



PSO algorithm application in induction motor characteristics optimization

A.Asiani and M.A.Taghikhani*

Department of Engineering, Imam Khomeini International University, Qazvin, Iran.

ARTICLE INFO

Article history:

Received: 3 May 2013;

Received in revised form:

16 August 2014;

Accepted: 25 August 2014;

Keywords

Particle Swarm Optimization (PSO),
Single Phase Induction Motor (SPIM),
Design Optimization,
Maximum Efficiency.

ABSTRACT

Parameters design and characteristics optimization of a split-phase single-phase induction motor are proposed in this paper using particle swarm optimization (PSO) algorithm. The PSO considers the motor efficiency as objective function and five performance related items as constraints. The main advantages of the proposed technique are; its simple structure and its straightforward verification of maximum efficiency of induction motor for a given output power. The PSO algorithm was implemented on a test motor and therefore, a code has been provided under MATLAB software. The results show that the PSO method gives more suitable design optimization against conventional methods.

© 2014 Elixir All rights reserved

Introduction

Presently, millions of SPIMs in smaller sizes are in the commercial and domestic field. Therefore, even a minor improvement in the design of this kind of motors may save a vast electrical energy worldwide [1]. There are many types of SPIMs, depending upon the starting arrangement provided with the motor. Advances in the design of these motors have made an improvement in the power factor and efficiency of the motor. Split-phase, capacitor-start, shaded-pole and repulsion-type motors are quite popular from industrial viewpoint. Split phase single phase induction motor design is studied in this paper.

A number of works have reported the improvement of SPIMs efficiency so far. Active power loss effect in induction motor optimum design has proposed in [2]. The method of boundary search along active constraints and the Han-Powell method for optimization of single phase induction motor design have been presented in [3] and [4]. A triac-based drive with an optimal efficiency voltage controller is proposed in [5]. An appropriate method for motor efficiency maximization control, combined with a variable-speed drive, has been presented in [6]. Electrical machine design optimization using genetic algorithm has discussed in [7]. Some of the evolutionary algorithms for optimal design are available in the literatures [8]-[11].

This paper proposes a method for design optimization of single-phase induction motor to maximize the efficiency using particle swarm optimization (PSO). The present paper will be organized as follows. Section 2 briefly explains PSO algorithm. Section 3 presents relationships governing single-phase induction motor. Section 4 discusses the optimal design with variables and constraints. Section 5 gives the detailed discussion on the results of PSO algorithm and their comparison with conventional design.

Particle Swarm Optimization

PSO is a population-based, stochastic optimization algorithm based on the idea of a swarm moving over a given landscape. The algorithm adaptively updates the velocities and members positions of the swarm by learning from the good experiences. In PSO, the velocity v_i^d and position x_i^d of the d th dimension of the i th particle are updated as follows:

$$v_i^d = w \cdot v_i^d + c_1 \cdot r_1 \cdot (pbest_i^d - x_i^d) + c_2 \cdot r_2 \cdot (gbest^d - x_i^d) \quad (1)$$

$$x_i^d = x_i^d + v_i^d \quad (2)$$

Where

x_i : the position of the i th particle

v_i : the velocity of particle i

$pbest_i$:the best location in the search space ever visited by particle i

$gbest$:the best location discovered so far

w : the inertia weight that controls the impact of previous velocity of particle on its current one

r_1, r_2 : independently uniformly distributed random variables with range (0, 1)

c_1, c_2 : positive constants(acceleration) coefficients which control the maximum step size

In PSO, equation (1) is used to calculate the new velocity according to its previous velocity and to the distance of its current position from both its own best historical position and the best position of the entire population or its neighborhood. Generally, the value of each component in v can be clamped to the range $[-v_{max}, v_{max}]$ to control excessive roaming of particles outside the search space. Then the particle flies toward a new position according equation (2). This process is repeated until a user-defined stopping criterion is reached. A linearly decreasing inertia weight from maximum value w_{max} to minimum value w_{min} is used to update the inertia weight:

$$w^k = w_{max} - \frac{w_{max} - w_{min}}{k_{max}} \cdot k \quad (3)$$

K_{max} is maximum iteration number [12, 13].

Relationships Governing Single Phase Induction Motor

Like other types of electrical machines, SPIMs are also designed to meet a given set of specifications. The motor has to be designed to give sufficient starting torque with reasonable performance during its working period. Fig. 1 illustrates equivalent circuit of SPIM based on double rotating field theory. The power flow diagram for a SPIM is shown in Fig. 2. Considering figures 1 and 2 relationships governing single phase induction motor may be expressed as follows [1]:

$$r_{lm} = \frac{\rho \cdot l_{mtm} \cdot T_m}{a_m} \quad (4)$$

$$a_m = \pi \cdot \frac{d_m^2}{4} \quad (5)$$

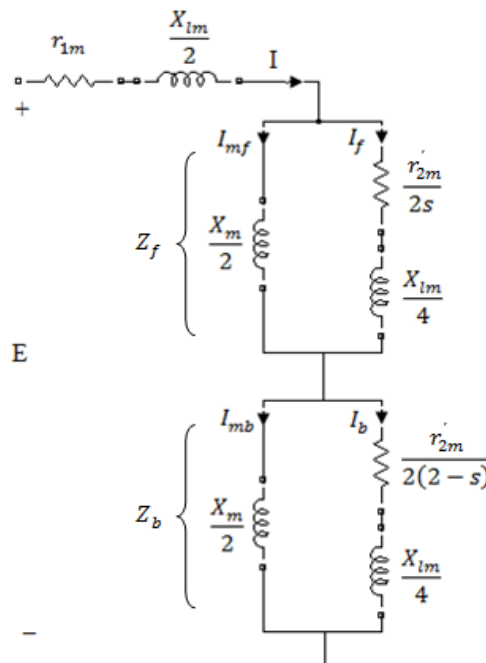


Fig.1. equivalent circuit of SPIM based on double revolving field theory

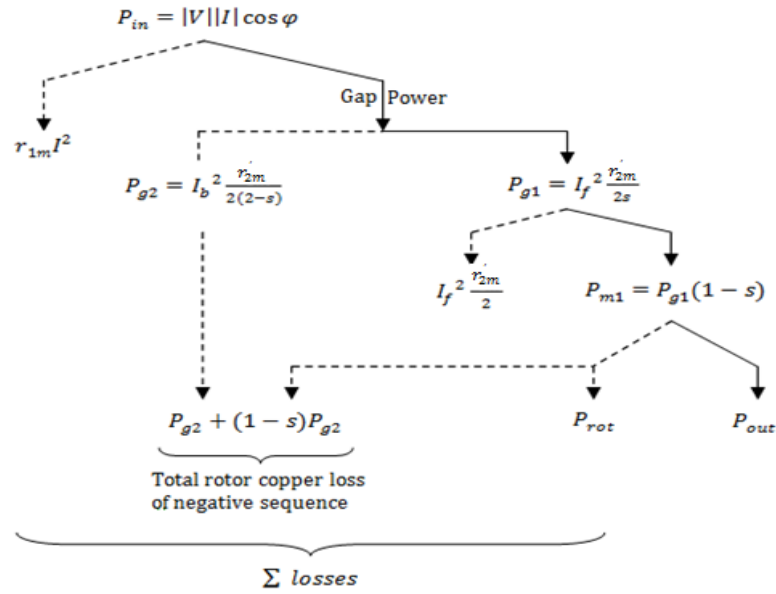


Fig.2. power flow diagram for a SPIM

$$r_{1a} = \frac{\rho \cdot l_{mta} \cdot T_a}{a_a} \quad (6)$$

$$a_a = \pi \cdot \frac{d_a^2}{4} \quad (7)$$

$$r'_{2m} = 2 \cdot N_m^2 \cdot K_{wm}^2 \cdot \rho \cdot \left[\frac{l_b}{A_b \cdot N_b} + \frac{0.64 \cdot D_m}{P^2 \cdot A_e} \right] \quad (8)$$

$$X_{lm} = X_s + X_{zz} + X_e + X_b + X_{sk} \quad (9)$$

$$X_m = K_x \cdot (0.2546 \cdot K_m \cdot C_{sk}) \quad (10)$$

$$K_x = 2 \cdot \pi \cdot f \cdot (N_m \cdot K_{wm})^2 \cdot 10^{-8} \quad (11)$$

$$k_m = \frac{\pi \cdot D_i \cdot L/2}{l'_g \cdot S_f \cdot P} \quad (12)$$

$$C_{sk} = \frac{\sin(\alpha/2)}{\frac{\pi \cdot \alpha}{360}} \quad (13)$$

$$Z_f = \frac{(j \frac{X_m}{2})(\frac{r'_{2m}}{2s} + j \frac{X_{lm}}{4})}{\frac{r'_{2m}}{2s} + j(\frac{X_{lm}}{4} + \frac{X_m}{2})} \quad (14)$$

$$Z_b = \frac{(j \frac{X_m}{2})(\frac{r'_{2m}}{2(2-s)} + j \frac{X_{lm}}{4})}{\frac{r'_{2m}}{2(2-s)} + j(\frac{X_{lm}}{4} + \frac{X_m}{2})} \quad (15)$$

$$Z_{total} = r_{1m} + j \frac{X_{lm}}{2} + Z_f + Z_b \quad (16)$$

$$I = \frac{V}{Z_{total}} = |I| \angle \varphi \quad (17)$$

$$Pf = \cos \varphi \quad (18)$$

$$P_{in} = |V| |I| \cos \varphi \quad (19)$$

$$P_{out} = P_{m1} - P_{rot} - (1-s)P_{g2} \quad (20)$$

$$\eta = \frac{P_{out}}{P_{in}} \quad (21)$$

Where:

r_{1m}, r_{1a}	Stator resistances of main and auxiliary winding, respectively.
a_m, a_a	Area of the main and auxiliary winding conductor, respectively.
d_m, d_a	Wire diameter of main and auxiliary winding, respectively.
T_m, T_a	Number of turns in main and auxiliary winding, respectively.
r'_{2m}	Rotor resistance
X_{lm}	Leakage reactance
X_m	Magnetizing reactance
X_s	Slot leakage reactance
X_{zz}	Zig zag leakage reactance
X_e	End leakage reactance
X_b	Belt leakage reactance
X_{sk}	Skew leakage reactance
D_i	Stator bore diameter
L	Stator stack length
P	Number of poles
V	Voltage
I	Current in main winding
f	Frequency
s	Rotor slip
Pf	Power Factor
K_{wm}	Winding factor
α	Skew of rotor
ρ	Copper resistivity
S_1	Number of stator slots
S_2	Number of rotor slots
l_{mm}	Length of mean turn of the main winding
l_{ma}	Length of mean turn of the auxiliary winding
N_m	Number of conductors in stator main winding
l_b	Length of rotor bar

l_g	Effective gap length
A_e	Area of the end ring
S_f	Saturation factor
D_m	Mean diameter of rotor
N_b	Number of rotor bars
A_b	Area of the rotor conductor
P_{in}	Input power
P_{out}	Output power
η	Efficiency

Optimal SPIM Design

The first optimization component is usually formulated is minimizing $f(x)$ such that x exists within the n -dimensional feasible region $D = \{x \mid x \geq 0, g_i(x) \leq 0, i=1,2,\dots,m\}$. $f(x)$, $g_i(x)$ are real-valued scalar functions and vector x includes the n principal variables for which the optimization is to be performed. The function $f(x)$ is called the “objective function”, for which the optimal values of x result in the minimum (maximum) of $f(x)$, and these optimal values satisfy the given constraints. Inequality constraints $g_i(x) \leq 0$ include the performance properties of the motor, dimensional restrictions and additional requirements [3]. The following variables ($x_1 \dots x_6$) are considered as the principal or independent variables of optimization.

x_1 : Stator bore diameter (m)

x_2 : Stator stack length (m)

x_3 : Number of turns in main winding

x_4 : Number of turns in auxiliary winding

x_5 : Wire diameter of main winding (mm)

x_6 : Wire diameter of auxiliary winding (mm)

The performance constraints imposed into induction motor design in this paper is as follows which are expressed in terms of variables:

1. Power factor at rated load
2. Locked rotor current to rated current ratio
3. Starting torque to full load torque ratio
4. Current density in the main winding
5. Current density in the auxiliary winding

The motor efficiency η has been taken as objective function to be maximized.

5. Results and Discussions

The PSO algorithm is implemented to optimize the design of induction motor whose specifications are available in appendix. Results are compared with conventional design and tabulated in table 1. It is observed from the table that the efficiency of the induction motor has increased using PSO compared to the conventional design. Full load slip in PSO is smaller than conventional design. Starting torque to full load torque ratio is slightly better than the other. The PSO algorithm parameters used in this paper are tabulated in table 2 which determined by trial and error method by using computer simulations.

Fig. 3 shows the variations of efficiency against the PSO iteration number during the optimization process. The variations of motor efficiency verse the output power and rotor slip are shown in figures 5 and 6, respectively. From figures 5 and 6, it is obvious that maximum efficiency is achieved at the rated output power with rotor slip of 0.0332.

Table 1: Results of conventional and PSO based induction motor design

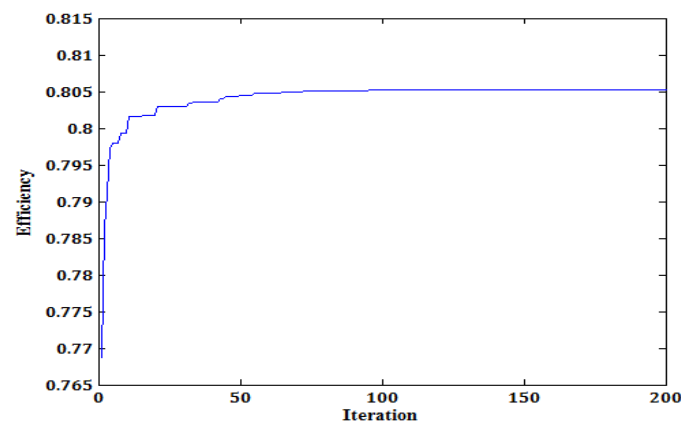
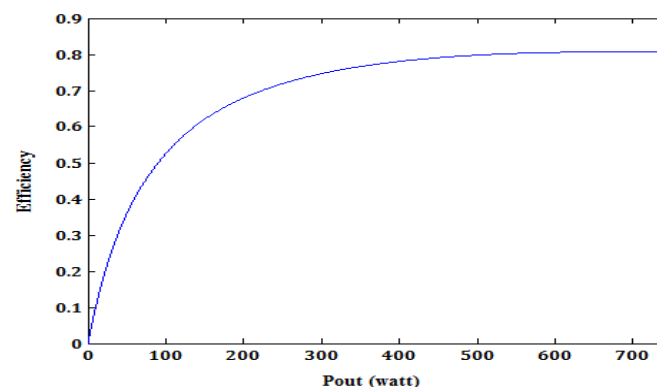
Motor Parameters	Conventional design	PSO algorithm
Stator bore diameter (m)	0.1	0.11227
Stator stack length (m)	0.06	0.045
Number of turns in main winding	184	206
Number of turns in auxiliary winding	262	270
Wire diameter of main winding (mm)	1.42	1.4381
Wire diameter of auxiliary winding (mm)	0.711	0.7377
Efficiency (%)	0.7859	0.8053
Power factor	0.8324	0.8324
Starting torque to full load torque ratio	0.7570	0.7576
Full load slip	0.0366	0.0332

Table 2: Parameters of the PSO

Parameters	Value
Population size	100
dimension number	6
Max iteration	200
w_{max}	0.9
w_{min}	0.4
c_1	2
c_2	2

Conclusion

The application of particle swarm optimization (PSO) in parameters design of a split-phase single-phase induction motor is investigated in this paper. The induction motor design optimization is formulated as a nonlinear programming problem and the motor efficiency is considered as objective function. The main advantages of the proposed technique are; its simple structure and its straightforward verification of maximum efficiency of induction motor for a given output power. The PSO offered better results compared with conventional design and it is more suitable to design optimization of single phase induction motor.

**Fig.3. Variation of efficiency against PSO iteration number****Fig.4. Variation of designed motor efficiency against output power**

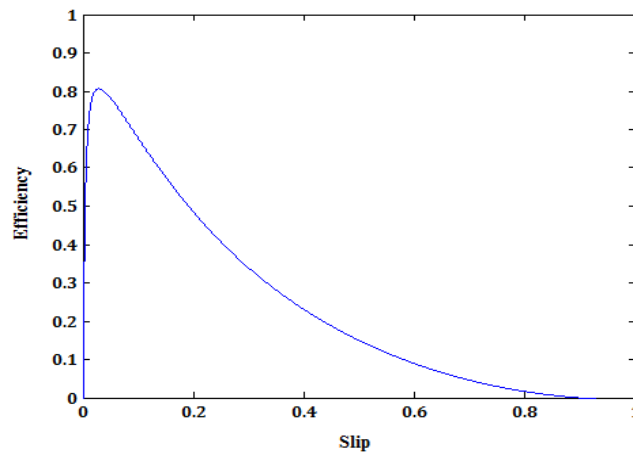


Fig.5. Variation of motor efficiency against rotor slip

Appendix:

Specifications of SPIM motor:

Capacity	1 hp
Voltage	115 V
Frequency	50 Hz
Number of poles	2
Number of stator slots	24
Number of rotor slots	30

References:

- [1] Mittle V.N., Mittal A.; "Design of electrical machines", Fourth editions, Standard Publishers Distributors, 1996.
- [2] Fetih N.H., El-Shewy H.M.; "Induction motor optimum design, including active power loss effect", IEEE Transactions on Energy Conversion, 1986, Vol. 1, No. 3, pp. 155-160.
- [3] Huang H., Fuchs E.F., Zak Z.; "Optimization of single-phase induction motor design, I: Formulation of the optimization technique", IEEE Transactions on Energy Conversion, 1988, Vol. 3, No. 2, pp. 349-356.
- [4] Huang H., Fuchs E.F., White J.C.; "Optimization of single-phase induction motor design, II: The maximum efficiency and minimum cost of an optimal design", IEEE Transactions on Energy Conversion, 1988, Vol. 3, No. 2, pp. 357-366.
- [5] Mademlis C., Kioskeridis I., Theodoulidis T.; "Optimization of single-phase induction motors Part I: Maximum energy efficiency control", IEEE Transactions on Energy Conversion, 2005, Vol. 20, No. 1, pp. 187-195.
- [6] Zahedi B., Vaez-Zadeh S.; "Efficiency optimization control of single-phase induction motor drives", IEEE Transactions on Power Electronics, 2009, Vol. 24, No. 4, pp. 1062-1070.
- [7] Uler G.F., Mohammed O.A., and Chang-Seop Koh; "Design optimization of electrical machines using genetic algorithms", IEEE Transactions on Magnetics, 1995, Vol. 31, No. 3, pp. 2008-2011.
- [8] Wiecek J.P., Gol O., Michalewicz Z.; "An evolutionary algorithm for the optimal design of induction motors", IEEE Transactions on Magnetics, 1998, Vol. 34, No. 6, pp. 3882 - 3887.
- [9] Bhuvaneswari R., Subramanian S.; "Optimization of single-phase induction motor design using radial basis function network", IEEE Indicon Conference, Chennai, India, pp. 35-40, 11-13 Dec. 2005.
- [10] Bellarmine G.T., Bhuvaneswari R., Subramanian S.; "Radial basis function network based design optimization of induction motor", Proceedings of the IEEE Southeast Conference, pp. 75-80, 2006.
- [11] Raj C.T., Srivastava S. P. Agarwal P.; "Realization on PSO based induction motor design via SPEED/PC-IMD", International Conference on Advanced Computer Control(ICACC'09), pp. 65-69, 2009.

- [12] Wang Y., Li B., Weise T., Wang J., Yuan B., Tian Q.; “ Self-adaptive learning based particle swarm optimization”, Information Sciences, 2011, Vol.181, No.20, pp. 4515-4538.
- [13]Jiang Y., Hu T., Huang C., Wu X.; “An improved particle swarm optimization algorithm”, Applied Mathematics and Computation, 2007, Vol. 193, No.1, pp. 231–239.