

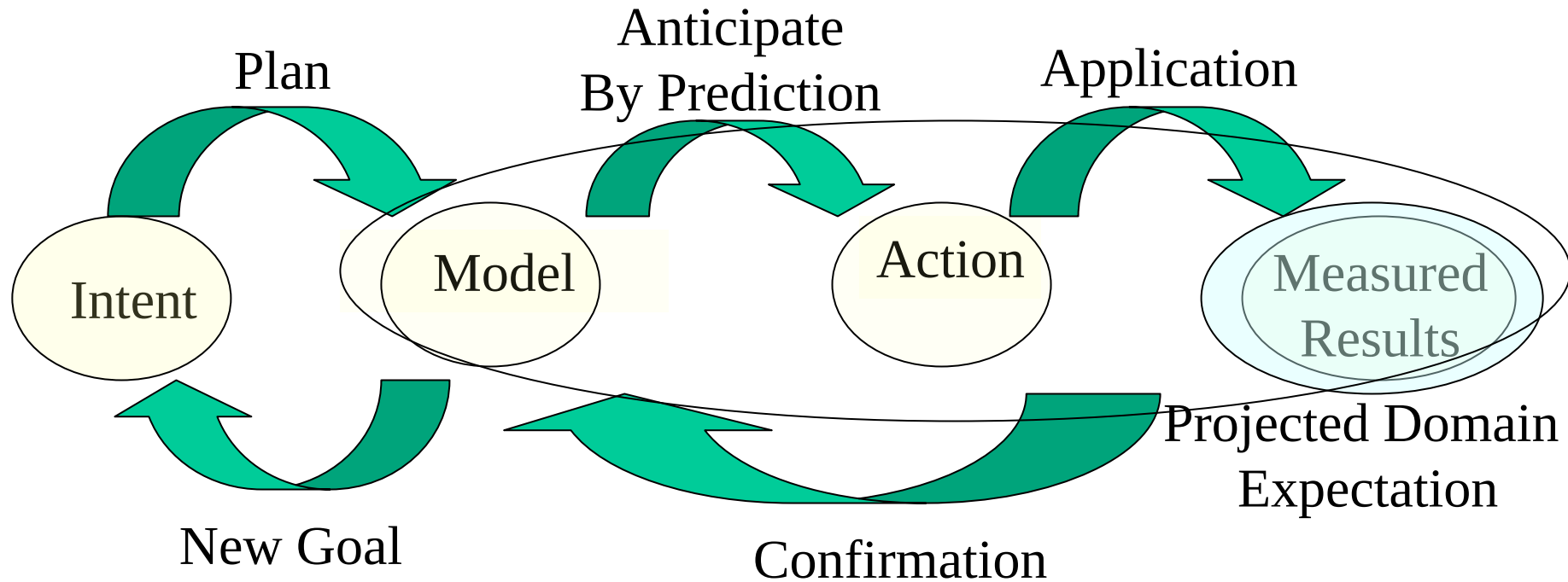
The Variable Structure Systems and Cognition

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McMaster University

Overview

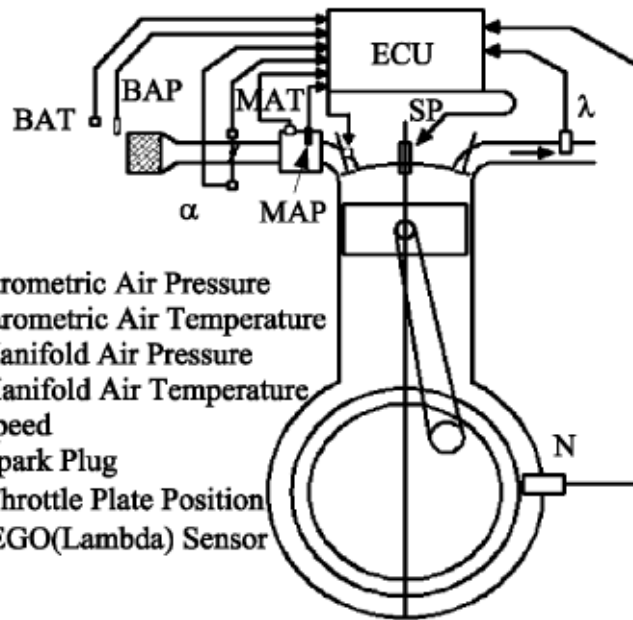
- Hypothesis
- Feedforward Control and Wear in Automotive Systems
- Overview of the Variable Structure Systems and Sliding Mode Control
- Variable Structure Filter
- Examples
- Questions

Behavioral Model

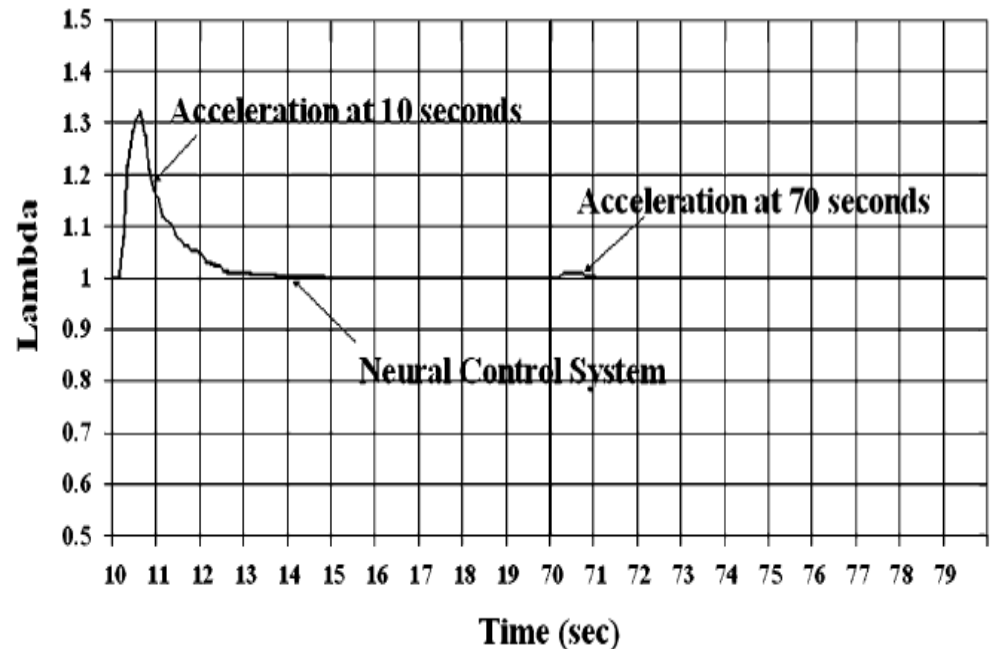
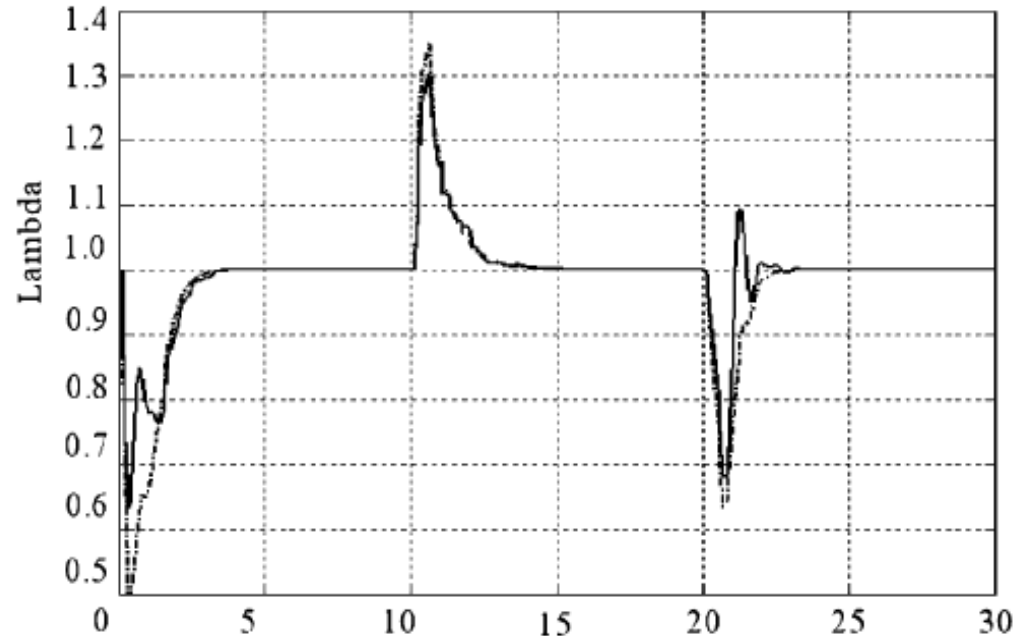


Model-based Feedforward Control

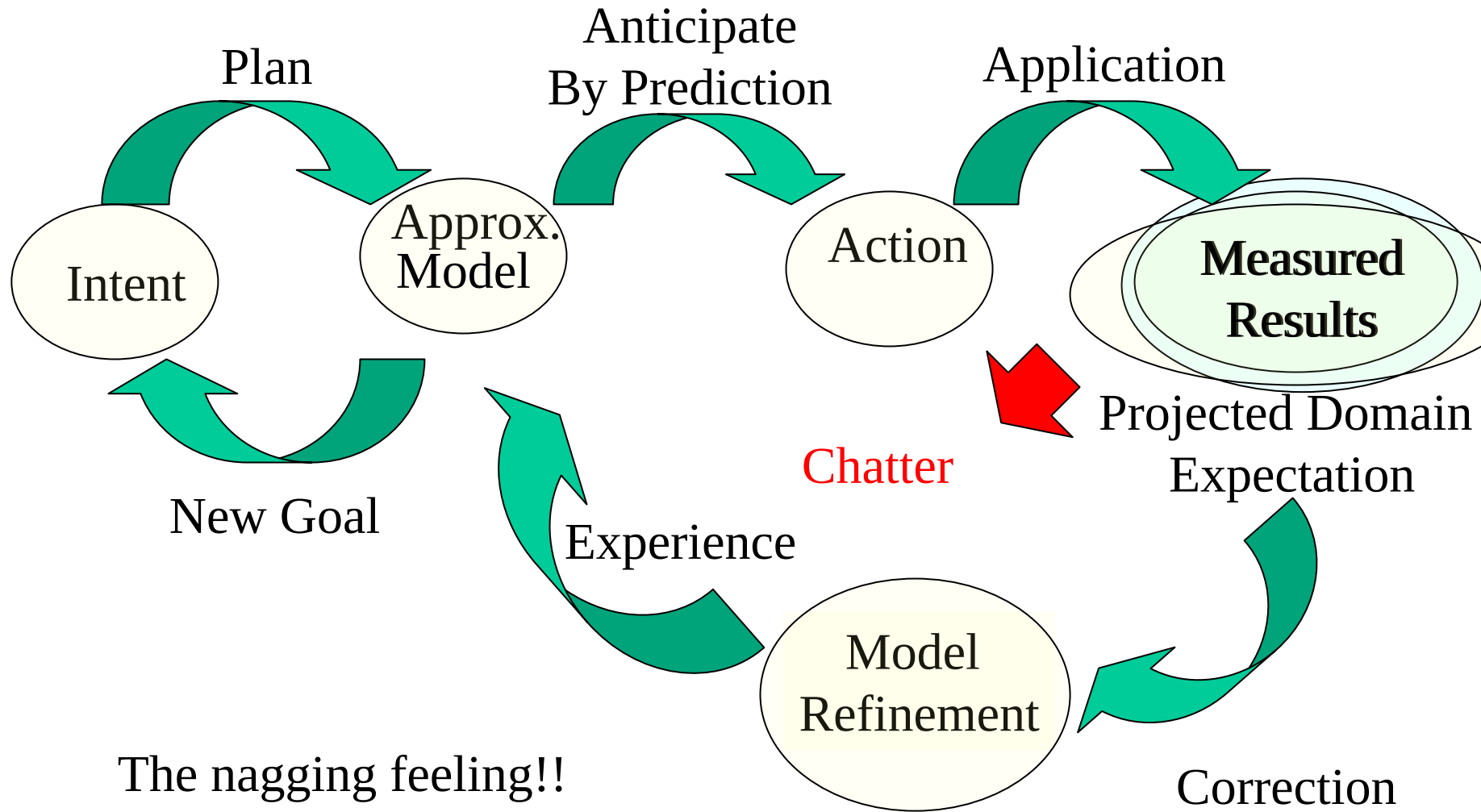
Automotive Control Context



BAP – Barometric Air Pressure
 BAT – Barometric Air Temperature
 MAP – Manifold Air Pressure
 MAT – Manifold Air Temperature
 N – Speed
 SP – Spark Plug
 α – Throttle Plate Position
 λ – EGO(Lambda) Sensor



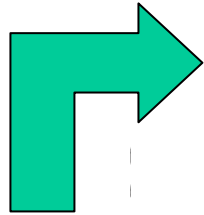
Hypothesis



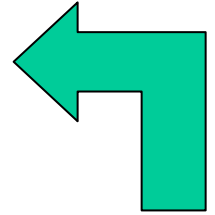
Hybrid Vehicles



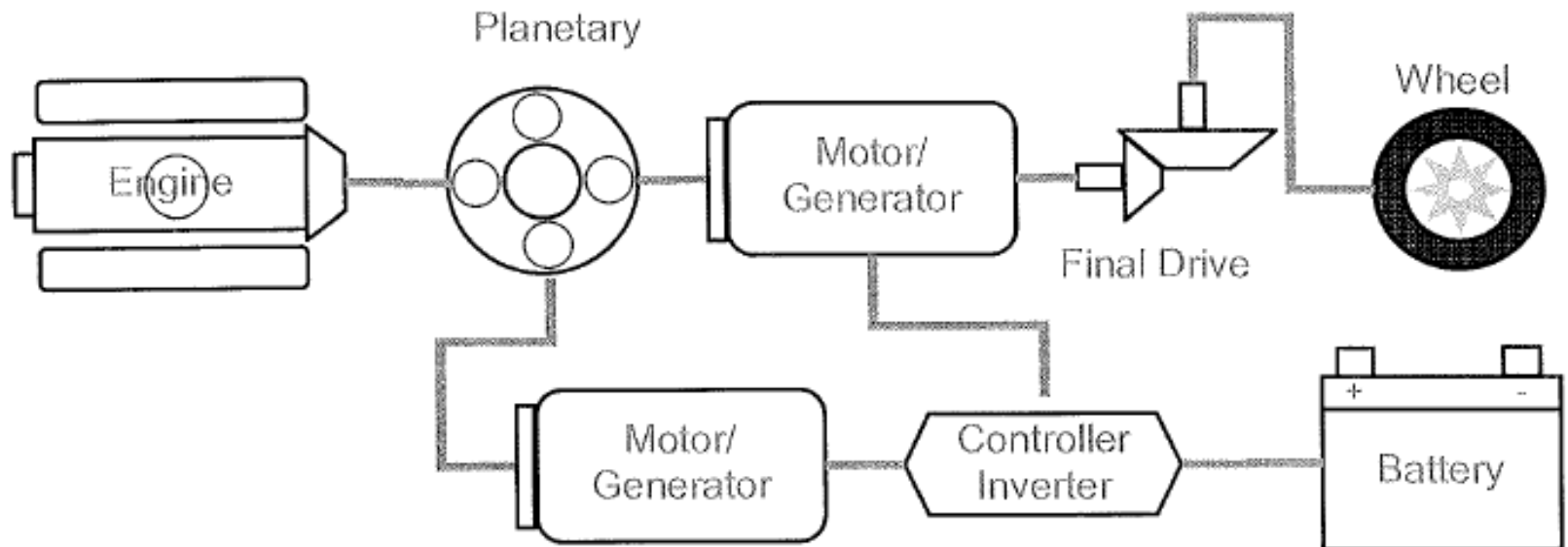
Steady-State
Engine Operation



Ford Escape Hybrid

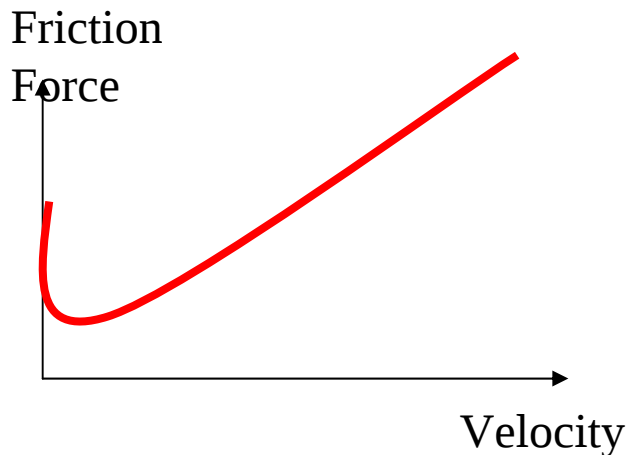
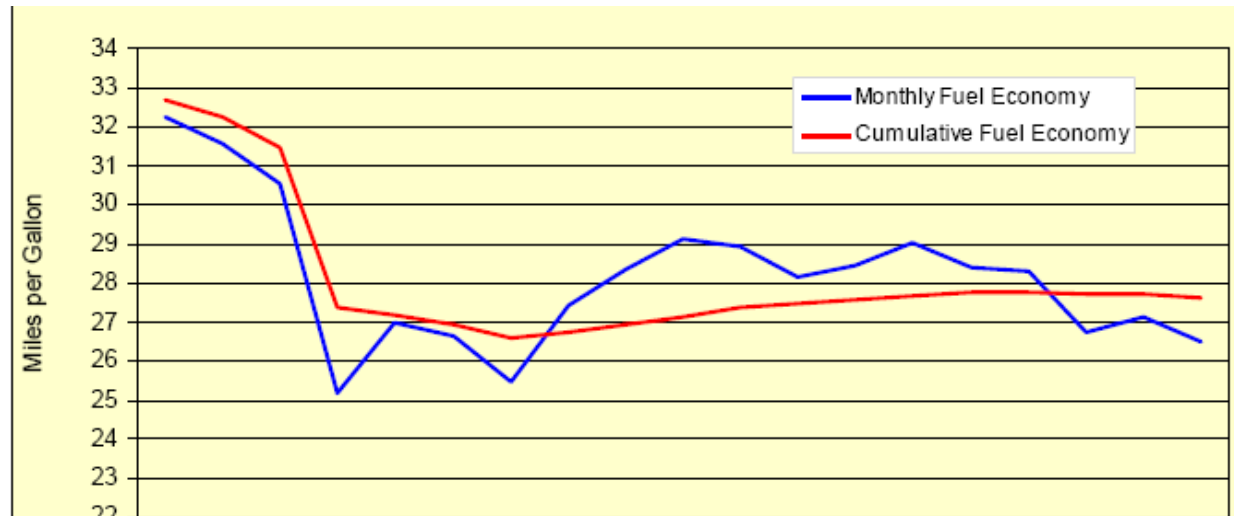


Regeneration



Fuel Economy Hybrid

Monthly Fuel Economy

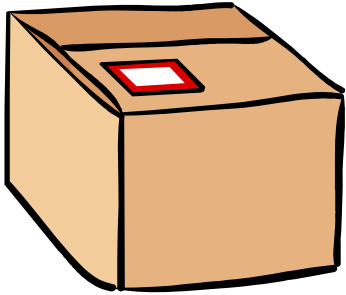


➡ *Friction characteristic is accurately described by a quadratic function of velocity*

➡ *Can be tracked in real-time by parameter estimation using minimal sets of sensor but multiple banks of filters*

Modeling & Model Identification

Black Box Model by
System Identification



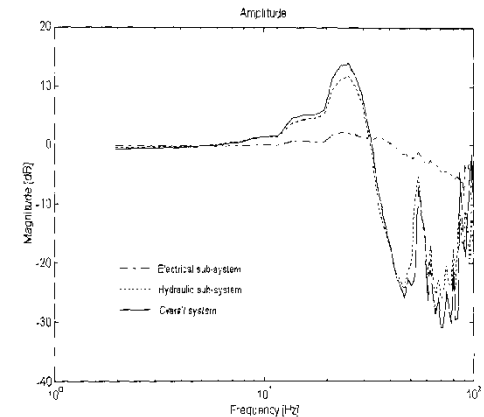
Experiments



Validated
Model

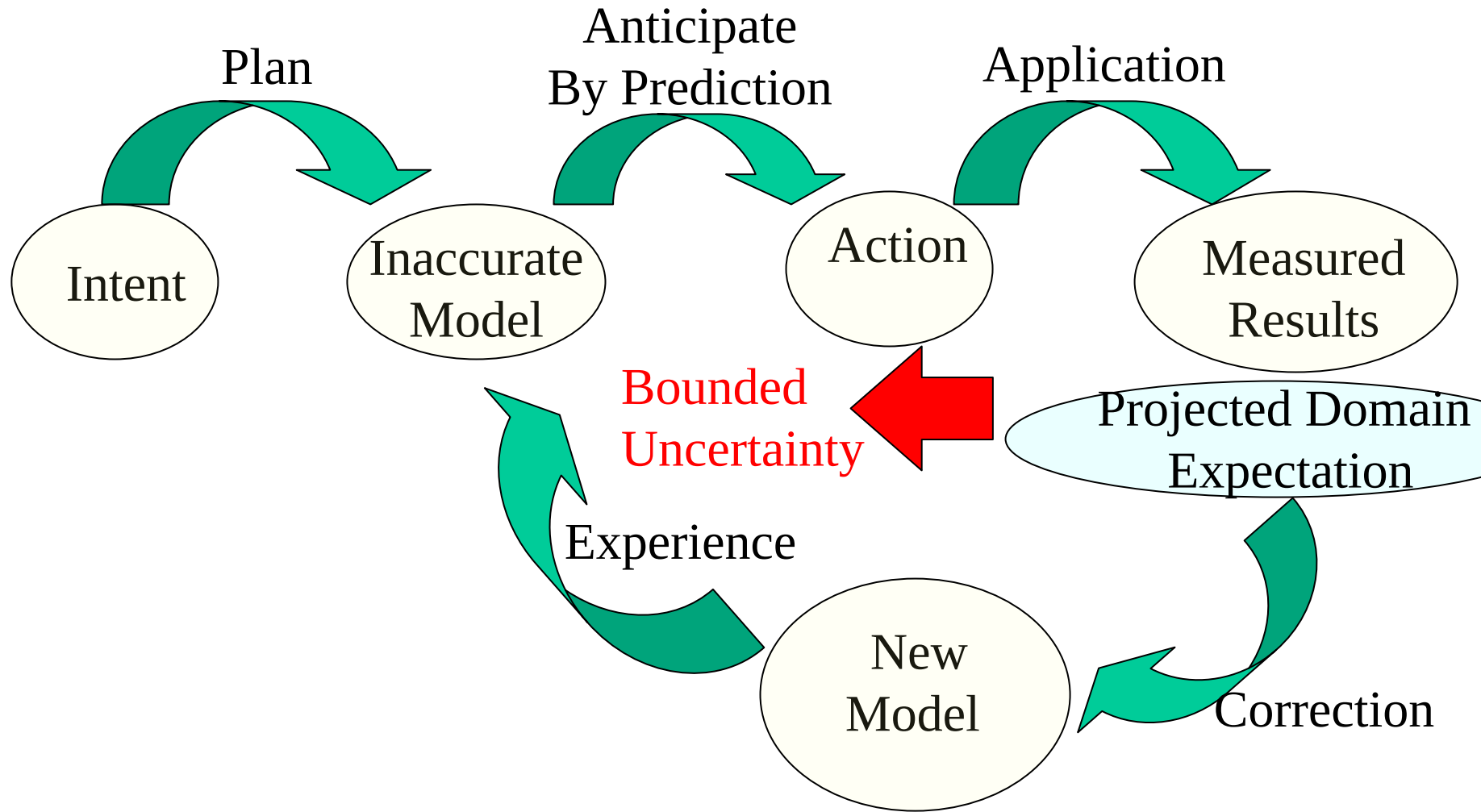


Physical Model



Sensitivity Analysis

Hypothesis



Research Thrust - White Box PHMS

- **Model & Fault Characterization**
 - Physical Modeling
 - System Identification & Dynamic Significance
 - Analysis & Model Structure Simplification
 - Experimentation for Model Parameter Quantification
 - Sensitivity Analysis
 - Mapping to Fault Conditions
- **Robust Parameter Tracking Theory**
 - State and P

“Achilles Heel” of White Box Condition Monitoring

Variable Structure Filter

- over
- **Prognosis by Association**
 - Expert knowledge
 - Experimentation
 - In-vehicle data collection and monitoring

Continuous Monitoring
Data Analysis
Labor intensive

Variable Structure Systems



States:

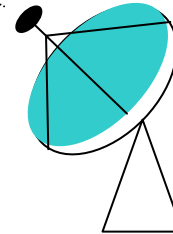
- Position
- Velocity
- Acceleration

State space = n-dimensional space made up of state space variables

Position

Actual Aircraft
Position

Time

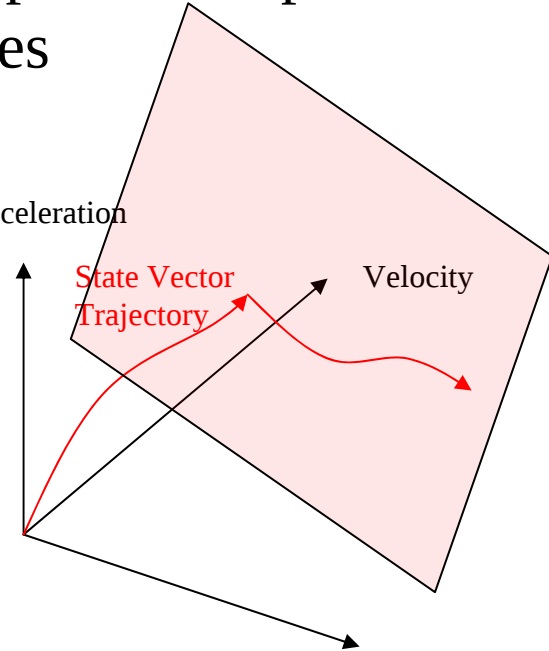


Acceleration

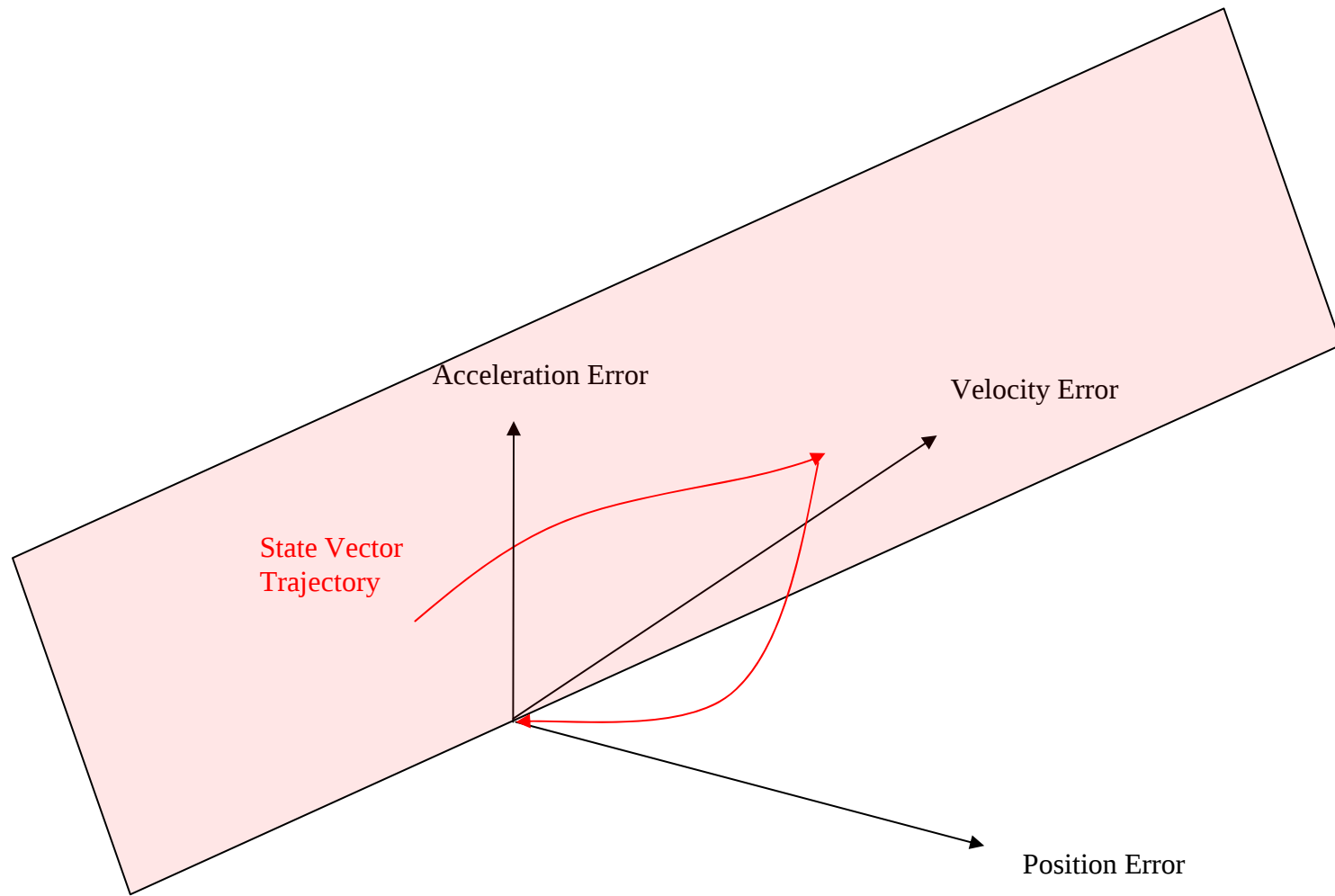
State Vector
Trajectory

Velocity

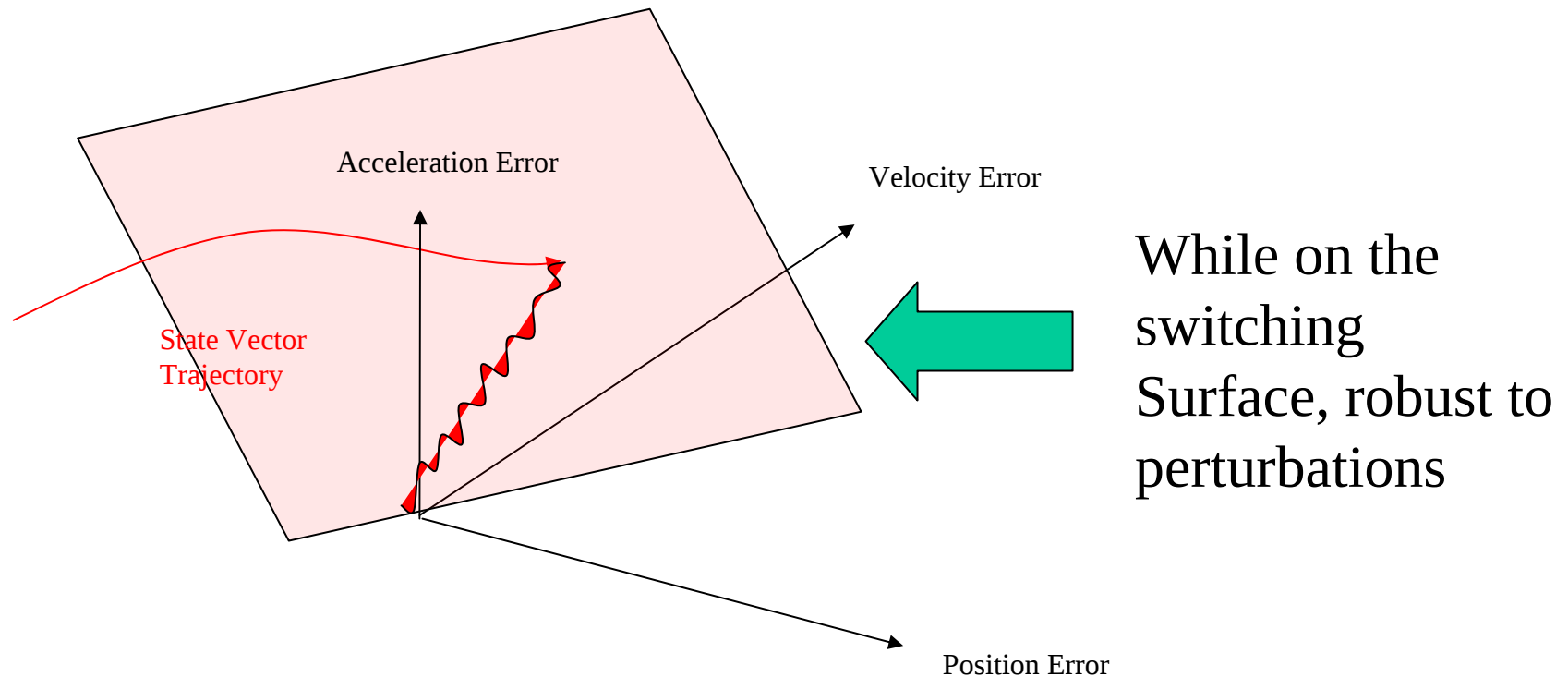
Position



Variable Structure Control

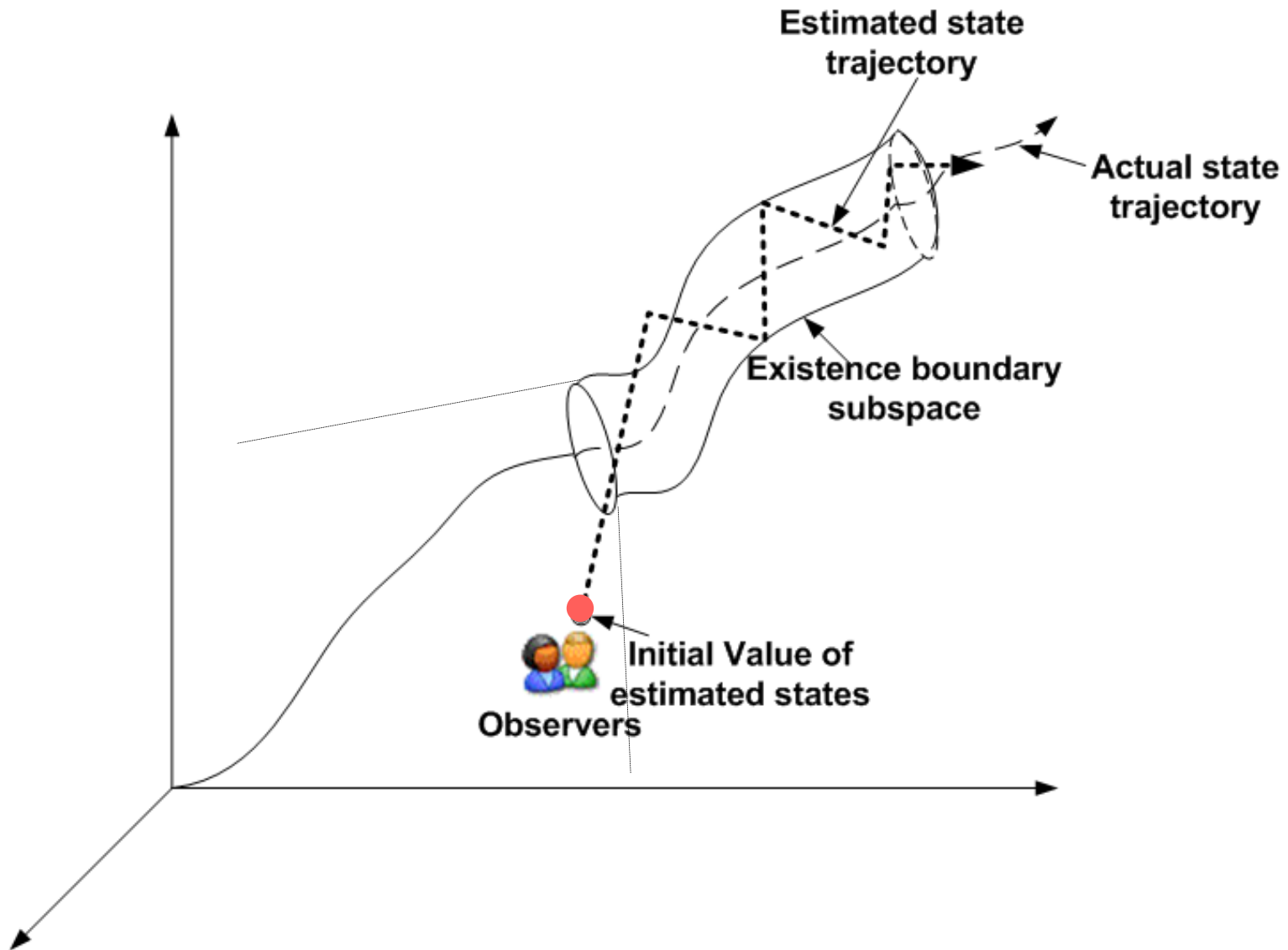


Sliding Mode Control

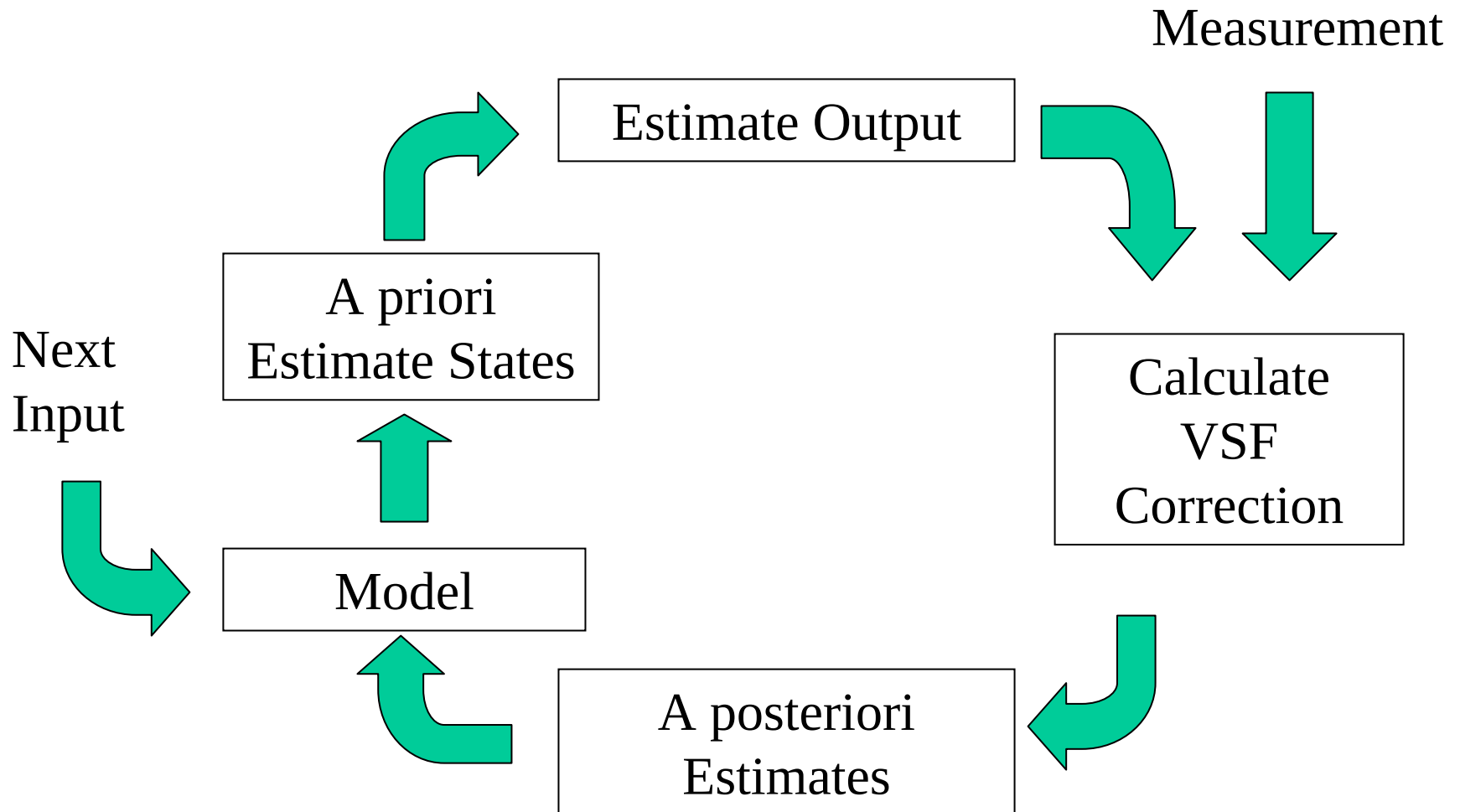


Variable Structure Filter For Robust Parameter Tracking

SVSF



Variable Structure Filter Predictor-Corrector Method



Theorem

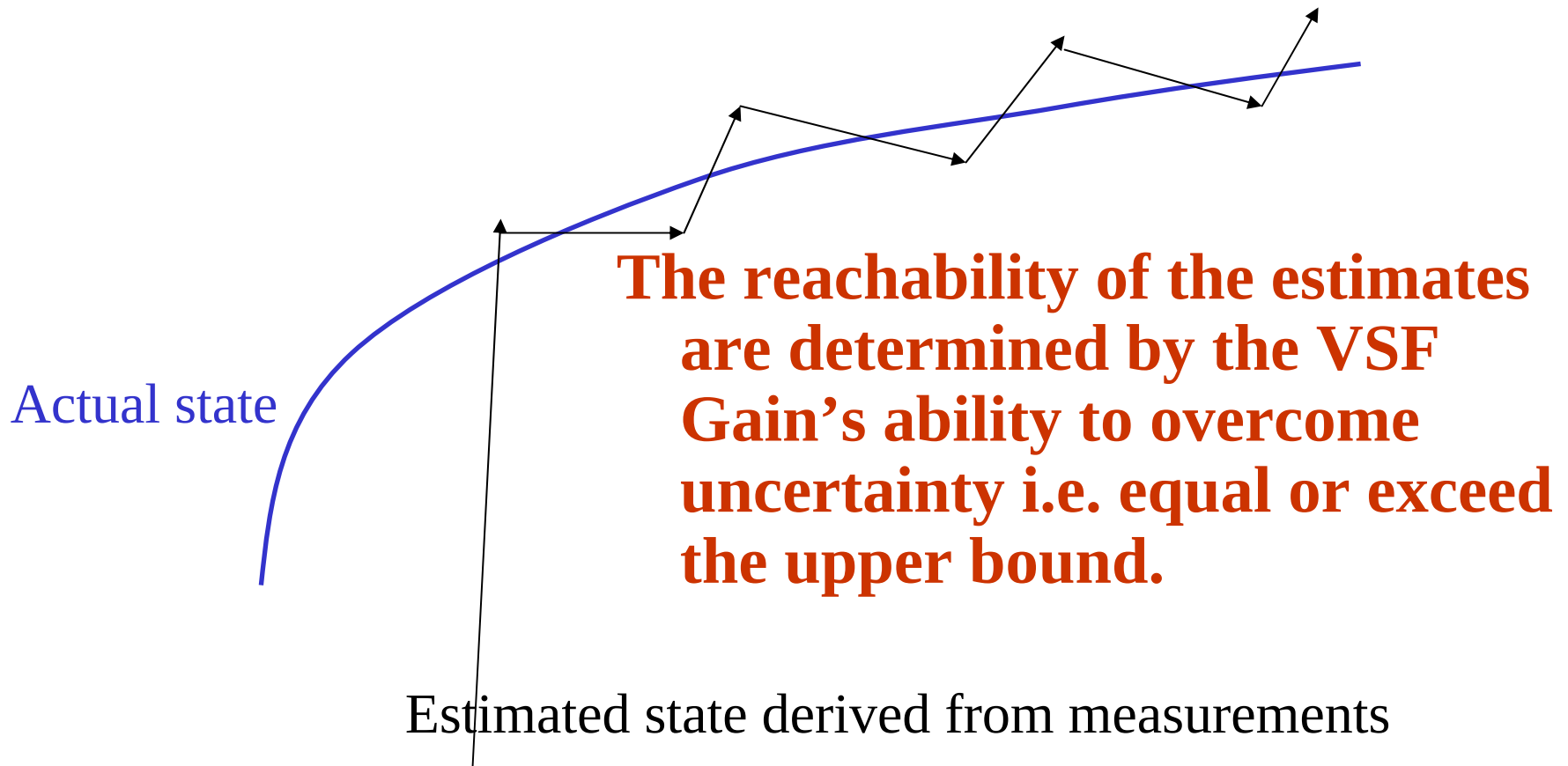
- The following gain structure results in convergence to a boundary layer:

$$\mathbf{K}_k = \left| \hat{\mathbf{H}}^+ \left(\left| \mathbf{e}_{z_{k|k-1}} \right|_{\mathbf{ABS}} + \mathbf{Y} \left| \mathbf{e}_{z_{k-1|k-1}} \right|_{\mathbf{ABS}} \right) \right|_{\mathbf{ABS}} \circ \mathbf{s}_{\mathbf{gn}}(\mathbf{e}_{z_{k|k-1}})$$

- Boundary layer width can be explicitly linked to uncertainties

VSF Action

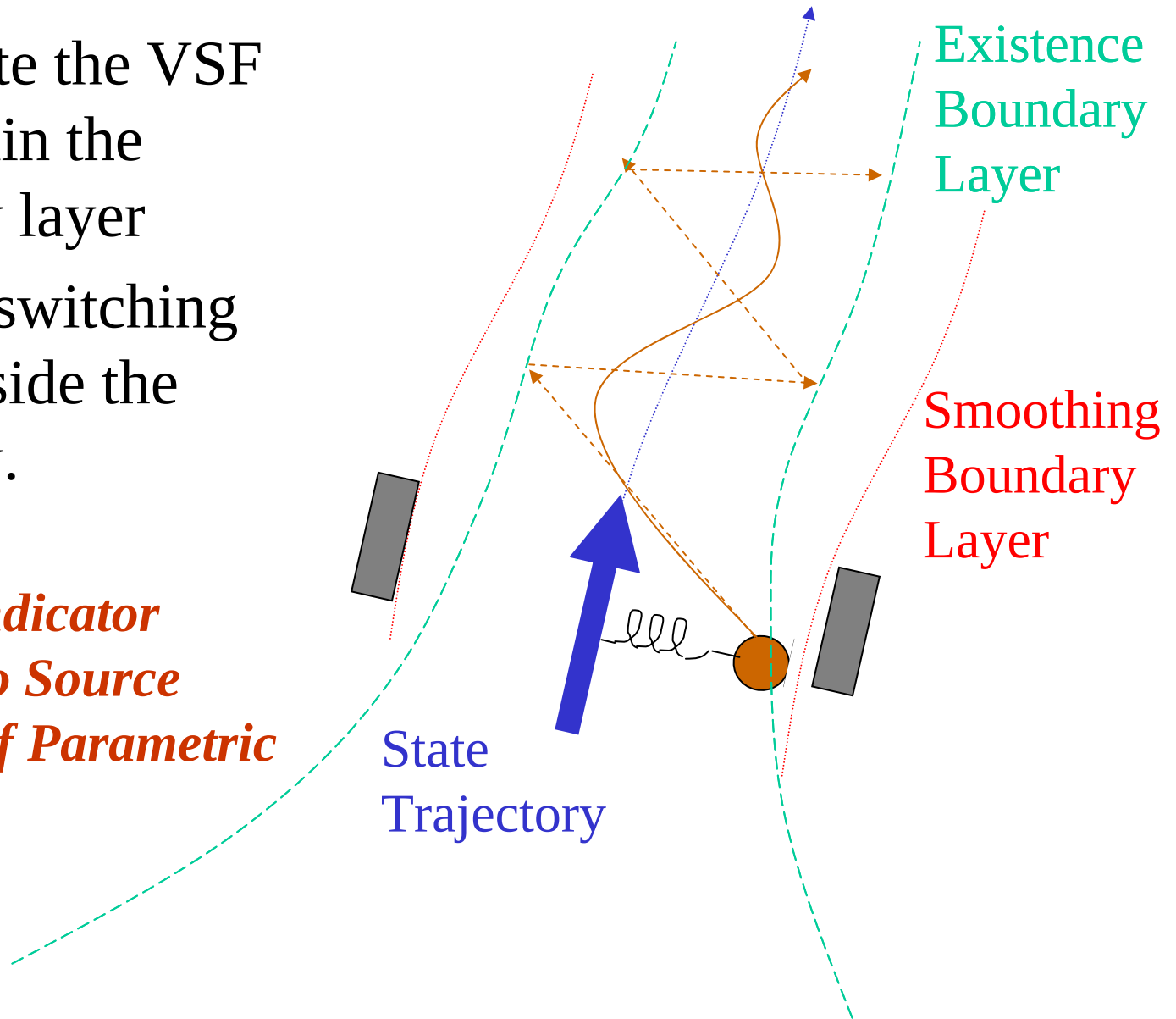
The VSF gain retain the estimated output to within a neighborhood of the measured output.



Smoothing Boundary Layer

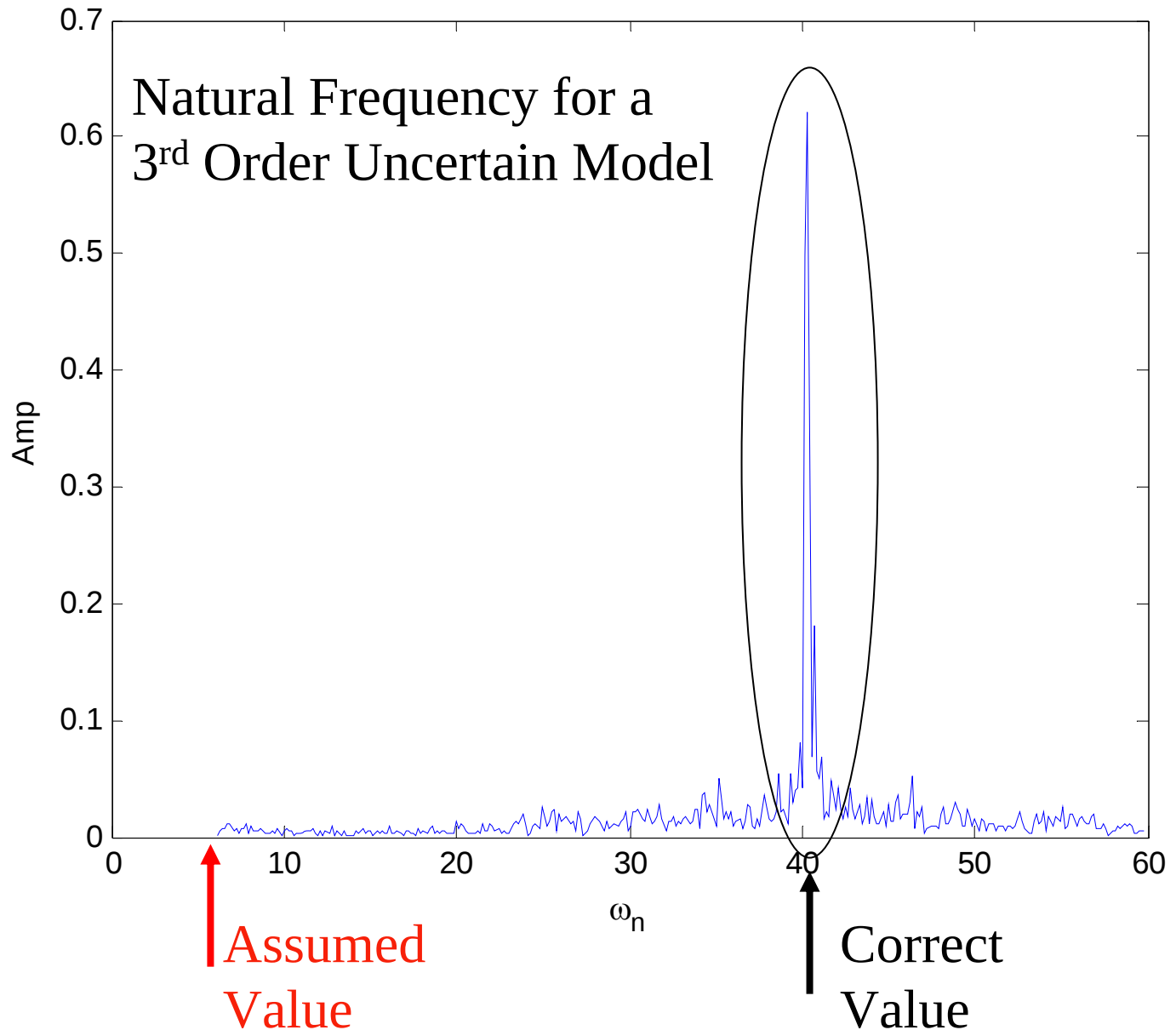
- Interpolate the VSF gain within the boundary layer
- revert to switching form outside the boundary.

*Secondary Indicator
Pertaining to Source
And Range of Parametric
Uncertainty*

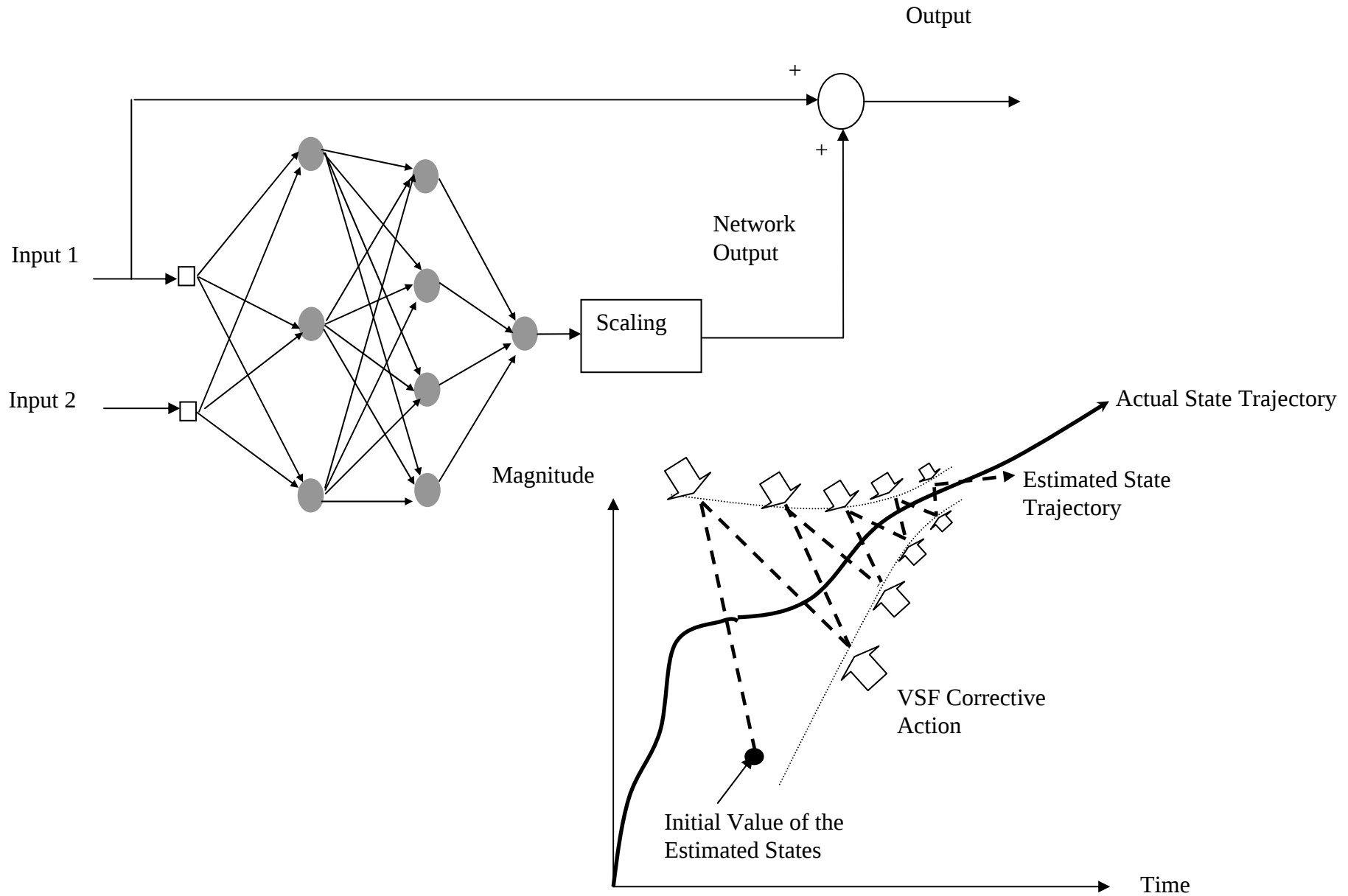


Chattering – Information Content

FFT for Chateering of e_3



Network Elements or Neurons



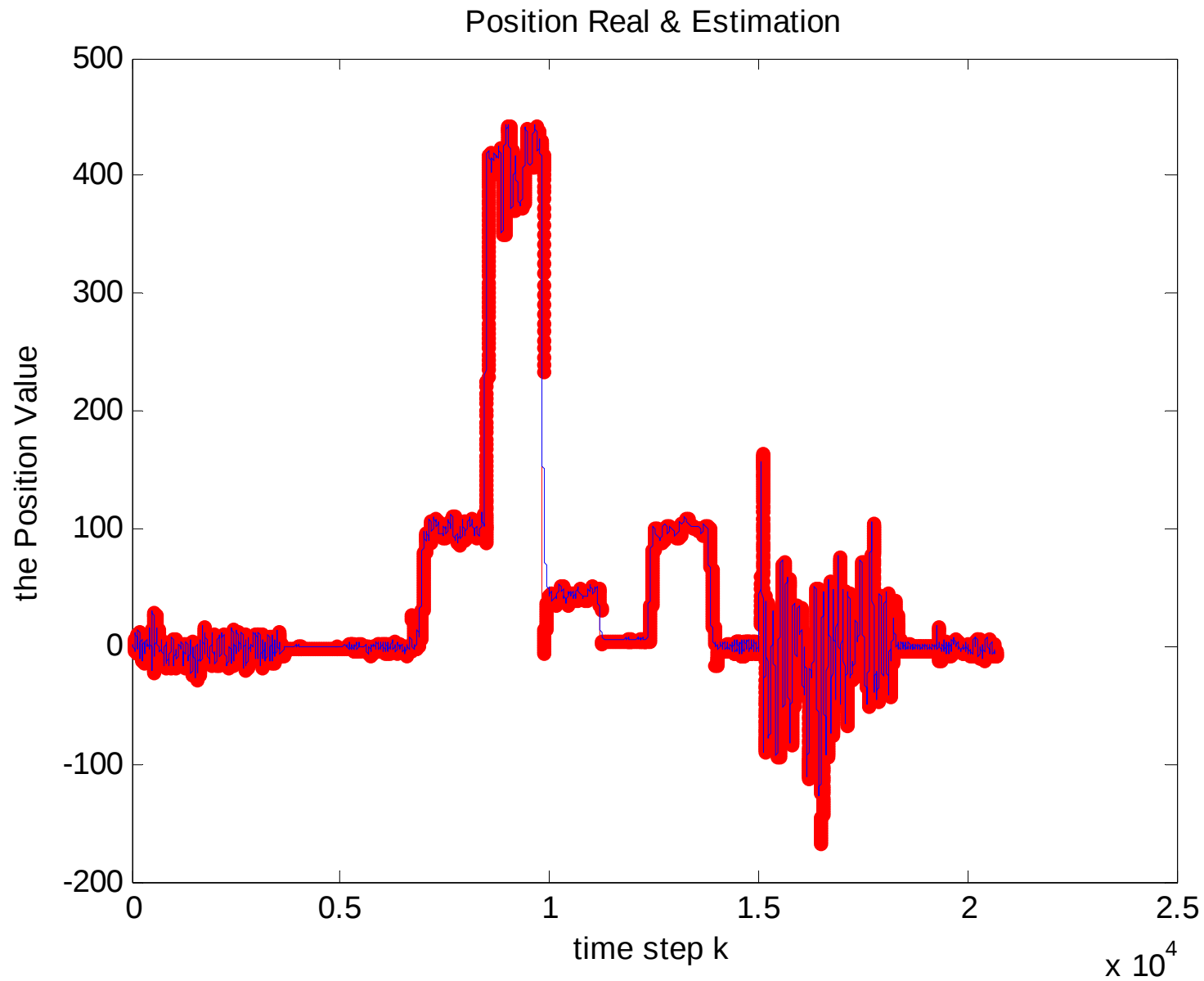
Example

State and Parameter Tracking

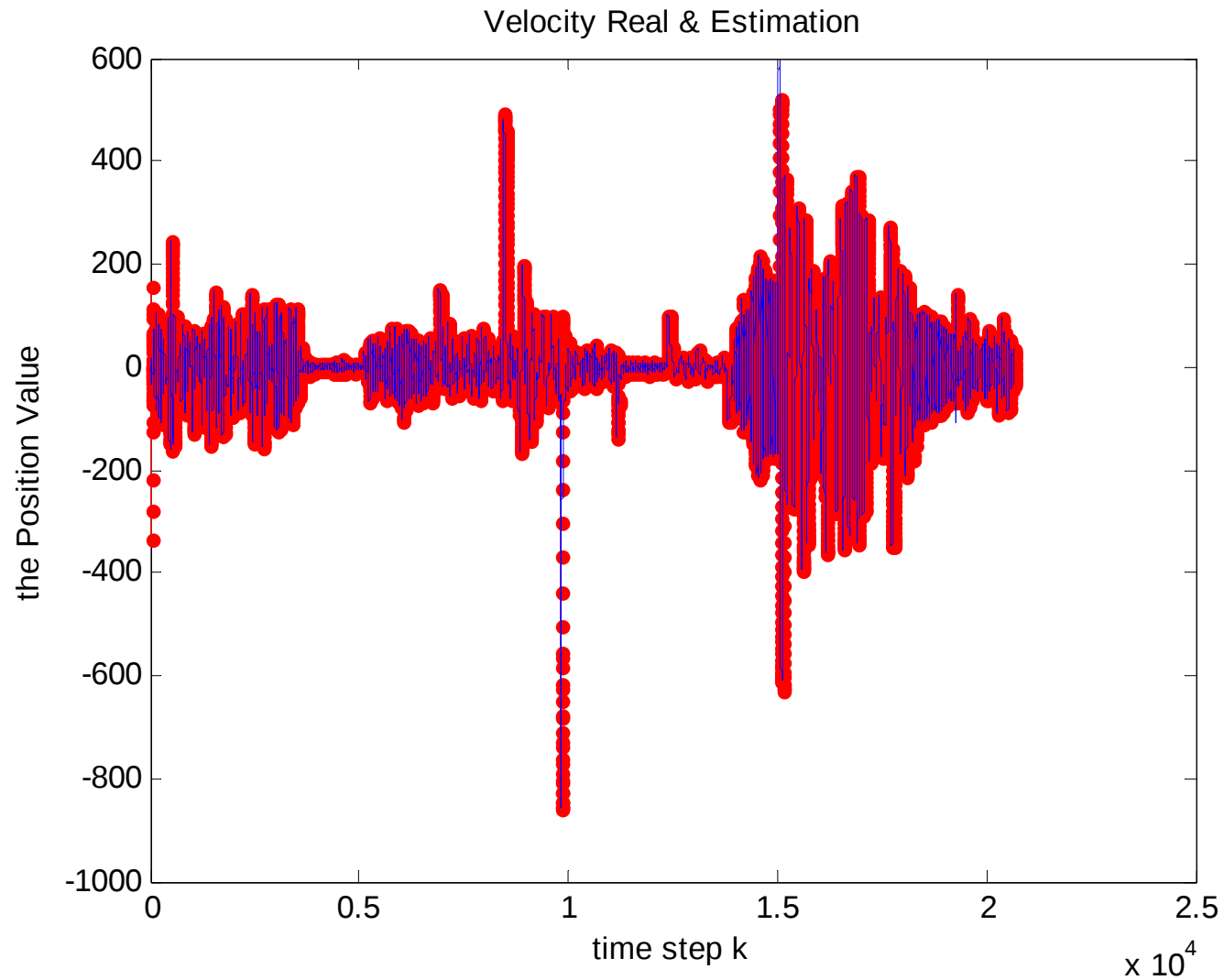
Uncertain 2nd order system

$$\frac{P\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

States - 1



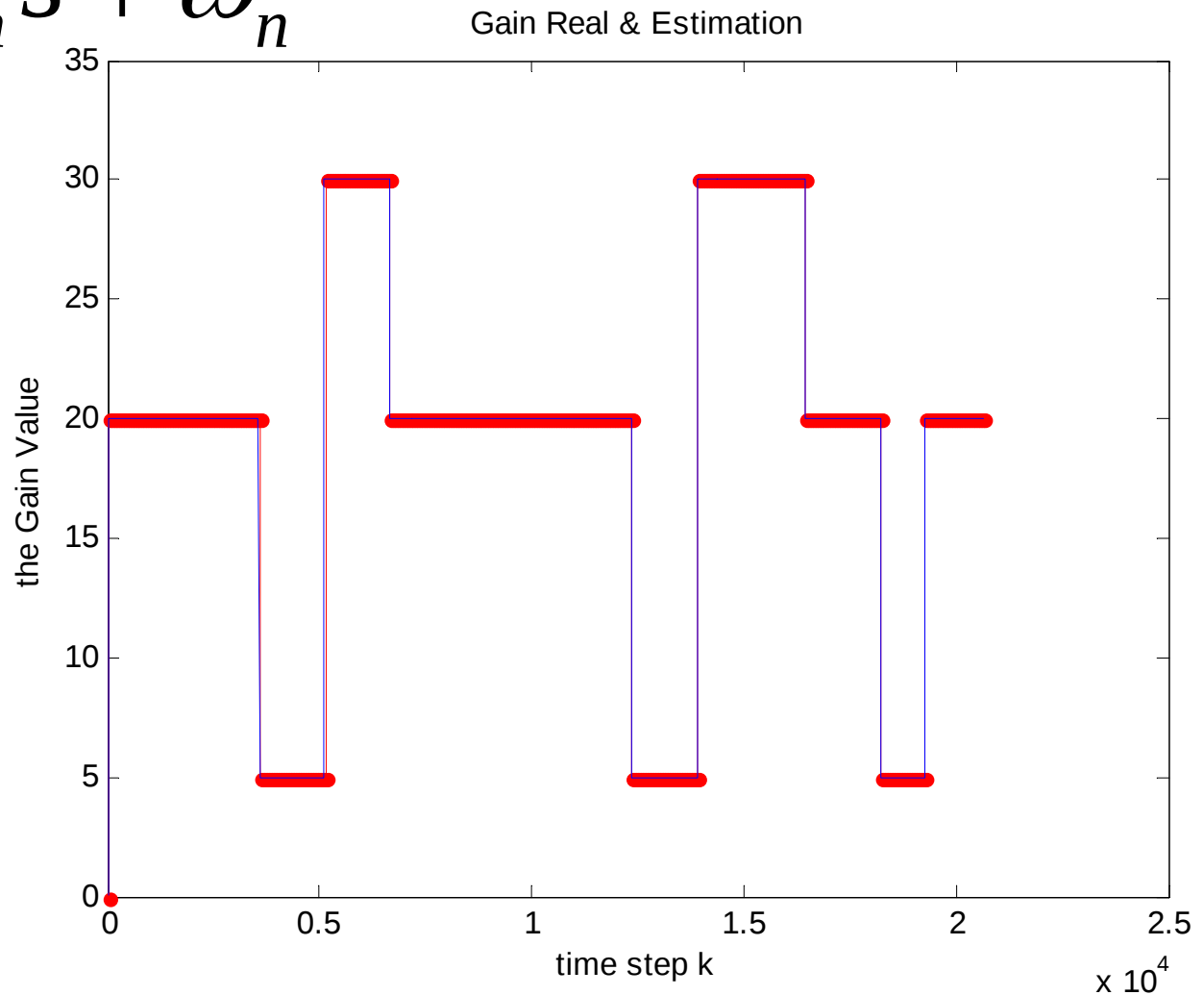
State 2



Parameter/State Estimation Error

$$P\omega_n^2$$

$$s^2 + 2\zeta\omega_n s + \omega_n^2$$

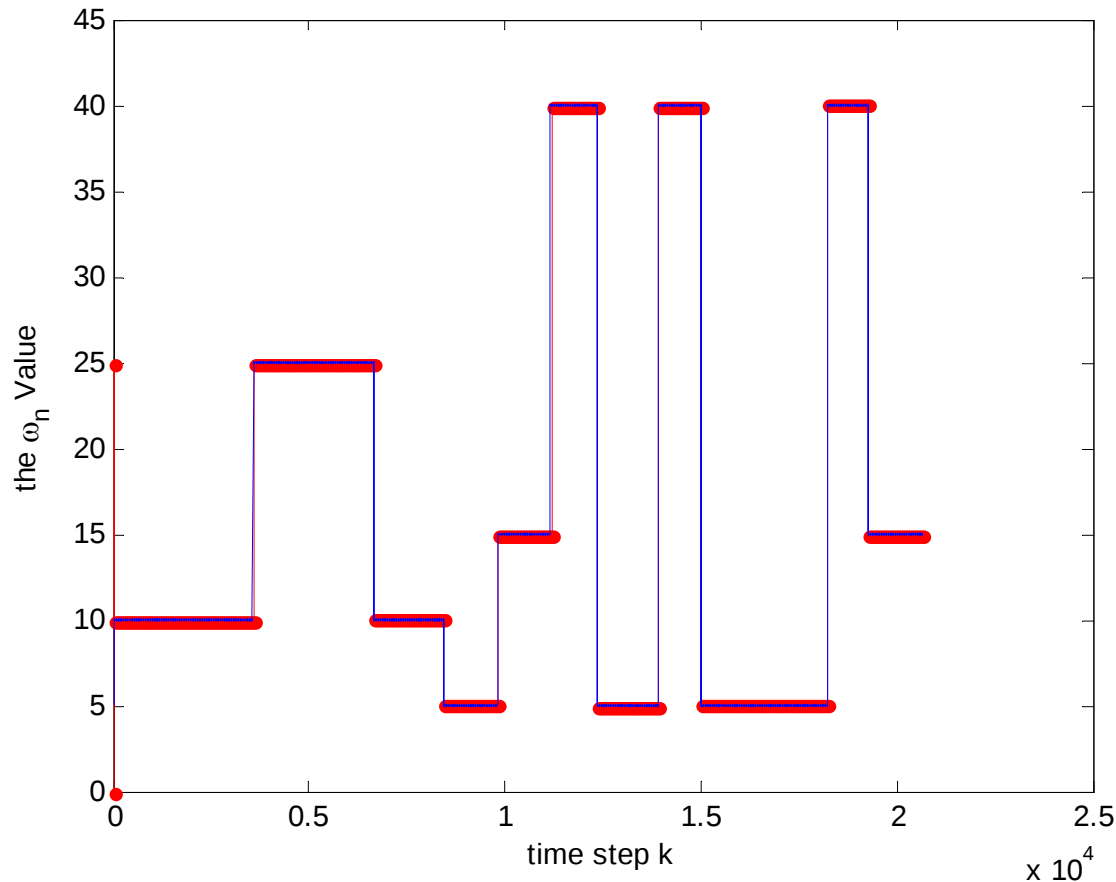


Parameter/State Estimation Error

$$P\omega_n^2$$

$$s^2 + 2\zeta\omega_n s + \omega_n^2$$

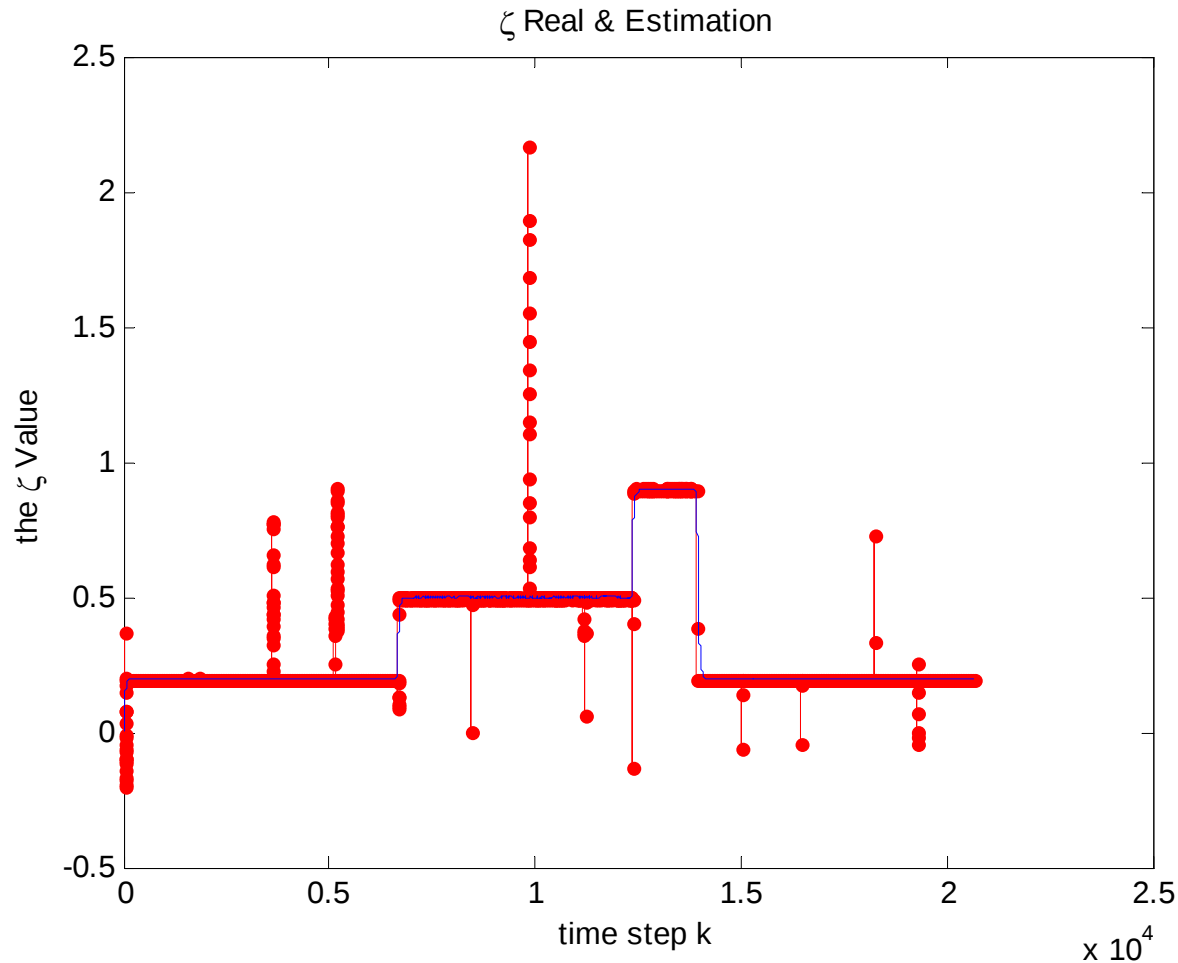
ω_n Real & Estimation



Parameter/State Estimation Error

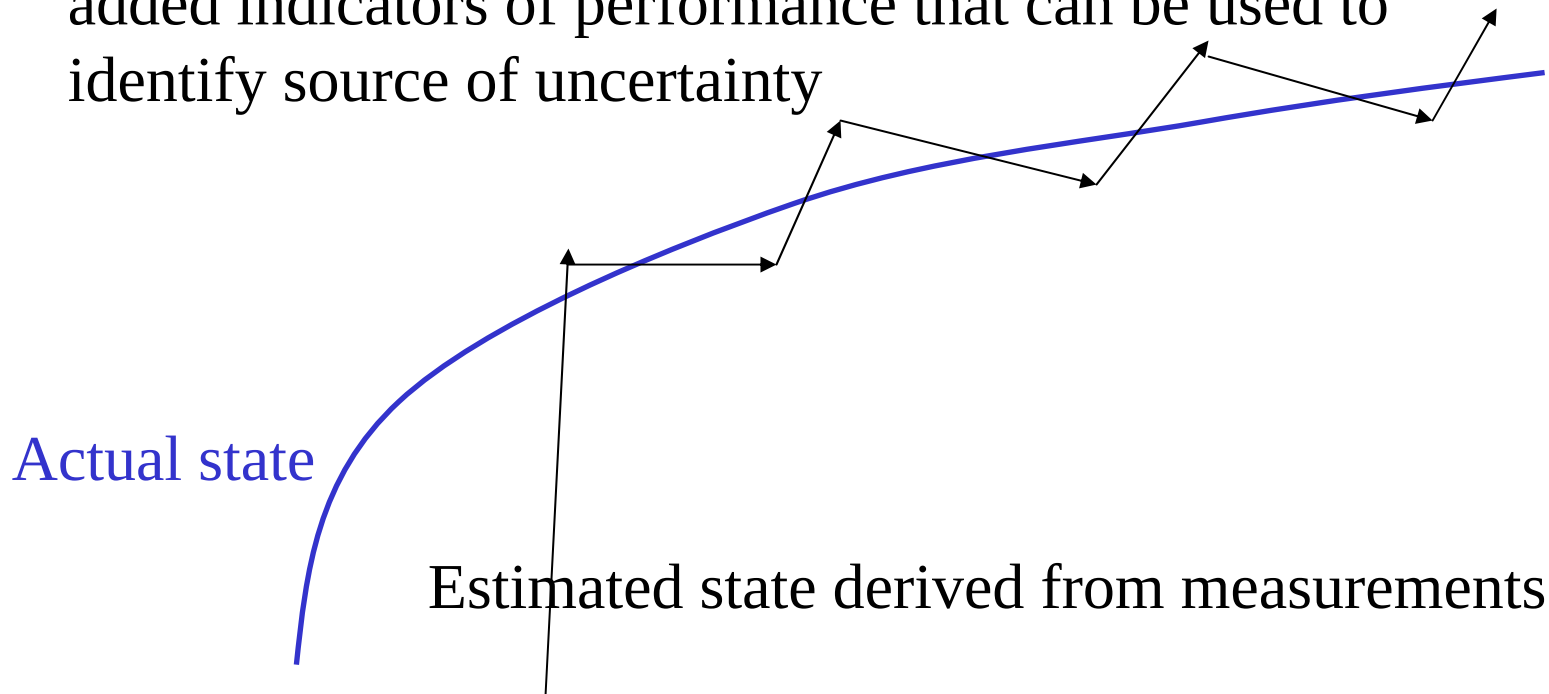
$$P\omega_n^2$$

$$s^2 + 2\zeta\omega_n s + \omega_n^2$$

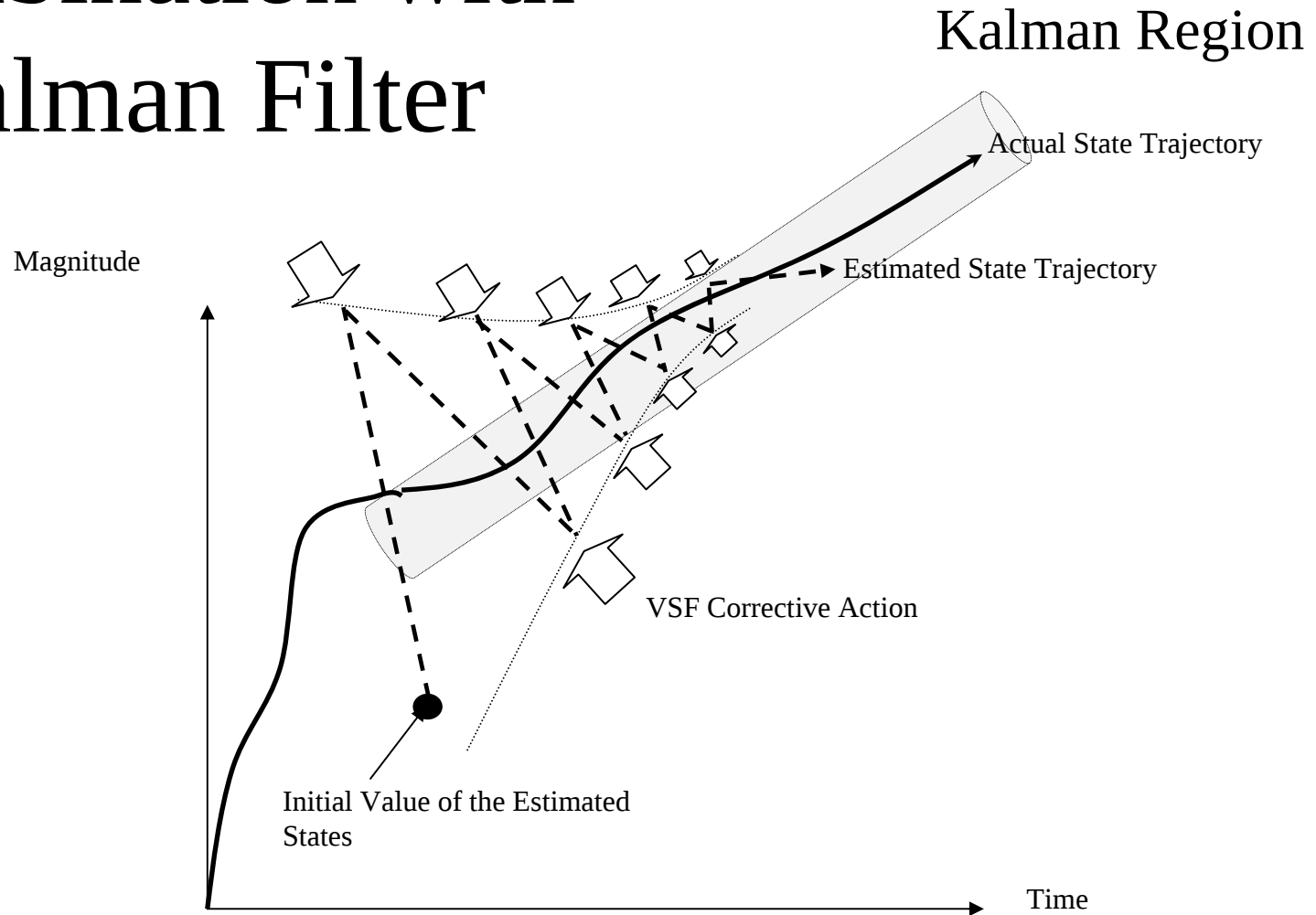


Parameter Estimation Using Variable Structure Filter

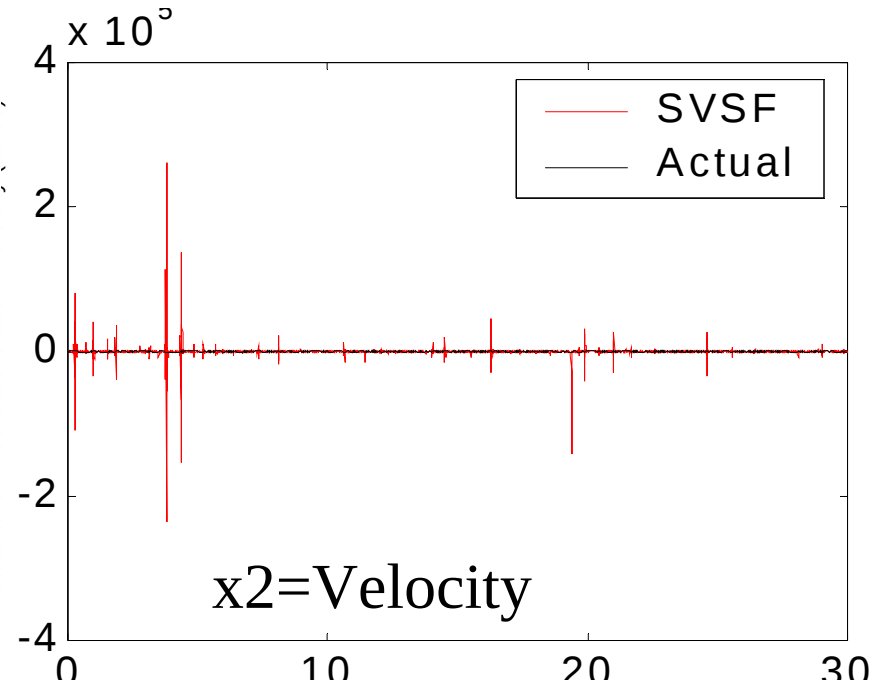
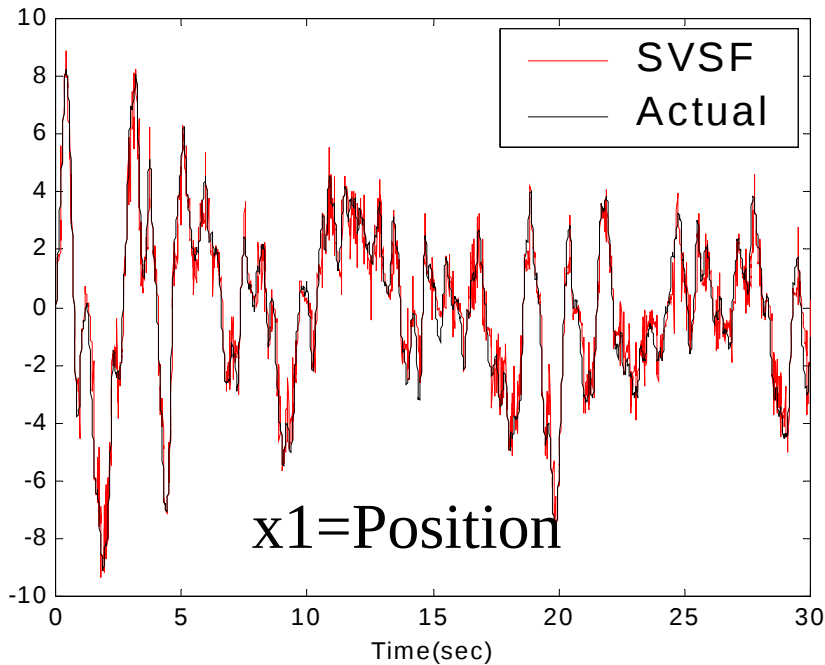
- Accommodates multiple parameter changes in the model and guarantees numerical stability of the estimation process given bounded uncertainties
- Allows performance recovery against uncertainty by adaptation and self tuning
- Minimizes the requirement for additional sensors due to its added indicators of performance that can be used to identify source of uncertainty



Combination with Kalman Filter



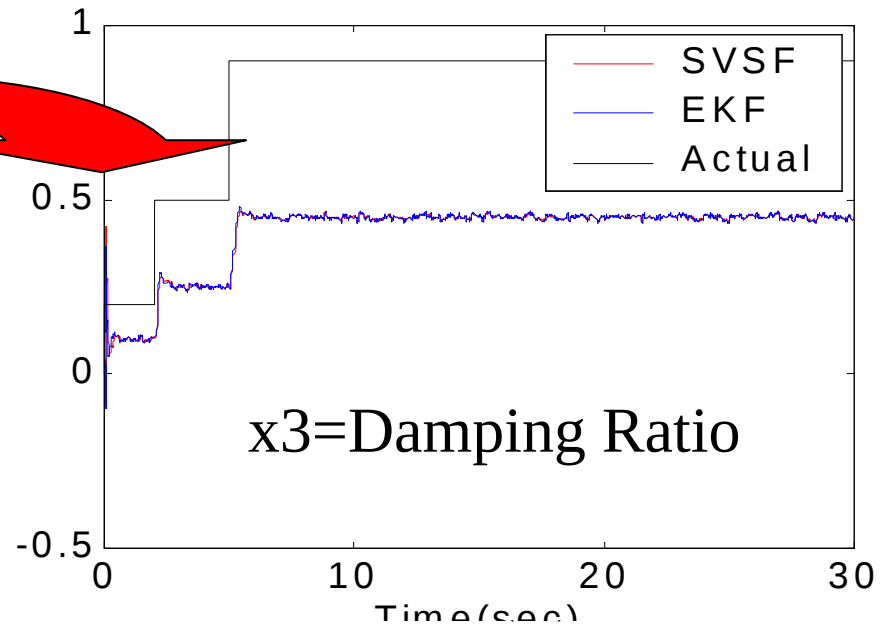
Parameter/State Estimation Error



$$\frac{12\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

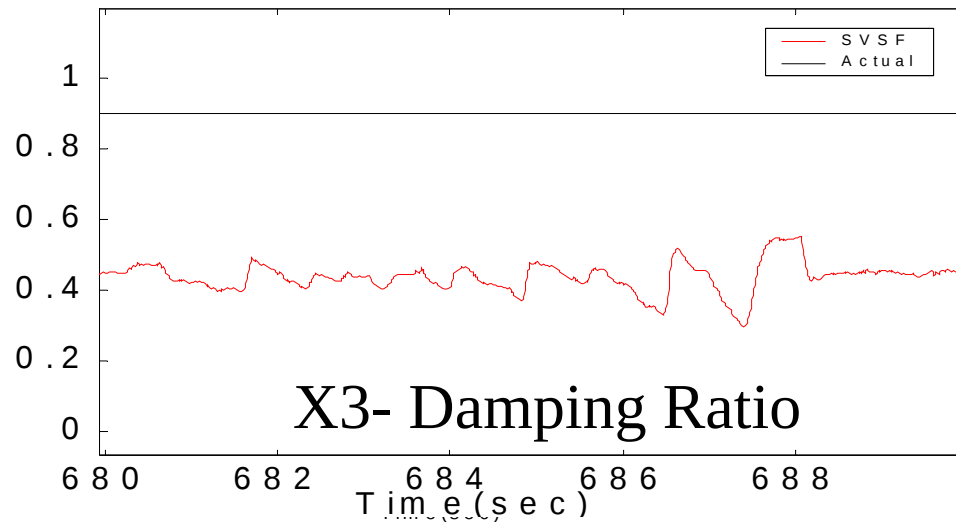
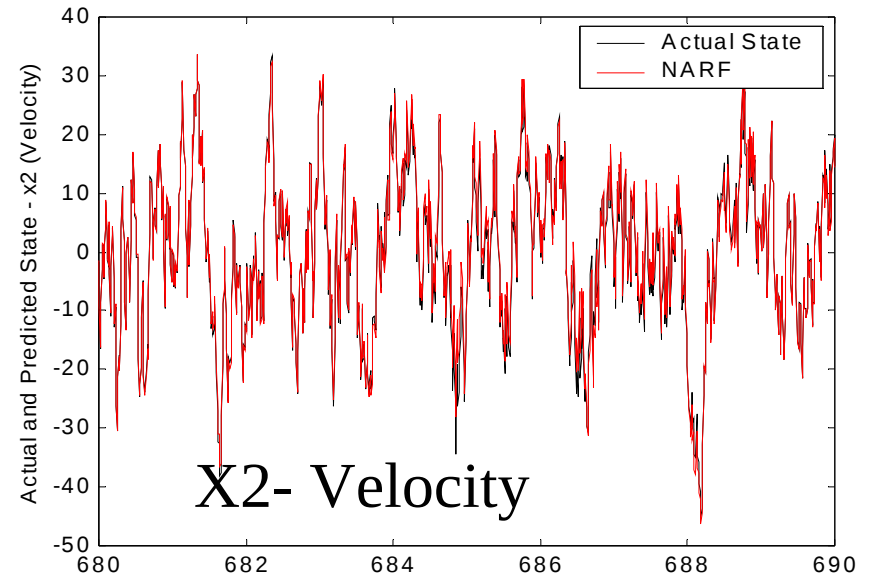
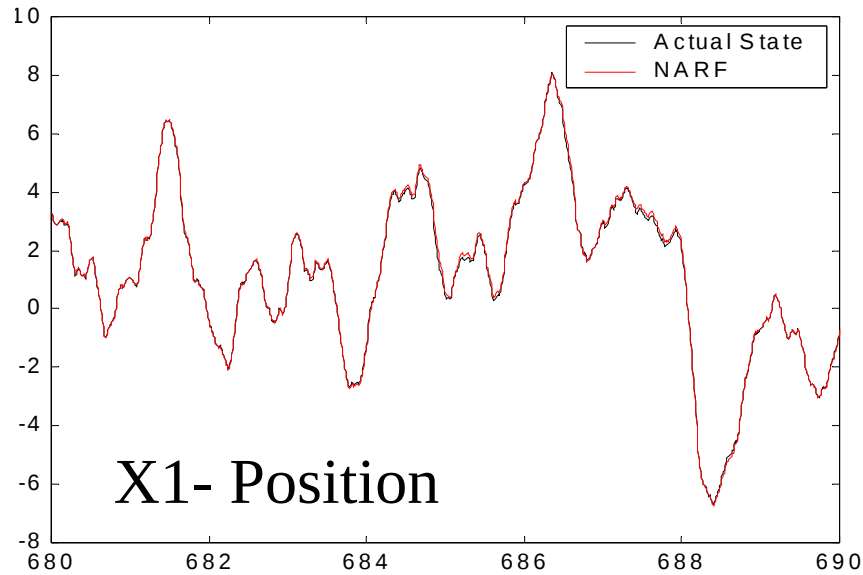
$\omega_n = 5$

$\hat{\omega}_n = 30$



Parameter/State Estimation

Uncertain Model



Advanced Technology Vehicles

Performance
Reliability

**Maintaining
Fuel Efficiency**

Driver Profile

Wear

Energy Recovery

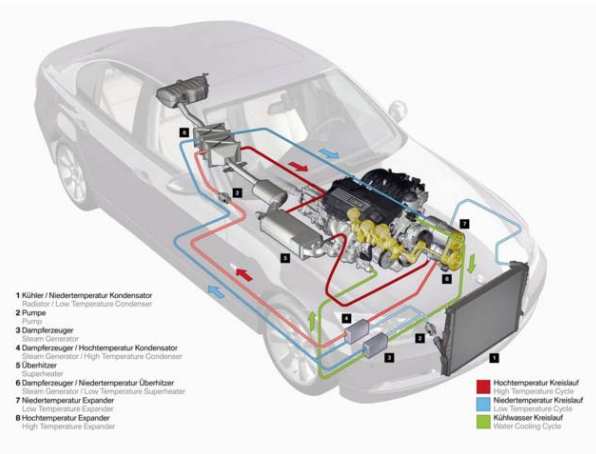
Energy Storage

Power Management

Architecture

Component Specs

Regeneration



Thermal Management



**Steady-State
Engine Operation**

Advanced Technology Vehicles

Performance
Reliability

**Cognitive
Car**

Regulate

Predict

Adapt

Learn

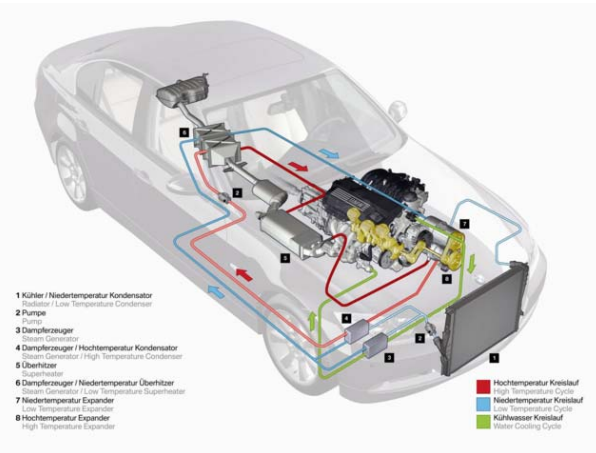
Sensing

Model

Architecture

Component Specs

Regeneration



Thermal Management



Steady-State
Engine Operation

Evolution of Intelligent Systems

Cognitive Car



Cognitive dynamic systems build up rules of behavior over time through learning from continuous experiential interactions with the environment, and thereby deal with environmental uncertainties.

**Intelligence &
Cognitive Dynamics
In Group/Parallel Systems**

**Power train
Control
Applications**

Diagnostics

Safety

Energy Recovery

Hybrid Architecture

Questions