

Lecture – 38

*Robust Nonlinear Control of Aircrafts Using  
Neuro-adaptive Augmented Dynamic Inversion*

*Dr. Radhakant Padhi*

*Asst. Professor*

*Dept. of Aerospace Engineering*

*Indian Institute of Science - Bangalore*



## References

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**Radhakant Padhi, Narayan P. Rao, Siddharth Goyal and Abha Tripathi, “*A Model-Following Neuro-Adaptive Approach for Robust Control of High Performance Aircrafts*”, *Automatic Control in Aerospace*, Vol. 3, No. 1, May 2010.**

# Problem Statement

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- **Pilot Commands:**

- Case -1: Longitudinal

(Roll Rate = 0, **Normal Acceleration**, Lateral Acceleration = 0, Total Velocity)

- Case -2: Lateral

(Roll Rate, **Height**, Lateral Acceleration = 0, Total Velocity)

- **Turn Coordination:** *Lateral acceleration command is zero*

- **Goal:**

- Airplane responds to the pilot commands “quickly” & “nicely”
- The control design should have sufficient robustness for parametric inaccuracies (mass, MI and aerodynamic coefficients)

**Note:** Essentially it is a **Robust Tracking** problem

# Airplane Dynamics: Six Degree-of-Freedom Model

Ref: Roskam J., Airplane Flight Dynamics and Automatic Controls, 1995

$$\dot{U} = VR - WQ - g \sin \Theta + \frac{1}{m}(F_{Ax} + F_{Tx})$$

$$\dot{V} = WP - UR + g \sin \Phi \cos \Theta + \frac{1}{m}(F_{Ay} + F_{Ty})$$

$$\dot{W} = UQ - VP + g \cos \Phi \cos \Theta + \frac{1}{m}(F_{Az} + F_{Tz})$$

$$\dot{P} = c_1 QR + c_2 PQ + c_3(L_A + L_T) + c_4(N_A + N_T)$$

$$\dot{Q} = c_5 PR - c_6(P^2 - R^2) + c_7(M_A + M_T)$$

$$\dot{R} = c_8 PQ - c_2 QR + c_4(L_A + L_T) + c_9(N_A + N_T)$$

$$\dot{\Phi} = P + Q \sin \Phi \tan \Theta + R \cos \Phi \tan \Theta$$

$$\dot{\Theta} = Q \cos \Phi - R \sin \Phi$$

$$\dot{\Psi} = (Q \sin \Phi + R \cos \Phi) \sec \Theta$$

$$\begin{bmatrix} \dot{x}_E \\ \dot{y}_E \\ \dot{z}_E \end{bmatrix} = \begin{bmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix} \begin{bmatrix} U \\ V \\ W \end{bmatrix}$$

$$\dot{h} = U \sin \Theta - V \cos \Theta \sin \Phi - W \cos \Theta \cos \Phi$$



# Aerodynamic Model (F-16)

Ref: Eugene A. Morelli , American Control Conference Philadelphia, Pennsylvania June 1998

$$C_{x_t} = C_{x_t}(\alpha, \delta_e) + C_{x_q}(\alpha)\tilde{q}$$

$$C_{y_t} = C_{y_t}(\beta, \delta_a, \delta_r) + C_{y_p}(\alpha)\tilde{p} + C_{y_r}(\alpha)\tilde{r}$$

$$C_{z_t} = C_{z_t}(\alpha, \beta, \delta_e) + C_{z_q}(\alpha)\tilde{q}$$

$$C_{l_t} = C_{l_t}(\alpha, \beta) + C_{l_p}(\alpha)\tilde{p} + C_{l_r}(\alpha)\tilde{r} + C_{l_{\delta_a}}(\alpha, \beta)\delta_a + C_{l_{\delta_r}}(\alpha, \beta)\delta_r$$

$$C_{m_t} = C_{m_t}(\alpha, \delta_e) + C_{m_q}(\alpha)\tilde{q} + C_{m_z}(x_{cg_{ref}} - x_{cg})$$

$$C_{n_t} = C_{n_t}(\alpha, \beta) + C_{n_p}(\alpha)\tilde{p} + C_{n_r}(\alpha)\tilde{r} + C_{n_{\delta_a}}(\alpha, \beta)\delta_a + C_{n_{\delta_r}}(\alpha, \beta)\delta_r + C_{n_y}(x_{cg_{ref}} - x_{cg})\left(\frac{\bar{c}}{b}\right)$$

$$\tilde{p} = pb / 2V_T \quad \tilde{q} = q\bar{c} / 2V_T \quad \tilde{r} = rb / 2V_T$$



## Equations in Desired Form

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$$\dot{X}_V = f_V (X) + [g_V (X)] U_c$$

$$\dot{X}_R = f_R (X) + [g_R (X)] U_c$$

where

$$X \triangleq [V_T \quad \alpha \quad \beta \quad P \quad Q \quad R \quad \Phi \quad \Theta \quad \Psi \quad X_E \quad Y_E \quad h]^T$$

$$X_V \triangleq [U \quad V \quad W]^T$$

$$X_R \triangleq [P \quad Q \quad R]^T$$

$$X_A \triangleq [\Phi \quad \Theta \quad \Psi]^T$$

$$U_c \triangleq [U_{AER} \quad U_T]^T \triangleq [[\delta_a \quad \delta_e \quad \delta_r] \quad T]^T$$

# Definitions and Goal

- Total Velocity:

$$V_T$$

- Roll Rate (about x-axis):

$$P$$



- Normal Acceleration:

$$n_z \triangleq -(F_z / m) = -(1/m)(F_{A_z})$$

- Lateral Acceleration:

$$n_y \triangleq (F_y / m) = (1/m)(F_{A_y})$$

- Goal:

$$P \rightarrow P^*, \quad n_z \rightarrow n_z^*, \quad n_y \rightarrow n_y^* = 0, \quad V_T \rightarrow V_T^*$$

where  $P^*, n_z^*, V_T^*$  are pilot commands

# Control Synthesis Procedure

- Define new variables:
 
$$a_z \triangleq n_z + \dot{W}, \quad a_z^* \triangleq n_z^* + \dot{W}$$

$$a_y \triangleq n_y - \dot{V}, \quad a_y^* \triangleq n_y^* - \dot{V}$$

- Key observation:  $\ddot{V} = \ddot{W} = 0$

$$\left( [n_z \ n_y]^T \rightarrow [n_z^* \ n_y^*]^T \right) \Leftrightarrow \left( [a_z \ a_y]^T \rightarrow [a_z^* \ a_y^*]^T \right)$$

- Known:

$$n_z = f_{n_z} + g_{n_z} U_c$$

$$\dot{P} = f_P + g_P U_c$$

$$n_y = f_{n_y} + g_{n_y} U_c$$

$$\dot{V}_T = f_{V_T} + g_{V_T} U_c$$

- In Wind Axis Frame:  $n_{wz} = f_{n_{wz}} + g_{n_{wz}} U_c$



# Control Synthesis Procedure: Longitudinal

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- Dynamics: 
$$a_z = UQ - VP + g \cos \Phi \cos \Theta$$
$$a_y = UR - WP - g \sin \Phi \cos \Theta$$
- Differentiate: 
$$\dot{a}_z = f_{a_z} + g_{a_z} U_c$$
$$\dot{a}_y = f_{a_y} + g_{a_y} U_c$$
- **Case-1: (Longitudinal) Pilot commands:**
  - Roll Rate:  $P^* = 0$
  - Normal Acceleration:  $n_z^*$
  - Lateral Acceleration:  $n_y^* = 0$
  - Total Velocity:  $V_T^*$

# Control Synthesis Procedure: Longitudinal

- **Define (Errors):**

**Fast Variables**  $\hat{X}_T \triangleq [\hat{P} \quad \hat{a}_z \quad \hat{a}_y]^T \triangleq [(P - P^*) \quad (a_z - a_z^*) \quad (a_y - a_y^*)]^T$

**Slow Variables**  $\hat{V}_T \triangleq [(V_T - V_T^*)]$

- Design a controller such that:

$$\dot{\hat{X}}_T + K \hat{X}_T = 0,$$

$$\dot{\hat{V}}_T + K_{V_T} \hat{V}_T = 0,$$

$$K \triangleq \text{diag} \left( \frac{1}{\tau_P}, \frac{1}{\tau_{n_z}}, \frac{1}{\tau_{n_y}} \right) \quad K_{V_T} \triangleq \text{diag} \left( \frac{1}{\tau_{V_T}} \right)$$



# Control Synthesis Procedure: Longitudinal

- Calculate  $[U_{AER}]$  for every  $\Delta t$  time

$$U_{AER} = [A_U]^{-1} b_U$$

$$A_U \triangleq \begin{bmatrix} g_P \\ g_{a_z} \\ g_{a_y} \end{bmatrix} + [K] \begin{bmatrix} 0 & g_{n_z} & g_{n_y} \end{bmatrix}^T, \quad b_U \triangleq [K] \begin{bmatrix} P - P^* \\ n_z - n_z^* \\ n_y - n_y^* \end{bmatrix} - \begin{bmatrix} f_P \\ f_{a_z} \\ f_{a_y} \end{bmatrix}$$

- Calculate  $U_T$  for every  $5 \cdot \Delta t$  time

$$U_T = -[d_{V_T}]^{-1} c_{V_T}$$

$$c_{V_T} = \left[ (f_{V_T} + g_{V_T} U_{AER}) - \dot{V}_T^* + K_{V_T} (V_T - V_T^*) \right]$$

# Control Synthesis Procedure: Lateral

- **Case-2: (Lateral) Pilot commands:**

- Roll Rate:  $P^*$
- Height:  $h^*$
- Lateral Acceleration:  $n_y^* = 0$
- Total Velocity:  $V_T^*$



- Generation of  $\Theta^*$  command:

- Define  $\hat{h} \triangleq (h - h^*)$  and aim for  $\dot{\hat{h}} + (1/\tau_h)\hat{h} = 0$

$$[U \sin \Theta - V \cos \Theta \sin \Phi - W \cos \Theta \cos \Phi] - \dot{h}^* + (1/\tau_h)(h - h^*) = 0$$

- Solve for  $\Theta \triangleq \Theta^*$

# Control Synthesis Procedure: Lateral

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- Generation of  $Q^*$  command:

- Define  $\hat{\Theta} = (\Theta - \Theta^*)$  and aim for  $\dot{\hat{\Theta}} + (1/\tau_{\Theta})\hat{\Theta} = 0$

$$(Q \cos \Phi - R \sin \Phi) - \dot{\Theta}^* + (1/\tau_{\Theta})(\Theta - \Theta^*) = 0$$

- Solve for  $Q \triangleq Q^*$

- Control Computation:

- Define (Errors):

$$\text{Fast Variables} \quad \hat{X}_T \triangleq \begin{bmatrix} \hat{P} & \hat{Q} & \hat{a}_y \end{bmatrix}^T \triangleq \begin{bmatrix} (P - P^*) & (Q - Q^*) & (a_y - a_y^*) \end{bmatrix}^T$$

$$\text{Slow Variable} \quad \hat{V}_T \triangleq \begin{bmatrix} (V_T - V_T^*) \end{bmatrix}$$

# Control Synthesis Procedure: Lateral

- Design a controller such that

$$\begin{aligned} \dot{\hat{X}}_T + K \hat{X}_T &= 0 & \dot{\hat{V}}_T + K_{V_T} \hat{V}_T &= 0 \\ K &\triangleq \text{diag} \left( \frac{1}{\tau_P}, \frac{1}{\tau_Q}, \frac{1}{\tau_{n_y}} \right) & K_{V_T} &\triangleq \text{diag} \left( \frac{1}{\tau_{V_T}} \right) \end{aligned}$$

- After some algebra, Finally:

$$\begin{aligned} U_c &= \left[ \left[ A_U^{-1} b_U \right]^T \quad T \right]^T \\ A_U &\triangleq \left[ g_P^T \quad g_Q^T \quad \left( g_{a_y}^T + (1/\tau_{n_y}) g_{n_y}^T \right) \right]^T \\ b_U &\triangleq - \left[ f_P \quad f_Q \quad f_{a_y} \right]^T - K \left[ (P - P^*) \quad (Q - Q^*) \quad (f_{n_y} - n_y^*) \right]^T \end{aligned}$$

# Control Synthesis Procedure: Combined Longitudinal & Lateral

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- **Case-3: (Combined Longitudinal and Lateral)**

Pilot commands:

- Roll Rate (about velocity vector):  $P_w^*$
- Normal Acceleration:  $n_z^*$
- Lateral Acceleration:  $n_y^* = 0$
- Total Velocity:  $V_T^*$

- Generation of  $P^*$  command:

- Define  $P^*$  in the desired form as:  $P^* = f_{P^*} + g_{P^*} U_A$   
 where  $f_{P^*} = (1 / \cos \alpha (\cos \beta + \tan \beta \sin \beta))(P_w^* - R \sin \alpha (\sin \beta \tan \beta + \cos \beta)$

$$- \tan \beta (f_{n_{wz}} / V_T) + (g / V_T) \cos \gamma \cos \Phi$$

and  $g_{P^*} = (- \tan \beta / \cos \alpha (\cos \beta + \tan \beta \sin \beta))(g_{n_{wz}} / V_T)$

# Control Synthesis Procedure: Combined Longitudinal & Lateral

- Design a controller such that

$$\begin{aligned} \dot{\hat{X}}_T + K\hat{X}_T &= 0 & \dot{\hat{V}}_T + K_{V_T}\hat{V}_T &= 0, \\ K &\triangleq \text{diag}\left(\frac{1}{\tau_P}, \frac{1}{\tau_{n_z}}, \frac{1}{\tau_{n_y}}\right) & K_{V_T} &\triangleq \text{diag}\left(\frac{1}{\tau_{V_T}}\right) \end{aligned}$$

- After some algebra, Finally:

$$\begin{aligned} U_c &= \left[ [A_U^{-1} b_U]^T \quad T \right]^T \\ A_U &\triangleq \begin{bmatrix} g_P^T & g_{a_z}^T & g_{a_y}^T \end{bmatrix}^T + K \begin{bmatrix} -g_{P^*}^T & g_{n_z}^T & g_{n_y}^T \end{bmatrix}^T \\ b_U &\triangleq -\begin{bmatrix} f_P & f_{a_z} & f_{a_y} \end{bmatrix}^T - K \begin{bmatrix} (P - f_{P^*}) & (f_{n_z} - n_z^*) & (f_{n_y} - n_y^*) \end{bmatrix}^T \end{aligned}$$



## Alternative way to compute $P^*$ Command

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- Goal:  $\Phi \rightarrow \Phi^*$  (Note:  $\dot{\Phi}^* = 0$  for longitudinal case)
- Define  $\tilde{\Phi} \triangleq (\Phi - \Phi^*)$
- Desired error dynamics:

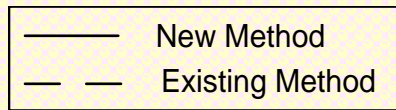
$$\dot{\tilde{\Phi}} + (1/\tau_\Phi)\tilde{\Phi} = 0, \quad \tau_\Phi > 0$$

- Substitute for  $\dot{\Phi}$  and solve for  $P \triangleq P^*$

$$P^* = \dot{\Phi}^* - \left[ Q \sin \Phi \tan \Theta + R \cos \Phi \tan \Theta + \frac{1}{\tau_\Phi} (\Phi - \Phi^*) \right]$$

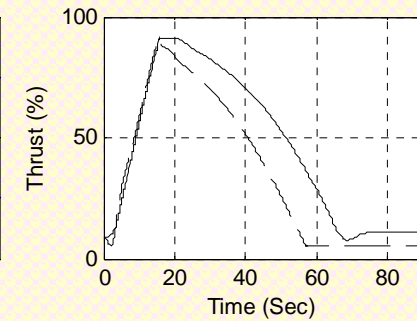
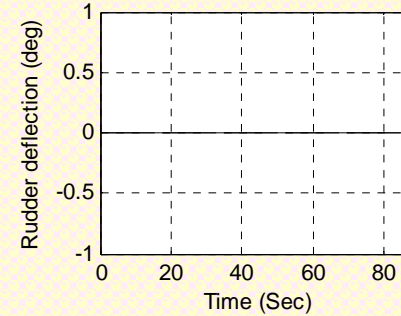
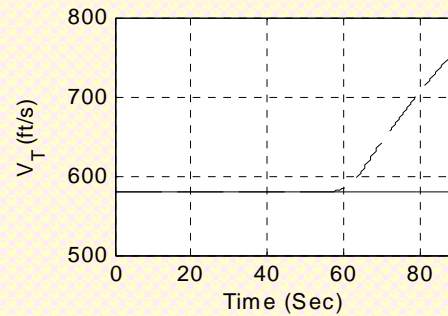
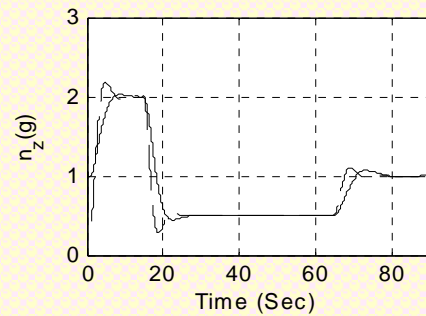
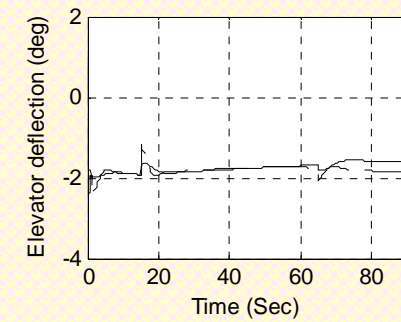
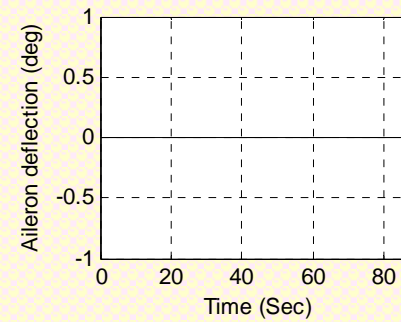
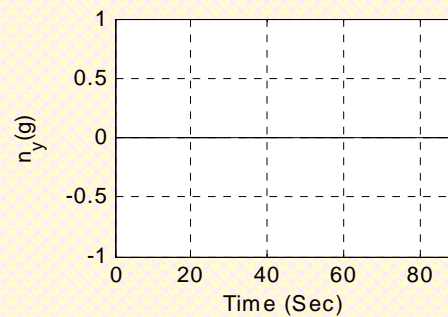
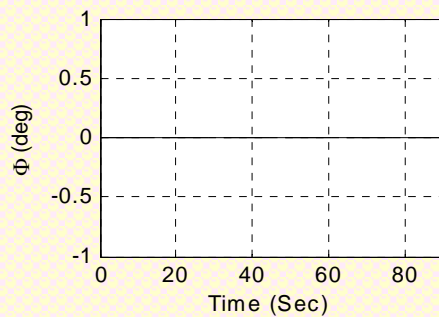
# Results: Longitudinal

(With  $\phi^*$  as a command)



Tracked Variables

Control Variables

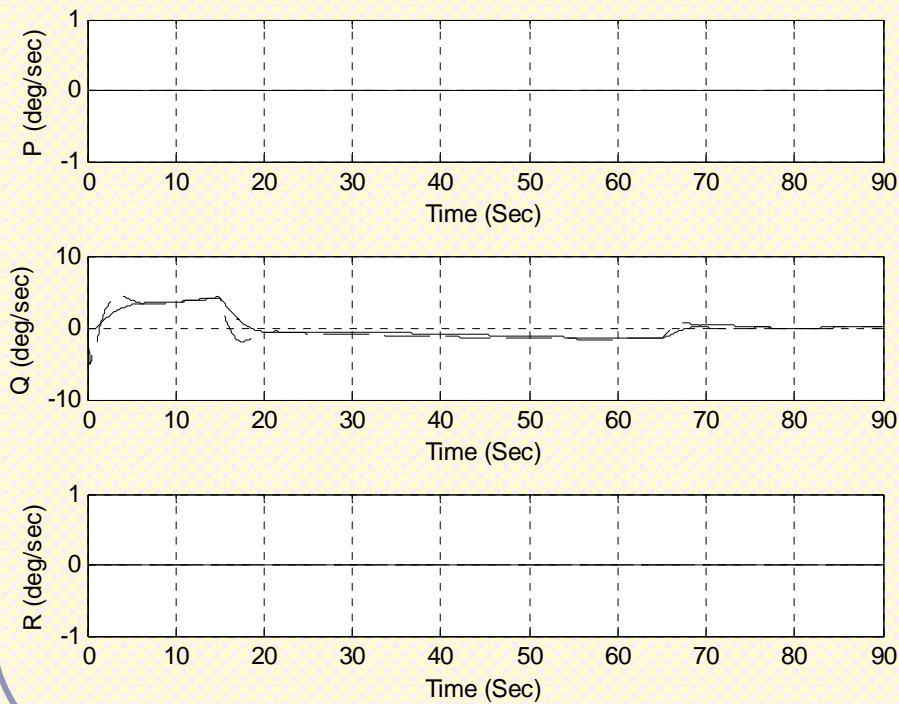


# Results: Longitudinal

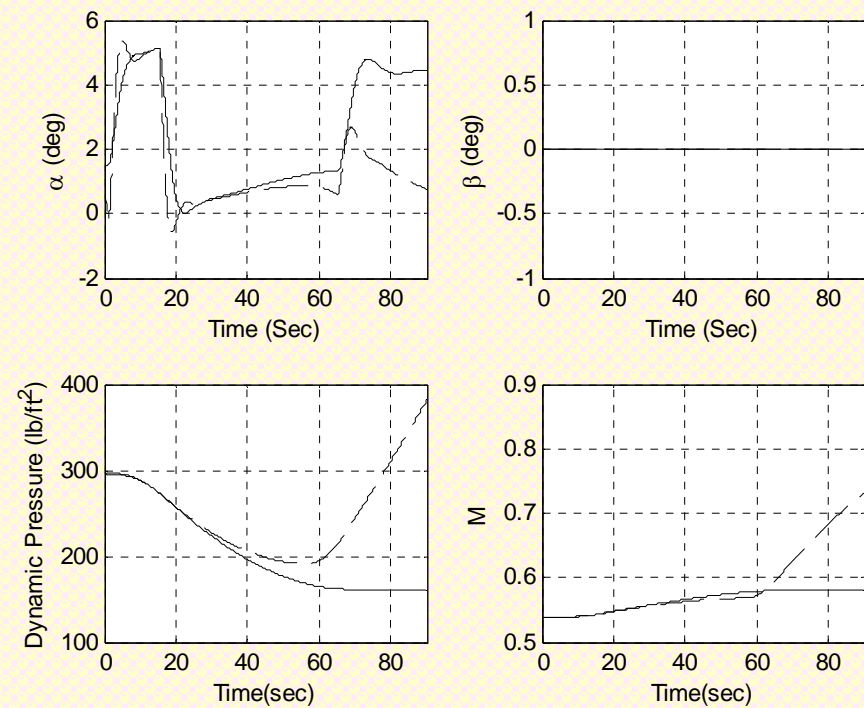
(With  $\phi^*$  as a command)



P, Q, R



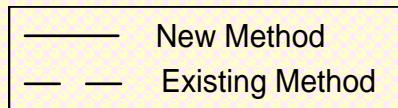
Aerodynamic Variables



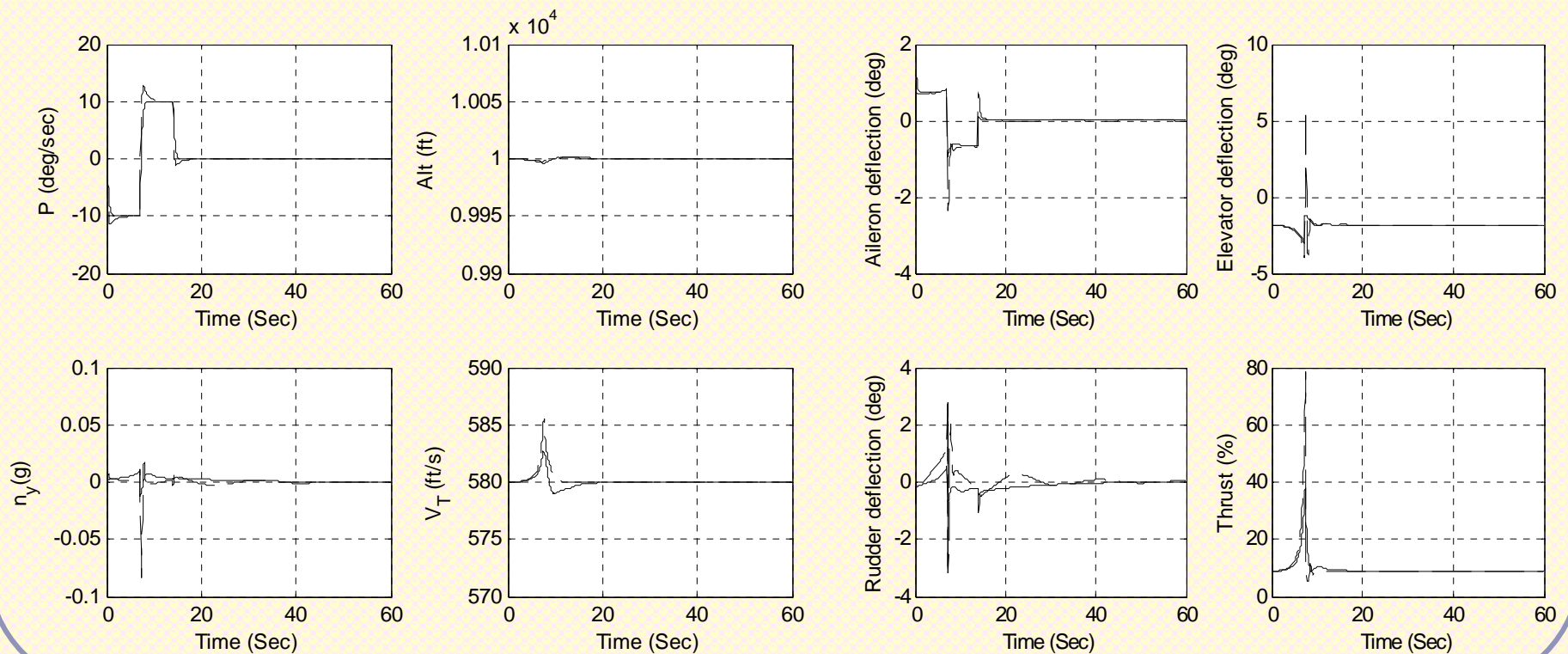
# Results: Lateral

(With  $P^*$  as a command)

Tracked variables



Control variables



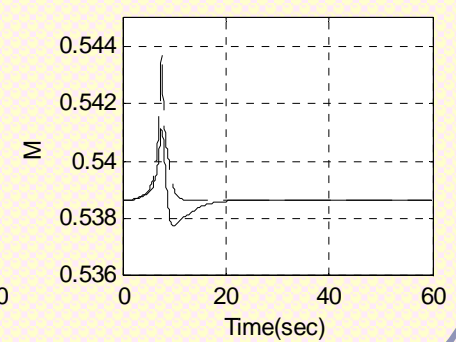
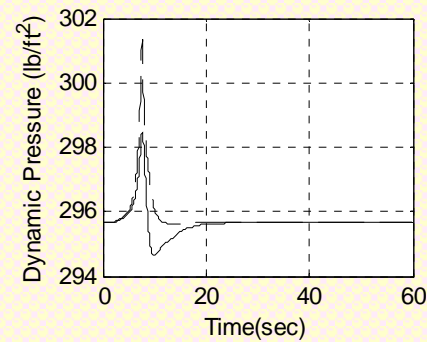
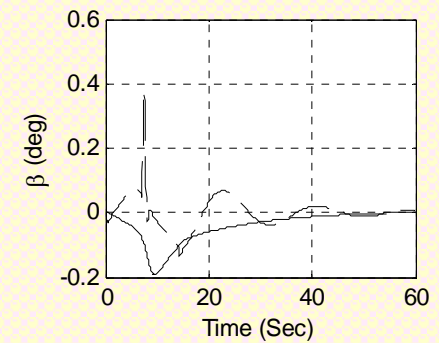
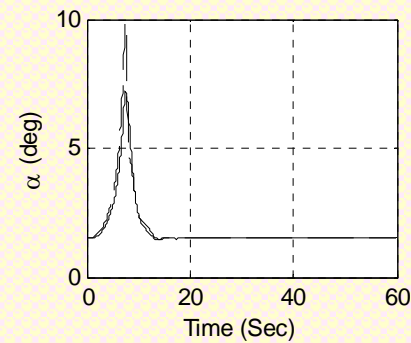
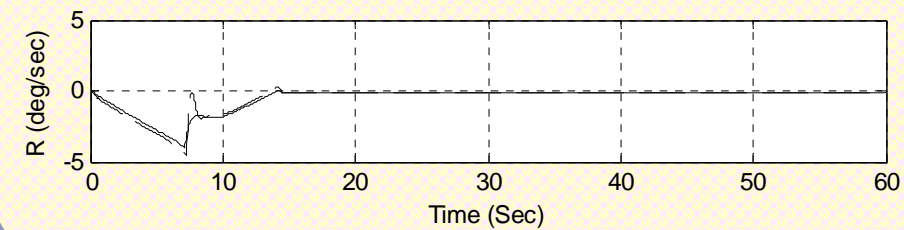
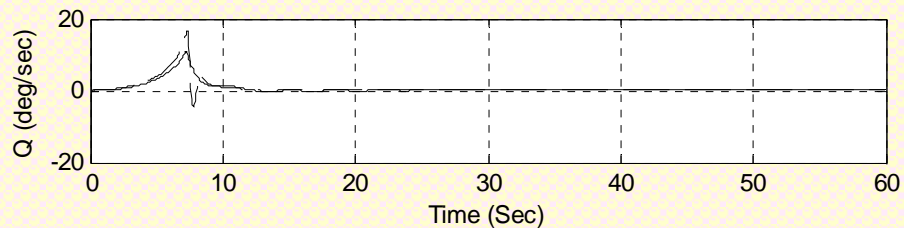
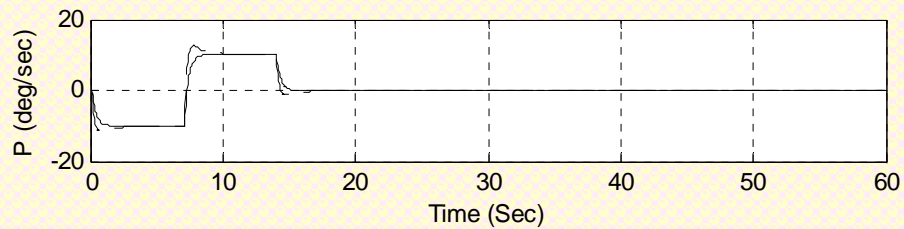
# Results: Lateral

(With  $P^*$  as a command)



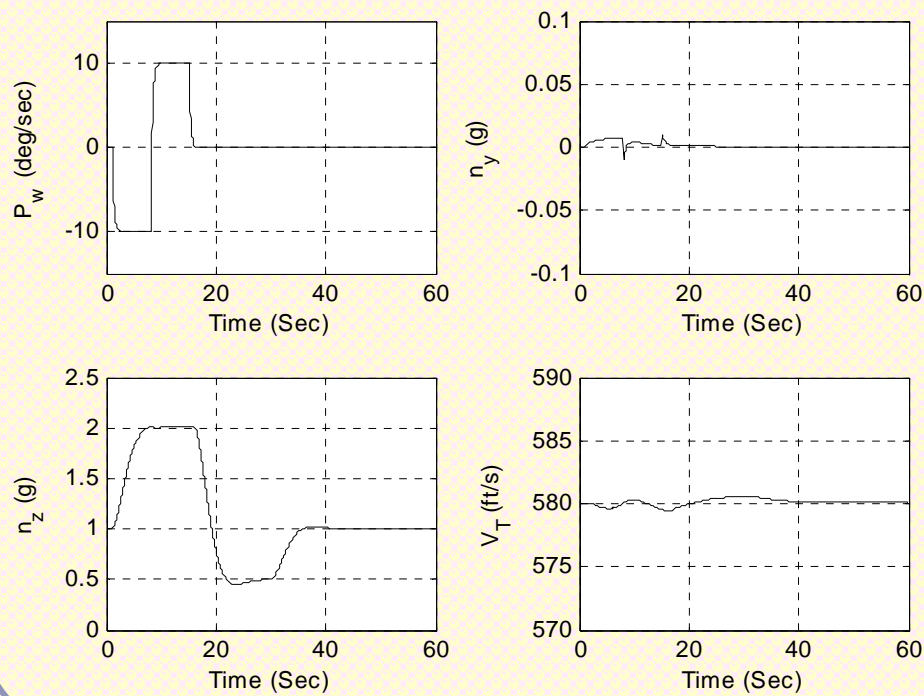
P, Q, R

Aerodynamic variables

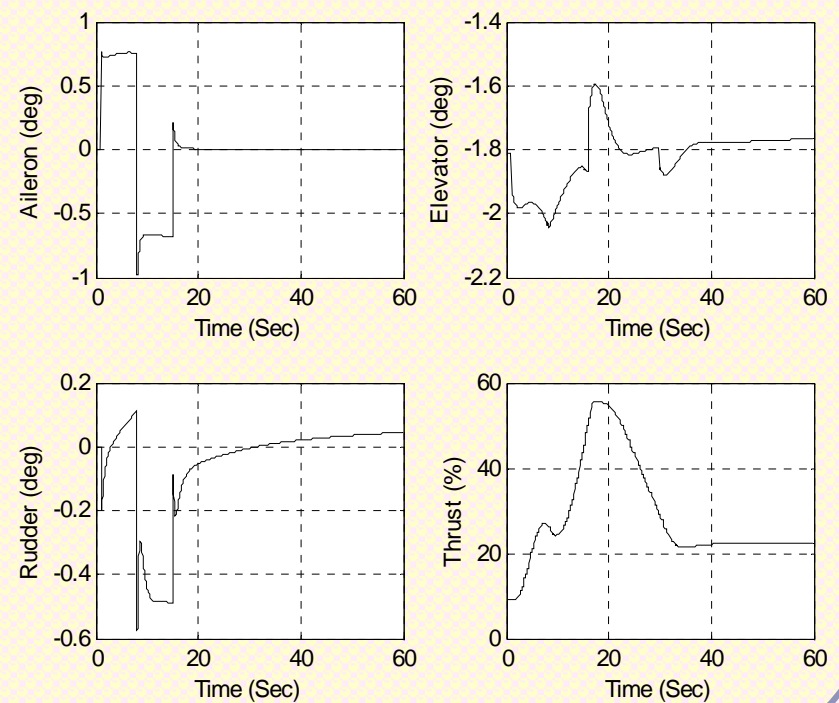


# Results: Combined Longitudinal and Lateral

## Tracked Variables



## Control Variables



# Summary: Nominal Controller

## Existing Method:

- Assumption:  $\dot{V} = \dot{W} = 0$   
 $\ddot{\Phi}^* = \ddot{\Theta}^* = \ddot{\Psi}^* = 0$
- More number of design parameters (11 & 12)
- Works

## New Method:

- Assumption:  $\ddot{V} = \ddot{W} = 0$
- Less number of design parameters (5 & 7)
- Works better...!
  - Lesser control magnitude
  - Smoother transient response
  - Better turn co-ordination

# *Neuro-Adaptive Control Design for Enhanced Robustness*



**Indian Institute of Science, Bangalore**





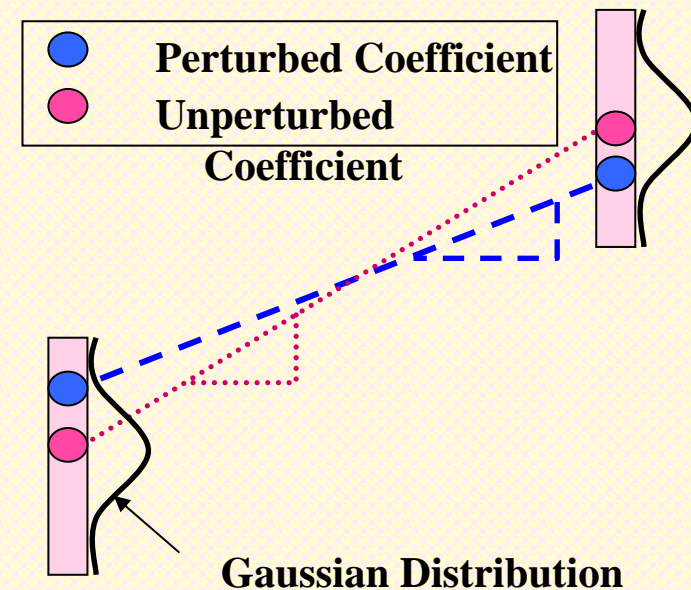
# Objective:

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**To increase the robustness of the  
“Nominal Controller” with respect to  
parameter and/or modeling  
uncertainties.**

# Robustness Study

- Parametric uncertainties was introduced in aero-coefficients and inertia parameters using Gaussian distribution around the nominal parameter values
- Since no analytical method for analyzing robustness behavior is available, a stochastic approach has been followed in this paper as an alternative.



# Longitudinal Mode

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Pilot Commands given:

$$V_T^* = V_{T0}; n_z^* = 2g;$$

( $\phi^*$  and  $n_y^*$  being maintained at zero);

Limits Imposed on the steady state error:

$$\phi: \pm 3^\circ; \quad n_y: \pm 0.05g; \quad V_T = \pm 1\%; \quad n_z = \pm 15\%$$

# Lateral Mode

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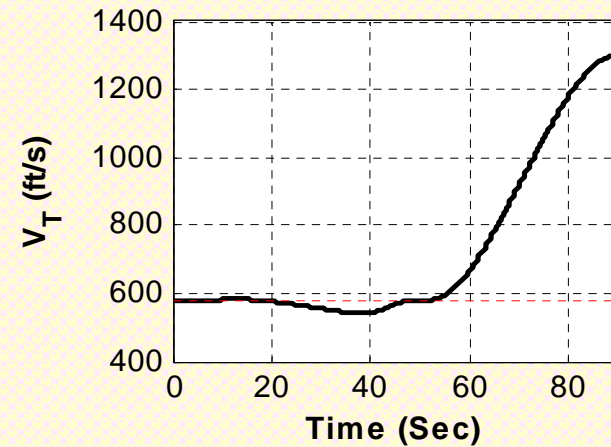
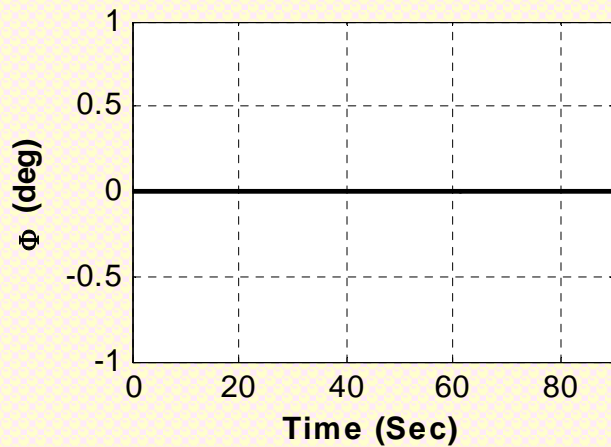
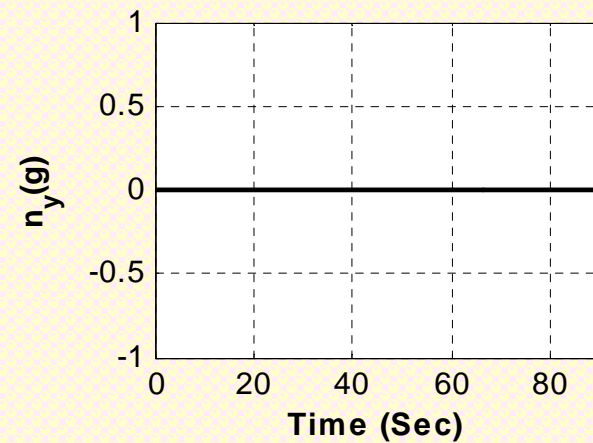
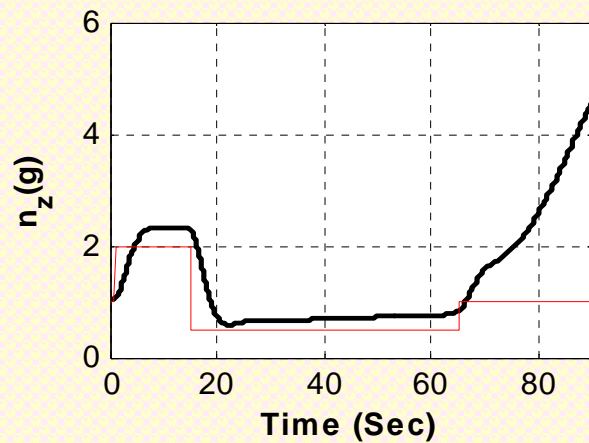
Pilot Commands Given:

$$\phi^* = -40^\circ; \quad V_T^* = V_{T0}; \quad h^* = h_0; \quad n_y^* = 0$$

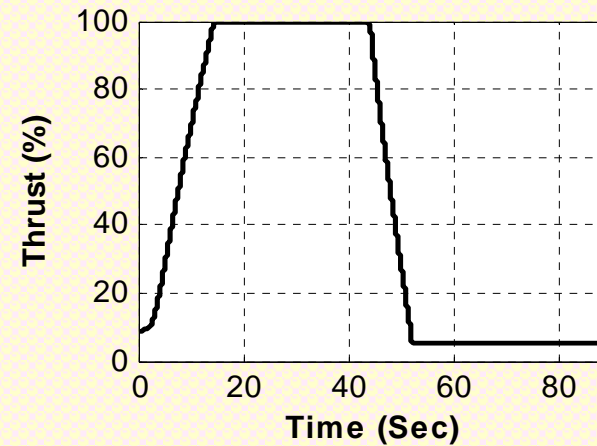
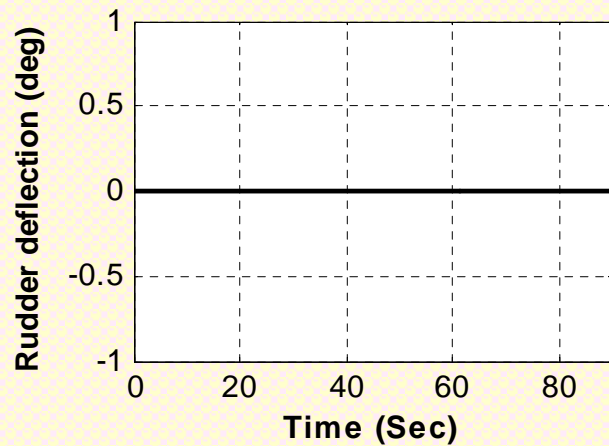
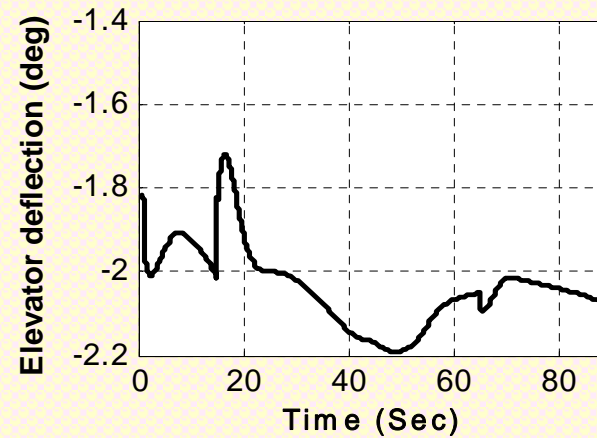
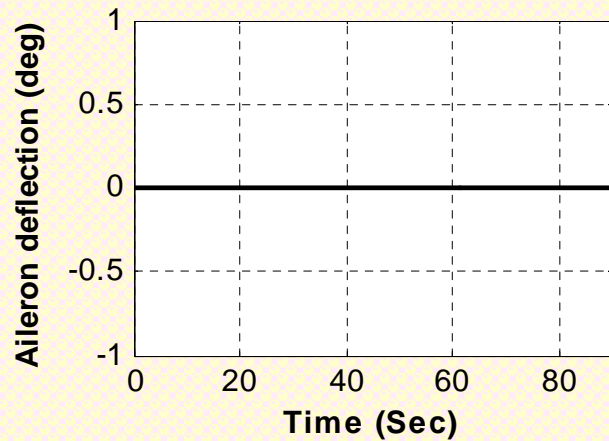
Limits Imposed on the steady state error:

$$h: \pm 1\%; \quad n_y : \pm 0.05g; \quad V_T = \pm 1\%; \quad \phi = \pm 10\%$$

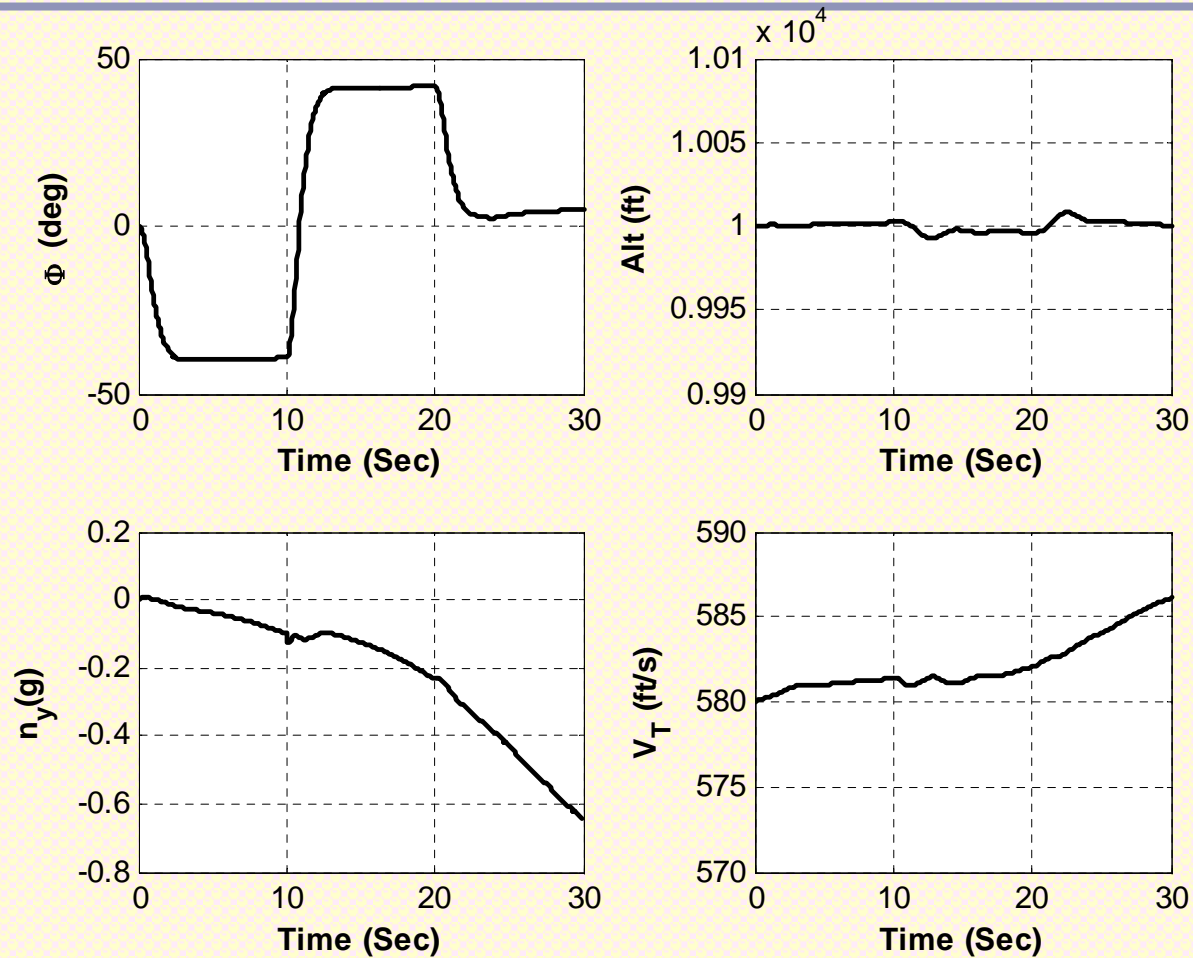
# Longitudinal Maneuver (Failure)



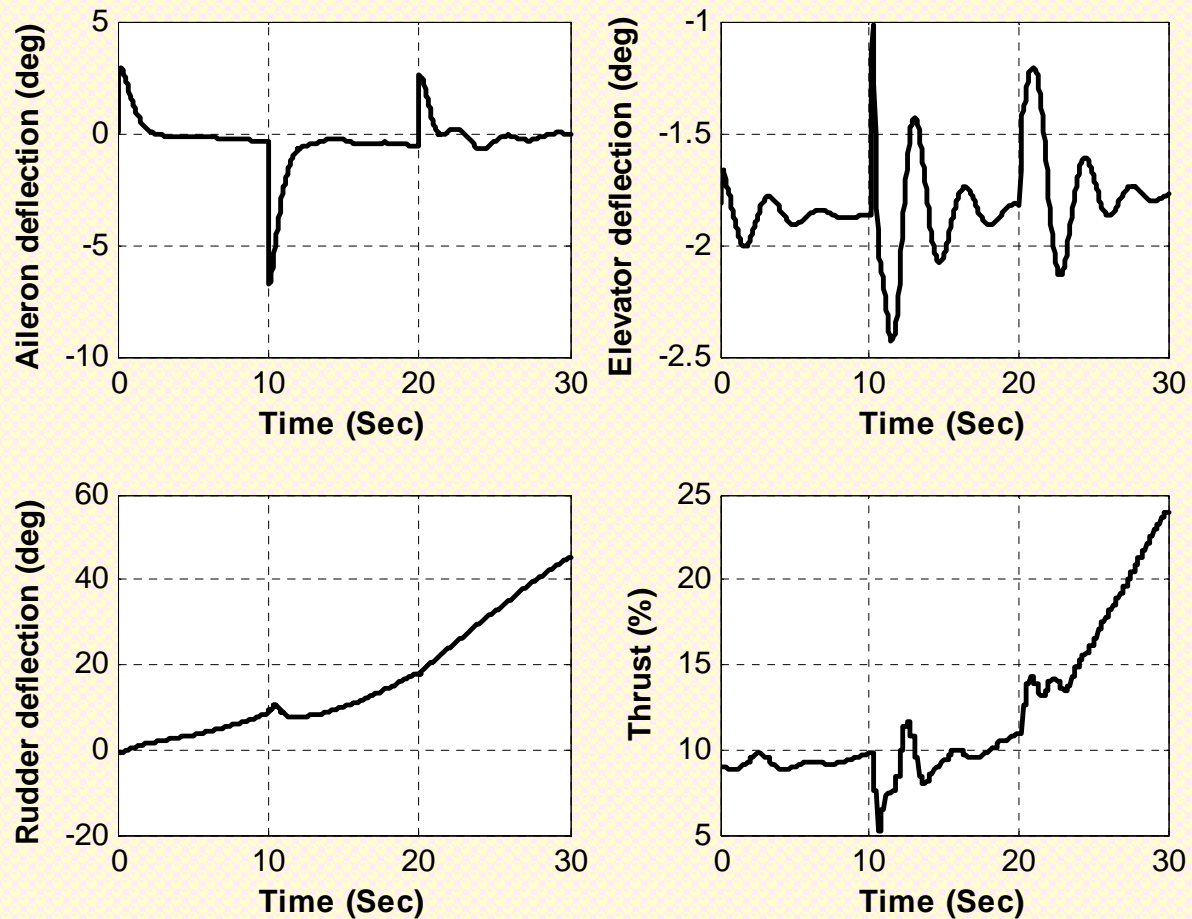
# Longitudinal Maneuver (Failure)



# Lateral Maneuver (Failure)



# Lateral Maneuver (Failure)





## Robustness in Longitudinal Mode

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<i>Aerodynamic coefficient</i>	<i>1%</i>	<i>1%</i>	<i>2%</i>	<i>2%</i>	<i>5%</i>	<i>5%</i>
<i>Mass and Inertia coefficient</i>	<i>5%</i>	<i>10%</i>	<i>5%</i>	<i>10%</i>	<i>5%</i>	<i>10%</i>
<i>% Success</i>	<i>100%</i>	<i>100%</i>	<i>98%</i>	<i>96%</i>	<i>78%</i>	<i>76%</i>

## Robustness in Lateral Mode

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<i>Aerodynamic coefficient</i>	1%	1%	2%	2%	5%	5%
<i>Mass and Inertia coefficient</i>	5%	10%	5%	10%	5%	10%
<i>% Success</i>	100%	100%	100%	100%	95%	86%

# Neuro-Adaptive Design

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- Problem:
  - The DI design sensitive to modeling parameter inaccuracies
- Solution:
  - Enhance Robustness by augmenting the DI with “***Neuro-Adaptive Design***”
- The adaptive design should preferably be compatible with “any” nominal controller

# Output Robustness: Robustness of Inner Loop

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Desired output dynamics:

$$\dot{Y}_d = f_{Y_d}(X_d) + G_{Y_d}(X_d)U_d$$

Actual output dynamics:

$$\dot{Y} = f_{Y_d}(X) + G_{Y_d}(X)U + d(X)$$

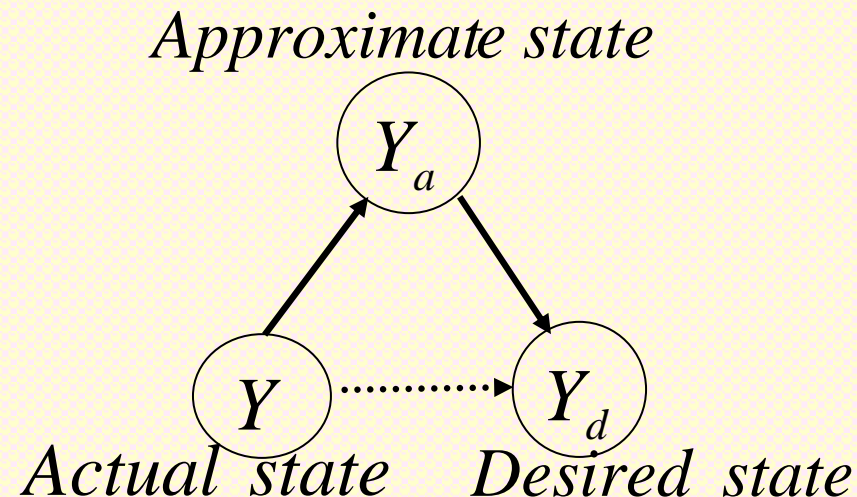
**Objective:**  $Y \rightarrow Y_d$  as soon as possible

# N-A for Robustness of Output Dynamics

- Dynamics of auxiliary output:

$$\dot{Y}_a = f_{Y_d}(X) + G_{Y_d}(X)U + \underbrace{\hat{d}(X)}_{\text{NN Approx.}} + K_a(Y - Y_a)$$

- **Strategy:**



## Steps for assuring $Y_a \rightarrow Y_d$

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- Enforce the error dynamics

$$\dot{E}_d + KE_d = 0 \quad E_d \triangleq (Y_a - Y_d)$$

- After carrying out the necessary algebra

$$f(X, U) = h(X, X_a, X_d, U_d)$$

- In case of control affine system

$$f(X) + [g(X)]U = h(X, X_d, X_a, U_d)$$

- The control is given by

$$U = [g(X)]^{-1} \{h(X, X_d, X_a, U_d) - f(X)\}$$

## Steps for assuring $Y \rightarrow Y_a$

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- The error in the output is defined as

$$E_a \triangleq (Y - Y_a) \quad e_{a_i} \triangleq (y_i - y_{ai})$$

- Ideal neural network is given by:

$$d_i(X) = W_i^T \varphi_i(X) + \varepsilon_i$$

where  $W_i$  is the weight matrix and  $\varphi_i(X)$  is the radial basis function

# Function Learning:

Define error  $e_{a_i} \triangleq (y_i - y_{a_i})$

Output dynamics

$$\begin{aligned}\dot{y}_i &= f_{Y_i}(X) + g_{Y_i}(X)U + d_i(X) \\ \dot{y}_{a_i} &= f_{Y_i}(X) + g_{Y_i}(X)U + \hat{d}_i(X) + k_{a_i} e_{a_i}\end{aligned}$$

**From universal function approximation property**

$$\begin{aligned}d_i(X) &= W_i^T \varphi_i(X) + \varepsilon_i \\ \hat{d}_i(X) &= \hat{W}_i^T \varphi_i(X)\end{aligned}$$

Error dynamics

$$\begin{aligned}\dot{e}_{a_i} &= d_i(X) - \hat{d}_i(X) - k_{a_i} e_{a_i} \\ &= \tilde{W}_i^T \Phi_i(X) + \varepsilon_i - k_{a_i} e_{a_i}\end{aligned}$$



# Lyapunov Stability Analysis

Lyapunov Function Candidate:

$$L_i = \frac{1}{2} (e_{a_i} p_i e_{a_i}) + \frac{1}{2} (\tilde{W}_i^T \gamma_i \tilde{W}_i)$$

Derivative of Lyapunov Function:

$$\begin{aligned} \dot{L}_i &= e_{a_i} p_i \dot{e}_{a_i} + \tilde{W}_i^T \gamma_i \dot{\tilde{W}}_i \\ &= e_{a_i} p_i \left[ \tilde{W}_i^T \Phi_i(X) + \varepsilon_i - k_{a_i} e_{a_i} \right] - \tilde{W}_i^T \gamma_i \dot{\tilde{W}}_i \\ &= \tilde{W}_i^T \left[ e_{a_i} p_i \Phi_i(X) - \gamma_i^{-1} \dot{\tilde{W}}_i \right] + e_{a_i} p_i \varepsilon_i - k_{a_i} e_{a_i}^2 p_i \end{aligned}$$

Weight Update Rule:

$$\dot{\tilde{W}}_i = \gamma_i e_{a_i} p_i \Phi_i(X, X_d)$$

# Lyapunov Stability Analysis

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This condition leads to  $\dot{L}_i = e_{a_i} p_i \varepsilon_i - k_{a_i} e_{a_i}^2 p_i$

$\dot{L}_i < 0$  whenever  $|e_{a_i}| > |\varepsilon_i| / k_{a_i}$

Using the Lyapunov stability theory, we conclude that the trajectory of  $e_{a_i}$  and  $\tilde{W}_i$  are pulled towards the origin.

Hence, the output dynamics is “Practically Stable”!

## Problem Specific equations

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- Output vector in longitudinal mode:

$$Y \triangleq \begin{bmatrix} P & a_z & a_y & V_T \end{bmatrix}^T$$

- Output vector in lateral mode:

$$Y \triangleq \begin{bmatrix} P & Q & a_y & V_T \end{bmatrix}^T$$

## Basis Function

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$$\varphi_i(X) = \left[ e^{-\frac{1}{2} \left( \frac{\|Y - Y_a\|^2}{\sigma_1^2} \right)} \quad e^{-\frac{1}{2} \left( \frac{\|Y - Y_a\|^2}{\sigma_2^2} \right)} \quad e^{-\frac{1}{2} \left( \frac{\|Y - Y_a\|^2}{\sigma_3^2} \right)} \right]^T$$

where mean values chosen were:

$$\sigma_1 = 0.1, \sigma_2 = 1, \sigma_3 = 10$$

## Design parameters selected (Longitudinal mode)

---

- The learning rates and the scalar values:

$$\gamma_P = \gamma_{n_z} = \gamma_{n_y} = \gamma_{V_T} = 60$$

$$p_p = 0.0001, \quad p_{n_z} = 0.05, \quad p_{n_y} = 0.001, \quad p_{V_T} = 0.005$$

- The constants selected are:

$$K = \text{diag} [5 \quad 5 \quad 5 \quad 4]$$

$$K_a = \text{diag} [0.05 \quad 0.05 \quad 0.05 \quad 1]$$

## Design parameters selected (Lateral mode)

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- The learning rates and the scalar values:

$$\gamma_P = \gamma_{n_z} = \gamma_{n_y} = 60; \quad \gamma_{V_T} = 40$$

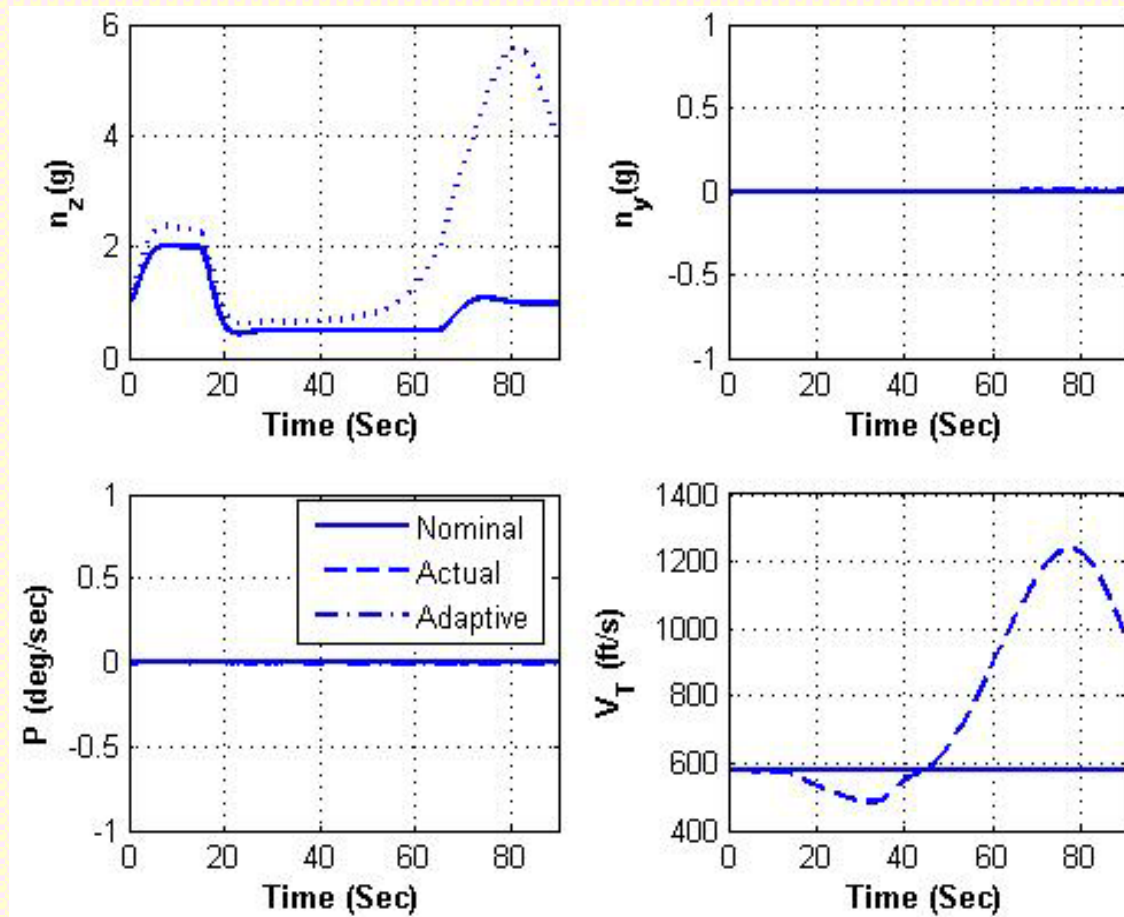
$$p_p = p_q = p_{n_y} = 0.0001, \quad p_{V_T} = 0.05$$

- The constants selected are:

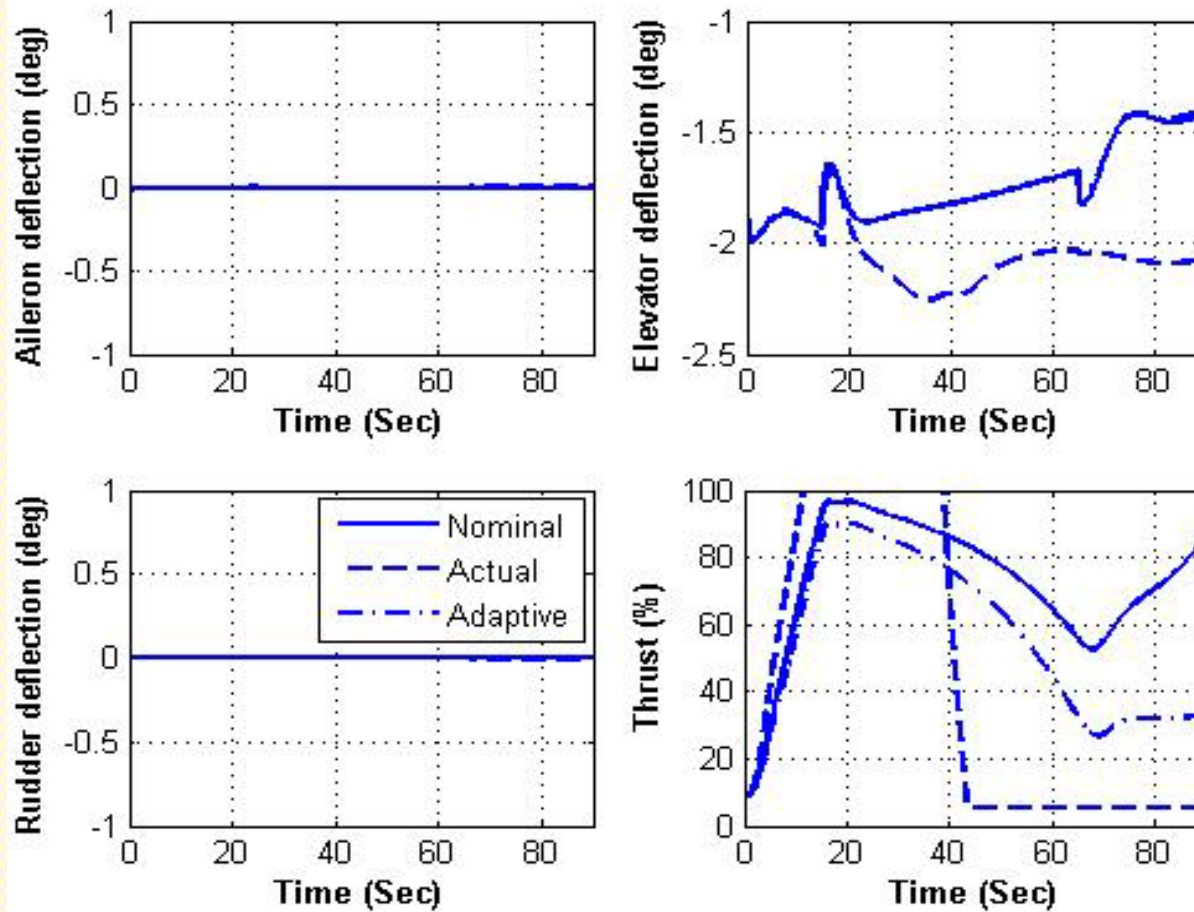
$$K = \text{diag} [5 \quad 4 \quad 4 \quad 4]$$

$$K_a = \text{diag} [0.05 \quad 0.8 \quad 0.05 \quad 0.05]$$

# Outputs in Longitudinal Mode

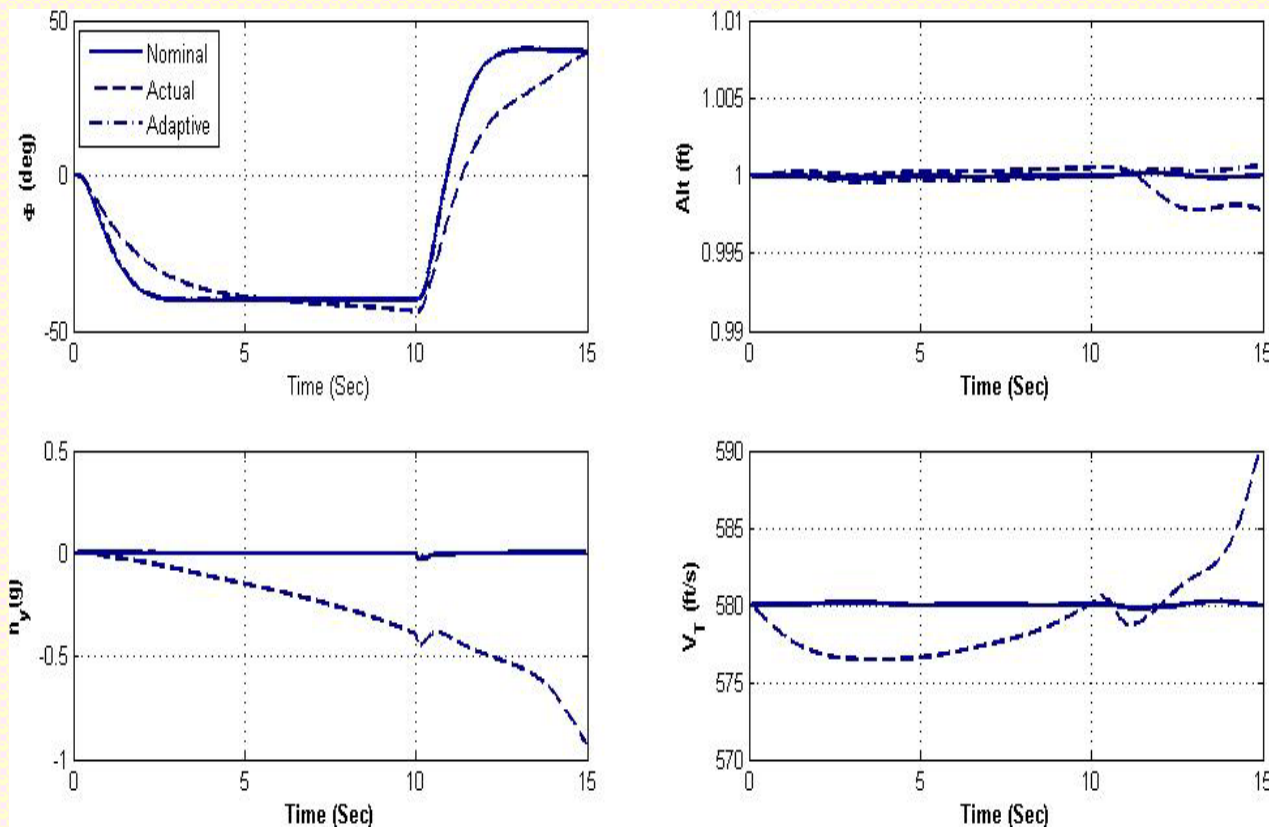


# Controls in Longitudinal Mode

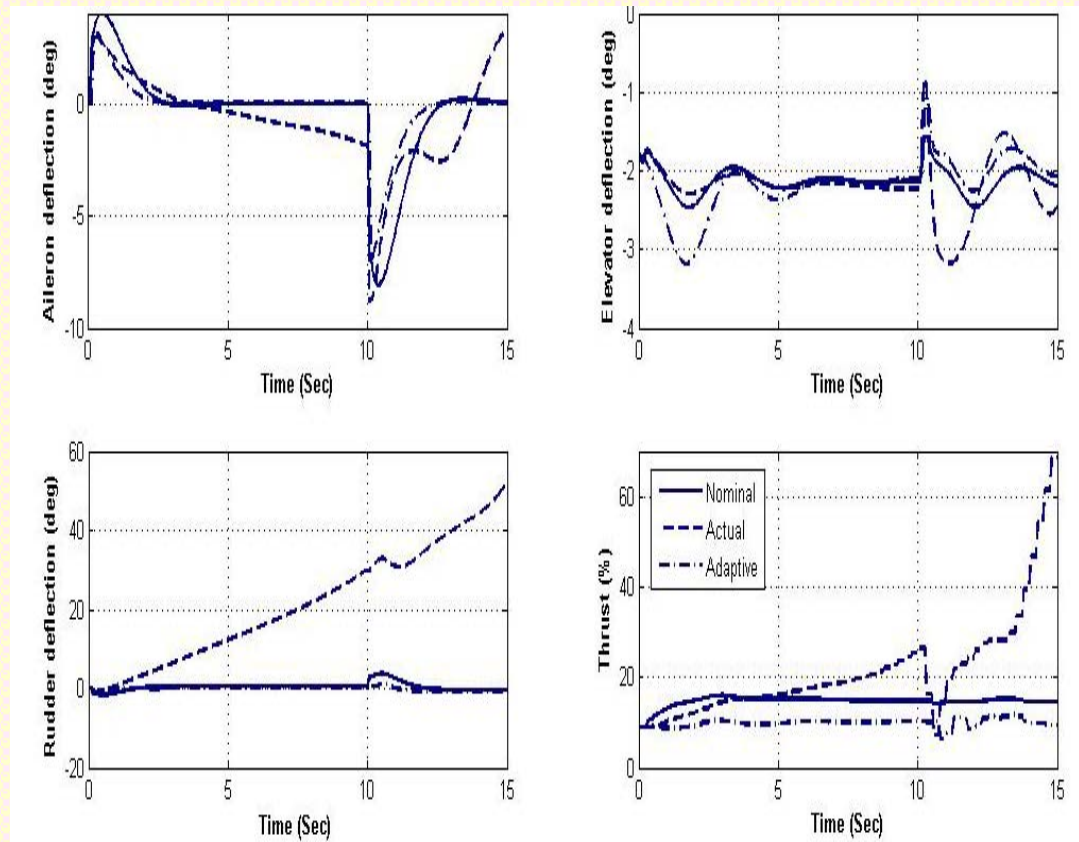




# Outputs in Lateral Mode



# Controls in Lateral Mode



# Robustness Enhancement: Longitudinal Mode

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Aerodynamic Coefficient Perturbation	1 %	1 %	2 %	2 %	5 %	5 %	10 %	10 %
Inertia Parameter Perturbation	5 %	10 %	5 %	10 %	5 %	10 %	5 %	10 %
Nominal Success	100 %	100 %	96 %	92 %	76 %	70 %	48 %	40 %
Adaptive Success	100 %	100 %	100 %	100 %	100 %	100 %	100 %	100 %

# Robustness Enhancement: Lateral Mode

---

<b>Aerodynamic Coefficient Perturbation</b>	1 %	1 %	2 %	2 %	5 %	5 %	10 %	10 %
<b>Inertia Parameter Perturbation</b>	5 %	10 %	5 %	10 %	5 %	10 %	5 %	10 %
<b>Nominal Success</b>	100 %	100 %	100 %	100 %	98 %	94 %	78 %	76 %
<b>Adaptive Success</b>	100 %	100 %	100 %	100 %	100 %	100 %	100 %	100 %

# Summary

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- Nominal control design has been carried out using “dynamic inversion” (the new method shows remarkable improvement in performance!)
- The nominal design has been augmented with “neuro-adaptive design” for improvement in robustness.
- Simulation Results
  - Tracking performance is very good
  - Enhancement of robustness is substantial

# References

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- **Radhakant Padhi**, Narayan P. Rao, Siddharth Goyal and Abha Tripathi, “*A Model-Following Neuro-Adaptive Approach for Robust Control of High Performance Aircrafts*”, Automatic Control in Aerospace, Vol. 3, No. 1, May 2010.
- **Radhakant Padhi** and S. N. Balakrishnan, “*Implementation of Pilot Commands in Aircraft Control: A New Dynamic Inversion Approach*”, AIAA Conference on Guidance Navigation and Control, 2003, Austin, TX, USA.
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**Thanks for the Attention...!**

