Modeling an Integrated Public Transportation System -
a case study in Dublin, Ireland

Prabhat Shrivastava 1*, Margaret O'Mahony 2

1 Professor of Transportation System Engineering
Department of Civil Engineering - Sardar Patel College of Engineering
Bhavan's Campus, Andheri (W), Mumbai – 400058, India
2 Professor of Civil Engineering, Bursar,
Head of Department of Civil, Structural & Environmental Engineering and
Director of the Centre for Transport Research, Trinity College Dublin,
Dublin 2, Ireland

Abstract

The efficiency of the public transport system in any city depends on integration of its major public transport modes. Suburban railway and public buses are the modes normally used by the majority of commuters in metropolitan cities of developed and developing countries. Integration of these two services reduces overall journey time of an individual. In this research, a model is developed for operational integration of suburban trains and public buses. The model has two sub models: a Routing Sub Model and a Scheduling Sub Model. In the Routing Sub Model, feeder routes are generated for public buses which originate from a railway station. A Heuristic Feeder Route Generation Algorithm is developed for generation of feeder routes. In the Scheduling Sub Model, optimal coordinated schedules for feeder buses are developed for the given schedules of suburban trains. As a case study the Dun Laoghaire DART (Dublin Area Rapid Transit) (heavy rail suburban service) station of Dublin in Ireland is selected. Feeder bus services are coordinated with existing schedules of the DART on the developed feeder route network. Genetic Algorithms, which are known to be a robust optimization technique for this type of problem, are used in the Scheduling Sub Model. Finally the outcome of the research is a generated feeder route network and coordinated services of feeder buses on it for the DART station.

Keywords: Coordinated schedules, Genetic algorithms, Modal integration, Optimization, Public transportation, Routing and scheduling.

1. Introduction

It has been observed that most of the metropolitan cities of developed and developing countries are facing problems due to lack of coordination among public transport facilities. Each public transport facility is planned and designed without considering its...
impact on other public transport services. In fact in most of the cases these facilities compete each other instead of complementing. This unhealthy competition leads to duplication of services to many areas and hence proves to be uneconomical. Commuters have to spend more time on journeys because of higher transfer time due to lack of integration among public transport modes. The efficiency of an entire public transport system can be enhanced by overall coordination among its modes. Coordination among different modes can be achieved by system integration, which occurs at three levels: institutional, operational and physical. The literature review has revealed that many studies are carried out for optimization of services of a single mode specially bus or train but the effort is meager as far as coordination of two modes are concerned. However, routing and scheduling problems for coordinated operations were attempted by Wirasinghe (1980), Geok and Perl (1988) using analytical models. They had considered highway grid which is assumed to be rectangular and parallel to a single railway line which may not always be true in practice. They had made an attempt to describe complex transit system by approximate analytical models. Thus most of the studies on coordination of modes are limited to analytical modeling without considering a real life network (Shrivastava and Dhingra, 2000). In this research, a model is developed for operational integration of public transport modes. Development of feeder routes and schedule coordination, the two important aspects of operational integration, are attempted in this research. As a case study, Dun Laoghaire DART station is selected. Dun Laoghaire is a rapidly growing suburb of Dublin city in Ireland. The coordination between DART services and Dublin buses (public buses) at this DART station is attempted.

2. Data collection

The Dublin Area Rapid Transit (DART) is a suburban railway system in Dublin, running basically along the coastline of Dublin Bay from Greystones to Howth and Malahide. There are 32 stations on the existing DART line. Lack of coordination between public buses and DART services has been observed even during peak hours at many stations. Dun Laoghaire is one of the prominent DART stations from where large number of trips originate. It was decided to select Dun Laoghaire as the study area due to its land use pattern which allows greater scope of feeder bus services from the station. Considerable movement of commuters takes place towards many areas from the DART station.

Typical traffic surveys were conducted during the morning peak period i.e. 7 to 9 a.m. on April 28, 2004. It was observed that the maximum number of commuters travel during 8 to 9 a.m. Therefore this time period is identified as peak hour. It has been confirmed during traffic surveys that after 9 a.m. commuter traffic starts decreasing and becomes very less after 9.30 a.m. onwards. During the traffic surveys, commuters exiting the DART station were counted manually. Typical commuter counts revealed that between 8 and 9 a.m. 1293 commuters exit from the DART station. Traffic surveyors conducted sample interviews of commuters leaving the DART station. Between 8 and 9 a.m. 300 commuters were interviewed thus making a sample size above 20%. Enquiries were made regarding their destinations, mode of transport and travel time to their destinations from DART station. Commuters who did not opt public
buses for their further journeys were also asked about their willingness to shift to public buses if buses are coordinated with DART services in future. It was found that 40% of commuters have their working places very near to DART station and they have to walk even less than 5 minutes. These commuters were not interested in shifting to public buses even if they are well coordinated with DART services. The percentage of commuters willing to shift to public buses were added to those who use public buses and a potential demand matrix for public buses was developed. It was found that there are 16 destinations (nodes) for which demand exist from DART station. Table 1 indicates potential demands to various destinations. The demand for Dun Laoghaire College, Sallynoggin, Monkstown, Deans Grange, Stillorgan and Loughlinstown was found to be more than average. Thus these nodes were identified as major destinations and priority is given to these destinations for development of feeder routes. Connectivity and distances to all destinations were obtained from Dublin Street map (Dublin street map, 2000). An average speed of 15 km per hour was adopted to address the existing congestion level and road geometrics of the influence area (Scott Wilson, 2000). Using this speed, a travel time matrix was developed. The size of matrix was 17 ×17 which includes DART station and other identified 16 destinations as indicated in Table 1. The potential demand matrix and travel time matrix were used for development of feeder route network. It was also observed during traffic surveys that in the morning peak period the trains towards city centre (north bound trains) contribute about 30% passengers; the remaining 70% were by trains from city centre (south bound trains). There were nine north bound and eight south bound trains during the peak hour of 8 to 9 a.m. The schedule coordination for feeder buses is attempted fortheses trains during the indicated peak hour.

Table 1: Potential Demand to Various Destinations

<table>
<thead>
<tr>
<th>Node No. (code)</th>
<th>Destinations</th>
<th>Potential demand to various destinations</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Dun Laoghaire DART Station</td>
<td>7 - 8 a.m. 8 - 9 a.m. 7 - 9 a.m.</td>
</tr>
<tr>
<td>2</td>
<td>Dun Laoghaire College</td>
<td>39  202  241</td>
</tr>
<tr>
<td>3</td>
<td>Sallynoggin</td>
<td>17  103  120</td>
</tr>
<tr>
<td>4</td>
<td>Monks town</td>
<td>10  63  73</td>
</tr>
<tr>
<td>5</td>
<td>Deans Grange</td>
<td>16  93  109</td>
</tr>
<tr>
<td>6</td>
<td>Temple Hill</td>
<td>02  06  08</td>
</tr>
<tr>
<td>7</td>
<td>Black Rock</td>
<td>08  46  54</td>
</tr>
<tr>
<td>8</td>
<td>Stillorgan</td>
<td>13  77  90</td>
</tr>
<tr>
<td>9</td>
<td>Leopards town</td>
<td>02  08  10</td>
</tr>
<tr>
<td>10</td>
<td>Foxrock</td>
<td>02  08  10</td>
</tr>
<tr>
<td>11</td>
<td>Maple Manor / Cabinteely</td>
<td>02  04  06</td>
</tr>
<tr>
<td>12</td>
<td>Lough Linstown</td>
<td>13  78  91</td>
</tr>
<tr>
<td>13</td>
<td>Mount Merrion</td>
<td>02  15  17</td>
</tr>
<tr>
<td>14</td>
<td>University College of Dublin</td>
<td>04  23  27</td>
</tr>
<tr>
<td>15</td>
<td>Dundrum</td>
<td>06  31  37</td>
</tr>
<tr>
<td>16</td>
<td>Sandyford</td>
<td>03  15  18</td>
</tr>
<tr>
<td>17</td>
<td>Rouches Town Avenue</td>
<td>02  04  06</td>
</tr>
</tbody>
</table>
3. Model for operational integration

The objective of the research is to develop a model for operational integration of DART services and public buses. The scope of work involves development of a feeder route network for a selected DART station, Dun Laoghaire. The feeder route network is developed in a Routing Sub Model. The next stage is schedule coordination of feeder buses for the existing schedules of DART on the developed feeder route network. The schedule coordination is attempted in the Scheduling Sub Model. Figure 1 indicates the overall methodology adopted for operational integration of two services.

Figure 1: Proposed methodologies for operational integration.
3.1 Routing Sub Model

From the literature review it is evident that heuristic approach has been very popular for development of route network. Lampkin and Saalmans (1967), Silman et al. (1974), Dubois et al. (1979), Hsu and Surti (1976), Dhingra (1980), Mandl (1980), Baaj and Mahamassani (1990 and 1995) developed bus routes using heuristic approach by insertion of nodes in base network. Heuristic approach may or may not provide optimal route structure but it is certainly able to provide good practically acceptable suboptimal solutions (Shrivastava and Dhingra, 2001). Location of various destinations (nodes), limited connectivity among some of nodes in the influence area of DART station and design of routes without further bus to bus transfer (passengers are already subjected to one transfer i.e. from DART to buses) also encouraged to use heuristic approach in this study. The heuristic algorithm described here is developed in ‘C’ language using different node selection and insertion strategies. Proposed heuristic algorithm is heavily guided by demand matrix because satisfaction of demand is one of the prime aspects for generation of routes (Baaj and Mahamassani, 1995). Thus the model for operational integration is decomposed in two sub models: one for routing and other for schedule coordination. In actual practice also user is more concerned about the waiting / transfer time rather than slightly higher journey time. It leads to higher level of discomfort and dissatisfaction if commuters have to wait longer for connecting buses to their destinations. Therefore it is decided to carry out rigorous optimization to minimize transfer time from DARTs to buses on heuristically developed feeder routes.

The proposed heuristic algorithm has two distinct parts
1. Development of shortest paths using Dijkstra’s algorithm from DART station to identified major destinations.
2. Deviation of shortest paths by inserting other identified nodes to develop feeder routes. The deviation of shortest paths has been done based on various ‘node selection and insertion strategies’.

There should be a judicious balance in satisfaction of demand due to insertion of nodes and increase in route length for development of routes (Baaj and Mahamassani, 1995). Thus the deviation of shortest paths for development of routes is governed by ‘maximum demand deviated shorter path time’ criterion. In this criterion, the deviation of shortest paths due to insertion of nodes between origin and destination is restricted to 1.5 times the travel time on the shortest paths. The nodes which are attached at the end of shortest paths are governed by ‘path extension time criterion’. The path extension time criterion fixes an upper limit on the length of routes. In the present case study this upper limit for the length of routes is kept as 15 Km (1 hr). This upper limit for the routes is decided based on the locations of various destinations identified in sample interviews which were part of traffic surveys. The upper limit on length of routes is imposed because if routes are very long then the purpose of feeder routes is lost and such routes pose difficulty in maintaining the schedules. Though the upper limit of the route length adopted in the case study is on higher side, this limit can be reduced if other DART stations are also coordinated. This is due to the fact that a particular node may have connectivity with more than one DART stations which may lead to shorter and better routes from one station as compared to other one. Traffic surveys also revealed that some of the nodes having higher demands are concentrated near the DART station and as a result many shortest paths would be developed from DART station to these major destinations. After inserting the nodes very short routes mushrooming near
railway station would develop. Such routes are not practically acceptable. A similar problem was felt in route generation algorithm of Baaj and Mahmassani (1995). Proposed heuristic algorithm avoids development of such routes by imposing a constraint on minimum length of shortest paths and deviation of shortest paths using node selection and insertion strategies. The various steps involved in the proposed algorithm are described as follows:

1. Prepare the demand matrix with code numbers of various nodes (destinations) to which potential demand is identified from the DART station (origin).
2. Identify the connectivity of the above nodes using the existing route map and develop travel distance matrix in kms. Non-connectivity of nodes is assigned a very high number in the matrix. Using the average speed of travel the matrix is converted into travel time matrix in terms of ‘minutes’.
3. Identify the nodes having more than average demand and select them as major destinations.
4. Develop the Shortest paths from the DART station to major destinations using Dijkstra’s algorithm.
5. Remove all the nodes from the node list which are present in any shortest path and arrange remaining nodes in the decreasing order of their demand i.e. node having highest demand is kept at top and one with least demand at the bottom. This is done so as to give priority to higher demand nodes during the insertion process. Nodes at the top are chosen first for insertion. The nodes are removed from node list because they are now the part of shortest path and hence will be the part of a route. The opportunity is given to other nodes for insertion in shortest paths / routes.
6. Identify the nodes/chain of nodes at the end of shortest paths / routes and insert them at the end of shortest path/routes using path extension time criteria. This automatically avoids delay to the higher demand nodes (major destinations).
7. The nodes, which are remaining and are already arranged as per demand, are then inserted as per node selection and insertion strategies. The lengths of routes are governed by the above mentioned time criteria which are applied depending on location of nodes and the way they are inserted in the shortest paths / routes.
8. Insertion of nodes continues one after another until all the nodes are exhausted.
9. After generating all of the routes, they are checked for backtracking. If backtracking is found and better alternatives are available they are considered and the route is suitably modified.

3.2 Node selection and insertion strategies

In development of feeder routes the nodes having higher demands should be given priority over nodes having lower demands (Shrivastava and Dhingra, 2001). Thus in node selection and insertion strategies the nodes having higher demands are given priority for insertion over lower demand nodes. The strategies adopted for insertion of any node in the shortest paths / routes are briefly mentioned below.

a) The best possible shortest path / route for any node to be inserted is first identified. The best possible shortest path / route for a particular node is decided on the basis of the ratio of saving in passengers walk time (SPWT) to increased bus passenger time (IBPT) due to insertion of the node. This ratio is calculated for all the shortest
paths / routes. The node is inserted to the shortest path / route which gives the highest value of this ratio.

Let, in fig. II
DART Station: ‘i’
Destinations having more than average demand: j1, j2 and j3
Shortest paths/Routes originating from DART station ‘i’: (i,j1), (i,j2) and (i,j3)
Node to be inserted: ‘k1’
Demand from railway station ‘i’ to ‘j’: D_{ij}
Demand from railway station ‘i’ to ‘k1’: D_{i,k1}
Nodes on shortest paths/routes (i,j1), (i,j2) and (i,j3) which are nearest to ‘k1’: j1~, j2~ and j3~
(Routes/shortest paths those have no connectivity with k1 are omitted)
Travel time on shortest path/route (i, j): t_m (i, j)
Say k1 is inserted in route (i,j1) the travel time will increase from ‘i’ to ‘j1’ due to deviation of this shortest path/route. Also D_{i,k1} passengers who had to walk for a distance of (j1~ - k1) to reach k1 will be benefited. Therefore
Travel time from ‘i’ to ‘j1’ via node k1 due to its insertion: t (i, j1)
Increase in travel time: \{ t (i, j1) - t_m (i, j1) \}
Delay in terms of passenger-min for bus passengers (IBPT): D_{ij1} \{ t (i, j1) - t_m (i, j1) \}
Walking time for passengers from j1~ to k1: t (j1~ - k1)
Saving in passengers-min due to walking (SWPT): D_{i,k1} t (j1~ - k1)
Calculate following (SPWT/ IBPT) ratios for all the routes as given below and consider the route for which this ratio is maximum. In this case Maximum demand deviated shorter time path criterion is adopted.

a. \frac{D_{i,k1} t (j1~ - k1)}{D_{ij1} \{ t (i, j1) - t_m (i, j1) \}}
b. \frac{D_{i,k1} t (j2~ - k1)}{D_{ij2} \{ t (i, j2) - t_m (i, j2) \}}
c. \frac{D_{i,k1} t (j3~ - k1)}{D_{ij3} \{ t (i, j3) - t_m (i, j3) \}}
b) After selecting the route / shortest path for insertion of any node the best possible way in which the node could be inserted in the selected shortest path / route is determined. The best possible way is determined on the basis of minimum additional passengers delay to successor nodes. Sometimes backtracking becomes essential at any node due to its location and connectivity with other nodes. In such cases also the above criterion of minimum additional passenger delay is used.

c) The presence of a node or series of nodes at the end of shortest path / route makes it essential to extend the route. In case of the presence of one node, both the options of inserting the node at the end of shortest path / route and between last and last but one node are analyzed. The option which gives the minimum additional passenger delay is selected. The series of nodes are attached at the end of the concerned shortest path / route if they are present at the end to avoid additional delay to higher demand nodes.

d) Sometimes due to the presence of a series of nodes near to the shortest path / route, backtracking on some nodes becomes essential. This backtracking may also increase the length of the route beyond the specified value. In such cases, to avoid backtracking and delays to higher demand nodes part of the length of shortest path is merged with the series of nodes and thus new routes are developed (Shrivastava and Dhingra, 2001).

e) Finally, all the routes are checked for undesirable backtracking. To check undesirable backtracking and to explore better options the travel time on the return journey of backtracked section is assigned a very high value. Other options, if any, are analyzed and compared with the backtracked option and the better one in terms of minimum passenger delay is selected.

3.3 Scheduling Sub Model

Attempts have been made to obtain optimal schedule on transit networks only with transfer time consideration using computer simulation (Rapp and Gehner, 1976) and combination of optimization model and simulation procedure (Bookbinder and Diesilets, 1992). However development of optimal schedules is an extremely difficult task especially for schedule coordination problem even for a small transit network. The schedule coordination problem consists of transfers between at least two modes along with other objective like vehicle operation cost or fleet size. There are constraints like keeping load factors and transfer times on various routes acceptable to both users and operators. Thus the objective function and constraints make such problems multi-objective, non linear and non convex (Shrivastava et al, 2002). The difficulty due to large number of variables and constraints, the discrete nature of variables and non-linearity involved in the objective function and the constraints makes such problems difficult to be solved by traditional optimization techniques (Chakroborthy et al., 1995). In view of this, techniques like fuzzy logic have been tried for such problems (Kikuchi and Parmeswaran, 1993). Chakroborthy et al. (1995) highlighted the enormity of a similar type of problem. Even after linearizing the problem, the complexity remains very large. The benefit obtained through linearization is offset by the increase in the number of variables and constraints. In general, the number of variables and constraints required are of the order of $O(r^n)$, where ‘r’ is the number of routes through a transfer station and ‘n’ is the number of buses/trains on any of the routes. Chakroborthy et al. (1995) attempted to solve the linearized formulation of a similar problem, but the
algorithm failed to converge to any solution. Therefore Genetic Algorithms (GAs), which is a robust optimization technique and well suited for such problems, is applied for this phase of the research (Goldberg, 1989). The basic differences of GAs with most of the traditional methods are that GAs use coding of the variables instead of variables directly, a population of points instead of a single point, and a stochastic operators instead of deterministic operators. All these features make GAs search robust, allowing them to be applied to a wide variety of schedule coordination problems (Shrivastava and Dhingra, 2002). The following steps are involved in determination of coordinated schedules.

1. Assignment of traffic on developed feeder routes.
2. Development of objective function and constraints.
3. Calculation of penalized objective function
4. Application of Genetic Algorithm to determine optimal frequencies on different routes for minimum penalized objective function.

### 3.3.1 Assignment of traffic on developed routes

Potential demand to various destinations from the DART station is identified through traffic surveys. Since all the feeder routes to different destinations originate from the DART station the link connecting the station and the first node on the route is critical link. This link carries the maximum load on the route. Scheduling of buses is done on the basis of this maximum load. The assignment of traffic on feeder routes is based on the proportionate frequency criterion which is based the fact that a higher bus frequency attracts larger traffic.

### 3.3.2 Development of objective function and constraints

The scheduling of any public transport facility must satisfy both users and operators. The users are concerned with availability of services without waiting longer time and acceptable crowding levels. Operators are concerned with saving in operational cost of facility or minimizing the fleet size and higher crowding levels to earn profit or at least to get break even. Thus in the objective function for schedule coordination the user cost is associated with the transfer time between buses and DART services. The operator cost is taken as the vehicle operating cost which is incurred due to total distance travelled by buses (Shrivastava and Dhingra, 2002). The constraints are related to minimum and maximum load factor, minimum and maximum transfer time and unsatisfied demand. Mathematically the objective function and constraints can be presented as follows:

**Objective Function:**

\[
C_1\left\{\sum_j \sum_u \sum_f pass_j^u \left(\text{bus}_j - \text{dart}^u\right) \delta_j^u + \sum_j \sum_u \sum_f pass_j^u \left(\text{bus}_j - \text{dart}^u\right) \delta_j^u \right\} + C_2\left\{\sum_f T_{L_j}\right\}
\]
Constraints

1. \((\text{bus}_j - \text{dart}_u) \leq T_{\text{max}}\) and \((\text{bus}_j - \text{dart}_v) \leq T_{\text{max}}\)  
   Maximum transfer time constraint

2. \((\text{bus}_j - \text{dart}_u) \geq T_{\text{min}}\) and \((\text{bus}_j - \text{dart}_v) \geq T_{\text{min}}\)  
   Minimum transfer time constraint

3. \(\frac{Q_{\text{max}}}{N_j \times \text{CAP}} \leq L_{\text{max}}\)  
   Maximum load factor constraint

4. \(\frac{Q_{\text{max}}}{N_j \times \text{CAP}} \geq L_{\text{min}}\)  
   Minimum load factor constraint

5. \(\sum_j d_{\text{unsat}} = 0\)  
   Unsatisfied demand constraint

Where,

- \(j = \) Number of routes available at each stations
- \(l = \) Number of buses available for \(u^{th}\) north bound DART and \(v^{th}\) south bound DART
- \(VOC = \) Vehicle operating cost for Dublin buses
- \(C_1 = \) Cost of transfer time in Euro per minute, adopted as 11.32 cents/minute for the case study, (Steer Davies, 1994).
- \(C_2 = \) Cost of operation of Dublin bus per Km., adopted as € 3.66 for Dublin buses for the case study, (Scott Wilson, 2000).
- \(\text{pass}_u = \) Passengers transferring from \(u^{th}\) north bound DART to \(j^{th}\) route.
- \(\text{pass}_v = \) Passengers transferring from \(v^{th}\) south bound DART to \(j^{th}\) route.
- \(\text{bus}_j = \) Departure of \(l^{th}\) bus on \(j^{th}\) route
- \(\text{dart}_u = \) Arrival of \(u^{th}\) north bound DART
- \(\text{dart}_v = \) Arrival of \(v^{th}\) south bound DART
- \(\delta_{u.l} = \) is a term which shows whether transfer of passengers is possible or not. It attains a value one if transfer from \(u^{th}\) north bound DART to \(l^{th}\) bus on \(j^{th}\) route at DART station is feasible otherwise it attains a value zero.
- \(\delta_{v.l} = \) is also a term which shows whether transfer of passengers is possible or not. It attains a value one if transfer from \(v^{th}\) south bound DART to \(l^{th}\) bus on \(j^{th}\) route at DART station is feasible otherwise it attains a value zero.
- \(f_j = \) Frequency of buses on \(j^{th}\) route in terms of number of bus trips per hour
- \(l_j = \) length of \(j^{th}\) route in kilometers
- \(TP = \) Time period, hours
- \(T_{\text{max}} = \) Maximum allowable transfer time between arrival of DART and departure of connecting bus. For the case study this value is assumed as 10 minutes (Based on commuters’ opinion survey in study area).
- \(T_{\text{min}} = \) Minimum allowable transfer time between arrival of DART and departure of connecting bus. For the case study this value is assumed at 5 minutes (Based on observations & opinion survey in study area).
- \(Q_{j,\text{max}} = \) Number of passengers on first link connecting DART station on \(j^{th}\) route for given time period.
- \(N_j = \) Number of bus trips during entire time period under consideration (\(f_j \times TP\))
- \(\text{CAP} = \) Seating capacity of bus, for Dublin buses it is taken as 74 (Scott Wilson, 2000)
- \(L_{\text{max}} = \) Maximum load factor, it is adopted as 1.2 for the case study (Scott Wilson, 2000)
- \(L_{\text{min}} = \) Minimum load factor, adopted as 1 for the case study
- \(d_{\text{unsat}} = \) Unsatisfied demand
The first term of the objective function involves transfer time between DART services (both nth and sth bound) and coordinating buses. The second term gives the vehicle operating cost, which is proportional to the distance traveled by buses. Constants C1 and C2 are used to convert the objective function in monetary unit of Euro (€). The first two constraints are related to transfer time (Chakroborthy et al., 1995). The first constraint ensures that transfer time between arrival of a DART and departure of connecting buses should be less than a maximum value. The second constraint ensures that there should be minimum time available for transfer. This constraint is obvious because it takes a minimum time for passengers to board coordinating buses after arriving from DART. Through the traffic surveys this minimum transfer time has been established as 5 minutes. The third and fourth constraints ensure that the load factor lies within a maximum and a minimum value so that better level of service and availability of a certain minimum number of passengers can be ensured for economical operations. The maximum load factor is the ratio of crush capacity and normal capacity of Dublin buses. The crush capacity is taken as 88 and normal capacity is 74 thus the maximum load factor is taken as 1.2 (Scott Wilson, 2000). The last constraint ensures that maximum demand is satisfied and maximum number of commuters get coordinating buses during the period of analysis (Shrivastava et al, 2002). None of the above constraints are rigid. These constraints are obeyed and violated as per their relative importance and magnitude is directly proportional to potential demand associated with a particular constraint. Penalties are decided as per the extent of violation of constraints i.e. higher penalties are imposed for greater violation of these constraints.

3.3.3 Calculation of penalized objective function

The objective function and constraints as mentioned above pose a constrained optimization problem. Transformation methods are the simplest and most popular optimization methods of handling constraints. The constrained problem is transformed into a sequence of unconstrained problems by adding penalty terms for each constraint violation. If a constraint is violated at any point, the objective function is penalized by an amount depending on the extent of constraint violation (Deb, 1995). Three sets of penalties are decided which are added to objective function and penalized objective function is calculated. The following penalties are used in analysis:

1. Transfer time penalty
2. Load factor penalty
3. Penalty for unsatisfied demand

These penalties are function of objective function, penalty coefficient, number of affected commuters and adopted bus capacity.

3.3.3.1 Transfer time (tt) penalties

As stated above, it is observed during the surveys that it takes about 5 minutes on average to reach a bus stop after arriving from the DART. Thus the minimum transfer time from DART to bus is adopted as 5 minutes. Therefore, any bus which starts after 5 minutes of the scheduled arrival of DART is considered as a connecting bus to that particular DART service. A transfer time between 5 to 10 minutes is regarded as acceptable. In fact considering 5 minutes as the minimum time required for transfer, effective waiting time lies between zero to five minutes which is considered as acceptable. Any transfer after 10 minutes i.e. effective waiting time more than 5 minutes...
is penalized. Higher values of penalty coefficients are adopted for higher transfer time because higher transfer time causes more discomfort to passengers.

3.3.3.2 Penalty due to unsatisfied demand

If some passengers are not able to get any bus in the specified duration of analysis then it is taken as unsatisfied demand and the penalty is imposed on objective function.

3.3.3.3 Load factor (LF) penalties

The minimum value of load factor is adopted as ‘1’. The value of maximum load factor is adopted as 1.2 so as to maintain a better level of service. Level of service becomes poor due to a rise in load factor above maximum adopted value. If the load factor becomes less than ‘1’ it leads to uneconomical operation which may not be acceptable to operators. Therefore higher values of penalty coefficients are adopted as load factor increases above the maximum specified value similarly higher values are adopted as load factor decreases below minimum value.

3.3.4 Application of Genetic Algorithms

In the real world, the process of natural selection controls evolution. Organisms most suited for their environment tend to live long enough to reproduce, whereas less suited organisms often die before producing young or produce fewer and/or weaker young. In the applications of Genetic Algorithms process of evolution is studied by creating an artificial world, populating it with pseudo organisms and giving those organisms a goal to achieve (Goldberg, 1989). Genetic Algorithms store the characteristics of artificial organisms in a Genotype, which mimics the DNA of natural life. The genotype is nothing more than a long string of bits. A bit is the smallest piece of data a computer can process. It can be only one of two values: ‘0’ or ‘1’. A bit in the genotype string can be ‘on’ which has the value ‘1’, or can be ‘off’ which has the value ‘0’. The existence of a certain characteristic can be indicated by whether a particular bit is set to ‘on’ or ‘off’.

The operation of GAs begins with population of random strings representing design of decision variables. Thereafter, each string is evaluated to find the fitness value. The population is then operated by three main operators – reproduction, crossover and mutation to create a new population of points. The new population is further evaluated and tested for termination. If the termination criterion is not met, the population is iteratively operated by the above three operators and evaluated. This procedure is continued until the termination criterion is met. One cycle of these operations and subsequent evaluation procedure is known as a ‘generation’. The GAs use search strategies by using probability in all their operators. Since an initial random population is used, to start with, the search can proceed in any direction and no major decisions are made in the beginning. Later on, when the population begins to converge in some bit positions, the search direction narrows and optimal or near optimal solution is achieved. Thus nature of narrowing the search space as the search progresses is adaptive and is unique characteristic of Genetic Algorithms (Deb, 1995). Therefore Genetic Algorithms always guarantee the optimum / near to global optimum solution for ill behaved functions. Solutions even near to global optimum obtained by GAs are acceptable for practical problems, like the one which is being attempted in this research.

‘Reproduction operator’ is usually the first operator applied on a population. Reproduction selects a good string in a population and forms a mating pool. In the
‘Crossover operation’, information among strings of the mating pool is exchanged and new strings are created. ‘Mutation’ adds new information in a random way to the genetic search process, and ultimately helps to avoid GAs from getting stuck at local optimums. In the present analysis ‘uniform random’ and ‘roulette’ selection operators are compared. Similarly ‘simple’ and ‘uniform’ cross over are tested. Best among ‘simple invert’, ‘simple random’ and ‘swap’ mutation is used (Lance Chambers, 1995).

3.4 Use of Genetic Algorithms for objective function and constraints

The above objective function is used with LibGA software (Lance Chambers, 1995) of Genetic Algorithms in Linux environment to determine optimal frequencies on developed feeder route network. Genetic Algorithms parameters are tuned for the objective function and thus type of process and best values of operators are decided. The following are the outcomes of several runs for tuning Genetic Algorithms parameters.

− Roulette and uniform random selections are compared and it is found that Roulette selection converges faster for our objective function.  
− Simple random and Swap mutation give better results as compared to Simple invert. In the analysis the Simple random mutation is adopted.  
− Uniform crossover converges earlier to Simple crossover. Thus uniform crossover is adopted for the analysis.  
− Among seed values 1 to 10 seed value ‘1’ gave best results and hence is adopted for analysis.  
− The value of penalized objective function for pool size 30 is found to be same as obtained for pool size 70 and above. Therefore pool size 30 is adopted which has the advantage of lesser computational time also.  
− It is found that combination of crossover probability of 0.85 and mutation probability of 0.005 gave the lowest value of penalized objective function. Thus these values are used for the analysis.

Using the above Genetic Algorithm parameters, a set of penalty coefficients for transfer time, load factor and unsatisfied demand are decided. The coefficients are decided so as to keep the load factor in the range between 1 (minimum load factor) and 1.2 (maximum load factor), the percentage unsatisfied demand as low as possible and the effective waiting time for larger percentage demand between ‘zero’ and ‘five’ minutes. The demand satisfaction and load factors on various routes are two dominating factors for both users and operators. It has been found during the interviews of commuters that they prefer to have connecting buses with in five minutes of waiting after arriving at bus stops but most of them even accept ten minutes of waiting as a reasonable time. Thus the variation of penalty coefficients for minimum load factor is studied on percentage satisfaction of demand with in ten minutes of waiting. The coefficient for minimum load factor is selected because it is observed that the load factor frequently goes below 0.4 (minimum value) due to low demand which is not compatible to adopted existing bus capacity. Table 2 indicates typical variation of overall load factor (average load factor of all the routes), percentage demand satisfied with in ten minutes of waiting and values of penalized objective function. This typical variation is observed when penalty coefficient corresponding to minimum load factor (less than 0.4) is varied keeping other coefficients same. The typical variation in the table shows that Genetic Algorithms are very sensitive to penalties. A weighted factor is
calculated by awarding equal weights to the overall load factor and percentage demand satisfaction within ten minutes of waiting. Penalty coefficient corresponding to higher weighted factor is selected for further analysis.

Table 2: Typical variation of overall load factor, satisfied demand within ‘10’ minutes of waiting and penalized objective function with respect to coefficient of minimum load factor penalty

<table>
<thead>
<tr>
<th>Value of Coefficient for minimum load factor</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
<th>10</th>
</tr>
</thead>
<tbody>
<tr>
<td>Over all load factor (average for all the routes)</td>
<td>0.3074</td>
<td>0.3074</td>
<td>0.3656</td>
<td>0.3656</td>
<td>0.4210</td>
<td>0.5055</td>
<td>0.5055</td>
<td>0.5141</td>
<td>0.5310</td>
<td>0.5423</td>
</tr>
<tr>
<td>% demand satisfied with in 10 minutes of weighting</td>
<td>95.24</td>
<td>95.24</td>
<td>90.04</td>
<td>86.77</td>
<td>85.11</td>
<td>77.11</td>
<td>76.19</td>
<td>75.19</td>
<td>73.34</td>
<td>71.74</td>
</tr>
<tr>
<td>Typical values of penalized objective function</td>
<td>61804</td>
<td>81708</td>
<td>92732</td>
<td>105126</td>
<td>125158</td>
<td>132600</td>
<td>140042</td>
<td>149857</td>
<td>155529</td>
<td>173348</td>
</tr>
</tbody>
</table>

The penalties discussed above are calculated using the selected set of penalty coefficients and the penalized objective function is determined by adding penalties to the objective function. A set of frequencies on various routes corresponding to the minimum value of the penalized objective function is used for determination of coordinated schedules on various routes.

4. Results and discussion

It was found that there are 6 destinations having demand greater than average. These destinations are Dun Laoghaire College, Sallynoggin, Monkstown, Deans Grange, Stillorgan, and Loughlinstown. Using Dijkstra’s algorithm, four shortest paths were developed. These shortest paths were modified by node selection and insertion strategies and four feeder routes were obtained. The developed feeder route network is shown in Figure 3 with the code numbers of nodes as given in Table 1. The lengths of feeder routes 1, 2, 3 and 4 are 5.54, 9.10, 14.6 and 5.8 km respectively. If similar exercise is carried out by identifying influence area of all stations shorter feeder routes will be developed. This is due to the fact that one node may be connected to more than one DART station and its connectivity will certainly be better with shorter connecting length from one particular station. This will lead to smaller feeder routes which will ultimately help in maintaining schedules of feeder buses (Shrivastava and Dhingra, 2001). It can be seen in Figure 3 that destinations like Stillorgan (8), Mount Merrion (13), University College Dublin (14) and Dundrum (15) are closer to Blackrock DART station as compared to Dun Laoghaire. Thus feeder routes for these destinations from Blackrock will be shorter. In the existing route structure of Dun Laoghaire bus routes numbers 46A, 75, 111, 59, 46X originate from station where as route numbers 7, 7A and 45A pass through the station with origins elsewhere. Some of the existing routes that originate at the station pass through some of the locations for which demand does not originate from the station as indicated in our typical traffic survey. The route
Number 75 passes through Stillorgan, Leopardstown, Sandyford, Ballinteer, Oldbawn and Tallaght.

Our typical survey shows the demand for last three destinations is nil from the station. Moreover the length of this route is very long having existing trip time more than one hour. Such longer routes pose problems in maintaining schedules. Similarly route number 46A goes to the city centre thus duplicating the services of the DART towards city centre. Route number 45A goes to Bray which is parallel to the DART line. Route number 59 passes through Dalkey and Killiney. Route number 111 also passes through Dalkey and goes to Loughlinstown. The routes 59 and 111 could have been clubbed together and a single feeder route could have served the purpose. Thus it can be concluded that the existing routes do not serve the purpose of feeder routes and lead to duplication of services. Table 3 gives typical details of bus schedules with load factors on different routes during morning peak hour. The average load factors on 2\textsuperscript{nd}, 3\textsuperscript{rd} and 4\textsuperscript{th} routes are more than 0.4 and the overall load factor for all the routes is 0.3650. Average load factor on route ‘1’ is very low and this is due to the fact that during the hour of analysis the potential commuters for destinations lying on this route are less as compared to adopted existing capacity of buses in the analysis. It will be appropriate to use buses with smaller capacity on such routes. Moreover the local demand which is not considered at various nodes will further improve the load factors. The local demand is not considered because the routes are designed for feeder buses from DART station. Hence satisfaction of demands which generate from DART station is of prime concern. The load factors can be further improved if DART schedules are optimized beforehand.
(Shrivastava and Reddy, 2002). However, if DART schedules are modified then coordination for other direction travel i.e., from buses to DART should also be studied. In the existing scenario due to frequent availability of DART services there will always be coordination from buses to DARTs irrespective of arrival time of buses.

Table 3: Details of Bus Schedules with Load Factors

<table>
<thead>
<tr>
<th>S. No.</th>
<th>Train Timings</th>
<th>Bus Timings</th>
<th>Load Factors</th>
<th>Over all load factor</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>North Bound</td>
<td>South Bound</td>
<td>Route 1</td>
<td>Route 2</td>
</tr>
<tr>
<td></td>
<td>DARTS</td>
<td>DARTS</td>
<td>Route 1</td>
<td>Route 2</td>
</tr>
<tr>
<td>1</td>
<td>08.08</td>
<td>08.02</td>
<td>8.07</td>
<td>8.07</td>
</tr>
<tr>
<td>2</td>
<td>08.15</td>
<td>08.09</td>
<td>8.22</td>
<td>8.13</td>
</tr>
<tr>
<td>3</td>
<td>08.23</td>
<td>08.20</td>
<td>8.37</td>
<td>8.19</td>
</tr>
<tr>
<td>4</td>
<td>08.29</td>
<td>08.25</td>
<td>8.52</td>
<td>8.25</td>
</tr>
<tr>
<td>5</td>
<td>08.33</td>
<td>08.31</td>
<td>8.31</td>
<td>8.55</td>
</tr>
<tr>
<td>6</td>
<td>08.38</td>
<td>08.36</td>
<td>8.37</td>
<td></td>
</tr>
<tr>
<td>7</td>
<td>08.43</td>
<td>08.45</td>
<td>8.43</td>
<td></td>
</tr>
<tr>
<td>8</td>
<td>08.49</td>
<td>08.53</td>
<td>8.49</td>
<td></td>
</tr>
<tr>
<td>9</td>
<td>08.58</td>
<td>nil</td>
<td>8.55</td>
<td></td>
</tr>
<tr>
<td>10</td>
<td>Trains after 9 a.m.</td>
<td>Buses to be scheduled after 9 a.m.</td>
<td>9.00</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Average Load factors on Individual Routes</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table 4 gives waiting time details corresponding to developed coordinated schedules. It can be seen from the table that 65.07% demand is satisfied within ‘0’ to ‘5’ minutes of waiting and 24.97% of demand is satisfied between ‘6’ to ‘10’ minutes of waiting. Thus a total demand of 90.04% is satisfied within ‘10’ minutes of waiting. Entire demand is satisfied before ‘15’ minutes of waiting. In the present scenario since the existing routes do not serve the purpose of feeder routes average waiting time of commuters at Dun Laoghaire DART station is more than 15 minutes even during the morning peak hour with load factors in the range of 0.2 to 0.3.

Table 4: Waiting Time Details of Passengers

<table>
<thead>
<tr>
<th>Duration of Delay in Minutes</th>
<th>Percentage Demand Satisfied</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Route No : 1</td>
</tr>
<tr>
<td>0 - 5</td>
<td>42.8</td>
</tr>
<tr>
<td>6 - 10</td>
<td>32.2</td>
</tr>
<tr>
<td>11 - 15</td>
<td>25.0</td>
</tr>
<tr>
<td>More than 15</td>
<td>nil</td>
</tr>
</tbody>
</table>
5. Conclusions

Following conclusions can be drawn from this research.

− In this research, the model has been developed for operational integration of two services i.e. public buses and a rail service (DART) for one DART station only. The same modeling exercise can be repeated at various other DART stations after identifying influence area of each for different time periods of a day. Thus the model can develop an integrated public transport system in which suburban trains / DART services will work as main line haul service and buses can feed the local areas. This type of integrated system will allow both the modes to compliment each other instead of competing. The integrated system will also reduce wasteful duplication of services. If the modelling exercise is repeated to other DART stations route structures will be better in terms of lengths and satisfaction of demands.

− It is also confirmed that Genetic Algorithms are very efficient in solving multi objective, non linear schedule coordination problem. The time taken to obtain results is directly proportional to adopted population size. Near optimal results can be obtained with smaller population sizes, which take less computational time and are practically acceptable in real life situations. In the case study population size 30 is selected which takes less computation time and is able to provide equally good results as provided by higher population sizes.

− The variation of percentage demand satisfaction and over all load factor against variation of minimum load factor penalty shows that the Genetic Algorithms are very sensitive to penalties. Thus selection of appropriate penalties is very much required before the optimization process.

− The model developed in the research considers and develops real life network with real life objectives for both users and operators. The model takes into account real life constraints like level of service (maximum load factor), economical operation (minimum load factor), minimum and maximum transfer time and availability of public buses to maximum number of commuters (constraint for unsatisfied demand). The model maintains a judicious balance between load factor and satisfaction of demand within acceptable waiting time. Thus the model is able to provide satisfactory results (feeder routes and coordinated schedules) from users and operators point of view. Hence it can be claimed that proposed modeling exercise is a specific contribution towards realistic modeling on coordinated operations for passenger trips.

Acknowledgements

The research is funded under the Programme for Research in Third-Level Institutions (PRTLI), administered by the HEA. The authors would like to acknowledge the useful discussions with Dr Biswajit Basu during the research work.
References


