

IEEE PSCE 2006



**Application of Particle Swarm Optimization to Economic
Dispatch Problem: Advantages and Disadvantages**

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October 31, 2006



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Introduction (1)

□ Overview of Particle Swarm Optimization (1)

- **Branch of Swarm Intelligence**
 - **Swarm Intelligence Techniques : PSO (Eberhart and Kennedy, 1995), Ant Colony Optimization (Dorigo, 1992), Stochastic Diffusion Search (Bishop, 1989), etc.**
- **Basic Characteristics of PSO**
 - **Relatively recently developed Heuristic AI-Optimization technique**
 - **Analogy of school of fish and flock of bird**
 - **Population based parallel search algorithm**
 - **Each particle or agent searches, based on its own experience and its neighbor's experiences**



Introduction (2)

□ Overview of Particle Swarm Optimization (1)

- Advantages

- Simple concept
- Easy implementation
- Robustness to control parameters
- Computational efficiency, etc.



Introduction (3)

□ Representation of each particle in optimization problems with n -dimensional search space

- Position and velocity of particle- i :

$$X_i = (x_{i1}, \dots, x_{in}) \quad V_i = (v_{i1}, \dots, v_{in})$$

- Best position of particle- i : $Pbest_i = (x_{i1}^{Pbest}, \dots, x_{in}^{Pbest})$

- Its neighbor's best position: $Gbest = (x_1^{Gbest}, \dots, x_n^{Gbest})$



Introduction (4)

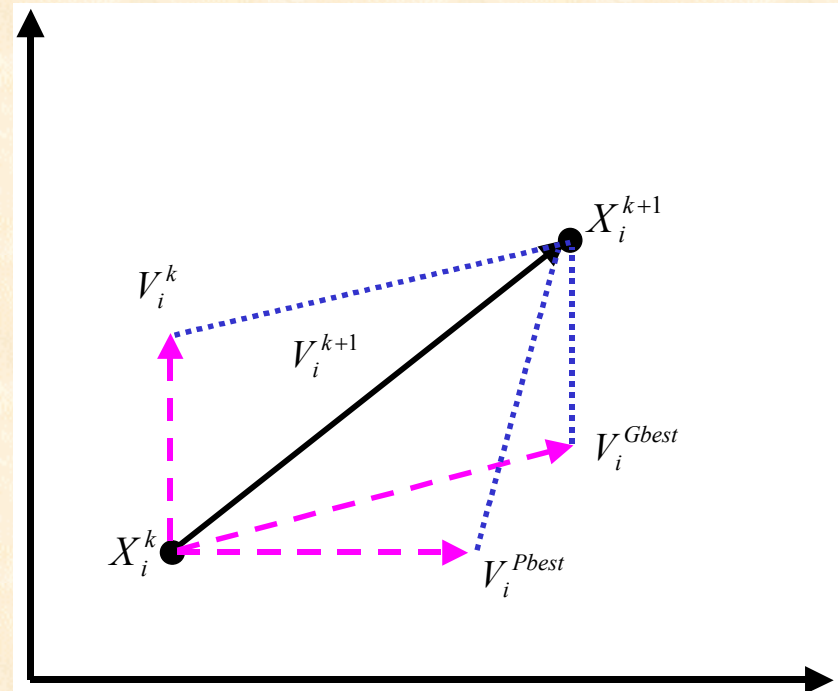
□ General update rule (update of velocity and position)

- Velocity update of particle-*i*:

$$V_i^{k+1} = \omega V_i^k + c_1 r_1 \times (Pbest_i^k - X_i^k) + c_2 r_2 \times (Gbest^k - X_i^k)$$

- Position update of particle-*i*:

$$X_i^{k+1} = X_i^k + V_i^{k+1}$$





Introduction (5)

□ Control Parameters

- Acceleration coefficients (c_1, c_2)
 - Constant values where normally $c_1 + c_2 = 4.0$ and $c_1 = c_2$ (Ozcan and Mohan, 1999)
 - Problem dependent coefficients
- Inertia weight factor (ω)
 - Constant approach : Suggested by Shi (1998) and Eberhart (1999)
 - Dynamically changing approach (linearly decreasing case): changing from exploring mode to exploiting mode

$$\omega = \omega_{\max} - \frac{\omega_{\max} - \omega_{\min}}{Iter_{\max}} \times Iter$$



Introduction (6)

- ❑ **Overview of Economic Dispatch Problems and Application Techniques**
 - **Nonlinear and non-smooth optimization problems when considering valve-point effects, multi-fuels, prohibited operating zones, etc.**
 - **Limitation of mathematical-based approaches (gradient information)**
 - **Limitation of Dynamic Programming (Curse of Dimensionality)**



Introduction (7)

- ❑ **Overview of Economic Dispatch Problems and Application Techniques**
 - **Alternatives (Heuristic AI-based techniques)**
 - **Genetic Algorithms (GAs)**
 - **Evolutionary Programmings (EPs)**
 - **Tabu Searches (TPs)**
 - **Neural Networks (NNs)**
 - **Particle Swarm Optimizations (PSOs)**



ED Problem Formulations (1)

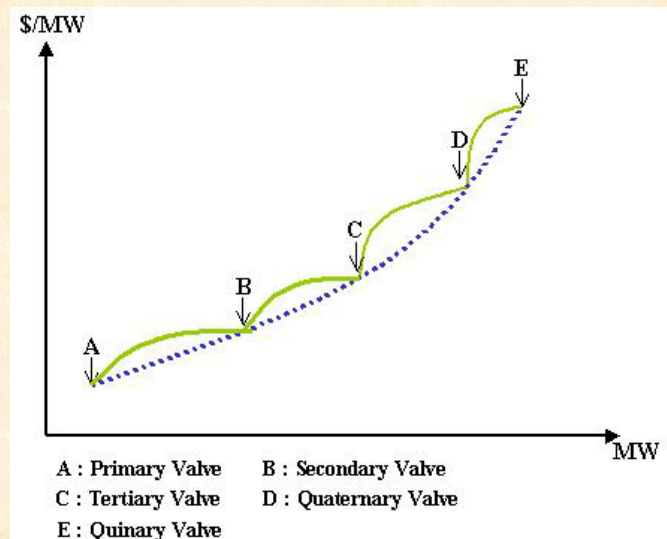
□ Conventional quadratic cost function

- Objective function (analytical function)

$$F_T = \sum_{i=1}^N F_i(P_i) \quad F_i(P_i) = a_i + b_i P_i + c_i P_i^2$$

□ Valve-point effects

$$F_i(P_i) = a_i + b_i P_i + c_i P_i^2 + \left| e_i \times \sin(f_i \times (P_{i,\min} - P_i)) \right|$$



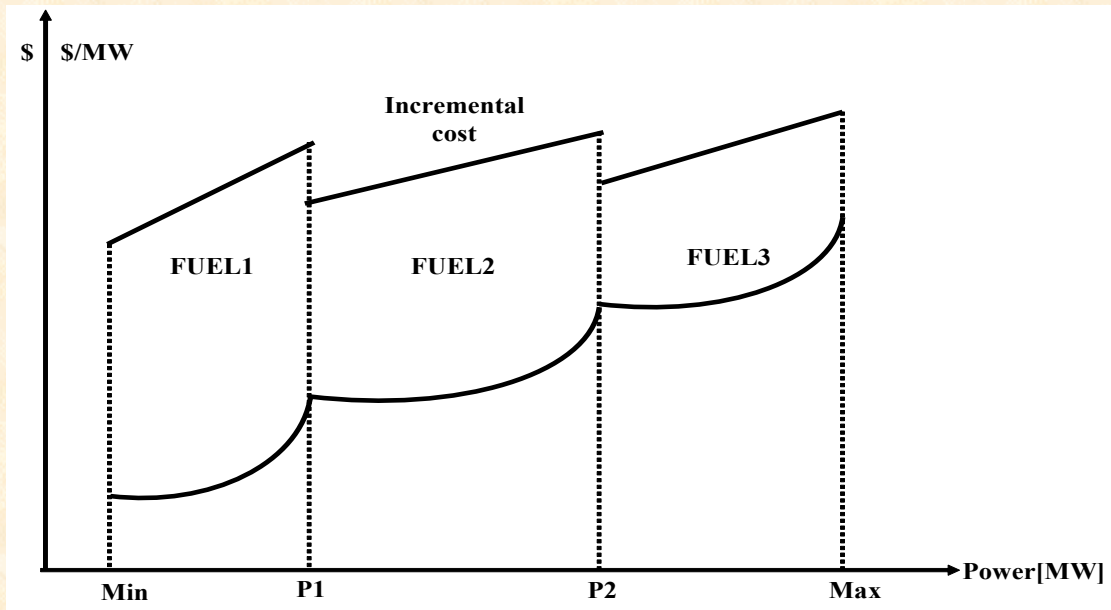


ED Problem Formulations (2)

□ Cost function with multiple fuels

- Objective function

$$F_i(P_i) = \begin{cases} a_{i1} + b_{i1}P_i + c_{i1}P_i^2 & \text{if } P_{i\min} \leq P_i \leq P_{i1} \\ a_{i2} + b_{i2}P_i + c_{i2}P_i^2 & \text{if } P_{i1} \leq P_i \leq P_{i2} \\ \vdots & \vdots \\ a_{in} + b_{in}P_i + c_{in}P_i^2 & \text{if } P_{in-1} \leq P_i \leq P_{i\max} \end{cases}$$





ED Problem Formulations (3)

□ Constraints

- Equality constraint (active power balance equation)

$$\sum_{i=1}^N P_i = P_D + P_{Loss}$$

- Inequality constraints (1)

- Minimum and Maximum power limits: $P_{i,\min} \leq P_i \leq P_{i,\max}$

- Ramp rate limits: $P_i - P_i^0 \leq DR_i$ $P_i - P_i^0 \leq UR_i$

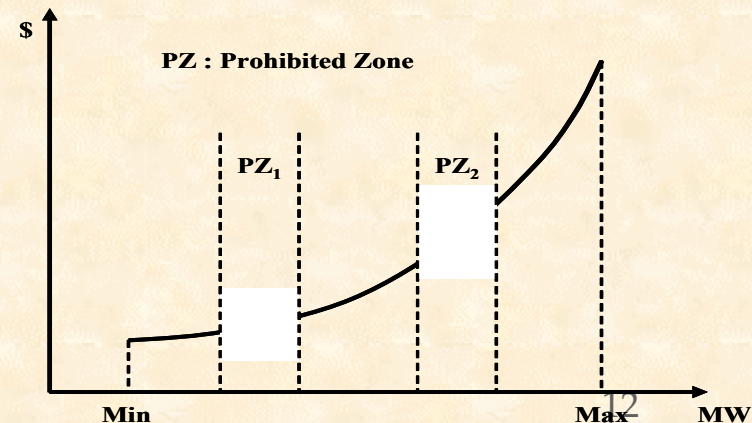
- Spinning reserve requirements: $\sum_{i=1}^N \{ \text{Min} (P_{i,\max} - P_i, UR_i) \} \geq S_R$

- Generator prohibited operating zones:

$$P_i \in \begin{cases} P_{i,\min} \leq P_i \leq P_{i,1}^l \\ P_{i,k-1}^u \leq P_i \leq P_{i,k}^l, & k = 2, 3, \dots, N_{PZ,i} \\ P_{i,N_{PZ,i}}^u \leq P_i \leq P_{i,\max} \end{cases}$$

- Line flow constraints:

$$|P_{Lf,k}| \leq P_{Lf,k}^{\max}$$





Survey of PSO Applications to ED Problems (1)

❑ Practical ED Problems

- Objective function
 - Cost functions with valve-point and multi-fuel impacts
- Constraints
 - Nonlinear equality and inequality constraints

❑ Park, Lee, Shin & Lee (2005)

- Non-smooth cost functions - valve-point and multi-fuel effects
- Suggested a modified PSO scheme, employing constraint treatment technique and dynamic search space reduction strategy
- Parameter Values
 - Valve-point effect problems: population = 20, initial inertia weight (w_{\max}) = 1.0, final inertia weight (w_{\min}) = 0.5, acceleration coefficients = 2.0
 - Multi-fuel effect problems: population = 30, initial inertia weight (w_{\max}) = 0.5, final inertia weight (w_{\min}) = 0.1, acceleration coefficients = 2.0



Survey of PSO Applications to ED Problems (2)

- ❑ **El-Gallad, et al. (2002)**
 - Added the constraints of system spinning reserve and generator prohibited operating zones. Applied the conventional PSO technique
- ❑ **Giang (2003)**
 - Additional constraints of generator ramp rate limits on the El-Gallad's
 - PSO technique with the fitness function values in [0.1]
 - Parameter Values
 - population = 100, initial inertia weight (w_{\max}) = 0.9, final inertia weight (w_{\min}) = 0.4, acceleration coefficients = 2.0
- ❑ **Giang (2004)**
 - Dynamic EDs considering spinning reserve, ramp rate, prohibited operating zones, and line flow limits.



Survey of PSO Applications to ED Problems (3)

❑ Victoire and Jeyakumar (2005)

- Solved dynamic ED problems combining PSO and Sequential Quadratic Programming (SQP)
 - Main Optimizer: PSO, Local Optimizer: SQP
- Parameter Values
 - population = 100, initial inertia weight (w_{\max}) = 1.3, final inertia weight (w_{\min}) = 0.7, acceleration coefficients = 2.0



Survey of PSO Applications to ED Problems (4)

❑ Comparison of simulation results (1)

- 40-generator system with valve-point effects - demand is 10,500MW

	CEP	FEP	MFEP	IFEP	PSO
Min. Cost	123488.29	122679.71	122647.57	122624.35	122252.27

CEP (Classical Evolutionary Programming), FEP (fast EP), MFEP (modified FEP), IFEP (improved FEP)

● References

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- Sinha, R. Chakrabarti, and P. K. Chattopadhyay, “Evolutionary programming techniques for economic load dispatch,” *IEEE Trans. on Evolutionary Computations*, Vol. 7, No. 1, pp. 83-94, Feb. 2003.
- J. B. Park, K. S. Lee, J. R. Shin, and K. Y. Lee, “A particle swarm optimization for economic dispatch with nonsmooth cost functions,” *IEEE Trans. on Power Systems*, Vol. 20, No. 1, pp. 34-42, Feb. 2005.



Survey of PSO Applications to ED Problems (5)

❑ Comparison of simulation results (2)

- 15-generator system with prohibited operating zones

	GA	PSO
Total power output [MW]	2668.4	2662.4
Total line loss [MW]	38.2782	32.4306
Total generation cost [\$]	33,113	32,858

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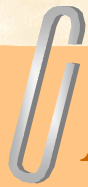


Survey of PSO Applications to ED Problems (6)

- ❑ Comparison of simulation results (3)
 - 10-generator system with multi-fuels

S	U	MHNN		AHNN		IEP		PSO	
		F	GEN	F	GEN	F	GEN	F	GEN
1	1	2	224.5	2	225.7	2	219.5	2	218.3
	2	1	215.0	1	215.2	1	211.4	1	211.7
	3	3	291.8	1	291.8	1	279.7	1	280.7
	4	3	242.2	3	242.3	3	240.3	3	239.6
2	5	1	293.3	1	293.7	1	276.5	1	278.5
	6	3	242.2	3	242.3	1	239.9	3	239.6
	7	1	303.1	1	302.8	1	289.0	1	288.6
3	8	3	242.2	3	242.3	3	241.3	3	239.6
	9	1	355.7	1	355.1	3	425.1	3	428.5
	10	1	289.5	1	288.8	1	277.2	1	274.9
TP			2699.7		2700.0		2700.0		2700.0
TC			626.12		626.240		623.851		623.809

- References
 - J. B. Park, K. S. Lee, J. R. Shin, and K. Y. Lee, "A particle swarm optimization for economic dispatch with nonsmooth cost functions," *IEEE Trans. on Power Systems*, Vol. 20, No. 1, pp. 34-42, Feb. 2005.



Advantages and Disadvantages of PSO (1)

- ❑ PSO-based approach is considered as the one of the most powerful methods for resolving the non-smooth global optimization problems
- ❑ Main Advantages of PSO-based Approaches
 - A derivative-free technique
 - Easy in its concept and coding implementation
 - Less sensitive to the nature of the objective function
 - Limited number of parameters. Also, less sensitive to parameters
 - Less dependent on initial points
 - Generate high-quality solutions with shorter calculation time and stable convergence characteristics



Advantages and Disadvantages of PSO (2)

❑ Disadvantages of PSO-based Approaches

- Lacking somewhat of a solid mathematical foundation for analysis,
- Some limitations in real-time ED applications, such as in the 5-minute dispatch with network constraints, due to relatively longer computation time (Possibility for the off-line real-world ED problems such as in the Day-ahead electricity markets)
- Still having the problems of dependency on initial conditions, parameter values, difficulty in finding the optimal design parameters, stochastic characteristics of the final outputs



Conclusions

- ❑ This discussion is to survey and summarize the applications of PSO to ED problems including the advantages and disadvantages of the approaches
- ❑ PSO techniques
 - Much attention in various application areas including ED problems
 - Outperforming other existing heuristic optimization techniques as well as mathematical methods on the non-smooth ED problems
 - Still needing further research and development to improve its performance and to obtain robustness
 - For the application to real-world on-line ED problems, it is necessary to combine the conventional mathematical approach with the PSO methods