

## **CHAPTER THREE**

### **AIRCRAFT LANDING PROBLEMS (ALP)**

#### **3.1 Introduction**

**Aircraft Landing Problem (ALP):** Given a set of planes in the radar horizon of an air-traffic controller (ATC), the problem is one of determining a landing time at a busy airport for each plane such that each plane in this ATC horizon lands within a prespecified landing time window and such that landing separation criteria specified for each pair of planes in this horizon are adhered to.

In this chapter, we study the aircraft landing problem (ALP), which is considered as one of a COPs, in a single runway case. We present in the first part, a mathematical formulation of the problem with a linear objective function. In the second part, we consider the static case of the problem where all data are known in advance. We present a new heuristic for scheduling static case of aircraft landing to solve the ALP.

#### **3.2 Aircraft Landing Problem (ALP)**

##### **3.2.1 ALP Motivation**

When the number of approaching flights exceeds the airport capacity, some of these aircraft can't be landed on its 'perfect' landing time. However, some costs are being considered:

- There is a cost mainly on the waste of fuel for each plane flying faster than its most economical speed.
- Airlines also experience different costs for delays of different flights.
- Depending on the amount of delay, there might be a number of transfer passengers that miss their connecting flight.

- The crew or aircraft might also be needed to perform a next flight, which now has to be rescheduled.
- This might propagate delays to departing flights.

### 3.2.2 Description and Notations of the ALP

Assumptions of ALP are as follows:

- The set of aircrafts which are waiting to be landed is known, a static model.
- There are several runways in the airport.
- The sets of aircrafts including the target time and time window are waiting to be landed on the runway.
- The cost is considered for each unit of tardiness or earliness for the target time of every aircraft.
- Each aircraft is supposed to land on a determined runway, when the limitation of separation time (the time between this aircraft and previous ones which land on this runway or others) is satisfied.
- All aircrafts are not equal and similar to each other and there are different aircrafts.

The **objective function** of the problems is to minimize the deviation of target time for each aircraft. As when an aircraft lands sooner than the target time, it causes problems for other aircrafts flight schedules. Now, we shall assume that we are minimizing total cost, where the cost for any plane is linearly related to deviation from its target time. Figure (3.1) illustrates the variation in cost within the time window of a particular plane.

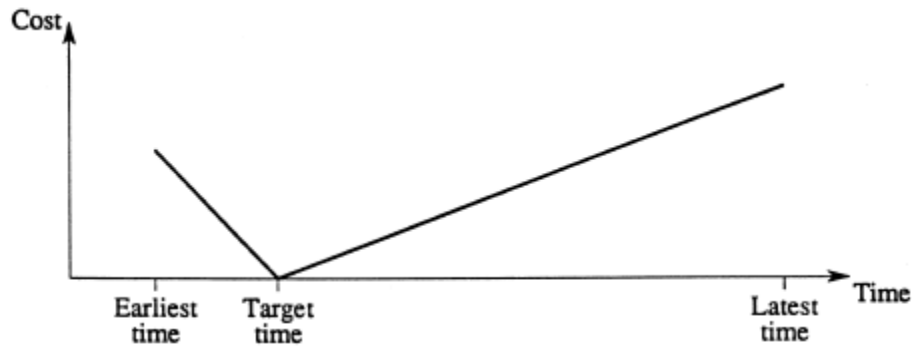


Figure (3.1): Variation in cost for a plane within its time window.

The aircraft generally partitioned into three weight classes: Small, Large and Heavy. The time separation requirements are then a function of the plane speed and the length of the final approach path.

We first observe that the ALP involves two decision problems: (1) a **sequencing** problem (which determines the sequence of plane landings) and (2) a **scheduling** decision problem (which determines the precise landing times for each plane in the sequence, subject to the separation constraints).

The ALP has the following notations:

- $N$  : the number of planes.
- $P$  : Set of  $N$  planes,  $P = \{1, 2, \dots, N\}$ .
- $R$  : the number of landing runways (here we take  $R=1$ ).
- $E_i$  : the earliest landing time for plane  $i \in P$ .
- $L_i$  : the latest landing time for plane  $i \in P$ .
- $T_i$  : the target (preferred) landing time for plane  $i \in P$ .
- $S_{ij}$  : the required separation time ( $\geq 0$ ) between plane  $i$  landing and plane  $j$  landing (where plane  $i$  lands before plane  $j$ ),  $i, j \in P$ ,  $i \neq j$ .
- $g_i$  : the penalty cost ( $\geq 0$ ) per unit of time for landing before the target time  $T_i$  for plane  $i \in P$ .
- $h_i$  : the penalty cost ( $\geq 0$ ) per unit of time for landing after the target time  $T_i$  for plane  $i \in P$ .

The variables are:

- $t_i$  : the actual landing time for plane  $i \in P$ .

$\alpha_i$  : how soon plane  $i \in P$  lands before  $T_i$ , mathematically, the earliness  
 $\alpha_i = \max\{0, T_i - t_i\}$ .

$\beta_i$  : how soon plane  $i \in P$  lands after  $T_i$ , mathematically, the tardiness  
 $\beta_i = \max\{0, t_i - T_i\}$ .

$$\delta_{ij} = \begin{cases} 1 & \text{if plane } i \text{ lands before plane } j \quad \forall i, j \in P, i \neq j. \\ 0 & \text{otherwise.} \end{cases}$$

Without significant loss of generality, we shall henceforth assume that the times  $E_i$ ,  $L_i$ , and  $S_{ij}$  are integers.

### 3.3 Single Runway Formulation of ALP

In this section, we present an initial mixed-integer zero-one formulation of the static single runway ALP.

#### 3.3.1 ALP Constraints

The first set of constraints are

$$E_i \leq t_i \leq L_i, \quad \forall i \in P, \quad \dots(3.1)$$

which ensure that each plane lands within its time window. Now, considering pairs  $(i, j)$  of planes we have that

$$\delta_{ij} + \delta_{ji} = 1, \quad \forall i, j \in P, i < j, \quad \dots(3.2)$$

We need to define three sets:

- U : the set of pairs  $(i, j)$  of planes for which we are uncertain whether plane  $i$  lands before plane  $j$ .
- V : the set of pairs  $(i, j)$  of planes for which  $i$  definitely lands before  $j$  (but for which the separation constraint is not automatically satisfied).
- W : the set of pairs  $(i, j)$  of planes for which  $i$  definitely lands before  $j$  (and for which the separation constraint is automatically satisfied).

Then, we can define the set  $W$  by

$$W = \{(i,j) \mid L_i < E_j \text{ and } L_i + S_{ij} \leq E_j, \forall i,j \in P, i \neq j\} \quad \dots(3.3)$$

In words,  $i$  must land before  $j$  ( $L_i < E_j$ ) and the separation constraint is automatically satisfied ( $L_i + S_{ij} \leq E_j$ ).

We can define the set  $V$  by

$$V = \{(i,j) \mid L_i < E_j \text{ and } L_i + S_{ij} > E_j, \forall i,j \in P, i \neq j\} \quad \dots(3.4)$$

In words,  $i$  must land before  $j$  ( $L_i < E_j$ ) but the separation constraint is not automatically satisfied ( $L_i + S_{ij} > E_j$ ).

Some plane lands first may have overlapping time windows. Hence, we can define the set  $U$  as:

$$U = \{(i,j) \mid i,j=1,\dots,N, i \neq j; E_j \leq E_i \leq L_j \text{ or } E_j \leq L_i \leq L_j, E_i \leq E_j \leq L_i \text{ or } E_i \leq L_j \leq L_i\} \dots(3.5)$$

We need a separation constraint for pairs of planes in  $V$ , and this is

$$t_j \geq t_i + S_{ij} \quad \forall (i,j) \in V \quad \dots(3.6)$$

which ensures that a time  $S_{ij}$  must elapse after the landing of plane  $i$  at  $t_i$  before plane  $j$  can land at  $t_j$ .

Finally, we need constraints to relate the  $\alpha_i$ ,  $\beta_i$ , and  $t_i$  variables to each other.

$$T_i - t_i \leq \alpha_i \leq T_i - E_i, \quad i \in P, \quad \dots(3.7)$$

$$t_i - T_i \leq \beta_i \leq L_i - T_i, \quad i \in P, \quad \dots(3.8)$$

### 3.3.2 Objective Function

First, we have the following objective function:

$$\text{minimize } Z_1 = \sum_{i=1}^N t_i \quad \dots(3.9)$$

which mean the sum of all actual times of all aircrafts.

We now need only to setup the objective function, minimize the deviation from the target times ( $T_i$ ), and this is

$$\text{minimize } Z_2 = \sum_{i=1}^N (g_i \alpha_i + h_i \beta_i) \quad \dots(3.10)$$

Lastly we have the following multi-Criteria Objective function (MCOF):

$$\text{minimize } Z_3 = (Z_1, Z_2) \quad \dots(3.11)$$

The complete formulation (model) of the single runway problem is therefore to satisfy function (3.9), (3.10) and (3.11) subject to relations (3.1), (3.2), and (3.6)-(3.8).

**Example (3.1):** For  $N=3$ , lets have the following ALP information.

	$P_1$	$P_2$	$P_3$
$E_i$	129	195	89
$T_i$	155	258	98
$L_i$	559	744	510
$g_i$	10	10	30
$h_i$	10	10	30

	$S_{ij}$		
	1	2	3
1	0	3	15
2	3	0	15
3	15	15	0

Formulate this ALP:

$$Z_1 = \sum_{i=1}^N t_i$$

$$Z_2 = \sum_{i=1}^3 (g_i \alpha_i + h_i \beta_i)$$

$$Z_1 = (10 * \max\{0, 155 - t_1\} + 10 * \max\{0, t_1 - 155\}) + (10 * \max\{0, 258 - t_2\} + 10 * \max\{0, t_2 - 258\}) + (30 * \max\{0, 98 - t_3\} + 10 * \max\{0, t_3 - 98\})$$

s.t.

$$129 \leq t_1 \leq 559, 195 \leq t_2 \leq 744, 89 \leq t_3 \leq 510,$$

$$\delta_{ij} + \delta_{ji} = 1, \forall i, j \in P, i < j,$$

$$t_j \geq t_i + S_{ij} \quad \forall (i, j) \in V,$$

$$155 - t_1 \leq \max\{0, 155 - t_1\} \leq 26, 258 - t_2 \leq \max\{0, 258 - t_2\} \leq 63, 98 - t_3 \leq \max\{0, 98 - t_3\} \leq 9, \\ t_1 - 155 \leq \max\{0, t_1 - 155\} \leq 404, t_2 - 258 \leq \max\{0, t_2 - 258\} \leq 486, t_3 - 98 \leq \max\{0, t_3 - 98\} \leq 412.$$

Calculate the objective functions  $Z_1$  and  $Z_2$  if  $t_i = 150, 250, 100$ :

$Z_1 = 500$ , while for  $Z_2$  we have:

$$\alpha_i = \max\{0, T_i - t_i\}, i = 1, 2, 3, \text{ then:}$$

$$\alpha_i = \max\{0, 155 - 150\} = 5, \max\{0, 258 - 250\} = 8, \max\{0, 98 - 100\} = 0.$$

And  $\beta_i = \max\{0, t_i - T_i\}, i = 1, 2, 3$ , then:

$$\beta_i = \max\{0, 150 - 155\} = 0, \quad \max\{0, 250 - 258\} = 0, \quad \max\{0, 100 - 98\} = 2.$$

$$Z = (g_1\alpha_1 + h_1\beta_1) + (g_2\alpha_2 + h_2\beta_2) + (g_3\alpha_3 + h_3\beta_3) = 10 \times 5 + 10 \times 8 + 30 \times 2 = 190.$$

Lastly for  $Z_3$ , we have  $Z_3 = (Z_1, Z_2) = (500, 190)$ .

### 3.4 Techniques to Improve the Solution and Reduce the Computations

In this section we demonstrate two types of methods which contribute in improving the solution and speed the approach to the good solution. In addition, we will discuss some special cases of ALP.

#### 3.4.1 Time Window Tightening (TWT)

Let  $Z_{UB}$  (for any three objective functions) be any upper bound to the problem. Then, it is possible to limit the deviation from target for each plane. Specifically, for plane  $i$ , we can update  $E_i$  using:

$$E_i = \max\{E_i, T_i - Z_{UB}/g_i\}, \quad i \in P, \quad \dots(3.12)$$

Similarly we have

$$L_i = \min\{L_i, T_i + Z_{UB}/h_i\}, \quad i \in P, \quad \dots(3.13)$$

The benefit of tightening (closing) the time windows is that (potentially) the sets  $U$  and  $V$  can be reduced in size, thereby giving a smaller problem to solve.

**Example (3.2):** The time window tightening of example (3.1) using Eq. (3.12) and (3.13). for instance,  $Z_{UB} = 1060$  (using  $Z_2$ ) we have:

$$E_i = \max\{E_i, T_i - 106\} \text{ where: } E_1 = \max\{129, 155 - 106\} = 129,$$

$$E_2 = \max\{195, 258 - 106\} = 195, \quad E_3 = \max\{89, 98 - 35\} = 98.$$

$$L_i = \min\{L_i, T_i + 106\} \text{ where: } L_1 = \min\{559, 155 + 106\} = 261,$$

$$L_2 = \min\{744, 258 + 106\} = 364, \quad L_3 = \min\{89, 98 + 35\} = 133.$$

These results are shown in table (3.1).

Table (3.1): time window tightening of example (3.2) for  $Z_{UB}=1060$ .

	$P_1$	$P_2$	$P_3$
$E_i$	129	195	89
$T_i$	155	258	98
$L_i$	261	364	133
$g_i$	10	10	30
$h_i$	10	10	30

**Exercise (3.1):** calculate the TWT for:

1. from example (3.1),  $Z_{UB}=900$ .
2. from Appendix, for  $N=10$ , for 1<sup>st</sup> 5 aircraft,  $Z_{UB}=90$ .
3. from Appendix, for  $N=15$ , for 1<sup>st</sup> 5 aircraft,  $Z_{UB}=90$ .

### 3.4.2 Successive Rules (SR)

Reducing the current sequence is done by using several SR's. When, for each  $i$  ( $i \in P$ ), and with its cost given in the objective function (3.9), we can derive SR that restrict the search for an optimal solution. Such rules can be used in some optimization algorithms. These improvements lead to very large decrease in the number of solutions to obtain the optimal solution.

**Definition (3.1):** Let  $W_i=[E_i,L_i]$  be the time window interval of plane  $i \in P$ , if  $W_i \cap W_j = \emptyset$  (time windows are disjoint) and  $L_i < E_j$  we denote for the interval  $W_i$  precedes the interval  $W_j$  in line number by  $W_i \prec W_j$ .

**Definition (3.2):** We say that plane  $i$  precedes the plane  $j$  (we write  $i \rightarrow j$  or  $(i,j) \in W$ ) or  $j$  precedes the plane  $i$  if  $W_i \cap W_j = \emptyset$ , for  $i \neq j$ .

**Remark (3.1):**

1.  $t_i < t_j$  and  $t_j \geq t_i + S_{ij}$  if and only if  $i \rightarrow j$ ,  $\forall i, j \in P$ ,  $i \neq j$ .
2. if  $E_i \leq E_j \leq L_i$  or  $E_i \leq L_j \leq L_i$ , then  $W_i \cap W_j \neq \emptyset$  for  $i \neq j$ , we say that  $W_i$  and  $W_j$  are overlapped.



**Proposition (3.1):** if  $W_i \supset W_j$ , then  $t_i \in W_i < t_j \in W_j, \forall i, j \in P, i \neq j$ .

**Proof:** since  $W_i \supset W_j$ , then  $t_i \notin W_j$  and  $t_j \notin W_i$ . Suppose  $t_i \geq t_j$ , for  $t_i = t_j, t_j = t_i \in W_i, C!$ . For  $t_i > t_j$ , if  $t_j \in W_i$  C!. Take  $t_j \notin W_i$ . Then  $t_j \in$  another interval say  $W_k$ , s.t.  $W_k \supset W_j$ , but  $t_j \in W_j$  and that is a contradiction since there is no integer belong to two disjoint intervals in the same time. Then  $t_i < t_j$   $\square$

**Remark (3.2):** if  $W_i \cap W_j = \emptyset$ , then  $L_i < E_j$  or  $L_j < E_i, \forall i, j \in P, i \neq j$ .

**Definition (3.3):** the  $i \rightarrow j$  if one of the following conditions is satisfied:

1.  $L_i < E_j$  for  $i \neq j$ .
2. For  $L_i \geq E_j$ , if  $L_i < E_j + S_{ij}$  for  $i \neq j$ .

Conditions of SR are shown in figure (5.2).

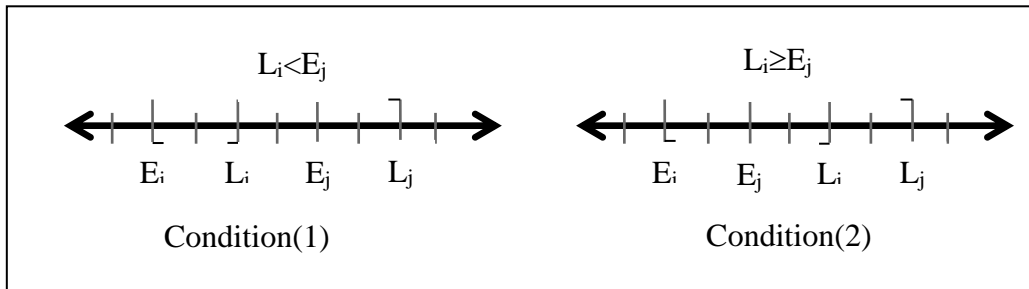


Figure (3.2): Conditions of dominance rules.

**Example (3.3):** For  $N=5$  let's have the following ALP information:

	$P_1$	$P_2$	$P_3$	$P_4$	$P_5$
$E_i$	129	89	96	111	123
$T_i$	155	98	106	123	135
$L_i$	191	110	118	135	147
$g_i$	10	30	30	30	30
$h_i$	10	30	30	30	30

$S_{ij}$	1	2	3	4	5
1	0	15	15	15	15
2	15	0	8	8	8
3	15	8	0	8	8
4	15	8	8	0	8
5	15	8	8	8	0

From definition (3.3), condition (1) we obtain the following SR's:

$2 \rightarrow 1, 2 \rightarrow 4, 2 \rightarrow 5, 3 \rightarrow 1, 3 \rightarrow 5$ .

From condition (2), we have  $3 \rightarrow 4$  because of  $E_4 + S_{34} = 111 + 8 = 119 > L_3 = 118$ , and  $4 \rightarrow 1$  because of  $E_1 + S_{41} = 129 + 15 = 144 > L_4 = 135$ . Figure (3.3) shows the SR's of example (3.3).

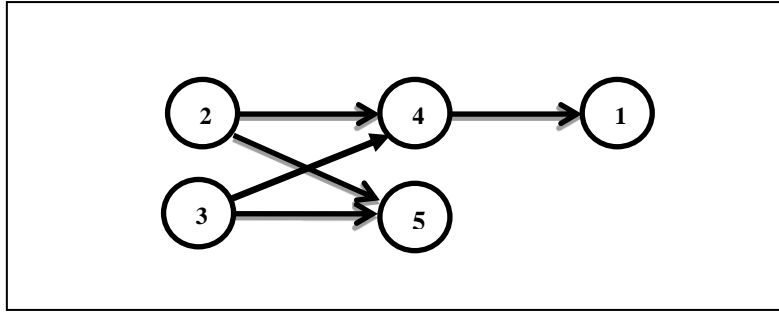


Figure (3.3): Graph of SR of example (3.3).

The adjacency matrix  $A$  of the graph shown above is:

$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} 0 & 0 & 0 & 0 & \delta_{15} \\ 1 & 0 & \delta_{23} & 1 & 1 \\ 1 & \delta_{32} & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 & \delta_{45} \\ \delta_{51} & 0 & 0 & \delta_{54} & 0 \end{bmatrix} \end{matrix}$$

**Note:**

- $\delta_{15} + \delta_{51} = 1$ ,  $\delta_{23} + \delta_{32} = 1$ ,  $\delta_{45} + \delta_{54} = 1$
- the sequencing problem of this ALP can be solved by  $2^3 = 8$  possible and no need to try  $5! = 120$  possible.

**Example (3.4):** Find the possible sequences for example (3.3):

From adjacency matrix  $A$ , we have  $(\delta_{15}, \delta_{23}, \delta_{45})$ ,  $1 \leftrightarrow 5$ ,  $2 \leftrightarrow 3$  and  $4 \leftrightarrow 5$ .

So we have:

<b>i</b>	<b><math>(\delta_{15}, \delta_{23}, \delta_{45})</math></b>	<b>Subsequence</b>	<b>sequence</b>	<b>Acceptance</b>
1.	(0,0,0)	5→1,3→2,5→4	3→2→5→4→1	✓
2.	(0,0,1)	5→1,3→2,4→5	3→2→4→5→1	✓
3.	(0,1,0)	5→1,2→3,5→4	2→3→5→4→1	✓
4.	(0,1,1)	5→1,2→3,4→5	2→3→4→5→1	✓
5.	(1,0,0)	1→5,3→2,5→4	3→2→1→5→4	✗
6.	(1,0,1)	1→5,3→2,4→5	3→2→4→1→5	✓
7.	(1,1,0)	1→5,2→3,5→4	2→3→1→5→4	✗
8.	(1,1,1)	1→5,2→3,4→5	2→3→4→1→5	✓

### 3.4.3 Special Cases

**Definition (3.4):** Let  $S = \max\{S_{ij}\}$ ,  $\forall i, j \in P, i \neq j$ , then  $W_i$  is called **logical time window** if the length  $\ell_i$  of  $W_i$ , for  $i \in P$  is  $\ell_i = L_i - E_i + 1 \geq 2S$  and  $T_i = (E_i + L_i)/2$ .

**Example (3.3):** let  $W_1 = [10, 20]$  and  $W_2 = [25, 50]$ ,  $S_{12} = 10$ ,  $S = 10$ . Note that  $\ell_1 = 11$  and  $\ell_2 = 26$ ,  $W_2$  is logical time window but  $W_1$  is not. While if  $W_1 = [10, 15]$  and  $W_2 = [16, 24]$ ,  $S_{12} = 15$ ,  $S = 15$ . Note that both  $W_1$  and  $W_2$  are not logical time windows, since if  $t_1 = E_1 = 10$ , then  $t_2 < t_1 + S_{12} = 10 + 15 = 25 > L_2 = 24$ , that mean  $W_2$  is not logical time definitely, not satisfies the separation constraint.

**Case (1):** Let  $W_{i_1}, W_{i_2}, \dots, W_{i_N}$  are all disjoint logical time windows in this sequence s.t.  $W_{i_k} \cap W_{i_j} = \emptyset, \forall i_k, i_j \in P, i_k \neq i_j$ , then the optimal solution with cost  $Z=0$  at  $t_{i_1} = T_{i_1} < t_{i_2} = T_{i_2} < \dots < t_{i_N} = T_{i_N}$  and  $i_1 \rightarrow i_2 \rightarrow \dots \rightarrow i_N$ .

**Proof:** Without losing the generality, let  $N=3$  to show  $Z=0$  and  $1 \rightarrow 2 \rightarrow 3$ . Since  $W_1, W_2$  and  $W_3$  are logical time windows this mean  $S = \max\{S_{ij}\}$ ,  $\forall i, j \in P$ . Let  $t_1 = T_1$ ,  $T_1 + S \leq L_1 < E_2 < T_2$ , then take:

$$t_2 = T_2 > T_1 + S = t_1 + S \quad \dots (a)$$

$\therefore t_1 = T_1$  and  $t_2 = T_2$  satisfy the window and separation conditions (WSC's). By applying relation (a) again for  $t_2$  and  $t_3$  we obtain that:  $t_2 = T_2$  and  $t_3 = T_3$  satisfy the WSCs.

$\therefore$  The optimal solution with cost  $Z=0$  for  $N=3$  and  $1 \rightarrow 2 \rightarrow 3$ .

Consequently, this case can be applied for  $N$  aircraft and for any sequence  $\pi$ .  $\square$

**Case (2):** Let  $W = W_1 = W_2 = \dots = W_N$  be the same large time window, then the optimal solution  $Z=0$  at  $t_{i_k} = T_{i_k}$  if  $T_{i_k}$  satisfies the separation constraint  $\forall i_k \in P$  and  $i_1 \rightarrow i_2 \rightarrow \dots \rightarrow i_N$ .

**Proof:** let's take any arbitrary sequence  $\pi$ . Since  $T_{i_k}$  satisfy the separation constraints, this means:  $T_1 \leq T_2 - S_{12}$ ,  $T_2 \leq T_3 - S_{23}, \dots, T_{N-1} \leq T_N - S_{N-1,N}$ . If we take  $t_{i_k} = T_{i_k}$ , then the landing times  $t_{i_k}$  satisfy the separation constraint  $\forall i_k \in P$ .

∴ The optimal solution with cost  $Z=0$  and  $1 \rightarrow 2 \rightarrow \dots \rightarrow N$ .  $\square$

Of course, this case can be applied for any sequence  $\pi$ .

**Exercise (3.2):** Find the SR for:

1. For  $N=5$  let's have the following ALP information:

	$P_1$	$P_2$	$P_3$	$P_4$	$P_5$	$S_{ij}$	1	2	3	4	5
$E_i$	129	111	123	89	96	1	0	15	15	15	15
$T_i$	155	123	135	98	106	2	15	0	8	8	8
$L_i$	191	135	147	110	118	3	15	8	0	8	8
$g_i$	10	30	30	30	30	4	15	8	8	0	8
$h_i$	10	30	30	30	30	5	15	8	8	8	0

2. For  $N=5$  let's have the following ALP information:

	$P_1$	$P_2$	$P_3$	$P_4$	$P_5$	$S_{ij}$	1	2	3	4	5
$E_i$	146	241	90	95	108	1	0	3	15	15	15
$T_i$	155	250	93	98	111	2	3	0	15	15	15
$L_i$	164	259	96	101	114	3	15	15	0	8	8
$g_i$	10	10	30	30	30	4	15	15	8	0	8
$h_i$	10	10	30	30	30	5	15	15	8	8	0

3. For  $N=5$  let's have the following ALP information:

	$P_1$	$P_2$	$P_3$	$P_4$	$P_5$	$S_{ij}$	1	2	3	4	5
$E_i$	241	146	108	90	95	1	0	3	15	15	15
$T_i$	250	155	111	93	98	2	3	0	15	15	15
$L_i$	259	164	114	96	101	3	15	15	0	8	8
$g_i$	10	10	30	30	30	4	15	15	8	0	8
$h_i$	10	10	30	30	30	5	15	15	8	8	0