

## Dispatching Economically Restricted Loads Using a Biogeography-Based Optimization Algorithm and an Improved Harmony Search Algorithm

U.SASHICHANDRA, VIJAYA JOYTHI,

Assistant Professor, Assistant Professor, Department of Humanities and Science,  
Samskruti College of Engineering and Technology, Ghatkesar.

### Abstract

HSA, IHS, and BBO algorithms are compared in this work for handling restricted economic load dispatch issues in a power system with a limited number of available resources. New solution vectors are generated using the IHS algorithm, which makes use of numerous harmony memory consideration rates and dynamic pitch adjustment rates. They were tested in a test system with twenty producing units with ramp rate restrictions and valve point loading constraints, and the algorithms worked well. IHS approach outperforms both Harmony search and Biogeography-based optimization algorithm in terms of total fuel cost and convergence characteristics, as shown by the simulation results.

### Introduction

Customers' requests for electrical energy must be met promptly and efficiently, as mandated by national legislation, by the vast majority of the world's electric power companies. Despite meeting the country's power needs, the utility must also guarantee that the electricity is produced at the lowest possible cost. This means that the entire demand must be distributed among the generating units in a way that reduces the system's overall generation cost while still meeting the economic needs of the system. There are several ways to calculate how much power is created by each committed producing unit in order to keep overall costs down while still meeting demand for electricity.

Economic dispatch is one such method. "The operation of generating facilities to provide energy at the lowest cost to reliably supply customers, recognising any operational restrictions of generation and transmission infrastructure" might be termed as "economic dispatch". Allocating generating among committed units in order to meet limits and reduce energy consumption in terms of dollars per hour is

an essential optimization job in power system operation. Figure 1 depicts a simple heat rate curve, which depicts the input-output relationship of a thermal unit (a). When the heat rate curve is converted from Btu/h to \$/h, the fuel cost curve depicted in Fig. 1 may be seen (b)

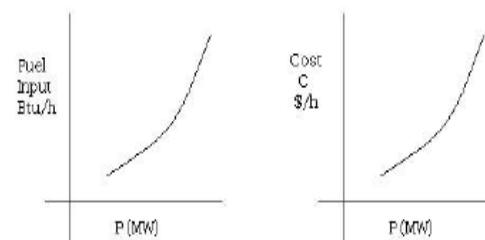


Fig. 1(a) Power Vs Fuel input

Fig. 1(b) Power Vs Cost

A variety of derivatives-based approaches, including as lambda iteration, gradient technique, Lagrangian Multiplier method, Dynamic Programming method, were previously used to tackle ELD issues. Because to valve point effect, ramp rate limits, and so on, contemporary generators' input-output characteristics are non-linear. Genetic algorithm (GA), particle swarm optimization (PSO), and artificial bee colony (ABC) optimization approaches have recently been used to tackle the ELD issue

with non-smooth cost functions, and have proved effective. The Harmony search algorithm is one of these more contemporary methods. As in improvised music, the goal of the "harmony search" (HS) algorithm is to find the best possible harmony by analysing the pitches of the individual artists involved. The process of musical improvisation resembles that of optimal design, which is concerned with arriving at the best possible solution. Harmony is defined by the pitch of each musical instrument, exactly like a collection of variables. The Upgraded Harmony Search Algorithm (IHSA) is an improved version of HS. This article discusses the IHS method for solving the ELD issue with the addition of Ramp Rate limitations.

**Problem Formulation**

The main objective of economic load dispatch problem is to minimize

$$\min f = \sum_{i=1}^N F_i(P_i) \tag{1}$$

Where  $F_i$  is the total fuel cost for the generator unity  $i$  (in \$/h), which is defined by equation:

$$F_i(P_i) = a_i P_i^2 + b_i P_i + c_i \tag{2}$$

Where  $a_i$ ,  $b_i$  and  $c_i$  are cost coefficients of generator  $i$ . Two constraints are considered in this problem, i.e., the generation capacity of each generator and the power balance of the entire power system.

Constraint 1: This constraint is an inequality constraint for each generator. For normal system operations, real power output of each generator is within its lower and upper bounds and is known as generation capacity constraint given by

$$P_{Gi}^{min} \leq P_{Gi} \leq P_{Gi}^{max} \tag{3}$$

Constraint 2: This constraint is an equality constraint. In which the equilibrium is met when the total power generation must equals the total demand  $P_D$  and the real power loss in transmission lines  $P$ . This is known as

power balance constraint can be expressed as given in

$$\sum_{i=1}^N P_G = P_D + P_L \tag{4}$$

Ramp rate limit constraint: The power generated,  $P_i$  by the  $i$ th generator in certain interval may not exceed that of previous interval  $P_{i0}$  by more than a certain amount  $UR_i$ , the up-ramp rate limit and neither may it be less than that of the previous interval by more than some amount  $DR_i$  the down ramp rate limit of the generator. These give rise to the following constraints.

As generation increases  $P_i - P_{i0} \leq UR_i$

As generation decreases  $P_{i0} - P_i \leq DR_i$

and  $\max(P_i^{min}, P_{i0} - DR_i) \leq P_i \leq \min(P_i^{max}, P_{i0} + UR_i)$  (5)

**Valve point loading constraint:**

The valve-point loading is taken in consideration by adding a sine component to the cost of the generating units. Typically, the fuel cost function of the generating units with valve-point loadings is represented in Fig.2.

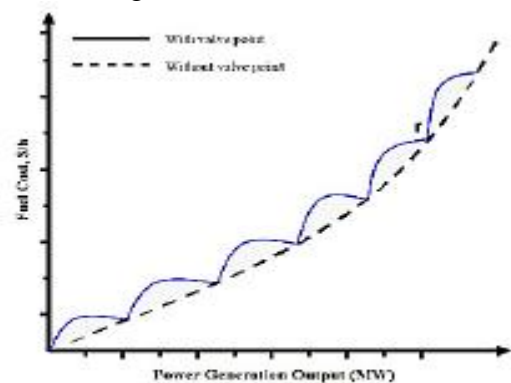


Fig. 2 Power generation output Vs Fuel cost

**Harmony Search Algorithm**

An algorithm developed by Geem that mimics the improvisation of musicians is

known as the harmony search (HS) algorithm. A musician's improvisations are equivalent to the local and global search strategies used in optimization techniques, and the harmony they create may be compared to an optimization solution vector. Instead of a gradient search, the HS method employs a stochastic random search. Harmony memory and pitch adjustment rate are used to discover the solution vector in the search space using this approach. In order to obtain the optimal value for the objective function, the HS algorithm employs the notion of how aesthetic assessment helps to reach the perfect condition of harmony. The HS method has a few parameters and is straightforward to apply conceptually and practically. A number of optimization issues have been solved using this technique. The HS algorithm's optimization approach is as follows:

- Initialize the optimization problem and algorithm parameters.
- Initialize the harmony memory.
- Improvisation of a New Harmony memory.
- Update the harmony memory.
- Check for stopping criteria. Otherwise, repeat step 3 to 4

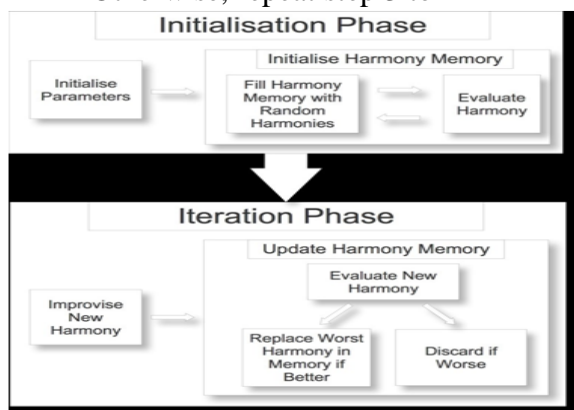


Fig. 3: Block Diagram of Harmony search Algorithm

### Implementation of the Proposed Approach

- The proposed approach to solve ELD problem is described in the

following steps.

- Input the system parameters, minimum and maximum limits of control variables.
- Choose the harmony memory size HMS, pitch adjusting rate PAR, bandwidth BW and the maximum number of improvisations NI.
- Initialize the harmony memory HM as explained in the section III-B. While initializing, all the control variables are randomly generated within their limits.

### Start the improvisation.

- ❖ For each solution vector in HM, evaluate the objective functions.
- ❖ Improvise the New Harmony memory as explained in the section III-C.

Table 1 : Comparison of fuel cost coefficients for different generators

| No. of Generators | $P_{min}$ (MW) | $P_{max}$ (MW) | A (\$/MWhr) | B (\$/MWhr) | C (\$/MWhr) |
|-------------------|----------------|----------------|-------------|-------------|-------------|
| 1                 | 50.0           | 300            | 95          | 6.8000      | 0.0070      |
| 2                 | 50.0           | 450            | 30          | 4.0000      | 0.0055      |
| 3                 | 50.0           | 450            | 45          | 4.0000      | 0.0055      |
| 4                 | 0.0            | 100            | 10          | 0.8500      | 0.0025      |
| 5                 | 50.0           | 300            | 20          | 4.6000      | 0.0060      |
| 6                 | 50.0           | 450            | 90          | 4.0000      | 0.0055      |
| 7                 | 50.0           | 200            | 42          | 4.7000      | 0.0065      |
| 8                 | 50.0           | 500            | 46          | 5.0000      | 0.0075      |
| 9                 | 0.0            | 600            | 55          | 6.0000      | 0.0085      |
| 10                | 0.0            | 100            | 58          | 0.5000      | 0.0020      |
| 11                | 50.0           | 150            | 65          | 1.6000      | 0.0045      |
| 12                | 0.0            | 50             | 78          | 0.8500      | 0.0025      |
| 13                | 50.0           | 300            | 75          | 1.8000      | 0.0050      |
| 14                | 0.0            | 150            | 85          | 1.6000      | 0.0045      |
| 15                | 0.0            | 500            | 80          | 4.7000      | 0.0065      |
| 16                | 50.0           | 150            | 90          | 1.4000      | 0.0045      |
| 17                | 0.0            | 100            | 10          | 0.8500      | 0.0025      |
| 18                | 50.0           | 300            | 25          | 1.6000      | 0.0045      |
| 19                | 100.0          | 600            | 90          | 5.5000      | 0.0080      |
| 20                | 120.0          | 500            | 18          | 6.7000      | 0.0020      |

- ❖ Perform the non-dominated sorting and ranking on the combined existing and New Harmony memory
- ❖ Choose the best harmony memory from the combined solution vectors as given in the section III-D for the next improvisation.
- ❖ Check for stopping conditions. If the

number of improvisations has been reached stop the algorithm. Otherwise, go to step 5.

Table 2 : Comparison of BBO, HS, IHS with ramp rate limit

| LOAD DEMAND(MW) | BBO TOTAL COST(\$/hr) | HSA TOTAL COST(\$/hr) | IHSA TOTAL COST(\$/hr) |
|-----------------|-----------------------|-----------------------|------------------------|
| 925             | 1232.90               | 1020.56               | 846.324                |
| 1000            | 1438.00               | 1120.12               | 989.94                 |
| 1500            | 2487.52               | 2332.41               | 2234.32                |
| 2000            | 4257.37               | 4167.8                | 3979.26                |
| 2500            | 6477.52               | 6421.93               | 6224.65                |

**Biogeography Based Optimisation Algorithm**

BBO, suggested by Dan Simon in 2008, is a stochastic optimization technique for solving multimodal optimization problems. It is based on the concept of biogeography, which deals with the distribution of species that depend on different factors such as rain fall, diversity etc. The main parts of BBO algorithm includes Migration

**The algorithm steps of BBO are as follows**

- Step 1: Initialization of the BBO parameters.
- Step 2: The initial position of SIV of each habitat should be randomly selected while satisfying different equality and inequality constraints of ELD problems. Several numbers of habitats depending upon the population size are being generated. Each habitat represents a potential solution to the given problem.
- Step 3: Calculate each HSI i.e. value of objective function for each i-th habitat of the population set n for given emigration rate  $\mu_s$ , immigration rate  $\lambda_s$  and species S.
- Step 4: Based on the HSI values some elite habitats are identified.

- Step 5: Each non-elite habitat is modified by performing probabilistically immigration and emigration operation.
- Step 6: Species count probability of each habitat is updated using equation 11. Mutation operation is performed on the non-elite habitat and HSI value of each new habitat is computed.
- Step 7: Feasibility of a problem solution is verified i.e. each SIV should satisfy equality and inequality constraints.
- Step 8: Go to step 3 for the next iteration.
- Step 9: Stop iterations after a predefined number of iterations.

**Simulation Results**

Using the 20-generator test system, the fuel cost coefficients and generation limitations for each generator are shown in Table 1. Harmony Search algorithm (HSA), IHSA and Biogeography based optimization algorithm (BBO) are compared in a simulation and the results are displayed in Table 2. With regard to valve point loading, the findings are shown in Table 3.

Table 3 : Comparison of BBO, HSA, IHS with valve point loading

| LOAD DEMAND(MW) | BBO TOTAL COST(\$/hr) | HS TOTAL COST(\$/hr) | IHS TOTAL COST(\$/hr) |
|-----------------|-----------------------|----------------------|-----------------------|
| 925             | 872.425               | 838.14               | 687.79                |
| 1000            | 1036.84               | 1016.54              | 976.98                |
| 1500            | 2266.82               | 2157.54              | 1004.87               |
| 2000            | 4016.83               | 3354.98              | 2900.889              |
| 2500            | 6266.83               | 5735                 | 4089.13               |

**RAMP RATE CONSTRAINT:**

The convergence characteristics obtained for all the three algorithms with the inclusion of ramp rate limit constraint is shown in Fig. 4. The comparison of fuel cost with ramp rate is shown in Fig. 5.

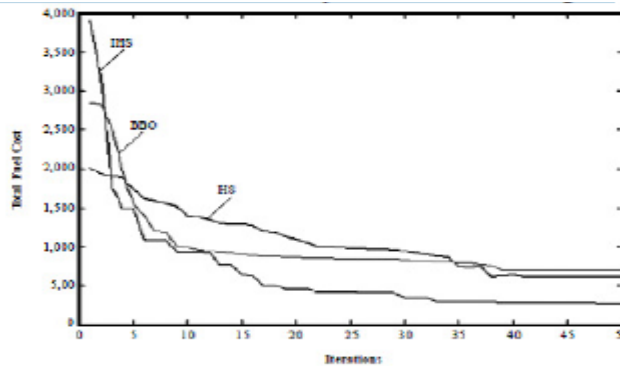


Fig. 4 Convergent characteristic of all algorithms

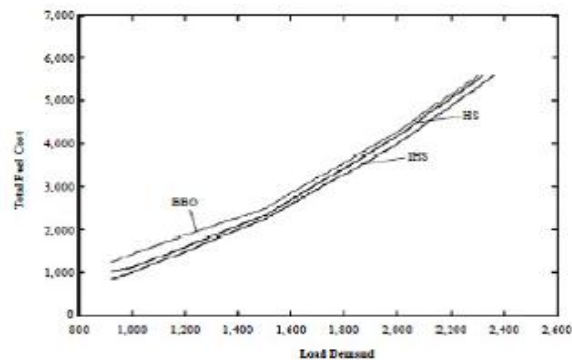


Fig 5 Comparison of Fuel Cost Solution

**VALVE POINT LOADING CONSTRAINT:**

The convergence characteristics obtained for all the three algorithms with the inclusion of valve point loading constraint is shown in Fig. 6. The comparison of fuel cost with valve point loading is shown in Fig. 7.

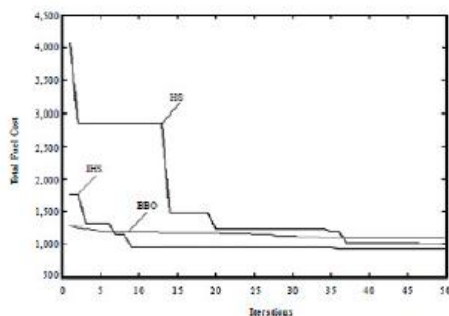


Fig. 6 Convergence Characteristics Between IHS, HS, BBO with Valve point loa

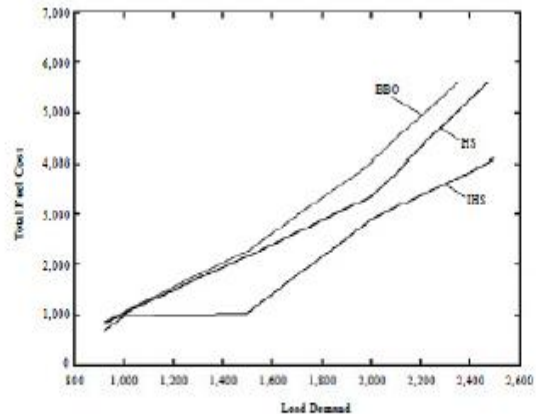


Fig. 7 Comparison of fuel cost from the table with valve point loading

**Conclusion**

It is possible to solve the economic load dispatch issue in the power system by applying the IMPROVED HARMONY SEARCH ALGORITHM (IHS), the HARMONY SEARCH ALGORITHM (HSA), and the BIOGEOGRAPHY BASED OPTIMIZATION ALGORITHM (BBO). The simulation results show that the IHS algorithm outperforms both the HS and BBO algorithms in terms of performance and overall fuel cost. HS algorithm may be well-known, but it is considered innovative and imaginative when it is compared to BBO and IHS under restrictions of ram prate and valve point loading in order to save fuel costs.

Because of these two factors, the Improved Harmony Search algorithm is superior than previous algorithms.

- ✓ When compared to alternative options, the total fuel cost is the lowest.
- ✓ When compared to others, the reaction time of convergence characteristics is very rapid.

**References**

[1] Leandro dos Santos Coelho, Viviana Cocco Mariani, An improved harmony

search algorithm for power economic load dispatch, *Energy Conversion and Management*, Elsevier, 50 (2009), p. 2522–2526, May 2009.

[2]M. Mahdavi, M. Fesanghary, E. Damangir, An Improved Harmony Search Algorithm for Solving Optimization Problems, Elsevier, *Applied Mathematics and Computation*, Vol. 188, p. 1567-1579, 2007.

Geem ZW, Kim JH, Loganathan GV, A new heuristic optimization algorithm: Harmony search. *Simulation* 2001; 76: 60-68.

Hadi Sadat, “Power System Analysis”, WCB/McGraw-Hill, 1999.

D.P. Kothari and J.S. Dhillon, *Power System Optimization*, Prentice-Hall of India, 2006.

Lee KS, Geem ZW. A new meta-heuristic algorithm for continuous engineering optimization: harmony search theory and practice. Elsevier, *Computational Methods Appl. Mech. Eng* 2005; 194(36–38):p.3902–33.

J. Mahdavi M, Fesanghary M and Damangir E (2007), An Improved Harmony Search Algorithm for Solving Optimization Problems, Elsevier, *Applied Mathematics and Computation*, Vol. 188, p. 1567-1579.

Bhattacharya, P.K. Chattopadhyay, Biogeography-based optimization for different economic load dispatch problems, *IEEE Trans. on Power Systems*, Vol. 25, p. 1064- 1077, May 2010.

L.D.S. Coelho, V.C. Mariani, Combining of chaotic differential evolution and quadratic programming for economic dispatch optimization with valve point effect, *IEEE Trans. on Power Systems*, Vol. 21, p. 989–996, May 2006.