

## 1. Background

- Tactile Internet (TI) aims to allow real-time interaction between remote systems.
- However, ultra-low latency (ULL) networks have a distance limitation, as the speed of information is limited to the speed of light.
- A possible solution is to use a local physics simulation that matches the remote environment.
- But how can we simulate any remote environment locally? By using visual information to estimate the properties of objects.
- An RGB-D camera, such as the Microsoft Kinect, can be used to obtain both color and depth information of the objects in its field of view.
- This information can be represented in a point cloud - an array of points with XYZ-coordinates (and RGB-values).



Figure 1. Microsoft Kinect v1 [1].

Figure 2. A point cloud of objects on a table [2].

## 2. Research Question

- We want to be able to estimate the physical properties of objects in scenes similar to Figure 2.
- This is a complex problem, so perhaps a good starting question is: how to estimate the mass of an object from its point cloud?
- Assuming the density of the object can be obtained based on the material properties estimated from its point cloud.
- Then all we need is the volume of the object.

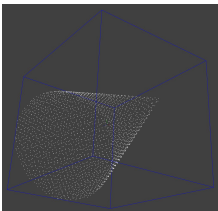


Figure 3. The computed oriented bounding box (OBB), shown in blue, for the downsampled point cloud of a cone.

## 3. Methodology

- A test set of synthetic point clouds was generated using artificial objects with varying rotations.
- Using this test set, we evaluated two volume estimation approaches:
  1. Based on the volume of the oriented bounding box (OBB).
  2. Based on the volume of the mesh.
- The mesh was created using a greedy triangulation surfacing algorithm implemented by the Point Cloud Library (PCL).
- The volume of the resulting mesh was estimated by subdividing it using tetrahedrons.

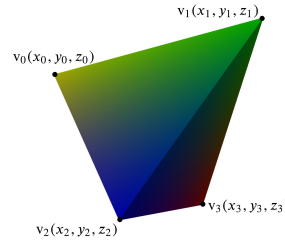


Figure 4. A tetrahedron - created from the vertices of a triangular face in the mesh,  $v_1$ ,  $v_2$ , and  $v_3$ , and the approximate center of the mesh,  $v_0$ .

- The volume of the tetrahedron can be calculated as follows:

$$\frac{1}{6} | -(x_2 - x_0)(y_3 - y_0)(z_1 - z_0) + (x_3 - x_0)(y_2 - y_0)(z_1 - z_0) + (x_2 - x_0)(y_1 - y_0)(z_3 - z_0) - (x_1 - x_0)(y_2 - y_0)(z_3 - z_0) - (x_3 - x_0)(y_1 - y_0)(z_2 - z_0) + (x_1 - x_0)(y_3 - y_0)(z_2 - z_0) |$$

- The sign of the volume is determined by the inner product of the vector from the center,  $v_0$ , to the vertex  $v_1$ , and the normal of the triangular face.

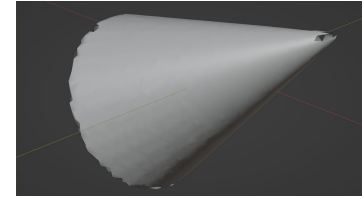


Figure 5. The resulting mesh of the point cloud for a cone (rotated 45°).

## 4. Results

Object	Rotation (degrees)	Expected Value (cm <sup>3</sup> )	Mean OBB Estimate (cm <sup>3</sup> )	Mean Mesh Estimate (cm <sup>3</sup> )	Mean OBB Error (%)	Mean Mesh Error (%)
Box	0.00	20,250.00	20,253.93	18,629.75	0.02%	8.00%
	7.00	20,250.00	20,857.21	18,689.14	3.00%	7.71%
Cone	0.00	4,581.66	17,931.24	4,655.32	291.37%	1.61%
	45.00	4,581.66	18,219.28	2,605.36	297.66%	43.13%
Mug	0.00	46,941.40	95,572.11	45,555.86	103.90%	2.95%
	7.00	46,941.50	95,669.36	45,377.22	103.91%	3.33%
Torus	0.00	10,846.60	28,223.36	10,768.54	160.20%	0.72%
	7.00	10,846.60	28,319.24	10,763.14	161.09%	0.77%
	45.00	10,846.60	28,389.51	10,763.63	161.74%	0.76%

Table 1. The results obtained from the 2 volume estimation approaches on the test set.

## 5. Conclusions

- In general, the mesh volume estimate performs better than the OBB volume estimate, in terms of the mean percent error.
- But performs worse than some other existing volume estimation methods, such as the method used in "Object Volume Estimation Based on 3D Point Cloud" [3].
- The test set used may be incompatible with the surfacing algorithm used.
- Further research should be conducted to verify the assumptions made.

### References:

- [1] Z. Zhang, "Microsoft kinect sensor and its effect." IEEE Multimedia, vol. 19, pp. 4–10, 2 Feb. 2012, issn: 1070-986X. doi: 10.1109/MMUL.2012.24. [Online]. Available: <http://ieeexplore.ieee.org/document/6190806/>
- [2] A. Makhaj, F. Thomas, and A. P. Gracia, "Grasping unknown objects in clutter by superquadric representation," Oct. 2017. [Online]. Available: <http://arxiv.org/abs/1710.02121>
- [3] W.-C. Chang, C.-H. Wu, Y.-H. Tsai, and W.-Y. Chiu, "Object volume estimation based on 3d point cloud," in 2017 International Automatic Control Conference (CAC), pp. 1–5, IEEE, 2017.