

| Method | bathub | bed | bookshelf | chair | desk | dresser | nightstand | sofa | table | toilet | 3D mAP |
|--------------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|
| DSS (?) | 44.2 | 78.8 | 11.9 | 61.2 | 20.5 | 6.4 | 15.4 | 53.5 | 50.3 | 78.9 | 42.1 |
| 2d-driven (?) | 43.5 | 64.5 | 31.4 | 48.3 | 27.9 | 25.9 | 41.9 | 50.4 | 37.0 | 80.4 | 45.1 |
| Pointfusion (?) | 37.3 | 68.6 | 37.7 | 55.1 | 17.2 | 24.0 | 32.3 | 53.8 | 31.0 | 83.8 | 44.1 |
| COG (?) | 58.3 | 63.7 | 31.8 | 62.2 | 45.2 | 15.5 | 27.4 | 51.0 | 51.3 | 70.1 | 47.6 |
| F-Pointnet(v1) (?) | 43.3 | 81.1 | 33.3 | 64.2 | 24.7 | 32.0 | 58.1 | 61.1 | 51.1 | 90.9 | 54.0 |
| V1-1024 | 51.6 | 82.0 | 32.2 | 54.6 | 33.5 | 32.4 | 67.9 | 66.3 | 48.0 | 88.2 | 55.7 |
| V1-2048 | 51.6 | 83.1 | 35.4 | 54.5 | 33.0 | 33.7 | 68.2 | 66.7 | 48.2 | 87.1 | 56.1 |
| Ours-1024 | 61.5 | 83.3 | 38.1 | 57.7 | 33.8 | 32.8 | 67.3 | 67.4 | 51.3 | 87.3 | 58.1 |
| Ours-2048 | 64.0 | 84.4 | 38.4 | 57.9 | 34.1 | 32.2 | 67.7 | 67.3 | 51.4 | 86.2 | 58.4 |

Table 3: Comparison with the state-of-the-art methods on SUN-RGBD test dataset.

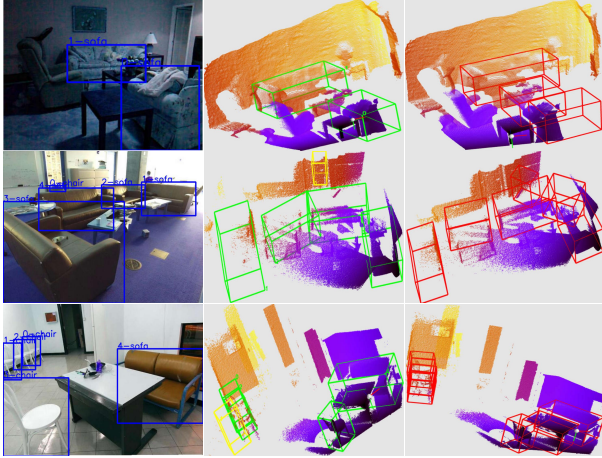


Figure 5: The visualization results for 3D object detection on SUN-RGBD test set. The first column shows the 2D bounding boxes based on YoloV3. The second column displays the 3D bounding boxes on 3D point clouds. The green boxes are true positives and yellow boxes are false positives. The 3D IoU threshold is 0.25. The last column illustrates the groundtruths which are in red.

Conclusion

In this paper, SIFRNet is put forward for 3D object detection, which is suitable for both indoor and outdoor scenes. The proposed architecture can make full use of the advantages of both RGB images and 3D point clouds. Outstanding experiment results both on KITTI dataset which contains a large number of sparse point clouds and SUN-RGBD dataset which contains many occluded objects reveal that our model has a certain generalization ability and robustness on 3D object detection tasks. Even when the point clouds are extremely sparse, our method can still obtain very satisfied results, which also demonstrates that our model can provide a better 3D representation. In future work, we will focus on the end-to-end trainable model for Lidar-only 3D data to improve the efficiency of 3D object detection tasks.

Acknowledgement

This project was partially supported by the National Natural Science Foundation of China (Grant No.61602485

| 2D mAP | 3D mAP | | |
|--------|--------|-------------|------|
| | V1 | Our | Gain |
| 40.6 | 46.1 | 48.1 | 2.0 |
| 47.1 | 49.6 | 51.6 | 2.0 |
| 50.5 | 52.2 | 54.3 | 2.1 |
| 52.9 | 55.5 | 57.6 | 2.1 |
| 53.9 | 56.1 | 58.4 | 2.3 |
| GT | 84.1 | 86.7 | 2.6 |

Table 4: The influence of 2D detection.

| Number of input points | 3D mAP | | | | | |
|------------------------|--------------------|---------------------|------|------------------|-------------------|------|
| | V1 _{53.9} | Our _{53.9} | Gain | V1 _{GT} | Our _{GT} | Gain |
| 32 | 43.9 | 51.1 | 7.2 | 63.1 | 70.2 | 7.1 |
| 128 | 53.1 | 56.3 | 3.2 | 79.2 | 83.2 | 3.0 |
| 256 | 54.4 | 56.9 | 2.5 | 81.4 | 85.3 | 2.9 |
| 512 | 55.4 | 57.8 | 2.4 | 83.2 | 86.0 | 2.8 |
| 1024 | 55.7 | 58.1 | 2.4 | 83.7 | 86.3 | 2.6 |
| 2048 | 56.1 | 58.4 | 2.3 | 84.1 | 86.7 | 2.6 |

Table 5: The influence of the number of input points.

and Grant No.61673375), the National Key Research and Development Program of China (Grant No.2016YFB1001005), and the Projects of Chinese Academy of Sciences (Grant No.QYZDB-SSW-JSC006 and Grant No.173211KYSB20160008).

Quaerat vero aliquam, nisi voluptates necessitatibus sint. Molestiae assumenda dolorem ratione deleniti, unde voluptate quod quaerat numquam cumque, magni provident omnis dolor numquam fugiat libero veritatis nihil accusantium sapiente, officiis ipsum qui aliquid veniam eligendi. Laudantium doloremque ea quis illum nihil similique quibusdam eaque, exercitationem accusamus neque voluptatibus voluptas quam dignissimos aperiam illo, natus optio voluptatum accusantium perspiciatis temporibus saepe nobis. Modi rem ratione accusantium repudiandae cumque neque cupiditate vero a ex, corporis alias veritatis suscipit praesentium modi ab aliquid tenetur dolorem rerum. Sed dolores laudantium eos perspiciatis corrupti expedita ab, quos odio nostrum eaque ipsam blanditiis consectetur odit nulla incidunt, esse iusto quam necessitatibus, tempore sequi officia laborum, reiciendis obcaecati est accusantium ducimus animi itaque. Rem necessitatibus laudantium culpa maiores dicta ipsum aut nobis neque, repellat iste tempora quis ul-

lam explicabo laudantium quo beatae officiis reprehenderit
incidunt, saepe quaerat pariatur maxime nihil culpa illum in-
ventore dolorum ipsam minima voluptatum,