

## MIDS W251 Spring 2020

### Section: Monday 6:30pm Suhas Gupta

#### HW7 – Neural face detection pipeline

1. The homework solution was implemented in a similar architecture as HW3 with the openCV face detector (used in HW3) replaced by the **Tensor Flow Face Detector** [1]. Total 5 docker containers were created with custom built images to create the modular architecture shown in Fig1.

- The custom images were built from the Docker files that are noted with the Docker keyword in Figure 1 for each container
- *Docker-compose* was used to setup a swarm of services to deploy the face detection application. The *docker-compose.yml* file contains the deployment parameters.
- The code implemented in this homework solution is available on the following GitHub repo:

GitHub Repo Path: <https://github.com/suhasgupta791/v2/tree/master/week07/hw>

Repo Clone URL: <https://github.com/suhasgupta791/v2.git>

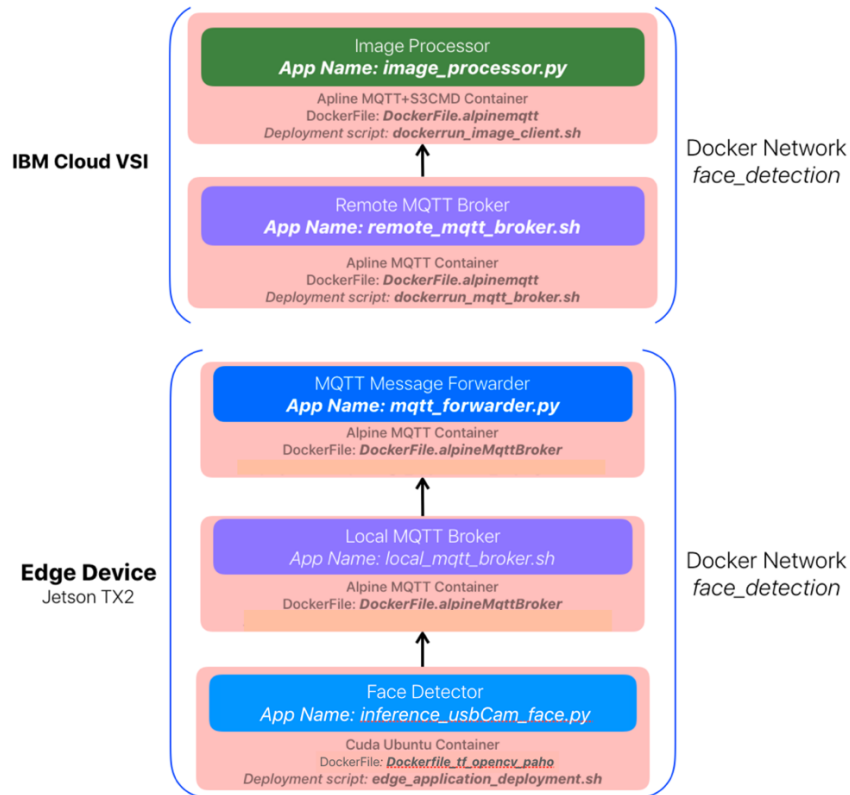
- The local (Jetson TX2) application scripts are located at:  
[https://github.com/suhasgupta791/v2/tree/master/week07/hw/jetson\\_local\\_scripts](https://github.com/suhasgupta791/v2/tree/master/week07/hw/jetson_local_scripts)
- The remote (IBM VSI) application scripts are located at:  
[https://github.com/suhasgupta791/v2/tree/master/week07/hw/ibm\\_vsi\\_remote\\_scripts](https://github.com/suhasgupta791/v2/tree/master/week07/hw/ibm_vsi_remote_scripts)

- The images received at the remote cloud host are stored in the following object storage:

Bucket Name: **w251-suhasgupta-hw07**

host\_base = **s3.us-south.cloud-object-storage.appdomain.cloud**

host\_bucket = **%(w251-suhasgupta-hw07)s.s3.us-south.cloud-object-storage.appdomain.cloud**



## 2. Face detection solution

- a. A MobileNet SSD (single shot multibox detector) based face detector with pretrained model provided, powered by TensorFlow [object detection api](#), trained by [WIDERFACE dataset](#). It achieves an accuracy of up to **98%** for detecting human faces.
- b. It does achieve very high accuracy on human faces in empirical tests.
  - a. However, a major limitation of the model seems to be coming from using the “eyes” as a discerning feature of a face. While closing the eyes or covering them, the network is unable to detect the human face.
  - b. From this empirical result, we can conclude that the training dataset did not have any images of human faces with closed or covered eyes.
  - c. Such a model would not be preferable to use in a production system where there is a high probability of the camera not being able to detect human eyes with enough resolution to result in proper face detection.
- c. It achieves frame rates of up to 50 frames/second. The bottleneck is the video rendering time of the hardware used (camera, display port link etc.)
- d. TensorFlow based detector is better in quality (higher frame rates, higher accuracy) when compared to the OpenCV implementation. It is easy to improve its accuracy with a wider training dataset. The inference is fast and high frame rates can be achieved compared to the OpenCV implementation.