Application: AEF-A.10 Optimizing the solutions

KEY WORDS

Linear Static Analysis, Optimization, Linear Material, 2D Geometric Model, 2D Finite Element, Linear Finite Element, Element, Design Parameters, Status Parameters, Objective Function

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A. DESCRIEREA PROBLEMEI / PROBLEM DESCRIPTION

A.1 Introduction

In general, the FEA determines values of the output parameters (deformations, displacements, stresses), depending on the preliminary predefined model parameters. Some FEA have distinct optimization modules that for a preliminarily analyzed structure allow the determination of independent parameters, consequence of solving an optimization model that involves minimizing / maximizing some purpose functions while imposing restrictions of other dependent parameters (see Chapter F)



This application presents, using finite element analysis, the algorithm for solving the problem of dimensional constructive optimization of the structure in the figure above. For preliminary FEA we consider: L = 50 mm, H = 40 mm, G = 10 mm, a = 20 mm. The values of the optimization model parameters are: $D_{min} = 14 \text{ mm}$, $D_{max} = 18 \text{ mm}$, $H_{min} = 35 \text{ mm}$, $H_{min} = 44 \text{ mm}$, $\sigma_a = 140 \text{ Mpa}$.

B. THE FEA MODEL

B.1 The model definition

For the analysis and optimization with FE, the following simplifying hypotheses are adopted:

- linear behavior of the material,
- adopting constraints associated with symmetry properties,
- external load by force distributed on the surface,
- the proposed problem is solved in two stages: structural analysis and optimization.

B.2 The analysis model description

Figure a shows the FEA and optimization model associated with the <u>geometric plane model</u> considered in the XY plane. The X-axis is the axis of symmetry of this model. In addition, the design parameters: hole diameter (P1) and width (P2) are also highlighted for optimization.



B.3 Characteristics of the material and the environment

The strength characteristics of E335 material for finite element analysis are:

- the modulus of longitudinal elasticity, $E = 210000 \text{ N} / \text{mm}^2$;
- transverse contraction coefficient (Poisson), v = 0.3.
- Average working temperature of the subassembly, $T_0 = 20^0$ C.

C. PREPROCESSING OF FEA MODEL



C.2 Modelling of material and environment characteristics					
\bigwedge , Project Schematic: \Box \checkmark Engineering Data \checkmark \downarrow \rightarrow \downarrow	Edit Outline of Schematic A2: Engineering Data				
🚽 🦤 Structural Steel Properties of Outline Row 3: Structural Steel 🛛 🖃	\boxtimes Isotropic Elasticity \rightarrow Young's Modulus , [select from				
list, C column(Unit) cu $\downarrow \checkmark$, \downarrow MPa], [input in the box in column B (Unit) value, 210000] $\rightarrow \downarrow$					
\checkmark Update Project $\rightarrow \Box$ Return to Project (the other parameters remain the default).					

C.3 Geometric modelling					
C.3.1 Model loading, DesignModeler (DM)					
ے 🖓 Geometry 🖓 🖌 ط	New Geometry \rightarrow ANSYS Workbench : \Box^{\bigcirc} Millimeter , \Box^{\bigcirc} OK				



Generation of embedding constraint

J \checkmark Supports ▼ → J Fixed Support; → J (the line selection filter is activated) → [select left face (fig. C.6,b)];



D. SOLVING THE AEF MODEL



E. POST-PROCESSING OF RESULTS



F. PREPROCESSING OF THE OPTIMIZATION MODEL

F.1 Setting input (design) and output (status) parameters					
Setting input (design) parameters					
$\textcircled{0}: \square \swarrow \textcircled{2} \xrightarrow{\mathbb{C}} \overset{\text{Sketch1}}{\longrightarrow} \xrightarrow{\text{Details View}} \xrightarrow{\square} \textcircled{\text{Dimensions: 4}}: $ [activated with \dashv the button associated with the circle					
diameter, $\square D D] \rightarrow A$: Static Structural - DesignModeler, Parameter Name:, [input the name, Diameter], $\square OK$;					
[activated with \downarrow the button associated with the rectangle width dimension, $D^{12} \rightarrow$					
A: Static Structural - DesignModeler, Parameter Name:, [input the name, Width], \downarrow^{OK} ($\downarrow^{\Lambda} \rightarrow$ Project Schematic : (the					
input parameter setting loop appears automatically, fig. a).					

Setting output (status) parameters	Geometry" I Properties Lis activated with the button associated						
with the mass \mathbb{P} Mass \mathbb{A} Solution (A6) $\rightarrow \mathbb{P}$ Equivalent Stress \rightarrow Details of "Equivalent Stress" \square Results Fig							
activated with \downarrow the button associated with the maximum equivalent voltage P Maximum $\downarrow \downarrow OK$ ($\downarrow \Lambda \rightarrow$							
Project Schematic : (the output parameter setting	loop appears automatically, fig. b).						
 A Static Structural Engineering Data 	 A 1 Z Static Structural 2 Primeering Data 						
3 🔘 Geometry 🗸 🖌	3 0 Geometry 🗸						
4 💓 Model 🗸 🖌	4 🎯 Model 🗸 🖌						
5 😪 Setup 🗸 🖌	5 🕵 Setup 🗸 🖌						
6 Solution	6 🗑 Solution 🗸 🖌						
7 💓 Results 🗸 🖌	7 💓 Results 🗸 🖌						
> 8 Cp2 Parameters	> 8 Parameters						
Static Structural	Static Structural						
Parameter Set	ि Parameter Set						
<i>a</i> .	<i>b</i> .						
F.2 Launc	hing the optimization module						
ے 🔨 🕂 Design Exploration 🔶	႕ 🧿 Goal Driven Optimization (fig. a).						
•	A						
1	2 Static Structural						
2	🥏 Engineering Data 🗸 🖌						
3	随 Geometry 🛛 🧧 🛓						
4	📦 Model 🛛 💱 🛓						
5	🍓 Setup 🛛 💝 🛓						
6	🕼 Solution 🛛 💝 🛓						
7	😥 Results 🛛 💝 🛓						
8	Parameters						
	Static Structural						
CpJ Paramete	er Set						
*	в						
1 😡	Goal Driven Optimization						
2	Design of Experiments 🦩 🗸						
3	Response Surface						
4 😡	Optimization ?						
	Goal Driven Ontimization						
F.3 Generating	<i>u</i> . and visualizing feasible solutions						
Generating feasible solutions							
$_{\downarrow}$ I Design of Experiments $~\neq~$ $_{\downarrow}$ $\rightarrow~$ Outline of Sc	chematic B2: Design of Experiments : (, View \rightarrow , Outline , $Properties$, ,						
Table, \Box Chart), \Box P1-Diametrul \rightarrow Properties of Outline A5: P1: Lower Bound. [input the lower value. 14.4].							
Upper Bound [input the upper value, 17,6]; J P2 - Latimea Properties of Outline A6: P2 Lower Bound , [input the lower							
value, 36], Upper Bound [input the upper value, 44]. $\downarrow \neq$ Update ANSYS Workbench $\rightarrow \downarrow$ Yes (after							
processing the table with test variants appear	s, Table of Schematic B2:). J CReturn to Project						
Visualization of feasible solutions							
🚬 💽 Response Surface 刘 🖌 Outline of Schematic B3: Response Surface 🚬 🖓 Response Surface 🛶 🚽 🖉 Update .							
Outline of Schematic B3: Response Surface $ ightarrow$, $ ightarrow$	Response , Properties of Outline A16: Response , Mode , [select from list ,						



G. SOLVING THE OPTIMIZATION MODEL

_	\rightarrow B	: J Optimization	2 🖌	\rightarrow Table of	Schematic B4: Optimization	Optimization Objectives		
Objective , [select in column D from list العابي, الم Minimize], [select in column E from list العابي,								
_JValues <= Target]; Target Value, [input in column D the value limit, 140]. Properties of Outline A2: Optimization,								
Detimization Pethod, [select from list, NLPQL].								
Outline of Schematic B4: Optimization \rightarrow , \square \bigcirc Optimization \square \checkmark Update (appear in window								
Table of Schematic B4: Optimization lines from fig.a).								
	11 Candidate Points							
	12	Candidate A	17,6	36	A 0,1222	68,565		
	13	Verification A			A 0,1222	68,589		
<i>a</i> .								
Obs. The NLPQL (Nonlinear Programming by Lagrangean Quadratic) method is based on the gradient								

algorithm for models with a single objective function and multiple constraints.

H. POST-PROCESSING OF RESULTS



I. ANALYSIS OF RESULTS

I.1 Interpretation of results

Following the analysis of the results obtained, as a result of the modeling and post-processing of the results (subchapters E and H), the following are highlighted:

- Following the deformation process of the non-optimized element (D = 16 mm, H = 40 mm) as a result of the action of the force F (subchapter A.2, fig. a) the maximum displacement is observed 0.0144468 mm (subchapter E. 2, Fig. a) in the area of the force action; the maximum equivalent stress has the value 56,614 MPa (subchapter E.2, fig. b) in the embedded area; the mass of the element is 141.22 g (subchapter F.3, Table of Schematic B2).
- Following the deformation process of the optimized element (D = 17.6 mm, H = 36 mm) as a result of the action of the force F (subchapter A.2, fig. a) the maximum displacement is observed 0.020152 mm

(subchapter. H.2, Fig. a) in the area of the force action; the maximum equivalent stress has the value 68.589 MPa (subchapter H.2, fig. b) in the embedded area; the mass of the element is 122.2 g (subchapter H.3, fig. a).

I.2 Design studies

The analysis of the above results shows the decrease of the element mass following the finite element solving of the optimization model; at the same time the increase of the maximum displacement (rigidity) is observed. In order to optimize related to other design restrictions, it is necessary to modify the analysis model, re-adopt the design and status parameters and the objective function. Thus, it is necessary, after the modifications of the analysis and / or optimization model, to solve it by activating the commands $\downarrow \bigcirc$ Refresh Geometry; \downarrow

 $\frac{3}{2}$ Solve . After solving the model, the results are reanalyzed and reinterpreted.

J. CONCLUSIONS

Modeling and analysis with finite elements in this paper were also carried out for didactic purposes following the initiation of the user with the main stages of development of a finite element optimization application in ANSYS Workbench, which emphasizes, above all, the modeling and analysis of a deformable element which is then dimensionally optimized.

The optimization model considered adopted involves the consideration of two geometric parameters as design variables, a state parameter (equivalent voltage) limited below the allowable value and the objective function that involves minimizing the mass of the element.

Following the solution of the finite element model of optimization, adopting the NLPQL method (Nonlinear Programming by Quadratic Lagrangean) which is based on the gradient algorithm for models with a single objective function and multiple constraints, the reduction of the element mass was obtained. maximum (but not exceeding the allowable value) and increasing the rigidity of the element.