

Forecasting the condition of high-pressure turbine parts by using Bayesian Belief Networks

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Content

- 1. Background
- 2. Aims/objectives & methodology
- 3. Construction of a Bayesian Belief Network (BBN)
- 4. Verification & evaluation
- 5. Conclusions



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Background Bayesian Belief Net (BBN)

Why BBN?

Functionality:

- shop visits (SV) determined by various influence parameter which are not analytical acquired
- modifications and further developments shall require little time and effort
- implementing expert knowledge
- manageable in case of imprecise data

Percentage of the AFR repaired components of the three main regions towards the cycles

directed acyclic graphs ine - Outsid conditional propability / Bayes' theorem: $P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$ quantitative side: conditional probability tables qualitative side: net architecture Asia (China Eastern Europe (F 1st FR 2nd FR 3rd FR 4th FR AFR 1.94 0.12 0.20 Aiddle Eas [Dreamstime.com] Illustration of BBN Technische Universität 11th October 2019 | Forecasting the Condition of HPT Parts by using BBN | **IFAS** Institut für Flugantriebe und Strömungsmaschinen Braunschweig 11. Dresdner Probabilistik-Workshop Slide 3

Background Why using BBN for engine maintenance?



\Rightarrow precise hardware forecast is badly required



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Slide 4

Degradation mechanisms High-pressure turbine NGVs and rotor blades

- turbine erosion
- turbine fouling
- hot gas corrosion
- creep & fatigue
- abrasion
- compressor erosion
- compressor fouling
- foreign object damage



Fouling on HPT stator blades, [2]





Erosion on HPT rotor blades, [5]



Hot gas corrosion on HPT blades, [6]

Fouling on HPC blades



Erosion on HPC rotor blades, [5]





Fan damage



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Slide 5

Degradation mechanisms Input data

Last Repair Repair Code Region Repair Code Rotor 1. Dust and sand concentration worldwide [Naval Research Lab] Cycle since last SV Flight profile Material Blade material [8] Rating Level Data Entry Plug Data entry plug [9] Wing Position Reverse thrust operation [9] Aircraft Utilisation Sand ingestion [10] Technische Universität 11th October 2019 | Forecasting the Condition of HPT Parts by using BBN | **IFAS** Institut für Flugantriebe und Strömungsmaschinen Braunschweig

Slide 6

Aim and objectives

<u>Aim:</u>

- developing a method for forecasting the repair of the HPT NGV
- showing the potential benefit in case of inadequate data density





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Framework conditions Learning data set

Boundary conditions:

- piece-part repair level
- no unscheduled SV
- engines with recent modification levels
- complete data set
- ~4500 NGV 1 data sets or ~195 jet engines
- ~5800 rotor 1 data sets or ~72 jet engines
- ~8000 NGV 2 data sets or ~333 jet engines
- ~5600 rotor 2 data sets or ~76 jet engines





Construction of a BBN Standard BBN (NGV 1)



The standard net

- simplest net architecture
- equal weighting of parameters



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Slide 9



Construction of a BBN Quadratic convergence

Cramer's contingency coefficient:

• detecting correlation for net architecture (qualitative side)

$$C = \sqrt{\frac{\phi^2}{\min\{k - 1, m - 1\}}}, \qquad 0 \le C \le 1$$

 ϕ^2 : Mean value of quadratic convergence

Maximum value of k: row m: column

- Min./max. values:
 - 1. $C = 0 \rightarrow$ statistical independence
 - 2. $C = 1 \rightarrow$ ideal statistical dependence
- $C \ge 0.5$: strong correlation between two parameters

Results (NGV 1):

Strong correlation between region and...

- ... rating level (C = 0.84)
- ... customer segment (C = 0.64)
- ...material (C = 0.57)

Strong correlation between material and wing position (C = 0.48)



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Construction of a BBN Expert knowledge

Framework:

- in average almost 20 years of experience per person
- up to almost 30 years of experience for one person
- rank the main influences which affect the next repair

Outcome:

- similar order of degradation effects and dominant and environmental conditions
- recent data (years 2011-2014) most suitable
- proposed cycle distribution





Construction of a BBN Final net architecture – modified BBN (NGV 1)



Evaluation – 1 Trend forecasting test cases

- up to eleven test cases generated during expert interview
- specifically for each component
- evaluation by scoring system



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Case	Evidences	Expected	Standard net	Modified net	Legend: 🗸	good forecast quality
1	2nd FR	~ 100% AFR	\checkmark	\checkmark	0	medium forecast quality
2	Middle East, N5, High	~ 100% AFR	\checkmark	\checkmark	Х	poor forecast quality
3	North America, 2nd FR, X-40	~ 100% AFR	\checkmark	\checkmark		
4	Middle East, Cycles ↑	AFR ↑	0	0		
5	North America, Cycles ↑	AFR ↑	0	\checkmark		
6	Cycles ↑	$AFR_{Outside} > AFR_{Inside}$	Х	Х		
7	Cycles>3000, DSR'142, High	~ 100% AFR	\checkmark	\checkmark		
8	North America, Cycles>2500	70% AFR, 30% FR	Х	\checkmark		
9	High, AT/FA	75% AFR, 25% FR	Х	\checkmark		
10	New	10% Serviceable	0	0		
11	AFR	Serviceable ↓, AFR ↑	0	0		

Trend forecasting test cases (NGV 1)



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Evaluation Results (NGV 1)

Evaluation data set:

- percentage distribution correspond to training data
- limited data used
- 41 test engines:
 - 1. 29 **current** engines from August 2012 till January 2013
 - 2. 2 unusual engines
 - 3. 10 representative engines from 2010 till 2012

Jet engines	Standard BBN	Modified BBN
All	77%	83%
Current	80%	85%
Unusual	50%	50%
Representative	73%	85%

List of high accuracy per net

 \Rightarrow With up to 83% the repair of the HPT NGV has been correctly forecasted!



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Conclusion

 potential adaptability of the quadratic convergence test and Cramer's coefficient for other hardware components has been shown



- remarks to improve the forecasting quality:
 - 1. available data (less learning data available for rotor blades)
 - 2. identification of relevant input parameters
 - 3. setting appropriate boundary conditions

Outlook:

- very promising potential for contract proposals and capacity planning
 - \Rightarrow determining the business value
 - \Rightarrow estimating the statistical variation
- investigating other modules / engine types, also future programmes



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Thanks for Listening! Corresponding author's email: d.giesecke@ifas.tu-bs.de



Publications:

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- [1] Giesecke, D., Friedrichs, J., Kenull, T., Binner, M. and Siegert, M., 2014. "A Method for Forecasting the Condition of HPT NGV's by using Bayesian Belief Networks and a Statistical Approach", GT2014-25464.
- [2] Giesecke, D., Wehking, M., Friedrichs, J., Binner, M., "A Method for forecasting the condition of several HPT parts by using Bayesian Belief Networks", GT2015-43110.



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Literature

- [3] MTU Maintenance Hannover GmbH.
- [4] Rupp, O. C., 2000. "Vorhersage von Instandhaltungskosten bei der Auslegung ziviler Strahltriebwerke". PhD thesis, TU München.
- [5] Ebmeyer, C., Wensky, T., Friedrichs, J., and Zachau, U., 2011. "Evaluation of total engine performance degradation based on modular efficiencies". GT2011-45839.
- [6] Kurz, R., 2005. "Gas Turbine Performance". In: 34th Turbomachinery Symposium Proceedings, Turbomachinery Laboratory.
- [7] Meher-Homji, C. B., Chaker, M. A., and Motiwala, H. M., 2001. "Gas turbine performance deterioration". In: Proceedings of the 30th Turbomachinery Symposium.
- [8] Buergel, R., Maier, H. J., and Niendorf, T., 2011. Handbuch Hochtemperatur -Werkstofftechnik: Grundlagen, Werkstoffbeanspruchungen, Hochtemperaturlegierungen und -beschichtungen. Springer Vieweg.
- [9] Bräunling, W. J. G., Berlin, 2009. Flugzeugtriebwerke Grundlagen, Aero-Thermodynamik, ideale und reale Kreisprozesse, Thermische Turbomaschinen, Komponenten, Emissionen und Systeme, Springer-Verlag.
- [10]Walsh, W. S., Thole, K. A., and Joe, C., 2006. "Effects of sand ingestion on the blockage of film-cooling holes". GT2006-90067).
- [11]Friedrichs, J., 2019. "Bauelemente von Strahltriebwerken Funktion, Betrieb und Wartung", Vorlesungsskript TU Braunschweig.



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