

CroScalar: A Multi-Scale Modeling Framework for Spatio-Temporal Data

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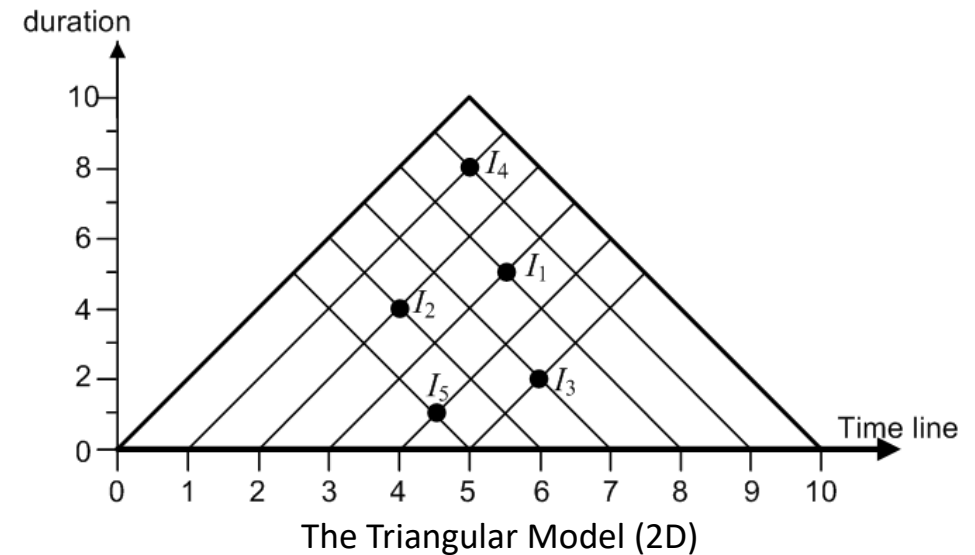
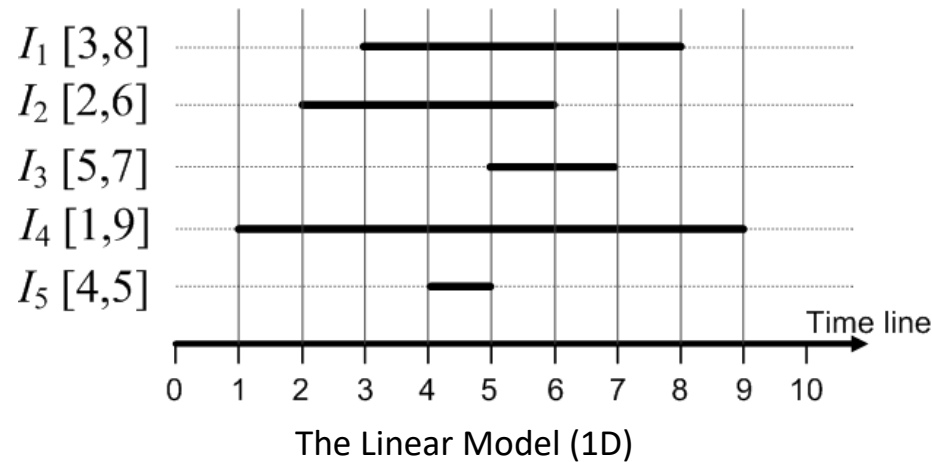
CroScalar: A Multi-Scale Modeling Framework for Spatio-Temporal Data

- Multi-Scale Temporal Analysis (Triangle Model)
- Multi-Scale for Spatial Analysis (Pyramid Model)
- Higher-Dimensional Models for Spatio-Temporal Analyses (CroScalar)



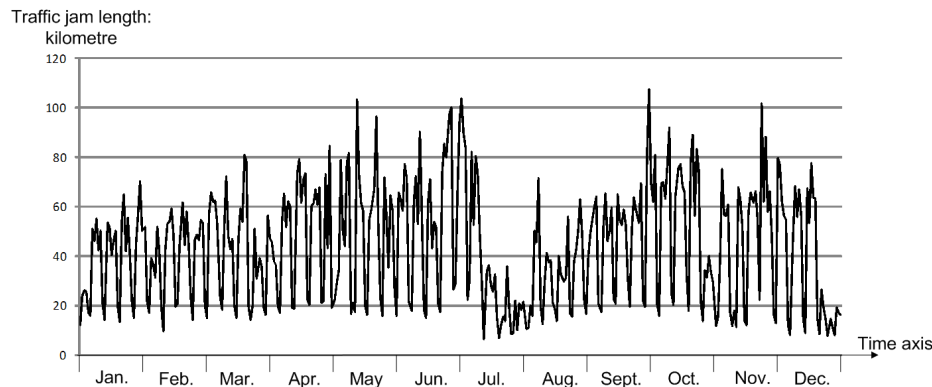
Triangular Model: Mapping Time Intervals in a 2D Space

- Time interval is an extent in time, which is usually represented as linear interval in a 1D linear space
- The linear model is inefficient for data visualization and analysis.
- Alternatively, time intervals can be represented as points in a 2D space

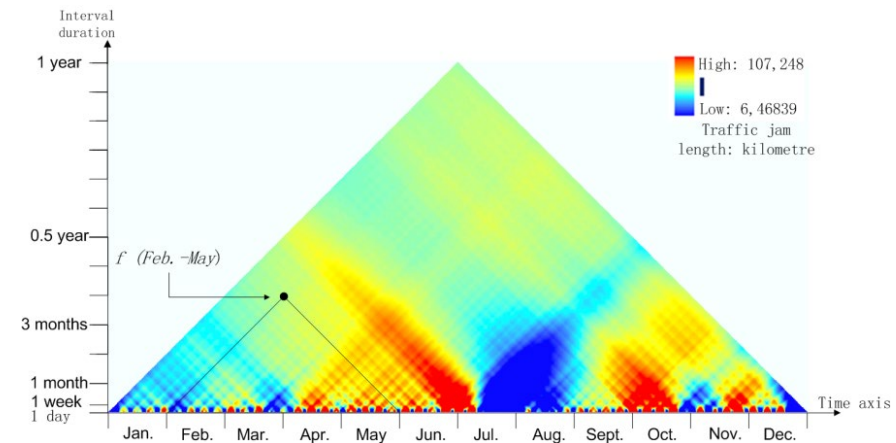


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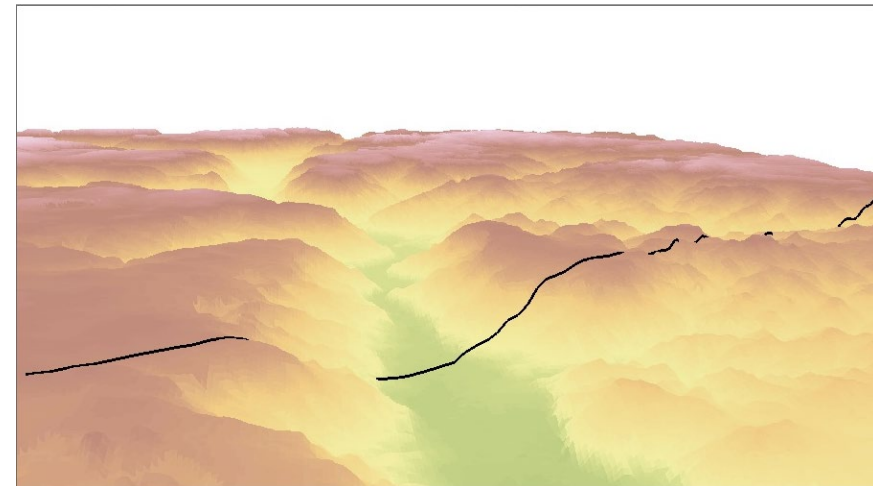


The Linear Model (1D)

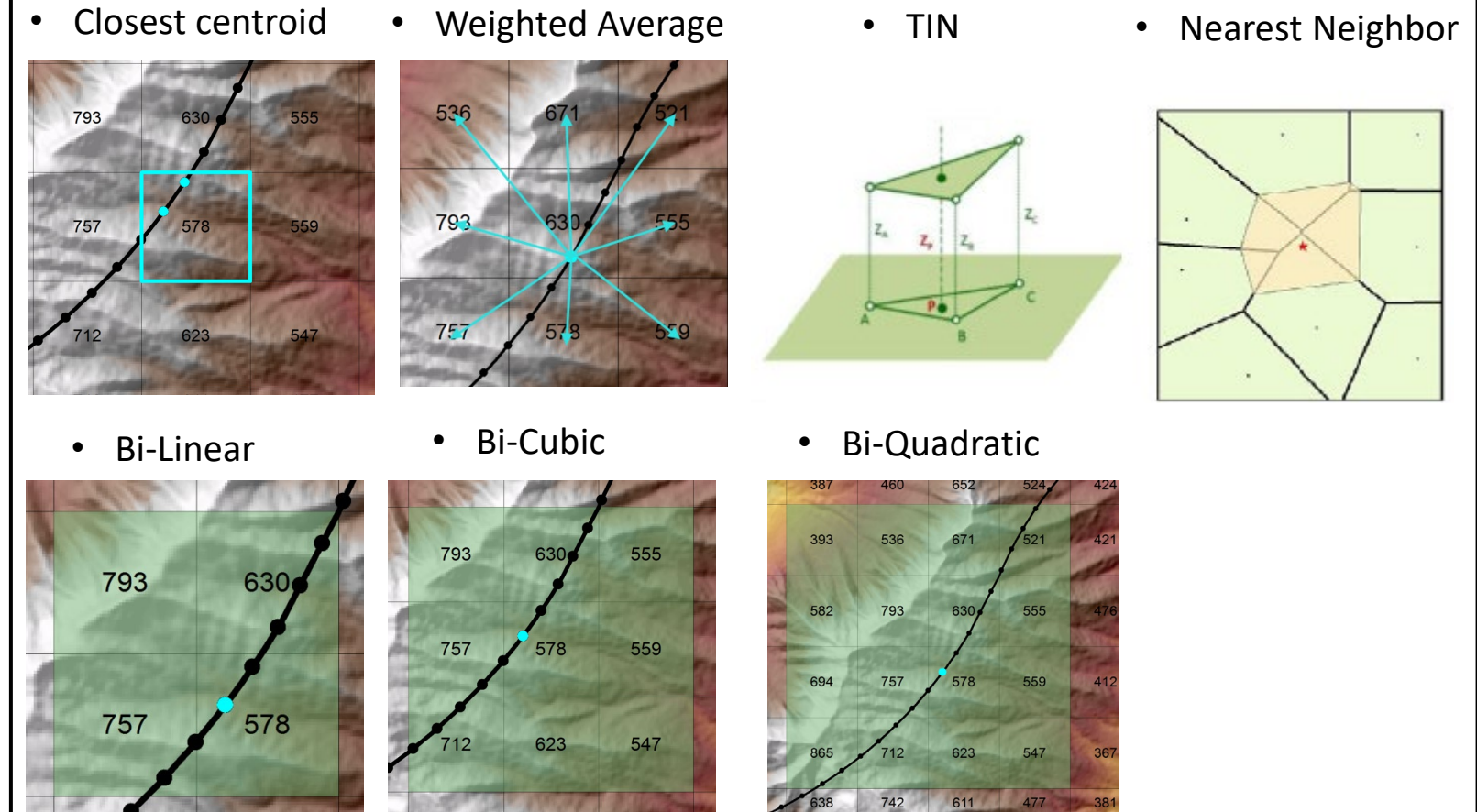


The Triangular Model (2D)

Uncertainty Assessment of Distance Measurement

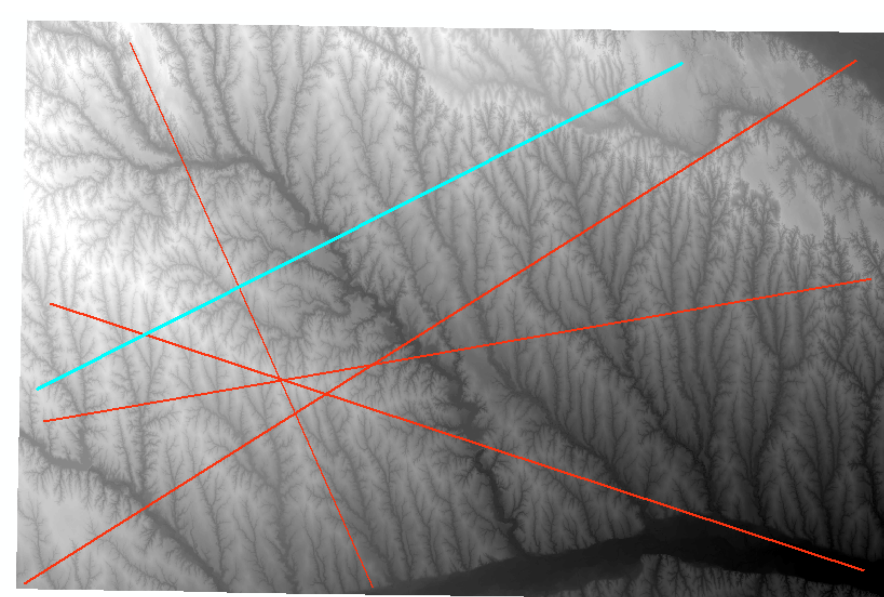


Distance Measurement in 3D terrain

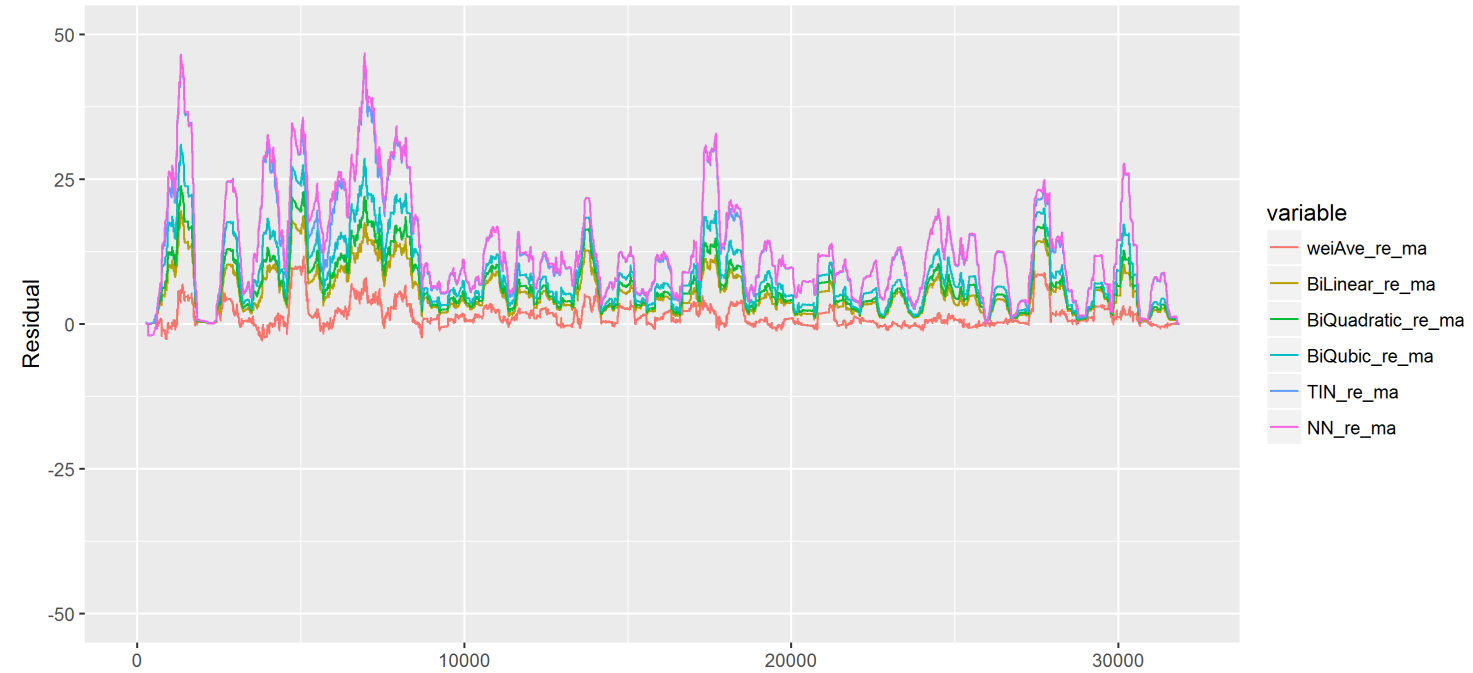


Surface-Adjusted Distance Measurement

Uncertainty Assessment of Distance Measurement

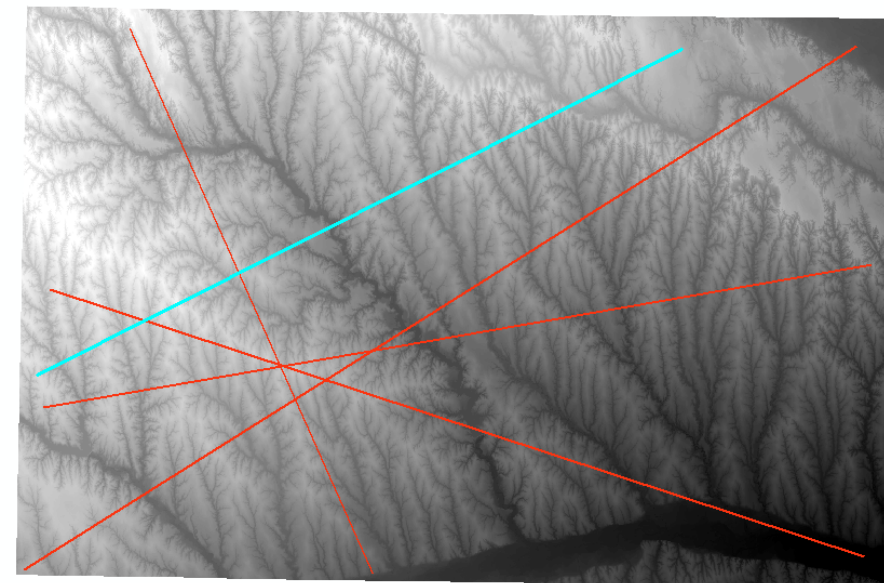


Transects in a study area in Nebraska

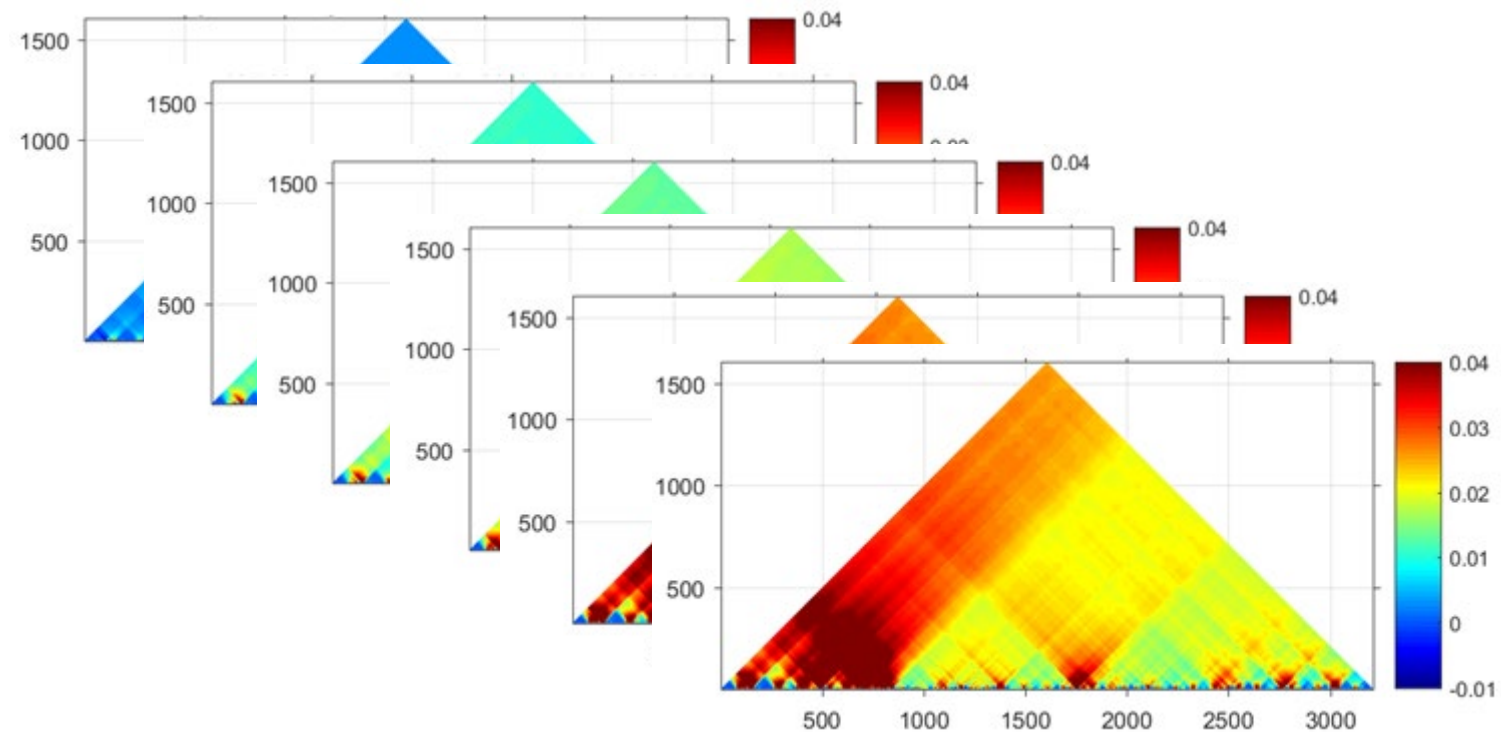


Measurement residuals in 9-meter intervals

Uncertainty Assessment of Distance Measurement

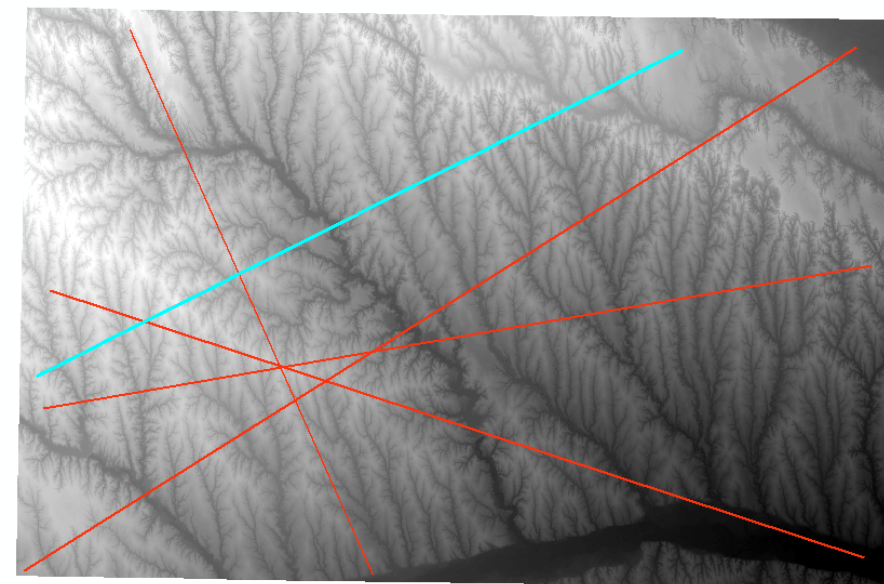


Transects in a study area in Nebraska



Triangle Models of residuals at different intervals

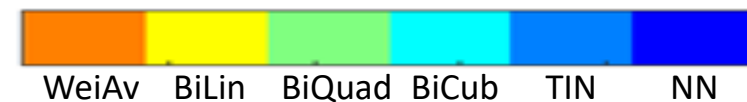
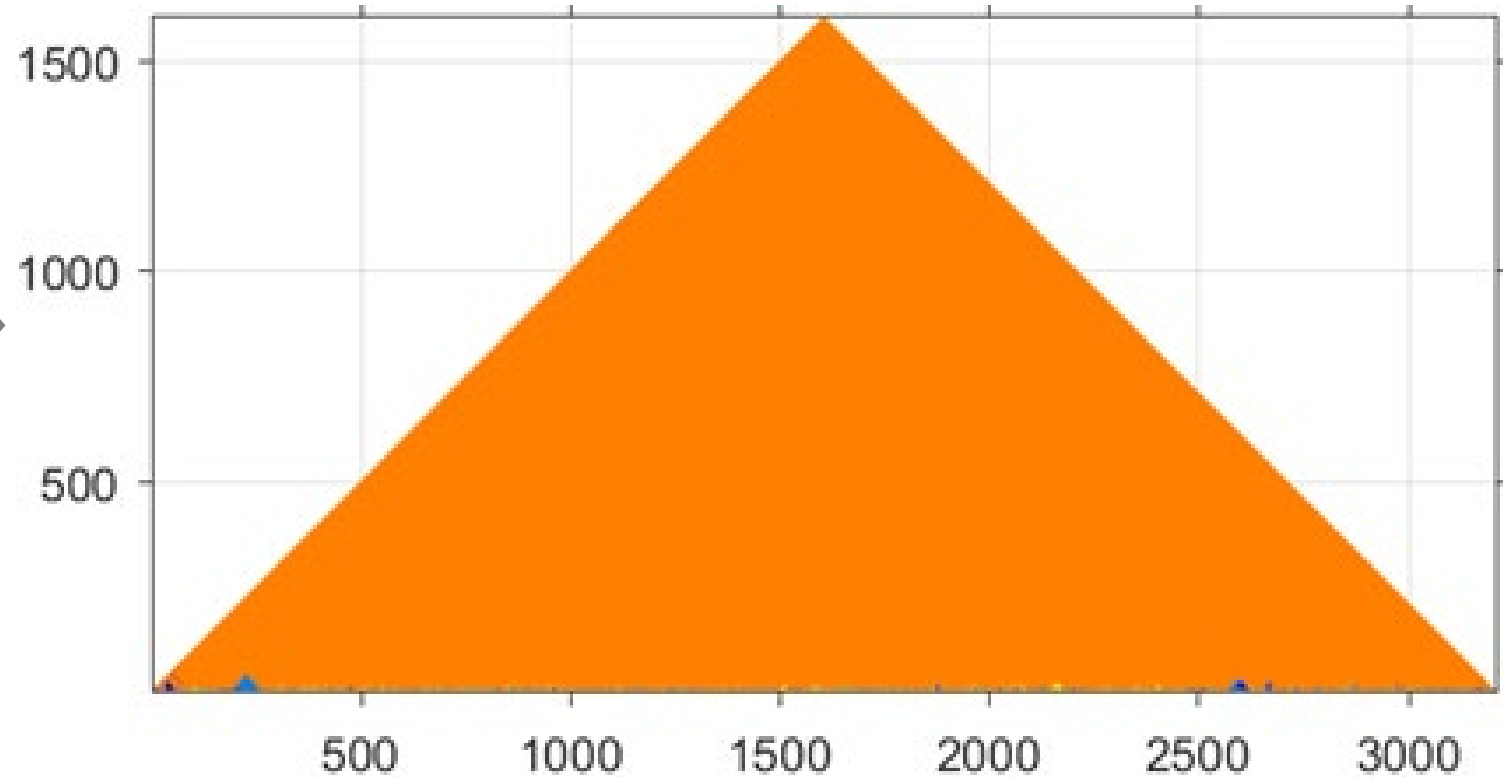
Uncertainty Assessment of Distance Measurement



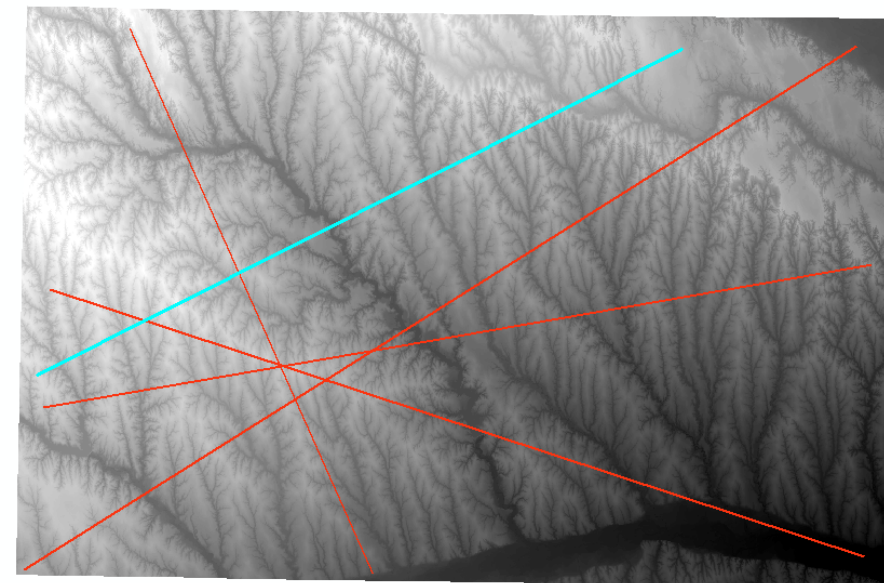
Transects in a study area in Nebraska



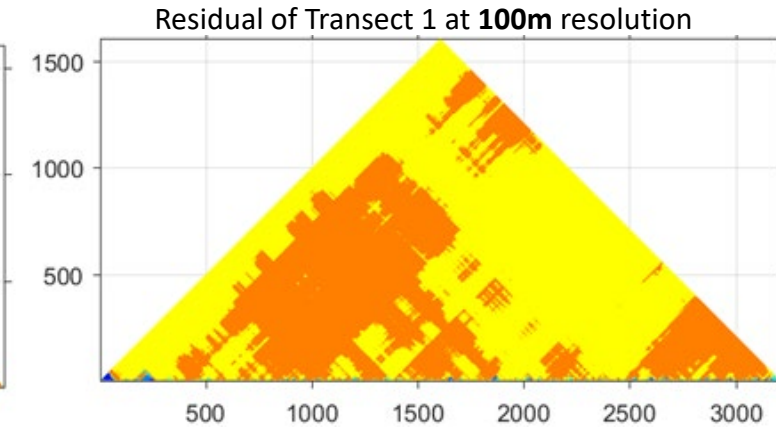
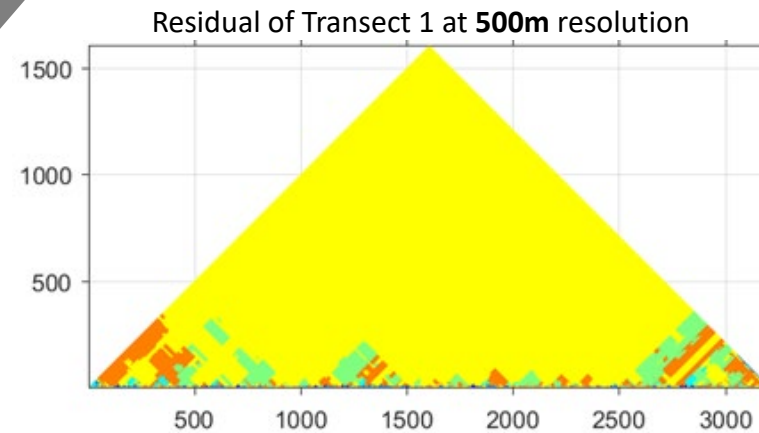
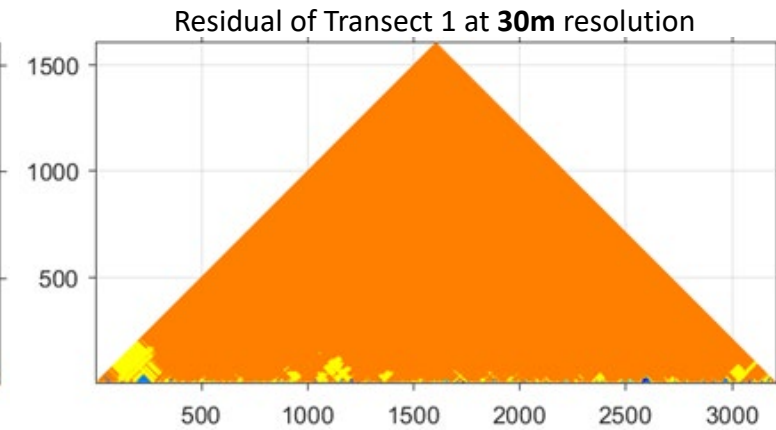
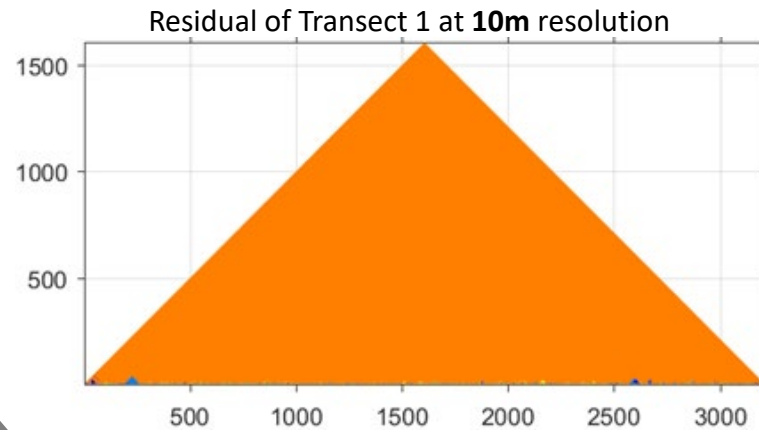
Residual of Transect 1 at 10m resolution DEM



Uncertainty Assessment of Distance Measurement

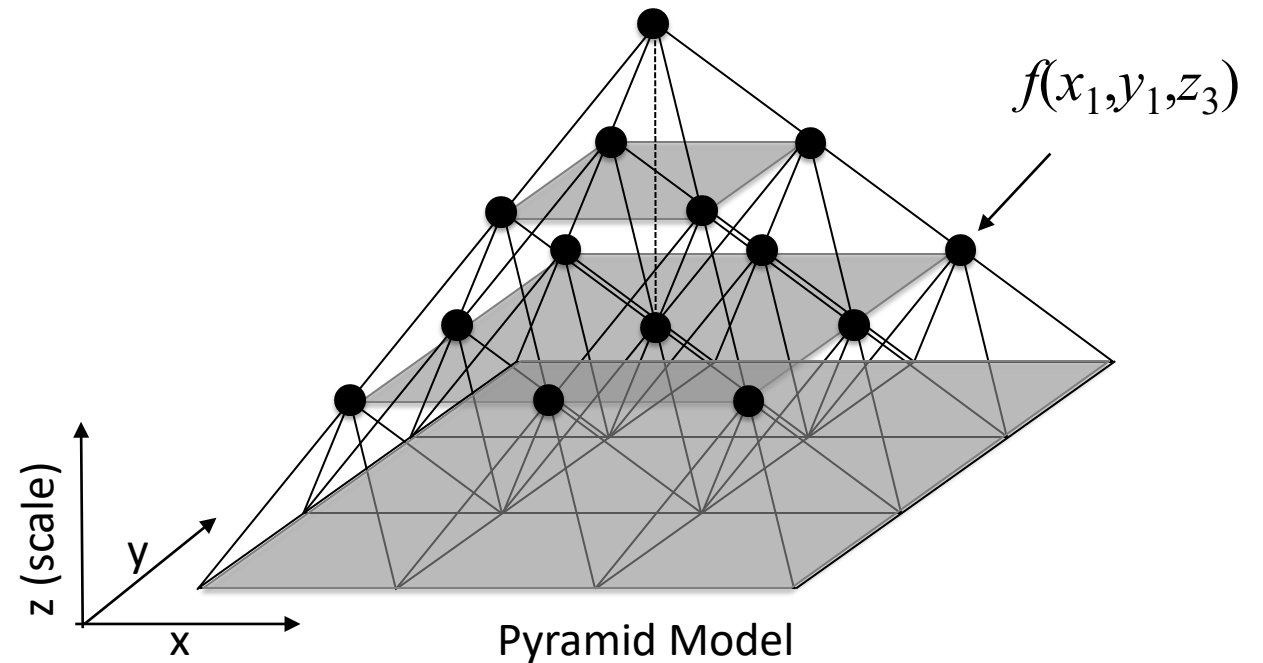
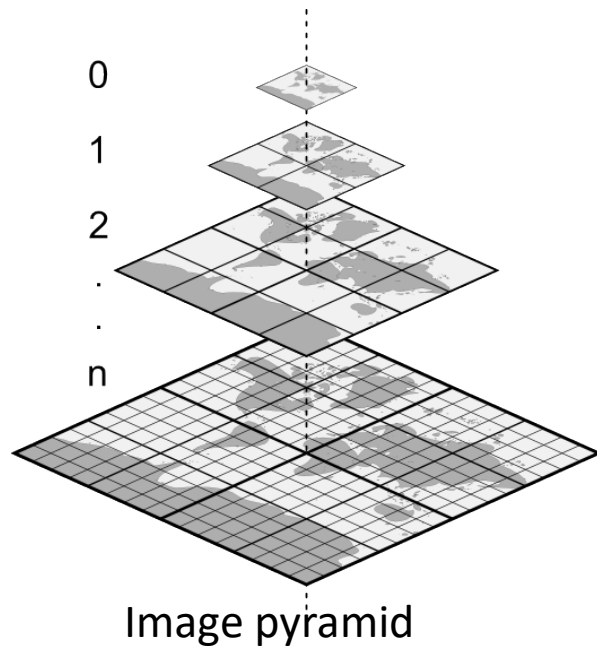


Transects in a study area in Nebraska



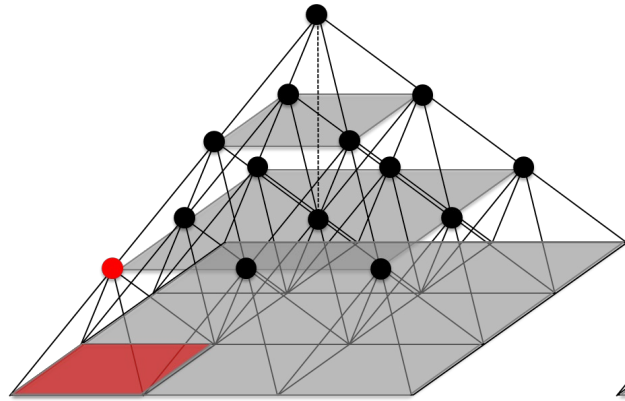
Pyramid Model

- Pyramid Model (PM): **Multi-scale representation** for 2D spatial data
- Similar concept as Image Pyramid
- Integrating the scale dimension (z) with the spatial (x,y) dimension

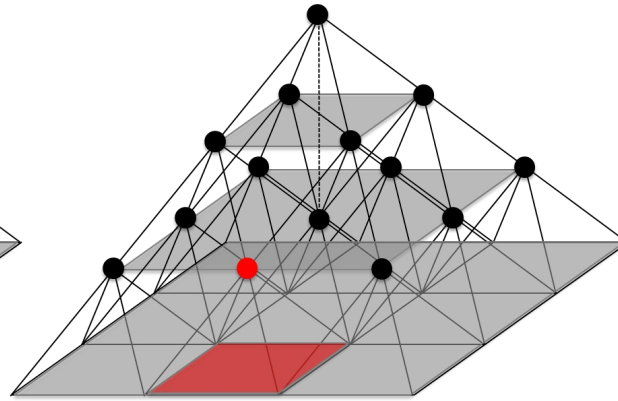


Pyramid Model

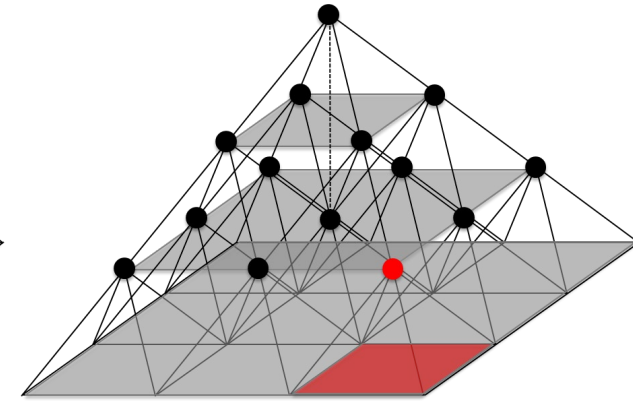
Each node (voxel) represent a specific cell in the tessellation in the base layer



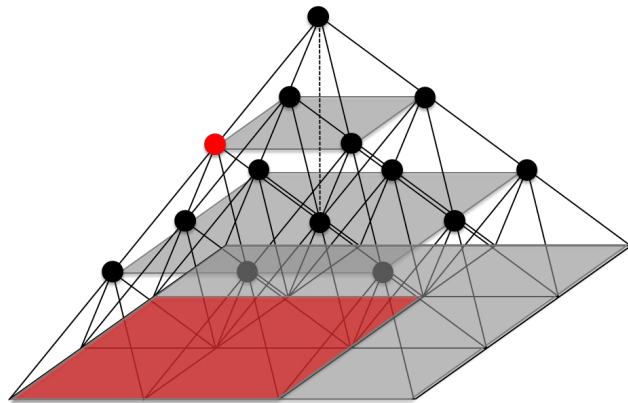
$$f(x_1, y_1, z_1)$$



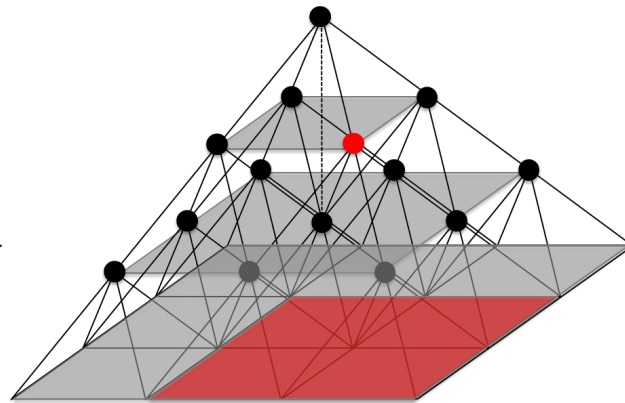
$$f(x_2, y_1, z_1)$$



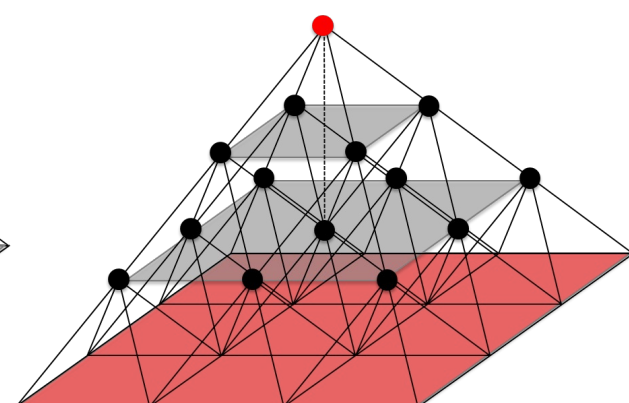
$$f(x_3, y_1, z_1)$$



$$f(x_1, y_1, z_2)$$

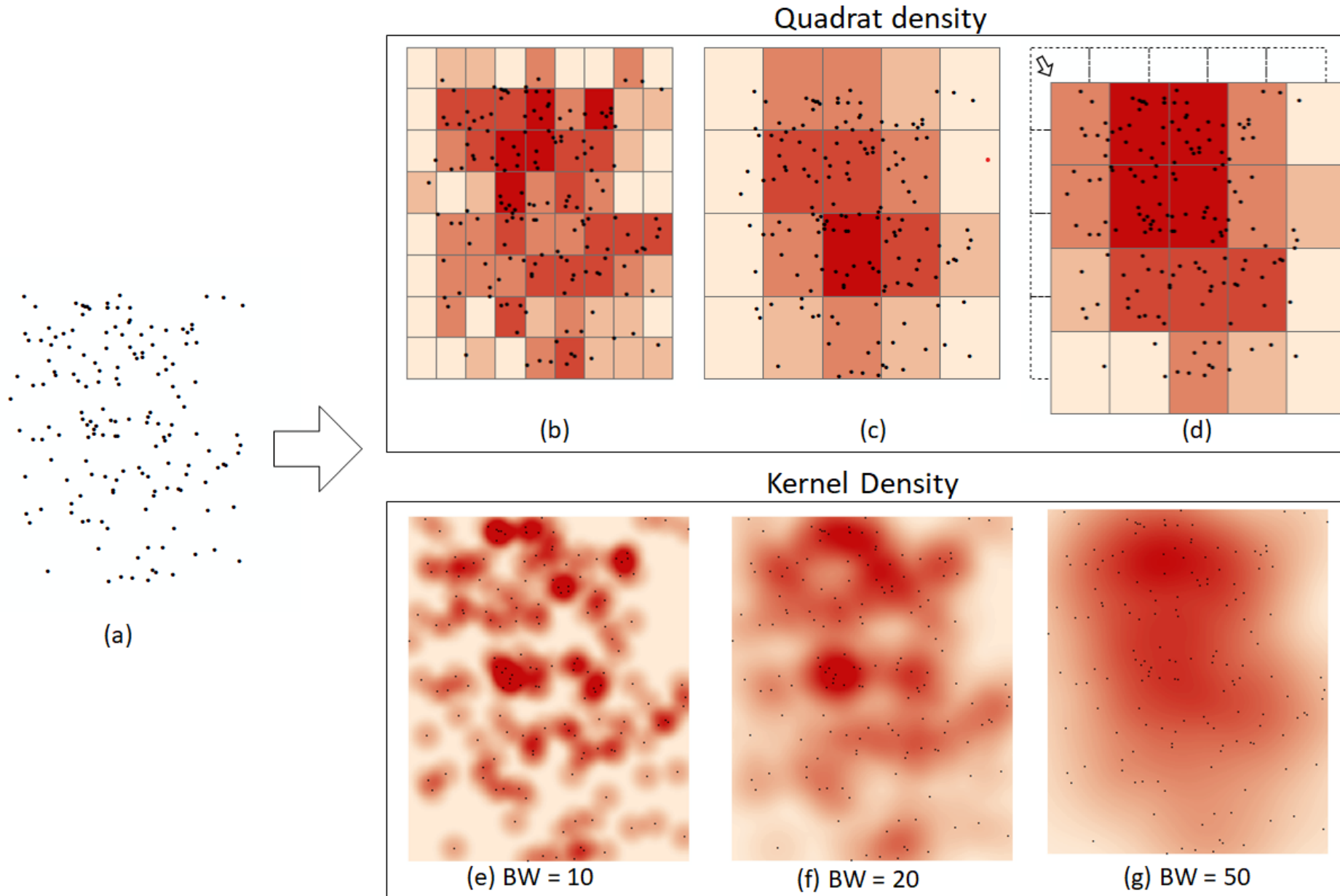


$$f(x_2, y_1, z_2)$$

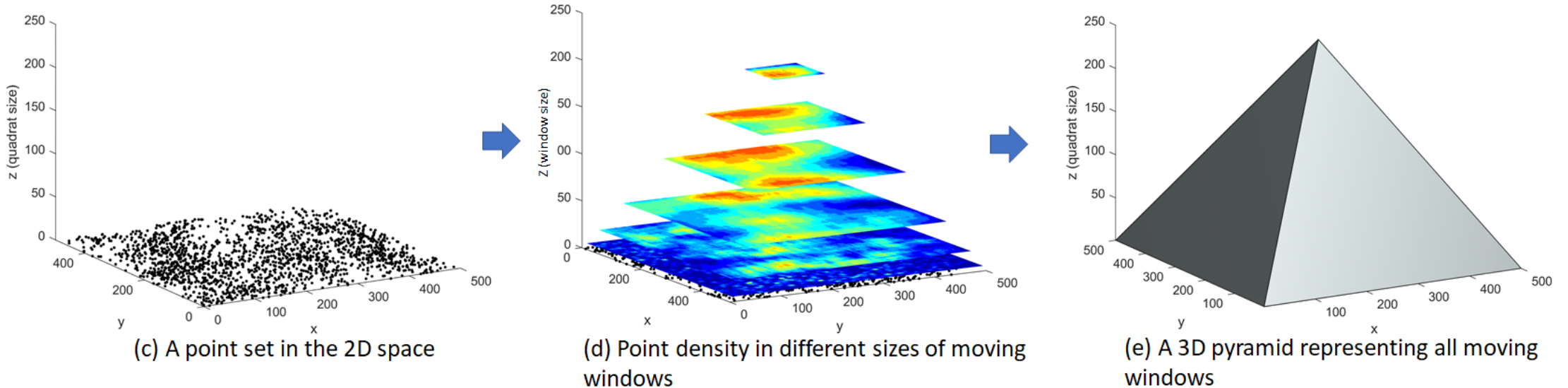
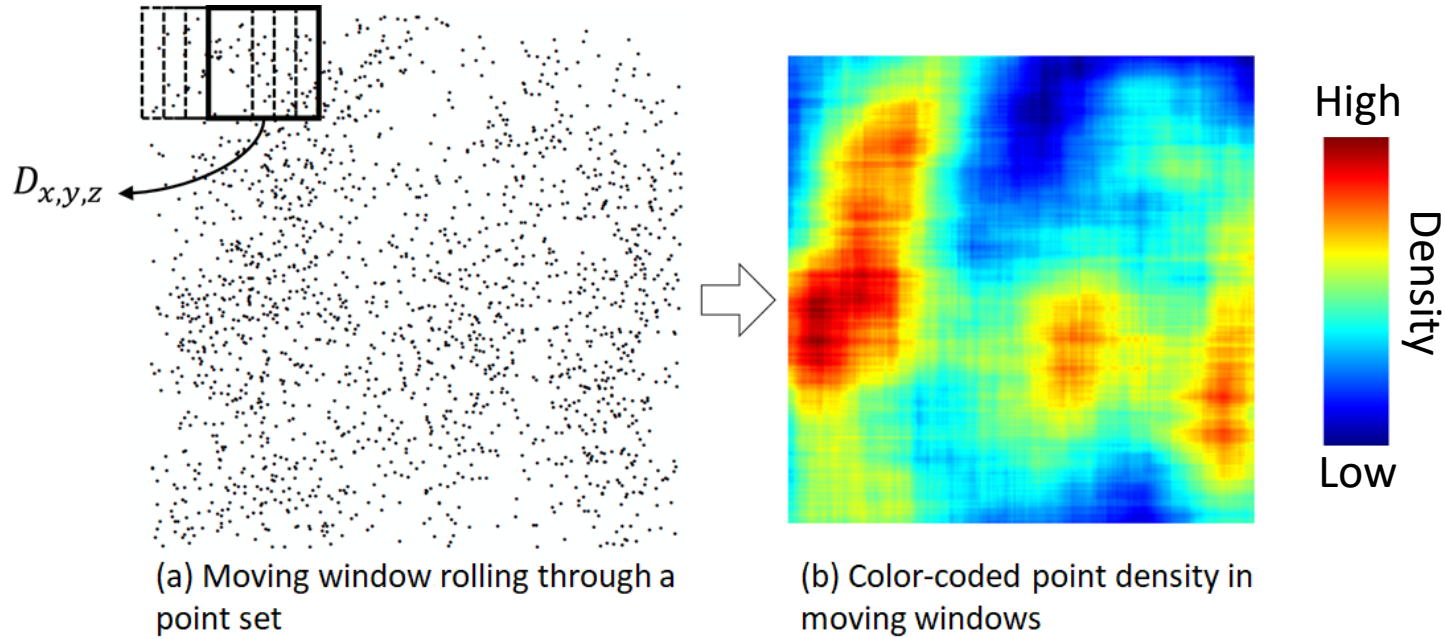


$$f(x_1, y_1, z_3)$$

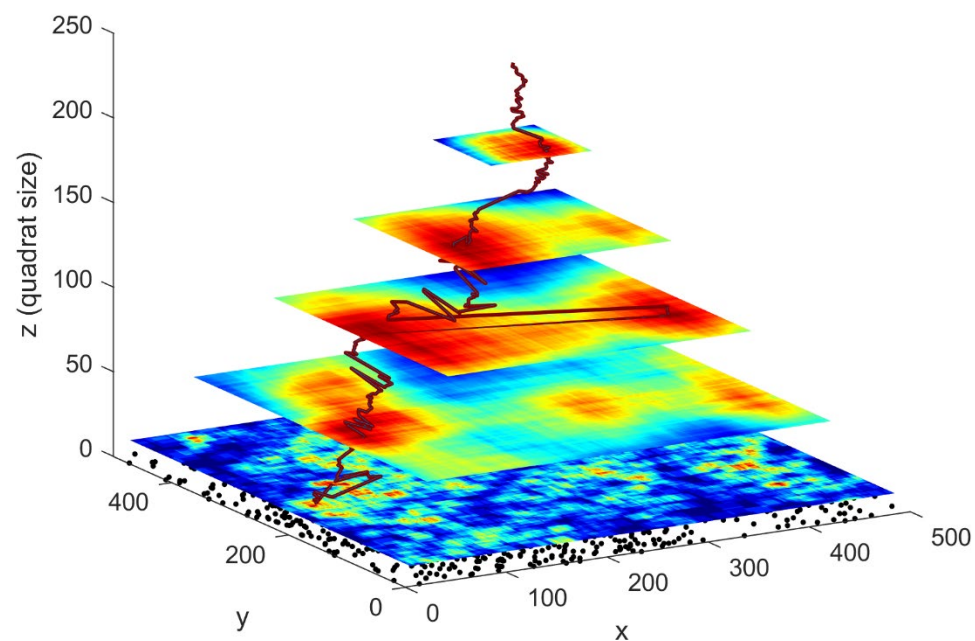
Scale Issue in Point Pattern Analysis



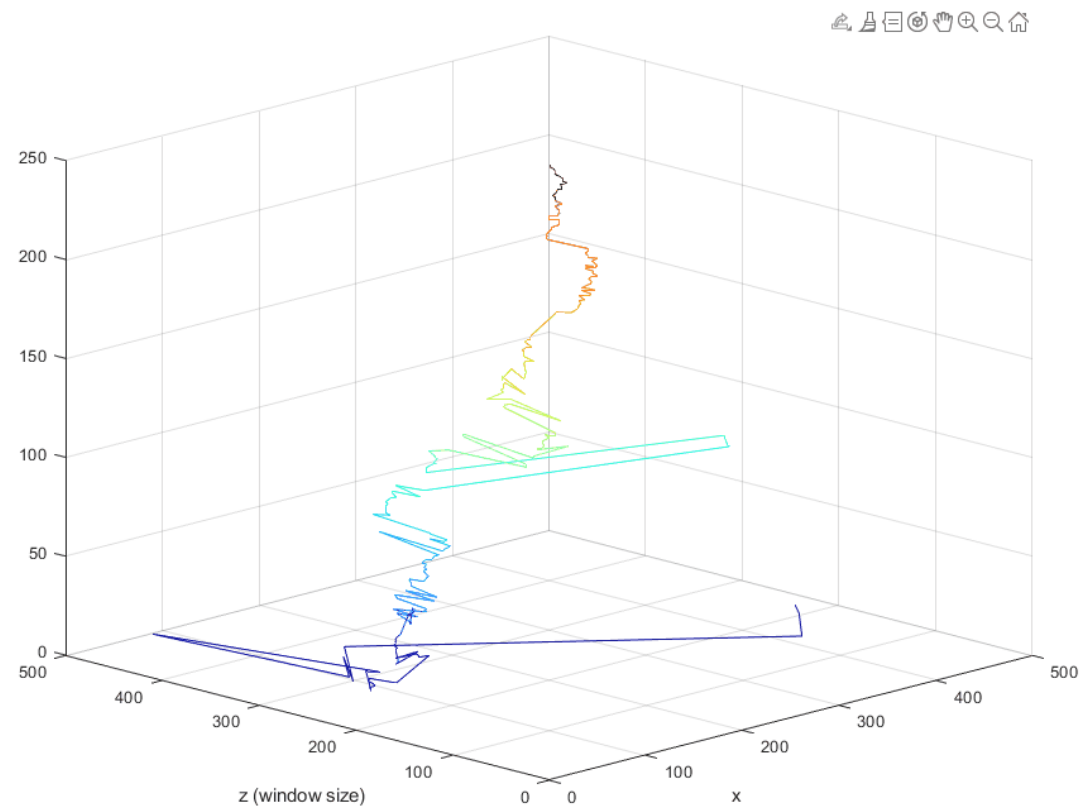
Multi-Scale Quadrat Density in PM



Global Peaks of Quadrat Density

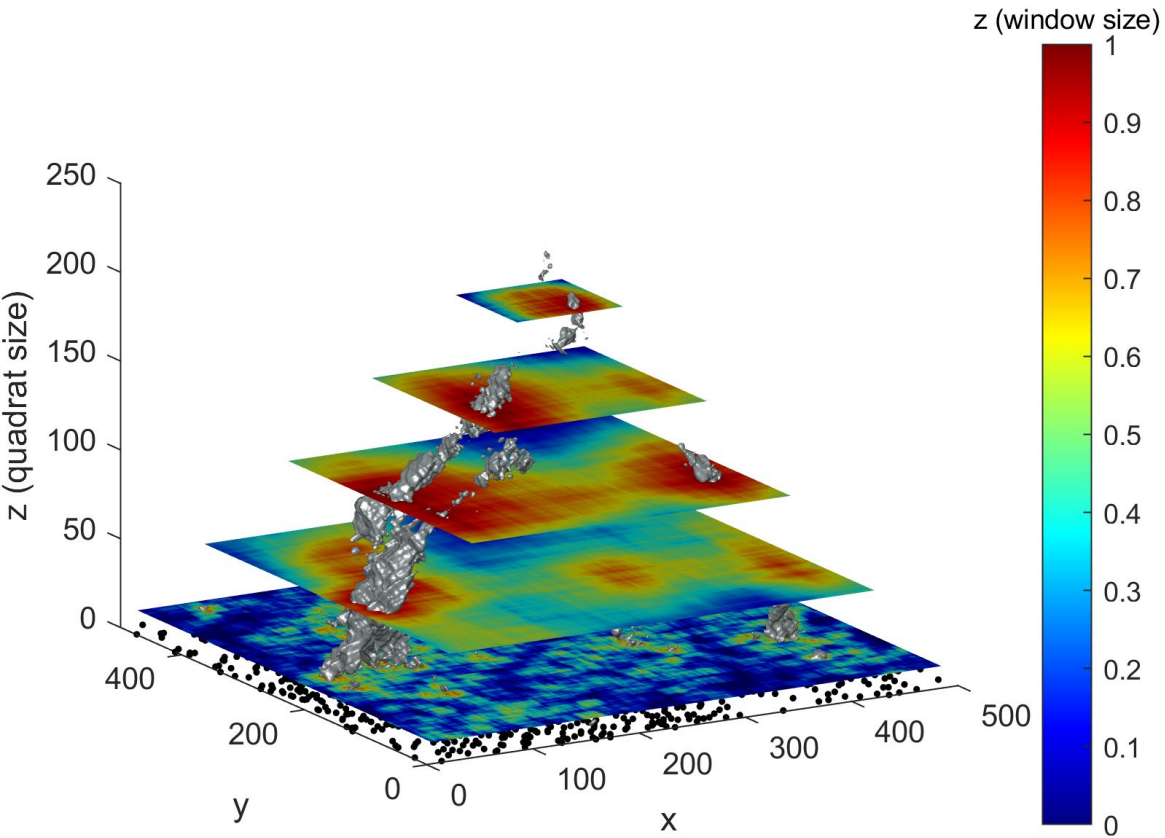


Global density peaks at different scales

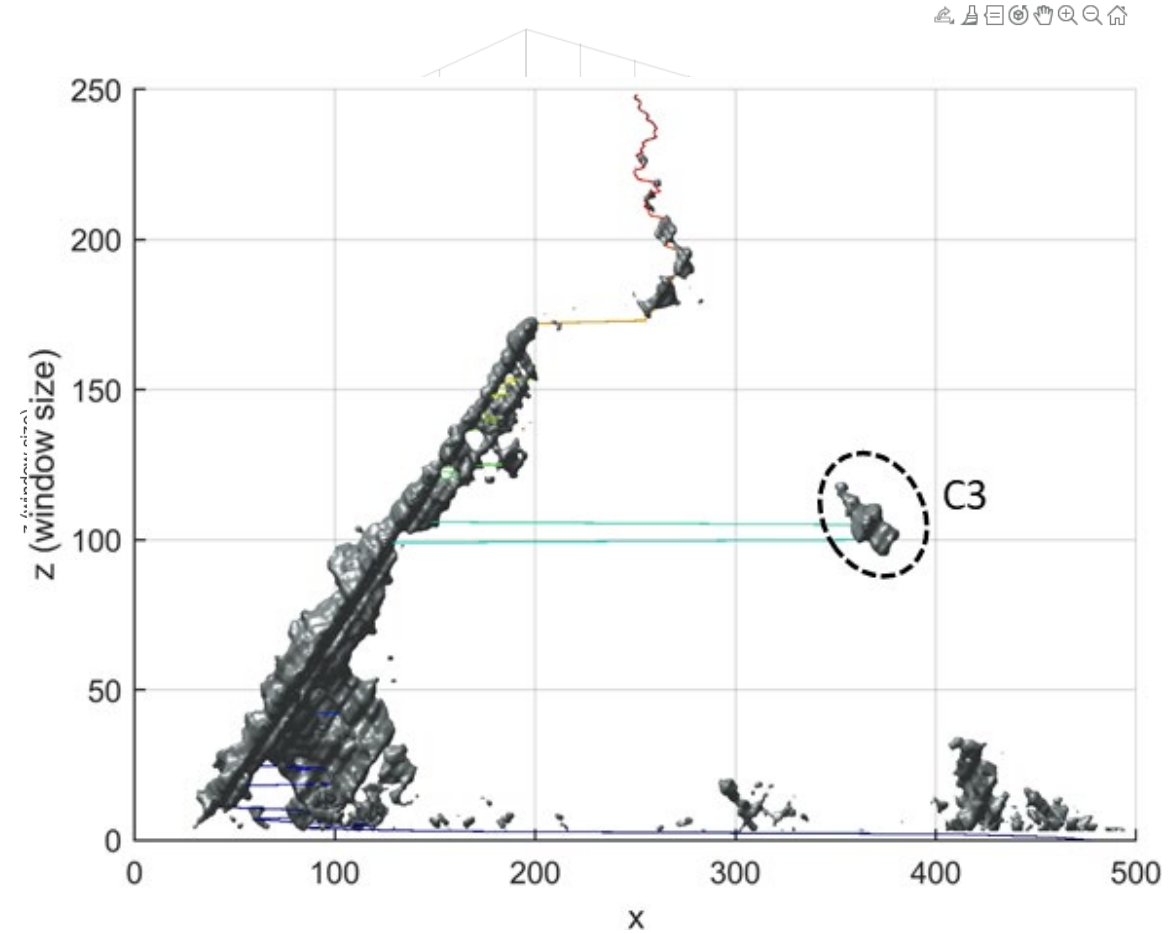


Global density peaks viewed from different angles

Isosurface of Quadrat Density

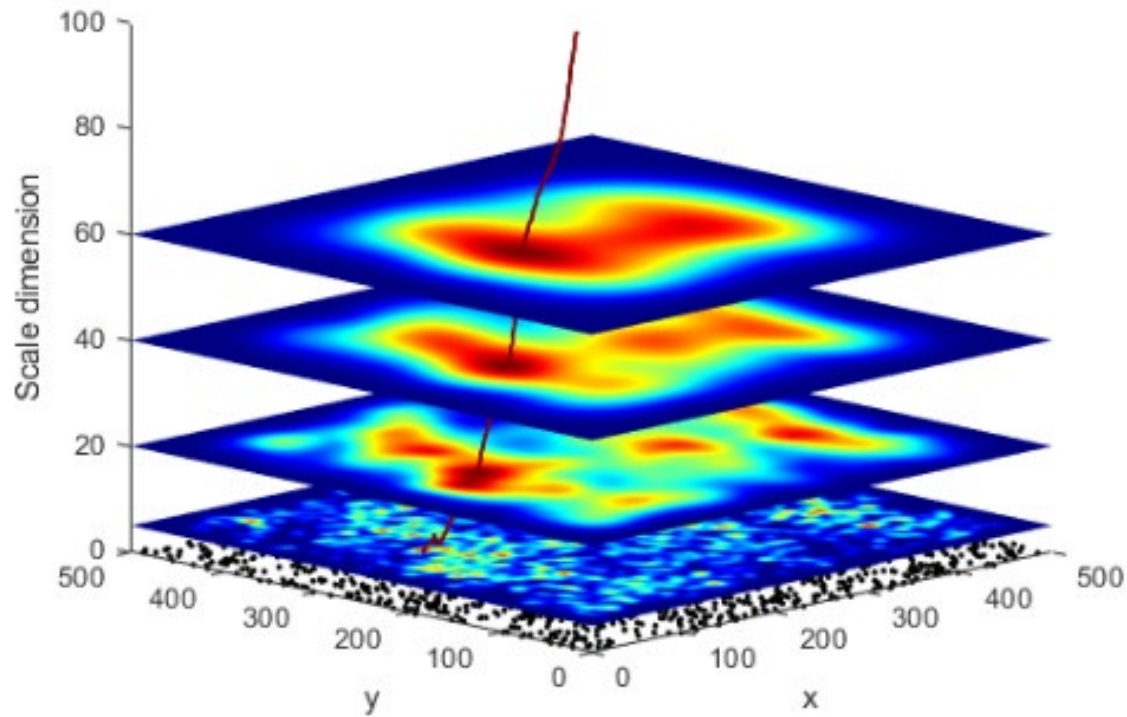


High density (99th percentile) voxels across scales

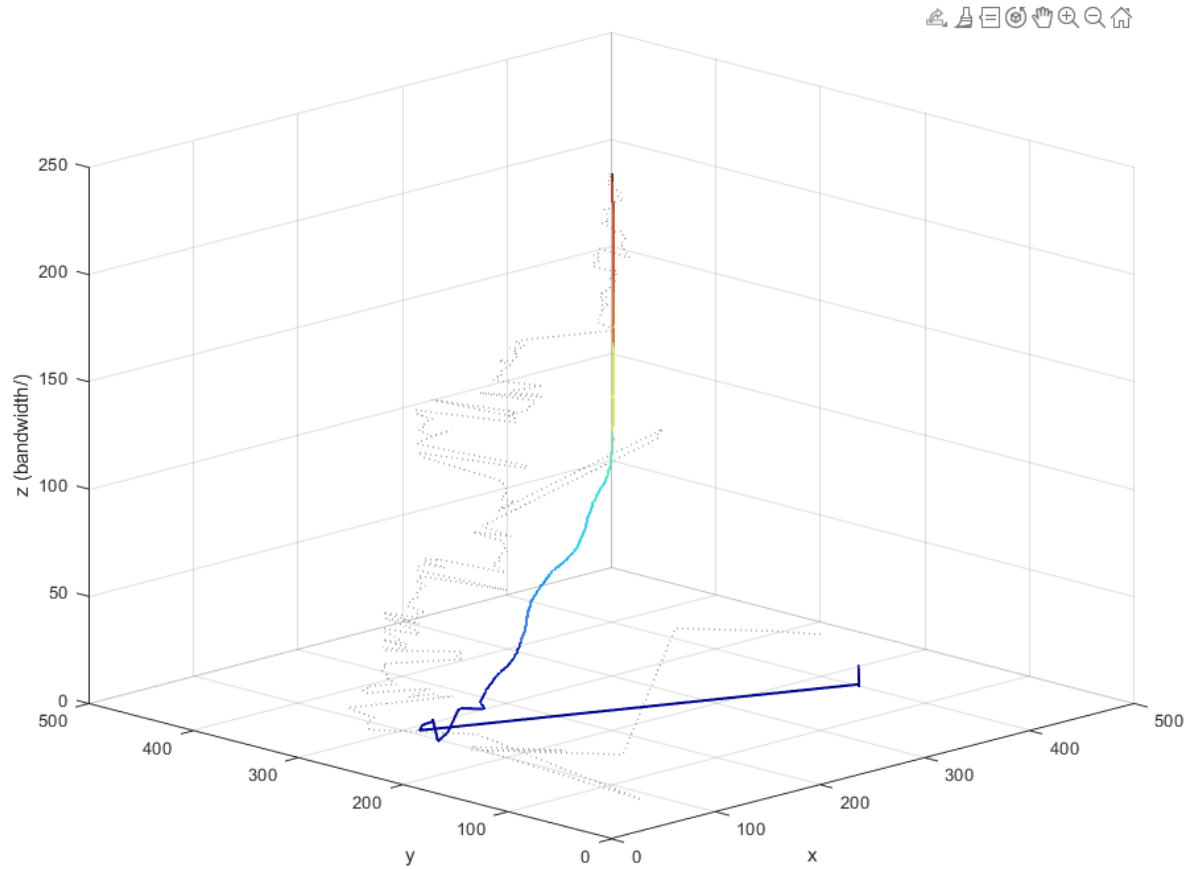


Voxels of density at the 99th percentile

Global Peaks and Isosurface of Kernel Density

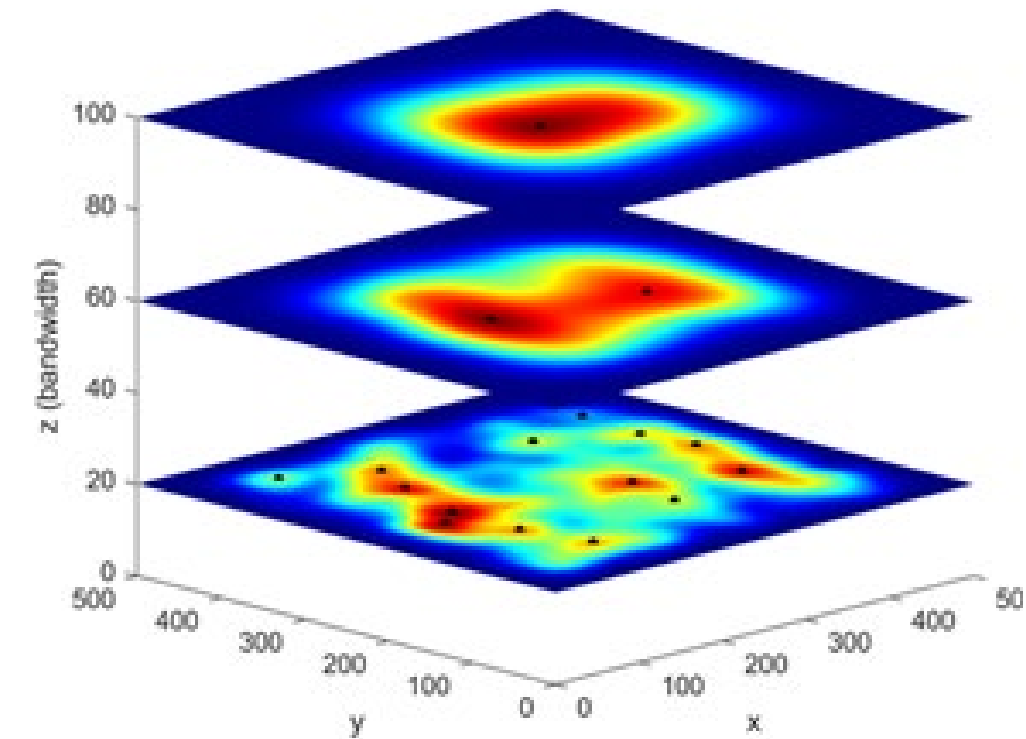


Linking global peaks of kernel density at different scales

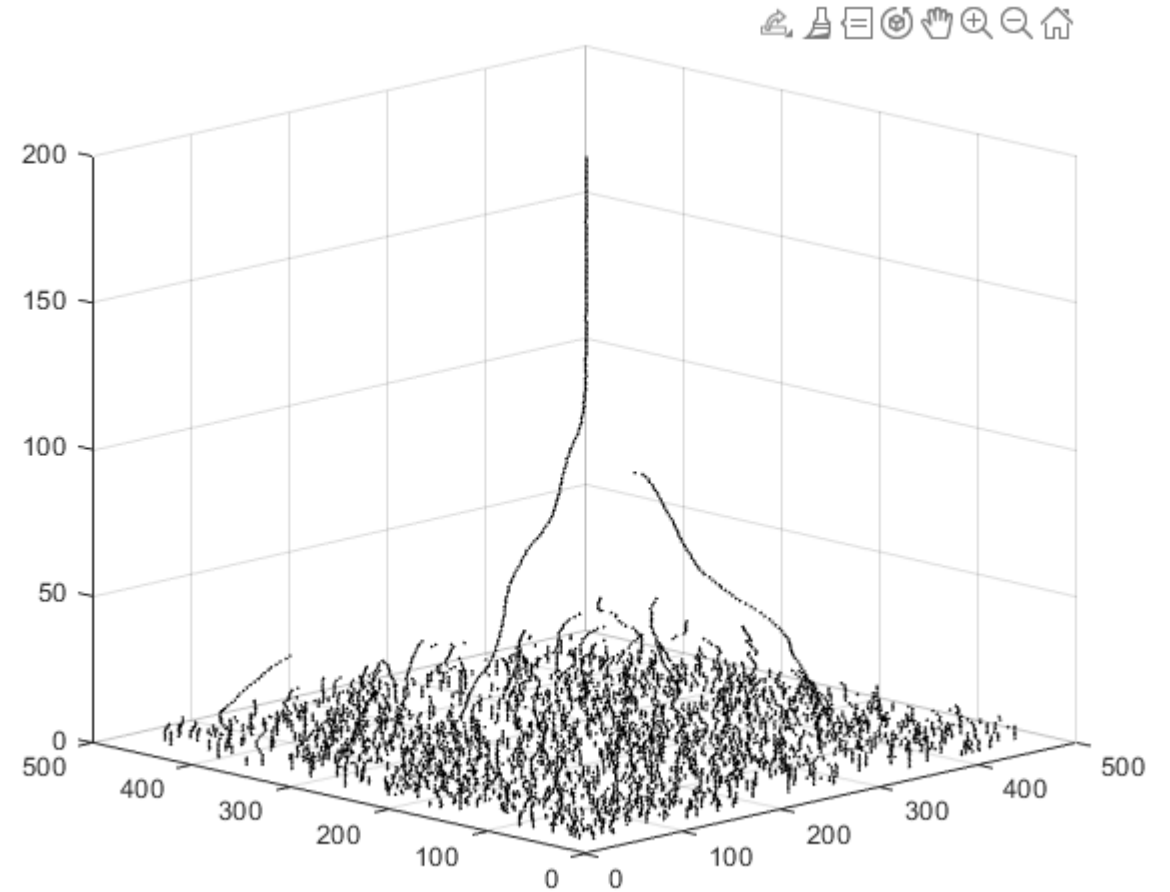


Global density peaks viewed from different angles

Local Peaks

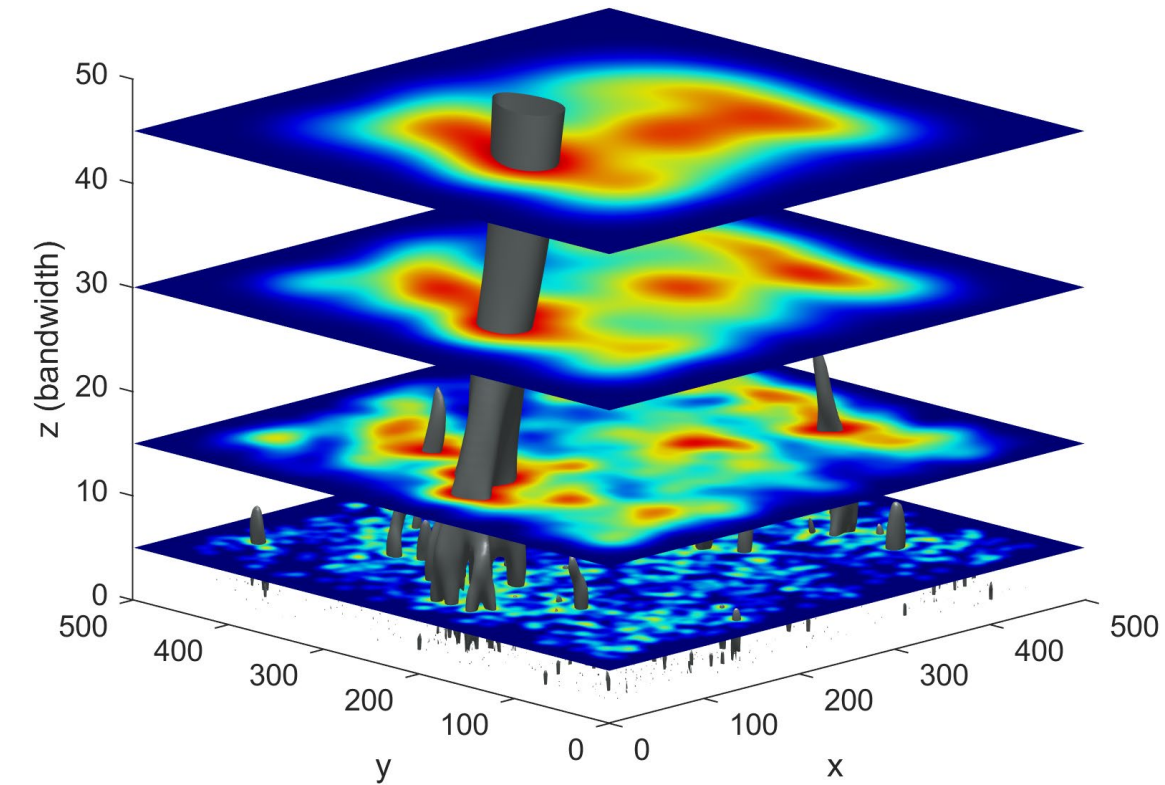


Local peaks detected at different scales

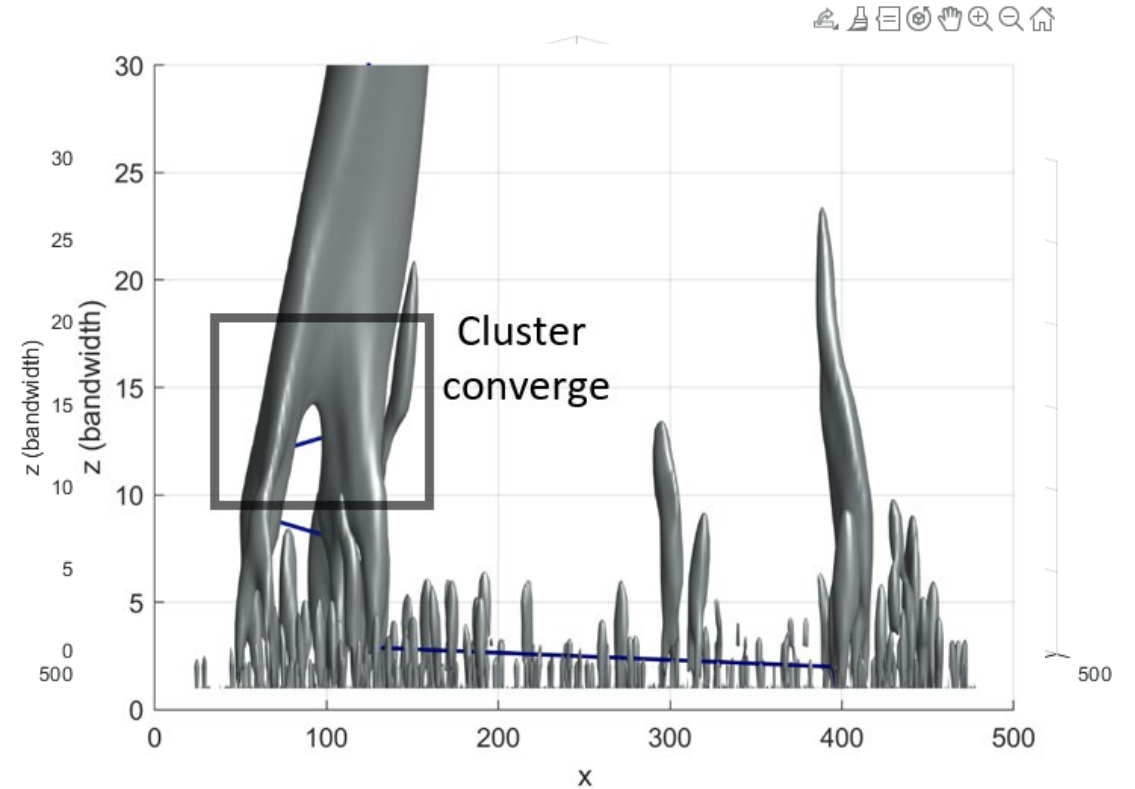


Local density peaks viewed from different angles

Isosurface of Kernel Density



High density (99th percentile) voxels across scales

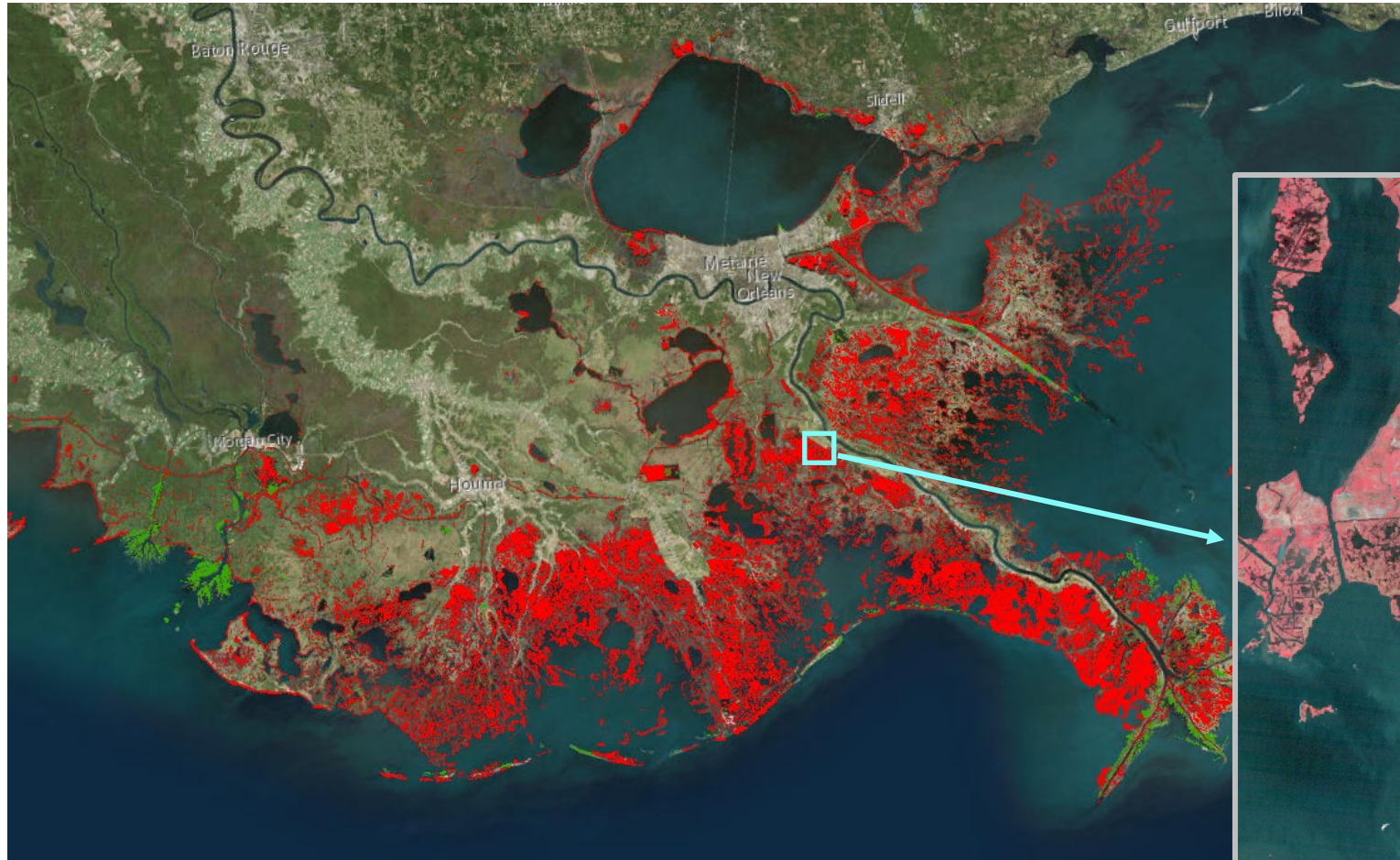


(d) Azimuth: 0° Altitude: 0°

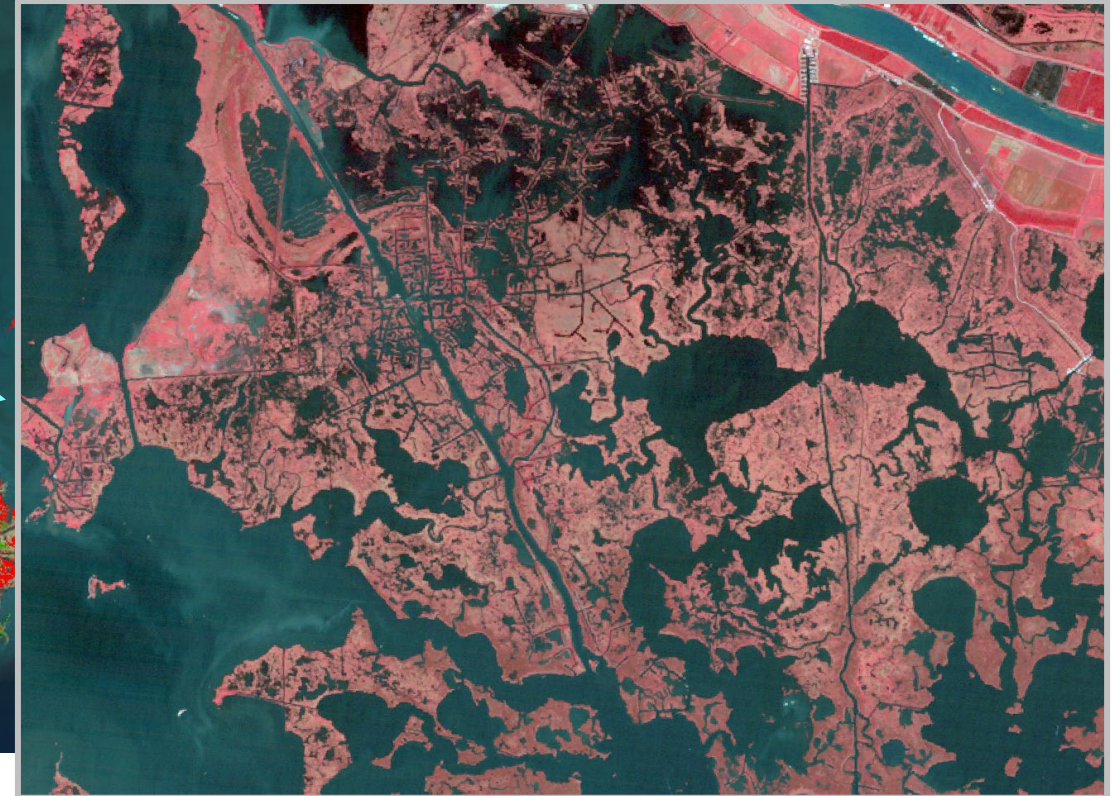
Voxels of density at the 99th percentile

Land Cover Change Analysis

Land cover change detection and modeling are scale-dependent

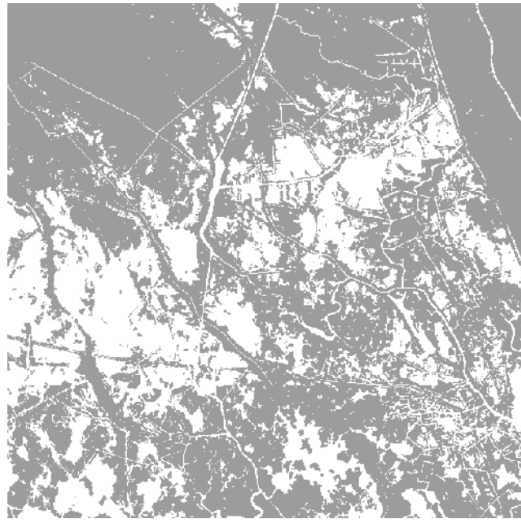


Land loss and gain from 1930 – 2010 in Mississippi Delta

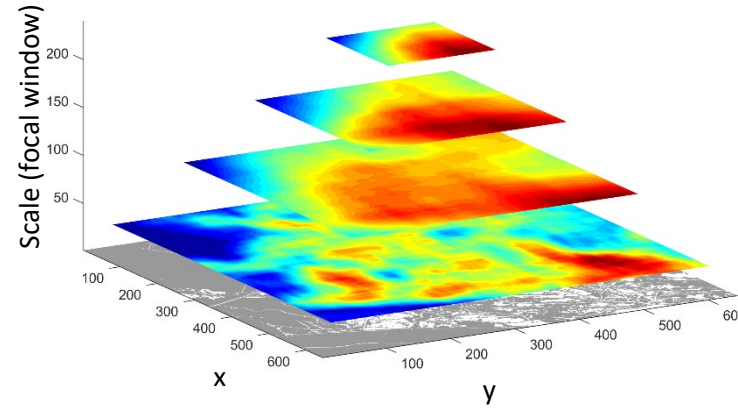


Wetland fragmentation is a driving factor of land loss

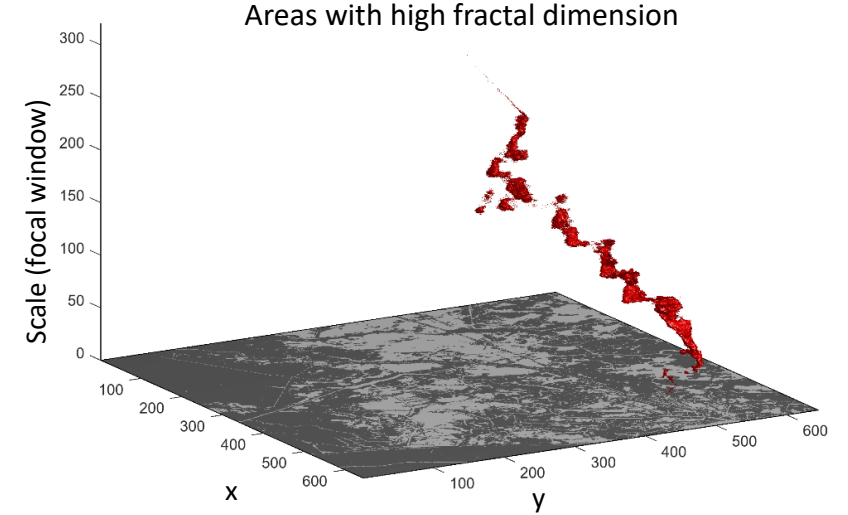
Multi-Scale Modeling of Land Loss



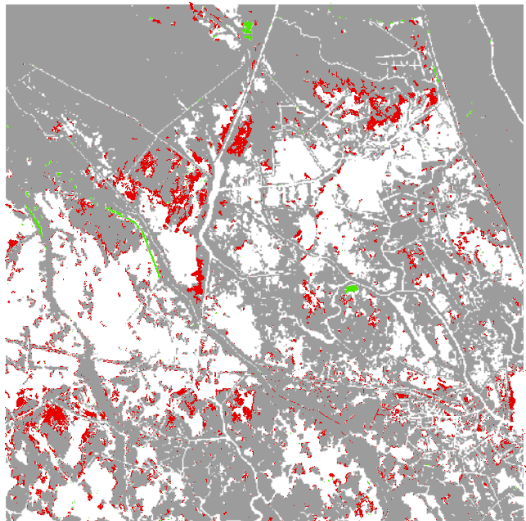
Land cover 2001



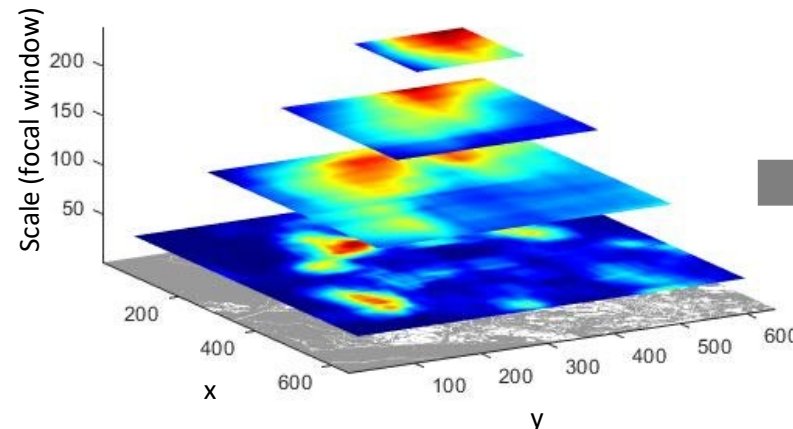
Multi-scale fractal dimensions in PM



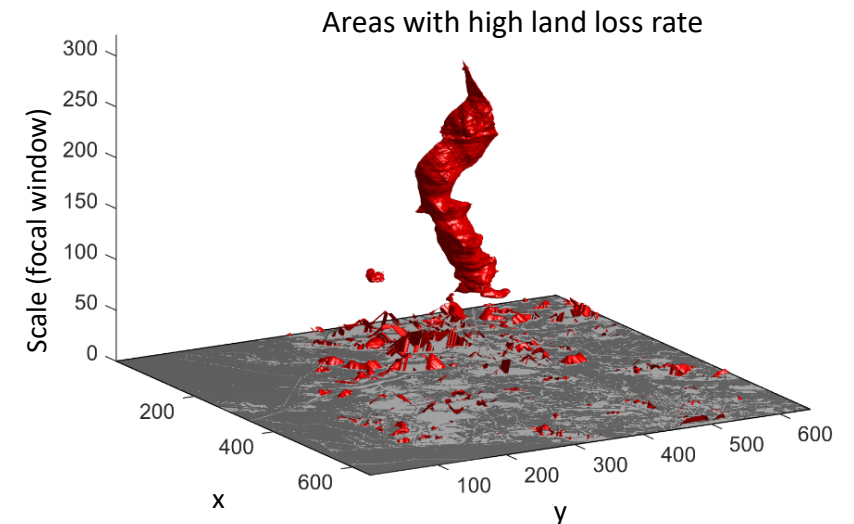
Visual analytics



Land loss 2001 - 2016

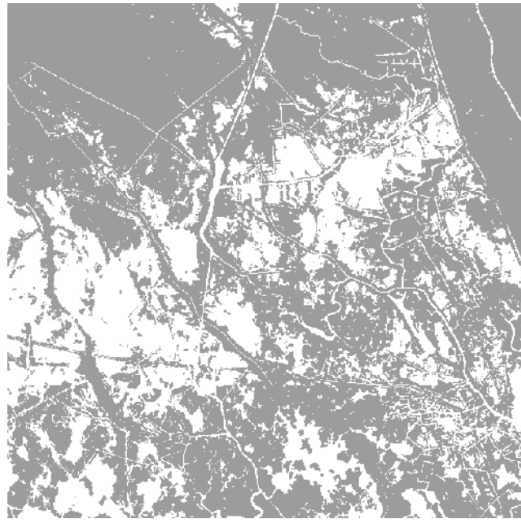


Multi-scale land loss ratio in PM

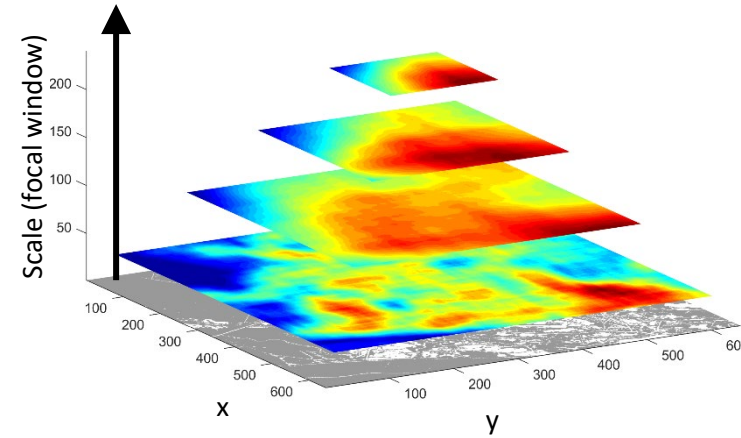


Visual analytics

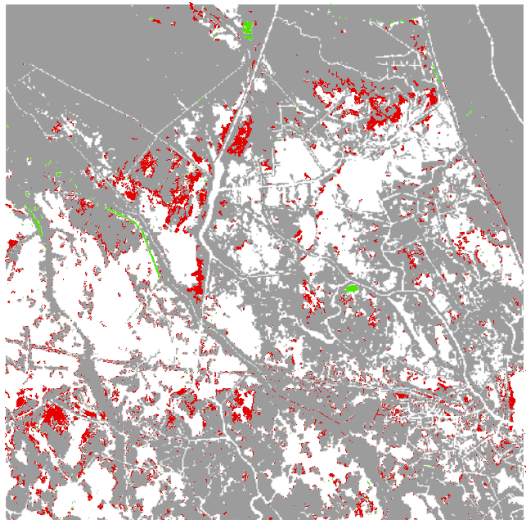
Multi-Scale Modeling of Land Loss



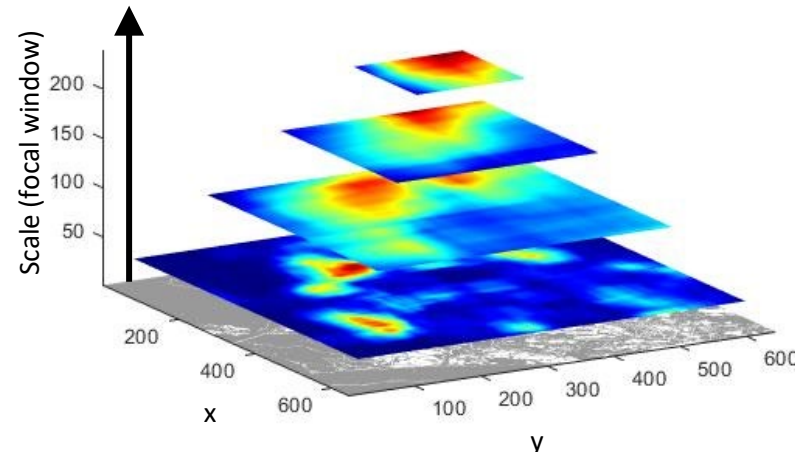
Land cover 2001



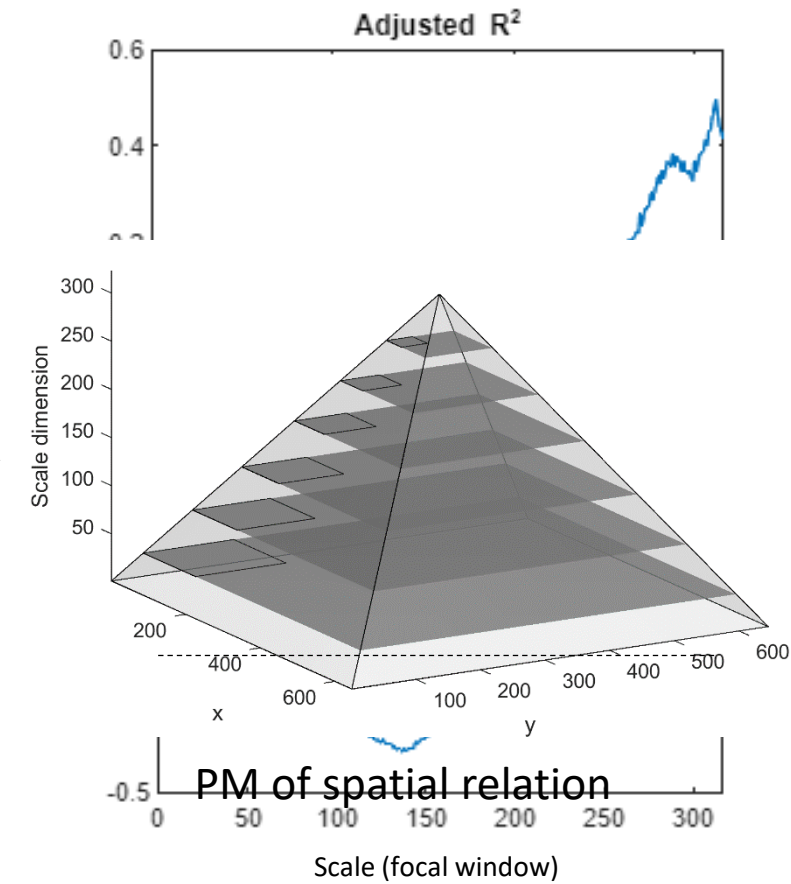
Multi-scale fractal dimensions in PM



Land loss 2001 - 2016



Multi-scale land loss ratio in PM



PM of spatial relation

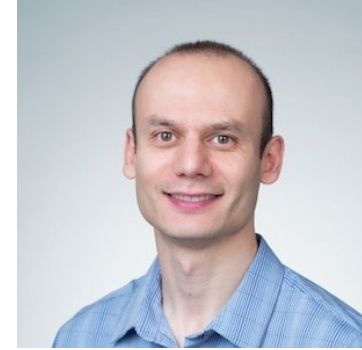
CroScalar: A Multi-Institution Collaboration



Yi Qiang, Assistant Professor
University of South Florida



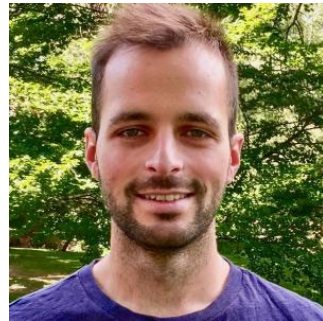
Barbara (babs) Battenfield, Emeritus Professor
University of Colorado - Boulder



Nodari Sitchinava,
Associate Professor
University of Hawaii - Manoa



Jinwen Xu (PhD candidate)
University of South Florida



Georgios (George) Charisoulis
(PhD candidate)
University of Colorado - Boulder



Kate Carlson
(MA, graduated in 2021)
University of Colorado - Boulder



Katie Tyler
(MA, graduated in 2022)
University of Colorado – Boulder



Thank you!

Reference

Y. Qiang, B. Battenfield, and J. Xu, “Analyzing multi-scale spatial point patterns in a pyramid modeling framework,” *Cartography and Geographic Information Science*, pp. 1–14, Apr. 2022, doi: [10.1080/15230406.2022.2048419](https://doi.org/10.1080/15230406.2022.2048419).

Y. Qiang, B. P. Battenfield, and M. B. Joseph, “How to Measure Distance on a Digital Terrain Surface and Why it Matters in Geographical Analysis,” *Geographical Analysis*, vol. 53, no. 3, pp. 588–622, 2021, doi: [10.1111/gean.12255](https://doi.org/10.1111/gean.12255).

Qiang, Y. and Van de Weghe, N. (2019) “Re-Arranging Space, Time and Scales in GIS: Alternative Models for Multi-Scale Spatio-Temporal Modeling and Analyses”, *ISPRS International Journal of Geo-Information*. vol: 8(2). DOI:10.3390/ijgi8020072

Funding Source

This article is based on work supported by the NSF Methodology, Measurement & Statistics (MMS) Program (Award No. 1853866).

Website: <https://croscalar.github.io>

