

# TP 4: Surface Reconstruction

## Objectives

- Test Poisson Surface Reconstruction method in CloudCompare
- Surface Reconstruction in Python : implement the Hoppe/IMLS implicit function

The report should be a pdf containing the answers to the **Questions** and named "TPX\_LASTNAME.pdf". Your code should be in a zip file named "TPX\_LASTNAME.zip". You can do the report as a pair, just state both your names inside the report and in the pdf and zip filenames, like "TPX\_LASTNAME1\_LASTNAME2.pdf"

Send your code along with the report to the email [mva.npm3d@gmail.com](mailto:mva.npm3d@gmail.com). The object of the mail must be "[NPM3D] TPX LASTNAME" or "[NPM3D] TPX LASTNAME1 LASTNAME2" if you are a pair working on the report.

## A. 3D Reconstruction in CloudCompare

The goal is to test and understand the well-known surface reconstruction method named Screened Poisson Reconstruction (SPR) on the point cloud "bunny\_normals.ply".

- 1) Open "bunny\_normals.ply" in CloudCompare and test the Poisson Reconstruction algorithm with Plugins -> PoissonRecon
- 2) You can change some parameters of the method with tab "Advanced"

In the file "bunny\_normals.ply", the normals have been computing using k nearest neighbors (k=30) and have been oriented to point away from the surface.

**Question 1:** Modify the parameters (octree depth, boundary, samples per node, point weight) to get the "best" reconstruction of the surface from PoissonRecon. Take a screenshot of the result. How many triangles does your final mesh have? Give the parameters that allowed you to have that result. Explain what "better" reconstruction means for this object.

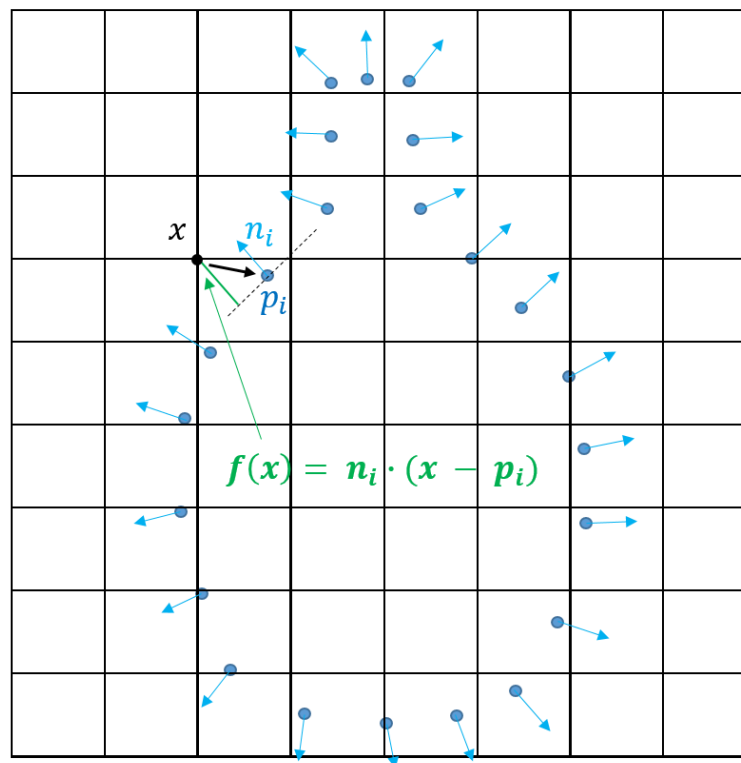
*Tips: to improve the quality of your screenshots, you can used EDL in CloudCompare*

## B.Surface Reconstruction in Python

### a. Implement the Hoppe implicit function

The goal is to implement in Python a classical surface reconstruction algorithm based on the Hoppe implicit function.

You need input point clouds with normal: you can use “bunny\_normals.ply” with pre-computed normals.



*Figure 1 Hoppe implicit function*

- 1) To compute the Hoppe function, we create first a regular grid of the volume space around the input point cloud.
- 2) Then, on every node  $x$  of the grid, the hoppe function is  $f(x) = n_i \cdot (x - p_i)$  when  $p_i$  is the closest point of the point cloud to  $x$  (and  $n_i$  the associated normal of  $p_i$ ).

- 3) The function  $f$  is a scalar field on a regular grid. To build of mesh, we need to extract the iso-zero of  $f$  with Marching Cubes. The Python package “scikit-image” has an efficient implementation of Marching Cubes.
- 4) We can finally export the mesh with the library trimesh and see the result in CloudCompare.

In `reconstruction.py`, implement the function `compute_hoppe` to fill the `scalar_field` with the Hoppe implicit function value.

**Question 2:** Make a surface reconstruction of the Bunny with the Hoppe function on a 128x128x128 voxel grid. Take a screenshot of the final mesh.

*Tips: start the implementation on a 16x16x16 voxel grid and when it is working, change the grid size.*

**Question 3:** Make a comparison between Hoppe and the reconstruction obtained with PoissonRecon in CloudCompare (visual comparisons with side-by-side screenshots, computation time, number of triangles, quality of reconstruction).

## b. Implement the IMLS implicit function

We have seen in the course that Hoppe implicit function is not a continuous surface representation. The Implicit Moving Least Square (IMLS) function is a better implicit function: it is a continuous surface representation.

For any node  $x$ , the IMLS function  $f(x)$  is defined by:

$$f(x) = \frac{\sum_i (n_i \cdot (x - p_i)) \theta(x - p_i)}{\sum_i \theta(x - p_i)}$$

$$\text{With: } \theta(x - p_i) = e^{-\frac{\|x - p_i\|^2}{h^2}}$$

The parameter  $h$  is proportional to the noise of the point cloud.

$h = 0.001$  is a good trade-off for the bunny point cloud.

Instead of computing the function  $f(x)$  using every point of the point cloud, you can take only the  $k$  nearest neighbors ( $k = 30$  is fine for many point clouds). In `reconstruction.py`, implement the function `compute_ims` to fill the `scalar_field` with IMLS implicit function value.

**Question 4:** Make a surface reconstruction of the Bunny with the IMLS function on a  $128 \times 128 \times 128$  voxel grid. Take a screenshot of the final mesh. Make a comparison between Hoppe, IMLS and the reconstruction obtained with PoissonRecon in CloudCompare (visual comparisons with side-by-side screenshots, computation time, number of triangles, quality of reconstruction).

*Tips: the computation for a  $128 \times 128 \times 128$  grid will take around 20 min on a laptop for the IMLS mesh (you can improve the code by vectorizing calculations or computing the function only on intersecting voxels).*

## C. Going further (BONUS)



As a bonus, you will test a recent surface reconstruction method using neural networks, Points2Surf.

Paper on arXiv: <https://arxiv.org/pdf/2007.10453.pdf>

Github: <https://github.com/ErlerPhilipp/points2surf>

**Question Bonus:** Make a reconstruction of the Bunny with Points2Surf using  $128 \times 128 \times 128$  voxel grid as a fair comparison with previous methods. Take a screenshot of the final mesh. Detail the model used, the training set and the parameters used for your result.

Make a comparison between Points2Surf, IMLS and PoissonRecon from CloudCompare (visual comparisons with side-by-side screenshots, computation time, number of triangles, quality of reconstruction).