AN INNOVATIVE FLEET CONDITION MONITORING CONCEPT FOR A 2MW GAS TURBINE

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Introduction



Development of GTPtracker[©] innovative gas turbine condition monitoring technology
 Advanced modelling and simulation methods
 Benefit from latest available ICT technology



Innovative & cost effective OP16 2 MW class gas turbine
 Compact, simple and robust design for high reliability
 Advanced technology for high fuel flexibility, efficiency and low emissions

Collaborative effort to demonstrate GTPtracker on OPRA OP16 engine fleet

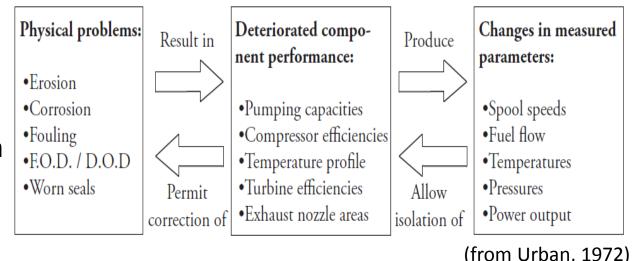


Gas turbine condition monitoring overview

- Gas turbine condition monitoring
 - Performance analysis (gas path)
 - $_{\circ}~$ Vibration analysis
 - Lubrication system

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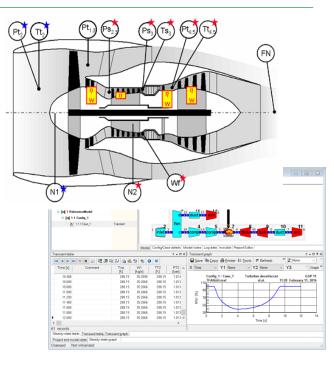
- Physical degradation ↔ Measurements
- System condition ↔ Component condition
- Diagnostics & Prognostics
 - Maintenance decision support
 - Optimize RAM (Reliability Availability, Maintenance)
 - Minimize LCC (Life Cycle Costs)



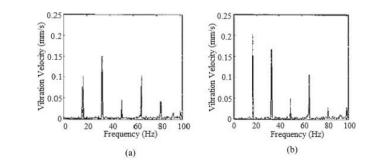


Condition monitoring challenges

- Modelling effects of faults & deterioration
 - Translating measurements into component condition information
 - Physical models
 - No need for operation history
 - Only cover known mechanisms
 - Empirical approach
 - Need history including faults and deterioration, or models
 - Genetic algorithms, Neural Networks, 'Machine learning' etc.
 - Valuable to identify unknown correlations among effects
- On-line condition monitoring
 - Requires *real-time* running of analysis models
 - Limits on model complexity
 - $_{\circ}~$ Find ways to simplify models
 - Minimize reduction of fidelity essential for fault & deterioration analysis









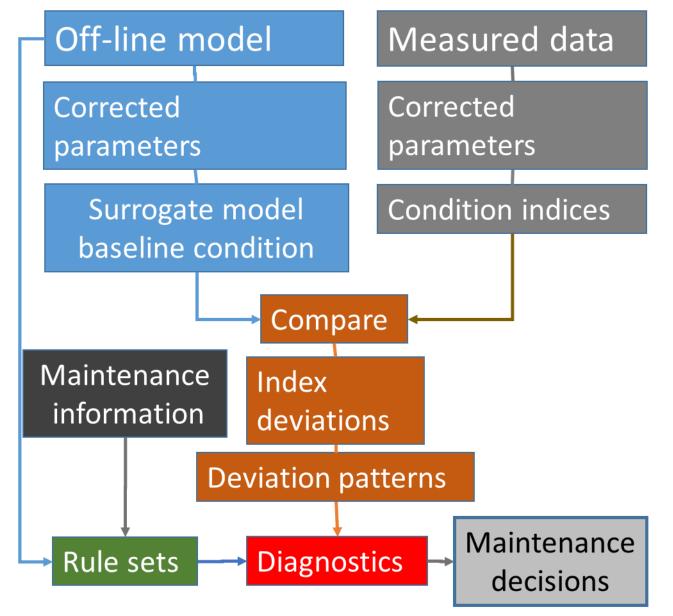
GTPtracker[©] key elements

Models relating measurements with condition

- Off-line comprehensive detailed models (GSP cycle model)
- On-line surrogate models (derived using off-line model)
- Condition indices
 - $_{\odot}\,$ Indicating % deviation from base line
- Rulesets
 - $_{\odot}\,$ Relating specific faults and deterioration to condition index patterns
- Diagnostics
 - Evaluating rulesets on performance snap shots (real-time)
- Prognostics
 - Predicting ruleset matches
- > Relating rulesets to maintenance actions



GTPtracker methodology – work flow





Surrogate Model derivation

Example: Base line Power – EGT (T45) relation

Reduced Power

$$PWc = \frac{PW}{\partial^a \ \theta^b} + c.\,\delta + d.\,\theta + e$$

x10³

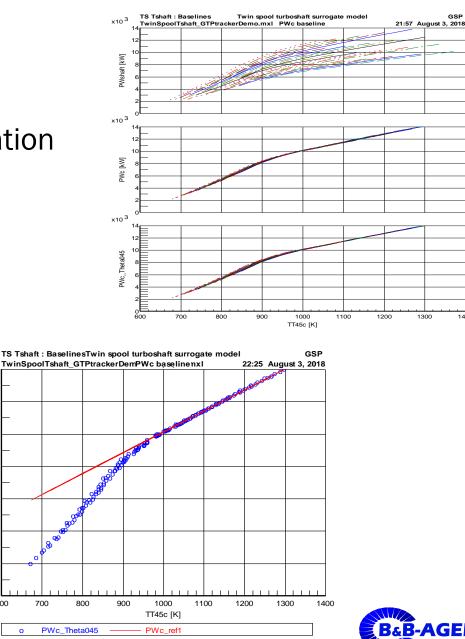
600

$$TT_{45c} = \frac{TT45}{\theta}$$

Use model to find coefficients a..e

• Regression

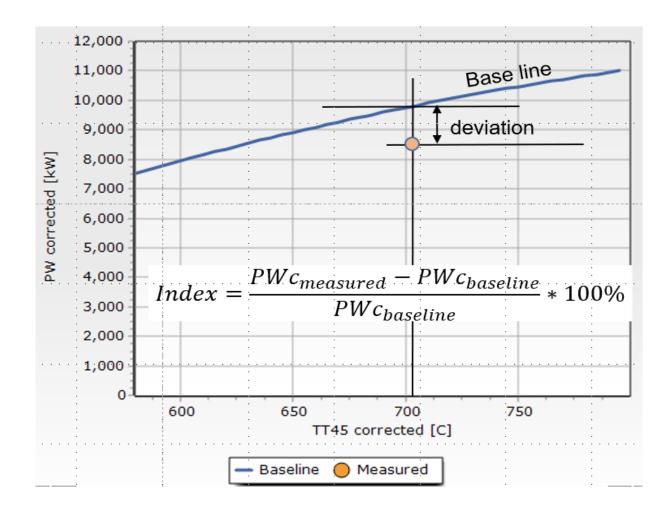
$$PWc_{ref} = 9900 + (TT45c - 980) \cdot 12.9$$





1400

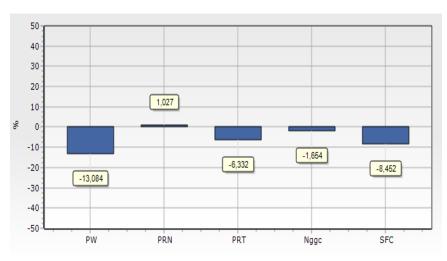
Condition Indices





Rulesets for diagnostics

- Patterns of deviations from baseline model
 - Characteristic for specific faults and engine health problems
 - $_{\circ}~$ Combinations of 'above-below' rules
 - A match of a ruleset with an operating point indicates a specific problem
 - Optimally isolate single root cause of problem
 - Prefer parameters independent of inlet conditions (indices)
 - Link to specific maintenance actions
- Generate rulesets using detailed off-line model
 - $_{\odot}\,$ Simulation of faults and deterioration modes
 - Include uncertainly (Monte Carlo simulations)
- Unknown problems & deterioration modes
 - Develop rulesets based on experience
 - Apply AI (GA, ANN, machine learning etc.)

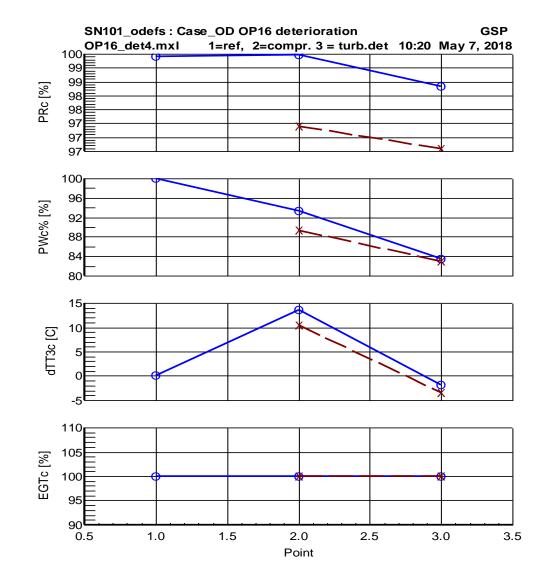


Rulesets for configuration: OP16 Config 2						
Name Description		ption	Component		Maintenance ty	
- Compressoi	Gen.Detei Compre		ssor water wash		ash	
Fieldname	Fieldname Display lab PW index [%] PW index		Display label		Below	Above
PW index [9				[%]	-2.00	
PR index [%	5]	PR index		[%]		-3.00
TT3_2 [°C]		TT3_2		[°C]		280
HE 4 1	н +	$+_{i} = +$	~ ×	a		
- Turbine det			Turbine		Inspection 1	
Fieldname	FieldnameDisplay labPW index [%]PW indexPR index [%]PR indexTT3_2 [°C]TT3_2		bel	Unit	Below	Above
PW index [9				[%]	-3.00	
PR index [%				[%]	-3.00	
TT3_2 [°C]				[°C]	280	
HE 4 1	н +	$+_2 = +$	~ ×	a		



Simple ruleset generation example

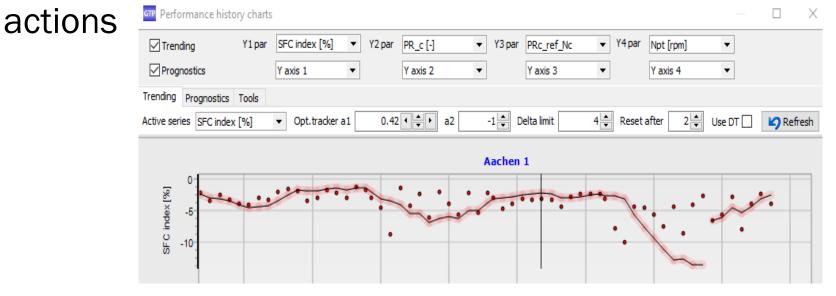
 Isolation of compressor vs. turbine deterioration





Trending

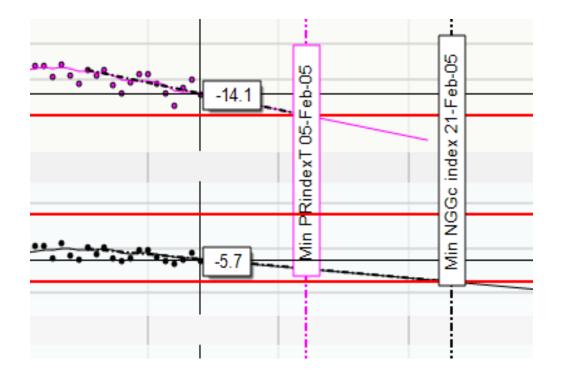
- Time series analysis methods
- Optimal tracker Kalman filter
 - see 'Everything Works Wonderfully' by Mike Provost
 - User configured (factors per parameter)
 - Automatic elimination of outliers
 - $_{\odot}$ Automatic reset at discontinuities and known maintenance





Prognostics

- Extrapolation of recent trends
 - $_{\circ}$ If correlation coefficient > minimum
 - Perform ruleset diagnosis at intersection with parameter limit
 - Automatic suggestion of maintenance action (maintenance calendar)
 - $_{\odot}$ Continuous real time analysis



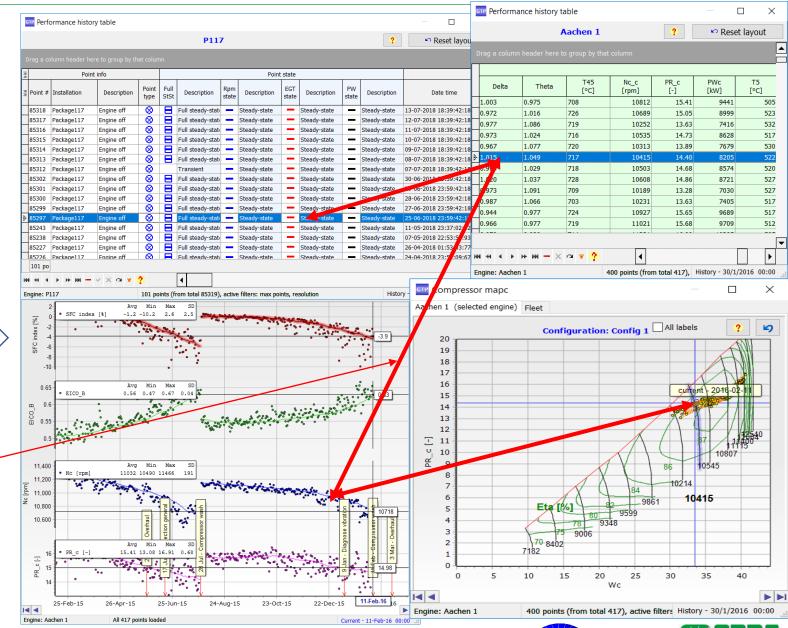
💷 Maintenance calendar					
P098 (selected engine) Fleet					
Drag a column header here to group by that column					
∃ Туре	Description / reason	Date tim 🔻	EOI CC	Status	
Inspection 8500	Inspection after 8500 EOH	30-10-2018	463 153	Scheduled	
water wash	Compressor water wash	13-08-2018	463 153	Suggested	
Replace GT air inlet fine filter	Exceeding of dPfilter_index	22-05-2018		Predicted	
	× ¤ 7 ?			Þ	



Analysis tools

- Performance history tables

 Point type & state
- Filtering data on point type, state, date and more
- End user configurable sets of analysis graphs
 - Time series, X-Y
 - Baselines, component performance maps
- Cursor synchronized in all tables and graphs on the selected operating point
- On-line / real time refresh adding new points





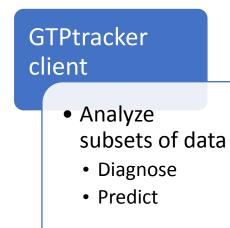
Data processing

Asset data

- Measured data
 - DA system
 - Historian
 - Manual import source
- Selection of parameters
 - application specific
 - Performance
 - Vibration
 - oil system
 - status flags
- Required for
 - trending
 - diagnostics
 - prognostics

GTPtracker Import Service

- Continuous import
 - On-line
 - Real-time
- Filtering data
 - Only useful time series data
 - User specified filter
- Data reduction ratio
 - Operation type
- Assignment of
 - Point type
 - Point states



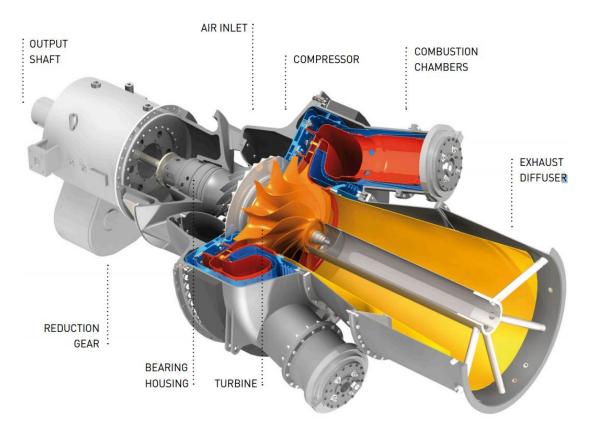
- Select subsets
 - Query Filter
- Filter on
- Time, EOH, cycles
- Point type
- Point state
- Resolution



The OP16 gas turbine

- Power generation gas turbine, 1.85 MW
- Compact, single-shaft, all-radial rotor
- Single stage centrifugal compressor, 6.7:1
- · Four combustor cans mounted in reverse-

flow direction





Application of GTPtracker to OP16 gas turbine fleet

- Corrected parameters are calculated in real-time by the surrogate OP16 performance model embedded in GTPtracker.
 - EGT, Power, PR, Thermal efficiency
- Performance indices, derived from the corrected parameters, are continuously trended, along with other measured/calculated parameters.
- Based on detailed off-line OP16 performance model and OPRA's experience, rule sets are developed relating deviation patterns to faults.
- Usually, the base load data is filtered out to accurately trend and analyze the measured and calculated parameters.

• Case studies on available field performance history



Case study 1: Detecting faulty fuel flow measurement

 Constant power and EGT index trends and a deviation in thermal efficiency index characterize a faulty fuel flow measurement.

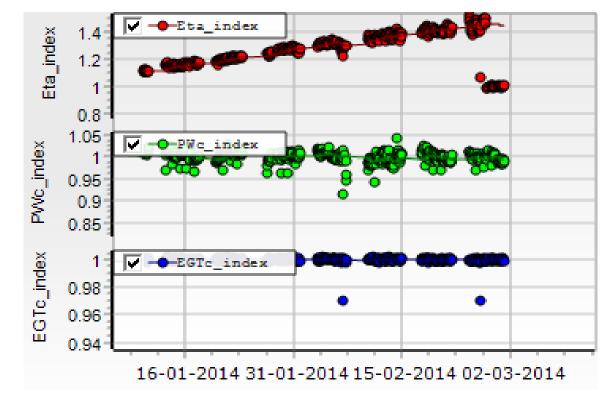
• Specify corresponding pattern in a ruleset :

Configuration of engine type: OP16 Constants Baselines / maps Rulesets						
Nar	Name Description			Component Maintenance type		
	Fuel flow sensor	Fuel flow sensor o	lrift	Fuel flow sensor reset		
	Fieldname	Display label	Unit	Below	Above	
	PWc_index [-]	PWc_index	[-]	1.0	50	0.950
	EGTc_index [-]	EGTc_index	[-]	1.00	05	0.995
	Eta_index [-]	Eta_index	[-]	0.8	00	1.200
	HE ← ► ₩ +	$+_2=+\vee\times$	a			►



Case study 1: Detecting faulty fuel flow measurement

 Trend of performance indices: Thermal efficiency, Power, Exhaust gas temperature





- Ruleset match detected at 16-1-2014
- Repaired 1-3-2014

Case study 1: Detecting faulty fuel flow measurement

• Upon detection, the maintenance calendar is automatically updated, adding the maintenance action of sensor replacement.

- Maintenance decision
 - Automatic initial status
 - To be confirmed/changed by operator

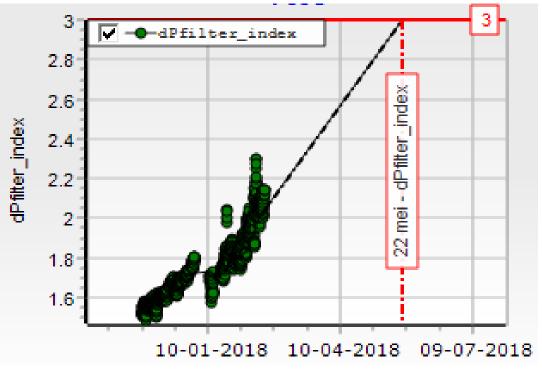
Scheduled maintenance

Maintenance calendar						
P181 (selected engine) Fl	eet					
Drag a column header here to group by that column						
∃ Туре	Description / reason	Date time				
Fuel flow sensor reset	Drifting fuel flow measurement	16-01-2014 18:54:57:000				
$H + F H + F - F \land X \land F ? $						



Case study 2: Predicting filter clogging

 Rate of increase of differential pressure across a filter determines the rate of clogging.



 Simple 1 parameter limit on level and/or rate of change



Conclusions

- An innovative online condition monitoring system has been developed for the OPRA OP16 gas turbine using the GTPtracker monitoring and tracking tool.
- The connection of the condition monitoring process with accurate cycle models capable of simulating deterioration via a surrogate models and rulesets for diagnostics offers an optimal compromise between complexity and functionality.
- The GTPtracker environment and configuration user interface provides a powerful tool for diagnostics engineers to optimize maintenance (minimize costs), reliability, availability and safety for a gas turbine fleet.
- A customized version of the GTPtracker tool has recently been deployed for the OP16 engine.
- GTPtracker can rapidly be deployed and coupled to gas turbine data acquisition systems



Questions



