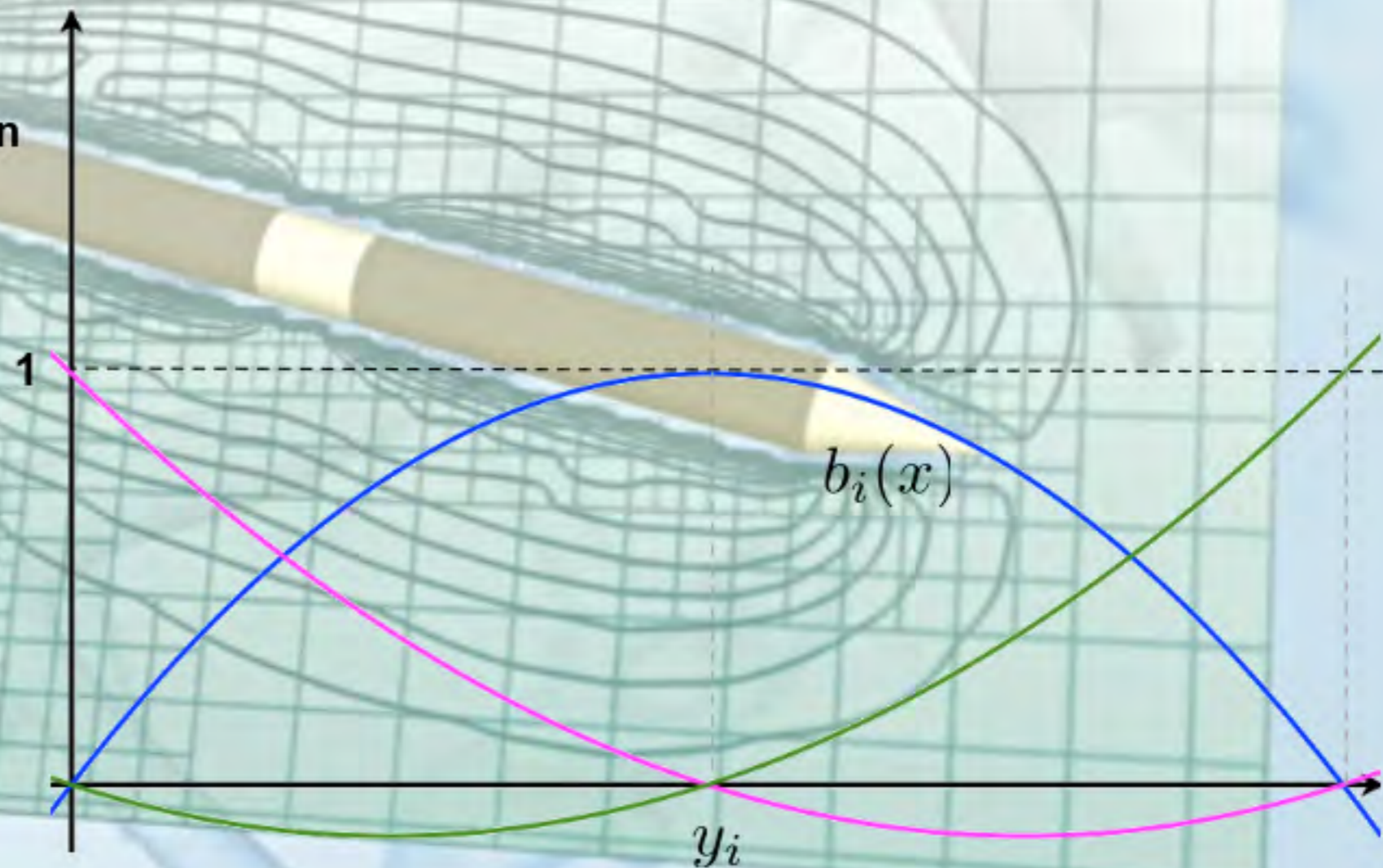


# Flexibility of Shape Functions

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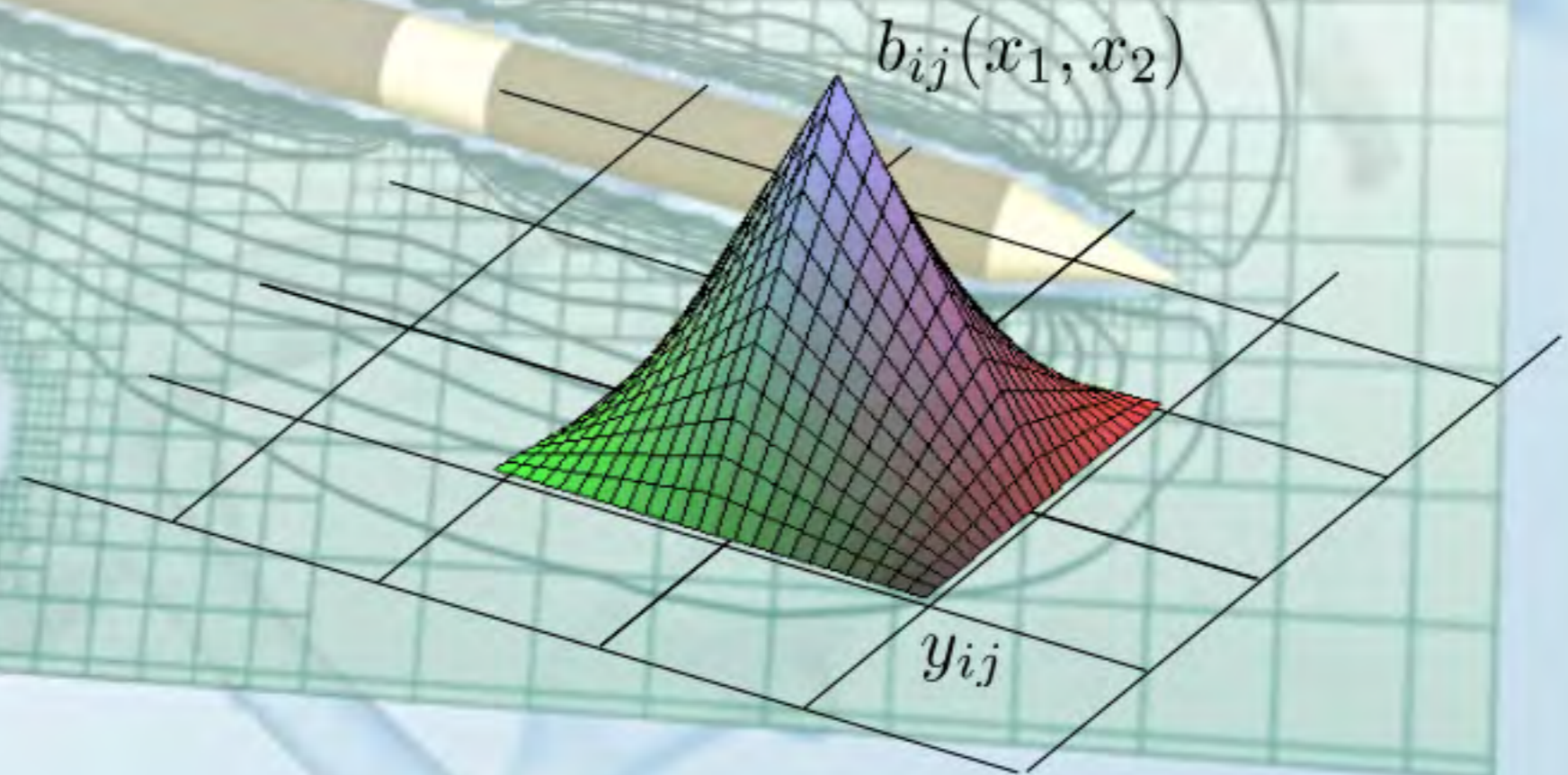
Piecewise polynomial function



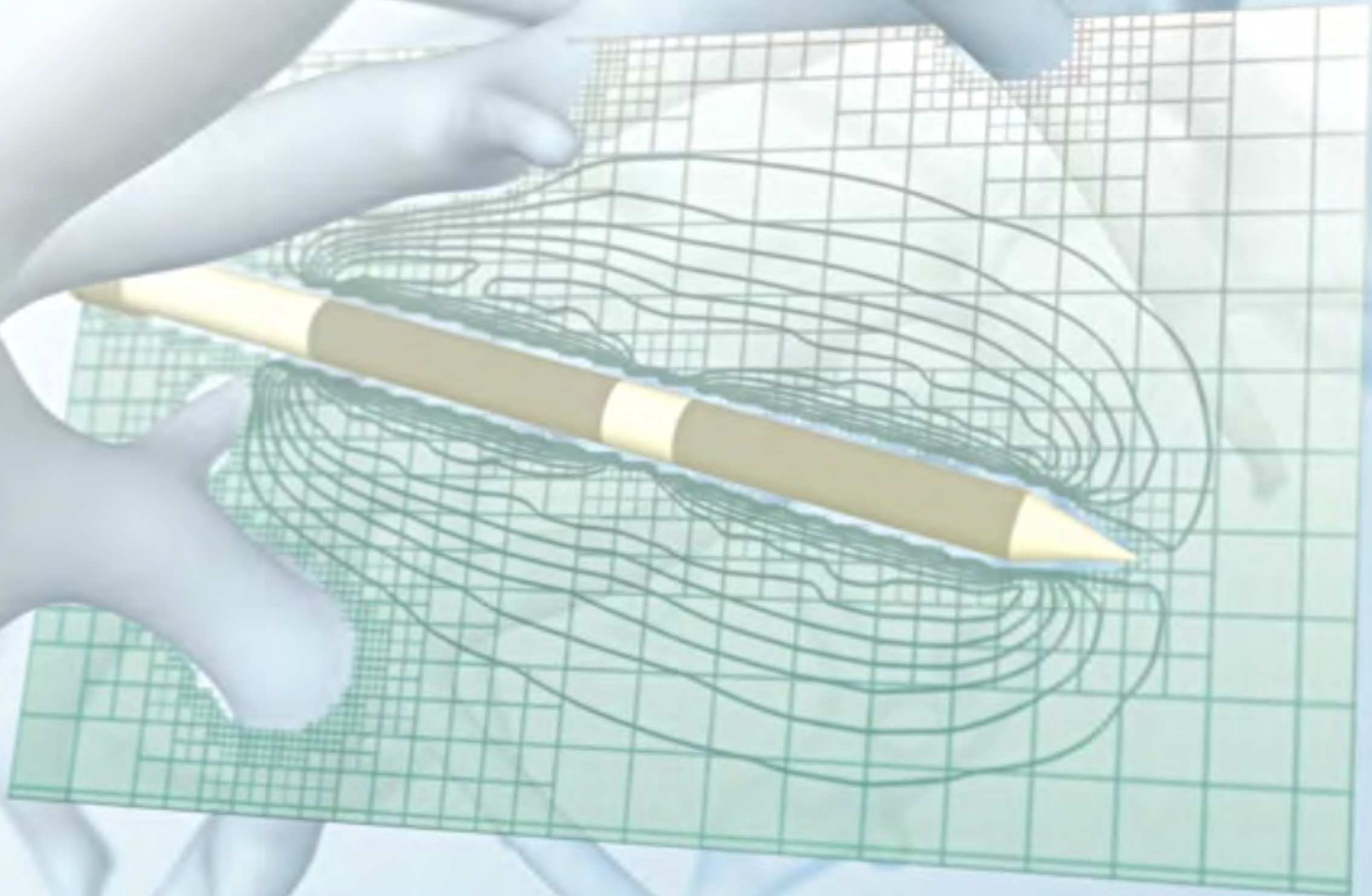
# Shape Functions in 2D

$$\phi(x) = \phi(x_1, x_2) = \sum_{i,j} \phi_{ij} b_{ij}(x_1, x_2)$$

**Piecewise bilinear function**



# Flexibility of Domain Discretization



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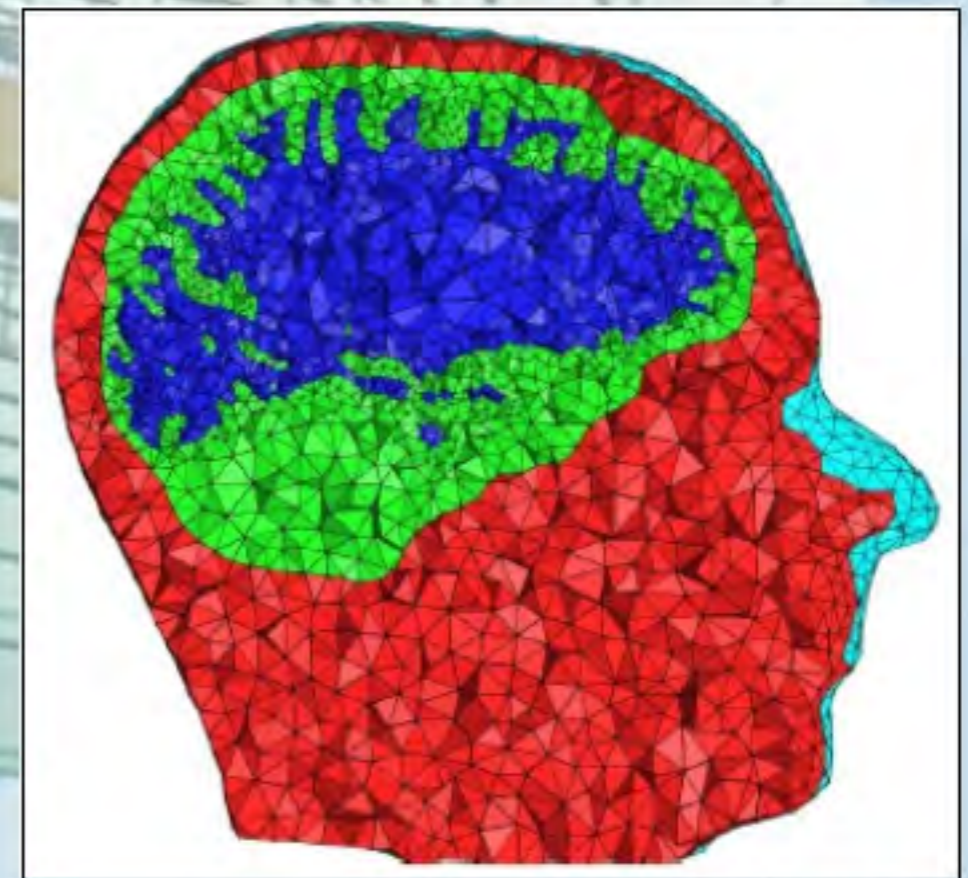
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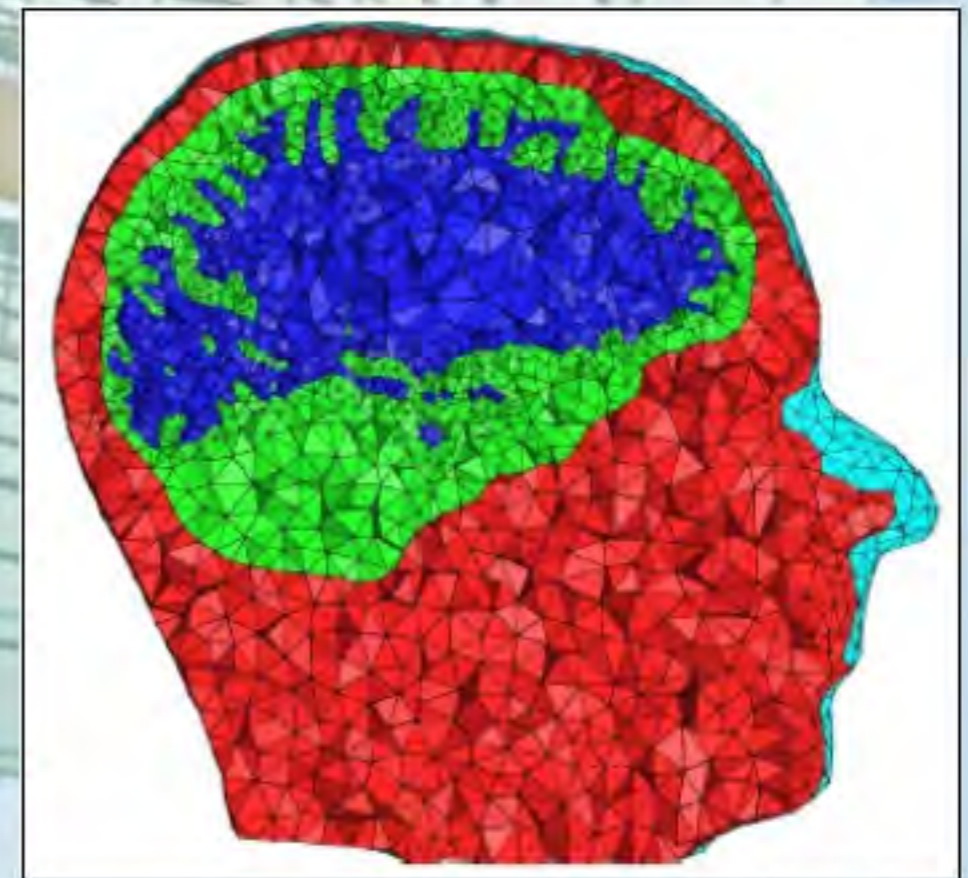
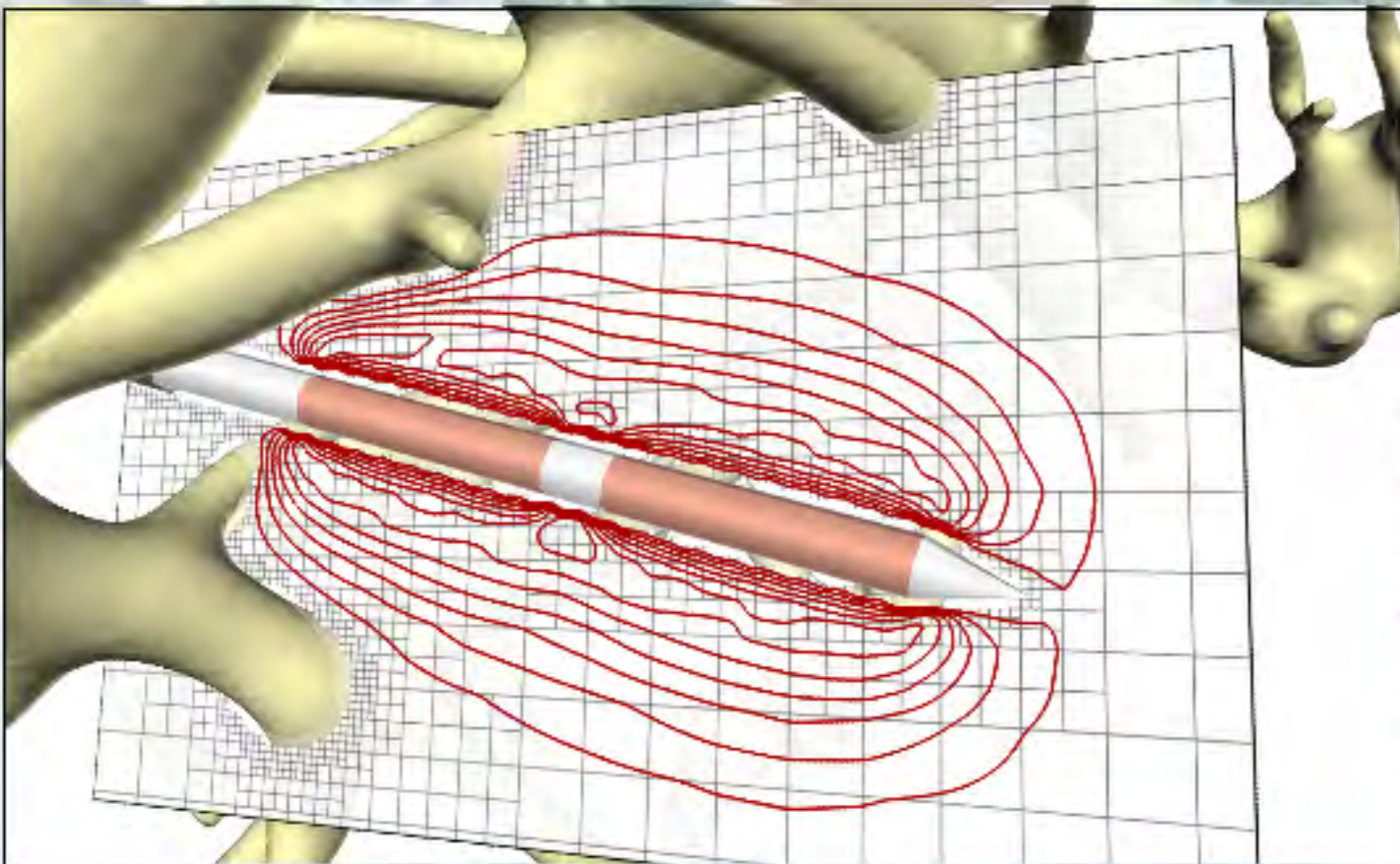


[iso2mesh.sourceforge.net](http://iso2mesh.sourceforge.net)

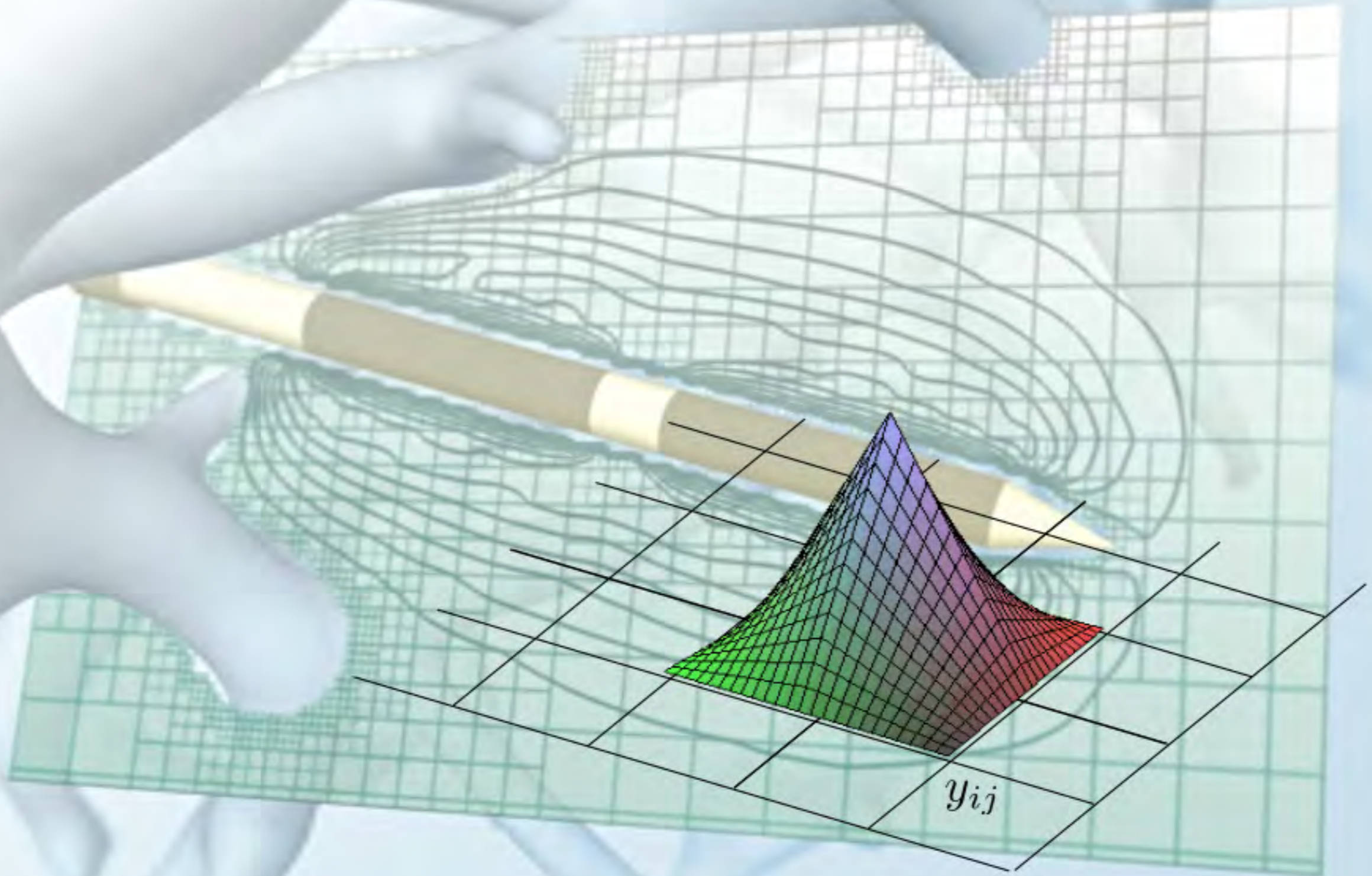


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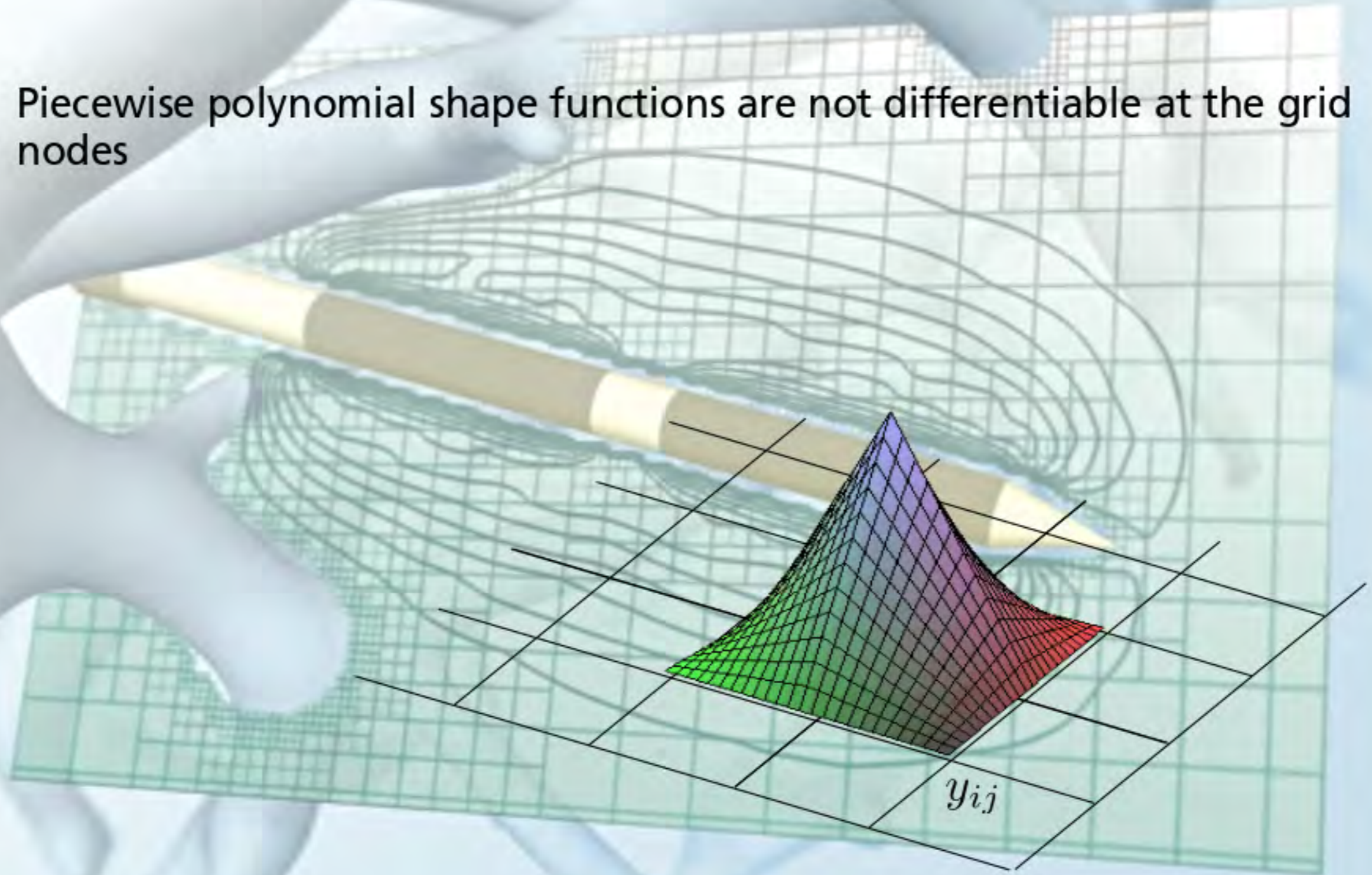


# Piecewise Polynomial Shape Functions



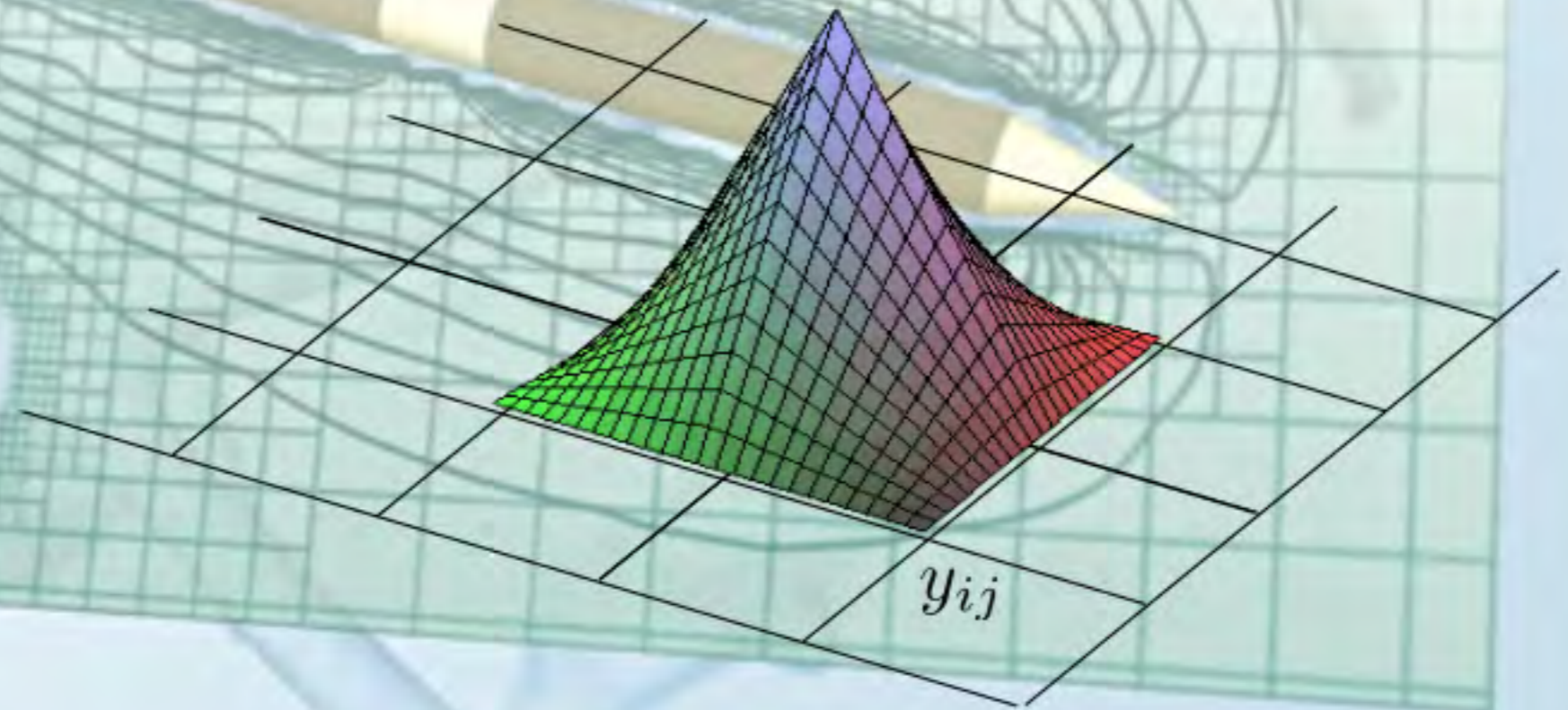
# Piecewise Polynomial Shape Functions

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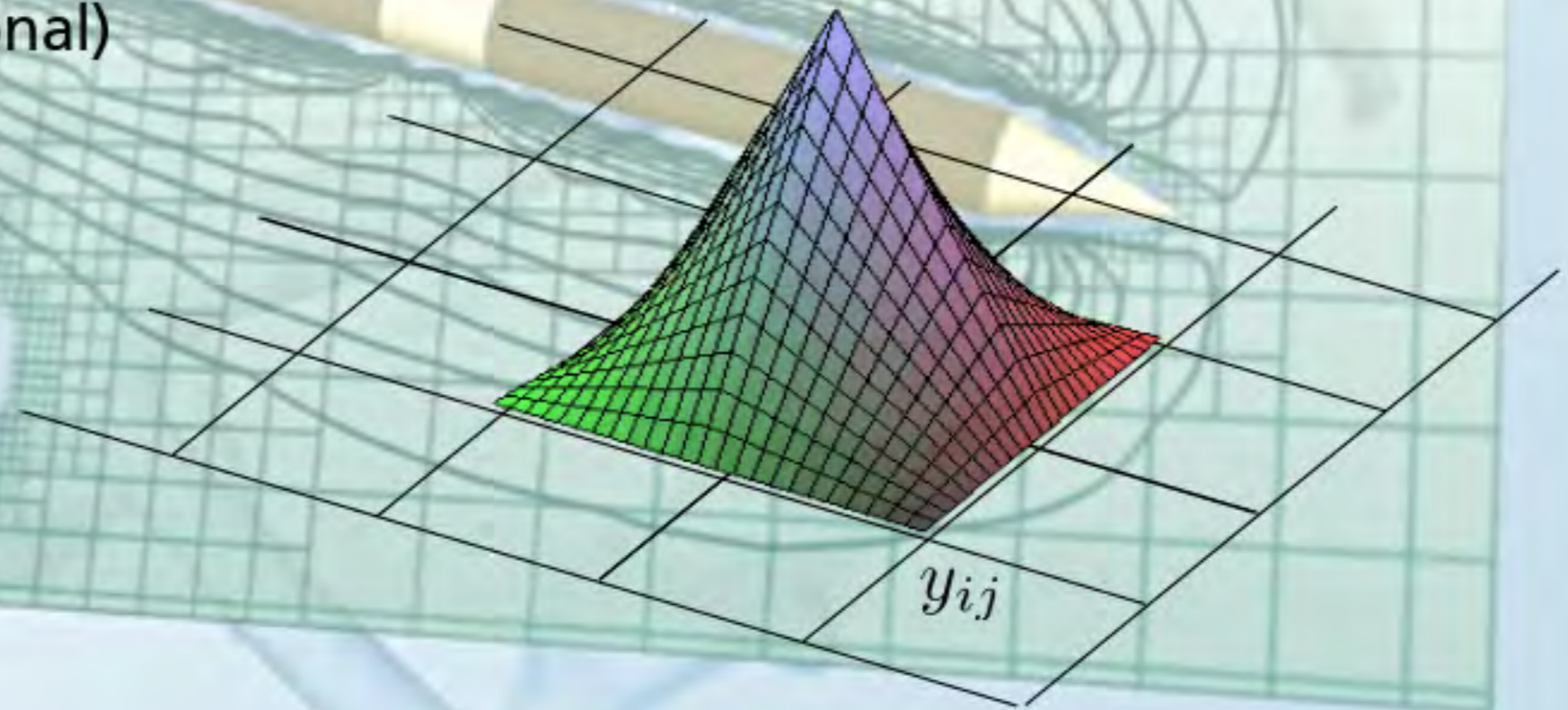
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# Piecewise Polynomial Shape Functions

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- › Need a different notion of differentiability
- › Weak (distributional) derivative



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- › Linear system of equations

$$A\Phi = F$$

$$A \in \mathbb{R}^{(N+1) \times (N+1)}$$

$$\Phi, F \in \mathbb{R}^{(N+1)}$$

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- › Matrix inversion:
  - » Iterative Method, e.g. Gauss–Seidel, conjugate gradient method
  - » But also: Multigrid method

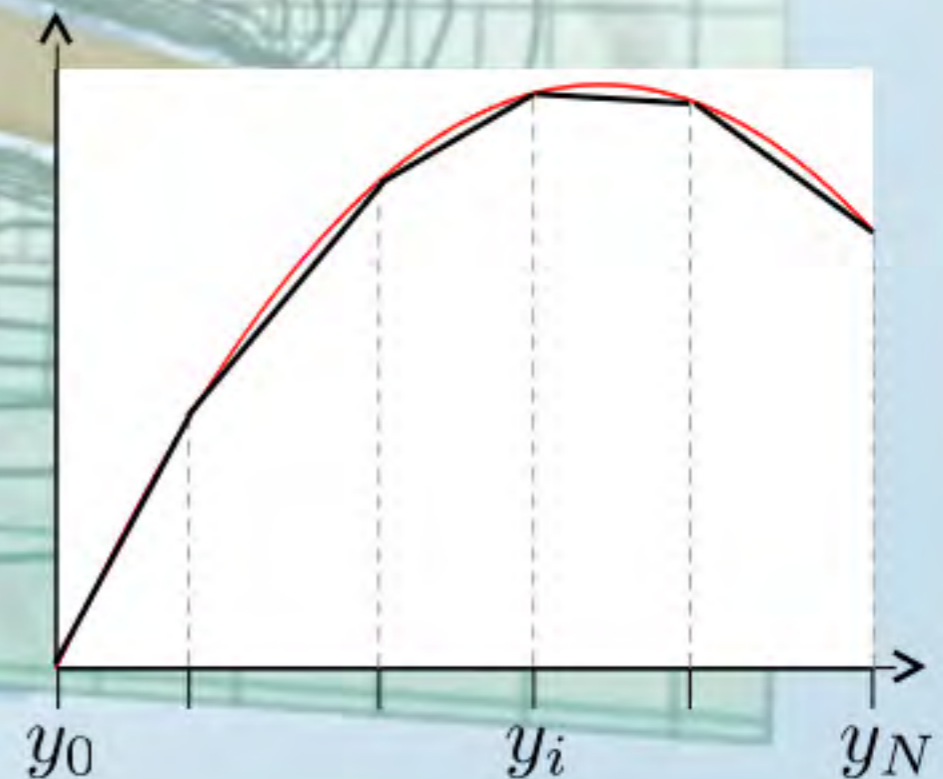


## Summary FEM

- › Compose functions by shape functions
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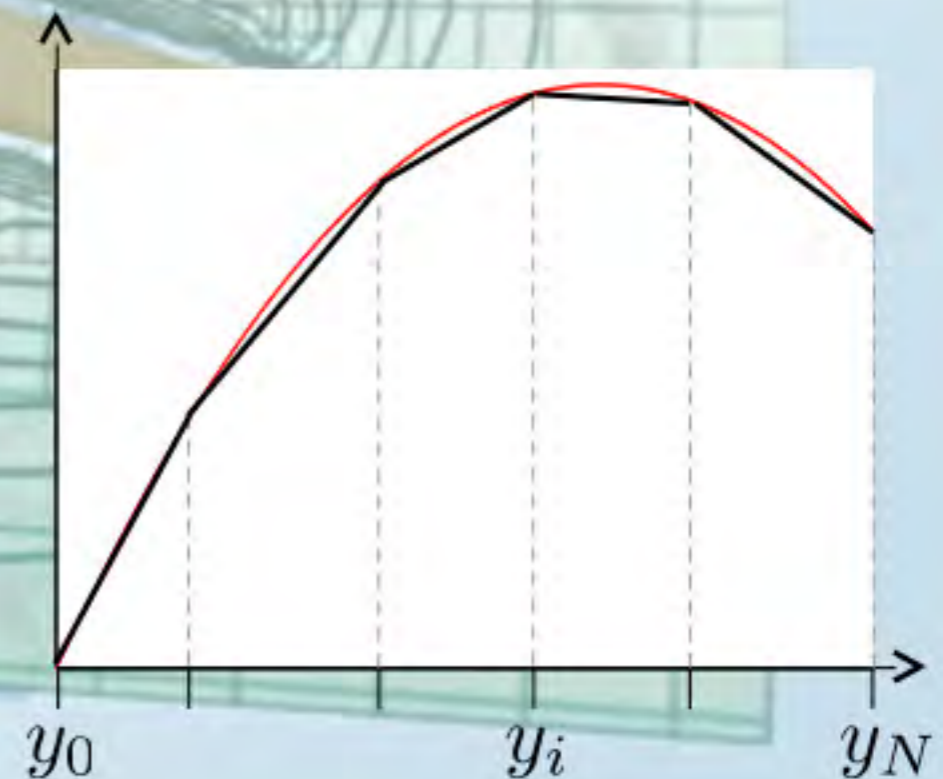
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› Accuracy  $\|\phi - \phi_h\|_{L^2} = Ch^{p+1}$



# 3

## Discretizing the Bioheat Transfer Equation

## Semi-Discrete Form

$$\rho(x)c(x)\partial_t T(t,x) - \operatorname{div}(\lambda(x)\nabla T(t,x)) = \sigma(x)|\nabla\phi(x)|^2 - \nu(x)(T(t,x) - T_b)$$

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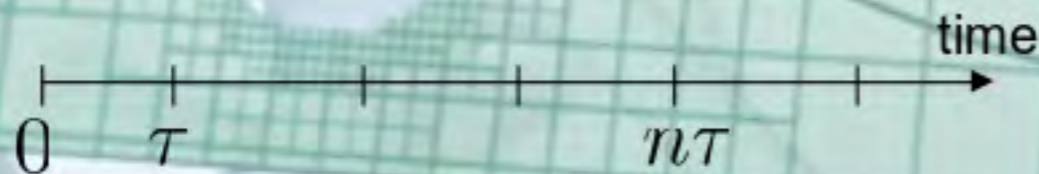
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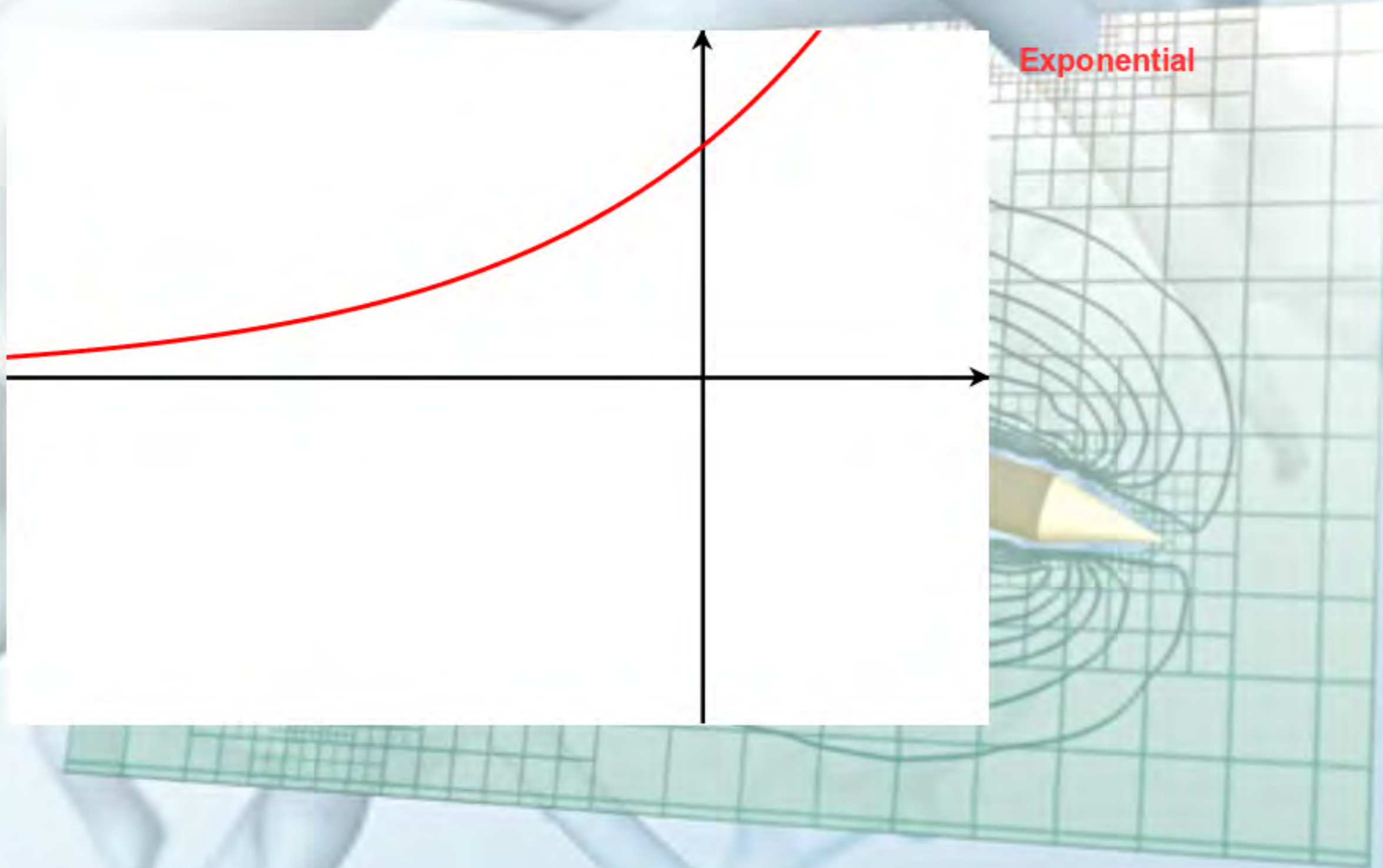
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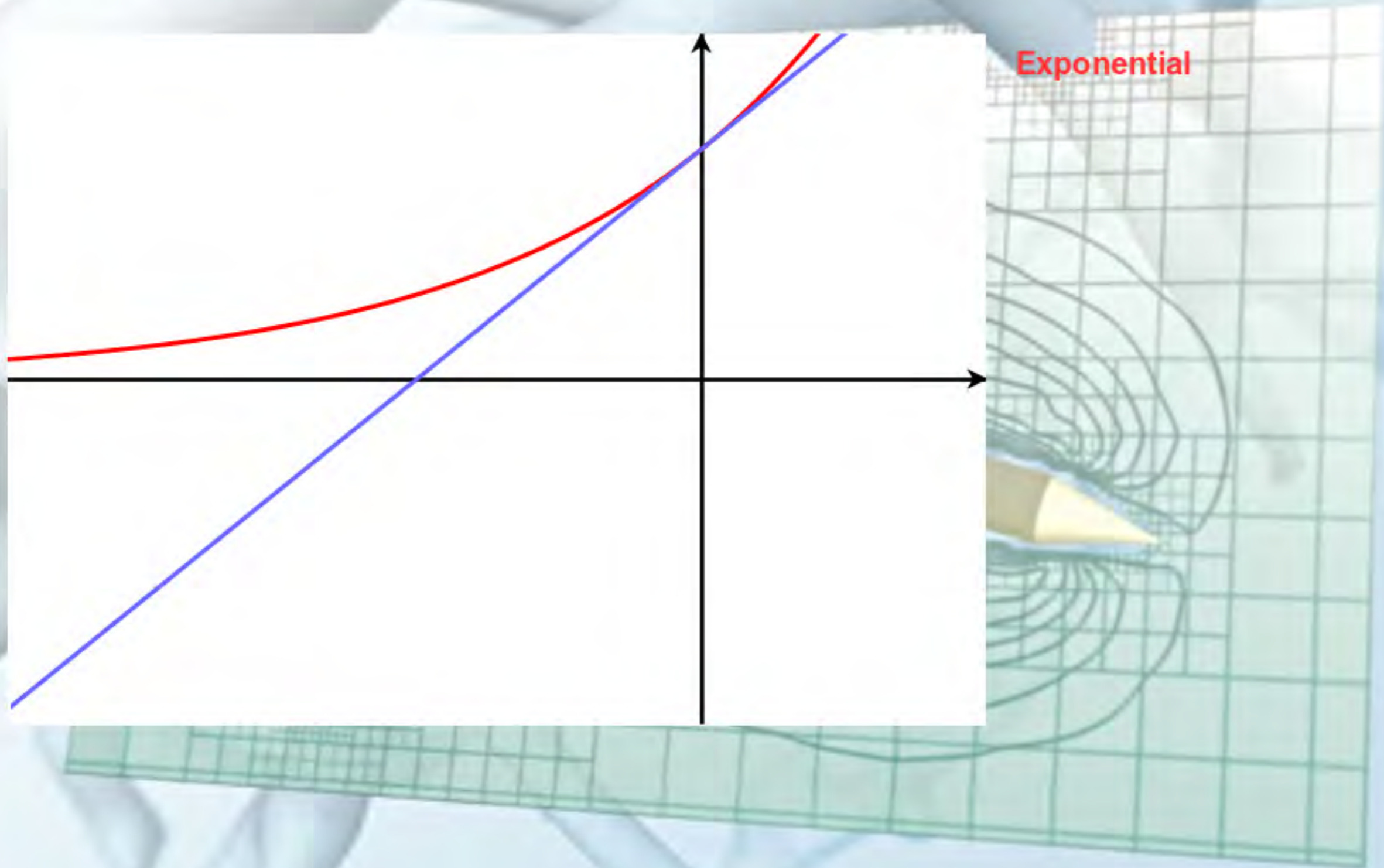


$$T_{n+1} = \exp(\tau A) T_n$$

# Approximating $\exp(\tau A)$ by Rational Functions

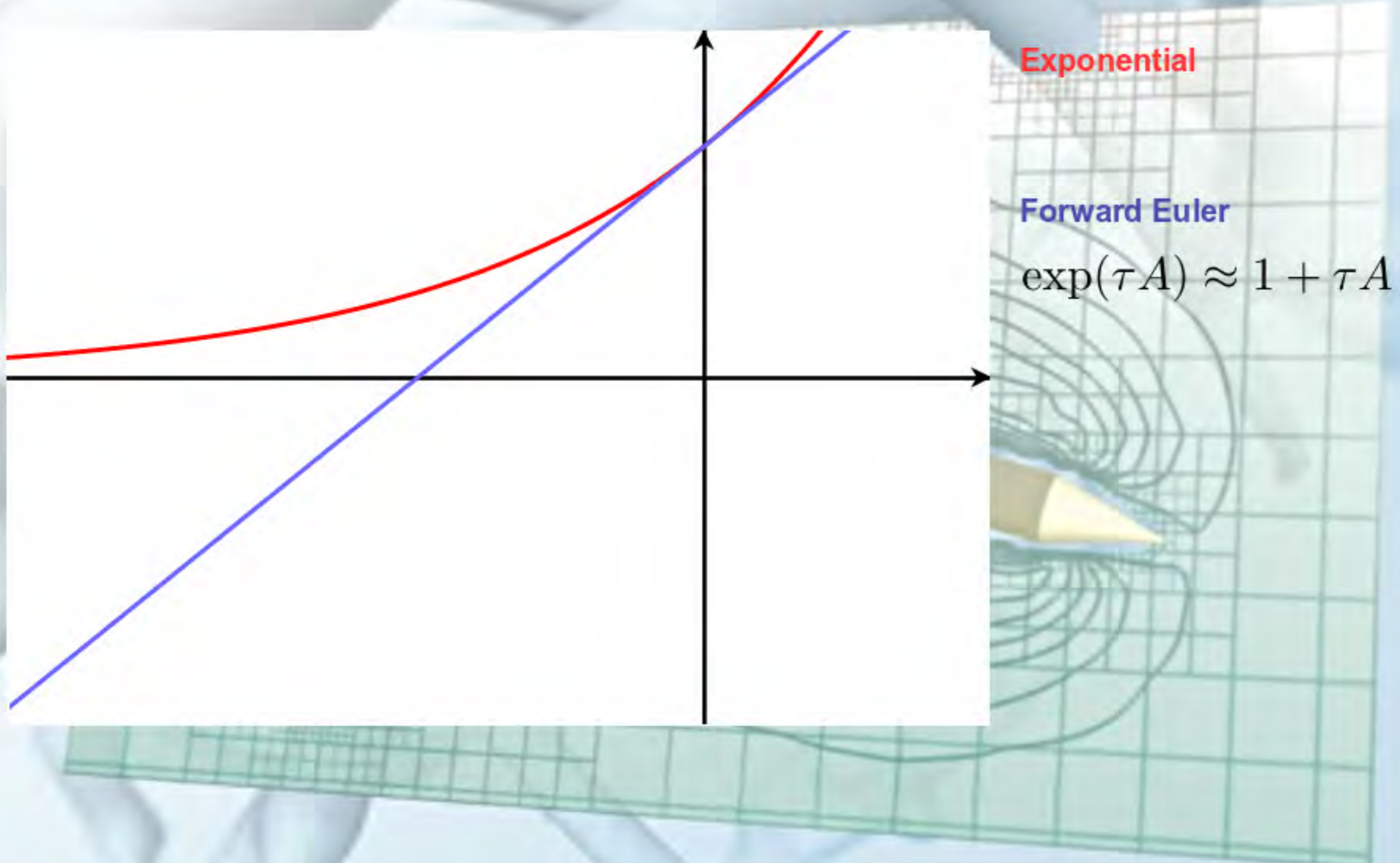


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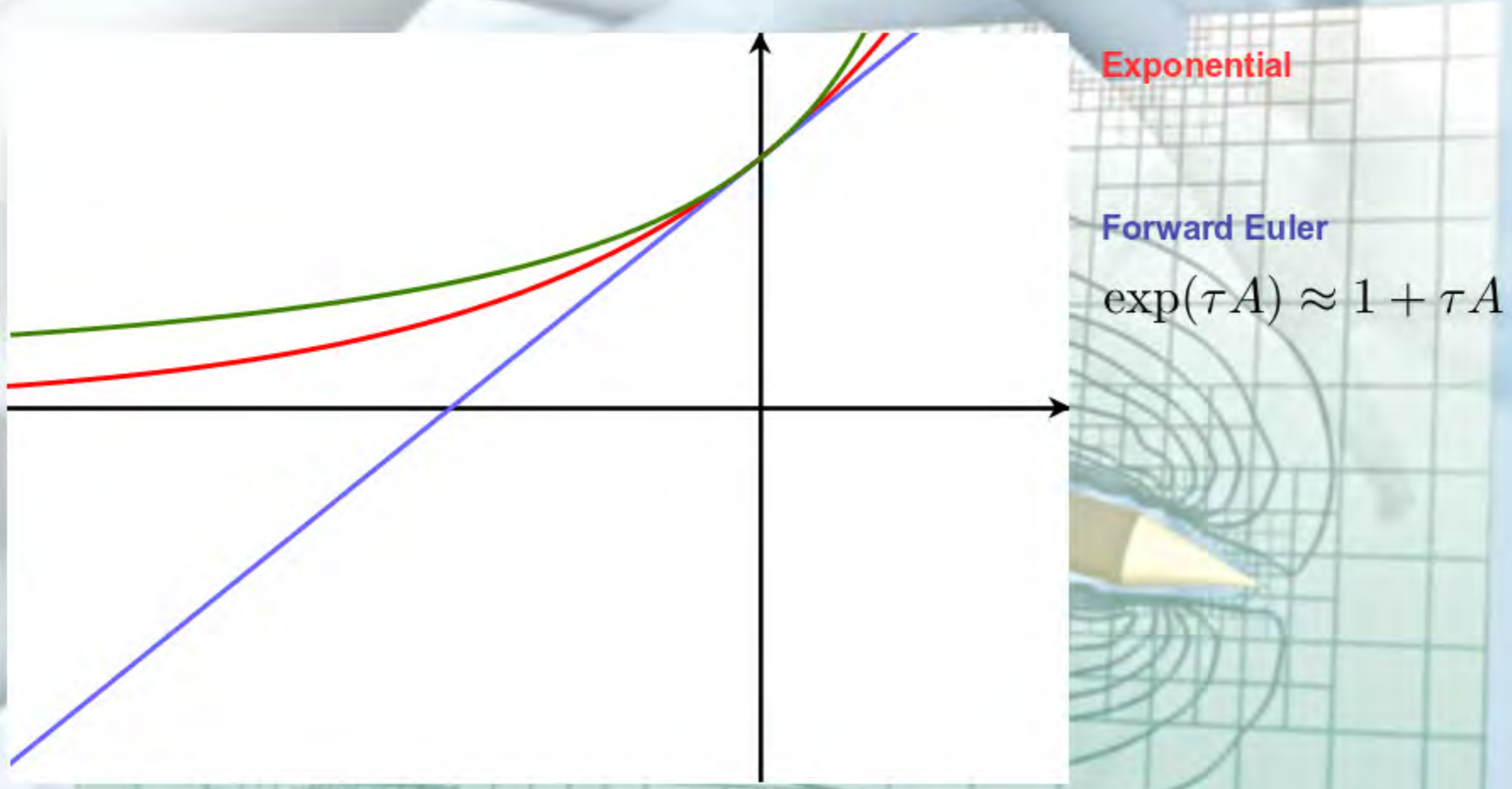




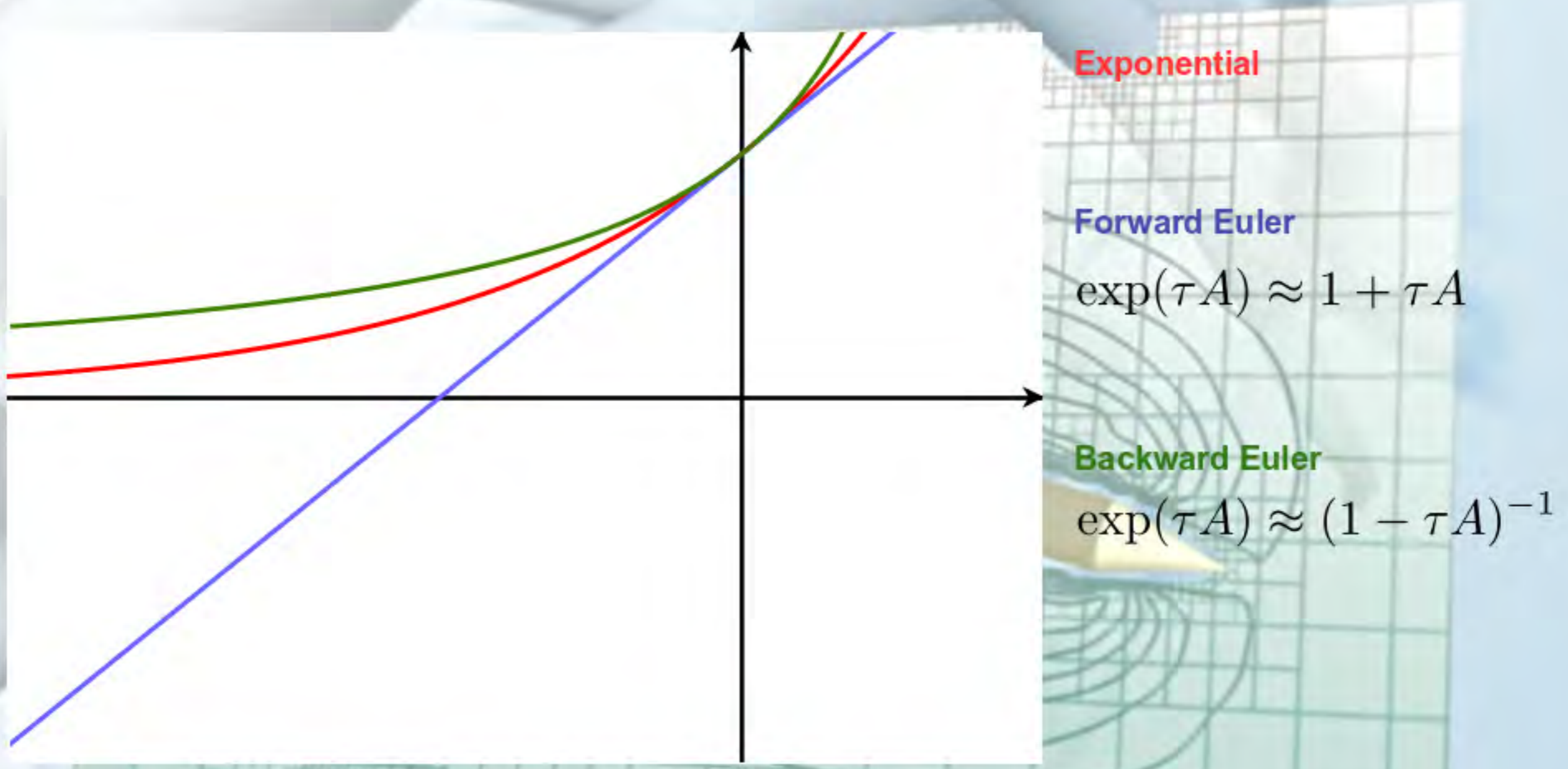
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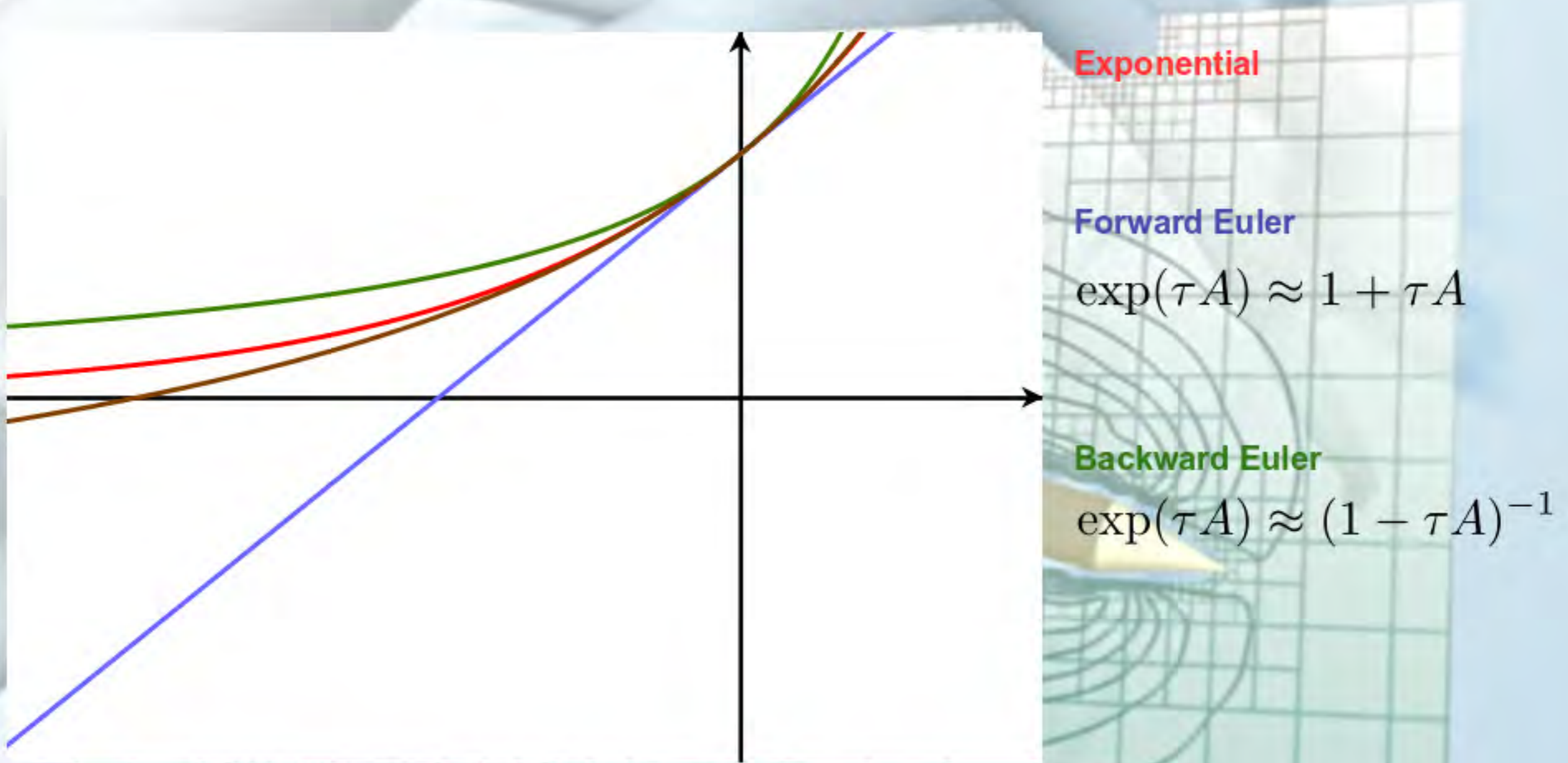
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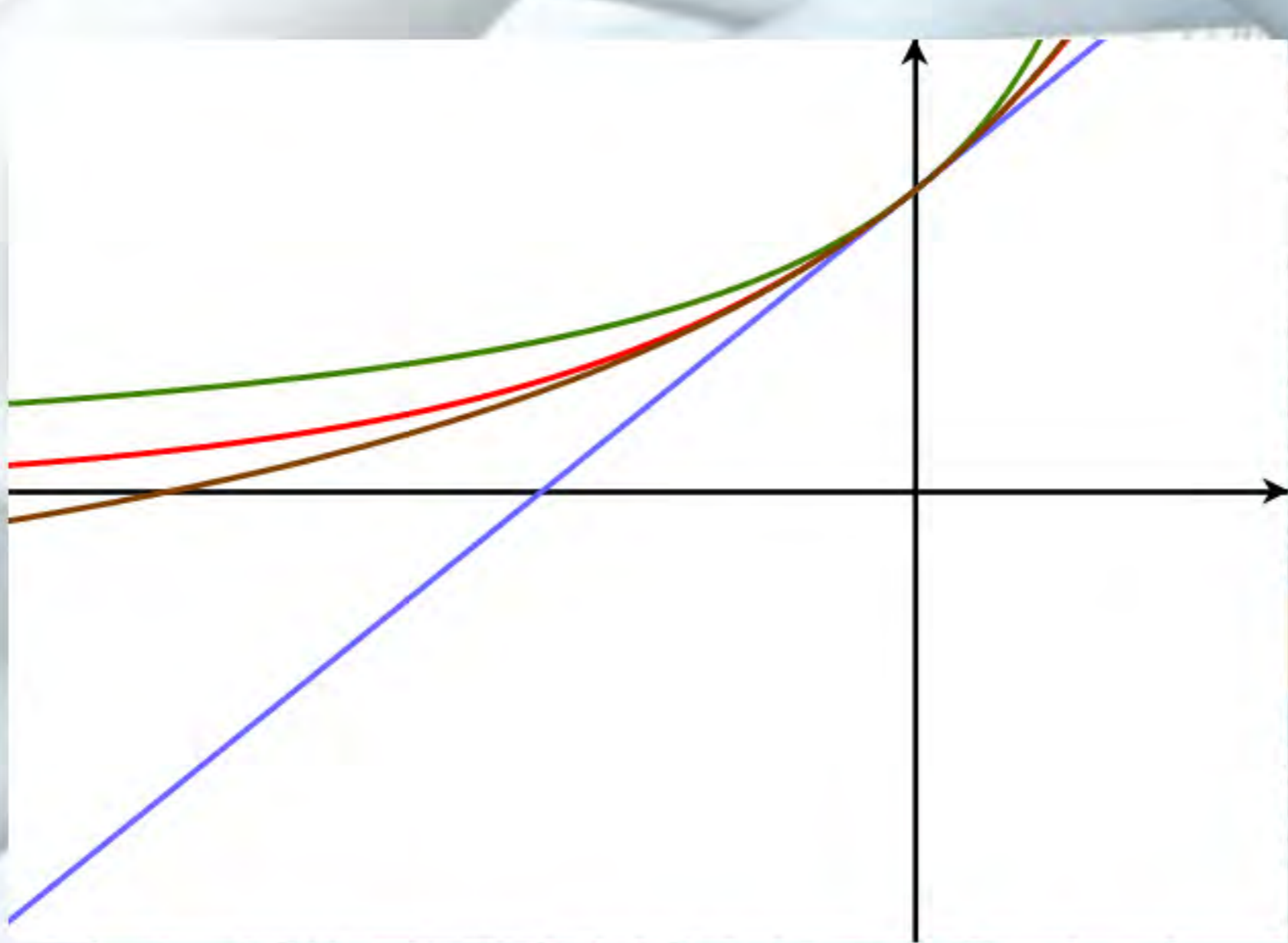
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**Exponential**

**Forward Euler**

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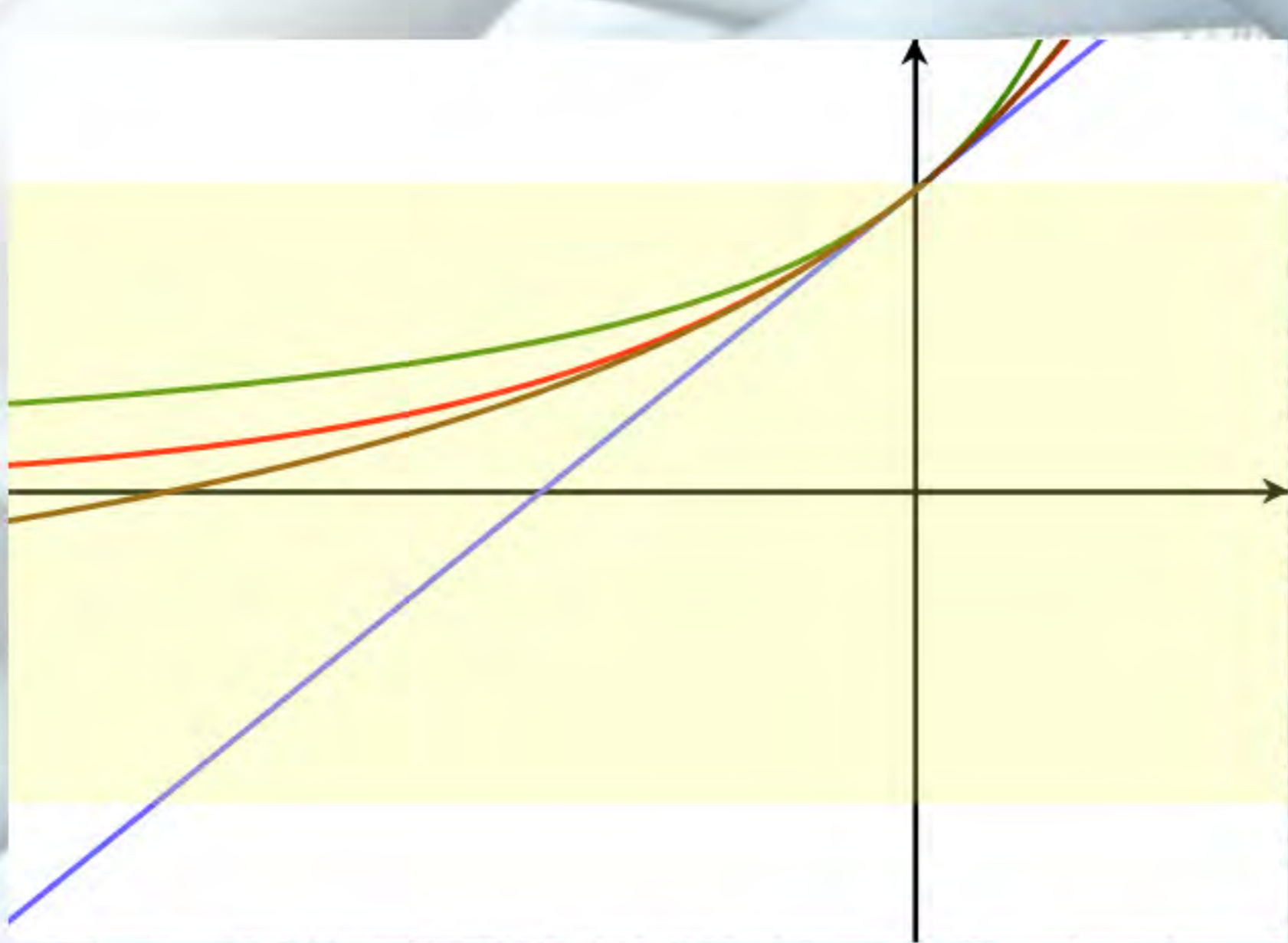
**Backward Euler**

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**Crank-Nicholson**

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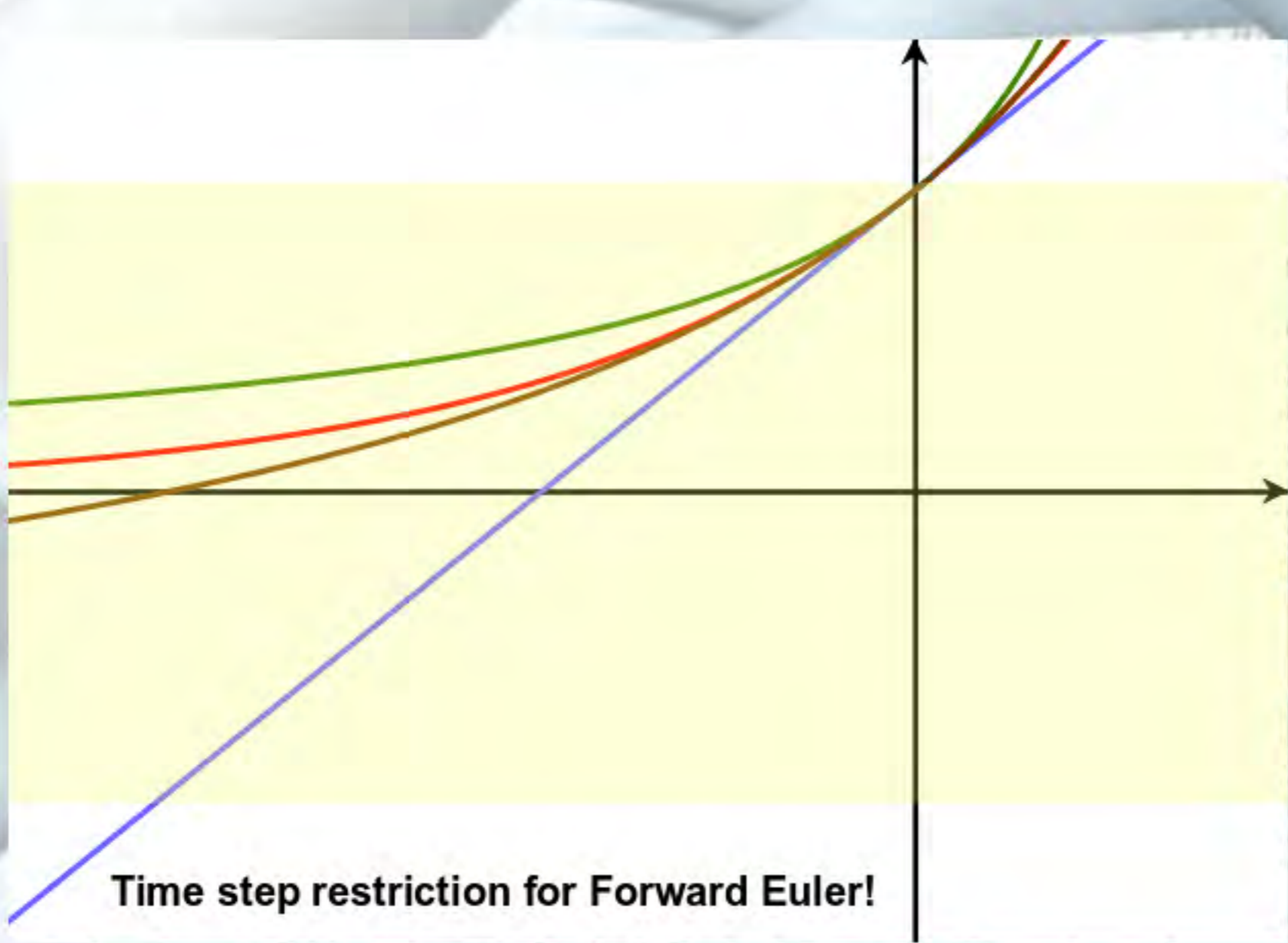
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Time step restriction for Forward Euler!

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- Matrix inversion necessary



4

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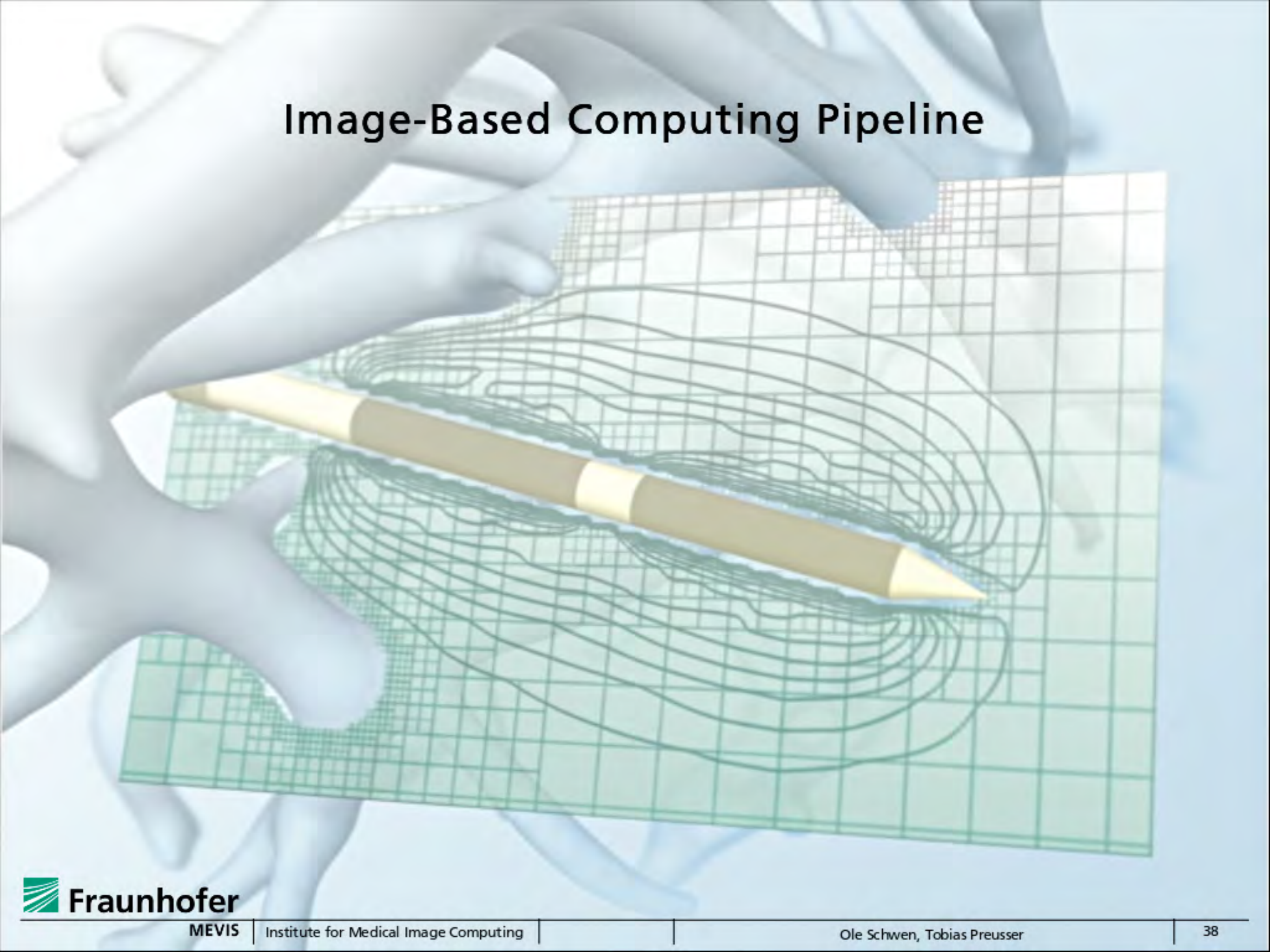
For 1D regular grids and piecewise linear shape functions: FDM = FEM



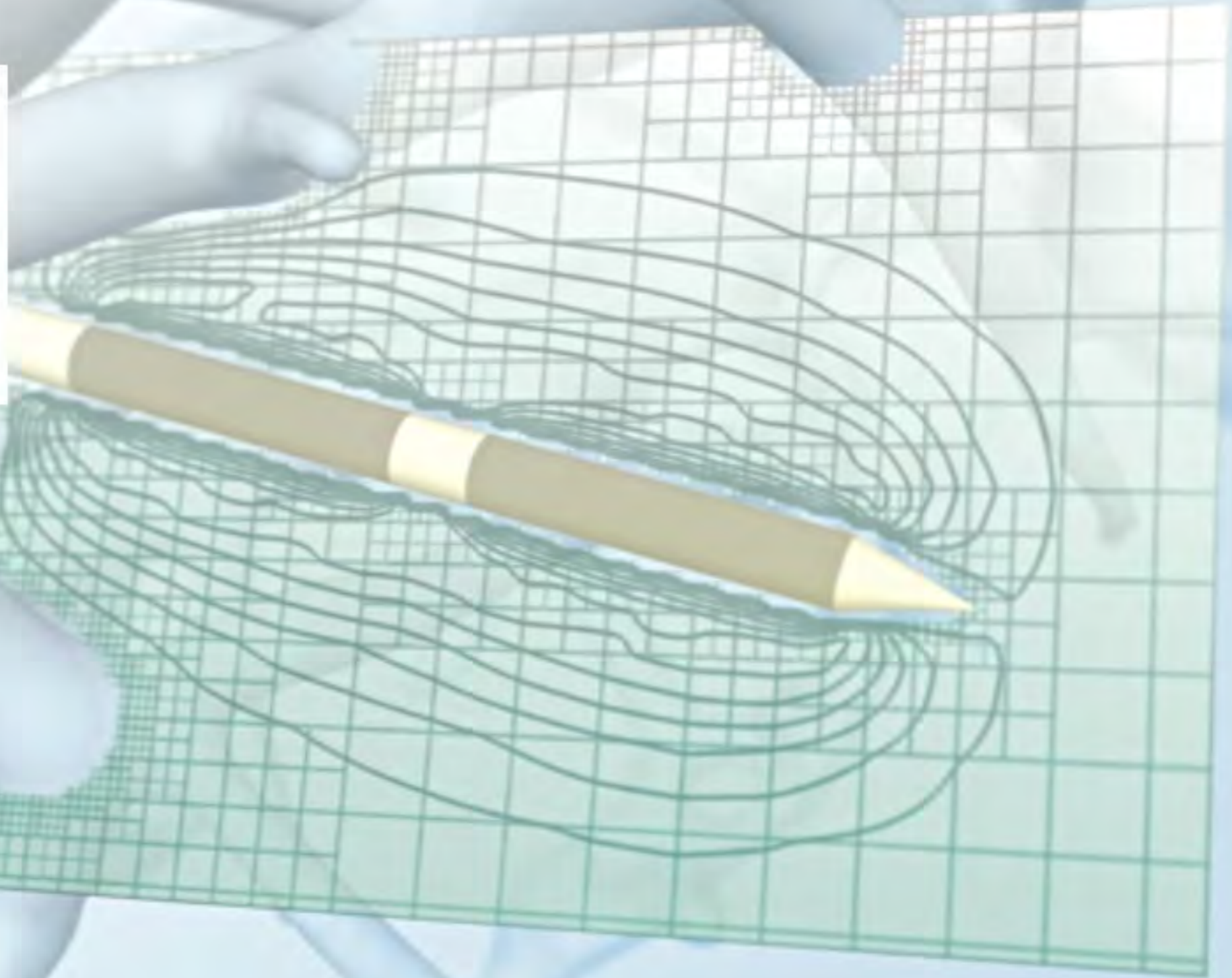
5

# Examples

# Image-Based Computing Pipeline



# Image-Based Computing Pipeline



# Image-Based Computing Pipeline



Image Acquisition



# Image-Based Computing Pipeline

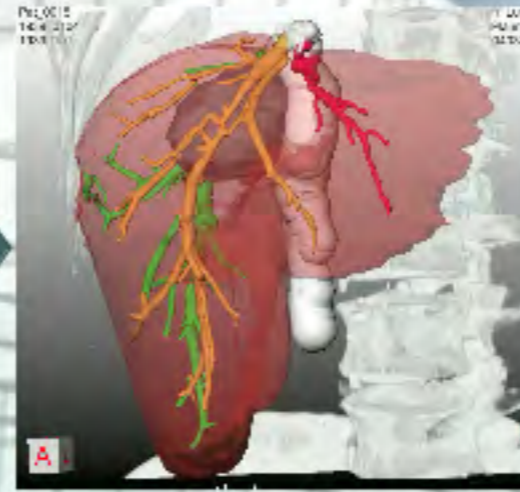


Image Acquisition

Image Processing

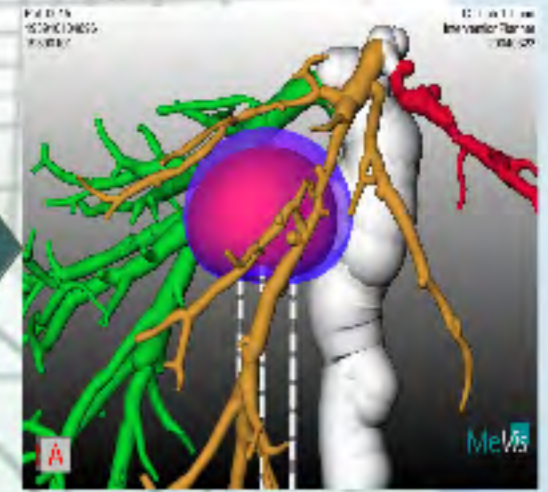
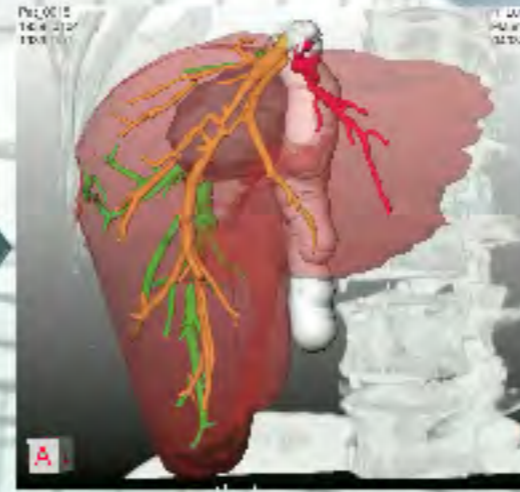
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Image Acquisition



Image Processing



Simulation

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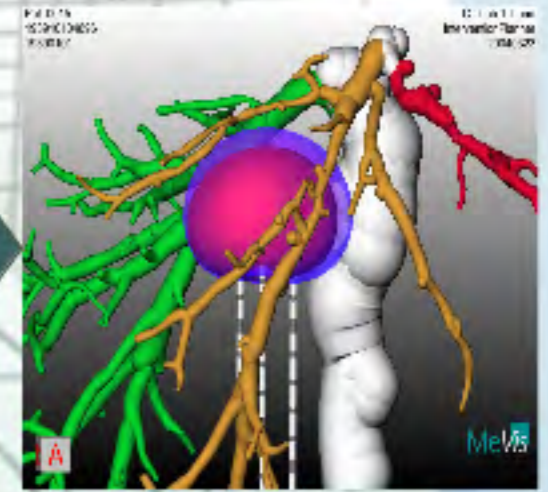
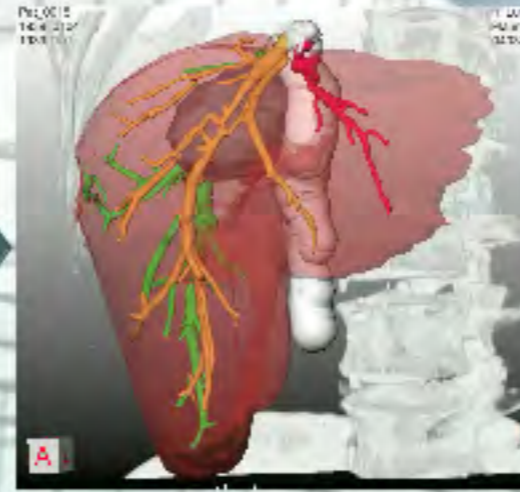


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Simulation



# Image-Based Computing Pipeline

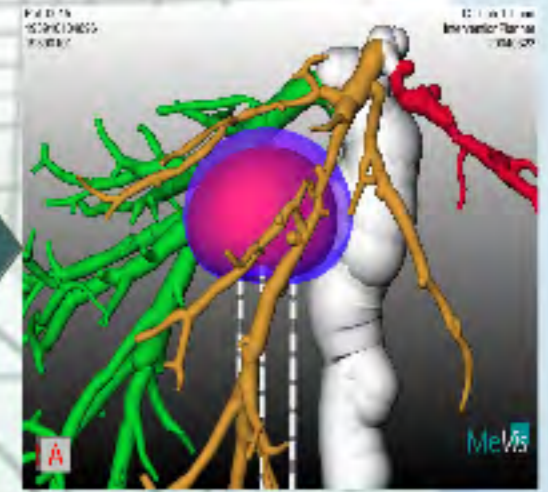
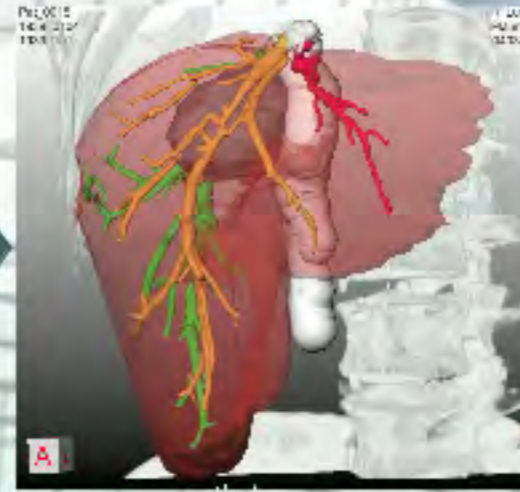
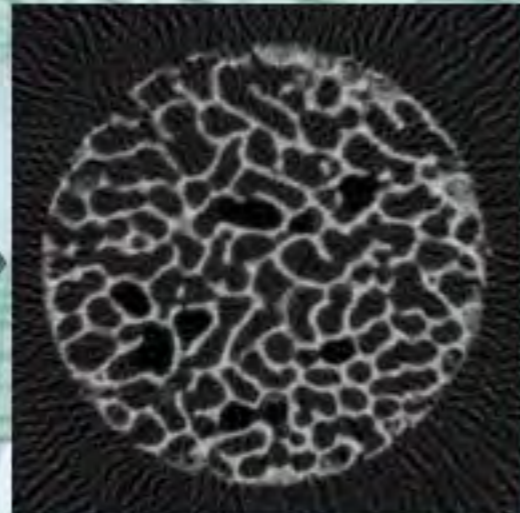


Image Acquisition

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Simulation



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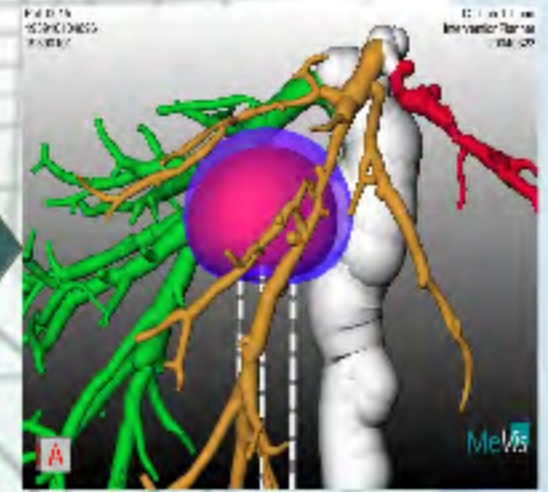
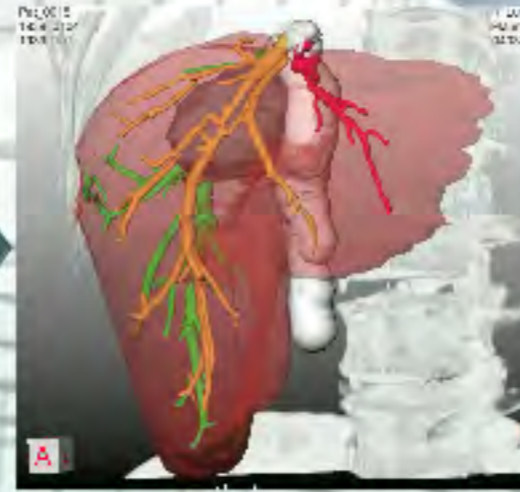
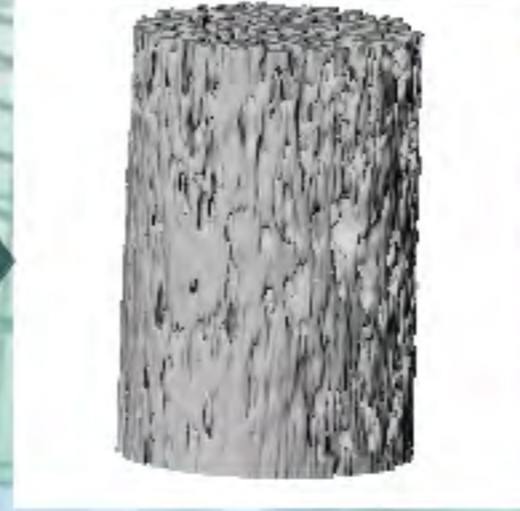
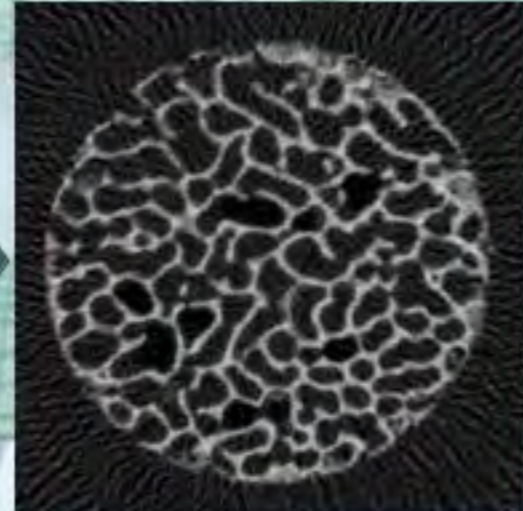


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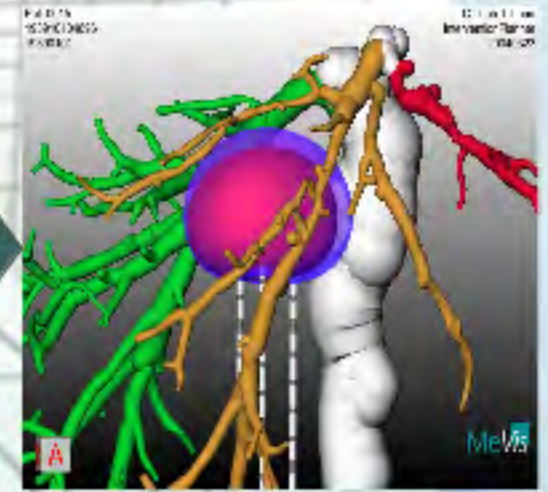
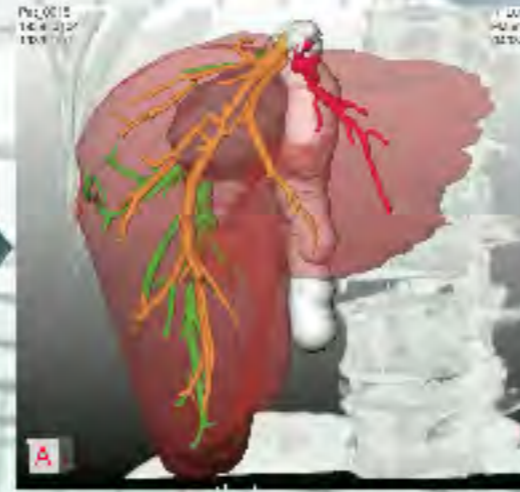
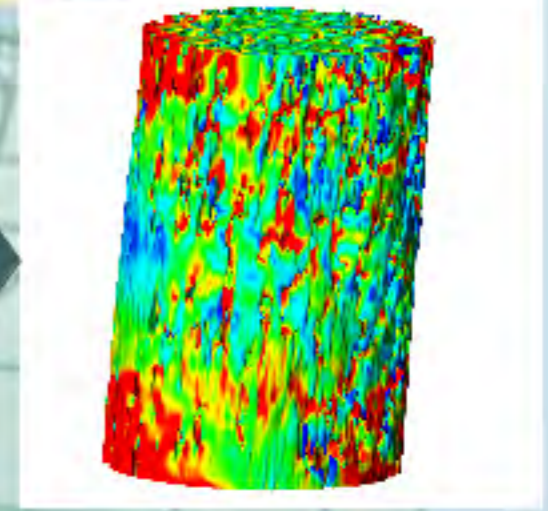
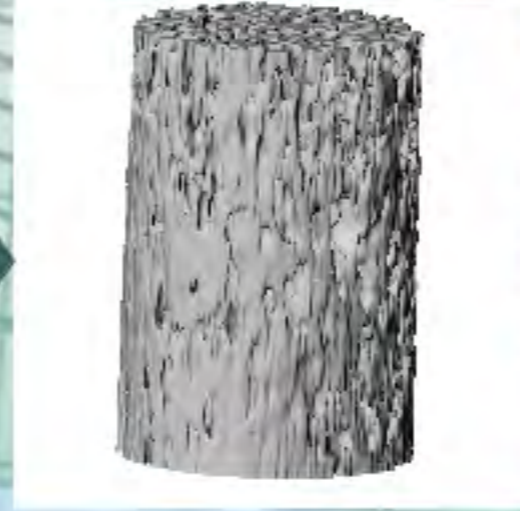
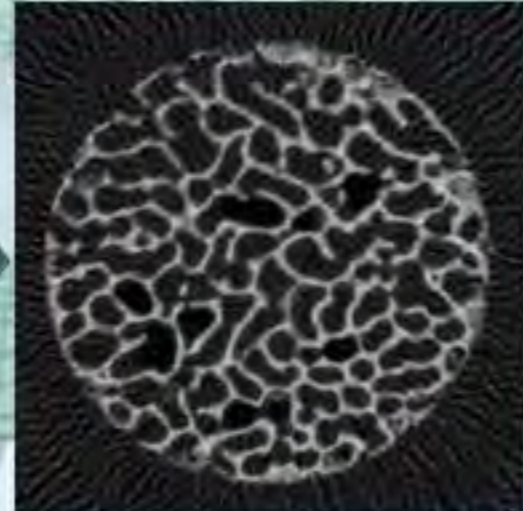


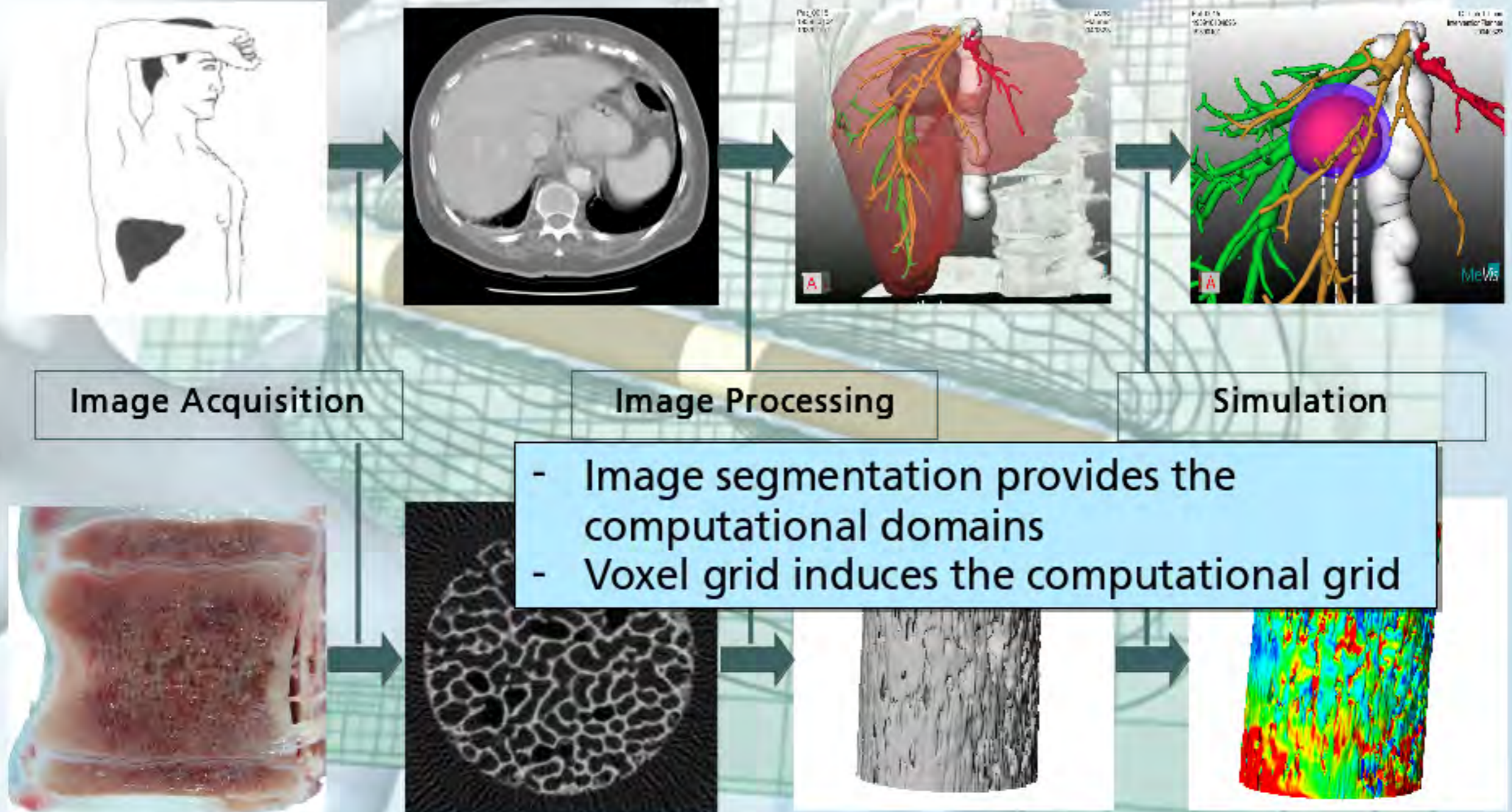
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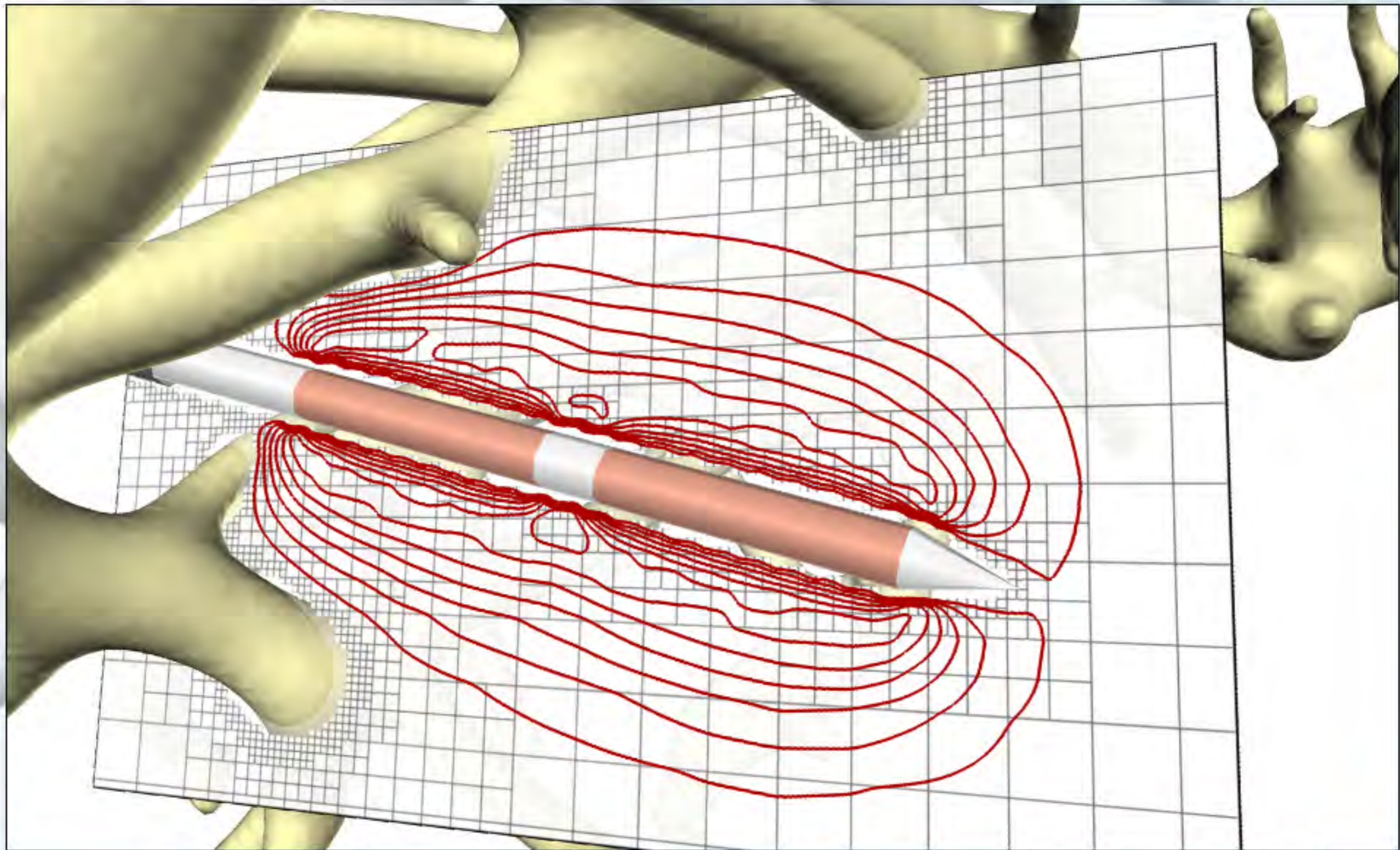
# Simulation of RF Ablation





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# Summary and Outlook

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