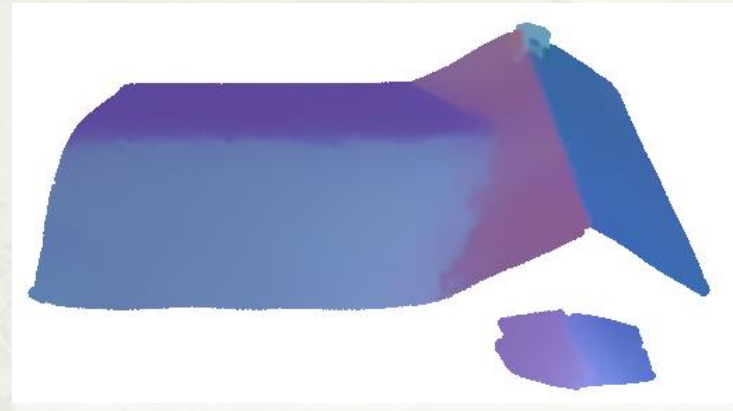


# Feature-Enhancing Aerial LiDAR Point Cloud Refinement



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# Outline

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- \* Introduction
- \* Approach Overview
- \* Implementation Details
- \* Evaluation
- \* Conclusion and Future Work
- \* Acknowledgements

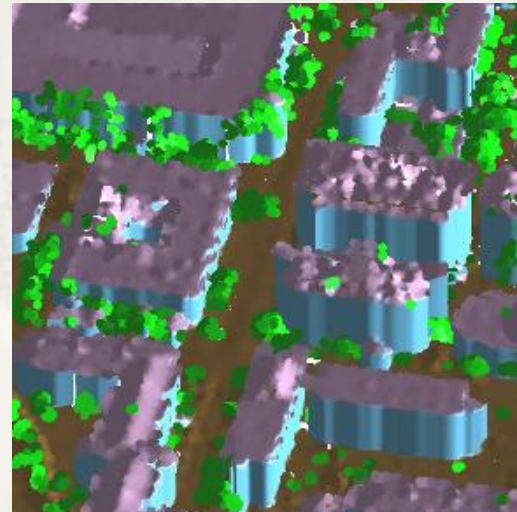
# Why Refinement?

## Defects of raw point clouds

- \* Noise
- \* Under-sampling

## ➔ Visual artifacts in applications

- \* Grainy planes
- \* Gaps and holes
- \* Bumpy boundaries



Direct Rendering

# Challenges of Aerial LiDAR Points

- \* Sparse sampling density
- \* 2.5D nature



Aerial LiDAR

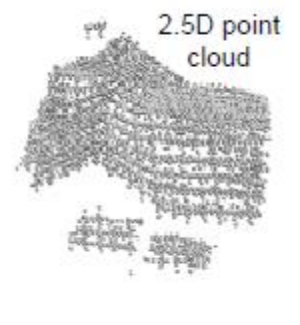
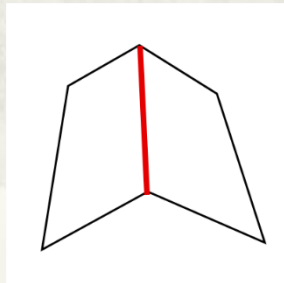


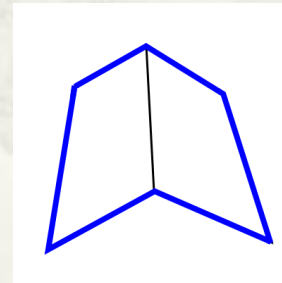
Image courtesy to Qian-Yi Zhou

# Preliminary Definitions

- \* Normal discontinuous feature
  - \* Discontinuities in normals
  - \* The underlying surface is still continuous
- \* Position discontinuous feature
  - \* Discontinuities in positions
  - \* The underlying surface breaks



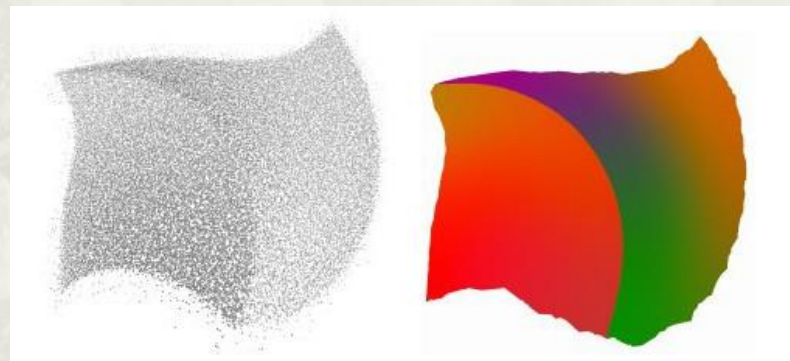
Normal discontinuous features



Position discontinuous features

# Related Work

- \* Feature preserving bilateral filtering [Mederos2003]  
[Duguet2004] [Sun2007] [Nociar2010] [Huang2012]
  - \* Cannot handle position discontinuous features



From [Huang2012]

- \* Explicit feature smoothing approaches [Pauly2003]  
[Daniels2007] [Zhou2008]
  - \* Replace points as lines/curves

# Approach Overview

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## Input

- \* An unoriented, piece-wise smooth aerial LiDAR point cloud of a single building

## Output

- \* A new set of oriented points
- \* Providing
  - \* Smoothed noise
  - \* Filled gaps and holes
  - \* Enhanced normal discontinuous features
  - \* Enhanced position discontinuous features



# Approach Overview (Cont.)

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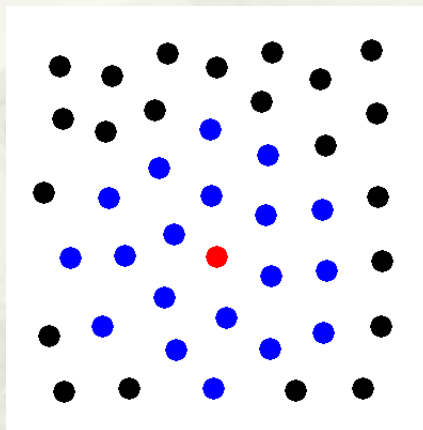
## Approach

- \* Explicitly extract and regularize position discontinuous features
- \* Two-step framework
  - \* Smoothing: handles noise
  - \* Up-sampling: handles under-sampling

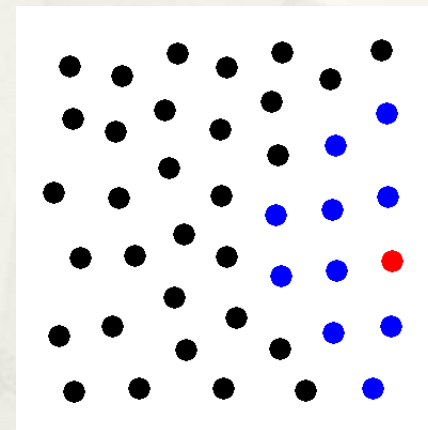


# Position Discontinuous Feature Refinement



- \* Detection: local environment analysis



Interior



Boundary

 Current point  
 Neighbors

- \* Boundary direction (B. direction) estimation
  - \* The direction of the underlying line formed by boundary points only
  - \* PCA over neighbors

# Step 1: Smoothing

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- \* Bilateral filtering

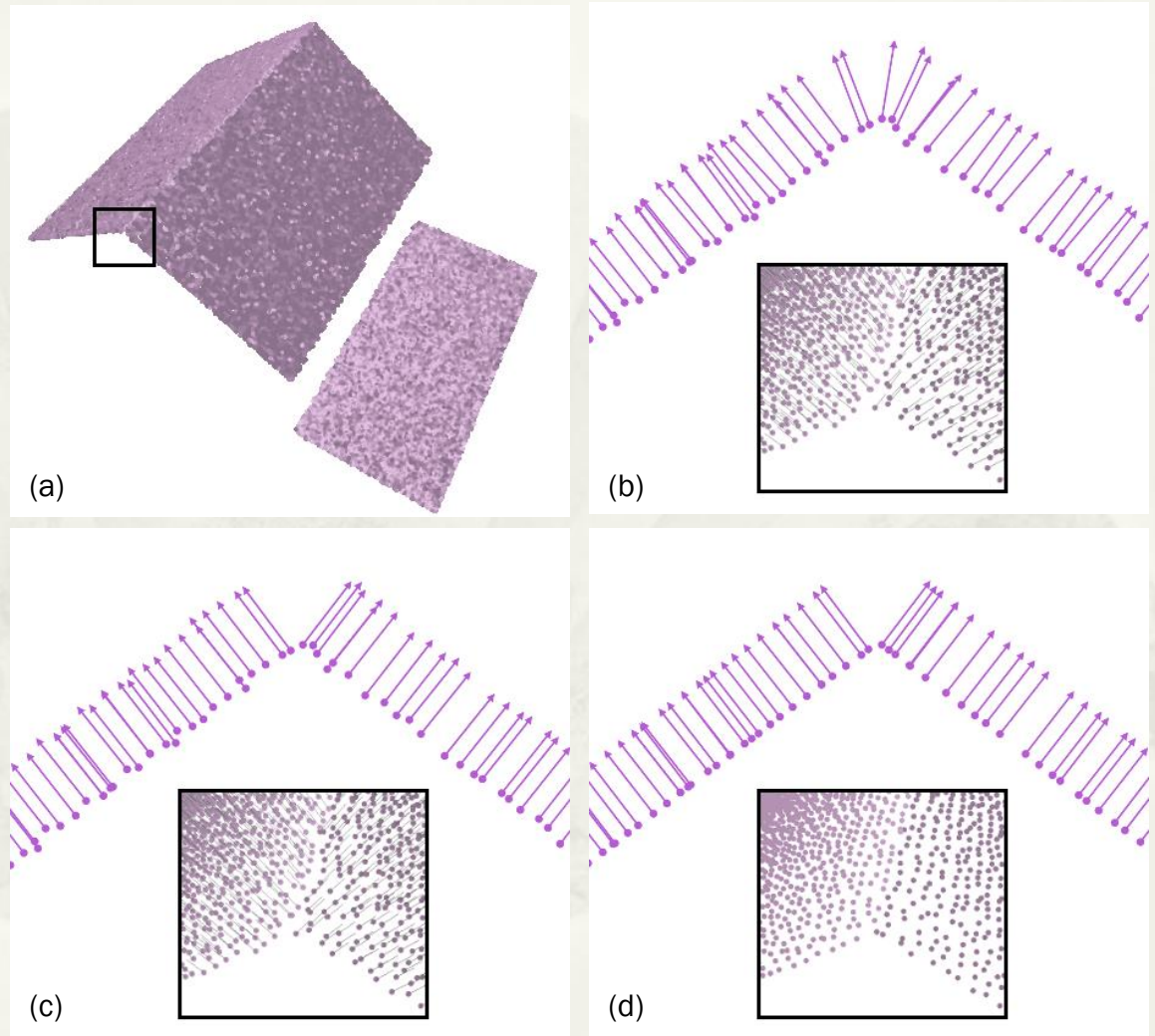
- \* The estimate for a point is the weighted average of prediction from neighboring points
- \* The weight of a neighbor depends on
  - \* Spatial distance (spatial weight)
  - \* Normal/B. direction difference (influence weight)

- \* Two stage:

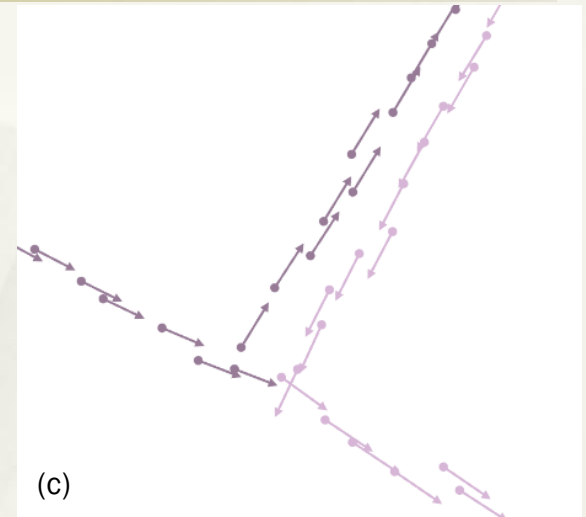
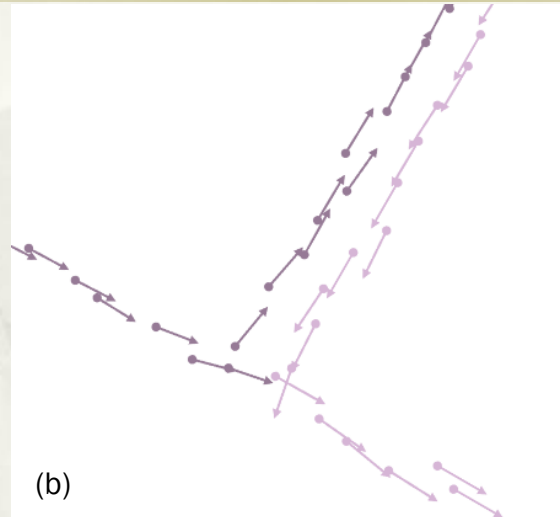
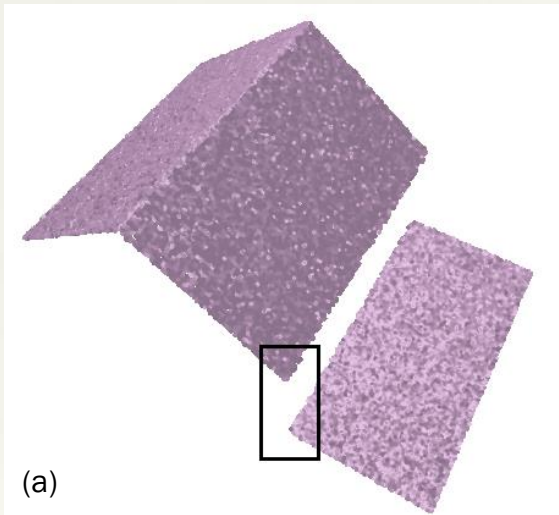
- \* Smooth normals/B. directions
- \* Smooth positions under the guidance of the filtered normals/B. directions

# Smoothing for All Points

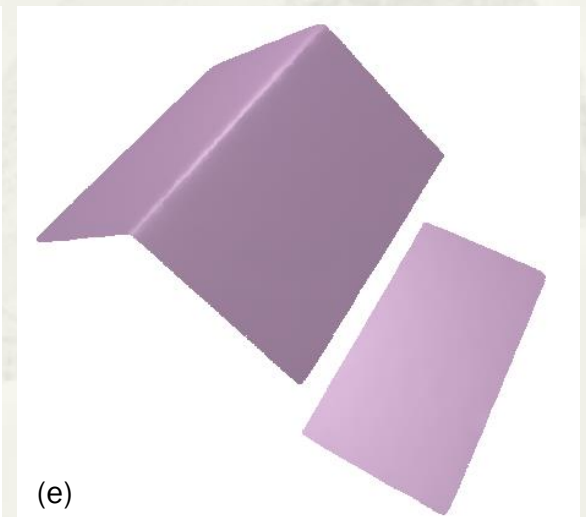
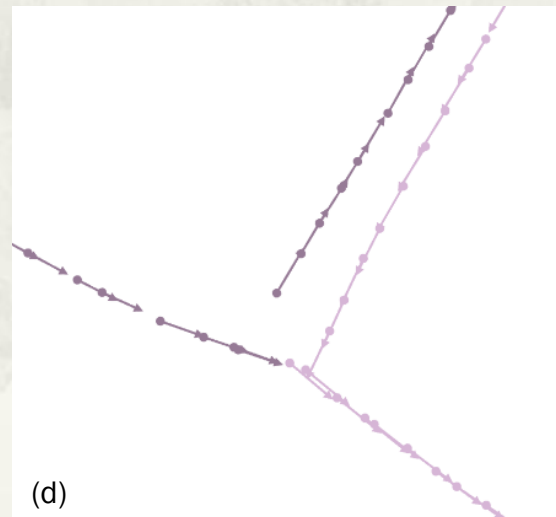
- (a) Input noisy points
- (b) Normal estimation
- (c) Normal smoothing
- (d) Position smoothing



# Smoothing for Boundary Points

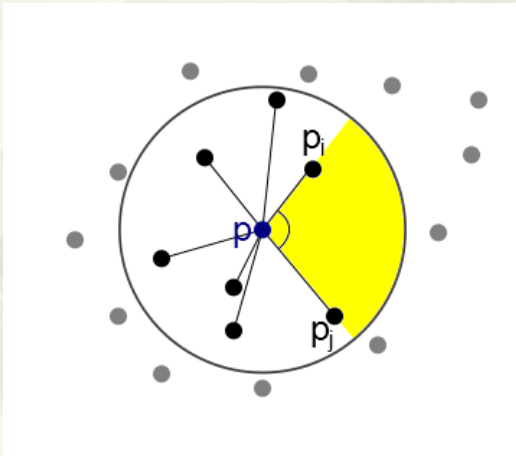


- (a) Input noisy points
- (b) B. direction estimation
- (c) B. direction smoothing
- (d) B. position smoothing
- (e) Output points

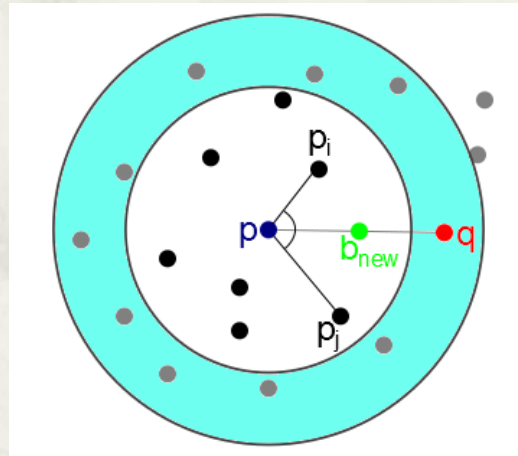


# Step 2: Up-sampling

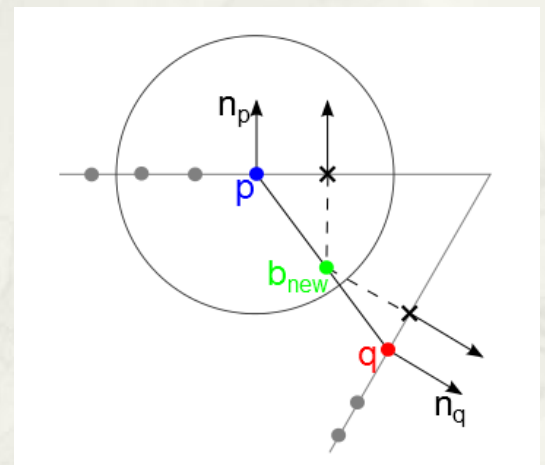
- \* Local gap detector: local environment analysis
- \* Bilateral projection operator: preserve features



(a) Locate gap



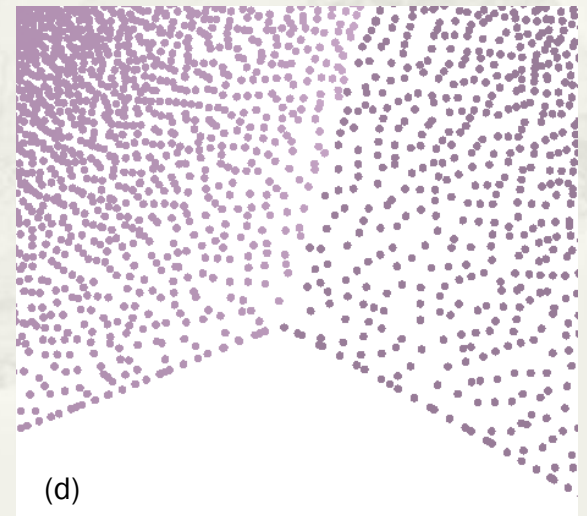
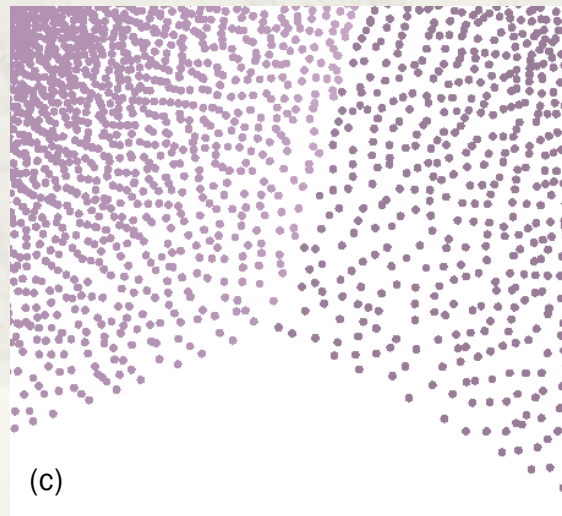
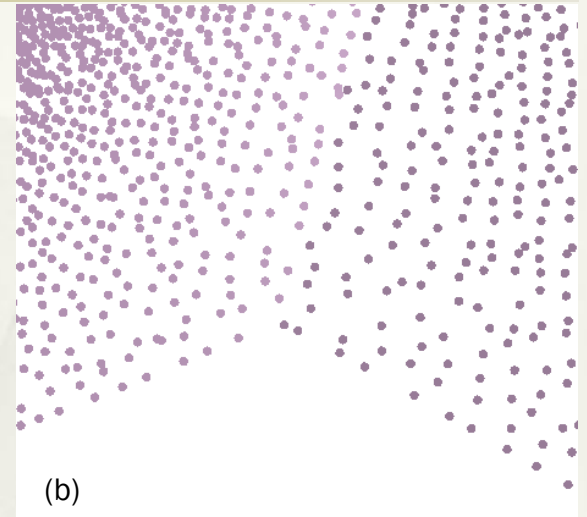
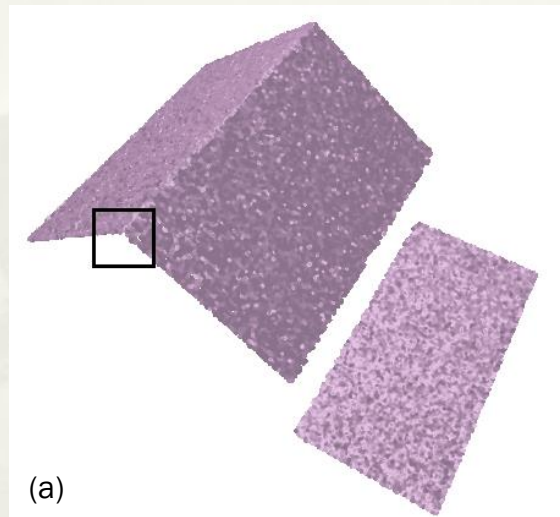
(b) Find end point



(c) Interpolate new point

# Up-sampling

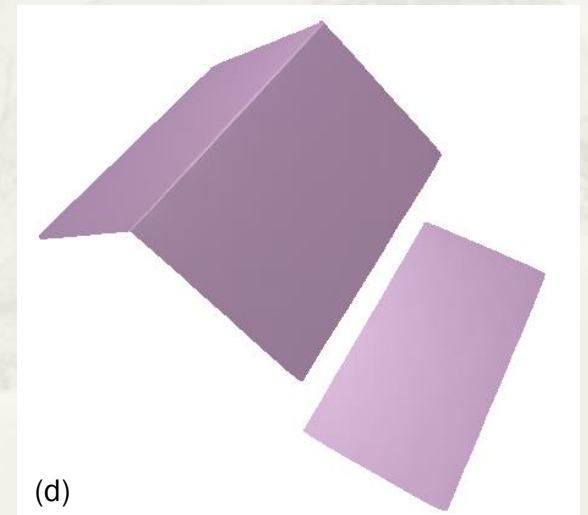
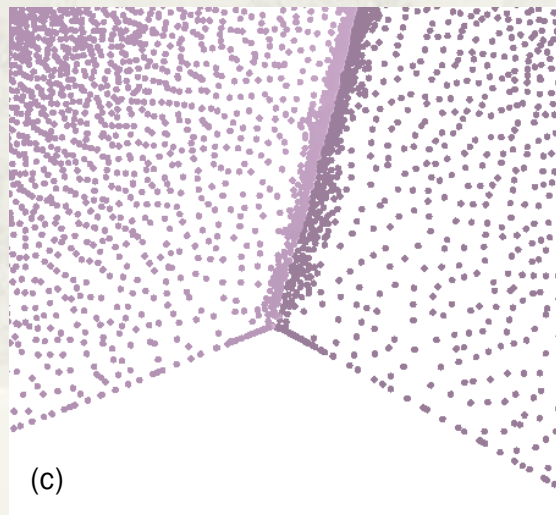
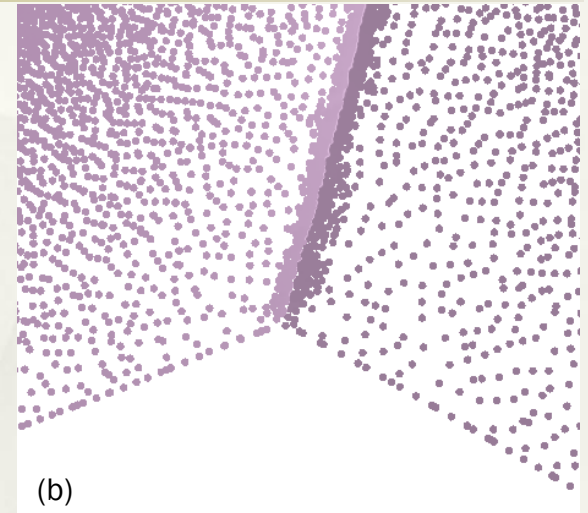
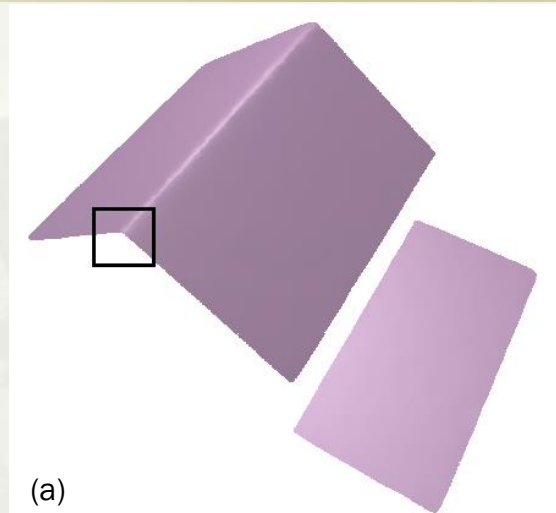
- (a) Input noisy points
- (b) Smoothed points
- (c) Interior up-sampling
- (d) Boundary up-sampling





# Feature Enhancing Up-sampling

Features can be enhanced if up-sampling is restrained to regions near features



- (a) Smoothed points
- (b) Interior enhancing
- (c) Boundary enhancing
- (d) Output points



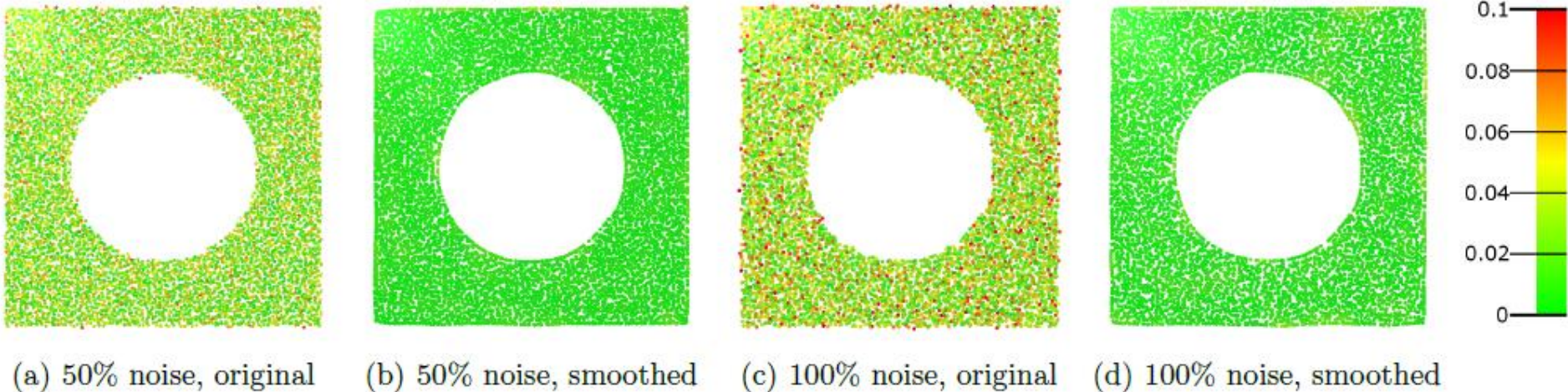
# Evaluation Outline

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- \* Robustness
- \* Stability
- \* Versatility
- \* Extensibility
- \* Comparison with previous work
- \* Performance
- \* Applications

# Robustness to Noise

- \* The smoothing quality does not degrade much as the noise increases



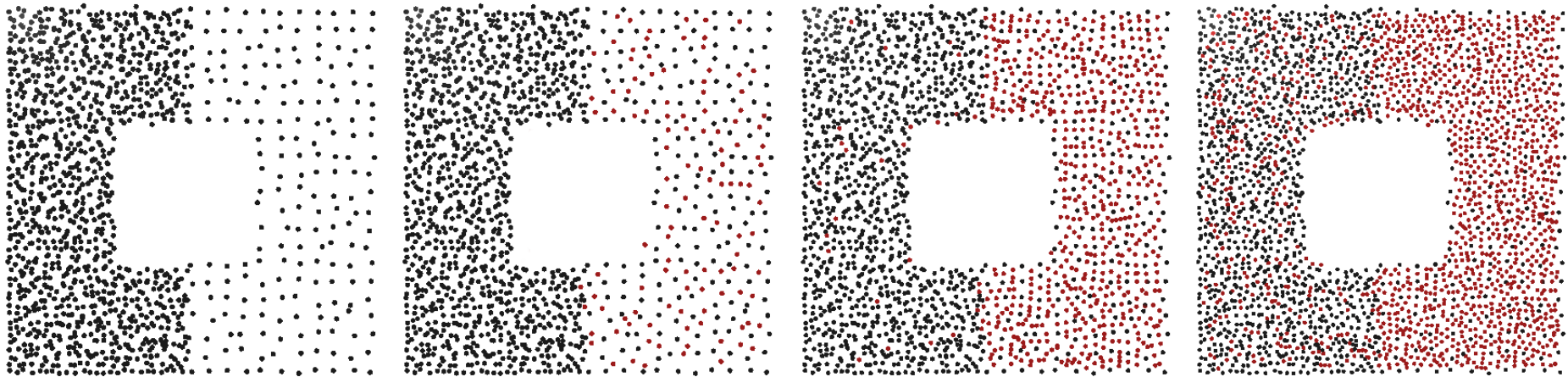
Data set	Interior-MD	Interior-SD	Boundary-MD	Boundary-SD
(a)	0.028482	0.018486	0.051923	0.027182
(b)	0.003169	0.003243	0.025885	0.015879
(d)	0.036123	0.026051	0.060937	0.031821
(e)	0.004458	0.004472	0.030655	0.019828

MD: mean distance  
SD: standard deviation

# Robustness to Non-homogeneous Point Density

- \* Achieve global uniform sampling with a single parameter
- \* Both inner and outer boundaries are well preserved

Original point  
New point



(a) Original

(b)  $\gamma=2.0$

(c)  $\gamma=1.3$

(d)  $\gamma=1.0$

# Stability

- \* Most normal and position discontinuous features are sharply recovered for data sets with vastly different densities

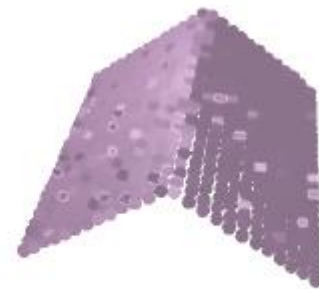
Noisy input



(a) 15,625 samples



(b) 3,136 samples



(c) 625 samples

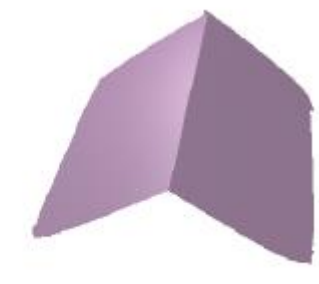
Refined output



(d) 42,479 samples



(e) 42,319 samples

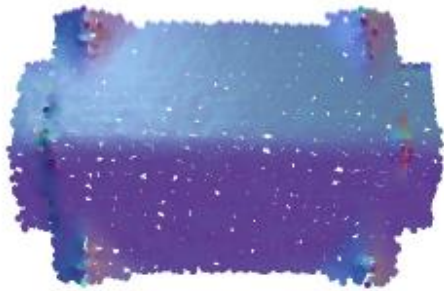


(f) 45,469 samples

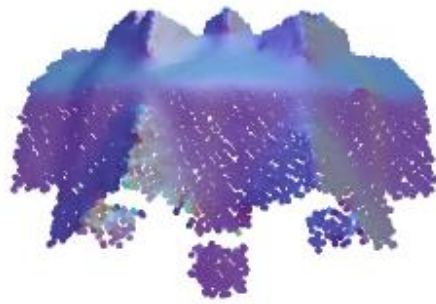


# Versatility

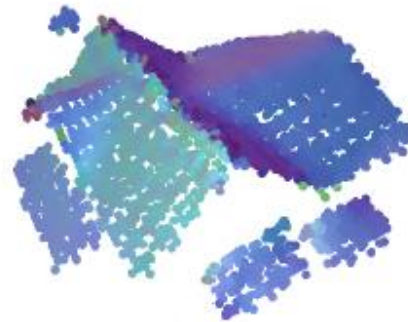
- \* Effective to buildings with various roof shapes and complexity



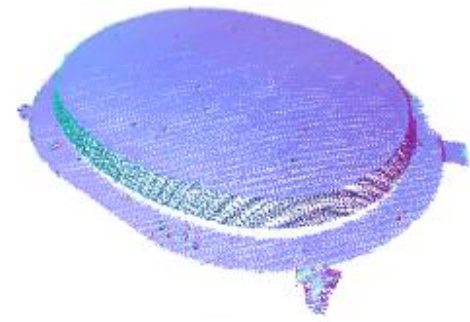
(a) Original



(b) Original



(c) Original



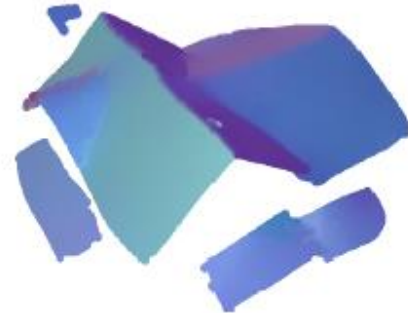
(d) Original



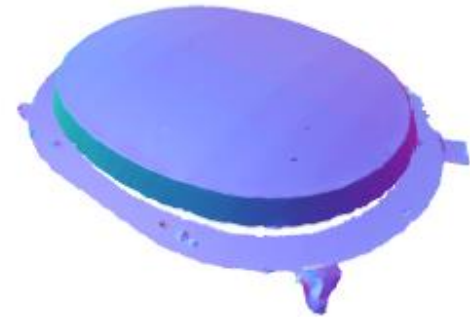
(e) Refined



(f) Refined



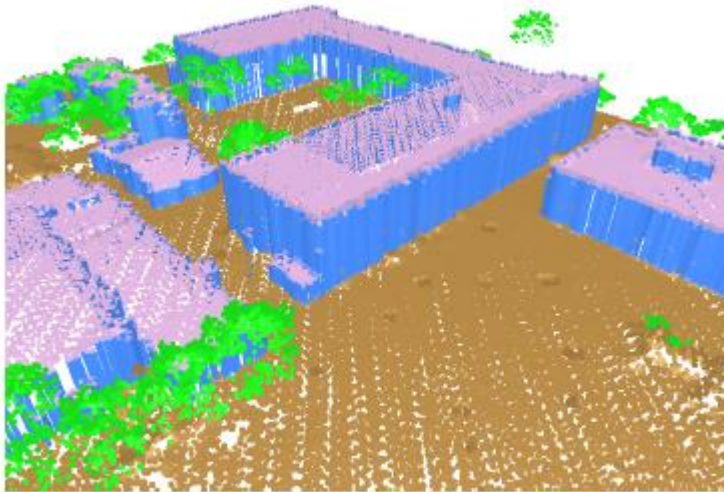
(g) Refined



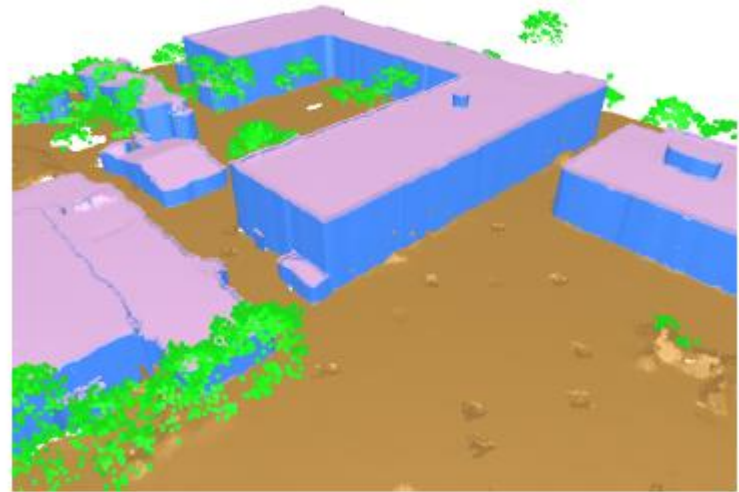
(h) Refined

# Extensibility

- \* Extensible to objects other than buildings
- \* Extensible to large-scale data



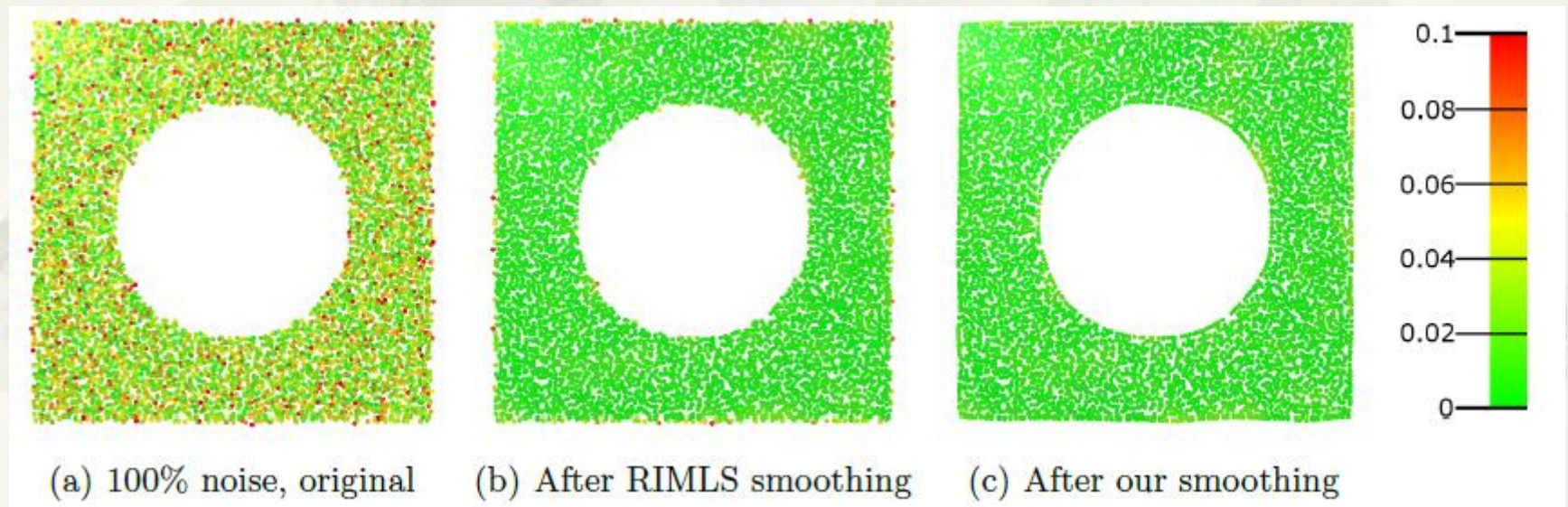
(a) Original



(b) Refined

# Comparison with Point-Based Refinements

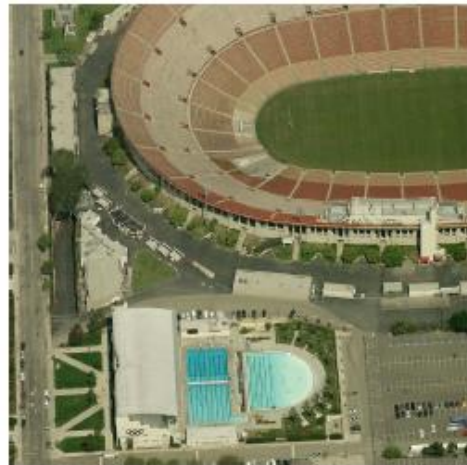
- \* Comparable quality on non-boundary points
- \* Obviously less errors on boundaries



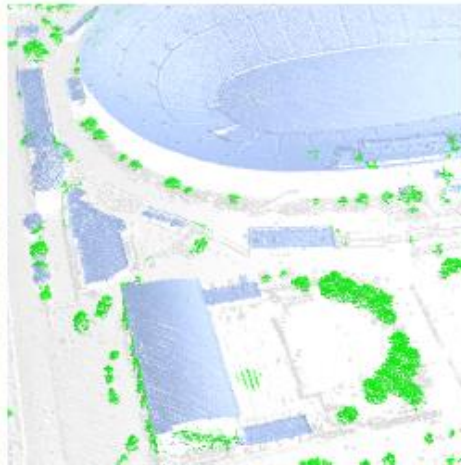


# Comparison with Geometry Modeling

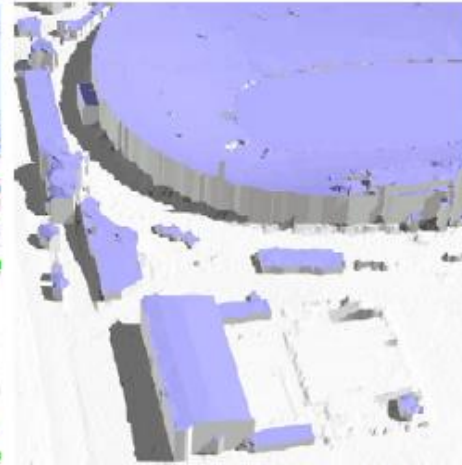
- \* Comparable quality on straight boundaries
- \* Better quality on curved boundaries



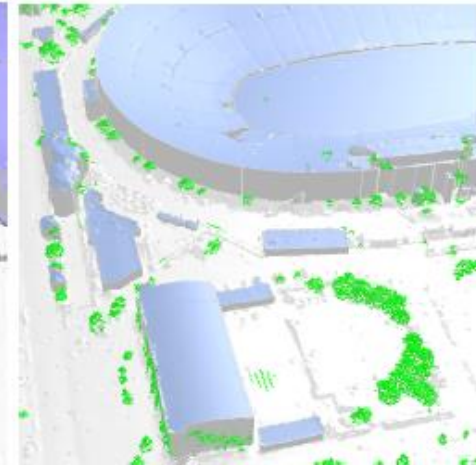
(a) Ground truth image



(b) Original points



(c) Geometry models



(d) Refined points

# Performance

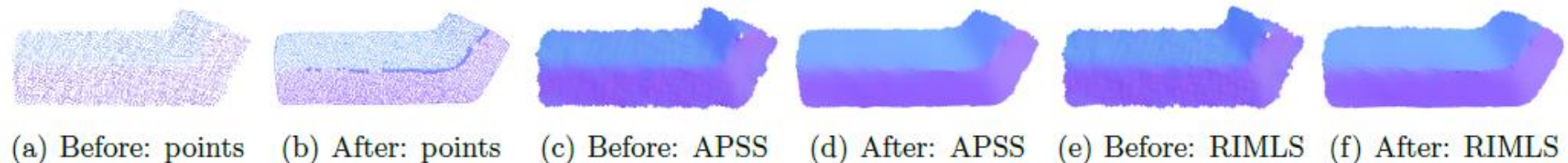
- \* Smoothing and up-sampling are fast

Data set	IP-N	SM-T	US-T	US-N	EU-T	OP-N
Figure 1 (b)	5,594	0.57	1.37	11,597	2.36	16,590
Figure 6 (f)	23,500	2.07	5.84	52,303	2.57	57,461
Figure 11 (e)	9,079	1.23	1.68	15,603	5.80	25,720
Figure 11 (f)	7,546	1.12	1.18	12,146	4.11	21,409
Figure 11 (g)	4,658	0.49	0.99	9,248	3.45	16,532
Figure 11 (h)	119,310	11.06	40.44	298,164	105.36	450,279
Figure 15 (b)	4,498	0.76	0.54	6,403	0.70	7,913

IP-N: number of input points; NE-T: time for normal estimation; SM-T: time for smoothing; US-T: time for up-sampling; UP-N: number of points after up-sampling; EU-T: time for feature enhancement up-sampling; OP-N: number of output points. Time is measured in seconds.

# Applications

- \* Our approach provides
  - \* smoother planes
  - \* no gaps
  - \* sharpened both normal and position discontinuous features
- \* It is a useful pre-process to improve visual quality of
  - \* direct point rendering
  - \* surface reconstruction (e.g. APSS and RIMLS)



# Conclusion

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- \* Explicitly extract and regularize position discontinuous features
- \* The smoothing step filters noise while preserves features
- \* The up-sampling step fills gaps while enhance features
- \* Experiments show improved rendering and surface reconstruction results

# Future Work

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- \* To improve performance: parallelize each refinement operation
- \* To improve quality:
  - \* incorporate other information such as aerial images
  - \* add more restricted geometric assumptions as in urban modeling
- \* Incorporate refinement into real-time city-scale point cloud visualization

# Acknowledgements

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We thank the reviewers for their useful comments. We thank Airborne 1 Corp. for providing real LiDAR data. We appreciate the programming support from Miao Wang, useful comments, reviews and support from Dr. Luciano Nocera and Dr. Saty Raghavachary.