

IMPLEMENTATION OF STOCHASTIC METHODS FOR INDUSTRIAL GAS TURBINE FAULT DIAGNOSIS

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Implementation Of Stochastic Methods For Industrial Gas Turbine Fault Diagnosis



§ Diagnostic Scheme with GPA
§Stochastic Methods Used
§Twin Shaft Industrial GT Implementation
§Signle Shaft Industrial GT Implementation
§Summary - Conclusions



§ Diagnostic Scheme with GPA

§Stochastic Methods Used

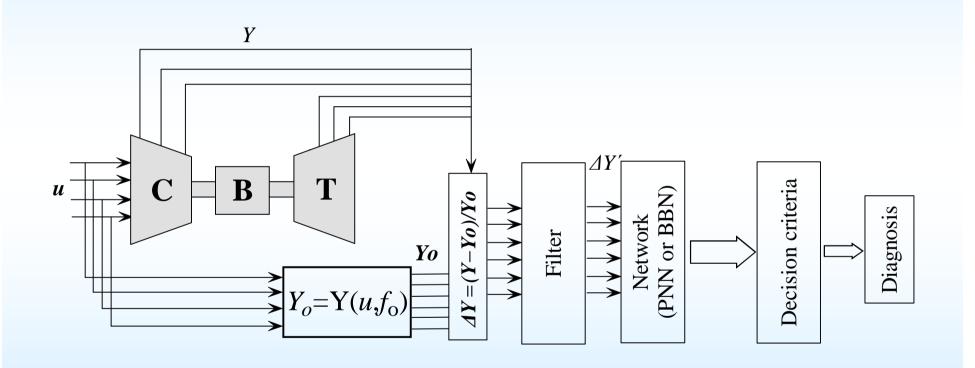
§Twin Shaft Industrial GT Implementation

§Single Shaft Industrial GT Implementation

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The Diagnostic Process



Implementation Of Stochastic Methods For Industrial Gas Turbine Fault Diagnosis



§ Diagnostic Scheme with GPA

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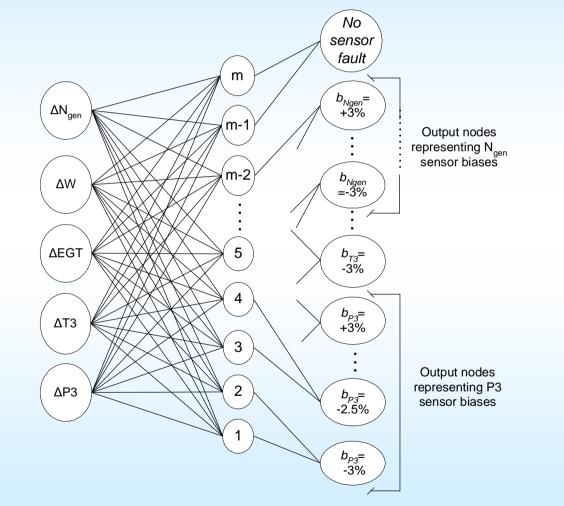
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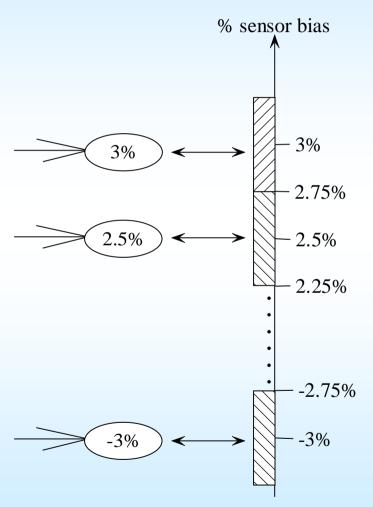


The method of PNN for sensor fault diagnosis



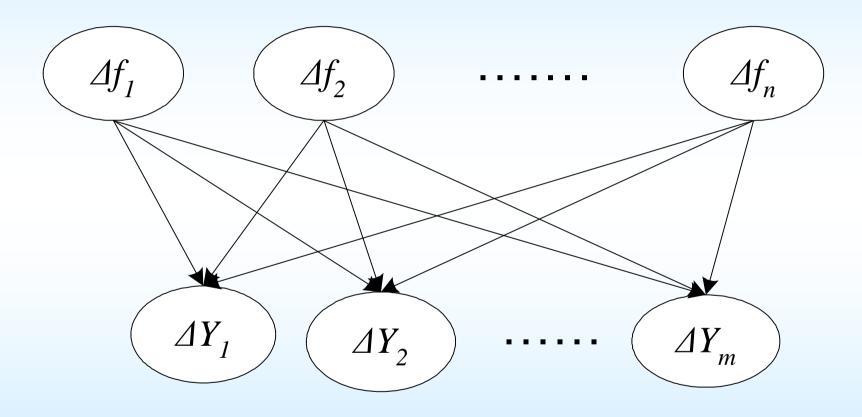


The method of PNN for sensor fault diagnosis





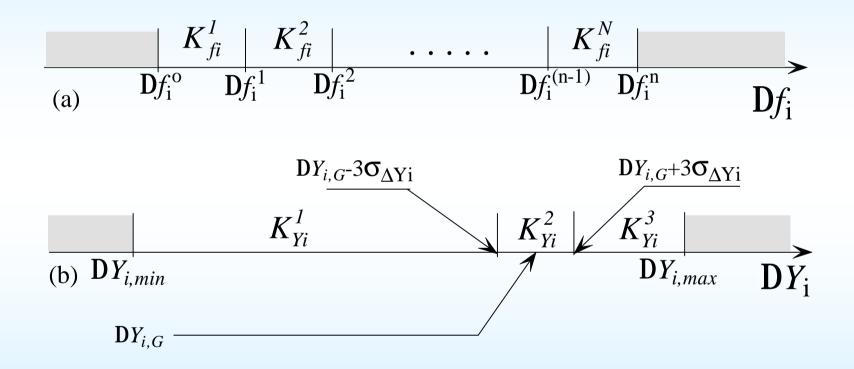
BBN for component faults diagnosis (I)



Architecture of BBN for Component Fault Diagnosis



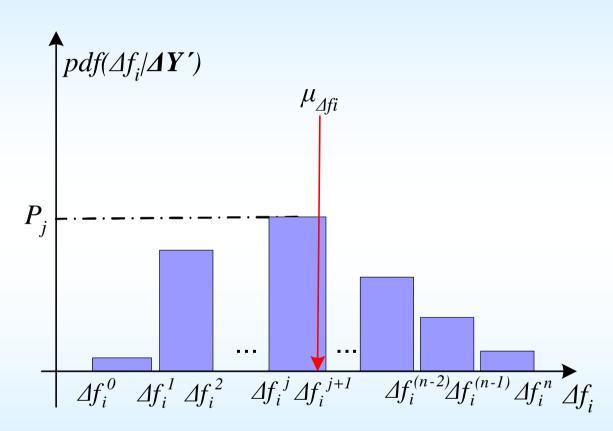
BBN for component faults diagnosis (II)



states of the nodes of the diagnostic BBN. (a. health parameter node, b. measurement node).



BBN for component faults diagnosis (III)



Probability density function of the fi health parameter



§ Diagnostic Scheme with GPA

§Stochastic Methods Used

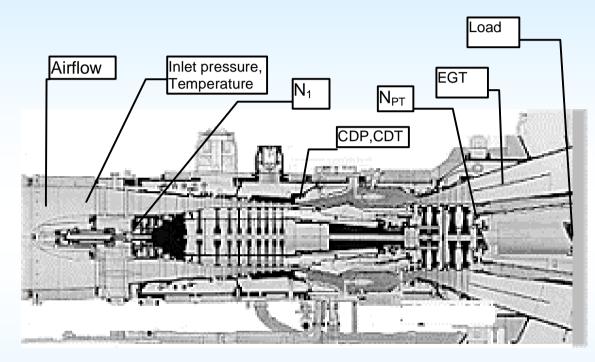
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Fault Diagnosis On A Twin Shaft Industrial Gas Turbine

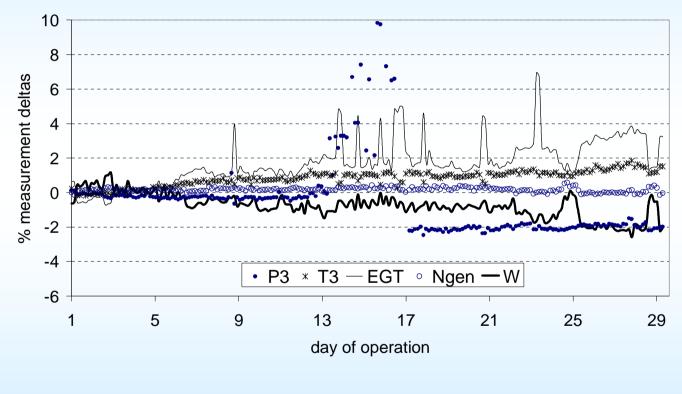


Gas turbine Layout and quantities measured for monitoring



Sensor Fault Diagnosis using PNN

Twin Shaft Industrial Gas Turbine

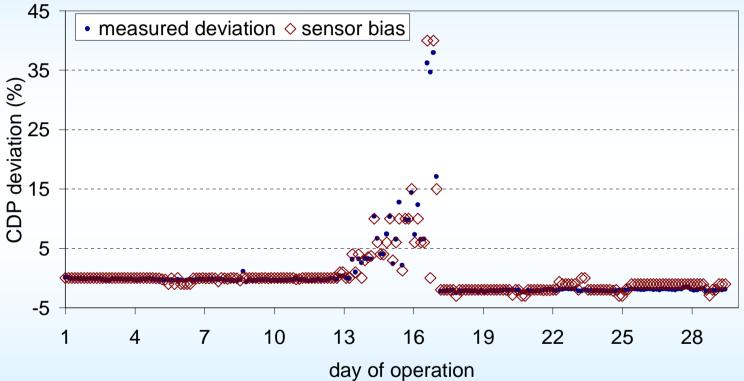


Evolution of Deltas



Sensor Fault Diagnosis using PNN

Twin Shaft Industrial Gas Turbine

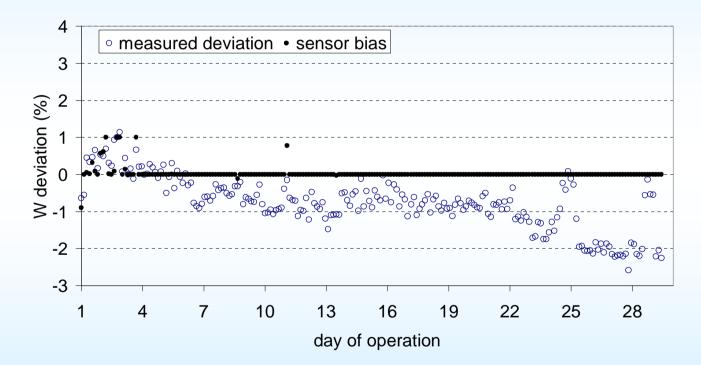


Estimated biases of P3 sensor and deviations of its measured values



Sensor Fault Diagnosis using PNN

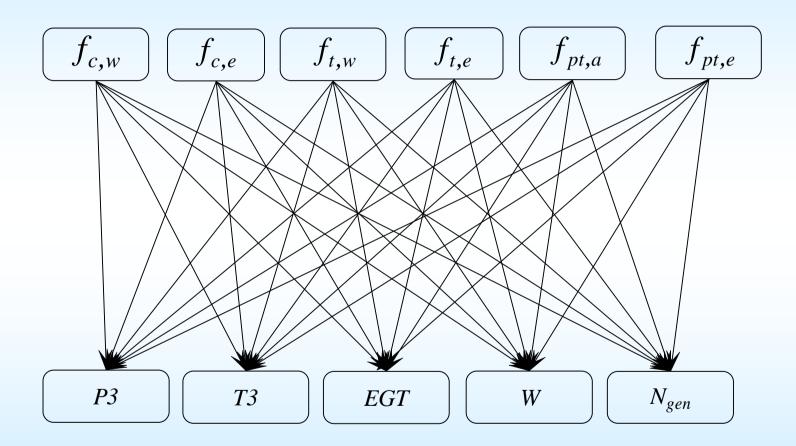
Twin Shaft Industrial Gas Turbine



Estimated biases of W and deviations of measured values



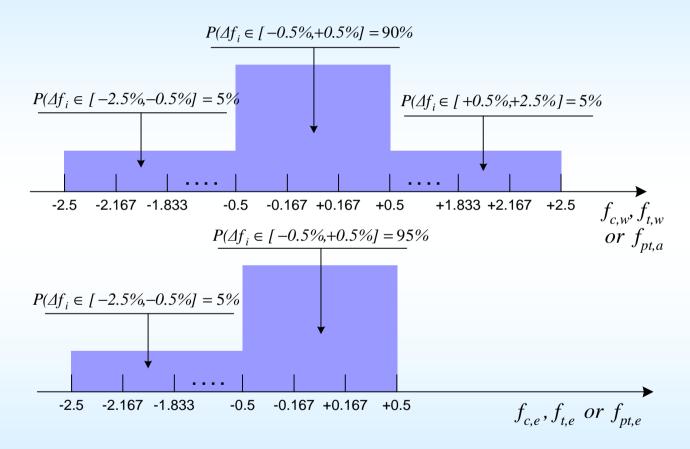
BBN for component fault diagnosis



Architecture of the BBN for component fault diagnosis on the twin shaft gas tur



BBN for component fault diagnosis

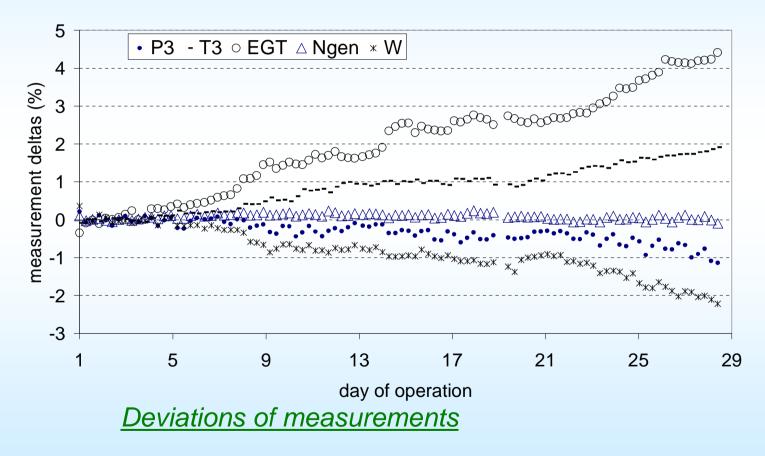


States (a) and the a-priori probability density functions (b) of the health parameter nodes



Component Fault Diagnosis using BBN

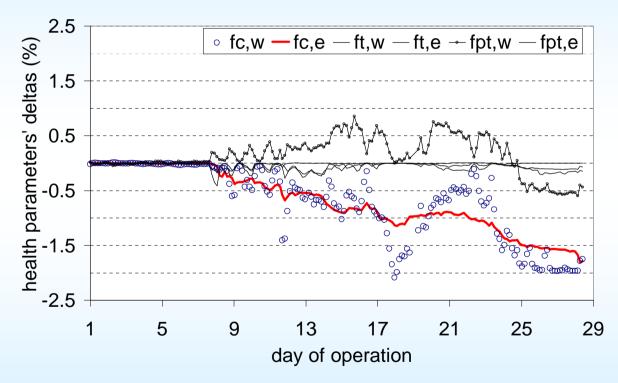
Twin Shaft Industrial Gas Turbine





Component Fault Diagnosis using BBN

Twin Shaft Industrial Gas Turbine



Diagnosis of a compressor fouling case



§ Diagnostic Scheme with GPA

§Stochastic Methods Used

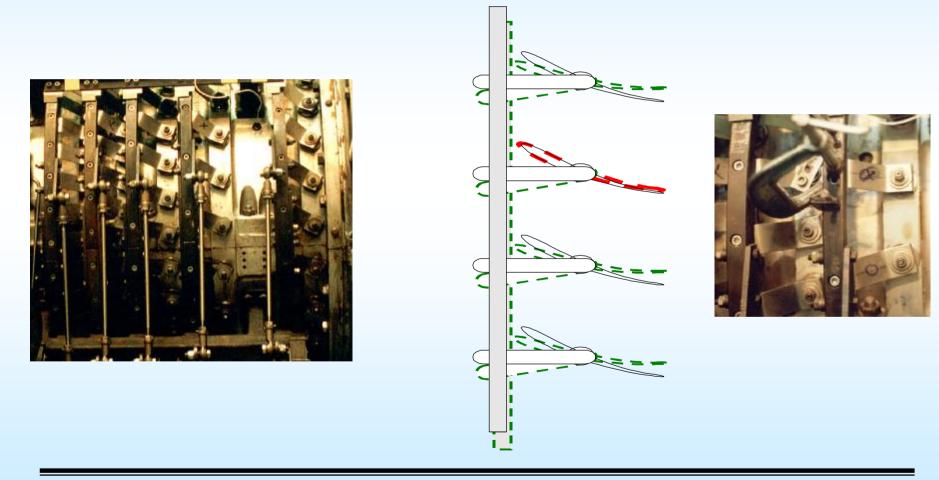
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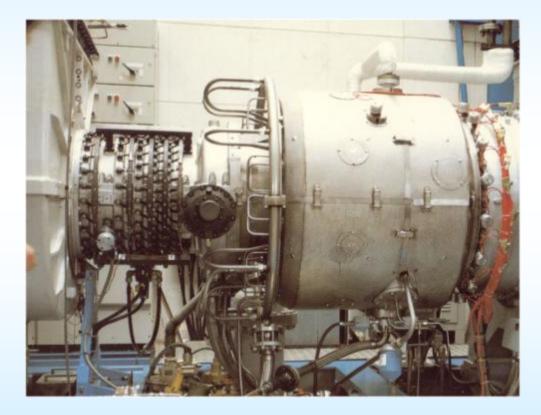


Fault Diagnosis On A Single Shaft Industrial Gas Turbine (I)



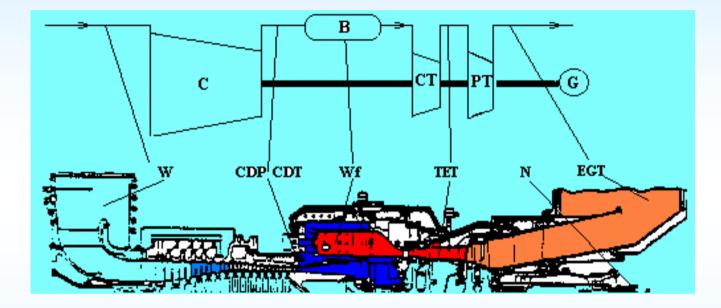


Fault Diagnosis On A Single Shaft Industrial Gas Turbine (II)





Fault Diagnosis On A Single Shaft Industrial Gas Turbine (III)

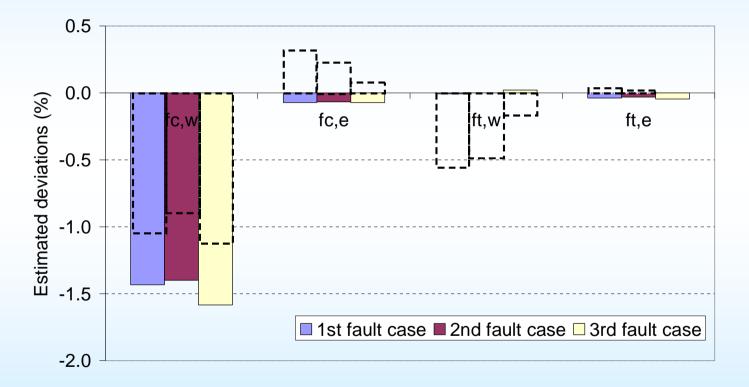


Engine Layout And Measured Quantities



Component Fault Diagnosis using BBN

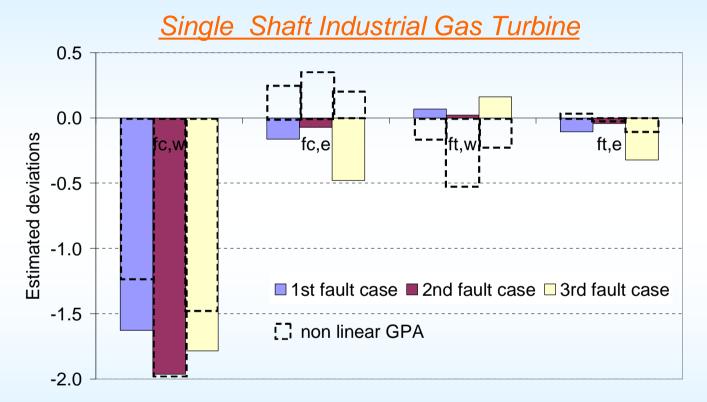
Single Shaft Industrial Gas Turbine



Estimated health parameter deviations for 1st stage mistuned VGV



Component Fault Diagnosis using BBN



Estimated health parameter deviations for three 1st stage mistuned VGV



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Summary - Conclusions

§ Possibilities offered by implementation of stochastic methods for fault diagnosis in industrial GTs have been demonstrated

§Sensor fault were successfully identified

§Component faults and gradual deterioration were also identified

§Methods of the type presented here can constitute a usefull constituent

of gas path diagnostic schemes