Modelling Single Line Train Operations

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Scheduling of trains on a single line involves using train priorities for the resolution of conflicts. The mathematical programming model described in the first part of this paper schedules trains over a single line track when the priority of each train in a conflict depends on an estimate of the remaining crossing and overtaking delay. This priority is used in a branch and bound procedure to allow the determination of optimal solutions quickly. This is demonstrated with the use of an example. Rail operations over a single line track require the existence of a set of sidings at which trains can cross and/ or overtake each other. Investment decisions on upgrading the number and location of these sidings can have a significant impact on both customer service and rail profitability. Sidings located at insufficient positions may lead to high operating costs and congestion. The second part of this paper puts forward a model to determine the optimal position of a set of sidings on a single track rail corridor. The sidings are positioned to minimise the total delay and train operating costs of a given cyclic train schedule. The key feature of the model is the allowance of non-constant train velocities and non-uniform departure times.

**Key words:** Railroad Scheduling, Siding Location, Optimisation Techniques, Railroad Operations
INTRODUCTION

This paper deals with two problems of single line train scheduling namely: the on-line scheduling of trains over a single line track with multiple sidings; and the optimum location of the sidings with respect to a given schedule. Part I deals with the optimum dispatching of trains on a single line of track. Trains can be dispatched from either end or from intermediate points on the track. When two trains approach each other on a single line, one of them must take the siding for the safe operation of the system. Determining which train takes the siding is done taking into account such factors as train priority, distance, lateness, and train operating costs. It is common practice for train operators to set a fixed timetable where conflicts are resolved. A train dispatcher in a control centre will act in the event of unforseen events. As these events can cause delays to trains, the dispatcher needs to continually alter the given timetable and resolve new conflicts. This is usually performed manually under strict time constraints so that the number of alternatives which can be assessed is very limited.

The operator’s experience and knowledge of local conditions, will continue to be used. Train dispatching decisions, which to a certain extent involve human as well as technical factors, will require human intervention to resolve problems. However, with such an optimisation model available, the operator is able to quickly update a schedule as unplanned events occur. The new optimal schedule offered by the model may not be fully implementable for practical reasons. However, the gap between the optimum and the practically feasible schedule, can be readily assessed. The penalty for not being able to implement the optimum schedule, in terms of operating cost and travel time reliability, can be evaluated against the practical factors which prevent implementation of the optimum schedule.
With the online train scheduling problem, the determination of the priority of a train at a particular point on the journey involves the consideration of the initial priority, current lateness of the train and a lower bound estimate of possible further conflict delay. Exploiting such a lower bound in a model will act as a look ahead function and will allow optimum schedules to be located quickly.

A second major use of the model relates to the planning of railroad operations. Such planning can be conveniently divided into two components, namely: short to medium term train planning; and railroad infrastructure planning associated with train operations. The model can be used to evaluate the implications of changes to a timetable in terms changed train departures, additional trains, and changes in train speeds. The optimum scheduling algorithm can be used as a simulator of proposed changes. Finally, the model can be used for long-range planning of railroad operations. In Australia, there are two main infrastructure planning issues which are currently under investigation, namely: the upgrading of main line track to allow higher speeds and heavier axle loads; and the need to extend sidings to allow for longer trains. The scheduling optimisation model can be used to evaluate both these investment strategies. The impact on the schedule of extending some sidings and not others can be assessed by using the model to simulate the effect of the proposed changes on future schedules. The removal of sidings has a cost in terms of flexibility and feasibility of schedules.

Part II deals with the development of a model to estimate the optimum position of sidings on a single line track. With high capital costs, a rail line must be designed as economically as possible, and at the same time have enough capacity to accommodate the forecast demand. Planning for a rail line involves determining the number of sidings required, the length of each siding, its position and the vertical and horizontal alignments for the line.
When determining the positions of sidings, several variables must be considered. The sidings must be placed in order to minimise train delays and total train operating costs. If too many sidings are planned for, the initial capital costs will outweigh the long term benefits and there will be wasted capacity.

**PART I: OPTIMUM TRAIN SCHEDULES**

**Past Research**

Research involving the scheduling of trains on a single line track is extensive and the following highlights the major developments.

Kraft (1) developed a dispatching rule giving the optimal time advantage for a particular train based on train priority, track running times and the delay penalties of each train. A similar method discussed by Sauder and Westerman (2) was implemented as a Decision Support System in a railway division of the United States. These models, which assume fixed train speeds, produce train plans which minimise the weighted total travel times.

Kraay et al (3) are the first to look at the idea of determining the cross-overtake plan and velocity profile to pace trains in order to conserve fuel, whilst keeping the lateness of the trains at a minimum. Similarily formulated constraints to that of Kraay et al (3) were used in a interactive Decision Support System (SCAN) by Jovanovic and Harker (4), to develop reliable train schedules using current schedules. Mills, Perkins and Pudney (5) formulated a discrete network type model by discretising the departure and arrival time variables of this formulation.
Model Formulation

Assumptions and Inputs

The following assumptions are made with regard to the model in this section:

- The track is divided into segments which are separated by sidings.
- Crossing and overtaking can occur at any siding or double line track segments.
- Trains can follow each other on a track segment with a minimum headway.
- For double track sections, it is assumed one lane will be allocated for inbound trains and one lane will be allocated for outbound trains. Usually, signal points will be set up this way.
- Scheduled stops are permitted at any intermediate siding for any train

The model will require various information to make use for the input to the model. The specific information is as follows:

- An unresolved train plan to make available the number of overtake and cross interferences for each train.
- The initial priorities of each train. These are determined by several factors such as the type of train, customer contract agreements and train load.
- The upper and lower velocity limits for each train (which are dependent on the physical characteristics of the track segment and the train).
- Segment lengths and the identification of single and double line track segments.
- The times of any scheduled train stops. These stops may include loading/unloading, refuelling and crew changes.
Definition of Variables

The set of trains is denoted by \( I = \{1, 2, \ldots, m, m+1, \ldots, n\} \) for which inbound trains are from 1 to \( m \) and outbound are from \( m+1 \) to \( n \). The variables used in the model are listed and described in this section.

Let: \( P = \{P_1, P_2\} \)

where:

- \( P_1 \) = set of single line tracks, \( P_2 \) = set of double line tracks

The integer decision variables for determining which train traverses a section first (also determines the position of conflict resolution) are given by:

\[
A_{ijp} = \begin{cases} 
1 & \text{if inbound train } i \leq m \text{ traverses track segment } p \in P_1 \text{ before inbound train } j \leq m \\
0 & \text{otherwise}
\end{cases}
\]

\[
B_{ijp} = \begin{cases} 
1 & \text{if inbound train } i \leq m \text{ traverses track segment } p \in P_1 \text{ before outbound train } j > m \\
0 & \text{otherwise}
\end{cases}
\]

\[
C_{ijp} = \begin{cases} 
1 & \text{if outbound train } i > m \text{ traverses track segment } p \in P_1 \text{ before outbound train } j > m \\
0 & \text{otherwise}
\end{cases}
\]

The arrival and departure time decision variables are as follows:

\[
X_{iq}^i = \text{arrival time of train } i \in I \text{ at station } q \in Q
\]

\[
X_{dq}^i = \text{departure time of train } i \in I \text{ from station } q \in Q
\]

\[
X_{Oi}^i = \text{departure time of train } i \in I \text{ from its origin station}
\]

\[
X_{Di}^i = \text{arrival time of train } i \in I \text{ at its destination station}
\]

The input parameters are defined as follows:

\[
h_p = \text{minimum headway between two trains on segment } p \in P_1
\]

\[
d_p = \text{length of segment } p \in P
\]

\[
Y_{Oi}^i = \text{planned departure time of train } i \in I \text{ from origin station}
\]

\[
Y_{Di}^i = \text{planned arrival time of train } i \in I \text{ at destination station}
\]

\[
\nu_{ip}^i = \text{minimum allowable velocity of train } i \in I \text{ on segment } p \in P
\]

\[
\nu_p = \text{maximum achievable average velocity of train } i \in I \text{ on segment } p \in P
\]
$W_i = \text{initial priority of train } i \in I \text{ (highest for passenger trains)}$

$S^{i}_{q} = \text{scheduled stop time for train } i \in I \text{ at station } q \in Q.$

An illustration of the ordering of a single track used for the model in this paper is given in Figure 1 where the set of stations is represented by $Q=\{1,2,\ldots,N_{S}\}$ and here, track $(p-2) \in P_{2}$.

Model Derivation

The objective function used in the model takes the following form:

$$\text{Min } \sum_{i} W_i \ast (\text{delay of train } i \in I \text{ at destination}) + \text{Train Operating Costs} \quad (1)$$

For the purposes of the solution procedure (namely Branch and Bound), the delay of train $i \in I$ is comprised of two parts. These are the current delay of train $i \in I$ at any point in time and a lower bound estimate of remaining overtake and crossing delay from this point (Higgins et al. (6)). The model is subject to various constraints to ensure safe operation, enforce speed restrictions and permit stops. The following and overtake constraints for outbound trains $i,j \in I$ are as follows:

$$M \ast C_{ijp} + X_{aq}^{i} \geq X_{aq}^{j} + h_{p} \quad p \in P_{1} \text{ and } i,j > m \quad (2)$$

$$M \ast (1 - C_{ijp}) + X_{dq}^{i} \geq X_{dq}^{j} + h_{p} \quad p \in P_{1} \text{ and } i,j > m \quad (3)$$

and for inbound trains $i,j \in I$:

$$M \ast A_{ijp} + X_{aq}^{i} \geq X_{aq}^{j} + h_{p} \quad p \in P_{1} \text{ and } i,j \leq m \quad (4)$$

$$M \ast (1 - A_{ijp}) + X_{dq}^{i} \geq X_{dq}^{j} + h_{p} \quad p \in P_{1} \text{ and } i,j \leq m \quad (5)$$
Equation (2) implies that if train \( j \in I \) goes first, then train \( i \in I \) must depart station \( q \in Q \) after train \( j \in I \) plus the minimum headway, and arrive at station \( (q+1) \in Q \) after train \( j \in I \) plus the headway. Equation (3) is similar except train \( i \in I \) goes first. Equations (4) and (5) are the same as equations (2) and (3), but for inbound trains. The constraints for the case when two trains approach each other are:

\[
\begin{align*}
  h_p + X_{aq}^i &\leq X_{dq}^j + M_B & p \in P, \quad (i \leq m, j > m) \\
  h_p + X_{aq}^j &\leq X_{dq}^i + M_B & (1 - B_{pq}) & p \in P, (i \leq m, j > m)
\end{align*}
\]

Equation (6) implies that if outbound train \( j \in I \) goes first, inbound train \( i \in I \) must depart station \( q \in Q \) after train \( j \in I \) arrives plus a safety headway. Constant \( M \) is chosen large enough so that both equations in each crossing and overtake constraint are satisfied.

Given the upper and lower velocities for each train on each segment, the upper and lower limits for traversal time of train \( i \in I \) on segment \( p \in P \) are given by:

\[
\begin{align*}
  \frac{d_p}{\bar{v}_p} &\leq X_{aq}^i - X_{dq}^j & i > m, p \in P \\
  \frac{d_p}{\bar{v}_p} &\geq X_{aq}^i - X_{dq}^j & i \leq m, p \in P
\end{align*}
\]

To stop trains from departing before their scheduled times and trains departing intermediate stations before they arrive, the following constraints are included:

\[
\begin{align*}
  X_{aq}^i &\geq Y_{qi} & i \in I, q \in Q \\
  X_{aq}^i + S_q &\leq X_{dq}^i & i \in I, q \in Q
\end{align*}
\]

The objective is to minimise equation (1) subject to constraints given by equations (2) - (8).
**Solution Procedure**

The solution procedure described in this section is based on Branch and Bound (BB) and uses the depth first search for the resolution of conflicts. Each node in the BB tree represents a partially resolved schedule which is calculated by solving a nonliner program (ie. solve objective function (1) st (7,8) and the appropriate overtake or crossing constraints from equation (2)-(6)). The lower bound to the conflict delay costs of the remaining conflicts is calculated after the partial schedule is determined and is added to the cost of the partially resolved schedule. The BB technique used is described in full detail in Higgins et al (6).

**Model Testing**

The exact algorithm of sections 3 and 4 are implemented in FORTRAN on a 80486 PC. To solve the non-linear programs, GAMS/MINOS 5.2 (Brooks et al (7)) is accessed from the FORTRAN program. The model was tested on train schedules varying from 9 trains to 49 trains and was compared to a Branch and Bound Procedure with a lower bound calculated by relaxing the remaining conflict constraints. The method in this paper was able to find the optimal solution with up to 30 times less evaluations of the nonlinear program for most problems. For most problems the first upperbound was the optimal solution. The method using a lower bound calculated by relaxing the remaining conflicts required anything from a few hundred evaluations to several thousand. This is very important for a real life scheduling system as a solution would be required within a set time limit. The problem represented in Figure 2 contains 30 trains (53 conflicts) and was solved with 13 times fewer evaluations when the improved lower bound estimate is used.
PART II: OPTIMUM LOCATION OF SIDINGS

Past Research

Since most of the work done in determining the best positions of sidings uses simulation, optimal strategies are not usually found. The limited literature which does consider optimality of siding positions only assumes simple train movements.

Petersen and Taylor (8) investigated an analytical model to determine the required number and length of sidings for a schedule of passenger trains. The determination of the length of the siding was to obtain the maximum benefit of the acceleration and deceleration characteristics of the trains.

An approach was taken by Kraft (9) to derive an analytical equation for determining the best position between two yards to put a siding. To construct the model, free running time between sidings, average running speed (including delays) and the number of trains per unit time was considered. The model cannot consider multiple sidings simultaneously. The equation may however be useful as an initial estimate. Mills et al (10) use simulation, analytic and heuristic techniques to investigate line capacity of a mine to port track system. The analytic model which determines the optimal number of equally spaced sidings is based on the expected crossing delay to a train.

None of the above literature considers solving for optimal position of sidings without the assumption constant velocities and equally spaced sidings. The remainder of this paper considers the model formulation and solution to this problem.
Model Development

In this section the analytical model is formulated and the main feature is the treatment of the track segment lengths (or siding positions) as variable. Some sidings will be located at fixed positions and are not considered as variable in the model. This occurs when the siding is already existing or if is to serve another purpose besides resolving conflicts. Scheduled stops will only be permitted at fixed sidings (stations).

Assumptions

The following assumptions are made specifically in conjunction with the siding location model:

- Double line track sections are allowed and can be solved for optimum length but its position in a string of track segments cannot be moved.
- Generally, only one train can occupy a siding at one time (unless specified) except for the origin and destination stations which are assumed to have infinite capacity.
- Scheduled stops are only permitted on fixed sidings.
- The train schedule is a cyclic schedule for which it is repeated on a daily or weekly basis.

The following information (in addition to the information required in the first part of this paper) is required by the user:
• The upper velocities of each train at 1-km intervals of the track. These are used to approximate the upper velocities of trains on track segments as they are considered variable during the calculation procedure.
• Any cost parameters such as cost of lateness per time and train operating costs.
• Initial positions of the sidings. A good initial solution will ensure fast convergence.

Definition of Variables

The variables used in the siding location problem is the same as the first part of this paper except for the following differences. The sidings are represented by the set \( Q = \{ 1, 2, \ldots, NS \} \) where \( NS \) is the total number of sidings in the track system. Let \( Q_1 \) represent the set of fixed stations (sidings) and \( Q_2 \) represent the set of variable sidings.

If the train schedule considered consists of daily and weekly trains then the cycle will be one week (ie. the schedule considered is one cycle). The scheduled stop time is defined as:

\[
S_{iq} = \text{scheduled stop time for train } i \in I \text{ at station } q \in Q_i
\]

The upper velocity of a train on a discrete interval of track (used when calculating the upper velocities on a track segment) is given by

\[
vel_{ig} = \text{upper velocity of train } i \in I \text{ at distance interval } g \text{ on the track}
\]

Assume the minimum headway is given by \( h \). It does not however have to be constant for all trains and track segments since the minimum headway may be train dependent or determined by signal points.

Formulation and Constraints
The objective function will generally take the form of minimising train delay costs and train operating costs. A dynamically prioritised delay criterion which allows the priority of each train to change from origin to destination as discussed in the model given in part I. Objectives involving minimising destination lateness of trains are found in Kraay et al (3) and Mills et al (5), while Petersen et al (11) minimises total travelling times. Although the model in this paper does not depend on the objective used, it is important however for the objective function to be convex to avoid the location of local optima. The overtake, crossing, upper velocity and scheduled stop constraints are the same as those given in part 1 of this paper.

Since the track segments are of variable position and length (during the solution procedure), the upper velocities must be approximated. To estimate the upper velocities (maximum achievable velocities) it will be assumed the upper velocities on each one kilometre interval of the track corridor are known (or calculated using a train movement simulator). If one kilometre intervals are too fine then larger intervals may be used. Since the problem will be solved iteratively, the upper velocity of a train on a track segment is calculated by taking the average upper velocity of the intervals that lie in the track segment of the current solution. The upper velocity of train \( i \in I \) on track segment \( p \in P \) is calculated by the following equations:

\[
\bar{v}_p^i = \frac{\sum_{g=dl}^{dh} \text{vel}^i_g}{dh - dl + 1}
\]

where

\[
dl = \text{integer part of } (\sum_{k=1}^{p-1} d_k) + 1, \quad dh = \text{integer part of } (\sum_{k=1}^{p} d_k)
\]
The expected arrival times at intermediate sidings are also dependent on the positions of sidings and are calculated by first determining the planned velocities on the track segments. The planned velocities are calculated as follows:

\[ \text{Ratio of fastest journey time to expected journey time} \quad RA_i = \frac{Y_{\text{fin}}^i - Y_{\text{ori}}^i}{\sum_g \frac{1}{vel_g^i}} \]

The planned velocity of train \( i \in I \) on track segment \( p \in P \), \( PV_i^p \), is \( \frac{\bar{v}_i^p}{RA_i} \).

From the planned velocity, the expected departures from each of the intermediate stations are calculated using equation (10). The expected arrival times will be the same as the expected departure times unless there are scheduled stops.

\[ Y_{d_{\text{out}}}^i = Y_{\text{ori}}^i + \sum_{k=1}^{g-1} \frac{d_k}{PV_{i,k}^i} \quad \text{train } i \in I \text{ is outbound} \]

\[ Y_{d_{\text{in}}}^i = Y_{\text{ori}}^i + \sum_{k=q}^{TRP} \frac{d_k}{PV_{i,k}^i} \quad \text{train } i \in I \text{ is inbound} \]  

(10)

where \( TRP \) is the number of track segments on the rail corridor. The fastest times which the trains travel from origin to destination are assumed to not be affected by the siding positions so the expected arrivals and departures at the these sidings do not change. The last constraint is to ensure that the sum of the length of the track segments is equal to length of the entire track corridor ie.

\[ \sum_{k=1}^{i-1} d_k = TLEN_i \quad i \in Q_i \]  

(11)

where \( TLEN_i \) is the length of the track system form the origin to the fixed siding \( i \in Q_i \).

**Solving the Model**

In this section, a decomposition procedure is presented to obtain a solution to the above formulation. Solving the problem as formulated can be difficult due to the required
solution of three sets of variables (track segment lengths, arrival/ departure times and binary conflict resolution variables). The conflict resolution binary variables are solved using a branch and bound type procedure (or a heuristic) and require the sidings to be at fixed positions. The problem must be decomposed so that solutions can be obtained for the three sets of variables.

The decomposition procedure proposed here is different to the Generalised Benders Decomposition (GBD) by Geofferon (12). The GBD partitions the model via the set of continuous variables and the set of integer variables. A more efficient way would be to partition the problem so that the structure of the problem could be exploited. This will allow a more efficient means of solving the sub-problems to be used. The model here will be decomposed into two sub models, one which is solved for track segment lengths and arrival and departure times, the other which is solved for the optimal train schedule given the track segment lengths. The process will iterate between the two sub-problems until there is no more improvement. This type of decomposition procedure is popular when solving complicated routing and scheduling problems. When one set of variables are fixed, the problem can sometimes be reduced to a well known form which can be easily solved using common procedures or heuristics. Two good examples are found in papers by Koskosidis et al (13) which looks at the soft time window constraints for the vehicle routing problem, and Sklar et al (14) which considers the aircraft scheduling problem.

The complete model for this paper can be stated by equation (12):

\[
\begin{align*}
\min Z &= f(d_k \forall k, X_{aq} \forall i,q, X_{aq} \forall i,q, A_{ijp} B_{ijp} C_{ijp} \forall i,j,p) \\
\text{st} \quad \text{constraints (2-11)}
\end{align*}
\]
where \( f(\bullet) \) represents the non-linear (or linear) objective function of the variables defined in the first part of this paper. The model is decomposed to form models \( Z_1 \) and \( Z_2 \). The model \( Z_1 \) which is represented by equation (13) is solved to obtain the optimum track segment lengths subject to fixed conflict resolution variables \( A_{ijp}, B_{ijp}, C_{ijp} \) (ie. fixed schedule). The model \( Z_2 \) is solved to obtain the optimum schedule subject to fixed track segment lengths (ie normal train scheduling problem). Each model is solved using the output from the other model as initial values.

\[
\begin{align*}
\text{Min } Z_1 &= f(d_k \forall k, X^i_{dq} \forall i,q, X^i_{aq} \forall i,q) \\
\text{st } &\text{ constraints (7-11)}
\end{align*}
\]

\[
\begin{align*}
\text{Min } Z_2 &= f(X^i_{dq} \forall i,q, X^i_{aq} \forall i,q, A_{ijp}, B_{ijp}, C_{ijp} \forall i,j,p) \\
\text{st } &\text{ constraints (2-9, 11)}
\end{align*}
\]

The upper velocities of model \( Z_1 \) will be those of the latest solution of model \( Z_2 \). This is reasonable since to have the upper velocities as a function of track segment lengths \( d_k \) (which is variable in model \( Z_1 \)) would require non-linear constraints. This may cause the solution to model \( Z_1 \) to be slightly inaccurate for the first couple of iterations if there is a large change in siding positions. Results generated in the next section have indicated little effect on the convergence.

The following variables will be defined for the decomposition algorithm to resemble the current stage of solution.

\[
\begin{align*}
d^t_k &= \text{ length of track segment } k \in P \text{ after the } t^{\text{th}} \text{ iteration using model } Z_1 \\
X^{t,i}_{dq} &= \text{ departure time of train } i \in I \text{ from station } q \in Q \text{ after the } t^{\text{th}} \text{ iteration using model } Z_1 \\
X^{t,i}_{aq} &= \text{ arrival time of train } i \in I \text{ at station } q \in Q \text{ after the } t^{\text{th}} \text{ iteration using model } Z_1
\end{align*}
\]
The expected departure times are calculated using equation (10) and these constraints will be linear since the planned velocities are constant. This is because the upper velocities from model $Z_2$ is used in the current iteration of model $Z_1$. The initial track segment lengths $d_k^0$ can be estimated using simulation techniques or by a simple inspection to see where the conflicts occur. Another method is to just assume equal track segment lengths for the initial estimates. If the purpose is to upgrade an existing track corridor, then the current positions of some existing sidings may be used for the initial estimate.

The optimum siding positions are calculated using the following decomposition procedure:

1. Given initial values $d_k^0 \forall k$ solve the model $Z_2$ to obtain $X_i^{l,2}_{dq}, X_i^{l,2}_{aq}, B_{ijp}^1, A_{ijp}^1$ and $C_{ijp}^1 \forall i, j, p$. Solving $Z_2$ is exactly the same as solving the normal train scheduling problem (Higgins et al (6) and Kraay et al (3)). Let $t=1$.

2. Given $B_{ijp}^t, A_{ijp}^t$ and $C_{ijp}^t$, solve the non-linear program $Z_1$ for $d_k^t, X_i^{l,1}_{dq}$ and $X_i^{l,1}_{aq}$. This part is not a computational burden but the objective function is more complex due to $d_k^t$ being variable. The form of this model makes it suitable for solution using a simplical decomposition procedure (Hohenbalken (15)).
3. Let \( t = t+1 \). Solve the problem \( Z_2 \) given \( d_k^{t-1} \) for \( X_{i,t,2}^{q}, X_{i,t,2}^{a}, B_{ijp}^{t-1}, A_{ijp}^{t-1} \) and \( C_{ijp}^{t-1} \) using \( X_{i,t,1}^{q}, X_{i,t,1}^{a}, B_{ijp}^{t-1}, A_{ijp}^{t-1} \) and \( C_{ijp}^{t-1} \) as initial values. The procedure terminates when the conflict resolution strategy does not change from iteration \( t-1 \) to \( t \). It is a major computational burden to solve for the integer variables using a branch and bound procedure. It is required for the initial solution of step 1., but if only a couple of conflict resolutions change as the positions of the sidings converge, then a much more efficient method of updating the conflict resolution strategy is necessary. A heuristic for this is described in the paper by Higgins et al (16). Goto step 2.

**Model Testing**

The examples considered here contain 7 trains and 6 sidings, 4 of which are movable. The examples were chosen to illustrate the time savings of having sidings at their optimal positions compared to their current positions. The objective function chosen for examples is minimum tardiness plus fuel cost and is given in equation (16). The fuel consumption function is the same as that used by Mills et al (5). For the two examples presented here, the restriction of one train per siding is relaxed. The lateness at destinations (second term of equation (16)) for the initial and optimal solution (both examples) are shown in Table 1(a) with the track segment lengths given in Table 1(b). Figure 3a represents the initial resolved train graph for the first example. By inspection of this train graph, it appears that the sidings are at quite reasonable positions with respect to the conflicts. The only real indication is that siding 2 could be closer to the inbound origin station. When the sidings are at optimum positions as shown in Figure 3b, considerable time savings are obtained for the trains and they are kept closer to schedule throughout the journey. The second and fourth columns of Table 1(a) indicate the time saved for trains when sidings are at optimal positions. More than an hour of delay has been cut for all trains.
The first example required only 2 iterations (terminated at $t=2$) of the decomposition procedure to achieve the optimal solution. Only one conflict required changing from the first iteration to the second. The original track segment lengths $d_k^0$ indicate that a good initial solution will ensure fast convergence. The initial positions of the track segments in the second example are a lot poorer than in the first. This example was set up so that most of the train interactions are toward the mid point of the journey where there are fewer sidings. The outbound trains suffer heavy delays due to this and the optimal solution relocates the sidings towards the middle of the train graph. Referring to the third and fifth columns of Table 1(a), there has been a reduction in delay for many trains with the average delay being significantly reduced.

**MODEL LIMITATIONS**

The models in both parts of this paper have some limitations as far as real life applications are concerned. While the emphasis of part I was to allow optimal solutions to real life problems to be found, it does not allow random delay events. Instead the risk delay of a given resolved schedule can be calculated separately by a model described in Higgins et al (17) The model takes into account the risk delay due to terminal and stoppage delays, train related delays and track related delays.

Some trains will have different characteristics such as number of wagons on a given day and number of locomotives used. The upper achievable velocity is easily adjusted to cater for such train differences. At this stage the models have been tested using a real life problem which consists of a single line track of 120 kilometres with 13 sidings and a daily density of 30 trains. Most types of objective functions which have been proposed in
the past can be accommodated using the models. However, the inclusion of other variables such as delay risk would not be possible.

**CONCLUSIONS**

This paper has presented an on-line model for the scheduling of trains on a single line track and a planning tool for determining the optimal positioning of sidings. The on-line model allows the priority of a train to change from origin to station resulting in a more reliable system. This is a more realistic interpretation of how the train dispatcher would consider the rail network. Conflicts are resolved upon their current priorities which are dependent on the future delays for each train.

The on-line model will be useful to train dispatchers for generating more reliable train schedules. Optimum schedules will be generated quickly and trains will be kept on schedule with respect to future delays. The results have demonstrated significant computation time improvements especially for larger problems which involve tight schedules.

For the siding location problem, a decomposition procedure was used iteratively to solve for the best siding positions and corresponding resolved schedule. Results of the model have shown much improvement in delays to trains when the sidings are at optimal positions. If using this model to determine the positions of sidings only reduces the overall delay by a small percentage (while keeping the train costs uniform) then the long term benefits may be large.
Positioning of sidings is one aspect of trying to optimise freight rail transport. A larger concern however is the upgrading of existing track. It is important to know the effects of lateness and reliability of schedules when upgrading the existing track corridor. Besides the delay occurred when a train waits at a siding for another train to pass, there are other delays that must be considered. These delays are categorised as risk delays and are caused by maintenance, any train failure or environment problems. A prime interest for upgrading existing track is the knowledge of this risk delay while the track is in its current state and an estimate of this delay if certain upgrading was carried out. Research is continuing by the authors on the development of a model to estimate the risk delay to a system and to identify which sections of track contribute the most risk.

REFERENCES


### TABLE 1a Comparison of lateness at destinations

<table>
<thead>
<tr>
<th>Train</th>
<th>Destination lateness for current solution (hrs)</th>
<th>Destination lateness for optimal solution (hrs)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Example 1</td>
<td>Example 2</td>
</tr>
<tr>
<td>1</td>
<td>0.01</td>
<td>0.00</td>
</tr>
<tr>
<td>2</td>
<td>1.62</td>
<td>0.46</td>
</tr>
<tr>
<td>3</td>
<td>1.86</td>
<td>0.38</td>
</tr>
<tr>
<td>4</td>
<td>0.00</td>
<td>0.21</td>
</tr>
<tr>
<td>5</td>
<td>0.38</td>
<td>0.42</td>
</tr>
<tr>
<td>6</td>
<td>1.15</td>
<td>1.07</td>
</tr>
<tr>
<td>7</td>
<td>1.09</td>
<td>0.40</td>
</tr>
<tr>
<td>All Trains</td>
<td>6.11</td>
<td>2.94</td>
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</table>

### TABLE 1b Comparison of original and optimal track segment lengths

<table>
<thead>
<tr>
<th>Track segment k</th>
<th>Original length ((d_k^0 \text{ km}))</th>
<th>Optimal length (km)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Example 1</td>
<td>Example 2</td>
</tr>
<tr>
<td>1</td>
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<td>8.18</td>
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<tr>
<td>2</td>
<td>28.86</td>
<td>35.40</td>
</tr>
<tr>
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<td>28.47</td>
<td>35.40</td>
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<td>43.14</td>
<td>43.52</td>
</tr>
<tr>
<td>5</td>
<td>25.26</td>
<td>23.46</td>
</tr>
</tbody>
</table>
FIGURE 1  Sample of a network showing the single and double track segments
FIGURE 2  Optimal solution of 30 train problem
TABLE 1a Comparison of lateness at destinations

TABLE 1b Comparison of original and optimal track segment lengths

FIGURE 1 Sample of a network showing the single and double track segments

FIGURE 2 Optimal solution of 30 train problem

FIGURE 3 Optimal schedule given current siding positions (left), given optimal siding positions (right)
FIGURE 3  Optimal schedule given current siding positions (left), given optimal siding positions (right)