

Fast SFC-based point cloud change detection

Vitali Diaz, Peter van Oosterom, Martijn Meijers,
Edward Verbree, Thijs van Lankveld and Nauman Ahmed



November 12, 2024



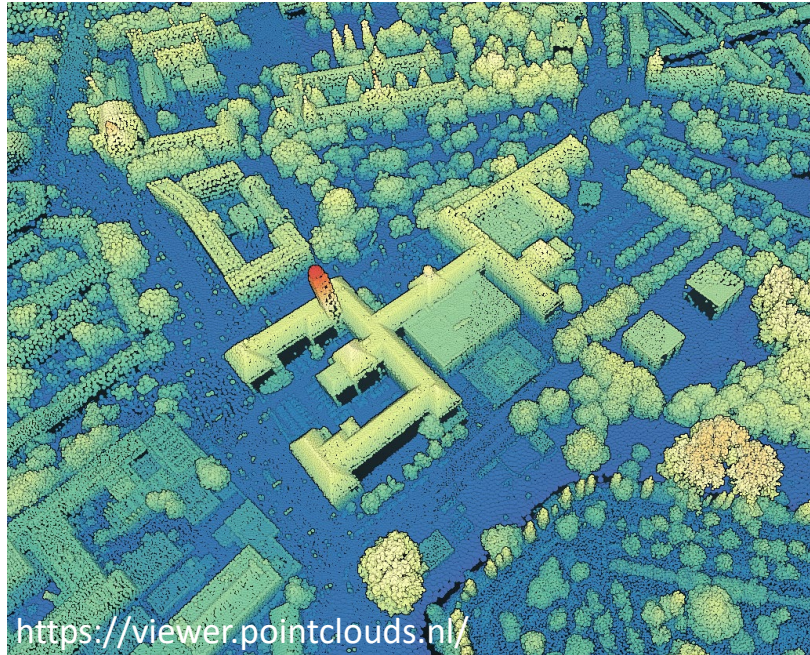
Outline

1. Motivation
2. Direct point cloud-based change detection
3. Fast Space-Filling Curve-based point cloud change detection
4. Conclusions
5. Future work



1. Motivation: The dream

Point Clouds as a digital representation of the (built) environment

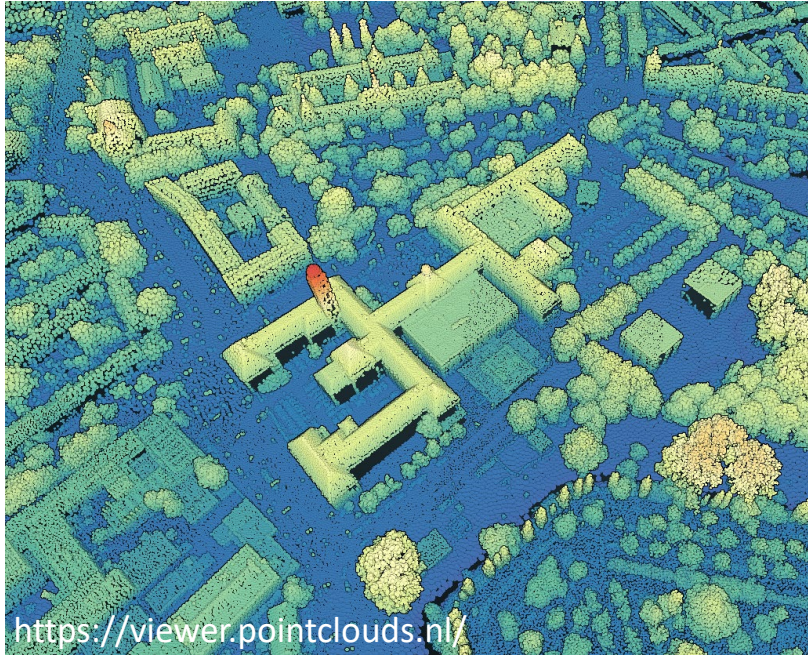


AHN3
point cloud from
Airborne Laser Scanning
(ALS)



Digital Surface Model (DSM)
Point cloud using aerial
photographs

Point Clouds as a digital representation of the (built) environment



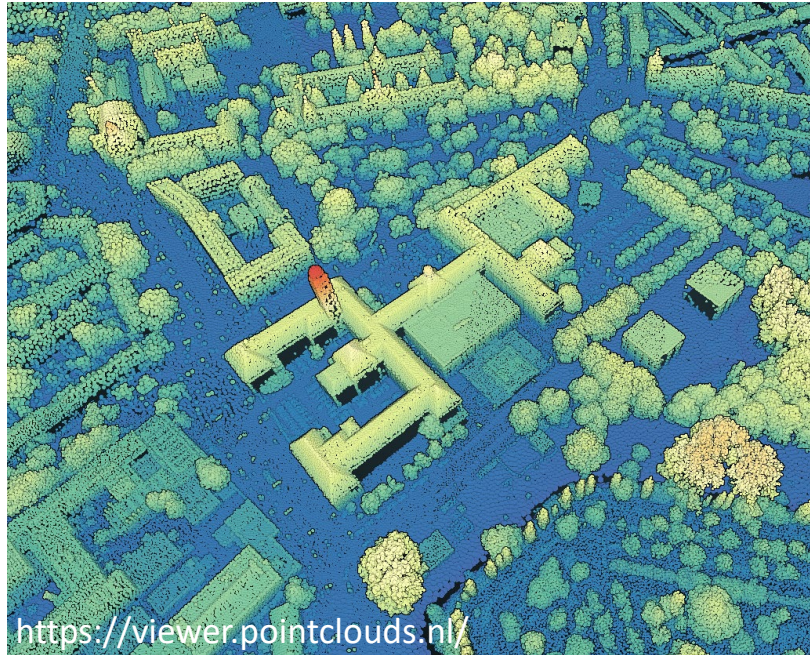
AHN3
point cloud from
Airborne Laser Scanning
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Digital Surface Model (DSM)
Point cloud using aerial
photographs

Indoor, occluded
areas, shadow
zones?

Point Clouds as a digital representation of the (built) environment



AHN3
point cloud from
Airborne Laser Scanning
(ALS)



Digital Surface Model (DSM)
Point cloud using aerial
photographs

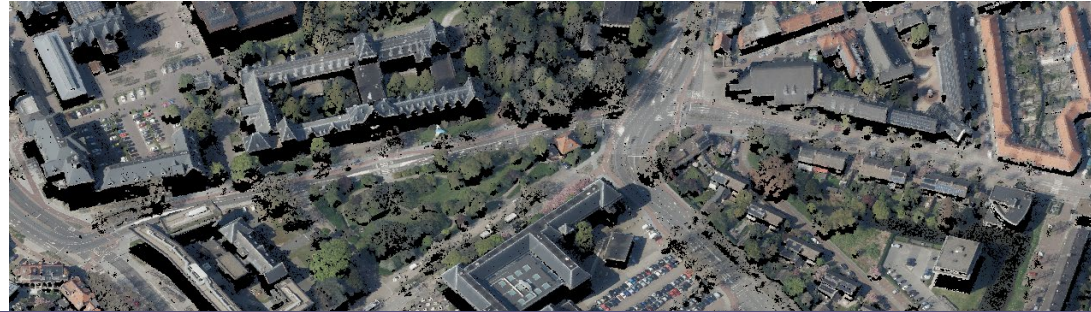
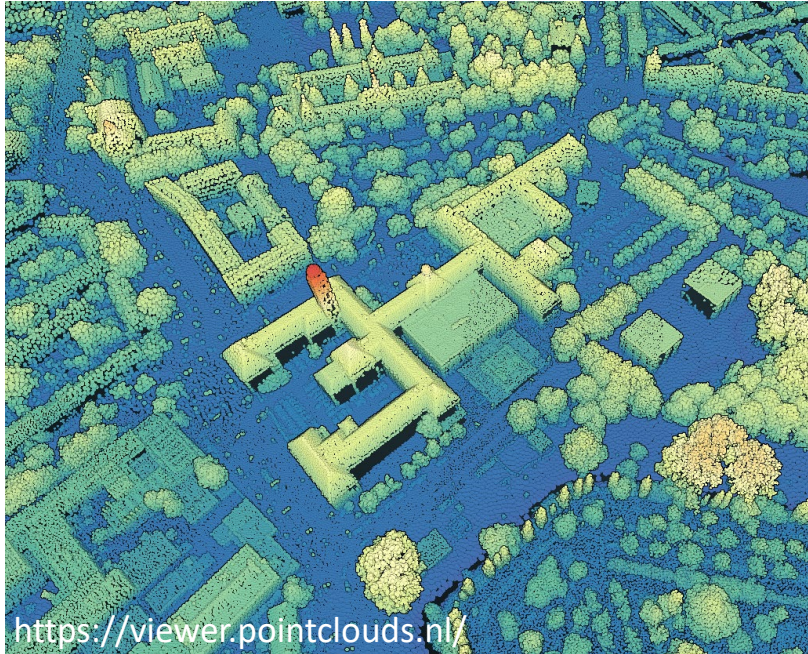
Indoor, occluded
areas, shadow
zones?

filling the data gap, and
improving the nD representation
of the (built) environment



Mobile Laser Scanning
MLS

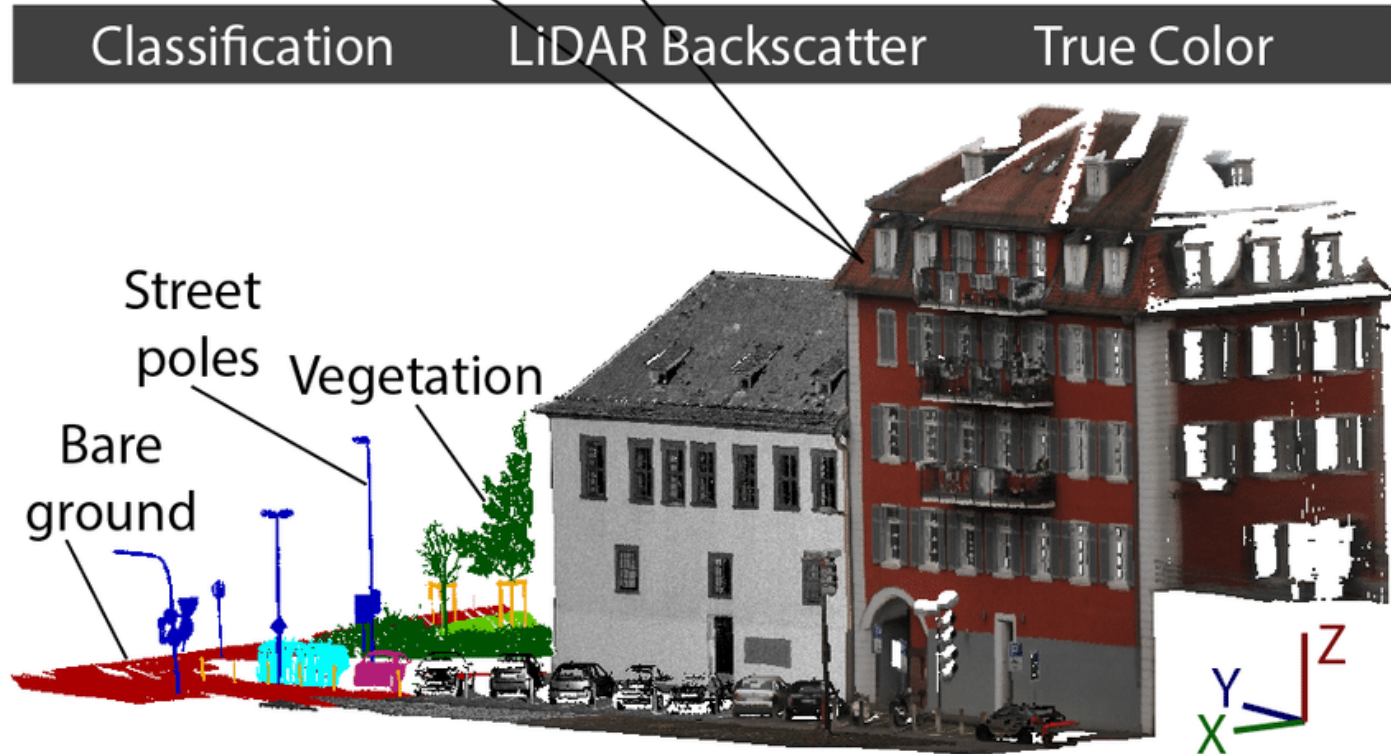
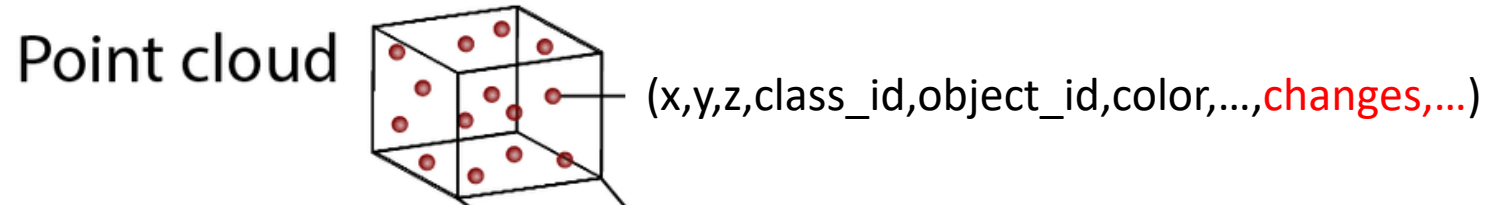
Point Clouds as a digital representation of the (built) environment



AHN3
point cloud from
Airborne Laser Scanning
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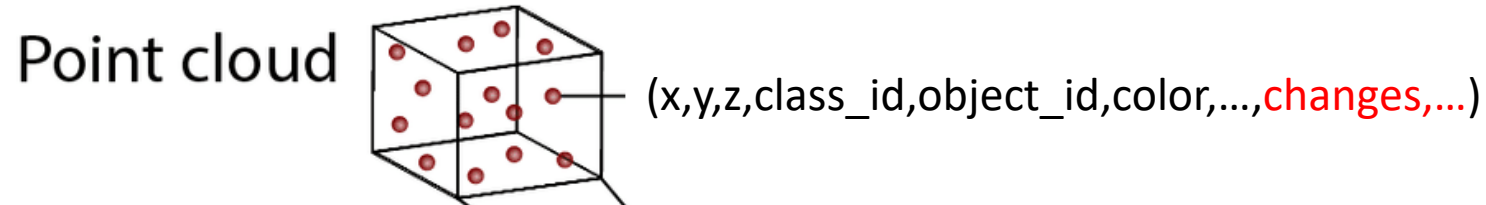
Rendering:
3D Gaussian Splatting

Beside x,y,z coordinates, other attributes are (and can be) attached to every single point



Ghamisi et al. (2019). IEEE GRS Magazine,
<https://doi.org/10.1109/mgrs.2018.2890023>

Beside x,y,z coordinates, other attributes are (and can be) attached to every single point



Classification LiDAR Backscatter True Color



Visualising additional attributes



Ghamisi et al. (2019). IEEE GRS Magazine,
<https://doi.org/10.1109/mgrs.2018.2890023>



1 Motivation

Point clouds as direct data sources for change detection

Point clouds provide rich information for visualization and serve as direct data sources for [change detection analysis](#), enabling analysts to detect, quantify, and interpret spatial changes directly without data conversion and additional storage (for grid conversion).

Leveraging inherent point cloud characteristics

By utilizing the unique characteristics of point clouds, researchers and practitioners can gain valuable insights into dynamic processes, supporting informed decision-making, planning, and development.

Research contribution

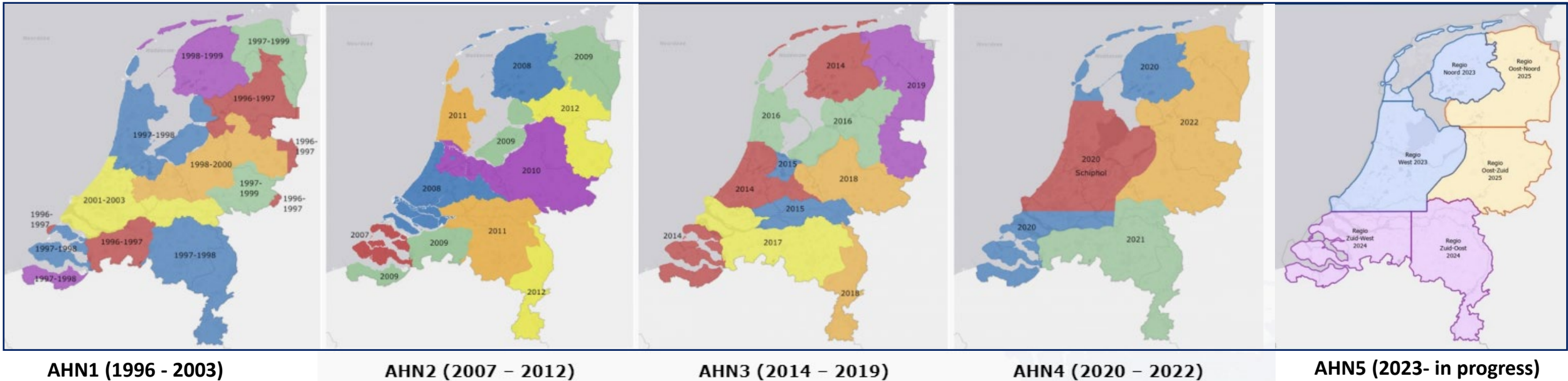
This research advances the development of direct point cloud change detection methods and explores optimization strategies to enable fast computations on large-scale datasets, such as the massive point cloud of the entire Netherlands (AHN).

The data (and the challenge): the height model of the Netherlands



The 'Actueel Hoogtebestand Nederland' (AHN) is the massive point cloud dataset acquired through laser altimetry and covers height information of the Netherlands

<https://www.ahn.nl/>



6 to 10 points per m²

10 to 14 points per m²



Using a bow is not an option because I'm not going to turn stones into arrows. The choice is to use the slingshot to throw stones.

So the questions:

-which method and why?

-how to develop the implementation for fast computation?



2. Direct point cloud-based change detection



2 Direct pc-based change detection

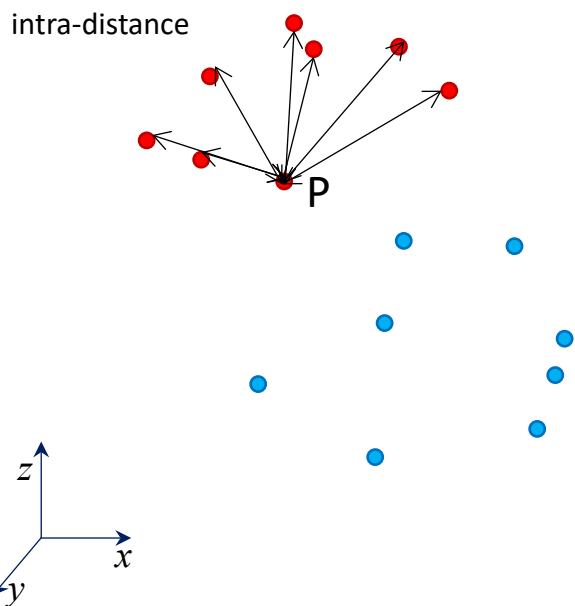
- two experiments to find the suitable pc-based method:
 - (1) controlled offset,
 - (2) comparison over spatio-temporal point clouds
- range of algorithms (as available in CloudCompared)
- reimplemented in Matlab for full control
- testing with various data sets (AHN, CoastScan, lake, bunny, etc)



2 Direct pc-based change detection

Some definitions

• reference cloud time t • compared cloud time $t+1$



Intra-distance: distance between points within the same point cloud.

For point P , distance is calculated as the average of the Euclidian distances of its k nearest neighbors.

Intra-distance is calculated for each point.

Diaz et al. (2024)

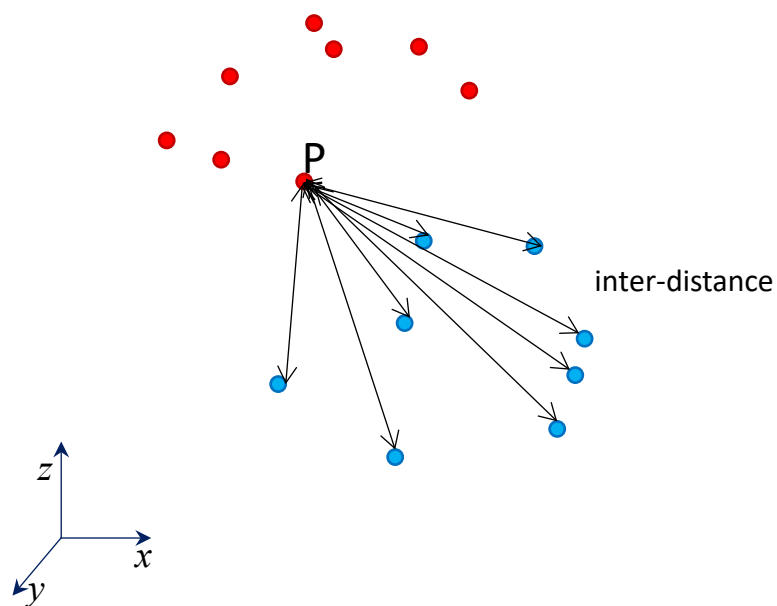
https://doi.org/10.1007/978-3-031-43699-4_20



2 Direct pc-based change detection

Some definitions

• reference cloud time t • compared cloud time $t+1$



Diaz et al. (2024)

https://doi.org/10.1007/978-3-031-43699-4_20

Intra-distance: distance between points within the same point cloud.

For point P , distance is calculated as the average of the Euclidian distances of its k nearest neighbors.

Intra-distance is calculated for each point.

Inter-distance: distance between two corresponding point clouds taken at different epochs (also known as c2c distance).

It is calculated for each point of the compared cloud, providing the spatial dissimilarities/similarities between the two point clouds.

There are various methods to calculate it.

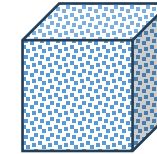
2 Direct pc-based change detection



Experiment 1

Controlled offset test:

(1) Select a point cloud as the "reference cloud"



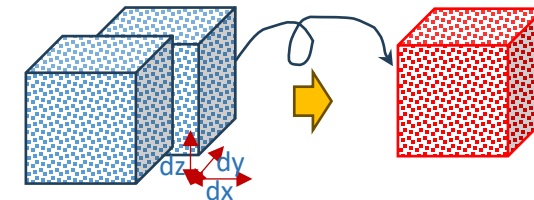
intra-distance
intra-distance_{avg}

(2) Calculate intra-distance for every point of the reference cloud and the average intra-distance.

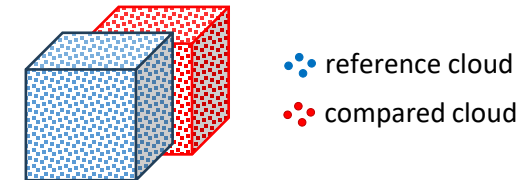
(3) Set artificial offsets based on the average intra-distance (i.e., below, above) to analyze a range of controlled offset scenarios.

dx, dy, dz

(4) Apply the offsets to all points of the reference cloud to create the "compared cloud" for each controlled offset scenario.



(5) Calculate the inter-point distance (c2c distance) between the compared and the reference cloud. Eight different methods.



(6) Finally, evaluate each method to analyze its accuracy in capturing the applied offset.

Diaz et al. (2024)

https://doi.org/10.1007/978-3-031-43699-4_20

2 Direct pc-based change detection



Experiment 1

Diaz et al. (2024)

https://doi.org/10.1007/978-3-031-43699-4_20

- Testing 4 different datasets (bunny, lake, CoastScan, and AHN)
- 3 different types of offsets: vertical, horizontal, and diagonal
- 8 different methods

2 Direct pc-based change detection



Inter-distance (C2C distance)

-Simple approach

- The nearest neighbor dx,dy,dz

-Weighted methods

- Natural Neighbor Interpolation (NNI) dz
- Inverse Distance Weight (IDW) dz

-Local modelling Point-Model

- Least squares plane dx,dy,dz
- Linear interpolation dx,dy,dz
 - 2.5D triangulation dx,dy,dz
- Quadratic (height function) dx,dy,dz

Model-Model

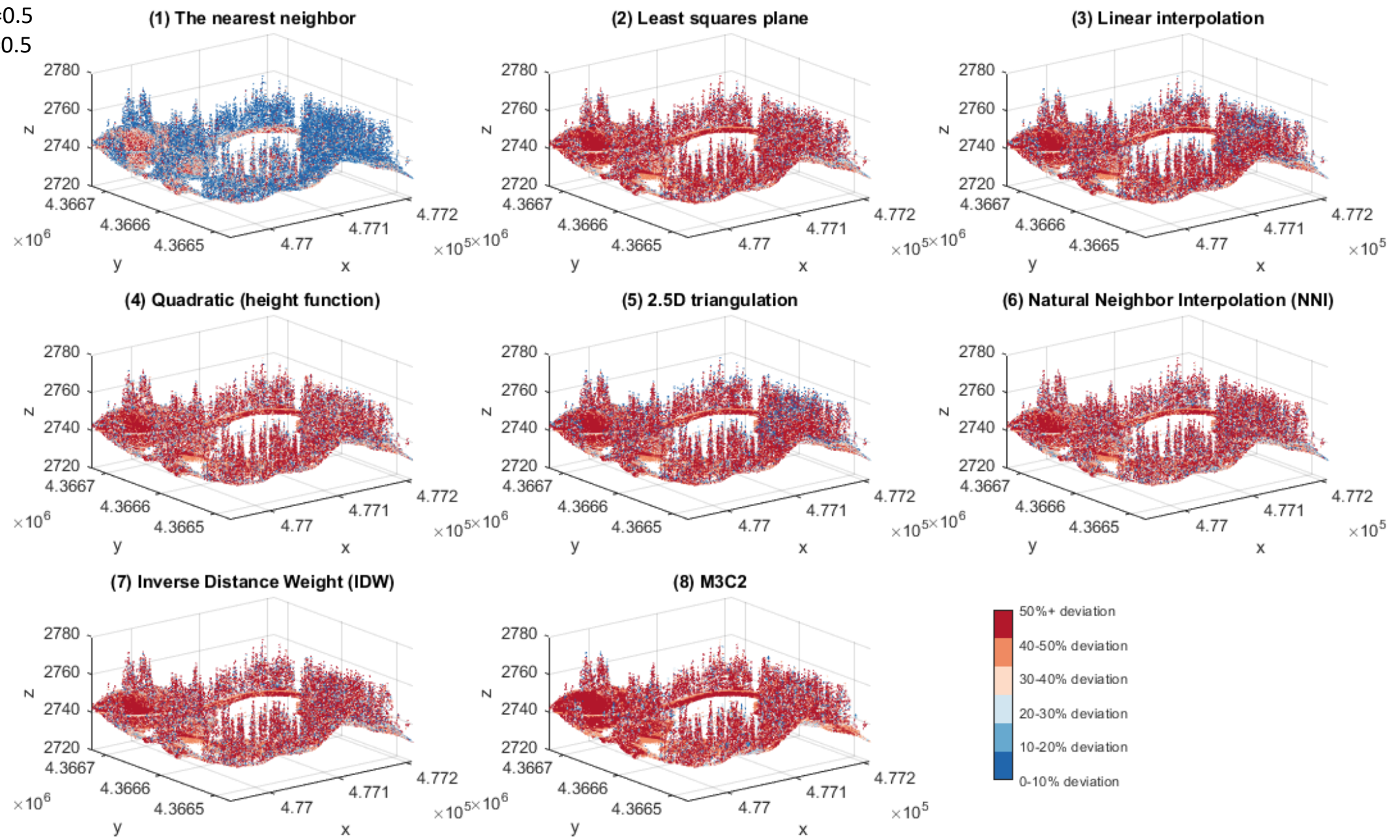
- Multiscale Model to Model Cloud Comparison (M3C2) dx,dy,dz

- time_cost_querying
- time_cost_modelling
- time_cost_distance_calculation

Average
intra-distance
= 1.06 m

$dx=0.5$ m
 $dy=0.5$
 $dz=0.5$

Experiment 1



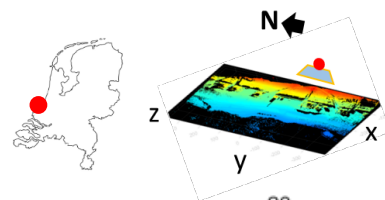
Experiment 1: CoastScan data with controlled offset

Average intra-distance = 0.396 m

$dx=0.2$ m

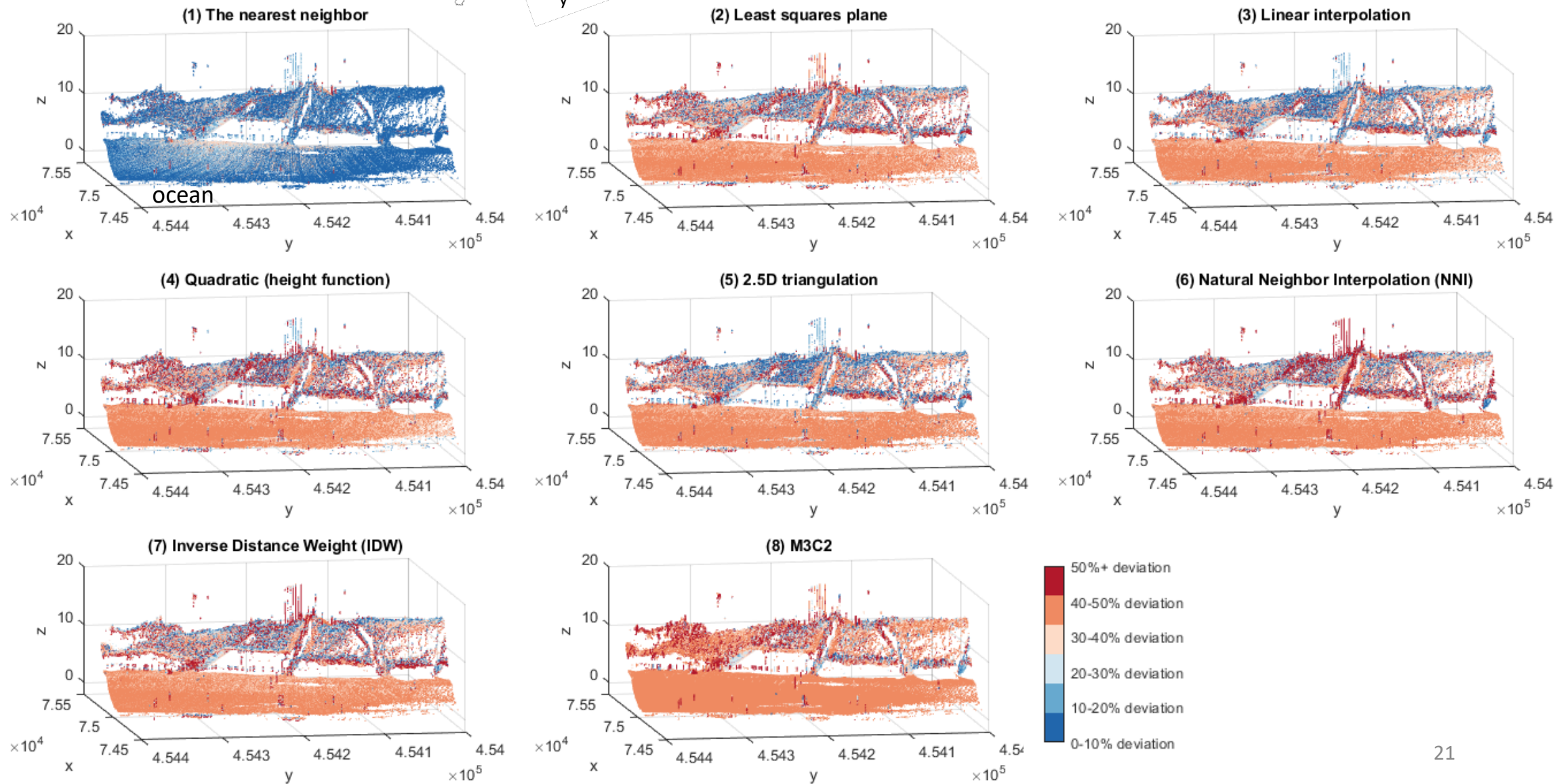
$dy=0.2$

$dz=0.2$



Diaz et al. (2024)

https://doi.org/10.1007/978-3-031-43699-4_20



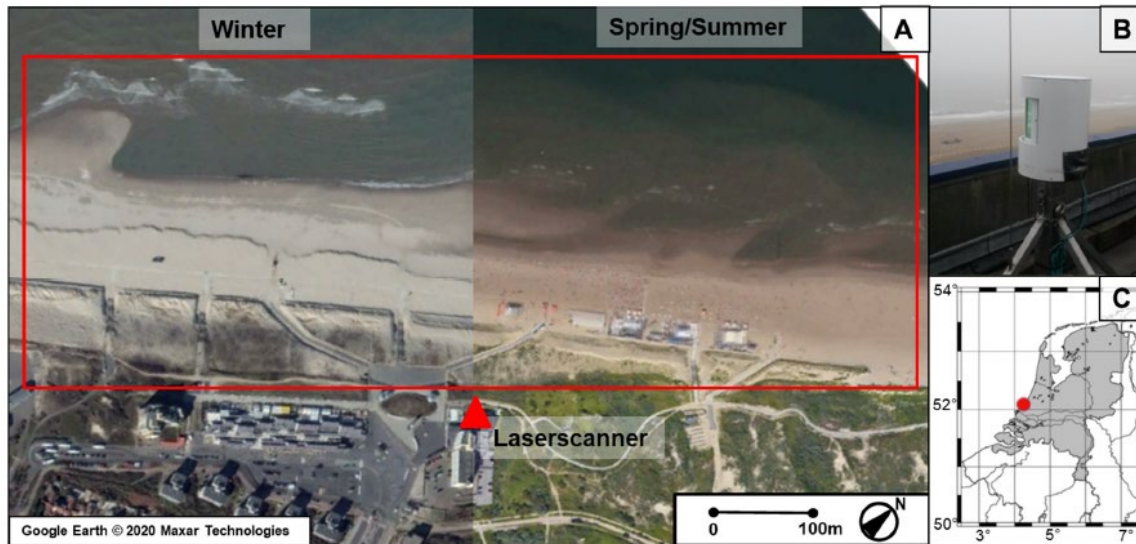
2 Direct pc-based change detection

Experiment 2

Change detection with spatio-temporal point clouds

Diaz et al. (2024)
<https://doi.org/10.5194/egusphere-egu24-8191>

Kijkduin beach-dune system



(A) Aerial photo of the beach-dune in Kijkduin and the scan Area within the red rectangle.
(B) The laser scanner positioned on a hotel next to the beach (indicated by a red triangle in A).
(C) The location of Kijkduin (52.07°N, 4.22°E) in The Netherlands.

Vos et al. (2022)
<https://doi-org.tudelft.idm.oclc.org/10.1038/s41597-022-01291-9>

Performance evaluation

- 1) Comparison with the distances calculated with the raster-to-raster approach
- 2) Time cost comparison
 - time_cost_querying
 - time_cost_modelling
 - time_cost_distance_calculation

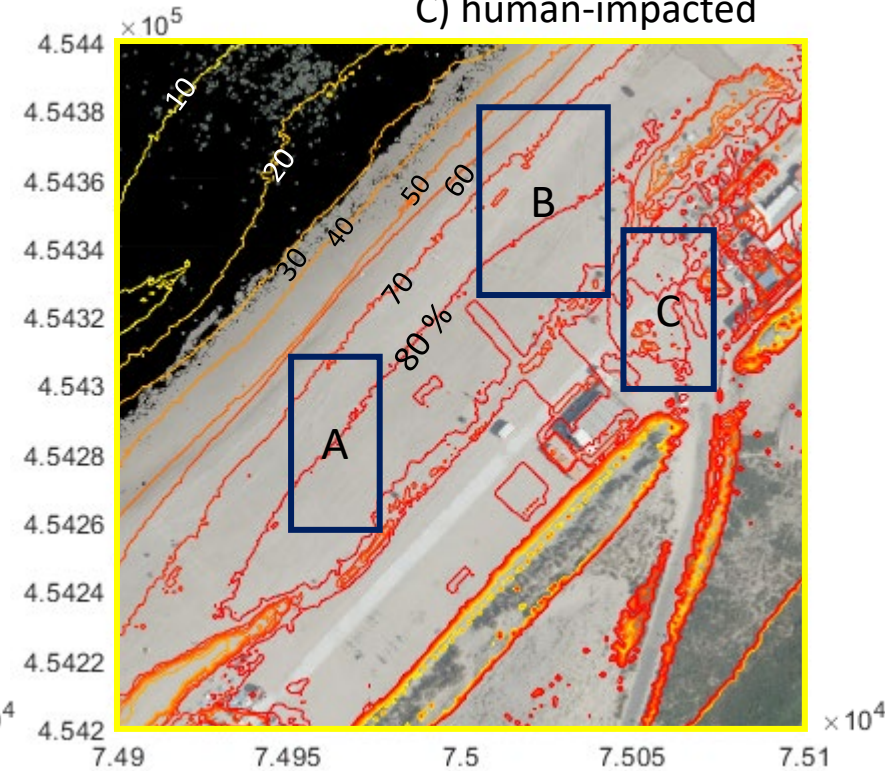
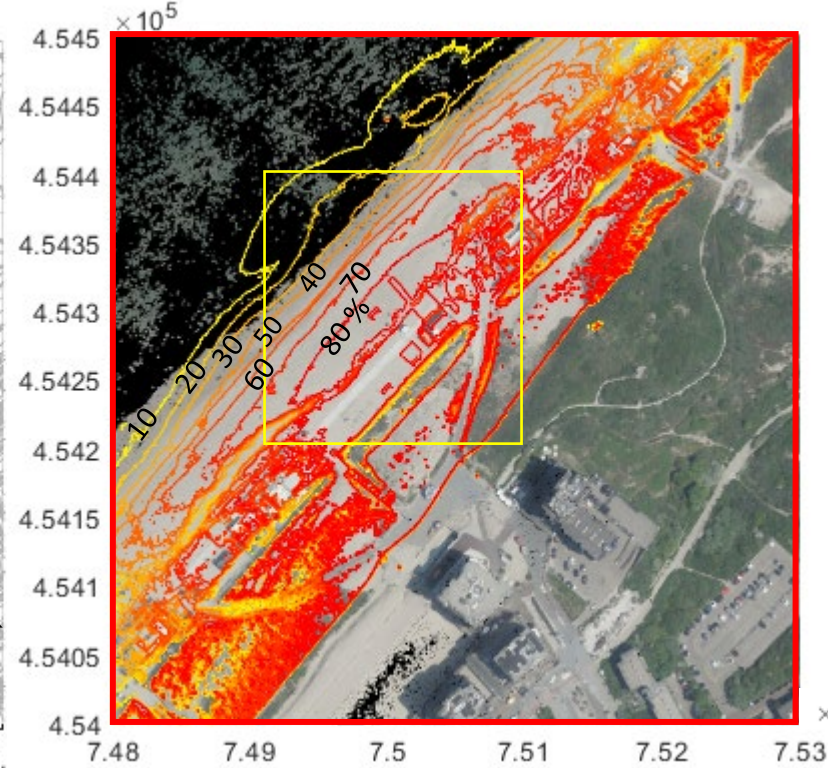
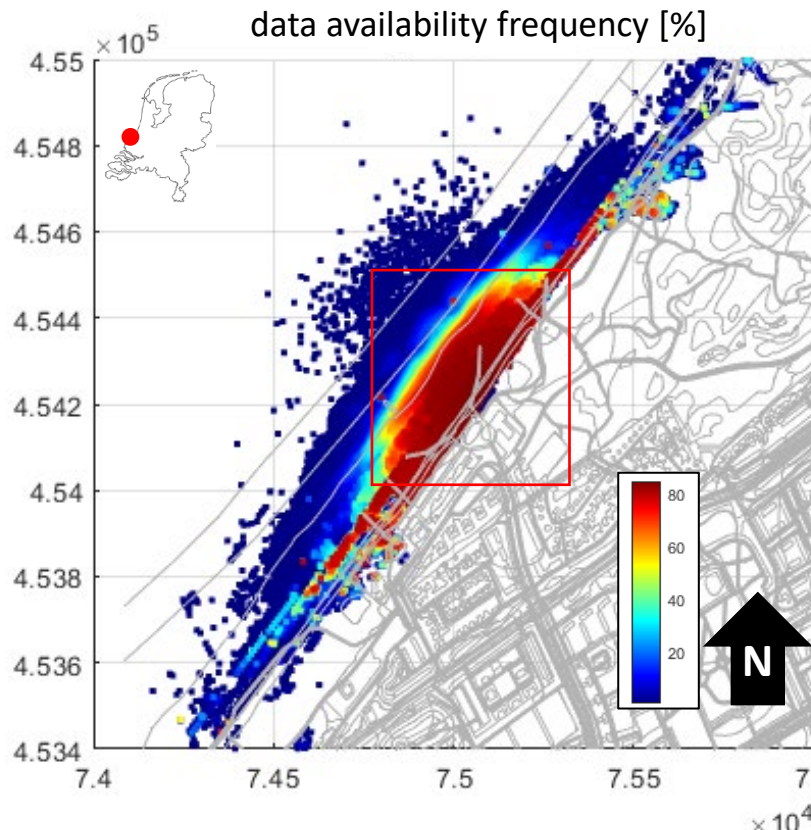
Experiment 2: spatio-temporal CoastScan data

■ data availability
□ data gap

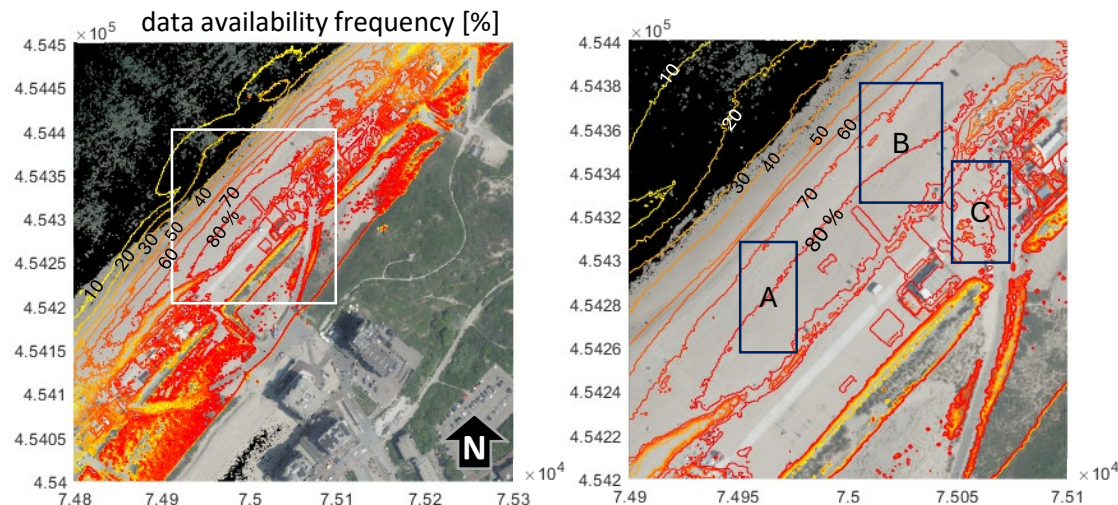
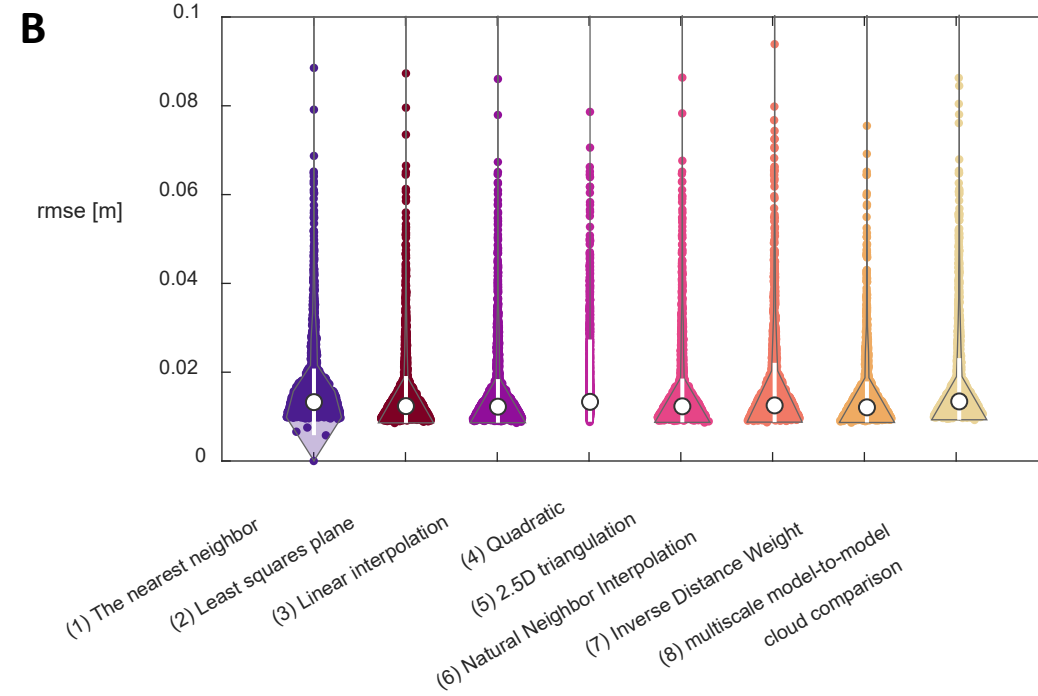
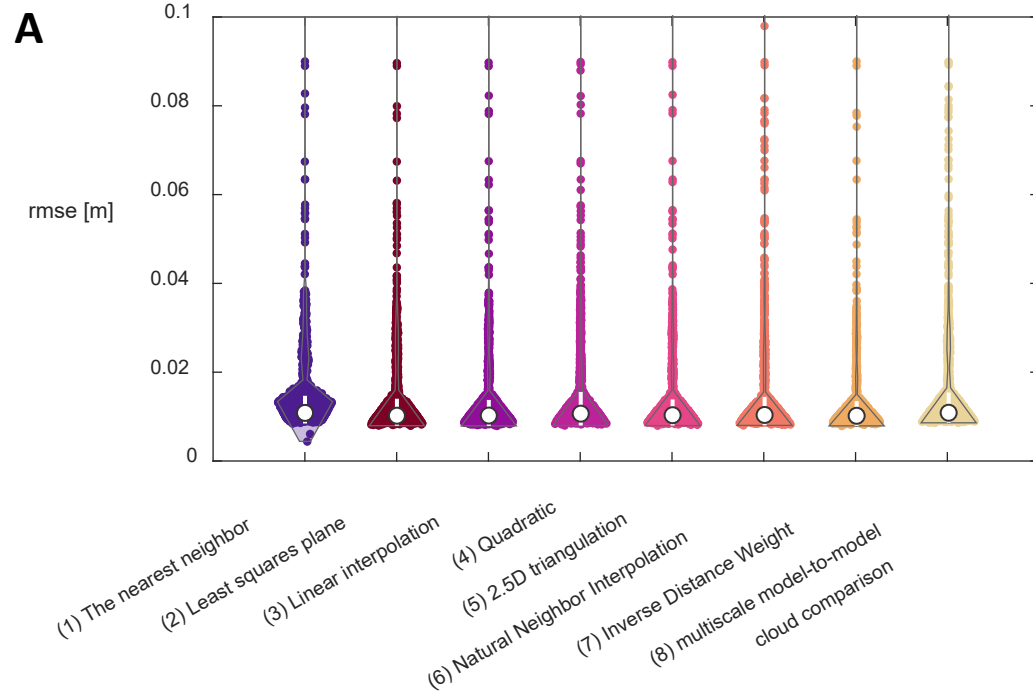


period 2017/01/01 to 2017/04/18
2,187 epochs

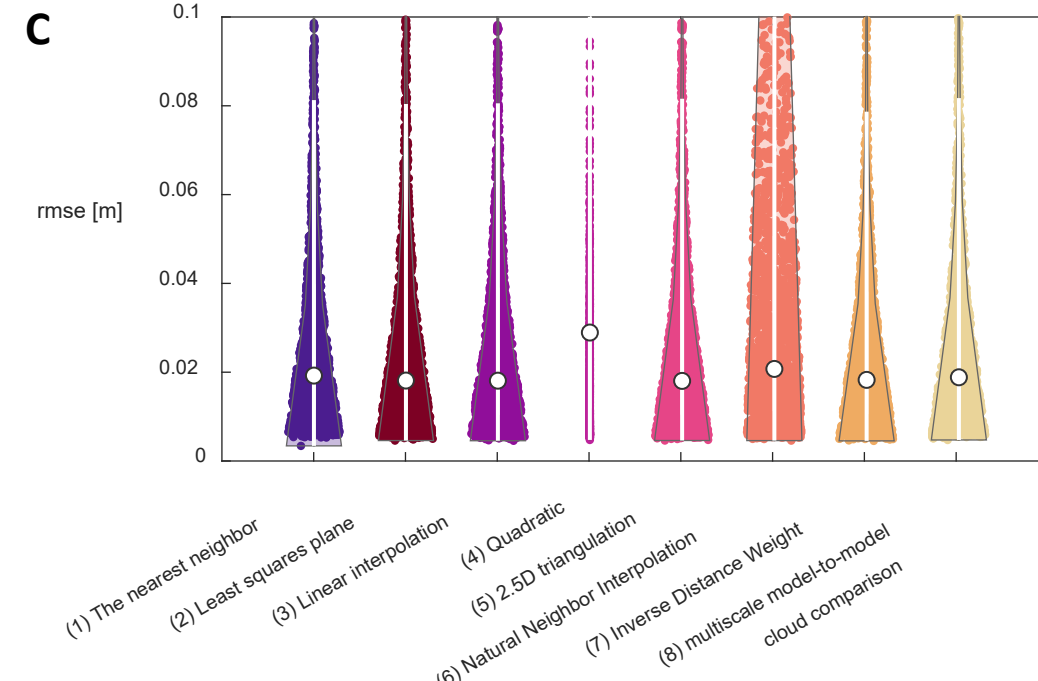
A) partially natural
B) intermediate
C) human-impacted



Error

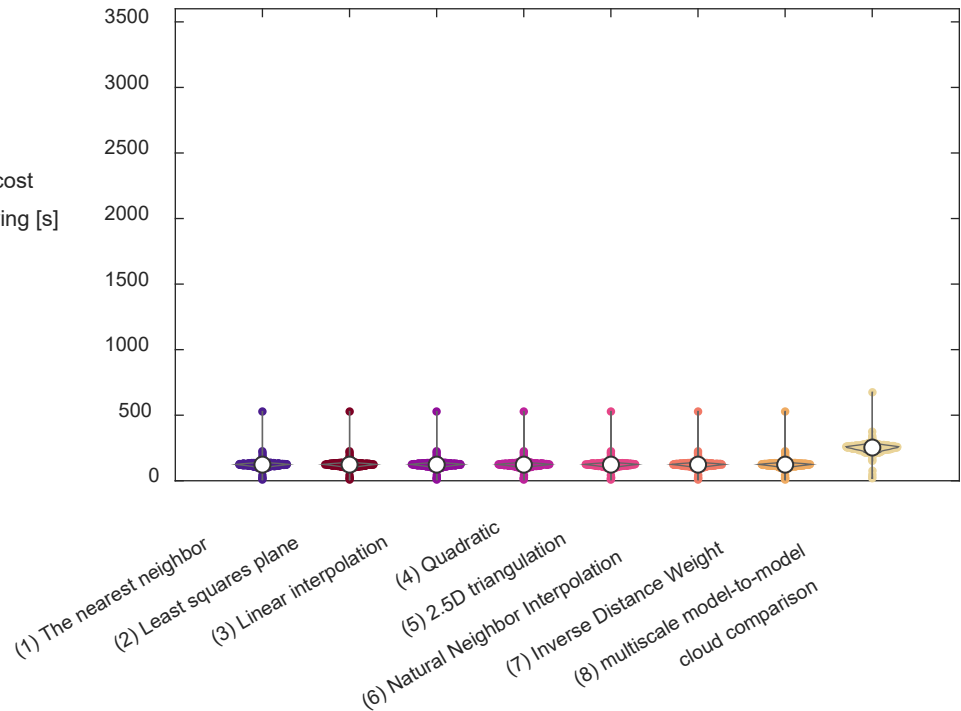


A) partially natural
B) intermediate
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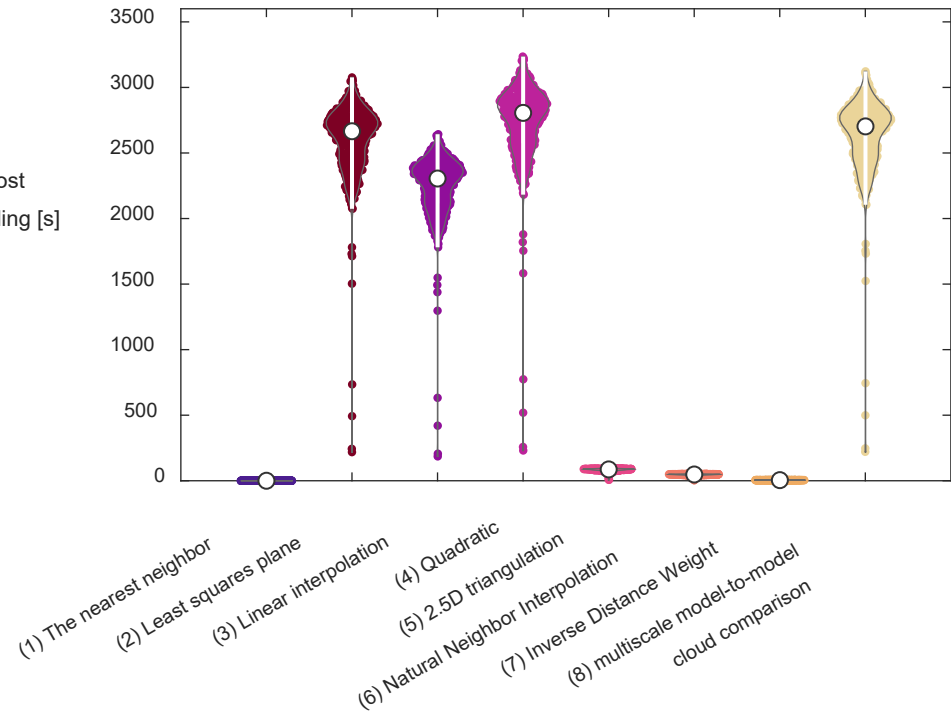


Time cost

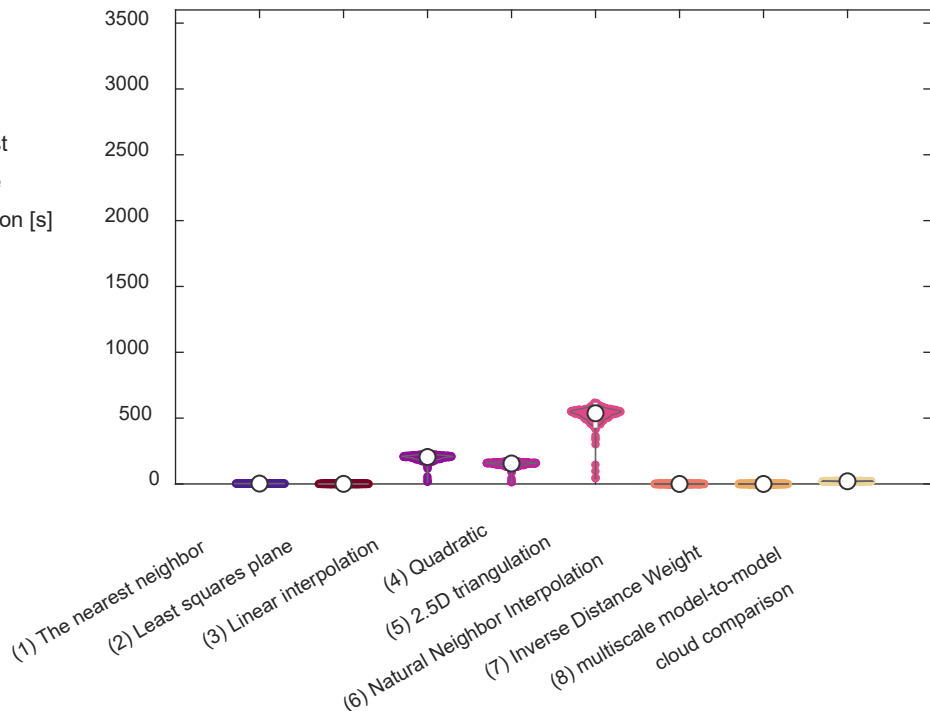
time cost
querying [s]



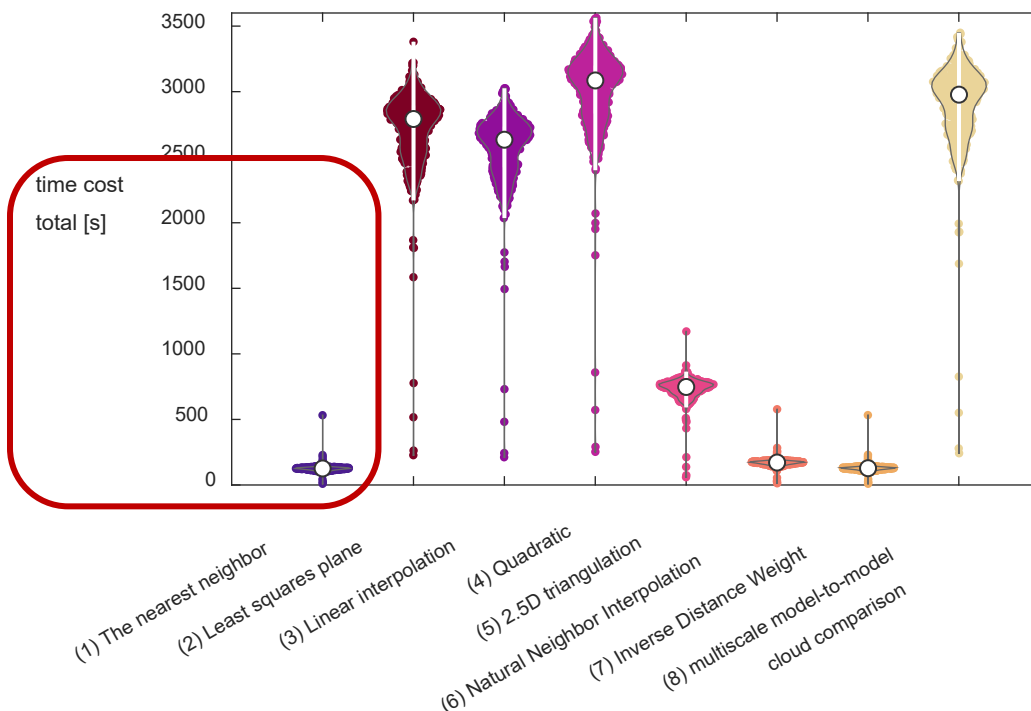
time cost
modelling [s]

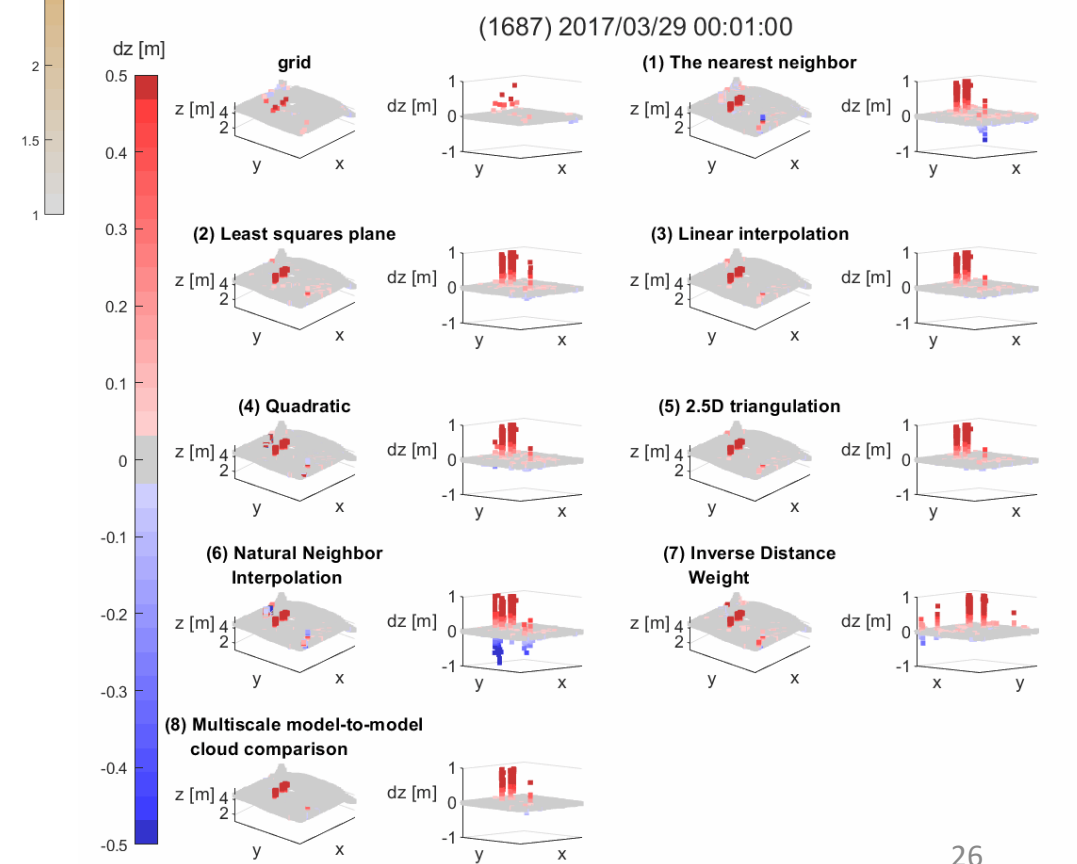
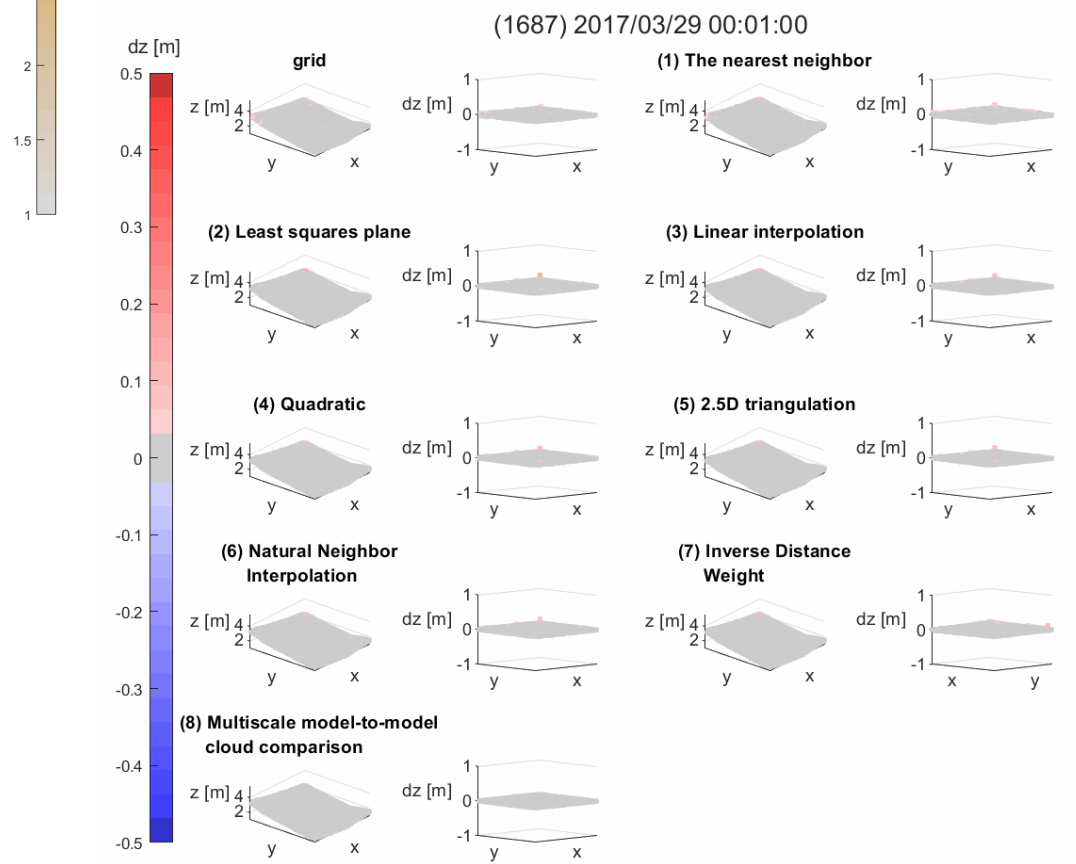
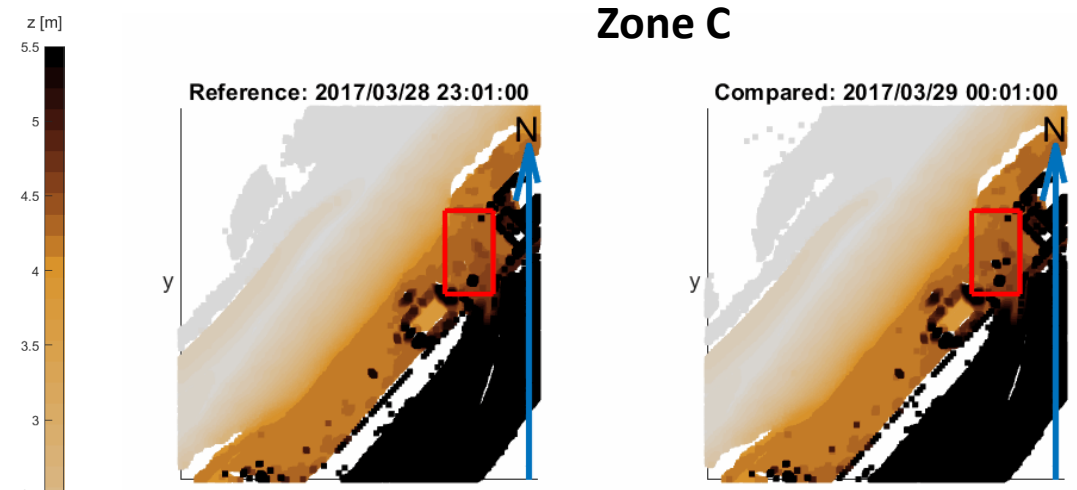
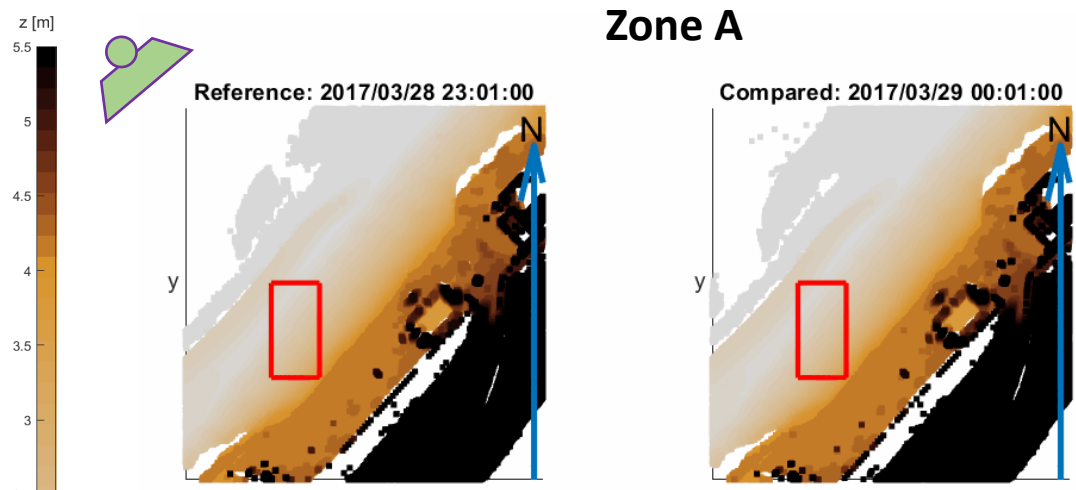


time cost
distance
calculation [s]



time cost
total [s]



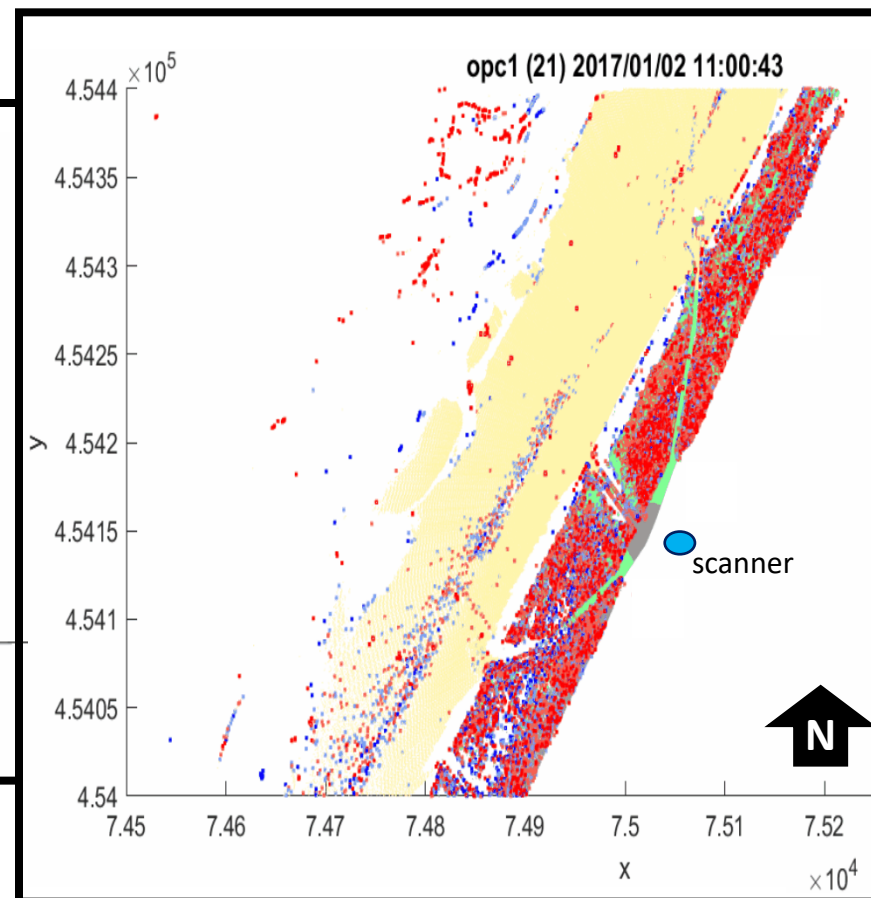
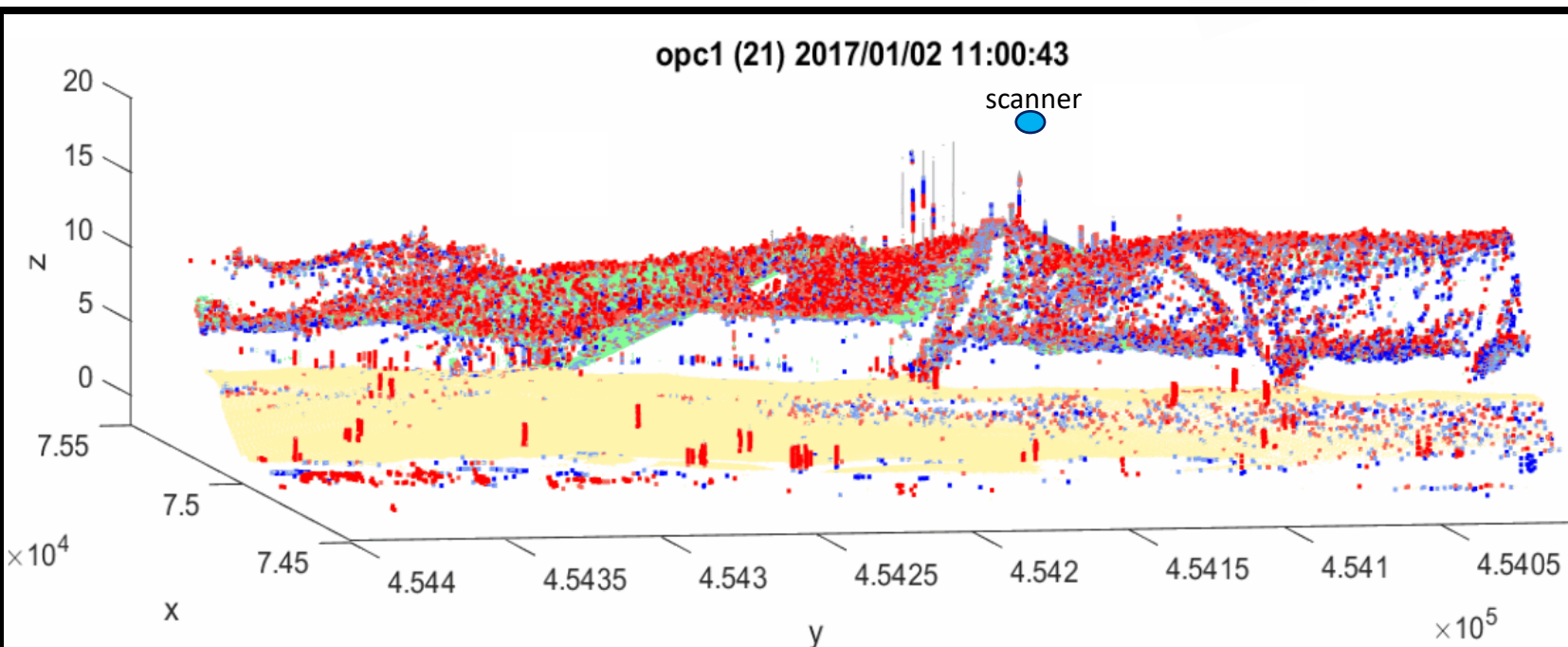
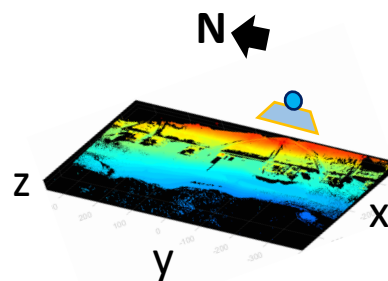




Spatio-temporal CoastScan data

Op1: 3D Nearest Neighbor(NN)

Changes are detected efficiently



- $dz > 0$
- $dz < 0$



3. Fast Space-Filling Curve-based point cloud change detection



SFC-based point cloud change detection

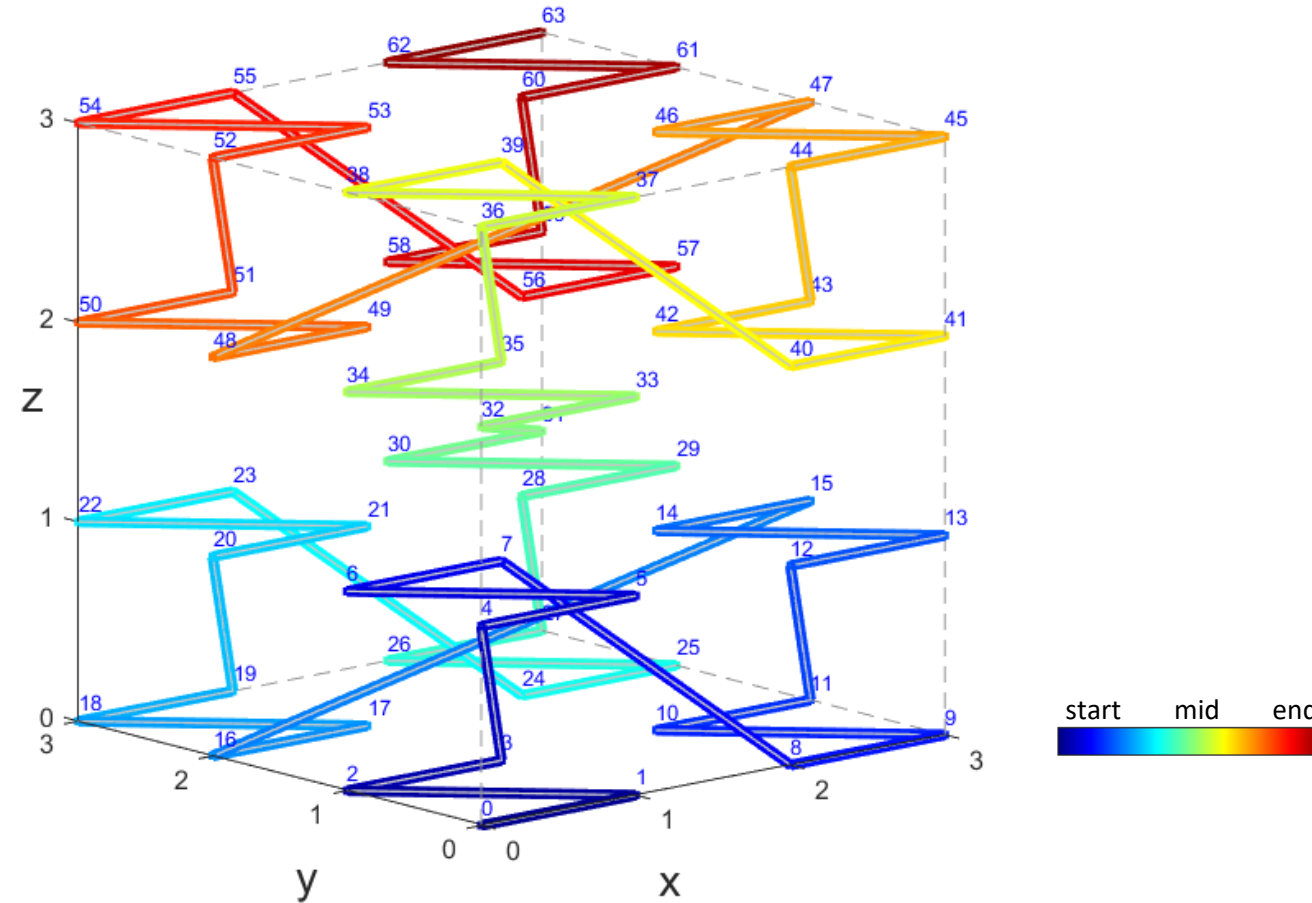
Once the optimal c2c distance method was identified, this phase aimed to develop an efficient implementation.

A 3D SFC-based point cloud method was developed to find the nearest neighbor, taking advantage of SFC's dimensionality reduction, enabling fast and efficient access to multidimensional data.

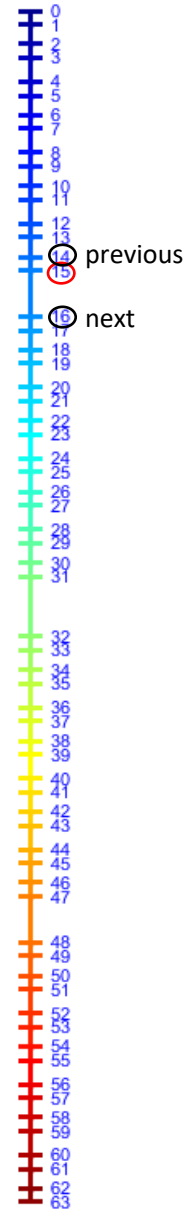
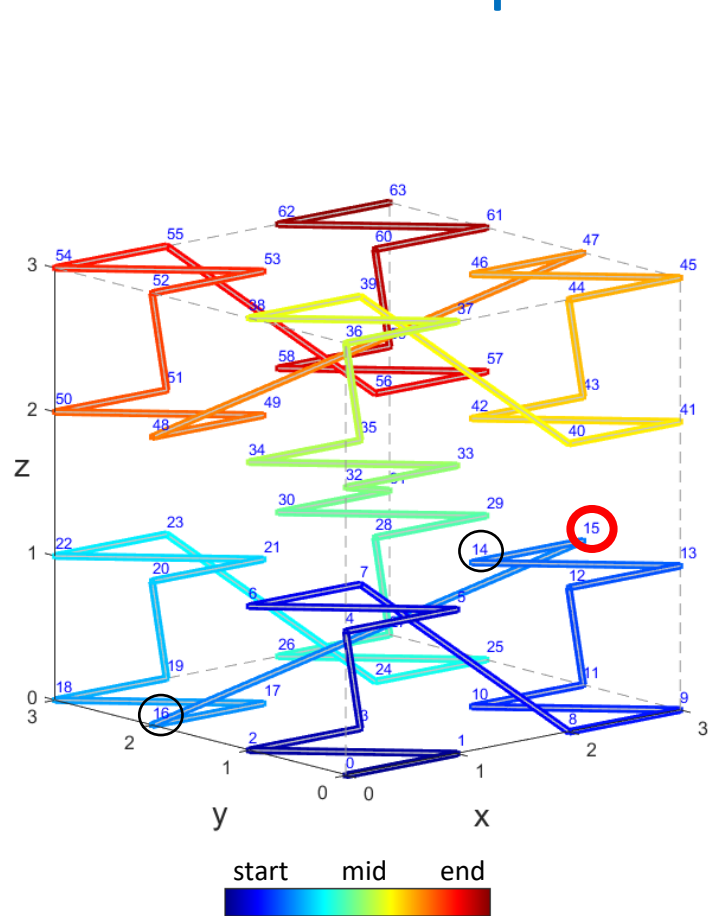
The kd-tree algorithm was the benchmark for finding the 'real' nearest neighbor (NN).

The results of the 3D SFC-based method were compared with those of the kd-tree algorithm.

SFC-based point cloud change detection

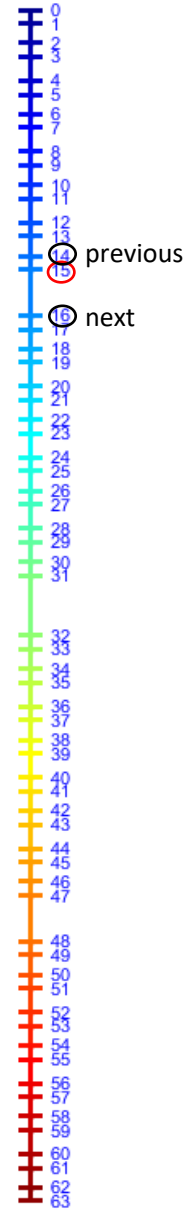
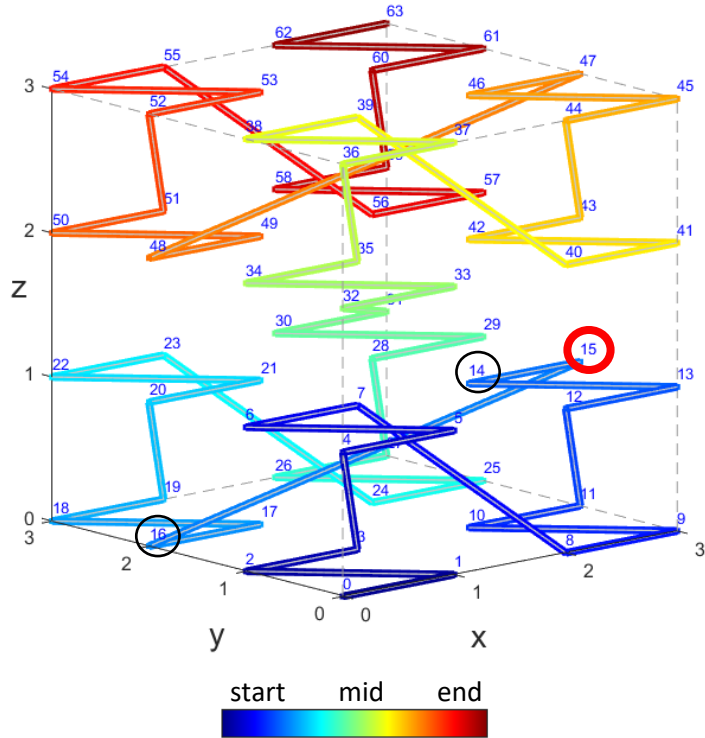


SFC-based point cloud change detection

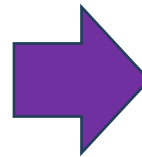


Euclidean distance gives us the closest neighbor

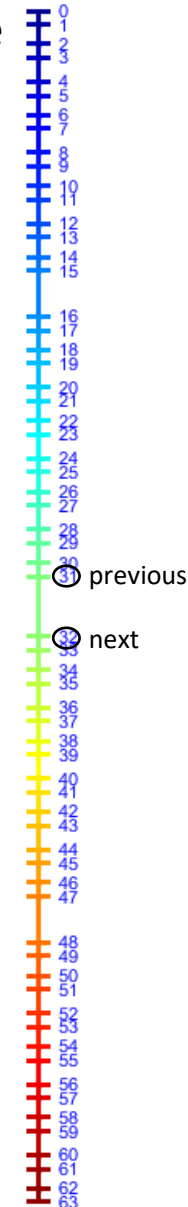
SFC-based point cloud change detection



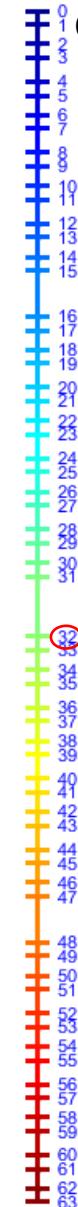
Euclidean distance gives us the closest neighbor



reference cloud



compared cloud

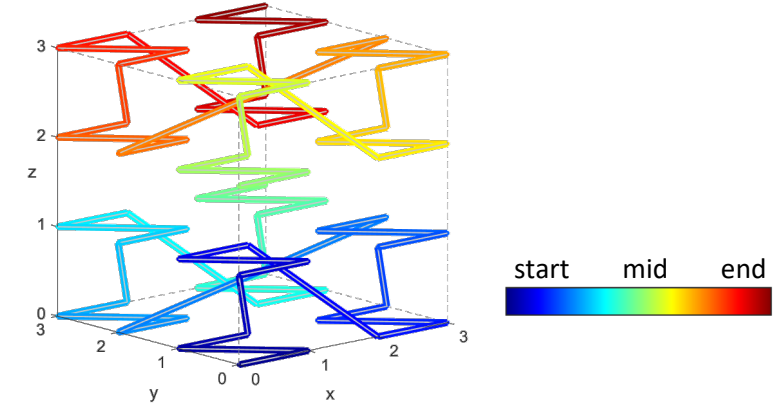


SFC-based point cloud change detection



- Preparation of point cloud, per epoch

1. scaling and offsetting
2. SFC key calculation
3. Sorting SFC key column

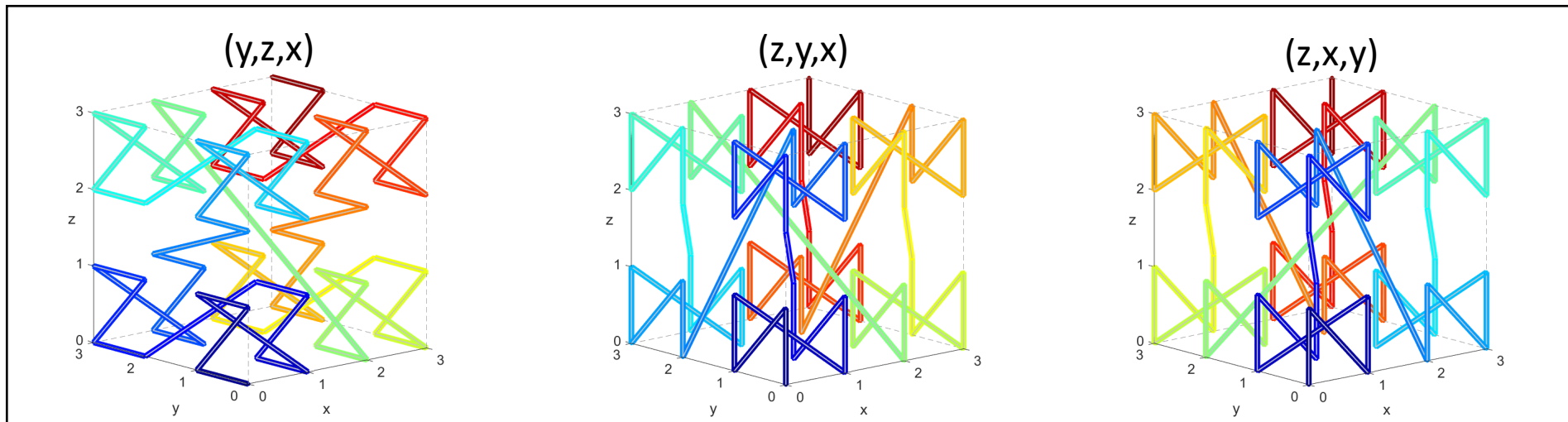
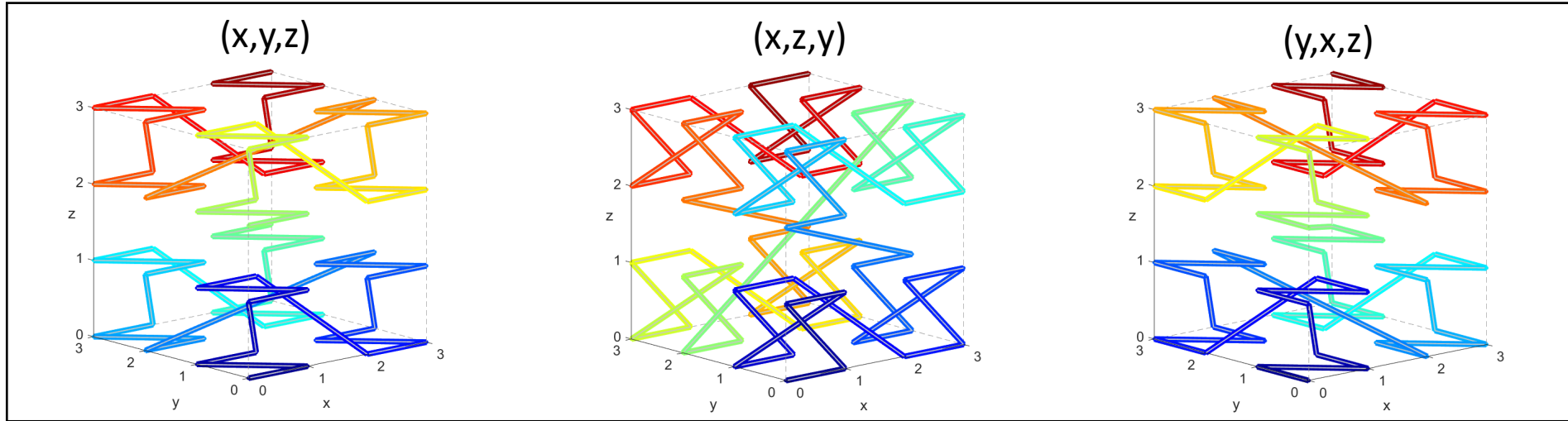


- Batch process NN distance calculation: for every point P in epoch 2:
 1. search the previous and next SFC key of the calculated SFC key in epoch 1
 2. decode and unscale these two SFC keys (previous and next)
 3. find the approximate nearest neighbor (NN) of the 2 candidates
 4. output d and dz

SFC-based point cloud change detection

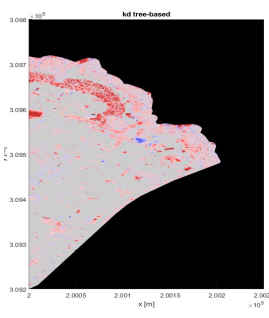


order for
SFC key
calculation

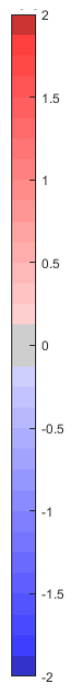
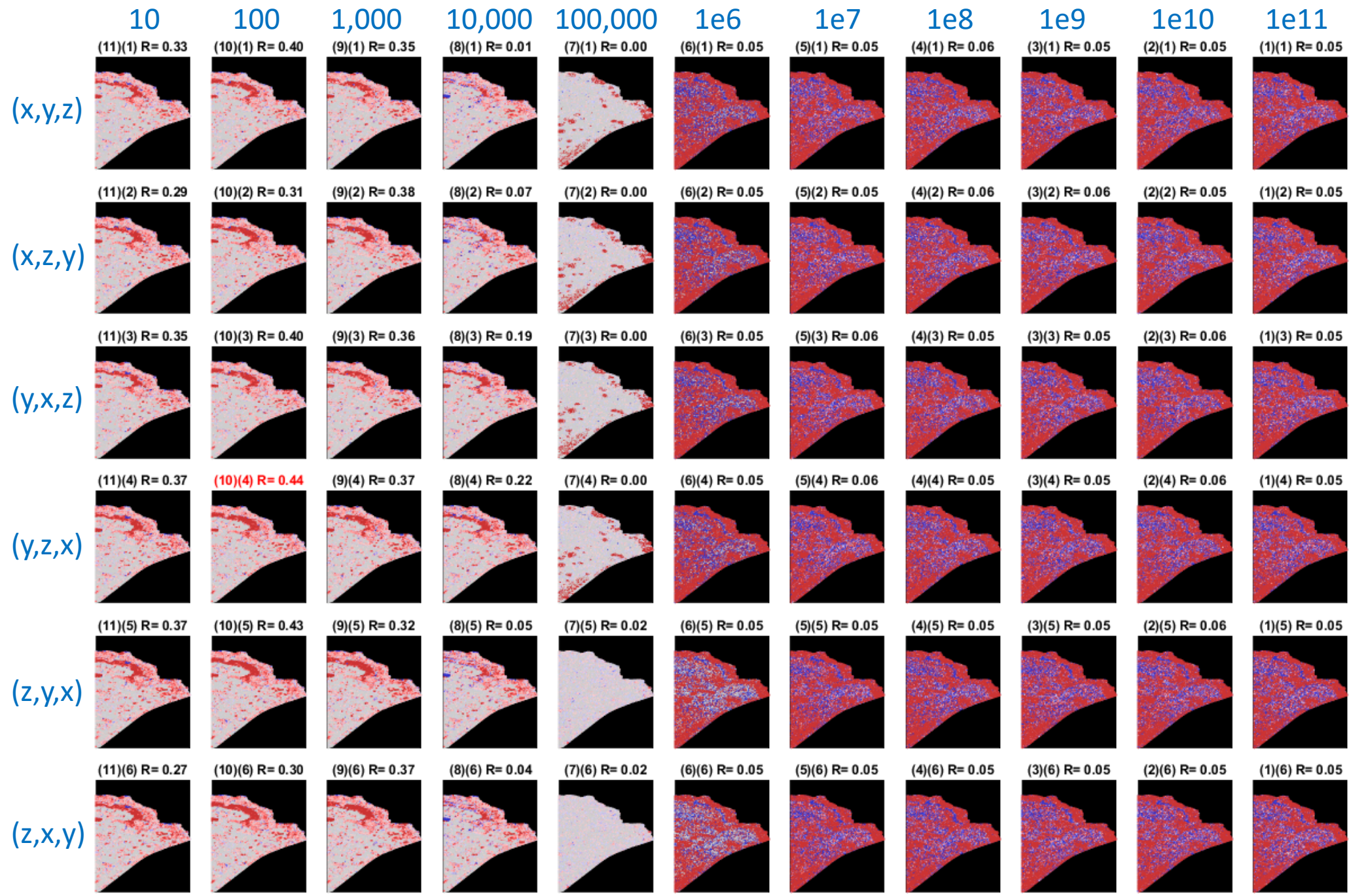


Does it
matter?

Comparison with kd-tree ('real' NN)



scale factor

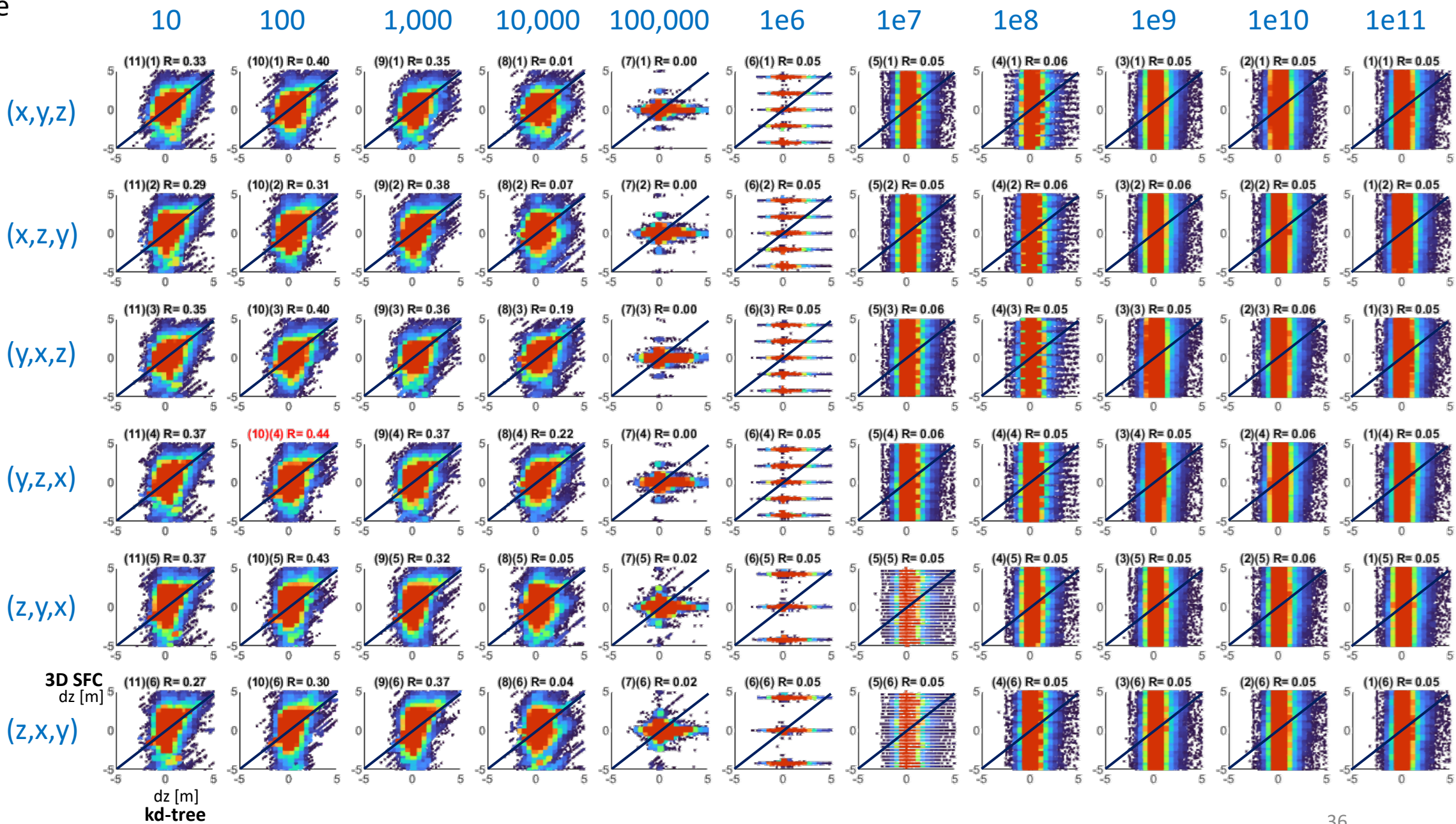


order for SFC key calculation

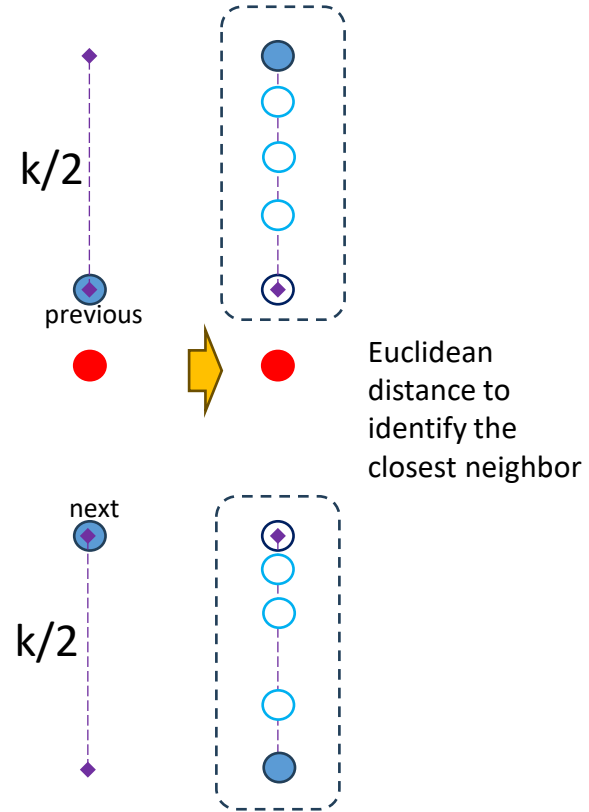
Comparison
with kd-tree
(‘real’ NN)

scale factor

order for
SFC key
calculation



scale factor
10



Order for SFC key calculation	k points (neighborhood)	R
(x,y,z)	8	0.3888
(x,y,z)	64	0.4722
(x,y,z)	512	0.6258
(x,y,z)	4,096	0.8101
(x,y,z)	32,768	0.9754
(x,y,z)	262,144	0.9985
(x,z,y)	8	0.3486
(x,z,y)	64	0.3957
(x,z,y)	512	0.4974
(x,z,y)	4,096	0.7848
(x,z,y)	32,768	0.9633
(x,z,y)	262,144	0.9888
(y,x,z)	8	0.4260
(y,x,z)	64	0.5130
(y,x,z)	512	0.6568
(y,x,z)	4,096	0.8185
(y,x,z)	32,768	0.9751
(y,x,z)	262,144	0.9990
(y,z,x)	8	0.4538
(y,z,x)	64	0.5395
(y,z,x)	512	0.6913
(y,z,x)	4,096	0.8772
(y,z,x)	32,768	0.9800
(y,z,x)	262,144	0.9990
(z,y,x)	8	0.4178
(z,y,x)	64	0.5598
(z,y,x)	512	0.6862
(z,y,x)	4,096	0.7971
(z,y,x)	32,768	0.9808
(z,y,x)	262,144	0.9990
(z,x,y)	8	0.3147
(z,x,y)	64	0.4099
(z,x,y)	512	0.5460
(z,x,y)	4,096	0.7813
(z,x,y)	32,768	0.9739
(z,x,y)	262,144	0.9910

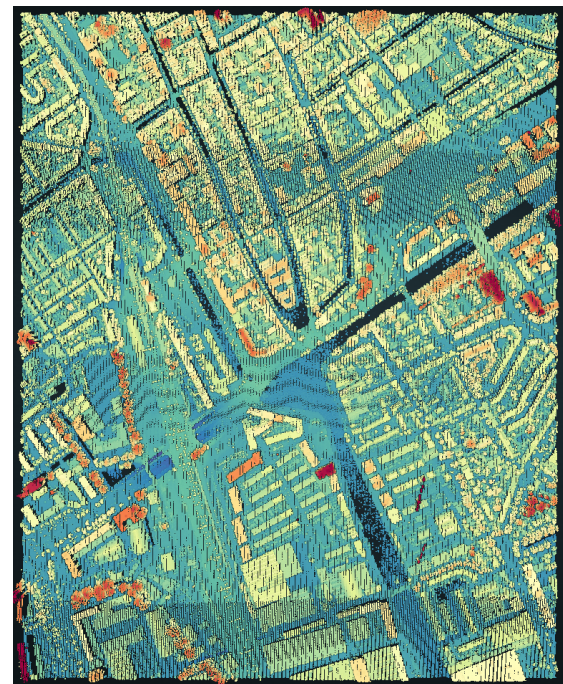
scale factor
100

Comparison with kd-tree ('real' NN)

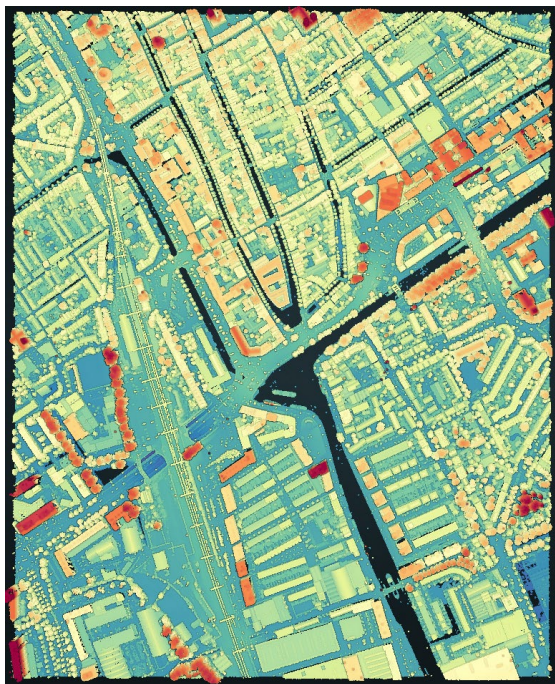
Order for SFC key calculation	k points (neighborhood)	R
(x,y,z)	8	0.4460
(x,y,z)	64	0.5379
(x,y,z)	512	0.6229
(x,y,z)	4,096	0.8697
(x,y,z)	32,768	0.9655
(x,y,z)	262,144	0.9969
(x,z,y)	8	0.3566
(x,z,y)	64	0.4077
(x,z,y)	512	0.5022
(x,z,y)	4,096	0.6299
(x,z,y)	32,768	0.9850
(x,z,y)	262,144	0.9969
(y,x,z)	8	0.4391
(y,x,z)	64	0.5549
(y,x,z)	512	0.7004
(y,x,z)	4,096	0.8731
(y,x,z)	32,768	0.9756
(y,x,z)	262,144	0.9964
(y,z,x)	8	0.4650
(y,z,x)	64	0.5949
(y,z,x)	512	0.8029
(y,z,x)	4,096	0.9243
(y,z,x)	32,768	0.9820
(y,z,x)	262,144	0.9964
(z,y,x)	8	0.5030
(z,y,x)	64	0.5923
(z,y,x)	512	0.7316
(z,y,x)	4,096	0.9303
(z,y,x)	32,768	0.9824
(z,y,x)	262,144	0.9966
(z,x,y)	8	0.3510
(z,x,y)	64	0.4325
(z,x,y)	512	0.5148
(z,x,y)	4,096	0.8740
(z,x,y)	32,768	0.9862
(z,x,y)	262,144	0.9969

Delft

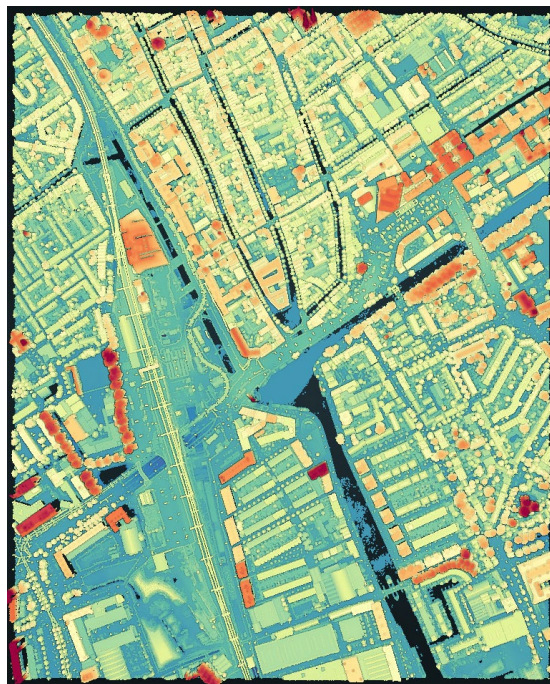
AHN1



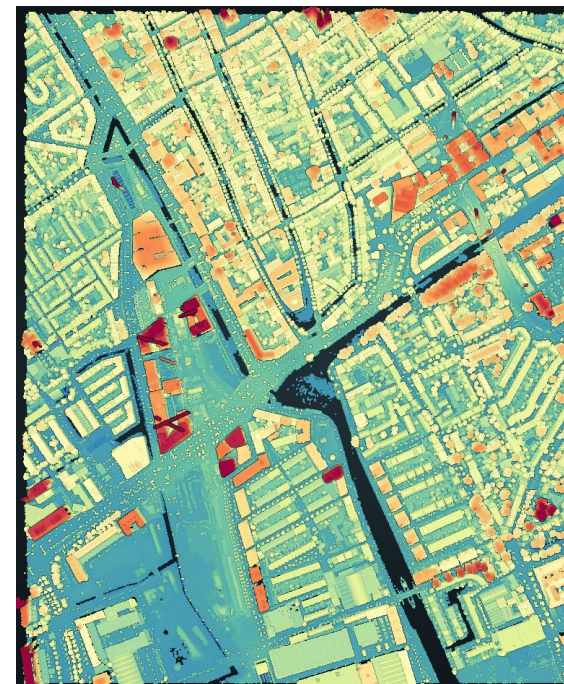
AHN2



AHN3



AHN4



up



down

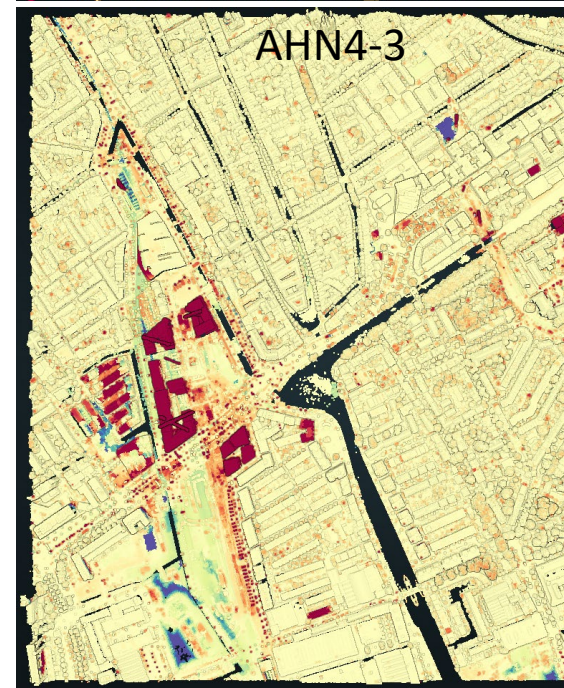
AHN2-1



AHN3-2



AHN4-3



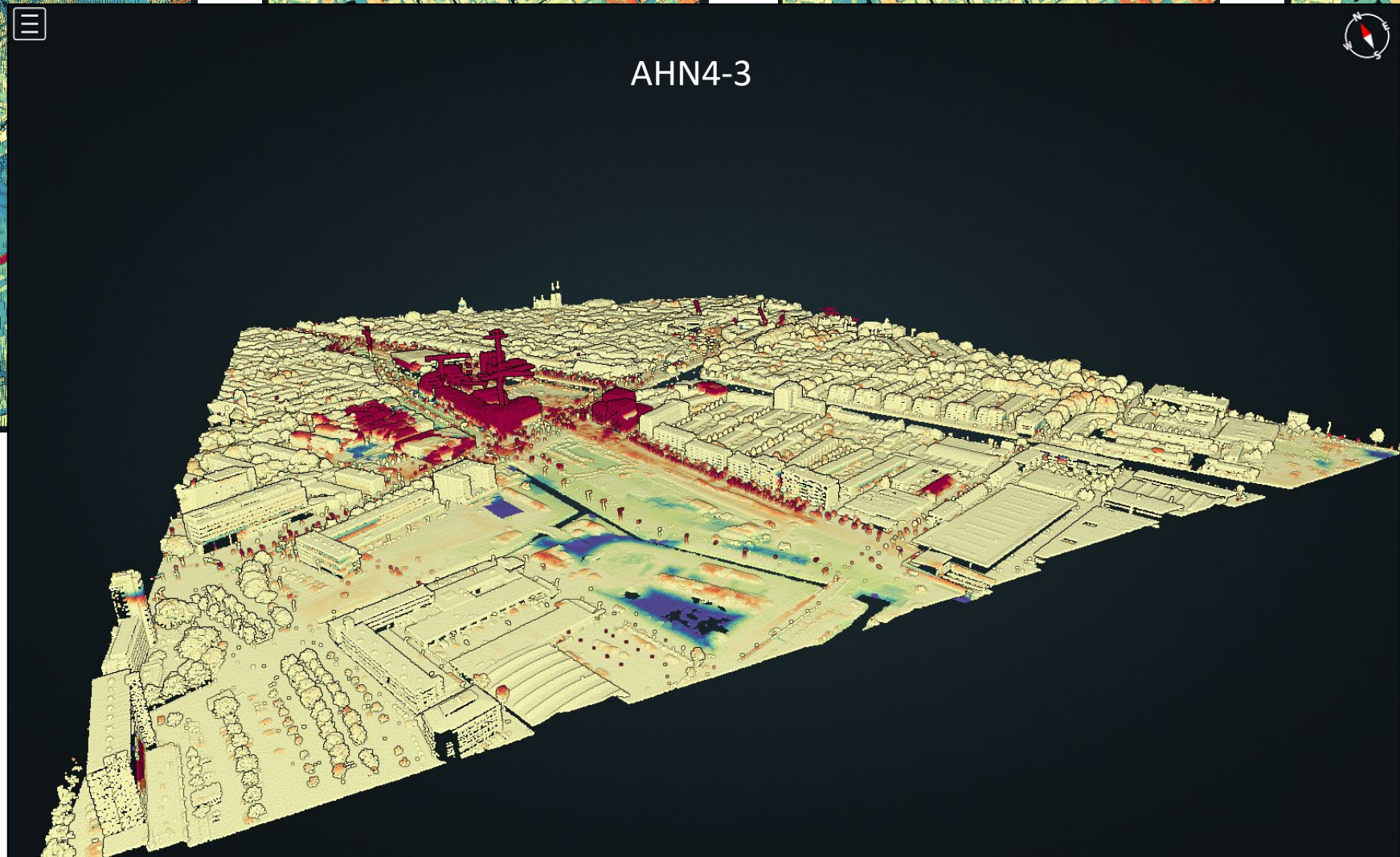
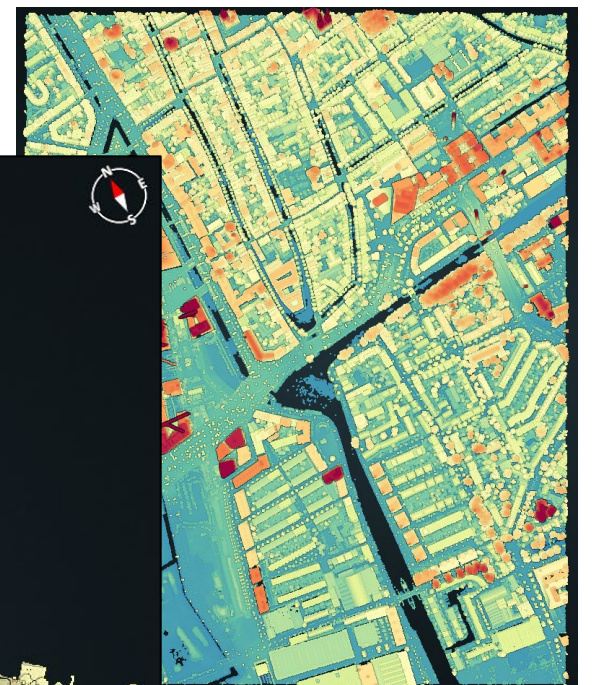
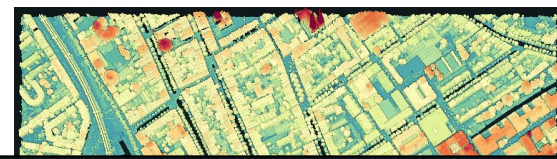
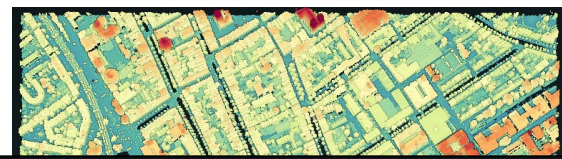
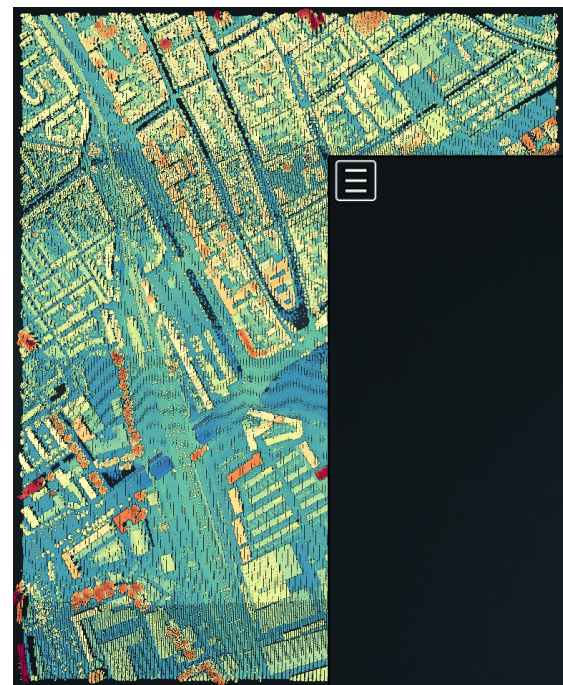
Delft

AHN1

AHN2

AHN3

AHN4

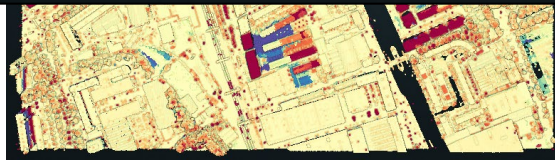


AHN4-3

up

 down

AHN4-3

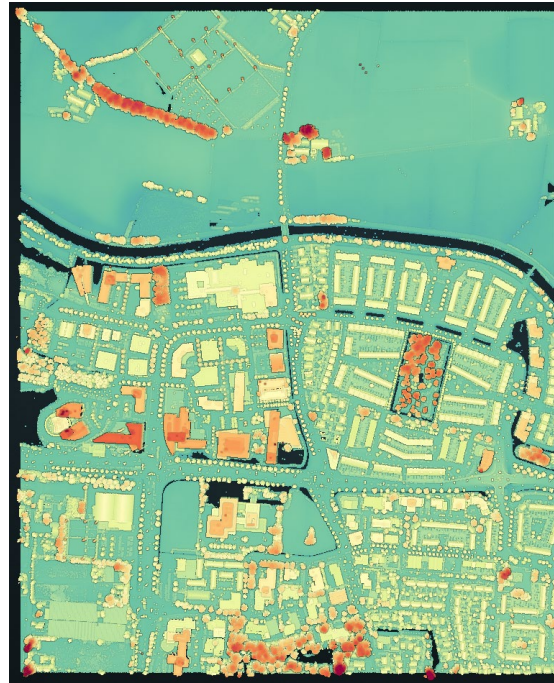


Leusden

AHN1



AHN2



AHN3



AHN4



AHN2-1



AHN3-2



AHN4-3



Leusden

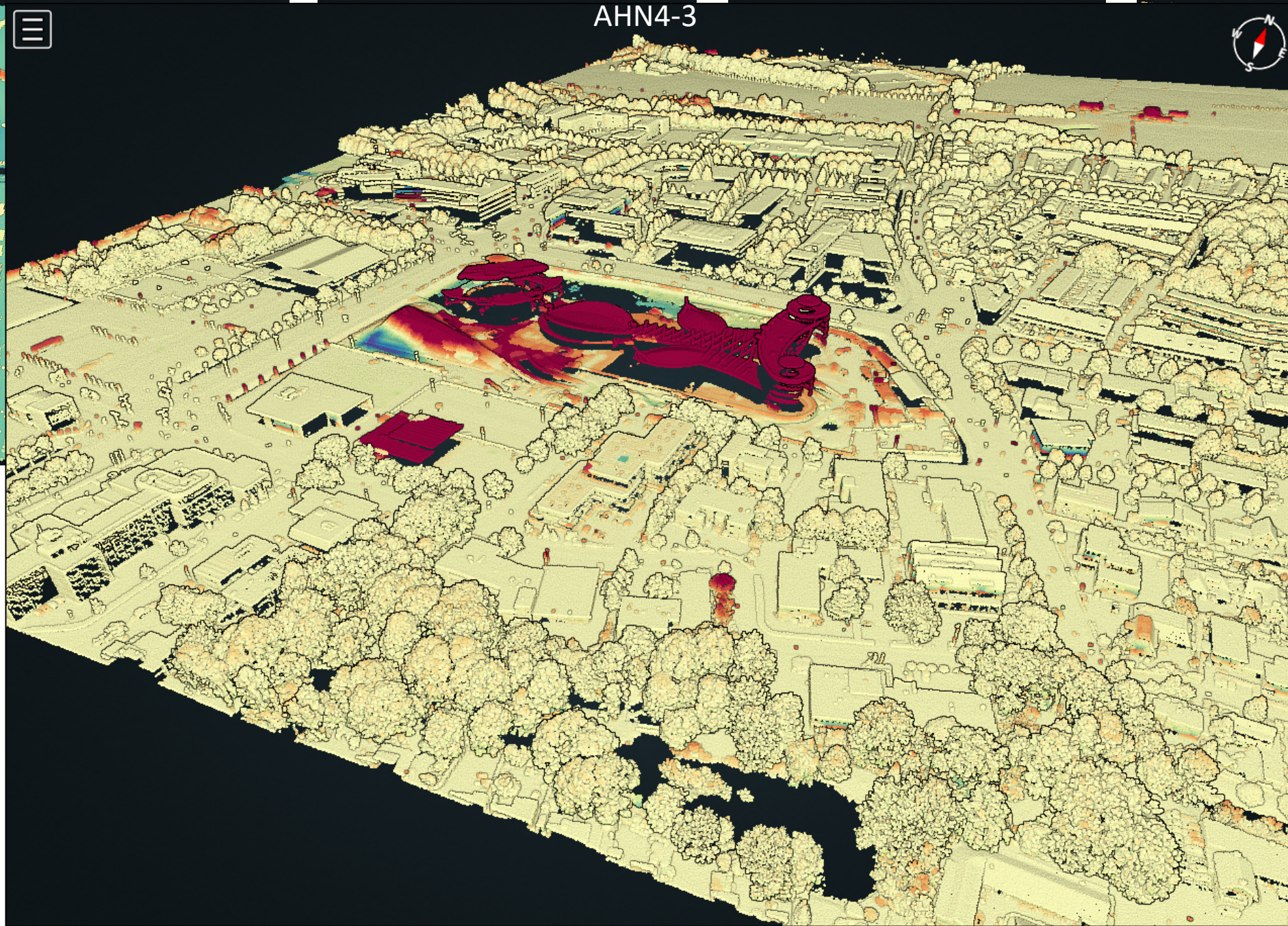
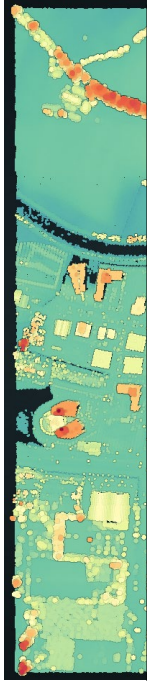
AHN1

AHN2

AHN3

AHN4

AHN4-3



Woerden

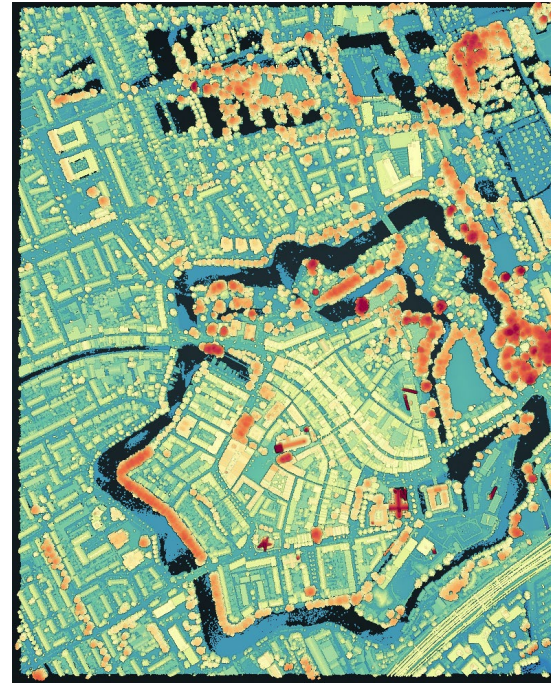
AHN1



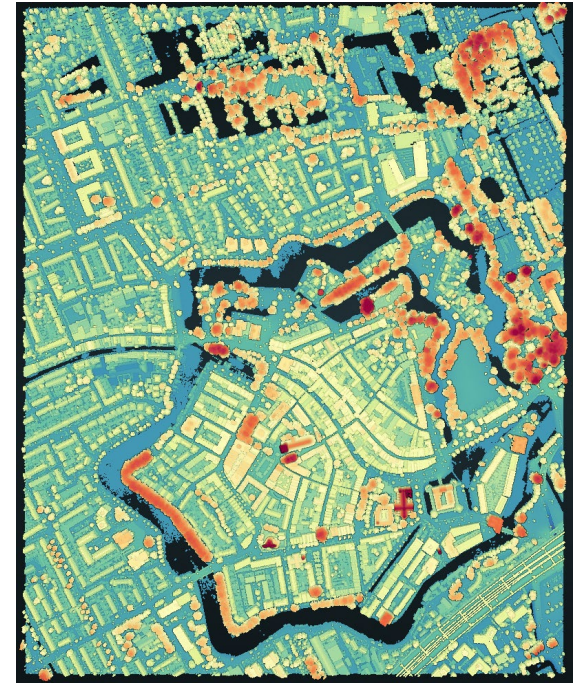
AHN2



AHN3



AHN4



AHN2-1



AHN3-2



AHN4-3



Woerden

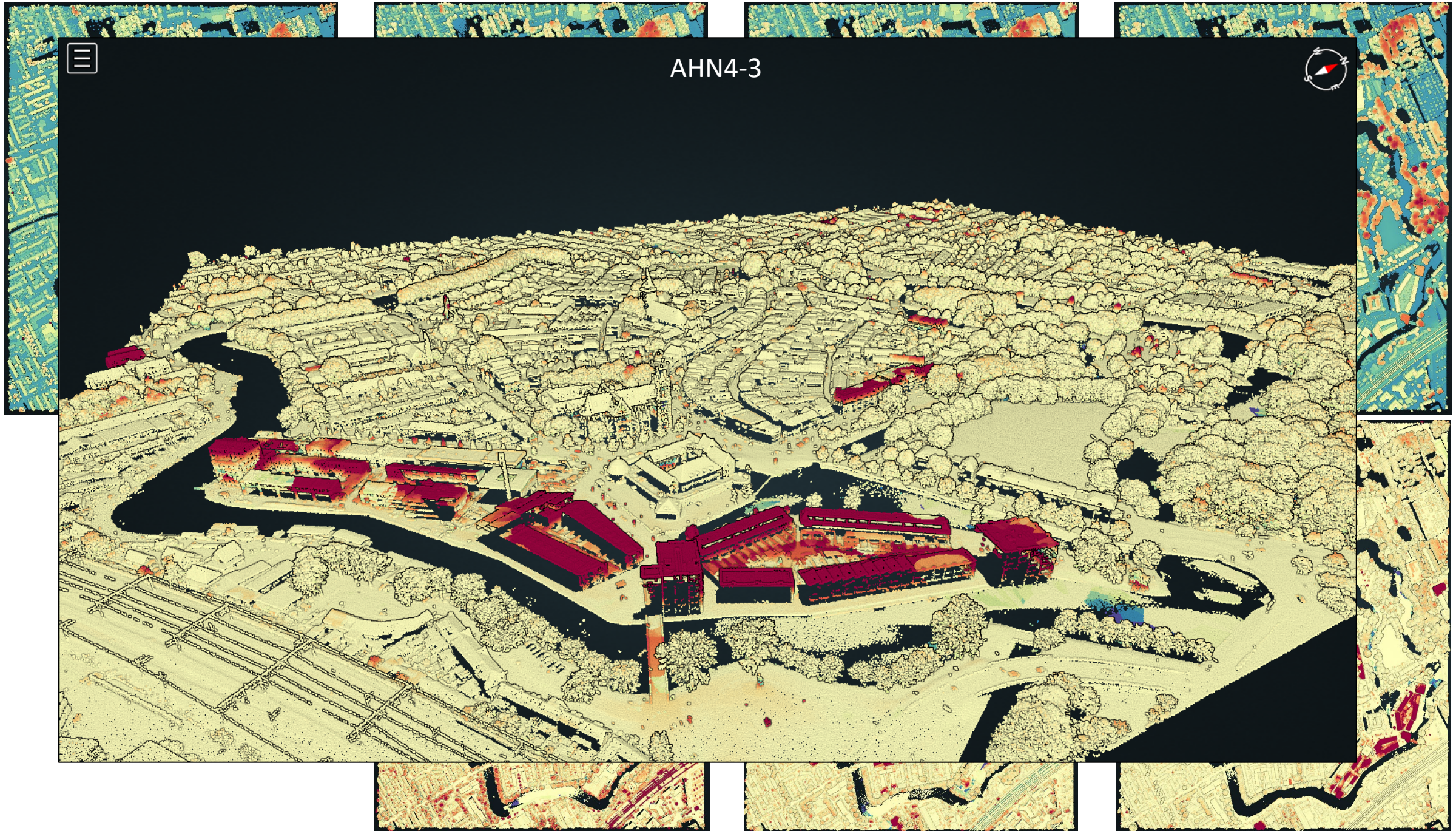
AHN1

AHN2

AHN3

AHN4

AHN4-3





4 Conclusions

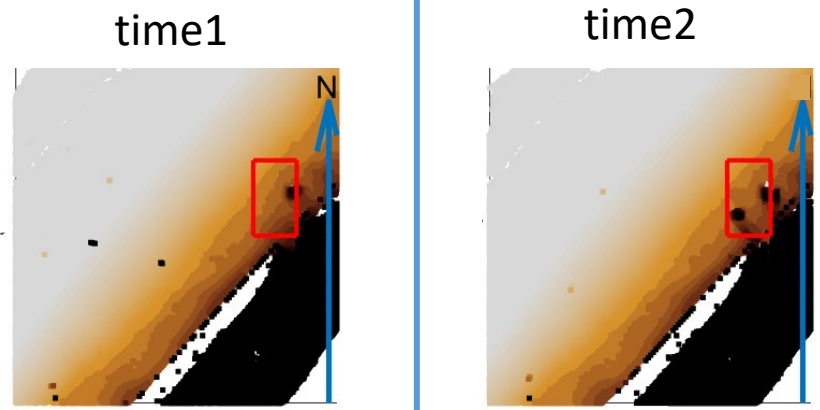
- 8 methods for change detection analysis with spatiotemporal point clouds.
- Nearest Neighbor is the best candidate for the c2c distance calculation.
- Development of an SFC-based c2c distance calculation method for fast implementation in massive point clouds.



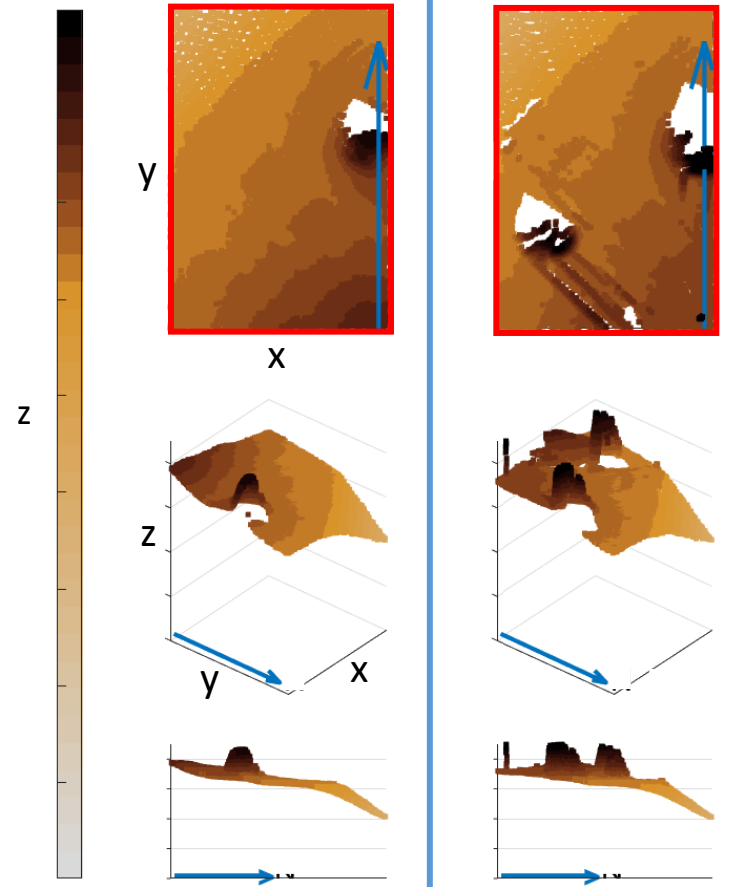
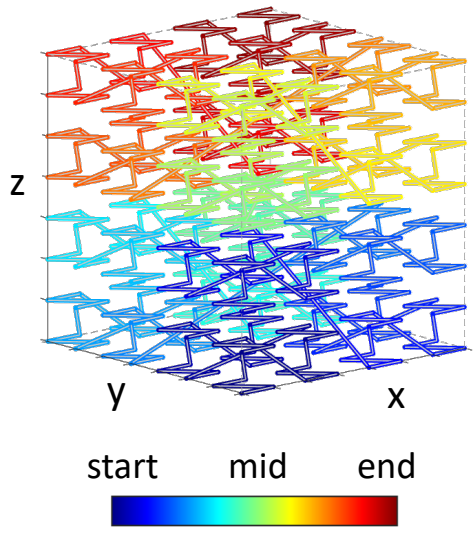
5 Future work

- Finishing calculation of changes for AHN 1 to 4
- Visualizing the changes in Potree viewer
- Publications:
 - Cloud-to-cloud method for change detection in spatio-temporal point clouds: CoastScan data
 - Fast Space Filling Curve-based point cloud change detection: development and implementation

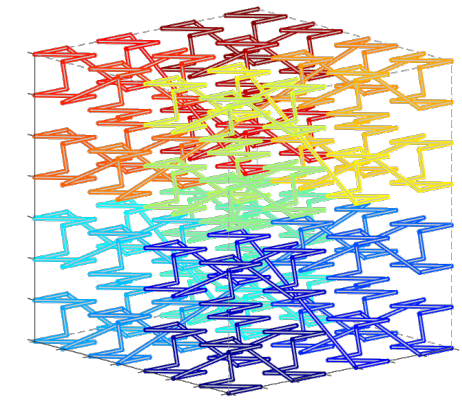
***Intelligent method for monitoring and predicting changes in spatio-temporal point cloud data**



3D_space-filling_curve1



3D_space-filling_curve2



*intelligent indicates using ML algorithms to segment, characterise, and predict changes

Fast SFC-based
point cloud
change detection

Vitali Diaz Mercado
v.diazmercado@tudelft.nl



Thanks!

