

MGE-MSR-01 – Sensors and State Estimation

Introduction to Exercise 2
WS 20/21

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eCampus PPT submission deadline: Tuesday 12.01.2020., 23:59

Discussion Time: Wednesday 13.01.2021., 08:30.

Topic: Exercises deal with three different topics

- Exercise 1: Inertial navigation
- **Exercise 2: Kalman filtering**
- Exercise 3: Laser scanning

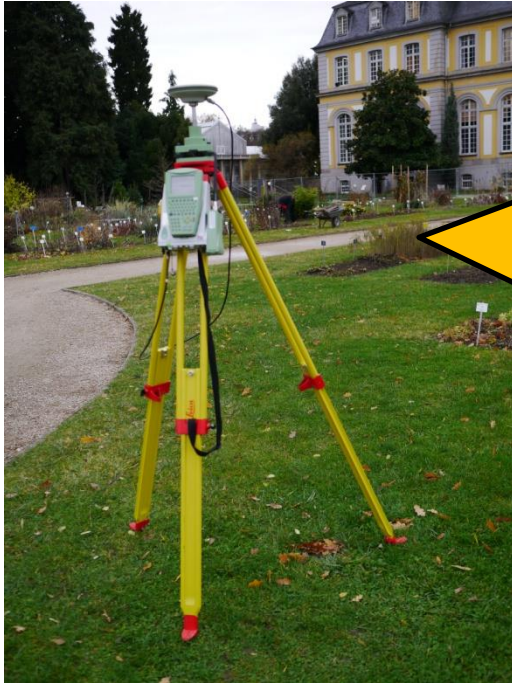
Date	Preliminary discussion	Exercise examination
	Begin of lectures	
	Exercise 1	–
16.12.2020	Exercise 2	Exercise 1
13.01.2021	Exercise 3	Exercise 2
	–	Exercise 3
	End of lectures	

- Understanding theory of Kalman Filtering
(functional and stochastic model, steps, system and measurement model, tuning)
- Implementing simplified 2D Extended Kalman Filter
(+ visualization, interpretation and comparison of the results)

Main literature:

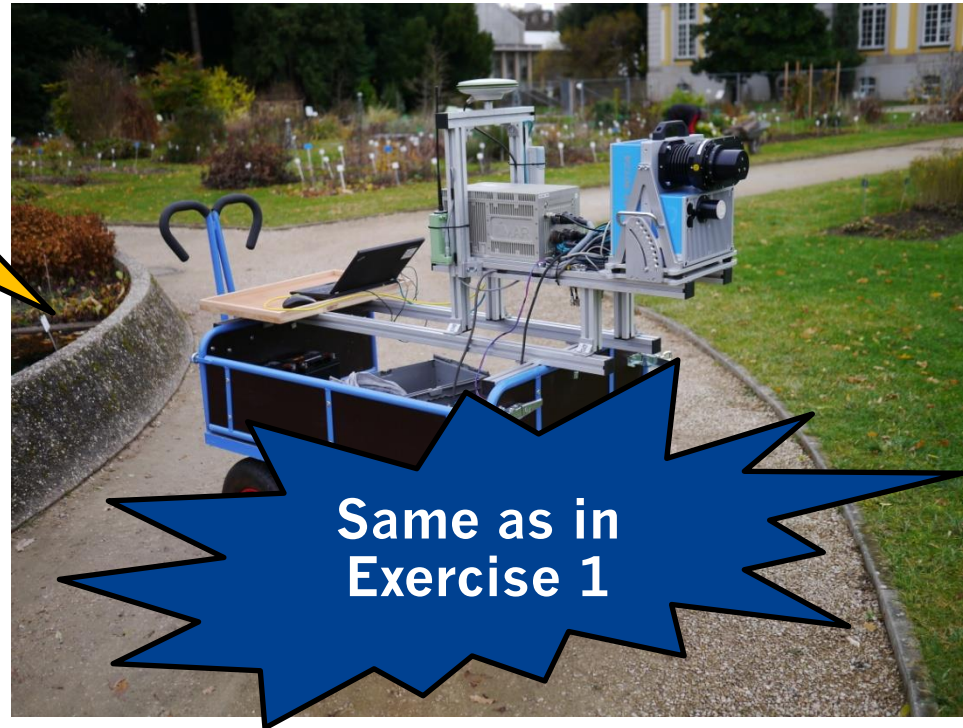
- MGE-MSR-01 – Sensors & State Estimation
(Lectures: Kalman Filter I., II., III.)
- Groves, P. D.: Navigation Using Inertial Sensors,
University College London, UK, IEEE A&E Systems magazine,
February 2015, Part II of II.

Components of the MSS



GNSS master station
(sending observations
to MSS → RTK-GNSS)

Kinematic MSS for laser
scanning in motion



**Same as in
Exercise 1**

Components of the MSS

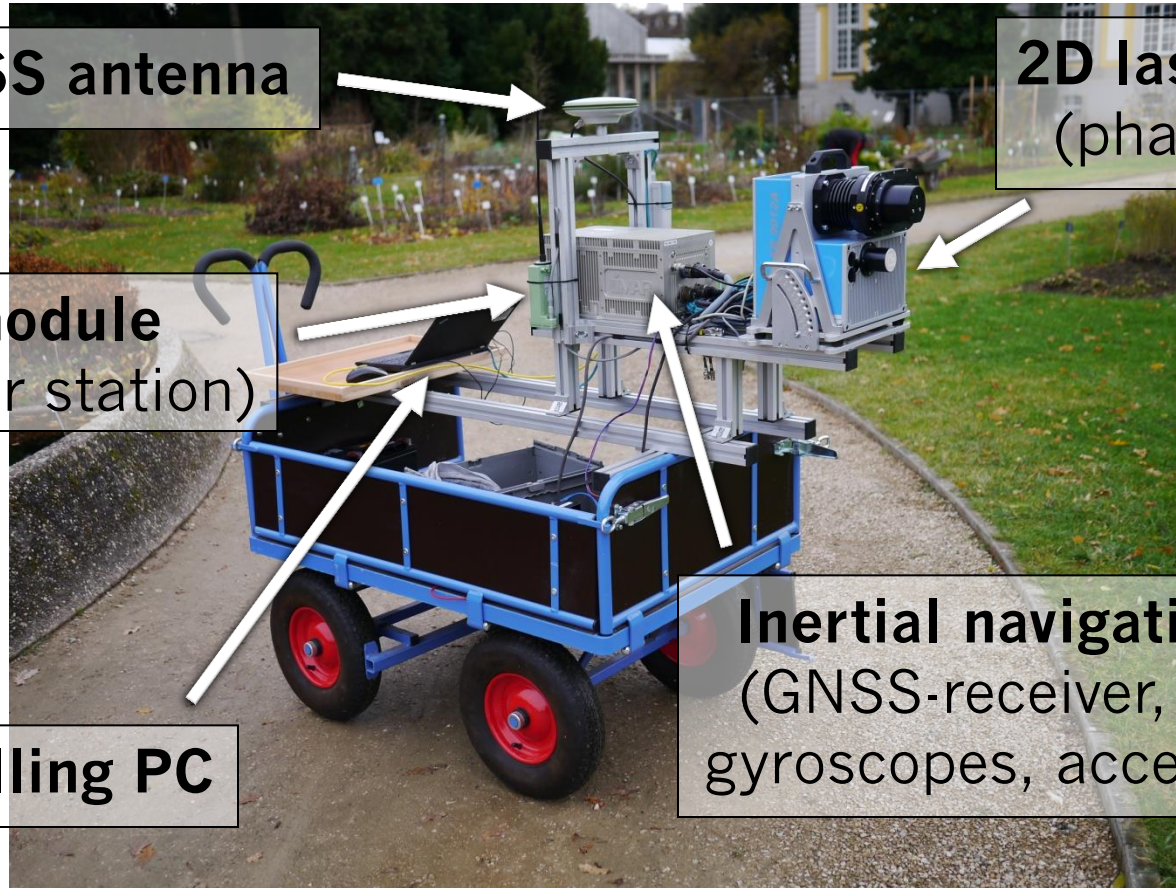
GNSS antenna

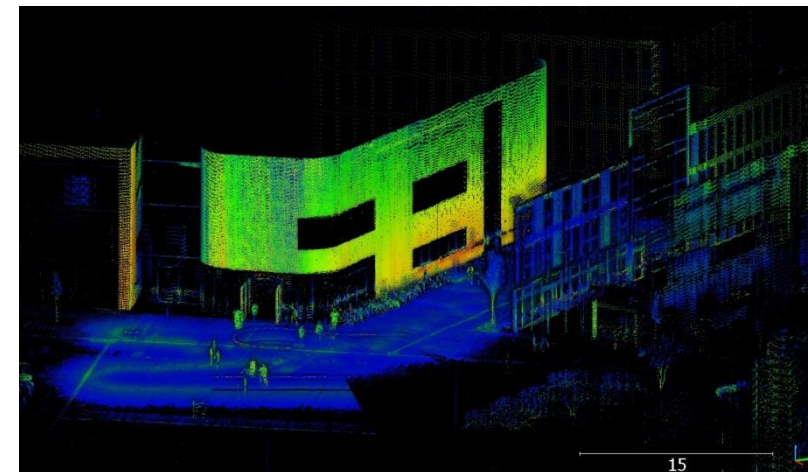
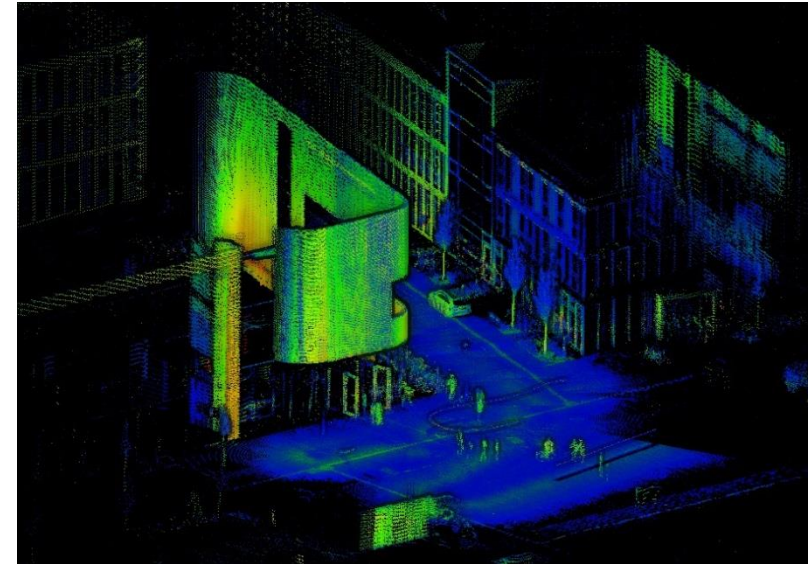
2D laser scanner
(phase-based)

Radio module
(for master station)

Controlling PC

Inertial navigation system
(GNSS-receiver, fiber-optic
gyroscopes, accelerometers)





• Describe Kalman Filter:

- Algorithm diagram / flow-chart
- Two main steps of the KF
- All formulas
- All variables in formulas
- Input data
- Output data
- Necessary initial values
- Difference between KF and EKF

Prediction

$$\begin{aligned}\mathbf{x}_k^- &= \Phi \mathbf{x}_{k-1} + \mathbf{B} \mathbf{u}_k \\ \mathbf{P}_k^- &= \Phi \mathbf{P}_{k-1} \Phi^T + \mathbf{G} \mathbf{Q} \mathbf{G}^T\end{aligned}$$

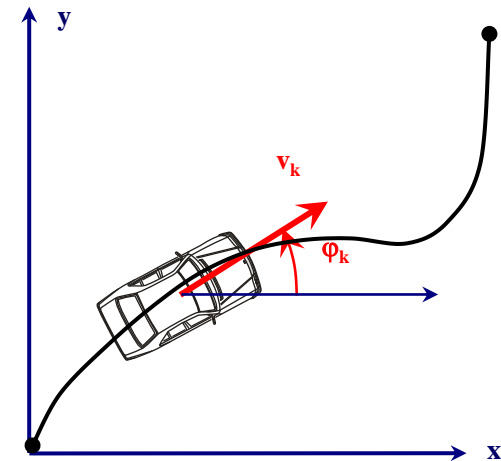
$$\begin{aligned}K_k &= P_k^- H^T (H P_k^- H^T + R)^{-1} \\ \mathbf{x}_k &= \mathbf{x}_k^- + K_k (z_k - H \mathbf{x}_k^-) \\ P_k &= (I - K_k H) P_k^-\end{aligned}$$

Correction

- Describe selected **system model** for this exercise
(Lecture: Kalman_Filter_III, Example 2)

Presumptions:

- constant angular rate
- constant acceleration



$$\mathbf{x}_k = f(\mathbf{x}_{k-1}, \mathbf{w}_k)$$

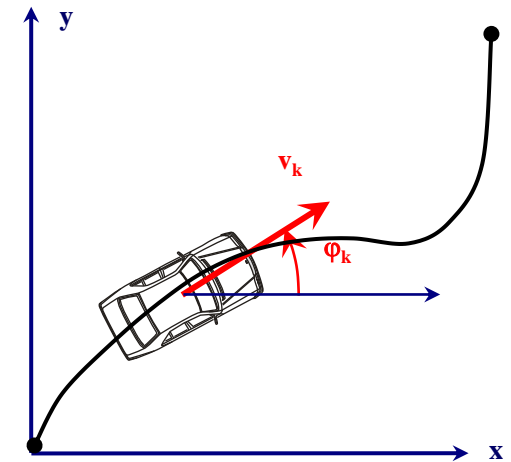
$$\begin{bmatrix} x_k \\ y_k \\ \varphi_k \\ \dot{\varphi}_k \\ v_k \\ a_k \end{bmatrix} = f \left(\begin{bmatrix} x_{k-1} \\ y_{k-1} \\ \varphi_{k-1} \\ \dot{\varphi}_{k-1} \\ v_{k-1} \\ a_{k-1} \end{bmatrix} \right) = \begin{bmatrix} x_{k-1} + \cos(\varphi_{k-1})v_{k-1}\Delta t \\ y_{k-1} + \sin(\varphi_{k-1})v_{k-1}\Delta t \\ \varphi_{k-1} + \dot{\varphi}_{k-1}\Delta t \\ \dot{\varphi}_{k-1} + w_{\dot{\varphi}}\Delta t \\ v_{k-1} + a_{k-1}\Delta t \\ a_{k-1} + w_a\Delta t \end{bmatrix}$$

- Describe selected **measurement model**
(Lecture: Kalman_Filter_III, Example 2)

Measurements:

- acceleration,
- angular rate,
- GNSS

if loop !



$$\mathbf{z}_k = \begin{bmatrix} x_{gps,k} \\ y_{gps,k} \\ a_k \\ \omega_k \end{bmatrix} = \mathbf{H}\mathbf{x}_k = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \mathbf{x}_k$$

- Measurements have different measurement rate!

if there_is_GNSS_measurements



Check this part
of the code!

$$\mathbf{z}_k = \begin{bmatrix} x_{gps,k} \\ y_{gps,k} \\ a_k \\ \omega_k \end{bmatrix} = \mathbf{H}\mathbf{x}_k = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \mathbf{x}_k$$

else

$$\mathbf{z}_k = \begin{bmatrix} a_k \\ \omega_k \end{bmatrix} = \mathbf{H}\mathbf{x}_k = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \mathbf{x}_k$$

MSS measurements (IMAR.mat data files)

1. GNSS observations

- Transform ellipsoidal coordinates to UTM coordinates (N, E)
(Necessary adaptation for 2D EKF, see “General Task” in
Exercise 2 sheet)

`x = imar_data.gpsUTM[:,0]` # East UTM [m]

`y = imar_data.gpsUTM[:,1]` # North UTM [m]

MSS measurements (IMAR.mat data files)

2. Acceleration (in the movement direction x-axis of the body frame – simplification)

```
a = -imar_data.acceleration[:,0] # acceleration x [m/s^2]
```

(with removed gravity influence!)

3. Angular rate (around z axis of the body frame, only yaw/heading angle)

```
omega = imar_data.angularvelocity[:,2] # angular velocity z [rad/s]
```


Functional model

$$\begin{bmatrix} x_k \\ y_k \\ \varphi_k \\ \dot{\varphi}_k \\ v_k \\ a_k \end{bmatrix} = \begin{bmatrix} x_{k-1} + \cos(\varphi_{k-1})v_{k-1}\Delta t \\ y_{k-1} + \sin(\varphi_{k-1})v_{k-1}\Delta t \\ \varphi_{k-1} + \dot{\varphi}_{k-1}\Delta t \\ \dot{\varphi}_{k-1} + \boxed{w_{\dot{\varphi}}}\Delta t \\ v_{k-1} + a_{k-1}\Delta t \\ a_{k-1} + \boxed{w_a}\Delta t \end{bmatrix}$$

- **System noise**
- Angular acceleration
0.1 [rad/s]
- Linear jerk
1.5 [m/s³]

Measurement noise

- GNSS - 0.05m
- Acceleration - empirically - std(a(1:1000))
- Angular Rate – empirically - std(omega(1:1000))

Functional model

$$\begin{bmatrix} x_k \\ y_k \\ \varphi_k \\ \dot{\varphi}_k \\ v_k \\ a_k \end{bmatrix} = \begin{bmatrix} x_{k-1} + \cos(\varphi_{k-1})v_{k-1}\Delta t \\ y_{k-1} + \sin(\varphi_{k-1})v_{k-1}\Delta t \\ \varphi_{k-1} + \dot{\varphi}_{k-1}\Delta t \\ \dot{\varphi}_{k-1} \\ v_{k-1} \\ a_{k-1} \end{bmatrix}$$

System noise

- Angular acceleration
- 0.1 [rad/s]

Filter tuning

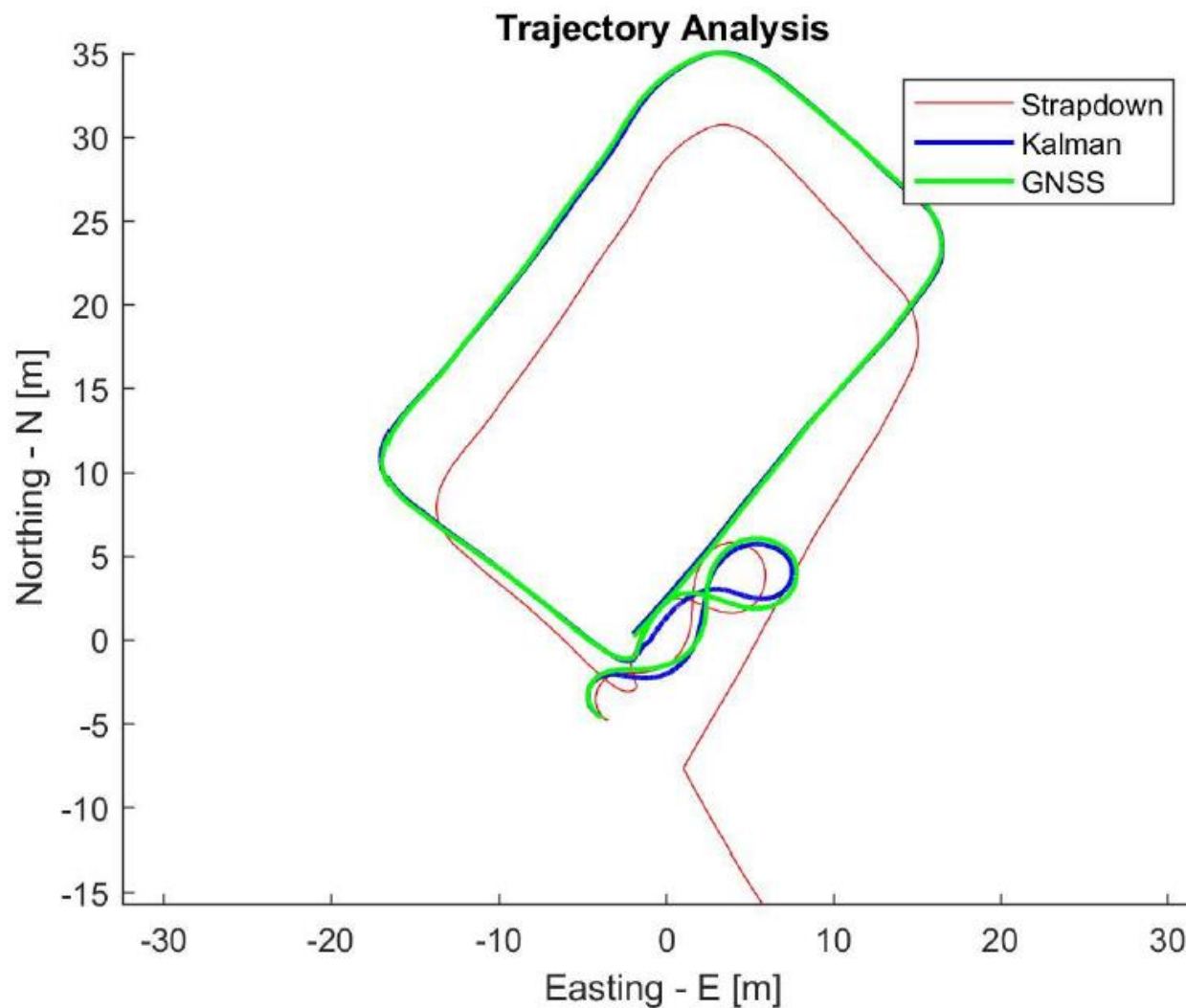
- change system and measurement noise parameters
- Analyse the differences in trajectory

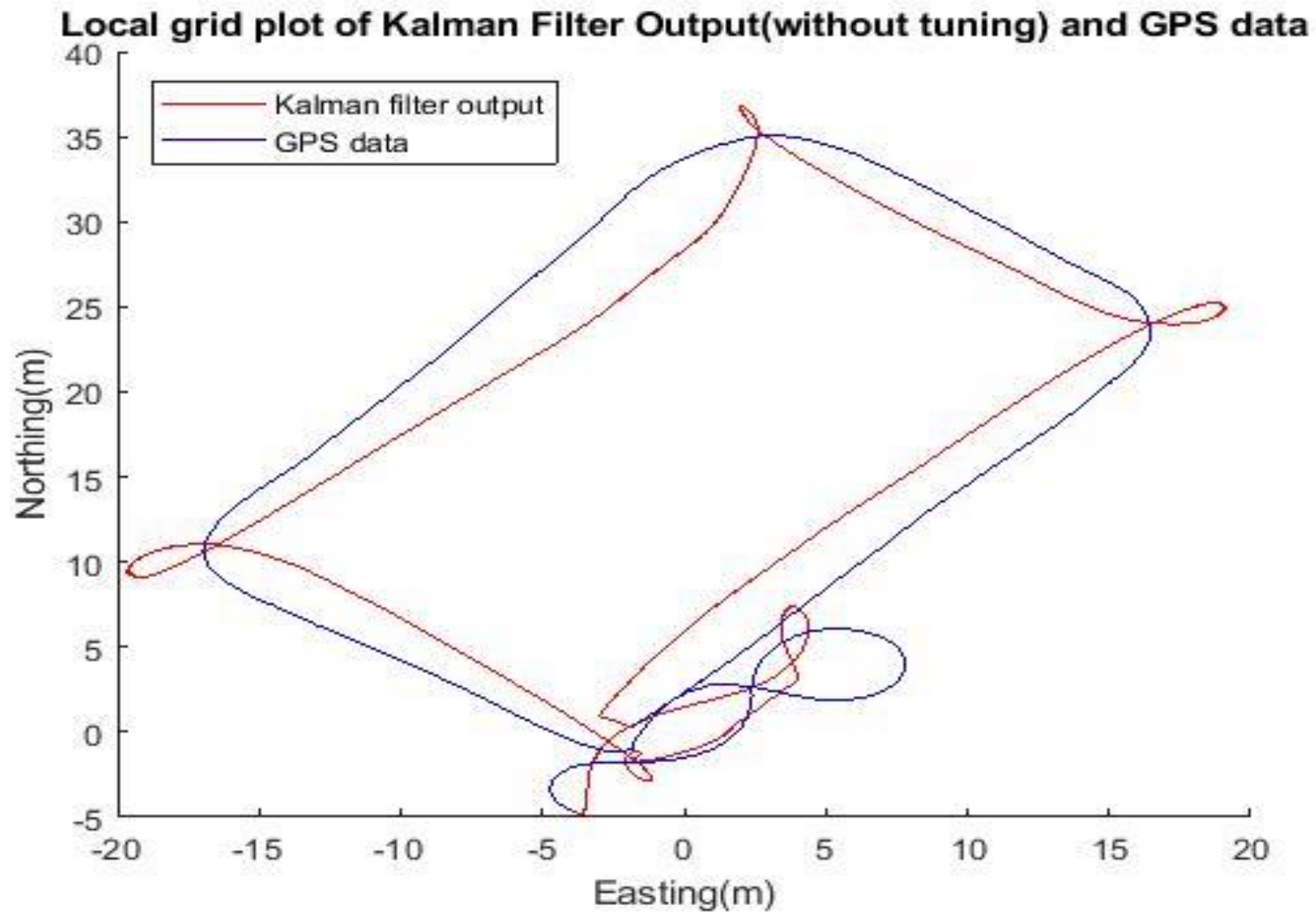
- Given
- Acceleration $(1:1000)$
- Angular Rate $\text{imp} = \text{std}(\text{omega}(1:1000))$

**Output: State vector solution
for each time stamp**

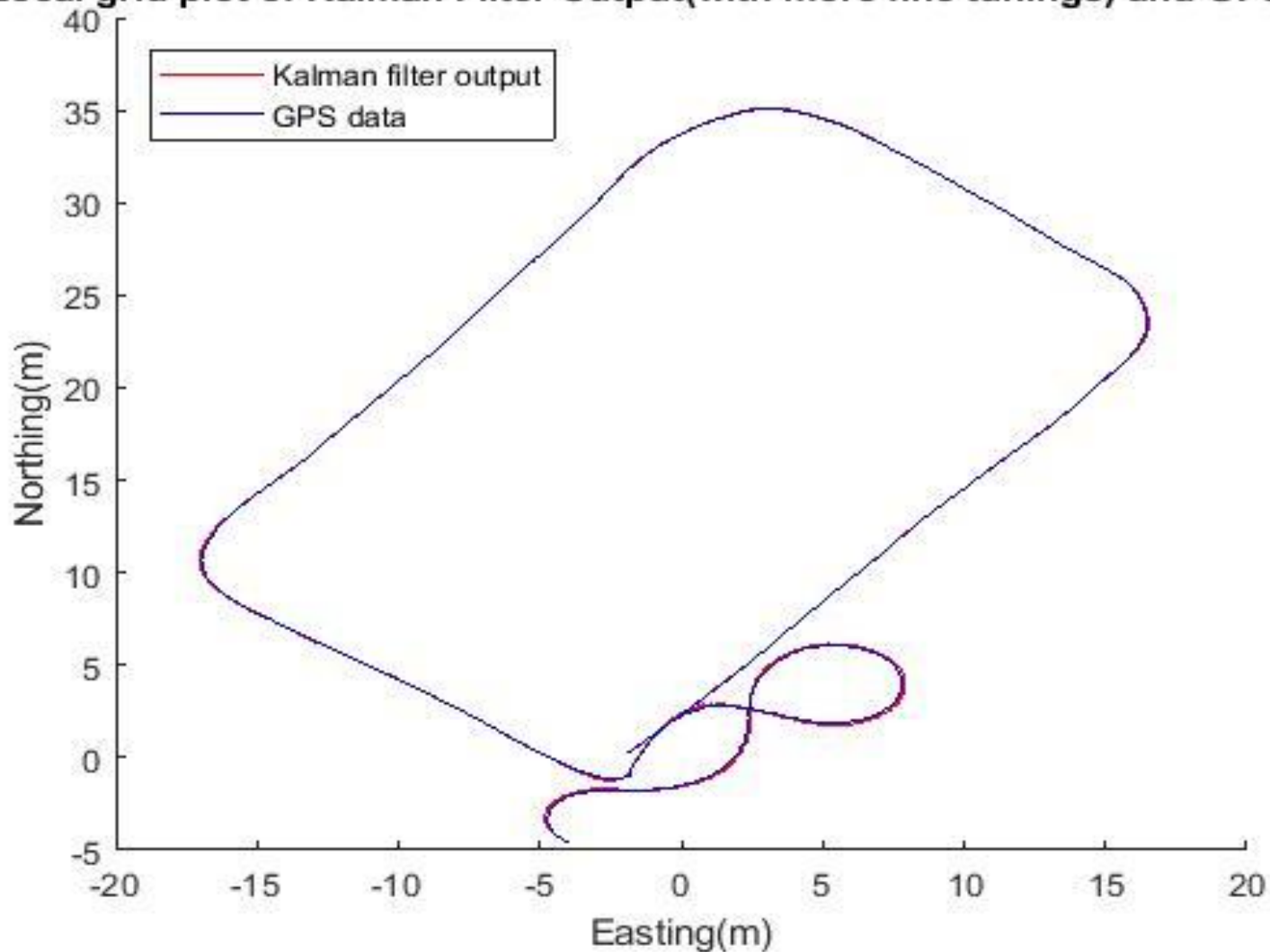
$$\begin{bmatrix} t \\ x_k \\ y_k \\ \varphi_k \\ \dot{\varphi}_k \\ v_k \\ a_k \end{bmatrix} = \begin{bmatrix} x_{k-1} + \cos(\varphi_{k-1})v_{k-1}\Delta t \\ y_{k-1} + \sin(\varphi_{k-1})v_{k-1}\Delta t \\ \varphi_{k-1} + \dot{\varphi}_{k-1}\Delta t \\ \dot{\varphi}_{k-1} + w_{\dot{\varphi}}\Delta t \\ v_{k-1} + a_{k-1}\Delta t \\ a_{k-1} + w_a\Delta t \end{bmatrix}$$

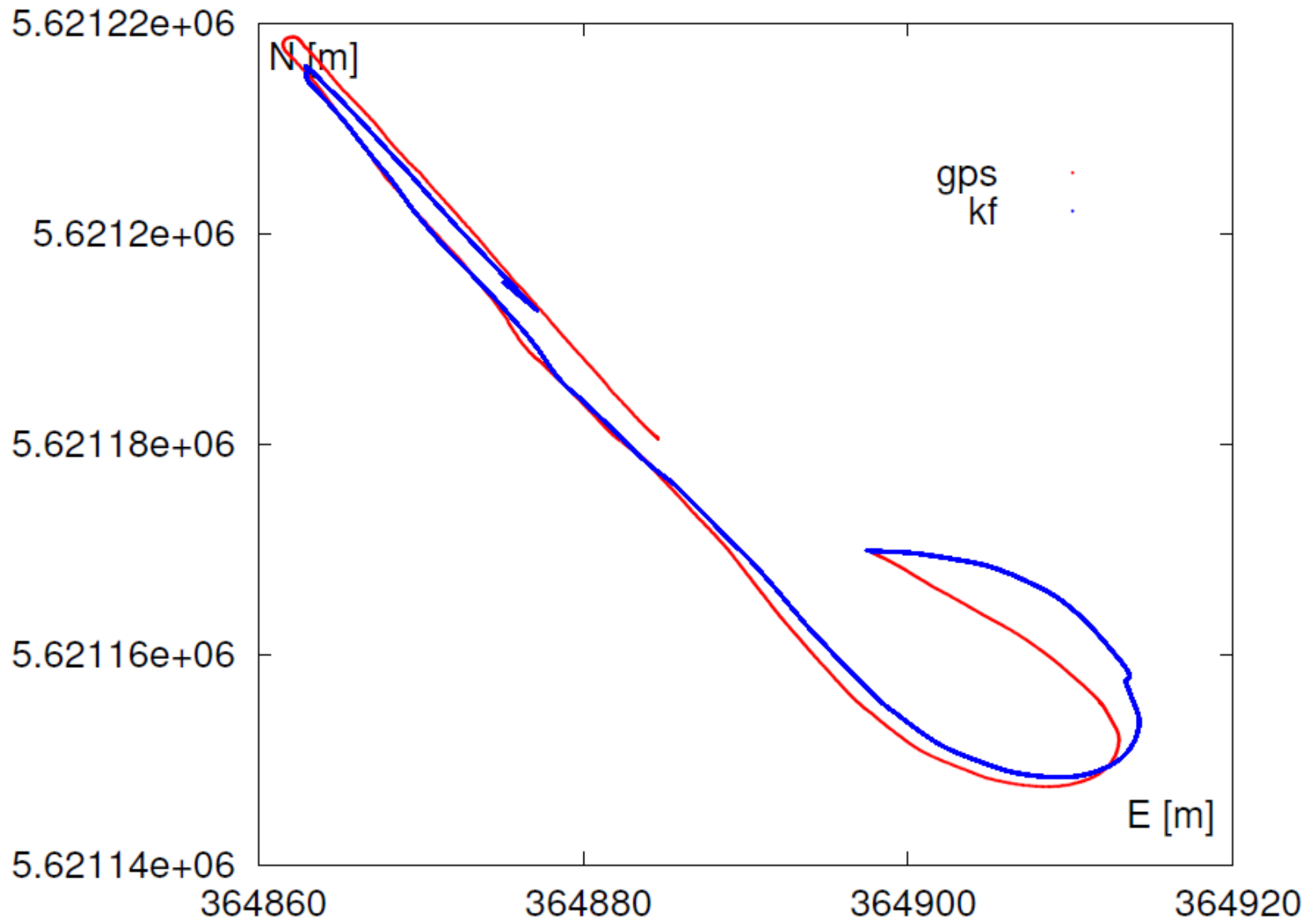
Solutions From previous years

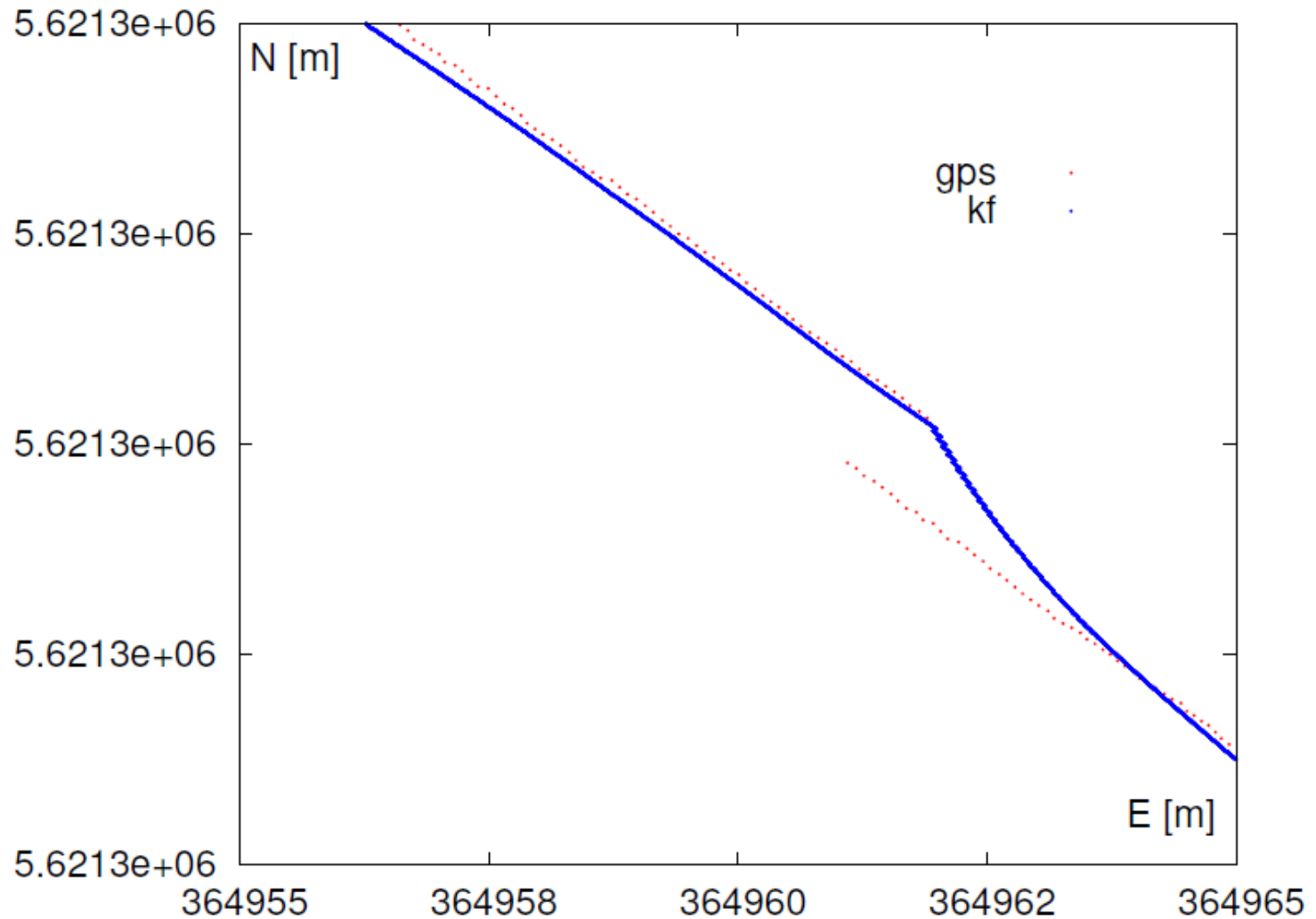


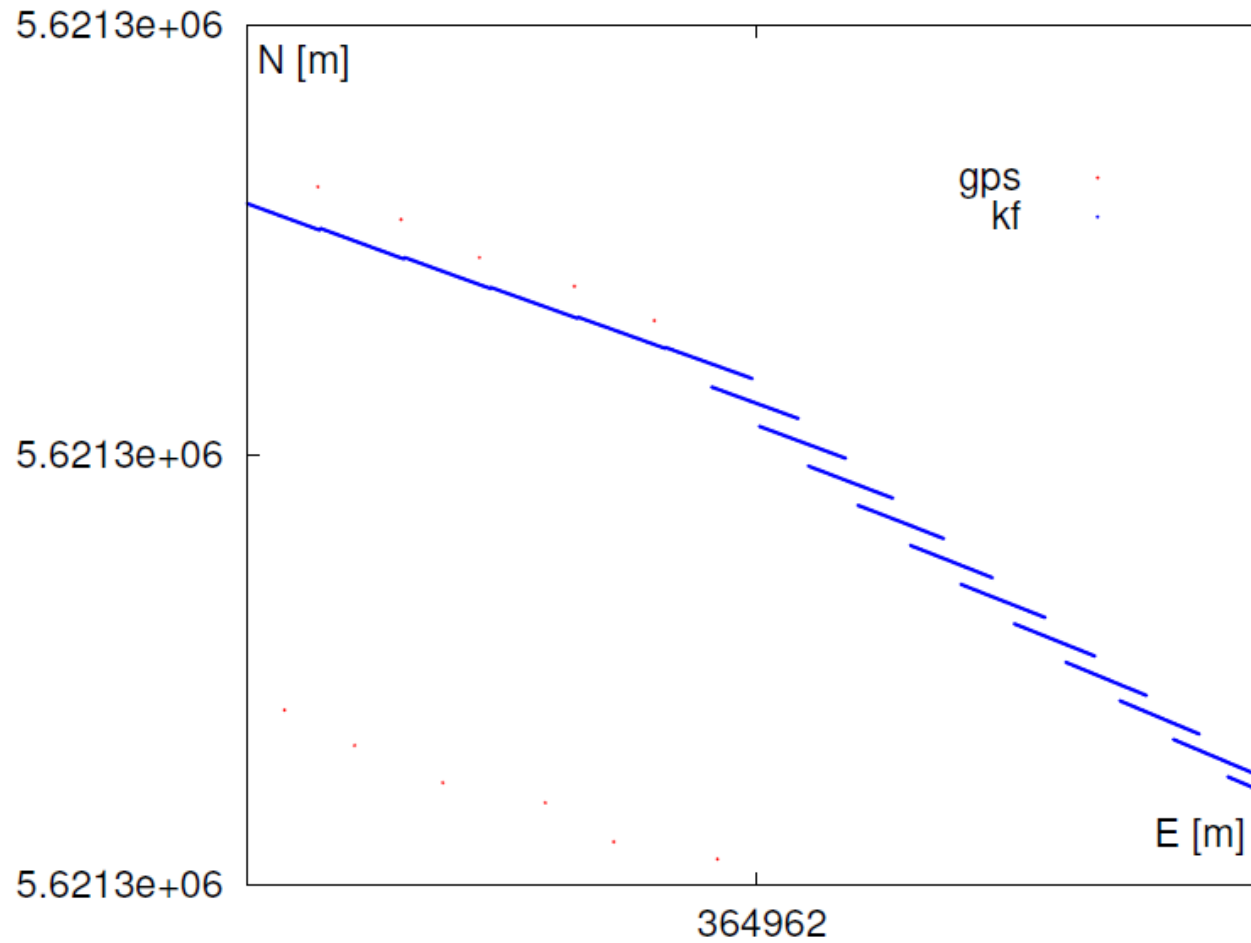


Local grid plot of Kalman Filter Output(with more fine tunings) and GPS data









Remarks for plots:

- Title
- Equally scaled axis
- Colour visible on the screen (not white & yellow)
- Axis titles (with readable measurement units)
- UTM (N,E) [m] -> reduce UTM to local coordinates for better readability(reduction to t0)
- Multiple plot views (e.g. overview & detailed view)
- Use in-built functions from previous exercise!

Thank you for your attention
Questions or comments?