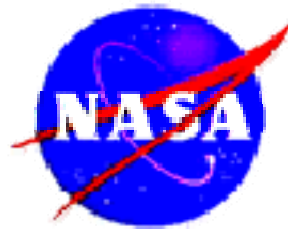


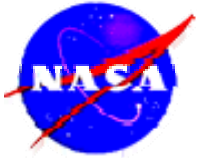
Linear Eulerian Model of En route air traffic flow

Alex Bayen

**Department of Civil and Environmental Engineering
University of California at Berkeley**

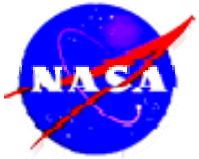


**Modeling, optimization and software in Air Traffic Management Workshop
45th IEEE Conference on Decision and Control, December 12th, 2006, San Diego, CA**



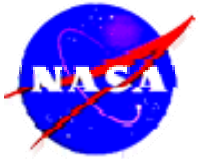
Outline

1. Introduction: Eulerian/Lagrangian; Micro/Macro-aggregate
2. Delay model for TFM: multicommodity cell transmission
 1. Automated graph topology model building
 2. Aggregate travel time estimation
3. LTI models of the NAS
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 3. Practical implementation
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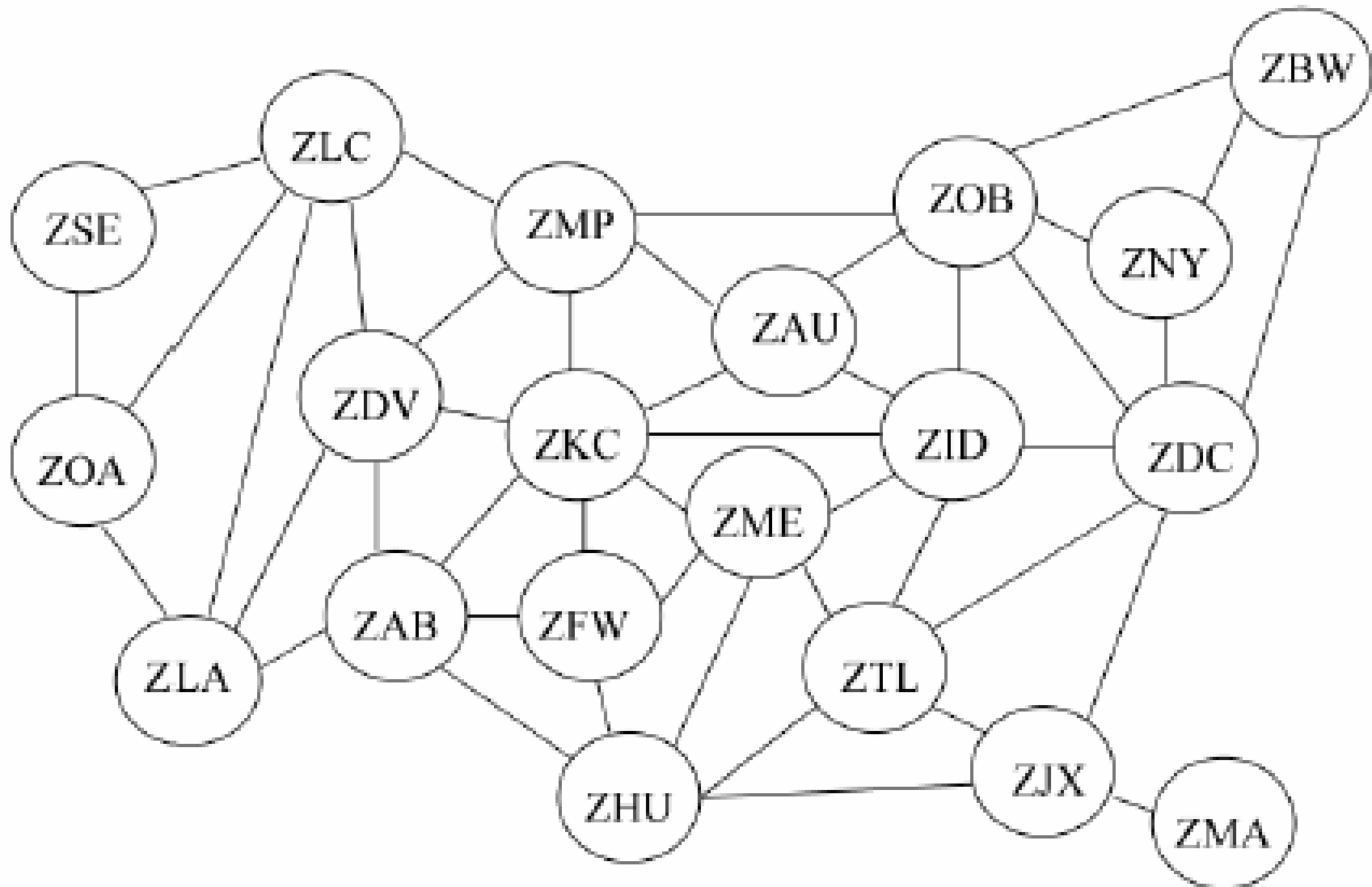


Eulerian / Lagrangian; micro / macro (aggregate)

	Eulerian Control volume based	Lagrangian Trajectory based
Micro Particles		
Macro Aggregation of Particles		

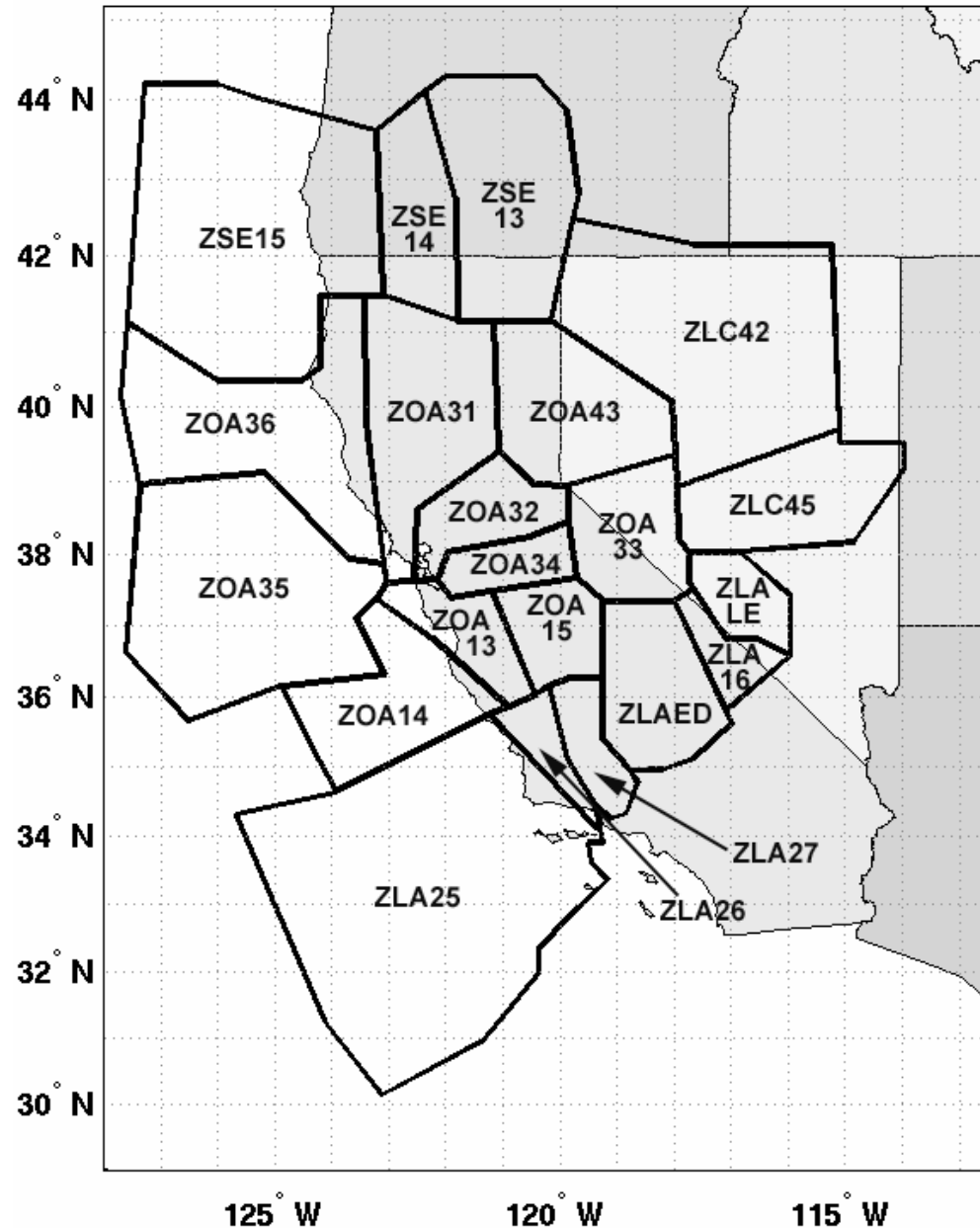


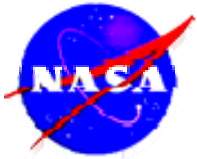
Air Route Traffic Control Centers (ARTCCs)



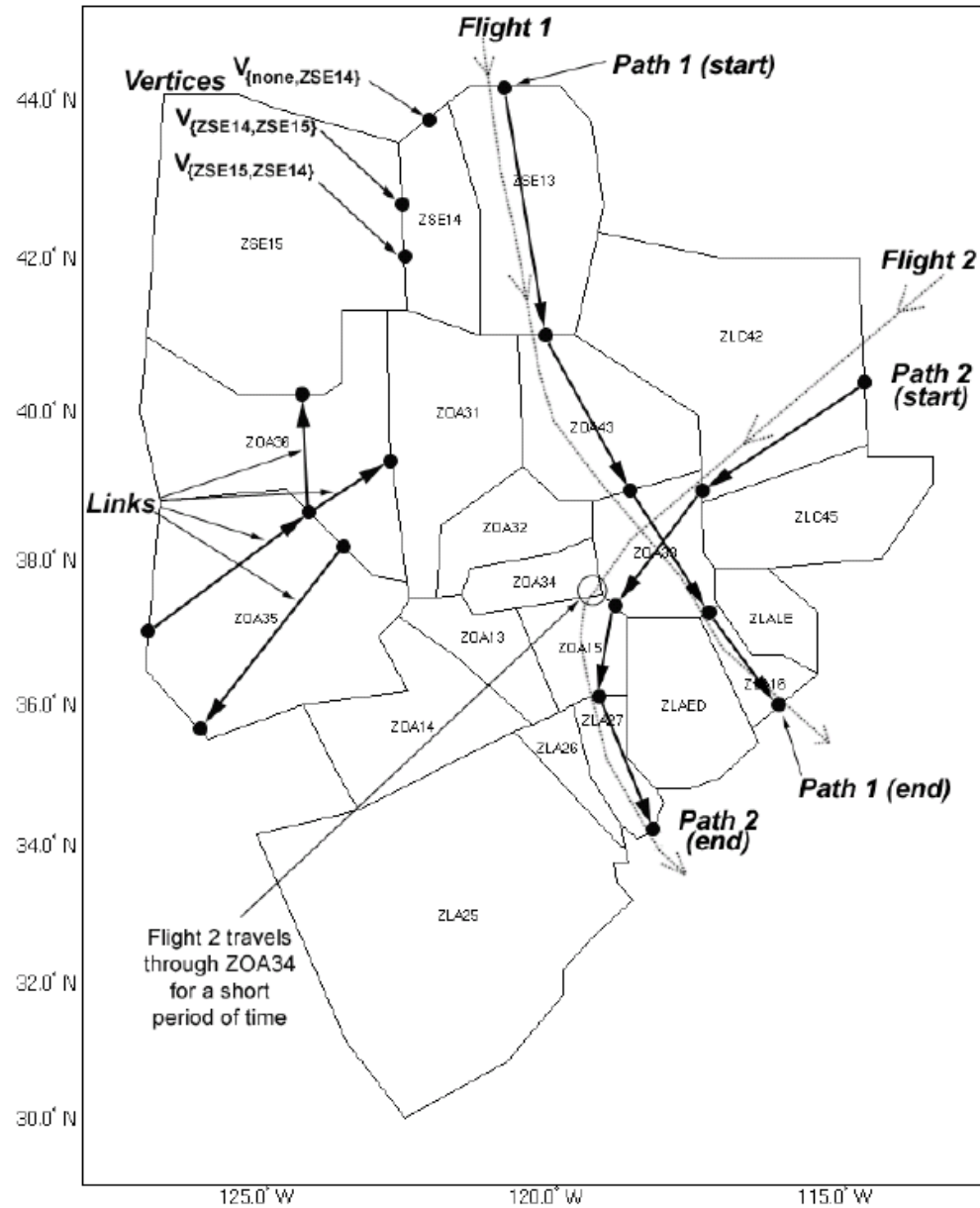


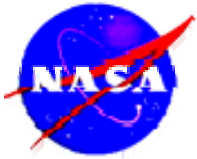
Underlying navigation infrastructure: sectors



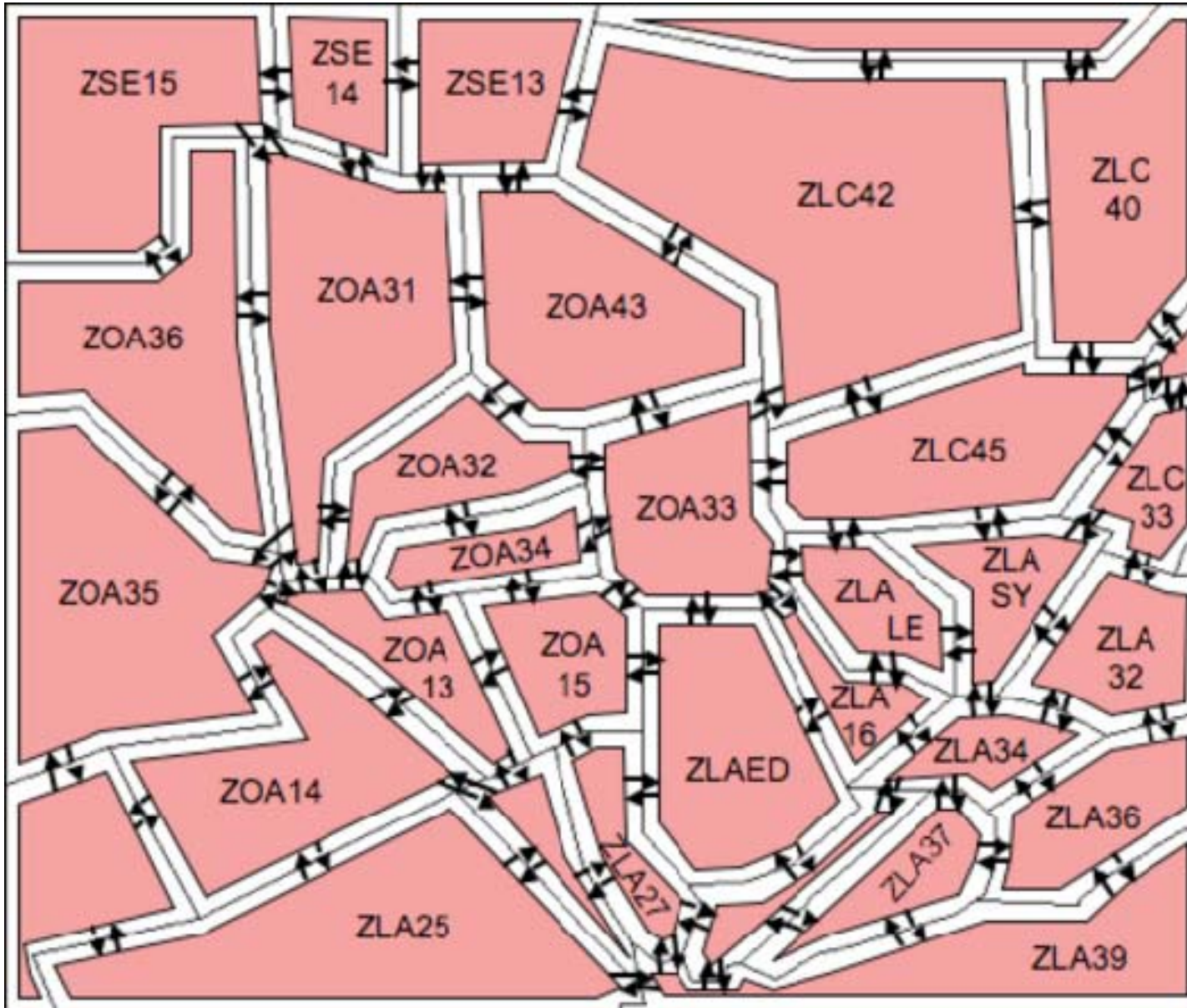


Nature of data used: flows (to be clustered)



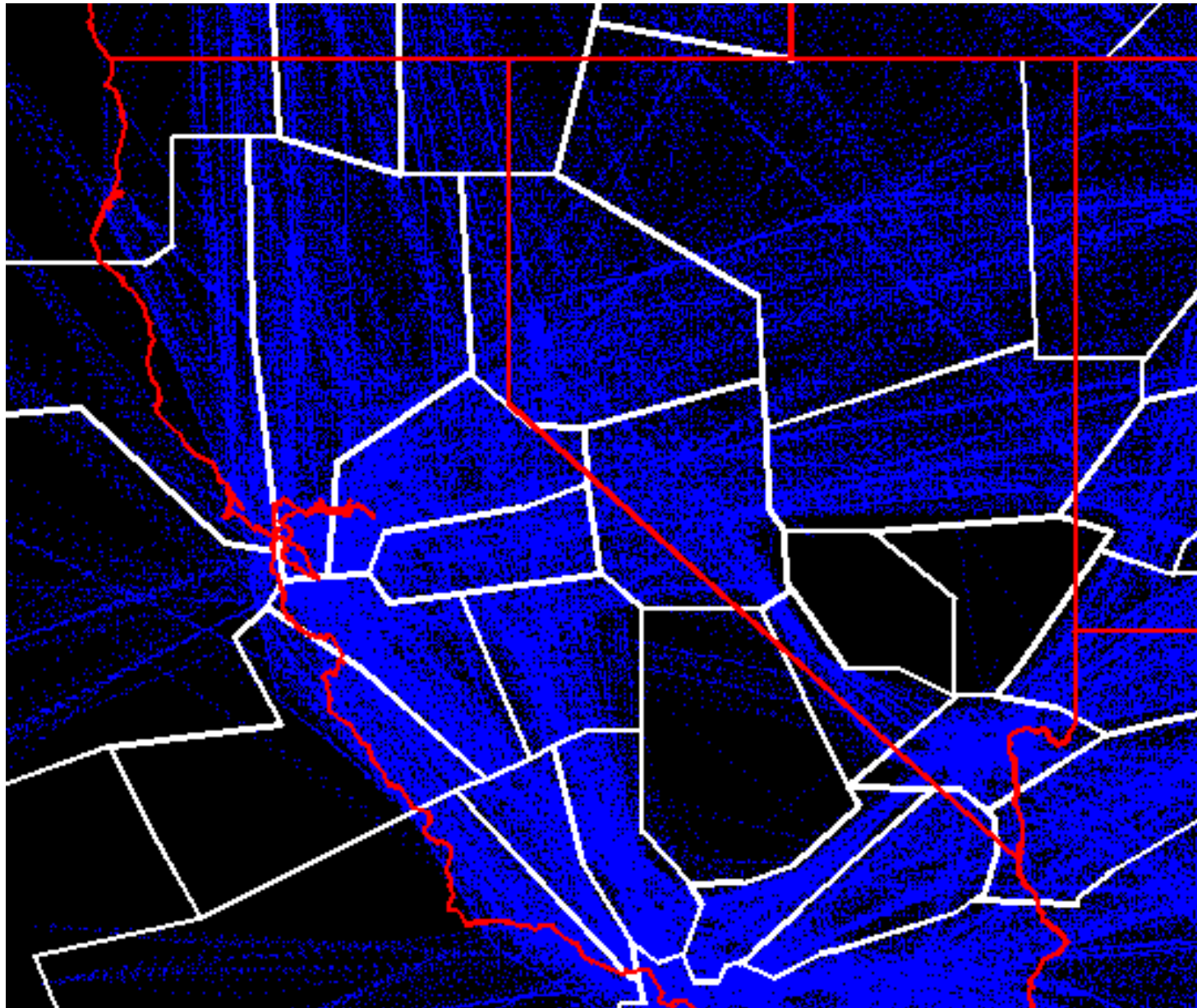


Underlying navigation infrastructure: sectors



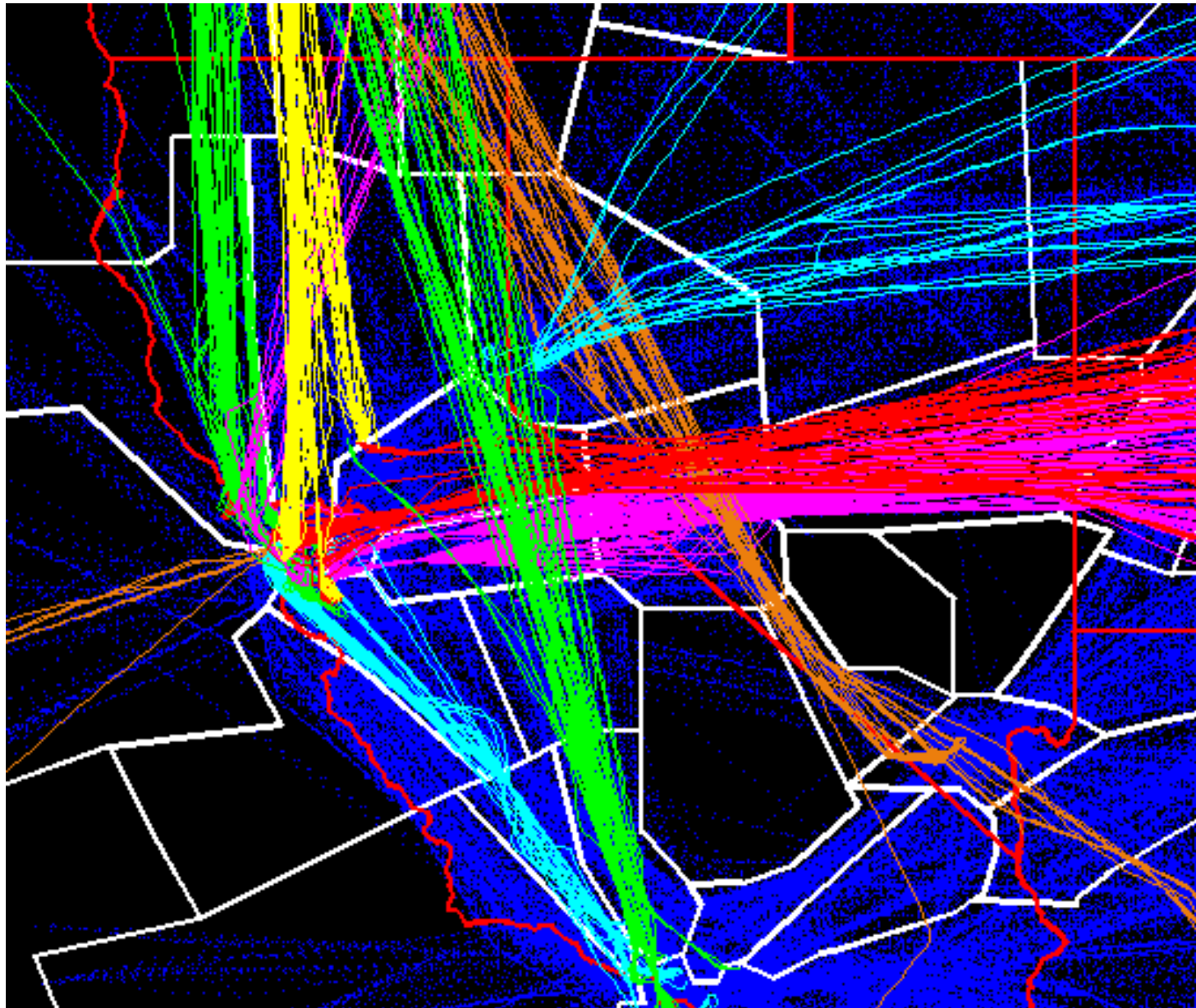


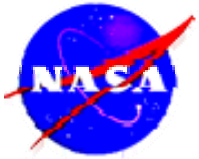
Conceptual framework: flow aggregation



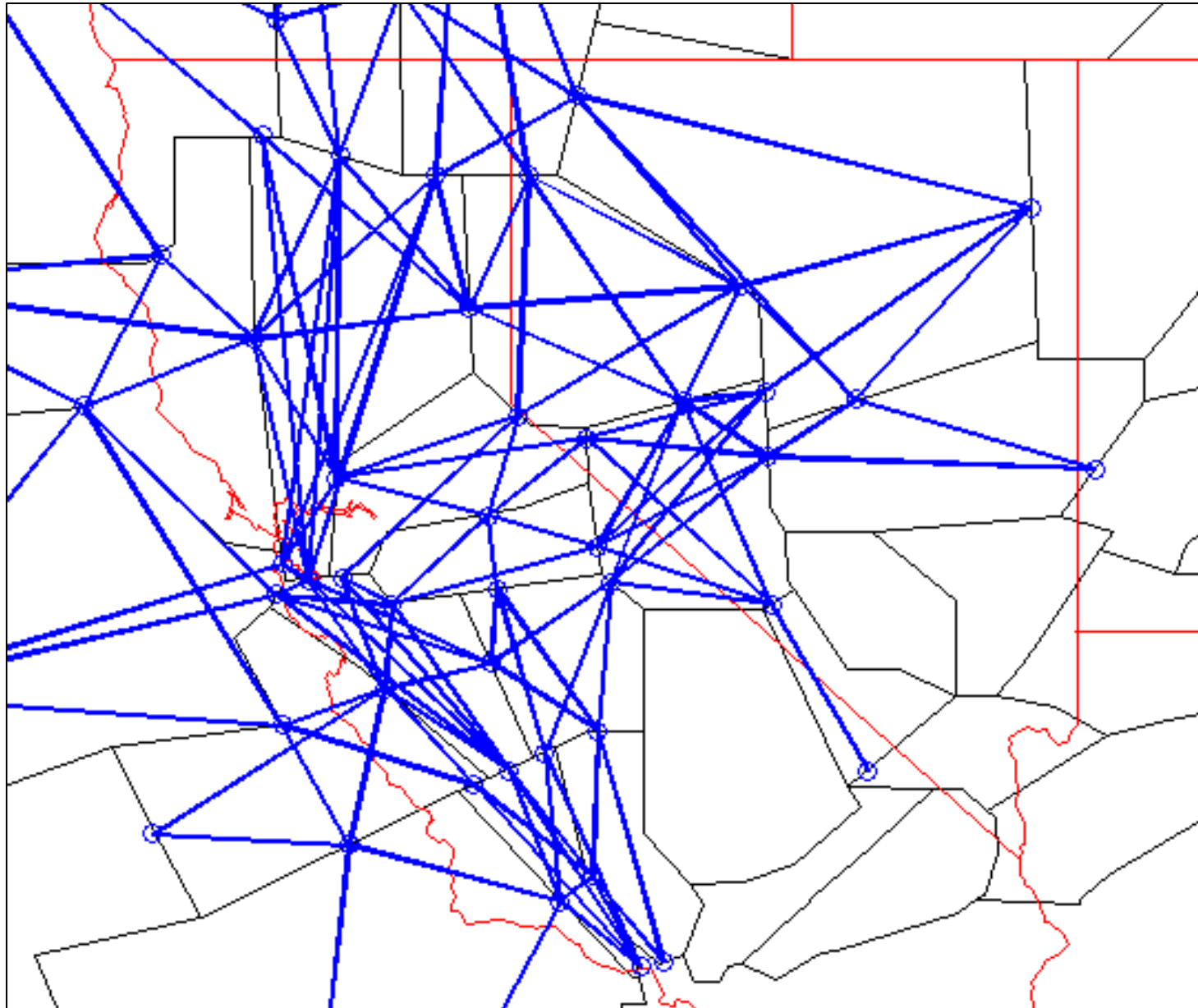


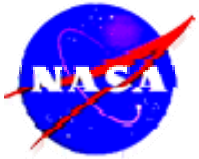
Conceptual framework: flow aggregation



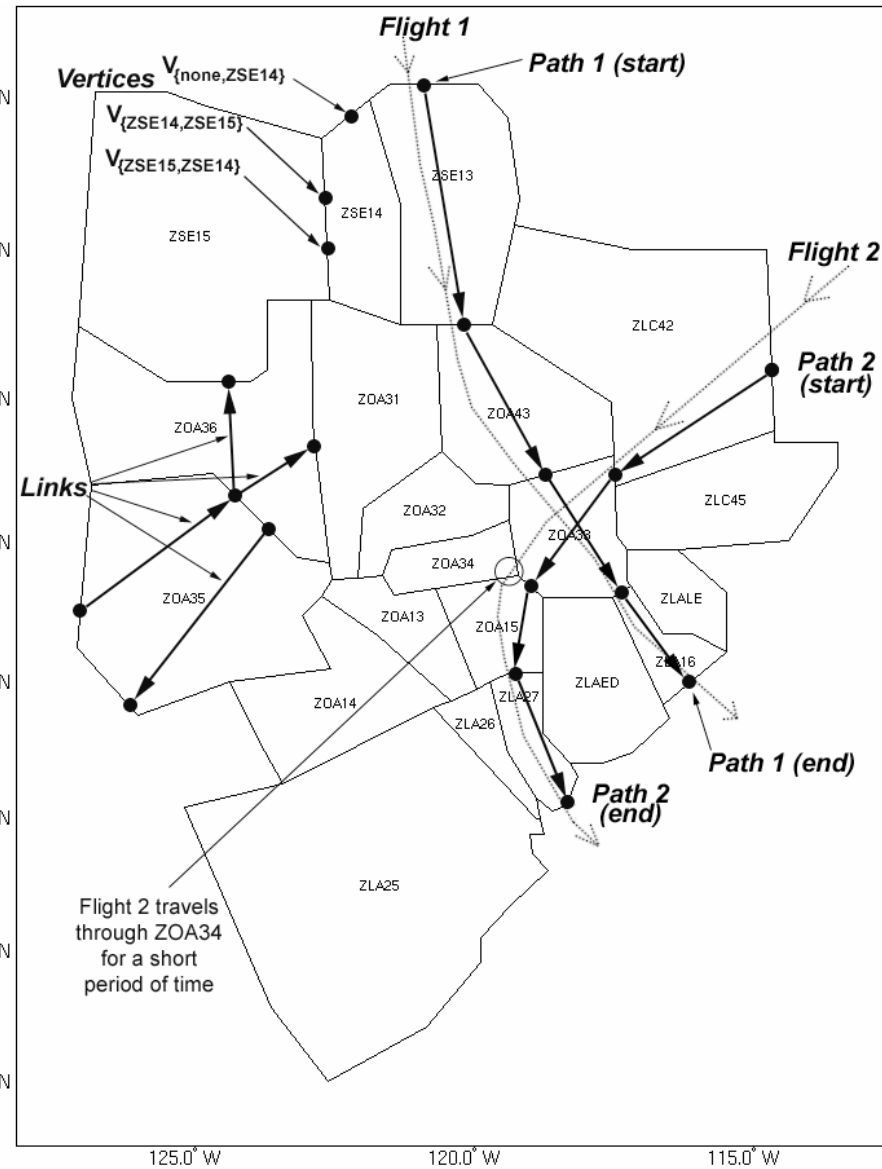
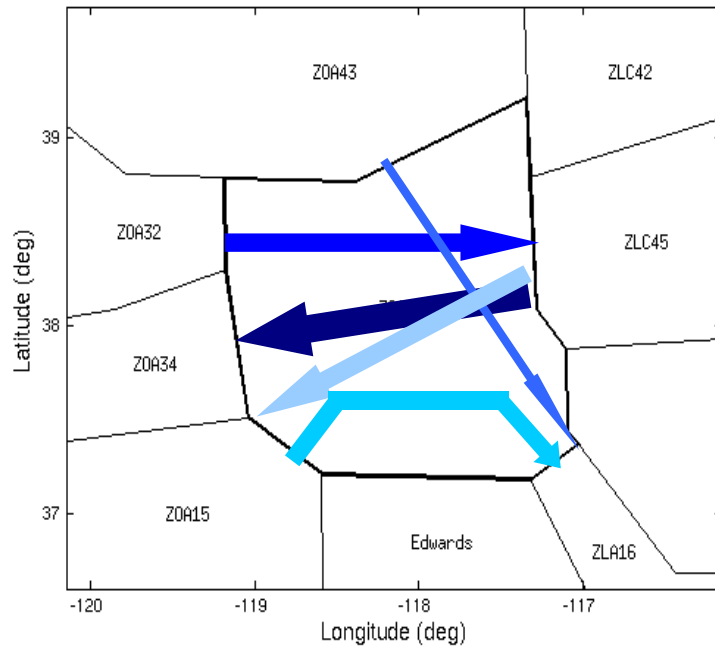


Resulting mathematical object: network





Automated model building



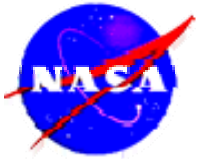
Related work: [Histon, Hansman, 2002]



Sequential (automated) algorithm

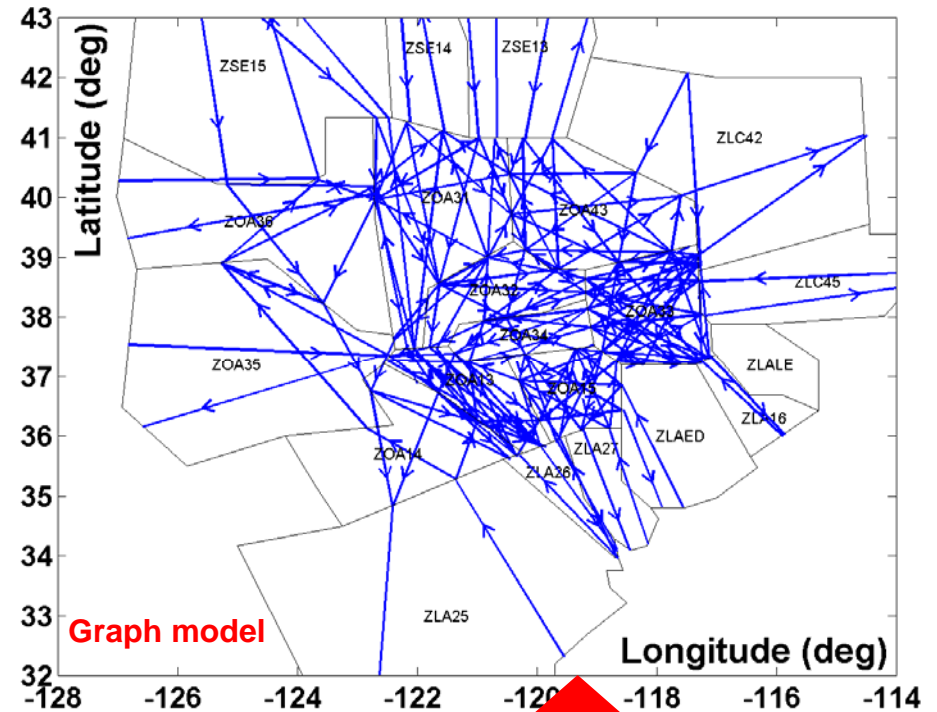
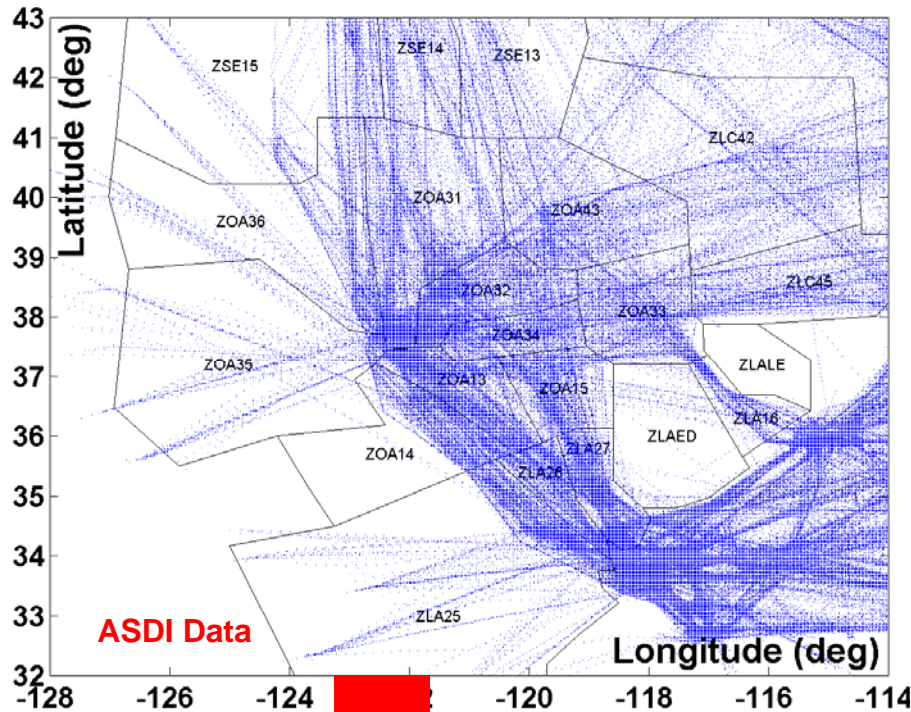
1. Airspace segmentation using sector boundaries
2. Link building using clustering techniques
3. Data aggregation using ASDI/ETMS information (flight plan information)
4. Filtering using LOAs, and observed flow patterns
5. Computation of the aggregate flow pattern features

Output: topology of the flows

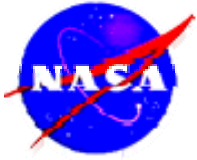


Models: graph model of air traffic flow

- Flight tracks – graph theoretical model



automated identification
procedure: clustering;
pattern recognition



Network flow graph building algorithm

Sequential (automated) algorithm

1. Airspace segmentation using sector boundaries
2. Link building using clustering techniques
3. Data aggregation using ASDI/ETMS information (flight plan information)
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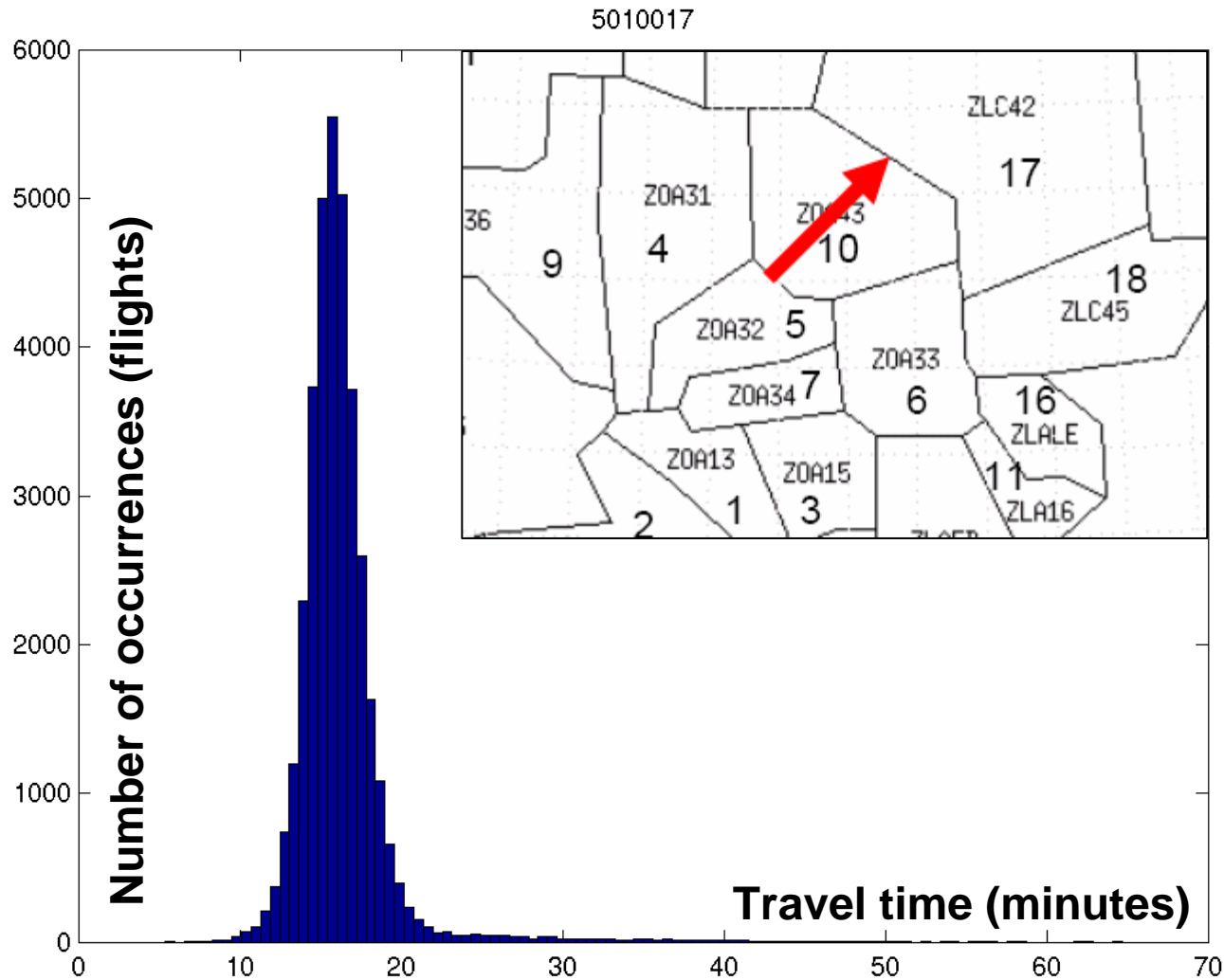


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Example of travel time distribution

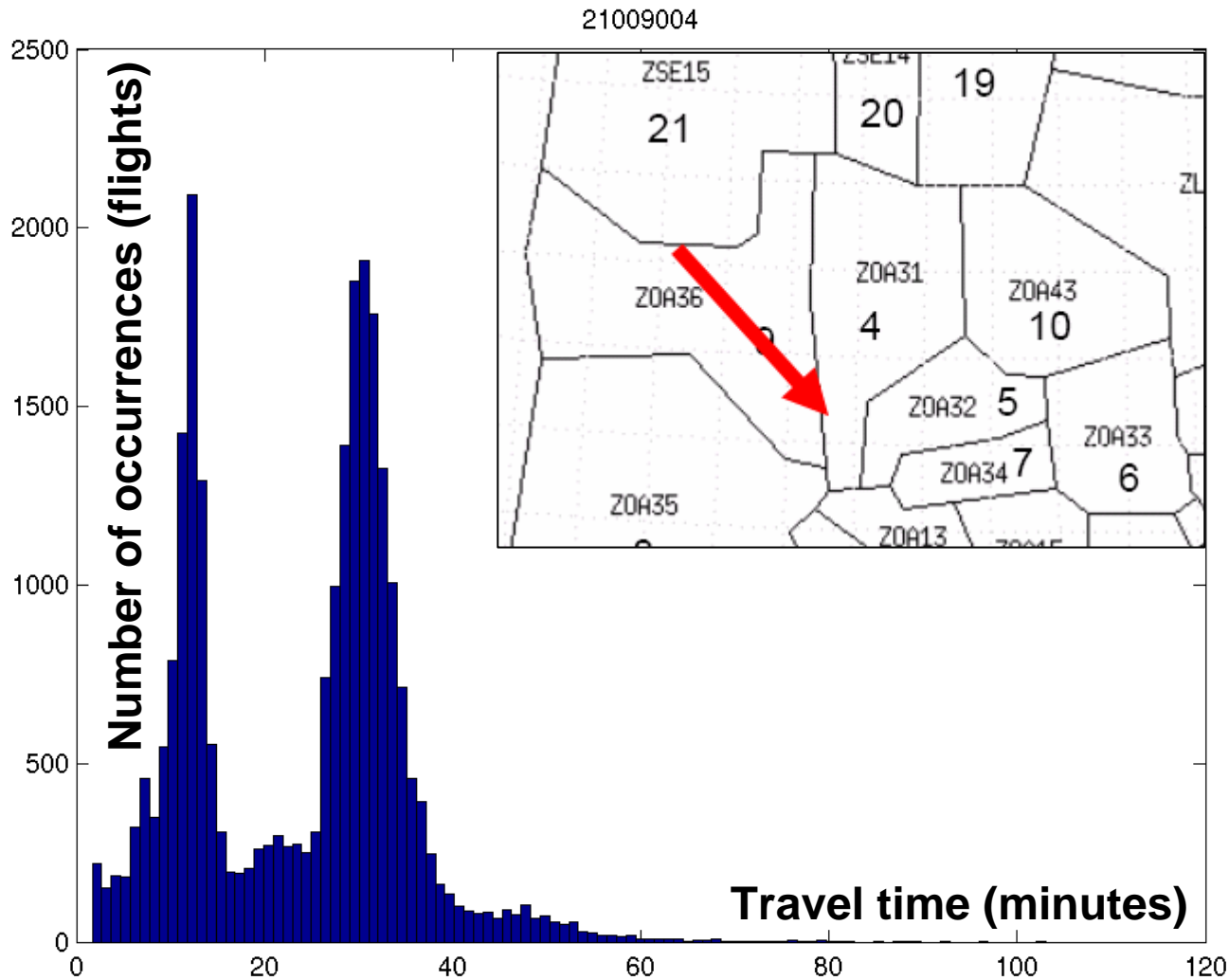


Travel times through the link ZOA32-ZOA43-ZLC42

Departure flights from the bay area.



Example of “bad” segmentation: second peak



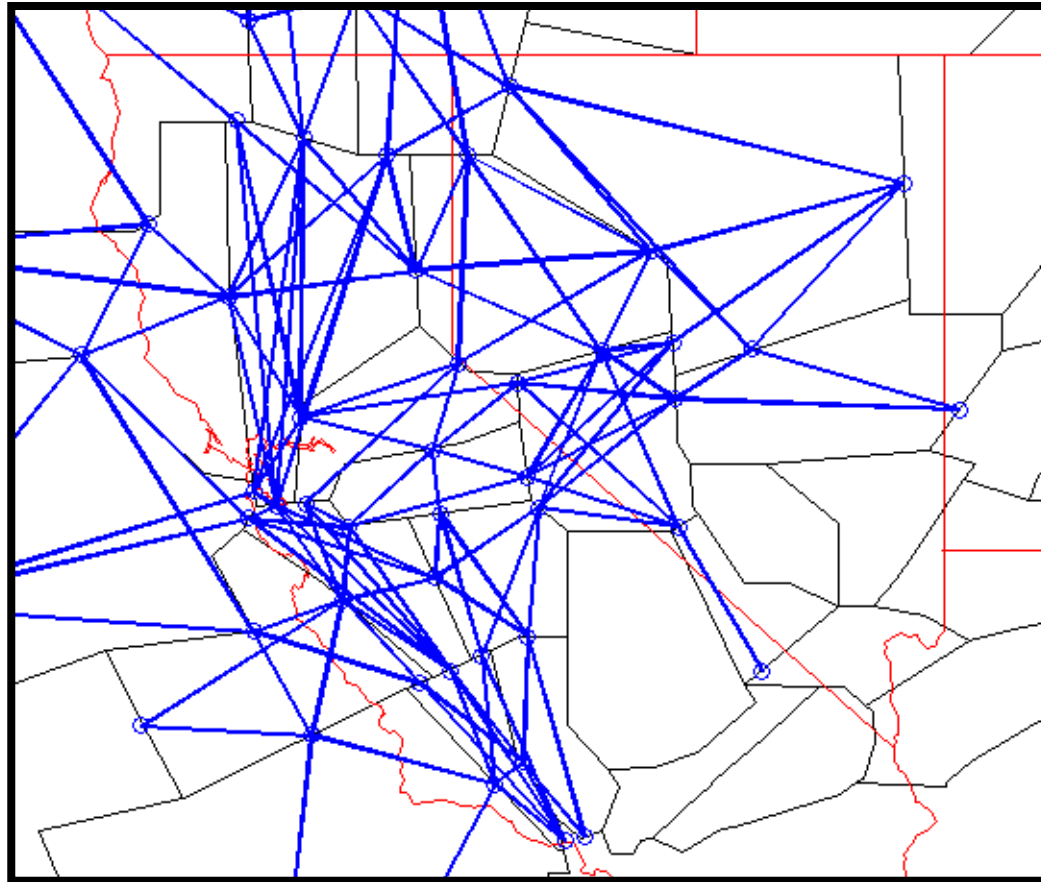
Travel times through the link ZSE15-ZOA26-ZOA31

Possibility: one descending bay area, one passing by



Graph theoretic model

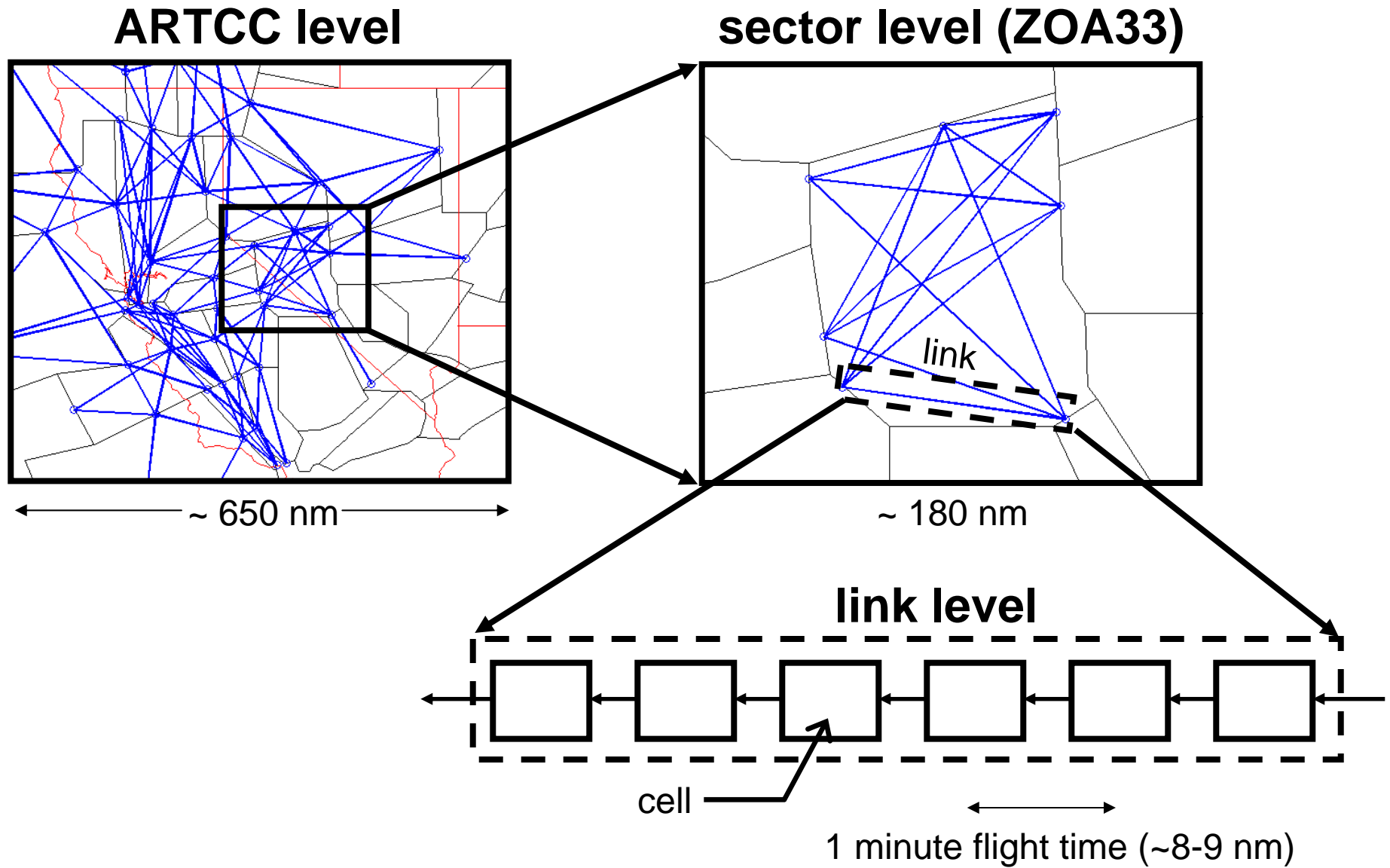
Model of flow patterns for Oakland, and part of Los Angeles, Seattle, and Salt Lake ARTCCs (ZOA, ZLA, ZSE, and ZLC)



Typical directed graph: ~300 links



Models #1: Delay Model

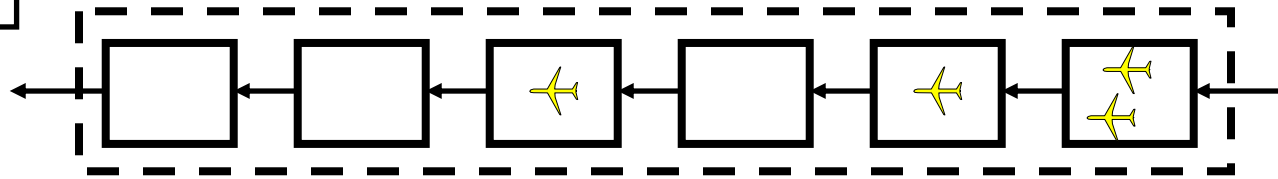




Delay Model: Eulerian dynamics on a link

time step
1

delay system at the link level



state: cell counts

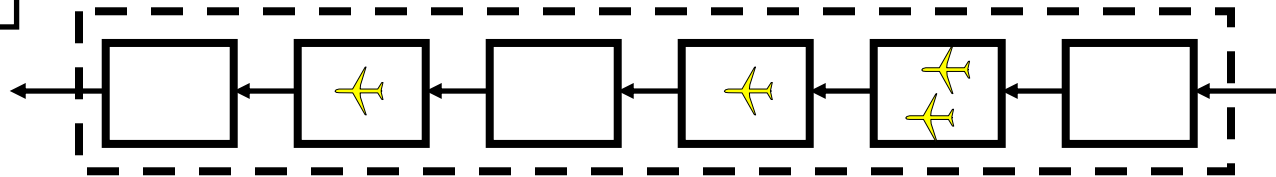
$$x(1) = \begin{bmatrix} x_6(1) \\ x_5(1) \\ x_4(1) \\ x_3(1) \\ x_2(1) \\ x_1(1) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}$$



Delay Model: Eulerian dynamics on a link

time step
2

delay system at the link level

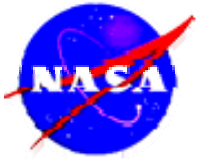


state at current time step

state at previous time step

$$\underbrace{\begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \\ 2 \\ 0 \end{bmatrix}}_{x(2)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ 1 \\ 2 \end{bmatrix}}_{x(1)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(1)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(1)} 0$$

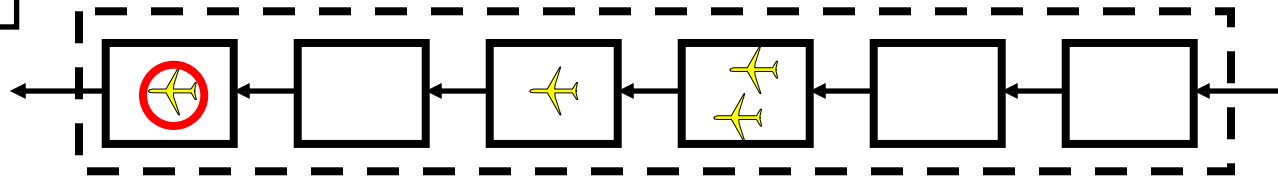
state: cell counts



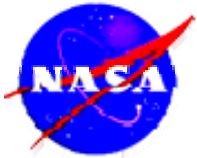
Delay Model: Eulerian dynamics on a link

time step
3

delay system at the link level



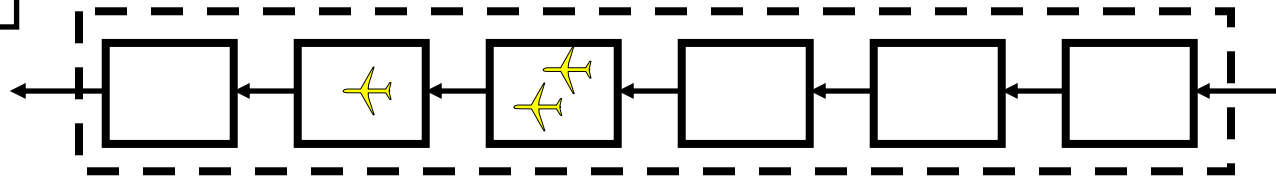
$$\underbrace{\begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{x(3)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \\ 2 \\ 0 \end{bmatrix}}_{x(2)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(2)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(2)} 0$$



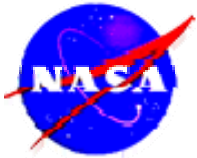
Delay Model: Eulerian dynamics on a link

time step
4

delay system at the link level



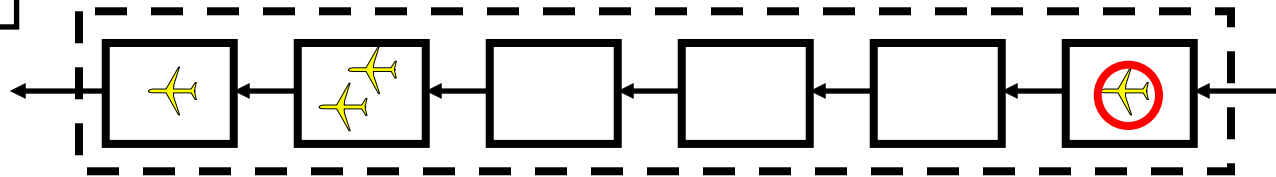
$$\underbrace{\begin{bmatrix} 0 \\ 1 \\ 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{x(4)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 1 \\ 0 \\ 1 \\ 2 \\ 0 \\ 0 \end{bmatrix}}_{x(3)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(3)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(3)} 0$$



Delay Model: Eulerian dynamics on a link

time step
5

delay system at the link level



$$\underbrace{\begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{x(5)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 1 \\ 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{x(4)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(4)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(4)} \textcircled{1}$$

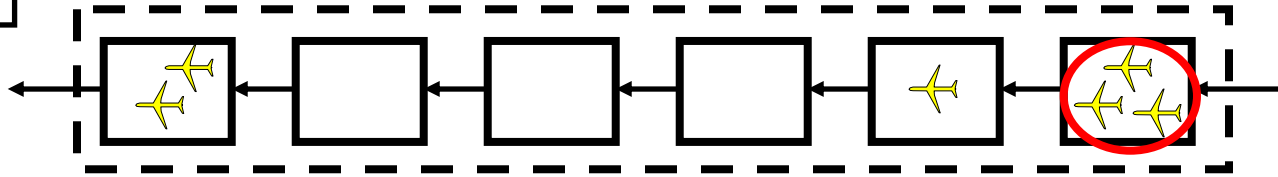
input to the link (forcing) \rightarrow



Delay Model: Eulerian dynamics on a link

time step
6

delay system at the link level



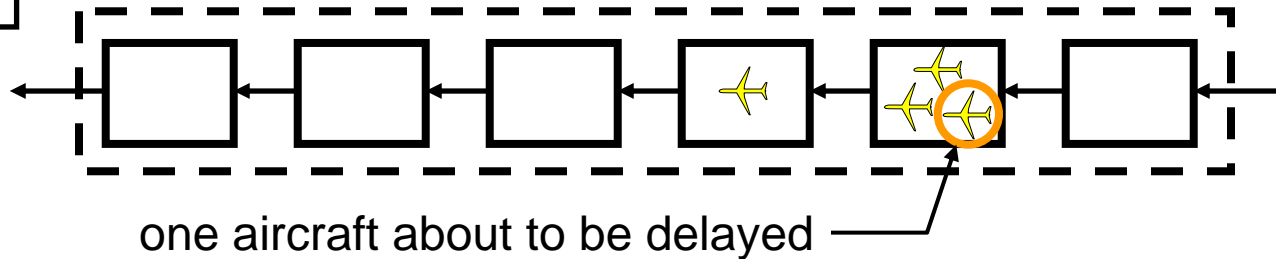
$$\underbrace{\begin{bmatrix} 2 \\ 0 \\ 0 \\ 0 \\ 1 \\ 3 \end{bmatrix}}_{x(6)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{x(5)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(5)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(5)} \quad \text{3}$$



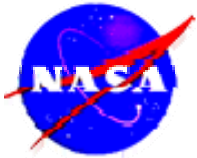
Delay Model: Eulerian dynamics on a link

time step
7

delay system at the link level



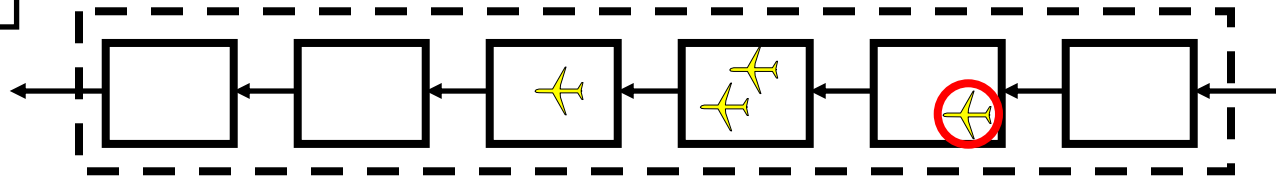
$$\underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 3 \\ 0 \end{bmatrix}}_{x(7)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 2 \\ 0 \\ 0 \\ 0 \\ 1 \\ 3 \end{bmatrix}}_{x(6)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(6)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(6)} 0$$



Delay Model: Eulerian dynamics on a link

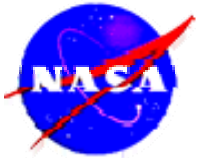
time step
8

delay system at the link level



$$\underbrace{\begin{bmatrix} 0 \\ 0 \\ 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}}_{x(8)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 3 \\ 0 \end{bmatrix}}_{x(7)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}}_{u(7)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(7)} \underbrace{\quad}_0$$

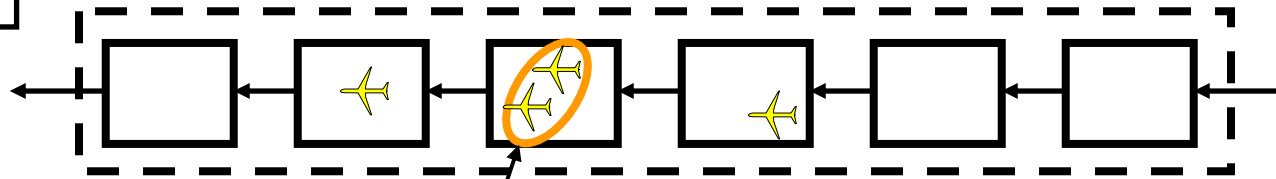
control input



Delay Model: Eulerian dynamics on a link

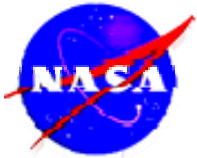
time step
9

delay system at the link level



two aircraft about to be delayed

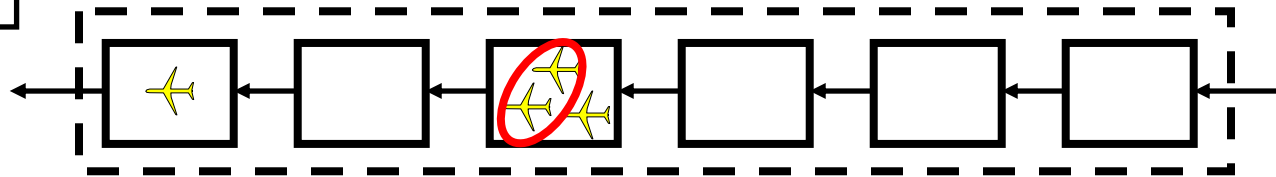
$$\underbrace{\begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \\ 0 \\ 0 \end{bmatrix}}_{x(9)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}}_{x(8)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(8)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(8)} 0$$



Delay Model: Eulerian dynamics on a link

time step
10

delay system at the link level

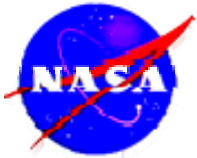


$$\underbrace{\begin{bmatrix} 1 \\ 0 \\ 3 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{x(10)} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 1 \\ 2 \\ 1 \\ 9 \\ 0 \end{bmatrix}}_{x(9)} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 0 \\ 0 \\ 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}}_{u(9)} + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{f(9)} 0$$



Outline

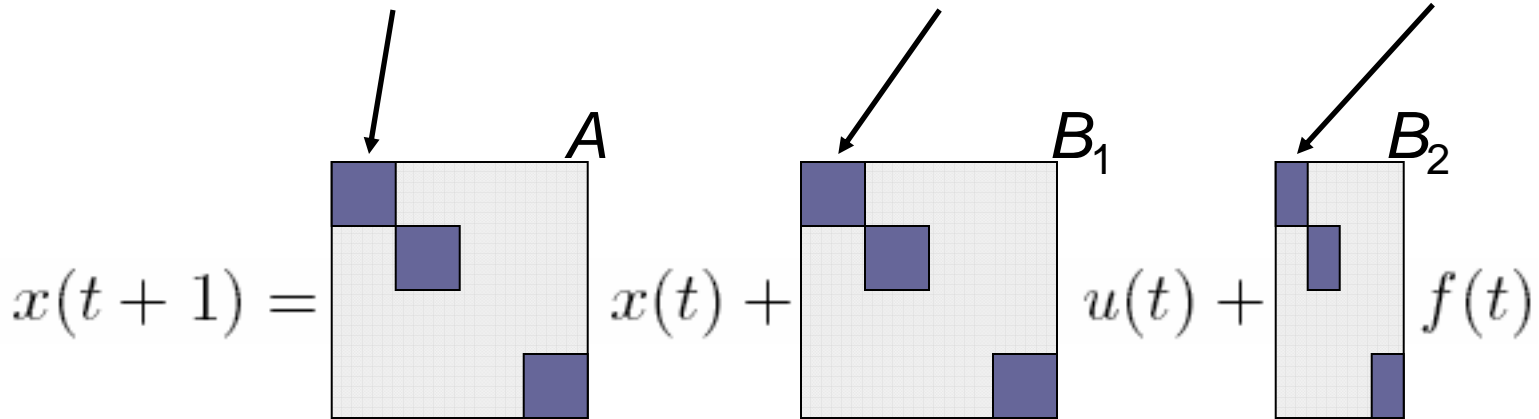
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Models #1: Delay Model

link level

$$\begin{bmatrix} x_6(t+1) \\ x_5(t+1) \\ x_4(t+1) \\ x_3(t+1) \\ x_2(t+1) \\ x_1(t+1) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_6(t) \\ x_5(t) \\ x_4(t) \\ x_3(t) \\ x_2(t) \\ x_1(t) \end{bmatrix} + \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u_6(t) \\ u_5(t) \\ u_4(t) \\ u_3(t) \\ u_2(t) \\ u_1(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} f(t)$$



Air Route Traffic Control Center (ARTCC) level

Sparse LTI dynamical system:
 blocks are nilpotent or upper diagonal matrices



Control theoretic model of air traffic flow

System can be described by a Linear Dynamical System

LTI system is a good first order approximation of the system

For the entire NAS

Difference with the seminal work of Bertsimas/Stock: model (vs. paradigm)

Current size processed for the model (from ASDI/ETMS data)

20 ARTCCs in US (continental)

284 sectors

9572 links

~300,000 actual Paths

110,081 OD pairs

Graph theoretical model

20 ARTCCs in US (continental)

284 sectors

9572 links

110,081 OD pairs



Naive approach of the LTI system

Observability and controllability matrices can be built

$$\mathcal{C}_t = \begin{bmatrix} B & AB & \dots & A^{t-1}B \end{bmatrix}$$

$$\mathcal{O}_t = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{t-1} \end{bmatrix}$$

Matrices are full rank! System is observable and controllable!

Obviously this does not work because

- 1) System is integer valued
- 2) Even if the system is relaxed, the system is constrained:
 - For the control
 - For the state



However, constrained linear optimization works

Standard constrained optimization provides computationally tractable alternatives to classical control techniques, which are unfortunately inapplicable.

Example: global delay mitigation

$$\mathbf{min:} \quad \sum_{k=0}^N c^T x_k$$

subject to:

$$x_0 = B_2 f_0$$

$$x_{k+1} = Ax_k + B_1 u_k + B_2 f_k, \quad k \in \{0, \dots, N-1\}$$

$$Ex_k + Lu_k \leq M, \quad k \in \{0, \dots, N-1\}$$

$$x_N \in \chi_f$$

N : number of time steps, c : vector of 1's

E, L, M : implement user-specified constraints (capacity, nonnegativity, etc)

χ_f : set of feasible final states, x, f, u, A, B_1, B_2 : as defined earlier



MILP formulation of control

$$\text{min:} \quad \sum_{k=0}^N c^T x_k \quad \text{travel time}$$

subject to:

$$x_0 = B_2 f_0$$

$$x_{k+1} = Ax_k + B_1 u_k + B_2 f_k, \quad k \in \{0, \dots, N-1\}$$

$$Ex_k + Lu_k \leq M, \quad k \in \{0, \dots, N-1\}$$

$$x_N \in \chi_f$$

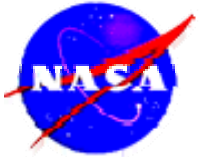
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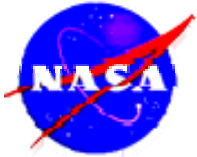
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initial condition



MILP formulation of control

$$\text{min:} \quad \sum_{k=0}^N c^T x_k$$

subject to:

$$x_0 = B_2 f_0$$

$$x_{k+1} = Ax_k + B_1 u_k + B_2 f_k, \quad k \in \{0, \dots, N-1\}$$

$$Ex_k + Lu_k \leq M, \quad k \in \{0, \dots, N-1\}$$

$$x_N \in \chi_f$$

N : number of time steps

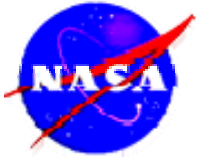
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χ_f : set of feasible final states

x, f, u, A, B_1, B_2 : as defined earlier

dynamics



MILP formulation of control

$$\text{min:} \quad \sum_{k=0}^N c^T x_k$$

subject to:

$$x_0 = B_2 f_0$$

$$x_{k+1} = Ax_k + B_1 u_k + B_2 f_k, \quad k \in \{0, \dots, N-1\}$$

$$Ex_k + Lu_k \leq M, \quad k \in \{0, \dots, N-1\}$$

$$x_N \in \chi_f$$

N : number of time steps

c : vector of 1's

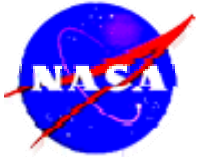
E, L, M : implement user-specified constraints
(capacity, nonnegativity, etc)

χ_f : set of feasible final states

x, f, u, A, B_1, B_2 : as defined earlier

user-specified constraints

- sector capacity
- nonnegativity of x and u
- u less than or equal to x



MILP formulation of control

$$\text{min:} \quad \sum_{k=0}^N c^T x_k$$

subject to:

$$x_0 = B_2 f_0$$

$$x_{k+1} = Ax_k + B_1 u_k + B_2 f_k, \quad k \in \{0, \dots, N-1\}$$

$$Ex_k + Lu_k \leq M, \quad k \in \{0, \dots, N-1\}$$

$$x_N \in \chi_f$$

final state

N : number of time steps

c : vector of 1's

E, L, M : implement user-specified constraints
(capacity, nonnegativity, etc)

χ_f : set of feasible final states

x, f, u, A, B_1, B_2 : as defined earlier



MILP formulation of control

MILP formulation

$$\text{min:} \quad \sum_{k=0}^N c^T x_k$$

subject to:

$$x_0 = B_2 f_0$$

$$x_{k+1} = Ax_k + B_1 u_k + B_2 f_k, \quad k \in \{0, \dots, N-1\}$$

$$Ex_k + Lu_k \leq M, \quad k \in \{0, \dots, N-1\}$$

$$x_N \in \chi_f$$

Challenges:

>1M variables,

>1M constraints

N : number of time steps

c : vector of 1's

E, L, M : implement user-specified constraints
(capacity, nonnegativity, etc)

χ_f : set of feasible final states

x, f, u, A, B_1, B_2 : as defined earlier

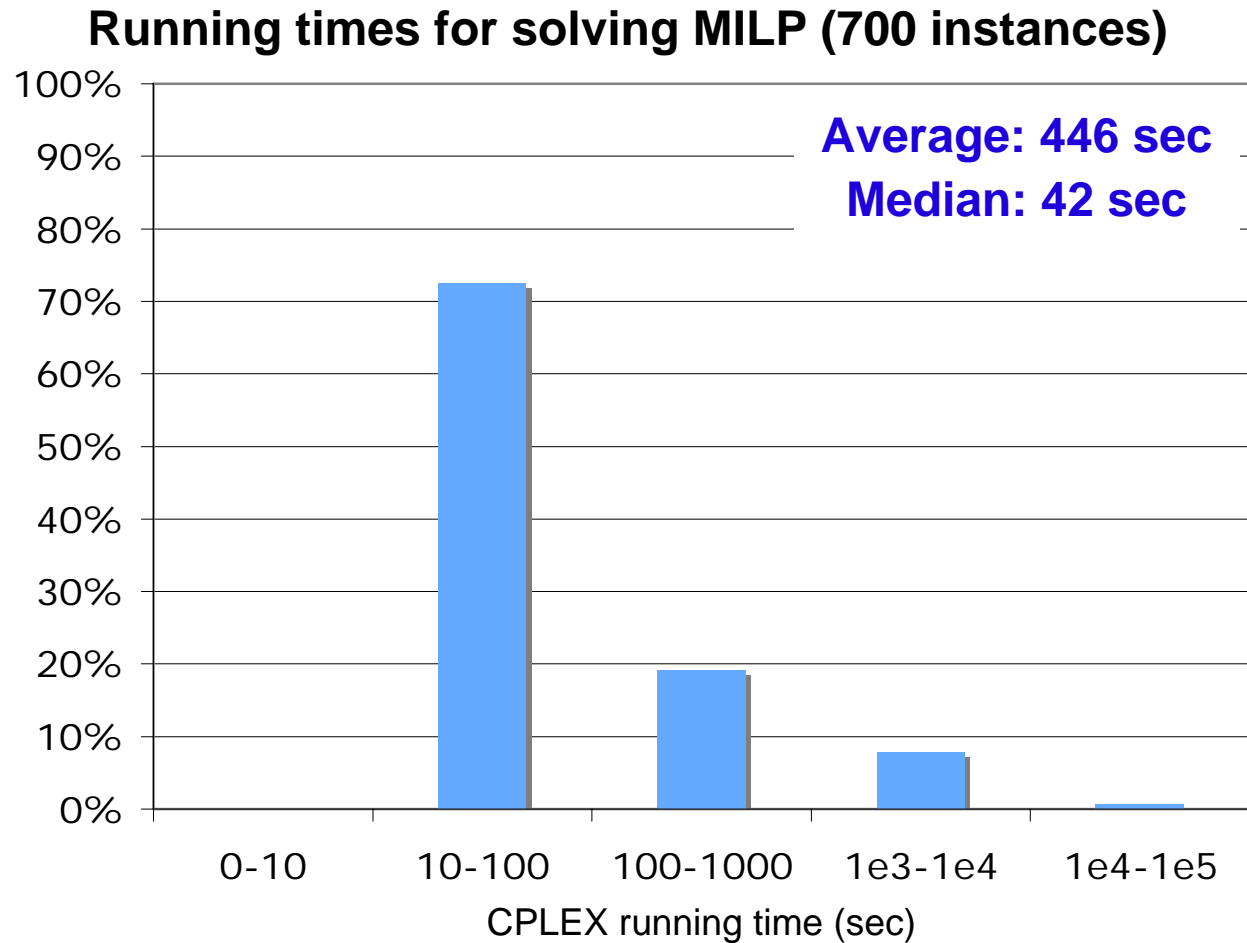


Outline

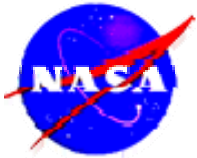
1. Introduction: Eulerian/Lagrangian; Micro/Macro-aggregate
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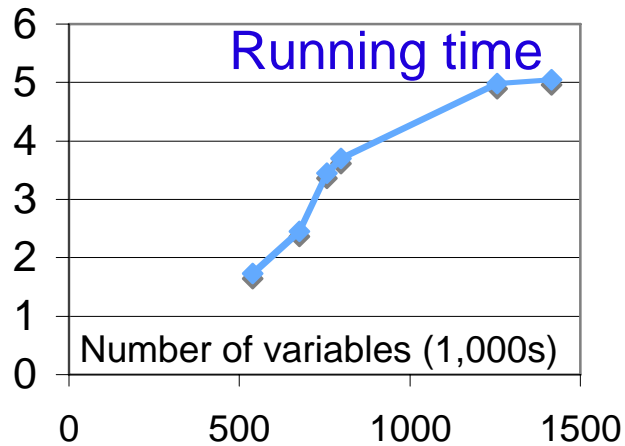
MILP running time



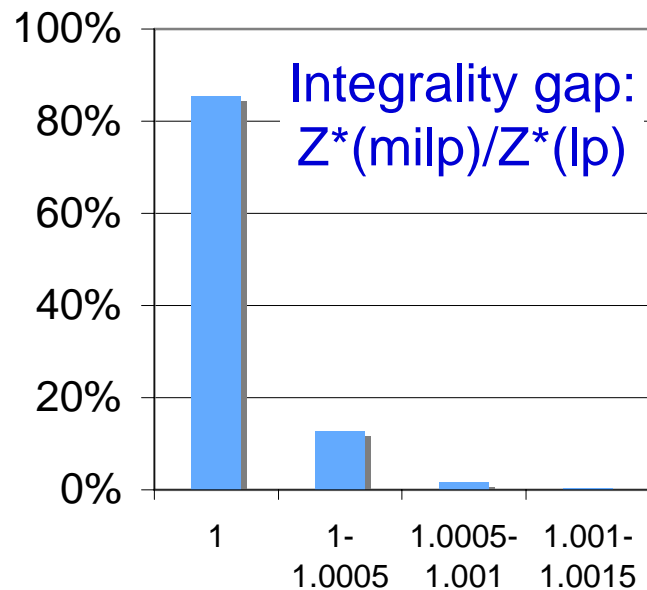
MILP: large variance of CPLEX times → LP relaxation



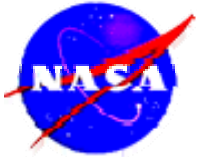
LP relaxation results



MILP \rightarrow LP:
Standard deviation on running time
divided by 4



LP results very close to MILP results
(only a few variables different)



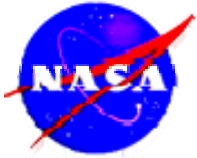
Example of MILP control

MILP control of aggregate Eulerian network airspace models

ATC actuation to control aircraft counts

**Charles-Antoine Robelin, Dengfeng Sun, Guoyuan Wu, and
Alexandre Bayen**





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Going beyond LP, IP, MILP...

Posed as a general IP (relaxed to a LP), the problem becomes computationally challenging for the whole NAS if no care is taken of the structure of the problem (limits of the “blind” LP approach).

Problem is in fact a multicommodity flow problem

- Static version is known to be NP-hard
- Relaxation of the IP is a LP
- LP relaxation can be solved easily with dual decomposition
- $\text{opt}(\text{LP}) = \text{opt}(\text{LP}^*)$



Lagrangian decomposition

$$\min_u \sum_{t=1}^N c^T x_t$$

subject to

$$x_0 = B_2 f_0$$

$$x_{t+1} = A x_t + B_1 u_t + B_2 f_t, t = 0, \dots, N - 1$$

$$\sum_{i \in Q_n} x_t^i \leq C_n, n = 1, \dots, |S|, t = 1, \dots, N.$$

$$\min_u \sum_{t=1}^N \left(\sum_{k=1}^K c^{kT} x_t^k \right)$$

subject to

$$x_0^k = B_2^k f_0^k, k = 1, \dots, K$$

$$x_{t+1}^k = A^k x_t^k + B_1^k u_t^k + B_2^k f_t^k, t = 0, \dots, N - 1, k = 1, \dots, K$$

$$\sum_{k=1}^K \sum_{i \in Q_n} x_t^{i,k} \leq C_n, n = 1, \dots, |S|.$$



Lagrangian decomposition

$$\min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right)$$

subject to

$$x_0^k = B_2^k f_0^k, k = 1, \dots, K$$

$$x_{t+1}^k = A^k x_t^k + B_1^k u_t^k + B_2^k f_t^k, t = 0, \dots, N-1, k = 1, \dots, K$$

$$\sum_{i \in Q_n} x_t^{i,k} = C_n^k, n = 1, \dots, |S|, k = 1, \dots, K.$$

$$\sum_{k=1}^K C_n^k \leq C_n, n = 1, \dots, |S|.$$

$$p^* := \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \max_{\lambda \geq 0} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{n=1}^{|S|} \left[\lambda_n \left(\sum_{k=1}^K C_n^k - C_n \right) \right]$$

subject to

$$x_0^k = B_2^k f_0^k, k = 1, \dots, K$$

$$x_{t+1}^k = A^k x_t^k + B_1^k u_t^k + B_2^k f_t^k, t = 0, \dots, N-1, k = 1, \dots, K$$

$$\sum_{i \in Q_n} x_t^{i,k} = C_n^k, n = 1, \dots, |S|, k = 1, \dots, K.$$



Lagrangian decomposition

$$p^* := \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \max_{\lambda \geq 0} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{n=1}^{|S|} \left[\lambda_n \left(\sum_{k=1}^K C_n^k - C_n \right) \right]$$

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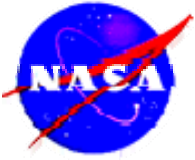
$$d^* := \max_{\lambda \geq 0} \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{n=1}^{|S|} \left[\lambda_n \left(\sum_{k=1}^K C_n^k - C_n \right) \right]$$

subject to

$$x_0^k = B_2^k f_0^k, k = 1, \dots, K$$

$$x_{t+1}^k = A^k x_t^k + B_1^k u_t^k + B_2^k f_t^k, t = 0, \dots, N-1, k = 1, \dots, K$$

$$\sum_{i \in Q_n} x_t^{i,k} = C_n^k, n = 1, \dots, |S|, k = 1, \dots, K.$$



Lagrangian decomposition

$$\begin{aligned}
 d^* &= \max_{\lambda \geq 0} \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{n=1}^{|S|} \left[\lambda_n \left(\sum_{k=1}^K C_n^k - C_n \right) \right] \\
 &= \max_{\lambda \geq 0} \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{n=1}^{|S|} \left(\sum_{k=1}^K \lambda_n C_n^k - \lambda_n C_n \right) \\
 &= \max_{\lambda \geq 0} \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{n=1}^{|S|} \sum_{k=1}^K \lambda_n C_n^k - \sum_{n=1}^{|S|} \lambda_n C_n \\
 &= \max_{\lambda \geq 0} \left\{ - \sum_{n=1}^{|S|} \lambda_n C_n + \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k \right) + \sum_{k=1}^K \sum_{n=1}^{|S|} \lambda_n C_n^k \right\} \\
 &= \max_{\lambda \geq 0} \left\{ - \sum_{n=1}^{|S|} \lambda_n C_n + \min_{u, C_n^k, k=1, \dots, K, n=1, \dots, |S|} \sum_{k=1}^K \left(\sum_{t=1}^N c^{kT} x_t^k + \sum_{n=1}^{|S|} \lambda_n C_n^k \right) \right\} \\
 &= \max_{\lambda \geq 0} \left\{ - \sum_{n=1}^{|S|} \lambda_n C_n + \sum_{k=1}^K \left(\min_{u, C_n^k, n=1, \dots, |S|} \sum_{t=1}^N c^{kT} x_t^k + \sum_{n=1}^{|S|} \lambda_n C_n^k \right) \right\} \\
 &= \max_{\lambda \geq 0} \left\{ - \sum_{n=1}^{|S|} \lambda_n C_n + \sum_{k=1}^K d^{k*}(\lambda) \right\}
 \end{aligned}$$

where

$$d^{k*}(\lambda) = \min_{u, C_n^k, n=1, \dots, |S|} \sum_{t=1}^N c^{kT} x_t^k + \sum_{n=1}^{|S|} \lambda_n C_n^k$$



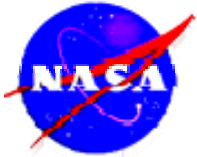
Lagrangian decomposition

Sub-problems (very tractable)

$$d^{k*}(\lambda) = \min_{u, C_n^k, n=1, \dots, |S|} \sum_{t=1}^N c^{kT} x_t^k + \sum_{n=1}^{|S|} \lambda_n C_n^k$$

subject to

$$x_0^k = B_2^k f_0^k$$
$$x_{t+1}^k = A^k x_t^k + B_1^k u_t^k + B_2^k f_t^k, t = 0, \dots, N-1$$
$$\sum_{i \in Q_n} x_t^{i,k} = C_n^k, n = 1, \dots, |S|.$$



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Future directions

1. Optimization

1. Dual decomposition
2. Multi commodity flow (dynamic): combinatorial optimization algorithms

2. Scenarios of interest

1. En route delay mitigation (MILP \rightarrow multicommodity flow)
2. Delay multiplier minimization (NEXTOR collaboration)
3. Dynamic sectorization of the en route airspace

3. Computational issues

1. Comparison of computational performance of various algorithms
2. Convergence acceleration
3. Link with operational tools (FACET in particular)