

SMART

Journal of Business Management Studies

(A Professional, Refereed, International and Indexed Journal)

A SERIAL OF SCIENTIFIC MANAGEMENT AND ADVANCED RESEARCH TRUST

Vol-11 Number- 1

January - June 2015

Rs.400

ISSN 0973-1598 (Print)

ISSN 2321-2012 (Online)

Professor MURUGESAN SELVAM, M.Com, MBA, Ph.D
Founder - Publisher and Chief Editor



**SCIENTIFIC MANAGEMENT AND ADVANCED RESEARCH TRUST
(SMART)**

TIRUCHIRAPPALLI (INDIA)

www.smartjournalbms.org

A REVIEW ON HEURISTIC APPROACHES TO SOLVE COORDINATED PRODUCTION - DISTRIBUTION PROBLEMS

Ajitha Angusamy

Faculty of Business, Multimedia University, Melaka, Malaysia

Email ID: ajitha.angusamy@mmu.edu.my

Mohamed K. Omar

Nottingham University Business School, Melaka, Malaysia

Email ID: mkhaledomar@gmail.com

and

*Anantharaman R.N.**

Faculty of Business, Multimedia University, Melaka, Malaysia

Email ID: r.n.anantharaman@mmu.edu.my

Abstract

In the current competitive environment, companies need to be highly efficient in meeting customer demand at lower operational costs. As a result, companies are compelled to seek various ways to enhance their operations. Researchers realised that optimization of various departments separately would not guarantee overall optimization. This led the Researchers to focus on the effect of integrated approach and proven significant cost reduction on the overall operational costs. This paper provides a review on coordinated models in the literature and the role of heuristic approaches to solve such coordinated models. The review found that genetic algorithm plays a vital role in solving complex coordinated problems. The article also identifies the areas that are rarely addressed in coordinated models and genetic algorithms.

Keywords: *Coordinated Models, Optimization, Heuristics, Algorithms.*

JEL Code : *M31, M39.*

1. Introduction

In today's competitive environment, companies are challenged with characteristics such as the optimum time taken to design, manufacture and distribute the products. In order to face the challenges, companies need to be highly efficient in meeting the demand at lower operational costs. The concept of coordinated production and transportation model have proved a significant cost reduction for the

industries that have used it. Most of the coordinated production and distribution problems in the literature, consider an overall production plan, inventory decisions and product flow through a facility over a single period of time in order to minimise the cost or maximise the profit.

In the past few decades, the coordinated production and transportation models that optimized the overall operational performance,

*** Corresponding Author**

have been an attractive area of research. Several mathematical models have been proposed for production-distribution problems in the literature and the complexity in solving such problems has also been investigated. Due to wide range of assumptions and conditions, the literature on the existing models and their solution approaches is very extensive. This paper focuses on studies related to production-distribution models that use heuristics method to obtain the solution.

2. Coordinated Production-Distribution Problem

The coordinated approach for production-distribution problems, consists of decisions that integrate various functions such as processing raw materials, inventory management, production planning, distribution of finished products to the warehouses and facilities location, into a single optimization model (**Fahimnia et al., 2013**). One of the earliest coordinated models was proposed by **Geoffrion and Graves (1974)** and several coordinated models for various production-distribution environments were developed later. A brief summary of the coordinated models, studied in the literature, is presented in **Table 1**.

Models were developed for a wide range of industries, under different production environments. Models provided production planning and transportation decisions for solving the problem under review. The studies have proved that coordinated models are more efficient in several aspects than the uncoordinated models. However, coordinated functions may not be beneficial in all situations, as the integrated approach depends on the system parameters (**Chandra and Fisher 1994**). The situations under which integrated efforts are more useful were provided by **Chandra and Fisher (1994)**.

The proposed models were developed

by using mathematical programming of problems such as Linear Programming (LP), Mixed Integer Linear Programming (MILP), Mixed Integer Non-linear Programming, Network Flow Problems and Dynamic Programming. Most of the coordinated models were developed by using Mixed Integer Programming (MIP) and the objective function minimizes the total operational costs .

3. Why Heuristics?

In general, linear programming problems are easier to solve when compared to MIP. Various algorithms, with the latest developments in computers, have made even a large scale LP problems solvable in a reasonable time for implementation (**Hung and Hu 1998**). Unlike LP problems, MIP problems are very difficult to solve. Due to the presence of integer variables, even a small MIP problem needs much computational effort to solve. Single level MIP problems are proved to be Non Deterministic Polynomial – time hard (NP hard) (**Bitran and Yanasse, 1982**). In the case of multi-level problems, even finding a feasible solution has proved to be NP- complete (**Maes and Wassenhove, 1991**). Hence most of the earlier studies proposed heuristics approach to solve MIP problems.

Heuristic approaches are proposed so as to find a near-optimal solution in a significantly less computation time. A mathematical programming procedure that is used to generate a solution for a complicated problem, is termed as Heuristic. Heuristics belongs to a class of optimum seeking mathematical programming methodology.

4. Heuristic Approaches

The general heuristics, called meta-heuristics, are more suitable for obtaining various combinatorial optimisation solutions (**Arostegui, 2006**). The most popular meta-heuristics are Simulated Annealing, Tabu Search and Genetic

Algorithm. The basic concepts and introduction to meta-heuristics can be found in **Glover (1989, 1990)**, **Michalewicz (1994)**. General guidelines for meta-heuristic design are discussed in **Hertz and Widmer (2003)**.

4.1 Simulated Annealing

Simulated Annealing (SA), which was formally introduced by **Kirkpatrick et al. (1983)**, are often used to solve combinatorial optimisation problems. The characteristics of Simulated Annealing are discussed in **Eglese (1990)** and **Jans and Degraeve (2007)**. Studies that illustrate the efficiency of SA are **Kuik et al., (1993)**, **Marett and Wright (1996)** **Tian et al. (1996)**, **Drezner et al. (2002)**, **Ou Tang (2004)**. **Özdamar and Barbarosoglu (2000)** who suggested an integrated Lagrangean relaxation and simulated annealing method to solve a dynamic multi-level multi-item lot sizing problem. Authors found that the integrated approach is beneficial in providing satisfactory solution when compared with the solution obtained by Lagrangean relaxation in **Tempelmeir and Derstroff (1996)**.

4.2 Tabu Search

Tabu Search (TS) is another searching technique that can be used as an alternative to SA. The earlier studies that used TS to solve capacitated lot problems, are **Hindi (1995, 1996)**. The studies that compared TS with other heuristics and concluded that TS outperformed other heuristics are noxious facilities location problem by **Kincaid (1992)**, hydraulic turbine runner-balancing problem by **Sinclair (1993)**, fixed charged transportation problem by **Sun (1998)**, vehicle routing by **Gendreau et al.(1994)**, routing and distribution by **Rochat and Semet (1994)**, un-capacitated warehouse location by **Michel and Hentenryck (2004)**, p-median problems by **Rolland et al. (1996)** and **Voss (1996)**, un-capacitated facility location

problem by **Ghosh (2003)**, budget constrained location problem by **Wang et al.(2003)**, single source capacitated locations problem by **Corinthal and Captivo (2003)** and **Chamberland (2004)**, supply chain scheduling by **Higgins et. al (2005)** and **Garcia and Lozano (2005)**, **Arosteguie (2006)**.

Studies that used TS to solve production-distribution problems are **Keskin, and Üster (2007)**, **Bard and Nananukul (2010)**, **Armentano et al., (2011)** and **Condotta et al. (2013)**. The heuristics proposed by **Armentano et al., (2011)**, outperformed the algorithm developed by **Boudia and Pins (2009)**, **Bard and Nananukul (2010)** in all instances.

4.3 Genetic Algorithms

In the recent years, Genetic Algorithms (GA) have attracted researchers. The basic concepts of GA were introduced by **Grinold and Marshall (1977)**. The general form of GA is described by **Goldberg (1989)**. The capability of GA, to obtain good quality solutions, are illustrated in **Goldberg (1989)**, **Shaw and Fleming (1996)**, **Lee et.al., (1997)**, **Herrman (1999)** and **De-Jong (2006)**. GA is a direct search algorithm which works by imitating principles of evolution and chromosomal processing in natural genetics (**Gupta and Bhunia, 2006**). GA is a global search optimization technique which starts with an initial set of random solutions known as population. Every individual or solution in the population is called chromosome. The chromosomes in the population can be either represented by binary coded strings or real coded strings. It has been proven that GAs show much better performance in the case of solving large-size problems (**He & Hui, 2007a,2007b**).

Several studies have proved genetic algorithms to be effective and useful approach

for solving optimization problems. However, for many problems, the traditional simple genetic algorithm does not provide good quality solutions. Researchers proposed several methods of hybridization to overcome this problem. In the case of hybrid genetic algorithm, a local optimization is incorporated as an add-on extra to the simple genetic algorithm loop of recombination and selection. Due to the complementary properties of conventional heuristics and genetic algorithms, the hybrid approach often outperforms either method operation alone (**Gen & Cheng, 1997**).

GAs can be used to solve various problems such as lot-sizing and scheduling, scheduling and sequencing, facility-location problem, vehicle routing and scheduling, supply chain scheduling, production and distribution. The literature that addresses some of the above mentioned problems are summarized in **Table-2**. The studies illustrate that GA is capable of handling any kind of problem efficiently.

5. Discussions

The coordinated models reviewed, demonstrates a significant advance in the integrated analysis. The studies proved that coordination of production and distribution can bring significant cost reduction in the total cost. However, most of the models consist of only one objective function which aims at minimizing the total operational cost. Therefore, models with multi-objective functions and also maximization problems, need to be addressed in future studies. The literature also reveals that coordinated models are difficult to solve. Heuristic approach plays an important role in obtaining optimal or near optimal solutions for complex problems.

Meta-heuristics can be used to solve complex, combinatorial optimization problems. Very few studies used simulated annealing and tabu search to solve production-distribution

problems. Among the meta-heuristics, GA is dominant in solving complex problems. Furthermore, GA, combined with other heuristics such as Lagrangean approach, greedy decoder and also with modified operators, neighbourhood search, has proved to be more beneficial than the traditional GA.

The application of real coded GA is scarce in literature. Since real coded representation makes hybridization easier (**Kaelo and Ali, 2007**), there is still a need to develop more real coded hybrid GA in order to enhance the existing solution approaches.

6. Conclusion

The review of literature in the area of coordinated production-distribution problems that use heuristic approach to solve has been presented in this paper. The review of earlier research shows that coordinated models are more beneficial in terms of cost savings and operational efficiency. Most of the existing problems are modelled mathematically as MIP problems. The studies also reveal that complex problems can be solved for optimality or near optimality by using hybrid meta-heuristics. Among the meta-heuristics, hybrid GA is more dominant in solving complex problems. Future researches need to focus on problems with multi-objective functions and develop real coded hybrid GAs.

7. References

- Arostegui, Jr., M.A., Kadipasaoglu, S.N., and Khumawala, B.M., (2006). An empirical comparison of Tabu Search, Simulated Annealing and Genetic Algorithms for facilities location problems. *The International Journal of Production Economics*, 103, 742-754.
- Armentano V.A., Paulo M.França, Franklina M.B. de Toledo, (1999) . A network flow model for the capacitated lot-sizing problem. *The International Journal of Management Science*, 27, 275- 284.

- Amorim, P, Günther, H.O., and Almada-Lobo, B., (2012). Multi-objective integrated production and distribution planning of perishable products. *International Journal of Production Economics*, 138, 89 – 101.
- Asgari, N., Farahani, R. Z., Rashidi-Bajgan, H.m, and Sajadieh, M. S., (2013). Developing model based software to optimize wheat storage and transportation: A real-world application. *Applied Soft Computing*, 13, 1074 – 1084.
- Aydinel, M., Sowlati, T., Cerda, X., Cope, E., and Gerschman, M., (2008). Optimization of production allocation and transportation of customer orders for a leading forest products company. *Mathematical and Computer Modelling*, 48, 1158 – 1169.
- Bard, J.F., and Nanakul, N., (2010). A branch and price algorithm for an integrated production and inventory routing problem. *Computers and Operations Research*, 37, 2202-2217.
- Benjamin, J. (1989), “An analysis of inventory and transportation costs in a constrained network”. *Transportation Science*, 23(3), 177-183.
- Bilgen, B., and Ozkarahan, I., (2007). A mixed-integer linear programming model for bulk grain blending and shipping, *International Journal of Production Economics*, 107, 555 – 571.
- Bitran, G. R., and Yanasse, H.H., (1982). Computational complexity of the capacitated lot size problem, *Management Science*, 28, 1174-1186.
- Blumenfeld, D. E., Burns, L.D., Diltz, J.D., and Daganzo, C.F. (1985), “Analyzing trade-offs between transportation, inventory and production costs on freight networks”. *Transportation Research*, 19(5), 361-380.
- Boissière, Frein, Y., and Rapine, C., (2008). “Lot-sizing in a serial distribution system with capacitated in-system production flow.” *International Journal of Production Economics*, 112, 555 – 571.
- Borisovsky, P., Dolgui, A., Ereemeev, A. (2009). Genetic algorithms for a supply management problem: MIP-recombination vs greedy decoder. *European Journal of Operational Research*, 195(3), 770-779.
- Boudia, M., Louly, M.A.O., and Prins, C. (2007). A reactive GRASP and path relinking for a combined production-distribution problem. *Computers and Operations Research*, 34, 3402 – 3419.
- Chamberland, S., (2004). An efficient heuristic for the expansion problem in cellular wireless networks. *Computers & Operations Research*, 31, 1769 – 1791.
- Chandra, P. and Fisher, M.L., (1994). Coordination of Production and Distribution Planning, *European Journal of Operational Research*, 72, 503-517.
- Chen, Z. L., (2004). Integrated Production and Distribution Operations: Taxonomy, Models and Review, In: Simchi-Levi D, W.D, Chen Z-L, editors. Handbook of Quantitative Supply Chain Analysis: Modelling in the E-Business Era. *Kluwer Academic Publishers, New York*.
- Chen, Z. L., and Vairaktaris, G.L., (2005). Integrated scheduling of production and distribution operations. *Management Science*, 51, 614 – 628.
- Chiadamrong, N., and Kawtummachai, R., (2008). A methodology to support decision-making on sugar distribution for export channel: A case study of Thai sugar industry. *Computers and Electronics in Agriculture*, 64, 248-261.
- Chiang, W-C, Russell, R, Xu, X and Zepeda, D., (2009). A simulation/ metaheuristic approach to newspaper production and distribution supply chain problems. *International Journal of Production Economics*, 121, 752-767.

- Chung, S.H., Lau, H.C.W., Choy, K.L., Ho, G.T.S., and Tse, Y.K., (2010). Application of genetic approach for advanced planning in multi-factory environment. *International Journal of Production Economics*, 127, 300-308.
- C ccola, M.E., Zamarripa, M., M ndez, C. A., and Espu a, A.,(2013). Toward integrated production and distribution management in multi-echelon supply chains. *Computers and Chemical Engineering*, article in press.<http://dx.doi.org/10.1016/i.compchemeng.2013.01.004>.
- Condotta, A., Knust, S., Meier, D.m and Shakhlevich, N. V.,(2013). Tabu search and lower bounds for a combined production-transportation problem. *Computers & Operations Research*, 40, 886-900.
- Corinthal, M.J., and Captivo, M.E., (2003). Upper and lower bounds for the single source capacitated location problem. *European Journal of Operational Research*, 151, 333-351.
- Das, K., and Sengupta, S., (2008). A hierarchical process industry production-distribution planning model. *International Journal of Production Economics*, 117, 402-419.
- De Jong, K.A., (2006). Evolutionary Computation A Unified Approach, *The MIT Press*, Cambridge.
- Delavar, R. D., Keshteli, M. H., and Zaverdehi, M-H.,(2010). Genetic algorithm for scheduling of production and air transportation. *Expert Systems with Application*, 37, 8255-8266.
- Dellaert, N., Jeunet, J.andJonard, N.(2000) A genetic algorithm to solve the general multi-level lot-sizing problem with time-varying costs. *International Journal of Production Economics*, 68, 241-257.
- Dhaenens-Flipo, C. and Finke, G.(2001) An integrated model for an industrial production-distribution problem. *IIE Transactions*, 33, 705-715.
- Drezner, T., Drezner, Z., and Salhi, S., (2002). Solving the multiple competitive facilities location problem. *European Journal of Operational Research*, 142(1), 138-151.
- Eglese, R.W.,(1990). Simulated annealing: A tool for operational research. *European Journal of Operational Research*, 46, 271-281.
- Ek iođlu, S.D., Romeijn, H., and Pardalos, M. (2006). Cross-facility management of production and transportation planning problem. *Computers and Operations Research*, 33, 3231-3251.
- Ereng c, S.S., Simpson N. C., and Vakhari A.J., (1999). Integrated production/distribution planning in supply chains: an invited review. *European Journal of Operational Research*, 115, 219-246.
- Fahimia, B., Farahani, R.Z., Marian, R., and Luong, L., (2013). A review and critique on integrated production-distribution planning models and techniques. *Journal of Manufacturing Systems*,32, 1-19.
- Fahimnia, B., Luong, L., and Marian, R., (2011). Genetic algorithm optimization of an integrated aggregate production-distribution plan. *International Journal of Production Research*. 1-16.
- Fakhrzad, M.B., and Zare, H.K., (2009). Combination of genetic algorithm with Lagrange multipliers for lot-size determination in multi-stage production scheduling problems. *Expert Systems with Applications*, 36, 10180-10187.
- Garcia, JM., Lozano, S. and Canca, D.(2004) Coordinated scheduling of production and delivery from multiple plants. *Robotics and Computer-Integrated Manufacturing*, 20, 191-198.
- Geoffrion, A.M., and Graves, G.W. (1974). Multicommodity distribution design by Benders decomposition, *Management Science*, 20/5, 822-844.

- Gen, M., and Cheng, R., (1997). Genetic Algorithms and Engineering Design (New York: John Wiley & Sons).
- Gendreau, M., Hertz, A., and Laporte, G.,(1994). A tabu search heuristic for the vehicle routing problem. *Management Science*, 40 (10), 1276 - 1290.
- Glover, F., (1989). Tabu Search – Part I, *ORSA. Journal on Computing*, 1(3), 190 – 206.
- Glover, F., (1990). Tabu Search – Part II, *ORSA. Journal on Computing*, 2(1), 4 - 32.
- Goldberg, D.E., (1989). Genetic Algorithms in Search, Optimization and Machine Learning. *Addison-Wesley Publisher*, Massachusetts.
- Glover, F., and Laguna, M.,(1989). Target Analysis to improve a tabu search method for machine scheduling. *Advanced Knowledge Research Group, US West Advanced Technologies*, Boulder, CO.
- Gupta, R. K., and Bhunia, A. K., (2006). An application of real-coded Genetic Algorithm for integer linear programming in Production-Transportation Problems with flexible transportation cost. *AMO-Advanced Modeling and Optimization*, 8, (1).
- He, Y. and Hui, C. (2007a). Genetic algorithm based on heuristic rules for high-constrained large-size single-stage multi-product scheduling with parallel units, *Chemical Engineering & Processing: Process Intensification*, 46(11), 1175–1191.
- He, Y., and Hui, C. (2007b) Genetic algorithm for large-size multi-stage batch plant scheduling. *Chemical Engineering Science* 62(5), 1504–1523.
- Herrmann, J.W., (1999). A Genetic Algorithm for Minimax Optimisation Problems. *Proceedings of the 1999 Congress on Evolutionary Computation*, Washington DC, USA, 1099-1103.
- Hertz, A., Widmer, M.,(2003). Guidelines for the use of meta-heuristics in combinatorial optimization, *European Journal of Operational Research*, 151, 247-252.
- Higgins, A., Beashel, G., and Harrison, A. (2005). Scheduling of brand production shipping within a sugar supply chain. *Journal of Operational Research Society*, 1 – 9.
- Hindi, K.S.,(1996). Solving the CLSP by a Tabu Search heuristic, *Journal of Operational Research Society*, 47, 151-161.
- Hung, Y. F., and Hu, Y.C., 1998. Solving mixed integer programming production planning problems with setups by shadow price information. *Computers and Operations Research*, 25(12), 1027-1042.
- Haq, A. N., Vrat, P., and Kanda, A. (1991). An integrated production-inventory-distribution model for manufacture of urea: A case. *International Journal of Production Economics*, 39, 39-49.
- Jans, R., and Degraeve, Z., (2007) Meta-heuristics for dynamic lot sizing: A review and comparison of solution approaches. *European Journal of Operational Research*, 177, 1855-1875.
- Jayaraman, V. and Prikul, H.(2001). Planning and coordination of production and distribution facilities for multiple commodities. *European Journal of Operational Research*, 133, 394-408.
- Keskin, B.B., and Üster, H., (2007). Metaheuristic approaches with memory and evolution for a multi-product production/distribution system design problem. *European Journal of Operational Research*, 182, 663-682.
- Kimms, A., (1999). A genetic algorithm for multi-level, multimachine lot sizing and scheduling, *Computers and Operations Research*, 26, 829-848.

- Kincaid, R.K., (1992). Good solutions to discrete noxious location problems via metaheuristics. *Annals of Operations Research*, 40, 265-281.
- Kirkpatrick, A., Gelatt Jr., C.D., and Vecchi, M.P., (1983). Optimization by simulated annealing. *Science*, 220, 671-680.
- Kuik, R. Solomon, M., and Van Wassenhove, L.N. and Maes, J. (1993). Linear Programming, Simulated annealing, and tabu search heuristics for lot sizing in bottleneck assembly systems, *IIE Transactions*, 25/1, 62-72.
- Kumar, A., PrakashTiwari, M.K., Shankar, R., and Baveja, A. (2006). Solving machine-loading problem of a flexible manufacturing system with constrained-based genetic algorithm. *European Journal of Operational Research*, 175(2), 1043–1069.
- Lee, C.Y., Lei, Y. and Pinedo, M., (1997). Current trends in deterministic scheduling. *Annals of Operations Research*, 70, 1 – 41.
- Liu, C. H., (2011). Using genetic algorithms for the coordinated scheduling problem of a batching machine and two-stage transportation. *Applied Mathematics and Computation*, 217, 10095 – 10104.
- Li, Y, Chen, J., and Cai, X.,(2007). Heuristic genetic algorithm for capacitated production planning problems with batch processing and remanufacturing. *International Journal of Production Economics*, 111(1), 180-191.
- Maes, J. and Van Wassenhove, L.N., (1991). Multi-level capacitated lotsizing complexity and LP based heuristics. *European Journal of Operational Research*, 53, 131-148.
- Marett, R., and Wright, M., (1996). A comparison of neighbourhood search techniques for multi-objective combinatorial problems. *Computers & Operations Research*, 23 (5), 465 - 483.
- Melo, R. A., and Wolsey, L.A., (2010). Optimizing production and transportation in a commit-to-delivery business mode. *European Journal of Operational Research*, 203, 614-618.
- Michalewicz, Z., (1994). Genetic Algorithms + Data Structures - Evolution Programs. *Springer*, Berlin.
- Michel, L. and Hentenryck, P.V., (2004). A simple tabu search for warehouse location. *European Journal of Operational Research*, 157, 576 - 591.
- Özdamar, L. and Barbaraosoglu, G., (2000). An integrated Lagrangean relaxation-simulated annealing approach to the multi-level multi-item capacitated lot sizing problem. *International Journal of Production Economics*, 68, 319-331.
- Pasandideh, S. H. R., Niaki, S. T.A., and Nia, A. R., (2011). A genetic algorithm for vendor managed inventory control system of multi-product multi-constraint economic order quantity model. *Expert Systems with Applications*, 38, 2708 – 2716.
- Pyke, D.F. and M. Cohen (1994). Multiproduct integrated production-distribution systems. *European Journal of Operational Research*, 74, 18-49.
- Rizk, N., Martel, A., and Amours, S.D., (2006). Multi-item dynamic production-distribution planning in process industries with divergent finishing stages. *Computers & Operations Research*, 33, 3600-3623.
- Rochat, Y. and Semet, F., (1994). A tabu search approach for delivering pet food and flour. *Journal of Operational Research Society*, 45 (11), 1233 – 1246.
- Rolland, E., Schilling, D.A. and Current, J.R., (1996). An efficient tabu search procedure for the p -median problem, *European Journal of Operational Research*, 96, 329 – 342.
- Scholz-Reiter, B., Frazzon, E.M., and Makuschewitz, T., (2010). Integrating manufacturing and logistics systems along global supply chains. *CIRP Journal of Manufacturing Science and Technology*, 2, 216 -223.

- Shaw, K.J., and Fleming, P.J., (1996). An initial study of practical multi-objective production scheduling using genetic algorithms. *Proceedings of the UKACC International Conference on CONTROL'96*, University of Exeter, United Kingdom, 2-5.
- Sinclair, M., (1993). Comparison of the performance of modern heuristics for combinatorial optimization on real data. *Computers & Operations Research*, 20(7), 687-695.
- Spitter, J.M., Hurkens, C.A.J., A.G de Kok, Lenstra, J.K. and Negenman, E.G. (2005). Linear Programming Models with Planned Lead Times for Supply Chain Operations Planning. *European Journal of Operational Research*, 163, 706-720.
- Stecke, K.E., and Zhao, X., (2007). Production and transportation integration for a make-to-order manufacturing company with a commit-to-delivery business mode. *Manufacturing and Service Operations Management*, 9, 206-224.
- Sun, A., (1998). A tabu search heuristic procedure for solving the transportation problem with exclusionary side constraints, *Journal of Heuristics*, 3(4), 305 – 326.
- Swenseth, S.R. and Godfrey, M.R.,(2002). Incorporating transportation costs into inventory replenishment decisions. *International Journal of Production Economics*, 77, 113-130.
- Tang, O., (2004). Simulated annealing in lot sizing problems. *International Journal of Production Economics*, 88, 173 – 181.
- Tempelmeier, H. and Derstroff, M., (1996). A Lagrangean-based-heuristic for dynamic multi-level multi-item constrained lotsizing with setup times, *Management Science*, 42/1, 738-757.
- Tian, P., Wand, H., and Zhang, D., (1996). Simulated annealing for the quadratic assignment problem. *Computers and Industrial Engineering*, 31 (3/4), 925 - 928.
- Torabi, S.A., FatemiGhomi, S.M.T. and Karimi, B.,(2006) A hybrid genetic algorithm for the finite horizon economic lot and delivery scheduling in supply chains. *European Journal of Operational Research*,173,173-189.
- Voss, S., (1996). A reverse elimination approach for the p -median problem. *Studies in Locational Analysis*, 8, 49-58.
- Xie, J.X., and Dong, J., (2002) Heuristic Genetic Algorithms for General Capacitated Lot-Sizing Problems. *Computers and Mathematics with Applications*, 44, 263-276.
- Yan, C., Banerjee, A., and Yang, L., (2011). An integrated production-distribution model for a deteriorating inventory item. *International Journal of Production Economics*, 133, 228-232.
- Zhao, X., and Stecke, K.E., (2005). Managing the technology of integrating the production and transportation functions in assembly or flow operations for make-to-order industries, in: *Proceedings of Portland International Conference on Management of Engineering and Technology*, 489 -499.
- Zuo, M., Kuo, W. and McRoberts, K. (1991). Application of mathematical programming to a large-scale agricultural production and distribution system. *Journal of Operational Research Society*,42(8), 639-648

Table -1 Review of Coordinated Models

Proposed Coordinated models	Author and Year
Production - Distribution	Benjamin (1989), Chandra and Fisher (1994), Zuo <i>et al.</i> (1991), Dhanenens-Flipo and Finke (1999), Jayaraman and Pirkul (2001), Ekşioğlu <i>et.al</i> (2006), Boudia <i>et.al</i> (2007), Chen (2004), Zhao and Stecke (2005), Chen and Vairaktaris (2005), Rizk <i>et. al.</i> , (2006), Stecke and Zhao (2007), Boissière <i>et.al.</i> , (2008), Aydinel <i>et.al.</i> , (2008), Melo and Wolsey (2010), Yan <i>et al.</i> (2011), Amorim <i>et al.</i> (2012) Cóccola <i>et al.</i> (2013), Condotta <i>et al.</i> (2013), Fahimnia <i>et al.</i> (2011),
Supply-chain	Pyke and Cohen (1993), Erengüç <i>et. al.</i> , (1999), Jayaraman and Pirkul (2001), Spitter <i>et al</i> (2005), Ekşioğlu <i>et.al</i> (2006), Bilgen and Ozkarahan (2007), Boissière <i>et.al.</i> ,(2008), Chiang <i>et. al.</i> , (2009), Das and Sengupta (2009). Scholz-Reiter <i>et al.</i> , (2010),
Production-Inventory - Transportation	Blumenfeld <i>et al.</i> (1985). Haq <i>et al.</i> (1991), Swenseth and Godfrey (2002), Bard and Nananukul (2010), Asgari <i>et al.</i> (2013)

Table – 2 Review of Models that use Genetic Algorithm

Author and Year	Problem Addressed
Kimms (1999)	Lot sizing and scheduling
Dellaert <i>et.al.</i> ,(2000)	Multi-stage lot-sizing problem
Xie and Dong (2002)	General capacitated lot-sizing problem
Torabi <i>et. al.</i> , (2006)	An economic lot and delivery scheduling problem
Kumar <i>et. al.</i> , (2006)	Machine-loading problem
Gupta and Bhunia (2006)	Production-transportation problem
Li <i>et. al.</i> , (2007)	Capacitated production planning problem
Chiadamrong and Kawtummachai (2008)	Warehouse and Distribution management problem
Borisovsky <i>et. al.</i> , (2009)	Supply management problem
Fakhrzad and Khademi (2009)	Multi-stage production scheduling problem
Delavar <i>et.al.</i> , (2010)	Coordinated scheduling of production and air transportation
Chung <i>et.al.</i> , (2010)	Advanced planning in Multi-factory environment
Pasinidideh <i>et.al.</i> , (2011)	Non-linear programming problem for a economic order quantity for a two-level supply chain system.
Liu (2011)	Production-scheduling-transportation
Fahimnia <i>et.al.</i> , (2011)	Production-distribution
Asgari <i>et. al.</i> , (2013)	Production-inventory transportation