基于 RGB-D 图像的三维物体识别算法的研究与 实现

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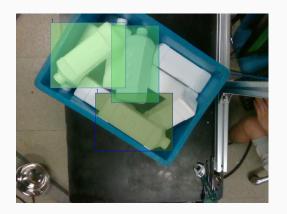
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- 2. Dual RGB-D Camera
- 3. 3D-MRAI Algorithm
- 4. Application: Bin-picking
- 5. Conclusion

Introduction

Problem

Problem



Problem

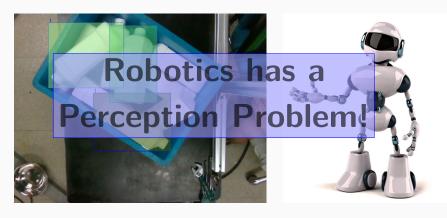
2D Detection



Where to pick?



Where to pick?



RGB Image





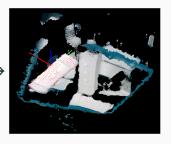
Depth Map

RGB Image





3D Detection Results



Depth Map

Dual RGB-D Camera

Problem & Key

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Problem: SR300 相机对反光物体在某些角度下深度信息有严重的缺失

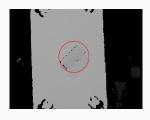


Problem & Key

Problem: SR300 相机对反光物体在某些角度下深度信息有严重的缺失



Key: 改变相机的拍摄角度



Structure of Dual RGB-D Camera

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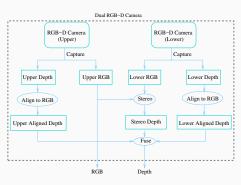


对偶 RGB-D 相机物理结构

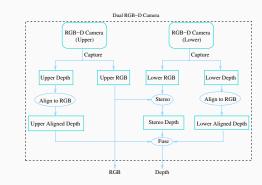
Structure of Dual RGB-D Camera



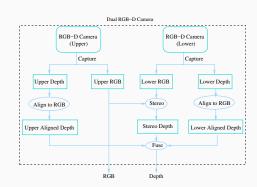
对偶 RGB-D 相机物理结构



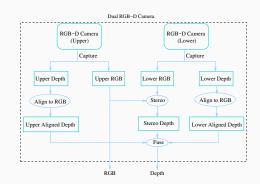
对偶 RGB-D 相机内部原理图



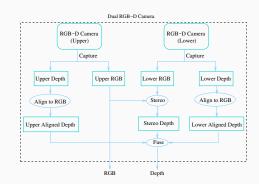
■ 将深度图与彩色图对齐

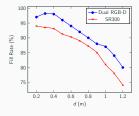


- 将深度图与彩色图对齐
- 通过双目匹配算法 (ELAS) 形成一张新的 深度图

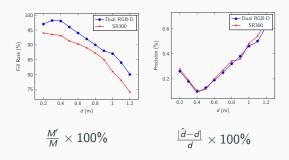


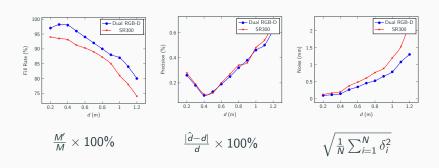
- 将深度图与彩色图对齐
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- 融合三张深度图





$$\frac{M'}{M} \times 100\%$$



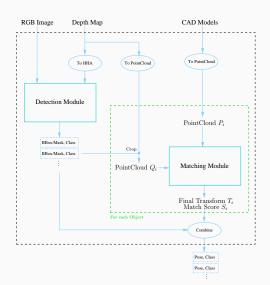


3D-MRAI Algorithm

3D-MRAI

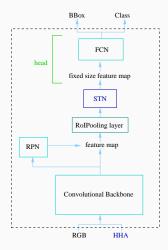
3D-MRAI

3D-MRAI (3D Mask R-CNN with Angle-fixed-4PCS and ICP)



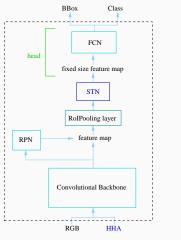
Detection Module

Detection Module

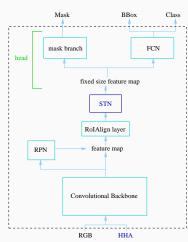


Based on Faster R-CNN

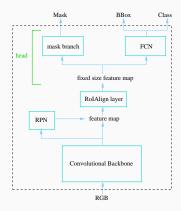
Detection Module



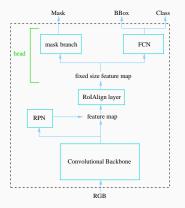
Based on Faster R-CNN



Based on Mask R-CNN



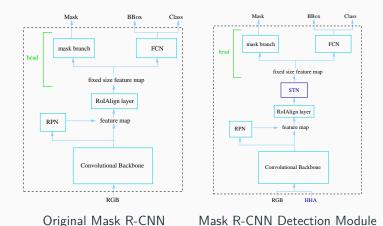
Original Mask R-CNN



Mask BBox Class mask branch FCN fixed size feature map STN RoIAlign layer feature map RPN Convolutional Backbone RGB HHA

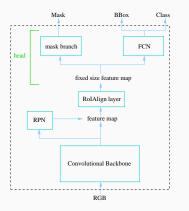
Original Mask R-CNN

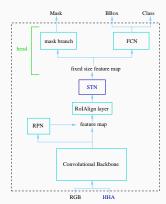
Mask R-CNN Detection Module



■ 引入 HHA, 有效利用三维信息, 解决原算法难以检测纹理缺少的 物体

Mask R-CNN Module





Original Mask R-CNN Mask R-CNN Detection Module

- 引入 HHA, 有效利用三维信息, 解决原算法难以检测纹理缺少的 物体
- 在网络中增加 STN(Spatial Transformer Network),使得提取的特 征具有旋转不变性、增加了算法检测准确度

Detection Module

Detection Module







Height above ground frame







HHA frame

Detection Module



Horizontal disparity frame



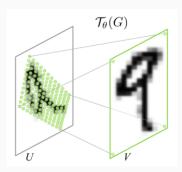
Height above ground frame



Angle with gravity frame



HHA frame



Detection Module Results

Detection Module Results

表 1: APC 数据集上的精确度

	input	output	AP	$AP_{0.5}$	$AP_{0.75}$
Faster R-CNN	RGB	bbox	33.26	56.29	34.03
Our method(based on Faster R-CNN)	RGB + HHA	bbox	34.55	57.99	34.69
Mask R-CNN	RGB	mask	32.34	55.78	33.12
Our method(based on Mask R-CNN)	RGB+HHA	mask	33.94	56.45	33.99

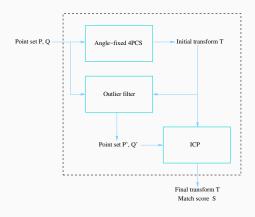
Detection Module Results

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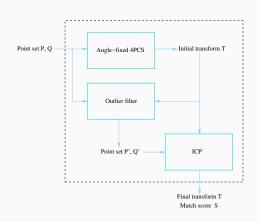
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表 2: workpiece 数据集上的精确度

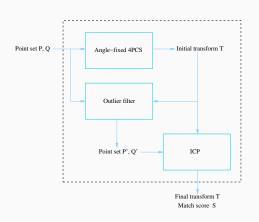
	input	output	AP	$AP_{0.5}$	$AP_{0.75}$
Faster R-CNN	RGB	bbox	18.78	37.49	19.46
Our method(based on Faster R-CNN)	RGB + HHA	bbox	32.39	56.37	33.54
Mask R-CNN	RGB	mask	16.12	35.95	18.74
Our method(based on Mask R-CNN)	RGB + HHA	mask	30.98	53.74	32.19

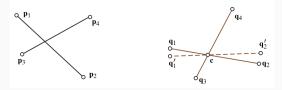


去除 4PCS 算法中角度 不相等的基,减少了算 法运算时间

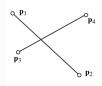


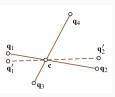
- 去除 4PCS 算法中角度 不相等的基,减少了算 法运算时间
- 通过增加滤波和 ICP 算法,增加了匹配精度





4PCS 存在的问题





4PCS 存在的问题



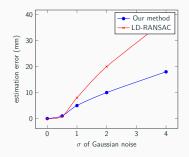
目标 3D 模型转化的点云



从 BBox 裁剪得到的点云

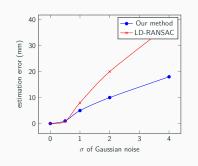
Matching Module Results

Matching Module Results



$$E = \sqrt{\frac{1}{N} \sum_{i=1}^{N} \min_{p_j \in T(P)} \|q_i - p_j\|^2}$$

Matching Module Results



5

4

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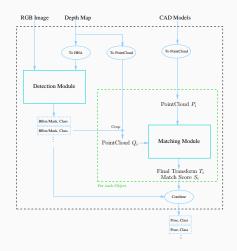
 $d\uparrow$, the size of point cloud \downarrow

Our method

- LD-RANSAC

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Combine Everything



Combine Everything

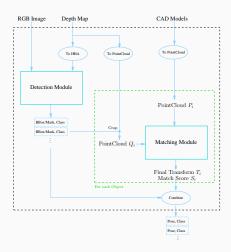
13

14 return Res

```
Input: RGB Image I, Depth Map
          D. CAD Models M
  Output: Set of Pose and Class Res

    Res ← Ø:

2 P ← Ø:
3 forall M_i ∈ M do
4 | P←
      \{P, CAD2PointCloud(M_i)\};
5 H = Depth2HHA(D);
6 Q = Depth2PointCloud(D):
7 Mask, Class \leftarrow DetectModule(I, H);
8 forall m_i \in Mask, c_i \in Class do
    Q_i \leftarrow Crop(Q, m_i);
   P_i \leftarrow P(c_i):
   T_i, S_i \leftarrow MatchModule(P_i, Q_i);
    if S_i > S_{min} then
       Res \leftarrow \{Res, [T_i, c_i]\};
```















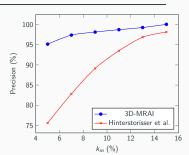
<i>k_m</i> [%]	5	7	9	11	13	15
Hinterstorisser et al.	75.63	83.84	89.13	93.48	96.83	98.12
3D-MRAI	95.12	97.35	98.10	98.69	99.22	100.00







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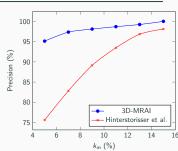




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$$m = \underset{\mathbf{x} \in M}{\operatorname{avg}} \| (R\mathbf{x} + t) - (\tilde{R}\mathbf{x} + \tilde{t}) \|$$

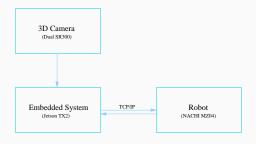
如果 $m < k_m d$,算法输出位姿正确



Application: Bin-picking

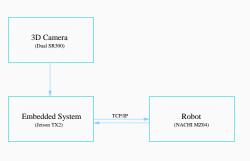
System Structure

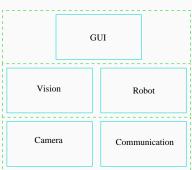
System Structure



硬件架构

System Structure



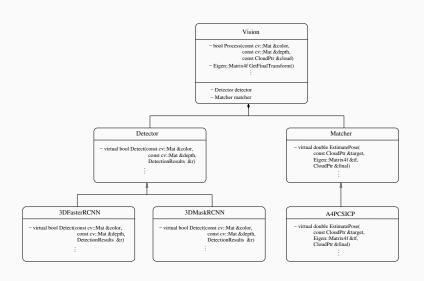


硬件架构

软件架构

Design Example

Design Example



Experiment

Experiment



实验环境

Experiment



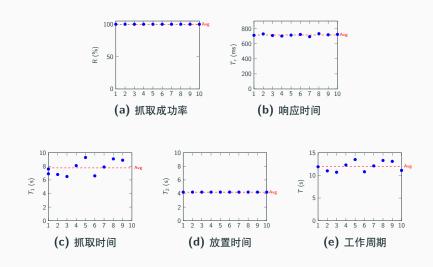


实验环境物料箱

	成功率 R	响应时间 T_r	抓取时间 T ₁	放置时间 T ₂	工作周期 T
1	100%	711ms	7.6s	4.2s	11.9s
2	100%	729ms	6.8s	4.2s	11.0s
3	100%	708ms	6.5s	4.2s	10.7s
4	100%	701ms	8.1s	4.2s	12.3s
5	100%	713ms	9.3s	4.2s	13.5s
6	100%	722ms	6.6s	4.2s	10.8s
7	100%	693ms	7.9s	4.2s	12.1s
8	100%	732ms	9.1s	4.2s	13.3s
9	100%	718ms	8.9s	4.2s	13.1s
10	100%	723ms	6.9s	4.2s	11.1s
Avg.	100%	715ms	7.77s	4.20s	11.98s

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Avg.	100%	715ms	7.77s	4.20s	11.98s

$$T = T_1 + \max(T_r, T_2)$$



Conclusion

■ 对偶 RGB-D 相机结构相比单个 RGB-D 相机有更高的填充率,更低的噪声,但对精度的提升不明显

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- 3D-MRAI 算法相比传统算法表现出了较高的检测准确率,估计的 位姿精度也更高,但 FPS 较低

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- 基于 3D-MRAI 算法设计的 Bin-Picking 视觉系统具有更高的抓取 成功率、更快的响应速度,以及更低的成本

Questions?