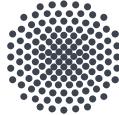


Tutorium Teil 1

UAV-basierte Luftbildauswertung - ein Update

Michael Cramer

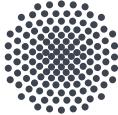


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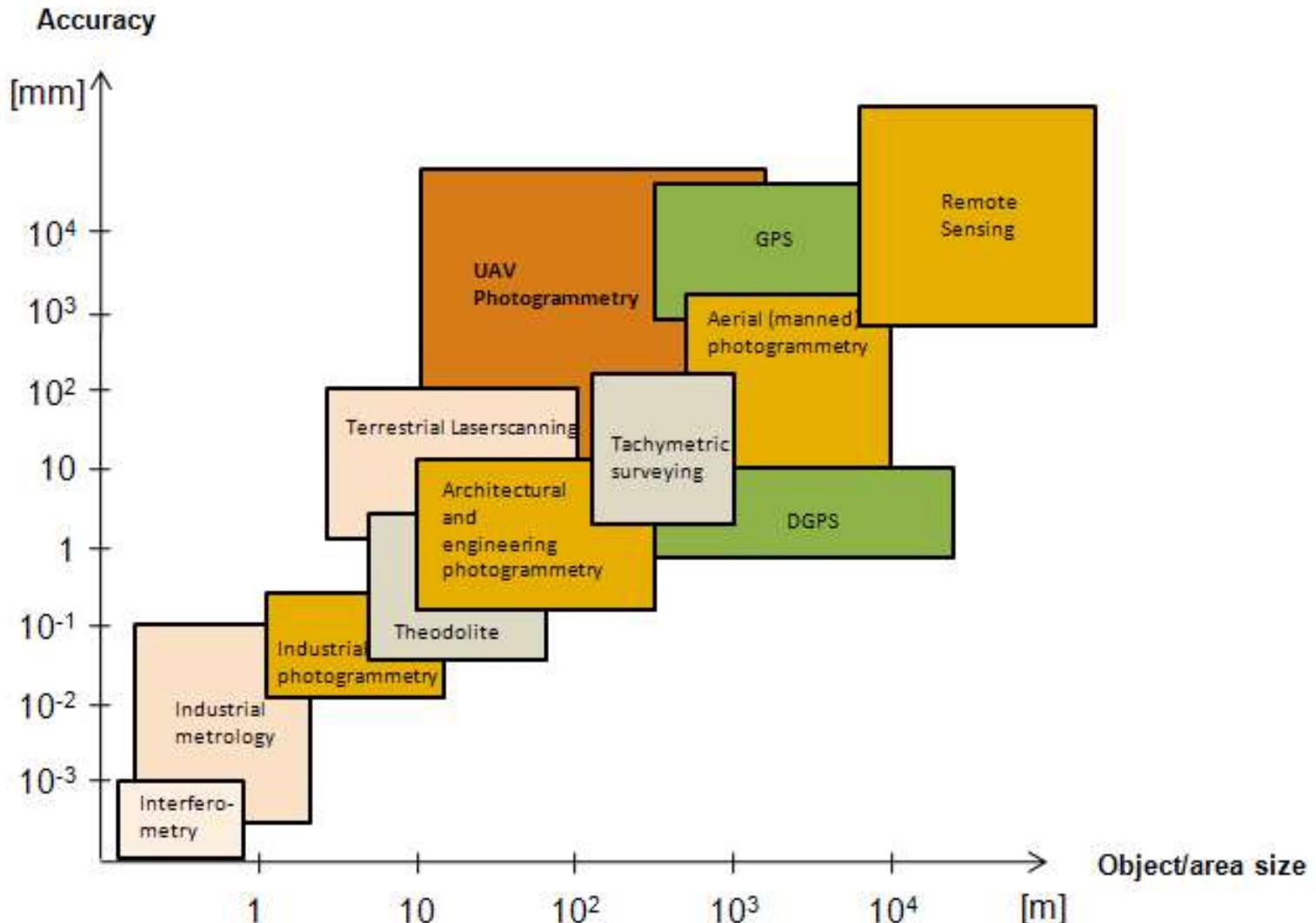
- Is UAV-based mapping so much different to regular flight campaigns?
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State-of-the-art in UAV-based mapping – an update

Is UAV-based mapping so much different to regular flight campaigns?

Use of UAV in Mapping / Photogrammetry



© Luhmann et al., 2006

Manned vs. unmanned photo flights

Platforms & cameras

Platform

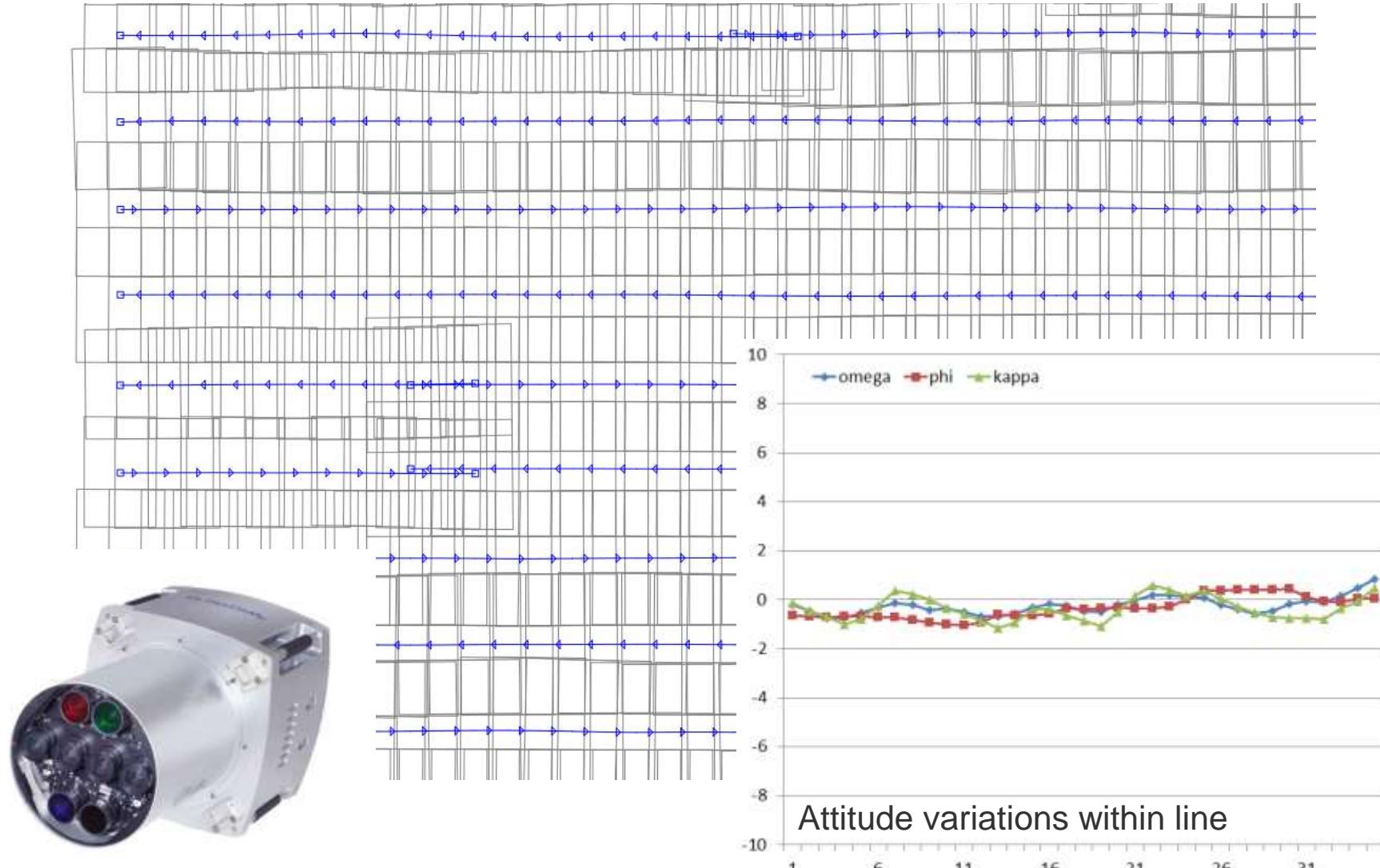


Camera

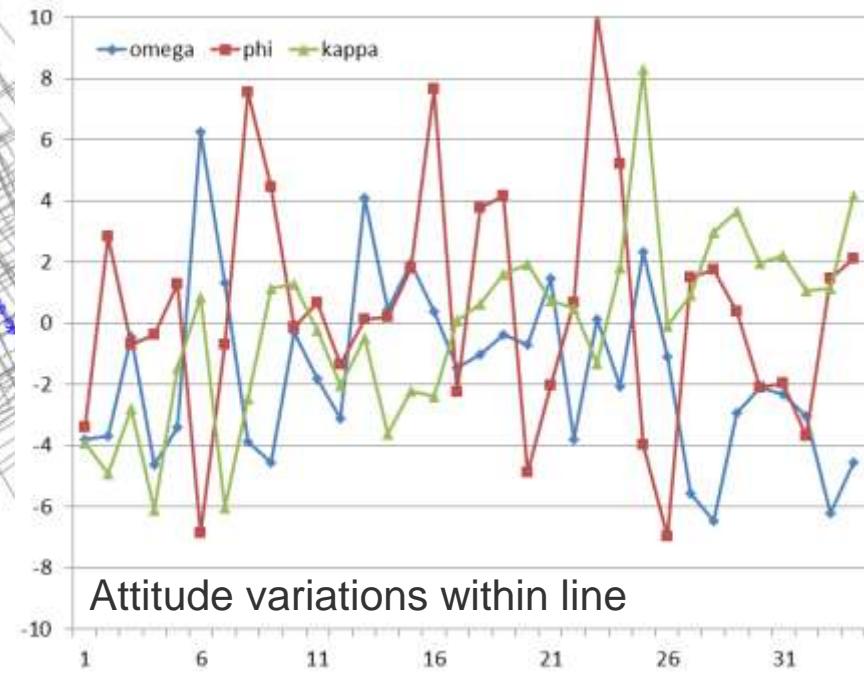
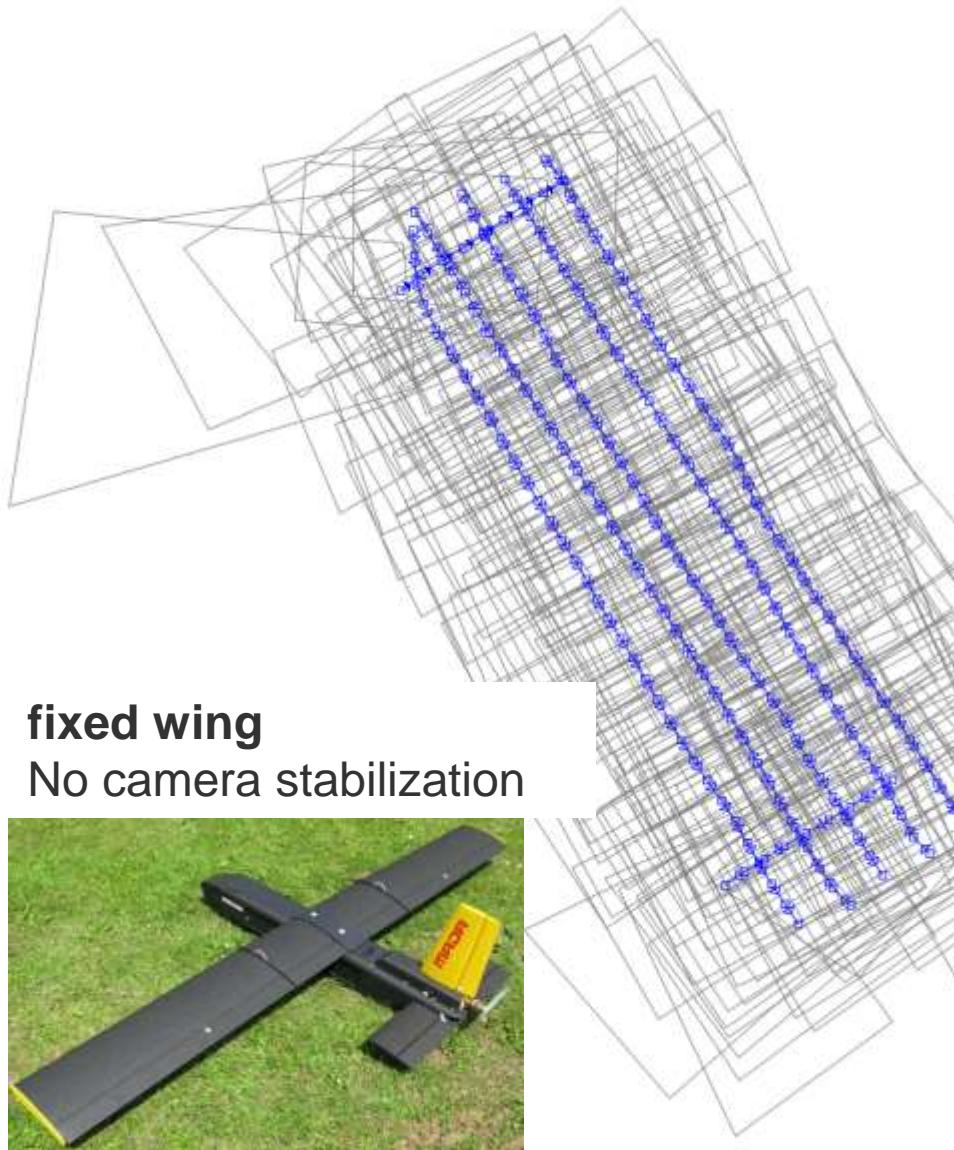


Block configurations: Manned platform

Orthomapping flights LGL BW



Block configurations: UAS – fixed wing

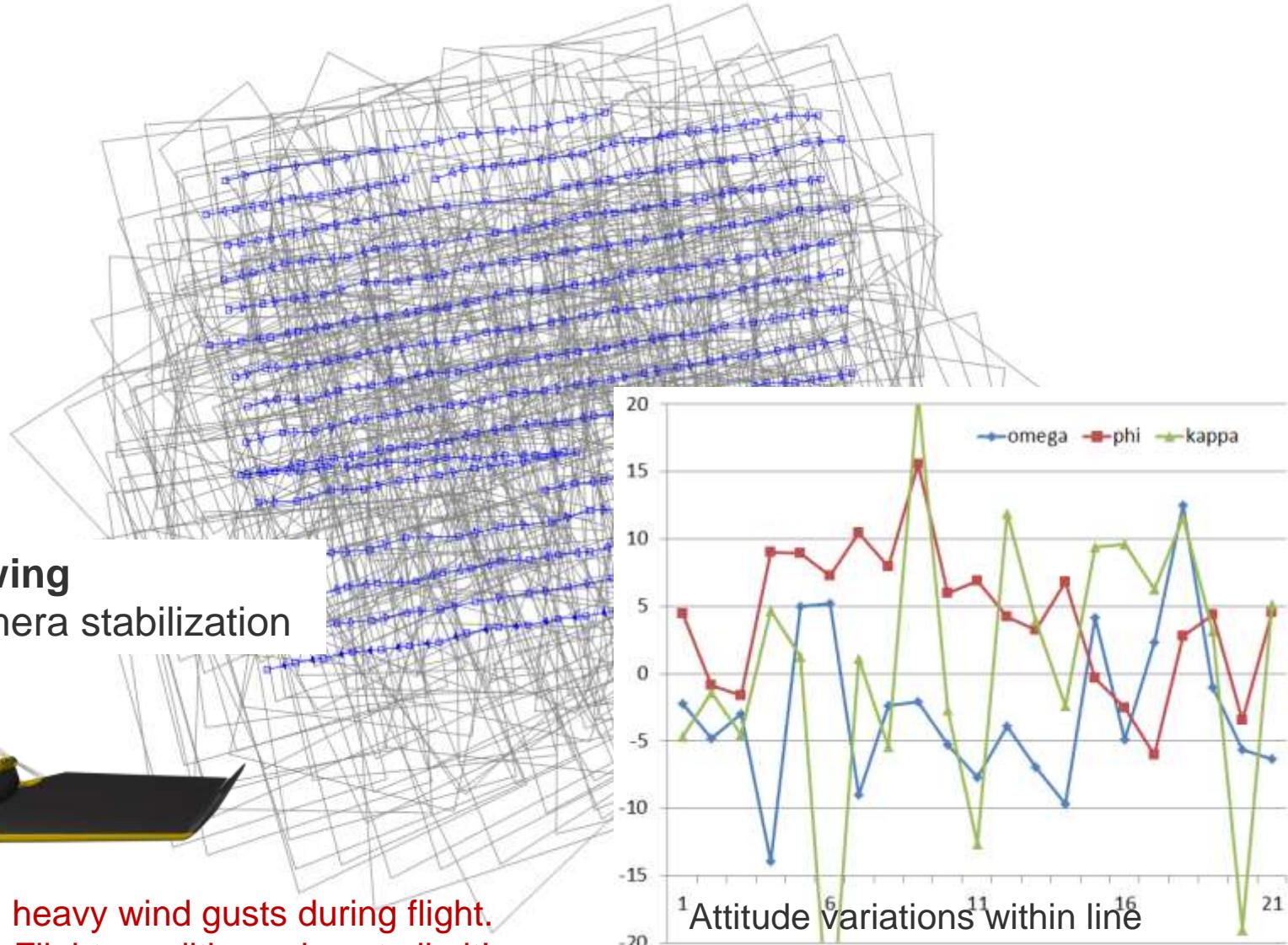


Block configurations: UAS – fixed wing

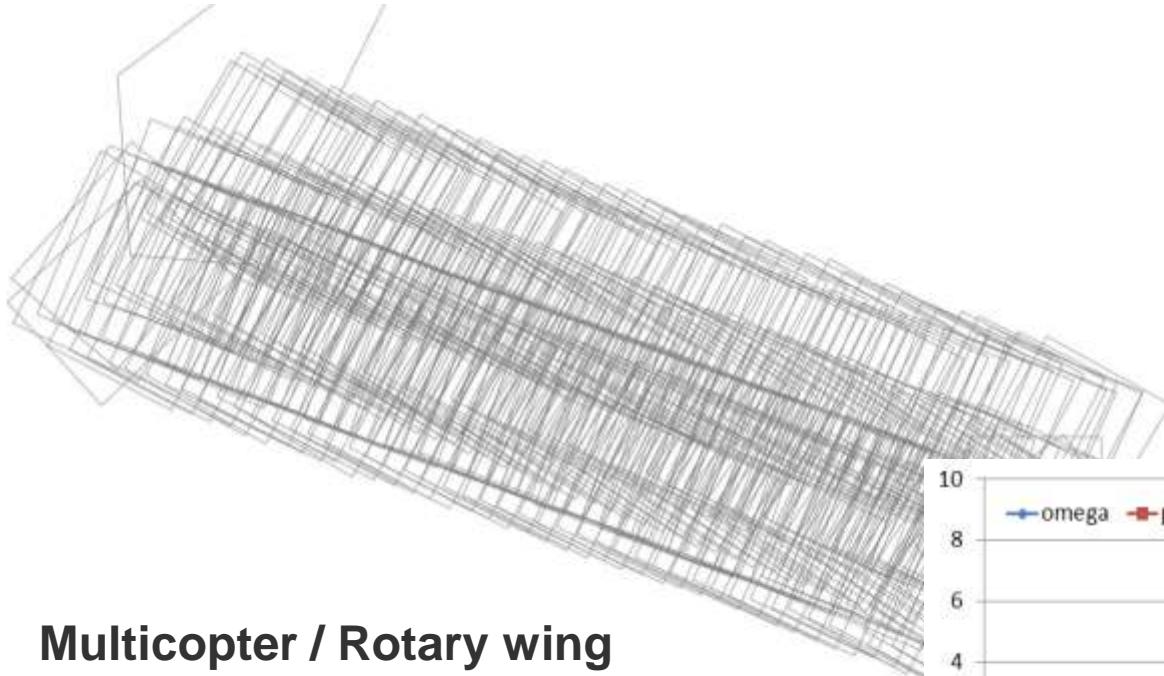
fixed wing
No camera stabilization



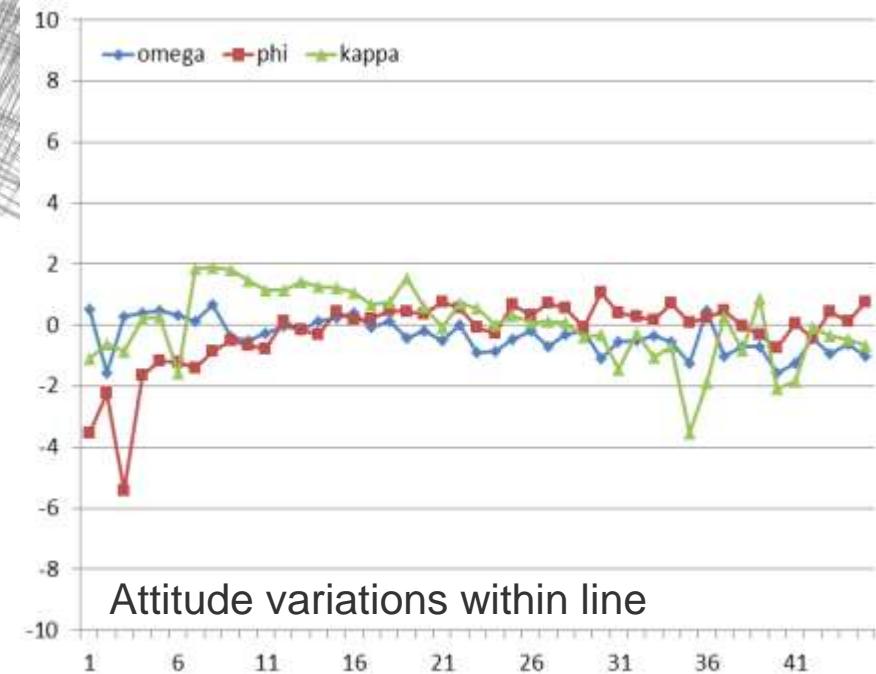
Notice: heavy wind gusts during flight.
Flight conditions close to limit!



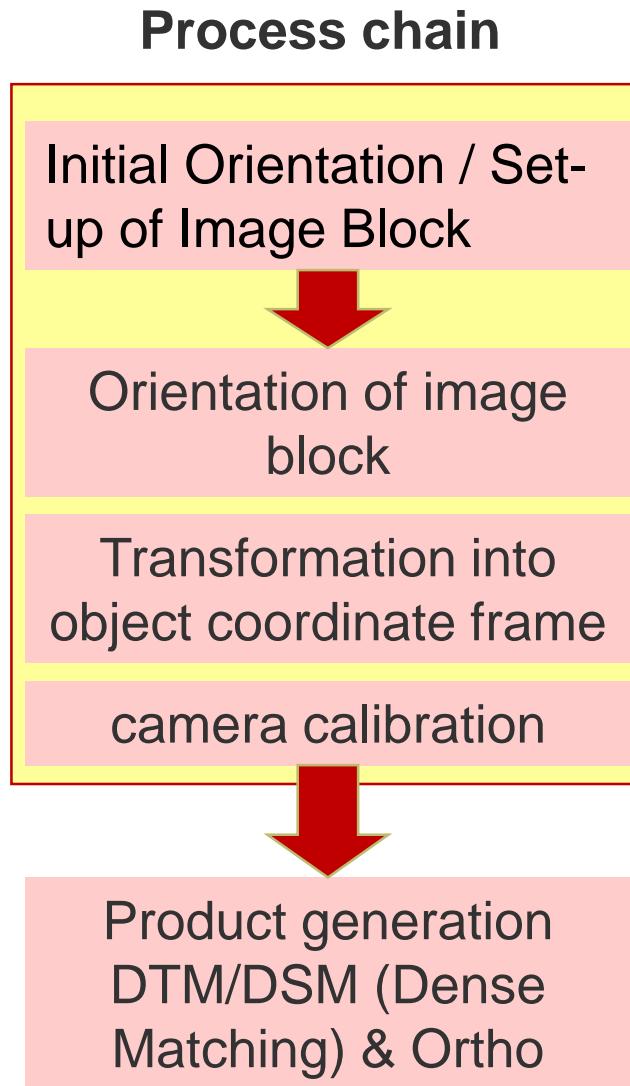
Block configurations: UAS – rotary wing



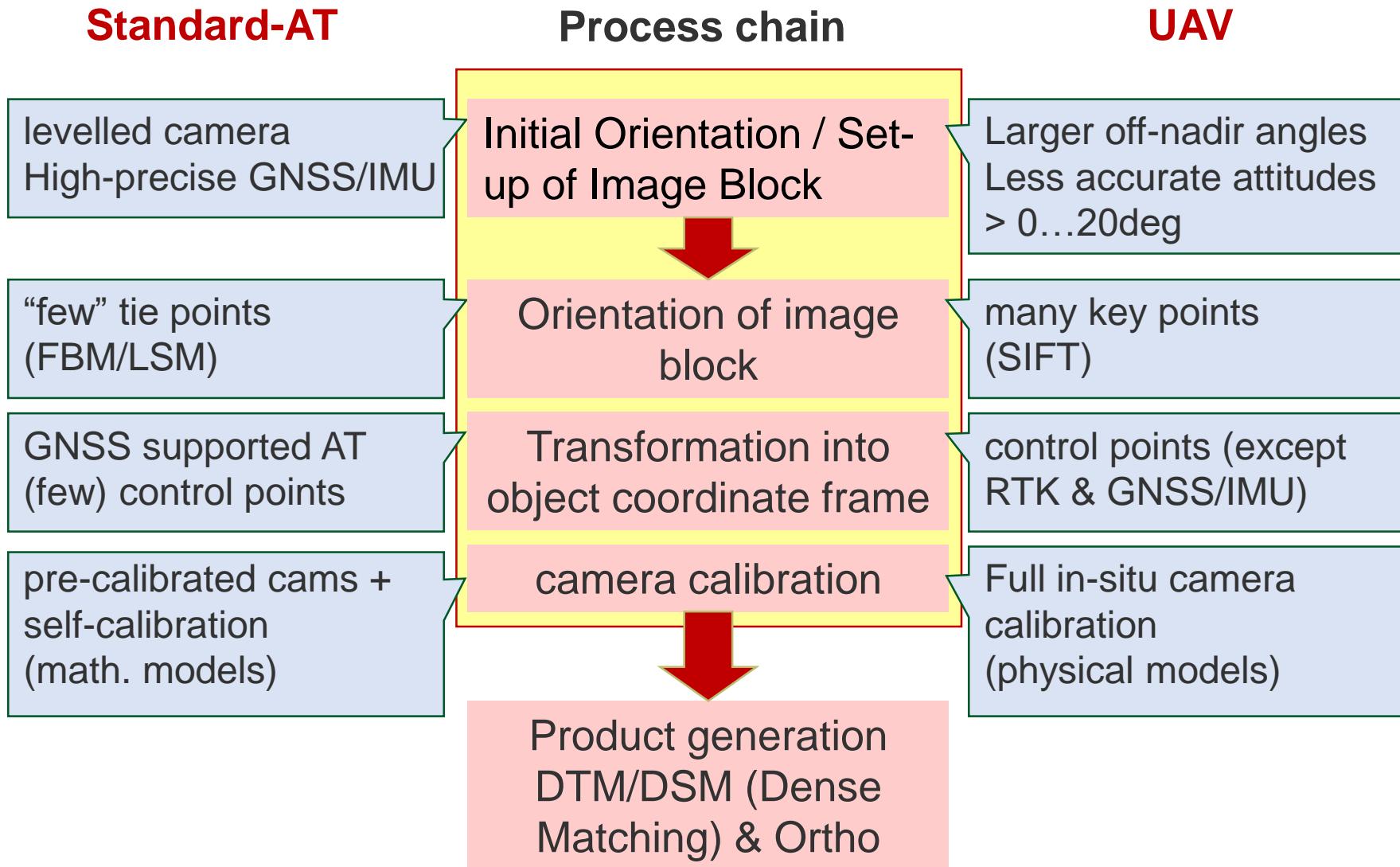
Multicopter / Rotary wing
Camera stabilization



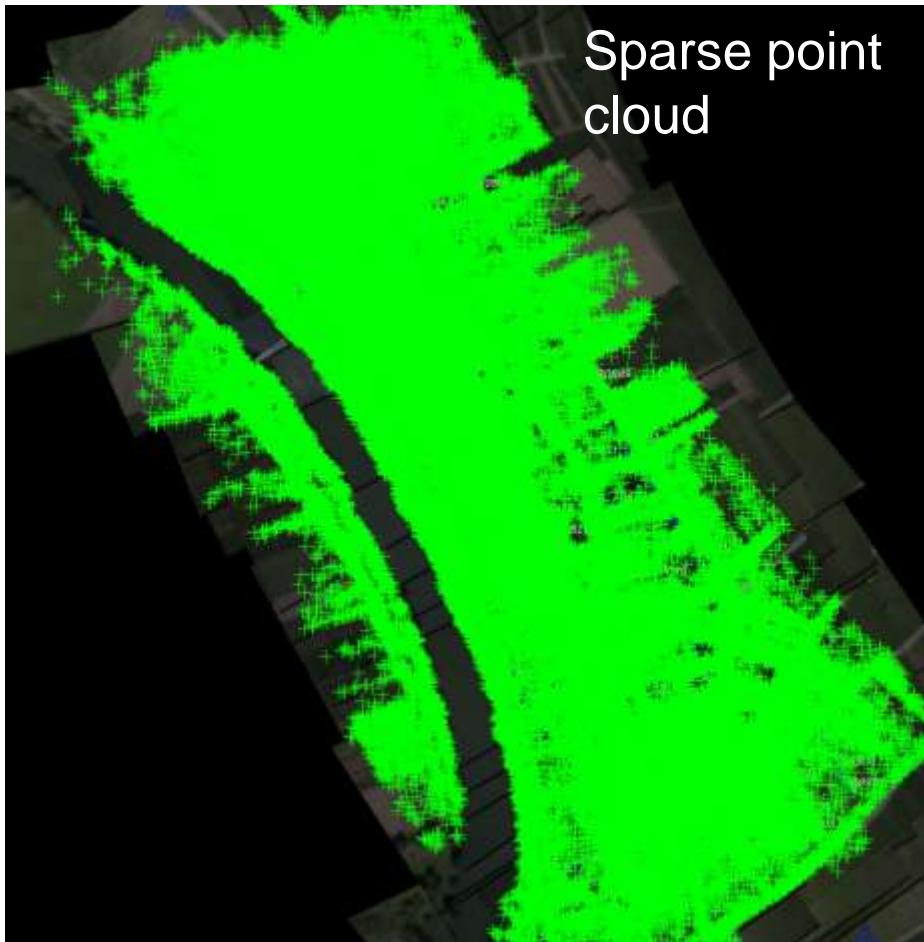
Aerial Triangulation



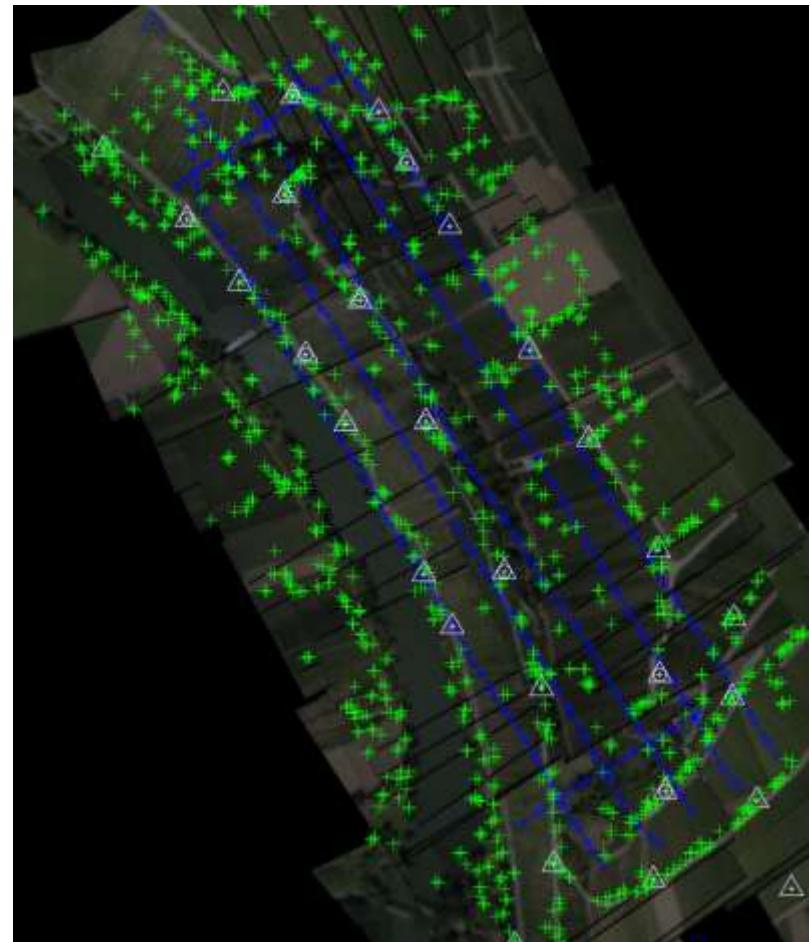
Aerial Triangulation (classical vs. UAV-based)



Tie point extraction



SIFT-Points (SfM Bundler)



FBM/LSM points (inpho Match-AT)

Manned vs. unmanned photo flights

Mapping Camera vs. Bridge Camera



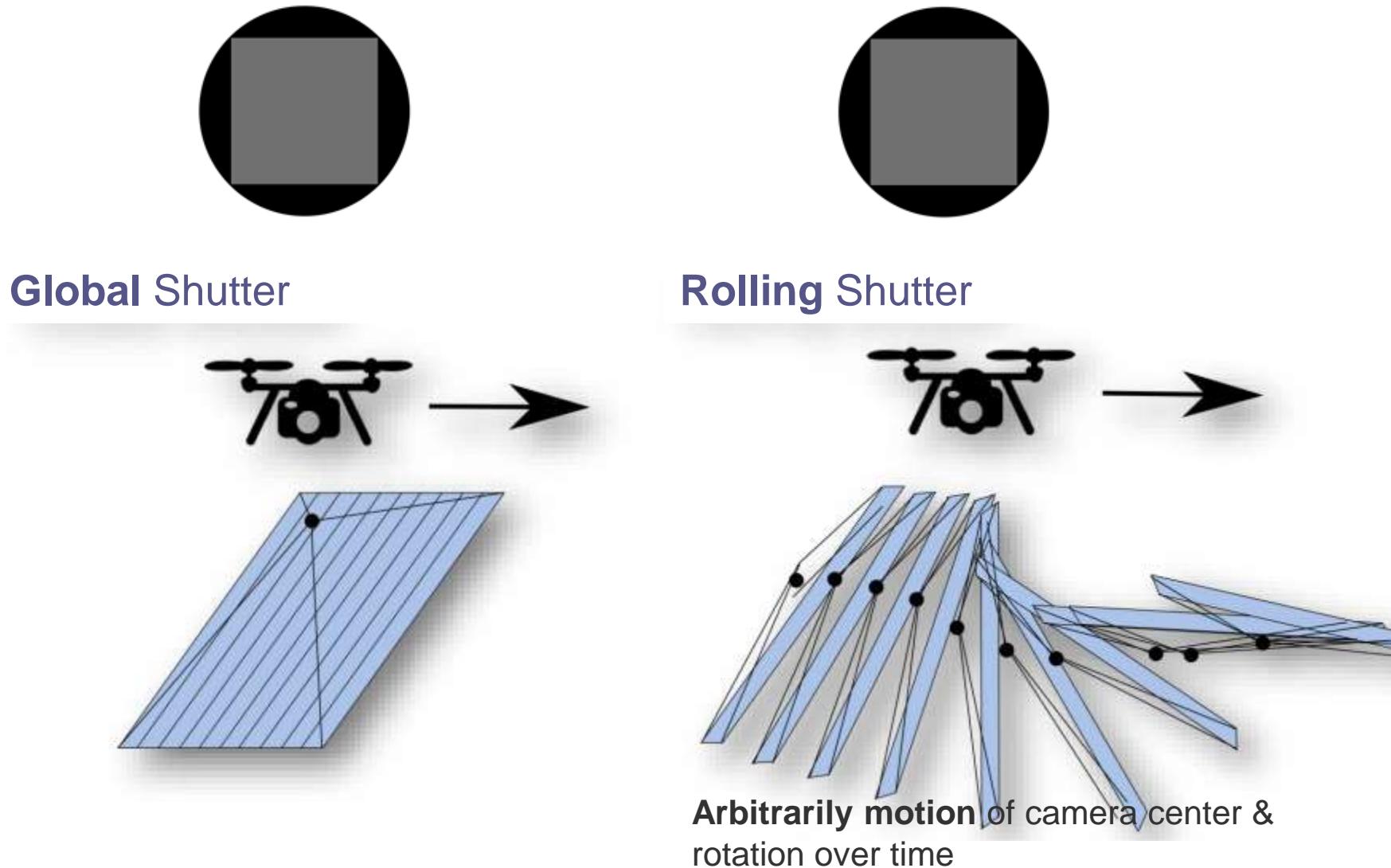
	Leica DMC III	Sony Alpha 7Rii
Image format	CMOS, 102 x 56 mm ² 26112 x 15.000 pix 3.9 µm / 392 MPix	@ CMOS, 36 x 24 mm ² 7952 x 5304 pix 4.5 µm / 42 MPix
Lens	Fixed lens, fix focus lens Multiple lens PAN&MS	interchangeable lens One lens only
Shutter	mechanical, global	electronical, rolling
Motion comp.	FMC (mechanical)	optical image stabilization
Dimensions	622 x 497 x 460 mm	127 x 96 x 60 mm
Weight	60kg (plus 6kg memory)	625g (plus lens)

Rolling shutter – extreme sample



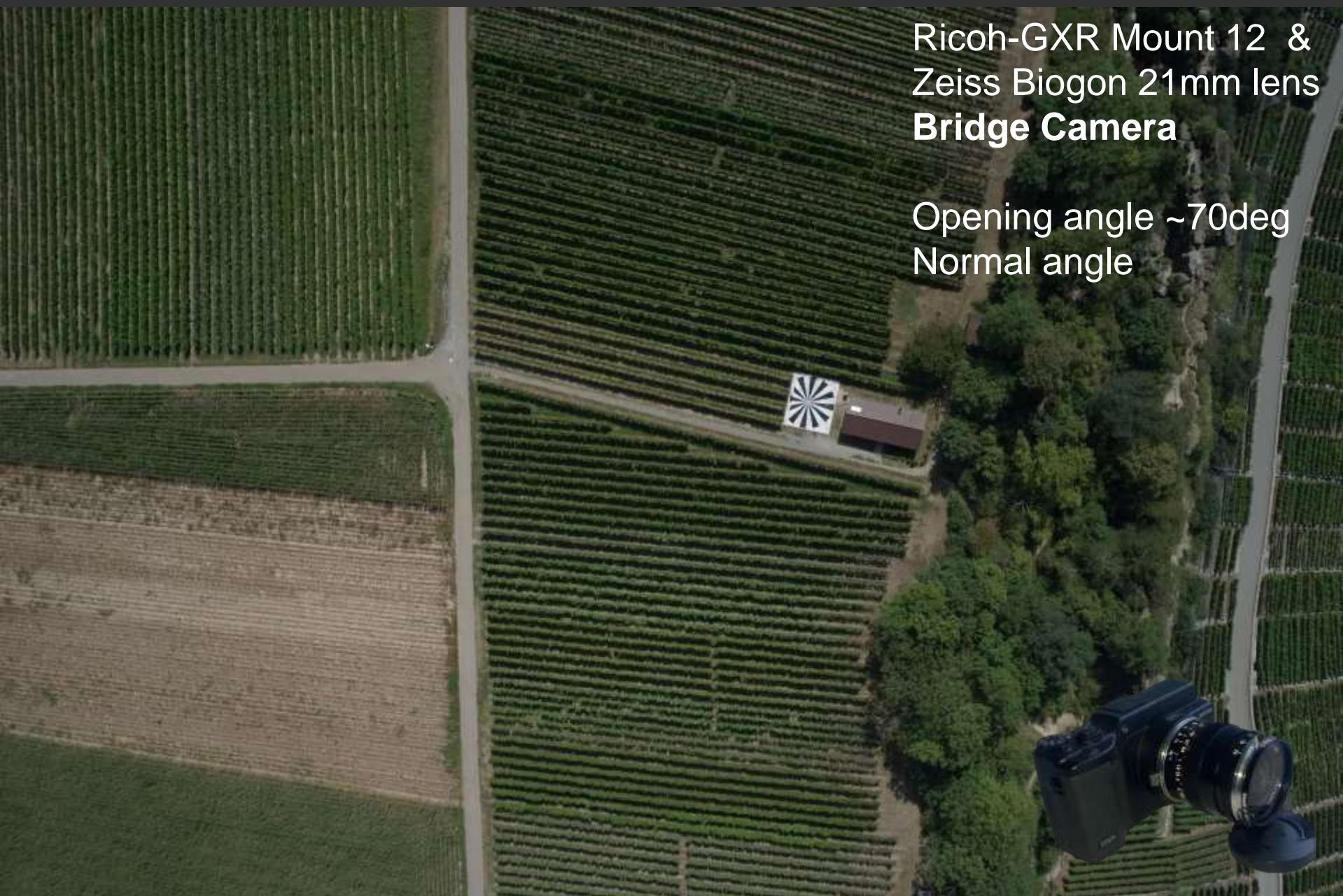
Rolling shutter

Source Pix4D



Ricoh-GXR Mount 12 &
Zeiss Biogon 21mm lens
Bridge Camera

Opening angle ~70deg
Normal angle



Bridge Cam



dIGIcam 50 Mpix
8176 x 6132 pix, 50mm

Opening angle ~63deg
Normal angle

Airborne Cam



Performance of camera – geometric resolution

Comparison Phase One iXM 100 & dji Phantom 4 RTK

DJI Phantom 4 RTK



Samples:
Siemens star
in image center

Phase One iXM-100



	DJI Phantom 4 RTK	Phase One iXM-100
Resolution MTF10	0.563 line/pix	0.769 line/pix
GSD (nom.)	6.9 mm	5.0 mm
tGSD or GRD	12.3 mm	6.5 mm



178 %



130 %

Performance of camera – geometric resolution

Comparison Phase One iXM 100 & dji Phantom 4 RTK

DJI Phantom 4 RTK

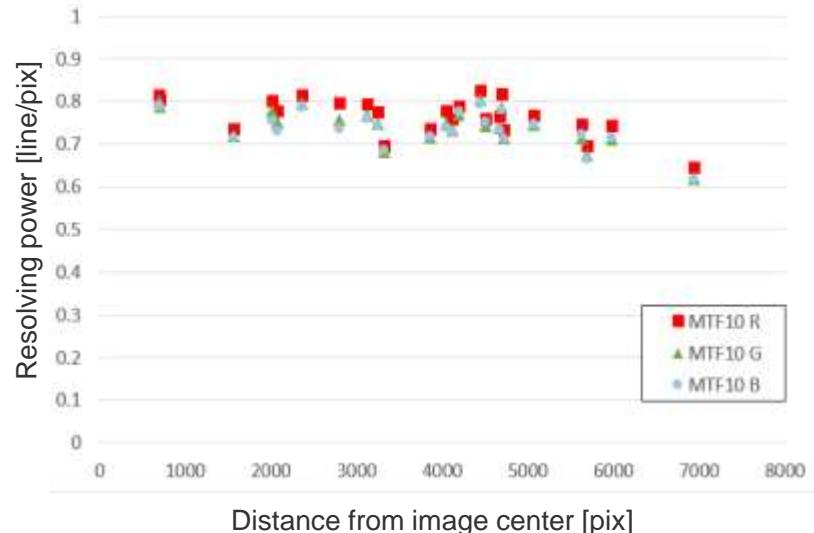
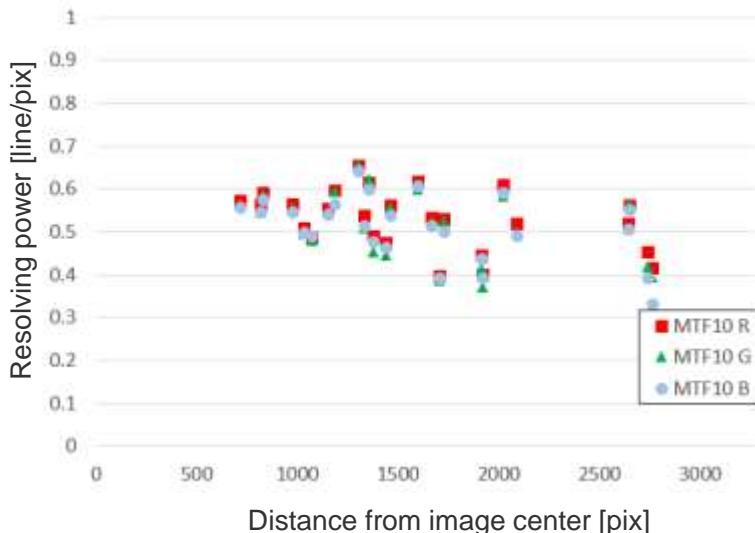


Sensor format $13.2 \times 8.8\text{mm}^2$
Pixel count $5472 \times 3648\text{pix}$
Pixel size $2.41\mu\text{m}$

Phase One iXM-100 / 35mm lens

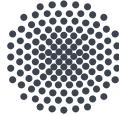


Sensor format $43.9 \times 32.9\text{ mm}^2$
Pixel count $11664 \times 8750\text{pix}$
Pixel size $3.76\mu\text{m}$



Questions & Comments ?!



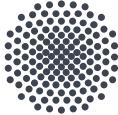


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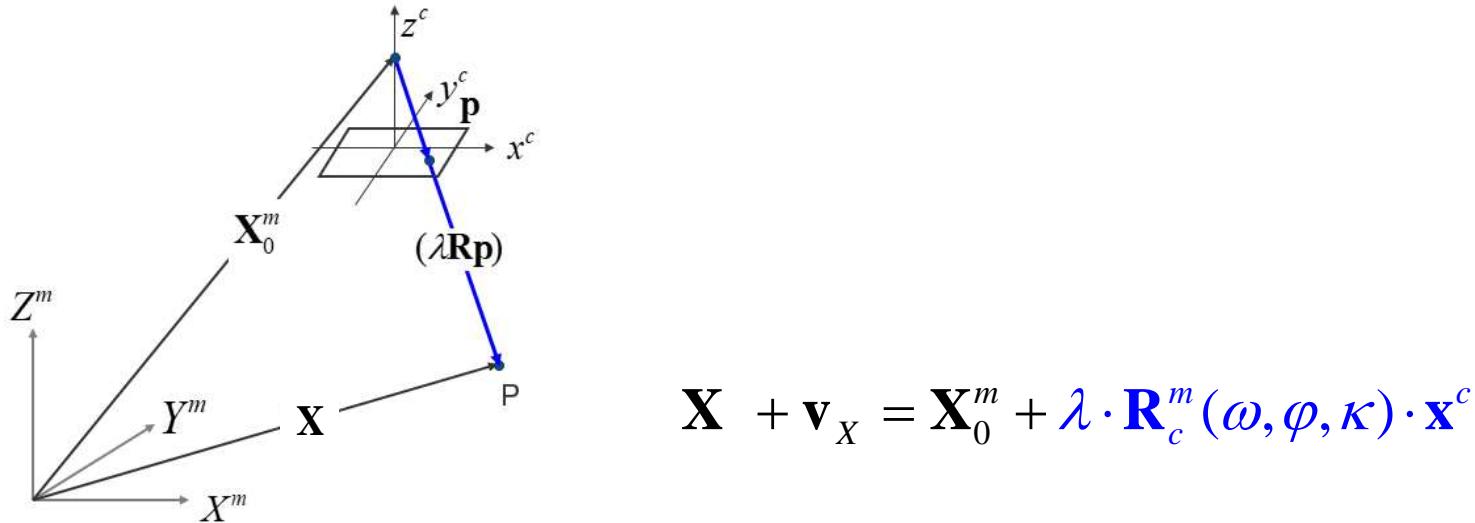


State-of-the-art in UAV-based mapping – an update

“With a little help from my friends” – using GNSS/inertial-based georeferencing for UAV-imagery – case study



Indirect georeferencing

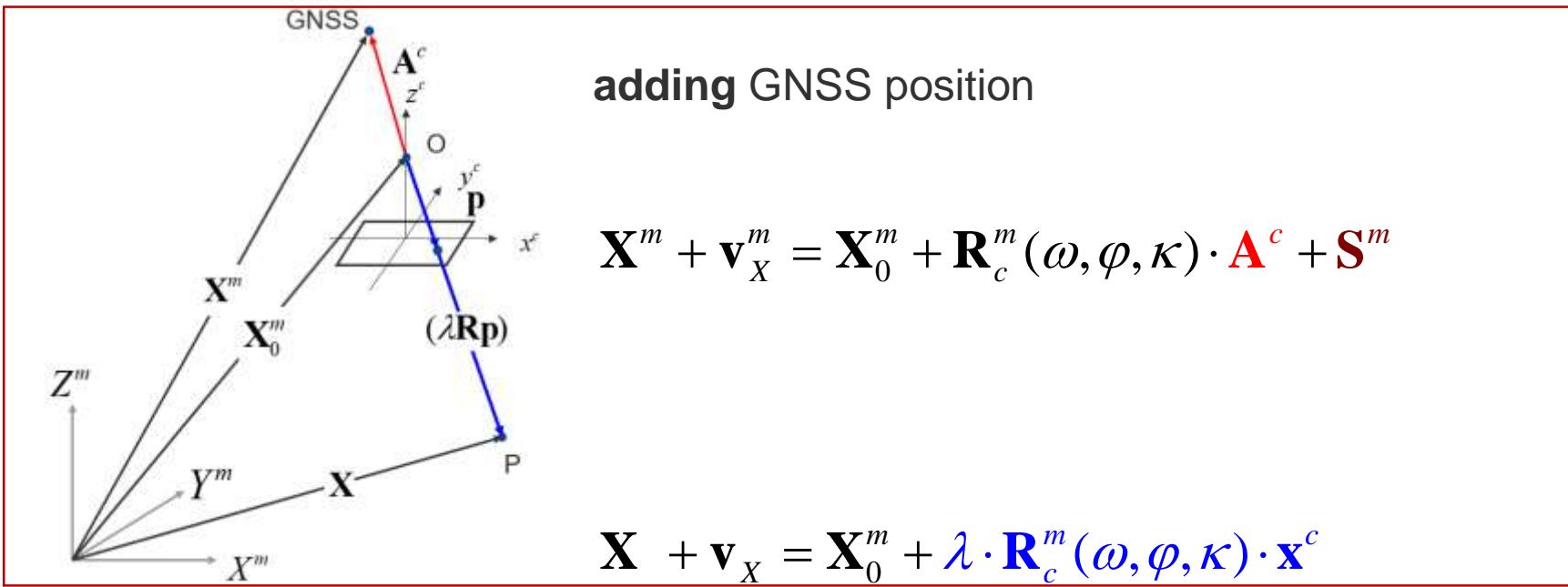


- Collinearity equation

$$x + v_x = x_0 - c \cdot \frac{r_{11}(X - X_0^m) + r_{12}(Y - Y_0^m) + r_{13}(Z - Z_0^m)}{r_{31}(X - X_0^m) + r_{32}(Y - Y_0^m) + r_{33}(Z - Z_0^m)}$$

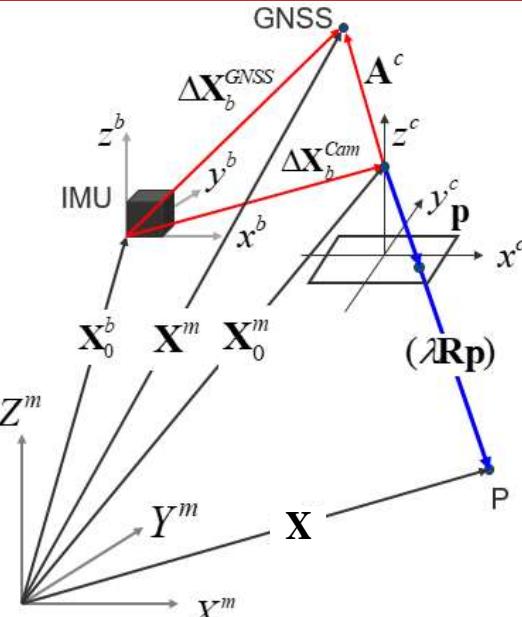
$$y + v_y = y_0 - c \cdot \frac{r_{21}(X - X_0^m) + r_{22}(Y - Y_0^m) + r_{23}(Z - Z_0^m)}{r_{31}(X - X_0^m) + r_{32}(Y - Y_0^m) + r_{33}(Z - Z_0^m)}$$

GNSS-supported AT



- GNSS perspective centre coordinate observations significantly reduce requirements on geodetic measurements, i.e. ground control points
- Position observations are introduced as weighted observations
- System calibration (i.e. **lever arm vector**) is assumed to be known or determinable. Additional systematic **GNSS shifts** can be added to the model

GNSS/inertial-supported AT / Direct georeferencing



adding GNSS/inertial EO parameters

$$\mathbf{X}^m + \mathbf{v}_X^m = \mathbf{X}_0^m + \mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{A}^c + \mathbf{S}^m$$

$$\mathbf{R}_b^m(\Phi + v_\Phi, \Theta + v_\Theta, \Psi + v_\Psi) =$$

$$\mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{R}_b^c(\Delta\omega, \Delta\varphi, \Delta\kappa)$$

$$\mathbf{X}^m + \mathbf{v}_X^m = \mathbf{X}_0^m + \lambda \cdot \mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{x}^c$$

- GNSS / inertial orientation parameters are beneficial - especially for non-standard block configurations, reducing requirements on control and tie points
- EO direct observations are introduced as weighted observations
- But: System calibration (i.e. **Boresight parameter**) is assumed to be known or determinable

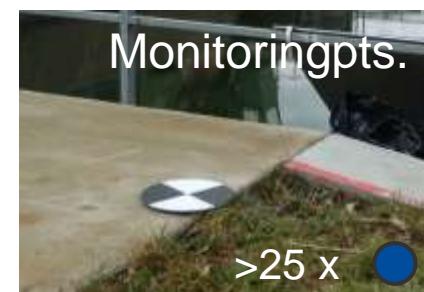
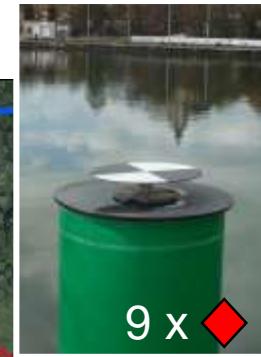
Study: High-precision 3D monitoring for engineering applications

Research project with BfG Koblenz, Testsite Ship-lock Hessigheim



Testsite Ship-lock Hessigheim

(netto) Site extensions: 570 m (EW) x 780 m (NS)



UAV – Systems: Photogrammetry and LiDAR

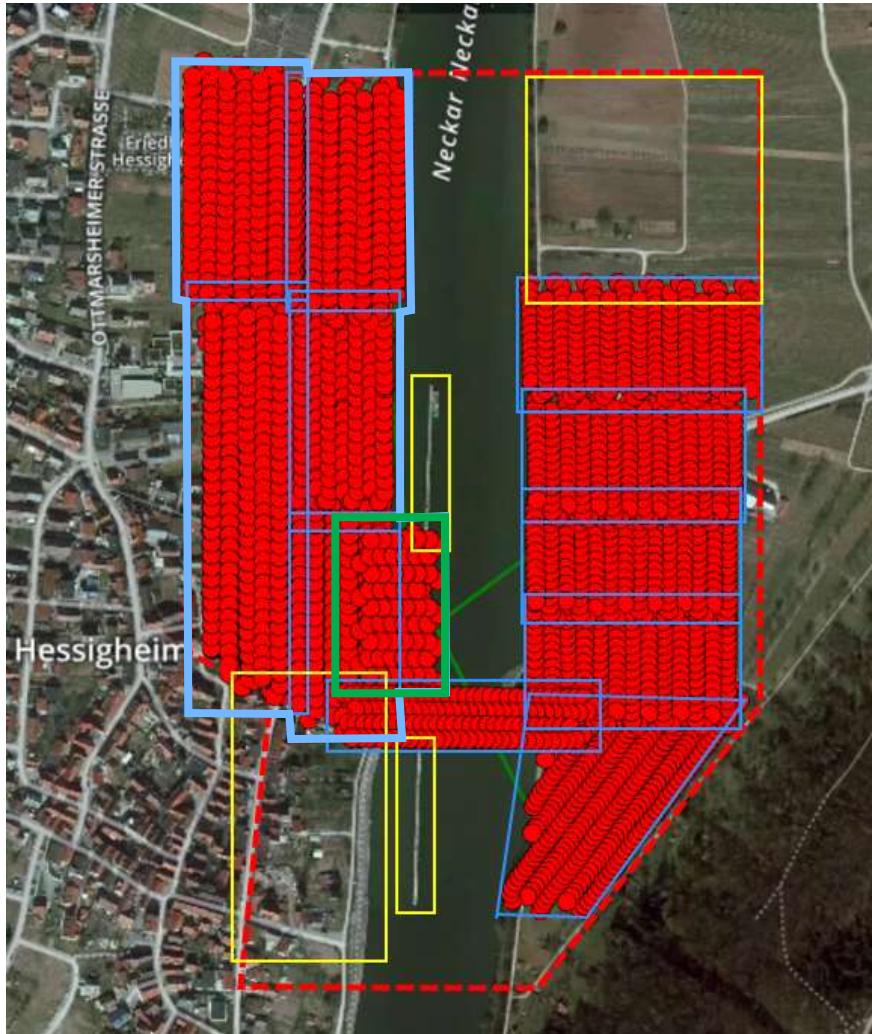


- *CopterSystems* with PhaseOne IXM 100, 35 mm lens (**fix-focussed** to 60m distance, **lab-calibrated**)
- Mean flying height: 40 m above ground
Mean nom. **GSD: 4.0 mm**
Image area coverage: ~50 x 35 m²
- Accuracy: scaleable (1 pix aspired)

- *Geografie Uni Innsbruck RiCOPTER* with VUX-1UAV
- Point density: 300-400 Pts/m² (per strip), 800 Pts/m² (multiple overlap)
 - 2 x Sony Alpha 6000 camera oblique
 - 1.2-2.0 cm GSD
- Accuracy: 10 mm, Precision: 5 mm

Testsite Ship-lock Hessigheim

Flightcampaign November 2018



- PhaseOne IXM 100 data
 - 12(+2) flights @ 2 days
 - Calibration flight (cross pattern) at first day (ship lock)
 - Quite poor sunlight / weather as flights were done Nov 14/15 2018 – no full coverage of test site possible

• Calibration block

- 148 images, 2 different flight heights (40m & 50m)

• Block Western-Shore

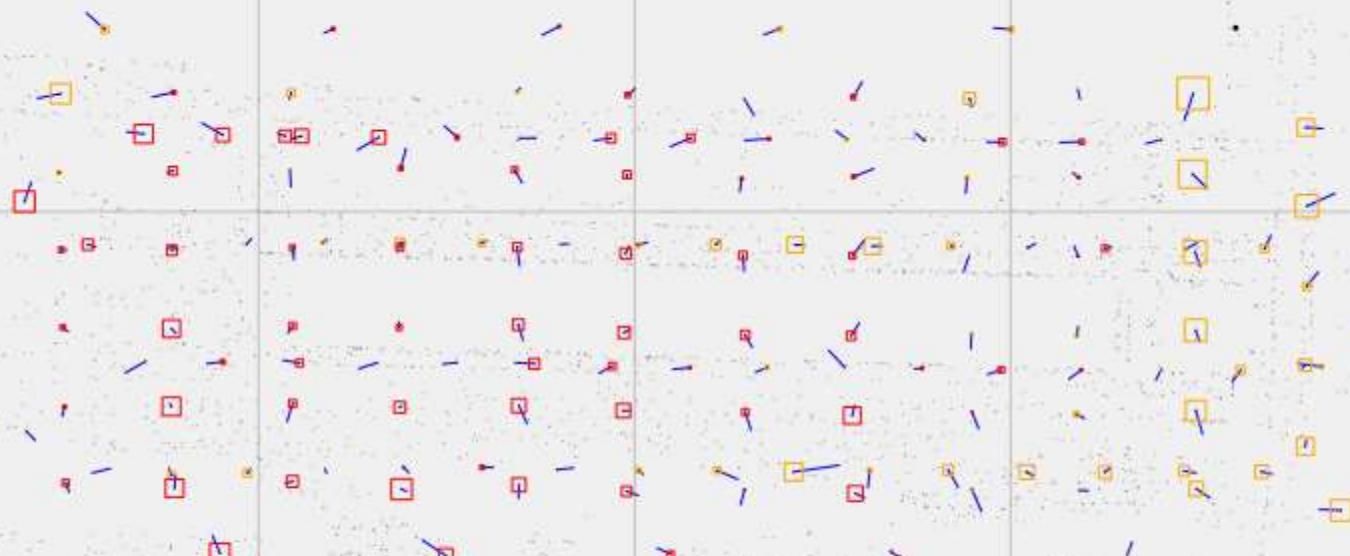
- 1037 (909) Images, one flight height (40m)

Calibration block: Performance of GNSS/inertial EO

UAV-application of Applanix APX-15 EI UAV

Residuals at camera positions (observed – adjusted)

[10 cm]



RMS values for GNSS/inertial position

Manufacturers specs

X [m]	nom.	Y [m]	nom.	Z [m]	nom.	Total [m]
0.0114	0.03m	0.0164	0.03m	0.0121	0.03m	0.0234

542

542

542

542

542

Calibration block: Performance of GNSS/inertial EO

UAV-application of Applanix APX-15 EI UAV

Residuals at camera **attitude** (observed – adjusted)

[0.85°]



514050

514000

RMS values for **GNSS/inertial attitude**

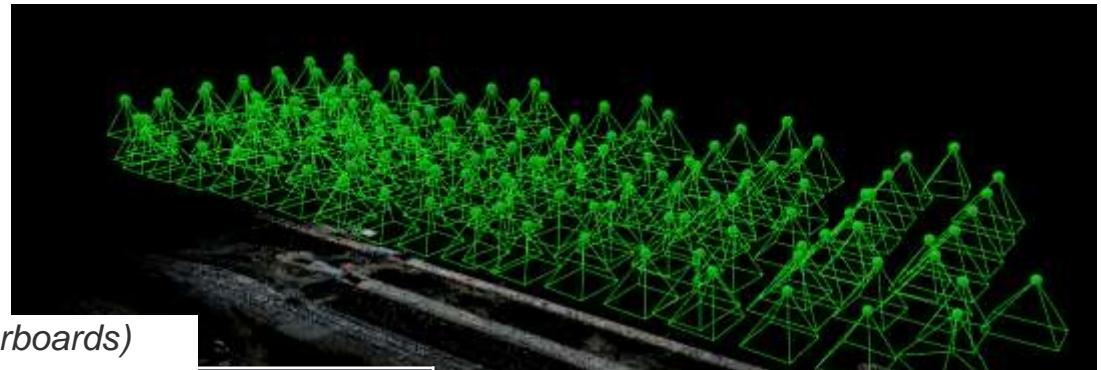
Manufacturers specs

Omega (X) [deg]	nom.	Phi (Y) [deg]	nom.	Kappa (Z) [deg]	nom.
0.016330	0.025deg	0.017240	0.025deg	0.293340	0.08deg



Calibration block: cross-pattern / ship-lock

Impact of GNSS- and GNSS/inertial-data in AT



Residuals from 20 Checkpoints (Checkerboards)

	AT with GCPs only	GNSS- supported AT	GNSS/inertial- supported AT
Sigma0 [pix]	0.5815	0.5791	0.6691
RMS X [m]	0.0016	0.0014	0.0014
Y [m]	0.0021	0.0023	0.0022
Z [m]	0.0042	0.0019	0.0018
STD X [m]	0.0016	0.0014	0.0014
Y [m]	0.0021	0.0021	0.0022
Z [m]	0.0038	0.0019	0.0017
Max. X [m]	-0.0030	-0.0026	0.0033
Y [m]	-0.0062	-0.0067	-0.0056
Z [m]	0.0110	-0.0035	0.0034



3 GCPs (Pillars)
20 CHPs

Calibration block: cross-pattern / ship-lock

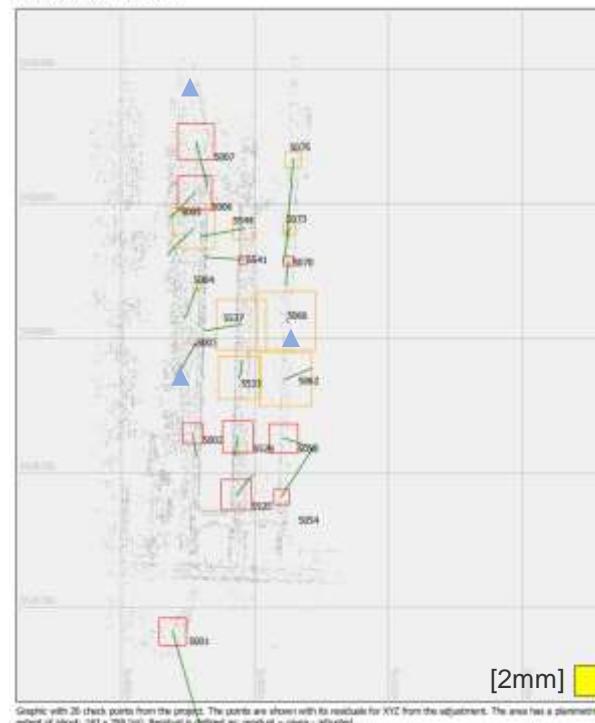
Impact of GNSS- and GNSS/inertial data in AT

Check point residuals



AT with GCPs only

Check point residuals



GNSS-supported AT

Check point residuals

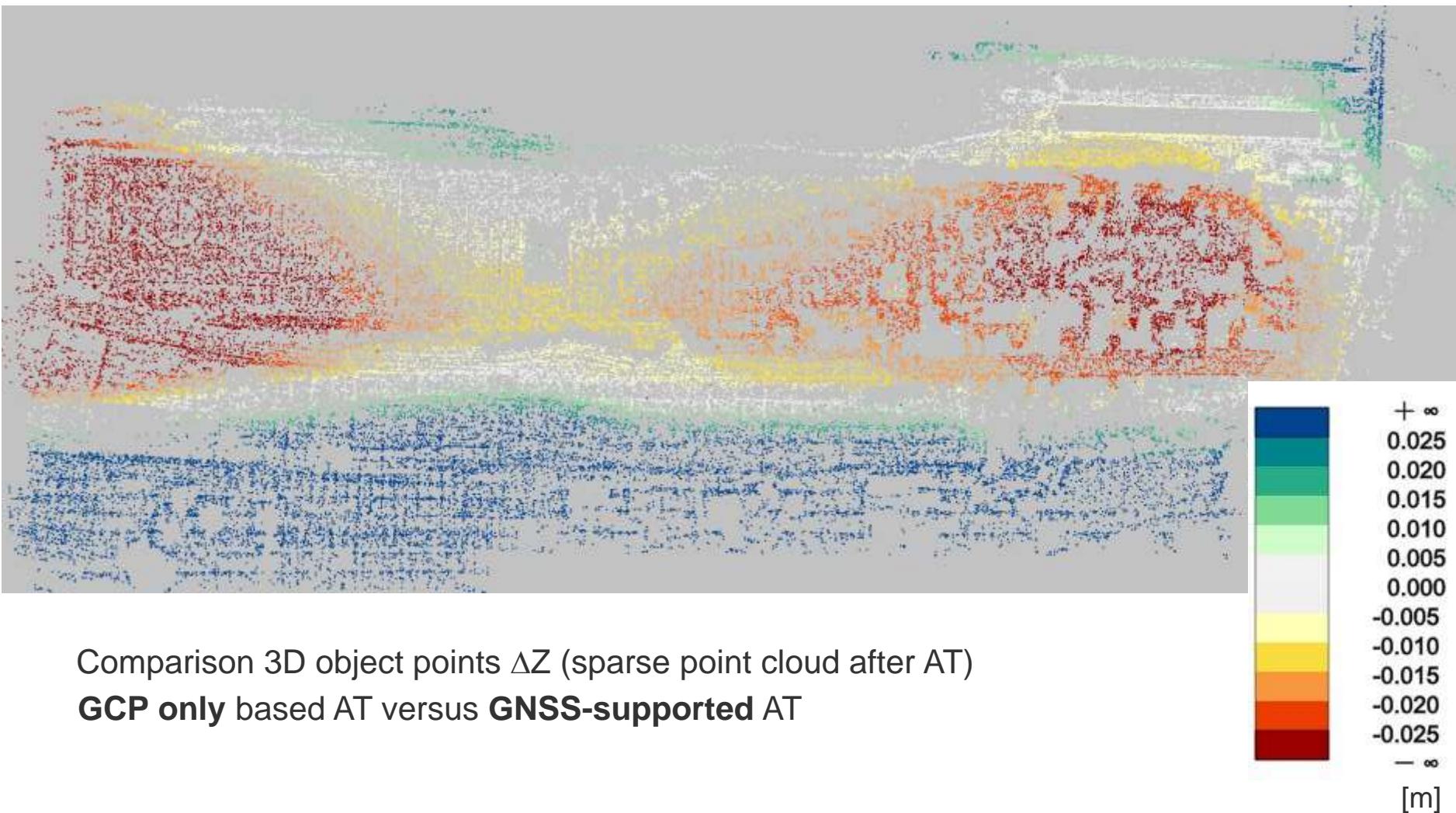


GNSS/inertial-supported AT

3 GCPs (Pillars)
20 CHPs

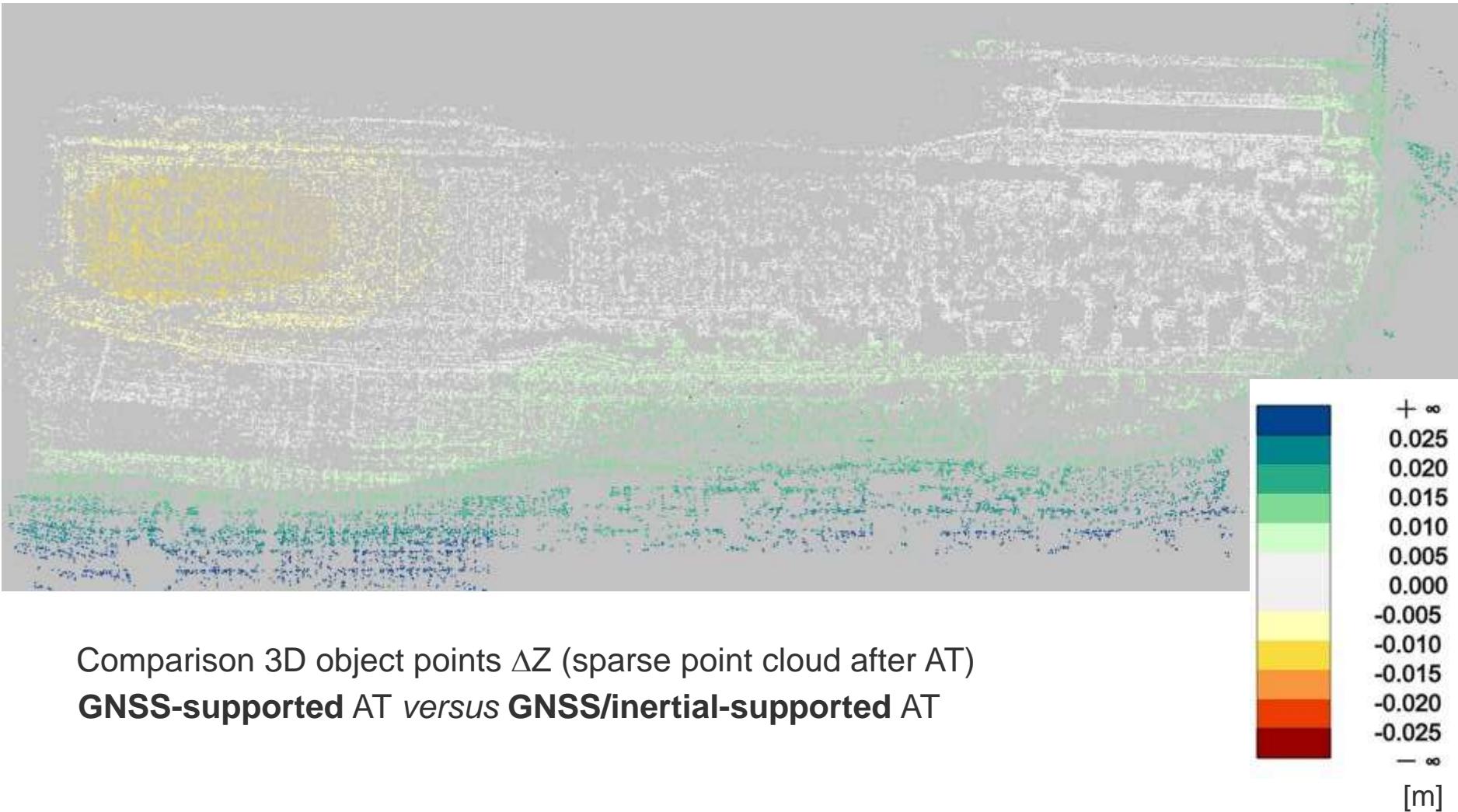
Block Western-Shore

Impact of GNSS- and GNSS/inertial data in AT



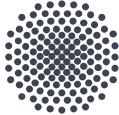
Block Western-Shore

Impact of GNSS- and GNSS/inertial data in AT

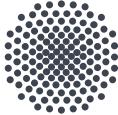


Questions & Comments ?!





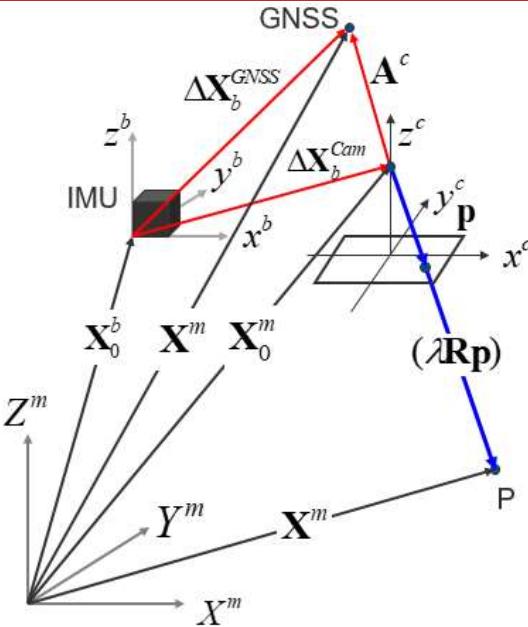
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State-of-the-art in UAV-based mapping – an update

**Are there any further updates
in georeferencing of UAV-
based imagery?**

Direct Georeferencing / Integr. sensor orientation



absolut GNSS/inertial EO elements

$$\mathbf{X}^m + \mathbf{v}_X^m = \mathbf{X}_0^m + \mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{A}^c + \mathbf{S}^m$$

$$\mathbf{R}_b^m(\Phi + v_\Phi, \Theta + v_\Theta, \Psi + v_\Psi) =$$

$$\mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{R}_b^c(\Delta\omega, \Delta\varphi, \Delta\kappa)$$

- GNSS / inertial orientation parameters are **beneficial** - especially for such corridor flights, reducing requirements on link points (and points)
- **But:** System calibration (generally Boresight parameter) is assumed to be known or determinable
- **Problem:** **corridor only provides poor geometry**, in addition camera might no be really **stable** over time

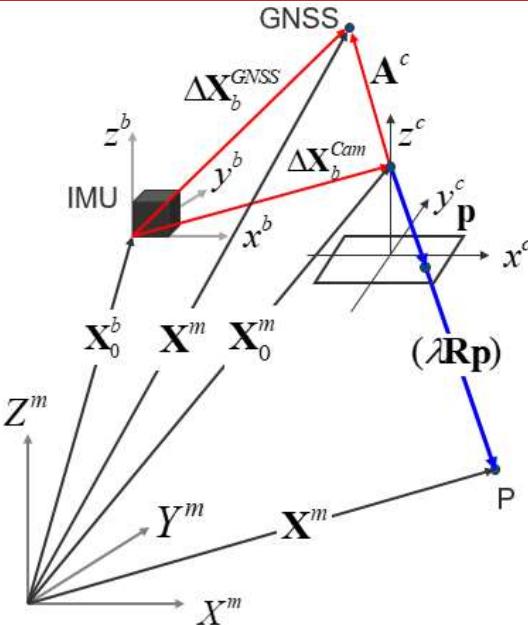
GNSS/inertial supported AT using **relative** GNSS/inertial EO parameters

Why using **relative** GNSS/inertial EO params / relative aerial control?

- **elimination of system calibration parameters;**
- stationary time dependent stochastic model for relative aerial control;
- new photogrammetric observation models for horizontal map-projected coordinates;
- new attitude aerial control observation models to avoid re-parameterisations steps

M. Blázquez & I. Colomina (2012): Relative INS/GNSS aerial control in integrated sensor orientation: Models and performance, in ISPRS Journal of Photogrammetry and Remote Sensing 67 (2012) 120–133
<http://doi.org/10.1016/j.isprsjprs.2011.11.003>

AT using relative aerial control



absolut GNSS/inertial EO elements

$$\mathbf{X}^m + \mathbf{v}_X^m = \mathbf{X}_0^m + \mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{A}^c + \mathbf{S}^m$$

$$\mathbf{R}_b^m(\Phi + v_\Phi, \Theta + v_\Theta, \Psi + v_\Psi) =$$

$$\mathbf{R}_c^m(\omega, \varphi, \kappa) \cdot \mathbf{R}_b^c(\Delta\omega, \Delta\varphi, \Delta\kappa)$$

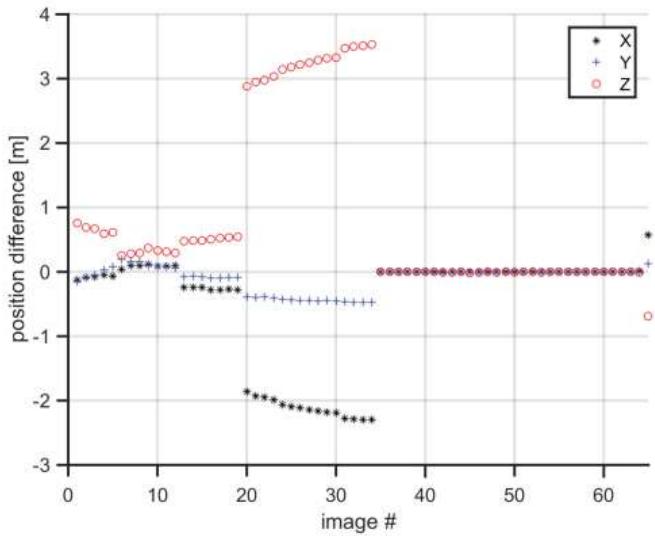
- AT using **relative** GNSS/inertial aerial control exterior orientation elements
 - Boresight-Alignment $\mathbf{R}_b^c(\Delta\omega, \Delta\varphi, \Delta\kappa)$ is eliminated

$$\Delta\mathbf{X}^m(t_{ij}) + \mathbf{v}_{\Delta X}^m = (\mathbf{X}_0^m(t_j) - \mathbf{X}_0^m(t_i)) + (\mathbf{R}_c^m(\omega_{t_j}, \varphi_{t_j}, \kappa_{t_j}) - \mathbf{R}_c^m(\omega_{t_i}, \varphi_{t_i}, \kappa_{t_i})) \cdot \mathbf{A}^c$$

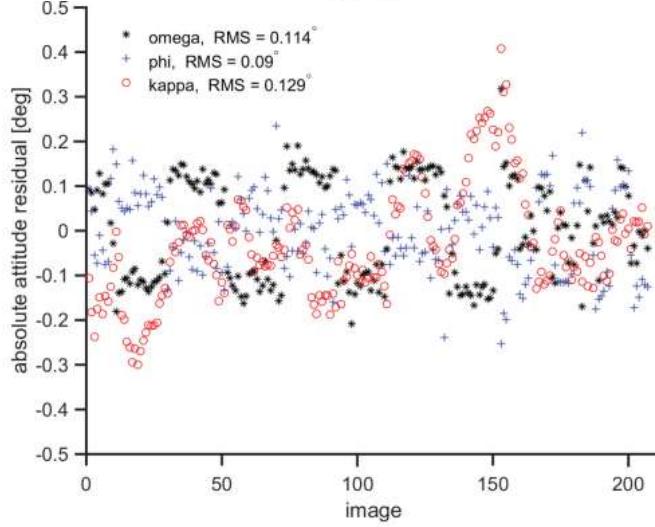
$$\Delta\mathbf{R}_b^m(\Phi_{t_{ij}} + v_\Phi, \Theta_{t_{ij}} + v_\Theta, \Psi_{t_{ij}} + v_\Psi) = \mathbf{R}_c^m(\omega_{t_j}, \varphi_{t_j}, \kappa_{t_j}) \cdot \mathbf{R}_m^c(\omega_{t_i}, \varphi_{t_i}, \kappa_{t_i})$$

AT with relative GNSS/inertial aerial control

Position

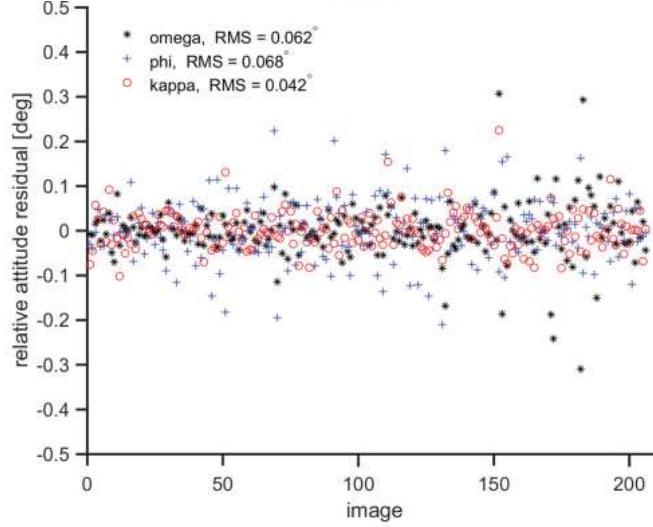
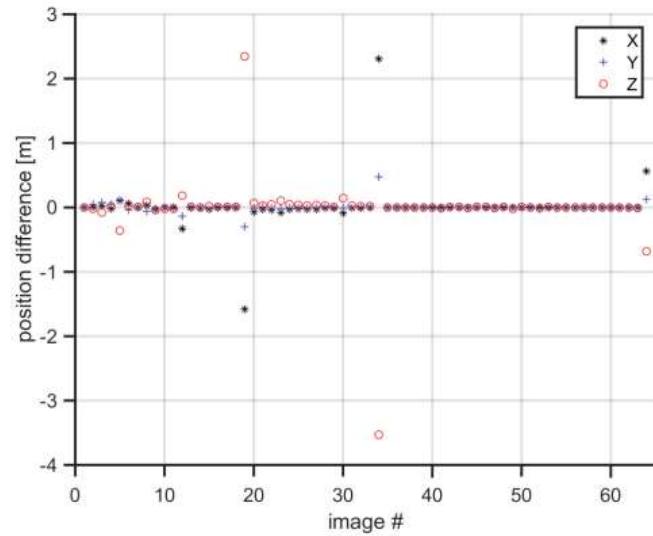


Attitude



absolute

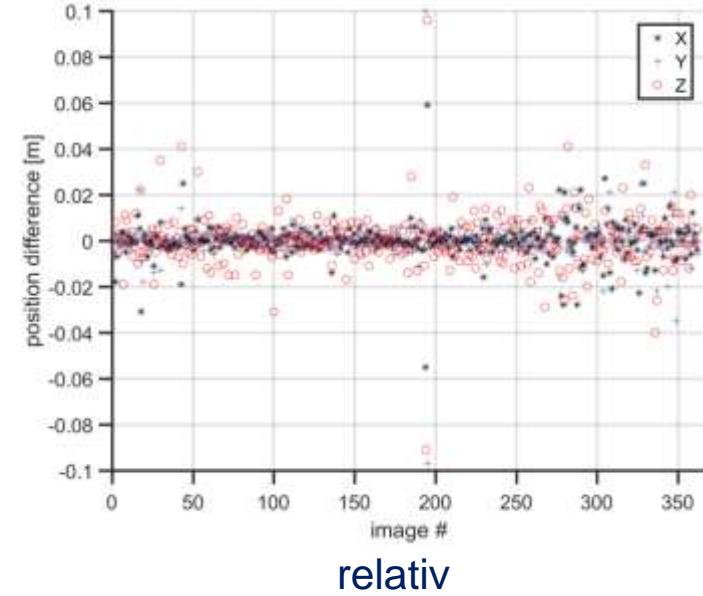
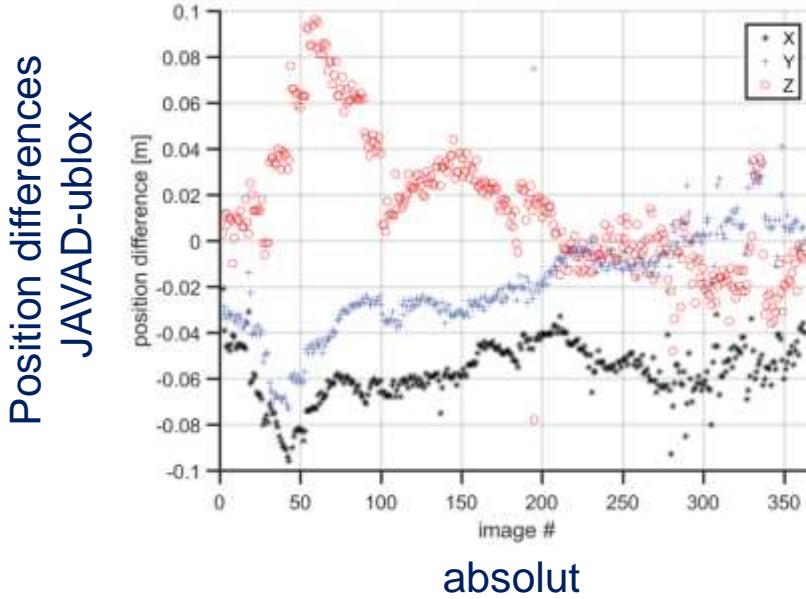
© Rehak, 2017



relative

AT with relative GNSS/inertial aerial control

Influence of GNSS receiver quality (positioning only)



© Rehak, 2017



JAVAD

Javad OEM TR-G3T

216 channels each of GPS L1/L2/L2C/L5,
GLONASS L1/L2, Galileo E1/E5A

Costs: ~10000 \$US



ublox

ublox NEO-8T

72-channel u-blox GPS/QZSS L1 C/A,
GLONASS L10F, BeiDou B1

Costs: ~75 \$US

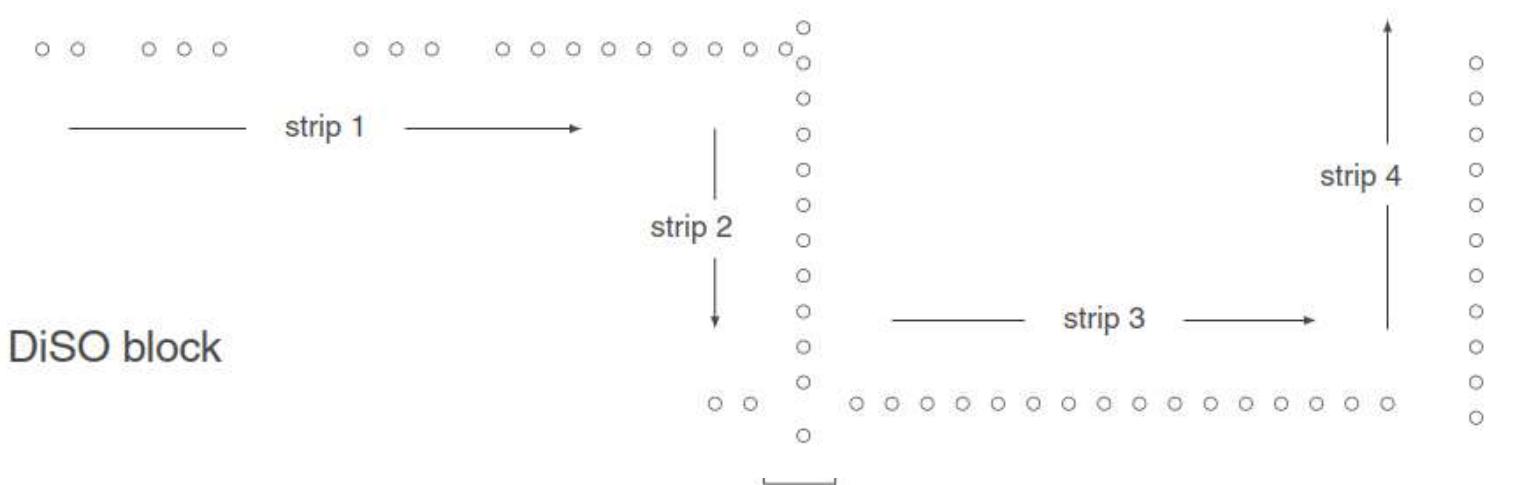
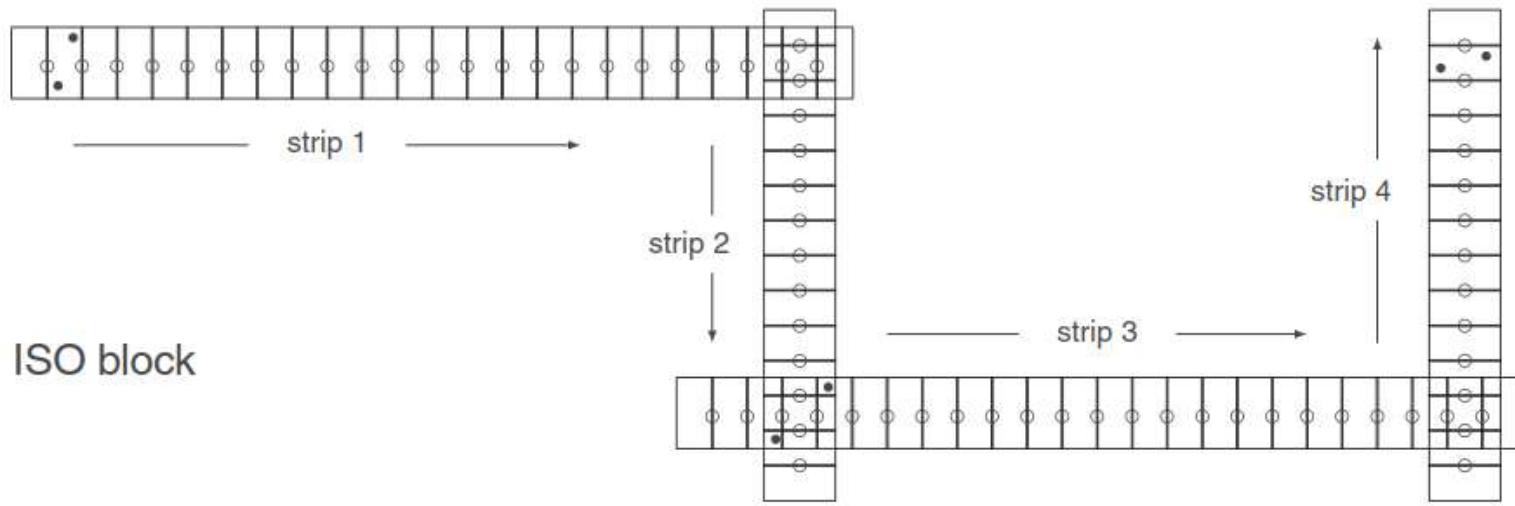
Fast AT

- particular case of Integrated Sensor Orientation (GNSS/inertial supported AT) characterized by the use of the following observations:
 - (time) position and attitude (tPA) **aerial control observations**, either in the absolute or relative mode, for all images.
 - Ground control point observations for a **limited number (in principle) of points** and images.
 - Image coordinate observations **for the ground control points only**.

M. Blázquez & I. Colomina (2012): Fast AT: A simple procedure for quasi direct orientation, in ISPRS Journal of Photogrammetry and Remote Sensing 71 (2012) 1–11

<https://doi.org/10.1016/j.isprsjprs.2012.04.005>

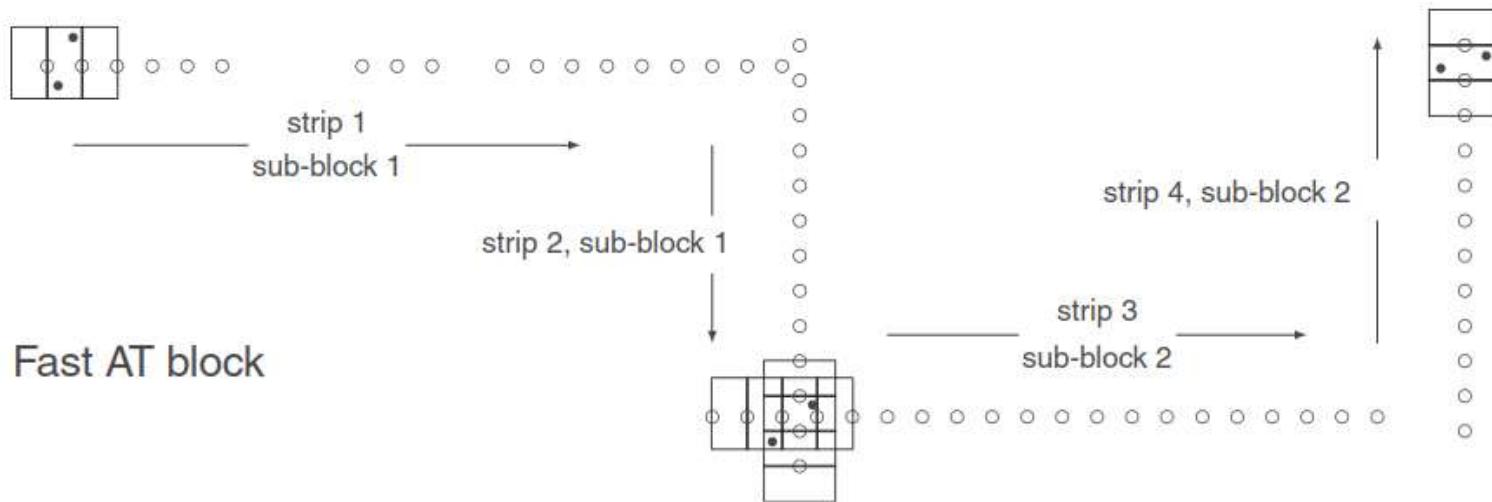
Fast AT Conceptual Layout (1/2)



© Blázquez & Colomina, 2012

Fast AT

Conceptual Layout (2/2)



- image and tPA aerial control point
- ground control point (GCP)



image with one or more
photo-measurements

ISO block: $23 + 13 + 23 + 13$ images, photo-measurements in all images.

Fast AT block: $18 + 13 + 20 + 12$ images, photo-measurements in $2 + 3 + 3 + 2$ images, 2 sub-blocks,
no need for image/strip overlap, image overlap recommended in areas with GCPs.

DiSO block: $17 + 12 + 18 + 11$ images, no photo-measurements, no GCPs, no overlap requirements.

Fast AT

Observations of InSO, ISO, Fast AT and DiSO.

InSO – indirect sensor orientation

ISO – integrated sensor orientation

DiSO – direct sensor orientation

Observations	InSO	ISO	Fast AT	DiSO
tPA/tPVA aerial control	NO	YES	YES	YES
Ground control points	YES many	YES few	YES few	NO
Image coordinates	YES many	YES many	YES few	NO

Properties of ISO, Fast AT and DiSO.

Properties	ISO	Fast AT	DiSO
Precision	+	0	0
Accuracy	+	0	—
Reliability	+	0	—
Cost	+	0	—
Time	+	0	—

+: high. 0: average. -: low.

© Blázquez & Colomina, 2012

Fast AT potential applications

- Fast AT is of interest in situations where ISO (integrated sensor orientation) is not feasible or required and where DiSO (direct sensor orientation) is not accurate or reliable enough. For instance, it may be used for ill-textured areas where image matching is difficult.
- for applications currently relying on DiSO, where accuracy and reliability matter, and that, for some reason, cannot afford the time and/or cost required by ISO and where the measurement or use of existing GCPs makes sense in the context of the application.
- can be used in combination with standard ISO procedures: **One possibility is that a small ISO block be used for camera calibration and that Fast AT be applied for larger blocks. The data acquisition for the ISO block can take place at any time, before, in between or after the Fast AT blocks are acquired. Sensor calibration parameters can then be computed and, later on, used as constants or observations in the Fast AT blocks.**

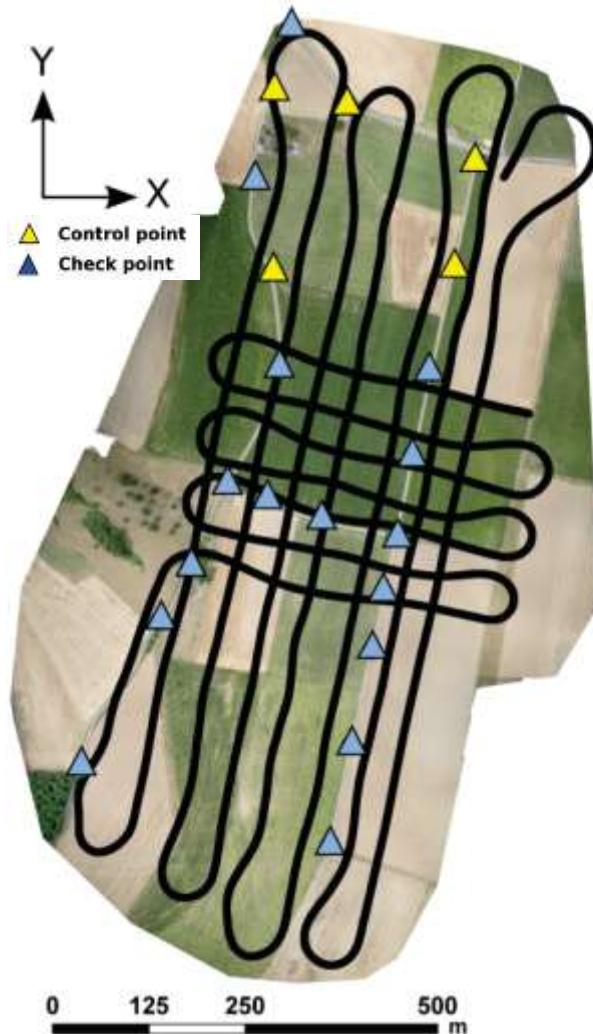
Fast AT

Mathematical Model

- Fast AT concept is independent from the particular functional models of the observation equations for image coordinate, ground control and aerial control observations
- Both possibilities
 - Classical use of EOP observations: classical use of tPA aerial control observations in absolute mode, **absolute Fast AT**
 - Relative use of EOP observations: use of tPA aerial control observations in relative mode, **relative Fast AT**

Influence of Aerial Control on Mapping Accuracy

Block Scenario



UAV test scenario

fixed-wing UAV with Sony NEX-5R,
GSD 4.5cm, 7 + 7 flight lines,
207 images, 80% / 60% overlap
20 GCP / ChP (notice distribution!)

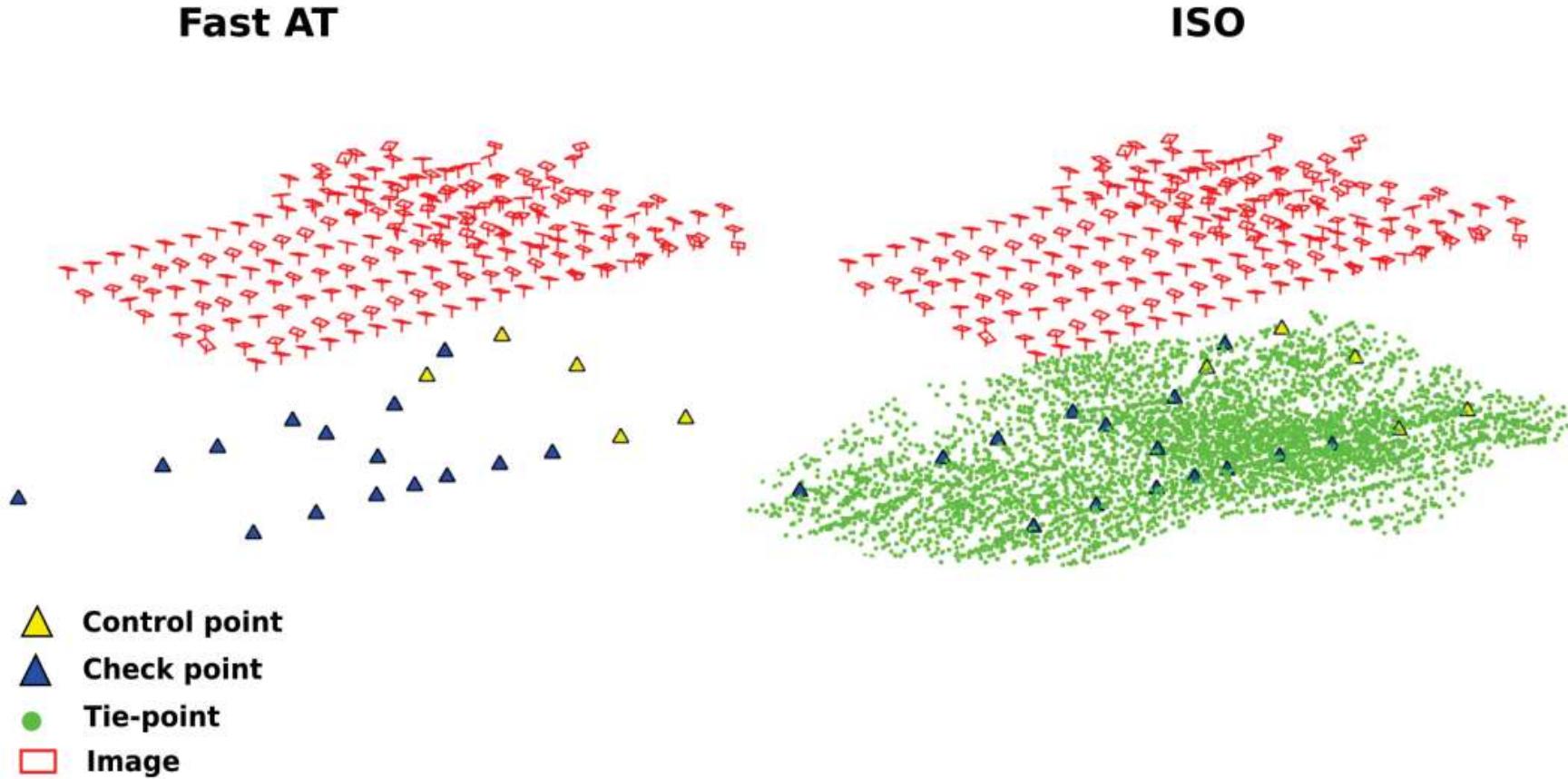
Test scenarios

Mode	Position cont.	Attitude cont.	Camera cal.	Boresight
Indirect SO	-	-	Yes	-
ISO	Absolute	Absolute	Yes	Known
ISO	Absolute	Relative	Yes	-
ISO	Relative	Relative	Yes	-
Fast AT	Absolute	Absolute	No	Known
Fast AT	Absolute	Relative	No	-
Fast AT	Relative	Relative	No	-
DiSO	Absolute	Absolute	No	Known

© Rehak, 2017

Influence of Aerial Control on Mapping Accuracy

Block Scenario



Distribution of GCPs, ChPs, and tie-points

© Rehak, 2017

UAV test using GNSS/inertial aerial control

Test	Mean [cm]			RMS [cm] / [pix]		
	X	Y	Z	X	Y	Z
Indirect Georeferencing (5 GCP / 15 ChP) with CamCal	6.8	0.8	-66.4	1.6 0.4	14.5 3.2	117.1 26.0
Direct Georeferencing (0 GCP / 20 ChP) with Boresight, no CamCal	-0.5	-1.3	-1.5	5.2 1.2	6.3 1.4	16.6 3.7
Integrated SO + abs. Position + abs. Attitude (5 GCP / 15 ChP) with Boresight & CamCal	0.6	1.6	3.5	3.2 0.7	2.9 0.6	5.3 1.2
Integrated SO + abs. Position + rel. Attitude (5 GCP / 15 ChP) no Boresight, with CamCal	-0.5	1.6	3.2	3.0 0.7	2.8 0.6	5.3 1.2
Fast AT + abs. Position + rel. Attitude (5 GCP / 15 ChP) no Boresight, no CamCal	0.8	3.8	-2.2	2.4 0.5	4.7 1.0	6.1 1.4

© Rehak, 2017

GSD: 4.5 cm

Hybrid LiDAR & image sensor orientation

Hybrid sensors

Leica Citymapper-2



RiCOPTER (Skyability Siegendorf A)

Standard georeferencing approach:

- each data stream processed / georeferenced independently, i.e.
 - LiDAR strip adjustment
 - Photogrammetric bundle
- Sensors moved on the same trajectory **not** considered!



AeroSpector (GGS Speyer)

Hybrid LiDAR & image sensor orientation

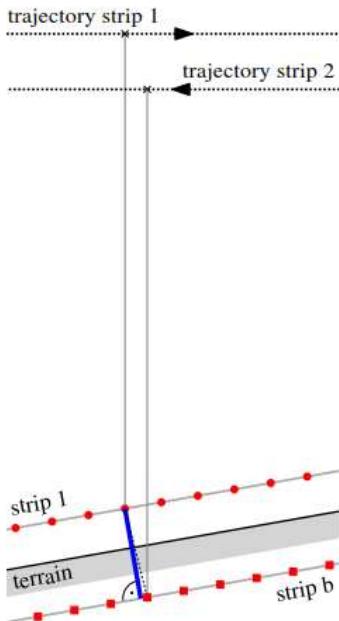
- simultaneously optimization of the relative orientation and absolute orientation (georeference) of the lidar and image data.
- sensor orientations are optimized by minimizing the discrepancies
 - (1) within the overlap area of flight strips and/or images and
 - (2) with respect to ground truth if available
- rigorous modelling using the original measurements of the sensors (i.e. scanner: polar measurements, camera: image coordinates) and the flight trajectory of the aircraft

Glira, P., Pfeifer, N., and Mandlburger, G.: HYBRID ORIENTATION OF AIRBORNE LIDAR POINT CLOUDS AND AERIAL IMAGES, ISPRS Ann. Photogramm. Remote Sens. Spatial Inf. Sci., IV-2/W5, 567-574, <https://doi.org/10.5194/isprs-annals-IV-2-W5-567-2019>, 2019.

© Glira et al., 2019

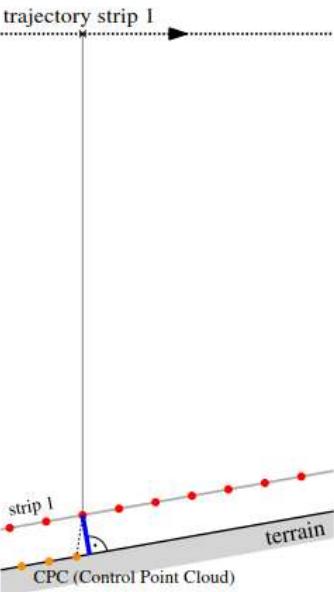
Hybrid LiDAR & image sensor orientation

Correspondence between
two lidar strips
(STR-to-STR)



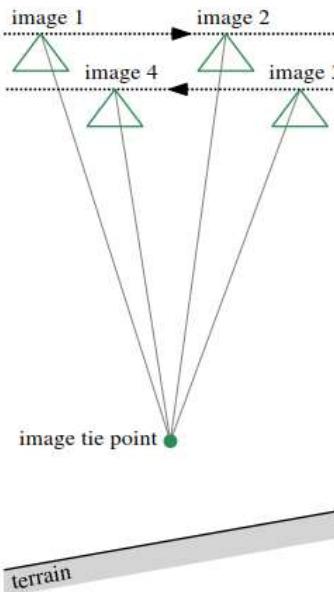
Minimization of
point-to-plane distance
in object space

Correspondence between
CPC and lidar strip
(CPC-to-STR)



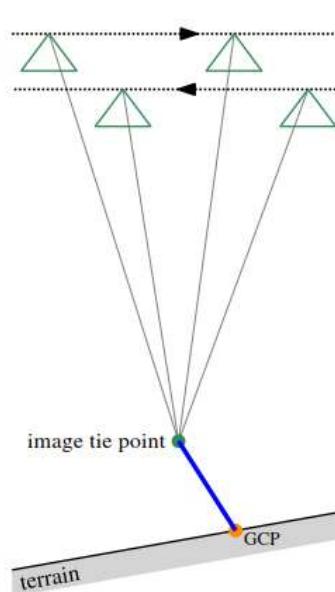
Minimization of
point-to-plane distance
in object space

Correspondence between
images (tie point)
(IMG-to-IMG)



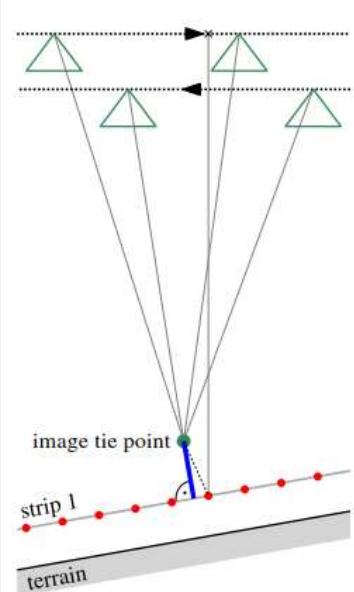
Minimization of
reprojection error
in image space

Correspondence between
tie point and GCP
(IMG-to-GCP)



Minimization of
point-to-point distance
in object space

Correspondence between
tie point and lidar strip
(IMG-to-STR)



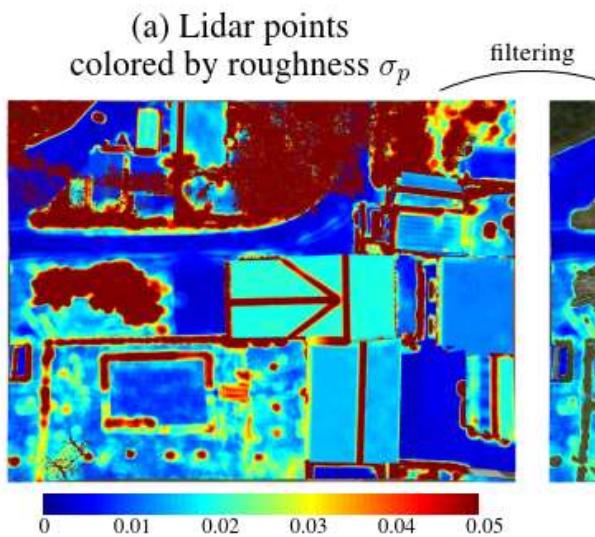
Minimization of
point-to-plane distance
in object space

strip adjustment of lidar point clouds

aerial triangulation

hybrid adjustment: lidar strip adjustment + aerial triangulation

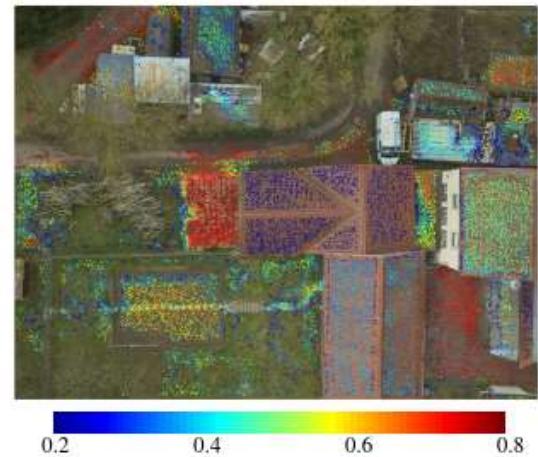
Hybrid LiDAR & image sensor orientation



(b) Image tie points

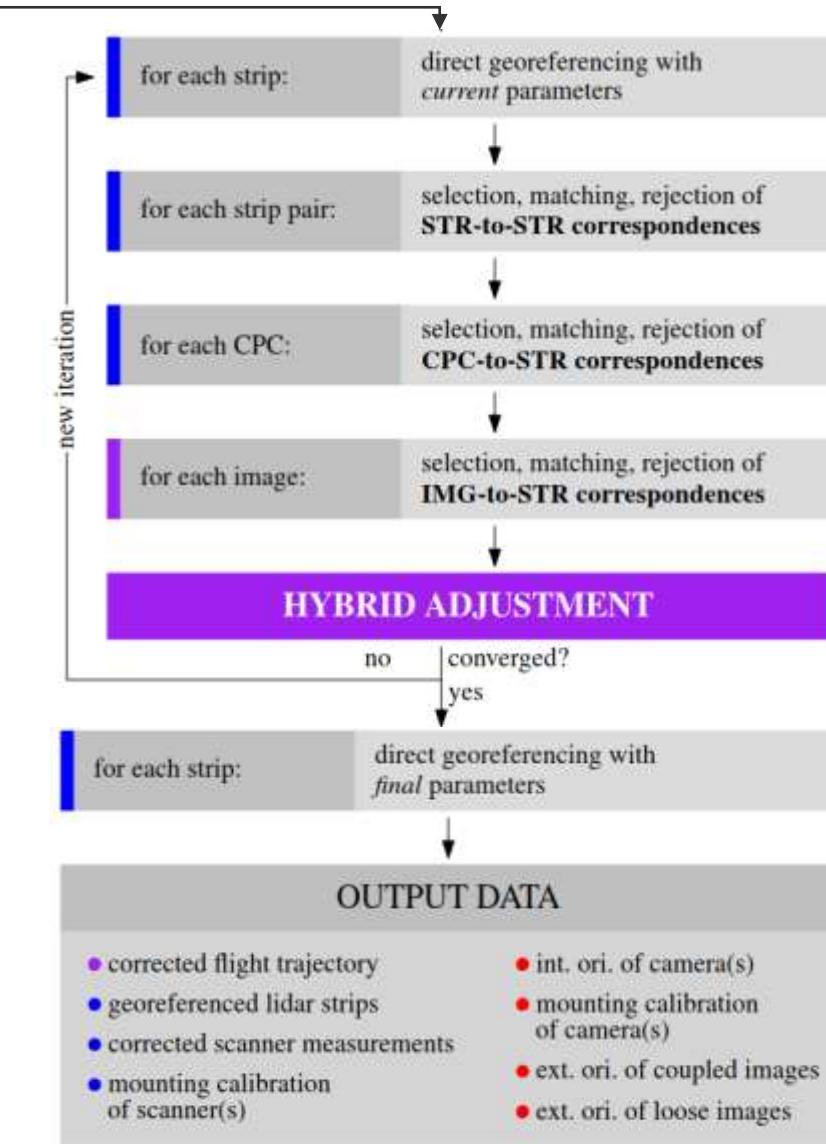
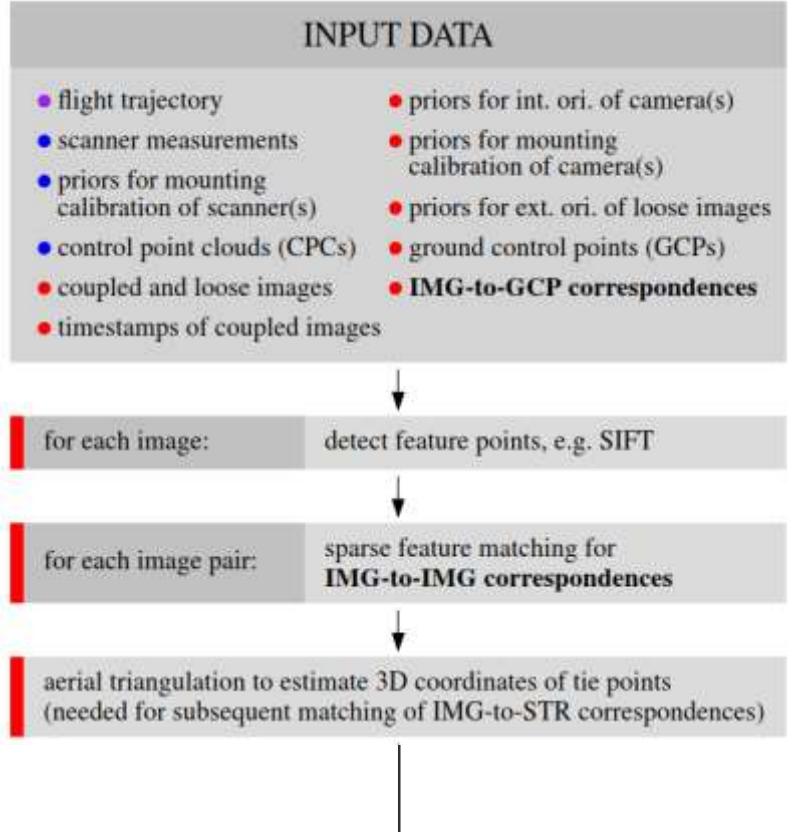


(c) IMG-to-STR correspondences colored by weight w_p



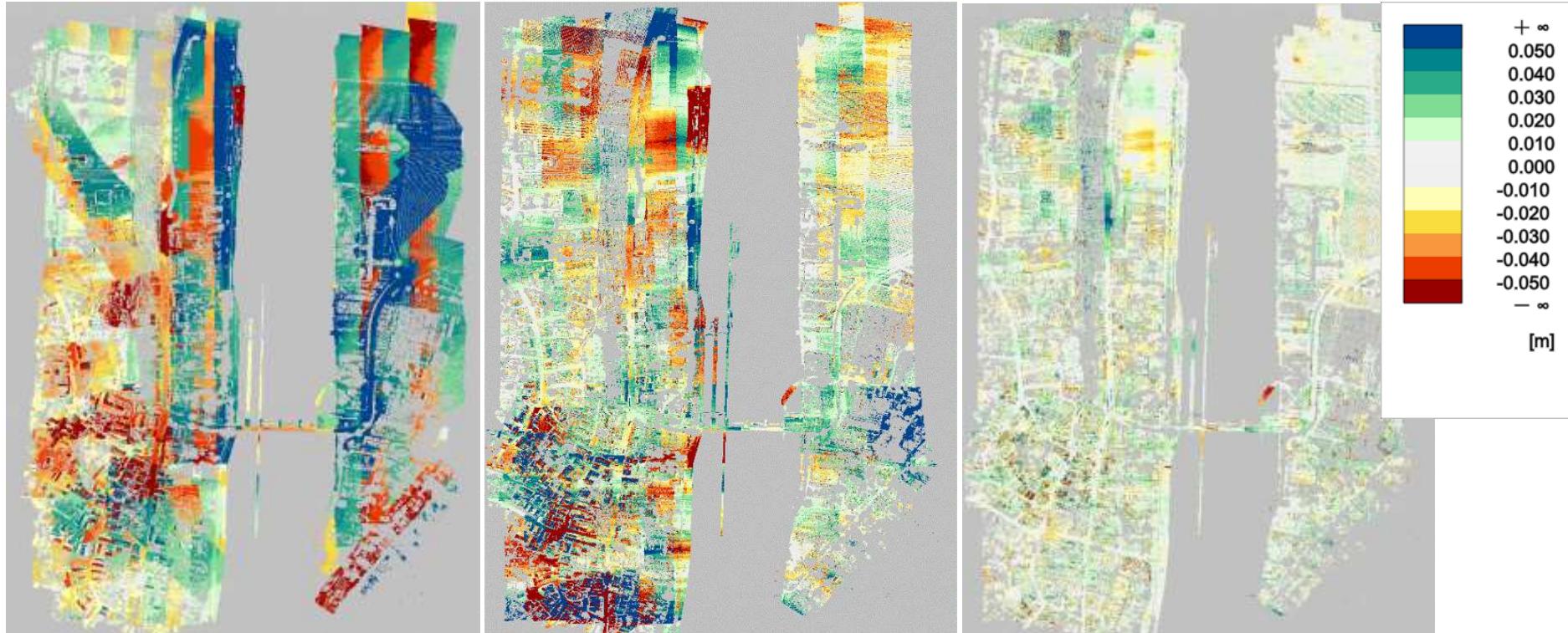
© Glira et al., 2019

Hybrid LiDAR & image sensor orientation



Stand-alone LiDAR processing

Flight campaign November 2018



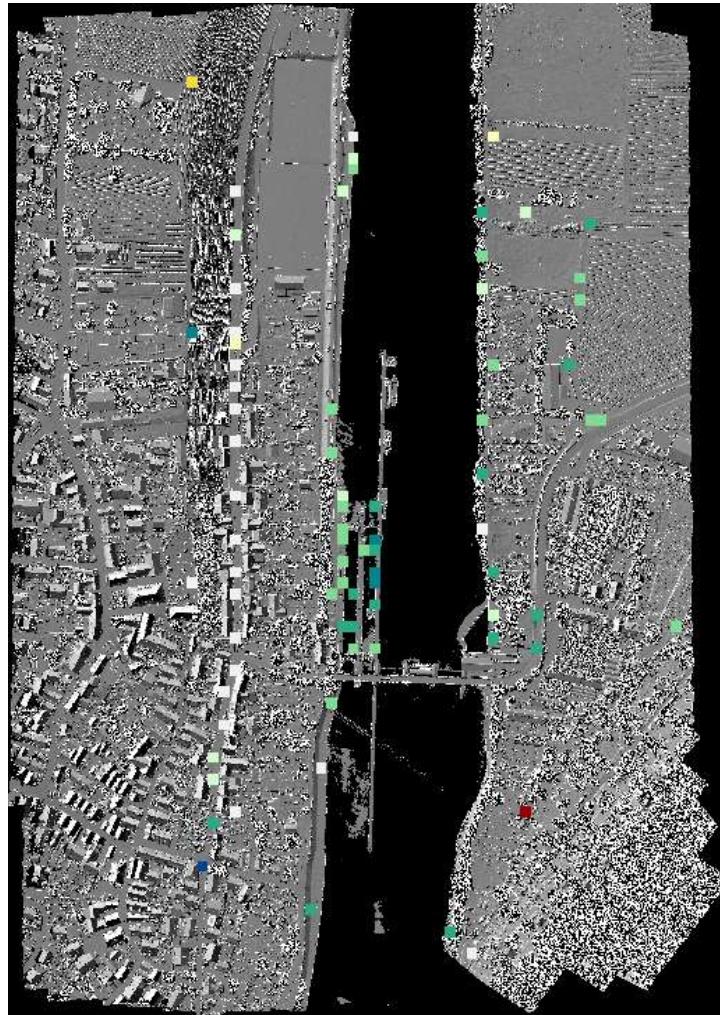
before strip-adjustment

After strip-adjustment,
bias correction

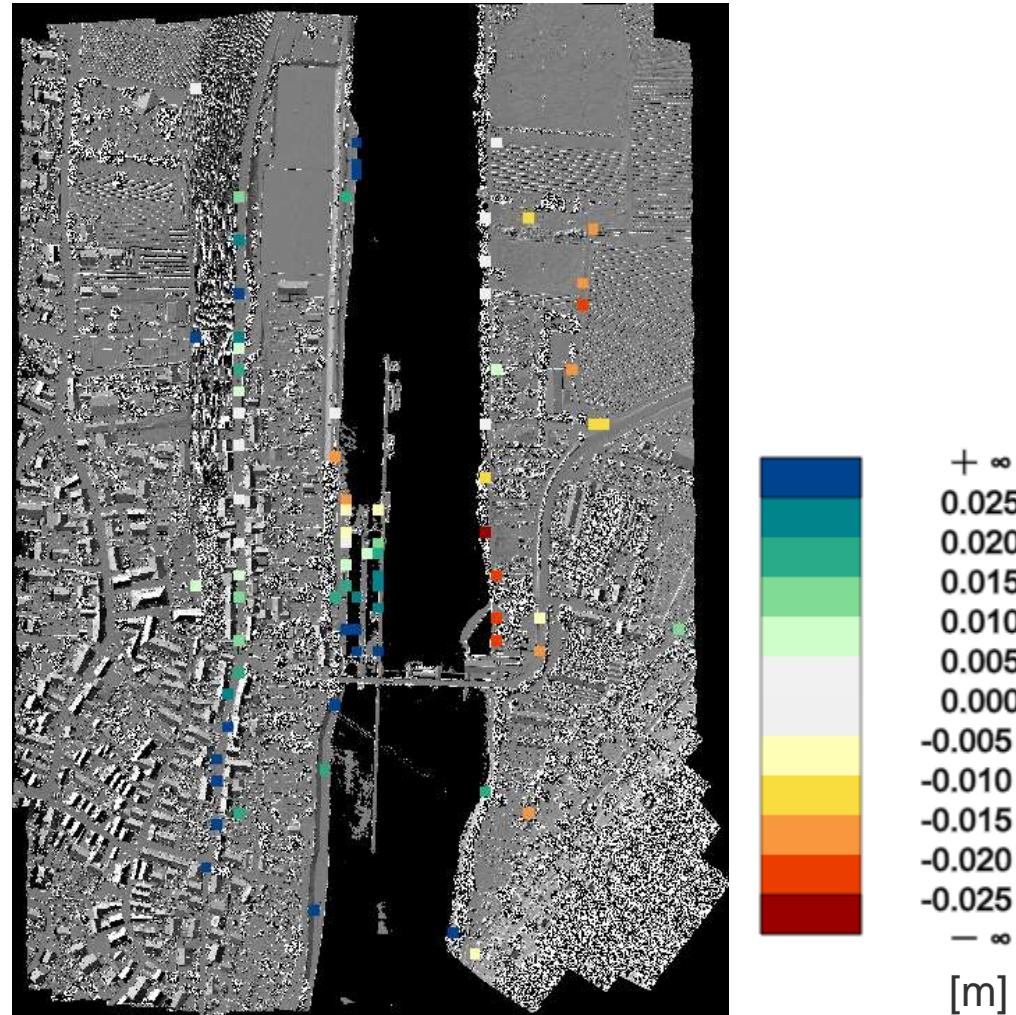
After strip-adjustment,
spline model correction

Stand-alone LiDAR processing

Absolute vertical accuracy from levelled (check-)planes

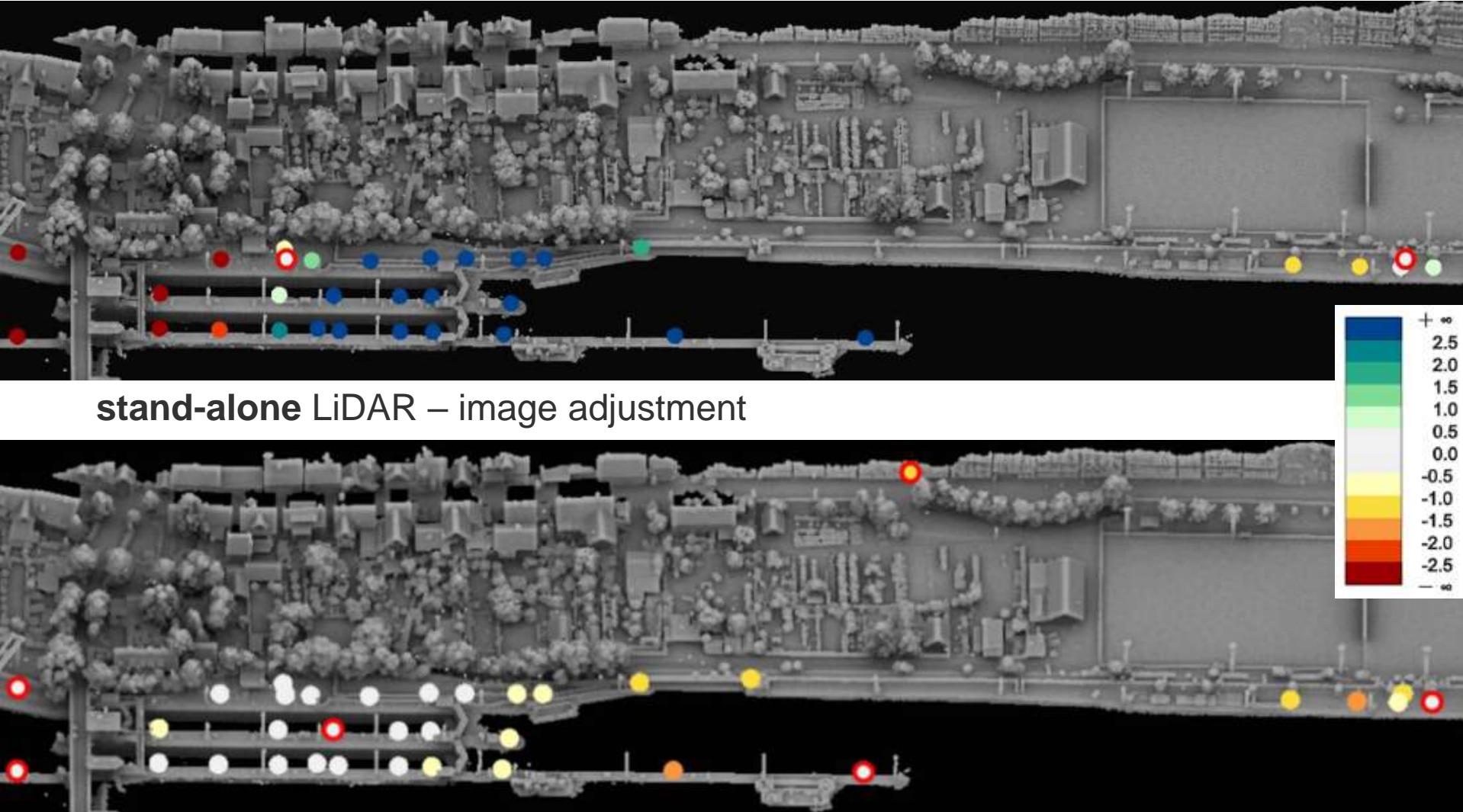


Bias model



Spline model

Stand-alone vs. hybrid LiDAR – image adjustment



hybrid LiDAR – image adjustment

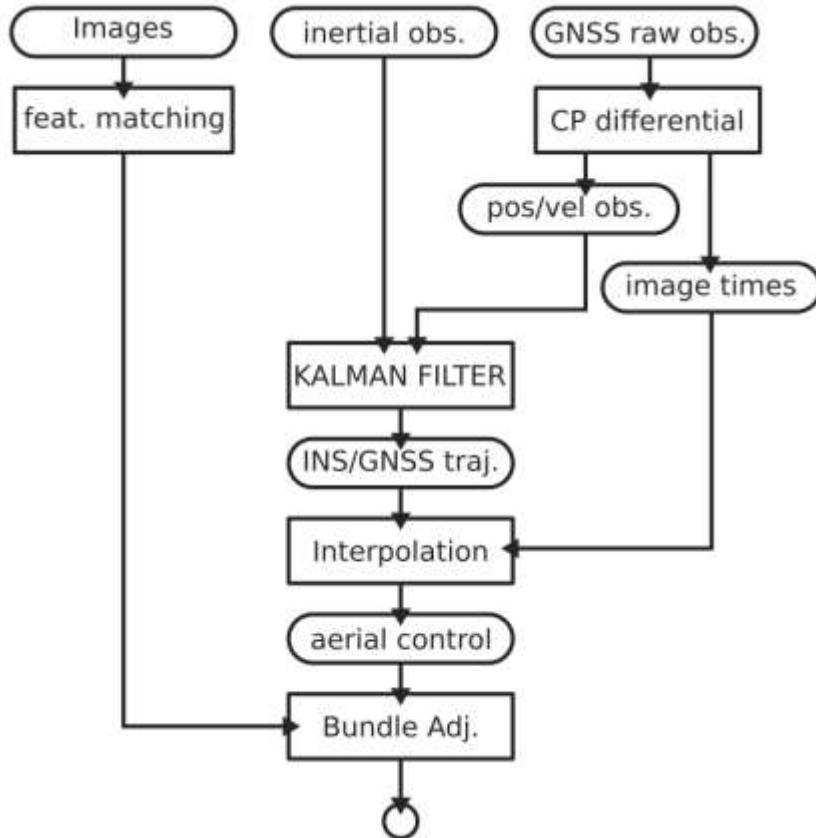
Bundle adjustment with raw inertial data

- typically GNSS/inertial data processing is done as pre-processing step before integrated sensor orientation
- this **approach fails in challenging scenarios**: short inertial/GNSS trajectories, weak photogrammetric block geometries, poor GNSS observations / signal quality
- **joint adjustment** of GNSS/inertial (**raw**) observations together with other sensors (camera / LiDAR) in **dynamic network / pose-graph optimization**

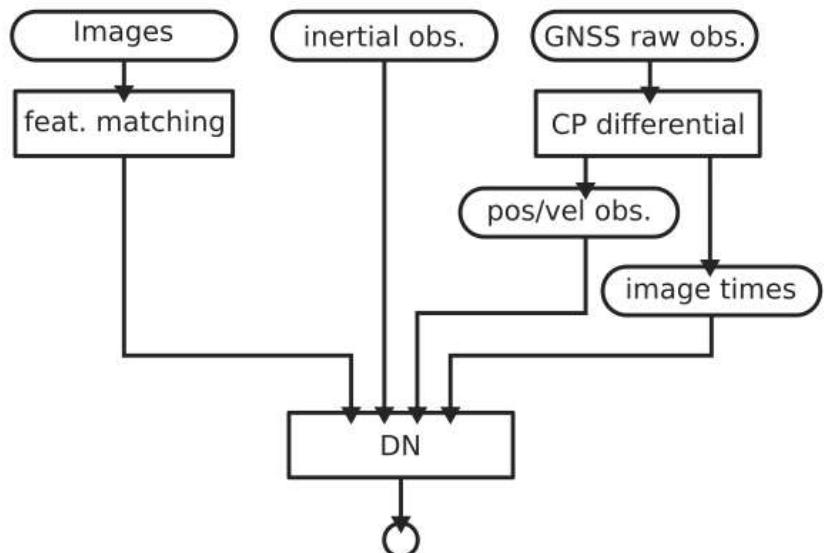
D. Cucci, M. Rehak & J. Skaloud (2017): Bundle adjustment with raw inertial observations in UAV applications, ISPRS Journal of Photogrammetry and Remote Sensing 130 (2017) 1–12
<https://doi.org/10.1016/j.isprsjprs.2017.05.008>

Bundle adjustment with raw inertial data

Processing chains in sensor orientation



Traditional workflow

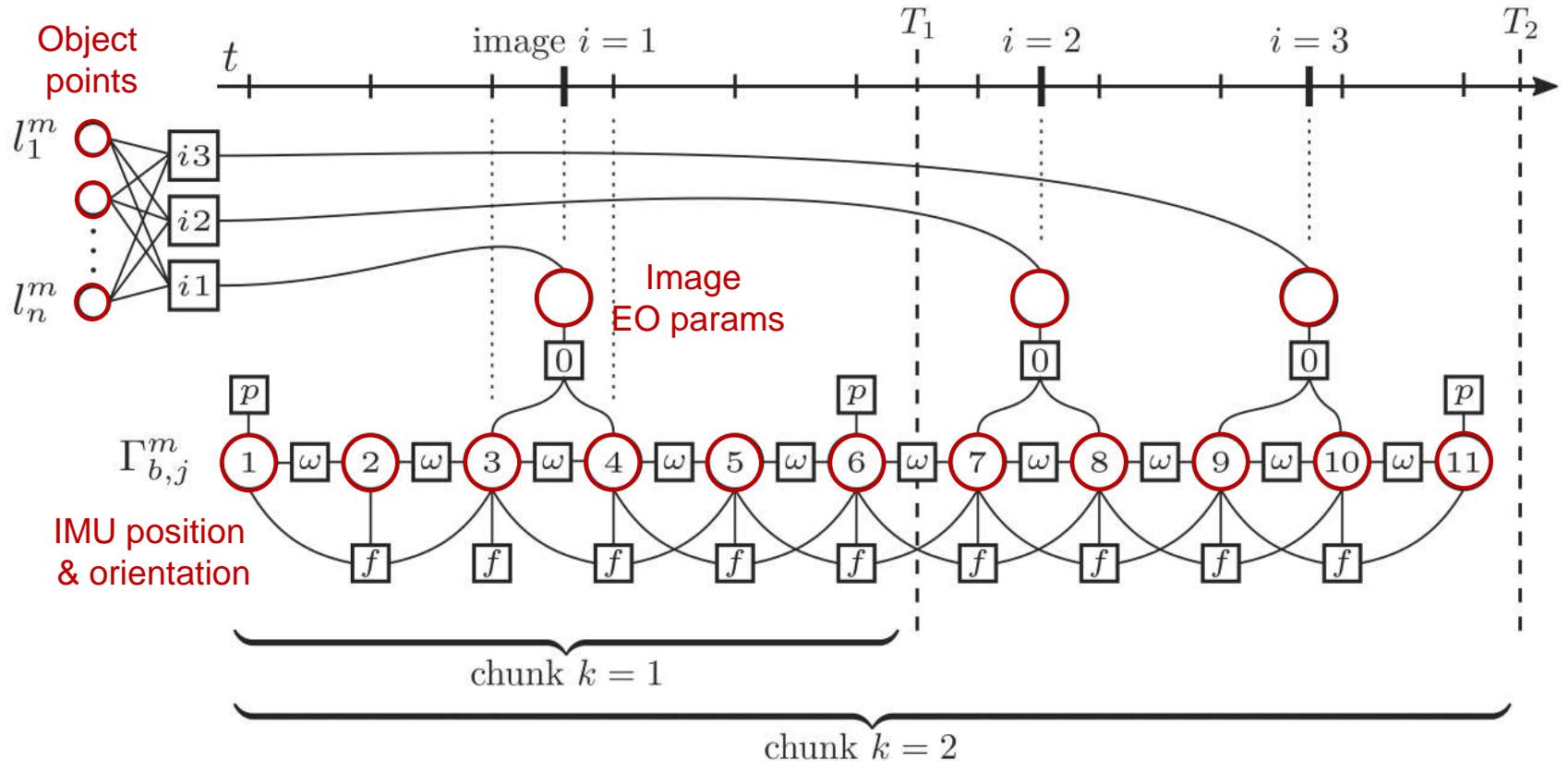


Modified workflow

Dynamic Networks – Integration of inertial raw data into sensor georeferencing

- determine the **navigation solution** (i.e., the trajectory), and, optionally, other parameters, such as inertial sensor biases, fusing all the available sensor readings **in a single step** ► “**tight fusion**” between inertial and image measurements.
- each “raw” sensor reading corresponds to an observation model, which forms the vector of unknowns:
 - Sensor poses (exterior orientations),
 - 3D position of the map-fixed features (object points) and
 - calibration parameters (interior orientation, boresight, lever-arms, IMU sensor biases)
- high number of unknowns because of high IMU rate has to be reduced (otherwise ill-conditioning), i.e. IMU pre-integration or solving through regularization

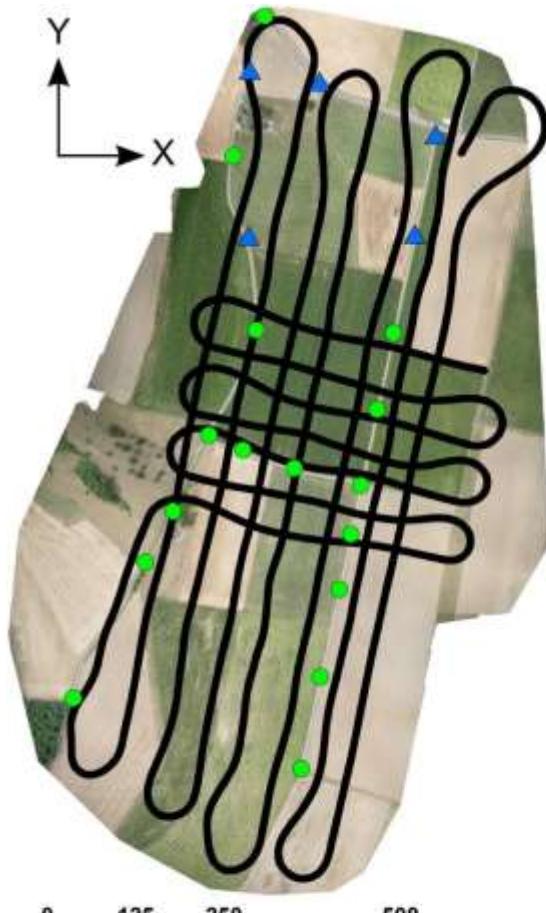
Dynamic Networks – Integration of inertial raw data into sensor georeferencing



© Cucci et al., 2017

Dynamic Networks – Integration of inertial raw data into sensor georeferencing

Empirical UAV test flight – test layout



- Check point
- ▲ Control point
- Trajectory



- Tie-point



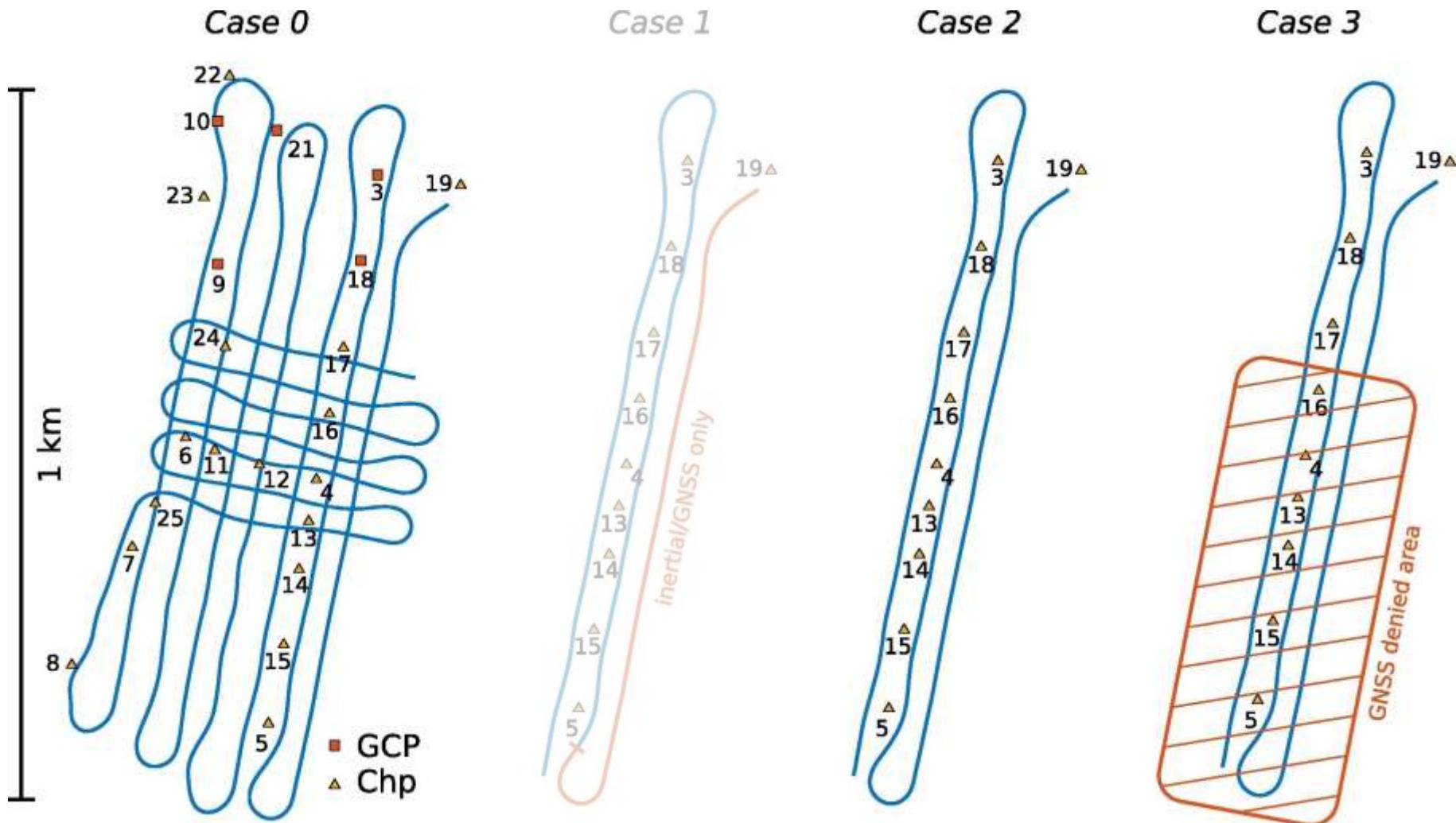
- < 3
- n obs.
- 30

Remark:
Data set
already
introduced
before

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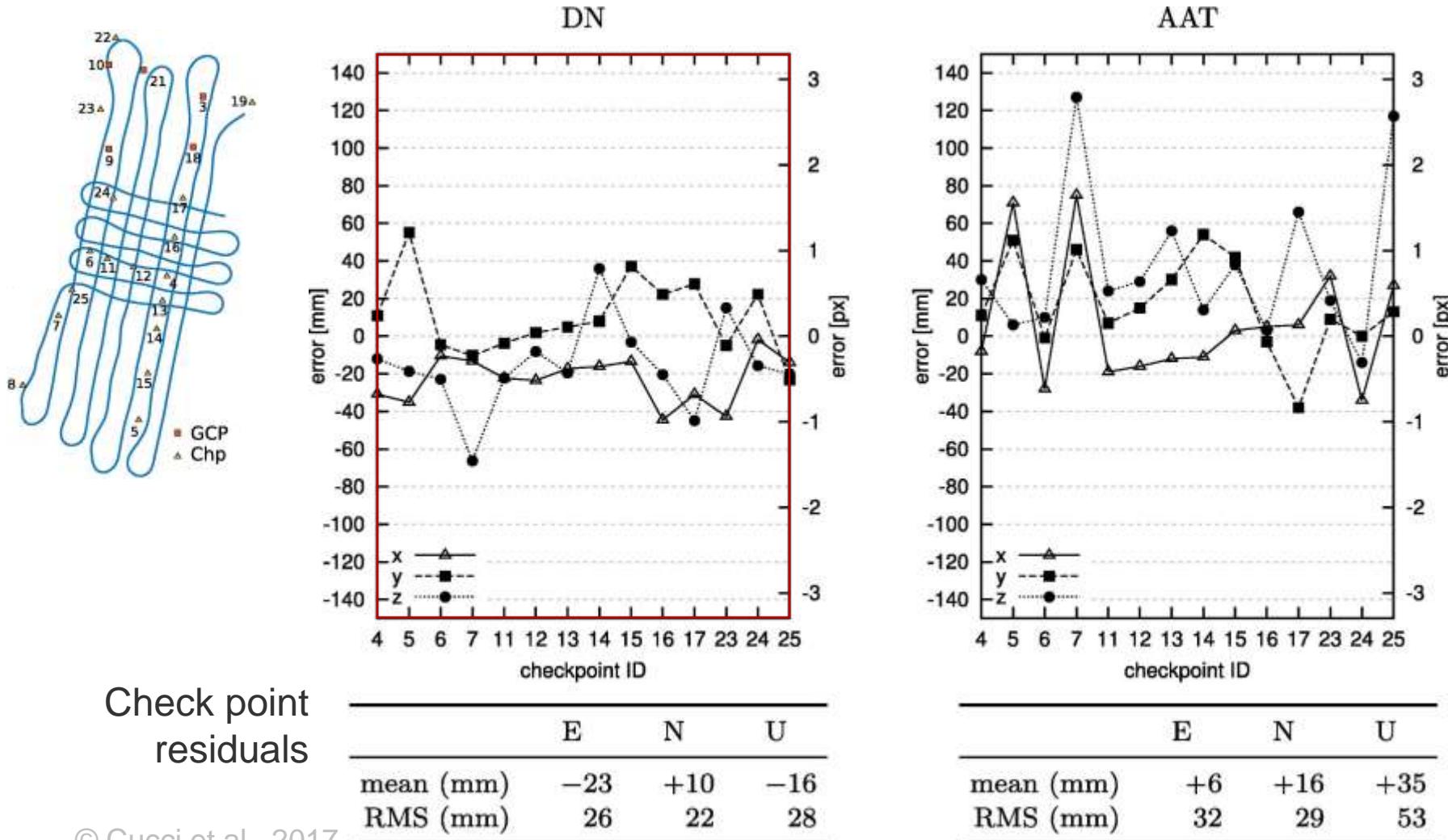
Dynamic Networks – Integration of inertial raw data into sensor georeferencing

Empirical UAV test flight – tested scenarios



Dynamic Networks – Integration of inertial raw data into sensor georeferencing

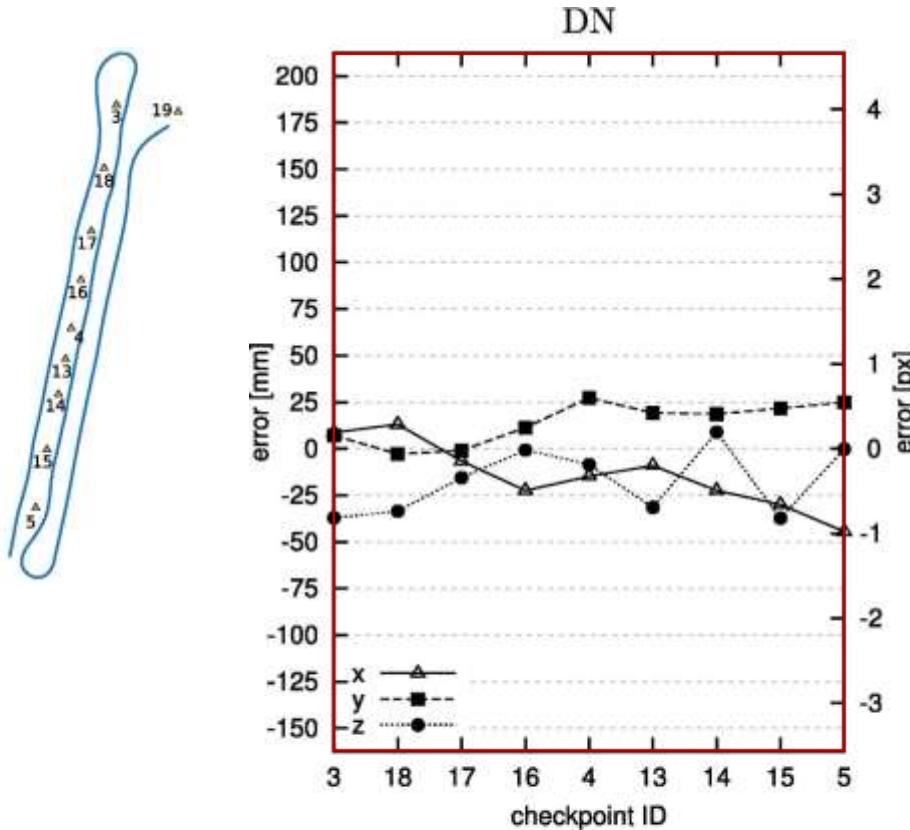
Empirical UAV test flight – Case 0: Full photogrammetric block with GCP



© Cucci et al., 2017

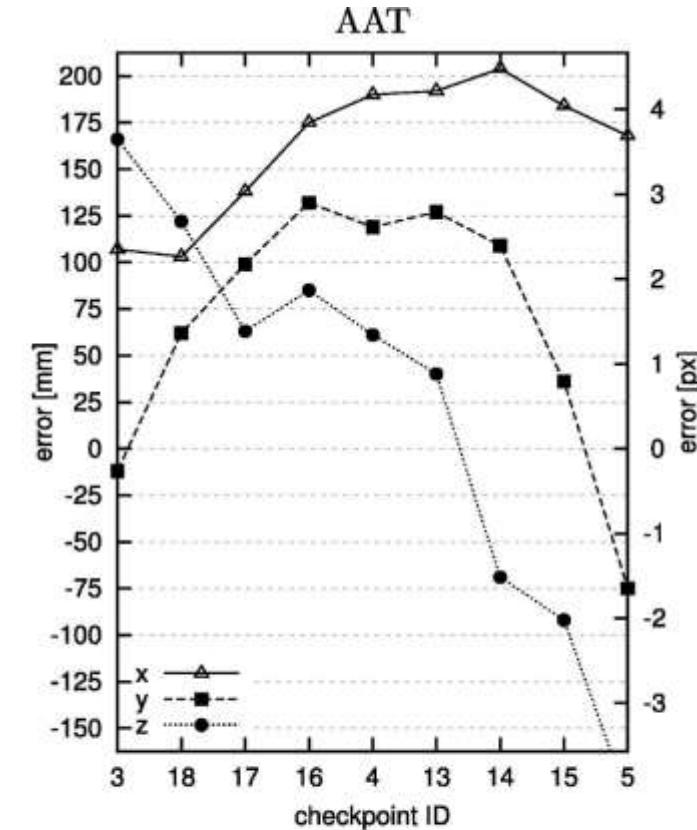
Dynamic Networks – Integration of inertial raw data into sensor georeferencing

Empirical UAV test flight – Case 2: Corridor survey



Check point residuals

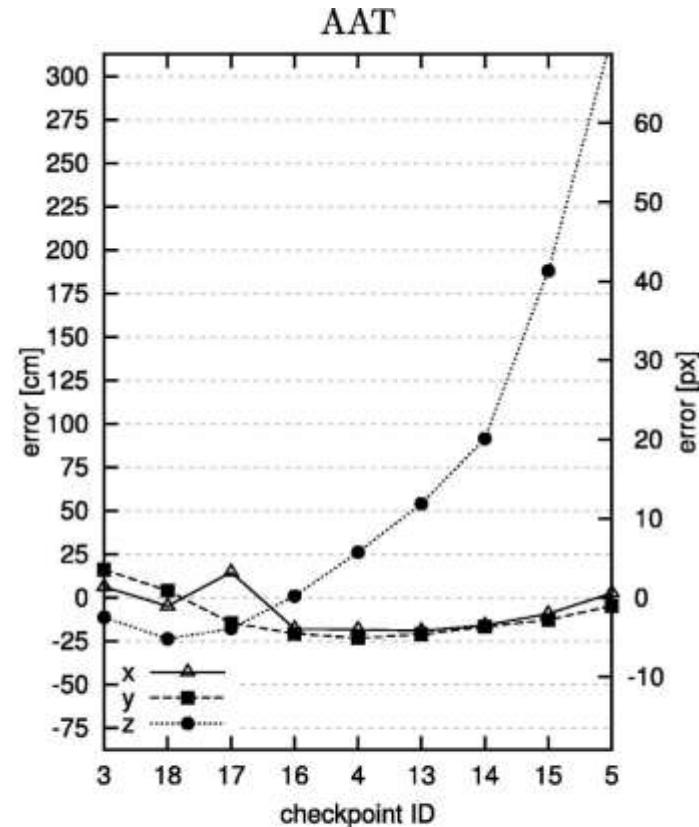
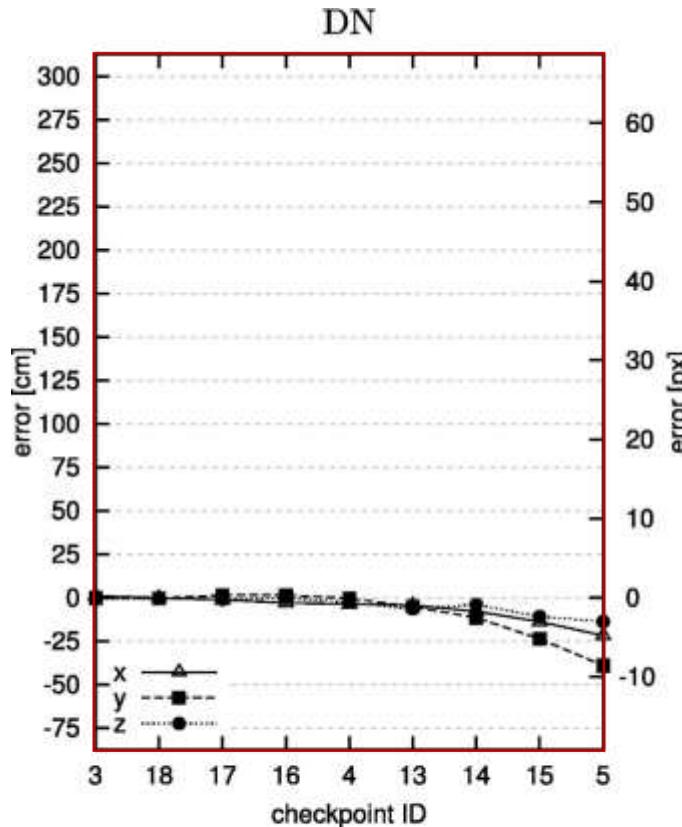
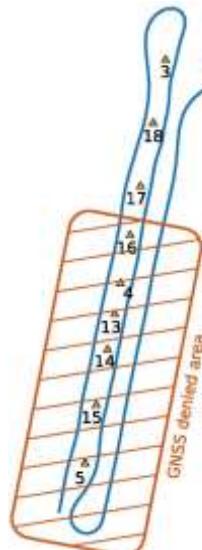
	E	N	U
mean (mm)	-14	+14	-17
RMS (mm)	22	17	24



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Dynamic Networks – Integration of inertial raw data into sensor georeferencing

Empirical UAV test flight – Case 3: Corridor survey with GNSS blockages



Check point residuals

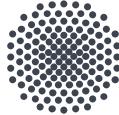
	E	N	U
mean (mm)	-61	-86	-44
RMS (mm)	93	158	64

	E	N	U
mean (mm)	-101	-104	+701
RMS (mm)	135	162	1301

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Questions & Comments ?!





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Tutorium DGPF Jahrestagung Stuttgart 2019
Institut für Photogrammetrie, Universität Stuttgart, 04.03.2020

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