# Accepted Manuscript

Pre-positioning of relief items in humanitarian logistics considering lateral transhipment opportunities

Serhat Başkaya, Mustafa Alp Ertem, Serhan Duran

PII: S0038-0121(16)30040-4

DOI: 10.1016/j.seps.2016.09.001

Reference: SEPS 540

To appear in: Socio-Economic Planning Sciences

Received Date: 21 February 2016

Revised Date: 17 June 2016

Accepted Date: 12 September 2016

Please cite this article as: Başkaya S, Ertem MA, Duran S, Pre-positioning of relief items in humanitarian logistics considering lateral transhipment opportunities, *Socio-Economic Planning Sciences* (2016), doi: 10.1016/j.seps.2016.09.001.

This is a PDF file of an unedited manuscript that has been accepted for publication. As a service to our customers we are providing this early version of the manuscript. The manuscript will undergo copyediting, typesetting, and review of the resulting proof before it is published in its final form. Please note that during the production process errors may be discovered which could affect the content, and all legal disclaimers that apply to the journal pertain.



# Pre-positioning of Relief Items in Humanitarian Logistics Considering Lateral Transhipment Opportunities

3

The main objective of this study is to investigate the inclusion of lateral transhipment 4 5 opportunities into the humanitarian relief chain and to examine the effect of different 6 parameters on minimizing the average distance travelled per item while serving the 7 beneficiaries. Direct shipment model (DT), lateral transhipment model (LTSP) and maritime 8 lateral transhipment model (MLTSP) are developed and compared between each other by using a real life earthquake scenario prepared for the city of Istanbul by JICA (Japanese 9 10 International Cooperation Agency). Developed mathematical models decide on the locations and number of disaster relief facilities, quantity of relief items to hold at those facilities, and 11 quantity of lateral transhipment between the facilities. Vulnerability of the roads and 12 heterogeneous capacitated facilities are also considered. It can be concluded that both LTSP 13 14 and MLTSP models gave better results than DT model and lateral transhipment option helps beneficiaries to obtain relief items faster and with higher service level. 15

16

Key Words: Freight transportation, maritime transportation, relief chain, capacitated facility
 location, vulnerability

19	
20	
21	
22	
23	
24	
25	
26	
27	
28	
29	
30	
31	
32	
33	
34	

#### 35 **1. Introduction**

36 From 2003 to 2012, annual average of 106,654 people were reported dead, more than 216 37 million people were reported to be affected by disasters, and close to \$157 billion worth of 38 economic damage was reported (Guha-Sapir et al., 2014). These facts reveal the importance 39 of disaster management in mitigating the negative effects of the disaster. Humanitarian 40 logistics, which plays a key role in every stage of disaster relief operations, is defined as "the 41 process of planning, implementing and controlling the efficient, cost-effective flow and storage of goods and materials, as well as related information, from point of origin to point of 42 consumption for the purpose of meeting the end beneficiary's requirements" (Thomas and 43 44 Mizushima, 2005). When a state of emergency is declared and aid is appealed, resources such as relief personnel, relief goods and equipment are mobilized to the disaster location. By its 45 definition, mobilization of resources as well as its predecessor and successor operations in a 46 47 relief chain (Duran et al., 2013) can be categorized as humanitarian logistics, which contribute to more than 80% of the total relief costs (Van Wassenhove, 2006). Although local 48 government of the disaster location is mainly responsible to alleviate the suffering of its 49 people (Thomas and Fritz, 2006), non-governmental organizations (NGOs) as well as other 50 relief aid agencies offer their help to transport the right number of relief goods on time to the 51 52 right place (Tatham and Pettit, 2010).

Supply chains are usually considered to be consisting of vertical transportation through 53 several echelons (i.e. levels) such as manufacturer, warehouse, retailer, customer etc. The 54 55 practice of allowing horizontal transportation within the same echelon is called lateral transhipment (Axsater, 2006) and is mostly used for low demand, high value items where 56 57 emergency orders are allowed (Wong et al., 2006; Kutanoglu and Mohajan, 2009). In settings where lateral transhipment is observed, retailers might keep only certain types of items and 58 59 replenish those items from the warehouses. As a cure to the burden of waiting for next regular 60 warehouse shipment or placing emergency orders with high cost to the warehouse, 61 transhipments from other retailers with adequate inventory is utilized. Thus, retailers face two sources of demand (from customers and other retailers) and two sources of supply (from 62 warehouses and other retailers) (Axsater, 2006). 63

Inspired from the emergency nature of lateral transhipment decisions in commercial logistics, 64 lateral transhipment in humanitarian logistics can also be a viable alternative to alleviate the 65 suffering of beneficiaries within the shortest time possible. Lateral transhipment in 66 humanitarian logistics is observed when aid distribution centres transfer relief items among 67 themselves when they cannot satisfy the immediate need of beneficiaries from their own 68 inventory. To the best of our knowledge, lateral transhipment in humanitarian logistics has not 69 70 been analysed thoroughly in the literature. The objective of this study is addressing this literature gap and proposing an integrated model for facility location and transportation 71 72 decisions including lateral transhipments.

The rest of the paper is organized as follows. In the second section, we present the related literature. The problem is defined and the related systems are described in the third section. Proposed mathematical formulations are presented in the fourth section. The fifth section provides the results of experimental studies conducted for the city of Istanbul with the real life data. Finally, we conclude with our major findings and possible future research directions.

#### 78 **2. Literature Review**

Disaster management can be analyzed in four phases, namely, mitigation, preparedness, 79 response and recovery (Altay and Green, 2006). Most of the studies in humanitarian logistics 80 have focused on the preparedness and response phases (Altay and Green, 2006). In their 81 review study, Caunhye et al. (2012) state that inventory pre-positioning, evacuation and relief 82 distribution aims are brought together in location analysis in most of the facility location 83 84 optimization models in humanitarian logistics. The decisions are varied such as commodity 85 pre-positioning, facility selection among potential local and global distribution centres, and optimizing facility size. In the pre-positioning literature, the most frequent objectives are 86 minimizing costs of setting up relief centres, transportation (Galindo and Batta 2013, Lin et al. 87 2012, Khayal et al. 2015) and commodity procurement costs, average (Duran et al., 2011) or 88 maximum response time, unfilled demand (Afshar and Haghani, 2012) and expected number 89 90 of casualties left behind or maximizing beneficiaries' coverage. Huang et al. (2012) describe 91 efficiency, efficacy and equity types of objective functions for relief routing. Facility location 92 problem can also be solved together with the routing of vehicles as in Ukkusuri and 93 Yushimito (2008).

Two stage stochastic models are utilized in some pre-positioning studies. Barbarosoglu and 94 Arda (2004) propose a two-stage stochastic programming model to plan the transportation of 95 vital first-aid commodities to disaster-affected areas during emergency response where the 96 97 capacities of the arcs in the road network, the supply amounts and the resource requirements 98 are considered to be random. Mete and Zabinsky (2010) develop a stochastic optimization approach selecting the storage locations and amounts of medical supplies to minimize 99 100 warehouse operation costs, the response time and unfilled demand rate balancing the 101 preparedness and risk despite the uncertainties of disaster events. Bemley et al. (2013) develop a two-stage stochastic pre-positioning model to maximize expected amount of 102 103 repaired ports providing short-term port recovery from weather events such as hurricanes.

Scenario based approaches are also utilized in the pre-positioning literature. Balçık and Beamon (2008) propose a scenario-based model for a pre-positioning system balancing the costs against the risks to determine the number and the location of distribution centres in a relief network and the amount of each relief commodity stored at each facility. Duran et al. (2011) develop a mathematical model to obtain the configuration of the supply network that minimizes the average response time over all the demand instances and decide which warehouse to open and how to allocate the inventory among them.

Commercial studies on lateral transhipment are not directly related to disaster response, but 111 still have some common characteristics to humanitarian logistics settings. Some of these 112 characteristics are the uncertainty in demand, existence of possible future states, and 113 114 uncertainty in the number of facilities to be established. These characteristics are related to the uncertainty in the time, place and the effect of a disaster. Most of the commercial lateral 115 transhipment studies are related to repairable spare parts. In one of the earliest studies on 116 lateral transhipment, Lee (1987) presents a model of pooling groups with identical retailers. 117 Demand of one retailer is satisfied from another retailer in the same pooling group. Different 118 119 priority rules between available retailers and optimal stocking levels for various service levels are also analysed. Axsater (1990) generalizes the pooling group idea to non-identical retailers. 120 His method shows an improvement on Lee (1987)'s work when the proportion of emergency 121 122 transhipments is large. Commercial studies differ from humanitarian logistics by their demand 123 rate and item value. Commercial lateral transhipment is often used for low demand and high 124 value items. On the other hand, lateral transhipment in humanitarian logistics is used during a demand surge (i.e. high demand) and for low value items (e.g. bottled water and meals-ready-125 126 to-eat).

Lateral and emergency shipments occur in response to stock outs. Wong et al. (2006) study a 127 multi-item, continuous review model of two-location inventory systems for repairable spare 128 129 parts. The objective of the study is to minimize the total costs for inventory holding, lateral transhipments and emergency shipments subject to a target level for the average waiting time 130 per demanded part at each of the two locations. Kutonoğlu and Mohajan (2009) consider a 131 132 two-echelon service parts logistics system with one central warehouse and a number of local warehouses that meet all the time-based service level constraints at minimum total cost 133 134 including inventory holding cost, transportation cost, and penalty cost due to lost demand. Time-based service level constraints are similar to allowable maximum response time or 135 136 maximum distance constraints in humanitarian logistics.

137 Ozkapici et al. (2016) study the problem of locating disaster relief facilities in the city of 138 Istanbul utilizing the Bosphorus strait. The authors consider maritime transportation for relief 139 item distribution in the city of Istanbul where two main ports and a container ship located on 140 the Marmara Sea are considered as main supply facilities. Ozkapici et al. (2016) conclude that 141 including maritime transportation into the relief item distribution system provides a more 142 flexible humanitarian logistics system for Istanbul. Inspired from Ozkapici et al. (2016), one 143 of the mathematical models developed in this study uses maritime transportation with lateral 144 transhipment opportunities.

Three works can be cited as the most related to this study in humanitarian logistics. Reyes et al. (2013) show that lateral transhipment in a disaster relief system is more efficient using a simulation model based on system dynamics. Stanger et al. (2013) illustrate the use of lateral transhipment in blood transportation for UK hospitals. They demonstrate the real life benefits of lateral transhipment based on comprehensive case studies and surveys. Mulyono and Ishida (2014) build a logistics and inventory model using probabilistic cellular automata for the Page 4 of 21

enterprise inventory model and self-repair network model, which is applicable to 151 humanitarian relief situations. Mulyono and Ishida (2014) use a real life data set from a 152 volcanic eruption (Sinabung Mountain - September 2013) in Indonesia to validate their 153 154 model. Although Reves et al. (2013), Stanger et al. (2013), and Mulyono and Ishida (2014) illustrate the use of lateral transhipment in humanitarian relief situations; they do not utilize a 155 mathematical programming model in their studies. In this study, the main objective is to 156 investigate whether lateral transhipment in humanitarian logistics provides flexibility and 157 decreases average travel distance comparing mathematical models with and without lateral 158 transhipment. 159

# 160 **3. Description of the Relief Item Distribution System**

In this section, a description of the proposed relief item distribution system, sources of the 161 162 data used and the assumptions are given, respectively. A distribution system with two 163 echelons is proposed here for a possible earthquake scenario where we have in the upper echelon the relief facilities and in the lower echelon demand locations. Each demand location 164 is assigned to only one relief facility and relief items are transported from relief facilities to 165 demand locations according to this assignment. This type of material shipment is called as 166 direct shipment. Lateral transhipment between relief facilities is also possible. Any relief 167 168 facility can engage in lateral transhipment with a neighbour relief facility. This type of 169 material shipment is called as lateral transhipment. In lateral shipment any relief facility can 170 satisfy demand of any demand location assigned to it by using excess stock of the neighbour 171 relief facility. The suggested distribution system for relief items is presented in Figure 1.

172 <><Insert Figure 1 about here>>

For each relief facility, it is allowed to use only one neighbour relief facility for lateral transhipment. One standard "relief item package" is delivered to each family of four people. This package contains bottles of water and food cans. We assume that relief facilities are willing to release true information about their inventory position to other relief facilities and capacity of land vehicles is assumed to be enough for deliveries.

- 178 **4. Mathematical Models**
- 179 Mixed integer programming formulations of direct shipment model, lateral transhipment 180 between supply points model and maritime lateral transhipment model are presented in this 181 section, respectively.

# 182 *4.1 Model with the Direct Shipment only (DT)*

- Let index sets of *I* and *J* represent the set of possible relief facilities and the set of demand
  locations, respectively. We define the decision variables of the DT model as:
  - (1, if relief facility i is opened,
  - $y_i$ :  $\begin{cases} 1, & \text{if relief facil} \\ 0, & otherwise. \end{cases}$

	(1 if demand location <i>i</i> is assigned to relief facility <i>i</i>	185
$m_{ij}$ :	1, if demand location j is assigned to rener facility i,	186
	(0, <i>Utilet wise</i> .	187
$q_i$ :	Quantity of relief item held at relief facility <i>i</i> ,	188
<u>.</u>	Quantity of raliafitam cant to damand point i from raliaf facility i	189
$x_{ij}$ :	Quantity of fener hem sent to demand point j from fener facility i,	190
		191
and its	s parameters as:	
Ш7.	A kie avarber	
W:	A big number,	
N:	Quantity of relief items required by a beneficiary at any demand p	oint,
<i>P</i> :	Maximum number of relief facilities to open,	
<i>R</i> :	Maximum distance for a relief item to travel,	
F:	Maximum number of beneficiaries a school class can serve,	
$v_{ii}$ :	Vulnerability factor between relief facility <i>i</i> and demand location	i,

- $d_j$ :
- Number of people affected at demand location j, Number of school classes available at relief facility i,  $c_i$ :
- Distance between relief facility *i* and demand location *j*.  $r_{ij}$ :

194

192 193

195 Thus, the complete DT model can be written as:

$\sum_{i \in I} \sum_{j \in J} \left[ x_{ij} r_{ij} \left( 1 + v_{ij} \right) \right]$		
$\frac{\sum_{j \in J} (d_j N)}{\sum_{j \in J} (d_j N)}$		(1)
subject to		
$\sum_{i \in I} x_{ij} \ge d_j N$	j∈J	(2)
$r_{ij}(1+v_{ij})m_{ij} \le R$	i ∈ I, j∈ J	(3)
$\sum_{j \in J} x_{ij} \le q_i$	$i \in I$	(4)
$\sum_{i \in I} y_i \le P$		(5)
$\sum_{i \in I} m_{ij} = 1$	$j \in J$	(6)
$\sum_{i \in I} m_{ij} \le W y_i$	$i \in I$	(7)
$x_{ij} \leq W m_{ij}$	$i \in I, j \in J$	(8)
$q_i \leq y_i c_i NF$	$i \in I$	(9)
$\sum_{i \in I} q_i \leq \left\{ \sum_{j \in J} d_j \right\} N \times 1.01$		(10)
$x_{ij}, q_i \geq 0$	$i \in I, j \in J$	(11)
$y_i, m_{ij} \in \{0,1\}$	$i \in I, j \in J$	(12)

197 The objective function (1) minimizes the average distance travelled per the relief item. 198 Vulnerabilities of the routes affect the distances by inflating them. Horner and Widener 199 (2011) concluded that disruption levels of a network after a disaster increased the average 200 distance between a neighbourhood and its relief centre. Inspired from Horner and Widener 201 (2011)'s conclusion, original distance of a route is inflated here by the vulnerability of that 202 route ranging from 0 to 1.0 where 1.0 represents the most vulnerable case using 203 [Inflated distance = Original distance × (1 + Vulnerability)] equation.

204 Constraint set (2) ensures that demand for relief items at each demand point is met. With the 205 Constraint (3), relief items do not travel more than R, and the relief items sent do not exceed 206 the respective inventory held at the relief facility i via Constraint (4). Via Constraint (5), at 207 most P relief facilities can be opened. Constraints (6-8) make sure that each demand location iis assigned to only one relief facility, a demand location can be assigned to a relief facility 208 209 that is opened and relief items cannot be sent from a relief facility to a demand location unless that demand location is assigned to that relief facility. Constraint set (9) imposes an upper 210 bound on the quantity of relief items to be held at a relief location considering the number of 211 212 classrooms in the district and the maximum number of people (F) that can be served from a 213 school classroom. The parameter F is considered like a service level; lower F values 214 corresponds to better service for the beneficiaries. Assuming that the total capacity of the 215 facilities is 101% of total demand, Constraint (10) is added.

# 216 4.2 Model with Lateral Transhipment option between Supply Points (LTSP)

With the inclusion of the lateral transhipment option to the model, we need a new index for 217 the relief facilities used as lateral transhipment source. Let us denote it as i' under the set I. In 218 219 addition to the parameters used in the DT model, the new parameters related with the relief 220 facilities acting as lateral transhipment sources in the LTSP model are;  $r_{i'i}$ , the travel distance 221 between relief facilities i' and demand location  $j_i r_{i'i}$ , the travel distance between relief facilities i' and relief facility  $i, v_{i'j}$ , vulnerability factor between relief facilities i' and demand 222 location j, and  $v_{i'i}$ , vulnerability factor between relief facilities i' and relief facility i. If we 223 define the additional decision variables as: 224

$$t_{ii'j}$$
: {1, if relief facilities *i* and *i'* engages in lateral transhipment for demand location *j*, otherwise,

 $\bar{x}_{ii'j}$ : Quantity of relief item sent to demand location *j* from facility *i* through facility *i'*,

 $f_{ii'}$ : {1, if relief facilities *i* and *i'* engages in lateral transhipment, otherwise,

- then the complete LTSP model can be represented as:
- 226

$$\text{Minimize} \frac{\sum_{i \in I} \sum_{j \in J} [x_{ij} r_{ij} (1 + v_{ij})] + \sum_{i \in I} \sum_{i' \in I} \sum_{j \in J} [\bar{x}_{ii'j} (r_{i'j} (1 + v_{i'j}) + r_{ii'} (1 + v_{ii'}))]}{\sum_{j \in J} (d_j N)}$$
(13)

Page **7** of **21** 

<i>subject to</i> (3), (5), (6), (7), (8), (9), (10), <i>and</i>		
$\sum_{i\in I} x_{ij} + \sum_{i\in I} \sum_{i'\in I} \bar{x}_{ii'j} \ge d_j * N$	j∈J	(14)
$(r_{ii\prime} * (1 + v_{ii'}) + r_{i\prime j} * (1 + v_{i'j}))t_{ii\prime j} \le R$	$i \in I, i' \in I, j \in J, i \neq i'$	(15)
$\sum_{j \in J} x_{ij} + \sum_{j \in J} \sum_{i' \in I} \bar{x}_{ii'j} \le q_i$	$i \in I, i \neq i'$	(16)
$\sum_{i \in I} f_{ii\prime} \le 1$	$i' \in I, i \neq i'$	(17)
$\bar{x}_{ii\prime j} \leq W t_{ii\prime j}$	$i \in I, i' \in I, j \in J, i \neq i'$	(18)
$\sum_{j \in J} \sum_{i \nu \in I} t_{i \nu j} \le W y_i$	$i \in I, i \neq i'$	(19)
$\sum_{j \in J} \sum_{i \in I} t_{ii'j} \le W y_{i'}$	$i' \in I, i' \neq i$	(20)
$\sum_{i\in I} t_{ii'j} \le m_{i'j}$	$i' \in I, j \in J, i \neq i$	(21)
$\sum_{i \in I} t_{ii'j} \le W f_{ii'}$	$i \in I, i' \in I, i \neq i$	(22)
$x_{ij}, \bar{x}_{ii\prime j}, q_i \ge 0$	$i \in I, i' \in I, j \in J, i \neq i'$	(23)
$y_i, m_{ij}, t_{ii'j} \in \{0, 1\}$	$i \in I, i' \in I, j \in J, i \neq i'$	(24)

The objective function (13) again minimizes the average distance travelled per a relief item 227 228 including the vulnerability effect. Constraint (14) ensures that demand of every demand location is 229 satisfied either directly from relief facilities or through lateral transhipment. Constraints (3) and 230 (15) limit the travel distance of relief items. Constraint (16) ensures that the capacity of a relief facility opened is sufficient to meet total demand assigned to that relief facility. Constraint (17) 231 ensures that any relief facility is allowed to engage in lateral transhipment with at most one 232 neighbour relief facility for a demand location. Constraint (18) ensures that relief item cannot be 233 sent through a relief facility unless lateral transhipment is allowed. Constraints (19-20) allow only 234 the open relief facility pairs to engage in lateral transhipment. Constraint (21) allows that lateral 235 236 transhipment is engaged with the neighbour relief facility to satisfy the demand of location assigned to that neighbour relief facility. Constraint (22) provides that lateral transhipment can be 237 made for demand location *i* if the related two relief facilities engage in lateral transhipment. 238

#### 239 4.3 Model with Maritime Lateral Transhipment option between Supply Points (MLTSP)

For the MLTSP model, this time the index for the ports visited for lateral transhipment is defined as k and k' under the set K. We also need to define new parameters namely;  $v_{ik}$ , vulnerability factor between relief facility i and port k,  $v_{kk'}$ , vulnerability factor between port k and port k',  $r_{ik}$ , distance between relief facility i and port k,  $r_{kk'}$ , distance between port k and port k', *cap*, capacity of a ship,  $n_s$ , number of ships,  $s_s$ , speed of a ship,  $s_t$ , speed of a land vehicle (e.g. truck). When we define the additional decision variables as:

247  $\bar{x}_{ikk'i'j}$ : quantity of items sent to location *j* from facility *i* through ports *k* and *k'* and facility *i'*, 248  $z_{kk'}$ : number of ships used between port *k* and port *k'* for the shipment of relief items,

 $b_{ikk'i'j}$ : {1, if relief facilities *i* and *i'* engages in lateral transhipment through ports *k* and *k'*, otherwise.

the MLTSP model can be written as:

$$\sum_{i \in I} \sum_{j \in J} [x_{ij}r_{ij}(1+v_{ij})] + \sum_{i \in I} \sum_{j \in J} \sum_{i' \in I} [\bar{x}_{ii'j}(r_{i'j}(1+v_{i'j})+r_{ii'}(1+v_{ii'}))] + \sum_{i \in I} \sum_{i \in I} \sum_{k \in K} \sum_{k' \in K} \sum_{k' \in K} \sum_{i' \in I} \sum_{j \in J} [\bar{x}_{ikk'i'j}(r_{i'j}(1+v_{i'j})+r_{i'k'}(1+v_{i'k'})+r_{kk'}(\frac{s_t}{s_s})*(1+v_{kk'})+r_{ik}(1+v_{ik})] + \sum_{i \in I} \sum_{i \in I} \sum_{i' \in I} \sum_{i$$

$$(r_{ik} * (1 + v_{ik}) + r_{kk'} * \left(\frac{s_t}{s_s}\right) * (1 + v_{i'k'}) + r_{k'i'} * (1 + v_{i'k'}) + r_{i'j} * (1 + v_{i'j})) * b_{ikk'i'j} \leq R$$
   
  $k,k' \in K; i,i' \in I; j \in J, i \neq i',k$  (27)

$$\sum_{j \in J} x_{ij} + \sum_{j \in J} \sum_{i \prime \in I} \bar{x}_{ii\prime j} + \sum_{j \in J} \sum_{i \prime \in I} \sum_{k \in K} \sum_{k' \in K} \bar{x}_{ikk\prime i\prime j} \le q_i \qquad i \in I, i \neq i'$$
(28)

 $k, k' \in K; i, i' \in I; j \in J, i \neq i', k$  $\neq k'$ 

(29)

 $\bar{x}_{ikk'i'j} \leq W * b_{ikk'i'j}$ 

$$\sum_{j \in J} \sum_{i r \in I} \sum_{k \in K} \sum_{k r \in K} b_{ikkrirj} \leq W * y_i \qquad i \in I, i \neq i' \qquad (30)$$

$$\sum_{j \in J} \sum_{i r \in I} \sum_{k \in K} \sum_{k r \in K} b_{ikkrirj} \leq W * y_i, \qquad i \in I, i \neq i' \qquad (31)$$

$$z_{kkr} \leq \sum_{j \in J} \sum_{i \in I} \sum_{i r \in I} \bar{x}_{ikkrirj} \qquad k \in K, k' \in K, k \neq k' \qquad (32)$$

$$\sum_{k \in K} \sum_{k' \in K} z_{kk'} \leq n_s$$

$$\sum_{j \in J} \sum_{i \in I} \sum_{i \in I} \bar{x}_{ikk'i'j} \leq cap * z_{kk'}$$

$$k \in K, k' \in K, k \neq k'$$

$$\sum_{k' \in K} \sum_{i \in I} \sum_{k \in K} b_{ikk'i'j} \leq m_{i'j}$$

$$\sum_{\substack{x_{ij}, \bar{x}_{ii'j}, q_i, \bar{x}_{ikk'i'j} \leq 0 \\ y_i, m_{ij}, t_{ii'j}, b_{ikk'i'j} \in \{0,1\}$$

$$z_{kk'} \text{ integer}$$

$$(33)$$

$$k \in K, k' \in K, k \neq k'$$

$$(34)$$

$$i' \in I, j \in J, i \neq i$$

$$i \in I, i' \in I, j \in J, i \neq i'$$

$$(35)$$

$$i \in I, i' \in I, j \in J, i \neq i'$$

$$(36)$$

$$i \in I, i' \in I, j \in J, i \neq i'$$

$$(37)$$

$$k \in K, k' \in K, k \neq k'$$

$$(38)$$

250

The objective function (25) minimizes the average distance travelled per a relief item including the vulnerability affect. Constraint (26) ensures that demand of every demand location is satisfied either directly from relief facilities or through lateral transhipment. Constraints (3), (15) and (27) limit the travel distance of relief item. In Constraint (27) the 255 distance between ports is multiplied by the ratio of speed of land vehicle to speed of ship in order to convert the distances travelled by ship in an hour to the distance travelled by land 256 vehicle in an hour. Constraint (28) ensures that the capacity of a relief facility opened is 257 258 sufficient to meet total demand assigned to that relief facility. Constraints (18) and (29) ensure that relief item cannot be sent through a relief facility unless lateral transhipment is allowed. 259 Constraints (19-20) and (30-31) allow only the open relief facility pairs to engage in lateral 260 transhipment. Constraints (21) and (35) allow lateral transhipments to neighbour relief facility 261 to satisfy demand of demand location that assigned to that neighbour relief facility. Constraint 262 (32) is used in case there is no relief item shipment between ports, any ship cannot be utilized. 263 Constraint (33) ensures that number of ship is limited. Constraint (34) ensures that shipment 264 amount between ports cannot exceed the total capacity of ships used between that ports. 265 Constraints (36-38) define restrictions on decision variables. 266

# **5. Experimental Study Applied for a possible Earthquake at the city of Istanbul**

An experimental study is conducted to validate the direct shipment model (DT), lateral transhipment between supply points model (LTSP) and maritime lateral transhipment model (MLTSP). Models are solved by GAMS 24.2 with Cplex 12.6 Solver and the average solution times of the models are provided in Table 1.

272

#### <<insert table 1 about here>>

#### 273 *5.1. Data sources*

The main data sources utilized in this study are the JICA Report (2002) and Ozkapici et al.(2016). Types of data in the system and the methods to update these data are explained in the

following. Data set for the experiments are provided on the online version of this article.

277 In the JICA report, damage estimations and beneficiary populations are provided based on the 278 districts of Istanbul. As a result, 37 districts of Istanbul are taken as locations of demand. For 279 each district, district centre point is obtained and represented with a single coordinate ( $N^{\circ}$ ;  $E^{\circ}$ ) calculated as the weighted average of the coordinates of its neighbourhoods. The coordinate 280 281 of each neighbourhood is taken as the coordinate of the Mukhtar office (i.e. local government office located in each neighbourhood in Turkey) belonging to that neighbourhood. Then, the 282 coordinate of a district is calculated by taking the weighted average of coordinates of its 283 284 neighbourhoods, where the weights are the populations of the neighbourhoods.

There are 37 potential relief facility locations which are the same as the demand locations.
The capacities of potential relief facility locations are estimated from available public school
buildings. As a result, the capacity of each potential relief facility is different.

JICA report (2002) states the possible number of heavily, moderately and partly damaged
buildings for each district. By using the Equation (39), for each district the average number of
people living in one building is calculated.

291 
$$A = \frac{Population of the District}{Number of Buildings in the District}$$
(39)

Page **10** of **21** 

The population data of districts in the above formulation are obtained from the Turkish Statistical Institute (2013). The number of people affected from the earthquake in each district is calculated by using the Formula (40) as in the JICA Report (2002) where B, C and Dcorresponds to number of heavily damaged buildings, moderately damaged buildings and partly damaged buildings, respectively.

# 297 Number of affected people = A \* (100% \* B + 50% \* C + 10% \* D) (40)

The number of relief items needed in each district is calculated by the Formula (41). It is assumed that the single relief item is delivered to a family of four people. As a result, formulation includes a multiplication by 0.25 (=*N*).

Relief items required = 0.25 \* number of affected people in that district (41)

Distances between relief facilities and demand locations and between relief facility pairs are obtained from Google Maps<sup>TM</sup>. The shortest distance between two points is selected from alternatives given by Google Maps<sup>TM</sup>. Travel time of relief item in the system is restricted to ensure that in a determined time interval the relief item reaches to the affected people. Maximum travel time is restricted to one and two hours. In the experimental study, we assume that the relief items are carried by trucks with an average speed of 40 km/h (= *s<sub>t</sub>*).

308 Vulnerability of the roads between demand locations and relief facilities and between relief 309 facility pairs are determined according to the road blockage probability of 7 - 15 meters wide 310 roads obtained from JICA report. For each colour denoted on Figure 2, vulnerability 311 coefficient is determined and its values are shown in Table 2 for different vulnerability cases.

- 312 <</i>
  312 <</li>
- 313 <sinsert Table 2 about here>>

To calculate the vulnerability coefficient of each path between the demand locations and relief 314 facilities and between relief facility pairs, emergency road network proposed by the JICA 315 316 report (2002) is used. This proposed emergency network is overlapped with the map of the 317 road blockage caused by building collapse on medium width road. The map shown in Figure 318 2 is divided into equal squares. Shortest path is determined on the emergency road network 319 for each pair of district by using Google Maps<sup>TM</sup>. Then the numbers of red, orange, yellow, 320 green, blue and grey squares are counted on that path. The vulnerability of that path is 321 calculated as the average of the multiplication of the number of coloured squares on the path 322 and the corresponding coefficient of that colour.

The number of classes in public schools in each district is used to determine the capacity of relief facilities. Total number of school classes available in districts is multiplied by 0.9 due to the assumption that 10% of the school classes may be damaged during disaster. The parameter F can be interpreted as a service quality level, the equal average number of people to be served from each school class. United Nations High Commissioner for Refugees (UNHCR) Emergency Handbook (UNHCR, 2015) recommends at least one final distribution point (i.e. a school in our study) for 5,000 people and a maximum 5 km walking distance for beneficiaries.

330 For the most probable earthquake scenario stated by JICA, number of beneficiaries is 331 calculated as 2,027,467 and the total number of classrooms is calculated to be 61,201 in the 332 37 districts of Istanbul. If the schools in all 37 districts were used for relief delivery and all of 333 the classrooms were used in Istanbul, the parameter F would take a value of 33 (= 334 2,027,467/61,201). This best case would not manageable due to the coordination complexities and operational costs. Therefore, we vary the parameter F between 50 and 100 to reach 335 336 feasible results. If parameter F is taken as 50, beneficiaries walk less than 0.3 km to reach a 337 nearby school classroom for relief distribution and a school serves to at most 750 beneficiaries. If parameter F is taken as 100, beneficiaries walk less than 0.5 km to reach a 338 339 nearby school classroom for relief distribution and a school serves to at most 2,500 340 beneficiaries on average. These parameters are well below the recommended standards set by the UNHCR. 341

## 342 5.2 Results of DT and LTSP Models

Firstly, solution of DT and LTSP models are presented and compared for varying number of relief facilities (P) and maximum number of people (F) that can be served from a school classroom. In these models, there is no material shipment between Anatolian side and European side due to the possible damage of the main bridges connecting the two sides of the city of Istanbul. On the other hand, in the model (MLTSP) where maritime transportation is allowed, relief items are transported between Anatolian and European sides and MLTSP model results are compared with the LTSP model results.

350 DT and LTSP models are solved for varying number of relief facilities (P); 15, 20, 25, 30, 351 maximum number of people (F) that can be served from a school classroom; 50, 75, 100, 352 vulnerability factor of roads; low, medium, high and maximum allowed distance travelled of 353 relief item (R); 40km, 60km, 80km. We observe that the vulnerability factor of the roads and 354 maximum allowed distance travelled of relief item (R) do not affect the location of the relief 355 facilities and the lateral transhipment percentages significantly. Therefore, while comparing 356 the models in this section we always assume that the maximum allowed distance travelled of 357 relief item (R) is 40 km (its minimum value) and vulnerability factor of roads are high (its 358 maximum value).

As seen on Figure 3, we observe that the average distance travelled per relief item in LTSP 359 model is always equal or better than the average distance travelled value per relief item in DT 360 model as expected since LTSP model is a relaxation of the DT model. Moreover, we also see 361 362 that to achieve the best service quality level (when only 50 (=F) people are served from a 363 classroom) lateral transhipment between the relief facilities is a requirement and at least 25 364 relief facilities should be opened. For the medium service quality level (F=75), the 365 distribution system can be managed both with and without the lateral transhipment but the 366 lateral transhipment flexibility results in around 50% improvement in the average distance travelled per relief item. For the low service quality level (F=100), the lateral transhipment option adds no value.

369

<<insert Figure 3 about here>>

In Figure 4, the percentages of lateral transhipments are presented for LTSP Model. There exists a smooth increase of the percentage of lateral transhipment as the maximum number of people (F) that can be served from a school classroom decreases. In other words, as the authorities require a higher quality service to beneficiaries, the lateral transhipment percentage increases. When only 50 people are served from a classroom, at the highest service level, 12-14 % of the relief items are supplied via lateral transhipment. Also as expected, the increase in the number of relief facilities causes the lateral transhipment amount to decrease.

377 <insert Figure 4 about here>>

#### 378 5.3 Inclusion of Maritime Transportation into the LTSP Model

LTSP model allows only land transportation in either side of the city (i.e. Anatolian and European sides). In the case of high vulnerability, sending relief items to demand locations using land vehicles is more difficult due to high risk of road blockages. Hsieh (2014) discussed ports' effect on creating extra transportation capacity and the risk of port failures with respect to vulnerability. Istanbul has many seaports on each side and daily maritime transportation is made between these ports. In case of a disaster, in addition to land transportation these ports can be used to transport relief items.

386 5.3.1 Distribution System Description and Data Sources of the MLTSP Model

In MLTSP model, transhipment between ports is possible. As a result, two transhipment nodes are added to the existing nodes at this case. Figure 5 illustrates the flow of the relief item in the suggested distribution system.

390 <<<insert Figure 5 about here>>

391 Istanbul Sea Buses (abbreviated as IDO in Turkish) is the main company on seaway transportation. IDO ports in İstanbul are considered as transhipment points in MLTSP model. 392 393 Ports are uncapacitated and ships are ready to make shipment of relief item at each port. Ports 394 located at the same side of Istanbul are not allowed to make relief item shipment between 395 each other. In the model MLTSP, one type of ship is used. Capacity is taken as 6,100 relief 396 items and the speed is taken 56 km/h ( $s_s$ ), averages of available sea bus types. Loading and 397 unloading time is assumed to be small within the overall trip duration. Maximum number of 398 ships that can be utilized for relief item transportation is determined as 25  $(n_s)$ , the number of IDO sea buses. 399

400 Distances between relief facilities and ports are calculated using Google Maps<sup>TM</sup>. The shortest 401 distance between two points is selected on Google Maps<sup>TM</sup>. Distance between ports are 402 calculated on Google Earth<sup>TM</sup> as sea miles and then converted to km. The vulnerability between ports is set as 0.001 due to the fact that there is no risk of blockage on the seawayresulting from building collapse.

## 405 5.3.2. Comparison of MLTSP Model with LTSP Model

As seen in Figure 6, MLTSP model begins to give better average distance travelled values than LTSP model only when the service level requirement is the highest (F=50). To be able to serve the city of Istanbul at the highest service level (F=50) with 20 relief facilities only, maritime transportation is also a requirement in addition to the lateral transhipment. But when the relief facilities number increases to 25, the distribution system can be managed both with and without the maritime transportation and maritime transportation results in around 7% improvement in the average distance travelled per relief item.

413 <insert Figure 6 about here>>

414 In Figure 7, we observe that the lateral transhipment percentage among the overall distribution amount can increase up to 18% with the inclusion of the maritime transportation 415 option to the distribution system. Interestingly, the percentage of total lateral transhipment in 416 417 LTSP is greater than the percentage of total lateral transhipment in MLTSP when more than 20 relief facilities are opened at the highest service level (F=50). To understand the reason of 418 419 having lower percentage of lateral transhipment in the MLTSP model, it should be noted that 420 demand of districts located in European side is larger than the demand of districts located in 421 Anatolian side. In addition to that, the number of classes of districts located in Anatolian side 422 is greater than the number of classes of districts located in European side. These two facts 423 results in relief facilities located in Anatolian side to have more excess inventory to make 424 lateral transhipment between relief facilities.

<<insert Figure 7 about here>>

425

426

#### 427 **6.** Conclusion

In this study, lateral transhipment opportunities are included into the humanitarian relief chain. Direct shipment model (DT), lateral transhipment model (LTSP) and maritime lateral transhipment model (MLTSP) are developed and these models are compared between each other by using a real life earthquake scenario developed for Istanbul by JICA (Japanese International Cooperation Agency) with varying parameters. Lateral transhipment and maritime transportation with lateral transhipment cases are examined the first time in the literature for the Istanbul case.

435

Since using highways is more difficult and time consuming in high vulnerability case, all
models are studied for the high vulnerability scenario to allow lateral transhipment between
both sides of Istanbul via seaway. MLTSP model is compared with LTSP model to examine
the effect of lateral transhipment on seaway between Anatolian and European sides of

440 İstanbul. Since demand of districts located in European side is larger than the demand of 441 districts located in Anatolian side and maximum level of inventory holding capacity of 442 districts (number of school classes of districts) located in Anatolian side is greater than 443 maximum level of inventory holding capacity of districts located in European side, all lateral 444 transhipment on seaway is directed from Anatolian side to European side.

445

We use the classes in public schools in each district as the relief facilities which directly serve to beneficiaries. Thus, the number of people served from a classroom represents a quality level of the service provided to the beneficiaries in this study. We observe that to achieve the highest service quality (when only 50 people are served from a classroom), minimum number of relief facilities to open is 25 only if lateral transhipment is utilized. This number decreases to 20 if maritime transportation is also allowed and the percentages of lateral transhipments are significant under these settings.

453

Although maritime transportation brings a small improvement to the system only for the high service level requirements, it could still be a promising alternative with additional ports and ships. With the medium service quality level, the distribution system can be managed both with and without the lateral transhipment but the lateral transhipment flexibility results in around 50% improvement in the average distance travelled per relief item over direct shipment model.

460

The most probable earthquake scenario stated by the JICA Report is used in this study. All of four scenarios in the JICA report can be studied together by developing stochastic models in the future. Developed models have relaxing assumptions on the capacity and number of land vehicles, loading/unloading time for LTSP and MLTSP model. These assumptions can be abandoned in the future.

466

- 467
- 468

469 470

471

# 472 **References**

- Afshar A and Haghani A (2012). Modeling integrated supply chain logistics in realtime large-scale disaster relief operations. *Socio-Economic Planning Sciences* 46(4): 327-338.
- 476
  476 2. Altay N and Green WG (2006). OR/MS research in disaster operations 477 management. *European Journal of Operational Research*175(1): 475-493.
- 478 3. Axsater S (1990). Modeling emergency lateral transhipments in inventory systems.
  479 *Management Science* 36: 1329–1338.
- 480 4. Axsater S (2006). Inventory control (2nd ed.). Springer:New York.

481	5.	Balçık B and Beamon BM (2008). Facility location in humanitarian relief.
482		International Journal of Logistics: Research and Applications11(2):101–121.
483	6.	Barbarosoglu G and Arda Y (2004). A two-stage stochastic programming framework
484		for transportation planning in disaster response. Journal of the Operational Research
485		Society <b>55:</b> 43-53.
486	7.	Bemley JL, Davis LB and Brock III LGB (2013). Pre-positioning commodities to
487		repair maritime navigational aids. Journal of Humanitarian Logistics and Supply
488		Chain Management <b>3:</b> 65-89.
489	8.	Caunhye AM, Nie X and Pokharel S (2012). Optimization models in emergency
490		logistics: A literature review. Socio-Economic Planning Sciences 46: 4-13.
491	9.	Duran S, Ergun O, Keskinocak P and Swann J (2013). Humanitarian Logistics:
492		Advanced Purchasing and Pre-positioning of Relief Items. Handbook of Global
493		Logistics. In: Springer (Ed.). International Series in Operations Research &
494		Management Science: New York, pp 447-462
495	10	. Duran S, Gutierrez MA and Keskinocak P (2011). Pre-positioning of emergency items
496		worldwide for CARE international. <i>Interfaces</i> <b>41</b> : 223-237.
497	11	. Galindo G and Batta R(2013). Prepositioning of supplies in preparation for a hurricane
498		under potential destruction of prepositioned supplies. Socio-Economic Planning
499		Sciences47(1): 20-37.
500	12	. Guha-Sapir D, Hoyois P, Below R(2014). Annual Disaster Statistical Review 2013:
501		The Numbers and Trends. Brussels: CRED.
502		
503	13	. Horner MW, Widener MJ (2011). The effects of transportation network failure on
504		people's accessibility tohurricane disaster relief goods: a modeling approach and
505		application to a Florida case study. <i>Natural Hazards</i> <b>59</b> ( <b>3</b> ):1619–1634
506		
507	14	. Hsieh CH (2014). Disaster risk assessment of ports based on the perspective of
508		vulnerability. Natural Hazards74(2):851–864.
509		
510	15	. Huang M, Smilowitz K and Balcik B(2012). Models for relief routing: Equity,
511		efficiency and efficacy. Transportation research part E: logistics and transportation
512		review <b>48</b> : 2-18.
513		
514	16	. JICA Report (2002). The study on a disaster prevention/mitigation basic plan in
515		Istanbul including seismic micronization in the Republic of Turkey. Final Report.
516		Japan International Cooperation Agency, December.
517		
518	17	. Khayal, D., Pradhananga, R., Pokharel, S., & Mutlu, F. (2015). A model for planning
519		locations of temporary distribution facilities for emergency response. Socio-Economic
520		Planning Sciences, 52, 22-30.
F 2 4		
521	10	$K_{\rm eff} = h_{\rm eff} E_{\rm eff} M_{\rm eff} = M_{\rm eff} (2000)$ An interaction such a final state of the stat
522	18	. Kutanogiu E and Monajan M (2009). An inventory sharing and allocation method for
523		a multi-location service parts logistics network with time-based service levels.
524	10	European Journal of Operational Research <b>194:</b> 728-742.
525	19	Lee HL (1987). A multi-echelon inventory model for repairable items with emergency
526		lateral transhipments. <i>Management Science</i> <b>33:</b> 1302-1316.

527 528 520	20. Lin, Y. H., Batta, R., Rogerson, P. A., Blatt, A., & Flanigan, M. (2012). Location of temporary depots to facilitate relief operations after an earthquake. <i>Socio-Economic Planning Sciences</i> , 46(2), 112–123
529	21 Mate HO and Zabinsky BZ (2010) Stochastic optimization of medical supply location
530	21. Mete HO and Zabilisky BZ (2010). Stochastic optimization of medical supply location
221	Economics 126: 76.84
532	22 Mulyono NB and Ishida V (2013). Humanitarian logistics and inventory model based
527	on probabilistic cellular automata. International Journal of Innovative Computing
525	Information and Control 10: 357 372
526	
550	
537	23. Ozkapıcı BD, Ertem MA and Aygüneş H (2016). Intermodal Humanitarian Logistics
538	Model Basedon Maritime Transportation in Istanbul. Natural Hazards, DOI:
539	10.1007/s11069-016-2318-9.
540	24. Reyes P, Man J and Jaska P (2013). A disaster relief inventory model based on
541	transhipment. Independent Journal of Management & Production 4: 481-509.
542	
543	25. Stanger SHW, Wilding R, Hartmann E, Yates N, Cotton S (2013). Lateral
544	transshipments: an institutional theory perspective. International Journal of Physical
545	Distribution & Logistics Management43(9): 747 – 767.
546	26. Tatham PH and Pettit SJ (2010). Transforming humanitarian logistics: the journey to
547	supply network management. International Journal of Physical Distribution &
548	Logistics Management 40: 609-622.
549	27. Thomas A and Fritz L (2006). Disaster Relief, Inc. Harvard Business Review 1-11.
550	28. Thomas A and Mizushima M (2005). Fritz institute: Logistics training: necessity or
551	luxury?. Forced Migration Review22:60-61.
552	29. Turkish Statistical Institute, (2013), "2012 Address Based Population Registration
553	Results"
554	30. Ukkusuri SV and Yushimito WF (2008). Location routing approach for the
555	humanitarian pre-positioning problem. Journal of the Transportation Research Board
556	<b>2089:</b> 18-25.
557	31. UNHCR (2015). Emergency Handbook: A UNHCR guide to agile, effective and
558	community based humanitarian emergency responses, Available at
559	https://emergency.unhcr.org/ (retrieved on 8 <sup>th</sup> May 2016)
560	32. Van Wassenhove LN (2006). Humanitarian aid logistics: supply chain management in
561	high gear. Journal of Operational Research Society 57: 475-489.
562	33. Wong H, Van Houtum GJ, Cattrysse D and Van Oudheusden D (2006). Multi-item
563	spare parts systems with lateral transhipments and waiting time constraints. European
564	Journal of Operational Research 171: 1071-1093.
565	
566	
567	
568	Table 1: Average Solution Times of the Models

	DT Model	LTSP Model	MLTSP Model
Avg Solution Time (min.)	1	5	30

569

570

Table 2: Vulnerability Coefficient of Each Severity Colour for Different Vulnerability Scenarios

	Vulnerability Coefficient		
	Low Vulnerability	Average Vulnerability	High Vulnerability
Red	0.50	0.75	0.99
Orange	0.30	0.40	0.50
Yellow	0.20	0.25	0.30
Green	0.10	0.15	0.20
Blue	0.05	0.075	0.10
Grey	0	0.025	0.05

#### 571 **Figure Captions**

- 572 Figure 1: Relief Item Flow in the Distribution System
- 573 Figure 2: Road Blockage Caused by Building Collapse Medium Width (7-15m) Road (JICA report, 2002)

574 Figure 3: Average Distance Travelled for DT and LTSP Models under High Vulnerability when *R*=40 km

- 575 Figure 4: Percentages of Lateral Transhipment for LTSP Model for *R*=40 km for High Vulnerability Factor
- 576 Figure 5: Relief Item Flow in the Distribution System Defined for MLTSP
- 577 Figure 6: Average Distance Travelled in MLTSP and LTSP Models when *R*=40 km, High Vulnerability Factor
- 578 Figure 7: Percentages of Lateral Transhipment in MLTSP and LTSP Models when R= 40 km, High
- 579 Vulnerability Factor







588

589 Figure 2: Road Blockage Caused by Building Collapse Medium Width (7-15m) Road (JICA report, 2002)





Figure 3: Average Distance Travelled for DT and LTSP Models under High Vulnerability when *R*=40 km



601 Figure 5: Relief Item Flow in the Distribution System Defined for MLTSP

603





#### Highlights

- Lateral transhipment is modeled for humanitarian logistics
- Pre-positioning and transportation decisions are addressed
- Heterogeneous capacitated facilities (i.e. schools) are utilized
- Higher service levels require lateral transhipment
- A real life earthquake scenario for Istanbul is used in experiments
- Maritime transportation option is added to lateral transhipment model

CER HER

## AUTHOR BIOGRAPHIES

Serhat BAŞKAYA

Serhat Başkaya works as an industrial engineer at Aselsan, Inc. Aselsan is a Turkish Armed Forces Foundation company addressing mostly Turkey's military defense needs through national means. He holds a B.S. and an M.S. degree from from Middle East Technical University, all in industrial engineering.

Assist. Prof. Dr. Mustafa Alp ERTEM<sup>1</sup>

Mustafa Alp Ertem, Ph.D., is an Assistant Professor of Industrial Engineering at Çankaya University (Ankara, Turkey). Before joining Çankaya University, he worked at J.B. Hunt Transport Services, Inc (NASDAQ100: JBHT). He holds a Ph.D. degree from the University of Arkansas, an M.S. degree from Middle East Technical University, and a B.S. degree from Istanbul Technical University, all in industrial engineering. His research interests are humanitarian logistics, procurement auctions, and intermodal transportation.

Assoc. Prof. Dr. Serhan DURAN

Serhan Duran is an associate professor in the Industrial Engineering Department of Middle East Technical University and has received his BS degree from the same department in 2002. He holds two masters degrees and a Ph.D. from H. Milton Stewart School of Industrial and Systems Engineering at the Georgia Institute of Technology. During his PhD studies he worked as an operations research analyst on logistics projects for both profit and non-profit organizations. He has taught courses on revenue management, financial accounting and engineering economy at Middle East Technical University and his research interests include operations research applications in humanitarian logistics, energy sector and demand management.

<sup>&</sup>lt;sup>1</sup> Corresponding author