

Developing Strong and Hybrid Formulation for the Single Stage Single Period Multi Commodity Warehouse Location Problem: Theoretical Framework and Empirical Investigation

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Received 27 March 2015; accepted 20 April 2015; published 21 April 2015

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Abstract

We note that the Single Stage Single Period Multi Commodity Warehouse Location Problem (SSSPMCWLP) has been first attempted by Geoffrion and Graves [1], and that they use the weak formulation (in context of contribution of this paper). We give for the first time “strong” formulation of SSSPMCWLP. We notice advantages of strong formulation over weak formulation in terms of better bounds for yielding efficient Branch and Bound solutions. However, the computation time of “strong” formulation was discovered to be higher than that of the “weak” formulation, which was a major drawback in solving large size problems. To overcome this, we develop the hybrid strong formulation by adding only a few most promising demand and supply side strong constraints to the weak formulation of SSSPMCWLP. So, the formulations developed were put to test on various large size problems. Hybrid formulation is able to give better bound than the weak and takes much less CPU time than the strong formulation. So, a kind of trade off is achieved allowing efficiently solving large sized SSSPMCWLP in real times using hybrid formulation.

Keywords

SSUWLP, SSSPMCWLP, Strong and Weak Constraints, Hybrid Formulation, RMIP

1. Introduction and Literature Survey

Warehousing is a key component of today’s business supply chain. It is not just a place to store finished goods but

How to cite this paper: Sharma, R.R.K., Tyagi, P., Kumar, V. and Jha, A. (2015) Developing Strong and Hybrid Formulation for the Single Stage Single Period Multi Commodity Warehouse Location Problem: Theoretical Framework and Empirical Investigation. *American Journal of Operations Research*, 5, 112-128. <http://dx.doi.org/10.4236/ajor.2015.53010>

involves various value added functions like consolidation, packing and shipment of orders to customers. Organizations produce several commodities at various manufacturing sites, and generally it is not economical to get these commodities directly from plants to markets owing to distant market locations, associated demand disparities and uncertainties. Therefore, warehouses are installed at various locations between plants and markets to consolidate the localized customer demand. There can be more than one stage of the warehouse located between plants and the markets (Multistage Warehouse Location Problem). Among all considerations for securing a warehouse premise, the most important features are the space the warehouse can offer and its location. Generally, the sole objective in warehouse allocation problem is to minimize the total cost of transporting the goods from the manufacturing sites to the end customers via these warehouses.

We develop in this work the strong constraints for the multi-commodity case of Single Stage Single Period Capacitated Warehouse Location Problem (SSSPMCWLP). Here, “Capacitated” means that the capacity of a warehouse at a particular location is fixed and limited. This problem can be compared with a real life problem faced by Food Corporation of India (FCI) [2]. In India, FCI distributes a variety of food grains (wheat, rice, etc.) all over the country on a massive scale, purchasing them from various assigned locations called “MANDIS” directly from the farmers. FCI has a central office and five zones (North, South, East, North-East, and West). Each zone is further divided into regions, and each region is divided into several states and each state has several districts in it. From “MANDIS”, these grains are taken to central warehouses of each region and then transported to district warehouse as per the demand. From district warehouse, food grains are moved to depots, which distribute food grains directly to the public at a fair price. So, the MANDIS acts as plants or sources of food grain and district warehouses as demand points. Regional warehouses serve the intermediate transshipment points. The problem faced by FCI is to choose an optimum number of regional warehouse locations with sufficient capacity so as to minimize the sum of location and transportation costs of distributing food grains from MANDIS to these large Regional warehouses, and subsequently to smaller warehouses or district distribution centres. This results in SSSPMCWLP.

The importance of allocation of warehouses/facility between the production sites and the market can be depicted from the fact that this has been studied in its variant forms by long lists of researchers from last four decades and still remains the crucial decision of today’s supply chain design. Researchers have used several methods/approaches to reach the optimal solution. Warehouse allocation problem can be classified as:

1.1. Uncapacitated (SPLP) or Capacitated (CPLP)

Uncapacitated or Simple Plant Location Problems are those where it is assumed that facility has infinite storage space/handling capacity while the Capacitated Plant Location Problem implies that the capacity of the warehouse is limited and known in advance.

1.2. Single Commodity or Multi-Commodity

Single commodity means that there is only a single type of commodity and multi-commodity problems dealing with more than one type of commodities to be distributed.

1.3. Single Stage or Multi-Stage

Single stage refers to the real life problems where commodities are stored at one intermediate stage between plants and markets, while in the multi-stage problems, food grain distribution system commodities are stored at multiple stages (Figure 1 shows the schematic representation of single stage).

Our problem is multi-commodity single stage single period capacitated warehouse location problem (SSSPMCWLP), whose variants have been attempted by many researchers. An overview of the work done in the field of facility allocation can be found in review work of ReVelle and Eiselt [3]. Geoffrion and Graves [1], Sharma [4] [5], Drezner *et al.* [6] and Kouvelis *et al.* [7] have used weak formulation to address single stage capacitated warehouse location problem (SSCWLP). Keskin and Uster [8] have attempted to solve SSCWLP problem with capacity limit at the manufacturing plant, and not on the warehouse. Geoffrion and Graves [1] and Sharma [4] have attempted the SSUWLP and have given completely different formulations considering two different real life problems. Geoffrion and Graves [1] gave a formulation for multi-commodity capacitated single-period version and successfully applied it to a real problem for a food firm using solution technique based on bender decomposition. Hindi and Basta [9] and Hindi *et al.* [10] solved the similar production-distribution pro-

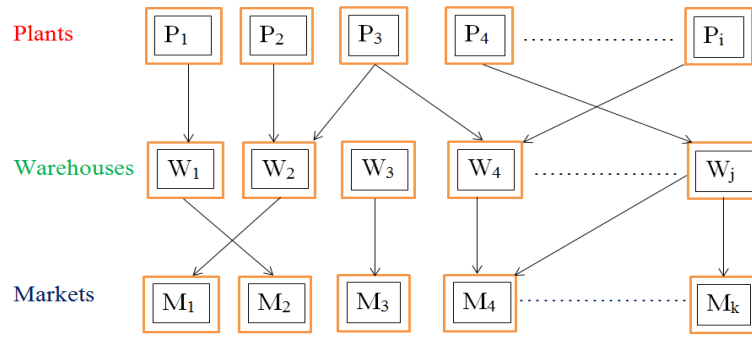


Figure 1. Multi-commodity distribution from plants to markets via warehouses.

blem by a branch-and-bound algorithm. Sharma [4] formulated a fertilizer production-distribution system problem as a mixed zero-one integer linear programming problem (MILP). Sharma and Verma [11] gave strong and weak formulation for SSUWLP considering single commodity; they also gave hybrid formulation to get a better bound than the weak formulation along with lesser CPU time than the strong formulation. Montoya-Torres *et al.* [12] developed three-echelon un-capacitated facility location problem (TUFLP) where they minimized the total cost of warehouse location, production, and distribution of products and proposed a greedy randomized adaptive search procedure (GRASP) to solve the multi-item version of the TUFLP. Elson [13] proposed a different multi-commodity version of the plant-warehouse facility logistics problem, concentrating on a single echelon of transshipment/stocking points. Many researchers have worked on developing both heuristic solution methods and exact algorithms to solve capacitated plant location problem. Pirkul and Jayaraman [14] proposed a heuristic based on the Lagrangian relaxation of the model to solve the two-stage production-distribution system design problem (TSPDSDP) with multi-commodity. Keskin and Üster [8] proposed a scatter search algorithm with hybrid improvements including local search and path-relinking to solve a TSPDSDP in which a fixed number of intermediate facilities points are to be located between the plants and customer locations. Verma and Sharma [15] applied vertical decomposition and Lagrangian relaxation approach to solve SSCWLP. Sharma and Aggarwal [16] solved two-stage capacitated warehouse location problem (TSCWLP) by vertical decomposition method, which was attained by relaxing the associated flow balance constraints.

It may be noted that Sharma [17], Sharma and Murali [18], Sharma and Berry [3] and Sharma and Verma [12] have attempted single commodity problem and have used “Normalized” variables for developing their strong formulation. We attempt for the first time to give strong formulation for multi-commodity version of the location-distribution problem (SSSCPMCWLP) and have not used ‘Normalized’ variables due to obvious difficulty.

The remainder of this paper is organized as follows. Section 2 gives mathematical formulation of the SSSCPMCWLP based on the existing literature and our modifications. Section 3 provides theoretical framework for using strong formulation, followed by empirical justification of hybrid formulation trade off in relaxed bound values versus computation time. The results from empirical investigation are followed by conclusions in Section 4.

2. Mathematical Formulation of SSSCPMCWLP

2.1. Indices Used

- i : Set of the supply points (plants);
- j : Set of the potential warehouse points;
- k : Set of the markets;
- m : Set of the commodities.

2.2. Definition of Constants

- F_j : Fixed cost of establishing and maintaining the warehouse at location “ j ”;
- S_{im} : Supply capacity of plant “ i ” for commodity “ m ”;
- D_{km} : Demand for commodity “ m ” at market “ k ” ;

CAP_j : Capacity of the warehouse at location “ j ” for all commodities (assumed that all the commodity are of same density and occupy same space**);

M : A very large number, here taken as two times the maximum supply or maximum warehouse capacity;

CPW_{ijm} : Cost of shipping one unit from plant “ i ” to warehouse “ j ” of the commodity “ m ”;

CWM_{jkm} : Cost of shipping one unit from warehouse “ j ” to market “ k ” for commodity “ m ”;

** : Even if the volume or density are different for commodities a conversion factor can be introduced to account for them in warehouse capacity constraint (say v_m for commodity “ m ”).

2.3. Definition of Decision Variables

XPW_{ijm} : Number of units shipped from plant “ i ” to warehouse “ j ” of the commodity type “ m ”;

XWM_{jkm} : Number of units shipped from warehouse “ j ” to market “ k ” of commodity type “ m ”;

Y_j : Binary variable which is 1 if it is decided to locate a warehouse at location “ j ” and 0 otherwise;

Z : Total cost of transporting commodities from plants to warehouses, warehouses to demand points or markets and fixed cost of locating the warehouses.

2.4. Mathematical Model

Minimize

$$Z = \sum_{i,j,m} (XPW_{i,j,m} * CPW_{i,j,m}) + \sum_{j,k,m} (XWM_{j,k,m} * CWM_{j,k,m}) + \sum_j Y_j * F_j \quad (1)$$

Subjected to:

$$\sum_j XPW_{i,j,m} \leq S_{i,m}, \quad \forall i, m \quad (\text{Supply Constraints}) \quad (2)$$

$$\sum_i XPW_{i,j,m} = \sum_k XWM_{j,k,m}, \quad \forall j, m, \quad (\text{Flow balance constraint}) \quad (3)$$

$$\sum_j XWM_{j,k,m} = D_{k,m}, \quad \forall k, m, \quad (\text{Demand Constraints}) \quad (4)$$

$$\sum_{i,m} XPM_{i,j,m} \leq CAP_j, \quad \forall j, \quad (\text{Capacity Constraints}) \quad (5)$$

Linking constraints

$$\sum_{i,m} XPW_{i,j,m} \leq M * Y_j, \quad \forall j, \quad (\text{Weak linking constraint}) \quad (6)$$

$$\left. \begin{aligned} XPW_{i,j,m} &\leq S_{i,m} * Y_j, \quad \forall i, j, m; \quad (7) \\ XWM_{j,k,m} &\leq D_{k,m} * Y_j, \quad \forall i, j, m. \quad (8) \end{aligned} \right\} (\text{Strong Linking Constraints})$$

Positive and binary constraints on decision variables:

$$XPW_{i,j,m} \geq 0, \quad \forall i, j, m \quad (9)$$

$$XWM_{j,k,m} \geq 0, \quad \forall j, k, m \quad (10)$$

$$Y_j \in \{0,1\}, \quad \forall j \quad (11)$$

Relaxed binary constraints

$$0 \leq Y_j \leq 1, \quad \forall j \quad (12)$$

Formulation I: Min (1); s.t. (2) - (5); (6); (9) - (11);

Formulation II: Min (1); s.t. (2) - (5); (7) - (8); (9) - (11).

3. Theoretical Framework and Empirical Verification

Theorem 1: LP relaxation of strong formulation (formulation II) gives superior bounds than the LP relaxation of weak formulation (formulation I) of SSSPMCWLP.

Proof: we see the feasible region of LP relaxation of formulation II is much smaller than feasible region of LP relaxation of formulation I, since $M \gg$ all S'_{im} s and D'_{km} s. Tighter feasible region for a minimization problem obviously result in higher objective function value *i.e.* formulation II objective function value will be higher (proved).

Though LP relaxation of weak formulation gives inferior bounds but are computed very fast (as it has linear number of linking constraints). LP relaxation of strong formulation gives superior bounds but it takes more time (as number of linking constraints are of order $(j * m + k * m)$). Therefore computational time of LP relaxation of strong formulation is much higher than the LP relaxation of weak formulation of SSSPMCWLP.

The hybrid formulation is developed to take advantage of both (better bound of strong formulation and better computation time of weak formulation) by adding most promising constraints of the strong formulation to the weak formulation of the SSSPMCWLP. Those few ‘strong’ constraints added may be binding in strong formulation, and can carefully be obtained by sorting the demand and supply values and gradually iterating to get the desired result. We started with 20% of strong constraints then reduced to 5%, by step size of 5% and later reduced with step size of 1%. In each step, we observed decrease in the bounds obtained by LP relaxation but also simultaneous significant improvement in computational time. After some iteration we arrived at desired/satisfactory level and decided that the top 2% of total “strong” demand and supply constraints were to be added to the “weak” formulation of SSSPMCWLP to get the required hybrid formulation. At this level, the bounds were only a little inferior from weak bounds, but took substantially less computational time compared to strong computational time; significantly improving the overall computational times for solving original SSSPMCWLP. This motivated us to develop the following procedure.

Procedure to build hybrid formulation constraints

Step 1: Sort demand points values in increasing order of demand $D_{k,m}$ for each m .

Set cut off number $N = \text{INT}(0.02 * K)$, where K is number of demand points.

Let the demand associated with this number N and given m be: $(2\%D \text{ cut off}_{k,m})$

Prepare a set of most promising demand points (MPDP) = $\{k : D_{k,m} \leq (2\%D \text{ cut off}_{k,m}) \text{ for each } m\}$

Now,

$$\text{MPDP_constr} = \{XWM_{j,k,m} \leq D_{k,m} * Y_j; \forall j, m, k; \text{ where } k \in \text{MPDP}\} \quad (13)$$

Step 2: Sort supply points values in increasing order of demand $S_{i,m}$ for each m .

Set cut off number $N = \text{INT}(0.02 * K)$, where S is the number of supply points.

Let the Supply capacity associated with this number N and given m be: $(2\%S \text{ cut off}_{i,m})$

Prepare a set of most promising supply points (MPSP) = $\{i : S_{i,m} \leq (2\%S \text{ cut off}_{i,m}) \text{ for each } m\}$

Now,

$$\text{MPSP_constr} = \{XPW_{i,j,m} \leq S_{i,m} * Y_j; \forall i, m, j; \text{ where } i \in \text{MPSP}\} \quad (14)$$

Hybrid formulation (HF) and its relaxation (HFR)

HF: Minimize (1); subject to (2) - (6); (9) - (11); (13); (14);

HFR: Minimize (1); subject to (2) - (6); (9); (10); (12); (13); (14).

3.1. Empirical Investigation

GAMS programming is used to show the superiority of formulation (II) over formulation (I) and t-test has been conducted for the statistical significance of the results. To overcome the drawback of higher CPU time for strong formulation (Formulation II), a hybrid formulation is developed by adding the most promising supply and demand side strong linking constraint to the weak formulation. Here we have included 2% of the demand and strong side linking constraints to the weak formulation for which demand and supply are least.

Problem instances of the size $50 \times 50 \times 50$ and $100 \times 100 \times 100$ are solved in GAMS for the following four categories (Table 1) and as per assumptions given below:

1) In each problem instance demand for each commodity in each market is taken as uniformly distributed with lower bound of 100 units and upper bound of 150 units;

Table 1. Problem categories for SSSPMCWLP.

Category	Average warehouse capacity	Average supply capacity
A.	30% more than the average demand	30% more than the average warehouse capacity
B.	30% more than the average demand	125% more than the average warehouse capacity
C.	125% more than the average demand	30% more than the average warehouse capacity
D.	125% more than the average demand	125% more than the average warehouse capacity

2) Total Supply Capacity of each commodity at any plant and Warehouse Capacity at any location is more than any market average demand and taken as per the above four categories in **Table 1**;

3) Warehouse Capacity and Supply Capacity of plants are also uniformly distributed;

4) Cost of carrying one unit from plant to warehouse is normally distributed with mean of 4000 and standard deviation of 300;

5) Cost of carrying one unit from warehouse to market is normally distributed with mean of 5000 and standard deviation of 400;

6) Fixed Cost of locating a warehouse is normally distributed with a mean of 10,000 and standard deviation of 1500.

Samples problems are solved by relaxing the 0 - 1 constraints for the location variable Y_j (constraint (12) in place of constraint (11) in formulation I and formulation II).

GAMS code is written to generate the random values for supply capacity, warehouse capacity, demand, costs of transportation and fixed cost of warehouse location as per the above specified criteria. The same code solves both the formulation (I) and formulation (II) with the same input data. Same data is used to solve hybrid formulation, so that we can compare the objective function values and CPU time values for these three formulations. 30 Problems are solved for each of the four categories for each problem size (total number of problem instances solved = 240, 120 of each of the size $50 \times 50 \times 50$ and $100 \times 100 \times 100$) using a Intel (R) Core (TM) i7-4770S CPU @ 3.10GHz.

Objective function values and CPU time for each and every problem instance is recorded and statistical analysis (t-test) is done to check the superiority of formulation (II) over formulation (I) and superiority of hybrid formulation over other formulations with a confidence level of 99.5% (**Tables 2-13**).

3.1.1. Statistical Analysis

Hypothesis tests are conducted as follows:

I. To check improvement in objective function values of LP relaxed Strong Bound (formulation II) from LP relaxed Weak Formulation (formulation I)

μ_1 : Percentage increase in objective function value of LP relaxed formulation (II) over LP relaxed formulation (I).

Null hypothesis, $H_0 : \mu_1 = 0$;

Alternate hypothesis, $H_a : \mu_1 > 0$.

From the statistical t-tables we have the critical value for t-stats at $\alpha = 0.005$ as 2.68 for d.o.f. = 49 and 2.6264 for $\alpha = 0.005$ and d.o.f. = 99, thus we can easily reject null hypothesis $\mu_1 = 0$.

II. To check improvement in objective function values of LP relaxed Hybrid formulation vs. LP relaxed Weak Formulation (formulation I).

μ_2 : Percentage increase in objective function value of LP relaxed Hybrid formulation over LP relaxed Weak formulation (formulation I).

Null hypothesis, $H_0 : \mu_2 = 0$;

Alternate hypothesis, $H_a : \mu_2 > 0$.

Result: Comparing the calculated t-values with critical t-values, the null hypothesis $\mu_2 = 0$ is rejected.

III. To check deterioration in CPU time values of LP relaxed Strong bound (formulation (II)) vs. LP relaxed Weak bound (formulation (I)).

t_1 : Percentage deterioration in CPU time for Strong formulation (II) over weak formulation (I).

Null hypothesis, $H_0 : t_1 = 0$;

Alternate hypothesis, $H_a : t_1 > 0$.

Table 2. Problem size $50 \times 50 \times 50$ of Category A.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	464761657.8	0.172	464,856,004	0.193	465252234.1	0.203
P2	465982540.6	0.141	466,077,974	0.160	466,479,602	0.203
P3	463307328.6	0.141	463,406,152	0.153	463794110.6	0.203
P4	463827848.1	0.14	463,921,356	0.159	464328797.5	0.187
P5	462125994.6	0.141	462,227,246	0.156	462614855.4	0.203
P6	464303782.3	0.141	464,405,186	0.158	464812299.2	0.203
P7	466278050.8	0.140	466,375,689	0.158	466787603.6	0.203
P8	463635528.2	0.141	463,730,203	0.152	464,121,359	0.187
P9	463313848.7	0.141	463,413,322	0.156	463809058.1	0.187
P10	464,843,759	0.109	464,937,239	0.120	465335659.3	0.187
P11	464899247.3	0.110	464,993,389	0.119	465404355.6	0.203
P12	463364638.6	0.109	463,457,636	0.122	463851905.1	0.203
P13	464718283.4	0.125	464,815,642	0.135	465205666.3	0.203
P14	461971206.5	0.125	462,068,220	0.142	462461413.5	0.203
P15	466460412.4	0.109	466,557,763	0.124	466968490.2	0.187
P16	466,121,646	0.125	466,215,197	0.140	466623984.1	0.203
P17	466,613,596	0.109	466,708,832	0.118	467117668.5	0.187
P18	457,455,118	0.125	457,547,981	0.141	457,942,972	0.203
P19	465677481.6	0.125	465,778,580	0.141	466196407.9	0.203
P20	464276924.6	0.125	464,370,012	0.142	464,767,951	0.203
P21	462941205.5	0.125	463,035,738	0.140	463438858.3	0.203
P22	461101747.4	0.125	461,195,628	0.136	461599000.9	0.203
P23	465161010.4	0.125	465,258,415	0.135	465650542.7	0.203
P24	461900311.9	0.125	461,998,743	0.135	462396333.7	0.203
P25	463755031.4	0.125	463,856,362	0.141	464258661.5	0.203
P26	459684460.9	0.125	459,780,121	0.134	460163496.2	0.235
P27	466716094.5	0.110	466,811,865	0.125	467213751.9	0.219
P28	465410474.9	0.109	465,503,929	0.123	465913015.5	0.203
P29	465,546,559	0.125	465,645,488	0.141	466045992.9	0.219
P30	459281474.2	0.141	459,377,556	0.156	459785502.4	0.203

Table 3. Problem size $50 \times 50 \times 50$ of Category B.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	462154595.8	0.125	462,252,526	0.168	462647776.5	0.203
P2	463922755.7	0.125	464,021,386	0.172	464424535.6	0.219
P3	454902314.7	0.094	454,995,979	0.130	455391806.2	0.234
P4	462729689.5	0.11	462,824,549	0.156	463237509.7	0.203
P5	463264246.2	0.109	463,357,548	0.159	463756427.2	0.203
P6	459871109.2	0.109	459,967,544	0.149	460359595.1	0.203
P7	463005259.3	0.109	463,102,583	0.150	463499775.1	0.203
P8	461,750,540	0.11	461,846,630	0.154	462250609.3	0.203
P9	462513368.3	0.109	462,607,027	0.142	463008757.8	0.203
P10	464652116.8	0.109	464,746,116	0.157	465,123,264	0.203
P11	465,808,805	0.125	465,907,370	0.174	466,306,867	0.25
P12	466659753.5	0.14	466,754,205	0.181	467165643.8	0.25
P13	465859563.6	0.141	465,953,947	0.177	466,367,333	0.234
P14	460683422.9	0.125	460,784,405	0.158	461175212.4	0.234
P15	459791614.3	0.125	459,888,906	0.174	460282546.8	0.188
P16	460724495.9	0.109	460,819,912	0.155	461231699.5	0.172
P17	461385093.5	0.125	461,480,185	0.158	461887513.4	0.203
P18	463032455.3	0.125	463,126,358	0.173	463528485.4	0.188
P19	461040271.5	0.109	461,141,147	0.152	461544498.2	0.25
P20	461,943,881	0.093	462,038,256	0.134	462455999.9	0.187
P21	461975515.3	0.11	462,077,104	0.139	462458306.3	0.188
P22	459544615.7	0.109	459,645,210	0.152	460,037,834	0.187
P23	461886169.8	0.11	461,983,905	0.146	462402129.2	0.203
P24	459427161.6	0.109	459,522,677	0.147	459,901,863	0.188
P25	465536812.4	0.125	465,631,596	0.161	466,037,225	0.234
P26	461721331.5	0.125	461,819,032	0.173	462220540.5	0.25
P27	460304848.7	0.11	460,396,956	0.148	460808541.2	0.234
P28	458865037.7	0.125	458,963,051	0.167	459365142.4	0.218
P29	461224194.5	0.109	461,321,005	0.153	461736086.6	0.203
P30	462191950.1	0.109	462,287,670	0.149	462677404.9	0.219

Table 4. Problem size $50 \times 50 \times 50$ of Category C.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	460156810.2	0.125	460,249,210	0.157	460647316.2	0.25
P2	461928500.6	0.109	462,025,737	0.154	462422695.2	0.187
P3	460184521.6	0.094	460,277,847	0.118	460670115.5	0.188
P4	458523854.3	0.109	458,623,812	0.154	459015839.3	0.188
P5	461210372.4	0.109	461,310,363	0.146	461695917.4	0.172
P6	466736925.2	0.109	466,839,421	0.134	467227158.5	0.188
P7	465040652.4	0.109	465,135,474	0.160	465533649.4	0.188
P8	465,711,634	0.11	465,806,825	0.132	466203952.5	0.188
P9	462,663,898	0.093	462,758,652	0.118	463143978.5	0.171
P10	461994675.3	0.11	462,089,430	0.155	462501565.4	0.188
P11	463137094.7	0.11	463,237,318	0.158	463654765.4	0.187
P12	460644100.8	0.11	460,736,460	0.149	461147401.4	0.187
P13	459979852.8	0.109	460,080,404	0.144	460480501.1	0.172
P14	465645551.9	0.125	465,743,431	0.180	466,147,369	0.171
P15	459540857.4	0.109	459,641,773	0.143	460034510.6	0.172
P16	457524145.7	0.094	457,617,435	0.116	458003132.4	0.187
P17	463560066.4	0.094	463,658,619	0.129	464066716.1	0.172
P18	461800417.3	0.125	461,896,010	0.169	462290440.1	0.187
P19	461310140.9	0.094	461,402,864	0.132	461809878.3	0.172
P20	463826134.1	0.109	463,921,172	0.137	464324355.3	0.188
P21	461729092.8	0.125	461,826,795	0.158	462207481.3	0.172
P22	459986245.2	0.109	460,084,636	0.148	460465903.5	0.187
P23	466476622.5	0.094	466,576,728	0.126	466958243.4	0.219
P24	461750643.4	0.094	461,844,887	0.114	462241797.1	0.172
P25	461884416.2	0.11	461,978,687	0.149	462377121.9	0.172
P26	462809272.1	0.11	462,905,999	0.162	463300162.2	0.188
P27	462,954,846	0.109	463,050,076	0.140	463446331.4	0.172
P28	462717628.6	0.11	462,814,105	0.157	463207321.1	0.172
P29	461110049.8	0.109	461,211,171	0.160	461601038.8	0.187
P30	461842013.8	0.125	461,938,493	0.170	462355526.8	0.187

Table 5. Problem size $50 \times 50 \times 50$ of Category D.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	465681644.6	0.109	465,779,158	0.125	466164697.7	0.188
P2	457544301.9	0.109	457,638,693	0.128	458054151.7	0.203
P3	459974575.7	0.093	460,067,583	0.106	460478771.5	0.188
P4	458538365.5	0.094	458,633,100	0.116	459024202.6	0.187
P5	460692904.2	0.094	460,793,612	0.111	461176844.4	0.203
P6	459,533,942	0.094	459,626,170	0.124	460005875.9	0.203
P7	460283567.4	0.109	460,376,361	0.141	460778713.3	0.187
P8	461,788,231	0.109	461,882,251	0.131	462277760.5	0.188
P9	456723909.6	0.125	456,816,899	0.150	457209892.4	0.172
P10	459648712.5	0.125	459,744,963	0.158	460163566.7	0.172
P11	456480416.3	0.11	456,579,336	0.127	456977922.9	0.188
P12	465421105.4	0.109	465,520,659	0.127	465917396.2	0.172
P13	463411618.9	0.109	463,506,016	0.138	463903987.7	0.188
P14	462087761.9	0.125	462,188,543	0.160	462585844.7	0.172
P15	464393272.7	0.109	464,494,557	0.135	464873431.8	0.187
P16	459898122.6	0.125	459,995,391	0.147	460381579.4	0.172
P17	460720255.7	0.109	460,814,657	0.145	461222610.1	0.187
P18	460516332.1	0.125	460,613,547	0.153	461019806.6	0.188
P19	460753982.8	0.109	460,850,649	0.131	461254252.7	0.187
P20	459873402.2	0.109	459,966,067	0.144	460379360.6	0.172
P21	461661335.6	0.125	461,757,407	0.149	462144676.9	0.218
P22	464025496.2	0.109	464,123,545	0.124	464,520,012	0.172
P23	456996616.6	0.141	457,093,226	0.184	457461190.2	0.172
P24	463178689.2	0.125	463,280,311	0.156	463674506.7	0.172
P25	460206278.2	0.11	460,301,633	0.147	460708601.5	0.188
P26	462363155.8	0.11	462,459,743	0.130	462850313.3	0.172
P27	464172352.1	0.11	464,268,389	0.137	464682393.6	0.203
P28	457850163.7	0.11	457,942,192	0.135	458357345.2	0.188
P29	462003319.6	0.125	462,098,030	0.148	462497026.1	0.172
P30	460635912.7	0.11	460,733,936	0.133	461120649.7	0.187

Table 6. Problem size $100 \times 100 \times 100$ of Category A.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	1,365,932,733	0.969	1,366,215,071	1.085	1,366,941,579	1.562
P2	1,356,964,622	1.062	1,357,249,992	1.144	1,357,960,054	1.572
P3	1,366,670,095	1.093	1,366,946,572	1.248	1,367,675,025	1.65
P4	1,364,344,024	1.047	1,364,630,536	1.145	1,365,354,097	1.578
P5	1,359,252,277	1.109	1,359,550,905	1.265	1,360,243,625	1.578
P6	1,362,703,027	1.031	1,362,989,604	1.137	1,363,687,229	1.547
P7	1,357,648,342	1.094	1,357,931,004	1.248	1,358,646,263	1.532
P8	1,361,734,440	1.078	1,362,015,230	1.234	1,362,771,012	1.578
P9	1,363,335,394	1.094	1,363,626,057	1.226	1,364,326,431	1.578
P10	1,354,662,978	1.062	1,354,940,548	1.126	1,355,672,184	1.625
P11	1,359,010,784	1.031	1,359,297,671	1.111	1,360,015,221	1.578
P12	1,360,290,820	1.031	1,360,566,551	1.118	1,361,275,760	1.484
P13	1,359,813,107	1.015	1,360,106,147	1.136	1,360,804,455	1.516
P14	1,360,229,705	1.094	1,360,524,059	1.190	1,361,194,499	1.563
P15	1,357,065,375	1	1,357,352,530	1.080	1,358,068,651	1.485
P16	1,366,116,379	1.063	1,366,414,875	1.185	1,367,101,877	1.797
P17	1,354,295,387	1.016	1,354,575,049	1.166	1,355,307,736	1.547
P18	1,362,453,698	1.093	1,362,753,165	1.196	1,363,466,065	1.656
P19	1,351,212,750	1.047	1,351,506,098	1.133	1,352,228,221	1.594
P20	1,354,546,726	1.062	1,354,825,085	1.156	1,355,567,185	1.593
P21	1,361,671,542	1.078	1,361,968,932	1.167	1,362,652,718	1.515
P22	1,365,090,355	1.031	1,365,366,240	1.125	1,366,081,015	1.531
P23	1,353,711,255	1.015	1,353,984,975	1.125	1,354,692,174	1.594
P24	1,359,167,606	1.078	1,359,454,798	1.239	1,360,142,833	1.547
P25	1,354,759,090	1.078	1,355,052,259	1.185	1,355,756,040	1.546
P26	1,360,824,145	0.984	1,361,122,302	1.101	1,361,829,294	1.594
P27	1,357,519,457	1.047	1,357,817,161	1.210	1,358,490,606	1.516
P28	1,355,437,009	1.047	1,355,724,226	1.148	1,356,416,967	1.562
P29	1,360,832,833	0.985	1,361,119,561	1.126	1,361,815,955	1.578
P30	1,349,511,932	1.062	1,349,797,353	1.191	1,350,507,838	1.578

Table 7. Problem size $100 \times 100 \times 100$ of Category B.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	1,347,812,101	0.984	1,348,043,655	1.099	1,348,824,104	1.437
P2	1,353,943,093	0.875	1,354,165,275	1.037	1,354,916,518	1.5
P3	1,353,589,922	0.797	1,353,818,814	0.918	1,354,572,437	1.406
P4	1,348,651,591	0.875	1,348,874,253	0.988	1,349,651,010	1.438
P5	1,353,236,978	0.859	1,353,454,443	0.970	1,354,223,675	1.422
P6	1,345,291,759	0.859	1,345,512,522	0.954	1,346,286,225	1.5
P7	1,354,089,798	0.875	1,354,306,453	0.976	1,355,101,717	1.578
P8	1,350,526,131	0.89	1,350,746,131	1.032	1,351,547,781	1.562
P9	1,347,779,924	0.844	1,348,003,925	0.939	1,348,783,579	1.563
P10	1,352,052,432	0.968	1,352,282,416	1.141	1,353,075,218	1.547
P11	1,348,474,977	0.875	1,348,704,892	0.998	1,349,452,863	1.531
P12	1,358,050,071	0.859	1,358,285,014	1.010	1,358,999,314	1.61
P13	1,360,492,873	0.875	1,360,727,422	0.989	1,361,471,188	1.563
P14	1,347,055,192	0.86	1,347,287,155	0.949	1,348,053,275	1.578
P15	1,346,813,104	0.86	1,347,032,769	0.957	1,347,825,533	1.409
P16	1,351,598,681	0.859	1,351,818,451	0.990	1,352,597,768	1.556
P17	1,352,634,141	0.844	1,352,873,693	0.938	1,353,614,996	1.438
P18	1,347,143,013	0.828	1,347,368,659	0.928	1,348,101,467	1.422
P19	1,347,578,203	0.844	1,347,811,739	0.996	1,348,549,057	1.375
P20	1,354,172,979	0.86	1,354,405,219	1.024	1,355,188,299	1.407
P21	1,353,131,970	0.844	1,353,350,636	0.974	1,354,106,286	1.407
P22	1,352,195,677	0.875	1,352,423,658	1.000	1,353,198,611	1.437
P23	1,354,067,995	0.875	1,354,309,967	1.036	1,355,054,115	1.532
P24	1,348,354,628	0.828	1,348,587,758	0.950	1,349,338,895	1.478
P25	1,348,863,378	0.86	1,349,093,359	1.030	1,349,868,041	1.578
P26	1,350,507,863	0.984	1,350,725,025	1.156	1,351,509,090	1.516
P27	1,351,696,314	1	1,351,938,538	1.191	1,352,686,083	1.453
P28	1,348,216,115	1	1,348,458,524	1.139	1,349,190,279	1.469
P29	1,349,155,465	0.969	1,349,378,346	1.123	1,350,167,144	1.406
P30	1,355,404,108	0.968	1,355,636,560	1.078	1,356,383,297	1.422

Table 8. Problem size $100 \times 100 \times 100$ of Category C.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	1,356,765,481	1	1,356,988,805	1.028	1,357,747,787	1.375
P2	1,348,578,687	0.906	1,348,784,480	0.934	1,349,566,313	1.375
P3	1,350,776,977	0.89	1,351,005,393	0.910	1,351,787,791	1.391
P4	1,356,692,279	0.89	1,356,918,304	0.913	1,357,678,393	1.337
P5	1,359,305,911	0.953	1,359,517,283	0.975	1,360,284,488	1.39
P6	1,349,187,512	0.922	1,349,390,565	0.951	1,350,173,750	1.359
P7	1,347,497,209	0.953	1,347,721,702	0.979	1,348,495,827	1.406
P8	1,351,561,595	0.922	1,351,780,684	0.943	1,352,539,359	1.39
P9	1,352,908,377	0.906	1,353,135,936	0.935	1,353,917,629	1.375
P10	1,355,180,785	0.906	1,355,411,030	0.928	1,356,176,036	1.359
P11	1,349,471,170	0.968	1,349,679,258	0.990	1,350,467,442	1.422
P12	1,347,863,661	0.922	1,348,079,993	0.948	1,348,867,681	1.422
P13	1,354,473,571	0.922	1,354,688,119	0.947	1,355,463,506	1.359
P14	1,350,205,726	0.922	1,350,433,235	0.948	1,351,193,519	1.344
P15	1,349,060,862	0.922	1,349,290,067	0.945	1,350,041,009	1.375
P16	1,347,587,320	0.859	1,347,789,863	0.886	1,348,577,355	1.453
P17	1,351,522,612	0.813	1,351,733,179	0.839	1,352,521,657	1.359
P18	1,348,302,161	0.828	1,348,510,204	0.847	1,349,326,170	1.422
P19	1,357,334,766	0.828	1,357,545,560	0.852	1,358,323,179	1.406
P20	1,346,200,443	0.844	1,346,428,489	0.867	1,347,193,192	1.485
P21	1,356,509,204	0.828	1,356,734,656	0.852	1,357,494,571	1.391
P22	1,350,745,322	0.829	1,350,970,356	0.852	1,351,735,380	1.438
P23	1,356,703,210	0.812	1,356,919,469	0.832	1,357,678,909	1.422
P24	1,357,978,256	0.844	1,358,191,187	0.864	1,358,962,351	1.5
P25	1,349,813,482	0.828	1,350,026,212	0.848	1,350,818,186	1.422
P26	1,354,216,441	0.937	1,354,434,470	0.963	1,355,225,303	1.516
P27	1,354,135,174	0.844	1,354,343,981	0.871	1,355,144,688	1.344
P28	1,345,665,627	0.844	1,345,879,453	0.869	1,346,652,596	1.407
P29	1,353,034,625	0.859	1,353,261,123	0.883	1,354,003,682	1.375
P30	1,352,589,054	0.844	1,352,793,971	0.869	1,353,593,395	1.343

Table 9. Problem size $100 \times 100 \times 100$ of Category D.

Problem instance number	Weak bound (relaxed formulation I)		Hybrid formulation bound (relaxed weak formulation with most promising strong linking constraints added)		Strong bound (relaxed formulation II)	
	Value	CPU time	Value	CPU time	Value	CPU time
P1	1,349,044,304	0.953	1,349,310,336	1.076	1,350,052,277	1.328
P2	1,341,330,308	0.891	1,341,585,965	1.079	1,342,364,193	1.344
P3	1,357,367,491	0.875	1,357,634,078	1.040	1,358,357,501	1.313
P4	1,344,761,701	0.907	1,345,022,451	1.072	1,345,749,092	1.312
P5	1,353,950,432	0.906	1,354,231,377	1.065	1,354,948,263	1.343
P6	1,345,725,674	0.86	1,345,986,206	1.060	1,346,727,389	1.328
P7	1,350,064,031	0.875	1,350,340,929	0.991	1,351,070,255	1.344
P8	1,349,599,349	0.859	1,349,856,043	1.048	1,350,587,142	1.312
P9	1,347,042,608	0.875	1,347,306,898	1.071	1,348,010,275	1.328
P10	1,359,202,269	0.891	1,359,485,934	1.034	1,360,184,283	1.313
P11	1,347,843,751	0.875	1,348,111,163	1.080	1,348,840,103	1.328
P12	1,347,079,789	0.859	1,347,339,910	0.996	1,348,069,301	1.344
P13	1,345,959,280	0.875	1,346,232,645	1.005	1,346,948,448	1.36
P14	1,352,533,674	0.844	1,352,817,571	0.995	1,353,537,564	1.359
P15	1,344,792,939	0.891	1,345,071,043	1.021	1,345,811,206	1.312
P16	1,346,897,235	0.828	1,347,173,079	1.006	1,347,857,811	1.344
P17	1,335,591,579	0.765	1,335,869,115	0.905	1,336,633,363	1.344
P18	1,352,021,379	0.813	1,352,281,103	0.976	1,353,033,815	1.344
P19	1,340,538,436	0.828	1,340,813,112	0.994	1,341,534,569	1.375
P20	1,358,293,736	0.796	1,358,553,985	0.893	1,359,285,053	1.329
P21	1,349,182,387	0.828	1,349,458,565	0.983	1,350,197,272	1.343
P22	1,350,465,764	0.844	1,350,736,127	0.978	1,351,464,235	1.328
P23	1,342,712,142	0.828	1,342,975,448	0.978	1,343,717,026	1.344
P24	1,349,696,083	0.812	1,349,955,090	0.925	1,350,685,192	1.312
P25	1,344,666,743	0.813	1,344,942,265	0.967	1,345,654,739	1.281
P26	1,349,872,300	0.812	1,350,155,773	0.941	1,350,889,585	1.328
P27	1,346,328,212	0.797	1,346,606,633	0.951	1,347,349,390	1.344
P28	1,349,420,910	0.812	1,349,697,811	0.919	1,350,411,598	1.359
P29	1,350,343,978	0.828	1,350,602,029	1.010	1,351,345,376	1.313
P30	1,358,015,527	0.828	1,358,295,142	0.945	1,359,024,827	1.313

Table 10. T-test results of objective function improvement of LP relaxed strong formulation over LP relaxed weak formulation.

Problem category	Problem size (50 × 50 × 50)	Problem size (100 × 100 × 100)
	t-values	t-values
A.	338.16	331.84
B.	264.68	281.77
C.	278.69	135.03
D.	225	289.06

Table 11. T-test results of objective function improvement of LP relaxed hybrid formulation over LP relaxed weak formulation.

Problem category	Problem size (50 × 50 × 50)	Problem size (100 × 100 × 100)
	t-values	t-values
A.	192.01	201.35
B.	202.02	161.58
C.	215.59	247.58
D.	203.49	202.31

Table 12. T-test results for CPU time deterioration of strong formulation (II) over weak formulation (I).

Problem Category	Problem size (50 × 50 × 50)	Problem size (100 × 100 × 100)
	t-values	t-values
A.	17.11	40.46
B.	21.74	29.08
C.	19.85	28.51
D.	15.55	38.59

Table 13. T-test results for CPU time improvement of hybrid formulation over strong formulation.

Problem category	Problem size (50 × 50 × 50)	Problem size (100 × 100 × 100)
	t-values	t-values
A.	17.76	37.72
B.	16.69	29.2
C.	10.45	41.63
D.	11.46	31.07

Result: Comparing the calculated t-values with critical t-values, the null hypothesis $t_1 = 0$ is rejected.

IV. To check improvements in CPU time values of Hybrid Formulation vs. Strong Formulation (Formulation I).

As we have added only 2% most promising demand and supply side strong linking constraint to the weak formulation to form hybrid formulation, therefore we expect the improvement in CPU time *i.e.* CPU time for Hybrid Formulation must be less than the CPU time for Strong Formulation. We conduct the following hypothesis test:

t_2 : Percentage improvement in CPU time i.e. decrement in the CPU time for hybrid formulation from Strong formulation.

Null hypothesis, $H_0 : t_2 = 0$;

Alternate hypothesis, $H_a : t_2 > 0$.

Result: Comparing the calculated t-values with critical t-values, the null hypothesis $t_2 = 0$ is rejected.

Looking at the t-tests done on the empirical results, we see all the null hypotheses have been rejected, implying differences in relaxed formulations results; strong constraints gave the superior bounds than the weak constraints. However, the CPU time for computation for LP relaxed strong formulations were significantly higher than the CPU time for LP relaxed weak formulations, which meant that the benefit of better bounds using stronger formulations is somewhat compromised by their higher computational time. Hybrid formulation values were intermediate between strong and weak formulations with significant improvement in bound values.

4. Conclusions

The SSSPMCWLP is often encountered in distribution networks which include cross-docks with known manufacturer and customer demand points. We develop strong and hybrid formulation for the SSSPMCWLP for the first time and show its merits over prevalent weak formulation in literature. The idea behind the hybrid formulation is to get advantages of both weak and strong formulation in terms of bound and computational time.

Computational study is done for a variety of problems of different sizes. We find that hybrid formulations of SSSPMCWLP is a good compromise, as it gives better bounds while not incurring greater computation burden.

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