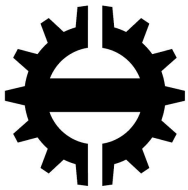

TAMPERE
UNIVERSITY OF
TECHNOLOGY

TUT Embedded Smile Detector

NVIDIA Jetson Developer Challenge

Feb 2018

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1. Introduction

The original Idea was to extend facial expression recognition to people's lives in order to prevent threatening behaviours .

Here we describe some possible use cases:

I. In an interview situation, the interviewee can be monitored to detect his moods and level of stress during the interview. So, the interviewers can act accordingly to interviewee’s feelings.

II. Quantitative evaluation of depression.

III. Customer satisfaction during test drive of a new car.

2. Implementation

2.1 Implementation

For implementing this system on an embedded device namely Jetson TX2 we have used multithreading since computational overheads from different modules of the system would cause inconsistency in the execution flow of the application. The main thread is responsible for handling the communication between different threads and modules. The general flow of the program could be explained as this: firstly we are grabbing frames in the main thread and share them between all threads using a data structure which all threads have access to it. Secondly, we are detecting faces in the captured frames and assigning coordinates of the detected faces to the shared data structure. Then the main thread is retrieving the prediction results of smiles in those coordinates form the smile detector thread. In the final step, we sketch the frame with the prediction result added to it.

We have utilized a well-known training architecture Mobilenet (Howard, A.G. 2017.) with different parameters and in the smile detector thread we are using it to predict smile probabilities. Mobilenets (with alpha = 0.25, 0.5, 0.75, 1.0) are based on a streamlined architecture that uses depth-wise separable convolutions to build light weight deep neural networks for mobile and embedded vision applications. Adjustable hyper-parameter 'alpha' allows the model builder to choose the right sized model for their application based on the constraints of the problem.

Handling multiple faces for smile detection was carried out by accessing coordinates of rectangles of detected faces in a frame by the smile detector thread and assigning smile probabilities to each rectangle. So while sketching the frame we will be able to draw the rectangle and the prediction result alongside each other.

The communication between different threads are asynchronous, so they do not need to be available at the same time. Since all threads are running almost at the same pace, we have a stable system which is able to draw approximately 18 frames ps.

2.2 Workflow

The current workflow of the application could be described as in the following sequence diagram.

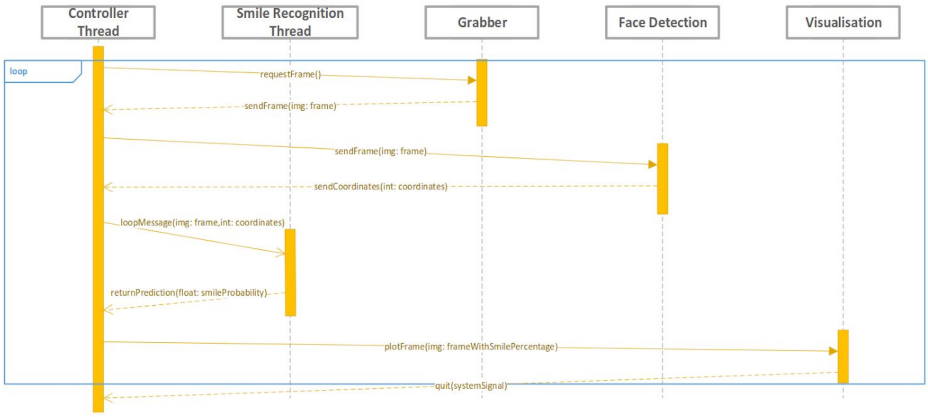

Figure 2.1: Sequence diagram

2.3 Environment and libraries

TX2 jetson owner is Tampere university of technology, signal processing lab and they lent us the device for this project. We also used Logitech webcam c920 as a camera.

Tools(libraries, apps, softwares)	**Version**	**Explanation**
Ubuntu	16.04 LTS	Operating system-64 bit.
Python	2.7.12	Programming language.
OpenCV	2.4.13.1	Library which is mainly use for computer vision.
Keras	2.1.1	A neural network library.
Tensorflow	1.3.0	A library for dataflow programming across a range of tasks including machine learning.
Numpy	1.11.0	A library for scientific computing in python.

3. Conclusion

In conclusion, we gained the hands-on experience of implementing an idea from the scratch to a working application.

As future works, we have plans as follows:

- Pretrain model with other datasets to make it more robust.

- Detecting other facial expressions than smiling.

- Consider experts' ideas to include diverse domain knowledge to make this application more practical.

- Improve thread programming.

In the end, we would like to say thanks to our university teacher Heikki Huttunen for his support during the project.

4. References

Howard, A.G., Zhu, M., Chen, B., Kalenichenko, D., Wang, W., Weyand, T., Andreetto, M. and Adam, H., 2017. Mobilenets: Efficient convolutional neural networks for mobile vision applications. *arXiv preprint arXiv:1704.04861*.

5. Annexes

5.1 Images

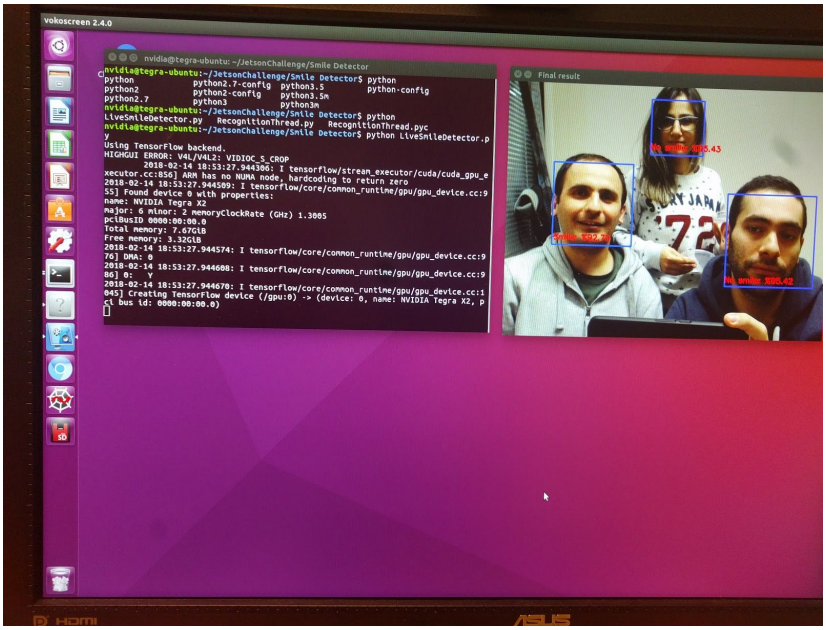

Figure 5.1: Creating the demo

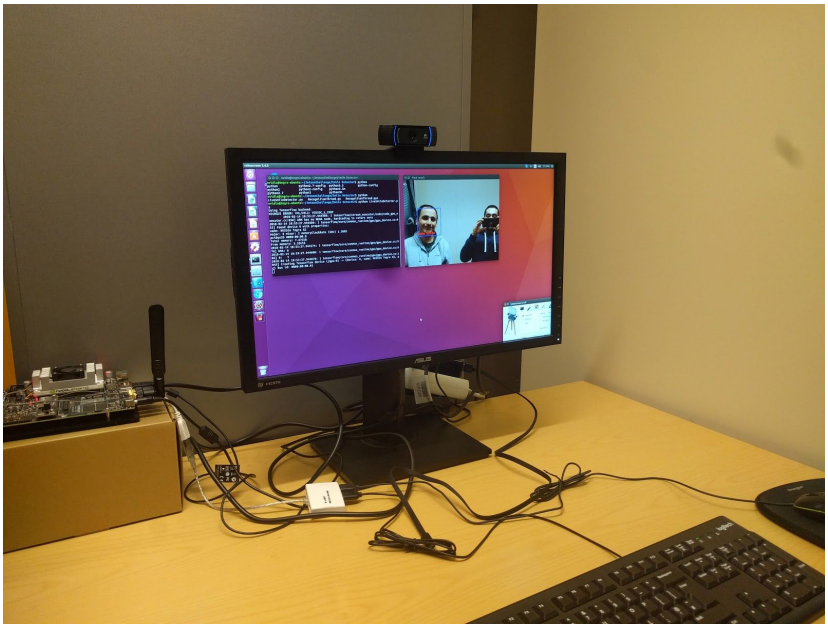

Figure 5.2: Hardware

5.2 links

Link to the public repository: https://github.com/alitakin/JetsonChallenge

Link to the demo: https://www.youtube.com/watch?v=4JGatQOchFo