

Sustainable Closed-Loop Supply Chain Network Design with Joint Economic Lot Sizing Problem

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Abstract: In the fashion of sustainable development and a friendly environment, the traditional supply chain, specifically network design, is no longer sufficient for sustainability needs. With the increasing demand for them and toward a holistic model in a sustainable closed-loop supply chain (SCLSC), we proposed the network design and joint economic lot-sizing problems (JELS). The new problem is modeled as a mixed-integer multi-objective problem, considering the three factors of sustainability: maximizing total profit, minimizing the total amount of CO₂ emissions, and maximizing social job opportunities. In order to achieve optimal decisions of facility location, amount flow, cycle time, and the number of shipments delivered in each inventory. The experiment analysis used five performance indicators compared to the non-dominated sorting genetic algorithm-II. Finally, the results show that the proposed model of SCLSC network design with JELS is reaching a more accurate and better decision in all three aspects of sustainability.

Keywords: Sustainable Supply Chain Management; Supply Chain Network Design; Closed-Loop Supply Chain; Joint Economic Lot Sizing; Lot Sizing; Sustainable Closed-loop Supply Chain.

1 Introduction

One of the most complex strategic decision problems in any supply chain is its network design problem [1]. This complexity is because, in this problem, we should determine the optimal long-term planning for the whole supply chain by giving the number, capacity, location, type, and more configuration parameters for each facility. These decisions, in return, will affect the profit, operation cost, risk resistance, and other products of the supply chain. Furthermore, the dramatic increase in the number of industries and globalization consuming our planet's limited resources lead to increased environmental pollution and social problems. These problems make the countries concerned about those problems worldwide [2, 3, 4]. And because of its continuous influences, the enterprises have been forced to consider them together with the economic performance, which leads to adapting their operation mode from traditional supply chain network (SCND) to green or sustainable supply chain network design (GSCND or SSCND).

The sustainable development definition is "to meet the current requirements without compromising the ability of future generations to meet their own needs" [1]. This definition interprets the triple bottom line concept for sustainability which consists of people - which represents the social issues (Soc.), planet - which means the environmental issues (Env.), and profit - which represents the economic issues (Eco.) [5]. Moreover, [6, 7] illustrated that there is still a gap in

the area of supply chain network design (SCND).

Mainly, the closed-loop supply chain (CLSC) network design problems consist of two parts; these parts are resolved simultaneously, forward logistics (FL) and reverse logistics

(RL) for every echelon in the network. In FL, the flow of materials and information moves from suppliers, manufacturers, distribution centers, and retailers until reaching final customers. However, in RL, the flow of materials (mainly the returned materials) and information move from customers returned to collection and recovery centers or disposal centers [8]. One of the significant trends in CLSC design problems is sustainable CLSC design problems [9]. They are concerned with solving economic, environmental, and social factors to simultaneously construct a more effective long-term network plan for both FL and RL.

To seek global optimization for network design problems considering, one of the factors that affect the SCND as same as SSCND is inventory decisions [10]. Their costs and replenishment decisions directly affect the economic part by adding inventory costs as well as the environmental part by transportation emissions and the social part by creating an employee job opportunity. In the previous research, lot-size models are concerned with coordination between buyer and vendor in inventory replenishment decisions, called joint economic lot size (JELS) models. These models minimize the total cost by determining inventory decisions like order, production, and shipment quantities.

Since most SCND, SSCND, CLSC designs, and JELS problems are classified as NP-hard problems [11], achieving the expected performance is quite challenging. Many techniques are used to solve these problems, like stochastic programming, heuristic techniques, fuzzy-based approaches, Bayesian network, etc. [12]. Ideally, the best strategy to achieve a sustainable supply chain is to construct it as a closed loop. This construction causes a more friendly environment (reducing waste and gas emissions) with more efficient economic factors [13, 14, 12]. Accordingly, this section provides the literature background in two areas: sustainable CLSC network design and closed-loop JELS.

Firstly, in the SCLSC network design problem, much recent research dealt with only one sustainable factor. In contrast, a few works have considered the mixed objectives [6, 15, 16], which we will focus on this topic here. We can categorize the mass sustainability concerns (Eco., Env., Soc.), JELS's decision, and facility location's decision (LD), as shown in Table 1.

[17] studied the design of sustainable CLSC by solving a single objective to minimize the total supply chain costs and the total CO₂ emissions to find the best amounts of products based on the network. They solved their model by simulation method. Furthermore, they did not consider the location decisions in their model. In contrast, [18] considered the location decisions for facilities were to establish it or not, in addition to the inventory decision about the best order size, as the product amount flow in their CLSC. Their model also a single objective minimization for total cost and gas emission, but the gas was carbon footprint in this research. Therefore, they solve the problem using AMPL for CPLEX. Later, some researchers studied the models for network design CLSC to determine the product flow and facility locations

with the same idea as the previously reviewed models with a single-objective minimization of total cost and total emission. However, those researchers solved them as a bi-objective model using Fuzzy Programming like [19] and [20].

Some other studies work on the CLSC network design for economic and social factors only like in [22]'s paper. They determined product flow, facility location, operations system,

Table 1 Literature work classification

Citation	Eco.	Env.	Soc.	JELS	LD
[17]	•	•			
[18]	•	•			•
[6]	•	•	•		•
[19]	•	•			•
[16]	•	•	•		•
[20]	•	•			•
[15]	•	•	•		
[21]	•	•	•		•
[22]	•		•		•
[9]	•	•	•		•
<i>Our work</i>	•	•	•	•	•

and transportation between facilities to minimize the total cost and maximize customers satisfaction. The model was solved by two heuristic algorithms NSGA-II and MOSA.

Furthermore, much research investigated the three sustainability factors in CLSC network design. [15] studied the three sustainable objectives minimizing the total cost, minimizing environmental factors, and maximizing the social aspects to find the best products flow. The model is solved by network optimization. Additionally, [9] introduced and solved the model of the three factors of sustainability: minimizing the total cost, minimizing the CO₂ emissions, and maximizing the social factors. To indicate the location, amount of production flow, and technology used decisions with three different distribution types: regular delivery, direct shipment, and direct delivery. They solved their model by proposing a Hybrid Genetic Algorithm (pro-HGA) and comparing it with the Genetic Algorithm and Hybrid Genetic Algorithm (HGA) proposed by [23] and [24] based on the Genetic Algorithm (GA) and the Tabu Search (TS).

On another hand, the area of solving the JELS problem in CLSC is considered one of the hot topics [25,26,27]. Many researchers dealt with the CLSC for the one-two-and-three echelon with different production sequences. According to [26] and specifically in the classification of the relevant literature table, about 75% of their citations were done for less than four echelons, while about 21% of work was done for the four echelons (eleven citations for the single echelon from 1967 until 2020, twelve citations for the two echelons from 2009 until 2019, and two citations for the three echelons, seven citations for the four echelon, and one citation for the five echelons). Thus, we focus on the four echelons CLSC introduced by [28]. Their model studied the integrated concerns of the supplier, manufacturer, retailer, and third-

party recycled dealer for both centralized and decentralized decisions. The model considers (1,1) policy that is one manufacturing cycle and one remanufacturing cycle. Hence, [29] extended the work done on this study by studying (M, 1) and (1, P) policies. Where [30] studied the (M, R) policy under deterioration, [31] did the same but under imperfect manufacturing assumption that generated from the quality of the returned items and compared it with (1,1) policy. In contrast, [26] dealt with sustainable industries in general and studied the industry particularly by adding recycled products. Its model considered five echelon CLSC and simultaneous production sequences.

In this paper, toward generating a holistic model for SCLSC to solve network design and JELS simultaneously, we proposed an integrated model from [9] for network design SCLSC with [28] model for JELS in CLSC. First, the conceptual model and problem description are introduced; the model assumptions, notations, and the mathematical model are described in Section 2. Furthermore, Section 3 illustrates the experiment. Then, we present the results and the discussion in Section 4. Finally, the conclusion and future work are discussed in Section 5.

2 Problem Description and Mathematical Model

The proposed SCLSC network-inventory problem is shown in Figure 1. It is based on the model described by [28]. The raw material flows from the suppliers – specifically the material inventory – to the manufacturer and the final products stored in the manufacturer's finished good inventory. Then transferred to the retailer inventory to be sold to customers. Some return to the third-party/collection stores in collector inventory and transferred to the manufacturer to recycle them. Those are stored in the manufacturer used product inventory to reproduce again and so on.

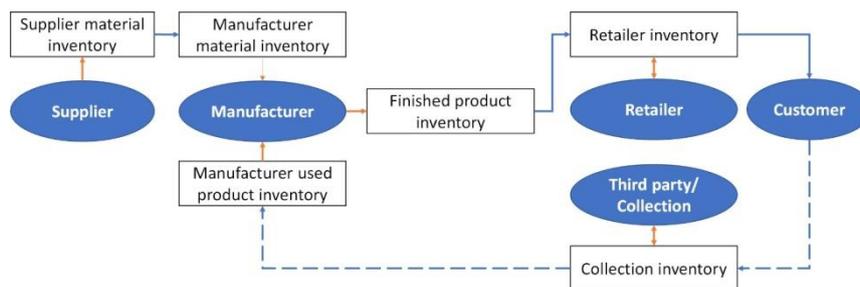


Figure 1 Supply Chain Network

By considering the described network, what is the best facility locations for every facility in our network (suppliers, manufacturer, retailer, and third party)? And, what is the best technology to be used at the chosen manufacturer? Furthermore, what is the best time for the retailer ordering cycle? Finally, what is the best number of delivers per each facility cycle time in both forward and backward flows? Those questions shape the main problem of this paper.

2.1 Assumptions and notations

The mathematical models in this analysis have the following assumptions:

- An infinite planning horizon.
- No deterioration.
- No stock shortages.
- The multi-echelon inventory system contains a single item.
- No space, capacity, or capital constraints.
- No quantity discounts.
- The manufacturing and remanufacturing rates and lead times are constant.
- The product's annual demand rate and the annual return rate are constant, and the annual return rate is less than the annual demand rate.
- The number of deliveries within the manufacturing cycle is an integer.
- The setup cost per run and the annual holding cost fraction are known and constant.
- To meet the retailer demand, remanufactured products are available before the manufactured products.
- The remanufactured products are comparable to newly manufactured products.
- Single suppliers, single manufacturers, single retailers, and single third parties in the closed-loop supply chain.
- The multi-echelon inventory system is considered.
- The unit transportation costs among the manufacturer, the DC, the retailer, the customer, the collection center, the recovery center, and the disposal center are different from each other in value and known beforehand.
- The unit amount of CO₂ emitted during transportation at each stage, and those emitted during manufacture and recovery at the manufacturer and the recovery center, respectively, are different from each other in value and known beforehand.
- The proposed SCLSC design problem is in a steady-state situation.
- The vehicle capacity is constant for all echelons and known beforehand.
- The model annotations are represented by the index sets, the echelon parameters, general parameters (parameters appeared in all echelons), and the decision variables.

Sets:

s	set of suppliers: $s \in \{1, 2, \dots, S\}$
m	set of manufacturers: $m \in \{1, 2, \dots, M\}$
c	set of customers: $c \in \{1, 2, \dots, U\}$
d	set of third parties: $d \in \{1, 2, \dots, \theta\}$
r	set of potential retailers: $r \in \{1, 2, \dots, R\}$

Supplier's Parameters:

	F_{so}	Fixed cost to supplier per order
H_s		Material inventory holding cost to supplier, in percentage per year per dollar
	P_s	Supplier purchase unit price
F_s		Fixed supplier cost to process manufacturer orders of any size
	TP_s	Total supplier profit per unit time
	d_{sm}	Distance from the supplier to manufacturer m , where $s \in \{1, 2, \dots, S\}$ and $m \in \{1, 2, \dots, M\}$
	θ_{sm}	The cost of shipping one unit from supplier to manufacturer m , where $s \in \{1, 2, \dots, S\}$ and $m \in \{1, 2, \dots, M\}$
	e_s	The fixed contract cost for suppliers, where $s \in \{1, 2, \dots, S\}$

General Parameters:

	vc	The capacity that can be shipped in a vehicle
	d_{cd}	Distance from customer c to third party d , where $d \in \{1, 2, \dots, \partial\}$ and $c \in \{1, 2, \dots, U\}$
	θ_{cd}	The cost of shipping one unit from customer c to third party d , where $d \in \{1, 2, \dots, \partial\}$ and $c \in \{1, 2, \dots, U\}$

Parameters Retailers:

	D	Demand annual rate
	O_r	Order cost to the retailer
	H_r	Inventory holding cost to retailer in annual percentage per dollar
	P_c	Retail price
	P_r	Wholesale price
	TP_r	Total retailer profit per unit time
	d_{rc}	Distance from retailer r to customer c , where $r \in \{1, 2, \dots, R\}$ and $c \in \{1, 2, \dots, U\}$
	θ_{rc}	The cost of shipping one unit from retailer r to customer c , where $r \in \{1, 2, \dots, R\}$ and $c \in \{1, 2, \dots, U\}$
	e_r	The fixed contract cost for retailer r , where $r \in \{1, 2, \dots, R\}$

Manufacturer's Parameters:

A_M Setup cost per manufacturing run

A_R Setup cost per remanufacturing run

O_{Mw} Ordering cost for the manufacturer material warehouse

O_{Rw} Ordering cost for the manufacturer used product warehouse

H_M Finished Product inventory holding cost percentage per year per dollar to the manufacturer

H_{Mw} Raw material inventory holding cost percentage per year per dollar for the manufacturer warehouse

H_{Rw} Used product inventory holding cost percentage per year per dollar for the manufacturer warehouse

P_M Manufacturer unit purchase price from the supplier

P_R Manufacturer unit purchase price from the third party

T_{R1} Manufacturer reproduction period in each remanufacturing cycle T_{R2}

Manufacturer non-production period in each remanufacturing cycle T_{M1}

Manufacturer production period in each manufacturing cycle

T_{M2} Manufacturer non-production period in each manufacturing cycle

F_m Fixed cost to the manufacturer for processing buyer orders of any size

P Annual manufacturer production rate ($P > D$)

B Annual manufacturer reproduction rate ($\beta > D$)

I Number of deliveries per remanufacturing/manufacturing cycle time from the manufacturer to the retailer, $I = k_M + k_R$, where I is a positive integer

TP_m Total manufacturer profit per unit time

wc Wight allocated to the created job opportunity

wl Wight allocated to the number of lost days due to damage to work

wu Wight allocated to unemployment

se_m Number of job opportunities created due to the use of technology t at manufacturer m

sl_m Number of lost days due to damage of work caused by the use of technology t at manufacturer m

CO_2V Amount of CO_2 emitted from a vehicle per km

CO_2M Unit amount of CO_2 emitted during the production process at manufacturer m

CO_2R Unit amount of CO_2 emitted during reproduction process at manufacturer m

d_{mr} Distance from manufacturer m to retailer r , where $r \in \{1, 2, \dots, R\}$ and $m \in \{1, 2, \dots, M\}$

θ_{mr} The cost of shipping one unit from manufacturer m to retailer r , where $r \in \{1, 2, \dots, R\}$ and $m \in \{1, 2, \dots, M\}$

e_m Fixed location cost for manufacturer m , where $m \in \{1, 2, \dots, M\}$

Thirdparty'sParameters:

C	Annualreturnrate
A_d	Setupcostperrunforhethirdparty
H_d	Usedproductinventoryholdingcostpercentageperyearperdollar to the third party
P_d	Third-partycollectingunitcostfromtheconsumer
F_d	Fixedthirdpartycosttoprocessmanufacturerordersofany size
TP_d	Totalthird-partyprofitperunittime
d_{dm}	Distancefromthirdpartydtomanufacturerm,whered \in $\{1,2,\dots,\partial\}$ and $m \in \{1,2,\dots,M\}$
θ_{dm}	Thecostofshippingoneunitfromthirdpartydtomanufacturerm, whered $\in \{1,2,\dots,\partial\}$ and $m \in \{1,2,\dots,M\}$
e_d	Thefixedcontractcostforthirdpartyd,whered $\in \{1,2,\dots,\partial\}$

ObjectiveNotations:

F_{eco}	Jointprofitperunittimeforthewholesystemrepresentsthe economic objective
F_{soc}	Socialobjectives
F_{env}	Environmentalobjectives
TT	Totaltransportationcosts
TF	Totalfixedlocationcosts
TC	The total amount of CO ₂ emitted during the production and reproduction process
TCT	The total amount of CO ₂ emitted during the transportation process

DecisionVariables:

T_r	Retailerorderingcycletime
k_M	Numberofdeliveriespermanufacturingcycletimefromthe manufacturer to the retailer
k_R	Numberofdeliveriesperremanufacturingcycletimefromthe manufacturer to the retailer
k_s	Numberofdeliveriesper T_{M1} fromthesuppliertothe manufacturer
k_d	Numberofdeliveriesper T_{R1} fromthethirdpartytothemanufacturer
x_m	The manufacturerisopenedtakevalue1ifnottakezerom \in $\{1,2,3,\dots,M\}$
x_r	The retailerischosentotakevalue1ifnottakevaluezeror \in $\{1,2,3,\dots,R\}$
x_s	Ifthesupplierischosentakevalue1,zerootherwises \in $\{1,2,3,\dots,S\}$
x_d	Ifthethirdpartyischosentakevalue1,zerootherwised \in $\{1,2,3,\dots,\partial\}$
t_m	Takesthevalue1iftechnologytisusedatmanufacturermand0 otherwise

2.2 Model-1:SustainableCLSCnetworkdesignandjointeconomiclotsizing

Totalretailerprofit(Eq.1)iscalculatedas;retailprice*ademand-(totalretailercost);thetotalretailer cost is calculated as; fixed cost + holding cost * wholesale price * inventory level + wholesale price * demand. Thus;

$$TP_r(T_r) = P_c D - \left[\frac{O_r}{T_r} + \frac{H_r P_r D T_r}{2} + P_r D \right] \tag{1}$$

Total manufacturer profit (Eq.2) consists of the wholesale price minus the total costs of the manufacturer, used product warehouse, and material warehouse. Total costs of return/used product consist of ordering cost for used products + holding cost for used productsinventory*manufacturerunitpurchasefromthethirdparty*numberorproducts in inventory + purchasing cost = (manufacturer unit purchase from the third party * demand). Where the total costs of material/raw material inventory consist of ordering cost for material warehouse+holdingcost=(materialinventoryholdingcost*manufacturerunitpurchase pricefromsupplier*quantity)+purchasingcost=(numberofdeliveriespermanufacturing cycle time from manufacturer to retailer * purchase price * Demand); thus;

$$TP_m(k_s, k_d, k_R, k_M, T_r) = P_r D - \frac{A_R + A_M + (k_R + k_s)F_m + k_d O_{Rw} + k_s O_{Mw}}{(k_R + k_s)T_r} - \frac{H_M(k_R P_R + k_M P_M)}{k_R + k_M} * \left[\frac{T_r D(2D + k_R \beta - k_R D)k_R}{2\beta(k_R + k_M)} + \frac{T_r D(2D + k_M P - k_M D)k_M}{2P(k_R + k_M)} - \frac{T_r D}{2} \right] - \frac{H_{Rw} P_R D^2 k_R^2 T_r}{2k_d(k_R + k_M)\beta} - \frac{H_{Mw} P_M D^2 k_M^2 T_r}{2k_s(k_R + k_M)P} - \frac{(k_R P_R + k_M P_M)D}{k_R + k_M} \tag{2}$$

The total third-party profit (Eq.3) shown below is calculated as manufacturer unit purchase price from third-party minus (setup cost plus fixed cost-plus holding cost plus collecting cost).

$$TP_d(k_d, k_R, k_M, T_r) = \frac{(P_R - P_d)Dk_R}{k_R + k_M} - \frac{A_d + k_d F_d}{(k_R + k_M)T_r} - \frac{H_d P_d T_r \left[(D^2 k_R^2 - 2Ck_R D(k_R + k_M))(k_d - 1) + C\beta k_d(k_R + k_M)^2 \right]}{2(k_R + k_M)\beta k_d} \tag{1}$$

Totalsupplierprofit(Eq.4)iscalculatedasthemanufacturerunitpurchasepricefrom the supplier (The fixed cost to supplier per order plus the number of deliveries multiplied by the fixed supplier cost of any order size) plus holding cost-plus purchase price.

$$TP_s(k_s, k_R, k_M, T_r) = \frac{(P_M - P_s)Dk_M}{k_R + k_M} + \frac{F_{s0} + k_s F_s}{(k_R + k_M)T_r} + \frac{H_s P_s D^2 k_M^2 T_r}{2P(k_R + k_M)} \left(1 - \frac{1}{k_s} \right) \tag{2}$$

The total amount of CO₂ emitted during production and recovery processes (Eq.5) is equal to the capacity of the selected manufacturer multiplied by the unit amount of CO₂ emitted during the production process at the manufacturer.

$$TCP(x_m) = \sum_m^M [P x_m CO_{2M} + \beta x_m CO_{2R}] \tag{3}$$

The total amount of CO₂ emitted during transportation processes (Eq.6) equals the amount of CO₂ ejected from a vehicle multiplied by the summation of distances from different selected facilities.

$$TCT(x_m, x_r, x_d, x_s) = CO_2V \left[\sum_s^S \sum_m^M \left(d_{sm} x_s x_m \frac{Dk_M}{(k_R+k_M)vc} \right) + \sum_m^M \sum_r^R \left(d_{mr} x_m x_r \frac{D}{vc} \right) + \sum_r^R \sum_c^U \left(d_{rc} x_r \frac{D}{vc} \right) + \sum_c^U \sum_d^\partial \left(d_{cd} x_d \frac{Dk_R}{(k_R+k_M)vc} \right) + \sum_d^\partial \sum_m^M \left(d_{dm} x_d x_m \frac{Dk_M}{(k_R+k_M)vc} \right) \right] \tag{4}$$

Total annual transportation cost (Eq.7) equals the summation of yearly transportation cost multiplied by the distances between selected facilities.

$$TTC(x_s, x_m, x_r, x_d) = \sum_s^S \sum_m^M \left(\theta_{sm} x_s x_m \frac{Dk_M}{(k_R+k_M)} \right) + \sum_m^M \sum_r^R (\theta_{mr} x_r x_m D) + \sum_r^R \sum_c^U (\theta_{rc} x_r D) + \sum_c^U \sum_d^\partial \left(\theta_{cd} x_d \frac{Dk_R}{(k_R+k_M)} \right) + \sum_d^\partial \sum_m^M \left(\theta_{dm} x_d x_m \frac{Dk_M}{(k_R+k_M)} \right) \tag{5}$$

Total Fixed association Costs (Eq.8) equal fixed costs for selecting this facility.

$$TFC(x_r, x_m, x_s, x_d) = \sum_r^R e_r x_r + \sum_m^M e_m x_m + \sum_s^S e_s x_s + \sum_d^\partial e_d x_d \tag{6}$$

The first objective (Eq.9) is to maximize the total profit for retailers, manufacturers, third-party, and suppliers minus (the total fixed association costs plus total annual transportation costs). The second objective (Eq.10) is to maximize the social factor, which is equal to the weight allocated to the created job opportunity at the manufacturer minus the weight allocated to lost day caused by worker's damage * number of lost days caused by worker's damage at the manufacturer) minus (weight allocated to unemployment * amount of unemployment at the manufacturer). The third objective (Eq.11) is to minimize the environmental factor that is equal to the total amount of CO₂ emitted during production and recovery processes (TCP) plus the total amount of CO₂ emitted during transportation processes (TCT).

Finally, the model's constraints (Eq.12-16) are the ratio of the number of deliveries per manufacturer cycle time over the number of deliveries per retailer cycle time, equal to the (demand minus annual return rate) over yearly return rate. The number of deliveries per

manufacturer cycle time and the number of deliveries per retailer cycle time is equal to the number of deliveries per re-manufacturer/manufacturer cycle time. The logic constraints and only one facility should be selected per echelon.

$$\text{Max. } F_{eco}(k_R, k_M, T_r, k_s, x_s, x_m, x_r, x_d) = TP_r + TP_m + TP_s + TP_d - (TFC + TTC) \quad (7)$$

$$\begin{aligned} \text{Max. } F_{soc}(x_m, t_m) \\ = wc * \sum_m^M (se_m * x_m * t_m) - wl * \sum_m^M (sl_m * x_m * t_m) - wu * \sum_m^M (su_m * x_m) \end{aligned} \quad (8)$$

$$\text{Min. } F_{env}(x_m, x_r, x_d, x_s) = TCP + TCT \quad (9)$$

Subject to:

$$\frac{k_M}{k_R} = \frac{D - C}{C} \quad (10)$$

$$k_R + k_M = I \quad (11)$$

$$T_r \geq 0 \quad (12)$$

$$k_s, k_d, I = 1, 2, 3, \dots \quad (13)$$

$$\sum_{s=1}^S x_s = 1, \sum_{m=1}^M x_m = 1, \sum_{p=1}^{\delta} x_p = 1, \text{ and } \sum_{r=1}^R x_r = 1 \quad (14)$$

2.3 Model-2: Joint economic clotsizing for CLSC

Were introduced the previous work done by [28] precisely the JELS concept in our network for the comparison responses. The model here determines the retailer cycle time and the number of delivers per facility cycle time in each flow. The only objective here is maximizing the total profit, representing only the economic factor as in reference work [28].

To convert the previous work done by [28] to be multi-objective, we put the social and environmental objectives equal to zero to solve this model by NSGA-II. The second change is removing the additional costs for the location decisions (TFC and TTC) to remove the impact of choosing

the facility location. The objectives will be Eq.(17-19).

$$Max.F_{eco}(k_R, k_M, T_r, k_s) = TP_r + TP_m + TP_s + TP_d \tag{17}$$

$$Max.F_{soc}(x_m, t_m) = 0 \tag{18}$$

$$Min.F_{env}(x_m, x_r, x_d, x_s) = 0 \tag{19}$$

3 Experiments

This paper uses the previous data generated by [28,9]—as shown in Table 2—and applies these data mainly on three scales: small, medium, and significant in every model—shown in Table 3. Furthermore, we used non-dominated sorting genetic algorithm-II (NSGA-II) [32] because this algorithm is the most algorithm used to solve this kind of problem in the literature. Finally, we try six differential algorithm parameters shown in Table 4 because we seek to find the best solutions for these problems.

To overlap the randomness, we run every case for every scale of 30 runs with the same parameters and evaluate the output by computing the non-dominated sorting to find the first front for these 30 runs. All runs were done using Julia Language in Intel® Core™ i7 CPU Q720 @ 1.60 GHz and 16 GB RAM. Moreover, different performance metrics are calculated for the proposed model to evaluate the algorithm.

Table 2 Parameters Setting

Value	Parameter	Value	Parameter	Value	Parameter	Value	Parameter	Value	Parameter
2000	<i>D</i>	770	<i>C</i>	2000	<i>AM</i>	0.25	<i>HM_w</i>	U[90,100]	<i>sem</i>
100	<i>Or</i>	250	<i>Ad</i>	2500	<i>AR</i>	0.25	<i>HR_w</i>	U[20,30]	<i>slm</i>
0.3	<i>H_r</i>	0.35	<i>H_d</i>	350	<i>OM_w</i>	115	<i>PM</i>	150	<i>vc</i>
175	<i>P_c</i>	70	<i>P_d</i>	350	<i>OR_w</i>	110	<i>PR</i>	0.3	<i>H_s</i>
150	<i>P_r</i>	150	<i>F_d</i>	0.2	<i>HM</i>	350	<i>F_m</i>	90	<i>P_s</i>
500	<i>P</i>	U[15,20]	<i>sum</i>	150	<i>F_s</i>	U(75,950)	<i>d_{cd}</i>	U(1,5)	<i>θ_{cd}</i>
400	<i>B</i>	3	<i>CO_{2V}</i>	U(75,950)	<i>d_{rc}</i>	U(75,950)	<i>d_{dm}</i>	U(1,5)	<i>θ_{mr}</i>
0.1	<i>w_c</i>	0.5	<i>CO_{2M}</i>	U(10,50)	<i>θ_{rc}</i>	U(1,5)	<i>θ_{dm}</i>	U(1000,5000)	<i>em</i>
0.1	<i>w_l</i>	0.3	<i>CO_{2R}</i>	U(100,500)	<i>er</i>	U(100,500)	<i>ed</i>	U(1,5)	<i>θ_{sm}</i>
0.1	<i>w_u</i>	200	<i>F_{so}</i>	U(75,950)	<i>d_{mr}</i>	U(75,950)	<i>d_{sm}</i>	U(100,500)	<i>es</i>

5	50)	50)	00)
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Table 3 Scales setting Scales Supplier

Manufacturer Retailers Customers Third- Numbers (Snum)

(Mnum)

(Rnum)

(Unum)

party

of	Decision					(Dnum)	Variables (dv_size)
1	3	3	3	2	2	19	
2	30	30	30	15	15	140	
3	60	60	60	30	30	275	

Tabl Casessett

e4	ing	iter_nu	pCross	pMut
Cases	popSize	m		
1	80	800	0.25	0.1
2	100	1000	0.5	0.2
3	120	1200	0.75	0.3
4	60	800	0.25	0.1
5	140	1200	0.75	0.3
6	140	1200	0.25	0.1

Theperformancemetricsusedhereare(1)theerrorratio(ER)forassessingthenumber ofPareto-optimalsolutionsinthesetwherethelowervalueisthebest,(2)Thegenerational

distance (GD) for measuring the closeness of the solutions to the true Pareto-front, the low value is the best, (3) the uniform distribution (UD) for focusing on the distribution of the solutions, (4) the maximum spread (MS) concerning the spread of the solutions a large value is the best, and (5). The maximum value is the best for the hyper area ratio (HAR), which considers closeness and diversity together. For more information, you can read [33].

4 Results and Discussions

In this section, we provide first—the comparison between the two models—second—the performance of NSGA-II in solving the proposed model on the different scales and cases. The comparison was made for the first-front solutions. This first front is interpreted from the non-dominated sorting algorithm for all first front of the last iteration for every run of 30 runs.

The results of the final objectives are represented in Table 5; Comparing the profit objective, Model-2 is generally better than Model-1 since it is mainly concerned with only this objective, so it makes sense to get these results. However, the Model-1 achieves better profit in some cases (Scale 1: Cases 3 and 6, Scale 2: Cases 1 and 5), which shown in the italic format in Table 5. Also, Model-1 is the best profit for all runs which is 127,336.23 \times 10,000 which indicates that our proposed model for integrating JEL and SCLSC network the design makes a better decision not only for the social and environmental issues but also for economic (profit).

The result of Model-1 shows that the algorithm reaches the maximum social value on all scales. The best environment objective values have been reached two times in Scale 1 and Scale 2. The worst environment value is 0.49% more than the best solution on Scale 1, 18.7% on Scale 2, and 3.15% on Scale 3. So, in general, we can conclude that the NSGA-II algorithm can solve this model effectively.

Table 5 Runs results

Model-1

Scale	1	2	3						
Cas e	Profit* 10,000	Soc ial	Enviro nment	Profit* 10,000	Soc ial	Enviro nment	Profit* 10,000	Soc ial	Enviro nment
1	10,945.61	7.89	35,408.33	8,891.38	9.19	248,140.35	35.42	9.03	551,293.22
2	2,886.81	7.89	35,275.80	946.77	9.19	203,446.17	41.62	9.03	542,935.28
3	127,336.23	7.89	35,408.33	10,844.89	9.19	203,951.40	10.65	9.03	541,661.46

4	1,688.94	7.89	35,447.93	73.86	9.19	248,978.37	2.33	9.03	551,477.13	
5	16,634.34	7.89	35,275.80	8,398.87	9.19	202,403.53	1,993.06	9.03	545,553.46	
6	43,391.44	7.89	35,305.57	2,565.07	9.19	202,403.53	21.4	9.03	559,294.12	
Model-2Scale	1			2			3			
	Cas e	Profit* 10,000	Soc ial	Enviro nment	Profit* 10,000	Soc ial	Enviro nment	Profit* 10,000	Soc ial	Enviro nment
1	11,210.30	0	0	5,388.38	0	0	3,483.38	0	0	
2	20,045.10	0	0	7,300.61	0	0	109,007.00	0	0	
3	94,722.90	0	0	34,071.99	0	0	75,139.50	0	0	
4	2,093.93	0	0	3,469.49	0	0	823.5	0	0	
5	34,010.00	0	0	2,975.91	0	0	66,354.30	0	0	
6	984.4	0	0	71,814.30	0	0	118,465.00	0	0	

The performance impacts illustrated in Table 5 and Figures 2, 3, and 4 represent each problem scale on the first proposed model; not that; all of them are modified by removing

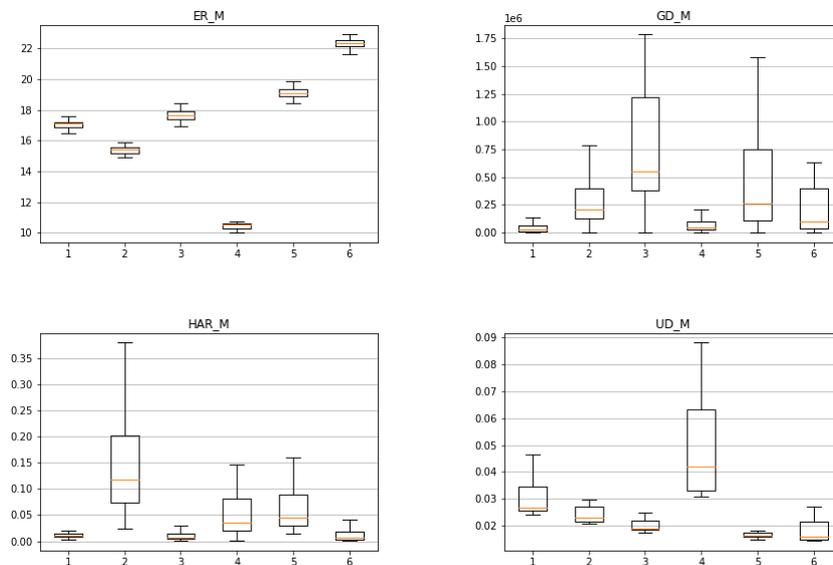


Figure 2 Scale-1 performance indicators

the outliers.

For the first scale, the small one (Figure 2), case 4 is the best performance for the Pareto optimal solutions (ER), the solution distributions (UD), and the second-best for the closeness (GD) of the solutions but not the diversity. The best performance for both closeness and diversity (HAR) is case 2. Furthermore, the spread metric (MS) has no values for this scale because the social objective's values are the same in all front solutions, so this metric cannot be evaluated. In contrast, the worst cases are 6 for ER and 3 for GD. Also, case 3 is the second-worst for HAR and UD.

On the second scale, Figure 3, case 3 is the best case for reaching the minimum ER. Case 5 has the best observations for the closeness of solutions (GD) and the second-best is case 3 if we consider the minimum and the most observations values from 25% up to 75%. While case 4; achieves the best performance for distribution (UD); in addition to both closeness and diversity (HAR). However, it's hard to define the best spread case precisely; also, we can argue that the best case here is case 3 because it has 25% up to 75% of observations higher than the remaining cases. As a result; we can conclude that case 3 is the best ER and MS and the second-best for GD, while case 4 is the best in the remaining indicators. The worst cases are case 6 for ER, case 2 for GD, cases 1, 3, and 5 for HAR, case 3 for UD, and case 4 for MS.

For the last scale in Figure 4, the best case for ER and HAR is case 3. Case 6 is the best for GD followed by case 3. While case-4 is the best case for UD. Case 5 is the best

for MS followed by case 3. Hence, case 3 is the best Pareto front solutions, closeness, and diversity; it comes in second place after case 5 for solutions spread. Case 6 is the best for closeness but not for diversity. So case 3 is considered the best case on the scale-3.

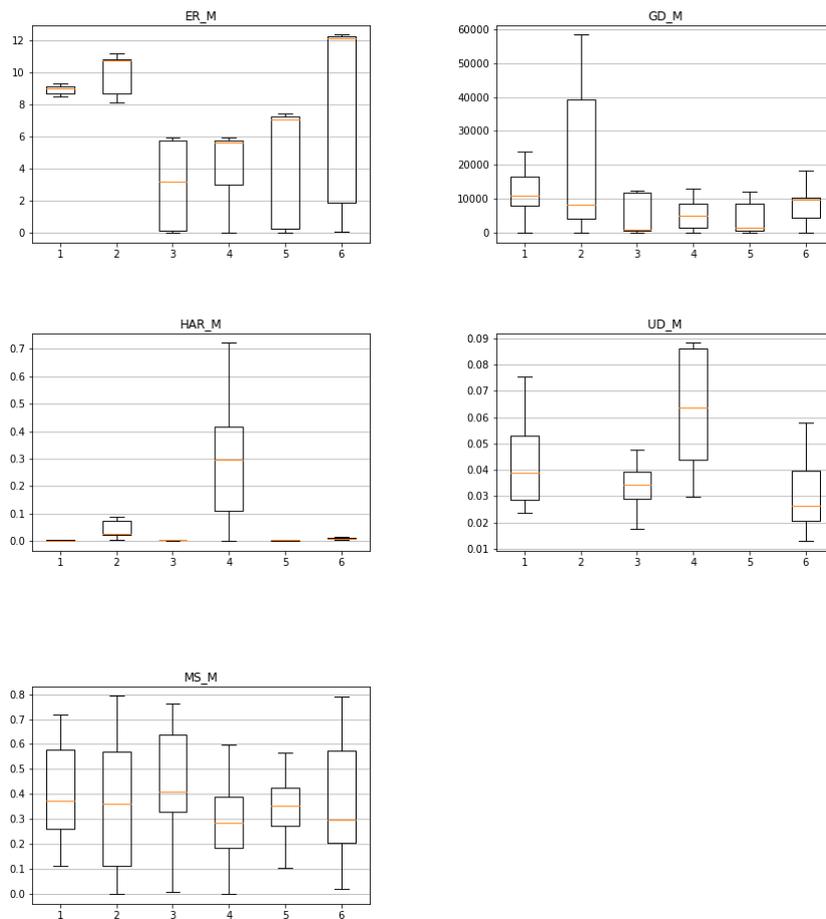


Figure 3 Scale-2 performance indicators

In conclusion, as represented in Table 6, the best case for the most performance metrics is case 4 concerning scale-1 and some of the performance metrics for scales 2 and 3. While case 3 is the best for remaining scales in most performance metrics. So, we can conclude that the medium "popSize" with high "iter_num" and maximum "pCross" and "pMute" is needed for medium and large-scale problems, while the small "pCross" and "pMute" with minimum "popSize" and "iter_num" are the best performance for small scale problem.

Table 6 Summary of performance conclusions

Scale	ER	GD	HAR	UD	MS
1	Case 4	Case 4	Case 2	Case 4	
2	Case 3	Case 5 then	Case 4	Case 4	Case 3
3	Case 3	Case 6 then	Case 3	Case 4	Case 5 then 3

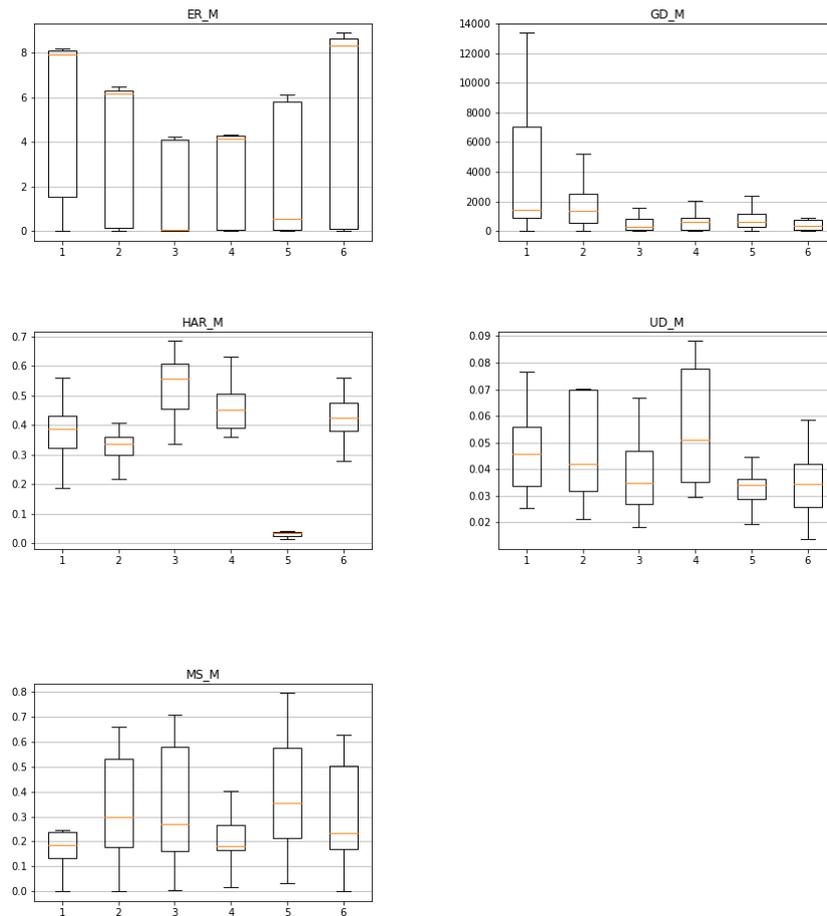


Figure 4 Scale-3 performance indicators

5 Conclusion and future work

In this paper, we proposed and solved the integrated model of joint economic lot-sizing problem with network design problem; we solved the problem under the three sustainability objectives for the closed-loop supply chain. The first objective is maximizing the total profit; the second is maximizing the social impact, while the third objective minimizes the environmental impact caused by transportation.

The problem was designed considering the network location for one supplier, one manufacturer, one retailer, one customer zone, and one third party, in addition to selecting the technology used in the manufacturer, the basic cycle time, and the number of deliveries per every echelon cycle time. The problem was formulated as a multi-objective mixed non-linear programming model and solved using the NSGA-II. To compare the performance of the proposed model, we consider another model (Model 2), which represents the lot-sizing problem for CLSC on our network.

Every model is tested for three scales (small, medium, and large) in five cases based on algorithm parameters settings. The results show that considering the lot sizing and network design decisions simultaneously lead to better decisions than considering them separately. Further analysis is done for algorithm performance used to solve the proposed model. This analysis was done for five indicators which are the number of Pareto front solutions (ER) solution closeness to the true Pareto front (GD) diversity with closeness (HAR) solution spread (MS), and distribution (UD). The results show that cases 3 and 4 are the best for the most performance matrices.

Further research is needed to investigate the best heuristic and meta-heuristic algorithm for solving the proposed model. Additionally, the model can be improved to handle multi-echelon selection rather than the current single-echelon selection. The last suggestion is to investigate multiple lot-sizing policies.

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