



CENTER FOR ADVANCED AVIATION SYSTEM DEVELOPMENT (CAASD)

# Predicting Sector Capacity for TFM

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*3 July 2007*

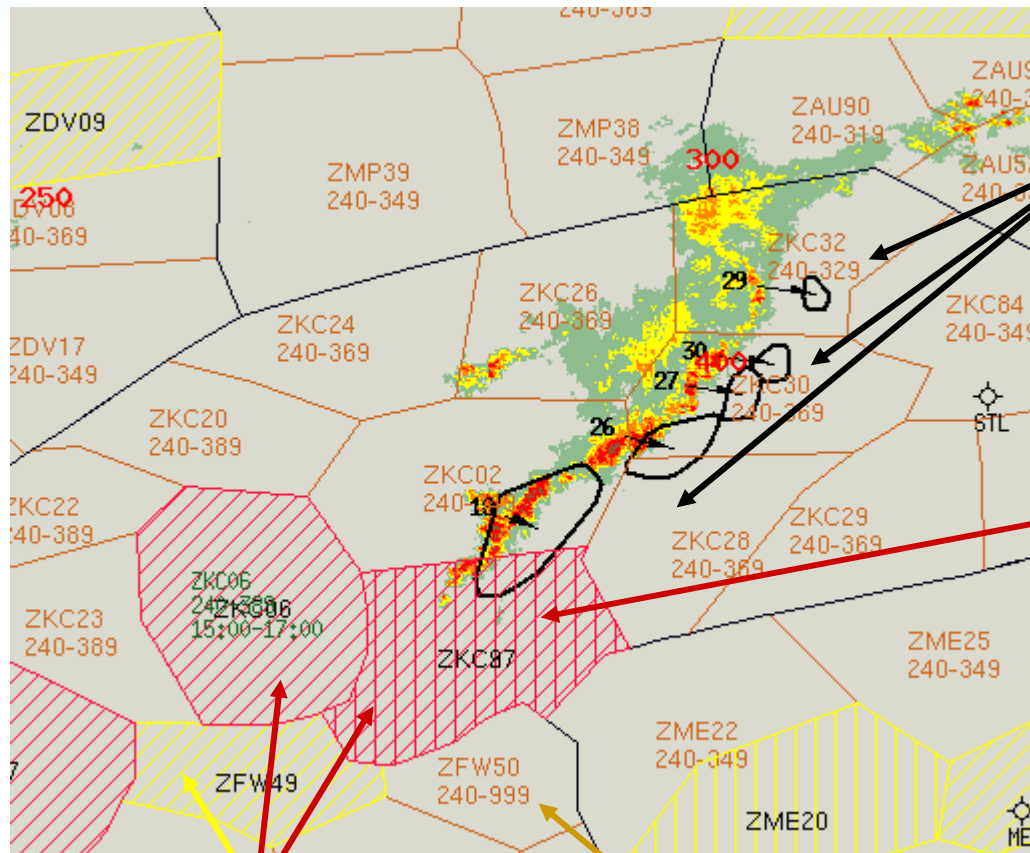


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# En Route Congestion



Uncertain weather forecasts indicate current and future loss of airspace capacity...

Uncertain traffic forecasts provide airspace demand...

If demand exceeds capacity, delays will occur and safety may be compromised.

Role of TFM:  
Balance demand vs. capacity.  
Key function:  
Predicting capacity/congestion.  
Required:  
A good metric of capacity/congestion.

Congestion Alerts

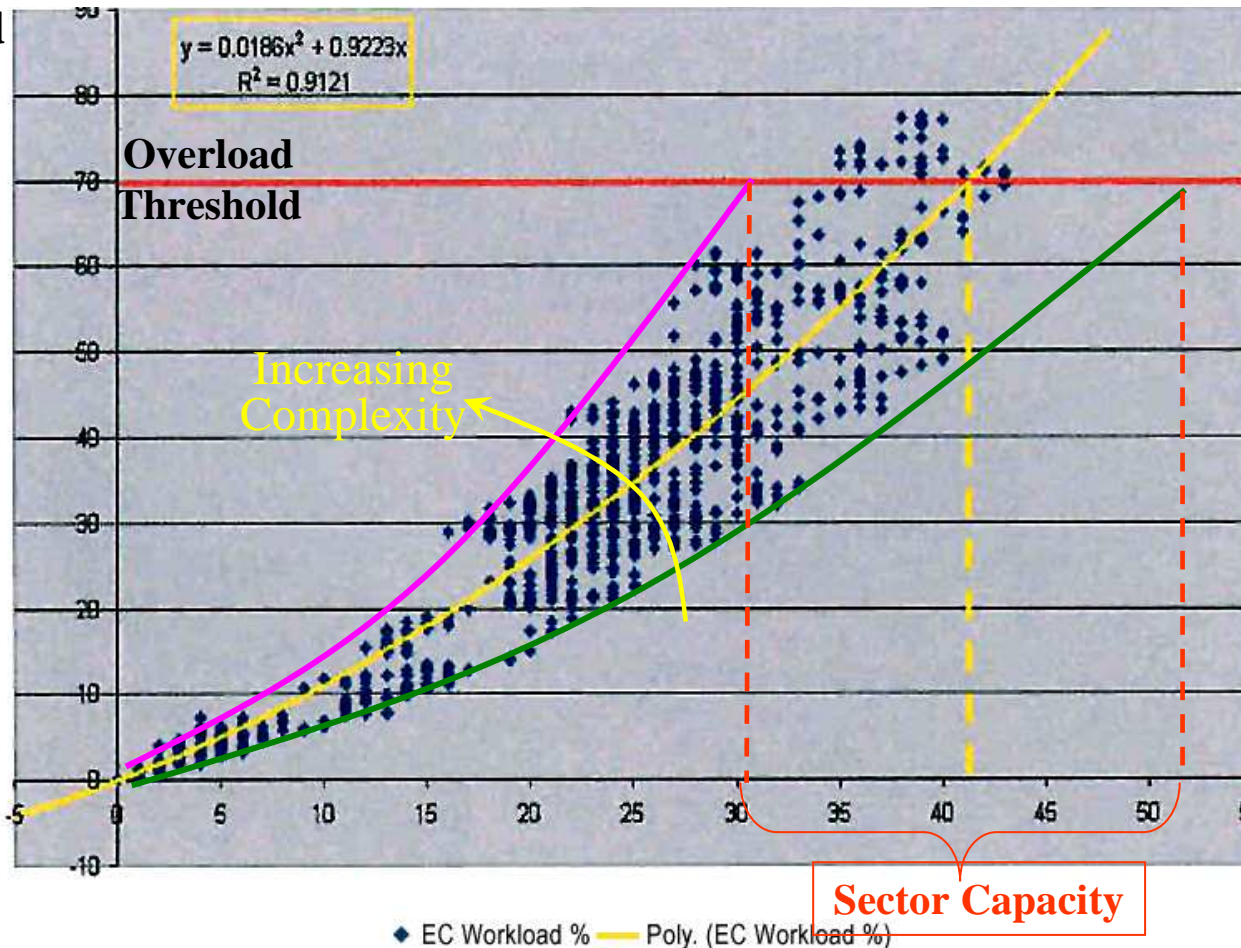
Air traffic control sector



# What is Sector Capacity?

Cited from EUROCONTROL Experimental Center Note No. 21/03  
EDYCOLO Wednesday 4 July 2001

Workload



Aircraft  
Count (N)

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## A Good Sector Capacity Metric for En Route TFM Decision Support...

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- **Better represents sector workload, and workload threshold, than peak aircraft count with a static threshold (e.g., MAP)**
- **Is intuitive and relevant to human decision-makers**
- **Provides insight into congestion resolution options**
- **Is predictable at useful look-ahead times (30 min – 2 hr)**
- **Captures impact of convective weather**



# Dynamic Density Study

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- **Metric variables included a wide range of different traffic measurement types:**
  - Aircraft count and density
  - Sector structure-based (e.g., handoff workload, sector shape)
  - Aircraft-aircraft proximity & conflicts
  - Aircraft state variables (e.g., speed variations, altitude transitions)
- **DD metrics were more effective than aircraft count for measuring and predicting controller-perceived workload.**
- **Many metric variables are essentially unpredictable beyond 30 minutes (example follows)**
  - Guidelines for which variables to include in predictive applications
  - Insight into what traffic features are predictable



## Example DD Variables

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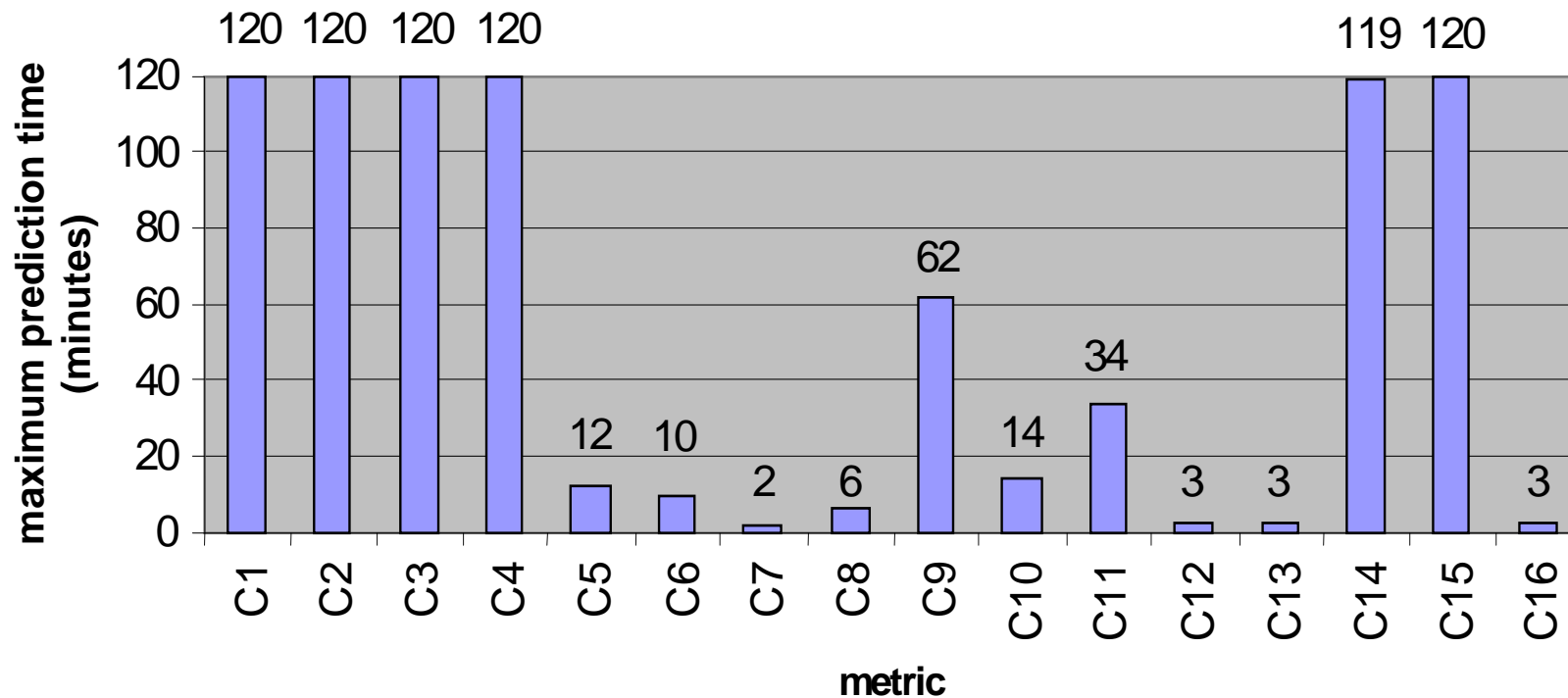
<b>C1</b>	<b>Number of aircraft/sector capacity</b>
<b>C2</b>	<b>Number of climbing aircraft</b>
<b>C3</b>	<b>Number of cruising aircraft</b>
<b>C4</b>	<b>Number of descending aircraft</b>
<b>C5</b>	<b>Horizontal proximity metric 1</b>
<b>C6</b>	<b>Vertical proximity metric 1</b>
<b>C7</b>	<b>Horizontal proximity measure 2</b>
<b>C8</b>	<b>Vertical proximity measure 2</b>
<b>C9</b>	<b>Horizontal proximity measure 3</b>
<b>C10</b>	<b>Vertical proximity measure 3</b>
<b>C11</b>	<b>Time-to-go to conflict measure 1</b>
<b>C12</b>	<b>Time-to-go to conflict measure 2</b>
<b>C13</b>	<b>Time-to-go to conflict measure 3</b>
<b>C14</b>	<b>Variance of speed</b>
<b>C15</b>	<b>Ratio of standard deviation of speed to average speed</b>
<b>C16</b>	<b>Conflict resolution difficulty based on crossing angle</b>

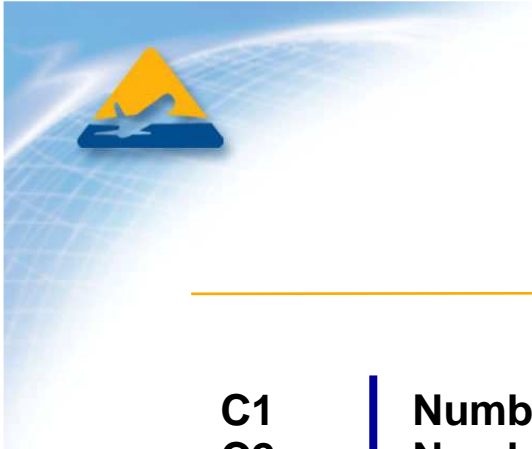


# Minimally-Useful Prediction Horizon:

R threshold = 0.3 ( $R^2 = 0.09$ )

Correlations between predicted and actual values were done for each look-ahead time; the point at which R falls below 0.3 is plotted.





## Useful Variables (120 min)

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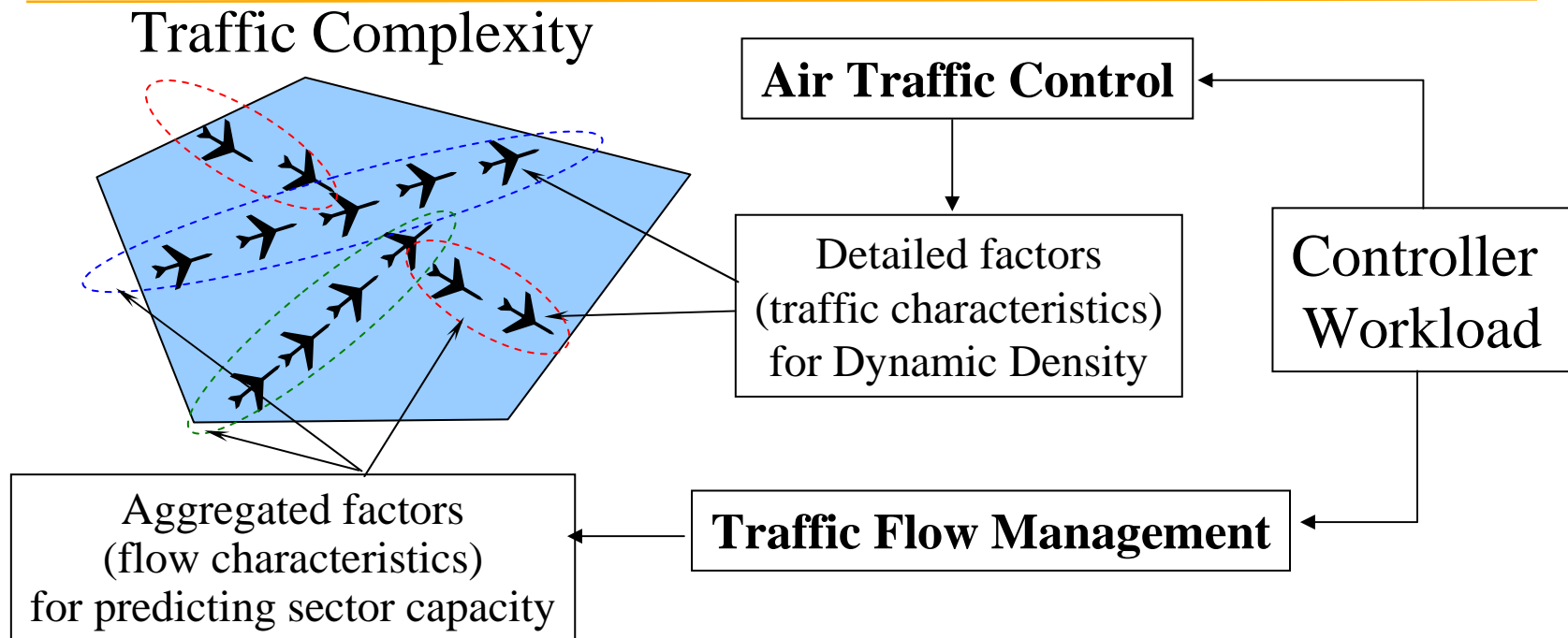
<b>C1</b>	<b>Number of aircraft/sector capacity</b>
<b>C2</b>	<b>Number of climbing aircraft</b>
<b>C3</b>	<b>Number of cruising aircraft</b>
<b>C4</b>	<b>Number of descending aircraft</b>
C5	Horizontal proximity metric 1
C6	Vertical proximity metric 1
C7	Horizontal proximity measure 2
C8	Vertical proximity measure 2
C9	Horizontal proximity measure 3
C10	Vertical proximity measure 3
C11	Time-to-go to conflict measure 1
C12	Time-to-go to conflict measure 2
C13	Time-to-go to conflict measure 3
<b>C14</b>	<b>Variance of speed</b>
<b>C15</b>	<b>Ratio of standard deviation of speed to average speed</b>
C16	Conflict resolution difficulty based on crossing angle

**Predictable variables are aggregate flow features rather than aircraft-to-aircraft interactions.**





# Flow Structure to Represent Traffic Complexity for TFM

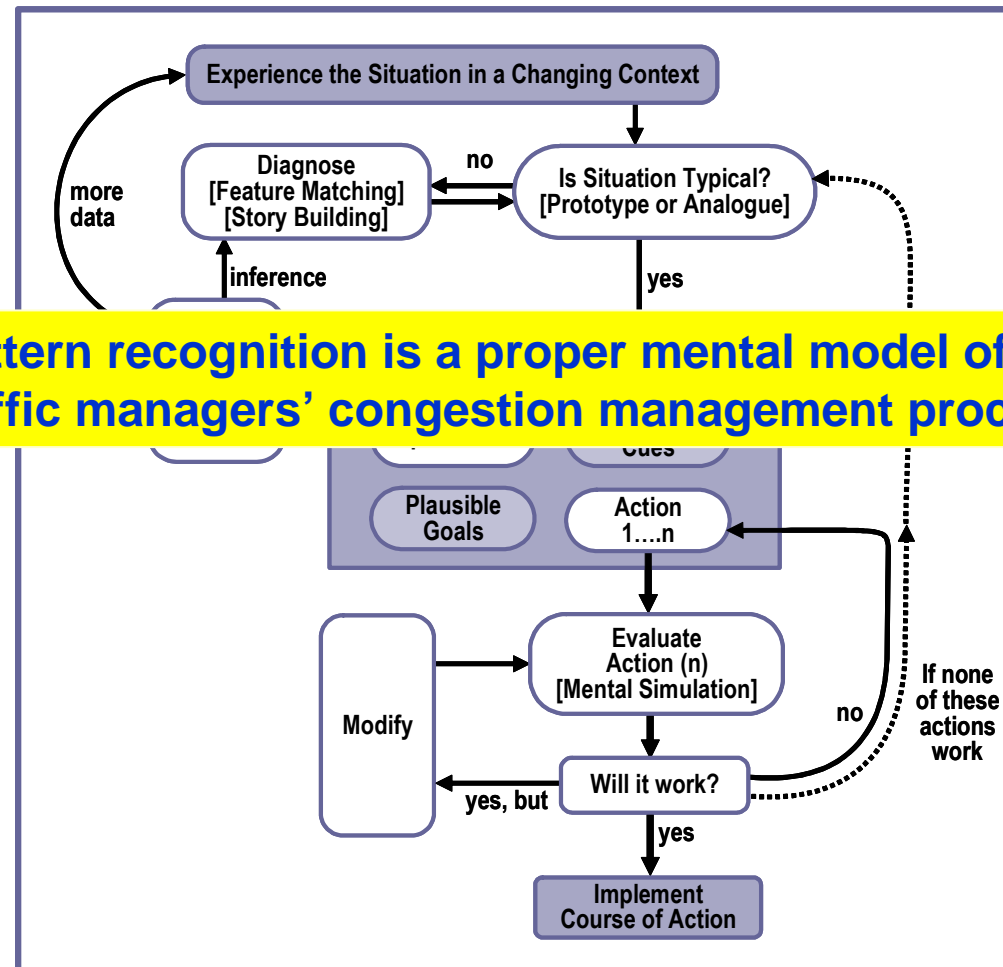


- **Aggregate flow variables are more easily predicted at TFM time frames than aircraft interaction variables**
- **Flow structure is tightly related to controllers control strategy**
- **Flow structure provides insight into congestion resolution options**



# Decision Theory under Stress and Uncertainty

Cited from “Recognition-Primed Decisions” by Klein, G., 1989.

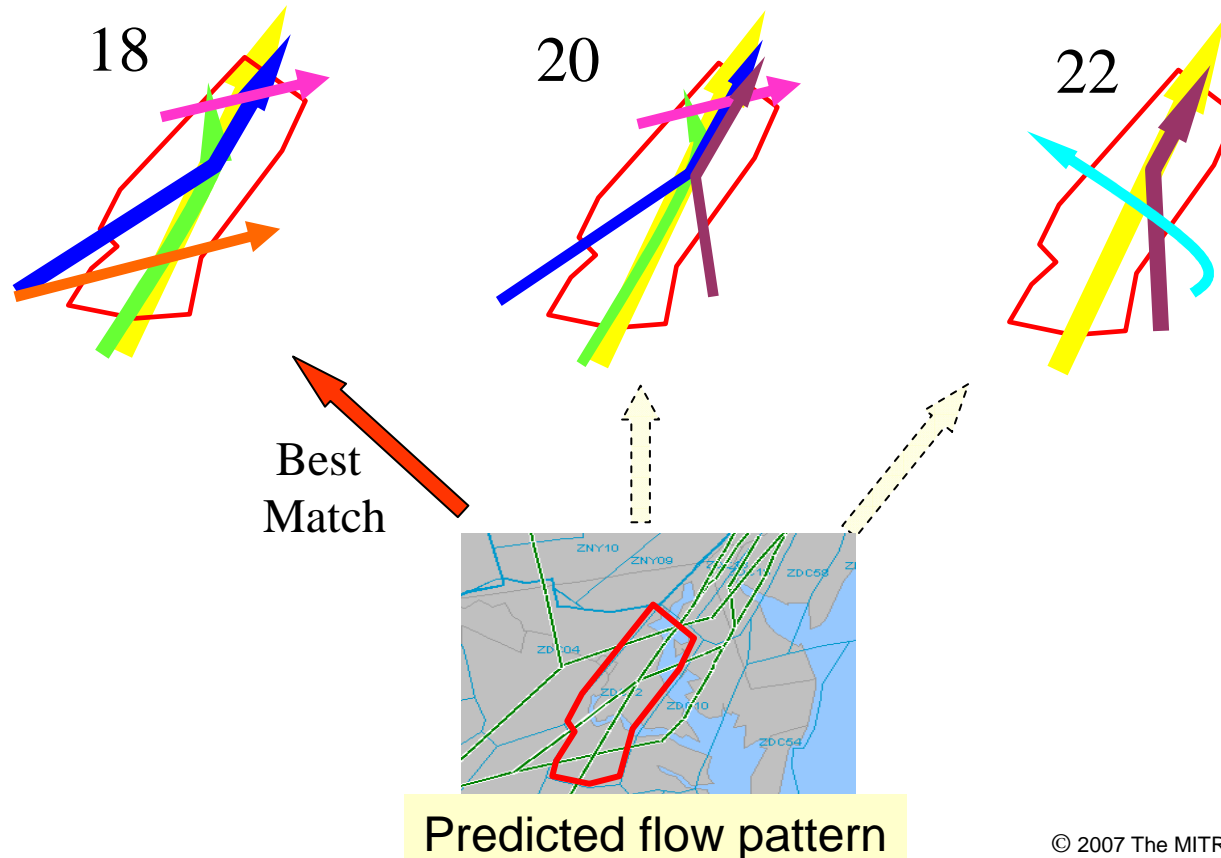


**Pattern recognition is a proper mental model of traffic managers' congestion management process.**



# Proposed Approach

- Identify the primary set of traffic flow patterns for each sector
- Assess the sector capacity for each pattern of the set
- Predict the sector capacity through pattern recognition

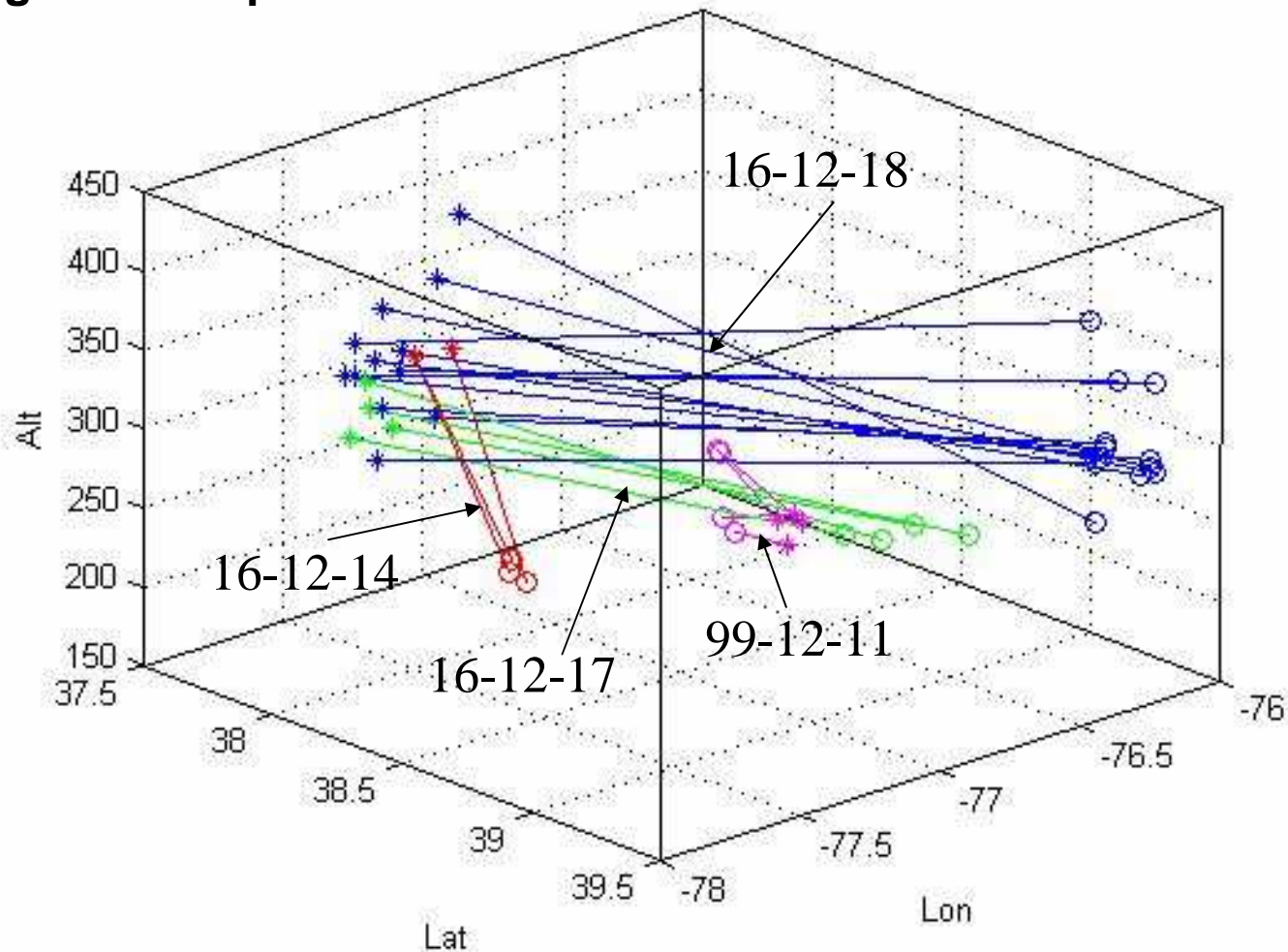




# Sector Transit Triplets to Represent Flows

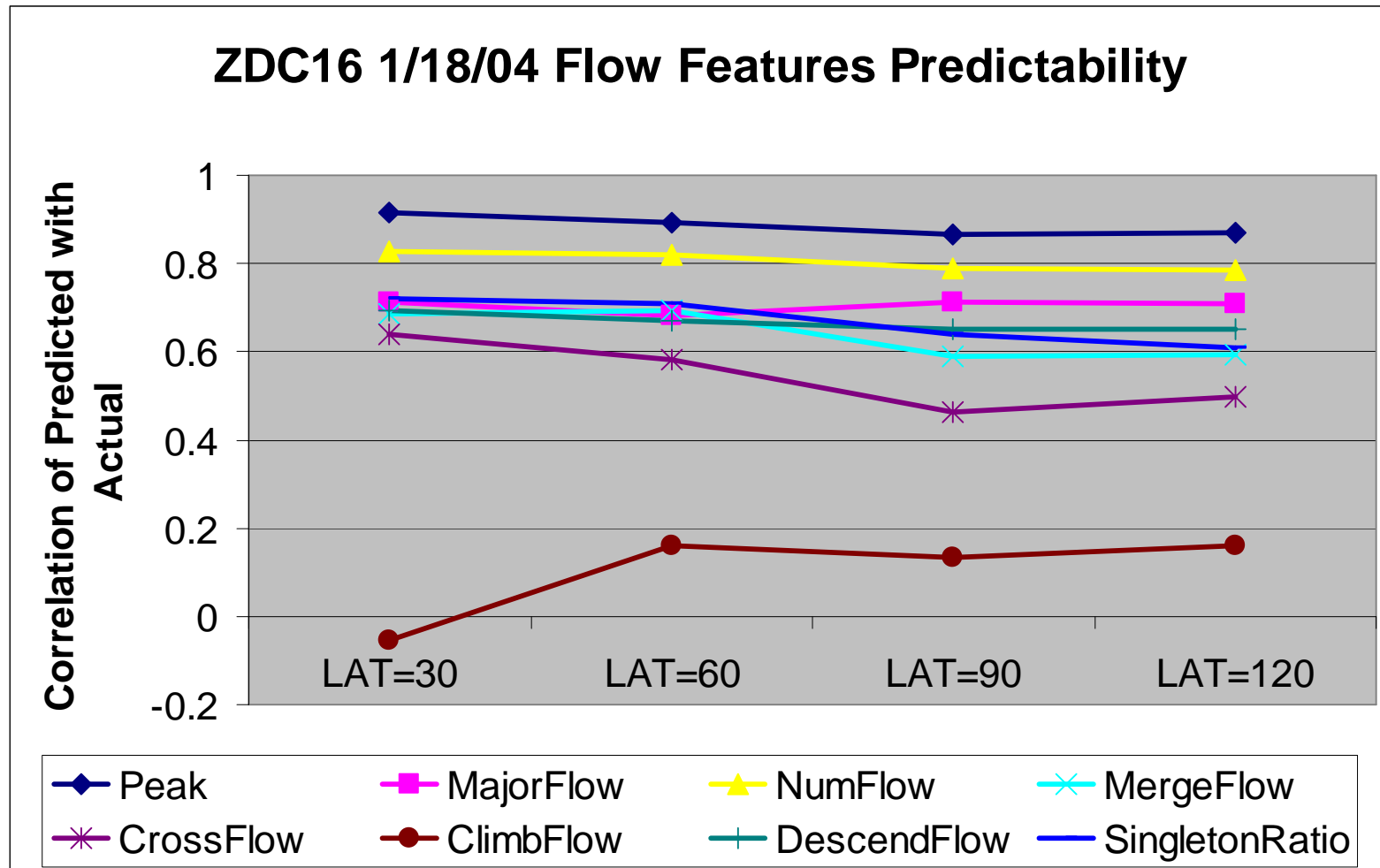
1/6/04 14Z ZDC12 Major Triplets (actual traffic)

- **Flows are the triplets that have at least two aircraft within the given time period.**



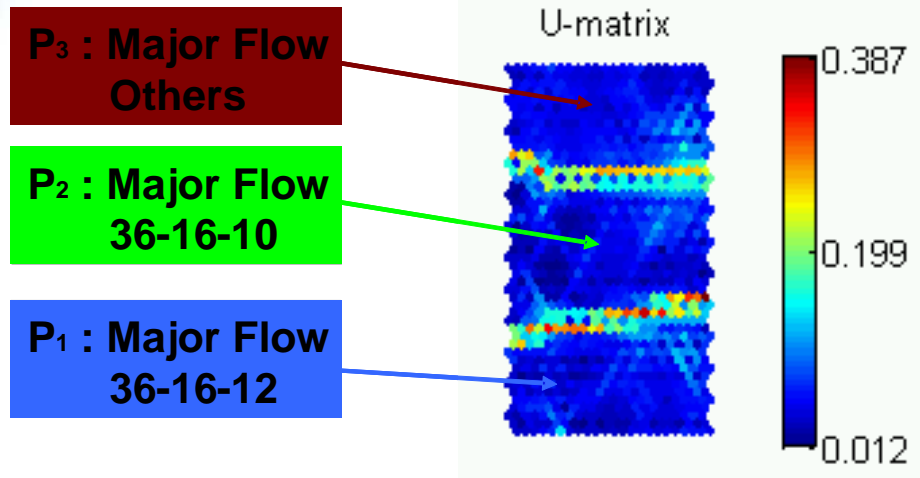


# Predictable Flow Features to Describe Traffic Flow Patterns

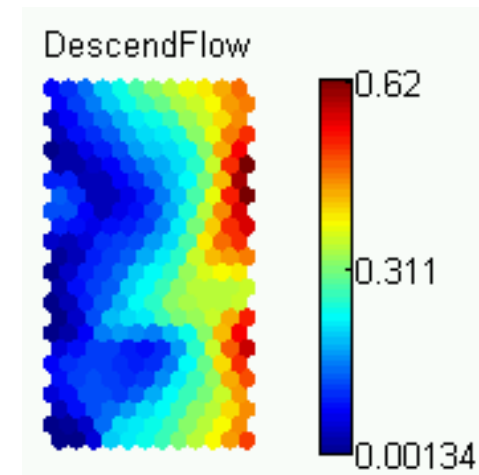
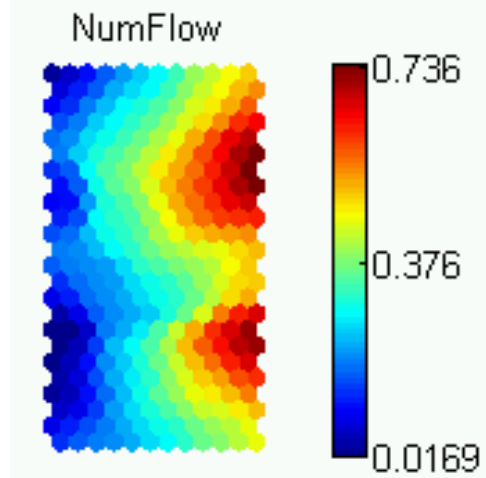
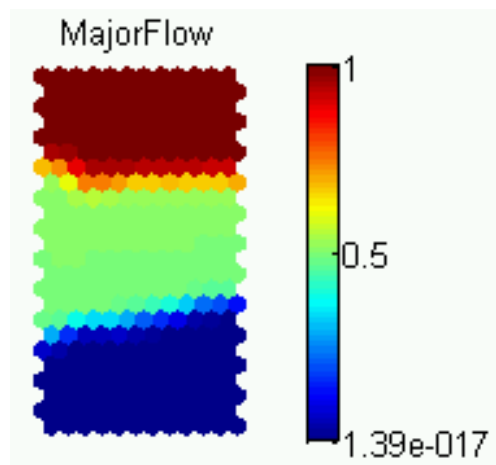




# Identify Primary Set of Traffic Flow Patterns with Self-Organizing Maps (SOMs)



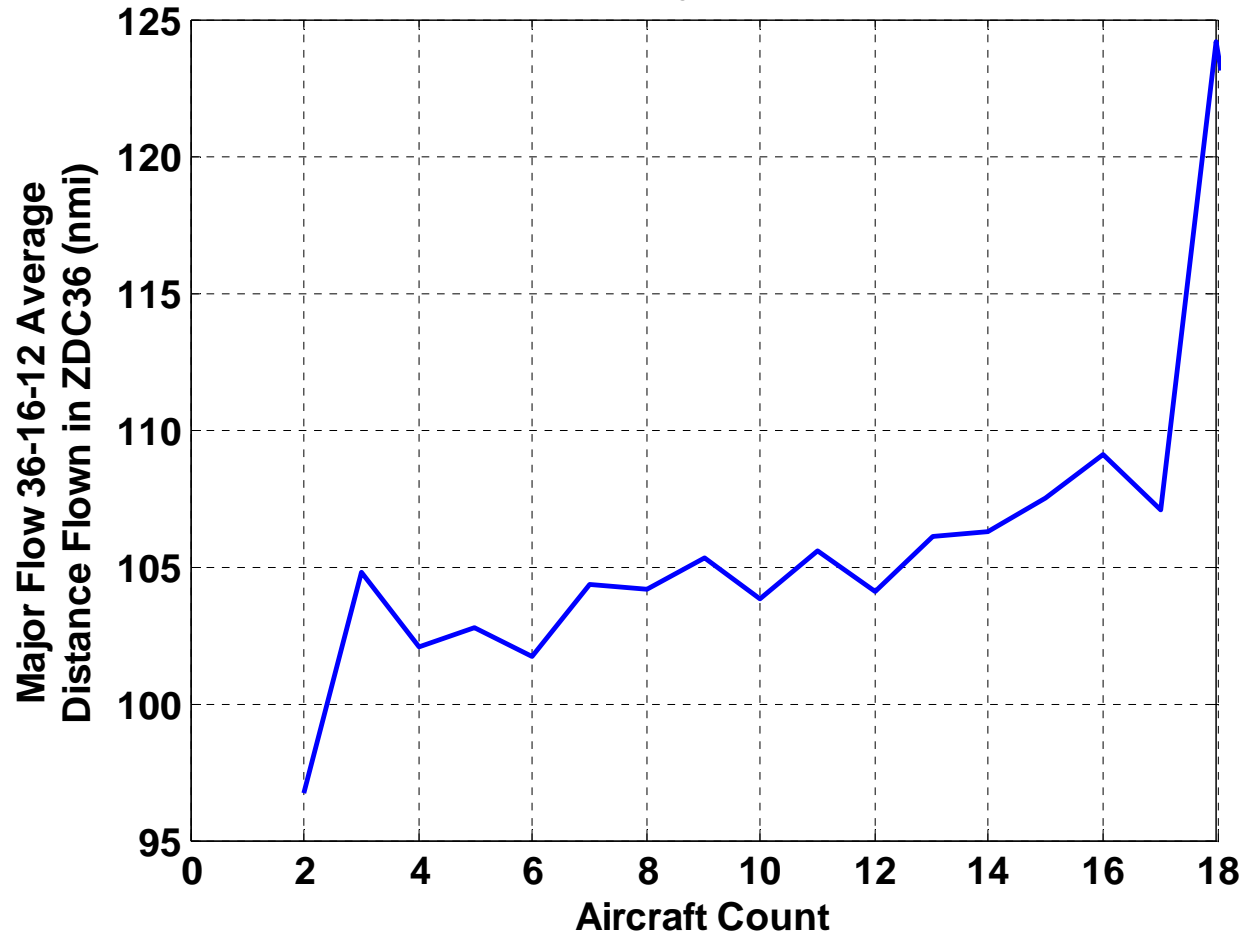
- Unified distance matrix visualizes distances between neighboring traffic flow patterns, and helps to see the cluster structure of the traffic flow patterns
- Each component plane shows the values of each feature in the patterns organized in U-Matrix





# Assessing Sector Capacity Through Observing System Performance

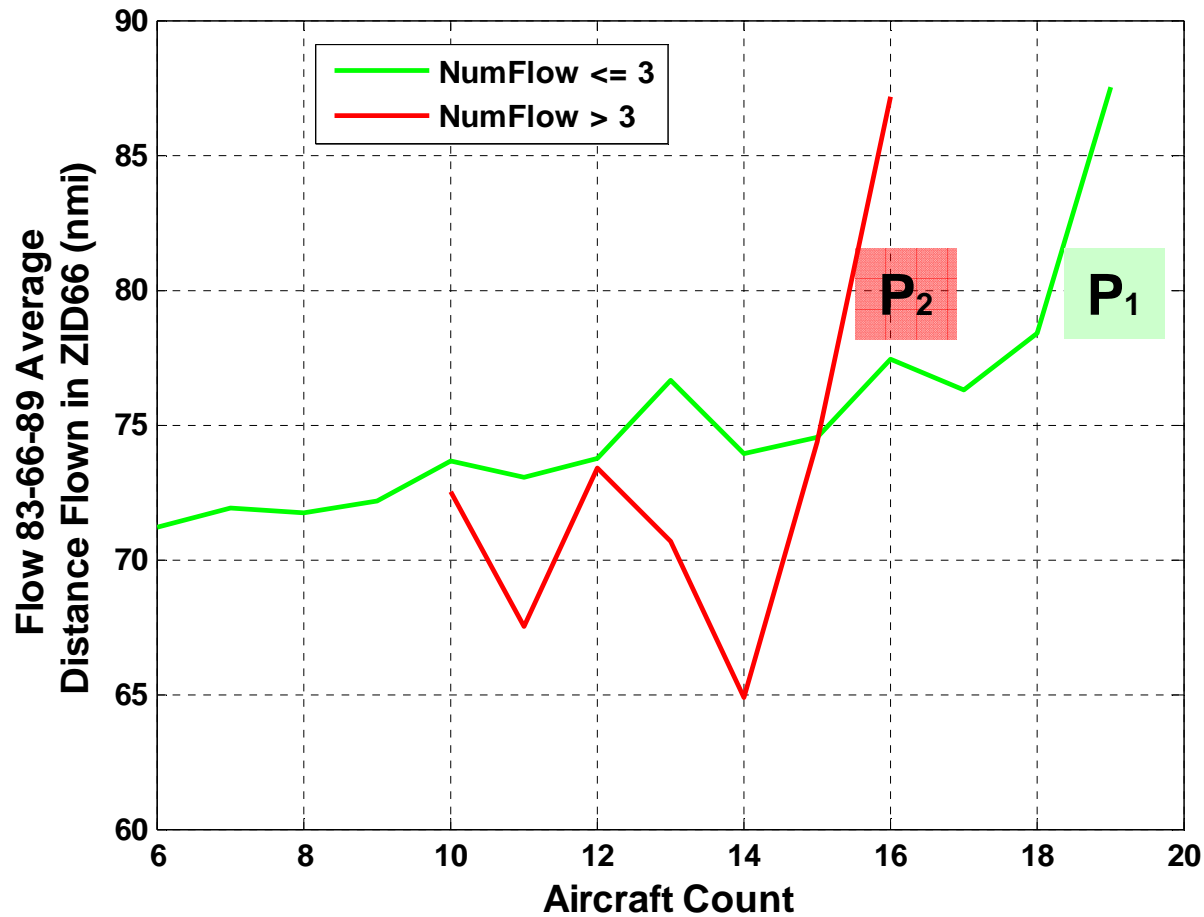
ZDC16 with Major Flow 36-16-12





# Performance Curves of Flow Patterns with Different Levels of Complexity

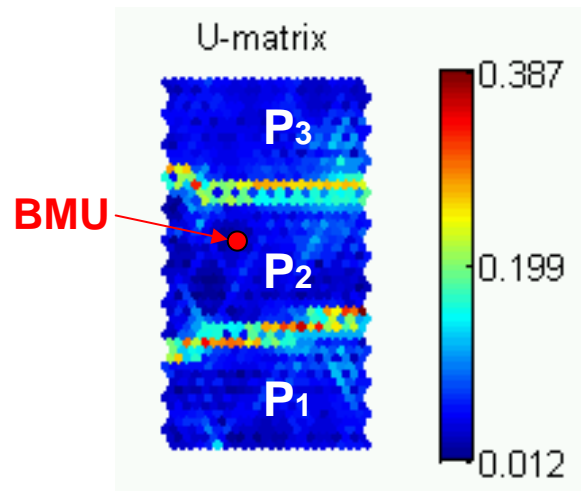
ZID66 with Different Number of Flows







# Predict Sector Capacity through Pattern Recognition with SOM



- Given a predicted traffic flow pattern with the flow feature vector  $\mathbf{x}(t)$
- The map takes the input vector  $\mathbf{x}(t)$  and goes through each map unit to find the Best Matched Unit (BMU)  $\mathbf{m}_c(t)$

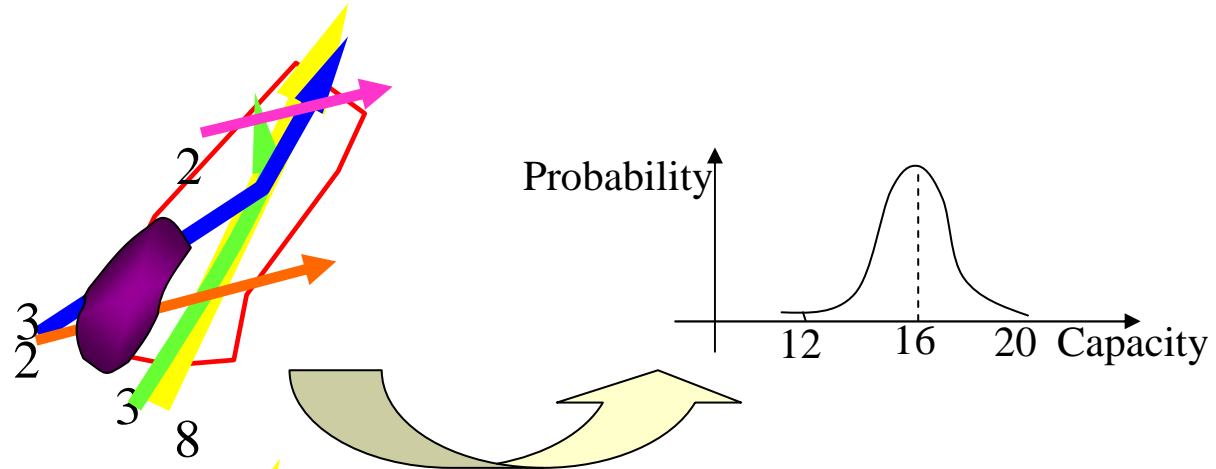
$$\forall i, \|\mathbf{x}(t) - \mathbf{m}_c(t)\| \leq \|\mathbf{x}(t) - \mathbf{m}_i(t)\|.$$

- The BMU belongs to P<sub>2</sub>, so the predicted sector capacity is C<sub>2</sub>

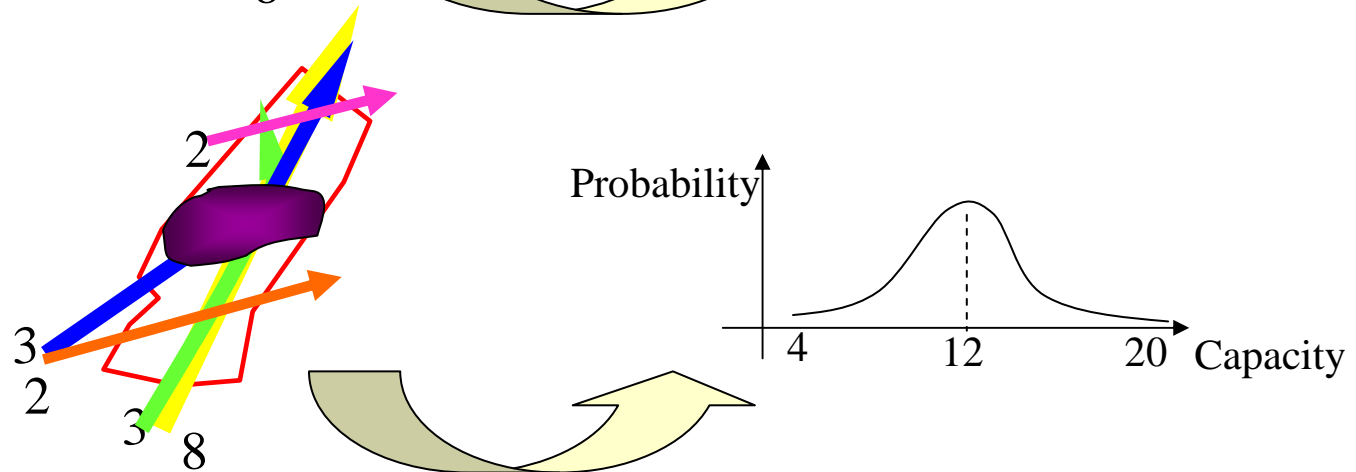


# Example Sector Capacity under Severe Weather Impact (1)

Case A



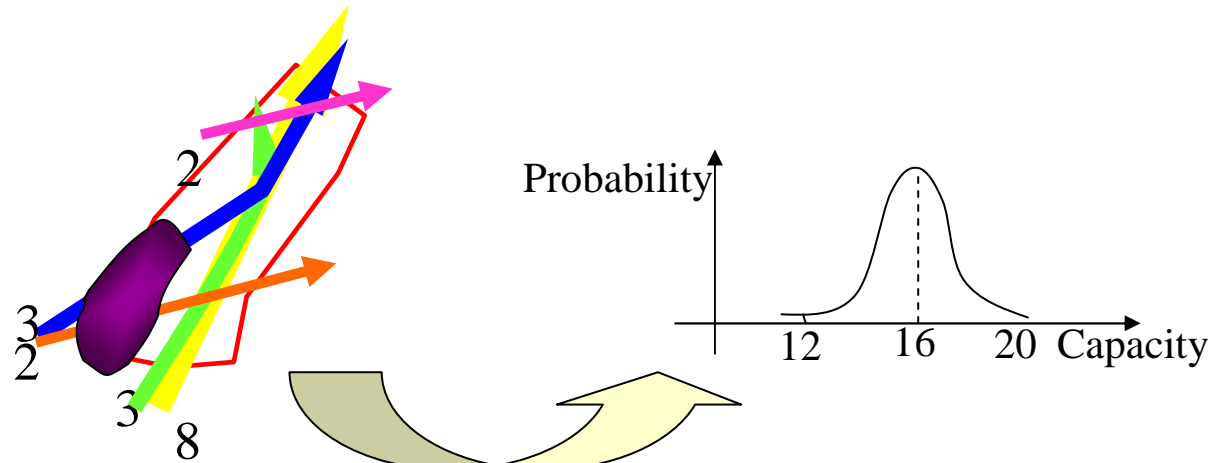
Case B



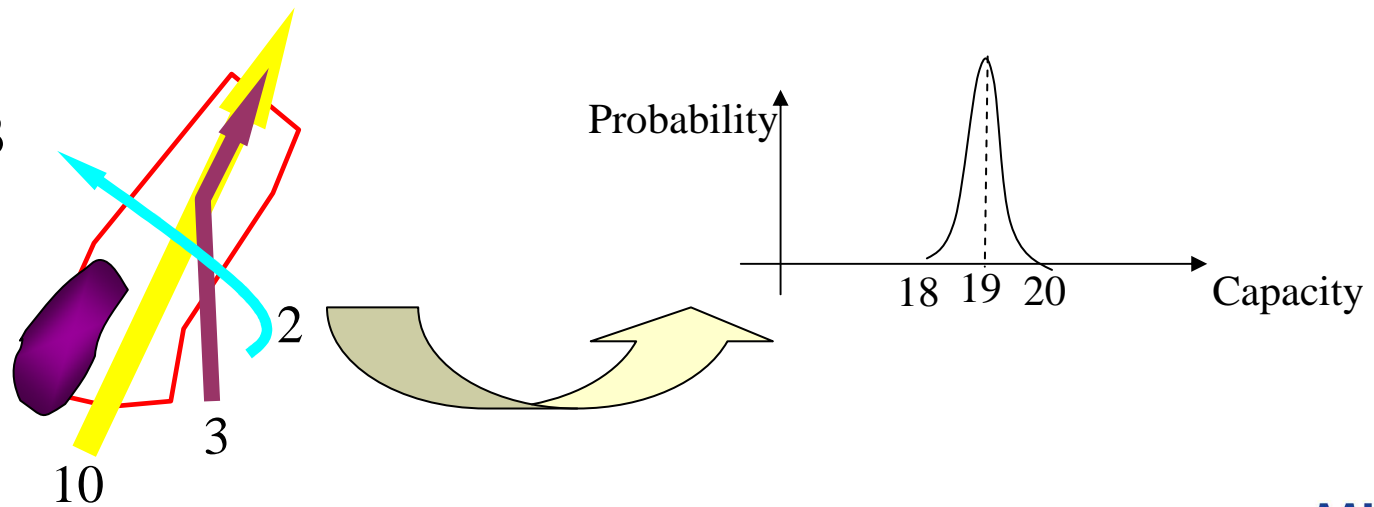


# Example Sector Capacity under Severe Weather Impact (2)

Case A



Case B





## Conclusion

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- **Sector capacity is still defined in terms of number of aircraft, but as a function of traffic flow pattern to consider complexity.**
- **Predictable flow features are used to describe traffic flow patterns, which makes the traffic complexity and sector capacity predictable.**
- **Pattern recognition is intuitive and relevant to controllers and managers decision making process.**
- **It helps not only on predicting the congestion, but also resolving the congestion**
  - **through both reducing the demand and increasing the capacity**
- **Assessing the capacity through observing system performance avoids measuring how hard controllers' working and predefining the workload threshold.**
- **Traffic flow pattern provides a basis for capturing weather impact on sector capacity.**