

## XXXIII Cycle

# **3D POINT CLOUD DENOISING USING** A JOINT GEOMETRY AND COLOR K-NN GRAPH

### Muhammad Abeer Irfan Supervisor: Prof. Enrico Magli

#### **Research context and motivation**

## Results

- Point cloud is an important representation of volumetric objects in three-dimensional space, which allows visualization from any viewpoint.
- The acquisition of a point cloud can be done using active sensors, or computed indirectly from multi-viewpoint images, the obtained point cloud suffers from the noise.
- The denoising of point cloud should be performed in order to improve the quality.
- The state-of-the-art technique denoised the geometry based on graph using geometry only.
- The disadvantage of the approach is that correct position of a point is estimated based on the graph signal representing the geometry only, the false surface is computed and as result holes are formed (See Fig. 1).



#### Natural Point clouds with real noise

- Demonstrate the visual comparison of the proposed algorithm with the denoised point cloud using a graph construction based on geometry only.
- The qualitative comparison has been made on both the real world and synthetic point clouds.





Figure 5: Palazzo\_Carignano model: (a) noisy input, (b)

(d) geometry only graph.

Figure 1: Green\_monster model: Geometry denoising based on geometry only graph.

#### Addressed research questions/problems

- How to denoise the geometry of a point cloud having no adverse effects?
- The displaced points in a noisy point cloud are noise? and should we eliminate them or relocate to their original positions?
- Can we relocate the noisy points with the same color to the points in their proximity to their original positions? (See Fig. 2)



Figure 2: Asterix model: Noisy points with the same color in their proximity.

### **Novel contributions**

#### **Graph Construction**

- Generate *K*-NN graph based on color similarity and geometry proximity in a 3D plane.
- Defining six-dimensional features for each point as  $p_i = [w_1 X_i w_2 C_i]$ , with  $X_i = [x_{1i} x_{2i}, x_{3i}]$  and



Figure 4: Arco\_Valentino model: (a) noisy input, (b) outlierfree input, denoised results by (c) proposed algorithm (d) geometry only graph.



Figure 6: Green\_monster model: (a) noisy input, (b) outlierfree input, denoised results by (c) proposed algorithm (d) geometry only graph.

Figure 7: Asterix model: (a) noisy input, (b) outlier-free input, denoised results by (c) proposed algorithm (d) geometry only graph.

#### Point clouds with synthetic noise

- To simulate the presence of noise, 50% of the points in the noise-free point clouds are affected by noise using uniform distribution.
- The comparative analysis of the proposed denoising algorithm with the state-of-the-art technique on synthetic point cloud is shown in Fig. 8.
- MSE and MCD comparison can be seen in Fig. 9 and Fig. 10.



Figure 8: red dotted section represents point cloud denoising using a joint geometry and color graph and blue dotted section represents point cloud denoising using geometry only.

outlier-free input, denoised results by (c) proposed algorithm

 $C_i = [c_{1i}, c_{2i}, c_{3i}]$  where  $c_{1i}, c_{2i}$  and  $c_{3i}$  are the color attributes and  $x_{1i}$   $x_{2i}$  and  $x_{3i}$  are the geometry coordinates of point *i*.

Figure 3: Illustration of a joint geometry and color k-NN graph, node A connected with red colored nodes

#### analogous to same color in its proximity.

#### Adopted methodologies

#### **Geometry Denoising**

- Geometry denoising is performed by exploiting the graph G constructed from both geometry and color information of the noisy point cloud.
- Convex minimization problem is considered for denoising the graph signal with the limitation that the signal must be smooth on a graph.
- The denoising problem can be written as optimization problem represented by Eq.1

$$\dot{x} = \arg\min_{x} \|x - g\|_{2}^{2} + \gamma \|\nabla_{G} x\|_{2}^{2}$$
(1)

• Where  $\dot{x}$  is the approximated denoised signal, the noisy signal is represented by  $g, \gamma$  is a parameter for regularization and  $\nabla_G x$  represents the gradient of the signal x on the graph G.

#### Submitted and published works

- Final draft of a conference paper is ready for submission to ICASSP.
- Drafting of Journal paper is in progress.





#### List of attended classes

- 01QRQRV Compressed sensing: theory and applications (30/5/19, 4 credits).
- 01TEVRV Deep learning (didattica di eccellenza) (4/6/2019, 6 credits).
- 01SFURV Programmazione scientifica avanzata in matlab (15/5/2019, 4 credits).
- 01QTEIU Data mining concepts and algorithms (6/3/2018, 4 credits).
- 01QSAIU Heuristics and metaheuristics for problem solving: new • trends and software tools (13/7/2018, 4 credits).
- 08IXTRV Project management (15/2/2018, 1 credit).
- 01RELKG Probabilità applicata e machine learning (3/9/2018, 4 credits).
- 01SHCRV Unsupervised neural networks (didattica di eccellenza) (9/4/2018, 6 credits).
- 02LWHRV Communication (15/2/2018, 1 credit).
- 01PJMRV Etica informatica (14/3/2018, 4 credits).
  - 01RISRV Public speaking (15/2/2018, 1 credit).







#### **Communications Engineering**