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Research Project (Thesis) MOD3-03

Feature Development Kit [FDK] – Plugin Development and Continuous Integration Maintenance

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# Abstract

According to National Institute on Deafness and Other communication Disorders(NIDCD) 7.7 percent of children between the ages of 3 to 17 in the US alone have had a disorder related to voice, speech, language or swallowing with in the duration of 12 months of the report being published[1]. The idea behind project AVATAR is to develop a standalone logopedics system which can be utilized for speech therapy for children suffering from these speech impediments. For the first part of this project the objective is to extract visual information about the shape and movement of the lips and the jaw that is associated with the pronunciation of the word and attempt to infer the word that has been said regardless of any accent or mispronunciation present. This could then be used as a base and stepping stone for further development towards a complete system for speech therapy.

With the advancements in mobile devices and their wide availability and popularity, android platform has been chosen for the development of the prototype application. The application developed employs several image processing techniques to extract the relevant information which is then passed to a remote computer for further processing that is required to infer the words or sentence said using only visual data (lip reading).

The scope of the project involves research and evaluation of the useable image processing methods, development of the application using the results from phase-1, performance optimizations to achieve a near real-time processing and evaluation and analysis of the results.

**Keywords:** Image Processing, face detection, facial landmarks detection, speech therapy, logopedics, OpenCV, dLib.

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# Abbreviations

API – Application Programming Interface

JNI – Java Native Interface

NDK – Native Development Kit

IDE – Integrated Development Environment

UI – User Interface

ROI – Region of Interest

XML – eXtensible Markup Language

ROS – Robot Operating System

# Introduction

According to the findings of National Health Interview Survey [²] nearly 8 percent of the children aged 3-17 in the US had a communication disorder with in the past 12 months of the study, and among these children only 55% received any intervention service during the time period of 12 months. These statistics show that the problem of speech impediment is very prevalent in children and young people while almost half of the people suffering from this problem receive any help.

The motivation behind this project is to develop a standalone system which can be used by laymen without any extra help to address certain speech impediments, and the most important part is that the system should be easily and widely available for everyone.

In order to achieve this objective, a smartphone application on the android platform has been chosen as the medium for the first phase of the project. The reason behind this is that in recent years smart phone industry has seen a large boom and by 2020 there are expected to be 3.5 billion smartphone users worldwide [3], apart from this the advancement in technology has made the smart phones very capable and tasks that required specialized equipment before can be performed by a simple application running on the smart phone. The introduction of wearable tech like smart watches and other health monitors that can be connected to a smartphone via Bluetooth and the data analysed by a paired app have also made diagnostics and health monitoring very easy and does not require highly skilled medical personnel to perform these tasks as in the past.

The first part of the project deals with the development of the android application. This part has further been divided into four phases for better management and implementation. *Phase-1* was research and performance evaluation of the methods which could be used to develop the application. *Phase-2* was the development of the application to the point that it could use the on-board camera to capture live images and perform image processing tasks to extract useful information. *Phase-3* was to establish real-time communication between the mobile application and a remote PC to transfer the extracted data to a PC to perform the resource extensive processing that is required to infer the words from lip reading. *Phase-4* was implementation of the Matlab scripts running on the remote PC using Hidden Markov Model and its training using data set developed using the android application. In the future the tasks done in phase-3 and 4 would be moved to be performed on the device itself or on the cloud to enable portability.

For image processing OpenCV’s android SDK is used in conjunction with dlib c++ library to extract facial landmarks. The details of the implementation are provided in the following sections.

# Theory

* 1. Literature Review

Digital Image Processing is the collection of techniques used to manipulate images to extract some results that might be of interest out of the image for example object detection, classification etc. Image processing itself can be further sub-divided into sub.-classes lie image analysis, image restoration, image enhancement etc. In the past decades the usage of these digital image processing techniques in production applications has increased tenfold. The applications of image processing range from medical technology, automated driving to space technology.

The following sections provide a brief overview of the state of art of the techniques that have been used or could be used in this project.

* 1. Image Processing Techniques – Theory

In the scope of this project object detection and classification techniques have been used to extract facial features from the images captured from the device camera which are then further used for image analysis using machine learning techniques:

* + 1. Face Detection

Face detection is the first step towards all facial analysis algorithms which in turn play a huge role in today’s technologies that interact with human’s; these facial analysis algorithms can range from facial recognition, authentication/security applications, lip reading and expression classification. Face detection is an easy task for human beings but for machines considering all the variable could prove to be a difficult task. The scale, orientation, pose and the lighting conditions of the image are among the many variables that need to be accounted for in order to perform a successful facial detection. There are multiple ways that have been a topic of research for performing face detection. They can be divided into four main methods: (i) Feature-based (ii) Appearance based (iii) Knowledge based (iv) Template matching. Out of these the more popular ones are feature based and knowledge based. For the first technique face detection can be done by using self-defined features in conjunction with HAAR cascade machine learning algorithm which is the techniques used in OpenCV’s face detection algorithm. The other technique is to use neural networks and train them on a huge dataset of faces, which are readily available now, one of the most influential face databases is the FERET or COLOR FERET database [5]. Tensor Flow is an open source platform for machine learning which is commonly used for research and applications involving deep neural networks. Tensor flow allows users to create dataflow graphs and the training API extracts features automatically from the training dataset which allows for detection and classification of objects of interest.

In 2001 P.Viola and M.Jones published a paper describing what is now known as Viola-Jones (V-J) Object detection framework. V-J algorithm is one of the first real time object detection algorithms and provides very high detection rate and very less percentage of false positives.

The algorithm has some limitations related to the pose of the face but the classifiers can be trained to recognize slight tilts in the face. V-J detector utilizes Haar-like features that are a scalar product between the image and some Haar-like templates. Figure XX shows the Haar-like templates used in the V-J algorithm and also employed by OpenCV. The Haar features signify that almost all the human faces share similar visual outlooks for example the bridge of the nose is lighter than the eyes which can be defined by the Haar template figure XX(b-1) and the eyes region is darker than the upper cheeks and which is defined by inversion of Haar template of fig. (XX-a-1). This way the regions of interest i.e. different features on a face can be detected using the gradients in pixel intensities. The V-J algorithm uses a 24x24 window as the base window size to start the evaluation of these features which gives rise to up to 160,000+ features to be calculated. The steps put forward by the algorithm brings this immense calculation down to a smaller size. The steps involved after selection of appropriate Haar-Feature templates in V-J algorithm are [7]:

* Image Integral and feature extraction
* AdaBoost Learning
* Cascading Classifiers

The first step of the V-J algorithm is to obtain an integral image from the original image. The integral image provides a quick and efficient way of calculating the sum of pixel values in a subset of a pixel grid which contain the pixels located above and to the left of x, y inclusive (with x, y being the coordinates of the pixel of interest) [8]. This can be represented mathematically by the following equation:

Where ii(x, y) is the integral image and i(x, y) the original image. The computation of integral image can be done in one iteration over the entire image. The advantage of computing an integral image is that the sum of any rectangle with in the image can be calculated very quickly and efficiently by the following equation.

The second part of the algorithm is application of AdaBoost to remove unnecessary features. For example the same Haar-feature template that applies to the eyes and the bridge of the nose can also apply to other parts of the face and using AdaBoost algorithm a best threshold can redefined which classifies the faces while selecting the features with minimum error rate so that the features that most accurately classify the faces and non-face parts of the image are selected. V-J algorithm combines a small number of these features to form an effective classifier, by employing a weak learning algorithm which determines the optimal threshold classification function so that a minimum number of misclassifications occur. All the images used in the training the classifiers are given an equal weight and every feature is applied on all the training images. After each iteration of classification weights assigned to misclassified images is increased and the iteration of classification is repeated. After each iteration new error rates and weights are calculated and assigned until the required accuracy is achieved. After the application of AdaBoost algorithm is successfully completed one feature is selected out of the 160,000+ features for each iteration.

The classifiers obtained from the AdaBoost stage are not strong enough to be used individually for object detection but the V-J algorithm suggests the work around to this by using the classifiers obtained in a cascade. Each stage of classifiers checks the given stage as to whether it definitely does not contain a face or maybe it contains a face. As soon as a sub window is classified as not containing a face it is discarded. The sub- window classified as maybe is then passed to the next stage in the cascade. It can be deduced that the more stages a sub-window passes the higher is the probability of it \_containing a face. All the stages in this cascade are obtained by training classifiers by using AdaBoost algorithm and then adjusting the threshold to minimize false negatives. OpenCV already provides a cascade classifier for face, eyes and other features and also provides a way to train custom classifiers for other objects that a user might want to detect [9].

The other method to perform face or object detection is to use neural networks. A lot of research is available on this technique and some trade-offs have been listed between using template based detection and neural networks. HA.Rowley et.al show in their paper that although neural network based object detection can enhance the results of the face detection system but it causes an increase in the execution time [10]. Although this paper was published in 1998 and with the evolution of neural networks e.g. tensor flow, this trade-off might not exist at this time and a new evaluation need to be performed. The system described in this paper is a two stage detection system. A filter is applied on a 20x20 pixel region of the image which generates an output from 1 to -1 based on the presence of a face in the image, in case the faces or the image is larger than the filter size the image is reduced in each repetition and the filter is applied on every point of the image on each iteration. In the second stage overlapping detections are merged, resulting in removal of false detections in neighbouring areas of the true faces, to further reduce the number of false positives, multiple networks can be used and their outputs are arbitrated to produce a final output.

M.Wu et. al. developed a system for face detection and feature localisation based on case structured classifier, template matching, feature space analysis method and AdaBoost. They also incorporated human face super resolution algorithm for helping in detection of faces in low-resolution images. The system performs image pre-processing with histogram equalization, the pre-processed image is fed to the face detector, which is based on the V-J’s cascade framework for robust and rapid object detection. The detector returns the number of faces and the locations of all the faces in the image. A resolution threshold of 64\*48 has been defined, if the faces are larger than this threshold value then they are passed on to the organ detection system otherwise they employ the face super-resolution algorithm to get a higher resolution image. Before this algorithm is applied a rough location of eyes needs to be detected, if a non-human face is detected in this step then that result could be discarded making the algorithm more robust. For organ location (eye location), this part is also interesting for our application for landmarks detection, the team used 8-eye-pair template as per the minimum relevant match theory. The team report an accuracy of 95.8% for eye detection, 89.8% for nose, 87.3% for mouth and 91.7% for chin detection [11].

Anamika Singh et.al. Proposed a system for face and eyes detection using Sobel edge detection and morphological post processing steps for extracting eyes, by analysing the eye regions and the shape of the region i.e. the distance from the left and right of the face and given the fact that the eyes are generally in a straight line on the face. They report a 92.5% success rate for eyes extraction from the IMM frontal face database [12].

Another algorithm for face detection and landmark localization is presented by Prathap Nair et.al [13]. The proposed algorithm detects and segments 3-d face meshes and localizes landmarks based on Point Distribution Model which removes any dependency on prior knowledge of texture, pose or orientation. They perform face detection by analysing the transformation between model points and points of interest in the sample image based on the deviation of the parameters from the mean model, afterwards landmark localization is performed on the extracted face based on the transformation that gives the minimum deviation of the model from the mean shape. They reported a 99.6% accuracy for face detection.

* + 1. Facial Landmarks detection

Landmarks detection in dlib and google.

* + 1. *Lip Reading Techniques*

Overview of pre-existing lip reading techniques.

* 1. Hidden Markov Model

The Hidden Markov Model (HMM) is a statistical model for a system with unobservable or hidden states. In simpler versions of Markov models for example the Markov Chain, the state is visible, and the only parameters to be considered are state transition probabilities. In the HMM since the states are not visible, instead the output dependent on the state is accessible. So each state has a probability distribution for possible outputs.

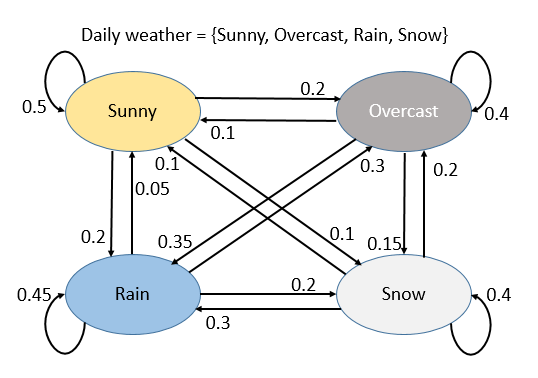
* + 1. *Markov Chains*

Markov Chains are used to describe types of random processes, giving information about the probabilities of sequences of random variables called states, which can take up values from a set (chain) for example the weather forecast for the day. Different states are then defined for the chain for example in case of weather the states could be {Sunny, Overcast, Rain, and Snow}. For a random process to be a Markov Chain it has to fulfil the property called the **Markov Assumption** which is defined as: the probability of being in a state j depends only on the previous state, and not on what happened before, or simply put only the present weighs in on predicting the future, while the past does not matter [4]. Mathematically this can be stated as:

Table xx below shows the essential components for a Markov chain:

|  |  |
| --- | --- |
| **Component** | **Explanation** |
| S={q1,q2… qn} | A set of ‘n’ states |
| P= | A transition probability matrix, containing probabilities of moving from state i to state j. |
| v=(v1,v2…vn) | Initial probability distribution vector. |

Fig. xx shows a state diagram with the transition probabilities, the states are shown as the nodes in the graph. The transition probabilities shown in the figure can also be shown in a matrix as:



In the stochastic probability matrix P each element (i,j) represents the transition probability. The deductions that can be made from this matrix are:

* Pk: each entry will represent the probability to arrive from i to j in k steps.
* If probability of being at a state at time t=0 are given by a vector ‘v’. The probability of being at state x at time t can be calculated as Pkv, the required result will be denoted by the xth entry in the matrix.
  + 1. *Hidden Markov Model*

As opposed to the simple Markov Model, in an HMM, Markov chains cannot be observed directly. In contrast the outputs generated from each stage are visible to the observer. Taking the example of a speech recognition system, the parts of the speech are non-observable events the words and their positions in a sentence are characterized as observable events and also can be the inputs of such a system. Table xx describes the components required to specify an HMM.

|  |  |
| --- | --- |
| **Component** | **Explanation** |
| S={q1,q2… qn} | A set of ‘n’ states |
| P= | A transition probability matrix, containing probabilities of moving from state i to state j. |
| V=(v1,v2…vn) | Initial probability distribution vector. |
| O=(o1,o2 … oT) | A sequence of T observations |
| B=bi(ot) | A sequence of observation likelihoods, they represent the probability of an observation ot being generated from a state i |

The Markov assumption as stated in the previous section also applies to HMM, but a 2nd assumption also comes into play which is termed as Output Independence and states that the probability of an output observation oi depends only on the state that produced the observation qi  and not on any other state or observation [5,6]. This assumption is mathematically described as:

* 1. Robot Operating System

ROS short for Robot Operating System is a collection of tools and libraries which is mainly designed to allow for writing robot software [30]. It provides the developers with low-level device control, drivers, package management and an infrastructure for message passing between processes. ROS systems are based on Publisher-Subscriber model. Active ROS processes are represented in a graph architecture and divided into different nodes. Each node can do its independent processing and can further communicate with other nodes. This communication is achieved by passing ROS messages. A node can be configured to be a publisher or/and a subscriber. In this scenario a publishing node does some processing and publishes some relevant information on a specific topic. This information could range from sensor data to images ( as in our case). Any other node which subscribes to this same topic name will have access to this data as soon as it is published by the first node.

2.4.1 ROS-Master :

A ROS-Master node must always run for the ROS based systems to work. This master node provides naming and registration services to all the other nodes in the system and manages publishers and subscribers to different topics.

2.4.2 ROS Nodes:

A node in a ROS system is any process that performs some computations.

A ROS eco system is made of multiple nodes. The advantage of using nodes is break down of code into simpler componenets. Added isolation so that if a fault occurs in one node, other nodes usually stay uneffected by it e.t.c. Node are combined together into a graph and communicate with each other by publishing and subscribing to topics or services.

2.4.3 ROS Topics:

Topics are named buses which are utilized by nodes to exchange information via ROS messages. Node in themselves are not aware of which other nodes they are commmunicating with instead they subscribe to the Topic containing the data that is relevant to them and publish the generated data to another relevant Topic if needed. Topics are generally used for unidirectional communication. ROS supports both TCP/IP based and UDP based message transport TCPROS is the default protocol used for communication.

For this project, the idea was to distribute the system between image processing tasks and machine learning (ML) tasks. The image capture and processing is done on the android device. Once the ROI has been extracted it should be transfered to a remote server running the ML module of the project which can then infer from the captured image-sequence the word or phrase that has been said.

For implementing this architecture several alternatives are available. For example the ML component (MLC) could be run on a cloud server (for example on Amazon Web Services) exposing the interface of the MLC via APIs. The android device can then transfer the images to the MLC via API calls and once the processing has been done on the cloud server the resulting string can be passed back to the android device via another API in json format.

Another option for implementation which does not require AWS, and in addition API development, was to use ROS. ROS provides the same architecture but on a simpler and lower scale. The application will again be divided into two parts Image-Processing Component which is to be run on the android device, and MLC to be run on a remote PC which must be on the same network as the android device running ROS-master node. For this architecture the Android application must also act as a ROS-node and once the ROI has been extracted it publishes the images to a ROS-Topic. The MLC running on the remote PC subscribes to this same topic. As soon as the images are available on the topic they are picked up by the MLC and the processing is performed. The results can either be shown on the PC or again published to another Topic to which the android device is subscribed to, so that the results can be shown on the device.

# Application Requirements

The requirement analysis for the application was done to come up with what needed to be done and to manage the development lifecyle of the application. The figure above shows a UML representation of the developed requirements. As can be seen from the requirements model, the developent has been divided into four phases to allow for easier development and clear goals. Each phase and the requirements contained in it are described in the following sections.

**3.1 Phase -1:**

*Analysis and comparison of facial and landmarks detection techniques:*

Phase-1 encompasses study of relevant image processing libraries and techniques which could be used for the application. A look into the timing and performance parameters of these techniques is to be performed so that the best approach could be decided for the development of the appliation in the later stages. The requirements corresponding to this stage are listed below:

*Req-1.1:* Determine best algorithms for face and facial landmarks detection, identifing parameter is timing performance.

*Req-1.1(a):* Application should run on android devices.

*Req-1.2:* Application should provide execution times for the relevant processes for analysis.

*Req-1.3:* Application should be able to detect faces in still images.

*Req-1.3(a):* User should be able to take an image using the device camera and use it for processing with in the application.

*Req-1.3(b):* User should be able to load an existing image from the gallery and use it in the application.

*Req-1.3(c):* OpenCV sdk for andoid should be used for face detection.

*Req-1.4:* Application should be able to detect facial landmarks in the given image.

*Req-1.4(a):* Utilize dLib for facial landmarks detection.

*Req-1.4(b):* Utilize Google’s Mobile Vision API for facial landmarks detection.

**3.2 Phase-2:**

*Real-Time Facial Landmarks Detection on Live Camera Stream:*

In phase-1 once the best approach has been identified, phase-2 is to extend the existing implementation from its application of still images to live feed from the on-board device camera. And to make the application as close to real time as possible given the hardware constraints of the target device.

*Req-2.1:* Andoroid application should be able to perform facial landmarks detection on live image stream from the device camera.

*Req-2.1(a):* App should utilize OpenCV and dLib in conjuction for facial landmarks detection

*Req-2.2:* Application should run close to or greated than 10FPS.

*Req-2.3:* App should be able to extract Lips ROI.

*Req-2.3(a):* Cartesian coordinates of the lips detection should be available for further processing in later stages.

**3.3 Phase-3:**

*Implementation of wireless data transfer to a remote computer:*

Phase-3 was to incorporate distributed processing capability in the application. Such that the image processing is to be performed on the android device but for further processing, the extracted images and information will be transfered to a remote computer running the Machine Learning Component of the application. This componenet will be responsible for processing the transfered images and deducing the said word or phrase from the given information.

*Req-3.1:* App should be able to transfer information over a wireless connection to a remote computer for further processing.

*Req-3.1(a):* Remote PC and the android device should be able to communicate over a TCP/IP link

*Req-3.1(b):* Use Robot Operating System for establishing data transfer capabilities.

*Req-3.2:* App should transfer the cartesian coordinates of the lips to the remote node.

*Req-3.3:* App should transfer the images of the Face/Lips ROI tot he remote computer.

**3.4 Phase-4:**

*Implementation of Machine Learning Component:*

Phase-4 is the final phase which brings together the work done in the previous parts and ties them together to produce the whole application. In this phase the Image processing results done on the device should be processed further on the computer. The chosen ML algorithm should then infer from the data the said word or phrase.

*Req-4.1:* App and PC should work as a system to analyse the video stream, extract facial landmarks and lips ROI and perform lip reading to infer the said word/phrase.

*Req-4.2:* A ROS node on PC should be able to fetch and save the incomming data from the mobile device.

*Req-4.3:* PC Node shall use TensorFlow to analyse the incomming images and perform lip reading.

*Req-4.3(a):* Training of TensorFlow model should be perfomed on a varied and a large dataset for better results.

*Req-4.4:* System shall be trained on a set of pre decided words and phrases.

# Architecture Design

Details of the requirements here.

* 1. **Application Flow**

Fig. x shows the flow of the android application

* 1. **Component Diagram**

Component diagram here

* 1. **Class Diagram**

Class diagram here

* 1. **Sequence Diagram**

Sequence Diagram here

# Application Development and Challenges

UI development, Code snippets, methods and their explanation. Performance improvements.

# Results

Description of results here

# Conclusion

Conclusions here.

# Documentation

* 1. **Developer Documentation**

Developer’s documentation

* 1. **End-User Documentation**

End User Documentation

# Java Doc

# Appendix

# References