

Mathematical Optimization for Traffic Management in Urban Air Mobility

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Chair "Integrated Urban Mobility"

Trends in Computational Discrete Optimization
ICERM – 24/04/23-28/04/23

- 1 Introduction
 - Mathematical Optimization for Air Traffic Management
 - Urban Air Mobility
- 2 Problem Definition
- 3 Mathematical Optimization formulation
- 4 Modeling of aircraft separation
- 5 Computational experience
- 6 Conclusions and future perspectives

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- Flights share the same resources, i.e., the airport parking, the runway, the air space, ...
- How to affect the resources so as to guarantee **safety** ?
- **Air space management:**
 - Planning → Strategic Deconfliction
 - Online → Tactical Deconfliction
 - Last minute → Conflict Avoidance

Separation constraint in classic **ATM** :

$$\forall i < j \in A, t \in T \quad \|x_i(t) - x_j(t)\| \geq D$$

where

- A is the set of aircraft,
- T is the considered time horizon,
- $x_i(t)$ the position in the sky of aircraft i at time t ,
- D is the safety distance to guarantee between pairs of aircraft.

Classic Air Traffic Management

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Urban Air Mobility – eVTOLs

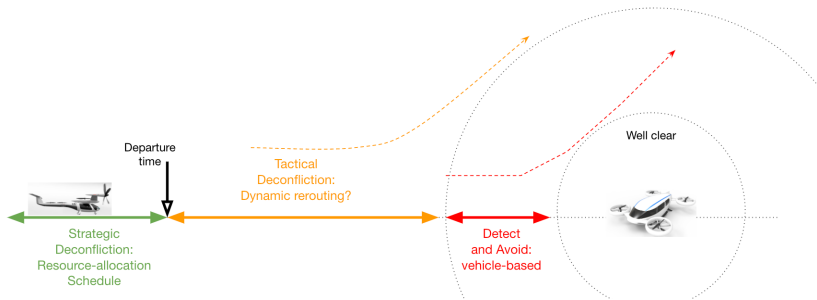




Source <https://www.objetconnecte.com/>

- Urban Air Mobility (**UAM**)
- Electric vertical take-off and landing (**eVTOLs**) aircraft
- Decision-makers tool to guarantee **safety and efficiency**

Urban Air Mobility



Our focus: UAM **tactical deconfliction**

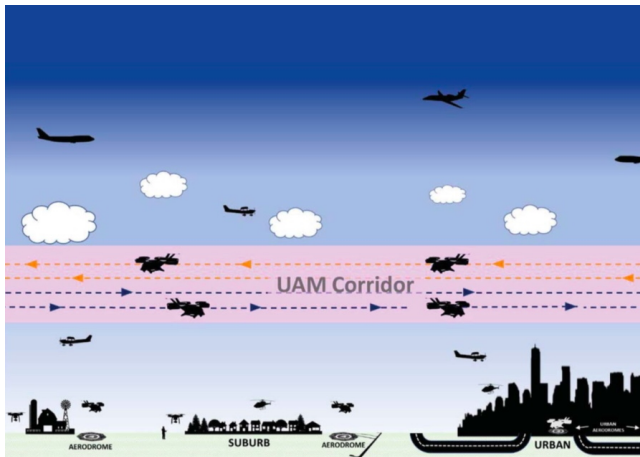
Differences w.r.t. classic ATM?

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Problem Definition

Well defined **skylane network** :

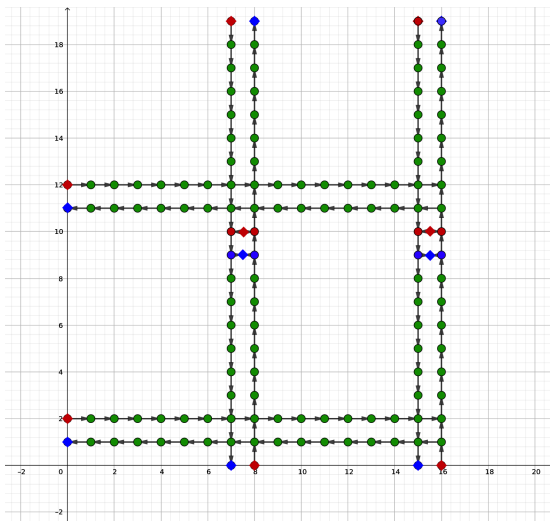


Source: FAA






- connecting **vertiports**
- virtual 3D **corridors**
- corridors can intersect at exclusive **junctions**

Notation

Network infrastructure:

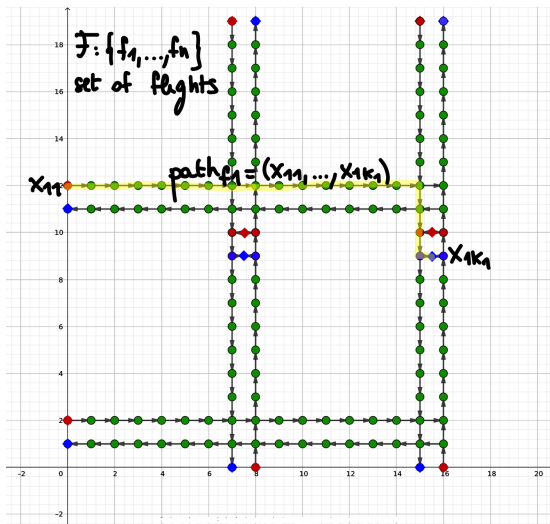


$G = (N, A)$
 N nodes set
 A arcs set (skylanes)

-  } vertiports
-  } vertiport junctions
-  } vertiport junctions
-  } vertiport junctions
-  } inner junctions ($J \subseteq N$)

Notation

Flights and schedule:



$(x_h, x_m) \in A$ belong to $path_{f_i}$
if x_h and x_m are consecutive
nodes in $path_{f_i}$

\hat{t}_{im} time at which f_i arrives
at/traverses x_m

v_{im} speed at which f_i
traverses x_m (constant
through the arc)

Given a nominal planning, consider the following uncertainties.

Disruption

- 1 For strategically deconflicted **scheduled traffic**
- 2 To accommodate a **new, priority, operation**
- 3 In reaction to some **unexpected traffic** e.g., intruder crossing a UAM corridor.

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Mathematical optimization formulation

- **Degrees of freedom**: speed changes manoeuvres at junctions (implicitly) + departure time reschedule
- **Output** : new schedule of the flights with safe arrival times at the waypoints of their paths
- Nominal and the deconflicted schedules are made of **safe time intervals** at which the flights can traverse each waypoint without incurring in conflicts
- **Minimum width** of the time intervals

Decision Variables:

- $t_{im}^{ear}, t_{im}^{lat}$ **new scheduled times**

Safe time interval $[t_{im}^{ear}, t_{im}^{lat}]$ for every $f_i \in \mathcal{F}$ to traverse each $x_m \in path_{f_i}$.

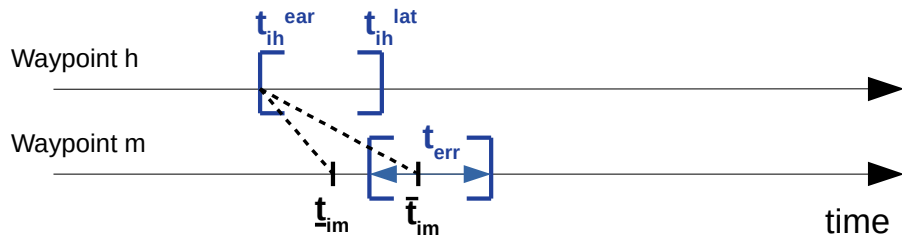
$$t_{im}^{lat} = t_{im}^{ear} + (\hat{t}_{im}^{lat} - \hat{t}_{im}^{ear})$$

- $\underline{t}_{im}, \bar{t}_{im}$ **bounding times**

Bounding interval $[\underline{t}_{im}, \bar{t}_{im}]$ for every $f_i \in \mathcal{F}$ to traverse each $x_m \in path_{f_i}$ **without violating speed and/or operating limits**.

$$\underline{t}_{im} \leq t_{im}^{ear} \leq \bar{t}_{im}$$

Mathematical optimization formulation



For each arc (h, m) , \underline{t}_{im} depends on t_{ih}^{ear} , $\text{dist}(x_h, x_m)$, and \underline{v}_{im}

For each arc (h, m) , \bar{t}_{im} depends on t_{ih}^{ear} , $\text{dist}(x_h, x_m)$, and \bar{v}_{im}

Objective function:

$$\min \sum_{f_i \in \mathcal{F}} \sum_{x_m \in \text{path}_{f_i}} |t_{im}^{\text{ear}} - \hat{t}_{im}^{\text{ear}}| + \sum_{f_j \in \mathcal{F}^{\text{prior}}} M \cdot (t_{i,x_{i1}}^{\text{ear}} - \hat{t}_{i,x_{i1}}^{\text{ear}})$$

M large number (to prioritize minimization of deviation of priority flights)

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Given a conflict $(i, j, m) \in \text{Conf}$, UAM separation constraints:

$$|t_{im} - t_{jm}| \geq \text{“safety time lapse”}.$$

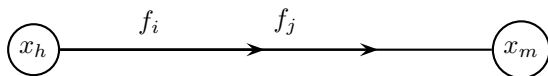
Time lapse depends on a **minimum safety distance** D , see classic ATM.

Two main families of conflicts:

- Trailing
- Intersection

Trailing conflicts

Two flights $f_i, f_j \in \mathcal{F}$ travel through the same arc (x_h, x_m) at the same time



- overtaking is **forbidden**
- hp: **constant speed** on each arc
- sufficient to impose **separation at x_h and x_m**

Trailing conflicts: MP formulation

$$\begin{aligned}v_{ih}(t_{jh} - t_{ih}) &\geq D & \forall (i, j, h, m) \in \text{Trail} \\v_{jm}(t_{jm} - t_{im}) &\geq D & \forall (i, j, h, m) \in \text{Trail}.\end{aligned}$$

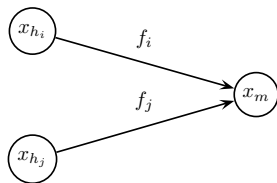
- $\text{Trail} := \{(i, j, h, m) : (i, j, m) \in \text{Conf}, (x_h, x_m) \in \text{path}_{f_i} \cap \text{path}_{f_j}\}$, set of potential trailing conflicts.

Constraints:

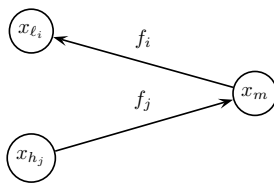
$$\begin{aligned}t_{jh}^{\text{ear}} - t_{ih}^{\text{lat}} &\geq \frac{D}{\underline{v}_{ih}} & \forall (i, j, h, m) \in \text{Trail} \\t_{jm}^{\text{ear}} - t_{im}^{\text{lat}} &\geq \frac{D}{\underline{v}_{jm}} & \forall (i, j, h, m) \in \text{Trail}\end{aligned}$$

Intersection conflicts

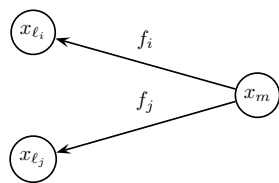
Given a conflict $(i, j, m) \in \text{Conf}$, three types of **intersection conflicts**



In/in



Out/in

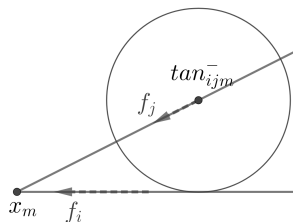


Out/out

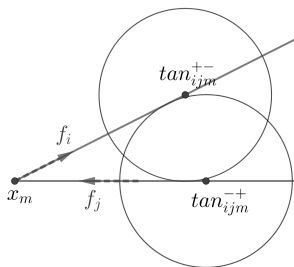
Intersection conflicts

- **Safety disks** of radius D around both f_i and f_j along their trajectories
- **Conflict zones** : when one trajectory is tangent to the safety disk of the other flight
- **Time separation** of passage through conflict zones
- Link **ATM strategies** : ensures a minimum distance of D between the flights all their way throughout their paths.
- **Arbitrating rule** (hp. $t_{im} < t_{jm}$):
“ f_j does not enter its conflict zone until
 f_i does not leave its own”

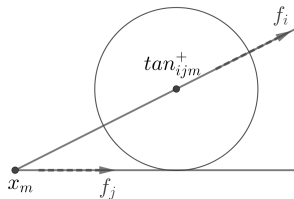
Intersection conflicts



In/in



Out/in



Out/out

- Case in/in: $t_{im} \leq \text{time}(j, \tan_{ijm}^-)$,
where $\text{time}(i, x) =$ the arrival time of f_i at the point x .
- Case out/in: $\text{time}(i, \tan_{ijm}^{+-}) \leq \text{time}(j, \tan_{ijm}^{-+})$
- Case out/out: $\text{time}(i, \tan_{ijm}^+) \leq t_{jm}$.

Classic Air Traffic Management

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Constraints:

$$t_{jm}^{ear} - t_{im}^{lat} \geq S(\alpha_{ijm}^-) \frac{D}{\underline{v}_{jm}} \quad \forall (i, j, m, h_i, h_j) \in \text{Cross}^-$$

$$t_{jm}^{ear} - t_{im}^{lat} \geq S(\alpha_{ijm}^{+-}) \left(\frac{D}{\underline{v}_{jm}} + \frac{D}{\underline{v}_{il_i}} \right) \quad \forall (i, j, m, l_i, h_j) \in \text{Cross}^{+-}$$

$$t_{jm}^{ear} - t_{im}^{lat} \geq S(\alpha_{ijm}^+) \frac{D}{\underline{v}_{il_i}} \quad \forall (i, j, m, l_i, l_j) \in \text{Cross}^+$$

- $\text{Cross}^- / \text{Cross}^{+-} / \text{Cross}^+ := \{(i, j, m, h_i, h_j) : (i, j, m) \in \text{Conf}, (x_{h_i}, x_m) \in \text{path}_{f_i}, (x_{h_j}, x_m) \in \text{path}_{f_j}\}$, set of potential **in/in** / **out/in** / **out/out** intersection conflicts.
- $\alpha_{xyz} := \angle(x, y), (y, z)$ is the angle made by the arcs (x, y) and (y, z) at the junction y , $x, y, z \in X$.
- $S(\alpha) := \frac{1}{\sin \alpha}$ if $\alpha < \pi/2$, 1 otherwise.

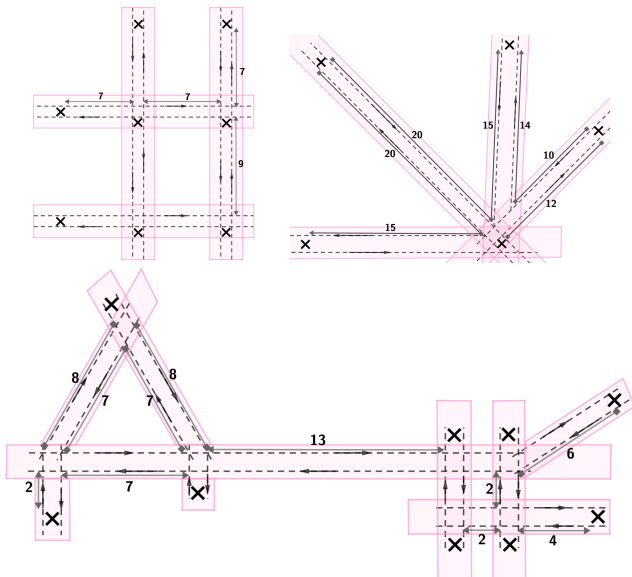
Further modeling details:

- When allow a change in passing order at a node: **binary variables** needed
- Departure times: integer values → **general integer variables**
- Climbing arcs: **additional conditions**
- etc.

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Computational experience



Intel Core i9-9880H CPU, 2.30GHz \times 16, Ubuntu 18.04.4 LTS
CPLEX v. 12.10 (AMPL environment)

- **Scenario 1** (pre-tactical): 20 instances per topology (60 instances)
- **Scenario 2** (priority flight): 20 instances per topology (60 instances)
- **Scenario 3** (intruder): 180 instances ($r \in \{1, 5, 10\}$)

Computational experience: Scenario 1

		Grid	Airport	Metroplex	Overall
cpu (s.)	mean	0.18	0.35	0.29	0.27
	max	0.73	1.65	1.06	1.65
	num. infeasible	0	0	1	1
deviation	total	137.24	225.10	151.82	171.72
	mean (per f_i, x)	0.16	0.31	0.19	0.22
delay at O	% trips	5.35	19.88	7.02	10.81
	total	10.20	13.55	13.26	12.32
	mean (per f_i)	0.14	0.33	0.16	0.21
	max (per f_i)	5.50	2.85	5.68	4.66
delay at D	% trips	5.75	27.37	7.89	13.77
	total	7.56	11.67	10.00	9.74
	mean (per f_i)	0.10	0.28	0.13	0.17
	max (per f_i)	5.26	2.14	5.30	4.22
speed changes	% trips	11.34	33.71	12.56	19.32
	% waypoints	9.97	29.53	11.83	17.20
	max (per f_i)	15.10	18.10	14.32	15.86
% obj. improv.	mean	43.71	51.77	41.81	45.90
	max	75.23	75.23	75.23	75.23

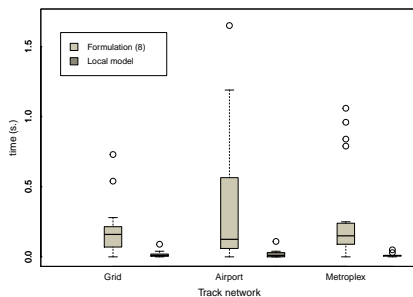
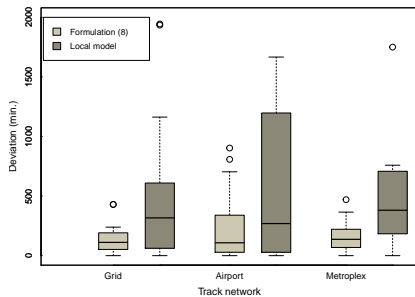
Computational experience: Scenario 2

		Grid	Airport	Metroplex	Overall
cpu (s.)	mean	0.29	0.31	0.36	0.32
	max	0.46	0.47	1.81	1.81
	num. infeasible	0	0	0	0
deviation	total	157.66	763.94	775.54	565.71
	mean (per f_i, x)	0.11	0.93	0.54	0.53
delay prior.	% instances	60	45	60	55
	total	3.95	0.70	2.70	2.45
	max (among instances)	14	5	8	14
delay at O	% trips	9.79	56.40	18.59	28.26
	total	12.55	38.95	66.00	39.17
	mean (per f_i)	0.13	0.89	0.57	0.53
	max (per f_i)	2.00	2.20	1.55	1.92
delay at D	% trips	11.26	56.46	19.60	29.11
	total	7.83	43.24	56.71	35.93
	mean (per f_i)	0.08	0.98	0.49	0.52
	max (per f_i)	1.50	1.92	1.30	1.57
speed changes	% trips	15.13	61.95	20.99	32.69
	% waypoints	12.92	63.77	19.70	32.13
	max (per f_i)	19.30	21.85	8.70	16.62
% obj. improv.	mean	48.15	18.64	7.29	24.70
	max	99.82	99.43	32.73	99.82

Computational experience: Scenario 3

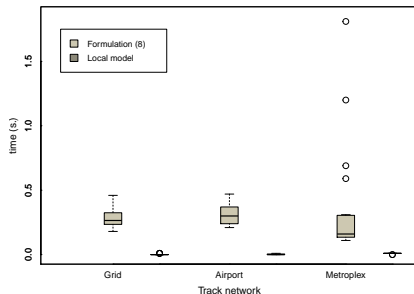
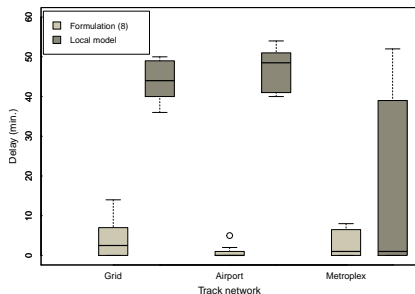
		Grid	Airport	Metroplex	Overall
cpu (s.)	mean	0.60	0.05	0.28	0.17
	max	0.20	0.08	0.37	0.37
	num. infeasible	41	46	41	128
	inf. $r = 1$	16	20	19	55
	inf. $r = 5$	15	16	14	45
	inf. $r = 10$	10	10	8	28
deviation	total	303.83	83.11	1162.59	558.19
	mean (per f_i, x)	0.21	0.14	0.79	0.40
delay at O	% trips	12.75	7.80	44.65	23.08
	total	21.89	4.57	101.68	46.38
	mean (per f_i)	0.22	0.12	0.88	0.44
	max (per f_i)	2.74	1.57	3.63	2.75
delay at D	% trips	17.40	18.75	43.06	27.14
	total	16.57	3.95	77.49	35.43
	mean (per f_i)	0.17	0.10	0.67	0.33
	max (per f_i)	1.94	0.60	3.21	2.04
speed changes	% trips	20.83	40.44	56.63	39.19
	% waypoints	18.09	17.61	47.47	28.69
	max (per f_i)	21.63	20.07	21.58	21.19
% obj. improv.	mean	29.55	49.08	63.31	46.38
	max	44.66	71.69	88.93	88.93
	become feas.	3	2	5	10

Computational experience: global vs. local approach



Comparison of performance against the local model for Scenario 1

Computational experience: global vs. local approach



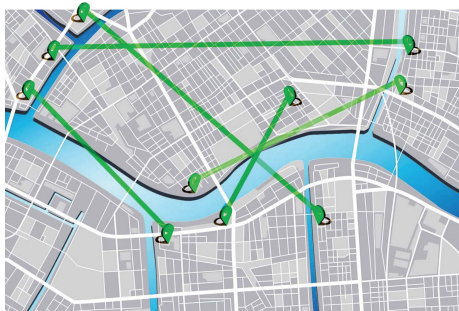
Comparison of performance against the local model for Scenario 2

Scenario 3: only 3 instances solved over 180 for the local approach!

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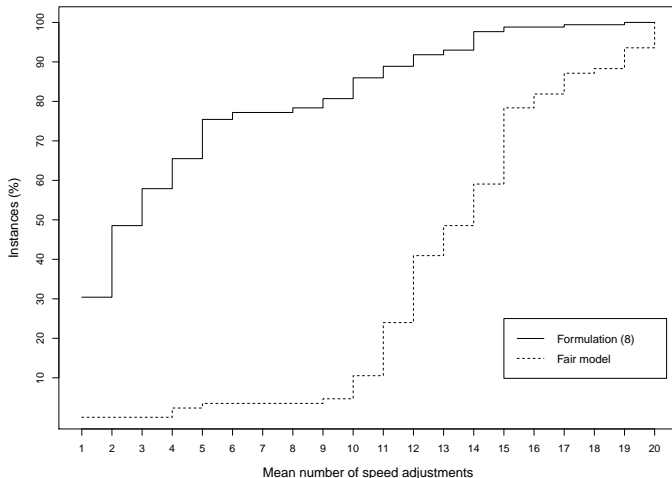
- Definition of a **new problem in UAM**
- Mathematical Optimization **formulation**
- **Promising** computational results
- **Fairness**
- **Ongoing research:**
 - Study robustness approaches at strategic level (with **Tom Portoleau**)

Questions



The cost of fairness

Spread equally the delay among flights



The cost of fairness: Mean number of speed adjustments made per trip

Spread equally the delay among flights

		Scenario 1		Scenario 2		Scenario 3	
		non-fair	fair	non-fair	fair	non-fair	fair
fair o.f. value	mean	39.36	23.99	36.72	19.02	38.86	19.99
	max	272.62	255.40	795.50	411.60	210.71	127.70
	% gain	-	35.06	-	39.87	-	45.29
delay at D	mean	9.74	37.62	35.93	65.39	35.43	77.50
	max	49.80	191.76	627.15	710.20	221.63	295.53
	% degrad.	-	478.76	-	430.39	-	397.72
min delay at D	mean	0.00	0.07	0.00	0.02	0.00	0.04
	max	0.00	1.11	0.00	0.42	0.00	0.37
max delay at D	mean	4.22	3.30	1.57	1.58	2.04	2.04
	max	15.52	15.52	8.00	8.00	6.76	6.76

Instance	Model	MILP			CP		
		10	100	500	10	100	500
	#trips						
Grid	avg time(s)	<0.1	0.19	0.82	<0.1	<0.1	<0.1
Airport	avg time(s)	<0.1	0.35	0.79	<0.1	<0.1	<0.1
Metroplex	avg time(s)	<0.1	0.29	0.86	<0.1	<0.1	<0.1

Table: Computation time comparison between MILP and CP formulation