# Detecting Parkinson's disease using Data-Driven Classification Model

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Abstract This study anchored on predicting Parkinson's disease whilst utilizing voice Dataset which assists with treating the individuals in early phases. Parkinson's illness is a neurological issue that prompts shaking and difficulty in walking, balance, and coordination. In most pessimistic scenarios, Patients have a big challenge walking or standing to an extent they can't live on their own and require a wheelchair to move around whilst help is required in all every day exercises. Other than motor indications, the individual may see, hear, or experience things that are not genuine (hallucinations), or accept things that are false (delusions). Parkinson's disease patients commonly have a low-volume voice with a droning quality. The speech pattern of Parkinson's patient is frequently delivered in short overflows with unseemly hushes among words and long stops prior to starting discourse. The voice dataset have the variables like MDVP: FO (Hz) - Average vocal fundamental frequency, MDVP: Fhi (Hz) -Maximum vocal fundamental frequency, jitter, simmer and so forth. The dataset was split into train and test where the train dataset was utilized to prepare the model. The test dataset was utilized to test the XGB model which delivered a higher precision of 100 percent.

#### Introduction <sub>1</sub> 1

- 2 Parkinson's disease (PD) is an ailment that influ-
- 3 ences the nerve cells in the cerebrum that make
- 4 dopamine gives indications including muscle immov-
- 5 ability, tremors, and changes in sync and talk. Parkin-
- 6 son's infection impacts a man's voice, aKecting them to
- murmur or have shuddering in talk. PD is simply sec-
- ond to alzheimer's disease in neurodegenerative sick-
- ness. It is relied upon to increment in the coming years
- 10 subsequently it is important to create identification
- 11 frameworks for down to earth examination and ideal
- 12 treatment. As the indications of PD happen all around
- 13 requested and generally, the older noticing the disor-
- der utilizing assessments of dysphonia has an indispensable part in examination. The characterization cal-
- culations from AI and machine learning are utilized to
- predict and explore the Parkinson's disease. The ideal
- highlights from the dataset are passed as contribution
- to the models and the expectation results are goNen.
- The expectation execution can be approved from the
- precision acquired through the classification algorithm.
- The assurance of Parkinson's ailment has logically im-
- proved the exactness boundary through the diKerent
- analysis. This paper will therefore explore the eKective-
- 25 ness of using supervised classification algorithms, such
- 26 as XGBOOST classifier, to accurately diagnose individ-
- 27 uals with the disease.

## **Background**

- 29 The clinical diagnosis of Parkinson diseases (PD) can
- 30 be aKirmed based on neuro-pathologic and histo-
- 31 pathologic standards [1]. Clinical indicative grouping
- 32 of PD should be possible on far reaching survey of the
- 33 writing information and determination basing on the
- 34 aKectability and particularity of the trademark clini-
- cal features. Imminent with center pathologic exam-
- 36 inations in agent populace of patients demonstrating
- 37 PD are expected to explore the clinical, pathologic, and
- 38 nosologic studies dependent on recurrence of event, at-
- 39 tributes, and danger factors in patients [2]. Neural Net-
- 40 works, DMneural, Regression and Decision Trees are re-
- 41 cently utilized for ascertaining the presentation score of 42 the classifiers dependable analysis of PD [3, 4]. R. Arefi
- 43 Shirvan.et al [4] proposed a framework/system for de-
- 44 tecting early stages of Parkinson's diseases. The data
- classification was completed by utilizing knn method.
- The least complex technique in gathering the compara-
- 47 bility is knn. Among classification strategy knn is uti-
- 48 lized at whatever point current realities for information
- 49 dissemination are insuKicient [5]. In this technique it 50 has two sections: a) decide k close neighbors, b) decid-
- 51 ing class type utilizing these nearby neighbors. it was
- 52 demonstrated that a 93.7 percent of precision for each 53 4 upgraded highlights, an exactness of 94.8 percent per
- 54 7 streamlined highlights and 98.2 percent precision for
- 55 9 advanced highlights is accomplished which is a mo-

56 mentous outcome contrasted with diKerent investiga-57 tions. In this study data from [6] from UCI archive is 58 utilized. The information incorporate 192 voice test ac-59 counts from 32 male and female. Each subject has had 60 6 voice signal accounts. 23 individuals experience the ill 61 eKects of PD and the rest are sound. individuals were 62 around 46-85 years of age the primary inconvenience of 63 the knn calculation is that it is a lethargic student, for 64 example order is finished by utilizing preparing infor-65 mation and from the training data it did not get to learn 66 anything from it. Mohammad s islam.et al [7] directed 67 a comparative analysis to detect Parkinson's infection 68 utilizing diKerent classifiers. Support vector machine 69 (SVM), feed forward back-spread based artificial neu-70 ral organization (fbann) and arbitrary tree (rt) classi-71 fiers were utilized and an examination between them is 72 made to separate among PD and healthy patients. The 73 study has utilized the UCI machine learning repository 74 from [8],[9]. The dataset comprises of 195 voice tests 75 from 31 people involving the two males and females. 76 From the taken subjects 23 were resolved with PD and 77 8 were healthy. To improve the grouping precision with 78 insignificant error rate a 10-fold cross approval which was repeated multiple times (100) has been executed for all the three classifiers. The knn classifier has accomplished a 97.37 percent acknowledgment exactness consequently outflanking the other two classifiers. Ddr. r. geetha ramani.etal[10] has suggested a framework to order PD and Non-PD patients by the following classifiers; binary logistic regression, linear discriminant analysis (Ida), partial least square regression (pls), random tree (rnd tree) and support vector machine (svm). The Parkinson's disease dataset is retrieved from the UCI Repository. This data is exctracted from patients and comprises of 197 unique samples and 22 features. Fisher separating feature choice calculation was discovered to be a viable element positioning framework. The random tree calculation accomplished 100 percent arrangement precision while the lda, c4.5, cs-mc4 and k-nn yielded exactness results more noteworthy than 90 percent. Among all, the c-pls calculation accomplished minimal precision of 69.74 percent. The multilayer perceptron (mlp) with back-engendering learning algorithm.

#### 3 Methodology 100

101 The following are the means that has been taken to 102 assemble the proficient model for early detection of Parkinson's illness: 103

#### **Dataset Detail** 104

The dataset used to build the model for this study 105 106 is retrieved is collected from UCI website. This dataset 107 has 195 unique values and 24 columns. Matrix column entries (aNributes):

MDVP: Fo(Hz): Average vocal fundamental fre-110 111 quency MDVP: Fhi (Hz): Maximum vocal fundamental fre-112 113 quency MDVP: Flo(Hz): Minimum vocal fundamental fre-114 115 quency MDVP:JiNer(percent),MDVP:JiNer(Abs), 116 117 MDVP:RAP,MDVP:PPQ,JiNer:DDP-118 sures of variation in fundamental frequency. MDVP:Shimmer,MDVP:Shimmer(dB),Shimmer:APQ3, 120 Shimmer:APQ5,MDVP:APQ,Shimmer:DDA – Several 121 measures of variation in amplitude NHR, HNR - Two measures of ratio of noise to tonal 123 components in the voice Status - Health status of the subject (one) - Parkin-124 125 son's, (zero) - healthy RPDE, D2 - Two nonlinear dynamical complexity 126 127 measures

Name - ASCII subject name and recording number

DFA - Signal fractal scaling exponent 128

129 spread1, spread2, PPE - Three nonlinear measures 130 of fundamental frequency variation

#### **Data Preprocessing** 131

109

This section entails two cycle which is Normaliza-133 tion and adjusting the dataset and is given in finer de-134 tail below:

#### Normalization 135

Normalization is a procedure which is applied as 137 a stage of preparing a dataset for machine learning model. The need of normalization is to adjust the estimations of numeric sections in the dataset to a typical scale, without changing contrasts in the scopes of values.

Xnew= (X- Xmin)/(Xmax- Xmin) 142

Where Xnew specific component spoke to by a seg-143 144 ment in the dataset, x is a value of this column. The 145 minimum value of the column is represented as Xmin 146 and the maximum value of the column is Xmax. How-147 ever for this study I used the MinMaxScaler function 148 from sklearn to normalize the features.

### **Modeling and Analysis**

149 XGBOOST: XGBoost is a gradient boosting library. 150 151 It helps to implements machine learning algorithms un-152 der the Gradient Boosting framework. XGBoost a par-153 allel tree boosting which solves many Machine Learn-154 ing problems in a fast and simple way. The jupyter note-155 book code aNached demonstrates how XGB classifier 156 has solved this machine learning problem.

#### **Results and Discussion** 157 4

158 This machine learning project analysis has utilized var-159 ious factors/variables to detect the presence of Parkin-160 son's disease. The XGBClassifier was used for the clas-161 sification and made use of the sklearn library to prepare

162 the dataset. This XGBClassifier model produced an ac163 curacy of 100 percent, which is great considering the
164 number of lines of code and the size of the dataset in
165 this python project.

## 166 5 Conclusion

167 In conclusion, this study has leveraged on the use of the XGBoost classifier which gives eKicient Parkinson's disease prediction model with high accuracy 100 per170 cent. This will go a long way in assisting detecting and predicting Parkinson disease prior to geNing it to most exceedingly awful. Analysis of voice data is sig173 nificant in the current decade to comprehend and in174 dicative techniques for human infections. The current technique gives the finding of PD utilizing voice dataset through machine learning algorithms. Early recogni175 tion of Parkinson's illnesses is valuable as it will assists with keeping the patients from most notice ably terrible 179 stage

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