

Open Architecture for Data Mining and PHM

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Aircraft Data Monitoring





Data Mining Functions

- Data exploration
 - Model structure selection
 - Off line, is not a part of deployed monitoring function

Model training

- Multivariate regression

Data exploitation

- Anomaly detection
- Fault isolation (diagnostics)
- Predictive trending



V&V for Data Mining and PHM





Outline

- Background and past effort – Where did we come from?
- Completed effort
 - Where are we now?
- On-going effort
 - What are we going currently?
- Extensions
 - What are the follow-on steps?



Data Mining Demo

- Java EE SOA software: Openalytics
- Functions from Stanford NRA
 - Regression modeling of aircraft dynamics
 - Anomaly detection from residuals MSPC
 - Detail in E. Chu, D. Gorinevsky, and S. Boyd, AIAA Infotech@Aerospace, 2010
- FOQA data from NASA FLTz simulator
 - Cruise flight segment

- Many flights with varying conditions May 2011 NNA08BC21C • NASA Architecture NRA



Openalytics SOA





Integrated Demonstration

- Historical database with 25,000 flights
- Openalytics:
 - GlassFish server
 - JSF Web GUI
- 100Gb db4o database
- Model trained in a few minutes
- Seeded faults detected well

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Outline

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- Extensions



Verification for FOQA Data

- NASA has FOQA data sets
 - Could implement regression-based algorithms
 - But the data access is restricted...
- Software integration and verificiation
 - Have to be done off site
 - Require testing with realistic data
- Solution: simulator for FOQA data



Simulator scope

- Generate realistic FOQA data sets
- Base aerodynamic configuration

 Cruise, end of climb, beginning of descent
- Quasi-steady flight
 - Smooth accelerations, decelerations, turns
- Linearized performance models
 Of the aircraft and of the engines
- Models are calibrated on real data



Airframe Dynamics

• Airframe dynamics: near-steady flight

 $m(a-g) = F_{aero} + F_{thrust}$

• Propulsion thrust

 $F_{thrust} = c_{e,L} \rho_{air} N_{1,L} + c_{e,R} \rho_{air} N_{1,R} - c_{e,M} M$

 $-N_{1,L}$, $N_{1,R}$ are engine fan RPMs, M is Mach number

• Aerodynamic forces

$$F_{aero} = qC_{a,0} + qC_{a,1}\alpha + qC_{a,2}u_1 + \dots + qC_{a,n+1}u_n$$

- $q = \frac{1}{2}\rho_{air}V^2$ is dynamic pressure
- α is AOA

 $-u_1, \ldots, u_n$ are control surface positions



Airframe Attitude

• Pitch dynamics

 $Ia_{pitch} = C_{p,m}(m - m_0) + qC_{p,1}\alpha + qC_{p,0} + qC_{p,2}u_{stab} + qC_{p,2}u_{elev}$

- Roll dynamics
 - $Ia_{roll} = qC_{r,1}r_{roll} + qC_{r,2}u_{aileron} + qC_{r,2}u_{rudder}$
- Yaw: coordinated turn



Actuator Allocation

- Symmetric allocation
 - Left elevator = Right elevator
 - Left aileron = -Right aileron
- Stabilizer = $c \cdot \text{Elevator}$
- Slats, flaps, spoilers stats
 - are not deployed





Regression Model

- Linear regression models
 - $y_j = B_j x_j + v_j$
 - $-y_j$ performance variables
 - $-x_i$ regressors
 - $-B_{j}$ regression parameters
 - $-v_i$ noise
- LS model fit (training)
 For one channel at a time



A319 Data Set

- The model is trained on A319 FOQA data set at NASA ARC
 - The data provided by an airline partner to NASA under a confidentiality agreement
 - The model completely depersonalizes flight data
 - Modeling of aircraft performance has no relation to airline operation.



Airframe Regression Model

scaled data

	RHO X N1 SUM	DELTA M	DELTA M X A- LAT	DELTA M X LONG	DELTA M X NORM	DELTA M X ROLL	AOA X PDYN	ROLL RATE X PDYN	DIFF AIL X PDYN	SUM ELEV X PDYN	STABILI ZER X PDYN	RUDDE R X PDYN	DYN PRESS
LAT ACCEL			-1.0000 ± .0000						-0.0483 ± .0133			-0.1023 ± .0282	-0.0386 ± .0348
LONG ACCEL	0.3927 ± .0258			-1.0000 ± .0000									-0.1854 ± .0328
NORM ACCEL					-1.0000 ± .0000		3.0586 ± .3441						2.1192 ± .5006
ROLL ACCEL						-0.2000 ± .0000		-0.0032 ± .0007	-0.0061 ± .0013			0.0018 ± .0008	0.0003 ± .0007
AOA X PDYN		0.8926 ± .1133								-1.0604 ± .1954	-0.7143 ± .0980		0.7017 ± .1021

Flight		STABILIZER X PDYN	RUDDER X PDYN	DYN PRESS	ONES
	AILR SUM				-0.5152 ± 0.1467
Surrace	N1 DIFF				0.0000 ± 0.0001
Allocation	ELEV DIFF				-0.0090 ± 0.0741
Allocation	SUM ELEV x PDYN	0.1403 ± 0.0476		-0.1209 ± 0.0835	

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Engine Regression Model

• Linearized performance model of the engine

scaled data

	AOA	N1	TOTAL AIR TEMP	VARB BLD VALVE	VARB STATR VANE	МАСН	TOTAL AIR PRESS	ONES
N2	0.0332 ± 0.0023	0.2841 ± 0.0029	0.0163 ± 0.0011			0.0258 ± 0.0023	0.0109 ± 0.0011	0.3053 ± 0.0032
EGT LPT	-0.0083 ±	1.4024 ±	0.3544 ±	-0.0505 ±	-0.3537 ±	0.0687 ±	-0.0184 ±	-1.2769 ±
TEMP(IAE)	0.0359	0.1641	0.0161	0.0074	0.0220	0.0503	0.0189	0.1388
T3 (HPC EXIT	-0.1221 ±	1.2868 ±	0.4152 ±	-0.0261 ±	-0.4742 ±	0.0852 ±	-0.0166 ±	-1.1573 ±
TEMP)	0.0289	0.1330	0.0114	0.0056	0.0201	0.0334	0.0135	0.0987
FUEL MASS	-0.6324 ±	1.8989 ±	0.1574 ±	-0.0586 ±	-0.3534 ±	-0.5733 ±	0.1633 ±	-0.8648 ± 0.1450
FLOW RATE	0.0507	0.1916	0.0259	0.0084	0.0241	0.0829	0.0271	



Regression Fit: Fuel Rate, EGT





Simulator Development

- Coded in Matlab
 - Almost completed
 - Stand-alone executable
- Uses the regression models
- Will generate most/all FOQA data channels



Flight Plan



– Turn: 3-4

- Turn rate \rightarrow roll



Overall Simulator Logic

- Flight Plan
 - → Accelerations (Kinematics)
 - → Airframe Attitude (Coordinated Flight)
 - → Flight Actuators (Regression)
 - → Engine Dynamics (Regression)
 - → Mass Change (Integration)
- Triangular model structure



Simulator Verification

• Accelerations

- From flight plan (kinematics)

• Airframe attitude

- Coordinated flight

- Flight actuators, engine dynamics
 - Regression models
 - Were shown earlier



Accelerations





Airframe Attitude



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Simulator Performance

- Most/all FOQA data channels
- 10-20% accurate in its range
- Generates 5000 s of data in 1 s on a PC
 - -40Mb csv file
 - Writing to disk adds time



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Fleet Data Monitoring

- Many flights of many aircraft
- Fixed effects
 - Tail-to-tail model variation
 - Flight-to-flight variation, e.g., loading
- Eric Chu's talk on Wednesday



Distributed Data Mining

- New NASA SBIR project
- Aircraft fleet data
 - Distributed data sets
 - Fixed effect models
 - Distributed computing
- Model-based monitoring
 - Train model
 - Use it for monitoring





Conclusions

- Formulated technology transition path for ground monitoring of FOQA data
 - Model training (data mining)
 - Monitoring (data exploitation)
- Demonstrated SOA software framework
- Working on distributed computing software for scalable aircraft fleet monitoring



Backup Slides