
Reliability-Based Optimization of the Coupled Structural-Acoustic System with Random Parameters

Xiaojun Wang, Yunlong Li, Zhiliang Ma and Zhiping Qiu

Institute of Solid Mechanics, Beihang University, Beijing, 100191, China

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Structural noise is an important factor that endangers aircraft fatigue life and flight safety. It also has a negative effect on aircraft stealth performance and noise navigability. An optimal design of a structure-acoustic coupled system is an effective way to reduce noise and vibration. Due to the uncertainties that exist in the structural and acoustical parameters, the traditional deterministic optimization method may be unfeasible when the parameters are subject to fluctuations. This means that when the parameters are uncertain, the results obtained from the deterministic optimization method may be beyond their constraints. This paper proposes to apply the stochastic reliability-based optimization method to the design optimization of the coupled structural-acoustic system with random parameters. A comparison between the results of the stochastic reliability-based method, the safety factor-based method, and the deterministic method show that the first two methods can effectively consider the dispersion of the parameters.

1. INTRODUCTION

Due to the modern industry's rapid development, traffic, city construction, and noise pollution have all attracted attention because they are harmful to structural performance and the general populace's health. Aircraft noise affects both the comfort and work efficacy of the pilot as well as the normal use of instruments inside the aircraft. Uncertainty widely exists in the objective world: it is inevitably subjected to the impact of the uncertainty of a load, structural size, material properties, the influence of various sudden external factors in production, design and use of aircrafts, spacecrafts, etc. These will all have an effect on the working characteristics and normal use of structures and could even lead to failure.

Since the 1970's, some scholars have begun to pay close attention to uncertain structure vibrations and acoustic radiation and have obtained some research results. Shuku and Ishihara¹ investigated the analysis of the acoustic field in irregularly-shaped rooms using the finite element method. Craggs² proposed an acoustic finite element approach for studying boundary flexibility and sound transmission between irregular enclosures. Chen and Chertock^{3,4} computed sound radiation by using the boundary element method. Marburg⁵ studied an optimization problem of the acoustic radiation of finite element beam structures, and analyzed the influence of variables on the objective function where the variables were density, thickness, and young's modulus. Bös⁶ studied the optimization problem of three-dimensional structural acoustic performance. Mullen and Muhanna⁷ considered the static structure problem with uncertain structural loads based on the fuzzy set theory and interval analysis. Zheng-Dong Ma⁸⁻¹¹ studied the sensitivity of response sound pressure, eigenvalue, and eigenvector to structural parameters based on the modal method, the iterative method, and the direct method. Papadopoulos¹² constructed a finite element model of a room sound field and improved

the sound quality of a room by redistributing the low frequency sound modal. Denli¹³ studied the structural vibration and acoustic radiation optimization by optimizing the boundary condition. Christensen^{14,15} studied the coupled structural-acoustic sensitivity analysis and optimization problem.

The Stochastic reliability-based optimization method is a rather classical approach in the field of optimization, but it has never been used for acoustic optimization. Besides, the previous optimization of the coupled structural-acoustic system were still limited to the deterministic method, and did not take the system parameter uncertainties into account. Deterministic structural optimization design often fails to consider the influence on structural performance by the randomness of material parameters, geometric dimensions, and loading. The optimal solution is usually located at the boundary of the constraint condition because if the randomness of the parameters is considered, the optimal solution may be in violation of the constraint condition and lead to an optimization failure.

The contribution of this paper is to overcome the shortcomings of the structural-acoustic deterministic optimization method by using two different methods: the interference theory of stress- intensity¹⁶⁻¹⁸ and the stochastic reliability-based optimization method, which are both applied to the coupled structural-acoustic system with established random parameters.

2. THE FINITE ELEMENT METHOD OF COUPLED STRUCTURAL-ACOUSTIC SYSTEM

2.1. The Finite Element Equation of the Coupled Structural-Acoustic System

The finite element equation of the coupled structural-acoustic system under frequency domain is as follows:

$$-\omega^2\mathbf{M}\mathbf{U} + j\omega\mathbf{C}\mathbf{U} + \mathbf{K}\mathbf{U} = \mathbf{F}; \quad (1)$$

where $\mathbf{M} = \begin{bmatrix} \mathbf{M}^s & \mathbf{0} \\ \mathbf{M}^{couple} & \mathbf{M}^a \end{bmatrix}$; $\mathbf{C} = \begin{bmatrix} \mathbf{C}^s & \mathbf{0} \\ \mathbf{0} & \mathbf{C}^a \end{bmatrix}$; $\mathbf{K} = \begin{bmatrix} \mathbf{K}^s & \mathbf{K}^{couple} \\ \mathbf{0} & \mathbf{K}^a \end{bmatrix}$; $\mathbf{U} = \begin{Bmatrix} \mathbf{U}^s \\ \mathbf{P} \end{Bmatrix}$; $\mathbf{F} = \begin{Bmatrix} \mathbf{F}^s \\ \mathbf{0} \end{Bmatrix}$; $\mathbf{M}^{couple} = -(\mathbf{K}^{couple}) = \mathbf{S}^T$.

\mathbf{M}^{couple} and \mathbf{K}^{couple} represent the coupled stiffness matrix and coupled mass matrix, respectively; \mathbf{K}^s and \mathbf{K}^a represent the total stiffness matrix of structure and sound field, respectively; \mathbf{M}^s and \mathbf{M}^a represent the total mass matrix of structure and sound field, respectively; \mathbf{F}^s is the external forces vector applied to the structure; \mathbf{U}^s is the node displacement amplitude vector of the structure; ω is the excitation natural frequency; \mathbf{C}^s and \mathbf{C}^a represent the damping matrix of structure and sound field respectively; \mathbf{P} is the unknown sound pressure.

2.2. The Finite Element Method for Frequency Response Analysis for the Coupled Structural-Acoustic System with Random Parameters

A parameter vector $\alpha = (a_1, a_2, \dots, a_m)^T$ is used to denote all the physical parameters of the structural-acoustic system, m is the structural parameter number. The matrix and vector of the coupling finite element equation could be expressed as the function of the parameter vector α . Thus, Eq. (1) can be represented in the following form:

$$-\omega^2 \mathbf{M}(\alpha) \mathbf{U}(\alpha) + j\omega \mathbf{C}(\alpha) \mathbf{U}(\alpha) + \mathbf{K}(\alpha) \mathbf{U}(\alpha) = \mathbf{F}(\alpha). \quad (2)$$

It is assumed that the uncertain parameter vector has a normal distribution and the random variables are independent of each other, namely:

$$\alpha \sim N(\alpha^c, \sigma_a). \quad (3)$$

The solution to the coupled structural-acoustic system with random parameters is converted to find all of the solutions that satisfy Eq. (2), namely:

$$\Omega = \{ \mathbf{U}(\alpha) | -\omega^2 \mathbf{M}(\alpha) \mathbf{U}(\alpha) + j\omega \mathbf{C}(\alpha) \mathbf{U}(\alpha) + \mathbf{K}(\alpha) \mathbf{U}(\alpha) = \mathbf{F}(\alpha) \}. \quad (4)$$

3. STOCHASTIC RELIABILITY-BASED OPTIMIZATION METHOD

3.1. Model of Stochastic Reliability-Based Optimization

A class of important problems in structural stochastic reliability-based optimization design minimizes the structural weight by selecting a reasonable distribution of the structural section size in a given reliability. Obviously, this kind of structural design is economical and reliable. Usually, structural weight is expressed as a linear function of the component's cross section dimensions (the design variables). One model of the stochastic reliability-based optimization is as follows:

$$\begin{cases} \text{Find } \mathbf{x}(x_1, x_2, \dots, x_n \in R^n) \\ \min W(\mathbf{x}) = \sum_{j=1}^n w_j(x_j) \\ \text{s.t. } g(\mathbf{x}) = \beta_s(\mathbf{x}) - \beta_s^\alpha \geq 0 \\ \mathbf{x}^l \leq \mathbf{x} \leq \mathbf{x}^u \end{cases}; \quad (5)$$

Among them, \mathbf{x} is the design variable of the structure; W is the mass of the structure; β_s is the index reliability of a constraint; β_s^α is the lower limit value of the reliability constraints.

3.2. The Mean Value First Order and Second Moment (MVFOSM) Reliability Method

The basic idea of the MVFOSM reliability method is to expand the nonlinear limit state functions at midpoints of random variables, ignore the higher order terms rather than second order, and then approximately calculate the mean and standard deviation of the limit state functions. The reliability index can be expressed as functions of the mean and standard deviation.

In stochastic reliability-based optimization of the coupled structural-acoustic system with random parameters, if the response $\mathbf{U}(\alpha, \omega)$ satisfies the constraint that has a normal distribution $\mathbf{U}_\delta \sim N(\mu_\delta, \sigma_\delta)$, the limit state functions about the response amplitude and the constraint can be represented as:

$$\mathbf{Z} = \mathbf{U}_\delta - \mathbf{U}(\alpha, \omega). \quad (6)$$

A Taylor series expansion about the limit state functions Eq. (6) at the midpoints of uncertain parameters α^c is then implemented:

$$\begin{aligned} \mathbf{Z} = \mathbf{U}_\delta - \mathbf{U}(\alpha^c, \omega) + \sum_{i=1}^m \frac{\partial \mathbf{U}(\alpha^c, \omega)}{\partial \alpha_i} \delta \alpha_i \\ + \sum_{i,j=1}^m \frac{\partial^2 \mathbf{U}(\alpha^c, \omega)}{\partial \alpha_i \partial \alpha_j} \delta \alpha_i \delta \alpha_j + \dots \end{aligned} \quad (7)$$

According to the MVFOSM reliability method, the mean μ_z and the variance σ_z^2 of the limit state functions are expressed as follows:

$$\mu_z = \mu_\delta - \mathbf{U}(\alpha^c, \omega) + \sum_{i=1}^m \frac{\partial \mathbf{U}(\alpha^c, \omega)}{\partial \alpha_i} \delta \alpha_i; \quad (8)$$

$$\sigma_z^2 = \sigma_\delta^2 + \sum_{i=1}^m \left[\frac{\partial \mathbf{U}(\alpha^c, \omega)}{\partial \alpha_i} \right]_{\mu_{\alpha^c}}^2 \sigma_{\alpha_i}^2. \quad (9)$$

The central difference method is applied to approximate the derivative of the above matrices or vectors about α :

$$\frac{\partial \mathbf{U}(\alpha^c, \omega)}{\partial \alpha} \approx \frac{\mathbf{U}(\alpha^c + \delta \alpha, \omega) - \mathbf{U}(\alpha^c - \delta \alpha, \omega)}{2\delta \alpha}; \quad (10)$$

$$\delta \alpha = \alpha - \alpha^c. \quad (11)$$

By substituting Eqs. (10) and (11) into Eqs. (8) and (9), the mean and variance of the limit state functions can be obtained.

The reliability index β and failure probability \mathbf{P}_f can be evaluated using Eqs. (12) and (13):

$$\beta = \frac{\mu_z}{\sigma_z}; \quad (12)$$

$$\mathbf{P}_f = \Phi(-\beta). \quad (13)$$

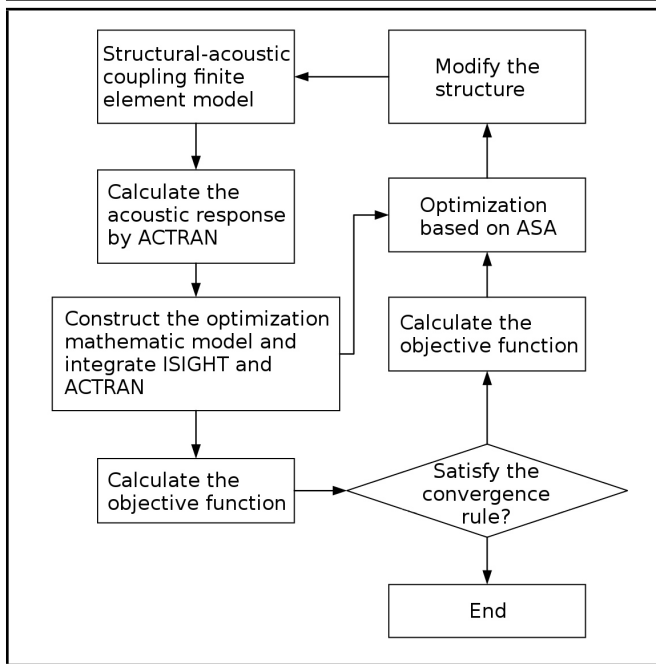


Figure 1. The flow chart of integration optimization for both ISIGHT and ACTRAN.

4. SAFETY FACTOR METHOD BASED ON RELIABILITY

In conventional mechanical design, strength calculation is carried out according to the principle of $s \leq r/n$, where s is the load stress, r is material strength, and n is the safety factor. The safety factor includes uncertain factors such as the difference in calculation methods, the manufacturing allowable deviation, and the difference between actual stress and theoretical stress. As a result, conventional mechanical design is bound to result in a lot of unreasonable designs. The main reason is that $s \leq r/n$ can't reflect the changing law of stress and strength. In fact, even though the component is made from the same material and subjected to the same loading, if the changing law of strength and load is different, the failure probability may be different. In short, conventional design doesn't consider the consequences of failure.

Due to all of these reasons and the lack of any suitable theories or sufficient experimental data, our country's current given safety factor in the specifications and standards are usually too conservative. It may not be economical and may even cause the design to be unreliable. Therefore, in recent years, people have associated the safety factor with stress and strength, and have put it forward based on reliability so it can improve the design's reliability.

The mean safety factor is defined as the ratio of the mean of parts intensity to stress in the dangerous section parts:

$$n = \frac{\mu_\delta}{\mu_s} \tag{14}$$

When the stress and intensity obey normal distribution, the mean safety factor is associated with the reliability of the parts, so we have the equation:

$$\beta = \frac{\mu_\delta - \mu_s}{\sqrt{\sigma_\delta^2 + \sigma_s^2}}; \tag{15}$$

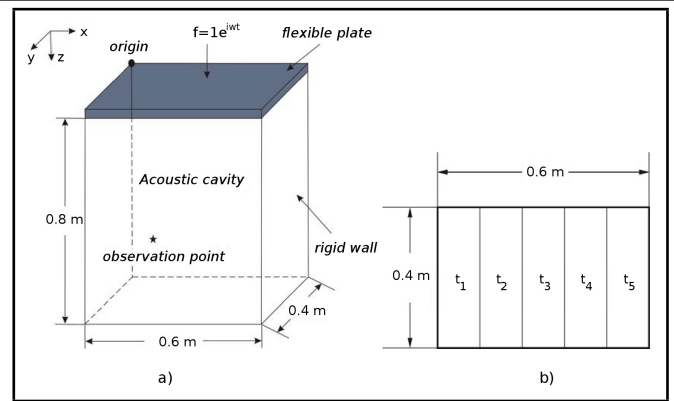


Figure 2. The coupled structural-acoustic system: a) a cuboid cavity and b) the distribution of design variables about plate thickness.

β is the reliability index. According to Eqs. (14) and (15), the mean safety factor n can be rewritten as:

$$n = \frac{\mu_\delta}{\mu_\delta - \beta \sqrt{\sigma_\delta^2 + \sigma_s^2}}. \tag{16}$$

This paper introduces the stress-intensity interference theory to the coupled structural-acoustic system. When a stochastic reliability-based optimization of the structural-acoustic system is implemented, we make the value of the mean and variance of the acoustic response and its constraints correspond to the mean and variance of stress and strength in Eq. (16), respectively. We can obtain the safety factor corresponding to each iterative step of stochastic reliability-based optimization (as can be seen from Fig. 3c). If we implement the safety factor based optimization, we should choose the maximum safety factor obtained from stochastic reliability-based optimization as the safety factor.

5. THE IMPLEMENTATION METHOD OF THE STRUCTURAL-ACOUSTIC OPTIMIZATION

The structural-acoustic finite model and the optimization mathematic model should be established before optimization. And the proper mathematical method should be selected to implement the optimization. In the process of acoustic optimization, each iteration process needs to compute the acoustic response of the system. In this paper, the acoustic responses are obtained by ACTRAN software. The optimization processes are based on the integration of ACTRAN and ISIGHT. Response values calculated by ACTRAN can be transformed into the objective function value through multidisciplinary design language within the ISIGHT. The adaptive simulated annealing algorithm (ASA) is used to find the optimization solution. The integrated block diagram is shown in Fig. 1. ASA is a variant of the simulated annealing (SA) algorithm in which the algorithm parameters that control temperature schedule and random step selection are automatically adjusted according to algorithm progress. This makes the algorithm more efficient and less sensitive to user defined parameters than canonical SA. These are in the standard variant often selected on the basis of experience and experimentation (since optimal values are problem dependent), which represents a significant deficiency in practice. Certainly, the optimization procedure based on deterministic analysis is also able to reduce the value of acoustic responses of structures¹⁹⁻²² significantly.

6. NUMERICAL EXAMPLE

A 3D acoustic cuboid model with length (0.6 m), width (0.4 m) and height (0.8 m) is shown in Fig. 2a. A flexible aluminum plate with the thickness of 6mm is imposed on one surface of the cuboid cavity $z = 0$, which composes a coupled structural-acoustic system. The remaining surfaces are perfectly rigid and the outer environment is a vacuum. The density, elastic modulus, damping coefficient, and Poisson ratio of the aluminum plate are $2700 \text{ kg}\cdot\text{m}^{-3}$, 70 GPa , 0.01 , and 0.3 , respectively. The cuboid cavity is surrounded by air with density $1.225 \text{ kg}\cdot\text{m}^{-3}$ and sound speed $340 \text{ m}\cdot\text{s}^{-1}$.

A harmonic excitation with the amplitude of 1 N is imposed at the central point of the flexible plate along the vertical direction. The frequency step is selected as 2 Hz to analyze the deterministic coupled system in the frequency domain of 1-300 Hz. From the results on field point 0.1, 0.1, and 0.6, we know that the first two characteristic frequencies are located in the values of 108 Hz and 214 Hz, respectively. Here, we select the thicknesses of the sub-block regions in the aluminum plate as design variables $\mathbf{t} = (t_1, \dots, t_5)^T$ as shown in Fig. 2. Supposing that the mean square sound pressure amplitude at ten special frequencies, such as 104 Hz, 106 Hz, 108 Hz, 110 Hz, 112 Hz, 210 Hz, 212 Hz, 214 Hz, 216 Hz, and 218 Hz, MSP_{10} is less than 40 Pa (Stochastic reliability-based optimization requires its reliability not less than 0.99). The objective function is to find the optimal solution to make the total mass as small as possible. The initial value of the design variable is 0.006 m, with a range of 0.003 m to 0.007 m. All of the random variables have a normal distribution and are independent of each other. The stochastic reliability-based optimization design and safety factor-based optimization design (with the aid of AC-TRAN and ISIGHT software) when the variation coefficient of random variables is 0.02, 0.05, and 0.1 respectively, was implemented.

6.1. Deterministic Optimization of the Structural-Acoustic System

The mathematical model of deterministic optimization about this example is:

$$\begin{cases} \min W \\ \text{s.t. } MSP_{10} \leq 40 \text{ Pa} \\ 3 \text{ mm} \leq t_1, t_2, t_3, t_4, t_5 \leq 7 \text{ mm} \end{cases};$$

The deterministic optimization results are show in Table 1 to Table 3.

6.2. Optimization When the Variation Coefficient of Random Variables Is 0.02

1) The stochastic reliability-based optimization method

The mathematical model of stochastic reliability-based optimization in this example is:

$$\begin{cases} \min W \\ \text{s.t. } \beta \geq 2.3263^* \\ 3 \text{ mm} \leq \mathbf{t}_1, \mathbf{t}_2, \mathbf{t}_3, \mathbf{t}_4, \mathbf{t}_5 \leq 7 \text{ mm} \end{cases}$$

*(corresponding to the reliability not less than 0.99); (17)

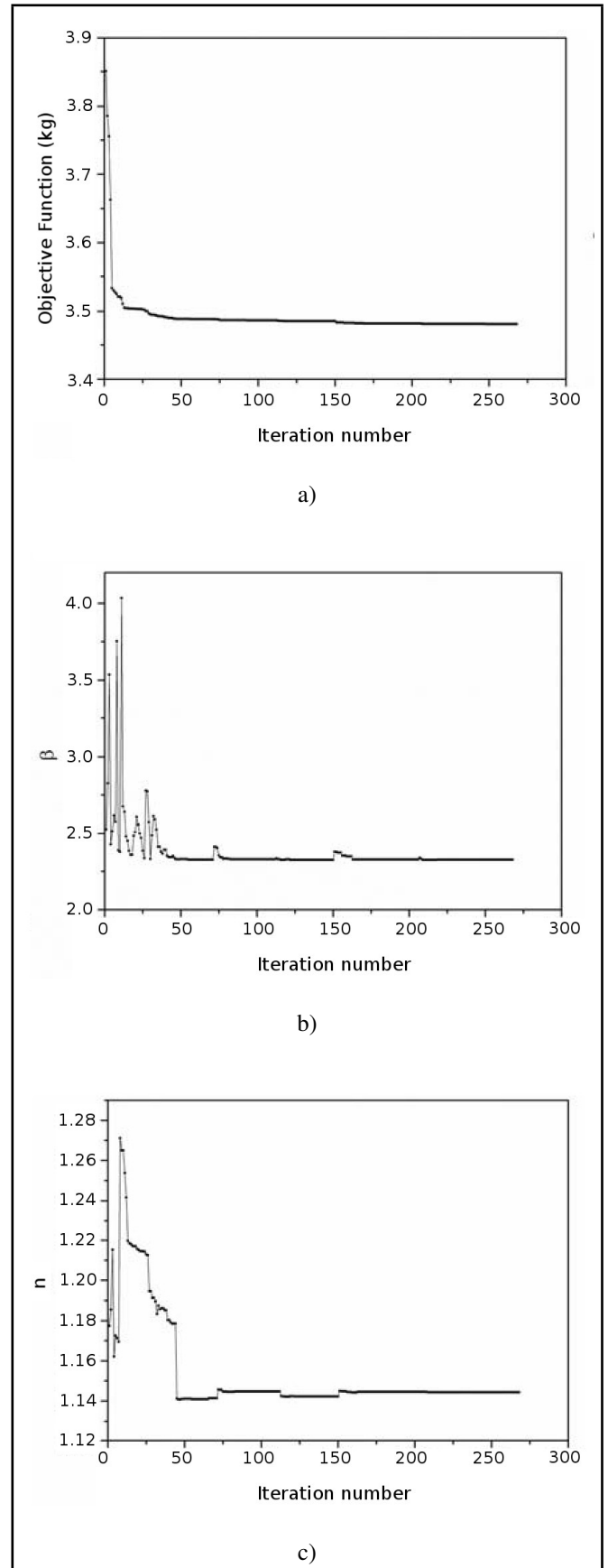


Figure 3. Iterative curves of reliability based optimization (the variation coefficient of random variables is 0.02): a) objective function, b) reliability index, and c) safety factor.

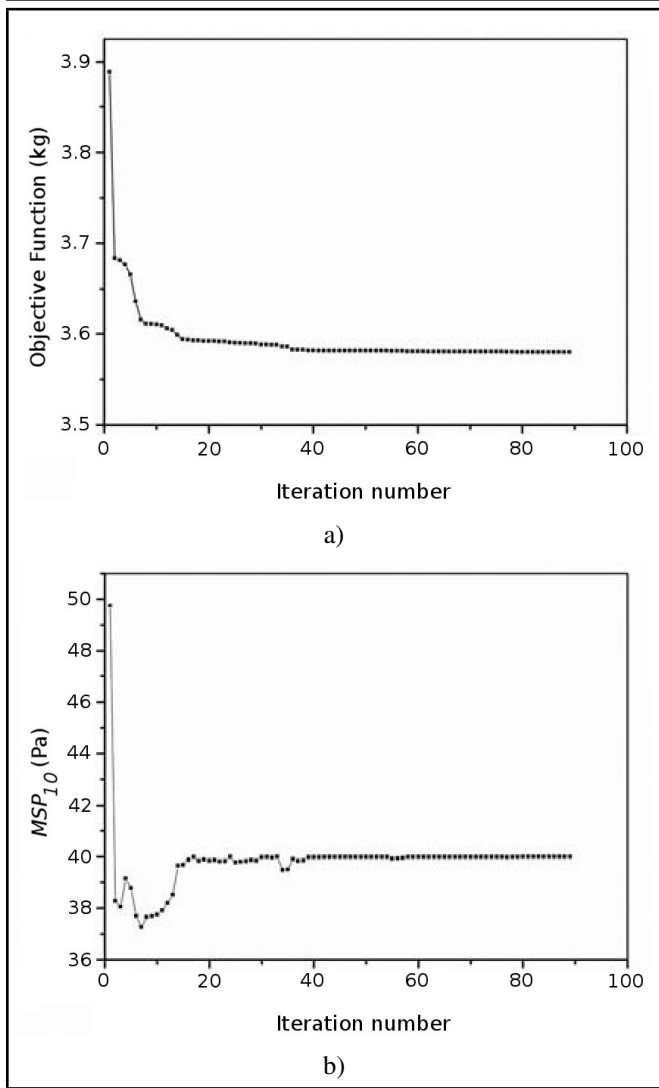


Figure 4. Iterative curves of the safety factor-based optimization ($n = 1.144$): a) objective function and b) MSP.

The iterative curves of the reliability-based optimization method are shown as Fig. 3 and the safety factor n is obtained from reliability index β according to Eq. (16).

2) *The safety factor-based optimization method*

In this example, the mean square sound pressure amplitude and the mean square sound pressure constraint value in stochastic reliability-based optimization corresponds to the stress μ_s and strength μ_δ in the stress-strength interference theory, and σ_δ is 0. According to Eq. (16), we can calculate the safety factor corresponding to the optimal solution $n = 1.144$, which satisfies the requirement of reliability: n is the mean safety factor obtained from the optimal solutions by stochastic reliability-based optimization (reliability is not less than 0.99). The mathematical model of the safety factor-based optimization method is

$$\begin{cases} \min W \\ \text{s.t. } MSP_{10} \leq 40/n \\ 3 \text{ mm} \leq t_1, t_2, t_3, t_4, t_5 \leq 7 \text{ mm} \end{cases} \quad (18)$$

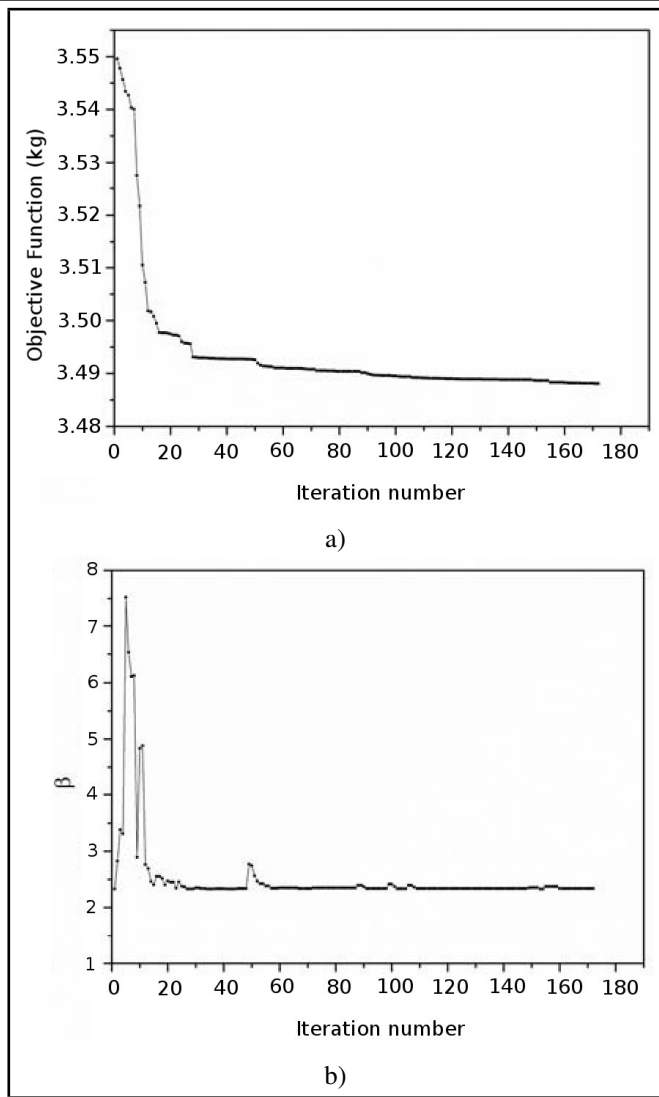


Figure 5. Iterative curves of the reliability-based optimization with random parameters (The variation coefficient of random variables is 0.05): a) objective function and b) reliability index.

6.3. Optimization When the Variation Coefficient of Random Variables Is 0.05

1) *The stochastic reliability-based optimization method*

2) *The safety factor-based optimization method*

According to Eq. (16), we can calculate the safety factor corresponding to the optimal solution and n is 1.085.

7. OPTIMIZATION WHEN THE VARIATION COEFFICIENT OF RANDOM VARIABLES IS 0.1

1) *The stochastic reliability-based optimization method*

2) *The safety factor-based optimization method*

According to Eq. (16), we can calculate the safety factor corresponding to the optimal solution and n is 1.352.

The results of the different optimization methods are shown in Table 1 to Table 3. The objective function (structural

Table 1. Comparison between the stochastic reliability-based optimization method and the safety factor-based method (the variation coefficient of random variables is 0.02).

Optimization type	Design variables (mm)					MSP_{10} (Pa)	Objective function (Kg)
	t_1	t_2	t_3	t_4	t_5		
Initial values	6.0000	6.0000	6.0000	6.0000	6.0000	49.7358	3.8880
The deterministic optimization method	5.1848	5.6531	4.9841	3.0000	3.0004	39.9982	2.8282
The stochastic reliability-based optimization method	3.6495	6.3726	6.5625	5.4085	4.8695	34.9550	3.4814
The safety factor-based method	5.2485	5.2700	6.7173	6.4458	3.2036	39.9989	3.4843

Table 2. Comparison between the stochastic reliability-based optimization method and the safety factor-based method (the variation coefficient of random variables is 0.05).

Optimization type	Design variables (mm)					MSP_{10} (Pa)	Objective function (Kg)
	t_1	t_2	t_3	t_4	t_5		
Initial values	6.0000	6.0000	6.0000	6.0000	6.0000	49.7358	3.8880
The deterministic optimization method	5.1848	5.6531	4.9841	3.0000	3.0004	39.9982	2.8282
The stochastic reliability-based optimization method	6.9406	3.6815	5.8479	6.3075	4.1366	36.8664	3.4881
The safety factor-based method	5.5517	5.0535	6.8061	5.7098	3.5169	39.9996	3.4523

Table 3. Comparison between the stochastic reliability-based optimization method and the safety factor-based method (the variation coefficient of random variables is 0.1).

Optimization type	Design variables (mm)					MSP_{10} (Pa)	Objective function (Kg)
	t_1	t_2	t_3	t_4	t_5		
Initial values	6.0000	6.0000	6.0000	6.0000	6.0000	49.7358	3.8880
The deterministic optimization method	5.1848	5.6531	4.9841	3.0000	3.0004	39.9982	2.8282
The stochastic reliability-based optimization method	5.7675	5.0835	6.8865	6.5635	3.2975	29.5858	3.5768
The safety factor-based method	6.7060	4.5992	6.6061	6.2562	3.3688	39.6467	3.5687

weight) of the safety factor optimization method and stochastic reliability-based optimization method is greater than the deterministic optimization method because the latter does not consider the influence of the random parameters on the response and its constraint value is larger than the other two methods. In addition, as seen on Table 1, the optimal value of the deterministic optimization method of t_4 and t_5 is close to the lower limit of design variables (3 mm). If the uncertainties of the system parameter are considered, the design variables may be beyond the constraint scope (3 mm), which is why the reliability of the deterministic optimization method does not have strong robustness.

8. CONCLUSIONS

Compared with the deterministic optimization design method, the reliability-based method and the safety factor method based on reliability can consider the effect of structural parameter randomness on the structural performance, so the optimization results are more reasonable than that of the deterministic optimization method. In this paper, the safety factor is obtained from the reliability index (mean and stan-

dard deviation) of the mean square acoustic pressure. Using this method to determine the safety factor and to implement the optimization design can help overcome the deficiency of the deterministic optimization method. Compared to the conventional design, the data processing and calculation process is simpler and the calculation precision is higher and closer to the actual. For example, when the variation coefficient of the random parameters is 0.02, 0.05, and 0.1 respectively, the difference of the two methods is about 1%, which shows that the safety factor method can effectively consider the dispersion of parameters. The main advantage in using this method is that it can save materials and ensure reliability.

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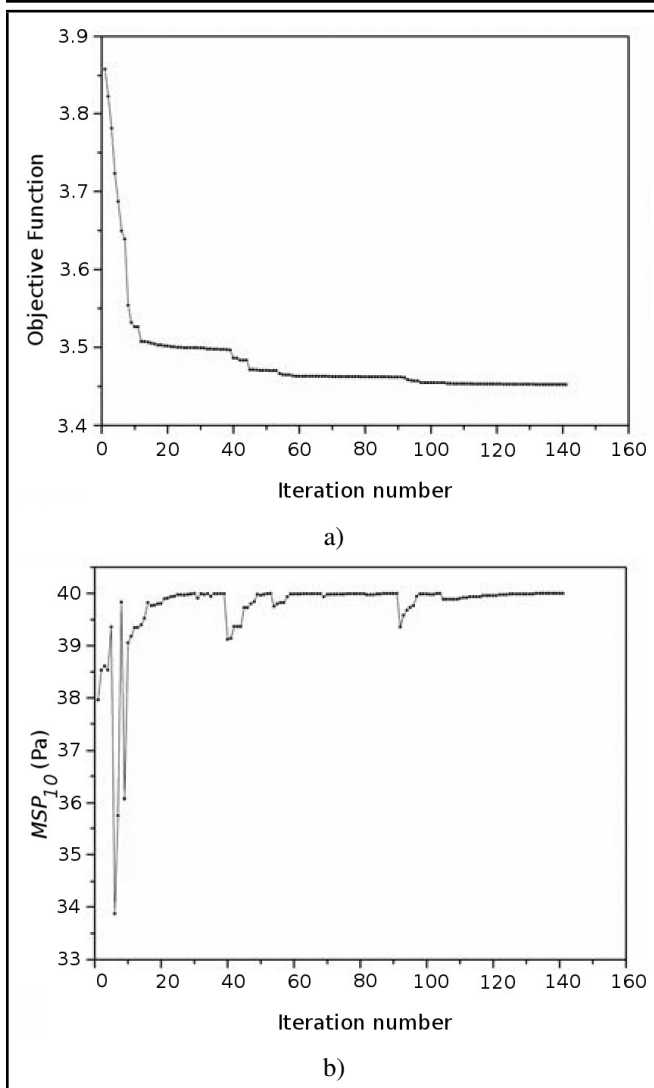


Figure 6. Iterative curves of the traditional safety factor-based optimization ($n=1.085$): a) objective function and b) MSP.

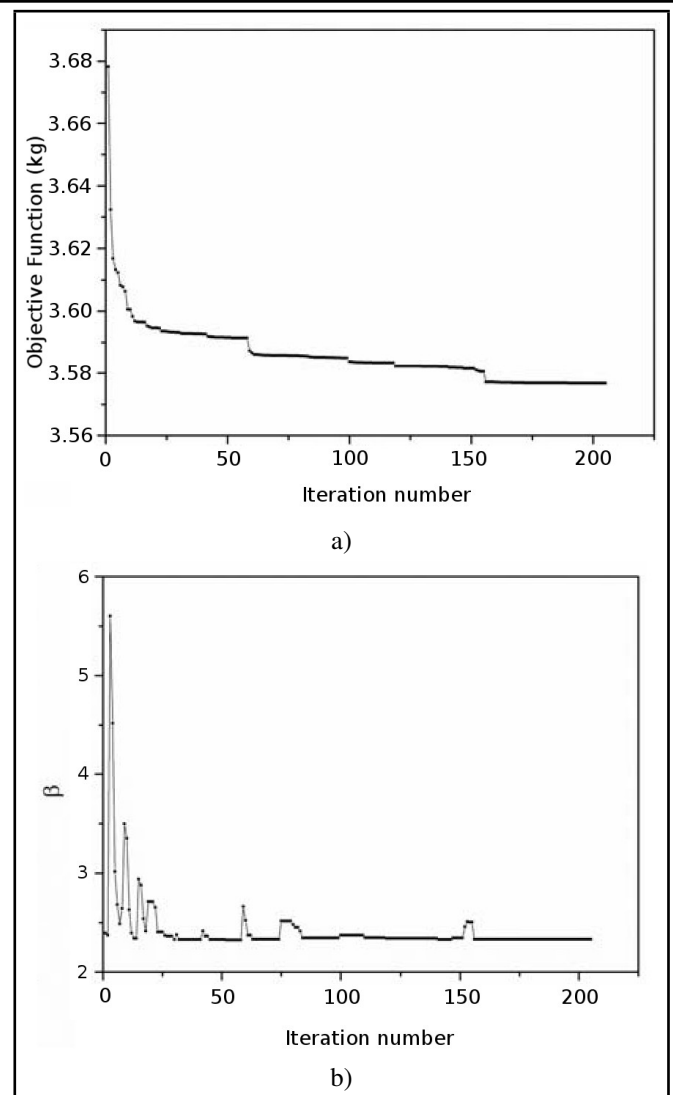


Figure 7. Iterative curves of the reliability-based optimization with random parameters (The variation coefficient of random variables is 0.1): a) objective function and b) reliability index..

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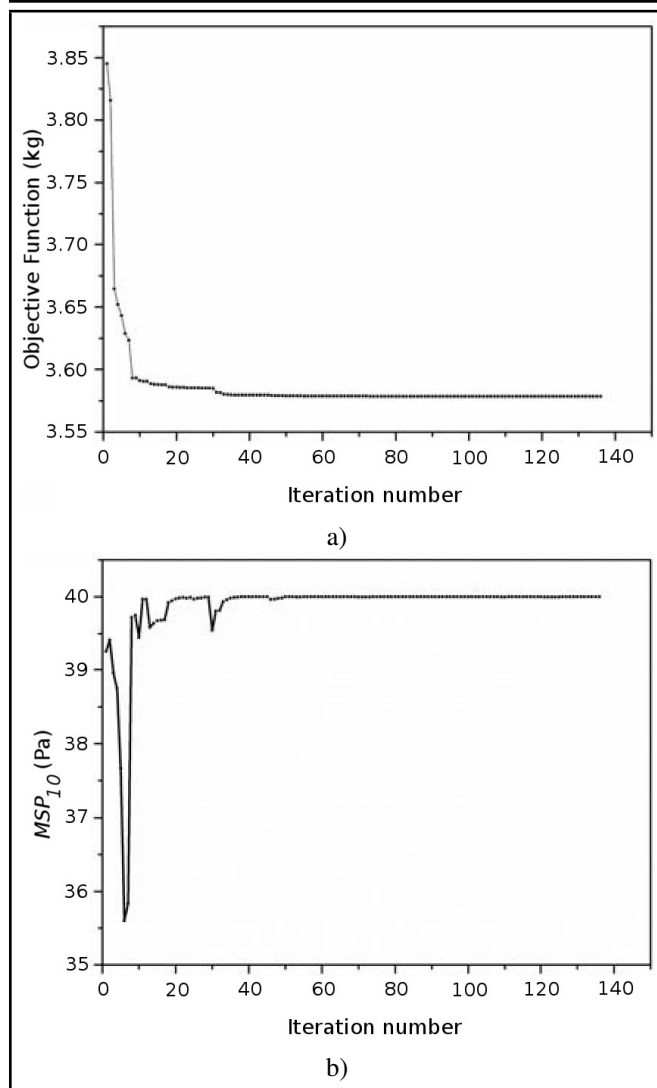


Figure 8. Iterative curves of the safety factor-based optimization ($n = 1.352$): a) objective function and b) MSP.

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