INVESTIGATION OF VOLTAGE STABILITY LIMIT VALUES IN ELECTRICAL POWER SYSTEMS USING GENETIC ALGORITHM

Ali Öztürk

e-mail:ozturk_a@ibu.edu.tr Duzce University Technical Education Faculty Electric Education Dept. Pakize Erdoğmuş e-mail: erdogmus_p@ibu.edu.tr Duzce University Technical Education Faculty Electronic and Comp. Edu. Dept. M. Ali Yalçın e-mail: yalcin@sakarya.edu.tr Sakarya University Engineering Faculty Electric-Electronic Engineering Dept.

Key words: Power Systems; Genetic Algorithms; Voltage Stability; Optimization; Critical Values

ABSTRACT

In this study, initially, critical values of voltage stability have been found by a method in which determinant of jacobian matrix, which is obtained by using the load flowing analysis, is zero. Then the same values have been found with genetic algorithm which is a modern optimization method. Accordingly, the results have been evaluated and observed that the stability critical values are found much more easily and reliable with genetic algorithm.

I. INTRODUCTION

The highest active power values that could be transferred to load bus and the voltage amplitude and phase values at this situation are determined by the critical values in electrical power systems. Voltage stability critical values problem is one of the major concerns in power system planning and secure operation[1,2]. In the case of the voltage stability in long distance energy transferring power systems, the critical values calculation have been generally made on Jacabian matrix obtained with load flowing [3-6]. The singular point of a system is searched by means of nonlinear equations constituting the system dynamics and with the Jacabian matrix in which partial differential is used[7,8]. However, it is possible that either any singular point could not be found or there would be more than one singular point. Consequently, solution may be partial or wrong. On the other hand, Genetic Algortihm(GA) doesn't search for the singular point of the function. It obtains the results by exploring the all possible solutions in the solution space. Therefore it is not possible to find wrong solution. Several GA applications related to electrical power systems have been realized[9]. It has been seen that most of these studies have been focused economic dispatch and reactive power optimization[10].

In this study, the theory of GA was shortly presented and how critical values are attained with GA are explained step by step. The critical limit values in the six bus sample power system have been found with the equations obtained with the possible singular Jacobian matrix [11-13] and GA. The solutions are compared to each other.

II. GENETIC ALGORITHM

GA is an optimization method based on the genetic concept. It is a strategy for solving the multi-variable optimization problems which are considered to be difficult by conventional optimization methods[14]. GA starts to run with a lot of possible solution according to the initial population which are randomly prepared. Then it tries to find optimum solutions by using genetic operators such as selection, crossover and mutation[15]. GA doesn't start the solution with one initial point. It starts to search with a lot of initial point called initial population. So the best solutions are selected and worst are eliminated. GA starts the search with the generations of the initial population depending on the represented fitness function (FF) variables. Initial population are generated randomly after coding the variables. Each row of the population is called an individual. Fitness function (FF) values are calculated for each individual. The fitness function (FF), is the difference between the goal function (GF) and the penalty function(PF) consisting constraints functions (CF). After operated elitism, selection, crossover and mutation, a new population is generated according to the fitness function values. With the evaluation of the previous population, the new population is generated till the number of generation Fitness function values are calculated in each new population. The best resulted ones are paid attention among these values. Until the stopping criteria are obtained, these process is repeated iteratively. The stopping criteria may be the running time of the algorithm, the number of generation and for fitness functions to give the same best possible values in a specified time. In this study generation size has been used as the stopping criteria.

III. DEFINING PROBLEM

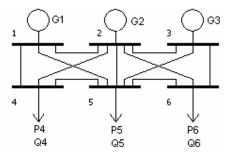


Figure1. An examples of six bus power system

Table 1. Line data of six bus system

Line Number	R (b)	X (b)	B (b)
12	0.1	0.2	0.04
14	0.05	0.092	0.04
15	0.08	0.2	0.06
23	0.05	0.25	0.06
24	0.05	0.1	0.02
25	0.1	0.3	0.04
26	0.07	0.2	0.05
35	0.12	0.26	0.05
36	0.02	0.1	0.02
45	0.2	0.4	0.08
56	0.1	0.3	0.06

	PRODUCTION		VOLTAGE
BUS N.	P(MW)	Q(MVAr)	V (b)
1	200	50	105
2	150	37.5	105
3	180	45	107

BUS	ACTIVE POWER LIMITS			
N.	P _{max} (MW)	P _{min} (MW)	Q _{max} (MVAr)	Q _{min} (MVAr)
1	0	0	100	-100
2	50	0	100	-100
3	60	0	60	-100

Table 4. Load data of six bus system

	BUS	LOAD VALUES		POWER
BUS N.	TYPE	P(MW)	Q(MVAr)	ANGLE
4	PQ	70	70	45
5	PQ	70	70	45
6	PQ	70	70	45

Admittance and impedance values: R, X, B of the six bus exemplary power system of which the stability critical values will found is shown in Table 1. These values have been given according to base values being $V_{base} = 220 \text{ KV}$ and $S_{base} = 100 \text{ MVA}$. The data of generators connected to buses numbered: 1, 2 and 3 have been shown Table 2 and Table 3. Also the data of loads connected to buses numbered: 4, 5 and 6 have been shown in Table 4 [16].

Fitness Function

The active and reactive power equations (P_r , Q_r) shown in equations 1 and 2 are obtained according to exemplary system shown in Figure 1[11].

$$P_r = \frac{V_s * V_r(b_1 * \cos\delta + b_2 * \sin\delta) (a_1b_1 + a_2b_2) * |V_r|^2}{b_1^2 + b_2^2}$$
(1)

$$Q_{r} = \frac{V_{s} * V_{r} (b_{l} * \sin\delta + b_{2} * \cos\delta) (a_{l}b_{2} + a_{2}b_{l}) * |V_{r}|^{2}}{b_{l}^{2} + b_{2}^{2}}$$
(2)

In equation 1 and 2, V_s is production bus voltage, V_r is load bus voltage, δ is load bus angle, a_1 and a_2 are real and imaginary parts of constant A for long transmission line, b_1 and b_2 are real and imaginary parts of constant B for known long transmission line. A and B constant of transmission line are given as $A = a_1 + j^* a_2$ and $B = b_1 + j^* b_2$ respectively.

Since it has been searched the maximum power values with GA, equation 1 belong to P_r , has been accepted as goal function.

GA runs as an unconstraint optimization method. If there are constraints, constraints are subtracted from the fitness function as penalty function so the problem has been converted to unconstrained optimization problem. With this way fitness function values are constrained. In this study, since solutions are evaluated for constant power values, power angle ϕ , and so $\tan\phi=Q_r/P_r$ must be constant. This situation has been taken as constraint. Under light of these explanations, CF will be as in equation 3:

$$CF = Q_r - \tan \varphi * P_r \tag{3}$$

And PF will be as in equation 4:

$$PF = r^* (CF)^2 \tag{4}$$

While GA is running, CF value decreases and lastly takes zero value. In equation 4, r is taken as a proper coefficient and in order to attain much more feasible result it has been taken the square of CF [17].

In this study since the aim is maximization, the FF is the subtraction CF from GF. And also, a proper K value is added to fitness function so that the fitness function doesn't take the negative value. When GA stops, this K constant is subtracted from the result. And the real value are calculated by this way. According to these definitions FF is as the equation 5:

$$FF = K + GF - PF \tag{5}$$

The biggest possible range of V_r and $\delta\,$ are as follows.

$$88 \leq V_r \leq 220 \tag{6}$$

And

$$-1.5 \leq \delta \leq +1.5 \tag{7}$$

Respectively.

Coding of variables

It is necessary to code the GA variables. In this study coding are made by binary system (0,1). The number of the bit belonging to the variable are stated equation 8. In that equation, X_{upper} and X_{lower} respectively shows the upper and lower bounds of variable, and ε shows the increase range [17]. The number of bits for coding the variables are given in Table 5.

$$2^{\ell i} \geq \frac{X_{upper} - X_{lower}}{\varepsilon} + 1 \tag{8}$$

Table 5 Bit numbers of variables to be coded

Variable Range	Inc	Bit Number
88< V _r < 220 KV	0.1	11
$-1.5 < \delta < +1.5$ Radian	0.01	11

Individuals are formed by gathering variables which are randomly coded. The formation of an individual has been given in Table 6. The cumulative bit number of an individual (ℓ) is obtained as 22. And this shows that it can be created 2^{22} different individual. GA starts the solution with the initial population created randomly. The individual number in the initial population is called the population size (PS). PS specifies how many searching number will be in a generation. PS are specified as 42

according to equation 9 [9]. The randomly created initial population is given in Table 7. Each time While GA is started to run, it has been created different initial population.

$$PS = 1.65 * 2^{0.21*\ell}$$
(9)

Table 6 An individual code

Variables		
V _r δ		
11101010011	00110010100	
Individual		
11101010011 00110010100		

Table 7.	Initial Po	opulation
----------	------------	-----------

Ind. Num.	Vr	δ
1	10011100011	10101100111
2	111011100	11110011111
•	•	
•	•	•
•	•	•
42	1110001110	11100001110

GA Operators

In GA, in order to create the next generation population some operators such as elitism, selection and crossover have been used. With the elitism operator, two individuals which have the best fitness functions values are guaranteed to be included in the next generation. The first two individuals of each new population are the best two individuals of the previous generation[18]. Selection is process for finding the much more proper parents generating the new generations. In this study, since the PS is 42, 20 pairs as parents will be selected with the selection operator and 40 new individual will be generated by these parents. And two individual will be selected by elitism. In this study tournament selection has been used. According to this method, randomly a group of individuals are selected from the population and the one of them, which has the best fitness function value, is selected to be one of parent and the others are returned to the population. The second member of parent and all parents are selected by the same strategy [19]. Crossover operator is the operation to generate the candidate individuals for the possible next generation with gene crossover. In this study uniform crossover method has been used. In this method, a mask individual which has the same bit number of an individual are created for each parent pair. First candidate individual is created in such a way that if the mask individual bit is 1, the bit of first pair is copied and if the mask individual bit is 0, the bit of second pair is copied. Second candidate individual is created in such a way that if the mask individual bit is 1, the bit of second pair is copied and if the mask individual bit is 0, the bit of first pair is copied[20]. Some bits of candidate individuals are randomly changed so that they do not become copy of their parent. This process is called mutation. After mutation candidate individuals are now individuals of new population. The mutation rate is given as follows:

$$\frac{1}{PS} < MR < \frac{1}{\ell} \tag{10}$$

In this study, mutation rate (MR) has been taken 0.04 according to Equation 10 [21]. The processes of generating new individuals by : selection, crossing and mutation have been shown in Table 8 and Table 9.

Table 8. Uniform Cross Over

Mask Ind.	1110101001111100010001
Parent 1	1001110110011001110001
Parent 2	1100100011110110010111
Candidate Parent1 (Child I)	1000100010011010010111
Candidate Parent2 (Child II)	1101110111110101110001

Candidate Ind. 1 before mutation					
1100101011011101000110					
Candidate Ind. 1 after mutation					
1000101111011101000110					

After completing the all individuals of each new generation, the fitness function values are calculated. This process is repeated for the generation number. In this study generation size(number) has been selected as 500. When running GA at least 20 times, it has been observed that the optimum solution has been obtained from the half of the generation size and no new optimum solution has been found.

IV. RESULTS AND DISCUSSION

As an examples, critical values of the buses numbered: 4, 5 and 6 in the system with six bus have been found by GA and conventional methods and they are shown in Table 10. The variation of FF, while searching the critical values of bus 4, is shown in Figure 3. Moreover the values of variables for bus four shown in Figure 4. The constant K has been selected to be $6*10^6$ to prevent that the fitness function may take the negative values.

Table 10 Stability critical values for with six bus power system.

N	Critical Values Found with Formulas			Critical Values Found with GA		
	P _{rkrtk} (MW)	V _{rkrtk} (KV)	δ _{rkrtk} (Rad.)	P _{rkrtk} (MW)	V _{rkrtk} (KV)	δ_{rkrtk} (Rad.)
4	198,46	128,27	0,27	199,00	125,84	0,26
5	167,19	128,57	0,28	168,91	129,95	0.25
6	146,13	144,29	0.24	146,10	145,36	0,28

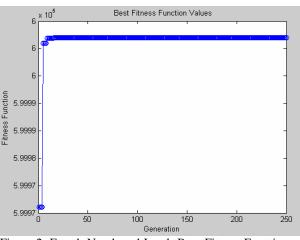


Figure 2. Fourth Numbered Load Bus Fitness Functions Values

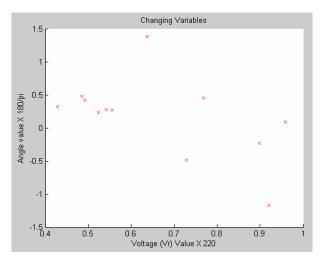


Figure 3. The changing of the variables, when exploring the best fitness function with GA.

In this study, it has been aimed to accurately find the critical values in electrical power systems. As can be seen in the results, the critical values have been obtained the more reliable and accurately. It also has been showed that GA has the capacity to find such critical values.

REFERENCES

[1] US-Canada Power System Outage Task Force, "Final report on the August 14, 2003, black in the United States and Canada : Causes and Recommendations", April **2004**,

[2]: UCTE "Final rapport of the Investigation Committee on the 28 September 2003 Black in Italy ", April **2004**

[3] P.,A. Löf G.Andersson, D.,J., Hill, "Voltage Stability Indices for Stressed Power Systems", Vol. 8 No 1 pp., 326-335, **1993**

[4] P.,A. Löf G.Andersson, D.,J., Hill, "Voltage Dependent Reactive Power Limits For Voltage Stability", IEEE Trans. Power Systems, Vol. 10, No1 pp. 220-228, 1995

[5] P., Kundur "Power System Stability and Control", McGraw-Hill, Inc. ISBN 0-07-5958-X

[6] C. W. ,Taylor "Power System Voltage Stability" McGraw-Hill Inc. ISBN 0-07-113708-4

[7] C.A. Caflizares F.L., Alvarado, "Point of Collapse and continuation Methods for Large AC/DC systems", *IEEE Trans. Power Systems*, vol. 8, no.1, 1993

[8] C.,A., Aumuller, T.,K., Saha., "Determination of Power System Coherent Bus Groups by Novel Sensitivity-Based Method for Voltage Stability Assessment", *IEEE Transactions on Power Systems*, vol.18, no.3, 2003

[9] C.,W., Liu, C.,S., Chang, J.,A., Jiang, "Genetic Algorithms as a Reactive Power Source Dispatching Aid for Voltage Security Enchancement", Proc. Natl. Sci. Counc. ROC (A). Vol. 25, No 11, pp 53-62, 2001

[10] M., Younes, M., Rahli, L.,A., "Koridak, Economic Power Dispatch Using Evolutionary Algorithm", Journal of Electrical engineering **Vol.57**, No 4, pp 211-217,**2006**

[11]. M.,A., Yalçın., "Enerji İletim Sistemlerinde Gerilim Kararlılığının Yeni Bir Yaklaşım ile İncelenmesi", PhD Thesis, *I.T.U.*, *F.B.E.*, **1995**,

[12] C.,S., Indulkar, B., Vistwanathan., S.,S., Venkata S.,S., "Maximum Power Transfer Limited By Voltage Stability in Series and Shunt Compensated Schemes For A.C.Transmission Systems", *IEEE Trans. on Power Delivery*, Vol.4,No. 2, 1989

[13] M.,A., Yalçın., N.,H., Yükseler, "A New Method in Determining Critical Values for The Large Scale Power Systems Voltage Stability", *Melecon 98 9th Mediterrnean Electrotechnical Conference*, Vol.2, ,Israel , 1998 [14].P. Mazlumder, E.,M., Rudnick, "Genetic Algorithms for VLSI Design Layout & Test Automation", Prentice Hall Inc., **1999**

[15]: D., E., Goldberg, "Genetic Algorithms in Search Optimization and Machine Learning", Addison-Wesley, Reading, **1989**

[16] J. W. Allen., and F.W. Buruce.," Power Generation, Operation, and Control", 2nd Edition, John Wiley & Sons, 1996

[17] H. Saruhan. I. Uygur , "Design Optimization of Mechanical Systems Using Genetic Algorithms", Journal of the Institute of Science & Technology of Sakarya University, 7/2, **2003**

[18]. W. Lu, ., "Optimum Design of Cold-Formed Steel Purlins Using Genetic Algorithms", *PhD Thesis, Helsinki* University of Technology of Steel Structures, Helsinki, **2003**

[19] H. Saruhan, "Genetic Algorithms: an Optimization Technique", *TECHNOLGY*, Vol. 7 İssue1, 2004

[20] M. Mitchell, "Genetic Algorithms : An Overview, An Introuction to Genetic Algorithms *,The MIT Pres* , Massachusetts , **1997**

[21] G. Syswerda., "Uniform Crossover in Genetic Algorithms", proceeding of the 3rd International Conference on Genetic Algorithms, Morgan Kaufman, pp 2-9,**1989**