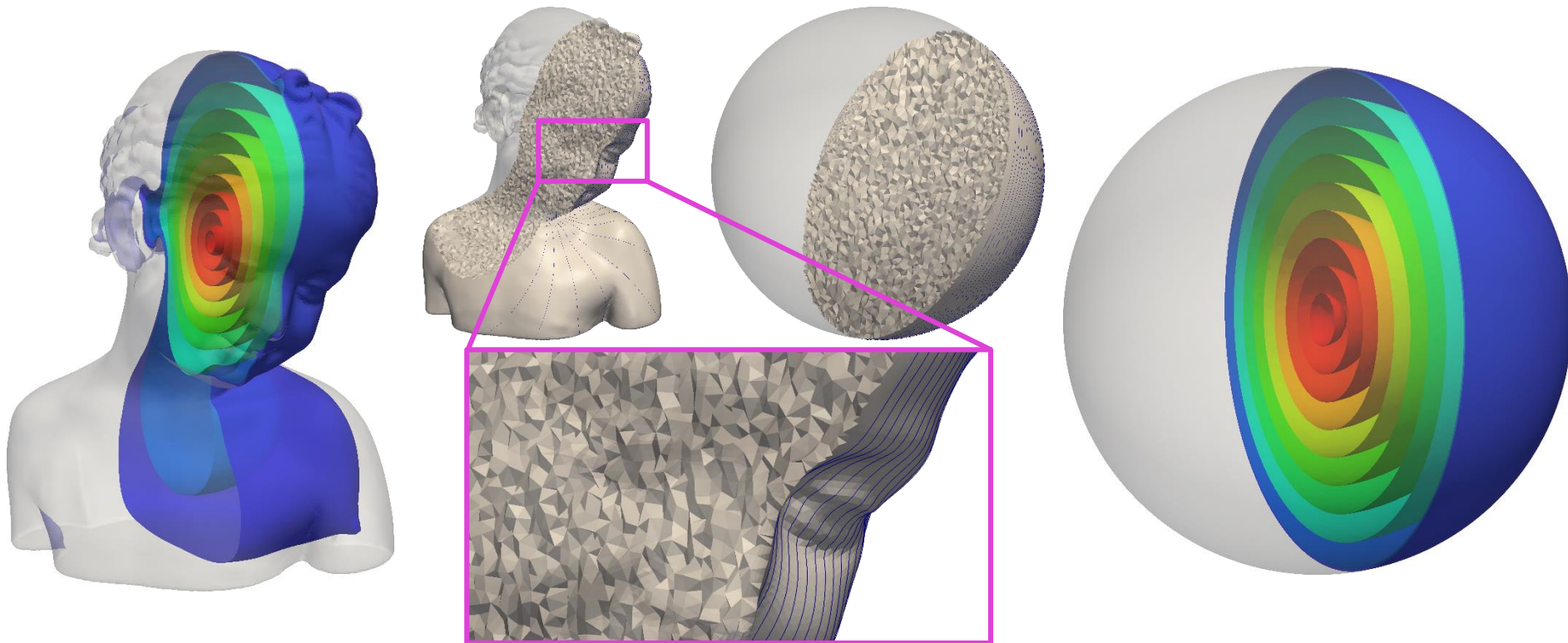


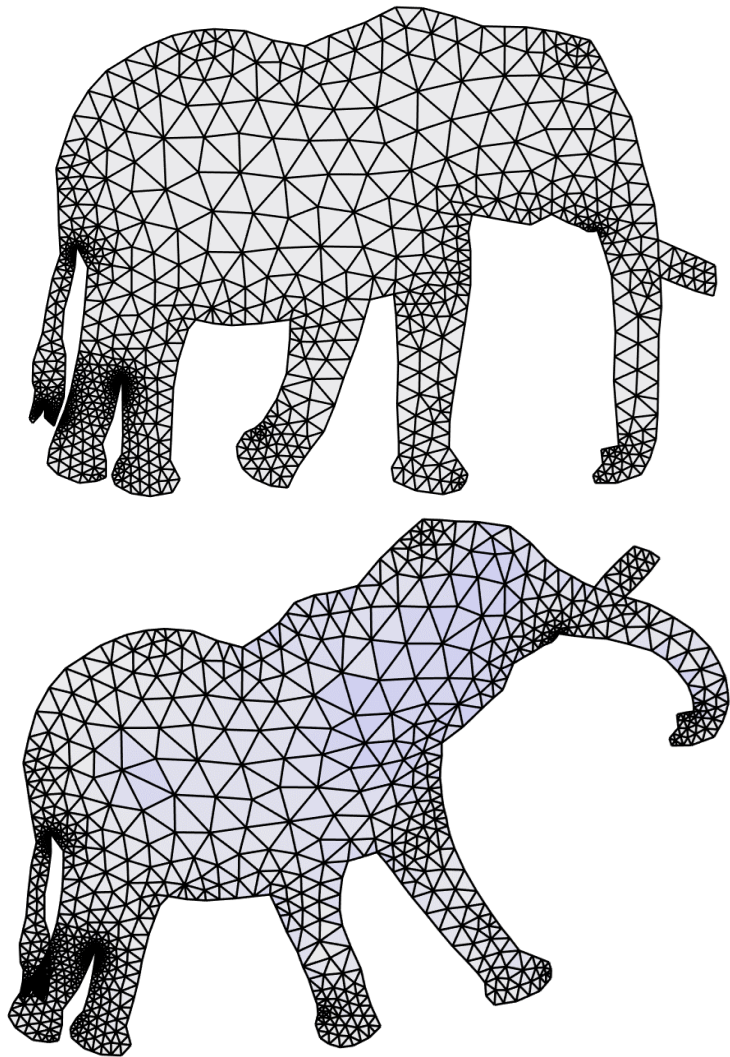


# Large Scale Bounded Distortion Mappings

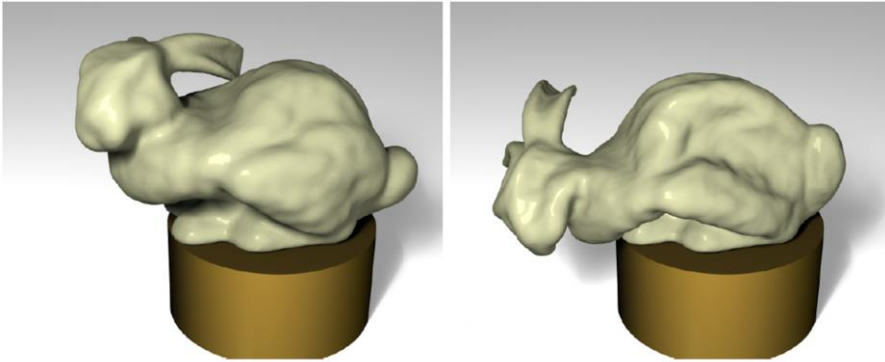
Shahar Kovalsky, Noam Aigerman, Ronen Basri and Yaron Lipman



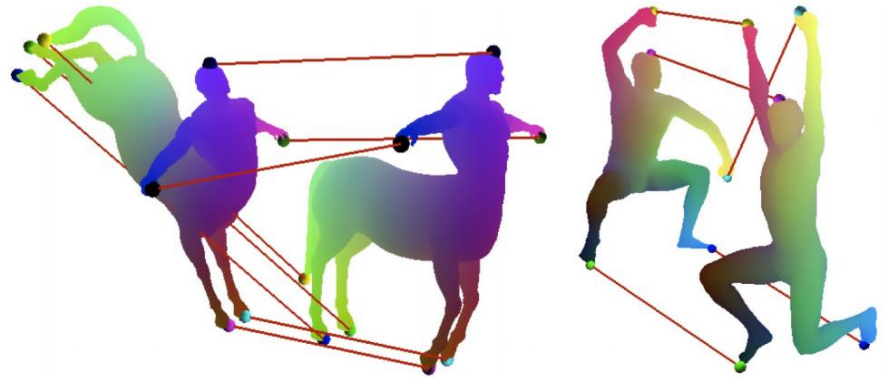
# Mappings



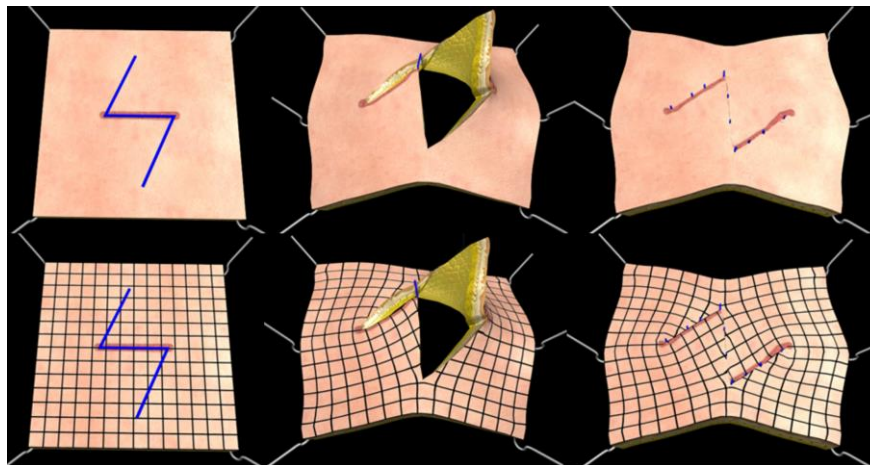
# Applications



[Wang et al. 2010]

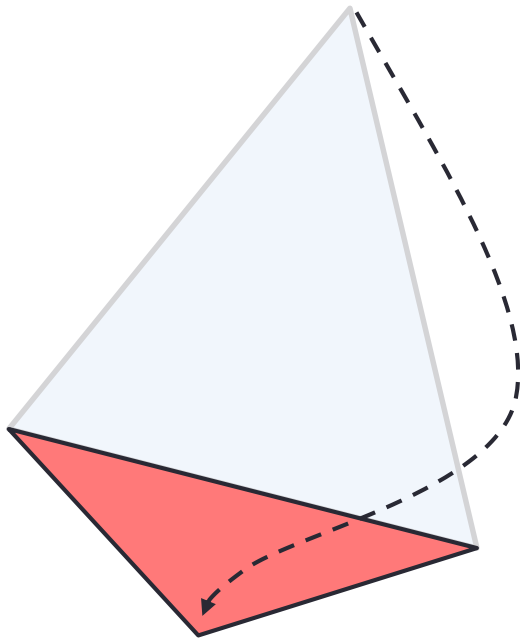


[Kim et al. 2010]

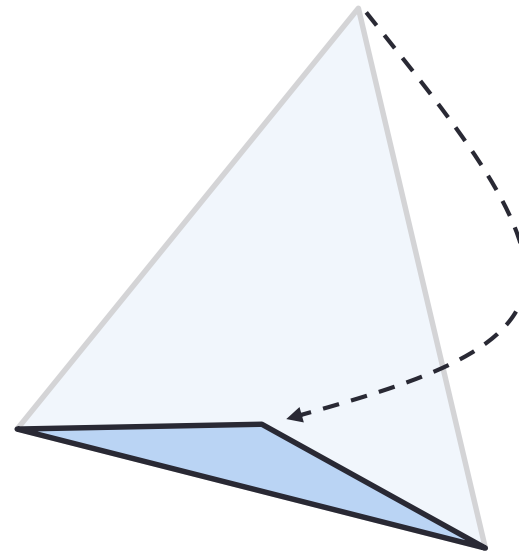


[Sifakis et al. 2009]

# Misbehaved Mappings



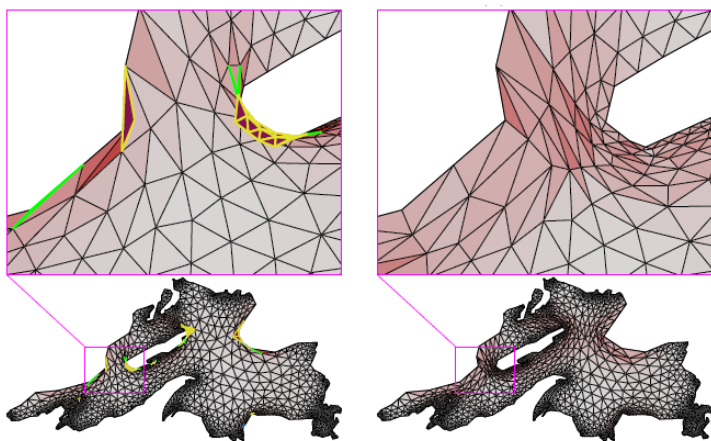
**Flip**



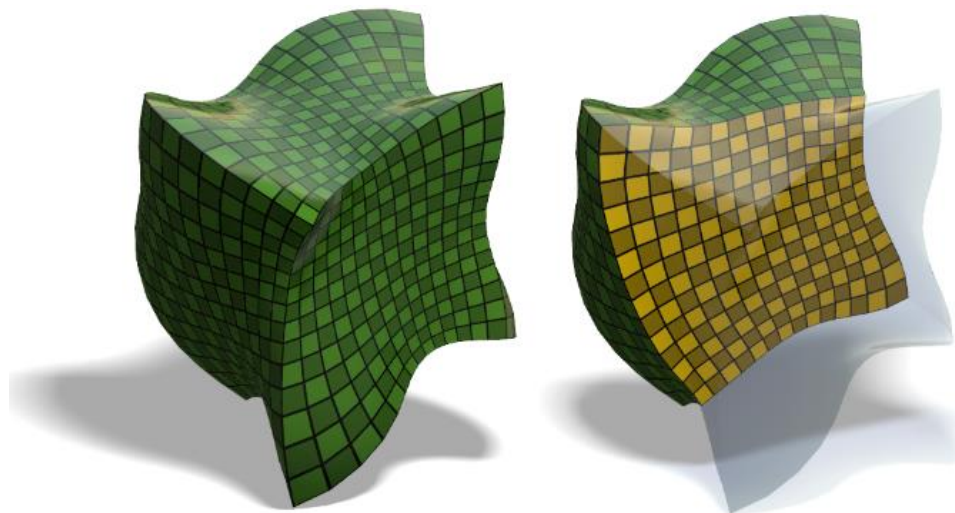
**High Distortion**



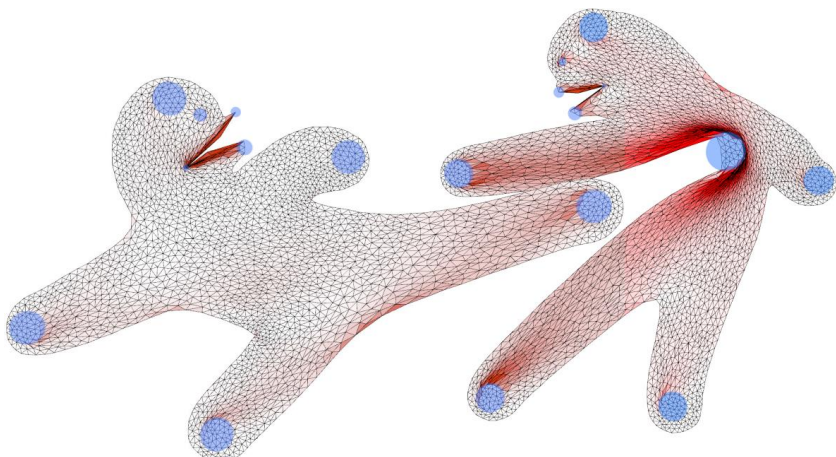
# Related Work



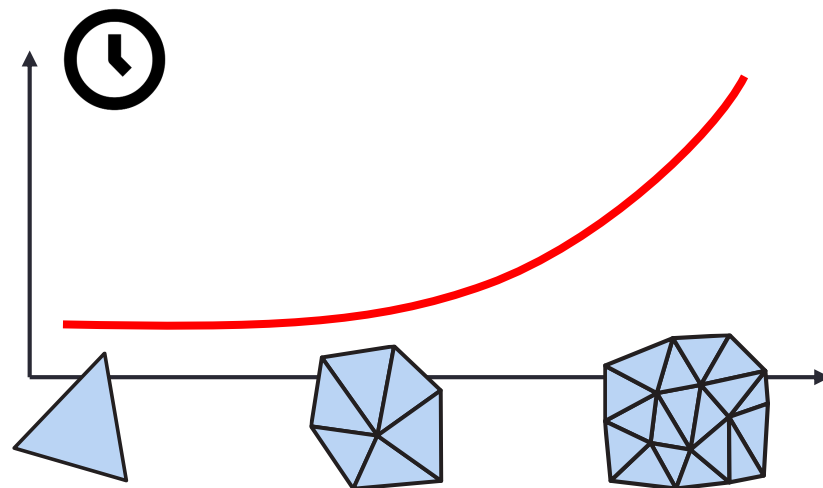
[Lipman 2012]



[Kovalsky et al. 2014]

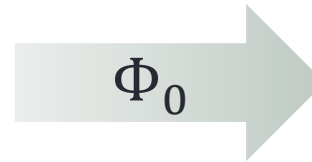
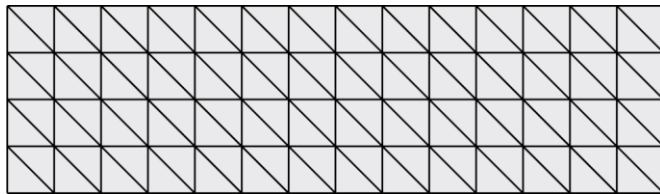


[Schüller et al. 2013]

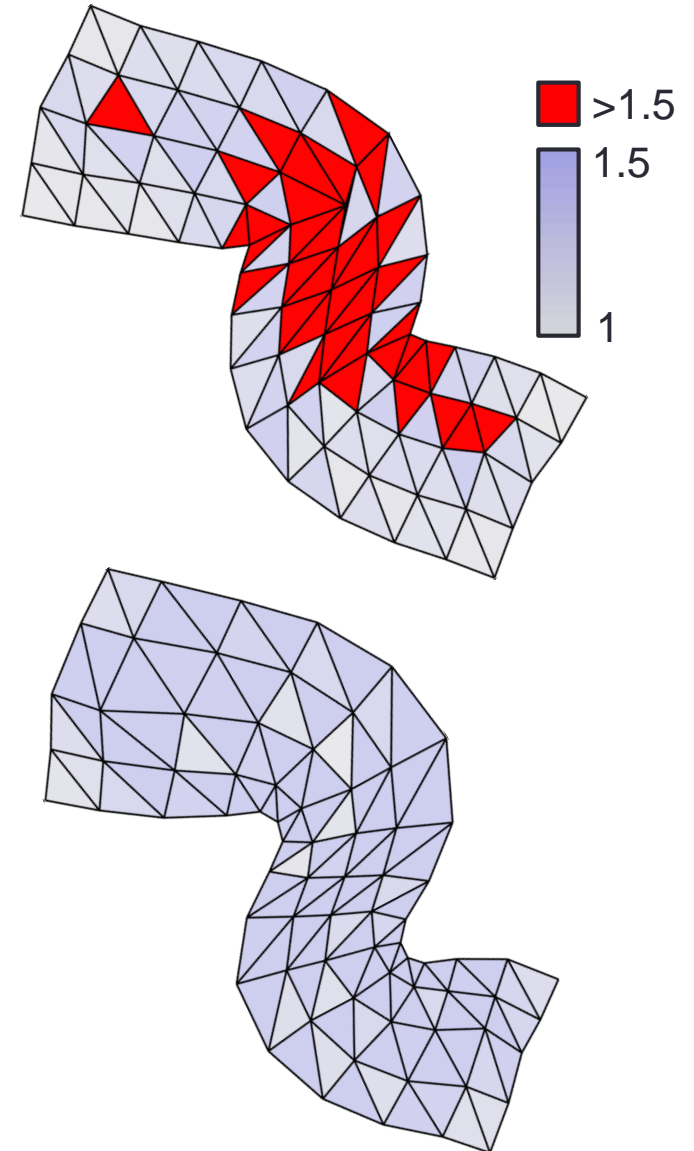


# Goal

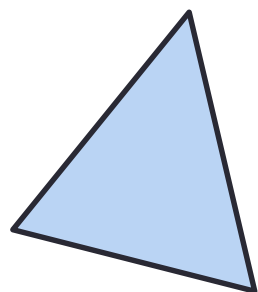
- Given:



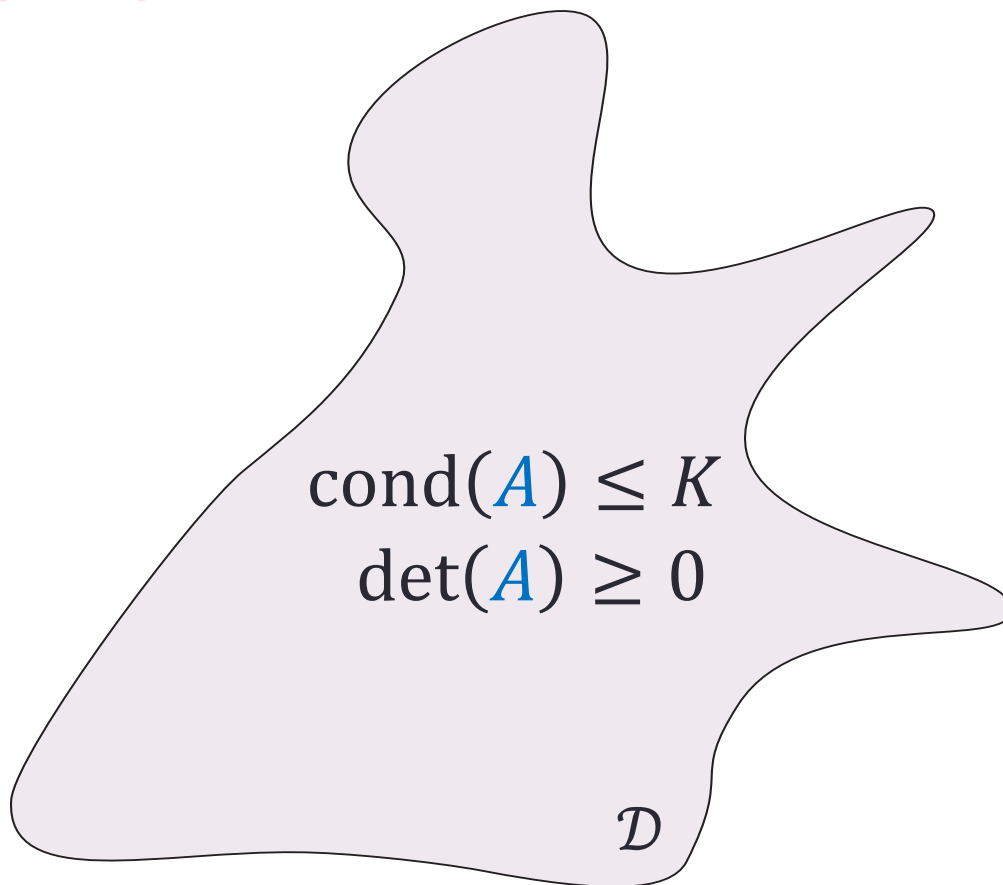
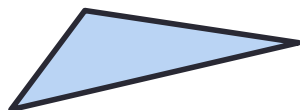
- Find  $\Phi \approx \Phi_0$ :
  - **No flips or high distortions**
  - **Fast and scalable**



# Bounded Distortion

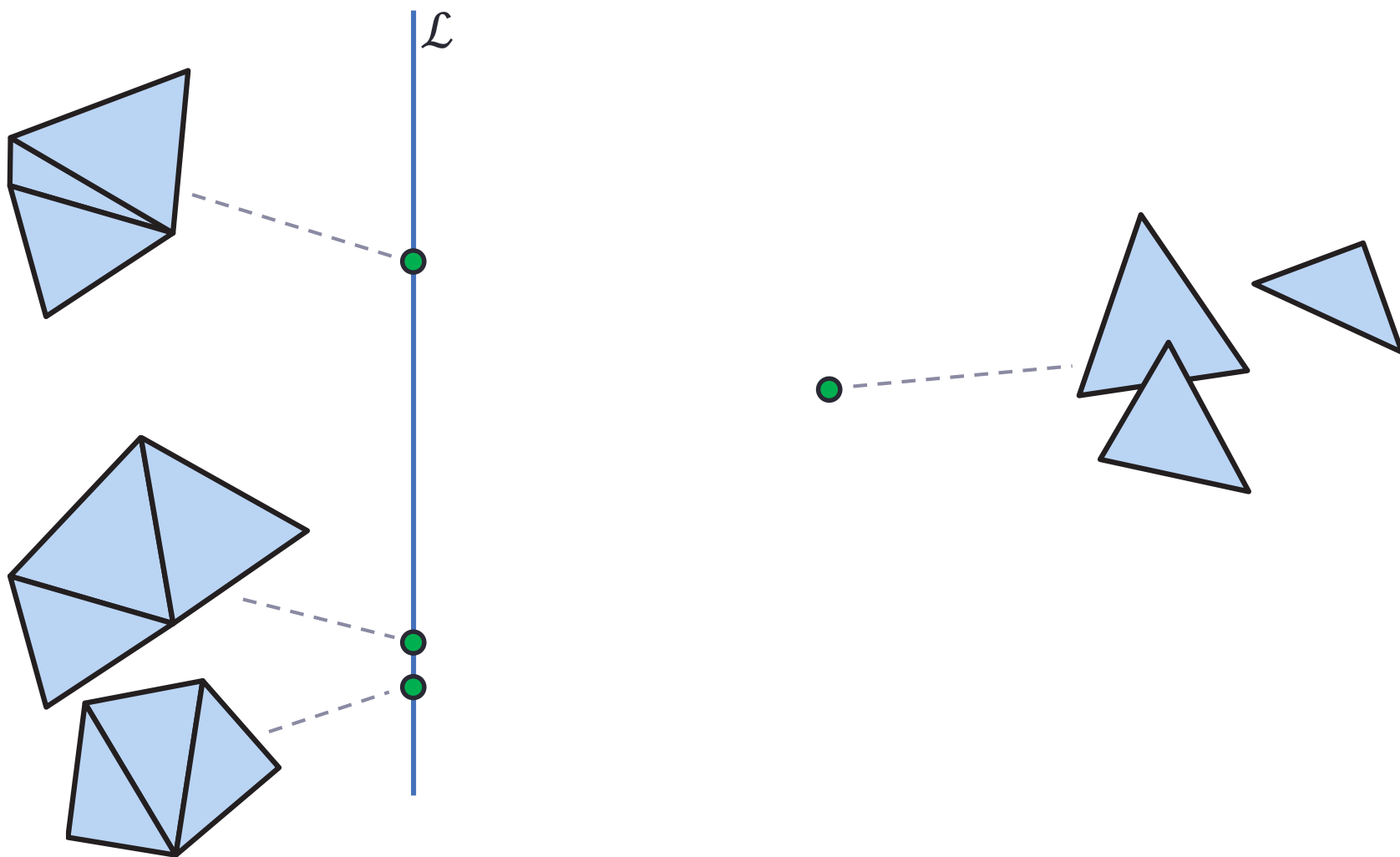


$$\Phi(x) = Ax + t$$



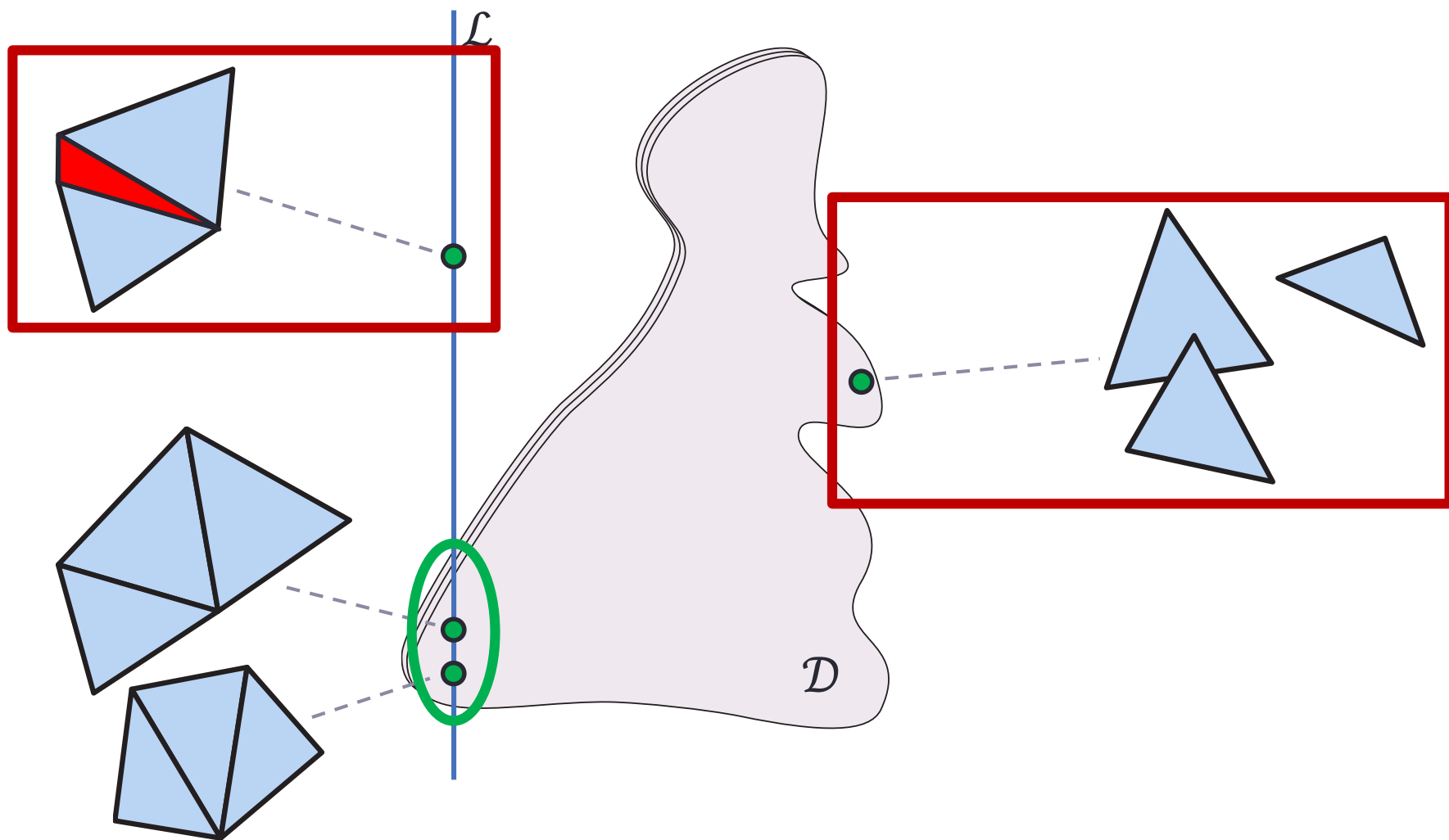
Scale-invariant characterization for  
well-behaved mappings

# Bounded Distortion Simplicial Mappings

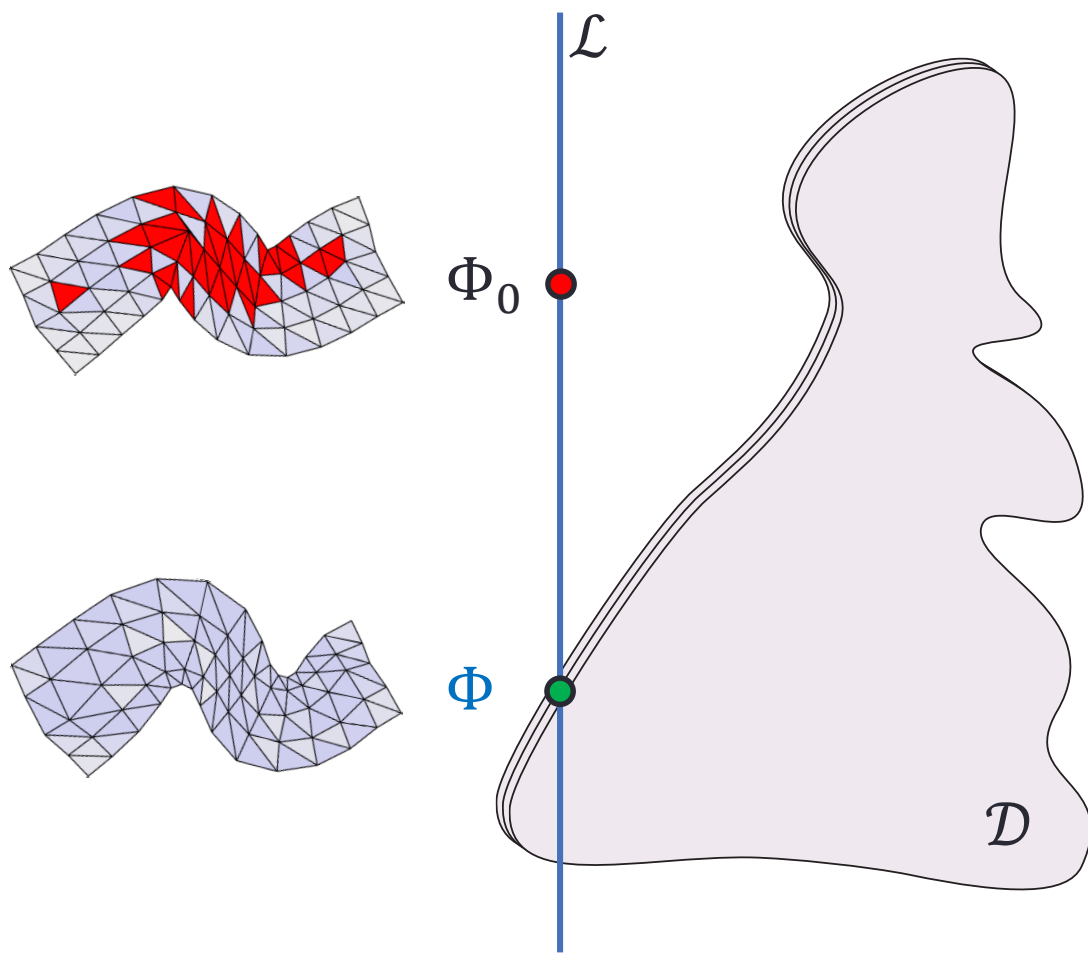




# Bounded Distortion Simplicial Mappings



# Goal

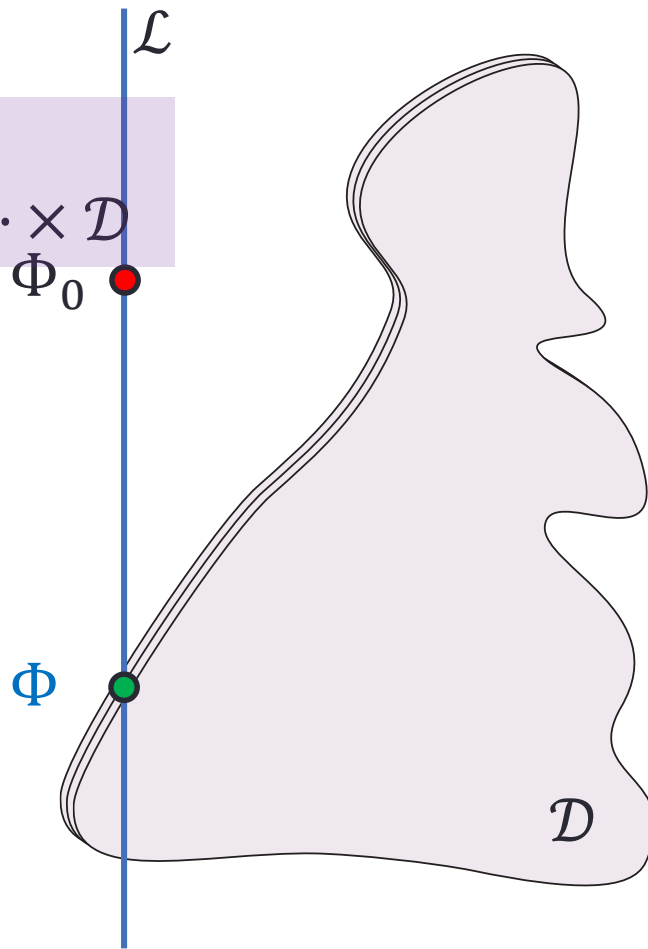


# Goal

$$\min_{\Phi} \|\Phi - \Phi_0\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$



# The Challenge

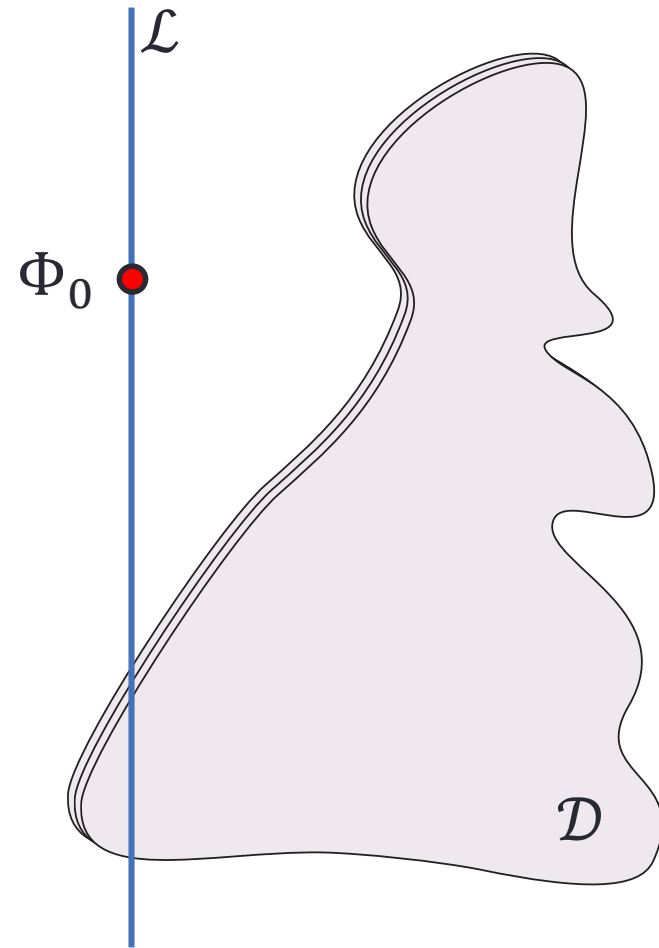
$$\min_{\Phi} \|\Phi - \Phi_0\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$

**efficient proxy** for

$\Phi \in \mathcal{D} \times \dots \times \mathcal{D} ??$

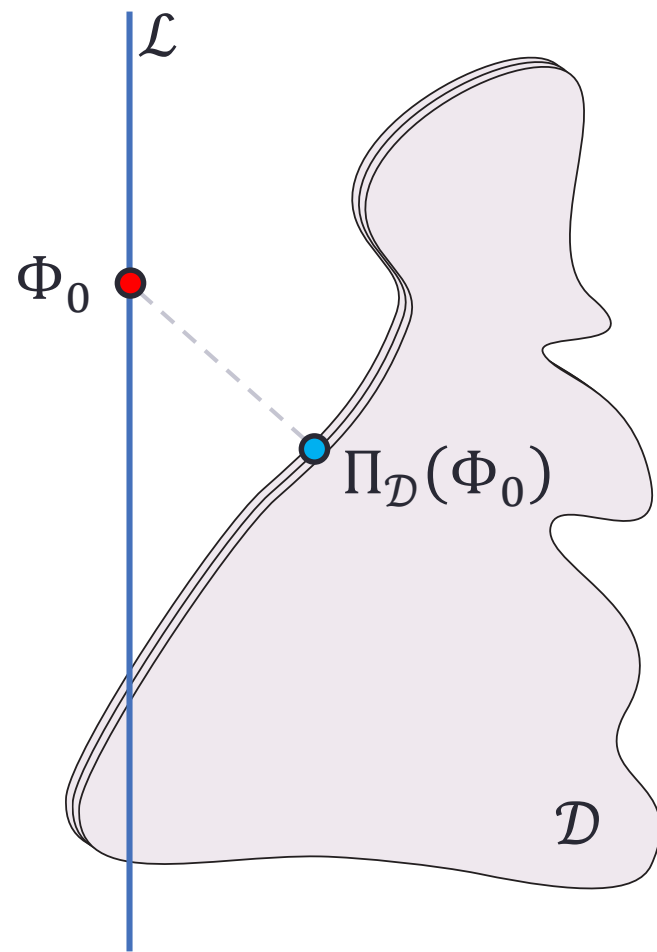


# BD Projection

$$\min_{\Phi} \|\Phi - \Phi_0\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$





# BD Projection

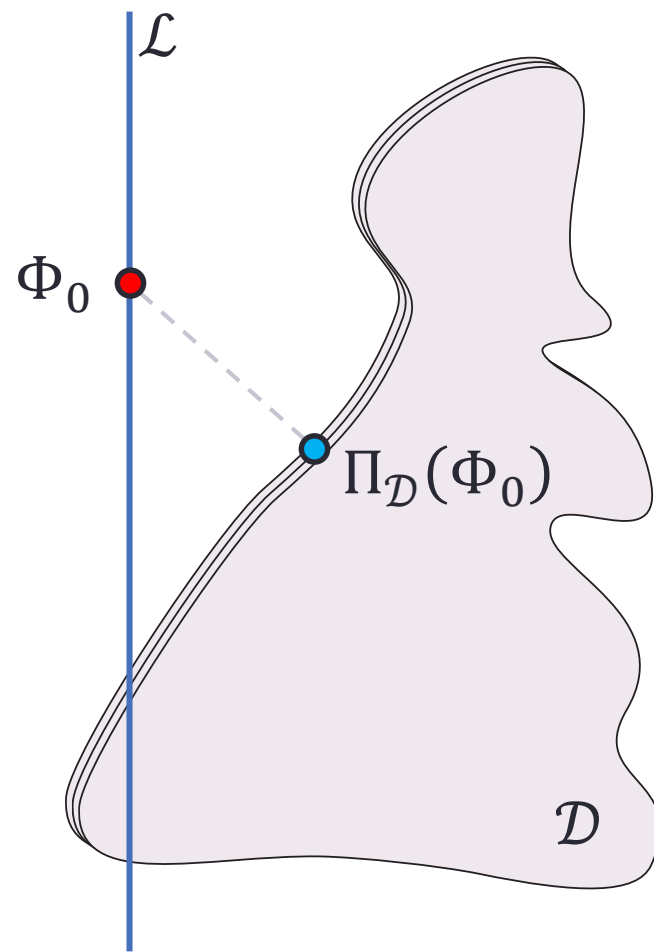
$$\min_{\Phi} \|\Phi - \Phi_0\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$

## Projection onto BD

- Separable  $\Rightarrow$  Parallelizable
- Low dimensional SVD  
+ simple arithmetic

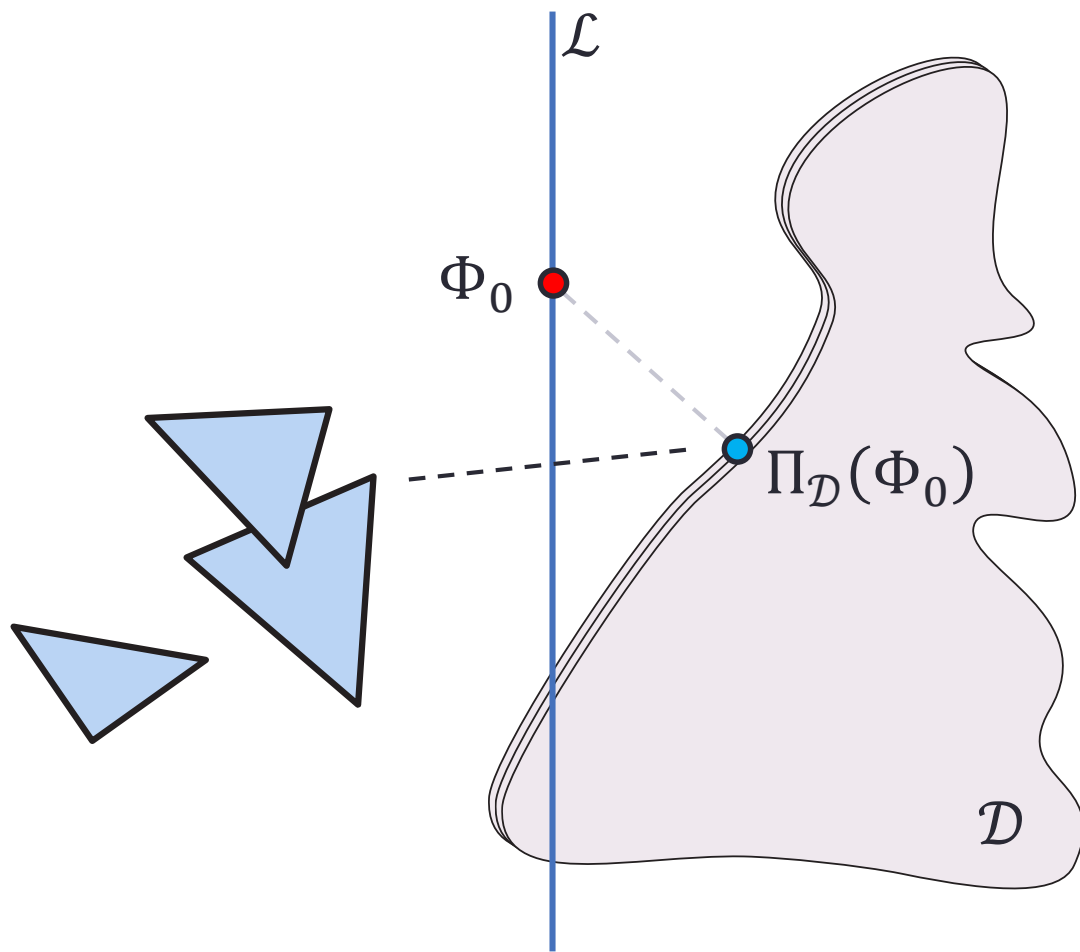


# BD Projection

$$\min_{\Phi} \|\Phi - \Phi_0\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$

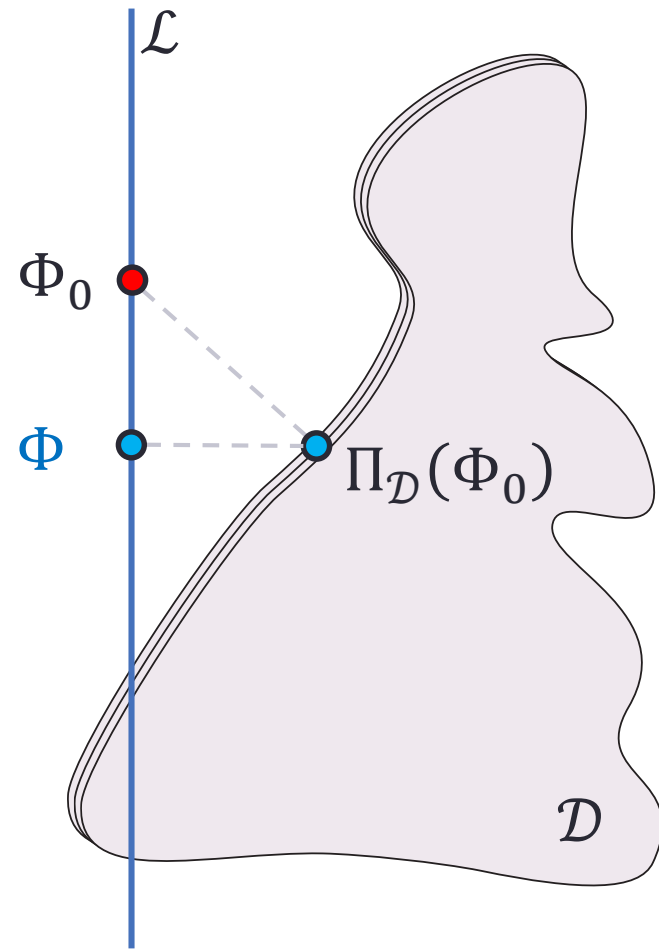


# BD Projection

$$\min_{\Phi} \|\Phi - \Phi_0\|$$

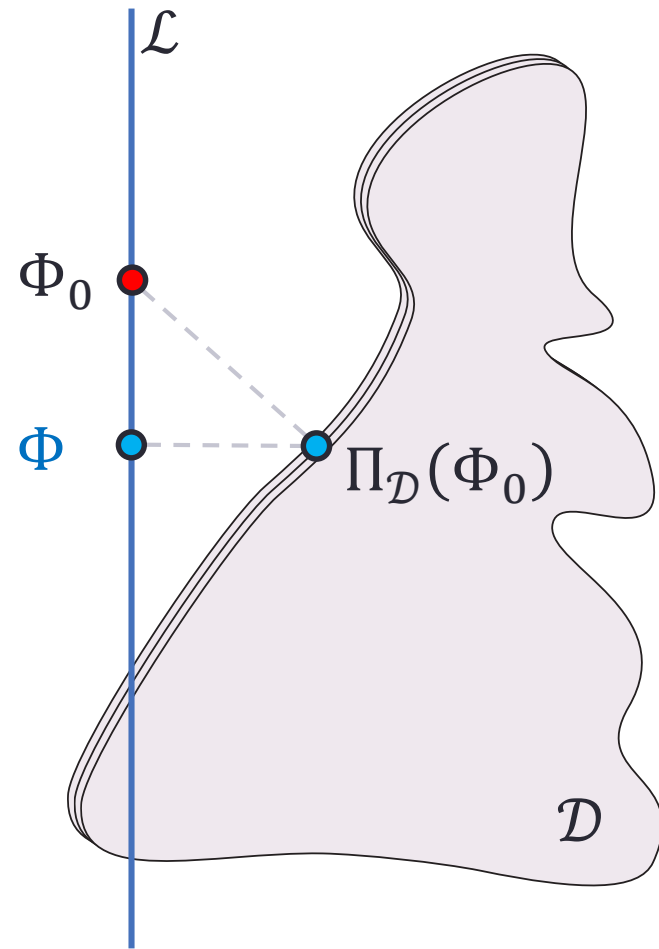
$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$



# BD Projection

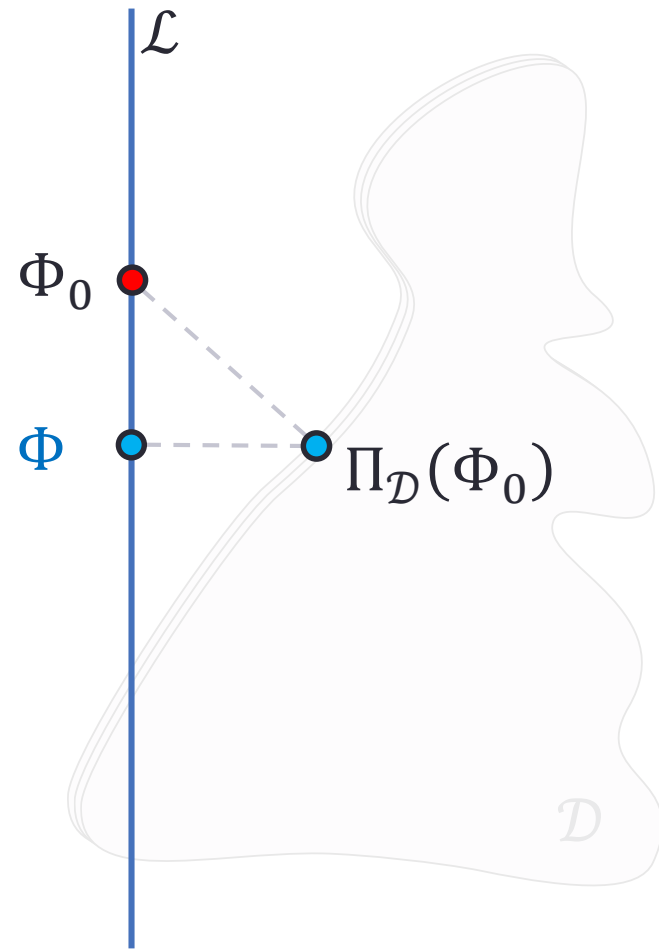
$$\begin{aligned} \min_{\Phi} \quad & \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\| \\ \text{s.t.} \quad & \Phi \in \mathcal{L} \end{aligned}$$



# Alternating Optimization

$$\begin{aligned} \min_{\Phi} \quad & \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\| \\ \text{s.t.} \quad & \Phi \in \mathcal{L} \end{aligned}$$

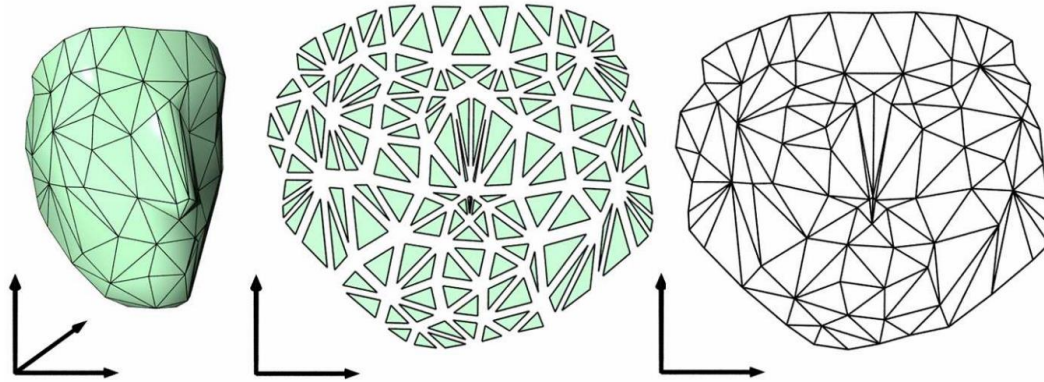
**Point proxy**  
for  
 $\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$



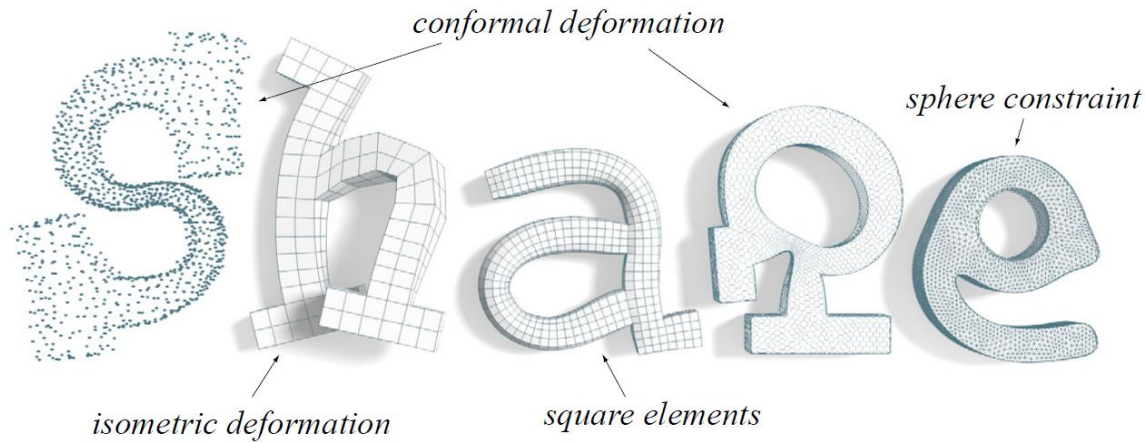


A

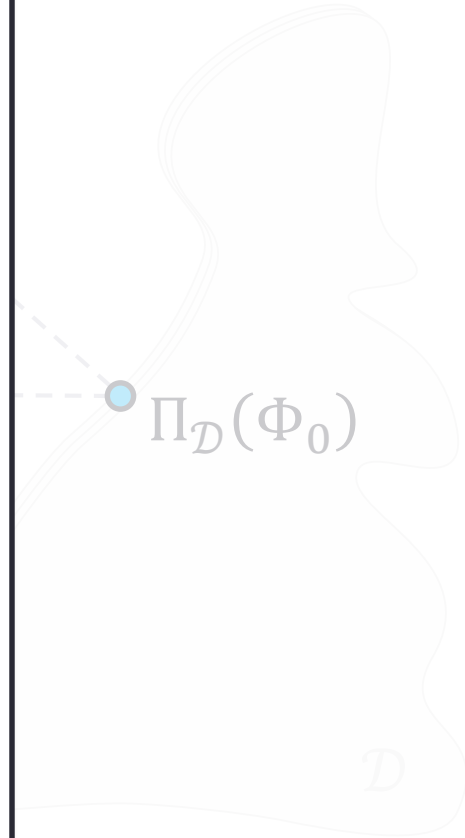
mi  
 $\Phi$   
s.t.



[Liu et al. 2008]



[Bouaziz et al. 2012]



# Alternating Optimization

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

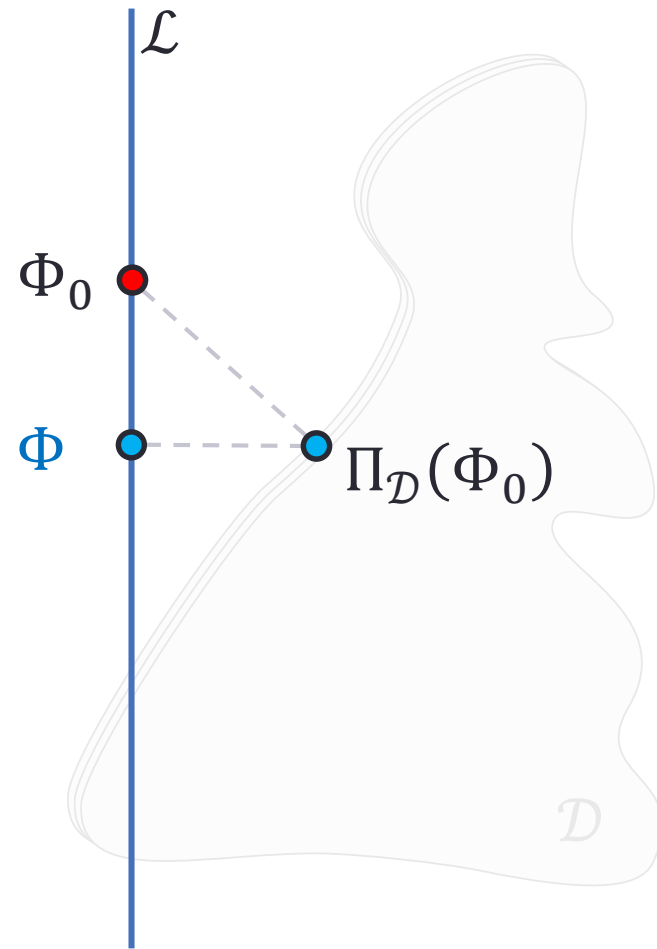
$$\text{s.t. } \Phi \in \mathcal{L}$$

Solve a linear system:

$$Ax = b$$

**Fixed**

- Can factorize  $A$   
⇒ Super efficient iterations
- Very slow convergence...

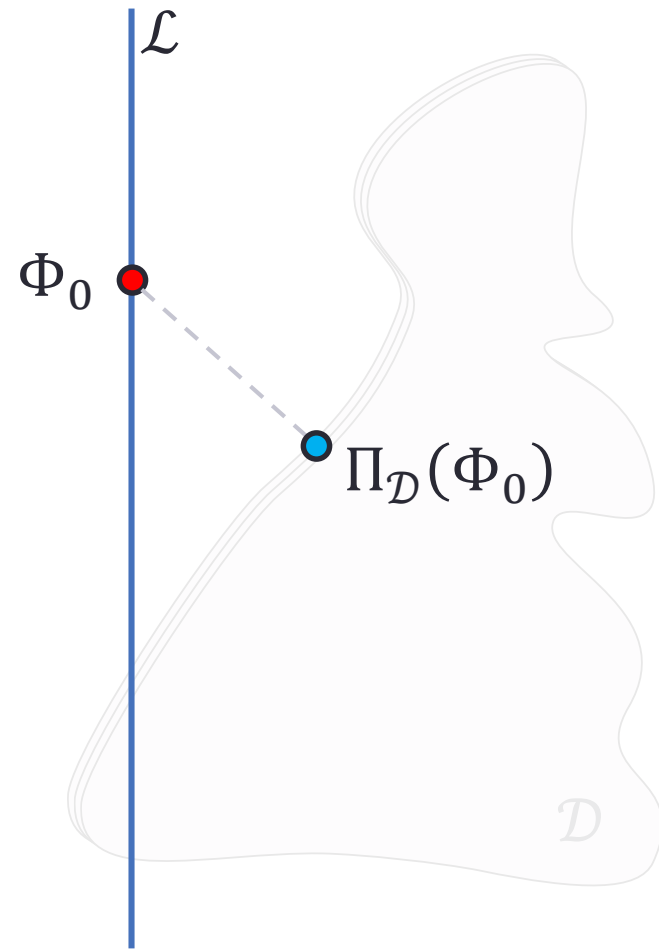


# Alternating Optimization

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$

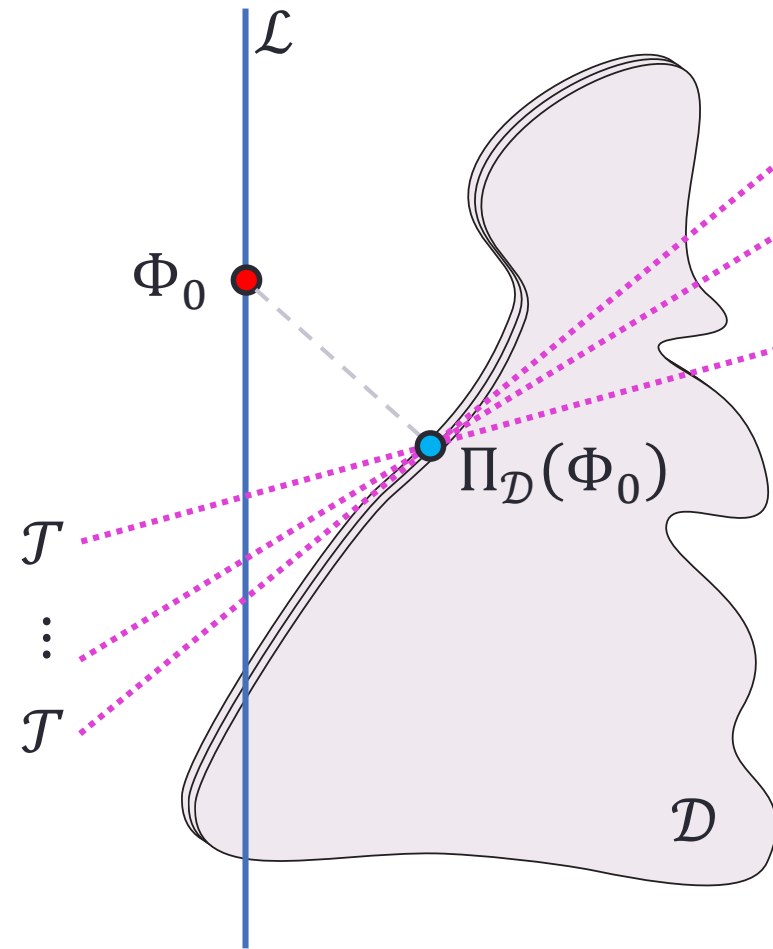


# 0<sup>th</sup> order vs. 1<sup>st</sup> order

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$



# Use Tangents

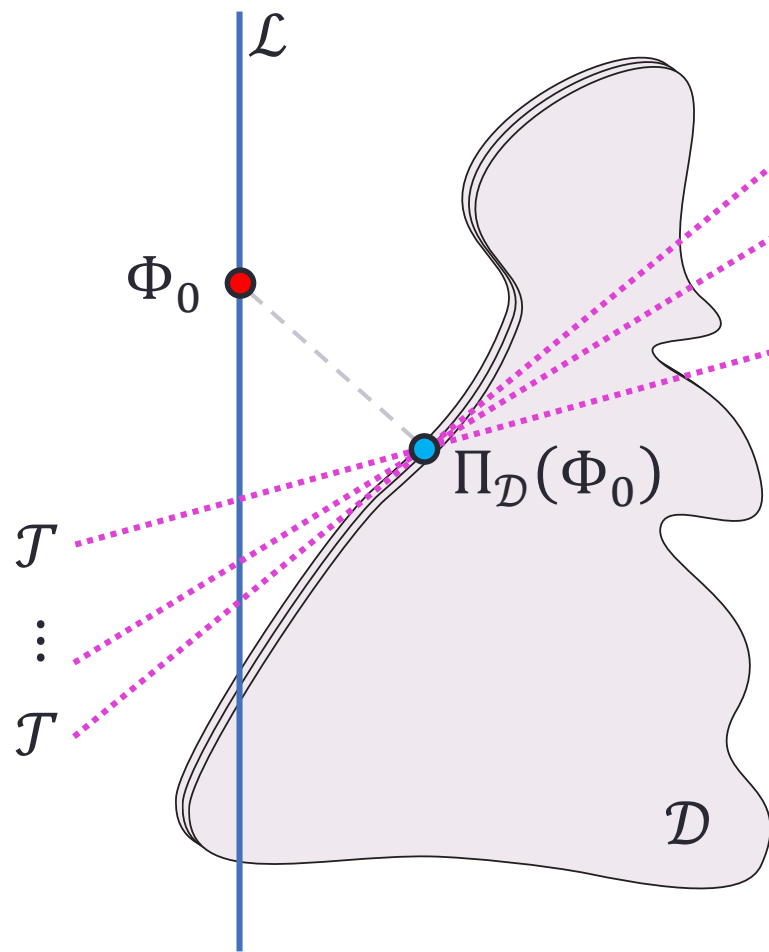
$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{T} \times \dots \times \mathcal{T}$$

1<sup>st</sup> order proxy  
for

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$





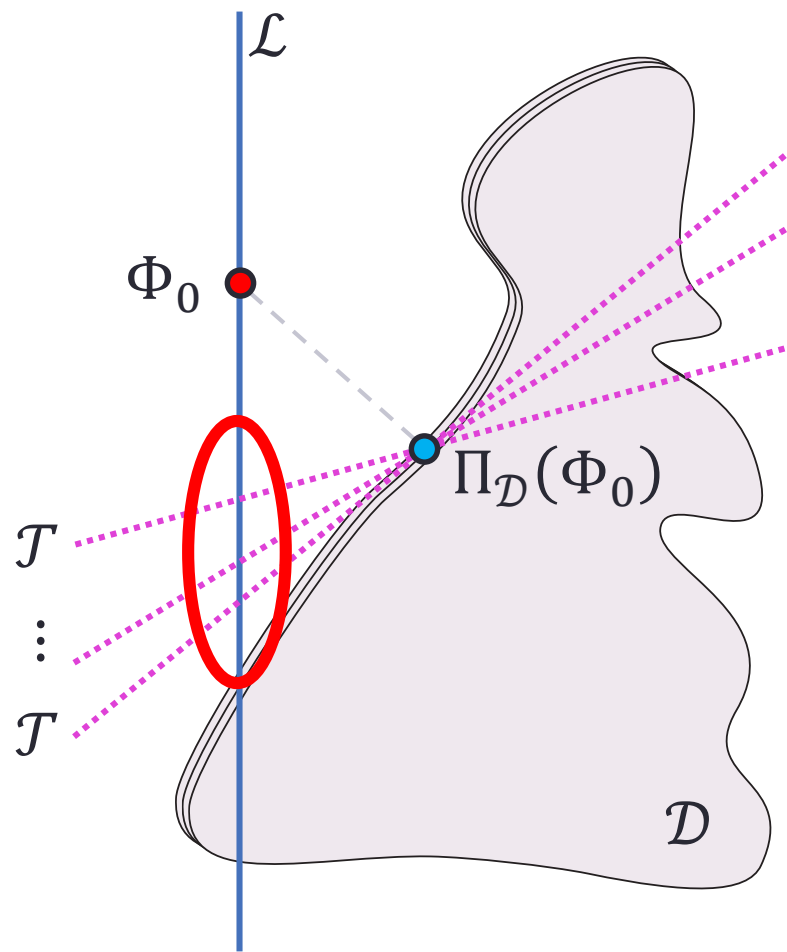
# Use Tangents

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

s.t.  $\Phi \in \mathcal{L}$

$\Phi \in \mathcal{T} \times \dots \times \mathcal{T}$

**Infeasible!**

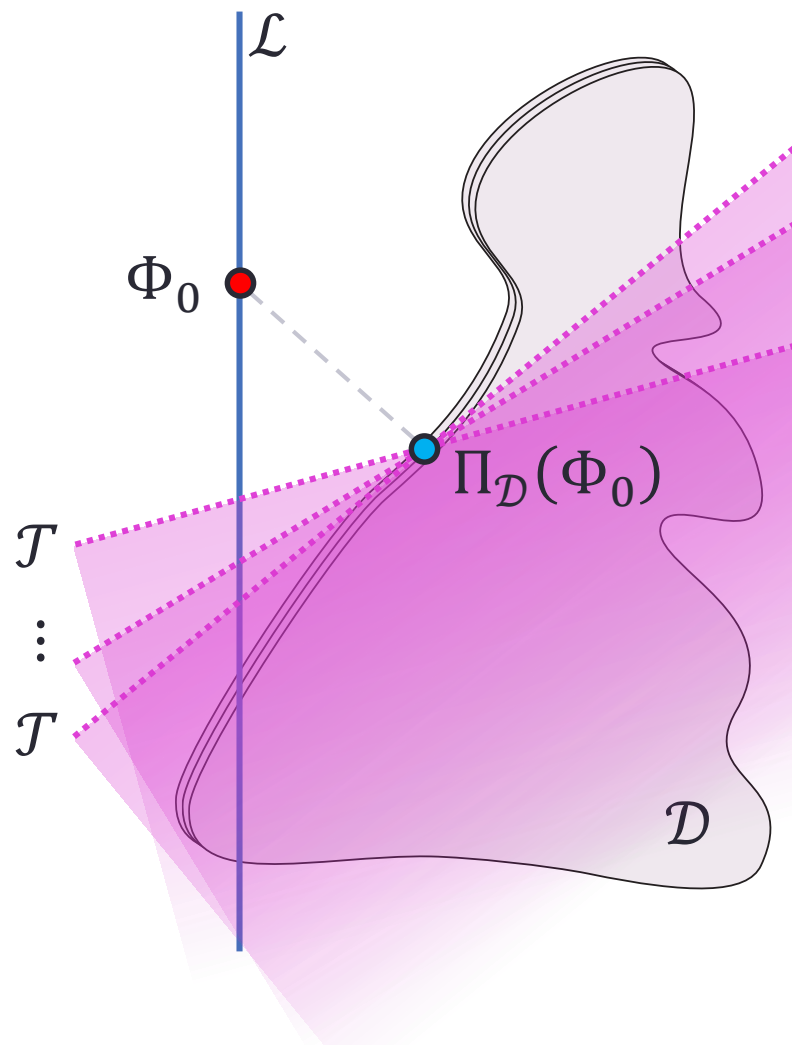


# Use Tangents

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{T} \times \dots \times \mathcal{T}$$

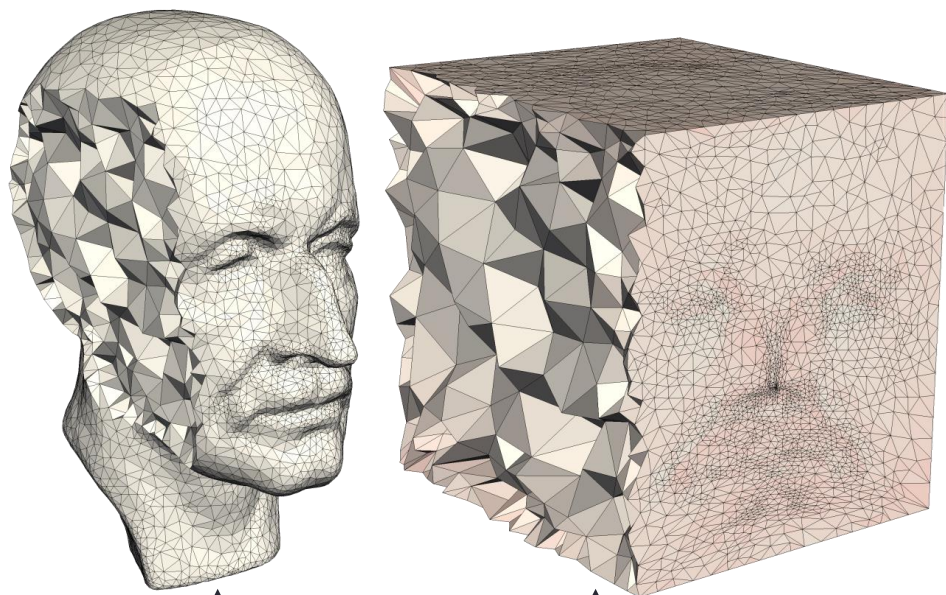


# Use Tangents

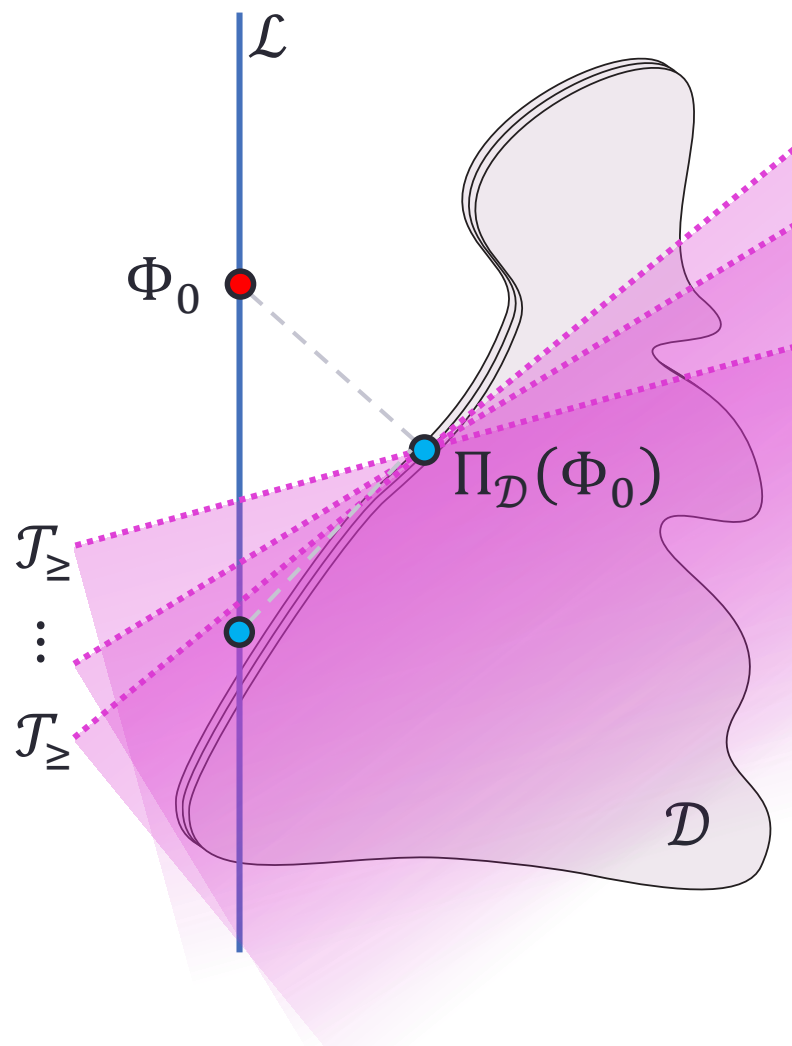
$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{T}_{\geq} \times \dots \times \mathcal{T}_{\geq}$$



[Aigerman and Lipman 2013]



# Aigerman and Lipman 2013

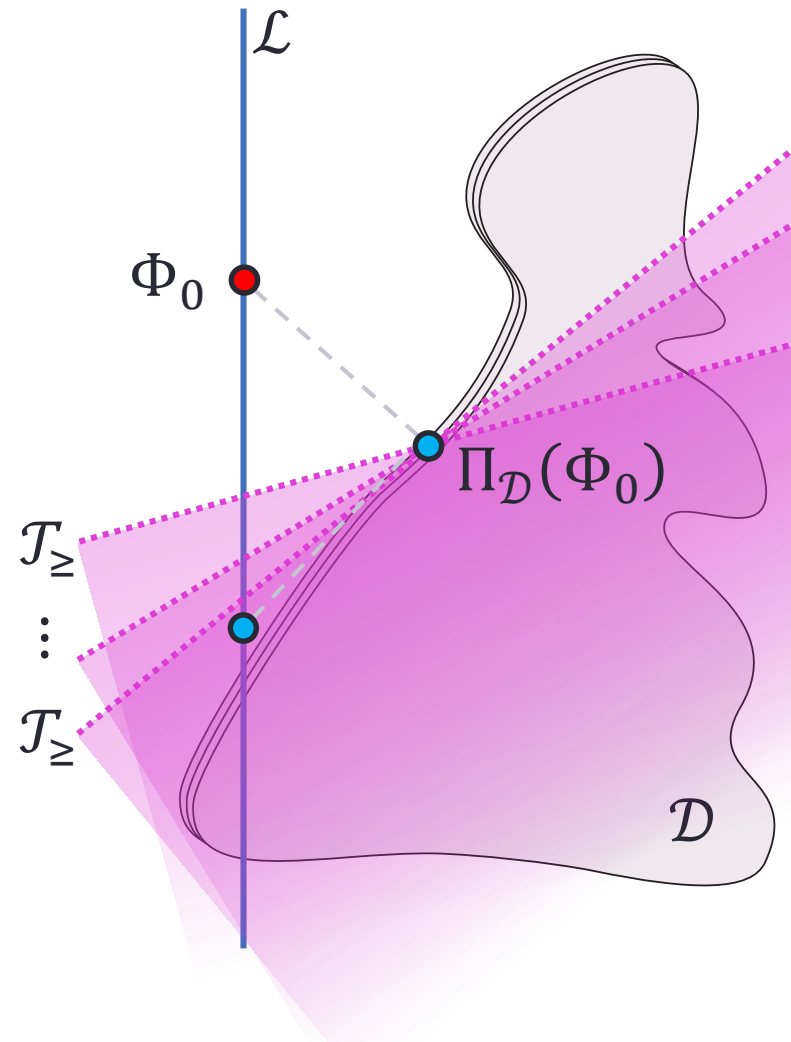
$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{T}_{\geq} \times \cdots \times \mathcal{T}_{\geq}$$

Solve a Quadratic Program

- Quadratic convergence
- Uses interior point solver
- Poor scalability



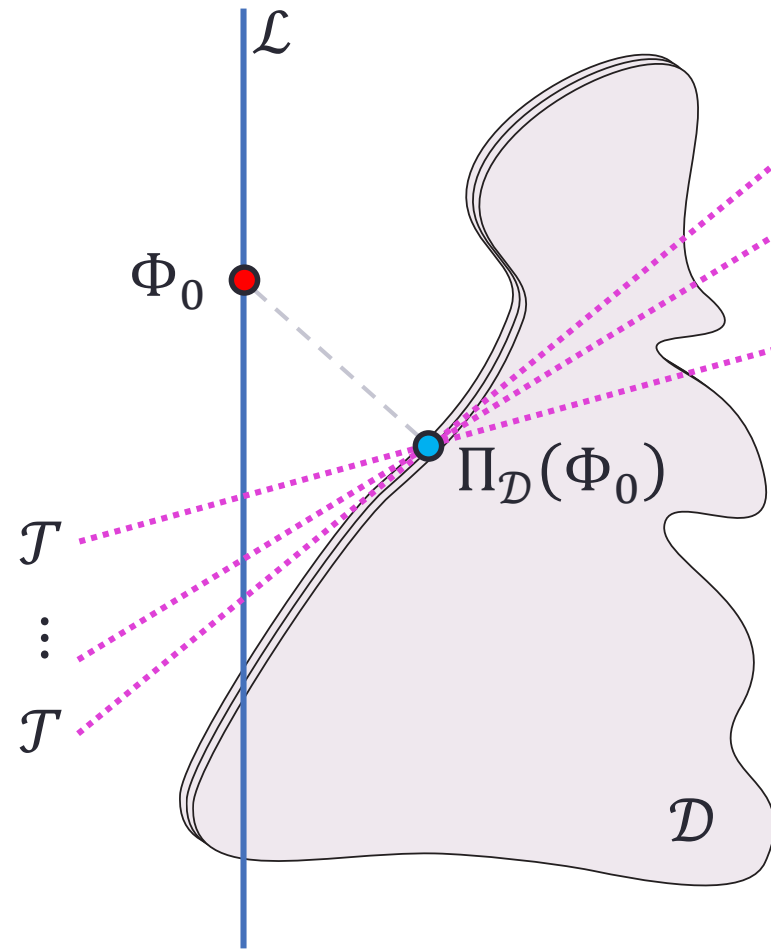
# Use Tangents

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{T} \times \dots \times \mathcal{T}$$

**Infeasible!**

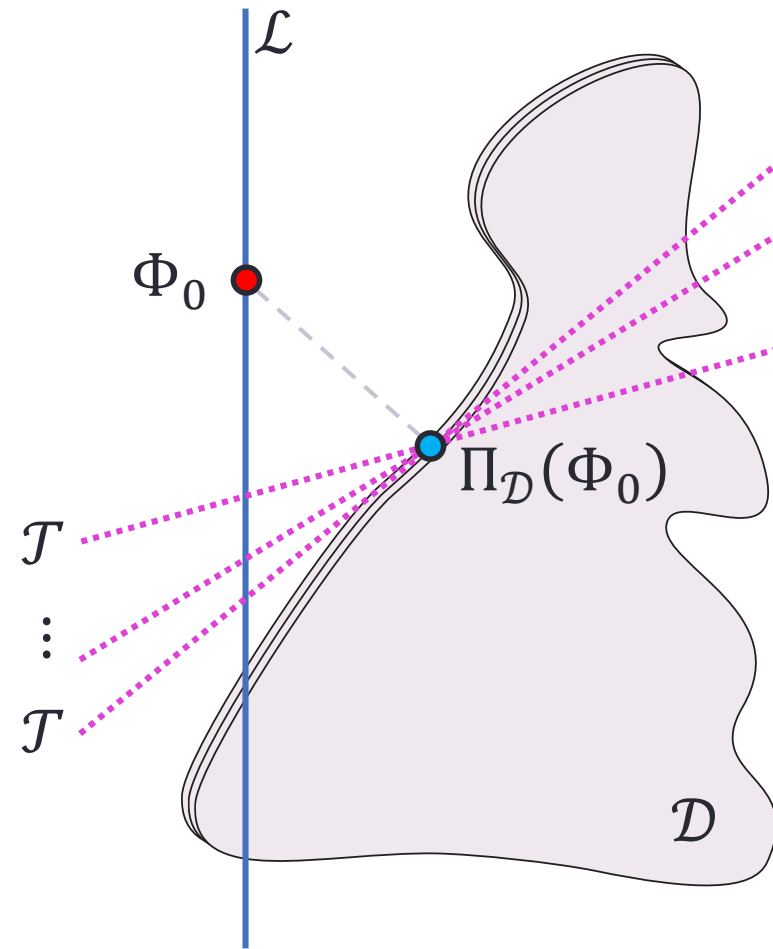


# Non-linear Least Squares

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

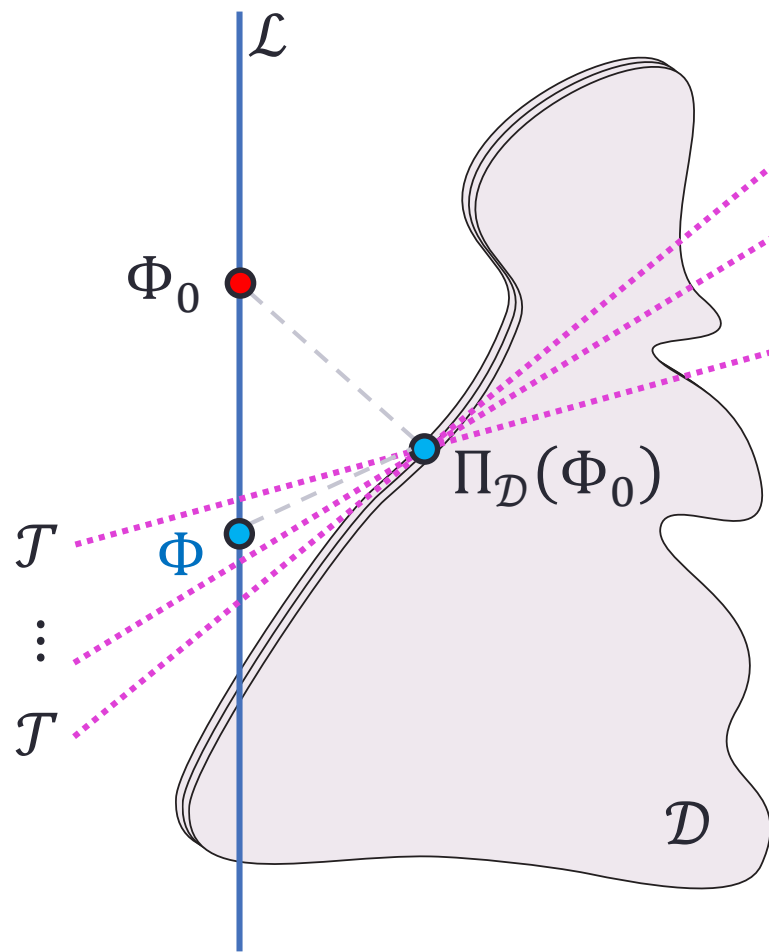
$$\Phi \in \mathcal{T} \times \dots \times \mathcal{T}$$



# Non-linear Least Squares

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|^2 + \sum \lambda_j \langle \Phi_j, \mathcal{T}_j^\perp \rangle^2$$

$$\text{s.t. } \Phi \in \mathcal{L}$$



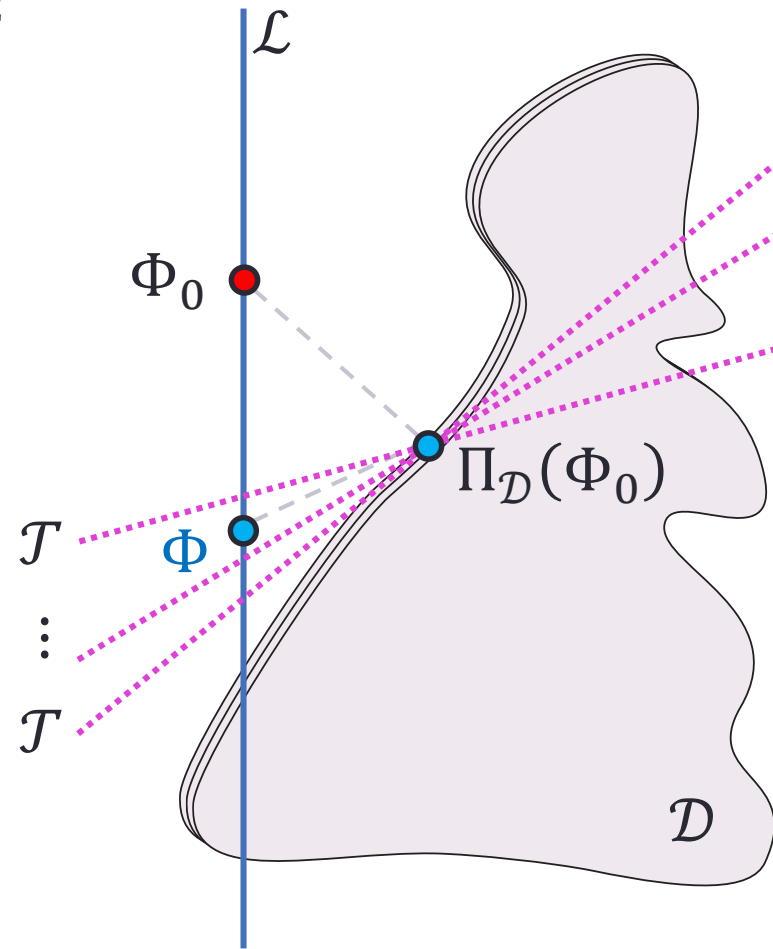
# Non-linear Least Squares

$$\begin{aligned} \min_{\Phi} \quad & \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|^2 + \sum \lambda_j \langle \Phi_j, \mathcal{T}_j^\perp \rangle^2 \\ \text{s.t.} \quad & \Phi \in \mathcal{L} \end{aligned}$$

Solve a linear system

(Gauss-Newton \ Levenberg-Marquardt)

- 2<sup>nd</sup>-order-like convergence
- Sensitive parameters
- Varying linear system



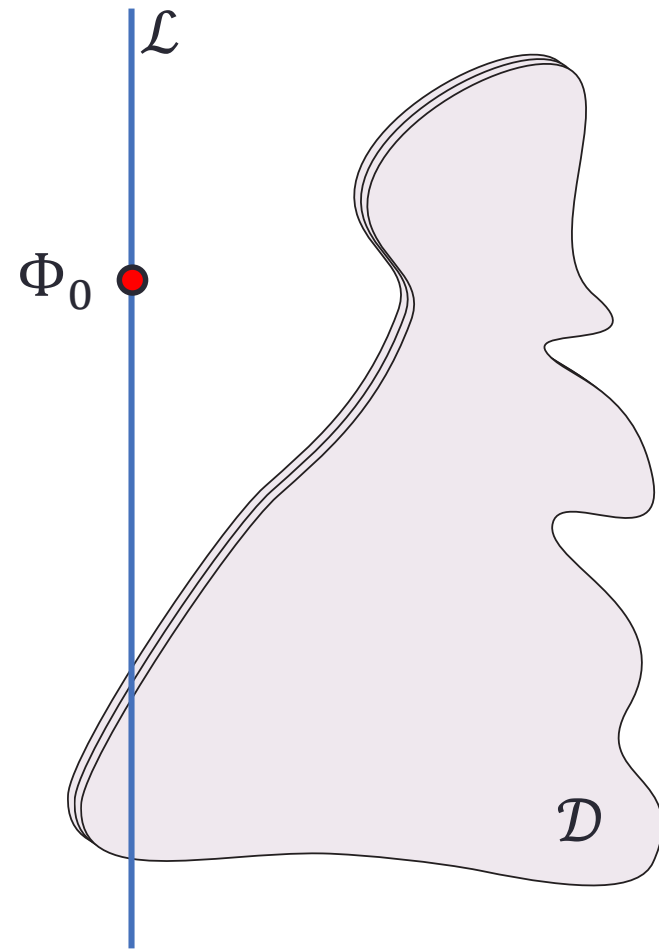
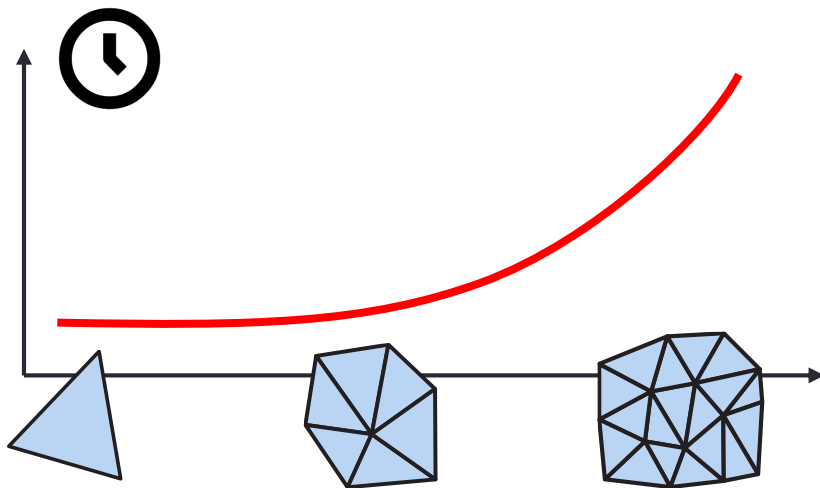


# Our Approach

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$



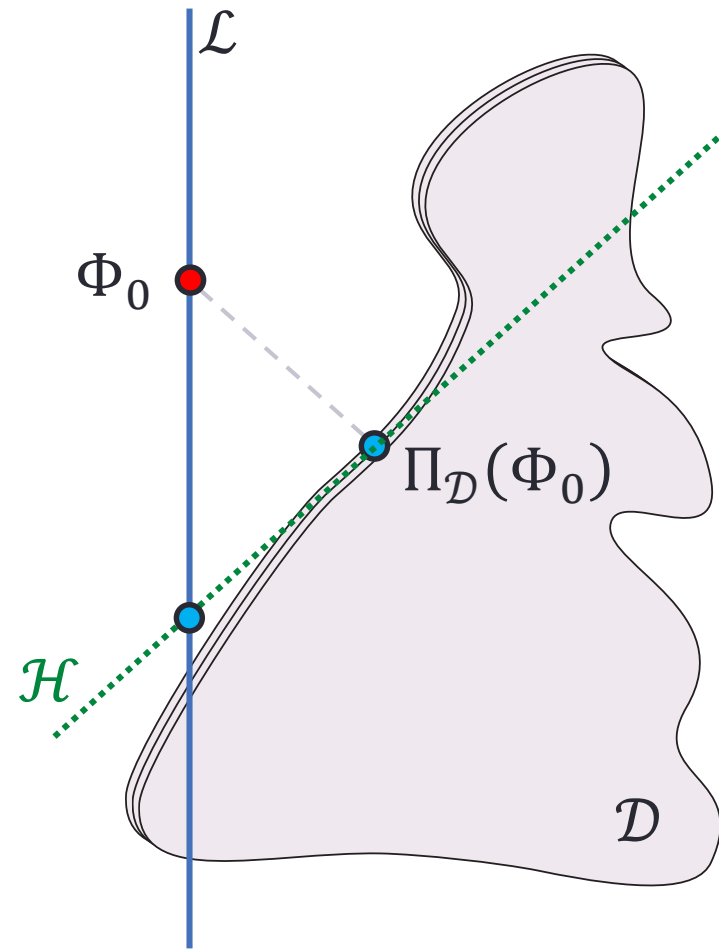
# Our Approach

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \underbrace{\mathcal{D} \times \dots \times \mathcal{D}}_{\bar{\mathcal{D}}}$$

Think of a **single**  
high-dimensional set

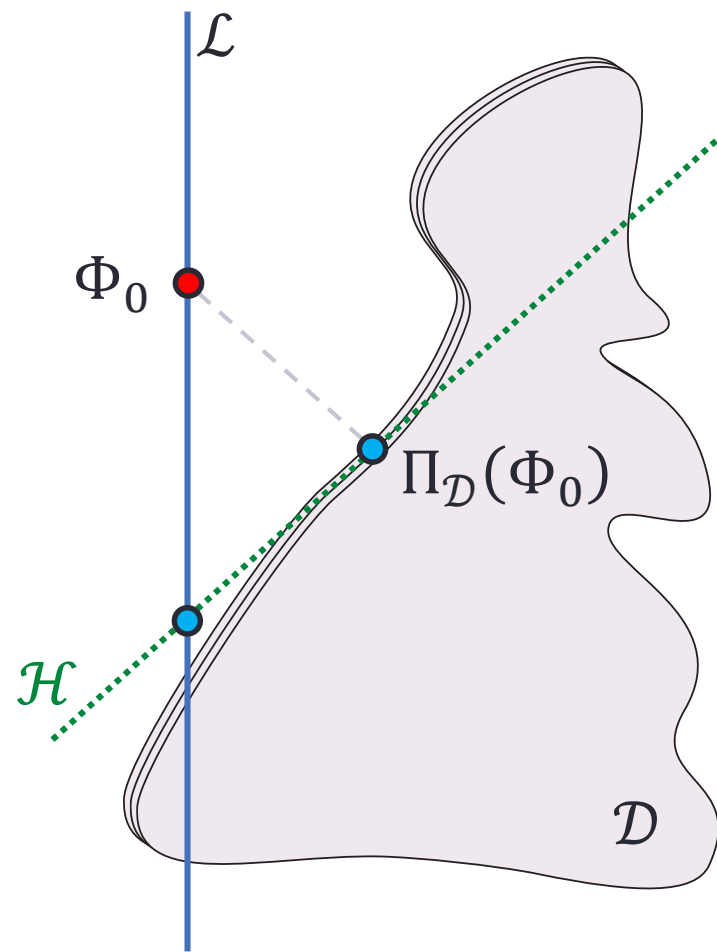


# Our Approach

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{D} \times \dots \times \mathcal{D}$$



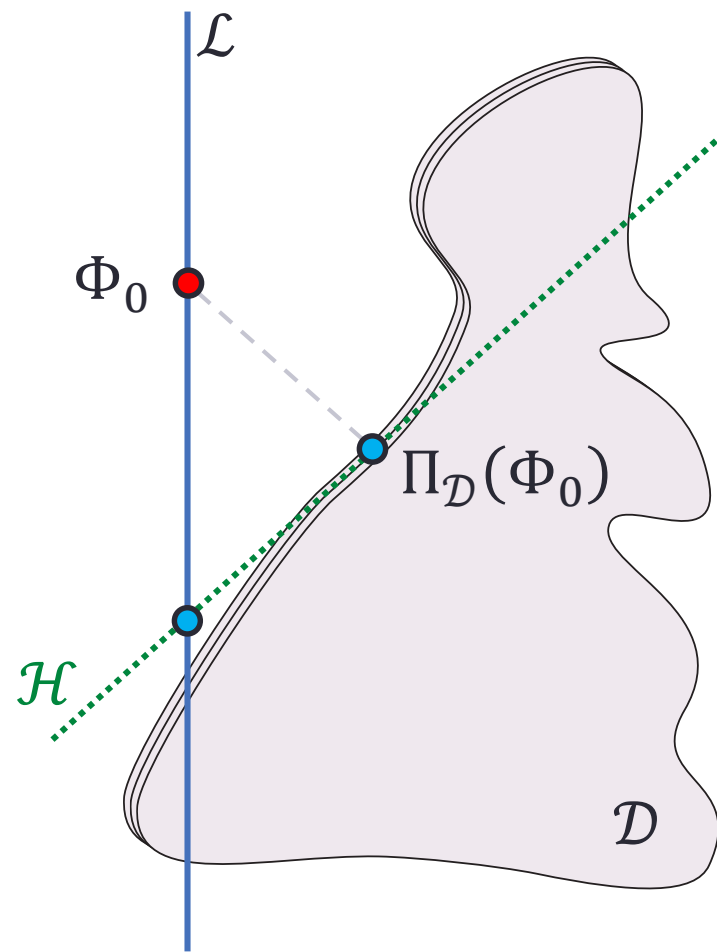
# Our Approach

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{H}$$

1<sup>st</sup>-order proxy



# Our Approach

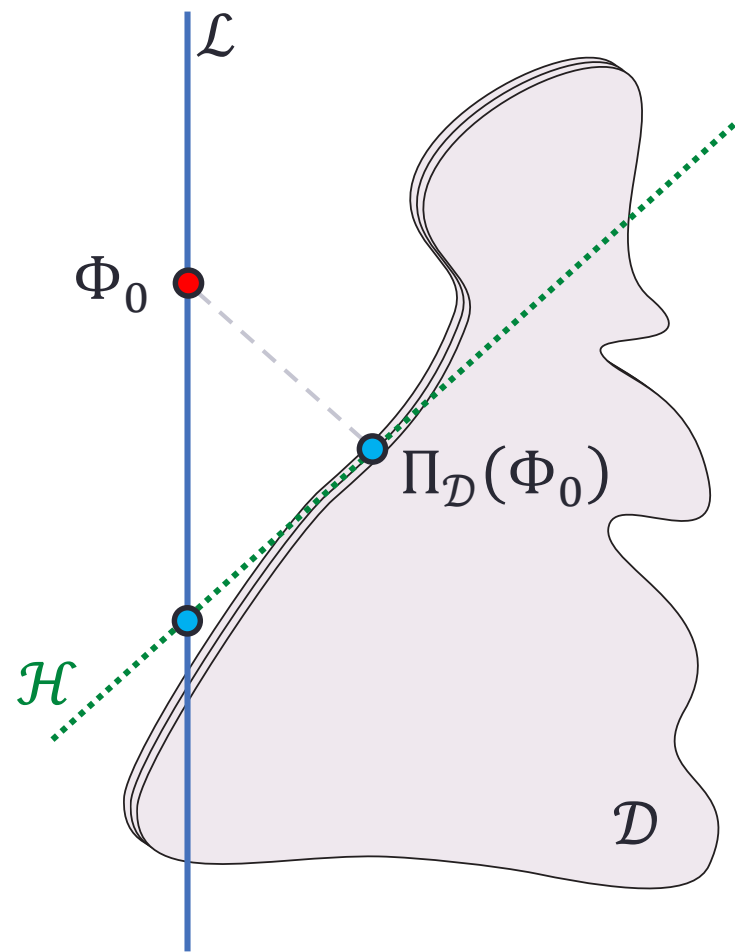
$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{H}$$

Solve a linear system

- **2<sup>nd</sup>-order-like** convergence
- Parameterless
- Super efficient iterations



# Algorithm

1. Compute  $\Pi_{\mathcal{D}}(\Phi_0)$

2. Form the hyperplane:

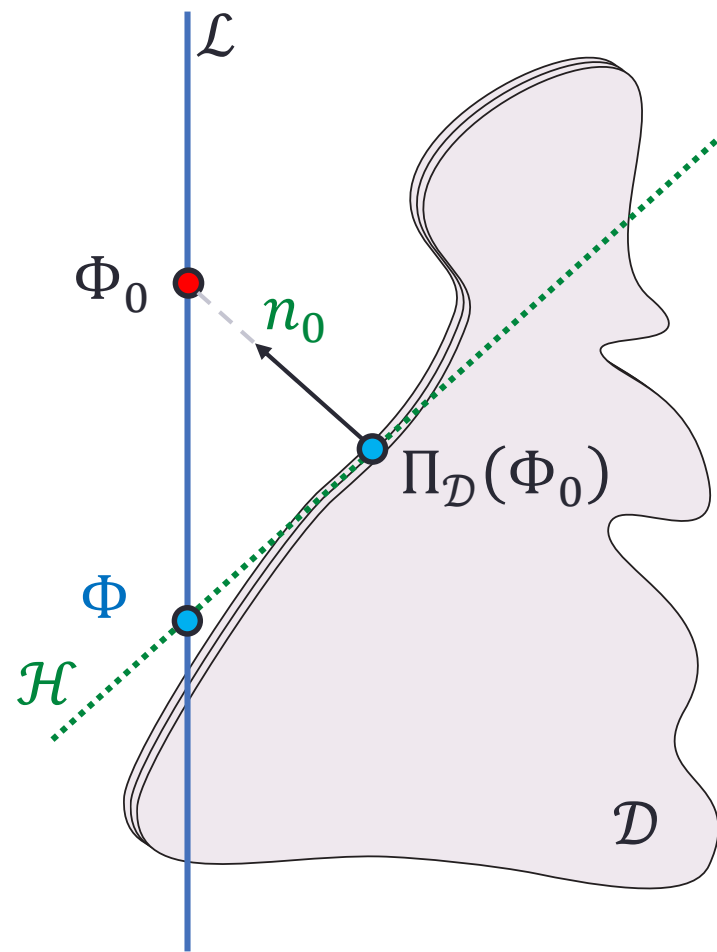
$$\mathcal{H} = \{x : n_0 x = n_0 \Pi_{\mathcal{D}}(\Phi_0)\}$$

3. Solve:

$$\min_{\Phi} \|\Phi - \Pi_{\mathcal{D}}(\Phi_0)\|$$

$$\text{s.t. } \Phi \in \mathcal{L}$$

$$\Phi \in \mathcal{H}$$



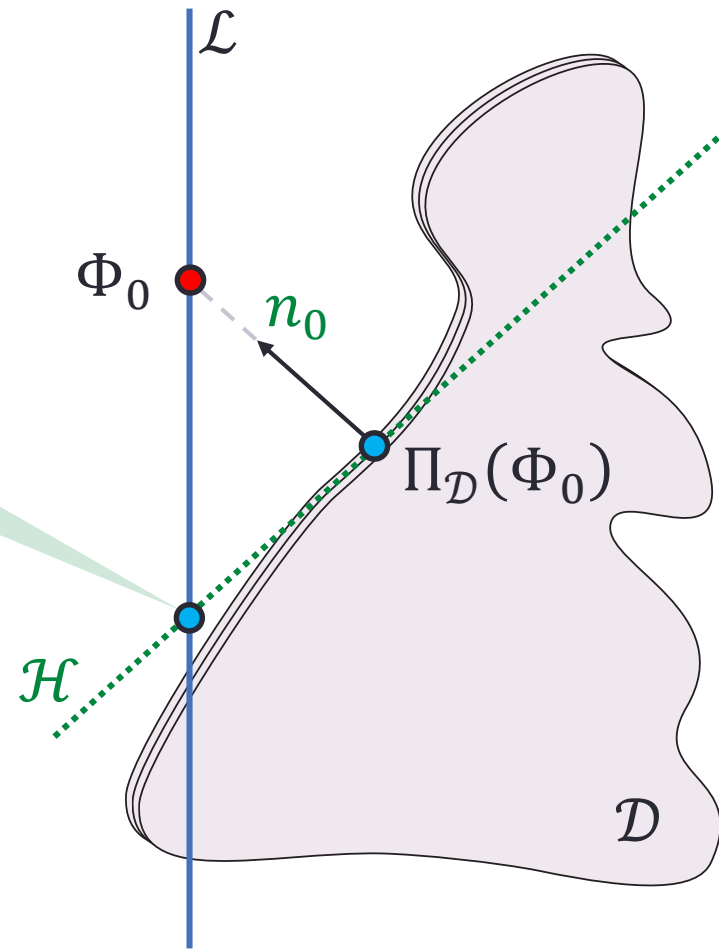
# Algorithm

- Equivalent KKT system:

$$\begin{bmatrix} \mathbf{M} & n_0 \\ n_0^T & 0 \end{bmatrix} x = b_0$$

- Solvable?

Exists?



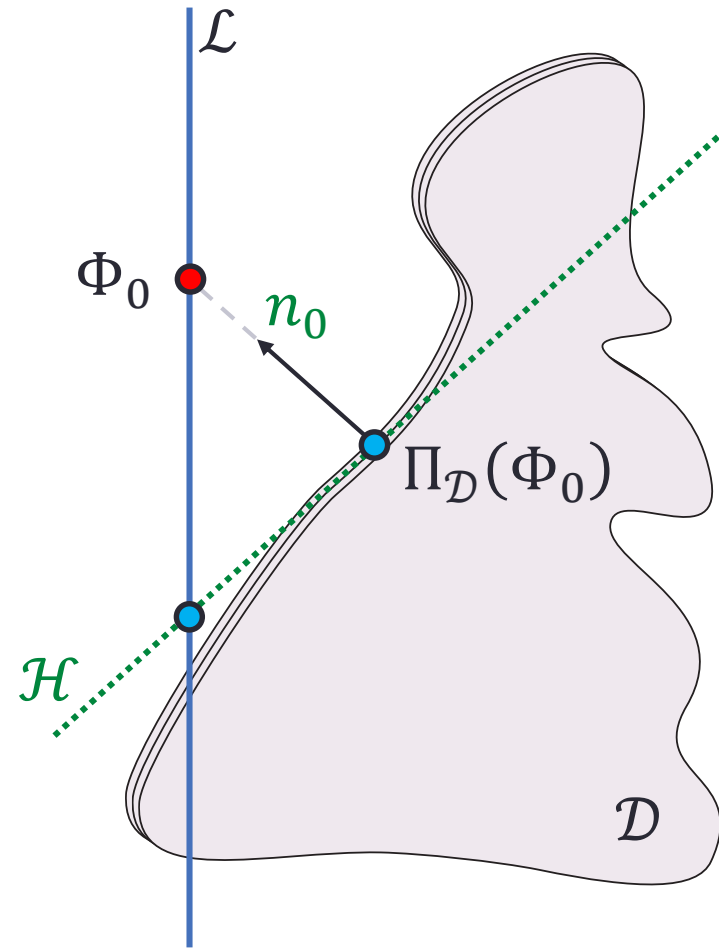
# Algorithm

- Equivalent KKT system:

$$\begin{bmatrix} M & n_0 \\ n_0^T & 0 \end{bmatrix} x = b_0$$

- Theorem:

Unique solution exists





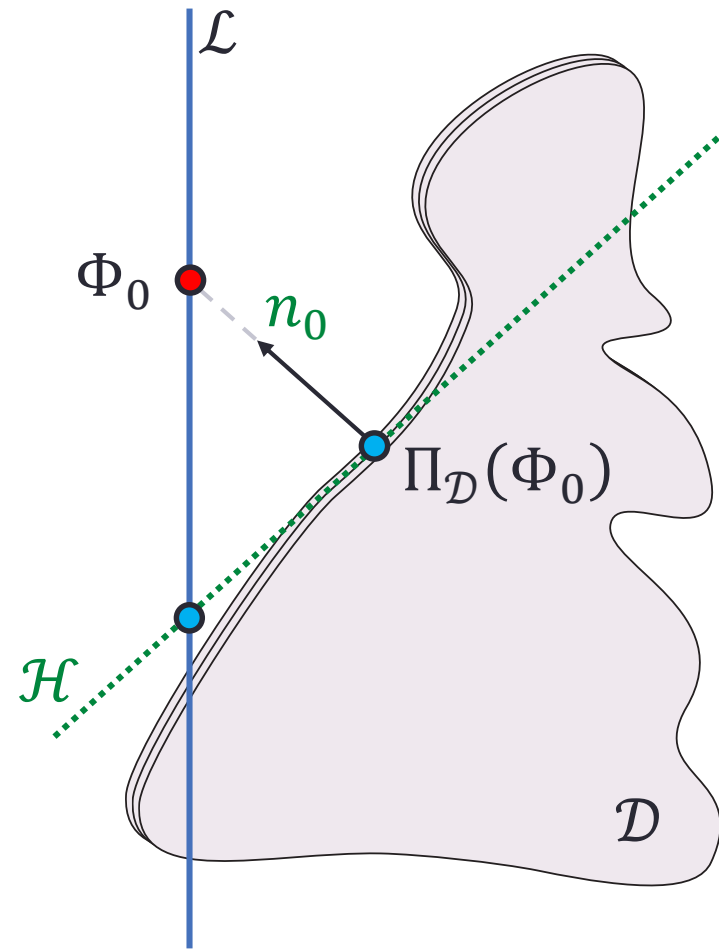
# Algorithm

- Equivalent KKT system:

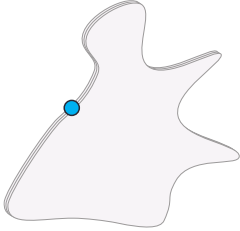
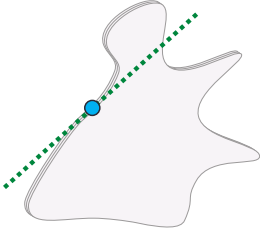
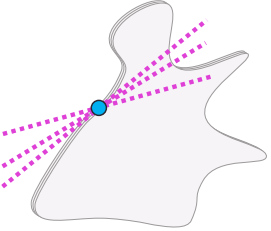
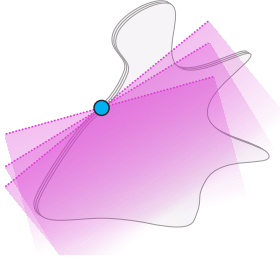
$$\begin{bmatrix} M & n_0 \\ n_0^T & 0 \end{bmatrix} x = b_0$$

FIXED!

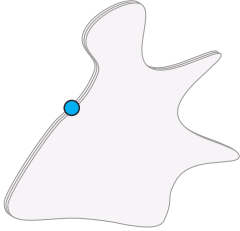
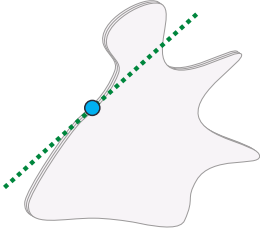
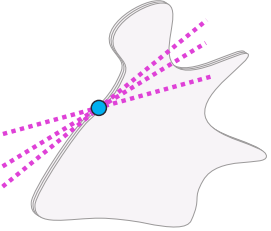
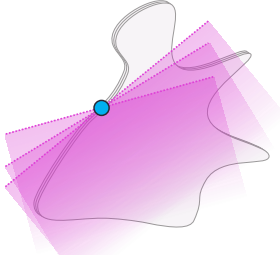
- Efficient pre-factorization
- Closed form expression
- Complexity: **2 back-substitutions**



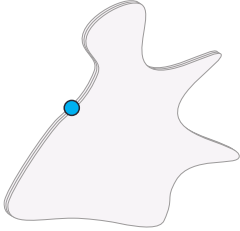
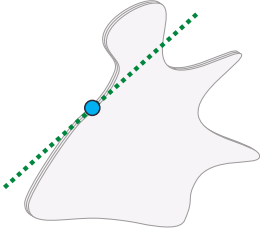
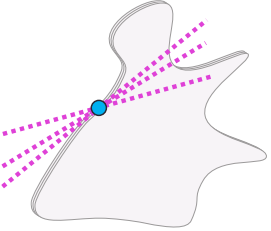
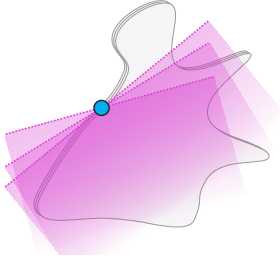
# Summary

	Alternating	Ours	Gauss-Newton	Aigerman 2013
Proxy				

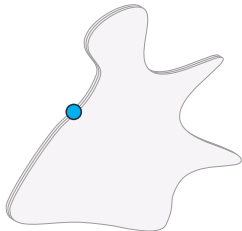
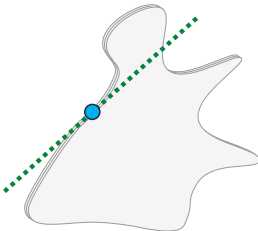
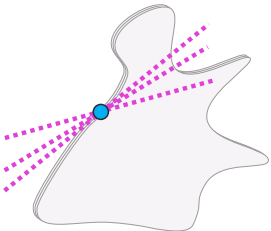
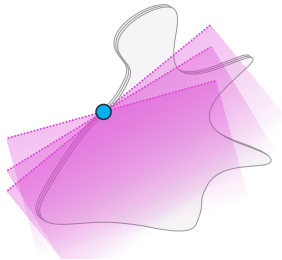
# Summary

	Alternating	Ours	Gauss-Newton	Aigerman 2013
Proxy				
Solve	Linear	Linear	Linear	QP

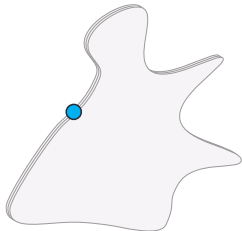
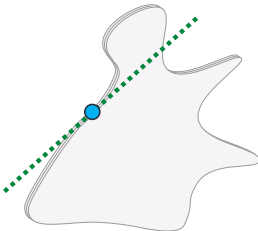
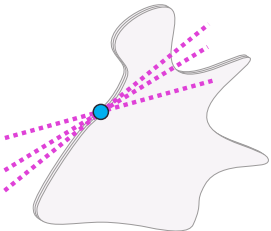
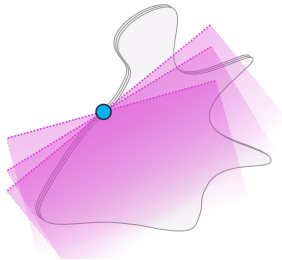
# Summary

	Alternating	Ours	Gauss-Newton	Aigerman 2013
Proxy				
Solve	Linear	Linear	Linear	QP
Factorization	Yes	Yes	No	-

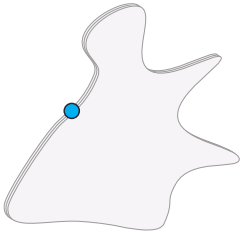
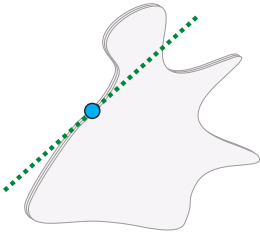
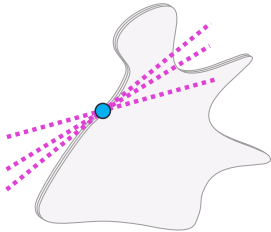
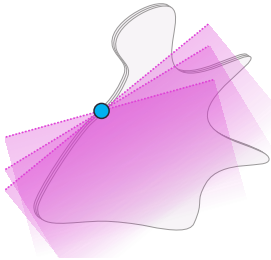
# Summary

	Alternating	Ours	Gauss-Newton	Aigerman 2013
Proxy				
Solve	Linear	Linear	Linear	QP
Factorization	Yes	Yes	No	-
Convergence	1 <sup>st</sup>	~2 <sup>nd</sup>	~2 <sup>nd</sup>	2 <sup>nd</sup>

# Summary

	Alternating	Ours	Gauss-Newton	Aigerman 2013
Proxy				
Solve	Linear	Linear	Linear	QP
Factorization	Yes	Yes	No	-
Convergence	1 <sup>st</sup>	~2 <sup>nd</sup>	~2 <sup>nd</sup>	2 <sup>nd</sup>
Parameters	No	No	Yes	No

# Summary

	Alternating	Ours	Gauss-Newton	Aigerman 2013
Proxy				
Solve	Linear	Linear	Linear	QP
Factorization	Yes	Yes	No	-
Convergence	1 <sup>st</sup>	~2 <sup>nd</sup>	~2 <sup>nd</sup>	2 <sup>nd</sup>
Parameters	No	No	Yes	No
Guarantee	No	No	No	No

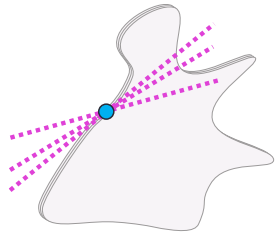
# Evaluation and Applications

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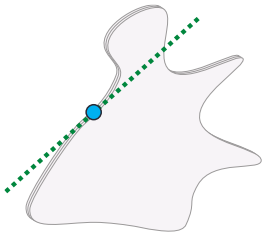


# Comparison

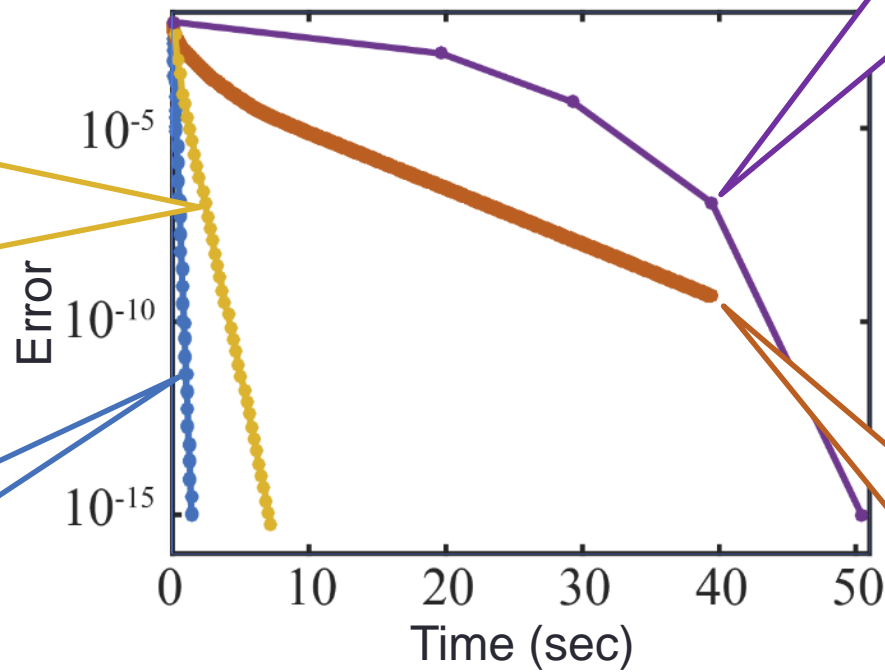
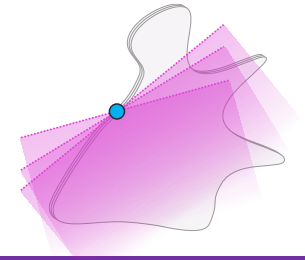
## Gauss-Newton



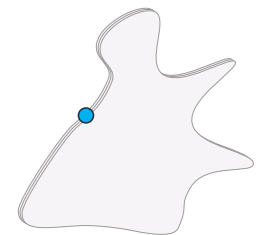
## Our approach



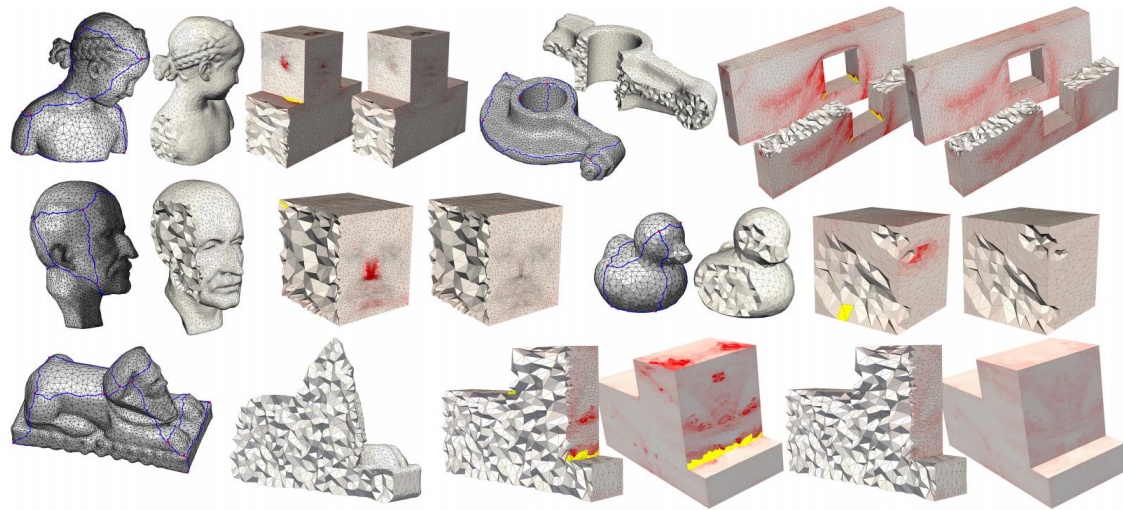
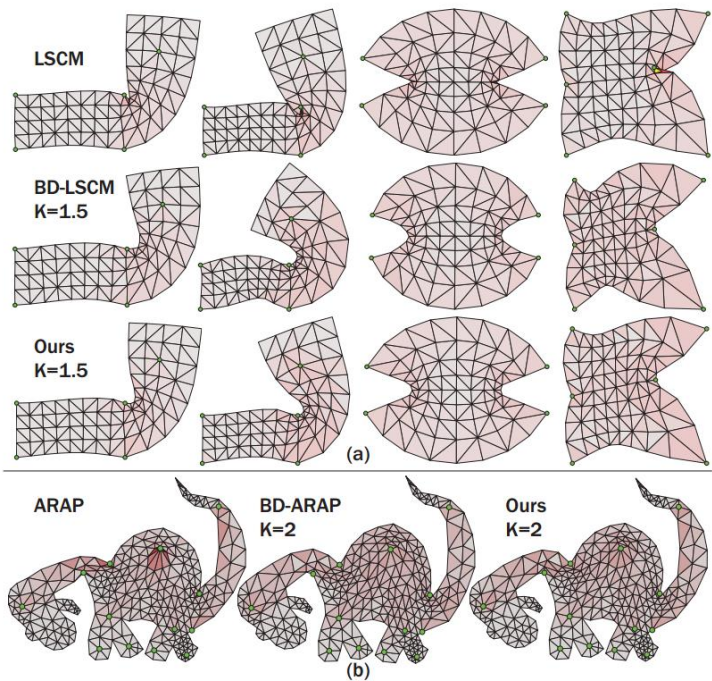
## Aigerman 2013



## Alternating



# Comparison with [Aigerman 2013]

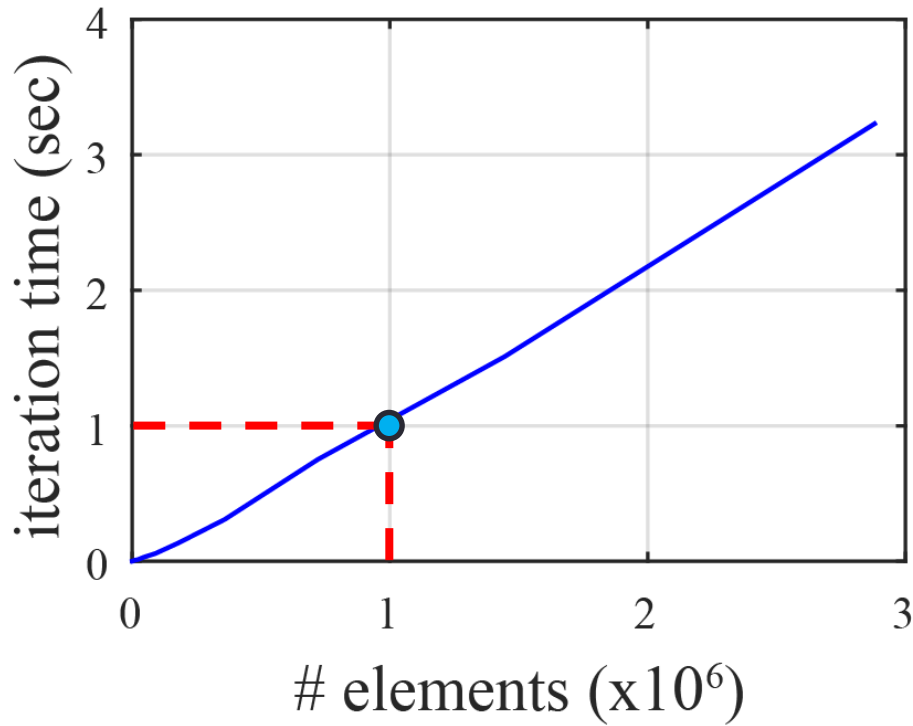


<60,000 elements

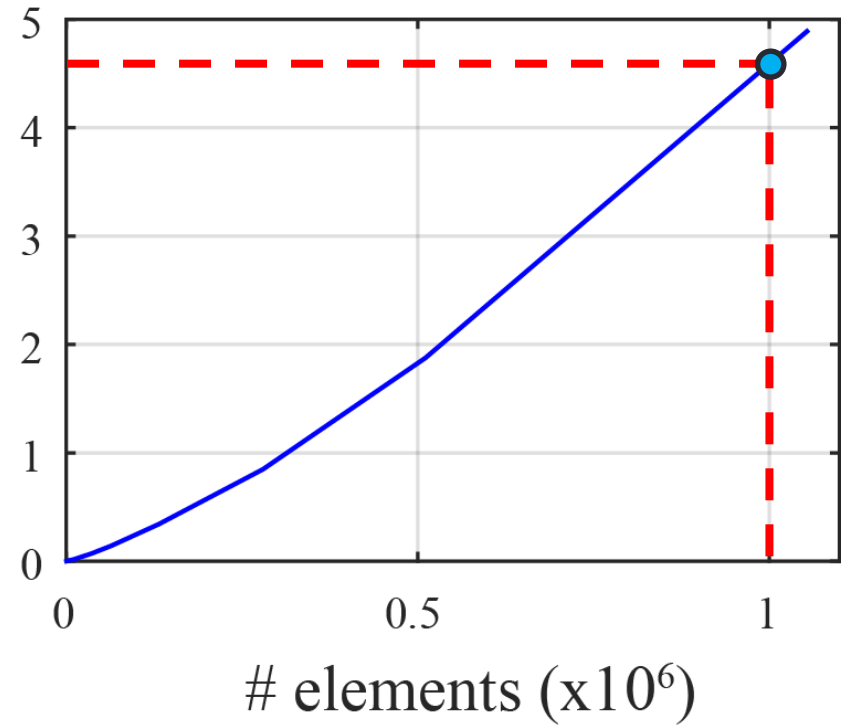
Comparable results  
x100 times faster

# Scalability

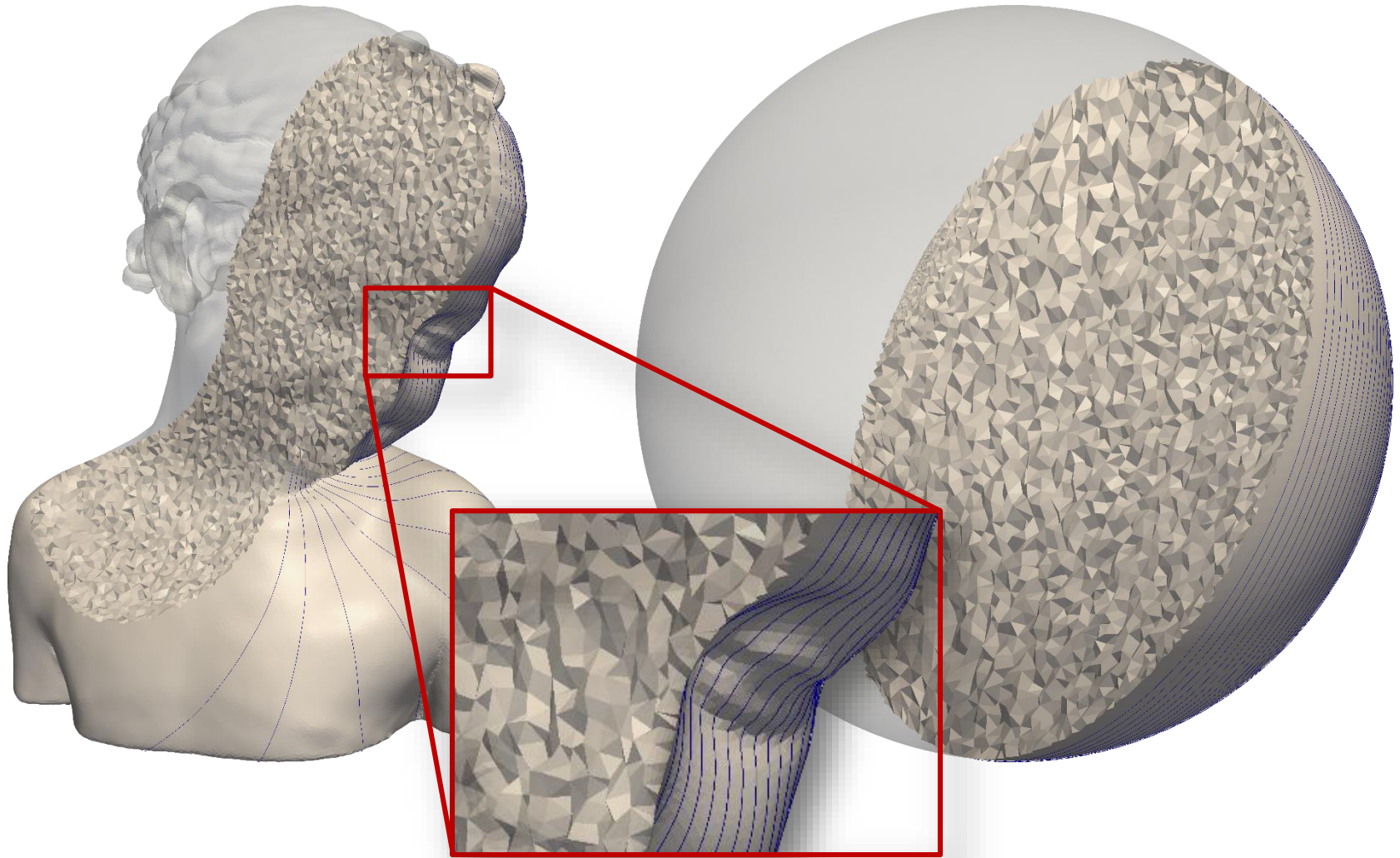
## 2D



## 3D



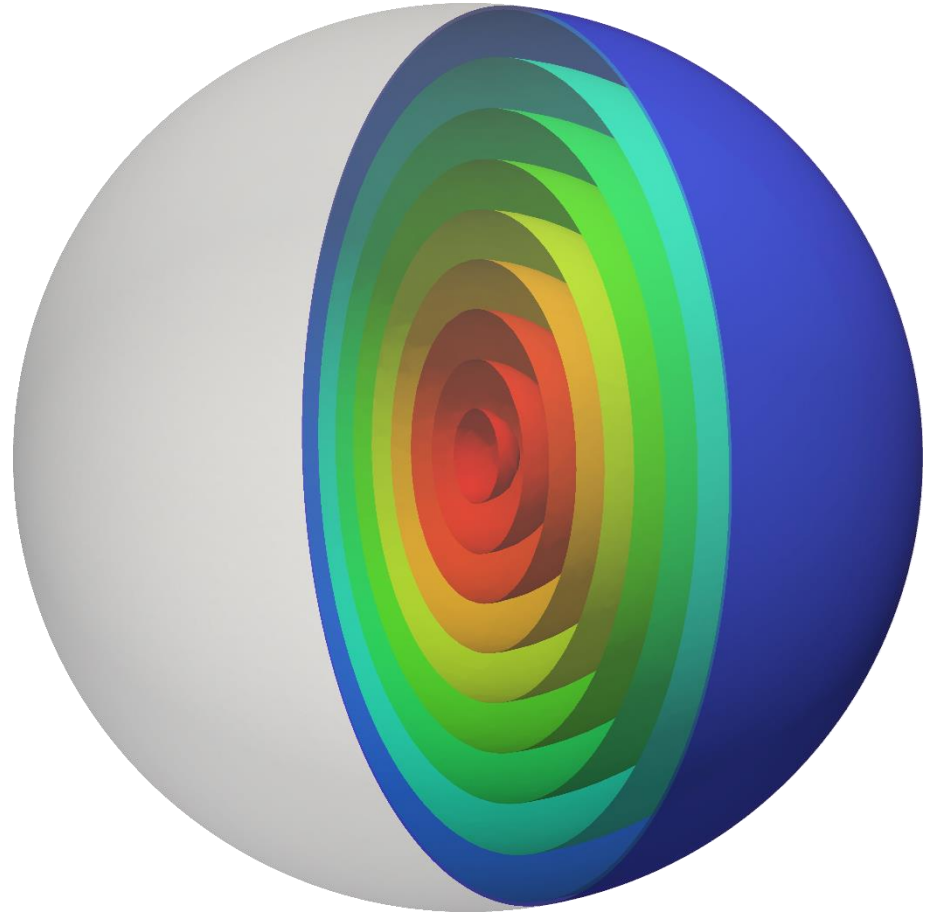
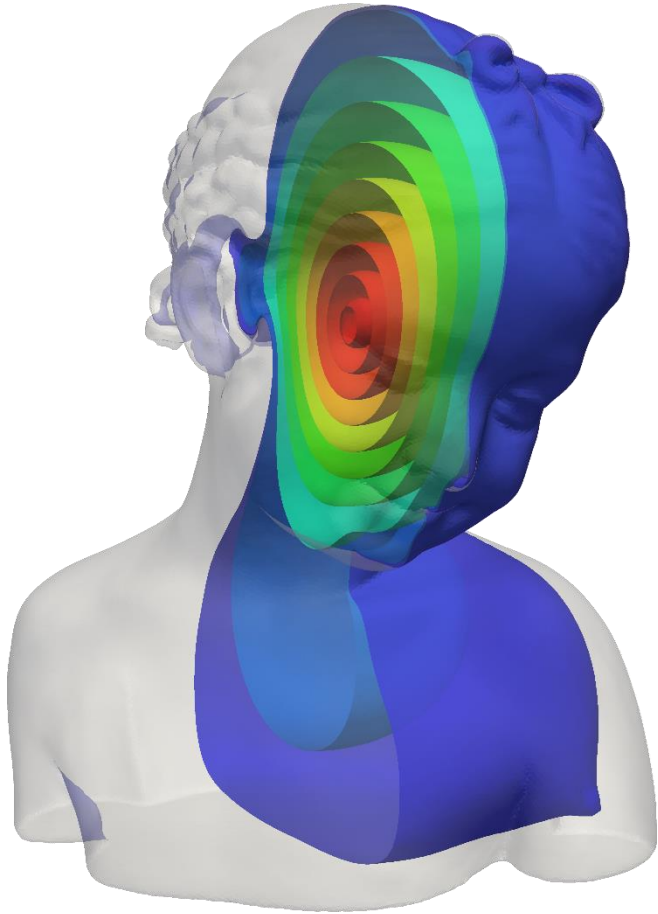
# Volumetric Mapping



**>1,200,000 Tetrahedra**



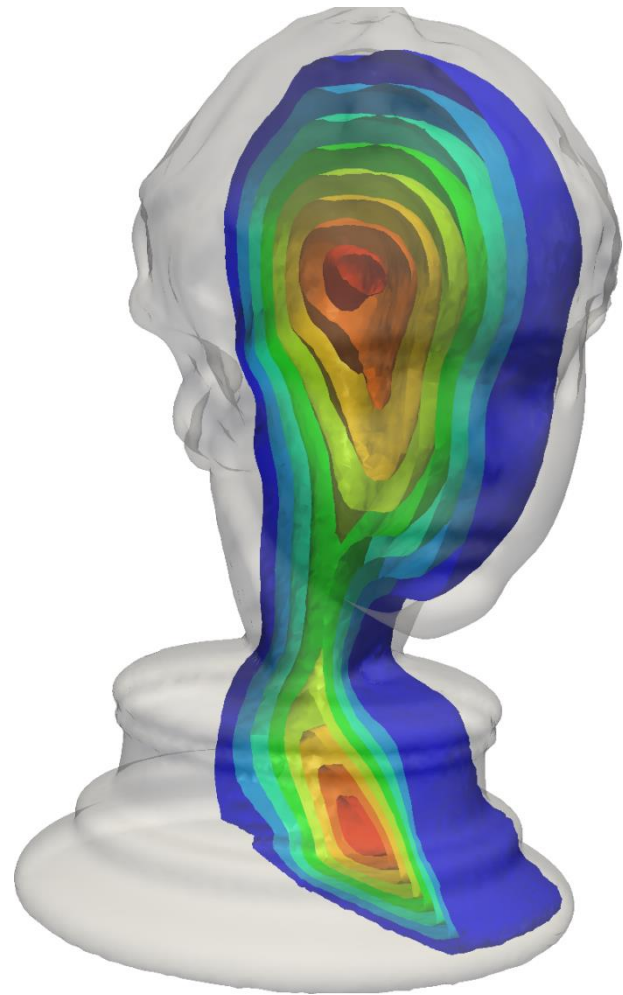
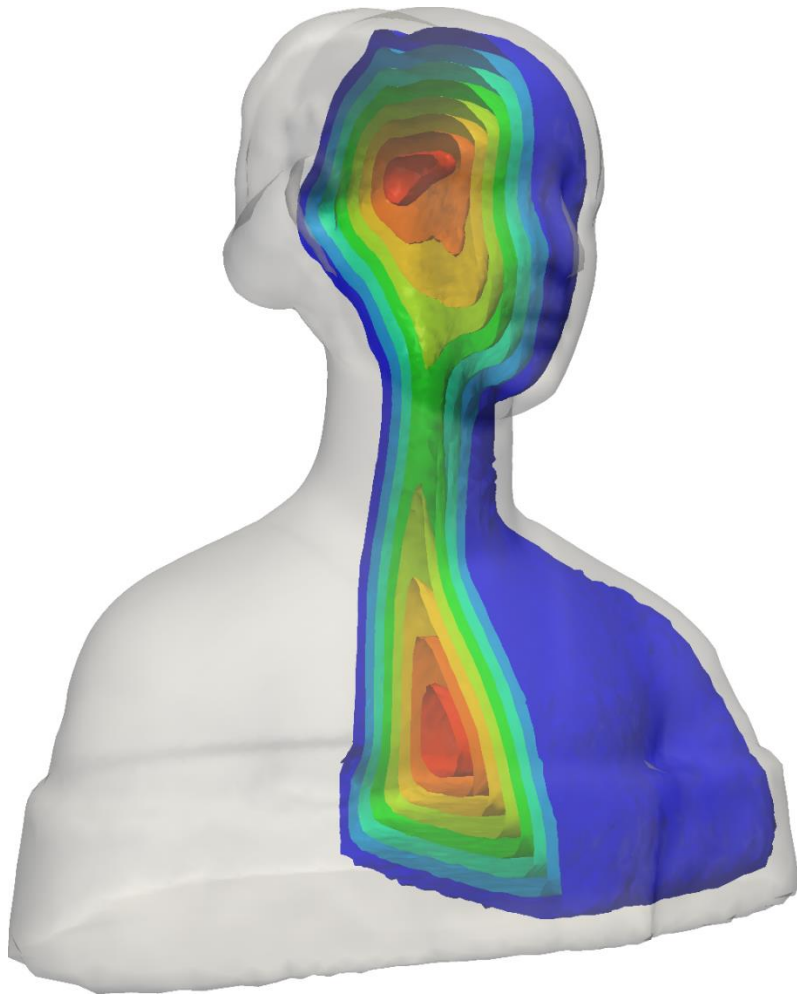
# Volumetric Mapping



# Volumetric Mapping



# Volumetric Mapping

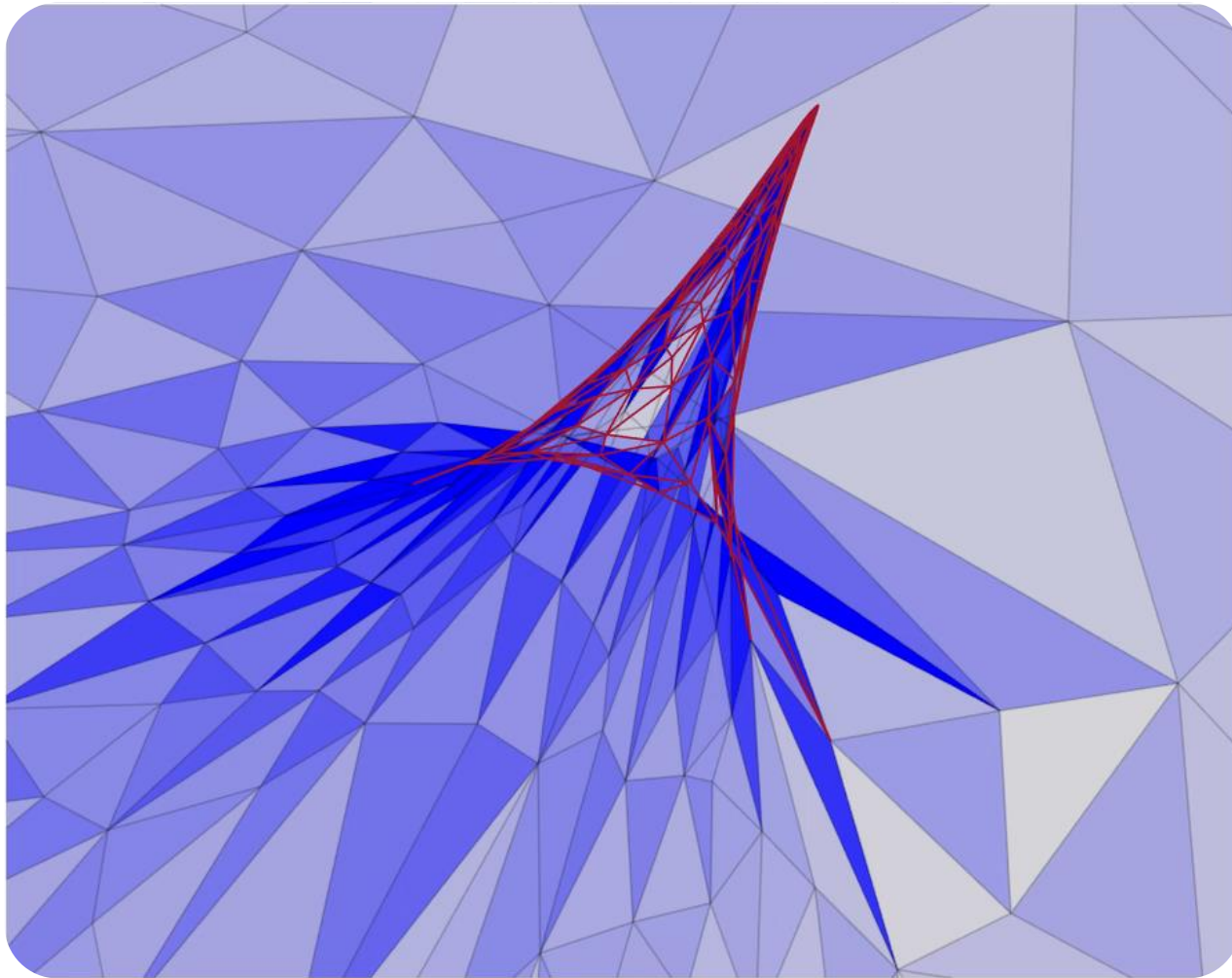


# Parameterization



**>500,000 Triangles**

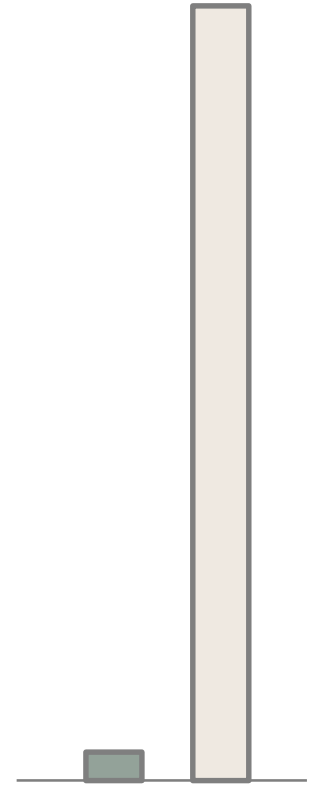
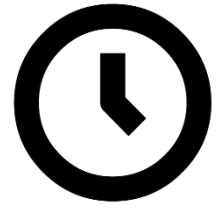
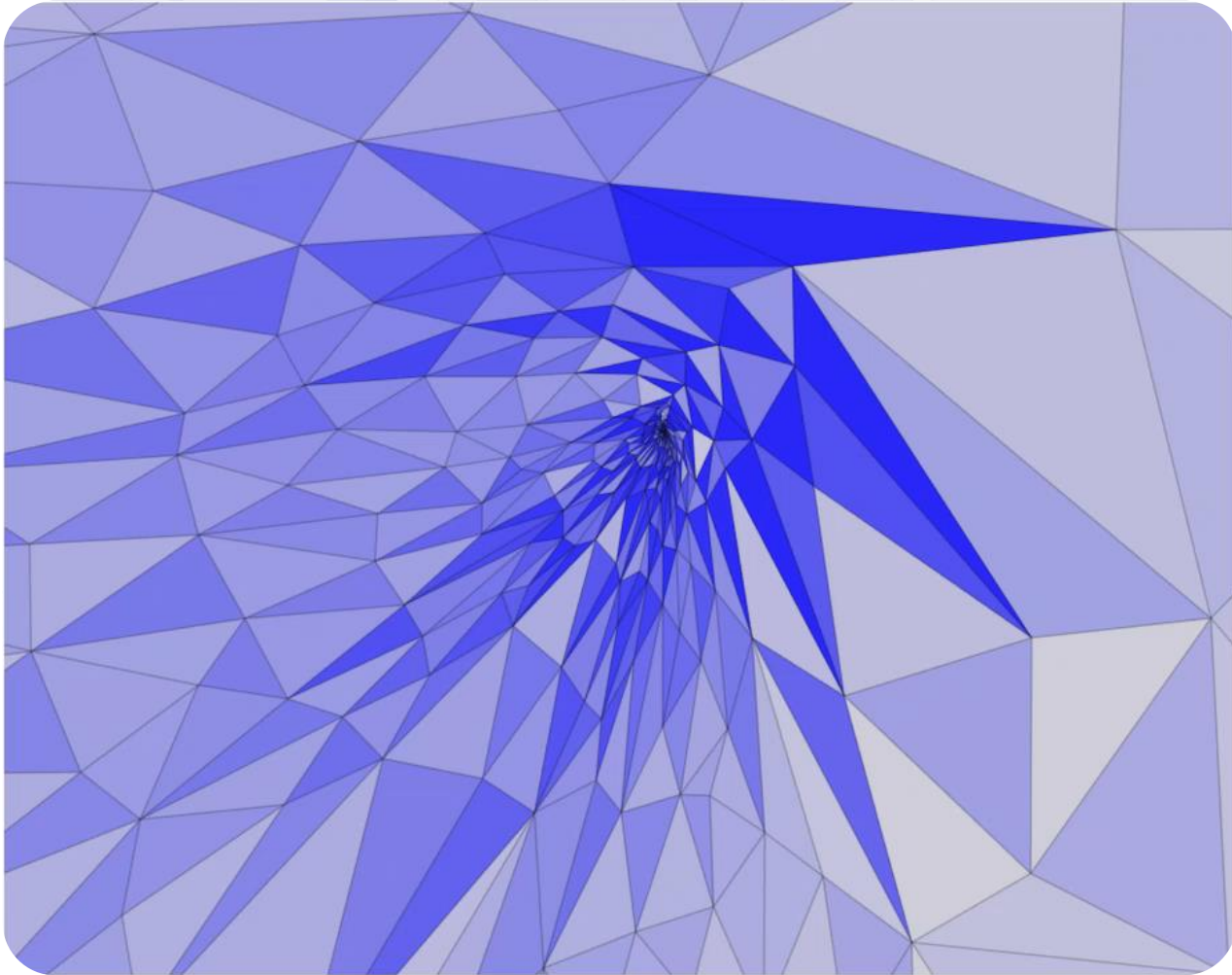




**Flipped  
Elements**

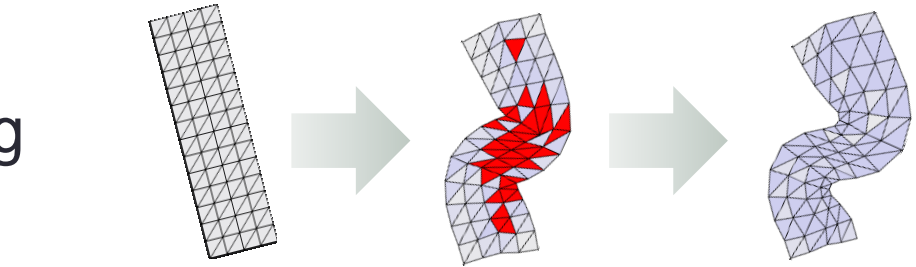


**High  
Distortion**

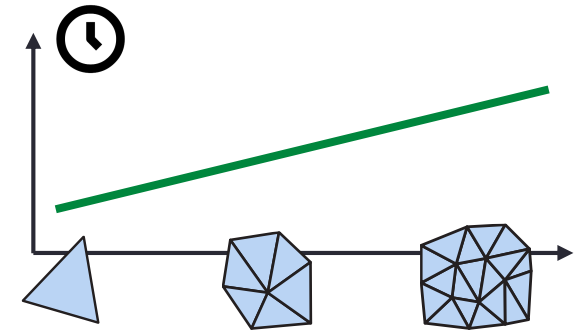
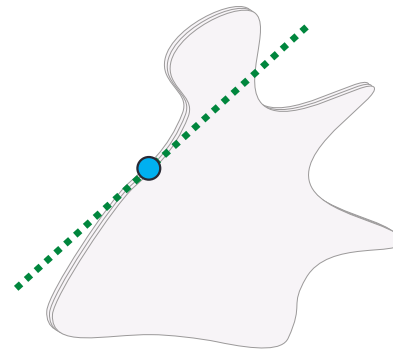


# Summary

- Find similar BD mapping



- Simple algorithm
- Efficient and Scalable



- No guarantee
- Optimize other energies?

Funded by:

- European Research Council
- Israel Science Foundation
- I-CORE program of the Israel PBC and ISF

# Thank you!

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