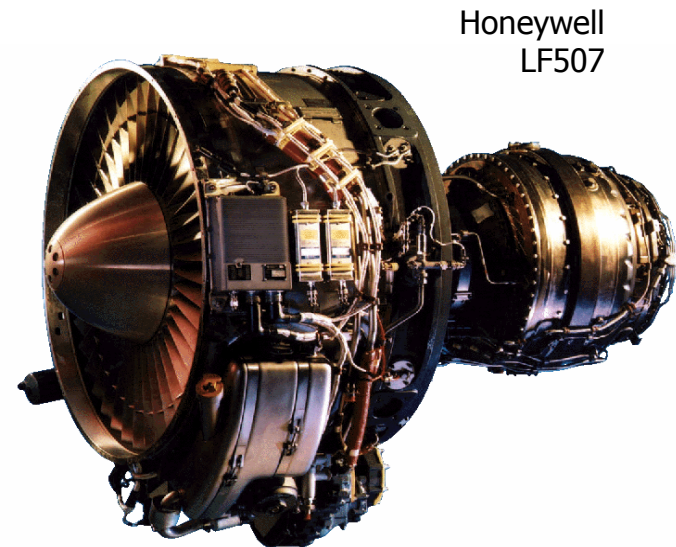


Lecture 16 - Health Management

- Diagnostics in control system design
 - BIT/BITE
- Fault tolerance - redundancy
- Systems health management applications
 - Added value – maintenance and support
- Statistical Process Control
- Parameter estimation

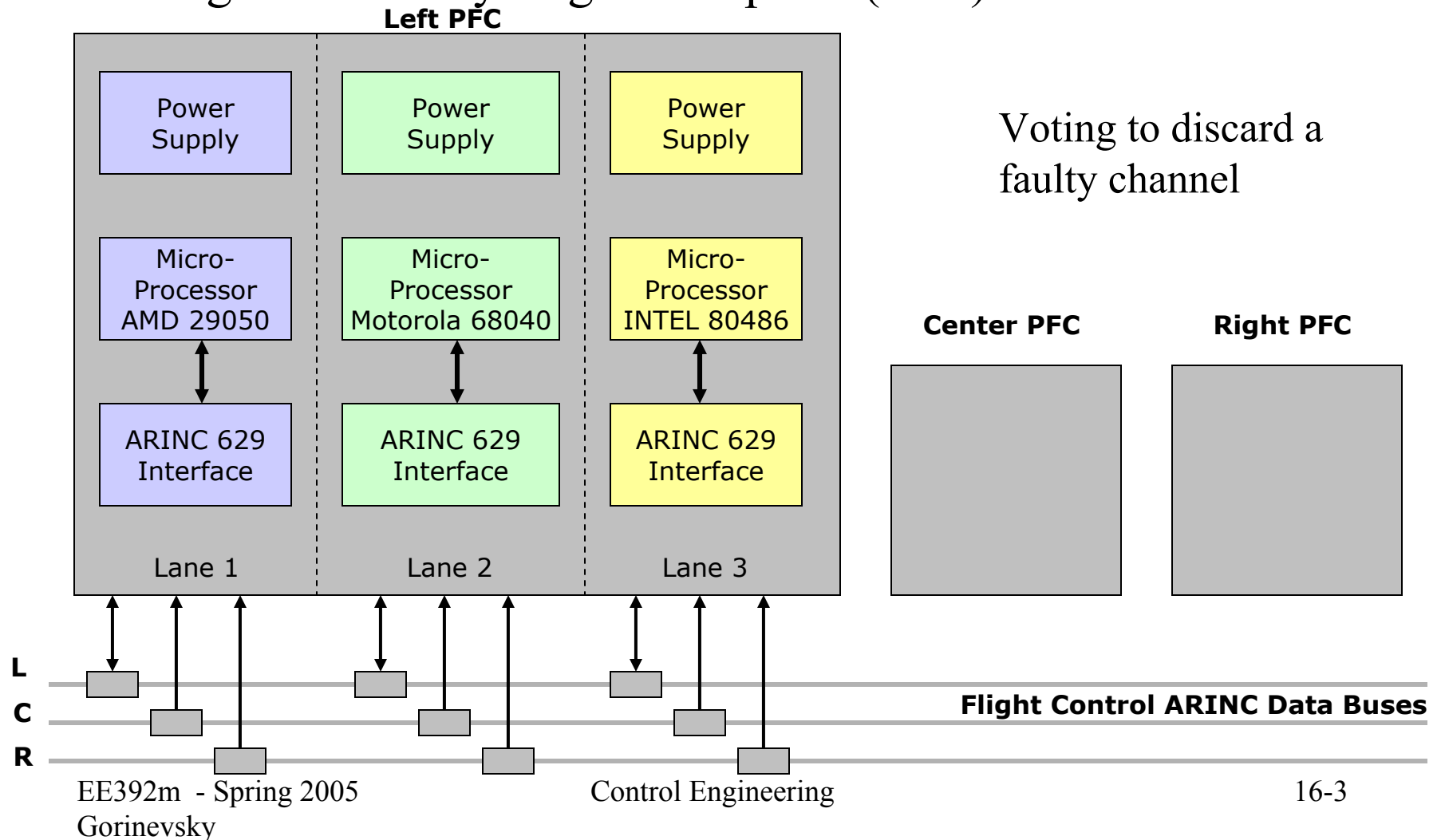
Diagnostics in Control Systems

- Control algorithms make less than 20% of the embedded control application code in safety-critical systems
- The remaining 80% deals with special conditions, fault accommodation
 - BIT (Built-in Test - software)
 - BITE (Built-in Test Equipment - hardware)
 - Binary results
 - Messages
 - Used in development and in operation



Fault Tolerance: Hardware Redundancy

- Boeing 777 Primary Flight Computer (PFC) Architecture



Analytical Redundancy

- Analytical Redundancy
 - correlate data from diverse measurements through an analytical model of the system
- Estimation techniques
 - KF observer
- Talked about in the literature
- Used only in much simplified form:
 - on loss of a sensor, use inferential estimate of the variable using other sensor measurements
 - on loss of an actuator, re-allocate control to other actuators

Systems Health Management

- Emerging technology - recent decade
 - Less established than most of what was discussed in the lectures
- Systems fault management functions
 - Abnormality detection and warning - something is wrong
 - Diagnostics - what is wrong
 - Prognostics - predictive maintenance
 - Accommodation – recover from fault
- On-line functions - control system
 - BIT (Built-in-test)
 - Accommodation; FDIR (Fault Detection, Identification, & Recovery)
- Off-line functions - enterprise system
 - Maintenance automation
 - Logistics automation

Vehicle Health Management

- IVHM - Integrated Vehicle Health Management - On-board
- PHM - Prognostics and Health Management - On-ground
- Vehicles: space, air, ground, rail, marine
 - Integrated systems, many complex subsystems
 - Safety critical, on-going maintenance, on-board fault diagnostics



EE392m - Spring 2005
Gorinevsky



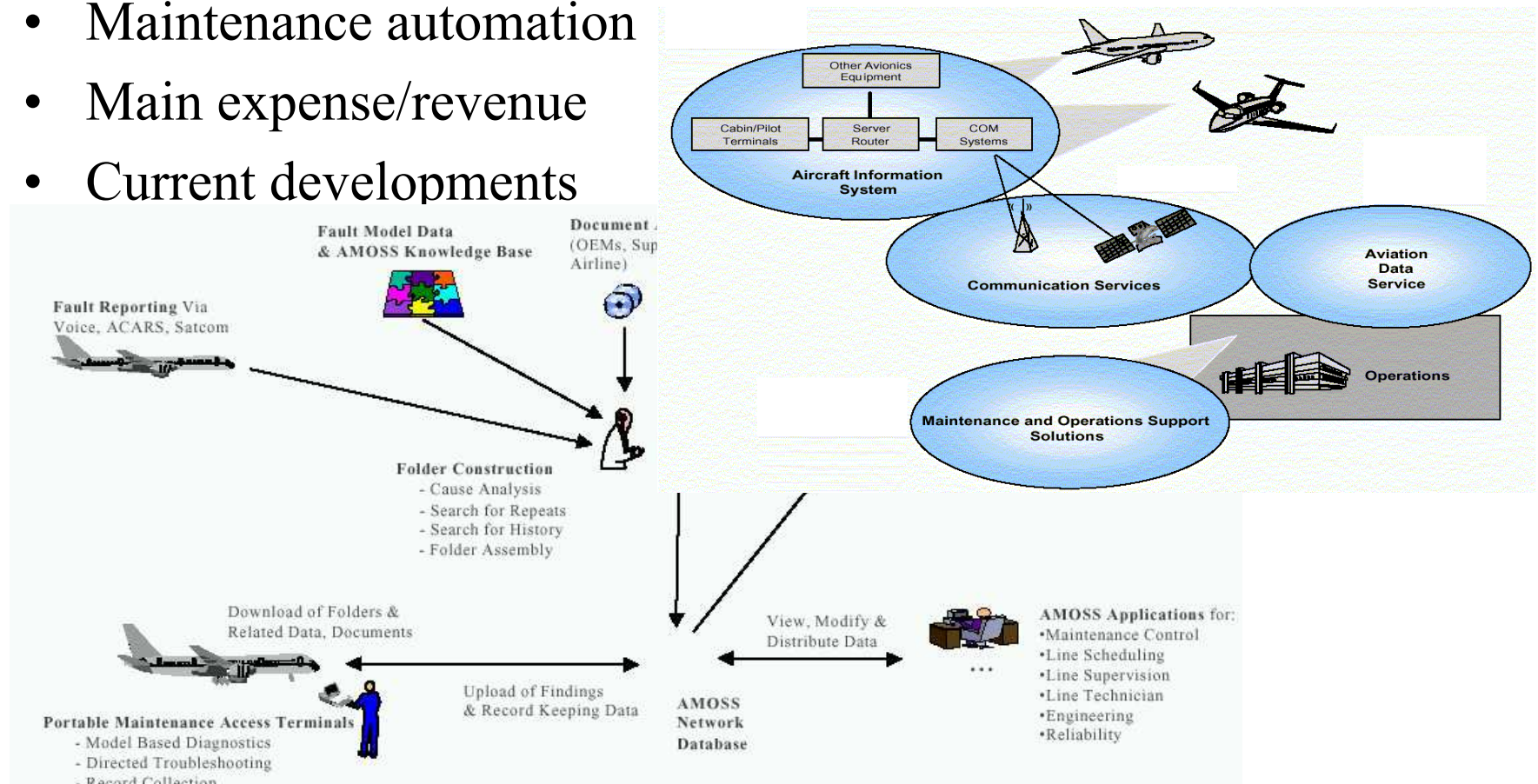
Control Engineering



16-6

Airline Enterprise - Maintenance

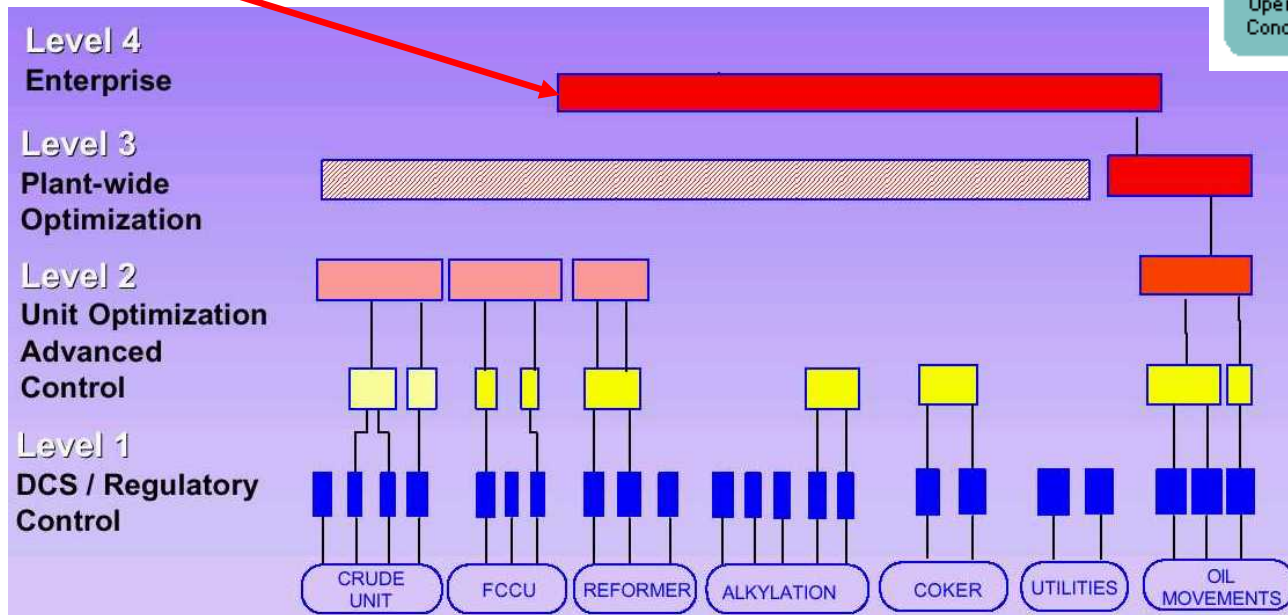
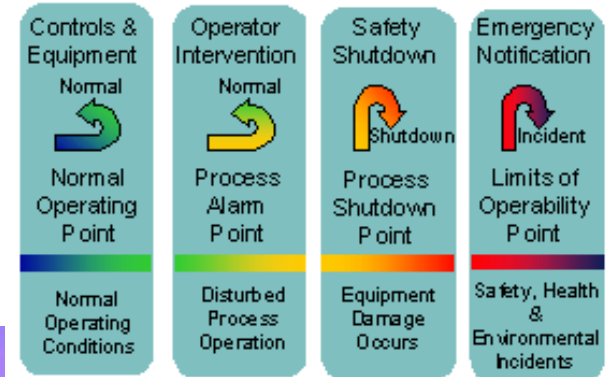
- Integrated on-board and on-ground system
- Maintenance automation
- Main expense/revenue
- Current developments



Industrial Plants

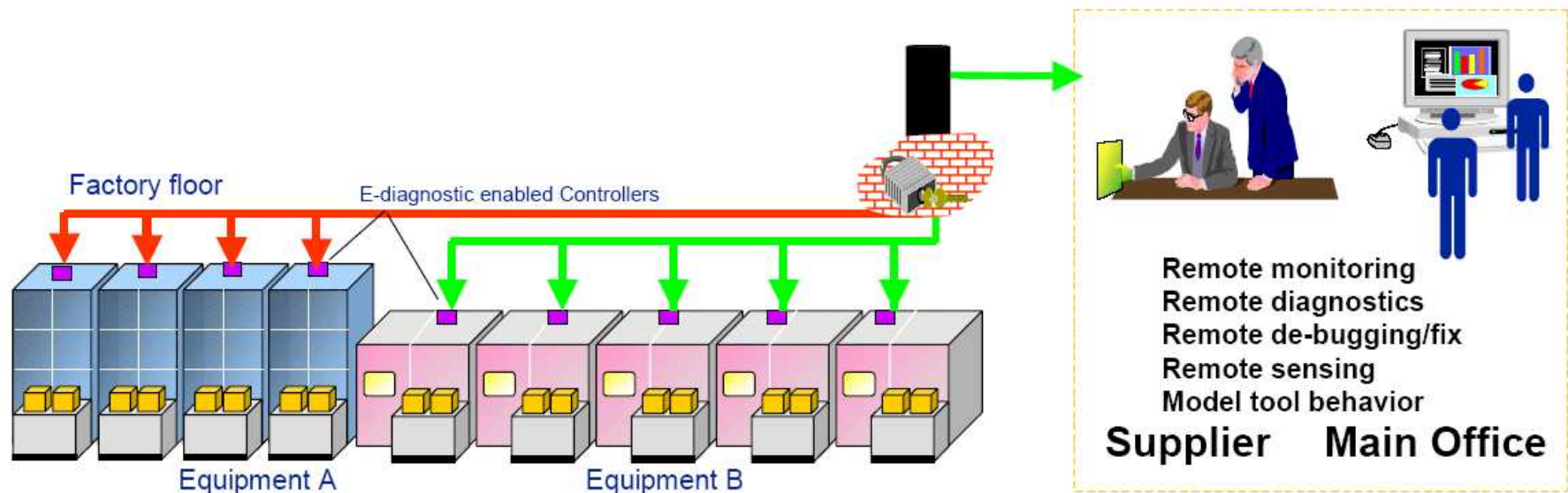
- Abnormal Situation Management
 - large cost associated with failures and production disruption
 - solutions are presently being deployed

Layers of Protection



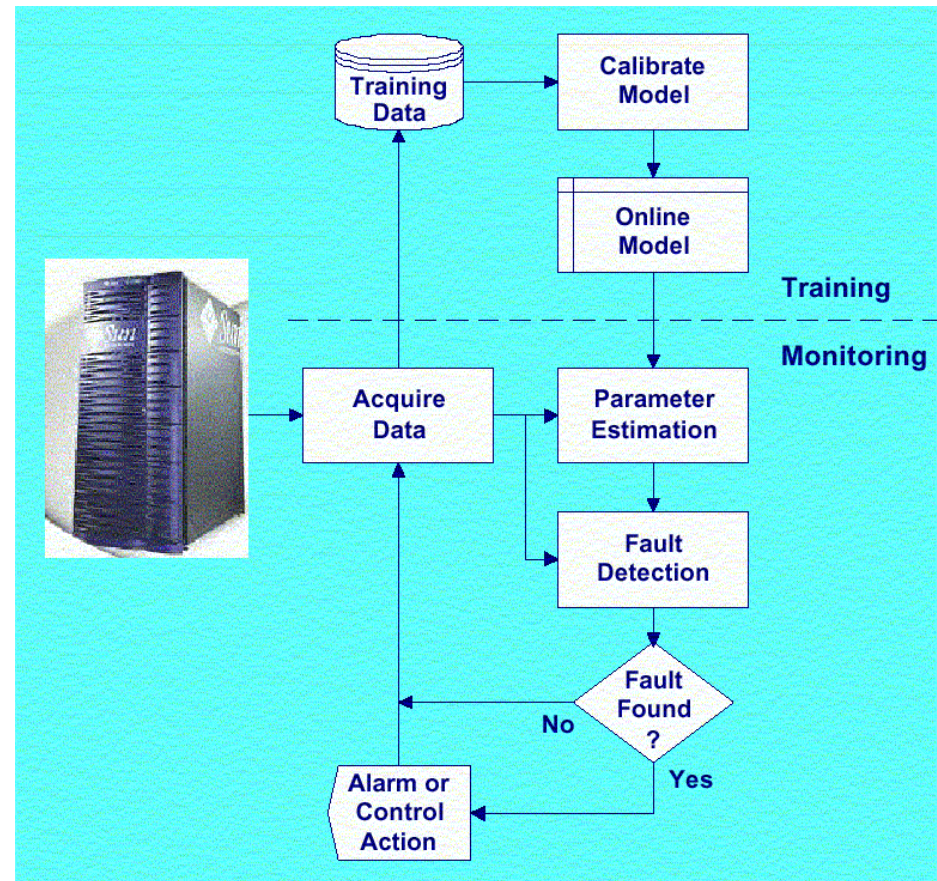
Semiconductor Manufacturing

- E-diagnostics initiative by SEMATECH, since 2001
- Consortium of all major semiconductor companies
- Now a part of SEMATECH e-Manufacturing
- Prototype demonstration in 2005



Computing

- Autonomic computing
 - Fault tolerance
 - Automated management, support, security
 - IBM, Sun, HP - Scientific American, May 2002
- Sun Storage Automated Diagnostic Environment
 - Health Management and Diagnostic Services



K.Gross, Sun Microsystems

Abnormality Detection - SPC

- SPC - Statistical Process Control (univariate)
 - discrete-time monitoring of manufacturing processes
 - early warning for an off-target quality parameter
- SPC vs. EPC
 - EPC (Engineering Process Control) - ‘normal’ feedback control
 - SPC - operator warning of abnormal operation
- SPC has been around for 80 years
- Three main methods of SPC:
 - Shewhart chart (20s)
 - EWMA (40s)
 - CuSum (50s)

SPC: Shewhart Chart

- Process model - SISO
 - quality variable randomly changes around a steady state value
 - the goal is to detect change of the steady state value

$$X(t) \approx \begin{cases} N(\mu_0, \sigma^2), & t \leq T \\ N(\mu_1 \neq \mu_0, \sigma^2), & t > T \end{cases}$$

- Shewhart Chart

$$Y(t) = \frac{X(t) - \mu_0}{\sigma} \quad \text{detection: } Y(t) > Z = c_1$$

- Simple thresholding for deviation from the nominal value μ_0
- Typical threshold of $3\sigma \Leftrightarrow 0.27\%$ probability of false alarm

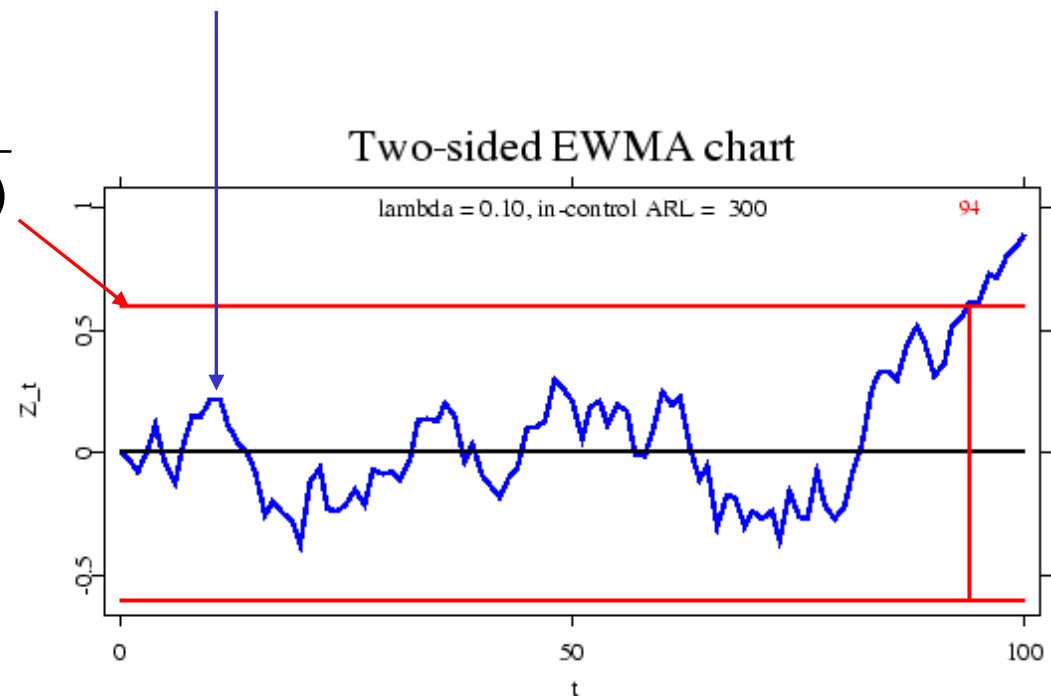
SPC: EWMA

- EWMA = Exponentially Weighted Moving Average
- First order low pass filter

$$Y(t+1) = (1-\lambda)Y(t-1) + \lambda X(t)$$

– Detection threshold

$$Z = c_2 \sqrt{\lambda(2-\lambda)}$$



SPC: CuSum

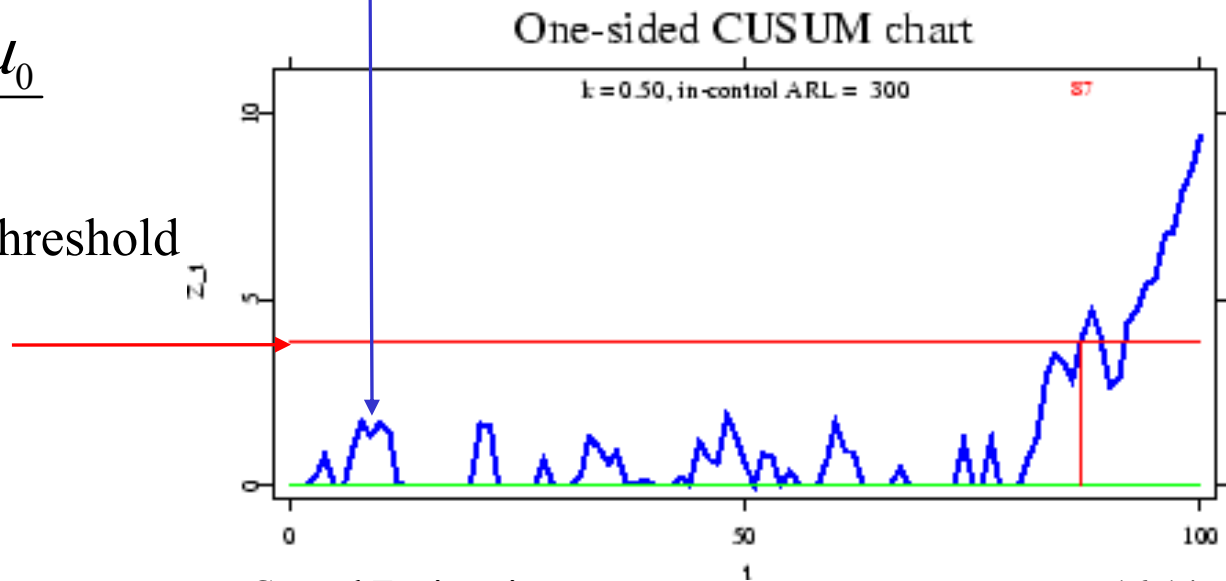
- CuSum = Cumulative Sum
 - a few modifications
 - one-sided CuSum most common

$$Y(t+1) = \max \left\{ 0, Y(t) + \frac{X(t) - \mu_0}{\sigma} - k \right\}$$

$$k = \frac{\mu_1 + \mu_0}{2\sigma}$$

- Detection threshold

$$Z = c_3$$



Multivariate SPC - Hotelling's T^2

- The data follow multivariate normal distribution

$$X(t) \approx N(\mu, \Sigma)$$

$$X(t) = \Sigma^{1/2} Y(t) + \mu$$

- Empirical parameter estimates

$$\mu = E(X) \approx \frac{1}{n} \sum_{t=1}^n X(t)$$

Uncorrelated
white noise

$$\Sigma = E((X - \mu)(X - \mu)^T) \approx \frac{1}{n} \sum_{t=1}^n (X(t) - \mu)(X^T(t) - \mu^T)$$

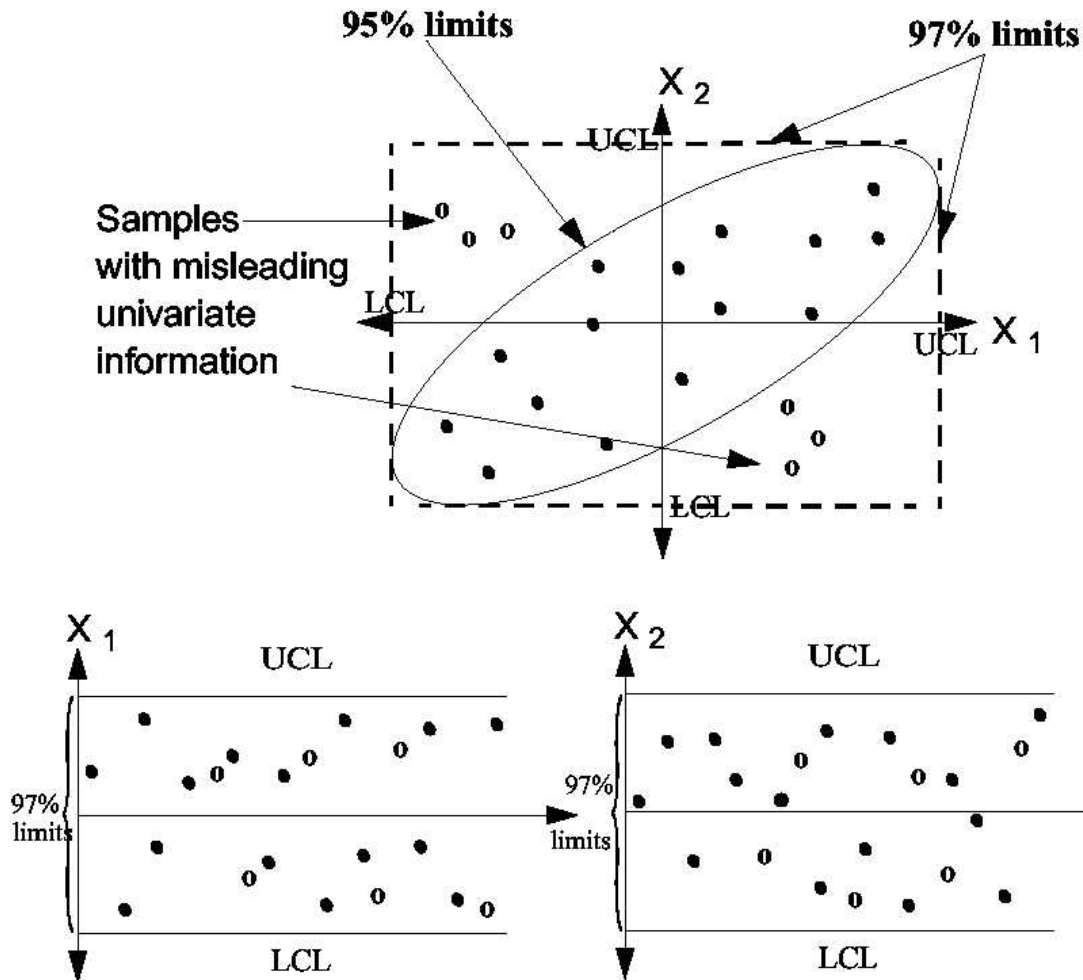
- The Hotelling's T^2 statistics is

$$T^2 = (X(t) - \mu)^T \Sigma^{-1} (X(t) - \mu)$$

$$T^2 = Y^T(t) Y(t)$$

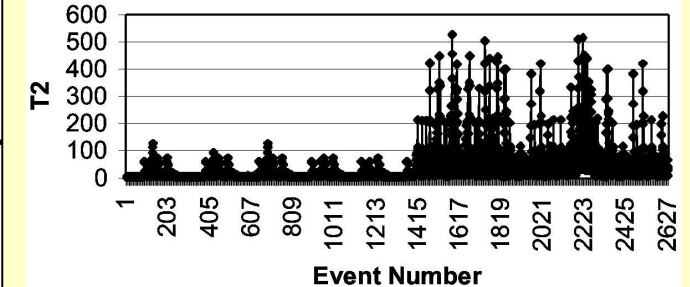
- T can be trended as a univariate SPC variable (almost)

Multivariate SPC - Example

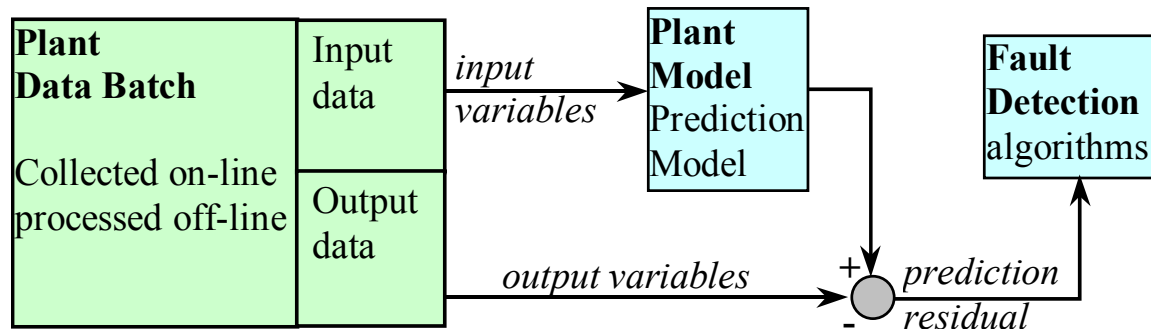


Example:

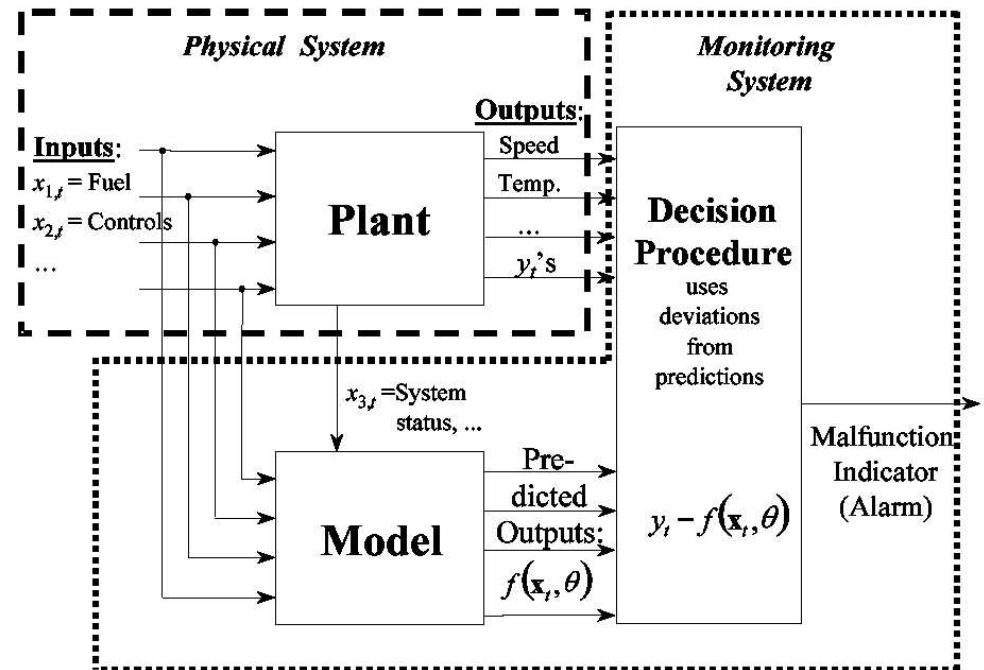
- MSPC for Cyber attack detection
- $X(t)$ consists of 284 audit events in Sun's Solaris Basic Security Module
- Ye & Chen, Arizona State



Model-based Fault Detection



- Compute model-based prediction residual
 - result of a simulation run
$$X = Y - f(U, \theta)$$
- If $\theta = 0$ (nominal case) we should have $X = 0$.
- X reflects faults



Model-based Fault Detection

- Compute model-based prediction residual $X(t)$ at cycle t
 - flight/trip/maneuver for a vehicle
 - update time interval or a batch for a plant
 - semiconductor process run
- $X(t)$ reflects modeling error, process randomness, and fault
- Use MSPC for detecting abnormality through $X(t)$
 - Hotelling's T^2
 - CuSum
- Does not tell us what the fault might be (diagnostics)

Parameter Estimation

- Residual model: $X = Y - f(U, \theta)$

$$X = \Phi \theta + \xi \qquad \Phi = -\frac{\partial f(U, \theta)}{\partial \theta}$$

- Fault models - meaning of θ
 - Sensor fault model - additive output change
 - Actuator fault model - additive input change
- Estimation technique
 - Fault parameter estimation - regression

$$\hat{\theta} = (\Phi^T \Phi + rI)^{-1} \Phi^T X$$