

Point cloud watermarking

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- Two task:
 - ▶ **Embedding** (inserting) some information (watermark) into the data;
 - ▶ **Detection/Extraction** of this information from the data;
 - ▶ (secret) **key** determines locations of this embedding information;
- Primarily, the watermarking is developed for LiDAR data (LAS files):
 - ▶ geo-referenced points; most important are the the coordinates of the points which represents the real terrains; other attributes.
- Extended to the 3D point clouds:
 - ▶ positioned, oriented and scaled in many ways;

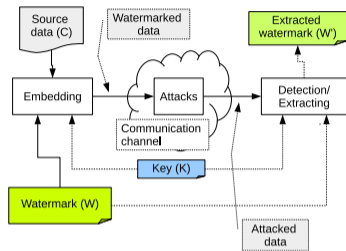


Figure 1: Watermarking process

- The coordinates of the point were chosen for watermarking the point clouds:
 - ▶ They are slightly modified (unnoticeable) but they need to be still resistible against various other data modifications (attacks);
- The data (3D point cloud) can be modified in various ways:
 - ▶ Destroy information (watermark), embedded into the data \Rightarrow thus, the data are attacked;
- The following attacks can occur:
 - ▶ Affine transformations (scaling, rotating, translating), simplification (voxelization), noise, random point removal, cropping, points rearrangement etc.;
 - ▶ For geo-referenced LiDAR data, the affine transformations were not considered;

- The watermark **embedding** procedure:
 1. Defining an array of marker positions (where the particular watermark bit would be embedded),
 2. Inserting a watermark bit within the circular area around the marker positions, and
 3. Modifying the coordinates of the points within this circular area.
- The watermark **extraction** procedure is very similar:
 1. Defining an array of marker positions (where particular watermark bit was embedded), and
 2. Extracting a watermark bit from these circular areas around marker positions.

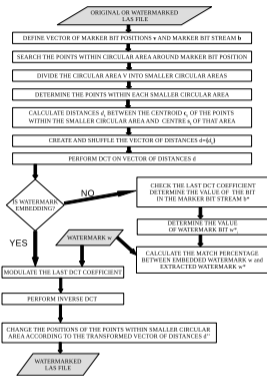


Figure 2: Embedding/Extraction diagram

- A watermark w :
 - ▶ $w = \{w_0, w_1, \dots, w_{M-1}\}, \quad w_i \in \{0, 1\}$
- A watermark bit stream b :
 - ▶ $b = \{b_0, b_1, \dots, b_j, \dots, b_{N-1}\}$
 - ▶ where: $b_j = w_{(j \bmod M)}$
 - ▶ thus, the watermark is embedded multiple times ($N = MX$);

- For each b_i : a marker positioned v_i is randomly defined and the corresponding circular area (within radius r_{out});
- Next, the smaller circular areas (seeds s_k) are determined applying the sunflower seed algorithm;

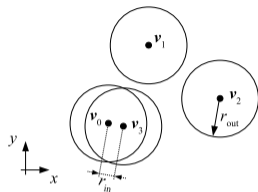


Figure 3: Marker bit positions

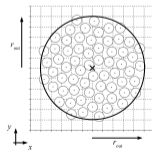


Figure 4: The sunflower seed algorithm

1. Determine all points within the smaller circular area (seed) by the approximate nearest neighbor search (FLANN);
2. Determine the geometric center c_k of all these points, and calculate distance d_k between s_k and c_k ;
3. The distances d_k for each marker position v_k was obtained; further shuffled by the Mersenne twister pseudo-random generator;
4. d_k are used as input for the DCT (Discrete cosine transformation);
5. Modify the last DCT coefficient (0 or 1);
6. Perform IDCT, thus, modified d'_k are obtained $\Rightarrow c_k$ and the points within the smaller circular area are changed;

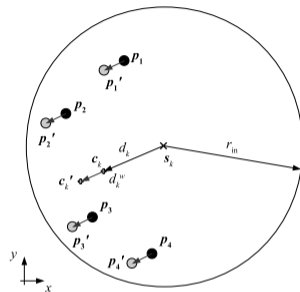


Figure 5: Modifying the points' coordinates within small circular area

- 1-4. The steps identical to the steps in the watermark embedding \Rightarrow the distances d_k for each marker position are determined, and the DCT is performed;
 5. The last DCT coefficient is checked (the bit value 0 or 1 is determined);
 6. Two counters cw_i^+ and cw_i^- are defined for each watermark bit w_i ;
 - ▶ they counts how many times occur 1 or 0 for w_i within the watermark bit stream b ;
 - ▶ if $cw_i^+ > cw_i^- \Rightarrow w_i = 1$ otherwise $w_i = 0$;
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- In general, 3D point cloud can be positioned, scaled, and oriented in different ways;
 - Thus, before extraction the point clouds should be aligned (registered);
 - Unfortunately, the most point cloud registration techniques works only for the point clouds with the same scale;

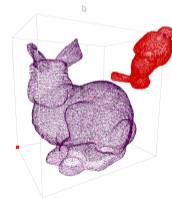


Figure 6: Point cloud registration issue

1. The two convex hulls H^W (watermarked and possibly attacked cloud) and H^I (original cloud) \Rightarrow two arrays of triangles T_w and T_i are obtained and sorted according to the triangle area in descending order;
2. For each triangle from T_w , the matching triangles from the T_I are determined (can be more than one):
 - ▶ the ratios between the triangle sides are used (a/c and b/c);
3. Next, the best matching triangles are determined:

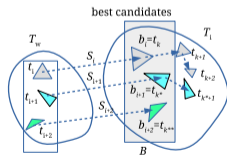


Figure 7: Matching triangles

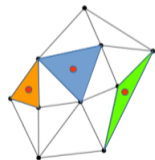
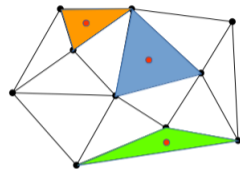


Figure 8: 3D Convex hulls

3. The best matching triangles are determined (cont.):
 - ▶ In general, the largest triangles from T^W with the similar ratios match with the largest triangles from T^I ;
 - ▶ ☹ Any modifications of the watermarked point cloud can damage the convex hull;
4. The scaling factor s is determined applying the following steps:

- ① The scaling factors (b_i) are sorted in the ascending order;
- ② The scaling factor b_i are joined together within $[-\xi, \xi]$, and the scaling average s_i and the number of them is determined as the frequency f_i ;
- ③ The final step: the scale factors s_i with high frequencies f_i that are close enough \Rightarrow they are joined together into the final scale $s = s_k$ with the highest frequency s_k ;

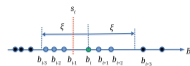


Figure 9: Array of scaling factors S_i binds the best candidates (b_i, b_{i+1} , and b_{i+2}) for scale estimation

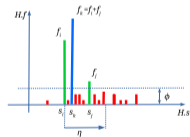
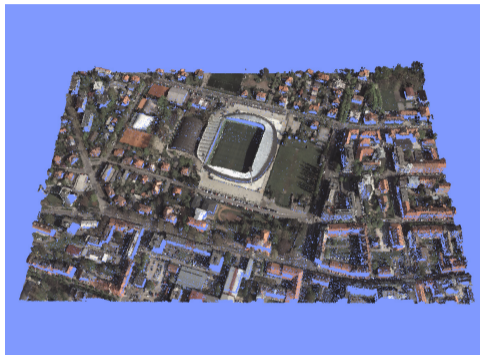
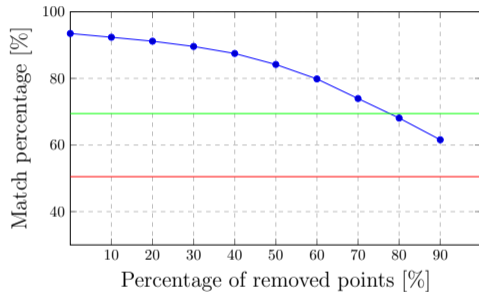


Figure 10: Determining scale factor s_i and f_i ; s_i in an average of $f_i = 6$ scale factors in this example

5. Two auxiliary point clouds are built from the centres of best matching triangles of the convex-hulls, respectively (marked as red in Figure 8); much less points;
6. The first auxiliary point clouds (from watermarked point cloud) is scaled by s and the rigid transformation matrix (i.e rotation and translation) is determined and applied to the watermarked cloud;
7. This coarse (initially) aligned point cloud is then aligned by well known algorithm iterative closest point (ICP).

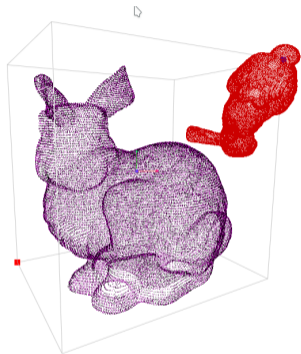


(a)

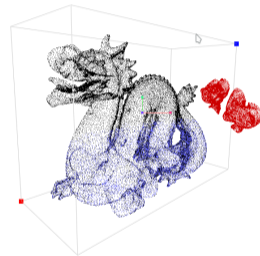


(b)

Figure 11: (a) After random removal attack (60% of random removed points; match percentage m^P : 87.50%)
(b) Random removal attack

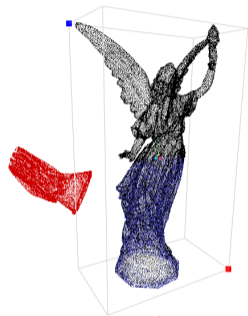


(a)

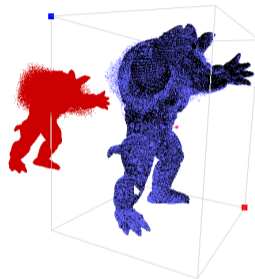


(b)

Figure 12: (a) Affine transformation ($m^P = 100\%$). (b) Affine transformation and cropping ($m^P = 95.31\%$).

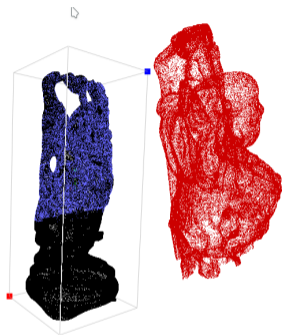


(a)

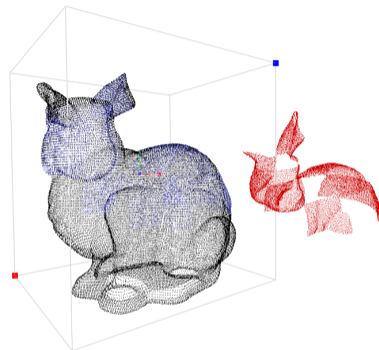


(b)

Figure 13: (a) Affine transformation, cropping and random removal ($m^P = 93.75\%$). (b) Affine transformation and local noise ($m^P = 96.88\%$).





(a)



(b)

Figure 14: (a) Affine transformation and cropping with slantwise cut ($m^P = 87.50\%$). (b) Affine transformation and complex cropping ($m^P = 93.51\%$).

-  Bogdan Lipuš and Borut Žalik.
Robust watermarking of airborne LiDAR data.
Multimedia Tools and Applications, 77(21):29077–29097, 2018.
-  Bogdan Lipuš and Borut Žalik.
3D convex hull-based registration method for point cloud watermark extraction.
Sensors, 19(15):3268, 2019.

