

# Blending of Genetic Algorithm with Fletcher Reeves Method to Solve Reactive Power Problem

K. Lenin, B.R. Reddy, and M.S. Kalavathi

**Abstract**—In this study a hybrid algorithm - Fletcher Reeves method and advanced Genetic Algorithm (GA) are suggested to solve reactive power problem. In this approach, each of the G Fletcher Reeves method again with progressive operators are calculated step length. These approaches are extended to a set of multi-point access instead of single point approximation to avoid the convergence of the available method at local optimum and a new method, named Population Based Fletcher Reeves Method (PFR), are proposed to solve the reactive power problem. PFR was tested in standard IEEE 30 bus test system and simulation results demonstrate obviously about the best performance of the recommended algorithm in reducing the real power loss with control variables within the limits.

**Index Terms**—Hybrid Algorithm, Fletcher Reeves method, Genetic Algorithm, Bound Constrained Optimization problem, Global-optima, optimal reactive power, Transmission loss.

## I. INTRODUCTION

TO till date various methodologies has been applied to solve the electrical reactive power problems. The key aspect of solving the reactive power problem is to reduce the real power loss with control variables are within the limits. Previously many type of mathematical methodologies like linear programming, gradient method [1-8] has been utilized to solve the electrical reactive power problem, but they lack in handling the constraints to reach a global optimization solution. In the Next level various types of Evolutionary algorithms [9-20] has been applied to solve the reactive power problem. But every algorithm has some merits and demerits. If one algorithm good in exploration but it lack in exploitation and another algorithm good in exploitation although it lack in exploration. Also some algorithm has poor speed in convergence. In the proposed method the step length of the Fletcher-Reeves method in each iteration is evaluated by GA. The above proposed concept is used to set initial points to overcome the problem of premature convergence. Proposed Population Based Fletcher Reeves Method (PFR) was tested in standard IEEE 30 bus test system and simulation study indicate the best performance of the proposed algorithm.

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## I. OBJECTIVE FUNCTION

### A. Active power loss

The objective of the reactive power dispatch problem (RPDP) is to minimize the active power loss (APL) and can be defined in equations as follows:

$$F = PL = \sum_{k \in \text{Nbr}} g_k (V_i^2 + V_j^2 - 2V_i V_j \cos \theta_{ij}) \quad (1)$$

Where F- objective function, PL – power loss,  $g_k$  - conductance of branch,  $V_i$  and  $V_j$  are voltages at buses  $i, j$ , Nbr- total number of transmission lines in electric power systems.

### B. Voltage profile improvement

To minimize the voltage deviation in PQ buses, the objective function can be written as:

$$F = PL + \omega_v \times VD \quad (2)$$

Where, VD - voltage deviation,  $\omega_v$ - is a weighting factor of voltage deviation.

And the Voltage deviation given by:

$$VD = \sum_{i=1}^{N_{pq}} |V_i - 1| \quad (3)$$

### C. Equality Constraint

The equality constraint of the problem is indicated by the power balance equation are given below:

$$P_G = P_D + P_L \quad (4)$$

Where the total power generation  $P_G$  has to cover the total power demand  $P_D$  and the power losses  $P_L$ .

### D. Inequality Constraints

The inequality constraint implies the limits on components in the power system in addition to the limits created to make sure system security. Upper and lower bounds on the active power of slack bus ( $P_g$ ), and electrical reactive power of generators ( $Q_g$ ) are written as follows:

$$P_{g\text{slack}}^{\min} \leq P_{g\text{slack}} \leq P_{g\text{slack}}^{\max} \quad (5)$$

$$Q_{gi}^{\min} \leq Q_{gi} \leq Q_{gi}^{\max}, i \in N_g \quad (6)$$

Higher and lower bounds on the bus voltage magnitudes:

$$V_i^{\min} \leq V_i \leq V_i^{\max}, i \in N \quad (7)$$

Higher and lower bounds on the transformers tap ratios:

$$T_i^{\min} \leq T_i \leq T_i^{\max}, i \in N_T \quad (8)$$

Higher and lower bounds on the compensators it can be expressed by the following equation.

$$Q_c^{\min} \leq Q_c \leq Q_c^{\max}, i \in N_C \quad (9)$$

Where N is the total number of buses, NT is the total number of Transformers; Nc is the total number of shunt reactive compensators.

## II. FLETCHER-REEVES METHOD

The well-known Fletcher –Reeves method is steepest descent method due to Cauchy is one of the oldest for solving unconstrained minimization problem [21]. In Fletcher- Reeves method, the key task is to find the optimal step length for getting the next better approximations of the decision variables in each iteration. Nearly all the scholars around the world utilized this approach in various applications .Here we are going to blend Genetic algorithm with Fletcher-Reeves to solve the reactive power problem.

## III. GENETIC ALGORITHM

Genetic algorithms (GA), the most widely used unique method used in the solution of many problems. To unravel an optimization problem through GA, it is very obligation to plan a suitable chromosome representation of solution. There are dissimilar types [22.23] of acting between which binary and real coding representations are common. In binary coding demonstration each changeable is characterized as binary substrings with ideal precision. In this instance the string length of an isolated will be huge and GA would execute In the following sections. In real coding exemplification all chromosome vectors are encoded as a vector of floating point number of same length as the solution vector. This category of illustration is very elementary to handle and is proficient of representing very quiet large domains. In this exemplification a vector  $(x_1, x_2, \dots, x_n)$  is used as a single to represent a solution of the optimization problem. In the subsequent step is to initialize the chromosomes which will take part in the artificial genetic operations like natural genetics. In this way population size of chromosomes are formed in which each element is initially selected arbitrarily within the desired domain. Amongst many processes for selection of an arbitrary number, here we have used the uniform distribution method.

## IV. POPULATION BASED FLETCHER-REEVES METHODOLOGY FOR SOLVING REACTIVE POWER PROBLEM

In this paper, a new methodology population based Fletcher Reeves method (PFR) by extending the inkling of single-point exploration to a multi-point exploration. The multiple approximations produce a series of paths among which at least one converges to the global optimum. In this technique of the study, all the chromosomes is upgraded by Fletcher Reeves

method whereas the step length is calculated by GA.

*Algorithm*

Step-1: Set  $k = 0$

Step -2: produce an initial population  $x^{(k)}$ , by generating each component  $x_i^{(k)}$  ( $i = 1, 2, \dots, p\_size$ ). (P\_size denotes population size)

Step -3: Compute the function values  $f(x_i^{(k)})$  for all  $i$

Step -4: Find the best value of  $f$  from all  $f(x_i^{(k)})$  come along with  $x^{(k)}$  and keep it in  $f_{old}^{(k)}$  and  $x_{old}^{(k)}$

Step-5: Increase the value of  $k$  by unity i.e.,  $k = k+1$

Step-6: Set  $i = 1$ ,

Step-7: Find the search direction  $d_i^{(k-1)} = -\nabla f(x_i^{(k-1)})$

Step-8: Find the best found value of step length  $\lambda$  and store this value in  $\lambda_i^{(k-1)}$

Step-9: Compute  $x_i^{(k)} = x_i^{(k-1)} + \lambda_i^{(k-1)} d_i^{(k-1)}$

Step-10: Compute  $d_i^{(k)} = -g_i^{(k)} + \beta_i^{(k-1)} d_i^{(k-1)}$ , Where;

$$\beta_i^{(k-1)} = \frac{\langle g_i^{(k)}, g_i^{(k)} \rangle}{\langle g_i^{(k-1)}, g_i^{(k-1)} \rangle} \text{ and } g_i^{(k)} = \nabla f(x_i^{(k)})$$

Step -11: Compute the best found value of step length  $\lambda_i^{(k)}$

Step-12: Improve the solution  $x_i^{(k+1)} = x_i^{(k)} + \lambda_i^{(k)} d_i^{(k)}$

Step -13: Compute  $f(x_i^{(k+1)})$

Step -14: Increase the value of  $i$  by unity i.e.,  $i = i+1$

Step -15: *if*  $i < p\_size$ , *then go to step 7*

Step-16: Find the best value of  $f$  from all  $f(x_i^{(k)})$  along with  $x_i^{(k)}$  and store it in  $f_{new}^{(k)}$  and  $x_{new}^{(k)}$

Step-17: If the termination criterion is satisfied, go to step-19. Else, go to step-18,

Step -18: *if*  $f_{new}^{(k)} < f_{old}^{(k)}$ , *assign*  $f_{old}^{(k)} = f_{new}^{(k)}$  *and*  $x_{old}^{(k)} = x_{new}^{(k)}$  *and then go to step -5*

Step-19: Print the result and stop the process.

## V. SIMULATION RESULTS

Validity of PFR algorithm has been verified by testing in IEEE 30-bus system, 41 branch system and it has 6 generator-bus voltage magnitudes, 4 transformer-tap settings, and 2 bus shunt reactive compensators. Bus 1 is taken as slack bus and 2, 5, 8,

11 and 13 are considered as PV generator buses and others are PQ load buses. Variables limits of the control are shows in Table I.

TABLE I  
PRIME VARIABLE LIMITS (PU)

List of Variables	Min.	Max.	Type
Generator Bus	0.90	1.11	Continuous
Load Bus	0.91	1.01	Continuous
Transformer-Tap	0.92	1.01	Discrete
Shunt Reactive Compensator	-0.10	0.30	Discrete

In Table II the power limits of generators buses are listed.

TABLE II  
GENERATORS POWER LIMITS

Bus	Pg	Pg <sub>min</sub>	Pg <sub>max</sub>	Qg <sub>min</sub>
1	96.00	49	200	-19
2	79.00	18	79	-19
5	49.00	14	49	-11
8	21.00	11	31	-14
11	21.00	11	28	-12
13	21.00	11	39	-14

TABLE III  
AFTER OPTIMIZATION VALUES OF CONTROL VARIABLES

Control Variables	PFR
V1	1.0501
V2	1.0408
V5	1.0206
V8	1.0305
V11	1.0702
V13	1.0507
T4,12	0.0000
T6,9	0.0100
T6,10	0.9000
T28,27	0.9100
Q10	0.1000
Q24	0.1000
Real power loss	4.2901
Voltage deviation	0.9091

Table III shows the proposed PFR approach successfully kept the control variables within limits.

Table IV list out the overall comparison of the results of optimal solution obtained by various methods.

TABLE IV  
COMPARISON OF RESULTS

Techniques	Real power loss (MW)
SGA (24)	4.9800
PSO (25)	4.9262
LP (26)	5.9880
EP (26)	4.9630
CGA (26)	4.9800
AGA (26)	4.9260
CLPSO(26)	4.7208
HSA (27)	4.7624
BB-BC (28)	4.6900
PFR	4.2901

## VI. CONCLUSION

In this paper, Population Based Fletcher Reeves method is efficaciously applied in order to solve Optimal RPDP. The projected PFR algorithm is tested in the standard IEEE 30 bus system operators. Simulation results show the strength of projected PFR methodology for providing improved optimal solution in diminishing the real power loss. Variables of the control obtained from after the optimization via PFR is within the limits.

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