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# On the implementation of point-based and cell-based mesh smoothing approaches in OpenFOAM

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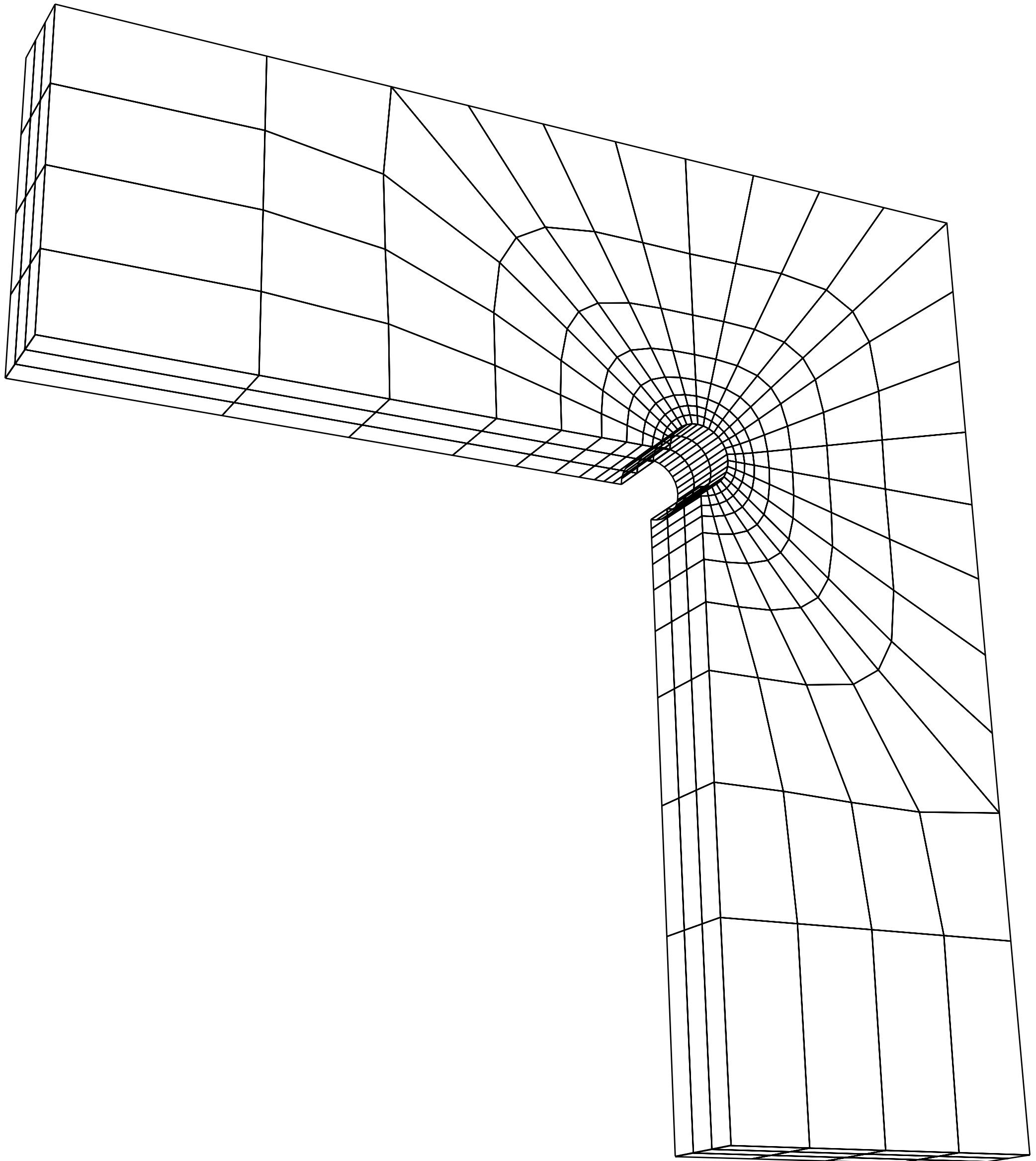
# Background & Motivation

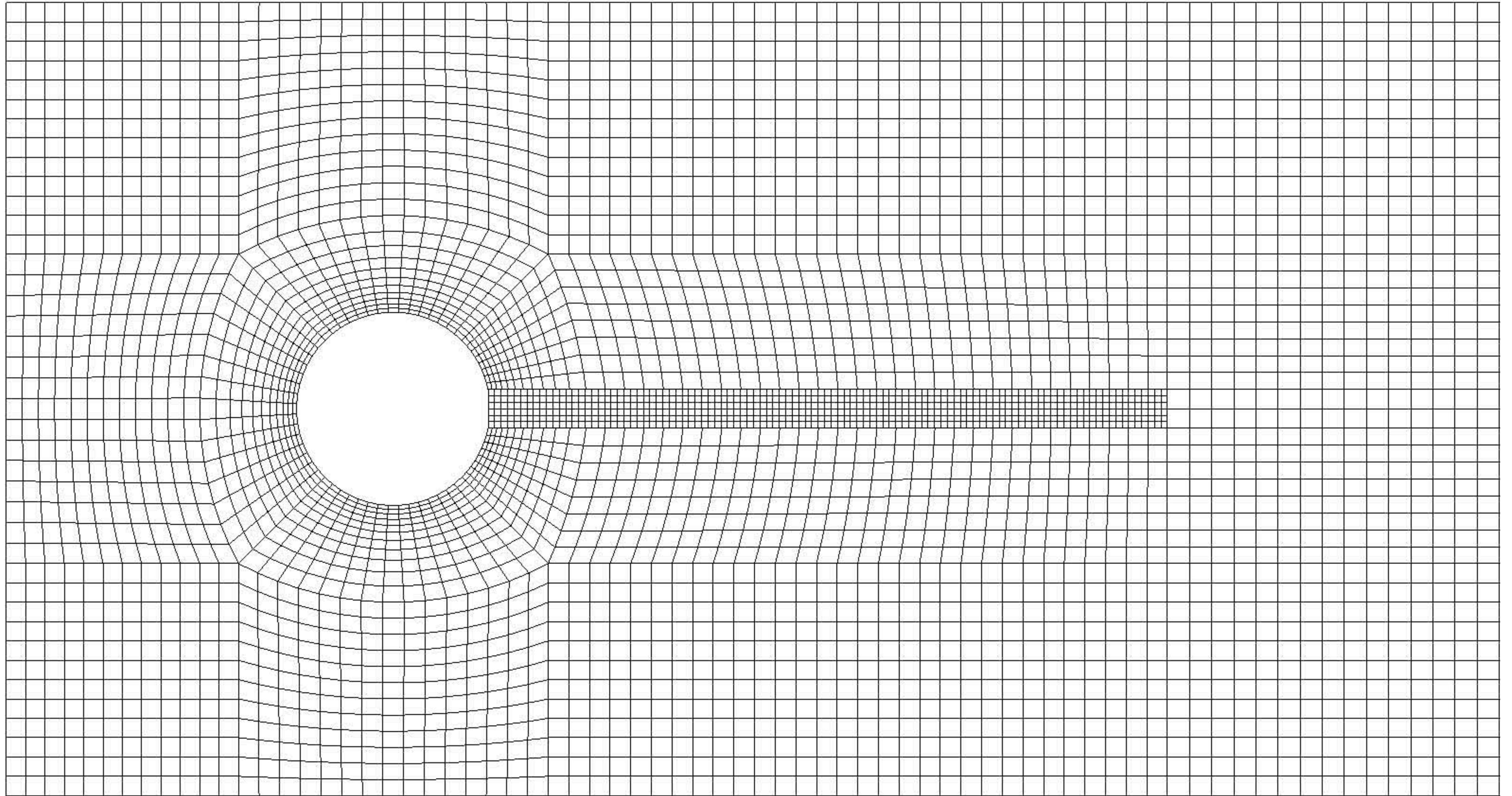
# Background

Mesh quality affects solution quality

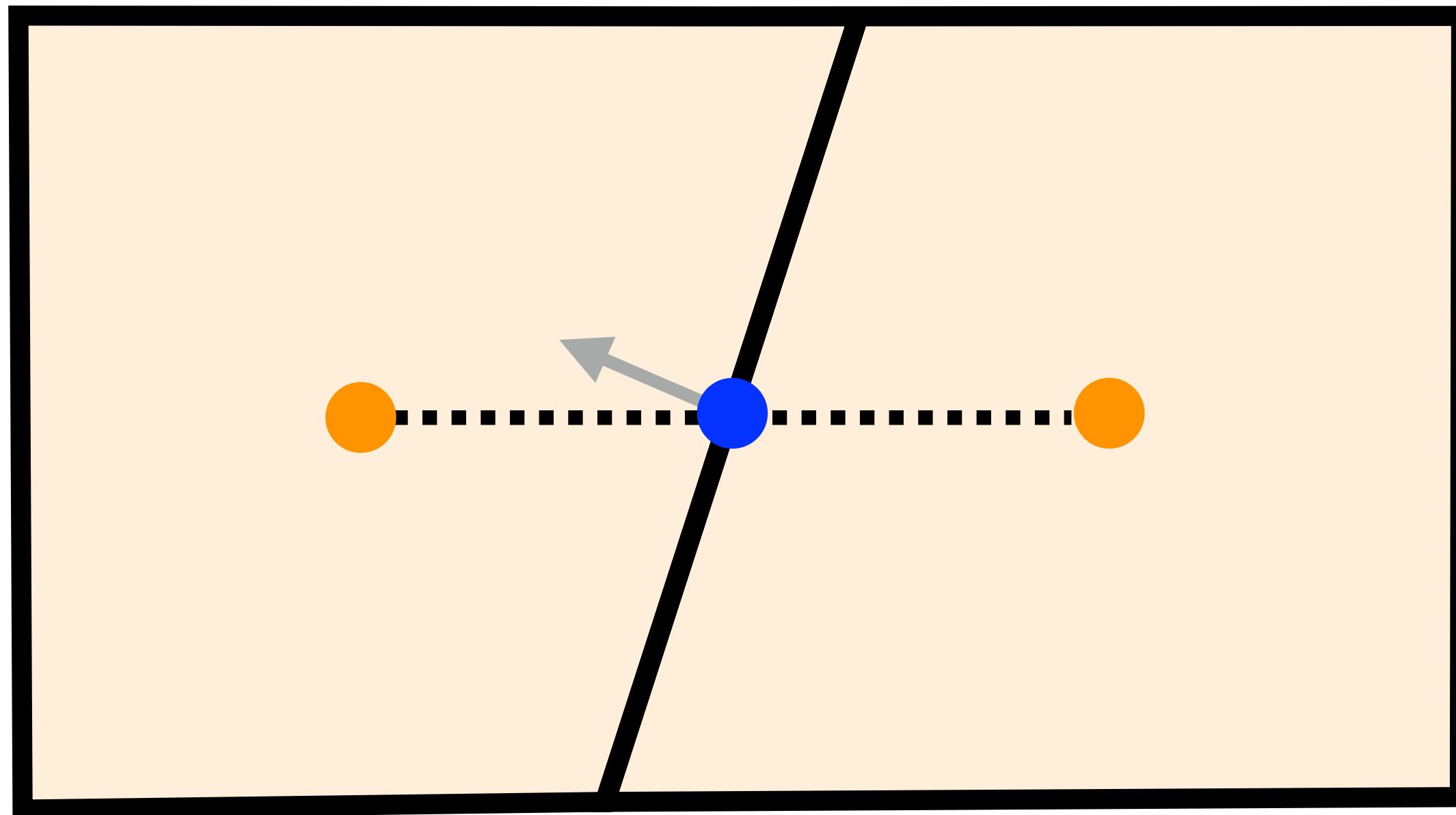
Why smooth?

- improve initial mesh
- maintain mesh quality in moving mesh simulations  
e.g. Arbitrary Lagrangian-Eulerian

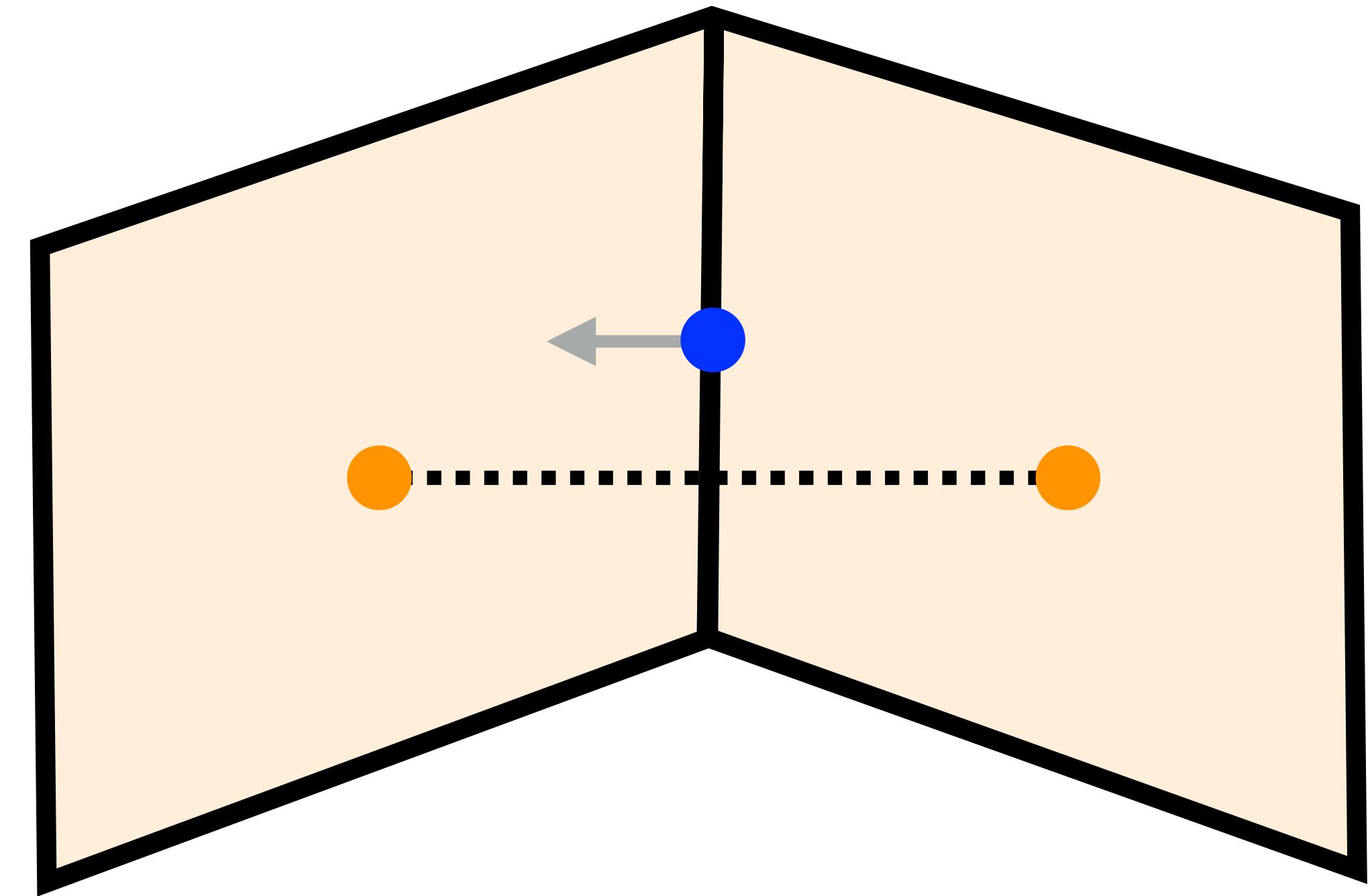




# Important mesh metrics



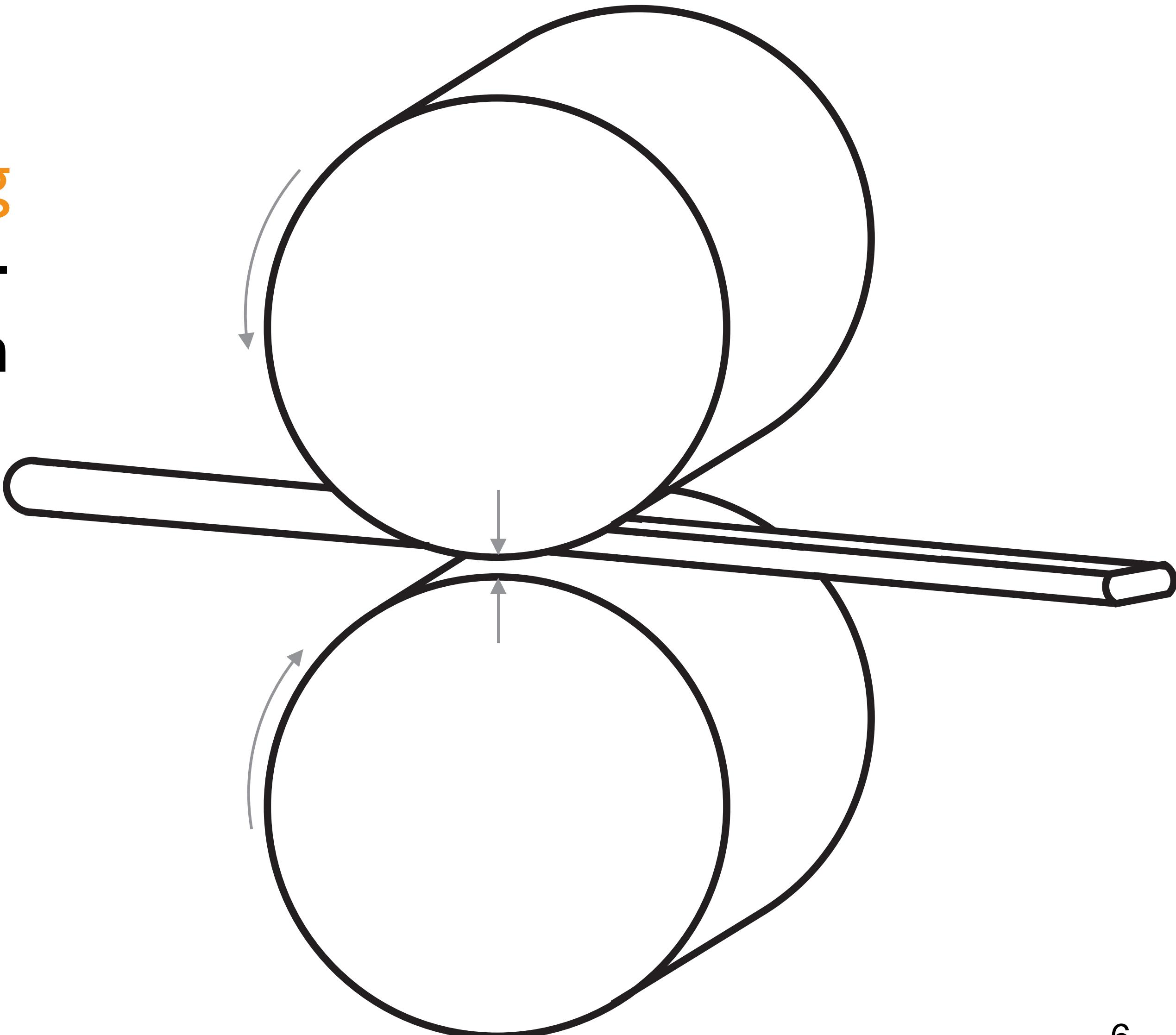
Non-orthogonal face



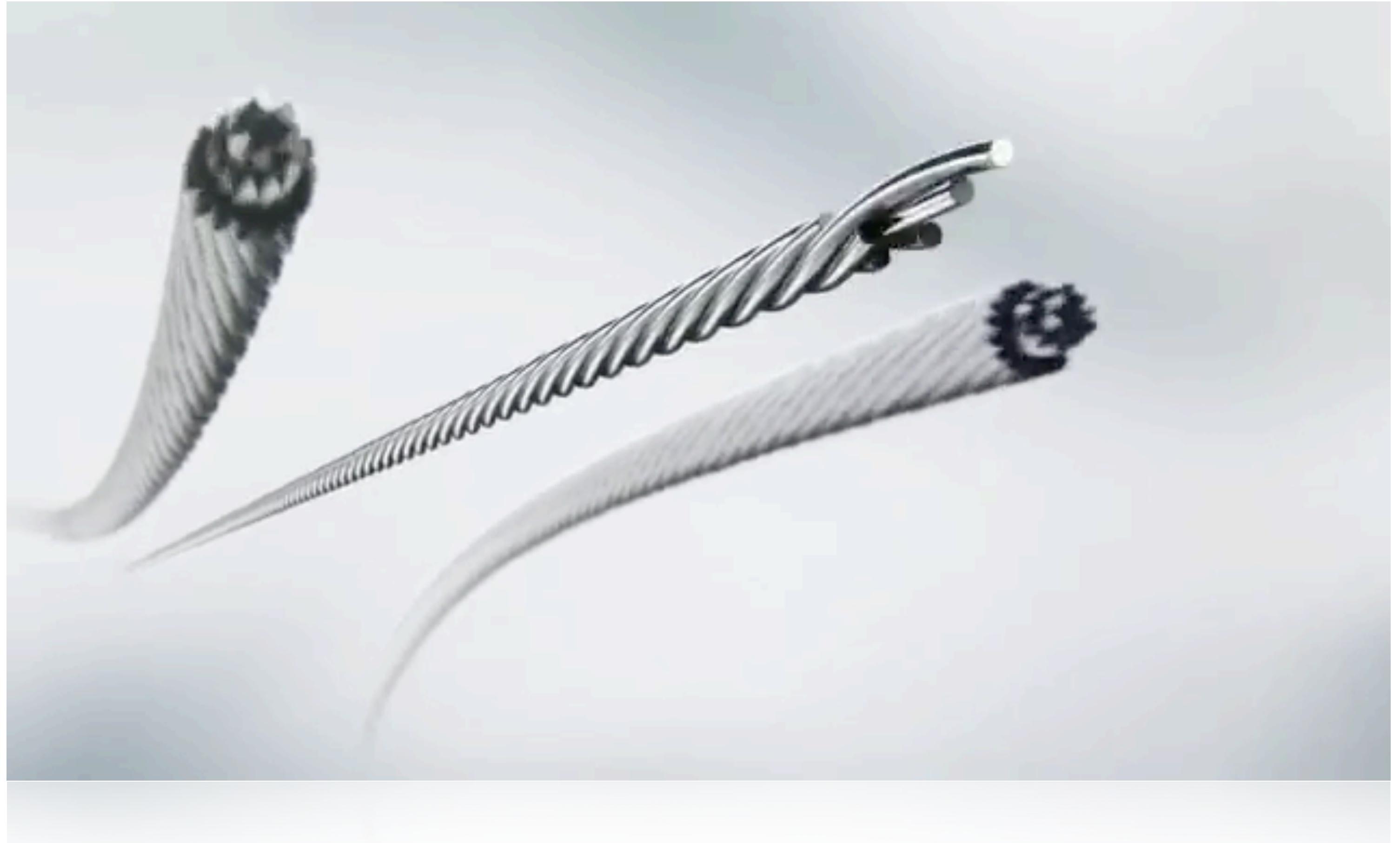
skewed face

# Goal of this project

Develop a **robust mesh smoothing procedure** suitable for ALE elasto-plastic metal forming simulations in OpenFOAM

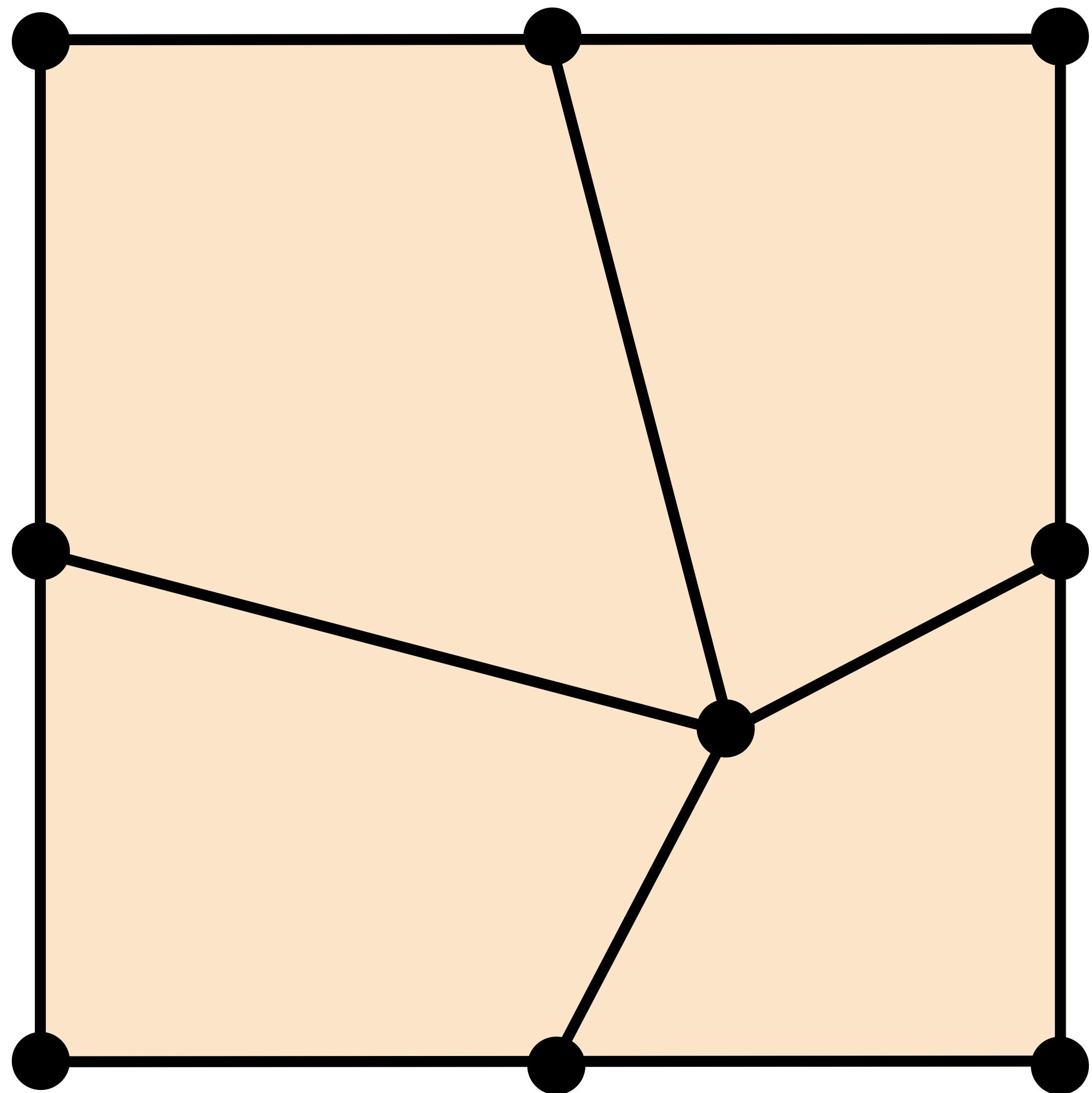


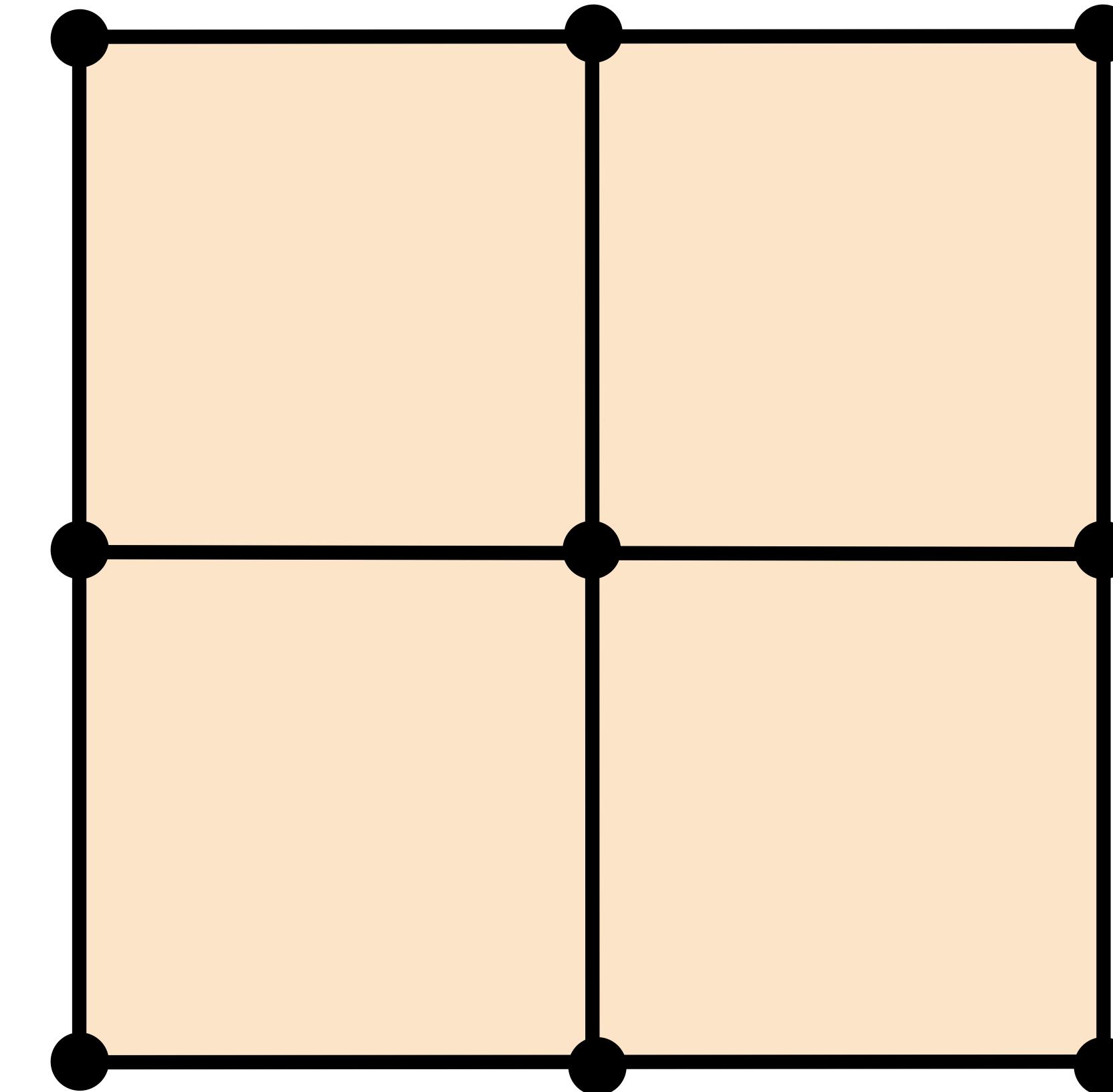
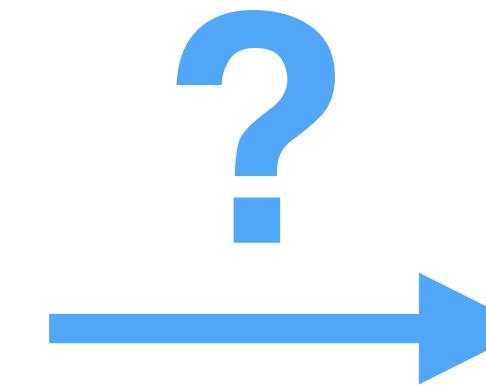
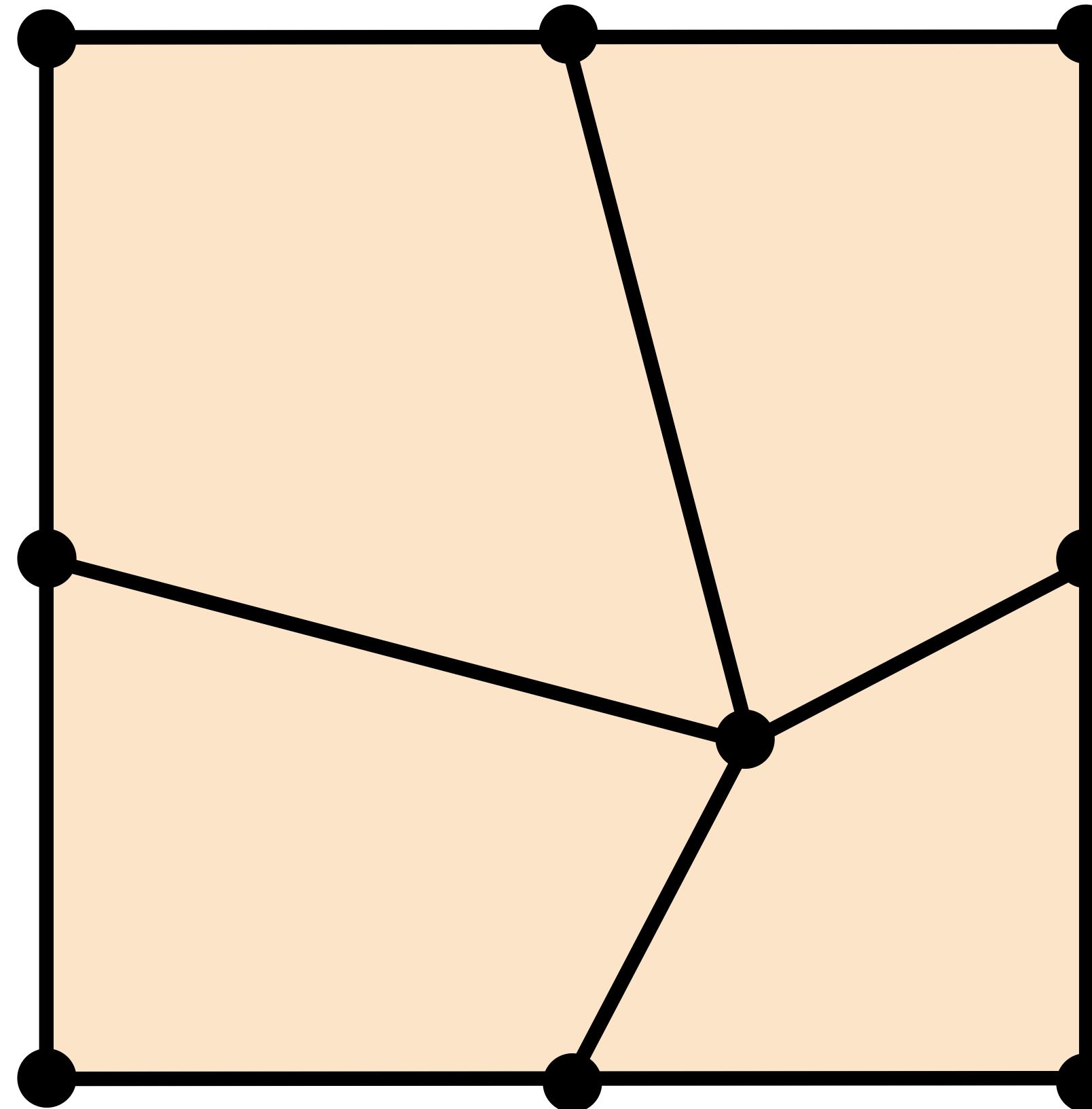
- Bekaert are global experts in the **transformation and coating of high-strength steel wire**
- Founded in 1880
- Headquartered in Belgium
- 30,000 employees worldwide





the challenge



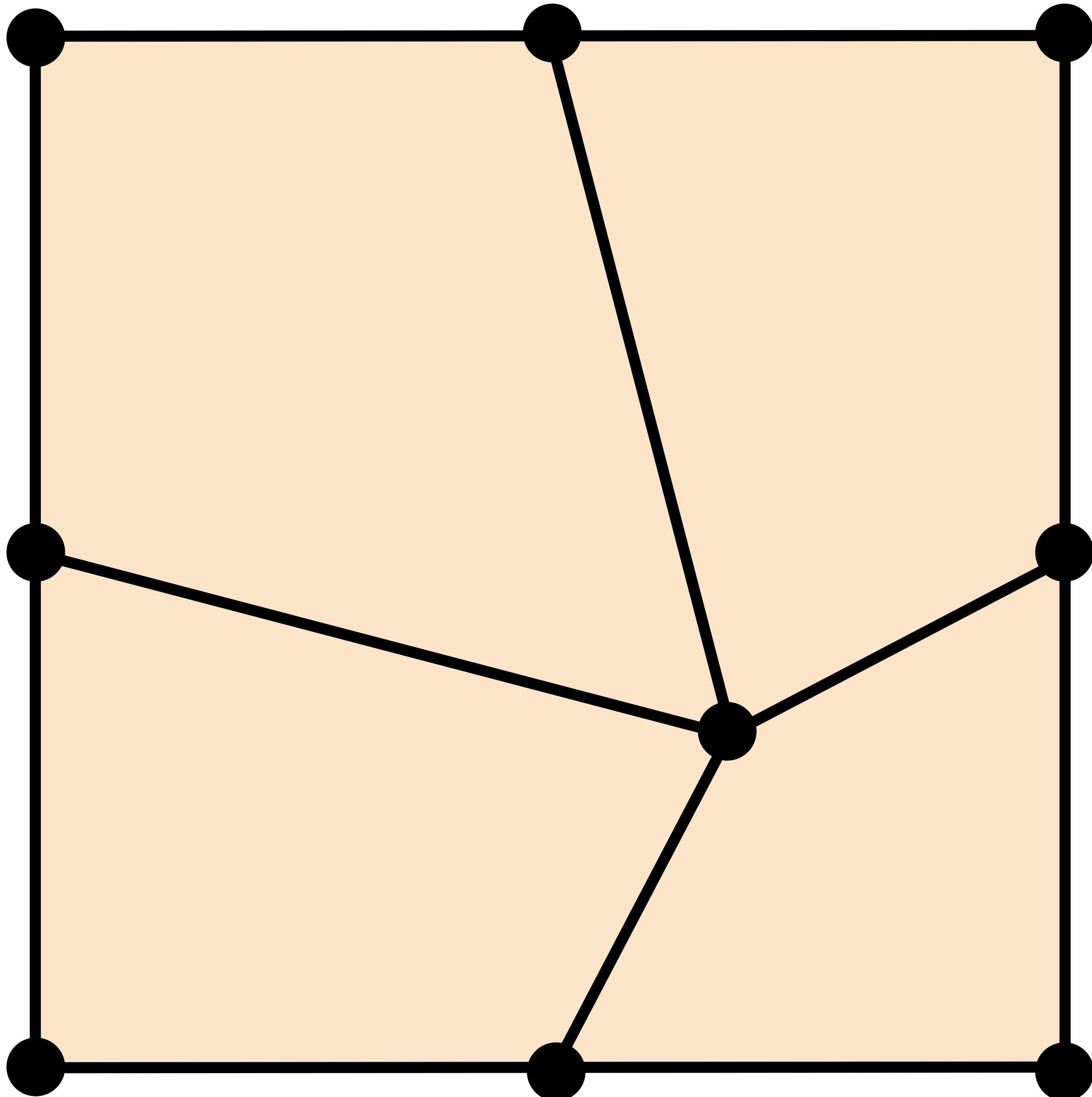


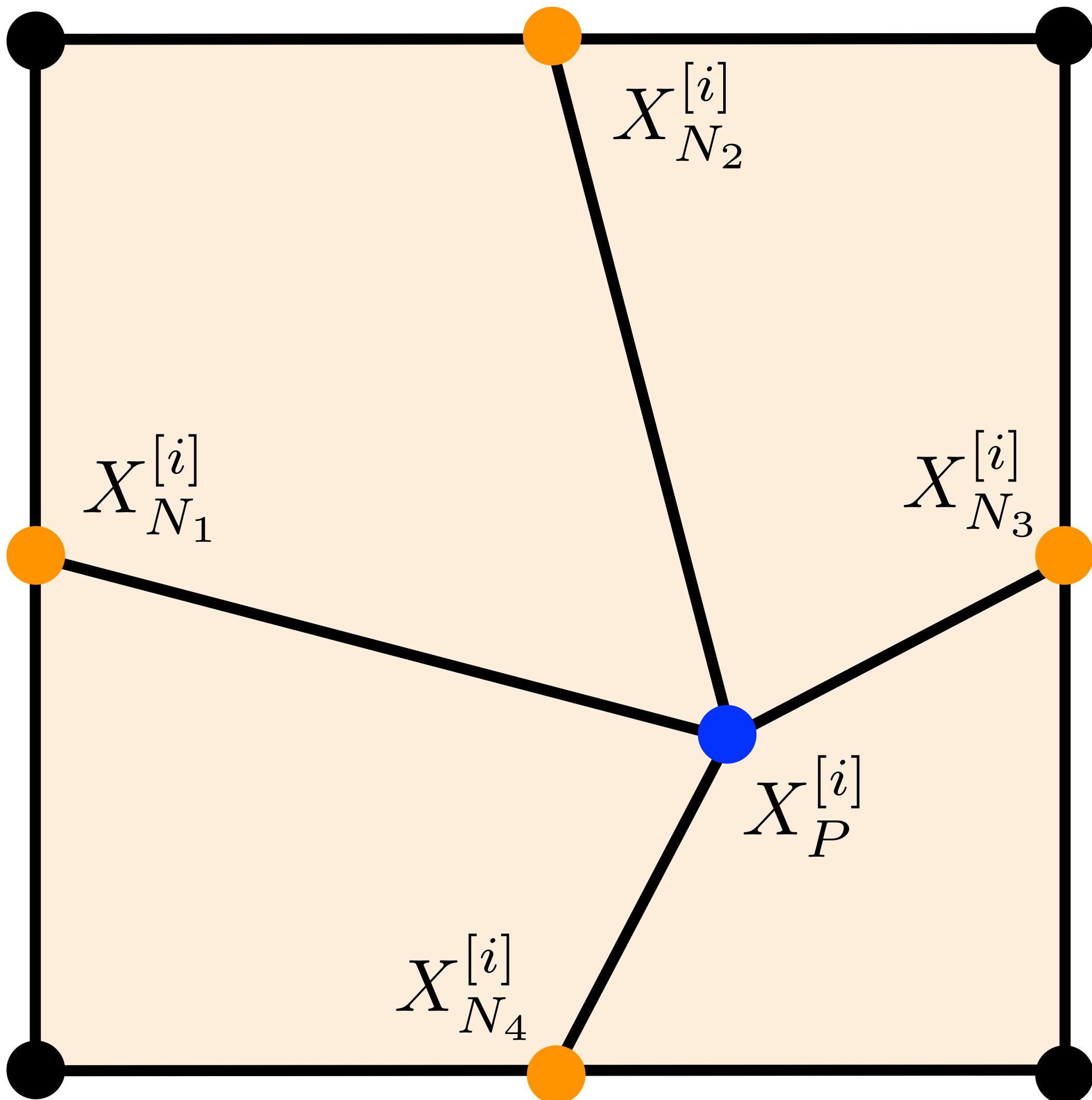
How can we move the mesh points to improve the mesh quality, **without** changing the connectivity?



point-based Laplacian  
smoothing

**Approach I**  
point-based Laplacian





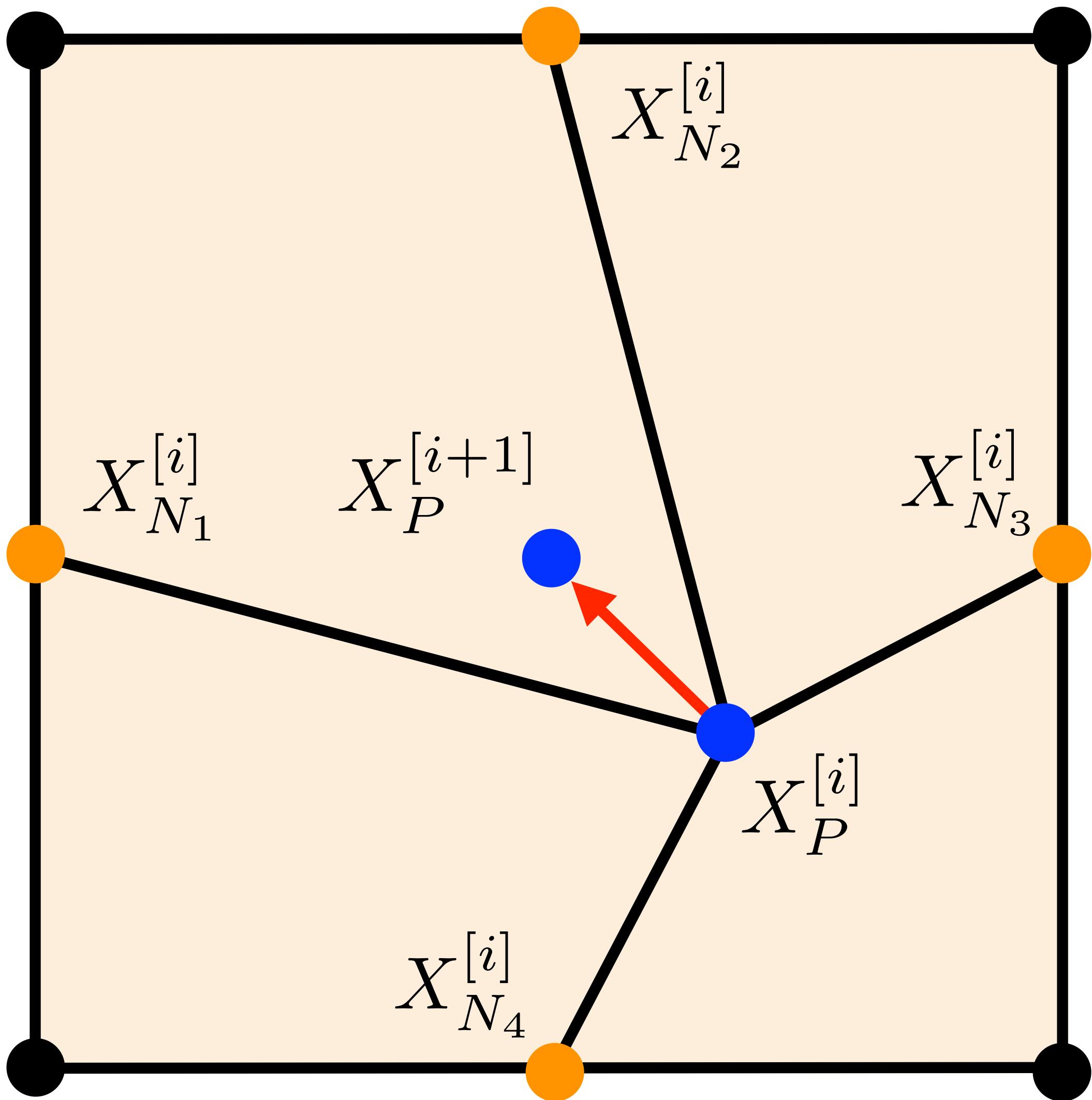
**Approach I**  
point-based Laplacian

$$X_P^{[i+1]} = \sum_{j=1}^n w_j X_{N_j}^{[i]}$$

Where we can calculate the weights as an **arithmetic average**:

$$w_j = \frac{1}{n}$$

number of  
neighbour points



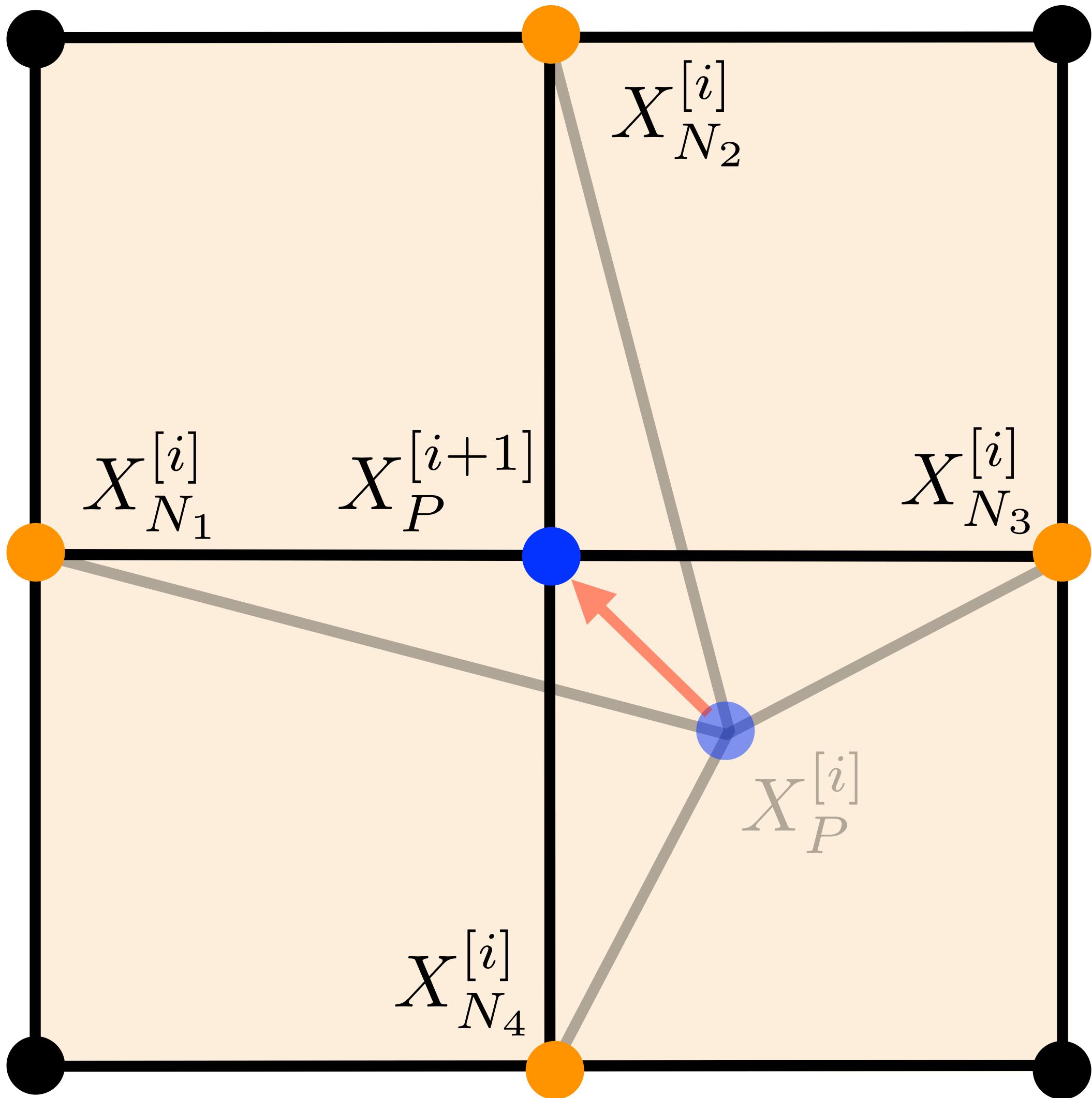
## Approach I point-based Laplacian

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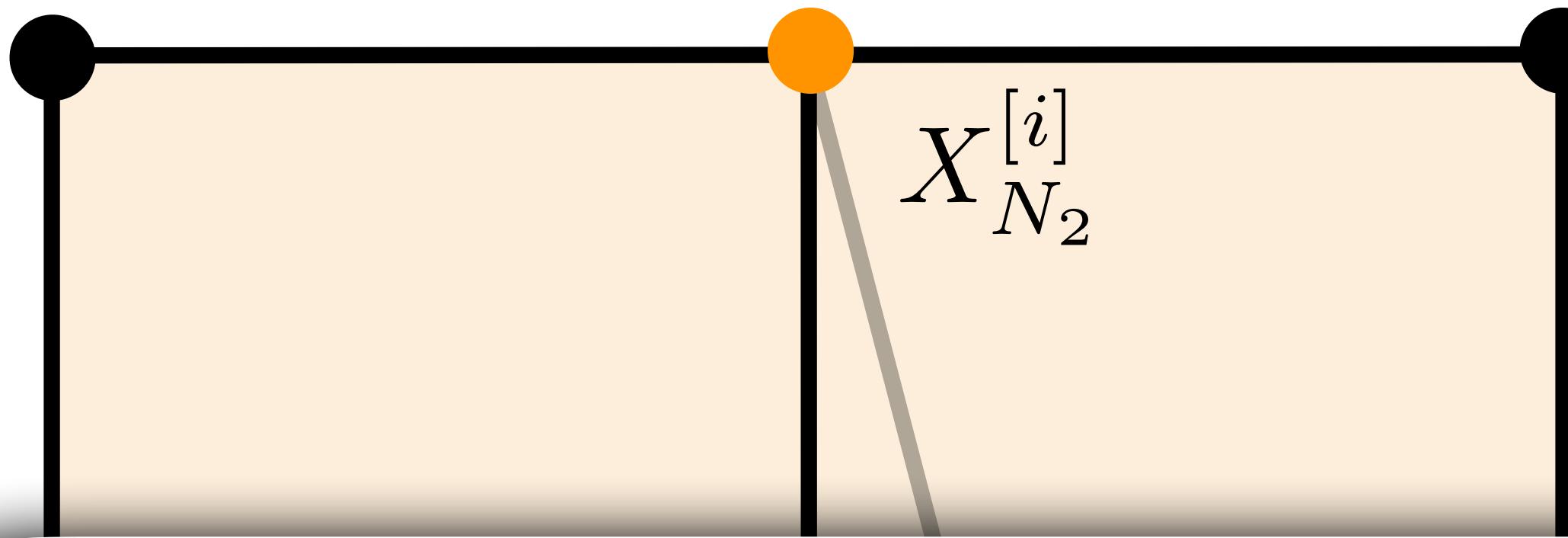
**Approach I**  
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## Approach I point-based Laplacian

**algorithm 1:**

```

for all internal points in the mesh
    newPoint = vector::zero
    for all neighbour points of the current point
        neighbourPointWeight = 1.0;
        newPoint += neighbourPointWeight*neighbourPoint;
        sumWeights += neighbourPointWeight;
    end
    newPoint /= sumWeights;
end

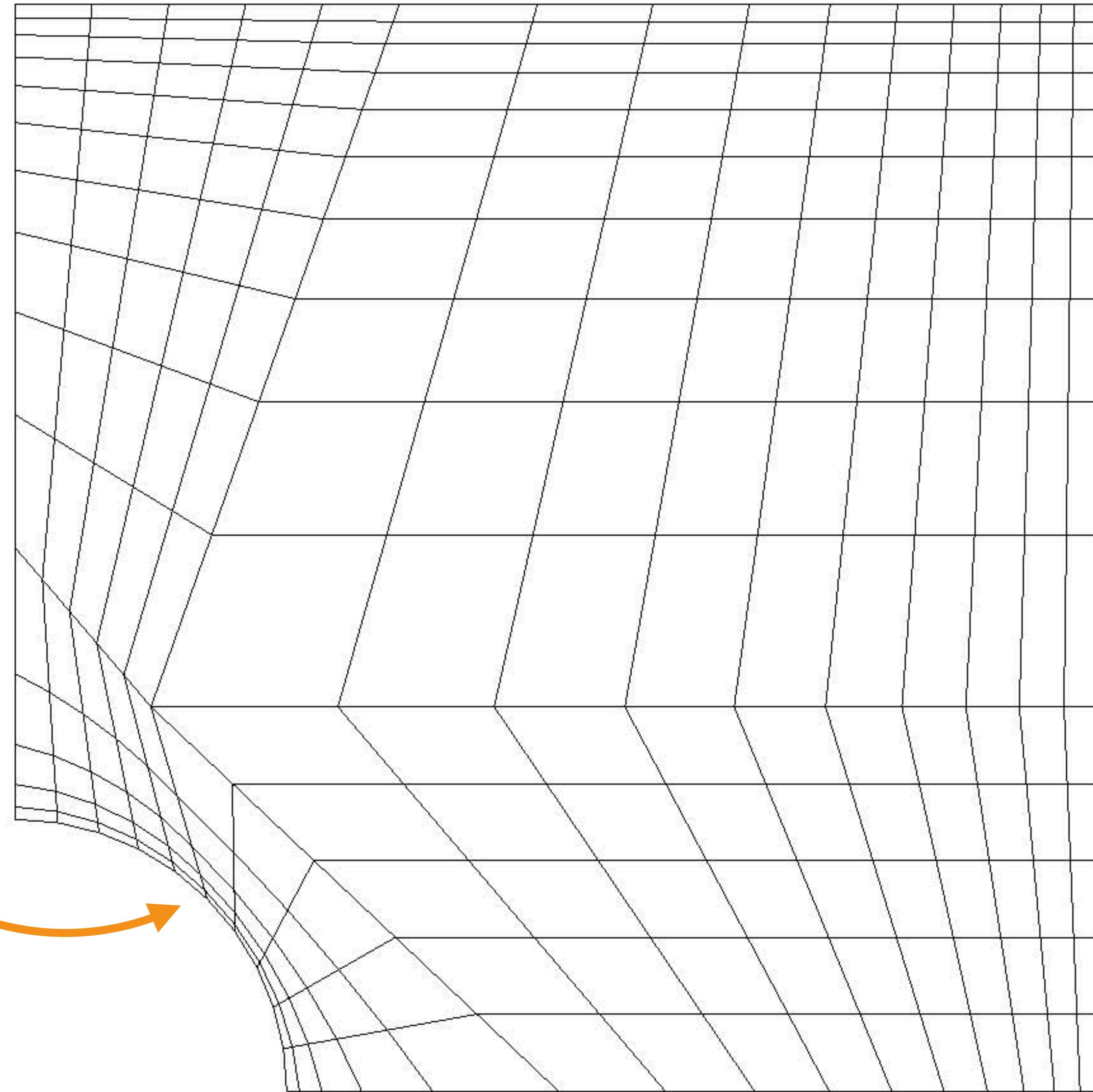
```

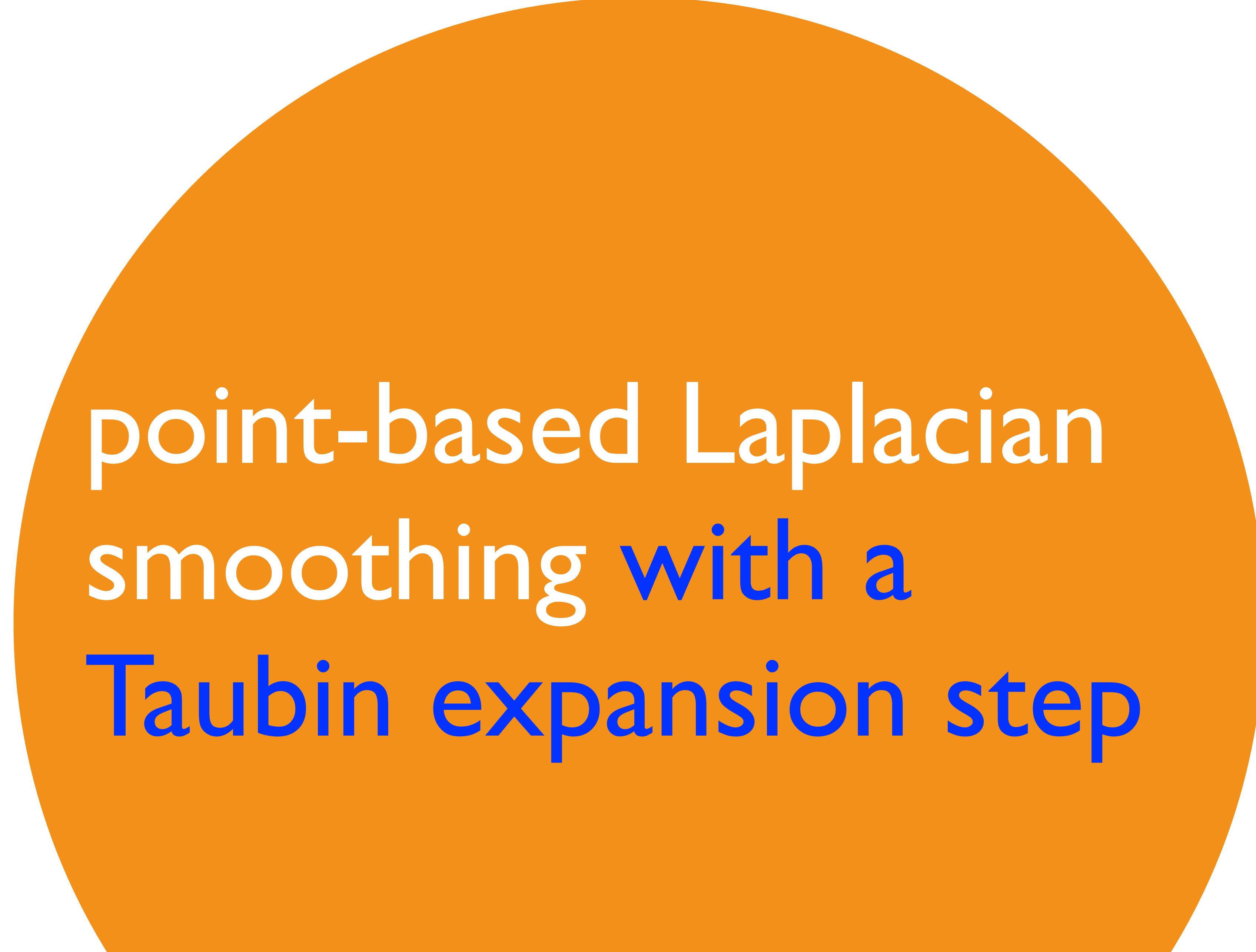
Note: we have to be careful with boundary points, and also in parallel  
neighbour points

# Point-based Laplacian

Laplacian tends to shrink  
cells in concave regions

Iteration: 0





point-based Laplacian  
smoothing **with a**  
**Taubin expansion step**

**Approach 2:** to eliminate the shrinkage effect produce by the standard Laplacian smoothing, a **Taubin expansion step** can be taken after each smoothing step:

$$X_P^{[i+1]} = X_P^{[i]} + \lambda \left( \sum_{j=1}^n w_j X_{N_j}^{[i]} - X_P^{[i]} \right) \quad \text{Laplacian step}$$

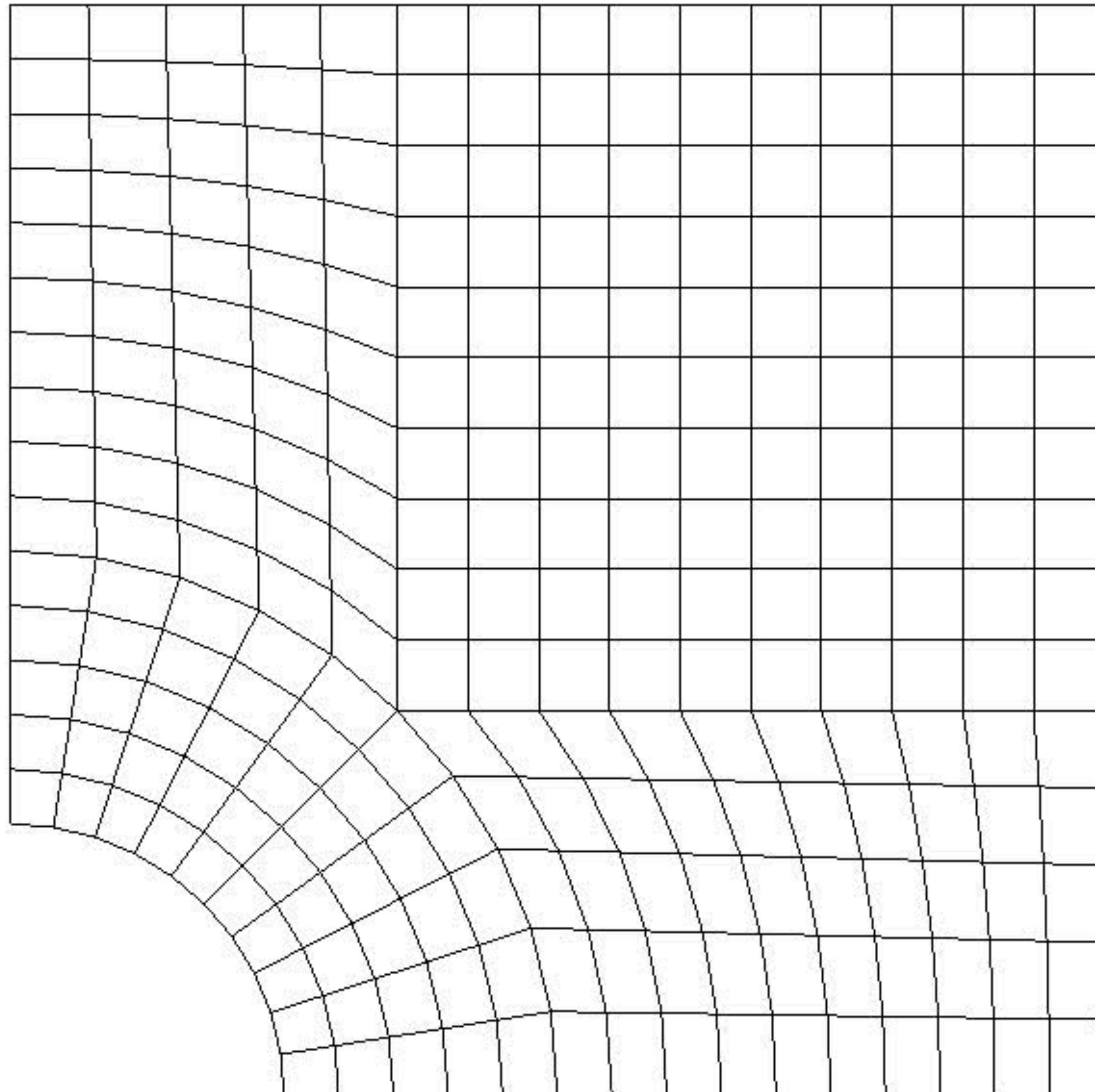
$$X_P^{[i+2]} = X_P^{[i+1]} + \mu \left( \sum_{j=1}^n w_j X_{N_j}^{[i+1]} - X_P^{[i+1]} \right) \quad \text{Taubin step}$$

where  $0 < \lambda < -\mu$

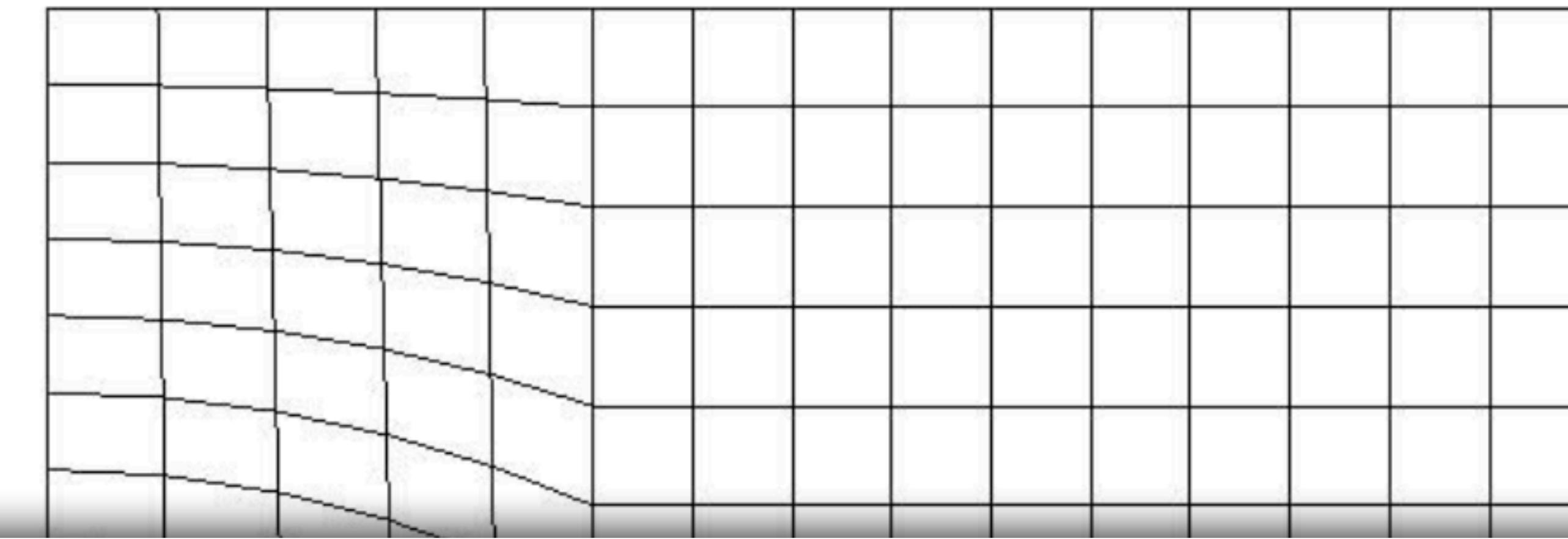
Point-based Laplacian  
with arithmetic  
weights and a Taubin  
expansion step

Adding the Taubin helps  
seems to make the  
method less robust on  
initially bad meshes

Iteration: 0



## Point-based Laplacian with arithmetic weights and a Taubin expansion step



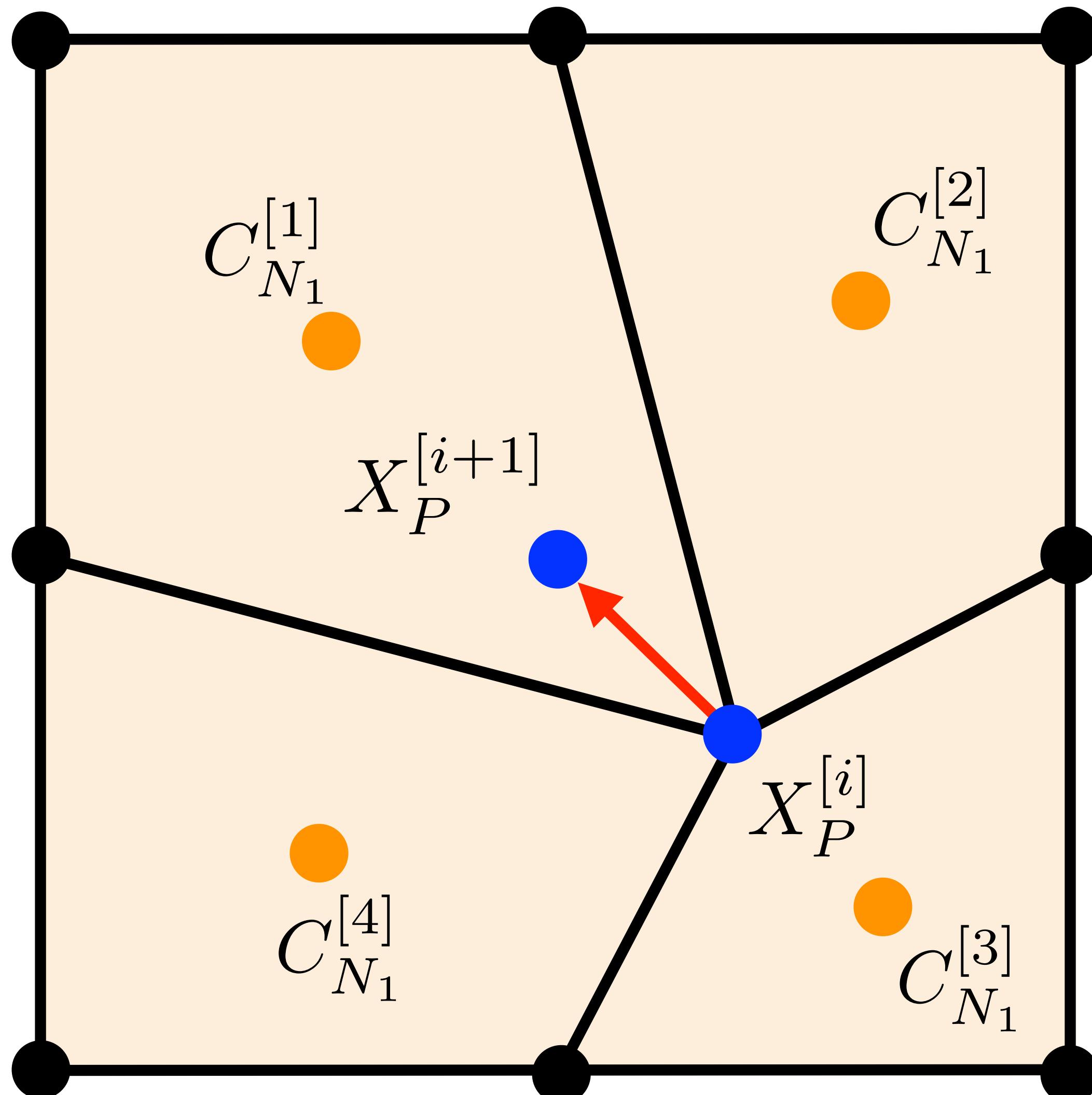
**algorithm 2:**

```
lambda = 0.33; mu = -0.34;
for all internal points in the mesh
    averageNeighbourPoint = vector::zero
    for all neighbour points of the current point
        neighbourPointWeight = 1.0;
        averageNeighbourPoint += neighbourPointWeight*neighbourPoint;
        sumWeights += neighbourPointWeight;
    end
    averageNeighbourPoint /= sumWeights;
    if laplacianStep
        newPoint = oldPoint + lambda*(averageNeighbourPoint - oldPoint)
    else // taubin step
        newPoint = oldPoint + mu*(averageNeighbourPoint - oldPoint)
    end
end
```

point-based Laplacian  
smoothing **with**  
**cell-volume weights**

### Approach 3

point-based Laplacian with cell-volume weights



$$X_P^{[i+1]} = \sum_{j=1}^{\text{ncells}} w_j C_{N_j}^{[i]}$$

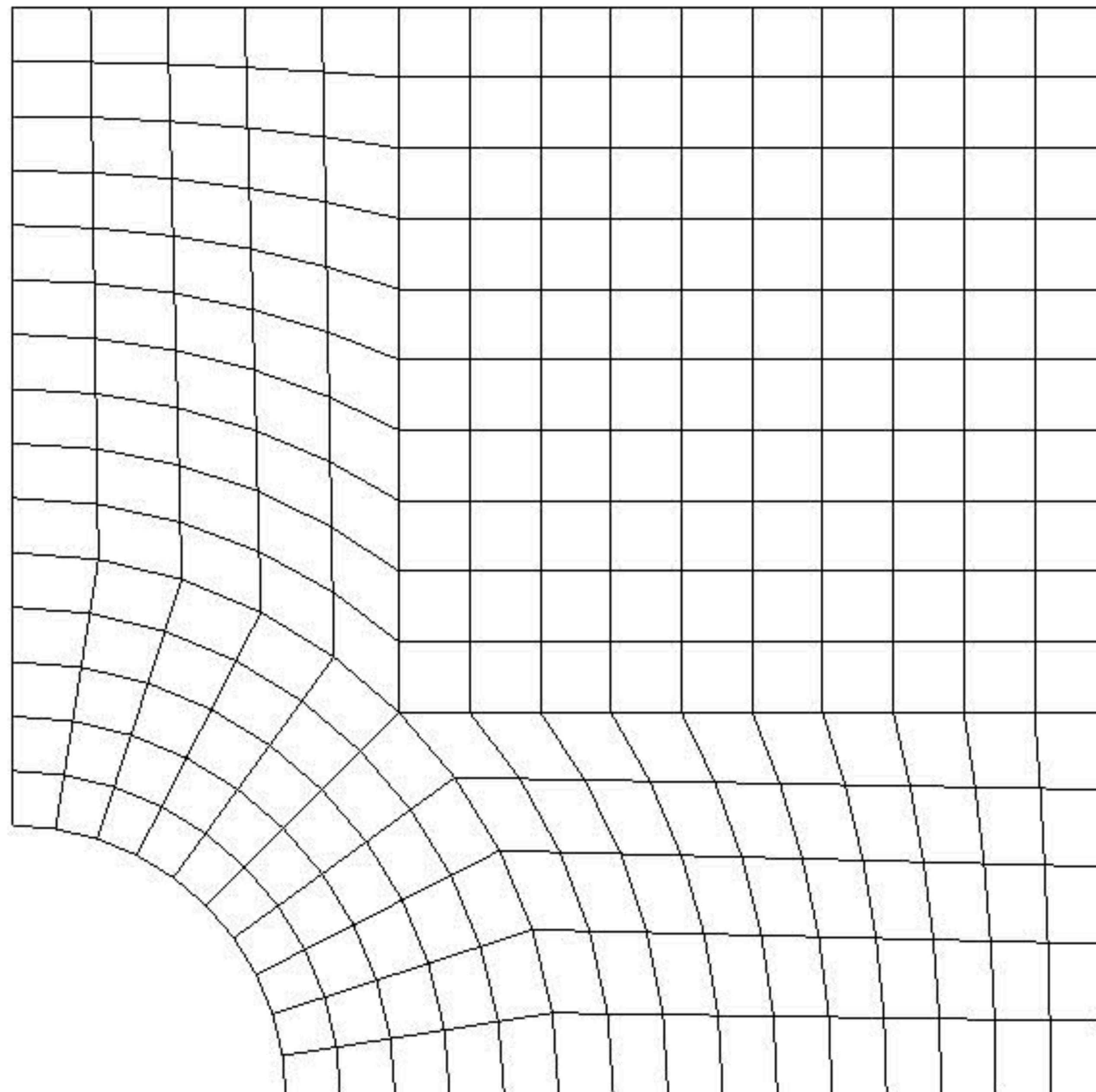
cell centre

We can use the cell volumes to calculate the weights:

$$w_j = \frac{V_{N_j}^{[i]}}{\sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]}}$$

Point-based Laplacian  
with cell-volume  
weights

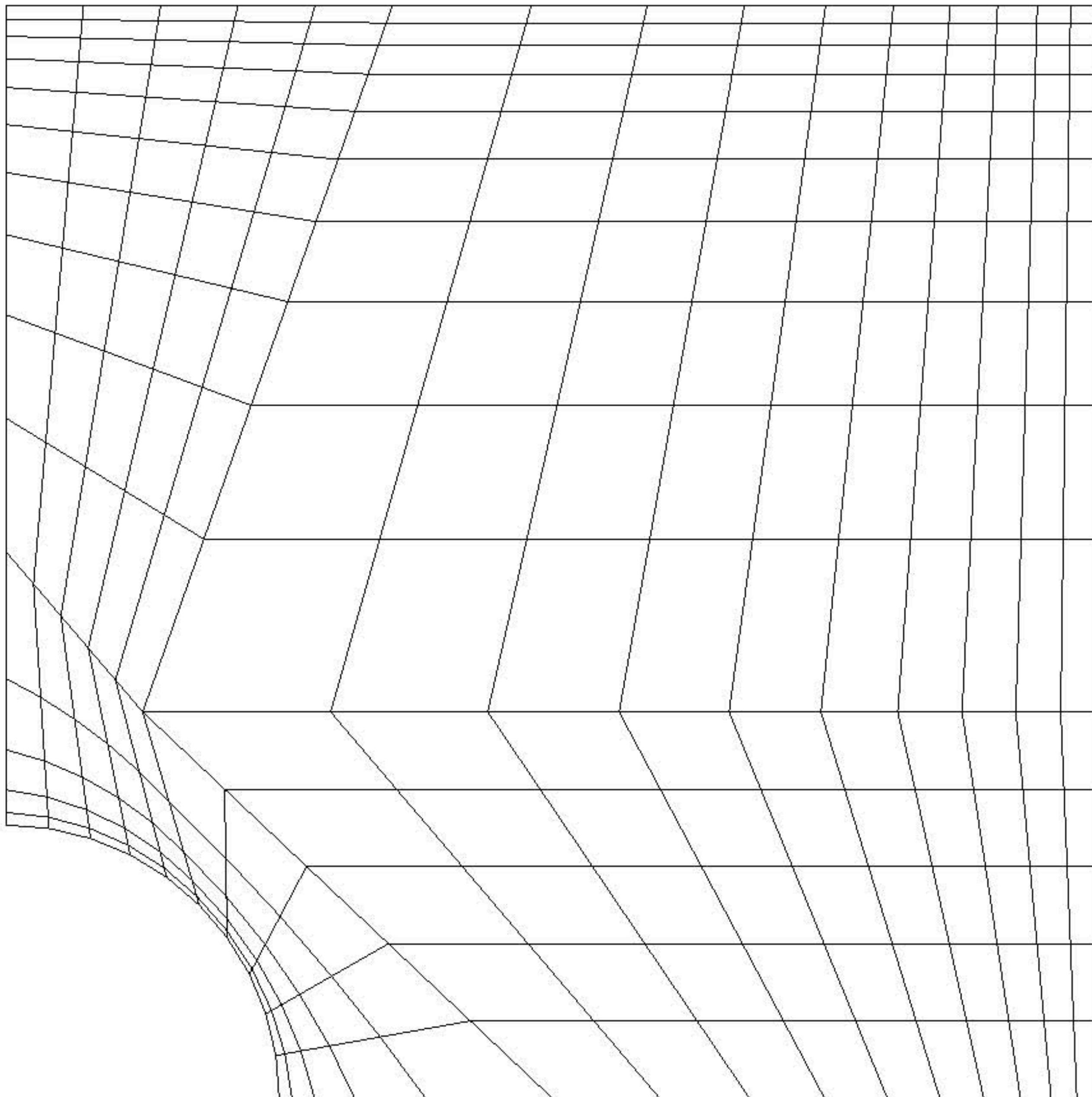
Iteration: 0



Point-based Laplacian  
with cell-volume  
weights

lambda = 1.0

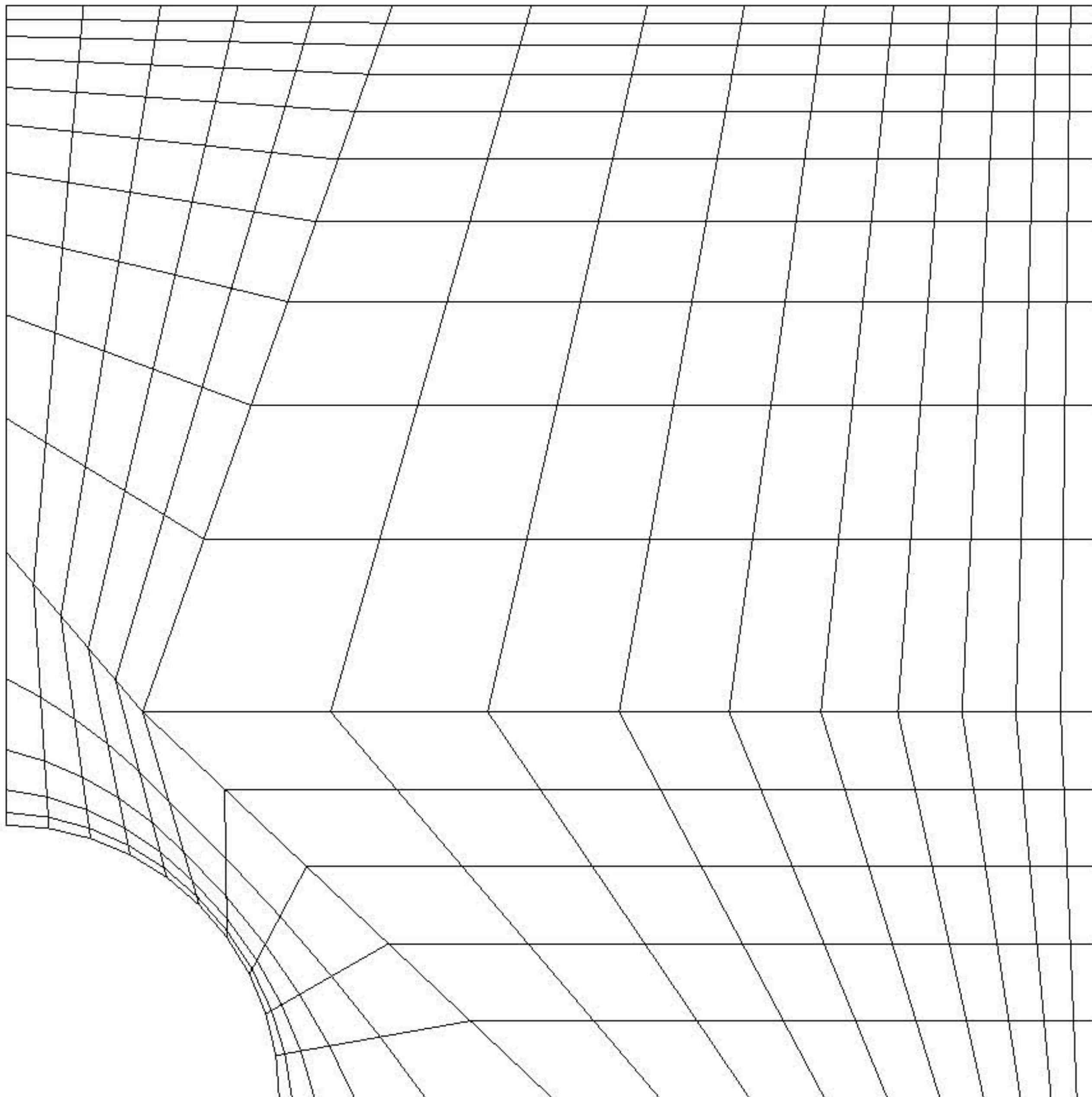
Iteration: 0



Point-based Laplacian  
with cell-volume  
weights

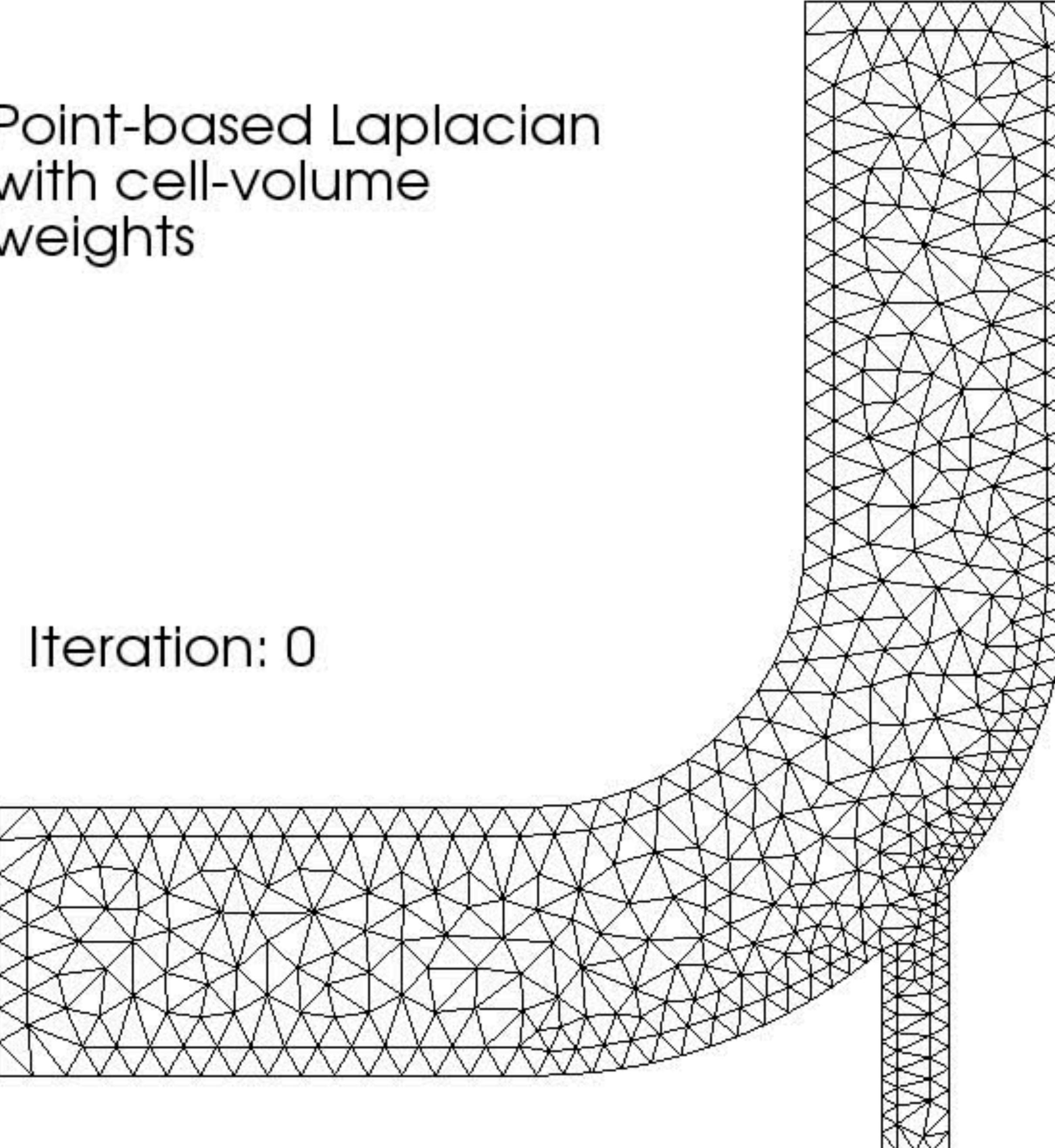
lambda = 0.5

Iteration: 0

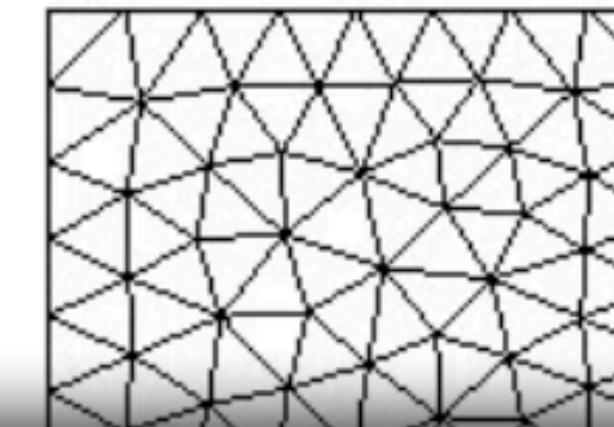


Point-based Laplacian  
with cell-volume  
weights

Iteration: 0



Point-based Laplacian  
with cell volume



**Before**

Max aspect ratio = 6.7264 OK.

Min volume = 0.521792. Max volume = 7.36354.

**Mesh non-orthogonality** Max: 36.302 average: 11.1868

**Max skewness** = 0.500248 OK.

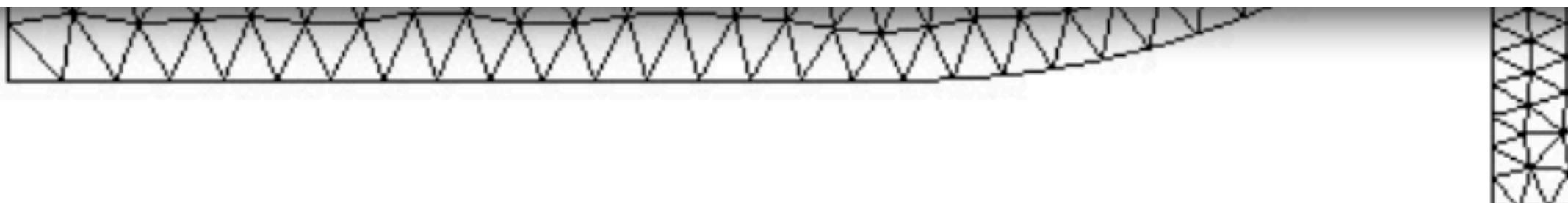
**After**

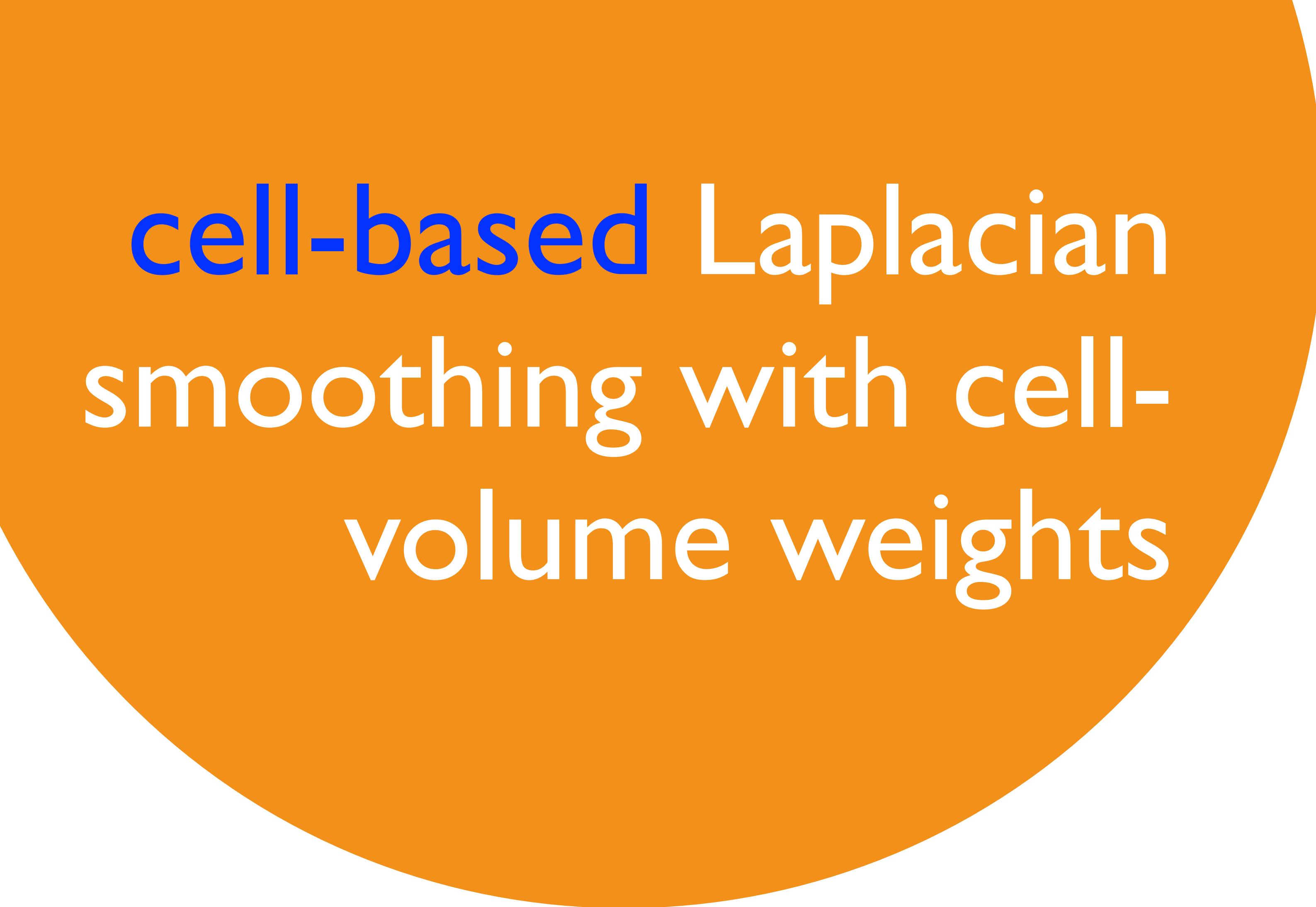
Max aspect ratio = 4.57246 OK.

Min volume = 0.943911. Max volume = 6.3874.

**Mesh non-orthogonality** Max: 33.0798 average: 10.467

**Max skewness** = 0.363852 OK.

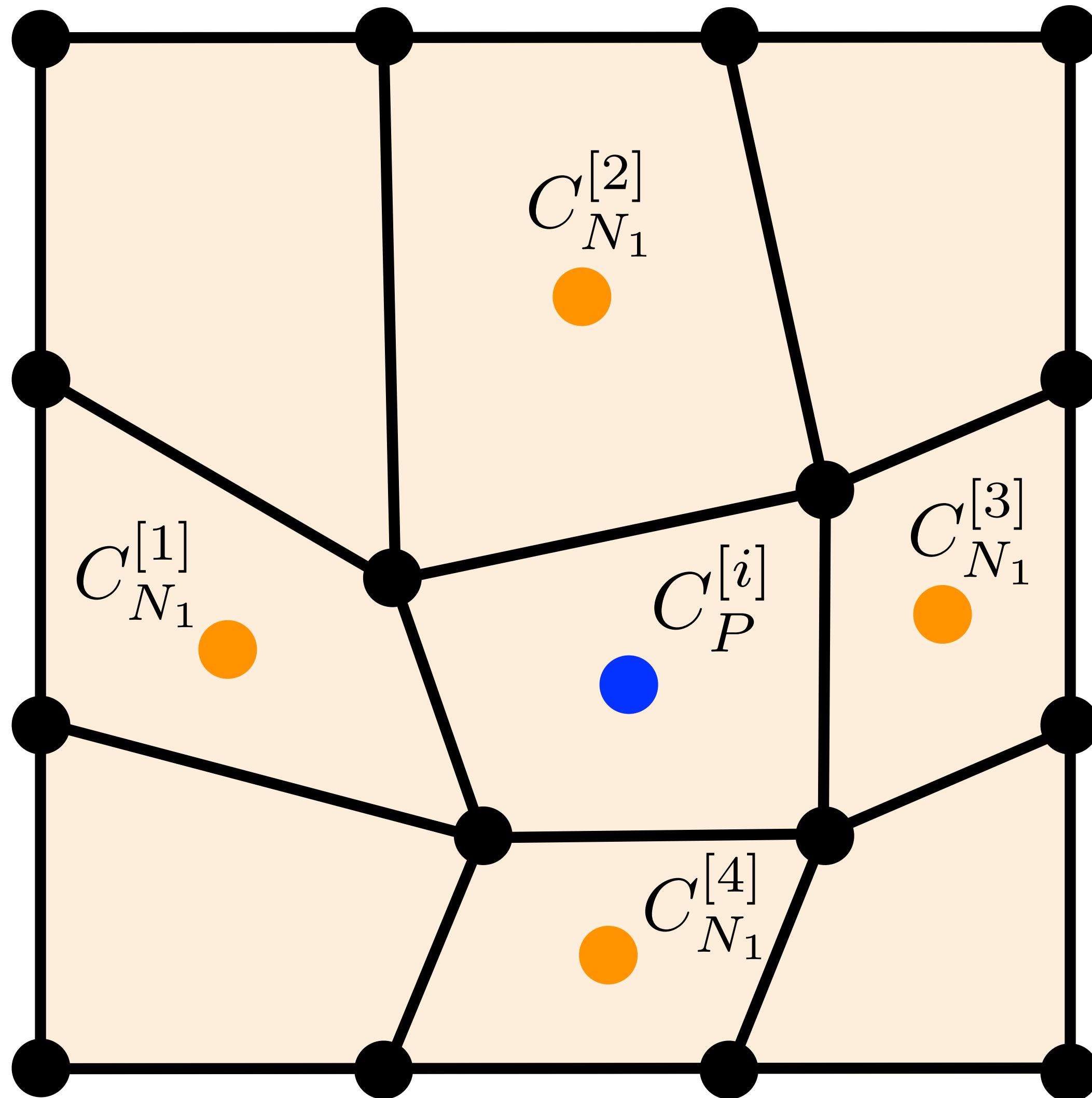




**cell-based** Laplacian  
smoothing with cell-  
volume weights

### Approach 3

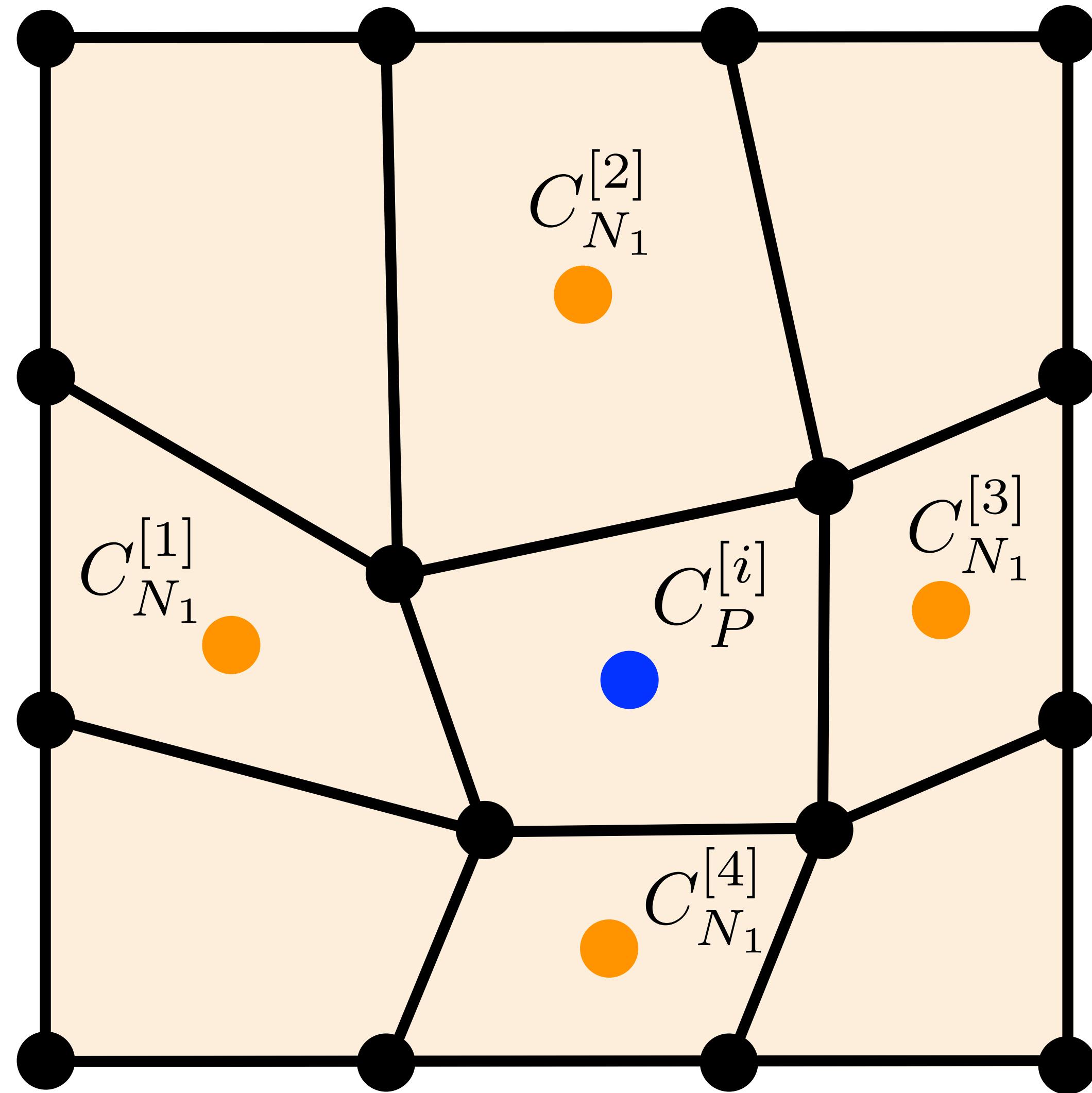
**cell-based Laplacian with cell-volume weights**



$$C_P^{[i+1]} = \sum_{j=1}^{\text{ncells}} w_j C_{N_j}^{[i]}$$

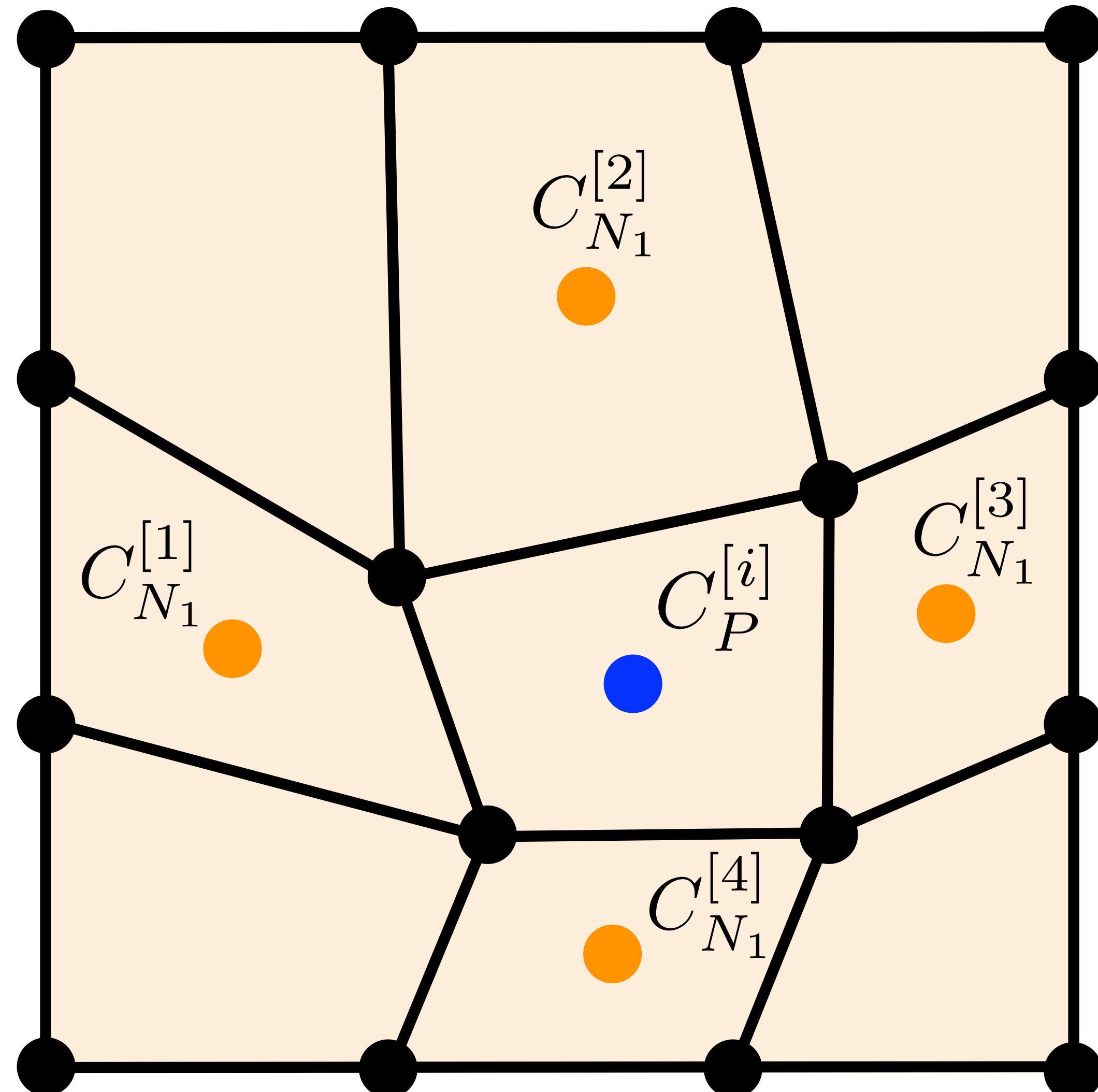
Where the cell-volume weights are calculated as:

$$w_j = \frac{V_{N_j}^{[i]}}{\sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]}}$$



### Advantage

It can be implemented in an **implicit** fashion using the finite volume infrastructure within OpenFOAM.



### Disadvantage

However, this approach gives the mesh motion at the cell centres, which must be interpolated to the points

A simple method is to take an arithmetic average.

# Deriving an implicit finite volume approach

The explicit method iterates over the equation to the right until convergence is achieved; note the left-hand side has iterator  $[i + 1]$ , whereas the right-hand side is  $[i]$ .

$$C_P^{[i+1]} = \sum_{j=1}^{\text{ncells}} w_j C_{N_j}^{[i]}$$

This equation can be re-written solely in terms of unknown cell-centre positions as:

$$-C_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} w_j C_{N_j}^{[i+1]} = 0$$

$$-C_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} w_j C_{N_j}^{[i+1]} = 0$$



$$-C_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} \left[ \frac{V_{N_j}^{[i]}}{\sum_{k=1}^{\text{ncells}} V_{N_k}^{[i]}} \right] C_{N_j}^{[i+1]} = 0$$



$$- \left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] C_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} C_{N_j}^{[i+1]} = 0$$

Note: this is the  
definition of a discrete  
Laplacian operator

$$-\left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] C_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} C_{N_j}^{[i+1]} = 0$$

↓

$$-\left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] \left[ C_P^{[\text{orig}]} + U_P^{[i+1]} \right] + \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \left[ C_{N_j}^{[\text{orig}]} + U_{N_j}^{[i+1]} \right] = 0$$

↓

cell centre  
displacements

$$-\left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] U_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} U_{N_j}^{[i+1]} = \left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] C_P^{[\text{orig}]} - \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} C_{N_j}^{[\text{orig}]}$$

$$-\left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] U_P^{[i+1]} + \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} U_{N_j}^{[i+1]} = \left[ \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} \right] C_P^{[\text{orig}]} - \sum_{j=1}^{\text{ncells}} V_{N_j}^{[i]} C_{N_j}^{[\text{orig}]}$$



$$\nabla \cdot (V \nabla U) = -\nabla \cdot (V \nabla X)$$

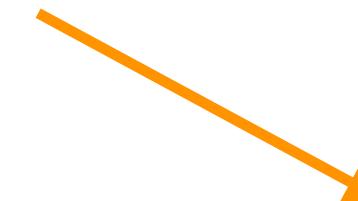


$$\nabla \cdot (V \nabla U) = -\nabla \cdot (V \mathbf{I})$$



$$\nabla \cdot (V \nabla U) = -\nabla V$$

$$\nabla \cdot (V \nabla U) = -\nabla V$$



Laplacian of displacement, where the cell-volume field is the diffusivity: cells with smaller volumes will tend to increase in size more

The gradient of the cell-volume field is the driving force: smaller cells will move towards larger cells

The cell volumes are a function of the displacements so some outer iterations may be required

solve

(

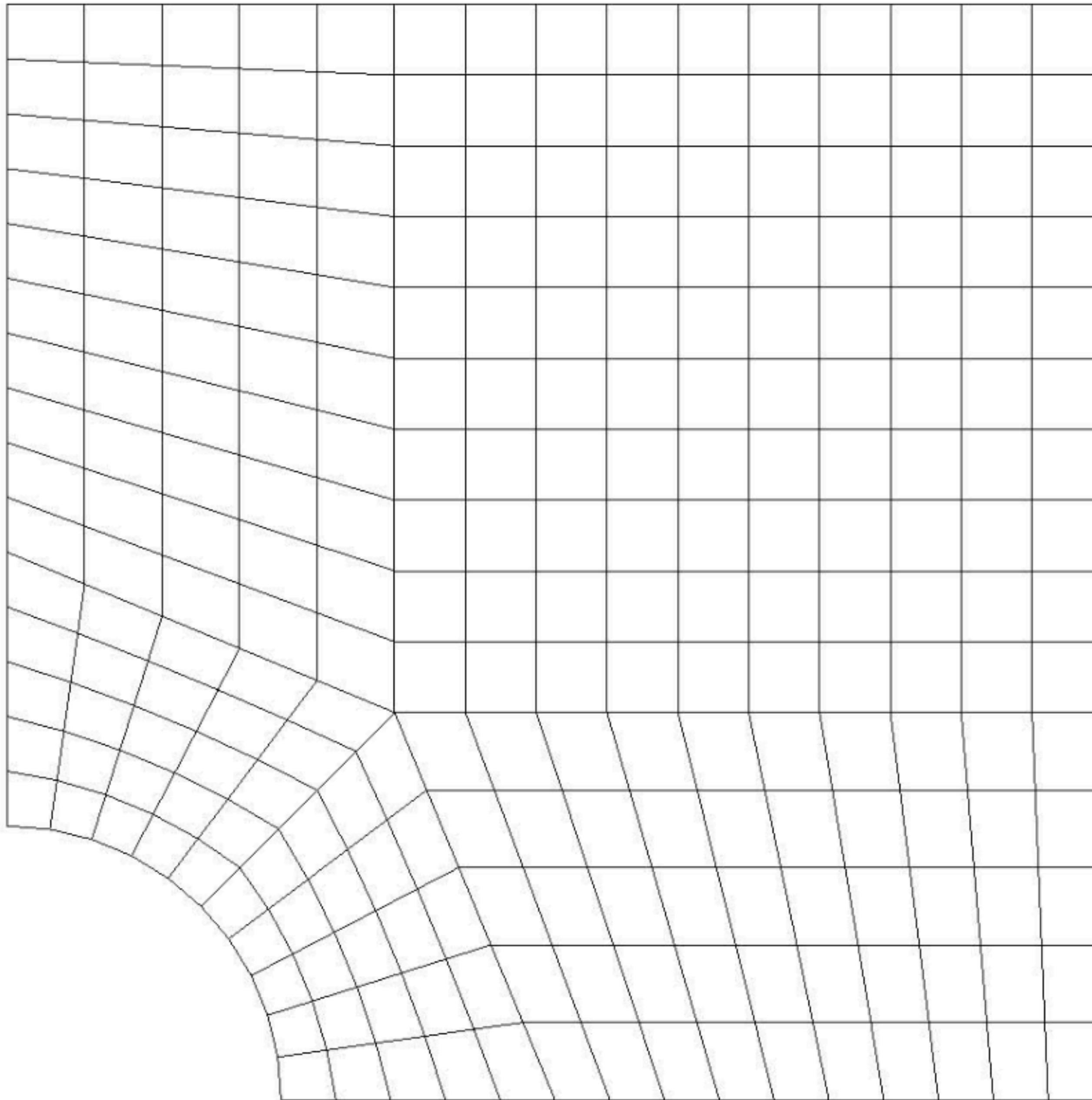
`fvm::laplacian(cellVolumes, U)`

`== -fvc::grad(cellVolumes)`

) ;

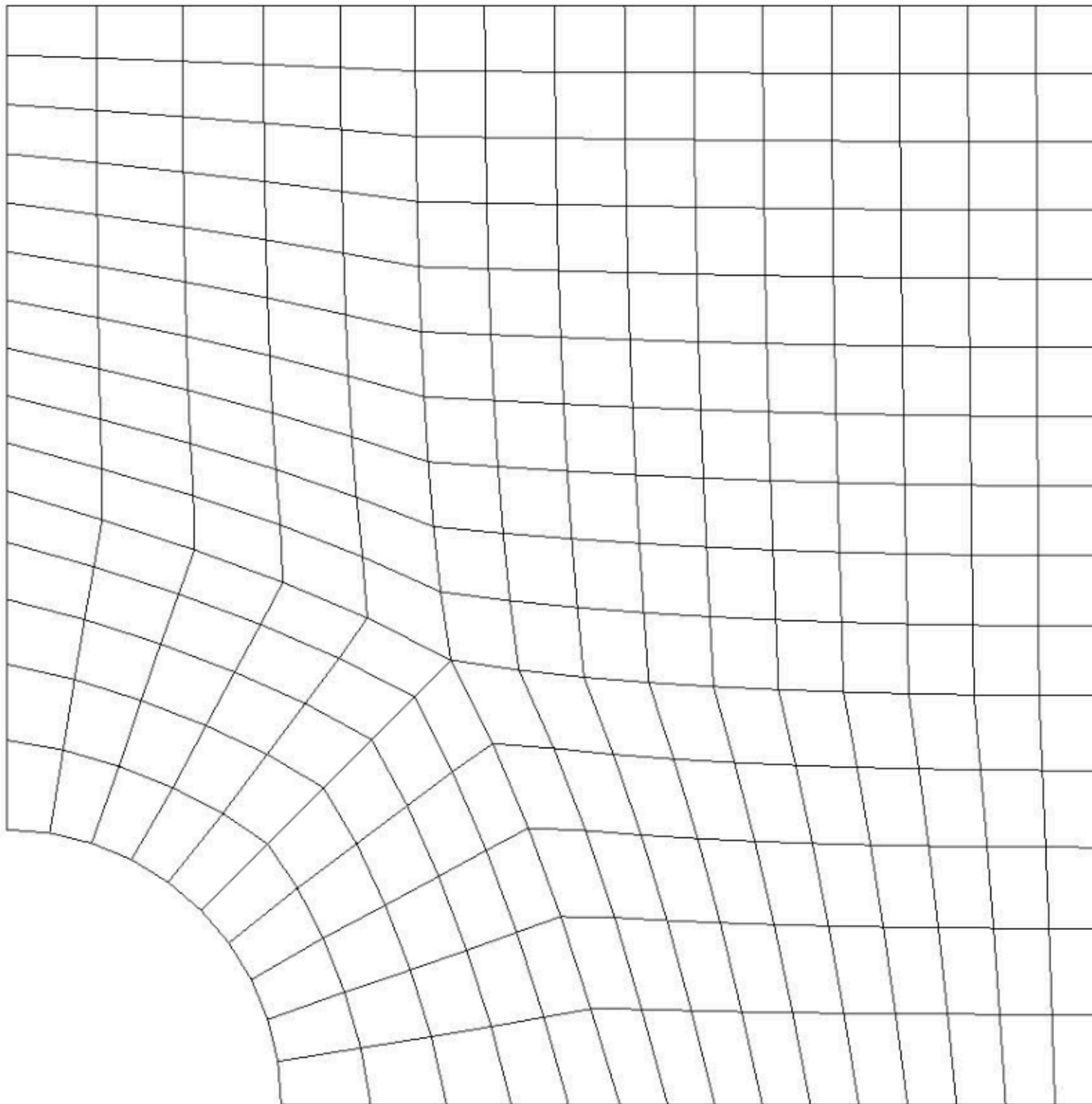
# Implicit cell-based Laplacian with cell-volume weights

Iteration: 0



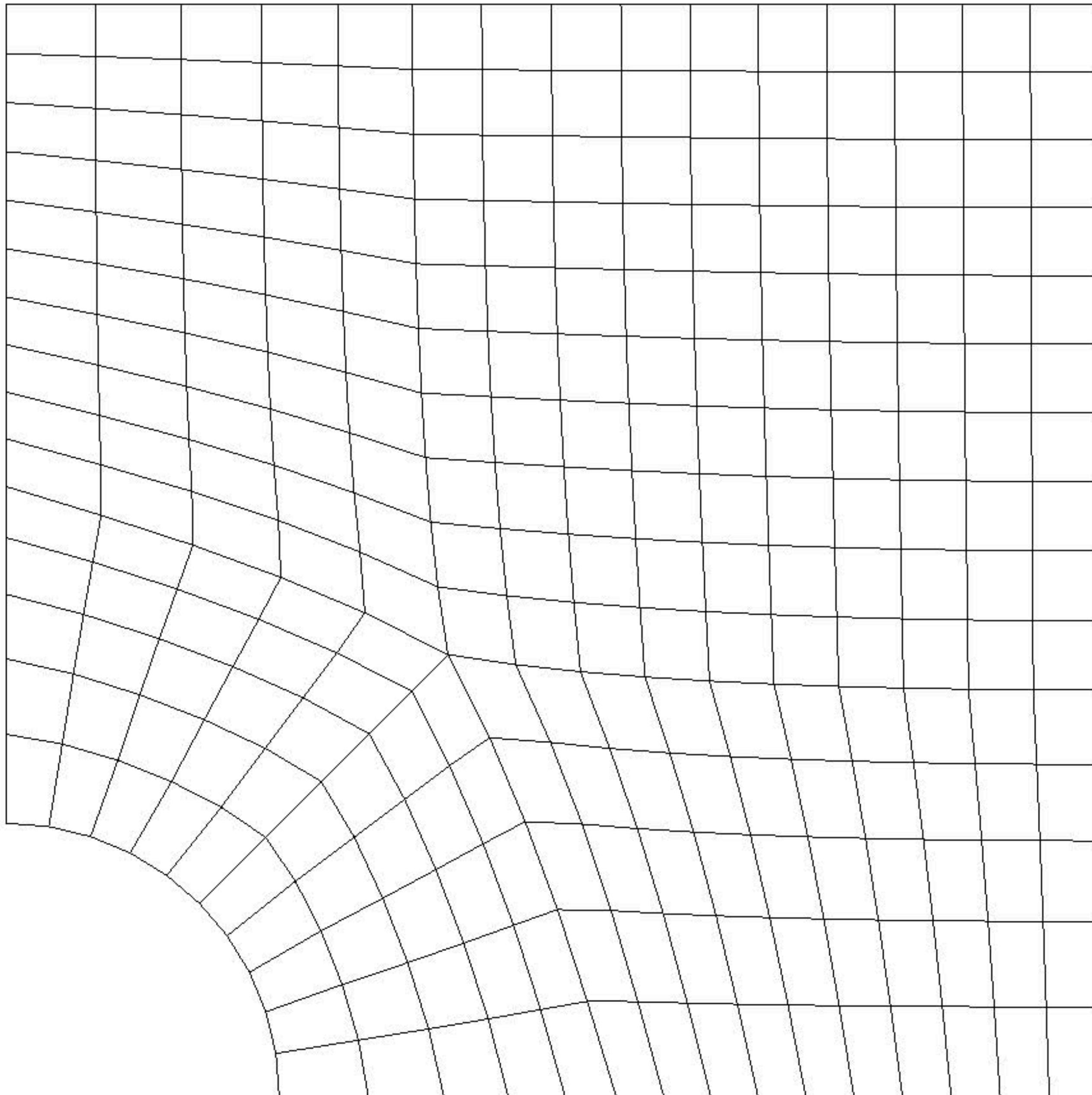
# Implicit cell-based Laplacian with cell-volume weights

Iteration: 1

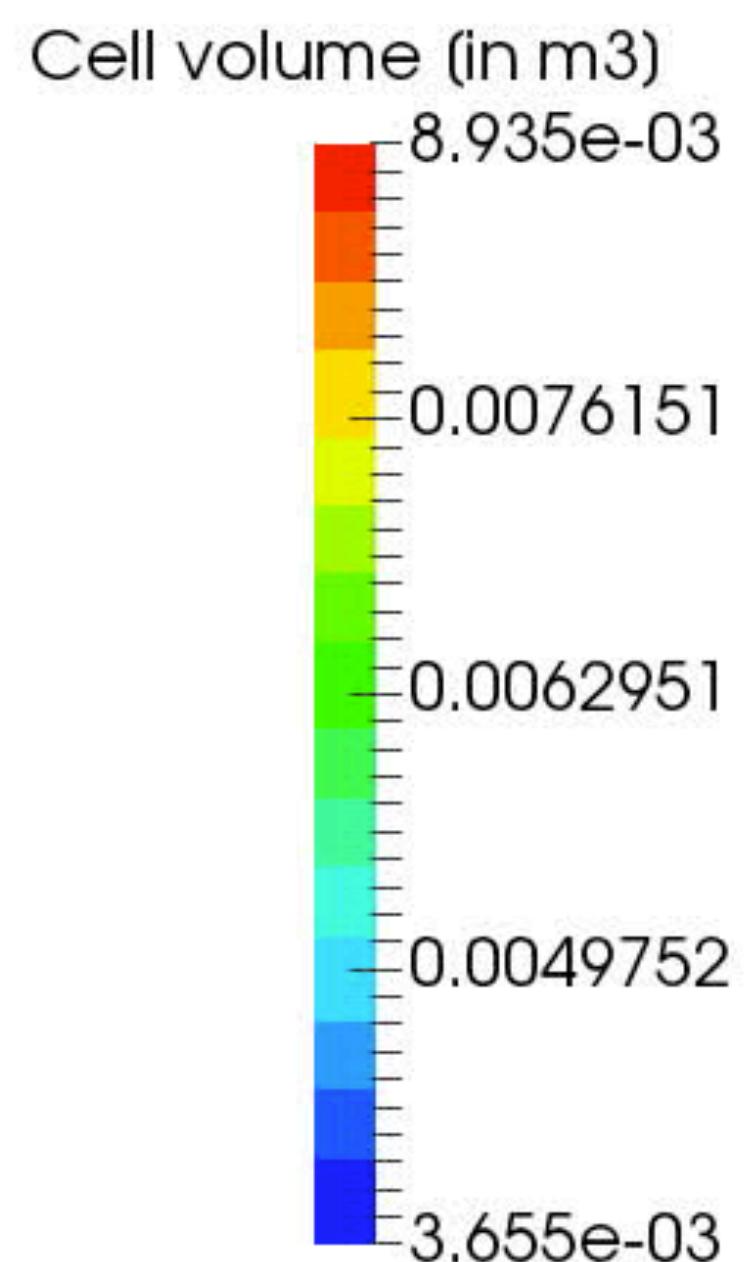


# Implicit cell-based Laplacian with cell-volume weights

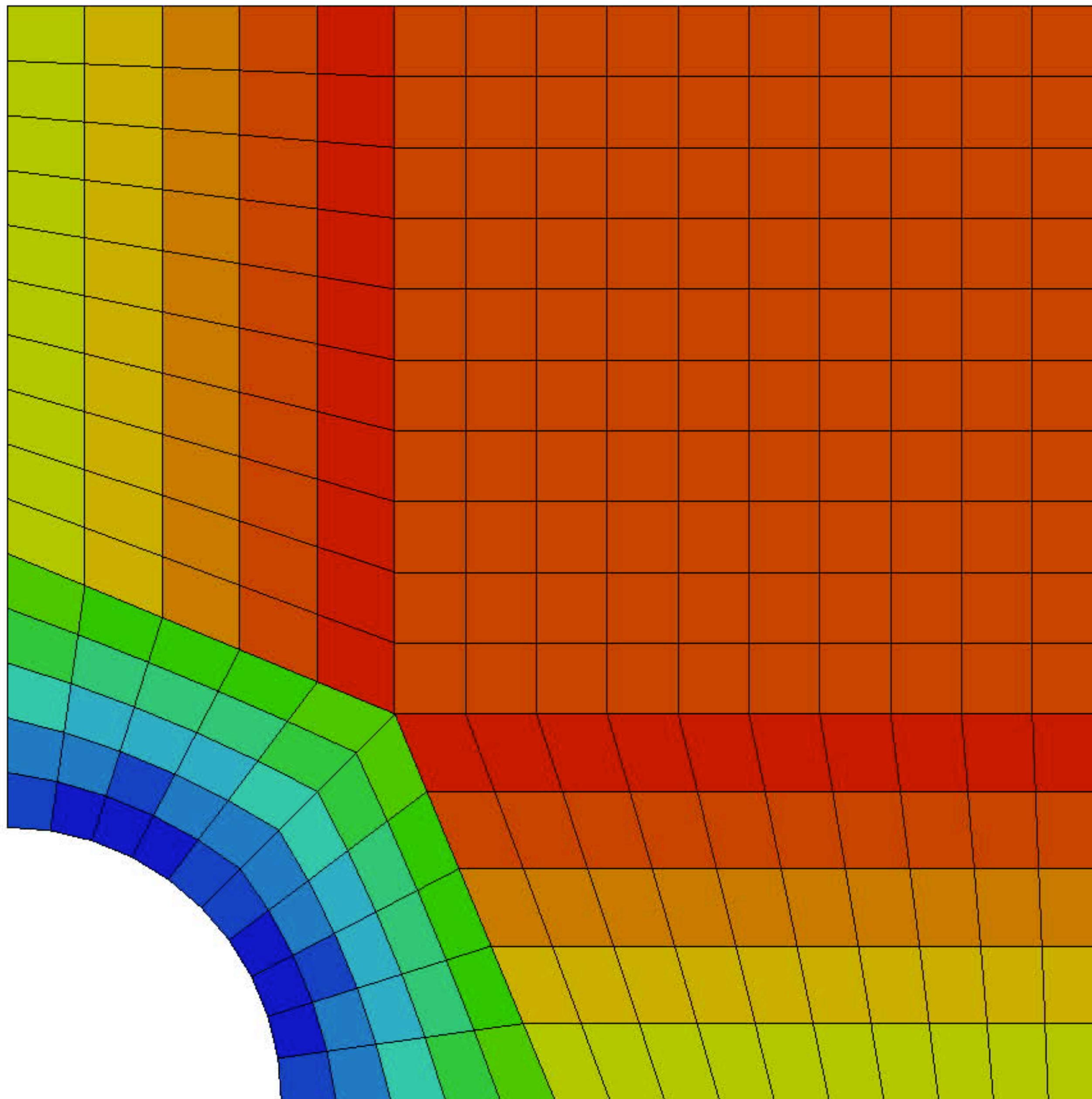
Iteration: 1



# Implicit cell-based Laplacian with cell-volume weights



Iteration: 0



# Summary & Next Steps

# Summary & next steps

- A number of mesh smoothing procedures have been implemented and examined:
  - *explicit* point-based Laplacian
  - *explicit* point-based Laplacian with a Taubin expansion step
  - *explicit* point-based Laplacian with cell-volume weights: **most robust**
  - *implicit* cell-based Laplacian with cell-volume weights: **fastest**

## Next Steps:

- Examine the smoothing procedures on more complex 2-D and 3-D meshes
- Incorporate the mesh smoothing procedure into an Arbitrary Lagrangian-Eulerian method for elasto-plastic analysis (metal forming)

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