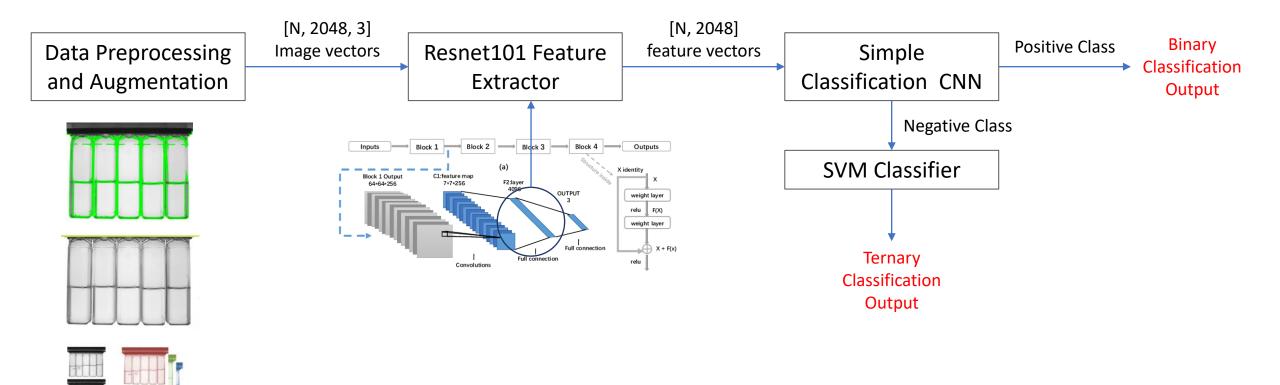


# **Proposed Architecture**

**RGB** 

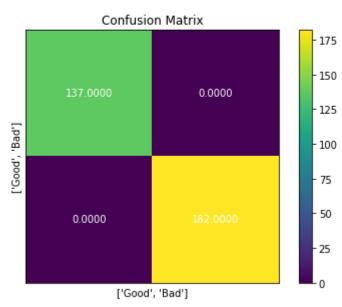
greyscale



# **Performance Metrics**

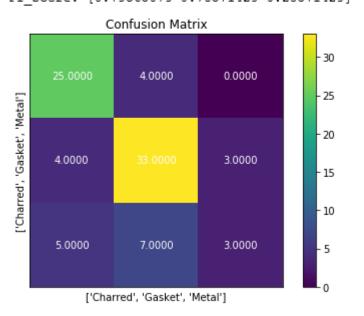
### **Binary Classification Metrics**

Acuuracy: 1.0 Precision: 1.0 Recall: 1.0 F1\_score: 1.0 Specificity: 1.0



### **Ternary Classification Metrics**

Acuuracy: 0.7261904761904762
Precision: [0.73529412 0.75 0.5
Recall: [0.86206897 0.825 0.2 ]
F1 score: [0.79365079 0.78571429 0.28571429]



# Design choices

- Tried PCA first for feature extraction and Initial SVM for Classification with PCA reduced feature set.
   Poor results.
- Hints from the organizers and the paper suggested Neural Networks for Feature Extraction, which seems to be a common method.
- 3 channel grayscale to RGB conversion reduces dataset size and is compatible with the Resnet101, packing thrice the amount of information into one tensor.
- Tried a CNN for Ternary Classification, poor results due to limited samples. Switched to Linear Kernel based SVM.
- SVMs perform well for smaller datasets.
- Applied Further Data Augmentation such as mask-based cropping. Accuracy decreased due to remaining artefacts from the mask-cropping.
- Realized that static parts of the image do not play a role in the learning outcome of the models. So
  mask-cropping not entirely necessary for the sole purpose of size reduction.
- Optimized training for Validation Loss/Validation accuracy.

# **Additional Information**

### Libraries Used

 Pytorch, Keras and Scikit-Learn for Neural Network building, Classical Machine Learning models and for Classification Metrics

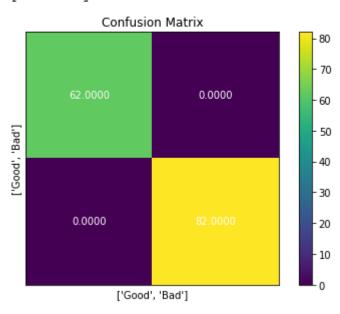
## What Did I Learn

- Industrial application example of Computer Vision.
- How to use state of the art Neural Networks for Feature Extraction.
- Solidified knowledge in Transfer Learning.
- Combining Classical Machine Learning and Deep Learning.
- Image Processing aspects for Masking Region of Interests.
- Differences between normal object detection and miniscule object detection methods.
- Understanding where not to use CNNs.
- Learn about Binary Entropy Losses and Categorical Entropy Losses

## **Test Dataset Performance Metrics**

### **Binary Classification Metrics**

Acuuracy: 1.0 Precision: 1.0 Recall: 1.0 F1\_score: 1.0 Specificity: 1.0



### **Ternary Classification Metrics**

Acuuracy: 0.7419354838709677

Precision: [0.888888889 0.67647059 0.7 Recall: [0.66666667 0.88461538 0.58333333] F1 score: [0.76190476 0.76666667 0.63636364]

