



# 面向城市点云理解的对称和相似性检测

## Symmetry and Similarity Detection for Urban Point Cloud Understanding

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# Outline

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1

点云 Urban Point Clouds



2

检测 Symmetry and Similarity detection



3

小结 Summary





# 0.1 HKU iLab: The urban big data hub



## ◆ iLab 实验室

- Director: Prof. Wilson Lu
- Urban big data hub at Faculty of Architecture, HKU
- multi-dimensional and multi-disciplinary *urban big data* collection, storage, analysis, and presentation to inform decision-making in urban development
- Focusing on **Information Technology (IT)**
  - BIM, GIS, GNSS, *Urban Remote Sensing*, IoT
  - Blockchain (BC/DLT)



Lab workshop



iLabHKU



<https://ilab.hku.hk/>

**iLab** | @HKURBAN<sup>lab</sup>  
the urban big data lab



iLab

## 0.2 About me

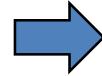


Homepage  
(full-text PDFs)

### ◆ A mixed background 背景

- BEng in Automation, CAUC
- MSc in Computer Science, CAUC
- PhD in System Engineering, HKPU
- PDF/RAP/AP in Construction IT

2004  
2007  
2012



### ◆ Research interests 方向

- Urban sensing and computing
- As-built BIM and digital twin
- Automation/IT in construction
- Applied operations research, ML
- Distributed (blockchain) applications to construction

### ◆ Engineering

- ISE, CEM, EIE

### ◆ Computer Science

- AI, OR, ML

### ◆ Economics

- SCM

### ◆ Professional

- MACM, SMCGS,  
MIEEE, MHKGISA

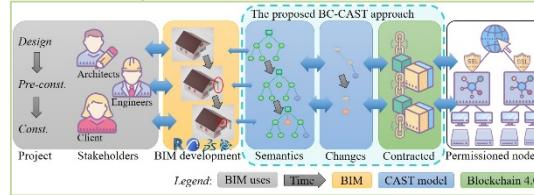
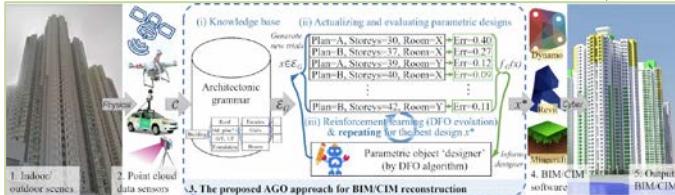
- V.C. ACM-HK,  
Com. CGS-BIM



# 0.2 Recent research projects

## ◆ On-going 在研

□ PI: HK RGC GRF/ECS (17200221, 27200520)



□ PC/Co-PI: RGC TRS (T22-504/21-R), SZ-HK-MC TRP T(C), ITF T-1



## ◆ Completed 完成

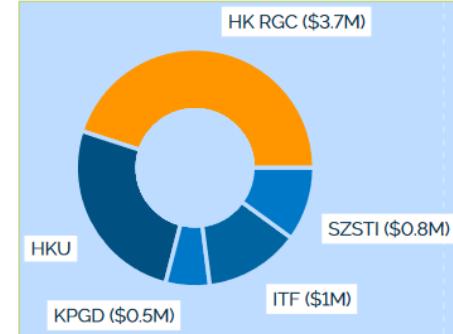
□ PI: HK RGC GRF (17201717, 17200218), etc.

□ Co-I: NSFC \*2, NSSFC, SPPR, ECF, etc.

Xue: Sym & Sim. SCUT, Guangdong, China. 2022.

## ◆ Keywords

- BIM/CIM
- 3D point cloud
- Derivative-free optimization
- Urban semantics



Sponsors of projects as PI/PC/Co-PI

Section 1

# URBAN POINT CLOUDS

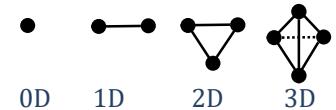
城市点云



# 1 Introduction

## ◆ Point 点

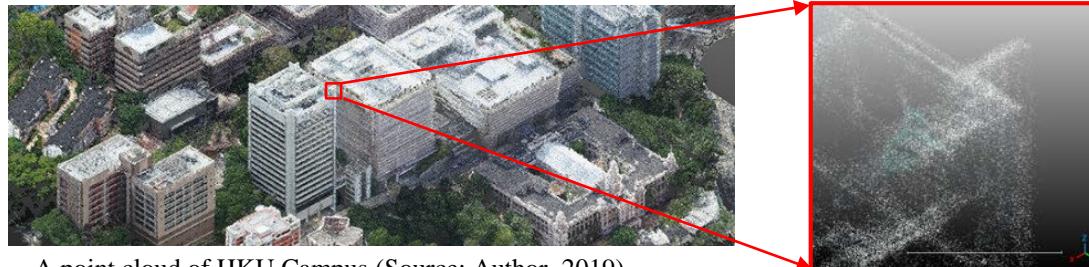
- A **location** in space, 0D (no width, length, or thickness)
- Structured format: {x, y, z}, [R, G, B, Nx, Ny, Nz, Cls, Int., ...]



## ◆ Cloud 云

- An unstructured collection [of water droplets or ice crystals]
- Dense when looking at a distance, sparse closely

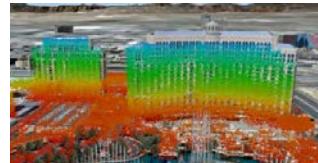
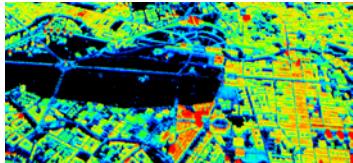
## ◆ Urban point cloud 城市点云





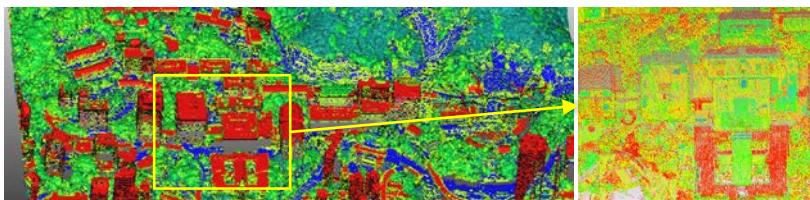
# 1.1 Major sources of urban point clouds

## ◆ SAR 合成孔径雷达



Bellagio Hotel, Las Vega  
(Zhu & Bamler 2014)

## ◆ LiDAR 光达

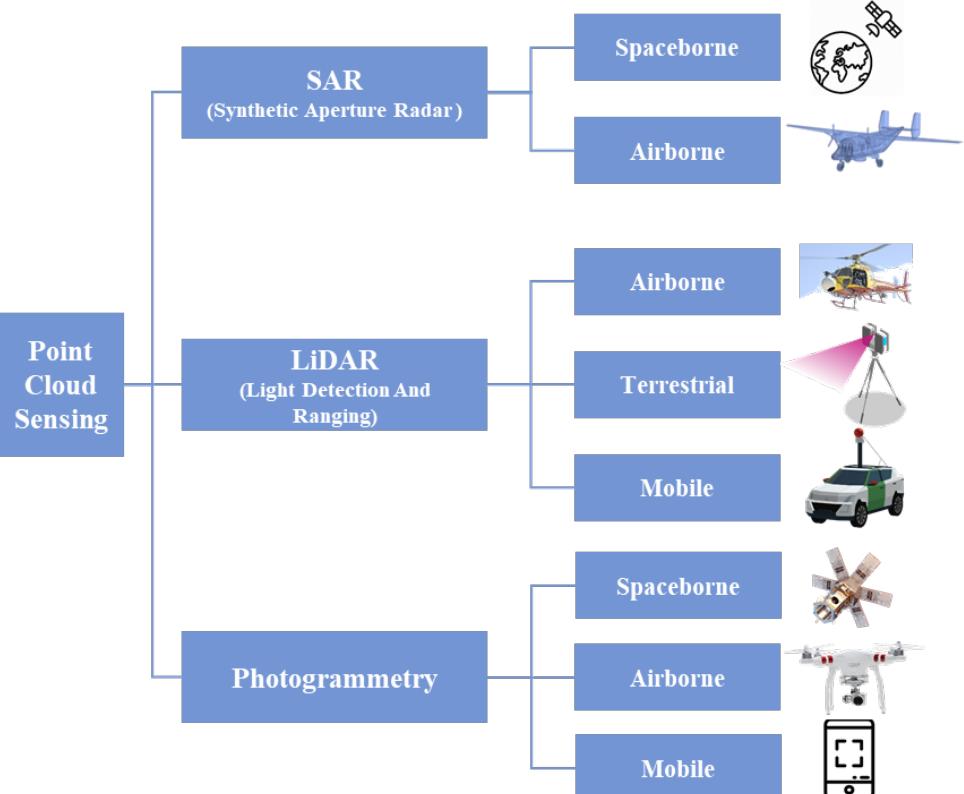


HKU Campus (Xue et al, 2019f)

## ◆ Photogrammetry 摄影测量



Xue: Sym & Sim. SCUT, G Building rooftop (Xue et al, 2019d)



(Xu et al, 2018)

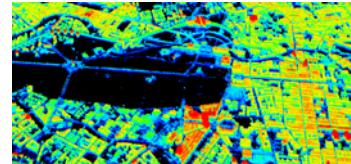


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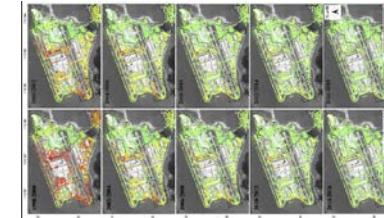
# 1.1 Advantages and applications

## ◆ SAR 合成孔径雷达

- mm-accuracy
- Coverage



## ◆ Use cases 用例



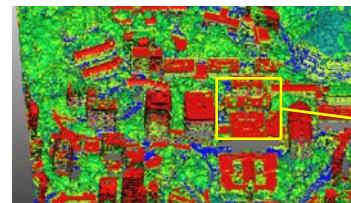
Ground settlement, building deformation (Wu et al. 2020a; 2020b)



(a) Coherent points overlaid on a Google Earth city model

## ◆ LiDAR 光达

- mm/cm/dm
- No distortion
- Intensity



## ◆ Photogrammetry 摄影测量

- cm-accuracy
- Colorful
- Cheaper



## ◆ Use cases 用例



Roof albedo (Xue et al. 2019f), indoor CFD simulation (Source: Author, 2022)



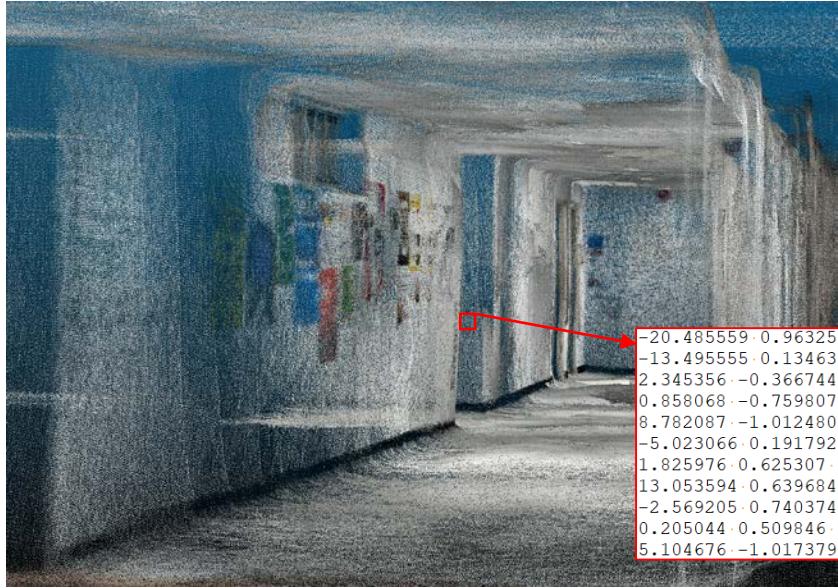
HKU @MineCraft (Source: Author, 2021)



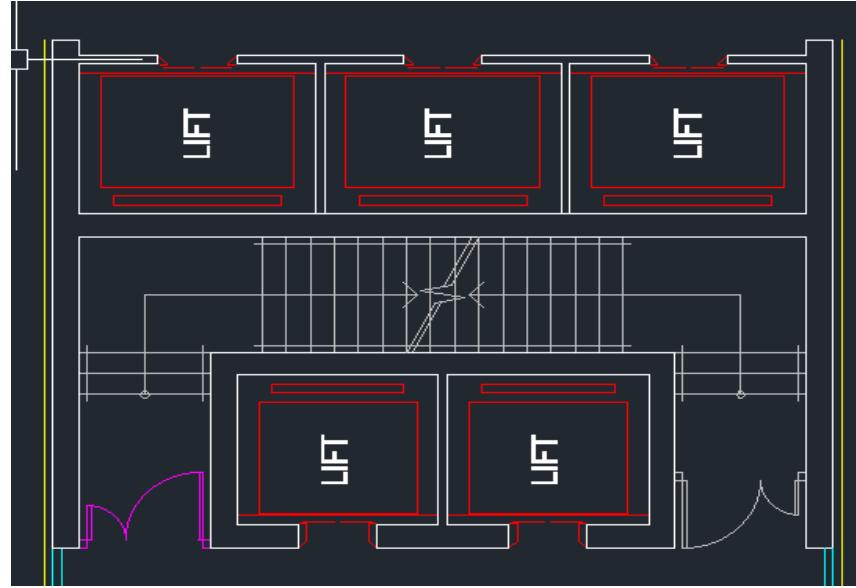
f2

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# 1.1 Point clouds compared to CAD/BIM drawings



- ✓ Rich in details and 3D appearance (*texture*)
- ✓ Consistent with the real 3D layouts (*z*)
- ✗ A lot of defects, e.g., sparse, noisy, and misaligned
- ✗ Unstructured, low semantic info, massive disk size



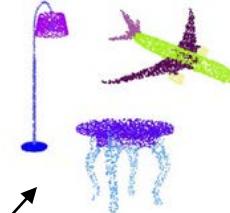
- ✓ Precise, compact, and parametric geometry (*x, y*)
- ✗ A lack of appearance
- ✗ Possibly inconsistent with the real 3D layouts



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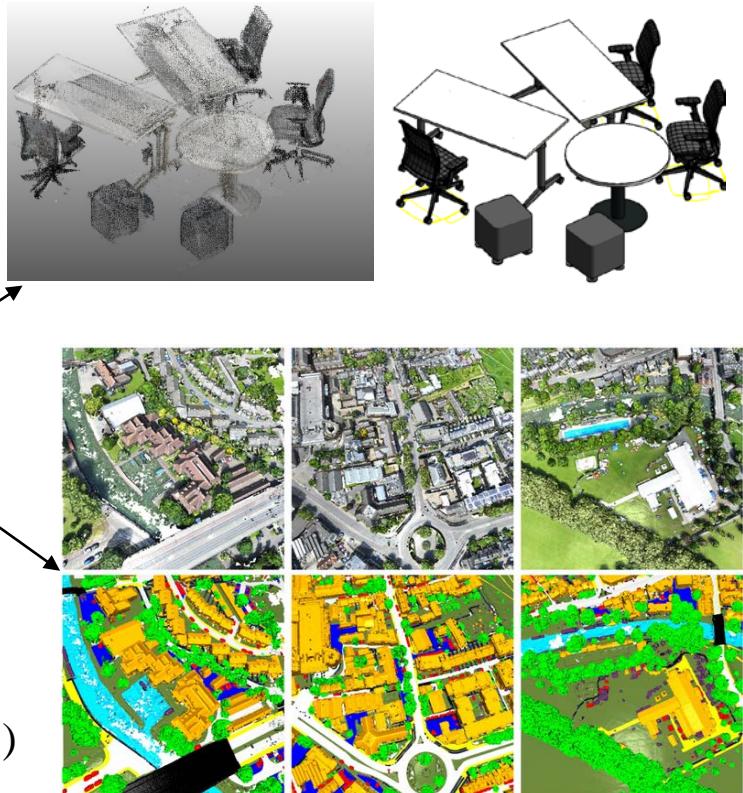
## 1.2 PC Understanding

### ◆ Urban point cloud understanding include

- 3D classification → 
- 3D object detection
- 3D semantic segmentation
- 3D parts and combinations
- 3D scene recognition
- 3D relations/topology recognition
- 4D motions (construction site, auto-driving)

### ◆ Related, but different from

- Image understanding (2D)
- Point cloud processing (registration, editing, etc.)



SensatUrban (Hu et al, 2021)

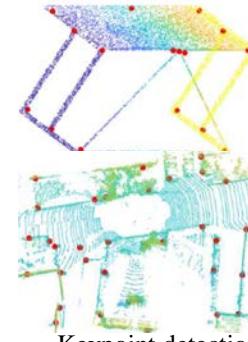


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## 1.2 Some existing methods for the tasks

### ◆ 3D classification 分类

- Spatial/shape features (Corner, SIFT, etc.)
- CNN deep features (e.g., *PointNet++*)
  - GraphNN features



Keypoint detection  
(Li & Lee 2019)



Example of semantic segmentation (Qi et al. 2017)

### ◆ 3D object detection 目标检测

- RANSAC (Random Sampling Consensus)
- Perfect normals + geometric shapes (e.g., walls, ceiling)

### ◆ 3D semantic segmentation 语义分割

- Sliding windows / region proposal / anchorless + 3D classification

### ◆ 3D scene / relationship 场景、关系

- 3D Object/parts/topology/semantics-based

### ◆ Most are general, any Building/Urban characteristic? 特色





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# 1.3 Symmetry and similarity as domain-specific

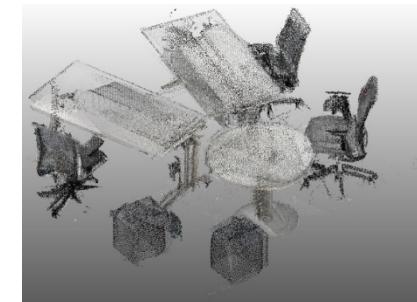
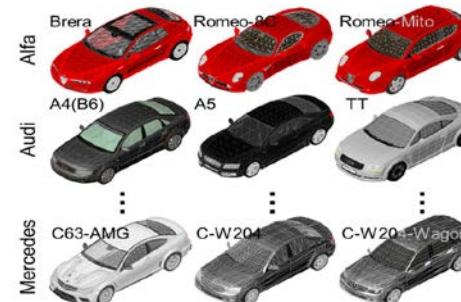
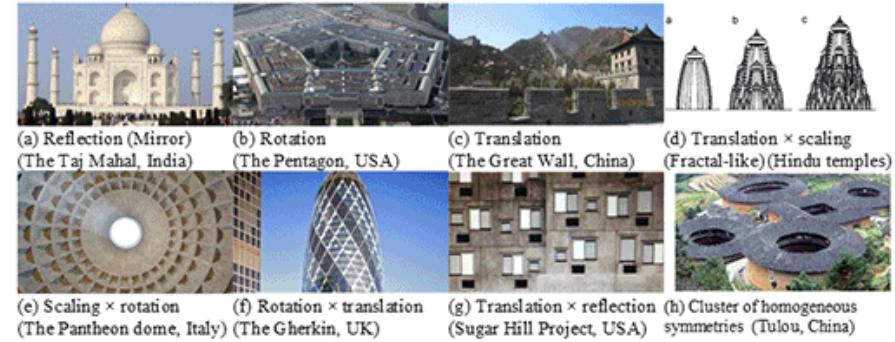
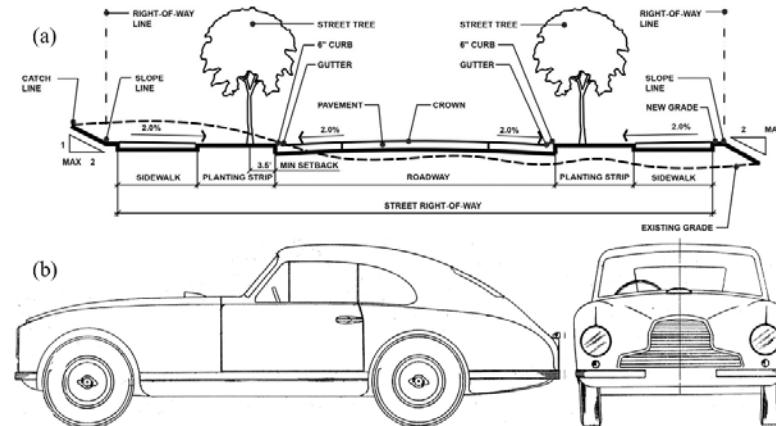
## ◆ Symmetry 对称

□ Reflect<sub>x</sub>(C) ≈ C 镜面对称

## ◆ Similarity 相似

□ AffineTrans<sub>x</sub>(C<sub>1</sub>) ≈ C<sub>2</sub> 仿形变换

## ◆ Guided by design/engineering laws



(Xue et al, 2019d)

Section 2

# **SYMMETRY AND SIMILARITY DETECTION**

## 对称、相似性检测



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## 2.1 “White-box” formulations for optimization

### ◆ “White-box” objective function 白盒目标

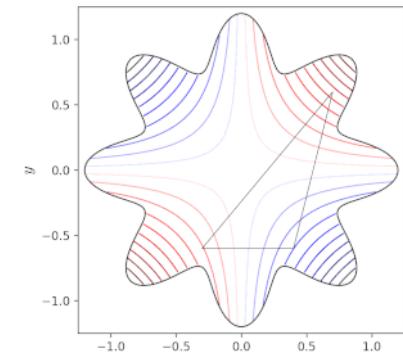
- $f_{\mathbf{x}} = \text{RMSE}(\text{Reflect}(C, \mathbf{x}), C)$
- $f_{\mathbf{x}} = \text{RMSE}(\text{AffineTrans}(C_1, \mathbf{x}), C_2)$

□ Pre-requisite:

- Closer densities
- RMSE can be any error metric



or



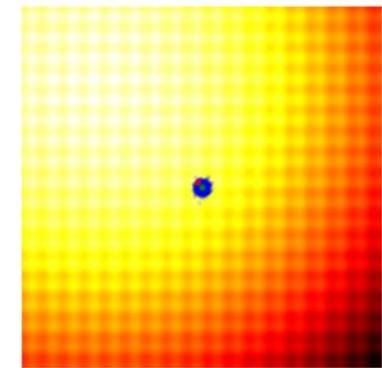
Nelder-Mead Source: Wikipedia.org

### ◆ Nonlinear optimization formulation 统一形式: 非线性优化

- $\arg \min f_{\mathbf{x}}$
- s.t.  $x$  in Range
- Constraints  $(x) \leq 0$

### ◆ 50+ off-the-shelf solvers for complex optimization 可用算法

- In C++/Python/... (see right)



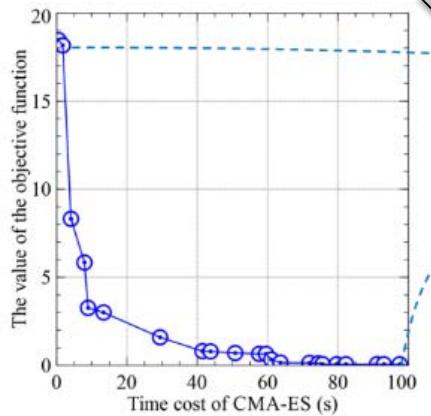
CMA-ES Source: otoro.net



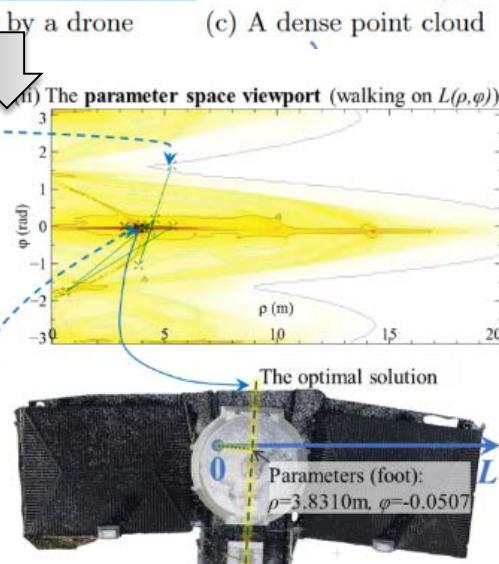
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## 2.1.1 Symmetry detections (early, vs updated)

PCR = 93.7%, Time = 98.6s (Xue et al. 2019d)



(i) The optimization viewport (descending of the objective function by CMA-ES)



◆ Updated in (Xue et al. 2019a)

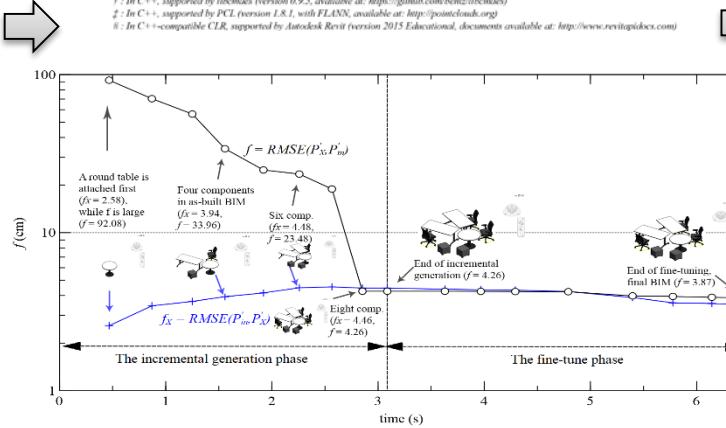
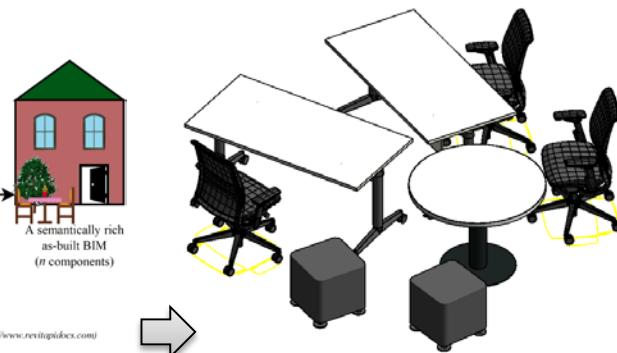
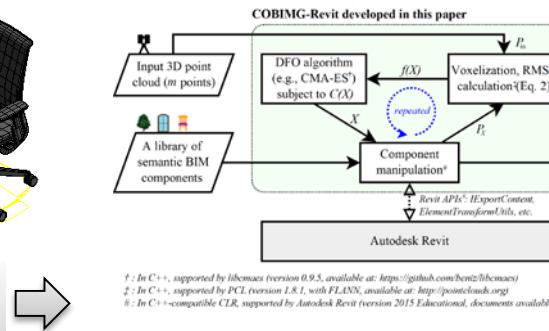
Id	Thumbnail of input point cloud <sup>a</sup> (C)	Normal <sup>b</sup> of Sym. <sup>c</sup> segmented	PCR (%)	Time (s)	Intrinsic <sup>d</sup> asymmetry
1		$[1.026 \quad -0.046 \quad 0]$	86.29	0.81	
2		$[0.171 \quad 0.036 \quad 0]$	85.22	1.79	
3		$[-0.080 \quad 0.004 \quad 0]$	95.99	3.68	
4		$[-1.979 \quad 2.242 \quad 0]$	95.44	2.77	As circled
5					
6					
7		$[2.687 \quad -0.243 \quad 0]$	97.51	3.05	As circled

(a) As-designed asymmetry (car park entrance) of the No. 4 building

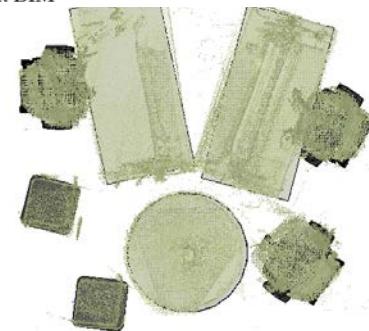
(b) As-built asymmetry on the deck of No. 7, where the east half is about 0.5m wider

Asymmetries  
局部非对称

- ◆ Time = 6.44s (Manual = 300s), RMSE = 3.87 cm



(a) A screenshot of the 3D view of the output as-built BIM



(b) A visual comparison between the input (grey points) and the output BIM



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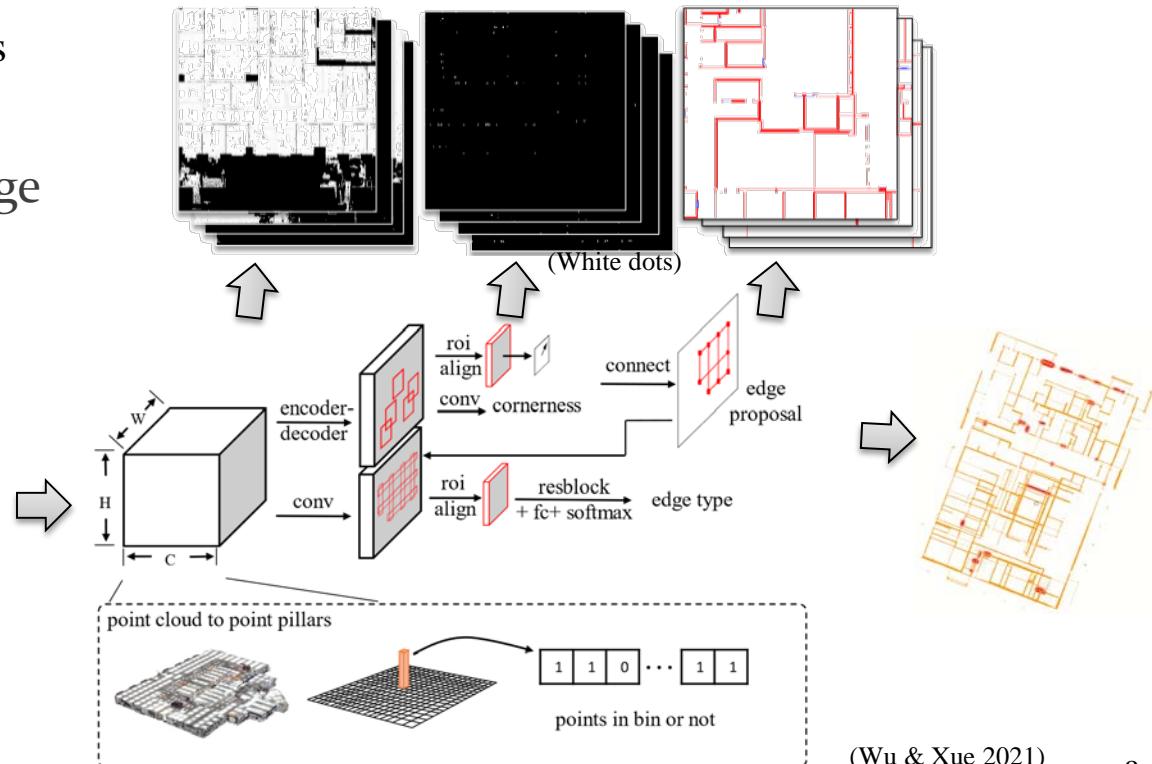
## 2.2 A “Black-box” formulation for deep learning

- ◆ Floor corners are symmetric, as an ML task 墙角

- Input: A top view of voxels
- Output: Corners and walls

- ◆ First Scan-to-BIM challenge

- 12% IoU for 2D track
- The Second Runner-up
- Plenty of room to improve





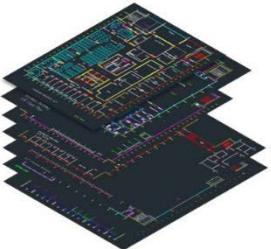
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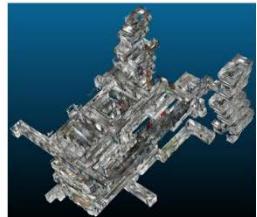
## 2.3 Case 1: Symmetry-guided as-built BIM 案例1

◆ Wu et al. (2021)

□ Merit Award, Hong Kong OpenBIM / OpenGIS Award

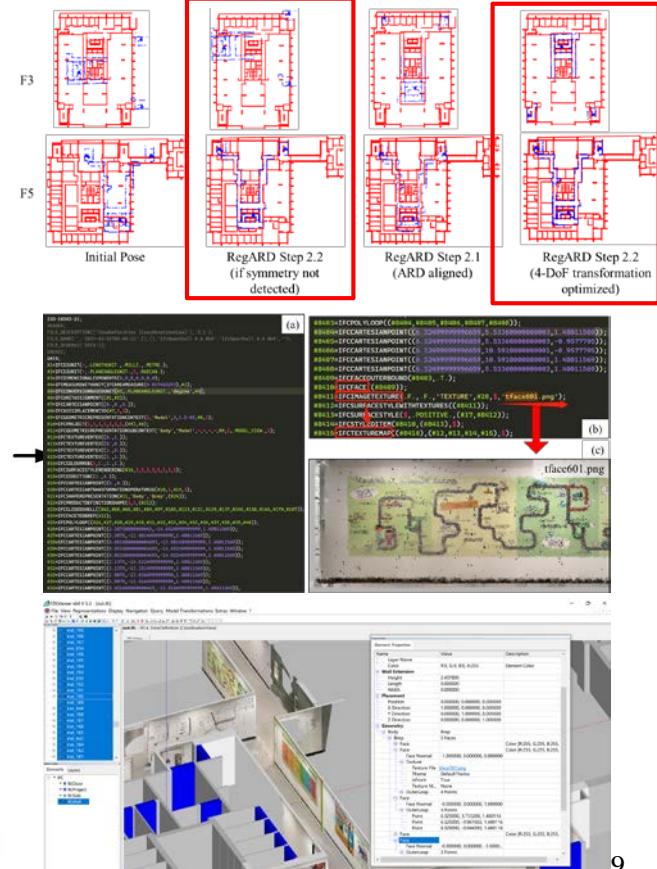
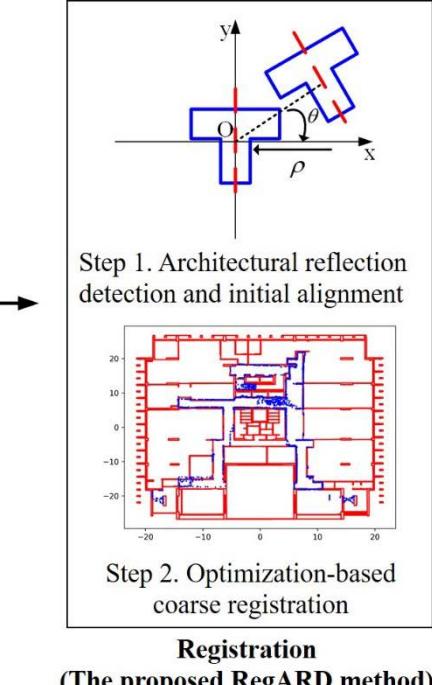
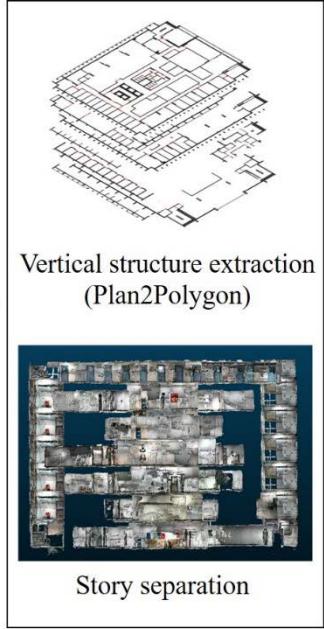


CAD drawings



Smartphone's point clouds

Inputs

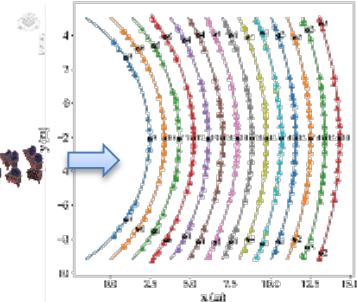
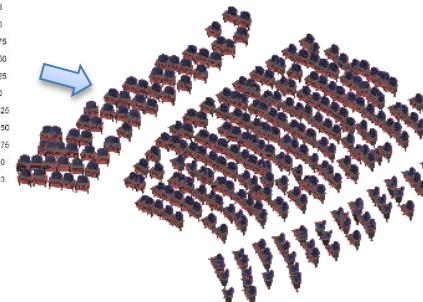
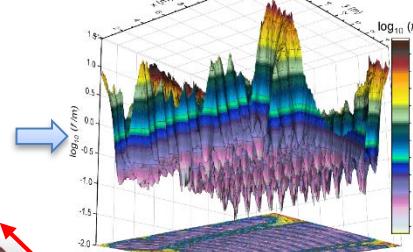




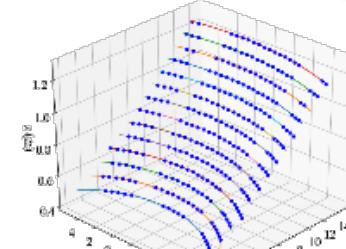
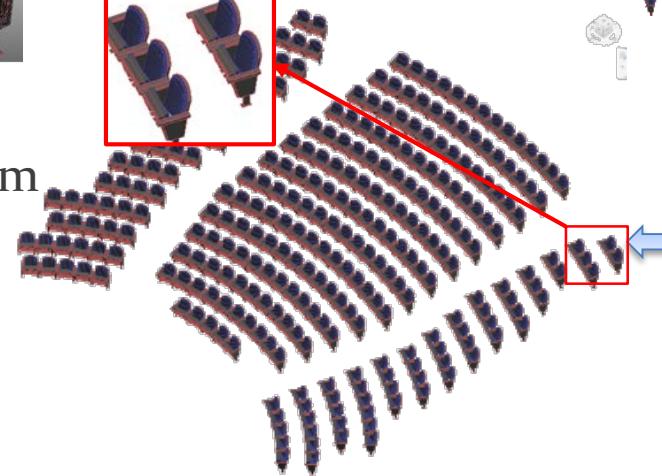
## 2.3 Case 2: Sim-guided many chairs 案例2 (Xue et al. 2019c)

f2a

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- ❖ Multi-modal algorithm
  - ▣ For noisy data
- ❖  $F_1 > 0.9$



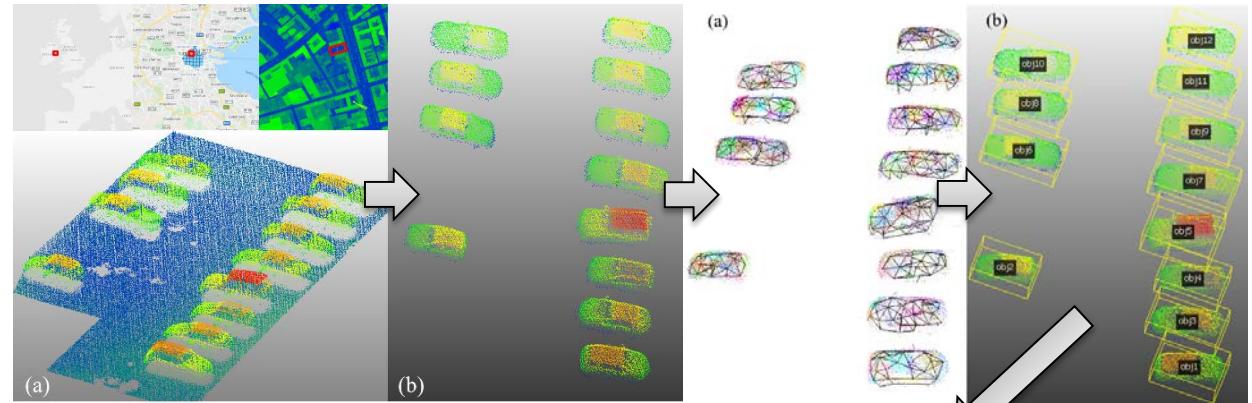


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## 2.3 Case 3: Sym+Sim for city objects 案例3 (1/2)

### ◆ Symmetry-based cross-sections 对称截面 (Xue et al. 2020)

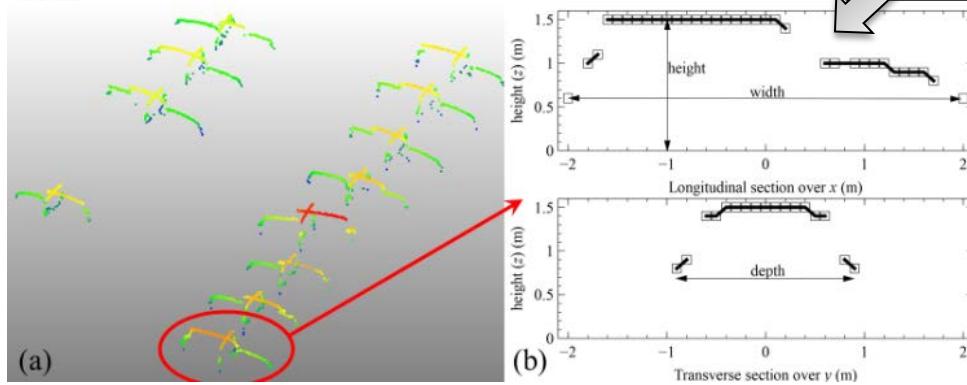
- 1. Ground removal
- 2. Connectedness
- 3. Major symmetry
  - 3.1 Section #1
- 4. Perpendicular
  - 4.1 Section #2
- 5. Voxelization



### ◆ For **unknown** objects

无需语义分割

- Symmetric
- Above ground

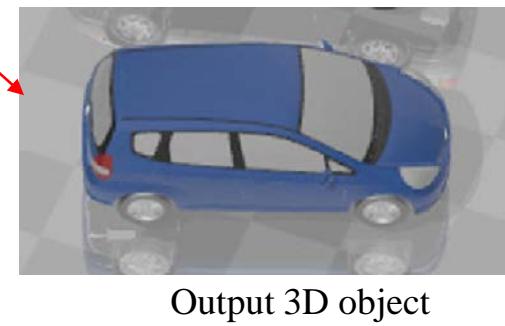
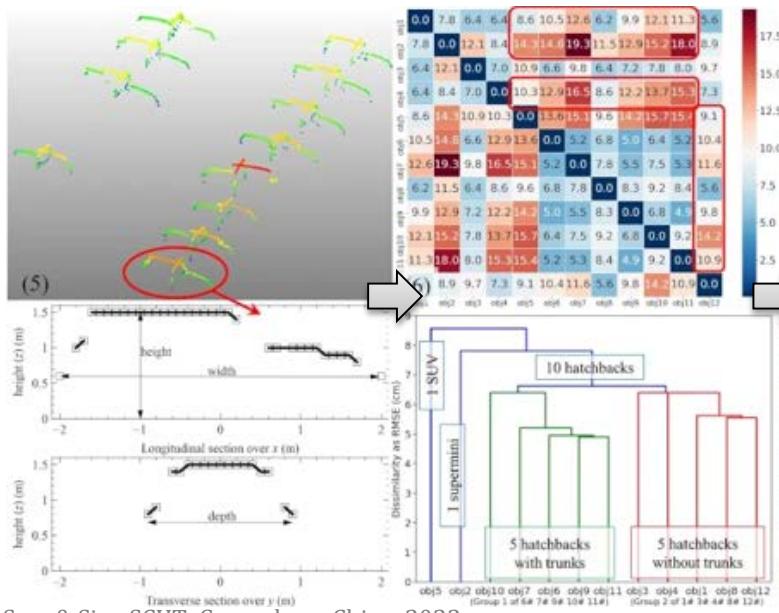




## 2.3 Case 3: Sym+Sim for city objects 案例3 (2/2)



- ◆ Similarity for clustering unknown **相似聚类**
  - 1. Cross-section-based registration
  - 2. Clustering using least RMSE
  - ◆ Similarity to sections of known 3D objects **匹配已知语义模型**
  - 1. Filters (Width, Height, Depth)



## Output 3D object

Section 3

# SUMMARY

小结



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## 3.1 A recap

### ◆ Urban point cloud 城市点云

- Has advantages for buildings/urban applications
- Understanding is a must-to-do for machines

### ◆ Symmetry and similarity 对称、相似

- Can be formulated as
  - “white-box” formulations
  - “black-box” formulation
- Very powerful for understanding a point cloud if detected
  - Sometimes better than the factual (e.g., cars, chairs)

### ◆ Yet, 有待研究

- There is plenty of room to improve





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## 3.2 Some personal view points 个人看法

### ◆ Artificial Neural Networks overheating? 过热

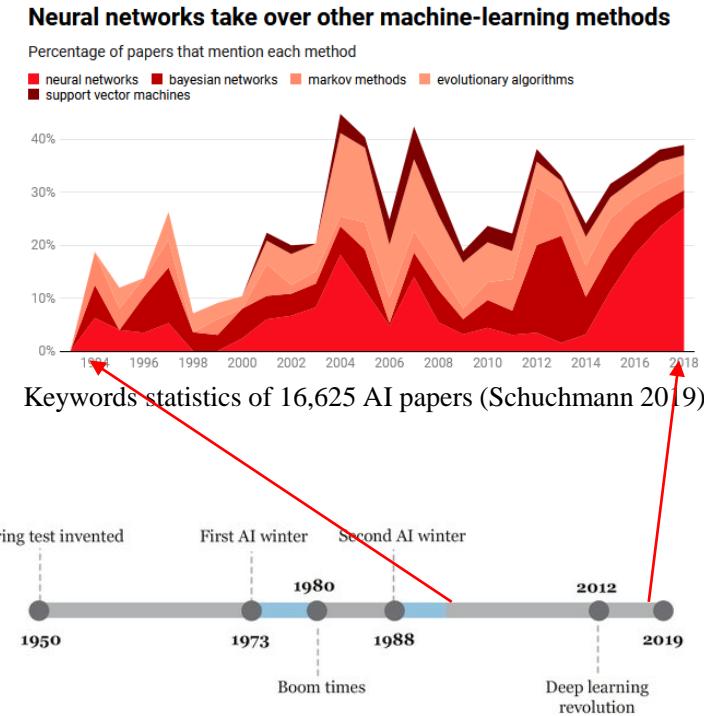
- Anyway, the “Evolutionary Algorithms” for our “white-box” modeling are still viable

### ◆ AI winters or capital winters? AI寒冬

- 1973: Exiting Bretton Woods system (布雷顿森林体系)
- 1987: “Black Monday” stock market crash

### ◆ Beyond symmetry and similarity? 未来工作

- Shape grammars (on-going)
- Between interior and exterior (on-going)
- Semi-supervised learning of relations





# References

- ◆ Hu, Q., Yang, B., Khalid, S., Xiao, W., Trigoni, N., & Markham, A. (2021). Towards semantic segmentation of urban-scale 3D point clouds: A dataset, benchmarks and challenges. In *Proceedings of the IEEE/CVF conference on computer vision and pattern recognition* (pp. 4977-4987).
- ◆ Li, J., & Lee, G. H. (2019). Usip: Unsupervised stable interest point detection from 3d point clouds. In *Proceedings of the IEEE/CVF International Conference on Computer Vision* (pp. 361-370).
- ◆ Qi, X., Liao, R., Jia, J., Fidler, S., & Urtasun, R. (2017). 3d graph neural networks for rgbd semantic segmentation. In *Proceedings of the IEEE International Conference on Computer Vision* (pp. 5199-5208).
- ◆ Schuchmann, S. (2019). *Analyzing the prospect of an approaching AI winter*
- ◆ Tan, T., Chen, K., Lu, W., & Xue, F. (2019, September). Semantic enrichment for rooftop modeling using aerial LiDAR reflectance. In *2019 IEEE International Conference on Signal Processing, Communications and Computing (ICSPCC)* (pp. 1-4). IEEE.
- ◆ Wu, Y., Shang, J., & Xue, F. (2021). RegARD: Symmetry-Based Coarse Registration of Smartphone's Colorful Point Clouds with CAD Drawings for Low-Cost Digital Twin Buildings. *Remote Sensing*, 13(10), 1882.
- ◆ Wu, S., Yang, Z., Ding, X., Zhang, B., Zhang, L., & Lu, Z. (2020a). Two decades of settlement of hong kong international airport measured with multi-temporal InSAR. *Remote Sensing of Environment*, 248, 111976.
- ◆ Wu, S., Le, Y., Zhang, L., & Ding, X. (2020b). Multi-temporal InSAR for Urban Deformation Monitoring: Progress and Challenges. **雷达学报**, 9(2), 277-294.
- ◆ Xu, J., Chen, K., Xue, F., & Lu, W. (2018). 3D point cloud data enabled facility management: A critical review. In *The 23rd International Symposium on Advancement of Construction Management and Real Estate, CRIOCM2018*. [https://doi.org/10.1007/978-981-15-3977-0\\_49](https://doi.org/10.1007/978-981-15-3977-0_49)
- ◆ Xue, F., Lu, W., Chen, Z., & Webster, C. J. (2020a). From LiDAR point cloud towards digital twin city: Clustering city objects based on Gestalt principles. *ISPRS Journal of Photogrammetry and Remote Sensing*, 167, 418-431. (2020 Featured Article)
- ◆ Xue, F., Wu, L., & Lu, W. (2021). Semantic enrichment of building and city information models: A ten-year review. *Advanced Engineering Informatics*, 47, 101245.
- ◆ Xue, F., Lu, W., Chen, K. (2018). Automatic generation of semantically rich as-built building information models using 2D images: A derivative-free optimization approach. *Computer-Aided Civil and Infrastructure Engineering*, 33(11), 926-942.
- ◆ Xue, F., Lu, W., Webster, C. J., & Chen, K. (2019a). A derivative-free optimization-based approach for detecting architectural symmetries from 3D point clouds. *ISPRS Journal of Photogrammetry and Remote Sensing*, 148, 32-40.
- ◆ Xue, F., Lu, W., Chen, K., & Zetkulic, A. (2019b). From Semantic Segmentation to Semantic Registration: Derivative-Free Optimization-Based Approach for Automatic Generation of Semantically Rich As-Built Building Information Models from 3D Point Clouds. *Journal of Computing in Civil Engineering*, 33(4), 04019024.
- ◆ Xue, F., Lu, W., Chen, K., & Webster, C. J. (2019c). BIM reconstruction from 3D point clouds: A semantic registration approach based on multimodal optimization and architectural design knowledge, *Advanced Engineering Informatics*, 42, 100965.
- ◆ Xue, F., Chen, K., & Lu, W. (2019d). Architectural symmetry detection from 3D urban point clouds: A derivative-free optimization (DFO) approach. In *Advances in Informatics and Computing in Civil and Construction Engineering* (pp. 513-519). Springer, Cham.
- ◆ Xue, F., Chen, K., & Lu, W. (2019e). Understanding unstructured 3D point clouds for creating digital twin city: An unsupervised hierarchical clustering approach. *CIB World Building Congress 2019*.
- ◆ Xue, F., Lu, W., Tan, T., & Chen, K. (2019f). Semantic enrichment of city information models with LiDAR-based rooftop albedo. In *Sustainable Buildings and Structures: Building a Sustainable Tomorrow* (pp. 207-212). CRC Press.
- ◆ Zhu, X. X., & Bamler, R. (2014). Superresolving SAR tomography for multidimensional imaging of urban areas: Compressive sensing-based TomoSAR inversion. *IEEE Signal Processing Magazine*, 31(4), 51-58.



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