Classification of Depressed and Non-Depressed
Subjects and Predicting their MADRS Score
Using Machine Learning

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### Overview

- Objectives
- Using Awesome Backgrounds
- Engage your Audience
- Capture Audience Attention

### Introduction

- What is Depression?
- What is motor activity data?
- How can we use this sensor based data?
- Use of Machine learning in Psychology.

## **Objectives**

- Classification of depressed and Non-depressed subjects based on their motor activity data.
- Multi-class classification of subjects into Nodepression, Mild-depression, severe-depression.
- Prediction of MADRS score of the subject which later can be used for classification.

### Motivation

- The COVID-19 pandemic
- Change in lifestyle
- Increasing work load
- Depression in teenagers

# Associated Challenges & Research Gap

- Dataset size
- Imbalanced dataset
- Improvisation in model performance
- Multi-class classification

## **Dataset Explained**

- Dataset size
- Data type
- Data content



C+ <class 'pandas.core.frame.DataFrame'>
RangeIndex: 55 entries, 0 to 54
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype	
0	number	55 non-null	object	
1	days	55 non-null	int64	
2	gender	55 non-null	int64	
3	age	55 non-null	object	
4	afftype	23 non-null	float64	
5	melanch	20 non-null	float64	
6	inpatient	23 non-null	float64	
7	edu	53 non-null	object	
8	marriage	23 non-null	float64	
9	work	23 non-null	float64	
10	madrs1	23 non-null	float64	
11	madrs2	23 non-null	float64	
<pre>dtypes: float64(7), int64(2), object(3)</pre>				
memory usage: 5.3+ KB				

memory usage: 5.3+ K

### Work Plan

- Binary Classification
  - Dataset Preparation
  - Classification task
- Multiclass Classification
  - Dataset preparation
  - Classification task
- MADRS score prediction
  - Dataset Preparation
  - Prediction task

# **Binary Classification**

- Dataset Preparation
- Data Modeling

# **Binary Classification Results**

### Without Oversampling

	Accuracy	F1-Score	Precision	Recall
AdaBoost	0.71	0.56	0.56	0.56
XGBoost	0.74	0.58	0.61	0.56
CatBoost	0.78	0.63	0.65	0.60
GBM	0.72	0.52	0.57	0.48
1D CNN	0.63	0.65	0.66	0.60

# **Binary Classification Results**

#### With oversampling on training set

	Accuracy	F1-Score	Precision	Recall
AdaBoost	0.70	0.70	0.70	0.71
XGBoost	0.72	0.72	0.72	0.73
CatBoost	0.78	0.78	0.78	0.78
GBM	0.76	0.76	0.67	0.76

### **Multi-class Classification**

- Dataset Preparation
- Data Modeling

### Multi-class Classification Results

### Without Oversampling

	Accuracy	F1-Score	Precision	Recall
AdaBoost	0.82	0.82	0.82	0.82
XGBoost	0.86	0.86	0.87	0.86
CatBoost	0.85	0.85	0.85	0.85
GBM	0.84	0.84	0.85	0.84
1D CNN	0.83	0.84	0.85	0.84

### Multi-class Classification Results

#### With Oversampling on training set

	Accuracy	F1-Score	Precision	Recall
AdaBoost	0.89	0.89	0.89	0.89
XGBoost	0.87	0.87	0.87	0.87
CatBoost	0.89	0.89	0.89	0.89
GBM	0.88	0.88	0.88	0.88
1D CNN	0.83	0.84	0.83	0.85

### **MADRS Score Prediction**

With and without oversampling no major changes in results.

	MAE	MSE
AdaBoost	1.68	8.60
XGBoost	1.32	7.01
1D CNN	2.31	9.56

### Conclusion

- Tree based ensemble models can be used to classify subjects as depressed and non-depressed, Also they can be used to classify subjects into no-depression, milddepression & severe depression. CatBoost ensemble model can be preferred over others.
- Tree based ensemble model performed well and thus can be used to predict MADRS.
- The results can be improved further with more data, as we saw improvement in results with oversampling.
- This work can be used in smartwatche & smartphone applications to warn users about mental health.