

Ramp Activity Expert System for Scheduling and Co-ordination at an Airport

Geun-Sik Jo

Inha University,
Dept. of Computer Science
and Engineering
Inchon, 402-751, Korea
gsjo@dragon.inha.ac.kr

Kang-Hee Lee

Inha University,
Dept. of Computer Science
and Engineering
Inchon, 402-751,
Korea

Hwi-Yoon Lee

Korean Air
Operations Control Center
Kangseo-Ku, Seoul,
157-220
Korea

Sang-Ho Hyun

Korean Air
Information Systems Services
and Development Dept.
Kangseo-Ku, Seoul, 157-220
Korea

Abstract

In this project, we have developed the Ramp Activity Coordination Expert System (RACES) in order to solve aircraft parking problems. RACES includes a knowledge-based scheduling system which assigns all daily arriving and departing flights to the gates and remote spots with domain specific knowledge and heuristics acquired from human experts. RACES processes complex scheduling problems such as dynamic inter-relations among the characteristics of remote spots/gates and aircraft with various other constraints, for example, customs and ground handling factors at an airport. By user-driven modeling for end users and near optimal knowledge-driven scheduling acquired from human experts, RACES can produce parking schedules for about 400 daily flights in approximately 20 seconds, whereas it normally takes human experts 4 to 5 hours to do the same. Scheduling results in the form of Gantt charts produced by RACES are also accepted by the domain experts. RACES is also designed to deal with the partial adjustment of the schedule when unexpected events occur. After daily scheduling is completed, the messages for aircraft changes and delay messages are reflected and updated into the schedule according to the knowledge of the domain experts. By analyzing the knowledge model of the domain expert, the reactive scheduling steps are effectively represented as the rules, and the scenarios of the Graphic User Interfaces (GUI) are designed. Since the modification of the aircraft dispositions, such as aircraft changes and cancellations of flights, are reflected in the current schedule, the modification should be sent to RACES from the mainframe for the reactive scheduling. The adjustments of the schedule are made semi-automatically by RACES since there are many irregularities in dealing with the partial rescheduling.

PROBLEM DESCRIPTION

The aircraft-parking problem is a scheduling problem which entails assigning every arriving and departing flight to in-terminal gates and remote gates satisfying various demands. It is a kind of scheduling problem which also includes job-shop scheduling if we consider the (remote)

gates as machines and incoming flights as jobs. In addition, this problem entails characteristics of temporal reasoning mechanisms. Theoretically, this problem belongs to the *NP*-class of problems in computational complexity. That is, if we try to assign m flights to n remote parking spots or gates (often referred to as bridges), then a non-polynomial number of combinations $(m!)^n$ are possible. The scheduling becomes more dynamic and difficult if the frequency of arriving and departing flights increases and unexpected events occur during the operational stage in a given time period. Increased air traffic and customs requirements also reflect difficulties in producing parking schedules. In practice, professional schedulers manually make the schedules once a day. This requires domain-specific knowledge, experience and heuristics and requires a considerable amount of time and tedious paper work to complete.

Traditionally researchers have used mathematical programming techniques to solve these kinds of problems. However, it is very difficult to model constraints and domain knowledge with only mathematical variables. In addition, there may be serious problems of remodeling and processing when unexpected events occur for large-scale practical problems. Recently, many researchers have proposed using AI techniques such as constraint directed reasoning, expert systems and Constraint Satisfaction Problem(CSP) to solve these problems (Jo, Jung, and Yang 1997, Fox 1987, Prosser 1993). AI techniques provide more flexible and expressive power than mathematical programming in modeling a complex scheduling problem. Comparison between Integer Programming and AI is well described in another work (Dhar and Ranganathan. 1990). In addition, modeling the reactive scheduling with Integer Programming is beyond our ability to formulate due to the dynamic addition of constraints and the necessity of the partial solutions.

Many expert systems have been presented in these areas. Practical expert systems have developed in the airline industry. American Airlines developed GateManager to effectively manage busy air traffic and resources (American Airlines 1993). The GATES system from Texas State University controls gate assignment and

tracking in New York's JFK airport (Brazile and Swigger 1988). Knowledge Engineering, a Singapore-based company, has successfully developed a constraint-based gate allocation system using ILOG for Changi airport in Singapore (Berger 1995).

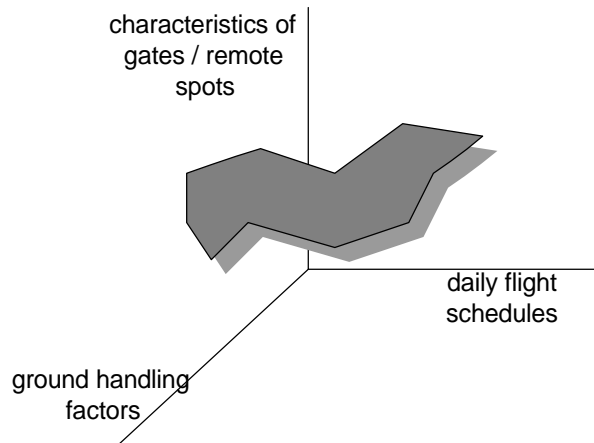


Figure 1. The feasibility region in gate allocation

DESCRIPTION OF RACES

RACES assigns daily flights to the gates and remote spots with domain specific knowledge and scheduling heuristics. Using domain filtering techniques, we can remove the inconsistency in the domain for variables and confine the search space. To find a user-driven optimal solution, RACES utilizes an efficient heuristic scheduling method to satisfy constraints. RACES produces a near optimal schedule with considerations for the flight schedules, aircraft type, characteristics of remote spots, and conditions for ground handling. Aircraft have to be assigned to adequate remote spots and gates with the satisfaction of given constraints. In addition, gates and remote spots are distinguished by size and hydrant facilities. RACES can be viewed as a three dimensional constraint solver as in figure 1, and it also maps three dimensional spaces into the two dimensional spaces which is represented in the form of a Gantt chart. RACES makes user-dependent, near-optimal schedules satisfying the given constraints with domain specific knowledge and heuristics. RACES produces the Gantt chart which represents the daily parking schedule a day beforehand.

In terms of constraint solving, there are three different types of constraints to satisfy. The first is the *strong-hard* constraint which has to be satisfied during the scheduling process. If this constraint is violated, then the solution is no longer valid. The second is the *weak-hard* constraint which can be violated in specific environments. This constraint can be violated by interacting with users, but not by RACES. The last is called a *soft-constraint* which is

applied to specific flights and specific times. During scheduling, a *soft-constraint* is checked and there is an attempted to satisfy the constraint if possible. If not, this type of constraint can be relaxed by RACES. RACES can be divided into two different knowledge-based systems. The first is to generate a one day schedule one day beforehand, which is described in the next section. The second is to adjust a schedule during an operational day. More technical details with examples for the management of constraints and the representation of domain knowledge is presented in another work (Jo, Jung, and Yang 1997).

INITIAL SCHEDULE GENERATION

After we have completed our documentation of knowledge acquired from domain experts, this knowledge is about 20 pages long, which does not include the database description. Moreover, the knowledge itself is too domain specific for the average person to understand. In this section, we describe the scheduling strategy being deployed in our system. In figure 2, RACES produces the Gantt chart which represents the daily parking schedule a day beforehand with today's schedule for co-ordination.



Figure 2. Generation of the initial schedule in RACES

Consistency by Domain Filtering

One scheduling procedure is responsible for binding continuous time values to the discrete time variables in order to satisfy constraints for the time restriction. To deal with the continuous time domain, we break down the continuous time values into discrete time elements. We also classify all the available remote spots and gates with *time-keys*. When the system processes scheduling, it filters domains and removes elements violating

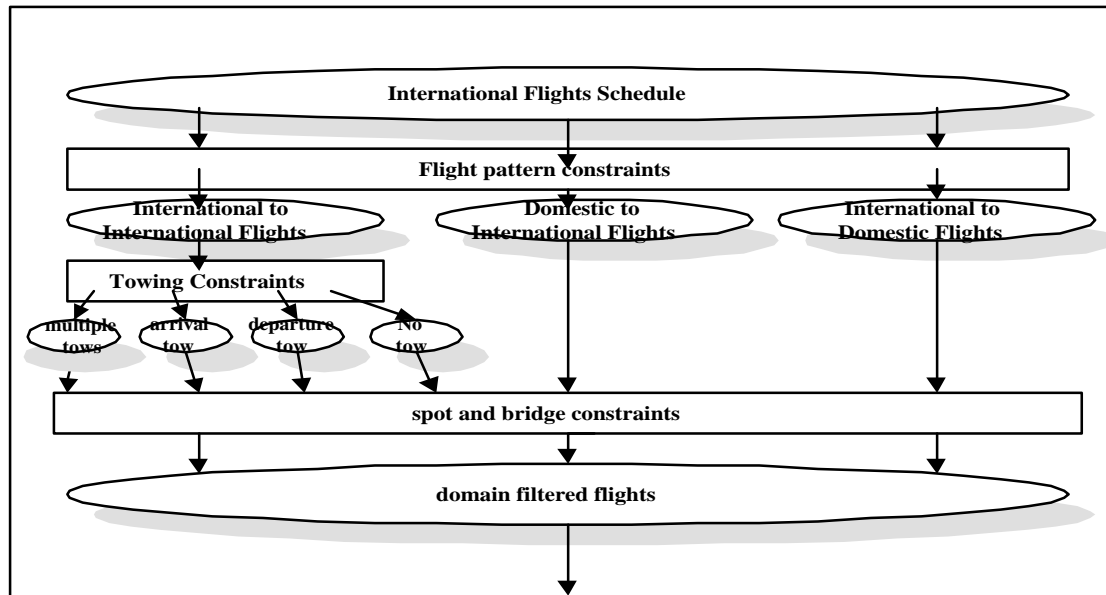


Figure 3. domain filtering before scheduling

constraints in three steps. An example of the domain filtering process for flights schedules is shown in Figure 3.

In the first step, RACES filters domains with the knowledge and constraints of various aircraft. Second, the system filters domains with the knowledge and constraints of towing. Finally, the system filters domains with the knowledge and constraints of available parking spots. As a result of the process, RACES can prune the search space significantly.

Knowledge-driven near optimal scheduling

We consider the optimal solution in terms of the user's benefit. An important factor is to minimize the number of *stand-by* flights which are not yet assigned to the gates/spots due to conjection at the airport. During scheduling, RACES also tries to allow for the least number of towings at the airport. If the system fails to assign flights to the gates, the size of the aircraft must be taken into consideration. If *stand-by* flights involve relatively large aircraft, it is difficult for users to assign them manually after the automatic schedule is produced since the number of large spots is usually inadequate.

There are two heuristic scheduling methods in RACES. One is the time-focused method using best-fit assignment in terms of the time-span for parking. The other is the aircraft-size focused method. Each method has advantages and disadvantages in finding user-driven optimal solutions given that one method conflicts with the other method at certain points during the scheduling process. In our work, to avoid conflicts between the time-span focused method and the size-focused method, we have empirically found and exploited a trade-off point between the two. A detailed

description of these heuristic scheduling methods is presented in another work (Jo, Jung, and Yang 1997).

KNOWLEDGE-BASED REACTIVE SCHEDULING

When the real operational day is reached after the scheduling has been completed, unexpected events can occur as the environment changes. If sudden changes in schedules occur, such as the delay of an aircraft or the change of aircraft for certain flights, then schedules must be adjusted. In adjusting the daily schedules, the domain specific knowledge from domain expert is encoded into RACES.

Expert Model in Reactive Scheduling

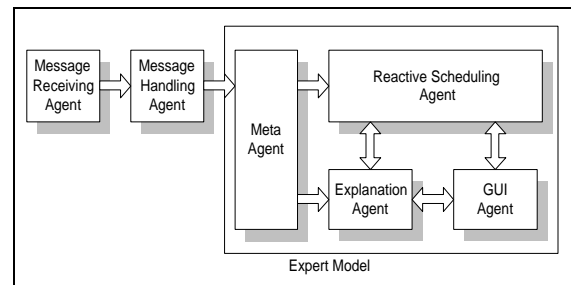


Figure 4. Expert model in reactive scheduling

The meta-agent in Figure 4 has the knowledge of which agent is activated depending on the message from the

message handling agent. Although the role of each agent can not be summarized due to the page requirement, message handling agents and reactive scheduling agents are explained in brief.

The task of the message handling agent is to check sling, detect cycle and group the related messages together. The messages for AC change come into RACES in a unit of flights which are not in an ordered form, but rather in mixed forms. Therefore, the message handling agent has to rearrange them into a unit with an HL number by their scheduled time. Then the agent can check the fallacy of a sling order or an omitted sling. Generally, one aircraft makes a sequence of flight during a given day. A sling is a sequence of flights that an aircraft should make. A sling consists of flight schedule that reflects a continuous time domain.

The task of the reactive scheduling agent is to reschedule according to messages received as the environment changes during the operation. This task can be divided into two different operations: 1) Creating new standby bars. 2) Assigning these bars. The most important task of the reactive scheduling agent is to assign new standby bars to the Gantt chart. This process requires complicated and delicate domain knowledge. A variety of adjusting rules are implemented for these processes.

Interactive Graphic User Interfaces

The change of aircraft(AC) occurs in a case when an aircraft can not operate the flight that it is scheduled to run due to a delay of operations, aircraft repair, or a breakdown. In this case, the flight will be operated by another aircraft, and we call this situation an *AC change*. For an example, see Figure 5.

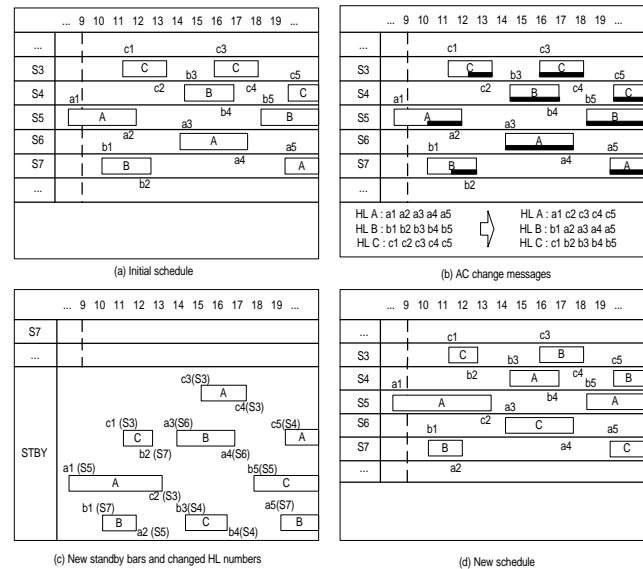


Figure 5. The steps for adjusting an initial schedule

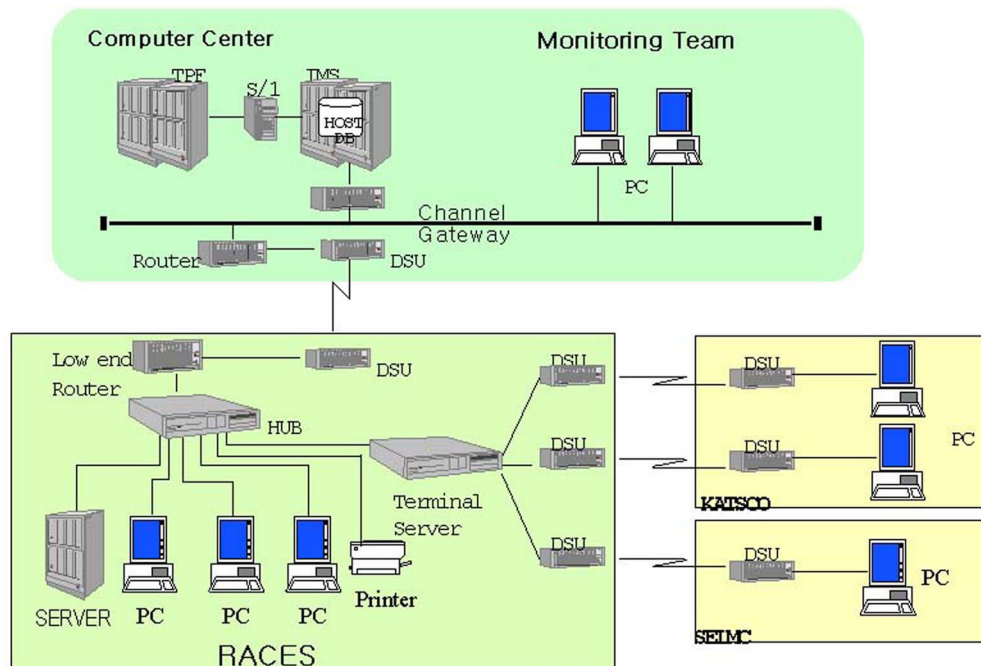


Figure 6. The relationship diagram with other systems

In Figure 5, the capital letters stand for the HL number which is the registration number for an aircraft, and the lower case letters stand for flight numbers. In the original scheduling, aircraft *A* was supposed to arrive as flight *a1* and depart as flight *a2*. Similarly, aircraft *B* was supposed to arrive as flight *b1* and depart as flight *b2*, and analogously, *C* as *c1* and *c2*. If we rotate aircraft *A* with *B*, *B* with *C* and *C* with *A* circularly, *A* will arrive as *a1* and depart as *c2*, analogously, *B* as *b1* and as *a2*, and *C* as *c1* and as *b2*. As we have shown in figure 5, each flight has a time slot for an arrival and a departure. If we change the schedule of flights for some aircraft, the length of the original solution bar may be changed. This means that we need to adjust the solution of the initial schedule. The aircraft rotation can occur not only between two aircraft but also among three, four, or more aircraft. They also occur in a cycle as we can see in Figure 5. Since the fully automatic adjustments are not consistent from time to time and they are quite complex, these are too difficult to automate. The intelligent interactive GUI is utilized for manual adjustments.

THE RELATIONSHIP BETWEEN RACES AND OTHER SYSTEMS

Figure 6 illustrates the relationship between RACES and other systems. RACES receives flight schedule data from the Computer Center in order to make an aircraft parking schedule. After RACES performs the scheduling, it transfers scheduling result into KASTCO(Korea Air Terminal Service Company) and SELMC(SEoul Maintenance Control department). KASTCO is responsible for performing the ground tasks of an aircraft, for example: hydrating, cleaning, and towing. With the scheduling result obtained from RACES, KASTCO can prepare their tasks efficiently. Also, SELMC is responsible for maintenance of an aircraft. They check device problems on an aircraft. They need the data from RACES to make their maintenance schedule. Actual towing of an aircraft needs cooperation between KASTCO and SELMC. The aircraft parking schedules from RACES allow them to prepare to tow an aircraft before they perform actual towing.

DEVELOPMENT HISTORY

The ramp activity coordination system for KAL(Korean Air Lines) at Kimpo Airport in Korea was managed by human experts using manual scheduling until 1995. In order to maximize the utilization of the resources of the airport, KAL and Expert Systems Lab at Inha University decided to develop the expert system with experiences and heuristics of domain experts at KAL. Three domain experts and a system engineer from KAL and five graduate

students at Inha University participated in this project between 1995 and 1997 under Prof. Geun-Sik Jo's supervision. It was approximately a \$ 125,000 project for Inha University, which excluded wages for three domain experts and a system engineer, and expenses for an office and other utilities, which were paid by KAL. It took 3 months to make a prototype of RACES in order to convince KAL to pilot the project. Then after about 6 months, we were able to complete the initial scheduling part of RACES. Most of the developing time was spent in reactive scheduling which took about 9 months. For another 4 months, networking routines needed for the mainframe to integrate with RACES, which in turn interacts with other computers, were added to create a real environment. To evaluate the system, two end users tested the results against real flight schedule data of the previous 120 days' data. Their test results were accepted by domain experts so that RACES could be used for real environments. For the maintenance of RACES, we explained source codes and trained a system engineer to do the maintenance job himself. At the present time, a system engineer from KAL is responsible for maintaining RACES and he is doing well. We think that the declarative features of Prolog have made it possible for only one person to do the maintenance job. In addition, at the developmental stage, we designed and implemented menus and submenus to prepare for the future extension and update of the knowledge base.

RACES was written in CHIP(Constraint Handling In Prolog) Ver 5.0 under a Unix operation system on an HP/712 machine. To retrieve and store the information on flights, bridges and remote spots, Oracle DBMS was used, where an SQL interface from the Prolog code was provided in the CHIP system.

DEPLOYMENT PROCESS

We developed RACES(Ramp Activity Coordination Expert System) in CHIP, which consists of about 50,000 lines of Prolog code with about 70 GUI menus. RACES solves the problem using methods similar to a human expert problem solving procedures. We represented and processed the domain specific knowledge and experiences. When the system processes scheduling, it can prune the search space using a domain filtering technique. RACES produces a user-driven optimal schedule using trade-off scheduling heuristics. To test accuracy of the system, we implemented RACES with the daily operational data of an actual airline company for about 120 days and the results were analyzed by domain experts. RACES has been approved by and continues to receive the approval of domain experts.

The system described in this paper was successfully deployed at Kimpo Airport in Korea. RACES has been

used at Kimpo airport by KAL since 1997. The controllers at the operational control center at KAL are now using this system for monitoring and controlling the assignment of remote spots and bridges. To date, the ground controllers using this system to actually tow aircraft, assign buses and do other ground work are able to interact well with persons at the operations control center. As long as the airport is in operation, this system should run almost 24 hours a day. When the FIDS(Flight Information Display System) was connected to RACES, the time that the aircraft spent waiting to park after landing was greatly reduced.

RACES currently has the ability to reschedule in approximately 70% of the cases in a real environment situation. Reactive scheduling is one of the most important topics for researchers to use the scheduling systems in practice. However, when the system processes the rescheduling by itself, we can often find that some adjustments are not adequate. Therefore, some of the adjustments should be done in co-operation with the users by adjusting the schedule manually through GUI. Interactive GUI in RACES plays a supporting role by helping users to make the right decision.

BENEFITS OF RACES

RACES made the work paradigm-shift from manual into automatic scheduling in the ramp activity management that required domain specific human knowledge and heuristics . It is clear that the initial investment is returned within in a year after deployment. There are, however, the following benefits which are not currently measurable in terms of money:

- | Time and cost involved in scheduling are drastically reduced.
 - | Real-time adjustment for unexpected events and weather conditions is provided for.
 - | Interactive GUI in RACES plays a supporting role by helping users make the right decision at the right time.
 - | Aircraft waiting time for parking after landing is reduced.
 - | High-quality passenger service is provided because RACES gives prior information about aircraft parking status, thereby, insuring the quick movement of an aircraft after landing.
 - | Objective verification of ramp activity management is possible.
- Finally, RACES can perform the operations necessary to maximize the utilization of ramp activity.

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