### **GEO1101** Synthesis Project

Digitizing Real-world Scenes from Images

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# 1. Introduction







- Why go 3D?
- What is behind the image?

# • 1 Introduction

- 2 Technical pipeline
- 3 Software presentation
- 4 Results and limitations
- 5 Conclusions & Future work

![](_page_5_Picture_0.jpeg)

Photogrammetry

# By a sequence of overlapping images depth can be captured

# Solving many equation returns 3D parameters of points that are visible from cameras

![](_page_6_Figure_3.jpeg)

But how can we find correspondences in images??

# Features!

![](_page_8_Picture_0.jpeg)

![](_page_8_Picture_1.jpeg)

### **Point clouds from images**

- Image quality against cost is advantageous
- Drone can capture every surface
- Surfaces behind obstacles can be transformed as well as a wide range of objects.
- Online pictures from artifacts and monuments can be used
- Colour can be adjusted to points

![](_page_10_Figure_0.jpeg)

### LIBRARIES

## **Functional Requirements**

### ✓ Data management

- ✓ Implementation of image matching (SIFT method)
- ✓ Implementation of structure from motion
- ✓ Implementation of Multi-view stereo
- ✓ Reconstruct 3D objects with smooth surfaces
- Reconstruct 3D objects with piece-wise planar surfaces.
- ✓ Export result of each step

# 2. Technical Pipeline

# Module A: Image Matching

#### Image matching

1. **Scale-space extrema detection:** The first stage of computation searches over all scales and image locations. It is implemented efficiently by using a difference-of-Gaussian function to identify potential interest points that are invariant to scale and orientation.

![](_page_14_Figure_2.jpeg)

#### Image matching

2. **Orientation assignment:** One or more orientations are assigned to each keypoint location based on local image gradient directions.

- ✓ Create histogram of local gradient directions at selected scale
- ✓ Assign orientation at peak of smoothed histogram
- ✓ (assistant orientation >80%peak)

![](_page_15_Picture_5.jpeg)

key point(x,y,scale,orientation)

#### Image matching

#### 3. Keypoints matching

![](_page_16_Picture_2.jpeg)

### Module B: Structure from Motion (SfM)

#### Structure from Motion (SfM)

Structure from Motion (SfM): The process of reconstructing 3D geometries from a sequence of 2D images taken from different viewpoints on same objects.

Output of SfM is a bundle file containing:

Sparse point cloud (X,Y,Z, n) Camera parameters (center, rotation) Their correlation

![](_page_18_Figure_4.jpeg)

Source: Schonberger, J. L., & Frahm, J. M. (2016). Structure-from-motion revisited.

### Module C: Multiview Stereo

## **Multiview Stereopsis**

- Aim : The construction of a dense point cloud
- Input : Images & camera parameters, matches, sparse point cloud
- Output characteristics : Points3D, normals, colour

![](_page_20_Picture_4.jpeg)

# **Multiview Stereopsis**

Patch based algorithm

Initializes patches to matching cells

Expand , Filtering (3 times)

Advantages:

- Works on occlusion and moving object existence
- Can reconstruct all the surfaces visible to cameras

![](_page_21_Picture_7.jpeg)

### Modula e: Surface Reconstruction

### **Smooth Surface Reconstruction**

- Good for curved and complex objects
- Using PoissonRecon by Kazhdan et al.
- Input : dense point cloud in PLY format with position, normal and colour of points
- Approached as a spatial Poisson problem
- Output : smooth and detailed triangulated meshes

### **Planar Surface Reconstruction**

- Useful for typical buildings with straight surfaces
- Input, dense point cloud in format
- Using the PolyFit method by Nan and Wonka
- Creates planes from point neighbourhoods and cuts at intersections
- Output: watertight solid

## **Technical pipeline**

![](_page_25_Picture_1.jpeg)

![](_page_25_Picture_2.jpeg)

5.1 Smooth surface reconstruction

![](_page_25_Picture_4.jpeg)

5.1 Plane surface reconstruction

# 3. Software presentation

### 4. Results and limitations

### Almere Watchtower

![](_page_28_Picture_1.jpeg)

![](_page_28_Picture_2.jpeg)

![](_page_29_Picture_0.jpeg)

### Wooden House

![](_page_30_Picture_1.jpeg)

![](_page_30_Picture_2.jpeg)

Object	Camera	number of pictures	angle of sight	number of points (after clipping to extent)
Wooden House	Mobile Phone	59 - 45	270	1,679,687

### Church in Delft

![](_page_31_Picture_1.jpeg)

![](_page_32_Picture_0.jpeg)

### Home with a skin

![](_page_33_Picture_1.jpeg)

![](_page_34_Picture_0.jpeg)

Object	Camera	number of pictures	angle of sight	number of points (after clipping to extent)
Home with a s	kin Drone	99 - 99	360	552,249

![](_page_35_Picture_0.jpeg)

![](_page_36_Picture_0.jpeg)

# Comparison with Pix4DMapper

![](_page_37_Picture_0.jpeg)

![](_page_38_Picture_0.jpeg)

- NARUX<sub>3</sub>D performed good overall
- One object at a time
- Quality of input will define the quality of final model
- Drone capturing oblique images gave the better output
- Problem definition is needed for algorithm selection
- NARUX3D requires further testing

![](_page_40_Picture_0.jpeg)