

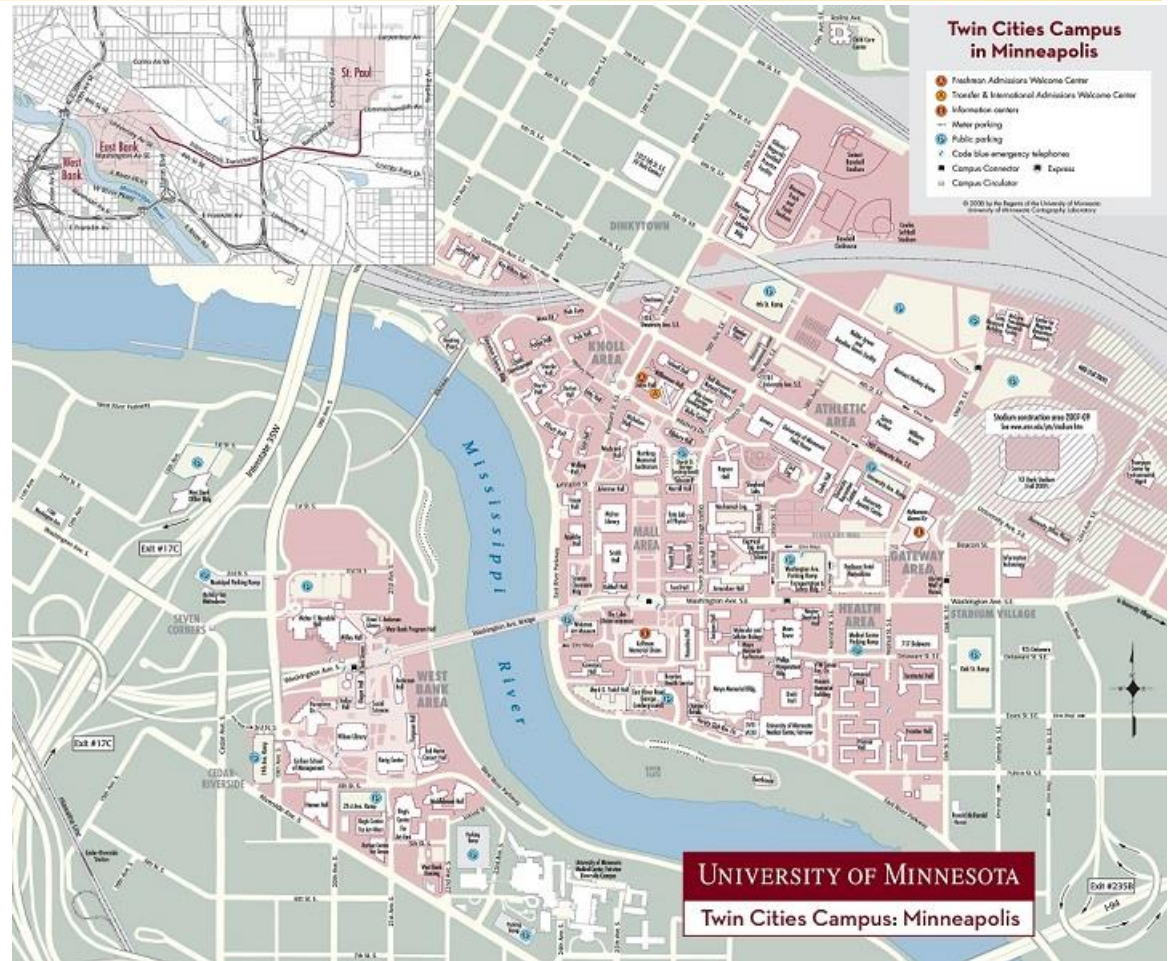
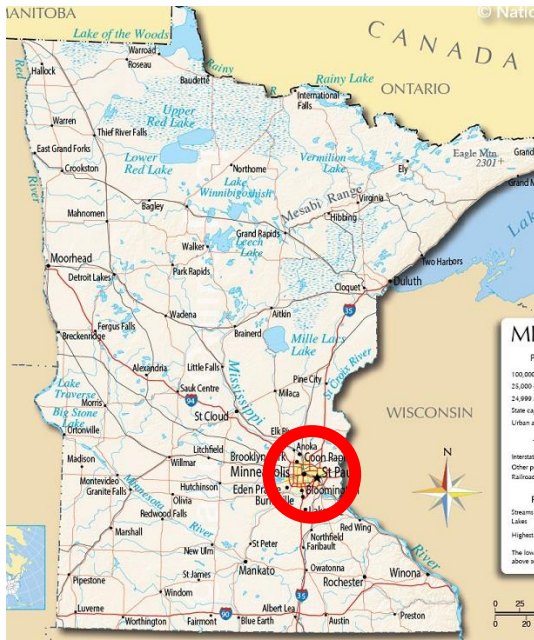
# Design and Analysis of Safety Critical Systems

**Peter Seiler**  
**University of Minnesota**



**MTA Sztaki**  
**December 5, 2017**

# University of Minnesota



- Founded in 1851
- Campuses in Twin Cities, Duluth, Morris and Crookston.
- Twin Cities campus has 52,557 students (~7,200 in CSE).

# Dept. of Aerospace Engineering & Mechanics



*1933 First Class of Seniors  
Taking Flying Lessons*



- First aeronautical engineering courses offered in 1926. Department founded in fall 1929 with 3 faculty members.
- Aeronautical Engineering merged with the Department of Mechanics and Materials in 1958 to form current department
- 17 regular faculty (6 systems, 6 fluids, 5 solids)
- 328 undergraduates, 17 MS, and 73 PhD students

# Aerospace Systems

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**Demoz Gebre-Egziabher:**  
*Sensor fusion; design of multi-sensor systems for navigation*



**William Garrard:** *Dynamics and control of aerospace vehicles; parachute dynamics.*



**Peter Seiler:** *Robust control with applications to aerospace systems and wind energy*



**Yohannes Ketema:** *Dynamics; dynamics of active materials; stability of formations; orbital mechanics*

# Aerospace Systems

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**Richard Linares:** *Orbital debris tracking, uncertainly quantification*



**Maziar Hemati:** *Control and optimization, primarily of fluid mechanical systems*



**Derya Aksaray:** *Control theory, formal methods, and machine learning with applications to autonomous systems.*



**Ryan Caverly:** *Robust control with applications to aerospace, mechanical and marine systems.*

# Research Summary

*Jordan Hoyt*

*Parul Singh*

*Sanjana Vijayshankar*

*Wind Energy*



*Raghu Venkataraman*

*Harish Venkataraman*

*Small UAVs*



*Abhineet Gupta*

*Aeroelasticity*



**Robust Control Design and Analysis**

*Chris Regan*

*Brian Taylor*

*Curt Olson*

# Fault Tolerance for Small UAVs

With: Raghu Venkataraman



## Funding:

(NSF) CPS: Managing Uncertainty in the Design of Safety-Critical Aviation Systems

(MnDrive) Precision Agriculture: Robotics and Sensor Development for Revolutionary Improvements in the Global Food Supply and Reduced Environmental Impact in the Agriculture Industry.

# Growth in Small UAVs



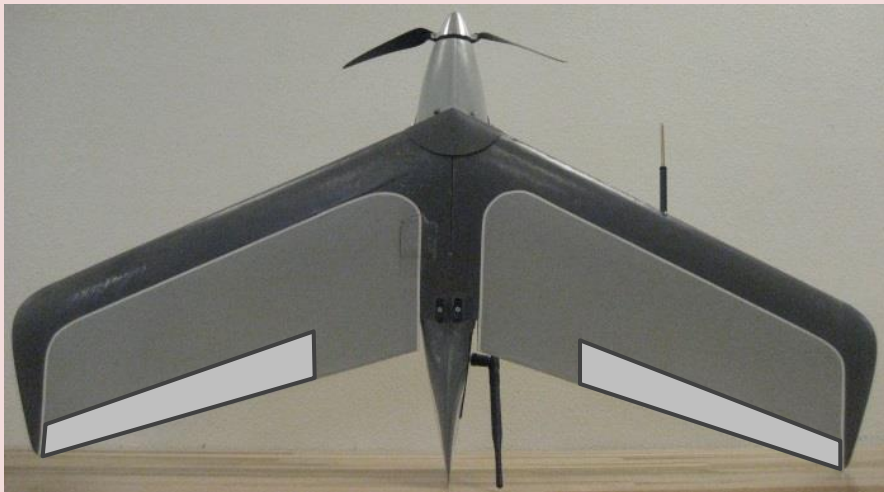
DJI Phantom 4  
(Source: [www.dji.com](http://www.dji.com))



Trimble UX5  
(Source: [uas.trimble.com](http://uas.trimble.com))



senseFly eBee  
(Source: [uncrate.com](http://uncrate.com))

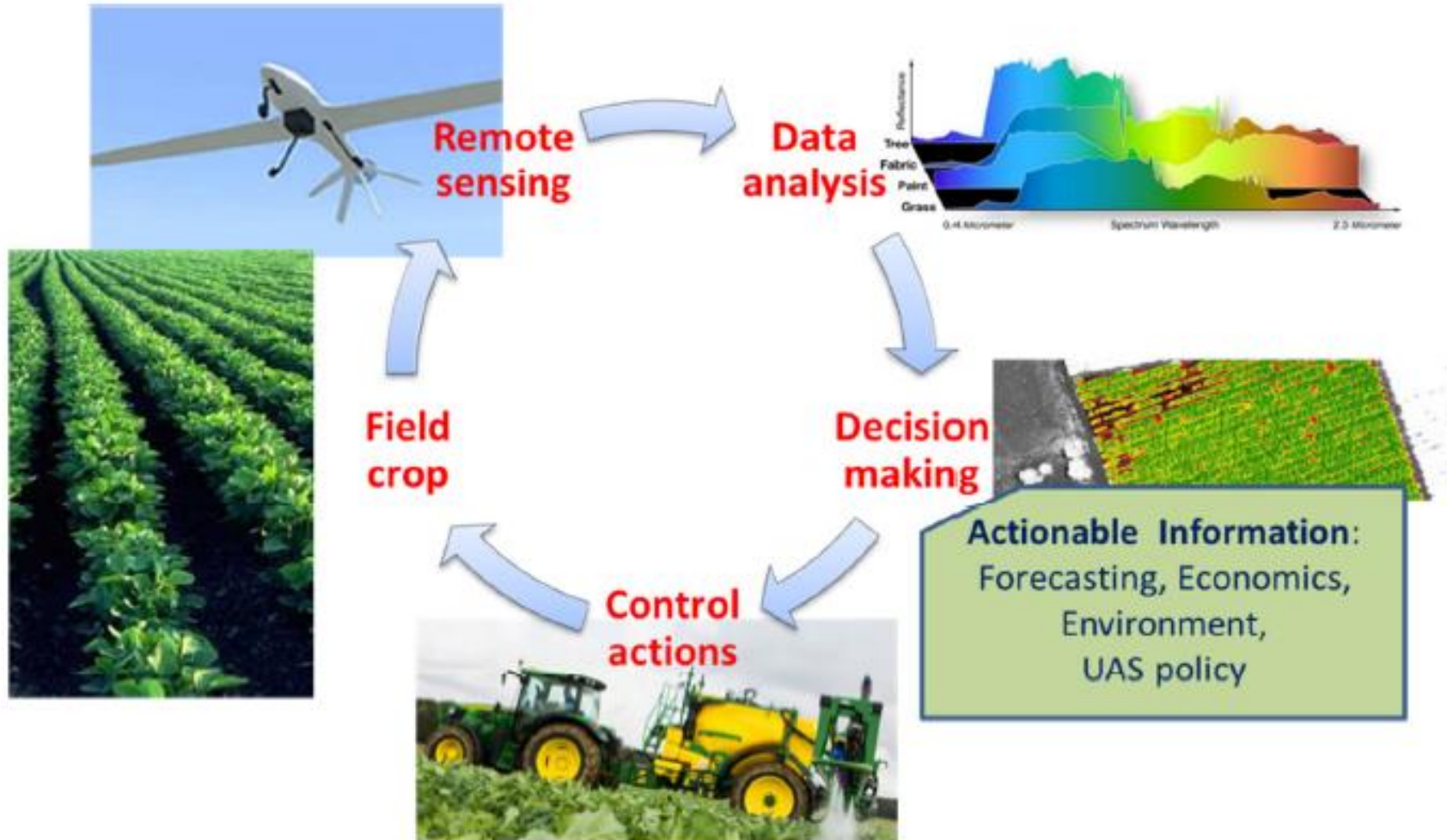


## Sentera Vireo

- Donated to UMN in 2014
- Remote sensing applications, e.g. **precision agriculture**
- Mahon et al. "Research Flight Test Vehicle: Small Two Surface UAV," UMN Technical Report, 2016.



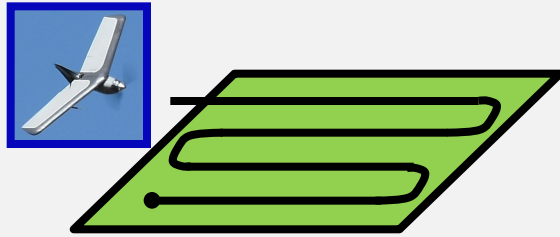
# Precision Agriculture



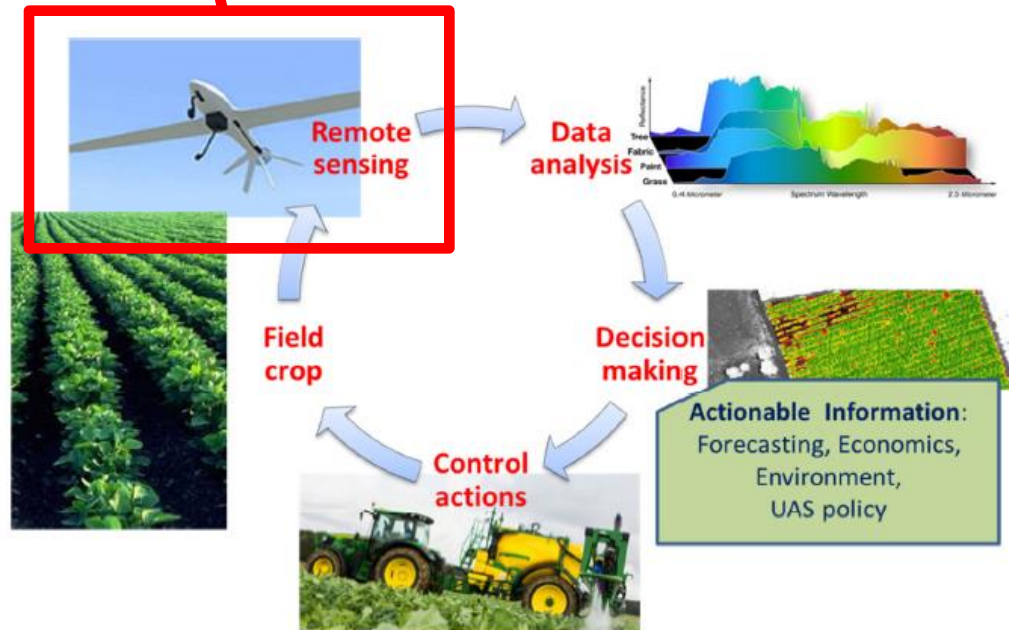
# Precision Agriculture

## Nominal mission

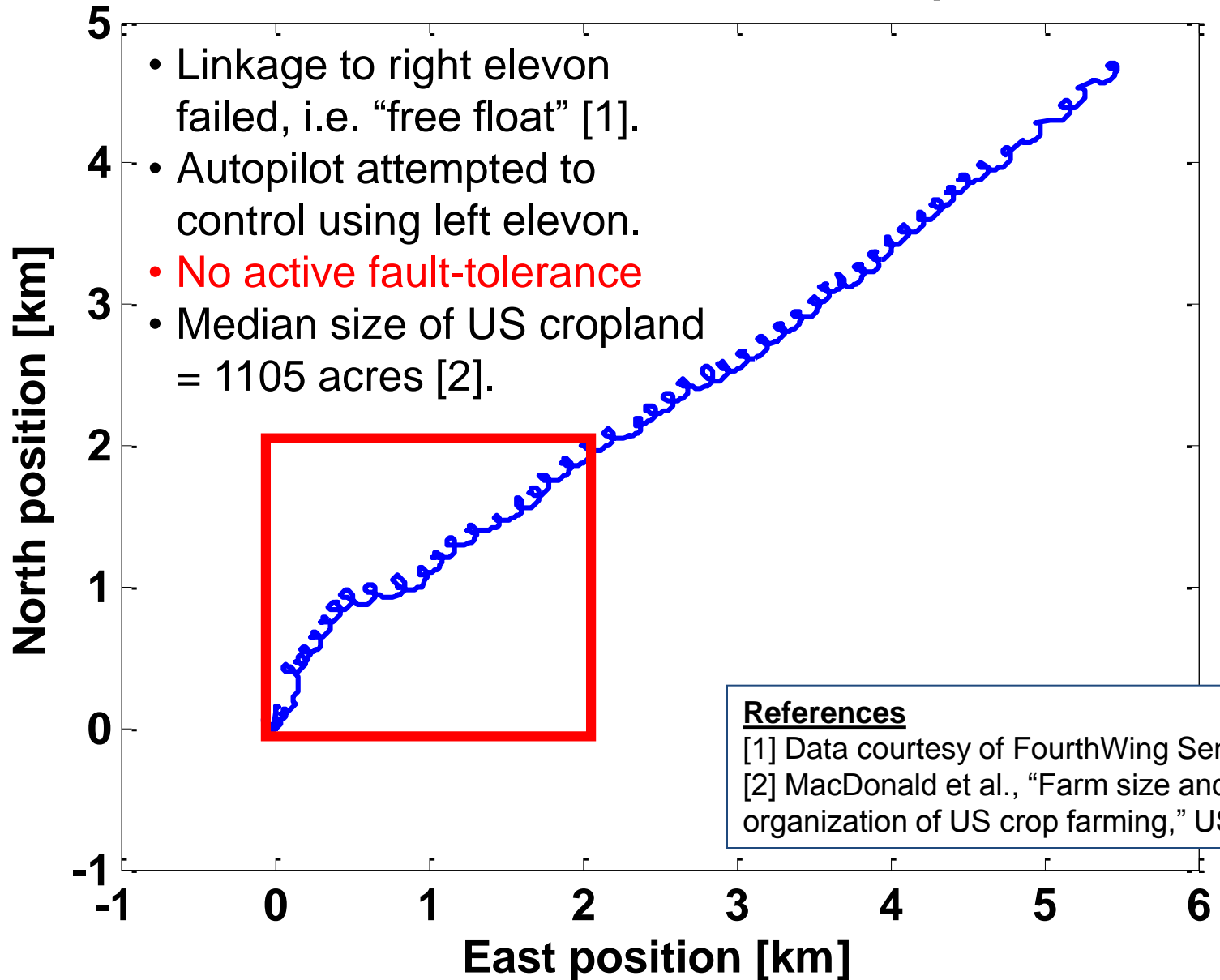
*Lawnmower pattern contained within geofence perimeter*



*Atkins, "Autonomy as an enabler of economically-viable, beyond-line-of-sight, low-altitude UAS applications with acceptable risk," AUVSI, 2014.*



# Flight Data From Aborted Mission



# Fault Tolerance: Commercial Aircraft

## Boeing 787-8 Dreamliner

- 210-250 seats
- Length=56.7m, Wingspan=60.0m
- Range < 15200km, Speed< M0.89
- First Composite Airliner
- Honeywell Flight Control Electronics



## Boeing 777-200

- 301-440 seats
- Length=63.7m, Wingspan=60.9m
- Range < 17370km, Speed< M0.89
- Boeing's 1<sup>st</sup> Fly-by-Wire Aircraft
- Ref: Y.C. Yeh, "Triple-triple redundant 777 primary flight computer," 1996.

# Fault Tolerance: Commercial Aircraft

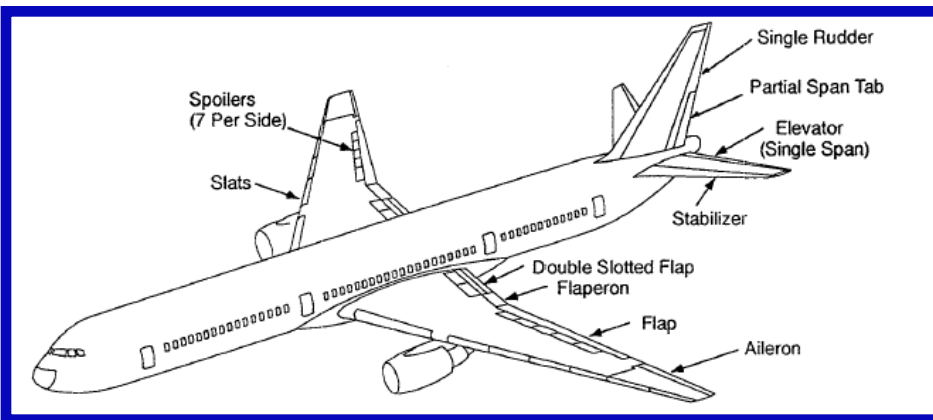
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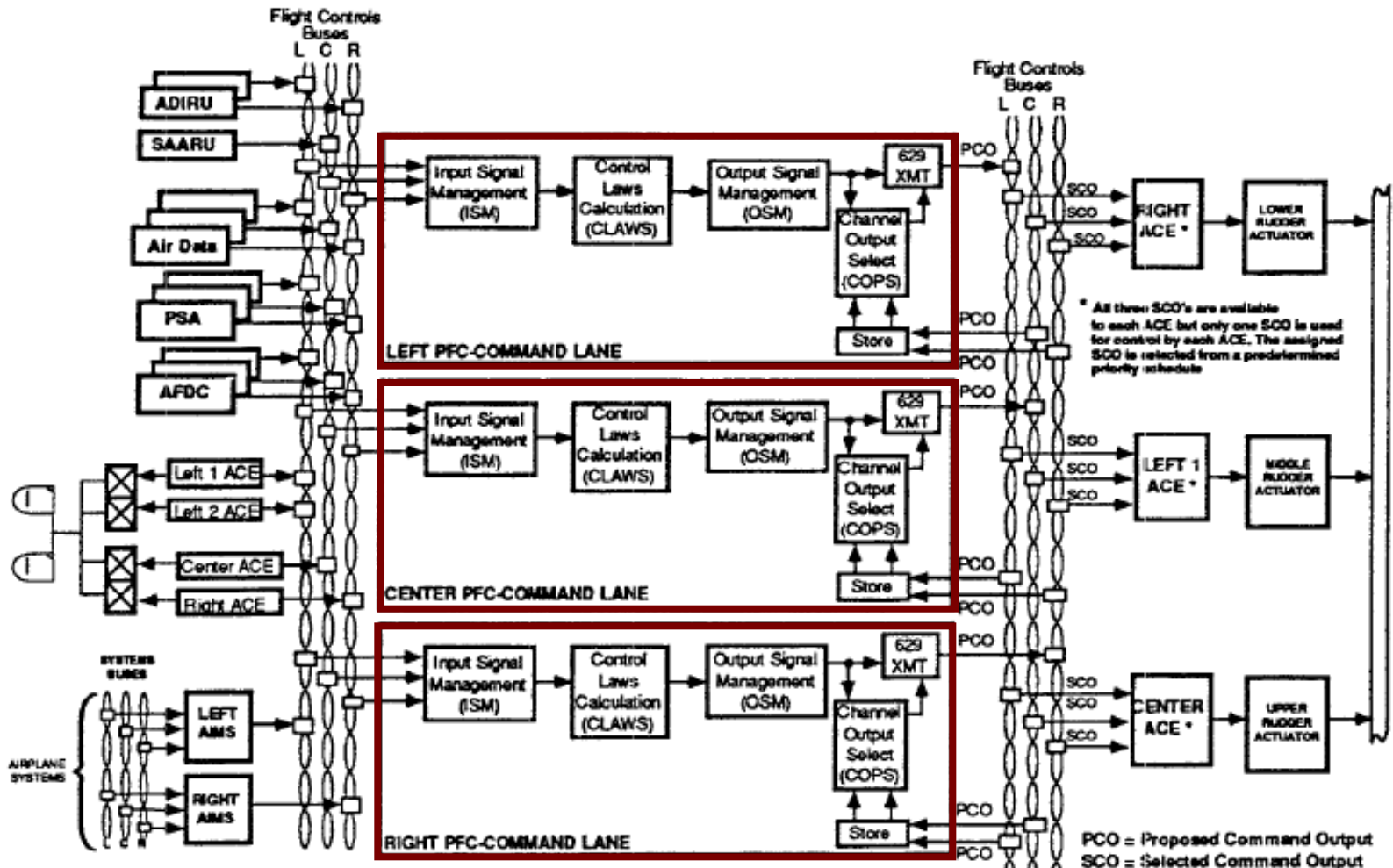


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# 777 Triple-Triple Architecture [Yeh, 96]



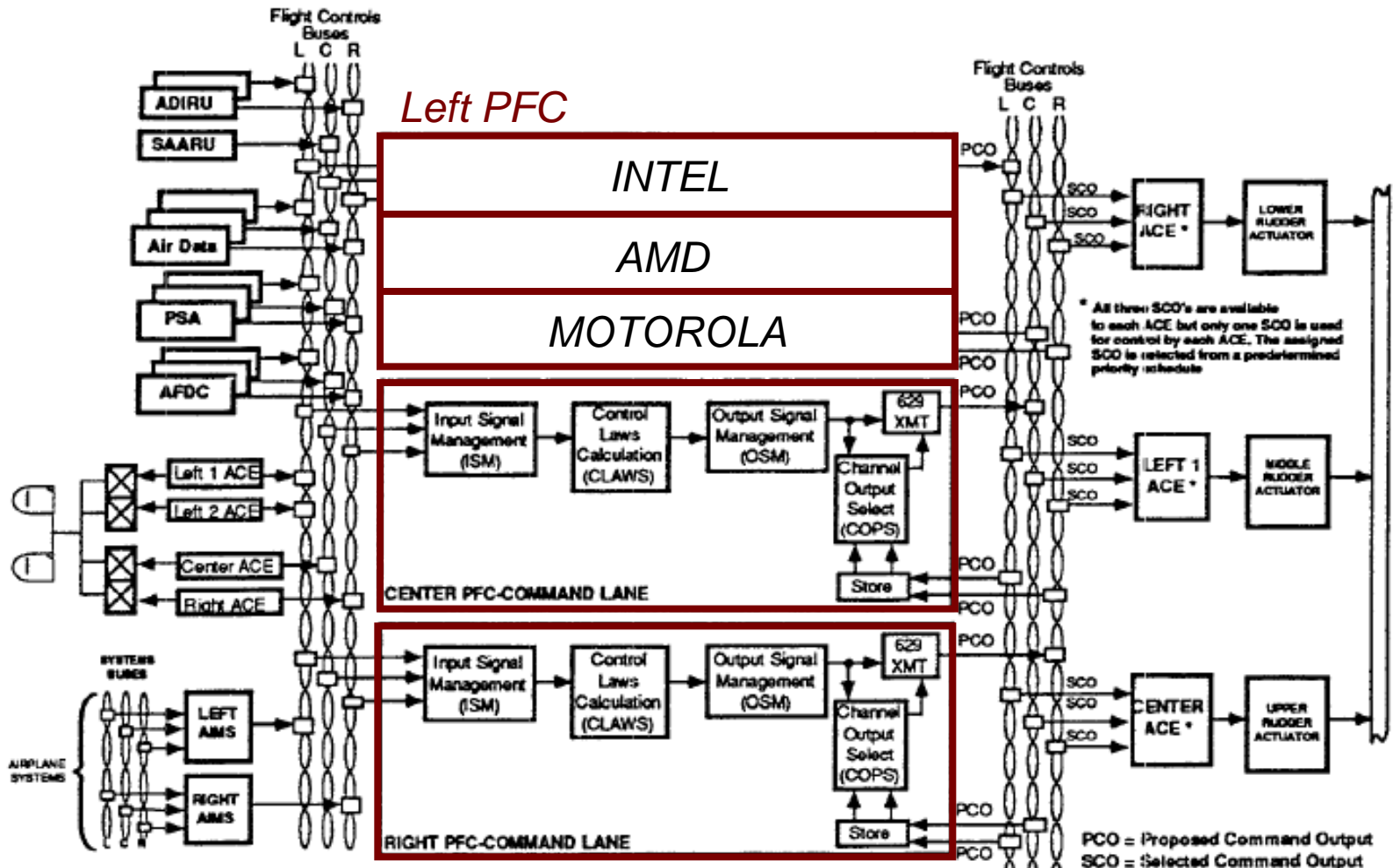
Sensors  
x3

Databus  
x3

*Triple-Triple  
Primary Flight  
Computers*

Actuator Electronics  
x4

# 777 Triple-Triple Architecture [Yeh, 96]



Sensors  
x3

Databus  
x3

*Triple-Triple  
Primary Flight  
Computers*

*Actuator Electronics  
x4*

# Reliability Comparison

## Boeing 777



### Reliability

- $< 10^{-9}$  catastrophic failures per hour
- No single point of failure
- Protect against random & common failures

### Design

- Hardware Redundancy
- Dissimilar hardware and software
- Limited use of analytical redundancy [1]
- Fault Trees, etc to certify

### References

[1] Goupil, "Oscillatory failure case detection in the A380 electrical flight control system by analytical redundancy," Control Engineering Practice, 2010.



# Reliability Comparison

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## Ultrastick 120



### Reliability

- $\sim 0.8$  failures/100 hrs [2]
- Single points of failure

### Design

- Limited by size, weight, power, and cost (SWAPC) constraints

## References

[1] Goupil, "Oscillatory failure case detection in the A380 electrical flight control system by analytical redundancy," Control Engineering Practice, 2010.

[2] Amos et al., "UAV for Reliability Build," Technical Report, University of Minnesota, 2014.

# Key Questions

## Boeing 777



### 1. What is an appropriate level of reliability for small UAS?

- FAA Modernization and Reform Act (1/12)
- FAA 14 CFR Part 107 (8/16)

### 2. Can analytical redundancy be used to increase the reliability of small UAS?

- Flight with a single aero surface [1]
- Fault detection of actuator failures [2,3,4]

### 3. How can analytical methods be certified?

- Probabilistic analysis methods and extended fault trees [5,6]

## Ultrastick 120



- [1] Venkataraman & Seiler, Safe Flight Using One Aerodynamic Control Surface, AIAA, 2016.  
[2] Venkataraman & Seiler, Model-Based Detection and Isolation of Rudder Faults for a Small UAS, AIAA, 2015.  
[3] Lakshminarayan, et al, "Designing Reliability Into Small UAS Avionics", Inside Unmanned Systems, 2016.  
[4] Bauer, et al, "Fault Detection and Basic In-Flight Reconfiguration of a Small UAV...", SafeProcess, 2018.  
[5] Venkataraman, et al, Reliability Assessment of Actuator Architectures for Unmanned Aircraft, AIAA, 2016.  
[6] Hu & Seiler, Pivotal decomposition for reliability analysis of fault tolerant control systems on UAVs, RESS, 2015.

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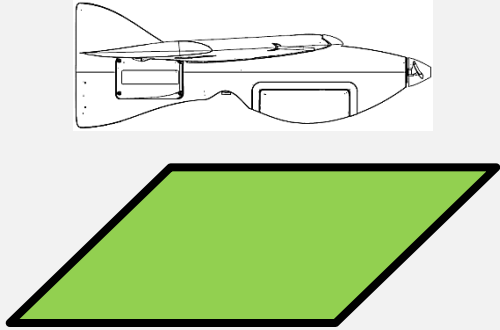
## Ultrastick 120



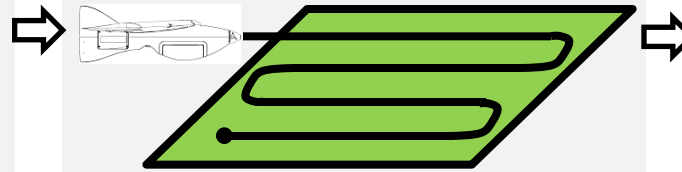
- [1] Venkataraman & Seiler, *Safe Flight Using One Aerodynamic Control Surface*, AIAA, 2016.
- [2] Venkataraman & Seiler, *Model-Based Detection and Isolation of Rudder Faults for a Small UAS*, AIAA, 2015.
- [3] Lakshminarayan, et al, "Designing Reliability Into Small UAS Avionics", *Inside Unmanned Systems*, 2016.
- [4] Bauer, et al, "Fault Detection and Basic In-Flight Reconfiguration of a Small UAV...", *SafeProcess*, 2018.
- [5] Venkataraman, et al, *Reliability Assessment of Actuator Architectures for Unmanned Aircraft*, AIAA, 2016.
- [6] Hu & Seiler, *Pivotal decomposition for reliability analysis of fault tolerant control systems on UAVs*, RESS, 2015.

# Final Goal

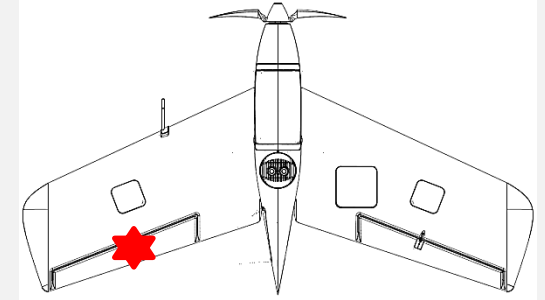
**Takeoff**



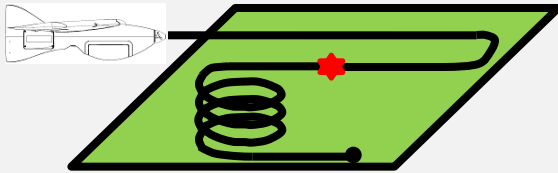
**Nominal mission**



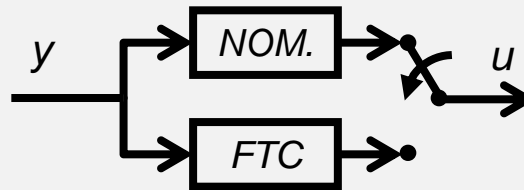
**Control surface fault**



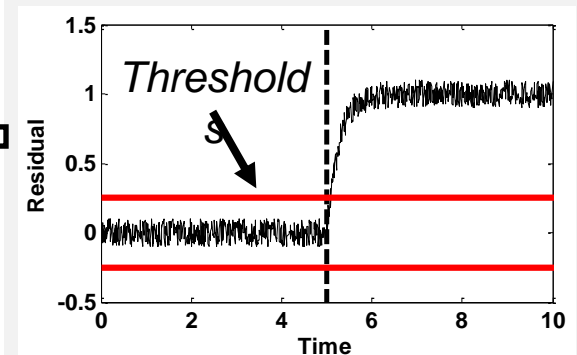
**Safe landing**



**Fault tolerant control**



**Fault diagnosis**



# Flight With One Aero Surface

## 1. Ultrastick 120 [1]

Demonstrated closed-loop steady, level flight (2015).

## 2. Senior Design [2]

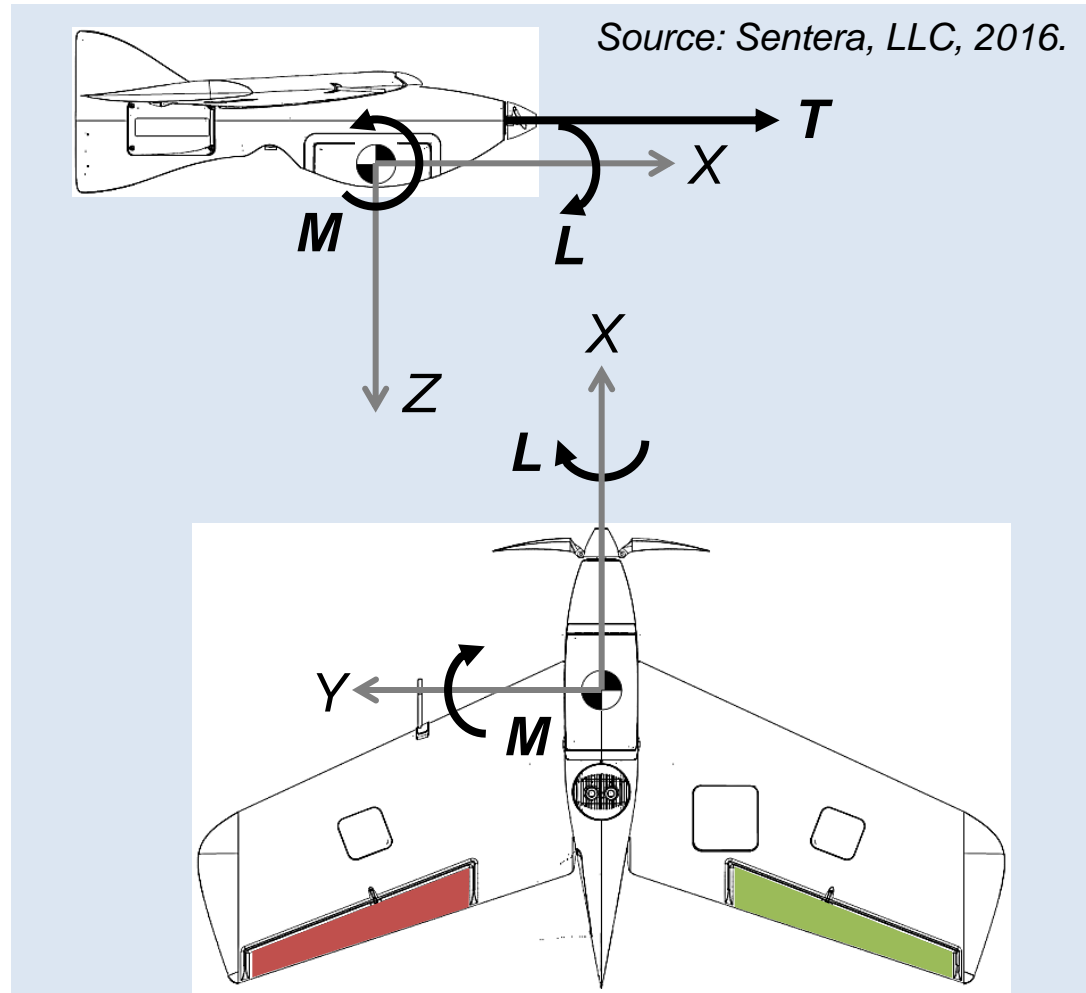
Team designed and built flying wing. Demonstrated ability to land by human pilot (2016).

## 3. Sentera Vireo

Built avionics and performed first flights for sys id (2016). Plan to demonstrate closed-loop landing (2017).

### References

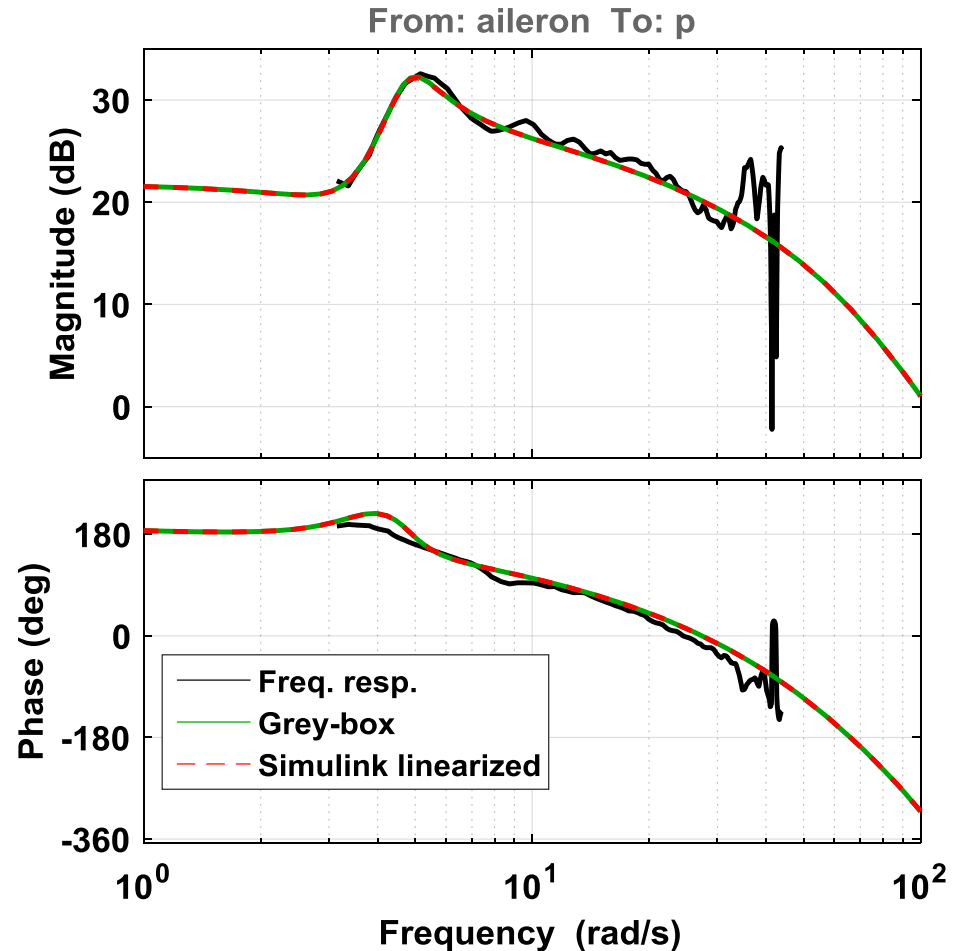
- [1] Venkataraman & Seiler, AIAA 2016.
- [2] Condrion, et al, UMN Report, 2016.



- *Control input simultaneously excites longitudinal and lateral-directional motion*
- *No direct yaw control*

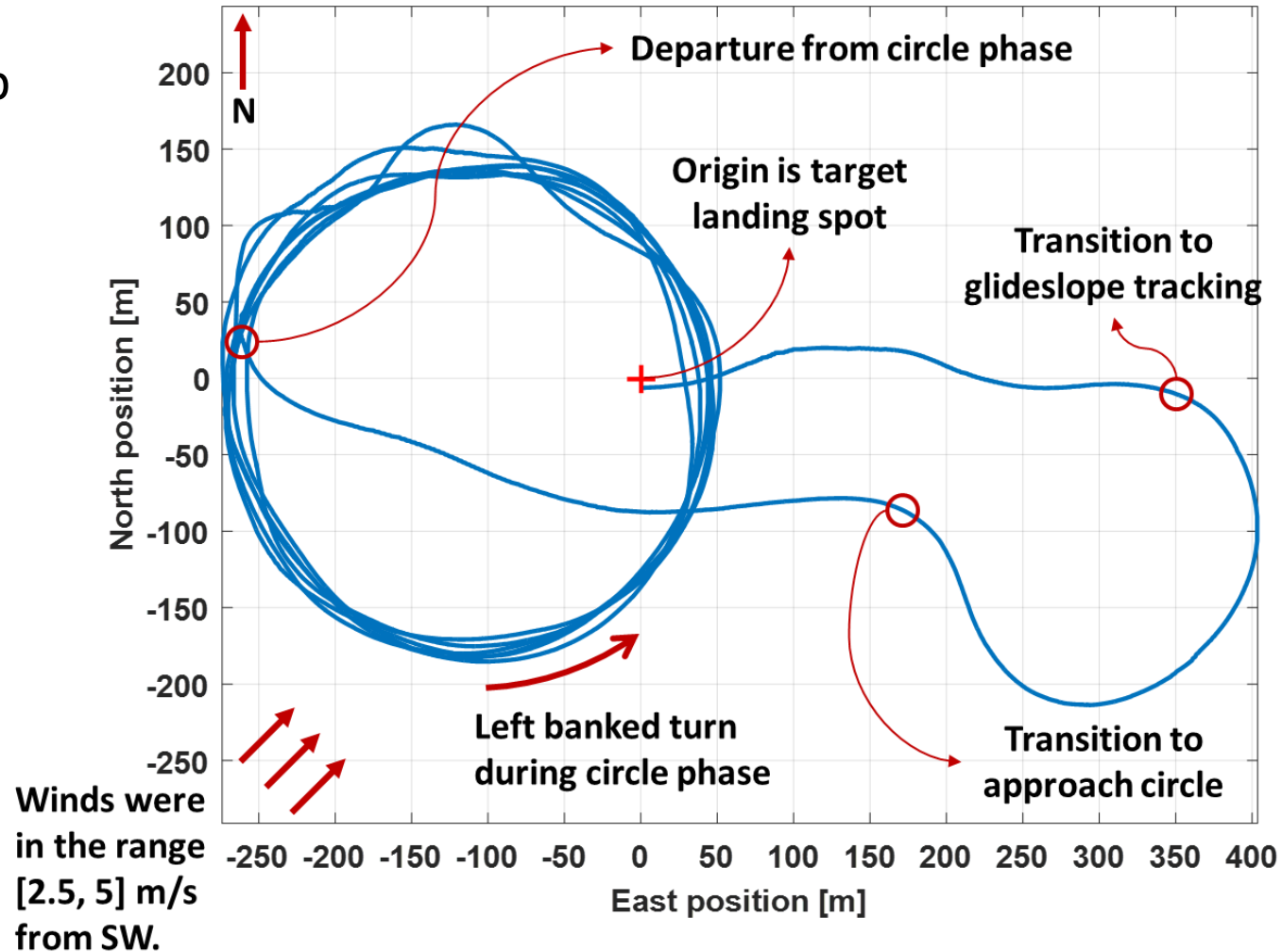
# System Identification

- Chirp excitations on elevator and aileron
- Identified frequency response from:
  - Elevator to pitch rate
  - Aileron to roll rate
- Grey-box modeling
  - Aero. Coeff. Initialized with using vortex-lattice method
  - Updated using flight data
- Plot shows aileron to roll rate
  - Dutch roll mode visible

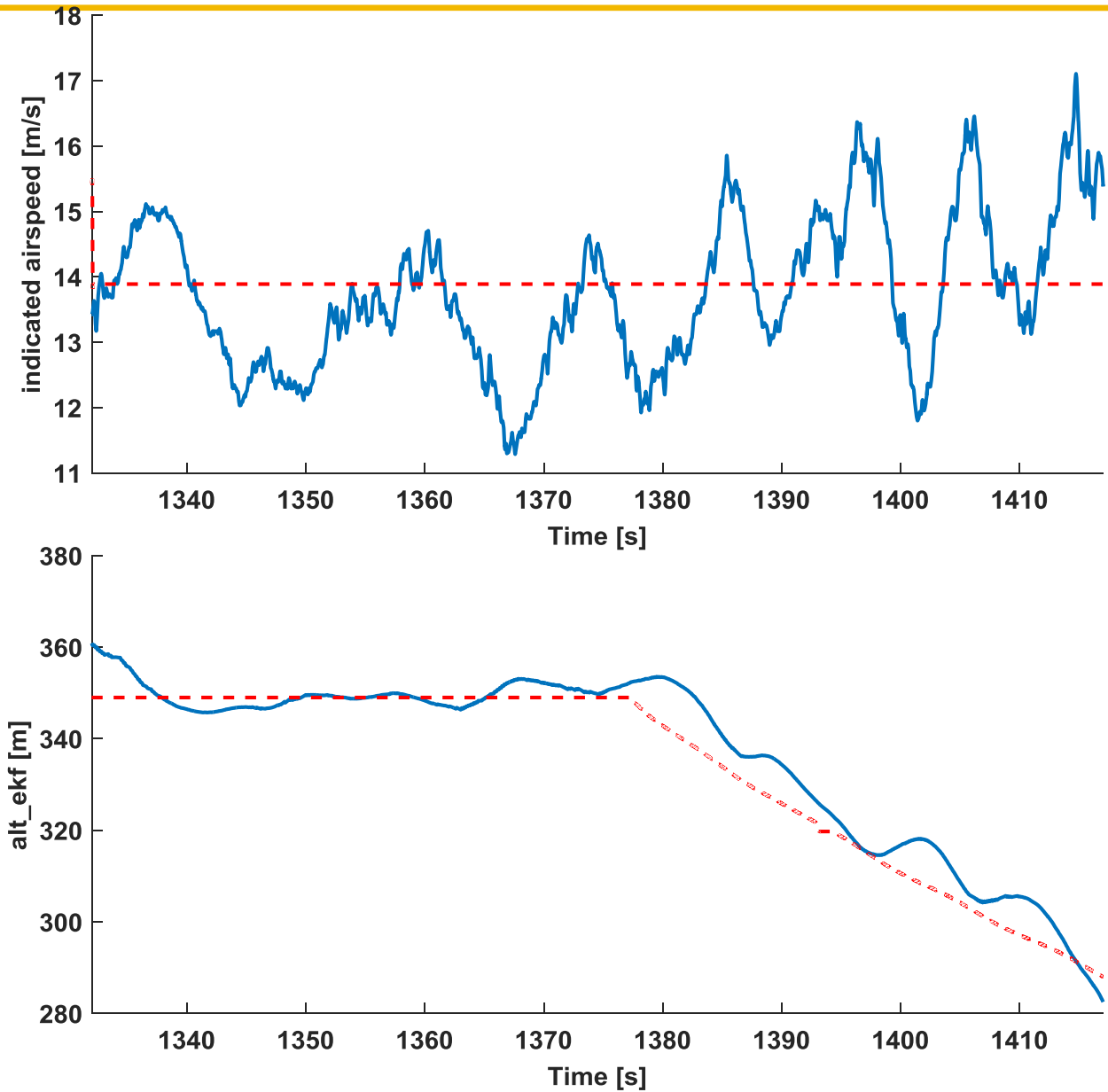


# Single Surface Flight

- Right elevon stuck at 5 deg trailing edge up
- Flight divided into circle (set by user) and land phases
- The red plus sign is the target touch-down point

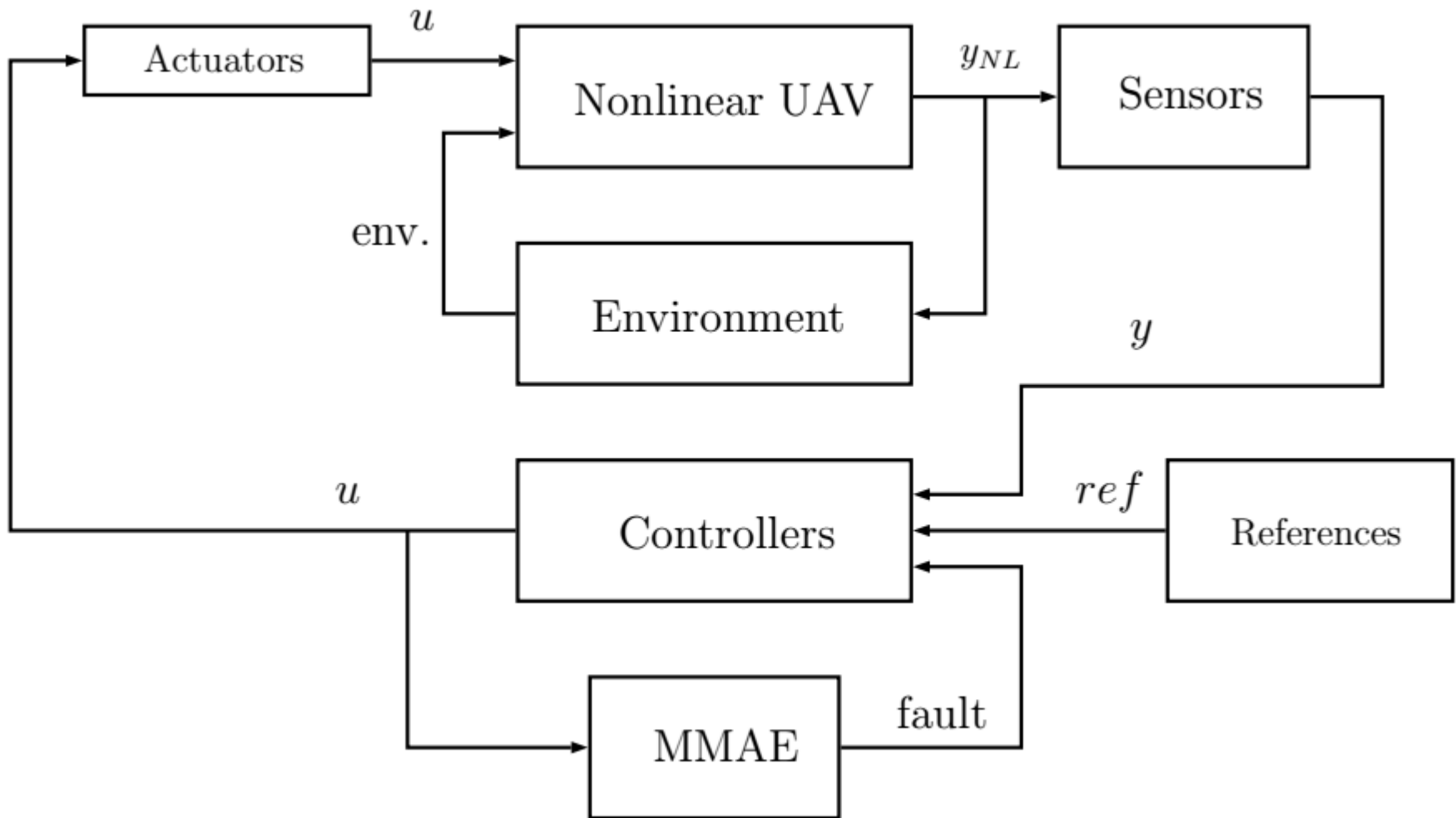


# Glideslope Tracking



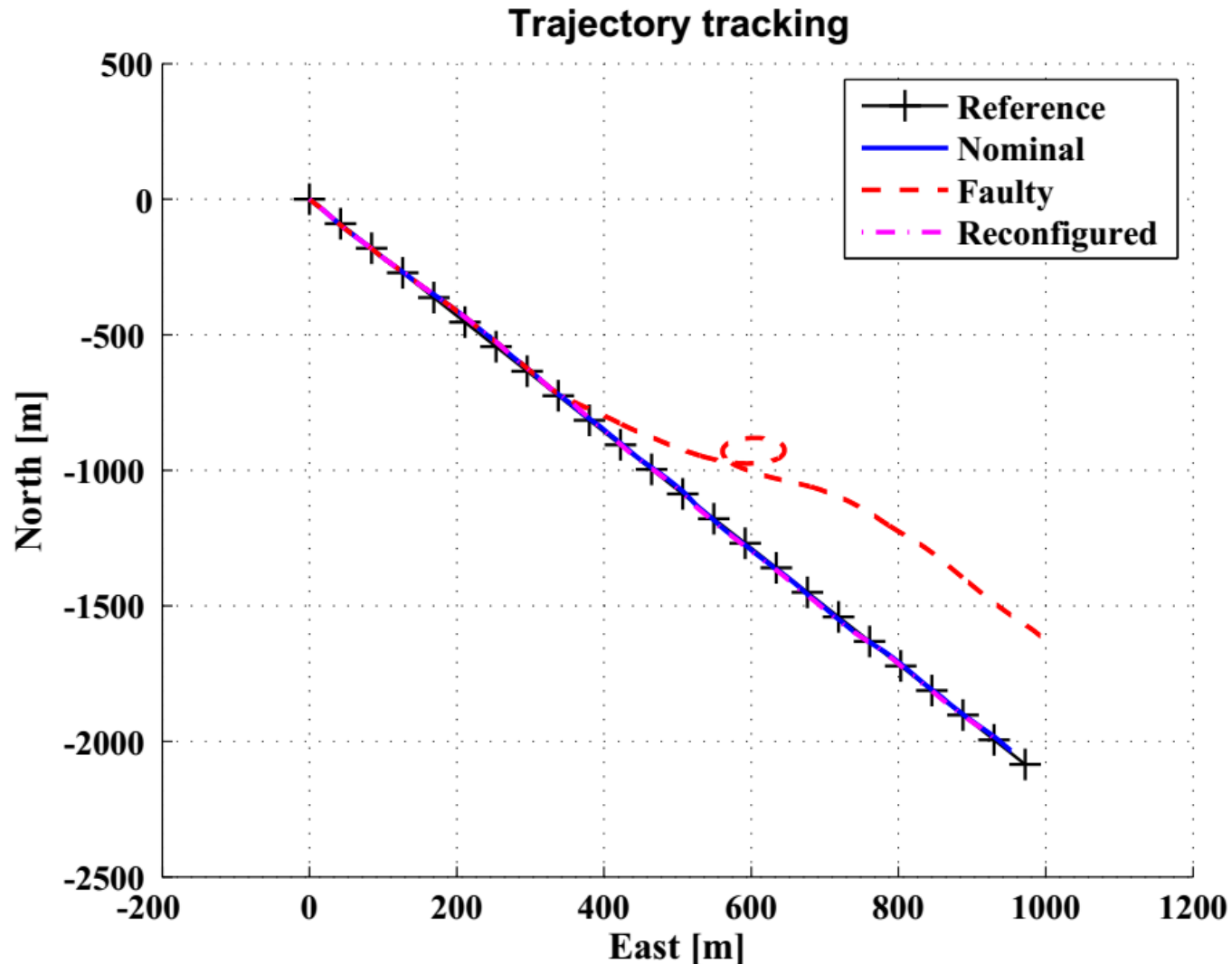


# Fault Detection and Reconfiguration



**Reference:** Bauer, et al, “Fault Detection and Basic In-Flight Reconfiguration of a Small UAV Equipped with Elevons”, SafeProcess, 2018.

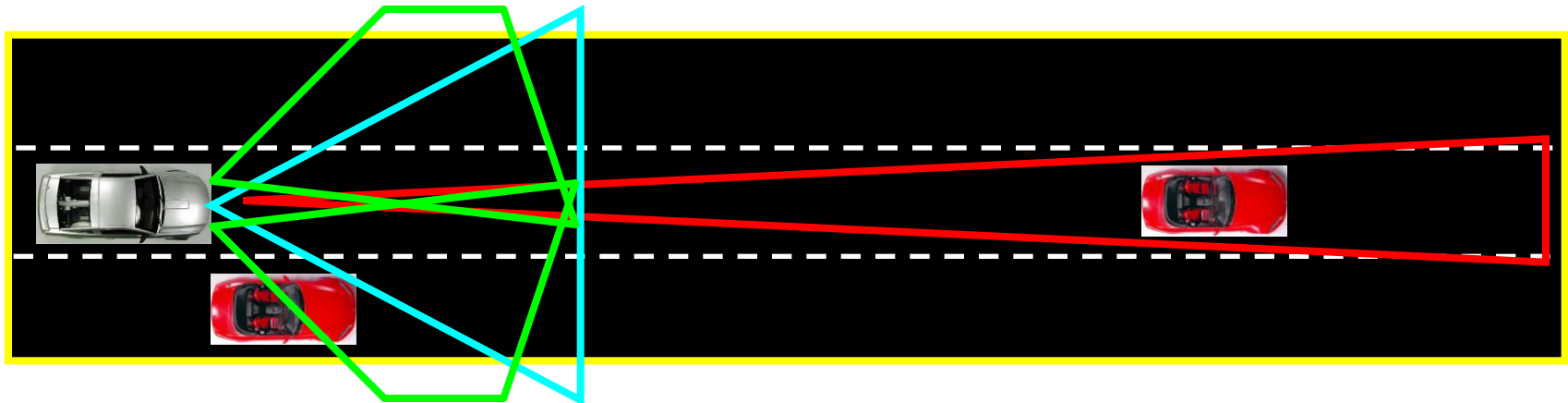
# Fault Detection and Reconfiguration



**Reference:** Bauer, et al, "Fault Detection and Basic In-Flight Reconfiguration of a Small UAV Equipped with Elevons", SafeProcess, 2018.

# From Aerospace to Automotive....

Similar reliability concerns are now common in automotive applications due to rise of autonomous driving.



# Performance Adaptive Aeroelastic Wing (PAAW)

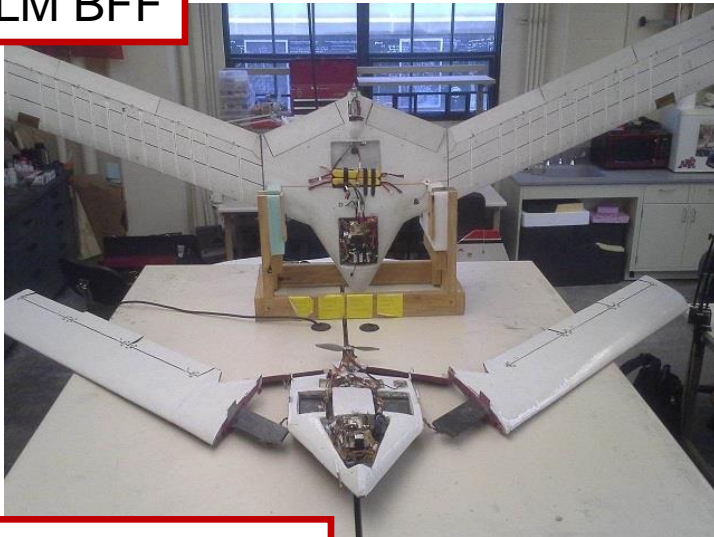
- Goal: Suppress flutter, control wing shape and alter shape to optimize performance
  - Funding: NASA NRA NNX14AL36A
  - Technical Monitor: Dr. Jeffrey Ouellette
  - Two years of testing at UMN followed by two years of testing on NASA's X-56 Aircraft



*Schmidt & Associates*



LM BFF



LM/NASA X-56

UMN Mini-Mutt

# Aeroservoelasticity (ASE)

---

## Efficient aircraft design

- Lightweight structures
- High aspect ratios



Source: [www.flightglobal.com](http://www.flightglobal.com)

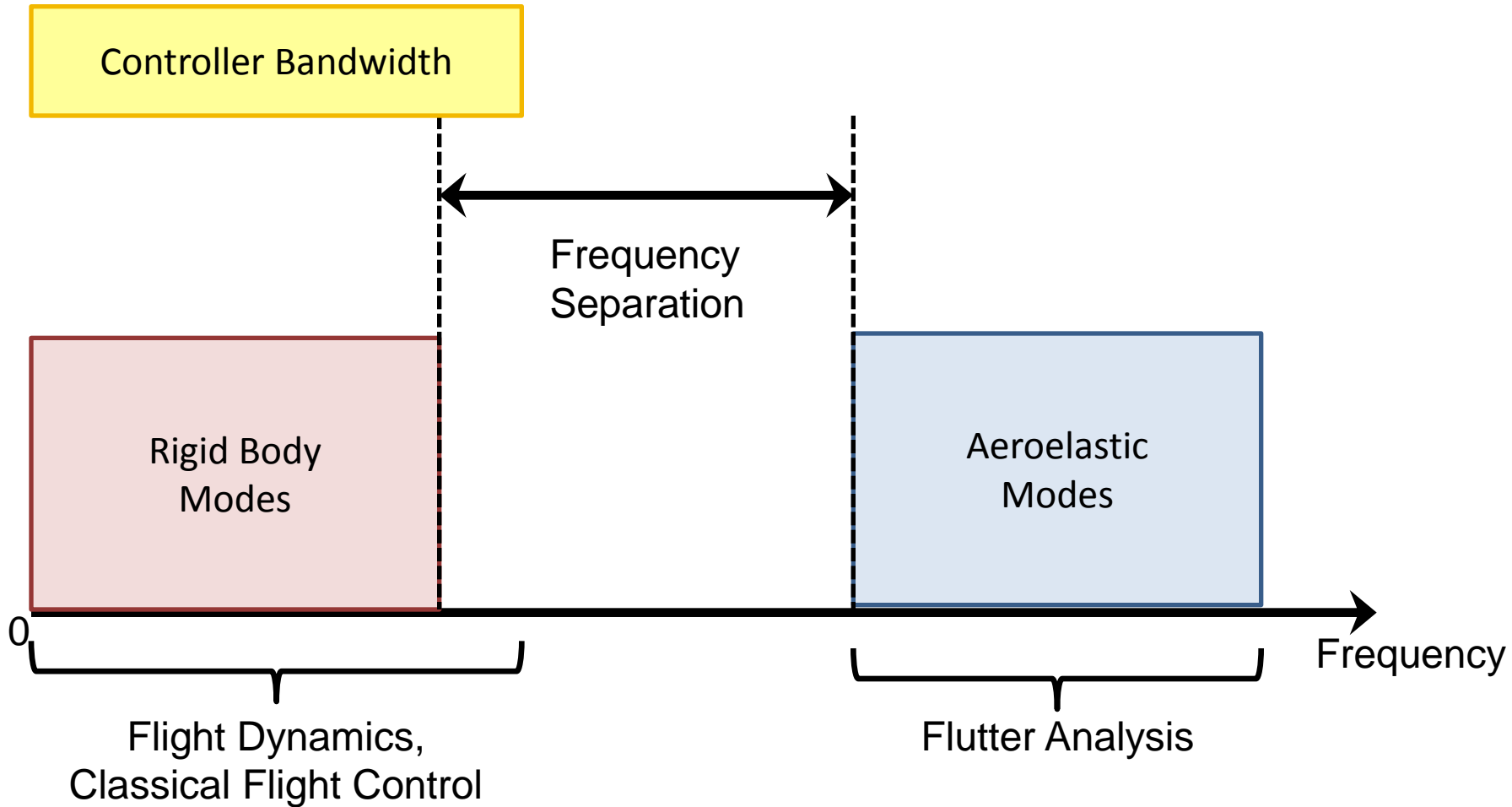
# Flutter

---



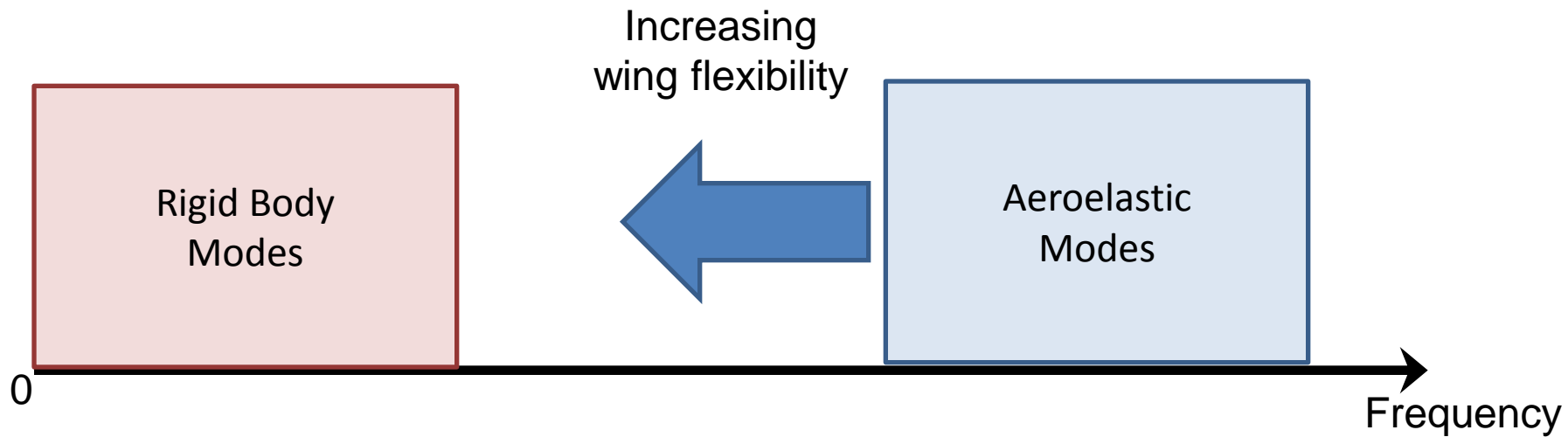
Source: NASA Dryden Flight Research

# Classical Approach



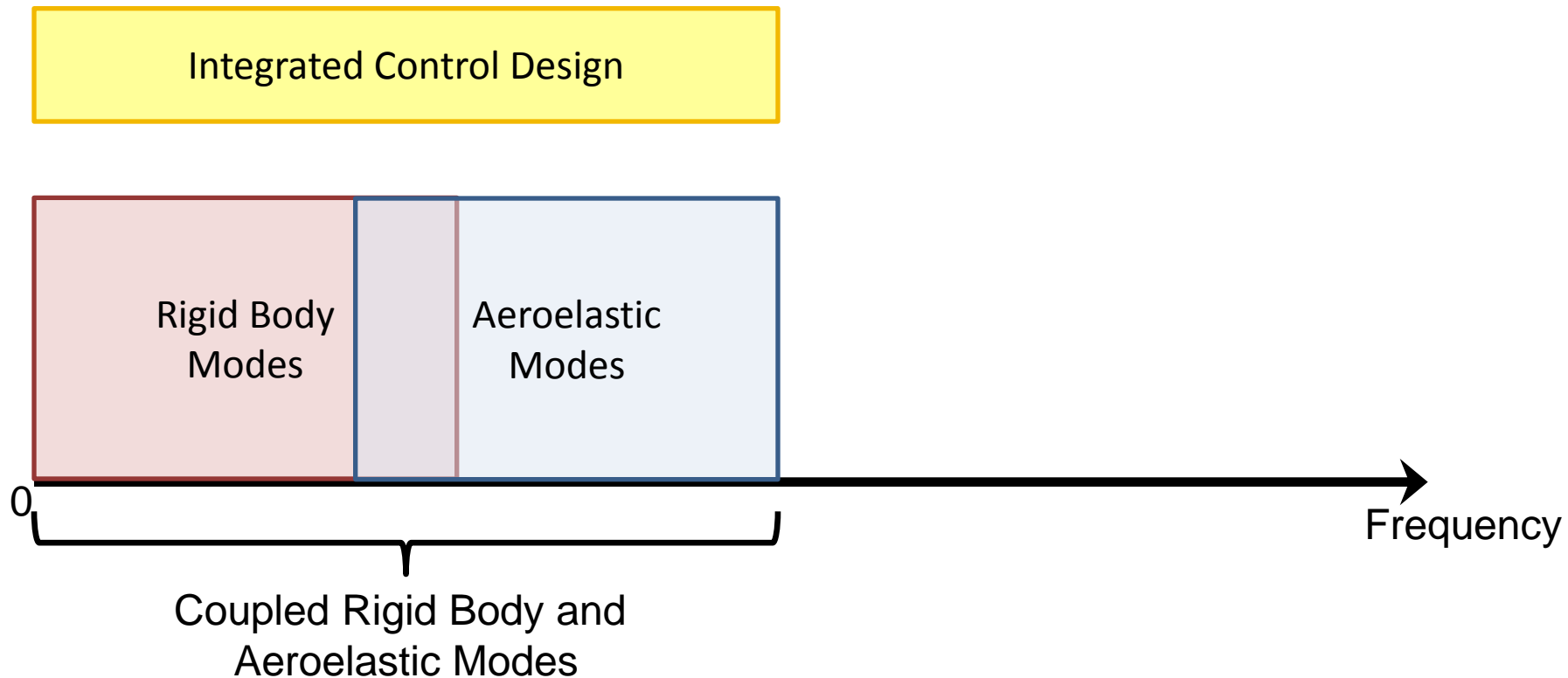
# Flexible Aircraft Challenges

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# Flexible Aircraft Challenges



# Modeling and Control for Flex Aircraft

## 1. Parameter Dependent Dynamics

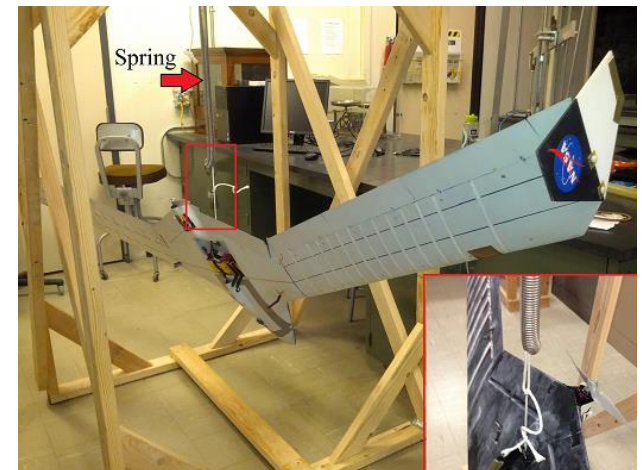
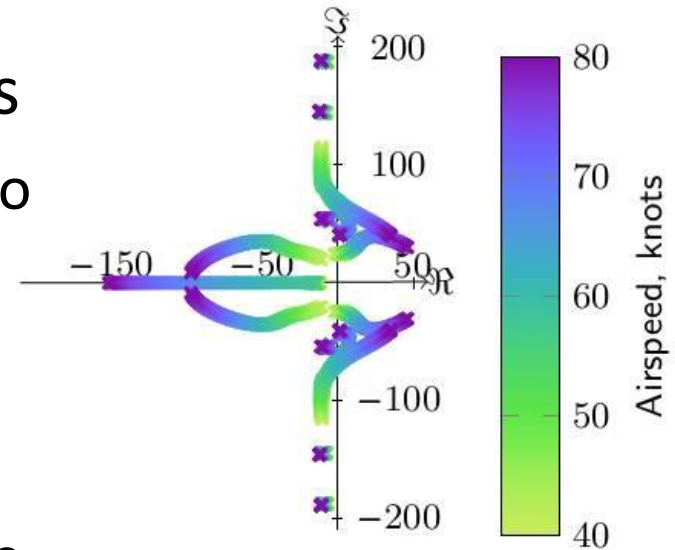
- Models depend on airspeed due to structural/aero interactions
- LPV is a natural framework.

## 2. Model Reduction

- High fidelity CFD/CSD models have many (millions) of states.

## 3. Model Uncertainty

- Use of simplified low order models OR reduced high fidelity models
- Unsteady aero, mass/inertia & structural parameters



# Current PAAW Aircraft



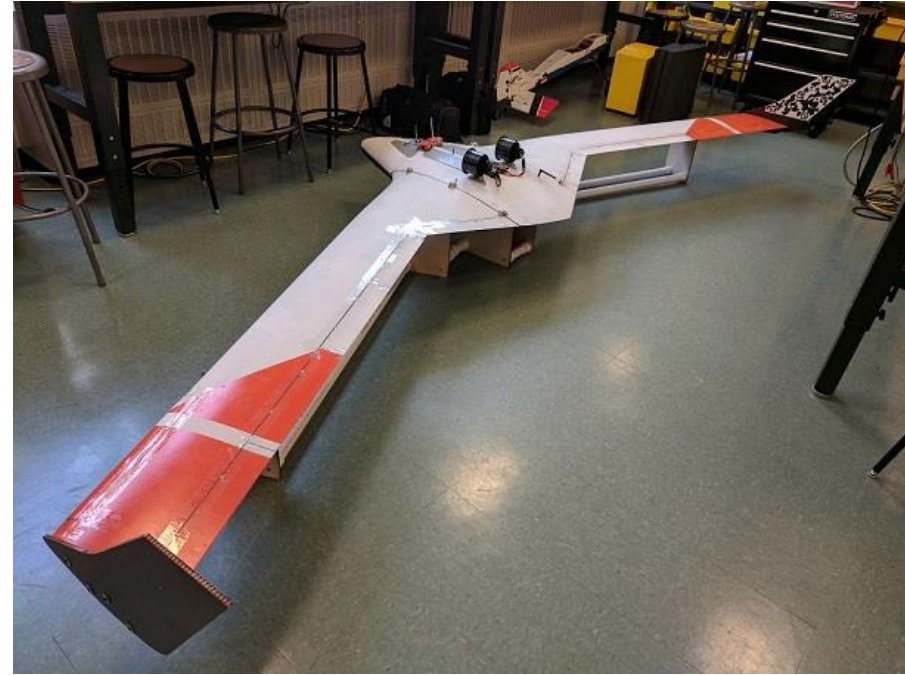
## **mAEWing1**

*10 foot wingspan*

*~14 pounds*

*Laser-scan replica of BFF*

*4 aircraft, >50 flights*



## **mAEWing2**

*14 foot wingspan*

*~42 pounds*

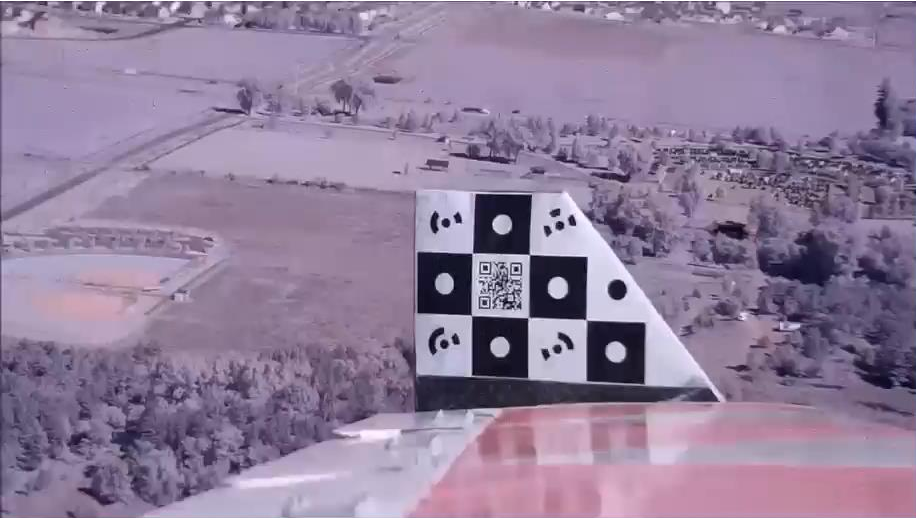
*Half-scale X-56*

*Currently ground testing*

# mAEWing1 and 2



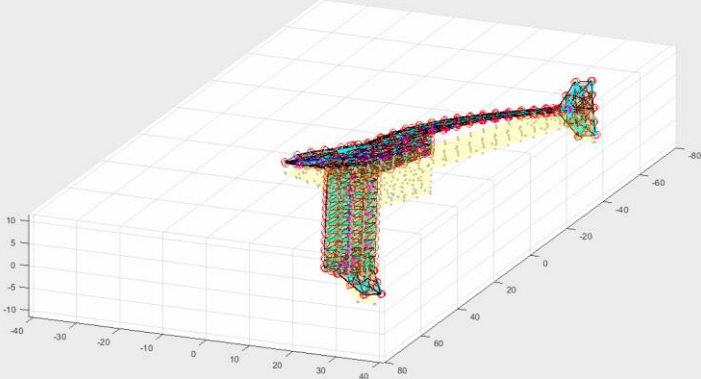
# Open-Loop Flutter



# Body Freedom Flutter

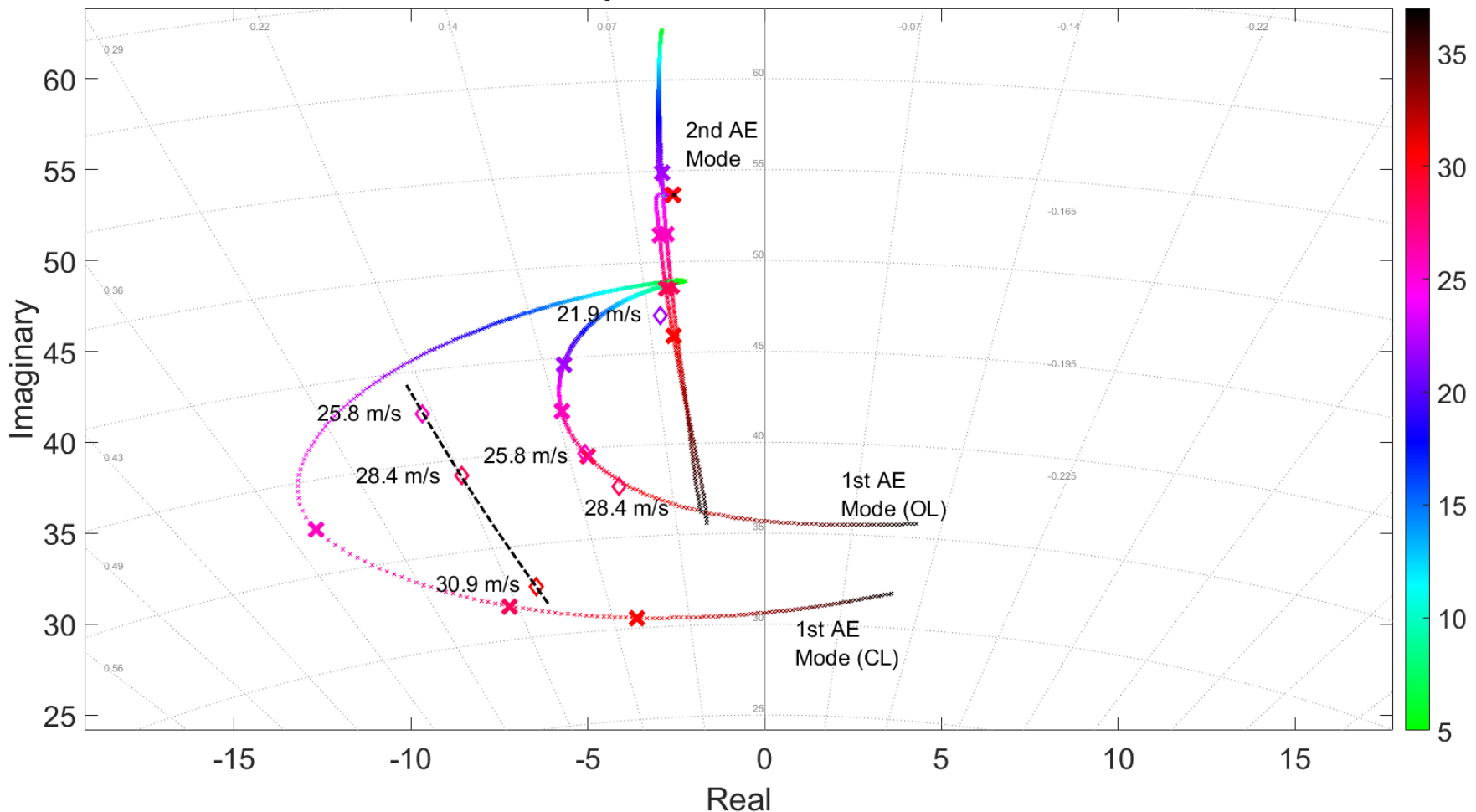


Velocity = 100.127 ft/s  
Mode 13: freq = 27.1046,  $\zeta = 0.02195$



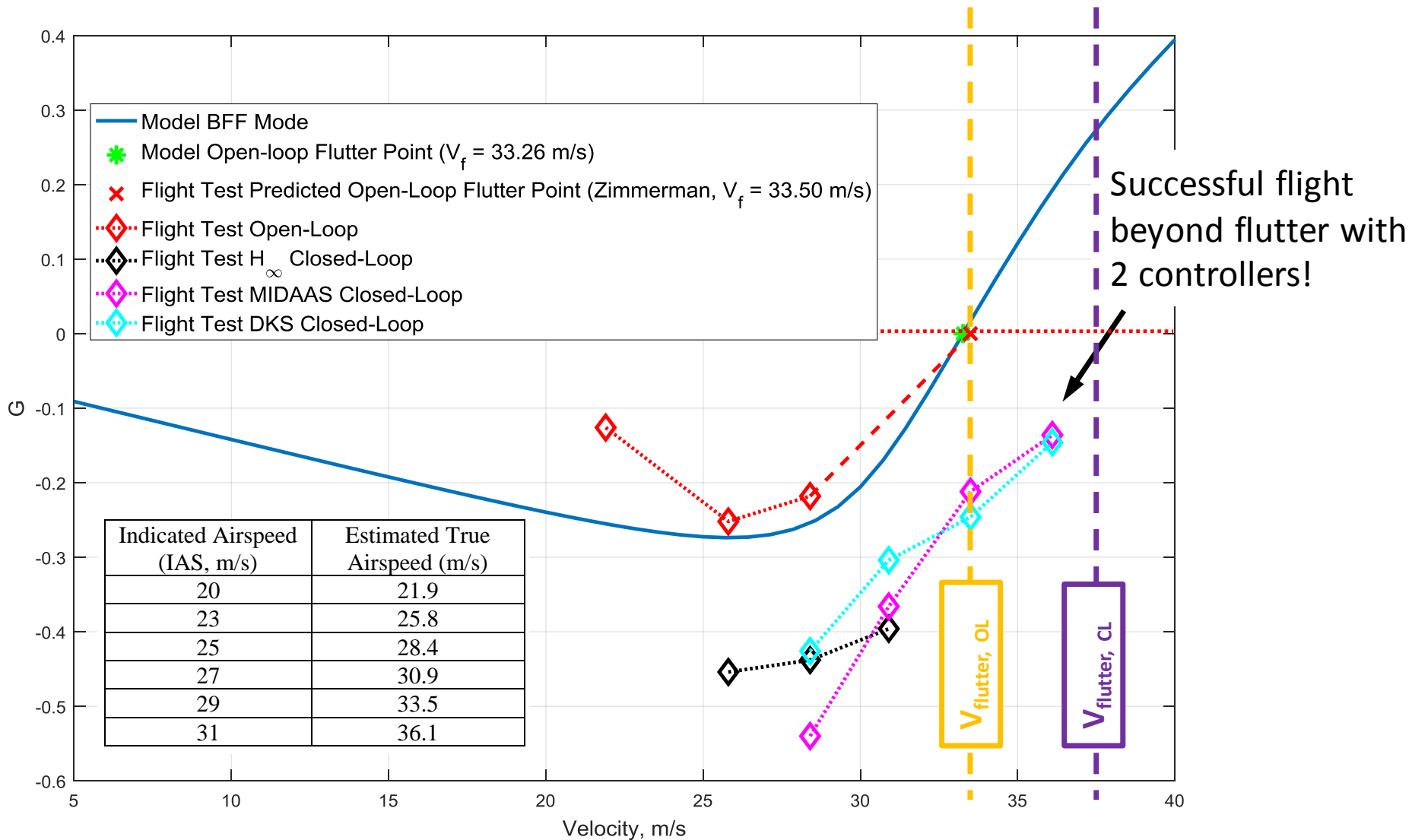
# Pole Map for H-Inf Controller

Map of Poles and Zeros



Comparison of BFF mode variation with airspeed I.D.'d from flight test data with theoretical predictions for Open Loop and  $H^\infty$  controller; Marker descriptions – (X): theoretical poles, ( $\diamond$ ): sys. I.D.'d open/closed loop poles.

# Flight Test Summary





# Finite Horizon Robustness Analysis of LTV Systems Using Integral Quadratic Constraints

Peter Seiler  
University of Minnesota



M. Moore, C. Meissen, M. Arcak, and A. Packard  
University of California, Berkeley



MTA Sztaki  
October 5, 2017

# Time-Varying Systems



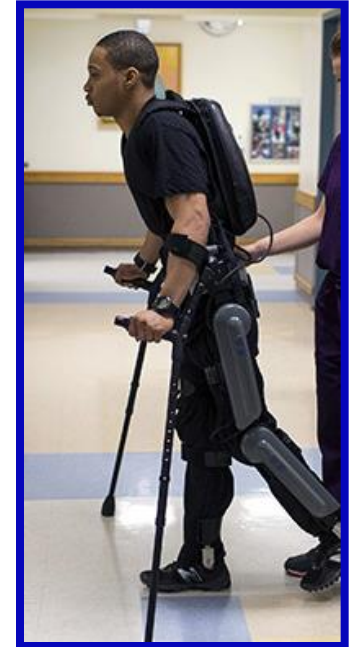
**Wind Turbine**  
Periodic /  
Parameter-Varying



**Flexible Aircraft**  
Parameter-Varying



**Vega Launcher**  
Time-Varying  
(Source: ESA)



**Robotics**  
Time-Varying  
(Source: ReWalk)

**Issue:** Few numerically reliable methods to assess the robustness of time-varying systems.

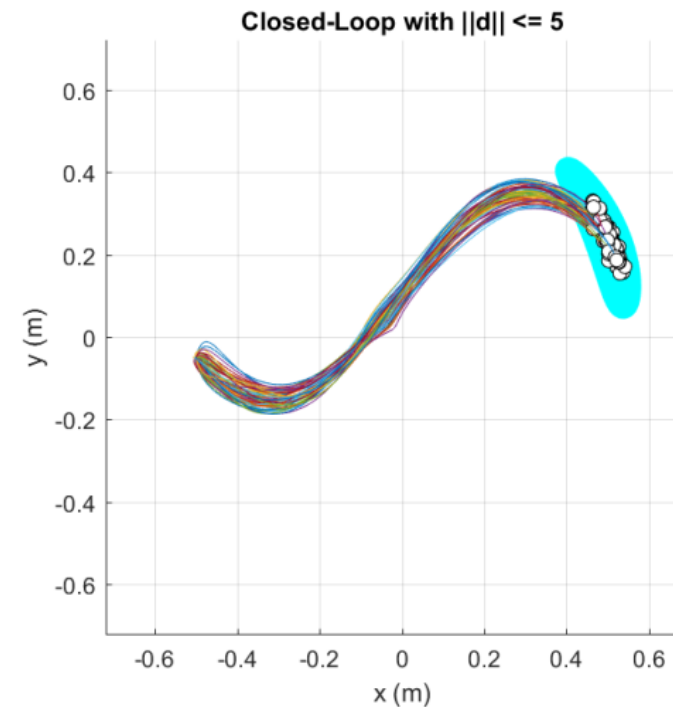
# Analysis Objective

**Goal:** Assess the robustness of linear time-varying (LTV) systems on finite horizons.

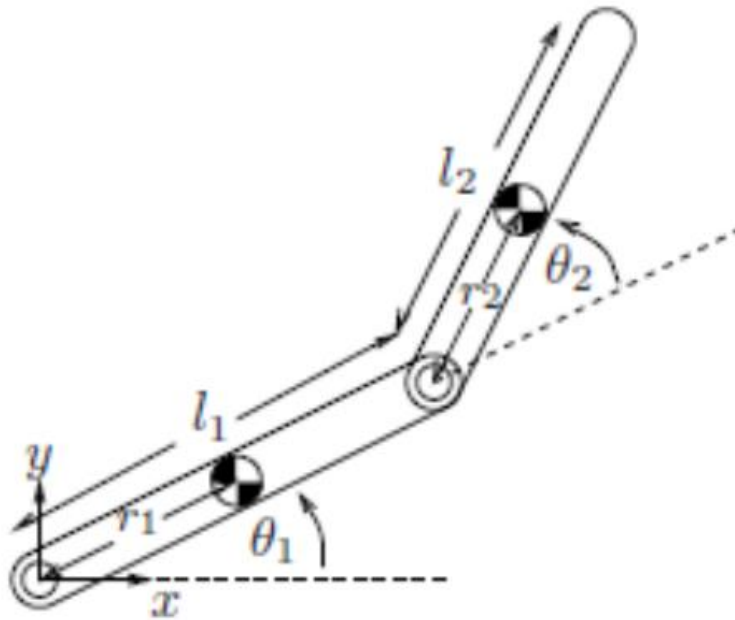
**Approach:** Classical Gain/Phase Margins focus on (infinite horizon) stability and frequency domain concepts.

Instead focus on:

- Finite horizon metrics, e.g. induced gains and reachable sets.
- Effect of disturbances and model uncertainty (D-scales, IQCs, etc).
- Time-domain analysis conditions.



# Two-Link Robot Arm



Two-Link Diagram [MZS]

Nonlinear dynamics [MZS]:

$$\dot{\eta} = f(\eta, \tau, d)$$

where

$$\eta = [\theta_1, \dot{\theta}_1, \theta_2, \dot{\theta}_2]^T$$

$$\tau = [\tau_1, \tau_2]^T$$

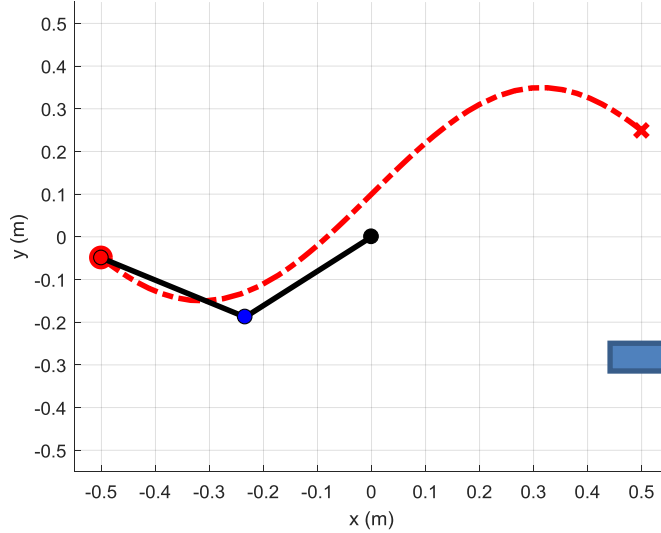
$$d = [d_1, d_2]^T$$

$\tau$  and  $d$  are control torques and disturbances at the link joints.

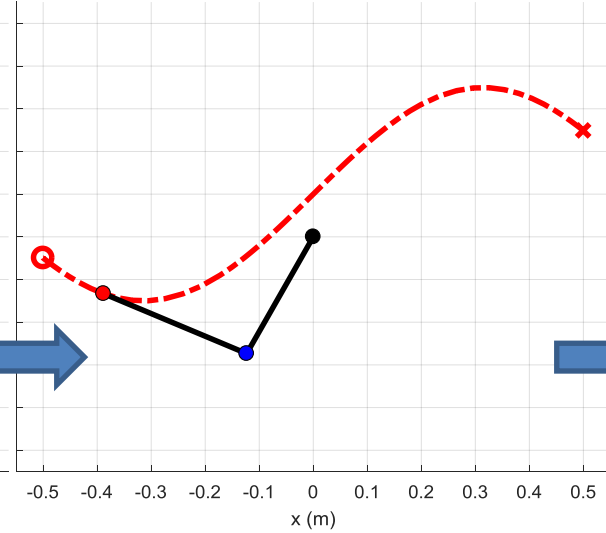
[MZS] R. Murray, Z. Li, and S. Sastry. *A Mathematical Introduction to Robot Manipulation*, 1994.

# Nominal Trajectory (Cartesian Coords.)

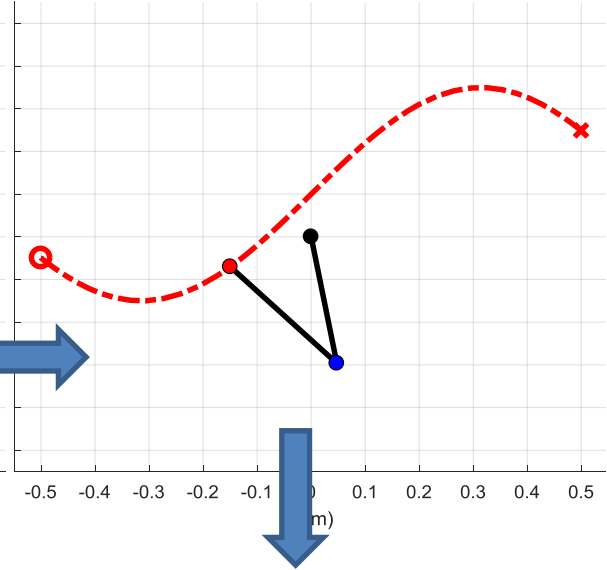
Two Link Robot at t=0sec



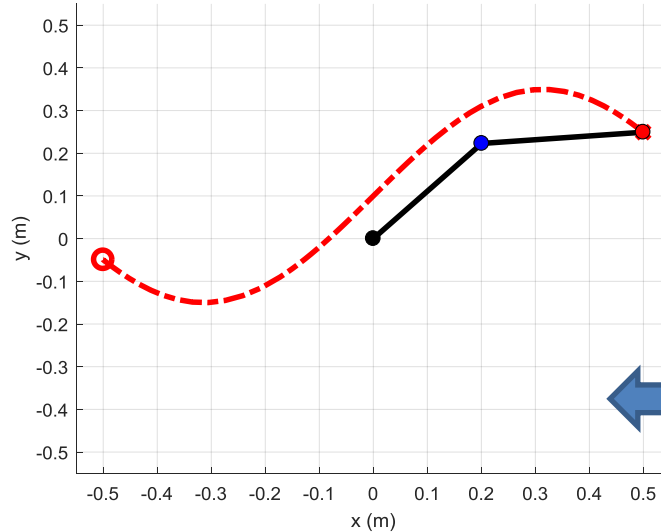
Two Link Robot at t=1sec



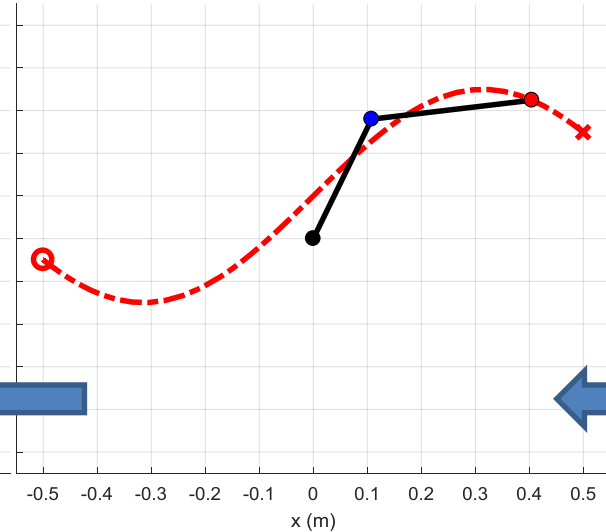
Two Link Robot at t=2sec



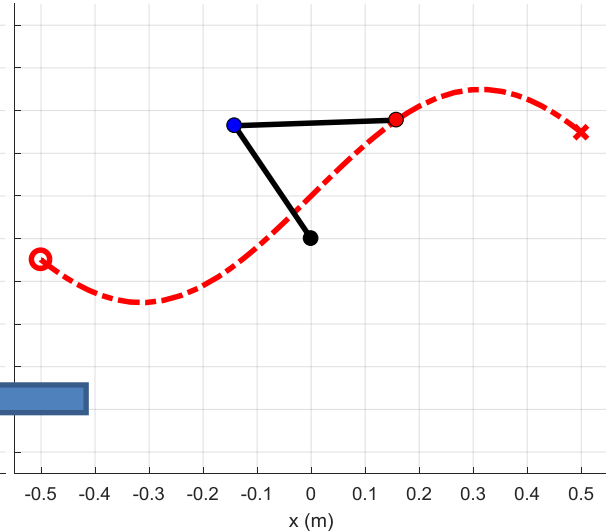
Two Link Robot at t=5sec



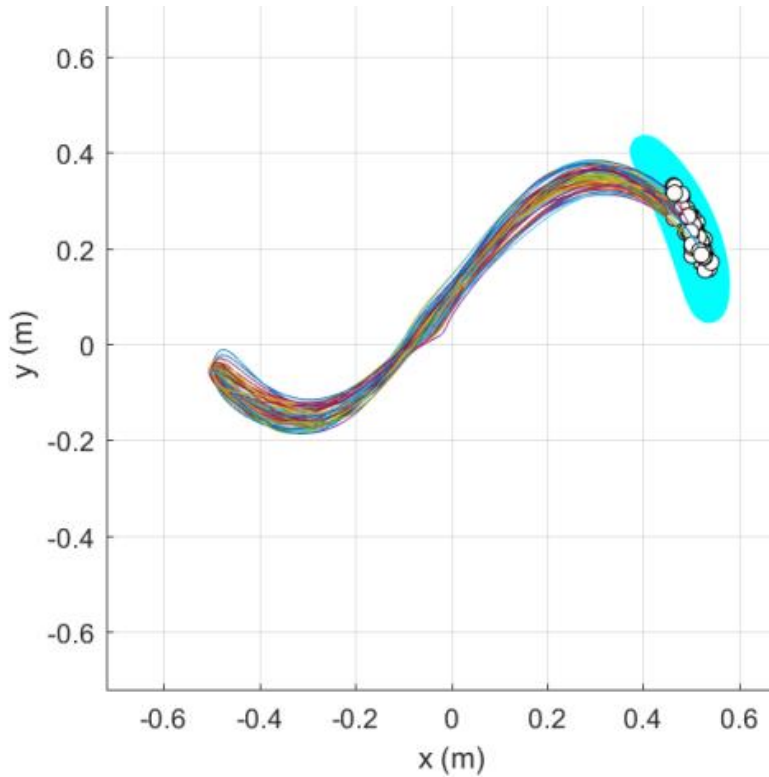
Two Link Robot at t=4sec



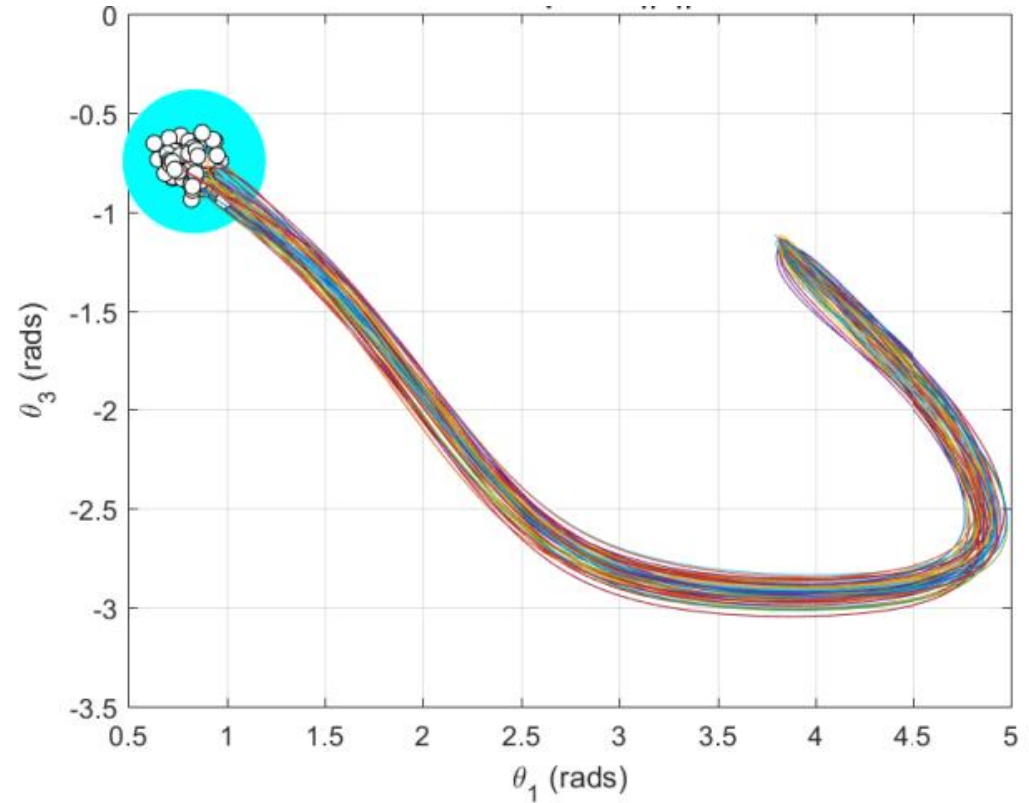
Two Link Robot at t=3sec



# Effect of Disturbances / Uncertainty



**Cartesian Coords.**



**Joint Angles**

# Overview of Analysis Approach

Nonlinear dynamics:

$$\dot{\eta} = f(\eta, \tau, d)$$

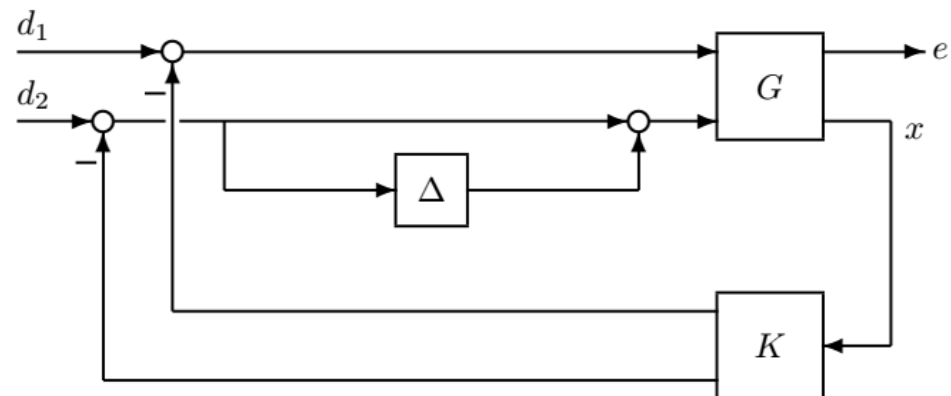
Linearize along a (finite –horizon) trajectory  $(\bar{\eta}, \bar{\tau}, d = 0)$

$$\dot{x} = A(t)x + B(t)u + B(t)d$$

Compute bounds on the terminal state  $x(T)$  or other quantity  $e(T) = C x(T)$  accounting for disturbances and uncertainty.

Comments:

- The analysis can be for open or closed-loop.
- LTV analysis complements the use of Monte Carlo simulations.



# Conclusions

---

- Fault tolerance for small UAVs
  - Commercial aircraft achieve high reliability with redundancy.
  - Model-based fault detection methods are an alternative that enables size, weight, power, and cost to be reduced.
  - Develop methods for analytical fault tolerance on small UAS and tools to certify the probabilistic performance.
- Modeling and control of flexible aircraft
- Robustness analysis of time-varying systems

<http://www.aem.umn.edu/~SeilerControl/>



# Acknowledgements

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- US National Science Foundation
  - Grant No. NSF-CMMI-1254129: “CAREER: Probabilistic Tools for High Reliability Monitoring and Control of Wind Farms.” Prog. Manager: J. Berg.
  - Grant No. NSF/CNS-1329390: “CPS: Breakthrough: Collaborative Research: Managing Uncertainty in the Design of Safety-Critical Aviation Systems”. Prog. Manager: D. Corman.
- NASA
  - NRA NNX14AL36A: "Lightweight Adaptive Aeroelastic Wing for Enhanced Performance Across the Flight Envelope," Tech. Monitor: J. Ouelette.
  - NRA NNX12AM55A: “Analytical Validation Tools for Safety Critical Systems Under Loss-of-Control Conditions.” Tech. Monitor: C. Belcastro.
  - SBIR contract #NNX12CA14C: “Adaptive Linear Parameter-Varying Control for Aeroservoelastic Suppression.” Tech. Monitor. M. Brenner.
- Eolos Consortium and Saint Anthony Falls Laboratory
  - <http://www.eolos.umn.edu/> & <http://www.safl.umn.edu/>

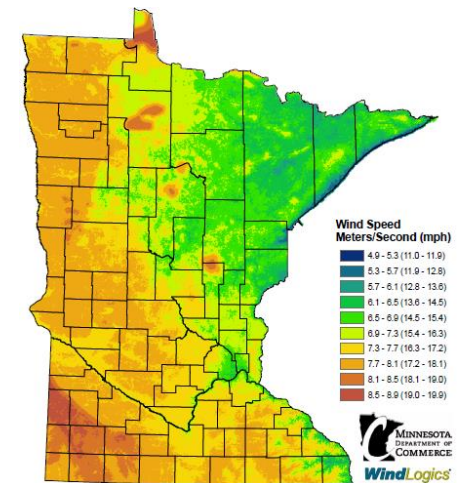
# Backup

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# Modeling and Control for Wind Energy

Jen Annoni, Shu Wang, Daniel Ossmann, Parul Singh,  
Jordan Hoyt, Sanjana Vijayshankar  
(with support from SAFL/EOLOS)



## Clipper Liberty, 2012:

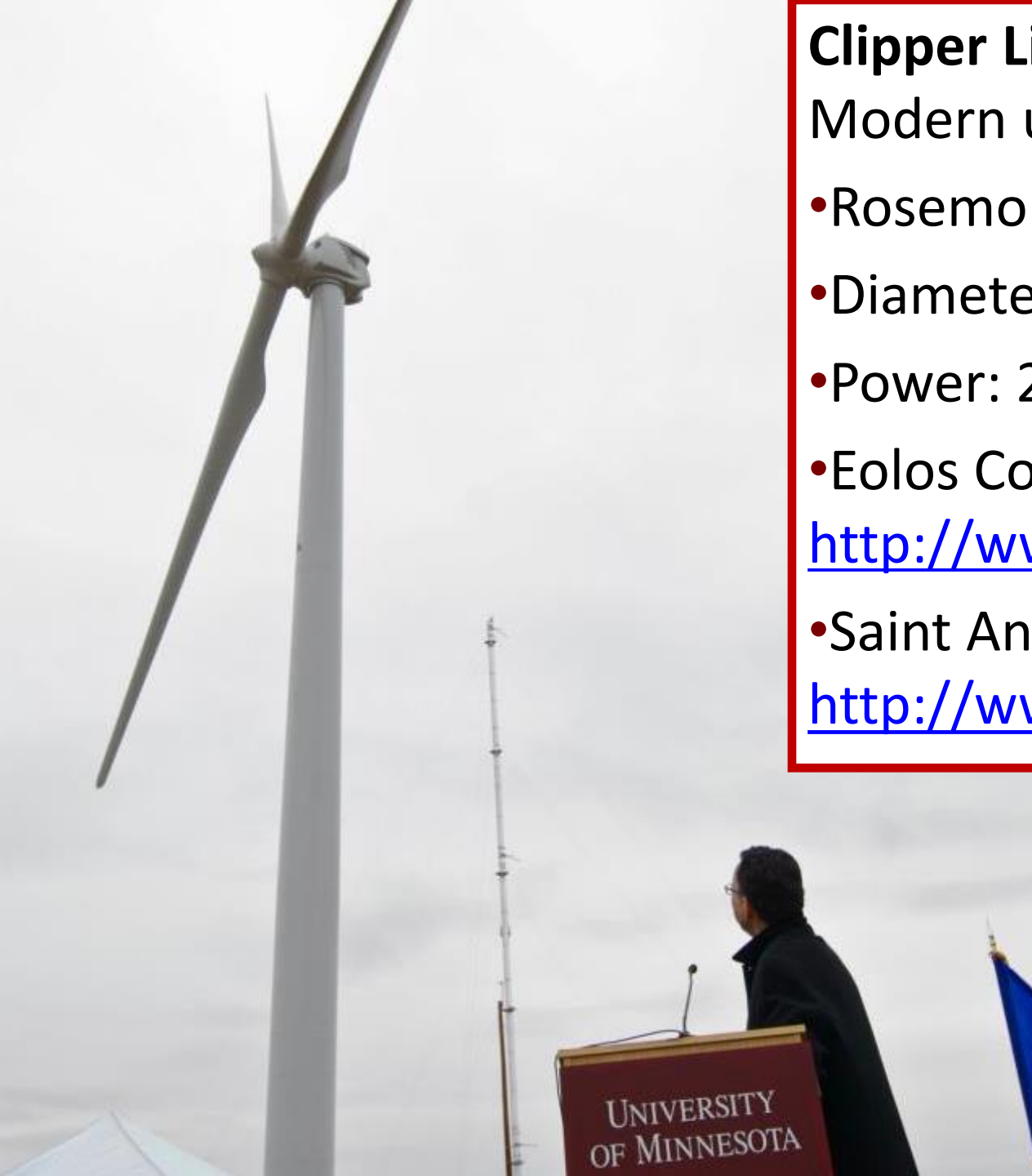
Modern utility-scale turbine.

- Rosemount, MN.
- Diameter: 96m
- Power: 2.5MW
- Eolos Consortium:

<http://www.eolos.umn.edu/>

- Saint Anthony Falls Lab:

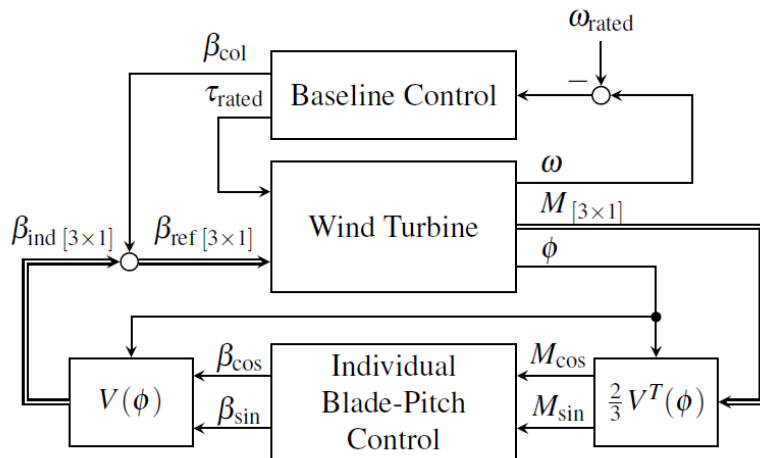
<http://www.safl.umn.edu/>



# Individual Blade Pitch Control

## Goals:

- Reducing structural loads on the turbine **to**
- increase life time of turbine and components **while**
- keeping power production constant **by**
- adding an individual blade pitch controller



Controller architecture

Implementation  
2017



C96 Liberty research turbine

Ref: Ossmann, Theis, Seiler, '16 ASME DSCC, Best Energy Paper Award

# Modeling and Control for Wind Farms

## 1. Parameter Dependent Dynamics

- Models depend on windspeed due to structural/aero interactions
- LPV is a natural framework.

## 2. Model Reduction

- High fidelity CFD/CSD models have many (millions) of states.

## 3. Model Uncertainty

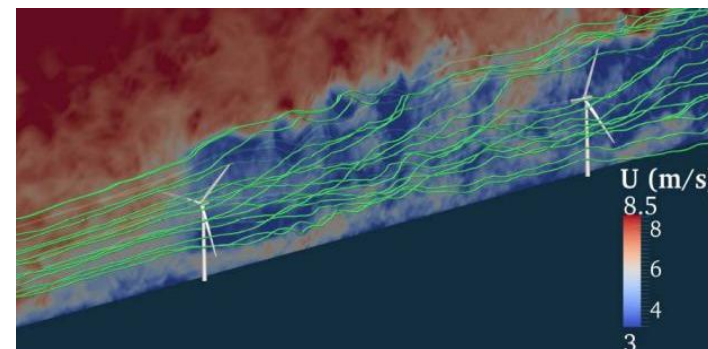
- Use of simplified low order models OR reduced high fidelity models



Eolos: <http://www.eolos.umn.edu/>

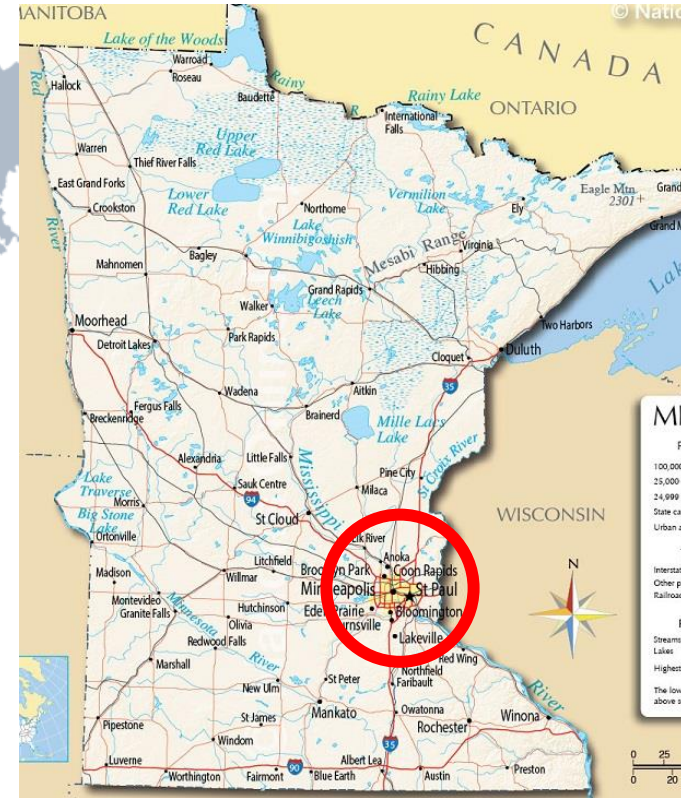
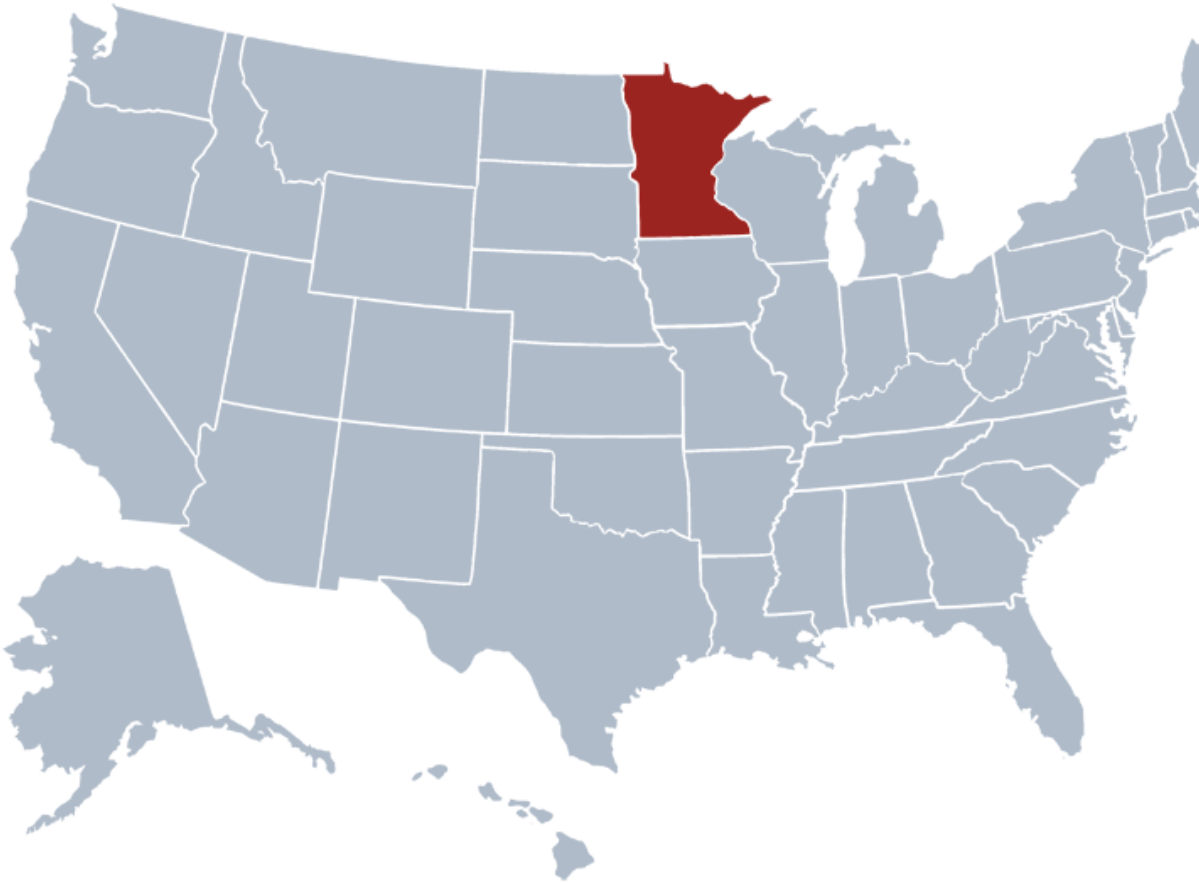


Saint Anthony Falls: <http://www.safl.umn.edu/>



Simulator for Wind Farm Applications, Churchfield & Lee  
<http://wind.nrel.gov/designcodes/simulators/SOWFA>

# Minneapolis and St. Paul, Minnesota



- Twin Cities Population ~3.5 Million
- Average daily low/high in January is  $-15.4^{\circ}\text{C}$  /  $-5.6^{\circ}\text{C}$
- Strong outdoor culture with many lakes and bike trails

# Department History



*Akerman Tailless Aircraft*

John D. Akerman was first Department Head 1929 - 1957

- Born in Latvia late 1890's
- Studied with Nikolai Joukowski
- Acquainted with Igor Sikorsky

Jean and Jeanette Piccard performed pioneering research in high altitude ballooning (1930's)



*1930's  
Cellophane Stratosphere Balloon  
Ascent in Memorial Stadium*