



# Tools for Sensor Performance Assessment



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Faculty:  
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## Objectives

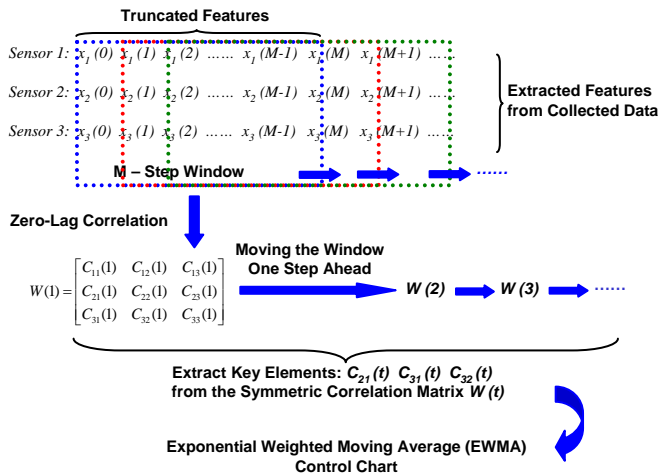
- To detect sensor progressive degradation and predict sensor failures and reconstruct the degrading sensor readings

## State-of-the-Art

- Direct hardware redundancy method:** use two or more sensors to measure the same key variable and assess sensor performance by comparing the consistency between different sensors, simple to implement but potentially costly
- Functional redundancy method:** use multiple sensors to measure related variables and validate sensor readings by employing sensor fusion techniques, more complex but cheaper approach

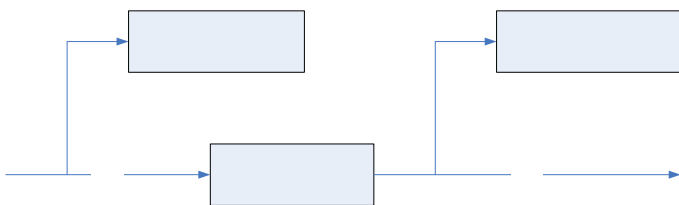
## Approaches

- Zero-lag correlation based direct hardware redundancy method



- Time-series modeling based functional redundancy method

### General Model



### Transfer Function

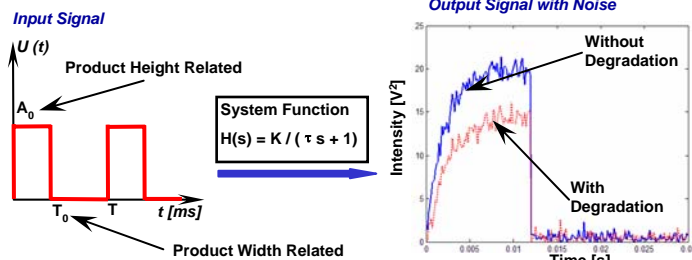
$$G(s) = \frac{S_1(s)}{S_2(s)} = \frac{G_0(s)G_{out}(s)}{G_{in}(s)} = \frac{A_0(s)A_{out}(s)B_{in}(s)}{B_0(s)B_{out}(s)A_{in}(s)}$$

## Methods

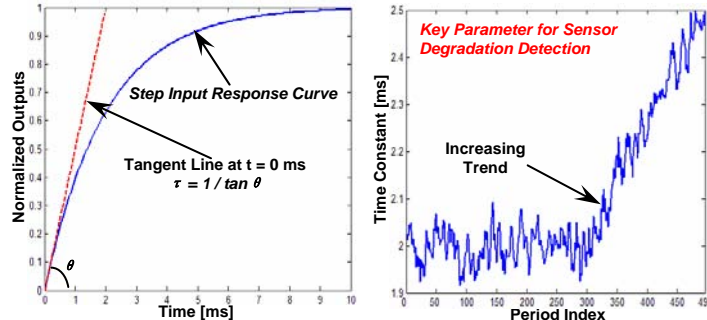
By fitting Auto-Regressive Moving Average Vector (ARMAV) time series model to the data within an extending window, the dynamics of the plant and input/output sensor can be identified by finding the zeros and poles of the transfer function.

## Case Study

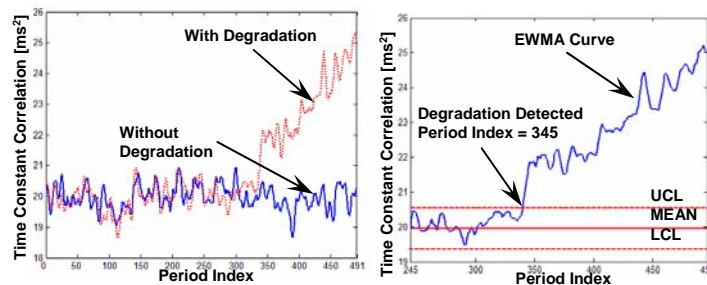
### Sensor with First Order Dynamics



### Feature Extraction



### Zero-lag Correlation Based Hardware Redundancy Method



## Future Work

- Evaluate both methods with real industrial data instead of simulation data
- Develop algorithm for sensor data reconstruction



# Detection of Intermittent Connections in An Industrial Network



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## Objectives

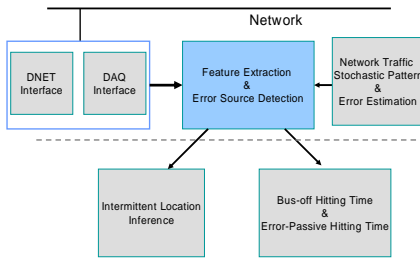
- Provide network performance and diagnostic information at the device level and network level.
- Develop novel network health monitoring tools for plant floor network systems.
- Minimize network diagnosing time due to intermittent connections on the network.

## State-of-the-Art

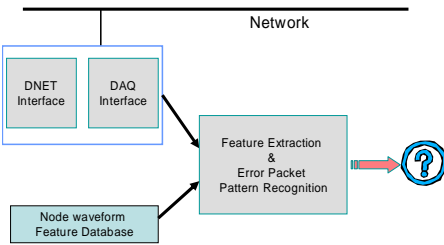
- Intermittent Connection is difficult to identify.
- No tool available for Intermittent Connection detection

## Approaches

- Network health monitoring framework

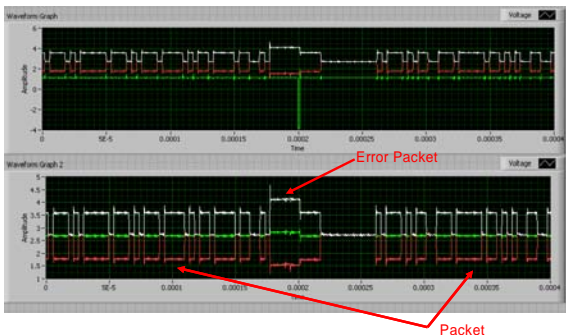


- Error Source Detection

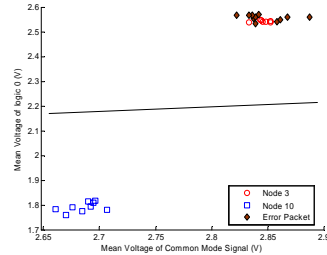


## Accomplishments

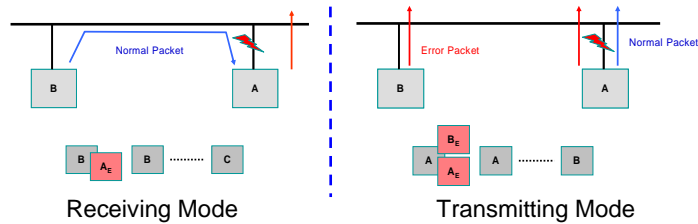
- Developed Error Source Detection System



- Classification Result of Error Source



- Identification of Error Location
  - Intermittent connection scenarios



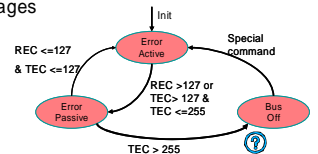
- Determination of the error location

- Assume the intermittent connection locate at the drop cable of a node
- If we observe two patterns happens coincidentally
  - Node A send error packet after normal packets
  - Error packets followed by packet sent by node A
- Then Node A has intermittent connection problem.

- Network Node Health Modeling

$$Q = \begin{bmatrix} 0 & 0 & 0 & \dots & 0 & 0 & \dots & 253 & 254 & 255 & 256 \\ 0 & -\lambda_0^k & 0 & 0 & \dots & \lambda_0^k & 0 & 0 & 0 & 0 & 0 \\ 1 & \lambda_0^k & -\lambda^k & 0 & \dots & 0 & \lambda_0^k & 0 & 0 & 0 & 0 \\ 2 & 0 & \lambda_0^k & -\lambda^k & \dots & 0 & 0 & 0 & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 254 & 0 & 0 & 0 & \dots & 0 & 0 & \lambda_0^k & -\lambda^k & 0 & \lambda_0^k \\ 255 & 0 & 0 & 0 & \dots & 0 & 0 & 0 & \lambda_0^k & -\lambda^k & \lambda_0^k \\ 256 & 0 & 0 & 0 & \dots & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

- $\lambda_0^k$  is the rate of transmission of non-corrupted messages for station k
- $\lambda^k$  is the rate of corrupted messages
- $\lambda^k = \lambda_0^k + \lambda_0^k$



## Future Work

- Conduct plant validations
- Construct intermittent connection model and estimate nodes bus-off time

## Sponsors

- NSF I/UCRC for Intelligent Maintenance Systems



# Cutting Process Monitoring in Turbine Blade Machining



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Faculty:  
*D. Djurdjanovic, J. Ni*

## Objectives

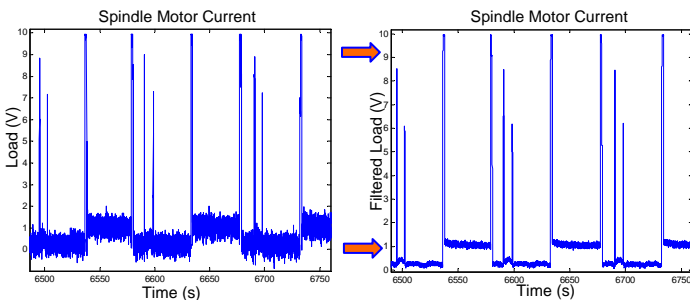
- Track 5-axis mill health under dynamically-varying, oblique loading conditions
- Improve tracking and prediction of process health for various machines through signal processing
- Propose a general method for tracking the health of a mill bit without adding sensors to the system
- Decrease the overall cost of milling compressor blades by using a condition-based method to change mill bits

## State-of-the-Art

- Use only the load data (amount of current) from the spindle and 5 servo motor drives to track mill bit health
- Explore correlation between servo motor current and cutting force in an oblique cutting
- Employ Time Frequency Analysis and other recently-developed methods to extract features from load data.

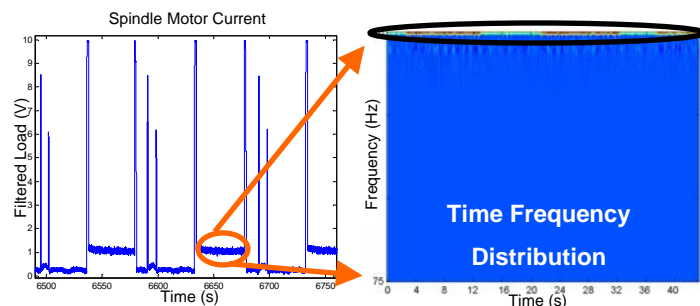
## Approach

- Filter raw load data using low-pass filter



Effect of filtering on data.

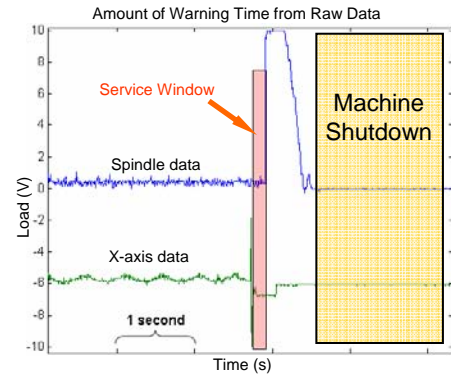
- Partition filtered data to separate milling passes based on varying levels in servo-motor current
- Extract features from filtered data using TFA



Extraction of features using Time Frequency Analysis

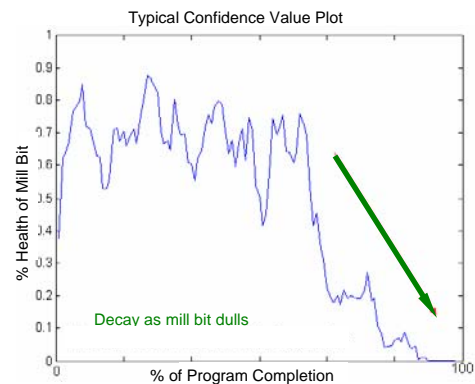
## Accomplishments

- Provided warning time (based on raw data) before unexpected shutdown happens



Warning of shutdown based on level of X-axis signal

- Found method for properly partitioning data
- Determined predictive model for health of mill bit
- Produced CV degradation trends for load data



Decay of mill bit health based on drift of process features

## Future Work

- Finish characterization of mill health
- Install sensors and begin collecting data on grinder
- Integrate and process data from multiple sensors with various frequency ranges for the grinder project

## Sponsors

- This research is supported by General Electric Aviation. My contact at GE is Roger Lindle.



# Immune System Engineering for Automotive Systems



Research Assistants/Staff:

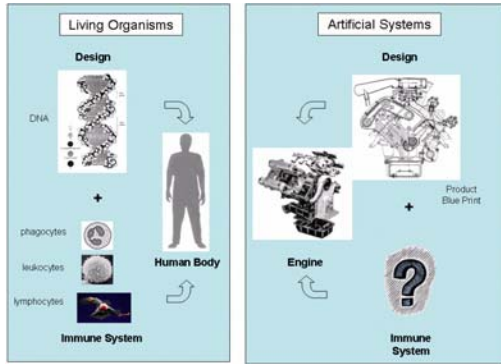
Shimin Duan  
D. Djurdjanovic

Faculty:

J. Ni

## Objectives

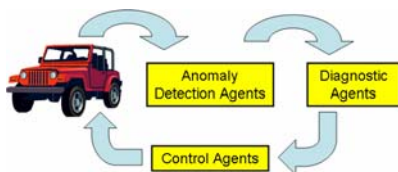
- Develop novel approach to realize “immune systems” functionalities in automated systems.
- Robustly detect abnormal behavior, isolate source, and compensate for negative effects to achieve desired performance in spite of the presence of an anomaly.



An anomaly effects automated system like a virus effects human

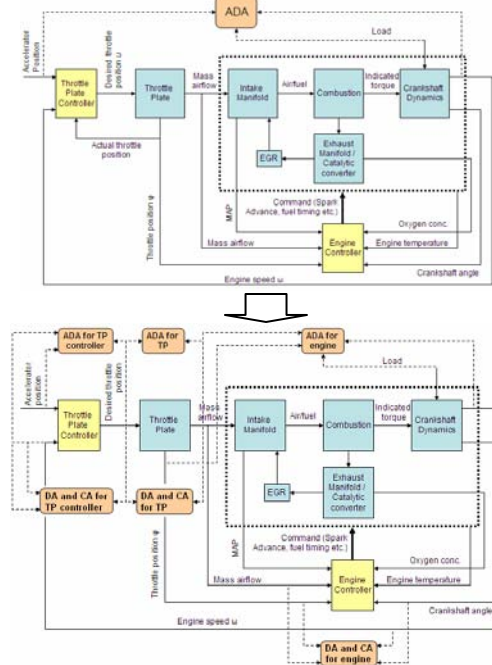
## Problem Statement

- The goal of the proposed research will be to incorporate the natural immune system functionalities and uniqueness into automotive engine system. There are three primary parts for accomplishing this target:
  - Identify the intrusion depicted in the degraded system
  - Isolate the intrusion source by reconfiguring, reconnecting and decomposing detection agents.
- Diagnostic Agents (DA-s)
  - Recognize and describe the anomaly source character if input/ output patterns have been observed in the past.
  - In case input/output patterns have not been seen in the past, create a new DA for recognizing the new anomaly next time.
- Control Agents (CA-s)
  - Utilize fault characterization from DAs to postulate control laws for restoring the performance of the anomalous subsystem as much as possible.



Proposed immune system operation for an automotive system

## Immune System for the Automotive Engine System



Multiplication of ADA-s, DA-s and CA-s in the electronic throttle mechanism

## Proposed Research Task and Time Table

| Tasks   | Month |      |       |       |       |       |
|---|-------|------|-------|-------|-------|-------|
|   | 0-6   | 7-12 | 13-18 | 19-24 | 25-30 | 31-36 |
| <b>Task 1: Methods for Anomaly Detection and Isolation</b>  | ←     |      |       |       | →     |       |
| Subtask 1.1: Multi-regime anomaly detection                 |       |      |       |       |       |       |
| Subtask 1.2: Isolation of the Anomaly Source                |       |      |       |       |       |       |
| Subtask 1.3: Practical implementation of ADA-s              |       |      |       |       |       |       |
| <b>Task 2: Methods for Diagnosis of Anomalous Condition</b> |       | ←    |       |       |       | →     |
| Subtask 2.1: Methods for realization of DA-s                |       |      |       |       |       |       |
| Subtask 2.2: Development of identification methods          |       |      |       |       |       |       |
| Subtask 2.3: Demonstration on an engine subsystem           |       |      |       |       |       |       |
| <b>Task 3: Methods for Control System Reconfiguration</b>   |       |      | ←     |       |       | →     |
| Subtask 3.1: Design of CA-s for recovering performance      |       |      |       |       |       |       |
| Subtask 3.2: Implementation on an engine subsystem          |       |      |       |       |       |       |

The research activity for over a three-year period

## Project Leadership and Management

- PI: Dr. Dragan Djurdjanovic, University of Michigan
- Co-PI: Prof. Jun Ni, University of Michigan  
Kenneth Marko, ETAS Inc