PSO-PI ALGORITHM FOR WIND FARM SUPERVISION

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Abstract
Nowadays, the research related to the wind farms is oriented to the development of supervision algorithm to manage the active and reactive powers as well as to provide an ancillary system. This paper proposes a particle swarm optimization [PSO] of Proportional integral [PI] regulators algorithm for wind farm supervision. This control system is based on two control levels: A supervisory system controls active and reactive power of the whole wind farm by sending out set points to all wind turbines, and a machine control system ensures that set points at the wind turbine level are reached. The whole control is added to the normal operating power reference of the wind farm established by a Supervisory Control. Finally the performance of the proposed algorithm is verified through simulation results considering a wind farm of three generators DFIGs (1.5 MW).

key-words: wind farm supervision, Doubly fed induction generators DFIG, PI Proportional Integral controller, PSO Particle Swarm Optimizations.

1. INTRODUCTION
For several years, the environmental protection has caused much attention, and consequently, several technologies are developed. It’s the case of the wind power. Nowadays, this source of energy is still used for water pump but it’s mainly used for electricity production and this without any harmful impact to the environment. The high costs of exploitation of the nuclear, thermal power stations and the fossil fuels also, made possibility of wind power being more competitive.

Today, the rate of penetration of wind farms becomes increasingly significant in the electrical network. However, several problems of instability are generated at the time of the connection of these farms to the network, because so far it does not participate to the ancillary system (voltage regulation, frequency regulation, black-start, operation in islanding). Following these problems of instability of the electrical network; ones procedure of obliteration must be necessarily planned by the manager of network, which causes a forced disconnection of the wind generators based on the network instability, furthermore, the supervision of the wind farms is considered to be necessary in order to connect them to the electrical network without disregarding the quality of electric power produced.

The recent research tasks in the field of wind Farms are directed to design supervision algorithms for wind farm with the aim of distributing the references of active and reactive powers on different wind generators. In this context, several algorithms were proposed [2][13][16][25] and can be classified mainly in three categories: The first algorithms are based on Proportional integral regulators PI; this class of algorithms regulates the problem of the supervision by using a simple PI regulator [8]. Two algorithms can be distinguished; the first uses this regulator to regulate the power-factor [15][21], while the second one regulates the active and reactive power directly [1] [16] [27], but the risk of the wind generators saturation is presented as the major problem of these algorithms, because the information on the maximum available active and reactive powers of each wind generators are not taken into consideration [8]. The second Algorithms are based on optimization of the objective function, which is used for the optimal active and reactive powers references distribution on the wind generators [13][23][25]. This function must formulate objectives, it is optimized by a mathematical equation which takes account of several parameters [8], it needs optimization methods like: genetic algorithm [18], neurons networks [10],[17], particles swarm optimization [4][11], and methods which combines the latter with fuzzy logic [13][24]. The last supervision Algorithms which are based on proportional distribution, were developed to distribute the power references in proportional way. From a safety point of view, these algorithms ensure that each wind generator works always far from its limits defined by the (P,Q) diagram[1][2][8]. They determine the references of the active and reactive powers of each wind generators \( P_{WG_{ref}}, Q_{WG_{ref}} \) from the global active and reactive power references required by the network system operator \( P_{WF_{ref}}, Q_{WF_{ref}} \) [8] [19] [20] [6]. Nevertheless, the implementation of this strategy is a little bit complex since it needs information on the available aerodynamic power of all the wind generators [20].

First, this paper describes briefly the studies carried out in order to develop a complete wind farm model made up with DFIG type generators. Then, a simple and robust control algorithm -based on particle swarm
optimization [PSO] of Proportional integral [PI] regulators algorithm for wind farm active and reactive power regulation is presented. The dynamic performance of the developed wind farm model together with its reactive power controller is simulated using MATLAB/SIMULINK. The simulation results illustrate good performance of this supervision.

II. POWER SYSTEM CONFIGURATION

![Fig.1 Power System Configuration](image)

The total diagram of an inter-connected electrical network which has several electrical devices is presented on fig.1, the wind farm is connected to HTA 20KV buses through a transformer of 20KV/690V. Different fixed and variables loads are connected to the same bus with another transformer. A central unit of wind farm supervision is installed in order to control the exchanges \((P_{WF}, Q_{WF})\) powers with the electrical network [8].

The objective of this unit is a management of the total active and reactive powers of the wind farm according to a plan of production required by the system operator. On the hand, A central supervisory control level decides the active and reactive power references \((P_{WF-ref}, Q_{WF-ref})\) for each wind generators local control level, based on received production orders (maximum production or power regulation \((P_{WF-max}, Q_{WF-max})\) from the system operator in other hand.

III. PI ALGORITHM FOR WIND FARM SUPERVISION

II.1. Principle of Operation

Generally, the reference value of the active power that a DFIG should generate is established through optimum generation curves, which provide the active power or the electromagnetic torque to be demanded to the generator as a function of its rotational speed. Such curves are derived as a result of thorough analysis of the wind turbine aerodynamics, and define the maximum mechanical power the DFIG can extract from the wind at any angular speed(fig.2). A considerable amount of references covering this topic in detail can be found throughout the bibliography[1] [8].

We consider a simple distribution of the active and reactive powers which consist on giving the same reference for each wind generator \((P_{WG-ref-i}, Q_{WG-ref-i}).\)

However, it is not immediate to determine the value of this reference but according to the power reference requested for the wind farm manager, by a control of the active and reactive powers at the level of the PCC (point of common coupling) the wind farm around a reference powers \((Q_{WF-ref}, P_{WF-ref}).\) The Fig.3 shows the principle of this control; a PI regulator is used for Correct the error in power and gives the power reference in order to satisfy the request of Operator system. This algorithm of supervision generates the reference which will be distributed between the wind generators in identical way [1][16][27].

![Fig.2: P-Q thermal limit curve for DFIM](image)
Wind generators, \( P_{WG_{\text{ref}i}}, Q_{WG_{\text{ref}i}} \). The simplest strategy is considering that these references are directly the controller outputs, and the same for all wind generators. Note that if the power reference is increased when one or more wind generators have reached the limits, in the next controller computation, the rest of wind farms will automatically assume the load.

IV. PSO BASED PI CONTROLLER

IV.1 Particle Swarm Optimization (PSO)

Kennedy and Eberhart developed a particle swarm optimization algorithm based on the behavior of individuals (i.e., particles or agents) of a swarm [3][5][22]. Its roots are in zoologist’s modeling of the movement of individuals (i.e., fish, birds, and insects) within a group. It has been noticed that members of the group seem to share information among them to lead to increased efficiency of the group. The particle swarm optimization algorithm searches in parallel using a group of individuals similar to other AI-based heuristic optimization techniques. Each individual corresponds to a candidate solution to the problem. Individuals in a swarm approach to the optimum through its present velocity, previous experience, and the experience of its neighbors. In a physical \( n \)-dimensional search space, the position and velocity of individual \( i \) are represented as the velocity vectors. Using these information individual \( i \) and its updated velocity can be modified under the following equations in the particle swarm optimization algorithm.

\[
X_{i}(k+1) = X_{i}(k) + V_{i}(k+1)
\]

\[
V_{i}(k+1) = \omega V_{i}(k) + a_{i} (x_{i}^{\text{best}} - x_{i}(k)) + \beta_{i} (x_{i}^{\text{gbest}} - x_{i}(k))
\]

Where
- \( x_{i}(k) \) is the individual \( i \) at iteration \( k \)
- \( V_{i}(k+1) \) is the updated velocity of individual \( i \) at iteration \( k \)
- \( \alpha_{i}, \beta_{i} \) are uniformly random numbers between \([0,1]\]
- \( x_{i}^{\text{best}} \) is the individual best of individual \( i \)
- \( x_{i}^{\text{gbest}} \) is the global best of the swarm

The procedure of the particle swarm optimization can be summarized in the flow diagram of Fig. 4.

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III.2 PI Algorithm Implementation

As is shown in Fig. 3, the PI controller receives from the transmission system operator TSO a reference of maximum active and reactive power (or voltage reference in some cases) that must not exceed the limit to guarantee the reliability on the grid. The control system consists of two control loops of active and reactive power, respectively. The active control loop receives active power reference, \( P_{WF_{\text{ref}i}} \), from the TSO and this value is compared with the active power measured in the interconnection point(PCC), \( P_{WF_{\text{mes}}} \), the error is computed and a power reference for each wind turbine, \( P_{WF_{\text{ref}}-i} \) (i = 1:N\(WG \) number of wind generators) is set up, through a dispatch function block. As for reactive power, the base value used in per unit representation is the same. The maximum possible reactive power demands, \( Q_{WF_{\text{ref}}} \), from the TSO depends on the P-Q thermal limit curve capability of wind generators and compensating devices added to supply or absorb reactive power. This value defines the limits for the reactive power controller (see Fig. 2).

There are different ways to design the dispatch function (Fig. 3) in order to calculate the power reference for each
In this paper, a PI controller using the PSO algorithm was developed to improve the step transient response of supervisory wind farm system. It was also called the PSO-PI controller. The PSO algorithm was mainly used to determine two optimal controller parameters \( K_p \) and \( K_i \) such that the controlled system could obtain a good step response output.

The searching procedures of the proposed PSO-PI controller were shown as below.

**Step 1:** Specify the lower and upper bounds of the two controller parameters and initialize randomly the individuals of the population including searching points, velocities, \( lbest \)s and \( gbest \).

**Step 2:** Calculate the evaluation value of each individual in the population using the evaluation function given by (3).

**Step 3:** Compare each individuals evaluation value with its \( lbest \). The best evaluation value among the \( lbest \) is denoted as \( gbest \).

\[
Min \left\{ (P_{WF} - P_{WF_{ref}})^2 + (Q_{WF} - Q_{WF_{ref}})^2 \right\} \tag{3}
\]

**Step 4:** Modify the member velocity \( v \) of each individual \( i \) according to (2).

**Step 5:** If \( v_i^{(k+1)} > v_g^{\max} \) then \( v_i^{(k+1)} = v_g^{\max} \)

If \( v_i^{(k+1)} < v_g^{\min} \), then \( v_i^{(k+1)} = v_g^{\min} \).

**Step 6:** Modify the member position of each individual according to (1).

**Step 7:** If the number of iterations reaches the maximum, then go to **Step 8**. Otherwise, go to **Step 2**.

**Step 8:** The individual that generates the latest \( gbest \) is an optimal Controller parameters.

### IV.2.1 Selection of PSO parameters

To start up with PSO, certain parameters need to be defined. Selection of these parameters decides to a great extent the ability of global minimization. The maximum velocity affects the ability of escaping from local optimization and refining global optimization [9][12][7].

The size of swarm balances the requirement of global optimization and computational cost. Initializing the values of the parameters is as per table 1.

<table>
<thead>
<tr>
<th>Table 1. PSO selection parameters</th>
</tr>
</thead>
<tbody>
<tr>
<td>Population size</td>
</tr>
<tr>
<td>Number of iterations</td>
</tr>
<tr>
<td>Velocity constant,c1</td>
</tr>
<tr>
<td>Velocity constant,c2</td>
</tr>
</tbody>
</table>

### III.2.2 Termination Criteria

Termination of optimization algorithm can take place either when the maximum number of iterations gets over or with the attainment of satisfactory fitness value[12][14]. Fitness value, in this case is nothing but reciprocal of the magnitude of the objective function, since we consider for a minimization of objective function. In this paper the termination criteria is considered to be the attainment of satisfactory fitness value which occurs with the maximum number of iterations as 100.

The block diagram of the regulation system with a PSO-PI controller is shown in Fig. 5.
IV. SIMULATION RESULTS AND DISCUSSION

The validation of this type of supervision was made on the model of a wind farm of three wind generators situated in different wind profiles. In order to observe the behavior of this regulation we applied to our system different level of active and reactive powers.

The scenarios of simulation used an identical distribution of the active and reactive powers references for wind generators of wind farm [Fig.6, Fig.7].

Based on PSO for the application of the PI tuning we get the PI tuning parameters for the model as

\[\text{Kp(active)} = 0.5 \quad \text{Ki (active)}=3.0\]
\[\text{Kp(reactive)} = 0.4 \quad \text{Ki (reactive)}=1.00\]

IV.1. Comparison results obtained by classical PI and PSO-PI controller

Analysis shows that the design of proposed controller gives a better robustness, and, the performance is satisfactory over a wide range of process operations. Simulation results show performance improvement in time domain specifications for a step response.

Using the PSO approach, global and local solutions could be simultaneously found for better tuning of the controller parameters.

The PI value which was obtained by the PSO algorithm is compared with that of the one derived from classical method in various perspectives fig.6, namely robustness and stability performances. All the simulations were implemented using MATLAB.

V. CONCLUSION

Even though the control algorithm based on the PSO-PI controller proposed in this paper is really robust under wind speed sudden changes, as it does not distribute the demanded reactive power among the DFIGs proportionally, in certain cases it might cause some of the wind generators to work just on their (P-Q) limits, saturated, while others would still be able to generate or absorb much more reactive power.

It can be concluded that wind farm supervision simulation results match, and evidence that active and reactive powers regulation in wind farms based on doubly fed induction generators may be accomplished robust and efficiently by means of the control algorithm suggested in this paper.

7. REFERENCES


Fig. 7. Simulation Results the centralized supervision of the reactive power [PI].

Fig. 8. Simulation Results the centralized supervision of the active power [PSO-PI].


