

Evolution 2: Manual to automatic image meas. & orientation





Manual monocomparator Computer-assisted monocomparator



Analytical plotter



Targetless automatic image meas. & network orientation



Fully automatic network orientation using targets



Automatic comparator with image EO

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Major CRP Developments in the Digital Era

- Coded targets/markers facilitated autoorientation & 3D point determination, in both off-line & on-line/real-time CRP systems
- FBM/SfM auto-orientation facilitated autoorientation & sparse 3D point cloud generation
- Dense image matching facilitated dense 3D point cloud generation to pixel-level resolution
- Automatic camera calibration facilitated by coded targets or FBM orientation; has enhanced accessibility to CRP

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Photogrammetric measurement for accident reconstruction & forensics



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Near-planar geometry not conducive to targetless orientation









Multi-level site was recorded in 3 networks spanning 300m, in 70 minutes



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Example Project: Scene mapping of tanker explosion



Example Project: Scene mapping of tanker explosion





Automated CRP Measurement using Targets



Aircraft Manufacture

AIRBUS Hamburg Fuselage Sections And Seat Rail Measurements (A380, A340, A330, A321, A320, A319, A318)



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Measurement Example: Passenger Door Frame & Door Hinges at AIRBUS Hamburg







A 15 minute task for a 3D meas. accuracy of 0.015mm; no operator intervention beyond loading the images

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Targetless, Au	Itomated Image Orientation
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Automated Object Reconstruction via Dense Matching



Approx. 55 million (visible) 3D points from dense image matching

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Targetless Network Orientation via FBM: Amphora



57 images, 90,000 pts ('sparse' p.c.), RMS vxy – 0.35 pixel

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Targetless object reconstruction via dense matching: Amphora



Dense point cloud from SGM (via SURE) comprising 116 million points



Triangulated mesh

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Examples taken from Photoscan Showcase (http://www.agisoft.com/community/showcase)

What did photogrammetrists get out SfM developments?

- A powerful new approach to solving the image-point correspondence problem, albeit to a precision that could be 10 times poorer than when using targets (eg 0.3 pl versus 0.03 pl)
- Some new approaches to determination of initial values for non-linear photogrammetric orientation models (eg bundle adjustment), but these are not universally applicable
- Adoption of RANSAC-type approaches, eg for filtering of matches and initial value determination; ie the notion of using solution plausibility involving many point combinations rather than relying on high-quality control/constraints comprising a few points.
- And ... the headaches of processing & interacting with dense point clouds (though not really a CV inspired issue!)

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<image>

Project Example: UAV WitnessPRO mapping, British Columbia

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UAV WitnessPRO mapping, RCMP British Columbia





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Prospects for post-orientation, 3D feature point extraction via monoplotting





• 3D Feature points from single images

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FBM-based auto orientation can work OK ... but beware! Case 1: Bridge Deformation Survey



8 images, >400 pts, RMS vxy=0.31pl, Accuracy 1:1,600 Summary of results suggests reasonable network orientation



Target-based auto orientation: Bridge Deformation Survey



8 images, 52 pts, RMS vxy=3.2pl, Accuracy 1:800, all pts >5 rays

Solution via targets with >5-ray intersections suggests something is wrong!

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FBM-based auto orientation: Network orientation problem



One image has excessive outliers & another has a limited number of points







Dimensional inspection of an aircraft tool; required to measure tooling points and 'black holes'.







Automatic Camera Calibration – with & without targets

Results of self-calibrations of the Nikon D200 camera for targeted and untargeted cases.



27 images, 25,000 feature points, 200 target points (25 codes)

	Focal length, c (σ _c) mm	x _p (σ _{xp}) mm	У _р (σ _{xp}) mm	∆r @ r=8mm µm	∆r @ r=10mm µm	∆r @ r=12.0mm µm	P(r) @ r=10mm μm	P(r) @ r=12mm μm	RMS v _{xy} No of points	
Coded targets	17.632 (0.0010)	-0.038 (0.0007)	-0.193 (0.0007)	121.8	217.2	332.2	5.7	8.1	0.09 pl 200	
Untargeted	17.627 (0.0008)	-0.036 (0.0005)	-0.193 (0.0005)	120.9	216.2	333.1	5.3	7.6	0.25pl 55,500	
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Prospects for in-field self-calibration

- Massive data redundancy afforded by FBM targetless orientation can mean that less stringent constraints need be applied to imaging geometry
- Example of 49-image, 41,000-point UAV network
- 18,000 points seen in >3 images; 6,000 pts seen in 6 or more images
- Feasible because Δh in object space approx. 60% of flying height H



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Prospects for 'natural' object point fields for self-calibration

- PhaseOne camera with 50mm lens
- 40-images, 3300 points (all with 6 or more rays), RMS vxy = 0.20 pixel
- Convergence could be relaxed because ∆h in object space >50% of H
- Standard errors of 2 μm for c & <1 μm for xp,yp



In-field self-calibration – multiple cameras

- 5 Sony Alpha 850s with 50mm lenses on a fixed-wing UAV
- Nadir & two oblique angles → 127⁰ fov
- @ H = 800m, lateral coverage is 3200m



Project courtesy of Prof. J.Y.Rau, NKCU

In-field self-calibration – multiple cameras

- Block of 540 UAV images from 5 cameras; 10,900 points
- Self-Cal. Bundle Adjustment: RMS v_{xy} = 0.44pl, σ_{XY} = 0.08m, σ_{Z} = 0.11m
- All points seen in 4 or more images, 9200 in 6 or more & 390 in >20



Final Remark

Choice for 3D object measurement/reconstruction via either sparse or dense surface matching & point cloud generation, or via targeted or manually measured feature points, will depend upon downstream priorities; some forms constitute final information products whereas others require subsequent data-to-information conversion

THANK YOU

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