

Article

Multi-Objective Optimization of a Two-Stage Helical Gearbox Using MARCOS Method

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Abstract: In order to address the Multi-Objective Optimization Problem (MOOP) in building a two-stage helical gearbox, this work presents a novel application of the Multi-Criterion Decision-Making (MCDM) method. The aim of the study is to determine the optimal primary design factors that will increase gearbox efficiency while decreasing gearbox volume. Three main design parameters were chosen for assessment in this work: the first stage's gear ratio, and the first and second stages' Coefficients of Wheel Face Width (CWFV). In addition, the MOOP is divided into two phases: phase 1 solves the single-objective optimization problem to reduce the gap between variable levels, and phase 2 solves the MOOP to determine the optimal primary design factors. Furthermore, the Entropy approach was picked to compute the weight criteria, and the MARCOS method was chosen as an MCDM method to handle the multi-objective optimization issue. The following are important characteristics of the study: Firstly, the MCDM method (MARCOS technique) was successfully applied to solve a MOOP for the first time. Secondly, this work has looked into power losses during idle motion to calculate the efficiency of a two-stage helical gearbox. The results of the study were used in the design of a two-stage helical gearbox in order to identify the optimal values for three important design parameters.

Keywords: helical gearbox; multi-objective optimization; gear ratio; gearbox efficiency; gearbox volume; MARCOS method



Citation: Dinh, V.-T.; Tran, H.-D.; Tran, Q.-H.; Vu, D.-B.; Vu, D.; Vu, N.-P.; Nguyen, T.-T. Multi-Objective Optimization of a Two-Stage Helical Gearbox Using MARCOS Method. *Designs* **2024**, *8*, 53. <https://doi.org/10.3390/designs8030053>

Academic Editor: José António Correia

Received: 15 April 2024

Revised: 20 May 2024

Accepted: 29 May 2024

Published: 5 June 2024



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1. Introduction

Industrial applications frequently utilize helical gearboxes because of their low cost, minimal complexity, and ease of design and manufacture. Therefore, many scientists have been interested in finding the best design for helical gearboxes.

Numerous studies on single- and multi-objective optimization for helical gearboxes have been conducted up to this point. The optimization problem for helical gearboxes has been approached for several single-objective functions, including minimum gear mass [1,2], minimum gear volume [2–4], minimum gearbox mass [1,5], minimum gearbox length [6,7], minimum gearbox across section area [8,9], minimum gearbox cost [10–12], and so on. The single-objective helical gearbox optimization problem has been solved by many different methods such as particle swarm optimization [4], Matlab optimization tool box [2], the direct search method [5,13,14] etc. Also, for helical gearboxes with several stages, such as

one [4,14], two [6,9], three [12,15], and four stages [1,16], single-objective optimal design has also been addressed.

With various single objectives, the MOOP has been resolved. The lowest gearbox power loss overall and the lowest gearbox volume were examined in [17]. In order to lower both the gearing mass and the flank adhesive wear speed, the authors in [18] conducted optimization research. In [19], two single targets were selected: the maximum gear stress and the minimum gear mass. The multi-objective optimization solutions in [20] have enhanced the transmission error signal's root mean square values as well as its maximal contact pressures. Implemented in [21] is the multi-objective optimization issue of choosing the best gear material for a helical gearbox to maximize surface fatigue and increase wear resistance. The optimal major design factor for maximizing gearbox efficiency and decreasing gearbox mass was identified in [13]. Besides this, MOOPs have also been solved using a variety of techniques, including the response surface methods [22], the PSO (particle swarm optimization) method [18], the NSGA-II method [17,18], the NSGA-II and the TOPSIS (Technique for Order of Preference by Similarity to Ideal Solution) methods [23], Grey Relation Analysis (GRA) and the Taguchi technique [13], etc. Recently, a genetic algorithm was used in [24] to solve a macro-geometry gear design optimization problem with two objective functions—gear mesh stiffness and dynamic behavior—in order to determine the optimal strategy for achieving a minimal amount of dynamic excitation. A computationally efficient and effective gear design optimization approach is provided based on the obtained results. A new framework for precise reliability analysis has been set out in [25]; it is predicated on employing metaheuristic algorithms to enhance the directional simulation. With the new improved version, the unit vector of direction is determined through the use of metaheuristic methods and expressed as a constrained optimization problem using the Harris Hawks Optimization technique. The suggested technique was compared to the performance of the first-order reliability method and six simulation-based reliability analysis methods. The outcomes demonstrate the enhanced directed simulation's high performance capabilities in resolving extremely nonlinear engineering issues.

While helical gearbox multi-objective optimization has been extensively studied, the MCDM technique has not been used to find the optimal primary design parameters for these gearboxes. Moreover, the research indicated above did not account for the power loss that occurs when a gear is in an idle motion or when a gear is immersed in lubricant during bath lubrication. This paper presents the findings of a multi-objective optimization study conducted for a two-stage helical gearbox, with two specific objectives in mind: reducing gearbox volume and maximizing gearbox efficiency. The study looked at the first stage's gear ratio and both stages' CFWW as the three optimal primary design characteristics for the two-stage helical gearbox. Furthermore, the Entropy approach was utilized to determine the weights of the criteria, and the MARCOS method was chosen to handle the MOOP. One of the main conclusions of the research is the suggestion to apply an MCDM technique to tackle MOOPs in combination with two-step problem solving, tackling single- and multi-objective problems. In addition, the problem's solutions are more effective than those of earlier studies. Moreover, the power losses incurred in idle motion have been added when calculating the efficiency of a two-stage helical gearbox.

2. Optimization Problem

In this part, the gearbox volume and efficiency are first calculated in order to build the optimization problem. Next, the specified objective functions and constraints are given. To facilitate calculations, the nomenclatures used in the optimization problem are presented in Table 1.

Table 1. The nomenclatures for the optimization problem.

Parameters	Nomenclature	Units
Gearbox housing volume	V_{gh}	dm^3
Gearbox width	B_1	dm
Gearbox height	H	dm
Pitch diameter of the pinion of stage 1	d_{w11}	mm
Pitch diameter of the gear of stage 2	d_{w21}	mm
Pitch diameter of the pinion of stage 2	d_{w12}	mm
Pitch diameter of the gear of stage 2	d_{w22}	mm
Center distance of stage 1	a_{w1}	mm
Center distance of stage 2	a_{w2}	mm
Gear ratio of stage 1	u_1	-
Gear ratio of stage 2	u_2	-
Gearbox ratio	u_{gb}	-
Gear width of stage 1	b_{w1}	mm
Gear width of stage 2	b_{w2}	mm
Wheel face width coefficient of stage 1	X_{ba1}	-
Wheel face width coefficient of stage 2	X_{ba2}	-
Material coefficient	k_a	$Mpa^{1/3}$
Allowable contact stress of stages 1	AS_1	Mpa
Allowable contact stress of stages 2	AS_2	Mpa
Contacting load ratio for pitting resistance	$k_{H\beta}$	-
Torque on the pinion of stage i	T_{1i}	Nmm
Output torque	T_{out}	Nmm
Efficiency of a helical gear unit	η_{hg}	-
Efficiency of a rolling bearing pair	η_b	-
Length of shaft i	l_{si}	mm
Diameter of shaft i	d_{si}	mm
Allowable shear stress of shaft material	$[\tau]$	MPa
Total power loss in the gearbox	Pl	
Power loss in the gears	Plg	Kw
Power loss in the bearings	Plb	Kw
Power loss in the seals	Pls	Kw
Power loss in the idle motion	Pzo	Kw
Efficiency of a helical gearbox	η_{hb}	-
Efficiency of the i stage of the gearbox	η_{gi}	-
Friction coefficient	f	-
Friction coefficient of bearing	f_b	-
Arc of approach on i stage	β_{ai}	
Arc of recess on i stage	β_{ri}	
Outside radius of the pinion	R_{e1i}	mm
Outside radius of the gear	R_{e2i}	mm

Table 1. Cont.

Parameters	Nomenclature	Units
Base-circle radius of the pinion	R_{01i}	mm
Base-circle radius of the gear	R_{02i}	mm
Pressure angle	α	rad.
Sliding velocity of gear	v	m/s
Peripheral speed of bearing	v_b	m/s
Load of bearing i	F_i	N
ISO Viscosity Grades number	VG_{40}	
Hydraulic moment of power losses	T_H	Nm

2.1. Calculation of Gearbox Volume

The volume of the gearbox V_{gb} can be calculated by (Figure 1):

$$V_{gb} = L \cdot B_1 \cdot H \tag{1}$$

where L , B_1 , and H are determined by [26]:

$$L = (d_{w11} + d_{w21}/2 + d_{w12}/2 + d_{w22}/2 + 22.5)/0.975 \tag{2}$$

$$H = \max(d_{w21}; d_{w22}) + 8.5 \cdot S_G \tag{3}$$

$$B_1 = b_{w1} + b_{w2} + 6 \cdot S_G \tag{4}$$

$$S_G = 0.005 \cdot L + 4.5 \tag{5}$$

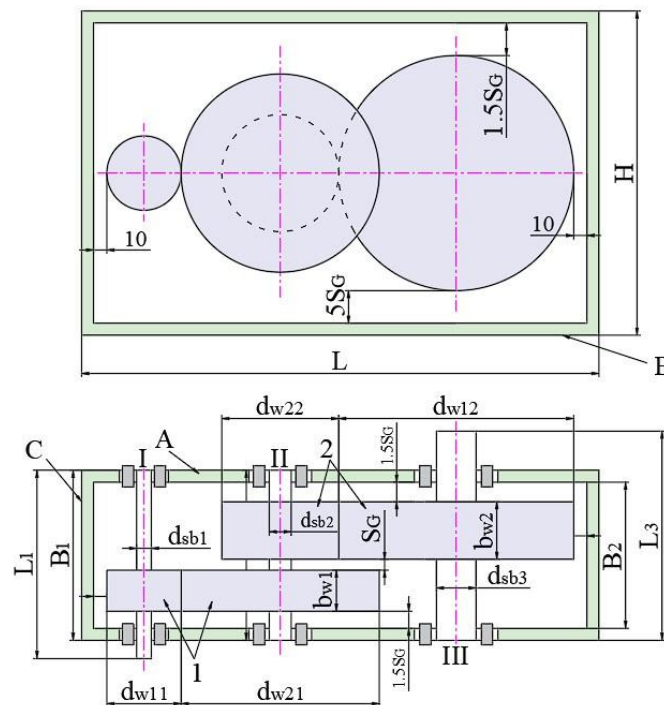


Figure 1. Calculated schema.

In the above equations,

$$d_{w1i} = 2 \cdot a_{wi} / (u_i + 1) \tag{6}$$

$$d_{w2i} = 2 \cdot a_{wi} \cdot u_i / (u_i + 1) \tag{7}$$

$$b_{w1} = X_{ba1} \cdot a_{w1} \tag{8}$$

$$b_{w2} = X_{ba2} \cdot a_{w2} \tag{9}$$

In Equations (6) to (9), a_{wi} is determined by [27]

$$a_{wi} = k_a \cdot (u_i + 1) \cdot \sqrt[3]{T_{1i} \cdot k_{H\beta} / ([AS_i]^2 \cdot u_i \cdot X_{bai})} \tag{10}$$

In which

$$T_{1i} = \frac{T_r}{\prod_{j=i}^3 (u_j \cdot \eta_{hg}^{3-i} \cdot \eta_{be}^{4-i})} \tag{11}$$

2.2. Calculation of Gearbox Efficiency

The efficiency of a two-stage helical gearbox (%) is determined by

$$\eta_{gb} = 100 - \frac{100 \cdot P_l}{P_{in}} \tag{12}$$

In which P_l can be found by [28]

$$P_l = P_{lg} + P_{lb} + P_{ls} + P_{z0} \tag{13}$$

where P_{lg} , P_{lb} , P_{ls} , and P_{z0} are determined in the following way.

Determination of P_{lg} :

$$P_{lg} = \sum_{i=1}^2 P_{lgi} \tag{14}$$

in which

$$P_{lgi} = P_{gi} \cdot (1 - \eta_{gi}) \tag{15}$$

η_{gi} can be calculated by [29]

$$\eta_{gi} = 1 - \left(\frac{1 + 1/u_i}{\beta_{ai} + \beta_{ri}} \right) \cdot \frac{f_i}{2} \cdot (\beta_{ai}^2 + \beta_{ri}^2) \tag{16}$$

while β_{ai} and β_{ri} can be found by [29]

$$\beta_{ai} = \frac{(R_{e2i}^2 - R_{02i}^2)^{1/2} - R_{2i} \cdot \sin\alpha}{R_{01i}} \tag{17}$$

$$\beta_{ri} = \frac{(R_{e1i}^2 - R_{01i}^2)^{1/2} - R_{1i} \cdot \sin\alpha}{R_{01i}} \tag{18}$$

In (16), f is calculated in the following way [13]:

- If $v \leq 0.424$ (m/s),

$$f = -0.0877 \cdot v + 0.0525 \tag{19}$$

- If $v > 0.424$ (m/s),

$$f = 0.0028 \cdot v + 0.0104 \tag{20}$$

Determination of P_{lb} [28]:

$$P_{lb} = \sum_{i=1}^6 f_b \cdot F_i \cdot v_i \tag{21}$$

where $i = 1 \div 6$ and $f_b = 0.0011$ as the radical ball bearings with angular contact were used [28].

Determination of P_s [28]:

$$P_s = \sum_{i=1}^2 P_{si} \tag{22}$$

In which i is the ordinal number of the seal ($i = 1 \div 2$) and P_{si} is calculated by

$$P_{si} = [145 - 1.6 \cdot t_{oil} + 350 \cdot \log \log (VG_{40} + 0.8)] \cdot d_s^2 \cdot n \cdot 10^{-7} \tag{23}$$

Determination of P_{z0} [28]:

$$P_{z0} = \sum_{i=1}^k T_{Hi} \cdot \frac{\pi \cdot n_i}{30} \tag{24}$$

In which k is the total number of gear pairs in the gearbox ($k = 2$); n is the number of revolutions of the driven gear; T_{Hi} can be found by [28]

$$T_{Hi} = C_{Spi} \cdot C_{1i} \cdot e^{\frac{C_{2i} \cdot v}{v_{n0}}} \tag{25}$$

In (25), $C_{Spi} = 1$ for stage 1 under the circumstance wherein the involved oil must pass the mesh, and for stage 2, C_{Spi} is computed using the following equation (Figure 2):

$$C_{Spi} = \left(\frac{4 \cdot e_{max}}{3 \cdot h_c} \right)^{1.5} \cdot \frac{2 \cdot h_{Ci}}{l_{hi}} \tag{26}$$

where l_{hi} is determined by [28]

$$l_{hi} = (1.2 \div 2.0) \cdot d_{a2i} \tag{27}$$

In (25), C_1 and C_2 are calculated by [28]

$$C_{1i} = 0.063 \cdot \left(\frac{e_{1i} + e_{2i}}{e_0} \right) + 0.0128 \cdot \left(\frac{b_i}{b_0} \right) \tag{28}$$

$$C_{2i} = \frac{e_{1i} + e_{2i}}{80 \cdot e_0} + 0.2 \tag{29}$$

In which $e_0 = b_0 = 10$ (mm).

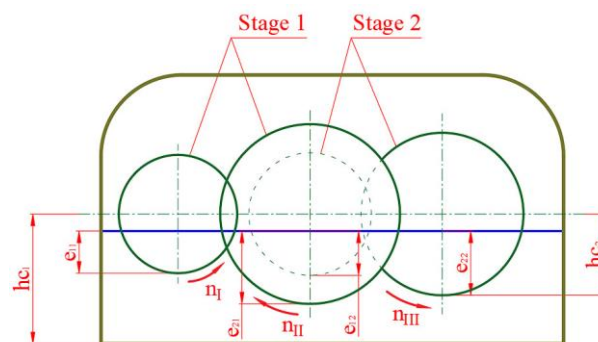


Figure 2. Calculated schema of lubrication factors.

2.3. Multi-Objective Optimization Problem

In this work, the MOOP can be express as follows:

$$\text{miny} = F(X) = (f_1(X), f_2(X))^T \tag{30}$$

Here, two single objectives compose the MOOP:

- Minimizing the gearbox volume—

$$\min f_1(X) = V_{gb} \tag{31}$$

- Maximizing the gearbox efficiency—

$$\min f_2(X) = \eta_{gb} \tag{32}$$

where the vector representing the design variables is denoted by X . There are five primary design parameters for a two-stage helical gearbox: u_1 , X_{ba1} , X_{ba2} , AS_1 , and AS_2 [13]. Furthermore, it was shown that AS_1 and AS_2 's maximum values correspond to their ideal values [13]. Thus, three primary design factors, u_1 , X_{ba1} and X_{ba2} , were chosen as variables for the optimization problem in this work. As a result, we have:

$$X = \{u_1, X_{ba1}, X_{ba2}\} \tag{33}$$

The constraints that follow must be met by the multi-objective function

$$1 \leq u_1 \leq 9 \text{ and } 1 \leq u_2 \leq 9 \tag{34}$$

$$0.25 \leq X_{ba1} \leq 0.4 \text{ and } 0.25 \leq X_{ba2} \leq 0.4 \tag{35}$$

3. Methodology

3.1. Method to Solve the Multi-Objective Optimization

The multi-objective optimization issue with two objectives—minimum gearbox volume and highest gearbox efficiency—was described in Section 2. In addition, three primary design factors have been chosen to be variables in the optimization issue. Table 2 lists these variables along with their minimum and maximum values. In fact, applying an MCDM approach to the MOO (multi-objective optimization) problem is challenging. The reason for this is that there are numerous options or possible solutions available for MOO problems. Each of the parameters in this study have limits, as indicated in Table 2, and the step between variables is 0.02 to ensure parameter accuracy and prevent missing the optimization problem's solution. In this instance, there are $(9 - 1)/0.02 \cdot (0.4 - 0.25)/0.02 \cdot (0.4 - 0.25)/0.02 = 22.500$ (runs) options (or experimental runs) that need to be identified and compared. Because of the wide range of options, it is not feasible to directly handle the OMO problem using the MCDM approach. In this paper, the MCDM problem was solved using the MARCOS method, and the criterion weights were determined using the Entropy approach. A simulation experiment was constructed in order to provide the input data for the MOOP for a two-stage helical gearbox in the MCDM problem. Since this is a simulation experiment, there is no restriction on the number of experiments that can be conducted by utilizing the full factorial design. Because there are three experimental variables (as previously specified) and five levels for each variable, the total number of experiments will be $5^3 = 125$. However, u_1 , which runs from 1 to 9 in Table 2, has the broadest distribution of the three variables mentioned. As a result, even with five levels, there was still a significant disparity between the levels of this variable (in this case, $(9-1)/4 = 2$). To reduce this disparity, save time, and improve the accuracy of the outcomes, a technique for solving multi-objective problems was proposed (Figure 3). This procedure is broken down into two phases: phase 1 factors reduce the gap between levels by solving the single-objective optimization problem, and phase 2 factors find the optimal primary design by solving the MOOP. Additionally, in the process of addressing the multi-objective problem, the MARCOS issue will be rerun using the smaller distance between two levels of variables if the levels of a variable are not sufficiently close to one another (≤ 0.02) (or the best answer is not appropriate for the requirement) (see Figure 3).

Table 2. Input parameters.

Variables	Symbol	Lower Bound	Upper Bound
Gearbox ratio of first stage	u_1	1	9
CWFW of stage 1	X_{ba1}	0.25	0.4
CWFW of stage 2	X_{ba2}	0.25	0.4

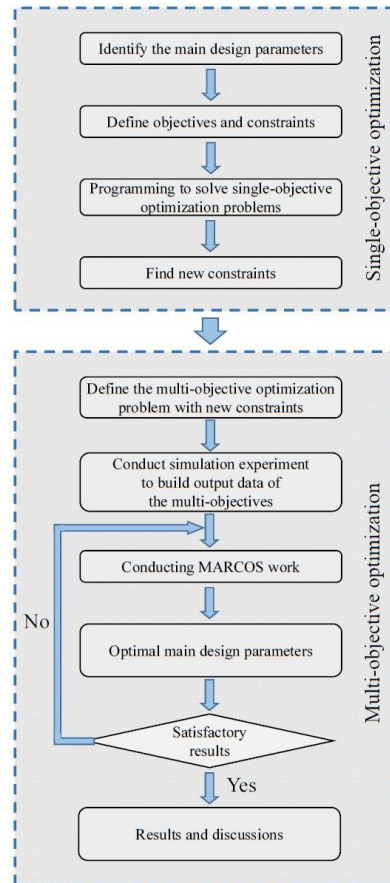


Figure 3. The process for solving the multi-objective problem.

3.2. Method to Solve MCDM Problem

In this work, the MARCOS method was selected for solving the MCDM problem. To apply the MARCOS approach, the following steps must be taken [30]:

- Making initial decision-making matrix

$$X = \begin{bmatrix} x_{11} & \cdots & x_{1n} \\ x_{21} & \cdots & x_{2n} \\ \vdots & \cdots & \vdots \\ x_{mn} & \cdots & x_{mn} \end{bmatrix} \tag{36}$$

where m and n are alternative and criterion numbers;

- An extended initial matrix is produced by appending an ideal (AI) and anti-ideal solution (AAI) to the original decision-making matrix

$$X = \begin{matrix} & AAI & \begin{bmatrix} x_{aa1} & \cdots & x_{aan} \\ x_{11} & \cdots & x_{1n} \\ x_{21} & \cdots & x_{2n} \\ \vdots & \vdots & \vdots \\ x_{m1} & \cdots & x_{mn} \\ AI & \begin{bmatrix} x_{ai1} & \cdots & x_{ain} \end{bmatrix} \end{bmatrix} \end{matrix} \quad (37)$$

Here, $AAI = \min(x_{ij})$ and $AI = \max(x_{ij})$ if the requirement set with criterion j is as large as possible; $AAI = \max(x_{ij})$ and $AI = \min(x_{ij})$ when criterion j is as small as possible; $i = 1, 2, \dots, m; j = 1, 2, \dots, n$;

- One then normalizes the extended starting matrix (X). To calculate the normalized matrix $N = [n_{ij}]_{m \times n}$, we use the following formula:

$$n_{ij} = x_{AI} / x_{ij} \quad (38)$$

$$n_{ij} = x_{ij} / x_{AI} \quad (39)$$

when the criterion j is as small as possible, use Equation (38), and when it is as large as possible, use Equation (39).

- Determine the weighted normalized matrix by

$$C = [c_{ij}]_{m \times n} \quad (40)$$

$$c_{ij} = n_{ij} \cdot w_j \quad (41)$$

where w_j is the weight coefficient of criterion j .

- Find the utility of alternatives K_i^- and K_i^+ by

$$K_i^- = S_i / S_{AAI} \quad (42)$$

$$K_i^+ = S_i / S_{AI} \quad (43)$$

In (42) and (43), S_i ($i = 1, 2, \dots, m$) is the sum of the elements of the weighted matrix C , and S_i is calculated by

$$S_i = \sum_{i=1}^m c_{ij} \quad (44)$$

- Find the utility function $f(K_i)$ of alternatives by

$$f(K_i) = \frac{K_i^+ + K_i^-}{1 + \frac{1-f(K_i^+)}{f(K_i^+)} + \frac{1-f(K_i^-)}{f(K_i^-)}} \quad (45)$$

where the utility function linked to the anti-ideal solution is denoted by $f(K_i^-)$, while the utility function linked to the ideal solution is represented by $f(K_i^+)$. These functions can be found by

$$f(K_i^-) = K_i^+ / (K_i^+ + K_i^i) \quad (46)$$

$$f(K_i^+) = K_i^- / (K_i^+ + K_i^i) \quad (47)$$

- To determine which alternative has the highest utility function value, rank the options according to the final utility function values.

3.3. Method to Find the Weight of Criteria

In this paper, the entropy technique was used to establish the weights of the criteria. The actions listed below can be used to put this strategy into practice [31].

- Finding indicator-normalized values,

$$p_{ij} = \frac{x_{ij}}{m + \sum_{i=1}^m x_{ij}^2} \tag{48}$$

- Calculating the Entropy for each indicator,

$$me_j = -\sum_{i=1}^m [p_{ij} \times \ln(p_{ij})] - \left(1 - \sum_{i=1}^m p_{ij}\right) \times \ln\left(1 - \sum_{i=1}^m p_{ij}\right) \tag{49}$$

- Determining the weight of each indicator,

$$w_j = \frac{1 - me_j}{\sum_{j=1}^m (1 - me_j)} \tag{50}$$

4. Single-Objective Optimization

In this study, the direct search strategy is used to solve the single-objective optimization problem. Moreover, a Matlab computer program has been created to address two single-objective issues: maximizing gearbox efficiency and minimizing gearbox volume. The following figures feature several of the program’s findings: In Figure 3, the connection between u_1 and V_{gb} is shown. When u_1 is at its optimal value, V_{gb} reaches its lowest value (Figure 4). Figure 5 shows the relationship between η_{gb} and u_1 . Furthermore, the optimal value of u_1 at which η_{gb} achieves its maximum is depicted in Figure 5. Figures 6 and 7 show the association of X_{ba1} and X_{ba2} with V_{gb} and η_{gb} , respectively. It can be seen that with an increase in X_{ba1} , V_{gb} will decrease (Figure 6a). In contrast, V_{gb} will fall as X_{ba2} rises (Figure 7a). Additionally, as X_{ba1} and X_{ba2} rise, η_{gb} falls (Figures 6b and 7b).

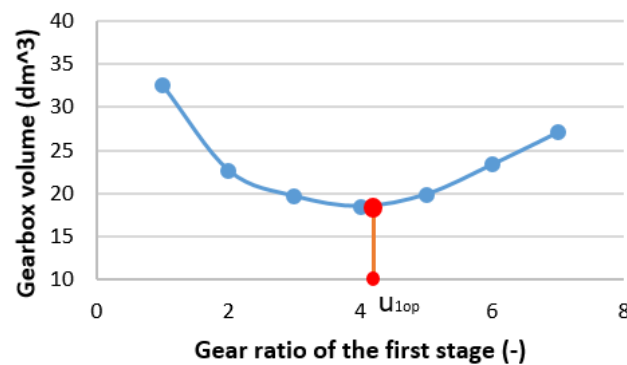


Figure 4. Gearbox volume versus first stage gear ratio.

The values of optimal major design factors of two single-objective functions, V_{gb} and η_{gb} , are shown in Table 3. From the table, it is clear that for the minimum V_{gb} , the minimum values of X_{ba1} and maximum values of X_{ba2} ($X_{ba1} = 0.25$ and $X_{ba2} = 0.4$) are the ideal values for X_{ba1} and X_{ba2} . The reason for this is that the cross-section dimension (determined by $H \cdot L$) must be small in order for V_{gb} to be tiny. Approximately equal values for d_{w21} and d_{w22} are required for this [32]. Given that the second stage gets a significantly greater torque than the first, a_{w2} will be significantly greater than a_{w1} . Consequently, to make d_{w21} approximate, d_{w22} requires an increase in X_{ba2} to decrease a_{w2} and a decrease in X_{ba1} to increase a_{w1} (Formula (9)).

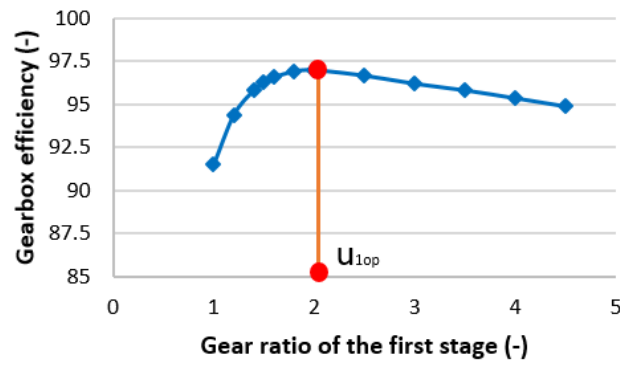


Figure 5. Gearbox efficiency versus first stage gear ratio.

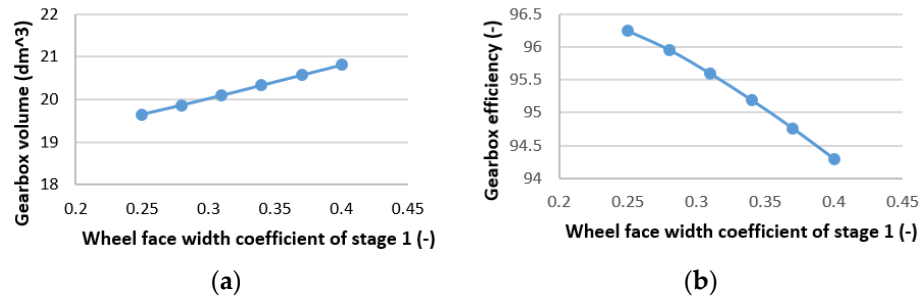


Figure 6. Relation between X_{ba1} and gearbox volume (a) and gearbox efficiency (b).

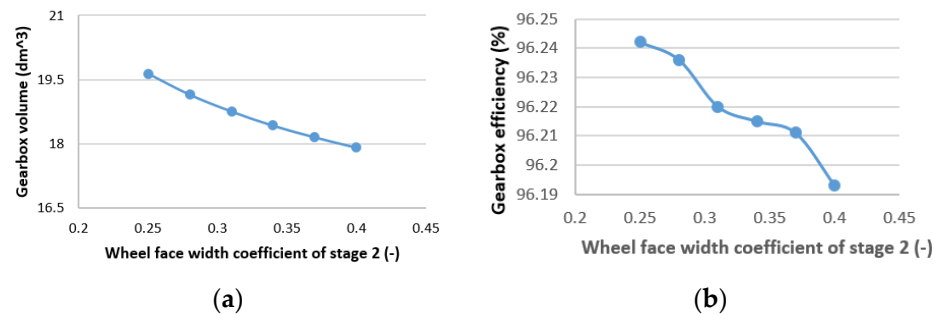


Figure 7. Relation between X_{ba2} and gearbox volume (a) and gearbox efficiency (b).

Table 3. Optimum main design factors for minimum V_{gb} and maximum η_{gb} .

Objective	Factor	u_t					
		10	15	20	25	30	35
V_{gb}	u_1	3.83	5.02	6.09	7.06	7.98	8.84
	X_{ba1}	0.25	0.25	0.25	0.25	0.25	0.25
	X_{ba2}	0.4	0.4	0.4	0.4	0.4	0.4
η_{gb}	u_1	2.49	2.98	3.49	3.98	4.42	4.79
	X_{ba1}	0.25	0.25	0.25	0.25	0.25	0.25
	X_{ba2}	0.25	0.25	0.25	0.25	0.25	0.25

Figure 8 illustrates the relationship between the optimum gear ratio for the first stage (u_1) and the overall gearbox ratio (u_t). Additionally, Table 4 displays newly calculated constraints for the variable u_1 .

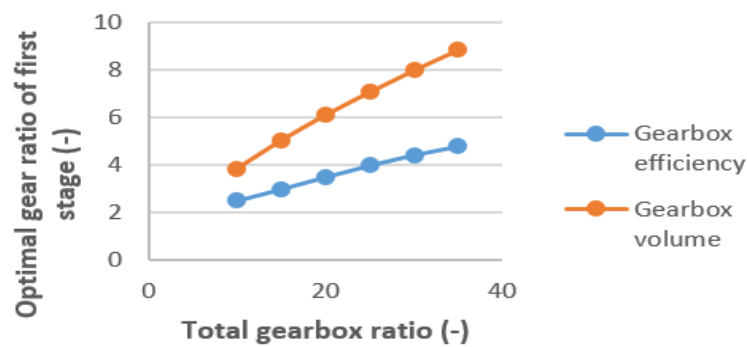


Figure 8. Optimum gear ratio of the first stage versus total gearbox ratio.

Table 4. New constraints of u_1 .

u_t	u_1	
	Lower Limit	Upper Limit
10	2.39	2.93
15	3.08	5.12
20	3.59	6.19
25	4.08	7.16
30	4.52	8.08
35	4.89	8.94

5. Multi-Objective Optimization

A computer program has been created to perform simulations. The gearbox ratios 10, 15, 20, 25, 30, and 35 were all considered for the analysis. For this problem, with $u_t = 15$, the solutions is displayed below. This overall gearbox ratio was used for 125 initial testing runs (as specified in Section 3). The experiment’s output values, the gearbox volume and efficiency, will be sent into MARCOS as input parameters in order to solve the MOOP. This procedure will be repeated until there is less than 0.02 separating two levels of u_1 . The primary design parameters and output responses for $u_t = 15$ in the fifth and final run of the MARCOS experiment are shown in Table 5 The criteria’s weights have been established using the Entropy technique (see Section 3.3), as follows: First, use Equation (48) to derive the normalized values of p_{ij} . Equation (49) is used to determine each indicator m_{ej} ’s entropy value. Finally, use Equation (50) to find the weight of the criteria w_j . The weights of V_{gb} and η_{gb} for the most recent MARCOS work run were determined to be 0.5565 and 0.4435, respectively. The MARCOS method’s multi-objective decision-making phases are outlined in Section 3.2. Specifically, they are as follows: Determine the ideal solution (AI) and the anti-ideal solution (AAI) using formula (36). The results show that, with AI, V_{gb} and η_{gb} were 17.21 (dm^3) and 95.8 (%), and, with AAI, they were 20.1 (dm^3) and 93.34 (%). The next step is to use Formulae (37) for V_{gb} and (39) for η_{gb} to derive the normalized values u_{ij} . The normalized values were then calculated using Formula (40), taking the weight c_{ij} into consideration. Moreover, the coefficients K_i^- and K_i^+ were obtained from Equations (42) and (43). The values of $f(K_i^-)$ and $f(K_i^+)$ were determined using Equations (46) and (47). It was found that $f(K_i^-) = 0.501$ and $f(K_i^+) = 0.499$. Lastly, the values of $f(K_i)$ were computed using Formula (44). Table 6 displays the options’ ranking and the results of several parameters (for the last run of MARCOS work). Out of all the possibilities given, option 105 is the most ideal one, according to the table. Consequently, Table 5 shows the optimal values for the primary design features: $u_1 = 3.69$, $X_{ba1} = 0.25$, and $X_{ba2} = 0.4$.

Table 5. Main design parameters and output results for $u_t = 15$ in the 5th run of MARCOS.

Trial	u_1	X_{ba1}	X_{ba2}	V_{gb} (dm ³)	e_{gb} (%)
1	3.61	0.25	0.25	21.05	95.906
2	3.61	0.25	0.2875	20.41	95.963
3	3.61	0.25	0.325	19.91	95.947
4	3.61	0.25	0.3625	19.52	95.931
5	3.61	0.25	0.4	19.19	95.902
6	3.61	0.2875	0.25	21.35	95.407
...					
25	3.61	0.4	0.4	19.95	93.52
26	3.63	0.25	0.25	21.01	95.888
27	3.63	0.25	0.2875	20.38	95.946
...					
51	3.65	0.25	0.25	20.98	95.852
52	3.65	0.25	0.2875	20.34	95.929
53	3.65	0.25	0.325	19.84	95.913
...					
75	3.67	0.4	0.4	19.86	93.42
76	3.67	0.25	0.25	20.94	95.837
77	3.67	0.25	0.2875	20.31	95.911
...					
104	3.69	0.25	0.3625	19.39	95.852
105	3.69	0.25	0.4	19.07	95.836
106	3.69	0.2875	0.25	21.21	95.306
...					
123	3.69	0.4	0.325	20.74	93.434
124	3.69	0.4	0.3625	20.24	93.405
125	3.69	0.4	0.4	19.82	93.387

Table 6. Several calculated results and rankings of alternatives by MARCOS for $u_t = 1.5$.

Trial	K^-	K^+	$f(K^-)$	$f(K^+)$	$f(K_i)$	Rank
1	0.0084	0.0084	0.4990	0.5010	0.0056	87
2	0.0085	0.0085	0.4990	0.5010	0.0057	58
3	0.0087	0.0086	0.4990	0.5010	0.0058	30
4	0.0088	0.0087	0.4990	0.5010	0.0058	15
5	0.0088	0.0088	0.4990	0.5010	0.0059	5
6	0.0083	0.0083	0.4990	0.5010	0.0055	101
...						
25	0.0085	0.0085	0.4990	0.5010	0.0057	56
26	0.0084	0.0084	0.4990	0.5010	0.0056	84
27	0.0085	0.0085	0.4990	0.5010	0.0057	55
...						
51	0.0084	0.0084	0.4990	0.5010	0.0056	83
52	0.0085	0.0085	0.4990	0.5010	0.0057	52
53	0.0087	0.0086	0.4990	0.5010	0.0058	28
...						
75	0.0085	0.0085	0.4990	0.5010	0.0057	49
76	0.0084	0.0084	0.4990	0.5010	0.0056	82
77	0.0086	0.0085	0.4990	0.5010	0.0057	50
...						
104	0.0088	0.0087	0.4990	0.5010	0.0058	9
105	0.0088	0.0088	0.4990	0.5010	0.0059	1
106	0.0083	0.0083	0.4990	0.5010	0.0055	95
...						
123	0.0083	0.0083	0.4990	0.5010	0.0055	91
124	0.0085	0.0084	0.4990	0.5010	0.0056	69
125	0.0085	0.0085	0.4990	0.5010	0.0057	46

Table 7 shows the optimal values for the main design parameters that correspond to the remaining u_t values of 10, 20, 25, 30, and 35, building on the previous discussion. The information in Table 6 permits the following deductions to be drawn.

Table 7. Optimum main design parameters.

No.	u_t					
	10	15	20	25	30	35
u_1	3.55	3.69	4.18	4.47	5.06	5.39
X_{ba1}	0.25	0.25	0.25	0.25	0.25	0.25
X_{ba2}	0.4	0.4	0.4	0.4	0.4	0.4

Table 7 reveals that X_{ba1} selects the lowest value ($X_{ba1} = 0.25$), while X_{ba2} selects the highest value ($X_{ba2} = 0.4$). This is due to the fact that a small box’s cross-sectional area ($L \times H$) is required to produce the smallest gearbox volume. In order to accomplish that, d_{w21} and d_{w22} must be about equal [32]. Additionally, because the second stage has a higher torque, a larger X_{ba2} is required in order to decrease the diameter of d_{w22} . Conversely, a smaller X_{ba1} must be chosen in order to raise d_{w21} because the first stage has a lower torque.

The obtained values of the gearbox efficiency in this work were compared with the findings in [23] in order to evaluate the effectiveness in employing the formula for the power loss in the idle motion when calculating power loss in gears. Table 5 [23] shows that the gearbox efficiency, based on an input power of 10 kW, a maximum output power of 9.971 kW, and a minimum output power of 9.933 kW, will be between 99.33 and 99.71 percent when the overall gearbox ratio is 7.5. Furthermore, Table 4 indicates that the gearbox efficiency in this work will range from 93.258% to 95.963% when u_t is increased from 10 to 35. Actually, a helical gear train’s efficiency is 0.93–0.98 (93–98%), whereas a pair of bearings’ efficiency is 0.99–0.995 (99–99.5%) [27]. The following formula will be used to calculate the efficiency of a two-stage helical gearbox based on these data:

$$\eta_{gb} = \eta_g^2 \cdot \eta_b^3 = 0.93^2 \cdot 0.99^3 \div 0.98^2 \cdot 0.995^3 = 0.83 \div 0.95 \text{ or } \eta_{gb} = 83 \div 95 \text{ or } (\%) \quad (51)$$

From the above analysis, it is clear that the gearbox efficiency of this work (93.258–95.963%) is reasonably near to reality (83–95%), when comparing with that in [23]. Furthermore, the gearbox efficiency shown in [23] (99.33–99.71%) is dramatically high and incompatible with reality. This demonstrates that the power loss in idle motion formula used in this work is a useful new tool that needs to be used.

Figure 9 shows that there is a definite first-order link between the ideal values of u_1 and u_t . Additionally, it was found that the following regression equation (with $R^2 = 0.9804$) may be used to calculate the optimal values of u_1 :

$$u_1 = 0.0777 \cdot u_t + 2.6414 \quad (52)$$

Once u_1 has been determined, the optimal value of u_2 can be found using the following formula:

$$u_2 = u_t / u_1 \quad (53)$$

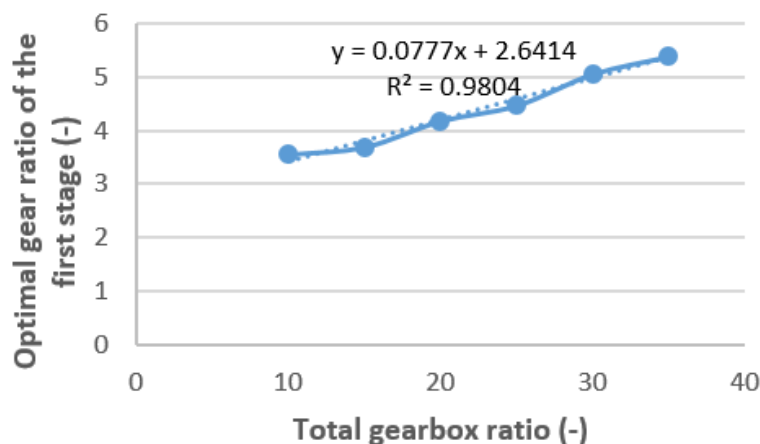


Figure 9. Optimum gear ratio of the first stage versus total gearbox ratio.

6. Conclusions

The MARCOS approach was utilized in this work to solve the MOOP related to the design of a two-stage helical gearbox. The study’s goal is to identify the best critical design factors that maximize gearbox efficiency while reducing gearbox volume. To do this, three essential design components were chosen: the CWFV for the first and second stages, and the first stage gear ratio. In addition, there are two steps in the MOOP solution process. Phase 1 is dedicated to solving the single-objective optimization problem of reducing the difference between variable values, whereas phase 2 is concerned with determining the optimal primary design factors. The following findings were drawn from this work:

- The single-objective optimization problem speeds up and simplifies the resolution of the MOOP by bridging the gap between variable levels;
- The three main design parameters for a two-stage helical gear gearbox, Equation (51) and Table 6, were recommended to have optimal values based on the study’s findings;
- In regard to the important design characteristics, two single objectives—the minimal gearbox volume and the greatest gear-box efficiency—were assessed;
- By using the MARCOS technique repeatedly until the desired results are reached, the MOOP can be solved more precisely (u_1 has an accuracy of less than 0.02);
- The experimental data’s extraordinary degree of concordance with the proposed model of u_1 verifies their reliability;
- Further research is required to determine how to apply the proposed approach for solving the MOOP for various domains and MCDM methods.

Author Contributions: The original idea was proposed by N.-P.V., and it was discussed by all the authors. With the help of T.-T.N., N.-P.V. handled the optimization problem and wrote the manuscript. Every author additionally contributed to the design of the simulation, the analysis of the experimental figures, and the interpretation of the experimental outcomes. N.-P.V. edited the manuscript with the help of T.-T.N. All authors have read and agreed to the published version of the manuscript.

Funding: This study obtained no outside funds.

Data Availability Statement: The data presented in this study are available in this article.

Acknowledgments: The authors would like to thank Thai Nguyen University of Technology for their support in this work.

Conflicts of Interest: The authors declare that they have no conflicts of interest.

References

1. Pi, V.N. Optimal determination of partial transmission ratios for four-step helical gearboxes with first and third step double gear-sets for minimal mass of gears. In Proceedings of the WSEAS International Conference on Applied Computing Conference, Istanbul, Turkey, 27–30 May 2008.

2. Golabi, S.I.; Fesharaki, J.J.; Yazdipoor, M. Gear train optimization based on minimum volume/weight design. *Mech. Mach. Theory* **2014**, *73*, 197–217. [[CrossRef](#)]
3. Rai, P.; Agrawal, A.; Saini, M.L.; Jodder, C.; Barman, A.G. Volume optimization of helical gear with profile shift using real coded genetic algorithm. *Procedia Comput. Sci.* **2018**, *133*, 718–724. [[CrossRef](#)]
4. Tamboli, K.; Patel, S.; George, P.; Sanghvi, R. Optimal design of a heavy duty helical gear pair using particle swarm optimization technique. *Procedia Technol.* **2014**, *14*, 513–519. [[CrossRef](#)]
5. Khai, D.Q.; Linh, N.H.; Danh, T.H.; Tan, T.M.; Cuong, N.M.; Hien, B.T.; Pi, V.N.; Dung, N.T.Q. Calculating Optimum Main Design Factors of a Two-Stage Helical Gearboxes for Minimum Gearbox Mass. In *International Conference on Engineering Research and Applications*; Springer: Berlin/Heidelberg, Germany, 2022.
6. Hung, L.X.; Hong, T.T.; Van Cuong, N.; Ky, L.H.; Thanh Tu, N.; Hong Cam, N.T.; Tuan, N.K.; Pi Vu, N. Calculation of optimum gear ratios of mechanical driven systems using two-stage helical gearbox with first stage double gear sets and chain drive. In *Advances in Engineering Research and Application, Proceedings of the International Conference on Engineering Research and Applications, ICERA 2019, Thai Nguyen, Vietnam, 1–2 December 2019*; Springer: Berlin/Heidelberg, Germany, 2020.
7. Van Cuong, N.; Le Hong, K.; Tran, T. Splitting total gear ratio of two-stage helical reducer with first-stage double gearsets for minimal reducer length. *Int. J. Mech. Prod. Eng. Res. Dev. IJMPERD* **2019**, *9*, 595–608.
8. Pi, V.N.; Hong Came, N.T.; Hong, T.T.; Hung, L.X.; Tung, L.A.; Tuan, N.K.; Tham, H.T. Determination of optimum gear ratios of a two-stage helical gearbox with second stage double gear sets. In *IOP Conference Series: Materials Science and Engineering*; IOP Publishing: Bristol, UK, 2019.
9. Pi, V.N.; Hong, T.T.; Thao, T.T.P.; Tuan, N.K.; Hung, L.X.; Tung, L.A. Calculating optimum gear ratios of a two-stage helical reducer with first stage double gear sets. In *IOP Conference Series: Materials Science and Engineering*; IOP Publishing: Bristol, UK, 2019.
10. Danh, T.H.; Huy, T.Q.; Danh, B.T.; Tan, T.M.; Van Trang, N.; Tung, L.A. Determining Partial Gear Ratios of a Two-Stage Helical Gearbox with First Stage Double Gear Sets for Minimizing Total Gearbox Cost. In *International Conference on Engineering Research and Applications*; Springer: Berlin/Heidelberg, Germany, 2022.
11. Tuan, N.A.; Danh, B.T.; Lam, P.D.; Linh, N.H.; Quang, N.H.; Anh, L.H.; Ngoc, N.D.; Trang, N.V. Determining the Optimum Gear Ratios to Minimize the Cost of Two-Stage Helical Gearbox with Second-stage Double Gear Sets. *J. Mech. Eng. Res. Dev.* **2021**, *44*, 10–20.
12. Vu, N.-P.; Nguyen, D.-N.; Luu, A.-T.; Tran, N.-G.; Tran, T.-H.; Nguyen, V.-C.; Bui, T.-D.; Nguyen, H.-L. The influence of main design parameters on the overall cost of a gearbox. *Appl. Sci.* **2020**, *10*, 2365. [[CrossRef](#)]
13. Le, X.-H.; Vu, N.-P. Multi-objective optimization of a two-stage helical gearbox using taguchi method and grey relational analysis. *Appl. Sci.* **2023**, *13*, 7601. [[CrossRef](#)]
14. Abuid, B.A.; Ameen, Y.M. Procedure for optimum design of a two-stage spur gear system. *JSME Int. J. Ser. C Mech. Syst. Mach. Elem. Manuf.* **2003**, *46*, 1582–1590. [[CrossRef](#)]
15. Pi, V.N.; Tuan, N.K. Optimum determination of partial transmission ratios of three-step helical gearboxes for getting minimum cross section dimension. *J. Environ. Sci. Eng. A* **2016**, *5*, 570–573.
16. Pi, V.N. Optimal calculation of partial transmission ratios for four-step helical gearboxes with first and third step double gear-sets for minimal gearbox length. In *Proceedings of the American Conference on Applied Mathematics*, Cambridge, MA, USA, 24–26 March 2008.
17. Miler, D.; Zvezelj, D.; Loncar, A.; Vuckovic, K. Multi-objective spur gear pair optimization focused on volume and efficiency. *Mech. Mach. Theory* **2018**, *125*, 185–195. [[CrossRef](#)]
18. Tudose, L.; Buiga, O.; Jucan, D. Multi-objective optimization in helical gears design. In *Proceedings of the Fifth International Symposium about Design in Mechanical Engineering-KOD*, Novi Sad, Serbia, 15–16 April 2008.
19. Wang, H.; Chen, D.; Pan, F.; Yu, D. Multi-objective Optimization Design of Helical Gear in Centrifugal Compressor Based on Response Surface Method. In *IOP Conference Series: Materials Science and Engineering*; IOP Publishing: Bristol, UK, 2018.
20. Lagresle, C.; Guingand, M.; de Vaujany, J.-P.; Fulleringer, B. Optimization of tooth modifications for spur and helical gears using an adaptive multi-objective swarm algorithm. *Proc. Inst. Mech. Eng. Part C J. Mech. Eng. Sci.* **2019**, *233*, 7292–7308. [[CrossRef](#)]
21. Terán, C.V.; Martínez-Gómez, J.; Milla, J.C.L. Material selection through of multi-criteria decisions methods applied to a helical gearbox. *Int. J. Math. Oper. Res.* **2020**, *17*, 90–109. [[CrossRef](#)]
22. Korta, J.A.; Mundo, D. Multi-objective micro-geometry optimization of gear tooth supported by response surface methodology. *Mech. Mach. Theory* **2017**, *109*, 278–295. [[CrossRef](#)]
23. Maputi, E.S.; Arora, R. Multi-objective optimization of a 2-stage spur gearbox using NSGA-II and decision-making methods. *J. Braz. Soc. Mech. Sci. Eng.* **2020**, *42*, 477. [[CrossRef](#)]
24. Marafona, J.D.; Carneiro, G.N.; Marques, P.M.; Martins, R.C.; António, C.C.; Seabra, J.H. Gear design optimization: Stiffness versus dynamics. *Mech. Mach. Theory* **2024**, *191*, 105503. [[CrossRef](#)]
25. Meng, D.; Yang, H.; Yang, S.; Zhang, Y.; De Jesus, A.M.; Correia, J.; Fazerer-Ferradosa, T.; Macek, W.; Branco, R.; Zhu, S.-P. Kriging-assisted hybrid reliability design and optimization of offshore wind turbine support structure based on a portfolio allocation strategy. *Ocean Eng.* **2024**, *295*, 116842. [[CrossRef](#)]
26. Römhild, I.; Linke, H. Gezielte Auslegung Von Zahnradgetrieben mit minimaler Masse auf der Basis neuer Berechnungsverfahren. *Konstruktion* **1992**, *44*, 229–236.

27. Chat, T.; Van Uyen, L. *Design and Calculation of Mechanical Transmissions Systems*; Educational Republishing House: Hanoi, Vietnam, 2007; Volume 1.
28. Jelaska, D.T. *Gears and Gear Drives*; John Wiley & Sons: Hoboken, NJ, USA, 2012.
29. Buckingham, E. *Analytical Mechanics of Gears*; Courier Corporation: North Chelmsford, MA, USA, 1988.
30. Stević, Ž.; Pamucar, D.; Puška, A.; Chatterjee, P. Sustainable supplier selection in healthcare industries using a new MCDM method: Measurement of alternatives and ranking according to COmpromise solution (MARCOS). *Comput. Ind. Eng.* **2020**, *140*, 106231. [[CrossRef](#)]
31. Hieu, T.T.; Thao, N.X.; Thuy, L. Application of MOORA and COPRAS models to select materials for mushroom cultivation. *Vietnam J. Agric. Sci.* **2019**, *17*, 32–2331.
32. Kudreavtev, V.N.; Gierzaves, I.A.; Glukharev, E.G. *Design and Calculus of Gearboxes*; Mashinostroenie Publishing: Sankt Petersburg, Russia, 1971. (In Russian)

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