Pollution Image Classification

Artificial Vision - Project

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Introduction

Objective

Develop a waste image <u>classification model</u> using artificial vision techniques



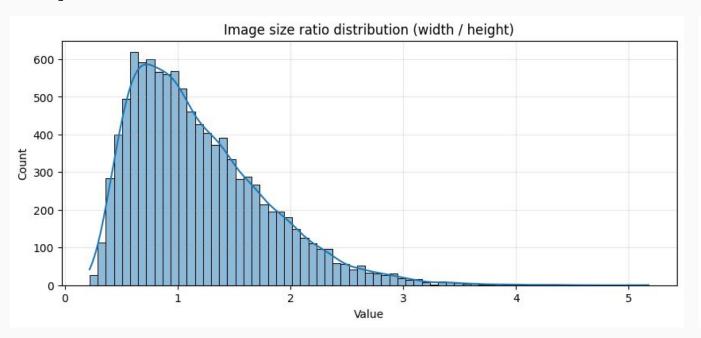
Dataset

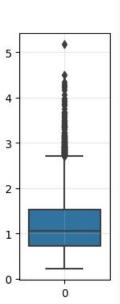
WaRPWaste Recycling Plant Dataset

- Sourced in Kaggle
- 28 distinct classes
- 10000 images



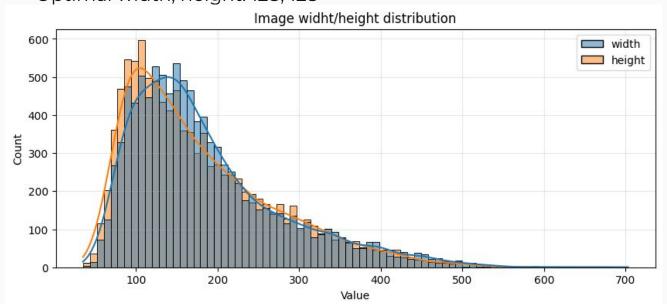
Aspect-Ratio Distribution

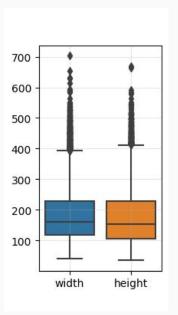




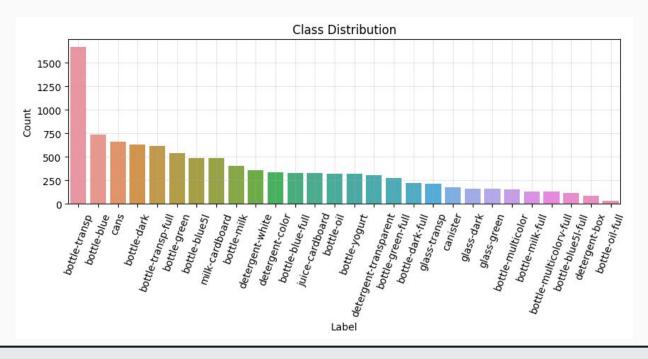
Images Width and Height Distribution

Optimal width, height: 128, 128

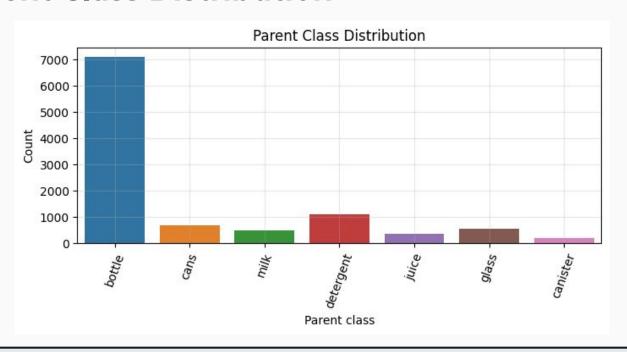




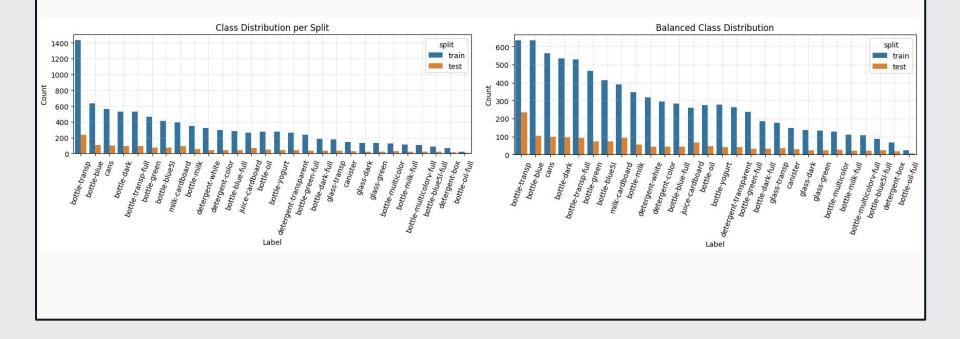
Class Distribution



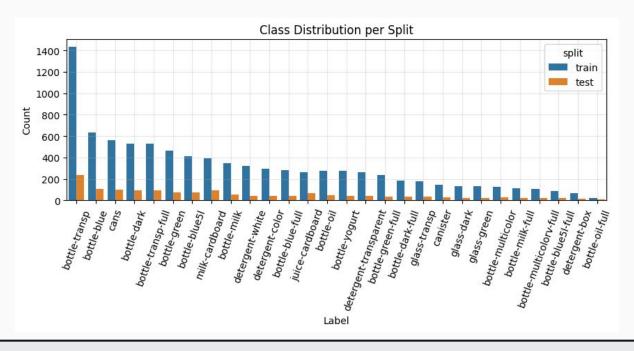
Parent Class Distribution



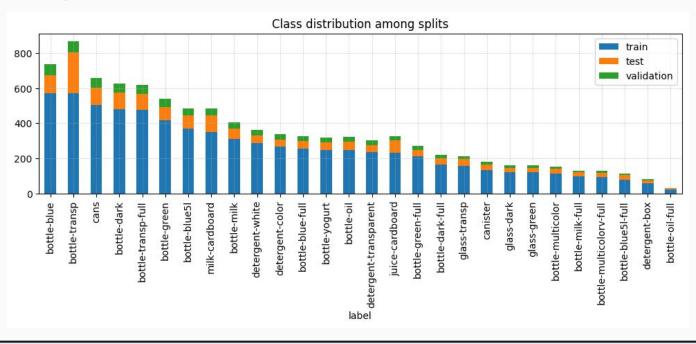
Balancing over populated class (bottle-transp)



Class Distribution Training, Test

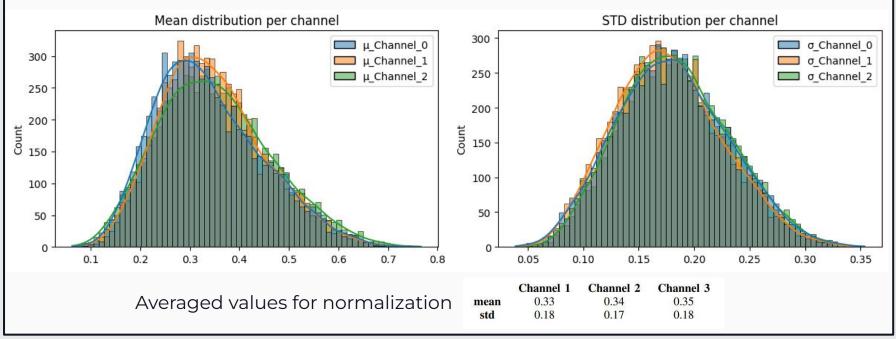


Training, test, validation split (75/15/10)



Color normalization

Computed mean and standard deviation for each channel



Color normalization

Averaged values for normalization:

	Channel 1	Channel 2	Channel 3
mean	0.33	0.34	0.35
std	0.18	0.17	0.18





bottle-multicolory-full

canister

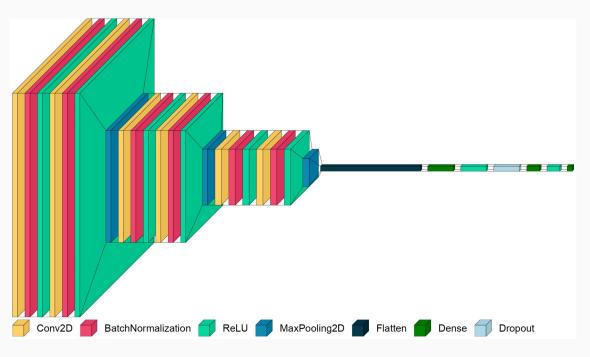
glass-dark

Dataset augmentation



Model

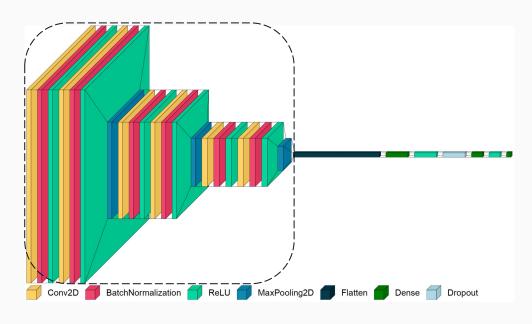
CNN Architecture



Model

Feature extraction layers

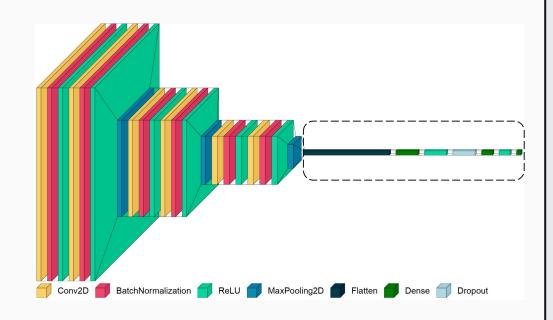
- Convolutional layers
- Batch normalization
- ReLU activation functions
- Max-pooling layers
- Dense layers



Model

Classification layers

- Flatten
- Dropout
- Dense layers
- Final Dense layers



Training

Hyperparameter tuning & Early Stopping

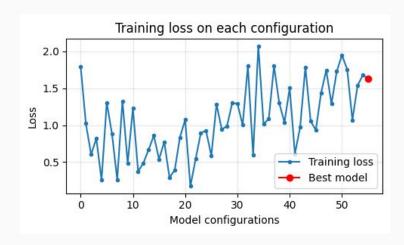
Learning rate	0.0005, 0.0001
Convolutional unit	16, 32, 64
Batch size	16, 64, 128
Dropout rate	0, 0.2
Weight decay	0.01, 0.001
Early stopping patience	15

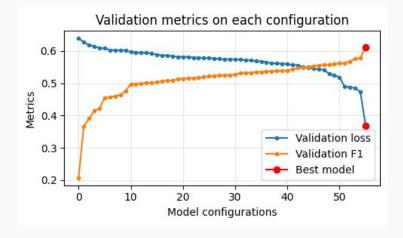
- Memory constraints: combinations with convolutional units 64 and batch size
 128 were excluded from the grid search
- Total hyperparameter combinations is 56

Training

Hyperparameter tuning & Early Stopping

Loss and metrics on validation set for every parameter combination





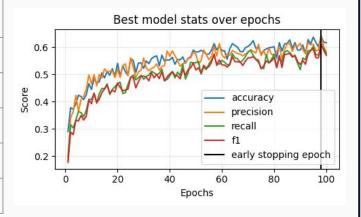
Training

Best model elected

Optimal model chosen on the highest F1-score achieved

Parameters									
Learning rate	0.0001								
Convolutional unit	32								
Batch size	16								
Dropout rate	0								
Weight decay	0.01								

Metrics	
Training loss	1.798
Validation loss	0.640
Validation precision	0.641
Validation recall	0.602
Validation f1-score	0.612



Performances

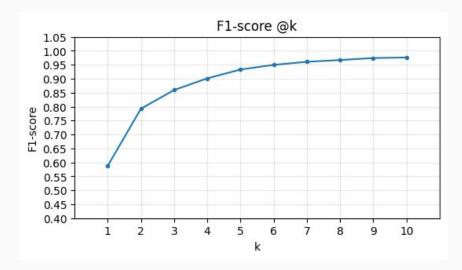
Test set classification report

bottle-blue-blue 0.55 0.68 0.61 104 bottle-blue-full 0.62 0.37 0.46 43 bottle-blue-blue-blue-blue-blue-blue-blue-bl
bottle-blue 0.55 0.68 0.61 104 bottle-blue5l-full 3 1 5 7 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
bottle-blue5l-full 0.62 0.37 0.46 43 bottle-dark-0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
bottle-blue5l 0.67 0.64 0.65 72 bottle-blue5l-full 0.78 0.29 0.42 24 bottle-dark 0.67 0.94 0.78 95 bottle-dark-full 0.76 0.56 0.64 34 bottle-dark-full 0.76 0.56 0.64 34
bottle-blue5l-full 0.78 0.29 0.42 24 bottle-green 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
bottle-dark 0.67 0.94 0.78 95 bottle-green-full - 0 0 0 0 0 9 22 0 0 0 1 0 0 0 0 1 0 0 0 0 0 0 1 0 0 0 bottle-dark-full 0.76 0.56 0.64 34 bottle-milk - 0 0 0 0 0 0 0 24 0 0 0 0 0 1 0 1 0 1 1 1 1 24 0 0 0 1 3
bottle-dark-full 0.76 0.56 0.64 34 bottle-milk-0 0 0 0 0 0 24 0 0 0 0 0 1 0 1 0 1 1 1 24 0 0 0 1 3
bottle-green 0.64 0.78 0.71 74 bottle-milk-full - 0 0 0 0 0 0 0 6 9 0 0 0 0 0 2 0 0 0 0 3 0 0 0 1 0
bottle-green-full 0.92 0.65 0.76 34 bottle-multicolor - 1 0 0 0 5 0 2 0 2 1 9 1 1 0 0 0 1 0 1 0 1 1 1 0 0 0 1 0
bottle-milk 0.39 0.42 0.41 57 bottle-multicolory-full - 0 0 0 4 0 0 0 0 0 14 0 0 0 2 0 0 0 0 0 0 0 0 1 0 bottle-milk-full 0.82 0.43 0.56 21 bottle-milk-full 0.82 0.43 0.56 21
bottle-multicolor 0.60 0.32 0.42 28 50 bottle-oil-full - 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
ittle-multicolory-full 0.70 0.67 0.68 21 5 bull-harm 15 1 4 0 2 0 2 0 2 0 2 0 2 0 2 0 2 0 2 0 2 0
bottle-transpfull 3 6 0 0 0 0 1 0 0 0 0 2 0 21 46 1 0 0 0 0 3 0 0 0 8 1 0
bottle-oil-full 0.00 0.00 0.00 8 bottle-rought 0.1 0.0 0.2 0.1 0.0 0.0 0.1 0.14 0.1 0.3 1.2 0.0 1.8 7
bottle-transp 0.60 0.60 0.60 234 capiter 1 0 0 0 3 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
bottle-transp-full 0.71 0.50 0.59 92
DOTTIE-YOGUTT
canister 0.71 0.33 0.45 30 determine clar 0.00 0.00 0.00 0.00 0.00 0.00 0.00 0.
cans 0.89 0.69 0.78 98
detergent-box 0.75 0.35 0.48 1/
detergent-color 0.4/ 0.51 0.49 43
detergent-transparent 0.35 0.22 0.27 41 -20
detergent-white 0.31 0.51 0.39 43
glass-dark 0.94 0.60 0.73 25 glass-transp-0 0 0 0 0 0 1 0 1 0 0 0 3 0 4 1 0 0 2 0 0 1 0 0 0 22 0 1
glass-green 0.74 0.68 0.71 25 juice-cardboard - 0 0 0 0 1 0 3 0 0 0 0 0 0 0 0 0 2 0 1 0 1 0 0 0 0 0 49 11
glass-transp 0.54 0.61 0.57 36 milk-cardboard - 0 0 0 0 0 0 0 0 0 0 0 0 1 0 3 1 0 0 0 0
101
juice-cardboard 0.60 0.60 0.60 0.60 0.60 0.60 0.60 0.6
accuracy 0.59 1551 dd - bottle - bottle - bottle - bottle - bottle - bottle - card dass see a gas see a ga
macro avg 0.60 0.52 0.54 1551 a g g g g g g g g g g g g g g g g g g
weighted avg 0.61 0.59 0.58 1551
de ter
Predicted Label

Performances

Model performance @ top-k

For this study k=5 was chosen



Performances

Model performance @ top-5

	precision	recall	f1-score	support
bottle-blue	0.99	0.95	0.97	104
bottle-blue-full	0.88	0.84	0.86	43
bottle-blue5l	0.91	0.97	0.94	72
bottle-blue5l-full	0.92	0.96	0.94	24
bottle-dark	0.99	0.99	0.99	95
bottle-dark-full	0.97	0.97	0.97	34
bottle-green	0.99	0.97	0.98	74
bottle-green-full	0.94	0.97	0.96	34
bottle-milk	0.98	0.96	0.97	57
bottle-milk-full	0.90	0.90	0.90	2
bottle-multicolor	0.92	0.82	0.87	28
oottle-multicolorv-full	0.78	0.86	0.82	2
bottle-oil	0.93	0.90	0.91	48
bottle-oil-full	0.44	0.50	0.47	8
bottle-transp	0.98	0.97	0.97	234
bottle-transp-full	0.91	0.90	0.91	92
bottle-yogurt	0.79	0.79	0.79	42
canister	0.71	0.73	0.72	30
cans	0.92	0.92	0.92	98
detergent-box	0.67	0.94	0.78	17
detergent-color	0.98	0.95	0.96	43
detergent-transparent	0.94	0.80	0.87	4:
detergent-white	0.84	0.98	0.90	43
glass-dark	0.96	1.00	0.98	25
glass-green	1.00	0.96	0.98	25
glass-transp	0.97	0.86	0.91	36
juice-cardboard	0.93	0.97	0.95	68
milk-cardboard	0.98	0.99	0.98	94
accuracy			0.93	155
macro avg	0.90	0.90	0.90	155
weighted avg	0.94	0.93	0.93	1551

bottle-blue - 99	5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-blue-full - 0	36	7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-blue5l - 0	0	70	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-blue5I-full - 0	0	0	23	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		- 200
bottle-dark - 0	0	0	0	94	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-dark-full - 0	0	0	0	0	33	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-green - 0	0	0	0	0	0	72	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-green-full - 0	0	0	0	0	0	0	33	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-milk - 0	0	0	0	0	0	0	0	55	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-milk-full - 0	0	0	0	0	0	0	0	0	19	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		150
bottle-multicolor - 0	0	0	0	0	0	0	0	0	0	23	5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-multicolory-full - 0	0	0	0	0	0	0	0	0	0	0	18	3	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-oil - 0	0	0	0	0	0	0	0	0	0	0	0	43	5	0	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-oil-full - 0	0	0	0	0	0	0	0	0	0	0	0	0	4	4	0	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-oil-full - 0 bottle-transp - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	226	8	0	0	0	0	0	0	0	0	0	0	0	0		
bottle-transp-full - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	83	9	0	0	0	0	0	0	0	0	0	0	0		- 100
bottle-yogurt – 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	33	9	0	0	0	0	0	0	0	0	0	0		
canister - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	22	8	0	0	0	0	0	0	0	0	0		
cans - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	90	8	0	0	0	0	0	0	0	0		
detergent-box - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	16	1	0	0	0	0	0	0	0		
detergent-color - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	41	2	0	0	0	0	0	0		
detergent-transparent - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	33	8	0	0	0	0	0		- 50
detergent-white - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	42	1	0	0	0	0		
glass-dark - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	25	0	0	0	0		
glass-green - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	24	1	0	0		
glass-transp = 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	31	5	0		
juice-cardboard - 0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	66	2		
milk-cardboard - 1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	93		- 0
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bottle-blue	bottle-blue-full	bottle-blue5l	1	bottle-dark	k-fu	bottle-green	n-fu	bottle-milk	bottle-milk-full	S	V-fu	bottle-oil	bottle-oil-full	bottle-transp	p-fr	bottle-yogurt	canister	cans	detergent-box	Col	age	vhi	glass-dark	glass-green	glass-transp	og	oal		
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	pot	Q	bottle-blue5I-full		bottle-dark-full	-0	bottle-green-full		bot	bottle-multicolor	nu		Q	ā	bottle-transp-full	ğ			ge	detergent-color	H	detergent-white			0,	uice-cardboard	milk-cardboard		
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											bottle-multicolorv-full										detergent-transparent								
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GradCAM

How to?!??!?!?

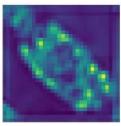
```
def evaluate gradcam(image, label, path):
    pred = model(image, register hook=True)
    # label pred = pred.argmax(dim=1)
    labels pred = pred.topk(5)
    pred.sum().backward()
    gradients = model.get activations gradient()
    pooled gradients = torch.mean(gradients, dim=[0, 2, 3])
    activations = model.get activations(image).detach()
    for i in range(64):
        activations[:, i, :, :] *= pooled gradients[i]
    heatmap act = torch.mean(activations, dim=1).squeeze()
    heatmap act = np.maximum(heatmap act.cpu(), 0)
    heatmap act /= torch.max(heatmap act)
    image rgb = cv2.imread(path)
    heatmap = cv2.resize(heatmap act.numpy(), (image rgb.shape[1], image rgb.shape[0]))
    heatmap = np.uint8(255 * heatmap)
    heatmap = cv2.applyColorMap(heatmap, cv2.COLORMAP JET)
    superimposed img = heatmap * 0.5 + image rgb
    superimposed img = np.uint8(superimposed img)
    superimposed img = cv2.cvtColor(superimposed img, cv2.COLOR BGR2RGB)
    return image rgb, heatmap act, superimposed img, labels pred
```

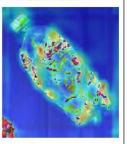
GradCAM

Correct predictions

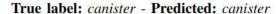
True label: bottle-blue - Predicted: bottle-blue



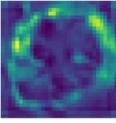


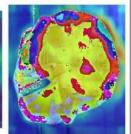


Class	Probability
bottle-blue	0.992
bottle-blue-51	0.003
bottle-transp	0.002
bottle-oil	0.002
bottle-green	0.001





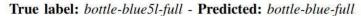




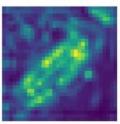
Class	Probability
canister	0.805
detergent-white	0.146
bottle-milk	0.045
bottle-multicolor	0.003
detergent-transparent	0.001

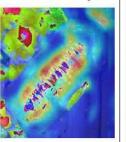
GradCAM

Incorrect predictions





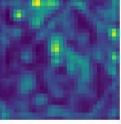




Class	Probability
bottle-blue-full	0.625
bottle-blue51-full	0.368
bottle-blue	0.004
bottle-transp-full	0.003
bottle-green-full	0.001

True label: milk-cardboard - Predicted: juice-cardboard







Class	Probability
juice-cardboard	0.601
milk-cardboard	0.362
detergent-white	0.037
bottle-milk	0.001
bottle-green	i0.001

GradCAM on Video

Realtime GradCAM on Video

