## Self-Adaptive Multi-Population Jaya Algorithm based Reactive Power Reserve Optimization Considering Voltage Stability Margin Constraints

Dr. Aditya Tiwary Associate Professor Electrical & Electronics Engineering Department Institute of Engineering & Science IPS Academy, Indore Email ID: raditya2002@gmail.com

*Abstract*— In this paper an innovative technique is proposed to achieve the optimum value of reactive power reserve accounting voltage stability margin constrains. Reactive power reserve and voltage stability are important issues for proper operation on the power system. This is achieved by suitable settings of reactive power control variables. The fitness function has been minimizes using Self-adaptive multi-population based Jaya Algorithm (SAMP - Jaya). The developed algorithm has been implemented on two IEEE test systems.

Keywords- Self-adaptive multi-population based Jaya Algorithm, voltage stability margin, reactive power reserve, reactive power control

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variables

### NOTATIONS

$QG_p$	Reactive power output of $\boldsymbol{p}^{th}$ generator buses
$QG_p^{ ext{max}}$ , $QG_p^{ ext{min}}$	Upper and lower limit on reactive power output of p <sup>th</sup> bus
$QG_p^{\max}$	Average reactive power output
VSM	Voltage stability margin
VSM <sup>th</sup>	Threshold value of Voltage stability margin
$V_i^{\max}$ , $V_i^{\min}$	Upper and lower limits on load bus voltages
$U_{\kappa}$	K <sup>th</sup> control variable
$U_K^{\min}$ , $U_K^{\max}$	Lower and upper limits on control variable
NC	Number of total control variables
$P_d^0$	Current total load

#### I. INTRODUCTION

Modern power system is highly complex in nature. In this complex power system obtaining the desired voltage profile along with voltage stability margin in very important and is a matter of top priority. Since long proximity indicators have been used for voltage security enhancement. Tiranuchit et al. [1] applied minimum singular value of jacobian to maintain desired voltage stability margin (VSM) and voltage profile. Sensitivity analysis has been used by Begovic et al. [2] for improving voltage security. An algorithm for optimum reactive power dispatch employing LP and an optimal impedance solution on voltage stability index was developed by Chebbo et al. [3]. Ajjarapu et al. [4] developed an optimal planning strategy for reactive power against voltage instability employing repeated load flow runs up to voltage collapse point. Bansilal et al. [5] proposed optimal reactive power dispatch algorithm for voltage stability improvement. Kessel et al. [6] estimated the voltage stability of the power system. Pande et al. [7] used functional link network for reactive power management and voltage stability enhancement. Titare et al [8] developed an approach to mitigate probability of voltage collapse accounting parameter uncertainties using improved PSO algorithm. Fuzzy technique to develop a reactive power optimization algorithm for hybrid system was developed by Taghavi et al. [9]. Khazali et al. [10] applied harmony search algorithm for obtaining optimal performance of the system based on reactive power considerations. Genetic algorithm has been employed for voltage stability margin enhancement and reactive power dispatch by Devaraj et al. [11]. Mousavi et al. [12] developed a preventive strategy for reactive power management along with VSM improvement. Singh et al. [13] developed a multi objective VAR management algorithm using modified differential evolution algorithm. Titare et al. [14] used voltage dependent reactive power reserves modeling for voltage stability enhancement employing ensemble of mutation and crossover strategies and parameters in differential evolution (EPSDE). Fang et al. [15] developed a robust optimal reactive power reserves dispatch under stochastic environment of load injected at buses employing chance constraints relaxation - based method. Bhattacharya and Raj [16] used modal analysis and L-index for optimization of reactive power reserves based on differential evolution technique. Sun et al [17] presented a bi- objective reactive power reserves optimization algorithm to coordinate long and short term voltage stability considerations. Fang et al [18] developed an interval optimal reactive power reserve dispatch considering uncertainties in the load and load direction. Rojar et al. [19] presented an excellent review of various metaheuristic techniques used for optimal reactive reserves dispatch.

In this paper an innovative methodology based on Selfadaptive multi-population based Jaya Algorithm (SAMP -Jaya) for reactive power reserve optimization is proposed. The specific objectives are to limit reactive power generation on both side i.e. over excitation as well as under excitation, such that sufficient reactive power reserve is available on either side. It has been proposed to directly evaluate the VSM and it should have at least some threshold. The proposed methodology has been implemented on IEEE 14-bus and 30-bus standard test systems.

### II. PROBLEM FORMULATION

The quadratic performance index is selected as fitness function as proposed by Purey et al. [20]

$$F = \sum_{p=1}^{NG} \left[ \frac{QG_p - QG_p^{avg}}{QG_p^{\max} - QG_p^{\min}} \right]^2$$
(1)

Now  $QG_p^{avg}$  is given as follows [20]:

$$QP_p^{avg} = 0.5[QG_p^{\max} + QG_p^{\min}]$$
<sup>(2)</sup>

The objective function is minimized subject to following constrains [20]:

- (i)  $VSM \ge VSM^{th}$
- (ii)  $V_i^{\min} \le V_i \le V_i^{\max}$
- (iii)  $U_{K}^{\min} \leq U_{K} \leq U_{K}^{\max}$  K=1,2,3,4,....NC
- (iv) All load bus voltages within limits at  $p_d^0$ and also  $(1 + VSM_{Th}^d)P_d^0$

# **III.** SELF-ADAPTIVE MULTI-POPULATION BASED JAYA ALGORITHM (SAMP - JAYA): AN OVERVIEW:

The self-adaptive multi-population based Jaya algorithm for solving the constrained and unconstrained numerical and engineering optimization problem was proposed by Rao et al. [21]. It is based on the concept that the solution obtained for a given problem should move towards the best solution and avoid the worst solution. First, the initial population is generated having population size (NP) and number of design variables (D) is decided. Now divide the total population into m sub populations depending on quality of solution. Identify the best and worst solution in each and every sub population. Obtain the new value as follows:

$$Y_{i}' = Y_{i} + rand_{1}(Y_{best,i} - |Y_{i}|) - rand_{2}(Y_{worst,i} - |Y_{i}|)$$
(3)
Where

 $Y_i$  is previous value

 $Y_{best,i}$  is best solution

 $Y_{worsti}$  is worst solution

 $rand_1, rand_2$  are random numbers having range of [0,1] Accept the better solution in each sub population. Merge the entire sub population together. Now check if previous best solution of entire population is better than the current best solution in the entire population. If yes, then m is decreased by 1, else m is increased by 1. The procedure is terminated if a maximum number of generations have been executed.

#### **IV. RESULTS AND DISCUSSIONS:**

In this paper, Self-adaptive multi-population based Jaya Algorithm (SAMP - Jaya) has been applied to obtain optimum reactive power reserve using reactive power control variables such as PV- bus voltages, OLTC and shunt compensations on IEEE 14-bus and 30-bus standard test systems.

### 4.1 Case-A: IEEE 14-Bus System [14]

Table-1 shows reactive power control variables (PV-bus voltages, shunt compensations and OLTCs) and all load bus voltages under base case condition. Table-2 shows the comparison of each algorithm to find the best optimal control variable settings with and without optimization using the proposed SAMP - Jaya optimization algorithm and the results has been compared with Jaya [20], TLBO [20], DE [20] and CAPSO [20] techniques. Table-3 shows the comparison of reactive reserves at different generator bus (bus nos. 1<sup>st</sup>, 2<sup>nd</sup> & 3<sup>rd</sup>) using SAMP - Jaya optimization algorithm. Table-4 shows the comparison of SAMP - Jaya with other techniques reported in literature based on arithmetic mean value, standard deviation, best value, worst value, frequency of convergence, standard error, length of confidence interval and confidence interval of fitness function. Fig. 1 shows a plot for the convergence of fitness function with respect to number of iteration for SAMP -Jaya technique. It is observed that SAMP - Jaya gives much better global optimal results than other optimization techniques reported in literature [20].

#### 4.2 Case-B: IEEE 30-Bus System [22]

Table-5 shows reactive power control variables and all load bus voltages under base case condition. Table-6 shows the best optimal control variable settings with and without optimization. Table-7 shows the comparison of reactive reserves at different generator bus (bus nos. 1<sup>st</sup>, 2<sup>nd</sup>, 5<sup>th</sup>, 8<sup>th</sup>, 11<sup>th</sup> & 13<sup>th</sup>) using SAMP - Java technique. Table-8 shows the comparison of proposed SAMP - Jaya optimization algorithm with Jaya [20], TLBO [20], DE [20] and CAPSO [20] techniques based on arithmetic mean value, standard deviation, best value, worst value, frequency of convergence, standard error, length of confidence interval and confidence interval of objective function. Fig. 2 shows a plot for convergence of fitness function with respect to number of iteration for SAMP - Java. It is observed that SAMP - Java gives much better global optimal results than in comparison with Jaya [20], TLBO [20], DE [20] and CAPSO [20] techniques.

#### v. CONCLUSION

An innovative methodology has been presented for reactive power reserves. The objective is to maintain voltage profile and voltage stability margin by using SAMP – Jaya optimization algorithm. The developed algorithm performance is been analysed based on different criteria such as mean value, median value, mean deviation, variance, standard deviation, best value, worst value, frequency of convergence, standard error, length of confidence interval, confidence interval, class interval & proportionate frequencies of fitness function. The results obtained by the proposed SAMP – Jaya optimization algorithm has given better results than in comparison with the different methods shown in the literature.

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Table-1. Load	flow	solution	for 14-bus	test system	under stresse	d
		cor	ndition. [20]			

Total load (S<sub>d</sub>)=3.6758pu, Static voltage stability limit-4 6858pu

	111111-1.0050pu								
S.	Control	Control	Load bus	Load bus					
No.	variables	variables	voltages	voltage					
		magnitude	_	magnitude (pu)					
		(pu)							
1	$V_1$	1.0812	$V_4$	0.8248					
2	$V_2$	1.0485	V <sub>5</sub>	0.8618					
3	<b>V</b> <sub>3</sub>	1.0739	V <sub>6</sub>	0.9522					
4	BSH4	0.0015	V <sub>7</sub>	0.8618					
5	BSH12	0.0057	V <sub>8</sub>	0.9696					
6	$TAP_4$	1.0657	V9	0.8291					
7	TAP <sub>10</sub>	1.0673	V10	0.8126					
			V11	0.8114					
			V12	0.7970					
			V13	0.7917					
			V14	0.7897					

Table-2 Reactive power control variables using SAMP - Jaya algorithms for

	IEEE 14-bus system $(S_{dt})=3.0/38$ pu.							
S.	Reactive	Base	SAMP	JAYA	TLBO	DE [20]	CAPS	
No.	control	case	-	[20]	[20]		O [20]	
	variable	[20]	JAYA					
1	Tap <sub>4</sub>	1.0657	0.9315	0.9317	0.9320	0.9326	0.9284	
2	Tap <sub>10</sub>	1.0673	0.9268	0.9266	0.9254	0.9258	0.9217	
3	$Qc_4$	0.0015	0.0512	0.0508	0.0370	0.0447	0.0409	
4	Qc12	0.0057	0.0478	0.0473	0.0483	0.0357	0.0318	
5	V1	1.0812	1.0790	1.0788	1.0798	1.0797	1.0776	
6	V <sub>2</sub>	1.0485	1.0428	1.0428	1.0445	1.0457	1.0447	
7	V <sub>3</sub>	1.0739	1.0695	1.0693	1.0704	1.0716	1.0693	

Table-3 Reactive power reserve at generator buses and fitness function using SAMP – Jaya technique for IEEE 14-bus system (S<sub>dt</sub>)=3.6758pu.

S.	Methodology	Reactiv	e power Rese	rve (pu)	Total	Fitness
No.		Qgk(res)1	Qgk(res)1	Qgk(res)1	reactive	function
					power	
					reserve	
					(pu)	
1	SAMP – Jaya	2.6988	0.8185	0.014	3.5318	0.2082
2	JAYA [20]	2.6991	0.8181	0.014	3.5312	0.2088
3	TLBO [20]	2.7114	0.7916	0.0152	3.5183	0.221
4	DE [20]	2.737	0.7646	0.012	3.5136	0.2326
5	CAPSO [20]	2.7628	0.707	0.0319	3.5016	0.2489
6	Base Case [20]	2.5295	0.6628	0.0398	3.2321	0.3381

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# Table -4 statistical inferences based on proposed SAMP – Jaya techniques for IEEE 14-bus system.

Optimization	SAMP	JAYA	TLBO	DE	CAPSO [20]
methods	– Jaya	[20]	[20]	[20]	
Arithmetic	0.2102	0.2132	0.2275	0.2430	0.2669
mean value					
of OF					
Median value	0.2078	0.2115	0.2255	0.2421	0.2653
of OF					
Mean	2.00E-	2.00E-	4.00E-	1.50E-	1.50E-05
deviation of	05	05	05	05	
OF					
Variance of	2.38E-	2.47E-	5.05E-	8.03E-	1.77E-04
OF	05	05	05	05	
Standard	0.0045	0.0049	0.0071	0.0089	0.0133
deviation of					
OF					
Best value of	0.2056	0.2088	0.2210	0.2326	0.2489
OF					
Worst value	0.2245	0.2273	0.2480	0.2662	0.2951
of OF					
Frequency of	14	13	11	10	10
convergence					
Confidence	0.95	0.95	0.95	0.95	0.95
level					
Determined	2.0452	2.0452	2.0452	2.0452	2.0452
value for the					
Engg.					
Application					
Standard	0.0020	0.0023	0.0032	0.0041	0.0061
error of the					
mean OF					
Confidence	$0.2101 \le \mu \le$	$0.2109 \le \mu \le$	0.2243 ≤ µ ≤	$0.2389 \le \mu$	$0.2608 \le \mu \le 0.2730$
interval of					
the OF					
Length of	0.0085	0.0094	0.0131	0.0167	0.0249
confidence					
interval of					
the OF					

### OF-objective function

Table-5. Load flow solution for 30-bus test system under stressed condition. [20]

	(14)	,,	0	
S.	Control	Control	Load	Load bus
No.	variables	variables	bus	voltage
		magnitude	voltages	magnitude
		(pu)		(pu)
1	<b>V</b> <sub>1</sub>	1.0842	V <sub>3</sub>	1.0231
2	V <sub>2</sub>	1.0476	$V_4$	1.0105
3	V <sub>5</sub>	1.0112	V <sub>6</sub>	1.0052
4	V <sub>8</sub>	1.0262	V <sub>7</sub>	0.9902
5	V11	1.0845	V9	0.9400
6	V13	1.0928	V10	0.8948
7	BSH10	0.0106	V12	0.9516
8	BSH24	0.0040	V14	0.9135
9	TAP <sub>11</sub>	1.0686	V15	0.8996
10	TAP <sub>12</sub>	1.0693	V16	0.9110
11	TAP <sub>15</sub>	1.0563	V17	0.8873
12	TAP <sub>36</sub>	0.9215	V18	0.8686
			V19	0.8578
			V20	0.8651
			V21	0.8674
			V22	0.8693
			V23	0.8703
			V24	0.8511
			V25	0.8593
			V26	0.8379
			V27	0.8749
			V28	0.9981
			V29	0.8311
			V30	0.8084

## Table-6 Reactive power control variables using SAMP - Jaya algorithms for IEEE 30-bus system $(S_{4x}) = 4.6759$ nu

	$\frac{1}{10000000000000000000000000000000000$								
S.	Control	Base	SAMP	JAYA	TLBO	DE	CAPSO		
No.	variable	case	-	[20]	[20]	[20]	[20]		
		[20]	JAYA						
1	Tap <sub>11</sub>	1.0686	0.9246	0.9247	0.9232	0.9232	0.9253		
2	Tap <sub>12</sub>	1.0693	1.0260	1.0263	1.0238	1.0238	1.0275		
3	Tap <sub>15</sub>	1.0563	0.9305	0.9314	0.9327	0.9349	0.9266		
4	Tap <sub>36</sub>	0.9215	1.0743	1.0759	1.0839	1.0692	1.0791		
5	Qc10	0.0106	0.1753	0.1750	0.1756	0.1543	0.1556		
6	Qc24	0.0040	0.0378	0.0380	0.0372	0.0356	0.0375		
7	V1	1.0842	1.0821	1.0820	1.0768	1.0833	1.0710		
8	V2	1.0476	1.0315	1.0317	1.0266	1.0352	1.0194		
9	V5	1.0112	1.0098	1.0097	1.0011	1.0111	0.9980		
10	V8	1.0262	1.0141	1.0141	1.0140	1.0261	1.0234		
11	V11	1.0845	1.0839	1.0838	1.0833	1.0846	1.0768		
12	V13	1.0928	1.0912	1.0912	1.0807	1.0921	1.0875		

Table-7 Reactive power reserve at generator buses and fitness function using SAMP – Java technique for IEEE 30-bus system (S<sub>4t</sub>)=4.6759pu.

Met		Reactiv	e power	Reserv	e (pu)		Tota	Fitne
hod	Og	Ogk(r	Ogk	Ogk	Ogk	Ogk	1	SS
olog	k(re	es)2	(res)	(res)	(res)	(res)	reac	funct
у	s)1		5	8	11	13	tive	ion
							pow	
							er	
							rese	
							rve	
							(pu)	
SA	0.9	0.222	0.03	0.30	0.05	0.03	1.62	1.46
MP	753	8	25	28	45	01	78	38
_								
Jaya								
JAY	0.9	0.222	0.03	0.30	0.05	0.03	1.62	1.46
A	759	8	37	76	47	13	60	92
[20]								
TL	0.9	0.209	0.08	0.22	0.04	0.05	1.61	1.51
BO	908	9	04	92	41	89	33	19
[20]							1 - 0	
DE	1.0	0.202	0.07	0.18	0.06	0.03	1.60	1.54
[20]	301	7	91	97	84	8	8	09
CA	0.9	0.367	0.08	0.03	0.07	0.03	1.58	1.58
PSO	931	7	55	03	16	97	79	97
[20]								
Bas	1.2	0.227	0.07	0.09	-	-	1.31	2.29
e	278	2	29	65	0.07	0.23	47	25
Cas					49	48		
e								
[20]								

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Optimization	SAMP	JAYA	TLBO	DE	CAPS
methods	– Jaya	[20]	[20]	[20]	O [20]
Arithmetic	1.4805	1.4814	1.5348	1.5781	1.6461
mean value of					
OF					
Median value	1.4768	1.4787	1.5298	1.5751	1.6476
of OF					
Mean deviation	2.00E-	2.00E-	4.50E-	5.00E-	-5.00E-
of OF	05	05	05	05	05
Variance of OF	1.40E-	1.40E-	4.02E-	7.64E-	1.38E-
	04	04	04	04	03
Standard	0.0115	0.0118	0.0200	0.0276	0.0371
deviation of OF					
Best value of	1.4678	1.4692	1.5119	1.5409	1.5897
OF					
Worst value of	1.5105	1.5142	1.5849	1.6479	1.7177
OF					
Frequency of	13	12	11	10	9
convergence					
Confidence	0.95	0.95	0.95	0.95	0.95
level					
Determined	2.0452	2.0452	2.0452	2.0452	2.0452
value for the					
Engg.					
Application					
Standard error	0.0041	0.0054	0.0091	0.0126	0.0169
of the mean OF					
Confidence	1.4758≤	1.4760≤	1.5257 ≤	1.5655≤	$1.6292 \leq$
interval of the					
OF					
Length of	0.0210	0.0221	0.0372	0.0515	0.0691
confidence					
interval of the					
OF					

## Table -8 statistical inferences based on proposed SAMP – Jaya techniques for IEEE 30-bus system.

OF-objective function



Fig. 1. Plot of convergence of fitness function with respect to number of iteration using SAMP – Jaya techniques for IEEE 14-bus system.



Fig. 2. Plot of convergence of fitness function with respect to number of iteration using SAMP – Jaya techniques for IEEE 30-bus system.