



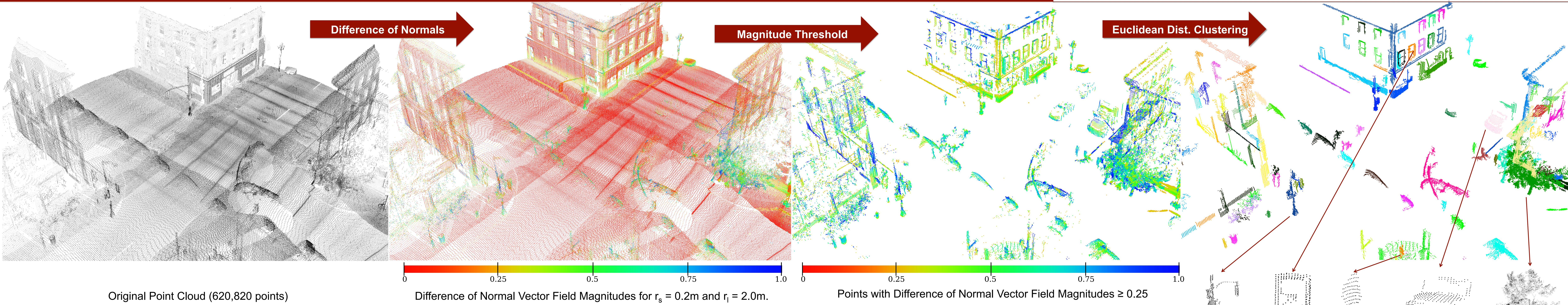
The Difference of Normals as a Scale-based Operator in Point Clouds

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A PCL implementation of the DoN operator is available along with examples at <http://yanii.github.com/DoNPCL> - scan the QR code with your smartphone to go there now!



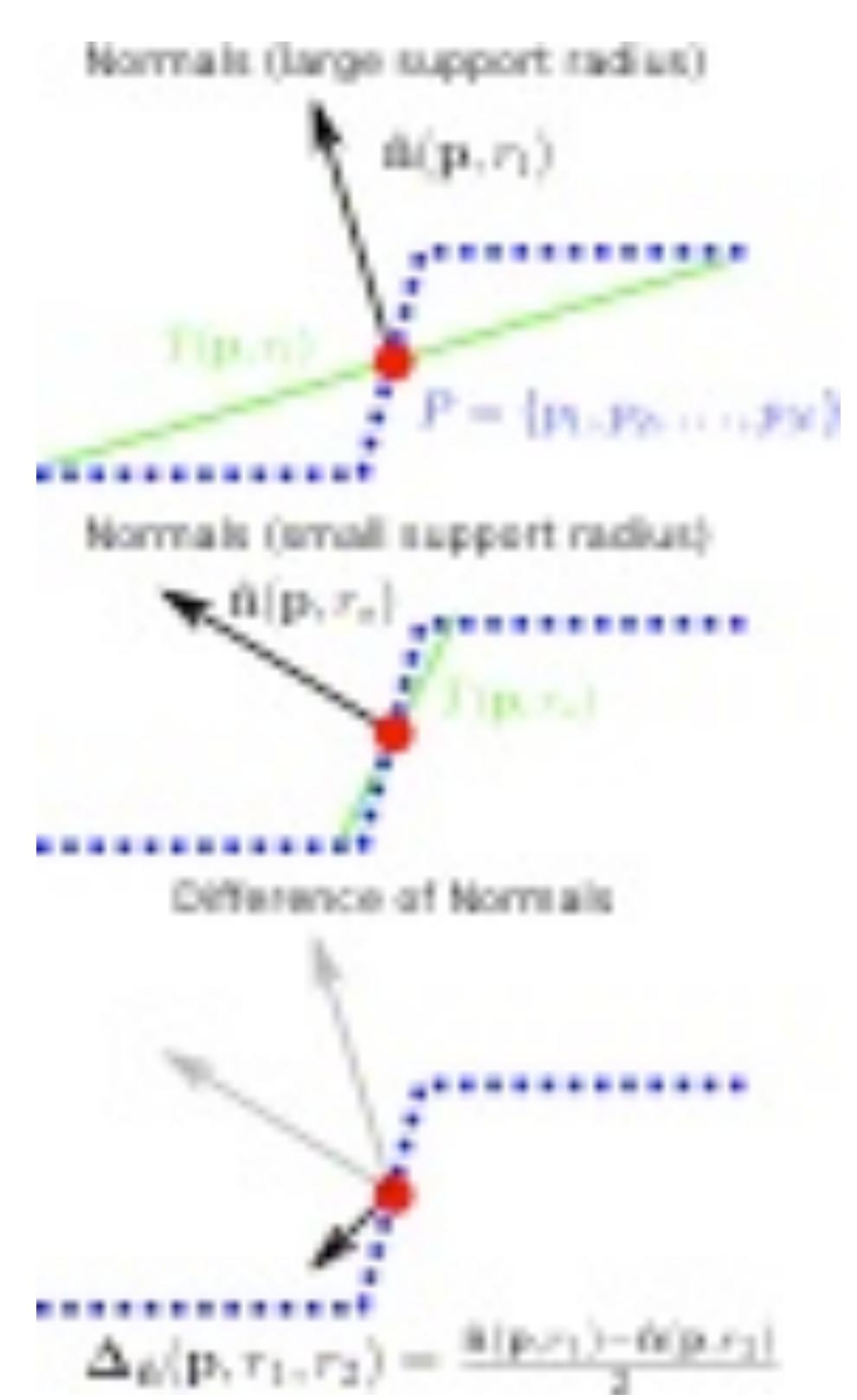
Original Point Cloud (620,820 points)

Method

- Difference of Normals (DoN) is a scale-based *surface processing* operator for unorganized point clouds.
- Conceptually similar to the Difference of Gaussians in 2D image processing, but operating on the implicit surface of a point cloud.

Motivation

- Availability of large (millions of points), composite unorganized (non-regularly sampled) LIDAR data of urban street neighborhoods.
- Common use of such data is to create Geographic Information System (GIS) models.
- Street furniture* in particular is of interest, such as fire hydrants, traffic lights, etc.
- Other objects of interest to GIS models include buildings (inc. building facades), curbs, roads, trees.
- Current models are laboriously created manually!
- Is automatic modeling of this data possible?
- Key problem is segmentation of the data!



$$\Delta_{\hat{n}}(\mathbf{p}, r_s, r_l) = \frac{\hat{n}(\mathbf{p}, r_s) - \hat{n}(\mathbf{p}, r_l)}{2}$$

where \mathbf{p} is the point, r_s is the small support radius, and r_l is the large support radius.

- Result is normalized difference of unit vectors, thus DoN magnitude always in range [0, 1.0].

- The resulting vector field may be thresholded by magnitude to find points that have the strong response at a given scale.
- The two parameters r_s, r_l provide parameters to the scale and bandwidth of the filter, similar to the two standard deviations σ_s, σ_l of the Difference of Gaussian.
- Empirically it was found that the best results were found with radii with a ratio of one octave, i.e. $r_l = 10 \times r_s$.

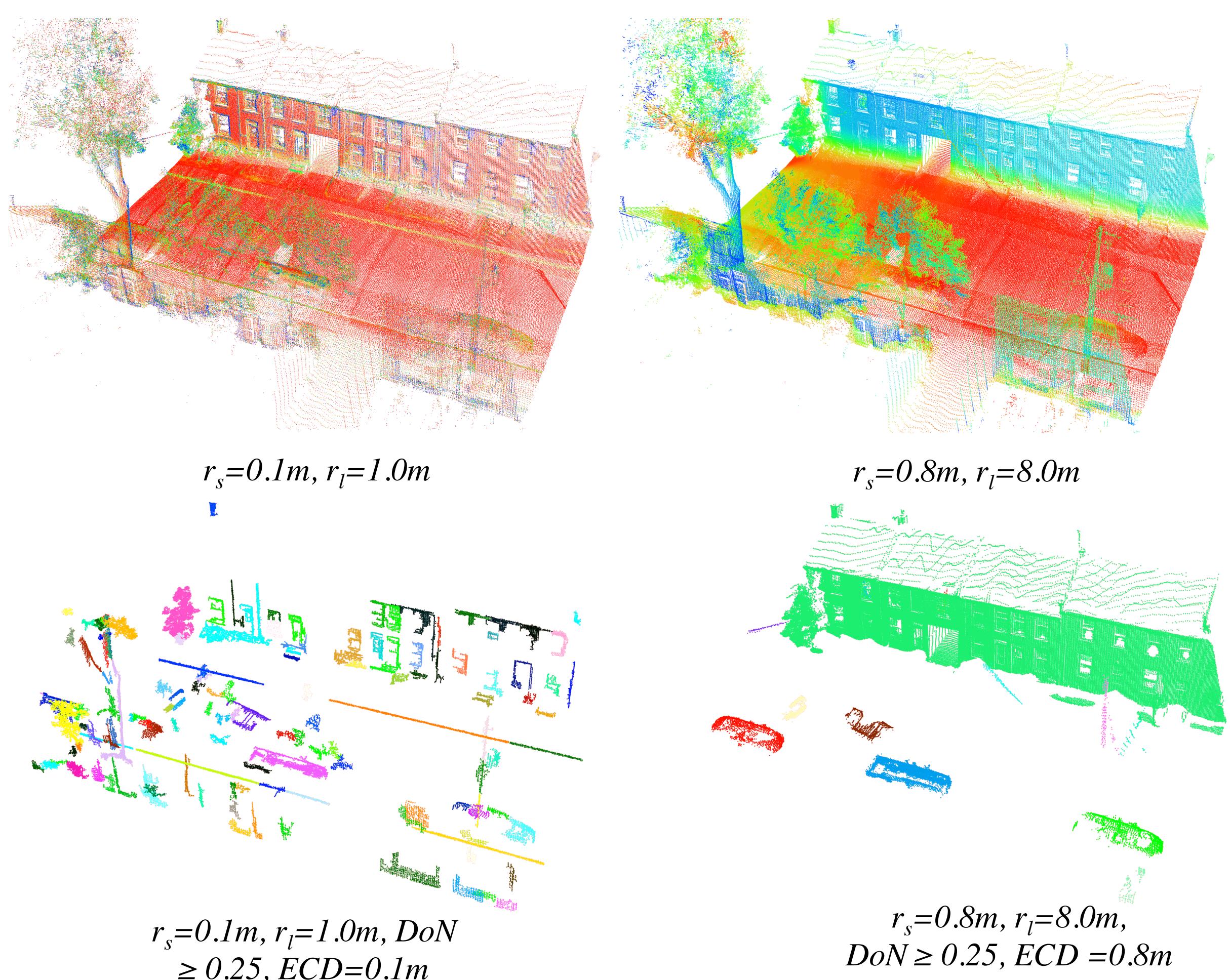


Illustration of DoN Thresholding/Clustering with Different Scales

Results

- Together with a clustering algorithm, empirically found to provide a good scale-based segmentation of unorganized point clouds with highly variable sparsity and sampling.
- Quantitatively shown to be a consistent saliency operator across point clouds scanned from the same underlying surface with different scanners and sampling¹.

Future Work

- Thorough quantitative evaluation of clustering/segmentation results, however without ground truth this is a significant hurdle - suggestions welcome!
- Automatically determine parameters to isolate a known object.
- Use connected components to find clusters instead of far simpler Euclidean distance clustering.
- Perform object recognition on resulting clusters.

References

- Shape Matching of Repeatable Interest Segments in 3D Point Clouds, Int. Workshop on Point Cloud Processing, 2012, Joseph Lam, et al.
For more references see: *Automatic Urban Modelling using Mobile Urban LIDAR Data*, M.Sc. Thesis, Queen's University, Yani Ioannou.
<http://qspace.library.queensu.ca/handle/1974/5443> - use QR code at right.



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