

K.K.WAGH INSTITUTE OF ENGINEERING EDUCATION & RESEARCH

PROJECT STAGE -II

Depression Classification Model using Machine Learning

Group Id:05

Internal Guide: Prof. N.M.Pagare

External Guide: Prof. N.G.Sharma





TEAM MEMBERS

Division./ Roll No.	Exam Seat No.	Name of the student
A-17	B150134238	Harsimran Singh Dhillon
A-18	B150134242	Himanshu Garud
A-19	B150134207	Ojas Ahire
A-20	B150134270	Aniruddha Kulkarni

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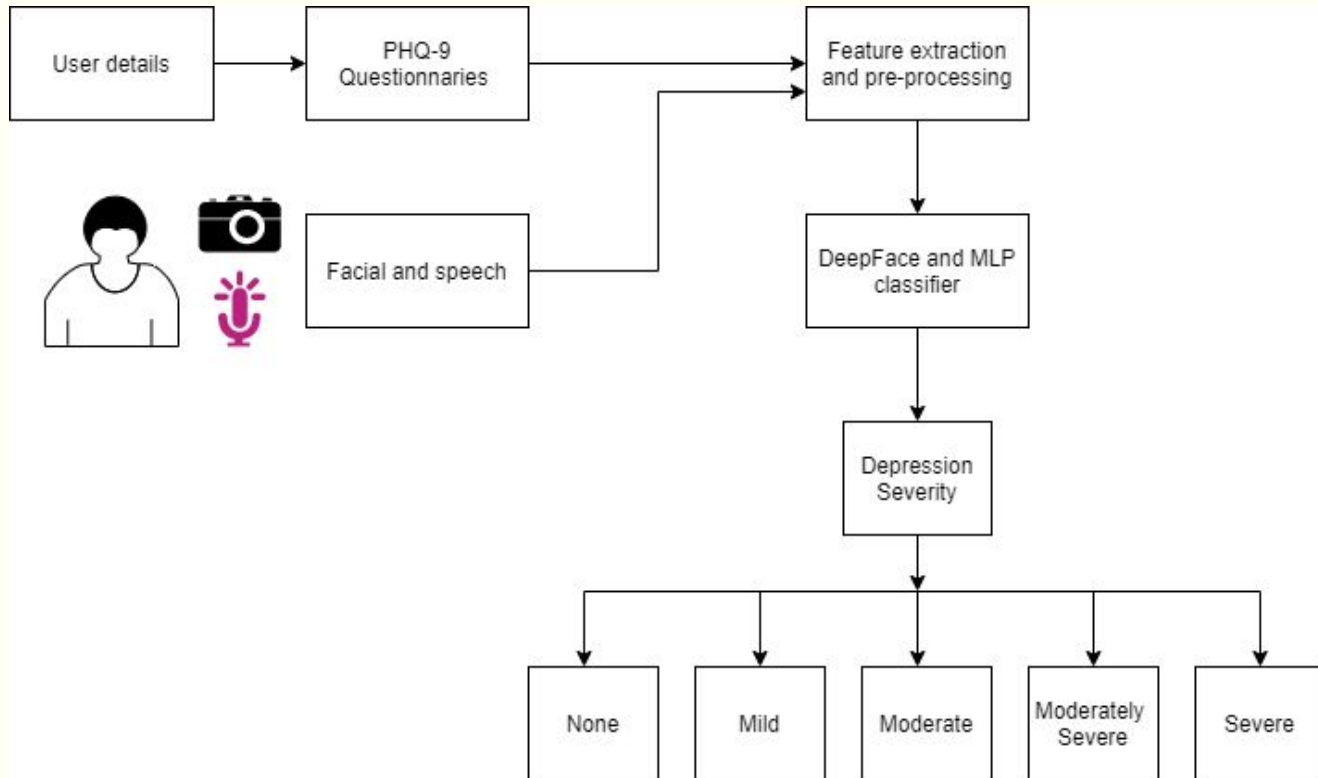
Problem Definition

- To develop an automated depression detection system using machine learning
- Technical Keywords: Speech processing, PHQ-9, RAVDESS, TESS, SFC, LFW, DeepFace and MLP Classifier

Objectives

- To build a system based on machine learning for predicting the depression severity levels in people
- To reduce the time required for diagnosis by automating the diagnosis procedure

Revised Final Design



Depression Classification Model using Machine Learning

Tools and Techniques Used

- Tools: VSCode , Jupyter Notebook and MongoDB compass
- Technologies: Vue.js, Flask, MongoDB and Heroku
- Languages: JavaScript, HTML and Python
- Techniques: MLP Classifier, CNN, DeepFace Architecture and Rule based learning

Implementation Status

- There are 4 modules
 - PHQ-9 questionnaire
 - Facial and speech input
 - Feature extraction and pre-processing
 - DeepFace and MLP classifier

Algorithms

Algorithms used in implementation:

- DeepFace- Facial feature extraction and classification
- MLP Classifier- Speech feature extraction and classification

Modular Testing

Functional Testing

- PHQ-9 Questionnaire Module:
 - Test Case 1: Output Changes dynamically after changing the input
 - Test Case 2: All the options and their associated values of outputs were compared
 - Test Case 3: Responses getting submitted
- Facial Feature Module:
 - Test case 1: Classified emotion as the output
 - Test case 2: No output for improper lightning conditions
 - Test case 3: No output if there is no person in the frame or bad angle

- **Speech Feature Module:**

- Test case 1: Classified emotion as the output of the speech
- Test case 2: No output if the mic is being accessed somewhere else
- Test case 3: No output if there is lot of noise

GUI Testing

- Authentication:
 - Test case 1: Prompts invalid details(invalid phone number digits and password validation in registration page)
 - Test case 2: Prompts successful registration and redirects to login page(after registration)
 - Test case 3: Prompts invalid password(when user enters a wrong password in login page)
 - Test case 4: Successful login and redirects to home page(when user enters correct credentials)

Results

- Following are the screenshots:
 - PHQ-9 Questionnaire
 - Facial Features Classification
 - Speech Features Classification
 - CSV File

PHQ-9 Questionnaire

family down?

More than half the days

Trouble concentrating on things, such as reading the newspaper or watching television?

Nearly Everyday

Moving or speaking so slowly that other people could have noticed? Or the opposite - being so fidgety or restless that you have been moving around a lot more than usual?

More than half the days

Thoughts that you would be better off dead, or of hurting yourself in some way?

Not at all

Total= 18/27
Depression Severity: Moderately Severe

Submit

Facial Features Classification

Speech Features Classification

C:\Windows\System32\cmd.exe

[illegible]

Features extracted: 180

```
['disgust']
```

Accuracy: 84.93%

CSV File

[illegible]

Conclusion

- PHQ-9 Questionnaire provides with the predicted results having an accuracy of 89%
- To make the predictions more accurate facial features as well as acoustic features are taken
- Overall accuracy of the system is 93%

Future Scope

- Building an android application of the system
- Improving the efficiency and reliability of the system
- Making the system more stable with each update possible
- Adding new features to visualise data

Efficiency Issues

- Misclassification of facial and speech features because of noise
- Large execution time
- Internet connectivity and hardware availability

Cost Analysis

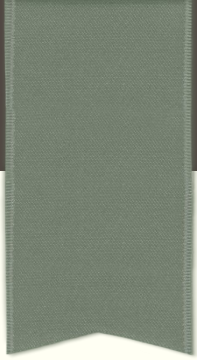
- Will only require cost for deploying the server on cloud
- Cost required for domain name
- Estimated lines of code, LOC = 2000
- Duration of project = 5 Months
- Efforts = 4.5 person-month

References

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Published 09 April 2019.



THANK YOU !!

