

System Identification within Flight Test and Flight Physics

02.03.2018

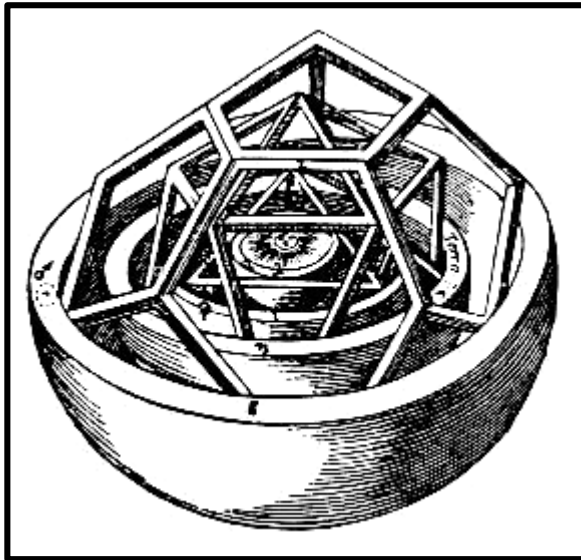
A project proposal



Alejandro Valverde

Aircraft System Identification

System Identification is the process of building mathematical models for physical systems based on imperfect observations or measurements



When applied to aircraft, it is called
Aircraft System Identification

Problems in Aircraft Dynamics



Simulation: Given u and S , find z

Control: Given z and S , find u

Identification: Given z and u , find S

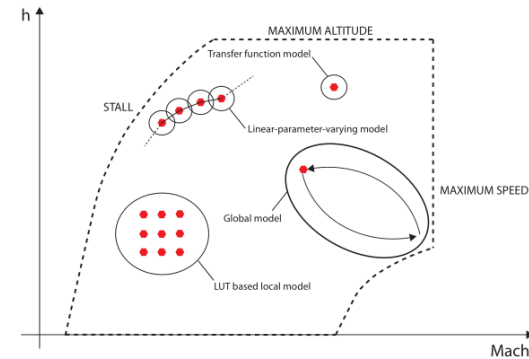
Applications of Aircraft System ID techniques

Aircraft System Identification offers a wide range of different applications, including:

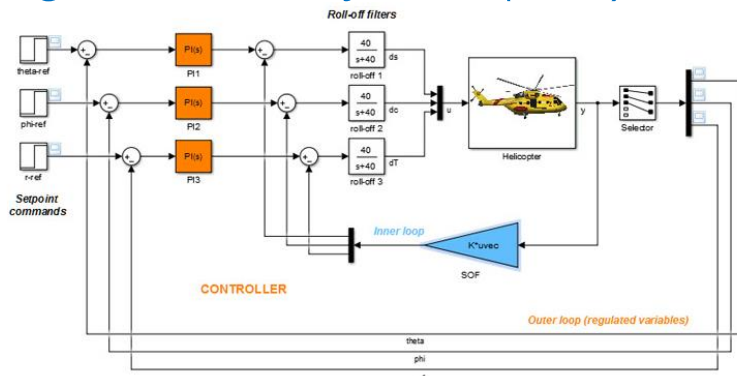
Flight Simulator development



Flight envelope expansion



Flight Control System (FCS) design

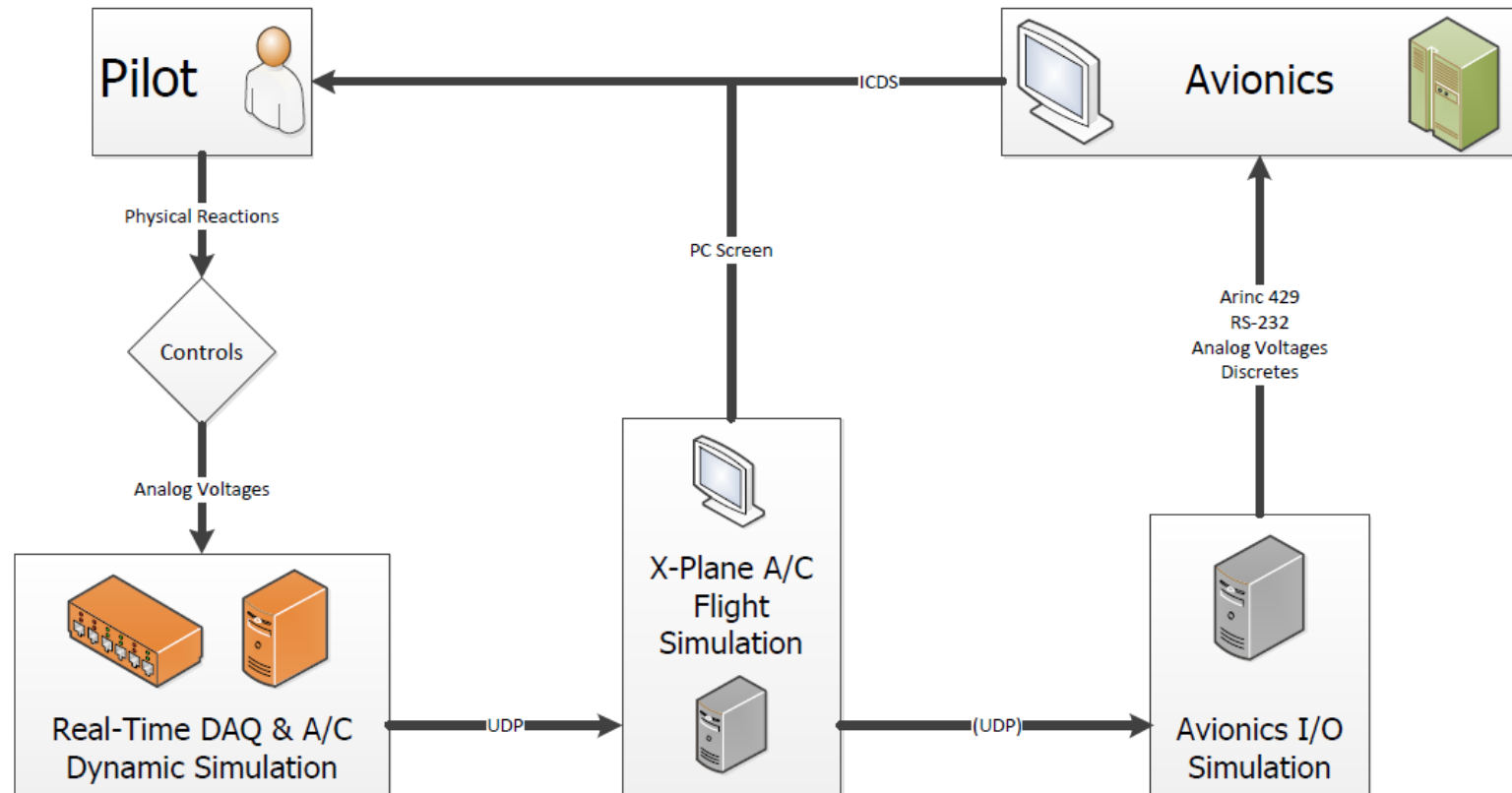


Handling Qualities assessment and enhancement



And many others...

Aircraft Simulation - Ongoing Kopter project



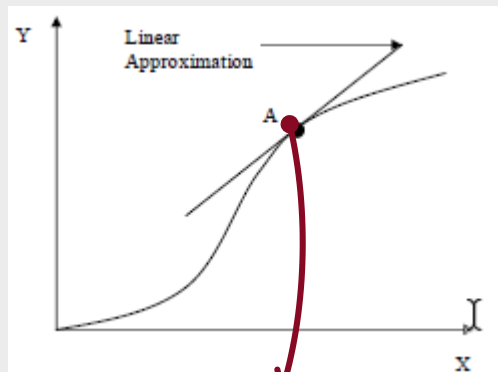
Extracted from Nicolas Callejo's MSc Thesis overview in collaboration with Kopter Avionics Dep.

Aircraft Simulation – Current Kopter approach

Dynamic Simulation

Trimmed condition of FlightLab, e.g.:

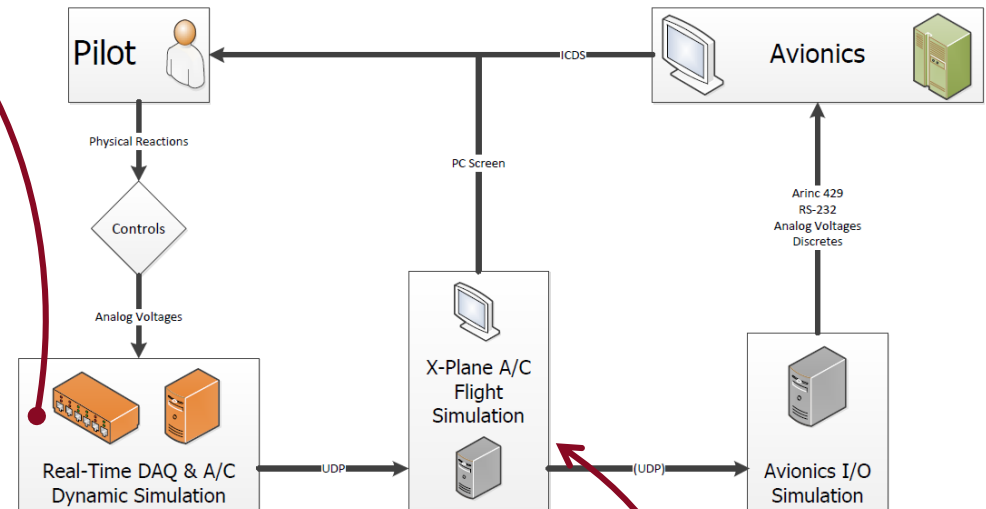
➔ fwd level flight @ 100kts



linearization of full H/C dynamics in LTI model:

$$\dot{x} = Ax + Bu$$

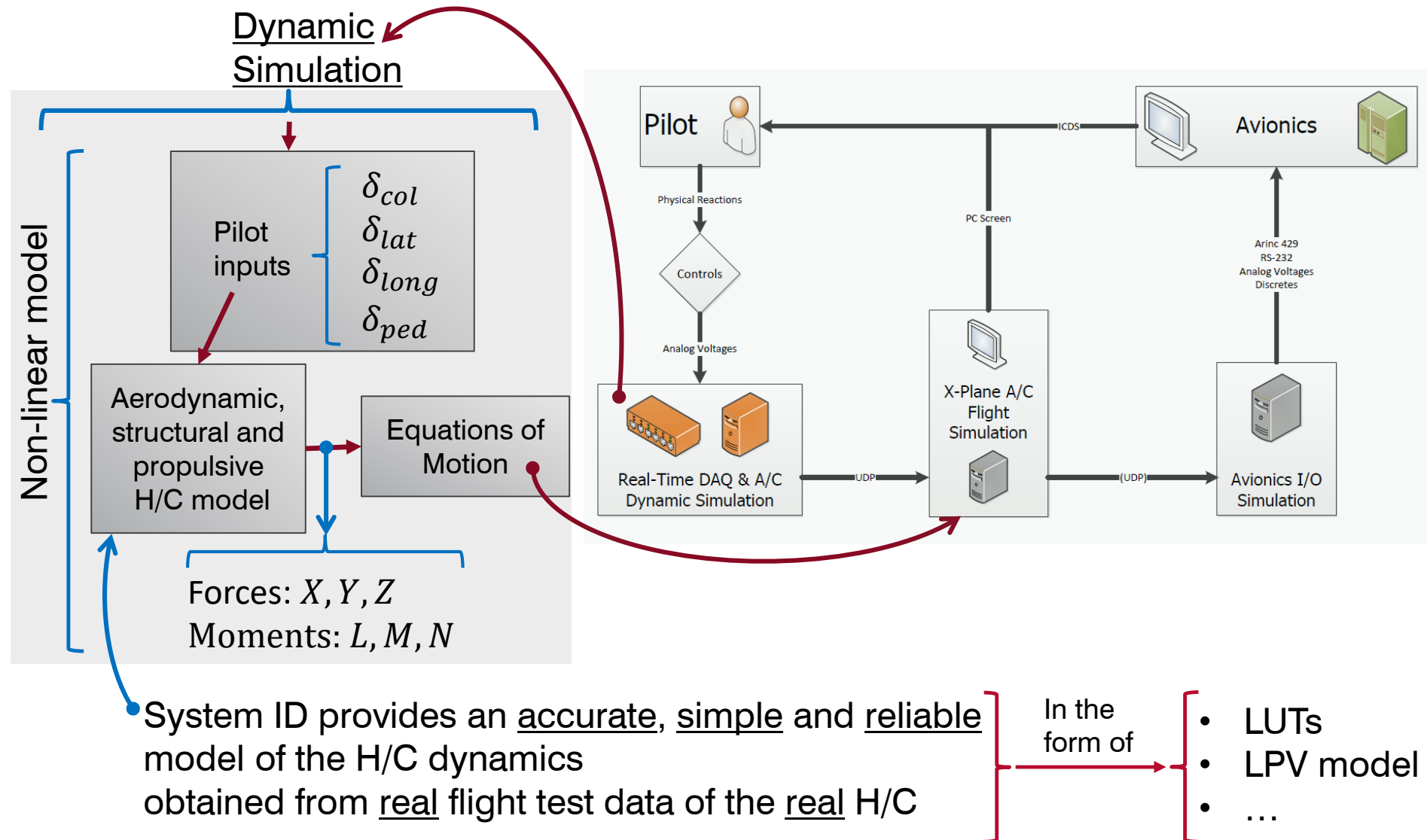
$$y = Cx + Du$$



Suitable for:

- ➔ FCS design
- ➔ H/C flight simulation
- ➔ Aircraft dynamic stability analysis

Aircraft Simulation – Aircraft SID approach



System ID R/C Model vs FlightLab

H/C model from System ID

Advantages:

- High accuracy provided that good quality data points are available
- Validation -> instantaneous
- Non-linear flight dynamics
- Transient flight conditions
- Real-time dynamic modelling
- Account for aeroelastic effects

Drawbacks:

- Depend on flight test data availability and dedicated flight manoeuvres



➔ **Best use: Dynamic model of the real H/C:**

- FCS design
- Handling qualities
- Simulator development



FlightLab H/C model

Advantages:

- Not dependent on flight test data
- Model complexity adjustable:
Full aeroelastic -> loads&dynamics
Rigid -> flight mechanics

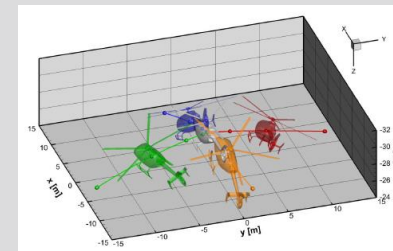
Drawbacks:

- Validation -> complicated
- Full aeroelastic model -> complicated
- Internal parameters hard to properly be tuned



➔ **Best use: Prediction of H/C:**

- H/C and compnt. loads
- Rotor dynamics
- H/C Performance



How does it work?

Forces and moments

$$\frac{d}{dt}(mV) = \Sigma F = F_{Aero} + F_{Thrust} + F_{Gravity} + F_{Controls} + F_{Propulsion} + F_{Disturbances}$$

$$\frac{d}{dt}(IW) = \Sigma M = M_{Aero} + M_{Thrust} + M_{Controls} + M_{Propulsion} + M_{Disturbances}$$

ΣF
 ΣM

Develop a model form from which parameters can be estimated using flight test data

for example

Pitching moment coefficient

$$C_m = \underbrace{C_{m_0}}_{\text{pitching moment bias}} + \underbrace{C_{m_\alpha}}_{\text{static stability}} \underbrace{\alpha}_{\text{H/C angle of attack}} + \underbrace{C_{m_q}}_{\text{dynamic stability (damping)}} \underbrace{q}_{\text{longitudinal angular velocity}} + \underbrace{C_{m_\delta}}_{\text{pitch control authority}} \underbrace{\delta}_{\text{pitch control input}} + \dots + \underbrace{v_m}_{\text{measurement noise}}$$

How does it work?

Pitching moment model:

$$C_m = C_{m_0} + C_{m_\alpha} \alpha + C_{m_q} q + C_{m_\delta} \delta + v_m$$

$$\theta = [C_{m_0}, C_{m_\alpha}, C_{m_q}, C_{m_\delta}]'$$

Vector of regressors (parameters to be identified)

Observations z

y : Model output

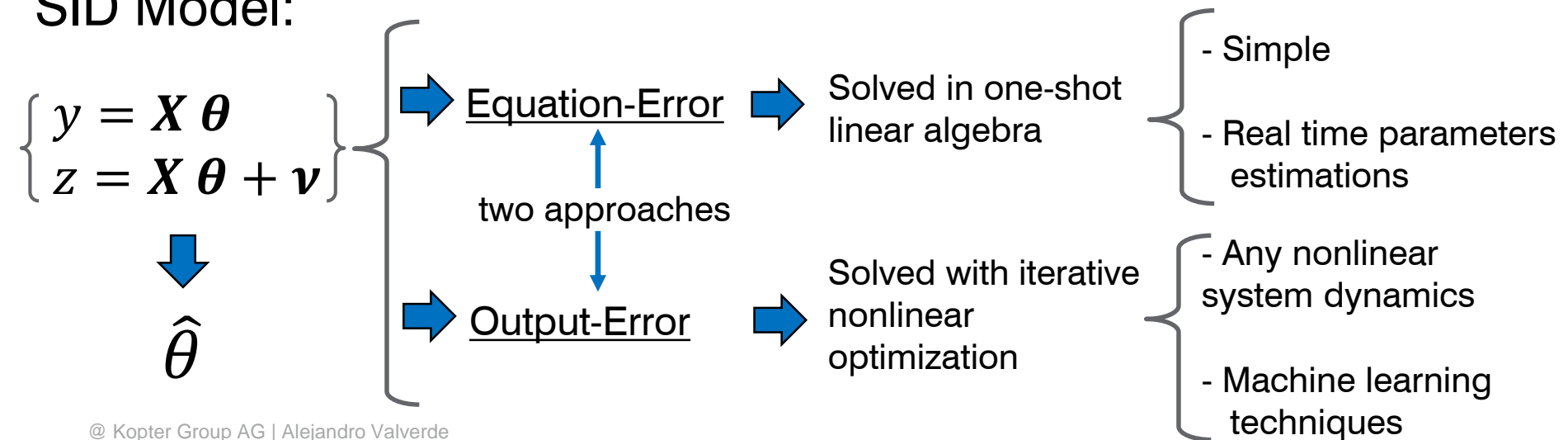
θ : Regressors

X : Values of variables measured

z : Observations

v : Noise of the measurement

SID Model:

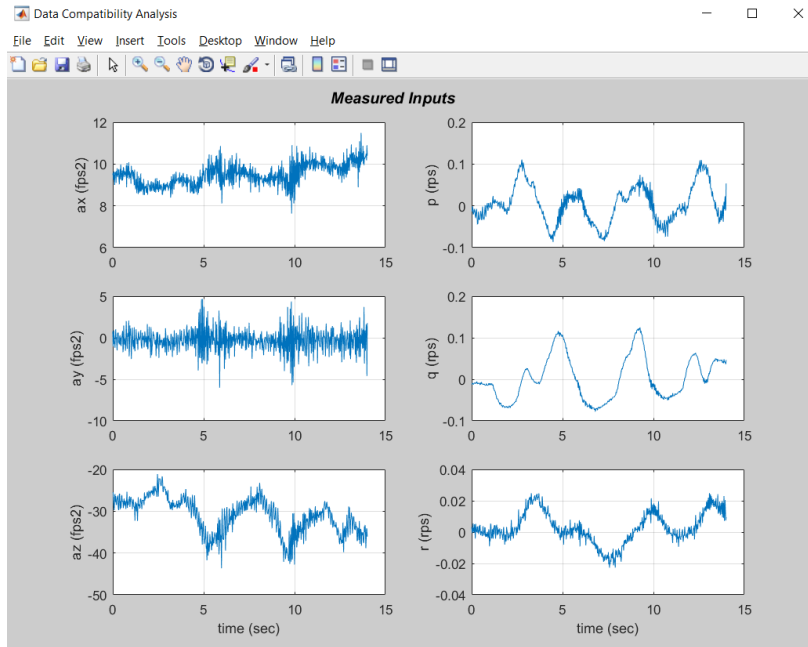


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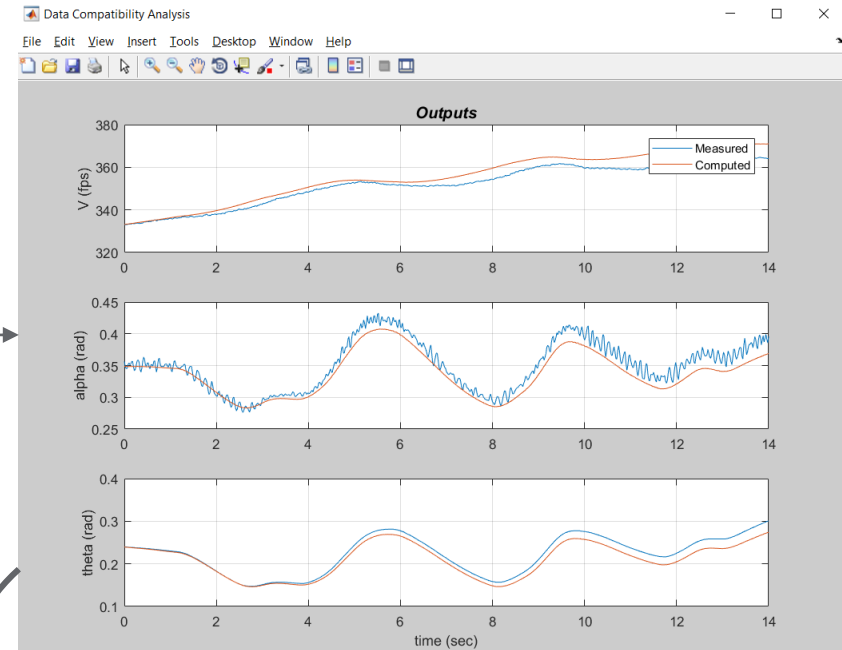
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Tools – SIDPAC software



SID
H/C
Model



Measured inputs:

- a_x, a_y, a_z
- p, q, r

Outputs to be validated
against real flight test data:

- V, α, θ

What would we need?

➡ Complete instrumented helicopter

- IMUs
- MR/TR rpm
- δ_{col} , δ_{lat} , δ_{long} , δ_{ped}
- ...



We already have this

➡ A laptop



➡ Some coffee...



Instrumentation overview

Table 9.1 Instrumentation characteristics for a general aviation airplane

Measured quantity	Transducer	Working range	Resolution	rms measurement error
a_x, g	Accelerometer	$[-1,1]$	0.001	0.0046
a_y, g	Accelerometer	$[-1,1]$	0.001	0.0050
a_z, g	Accelerometer	$[-3,6]$	0.001	0.0050
$p, \text{deg/s}$	Rate gyro	$[-102,102]$	0.12	0.20
$q, \text{deg/s}$	Rate gyro	$[-29,29]$	0.032	0.19
$r, \text{deg/s}$	Rate gyro	$[-29,29]$	0.034	0.080
ϕ, deg	Vertical gyro	$[-90,90]$	0.10	0.077
θ, deg	Vertical gyro	$[-87,87]$	0.098	0.092
α, deg	Flow vane	$[-12,27]$	0.029	0.027
β, deg	Flow vane	$[-29,32]$	0.018	0.019
δ_{a_r}, deg	Potentiometer	$[-23,10]$	0.002	0.019
δ_{a_l}, deg	Potentiometer	$[-10,25]$	0.002	0.0061
δ_s, deg	Potentiometer	$[-16,3]$	0.010	0.0037
δ_r, deg	Potentiometer	$[-31,28]$	0.011	0.0091
$V, \text{m/s}$	Pressure transducer	$[0,75]$	0.037	0.89
h, m	Altimeter	$[-150,2900]$	—	—
$T, ^\circ\text{C}$	Thermometer	$[-18,38]$	—	—

Table 9.2 Dynamic characteristics of transducers

Measured quantity	Transducer	Natural frequency, Hz	Damping ratio	Time delay, s
a_x, g	Accelerometer	402	1.58	0.0012
a_y, g	Accelerometer	216	1.10	0.0016
a_z, g	Accelerometer	921	1.58	0.0005
$p, \text{deg/s}$	Rate gyro	27	0.64	0.0075
$q, \text{deg/s}$	Rate gyro	27	0.64	0.0075
$r, \text{deg/s}$	Rate gyro	27	0.64	0.0075
α, deg	Flow vane	23 ^a	0.085 ^a	0.0012
β, deg	Flow vane	23 ^a	0.085 ^a	0.0012

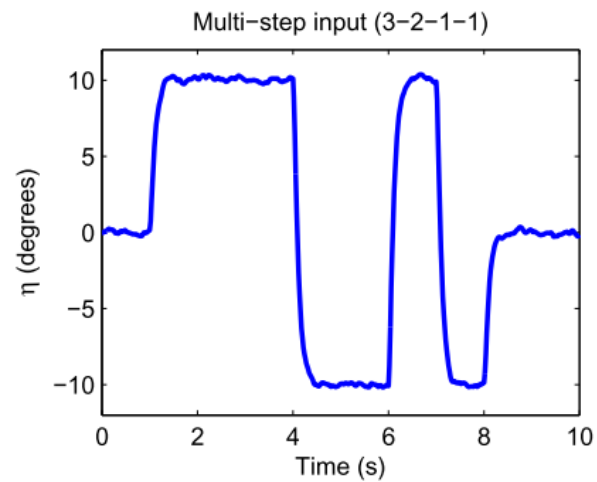
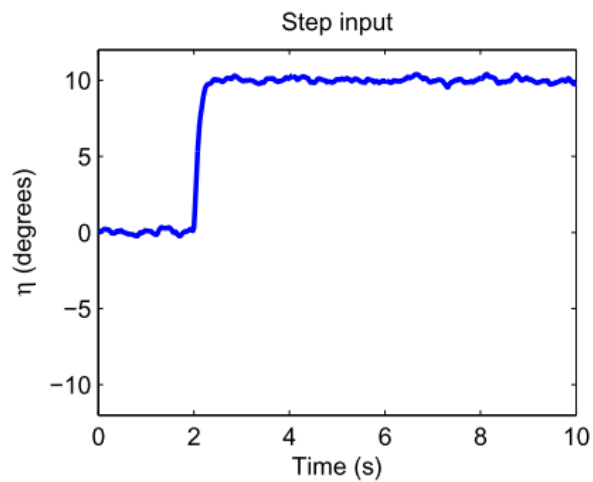
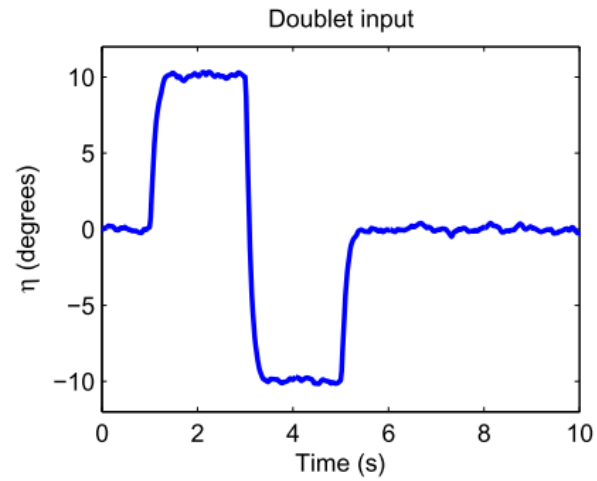
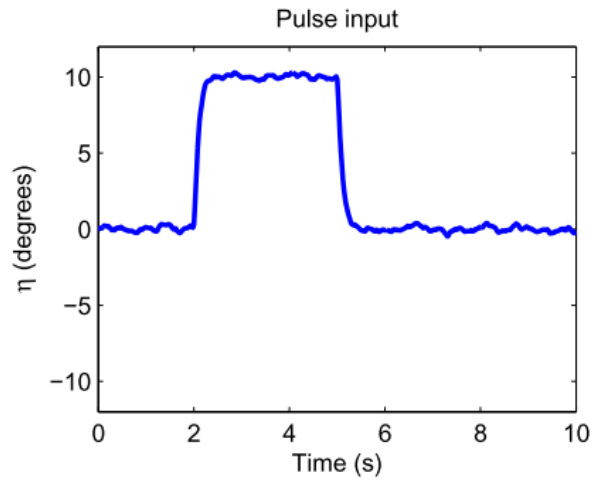
^aAt $V = 164 \text{ ft/s}$

Inputs


Constraints on inputs:

- Rate of data acquisition
- Limited resolution on sensors or data acquisition system
- Limited time for each maneuver
- Limitations on how the aircraft can be excited

Inputs



Why to introduce System ID into P3 test campaign?

- ➡ Sooner or later, it would need to be implemented  Every big helicopter manufacturer has a dedicated SID department
 - ➡ SID framework could be developed now using P2 flight test data
 - ➡ Assist P3 flight envelope expansion in Sicily, necessary for TC. Issues like the cyclic migration could be addressed and main driving parameters identified.
 - ➡ Provide a proper dynamic model of the H/C for the simulator
 - ➡ Surely it will be needed for PS4 and further developments after TC, such as FCS design
 - ➡ Cost effective
- We have to do it now,
we are ready now

System ID takes time and money – but not nearly as much as not doing it

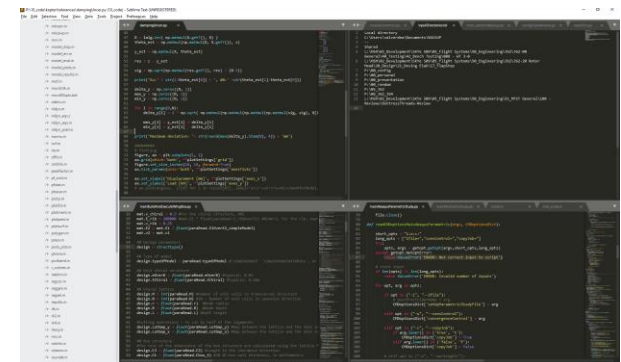
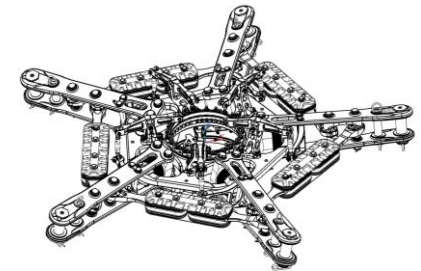
Eugene A. Morelli
NASA Langley Research Center Hampton

Why I can be the one doing this?

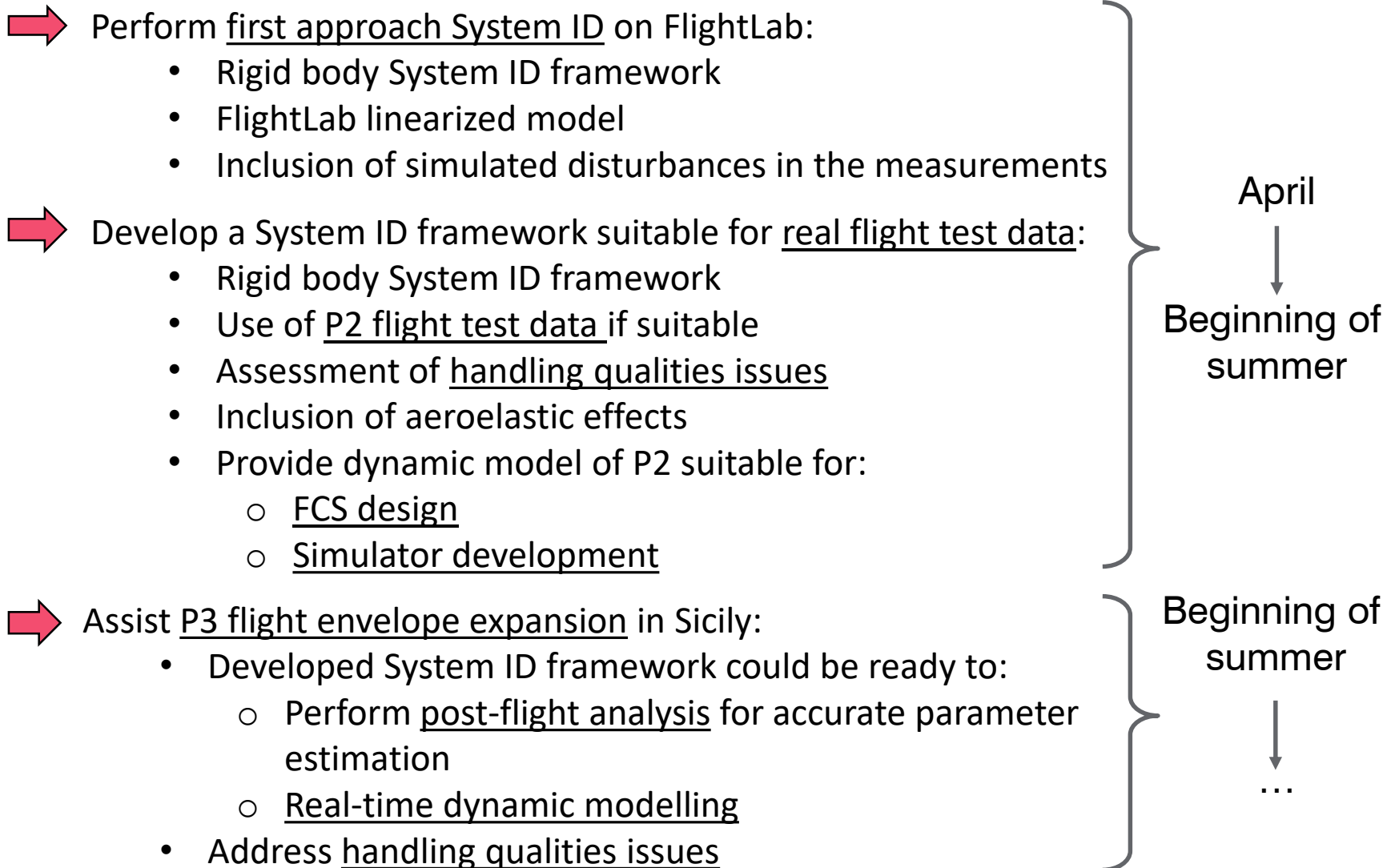
Young

Passionate

- ➔ I know what to do and how to do it
- ➔ Kopter experience, within rotorheads
- ➔ Data analysis experience
 - Python
 - Matlab
 - Data analysis and machine learning techniques
 - ...
- ➔ Availability from April onwards



Next steps – Near future



First steps, using FlightLab

```

P:\10_code\kopter\SH400_4_id\main.py - Sublime Text (UNREGISTERED)
File Edit Selection Find View Goto Tools Project Preferences Help

# main.py
from configclassmodule import *

# main script
cmd = m.grtcmd()

# main script settings
plotSettings = plotSettings()

# main script parameters
n_states = 73
n_inputs = 4
n_outputs = 6

# main script variables
x = np.zeros(n_states)
y = np.zeros(n_outputs)

# main script functions
def input_list():
    input_list = ['lon', 'lat', 'col', 'ped']
    output_list = ['lon', 'lat', 'col', 'ped']
    output_list = ['lon', 'lat', 'col', 'ped']

# main script initialization
A = np.loadtxt(cod_path + 'from_ligPhysics\\A_LTI_Model\\A_from_LTI.txt')
B = np.loadtxt(cod_path + 'from_ligPhysics\\A_LTI_Model\\B_from_LTI.txt')
C = np.loadtxt(cod_path + 'from_ligPhysics\\A_LTI_Model\\C_from_LTI.txt')
D = np.loadtxt(cod_path + 'from_ligPhysics\\A_LTI_Model\\D_from_LTI.txt')

# main script control
ctrl = control.ss(A, B, C, D)

# main script simulation
t_sim = np.linspace(0, 10, 1000)

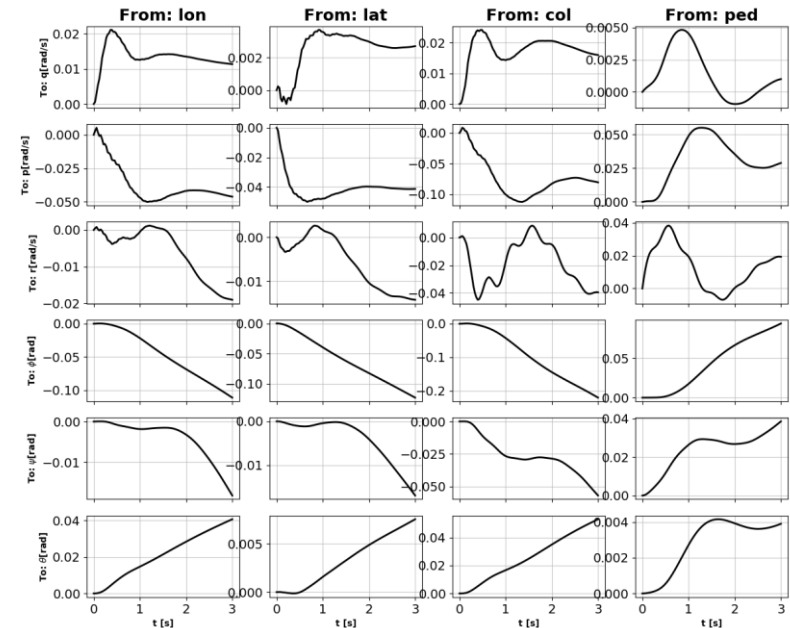
# main script outputs
youts = []
for i_input in range(n_inputs):
    t, yout = control.step_response(ctrl, f = 't_sim', input = i_input, output = None)
    youts.append(yout)

# main script plotting
figure, axes = plt.subplots(n_inputs, sharex = 'col')
figure.set_xlabel('t [s]', loc = 'bottom')

# main script plotting
for i_out in range(n_outputs):
    for i_in in range(n_inputs):
        axes[i_out, i_in].plot(t, youts[i_in][i_out], (color = 'r', c = plotSettings['color']))
        axes[i_out, i_in].tick_params(axis = 'both', **plotSettings['tick_params'])
        axes[i_out, i_in].set_ylabel('youts[i_out, i_in]', **plotSettings['axes_label'])
        axes[i_out, i_in].set_xlabel('t [s]', **plotSettings['axes_label'])
        axes[i_out, i_in].set_title('From: ' + input_list[i_in], **plotSettings['title'])
        axes[i_out, i_in].grid(which = 'both', **plotSettings['grid'])
    i_in += 1

# main script legend
ax = figure.legend(**plotSettings['legend'])

```



- fwd level flight
- 147 kts
- 2800 kg
- $x_{cg} = 3.37\text{m}$

LTI model

$$y = X \theta$$

$$z = X \theta + v$$

- 73 states
- 6 outputs
- 4 inputs



Thank you

Kopter Group AG
Binzstrasse 31
8620 Wetzikon
Switzerland

Registered Office
Flugplatzareal 10
8753 Mollis
Switzerland

E contact@koptergroup.com
T +41 44 552 33 33
koptergroup.com

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- [2] R. K. Remple and M. B. Tischler, *Aircraft and Rotorcraft System Identification*. 2006.
- [3] G. Morelli and S. Derry, “System Identification Methods for Aerodynamic Modeling and Validation using Flight Data,” 2011.
- [4] G. Morelli, “Aircraft System Identification,” 2011.

System ID methodology diagram

Manoeuvre

Optimized Input

Input

Helicopter

Actual Response
(training dataset)

Measurements

Data Collection
& Compatibility

Methods

Parameter Estimation

Estimation
Algorithm /
Optimization

Identification
Criteria

Response
Error

Parameter
Adjustments

Models

Mathematical
Model /
Simulation

Model Response

A Priori Values,
lower/upper
bounds

Model
structure

Identification Phase

Validation Phase

Complementary Flight Data
(validation dataset)

Model
Validation