

Statistical Prediction of Flight Behavior in the Vicinity of Convective Weather

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- 2 of 3 events correctly predicted
- 1 of 4 alerts are false alarms

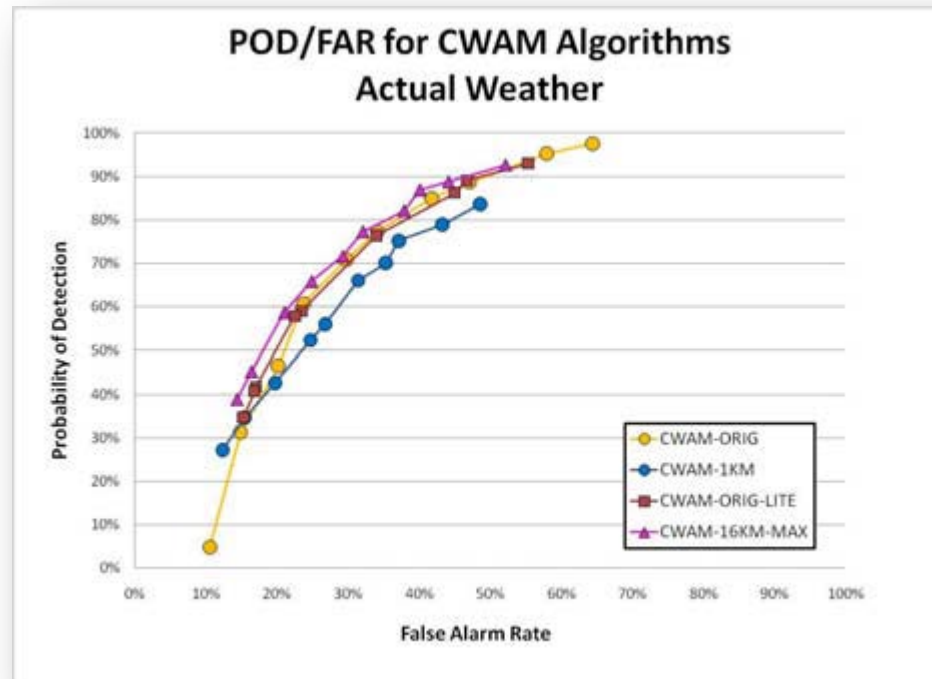


Figure reproduced from Matthews, M., DeLaura, R., "Evaluation of Enroute Convective Weather Avoidance Models Based on Planned and Observed Flight", 14th Conference on Aviation, Range, and Aerospace Meteorology (ARAM), Amer. Meteor. Soc., Atlanta, GA, Jan 2010.

Gaussian Copula Likelihood Approximation for Arbitrary Feature Vector Data

- Putting it all together ...

$$\Pr \{y|x\} \approx \prod_{i=1}^m \frac{\frac{1}{\sigma_x^{(i)}} f_x^{(i)} \left(\frac{y^{(i)} - \mu_x^{(i)}}{\sigma_x^{(i)}} \right)}{\underbrace{\varphi \left(\Phi^{-1} F_x^{(i)} \left(\frac{y^{(i)} - \mu_x^{(i)}}{\sigma_x^{(i)}} \right) \right)}_{\text{non-gaussianity-}\mathcal{N}(0, C_x) \text{ Jacobian correction term}}}$$

$$\frac{y^{(i)} - \mu_x^{(i)}}{\sigma_x^{(i)}} = \text{standardized data}$$

$$f_x^{(i)} = \text{p.d.f. of } \frac{Y^{(i)} - \mu_x^{(i)}}{\sigma_x^{(i)}}$$

$$F_x^{(i)} = \text{c.d.f. of } \frac{Y^{(i)} - \mu_x^{(i)}}{\sigma_x^{(i)}}$$

$$\frac{1}{\sqrt{\det 2\pi C_x}} \exp \left(-\frac{1}{2} \left[\Phi^{-1} F_x \left(\frac{y - \mu_x}{\sigma_x} \right) \right]^{\text{tr}} C_x^{-1} \left[\Phi^{-1} F_x \left(\frac{y - \mu_x}{\sigma_x} \right) \right] \right)$$

$\mathcal{N}(0, C_x)$ density evaluated at $\Phi^{-1} F_x((y - \mu_x)/\sigma_x)$

John Frank Stevens





High resolution weather information w/ minimal latency



Conditioned flight dataset



Integrated analysis environment



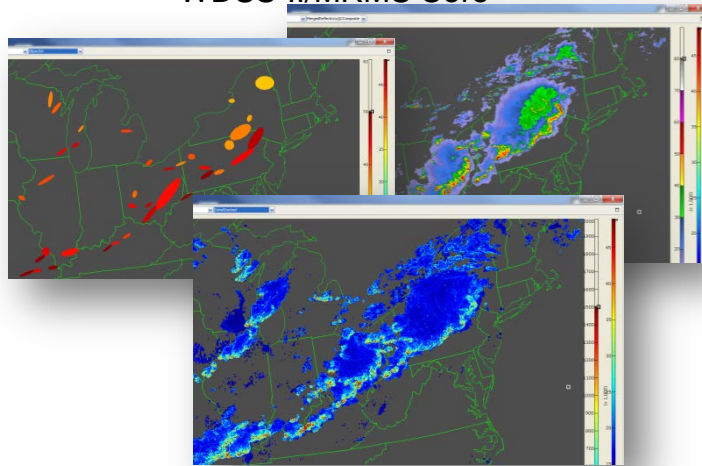
Clear definitions for weather-related flight deviation



Automated identification of deviations when they occur

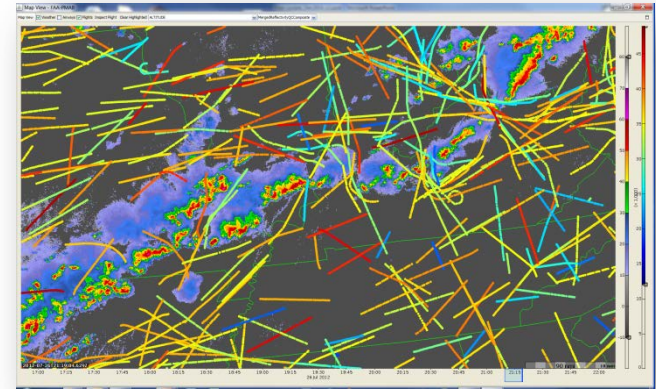
High Resolution 3D CONUS-wide Weather Feature Mosaics

New feature extraction routines built on FAA/NWS WDSS-II/MRMS Core



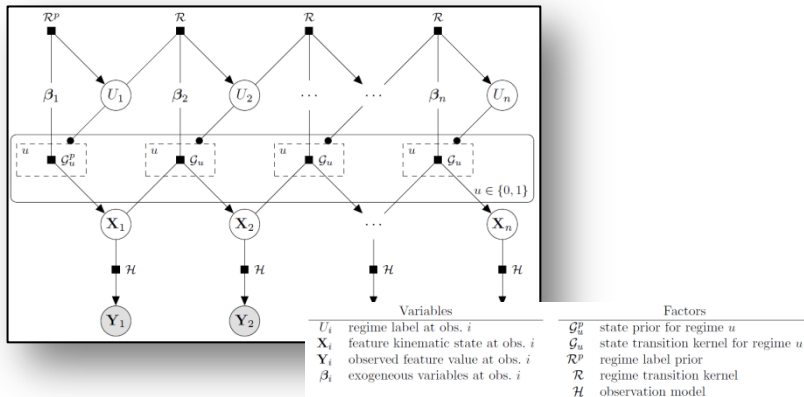
PDARS – Wx Analysis Environment - Flight View

Built using Java and OpenGL.
Specifically designed for rapid visual analysis of very large multi-dimensional data sets



Markov Switching Model for Flight Deviation Detection

Implemented in Java for scalable processing of large data sets. Approximately 1 minute to process 30,000 flights



Gaussian Copula-based Classifier

Implemented in Java. Adaptable to arbitrary feature data, including non-Gaussian data

Gaussian Copula Likelihood Approximation for Arbitrary Feature Vector Data

- Putting it all together...

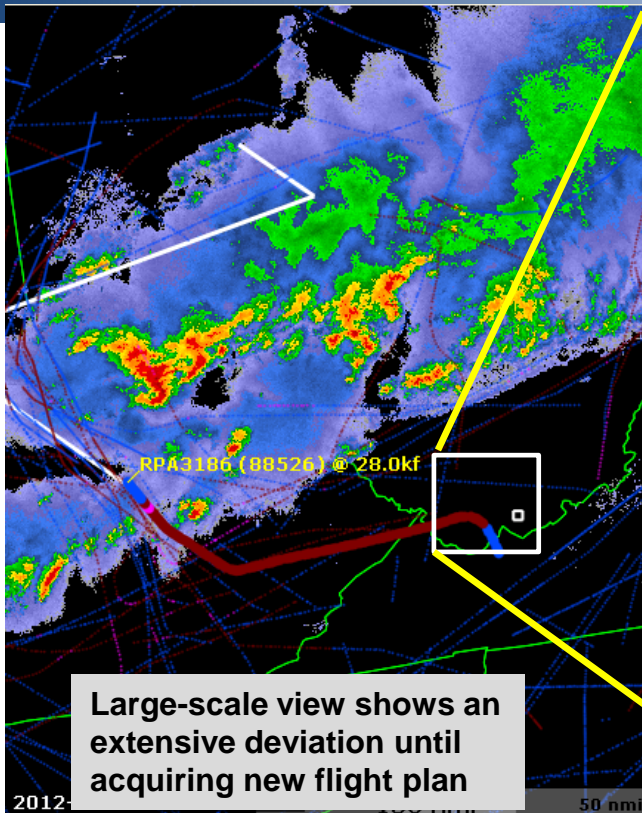
$$\Pr\{y|\tau, \alpha\} \approx \prod_{i=1}^m \frac{1}{\sigma_{z,\alpha}^{(i)}} f_{\tau}^{(i)} \left(\frac{y^{(i)} - \mu_{z,\alpha}^{(i)}}{\sigma_{z,\alpha}^{(i)}} \right) \phi \left(\Phi^{-1} F_{\tau}^{(i)} \left(\frac{y^{(i)} - \mu_{z,\alpha}^{(i)}}{\sigma_{z,\alpha}^{(i)}} \right) \right)$$

nongaussianity- $\mathcal{N}(\mathbf{0}, \mathbf{C}_{\tau})$ Jacobian correction term

$$\frac{1}{\sqrt{\det 2\pi \mathbf{C}_{\tau}}} \exp \left(-\frac{1}{2} \left[\Phi^{-1} \mathbf{F}_{\tau} \left(\frac{y - \mu_{z,\alpha}}{\sigma_{z,\alpha}} \right) \right]^T \mathbf{C}_{\tau}^{-1} \left[\Phi^{-1} \mathbf{F}_{\tau} \left(\frac{y - \mu_{z,\alpha}}{\sigma_{z,\alpha}} \right) \right] \right)$$

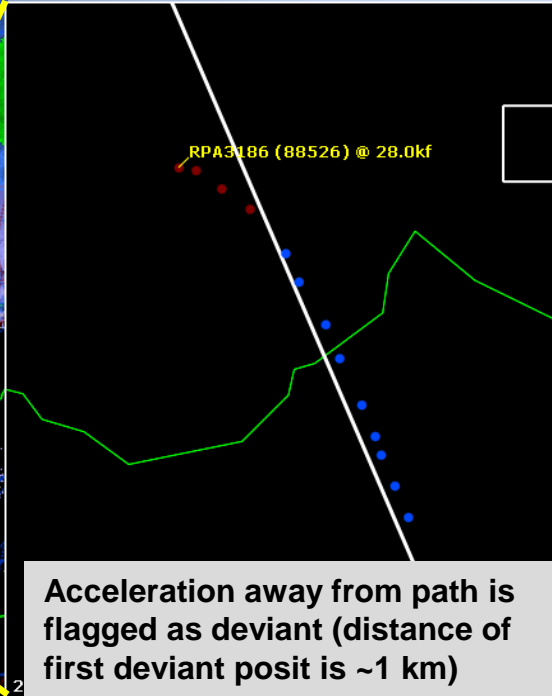
$\mathcal{N}(\mathbf{0}, \mathbf{C}_{\tau})$ density evaluated at $\Phi^{-1} \mathbf{F}_{\tau}((y - \mu_{z,\alpha}) / \sigma_{z,\alpha})$

$\frac{y^{(i)} - \mu_{z,\alpha}^{(i)}}{\sigma_{z,\alpha}^{(i)}} = \text{standardized data}$
 $f_{\tau}^{(i)} = \text{p.d.f. of } \frac{Y^{(i)} - \mu_{z,\alpha}^{(i)}}{\sigma_{z,\alpha}^{(i)}}$
 $F_{\tau}^{(i)} = \text{c.d.f. of } \frac{Y^{(i)} - \mu_{z,\alpha}^{(i)}}{\sigma_{z,\alpha}^{(i)}}$

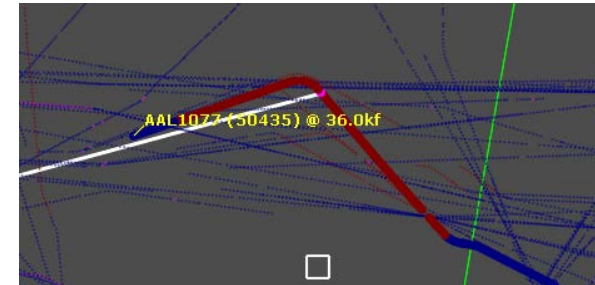


Large-scale view shows an extensive deviation until acquiring new flight plan

- Declared as Deviation
- Declared as Non-deviation



Examples of correct behavior during flight plan crossings



Statistic	7/26/12 (severe)		6/8/12 (mostly benign)	
	Conservative	Aggressive	Conservative	Aggressive
Deviations near weather	3,172	5,728	746	1,451
Deviations not near weather	1,391	2,979	1,182	2,849
All deviating flights	3,849	6,821	1,819	3,913
Terminal deviations near weather	1,238	2,568	629	1,270
Terminal deviations not near weather	1,398	3,133	1,464	3,517
Total number of flights	29,424		29,707	

For comparison, published CWAM training/test data set captured a total of 1,564 deviations over 5,235 events

Comparison of Approaches

	CWAM	PMAB-CW
Data Set	<i>5,000 Interactions</i>	<i>2,400,000 Flights</i>
ARTCCs	<i>3</i>	<i>All</i>
Deviation Detection	<i>Lateral Distance & Manual Review</i>	<i>Automated Markov Switching Model</i>
Potential for Performance Gains	<i>Low</i>	<i>High</i>
Output Type	<i>Geographic</i>	<i>Trajectory + Geographic* (* future add)</i>
Airline Specific Classifiers	<i>No</i>	<i>Yes</i>
Aircraft Type Classifiers	<i>No</i>	<i>Yes</i>
Classifiers by Region	<i>No</i>	<i>Yes</i>
Classifiers by Season	<i>No</i>	<i>Yes</i>
Classifiers by Day/Night	<i>No</i>	<i>Yes</i>

Metron's Scalable Approach Provides Flexibility & Increases Feasibility for New Applications

Benefits to the NAS

