

MINISTRY OF EDUCATION
AND TRAINING

VIETNAM ACADEMY
OF SCIENCE AND TECHNOLOGY

GRADUATE UNIVERSITY OF SCIENCE AND TECHNOLOGY



Nguyễn Văn Tú

**MULTI-COMMODITY FLOW MODEL
FOR SOME FLEET ASSIGNMENT PROBLEMS**

MASTER THESIS IN APPLIED MATHEMATICS

Hanoi, 2023

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Hà Nội - 2023

Commitment

This thesis is done by my own study under the supervision of Dr. Le Xuan Thanh and Assoc. Prof. Dr. Bui Van Dinh. It has not been defended in any council and has not been published on any media. The results as well as the ideas of other authors are all specifically cited. I take full responsibility for my commitment.

Hanoi, September 2023



Nguyen Van Tu

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Introduction

In the field of Operational Research, there is a stream of researches concerning problems on scheduling for aviation operations. This is because of the reality and wide range of applications of these problems. The main problems in this stream can be classified as follows.

- *Flight network design problem* is to determine between which airports an airline should establish flight routes.
- *Flight trajectory problem* is to determine the path of an aircraft on each flight leg (i.e., the longitude, latitude, and altitude of each point on the flight path).
- *Fleet design problem* is to determine the number of airplanes and their types to satisfy the forecasted traveling demand of passengers.
- *Flight scheduling problem* is to determine the number of flights on each flight leg, the departure and arrival times of each flight on each flight leg during a given time period (day, week, month, quarter, season, etc.).
- *Fleet assignment problem* is to schedule the airplanes in airline fleets to perform flights with predetermined departure and arrival times.
- *Aircraft routing problem* is to determine the sequence of flight legs that each airplane will fly during a given time period.
- *Crew management* is to schedule the crews to serve the flights.

The fleet assignment problem plays a central role in the aviation management, as its solution is used as input data for other problems such as aircraft routing, crew management, and as a reference for many operations such as scheduling for logistics staffs and airplane maintenance, etc. The solution to this problem is a schedule for the fleets in the airline, which is built for a specific time period (weekly, monthly, quarterly, etc.) and is used as a reference for conducting airline operations throughout that period.

Because of the importance of the fleet assignment problem in the aviation management, a number of studies have focused on this problem. In [1] the authors present a survey on such studies. An exact solution method for the fleet assignment problem is to use mixed integer programming approach. In this direction, Abara in [2] proposes the first mixed integer programming formulation for the problem. A drawback of this formulation is that its size increases exponentially with respect to the number of flights. Hence, in practice, Abara's formulation can only solve the problems of limited size. In fact, this formulation was applied to a case study of American Airlines in 1990's, with 4 fleets to fly about 400 flights connecting 60 airports. Another mixed integer programming formulation for the fleet assignment problem is proposed by Hane et al. in [3]. The key idea in constructing the formulation in that paper is to see the flights as a time-expanded multi-commodity network and to view each assigned airplane type as a flow in this network. This formulation overcomes the drawback of Abara's one, since it can solve practical problems with up to 11 fleets, 2500 flights between 150 airports within 1 hour on a PC IBM RS/6000 Model 320. The formulation proposed in [3] is then used in many related research papers on the topic of fleet assignment. Hence, it becomes a basic formulation for the fleet assignment problem, and is often cited with the name "Basic Fleet Assignment Model" (BFAM for short). *We will focus our study on this model in Chapter 1 of this thesis.*

Due to the fact that delays and disruptions are unavoidable in airline operations, many research papers consider the fleet assignment problem in the context of data uncertainty. Their common approach is to construct a robust fleet assignment so that incurred cost will be reduced once operational delays or disruptions happen.

The paper of Rexing et al. [4] is the first one considering the fleet assignment problem under uncertainty. In this paper, the departure and arrival times of flights are subjected to interval uncertainty. The motivation of considering this problem comes from the observation that changing the flights' departure times can sometime reduce the total number of used airplanes, hence the total related cost is also reduced. The problem goal is to determine not only which airplane type flies which flight but also the departure time for each flight within its given time window, and the objective is to minimize the total number of used airplanes. To achieve that goal and objective, the authors construct a variant of BFAM in which the underlying idea is to discretize the given time window of departure time of each flight. By the

discretization, one obtains copies for each flight arc in the network, each copy corresponds to a discretized departure time. Exactly one among the copies of a flight arc is chosen to have a flow pass through, meaning that the flight will depart at the time corresponding to the chosen copy. Combining that idea with the construction of BFAM, the authors of [4] come up with a mixed integer programming formulation for their considered problem.

To construct a robust fleet assignment against the affection of flight delays and disruptions, in [5] the authors exploit a special structure of flight networks of airlines. Nowadays, most airlines design their flight networks in hub-and-spoke style. With this structure, each flight network has some hub stations together with some spoke ones. Hub stations are the main airports at that there are connections to most airports in the network. Spoke stations are the small airports at that there are only connections to hub stations, and a few connections to other spoke ones. The key idea in [5] is to partition the set of flights into hub-connection strings and spoke-connection strings. A hub-connection string means a string of flights that initiates at a hub station, passes several spoke ones, and then stops at some hub station. A spoke-connection string consists of a string of flights that connects only spoke stations. An upper bound on the number of flights in each hub-connection string as well as in each spoke-connection one is imposed so that the fleet assignment contains many cycles of few flights. Consequently, a disruption in a cycle of flights only has an impact on the cycle, thus the incurred cost is reduced. The partition of flights with constraints concerning the upper bound can be modeled as an integer programming, and hence can be done efficiently by using commercial solvers. In [6] a similar problem is studied. However, the authors there observe that the fewer airplanes are assigned to fly between spoke airports, the easier re-assignment can be done to replace canceled flights. Motivated from this observation, they impose an additional constraint to BFAM, which limits the number of airplanes used to fly between spoke airports.

All of the papers [4, 5, 6] mentioned above deal with data uncertainty in the fleet assignment problem, and they all focus on constructing such an assignment in sense of strict robustness. In means that, in those papers, a solution is made before any realization of data, delays, or disruptions. *In Chapter 2 of this thesis, we consider the fleet assignment problem under uncertainty in a different context, where we focus on finding a new fleet assignment after the realization of data.* More precisely, we are given a scheduled fleet assignment

constructed from a flight schedule and a set of airplanes. During operation in practice, delays and disruptions may happen and lead to an updated flight schedule, and sometime an updated set of available airplanes. The changes in input data may make the scheduled fleet assignment unusable, and therefore a new one needs to be constructed adapting the new circumstance. The cost of recovering from the scheduled fleet assignment to the new one, in some sense, should be minimized. We aim to construct such a new fleet assignment whose recovery cost is the number of differences with the scheduled one. By minimizing this number, we also reduce the cost incurred by related operations. Since we aim to find a new fleet assignment after the realization of the uncertain data, we call this problem *wait-and-see fleet assignment* for convenience. As a solution approach, we propose a mixed integer programming formulation for the problem. In our proposed formulation, the constraints of the problem are modeled similarly to BFAM, while the recovery cost in the objective function is the number of changes from the scheduled assignment to the new assignment solution.

As we have discussed, after this introduction part, the thesis contains two main chapters, each studies a specific problem of fleet assignment. Namely, in Chapter 1 we focus on the deterministic version of the fleet assignment problem and present in detail the construction of BFAM introduced in [3]. In Chapter 2 we study the wait-and-see fleet assignment problem and present our mixed integer programming formulation for the problem. We close this thesis with a conclusion part.

Our contributions in this thesis consist of the followings. In Chapter 1 we give a detail explanation for the construction of BFAM which is introduced in [3], provide a simple example for the model, and provide ZIMPL code as well as a numerical instance for implementing and experimenting the model. In Chapter 2 we provide ZIMPL code and some numerical instances for implementing and experimenting at the mixed integer programming formulation for the wait-and-see fleet assignment problem.

Chapter 1

Basic fleet assignment model

This chapter presents our study on the basic fleet assignment model (BFAM) introduced in [3]. The precise description of the fleet assignment problem is presented in Section 1.1. Section 1.2 gives the detail construction of BFAM. Section 1.3 presents our numerical experiments to evaluate the performance of this model.

1.1 Problem statement

Before each season, each airline often constructs a fleet assignment for a specific time period, and then use the assignment repeatedly in the whole season. For example, such an assignment can be scheduled for a sample week and then used for every week in the season. Roughly speaking, the fleet assignment problem is to determine *which airplane type* in the airline's fleets should be assigned to fly *which flight* in the sample time period. To construct such an assignment, one needs the input data not only about the flights in the sample time period but also about the airline's fleets. The input data about the flights include the following information.

- The set A of airports in the flight network of the airline. For each airport $a \in A$, one may take into account the maximum number s_a of airplanes that can stay in the airport at the same time.
- The set D of flight legs in the network of the airline. Each flight leg can be represented by an ordered pair of airport $(a, b) \in A \times A$, in which a is the departure airport and b is the arrival airport.
- The set L of flights in the considered time period. Each flight in L is an

ordered tuple $\ell = ((a, t_a), (b, t_b))$ in which $(a, b) \in D$, t_a is the departure time, and t_b is the arrival time of the flight.

- A list \mathcal{L} of *required throughs* (so-called *one-stop flights*). More precisely, each element in \mathcal{L} is an ordered pair $(\ell_1, \ell_2) \in L \times L$ in which the flight ℓ_1 is connected with its successive flight ℓ_2 (i.e., the arrival airport of ℓ_1 must be the departure airport of ℓ_2 , and these two flights must be served by the same airplane).

The input data about the airline's fleets include the following information.

- The set F of available fleets (i.e. airplane types) of the airline. Each fleet $f \in F$ consists of n_f airplanes of the same type.
- For each fleet $f \in F$, a set $D_f \subset D$ is given in advance. This set consists of the flight legs that can be served by airplanes in the fleet f . This information come from the fact the airplanes of different types have different passenger capacities, different flight ranges, ... and therefore each airplane type is suitable to serve some specific flight legs. For example, an airplane with short range cannot be assigned to serve flights of long distances.

Example 1.1. Assume that an airline has two fleets: one with Airbus 321 airplanes (denoted $A321$), and the other with Airbus 330 airplanes (denoted $A330$). The flight network of the airline has three airports with the corresponding IATA codes DAD, HAN, and HCM. In this case, the set of airports is $A = \{DAD, HAN, HCM\}$, and the set of fleets is $F = \{A321, A330\}$.

No	Departure airport	Arrival airport	Departure time	Arrival time	Airbus 321	Airbus 330
1	HAN	DAD	07h00	08h20	A321	
2	HAN	DAD	09h00	10h20	A321	
3	DAD	HAN	15h00	16h20	A321	
4	DAD	HCM	08h40	10h20	A321	
5	DAD	HCM	14h00	15h30	A321	
6	HCM	DAD	09h00	10h30	A321	
7	HCM	HAN	18h00	20h10	A321	A330
8	HAN	HCM	10h00	12h10	A321	A330

Table 1.1: The daily flight schedule of the airline and the compatibility of fleets with flights in Example 1.1.

The daily flight schedule and the compatibility of fleets with flights are given in Table 1.1. The first and the fourth flights are components of an

one-stop flight. The last two columns of the table tell us which flight can be served by which airplane type. It follows from this schedule that:

- the set of flight legs in the network is

$$D = \{(HAN, DAD), (DAD, HAN), (DAD, HCM), (HCM, DAD), (HAN, HCM), (HCM, HAN)\},$$

- the set of flight legs that can be served by fleet Airbus 321 is $D_{A321} = D$, while the set of flight legs that can be served by fleet Airbus 330 is

$$D_{A330} = \{(HAN, HCM), (HCM, HAN)\},$$

- the set of flights is

$$L = \{((HAN, 07h00), (DAD, 08h20)), ((HAN, 09h00), (DAD, 10h20)), ((DAD, 15h00), (HAN, 16h20)), ((DAD, 08h40), (HCM, 10h20)), ((DAD, 14h00), (HCM, 15h30)), ((HCM, 09h00), (DAD, 10h30)), ((HCM, 18h00), (HAN, 20h10)), ((HAN, 10h00), (HCM, 12h10))\},$$

- the set of require throughs consists of a single element

$$\mathcal{L} = \{(((HAN, 07h00), (DAD, 08h20)), ((DAD, 08h40), (HCM, 10h20)))\}.$$

□

One may be given the values of $c_{f\ell}$ and $r_{f\ell}$ (with $f \in F$ and $\ell \in L$) in which $c_{f\ell}$ is the cost of assigning an airplane of type f to serve a flight $\ell \in L$, and $r_{f\ell}$ is the revenue of such assignment. These values are often obtained from the past statistic of the airline. A valid fleet assignment must satisfy the following constraints.

- (C1) Each flight is served by exactly one airplane in some fleet.
- (C2) The number of used airplanes in each fleet does not exceed the fleet size.
- (C3) Right before and right after each flight event (take off or landing) in the considered time period, the number of airplanes in each fleet at each airport must be the same.
- (C4) The two concerning flights in each required through must be served by the same fleet.

The feasibility version of the fleet assignment problem asks whether such a valid solution exists. If this is possible, we may consider an optimization version of the fleet assignment problem, which asks to find a feasible assignment that optimizes one of the following objectives.

- (O1) Minimize the total cost of the assignment.
- (O2) Maximize the total revenue of the assignment.
- (O3) Minimize the number of used airplanes.

1.2 Basic fleet assignment model

The basic fleet assignment model (BFAM) is introduced in [3] to solve the fleet assignment problem under the deterministic setting of input data. The underlying idea in this model is to view the assigned flights in the fleet assignment solution as the flows in a time-expanded multi-commodity network. We recall the detail construction of the network in Section 1.2.1 before going to the description of BFAM in Section 1.2.2.

1.2.1 Time-expanded multi-commodity network

As presented in the previous section, a flight $\ell \in L$ is determined by its departure airport, arrival airport, departure time, and arrival time. To serve this flight, we need to assign with it an airplane from some fleet $f \in F$. Note that, after landing on the arrival airport of the flight, the airplane needs some additional time for operations such as cleaning, refueling, handling passenger baggages, or receiving more passengers in case of one-stop flight, etc. It means that the airplane needs the additional time after the arrival time of the flight before ready to take off for the next flight. This additional time depends on the airplane type and the airport since larger airplanes and busier airports require more time. We will use the term “ready time” to indicate the time at which the arriving flight is ready to take off. Hence, if we plan to assign an airplane from fleet f to a flight ℓ , then we replace the arrival time of the flight by the corresponding ready time.

It is worth noting that different flights on the same flight leg can be served by the airplanes of the same fleet. To distinguish such flights, in the time-expanded multi-commodity network, each pair of fleet-airport $(f, a) \in F \times A$ is assigned with a time line whose length represents the time period of the fleet

assignment result. Assume that an airplane of type $f \in F$ can be assigned to serve a flight $\ell = ((a, t_a), (b, t_b))$, which departs at time t_a from airport a and ready in the arrival airport b at time t_b . The departure event of this flight is represented by a node (f, a, t_a) corresponding to the departure time t_a on the time line (f, a) . Similarly, the arrival event of this flight is represented by a node (f, b, t_b) corresponding to the ready time t_b on the time line (f, b) . Then, the directed arc from the departure node (f, a, t_a) to the arrival node (f, b, t_b) represents the possibility of assigning an airplane of type f to serve the flight $\ell = ((a, t_a), (b, t_b))$. This arc transfers a flow unit if this possibility becomes true, otherwise no flow unit passes through the arc. We will use the term *flyable arcs* to indicate such arcs connecting the nodes on different time lines.

We need further constructions to complete the network. As we have defined above, each node on a time line represents either a possible departure event or a possible ready event of a flight. On each time line, each node is connected to its successive node by a directed arc. To be precise, consider (f, a, t_1) and (f, a, t_2) , in which $t_1 < t_2$, as two nodes corresponding to two consecutive events on the time line (f, a) . A directed arc is made from the former node to the latter one. The flow value of this arc represents the number of airplanes of type f that are in the ready status on the airport a during the time period from t_1 to t_2 . Furthermore, for each time line $(f, a) \in F \times A$, we connect its last node to its first node by a directed arc. The flow value of this arc equals the number of airplanes in fleet f that are ready on the airport a after the considered assignment time period. Since this arc makes the time line a cycle, the assignment solution can be used as a periodical schedule. We will use the term *ground arcs* to indicate the constructed arcs connecting the nodes on the same time line.

With the above network construction, a fleet assignment solution can be imagined as the circulation of flow units in the network. The flow conservation at every node in the network ensures the balance constraints (C3) and hence forces the airplanes to circulate through the network of flights. The following example illustrates the construction of the time-expanded multi-commodity network for the data instance given in Example 1.1.

Example 1.2. Consider the data instance given in Example 1.1. Assume that, except for the one-stop flight that needs 20 minutes in between, any flight served by Airbus 321 fleet (resp., Airbus 330 fleet) needs 30 minutes (resp., 40 minutes) to be ready after their arrivals. The last column of Table

1.2 gives us the precise ready time after each flight in the flight schedule if it is served by fleet Airbus 321. Since the flights on the legs (HAN, HCM) and (HCM, HAN) can be served by fleet Airbus 330, Table 1.3 gives us the ready time after each flight on these legs if it is served by this fleet.

No	Departure airport	Arrival airport	Departure time	Arrival time	Ready time
1	HAN	DAD	07h00	08h20	08h40
2	HAN	DAD	09h00	10h20	10h50
3	DAD	HAN	15h00	16h20	16h50
4	DAD	HCM	08h40	10h20	10h50
5	DAD	HCM	14h00	15h30	16h00
6	HCM	DAD	09h00	10h30	11h00
7	HCM	HAN	18h00	20h10	20h40
8	HAN	HCM	10h00	12h10	12h40

Table 1.2: The daily flight schedule of the airline in Example 1.1 with ready time in case of using fleet Airbus 321.

No	Departure airport	Arrival airport	Departure time	Arrival time	Ready time
1	HCM	HAN	18h00	20h10	20h50
2	HAN	HCM	10h00	12h10	12h50

Table 1.3: The daily flight schedule of the airline in Example 1.1 with ready time in case of using fleet Airbus 330.

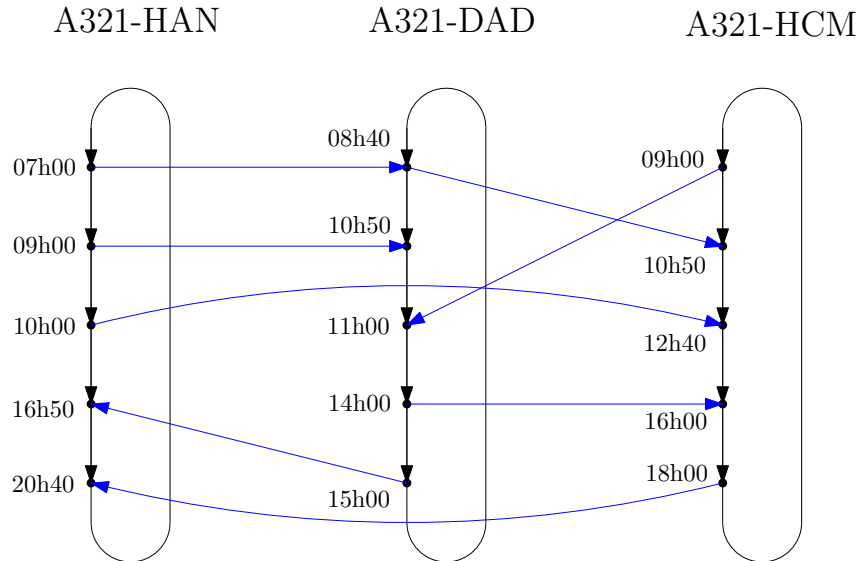


Figure 1.1: The component of the time-expanded multi-commodity network corresponding to the fleet Airbus 321.

We first construct the time lines concerning the fleet Airbus 321. These time lines correspond to airports DAD, HAN, HCM and respectively denoted A321-DAD, A321-HAN, A321-HCM. On each of these time lines, we put the nodes corresponding to the departure events and ready events of related flights. For instance, node A321-HAN-07h00 corresponds to the departure event of the first flight in the schedule if it is served by fleet Airbus 321. Each flight, if served by fleet Airbus 321, is represented by a directed arc connecting its departure node and its ready node. For example, the second flight in Table 1.2 is represented by a directed arc from node A321-HAN-09h00 (on the time line A321-HAN) to node A321-DAD-10h50 (on the time line A321-DAD). On each of the time lines we are considering, each node is connected to its successive node by a directed arc, and the last node is connected to the first node also by a directed arc. Figure 1.1 illustrates the component of the time-expanded multi-commodity network corresponding to the fleet Airbus 321. Similarly, we obtain the component corresponding to the fleet Airbus 330 as illustrated in Figure 1.2. The time-expanded multi-commodity network consists of these two components. \square

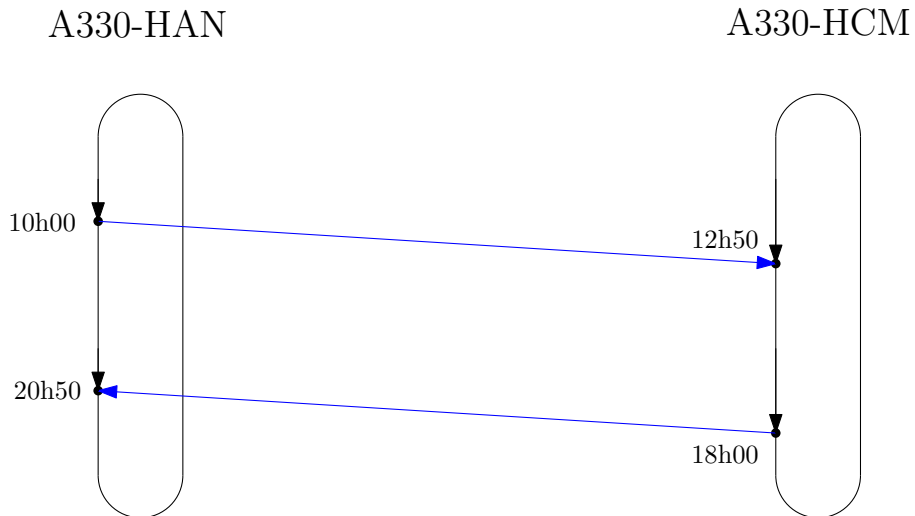


Figure 1.2: The component of the time-expanded multi-commodity network corresponding to the fleet Airbus 330.

1.2.2 Model description

A fleet assignment is a solution of assigning which fleet to serve which flight so that constraints (C1)-(C4) are satisfied. We can imagine the assignment of an airplane of fleet f to serve a flight ℓ as a flow unit transferred through the

corresponding arc in the time-expanded multi-commodity network. In this manner, the whole fleet assignment solution can be viewed as the distribution and transfer of flow units in the network, regarding constraints (C1)-(C4). BFAM formulates the problem of determining such flows as a mixed integer programming formulation. In this subsection we describe this formulation in detail.

Let N be the set of nodes in the network. For each $f \in F$, let N_f^* be the set of the last nodes on the time lines assigned with the fleet f . The set of arcs in the network is partitioned into two disjoint subsets: ground arcs and flyable arcs. Following the construction of the network, a ground arc is the one whose nodes are on the same time line, while a flyable arc connects nodes on different time lines. Each flyable arc has the form $((f, a, t_a), (f, b, t_b)) \in N \times N$ which corresponds to the possibility of assigning an airplane of fleet f to serve the flight from airport a with departure time t_a to airport b with ready time t_b . A flow unit transferred through this flyable arc means that the assignment is done. The flow value transferred through each ground arc on a time line (f, a) refers to the number of airplanes of fleet f in the ready status on airport a between the time of the events corresponding to its nodes. Let L^p be the set of flyable arcs in the network, and L^r the set of arcs in required throughs (one-stop flights).

Example 1.3. For the time-expanded multi-commodity network in Example 1.2, the set N^* of the last nodes in the time lines is

$$N^* = N_{A321} \cup N_{A330},$$

in which

$$\begin{aligned} N_{A321} &= \{(A321, HAN, 20h40), (A321, DAD, 15h00), (A321, HCM, 18h00)\}, \\ N_{A330} &= \{(A330, HAN, 20h50), (A330, HCM, 18h00)\}. \end{aligned}$$

The flyable arcs in the network are the ones of blue color in Figure 1.1 and Figure 1.2, while the ground arcs are black ones. The set of arcs concerning the required through in the network is

$$\begin{aligned} L^r &= \{(((A321, HAN, 07h00), (A321, HAN, 08h40)), \\ &\quad ((A321, HAN, 08h40), (A321, HCM, 10h50)))\}. \end{aligned}$$

□

Following the described manner, BFAM uses the following variables:

$$x_{fai,fbj} := \begin{cases} 1 & \text{if a flow unit transfers through arc } ((f, a, i), (f, b, j)), \\ 0 & \text{otherwise,} \end{cases} \quad (1.1)$$

$$y_{fai}^- := \text{flow value of the ground arc coming to node } (f, a, i) \in N, \quad (1.2)$$

$$y_{fai}^+ := \text{flow value of the ground arc going out of node } (f, a, i) \in N. \quad (1.3)$$

Using these variables, constraints (C1) can be formulated as follows.

$$\sum_{f \in F} x_{fai,fbj'} = 1 \quad \forall ((a, i), (b, j)) \in L. \quad (1.4)$$

Here (f, b, j') is the ready node corresponding to the flight $((a, i), (b, j)) \in L$ in case this flight is served by fleet f . This ensures that each flight is assigned to be served by exactly one fleet. Constraints (C2) can be modeled by

$$\sum_{(f,a,i) \in N_f^*} y_{fai}^+ \leq n_f \quad \forall f \in F. \quad (1.5)$$

Here, the left hand side equals the total number of airplanes in fleet f on all airports at the end of the scheduled time period. This value is also equal to the number of airplanes in fleet f . Recall that n_f is the number of available airplanes in fleet f . Hence, (1.5) ensure that, for each fleet, the number of used airplanes does not exceed the fleet size.

Constraints (C3) ensure the flow conservation at every node of the network. They can be formulated as follows.

$$\sum_{a,i} x_{fai,fbj} + y_{fbj}^- = \sum_{a,i} x_{fbj,faai} + y_{fbj}^+ \quad \forall (f, b, j) \in N. \quad (1.6)$$

The left hand side of the above equality is the total flow units coming to the node $(f, b, j) \in N$, i.e., the total number of airplanes of type f at airport b right before time j . The right hand side of the above equality is the total flow units going out of the node $(f, b, j) \in N$, i.e., the total number of airplanes of type f at airport b right after the time j . Therefore, the equality (1.6) ensures the flow conservation at node (f, b, j) .

Constraints (C4) impose that the component flights in each required through muse be served by the same fleet. Each required through is represented by an ordered pair of flyable arcs $((f, a, i), (f, b, j))$ and $((f, b, j), (f, c, k))$ in the time-expanded multi-commodity network, in which f is any fleet that can be

serve the required through, $((a, i), (b, j))$ is the first flight and $((b, j), (c, k))$ is the second flight of the required through. Hence, constraints (C4) can be represented as follows.

$$x_{fai,fbj} = x_{fbj,fck} \quad \forall(((f, a, i), (f, b, j)), ((f, b, j), (f, c, k))) \in L^r. \quad (1.7)$$

The original version of BFAM considers the objective (O1) which aims to minimize the total cost of the fleet assignment. For this objective, each flyable arc $((f, a, i), (f, b, j))$ in the network has a cost $c_{fai,fbj}$ representing the expected cost to serve the flight $((a, i), (b, j))$ by fleet f . This objective can be expressed by

$$\min \sum_{((f,a,i),(f,b,j)) \in L^p} c_{fai,fbj} x_{fai,fbj}. \quad (1.8)$$

Here we recall that L^p is the set of flyable arcs in the network. For a recap, BFAM reads as follows.

$$\begin{aligned} (BFAM) \quad & \min \sum_{((f,a,i),(f,b,j)) \in L^p} c_{fai,fbj} x_{fai,fbj} \\ \text{subject to} \quad & \sum_{f \in F} x_{fai,fbj'} = 1 && \forall((a, i), (b, j)) \in L \\ & \sum_{(f,a,i) \in N_f^*} y_{fai}^+ \leq n_f && \forall f \in F \\ & \sum_{a,i} x_{fai,fbj} + y_{fbj}^- = \sum_{a,i} x_{fbj,fai} + y_{fbj}^+ && \forall(f, b, j) \in N \\ & x_{fai,fbj} = x_{fbj,fck} \\ & \forall(((f, a, i), (f, b, j)), ((f, b, j), (f, c, k))) \in L^r \\ & x_{fai,fbj} \in \{0, 1\} && \forall((f, a, i), (f, b, j)) \in L^p \end{aligned} \quad (1.9)$$

$$y_{fbj}^+ \geq 0 \quad \forall(f, b, j) \in N \quad (1.10)$$

$$y_{fbj}^- \geq 0 \quad \forall(f, b, j) \in N. \quad (1.11)$$

Note that in (BFAM) the y -variables are nonnegative real numbers. However, the constrains (1.6) together with x -variable domain (1.9) guarantee that the y -variables are nonnegative integers.

By some slight modifications on (BFAM), we can model the objectives (O2) and (O3). Indeed, the objective (O2), which aims to maximize the total

expected revenue of the fleet assignment, can be modeled as

$$\max \sum_{(f,a,i),(f,b,j) \in L^p} r_{fai,fbj} x_{fai,fbj}. \quad (1.12)$$

Here we recall that $r_{fai,fbj}$ is the expected revenue of the flight $((a, i), (b, j))$ if it is served by fleet f . The objective (O3), which aims to minimize the number of used airplanes, can be modeled as

$$\min \sum_{f \in F} \sum_{(f,a,i) \in N^*} y_{fai}^+. \quad (1.13)$$

The inner sum in (1.13) is the total flow values of all arcs going out of the last nodes in the time lines associated with a fleet f . By the meaning of variables y^+ , this sum is nothing but the total airplanes in fleet f that are ready after the last events of the considered time period. Thus, the objective value in (1.13) is the number of used airplanes in all fleets.

1.3 Numerical experiments

To see how BFAM performs, we created an instance of the fleet assignment problem that consisting of 75 flights in one day between 3 airports with 3 fleets. The data of the problem instance are saved in 5 separated excel files as follows.

- *Airports.xlsx*. This file contains information about the airports in the problem instance. Except for the header line, each line of the file includes the following information: airport name, IATA code of the airport, capacity of the airport (i.e., the number of airplanes that can stay in the airport at the same time).
- *FleetComponent.xlsx*. This file contains information about the fleets in the problem instance. Except for the header line, each line of the file includes the following information: name of fleet (or airplane type), cardinality of the fleet (i.e., the number of airplanes in the fleet).
- *FlightLegs.xlsx*. This file contains information about the flight legs in the flight network of the problem instance. Except for the header line, each line of the file includes the following information: departure airport, arrival airport, IATA code of the flight leg.

- *Flights.xlsx*. This file contains information about the flights of the problem instance. Except for the header line, each line of the file includes the following information: flight code, IATA code of the flight leg, departure data and departure time, flight duration. The departure date is given as an integer which is the order of the date if we start counting from the first day of the considered time period.
- *AssignmentData.xlsx*. This file contains information about assigning each fleet to each flight of the problem instance. Except for the header line, each line of the file includes the information of a flight as described in the file *Flights.xlsx*, an airplane type that can be assigned to that flight, duration of the short maintenance after the flight, the cost and the revenue of the assignment. The values of assignment costs and expected revenues are randomly generated.

We implemented the (*BFAM*) formulations corresponding to objectives (O1)-(O3) by using ZIMPL 3.5.3 (see [7]), and then used GUROBI 9.1 (see <https://www.gurobi.com/>) as a mixed integer programming solver. For the use of our ZIMPL code, we need to save the input data in the excel files of the data sets in text files. Namely, the information in the excel files mentioned above are respectively saved in text files *Airports.txt*, *FleetComponent.txt*, *Flights.txt*, and *AssignmentData.txt*. Furthermore, from the input data about the arrival time of each flight and the length of maintenance duration after each flight, we computed the ready time of each flight and then add the information about the ready time in a data column of file *AssignmentData.txt*. Additionally, to establish the ground arcs in the (*BFAM*) formulations, we saved the information about the departure and ready times of all flyable arcs into a text file named *TimelineEvents.txt*. The information in this file are arranged in lexicographical order of timeline first, then the time corresponding to each node on these timelines. The excel files of the tested instance are available on

<https://github.com/lxthanh86/FleetAssignment>.

The corresponding text files are respectively given below.

File *Airports.txt* of the tested instance.

```

1 # File data of airports
2 # IATAcode      Capacity
3           DAD           22
4           HAN           47
5           SGN           104

```

File FleetComponent.txt of the tested instance.

```

1 # File data of fleet component
2 # TypeDenote      NumberOfAircrafts
3           01           52
4           02           15
5           03           14

```

File Flights.txt of the tested instance.

```

1 # File data of flights
2 # DepartureAirport  ArrivalAirport  DepartureDate  DepartureTime
3           HAN           DAD      001_06h00
4           HAN           DAD      001_09h00
5           HAN           DAD      001_12h00
6           HAN           DAD      001_13h00
7           HAN           DAD      001_16h00
8           HAN           DAD      001_17h00
9           HAN           DAD      001_18h00
10          HAN           DAD      001_20h00
11          HAN           SGN      001_06h00
12          HAN           SGN      001_07h00
13          HAN           SGN      001_08h00
14          HAN           SGN      001_09h00
15          HAN           SGN      001_10h00
16          HAN           SGN      001_11h00
17          HAN           SGN      001_12h00
18          HAN           SGN      001_13h00
19          HAN           SGN      001_14h00
20          HAN           SGN      001_15h00
21          HAN           SGN      001_16h00
22          HAN           SGN      001_17h00
23          HAN           SGN      001_18h00
24          HAN           SGN      001_19h00
25          HAN           SGN      001_20h00
26          HAN           SGN      001_21h00
27          HAN           SGN      001_22h00
28          DAD           HAN      001_06h00
29          DAD           HAN      001_08h00
30          DAD           HAN      001_11h00
31          DAD           HAN      001_14h00
32          DAD           HAN      001_15h00
33          DAD           HAN      001_18h00
34          DAD           HAN      001_19h00
35          DAD           HAN      001_20h00
36          DAD           SGN      001_08h00
37          DAD           SGN      001_09h00
38          DAD           SGN      001_10h00
39          DAD           SGN      001_11h00
40          DAD           SGN      001_12h00
41          DAD           SGN      001_13h00
42          DAD           SGN      001_14h00
43          DAD           SGN      001_15h00
44          DAD           SGN      001_17h00

```

45	DAD	SGN	001_18h00
46	DAD	SGN	001_19h00
47	DAD	SGN	001_20h00
48	SGN	HAN	001_06h00
49	SGN	HAN	001_07h00
50	SGN	HAN	001_08h00
51	SGN	HAN	001_09h00
52	SGN	HAN	001_10h00
53	SGN	HAN	001_11h00
54	SGN	HAN	001_12h00
55	SGN	HAN	001_13h00
56	SGN	HAN	001_14h00
57	SGN	HAN	001_15h00
58	SGN	HAN	001_16h00
59	SGN	HAN	001_17h00
60	SGN	HAN	001_18h00
61	SGN	HAN	001_19h00
62	SGN	HAN	001_20h00
63	SGN	HAN	001_21h00
64	SGN	HAN	001_22h00
65	SGN	DAD	001_06h00
66	SGN	DAD	001_09h00
67	SGN	DAD	001_10h00
68	SGN	DAD	001_11h00
69	SGN	DAD	001_12h00
70	SGN	DAD	001_13h00
71	SGN	DAD	001_14h00
72	SGN	DAD	001_15h00
73	SGN	DAD	001_16h00
74	SGN	DAD	001_17h00
75	SGN	DAD	001_18h00
76	SGN	DAD	001_20h00

File Flights.txt of the tested instance.

1	# File data all related information about flights and assignment options							
2	# 1.DEP_AIR 2.DEP_DATE 3.DEP_TIME 4.ARR_AIR 5.ARR_DATE 6.ARR_TIME							
	7.AIR_TYPE		8.READY_DATE		9.READY_TIME		10.COST	11.REVENUE
3	HAN	001_06h00	DAD	001_07h20	01	001_08h05	8233	13468
4	HAN	001_09h00	DAD	001_10h20	01	001_11h05	4084	16650
5	HAN	001_12h00	DAD	001_13h20	01	001_14h05	9176	17067
6	HAN	001_13h00	DAD	001_14h20	01	001_15h05	9698	11096
7	HAN	001_16h00	DAD	001_17h20	01	001_18h05	6812	10174
8	HAN	001_17h00	DAD	001_18h20	01	001_19h05	8979	17545
9	HAN	001_18h00	DAD	001_19h20	01	001_20h05	8171	19721
10	HAN	001_20h00	DAD	001_21h20	01	001_22h05	8057	6219
11	HAN	001_06h00	SGN	001_08h15	01	001_09h00	5707	6639
12	HAN	001_07h00	SGN	001_09h15	01	001_10h00	9879	11619
13	HAN	001_08h00	SGN	001_10h15	01	001_11h00	6460	11779
14	HAN	001_09h00	SGN	001_11h15	01	001_12h00	5116	14167
15	HAN	001_10h00	SGN	001_12h15	01	001_13h00	8712	11305
16	HAN	001_11h00	SGN	001_13h15	01	001_14h00	6571	7371
17	HAN	001_12h00	SGN	001_14h15	01	001_15h00	4135	13607

18	HAN	001_13h00	SGN	001_15h15	01	001_16h00	7082	12482
19	HAN	001_14h00	SGN	001_16h15	01	001_17h00	5463	19708
20	HAN	001_15h00	SGN	001_17h15	01	001_18h00	4934	12642
21	HAN	001_16h00	SGN	001_18h15	01	001_19h00	9632	15163
22	HAN	001_17h00	SGN	001_19h15	01	001_20h00	6758	16552
23	HAN	001_18h00	SGN	001_20h15	01	001_21h00	4443	7476
24	HAN	001_19h00	SGN	001_21h15	01	001_22h00	7942	13616
25	HAN	001_20h00	SGN	001_22h15	01	001_23h00	6725	10998
26	HAN	001_21h00	SGN	001_23h15	01	002_00h00	4537	16608
27	HAN	001_22h00	SGN	002_00h15	01	002_01h00	5978	7336
28	DAD	001_06h00	HAN	001_07h20	01	001_08h05	6663	9821
29	DAD	001_08h00	HAN	001_09h20	01	001_10h05	5540	7258
30	DAD	001_11h00	HAN	001_12h20	01	001_13h05	5411	12727
31	DAD	001_14h00	HAN	001_15h20	01	001_16h05	5236	18104
32	DAD	001_15h00	HAN	001_16h20	01	001_17h05	7258	7129
33	DAD	001_18h00	HAN	001_19h20	01	001_20h05	9541	14683
34	DAD	001_19h00	HAN	001_20h20	01	001_21h05	9963	7826
35	DAD	001_20h00	HAN	001_21h20	01	001_22h05	7230	11690
36	DAD	001_08h00	SGN	001_09h30	01	001_10h15	9981	10757
37	DAD	001_09h00	SGN	001_10h30	01	001_11h15	6537	13605
38	DAD	001_10h00	SGN	001_11h30	01	001_12h15	5366	14669
39	DAD	001_11h00	SGN	001_12h30	01	001_13h15	5815	10100
40	DAD	001_12h00	SGN	001_13h30	01	001_14h15	6181	18264
41	DAD	001_13h00	SGN	001_14h30	01	001_15h15	7684	16950
42	DAD	001_14h00	SGN	001_15h30	01	001_16h15	4395	6861
43	DAD	001_15h00	SGN	001_16h30	01	001_17h15	4692	8433
44	DAD	001_17h00	SGN	001_18h30	01	001_19h15	5300	12552
45	DAD	001_18h00	SGN	001_19h30	01	001_20h15	8476	7239
46	DAD	001_19h00	SGN	001_20h30	01	001_21h15	6412	9854
47	DAD	001_20h00	SGN	001_21h30	01	001_22h15	8407	9880
48	SGN	001_06h00	HAN	001_08h10	01	001_08h55	7876	10870
49	SGN	001_07h00	HAN	001_09h10	01	001_09h55	9756	13588
50	SGN	001_08h00	HAN	001_10h10	01	001_10h55	6375	9941
51	SGN	001_09h00	HAN	001_11h10	01	001_11h55	7345	18729
52	SGN	001_10h00	HAN	001_12h10	01	001_12h55	8537	15807
53	SGN	001_11h00	HAN	001_13h10	01	001_13h55	8707	6703
54	SGN	001_12h00	HAN	001_14h10	01	001_14h55	9837	17260
55	SGN	001_13h00	HAN	001_15h10	01	001_15h55	4738	19359
56	SGN	001_14h00	HAN	001_16h10	01	001_16h55	5006	8299
57	SGN	001_15h00	HAN	001_17h10	01	001_17h55	7858	17888
58	SGN	001_16h00	HAN	001_18h10	01	001_18h55	5941	16792
59	SGN	001_17h00	HAN	001_19h10	01	001_19h55	7663	11242
60	SGN	001_18h00	HAN	001_20h10	01	001_20h55	7103	10624
61	SGN	001_19h00	HAN	001_21h10	01	001_21h55	6082	6035
62	SGN	001_20h00	HAN	001_22h10	01	001_22h55	6429	17362
63	SGN	001_21h00	HAN	001_23h10	01	001_23h55	6036	15948
64	SGN	001_22h00	HAN	002_00h10	01	002_00h55	9738	9404
65	SGN	001_06h00	DAD	001_07h20	01	001_08h05	7593	18636
66	SGN	001_09h00	DAD	001_10h20	01	001_11h05	6633	16640
67	SGN	001_10h00	DAD	001_11h20	01	001_12h05	5622	14137
68	SGN	001_11h00	DAD	001_12h20	01	001_13h05	7914	18605
69	SGN	001_12h00	DAD	001_13h20	01	001_14h05	6952	16773
70	SGN	001_13h00	DAD	001_14h20	01	001_15h05	6479	8137
71	SGN	001_14h00	DAD	001_15h20	01	001_16h05	7027	8563

72	SGN	001_15h00	DAD	001_16h20	01	001_17h05	5732	16536
73	SGN	001_16h00	DAD	001_17h20	01	001_18h05	4061	8123
74	SGN	001_17h00	DAD	001_18h20	01	001_19h05	7489	16212
75	SGN	001_18h00	DAD	001_19h20	01	001_20h05	8701	17076
76	SGN	001_20h00	DAD	001_21h20	01	001_22h05	7384	8850
77	HAN	001_06h00	DAD	001_07h20	02	001_08h05	7477	10054
78	HAN	001_09h00	DAD	001_10h20	02	001_11h05	8887	15927
79	HAN	001_12h00	DAD	001_13h20	02	001_14h05	6241	19468
80	HAN	001_13h00	DAD	001_14h20	02	001_15h05	7149	16740
81	HAN	001_16h00	DAD	001_17h20	02	001_18h05	7736	15070
82	HAN	001_17h00	DAD	001_18h20	02	001_19h05	7535	19806
83	HAN	001_18h00	DAD	001_19h20	02	001_20h05	5463	13474
84	HAN	001_20h00	DAD	001_21h20	02	001_22h05	7451	7400
85	HAN	001_06h00	SGN	001_08h15	02	001_09h00	5774	11348
86	HAN	001_07h00	SGN	001_09h15	02	001_10h00	5669	8246
87	HAN	001_08h00	SGN	001_10h15	02	001_11h00	8277	10567
88	HAN	001_09h00	SGN	001_11h15	02	001_12h00	4484	12412
89	HAN	001_10h00	SGN	001_12h15	02	001_13h00	5738	18872
90	HAN	001_11h00	SGN	001_13h15	02	001_14h00	7366	15723
91	HAN	001_12h00	SGN	001_14h15	02	001_15h00	9497	12024
92	HAN	001_13h00	SGN	001_15h15	02	001_16h00	6121	11668
93	HAN	001_14h00	SGN	001_16h15	02	001_17h00	4365	11464
94	HAN	001_15h00	SGN	001_17h15	02	001_18h00	5543	14803
95	HAN	001_16h00	SGN	001_18h15	02	001_19h00	7037	11466
96	HAN	001_17h00	SGN	001_19h15	02	001_20h00	7577	17659
97	HAN	001_18h00	SGN	001_20h15	02	001_21h00	5990	7795
98	HAN	001_19h00	SGN	001_21h15	02	001_22h00	8965	7146
99	HAN	001_20h00	SGN	001_22h15	02	001_23h00	4900	15862
100	HAN	001_21h00	SGN	001_23h15	02	002_00h00	6411	12466
101	HAN	001_22h00	SGN	002_00h15	02	002_01h00	7539	8378
102	DAD	001_06h00	HAN	001_07h20	02	001_08h05	9236	16510
103	DAD	001_08h00	HAN	001_09h20	02	001_10h05	4185	10518
104	DAD	001_11h00	HAN	001_12h20	02	001_13h05	5527	10768
105	DAD	001_14h00	HAN	001_15h20	02	001_16h05	7532	16569
106	DAD	001_15h00	HAN	001_16h20	02	001_17h05	7806	11740
107	DAD	001_18h00	HAN	001_19h20	02	001_20h05	6086	8089
108	DAD	001_19h00	HAN	001_20h20	02	001_21h05	4173	10835
109	DAD	001_20h00	HAN	001_21h20	02	001_22h05	9084	17567
110	DAD	001_08h00	SGN	001_09h30	02	001_10h15	6971	11781
111	DAD	001_09h00	SGN	001_10h30	02	001_11h15	8888	13573
112	DAD	001_10h00	SGN	001_11h30	02	001_12h15	6939	15532
113	DAD	001_11h00	SGN	001_12h30	02	001_13h15	4902	13418
114	DAD	001_12h00	SGN	001_13h30	02	001_14h15	6868	8669
115	DAD	001_13h00	SGN	001_14h30	02	001_15h15	4970	17309
116	DAD	001_14h00	SGN	001_15h30	02	001_16h15	8759	11314
117	DAD	001_15h00	SGN	001_16h30	02	001_17h15	4288	16008
118	DAD	001_17h00	SGN	001_18h30	02	001_19h15	8478	16533
119	DAD	001_18h00	SGN	001_19h30	02	001_20h15	7808	15983
120	DAD	001_19h00	SGN	001_20h30	02	001_21h15	9913	17237
121	DAD	001_20h00	SGN	001_21h30	02	001_22h15	6139	12068
122	SGN	001_06h00	HAN	001_08h10	02	001_08h55	4626	8595
123	SGN	001_07h00	HAN	001_09h10	02	001_09h55	6966	19622
124	SGN	001_08h00	HAN	001_10h10	02	001_10h55	7020	7942
125	SGN	001_09h00	HAN	001_11h10	02	001_11h55	7944	12176

126	SGN	001_10h00	HAN	001_12h10	02	001_12h55	6982	8176
127	SGN	001_11h00	HAN	001_13h10	02	001_13h55	7108	16599
128	SGN	001_12h00	HAN	001_14h10	02	001_14h55	8049	18672
129	SGN	001_13h00	HAN	001_15h10	02	001_15h55	8784	15748
130	SGN	001_14h00	HAN	001_16h10	02	001_16h55	7058	11685
131	SGN	001_15h00	HAN	001_17h10	02	001_17h55	6988	8629
132	SGN	001_16h00	HAN	001_18h10	02	001_18h55	5308	12258
133	SGN	001_17h00	HAN	001_19h10	02	001_19h55	6337	18049
134	SGN	001_18h00	HAN	001_20h10	02	001_20h55	9211	9627
135	SGN	001_19h00	HAN	001_21h10	02	001_21h55	8454	17773
136	SGN	001_20h00	HAN	001_22h10	02	001_22h55	8451	12127
137	SGN	001_21h00	HAN	001_23h10	02	001_23h55	5874	17183
138	SGN	001_22h00	HAN	002_00h10	02	002_00h55	9640	7600
139	SGN	001_06h00	DAD	001_07h20	02	001_08h05	7449	9432
140	SGN	001_09h00	DAD	001_10h20	02	001_11h05	5474	11300
141	SGN	001_10h00	DAD	001_11h20	02	001_12h05	5258	7102
142	SGN	001_11h00	DAD	001_12h20	02	001_13h05	5389	19299
143	SGN	001_12h00	DAD	001_13h20	02	001_14h05	9010	11356
144	SGN	001_13h00	DAD	001_14h20	02	001_15h05	7719	7391
145	SGN	001_14h00	DAD	001_15h20	02	001_16h05	9949	12599
146	SGN	001_15h00	DAD	001_16h20	02	001_17h05	9348	17507
147	SGN	001_16h00	DAD	001_17h20	02	001_18h05	6264	8634
148	SGN	001_17h00	DAD	001_18h20	02	001_19h05	4870	9609
149	SGN	001_18h00	DAD	001_19h20	02	001_20h05	5996	12313
150	SGN	001_20h00	DAD	001_21h20	02	001_22h05	5245	11105
151	HAN	001_06h00	DAD	001_07h20	03	001_08h05	5811	16847
152	HAN	001_09h00	DAD	001_10h20	03	001_11h05	4272	11796
153	HAN	001_12h00	DAD	001_13h20	03	001_14h05	9229	6787
154	HAN	001_13h00	DAD	001_14h20	03	001_15h05	4321	14295
155	HAN	001_16h00	DAD	001_17h20	03	001_18h05	5583	9911
156	HAN	001_17h00	DAD	001_18h20	03	001_19h05	9466	9176
157	HAN	001_18h00	DAD	001_19h20	03	001_20h05	4638	19992
158	HAN	001_20h00	DAD	001_21h20	03	001_22h05	4618	17185
159	HAN	001_06h00	SGN	001_08h15	03	001_09h00	5806	19280
160	HAN	001_07h00	SGN	001_09h15	03	001_10h00	4977	15052
161	HAN	001_08h00	SGN	001_10h15	03	001_11h00	7799	8906
162	HAN	001_09h00	SGN	001_11h15	03	001_12h00	9435	9659
163	HAN	001_10h00	SGN	001_12h15	03	001_13h00	7791	14787
164	HAN	001_11h00	SGN	001_13h15	03	001_14h00	9483	17688
165	HAN	001_12h00	SGN	001_14h15	03	001_15h00	8068	13034
166	HAN	001_13h00	SGN	001_15h15	03	001_16h00	5618	6778
167	HAN	001_14h00	SGN	001_16h15	03	001_17h00	6190	12859
168	HAN	001_15h00	SGN	001_17h15	03	001_18h00	7252	8188
169	HAN	001_16h00	SGN	001_18h15	03	001_19h00	4644	16976
170	HAN	001_17h00	SGN	001_19h15	03	001_20h00	4112	8945
171	HAN	001_18h00	SGN	001_20h15	03	001_21h00	4001	13515
172	HAN	001_19h00	SGN	001_21h15	03	001_22h00	5151	15505
173	HAN	001_20h00	SGN	001_22h15	03	001_23h00	9573	13423
174	HAN	001_21h00	SGN	001_23h15	03	002_00h00	6953	8906
175	HAN	001_22h00	SGN	002_00h15	03	002_01h00	9566	7371
176	DAD	001_06h00	HAN	001_07h20	03	001_08h05	5637	15431
177	DAD	001_08h00	HAN	001_09h20	03	001_10h05	8741	10161
178	DAD	001_11h00	HAN	001_12h20	03	001_13h05	4269	12754
179	DAD	001_14h00	HAN	001_15h20	03	001_16h05	9568	10634

180	DAD	001_15h00	HAN	001_16h20	03	001_17h05	9763	7604
181	DAD	001_18h00	HAN	001_19h20	03	001_20h05	6880	9071
182	DAD	001_19h00	HAN	001_20h20	03	001_21h05	7286	18922
183	DAD	001_20h00	HAN	001_21h20	03	001_22h05	8035	16107
184	DAD	001_08h00	SGN	001_09h30	03	001_10h15	8172	8507
185	DAD	001_09h00	SGN	001_10h30	03	001_11h15	6565	13127
186	DAD	001_10h00	SGN	001_11h30	03	001_12h15	9320	11187
187	DAD	001_11h00	SGN	001_12h30	03	001_13h15	5339	14183
188	DAD	001_12h00	SGN	001_13h30	03	001_14h15	8105	16464
189	DAD	001_13h00	SGN	001_14h30	03	001_15h15	5215	19395
190	DAD	001_14h00	SGN	001_15h30	03	001_16h15	6781	7673
191	DAD	001_15h00	SGN	001_16h30	03	001_17h15	7198	13854
192	DAD	001_17h00	SGN	001_18h30	03	001_19h15	6393	18644
193	DAD	001_18h00	SGN	001_19h30	03	001_20h15	4094	12036
194	DAD	001_19h00	SGN	001_20h30	03	001_21h15	8177	11846
195	DAD	001_20h00	SGN	001_21h30	03	001_22h15	9670	7701
196	SGN	001_06h00	HAN	001_08h10	03	001_08h55	4466	12059
197	SGN	001_07h00	HAN	001_09h10	03	001_09h55	5307	11306
198	SGN	001_08h00	HAN	001_10h10	03	001_10h55	7104	19516
199	SGN	001_09h00	HAN	001_11h10	03	001_11h55	8158	6902
200	SGN	001_10h00	HAN	001_12h10	03	001_12h55	5342	10566
201	SGN	001_11h00	HAN	001_13h10	03	001_13h55	8804	10553
202	SGN	001_12h00	HAN	001_14h10	03	001_14h55	9255	11833
203	SGN	001_13h00	HAN	001_15h10	03	001_15h55	6410	6228
204	SGN	001_14h00	HAN	001_16h10	03	001_16h55	4636	9866
205	SGN	001_15h00	HAN	001_17h10	03	001_17h55	9380	11219
206	SGN	001_16h00	HAN	001_18h10	03	001_18h55	5416	18298
207	SGN	001_17h00	HAN	001_19h10	03	001_19h55	7516	19042
208	SGN	001_18h00	HAN	001_20h10	03	001_20h55	5557	8501
209	SGN	001_19h00	HAN	001_21h10	03	001_21h55	5671	15842
210	SGN	001_20h00	HAN	001_22h10	03	001_22h55	4467	11749
211	SGN	001_21h00	HAN	001_23h10	03	001_23h55	4910	14302
212	SGN	001_22h00	HAN	002_00h10	03	002_00h55	9906	14862
213	SGN	001_06h00	DAD	001_07h20	03	001_08h05	9161	7049
214	SGN	001_09h00	DAD	001_10h20	03	001_11h05	6383	13372
215	SGN	001_10h00	DAD	001_11h20	03	001_12h05	9373	7560
216	SGN	001_11h00	DAD	001_12h20	03	001_13h05	9077	12177
217	SGN	001_12h00	DAD	001_13h20	03	001_14h05	5173	10566
218	SGN	001_13h00	DAD	001_14h20	03	001_15h05	5230	15695
219	SGN	001_14h00	DAD	001_15h20	03	001_16h05	4024	11962
220	SGN	001_15h00	DAD	001_16h20	03	001_17h05	5061	7523
221	SGN	001_16h00	DAD	001_17h20	03	001_18h05	9878	18210
222	SGN	001_17h00	DAD	001_18h20	03	001_19h05	4169	17591
223	SGN	001_18h00	DAD	001_19h20	03	001_20h05	7331	17247
224	SGN	001_20h00	DAD	001_21h20	03	001_22h05	7441	10818

File TimelineEvents.txt of the tested instance.

```

1 # File data events on timelines in the model
2 # No.   Airport   AircraftType   EventDate_EventTime
3 1 DAD 01 001_06h00
4 2 DAD 01 001_08h00
5 3 DAD 01 001_08h05

```

6	4	DAD	01	001_09h00
7	5	DAD	01	001_10h00
8	6	DAD	01	001_11h00
9	7	DAD	01	001_11h05
10	8	DAD	01	001_12h00
11	9	DAD	01	001_12h05
12	10	DAD	01	001_13h00
13	11	DAD	01	001_13h05
14	12	DAD	01	001_14h00
15	13	DAD	01	001_14h05
16	14	DAD	01	001_15h00
17	15	DAD	01	001_15h05
18	16	DAD	01	001_16h05
19	17	DAD	01	001_17h00
20	18	DAD	01	001_17h05
21	19	DAD	01	001_18h00
22	20	DAD	01	001_18h05
23	21	DAD	01	001_19h00
24	22	DAD	01	001_19h05
25	23	DAD	01	001_20h00
26	24	DAD	01	001_20h05
27	25	DAD	01	001_22h05
28	26	DAD	02	001_06h00
29	27	DAD	02	001_08h00
30	28	DAD	02	001_08h05
31	29	DAD	02	001_09h00
32	30	DAD	02	001_10h00
33	31	DAD	02	001_11h00
34	32	DAD	02	001_11h05
35	33	DAD	02	001_12h00
36	34	DAD	02	001_12h05
37	35	DAD	02	001_13h00
38	36	DAD	02	001_13h05
39	37	DAD	02	001_14h00
40	38	DAD	02	001_14h05
41	39	DAD	02	001_15h00
42	40	DAD	02	001_15h05
43	41	DAD	02	001_16h05
44	42	DAD	02	001_17h00
45	43	DAD	02	001_17h05
46	44	DAD	02	001_18h00
47	45	DAD	02	001_18h05
48	46	DAD	02	001_19h00
49	47	DAD	02	001_19h05
50	48	DAD	02	001_20h00
51	49	DAD	02	001_20h05
52	50	DAD	02	001_22h05
53	51	DAD	03	001_06h00
54	52	DAD	03	001_08h00
55	53	DAD	03	001_08h05
56	54	DAD	03	001_09h00
57	55	DAD	03	001_10h00
58	56	DAD	03	001_11h00
59	57	DAD	03	001_11h05

60	58	DAD	03	001_12h00
61	59	DAD	03	001_12h05
62	60	DAD	03	001_13h00
63	61	DAD	03	001_13h05
64	62	DAD	03	001_14h00
65	63	DAD	03	001_14h05
66	64	DAD	03	001_15h00
67	65	DAD	03	001_15h05
68	66	DAD	03	001_16h05
69	67	DAD	03	001_17h00
70	68	DAD	03	001_17h05
71	69	DAD	03	001_18h00
72	70	DAD	03	001_18h05
73	71	DAD	03	001_19h00
74	72	DAD	03	001_19h05
75	73	DAD	03	001_20h00
76	74	DAD	03	001_20h05
77	75	DAD	03	001_22h05
78	76	HAN	01	001_06h00
79	77	HAN	01	001_07h00
80	78	HAN	01	001_08h00
81	79	HAN	01	001_08h05
82	80	HAN	01	001_08h55
83	81	HAN	01	001_09h00
84	82	HAN	01	001_09h55
85	83	HAN	01	001_10h00
86	84	HAN	01	001_10h05
87	85	HAN	01	001_10h55
88	86	HAN	01	001_11h00
89	87	HAN	01	001_11h55
90	88	HAN	01	001_12h00
91	89	HAN	01	001_12h55
92	90	HAN	01	001_13h00
93	91	HAN	01	001_13h05
94	92	HAN	01	001_13h55
95	93	HAN	01	001_14h00
96	94	HAN	01	001_14h55
97	95	HAN	01	001_15h00
98	96	HAN	01	001_15h55
99	97	HAN	01	001_16h00
100	98	HAN	01	001_16h05
101	99	HAN	01	001_16h55
102	100	HAN	01	001_17h00
103	101	HAN	01	001_17h05
104	102	HAN	01	001_17h55
105	103	HAN	01	001_18h00
106	104	HAN	01	001_18h55
107	105	HAN	01	001_19h00
108	106	HAN	01	001_19h55
109	107	HAN	01	001_20h00
110	108	HAN	01	001_20h05
111	109	HAN	01	001_20h55
112	110	HAN	01	001_21h00
113	111	HAN	01	001_21h05

114	112	HAN	01	001_21h55
115	113	HAN	01	001_22h00
116	114	HAN	01	001_22h05
117	115	HAN	01	001_22h55
118	116	HAN	01	001_23h55
119	117	HAN	01	002_00h55
120	118	HAN	02	001_06h00
121	119	HAN	02	001_07h00
122	120	HAN	02	001_08h00
123	121	HAN	02	001_08h05
124	122	HAN	02	001_08h55
125	123	HAN	02	001_09h00
126	124	HAN	02	001_09h55
127	125	HAN	02	001_10h00
128	126	HAN	02	001_10h05
129	127	HAN	02	001_10h55
130	128	HAN	02	001_11h00
131	129	HAN	02	001_11h55
132	130	HAN	02	001_12h00
133	131	HAN	02	001_12h55
134	132	HAN	02	001_13h00
135	133	HAN	02	001_13h05
136	134	HAN	02	001_13h55
137	135	HAN	02	001_14h00
138	136	HAN	02	001_14h55
139	137	HAN	02	001_15h00
140	138	HAN	02	001_15h55
141	139	HAN	02	001_16h00
142	140	HAN	02	001_16h05
143	141	HAN	02	001_16h55
144	142	HAN	02	001_17h00
145	143	HAN	02	001_17h05
146	144	HAN	02	001_17h55
147	145	HAN	02	001_18h00
148	146	HAN	02	001_18h55
149	147	HAN	02	001_19h00
150	148	HAN	02	001_19h55
151	149	HAN	02	001_20h00
152	150	HAN	02	001_20h05
153	151	HAN	02	001_20h55
154	152	HAN	02	001_21h00
155	153	HAN	02	001_21h05
156	154	HAN	02	001_21h55
157	155	HAN	02	001_22h00
158	156	HAN	02	001_22h05
159	157	HAN	02	001_22h55
160	158	HAN	02	001_23h55
161	159	HAN	02	002_00h55
162	160	HAN	03	001_06h00
163	161	HAN	03	001_07h00
164	162	HAN	03	001_08h00
165	163	HAN	03	001_08h05
166	164	HAN	03	001_08h55
167	165	HAN	03	001_09h00

168	166	HAN	03	001_09h55
169	167	HAN	03	001_10h00
170	168	HAN	03	001_10h05
171	169	HAN	03	001_10h55
172	170	HAN	03	001_11h00
173	171	HAN	03	001_11h55
174	172	HAN	03	001_12h00
175	173	HAN	03	001_12h55
176	174	HAN	03	001_13h00
177	175	HAN	03	001_13h05
178	176	HAN	03	001_13h55
179	177	HAN	03	001_14h00
180	178	HAN	03	001_14h55
181	179	HAN	03	001_15h00
182	180	HAN	03	001_15h55
183	181	HAN	03	001_16h00
184	182	HAN	03	001_16h05
185	183	HAN	03	001_16h55
186	184	HAN	03	001_17h00
187	185	HAN	03	001_17h05
188	186	HAN	03	001_17h55
189	187	HAN	03	001_18h00
190	188	HAN	03	001_18h55
191	189	HAN	03	001_19h00
192	190	HAN	03	001_19h55
193	191	HAN	03	001_20h00
194	192	HAN	03	001_20h05
195	193	HAN	03	001_20h55
196	194	HAN	03	001_21h00
197	195	HAN	03	001_21h05
198	196	HAN	03	001_21h55
199	197	HAN	03	001_22h00
200	198	HAN	03	001_22h05
201	199	HAN	03	001_22h55
202	200	HAN	03	001_23h55
203	201	HAN	03	002_00h55
204	202	SGN	01	001_06h00
205	203	SGN	01	001_07h00
206	204	SGN	01	001_08h00
207	205	SGN	01	001_09h00
208	206	SGN	01	001_10h00
209	207	SGN	01	001_10h15
210	208	SGN	01	001_11h00
211	209	SGN	01	001_11h15
212	210	SGN	01	001_12h00
213	211	SGN	01	001_12h15
214	212	SGN	01	001_13h00
215	213	SGN	01	001_13h15
216	214	SGN	01	001_14h00
217	215	SGN	01	001_14h15
218	216	SGN	01	001_15h00
219	217	SGN	01	001_15h15
220	218	SGN	01	001_16h00
221	219	SGN	01	001_16h15

222	220	SGN	01	001_17h00
223	221	SGN	01	001_17h15
224	222	SGN	01	001_18h00
225	223	SGN	01	001_19h00
226	224	SGN	01	001_19h15
227	225	SGN	01	001_20h00
228	226	SGN	01	001_20h15
229	227	SGN	01	001_21h00
230	228	SGN	01	001_21h15
231	229	SGN	01	001_22h00
232	230	SGN	01	001_22h15
233	231	SGN	01	001_23h00
234	232	SGN	01	002_00h00
235	233	SGN	01	002_01h00
236	234	SGN	02	001_06h00
237	235	SGN	02	001_07h00
238	236	SGN	02	001_08h00
239	237	SGN	02	001_09h00
240	238	SGN	02	001_10h00
241	239	SGN	02	001_10h15
242	240	SGN	02	001_11h00
243	241	SGN	02	001_11h15
244	242	SGN	02	001_12h00
245	243	SGN	02	001_12h15
246	244	SGN	02	001_13h00
247	245	SGN	02	001_13h15
248	246	SGN	02	001_14h00
249	247	SGN	02	001_14h15
250	248	SGN	02	001_15h00
251	249	SGN	02	001_15h15
252	250	SGN	02	001_16h00
253	251	SGN	02	001_16h15
254	252	SGN	02	001_17h00
255	253	SGN	02	001_17h15
256	254	SGN	02	001_18h00
257	255	SGN	02	001_19h00
258	256	SGN	02	001_19h15
259	257	SGN	02	001_20h00
260	258	SGN	02	001_20h15
261	259	SGN	02	001_21h00
262	260	SGN	02	001_21h15
263	261	SGN	02	001_22h00
264	262	SGN	02	001_22h15
265	263	SGN	02	001_23h00
266	264	SGN	02	002_00h00
267	265	SGN	02	002_01h00
268	266	SGN	03	001_06h00
269	267	SGN	03	001_07h00
270	268	SGN	03	001_08h00
271	269	SGN	03	001_09h00
272	270	SGN	03	001_10h00
273	271	SGN	03	001_10h15
274	272	SGN	03	001_11h00
275	273	SGN	03	001_11h15

```

276 274 SGN 03 001_12h00
277 275 SGN 03 001_12h15
278 276 SGN 03 001_13h00
279 277 SGN 03 001_13h15
280 278 SGN 03 001_14h00
281 279 SGN 03 001_14h15
282 280 SGN 03 001_15h00
283 281 SGN 03 001_15h15
284 282 SGN 03 001_16h00
285 283 SGN 03 001_16h15
286 284 SGN 03 001_17h00
287 285 SGN 03 001_17h15
288 286 SGN 03 001_18h00
289 287 SGN 03 001_19h00
290 288 SGN 03 001_19h15
291 289 SGN 03 001_20h00
292 290 SGN 03 001_20h15
293 291 SGN 03 001_21h00
294 292 SGN 03 001_21h15
295 293 SGN 03 001_22h00
296 294 SGN 03 001_22h15
297 295 SGN 03 001_23h00
298 296 SGN 03 002_00h00
299 297 SGN 03 002_01h00

```

Our ZIMPL code for (*BFAM*) with objective (O1) of minimizing the total assignment cost is given below.

ZIMPL code for (*BFAM*) with objective (O1).

```

1  # This is a ZIMPL model file for Fleet Assignment Problem
2  # based on time-expanded multi-commodity flight network.
3  # Objective: Minimize the cost of assigned solution.
4
5  # Input files
6  param fileAirports := "Airports.txt";
7  param fileFleet    := "FleetComponent.txt";
8  param fileFlights  := "Flights.txt";
9  param fileEvents   := "TimelineEvents.txt";
10 param fileAllInfo  := "AssignmentData.txt";
11
12 # Information about airports
13 set Airports := {read fileAirports as "<1s>" comment "#"};
14 param AirportCapacity[Airports] := read fileAirports as "<1s> 2n" comment "#";
15
16 # Information about fleet
17 set AircraftTypes := {read fileFleet as "<1s>" comment "#"};
18 param nAircraftsOfType[AircraftTypes] := read fileFleet
19                                     as "<1s> 2n" comment "#";
20
21 ## CONSTRUCTION OF THE TIME-EXPANDED MULTI-COMMODITY NETWORK
22
23 # The set of all nodes together with their increasing order with respect to

```

```

24 # event time. Each member of this set has 4 components:
25 # order, airport, aircraft type, and date_time of event
26 set NodesWithOrder := {read fileEvents as "<1n, 2s, 3s, 4s>" comment "#"};
27
28 # Set of all nodes of the network (without their order)
29 # Each node has 3 components: airport, aircraft type, data_time of event
30 set Nodes := proj(NodesWithOrder, <2, 3, 4>);
31
32 # Set of forward ground arcs on the timelines associated with airports
33 set OrderedFGAs := {<i, aB, atB, tB, j, aE, atE, tE>
34     in NodesWithOrder * NodesWithOrder
35     with j == i + 1 and aB == aE and atB == atE};
36 set ForwardGroundArcs := proj(OrderedFGAs, <2,3,4,6,7,8>);
37
38 # Sets of the first nodes and the last nodes on timelines
39 set FirstAndLastOrderedNodes := {<i, aB, atB, tB, j, aE, atE, tE> in
40     NodesWithOrder * NodesWithOrder with i == 1 and j == card(NodesWithOrder)};
41 set TimelineBridges := {<i, aB, atB, tB, j, aE, atE, tE>
42     in NodesWithOrder * NodesWithOrder
43     with j == i + 1 and (aB != aE or atB != atE)};
44 set FirstNodes := proj(TimelineBridges, <6, 7, 8>)
45     + proj(FirstAndLastOrderedNodes, <2, 3, 4>);
46 set LastNodes := proj(TimelineBridges, <2, 3, 4>)
47     + proj(FirstAndLastOrderedNodes, <6, 7, 8>);
48
49 ## Set of backward ground arcs on the timelines associated with airports
50 set BackwardGroundArcs := {<aB, atB, tB, aE, atE, tE> in LastNodes * FirstNodes
51     with aB == aE and atB == atE};
52
53 # Set of ground arcs
54 set GroundArcs := ForwardGroundArcs + BackwardGroundArcs;
55
56 # Set of flight arcs in the network
57 set DataForAssign := {read fileAllInfo as "<1s, 2s, 3s, 4s, 5s, 6s, 7n, 8n>"
58     comment "#"};
59 set DepartReadyFlights := proj(DataForAssign, <1,2,3,5,6>);
60 set FlightArcs := {<aB, atB, tB, aE, atE, tE> in Nodes * Nodes with
61     <aB, tB, aE, atB, tE> in DepartReadyFlights and atE == atB};
62
63 # Cost of each assignment
64 param Cost[FlightArcs] := read fileAllInfo as "<1s, 5s, 2s, 3s, 5s, 6s> 7n"
65     comment "#";
66
67 # Set of active flights (ones with possible assignments)
68 set RawFlights := {read fileFlights as "<1s, 2s, 3s>" comment "#"};
69 set ActiveFlights := RawFlights inter proj(FlightArcs, <1, 4, 3>);
70
71 ## VARIABLES
72 var x[FlightArcs] binary;
73 var y[GroundArcs] >= 0;
74
75 ## MODELING OBJECTIVE
76 # Objective: Minimize the cost of assigned solution
77 minimize TotalCost: sum <aB, atB, tB, aE, atE, tE> in FlightArcs

```

```

74         with <aB, aE, tB> in FlightsActive:
75             Cost[aB,atB,tB,aE,atE,tE] * x[aB,atB,tB,aE,atE,tE];
76
77 ## MODELING CONSTRAINTS
78
79 # Exactly one aircraft type is assigned to each active flight
80 subto AssignToActiveFlights:
81     forall <aB, aE, tB> in ActiveFlights do
82         sum <aB1, atB, tB1, aE1, atE, tE> in FlightArcs with aB1 == aB
83         and tB1 == tB and aE1 == aE: x[aB1, atB, tB1, aE1, atE, tE] == 1;
84
85 # For each aircraft type, the number of used aircrafts is at most the number of
86     available aircrafts
87 subto FleetCapacity:
88     forall <at> in AircraftTypes do
89         sum <aB, atB, tB, aE, atE, tE> in BackwardGroundArcs
90         with atB == at: y[aB, atB, tB, aE, atE, tE]
91         <= nAircraftsOfType[at];
92
93 # Flow conversation at each node
94 subto FlowConversation:
95     forall <a, at, t> in Nodes do
96         sum <a_fin,at_fin,t_fin> in Nodes with <a_fin,at_fin,t_fin,a,at,t>
97         in FlightArcs: x[a_fin, at_fin, t_fin, a, at, t]
98     + sum <a_gin,at_gin,t_gin> in Nodes with <a_gin,at_gin,t_gin,a,at,t>
99     in GroundArcs: y[a_gin, at_gin, t_gin, a, at, t]
100 == sum <a_fout,at_fout,t_fout> in Nodes
101     with <a,at,t,a_fout,at_fout,t_fout> in FlightArcs:
102     x[a, at, t, a_fout, at_fout, t_fout]
103 + sum <a_gout,at_gout,t_gout> in Nodes
104     with <a,at,t,a_gout,at_gout,t_gout> in GroundArcs:
105     y[a, at, t, a_gout, at_gout, t_gout];

```

Our ZIMPL code for (*BFAM*) with objective (O2) of maximizing the total assignment revenue is the same as the above code, except for the lines 59-60 and the lines 71-74. Precisely, the code is given below.

ZIMPL code for (*BFAM*) with objective (O2).

```

1 # This is a ZIMPL model file for Fleet Assignment Problem
2 # based on time-expanded multi-commodity flight network.
3 # Objective: Maximize the revenue of assigned solution.
4
5 # Input files
6 param fileAirports := "Airports.txt";
7 param fileFleet := "FleetComponent.txt";
8 param fileFlights := "Flights.txt";
9 param fileEvents := "TimelineEvents.txt";
10 param fileAllInfo := "AssignmentData.txt";
11
12 # Information about airports
13 set Airports := {read fileAirports as "<1s>" comment "#"};
14 param AirportCapacity[Airports] := read fileAirports as "<1s> 2n" comment "#";

```

```

15
16 # Information about fleet
17 set AircraftTypes := {read fileFleet as "<1s>" comment "#"};
18 param nAircraftsOfType[AircraftTypes] := read fileFleet as "<1s> 2n" comment "#
    ";
19
20 ## CONSTRUCTION OF THE TIME-EXPANDED MULTI-COMMODITY NETWORK
21
22 # The set of all nodes together with their increasing order with respect to
    event time
23 # Each member of this set has 4 components: order, airport, aircraft type, and
    date_time of event
24 set NodesWithOrder := {read fileEvents as "<1n, 2s, 3s, 4s>" comment "#"};
25
26 # Set of all nodes of the network (without their order)
27 # Each node has 3 components: airport, aircraft type, data_time of event
28 set Nodes := proj(NodesWithOrder, <2, 3, 4>);
29
30 # Set of forward ground arcs on the timelines associated with airports
31 set OrderedFGAs := {<i, aB, atB, tB, j, aE, atE, tE> in NodesWithOrder *
    NodesWithOrder with j == i + 1 and aB == aE and atB == atE};
32 set ForwardGroundArcs := proj(OrderedFGAs, <2,3,4,6,7,8>);
33
34 # Sets of the first nodes and the last nodes on timelines
35 set FirstAndLastOrderedNodes := {<i, aB, atB, tB, j, aE, atE, tE> in
    NodesWithOrder * NodesWithOrder with i == 1 and j == card(NodesWithOrder)};
36 set TimelineBridges := {<i, aB, atB, tB, j, aE, atE, tE> in NodesWithOrder *
    NodesWithOrder with j == i + 1 and (aB != aE or atB != atE)};
37 set FirstNodes := proj(TimelineBridges, <6, 7, 8>) + proj(
    FirstAndLastOrderedNodes, <2, 3, 4>);
38 set LastNodes := proj(TimelineBridges, <2, 3, 4>) + proj(
    FirstAndLastOrderedNodes, <6, 7, 8>);
39
40 # # Set of backward ground arcs on the timelines associated with airports
41 set BackwardGroundArcs := {<aB, atB, tB, aE, atE, tE> in LastNodes * FirstNodes
    with aB == aE and atB == atE};
42
43 # Set of ground arcs
44 set GroundArcs := ForwardGroundArcs + BackwardGroundArcs;
45
46 # Set of flight arcs in the network
47 set DataForAssign := {read fileAllInfo as "<1s, 2s, 3s, 4s, 5s, 6s, 7n, 8n>"
    comment "#"};
48 set DepartReadyFlights := proj(DataForAssign, <1,2,3,5,6>);
49 set FlightArcs := {<aB, atB, tB, aE, atE, tE> in Nodes * Nodes with <aB, tB, aE
    , atB, tE> in DepartReadyFlights and atE == atB};
50
51 # Revenue of each assignment
52 param Revenue[FlightArcs] := read fileAllInfo as "<1s, 5s, 2s, 3s, 5s, 6s> 8n"
    comment "#";
53
54 # Set of active flights (ones with possible assignments)
55 set RawFlights := {read fileFlights as "<1s, 2s, 3s>" comment "#"};
56 set ActiveFlights := RawFlights inter proj(FlightArcs, <1, 4, 3>);

```



```

57
58 ## VARIABLES
59 var x[FlightArcs] binary;
60 var y[GroundArcs] >= 0;
61
62 ## MODELING OBJECTIVE
63 maximize TotalRevenue: sum <aB, atB, tB, aE, atE, tE> in FlightArcs with <aB,
    aE, tB> in FlightsActive: Revenue[aB, atB, tB, aE, atE, tE] * x[aB, atB, tB
    , aE, atE, tE];
64
65 ## MODELING CONSTRAINTS
66
67 # Exactly one aircraft type is assigned to each active flight
68 subto AssignToActiveFlights:
69     forall <aB, aE, tB> in ActiveFlights do
70         sum <aB1, atB, tB1, aE1, atE, tE> in FlightArcs with aB1 == aB and tB1
            == tB and aE1 == aE: x[aB1, atB, tB1, aE1, atE, tE] == 1;
71
72 # For each aircraft type, the number of used aircrafts is at most the number of
    available aircrafts
73 subto FleetCapacity:
74     forall <at> in AircraftTypes do
75         sum <aB, atB, tB, aE, atE, tE> in BackwardGroundArcs with atB == at
            : y[aB, atB, tB, aE, atE, tE] <= nAircraftsOfType[at];
76
77 # Flow conversation at each node
78 subto FlowConversation:
79     forall <a, at, t> in Nodes do
80         sum <a_fin, at_fin, t_fin> in Nodes with <a_fin, at_fin, t_fin, a,
            at, t> in FlightArcs: x[a_fin, at_fin, t_fin, a, at, t]
81         + sum <a_gin, at_gin, t_gin> in Nodes with <a_gin, at_gin, t_gin, a, at
            , t> in GroundArcs: y[a_gin, at_gin, t_gin, a, at, t]
82         == sum <a_fout, at_fout, t_fout> in Nodes with <a, at, t, a_fout, at_fout,
            t_fout> in FlightArcs: x[a, at, t, a_fout, at_fout, t_fout]
83         + sum <a_gout, at_gout, t_gout> in Nodes with <a, at, t, a_gout,
            at_gout, t_gout> in GroundArcs: y[a, at, t, a_gout, at_gout, t_gout
            ];

```

Our ZIMPL code for (*BFAM*) with objective (O3) of minimizing the number of used airplanes is the same as the above code, except that the lines 59-60 are removed, and the lines 71-74 are changed. Precisely, the code is given below.

ZIMPL code for (*BFAM*) with objective (O3).

```

1 # This is a ZIMPL model file for Fleet Assignment Problem
2 # based on time-expanded multi-commodity flight network.
3 # Objective: Minimize the number of used aircrafts.
4
5 # Input files
6 param fileAirports := pathToDataFolder + "Airports.txt";
7 param fileFleet    := pathToDataFolder + "FleetComponent.txt";

```

```

8  param fileFlights := pathToDataFolder + "Flights.txt";
9  param fileEvents := pathToDataFolder + "TimelineEvents.txt";
10 param fileAllInfo := pathToDataFolder + "AssignmentData.txt";
11
12 # Information about airports
13 set Airports := {read fileAirports as "<1s>" comment "#"};
14 param AirportCapacity[Airports] := read fileAirports as "<1s> 2n" comment "#";
15
16 # Information about fleet
17 set AircraftTypes := {read fileFleet as "<1s>" comment "#"};
18 param nAircraftsOfType[AircraftTypes] := read fileFleet as "<1s> 2n" comment "#";
19
20 ## CONSTRUCTION OF THE TIME-EXPANDED MULTI-COMMODITY NETWORK
21
22 # The set of all nodes together with their increasing order with respect to
    event time
23 # Each member of this set has 4 components: order, airport, aircraft type, and
    date_time of event
24 set NodesWithOrder := {read fileEvents as "<1n, 2s, 3s, 4s>" comment "#"};
25
26 # Set of all nodes of the network (without their order)
27 # Each node has 3 components: airport, aircraft type, data_time of event
28 set Nodes := proj(NodesWithOrder, <2, 3, 4>);
29
30 # Set of forward ground arcs on the timelines associated with airports
31 set OrderedFGAs := {<i, aB, atB, tB, j, aE, atE, tE> in NodesWithOrder *
    NodesWithOrder with j == i + 1 and aB == aE and atB == atE};
32 set ForwardGroundArcs := proj(OrderedFGAs, <2,3,4,6,7,8>);
33
34 # Sets of the first nodes and the last nodes on timelines
35 set FirstAndLastOrderedNodes := {<i, aB, atB, tB, j, aE, atE, tE> in
    NodesWithOrder * NodesWithOrder with i == 1 and j == card(NodesWithOrder)};
36 set TimelineBridges := {<i, aB, atB, tB, j, aE, atE, tE> in NodesWithOrder *
    NodesWithOrder with j == i + 1 and (aB != aE or atB != atE)};
37 set FirstNodes := proj(TimelineBridges, <6, 7, 8>) + proj(
    FirstAndLastOrderedNodes, <2, 3, 4>);
38 set LastNodes := proj(TimelineBridges, <2, 3, 4>) + proj(
    FirstAndLastOrderedNodes, <6, 7, 8>);
39
40 # # Set of backward ground arcs on the timelines associated with airports
41 set BackwardGroundArcs := {<aB, atB, tB, aE, atE, tE> in LastNodes * FirstNodes
    with aB == aE and atB == atE};
42
43 # Set of ground arcs
44 set GroundArcs := ForwardGroundArcs + BackwardGroundArcs;
45
46 # Set of flight arcs in the network
47 set DataForAssign := {read fileAllInfo as "<1s, 2s, 3s, 4s, 5s, 6s, 7n, 8n>"
    comment "#"};
48 set DepartReadyFlights := proj(DataForAssign, <1,2,3,5,6>);
49 set FlightArcs := {<aB, atB, tB, aE, atE, tE> in Nodes * Nodes with <aB, tB, aE
    , atB, tE> in DepartReadyFlights and atE == atB};
50

```

```

51 # Set of active flights (ones with possible assignments)
52 set RawFlights := {read fileFlights as "<1s, 2s, 3s>" comment "#"};
53 set ActiveFlights := RawFlights inter proj(FlightArcs, <1, 4, 3>);
54
55 ## VARIABLES
56 var x[FlightArcs] binary;
57 var y[GroundArcs] >= 0;
58
59 ## MODELING OBJECTIVE
60 minimize NumberOfUsedAircrafts: sum <aB, atB, tB, aE, atE, tE> in
    BackwardGroundArcs: y[aB, atB, tB, aE, atE, tE];
61
62 ## MODELING CONSTRAINTS
63
64 # Exactly one aircraft type is assigned to each active flight
65 subto AssignToActiveFlights:
66     forall <aB, aE, tB> in ActiveFlights do
67         sum <aB1, atB, tB1, aE1, atE, tE> in FlightArcs with aB1 == aB and tB1
            == tB and aE1 == aE: x[aB1, atB, tB1, aE1, atE, tE] == 1;
68
69 # For each aircraft type, the number of used aircrafts is at most the number of
    available aircrafts
70 subto FleetCapacity:
71     forall <at> in AircraftTypes do
72         sum <aB, atB, tB, aE, atE, tE> in BackwardGroundArcs with atB == at
            : y[aB, atB, tB, aE, atE, tE] <= nAircraftsOfType[at];
73
74 # Flow conversation at each node
75 subto FlowConversation:
76     forall <a, at, t> in Nodes do
77         sum <a_fin, at_fin, t_fin> in Nodes with <a_fin, at_fin, t_fin, a,
            at, t> in FlightArcs: x[a_fin, at_fin, t_fin, a, at, t]
78         + sum <a_gin, at_gin, t_gin> in Nodes with <a_gin, at_gin, t_gin, a, at
            , t> in GroundArcs: y[a_gin, at_gin, t_gin, a, at, t]
79     == sum <a_fout, at_fout, t_fout> in Nodes with <a, at, t, a_fout, at_fout,
            t_fout> in FlightArcs: x[a, at, t, a_fout, at_fout, t_fout]
80     + sum <a_gout, at_gout, t_gout> in Nodes with <a, at, t, a_gout,
            at_gout, t_gout> in GroundArcs: y[a, at, t, a_gout, at_gout, t_gout
            ];

```

Our experiments were conducted on a computer with the following configurations: Intel(R) Core(TM) i7-6700HQ CPU 2*2.60 GHz, 16 GB RAM, Windows 10 Operating System. The numerical results are reported in Table 1.4.

Objective	# vars	# cons	Running time
(O1)	519	374	0.02 s
(O2)	519	374	0.02 s
(O3)	519	374	0.02 s

Table 1.4: Numerical results of (*BFA**M*) formulations on the tested instance.

With the running times of 0.02 seconds for solving the tested instance of small size with 75 flights between 3 airports and 3 fleets, we can say that (BFAM) is efficient in solving the fleet assignment problems. In the future we will construct larger size instances of the problems and test the model on these instances to have a better evaluation on the performance of (BFAM).

Chapter 2

Wait-and-see fleet assignment

In this chapter we study the so-called wait-and-see fleet assignment which has already introduced shortly in Introduction. The detail description of this problem is presented in Section 2.1. Section 2.2 gives the baseline theory for the construction of our mixed integer programming formulation for this problem in Section 2.3. Section 2.4 presents our numerical experiments to evaluate the performance of this formulation.

2.1 Problem statement

The fleet assignment problem presented in Chapter 1 aims to construct a fleet assignment with deterministic data. Such assignment is intended to be used for a long period of time. However, when operating in practice, many situations from various reasons can happen. Here are the most common situations one can experience in real life.

- Due to the bad weather and/or airline traffic control reasons, some flights are delayed or even canceled.
- Due to mechanical and/or maintenance reasons, the size of some fleet decreases since some airplanes in the original fleet are not available to fly. The size of fleet may also increase in case the airline buys some new airplanes for development purposes.
- Some new flight legs are opened to catch the high transportation demand of passengers. In this case, new flight schedule associated with the new legs are also given.

- Some of current flight legs are closed due to the low transportation demand of passengers. In this case, the flights associated with these legs are also omitted.

The scheduled fleet assignment may not work anymore in the new situation, hence there is a need of constructing a new fleet assignment adapting to the updated input data. The problem under our consideration in this chapter is to find such a new fleet assignment satisfying the two following criteria.

- (W1) It is valid (i.e., it satisfies constraints (C1)-(C4) stated in Section 1.1) in the setting of the new situation.
- (W2) It has the least difference (i.e. the minimum number of changes) from the scheduled one.

The former criterion is to ensure that the new fleet assignment works well under the setting of the new situation. The latter criterion is to reduce the negative affect of changing from the scheduled assignment to the new one. We call this problem by *wait-and-see fleet assignment*, in which the term ‘wait-and-see’ is to emphasize that the solution of this problem is made after the realization of data in the new setting.

2.2 Wait-and-see recovery robustness

In this section, we introduce the concept of wait-and-see recovery robustness. This concept has a similar spirit to the recovery-to-feasibility robustness concept introduced in [8]. Due to the similarity, to better understand our proposed robustness concept, we first review the latter one.

Consider an uncertain optimization problem $(P_\xi)_{\xi \in \mathcal{U}}$, which is a parameterized family of optimization problems corresponding to $\xi \in \mathcal{U} \subset \mathbb{R}^p$ (for some $p \in \mathbb{N}$):

$$\begin{aligned}
 (P_\xi) \quad & \min && f(x, \xi) \\
 & \text{s.t.} && F(x, \xi) \leq 0 \\
 & && x \in \mathcal{X},
 \end{aligned}$$

where

- ξ is the parameter vector representing data elements of the problem (P_ξ) ,
- \mathcal{U} is the set consisting of considered values of parameter ξ ,

- x is the decision vector,
- $\mathcal{X} \subset \mathbb{R}^n$ is the variable space (here n is the dimension of the space),
- $f(\cdot, \xi) : \mathbb{R}^n \rightarrow \mathbb{R}$ is the objective function corresponding to $\xi \in \mathcal{U}$,
- $F(\cdot, \xi) : \mathbb{R}^n \rightarrow \mathbb{R}^m$ (for some $m \in \mathbb{N}$) is the function describing the constraints of (P_ξ) for any fixed $\xi \in \mathcal{U}$.

To be precise, the vector inequality $F(x, \xi) \leq 0$ in the description of constraints of (P_ξ) means that $F_i(x, \xi) \leq 0$ for $i = 1, \dots, m$, in which the functions F_i 's are the components of the vector function F . The set \mathcal{U} is called *uncertainty set* and its elements are called *scenarios*. The uncertainty set can be given by disturbing a so-called *nominal scenario*. The nominal scenario can also refer to the most likely value of data vector ξ . The optimization problem $(P_{\bar{\xi}})$ corresponding to the nominal scenario $\bar{\xi} \in \mathcal{U}$ is called the *nominal problem* of the uncertain optimization problem $(P_\xi)_{\xi \in \mathcal{U}}$. An optimal solution to the nominal problem $(P_{\bar{\xi}})$ is called a *nominal solution* to $(P_\xi)_{\xi \in \mathcal{U}}$.

We denote by $\mathcal{F}(\xi)$ the feasible set of (P_ξ) , i.e.

$$\mathcal{F}(\xi) := \{x \in \mathcal{X} \mid F(x, \xi) \leq 0\}.$$

Let $d : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}_+$ be a function that we shall call *recovery cost* from now. A *recovery-to-feasibility robust solution* to $(P(\xi))_{\xi \in \mathcal{U}}$ is a solution x to the following recovery-to-feasibility robust counterpart

$$(\text{RecFeas}(\mathcal{U})) \quad \min_{x \in \mathcal{X}} \sup_{\xi \in \mathcal{U}} d(x, \mathcal{F}(\xi)),$$

where

$$d(x, \mathcal{F}(\xi)) = \min_{y \in \mathcal{F}(\xi)} d(x, y).$$

This means that we can recover a recovery-to-feasibility robust solution to be feasible in any scenario minimizing the recovery cost. More precisely, the recovery-to-feasibility robustness concept composes of the following main points:

- The robust solution x has to be made *before* the realization of uncertain parameter ξ . It does not need to be feasible to any problem in the family $(P_\xi)_{\xi \in \mathcal{U}}$.
- When the parameter ξ becomes realized, one can recover the computed solution x to obtain a feasible solution y of the corresponding optimization problem (P_ξ) .

- The recovery cost in the worst case of parameter ξ is minimized.

We now define the concept of *wait-and-see recovery robustness*. The *wait-and-see recovery robust counterpart* of $(P(\xi))_{\xi \in \mathcal{U}}$ is the following problem

$$(\text{WasRec}(\mathcal{U})) \quad \sup_{\xi \in \mathcal{U}} \min_{x \in \mathcal{F}(\xi)} d(x, \mathcal{F}(\bar{\xi})),$$

The objective of this robust counterpart is to minimize the cost of recovering a nominal solution to a feasible solution in the worst case of parameter ξ . This has a similar spirit to the objective of the recovery-to-feasibility robust counterpart. However, the wait-and-see recovery robustness has the following key differences from recovery-to-feasibility robustness:

- The feasible set $\mathcal{F}(\bar{\xi})$ of the nominal problem $(P_{\bar{\xi}})$ has to be given *before* the realization of uncertain parameter ξ .
- The robust solution x has to be made *after* the realization of uncertain parameter ξ . It must be feasible to some problem in the family $(P_{\xi})_{\xi \in \mathcal{U}}$.

To see the application of wait-and-see recovery robustness concept to the context of fleet assignment problem, we consider the special case in which \mathcal{U} consists of only two instances $\{\bar{\xi}, \xi^*\}$. In this case, $\bar{\xi}$ is the nominal instance of parameter ξ , while ξ^* is the only possible realization of the parameter. Furthermore, a feasible solution $\bar{x} \in \mathcal{F}(\bar{\xi})$ to the nominal problem $P(\bar{\xi})$ is computed beforehand. The wait-and-see recovery robust counterpart in this situation becomes

$$\min_{x \in \mathcal{F}(\xi^*)} d(x, \bar{x}). \quad (2.1)$$

Any solution to this problem is called a *wait-and-see recovery robust solution*. Such solution is determined to be feasible *after* the uncertain parameter ξ becomes realized, and has the minimum recovery cost from the given nominal solution.

2.3 Formulation

As mentioned in Section 1.1 and Section 1.2, in the deterministic setting of the old situation, the input data of the fleet assignment problem include:

- the set A of airports in the flight network of the airline,
- the set D of flight legs in the network of the airline,

- the set L of flights in the considered time period,
- the list \mathcal{L} of required throughs,
- the set F of available fleets (i.e. airplane types) of the airline,
- the set $D_f \subset D$ of the flight legs that can be served by airplanes in each fleet $f \in F$,
- the number n_f of airplanes of each type $f \in F$,
- the cost $c_{f\ell}$ to serve each flight $\ell \in L$ by each fleet $f \in F$,
- the revenue $r_{f\ell}$ obtained by using fleet $f \in F$ to serve flight $\ell \in L$.

For convenience, we denote by

$$\mathcal{I} = \{A, D, L, \mathcal{L}, F, D_f, n_f, c_{f\ell}, r_{f\ell} \text{ with } f \in F, \ell \in L\}$$

the input data of the fleet assignment in the deterministic setting of old situation. The updated input data in the setting of the new situation are denoted by

$$\tilde{\mathcal{I}} = \{\tilde{A}, \tilde{D}, \tilde{L}, \tilde{\mathcal{L}}, \tilde{F}, \tilde{D}_f, \tilde{n}_f, \tilde{c}_{f\ell}, \tilde{r}_{f\ell} \text{ with } f \in \tilde{F}, \ell \in \tilde{L}\}.$$

in which its elements representing the data components in the new setting.

Following the paradigm of constructing the time-expanded multi-commodity network in Section 1.2.1 and the description of BFAM formulation in Section 1.2.2, we construct the time-expanded multi-commodity network associated with the input data \mathcal{I} , and let L^p be the set of flyable arcs of this network. In the manner of (1.1), the scheduled fleet assignment (corresponding to the old situation) can be encoded by a binary vector \bar{x} in which

$$\bar{x}_{fai,fbj} = \begin{cases} 1 & \text{if the arc } ((f, a, i), (f, b, j)) \in L^p \text{ is used in the assignment,} \\ 0 & \text{otherwise.} \end{cases}$$

Similarly, we construct the time-expanded multi-commodity network corresponding to the updated data $\tilde{\mathcal{I}}$. Let \tilde{N} be the set of nodes, \tilde{L}^p the set of flyable arcs, and \tilde{L}^{rt} the set of arcs in required throughs in this network. Furthermore, for each fleet $f \in \tilde{F}$, let \tilde{N}_f^* be the set of the last nodes in the time lines corresponding to the fleet. Using the BFAM formulation on this network, one can compute a feasible fleet assignment in the setting of the new

data $\tilde{\mathcal{I}}$. More precisely, the formulation uses the following binary variables to encode such assignment:

$$z_{fai,fbj} = \begin{cases} 1 & \text{if the arc } ((f, a, i), (f, b, j)) \in \tilde{L}^p \text{ is used in the assignment,} \\ 0 & \text{otherwise.} \end{cases}$$

Furthermore, the formulation needs the following additional variables:

$$u_{fai}^- := \text{flow value of the ground arc coming to } (f, a, i) \in \tilde{N},$$

$$u_{fai}^+ := \text{flow value of the ground arc going out of } (f, a, i) \in \tilde{N}.$$

A feasible fleet assignment in the setting of the updated data \tilde{I} is a feasible solution of the following BFAM formulation.

$$\sum_{f \in \tilde{F}} z_{fai,fbj'} = 1 \quad \forall ((a, i), (b, j)) \in \tilde{L} \quad (2.2)$$

$$\sum_{(f,a,i) \in \tilde{N}_f^*} u_{fai}^+ \leq \tilde{n}_f \quad \forall f \in \tilde{F} \quad (2.3)$$

$$\sum_{a,i} z_{fai,fbj} + u_{fbj}^- = \sum_{a,i} z_{fbj,fai} + u_{fbj}^+ \quad \forall (f, b, j) \in \tilde{N} \quad (2.4)$$

$$z_{fai,fbj} = x_{fbj,fck} \quad \forall (((f, a, i), (f, b, j)), ((f, b, j), (f, c, k))) \in \tilde{L}^{rt} \quad (2.5)$$

$$z_{fai,fbj} \in \{0, 1\} \quad \forall ((f, a, i), (f, b, j)) \in \tilde{L}^p \quad (2.6)$$

$$u_{fbj}^+ \geq 0 \quad \forall (f, b, j) \in \tilde{N} \quad (2.7)$$

$$u_{fbj}^- \geq 0 \quad \forall (f, b, j) \in \tilde{N}. \quad (2.8)$$

The objective of the wait-and-see fleet assignment problem is to minimize the number of changes from the scheduled fleet assignment encoded by \bar{x} to the new one encoded by z . Keeping in mind that the vectors \bar{x} and z are binary, this cost can be computed by the Hamming distance between these vectors. Recall from [9] that the Hamming distance between two binary vectors $v, w \in \{0, 1\}^m$ is the number of indices at which the corresponding components of these vectors are different, and is computed by

$$\sum_{j=1}^m |v_j - w_j|.$$

Note that the Hamming distance is defined for two vectors of the same index set, while our vectors \bar{x} and z may not satisfy this condition. In order to use the Hamming distance to compute the objective value of the wait-and-see fleet assignment problem, we need to add some components to each of the vectors so that they share the same index set. This can be done as follows.

- For each $e \in \widetilde{L^p} \setminus L^p$, we add component \bar{x}_e to vector \bar{x} and set $\bar{x}_e = 0$.
- For each $e \in L^p \setminus \widetilde{L^p}$, we add component z_e to vector x and set $z_e = 0$.

This means that any index of z that is not an index of \bar{x} will be added to the index set of \bar{x} , and vice versa. By setting zero value for the additional components of the vectors, we impose that the assignments corresponding to the components will not appear in the assignment solutions. Now, the objective value is explicitly computed as the Hamming distance between the two updated vectors, i.e.,

$$d(z, \bar{x}) = \sum_{j \in L^p \cup \widetilde{L^p}} |z_j - \bar{x}_j|. \quad (2.9)$$

Since the value of \bar{x} is given beforehand, this objective value is in fact a function of z . Furthermore, in form of (2.9), this function is nonlinear. However, the objective of minimizing this recovery cost function can be linearized as follows.

$$\begin{aligned} \min \quad & \sum_{j \in L^p \cup \widetilde{L^p}} v_j \\ \text{s.t.} \quad & -v_j \leq z_j - \bar{x}_j \leq v_j \quad \forall j \in L^p \cup \widetilde{L^p}. \end{aligned}$$

To the end, we come up with the following MIP formulation for the wait-and-see recovery robust fleet assignment.

$$\begin{aligned} (BFAMwas) \quad \min \quad & \sum_{((f,a,i),(f,b,j)) \in L^p \cup \widetilde{L^p}} v_{faj,fbj} \\ \text{s.t.} \quad & z_{faj,fbj} - \bar{x}_{faj,fbj} \geq -v_{faj,fbj} \quad \forall ((f,a,i),(f,b,j)) \in L^p \cup \widetilde{L^p} \\ & z_{faj,fbj} - \bar{x}_{faj,fbj} \leq v_{faj,fbj} \quad \forall ((f,a,i),(f,b,j)) \in L^p \cup \widetilde{L^p} \\ & \sum_{f \in \widetilde{F}} z_{fai,fbj'} = 1 \quad \forall ((a,i),(b,j)) \in \widetilde{L} \end{aligned}$$

$$\begin{aligned}
\sum_{(f,a,i) \in \widetilde{N}_f^*} u_{fai}^+ &\leq \widetilde{n}_f && \forall f \in \widetilde{F} \\
\sum_{a,i} z_{fai,fbj} + u_{fbj}^- &= \sum_{a,i} z_{fbj,fai} + u_{fbj}^+ && \forall (f,b,j) \in \widetilde{N} \\
z_{fai,fbj} &= z_{fbj,fck} && \forall ((f,a,i), (f,b,j)), \\
&&& ((f,b,j), (f,c,k)) \in \widetilde{L}^{rt} \\
z_{fai,fbj} &= 0 && \forall ((f,a,i), (f,b,j)) \in L^p \setminus \widetilde{L}^p \\
z_{fai,fbj} &\in \{0,1\} && \forall ((f,a,i), (f,b,j)) \in L^p \cup \widetilde{L}^p \\
u_{fbj}^+ &\geq 0 && \forall (f,b,j) \in \widetilde{N} \\
u_{fbj}^- &\geq 0 && \forall (f,b,j) \in \widetilde{N}.
\end{aligned}$$

2.4 Numerical experiments

To evaluate the performance of (*BFAMwas*), we first created some data sets for the wait-and-see fleet assignment. As described in the previous section, the data of a wait-and-see recovery robust fleet assignment problem consist of a given fleet assignment (scheduled for the old setting) and the updated collection

$$\widetilde{\mathcal{I}} = \{\widetilde{A}, \widetilde{D}, \widetilde{L}, \widetilde{\mathcal{L}}, \widetilde{F}, \widetilde{D}_f, \widetilde{n}_f, \widetilde{c}_{f\ell}, \widetilde{r}_{f\ell} \text{ with } f \in \widetilde{F}, \ell \in \widetilde{L}\}$$

of data in the new situation. To see the data changes from the previous situation to the new situation, the collection

$$\mathcal{I} = \{A, D, L, \mathcal{L}, F, D_f, n_f, c_{f\ell}, r_{f\ell} \text{ with } f \in F, \ell \in L\}$$

of data in the previous situation may be also given. The given scheduled fleet assignment is constructed in the setting of \mathcal{I} . Hence, each instance of the wait-and-see fleet assignment problem consists of two parts: the *new part* corresponds to the new situation, the *old part* corresponds to the previous situation.

The new part of each data set corresponding to each wait-and-see fleet assignment problem instance consists of 5 excel files named *AirportsNew.xlsx*, *FleetComponentNew.xlsx*, *FlightLegsNew.xlsx*, *FlightsNew.xlsx*, and *AssignmentDataNew.xlsx*. The contents and functionalities of these files are respectively similar to the files *Airports.xlsx*, *FleetComponent.xlsx*, *FlightLegs.xlsx*, *Flights.xlsx*, and *AssignmentData.xlsx* as we have described in Section 1.3.

Note that in the file `AssignmentDataNew.xlsx` we do not need to provide the information about the cost and the revenue of each assignment in the new situation, since that parameters are not relevant to our objective which aims to minimize the number of differences between the new assignment with the scheduled one.

The old part of each data set corresponding to each wait-and-see fleet assignment problem instance contains an excel file named *OldAssignment.xlsx*, which saves the scheduled fleet assignment. Each line of this file includes the information of an assignment: a flight in the previous situation and the aircraft type assigned to it. For the sake of completeness, we provide the data materials for constructing the scheduled assignment. These materials are saved in other separated excel files in the old part. Their contents are almost similar to the files in the new part, but correspond to the old situation. More precisely, apart from the file `OldAssignment.xlsx`, the old part may contain excel files `AirportsOld.xlsx`, `FleetComponentOld.xlsx`, `FlightLegsOld.xlsx`, `Flight-sOld.xlsx`, and `AssignmentDataOld.xlsx`. These additional excel files have the same structure as the corresponding files in the new part.

We generated three problem instances named `WasFA1`, `WasFA2`, `WasFA3`. They are constructed from the public data about the schedule of domestic flights of Vietnam Airlines in the duration from October 27th, 2019 to March 28th, 2020. Each instance consists of two parts: old and new ones. The old parts of the tested instances share the same data of 19 domestic airports and 77 domestic flight legs in Vietnam, 5 fleets and 243 domestic flights of Vietnam Airlines in a nominal day. Furthermore, they share a list of all possible assignments between the flights and aircraft types, together with the cost and the revenue of each of such assignments. Additionally, the old part of each tested instance includes a scheduled fleet assignment. The assignment in instance `WasFA1` (resp., `WasFA2`, `WasFA3`) is the scheduled fleet assignment which is optimal with respect to the objective of minimizing the number of used aircrafts (resp., the objective of minimizing the total assignment cost, the objective of maximizing the total assignment revenue). The new parts of the tested instances are the same, and they are made from the common data in the old parts by delaying 5 flights, canceling 6 flights, and restricting fleet assignability on 3 flight legs. These tested instances are available on

<https://github.com/lxthanh86/WasFleetAssignment>.

We implemented the proposed formulation by using ZIMPL 3.5.3 (cf. [7]). The ZIMPL code of the formulation is given in the appendix section at the

end of this report. We used GUROBI 9.1 (see <https://www.gurobi.com>) as a mixed integer programming solver. All experiments were conducted on a computer with an Intel(R) Core(TM) i7-6700HQ CPU 2.6 GHz processor and 16 GB of RAM. Our ZIMPL code for (*BFAMwas*) is given below.

ZIMPL code for (*BFAMwas*).

```

1  # This is a ZIMPL model file for Wait-and-see Recovery Robust Fleet Assignment
    Problem
2  # based on time-expanded multi-commodity flight network and Hamming recovery
    cost.
3
4  # Input files
5  param fileAirports := "AirportsNew.txt";
6  param fileFleet    := "FleetComponentNew.txt";
7  param fileFlights  := "FlightsNew.txt";
8  param fileEvents   := "TimelineEventsNew.txt";
9  param fileAllInfo  := "AssignmentDataNew.txt";
10 param fileOldFA    := "PreviousSolution.txt";
11
12 # Information about airports
13 set Airports := {read fileAirports as "<1s>" comment "#"};
14 param AirportCapacity[Airports] := read fileAirports as "<1s> 2n" comment "#";
15
16 # Information about fleet
17 set AircraftTypes := {read fileFleet as "<1s>" comment "#"};
18 param nAircraftsOfType[AircraftTypes] := read fileFleet as "<1s> 2n" comment "#";
19
20 ## CONSTRUCTION OF THE TIME-EXPANDED MULTI-COMMODITY NETWORK
21
22 # The set of all nodes together with their increasing order with respect to
    event time
23 # Each member of this set has 4 components: order, airport, aircraft type, and
    date_time of event
24 set NodesWithOrder := {read fileEvents as "<1n, 2s, 3s, 4s>" comment "#"};
25
26 # Set of all nodes of the network (without their order)
27 # Each node has 3 components: airport, aircraft type, data_time of event
28 set Nodes := proj(NodesWithOrder, <2, 3, 4>);
29
30 # Set of forward ground arcs on the timelines associated with airports
31 set OrderedFGAs := {<i, aB, atB, tB, j, aE, atE, tE> in NodesWithOrder *
    NodesWithOrder with j == i + 1 and aB == aE and atB == atE};
32 set ForwardGroundArcs := proj(OrderedFGAs, <2,3,4,6,7,8>);
33
34 # Sets of the first nodes and the last nodes on timelines
35 set FirstAndLastOrderedNodes := {<i, aB, atB, tB, j, aE, atE, tE> in
    NodesWithOrder * NodesWithOrder with i == 1 and j == card(NodesWithOrder)};
36 set TimelineBridges := {<i, aB, atB, tB, j, aE, atE, tE> in NodesWithOrder *
    NodesWithOrder with j == i + 1 and (aB != aE or atB != atE)};
37 set FirstNodes := proj(TimelineBridges, <6, 7, 8>) + proj(
    FirstAndLastOrderedNodes, <2, 3, 4>);
38 set LastNodes := proj(TimelineBridges, <2, 3, 4>) + proj(

```

```

    FirstAndLastOrderedNodes, <6, 7, 8>);
39
40 # Set of backward ground arcs on the timelines associated with airports
41 set BackwardGroundArcs := {<aB, atB, tB, aE, atE, tE> in LastNodes * FirstNodes
    with aB == aE and atB == atE};
42
43 # Set of ground arcs
44 set GroundArcs := ForwardGroundArcs + BackwardGroundArcs;
45
46 # Set of flight arcs in the network
47 set DataForAssign := {read fileAllInfo as "<1s, 2s, 3s, 4s, 5s, 6s, 7n, 8n>"
    comment "#"};
48 set DepartReadyFlights := proj(DataForAssign, <1,2,3,5,6>);
49 set FlightArcs := {<aB, atB, tB, aE, atE, tE> in Nodes * Nodes with <aB, tB, aE
    , atB, tE> in DepartReadyFlights and atE == atB};
50
51 # Set of active flights (ones with possible assignments)
52 set RawFlights := {read fileFlights as "<1s, 2s, 3s>" comment "#"};
53 set ActiveFlights := RawFlights inter proj(FlightArcs, <1, 4, 3>);
54
55 ## CONSTRUCTION FOR OBJECTIVE FUNCTION
56
57 set OldFASolution := {read fileOldFA as "<1s, 2s, 3s, 4s, 5s, 6s, 7n>" comment
    "#"};
58 set OldFlightArcs := proj(OldFASolution, <1,2,3,4,5,6>);
59 param xOld[OldFlightArcs] := read fileOldFA as "<1s, 2s, 3s, 4s, 5s, 6s> 7n"
    comment "#";
60
61 ## VARIABLES
62
63 var x[FlightArcs + OldFlightArcs] binary;
64 var y[GroundArcs] >= 0;
65
66 ## MODELING OBJECTIVE
67
68 minimize HammingCost: sum <aB, atB, tB, aE, atE, tE> in OldFlightArcs: vabs(x[
    aB, atB, tB, aE, atE, tE] - xOld[aB, atB, tB, aE, atE, tE]) + sum <aB, atB,
    tB, aE, atE, tE> in FlightArcs without OldFlightArcs: vabs(x[aB, atB, tB,
    aE, atE, tE]);
69
70 ## MODELING CONSTRAINTS
71
72 # Variables of old indices that are no longer used in the new assignment must
    be zero
73 subto ZeroOld:
74     forall <aB, atB, tB, aE, atE, tE> in OldFlightArcs without FlightArcs do
75         x[aB, atB, tB, aE, atE, tE] == 0;
76
77 # Exactly one aircraft type is assigned to each active flight
78 subto AssignToActiveFlights:
79     forall <aB, aE, tB> in ActiveFlights do
80         sum <aB1, atB, tB1, aE1, atE, tE> in FlightArcs with aB1 == aB and tB1
            == tB and aE1 == aE: x[aB1, atB, tB1, aE1, atE, tE] == 1;
81

```

```

82 # For each aircraft type, the number of used aircrafts is at most the number of
    available aircrafts
83 subto FleetCapacity:
84     forall <at> in AircraftTypes do
85         sum <aB, atB, tB, aE, atE, tE> in BackwardGroundArcs with atB == at
            : y[aB, atB, tB, aE, atE, tE] <= nAircraftsOfType[at];
86
87 # Flow conversation at each node
88 subto FlowConversation:
89     forall <a, at, t> in Nodes do
90         sum <a_fin, at_fin, t_fin> in Nodes with <a_fin, at_fin, t_fin, a,
            at, t> in FlightArcs: x[a_fin, at_fin, t_fin, a, at, t]
91         + sum <a_gin, at_gin, t_gin> in Nodes with <a_gin, at_gin, t_gin, a, at
            , t> in GroundArcs: y[a_gin, at_gin, t_gin, a, at, t]
92     == sum <a_fout, at_fout, t_fout> in Nodes with <a, at, t, a_fout, at_fout,
            t_fout> in FlightArcs: x[a, at, t, a_fout, at_fout, t_fout]
93     + sum <a_gout, at_gout, t_gout> in Nodes with <a, at, t, a_gout,
            at_gout, t_gout> in GroundArcs: y[a, at, t, a_gout, at_gout, t_gout
            ];

```

Table 2.1 summarizes some numerical results of testing the basic fleet assignment model (*BFAM*) on the new part of the problem instances (that is, we did not take into account the information of the previous schedule when generating the new one). The second and the third columns of the table respectively give the number of variables and the number of constraints of the models for each tested instance. The times reported in the table are in seconds. It can be seen from the last column of Table 2.1 that the Basic Fleet Assignment Model is very efficient in solving the deterministic fleet assignment problem. For our tested instances, it gives a fleet assignment solution within one second.

Instances	# variables	# constraints	Modeling time	Solving time
WasFA1	3080	2113	51	0.23
WasFA2	3080	2113	51	0.23
WasFA3	3080	2113	51	0.14

Table 2.1: Performance of Basic Fleet Assignment Model on the tested instances.

Table 2.2 reports the numerical results of testing (*BFAMwas*) on the problem instances WasFA1, WasFA2, WasFA3. In comparison with Table 2.1, it is shown in Table 2.2 that our MIP formulation for the wait-and-see fleet assignment uses much more number of variables and constraints than the BFAM formulation, although the two formulations have similar constraint set. This is because our MIP formulation for the robust fleet assignment problem needs more variables and constraints to linearize the objective function, which is

nonlinear. However, the last column of Table 2.2 shows that, for our tested instances, our proposed MIP formulation for the robust problem performs very well. Therefore, we can conclude the approach of using our MIP formulation is efficient in solving the wait-and-see fleet assignment.

Instances	# variables	# constraints	Modeling time	Solving time
WasFA1	8017	7090	51	0.09
WasFA2	8017	7090	51	0.07
WasFA3	8017	7090	51	0.08

Table 2.2: Performance of formulation ($BFAM_{was}$) on the tested instances for the wait-and-see fleet assignment.

Conclusions

In this thesis we studied two fleet assignment problems: one with deterministic setting and the other in data uncertainty context.

The former problem is studied in Chapter 1, in which we aim to determine which airplane type in the fleets of an airline should be assigned to which flight in a sample time period, taking separate consideration of three objectives: minimize the total cost of the assignment, maximize the total revenue of the assignment, and minimize the number of used airplanes. Following [3], a time-expanded multi-commodity network is constructed so that the solution to this problem can be viewed as flows in the network. Thanks to that, we obtained a mixed integer programming formulation, which is called basic fleet assignment model (BFAM), for the problem. Numerical experiments on a small-size problem instance are described in detail in Chapter 1 showed the validity and efficiency of the formulation.

The latter problem is studied in Chapter 2, in which we aim to construct a new fleet assignment when the old one is no longer valid due to some changes in input data. The new fleet assignment is required to have the least number of changes in comparison with the old one, in order to reduce the induced cost. We approached this problem with a view from robust optimization paradigm. We proposed a new robustness concept so-called wait-and-see recovery robustness. This concept aims to find a solution to an uncertain optimization problem after the realization of the uncertain parameter, so that it can be recovered from an existed solution with the minimum recovery cost. The uncertain fleet assignment can be viewed as an application of the wait-and-see recovery robustness concept. As a solution approach, we constructed a mixed integer programming formulation for the considered problem. The first key idea in the formulation construction is to use a basic fleet assignment model (BFAM) to find a feasible fleet assignment solution in the setting of the new data. The second key idea is to use the Hamming distance as the recovery cost in the objective function, which can be easily linearized. Numerical experi-

ments on medium-size problem instances showed the validity and efficiency of our formulation for this problem.

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