

A NEW APPROACH FOR AN INCREMENTAL ORIENTATION OF MICRO-UAV IMAGE SEQUENCES

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Motivation





Motivation

- incremental orientation
 - real time compatible
 - (adaptive) exploration of new (unknown) environment
 - cartography
 - modelling
- a new approach
 - image orientation is an ill-posed problem
 - improvement by using convexly optimised initial values





Contents

- Hardware
- Methodology
- Experimental results
- Conclusion



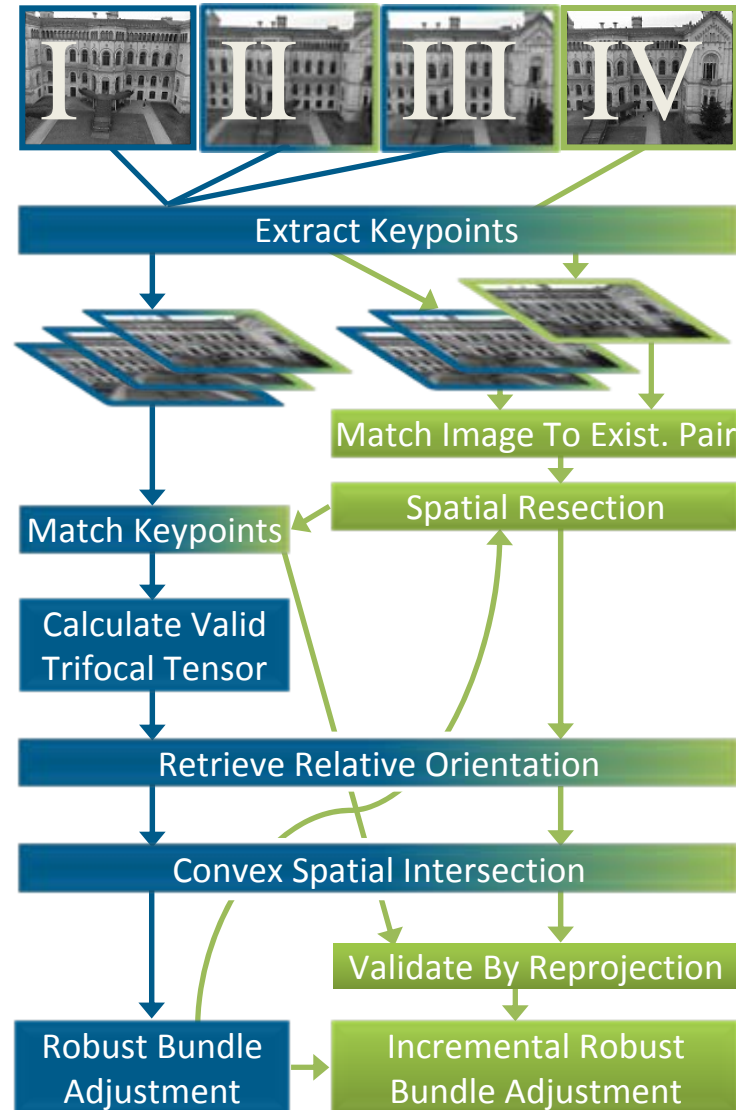
Hardware

- Microdrones md4-200 VTOL Quadrocopter
 - max. payload: 200g
 - max. flight duration: 30min
 - cruising speed $\sim 5\text{m/s}$
 - GPS, SBAS, IMU, Magnetometer, Barometer
- Canon Powershot S110
 - weight: 200g
 - resolution: 720 x 576 px
 - transmission: e. g. WiFi SD-card

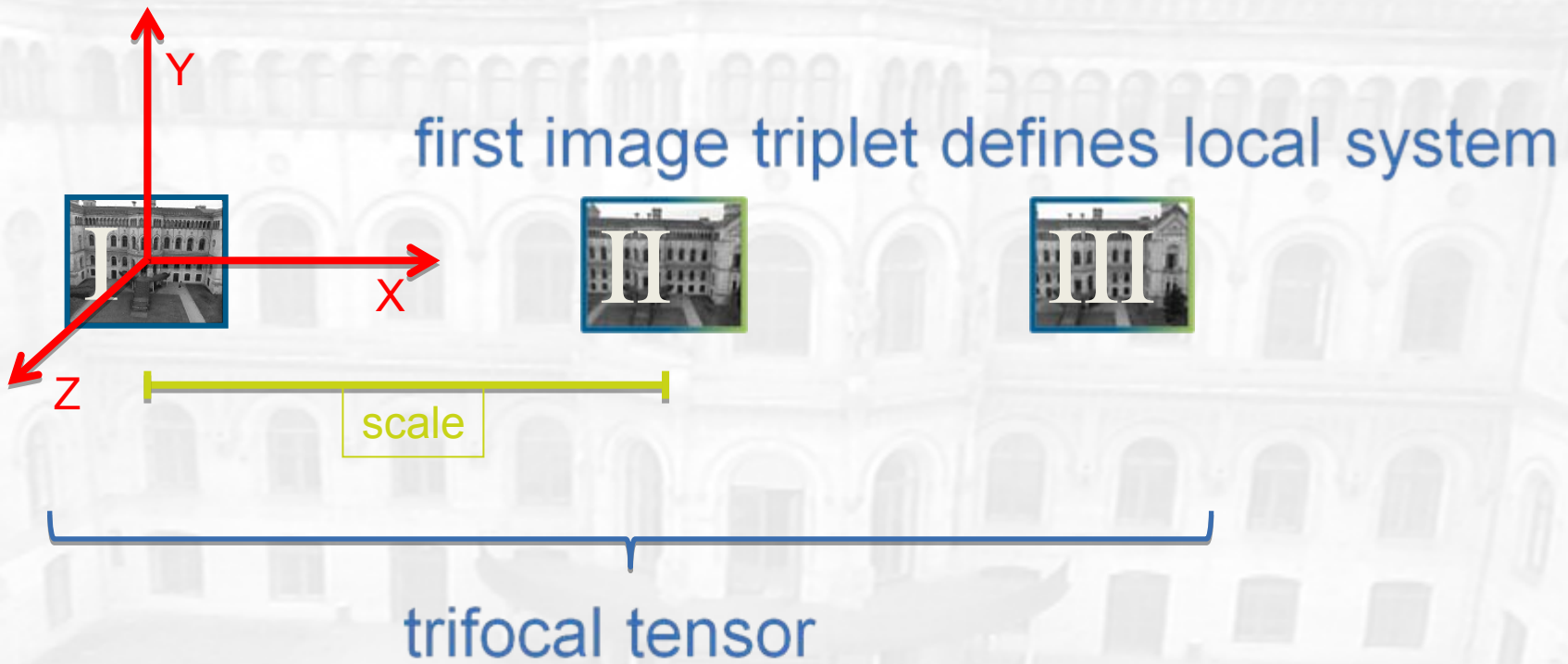


Methodology

- Work-flow



First image triplet



$$p^{I,i} \left(p^{II,j} \epsilon_{jpr} \right) \left(p^{III,k} \epsilon_{kqs} \right) T_i^{pq} = 0_{rs}.$$

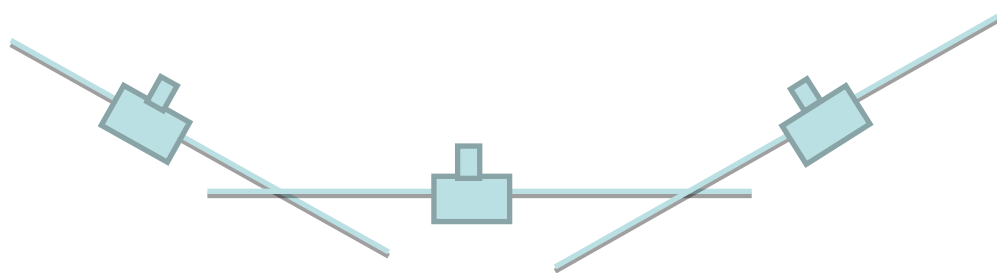
→ relative orientation → convex spatial intersection

Convex spatial intersection

- spatial intersection not always has a unique solution
- objective: minimise reprojection error

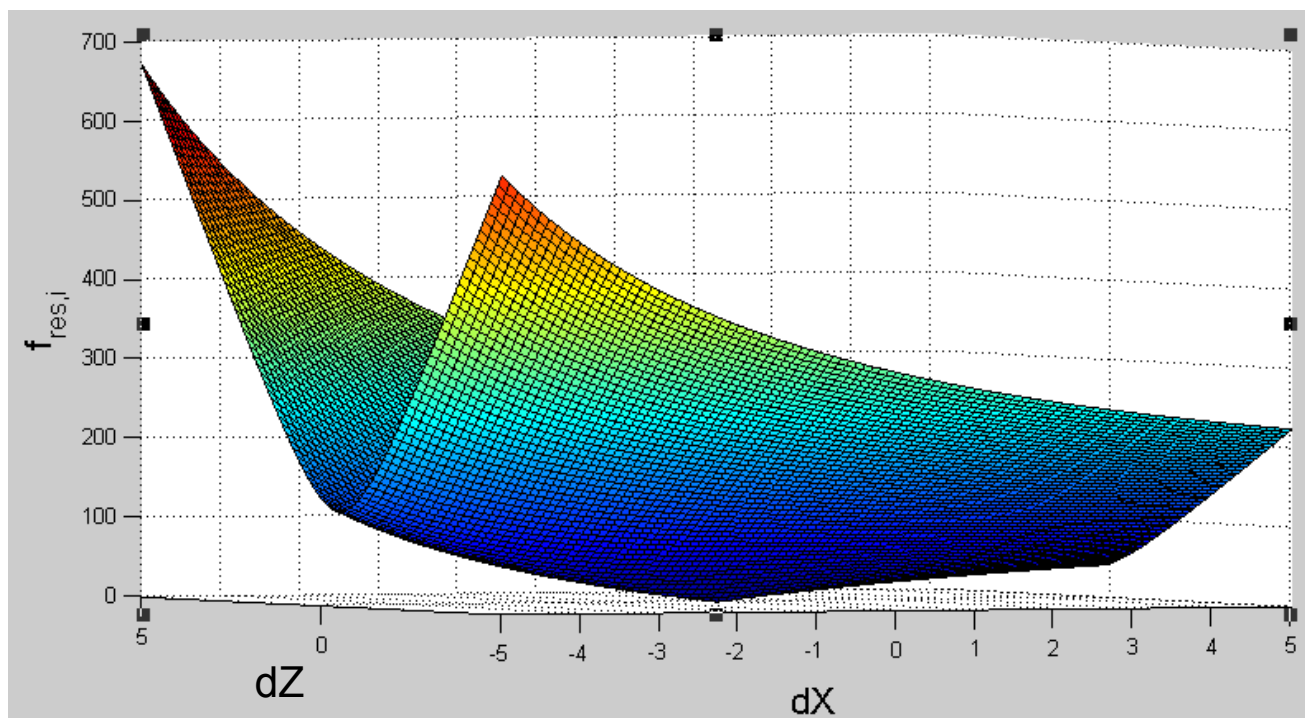
$$f_{res,i}(\mathbf{X}) = \|\mathbf{x}_{proj,i} - \mathbf{x}_i\|_m$$

subject to a convex set defined by halfspaces of involved image planes



Convex spatial intersection

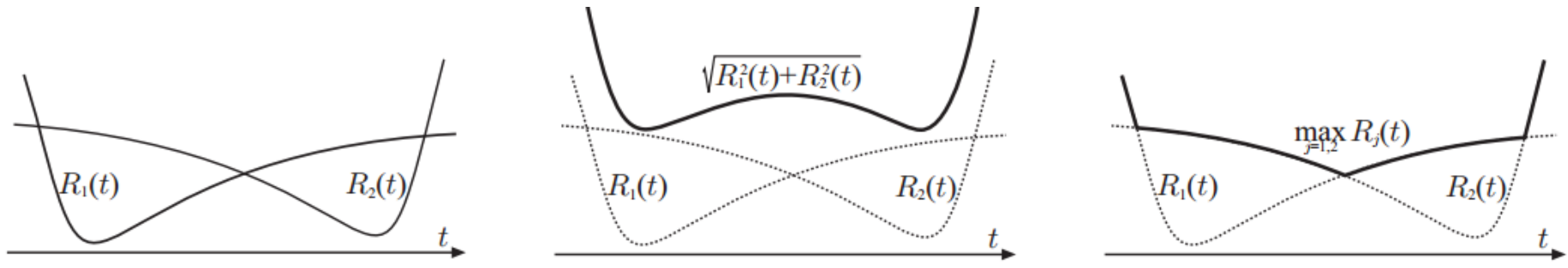
- reprojection error in image i is a quasiconvex function



- minimising reprojection error in images $i = 1, \dots, N$
 \rightarrow how quasiconvexity behaves?

Convex spatial intersection

- how quasiconvexity behaves?



[Schneider, 2011]

Convex spatial intersection

- using L_∞ norm \rightarrow quasiconvexity is preserved

$$\text{minimise} \quad \max_{i=1}^3 f_{res,i}(\mathbf{X})$$

$$f_{res,i}(\mathbf{X}) = \left\| \begin{pmatrix} -c \frac{\mathbf{r}_{i,1}^T(\mathbf{X}-\mathbf{X}_0)}{\mathbf{r}_{i,3}^T(\mathbf{X}-\mathbf{X}_0)} + x_0 - x_i \\ -c \frac{\mathbf{r}_{i,2}^T(\mathbf{X}-\mathbf{X}_0)}{\mathbf{r}_{i,3}^T(\mathbf{X}-\mathbf{X}_0)} + y_0 - y_i \end{pmatrix} \right\|_m$$

$$= \left\| \begin{pmatrix} \frac{(-c \cdot \mathbf{r}_{i,1}^T + (x_0 - x_i) \mathbf{r}_{i,3}^T) \mathbf{X} - (-c \cdot \mathbf{r}_{i,1}^T + (x_0 - x_i) \mathbf{r}_{i,3}^T) \mathbf{X}_0}{\mathbf{r}_{i,3}^T \mathbf{X} - \mathbf{r}_{i,3}^T \mathbf{X}_0} \\ \frac{(-c \cdot \mathbf{r}_{i,2}^T + (y_0 - y_i) \mathbf{r}_{i,3}^T) \mathbf{X} - (-c \cdot \mathbf{r}_{i,2}^T + (y_0 - y_i) \mathbf{r}_{i,3}^T) \mathbf{X}_0}{\mathbf{r}_{i,3}^T \mathbf{X} - \mathbf{r}_{i,3}^T \mathbf{X}_0} \end{pmatrix} \right\|_m$$

$$f_{res,i}(\mathbf{X}) = \frac{\|\mathbf{A}_i \mathbf{X} + \mathbf{b}_i\|_m}{\mathbf{c}_i^T \mathbf{X} + d_i}$$



Convex spatial intersection

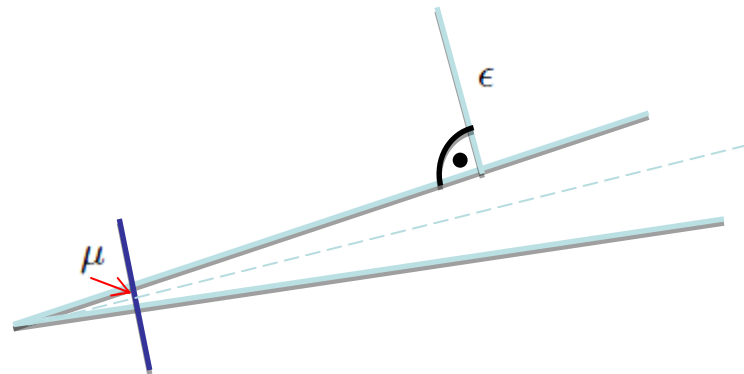
- using L_∞ norm \rightarrow quasiconvexity is preserved

$$\text{minimise} \quad \max_{i=1}^3 f_{res,i}(\mathbf{X})$$

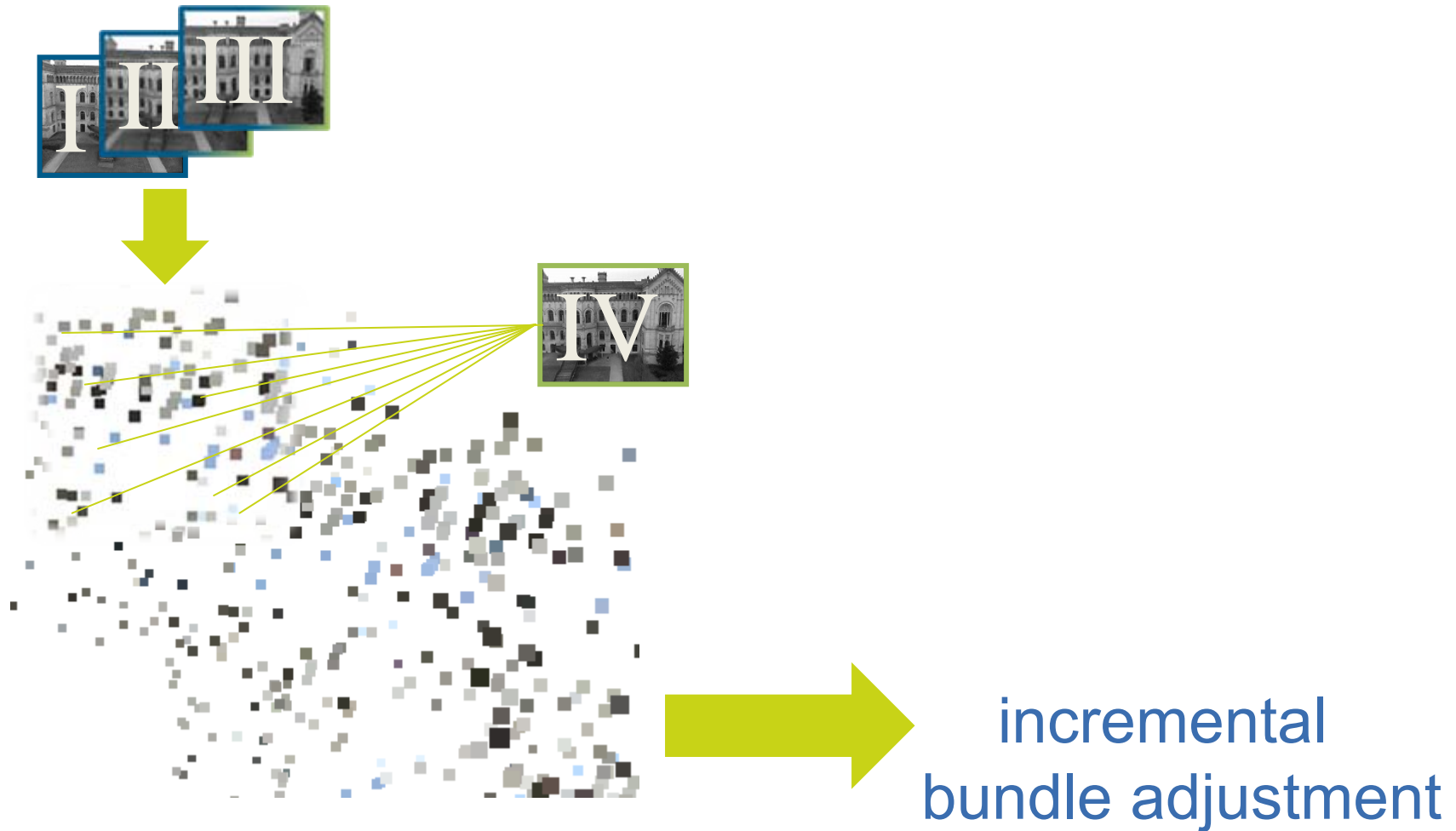
$$f_{res,i}(\mathbf{X}) = \frac{\|\mathbf{A}_i \mathbf{X} + \mathbf{b}_i\|_m}{\mathbf{c}_i^T \mathbf{X} + d_i}$$

$$\text{minimise} \quad \epsilon$$

$$\text{subject to} \quad \|\mathbf{A}_i \mathbf{X} + \mathbf{b}_i\|_\infty - \mu(\mathbf{c}_i^T \mathbf{X} + d_i) \leq \epsilon, \\ i = 1 \dots 3.$$



Further triplets



Incremental bundle adjustment

$$\begin{pmatrix} \mathbf{l}_1 + \mathbf{v}_1 \\ \mathbf{l}_2 + \mathbf{v}_2 \end{pmatrix} = \begin{pmatrix} \mathbf{A}_{11} & 0 \\ \mathbf{A}_{21} & \mathbf{A}_{22} \end{pmatrix} \begin{pmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \end{pmatrix}$$

- two different types of parameters
 1. parameters related to already estimated unknowns \mathbf{p}_1
 2. parameters related to unknowns that are estimated for the first time \mathbf{p}_2

$$\begin{pmatrix} \hat{\mathbf{p}}_{1,+} \\ \hat{\mathbf{p}}_2 \end{pmatrix} = \mathbf{N}_+^{-1} \begin{pmatrix} \mathbf{A}_{11}^T \mathbf{P}_{11} \mathbf{l}_1 + \mathbf{A}_{21}^T \mathbf{P}_{22} \mathbf{l}_2 \\ \mathbf{A}_{22}^T \mathbf{P}_{22} \mathbf{l}_2 \end{pmatrix}$$



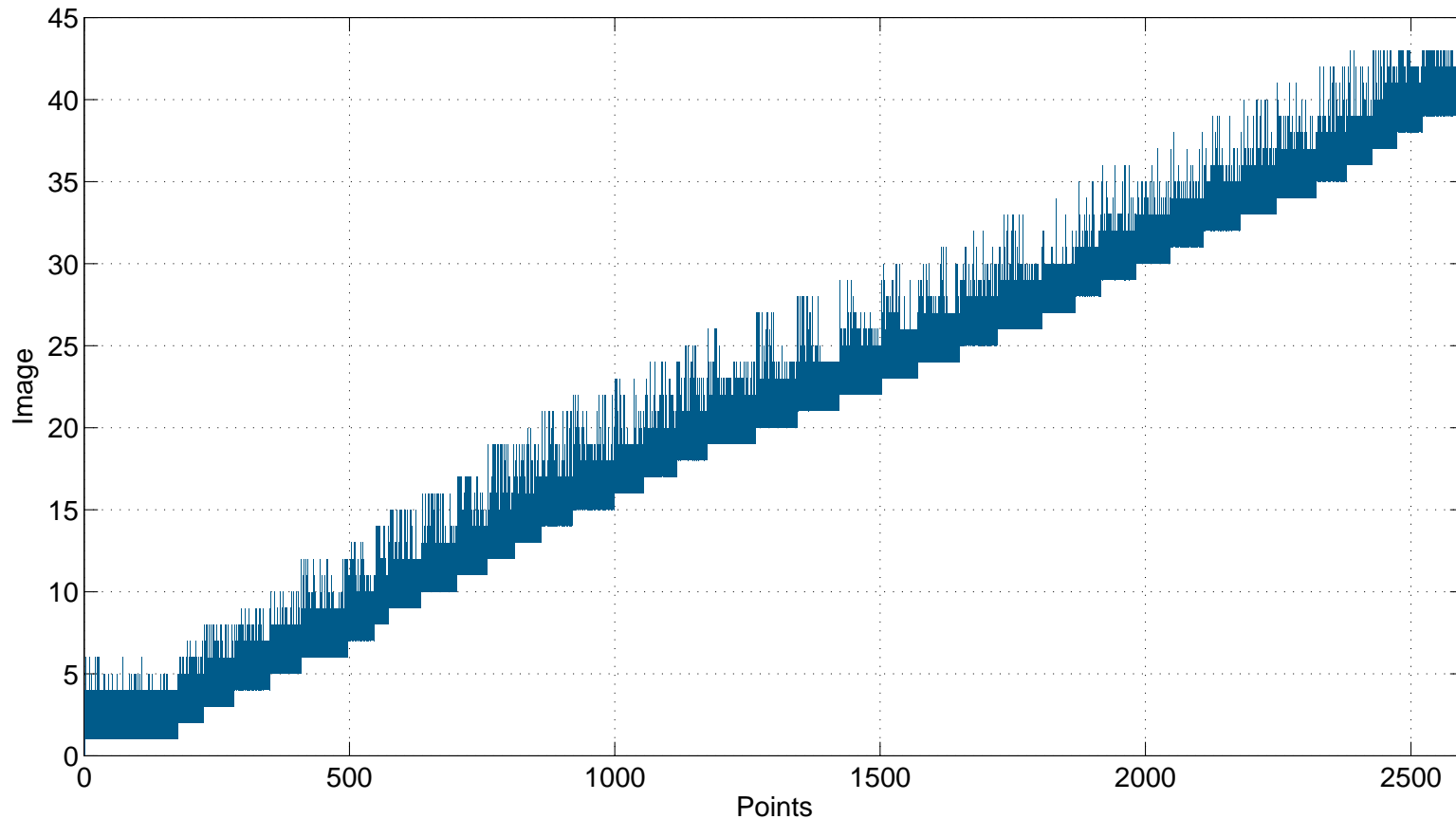
Settings

- camera calibrated before flight
- setting: Welfenschloss Hannover
- images collected in a fixed time interval
 - three separate flights each consisting of about 250 images
- number of observations limited based on distribution in the image
- old images are eliminated based on number of re-measured points



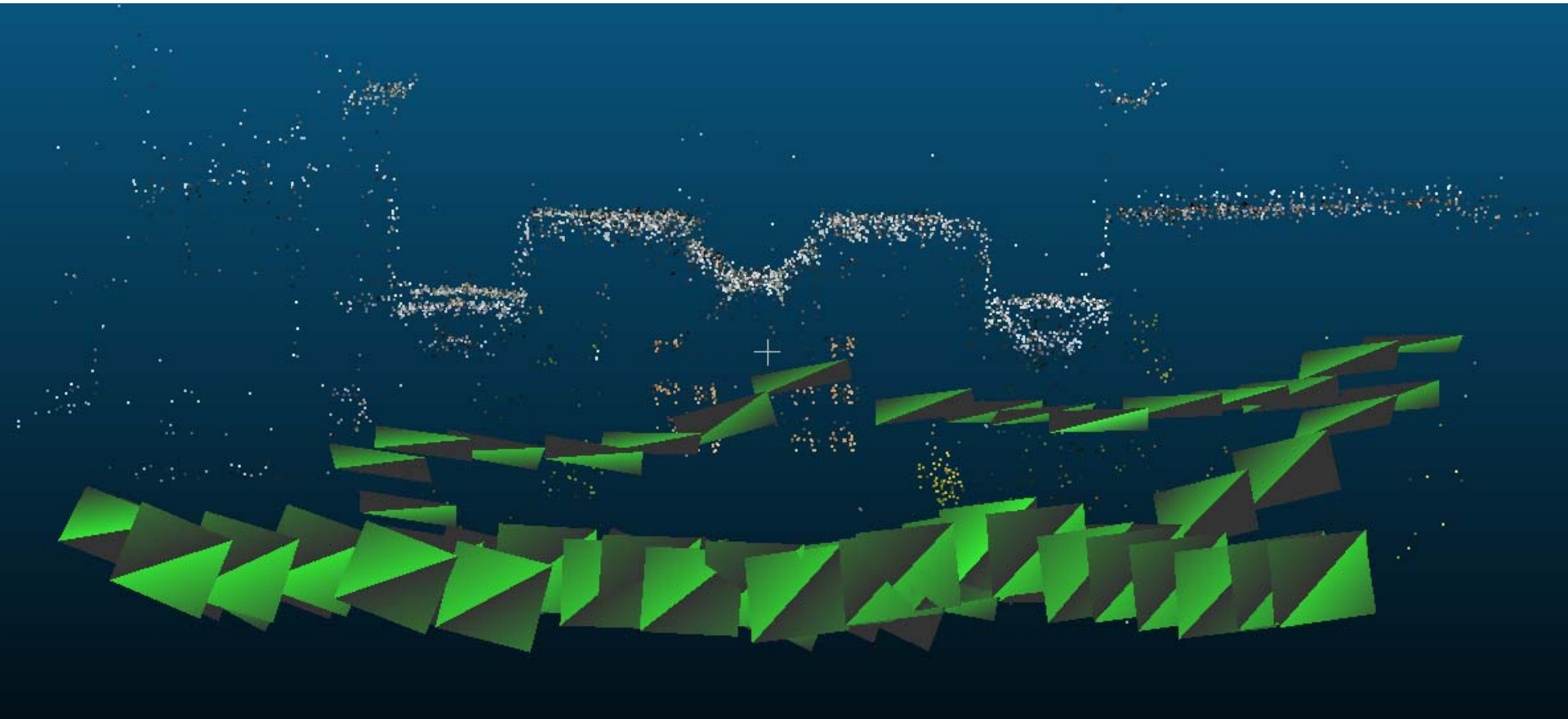
Results

- point/image distribution



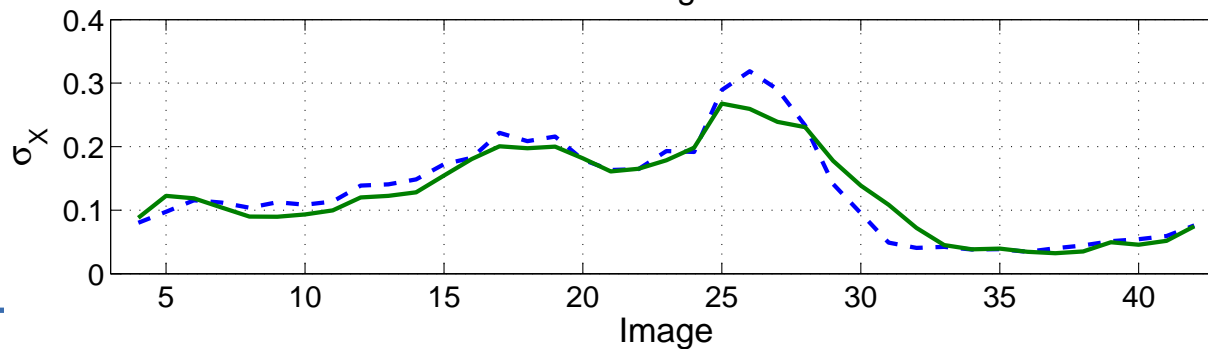
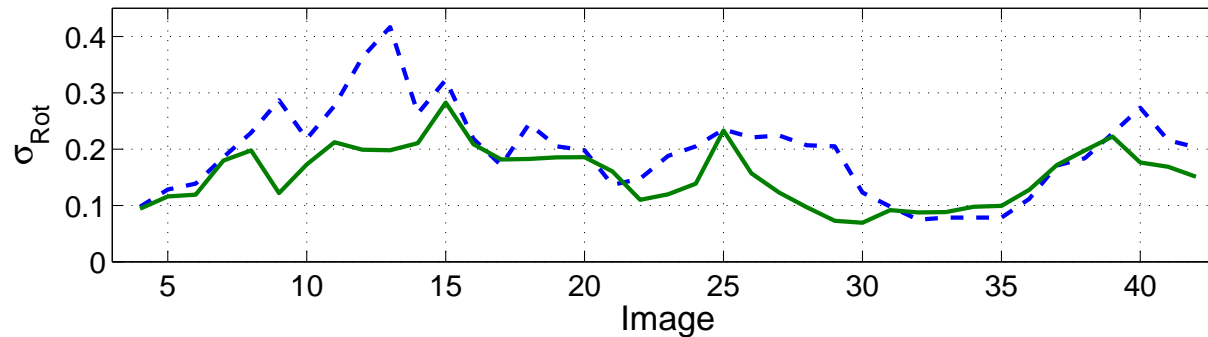
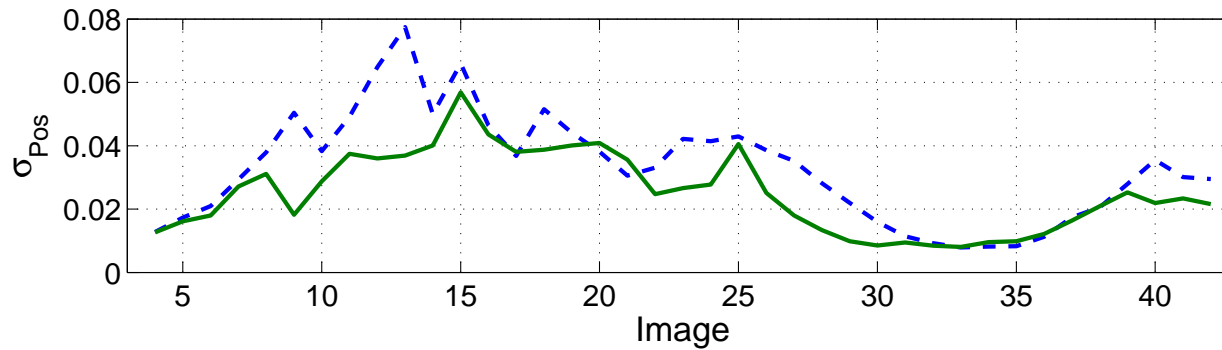
Results

- sparse point cloud and image orientation



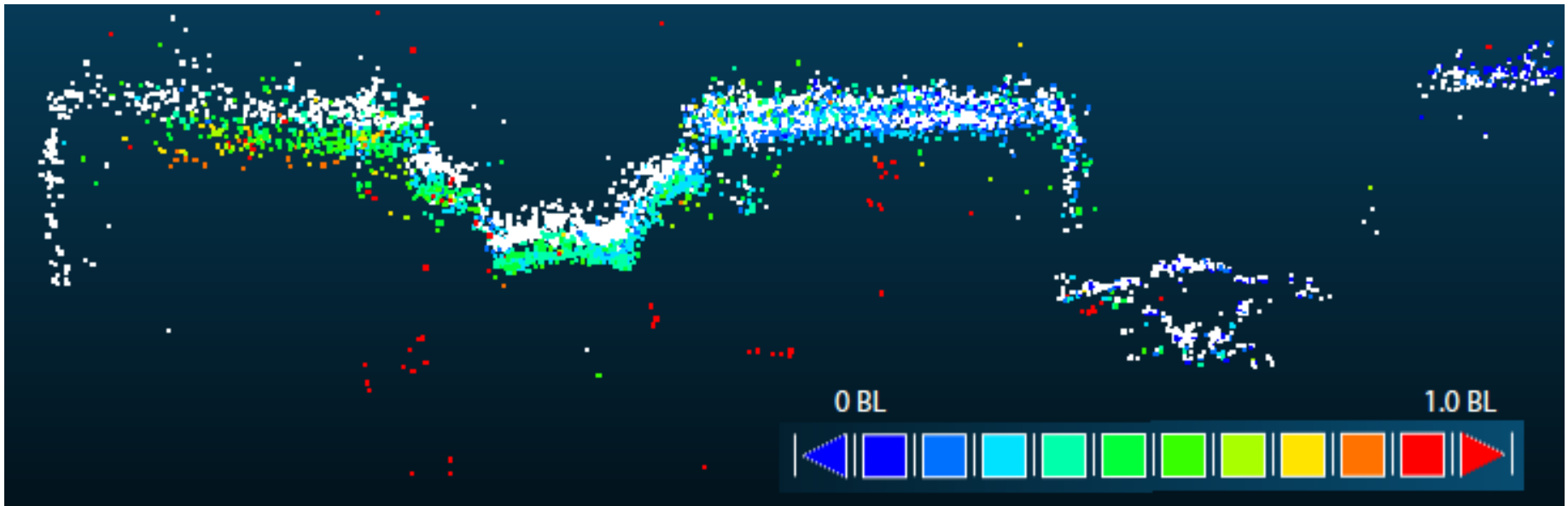
Results

- precision of the bundle adjustment



Results

- point cloud comparison



Conclusion

- incremental adjustment allows
 - estimation of new parameters
 - improvement of estimated parameters
- computation time highly depends on the number of newly found object points
 - limitation of number of observations necessary
 - nearly constant due to incremental bundle adjustment ($\sim 10\text{s}/\text{image}$ \rightarrow non-optimised \rightarrow covariance estimation)



Conclusion

- convex optimisation shows promising results
 - improved initial values
 - distribution of object points in space
- future challenges
 - unstable solution/drift → loop closure in on-line mode
 - absolute orientation of the model → integrate telemetry data
 - solid reference model → terrestrially measured/defined in a global coordinate system



Thank You!

