

Location Inventory Routing Problem

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Abstract

The abstract here.

1 Introduction

Location Inventory Routing Problem (LIRP) aims at integrating in a unified model the three level of decisions involved in the optimization of the supply chain. The strategical aspect relates to the design of the network, by positioning and opening distribution centers on a geographical area. At the tactical, production and inventory management decisions are made while the operational part of the problem focuses on logistics and transportation solutions.

1.1 Literature review

Fill the .bib file and reference the papers, explain what the authors are doing.

2 The model

2.1 Description and notations

In this work, we present a unified model that covers a broad spectrum of systems in which one supplier (or central plant) produces semi-finished or end products. There is a set of candidates depots, that is location onto which *distribution centers* (DC), or *depots* may be opened to operate as intermediate between the central supplier and a (often larger) set of retailers, also called customers. The decision to open or not a DC incurs a fixed opening cost for each depot at the beginning of the planning horizon and cannot be changed during the exploitation of the network. Units stored at a DC or at a customer incur a holding cost that represents the financial immobilisation of the good as well as maintenance costs. Finally deliveries are made through predefined routes through the network. A route starts either from the supplier and visits a set of depots before returning to the supplier or starts from a depot and visits a set of customers before returning to the same depot. Using a route to deliver units during a time period also incurs a transportation cost that consists of a (possibly zero) fixed ordering cost along with a routing cost that depends on the length of the route and its number of stops.

Throughout the paper, we use the following notations:

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Set	Definition
T	Length of the planning horizon
L	Number of levels in the distribution network
\mathcal{L}_k	Set of distribution centers j of level $k = 1, \dots, L$
$\mathcal{L} = \cup_{k=1}^L \mathcal{L}_k$	Set of all locations in the distribution network
$\mathcal{C} = \mathcal{L}_L$	Set of customers
\mathcal{R}_k	Set of routes starting from a DC at level $k - 1$ and visiting and a subset of locations at level k , $k = 1, \dots, L$
$\mathcal{R} = \cup_{k=1}^L \mathcal{R}_k$	Set of all the routes in the distribution network
Data	Definition
f_j	Fixed cost of opening distribution center $j \in \mathcal{L}_k$, $k = 1, \dots, L - 1$
κ_k	Capacity of vehicles delivering locations at level k
ν_k	fleet size for vehicles delivering locations of level k
d_{it}	Demand faced by customer $i \in \mathcal{C}$ in period $t = 1, \dots, T$
h_{it}	Per-unit, per-period holding cost of location $i \in \mathcal{L}$ in period $t = 1, \dots, T$
I_{i0}	Initial inventory of location $i \in \mathcal{L}$ at the beginning of the planning horizon
c_r	Cost of route $r \in \mathcal{R}$
α_{ir}	indicator that route r visits location i
β_{jr}	indicator that route r starts from DC j
I_i^{\max}	Maximum inventory at location $i \in \mathcal{L}$
Variables	
<i>Binary Variables</i>	
y_j	$\rightarrow 1$ if distribution center j is selected
z_{rt}	$\rightarrow 1$ if route r from level is selected in period t
<i>Continuous variables</i>	
q_{irt}	\rightarrow quantity delivered by route r to location i in period $t = 1, \dots, T$
I_{it}	\rightarrow Inventory at location i in period $t = 1, \dots, T$

2.2 MIP formulation

$$\begin{aligned}
& \text{minimize} && \sum_{j \in \mathcal{L} \setminus \mathcal{C}} f_j y_j + \sum_{t=1}^T \left(\sum_{r \in \mathcal{R}} c_r z_{rt} + \sum_{i \in \mathcal{L}} h_{it} I_{it} \right) && (1) \\
& \text{s.t.} && \sum_{r \in \mathcal{R}_L} \alpha_{ir} z_{rt} \leq 1 && \forall i \in \mathcal{C}, \forall t = 1, \dots, T && (2) \\
& && \sum_{r \in \mathcal{R}_k} \alpha_{jr} z_{rt} \leq y_j && \forall k = 1, \dots, L-1, \forall j \in \mathcal{L}_k, \forall t = 1, \dots, T && (3) \\
& && z_{rt} \leq \sum_{j \in \mathcal{L}_k} \beta_{jr} y_j && \forall k = 1, \dots, L-1, r \in \mathcal{R}_{k+1}, \forall t = 1, \dots, T && (4) \\
& && \sum_{r \in \mathcal{R}_k} z_{rt} \leq \nu_k && \forall k = 1, \dots, L, \forall t = 1, \dots, T && (5) \\
& && \sum_{i \in \mathcal{L}_k} q_{irt} \leq \kappa_k z_{rt} && \forall k = 1, \dots, L, \forall r \in \mathcal{R}_k, t = 1, \dots, T && (6) \\
& && I_{j,t-1} + \sum_{r \in \mathcal{R}_k} q_{jrt} = I_{jt} + \sum_{r' \in \mathcal{R}_{k+1}} \left(\beta_{jr'} \sum_{i \in \mathcal{L}_{k+1}} q_{ir't} \right) && \forall k = 1, \dots, L-1, \forall j \in \mathcal{L}_k, \forall t = 1, \dots, T && (7) \\
& && I_{i,t-1} + \sum_{r \in \mathcal{R}_L} q_{irt} = I_{it} + d_{it} && \forall i \in \mathcal{C}, \forall t = 1, \dots, T && (8) \\
& && I_{jt} \leq I_j^{\max} y_j && \forall k = 1, \dots, L-1, \forall j \in \mathcal{L}_k, \forall t = 1, \dots, T && (9) \\
& && I_{it} \leq \min \left(I_i^{\max}, \sum_{t' \geq t} d_{it'} \right) && \forall i \in \mathcal{C}, \forall t = 1, \dots, T && (10) \\
& && q_{irt} \geq 0 && \forall k = 1, \dots, L, \forall i \in \mathcal{L}_k, \forall r \in \mathcal{R}_k, \forall t = 1, \dots, T && (11) \\
& && I_{it} \geq 0 && \forall i \in \mathcal{L}, \forall t = 1, \dots, T && (12) \\
& && y_j \in \{0, 1\} && \forall j \in \mathcal{L} \setminus \mathcal{C} && (13) \\
& && z_{rt} \in \{0, 1\} && \forall r \in \mathcal{R}, \forall t = 1, \dots, T && (14)
\end{aligned}$$

The objective function (1) aims at minimizing the total cost incurred by the system. Constraints (2) and (3) state that every (open) location is served by at most one route $r \in \mathcal{R}$ serving the corresponding level in every period, respectively. Constraint (4) ensures that routes start only from opened depots. Constraints (5) states that the number of routes used to serve the locations of a given level k in period t does not exceed the fleet size allocated to the routes $r \in \mathcal{R}_k$. Constraint (6) ensures that the sum of the quantities delivered through a route $r \in \mathcal{R}_k$ in period t is lower than the capacity of a vehicle. Constraints (7) and (8) define the units flow through the depots and customers, respectively. Finally, constraints (9) and (10) ensure that the capacity constraint on the inventory at the depots and the customers are satisfied in any period.

2.3 Restrictions

Although the MIP formulation from § 2.2 deals with a rather general version of the problem, we focus on two particular cases:

1. The first one considers a model “Direct+Loop” in which the set \mathcal{R}_k consists only of direct routes between the supplier and the depots. That is, a route $r \in \mathcal{R}_k$ can be described by two arcs (s, j) and (j, s) , where $j \in \mathcal{R}_k$ and s is the supplier.
2. The second one, called “Loop+Direct” considers a set \mathcal{R}_L containing only direct routes between depots and customers. In other words, a route $r \in \mathcal{R}_L$ can be described by two arcs (j, i) and (i, j) with $j \in \mathcal{L}_k$ and $i \in \mathcal{C}$. Figure 1b.

Figure 1a and 1b illustrates the two types of structures related to these models.

3 Matheuristic method

The matheuristic method we propose for the LIRP is based on a sampling of the routes. After being shuffled, the original set of routes is splitted into α independent subsets. Each subset is solved with Cplex, leading to α

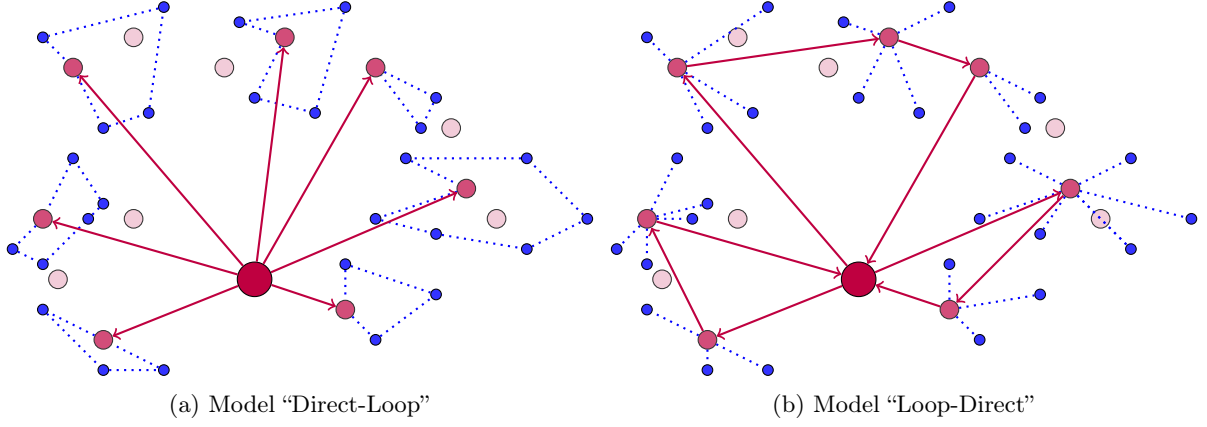


Figure 1: The two types of routing structures considered

suboptimal LIRP solutions. The routes contained in each suboptimal solution are stored in a pool of *promising routes*. Finally, the pool of promising routes is used to create a new (hopefully tractable) instance which solved with Cplex. Algorithm 1 describes the procedure.

3.1 Location based sampling

Location Sampling s’effectue en deux temps. Tout d’abord, les clients sont pré-affectés à un certain nombre de dépôts, comme décrit dans l’algo 2. Ensuite, cette pré-affectation est utilisée pour constituer des sous-ensembles de routes, comme décrit dans l’algo 3

Lines 5–9 of Algorithm 2, calculate distances between each client and all depots. For each client, these distances are ranked in non-decreasing order (line 12). Then, all clients are pre-allocated to their closest depot (line 13) and to the N_{close} closest depots provided the distance between the client and the depot does not exceed value μ_1 .

In lines 21–32 of Algorithm 2 deal with the special case where two neighbouring clients are not allocated to the same depots. Lines If we detect two clients c and c' such that *i*) distance between c and c' is no more than value μ_2 (line 24), *ii*) c is allocated to some depot d (line 26) *iii*) c' is pre-allocated to some depot $d' \neq d$ (line 26), *iv*) $dist(d, c') + dist(c', c) + dist(c, d') < \mu_3$ (line 26), then c is pre-allocated to d' (line 27).

Algorithm takes the pre-allocation matrix $preA$ as an input, selects p depots and builds the allocation matrix A . The selection of depots is based on a score. For each depot, this score corresponds to the number of clients pre-allocated to it (lines 4–6).

The algorithm is an iterative process based on a roulette wheel selection of the next depot. On lines 8–10, the depots are ranked in non-increasing order of their scores. Then depot position at rank $\lceil y^\beta |D| \rceil$ is selected, where y is a random value in the interval $[0, 1)$ and β a parameter strictly greater than 1.

Once a depot, called d^{new} , has been selected, all clients pre-allocated to d^{new} are allocated to it. Thus, their pre-allocation to all other depots is erased (line 18) and the score of other depots is updated (line 19).

The main loop (lines 7–24) stops as soon as p depots have been selected or all clients have been allocated to some depot. If p depots have been selected and some clients are still not allocated, the remaining clients are allocated to a dummy depot δ (lines 25–30). The dummy depot has an arbitrarily large fixed cost and we assume there is a direct route from δ to all clients. This dummy depot is used to guarantee the existence of a “feasible” solution.

Algorithm 1 The proposed sampling matheuristic

Require: The set R of all routes generated

Require: A parameter α (number of subsets)

Require: $t1$ and $t2$: CPU allocated to Cplex in the inner and outer loop of the algorithm respectively

Require: J : parameter

```
1:  $z^* = +\infty$ 
2:  $collectedRoutes \leftarrow \emptyset$ 
3:  $preA = preAllocate()$  {Pre-allocation of clients to depots}
4: repeat
5:    $amelioration = FALSE$ 
6:    $R = R \setminus collectedRoutes$ 
7:   Shuffle all routes in  $R$ 
8:    $S \leftarrow SelectLocations(preA, \alpha)$  {selects  $\alpha$  routes}
9:    $R' = RouteFiltering(R, S)$ 
10:   $RouteSampling(R', \alpha)$  {partition  $R'$  into  $\alpha$  independent subsets of routes}
11:  for  $s = 1$  to  $\alpha$  do
12:     $R_s = \text{routes in subset } s$ 
13:     $bestRoutes_s = \emptyset$ 
14:    for  $j = 1$  to  $J$  do
15:       $R_s \leftarrow R_s \setminus bestRoutes_s$  {remove the route just found, active when  $J > 1$ }
16:       $(z, bestRoutes_s) = SolveLIRP(R_s, t1)$ 
17:       $collectedRoutes \leftarrow bestRoutes_s$ 
18:    end for
19:  end for
20:   $(z, bestRoutes) = solveLIRP(collectedRoute, t2)$ 
21:  if  $z < z^*$  then
22:     $z^* = z$ 
23:     $amelioration = TRUE$ 
24:  end if
25: until  $amelioration = FALSE$ 
26: return  $bestRoutes$ 
```

Algorithm 2 Pre-allocations of clients to depots

Require: D : set of depots

Require: C : set of clients

Require: $N_{close}, \mu_1, \mu_2, \mu_3$: parameters

```
1:
2: Initialization:  $preA$ :  $|C| \times |D|$  pre-allocation matrix filled with zeros
3: Initialization:  $dist$ :  $|C| \times (|D| + |C|)$  distance matrix between clients and other clients + depot
4:
5: for all clients  $c \in C$  do
6:   for all depots  $d \in D$  do
7:     calculate the distance  $dist(c, d)$  between  $c$  and  $d$ 
8:   end for
9: end for
10:
11: for all clients  $c \in C$  do
12:    $ClosestDepots(c)$  = List of all depots  $d \in D$  ranked in non-decreasing order of the distances  $dist(c, d)$ .
13:    $A(c, ClosestDepots(c)[1]) = 1$  {pre-allocate client  $c$  to its closest depot}
14:   for  $n = 2$  to  $N_{close}$  do
15:     if  $dist(c, ClosestDepots(c)[n]) < \mu_1$  then
16:        $A(c, ClosestDepots(c)[n]) = 1$  {pre-allocate client  $c$  to depot ranked  $n$ }
17:     end if
18:   end for
19: end for
20:
21: for all clients  $c \in C$  do
22:   Rank all clients  $c' \in C, c' \neq c$  in non-decreasing order of the distances  $dist(c, c')$ 
23:   for all clients  $c' \in C$  do
24:     if  $dist(c, c') < \mu_2$  then
25:       for all  $d' \in ClosestDepots(c')$  do
26:         if  $A(c', d') = 1$  and  $preA(c, d') = 0$  and  $dist(d', c') + dist(c', c) + dist(c, d') < \mu_3$  then
27:            $preA(c, d') = 1$ 
28:         end if
29:       end for
30:     end if
31:   end for
32: end for
33: return  $preA$ 
```

Algorithm 3 Heuristic selection of depots

Require: D : set of depots
Require: C : set of clients
Require: $preA$: pre-allocation matrix
Require: p : number of depots to be selected
Require: β : parameter strictly greater than 1.

- 1: Initialize: fill the $|C| \times |D|$ allocation matrix A with zeros
- 2: $nbClient \leftarrow 0$ number of clients allocated to some depot
- 3: $S \leftarrow \emptyset$: list of selected depots
- 4: **for** each depot $d \in D$ **do**
- 5: $score(d) =$ number of clients pre-allocated to d
- 6: **end for**
- 7: **while** $|S| < p$ and $nbClient < |C|$ **do**
- 8: Rank all depots by non-increasing scores
- 9: Choose a random number y from the interval $[0, 1)$
- 10: $d^{new} = D[\lceil y^\beta |D| \rceil]$
- 11: $S = S \cup \{d^{new}\}$.
- 12: **for** all clients $c \in C$ **do**
- 13: **if** $preA[c][d^{new}] = 1$ **then**
- 14: $A[c][d^{new}] = 1$
- 15: $nbClient \leftarrow nbClient + 1$
- 16: **for** all depots $d \neq d^{new}$ **do**
- 17: **if** $preA[c][d] = 1$ **then**
- 18: $preA[c][d] \leftarrow 0$
- 19: $score(d) \leftarrow score(d) - 1$
- 20: **end if**
- 21: **end for**
- 22: **end if**
- 23: **end for**
- 24: **end while**
- 25: **if** $nbClient < |C|$ **then**
- 26: Add a dummy depot δ
- 27: **for** all $c \in C$ such that $\sum_{d \in D} preA[c][d] > 0$ **do**
- 28: $A[c][\delta] = 1$
- 29: **end for**
- 30: **end if**
- 31: **return** S
- 32: **return** A

After selecting a subset of depots, we filter the list of existing routes by removing all routes visiting unselected depots.

3.2 Filtering SD routes after depot selection

Algorithm 4 filters routes finishing at unselected depots.

Algorithm 4 SD Routes filtering

Require: $SDloop$: set of SD routes

Require: D set of depots

Require: C : set of clients

Require: S : set of selected depots

```

1:  $filteredSD \leftarrow \emptyset$ 
2: for all  $d \in D$  do
3:   if  $d \in S$  then
4:      $filteredSD = filteredSD \cup \{s, d\}$ 
5:   end if
6: end for
7: return  $filteredSD$ 

```

3.3 Filtering DC routes after depot selection

Algorithm 5 filters routes starting from unselected depots. Routes that start from an unselected depot (line 4), or that contain clients that are not pre-allocated to the starting depot of the route (line 6) are filtered.

Algorithm 5 DC Routes filtering

Require: $DCloop$: set of DC routes

Require: $preA$: pre-allocation matrix

Require: C : set of clients

Require: S set of selected depots

```

1:  $filteredDC \leftarrow \emptyset$ 
2: for each route  $r \in DCloop$  do
3:    $keep = 1$ 
4:   if  $depot(r) \in S$  then
5:     for all  $c \in r$  do
6:       if  $preA[c][depot(r)] = 0$  then
7:          $keep = 0$ 
8:          $c \leftarrow$  end of the route
9:       end if
10:    end for
11:  else
12:     $keep = 0$ 
13:  end if
14:  if  $keep = 1$  then
15:     $filteredDC = filteredDC \cup \{r\}$ 
16:  end if
17: end for
18: return  $filteredDC$ 

```

References