

IMPROVED COVARIANCE MATRIX EVOLUTION STRATEGY ALGORITHM
FOR STOCHASTIC DYNAMIC UNEQUAL AREA FACILITY LAYOUTS IN AN
OPEN AREA

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requirements for the award of the degree of
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DEDICATION

To my Parents and my wife

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ABSTRACT

Facility layout problems deal with layout of facilities, machines, cells, or departments in a shop floor. This research has formulated unequal area stochastic dynamic facility layout problems in an open or wall-less area in order to minimize the upper bound of the sum of the material handling costs, and the sum of the shifting costs in the whole time planning horizon. In addition, the areas and shapes of departments are fixed during the iteration of an algorithm and throughout the time horizon. In unequal area stochastic dynamic facility layout problems, there are several periods for the material flow among departments or product demand such that the material flow among departments or product demand is not stable in each period. In other words, the product demand is stochastic with a known expected value and standard deviation in each period. In this research, a new mixed integer nonlinear programming mathematical model was proposed for solving this type of problems. Particularly, they are non-deterministic polynomial-time hard and very complex, and exact methods could not solve them within a reasonable computational time. Therefore, meta-heuristic algorithms and evolution strategies are needed to solve them. In this research, a modified covariance matrix adaptation evolution strategy algorithm was developed and the results were compared with two improved meta-heuristic algorithms (improved particle swarm optimization and modified genetic algorithm). These two meta-heuristic algorithms were developed and used to justify the efficiency of the proposed evolution strategy algorithm. The proposed algorithms applied four methods, which are (1) department swapping method, (2) local search method 1, (3) period swapping method, and (4) local search method 2, to prevent local optima and improve the quality of solutions for the problems. The proposed algorithms and the proposed mathematical model were validated using manual and graphical inspection methods, respectively. The trial and error method was applied to set the respective parametric values of the proposed algorithms in order to achieve better layouts. A real case and a theoretical problem were introduced to test the proposed algorithms. The results showed that the proposed covariance matrix adaptation evolution strategy has found better solutions in contrast to the proposed particle swarm optimization and genetic algorithm.

ABSTRAK

Masalah susun atur kemudahan adalah berkaitan dengan susun atur kemudahan, mesin, sel atau jabatan di lantai pengeluaran. Kajian ini telah memformulasikan masalah susun atur kemudahan stokastik dinamik ketidaksamaan saiz di kawasan terbuka atau tidak berdinging untuk meminimumkan sempadan atas bagi jumlah kos pengendalian bahan, dan jumlah kos peralihan dalam keseluruhan tempoh masa perancangan. Di samping itu, saiz dan bentuk jabatan adalah tetap semasa lelaran algoritma dan sepanjang tempoh masa. Dalam masalah susun atur kemudahan stokastik dinamik ketidaksamaan saiz, terdapat beberapa tempoh masa untuk aliran bahan antara jabatan atau permintaan produk di mana aliran bahan antara jabatan atau permintaan produk adalah tidak stabil pada setiap tempoh masa. Dalam kata lain, permintaan produk adalah stokastik dengan nilai jangkaan dan sisihan piawai yang diketahui dalam setiap tempoh masa. Dalam kajian ini, satu model matematik baru pengaturcaraan integer campuran tak linear telah dicadangkan untuk menyelesaikan jenis masalah ini. Khususnya, masalah ini adalah masa polinomial keras tidak ketentuan dan sangat kompleks, dan kaedah tepat tidak dapat menyelesaikan masalah ini dalam tempoh masa yang munasabah. Jadi, algoritma meta-heuristik dan strategi evolusi diperlukan untuk menyelesaikannya. Dalam kajian ini, satu algoritma strategi evolusi adaptasi matrik kovarians yang diubahsuai telah dibangunkan dan hasilnya telah dibandingkan dengan dua algoritma meta-heuristik yang diperbaiki (iaitu pengoptimuman partikel berkumpulan dan algoritma genetik yang diperbaiki). Dua algoritma meta-heuristik ini dibangunkan dan digunakan untuk menjustifikasikan kecekapan algoritma strategi evolusi. Algoritma-algoritma yang dicadangkan menggunakan empat kaedah iaitu (1) kaedah penukaran jabatan (2) kaedah pencarian setempat 1 (3) kaedah penukaran tempoh dan (4) kaedah pencarian setempat 2, untuk mengelakkan penyelesaian optima setempat dan memperbaiki kualiti penyelesaian bagi masalah ini. Algoritma-algoritma yang dicadangkan telah disahkan menggunakan kaedah manual manakala model matematik telah disahkan menggunakan kaedah penyemakan graf. Kaedah cuba-cuba telah digunapakai untuk menentukan nilai parameter bagi algoritma yang dicadangkan demi mencapai susun atur yang lebih baik. Satu kes sebenar dan masalah teoritik telah diperkenalkan untuk menguji algoritma-algoritma yang dicadangkan. Hasil kajian menunjukkan bahawa strategi evolusi adaptasi matrik kovarians menghasilkan penyelesaian yang lebih baik jika dibandingkan dengan pengoptimuman partikel berkumpulan dan algoritma genetik.

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LIST OF ABBREVIATIONS

FLPs	-	Facility layout problems
SFLPs	-	Static facility layout problems
STFLPs	-	Stochastic facility layout problems
DFLPs	-	Dynamic facility layout problems
STDFLPs	-	Stochastic dynamic facility layout problems
UA SFLPs	-	Unequal area static facility layout problems
EA SFLPs	-	Equal area static facility layout problems
UA STFLPs	-	Unequal area stochastic facility layout problems
EA STFLPs	-	Equal area stochastic facility layout problems
UA DFLPs	-	Unequal area dynamic facility layout problems
EA DFLPs	-	Equal area dynamic facility layout problems
UA STDFLPs	-	Unequal area stochastic dynamic facility layout problems
EA STDFLPs	-	Equal area stochastic dynamic facility layout problems
CMA ES	-	Covariance matrix adaptation evolution strategy
PSO	-	Particle swarm optimization
GA	-	Genetic algorithm
NP-hard	-	Non-deterministic polynomial-time hard
STDFLP-RE	-	The real case
STDFLP-ONE	-	The theoretical problem instance

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CHAPTER 1

INTRODUCTION

1.1 An Overview

The foundation of this research is explained in this chapter. The background of this research is provided in section 1.2 and it is concerning static facility layout problems (FLPs), dynamic FLPs, stochastic FLPs, and stochastic dynamic FLPs. The problem is clearly stated in section 1.3 and the next section (section 1.4) is allocated to the objectives of the research. Questions and scope of the research are identified in sections 1.5 and 1.6, respectively. Sections 1.7 and 1.8 state the significance and contributions of this research, respectively. Finally, the structure of this thesis is mentioned in section 1.9.

1.2 Background of the Research

Facility layout is a strategy that has been widely utilized in many countries in order to decrease the total operating cost. According to Tompkins et al. (2010), facility design will be one of the most momentous areas in the manufacturing environment in the future. The main aim of FLPs is to find the best position or layout for facilities, departments, cells, and machines in a given area in order to reduce the total operating cost within manufacturing environments.

Particularly, FLPs cope with finding the locations of facilities, machines, or departments in a shop floor in order to minimize the sum of the material handling costs

among them. In static FLPs (SFLPs), the product demands are stable and cannot change for a long duration. It means the material flow among facilities, departments, cells, or machines is fixed throughout the entire planning horizon. SFLPs where the shapes and areas of all facilities or departments are same were introduced by Koopmans and Beckmann (1957) for the first time. Armour and Buffa (1963) further developed this type of problems and formulated SFLPs where the shapes and areas of all departments or facilities could be different. They assumed that the shapes of facilities or departments could change during the iteration of an algorithm whereas their areas could not change.

SFLPs where the shapes and areas of different departments could be different, but both their shapes and areas were fixed during the iteration of an algorithm were developed by Imam and Mir (1989). SFLPs are not suitable in some industries because they assume the product demands are fixed and constant. In reality, managers of companies must be able to respond quickly to changes in product demand and product price. Hence, dynamic FLPs (DFLPs), stochastic FLPs (STFLPs), and stochastic dynamic FLPs (STDFLPs) have been studied in which the product demands are not fixed and could be varied. In DFLPs, there are several periods (periods could be weeks, months, seasons, or years) for the product demands. In this type of problems, the material flow among facilities or departments can be changed in different periods but is fixed in each period. It is clear that there are several layouts for a solution of a DFLP (one layout for each period).

Rosenblatt (1986) investigated DFLPs where the shapes and areas of all facilities or departments were same, for the first time. He applied an exact method in order to minimize the sum of the material handling costs among departments and the sum of the shifting costs of departments in consecutive periods. DFLPs where the shapes of different departments could be different were originally studied by Montreuil and Venkatadri (1991). They assumed that the shapes of departments could be changed during the iteration of an algorithm whereas their areas were fixed in all periods and could not change during the iteration of an algorithm and throughout the time horizon.

Dunker et al. (2005) developed DFLPs where the shapes of various departments could be different. They assumed that the shapes and areas of departments could not change during the iteration of an algorithm and were fixed throughout the time horizon, whereas the departments have free orientations (the length and width of departments could be exchanged).

In STFLPs, there is only one period such that the product demands are uncertain. There are two types of STFLPs, which can either have: 1) stochastic product demands with a known variance and expected value or 2) several scenarios for product demands with different probabilities such that their summation is equal to one. The second type of STFLPs where the shapes of all departments were same was formulated by Rosenblatt and Kropp (1992) for the first time. The first type of STFLPs where the shapes of different departments could be different was originally formulated by Kulturel-Konak et al. (2004). They assumed that the product demands were stochastic with a known expected value and standard deviation. In addition, the shape of each department could change during the iteration of an algorithm whereas the area of each department was fixed and could not change.

In STDFLPs, there are several periods for the product demands such that the product demands are not stable and are uncertain in each period. There are two types of STDFLPs based on literature review. The first is STDFLPs in which the product demands are stochastic with a known variance and expected value in each period. The second is STDFLPs in which there are several scenarios for the product demands with different probabilities in each period such that their summation is equal to one in each period. Kouvelis and Kiran (1991) investigated the second type of STDFLPs where the shapes of all departments were same. Yang and Peters (1998) formulated the second type of STDFLPs where the shapes of different machines could be different. They assumed that the shapes and areas of machines were fixed during the iteration of an algorithm but each machine has free orientations (the length and width of machines could be exchanged). The first type of STDFLPs where the shapes of all facilities were same was firstly formulated by Moslemipour and Lee (2012).

This research is going to focus on STDFLPs where the areas and shapes of different departments could be different in order to minimize the sum of the material handling costs among departments and the sum of the shifting costs of departments in consecutive periods. In addition, each product demand is normally distributed with a known expected value and standard deviation in each period. An efficient method will be developed to solve the problems in this area.

1.3 Problem Statement

Having done a vast literature review, the problem studied in this research can be discussed from several perspectives. Firstly, SFLPs are not suitable in today's competitive market because companies must be able to respond rapidly to changes in product demand, product price, production volume and product mix. Therefore, it is necessary to focus on STDFLPs.

Secondly, in the real world, it is not practical to change the locations of departments in consecutive periods for dynamic problems because departments have walls and their shifting is very costly. Hence, it is necessary to address the problems where the facilities do not have walls such as departments in an open or wall-less area. In dynamic problems, the sum of the shifting costs of departments in an open or wall-less area in consecutive periods is certainly lower because there is no wall for departments in an open area. Therefore, departments in an open or wall-less area will be arranged in each period in this research.

Thirdly, research in the field of STDFLPs is very scarce based on literature review. Moreover, studies on this type of problems where the shapes and areas of all facilities or departments are same are rarely practical in today's competitive global marketplace. In addition, any change in shapes and areas of facilities or departments during the iteration of an algorithm is costly and managers aim to cut down the total operating cost in all situations. Fourthly, most of the data for product demands are normally distributed for stochastic problems in the real world (Casella and Berger, 2002).

Therefore, it is important to study STDFLPs where the areas and shapes of different departments can be different in an open or wall-less area, and the shapes and areas of departments are fixed during the iteration of an algorithm. In addition, the product demands are normally distributed with a known expected value and standard deviation in each period and departments have free orientations (the length and width of departments can be exchanged). This type of problems has not been studied and formulated until now.

Finally, the problems stated earlier are very complex and non-deterministic polynomial-time hard (NP-hard); hence, exact methods cannot solve them within a reasonable computational time. In addition, Yildiz and Solanki (2012) reported that researchers must choose a powerful algorithm to find appropriate layouts for FLPs. This research is going to develop a modified covariance matrix adaptation evolution strategy algorithm to solve the problems as it is an emerging new technique which has not been applied in the field of FLPs.

1.4 Research Objectives

The objectives of this research are defined based on literature review, background of study, and statement of problem. The main objectives of this research are mentioned as follows:

1. To formulate a new mathematical model for STDFLPs with unequal area departments in an open or wall-less area.
2. To develop a modified covariance matrix adaptation evolution strategy algorithm to solve this type of problems.

1.5 Research Questions

Responding to these questions will help to formulate and solve STDFLPs with unequal area departments in an open or wall-less area.

1. Which model (linear programming, nonlinear programming, mixed integer linear programming, or mixed integer nonlinear programming) will be used to formulate the problems as mentioned earlier?
2. Which solution representation (continuous solution representation or discrete solution representation) is suitable to represent the solutions of the problems stated earlier?
3. Is an evolution strategy algorithm better than other meta-heuristics in solving this type of problems?

1.6 Scope of the Research

This research is bounded by the following scopes.

1. A modified covariance matrix adaptation evolution strategy algorithm will be developed to solve STDFLPs. This is because it is one of the strongest evolution strategy algorithms for solving combinatorial optimization problems. The results of this algorithm will be compared with two meta-heuristic algorithms (modified particle swarm optimization and genetic algorithm), which will also be developed in this research. This comparison is made to show the efficiency of the proposed evolution strategy algorithm.
2. MATLAB R2013a will be used in order to code the problem as mentioned earlier.

3. A case study in Iran and a theoretical problem instance will be considered for the collection of required data in the field of STDFLPs.
4. Rectilinear distance or city block distance will be used for calculating distance between two departments.
5. Planning horizon will be divided into several periods.
6. Rectangular shapes are used for departments.
7. The shape and area of each department are fixed during the iteration of an algorithm and throughout the time horizon.
8. The orientation of each department can change (the width and length of a department can exchange) during the iteration of an algorithm and throughout the time horizon.
9. An open given area without walls is assigned to locate the departments.
10. Each product demand is normally distributed with a known expected value and standard deviation in each period.

1.7 Significance of the Research

In accordance with Tompkins et al. (2010), FLPs are one of the significant areas in the field of manufacturing. In addition, they stated that roughly 8 percent of the gross national product has been spent on new facilities in the United States since 1955 and over \$300 billions have been spent each year for layout and relayout. Krishnan et al. (2008) stated that between 20 to 50 percent of the total operating costs in manufacturing are allocated to material handling costs and these can be lessened by 10 to 30 percent with effective and efficient layouts. This research addresses STDFLPs

with unequal area departments in an open or wall-less area which are very applicable in a volatile business environment.

Practically, by solving this type of problems, it helps managers of a company to have a more flexible and robust layout which can change in consecutive periods. This makes the company more adaptive and responsive in meeting fluctuating customer demands. Having an optimal layout in each period also helps to reduce the material handling cost which in turn will decrease the total operating cost. Moreover, work productivity and efficiency can be improved.

Theoretically, this research proposes a mathematical formulation for STDFLPs with unequal area departments in an open or wall-less area. This will be a new mathematical model as this type of problems has been neglected until now based on literature review. This research also develops and applies a modified evolution strategy algorithm as the solution technique. To date, it has not been used in the field of FLPs. It is envisaged that this technique will create better layouts in comparison with other meta-heuristic algorithms.

1.8 Structure of the Thesis

The thesis is divided into six chapters. Introduction of research as mentioned above is provided in this chapter (chapter 1), where the background, problem statement, objectives, questions, scopes, and significance of the research are explained. FLPs are briefly surveyed in the beginning of chapter 2. Then, the types of FLPs based on material flow are explained and reviewed. Next, the solution methods are investigated and finally, chapter 2 ends with analysis and conclusions. Chapter 3 is provided to show the methodology and phases of the research. The research design is divided into five phases and all phases are explained comprehensively in chapter 3.

The model for STDFLPs with the features mentioned above is formulated and the objective function is developed in chapter 4. Then, the proposed algorithms (modified covariance matrix adaptation evolution strategy, particle swarm

optimization, and genetic algorithm) for solving the problems are explained. In addition, new swapping and local search methods are presented in chapter 4. Chapter 5 is designed to discuss the results of the current research. Finally, chapter 6 is provided for conclusions and recommendations for future studies.

REFERENCES

- Ariafar, S., Ismail, N., Tang, S., Ariffin, M., and Firoozi, Z. (2011). A stochastic facility layout model in cellular manufacturing systems. *International Journal of the Physical Sciences*. 6(15), 3666-3670.
- Ariafar, S., Ismail, N., Tang, S. H., Ariffin, M. K. A. M., and Firoozi, Z. (2012). The reconfiguration issue of stochastic facility layout design in cellular manufacturing systems. *International Journal of Services and Operations Management*. 11(3), 255-266.
- Armour, G. C., and Buffa, E. S. (1963). A heuristic algorithm and simulation approach to relative location of facilities. *Management Science*. 9(2), 294-309.
- Asl, A. D., and Wong, K. Y. (2015a). Solving unequal-area dynamic facility layout problems based on slicing tree representation and simulated annealing. *Proceedings of the 2014 International Conference on Mathematics, Engineering and Industrial Applications 2014 (ICoMEIA 2014)*. 28-30 May. Penang, Malaysia: AIP, 1-5.
- Asl, A. D., and Wong, K. Y. (2015b). Solving unequal-area static and dynamic facility layout problems using modified particle swarm optimization. *Journal of Intelligent Manufacturing*. in press. doi:10.1007/s10845-015-1053-5.
- Azimi, P., and Charmchi, H. R. (2012). A new optimization via simulation approach for dynamic facility layout problem with budget constraints. *Modelling and Simulation in Engineering*. 2012(4), 1-9.
- Balakrishnan, J., and Cheng, C. H. (2000). Genetic search and the dynamic layout problem. *Computers & Operations Research*. 27(6), 587-593.
- Balakrishnan, J., Cheng, C. H., and Conway, D. G. (2000). An improved pair-wise exchange heuristic for the dynamic plant layout problem. *International Journal of Production Research*. 38(13), 3067-3077.

- Balakrishnan, J., Cheng, C. H., Conway, D. G., and Lau, C. M. (2003). A hybrid genetic algorithm for the dynamic plant layout problem. *International Journal of Production Economics*. 86(2), 107-120.
- Balakrishnan, J., Jacobs, F. R., and Venkataramanan, M. A. (1992). Solutions for the Constrained Dynamic Facility Layout Problem. *European Journal of Operational Research*. 57(2), 280-286.
- Baykasoglu, A., Dereli, T., and Sabuncu, I. (2006). An ant colony algorithm for solving budget constrained and unconstrained dynamic facility layout problems. *Omega-International Journal of Management Science*. 34(4), 385-396.
- Baykasoglu, A., and Gindy, N. N. (2001). A simulated annealing algorithm for dynamic layout problem. *Computers & Operations Research*. 28(14), 1403-1426.
- Bazaraa, M. S., and Sherali, H. D. (1980). Benders' partitioning scheme applied to a new formulation of the quadratic assignment problem. *Naval Research Logistics Quarterly*. 27(1), 29-41.
- Benjaafar, S., and Sheikhzadeh, M. (2000). Design of flexible plant layouts. *IIE Transactions*. 32(4), 309-322.
- Braglia, M., Zanoni, S., and Zavanella, L. (2003). Layout design in dynamic environments: strategies and quantitative indices. *International Journal of Production Research*. 41(5), 995-1016.
- Braglia, M., Zanoni, S., and Zavanella, L. (2005). Robust versus stable layout design in stochastic environments. *Production Planning & Control*. 16(1), 71-80.
- Casella, G., and Berger, R. L. (2002). *Statistical Inference*. (2th ed.). India: Thomson.
- Chang, M., Ohkura, K., Ueda, K., and Sugiyama, M. (2002). A symbiotic evolutionary algorithm for dynamic facility layout problem. *Proceedings of the 2002 Evolutionary Computation (CEC'02)*. 12-17 May. Honolulu, HI: IEEE, 1745-1750.
- Chen, G., and Rogers, J. (2009). Managing Dynamic Facility Layout with Multiple Objectives. *Proceedings of the 2009 Technology Management in the Age of Fundamental Change (PICMET 2009)*. 2-6 August. Portland, OR: IEEE, 1175-1184.

- Chen, G. Y. (2012). A new data structure of solution representation in hybrid ant colony optimization for large dynamic facility layout problems. *International Journal of Production Economics*. 142(2), 362-371
- Chen, T., Chen, C., and Chuang, S. (2010). A simulated annealing-based approach for dynamic facility planning. *Proceedings of the 2010 Industrial Engineering and Engineering Management (IEEM)*. 7-10 December. Macao: IEEE, 1291-1294.
- Cheng, C.-P., Liu, C.-W., and Liu, C.-C. (2000). Unit commitment by Lagrangian relaxation and genetic algorithms. *IEEE Transactions on Power Systems*. 15(2), 707-714.
- Conway, D. G., and Venkataramanan, M. A. (1994). Genetic Search and the Dynamic Facility Layout Problem. *Computers & Operations Research*. 21(8), 955-960.
- Dong, M., Wu, C., and Hou, F. (2009). Shortest path based simulated annealing algorithm for dynamic facility layout problem under dynamic business environment. *Expert Systems with Applications*. 36(8), 11221-11232.
- Dorigo, M. (2007). Ant colony optimization. *Scholarpedia*. 2(3), 1461.
- Drira, A., Pierreval, H., and Hajri-Gabouj, S. (2007). Facility layout problems: A survey. *Annual Reviews in Control*. 31(2), 255-267.
- Dunker, T., Radons, G., and Westkamper, E. (2005). Combining evolutionary computation and dynamic programming for solving a dynamic facility layout problem - Discrete optimization. *European Journal of Operational Research*. 165(1), 55-69.
- Dutta, K. N., and Sadananda, S. (1982). A multigoal heuristic for facilities design problems: MUGHAL. *International Journal of Production Research*. 20(2), 147-154.
- Emami, S., and Nookabadi, A. S. (2013). Managing a new multi-objective model for the dynamic facility layout problem. *The International Journal of Advanced Manufacturing Technology*. 68(9-12), 2215-2228.
- Engelbrecht, A. P. (2007). *Computational Intelligence*. (2th ed.). Chichester: John Wiley & Sons.
- Erel, E., Ghosh, J., and Simon, J. (2003). New heuristic for the dynamic layout problem. *Journal of the Operational Research Society*. 54(12), 1275-1282.
- Francis, R. L., McGinnis, L. F., and White, J. A. (1992). *Facility Layout and Location: An Analytical Approach*. (2th ed.). Pearson College Division: Prentice Hall.

- Gandhi, S., Khan, D., and Solanki, V. S. (2012). A Comparative Analysis of Selection Scheme. *International Journal of Soft Computing and Engineering*. 2(4), 131-134.
- Garey, M. R., and Johnson, D. S. (1979). *Computers and Intractability: An Introduction to the Theory of NP-Completeness*. (1th ed.). San Francisco: W. H. Freeman.
- Gen, M., and Cheng, R. (2000). *Genetic Algorithms and Engineering Optimization*. (7th ed.). Canada: John Wiley & Sons.
- Ghadikolaei, Y. K., and Shahanaghi, K. (2013). Multi-floor dynamic facility layout: a simulated annealing-based solution. *International Journal of Operational Research*. 16(4), 375-389.
- Gilmore, P. C. (1962). Optimal and suboptimal algorithms for the quadratic assignment problem. *Journal of the Society for Industrial & Applied Mathematics*. 10(2), 305-313.
- Glover, F. (1986). Future paths for integer programming and links to artificial intelligence. *Computers & Operations Research*. 13(5), 533-549.
- Gonçalves, J. F., and Resende, M. G. (2014). A biased random-key genetic algorithm for the unequal area facility layout problem. *European Journal of Operational Research*. 246(1), 86-107.
- Hansen, N. (2006). The CMA Evolution Strategy: A Comparing Review. In Lozano, J.A., Larrañaga, P., Inza, I., and Bengoetxea, E. (Eds.) *Towards A New Evolutionary Computation* (pp. 75-102). Berlin: Springer-Verlag.
- Hansen, N., and Ostermeier, A. (2001). Completely derandomized self-adaptation in evolution strategies. *Evolutionary computation*. 9(2), 159-195.
- Haupt, R. L., and Haupt, S. E. (2004). *Practical Genetic Algorithms*. (2th ed.). New Jersey: John Wiley & Sons.
- Heragu, S. S., and Kusiak, A. (1991). Efficient models for the facility layout problem. *European Journal of Operational Research*. 53(1), 1-13.
- Ho, Y.-C., and Moodie, C. L. (1998). Machine layout with a linear single-row flow path in an automated manufacturing system. *Journal of Manufacturing Systems*. 17(1), 1-22.
- Holland, J. H. (1975). *Adaptation in Natural and Artificial Systems: An Introductory Analysis with Applications to Biology, Control, and Artificial Intelligence*. (1th ed.). Michigan: University Michigan Press.

- Hosseini-Nasab, H., and Emami, L. (2013). A hybrid particle swarm optimisation for dynamic facility layout problem. *International Journal of Production Research*. 51(14), 4325-4335.
- Imam, M., and Mir, M. (1989). Nonlinear programming approach to automated topology optimization. *Computer-Aided Design*. 21(2), 107-115.
- Imam, M. H., and Mir, M. (1993). Automated layout of facilities of unequal areas. *Computers & Industrial Engineering*. 24(3), 355-366.
- Jolai, F., Tavakkoli-Moghaddam, R., and Taghipour, M. (2012). A multi-objective particle swarm optimisation algorithm for unequal sized dynamic facility layout problem with pickup/drop-off locations. *International Journal of Production Research*. 50(15), 4279-4293.
- Kaku, B. K., and Mazzola, J. B. (1997). A tabu-search heuristic for the dynamic plant layout problem. *INFORMS Journal on Computing*. 9(4), 374-384.
- Kennedy, J., and Eberhart, R. (1995). Particle Swarm Optimization. *Proceedings of the 1995 IEEE International Conference on Neural Networks*. 27 November-1 december. Washington: IEEE, 1942-1948.
- Kennedy, J., Kennedy, J. F., and Eberhart, R. C. (2001). *Swarm Intelligence*. (1th ed.). San Francisco: Morgan Kaufmann.
- Kirkpatrick, S. (1984). Optimization by simulated annealing: Quantitative studies. *Journal of statistical physics*. 34(5-6), 975-986.
- Kochhar, J. S., and Heragu, S. (1999). Facility layout design in a changing environment. *International Journal of Production Research*. 37(11), 2429-2446.
- Köksoy, O., and T. Yalcinoz. 2008. Robust Design using Pareto type optimization: A genetic algorithm with arithmetic crossover. *Computers & Industrial Engineering*. 55(1), 208-218.
- Komarudin, and Wong, K. Y. (2010). Applying ant system for solving unequal area facility layout problems. *European Journal of Operational Research*. 202(3), 730-746.
- Koopmans, T. C., and Beckmann, M. (1957). Assignment problems and the location of economic activities. *Econometrica: Journal of the Econometric Society*. 25(1), 53-76.

- Kouvelis, P., and Kiran, A. S. (1991). Single and multiple period layout models for automated manufacturing systems. *European Journal of Operational Research*. 52(3), 300-314.
- Krishnan, K. K., Cheraghi, S. H., Nayak, C., and Motavalli, S. (2008). Case Study in Using Dynamic From-Between Charts to Solve Dynamic Facility Layout Problems. *California Journal*. 6(1), 115-122.
- Krishnan, K. K., Cheraghi, S. H., and Nayak, C. N. (2006). Dynamic From-Between Chart: a new tool for solving dynamic facility layout problems. *International Journal of Industrial and Systems Engineering*. 1(1), 182-200.
- Krishnan, K. K., Cheraghi, S. H., and Nayak, C. N. (2008). Facility layout design for multiple production scenarios in a dynamic environment. *International Journal of Industrial and Systems Engineering*. 3(2), 105-133.
- Kulturel-Konak, S. (2007). Approaches to uncertainties in facility layout problems: Perspectives at the beginning of the 21 st Century. *Journal of Intelligent Manufacturing*. 18(2), 273-284.
- Kulturel-Konak, S., Smith, A., and Norman, B. (2004). Layout optimization considering production uncertainty and routing flexibility. *International Journal of Production Research*. 42(21), 4475-4493.
- Kusiak, A., and Heragu, S. S. (1987). The facility layout problem. *European Journal of Operational Research*. 29(3), 229-251.
- Lacksonen, T., and Ensore , E. E. (1993). Quadratic assignment algorithms for the dynamic layout problem. *International Journal of Production Research*. 31(3), 503-517.
- Lacksonen, T. A. (1994). Static and dynamic layout problems with varying areas. *Journal of the Operational Research Society*. 45(1), 59-69.
- Lacksonen, T. A. (1997). Preprocessing for static and dynamic facility layout problems. *International Journal of Production Research*. 35(4), 1095-1106.
- Lee, T. S., Moslemipour, G., Ting, T. O., and Rilling, D. (2012). A Novel Hybrid ACO/SA Approach to Solve Stochastic Dynamic Facility Layout Problem (SDFLP). In Huang, D-S., Gupta, P., Zhang, X., and Premaratne, P. (Eds.) *Emerging Intelligent Computing Technology and Applications* (pp. 100-108). Berlin: Springer-Verlag.

- Li, L., Li, B., Liang, H., and Zhu, W. (2014). The Heuristic Methods of Dynamic Facility Layout Problem. In Sun, F., Li, T., and Li, H. (Eds.) *Knowledge Engineering and Management* (pp. 269-278). Berlin: Springer-Verlag.
- Liu, F.-B. (2012). Inverse estimation of wall heat flux by using particle swarm optimization algorithm with Gaussian mutation. *International Journal of Thermal Sciences*. 54(0), 62-69.
- Maniezzo, V., and Colorni, A. (1999). The ant system applied to the quadratic assignment problem. *Knowledge and Data Engineering, IEEE Transactions on*. 11(5), 769-778.
- Mavridou, T. D., and Pardalos, P. M. (1997). Simulated annealing and genetic algorithms for the facility layout problem: A survey. *Computational optimization and applications*. 7(1), 111-126.
- Mazinani, M., Abedzadeh, M., and Mohebbali, N. (2012). Dynamic facility layout problem based on flexible bay structure and solving by genetic algorithm. *The International Journal of Advanced Manufacturing Technology*. 65(5-8), 929-943.
- McKendall, A. R., and Hakobyan, A. (2010). Heuristics for the dynamic facility layout problem with unequal-area departments. *European Journal of Operational Research*. 201(1), 171-182.
- McKendall, A. R., and Liu, W. H. (2012). New Tabu search heuristics for the dynamic facility layout problem. *International Journal of Production Research*. 50(3), 867-878.
- McKendall, A. R., and Shang, J. (2006). Hybrid ant systems for the dynamic facility layout problem. *Computers & Operations Research*. 33(3), 790-803.
- McKendall, A. R., Shang, J., and Kuppusamy, S. (2006). Simulated annealing heuristics for the dynamic facility layout problem. *Computers & Operations Research*. 33(8), 2431-2444.
- Meller, R. D., Chen, W., and Sherali, H. D. (2007). Applying the sequence-pair representation to optimal facility layout designs. *Operations Research Letters*. 35(5), 651-659.
- Meller, R. D., and Gau, K.-Y. (1996). The facility layout problem: recent and emerging trends and perspectives. *Journal of Manufacturing Systems*. 15(5), 351-366.
- Mir, M., and Imam, M. (2001). A hybrid optimization approach for layout design of unequal-area facilities. *Computers & Industrial Engineering*. 39(1-2), 49-63.

- Montreuil, B., and Venkatadri, U. (1991). Strategic interpolative design of dynamic manufacturing systems layouts. *Management Science*. 37(6), 682-694.
- Moore, K. A. (2008). *Value Mapping Framework Involving Stakeholders for Supply Chain Improvement when Implementing Information Technology Projects*. Ph.D. Thesis. University of Central Florida Orlando, Florida.
- Moslemipour, G., and Lee, T. (2012). Intelligent design of a dynamic machine layout in uncertain environment of flexible manufacturing systems. *Journal of Intelligent Manufacturing*. 23(5), 1-12.
- Moslemipour, G., Lee, T. S., and Rilling, D. (2012). A review of intelligent approaches for designing dynamic and robust layouts in flexible manufacturing systems. *The International Journal of Advanced Manufacturing Technology*. 60(1-4), 11-27.
- Murata, H., Fujiyoshi, K., Nakatake, S., and Kajitani, Y. (1995). Rectangle-packing-based module placement. *Proceedings of the 1995 Computer-Aided Design, 1995. ICCAD-95. Digest of Technical Papers., 1995 IEEE/ACM International Conference on*. 5-9 November. San Jose, CA, USA: IEEE, 472-479.
- Nayak, C. N. (2007). *Solutions to Dynamic Facility Layout Problems: Development of Dynamic From Between Chart (DFBC) and Its Applications to Continuous Layout Modeling*. Ph.D. Thesis. Department of Industrial and Manufacturing Engineering, Wichita State University, United States.
- Nehi, H. M., and Gelareh, S. (2007). A survey of meta-heuristic solution methods for the quadratic assignment problem. *Applied Mathematical Sciences*. 1(46), 2293-2312.
- Norman, B. A., and Smith, A. E. (2006). A continuous approach to considering uncertainty in facility design. *Computers & Operations Research*. 33(6), 1760-1775.
- Palekar, U. S., Batta, R., Bosch, R. M., and Elhence, S. (1992). Modeling uncertainties in plant layout problems. *European Journal of Operational Research*. 63(2), 347-359.
- Reeves, C. R. (1993). *Modern Heuristic Techniques for Combinatorial Problems*. (1th ed.). New York: John Wiley & Sons, Inc.
- Rezazadeh, H., Ghazanfari, M., Saidi-Mehrabad, M., and Sadjadi, S. J. (2009). An extended discrete particle swarm optimization algorithm for the dynamic

- facility layout problem. *Journal of Zhejiang University-Science A*. 10(4), 520-529.
- Ripon, K. S. N., Glette, K., Hovin, M., and Torresen, J. (2010b). Dynamic facility layout problem with hybrid genetic algorithm. *Proceedings of the 2010 Cybernetic Intelligent Systems (CIS), 2010 IEEE 9th International Conference on*. 1-2 September. Reading: IEEE, 1-6.
- Ripon, K. S. N., Glette, K., Hovin, M., and Torresen, J. (2010a). A Genetic Algorithm to Find Pareto-optimal Solutions for the Dynamic Facility Layout Problem with Multiple Objectives. In Wong, K. W., Mendis, B. S. U., and Bouzerdoum, A. (Eds.) *Neural Information Processing: Theory and Algorithms* (pp. 642-651). Berlin: Springer.
- Rosenblatt, M. J. (1986). The dynamics of plant layout. *Management Science*. 32(1), 76-86.
- Rosenblatt, M. J., and Kropp, D. H. (1992). The single period stochastic plant layout problem. *IIE Transactions*. 24(2), 169-176.
- Royston, J. (1982). Algorithm AS 181: The W test for normality. *Applied Statistics*. 31(2), 176-180.
- Şahin, R., Ertoğral, K., and Türkbey, O. (2010). A simulated annealing heuristic for the dynamic layout problem with budget constraint. *Computers & Industrial Engineering*. 59(2), 308-313.
- Sahin, R., and Turkbey, O. (2009). A new hybrid tabu-simulated annealing heuristic for the dynamic facility layout problem. *International Journal of Production Research*. 47(24), 6855-6873.
- Schwefel, H.-P. P. (1993). *Evolution and Optimum Seeking: The Sixth Generation*. (1th ed.). New York: John Wiley & Sons, Inc.
- See, P. C., and Wong, K. Y. (2008). Application of ant colony optimisation algorithms in solving facility layout problems formulated as quadratic assignment problems: a review. *International Journal of Industrial and Systems Engineering*. 3(6), 644-672.
- Singh, S., and Sharma, R. (2006). A review of different approaches to the facility layout problems. *The International Journal of Advanced Manufacturing Technology*. 30(5-6), 425-433.
- Skorin-Kapov, J. (1990). Tabu search applied to the quadratic assignment problem. *ORSA Journal on computing*. 2(1), 33-45.

- Tavakkoli-Moghaddam, R., Javadian, N., Javadi, B., and Safaei, N. (2007). Design of a facility layout problem in cellular manufacturing systems with stochastic demands. *Applied Mathematics and Computation*. 184(2), 721-728.
- Tian, Z., Li, H., and Zhao, Y. (2010). Optimization of continuous dynamic facility layout problem with budget constraints. *Proceedings of the 2010 Information Science and Engineering (ICISE), 2010 2nd International Conference on*. 4-6 December. Hangzhou, China: IEEE, 387-390.
- Tompkins, J. A., White, J. A., Bozer, Y. A., & Tanchoco, J. M. A. (2010). *Facilities Planning*. (4th ed.). New York: Wiley.
- Tong, X. (1991). *SECOT: A Sequential Construction Technique for Facility Design*. Ph.D. Thesis. Department of Industrial Engineering, University of Pittsburgh.
- Ulutas, B. H., and Islier, A. A. (2009). A clonal selection algorithm for dynamic facility layout problems. *Journal of Manufacturing Systems*. 28(4), 123-131.
- Ulutas, B. H., and Kulturel-Konak, S. (2012). An artificial immune system based algorithm to solve unequal area facility layout problem. *Expert Systems with Applications*. 39(5), 5384-5395.
- Urban, T. L. (1987). A multiple criteria model for the facilities layout problem. *International Journal of Production Research*. 25(12), 1805-1812.
- Urban, T. L. (1993). A Heuristic for the Dynamic Facility Layout Problem. *IIE Transactions*. 25(4), 57-63.
- Urban, T. L. (1998). Solution procedures for the dynamic facility layout problem. *Annals of Operations Research*. 76(0), 323-342.
- Xiao, Y., Seo, Y., and Seo, M. (2013). A two-step heuristic algorithm for layout design of unequal-sized facilities with input/output points. *International Journal of Production Research*. 51(14), 4200-4222.
- Yang, T., and Peters, B. A. (1998). Flexible machine layout design for dynamic and uncertain production environments. *European Journal of Operational Research*. 108(1), 49-64.
- Yildiz, A. R., and Solanki, K. N. (2012). Multi-objective optimization of vehicle crashworthiness using a new particle swarm based approach. *The International Journal of Advanced Manufacturing Technology*. 59(1-4), 367-376.