

Module 4 : Applications of STM is
Infrastructure Engg

- offshore structures
- coastal structures

Lecture 1 : Machine Learning is STM

What is machine learning?

- a task of generating knowledge from the past experiences
 - data base (sensor data) collected from different sensors in the past (on the same or similar type of structure)
 - focus is prediction of new data based on the collected (existing) sensor data
- ML learning is STM - recent application
- past logs

Machine learning is common

- i) robot control
 - ii) Human-computer interaction
 - iii) speech recognition
- }

more details

③ Worden & Mansan

- identified the usefulness of mlc learning to detect damage

- Neural Network is very successful, and makes STM more competent & reliable

- mlc learning have also been attempted to complex problems is health monitoring

- for example, changes set is the structure due to operational and varying environmental conditions

i) Application of mlc learning - Embedded sensors

Aim: decentralizing the sensor fault detection and making it completely autonomous

Computer science 1 GPR, data mining

process of detecting patterns and structures within the data

- broader concept of pattern recognition - necessary is STM

- continuous, automatic systems

- Data mining can enable pattern of future predictions

- useful in decision making processes

- facilitate/accelerate feature extraction

- to load useful/meaningful

conclusions on the patterns

In data mining, techniques employed to detect patterns within the data set \Rightarrow machine learning

- Physio-based approaches in STM.
causes computational overload to the STM process.
- Recommendation & recommended data-driven approach
 - machine learning
- STM is concerned, machine learning is seen as a task of generating knowledge about the structural behavior from previously collected sensor data

Basic requirements

- Structural responses under operational, environmental, accidental loads are well understood, theoretically
- They are also documented well
- Detection & such repairs, on the full-scale structure is non-trivial
 - complex nature & combination of various factors acts, on the structure.
 - mostly unknown
 - highly level of uncertainty in their predictions

ii) output of system

may be challenged as is wrong/faulty

as then can opt

i) sensor calibrations

ii) interpretation + threshold values

iii) wrong pattern recognition

very difficult to
check from the
observed data

The observed data - should be put to self-diagnosis to
reveal useful/meaningful output

⇒ machine learning

Classifications of machine learning : STM

③ Categories

- (1) Supervised learning
- (2) Unsupervised learning
- (3) Semi-supervised learning

Supervised learning - leads a meaningful output - "labeled data"

- specific pattern of output (pairs of input/output)
- using labeled data, it is easy to classify new set of data

Unsupervised learning - deals with pattern recognition within the data set

- 'unlabeled data'

- data set with unspecified output
and not paired with any specific input

- They compare to from a general group

In STM, Unsupervised learning - unlabeled data can be useful to detect the existing dangers

✓ labeled data - can be useful to detect the types of dangers it's identifying as well

ml learn process, require labeled data to predict new data

- Hence, supervised learning is very appropriate to handle STM & complex problems.

- In supervised learning, '1' logic-based algorithms

(decision Tree

Rule-based classifier)

(i) perception-based algorithms

(ii) neural networks (single-layer perceptron, multi-layer)

(iii) statistical learning (Bayesian networks, etc)

- (v) instance-based learning (k-nearest neighbor algorithm)
- application to SVM

Complexity of implementing the machine learning approach

Aim: to create a decentralized, autonomous genetic-fault detection

Example: Analytical redundancy

- Instead of physically isolating multiple sensors for measuring one single parameter,

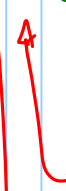
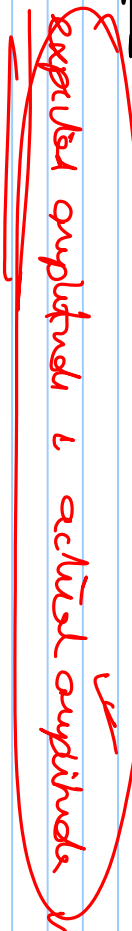
analytical redundancy utilizes the coherence & relationship b/w the given variables in the structure

Example

Measurement of peak amplitudes + frequency spectrum

- a) obtained from Fourier Transforms of acceleration response
- b) response peak amplitude, obtained from different sensors of the same structure

But the above test data are correlated to arrive at the required redundancy

- This condsn is used to predict the model peak amplitude of selected sensor locations 
- Deviation b/w the expected amplitude & actual amplitude will  now because isdi cptr & sensor fault it

concluded to be miscalibration of sensor

sensor location
(decision) ignore (use fault values)

In this whole process, there is an important advantage

- No Prior knowledge about the sensor location. Experimental (i.e. hardware details are required

- This is purely computer

Data-driven approach ✓

sensor-fault is identified based on the sensor data collected previously

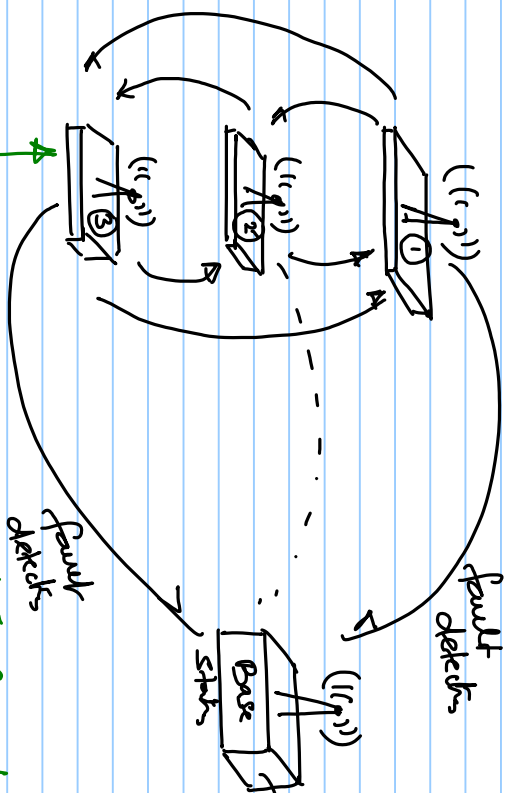
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Wireless STM

Comparison of wireless nodes

- Maximal accelerometers
- a bar station
- host computer

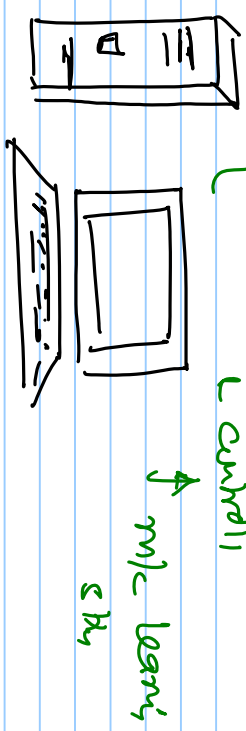


- i) acquisition of data (acceleration)
- ii) Fourier Transform (FFT)
- (iii) detects the highest peak
- (iv) communicates the model peak amplitude
- (v) uses model peaks of correlation peak data

Sensors

- no mac, simply measuring ops

Advanced - { They measure, process, drop, decide, detect



- i) storage of fault data
- ii) used in decision making

Additional reads

(1) K. H. Law, K. Srimally, Y. Wang. 2014

. Sensor data management techniques for Infrastructure Asset Mgmt,

chapter 6 book title

'Sensor Technologies for civil infrastructure'

Edited by M.L. Wang, J.P. Lynch & H. Sohn
Woodhead Publishing, UK

Summary

- importance of mlc learning
- supervised learning (labelled data) is enhancing its own output
- tensors are advanced to perform multiple tasks