

CSE 565 HW 2 REPORT

244201001033 AKCAN ERCAN

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- Your features detected on the objects
- The details of your object recognition algorithm for each of the 3 feature detectors
- The final confusion matrix for the 10 object recognition experiments for 3 different algorithms.
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➤ The list of selected object

I selected last 10 objects in rgbd-dataset; rubber_eraser, scissors, shampoo, soda_can, sponge, stapler, tomato, toothbrush, toothpaste, water_bottle.

There are **182,851** items, totally **675.7 MB**.

I saw bunch of sub-folder in each main folder. All sub-folder has three main picture;

- for example tomato object has 8 sub-folder;
 - *tomato_1, tomato_2, tomato_3, tomato_4, tomato_5, tomato_6, tomato_7, tomato_8*
- And tomato_1 has 3 main file;
 - *tomato_1_1_1_crop.png, tomato_1_1_1_depthcrop.png, tomato_1_1_1_loc.txt*
- As you can see, actually we need ..._crop.png, so I selected all pictures where includes name with ***_crop.png*** with sub-folder and main folder. The final folder and images;
- The sum of trained object, I mean sum of *_crop.png* files are **45,697** images.

```
# Server: GPU-ComputerVision Variables:Homework2_244201001033_AKCAN_ERCA...
> images = {dict: 10} {'rubber_eraser': [([211 187 171], [213 187 171], [214 188 172], ..., [210 183 167], [209 186 167], [209 186 167]), ([211 187 171], [214 189 170], [213 188 169], ..., [209 186 167], [209 186 167], [208 185 166], [207 184 165], [206 183 164], [205 182 163], [204 181 162], [203 180 161], [202 179 160], [201 178 159], [200 177 158], [199 176 157], [198 175 156], [197 174 155], [196 173 154], [195 172 153], [194 171 152], [193 170 151], [192 169 150], [191 168 149], [190 167 148], [189 166 147], [188 165 146], [187 164 145], [186 163 144], [185 162 143], [184 161 142], [183 160 141], [182 159 140], [181 158 139], [180 157 138], [179 156 137], [178 155 136], [177 154 135], [176 153 134], [175 152 133], [174 151 132], [173 150 131], [172 149 130], [171 148 129], [170 147 128], [169 146 127], [168 145 126], [167 144 125], [166 143 124], [165 142 123], [164 141 122], [163 140 121], [162 139 120], [161 138 119], [160 137 118], [159 136 117], [158 135 116], [157 134 115], [156 133 114], [155 132 113], [154 131 112], [153 130 111], [152 129 110], [151 128 109], [150 127 108], [149 126 107], [148 125 106], [147 124 105], [146 123 104], [145 122 103], [144 121 102], [143 120 101], [142 119 100], [141 118 99], [140 117 98], [139 116 97], [138 115 96], [137 114 95], [136 113 94], [135 112 93], [134 111 92], [133 110 91], [132 109 90], [131 108 89], [130 107 88], [129 106 87], [128 105 86], [127 104 85], [126 103 84], [125 102 83], [124 101 82], [123 100 81], [122 99 80], [121 98 79], [120 97 78], [119 96 77], [118 95 76], [117 94 75], [116 93 74], [115 92 73], [114 91 72], [113 90 71], [112 89 70], [111 88 69], [110 87 68], [109 86 67], [108 85 66], [107 84 65], [106 83 64], [105 82 63], [104 81 62], [103 80 61], [102 79 60], [101 78 59], [100 77 58], [99 76 57], [98 75 56], [97 74 55], [96 73 54], [95 72 53], [94 71 52], [93 70 51], [92 69 50], [91 68 49], [90 67 48], [89 66 47], [88 65 46], [87 64 45], [86 63 44], [85 62 43], [84 61 42], [83 60 41], [82 59 40], [81 58 39], [80 57 38], [79 56 37], [78 55 36], [77 54 35], [76 53 34], [75 52 33], [74 51 32], [73 50 31], [72 49 30], [71 48 29], [70 47 28], [69 46 27], [68 45 26], [67 44 25], [66 43 24], [65 42 23], [64 41 22], [63 40 21], [62 39 20], [61 38 19], [60 37 18], [59 36 17], [58 35 16], [57 34 15], [56 33 14], [55 32 13], [54 31 12], [53 30 11], [52 29 10], [51 28 9], [50 27 8], [49 26 7], [48 25 6], [47 24 5], [46 23 4], [45 22 3], [44 21 2], [43 20 1], [42 19 0], [41 18 0], [40 17 0], [39 16 0], [38 15 0], [37 14 0], [36 13 0], [35 12 0], [34 11 0], [33 10 0], [32 9 0], [31 8 0], [30 7 0], [29 6 0], [28 5 0], [27 4 0], [26 3 0], [25 2 0], [24 1 0], [23 0 0], [22 0 0], [21 0 0], [20 0 0], [19 0 0], [18 0 0], [17 0 0], [16 0 0], [15 0 0], [14 0 0], [13 0 0], [12 0 0], [11 0 0], [10 0 0], [9 0 0], [8 0 0], [7 0 0], [6 0 0], [5 0 0], [4 0 0], [3 0 0], [2 0 0], [1 0 0], [0 0 0]], [211 187 171], [213 187 171], [214 188 172], ..., [210 183 167], [209 186 167], [209 186 167]), ([211 187 171], [214 189 170], [213 188 169], ..., [209 186 167], [209 186 167], [208 185 166], [207 184 165], [206 183 164], [205 182 163], [204 181 162], [203 180 161], [202 179 160], [201 178 159], [200 177 158], [199 176 157], [198 175 156], [197 174 155], [196 173 154], [195 172 153], [194 171 152], [193 170 151], [192 169 150], [191 168 149], [190 167 148], [189 166 147], [188 165 146], [187 164 145], [186 163 144], [185 162 143], [184 161 142], [183 160 141], [182 159 140], [181 158 139], [180 157 138], [179 156 137], [178 155 136], [177 154 135], [176 153 134], [175 152 133], [174 151 132], [173 150 131], [172 149 130], [171 148 129], [170 147 128], [169 146 127], [168 145 126], [167 144 125], [166 143 124], [165 142 123], [164 141 122], [163 140 121], [162 139 120], [161 138 119], [160 137 118], [159 136 117], [158 135 116], [157 134 115], [156 133 114], [155 132 113], [154 131 112], [153 130 111], [152 129 110], [151 128 109], [150 127 108], [149 126 107], [148 125 106], [147 124 105], [146 123 104], [145 122 103], [144 121 102], [143 120 101], [142 119 100], [141 118 99], [140 117 98], [139 116 97], [138 115 96], [137 114 95], [136 113 94], [135 112 93], [134 111 92], [133 110 91], [132 109 90], [131 108 89], [130 107 88], [129 106 87], [128 105 86], [127 104 85], [126 103 84], [125 102 83], [124 101 82], [123 100 81], [122 99 80], [121 98 79], [120 97 78], [119 96 77], [118 95 76], [117 94 75], [116 93 74], [115 92 73], [114 91 72], [113 90 71], [112 89 70], [111 88 69], [110 87 68], [109 86 67], [108 85 66], [107 84 65], [106 83 64], [105 82 63], [104 81 62], [103 80 61], [102 79 60], [101 78 59], [100 77 58], [99 76 57], [98 75 56], [97 74 55], [96 73 54], [95 72 53], [94 71 52], [93 70 51], [92 69 50], [91 68 49], [90 67 48], [89 66 47], [88 65 46], [87 64 45], [86 63 44], [85
```

- I selected 3 different feature detector (descriptor): **SIFT**, **ORB** and **BRISK**

➤ The features detected on the objects

- When I try to show features detected on the objects, I got an error. Actually, features have two element; List {cv2.KeyPoint} and ndarray : { 33, 228}. I found these features for 3 detectors. In that place I will show detected features in PyCharm IDE. As you can see, each features have different KeyPoint in List (actually tuple) and different dimension of ndarray. It is impossible for me to show all the features anyway, because there are so many features.

```
Server: GTU-ComputerVision  Variables:Homework2_244201001033_AKCAN_ERCA... x

KNeighborsClassifier = {ABCMeta} <class 'sklearn.neighbors._classification.KNeighborsClassifier'>
base_path = {str} '.././.././../Downloads/rgbd-dataset/'
brisk = {BRISK} <cv2.BRISK 0x7f83540f33d0>
des = {ndarray: (20, 64)} [[180 59 215 ... 12 16 48], [176 1 0 ... 15 17 48], [180 27 83 ... 185 57 59], ..., [ 75 254 63 ...
detector = {str} 'BRISK'
feature_detectors = {dict: 3} {'BRISK': <cv2.BRISK 0x7f83540f33d0>, 'ORB': <cv2.ORB 0x7f83540f3fb0>, 'SIFT': <cv2.SIFT 0:
features = {dict: 3} {'BRISK': {'rubber_eraser': [((< cv2.KeyPoint 0x7f8353734630>, < cv2.KeyPoint 0x7f8353734660>), [[235 2
> {'SIFT': {dict: 10} {'rubber_eraser': [((< cv2.KeyPoint 0x7f839d2de190>, < cv2.KeyPoint 0x7f839d2dcde0>, < cv2.KeyPoint 0
> > {'rubber_eraser' = {list: 3110} [((< cv2.KeyPoint 0x7f839d2de190>, < cv2.KeyPoint 0x7f839d2dcde0>, < cv2.KeyPoint 0x7
> > {'scissors' = {list: 2371} [((< cv2.KeyPoint 0x7f834fd2aaf0>, < cv2.KeyPoint 0x7f834fd2ab20>, < cv2.KeyPoint 0x7f834fd
> > {'shampoo' = {list: 4776} [((< cv2.KeyPoint 0x7f834cb7da40>, < cv2.KeyPoint 0x7f834cb7da70>, < cv2.KeyPoint 0x7f834
> > {'soda_can' = {list: 3555} [((< cv2.KeyPoint 0x7f833eb0b450>, < cv2.KeyPoint 0x7f833eb0b480>, < cv2.KeyPoint 0x7f83
> > {'sponge' = {list: 8668} [((< cv2.KeyPoint 0x7f833d5e5740>, < cv2.KeyPoint 0x7f833d5e5770>, < cv2.KeyPoint 0x7f833c
> > {'stapler' = {list: 5084} [((< cv2.KeyPoint 0x7f833cd9d740>, < cv2.KeyPoint 0x7f833cd9d770>, < cv2.KeyPoint 0x7f833c
> > {'tomato' = {list: 5231} [((< cv2.KeyPoint 0x7f833c2cad30>, < cv2.KeyPoint 0x7f833c2cad60>, < cv2.KeyPoint 0x7f833c
> > 0000 = {tuple: 2} ((< cv2.KeyPoint 0x7f833c2cad30>, < cv2.KeyPoint 0x7f833c2cad60>, < cv2.KeyPoint 0x7f833c2c
> > 0 = {tuple: 33} (< cv2.KeyPoint 0x7f833c2cad30>, < cv2.KeyPoint 0x7f833c2cad60>, < cv2.KeyPoint 0x7f833c2c
> > 1 = {ndarray: (33, 128)} [[ 6. 75. 24. ... 0. 0. 8.], [ 5. 55. 129. ... 0. 0. 6.], [ 92. 2. 0. ... 0. 0. 0.], ..., [ 3:
    __len__ = {int} 2
    Protected Attributes
> > 0001 = {tuple: 2} ((< cv2.KeyPoint 0x7f833c2cb4e0>, < cv2.KeyPoint 0x7f833c2cb510>, < cv2.KeyPoint 0x7f833c2c
> > 0 = {tuple: 22} (< cv2.KeyPoint 0x7f833c2cb4e0>, < cv2.KeyPoint 0x7f833c2cb510>, < cv2.KeyPoint 0x7f833c2c
> > 1 = {ndarray: (22, 128)} [[ 66. 59. 1. ... 0. 16. 23.], [ 35. 0. 0. ... 0. 1. 132.], [ 3. 121. 83. ... 0. 0. 4.], ..., [
    __len__ = {int} 2
    Protected Attributes
> > 0002 = {tuple: 2} ((< cv2.KeyPoint 0x7f833c2cb900>, < cv2.KeyPoint 0x7f833c2cb930>, < cv2.KeyPoint 0x7f833c2c
> > 0003 = {tuple: 2} ((< cv2.KeyPoint 0x7f833c2cbc00>, < cv2.KeyPoint 0x7f833c2cbc30>, < cv2.KeyPoint 0x7f833c2c
> > 0004 = {tuple: 2} ((< cv2.KeyPoint 0x7f833c2cbf90>, < cv2.KeyPoint 0x7f833c2cbfc0>, < cv2.KeyPoint 0x7f833c2cc
```

- We also understand here that each descriptor didn't captured the same number of features. I mean; the sift descriptor described **5231** features for tomato object while the orb descriptor described **320** features object and brisk descriptor described **1347** features for tomato object.

[illegible]

➤ The details of your object recognition algorithm

- My object recognition algorithm has some steps;
 - Take an image,
 - Convert gray scale,
 - Detect and compute for each detectors: *SIFT*, *ORB* & *BRISK*,
 - If detector give description without none,
 - Add features to relevant key: *SIFT*, *ORB* & *BRISK*,
 - For loop for all image.

```
1 for object in images:
2     for img in images[object]:
3         gray_image = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
4         for name, detector in feature_detectors.items():
5             keypoints, descriptors = detector.detectAndCompute(gray_image, None) # Each detector detects and computes
6             if descriptors is not None: # IMPORTANT CASE: Ensure that there are descriptors
7                 features[name][object].append((keypoints, descriptors))
```

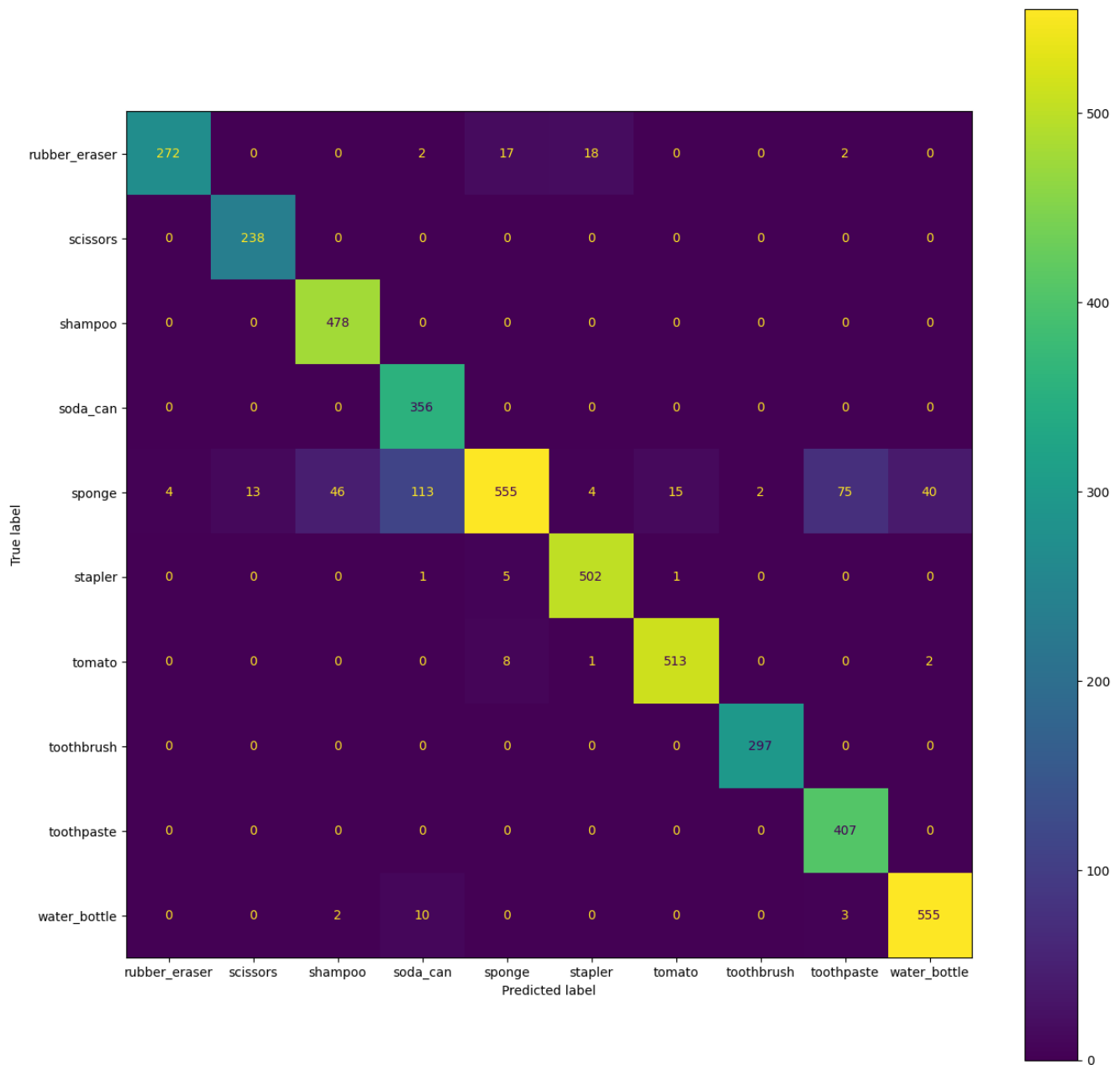
- The simple steps !

➤ The final confusion matrices

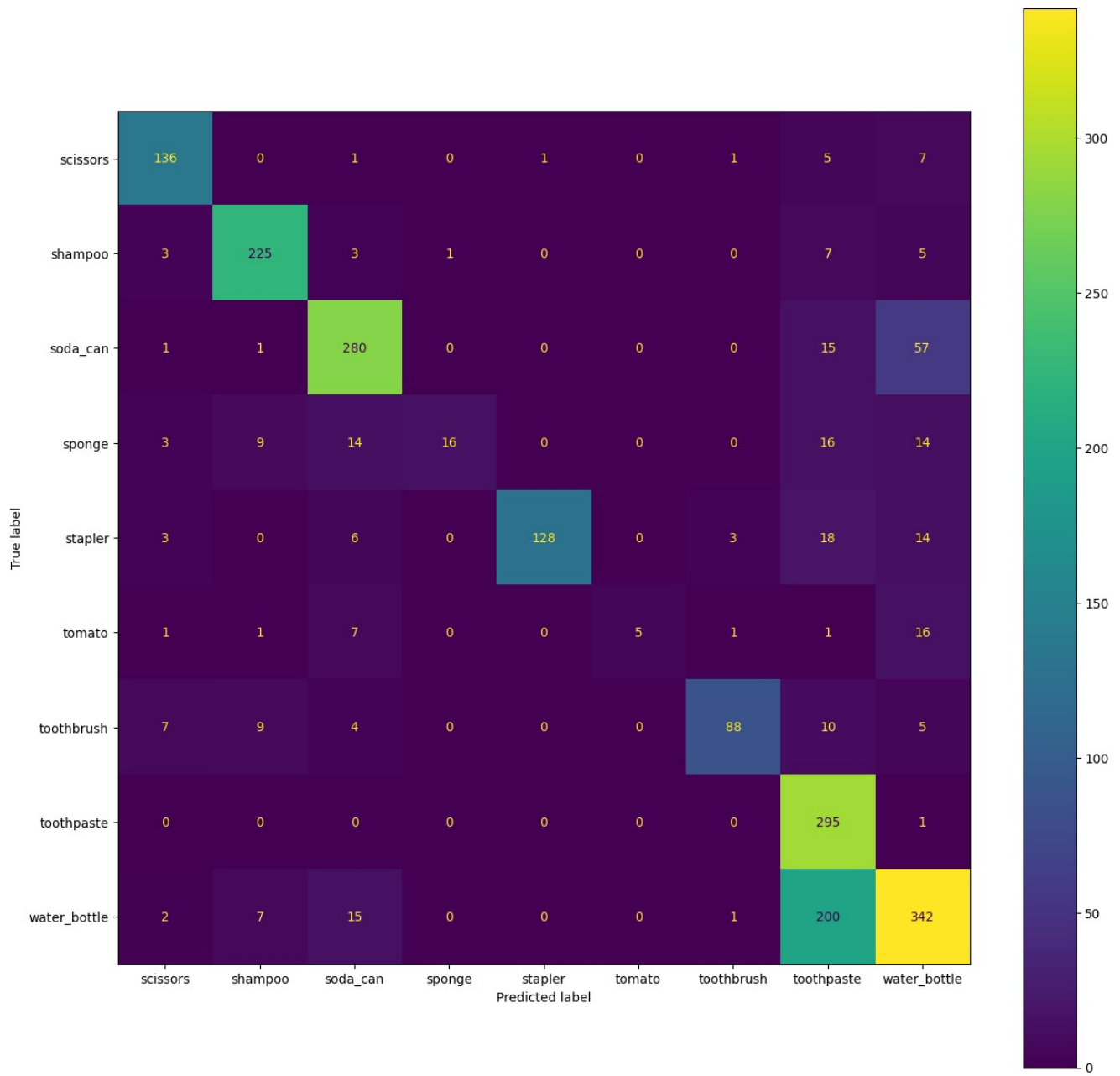
It took approximately **24 minutes** for all the algorithms to run and the results to appear.

There 3 different confusion matrix; ***SIFT, ORB & BRISK***

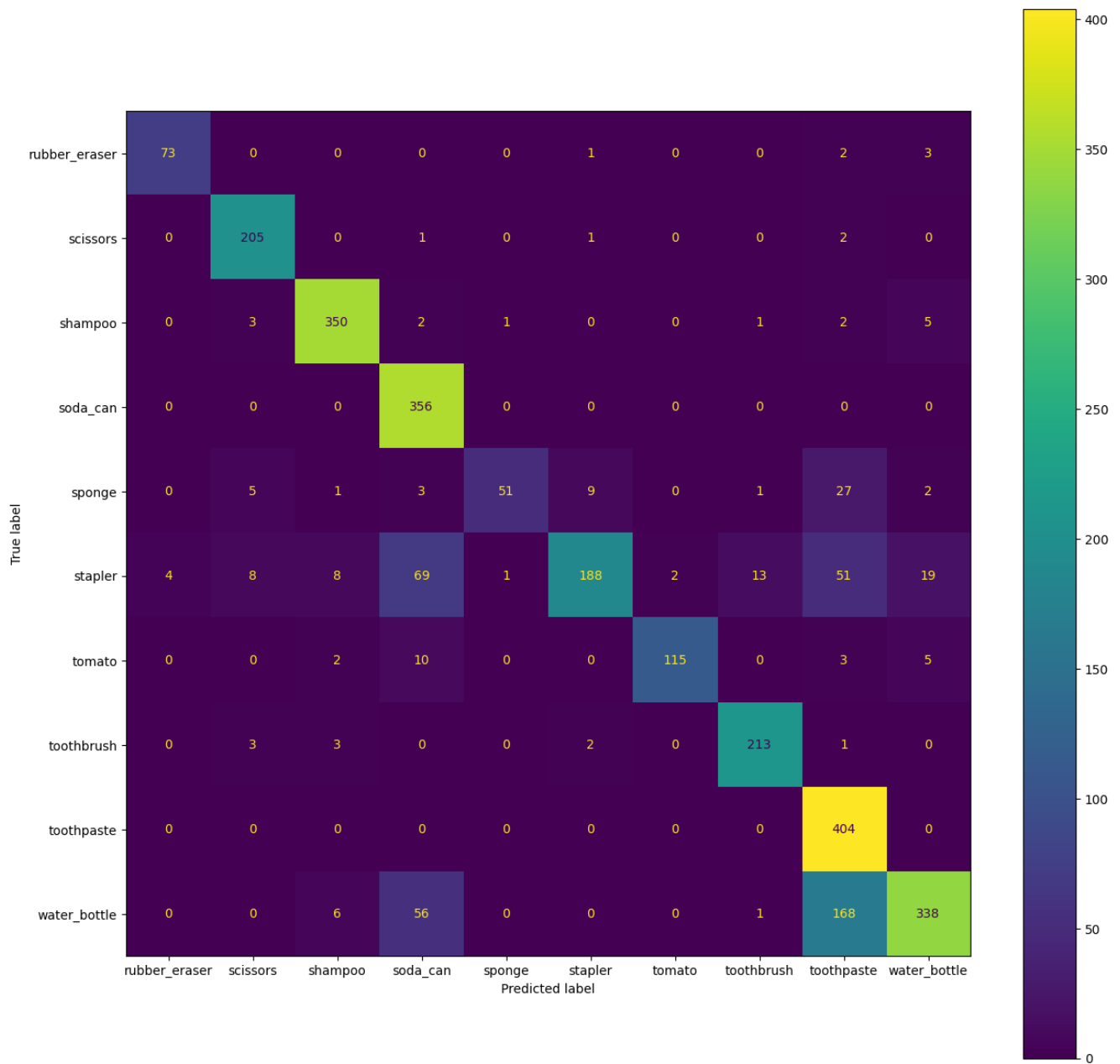
- Evaluating classifier: **SIFT**, Accuracies: **0.9157340355497038**



- Evaluating classifier: **ORB**, Accuracies: **0.7533565390353059**



- Evaluating classifier: **BRISK**, Accuracies: **0.8189285714285715**



➤ Discussion

As we can see, **SIFT** is better for accuracy and failures.

We also discuss above section that each descriptor didn't captured the same number of features. I mean; the sift descriptor described **5231** features for *tomato* object while the orb descriptor described **320** features for *tomato* object and brisk descriptor described **1347** features for *tomato* object.

These results occur from the methods that each algorithm uses when detecting features. Because if the algorithm cannot detect a feature for each image, it cannot say anything about the image, so the accuracy rate decreases.

```
Server: GTU-ComputerVision  Variables:Homework2_244201001033_AKCAN_ERCA... x
feature_detectors = {dict: 3} {'BRISK': < cv2.BRISK 0x7f83540f33d0>, 'ORB': < cv2.ORB 0x7f83540f3fb0>, 'SIFT': < cv2.SIFT 0x7f83540f3fb0>}
features = {dict: 3} {'BRISK': {'rubber_eraser': [((< cv2.KeyPoint 0x7f8353734630>, < cv2.KeyPoint 0x7f8353734660>), [(235 :
> 'rubber_eraser' = {list: 3110} [((< cv2.KeyPoint 0x7f839d2de190>, < cv2.KeyPoint 0x7f839d2dcde0>, < cv2.KeyPoint 0x7f839d2dcde0>), [(235 :
> 'scissors' = {list: 2371} [((< cv2.KeyPoint 0x7f834fd2aaf0>, < cv2.KeyPoint 0x7f834fd2ab20>, < cv2.KeyPoint 0x7f834fd2ab20>), [(235 :
> 'shampoo' = {list: 4776} [((< cv2.KeyPoint 0x7f834cb7da40>, < cv2.KeyPoint 0x7f834cb7da70>, < cv2.KeyPoint 0x7f834cb7da70>), [(235 :
> 'soda_can' = {list: 3555} [((< cv2.KeyPoint 0x7f833eb0b450>, < cv2.KeyPoint 0x7f833eb0b480>, < cv2.KeyPoint 0x7f833eb0b480>), [(235 :
> 'sponge' = {list: 8668} [((< cv2.KeyPoint 0x7f833d5e5740>, < cv2.KeyPoint 0x7f833d5e5770>, < cv2.KeyPoint 0x7f833d5e5770>), [(235 :
> 'stapler' = {list: 5084} [((< cv2.KeyPoint 0x7f833cd9d740>, < cv2.KeyPoint 0x7f833cd9d770>, < cv2.KeyPoint 0x7f833cd9d770>), [(235 :
> 'tomato' = {list: 5231} [((< cv2.KeyPoint 0x7f833c2cad30>, < cv2.KeyPoint 0x7f833c2cad60>, < cv2.KeyPoint 0x7f833c2cad60>), [(235 :
> 'toothbrush' = {list: 2968} [((< cv2.KeyPoint 0x7f8333d9aca0>, < cv2.KeyPoint 0x7f8333d9acd0>, < cv2.KeyPoint 0x7f8333d9acd0>), [(235 :
> 'toothpaste' = {list: 4067} [((< cv2.KeyPoint 0x7f83336f13b0>, < cv2.KeyPoint 0x7f83336f13e0>, < cv2.KeyPoint 0x7f83336f13e0>), [(235 :
> 'water_bottle' = {list: 5691} [((< cv2.KeyPoint 0x7f83314efd0>, < cv2.KeyPoint 0x7f83314ff000>, < cv2.KeyPoint 0x7f83314ff000>), [(235 :
> __len__ = {int} 10
> Protected Attributes
> 'ORB' = {dict: 10} {'rubber_eraser': [], 'scissors': [((< cv2.KeyPoint 0x7f834fd2ae80>, < cv2.KeyPoint 0x7f834fd2aeb0>, < cv2.KeyPoint 0x7f834fd2aeb0>), [(235 :
> 'rubber_eraser' = {list: 0} []
> 'scissors' = {list: 1507} [((< cv2.KeyPoint 0x7f834fd2ae80>, < cv2.KeyPoint 0x7f834fd2aeb0>, < cv2.KeyPoint 0x7f834fd2aeb0>), [(235 :
> 'shampoo' = {list: 2439} [((< cv2.KeyPoint 0x7f834cb7e0a0>, < cv2.KeyPoint 0x7f834cb7e0d0>, < cv2.KeyPoint 0x7f834cb7e0d0>), [(235 :
> 'soda_can' = {list: 3537} [((< cv2.KeyPoint 0x7f833eb0c360>, < cv2.KeyPoint 0x7f833eb0c390>, < cv2.KeyPoint 0x7f833eb0c390>), [(235 :
> 'sponge' = {list: 719} [((< cv2.KeyPoint 0x7f833d34f4b0>, < cv2.KeyPoint 0x7f833d34f4e0>), [(235 :
> 'stapler' = {list: 1714} [((< cv2.KeyPoint 0x7f833cd9e100>, < cv2.KeyPoint 0x7f833cd9e130>, < cv2.KeyPoint 0x7f833cd9e130>), [(235 :
> 'tomato' = {list: 320} [((< cv2.KeyPoint 0x7f833c2cbbd0>), [(235 :
> 'toothbrush' = {list: 1225} [((< cv2.KeyPoint 0x7f8333d9b8a0>, < cv2.KeyPoint 0x7f8333d9b8d0>, < cv2.KeyPoint 0x7f8333d9b8d0>), [(235 :
> 'toothpaste' = {list: 2951} [((< cv2.KeyPoint 0x7f83336f2280>, < cv2.KeyPoint 0x7f83336f22b0>, < cv2.KeyPoint 0x7f83336f22b0>), [(235 :
> 'water_bottle' = {list: 5669} [((< cv2.KeyPoint 0x7f83314ff930>, < cv2.KeyPoint 0x7f83314ff960>, < cv2.KeyPoint 0x7f83314ff960>), [(235 :
> __len__ = {int} 10
> Protected Attributes
> 'BRISK' = {dict: 10} {'rubber_eraser': [((< cv2.KeyPoint 0x7f8353734630>, < cv2.KeyPoint 0x7f8353734660>), [(235 :
> 'rubber_eraser' = {list: 783} [((< cv2.KeyPoint 0x7f8353734630>, < cv2.KeyPoint 0x7f8353734660>), [(235 :
> 'scissors' = {list: 2083} [((< cv2.KeyPoint 0x7f834fd2af40>, < cv2.KeyPoint 0x7f834fd2af70>, < cv2.KeyPoint 0x7f834fd2af70>), [(235 :
> 'shampoo' = {list: 3631} [((< cv2.KeyPoint 0x7f834cb7e190>, < cv2.KeyPoint 0x7f834cb7e1c0>, < cv2.KeyPoint 0x7f834cb7e1c0>), [(235 :
> 'soda_can' = {list: 3555} [((< cv2.KeyPoint 0x7f833eb0c660>, < cv2.KeyPoint 0x7f833eb0c690>, < cv2.KeyPoint 0x7f833eb0c690>), [(235 :
> 'sponge' = {list: 983} [((< cv2.KeyPoint 0x7f833d5e6bb0>), [(235 :
> 'stapler' = {list: 3623} [((< cv2.KeyPoint 0x7f833cd9db0>, < cv2.KeyPoint 0x7f833cd9dbf0>, < cv2.KeyPoint 0x7f833cd9dbf0>), [(235 :
> 'tomato' = {list: 1347} [((< cv2.KeyPoint 0x7f833c2cb360>, < cv2.KeyPoint 0x7f833c2cb390>, < cv2.KeyPoint 0x7f833c2cb390>), [(235 :
> 'toothbrush' = {list: 2217} [((< cv2.KeyPoint 0x7f8333d9ba50>, < cv2.KeyPoint 0x7f8333d9ba80>, < cv2.KeyPoint 0x7f8333d9ba80>), [(235 :
> 'toothpaste' = {list: 4031} [((< cv2.KeyPoint 0x7f83336f2c10>, < cv2.KeyPoint 0x7f83336f2c40>, < cv2.KeyPoint 0x7f83336f2c40>), [(235 :
> 'water_bottle' = {list: 5681} [((< cv2.KeyPoint 0x7f83314ffab0>, < cv2.KeyPoint 0x7f83314ffae0>, < cv2.KeyPoint 0x7f83314ffae0>), [(235 :
> __len__ = {int} 10
> Protected Attributes
```