



APPLICATION OF NON-TRADITIONAL METHOD FOR THE OPTIMIZATION OF FRICTION CLUTCH

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ABSTRACT

This paper describes the development of evolutionary algorithm for the optimal design of friction clutch. Powerful evolutionary algorithm such as Genetic Algorithm (GA) is applied for optimum design of friction clutch. The main objective function was to minimize the axial force between clutch plate and pressure plate, using diameter ratio and coefficient of friction as design variable and peripheral velocity as design constraint. The analysis in the present paper is based on understanding the relationship between diameter ratio and the design parameters like inner and outer diameters and peripheral velocity in order to arrive at the optimum design. It is shown that the dimensionless ratio of outer diameter to inner diameter is the most fundamental parameter, controlling all the major designs as well as performance parameters. The optimum design of clutch can be obtained, satisfying the constraint on design, by controlling design variable. The results of GA are compared with conventional optimization technique in order to satisfy the main objective function to minimize the axial force between pressure plate and clutch plate.

Keywords: Friction Clutch, Optimal Design, Genetic Algorithm (GA)

NOTATIONS

Di = Inner Diameter of Friction Surface (mm)
 Do = Outer Diameter of Friction Surface (mm)
 F = Normal Force of Clutch Plate Friction Surface (N)
 k = Dimensionless Ratio of Inner And Outer Diameters of Friction Surface
 n = Number of Friction Surfaces In Clutch
 N = Clutch RPM
 p = Normal Pressure on Friction Surface (N/mm²)
 pa = Maximum Normal Pressure on Friction Lining (N/mm²)
 T = Friction Torque Transmitted by Clutch (N-m)
 v = Tangential Velocity (m/s²)
 μ = Coefficient of Friction
 A = Expected Count
 B = Probability of Selection
 C = Cumulative Probability of Selection
 D = Random Number between 0 And 1

E = String Number
 F = True Count in Meeting Pool
 f(x) = Objective Function
 F(x) = Fitness Function
 F*(x) = Average fitness function
 Ng = Number of Generations
 Ni = Number of Iterations

1. INTRODUCTION

The manually operated clutch is widely used in automotive and industrial applications. Out of two basic approaches available for design of friction clutch, namely the uniform rate of wear and uniform pressure, the former is adopted for analysis of clutches. The difference in the value of friction torque, computed using these two basic approaches, gets smaller and smaller when the ratio of inner diameter to outer diameter of friction surface is increased. For maximum torque capacity of clutch, the ratio of inner diameter to outer diameter of friction surface



should be 0.577. It has been shown that the minimization of the axial force between pressure plate and clutch plate should be the objective of optimization and not the maximization of friction torque because friction torque is a pre-specified parameter in clutch design, not a variable. The real variables entering into the design are axial force, peripheral velocity and outer diameter of friction surface. Out of these the axial force is the most logical parameter to be optimized as it directly affects the design of clutch release linkage as well as ease of operation. Peripheral velocity is the two most important constraints on design. The dimensionless ratio of outer diameter to inner diameter is the most fundamental parameter, controlling all the major designs as well as performance parameters. Friction clutch must be designed for minimum axial force between the pressure plate and clutch plate, while satisfying the constraint on peripheral velocity of outer diameter of friction surfaces. Minimization of the axial force between pressure plate and clutch plate is the objective of optimization using Genetic Algorithm. In order to carry out the optimum design of a friction clutch, it is necessary to understand the dynamics of the problem and the interrelation of the parameters involved in design and analysis. The present work deals with the basic objective to minimize the axial force between clutch plate and pressure plate. The use of non-traditional algorithm with its efficient search and optimization process results in optimal friction clutch design. In this present work, an attempt is made to use genetic algorithm for optimal design process so that we can arrive for the best possible solution. The present study involves the action of GA technique. The output results in this paper work proves that this process of design using non-traditional algorithm is an efficient and effective one.

1.1 Overview Of Present Work :

In this present thesis work, Genetic Algorithm is applied for optimization of single plate dry friction clutch so as to obtain the objective function i.e. minimum axial force between clutch plate and pressure plate. In order to solve this problem using GA, we choose binary coding to represent variables x1 and x2. With 12 bits, solution accuracy is been calculated. Now we have chosen roulette wheel selection, a single point crossover, and a bitwise mutation operator by selecting 24 size populations. The next step is

Evaluation of each string and to find objective function value and fitness function value. Then good string in population is selected to form mating pool. After this process, crossover and mutation phase starts and new population is generated. Then new objective function and fitness function is created and cycle is repeated for number of generation. It is noted that for each generation, there is an improvement in fitness function and average fitness function.

2. FORMULATION OF WORK

2.1 Mathematical Model For Analysis

The analysis of automotive clutch is based on the assumption of uniform rate of wear at any point on the friction surface. The axial force and friction torque are given by equations (1) and (2).

Axial Force, $F = 1/2 \int \mu p a Di (Do - Di)$
 $F = \int \mu p a Di Do (1 - Di/Do)$ ----- (1)

Friction Torque, $T = 1/8 \int \mu p a Di (Do^2 - Di^2)$
 $T = 1/4 \mu F (Do + Di)$
 $T = 1/4 \mu F Do (1 + Di/Do)$ ----- (2)

In order to understand the effect of geometric proportions on the design and performance parameters of the friction clutch, it is useful to introduce a dimensionless variable k, as defined in equation (3), $k = Di/Do$ ----- (3)

The equation (2) for friction torque can be rearranged to obtain the inner diameter of friction lining, Di, as a function of this dimensionless parameter k.

$T = 1/8 \int \mu p a Di^3 (Do^2 / Di^2 - 1)$
 $T = 1/8 \int \mu p a Di^3 (1 / k^2 - 1)$ ----- (4)
 $Di = (8T / \int \mu p a)^{1/3} (k^2 / 1 - k^2)$ ----- (5)

Now, using the expression for Di, the axial force can be expressed in terms of k, as shown below. Put Equation (5) in Equation (1), we get,
 $F = 1/2 \int \mu p a [(8T / \int \mu p a)^{2/3} (k^2 / 1 - k^2)]^{2/3} (1/k - 1)$ ----- (6)

$F = C_1 (k^2 / 1 - k^2)^{2/3} (1 / k - 1)$ ----- (7)
 Here C1 is a constant, depending upon other pre-specified values and material constant.
 Using equations (3) and as $Do = k Di$
 $Do = (8T / \int \mu p a)^{1/3} k (k^2 / 1 - k^2)^{1/3}$
 $Do = C_2 (k^2 / 1 - k^2)^{1/3}$ ----- (8)

Here C2 is a constant, depending upon other pre-specified values and material constants. Now, the peripheral velocity of friction surface at outermost diameter of lining can be written as,
Peripheral Velocity, $V = \int Di Do N / 60$
 $V = \int N / 60 ((8T / \int \mu p a)^{1/3} k (k^2 / 1 - k^2)^{1/3})$
 $V = C_3 k (k^2 / 1 - k^2)^{1/3}$ ----- (9) Here C3 is a

constant, depending upon other pre-specified values and material constants. For design purpose, the parameters like friction torque T and angular speed N of the clutch are input parameters because these are known beforehand.

2.2 Quantitative Analysis

Based upon the theoretical model presented in this paper, the graphical variation of various important variables is plotted, on the common base k. The basic data for a four stroke four-cylinder petrol engine of a small passenger car has been used for quantitative analysis. The basic parameters used for quantitative analysis for optimization of friction clutch is shown in Table No.1

Table 1 Parameters for quantitative analysis

PARAMETER	VALUE
Design torque, N-m	100
Allowable maximum pressure on friction lining, N/mm ²	2X10 ⁵
Coefficient of friction	0.35
Number of friction surfaces	2
Maximum rotational speed, rpm	5500
Peripheral velocity, m/s	50

2.3 Technical Specification of Clutch For Conventional Method

Technical Specification of a Chevrolet Cruze 2000cc, 4 wheeler was taken for calculation. These specifications are shown below.

Table 2 Technical Specification of Clutch

Type	Single Plate Dry disc friction clutch
Outer diameter of friction lining Do(mm)	240
Inner diameter of friction lining Di(mm)	155
Thickness of friction lining t (mm)	8.7
Coefficient of friction surface (μ)	0.35

With the help of technical specification, following parameters were calculated using conventional method.

- Diameter Ratio (k) = 0.645**
- Axial Force (F) = 196.510 x 10³ N**
- Peripheral Velocity (V) = 50m/s**

3. OPTIMIZATION OF FRICTION CLUTCH USING GENETIC ALGORITHM

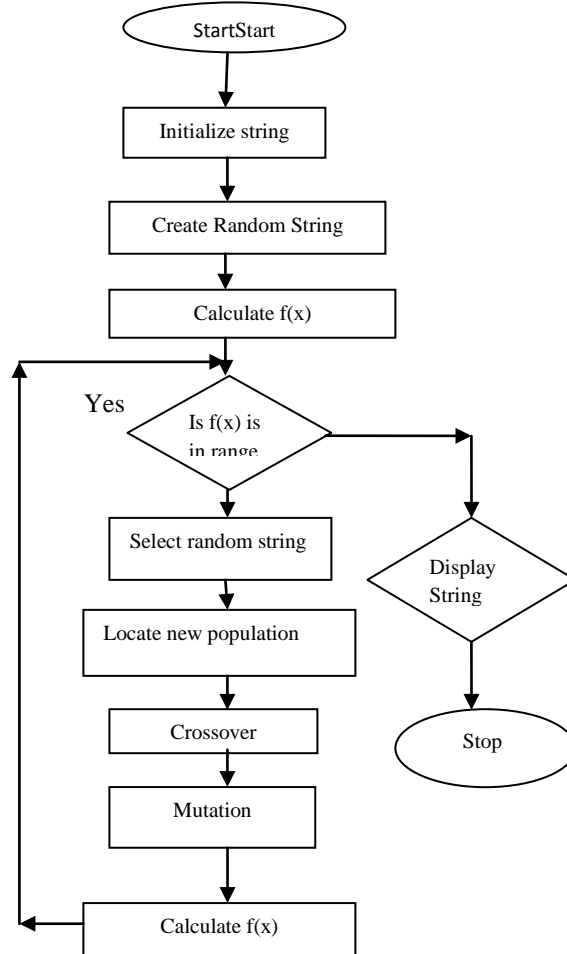


Figure 1 Flow chart representing GA process.

Objective Function: To minimize axial force between pressure plate and clutch plate

Design Variable: Diameter ratio(k), Coefficient of friction (μ). (i) 0.4 < k < 0.9 (ii) 0.1 < μ < 0.5

Design Constraint: Peripheral Velocity(V) As the objective is to minimize the function :- $f(x) = f(x_1, x_2) = 1/2 \left[\left[\frac{8T}{n} \right] x_2 \right]^{2/3} \left(\frac{x_1^2}{1 - x_1^2} \right)^{2/3} \left(\frac{1}{x_1} - 1 \right)$ in the interval 0.4 < x₁ < 0.9, 0.1 < x₂ < 0.5.

3.1 Step 1. Binary Coding To Represent Variable x1 and x2

Chose binary coding, 12 bits for variable x₁, x₂ total string length = 12+12=24

For 12 bits, solution accuracy is (Upper limit - initial interval) / (2^{no. of bit} - 1) Minimum count = 0 and Maximum count = 4096-1=4095



Step 2. 20 Iterations With Evaluation Of Each String In Population Size

String 1 and 2 are coded in order to determine a new objective function (NOF). Average Fitness Function = Sum of F(x) / No. of Population.
Average Fitness Function = $4.8857 \times 10^{-5} / 20$
= 0.0000244

Number population size 12 represents minimum objective function in phase 1 (evaluation phase of genetic programming.)

Step 3. Reproduction Of Each String In Population

Evaluation and Reproduction phases on a random population for population size 12 for string 1 and 2 is shown below

Substring 2:111000011000

Substring 1 :110000010011

(Design variable) $X_1=0.438$, $X_2=0.677$, (NOF)

$f(x)=197.23\text{KN}$, fitness function $f(X)=5 \times 10^{-6}$

Expected count(A)=2.07

Probability of selection (B) =0.103

Cumulative probability of selection (C)=0.625

Random number between 0 and 1(D)=0.388

String number(E)=9

True count in mating pool (F)= 0

For Mating pool ,Substring2:101000011000

Substring 1:110000110001

The action of reproduction operator is clear. Strings are assigned copies exactly equal to the mantissa of the expected count. Thereafter, roulette-wheel selection is implemented which is known as stochastic remainder selection for crossover.

Step 4. Crossover Of String In Population

After Reproduction, the population is enriched with good strings from the previous generations but does not have any new strings. Crossover operator is applied to the population to create better strings and is mainly responsible for the search aspect of GA and gives much of their power.

Step 5. Mutation

In the simple GA, mutation is the occasional random alteration of the value of a string position. This means changing from 0 to 1 or vice versa on a bit wise or whole wise. The need for mutation is to keep diversity in the population. Crossover and Mutation operators with intermediate population for string 1 and string 2 is shown below

Mating pool ,

Substring2:101000011000

Substring1:110000110001

For intermediate population,

Substring2:101000010001

Substring1:110000111000

For Mutation operator, new strings are

Substring2:111000010001

Substring1:110100111000

After applying the GA operators, a new set of population is created. They are decoded then and objective function values are calculated. This completes one generation of GA. Such iterations are continued till the termination criteria are achieved. The above process is simulated by computer program with a population size of 20 for number of generations.

4. RESULT AND DISCUSSION :

4.1 Output Of Genetic Algorithm Programming :

The following genetic algorithm output screenshots programming in C++ shows the optimized values obtained are closer and they are much better than the values obtained by the conventional method.

```

FITTEST POPULATION OF STRINGS In GENERATION 1 are ....
String -> 101000100100 | K = 0.570549
String -> 010010110000 | μ = 0.146520
Min Force = 387433.664863
String -> 110000110001 | K = 0.685934
String -> 101000011000 | μ = 0.315507
Min Force = 212306.273416
String -> 100010001100 | K = 0.480879
String -> 100011001100 | μ = 0.274969
Min Force = 266477.311689
String -> 110000010011 | K = 0.679341
String -> 111000011000 | μ = 0.440537
Min Force = 171024.190408
String -> 100011000100 | K = 0.493187
String -> 100100110010 | μ = 0.287424
Min Force = 257408.850246
String -> 111100001010 | K = 0.846154
String -> 011100100100 | μ = 0.223199
Min Force = 212797.257043
String -> 101001101101 | K = 0.586593
String -> 001111000001 | μ = 0.117338
Min Force = 444695.286107
press any number to Create NEW GENERATION OF STRINGS OR 0 to Terminate ... 1

```

Table 3 Result Of Objective Function Obtained After 160 Iterations

N_g	N_i	k	μ	$F(x)$ KN	$F(x)$	F^*x
1	40	0.677	0.438	197.23	0.507	0.244
2	40	0.6786	0.439	196.56	0.508	0.489
3	40	0.7504	0.409	189.850	0.5267	0.6587
4	40	0.822	0.420	137.17	0.728	0.697

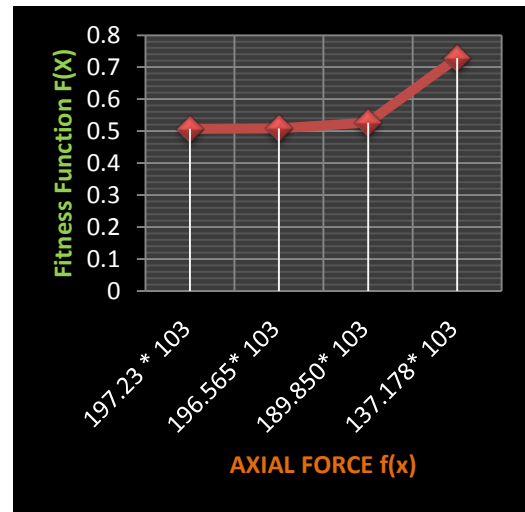


Figure 2 Performance of GA

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String -> 100011000000 | K = 0.492308
String -> 100010010101 | μ = 0.268254
Min Force = 269631.638140
String -> 101000100101 | K = 0.570769
String -> 100001100011 | μ = 0.262149
Min Force = 262848.069822
String -> 100010000110 | K = 0.479560
String -> 101001011101 | μ = 0.323932
Min Force = 239024.300580
String -> 110000001010 | K = 0.677363
String -> 010100100110 | μ = 0.160928
Min Force = 335300.362953
String -> 111100010110 | K = 0.848791
String -> 1001001011001000110000101 | μ = 0.286691
Min Force = 179068.344407
String -> 101001110000 | K = 0.587253
String -> 0100100010001000010010101 | μ = 0.141636
Min Force = 392088.502699
String -> 111000011000 | K = 0.792967
String -> 1001011101010100100001100 | μ = 0.295604
Min Force = 194388.439832
press any number to Create NEW GENERATION OF STRINGS OR 0 to Terminate ...0

PROGRAM TERMINATING .....
press any key
```

From Fig.3 and Fig.4, we can observe that the average fitness function goes on improving as number of generation goes on increasing. The fitness function $F(x)$ and average fitness function $F^*(x)$ will be effectively improved and at some point of generation it will come closer to the value one. Similarly axial force reduces continuously when diameter ratio increases. The rate of decrease is higher for larger values of diameter ratios show below in Fig.8. A lower value of axial force would lead to overall reduction in weight of the clutch operation linkage, lighter actuating springs, light weight actuation mechanism, ease of operation and comfortable conditions for the operator. Therefore, it is desirable to have the smallest possible value of axial force. This is clearly understood that the need for keeping the value of k as high as possible. Hence, we can conclude that average fitness of all the new population has a remarkable improvement as compared to the initial population.

By referring the Table 3, we can come into the conclusion that with four number of generation and 160 iterations, the objective function which is to be minimized has a minimum axial force value of 137.178KN with diameter ratio 0.822, coefficient of friction 0.4202 and a fitness function of 0.7289. With reference to the Fig. 2, we have come into the conclusion that as no of generation goes on increasing, fitness function also improves rapidly. It may also be noted that as the objective function(axial force) is minimizing the fitness function goes on improving. Hence we can conclude that minimization of objective function can be achieved by improving the fitness function by performing number of iterations.

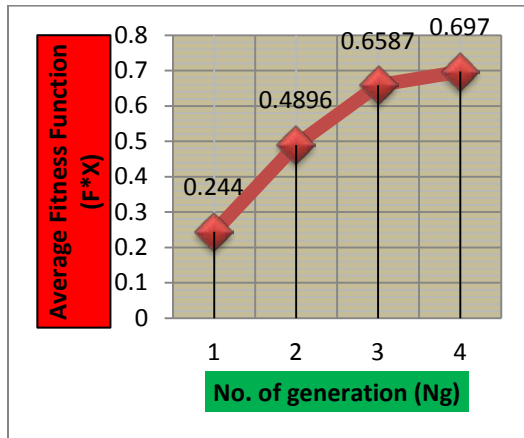


Figure 3 Performance of Average Fitness Function

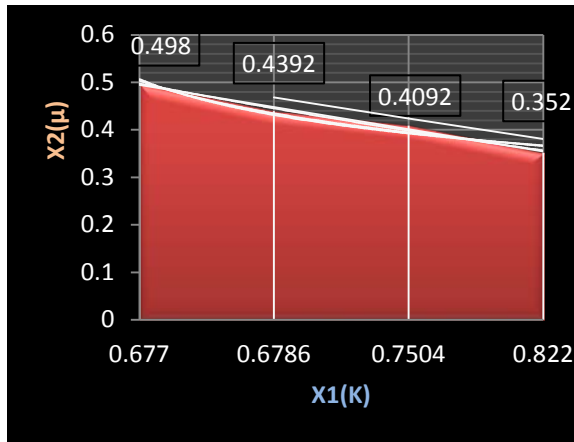


Figure 4 Variation of Axial Force

By referring the Table 4 , we can observe that the improvement in the objective function for minimization of axial force is 18.66 % as well as the design constraint has an improvement of 5.56% when compared with the conventional method.

Table 4 Result Analysis of GA with Conventional Method

Method	X1	X2	f(x) (KN)	V (m/s)
Conventional	0.64	0.35	196.510	50
GA	0.82	0.42	137.178	47.22

5. CONCLUSION :

The salient conclusions drawn from this analysis are :- Optimal design of friction clutch is closely related to the ratio of inner and outer diameters

of friction surface. The axial force and peripheral velocity are most sensitive to the dimensionless ratio of inner and outer diameters of friction surface. Therefore, the optimum value of k is 0.8 which would result into low axial force along with low outer diameter of friction surface and low peripheral velocity. In this study, the friction clutch design optimization problem is solved efficiently using the non-conventional method (genetic algorithm). The results obtained by GA outperformed the conventional method. The efficiency of non-conventional techniques demonstrated in this study suggests its immediate application to other design optimization problems. Also, GA approach handles the constraints more easily and accurately. Genetic algorithms can be applied to a wide range of problems Their flexibility comes at the cost of more function evaluations. There are many variants of genetic algorithms available, each with its own strengths and weaknesses.

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