



A cluster-based flight data analysis and visualization software

Lishuai Li and Florent Charruaud

Department of Systems Engineering and Engineering Management
City University of Hong Kong

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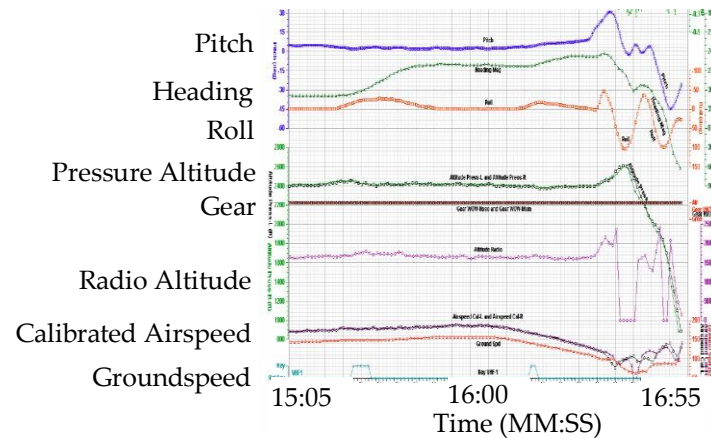
Opportunities arise from an abundance of flight data

- Airlines monitor flight operations extensively
- An abundance of data available from Digital Flight Data Recorder (the Black Box) and Quick Access Recorder
- Limited for accident investigations in the past
- Advancement in data storage, access, and analytics offer new opportunities

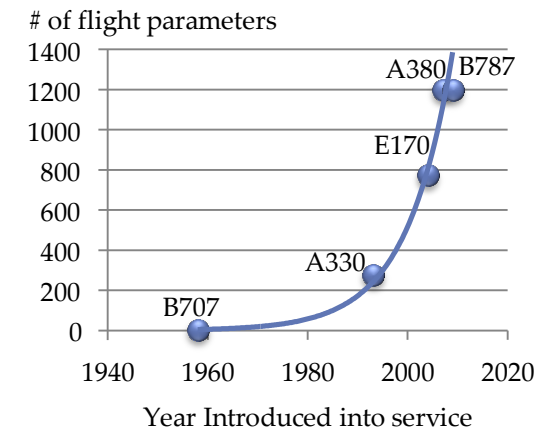
Illustration of flight data devices



Flight Data Example



Evolution of FDR data capacity



Limited existing analytical tools, leaving valuable information in the data untapped ...

- Current flight data analysis tools are based on “Exceedance Detection” (ED)
 - It detects unsafe events based on a pre-defined list of “exceedance events”, which are known issues of safety concerns
 - e.g. a “Excessive Power Increase” event is detected when
 - “HAT < 500 feet, Δ of N1 > x”
- ED cannot respond to emerging, previously unidentified issues that are potentially dangerous; The risks are high if the list is not updated in time

We aim to develop a cluster-based flight data analysis and visualization software: ClusterAD

Highlights

- A software tool for flight data analysis and visualization using data mining methods
- Enable airline decision makers to grasp latent patterns and obtain valuable insights about their flight operations for safety management and pilot training:

Identify common patterns of current flight operations

Detect anomalies without specifying any exceedance criteria

Monitor pilot proficiency / competency

ClusterAD framework

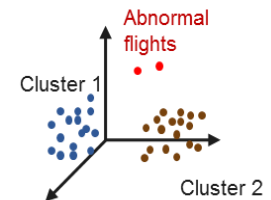
Flight Data



Cluster-based
Anomaly Detection

Abnormal
Flights

Common
Patterns



Expert Review

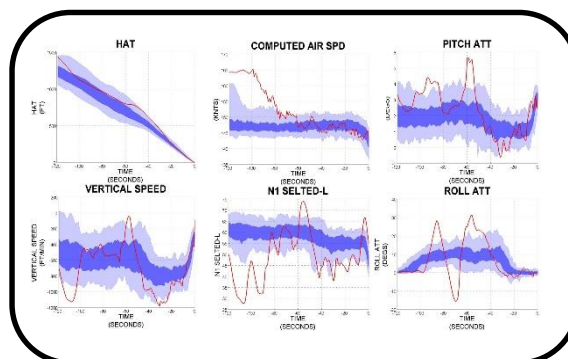
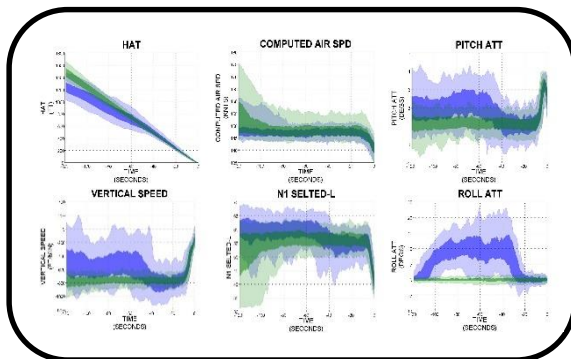
Benign
Cases

Watch
Cases

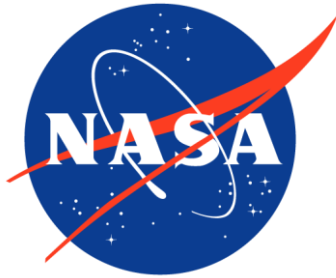


ClusterAD helps airlines see and understand flight data

- Flight data monitoring
 - Broaden the tool set to allow more flexible, broad-reaching, analysis of data
- Safety
 - Understand patterns of current operations better
 - Identify abnormal flights with operational significance
 - Suggest risks unknown to conventional diagnosis
- Efficiency
 - Gain insights on potentials for fuel efficiency improvements



ClusterAD has been tested with a number of airlines via research collaborations



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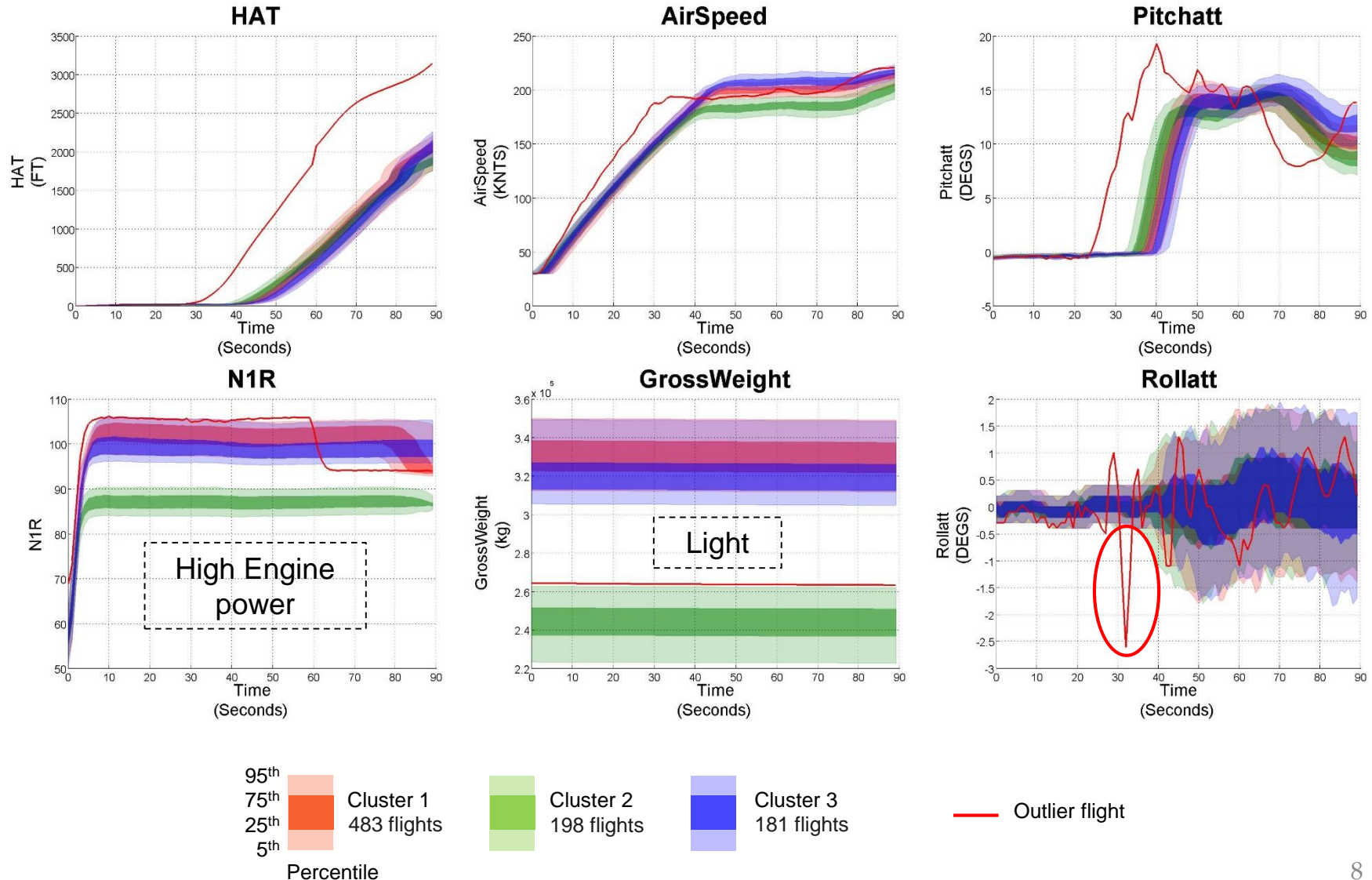
Selected papers related to ClusterAD:

1. Lishuai Li, Santanu Das, R. John Hansman, Rafael Palacios, and Ashok N. Srivastava. "Analysis of Flight Data Using Clustering Techniques for Detecting Abnormal Operations," *Journal of Aerospace Information Systems*. Vol. 12, No. 9 (2015). pp. 587-598. doi: [doi:10.2514/1.1010329](https://doi.org/10.2514/1.1010329) [PDF]
2. Lishuai Li, R. John Hansman, Rafael Palacios, and Roy Welsch. "Anomaly Detection via a Gaussian Mixture Model for Flight Operation and Safety Monitoring," *Transportation Research Part C: Emerging Technologies*. Vol.64 (2016). pp. 45-57. doi: [doi: 10.1016/j.trc.2016.01.007](https://doi.org/10.1016/j.trc.2016.01.007) [PDF]
3. Florent Charruaud, Lishuai Li. "Flight Operations Monitoring through Cluster Analysis: A Case Study," *Intelligent Systems, IEEE*. Vol. 30, No. 6 (2015). pp. 24-29. doi: [doi: 10.1109/MIS.2015.111](https://doi.org/10.1109/MIS.2015.111) [PDF]

Case study 1: Takeoff analysis using ClusterAD

- Objective:
 - Can we **identify common patterns** and **detect anomalies** of flights taking off from a selected airport via ClusterAD?
- Dataset:
 - 957 flights, B777-300ER
- Flight parameters included in the analysis:
 - Airspeed, HAT, groundspeed, normal acceleration, lateral acceleration, longitudinal acceleration, N1, gross weight, pitch attitude, roll attitude, fuel flow, cross wind, head wind, flap angle, angle of attack, vertical speed

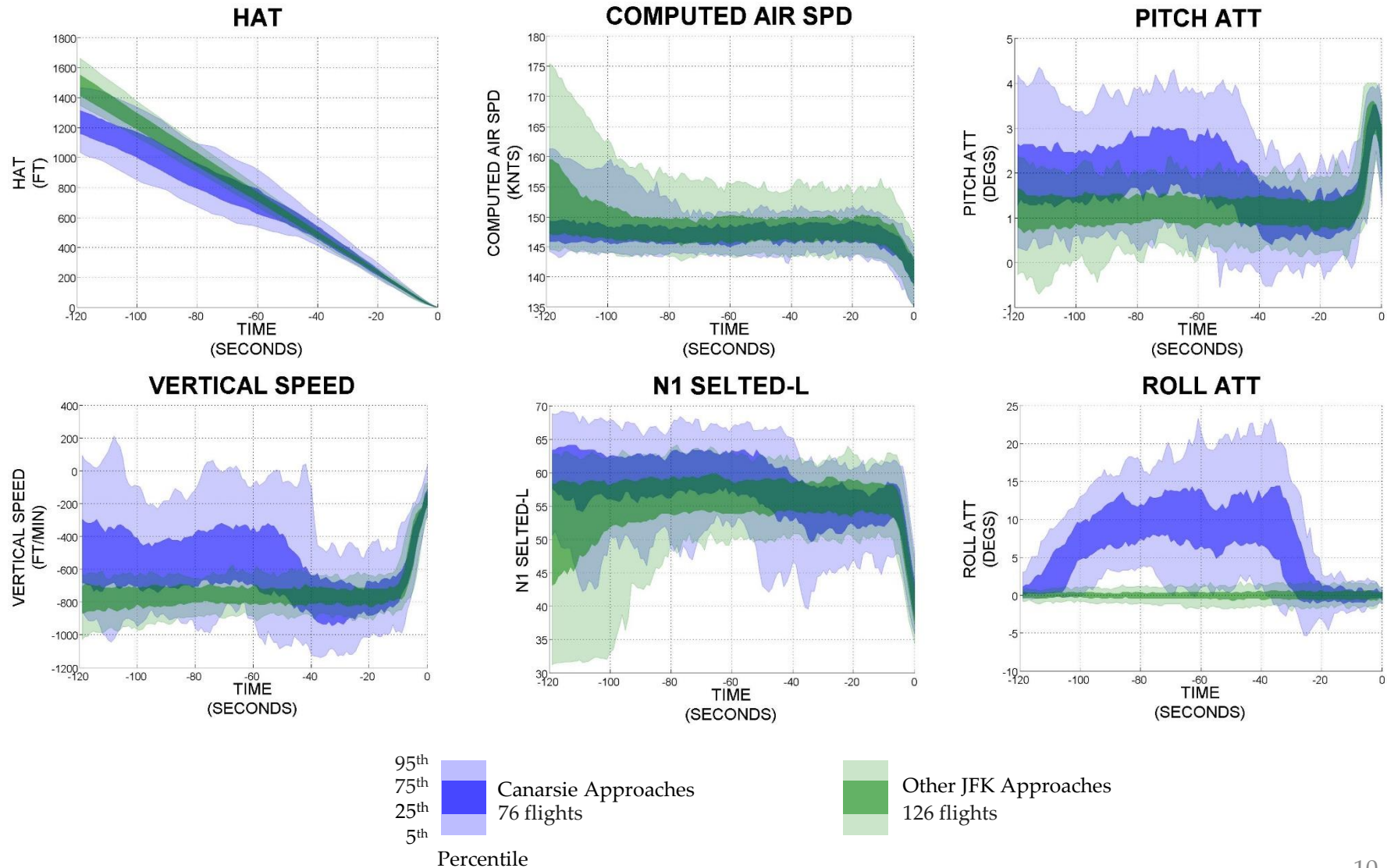
Case study 1: Three common patterns and anomalies are automatically identified



Case study 2: Landing analysis using ClusterAD

- Objective:
 - Issues have been reported regarding the Canarsie VOR approach at JFK airport. Can we identify the issues and reveal related causes using ClusterAD?
- Dataset:
 - 202 flights (76 Canarsie VOR approach and 126 other), B773ER
- Flight parameters:
 - Airspeed, HAT, groundspeed, normal acceleration, lateral acceleration, longitudinal acceleration, N1, gross weight, pitch attitude, roll attitude, fuel flow, cross wind, head wind, flap angle, angle of attack, vertical speed

Case study 2: Canarsie approach compared with other JFK approach



Case study 2: 3D visualization of JFK approaches



Thank you! Questions?



Contact:

Lishuai Li

Florent Charruaud

Assistant Professor

Research Associate

lishuai.li@cityu.edu.hk

florent.charruaud@gmail.com