



THE UNIVERSITY OF HONG KONG 香港大學
faculty of architecture 建築學院



iLab | @HKURBAN
the urban big data lab



华南理工大学 | 土木与交通学院
SCHOOL OF CIVIL ENGINEERING & TRANSPORTATION
SOUTH CHINA UNIVERSITY OF TECHNOLOGY

SIR Frontiers @ SCUT
学术前沿系列讲座

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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29 Apr 2022 , Guangzhou, China





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Outline

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

◆ 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)



iLabHKU



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



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the urban big data lab



0.2 About me



Homepage
(full-text PDFs)

◆ A mixed background 背景

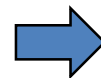
- ▣ BEng in Automation, CAUC
- ▣ MSc in Computer Science, CAUC
- ▣ PhD in System Engineering, HKPU
- ▣ PDF/RAP/AP in Construction IT

◆ 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



2004
2007
2012



◆ Engineering

- ▣ ISE, CEM, EIE

◆ Computer Science

- ▣ AI, OR, ML

◆ Economics

- ▣ SCM

◆ Professional

- ▣ MACM, SMC GS, MIEEE, MHKGISA
- ▣ V.C. ACM-HK, Com. CGS-BIM



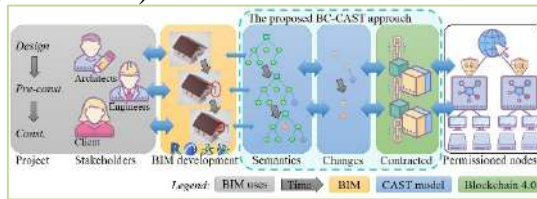
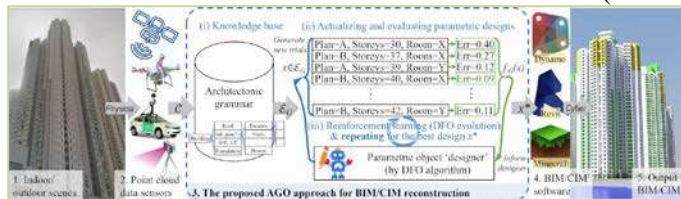
0.2 Recent research projects



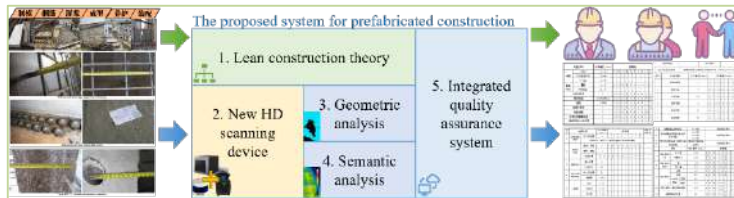
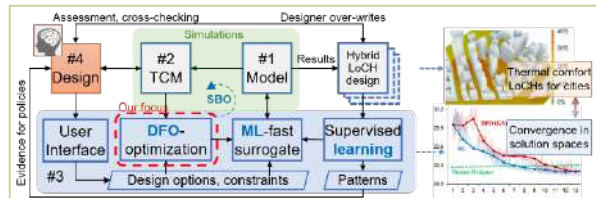
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◆ On-going 在研

■ 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.

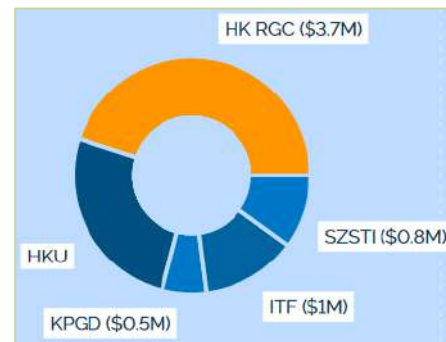
◆ Keywords

■ BIM/CIM

■ 3D point cloud

■ Derivative-free optimization

■ Urban semantics



Sponsors of projects as PI/PC/Co-PI

The background of the slide is a photograph of a large, classical-style building with a prominent clock tower on the right side. The building features a series of columns and arches. In the foreground, there are lush green palm trees and other tropical foliage. The overall scene is brightly lit, suggesting a sunny day.

Section 1

URBAN POINT CLOUDS

城市点云



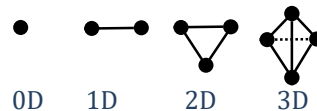
1 Introduction



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◇ Point 点

- ▣ A **location** in space, 0D (no width, length, or thickness)
- ▣ Structured format: $\{x, y, z\}$, $[R, G, B, N_x, N_y, N_z, Cls, Int., \dots]$



◇ 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



A point cloud of HKU Campus (Source: Author, 2019)



A close look of cloud at Mount Hua (Source: Author)

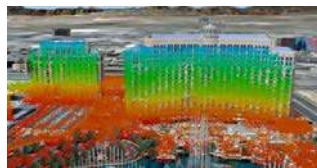
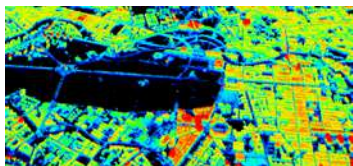


1.1 Major sources of urban point clouds



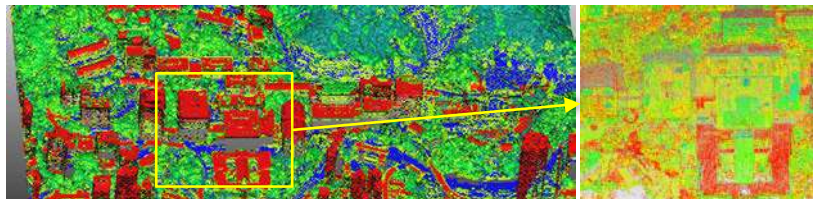
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◇ SAR 合成孔径雷达



Bellagio Hotel, Las Vega
(Zhu & Bamler 2014)

◇ LiDAR 光达

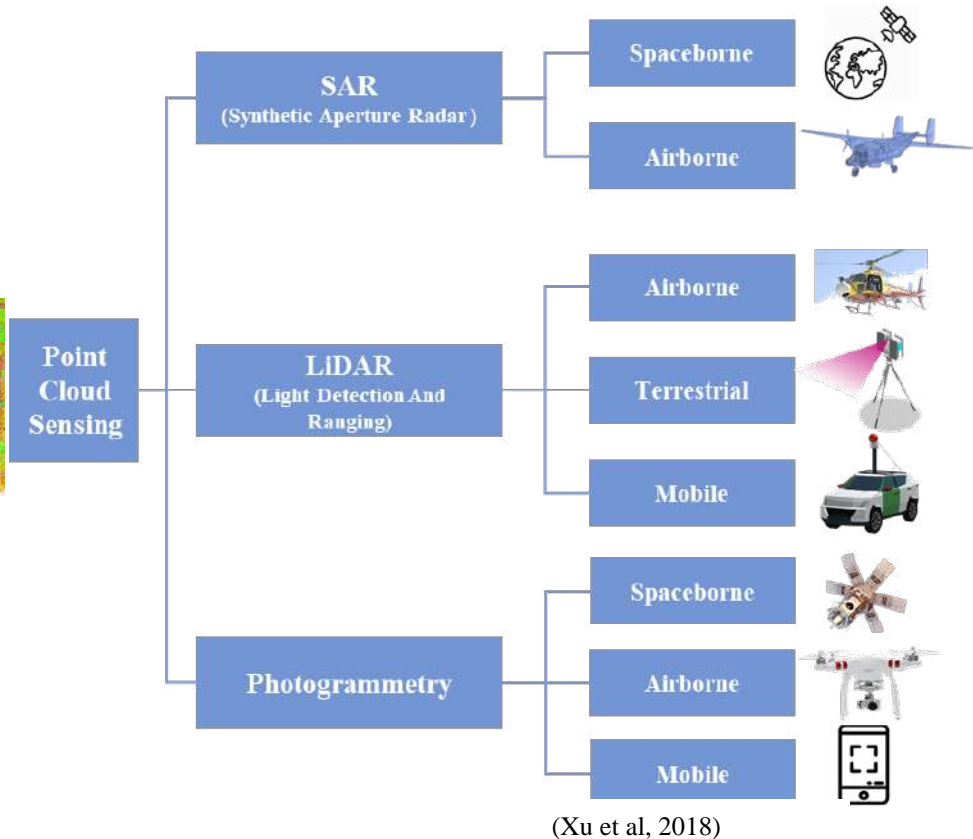


HKU Campus (Xue et al, 2019f)

◇ Photogrammetry 摄影测量



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





1.1 Advantages and applications



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◆ SAR 合成孔径雷达

- mm-accuracy
- Coverage

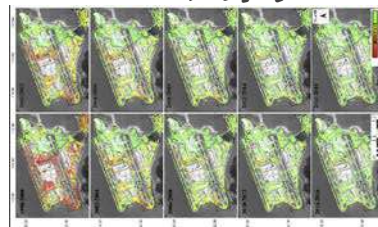
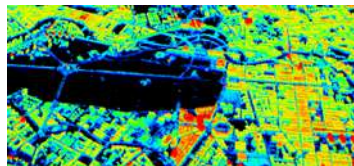
◆ LiDAR 光达

- mm/cm/dm
- No distortion
- Intensity

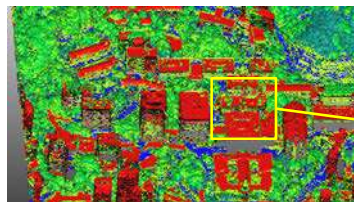
◆ Photogrammetry 摄影测量

- cm-accuracy
- Colorful
- Cheaper

◆ Use cases 用例



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



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



(a) Colours plot overlaid on a Google Earth city model



Kowloon Wall City 3D model
(Source: patrick-@sketchfab.com)

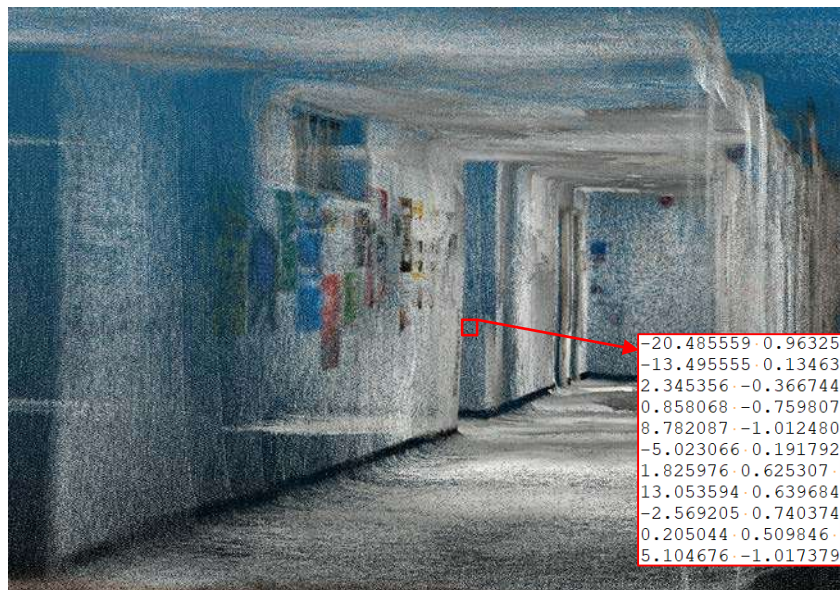


HKU @Minecraft (Source: Author, 2021)



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1.1 Point clouds compared to CAD/BIM drawings

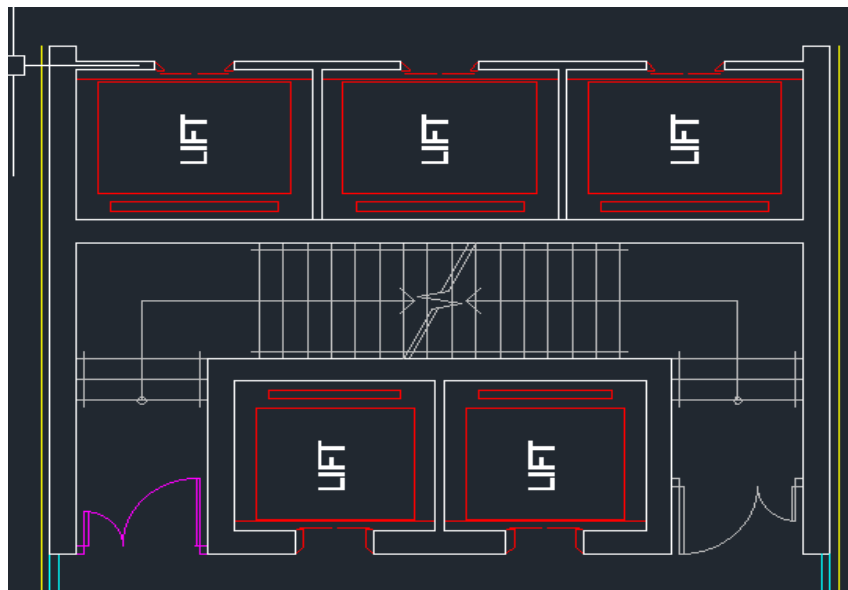


✓Rich in details and 3D appearance (*texture*)

✓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



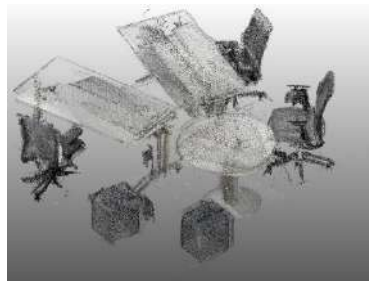
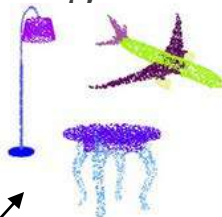
1.2 PC Understanding

◆ 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.)





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

◇ 3D object detection 目标检测

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

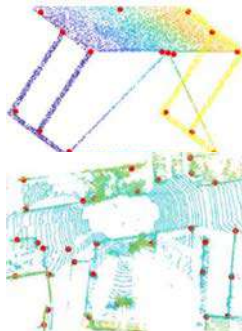
◇ 3D semantic segmentation 语义分割

- ▣ 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



Keypoint detection
(Li & Lee 2019)



Example of semantic segmentation (Qi et al. 2017)





1.3 Symmetry and similarity as domain-specific

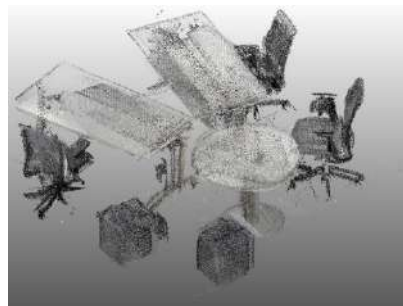
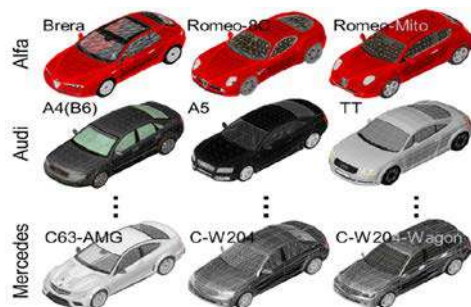
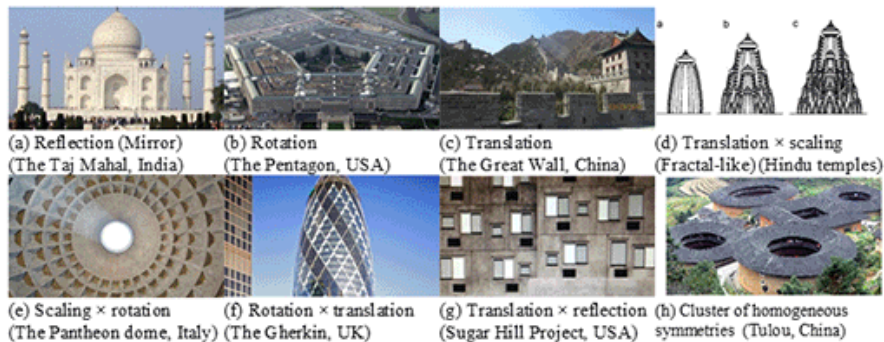
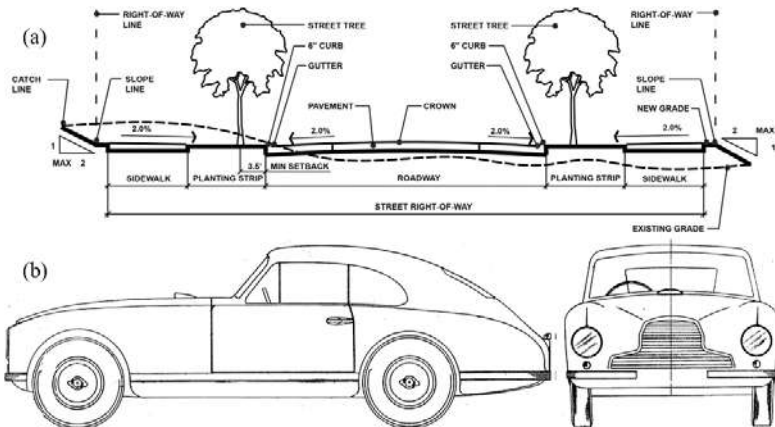
◇ Symmetry 对称

□ $\text{Reflect}_x(C) \approx C$ 镜面对称

◇ Similarity 相似

□ $\text{AffineTrans}_x(C_1) \approx C_2$ 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



2.1 “White-box” formulations for optimization

◇ “White-box” objective function 白盒目标

▣ $f_x = \text{RMSE}(\text{Reflect}(C, \mathbf{x}), C)$

▣ $f_x = \text{RMSE}(\text{AffineTrans}(C_1, \mathbf{x}), C_2)$

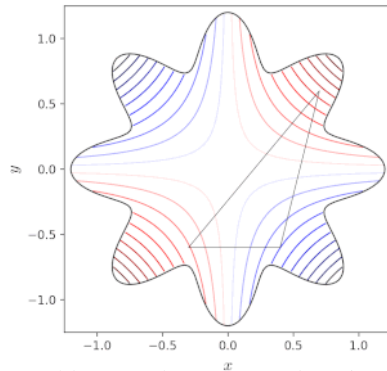
▣ Pre-requisite:

○ 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

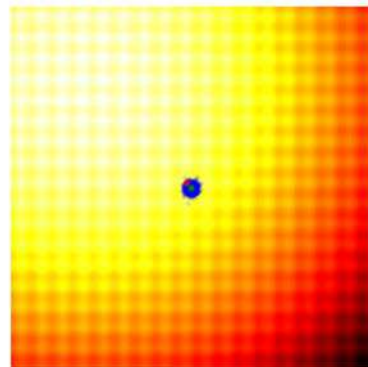
▣ $\arg \min f_x$

▣ s.t. x in $Range$

▣ Constraints $(x) \leq 0$

◇ 50+ off-the-peg solvers for complex optimization 可用算法

▣ In C++/Python/... (see right)



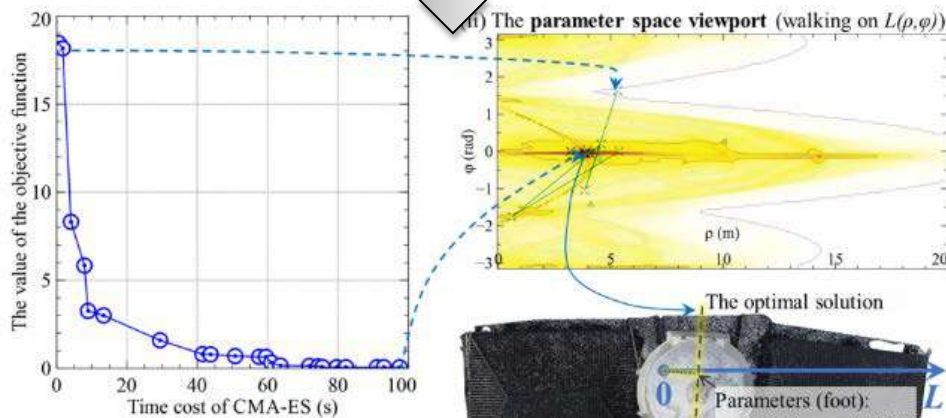
CMA-ES Source: otoro.net



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2.1.1 Symmetry detections (early, vs updated)

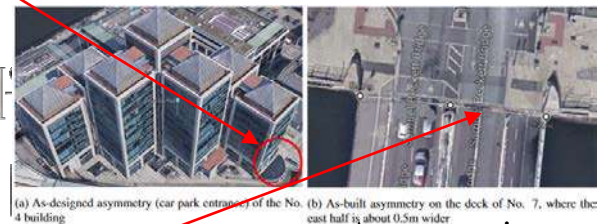
■ PCR = 93.7%, Time = 98.6s (Xue et al. 2019d)



(iii) The Point cloud viewport (testing a series of symmetries)

◆ Updated in (Xue et al. 2019a)

Id	Thumbnail of input point cloud ^a (C)	Normal ^b of Sym.	Symmetric pairs ^c segmented	PCR (%)	Time (s)	Intrinsic ^d asymmetry
1		$\begin{bmatrix} 1.026 \\ -0.046 \\ 0 \end{bmatrix}$		86.29	0.81	
2		$\begin{bmatrix} 0.1711 \\ 0.036 \\ 0 \end{bmatrix}$		85.22	1.79	
3		$\begin{bmatrix} -0.080 \\ 0.004 \\ 0 \end{bmatrix}$		95.99	3.68	
4		$\begin{bmatrix} -1.979 \\ 2.242 \\ 0 \end{bmatrix}$		95.44	2.77	As circled
5						
6						
7		$\begin{bmatrix} 2.687 \\ -0.243 \\ 0 \end{bmatrix}$		97.51	3.05	As circled



Asymmetries
局部非对称

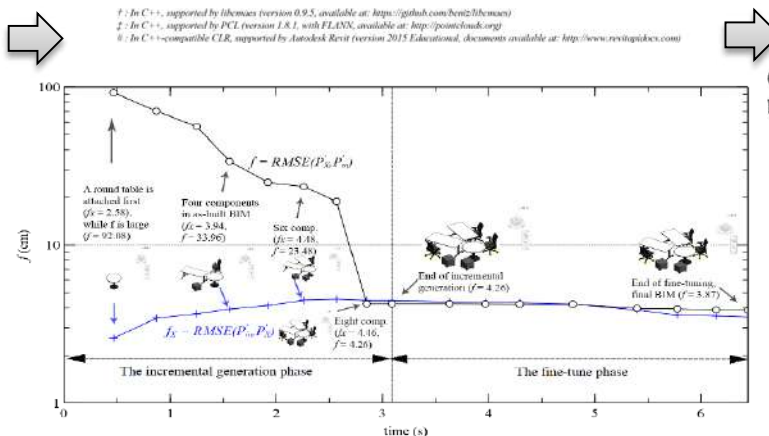
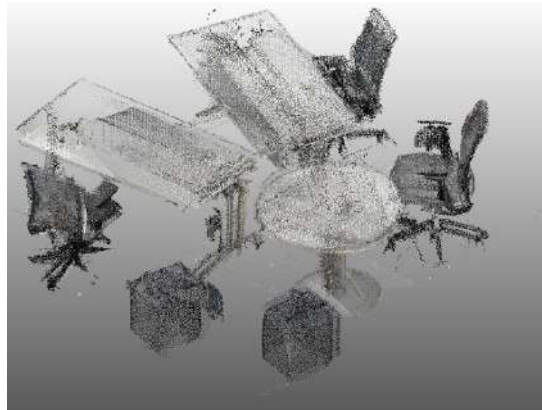
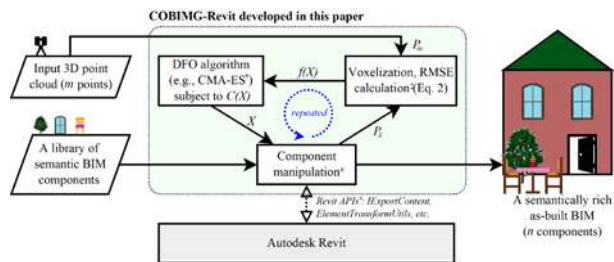


2.1.2 Similarity detection for as-built BIM (Xue et al. 2019b)

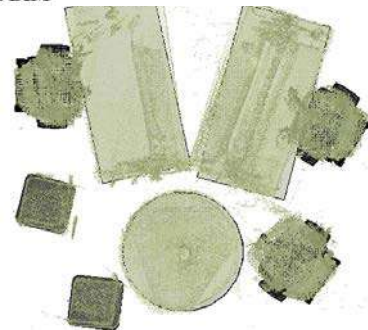


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◆ Time = 6.44s (Manual = 300s), RMSE = 3.87 cm



(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



2.2 A “Black-box” formulation for deep learning



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◆ Floor corners are symmetric, as an ML task 墙角

▣ Input: A top view of voxels

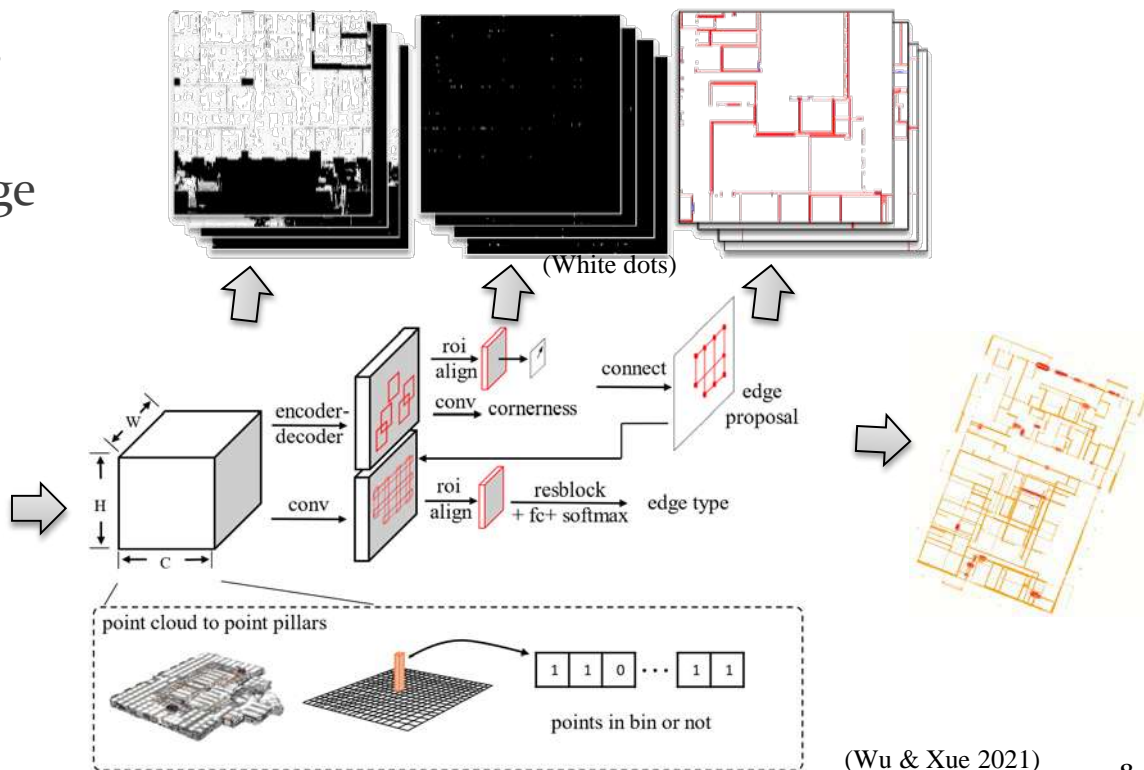
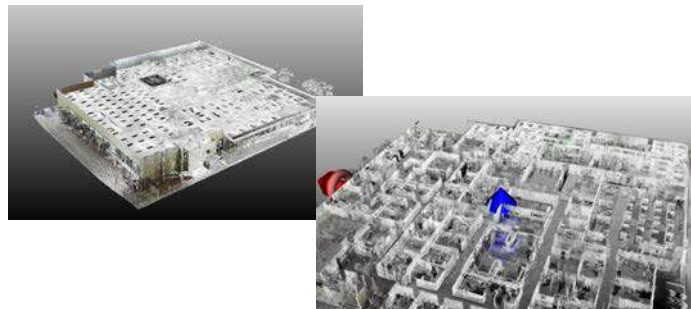
▣ Output: Corners and walls

◆ First Scan-to-BIM challenge

▣ 12% IoU for 2D track

▣ The Second Runner-up

▣ Plenty of room to improve



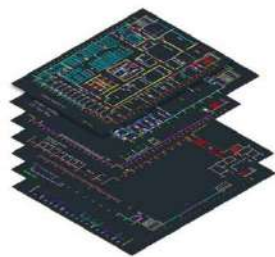


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2.3 Case 1: Symmetry-guided as-built BIM 案例1

◇ Wu et al. (2021)

▣ Merit Award, Hong Kong OpenBIM / OpenGIS Award

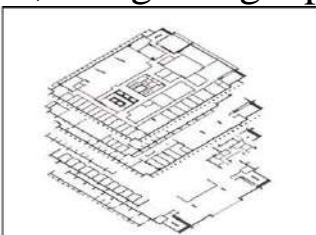


CAD drawings



Smartphone's point clouds

Inputs

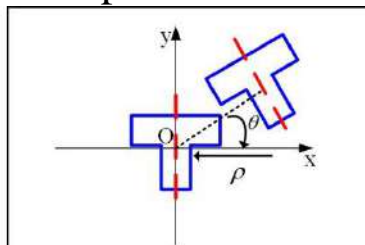


Vertical structure extraction
(Plan2Polygon)

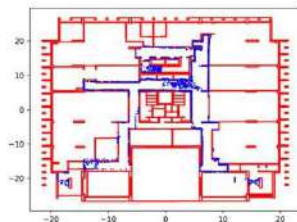


Story separation

Preprocessing
(In-house developed)

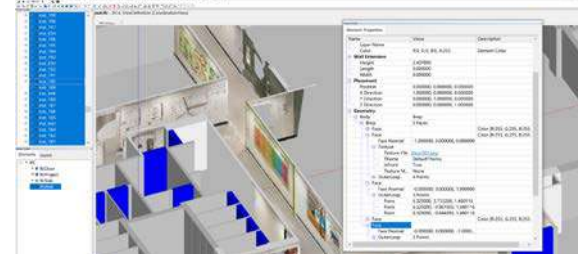
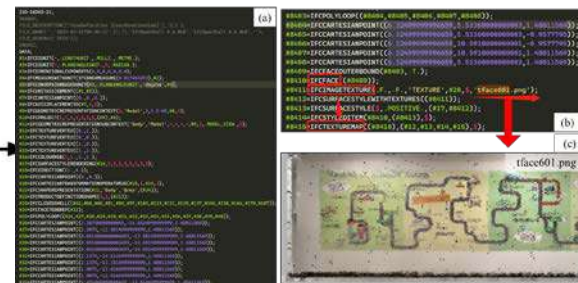
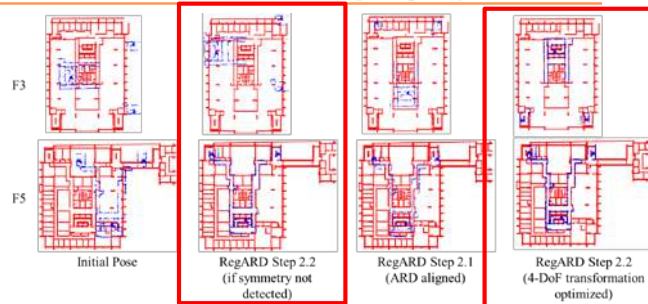


Step 1. Architectural reflection
detection and initial alignment



Step 2. Optimization-based
coarse registration

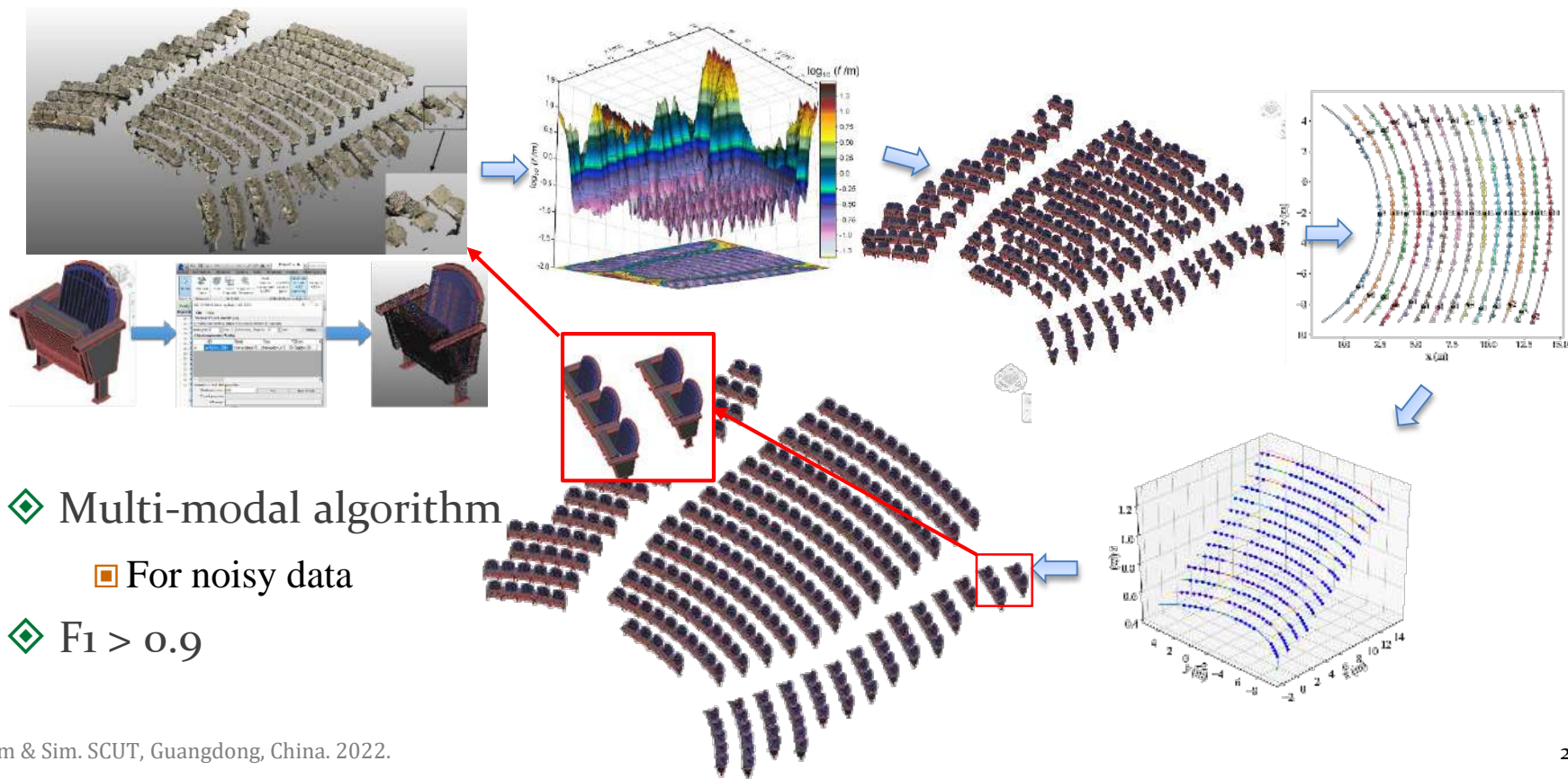
Registration
(The proposed RegARD method)





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





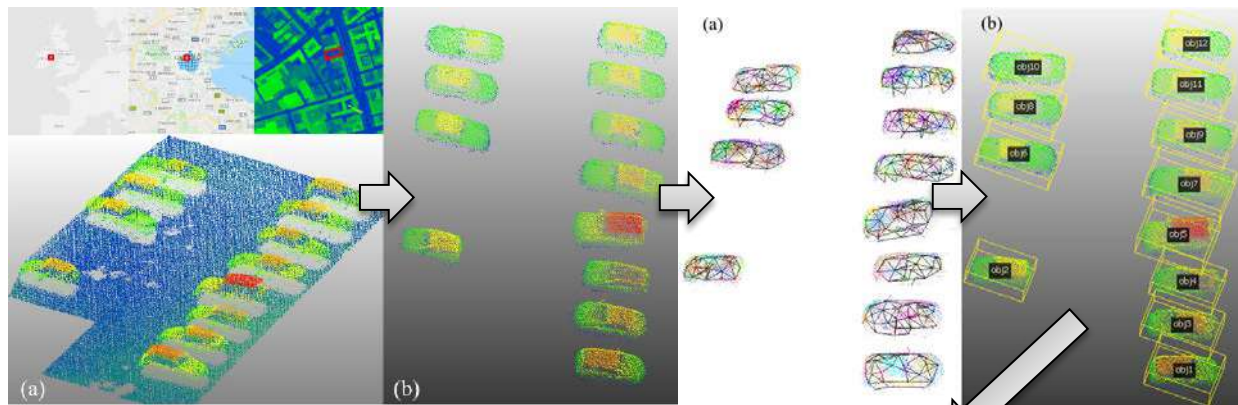
2.3 Case 3: Sym+Sim for city objects 案例3 (1/2)



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

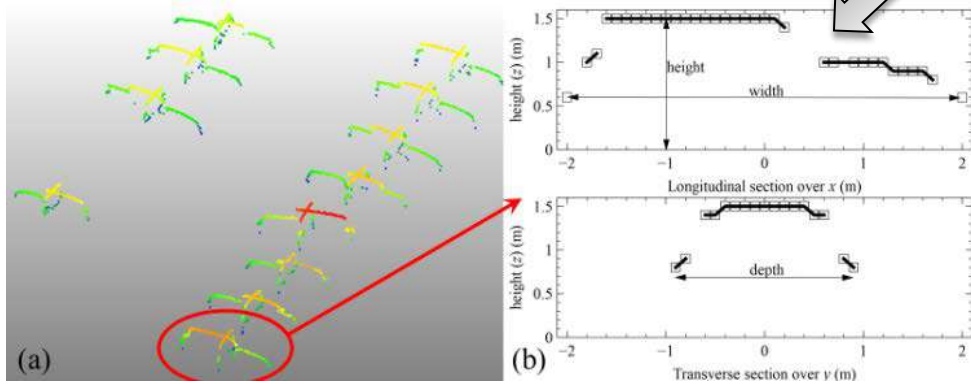
- ▣ 1. Ground removal
- ▣ 2. Connectedness
- ▣ 3. Major symmetry
 - 3.1 Section #1
- ▣ 4. Perpendicular
 - 4.1 Section #2
- ▣ 5. Voxelization



◇ 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)



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



Section 3

SUMMARY

小结



3.1 A recap



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◆ 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





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

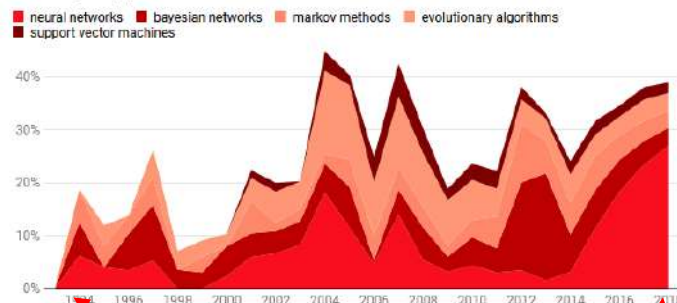
- ▣ 1973: Exiting Bretton Woods system (布雷顿森林体系)
- ▣ 1987: “Black Monday” stock market crash

◇ Beyond symmetry and similarity? 未来工作

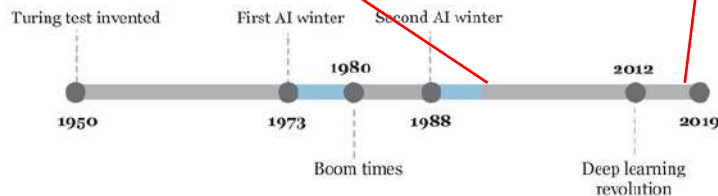
- ▣ Shape grammars (on-going)
- ▣ Between interior and exterior (on-going)
- ▣ Semi-supervised learning of relations

Neural networks take over other machine-learning methods

Percentage of papers that mention each method



Keywords statistics of 16,625 AI papers (Schuchmann 2019)





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Keep awesome!

感谢！ 欢迎提问

