

EDIBLE NUT DETECTION

COMPUTER VISION PROJECT
FINAL PRESENTATION

Mihir Mulye

Table of Contents

- Problem Statement
- Assumptions & required material
- Possible approaches
- Proposed pipeline
- Implementation
- Results
- Failures, Learnings and Scope
- References

Problem Statement

- Detect objects (edible nuts) in a video depicting a roll of objects
- Categories of nuts- Peanuts, Walnuts and Hazelnuts
- Requirements-
 - Desired category of nuts should be detected in a stationary frame (that frame where the configuration first attains a stationary state)
- Observed variations in collected data-
 - Partial occlusions
 - Variable lighting conditions
 - Distractors (objects not of interest) outnumbering the desired number of objects
 - Variable camera configurations resulting in different perspectives

Assumptions

- Distractors do not outnumber the nuts
- Complete occlusion is not dealt with

Required materials

- TensorFlow 1.14
- OpenCV 3.4
- Lab recordings
- LabelMe (annotation tool)

Possible approaches (1 of 2)

- Object detection can be dealt with using either classical vision techniques or by deep learning methods
- Classical methods:
 - Template matching [3]
 - SIFT [4]
- Deep learning methods:
 - SSD [1]
 - Lightweight network able detect objects without much processing requirement

Possible approaches (2 of 2)

- Classical methods pros & cons:
 - Fast to implement as no training needed
 - Classical methods need hand crafting of features
 - Fail to generalize and have difficulties handling variation in data
- Deep learning pros & cons:
 - Able to generalize relatively better than classical approaches
 - Need annotated data
 - Might overlearn if improper data is used for training
- Based on the pros and cons, best of both approaches can be combined to get good results

Proposed Pipeline

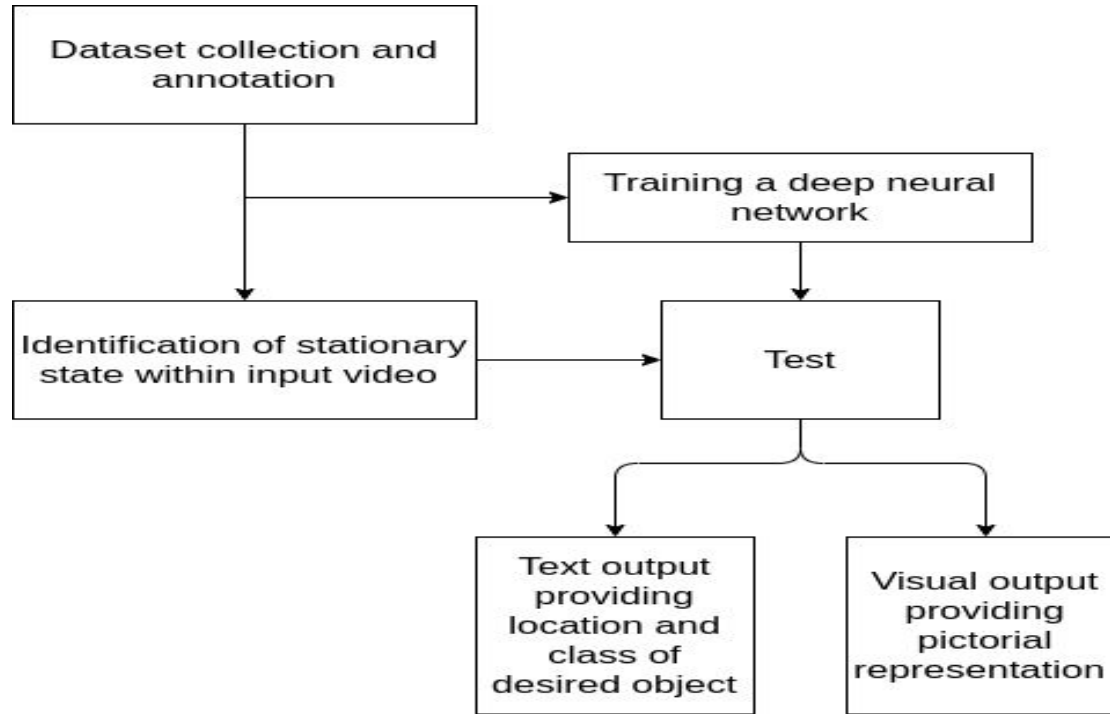


Figure 1 : Overview of proposed pipeline

Implementation (1 of 3)

- Combination of classical Computer Vision and Deep Learning has been employed to accomplish the tasks
- Identification of stationary state within input video:

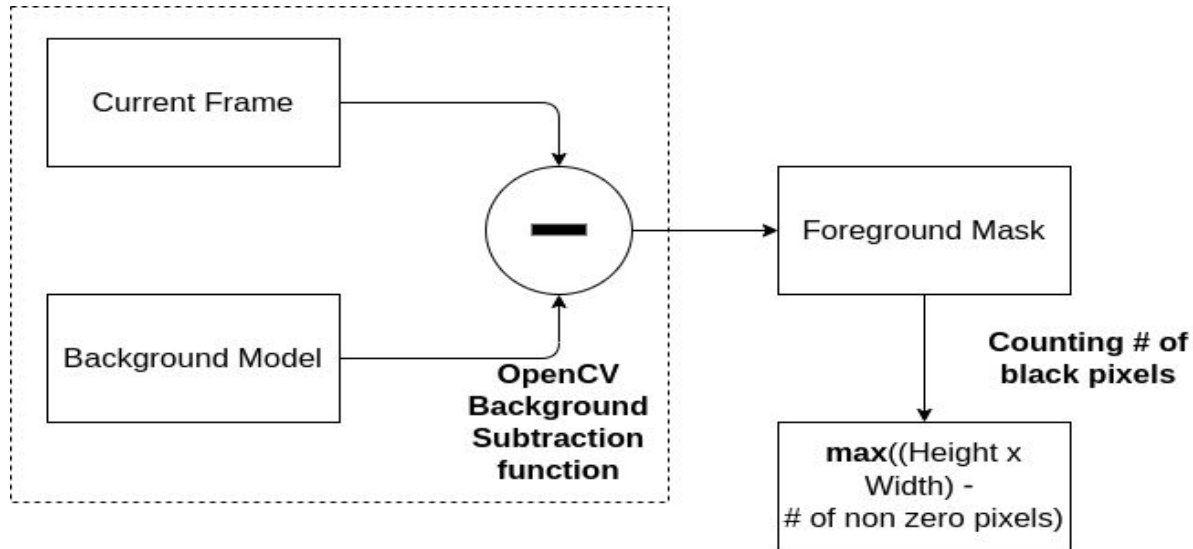


Figure 2 : Index of stationary frame is extracted at the end of this process [2]

Implementation (2 of 3)

- SSD Mobilenet [1] selected for the object detection

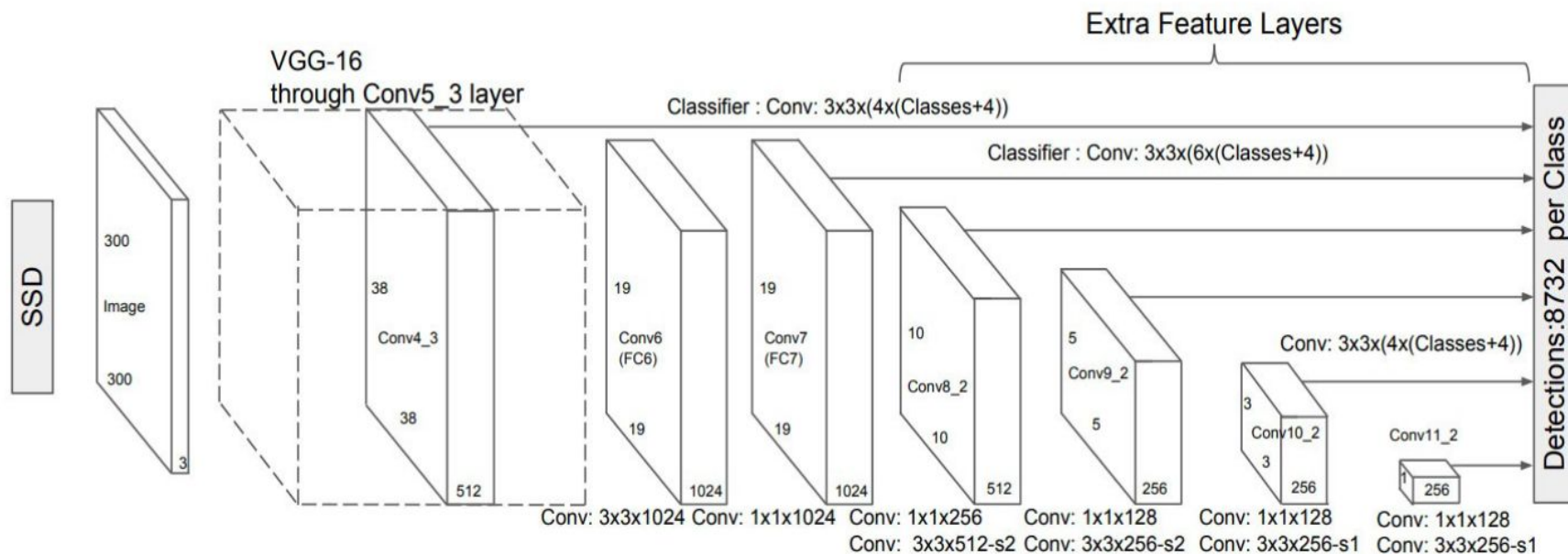
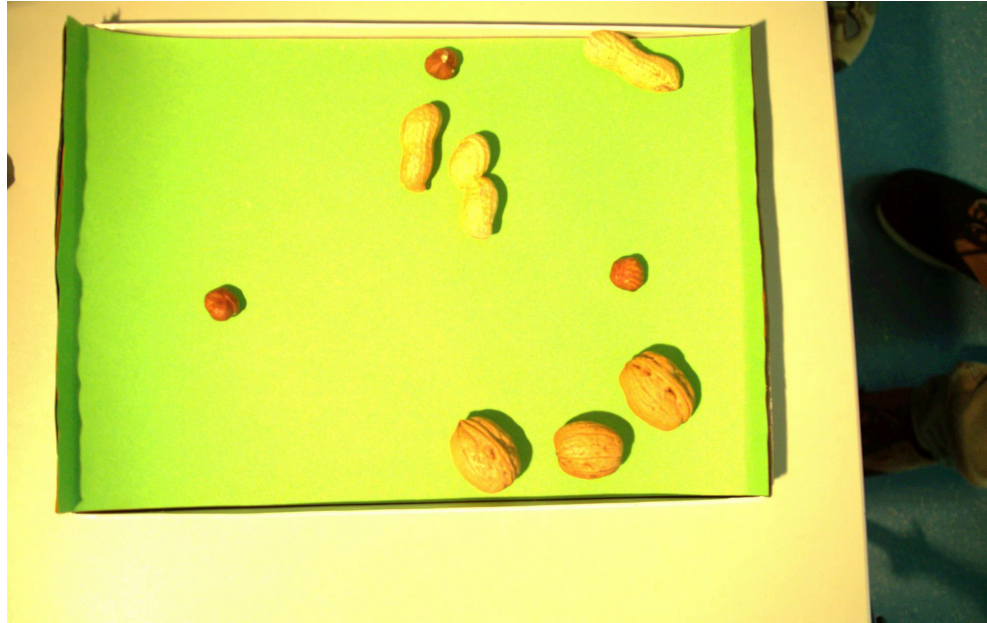


Figure 3 : Single Shot Multibox Detector overview [1]

Implementation (3 of 3)

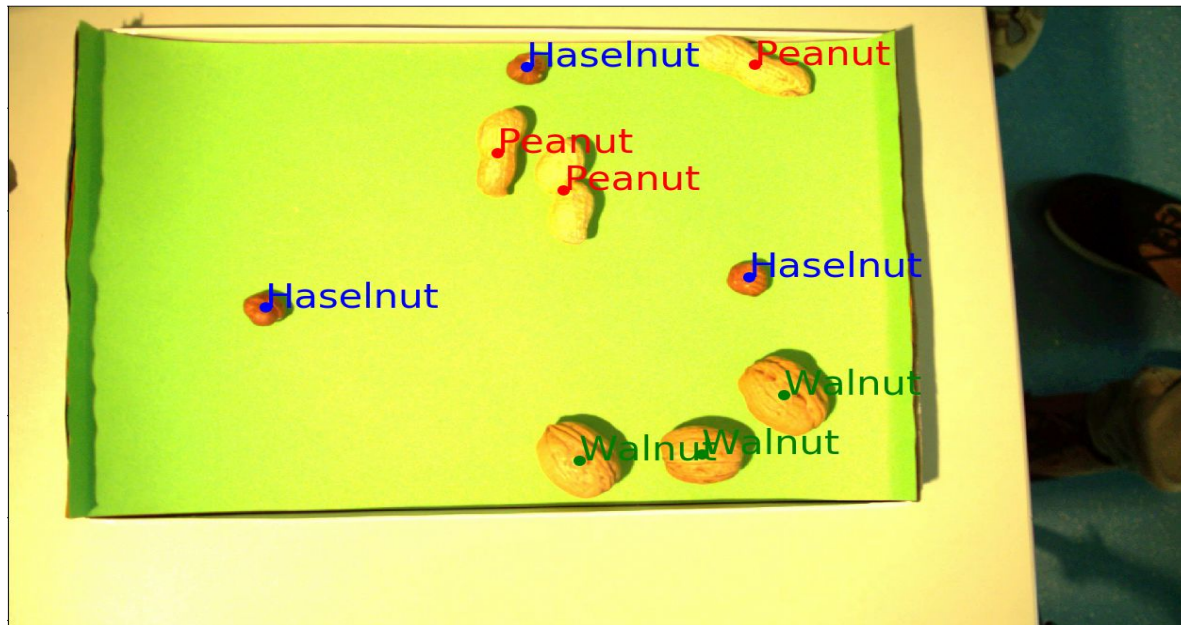
- Overview of SSD working
 - Initial phase extracts feature from input image (edges, colors)
 - Final phase classifies object based on extracted features
- Output from the 'stationary state identification' fed as input to SSD
- SSD framework yields class and the bounding box coordinates of the edible nuts
- Centroid of the bounding box coordinates calculated and output written to a csv file

Results 1 of 6:



**Figure 4 : Stationary frame
(Proper Illumination condition)**

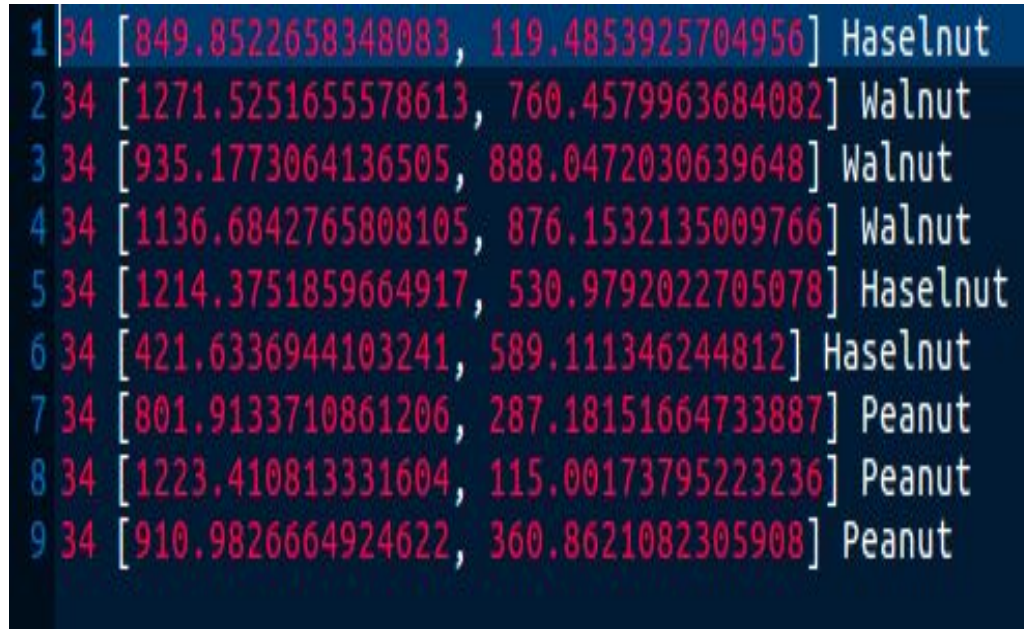
Results 2 of 6:



- Red : Peanut
 - Actual : 3
 - Detected : 3
- Green : Walnut
 - Actual : 3
 - Detected : 3
- Blue : Haselnut
 - Actual : 3
 - Detected : 3

**Figure 5 : Output frame
(Proper Illumination condition)**

Results 3 of 6:



```
1 34 [849.8522658348083, 119.4853925704956] Haselnut
2 34 [1271.5251655578613, 760.4579963684082] Walnut
3 34 [935.1773064136505, 888.0472030639648] Walnut
4 34 [1136.6842765808105, 876.1532135009766] Walnut
5 34 [1214.3751859664917, 530.9792022705078] Haselnut
6 34 [421.6336944103241, 589.111346244812] Haselnut
7 34 [801.9133710861206, 287.18151664733887] Peanut
8 34 [1223.410813331604, 115.00173795223236] Peanut
9 34 [910.9826664924622, 360.8621082305908] Peanut
```

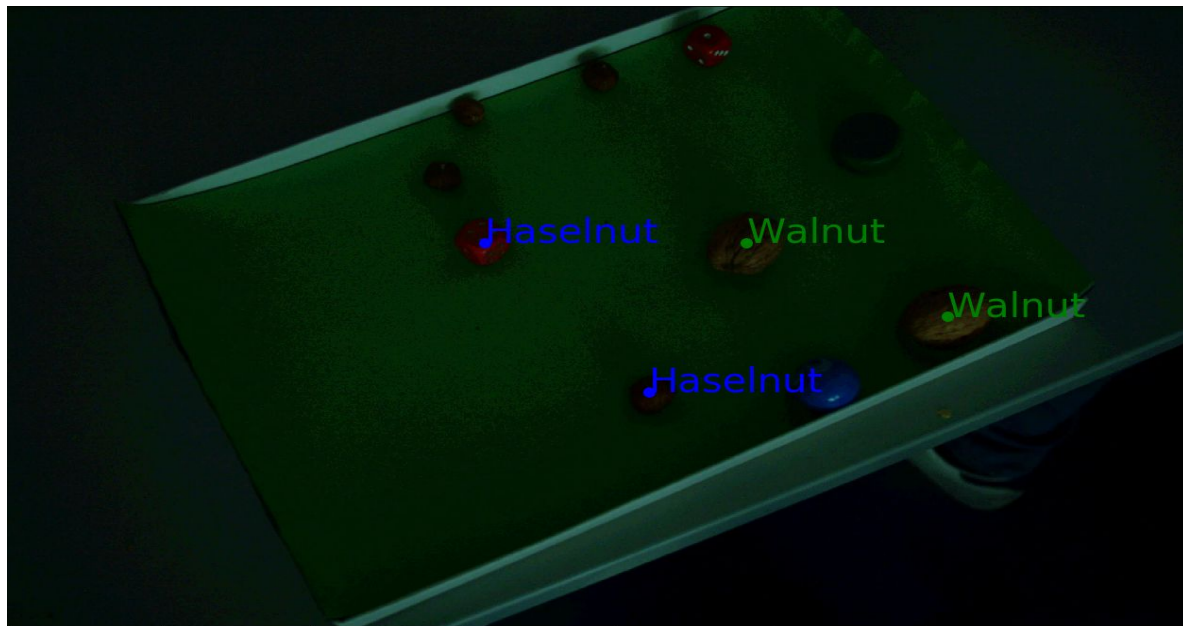
**Figure 6 : Stationary frame text output
(Proper Illumination condition)
Origin : Top Left corner**

Results 4 of 6:



**Figure 7 : Stationary frame
(Improper Illumination condition)**

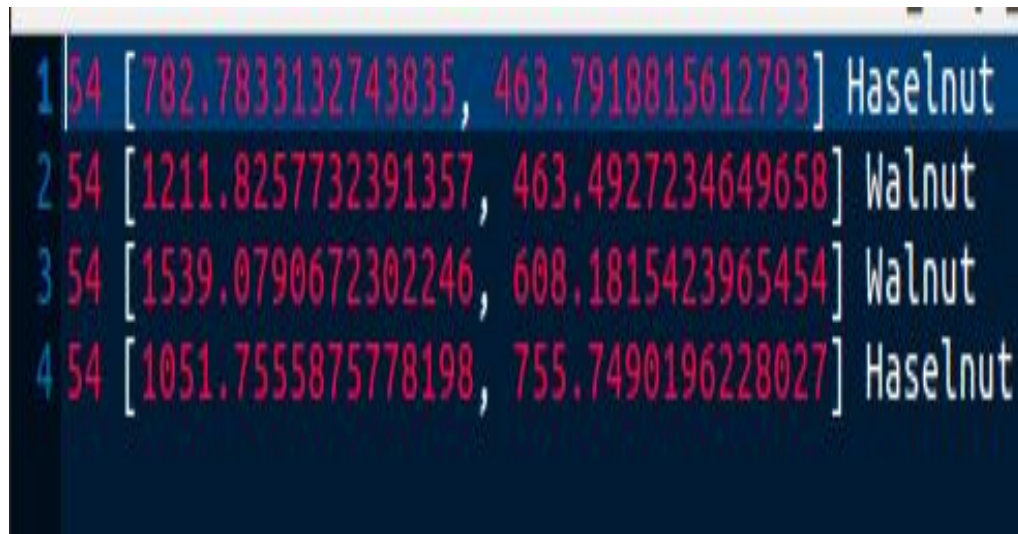
Results 2 of 6:



- Red : Peanut
 - Actual : 0
 - Detected : 0
- Green : Walnut
 - Actual : 2
 - Detected : 2
- Blue : Haselnut
 - Actual : 4
 - Detected : 2 (1 incorrect detection)
- Incorrectly detected red die as Haselnut

**Figure 8 : Output frame
(Improper Illumination condition)**

Results 3 of 6:



```
1 54 [782.7833132743835, 463.7918815612793] Haselnut
2 54 [1211.8257732391357, 463.4927234649658] Walnut
3 54 [1539.0790672302246, 608.1815423965454] Walnut
4 54 [1051.7555875778198, 755.7490196228027] Haselnut
```

**Figure 9 : Stationary frame text output
(Improper Illumination condition)
Origin : Top Left corner**

Failures, Learnings and Scope-

- Approaches such as-
 - Shape detection
 - Template matching
 - Color identification

were explored before switching to deep learning

- Owing to drastic variations in data, the methods were failing to generalize
- Static frame not always correctly isolated
- Spurious detections need to be addressed
- Bigger network can be used in place of lighter SSD
- More data can be used for training

References

- [1] Liu, Wei, Dragomir Anguelov, Dumitru Erhan, Christian Szegedy, Scott Reed, Cheng-Yang Fu, and Alexander C. Berg. "Ssd: Single shot multibox detector." In *European conference on computer vision*, pp. 21-37. Springer, Cham, 2016.
- [2] https://docs.opencv.org/3.4/d1/dc5/tutorial_background_subtraction.html
- [3] https://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_imgproc/py_template_matching/py_template_matching.html
- [4] https://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_feature2d/py_sift_intro/py_sift_intro.html