

MILP global search algorithm for Aircraft ground movement optimization

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Guide

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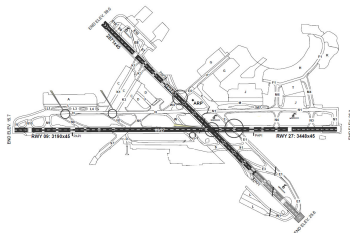
Section 1

Introduction

Air Traffic Control (ATC)

Ground movement

- Arriving & Departing aircrafts with **tentative** times
- Origin - Destination points
- Route through network of taxiways
- Solve overlap conflicts; minimize overall delay



Motivation

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Current ATC

- Prominently manual scheduling by ATCs: First come first serve (myopic)
- Over-scheduling by airlines leading to unrestrained push-backs
- Scheduling primarily empirical and based on 'gut'
- 12.5% flight delays in USA in 2012 estimated to have occurred near airports

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Need

- Testable and realistic method for optimal scheduling
- Run-time comparable to current semi-automated systems
- Globally optimal schedule to benchmark the method

Attempts: Commercial

- **CDM** (Collaborative Decision Making):
Collaborative data assimilation and dissemination
- **ASMGCS** (Advanced Surface Movement Guidance and Control Systems):
Automated surface surveillance and scheduling
- **NextGen** (Next Generation Air Transportation System):
De-centralized collaborative decision making and planning

Attempts: Academic

Macro-optimization

- *Capacity estimation*: Generate airport capacity envelope
- *Push-back optimization*: Generate optimal push-backs for aircrafts at terminals
- *Weather mitigation*: Models to predict weather and improve decision making

Micro-optimization

- *Stochastic*: Use heuristics and intelligent elimination to generate schedule
- *Deterministic* \ *Semi-deterministic*: Cut down solution space to deterministically solvable one; find optimal schedule

Micro-optimization

- Airport map interpreted as a graph
- Detailed schedule for each aircraft; time stamps at nodes
- Better delay minimization than macro-optimization

Concerns

- Highly dynamic; daily changing schedules may cause discomfort to ATCs and pilots
- Heavy need of ATC - pilot interaction over an already congested radio network
- Justifiable mistrust amongst ATCs on an automated system; heavy testing required

Micro-optimization

Stochastic

- **Genetic algorithm** (Liu):
 - First N paths considered for each aircraft
 - Binary decision variable corresponding to each arc
 - Output: Chromosome chain of decision variables
- **Genetic + A*** (Gotteland):
 - Genetic algorithm for path selection for each aircraft
 - A* for prioritizing aircrafts during conflicts
- **Bacterial Foraging** (Bijal):
 - Dijkstra's algorithm for shortest path allocation
 - Bacterial Foraging (BAFO) for conflict resolution & delay allocation

Micro-optimization

Deterministic & Semi-deterministic

- **MILP** (Smeltink):
 - Fixed (shortest) paths allocated to aircrafts
 - MILP model used for conflict resolution and delay allocation
 - Rolling Horizons for realistic run-times
- **Time Dependent Shortest Path (TDSP)** (Baik):
 - Aircrafts prioritized on first come first serve
 - One aircraft scheduled per iteration
 - TDSP generated from a time-space graph
- **MILP+TDSP** (Gupta):
 - TDSP used to generate schedule
 - Paths allocation based on TDSP schedule
 - MILP used for conflict resolution

Section 2

Global Search

Global optimum

Need of global optimum to benchmark other methods

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Each aircraft

- has multiple choices of routes
- can wait anywhere on the route to solve overlap conflicts
- can wait for any duration anywhere on the route to do so

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Infinitely many solutions?

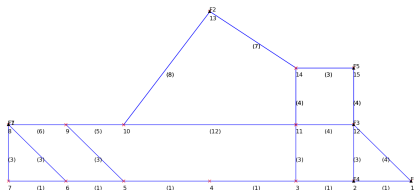
Requirements

Algorithm must:

- deliver **at least one** globally optimal solution: any other solution at best may be as good
- take a (possibly long but) finite run-time
- deliver a realistic solution
- be such that, **approach** ensures global optimality of solution

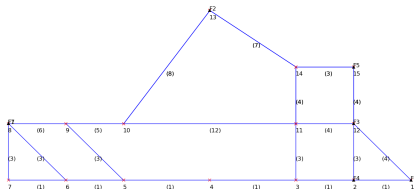
Assumptions

Assumptions



Map discretized to arcs & nodes

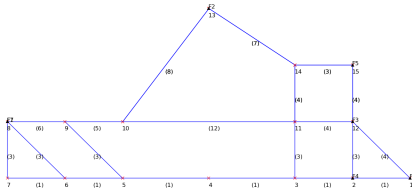
Assumptions



Map discretized to arcs & nodes

Time generally discretized into slots (10 secs)

Assumptions



A



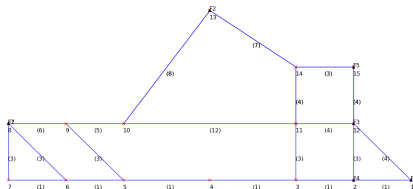
B

Waiting: **A** and **B** are equivalent w.r.t. delay

Map discretized to arcs & nodes

Time generally discretized into slots (10 secs)

Assumptions



Map discretized to arcs & nodes

Time generally discretized into slots (10 secs)



A



B

Waiting: **A** and **B** are equivalent w.r.t. delay

- Fixed origin-destination & start-time
- Infinite acceleration & Constant velocities
- Aircrafts: Point objects

Parameters

Parameters

Inputs

- Map : Interpreted as a graph
- Aircrafts
 - Starting time
 - Origin
 - Destination
 - Aircraft speed
 - Trailing separation
 - Priority

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Objective

Total time : Summation over the destination times of all aircrafts

Parameters

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- Map : Interpreted as a graph
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 - Origin
 - Destination
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 - Trailing separation
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Objective

Total time : Summation over the destination times of all aircrafts

Output

Schedule : Time stamps at every node for each aircraft

Algorithm

Path Finding

Flight-Path combinations

Local Mimima

Global Minimum

Post processing

Algorithm

Path Finding

- All origin-destination pairs identified from given aircrafts
- DFS applied for each pair: all possible paths explored
- DFS breaks on cycle-detection to avoid infinite loops

Flight-Path combinations

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Path Finding

Flight-Path combinations

All possible combinations of flight-path allocations generated from aircrafts and corresponding path pools

Local Mimima

Global Minimum

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Algorithm

Path Finding

Flight-Path combinations

Local Mimima

- For every flight-path combination, MILP run
- Conflict free schedule **deterministically** generated
- Local minimum for particular flight-path combination identified

Global Minimum

Post processing

Algorithm

Path Finding

Flight-Path combinations

Local Mimima

Global Minimum

All local minima compared and **global minimum** identified

Post processing

Algorithm

Path Finding

Flight-Path combinations

Local Mimima

Global Minimum

Post processing

- Feasibility check run on solution to detect spacio-temporal overlaps
- Feasible, globally optimal solution animated in real-time
- 3D space-time plot of solution generated

Section 3

MILP Design

Mathematically

Identify a schedule with minimum total delay for given flight-path combination

t_{ik} : Time when aircraft i reaches node k / enters arc k

P_i : Priority of aircraft i ; landing aircrafts have higher priority

Objective

Minimize : $\sum_{i=0}^N P_i t_{in}$

N = Total no. of flights

n = Last / Destination node in the path of flight i

Output

Values of t_{ik} for $i \in (0, N)$ & $k \in (0, n)$

Constraints

- Every aircraft has a maximum speed
- Aircrafts entering an arc from opposite directions must do so exclusively of each other in time
- Minimum separation equal to the minimum amongst (trailing separation of leading aircrafts) and (arc length) must always be maintained between an aircraft following another

Constraints : Travel time

Travel time on an arc should be greater than or equal to the
(arc length)/(aircraft speed)



- $t_{i(k+1)} - t_{ik} \geq \frac{length_{ik}}{speed_i}$

Constraints : Head-on collision

2 aircraft entering an arc from opposite directions must have no temporal overlap



- $M \times x_a + t_{ik} - t_{j(l+1)} \geq 0$
- $M \times (1 - x_a) + t_{jl} - t_{i(k+1)} \geq 0$

$$M \gg 0 \text{ \& } x_a \in [0, 1]$$

J leads ($x_a = 0$)

$$t_{ik} - t_{j(l+1)} \geq 0$$

I leads ($x_a = 1$)

$$t_{jl} - t_{i(k+1)} \geq 0$$

Constraints : Common Node 1

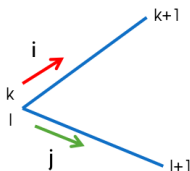
For aircraft i following aircraft j while entering arcs with 1st node common:

if *length of the arc of j is less than it's trailing separation:*
then i must enter it's arc only after j has exited it's own

Else, if *length of arc of j is greater than or equal to it's trailing separation:*

i may enter it's arc only after j is at least at a distance equivalent to it's trailing separation from common node 1

and visa-versa.



Constraints : Common Node 1

- If j leads & length of arc l is less than sep_j , then i enters arc k only after j has exited arc l i.e. reached node $(l + 1)$
- If j leads & length of arc l is greater than or equal to sep_j , then i enters arc k after j has traversed distance of at least sep_j on arc l
- If i leads & length of arc k is less than sep_i , then j enters arc l only after i has exited arc k i.e. reached node $(k + 1)$
- If i leads & length of arc k is greater than or equal to sep_i , then j enters arc l after i has traversed distance of at least sep_i on arc k

- If $Length_l < sep_j$,

$$M \times x_b + t_{ik} - t_{j(l+1)} \geq 0$$

- Else (if $Length_l \geq sep_j$),

$$M \times x_b + t_{ik} - t_{jl} \geq \frac{sep_j}{speed_j}$$

- If $Length_k < sep_i$,

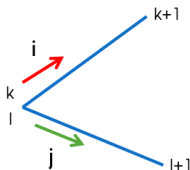
$$M \times (1 - x_b) + t_{jl} - t_{i(k+1)} \geq 0$$

- Else (if $Length_k \geq sep_i$),

$$M \times (1 - x_b) + t_{jl} - t_{ik} \geq \frac{sep_i}{speed_i}$$

$$M \gg 0 \text{ \& } x_b \in [0, 1]$$

Constraints : Common Node 1



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I leads ($x_b = 1$)

$$Length_l < sep_j$$

$$t_{ik} - t_{j(l+1)} \geq 0$$

$$t_{jl} - t_{i(k+1)} \geq 0$$

$$Length_k < sep_i$$

$$Length_l \geq sep_j$$

$$t_{ik} - t_{jl} \geq \frac{sep_j}{speed_j}$$

$$t_{jl} - t_{ik} \geq \frac{sep_i}{speed_i}$$

$$Length_k \geq sep_i$$

$\frac{sep_j}{speed_j}$ is the time taken by j to traverse a distance of sep_j

Constraints : Common Node 2

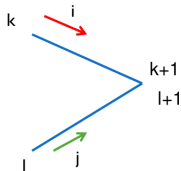
For an aircraft i following aircraft j while entering arcs with 2^{nd} node common:

if *length of the arc of i is less than j 's trailing separation*: then i must enter it's arc only after j has exited it's own arc

Else, if *length of arc of i is greater than or equal to j 's trailing separation*:

then j must exit it's arc when i is at least at a distance equal to j 's trailing separation from the common node 2

and visa versa



Constraints : Common Node 2

- If j leads & length of arc k is less than sep_j , then i enters arc k only after j has exited arc l i.e. reached node $l + 1$
- If j leads & length of arc k is greater than or equal to sep_j , then j exits arc l when the distance of i is at least equal to sep_j from node $k + 1$
- If i leads & length of arc l is less than sep_i , then j enters arc l only after i has exited arc k i.e. reached node $k + 1$.
- If i leads & length of arc l is greater than or equal to sep_i , then i exits arc k when the distance of j is at least equal to sep_i from node $l + 1$

- If $Length_k < sep_j$,

$$M \times x_c + t_{ik} - t_{j(l+1)} \geq 0$$

- Else (if $Length_k \geq sep_j$),

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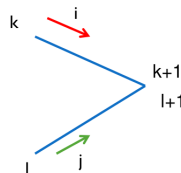
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$Length_l < sep_i$

$Length_k \geq sep_j$

$$t_{i(k+1)} - t_{j(l+1)} \geq \frac{sep_j}{speed_i}$$

$$t_{j(l+1)} - t_{i(k+1)} \geq \frac{sep_i}{speed_j}$$

$Length_l \geq sep_i$

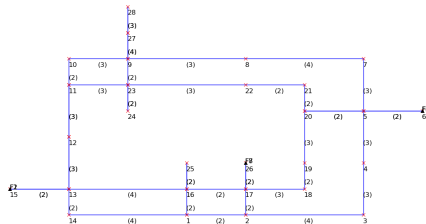
$\frac{sep_j}{speed_i}$ is the time taken by i to traverse a distance of sep_j

Section 4

Results

Roling

Flight	Origin	Destination	Start time	Speed	Trailing sep	Priority
1	26	15	7	1	3	1
2	24	15	6	2	3	1
3	25	6	10	2	3	1
4	25	6	8	1	3	1
5	25	6	16	2	3	1
6	24	6	14	1	3	1
7	28	26	0	1	3	1
8	28	26	3	1	3	1



Cost

Roling: 362

Global search: 204

Run time: 51 mins

Future work

- Global optimum identified
- Global search can not be used on the fly: Heavy run-time
- Stage II: Need method to deliver **fast** and **near-optimal** solutions

Direction

- Hybrid approach: Stochastic + Deterministic
- Rolling horizons
- Learning from empirical data

Merci! :)