

# Optimization of industrial parts by mesh morphing enabled automatic shape sculpting

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#### **Outline**

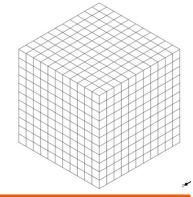
- Introduction
- BGM Background
- Adjoint Background
- RBF Background
- Automatic Surface Sculpting
- Applications
  - Simple thick plate
  - Industrial component
- Conclusions

#### Introduction

Mechanical component optimization is a target for engineering applications.

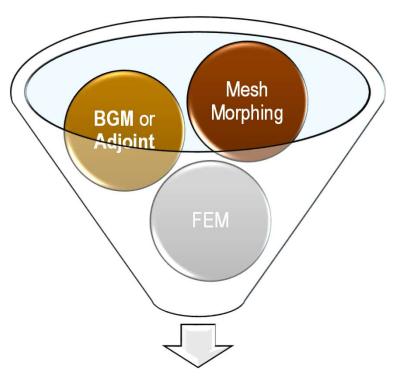


- Finite Element Method is a tool for optimization given load and constraint configuration
- CAD based optimization requires the generation of additional FEM models. It can be very time-consuming specially dealing with complex shape components.
- To overcome this, Mesh morphing can be adopted: It generates new FEM models without modifying the geometry and without the need to remesh it.



#### Introduction

- Biological Growth Method (BGM) and Adjoint Method exploit data coming from numerical analysis to define a shape modification that will sculpt model surfaces so that stress levels are optimized.
- The tool adopted for morphing the FEM mesh is RBF Morph™, which is based on Radial Basis Functions (RBFs). FEM pre-processor and solver used is ANSYS® Mechanical™.

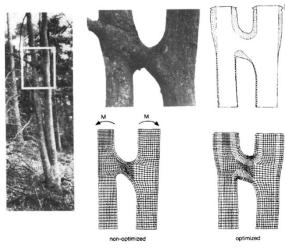


**Automatic Shape optimization** 



#### **BGM** Background

- BGM\* approach is based on the observation that biological structures growth is driven by local level of stress.
- Bones and trees' trunks are able to adapt the shape to mitigate the stress level due to external loads.
- The process is driven by stress value at surfaces. Material can be added or removed according to local values.



Reduction of maximum stresses 56 %



The BGM idea is that surface growth can be expressed as a linear law with respect to a given threshold value:  $\dot{\varepsilon} = k \left( \sigma_{Mises} - \sigma_{ref} \right)$ 

\*Mattheck C., Burkhardt S., 1990. A new method of structural shape optimization based on biological growth. Int. J. Fatigue 12(3):185-190.

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## **BGM** Background

Waldman and Heller\* refined this first approach proposing a multi peak one:

$$d_i^j = \left(\frac{\sigma_i^j - \sigma_i^{th}}{\sigma_i^{th}}\right) \cdot s \cdot c, \qquad \sigma_i^{th} = \max(\sigma_i^j) \text{ if } \sigma_i^j > 0 \qquad \text{or} \qquad \sigma_i^{th} = \min(\sigma_i^j) \text{ if } \sigma_i^j < 0$$

In RBF Morph ANSYS Workbench ACT a different implementation is present

$$S_{node} = \frac{\sigma_{node} - \sigma_{th}}{\sigma_{max} - \sigma_{min}} \cdot d$$

and different **stress types** can be used to modify the surface shape:

- Von Mises stress
- Maximum principal stress
- Minimum principal stress
- Stress intensity
- Maximum Shear stress
- Equivalent plastic strain

\*Waldman W., Heller M., 2015. Shape optimization of holes in loaded plates by minimization of multiple stress peaks, Defence Science and Technology Organisation Fisherman Bend, Australia

## **Adjoint Background**

- Adjoint method allows to obtain the sensitivities of an objective function with respect to a set of input parameters.
- This can be applied to the three displacement for each node of the computational mesh, so that a shape modification can be obtained.
- Largely used in Computational Fluid-Dynamics (CFD) but can be also applied in Computational Structural Mechanics (CSM).



It is possible to differentiate the discretised equation (Discrete Adjoint method) or to derive equation prior to their differentiation (Continuous Adjoint method).

## **Adjoint Background (discrete case)**

The objective function can be expressed as function of displacement and the derivative:

$$\Psi = f(\mathbf{X}(u), u) \qquad \frac{d\Psi}{du} = \frac{\partial \Psi}{\partial u} + \frac{\partial \Psi}{\partial \mathbf{X}} \frac{\partial \mathbf{X}}{\partial u}$$

- To obtain the displacement  $\frac{\partial \mathbf{X}}{\partial u}$  two methods are available:
  - direct method → has to be re-evaluated for each input parameter)
  - adjoint method → need only one calculation no matter how many input parameters, it uses a Lagrange-like multiplier to obtain displacements:

$$\mathbf{K}\mathbf{X} = \mathbf{F} \longrightarrow \mathbf{K}\lambda = \frac{\partial \Psi}{\partial \mathbf{X}}^{\mathrm{T}} \longrightarrow \frac{\partial \Psi}{\partial u} = \frac{\partial \Psi}{\partial u} + \lambda^{\mathrm{T}} \left( \frac{\partial \mathbf{F}}{\partial u} - \mathbf{X} \frac{\partial \mathbf{K}}{\partial u} \right)$$

#### **RBF Background**

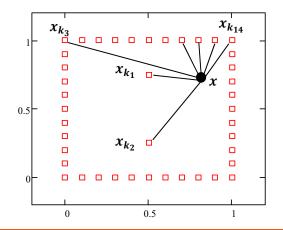
RBFs = mathematical tool capable to interpolate in a generic point in the space a function known in a discrete set of points (source points).

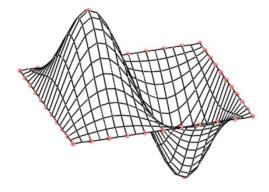


The interpolating function is composed by a radial basis and by a polynomial:

$$s(\mathbf{x}) = \sum_{i=1}^{N} \gamma_i \varphi(\|\mathbf{x} - \mathbf{x}_{\mathbf{k}_i}\|) + h(\mathbf{x})$$

distance from the i-th source point





## **RBF Background**

- If evaluated on the source points, interpolating function gives exactly the input values
- RBF problem (evaluation of coefficients  $\gamma$  and  $\beta$ ) is associated to the solution of a linear system:
  - M = interpolation matrix
  - P = constraint matrix
  - g = vector of known values on the source points
- Once solved the RBF problem each displacement component is interpolated

$$\begin{bmatrix} \mathbf{M} & \mathbf{P} \\ \mathbf{P}^{\mathsf{T}} & 0 \end{bmatrix} \begin{pmatrix} \boldsymbol{\gamma} \\ \boldsymbol{\beta} \end{pmatrix} = \begin{pmatrix} \boldsymbol{g} \\ 0 \end{pmatrix}$$

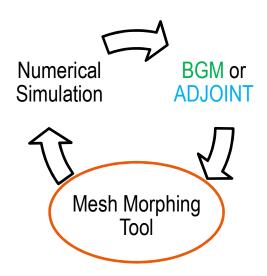
$$M_{ij} = \varphi \left( \mathbf{x}_{k_i} - \mathbf{x}_{k_j} \right) \qquad 1 \le i, j \le N$$

$$\mathbf{P} = \begin{bmatrix} 1 & x_{k_1} & y_{k_1} & z_{k_1} \\ 1 & x_{k_2} & y_{k_2} & z_{k_2} \\ \vdots & \vdots & \vdots & \vdots \\ 1 & x_{k_N} & y_{k_N} & z_{k_N} \end{bmatrix}$$

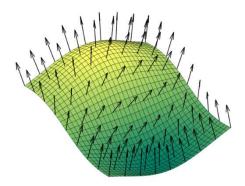
$$\begin{cases} s_{x}(\mathbf{x}) = \sum_{i=1}^{N} \gamma_{i}^{x} \varphi(\mathbf{x} - \mathbf{x}_{k_{i}}) + \beta_{1}^{x} + \beta_{2}^{x} x + \beta_{3}^{x} y + \beta_{4}^{x} z \\ s_{y}(\mathbf{x}) = \sum_{i=1}^{N} \gamma_{i}^{y} \varphi(\mathbf{x} - \mathbf{x}_{k_{i}}) + \beta_{1}^{y} + \beta_{2}^{y} x + \beta_{3}^{y} y + \beta_{4}^{y} z \\ s_{z}(\mathbf{x}) = \sum_{i=1}^{N} \gamma_{i}^{z} \varphi(\mathbf{x} - \mathbf{x}_{k_{i}}) + \beta_{1}^{z} + \beta_{2}^{z} x + \beta_{3}^{z} y + \beta_{4}^{z} z \end{cases}$$

#### **Automatic Surface Sculpting**

 Automatic optimization is accoplished connecting BGM or ADJOINT data from numerical simulation to mesh morphing tool.

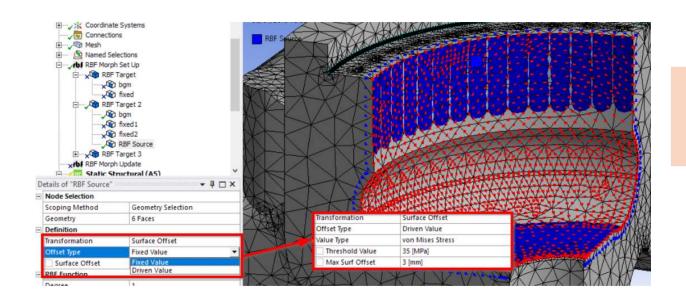


each node can displace according to the local normal direction



## **Automatic Surface Sculpting - BGM**

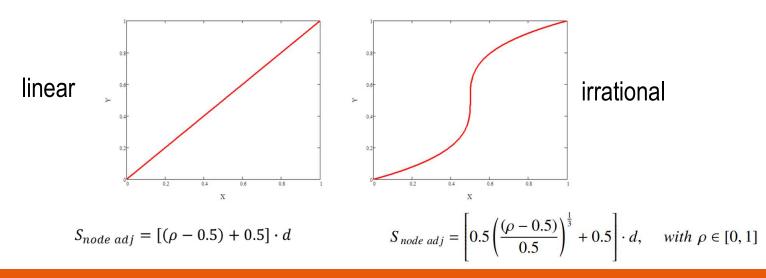
• Direction and amplitude of displacement is defined according to stress data, considering the threshold stress value  $\sigma_{th}$  and the d maximum displacement.



$$S_{node} = \frac{\sigma_{node} - \sigma_{th}}{\sigma_{max} - \sigma_{min}} \cdot d$$

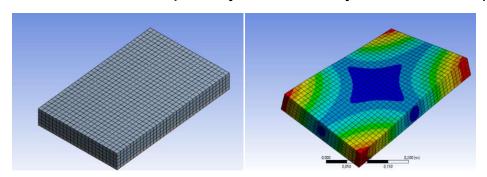
## **Automatic Surface Sculpting - Adjoint**

■ Direction and amplitude of displacement is defined according to topological optimization data, considering the topological density function,  $\rho \in [0,1]$ , interpolated using a linear or irrational function and the d maximum displacement.



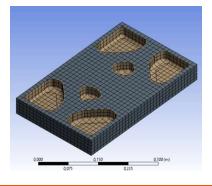
## **Applications – Simple Thick Plate – Adjoint**

Free and undamped dynamic analysis of a thick plate



Mode		Freq. (Hz)
	1	1457
	2	1542
	3	3159
	4	3800
	5	3816

- Optimization problem:
  - decrease mass (target -50%)
  - maintain first frequency above 1220 Hz

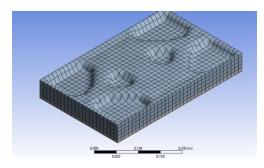


Topology Optimization Result

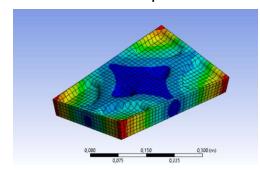


## **Applications – Simple Thick Plate – Adjoint**

Results (d = 15mm)



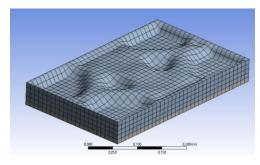
**Irrational Interpolation** 



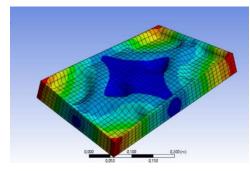
Mode	Baseline Freq. (Hz)	Linear int. Freq. (Hz)	Irrational int. Freq. (Hz)
1	1457	1316	1352
2	1542	1402	1441

Mass reduction = 20% (< 50%)

Topological Optimization removes material → mesh morphing moves nodes.

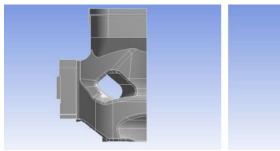


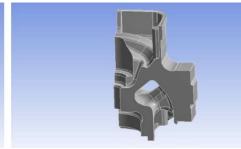
**Linear Interpolation** 

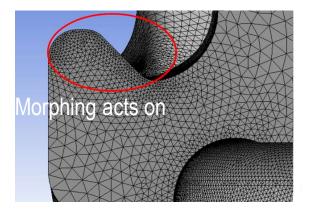


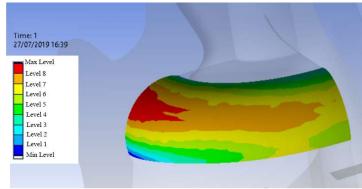
## **Applications – Industrial component – Adjoint**

Maximum reference stress reduction: 11.7 %





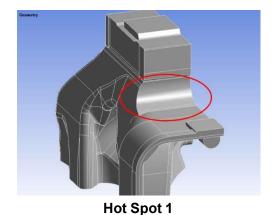


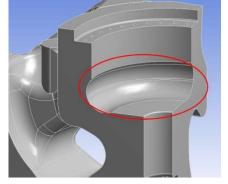


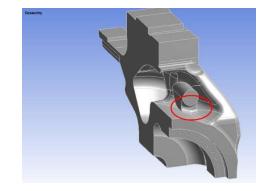
3 Hot Spots identified

Location	$\sigma_{th}$	d [mm]
Hot Spot 1	$0.5(\sigma_{max}$ - $\sigma_{min})$	3
Hot Spot 2	$0.5(\sigma_{max}$ - $\sigma_{min})$	3
Hot Spot 3	$0.5(\sigma_{max}$ - $\sigma_{min})$	1

$$S_{node} = \frac{\sigma_{node} - \sigma_{th}}{\sigma_{max} - \sigma_{min}} \cdot d$$



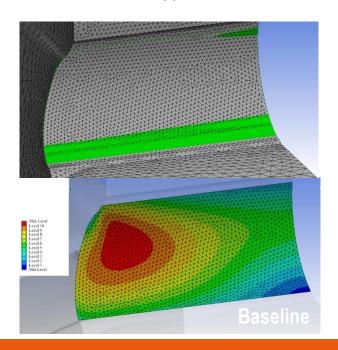


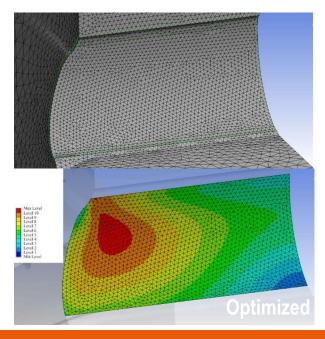


**Hot Spot 2** 

**Hot Spot 3** 

- Hot Spot 1
- Max stress reduction: 2.76%

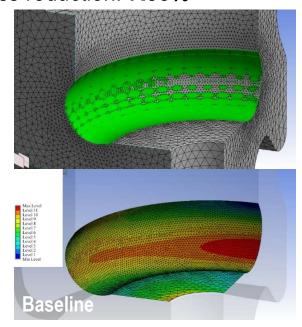


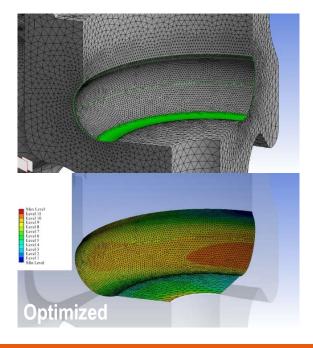


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- Hot Spot 2
- Max stress reduction: 7.85%

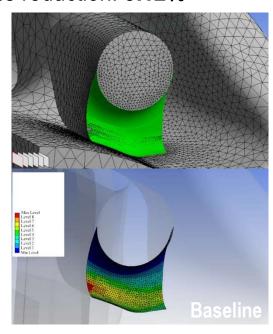


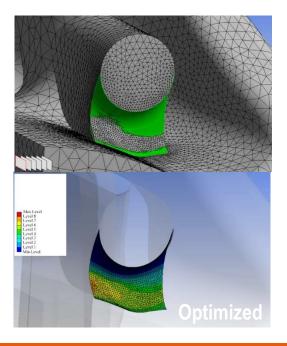


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- Hot Spot 3
- Max stress reduction: 8.12%





#### **Conclusions**

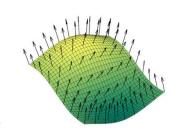
- A methodology to perform automatic shape optimization via surface sculpting was presented.
- Two approaches were investigated Biological Growth Method and Adjoint Method, which were successfully coupled with mesh morphing tool.
- Adjoint method, at the cost of an additional computation, gives sensitivities
  with respect of objective functions which were used to identify surface nodes to
  be moved inward or outward.
- BGM uses surfaces stress levels to identify model zones to be moved inward or outward, obtaining stress peak minimization and more uniform stress distribution





#### **Conclusions**

- Both approaches were illustrated on simple geometries and then applied to a more complex case.
- The proposed procedure allowed to reach optimized shapes matching different objective functions in an automatic way, with a minimal effort from the user.



 The framework adopted was composed by ANSYS Workbench with RBF Morph ACT extension, which provided an integrated tool to the user in which perform optimization task



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