

Global Tracking Control Structures for Nonlinear Singularly Perturbed Aircraft Systems

Anshu Siddarth and John Valasek

Department of Aerospace Engineering
Texas A&M University



CEAS Conference on Guidance, Navigation and Control,
13-15 April 2011,
Munich, Germany.

Student Research Team 2010 - 2011



2



Outline

- Problem Statement
 - Simultaneous Slow and Fast State Tracking
 - Literature Review

- Geometric Singular Perturbation Theory

- Mathematical Development & Stability Analysis

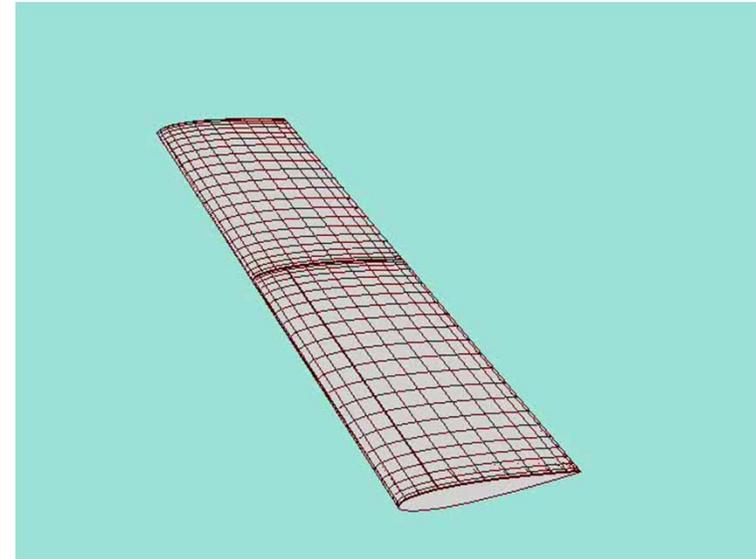
- Numerical Simulations
 - Nonlinear Six Degree-of-Freedom Simulation of an F/A-18A Hornet

- Conclusions & Future Work



Motivations

10+ Independent Morphing DOF



$$\dot{\sigma} = f(\sigma, \omega)$$

$$\dot{\omega} = g(\sigma, \omega) + h(\sigma, \omega, p)u$$

Affine in Control



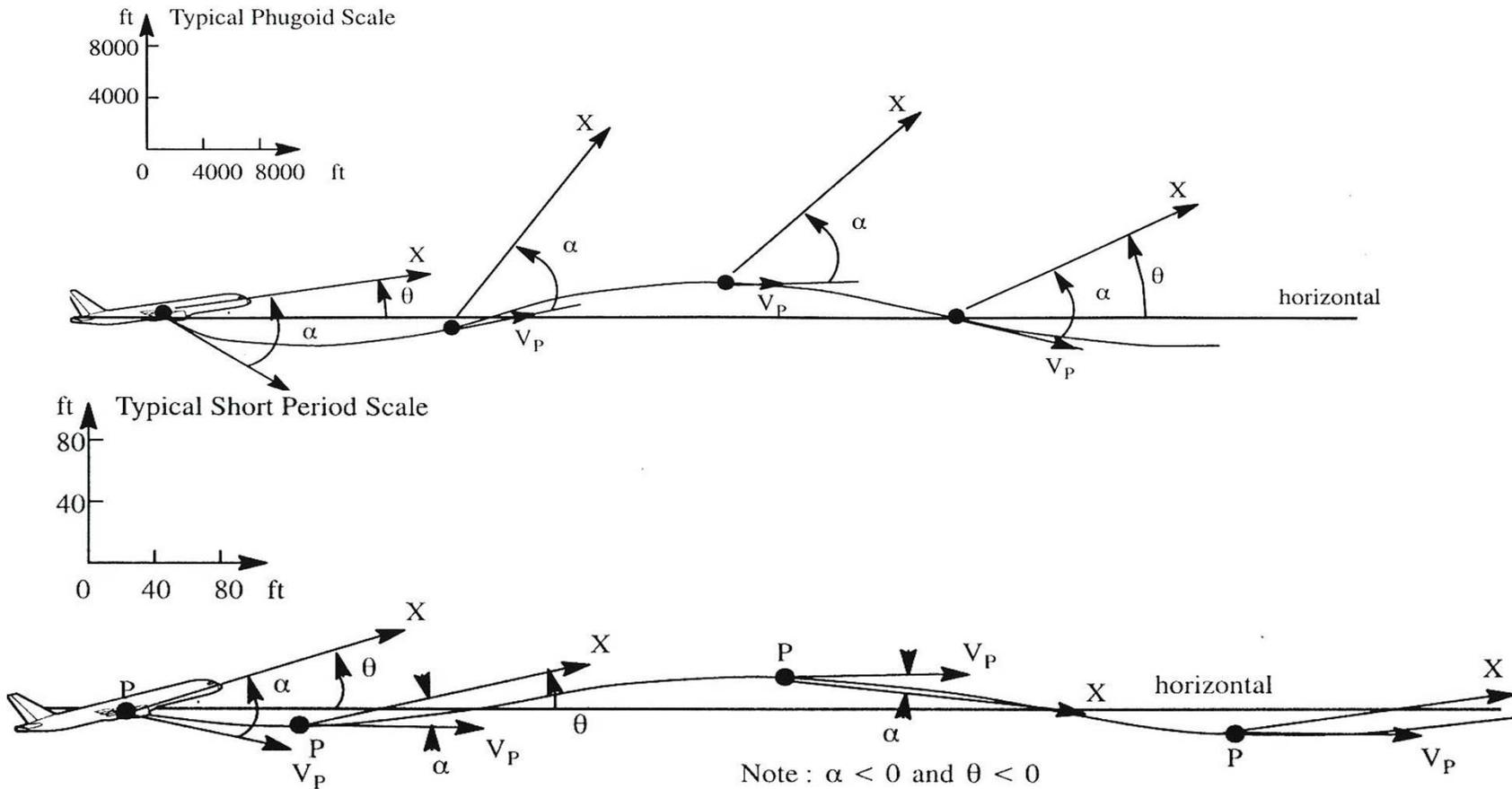
$$\dot{\sigma} = f(\sigma, \omega)$$

$$\dot{\omega} = g(\sigma, \omega, u) + h(\sigma, \omega, p, u)$$

Non-Affine in Control

Two Time Scale

Two-Time Scale Systems



Research Objective

- Nonlinear tracking control structures for:
 - **Two-Time Scale Systems/ Singularly Perturbed Systems**

Examples: mechanical oscillators, airplanes, flexible robot link manipulators, ...

- **Mathematical form:**

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{z}) + \mathbf{g}(\mathbf{x}, \mathbf{z})\mathbf{u}$$

$$\epsilon \dot{\mathbf{z}} = \mathbf{l}(\mathbf{x}, \mathbf{z}) + \mathbf{k}(\mathbf{x}, \mathbf{z})\mathbf{u}$$

$$\mathbf{y} = \begin{bmatrix} \mathbf{x} \\ \mathbf{z} \end{bmatrix}$$

\mathbf{x} is the vector of slow variables,

\mathbf{z} is the vector of fast variables,

ϵ is a small positive parameter that captures the time scale property (unknown)

\mathbf{y} is the vector of outputs.



6



Literature Survey

- Nonlinear Two-Time Scale System Analysis **Mease et. al**, (2003)
- Tracking of **slow variables** using the composite of:
 - Tracking controller for the slow subsystem,
 - Stabilizing controller for the fast subsystem to restrict fast states onto a manifold
- Global tracking results guaranteed only if the manifold is **unique**
 - Nonlinear in slow states and linear in fast states, **Li** (2009)
 - Assume unique manifold, **Grujic** (1988), **Choi** (2005)
- For general nonlinear systems local stability proven:
 - Assume the fast states as control variables for slow system, **Menon** (1987)
 - Approximate manifold, **Siddarth and Valasek** (*JGCD* May-Jun 2011)
- Simultaneous slow and fast tracking posed as optimal control problem, **Arstein** (1997)



Challenges:

Nonlinear Singularly-Perturbed Model

■ Key Issues:

- Numerically stiff equations
- Fast states are restricted to lie on a manifold
- Global results valid only if fast variables lie on a **unique** manifold

■ Approach:

- Model-reduction via Geometric Singular Perturbation Theory
- Use coordinate transformation and enforce the manifold to be exactly the fast state reference
- Composite control design

■ Benefits:

- No assumptions on the class of nonlinear systems considered
 - Does not require computation of the manifold
 - Global asymptotic tracking
 - No knowledge of the singular perturbation parameter required
-



Outline

- ✓ Problem Statement
 - Simultaneous Slow and Fast State Tracking
 - Literature Review

 - **Background**
 - **Geometric Singular Perturbation Theory**

 - Mathematical development

 - Numerical Simulations
 - Nonlinear six degree-of-freedom simulation of an F/A-18A Hornet

 - Conclusions & Future Work
-



9



Geometric Singular Perturbation Theory

(Fenichel, 1979)

Derivatives wrt slow time scale t	$\dot{x}_1 = -x_1 - z$ $\dot{x}_2 = -x_2 - z$ $\epsilon \dot{z} = -z$	$\tau = \frac{t - t_0}{\epsilon}$ 	$x_1' = -\epsilon x_1 - \epsilon z$ $x_2' = -\epsilon x_2 - \epsilon z$ $z' = -z$	Derivatives wrt fast time-scale τ
---	---	--	---	--

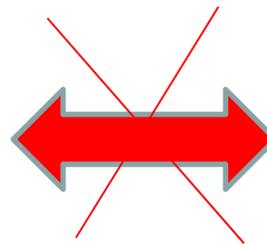
- Develop reduced-order models. Substitute $\epsilon = 0$

$$z = 0$$

$$\dot{x}_1 = -x_1$$

$$\dot{x}_2 = -x_2$$

Slow Subsystem



$$x_1' = 0$$

$$x_2' = 0$$

$$z' = -z$$

Fast Subsystem

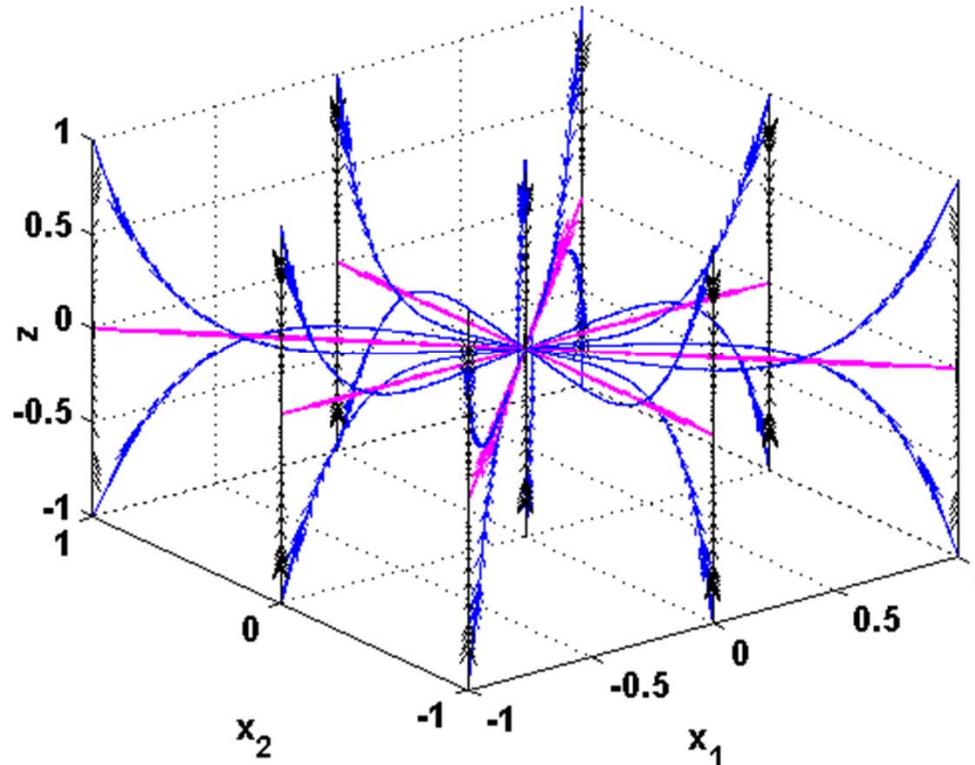


Geometric Singular Perturbation Theory

- Reduced-order models approximate the behaviour of the complete system

Complete System

$$\begin{aligned}\dot{x}_1 &= -x_1 - z \\ \dot{x}_2 &= -x_2 - z \\ \epsilon \dot{z} &= -z\end{aligned}$$



Slow Subsystem

$$\begin{aligned}z &= 0 \\ \dot{x}_1 &= -x_1 \\ \dot{x}_2 &= -x_2\end{aligned}$$

Complete
Slow Subsystem
Fast Subsystem

Fast Subsystem

$$\begin{aligned}x_1' &= 0 \\ x_2' &= 0 \\ z' &= -z\end{aligned}$$

Geometric Singular Perturbation Theory

Insights From The Geometric Approach

- Complete system (in blue) approximated by dynamics of slow subsystem (in pink) and fast subsystem (in black).
 - if complete system locally flattened it falls onto dynamics of the slow sub-system.
- Manifold is described by solution of the algebraic equation when substituting $\epsilon = 0$ in complete system.
- Dynamics on manifold governed by differential equations of slow subsystem which is exactly the dynamics of the slow variables.
- Fast dynamics need to lie on the manifold.



12



Geometric Singular Perturbation Theory

Insights From The Geometric Approach

Key Idea: Tracking of both slow and fast states is achieved if and only if the manifold is exactly the reference trajectory of the fast states.



13



Outline

- ✓ Problem Statement
 - Simultaneous Slow and Fast State Tracking

- ✓ Background
 - Geometric Singular Perturbation Theory

- **Mathematical development**

- Numerical Simulations
 - Nonlinear six degree-of-freedom simulation of an F/A-18A Hornet

- Conclusions & Future Work



14



Mathematical Formulation of the Control Law

- Complete Model:
$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{z}) + \mathbf{g}(\mathbf{x}, \mathbf{z})\mathbf{u}$$
$$\epsilon\dot{\mathbf{z}} = \mathbf{l}(\mathbf{x}, \mathbf{z}) + \mathbf{k}(\mathbf{x}, \mathbf{z})\mathbf{u}$$
$$\mathbf{y} = \begin{bmatrix} \mathbf{x} \\ \mathbf{z} \end{bmatrix}$$

- Tracking errors:
$$\mathbf{e}(t) = \mathbf{x}(t) - \mathbf{x}_r(t)$$
$$\boldsymbol{\xi}(t) = \mathbf{z}(t) - \mathbf{z}_r(t)$$

- Transform the equilibrium to origin:

$$\dot{\mathbf{e}} = \mathbf{f}(\mathbf{x}, \mathbf{z}) + \mathbf{g}(\mathbf{x}, \mathbf{z})\mathbf{u} - \dot{\mathbf{x}}_r \triangleq \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) + \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r)\mathbf{u}$$
$$\epsilon\dot{\boldsymbol{\xi}} = \mathbf{l}(\mathbf{x}, \mathbf{z}) + \mathbf{k}(\mathbf{x}, \mathbf{z})\mathbf{u} - \epsilon\dot{\mathbf{z}}_r \triangleq \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \epsilon\dot{\mathbf{z}}_r) + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r)\mathbf{u}$$

Step 1. Obtain Reduced-Order Models

$$\dot{\mathbf{e}} = \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) + \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}$$

$$\epsilon \dot{\boldsymbol{\xi}} = \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \epsilon \dot{\mathbf{z}}_r) + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}$$



■ Reduced Slow Subsystem

$$\dot{\mathbf{e}} = \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) + \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}_0$$

$$\mathbf{0} = \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{0}) + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}_0$$

■ Reduced Fast Subsystem

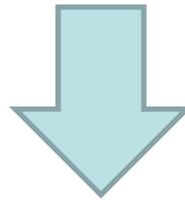
$$\dot{\mathbf{e}} = \mathbf{0}$$

$$\begin{aligned} \boldsymbol{\xi}' &= \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{z}_r') \\ &\quad + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) (\mathbf{u}_0 + \mathbf{u}_f) \end{aligned}$$

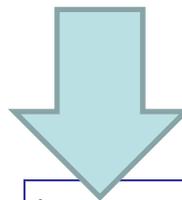
Step 2. Design Controller for Slow Subsystem

- Force origin as the equilibrium of the closed-loop slow subsystem

$$\begin{bmatrix} \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \\ \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \end{bmatrix} \mathbf{u}_0 = - \begin{bmatrix} \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) \\ \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{0}) \end{bmatrix} + \begin{bmatrix} A_e \mathbf{e} \\ A_\xi \boldsymbol{\xi} \end{bmatrix}$$



$$\begin{aligned} \dot{\mathbf{e}} &= \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) + \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}_0 \\ \mathbf{0} &= \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{0}) + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}_0 \end{aligned}$$



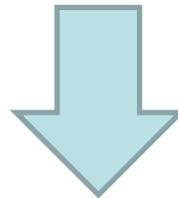
$$\begin{aligned} \dot{\mathbf{e}} &= A_e \mathbf{e} \\ \mathbf{0} &= A_\xi \boldsymbol{\xi} \end{aligned}$$

Resulting Closed-Loop
Slow Subsystem

Step 3. Design Controller for Fast Subsystem

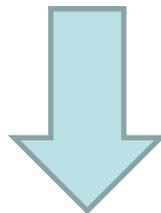
- To make sure the manifold is stable for all time,

$$\begin{bmatrix} \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \\ \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \end{bmatrix} \mathbf{u}_f = \begin{bmatrix} \mathbf{0} \\ \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{0}) - \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{z}_r') \end{bmatrix}$$



$$\mathbf{e}' = \mathbf{0}$$

$$\boldsymbol{\xi}' = \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \mathbf{z}_r') + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r)(\mathbf{u}_0 + \mathbf{u}_f)$$



$$\mathbf{e}' = \mathbf{0}$$

$$\boldsymbol{\xi}' = \mathbf{A}_\xi \boldsymbol{\xi}$$

Resulting Closed-Loop
Fast Subsystem

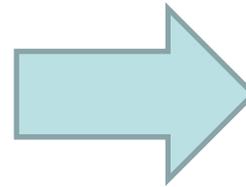
Step 4. Composite Control Design

- Composite Control: $\mathbf{u} = \mathbf{u}_0 + \mathbf{u}_f$

- Or,
$$\begin{bmatrix} \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \\ \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \end{bmatrix} \mathbf{u} = - \begin{bmatrix} \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) \\ \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{z}}_r) \end{bmatrix} + \begin{bmatrix} \mathbf{A}_e \mathbf{e} \\ \mathbf{A}_\xi \boldsymbol{\xi} \end{bmatrix}$$

Resulting Closed-Loop System

$$\begin{aligned} \dot{\mathbf{e}} &= \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) + \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u} \\ \epsilon \dot{\boldsymbol{\xi}} &= \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \epsilon \dot{\mathbf{z}}_r) + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u} \end{aligned}$$



$$\begin{aligned} \dot{\mathbf{e}} &= \mathbf{A}_e \mathbf{e} \\ \epsilon \dot{\boldsymbol{\xi}} &= \mathbf{A}_\xi \boldsymbol{\xi} \end{aligned}$$

Lyapunov Analysis

- Lyapunov Function Candidate:

$$v(\mathbf{e}, \boldsymbol{\zeta}) = (1-d)\mathbf{e}^T \mathbf{e} + d\boldsymbol{\zeta}^T \boldsymbol{\zeta}, \quad d > 0$$

- Time derivative about closed-loop dynamics:

$$\dot{v} \leq -\begin{bmatrix} \mathbf{e} \\ \boldsymbol{\zeta} \end{bmatrix}^T \begin{bmatrix} (1-d)(\alpha_1 - 2\alpha) & 0 \\ 0 & \frac{d}{\epsilon}\alpha_2 - 2\alpha d \end{bmatrix} \begin{bmatrix} \mathbf{e} \\ \boldsymbol{\zeta} \end{bmatrix} - 2\alpha v$$

$$\begin{aligned} \alpha_1 &= 2|\lambda_{\min}(A_e)|, \\ \alpha_2 &= 2|\lambda_{\min}(A_\xi)|, \\ \alpha &> 0 \end{aligned}$$

- If $\epsilon < \frac{\alpha_2}{2\alpha}$, $\dot{v} \leq -2\alpha v$

- Or, $\dot{v} \leq -2\alpha \gamma_{11} \left\| \begin{bmatrix} \mathbf{e} \\ \boldsymbol{\zeta} \end{bmatrix} \right\|^2$

- Thus, global exponential stability can be concluded.

- Or $\mathbf{x}(t) \rightarrow \mathbf{x}_r(t), \mathbf{z}(t) \rightarrow \mathbf{z}_r(t), t \rightarrow \infty$

Outline

- ✓ Problem Statement
 - Simultaneous Slow and Fast State Tracking
- ✓ Background
 - Geometric Singular Perturbation Theory
- ✓ Mathematical development
- Numerical Simulations
 - Nonlinear six degree-of-freedom simulation of an F/A-18A Hornet
- Conclusions & Future Work

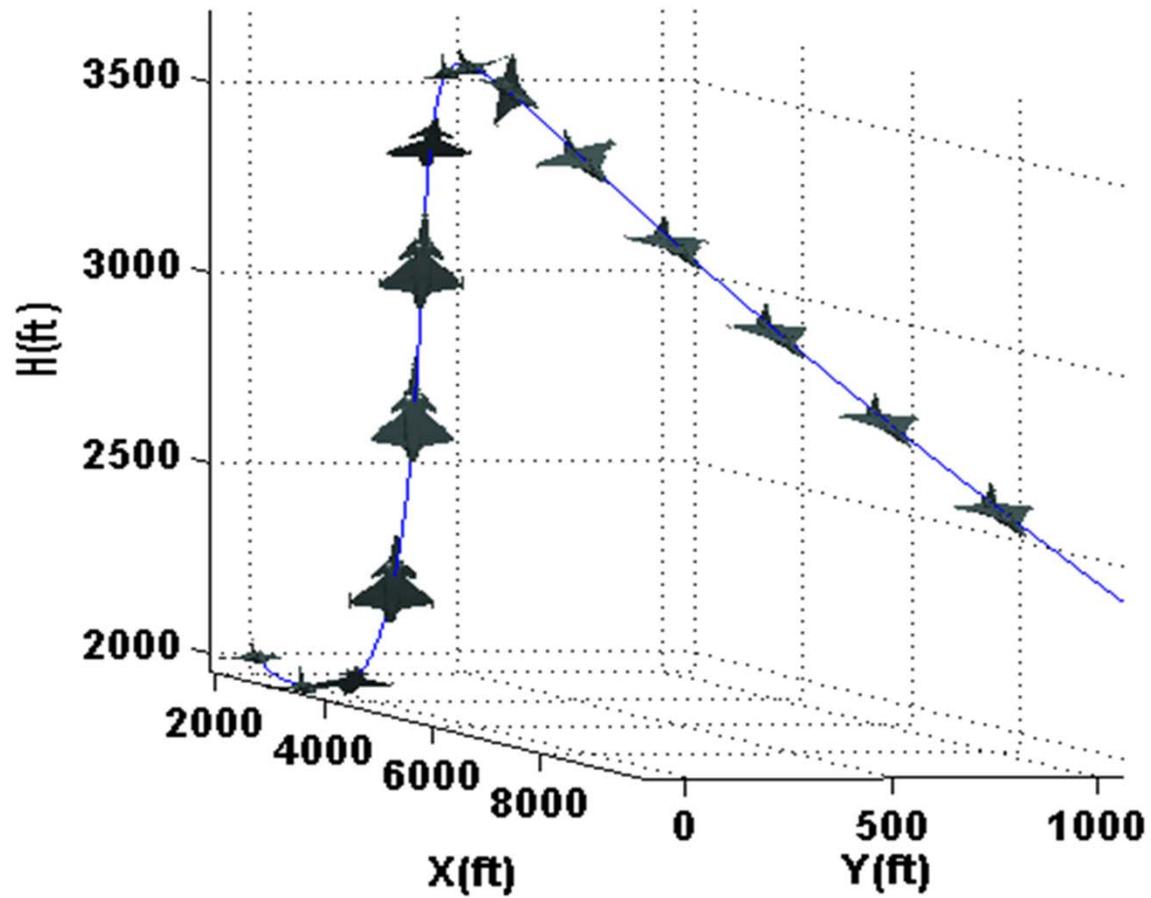


Numerical Example

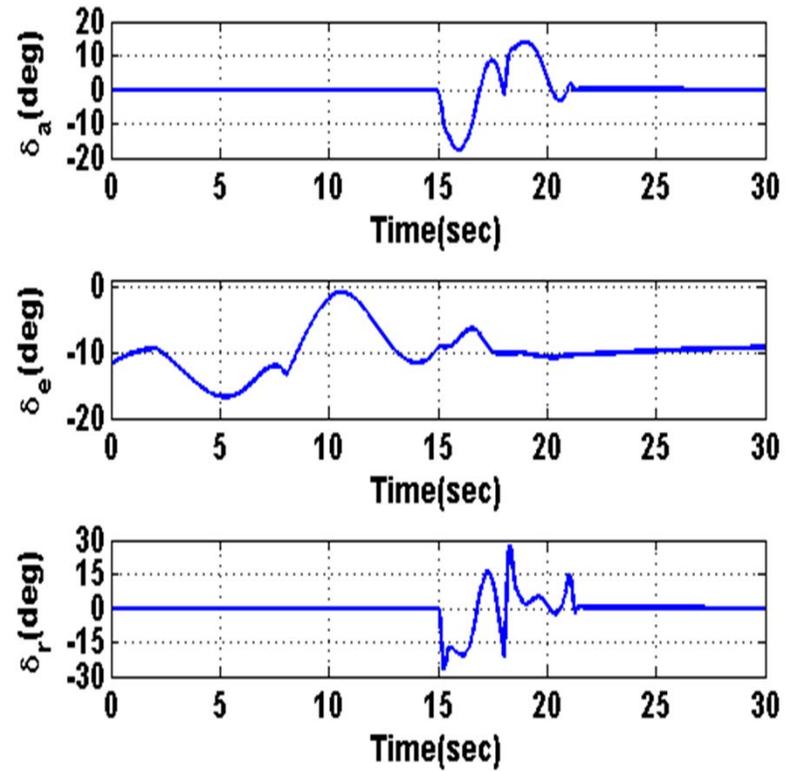
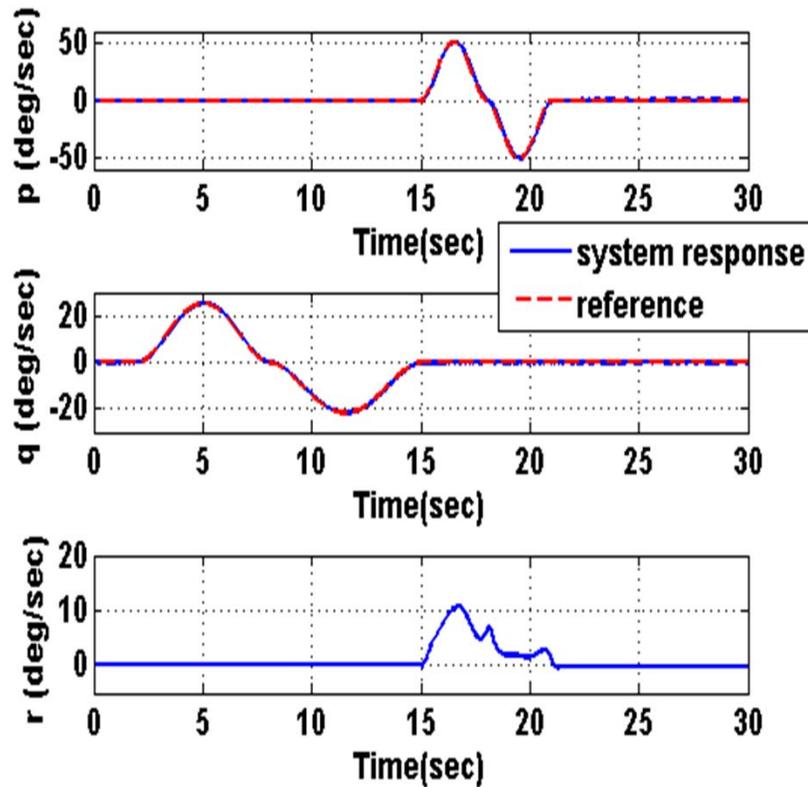
- Nonlinear, six degree-of-freedom F/A-18A Hornet written in the stability axis.
- Slow states: $\mathbf{x} = [M, \alpha, \beta, \phi, \theta, \psi]^T$
- Fast states: $\mathbf{z} = [p, q, r]^T$
- Control Variables: $u = [\eta, \delta_e, \delta_a, \delta_r]^T$
- Objective: Aggressive vertical climb with maximum pitch rate of 25deg/sec followed by roll at a rate of 50deg/sec with sideslip angle constrained to zero throughout.



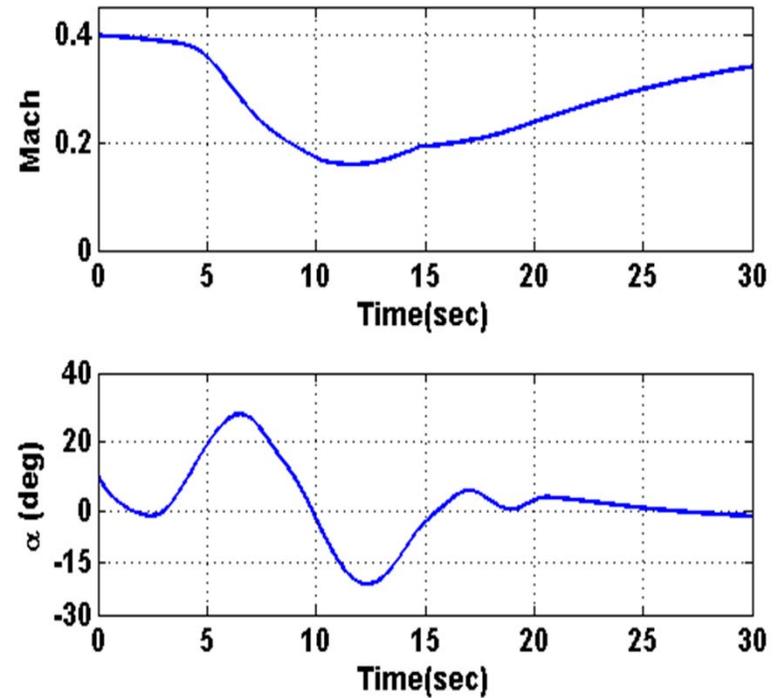
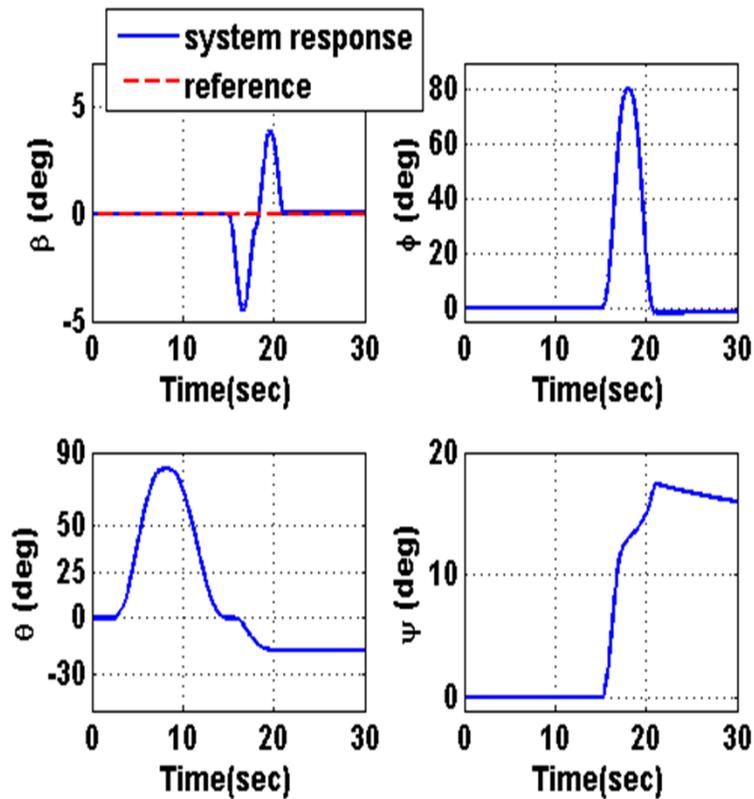
Aircraft Trajectory



Angular Rates and Control Deflections



Response of Slow State States



Outline

- ✓ Problem Statement
 - Simultaneous Slow and Fast State Tracking
- ✓ Background
 - Geometric Singular Perturbation Theory
- ✓ Mathematical development
- ✓ Numerical Simulations
 - Nonlinear six degree-of-freedom simulation of an F/A-18A Hornet
- Conclusions & Future Work



Summary and Conclusions

- Developed tracking controllers to simultaneously track slow and fast states for **nonlinear singularly perturbed system** using geometric singular perturbation theory as a model-reduction technique
- Controller is implemented without making any assumptions about the nonlinear model and does not require knowledge of the perturbation parameter.
- Consistent tracking guaranteed independent of the reference trajectory.
- Global asymptotic tracking demonstrated even though the reference trajectory requires airplane to switch between linear and nonlinear regimes.
- All closed-loop signals were bounded and control deflections computed were smooth.



27



Future Research Extensions

- Global stabilization without tracking fast states
 - How can global results be asserted? (Siddarth & Valasek, ACC 2011)

- Extensions for systems nonlinear in control
 - Current work assumes the control appears linearly

$$\dot{\mathbf{e}} = \mathbf{F}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \dot{\mathbf{x}}_r) + \mathbf{G}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}$$

$$\epsilon \dot{\boldsymbol{\xi}} = \mathbf{L}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r, \epsilon \dot{\mathbf{z}}_r) + \mathbf{K}(\mathbf{e}, \boldsymbol{\xi}, \mathbf{x}_r, \mathbf{z}_r) \mathbf{u}$$

- Non-minimum phase system control architecture
 - Take advantage of the multiple-time scale behaviour of aircraft (Siddarth & Valasek, AIAA GNC 2011)

- Autonomous control of Highly Reconfigurable Structures
 - Adaptive-Reinforcement Learning (Valasek et.al, 2004, 2006)



Acknowledgements

This work was sponsored (in part) by the

- U.S. Air Force Office of Scientific Research under contract FA9550-08-1-0038. Technical Monitor: Dr. Fariba Fahroo.
- NASA Johnson Space Station under contract C08-00884. Technical Monitor: Steve Fitzgerald.

Any opinions, findings, conclusions or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of the U.S. Air Force or NASA.



29



QUESTIONS?

