



High fidelity modelling for High Altitude Long Endurance Solar Powered Aircraft

Jongseok Lee

Master thesis (external at DLR Institute of Robotics and Mechatronics)
Tin Muskardin, Dr. Konstantin Kondak, Philipp Oettershagen, Thomas Stastny



Autonomous Systems Lab



German
Aerospace Center

High Altitude Long Endurance (HALE) platforms

- Aerial platforms capable of stratospheric flight for a long period.
- Communication networks to recording of weather and environment.



- Topic: high fidelity modelling procedures for fixed wing platforms.



Autonomous Systems Lab

German
Aerospace Center

Motivation

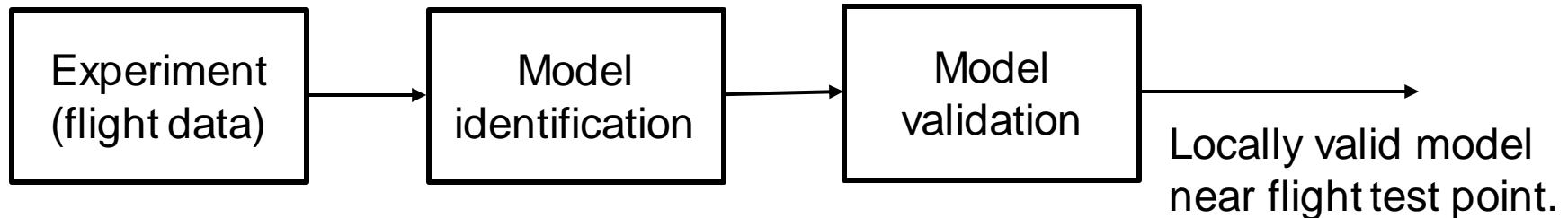
- Why high fidelity models?
 - Reduce or avoid in-flight tuning of controller gains.
 - Model based control for landing on mobile platform.
 - Simulation of stratospheric mission.
- Platforms – Elektra 1 and Penguin BE UAV.



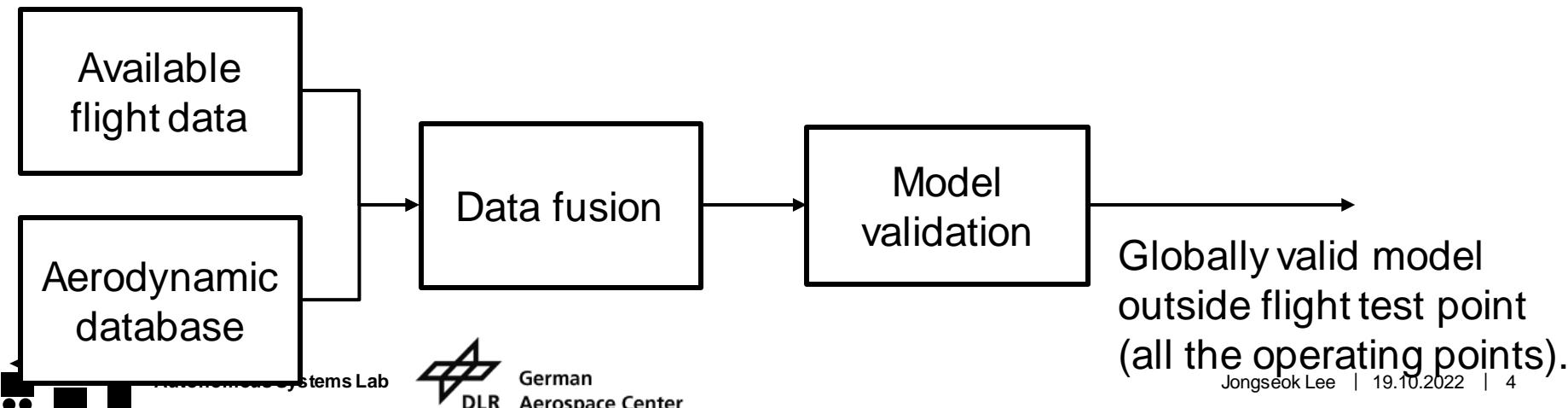
- Approach: local and global system identification.

Motivation

- Local system identification.



- Global system identification.



Overview

- Motivation.
- Aircraft system identification problem.
- Local system identification - two step method.
- Global system identification - incremental model update.
- Conclusion.

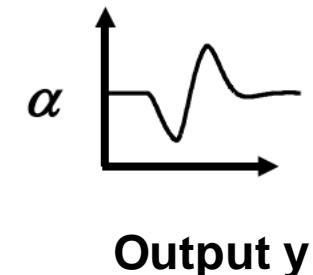
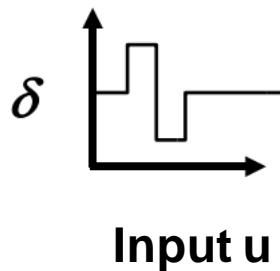


Autonomous Systems Lab



German
Aerospace Center

Aircraft system identification problem



System S

Given input u and output y find system S

$$\begin{aligned} m\dot{V} + \boldsymbol{\omega} \times mV &= \boxed{F_{aero}} + F_{thrust} + F_{gravity} \\ I\dot{\boldsymbol{\omega}} + \boldsymbol{\omega} \times I\boldsymbol{\omega} &= \boxed{M_{aero}} + M_{thrust} \end{aligned}$$



Aircraft system identification problem

Applying multidimensional taylor series expansion:

$$\widehat{\mathbf{F}}_{x,aero} = Fx_0 + Fx_u \mathbf{u} + Fx_w \mathbf{w} + Fx_q \mathbf{q} + Fx_{de} \mathbf{de}$$

$$\widehat{\mathbf{F}}_{z,aero} = Fz_0 + Fz_u \mathbf{u} + Fz_w \mathbf{w} + Fz_q \mathbf{q} + Fz_{de} \mathbf{de}$$

$$\widehat{\mathbf{M}}_{y,aero} = My_0 + My_u \mathbf{u} + My_w \mathbf{w} + My_q \mathbf{q} + My_{de} \mathbf{de}$$

- 15 parameters for longitudinal dynamics (linear model).
- Physical quantities related to stability and control.
- Linear Vs nonlinear aerodynamic model.
- Local – one value for parameters; Global - sets of values.



Overview

- Motivation.
- Aircraft system identification problem.
- Local system identification - two step method.
- Global system identification - incremental model update.
- Conclusion.



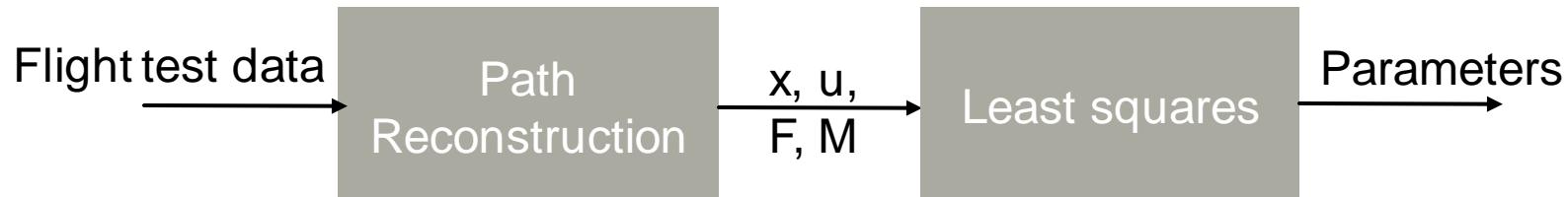
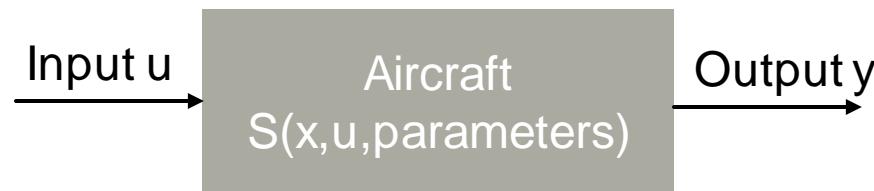
Autonomous Systems Lab



German
Aerospace Center

Local system identification - two step method

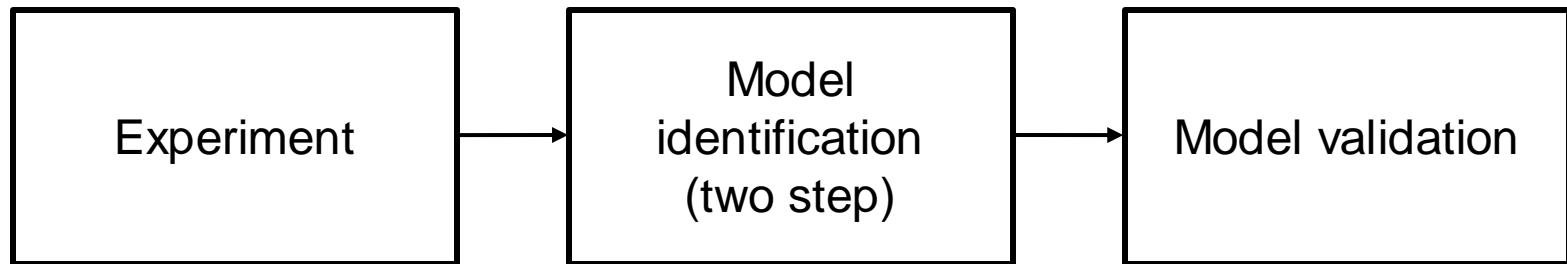
Approach: error = $y - S(x, u, \text{parameters})$



- Linear projection of features.

Local system identification - two step method

Procedures:

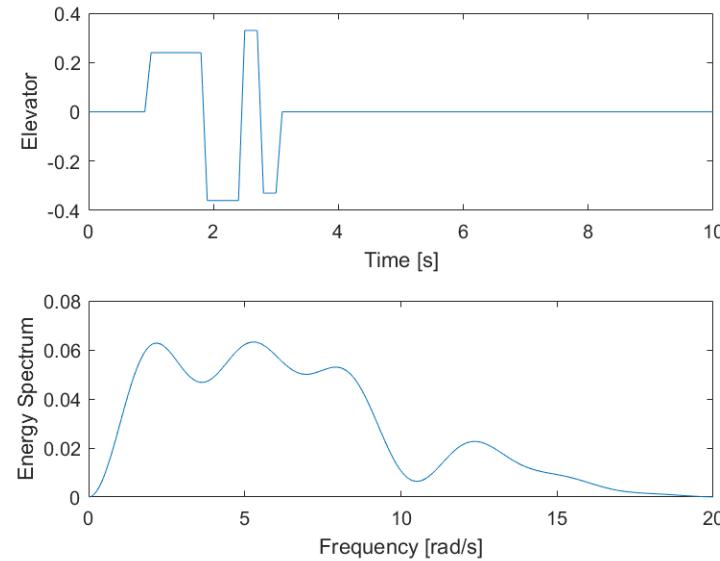
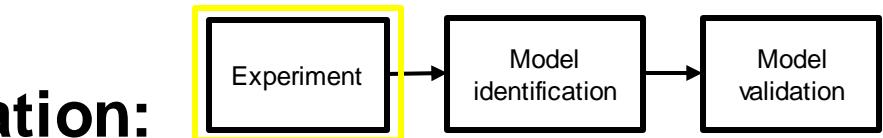
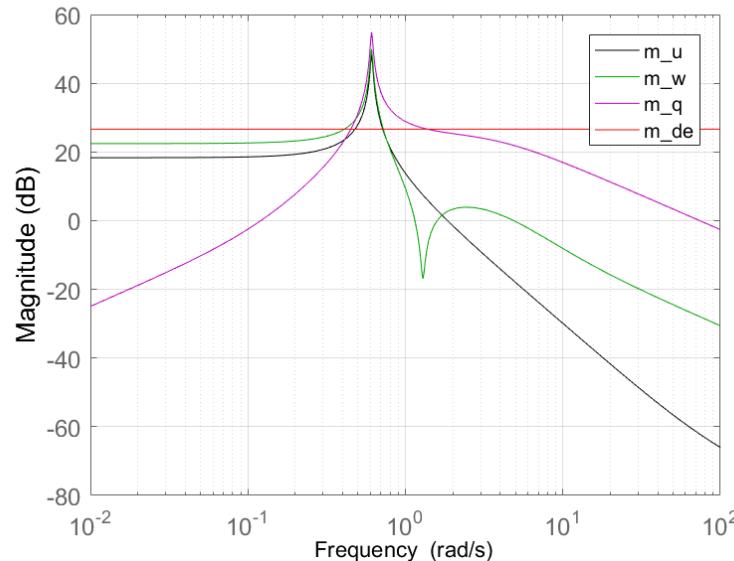


- Experiment: parameter identifiability.
- Reconstruction of path: IEKF estimation and smoothing.
- Parameter identification: linear regression.
- Model validation: statistics and controller synthesis.

Local system identification - two step method

Input design and experimentation:

- Parameter identifiability – contribution is visible.



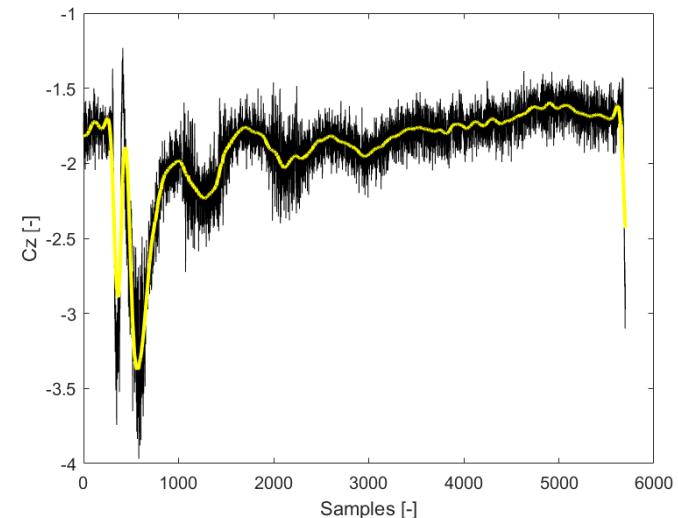
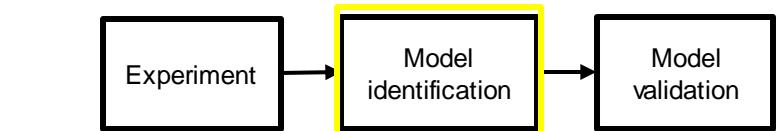
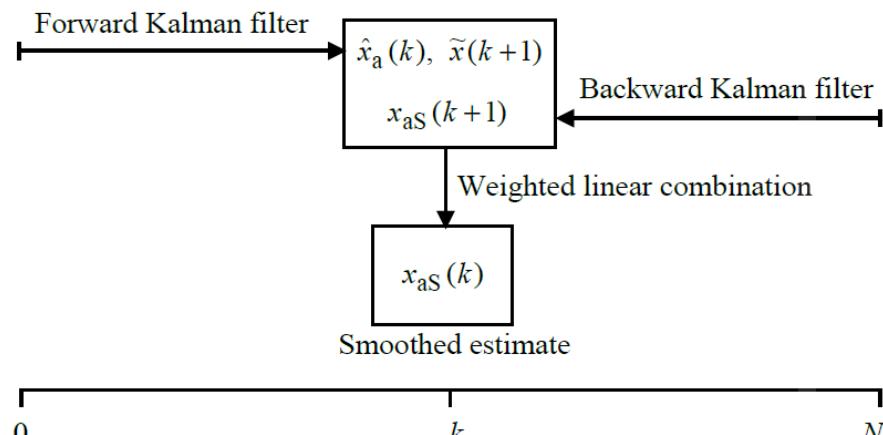
- 2 flights with Elektra 1 and 3 flights with Penguin BE.



Local system identification - two step method

Path reconstruction:

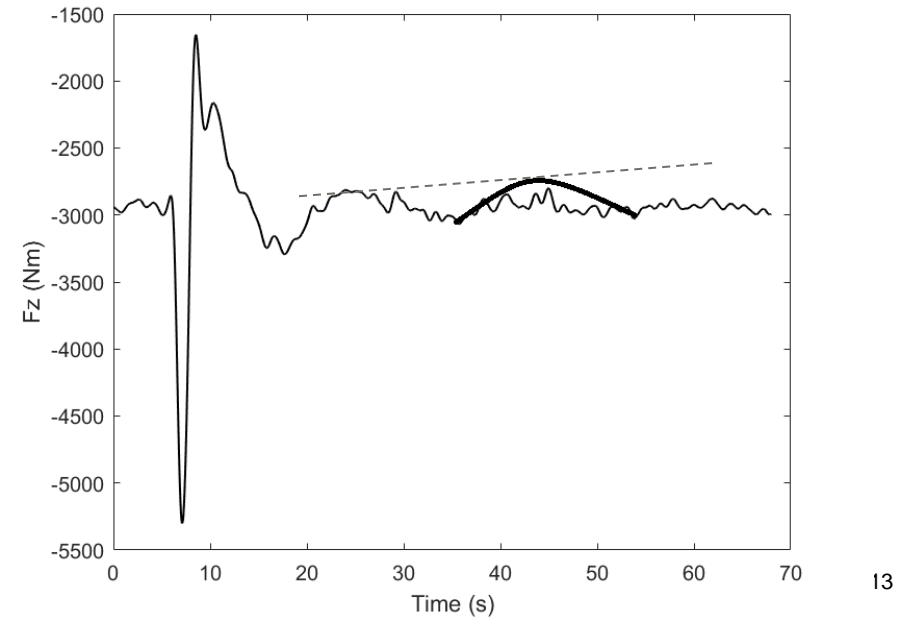
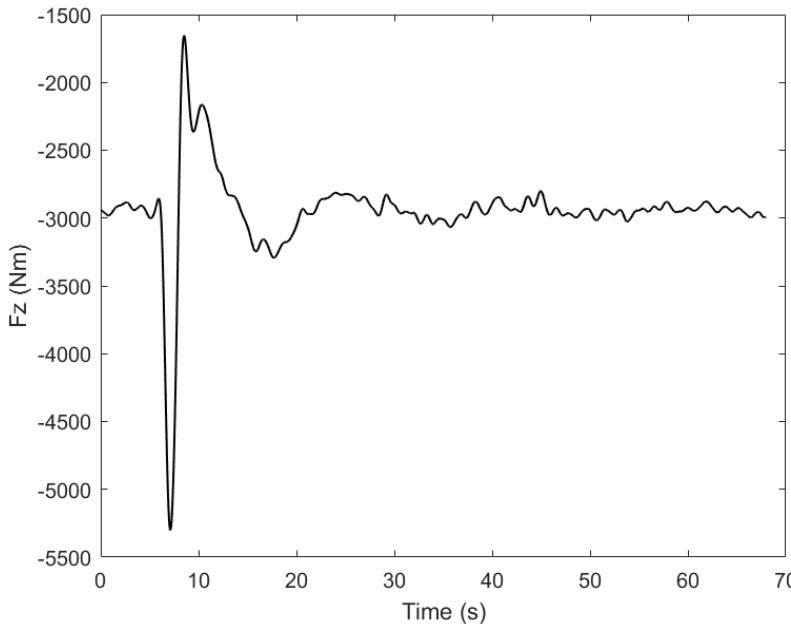
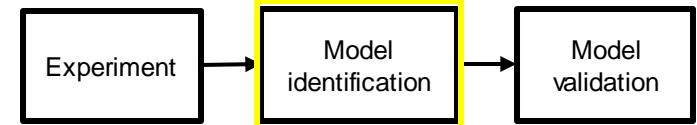
- Smoothed estimate of states, forces and moments.
- Estimation of instrumentation errors.



Local system identification - two step method

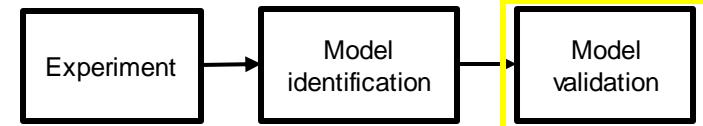
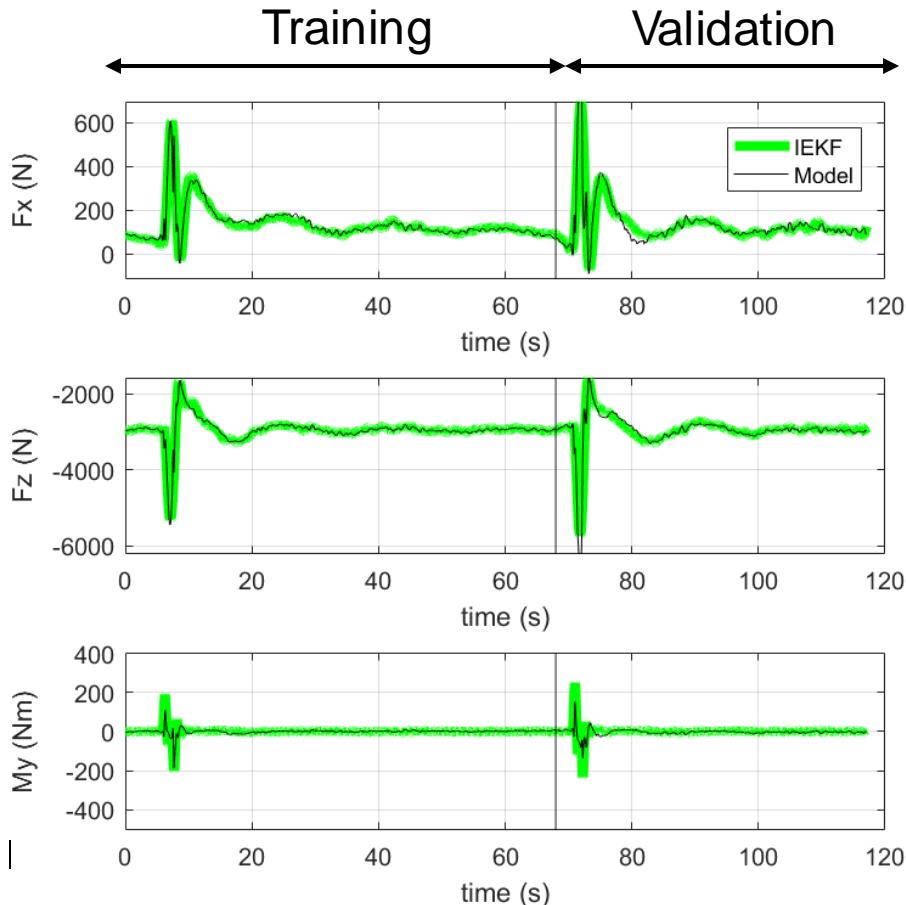
Parameter identification:

- Optimization – linear regression.
- Global minimum may not give correct parameters.
 - OLS Vs WLS Vs NLS Vs CLS.

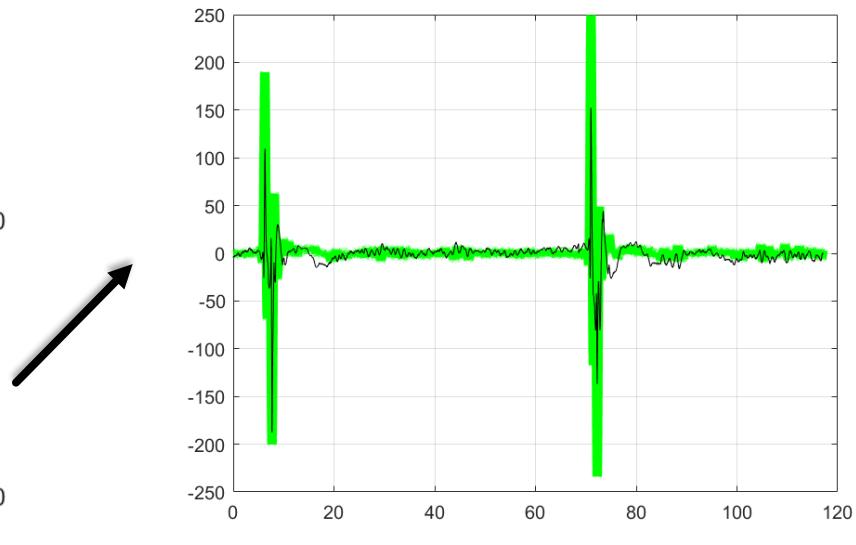


Local system identification - two step method

Model validation:

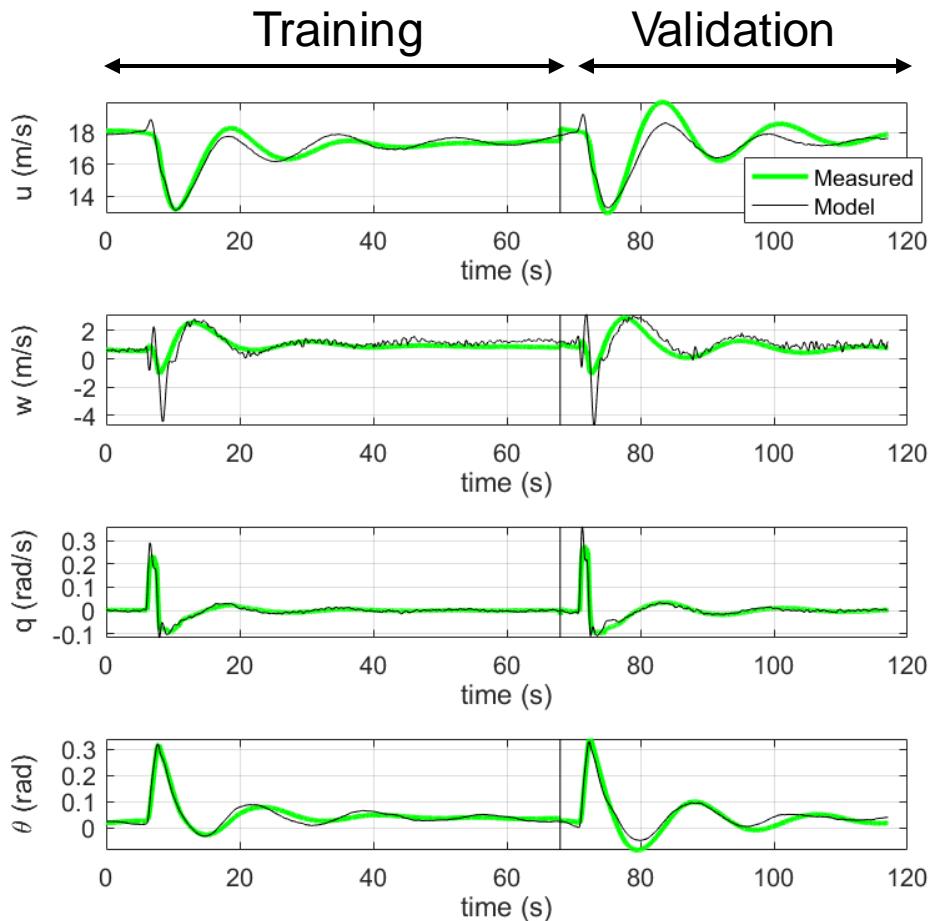


| Coefficient | Rsquared [-] | NRMSE [-] |
|-------------|--------------|-----------|
| F_x | 0.847 | 0.042 |
| F_z | 0.883 | 0.031 |
| M_y | 0.418 | 0.036 |



Local system identification - two step method

Model validation:



| State | GOF [-] | TIC [-] |
|----------------|---------|---------|
| u (m/s) | 0.8569 | 0.0133 |
| w (m/s) | 0.5576 | 0.2301 |
| q (rad/s) | 0.8507 | 0.1985 |
| θ (rad) | 0.9061 | 0.1289 |

- TIC 0.25 - 0.3 sufficient [Jategaonkar 2006].
- Fulfils FAA standards of high fidelity model.

Overview

- Motivation.
- Aircraft system identification problem.
- Local system identification - two step method.
- Global system identification - incremental model update.
- Conclusion.



Autonomous Systems Lab



German
Aerospace Center

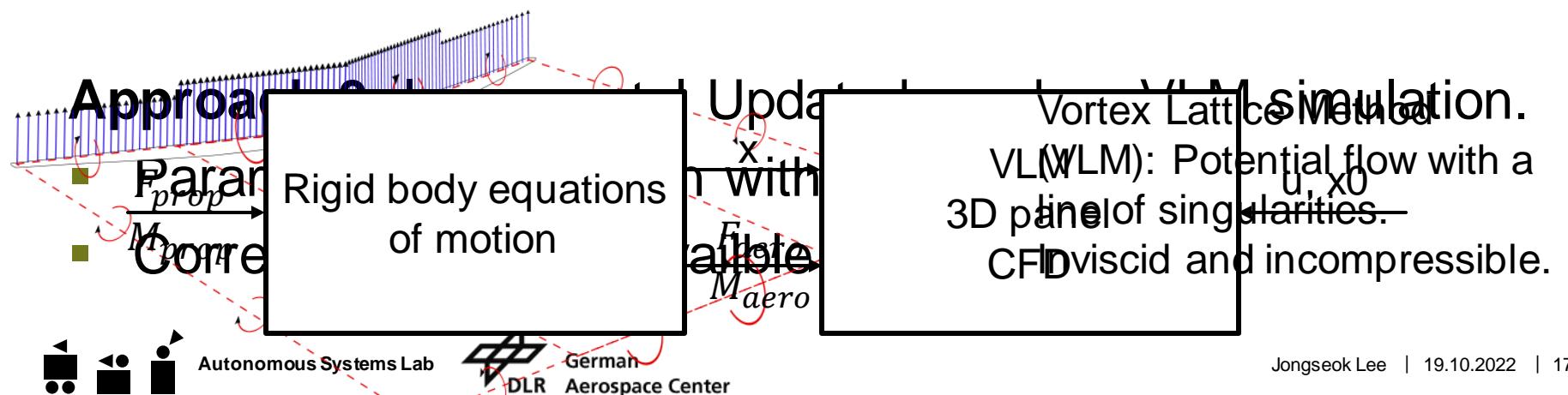
Global system identification – Incremental update

Approach 1: Current practice in the industry.

- Collection of data at all points of flight envelope.

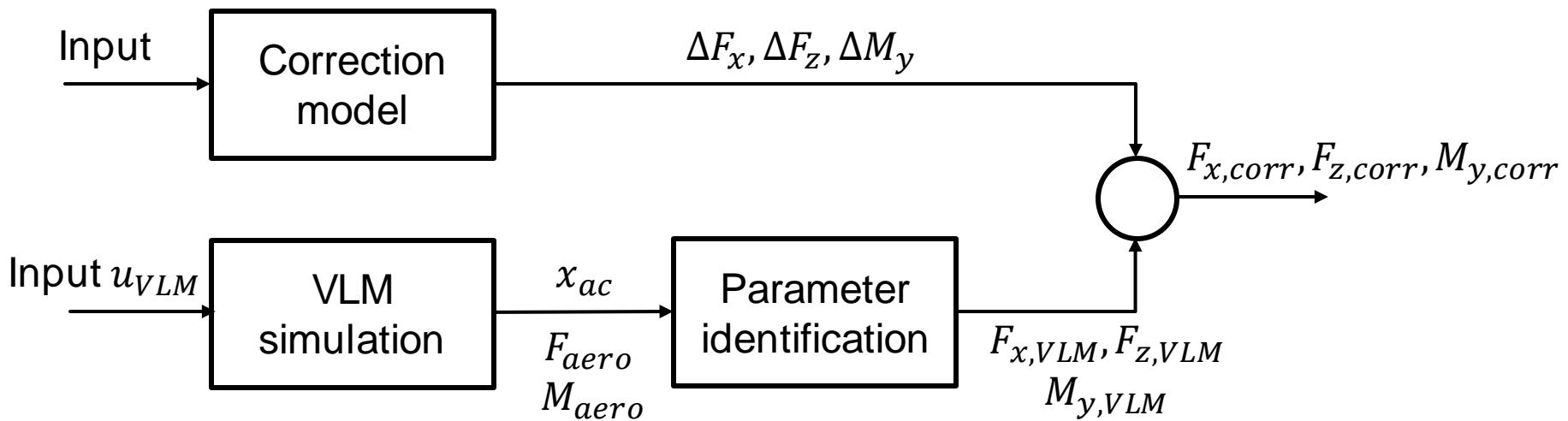
Approach 2: Incremental Update

- Data fusion of aerodynamic database with flight test data.
 - Aerodynamic database using windtunnel & CFD.



Global system identification – Incremental update

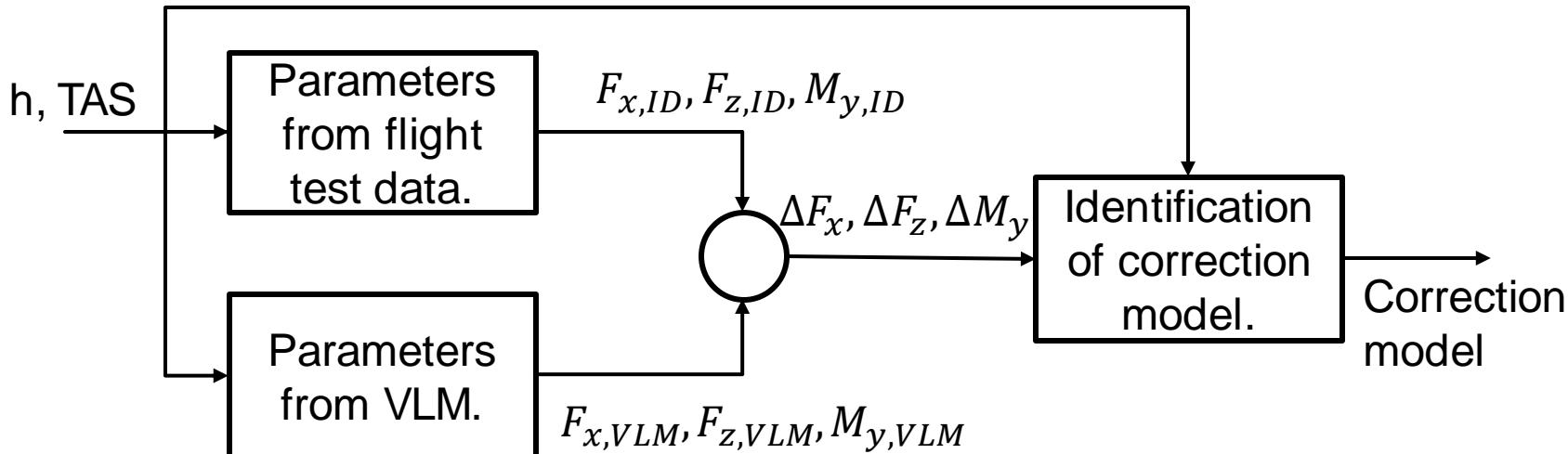
Method: Scheme for global system identification.



- Aerodynamic model outside the region of the flight data.
- Improvement in accuracy.

Global system identification – Incremental update

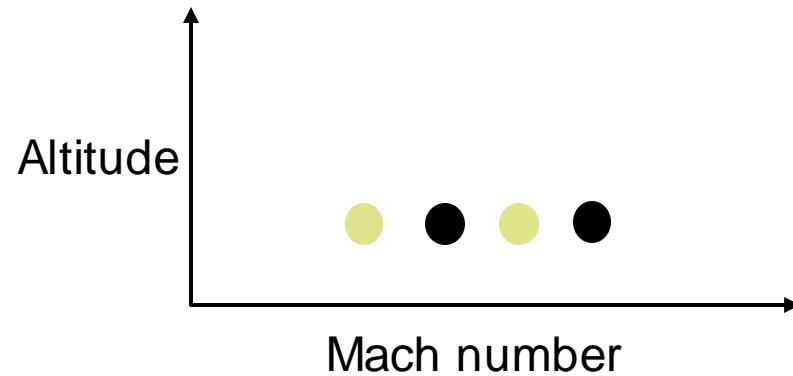
Method: Correction model identification.



- Correction model identification using available flight data.
- Separation of training and validation set via different trim.

Global system identification – Incremental update

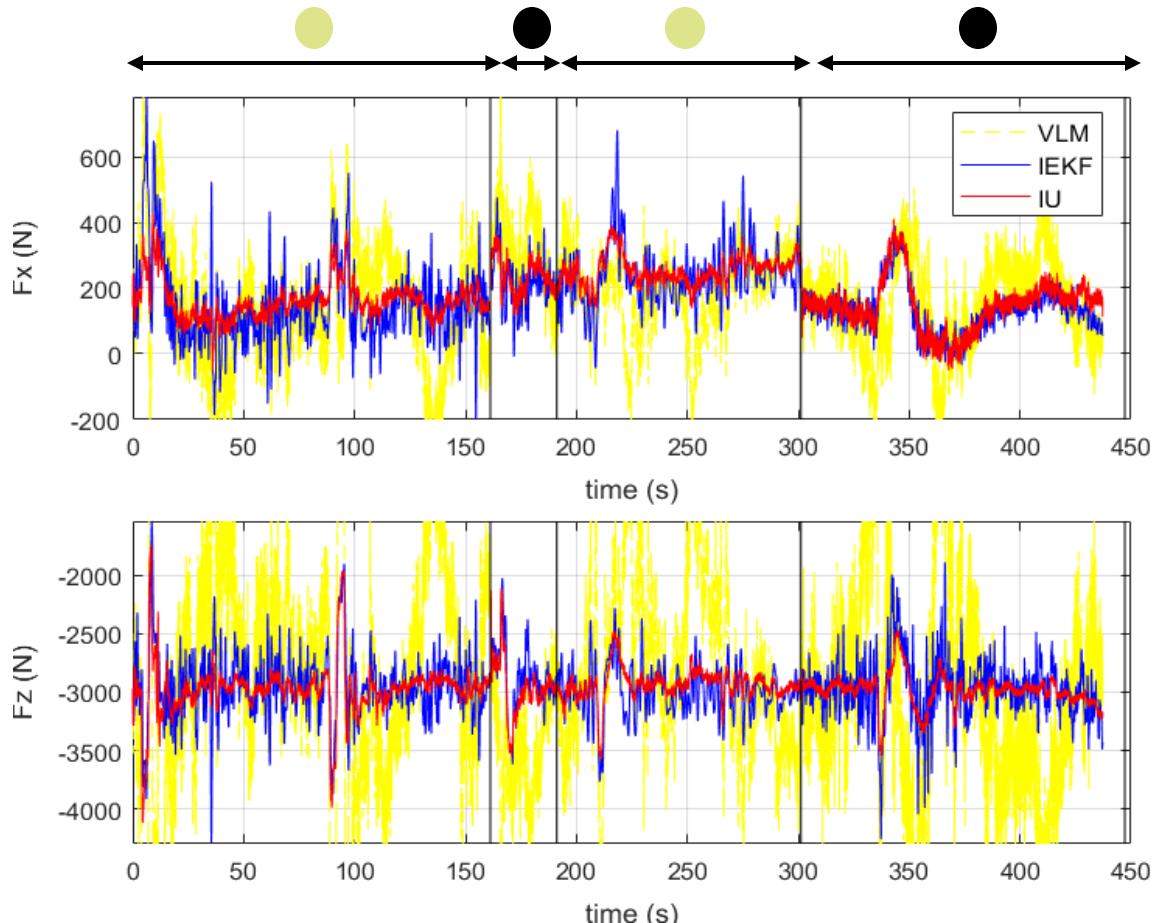
Scope: Preliminary study at low altitude for Elektra 1.



- Identification of correction model at two points ●.
- Validation of correction model at two points ○.

Global system identification – Incremental update

Results: Preliminary study at low altitude for Elektra 1.



| Fx | VLM | CVLM |
|-----------|------------|-------------|
| RMSE | 146.8 | 77.86 |
| NRMSE | 0.148 | 0.0789 |

| Fz | VLM | CVLM |
|-----------|------------|-------------|
| RMSE | 932.9 | 212.8 |
| NRMSE | 0.338 | 0.0777 |

Conclusion

- Local system identification.
 - Two step method implemented and validated.
 - System identification tool chain for 2 fixed wing platforms.
 - High fidelity model according to FAA standards.
- Global system identification.
 - Preliminary study on incremental model update scheme.
 - Within low altitude low velocity region the method proved to work with reduction of NRMSE by 0.5 and 0.2 for Fx and Fz respectively.



Future work

- Local system identification.
 - System identification for flexible aircraft (Elektra 2).
- Global system identification.
 - Wider ranges of velocities.
 - Wider ranges of altitude (Low Reynolds High Mach?).
- Fidelity definition for controller synthesis.
 - Step response of the aircraft.
 - Derivation of quantitative requirements?



Questions?



Autonomous Systems Lab



German
Aerospace Center

Back up slides

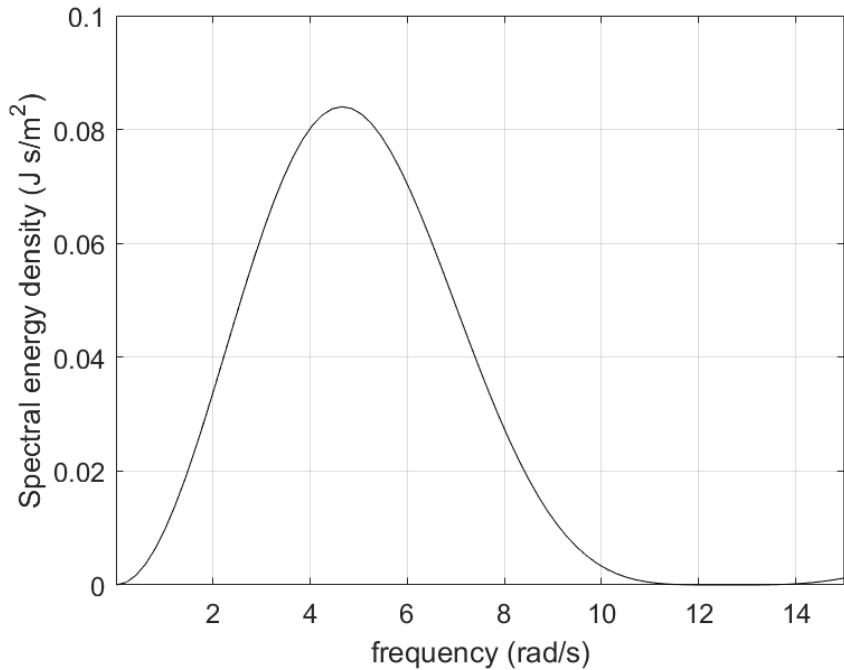
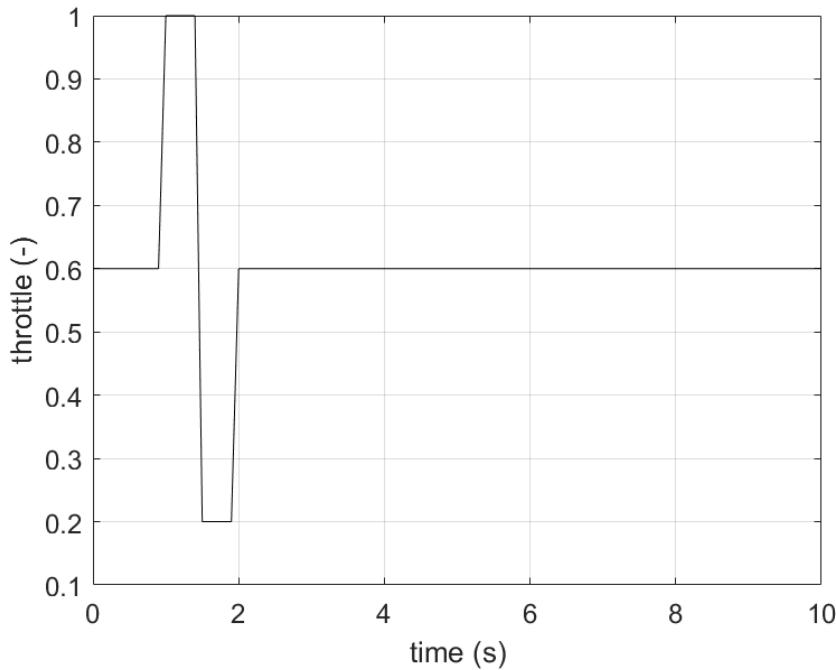


Autonomous Systems Lab

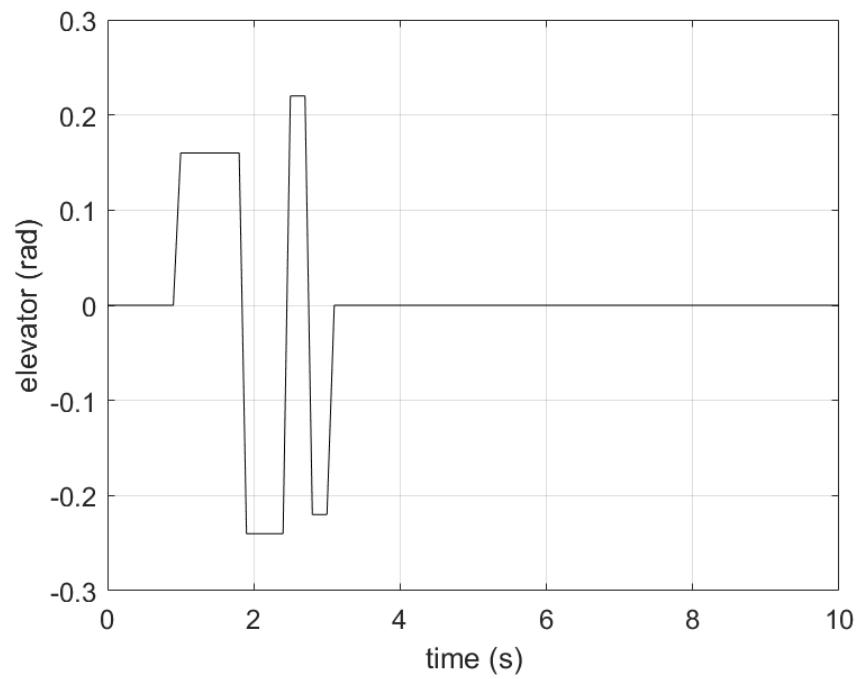
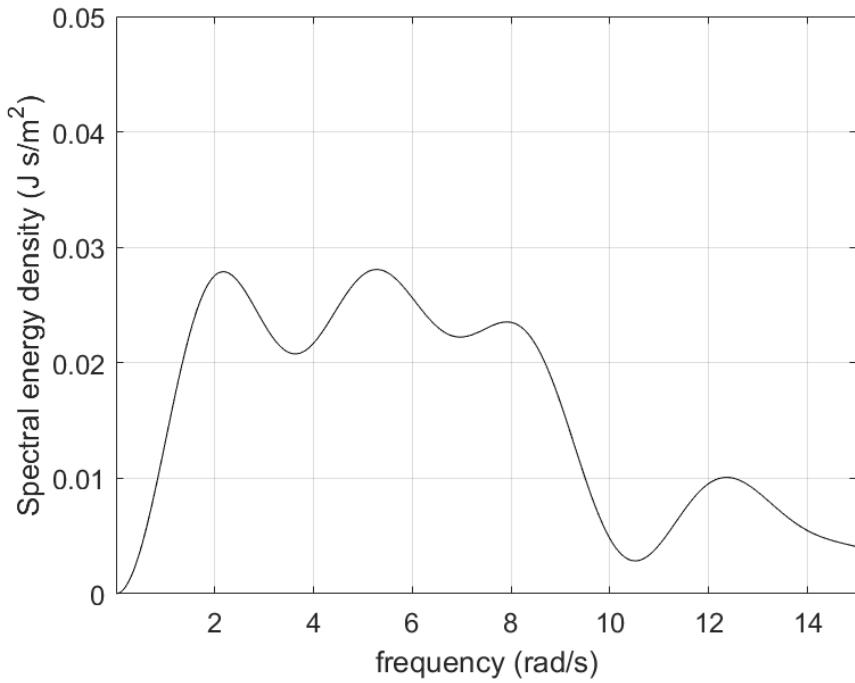


German
Aerospace Center

Back up slides – Elektra 1



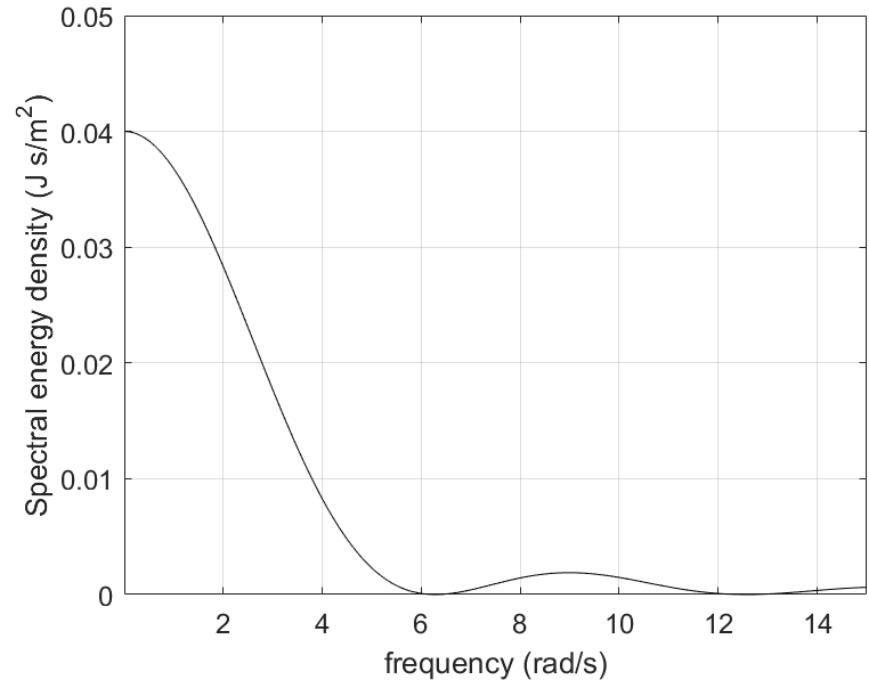
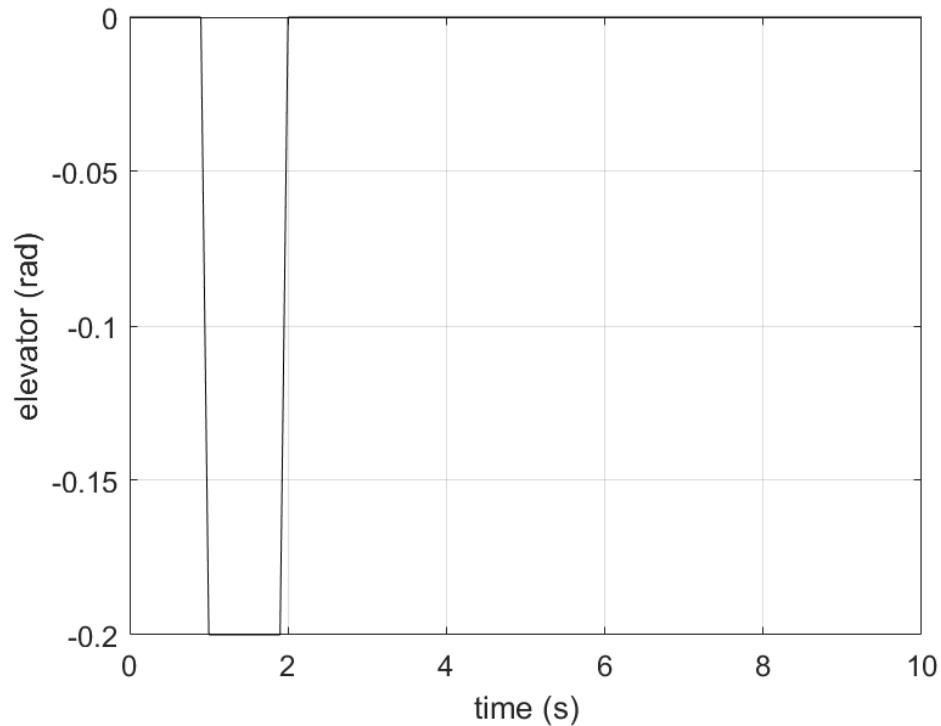
Back up slides – Elektra 1



Autonomous Systems Lab

German
Aerospace Center

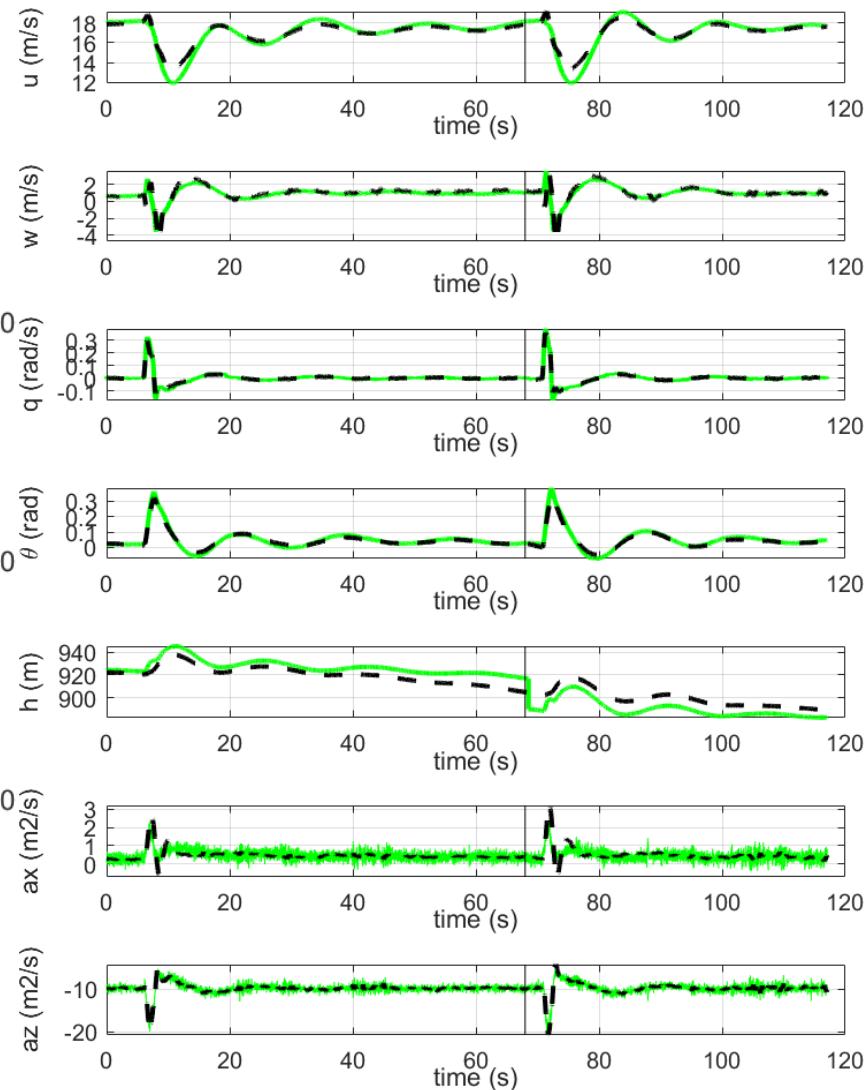
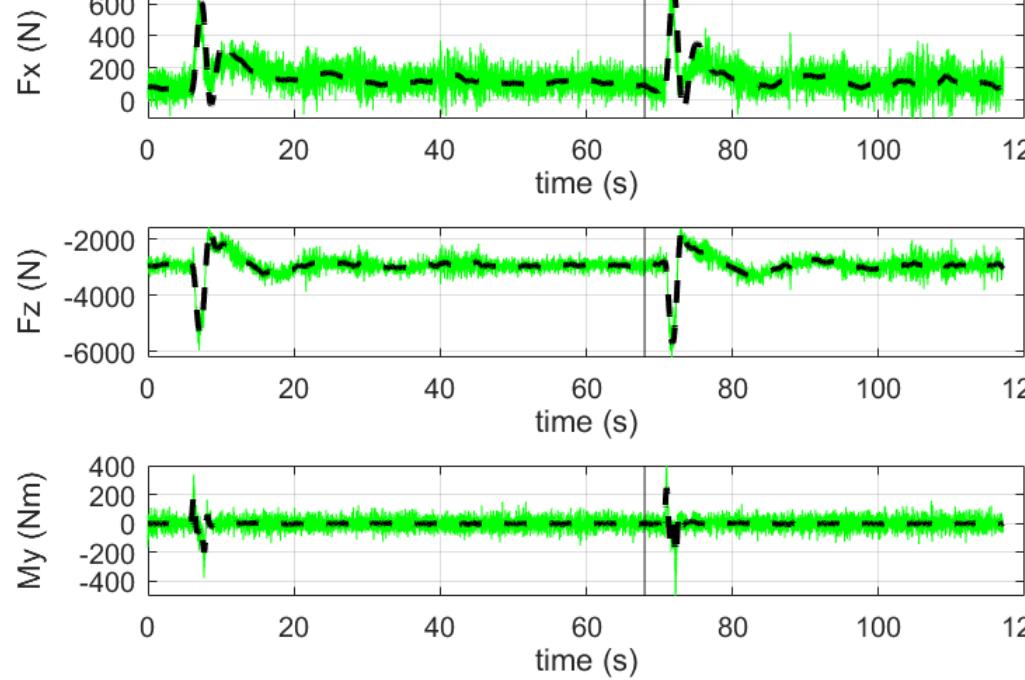
Back up slides – Elektra 1



Autonomous Systems Lab

German
Aerospace Center

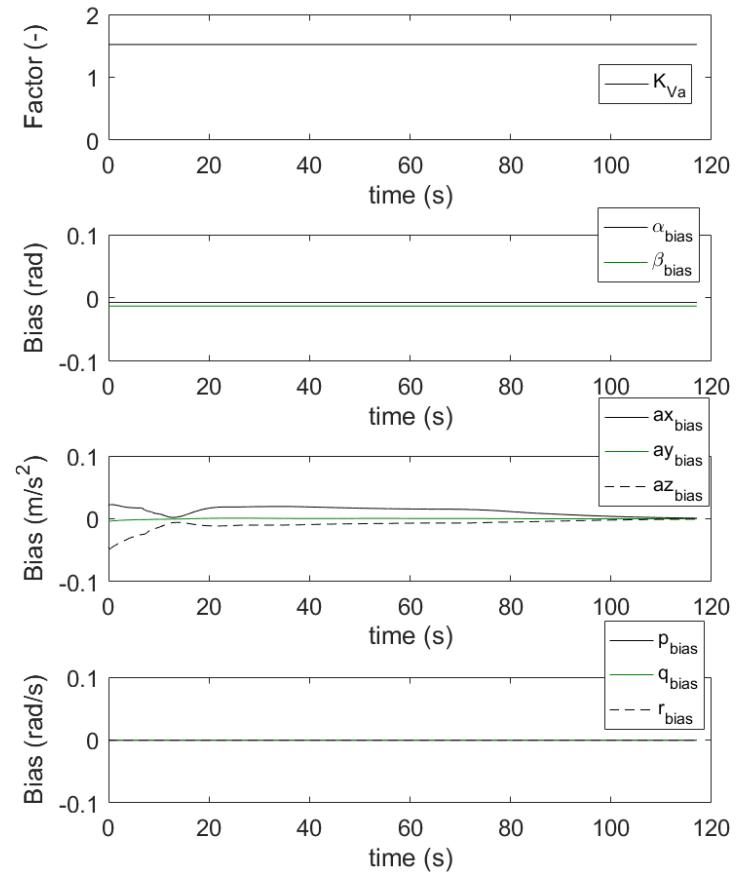
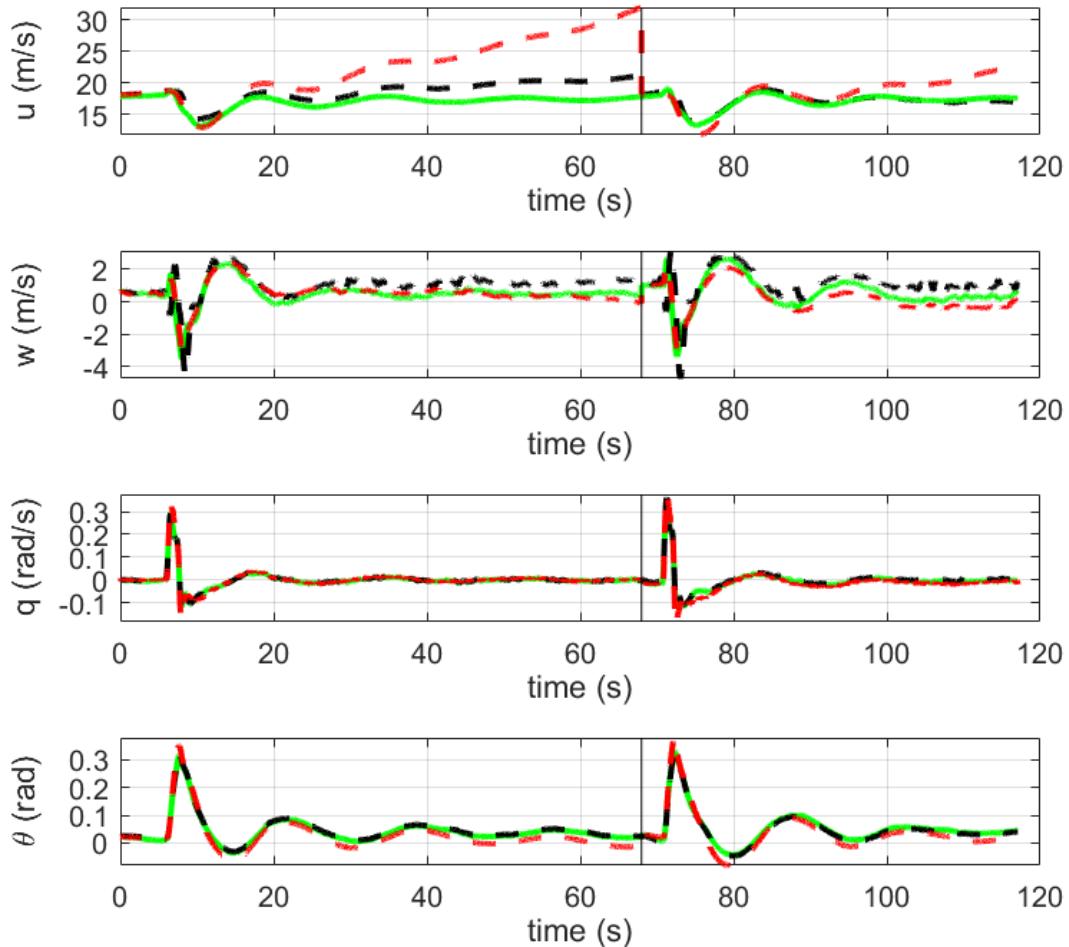
Back up slides – Elektra 1



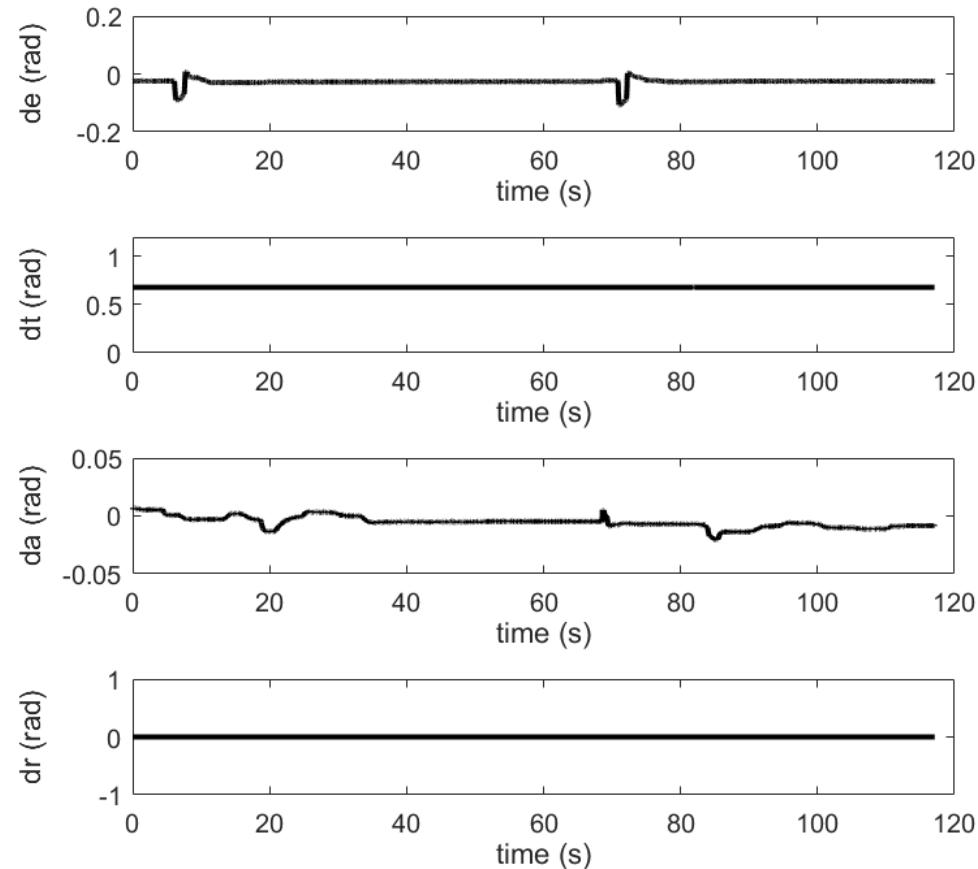
Autonomous Systems Lab

German
Aerospace Center

Back up slides – Elektra 1



Back up slides – Elektra 1



Back up slides – Elektra 1

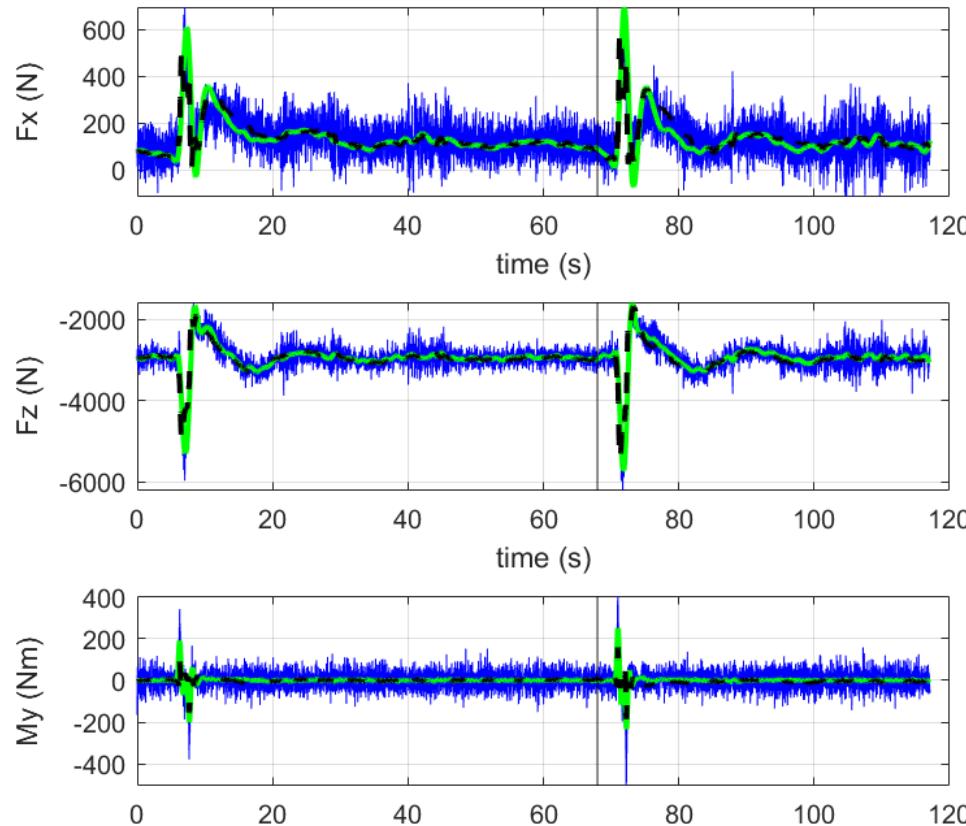


Table 2.5: R-squared, RMSE and NRMSE for forces and moments prediction of linear model.

| Identification | | | | Validation | | | |
|----------------|-------|-------|-------|------------|-------|-------|-------|
| Coef | C_X | C_Z | C_m | Coef | C_X | C_Z | C_m |
| R^2 | 0.69 | 0.79 | 0.30 | R^2 | 0.65 | 0.77 | 0.29 |
| RMSE | 41.89 | 153.7 | 15.89 | RMSE | 49.15 | 178.5 | 19.29 |
| NRMSE | 0.065 | 0.042 | 0.041 | NRMSE | 0.063 | 0.043 | 0.039 |

Back up slides – Elektra 1

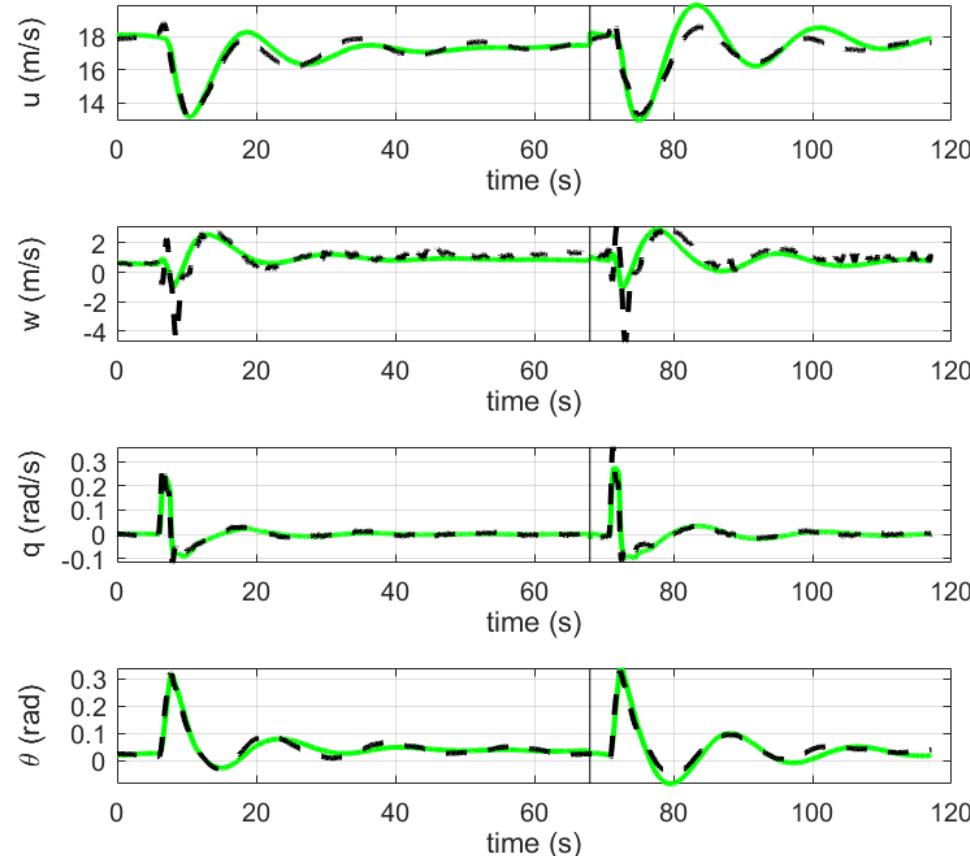
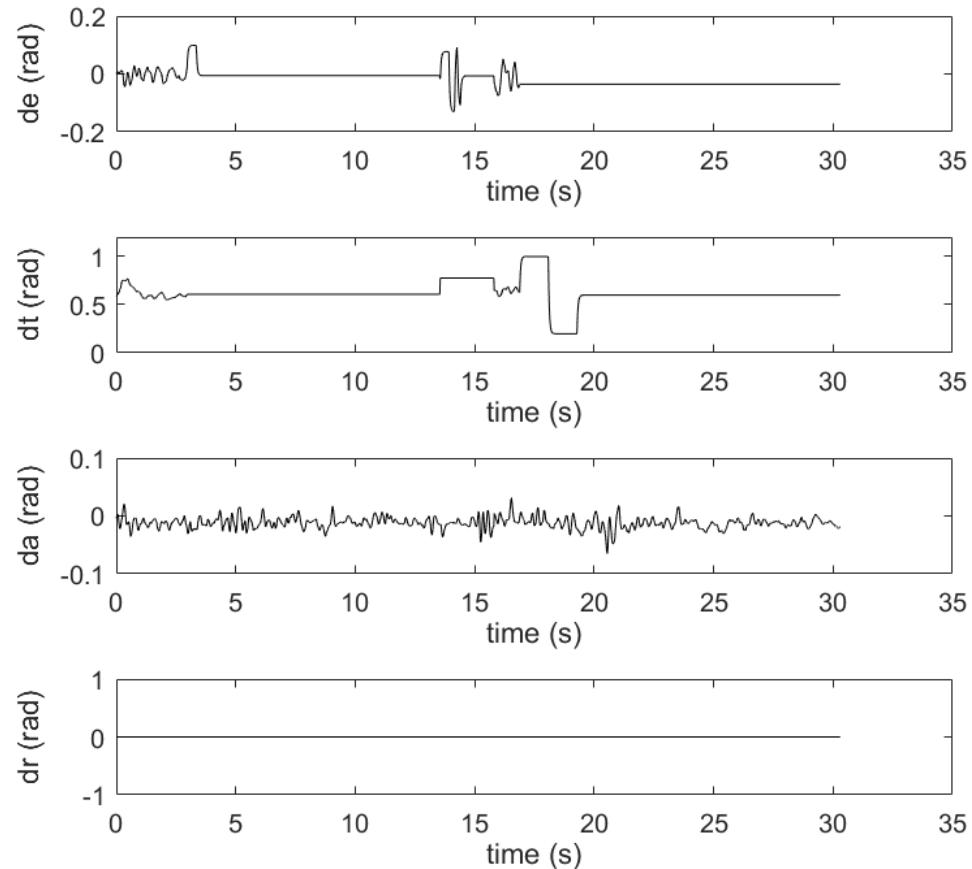


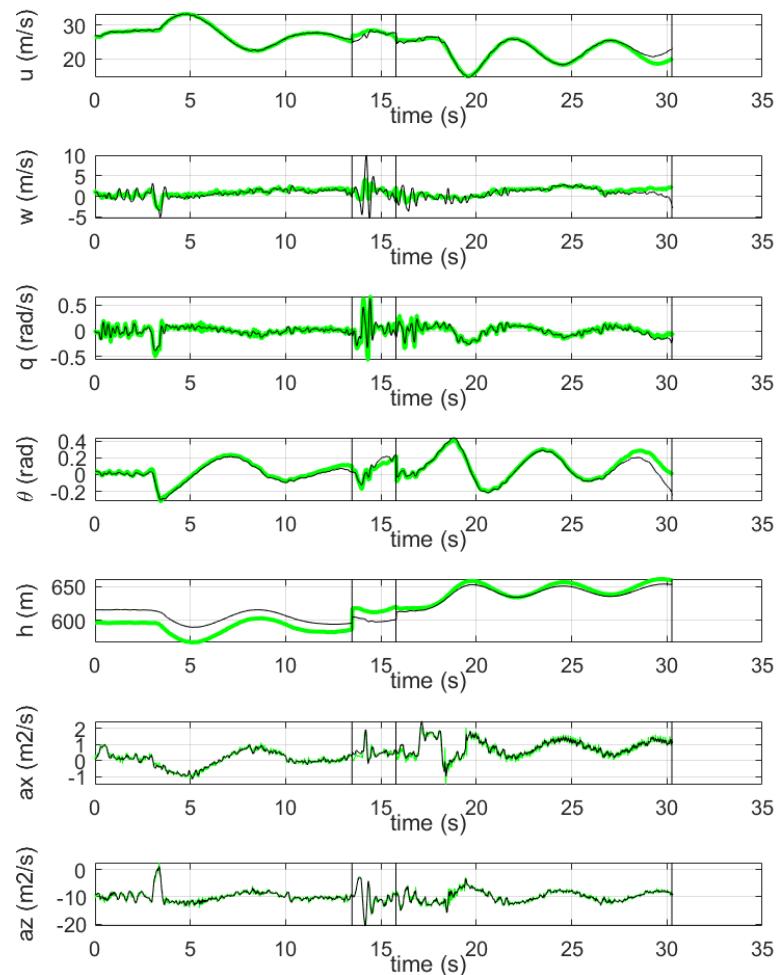
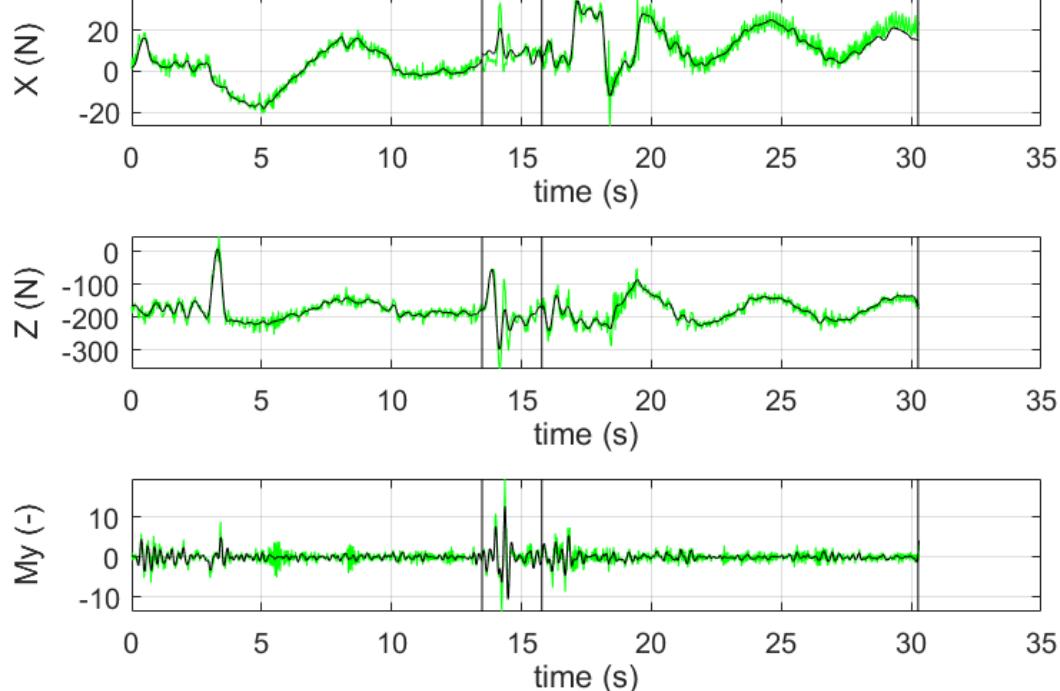
Table 2.7: TIC and GOF values for forward simulation using identified linear model.

| Nonlinear: | | | | | Validation | | | | |
|------------|----------------|--------|--------|----------|------------|------------|--------|--------|----------|
| | Identification | | | | | Validation | | | |
| State | u | w | q | θ | State | u | w | q | θ |
| GOF | 0.9037 | 0.5856 | 0.8959 | 0.9323 | GOF | 0.7156 | 0.5119 | 0.8765 | 0.8981 |
| TIC | 0.0094 | 0.2154 | 0.1660 | 0.0971 | TIC | 0.1872 | 0.2626 | 0.1753 | 0.1239 |

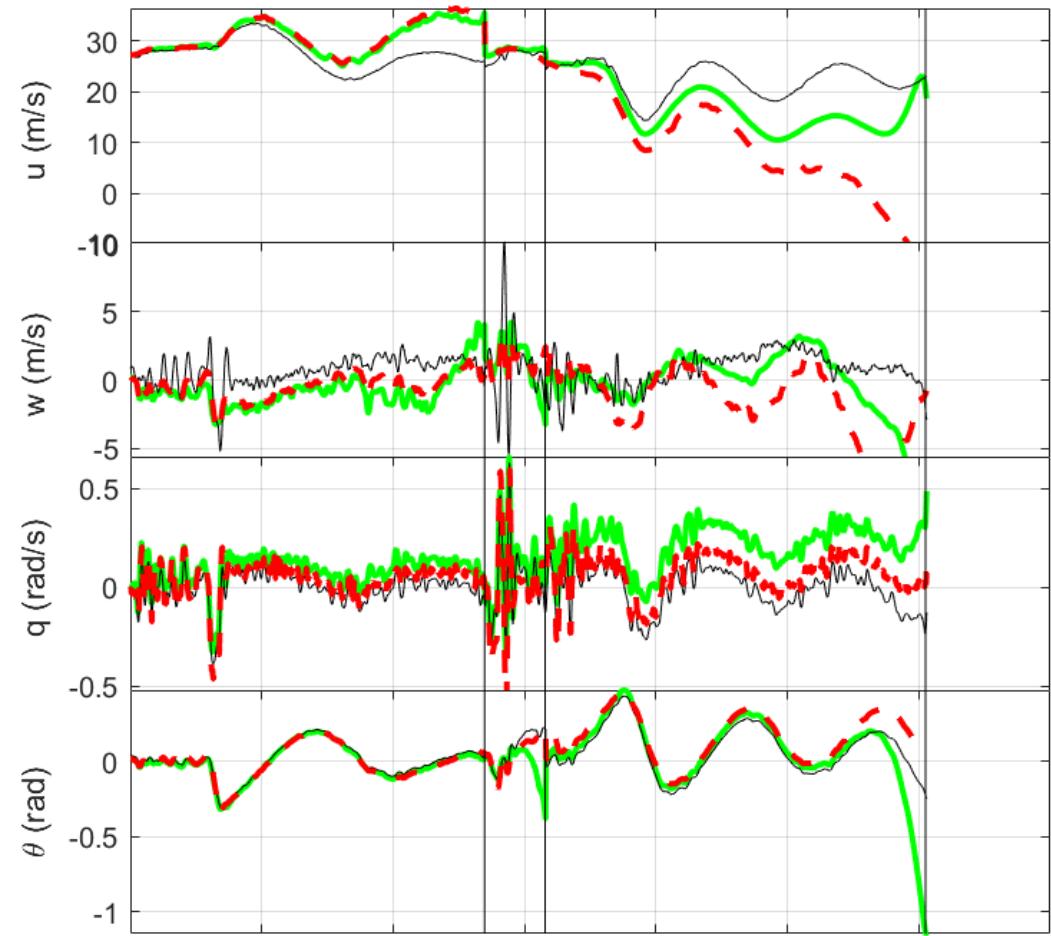
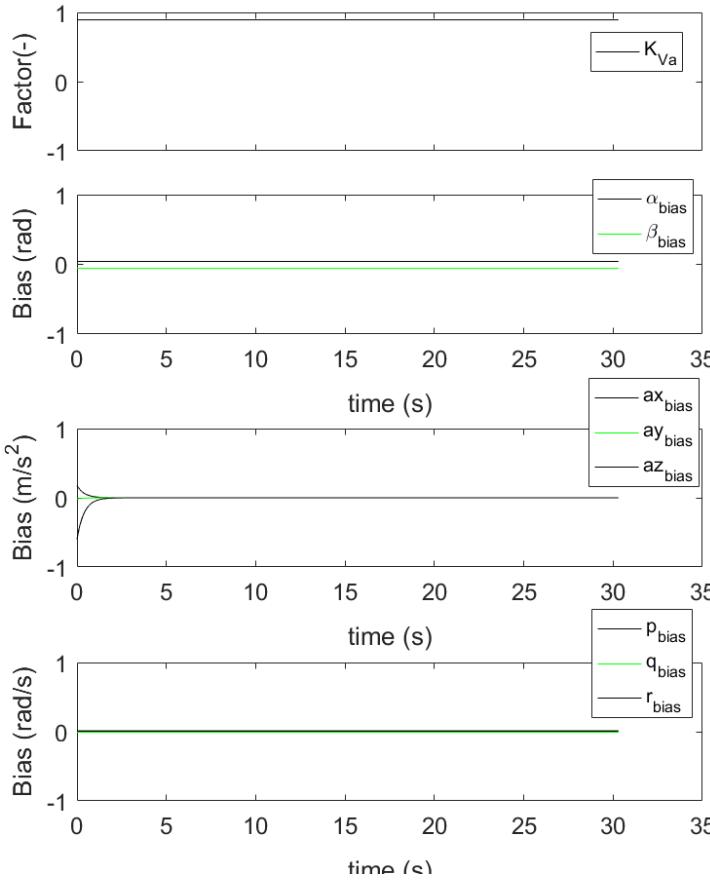
Back up slides – Penguin BE



Back up slides – Penguin BE



Back up slides – Penguin BE



Back up slides – Penguin BE

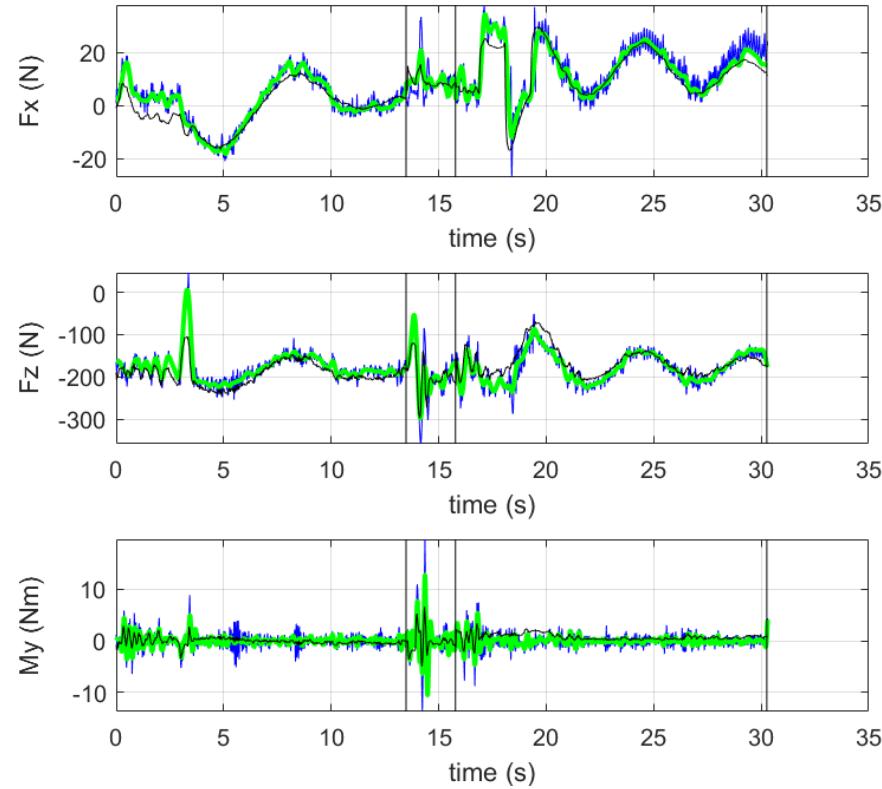


Table 3.12: Averaged R-squared, RMSE and NRMSE for forces and moments prediction of linear model.

| | Identification | | | | Validation | | | |
|-------|----------------|--------|--------|-------|------------|-------|-------|--------|
| | Coef | C_X | C_Z | C_m | Coef | C_X | C_Z | C_m |
| R^2 | 0.868 | 0.616 | 0.422 | | R^2 | 0.848 | 0.665 | 0.1359 |
| RMSE | 3.109 | 19.779 | 1.0899 | | RMSE | 4.10 | 20.63 | 1.239 |
| NRMSE | 0.071 | 0.067 | 0.045 | | NRMSE | 0.078 | 0.067 | 0.053 |

Back up slides – Penguin BE

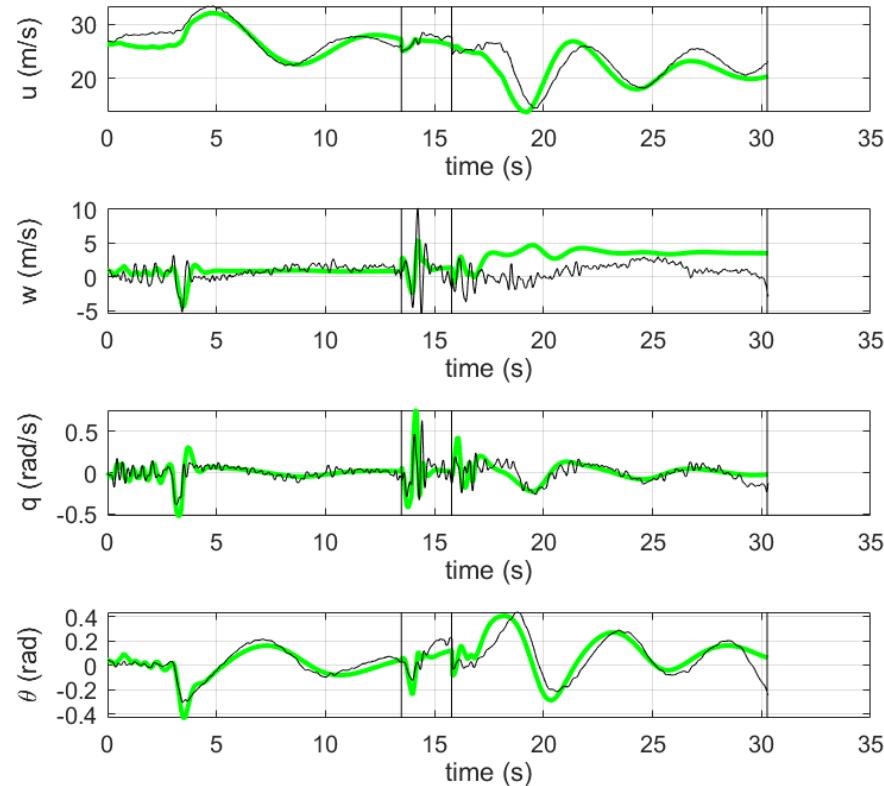


Table 3.14: TIC and GOF values for forward simulation with linear model.

| Identification | | | | Validation | | | | | |
|----------------|------|------|-------|------------|-------|-------|-------|-------|----------|
| State | u | w | q | θ | State | u | w | q | θ |
| GOF | 0.77 | 0.72 | 0.514 | 0.688 | GOF | 0.76 | 0.52 | 0.52 | 0.72 |
| TIC | 0.03 | 0.28 | 0.355 | 0.254 | TIC | 0.038 | 0.453 | 0.345 | 0.236 |

Back up slides – Global

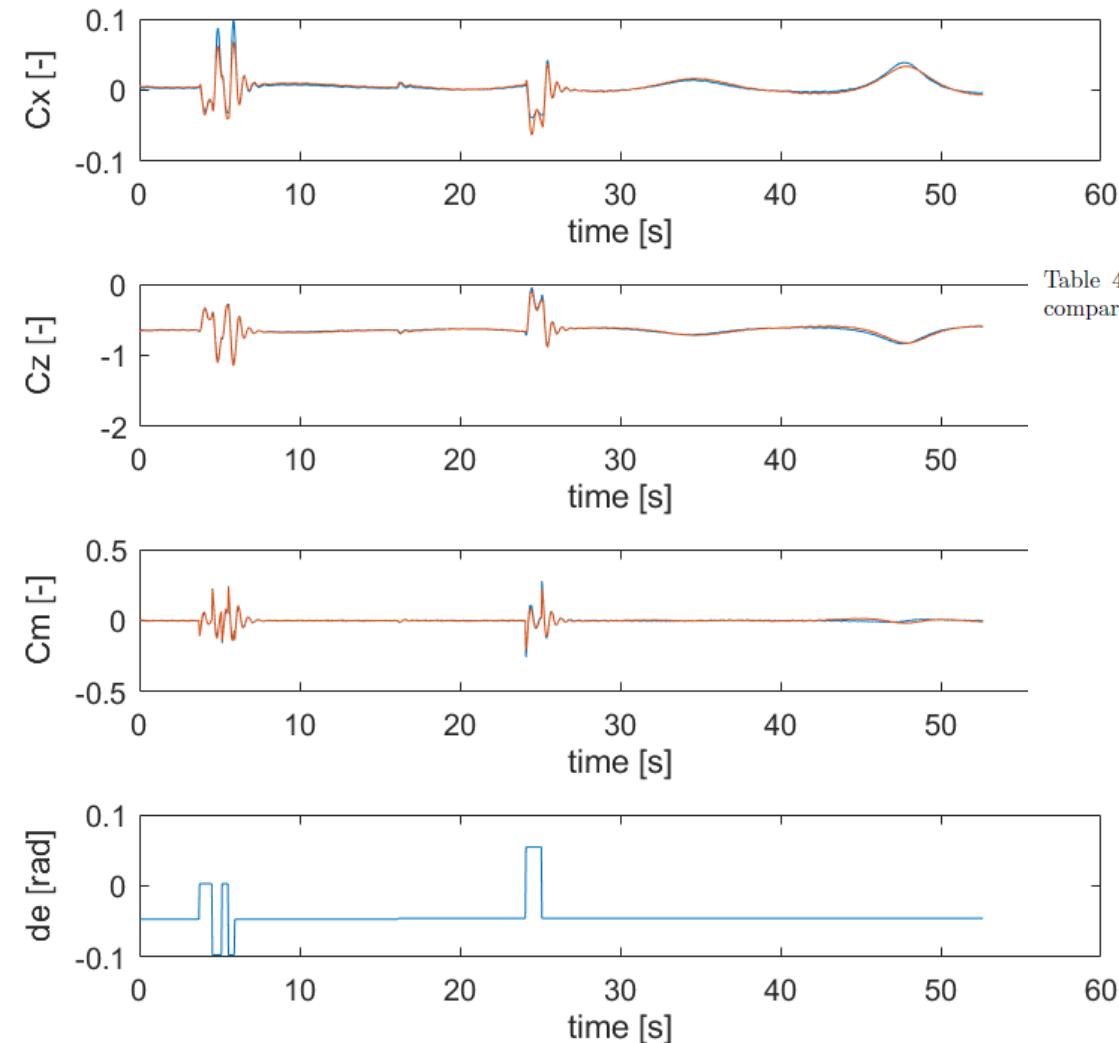
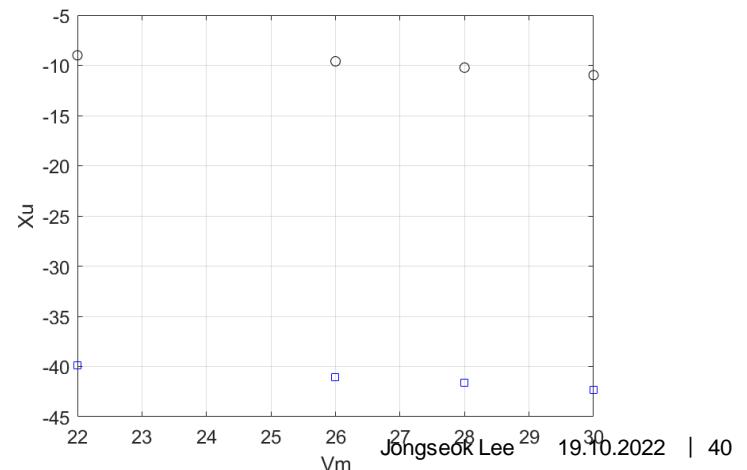
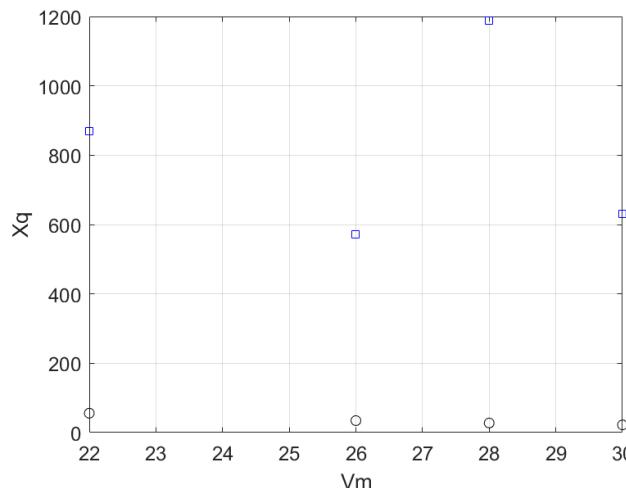
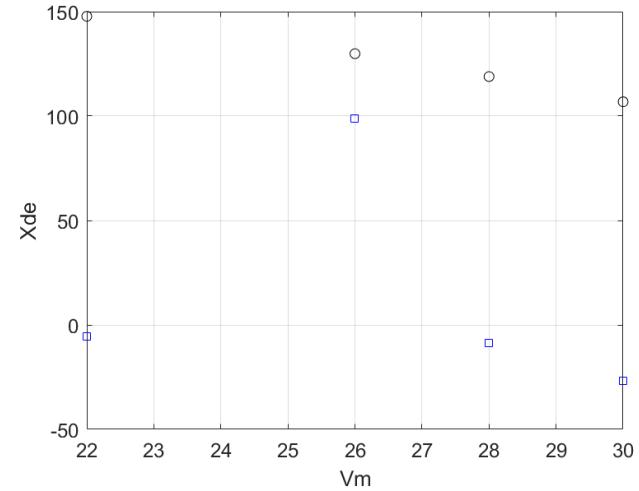
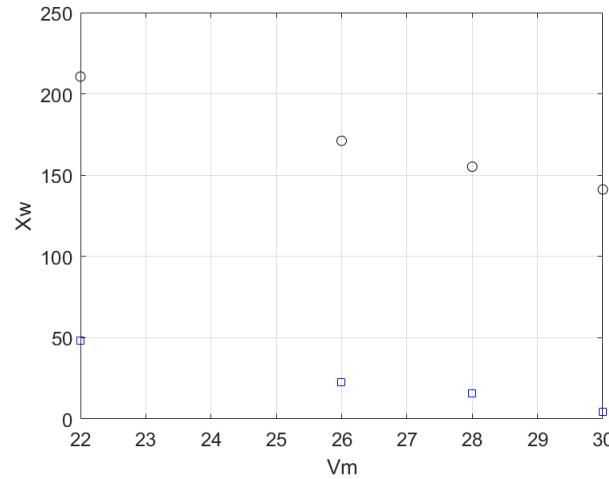


Table 4.1: Identified model parameters using VLM based simulation data (ID) compared to direct output of VLM software (AVL).

| ID: | | | | | |
|--------------|---------|--------------|----------|--------------|--------|
| Term | Value | Term | Value | Term | Value |
| C_{x_u} | -0.004 | C_{z_u} | -0.036 | C_{m_u} | 0.0105 |
| C_{x_w} | 0.0405 | C_{z_w} | -0.231 | C_{m_w} | -0.095 |
| C_{x_q} | 0.0026 | C_{z_q} | -0.08595 | C_{m_q} | -0.196 |
| $C_{x_{de}}$ | 0.00059 | $C_{z_{de}}$ | -0.0097 | $C_{m_{de}}$ | -0.047 |

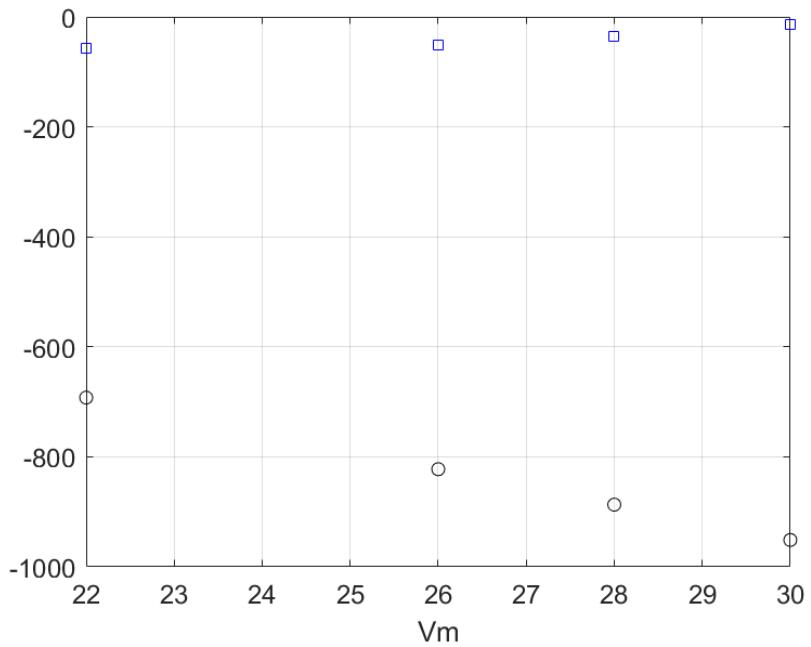
| AVL: | | | | | |
|--------------|---------|--------------|---------|--------------|---------|
| Term | Value | Term | Value | Term | Value |
| C_{x_u} | -0.004 | C_{z_u} | -0.04 | C_{m_u} | 0.011 |
| C_{x_w} | 0.045 | C_{z_w} | -0.25 | C_{m_w} | -0.105 |
| C_{x_q} | 0.0029 | C_{z_q} | -0.0955 | C_{m_q} | -0.218 |
| $C_{x_{de}}$ | 0.00054 | $C_{z_{de}}$ | -0.0088 | $C_{m_{de}}$ | -0.0431 |

Back up slides - Global

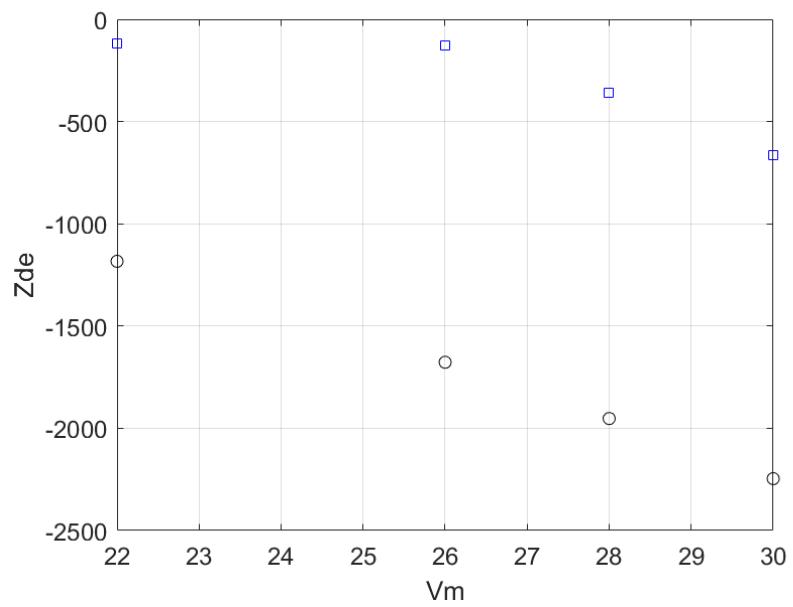




Z_w

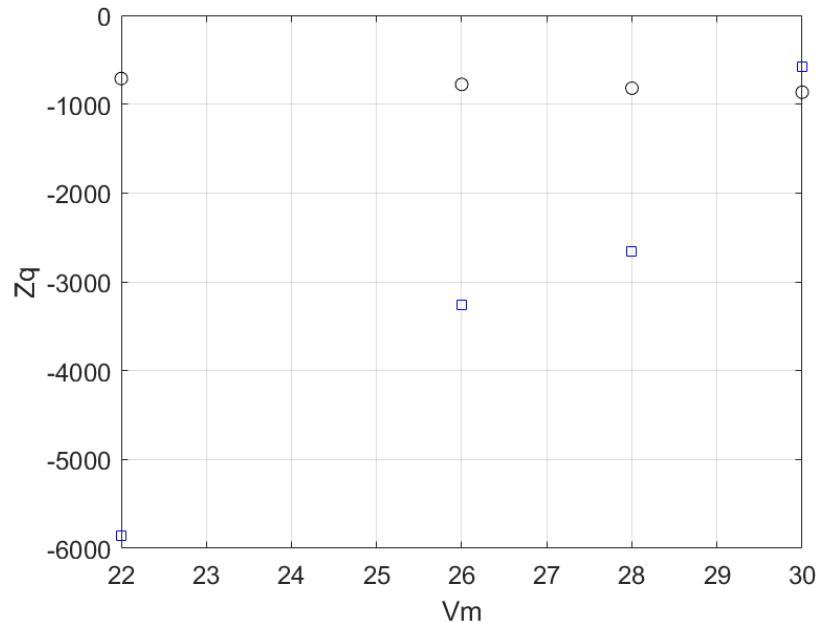


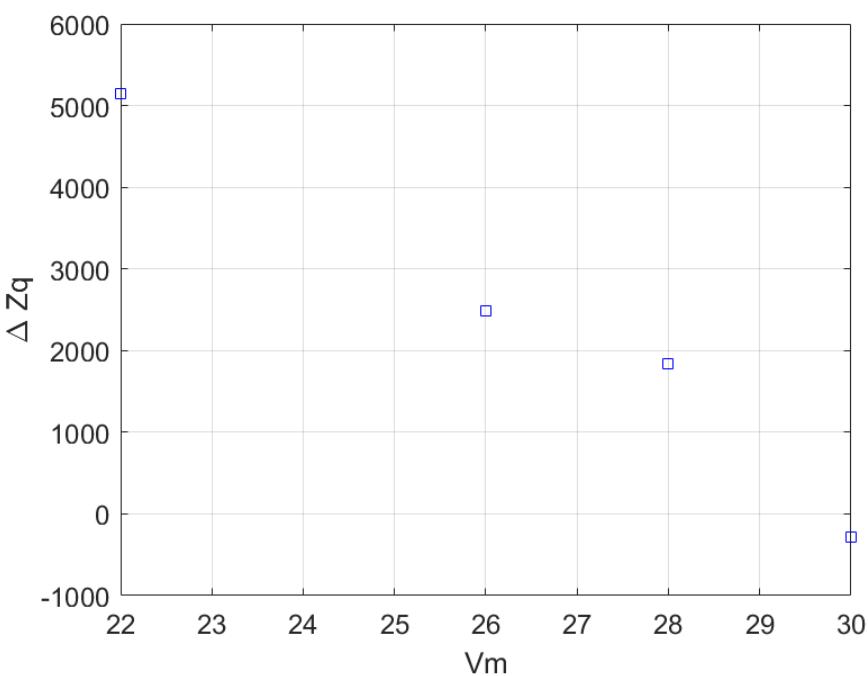
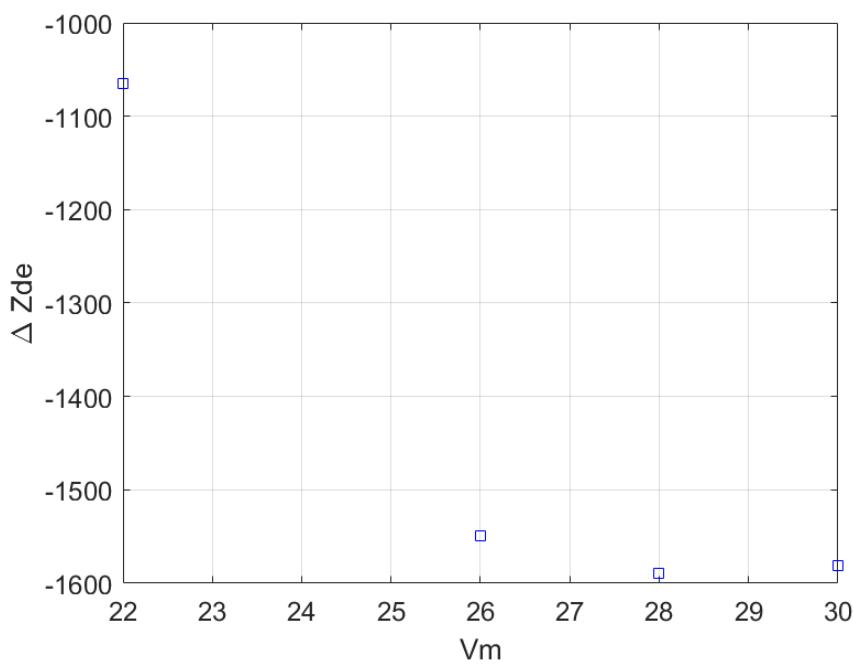
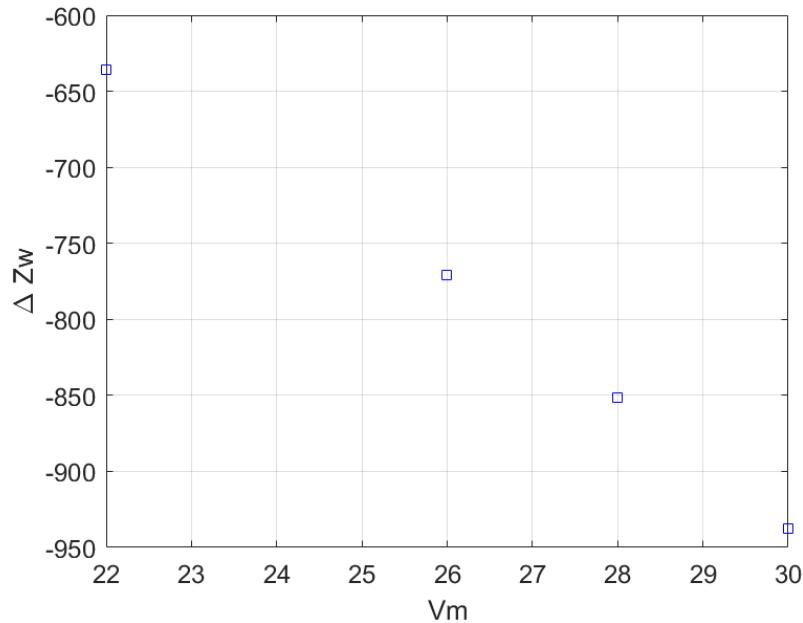
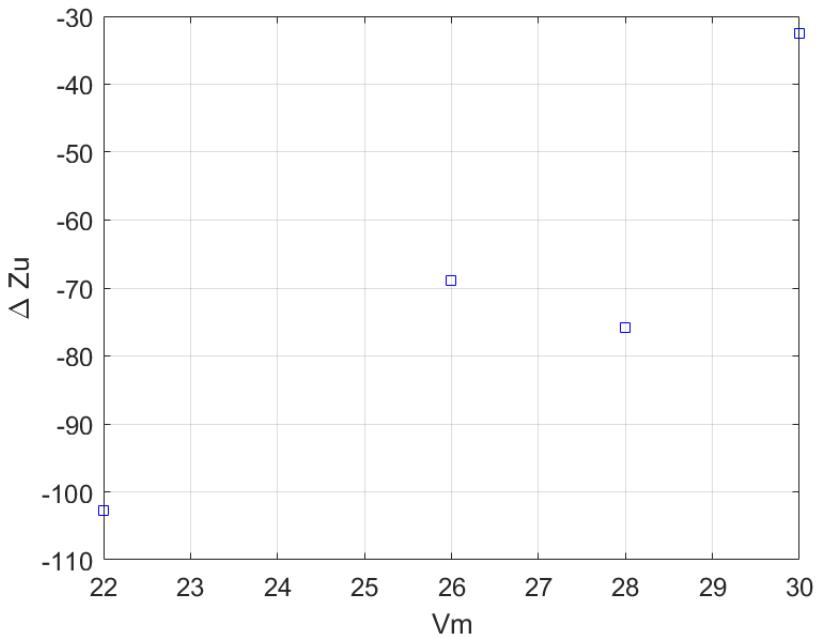
Z_{de}



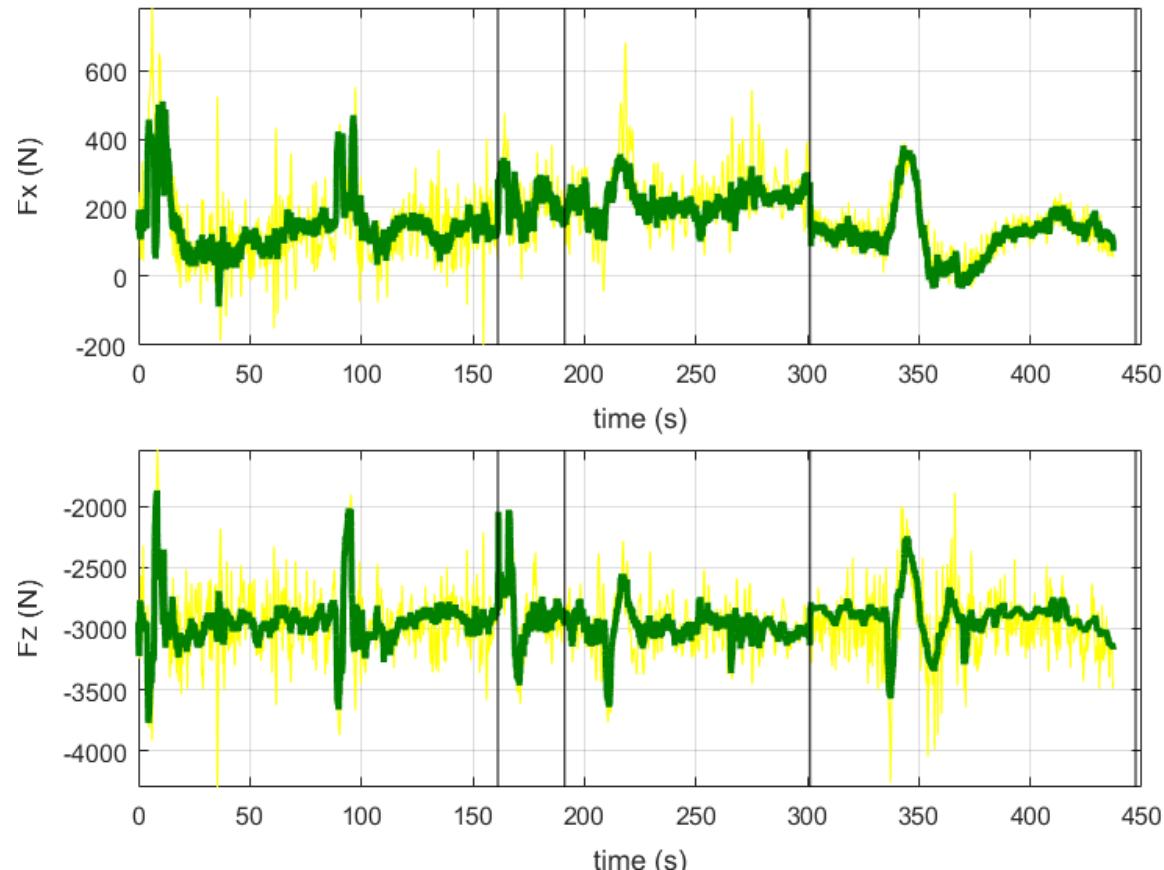
Z_u

Z_q





Back up slides - Global



Back up slides

Table 4.6: Case studies with various choice of training and validation set.

| Case A | V1 | V2 | V3 | V4 | ALL |
|------------|---------|---------|---------|----------|---------|
| RMSE F_x | 81.671 | 90.608 | 73.837 | 37.251 | 70.841 |
| RMSE F_y | 2.273E2 | 2.262E2 | 2.057E2 | 2.045E2 | 2.151E2 |
| Case B | V1 | V2 | V3 | V4 | ALL |
| RMSE F_x | 80.86 | 99.391 | 78.729 | 50.709 | 77.422 |
| RMSE F_y | 1.697E2 | 2.262E2 | 2.236E2 | 2.4066E2 | 2.257E2 |
| Case C | V1 | V2 | V3 | V4 | ALL |
| RMSE F_x | 86.883 | 97.78 | 74.22 | 37.251 | 74.033 |
| RMSE F_y | 3.710E2 | 2.359E2 | 1.884E2 | 2.0456E2 | 2.329E2 |
| Case D | V1 | V2 | V3 | V4 | ALL |
| RMSE F_x | 83.72 | 99.39 | 85.45 | 83.90 | 88.115 |
| RMSE F_y | 1.802E2 | 2.262E2 | 1.884E2 | 2.1704E2 | 2.112E2 |

$$\begin{aligned}
 \Delta X_u &= 0.083 \cdot V_m + 29.088 \\
 \Delta X_w &= -3.790 \cdot V_m + 246 \\
 \Delta X_q &= -57.9 \cdot V_m + 462 \\
 \Delta X_{de} &= -4.3 \cdot V_m + 248 \\
 \Delta Z_u &= 4.4 \cdot V_m - 201.3 \\
 \Delta Z_w &= -36 \cdot V_m + 156 \\
 \Delta Z_q &= -552 \cdot V_m + 1729.8 \\
 \Delta Z_{de} &= -1.39 \cdot V_m^2 - 17.82 \cdot V_m
 \end{aligned}$$

