Air Traffic Flow Management at Airports: A Unified Optimization Approach

Dimitris Bertsimas & Michael Frankovich MIT and Aviation Edge LLC

Motivation for *Unified Airport* Optimization

- Airports = key bottlenecks
- Current approaches in general solve subproblems in isolation
 - Sub-optimality
 - Overall infeasibility
- \Rightarrow Unified Approach



Key Sub-problems at Airports



- Minimum separation between flights
- Depends on flight *types* Arrival, departure; heavy, B-757, large, small

Runway Sequencing – Literature

- TSP/TRP
- Single runway
- Constrained position shifting
- Dynamic programming
- Time windows, side constraints
- Stochastic runway scheduling

- Dear (1976)
- Psaraftis (1980)
- Trivizas (1987)
- Balakrishnan and Chandran (2010)
- Clare and Richards (2011)
- Sölveling et al. (2011)



Runway Configuration Management

- Rules regulating runway use
- Runway configuration = – Combination of runways
 - Operating modes
- Which configuration?
- When?
- Literature:
 - Bertsimas, Frankovich, and Odoni (2011)



Flight Routing

- Route flights to assigned runway and beyond
- Determine gate-holding/speed control, if any



*Diagram thanks to Dr Tom Reynolds

Flight Routing – Literature

- Surface management:
 - Feron et al. (1997), Pujet et al. (1999), Carr (2001), Burgain (2010),
 Simaiakis et al. (2011)
- Taxi route optimization:
 - Marín (2006), Rathinam et al. (2008), Roling and Visser (2008), Malik et al. (2010)
- Sequencing & taxiing:

- Gotteland et al. (2009), Clare and Richards (2011)

Airport Operations Optimization Problem (AOOP)

- Selecting a runway configuration sequence
- Determining the service rate of arrivals and departures
- Assigning flight types to runways and determining their sequence
- Determining the gate-holding duration of departures
- Routing flights to their assigned runway and onwards within the terminal area

Summary of Approach

Two phases, each a binary optimization problem

- Decomposition motivation:
 - key bottleneck of system at runways
 - so initially ignore capacity of gates, taxiways, airspace
- Phase I (Configurations, sequencing, assignment)
 - Optimal under assumption
- Phase II (Routing)
 - Uses phase one solution
 - Tractably solves AOOP without assumption

Phase I Decisions



Phase I – Key Data

- Time horizon 1,...T, @20s intervals
- Configuration availability U_t
- For each flight f:
 - A flight type i
 - Set of feasible runways R_f
 - For each runway r in R_f :
 - Earliest possible takeoff/landing time \underline{T}_{r}^{f}
 - Computed based on shortest paths
- Objective weighted cost of delays

Phase I – Decision Variables

- ω_{kt} = 1 \Leftrightarrow configuration k is used at time t
- $\chi_t = 1 \Leftrightarrow$ we change configuration at time t
- $\phi_r^f = 1 \Leftrightarrow flight f is assigned to runway r$

 $\psi^{i}_{\mbox{ rt}}$ = 1 \Leftrightarrow a flight of type i reaches runway r at t

- Naïve approach: ψ^{f}_{rt}
 - Separation constraints
 - Model size
- Flight-slot assignment guaranteed

 $\min \ \Psi$

s.t.
$$\sum_{k \in \mathcal{K}} \omega_{kt} = 1, \quad \forall t \in \mathcal{T},$$
(3.2a)

$$\psi_{rt}^{i} = 0, \quad \forall i \in \mathcal{C}, r \in \mathcal{U}_{c}, t \in \mathcal{T},$$
(3.2b)

$$\psi_{r,t-h}^{i} + \psi_{rt}^{j} \leq 1, \quad \forall i, j \in \mathcal{C}, r \in \mathcal{R}_{i} \cap \mathcal{R}_{j},$$
(3.2c)

$$\psi_{r,t-h}^{i} + \psi_{r',t}^{j} \leq 1, \quad \forall i, j \in \mathcal{C}, (r, r') \in (\mathcal{R}_{i} \times \mathcal{R}_{j}) \cap \mathcal{V},$$
(3.2d)

$$\psi_{r,t-h}^{i} + \psi_{r',t}^{j} \leq 1, \quad \forall r \in \mathcal{R}, t \in \mathcal{T},$$
(3.2d)

$$\sum_{i \in \mathcal{C}: r \in \mathcal{R}_{i}} \psi_{rt}^{i} \leq 1, \quad \forall r \in \mathcal{R}, t \in \mathcal{T},$$
(3.2e)

$$\psi_{rt}^{i} + \omega_{kt} \leq 1, \quad \forall t \in \mathcal{T}, k \in \mathcal{K}, r \in \mathcal{R}_{k}, i \in \bar{\mathcal{I}}_{rk} : r \in \mathcal{R}_{i},$$
(3.2f)

$$\psi_{rt}^{i} - \sum_{k \in \mathcal{K}: r \in \mathcal{R}_{k}, } \omega_{k,t+h} \leq 0, \quad \forall i \in \mathcal{C}, r \in \mathcal{R}_{i},$$
(3.2g)

$$\sum_{r \in \mathcal{R}_{f}} \varphi_{r}^{f} = 1, \quad \forall f \in \mathcal{F},$$
(3.2h)

$$\sum_{r \in \mathcal{R}_{f}} \varphi_{r}^{f} \leq \sum_{r=1}^{t} \psi_{r\tau}^{i} \leq \sum_{f \in \mathcal{F}_{i}: r \in \mathcal{R}_{f}, \\t \geq T_{r}^{f} - U_{i}^{i}+1}$$
(3.2i)

$$\chi_{t} - \omega_{kt} + \omega_{k,t-1} \geq 0, \quad \forall k \in \mathcal{K}, t \in \mathcal{T} \setminus \{1\},$$
(3.2j)

Phase II – The "Routing Phase"



Key decision variables:

• $z_{jt}^{f} = 1 \Leftrightarrow flight f reaches node j by time t$

Phase II – Summary

- Fix runway assignments and *sequences* from P1
- Ensure routing constraints met
 - Eg taxiway/runway crossings, etc
 - Fix separation
- If capacity sufficiently high, P1 solution optimal
 - flights processed at same time in P2 as P1
 - flights travel along shortest paths, unimpeded
- Else, runway times will differ slightly

 here, need to ensure configurations respected
- Guaranteed feasible

Phase II Decisions - Routing



A Bound on the Optimality

- Phase II gives a feasible solution to AOOP
- An optimal solution to AOOP has value no better than the Phase I optimum
 - (It is the full problem, without routing constraints)
- This gives us a bound on the quality of our solution
- A large gap would indicate that the Phase I problem was far from feasible
 - Hence our assumption that runways key bottlenecks would be poor

Computational Experience – Aims

- Are our key assumptions valid?
- Is the methodology computationally tractable?
- Would the use of the methodology result in significant benefits in practice?

Computational Tractability & Bound on the Optimality Gap

Tliabta	Optimality Bound	Computational Times (s)			
riigiits		<i>P1</i>	<i>P2</i>	Total	
155	2.1	120	1286*	1430	
175	1.1	372	1071	1465	
153	0.6	64	129	211	
155	0.7	75	284	379	
168	0.5	340	187	546	
171	0.7	299	284	600	
159	0.9	252	533	806	

*=1% optimality gap

Using Data from DFW on 11/2/2009

Computational Tractability & Bound on the Optimality Gap

Flights	Optimality Bound	Computational Times (s)			
riigiits		<i>P1</i>	<i>P2</i>	Total	
90	0.2	252	147	418	
91	0.4	233	142	388	
80	0.1	143	16	168	
63	0.6	93	52	161	
63	0.3	198	15	229	
71	1.3	246	1200	1457	
59	0.5	255	59	325	

Using Data from BOS on 9/28/2010

Comparison of Optimized and Historic Surface Times

Optimized Surface Times (min./flight)			Historic Surface Times (min./flight)			
Dep. G.H.	Dep.	Arr.	Avg.	Dep.	Arr.	Avg.
1.8	9.5	10.8	10.2	13	8.9	10.7
2.2	9.2	10.7	9.9	12.8	9.3	11.2
1	9.5	10.6	10	13.5	9.2	11.6
1	9.6	11.1	10.4	13.5	9	11
0.9	9.3	11.8	10.5	13	10.1	11.6
2.1	10	11.4	10.7	13.6	8.9	11.3
1.3	9.1	10.8	10	13.9	9.1	11.4
0.7	9	11.1	10.1	13.4	9.6	11.2

Using data from DFW on 11/2/2009

Comparison of Optimized and Historic Surface Times

Optimized Surface Times (min./flight)			Historic Surface Times (min./flight)			
Dep. G.H.	Dep.	Arr.	Avg.	Dep.	Arr.	Avg.
1.6	13.7	4.4	9.3	18.7	5.2	12.4
0.8	14	4.5	9.6	16.7	6.1	11.6
2.2	16.4	5	12.2	17.6	5.9	13.1
0.5	16.6	4.9	10.3	18.2	6.5	12.2
0.5	16	5	10.1	18.5	4.4	10.9
1.3	11.6	8.2	10.1	16.6	5.1	11.4
2.6	13.8	4.6	8.5	14.7	5.8	9.9
0.9	16.2	4.9	10.5	17.1	6.6	11.3

Using data from BOS on 9/28/2010

Surface Congestion





BOS Optimized



Summary

- Unified and tractable approach to solve the entire Airport Operations Optimization Problem (AOOP)
 - Runway configuration management
 - Runway sequencing
 - Flight-runway assignment
 - Flight routing
 - Gate-holding

Limitations

- Who are the decision-makers?
 Can we implement our solution?
- Uncertainty see Frankovich (2012)
- Nevertheless:
 - Useful tool for measuring airport performance
 - Analysis of airport infrastructure changes

Thank You