

# Aircraft Operation Anomaly Detection Using FDR Data

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## **Motivation**

- Commercial aircraft accident rate has dropped significantly.
- Further improvement requires proactive safety management.
  - Identify risks in day-to-day operations
- Large amount of routine flight data "available"



- Flight Operations Quality Assurance (FOQA) in US
- Flight Data Monitoring (FDM) in Europe
- Information in flight data:
  - Rich about flight operations and risks
  - Underutilized by current practices





- Objective Identify emerging risks from routine flight data
- Issues: complexity of routine flight data
  - Large number of variables
  - Mix of relationships among variables
  - Variability among flights
    - ✓ Aircraft type
    - ✓ Procedures
    - ✓ Weather
    - ✓ Pilots





## **FDR Data**

- Flight Data Recorder (FDR)
  - Equips every commercial aircraft
  - Records 100+ to 1000+ flight parameters during each flight depending on aircraft and airline
- Rich information about
  - Aircraft
  - Environment
  - Crew operations
- Challenge for analysis
  - How to obtain useful information from massive data?





## **Current Data Analysis**

#### Exceedance detection

- Exceedance of a value under certain conditions
- List of events believed to be unsafe

#### Distribution analysis

• On specific queries, e.g. distribution of total energy at 900 ft AGL during approach, distribution of airspeed at takeoff, etc.

Event Code	Description	Measurement	Description
07A	Approach Speed Low Within 2 mins of T/D	nil	
07B	Approach Speed Low Below 25ft Radio	CASTDOWN	CAS AT TOUCHDOWN
		CASATR30	CAS MINUS VREF AT 30FT
08A	Climb Out Speed High Below 400ft AAL	MNCLS3540	MIN CLIMB SPEED 35FT TO 400FT
08B	Climb Out Speed High 400' to 1000' AAL	MIN400150	MIN CLIMB SPEED 400FT TO 1500FT
08C	Climb Out Speed Low 35' AGL to 400' AAL	MNCLS3540	MIN CLIMB SPEED 35FT TO 400FT

**Exceedance Event Examples** 

[Larder, Brian, and N. Summerhayes. 2004. Application of Smiths Aerospace Data Mining Algorithms to British Airways 777 and 747 FDM Data.]

#### Limitations

- Only known safety issues are examined.
- "You only get what you ask for."



## **Proposed Approach**

- Assumption
  - Majority of flights is safe
- Approach

Develop a method to track detailed in-flight recorded data

- Establish a *norm* of safe operations
- Identify *anomalies,* or *a*bnormal operations which indicate increased risks:
  - ✓ Vehicle impairment
  - ✓ External hazards
  - ✓ Inappropriate crew operations

✓ ...



## **Data Analysis Method**

#### Multivariate Cluster Analysis

- Use multiple variables over time
- Cluster flights

#### Establishment of a norm

- Norm: flight contained in clusters
- Anomalies: flights not belonging to a cluster
- Domain experts leverage on the results to identify emerging safety issues.



- Flights with unknown risks can be found
- "You don't need to specify what might be unsafe; but it tells"





## **Preliminary Study**

#### **Proof-of-concept demonstration on a limited FDR dataset**

- 1. Pre-filter a relatively homogeneous dataset
  - All B777 arrivals at Abu Dhabi Int'l Airport (183 flights)
  - Focused on final approach phase
- 2. Transform multiple time series into one vector for each flight
- 3. Cluster the vectors to identify norm and anomalies





# Subsets of flight parameters for Clustering

#### Data limitation

- 183 flights; 103 flight parameters
- Too sparse to form clusters if all parameters are used for clustering
- In the preliminary study, clustering is based on subsets of flight parameters:

Position	Longitude, latitude, height above threshold			
Position (with respect to runway)	Distance to threshold, deviation to centerline, height above threshold			
Position, heading, speed	Distance to threshold, deviation to centerline, height above threshold, heading relative to runway, speed measures			
Engine	N1, fuel flow, EGT, thrust lever, EMS thrust, N3			
Environment	Wind, temperature, pressure, air density			
Motion	Speeds, accelerations, load factor, pitch change rate, roll change rate, yaw change rate			
Control	Flap, slat, spoiler, elevator, stabilizer, trim, pitch, roll, yaw			
Force	Drag, lift, gross weight, CG position, normal load factor			



## **Cluster by Position**

Parameters included: Longitude, Latitude, Height above touchdown





## **Cluster by Position**





## Cluster by Position, Heading, & Speed







Parameters included: N1 for all engines, thrust, thrust lever, EGT for all engines, avg N3, avg fuel flow, etc.

#### N1: average (all engines, percent of maximum)





# Summary of outliers identified by subset

		Position (Absolute)	Position (Relative)	Position, Heading, Speed	Engine	Weather	Motion	Control	Force
	373564	1	0	0	0	0	0	0	0
	374577	1	1	0	0	0	0	1	1
_	377838	1	0	0	0	0	0	0	0
C	377844	1	1	1	1	1	0	0	0
	368467	0	1	1	1	0	0	0	0
	382554	0	1	1	0	1	0	0	0
	369204	0	0	0	0	1	0	0	0
	370512	0	0	1	0	1	1	0	0
	375699	0	0	1	0	0	0	0	0
	377288	0	0	1	0	0	0	1	1
	385160	0	0	1	0	0	0	0	0
	371927	0	0	0	1	0	0	0	0
	379659	0	0	0	1	0	0	0	0
	384089	0	0	0	1	0	0	0	0
	382520	0	0	0	0	1	0	0	0
	384512	0	0	0	0	1	0	0	0
	370713	0	0	0	0	0	1	0	0
	371044	0	0	0	0	0	1	1	1
	371929	0	0	0	0	0	1	0	0
	377860	0	0	0	0	0	1	1	1
_	379684	0	0	0	0	0	1	0	0
C	369202	0	0	0	0	0	0	1	1
	383279	0	0	0	0	0	0	0	1



### Example Anomaly: High Approach Easy to detect by current practices















### **Example Anomaly: Wind Gust** Difficult to detect by current practices















## **Summary & Future Work**

- Summary
  - Proposed an approach to identify emerging risks from routine flight data
  - Developed a method to track detailed flight data and define norm and anomalies for flight operations
  - Performed preliminary analysis on a limited FDR dataset

#### Next Steps

- Extend the analysis to other phase of flight
- Investigate parameters with no observable patterns over time
- Apply method to full FDR dataset (Data Wanted)



# Thank you!

# **Comments and questions?**