

Missing Sensors Restoration for System Control and Diagnostics

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Issue

- **Monitoring, diagnosis and control of complex drive systems requires measurements that are **available** and **accurate****
 - Physical quantities such as **current, voltage, speed, torque, position, frequency, flux, signal propagation**
 - **Abstract mathematical variables of features**

Problems

- Sensors data, analog or digital, are transmitted to a *control hub* that makes decision.
- **Failures could occur in**
 - communication links
 - routers, switches
 - sensors
- **Sensor data may also be lost due to**
 - transmission delays
 - Noise
 - unreliable transmission

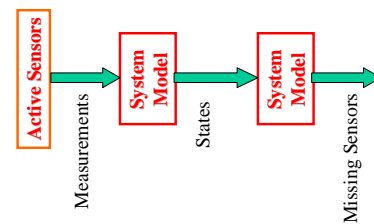
Missing Sensors Restoration (MSR)

- MSR establishes data dependency among sensors to later restore information of failed sensors.

Existing Techniques for Reconstruction of Missing Sensors

- State Estimation
- Kalman Filter
- Extended State Estimation
-

State Estimation



State Estimation - Limitations

- The system model must be available
 - Complex structure is approximated
- The system model must be elaborate enough to capture all necessary correlations of all sensors and all states
 - Nonlinear relationship is often linearized to simplify the computations
- The system Topology must be Fixed
 - Is not true for dynamically changing system

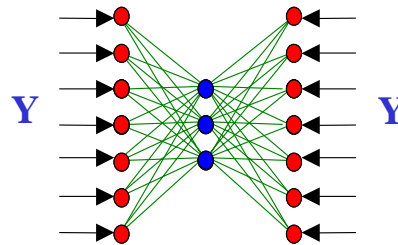
State Estimation - Limitations

- The system must be observable
 - ?
- Pseudo-measurements must be introduced
 - Often compromise the accuracy of the estimator.
- The quality of the data must be high.
 - noise or other uncertainties degrades the estimation accuracy

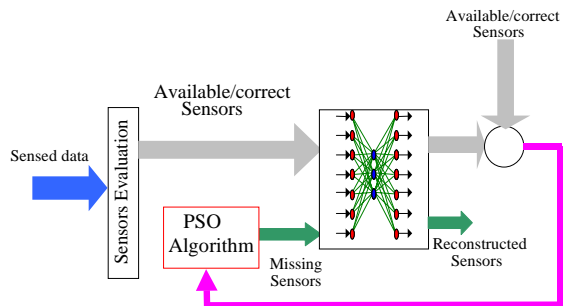
Missing Sensors Restoration (MSR)

- Two step process:
 1. Develop an encoder to capture the dependency among sensors
 2. Use fast search technique and the encoder to restore the data of the missing sensor.

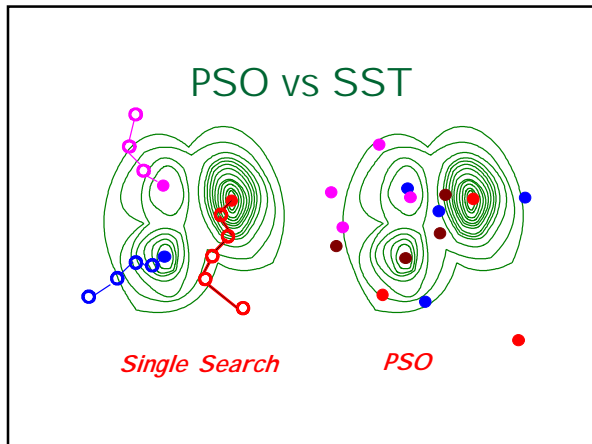
MSR: NN Encoder



MSR: Data Reconstruction



Particle Swarm Optimization Search

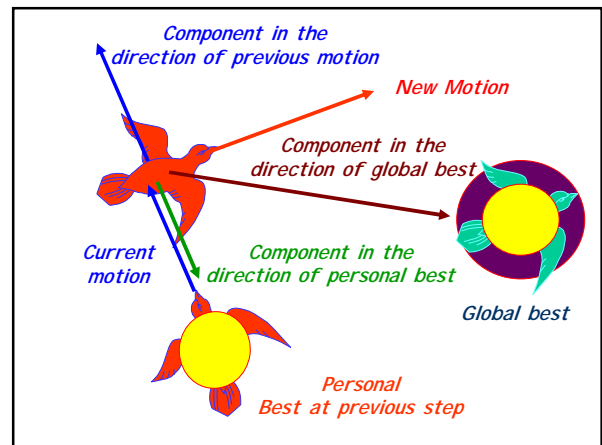


Particle Swarm Optimization

- **Inventors:** James Kennedy and Russell Eberhart
- An Algorithm originally developed to imitate the motion of a Flock of Birds, or insects
- **Assumes Information Exchange (Social Interactions) among the search agents**
- Basic Idea: Keep track of
 - Global Best
 - Self Best

How does it work?

- **Problem:**
Find X which minimizes $f(X)$
- **Particle Swarm:**
 - **Start:** Random set of solution vectors
 - **Experiment:** Include randomness in the choice of new states.
 - **Remember:** Encode the information about good solutions.
 - **Improvise:** Use the 'experience' information to initiate search in a new regions



Particle Swarm Dynamics

$$\vec{x}(k+1) = \vec{x}(k) + \vec{v}(k)$$

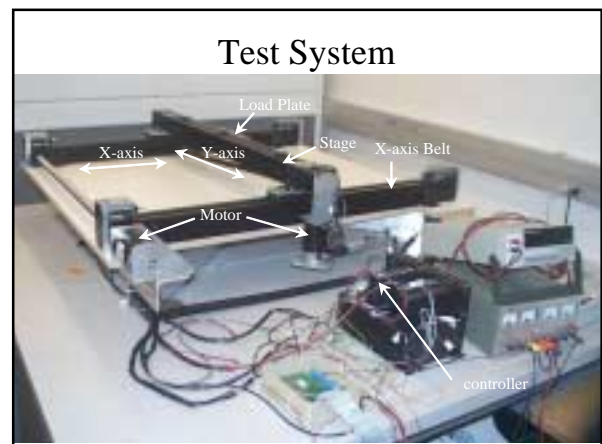
*non-zero velocity
PS never stop flying*

$$\vec{v}(k+1) = w \cdot \vec{v}(k) + r(0, a_1) \cdot (\vec{x}_{SelfBest}(k) - \vec{x}(k)) + r(0, a_2) \cdot (\vec{x}_{GroupBest}(k) - \vec{x}(k))$$

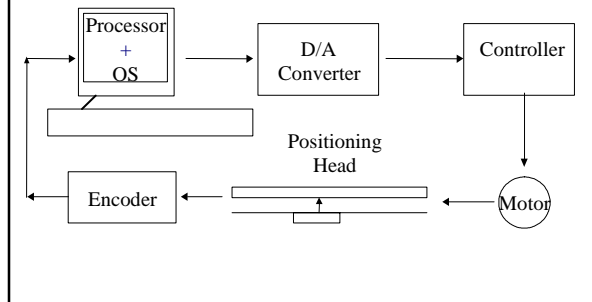
*Self consciousness of the swarm
Controlled randomness*

Inertia

The collective consciousness of the swarm



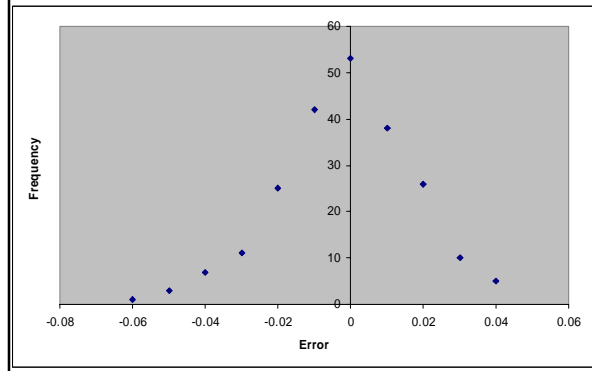
Test System



Test Conditions

- Measurements:
 - Voltage, current, speed, torque, stage position, reference and carrier frequencies.
- Several loading and operating conditions were used.
- Each of the measurements was assumed missing and is predicted by the MSR

Test result



Conclusions

- MSR is an effective method.
- Applied to
 - Airplane engine vibration
 - Power system vulnerability monitoring
 - Chemical process
 - Drives and control
- Need to be tested in more complex drives environment where ambient and electrical conditions are sensed.