Assessment Study on Speed Control of DC Series Motor Fed by Photovoltaic System via Bacterial Foraging

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Abstract- This paper presents the speed control of a DC series motor supplied by Photovoltaic (PV) system. The proposed design problem of speed controller is formulated as an optimization problem. Bacteria Foraging Optimization Algorithm (BFOA) is employed to search for optimal Proportional Integral (PI) parameters of speed controller by minimizing the time domain objective function. The performance of the proposed technique has been evaluated with respect to load torque variation, ambient temperature and radiation. Simulation results have shown the validity of the proposed technique in controlling the speed of DC series motor under different disturbances.

Key-Words: - DC Series Motor; Photovoltaic System; Speed Control; PI Controller; Bacteria Foraging Optimization Algorithm.

1. Introduction

DC series motors are widely used in traction and application that required high starting torque [l-2]. Due to the inherent characteristic possessed by the DC motor system, such as the complexity of the nonlinear system, unavailability of an accurate and precise mathematical model, the use of conventional PI controller become a suitable solution due to small steady-state error and low costs. However, searching the parameters of PI controller is not an easy task, particularly under varying load conditions, parameter changes and abnormal modes of operation [3-4]. Hence, a novel optimization technique called Bacteria Foraging Optimization Algorithm (BFOA) is applied in this paper to search for the optimal parameters of PI controller for speed control of DC series motor.

Photovoltaic (PV) system refers to an array of cells containing a solar photovoltaic material that converts solar radiation into direct current electricity. Solar PV systems work by converting light into electrical power. This is achieved using a thin layer of semi-conducting material, most commonly silicon, enclosed in a glass or plastic casing. When exposed to sunlight the semi-conducting material causes electrons in the materials' atoms to be knocked loose. The electrons that are knocked loose then flow through the material to produce an electric current known as a DC. The DC is carried through wiring to an inverter which converts the current to AC so it can be connected to main electricity distribution board which either used within the home or fed back into the national grid [5-7]. PV is used in this paper to power DC series motor.

In last few years, many researchers have posed different optimization techniques for enhancing speed tracking system. Tabu Search (TS) is discussed in [8] to design a robust controller for Induction Motor. However, it appears to be effective for the design problem, the efficiency is reduced by the use of highly epistatic objective functions (i.e. where parameters being optimized are highly correlated), and the large number of parameters to be optimized. Furthermore, it is time consuming method. Another heuristic technique like Genetic Algorithm (GA) is illustrated in [9] for optimal design of speed control of Switched Reluctance Motor (SRM). Despite this optimization technique requires a very long run time that may be several minutes or even several hours depending on the size of the system under study. Swarming strategies in fish schooling and bird flocking are used in the Particle Swarm Optimization (PSO) and presented in [10] for optimal design of speed control of different motors [11-13]. However, PSO suffers from the partial optimism, which causes the less exact at the regulation of its speed and the direction. In addition, the algorithm cannot work out the problems of scattering and optimization [14, 15]. Also, the algorithm pains from slow convergence in refined search stage, weak local search ability and algorithm may lead to possible entrapment in local minimum solutions. A relatively newer evolutionary computation algorithm, called Bacteria Foraging (BF) scheme has been presented by [16–18] and further established recently by [19–29]. Moreover, BFOA due to its unique dispersal and elimination technique can find favorable regions when the population involved is small. These unique features of the algorithms overcome the premature convergence problem and enhance the search capability. Hence, it is suitable optimization tool for power system controllers.

This paper proposes a new optimization algorithm known as BFOA for speed control of DC series motor supplied by PV system. BFOA is used for tuning the PI controller parameters to control the duty cycle of DC/ DC converter and therefore speed control of DC series motor. The design problem of the proposed controller is formulated as an optimization problem and BFOA is employed to search for optimal controller parameters. By minimizing the time domain objective function representing the error between reference speed and actual one is optimized. Simulation results assure the effectiveness of the proposed controller in providing good speed tracking system over a wide range of load torque, ambient temperature and radiation with minimum overshoot/undershoot and minimal steady state error.

2. System under Study

The system under study consists of PV system acts as a voltage source for a connected DC series motor. The input of PV system is the ambient temperature and radiation, while the output is the DC voltage. The proposed controller based on BFOA is used to control the duty cycle of DC/DC converter and consequently the voltage and speed of DC series motor. The schematic block diagram is shown in Fig. 1.

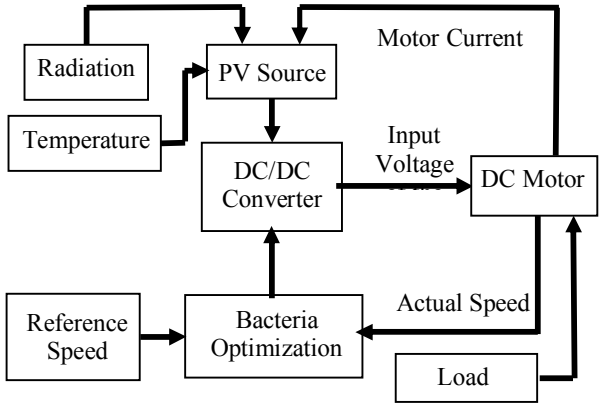

Fig. 1. Overall block diagram of DC series motor for control system.

2.1 DC Series Motor Construction

The DC series motor is a varying speed machine with a markedly drooping speed torque characteristic of the type. For applications requiring heavy torque overloads, this characteristic is particularly advantageous because the corresponding power overloads are held to more reasonable values by the associated speed drops. Very favorable starting characteristics also result from the increase in flux with increased armature current [30-35]. The parameters of DC series motor are shown in appendix.

The proposed system can be simulated with proper mathematic modeling. The DC series motor can be written in terms of equations as follows [30].

$$\frac{di_a(t)}{dt} = \frac{V_t(t)}{L_a + L_f} - \frac{R_a + R_f}{L_a + L_f} i_a(t) - \frac{M_{af}}{L_a + L_f} i_a(t)\omega_r(t) \quad (1)$$

$$\frac{d\omega_r(t)}{dt} = \frac{M_{af}}{J_m} i_a^2(t) - \frac{f}{J_m}\omega_r(t) - \frac{T_L}{J_m} \quad (2)$$

Where

i_a	The armature current,
V_t	The motor terminal voltage,
R_a, L_a	The armature resistance and inductance,
R_f, L_f	The field resistance and inductance,
ω_r	The motor angular speed,
J_m	The moment of inertia,
T_L	The load torque,
f	The friction coefficient,
M_{af}	The mutual inductance between the armature and field.

2.2 Photovoltaic Modeling

Solar cell mathematical modeling is an important step in the analysis and design of PV control systems. The PV mathematical model can be obtained by applying the fundamental physical laws governing the nature of the components making the system [5].

To overcome the variations of illumination, temperature, and load resistance, voltage controller is required to track the new modified reference voltage whenever load resistance, illumination and temperature variation occurs. I-V characteristics of solar cell are given by the following equations [6-7]:

The solar cell mathematical modeling is

$$I_c = I_{ph} - I_o\left\{e^{\left[\frac{q_o}{AkT}\left(V_c + I_c R_s\right)\right]} - 1\right\} \quad (3)$$

$$V_c = \frac{AkT}{q}\ln\left(\frac{I_{ph} + I_o - I_c}{I_o}\right) - I_c R_s \quad (4)$$

$$I = I_{ph} - I_o\left\{e^{\left[\frac{q_o}{n_s AKT}\left(V + n_s I R_s\right)\right]} - 1\right\} \quad (5)$$

$$V = \frac{n_s AkT}{q_o}\ln\left(\frac{I_{ph} + I_o - I}{I_o}\right) - n_s I R_s \quad (6)$$

Where;

$$I_{ph} = \frac{G}{1000}\left[I_{sc} + k_i\left(T - T_r\right)\right] \quad (7)$$

$$I_o = I_{or}\left(\frac{T}{T_r}\right)^3 e^{\left[\frac{q_o E_g}{AK}\left(\frac{1}{T_r} - \frac{1}{T}\right)\right]} \quad (8)$$

The module output power can be determined simply from

$$P = V.I \quad (9)$$

Where;

I , V	Module output current and voltage,
I_c , V_c	Cell output current and voltage,
I_{ph}, V_{ph}	The light generation current and voltage,
I_s	Cell reverse saturation current,
I_{sc}	The short circuit current,
I_o	The reverse saturation current,
R_s	The module series resistance,
T	Cell temperature,

K	Boltzmann's constant,
q_o	Electronic charge
KT	(0.0017 A/◦C) short circuit current temperature coefficient
G	Solar illumination in W/m^2
E_g	Band gap energy for silicon
A	Ideality factor,
T_r	Reference temperature,
I_{or}	Cell rating saturation current at T_r ,
n_s	Series connected solar cells,
k_i	Cell temperature coefficient.

Thus, if the module parameters such as module series resistance (R_s), reverse saturation current (I_o), and ideality factor (A) are known, the I-V characteristics of the PV module can be simulated by using equations (7 and 8).

2.3 DC-DC Converter

The choice DC-DC converter technology has a significant impact on both efficiency and effectiveness. Many converters have been used and tested; buck converter is a step down converter, while boost converter is a step up converter [31]. In this paper, a hybrid (buck and boost) DC/DC converter is used. The equations for this converter type in continuous conduction mode are:

$$V_B = \frac{-K}{1-K} V_{ph} \tag{10}$$

$$I_B = \frac{K-1}{K} I_{ph} \tag{11}$$

Where K is the duty cycle of the pulse width modulation (PWM). V_B , I_B are the output converter voltage and current respectively. The Matlab/Simulink [36] of PV system and DC-DC converter is shown in Fig. 2.

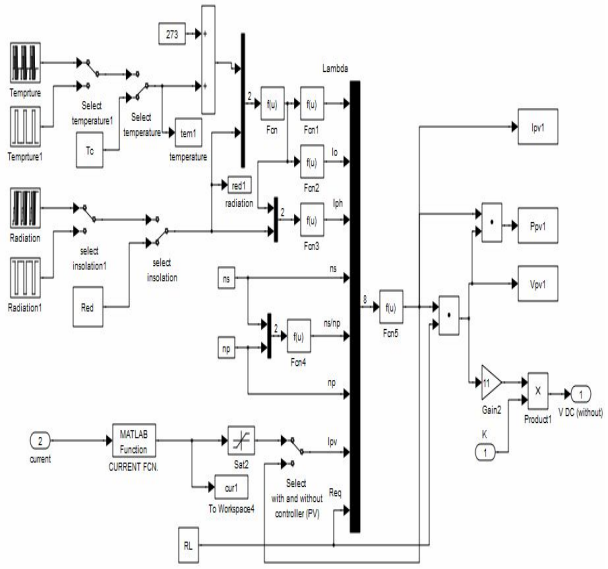

Fig. 2. Matlab/Simulink for PV system and DC-DC converter.

3. Objective Function

A performance index can be defined by the Integral of Time multiply Absolute Error (ITAE). Accordingly, the objective function J [37-40] is set to be:

$$J = \int_0^\infty t(|e|)dt \tag{12}$$

Where $e = w_{reference} - w_{actual}$

Based on this objective function J optimization problem can be stated as: Minimize J subjected to:

$$K_p^{\min} \le K_P \le K_p^{max}, K_I^{\min} \le K_I \le K_I^{max} \tag{13}$$

This paper focuses on optimal tuning of PI controller for speed tracking of DC motor using BFOA algorithm. The aim of the optimization is to search for the optimum controller parameters setting that minimize the difference between reference speed and actual one. On the other hand, in this paper the goal is speed control of DC motor and finally designing a low order controller for easy implementation.

4. Overview of BFOA

Natural selection tends to eliminate animals with poor foraging strategies and favor the propagation of genes of those animals that have successful foraging strategies since they are more likely to enjoy reproductive success. After many generations, poor foraging strategies are either eliminated or shaped into good ones. The Escherichia coli bacteria that are present in human intestine also undergo a foraging strategy. The control system of these bacteria that dictates how foraging should proceed can be subdivided into four sections namely Chemotaxis, Swarming, Reproduction and Elimination and Dispersal [16-17].

4.1 Chemotaxis

The characteristics of movement of bacteria in search of food can be defined in two ways, i.e. swimming and tumbling together known as chemotaxis. A bacterium is said to be 'swimming' if it moves in a predefined direction, and 'tumbling' if moving in an altogether different direction. Mathematically, tumble of any bacterium can be represented by a unit length of random direction $\varphi(j)$ multiplied by step length of that bacterium $C(i)$. In case of swimming, this random length is predefined.

4.2 Swarming

For the bacteria to reach at the richest food location, it is desired that the optimum bacterium till a point of time in the search period should try to attract other bacteria so that together they converge at the desired location more rapidly. To achieve this, a penalty function based upon the relative distances of each bacterium from the fittest bacterium till that search duration, is added to the original cost function. Finally, when all the bacteria have merged into the solution point, this penalty function becomes zero. The effect of swarming is to make the bacteria congregate into groups and move as concentric patterns with high bacterial density.

4.3 Reproduction

The original set of bacteria, after getting evolved through several chemotactic stages reaches the reproduction stage. Here, best set of bacteria gets divided into two groups. The healthier half replaces with the other half of bacteria, which gets eliminated, owing to their poorer foraging abilities. This makes the population of bacteria constant in the evolution process.

4.4 Elimination and Dispersal

In the evolution process, a sudden unforeseen event can occur, which may drastically alter the smooth process of evolution and cause the elimination of the set of bacteria and/or disperse them to a new environment. Most ironically, instead of disturbing the usual chemotactic growth of the set of bacteria, this unknown event may place a newer set of bacteria nearer to the food location. From a broad perspective, elimination, and dispersal are parts of the population level long distance motile behavior. In its application to optimization, it helps in reducing the behavior of *stagnation* often seen in such parallel search algorithms. The detailed mathematical derivations as well as theoretical aspect of this new concept are presented in [18-19]. The computational flow chart of BFOA algorithm is shown in Fig. 3. The parameters of BFOA are shown in appendix.
The algorithm of this technique involves two steps.

[Step 1] Initialization

i) p is the number of parameters to be optimized.
ii) S is the number of bacteria to be used for searching the total region.
iii) N_S is the swimming length after which tumbling of bacteria will be undertaken in a chemotactic loop.
iv) N_C is the number of iteration to be undertaken in a chemotactic loop ($N_C > N_S$).
v) N_{re} is the maximum number of reproduction to be undertaken.
vi) N_{ed} is the maximum number of elimination and dispersal events to be imposed over the bacteria.
vii) P_{ed} is the probability with which the elimination and dispersal will continue.
viii) P (1-p, 1-S, 1) is the location of each bacterium which is specified by random numbers on [-1, 1].
ix) The value of C (i) which is assumed to be constant in this case for all the bacteria to simplify the design strategy.
x) The values of $d_{attract}$, $\omega_{attract}$, $h_{repelent}$ and $\omega_{repelent}$.

Step-2 Iterative algorithm for optimization

This section models the bacterial population chemotaxis, swarming, reproduction, elimination and dispersal (initially, j=k=l=0). For the algorithm updating θ^i automatically results in updating of P.
[1] Elimination-dispersal loop: l=l+1
[2] Reproduction loop: k=k+1
[3] Chemotaxis loop: j=j+1
a) For i=1, 2,…, S, calculate cost function value for each bacterium i as follows.

- Compute value of cost function $J(i,j,k,l)$.

Let

$J_{sw}(i,j,k,l) = J(i,j,k,l) + J_{cc}(\theta^i(j,k,l), P(j,k,l))$ J_{cc} is defined by the following equation

$$J_{cc}(\theta, P(j,k,l))) = \sum_{i=1}^{S} J_{cc}(\theta, \theta^i(j,k,i)) =$$
$$\sum_{i=1}^{S}\left[-d_{attract}\, exp\left(-\omega_{attract}\sum_{m=1}^{p}(\theta_m - \theta_m^i)^2\right)\right] +$$
$$\sum_{i=1}^{S}\left[h_{repelent} exp\left(-\omega_{repelent}\sum_{m=1}^{p}(\theta_m - \theta_m^i)^2\right)\right] \quad (14)$$

- Let $J_{last} = J_{sw}(i,j,k,l)$ to save this value since one may find a better cost via a run.
- End of For loop

b) For i=1, 2,….S take the tumbling/swimming decision

- Tumble: generate a random vector $\Delta(i) \in \Re^p$ with each element $\Delta_m(i)$ m=1,2,…p,
- Move: Let
$$\theta^i(j+1,k,l) = \theta^i(j,k,l) + C(i)\frac{\Delta(i)}{\sqrt{\Delta^T(i)\Delta(i)}}$$ Fixed step size in the direction of tumble for bacterium i is considered.

Compute $J(i,j+1,k,l)$ and

$$J_{sw}(i,j+1,k,l) = J(i,j+1,k,l) + J_{cc}(\theta^i(j+1,k,l), P(j+1,k,l))$$

Swim
i) Let m=0 (counter for swim length).
ii) While $m < N_S$ (have not climbed down too long)

- Let m=m+1
- If $J_{sw}(i,j+1,k,l) < J_{last}$ (if doing better), let $J_{last} = J_{sw}(i,j+1,k,l)$ and let
$$\theta^i(j+1,k,l) = \theta^i(j,k,l) + C(i)\frac{\Delta(i)}{\sqrt{\Delta^T(i)\Delta(i)}}$$ and use this $\theta^i(j+1,k,l)$ to compute the new $J(i,j+1,k,l)$
- Else, let $m = N_S$. This is the end of the while statement.

iii) Go to next bacterium (i+1) if $i \neq S$

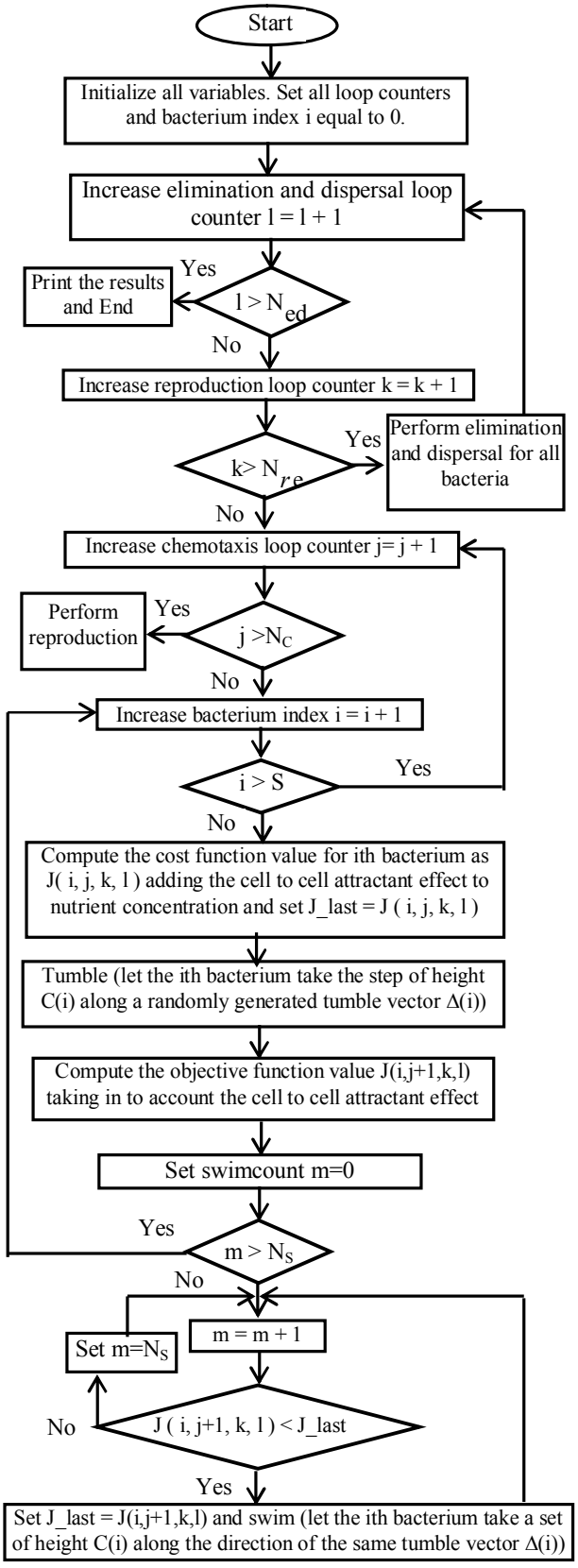

Fig. 3. Flow chart of BFOA.

[4] If j< N_c , go to [step 3]. In this case, continue chemotaxis, since the life of the bacteria is not over.

[5] Reproduction

a) For the given k and l, and for each i=1,2, ..S, let $J^i_{health} = \min_{j \in \{1...N_c\}} \{J_{sw}(i,j,k,l)\}$ be the health of the bacterium i(a measure of how many nutrients it got over its life time and how successful it was at avoiding noxious substance). Sort bacteria in order of ascending cost J_{health} .

b) The $S_r = S/2$ bacteria with highest J_{health} values die and other S_r bacteria with the best value split.

[6] If k< N_{re} , go to [step 2]. In this case, one has not reached the number of specified reproduction steps, so one starts the next generation in the chemotactic loop.

[7] Elimination-dispersal: for $i = 1,2,...., N$, with probability P_{ed} , eliminate and disperse each bacterium, and this result in keeping the number of bacteria in the population constant. To do these, if you eliminate a bacterium, simply disperse one to a random location on the optimization domain. If l< N_{ed} , then go to [step 2]; otherwise end.

The detailed mathematical derivations as well as theoretical aspect of this new concept were presented [16-17].

5. Results and Discussion

In this section different comparative cases are examined to show the effectiveness of the proposed BFOA controller for load torque, ambient temperature and radiation variations.

5.1 Response under step change for load torque

Figs. 4-5 show the step change of load torque, the current, voltage and power of PV system. The motor current, control signal and speed response under variation of the load torque are shown in Figs. 6-7 respectively. The actual speed tracks the reference speed with minimum overshoot and settling time. The settling time is approximately 0.03 second. Moreover, the speed response is very fast for the step variation of load torque. The parameters of proposed PI controller are $K_P = 0.0959$, $K_i = 2.2626$.

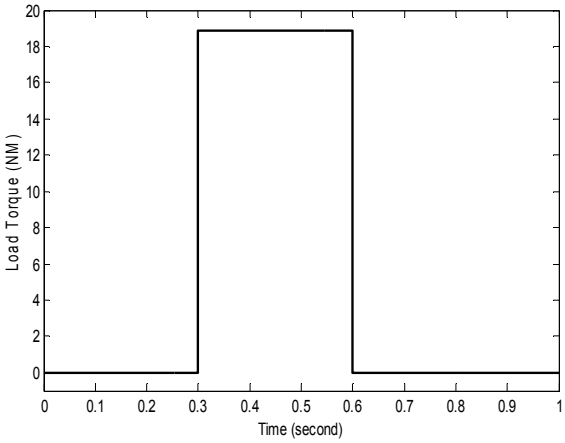

Fig. 4. Step change for load torque.

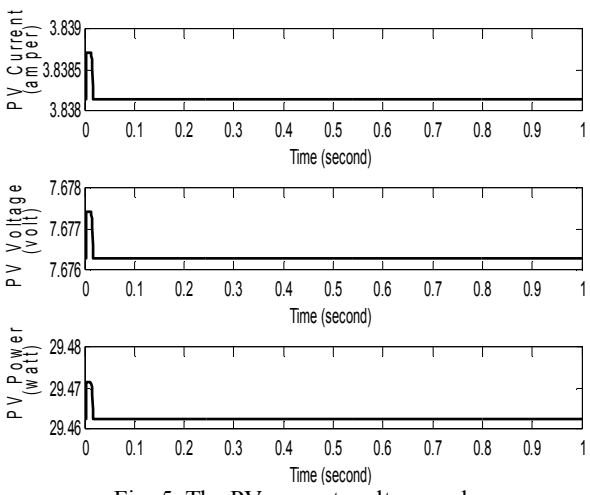

Fig. 5. The PV current, voltage and power.

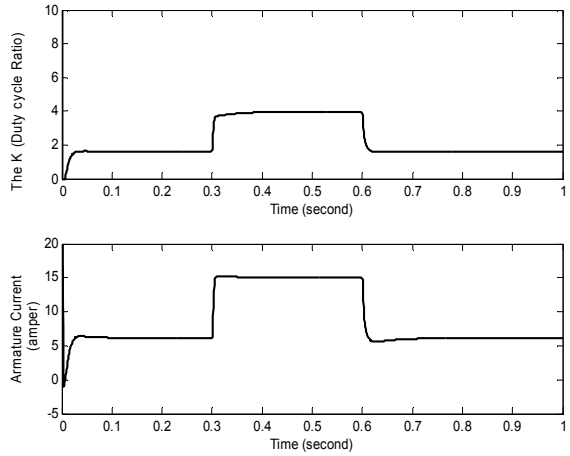

Fig. 6. The control signal and armature current.

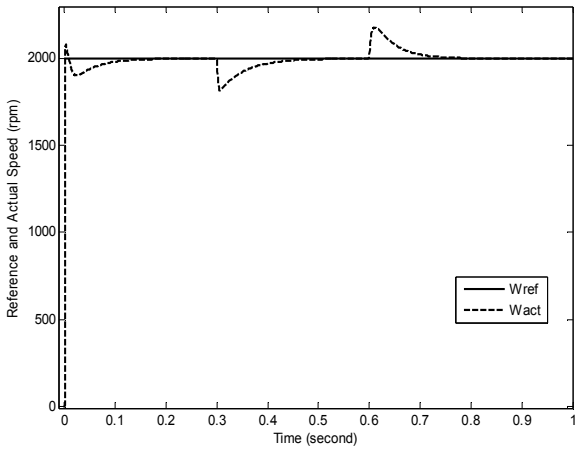

Fig. 7. The reference and actual speed for the DC series motor.

5.2 Response under step change of radiation

In this case, the system responses under variation of PV system radiation are obtained. Figs. 8-9 show the variation of the PV system radiation as an input disturbance and the control signal, motor current, current, voltage and power of PV system respectively. Moreover, the system responses based on BFOA are shown in Figs. 10 and 11. It is clear from these Figs., the proposed BFOA controller improves the speed control and the current response of DC series motor effectively. Also, the overshoot and settling time are highly minimized. Hence, PI based BFOA greatly enhances the performance characteristics of DC series motor.

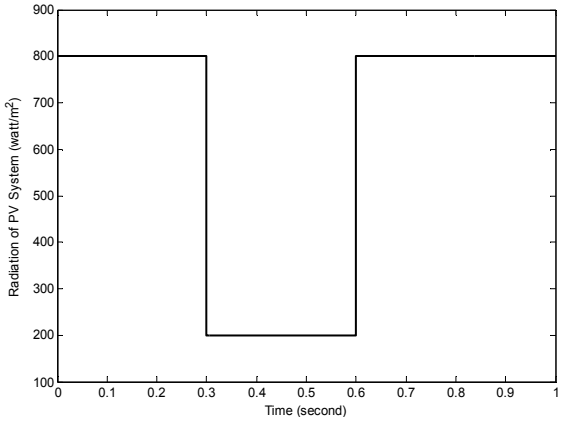

Fig. 8. Step change for PV system radiation.

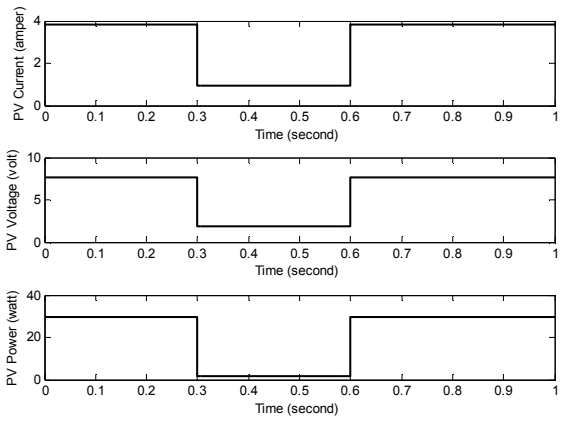

Fig. 9. The PV current, voltage and power.

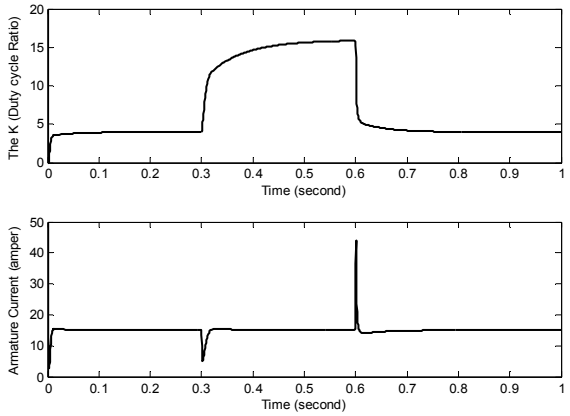

Fig. 10. The control signal and motor current.

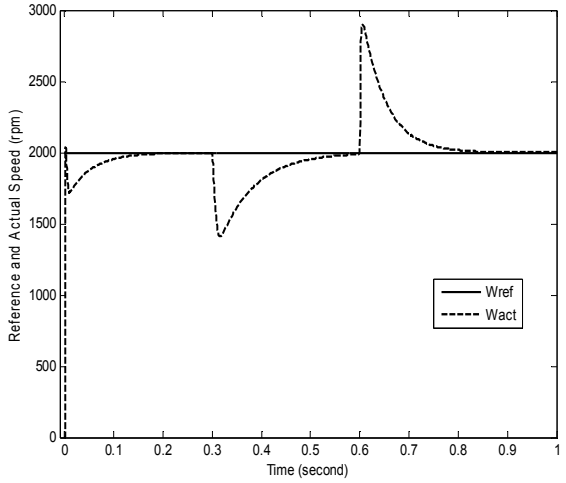

Fig. 11. The reference and actual speed for the DC series motor.

5.3 Response under step change of load torque, radiation and temperature

The effect of applying step change of load torque, radiation and temperature of PV system is shown in this case. Figs. 12-13 illustrate the variation of load torque, radiation, temperature and the output of PV system. The control signal, motor current and a comparison between the actual and reference speed are shown in Figs. 14-15 respectively. From these figures, the steady state and dynamic operation of DC series motor in terms of over shoot and settling time has been enhanced. Also, the actual speed tracks the reference speed at every step. Moreover, the proposed BFOA controller is effectively improved the speed control of DC series motor.

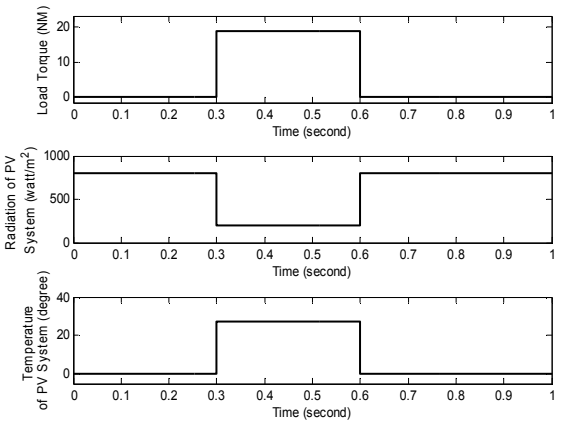

Fig. 12. Step change for load torque, PV system radiation and PV system temperature.

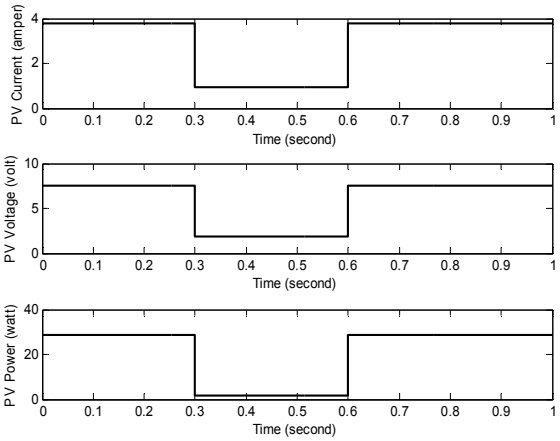

Fig. 13. The PV current, voltage and power.

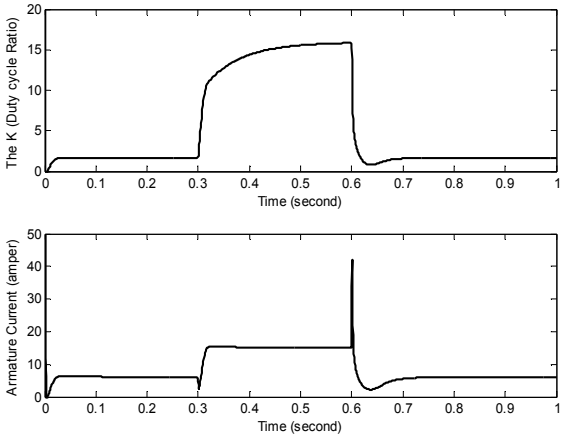

Fig. 14. The control signal and motor current.

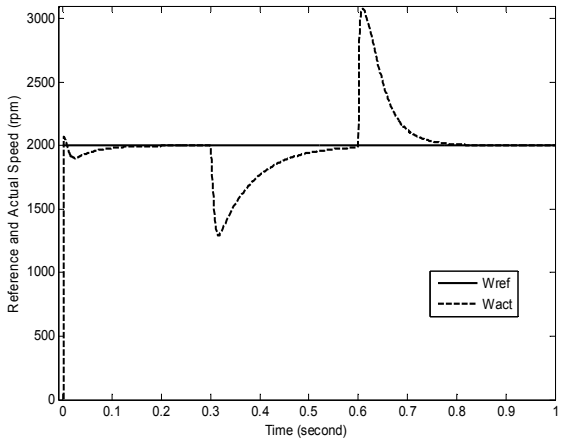

Fig. 15. The reference and actual speed for the motor.

5.4 Response under variables change of load torque, radiation and temperature

In this case, the system response under variations of load torque, radiation and temperature is obtained. Figs. 16 -17 show the change of load torque, radiation, temperature and parameters of PV system respectively. Moreover, the effect of the proposed BFOA controller on speed response is illustrated in Fig. 18. It is clear from this Fig, that the proposed BFOA controller is robust in tracking every change of reference speed. Also, the proposed controller has a small settling time and system response is quickly driven with the reference speed. Hence, the potential and superiority of the proposed BFOA controller is demonstrated.

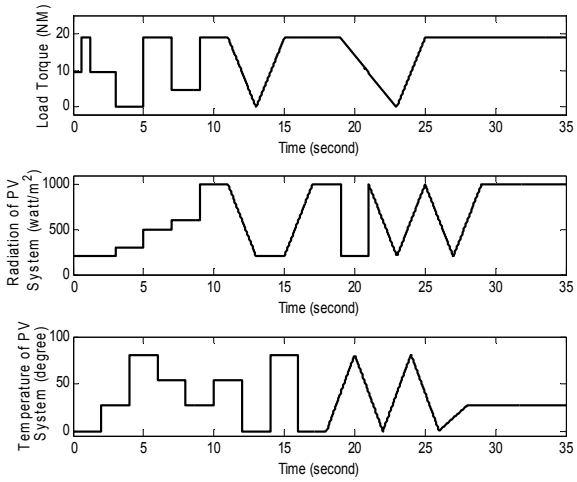

Fig. 16. The high disturbance for load torque, PV radiation and PV system temperature.

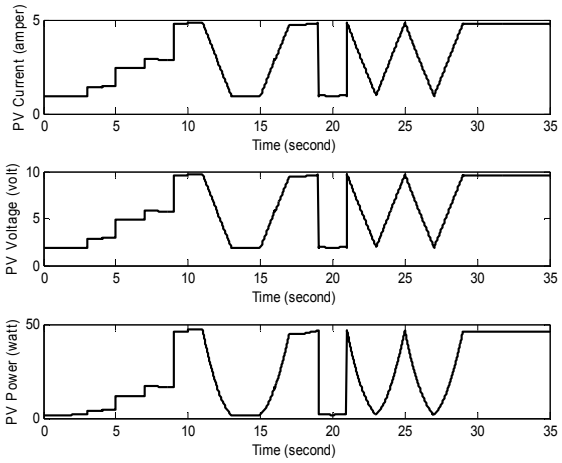

Fig. 17. The PV current, voltage and power.

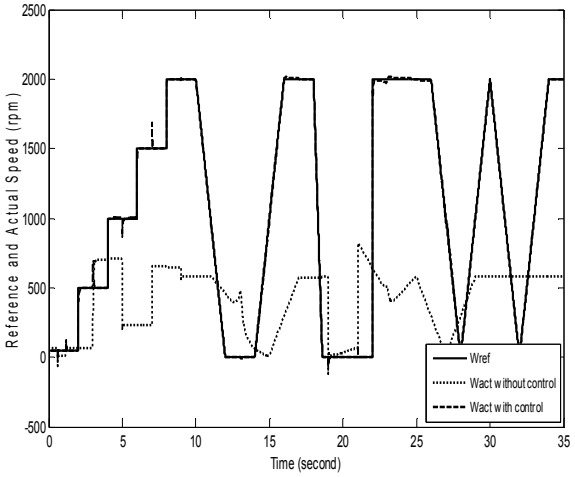

Fig. 18. The reference and actual speed for the motor with and without controller.

6. Conclusions

In this paper, a novel method of speed controller of DC series motor is proposed via BFOA. The design problem of the proposed controllers is formulated as an optimization problem and BFOA is employed to search for optimal parameters of PI controller. By minimizing the time domain objective function, in which the difference between the reference and actual speed are involved; speed control of DC series motor is improved. Simulation results emphasis that the designed BFOA tuning PI controller is robust in its operation and gives a superb performance for the change in load torque, radiation, temperature. Besides the simple architecture of the proposed controller, it has the potentiality of implementation in real time environment.

7. References

[1] G. K. Dubey, "Fundamentals of Electric Drives", Narosa Publishing House New Delhi, 1989.

[2] M. H. Rashid, "Power Electronics Circuits Devices and Applications", Prentice Hall, 3rd Edition, 2003.

[3] P. T. Krein, "Elements of Power Electronics", Oxford University Press, 1998.
[4] N. S. Kumar, V. Sadasivam, K. Prema, "Design and Simulation of Fuzzy Controller for Closed Loop Control of Chopper Fed Embedded DC Drives", IEEE Int. Conf., POWERCON, Singapore, 2004.
[5] C. F. Lu, C. C. Liu, and C. J. Wu, "Dynamic Modelling of Battery Energy Storage System and Application to Power System Stability", IEE Proceedings Generation, Transmission & Distribution, Vol. 142, No. 4, pp. 429-435, July, 1995.
[6] K. Hussein, I. Muta, T. Hoshino and M. Oskada, "Maximum Photovoltaic Power Tracking; An Algorithm for Rapidly Changing Atmospheric Conditions", IEEE Proceedings Generation, transmission and Distribution, Vol. 142, No 1, pp. 59 -64, 1995.
[7] M. M. Saied and A. A. Hanafy, "A Contribution to the Simulation and Design Optimization of Photovoltaic Systems", IEEE Trans. Energy Conv., Vol. 6, pp. 401-406, Sep. 1991.
[8] Y. W. Lee, "A Study on the Design of Speed Controller for Induction Motor Using Tabu Search Algorithm", KORUS'2005, 2005, pp. 818-821.
[9] Cetin Elmas, Tuncay Yigit, "Genetic PI Controller for a Switched Reluctance Motor Drive", International XII. Turkish Symposium on Artificial Intelligence and Neural Networks – TAINN 2003, pp. 1-10.
[10] J. Kennedy and R. Eberhart, "Particle Swarm Optimization", Proceedings of IEEE Int. Conference on Neural Networks, 1995, pp. 1942-1948.
[11] T. V. Mahendiran, K. Thanushkodi, and P. Thangam, "Speed Control of Switched Reluctance Motor using New Hybrid Particle Swarm Optimization", J. of Computer Science, Vol. 8, No. 9, 2012, pp. 1473-1477.
[12] A. S. Oshaba, and E. S. Ali, "Speed Control of Induction Motor Fed from Wind Turbine via Particle Swarm Optimization Based PI Controller", International Research Journal of Applied Sciences, Engineering and Technology, Vol. 5, No. 18, May 2013, pp. 4594-4606.
[13]A. S. Oshaba, and E. S. Ali, "Swarming Speed Control for DC Permanent Magnet Motor Drive via Pulse Width Modulation Technique and DC/DC Converter", International Research Journal of Applied Sciences, Engineering and Technology, Vol. 5, No. 18, May 2013, pp. 4576-4583.
[14] D. P. Rini, S. M. Shamsuddin, and S. S. Yuhaniz, "Particle Swarm Optimization: Technique, System and Challenges", Int. J. of Computer Applications, Vol. 14, No. 1, January 2011, pp. 19-27.
[15] V. Selvi, and R. Umarani, "Comparative Analysis of Ant Colony and Particle Swarm Optimization Techniques", Int. J. of Computer Applications, Vol. 5, No. 4, August 2010, pp. 1-6.
[16] K. M. Passino, "Biomimicry of Bacterial Foraging for Distributed Optimization and Control", IEEE. Control System Magazine, Vol. 22, No. 3, June 2002, pp. 52-67.
[17] S. Mishra, "A Hybrid Least Square Fuzzy Bacteria Foraging Strategy for Harmonic Estimation", IEEE Trans. Evolutionary Computer, Vol. 9, No.1, February 2005, pp. 61-73.
[18] D. B. Fogel, "Evolutionary Computation towards a New Philosophy of Machine Intelligence", IEEE, New York, 1995.
[19] E. S. Ali and S. M. Abd-Elazim, "Bacteria Foraging Optimization Algorithm Based Load Frequency Controller for Interconnected Power System", Int. J. of Electrical Power and Energy Systems, Vol. 33, No. 3, March 2011, pp. 633-638.
[20] E. S. Ali, S. M. Abd-Elazim, "TCSC Damping Controller Design Based on Bacteria Foraging Optimization Algorithm for a Multimachine Power System", Int. J. of Electrical Power and Energy Systems, Vol. 37, No. 1, May 2012, pp. 23-30
[21] E. S. Ali, and S. M. Abd-Elazim, "Coordinated Design of PSSs and SVC via Bacteria Foraging Optimization Algorithm in a Multimachine Power System", Int. J. of Electrical Power and Energy Systems, Vol. 41, No. 1, October 2012, pp. 44-53.
[22] S. M. Abd-Elazim, and E. S. Ali, "Bacteria Foraging Optimization Algorithm Based SVC Damping Controller Design for Power System Stability Enhancement", Int. J. of Electrical Power and Energy Systems, Vol. 43, No. 1, December 2012, pp. 933-940.
[23] E. S. Ali, S. M. Abd-Elazim, "Power System Stability Enhancement via Bacteria Foraging Optimization Algorithm", International Arabian Journal for Science and Engineering (AJSE), Vol. 38, No. 3, March 2013, pp. 599-611.
[24] S. M. Abd-Elazim, and E. S. Ali, "Bacteria Foraging: A New Technique for Optimal Design of FACTS Controller to Enhance Power System Stability", Journal of Electric Engineering, Vol. 13, No. 2, June 2013, pp. 220-227.
[25] S. M. Abd-Elazim, and E. S. Ali, "Hybrid BFOA-PSO Approach for Robust Design of TCSC Based Controller", Journal of Electric Engineering, Vol. 13, No. 2, June 2013, pp. 251-258.
[26] E. S. Ali, S. M. Abd-Elazim, "Optimal Power System Stabilizers Design for Multimachine Power System via Hybrid Approach", Journal of Electric Engineering, Vol. 13, No. 3, September 2013, pp. 83-90.
[27] E. S. Ali, S. M. Abd-Elazim, "Hybrid BFOA-PSO Approach for SSSC Damping Controller Design", Journal of Electric Engineering, Vol. 13, No. 3, September 2013, pp. 257-262.
[28] E. S. Ali, and S. M. Abd-Elazim, "BFOA based Design of PID Controller for Two Area Load Frequency Control with Nonlinearities", Int. J. of Electrical Power and Energy Systems, Vol. 51, 2013, pp. 224-231.
[29] E. S. Ali, and S. M. Abd-Elazim, **"**Hybrid BFOA–PSO Approach for Damping Power System Oscillations by Using Facts Devices", Journal of Engineering and Applied Sciences, Vol. 8, Issue 3, 2013, pp. 89-96.
[30] N. S. Kumar, V. Sadasivam, and M. Muruganandam, "A Low Cost Four quadrant Chopper Fed Embedded DC Drive Using Fuzzy Controller", Int. J. of Electric Power Components and Systems, Vol. 35, No. 8, August 2007, pp. 907 - 920.
[31] H. L. Tan, N. A. Rahim, and W. P. Hew, "A Dynamic Input Membership Scheme for a Fuzzy Logic DC Motor Controller", IEEE Int. Conf. on Fuzzy Systems, Vol. 1, 25-28 May 2003, pp. 426 - 429.
[32] P. Z. Lin, C. F. Hsu, T. T. Lee, "Type -2 Fuzzy Logic Controller design for Buck DC-DC Converters", Proceedings of the IEEE Int. Conf. on Fuzzy systems, 2005, pp. 365- 370.
[33] A. E. El-kholy, and A. M. Dabroom, "Adaptive Fuzzy Logic Controllers for DC Drives: A Survey of the State of the Art", J. of Electrical Systems 2-3, 2006, pp.116-145.
[34] J. Majed, C. Houda, J. Houssem, and B. B. Naceur, "Fuzzy Logic Parameter Estimation of an Electrical System", Systems, Signals and Devices, IEEE SSD 2008, 5th Int. Multi-Conf. on 20-22 July 2008, pp. 1- 6.
[35] B. K. Bose, "Modern Power Electronics and AC Drives", New Jersey: Prentice-Hall, 2002.
[36] http://www.mathworks.com.
[37] F. Golnaraghi, B. C. Kuo, "Automatic Control Systems", 9th Edition, 2010, John Wiley & Sons, Inc.
[38] G. Ellis, "Control System Design Guide: A Practical Guide", 2004, Elsevier.
[39] K. Ogata, "Modern Control Engineering", 5th Edition, 2010, Prentice Hall.
[40] W. J. Rugh, "Linear System Theory", 2nd Edition, 1996, Prentice Hall.

APPENDIX

The system data are as shown below:

a) DC series motor parameters are shown below.

DC motor parameters	Value
Motor rating	3.5 HP
Motor rated voltage	240 V
Motor rated current	12 A
Inertia constant J_m	0.0027 Kg-m^2
Damping constant B	0.0019 N.m.Sec./rad
Armature resistance R_a	1.63 Ω
Armature inductance L_a	0.0204 H
Motor Speed	2000 rpm
Full load torque	19 N. m

b) Bacteria parameters*:* Number of bacteria =10; number of chemotatic steps =10; number of elimination and dispersal events = 2; number of reproduction steps = 4; probability of elimination and dispersal = 0.25; the values of $d_{attract}$ =0.01; the values of $\omega_{attract}$ =0.04; the values of $h_{repelent}$ =0.01; the values of $\omega_{repelent}$ =10.