



**The Twelfth International Conference on Sensor Device Technologies and Applications - SENSORDEVICES 2021**

Special Track: MLS-RSUAV: Machine Learning and Classical Signal Processing  
Algorithms Applied to Robotics and Sensor Data Collected from  
UAVs Security Solutions for Mission Critical Applications



# 3D Reconstruction with Drone Images: Optimization by Reinforcement Learning

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# About Authors Presenting



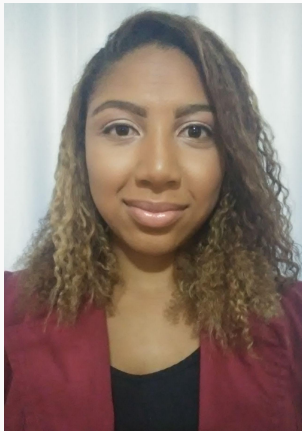
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# Outline



## Summary

- . Introduction
- . Methodology
- . Experiments
- . Discussion of Results
- . Conclusion / work in progress

# Introduction

- Increasing need for geometric 3D models
  - Movie industry, games, virtual environments, simulators;  
Inspect details to ambient researches [1], historical heritage [2],  
support military missions ...
- Existing solutions are not fully satisfying
  - User-driven modeling: long and error-prone
  - 3D scanners: costly and cumbersome
- Alternative: analyzing image sequences
  - UAVs increase mobility and multiples softwares
  - Cameras are cheap and lightweight
  - Cameras are precise (several megapixels)

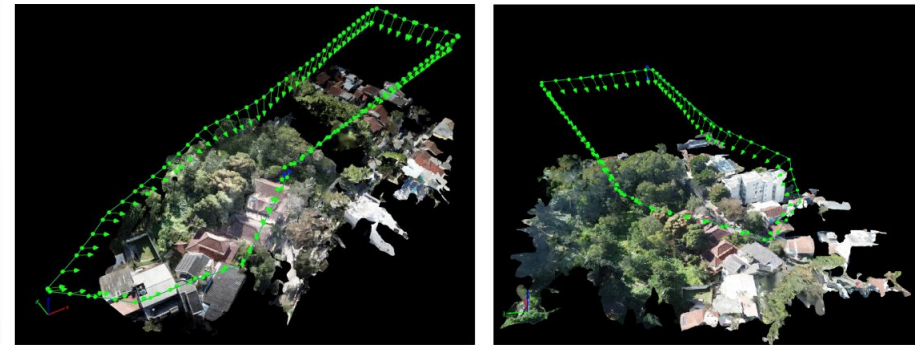
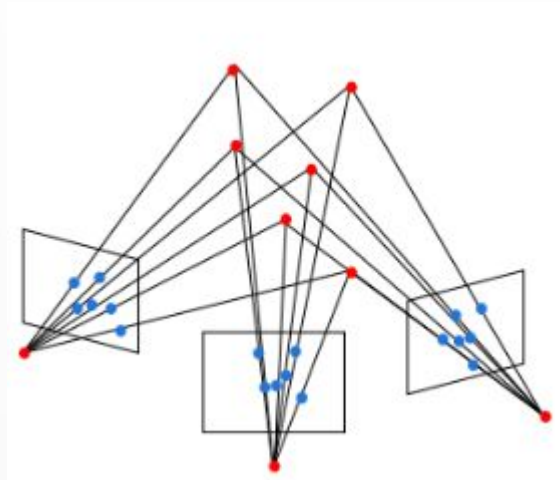


Figure 1. Mapped and reconstructed outdoor area.

# Methodology - 3D Reconstruction Pipeline



Multiview geometry [4]



Point cloud

# Methodology - Structure for Motion

- Structure from motion uses referenced relative motion for inference about 3D geometry;
- Camera poses (position + orientation) are resolved automatically;
- Use of adjustment from image set (bundle adjustment)

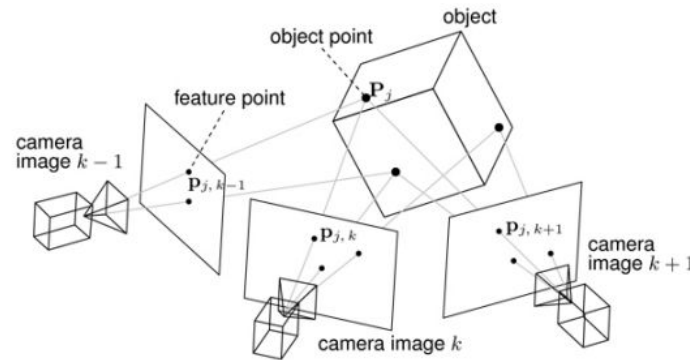
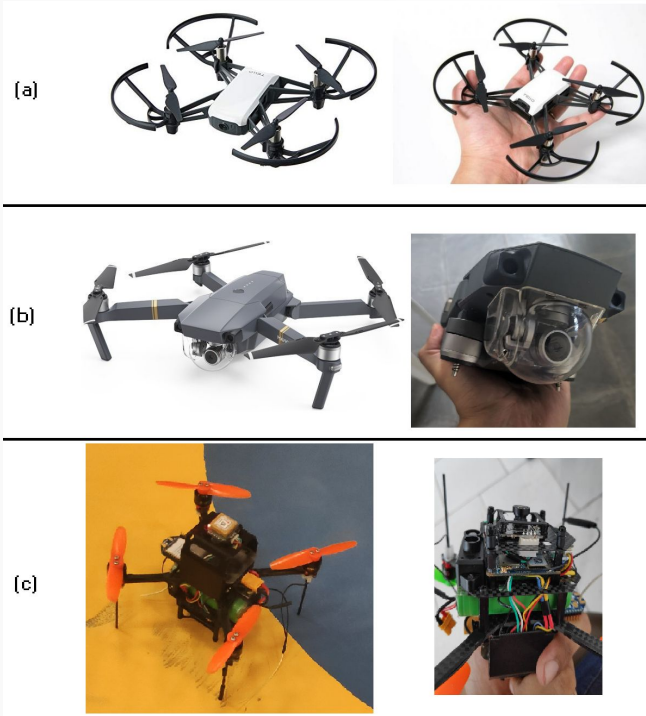


Figure 2. Result after structure-from-motion estimation. The projection of a 3D object point  $P_j$  in the camera image at time  $k$  gives the tracked 2D feature point  $P_{j,k}$  [5]

# Experiments - UAVs and Datasets



Dataset generated	Acquisition Device	Images	PIX4D	Meta-shape	ODM	RC
Crystal's valley	Mavic Pro (b)	337	x	x	x	
BOC 60 - High Res.	Mavic Pro (b)	302	x	x	x	
BOC 60 - Med. Res.	Mavic Pro (b)	169	x	x	x	
BOC 60 - Low Res.	Mavic Pro (b)	138	x	x	x	
LARC	Sub-250 (c)	150	x	x		
PIRF - Fan Scene	Tello (a)	62				x
PIRF - Human Scene	Tello (a)	50			x	x
PIRF - Bags Scene	Smartphone	217	x			
Object - Plant	Tello (a)	35	x	x		x
Object - Robot	Smartphone	154	x			
Object - Castell	Smartphone	64	x	x	x	x

Table I  
GENERATED DATASETS, IMAGES AND RECONSTRUCTION SOFTWARES  
USED.

# Experiment Outdoor– BOC 60

Resolutions: High (302), Medium(169) and Low(139) [6].

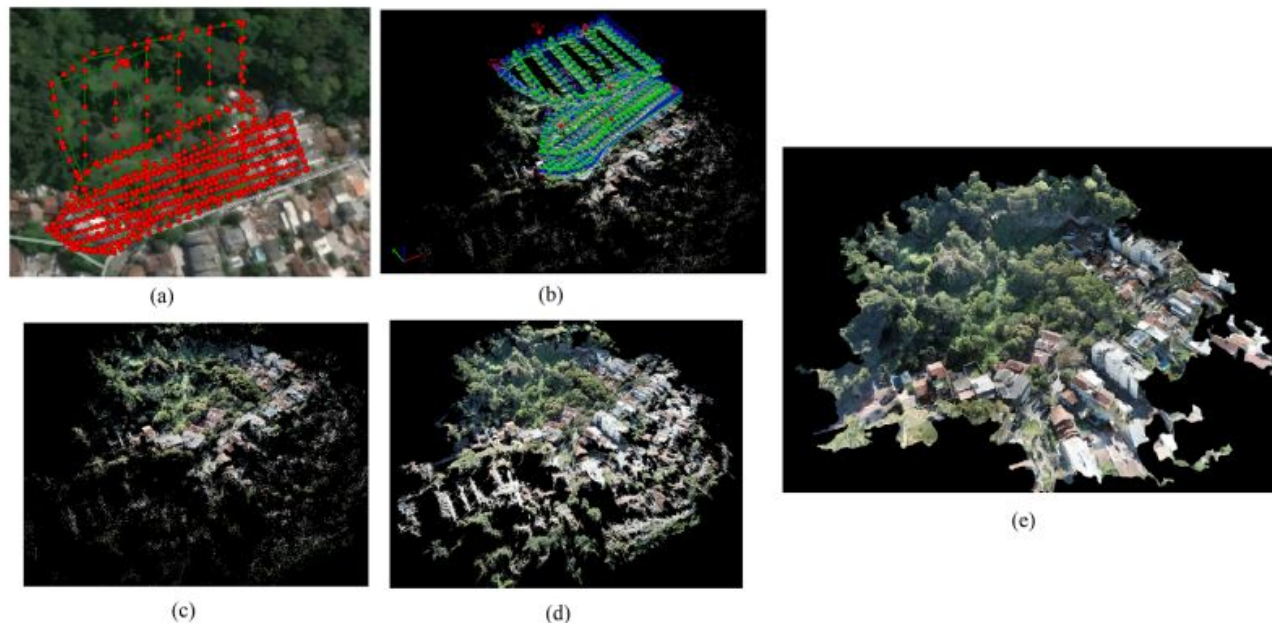
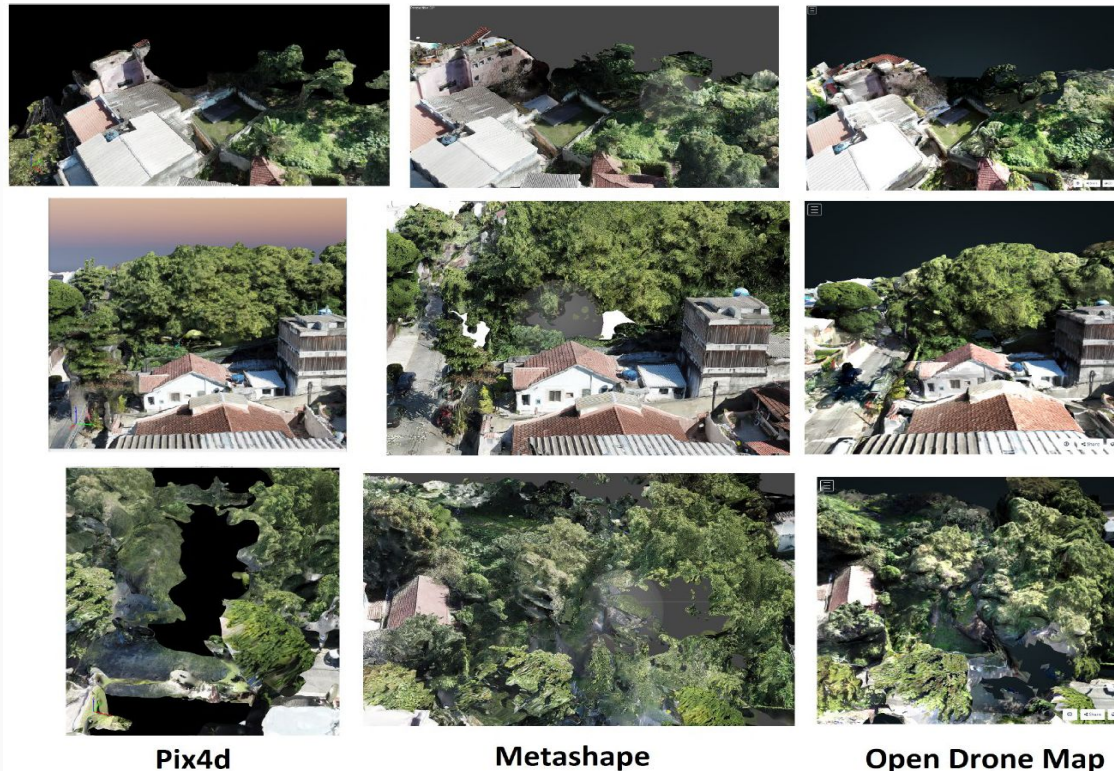


Figure 3. BOC 60 Steps to Rebuild PIX4D software; (a) Snapshot points on the map; (b) 3D image taking points; (c) Tie Points; (d) Dense cloud of points; (e) Textured 3D Model.

# Experiment Outdoor- BOC 60

Visual comparison between PIX4D, Metashape and ODM Softwares



# Experiment - Medieval Castle

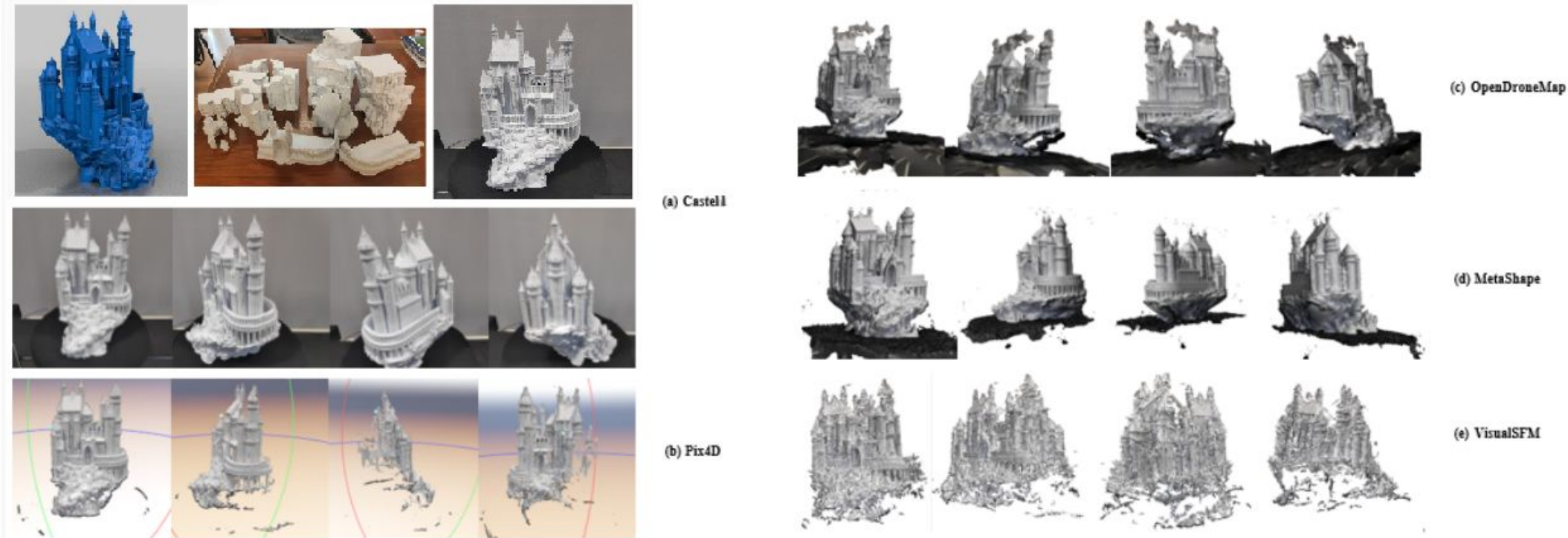


Figure 6. Medieval Castle experiment, reconstruction using different tools. [7]

# Discussion

CloudCompare software was used for comparison between point clouds [8].

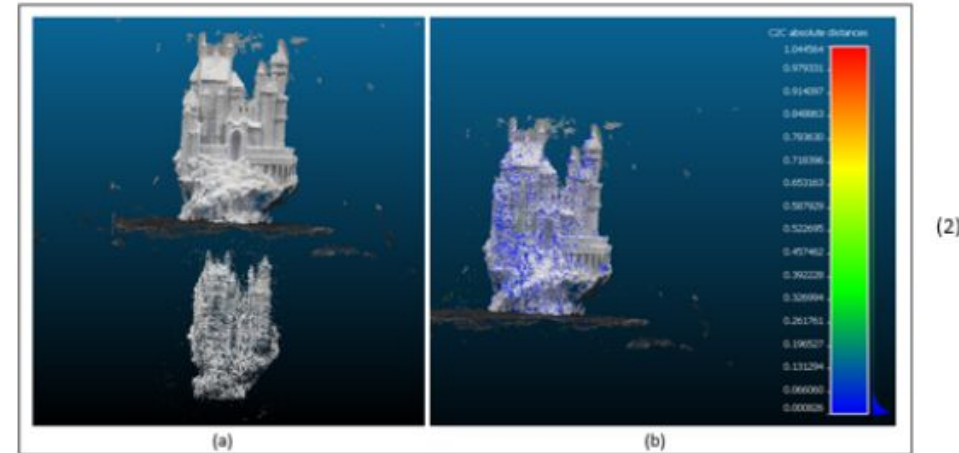
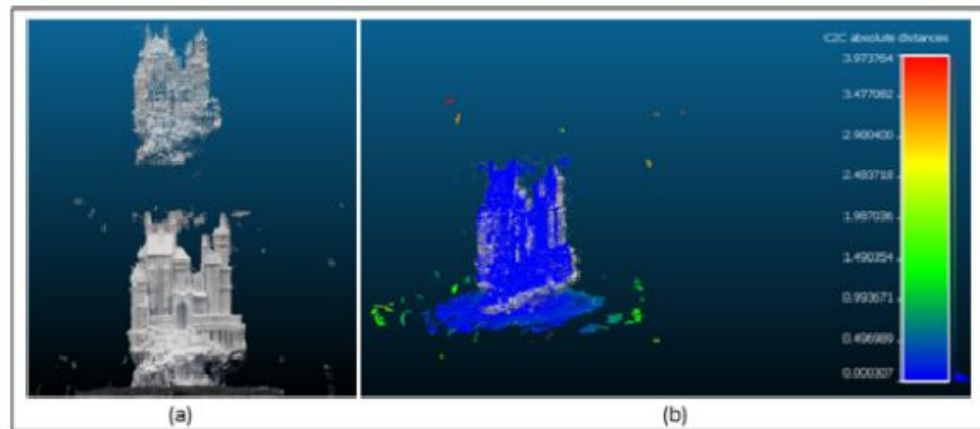


Figure 7. Comparison point cloud Metashape and OpenDroneMap. (1) Metashape reference. (2) OpenDroneMap reference. (a) Insertpoint cloud; (b) Generated heat map.

# Conclusion

- The contribution made by this project includes the creation of datasets with scenes and 3D objects;
- Use datasets for optimization experiments with machine learning and reinforcement learning in order to improve the distortions caused during image processing and also increasing the visible accuracy of the three-dimensional models.
- We conclude that it is feasible to use UAVs for imaging outdoor and indoor scenes to reconstruct objects and scenes.

# References



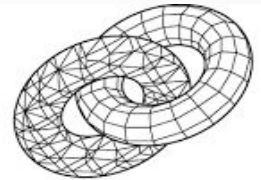
- [1] E. Casella, et al. "Mapping coral reefs using consumer-grade drones and structure from motion photogrammetry techniques". Coral Reefs, vol. 36, pp.269–275, 2017.
- [2] E. Colica, et al. "Using unmanned aerial vehicle photogrammetry for digital geological surveys: case study of Selmun promontory, northern of Malta". Environ Earth Sci 80, pp. 551, 2021.
- [3] I. Craig, Vision as process. Robotica, Cambridge University Press, vol.13, n. 5, pp. 540, 1995.
- [4] S. M. Seitz, B. Curless, and J. Diebel, and D. Scharstein, and R. Szeliski, "A Comparison and Evaluation of Multi-View Stereo Reconstruction Algorithms," 2006 IEEE Computer Society Conference on Computer Vision and Pattern Recognition (CVPR'06), 2006.
- [5] C. Kurz, T. Thormählen, and H. Siedel, "Visual Fixation for 3D Video Stabilization". Journal of Virtual Reality and Broadcasting, pp. 12, 2011.
- [6] L. C. P. Velho, DroneDatasets. 2020. Available at: <https://www.visgraf.impa.br/dds/boc60/index.html> [Retrieved: November, 2021]
- [7] Boldmachines, MedievalCastle, 2018. Available at: <https://www.thingiverse.com/thing:862724> [Retrieved: November, 2021]
- [8] CloudCompare-OpenSourceProject. Available at: <https://www.cloudcompare.org> [Retrieved: November, 2021]

# Thanks!

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