

# Discovery of Parkinson's syndrome with hand tremor analysis using density based improved K- Medoids algorithm

RaghuviraPratap A, Babu Sallagundla, Kranthi Kumar Guttikonda, Prasad J V D

*Abstract---* With medical care system, gigantic measure of irrefutable and research facility test, tomography, prescription and healthcare data of hand tremor are been gathered and stored tremendously. Tremor in Parkinson's syndrome is a most important component utilized in the assurance of hand rest ailment at beginning and movement. Traditionally, tremor has been assessed utilizing recurrence repetition analysis. The discovery of definitive hand tremor would be a major step towards early and reliable diagnosis for Parkinson's syndrome. Big data analytics and data mining algorithms on these data go for early recognition of Parkinson's syndrome that will assist in creating anticipatory measures and in enhancing serene care considerations. Clustering and Classifications are the two ways towards arranging objects into various gatherings by apportioning set of data hooked on a progression of subsets. Cluster has occupied its underlying foundations from algorithms like KMedoids, fuzzy c-means and KMeans. Anyway, regular KMedoids clustering algorithm experiences numerous constraints. To discover a feasible solution for Parkinson's syndrome tomography, prescription and healthcare of various companions are gathered, overseen and engendered through Parkinson's Progression Markers Initiative. Griddle multigram data sets and Survey graph give the factual investigation on the hand tremor analysis so that the well and Parkinson patient would be accurately categorized. This study and proposed approach emphasis around the diverse traditions to prevail over the difficulties existing by PPMI data, which is extensive and rift. This effort use the underlying revelations completed from end to end vivid investigations of different attribute. The study and visualisation of data prompted recognising the noteworthy characteristics. We utilized cluster analysis methodologies to look for Parkinson syndrome subtypes from a huge, multi-centre and healthy care cohort of patients across all phases, with a process of motor chief features such as rigidity, tremor signs we are additionally providing to built a software product that elevates back-to-back discovery of Parkinson's syndrome data to recognise potential biomarkers.

**Keywords:** Parkinson's syndrome, Hand tremor, PPMI computer aided diagnosis system, cluster analysis, KMedoids.

## 1. INTRODUCTION

Parkinson's syndrome as a Neuro-degenerative disorder and a huge number of populace endure with it everywhere throughout across the globe. Parkinson's syndromes, the

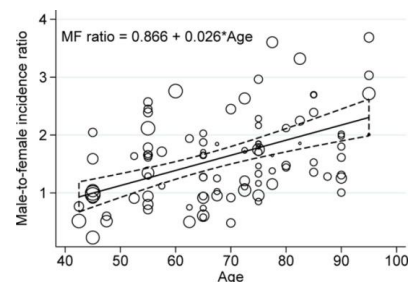
most widely recognised neurological turmoil in the elderly, were portrayed by dynamic extra pyramidal motor dysfunction including resting tremors, muscle inflexibility, hypo locomotion (bradykinesia and akinesia) and postural flimsiness [1]. Parkinson's syndrome as a neurological ailment that influences a individual movements, and may lead to hand tremors, gradualness of progress, sway rigidity and disproportion, as glowing as change in discourse and hand written skills.

A remarkable file of huge amount of data on Parkinson's syndrome is gathered, overseen and dispersed by the Parkinson Progression Markers Initiative. The coordination of thus perplexing and assorted amount of data from numerous locations offers unmatched challenges to consider the early on stages of predominant Neuro-degenerative procedures track their movement and rapidly recognise the efficacious of elective treatments. Particularly, in industrial countries, the quantity of patients with Parkinson syndrome has expanded essentially recently [2]. Be that as it may, there are no strategies which can gauge the movement effectively and precisely in its early times.

Different phases of Parkinson's sickness are

- i. Essential - Because of obscure reasons
- ii. Auxiliary - Dopamine inadequacy
- iii. Innate - Genetic origin
- iv. Multiple framework decay - Degeneration of parts other than mid-brain.

With the information about this is used to analyse, pathway and anticipate Parkinson syndrome and its succession [3]. A quick and opportune identification of these ailments could spare and fundamentally progress patient life by applying suitable methodology. With traditional Parkinson syndrome evaluation, association turmoil public sponsored Unified Parkinson Syndrome Rating Scale is riotously utilized.



**Figure 1: Parkinson syndrome male-to-femaleratios increase with age world-wide.**

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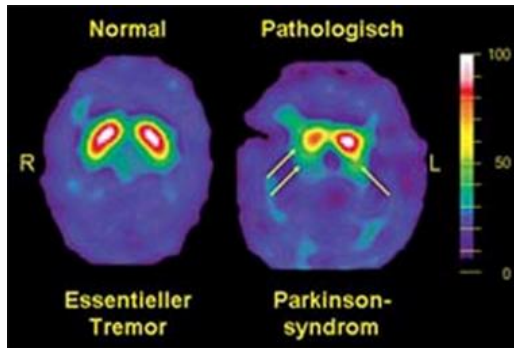
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As of late, assortments of innovative analysis advancements being utilised to identify, direct and indulge neurological syndrome, for example, brain gesture tests such as biopsy, computer aided tomography, magnetic resonance imaging, electromyography, positron outflow tomography. These innovations are imperative apparatus that assist doctors affirm or decree away the presence of a neurological disorder or former restorative circumstances.



**Figure 2: Example of a PET scan with different significance in Essentieller tremor and Parkinson syndrome.**

Parkinson statistics has every single of the characteristics of huge data, which portrayed by volume, velocity and value. Volume with an ever-increasing number of characteristics being gathered for the Parkinson exploration and with the expansion in support of distinctive companions through altered activities, volume of the information is developing. Variety Parkinson's syndrome contain organised, wording, imagery, acoustic and partially ordered data gathered as of the different shrewd wellness diagnosis tracking devices. With velocity portrayed by the swiftness in which information is made, accumulated and refined. These days' instantaneous dispensation frameworks help continuously making decisions. Veracity deals with integrity of statistics. Parkinson information is assorted, gage-genesis, inadequate, in congruent and meagre. With value compact, extract values by the incentive data, which is the objective of large data analytics<sup>[4]</sup>.

With main contribution of this study is to analyse the Parkinson's syndrome hand tremor data set and to identify significant patterns from it such that it leads to find out the biometric information of Parkinson syndrome.

Here with the study, appliance, test and legalize integral approach and data mining classification methodologies for Parkinson syndrome categorization and prophecy. Towards discover of Parkinson syndrome hazard with big data procedure, with together handled composite PPMI imagery, hereditary qualities, irrefutable and statistical information. Clustering techniques necessitate the supervision of density-based data, i.e. information related to human hand tremor through images. Gradually big data drafted from MRI and EEG images and X-ray resources. Consequently, computerized acquaintance detection becomes supplementary and added imperative role in hand tremor data records. Clustering applications are smart for the assignment of class detection. Conversely, the technique to

big data raises the subsequent necessities for clustering methodologies:

- a) Nominal desires of sphere acquaintance to resolve the effort attributes, since suitable principles are often not known before hand when dealing with big data repositories.
- b) Detection of clusters with capricious outline, because the outline of clusters in clinical databases may be sphere-shaped, long lasting, linear, drawn out etc.
- c) High quality competence on huge databases, i.e. on databases of drastically more than just a hardly any thousand of items.

## 2. BACKGROUND STUDY

A standout amongst the most well known databases utilized for this reason for existing is the UCI mechanism erudition warehouse, PPMI and Movement Disorder Society. The technique proposed in this examination was examined on coronary illness, Parkinson syndrome, neuro-degenerative disorder indexes got from these healthcare records, and the outcomes got be contrasted and considers in the prose. While choosing these data-set indexes, syndromes with lofty transience toll that influence the greater part of people were chosen. The synopsis data in the bibliography regarding studies about that execution on these data collections is visual beneath.

With difficulties presented by biomedical data investigation were overwhelmed by the conduit background. The conduit is a group based disseminated solution for dependable supervision of these diverse resources. The conduit permits various client and remote servers to interface, trade conventions, direct the execution, screen the conditions of various utensils or equipment, and offer complete protocols as versatile XML work stream<sup>[5]</sup>. According to Gracy [5] have examined the different kinds of classifiers to be specific, Naive Bayes, J48 and choice hierarchal tree based approaches. Shuddering hand, segment, upper limb or jawbone and passionate modifications are the elements considered in the discovery. With the Little et al<sup>[6]</sup> utilized SVM classifier with Gaussian drastic basis procedure to foresee syndromes in Parkinson's, with these methods for characteristic detection strategy to diminish the component liberty, and finest exactness rate of 91.4 was gotten by the proposed model.

According to Peter M. et al.,<sup>[7]</sup> Proportional classification study on diverse data sets in an individual have been implied for accurateness investigation and the time in use to accomplish the data set in regulate to find the finest categorization rule. The data of vigorous people and those with Parkinson could be exactly classified by using machine learning and big data tools. The SVM classification technique was favored in the categorization phase. With the performance evaluation stage, classification exactness, specificity, warmth analysis, f-measure, kappa data value and ROC analysis were utilized. Shahbaba and Neal<sup>[9]</sup> is an revelation of a non-linear model based on Dirichlet mixtures for the syndrome of Parkinson's classification, compared with multinomial legit models, verdict trees, and SVM. The proposed model attained

the classification accurateness of 86.8. According to Daset al.<sup>[9]</sup> used a proportional study of neural networks regression and decision trees for the diagnosis of Parkinson's syndrome ; the experimentation results had shown that the Neural Networks method attained the generally classification concert of 92.9. Sakar and Kursun et al.,<sup>[11]</sup> used common information measure to merge with SVM for the identification of Parkinson's syndrome and achieved the classification result of 92.7 percentage. Lukka<sup>[12]</sup> offered a another methodology where they have used Fuzzy based non linear entropy to combine with similarity classifier to anticipate Parkinson's disease and the classification mean is 85.7 percentage. Li. Et al<sup>[15]</sup> started a initial novel approach fuzzy based non linear and transformation approach combined with SVM and the Parkinson's syndrome record sets.

With these studies, it tends to oversee that the vast majority of the normal classifiers from machine learning and neural network utilized for disease analysis. Data collected from irrefutable, biomedical and statistical studies is complex. Though there are numerous approaches that are executed on diverse data sets on Parkinson syndrome there is nothing-appropriate application product build to optically look at the correlation amid a mixture of attributes and Parkinson's syndrome progression. So clearly, the decision of a proficient feature pre-processing technique and a great classifier is of critical significance for the diagnosis issue. Going for enhancing the proficiency and adequacy of the order execution for the analysis of, in this paper, a productive highlights weighting technique called improved KMedoids classification and a kernel density based extreme learning machine algorithms are analyzed.

### 3. METHODOLOGIES

A standout amongst the most difficult assignments when managing finding is to utilize illustration and indication based imaging data from patient tests. With earlier discussion, past mechanism have utilized high-end imagery technology for such purposes, yet being costly and might be sufficiently obtrusive with the patient.

#### A. Extracting Knowledge from Handwritten Exams

This methodology propagates a recorded data set from hand written exams. To fulfill this task, the loom is divided by two sections: a) dispensation the images, and b) characteristic extraction. Initially, it is required to pull out the hand written tremor from the spiral template, since the images are not registered. Immediately, it is significantly differentiated with its features based on their shapes. Here the images are digitized, through classification algorithm and image-processing techniques such as blurring filtering the clutter and disagreeable artifact are smoothed by means of a  $7 \times 7$  mean riddle.

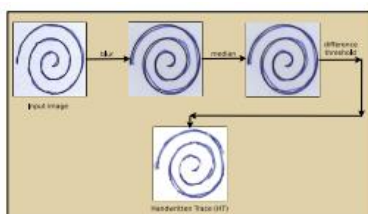


Figure 3. Image processing steps concerning the handwritten trace extraction.

The facet extraction stair goes for portraying both Signature Tremor and Handwritten Tremor, and afterwards to contrast those all together with assess the "measure of distinction" between equally imagery. In request to satisfy this errand, we primary require a brief and conservative portrayal of both Signature Tremor and Handwritten Tremor, which is practiced now by methods for the draft of the threshold imagery.

#### B. Improved density based KMedoids algorithm

KMedoids clustering technique projected to take away the clamour and tremendous kindness of KMeans algorithm to the incomparable statistics. *KMedoids* is a classical partition approach of clustering that clusters the information set of  $p$  objects into  $k$  number of clusters [1, 3]. This  $k$ : the number of clusters obligatory is to be given by abuser. This technique operates on the rule of minimizing the addition of dissimilarity amid every object and its equivalent reference point. The algorithm indiscriminately chooses the  $k$  objects in dataset  $P$  as early envoy objects called medoids. A medoid can be definite as the entity of a cluster, whose average dissimilarity to all the objects in the cluster is negligible i.e. it is a mainly centrally positioned point in the given statistics set. For all objects in the dataset, it assigns every entity to the adjoining cluster relying upon the object's remoteness to the cluster medoid. Following each task of a data object to meticulous cluster the new medoid is determined.

Let  $P$  be the Dataset with  $k$  points

$k$  is the number of clusters to be identified.

$j$  be the number of clusters primarily found by

density based clustering approach.

##### A. Input

$k$ : the number of clusters.

$P$ : a data set having  $m$  objects.

##### B. Output

Set of  $k$  clusters.

##### C. Approach

a. haphazardly prefer  $k$  objects in  $P$  as the preliminary envoy objects;

b. for all objects in the data set  $P$

i. Discover the cluster  $E$  which is adjoining to object  $j$  by using the distinction gauge;

ii. Allot object  $j$  to cluster  $E$ ;

iii. Locate the component object in cluster  $E$  having least amount intra cluster dissent as new centroid of  $E$

#### C. exhibit data of clusters obtain.

Pertain one iteration of KMedoid clustering with  $k$  and new  $C_k$  centers as the primary attributes and marque all the clusters with  $k$  labels.

#### D. Data Categorization

Data from different documents are ordered into six significant classes, for example,

i. Bio specimen data,

ii. Imaging such as DTscan imaging,

Magnetic Reverberation data.





iii. Health care history such as general medicinal history, General neurological exam, General physical exam, hand movements and hand written data.

iv. Issue characteristics such as sociology economics, PPMI obtained at irrefutable destinations in the countries),

v. Motor evaluation such as assessment of tremor with bradykinesia, assessment of tremors in idiom, jaw, bring down lip, hand over or in the leg. Development Disorder Society offers brought jointly Parkinson's syndrome Rating Scale that reins in the machine assessment.

vi) Non-motor evaluation such as evaluation of vocal learning, semantic acquaintance and lethargy scale are a portion of the non-machine valuation tests.

Category	Number of Files
Biospecimen	11
Imaging	9
Medical History	14
Motor Assessment	11
Non-Motor Assessment	16
Subject Characteristics	5

Table 1. Analysed and cleaned data files with various syndromes in Parkinson’s disease.

4. PROPOSED SYSTEM

Tidiness and gyrating the data is that the chief confront. Every dossier engaged separately, outmoded and managerial data that was not required for the revise, which was separated. Collectively these wide parameters along creates a huge distributed matrix. Discovery the association between varied attributes and visualizing them is that the goal of this paper. Deserves – This structure with an easy interactive optical image can abstract individuals from subtle arithmetic to produce a simplified and comprehensible description of the malady to the existence form of a familiar person.

Figure 4. illustrated an outline of our approach to accurately diagnose no degenerative disorder. Firstly, the syndrome hand tremor medical knowledge from clinics and local hospitals in data ware house is shaped. The medical data includes the symptom, syndrome, strokes of hand movement, hand written and doctor advises segmentation to get keywords. Afterwards, the set of keywords framed into data based mostly structure by exploitation classification rule KMedoids.

Secondly, once patients input detail of symptoms, it'll be matched with the associate and implication rules from cognitive content to identify and detect the co-symptom. Consequently, our approach proposes an interaction with the patients and suggest the co-symptom to reinforce the prognosis and remedy.

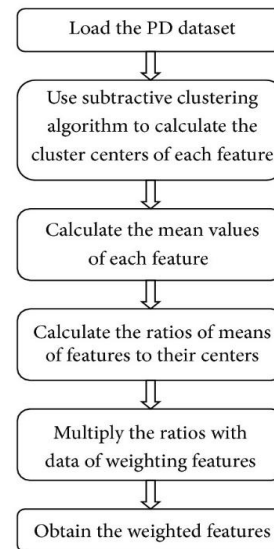


Figure 4. illustration of proposed overview with KMedoids algorithm.

There are around 3000 attributes and range continuously escalating because it is a current swot. This manuscript addresses the overall hand tremor signifies many challenges of data curation, aggregation and a primary evocative analysis of the PPMI information.

Here the approach provides the result of initial analysis. From every data-set file of hand, tremor analysis redundant and administrative data for the explanation was filtered and aggregating these wide selection parameters along develops an enormous meagre matrix and discovering the correlation between varied attributes and viewing it reaches the target.

Measure	Healthy Controlled Subjects		Parkinson’s syndrome with hand tremor data analysis	
	Mean(SD)	Range	Mean(SD)	Range
Gait velocity	1.1(0.2)	0.7 - 1.7	1.0(0.2)	0.7 - 1.7
Finger tapping	52.9(8.2)	35.0 - 71.0	36.7(10.3)	35.0 - 71.0
Strokes in hand writing	43.4(7.5)	35.0 - 71.0	24.5(7.5)	25.0 – 60.0
Movement time	4.1(0.3)	2.8 - 3.9	3.2(0.2)	2.7- 3.6
Reaction time	3.2(0.2)	3.6 - 4.9	4.3(0.2)	3.8- 5.1

Table 2. Summary of distinctive data classifications after data was analysed and cleaned.

A novel method entitled density based KMedoids clustering-based attribute weighting was proposed as a data pre-processing method.

5. RESULTS

**Metric measurements used for Evaluation**

Within categorize to calculate the recital of a clustering and classification system, a suitable metric Parkinson's syndrome hand tremor dataset will be desirable. For examining the algorithms under consideration, we apply Rand Index and Run Time as two measures

i. Concert in terms of time:

We evaluated the density based KMedoids in provisions of time required for clustering on PPMI dataset.

a) The parameters of Multidimensional Data:

No. of classes that represents clusters: 6

No. of proportions: 3

No. of nodes per each cluster or class: 80, 90, 140, 190

Standard Deviation: 5.000000e-003

ii. Concert in terms of accuracy:

The no. of elements in E that are in the identical separation in X and in the identical separation in Y,

i is the no. of elements in E which are not in the identical separation in X and not in the identical separation in Y,

j is the no. of elements in E that are in the same partition in X and not in the identical separation in Y,

k is the no. of elements in E that are not considered in the identical separation in X but are in the same partition in Y.

Instinctively, one can believe of  $i + j$  as the no. of agreements between X and Y and  $j + k$  the no. of disagreement between X and B. The Rand index, E, then becomes, Rand index has a assessment between 0 and 1 with 0 representations that the two data clusters do not differ on any pair of points and 1 representing that the data clusters are exactly thesame.

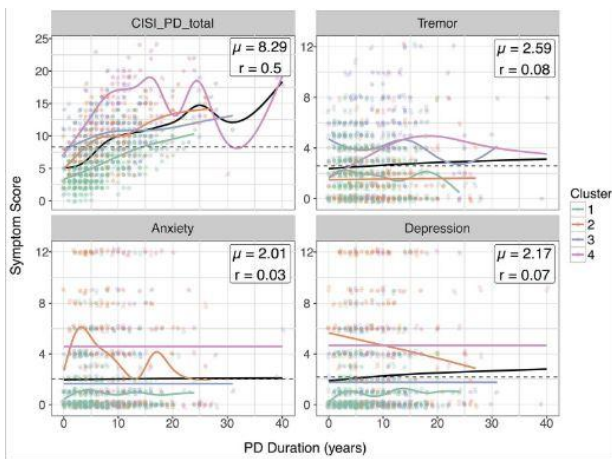
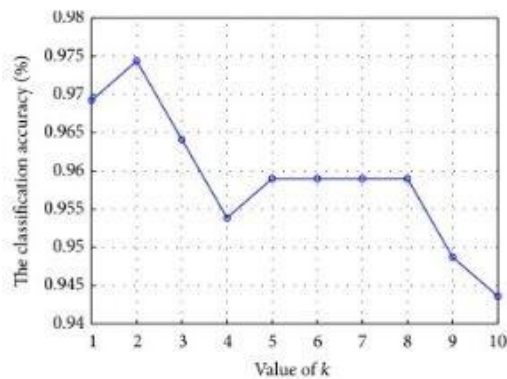
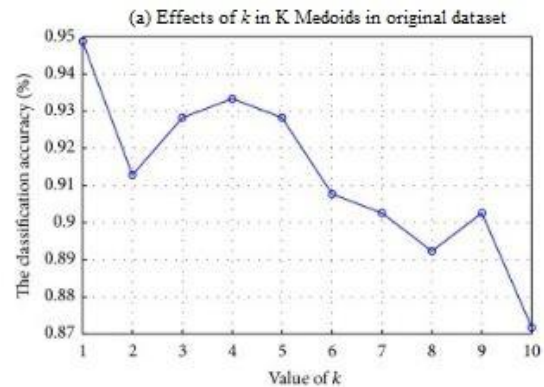


Figure 5. 2 dimensional allocation of four classes of clusters formation in the unique and weighted characteristics space by the best three principle components obtained with KMedoids method.

Authors	Methodology	Classification Accuracy
Duch et al.	K NN and FSM and SSV	84.6
Gracy et al.	Naïve Bayes and Random Tree	85.5
Little et al.	SVM and Gaussian Radical basis	91.4
Peter M. et	SVM, Kappa statistical	87.6

al.	Analysis	
Shahbaba et al.	Dirichletprocess Mixtures and Multinomial variable selection	92.9
Das et al.	kNN and Decision tree based hybrid genetic algorithm	92.9
Sakar and Kursum et al.	SVM with combined kernels	88.9
Lukka et al.	Fuzzy SVM and Bandler cohort rules	85.3
Little K et al.	Fuzzy based. Non linear and SVM	97.74
Proposed Method	Improved density based K-Medoid	97.5

Table 4 Contrast performance of different techniques in terms of accuracy (%) for the Parkinson's syndrome hand tremor data analysis.



(b) Effects of k in K Medoid in weighted data set

Qualified analysis of the proposed novel method found in the literature is given out in the table. Comparative analysis of Parkinson's syndrome of hand tremor dataset, which is clinically tested in software, can be seen normally ranging in between 79 and 91 percentages by the researchers. This 97-percentage accuracy of classification was achieved with the planned method for the data set considered.



## 6. CONCLUSION

The analysis showed that Parkinson's syndrome might corollary incarcerates vital movement characteristics that differentiate Parkinson's syndrome rigorousness and determine crucial symptoms. In hand movement tremor detection, the compassion was 78 and correctness was 97.5%. The marked motion options furthermore demonstrated robust correlation with Parkinson's syndrome sternness stage, hand movement tremor severity, and way of walking predicament. This approach is easy to use, user responsive, and economically reasonable. The simplest result was obtained by density based KMedoids classifier, which achieved concerning 97.5% of recognition. Such accuracy rate seems to be terribly appropriate for Parkinson's syndrome hand tremor identification, since we have utilized a inexpensive device for image acquirement. About future works, we have a tendency to aspire at taking into consideration wander images for Parkinson's syndrome identification, similarly on use each meander and spiral images for decision making system purposes. We have applied comprehensive experiments exploitation very dissimilar categories of dimensions at baseline from Parkinson's syndrome Progression Markers Initiative data set to forecast the sternness of the syndrome, measured by unified rating scale.

Our results ensure a number of the vital biomarkers known in active health check studies, corroborate a number of the biomarkers that are discovered as a possible marker of and see new biomarkers that have not nonetheless been discovered. Here this process of clustering is an proficient way of reaching in order from raw data and KMean's, KMedoids are basic methods for it. While it is easy to employ and understand, KMeans and KMedoids have solemn drawbacks. The anticipated density based KMedoids method performed exceptionally fine than on Parkinson's syndrome dataset in term of eminence of categorization deliberate by Rand index. As the major challenges in health care domain is the removal of intelligible knowledge from health check analysis data.

## REFERENCES

1. Stacey Kowal L, Timothy M. Dall, Ritashree Chakrabarti Michael V. Storm, Anjali, "The current and projected economic burden of Parkinson's disease in the United States", Movement disorders Clinical Practice, March 2013
2. Mahalakshmi Senthilarumugan V, Baskar G, S Nilakanta, "Big data and Parkinson's: Exploration, analyses, data challenges and visualization", Proceedings of the Hawaii International Conference on System Sciences, 2018
3. Micheal Minelli, Chambers M, Dhiraj A, "Big Data, Big Analytics: Emerging Business Intelligence and Analytic Trends for Today's Businesses", Wiley, 2013.
4. Little M. A., McSharry P. E., Hunter E. J., Spielman J., Ramig L. O., "Suitability of dysphonia measurements for telemonitoring of Parkinson's syndrome.", IEEE Transactions on Biomedical Engineering. 2009;56(4):1015–1022.
5. Peker M, "A decision support system to improve medical diagnosis using a combination of KMedoids clustering based attribute weighting and SVM.", NCBI, 2016
6. Shahbaba B., Neal R., "Nonlinear models using Dirichlet process mixtures.", Journal of Machine Learning Research. 2009;10:1829–1850.

7. Das R., "A comparison of multiple classification methods for diagnosis of Parkinson syndrome. Expert Systems with Applications.", 2010;37(2):1568–1572.
8. N Shamli, Sathiyabhama B, "Parkinson's brain disease prediction using data analytics", Modern Education and Computer science Press, 2016.
9. Sakar C. O., Kursun O., "Tediagnosis of parkinson's syndrome using measurements of dysphonia.", Journal of Medical Systems.", 2010;34(4):591–599.
10. Psorakis I., Damoulas T., Girolami M. A. "Multiclass relevance vector machines: sparsity and accuracy.", IEEE Transactions on Neural Networks. 2010;21(10):1588–1598.
11. Pasi Luukka, "Feature selection using entropy measures with similarity classifier. Expert Systems", 38(4), Expert Systems with applications, 2011
12. D Chiang Li, Chiao Wen Liu, Susan C Hu, "A fuzzy-based data transformation for feature extraction to increase classification performance with small medical data sets", Artificial Intelligence in medicine, 45-52, 2011.



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