

# Point Cloud Occupancy with Dynamic Planes

Computer Vision Course Project

Master's Degree in Artificial Intelligence and Robotics

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## 1 Introduction

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# What is the Addressed Problem

## 1 Introduction

- We are interested in learning the occupancy values of points inside a bounding box
- With encode a noisy point cloud into features and project them onto dynamic planes
- During Inference we reconstruct meshes with Multiresolution IsoSurface Extraction



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## 2 Dataset

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# FAUST Dataset

## 2 Dataset

This Dataset is composed by high-resolution human scans of 10 different bodies in 30 different poses.

- Each of the samples inside the training set has a corresponded ground truth alignment (registration)
- The test set is composed by 200 scans, while the training has 100 scans.
- About 80 % of the initial training set has been used for training, while the other 20 % has been used for validation



# Data Processing

## 2 Dataset

I have followed the same strategy of the original paper:

- We need to randomly sample 3000 points from the cloud's surface
- Then we inject Gaussian noise with zero mean and 0.05 standard deviation

These noisy clouds are then used to learn the features and the geometry of the object through the Encoder. The occupancy is then predicted over a set of 2048 points uniformly sampled over all the bounding box containing the original cloud/mesh.



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## 3 Architecture

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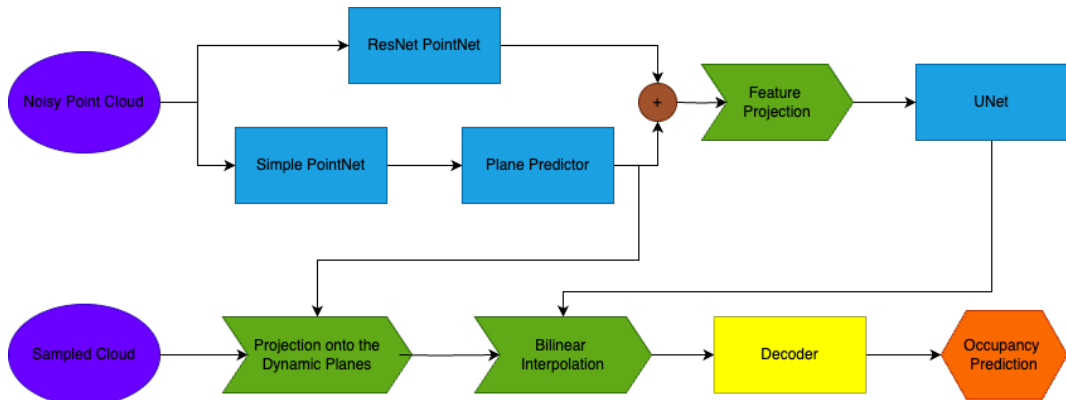
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# Architecture design

## 3 Architecture





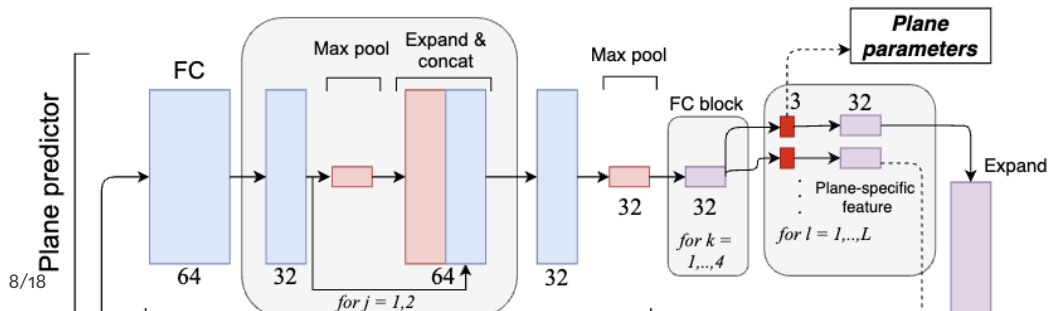


# Encoder

## 3 Architecture

The Encoder takes in input the Noisy Cloud and is composed by the following steps:

- ResNet PointNet
- Simple PointNet + Plane Predictor
- Feature summation + projection
- UNet





## Decoder

### 3 Architecture

The Dec is composed by:

- Feature Projection and Bilinear Interpolation
- Occupancy Network



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## 4 Reconstruction

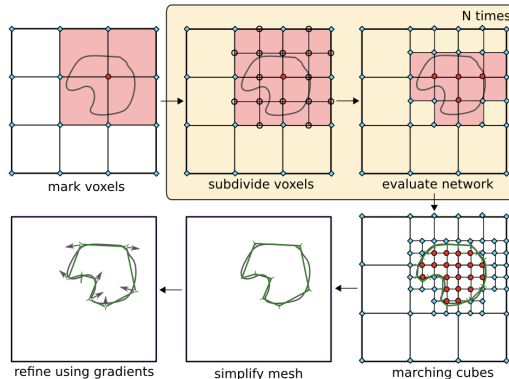
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# Multiresolution IsoSurface Extraction (MISE)

## 4 Reconstruction

- Create a grid over all the bounding box
- Evaluate the occupancy of each corner of the voxels
- Define the Active voxels as the one composed with at least one occupied corner and one not
- Subdivide each Active voxels into 8 subvoxels ( $2 \times 2 \times 2$  grid) and evaluate the new points occupancy
- Repeat until the desired resolution is obtained





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# Metrics

## 5 Results

In order to evaluate the performance of our model, the following metrics have been used:

- Chamfer Distance :

$$CD(A, B) = \frac{1}{|A|} \sum_{a \in A} \min_{b \in B} \|a - b\|_2^2 + \frac{1}{|B|} \sum_{b \in B} \min_{a \in A} \|b - a\|_2^2$$

- IOU :  $IoU(A', B') = \frac{|A' \cap B'|}{|A' \cup B'|}$
- F-Score:

Add each formula



# Loss Performances

5 Results

Insert here plots



# Metrics Performances

5 Results

Insert here just a table with metrics, gpu usage various types of sampling





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# Possible Changes and Future Improvements

6 Improvements



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