

Vision Based Landing System for a VTOL-MAV

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Technical Aims

- Operation without legal restrictions
- Autonomous flight also in urban environments
- Teaming UAV/UAV and UAV/UGV
- Tracking and geo-localization of objects



GPS signals not always available



Augmentation of navigation system with image based system

Outline

AirQuad

Image based navigation estimation

Image based height estimation

Simulation environment

Results

Conclusion



Specifications

- Electrically powered
- Max. dimensions 92 cm
- Take-off weight 1000 g
- Payload capacity 200 g
- Operating time 25 min
- Max. altitude ~ 500 m
- Max. speed ~ 50 km/h
- Operating range > 5 km



Assumption

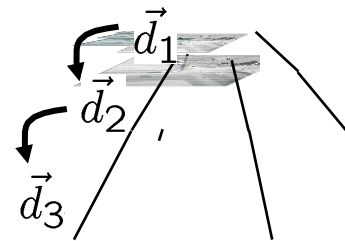
Augmentation of navigation system during hovering and landing situations

➔ Homographies suitable for motion estimation

Optical flow estimation

$$F_k(\vec{x}) \stackrel{!}{=} F_{k+1}(\vec{x} + \vec{d})$$

- Lucas-Kanade Algorithm



- Optical flow with census transform (based on Stein 2004, Zabih and Woodfill 1994) augmented with cross-correlation

Homography estimation and decomposition

- Estimation with RANSAC (RANDOM SAmple Consensus)

$$\vec{x}_{k+1} \sim \mathbf{H} \vec{x}_k$$

- Calibrated homography $\mathbf{H}_{cal} = \mathbf{K}^{-1} \mathbf{H} \mathbf{K}$

- Decomposition:

$$\mathbf{H}_{cal} = \pm \lambda \left(\mathbf{R} + \frac{1}{d} \vec{t} \vec{n}^T \right)$$

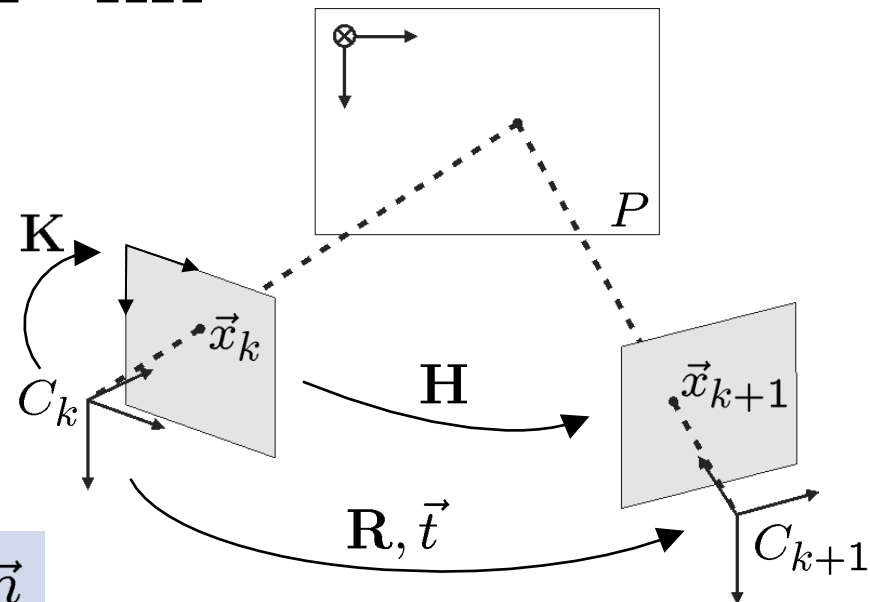
with \mathbf{R} rotation matrix

\vec{t} translation

d distance to scene plane P

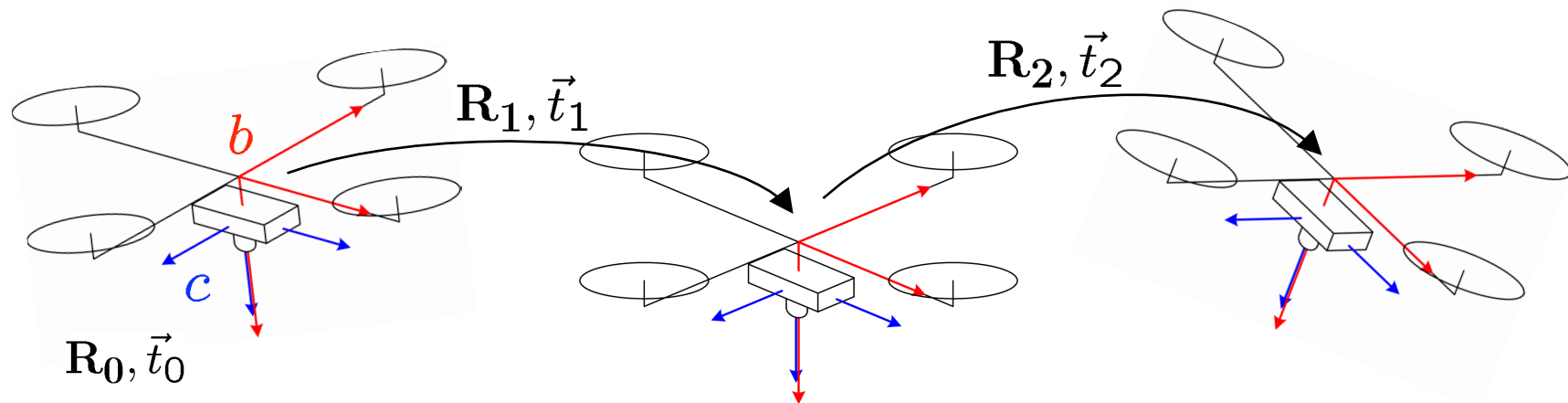
\vec{n} normal vector of P in
camera coordinates

➔ Decomposition into $\mathbf{R}, \frac{1}{d} \vec{t}, \vec{n}$



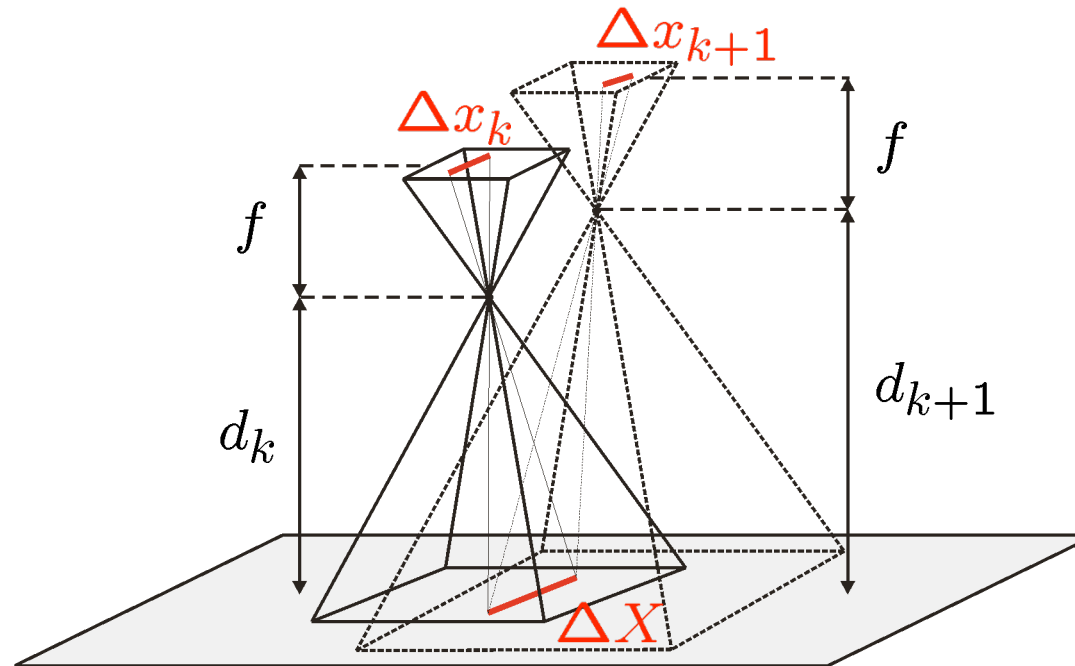
Propagation of attitude and position

- To be known
 - Initial position and attitude
 - Attitude $C_{b,c}$ between Camera and MAV
 - Distance to scene d , in this case height above ground



 Estimation of distance to scene necessary

Image based height above ground estimation



- With theorem of intersecting lines $\frac{d_k}{\Delta X} = \frac{f}{\Delta x_k}$
- From two different positions

$$\left. \begin{aligned} d_k &= f \Delta X \frac{1}{\Delta x_k} \\ d_{k+1} &= f \Delta X \frac{1}{\Delta x_{k+1}} \end{aligned} \right\} \xrightarrow{d_{k+1} - d_k} \Delta d_{k,k+1} = \quad d_{k+1} = \frac{\Delta d_{k,k+1}}{1 - \frac{\Delta x_{k+1}}{\Delta x_k}}$$

Estimation

$$d_{k+1} = \frac{\Delta d_{k,k+1}}{1 - \frac{\Delta x_{k+1}}{\Delta x_k}}$$

- $\Delta d_{k,k+1}$ from barometric pressure sensor
- $\Delta x_k, \Delta x_{k+1}$ from optical flow

Conditions

- Orientation of MAV and camera compensated
- Equation numerically well-conditioned

$$\left(1 - \frac{\Delta x_{k+1}}{\Delta x_k}\right) \neq 0 \quad \Rightarrow \quad \left|1 - \frac{\Delta x_{k+1}}{\Delta x_k}\right| > t$$

- Equivalent is motion in vertical direction:

$$\left(1 - \frac{\Delta x_{k+1}}{\Delta x_k}\right) \approx \left(\frac{\Delta d_{k,k+1}}{d_k}\right) \Rightarrow \left|\frac{\Delta d_{k,k+1}}{d_k}\right| > t_1$$

Conditions

- Motion in vertical direction: $\left| \frac{\Delta d_{k,k+1}}{d_k} \right| > t_1$
e. g. $t_1 \in [0.1 \dots 0.2]$
- Displacement not from noise: $\left| \Delta d_{k,k+1} \right| > t_2$
e. g. $t_2 = 3\sqrt{2} \sigma_{\text{baro}}$

Continuous estimation of height above ground

- with Kalman filter:
 $\Delta d_{k,k+1}$: known inputs, d_{k+1} : measurements
- with optical flow: $d_{k+1} = d_k \frac{\Delta x_k}{\Delta x_{k+1}}$

Outline

AirQuad

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Image based height estimation

Simulation environment

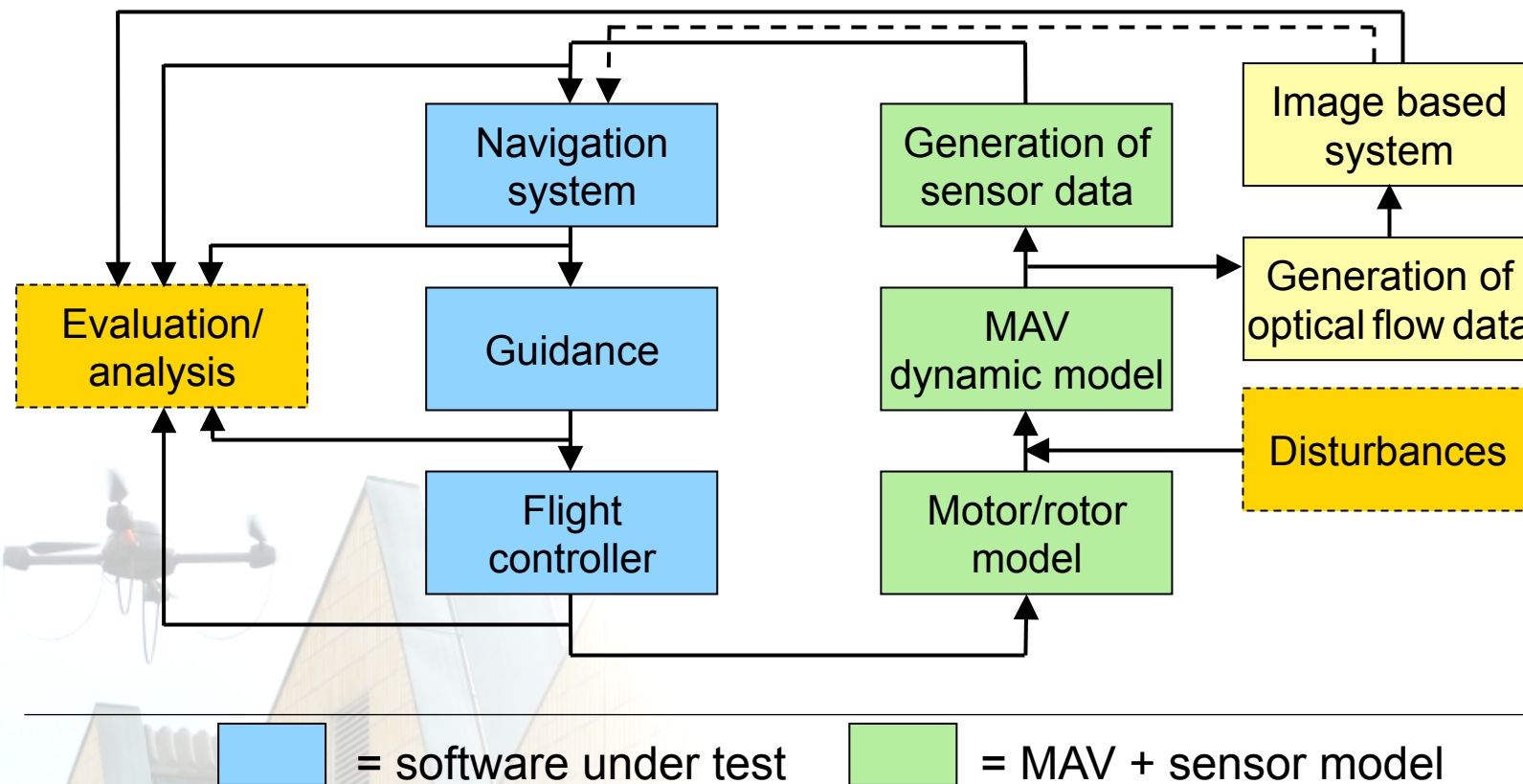
Results

Conclusion



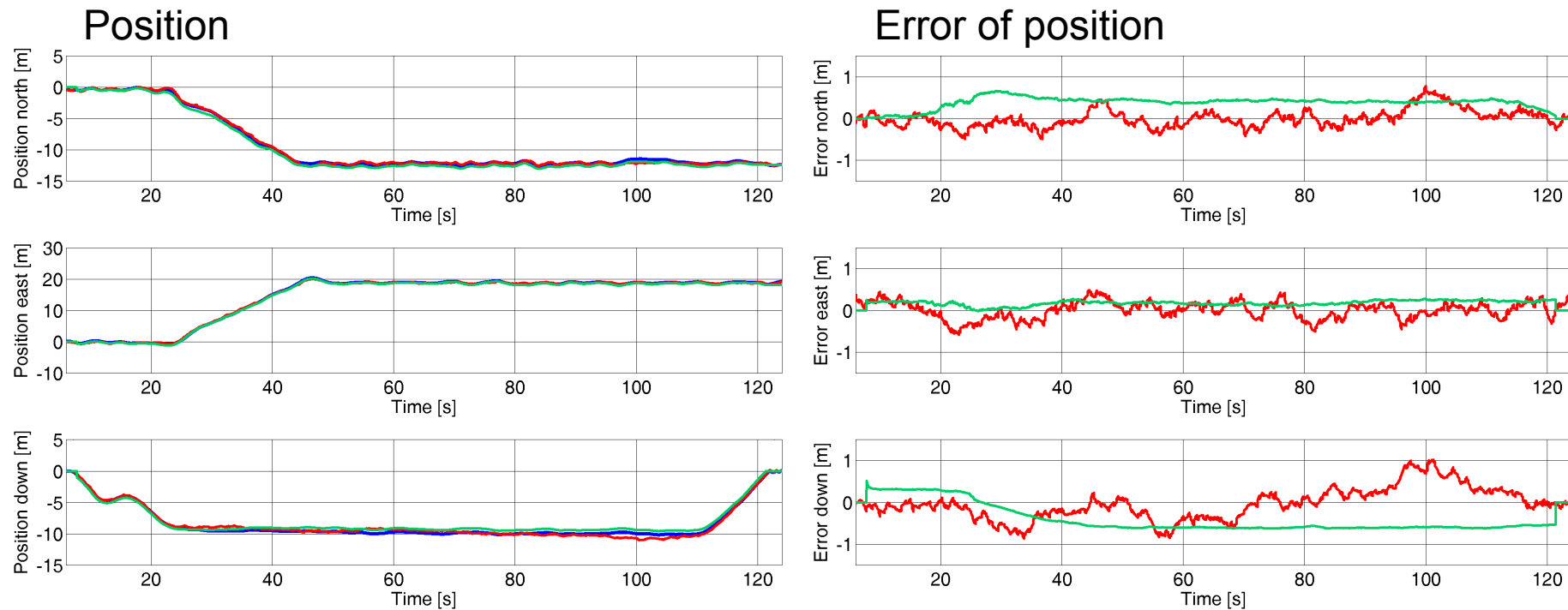
Simulation environment

- Essential for algorithm development and testing
- MAV model included
- Test of operational C-code



Results

1. Simulation: Hovering at defined position and landing

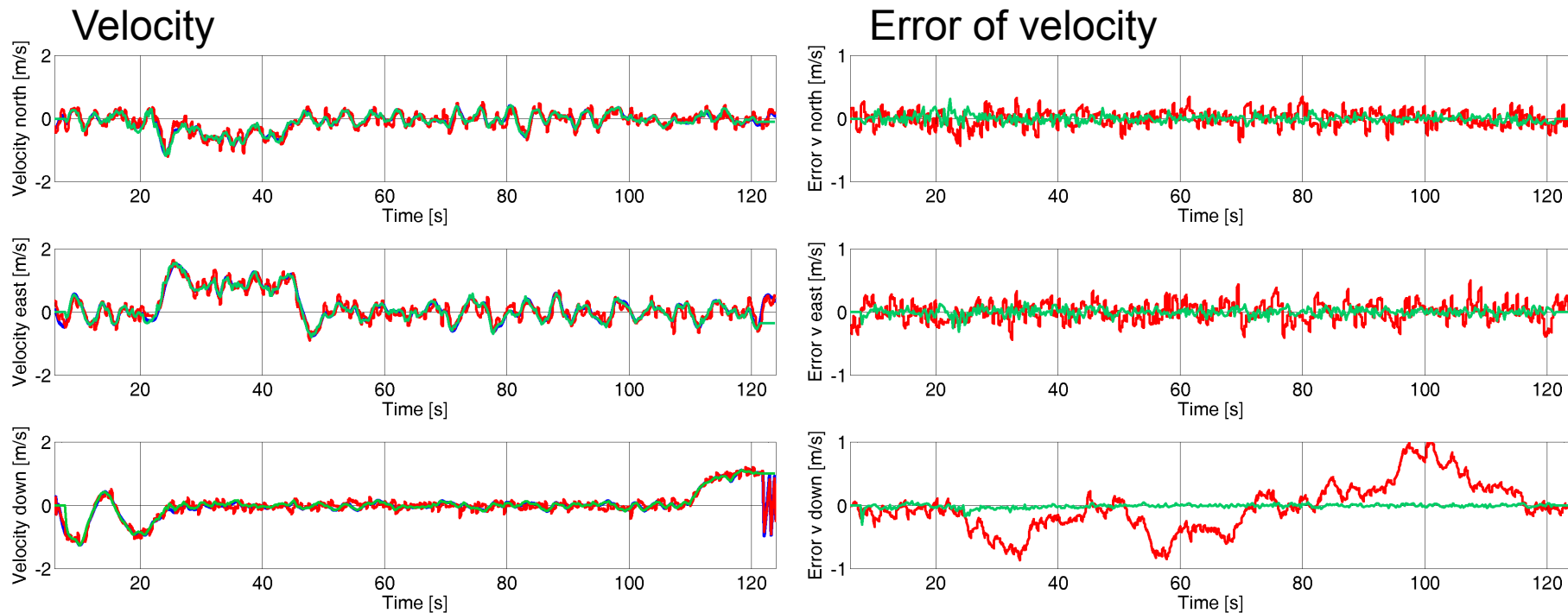


Frame rate 25 fps, image size 640x480, 200 features, feature noise $\sigma = \frac{1}{3}$ pix.

— Ground truth — GPS/INS/Mag/Baro — Vision

Results

Hovering at defined position and landing

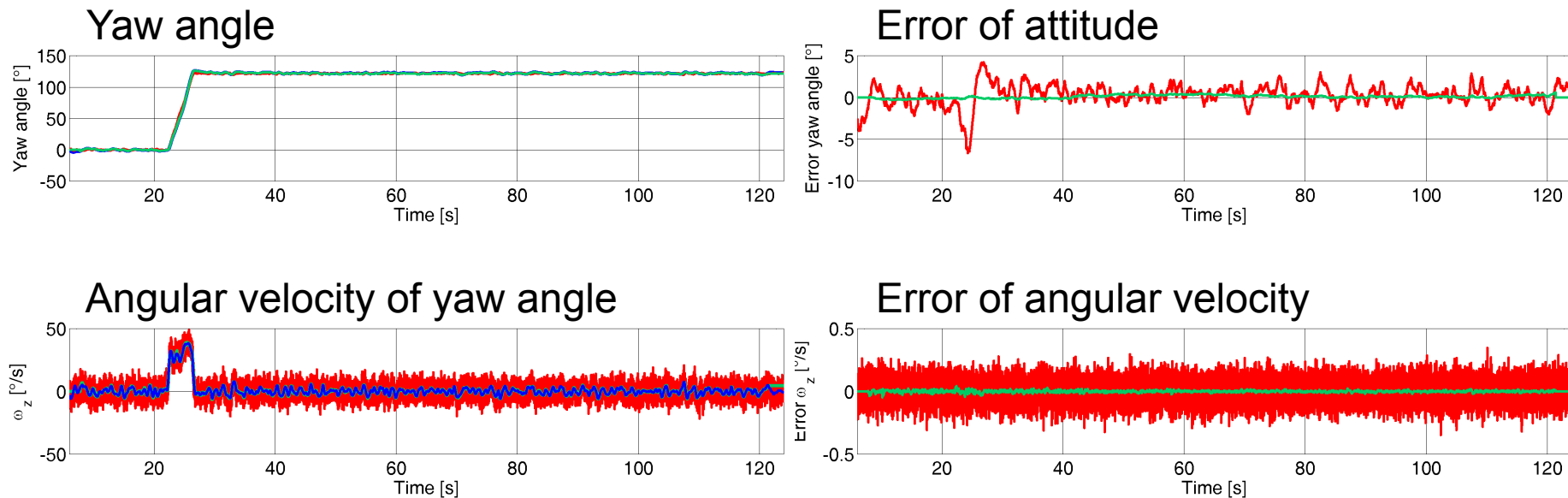


Positions divided by 1/25fps,
Averaging with $n = 6$, data rate 4.16Hz

— Ground truth — GPS/INS/Mag/Baro — Vision

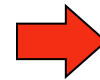
Results

Hovering at defined position and landing



Yaw angle is divided by 1/25fps

Magnetometer measurement can be corrupted by metallic surfaces.



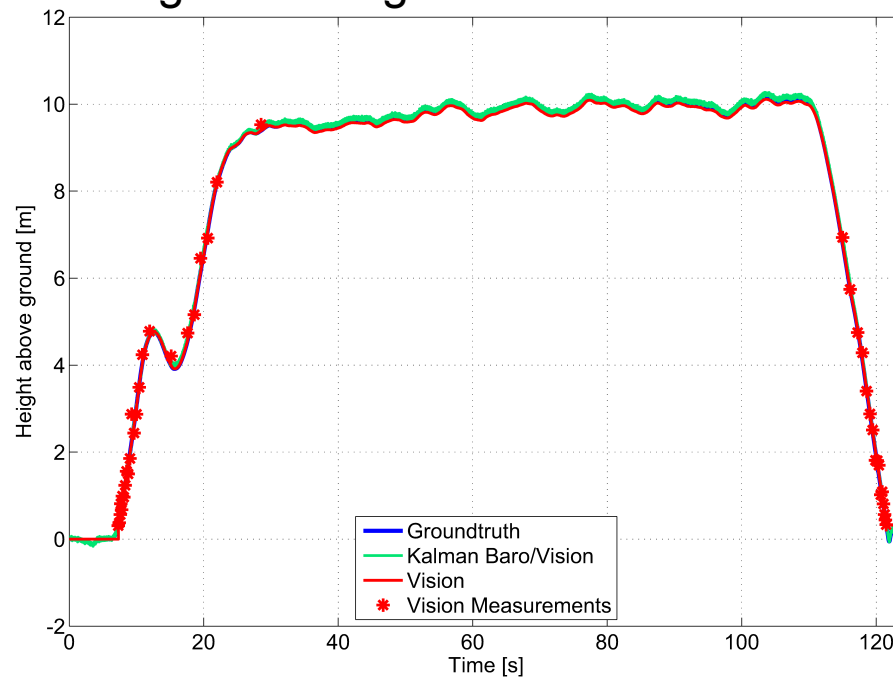
Improvement by vision system

— Ground truth — GPS/INS/Mag/Baro — Vision

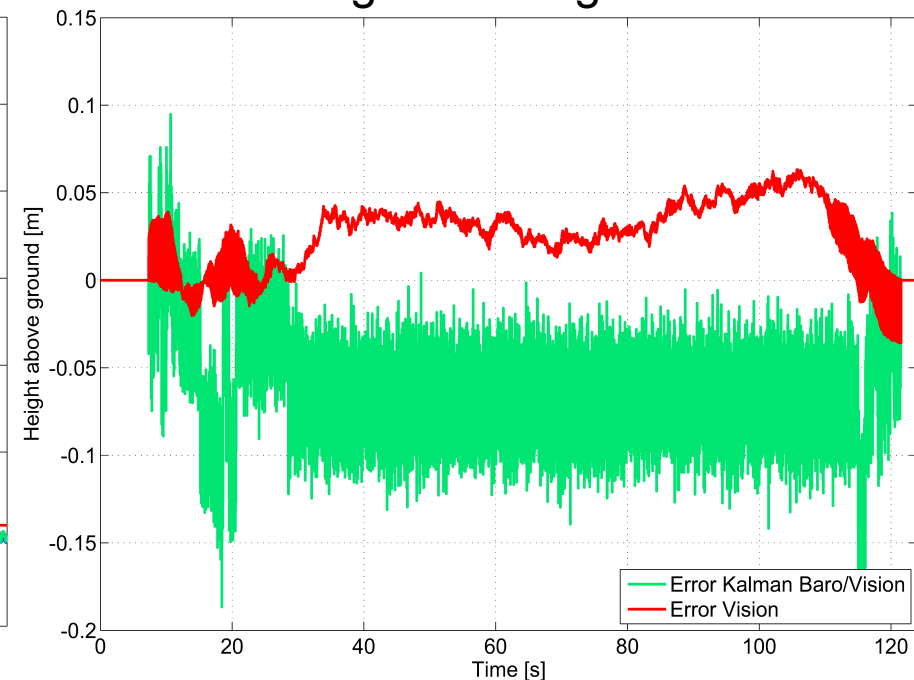
Results

Hovering at defined position and landing

Height above ground estimation



Error of height above ground



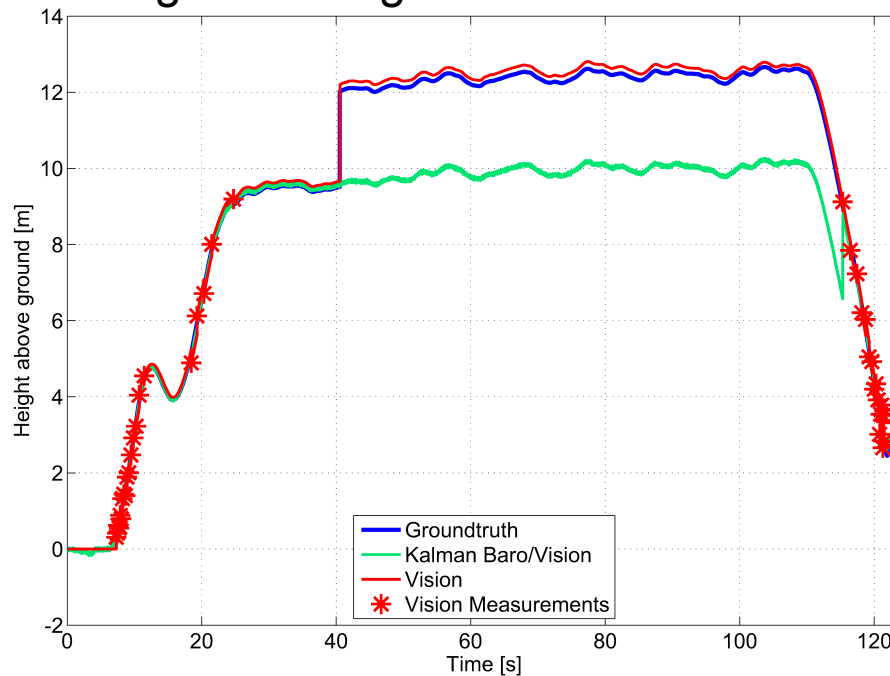
Baro rate 25Hz, baro offset -5m, baro noise $\sigma = 0.02$ m/min

m, baro drift 0.2

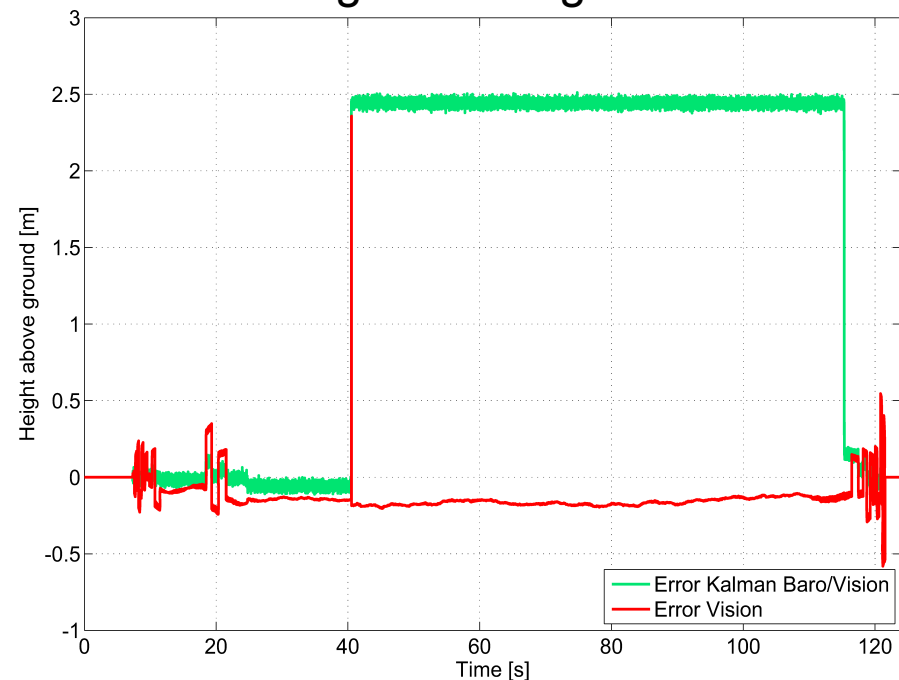
— Ground truth — Kalman filter Baro/Vision ● Vision meas. — Vision

Hovering at defined position and landing

Height above ground estimation



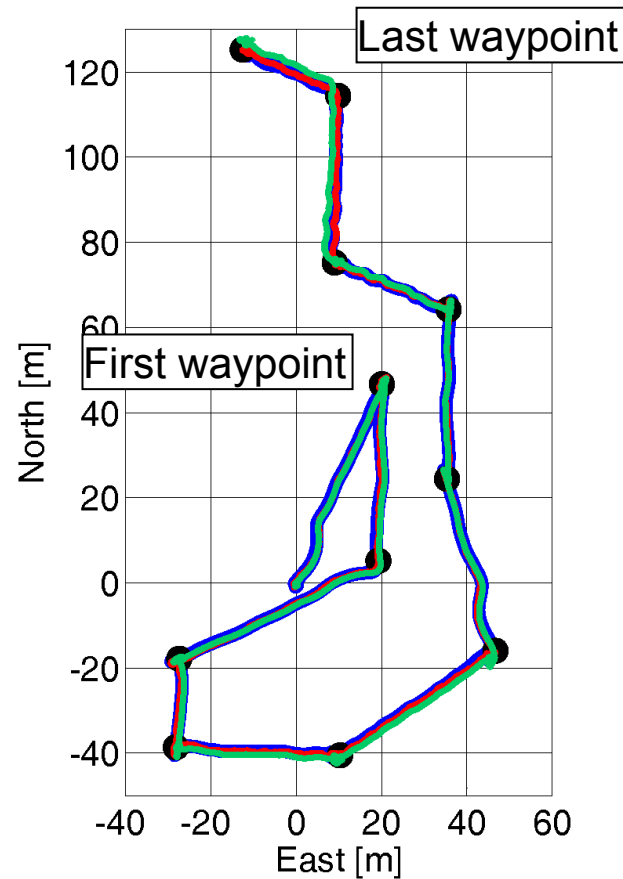
Error of height above ground



Simulated step of ground elevation of 2.5 m

— Ground truth — Kalman filter Baro/Vision ● Vision meas. — Vision

2. Simulation: Waypoint flight



Waypoint flight

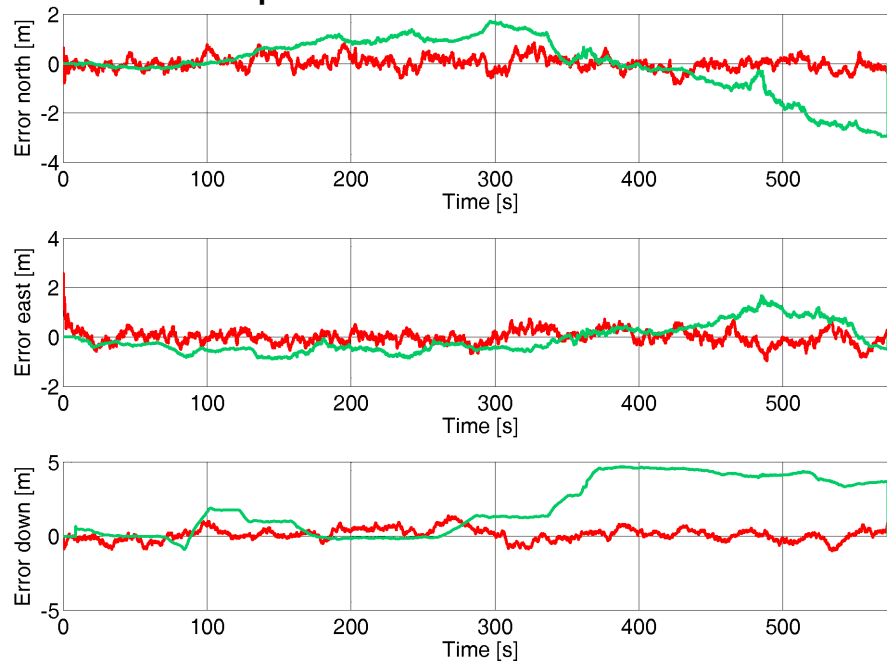
- 11 waypoints
- Hover-and-stare points
- ~ 10 min
- Height up to 30 m

— Ground truth — GPS/INS/Mag/Baro — Vision

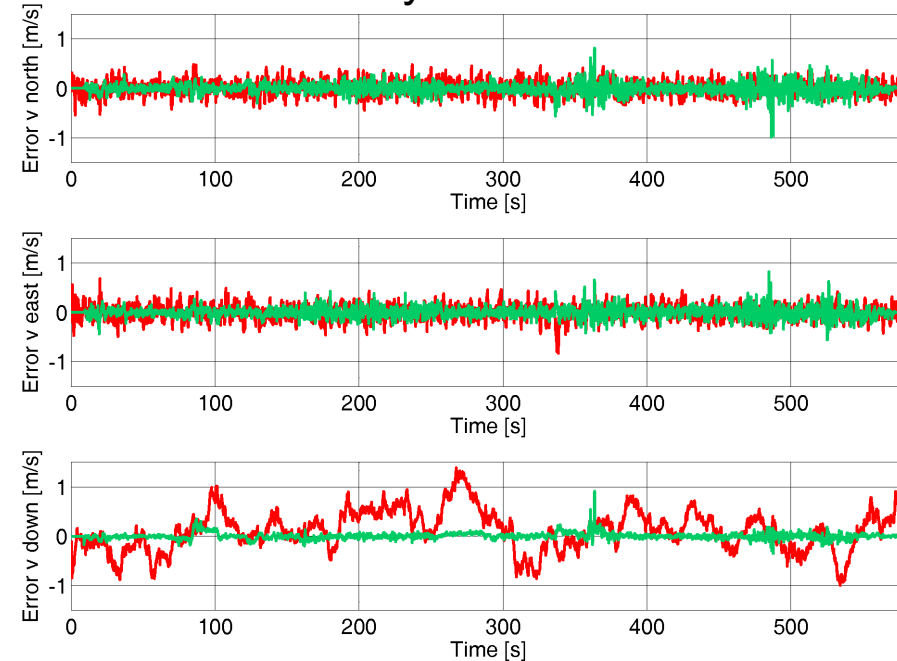
Results

Waypoint flight

Error of position



Error of velocity



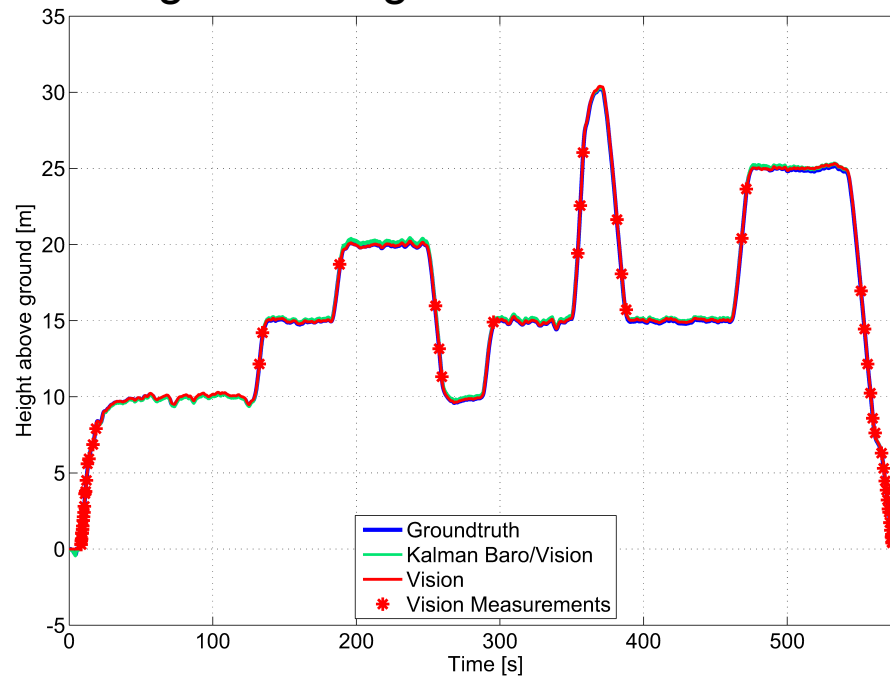
Frame rate 25 fps, image size 640x480, 200 features, feature noise $\sigma = \frac{1}{3}$ pix.

— GPS/INS/Mag/Baro — Vision

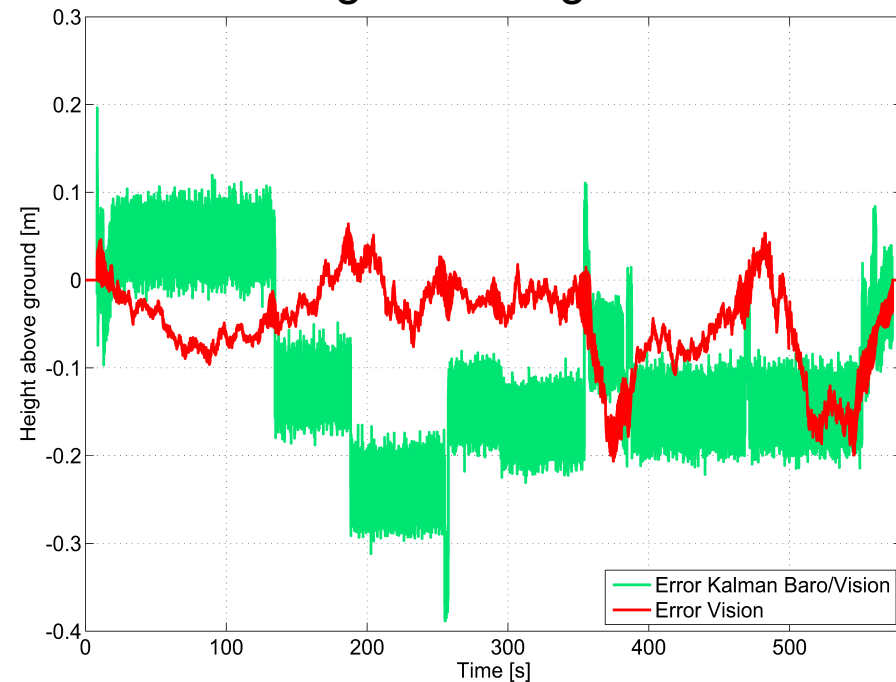
Results

Waypoint flight

Height above ground estimation



Error of height above ground



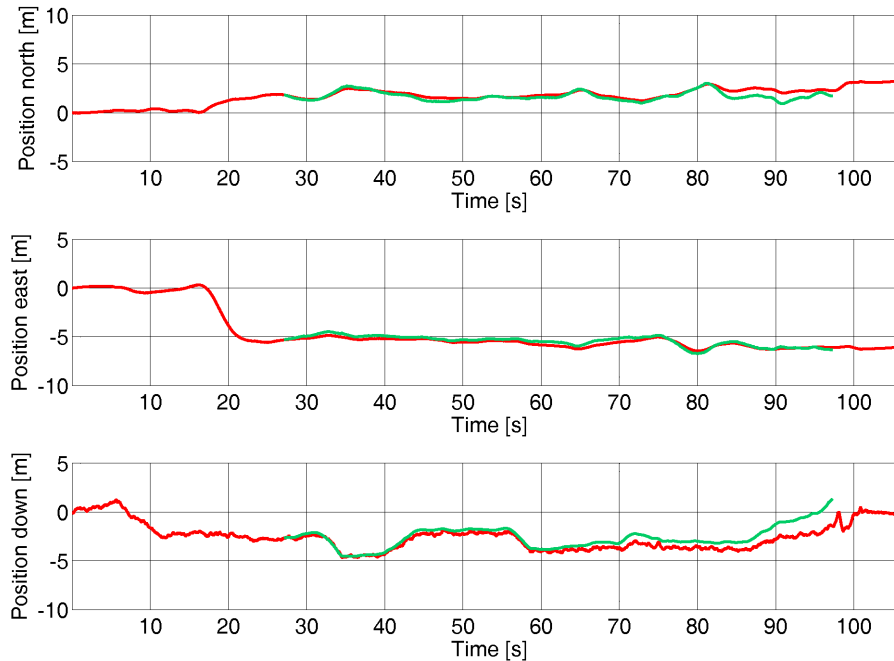
Baro rate 25Hz, baro offset -5m, baro noise $\sigma = 0.02$ m/min

m, baro drift 0.2

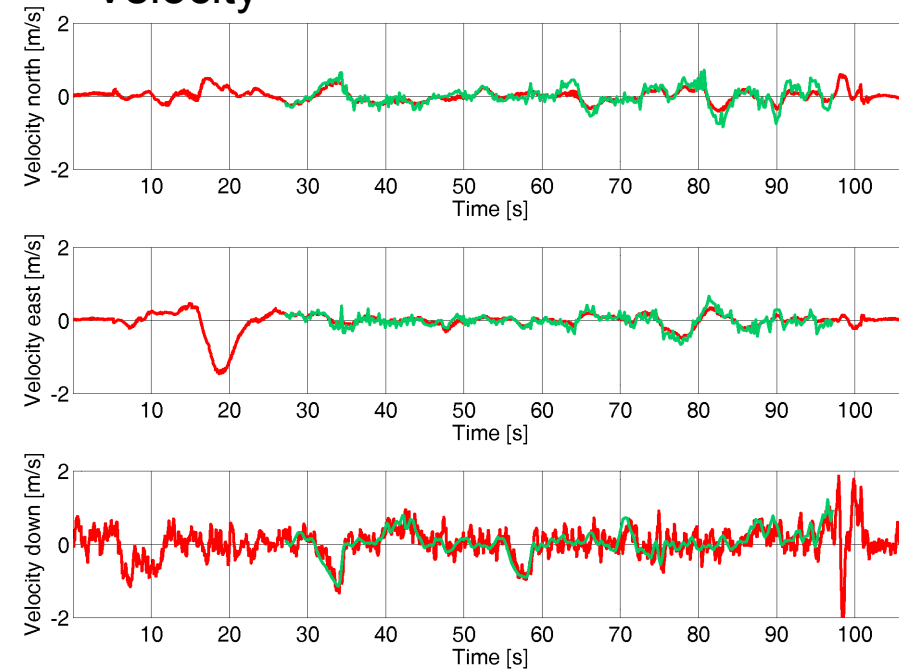
— Ground truth — Kalman filter Baro/Vision ● Vision meas. — Vision

3. Processing of in-flight data

Position



Velocity



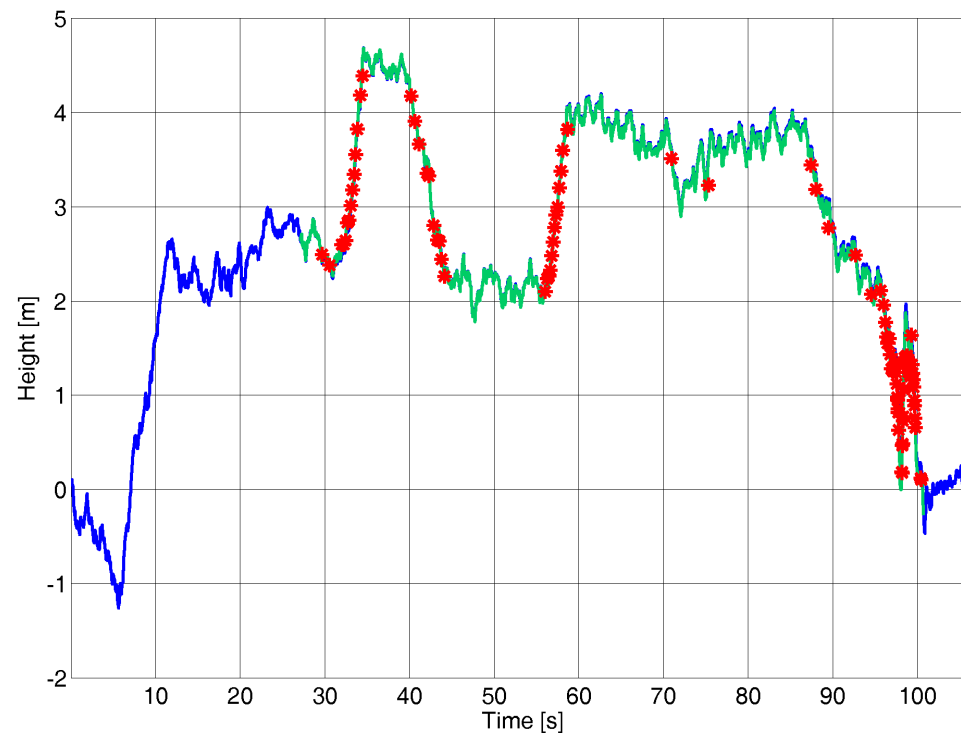
First results on processing of in-flight data

Positions divided by $1/30\text{fps}$,
Averaging with $n = 6$, data rate 5Hz

— GPS/INS/Mag/Baro — Vision

Processing of in-flight data

Height above ground estimation



First tests with
in-flight data confirm
results of simulations



Augmentation of
navigation system
possible

— Barometric sensor data — Kalman filter Baro/Vision ● Vision

Conclusion

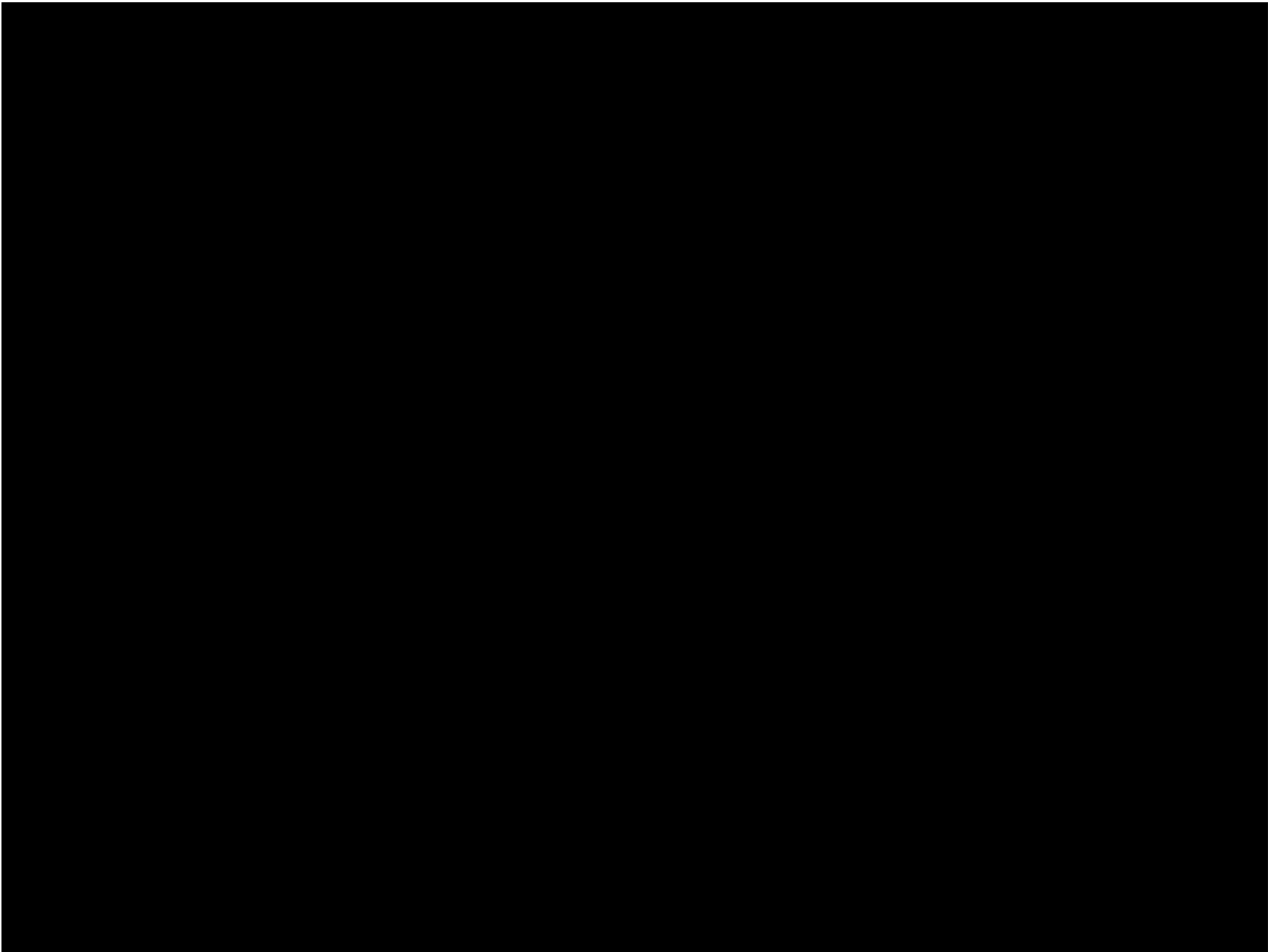
- ① Image based navigation aiding based on homographies in cases of
 - + hovering and
 - + landing
- ① Height above ground estimation solely with
 - + optical flow and
 - + barometric sensor data

Future Work

- Integration in navigation and guidance modules
- Implementation of algorithms on on-board image processing hardware

Thank you for your attention.

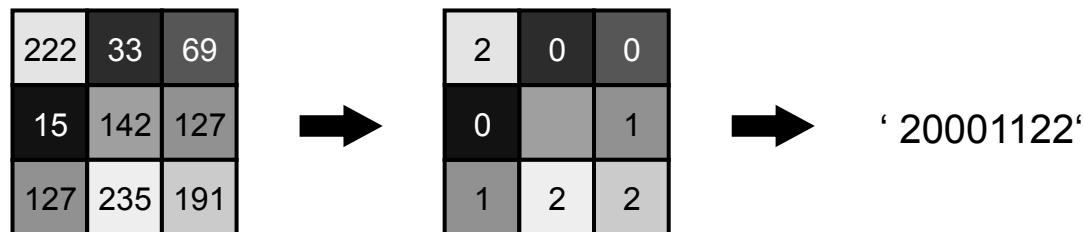




Optical flow with census transform

- Comparison of gray values in neighborhood W

$$\xi_i = \begin{cases} 0 & , F(\vec{x}) - F(\vec{x}_i) > \sigma \\ 1 & , |F(\vec{x}) - F(\vec{x}_i)| \leq \sigma \\ 2 & , F(\vec{x}) - F(\vec{x}_i) < -\sigma \end{cases} \quad \text{with } \vec{x}_i \in W$$



- Conversion of signature vector $\vec{\xi}$ to decimal integer
- Store points according to signature vector in table
- Correspondences \vec{x}_k, \vec{x}_{k+1} between images by comparing tables

Optical flow with census transform

- Use neighbors in distance $\delta > 1$, e. g. $\delta = 4$
- Filtering of signatures of one image
 - Use only signatures including useful information
e. g. reject '11111111'
 - Use only infrequent signatures e. g. less than 5 times in the image
- Filtering of correspondences
 - Hamming-Distance of 0
 - Distance between points not too large e. g. less than 50 pixels
 - Gray values of pixels similar
e. g. less than 20% deviation

 result not robust

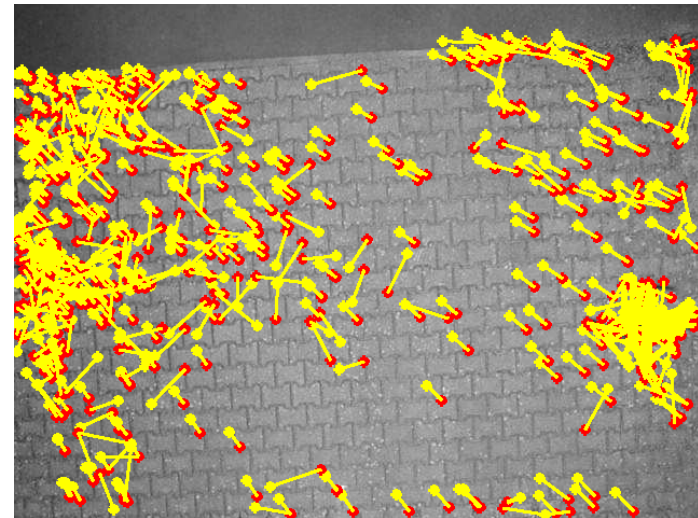
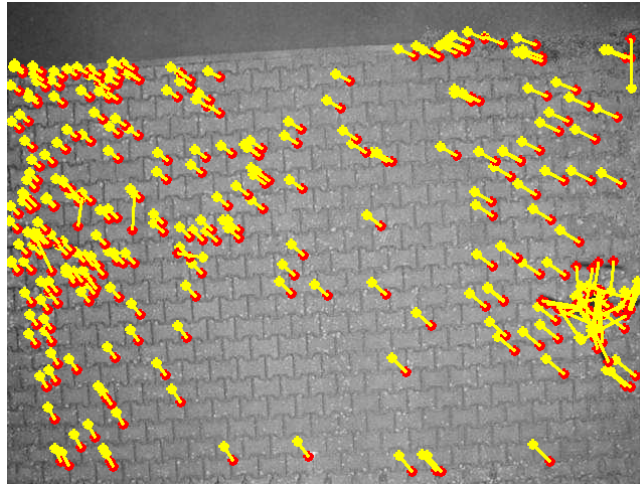


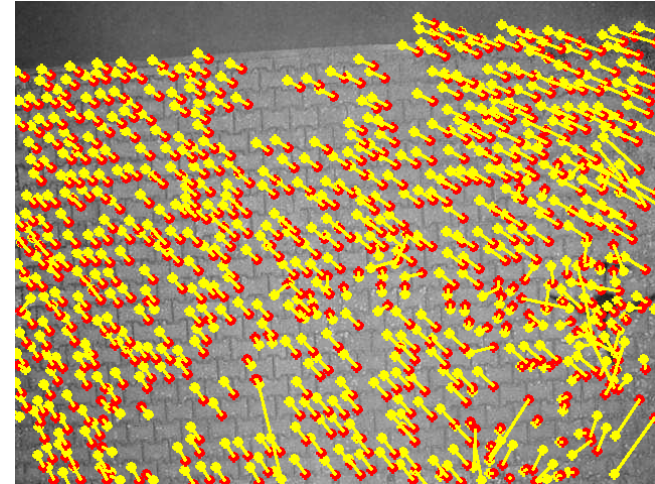
Image based navigation estimation

Results of optical flow calculation

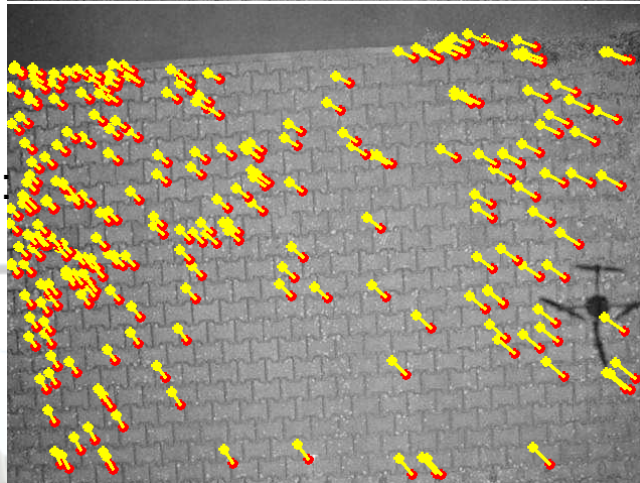
Census:



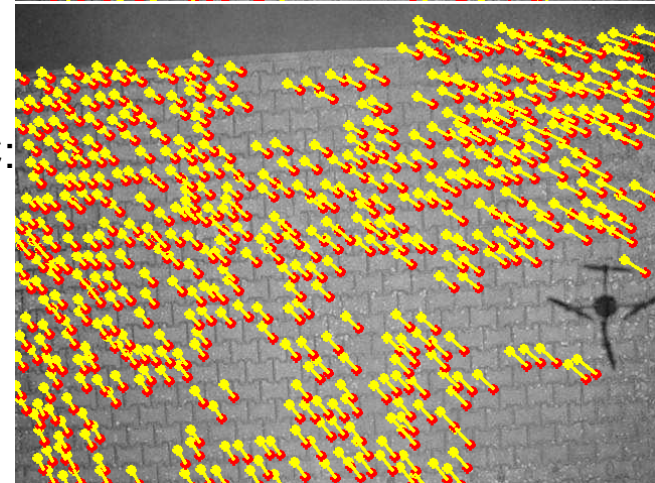
LK:



Census
after
RANSAC:



LK after
RANSAC:



Homography estimation and decomposition

- Estimation with RANSAC (RANDOM SAmple Consensus)

$$\vec{x}_{k+1} \sim \mathbf{H} \vec{x}_k$$

- Calibrated homography $\mathbf{H}_{cal} = \mathbf{K}^{-1} \mathbf{H} \mathbf{K}$

- Decomposition:

$$\mathbf{H}_{cal} = \pm \lambda \left(\mathbf{R} + \frac{1}{d} \vec{t} \vec{n}^T \right)$$

with \mathbf{R} rotation matrix

\vec{t} translation

d distance to scene plane P

\vec{n} normal vector of P in camera coordinates

- Rotation $\mathbf{R} = \mathbf{C}_{c_{k+1}, c_k}$
- Sign by $\vec{x}_{k+1}^T \mathbf{K} \mathbf{H}_{cal} \mathbf{K}^{-1} \vec{x}_k > 0$
- Singular value decomposition gives 2 physically possible solutions

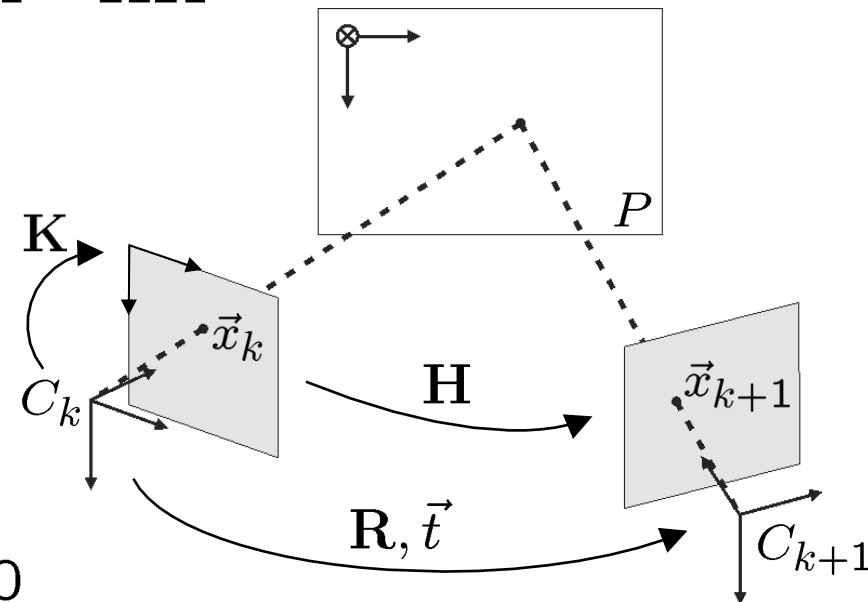


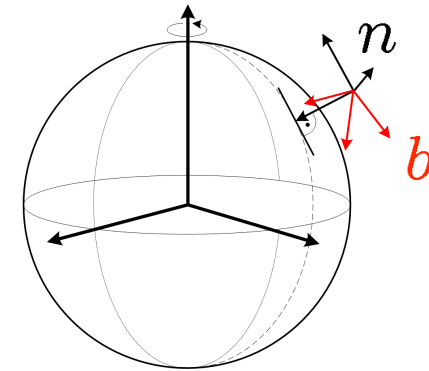
Image based navigation estimation

Propagation of attitude and position

- Integration in navigation coordinate system n

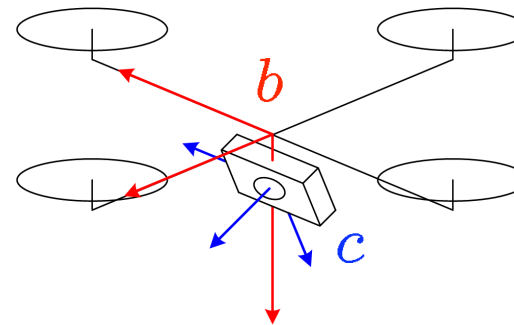
$$\mathbf{C}_{n,b_{k+1}} = \mathbf{C}_{n,b_k} \mathbf{C}_{b,c} \mathbf{R}^T \mathbf{C}_{c,b}$$

$$\vec{x}_{\text{MAV},k+1}^n = \vec{x}_{\text{MAV},k}^n - \mathbf{C}_{n,b_k} \mathbf{C}_{b,c} \mathbf{R}^T \vec{t}$$



with \mathbf{R} and \vec{t} from images

- Camera fixed on MAV: $\mathbf{C}_{b,c} = \text{const}$, centers coincide



$$d_{k+1} = \frac{\Delta d_{k,k+1}}{1 - \frac{\Delta x_{k+1}}{\Delta x_k}}$$

Conditions

- Orientation of MAV and camera compensated

$$\begin{aligned}\vec{x}_{k,corr} &= \mathbf{K} \hat{\mathbf{C}}_{c,b} \hat{\mathbf{C}}_{b,n} \mathbf{C}_{n,b_k} \mathbf{C}_{b,c} \mathbf{K}^{-1} \vec{x}_k \\ \vec{x}_{k+n,corr} &= \mathbf{K} \hat{\mathbf{C}}_{c,b} \hat{\mathbf{C}}_{b,n} \mathbf{C}_{n,b_{k+n}} \mathbf{C}_{b,c} \mathbf{K}^{-1} \vec{x}_{k+n} \\ \hat{\mathbf{C}}_{b,n} &= \mathbf{I}, \quad \hat{\mathbf{C}}_{b,c} : \hat{\phi}_{bc} = 0^\circ, \hat{\theta}_{bc} = 0^\circ, \hat{\psi}_{bc} = -90^\circ\end{aligned}$$

- Motion in vertical direction:

e. g. $t_1 \in [0.1 \dots 0.2]$

$$\left| \frac{\Delta d_{k,k+1}}{d_k} \right| > t_1$$

- Displacement not from noise:

e. g. $t_2 = 3\sqrt{2} \sigma_{\text{baro}}$

$$|\Delta d_{k,k+1}| > t_2$$

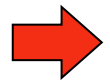
Estimation

$$d_{k+1} = \frac{\Delta d_{k,k+1}}{1 - \frac{\Delta x_{k+1}}{\Delta x_k}}$$

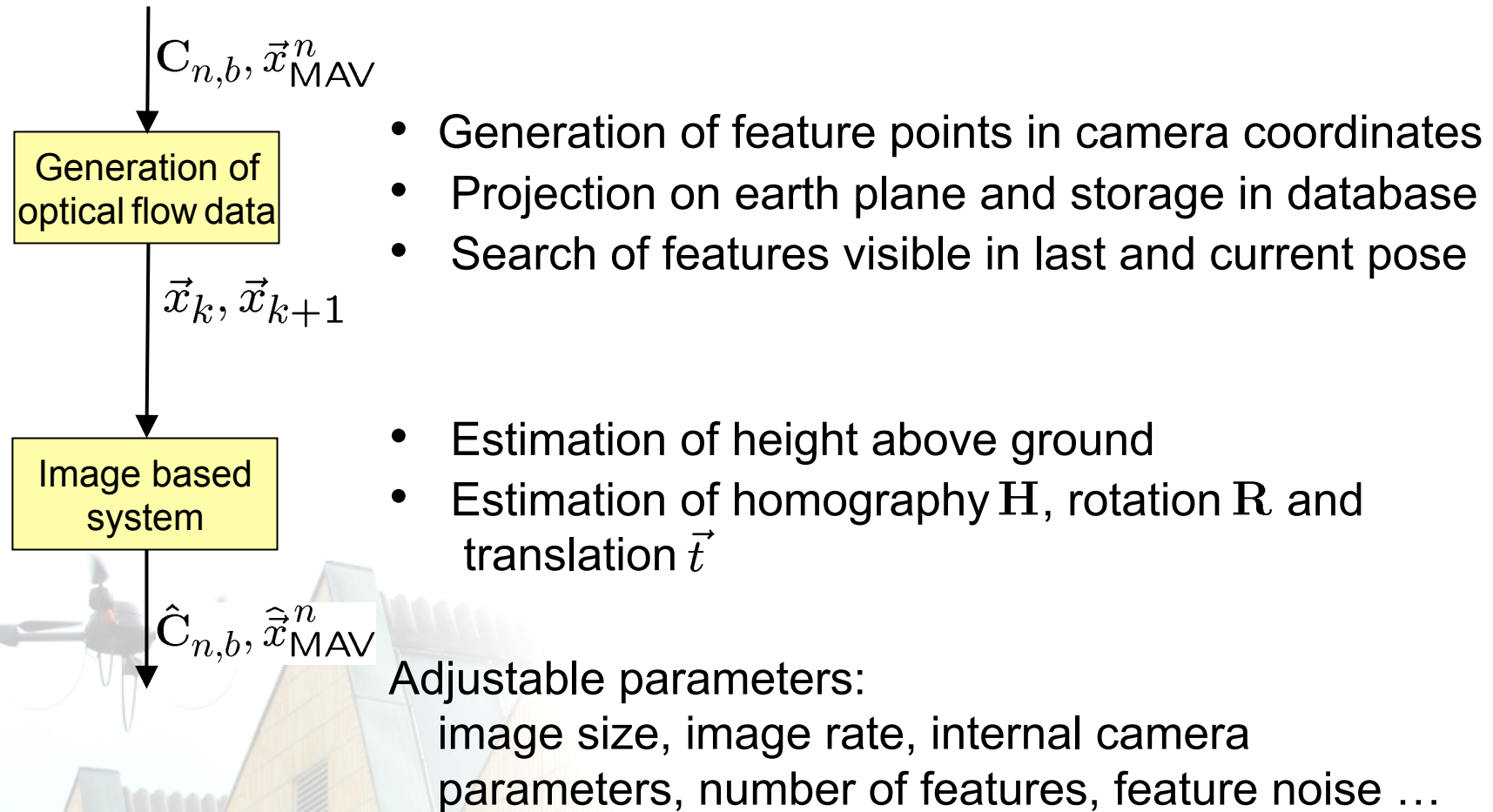
- $\Delta d_{k,k+1}$ from barometric pressure sensor
- $\Delta x_k, \Delta x_{k+1}$ from optical flow

Conditions

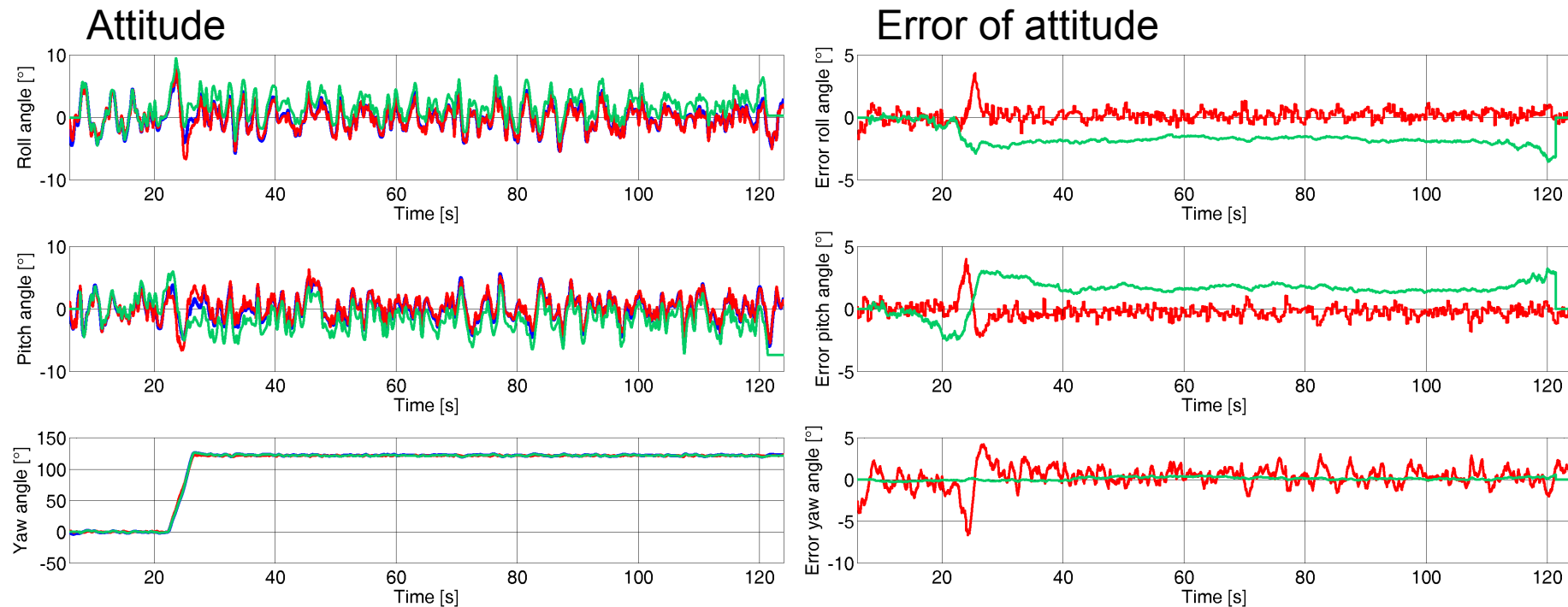
- Motion in vertical direction: $\left| \frac{\Delta d_{k,k+1}}{d_k} \right| > t_1$
e. g. $t_1 \in [0.1 \dots 0.2]$
- Displacement not from noise: $\left| \Delta d_{k,k+1} \right| > t_2$
e. g. $t_2 = 3\sqrt{2} \sigma_{\text{baro}}$

 Continuous estimation of height above ground with Kalman filter
 $\Delta d_{k,k+1}$: known inputs, d_{k+1} measurements

Simulation of vision system



1. Simulation: Hovering at defined position and landing

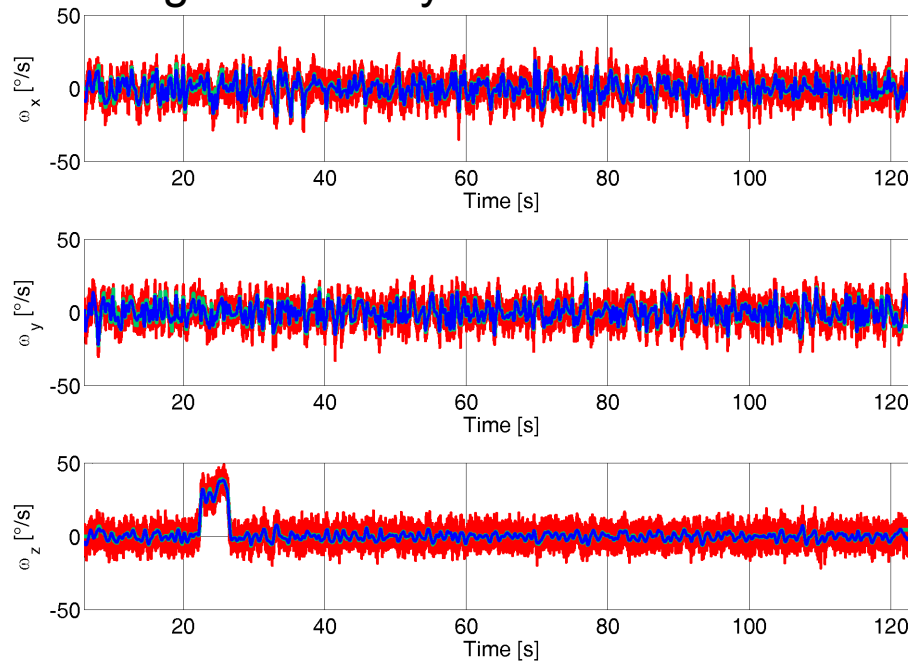


— Ground truth — GPS/INS/Mag/Baro — Vision

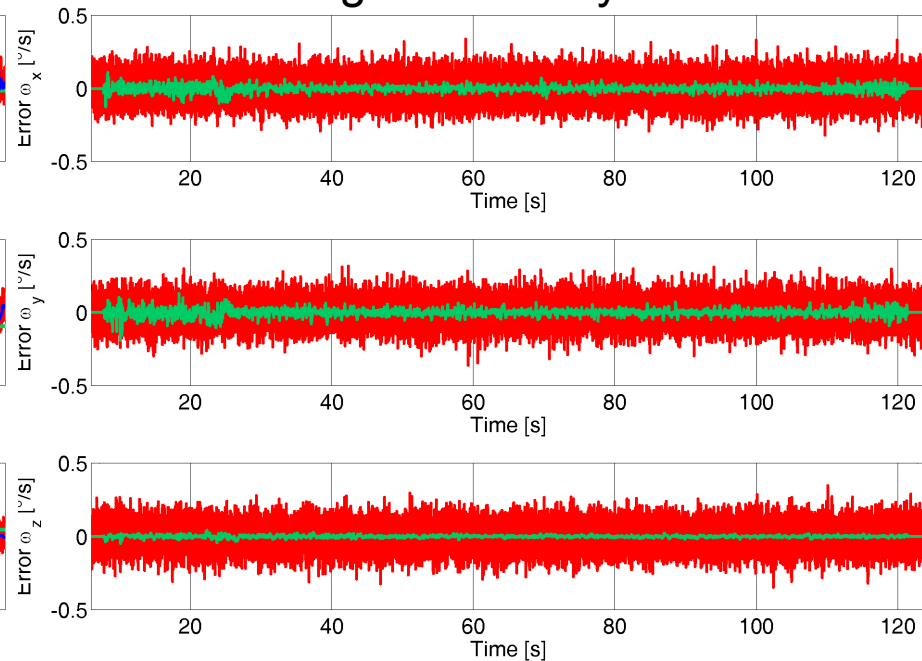
Results

Hovering at defined position and landing

Angular velocity



Error of angular velocity

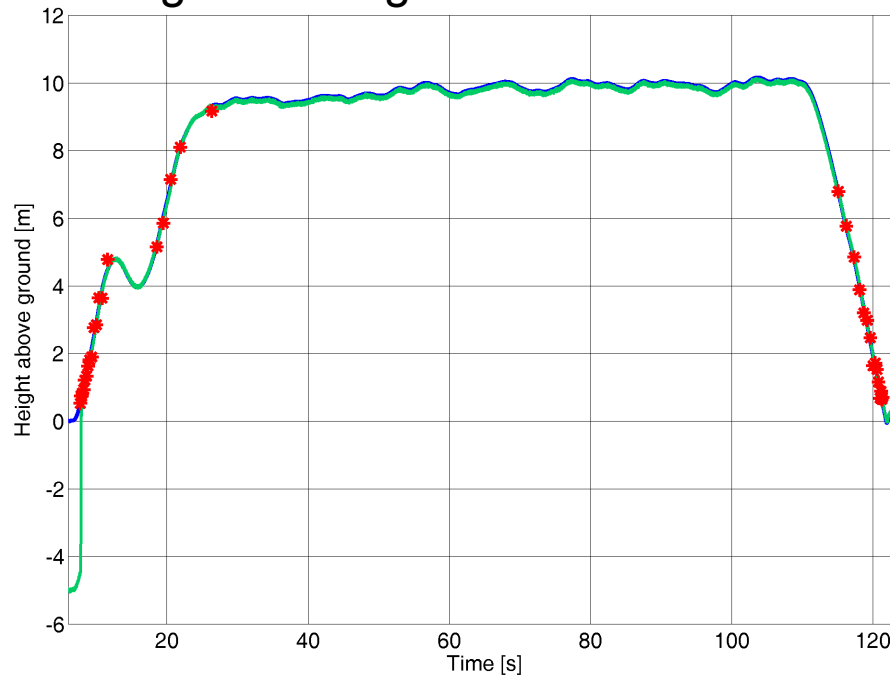


Angles divided by 1/25fps

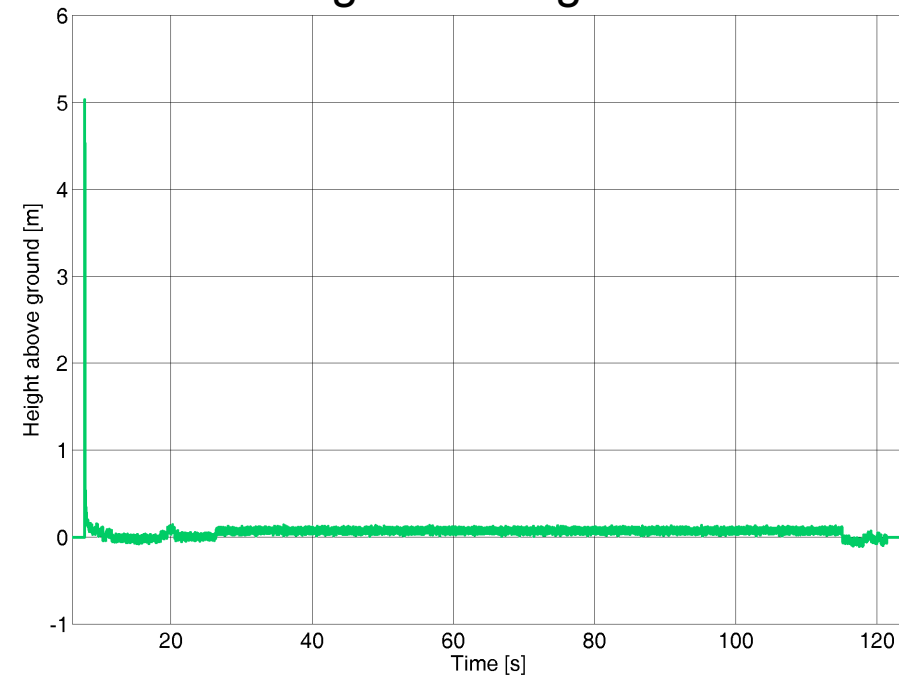
— Ground truth — GPS/INS/Mag/Baro — Vision

Hovering at defined position and landing

Height above ground estimation



Error of height above ground



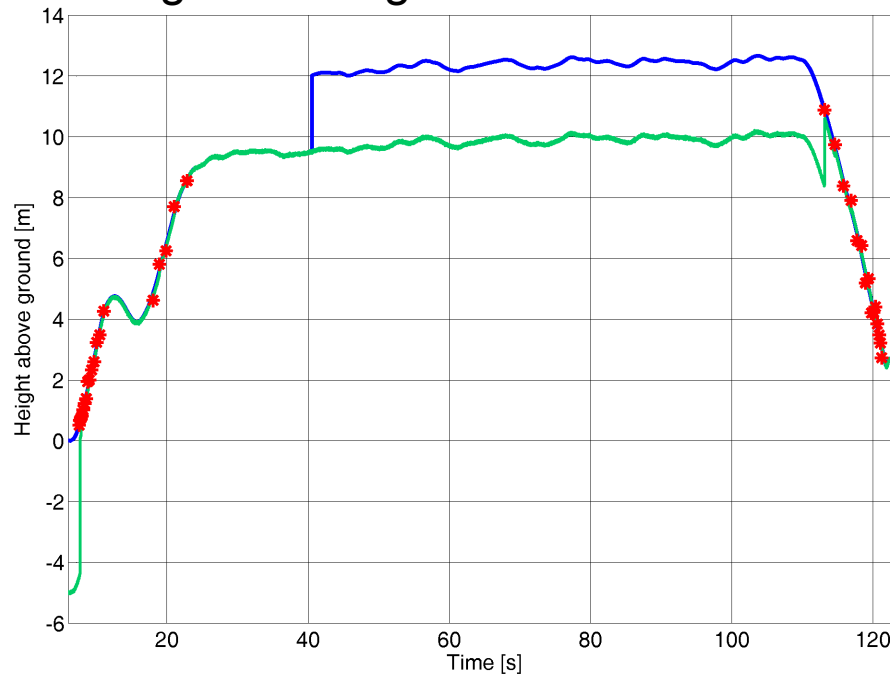
Baro rate 25Hz, baro offset -5m, baro noise $\sigma = 0.02$ m/min

m, baro drift 0.2

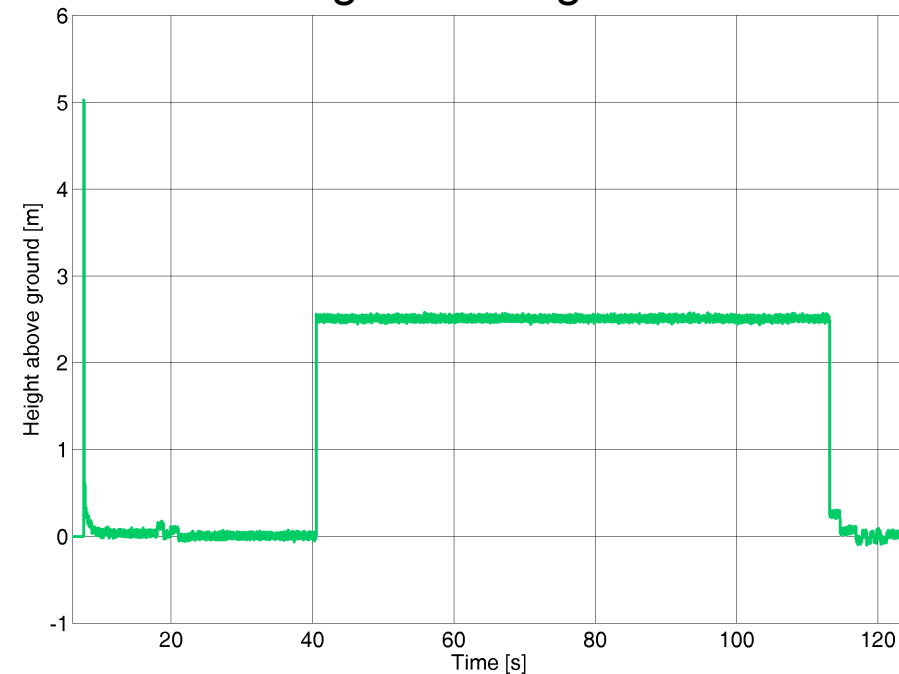
— Ground truth — Kalman filter Baro/Vision ● Vision

Hovering at defined position and landing

Height above ground estimation



Error of height above ground

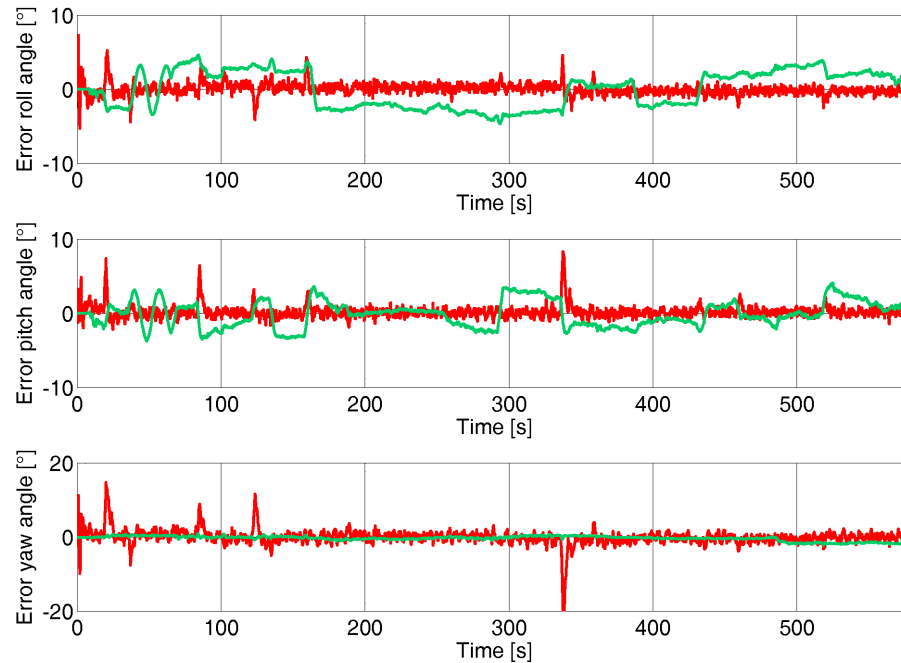


Simulated step of ground elevation of 2.5 m

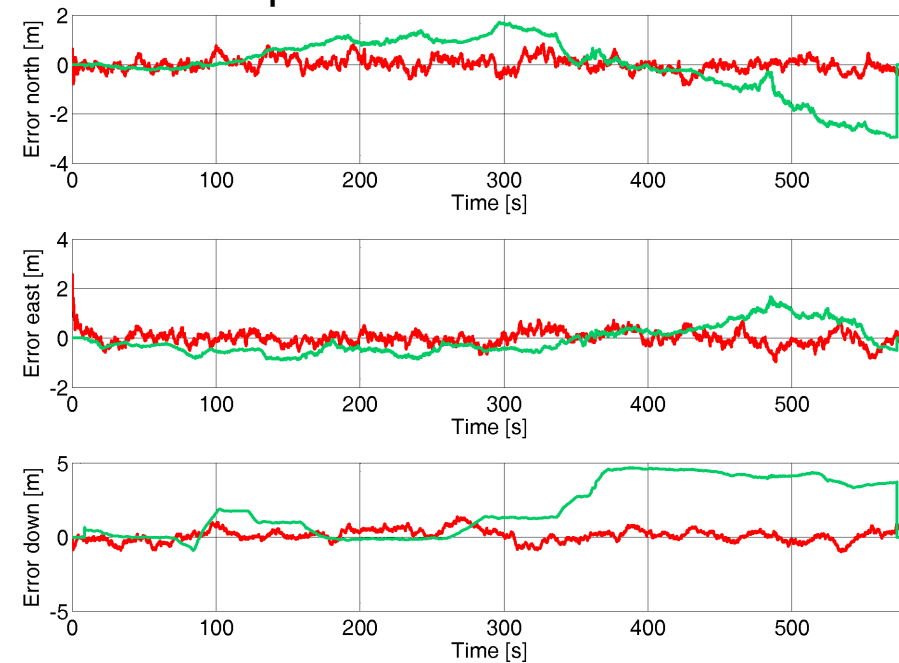
— Ground truth — Kalman filter Baro/Vision ● Vision

2. Simulation: Waypoint flight

Error of attitude



Error of position



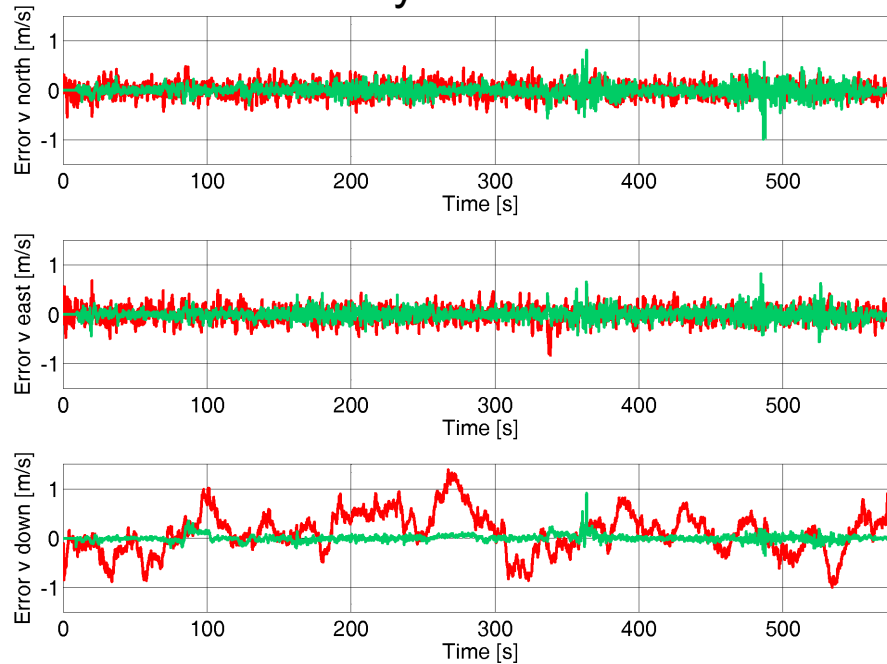
Frame rate 25 fps, image size 640x480, 200 features, feature noise $\sigma = \frac{1}{3}$ pix.

— GPS/INS/Mag/Baro — Vision

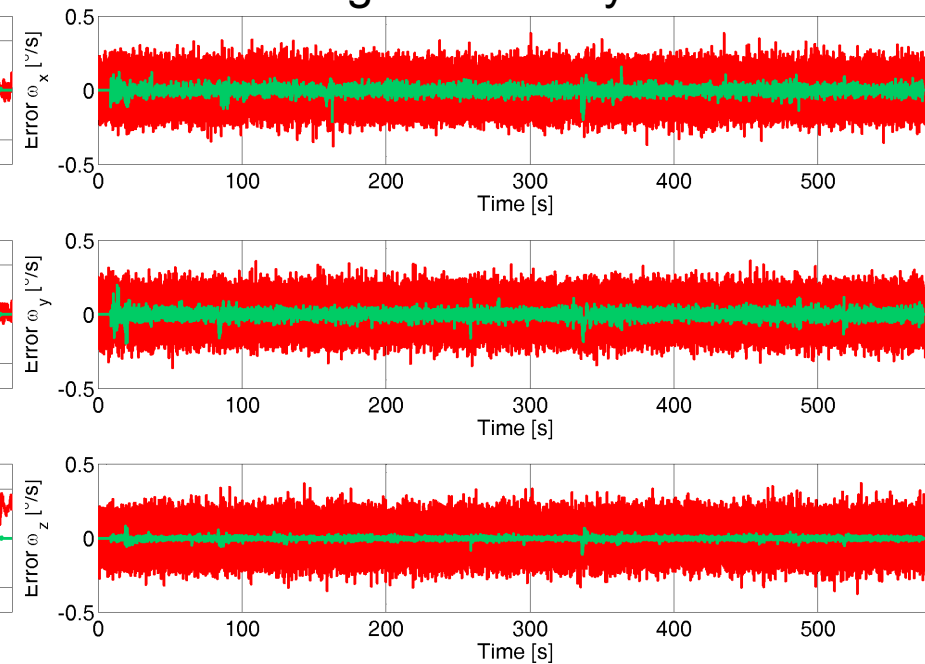
Results

Waypoint flight

Error of velocity



Error of angular velocity



Positions divided by 1/25fps,
Averaging with $n = 6$, data rate 4.16Hz

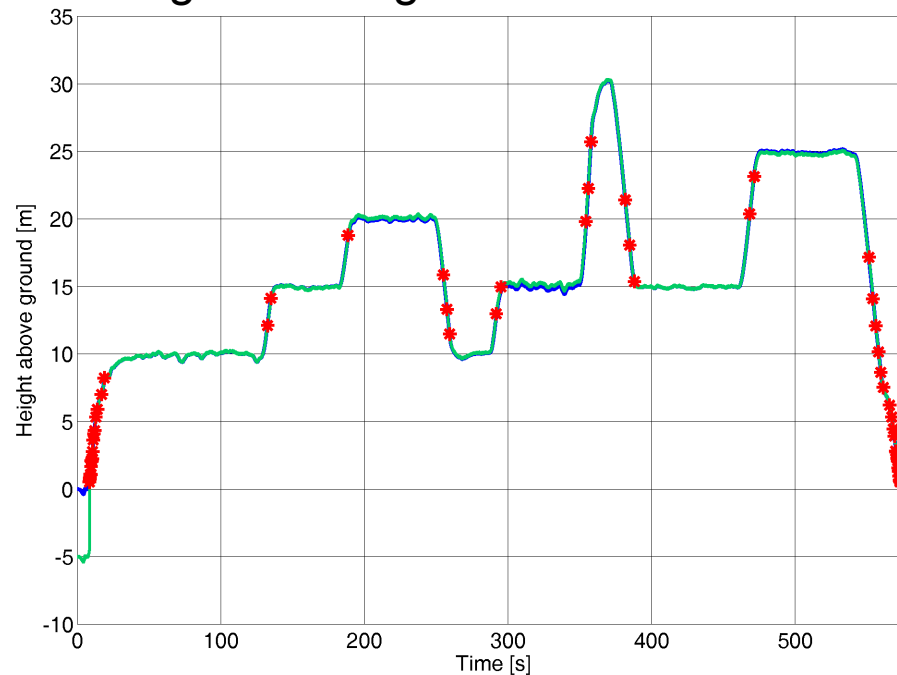
Angles divided by 1/25fps

— GPS/INS/Mag/Baro — Vision

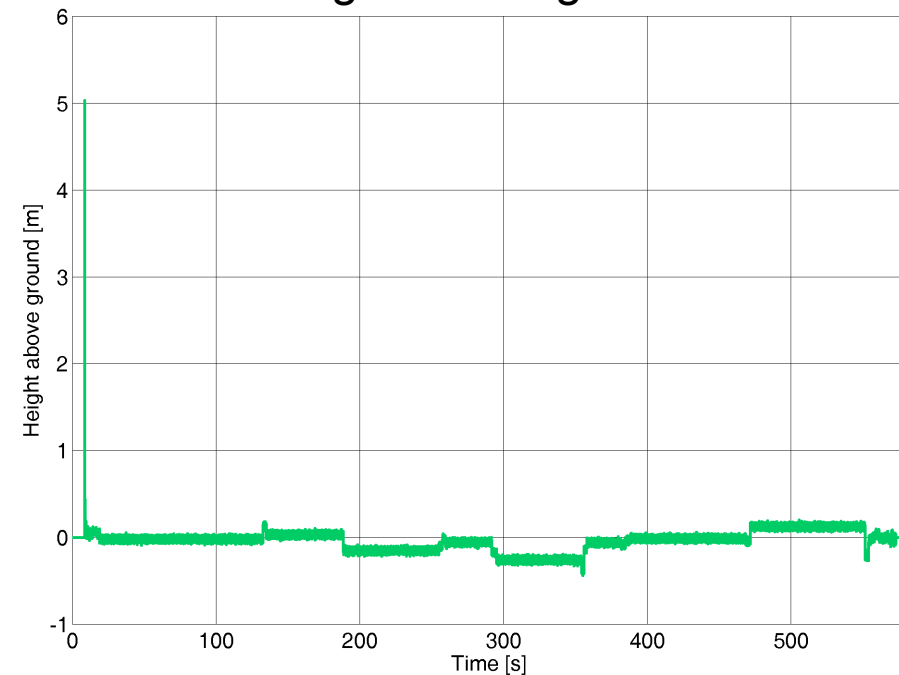
Results

Waypoint flight

Height above ground estimation



Error of height above ground



Baro rate 25Hz, baro offset -5m, baro noise $\sigma = 0.02$ m/min

m, baro drift 0.2

— Ground truth — Kalman filter Baro/Vision ● Vision

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Master Ueberschrift

