



Note on cost minimization for a multi-product fabrication-distribution problem with commonality, postponement and quality assurance

Yuan-Shyi Peter Chiu^a • Hong-Dar Lin^{a*} • Hua-Yao Wu^b

^aDepartment of Industrial Engineering and Management, Chaoyang University of Technology, Taichung 413, Taiwan

^bPhysics Department, College of Liberal Arts and Sciences, State University of New York at Oswego, NY 13126, USA

Received 11 24 2019; accepted 02 24 2020

Available online 02 29 2020

Abstract: A multiproduct fabrication-shipment problem with postponement and quality reassurance was explored (Chiu et al., 2016) to minimize overall relevant system costs, in which the differential calculus was utilized in the optimization procedure to help decide the optimal fabrication-distribution policy. This research note presents a two-phase algebraic approach, in lieu of their derivatives, to find the optimal policy to the problem. Such a simplified method aims at helping managers or practitioners in the field to understand and solve the problem more effectively, and without derivatives.

Keywords: Operations research, multi-product manufacturing system, algebraic approach, fabrication-distribution integration system, multi-delivery, quality assurance

*Corresponding author.

E-mail address: hdlin@cyut.edu.tw (Hong-Dar Lin).

Peer Review under the responsibility of Universidad Nacional Autónoma de México.

1. Introduction

In the manufacturing field, the most economical production lot was originally presented by Taft (1918) with simple assumption of producing single product on single machine under the condition of perfection in production process. In real life situations, however, production of nonconforming/imperfect products is inevitable owing to diverse uncontrollable factors (Porteus, 1986; Boone, Ganeshan, Guo, & Ord, 2000; Kumar, Goyal, & Singhal, 2017; Muralidharan, Vallavaraj, Mahanti, & Patidar 2017; MohanDas, Ayyanar, Susaiyappan, Kalimuthu, 2017; Villarreal et al., 2018; Arun, Lincon, & Prabhakaran, 2019). Some nonconforming items can be reworked to restore their quality, and of course with extra cost (Inderfurth, Janiak, Kovalyov, & Werner, 2006; Makarova, Shubenkova, Mavrin, & Boyko, 2017; Zhang, Wu, & Tang, 2017; Chiu, Wu, Chiu, & Hwang, 2018). Further, various discontinuous multi-shipment policies are commonly used in distribution of end products in the real supply chain systems (Thomas, & Hackman, 2003; Nielsen, & Saha, 2018; Lin, Chen, Chiu, & Chiu, 2019).

To maximize utilization, producing multiproduct on a single machine using common cycle time discipline can be of assistance (He, Wu, & Zhang, 2018; Mehdizadeh, Gholami, & Naderi, 2018). Moreover, in the planning stage of fabrication, when multiproduct sharing a common component, the delayed differentiation strategy is constantly taken into consideration to cut short the cycle time and/or reduce total system cost (Swaminathan, & Tayur, 1998; Abbad, & Zahratahdi, 2008; Bernstein, DeCroix, & Wang, 2011; Chiu, Kuo, Chiu, & Chang, 2019). Motivated by the benefit of applying postponement strategy to the real vendor-buyer systems, Chiu, Kuo, Chiu and Chiu (2016) studied a multiproduct fabrication-shipment problem featuring postponement and product quality reassurances using mathematical modeling and differential calculus to decide the optimal operating decisions. They also demonstrated that various benefits (in terms of reduction of cycle length and system) can be gained from their research results.

Unlike the differential calculus methodology, a simplified algebraic method was presented by Grubbstrom and Erdem (1999) to resolve an economic order quantity model without reference to derivatives in their optimization procedure. Likewise, we present such a two-phase approach to the model explored by Chiu et al. (2016) and show that the optimal operating decisions to the problem can be found without using the derivatives.

2. Materials and methods

2.1. The Proposed System

A two-stage multiproduct fabrication-shipment model is reexamined in this paper by the use of the mathematical modeling and a simplified algebraic method to find the optimal fabrication-distribution policy. Consider L distinct end products sharing a common intermediate component must be produced on a machine, and their separate annual demand λ_i must be satisfied (where $i=1, 2, \dots, L$). Stage one (see Figure 1) manufactures common parts at a of $P_{1,0}$, and stage two fabricates in sequence L customized finished items at a rate $P_{1,i}$.

In each stage, the manufacturing processes may generate x_i portion of defects randomly, at a rate $d_{1,i}$; so, (where $d_{1,i}=x_iP_{1,i}$ & $i=0, 1, 2, \dots, L$; please note that $i=0$ means the process is in stage 1 fabricating common parts). All stocks are inspected and a $\theta_{1,i}$ proportion of the defective items is identified as scrap (please note that $0 \leq \theta_{1,i} \leq 1$) and the other portion $(1 - \theta_{1,i})$ can be repaired via rework process in each cycle when of regular process ends, at a rate $P_{2,i}$ (Figure 1).

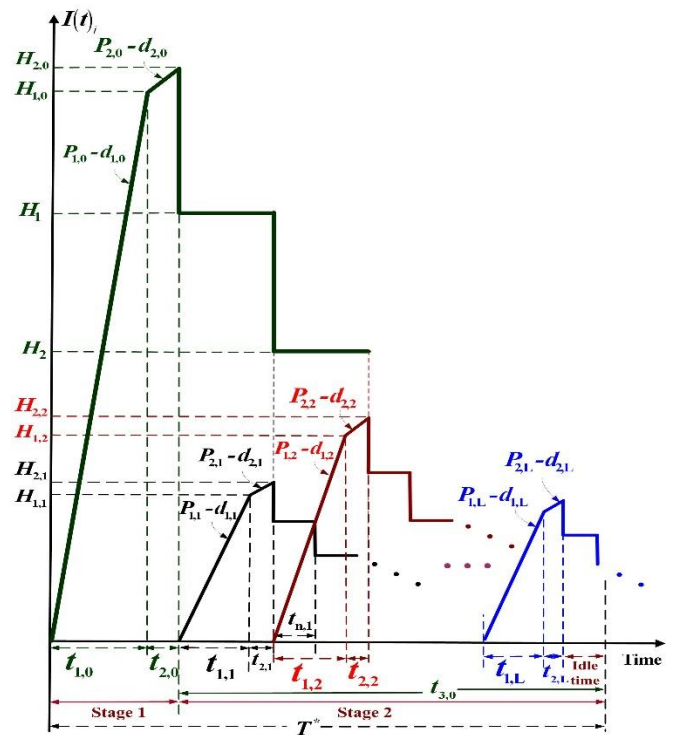


Figure 1. Inventory status of perfect quality common parts and end items in this study (Chiu et al., 2016).

Moreover, the rework is imperfect, a $\theta_{2,j}$ portion ($0 \leq \theta_{2,j} \leq 1$) of reworked items are scrap (see Figure 2). Under no stock-out permitted assumption in this model, we must have $(P_{1,j} - d_{1,j} - \lambda_i) > 0$. The same notation used as in (Chiu et al., 2016) is listed in Appendix A.

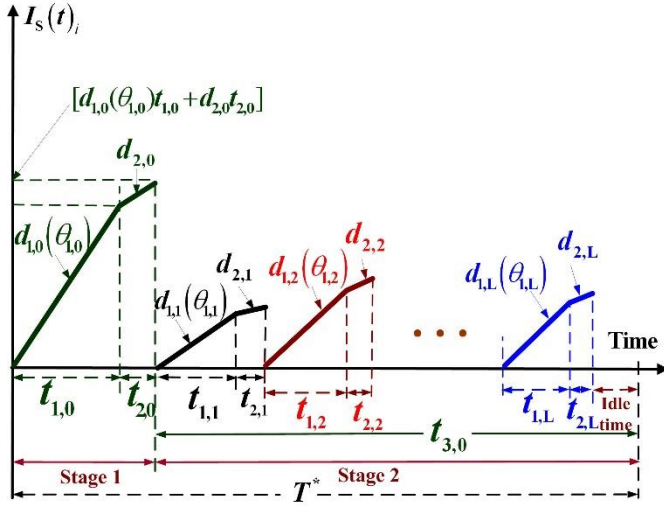


Figure 2. Status of scrapped common parts and end items (Chiu et al., 2016).

Status of stocks at customer's side is shown in Figure 3 and customer's stock holding cost is (Chiu et al., 2016)

$$\sum_{i=1}^L \left\{ h_{3,i} \left[\frac{n(D_i - I_i)t_{n,i}}{2} + \frac{n(n+1)}{2} I_{i^{n,i}} + \frac{nI_i(t_{1,i} + t_{2,i})}{2} \right] \right\} \quad (1)$$

Applying the same mathematical modeling as in [1], total relevant system cost $TC(T, n)$ includes the stage one, common part's setup, variable fabrication, rework, disposal, and holding costs; and the stage two, L distinct end items' sum of variable fabrication, setup, rework, disposal, and holding costs; and the fixed and variable distribution cost, and the buyers' stock holding costs, as follows [1]:

$$TC(T, n) = \left\{ \begin{aligned} &K_0 + C_0Q_0 + C_{R,0} [x_0(1 - \theta_{1,0})Q_0] \\ &+ C_{S,0} (x_0\phi_0)Q_0 \\ &+ h_{2,0} \left(\frac{(1 - \theta_{1,0})d_{1,0}t_{1,0}}{2} \right) (t_{2,0}) + h_{4,0}T(x_0Q_0) \\ &+ h_{1,0} \left[\frac{H_{1,0}t_{1,0}}{2} + \frac{H_{1,0} + H_{2,0}}{2} (t_{2,0}) \right. \\ &\quad \left. + \frac{d_{1,0}t_{1,0}}{2} (t_{1,0}) + \sum_{i=1}^L H_i (t_{1,i} + t_{2,i}) \right] \end{aligned} \right\}$$

$$\left. \begin{aligned} &K_i + C_iQ_i + C_{R,i} [x_i(1 - \theta_{1,i})Q_i] + C_{S,i} [x_i\phi_iQ_i] + nK_{1,i} \\ &+ h_{1,i} \left[\frac{Q_i}{2} (t_{1,i}) + \frac{H_{1,i}t_{1,i}}{2} + \frac{H_{2,i} + H_{1,i}}{2} (t_{2,i}) \right. \\ &\quad \left. + \left(\frac{n-1}{2n} \right) H_{2,i}t_{3,i} + \frac{d_{1,i}t_{1,i}}{2} (t_{1,i}) \right] \\ &+ h_{3,i} \left[\frac{n(D_i - I_i)t_{n,i}}{2} + \frac{n(n+1)}{2} I_{i^{n,i}} + \frac{nI_i(t_{1,i} + t_{2,i})}{2} \right] \\ &+ C_{T,j} [Q_i(1 - \phi_i x_i)] + h_{2,j} \left(\frac{P_{2,j}t_{2,j}}{2} \right) (t_{2,i}) + h_{4,i} (x_iQ_i)T \end{aligned} \right\} \quad (2)$$

Apply $E[x]$ to cope with the randomness of x and with extra derivations, one finds $E[TCU(T, n)]$ as follows (Chiu et al., 2016):

$$\begin{aligned} E[TCU(T, n)] &= E[TC(T, n)] / E[T] \\ &= \left[C_0E_{0,0}\lambda_0 + \frac{K_0}{T} + C_{R,0}\lambda_0(1 - \theta_{1,0})E_{1,0} + C_{S,0}\lambda_0\phi_0E_{1,0} + \omega_0T \right] \\ &\quad + \sum_{i=1}^L \left\{ \begin{aligned} &\left[C_i\lambda_iE_{0,i} + \frac{K_i}{T} + C_{R,i}\lambda_i(1 - \theta_{1,i})E_{1,i} + C_{S,i}\lambda_i\phi_iE_{1,i} + \frac{nK_{1,i}}{T} + C_{T,j}\lambda_i \right] \\ &+ \frac{h_{1,i}T\lambda_i^2}{2} \left[E_{3,i} + E_{4,i} - \frac{E_{5,i}}{n} \right] + \frac{h_{2,i}\lambda_i^2E_{1,i}^2}{2} \left[\frac{(1 - \theta_{1,i})^2}{P_{2,i}} \right] T \\ &+ \frac{h_{3,i}T\lambda_i^2}{2} \left[\frac{2E_{0,i}}{P_{1,i}} + \frac{2(1 - \theta_{1,i})E_{1,i}}{P_{2,i}} - \frac{1}{\lambda_i} + \left(1 + \frac{1}{n} \right) E_{5,i} \right] + Th_{4,i}\lambda_iE_{1,i} \end{aligned} \right\} \quad (3) \end{aligned}$$

where

$$\omega_0 = \left\{ \begin{aligned} &\frac{h_{1,0}\lambda_0^2(E_{0,0})^2}{2} \left[\frac{1}{P_{1,0}} + \frac{2E[x_0](1 - \theta_{1,0})(1 - E[x_0])}{P_{2,0}} \right. \\ &\quad \left. + \frac{E[x_0]^2(1 - \theta_{1,0})^2(1 - \theta_{2,0})}{P_{2,0}} \right] \\ &+ E_{1,0}h_{4,0}\lambda_0 + \frac{h_{2,0}\lambda_0^2(E_{1,0})^2}{2} \left[\frac{(1 - \theta_{1,0})^2}{P_{2,0}} \right] \\ &+ h_{1,0} \sum_{i=1}^L \left\{ \begin{aligned} &\left[\frac{\lambda_iE_{0,i}}{P_{1,i}} + \frac{\lambda_i(1 - \theta_{1,i})E_{1,i}}{P_{2,i}} \right] \\ &\left[\sum_{i=1}^L (\lambda_iE_{0,i}) - \sum_{j=1}^i (\lambda_jE_{0,j}) \right] \end{aligned} \right\} \end{aligned} \right\};$$

$$E_{0,0} = \frac{1}{(1 - E[x_0]\phi_0)}; E_{1,0} = \frac{E[x_0]}{(1 - E[x_0]\phi_0)}$$

$$E_{0,j} = \frac{1}{(1 - \phi_j E[x_j])} \quad \text{for } j = 1, \dots, i;$$

$$E_{0,i} = \frac{1}{(1 - E[x_i]\phi_i)}; E_{1,i} = \frac{E[x_i]}{(1 - E[x_i]\phi_i)}$$

$$E_{2,i} = (1 - \theta_{1,i})(1 - \theta_{2,i}) \quad \text{for } i = 1, 2, \dots, L;$$

$$E_{3,i} = \left[\frac{E_{0,i}^2}{P_{1,i}} + \frac{(1 - E[x_i])(1 - \theta_{1,i})E_{0,i}E_{1,i}}{P_{2,i}} \right];$$

$$E_{4,i} = \left[\frac{1 - E[x_i]}{\lambda_i} E_{0,i} + \frac{E_{2,i}E_{1,i}}{\lambda_i} + \frac{E_{0,i}E_{1,i}}{P_{1,i}} [1 - E_{2,i}] \right];$$

$$E_{5,i} = \left[E_{4,i} - E_{3,i} - \frac{(1 - \theta_{1,i})E_{2,i}E_{1,i}^2}{P_{2,i}} \right] \quad \text{for } i = 1, 2, \dots, L.$$

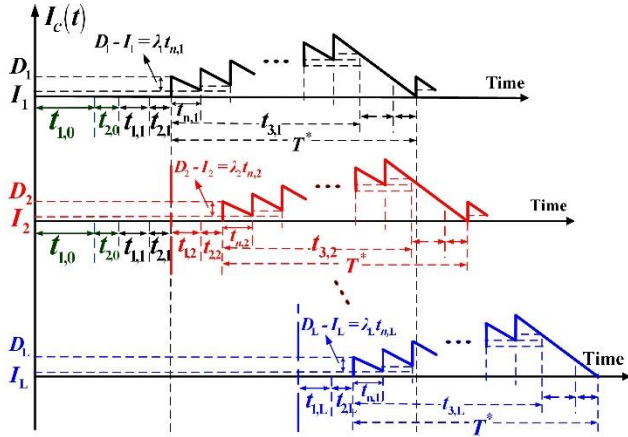


Figure 3. Status of stocks at customer's side in a production cycle (Chiu et al., 2016).

3. Results and discussions

3.1. The proposed algebraic solution process

3.1.1. Phase 1: deciding n^*

By observing $E[TCU(T, n)]$ (i.e., Eq. (3)), we found that two decision variables are in the forms of T^{-1} , T , $n^{-1}T$, and nT^{-1} . Let y_1, y_2, y_3, y_4 , and y_5 represent the following:

$$y_1 = [C_0\lambda_0E_{0,0} + C_{R,0}\lambda_0(1-\theta_{1,0})E_{1,0} + C_{S,0}\lambda_0\phi_0E_{1,0}] + \sum_{i=1}^L (C_i\lambda_iE_{0,i} + C_{R,i}\lambda_i(1-\theta_{1,i})E_{1,i} + C_{S,i}\lambda_i\phi_iE_{1,i} + C_{T,i}\lambda_i) \quad (4)$$

$$y_2 = K_0 + \sum_{i=1}^L K_i \quad (5)$$

$$y_3 = \sum_{i=1}^L (K_{1,i}) \quad (6)$$

$$y_4 = \omega_0 T + \sum_{i=1}^L \left\{ \frac{h_{1,i}\lambda_i^2}{2} (E_{3,i} + E_{4,i}) + \frac{h_{2,i}\lambda_i^2 E_{1,i}^2}{2} \left[\frac{(1-\theta_{1,i})^2}{P_{2,i}} \right] + \frac{h_{3,i}\lambda_i^2}{2} \left[\frac{2E_{0,i}}{P_{1,i}} + \frac{2(1-\theta_{1,i})E_{1,i}}{P_{2,i}} - \frac{1}{\lambda_i} + E_{5,i} \right] + Th_{4,i}\lambda_i E_{1,i} \right\} \quad (7)$$

$$y_5 = + \sum_{i=1}^L \left\{ -\frac{h_{1,i}\lambda_i^2}{2} + \frac{h_{3,i}\lambda_i^2}{2} \right\} (E_{5,i}) \quad (8)$$

Then, $E[TCU(T, n)]$ becomes the following:

$$E[TCU(T, n)] = y_1 + y_2 T^{-1} + y_3 (nT^{-1}) + y_4 T + y_5 (Tn^{-1}) \quad (9)$$

We rearrange Equation (9) as follows:

$$E[TCU(T, n)] = y_1 + (\sqrt{y_2} - \sqrt{y_4} T)^2 T^{-1} + (\sqrt{y_3} - \sqrt{y_5} Tn^{-1})^2 nT^{-1} + 2\sqrt{y_2} y_4 + 2\sqrt{y_3} y_5 \quad (10)$$

Equation (10) indicates that if both the 2nd and 3rd terms are zeros, then $E[TCU(T, n)]$ is minimized. Therefore, by substituting y_2, y_3, y_4 , and y_5 in Eq. (10), we have

$$n^* = \frac{\left(K_0 + \sum_{i=1}^L K_i \right) \sum_{i=1}^L \left[\frac{\lambda_i^2}{2} (h_{3,i} - h_{1,i}) E_{5,i} \right]}{\left\{ \left(\sum_{i=1}^L K_{1,i} \right) \left\{ \omega_0 + \sum_{i=1}^L \left[\frac{h_{1,i}\lambda_i^2}{2} (E_{3,i} + E_{4,i}) + \frac{h_{2,i}\lambda_i^2 E_{1,i}^2}{2} \left[\frac{(1-\theta_{1,i})^2}{P_{2,i}} \right] + h_{4,i}\lambda_i E_{1,i} + \frac{h_{3,i}\lambda_i^2}{2} \left[\frac{2E_{0,i}}{P_{1,i}} + \frac{2(1-\theta_{1,i})E_{1,i}}{P_{2,i}} - \frac{1}{\lambda_i} + E_{5,i} \right] \right] \right\} \right\}} \quad (11)$$

3.1.2 Phase 2: deciding T^*

Once n has been decided, reconsidering $E[TCU(T, n)]$ as a function of the single decision variable. Therefore, we can rearrange Eq. (9) as follows:

$$E[TCU(T, n)] = y_1 + (y_2 + y_3 n) T^{-1} + (y_4 + y_5 n^{-1}) T \quad (12)$$

and Eq. (12) can be rearranged further as

$$E[TCU(T, n)] = y_1 + \left(\sqrt{y_2 + y_3 n} - T \sqrt{y_4 + y_5 n^{-1}} \right)^2 T^{-1} + 2\sqrt{y_2 + y_3 n} \sqrt{y_4 + y_5 n^{-1}} \quad (13)$$

Now, if the 2nd term of Eq. (13) equals to zero, then $E[TCU(T, n)]$ will be minimized. That is

$$T^* = \sqrt{\frac{y_2 + y_3 n}{y_4 + y_5 n^{-1}}} \quad (14)$$

By substituting y_2, y_3, y_4 , and y_5 in Eq. (14), we obtain

$$T^* = \frac{K_0 + \sum_{i=1}^L (K_i + nK_{1,i})}{\left\{ \omega_0 + \sum_{i=1}^L \left[\frac{h_{1,i}\lambda_i^2}{2} \left[E_{3,i} + E_{4,i} - \frac{E_{5,i}}{n} \right] + \frac{h_{2,i}\lambda_i^2 E_{1,i}^2}{2} \left[\frac{(1-\theta_{1,i})^2}{P_{2,i}} \right] + h_{4,i}\lambda_i E_{1,i} + \frac{h_{3,i}\lambda_i^2}{2} \left[\frac{2E_{0,i}}{P_{1,i}} + \frac{2(1-\theta_{1,i})E_{1,i}}{P_{2,i}} - \frac{1}{\lambda_i} + \left(1 + \frac{1}{n} \right) E_{5,i} \right] \right] \right\}} \quad (15)$$

It is noted that the above equations (11) and (15) are identical to the results derived by Chiu et al. (2016). Using the same solution procedure as proposed in Chiu et al. (2016), $E[TCU(T, n)]$ of this particular system can also be gained.

4. Conclusion

A multiproduct fabrication-shipment problem with postponement and product quality reassurance was studied by Chiu et al. (2016) employing mathematical modeling and calculus to derive the optimal operating decisions. This study proposes a two-phase algebraic method in lieu of the derivatives they used to show that the same solutions can be gained without applying the derivatives. This simplified approach aims at helping practitioners in the fields (who may lack of knowledge in calculus), better comprehend and manage such a particular fabrication-shipment system.

Appendix A

T =cycle time, a decision variable,
 n =number of shipments, another decision variable,
 α =common part's completion rate,
 $E[T]$ =expected cycle length,
 $TC(T, n)$ = total system costs per cycle,
 $E[TC(T, n)]$ = expected total system costs per cycle,
 $E[TCU(T, n)]$ = the long-run average system costs per unit time,
 Notation used in stage 1 (where $i = 0, 1, 2, \dots, L$) is as follows:
 Q_i =lot size for product i ,
 K_i =setup cost,
 C_i =unit production cost,
 $C_{R,i}$ =unit rework cost,
 $C_{S,i}$ =unit disposal cost,
 ϕ_i =total scrap rate for product i , (where $0 \leq \phi_i \leq 1$),
 $h_{1,i}$ =unit holding cost,
 $h_{2,i}$ =reworked item's unit holding cost,
 $h_{4,i}$ =safety stock's unit holding cost,
 $t_{1,i}$ =uptime for product i ,
 $t_{2,i}$ =rework time,
 $t_{3,i}$ =delivery time,
 $H_{1,i}$ =maximal stock level at the end of regular process,
 $H_{2,i}$ =maximal stock level at the end of rework process,
 $I(t_i)$ =stock level at time t ,
 $I_S(t_i)$ =scrap stock level at time t .
 Notation used in stage 2 (where $i = 1, 2, \dots, L$) is as follows:
 H_i =common part level at the time of fabricating end product i ,
 $K_{1,i}$ =fixed shipping for product i ,
 $C_{T,i}$ =unit shipping cost,
 $t_{n,i}$ =a fixed time interval between two consecutive deliveries,
 $h_{3,i}$ =buyer's unit stock holding cost,
 l_i =the left-over stocks in each $t_{n,i}$,
 D_i =number of stocks per delivery.

References

- Abbad, M., Zahratahdi, T. (2008). An algorithm for achieving proportional delay differentiation. *Operations Research Letters*, 36(2), 196-200.
- Arun, P., Lincon, S.A., Prabhakaran, N. (2019) An automated method for the analysis of bearing vibration based on spectrogram pattern matching. *Journal of Applied Research and Technology*, 17(2), 126-136.
- Bernstein, F., DeCroix, G.A., Wang, Y. (2011). The impact of demand aggregation through delayed component allocation in an assemble-to-order system. *Management Science*, 57(6), 1154-1171.
- Boone, T., Ganeshan, R., Guo, Y., Ord, J.K. (2000). The impact of imperfect processes on production run times. *Decision Sciences*, 31(4), 773-785.
- Chiu, S.W., Kuo, J-S., Chiu, V., Chiu, Y-S.P. (2016). Cost minimization for a multi-product fabrication- distribution problem with commonality, postponement, and quality assurance. *Mathematical and Computational Applications*, 21(3), Art. No. 38, 1-17.
- Chiu, S.W., Wu, H.Y., Chiu, Y-S.P., Hwang, M.-H. (2018) Exploration of finite production rate model with overtime and rework of nonconforming products. *Journal of King Saud University - Engineering Sciences*, 30(3): 224-231, 2018-7.
- Chiu, S.W., Kuo, J-S., Chiu, Y-S.P, Chang, H-H. (2019). Production and distribution decisions for a multi-product system with component commonality, postponement strategy and quality assurance using a two-machine scheme. *Jordan Journal of Mechanical and Industrial Engineering*, 13(2), 105-115.
- Grubbstrom, R.W., Erdem, A. (1999). The EOQ with backlogging derived without derivatives. *International Journal of Production Economics*, 59, 529-530.
- He, G., Wu, W., Zhang, Y. (2018). Analysis of a multi-component system with failure dependency, N-policy and vacations. *Operations Research Perspectives*, 5, 191-198.
- Inderfurth, K., Janiak, A., Kovalyov, M.Y., Werner, F. (2006). Batching work and rework processes with limited deterioration of reworkables. *Computers and Operations Research*, 33(6), 1595-1605.

- Kumar, S., Goyal, A., Singhal, A. (2017). Manufacturing flexibility and its effect on system performance. *Jordan Journal of Mechanical and Industrial Engineering*, 11(2), 105-112.
- Lin, H-D., Chen, Y-R., Chiu, V., Chiu, Y-S.P. (2019). A decision model for a quality-assured EPQ-based intra-supply chain system considering overtime option. *Journal of Applied Engineering Science*, 17(3), 361-371.
- Makarova, I., Shubenkova, K., Mavrin, V., Boyko, A. (2017). Ways to increase sustainability of the transportation system. *Journal of Applied Engineering Science*, 15(1), 89-98.
- Mehdizadeh, E., Gholami, H., Naderi, B. (2018). A robust optimization model for multi-product production planning in terms of uncertainty of demand and delivery time. *Economic Computation and Economic Cybernetics Studies and Research*, 52(4), 227-240.
- MohanDas, C.D., Ayyanar, A., Susaiyappan, S., Kalimuthu, R. (2017). Analysis of the effects of fabrication parameters on the mechanical properties of Areca fine fiber-reinforced phenol formaldehyde composite using Taguchi technique. *Journal of Applied Research and Technology*, 15(4), 365-370.
- Muralidharan, R., Vallavaraj, A., Mahanti, G.K., Patidar, H. (2017). QPSO for failure correction of linear array of mutually coupled parallel dipole antennas with desired side lobe level and return loss. *Journal of King Saud University - Engineering Sciences*, 29(2), 112-117.
- Nielsen, I.E., Saha, S. (2018). Procurement planning in a multi-period supply chain: An epiphany. *Operations Research Perspectives*, 5, 383-398.
- Porteus, E.L. (1986). Optimal lot sizing, process quality improvement and setup cost reduction. *Operations Research*, 34,137-144.
- Swaminathan, J.M., Tayur, S.R. (1998). Managing broader product lines through delayed differentiation using vanilla boxes. *Management Science*, 44(12 Part 2), S161-S172.
- Taft, E.W. (1918). *The most economical production lot*. *Iron Age*, 101, 1410-1412.
- Thomas D.J., Hackman S.T. (2003). A committed delivery strategy with fixed frequency and quantity. *European Journal of Operational Research*, 148(2), 363-373.
- Villarreal, A., Garbarino, G., Riani, P., Gutiérrez-Alejandre, A., Ramírez, J., Busca, G. (2018). Influence of incorporating a small amount of silica on the catalytic performance of a MoO₃/Al₂O₃ catalyst in ethanol oxidative dehydrogenation. *Journal of Applied Research and Technology*, 16(6), 484-497.
- Zhang, Y., Wu, W., Tang, Y. (2017). Analysis of an k-out-of-n:G system with repairman's single vacation and shut off rule. *Operations Research Perspectives*, 4, 29-38.