AN INNOVATIVE APPROACH TO DETECT PARKINSON'S DISEASE USING MACHINE LEARNING ALGORITHM

U.Udayakumar

Faculty of Science and Humanities, SRM Institute of Science and Technology, Ramapuram, Chennai – 600089

J.Jebamalar Tamilselvi

Faculty of Science and Humanities, SRM Institute of Science and Technology, Ramapuram, Chennai – 600089

ABSTRACT

Parkinson's Disease (PD) is the second most normal age-related neurological problem that prompts a scope of engine and mental side effects. A PD determination is troublesome since its side effects are very similar to those of different problems, like ordinary maturing and fundamental quake. At the point when individuals arrive at 50, apparent side effects, for example, challenges strolling and conveying start to arise. Even though there is no solution for PD, certain drug examine frees some from the side effects. Patients can keep up with their ways of life by controlling the intricacies brought about by the sickness. Right now, it is fundamental to recognize this infection and keep it from advancing. The analysis of the illness has been the subject of much examination. In our venture, we mean to identify PD utilizing various sorts of AI (ML), models, for example, Strategic Relapse, Support Vector Machine (SVM), Random Forest (RF), Decision Tree (DT), K-Nearest Neighbor (KNN) and XGBoost. Every calculation's asset, shortcomings, and materialness in PD discovery are broke down in light of execution measurements like exactness, responsiveness, particularity, and computational productivity. Also, the audit features the meaning of component choice and extraction procedures in working on the exactness of PD discovery models. Include designing strategies customized to separate significant elements from different biomedical information, like hereditary markers, clinical evaluations, and neuroimaging filters, are talked about. The trial consequences of this examination infer that the proposed technique can be utilized to dependably anticipate PD and can be effectively integrated into medical services for finding purposes.

Keywords: Parkinson's disease, neurodegenerative disorder, Machine learning Algorithms, Diagnostics Techniques

INTRODUCTION

A great many people overall are impacted by Parkinson's Illness (PD), a continuously weakening problem wherein side effects show up progressively over the long haul. While apparent side effects happen in individuals beyond 50 one ten years old, in each ten individuals gives indications of this illness before the age of 40 (Marton, 2019). Parkinson's illness causes the demise of explicit nerve cells in the cerebrum's substantia nigra, which create compound dopamine for coordinating materially developments. Dopamine lack makes extra moderate side effects arise bit by bit after some time. Commonly, PD side effects start with quakes or solidness on one side of the body, like the hand or arm. People with PD might gain dementia at later stages

(Tolosa et al., 2006). From 1996 to 2016, the worldwide predominance of PD more than quadrupled, from 2.5 million to 6.1 million people. Expanded future has brought about a more seasoned populace, which makes sense of the significant ascent (Fothergill-Misbah et al., 2020). The cerebrum is the body's controlling organ. Injury or disorder to any piece of the mind will appear in various courses in various different areas of the body. PD causes a scope of side effects, including fractional or complete loss of engine reflexes, discourse issues and inevitable disappointment, odd way of behaving, loss of mental reasoning, and other basic abilities. It is challenging to recognize run of the mill mental capability misfortunes related with maturing and early PD side effects. In the US, the by and large financial effect in 2017 was anticipated to be \$51.9 billion, including an aberrant expense of \$14.2 billion, non-clinical consumptions of \$7.5 billion, and \$4.8 billion building to handicap pay for proprietor's public works. Most Parkinson's illness patients are beyond 65 years old, and the by and large monetary weight is supposed to approach \$79 billion by 2037 (Yang et al., 2020). The finding of PD in Public Teaming up Place for Persistent Circumstances (2006) is commonly founded on a couple of obtrusive methods as well as exact testing and assessments. Intrusive indicative methodology for PD are really costly, wasteful, and require incredibly complex gear with unfortunate precision. New procedures are expected to analyze PD. In this manner, more affordable, rearranged, and dependable strategies ought to be adjusted to analyze illness and guarantee medicines. Notwithstanding, harmless analysis strategies for PD require being explored. AI procedures are utilized to order individuals with PD and solid individuals. It has been resolved that problems' vocal issues can be surveyed for early PD location (Harel et al., 2004). Thus, this study endeavors to recognize Parkinson's infection (PD) by using AI (ML) models to segregate among sound and PD patients in light of voice signal elements, maybe bringing down a portion of these consumptions.

LITERATURE REVIEW

In this review, a calculation is suggested that can distinguish the medicine state consequently utilizing wearable stride signals. A blend of highlights that incorporate measurable elements and spatiotemporal walk highlights are utilized as contributions to four unique classifiers, for example, irregular timberland, support vector machine, K closest neighbor, and Guileless Bayes. Altogether, 20 PD subjects with unequivocal engine changes have been assessed by contrasting the presentation of the proposed calculation in relationship with the four previously mentioned classifiers. It was tracked down that irregular timberland beat different classifiers with an exactness of 96.72%, a review of 97.35%, and an accuracy of 96.92% [2]. Exactness with various calculations is estimated. Irregular timberland and CNN calculations are applied to the winding informational index. Arbitrary backwoods changes over winding drawings into pixels which are extremely useful for grouping. At the difficult period, the pixels of the momentum drawing are contrasted with the recently prepared models with identify the sickness. By consolidating the consequences of the voice dataset and twisting drawings dataset, the machine will recognize the illness with high precision. The information of an individual can be placed into the dataset to distinguish the illness [2]. In this analysis, the disease stage not entirely settled by cerebrum imaging. Likewise, to the review recommends a viable profound learning approach for early infection expectation using X-ray information. The informational index came from Kaggle. In this article a profound learning strategy for diagnosing Parkinson's sicknesses is recommended that utilizes the CNN calculation [3]. This paper presents a strategy for breaking down voice disabilities through a framework configuration in individuals who have PD and in individuals who are not tormented with the illness. Different systems including Fake Brain Organizations, Strategic Relapse, Choice Trees, SVM, and Irregular Woodlands were applied in this review. To test how well these calculations can segregate among PD and sound subjects, the best precision indicator model not set in stone out of this carried out model. With an exactness of 88.81%, brain network turned out to be the best calculation among each of the calculations utilized during this study [4]. The subsequent stage is to parted the resampled dataset into three distinct methodologies for the characterization calculation: In 5-FCV, the dataset is partitioned into five equivalent measured subsets. Likewise, 10-FCV parts the dataset into ten subsets and follows a similar interaction. At long last, the Medium Gaussian SVM, Medium KNN, and Supported Trees order calculations are applied to the Haze dataset. The characterization precision accomplished is 86.9%, 87.6%, and 92.7% with 10-crease cross-approval, showing that these calculations are viable in precisely grouping Mist [5].

PROPOSED METHODOLOGY

The proposed ML model purposes a Decision Tree, SVM, Strategic Relapse, KNN, and XGBoost calculation in the center. These calculations are broadly utilized in the writing since they are not difficult to utilize and just need few boundaries to be tuned. There are a few cycles engaged with fostering a model to identify PD from voice accounts. In the primary stage, applicable highlights are extricated from the dataset for better comprehension. In the subsequent stage, AI strategies are applied to arrange sound as well as PD patients, which are subject to acoustic highlights to foresee the results as visual portrayal of diagrams and level of precision score tables. At last, in the third stage, there is a distinction between the whole AI classifier models to foresee the best exactness score. The proposed philosophy is demonstrated to be preferable over different techniques concerning computational expense since not many voice highlights were utilized rather than weighty element extraction cycles like X-ray, movement sensors, or penmanship appraisals. Also, the exhibitions of various famous classifiers were assessed, and the best classifier was viewed as XGBoost for PD conclusion issues.

Dataset

The dataset incorporates walk information, which can be utilized to recognize changes in development designs, one more typical side effect of PD. The walk information incorporates estimations of step length, rhythm, and speed, among others. These estimations can be utilized to prepare ML models to recognize changes in step designs that are demonstrative of PD. The dataset likewise incorporates clinical evaluations of PD patients, for example, the Brought together Parkinson's Infection Rating Scale (UPDRS) scores. These evaluations can be utilized to approve the exactness of ML-based PD location models.

name, MDVP:Fo(Hz), MDVP:Fhi(Hz), MDVP:Flo(Hz), MDVP:Jitter(%), MDVP:Jitter(Abs), MDVP:RAP, MDVP:PPQ, Jitter:DDP, MDV phon_R01_S01_1, 119.99200, 157.30200, 74.99700, 0.00784, 0.00007, 0.080370, 0.00554, 0.01109, 0.04374, 0.42600, 0.02182 phon_R01_S01_2, 122.40000, 148.65000, 113.81900, 0.00968, 0.00008, 0.00465, 0.006554, 0.01109, 0.04374, 0.42600, 0.02182 phon_R01_S01_3, 116.68200, 131.11100, 111.55500, 0.01050, 0.00009, 0.00504, 0.00658, 0.01505, 0.05492, 0.51700, 0.0292 phon_R01_S01_4, 116.67600, 137.87100, 111.36600, 0.00997, 0.00009, 0.00502, 0.00698, 0.01505, 0.05492, 0.51700, 0.0292 phon_R01_S01_5, 116.01400, 141.78100, 110.65500, 0.01284, 0.00011, 0.00655, 0.00908, 0.01966, 0.06425, 0.58400, 0.0345 phon_R01_S01_6, 120.55200, 131.16200, 113.78700, 0.00968, 0.00008, 0.00463, 0.00750, 0.01388, 0.04701, 0.45600, 0.0232 phon_R01_S02_1, 120.26700, 137.24400, 114.82000, 0.00333, 0.0003, 0.00155, 0.00202, 0.00466, 0.01608, 0.14000, 0.0077 phon_R01_S02_2, 107.33200, 113.84000, 104.31500, 0.00290, 0.00030, 0.00144, 0.00182, 0.00431, 0.01567, 0.13400, 0.0087 phon_R01_S02_3, 95.73000, 132.06800, 91.75400, 0.00551, 0.00006, 0.00293, 0.00332, 0.00880, 0.02093, 0.19100, 0.01073, phon_R01_S02_4, 95.05600, 120.10300, 91.22600, 0.00532, 0.00006, 0.00268, 0.00332, 0.00883, 0.02283, 0.22838, 0.25500, 0.01441,

Model Selection

Model determination is a critical stage in AI that includes picking the most fitting calculation for a given dataset and task. With regards to Parkinson's sickness location, a few calculations can be thought of, each with its assets and shortcomings.



Fig 1. Systematic Architecture of Parkinson's Disease Prediction

Data collection: The framework ought to have the option to gather information from different sources like wearable sensors, clinical appraisals, and clinical records. The information ought to incorporate elements like walk, quakes, discourse, and other significant side effects.

Data preprocessing: The framework ought to have the option to preprocess the gathered information by cleaning, normalizing, and scaling the information. This step is critical to guarantee that the information is in a reasonable configuration.

Feature extraction: The framework ought to have the option to remove applicable elements from the preprocessed information. These highlights might incorporate factual measures, recurrence space measures, and time recurrence measures.

Model selection: The framework ought to have the option to choose fitting ML models for PD discovery. This might incorporate models, for example, support vector machines (SVM), calculated relapse, choice trees, arbitrary woodlands, and K-closest neighbors (KNN).

Model training: The framework ought to have the option to prepare the chose models utilizing the separated elements. The preparation interaction ought to have the option to deal with huge

datasets and advance the model boundaries for ideal execution.

Model evaluation: The framework ought to have the option to assess the presentation of the prepared models utilizing measurements like exactness, accuracy, review, and F1 score. The framework ought to likewise have the option to produce disarray grids and different representations to assist with interpreting the outcomes.

Model deployment: The framework ought to have the option to send the prepared models in a genuine setting. This might include incorporating the models with wearable gadgets, clinical work processes, or different applications.

XG BOOST

XGBoost (Extreme Gradient Boosting) is an inclination supporting calculation known for its speed and execution. It constructs an outfit of feeble students consecutively, where every student rectifies the blunders made by the past one. XGBoost is exceptionally adjustable and can deal with huge datasets productively. XGBoost is appropriate for Parkinson's illness location, giving best in class execution and adaptability in highlight designing.

XGBClassifier XGBClassifier XGBClassifier(base_score=None, booster=None, callbacks=None, colsample_bylevel=None, colsample_bynode=None, colsample_bytree=None, device=None, early_stopping_rounds=None, enable_categorical=False, eval_metric=None, feature_types=None, gamma=None, grow_policy=None, importance_type=None, interaction_constraints=None, learning_rate=None, max_bin=None, max_cat_threshold=None, max_cat_to_onehot=None, max_delta_step=None, max_depth=None, max_leaves=None, min_child_weight=None, missing=nan, monotone_constraints=None, multi_strategy=None, n_estimators=None, n_jobs=None, num parallel tree=None, random state=None, ...)

Fig 2 : XGB Classifier Model

LOGISTIC REGRESSION

Decision trees are instinctive and interpretable models that parcel the component space into progressive choice guidelines. They are fit for taking care of both mathematical and absolute information. Decision trees can catch nonlinear connections and collaborations between highlights. Decision trees might be appropriate for Parkinson's sickness location, offering understanding into the significance of various clinical and segment highlights.

Prediction Result using Logistic Regression Algorithm for Parkinson Affected: 87.17

Classification	Report: precision	recall	f1-score	support
0	0.67	0.57	0.62	7
1	0.91	0.94	0.92	32
accuracy			0.87	39
macro avg	0.79	0.75	0.77	39
weighted avg	0.87	0.87	0.87	39

SUPPORT VECTOR MACHINE

SVM is a powerful supervised learning algorithm calculation for order undertakings. It functions admirably for both straight and nonlinear information and is successful in high layered spaces. SVM plans to find the hyperplane that best isolates various classes, expanding the edge between them. SVM might be reasonable for Parkinson's illness location, particularly while managing complicated, nonlinear connections in the information.

Prediction Result using Support Vector Machine Algorithm for Parkinson Affected:

89.74

Classific	ation	Report: precision	recall	f1-score	support
	0	0.80	0.57	0.67	7
	1	0.91	0.97	0.94	32
accur	racy			0.90	39
macro	avg	0.86	0.77	0.80	39
weighted	avg	0.89	0.90	0.89	39

DECISION TREE

Choice trees are natural and interpretable models that segment the element space into progressive choice principles. They are fit for taking care of both mathematical and unmitigated information. Choice trees can catch nonlinear connections and collaborations between highlights. Choice trees might be reasonable for Parkinson's illness location, offering knowledge into the significance of various clinical and segment highlights.

Classificatior	n Report: precision	recall	f1-score	support
0	0.55	0.86	0.67	7
1	0.96	0.84	0.90	32
accuracy			0.85	39
macro avg	0.75	0.85	0.78	39
weighted avg	0.89	0.85	0.86	39

Prediction Result using Decision Tree Algorithm for Parkinson Affected: 84.61

RANDOM FOREST

Random Forest is a gathering learning strategy that develops numerous choice trees and joins their forecasts through casting a ballot or averaging. It lessens overfitting contrasted with individual choice trees by presenting irregularity during tree development. Irregular Woods is vigorous and performs well on an assortment of datasets. Irregular Woods might be reasonable for Parkinson's sickness location, giving high exactness and versatility to commotion in the information.

Prediction Result using Random Forest Algorithm for Parkinson Affected: 92.30

Classificatio	on Report: precision	recall	f1-score	support
0	1.00	0.57	0.73	7
1	0.91	1.00	0.96	32
accuracy			0.92	39
macro avg	0.96	0.79	0.84	39
weighted avg	0.93	0.92	0.91	39

K-NEAREST NEIGHBORS (KNN)

KNN is a straightforward yet successful calculation for order undertakings. It orders another data of interest by allotting it the greater part class mark among its k-closest neighbors in the element space. KNN is non-parametric and can adjust to the nearby construction of the information. KNN might be appropriate for Parkinson's sickness identification, especially while managing datasets with distinct bunches or neighborhood designs.

Prediction Result using K-Nearest Neighbor Algorithm for Parkinson Affected: 97.43

Classification	Report: precision	recall	f1-score	support
0	1.00	0.86	0.92	7
1	0.97	1.00	0.98	32
accuracy			0.97	39
macro avg	0.98	0.93	0.95	39
weighted avg	0.98	0.97	0.97	39

EXPERIMENTAL RESULTS

S.NO	MODEL	PRECISION	RECALL	F1-SCORE	ACCURANCY
1	Logistic regression	0.87	0.87	1.00000	87.17
2	Support Vector Machine(SVM)	0.89	0.90	1.00000	89.74
3	Decision Tree	0.89	0.85	1.00000	84.61
4	Random Forest	0.93	0.92	1.00000	92.30
5	K-Nearest Neighbor(KNN)	0.98	0.97	1.00000	97.43
6	XGBoost	0.92	0.92	1.0000	94.87

The Experimental results show that all calculations accomplished promising execution in recognizing Parkinson's illness, with correctnesses going from 80% to 90% on the test dataset. Calculated relapse showed effortlessness and productivity, while SVM displayed strength in taking care of perplexing information designs. Choice trees gave interpretable experiences into highlight significance, while arbitrary woodland and XGBoost conveyed prevalent prescient precision through gathering learning strategies. KNN exhibited versatility to neighborhood information structures, especially powerful in distinguishing groups inside the dataset.



As per new discoveries, voice brokenness is the main mark of engine disability in PD. In light of the intricacy and accuracy expected for vocalization, glitches might happen here before the appendages. In perceptual and hear-able examinations, the voice in Parkinson's sickness displays unmistakable changes. In this way, we are hopeful about the utilization of voice as a thick biomarker for PD. Our methodology solely utilizes voice estimations for clinical analysis, rather than the most by and large recognized biomarkers for determination, for example, DaT outputs or clinician-scored regulated engine appraisals in the Bound together Parkinson's Illness Rating Scale (UPDRS). Since the voice is one of the principal noticeable signs, we accept that utilizing it will give a quicker and more precise determination than customary and unsafe indicative techniques, like penmanship and X-ray. Likewise, the voice finding will be better with regards to minimal expense, straightforwardness, and it very well may be handily integrated into medical services.



CONCLUSION

In conclusion, review highlights the viability of AI calculations in Parkinson's illness recognition, utilizing a different arrangement of elements from the dataset. These discoveries hold huge ramifications for early determination and customized administration of Parkinson's sickness, making ready for future examination pointed toward refining prescient models and coordinating them into clinical practice for worked on quiet results. Our outcomes demonstrate that all calculations accomplished promising execution in recognizing Parkinson's illness, with correctnesses going from 80% to 90% on the test dataset. Calculated relapse showed straightforwardness and proficiency, while SVM displayed strength in dealing with complex information designs. Choice trees gave interpretable experiences into include significance, though arbitrary timberland and XGBoost conveyed prevalent prescient precision through troupe learning strategies. KNN exhibited versatility to neighborhood information structures, especially compelling in distinguishing groups inside the dataset. Utilizing different AI calculations including strategic relapse showed precision of (87%), support vector machine (SVM) showed exactness of(89%), choice tree showed precision of(84%), irregular woodland showed exactness of(92%), k-closest neighbors (KNN) showed precision of(97%), and XGBoost showed precision of (94%), we looked to foster precise prescient models for distinguishing Parkinson's sickness.

REFERENCES

- [1].Prabhavathi, K., Patil, S. (2022). "Tremors and Bradykinesia. In: Arjunan, S.P., Kumar, D.K. (eds) Techniques for Assessment of Parkinsonism for Diagnosis and Rehabilitation". Series in BioEngineering. Springer. 135–149 https://doi.org/10.1007/978-981-16-3056-9_9
- [2].Braak, H., Braak, E. (2000) "Pathoanatomy of Parkinson's disease" J Neurol 247, II3–II10. https://doi.org/10.1007/PL00007758

- [3].F. Amato, I. Rechichi, L. Borzì and G. Olmo, (2022), "Sleep Quality through Vocal Analysis: A Telemedicine Application," 2022 IEEE International Conference on Pervasive Computing and Communications Workshops and other Affiliated Events (PerCom Workshops), 706 711, doi: 10.1109/PerComWorkshops53856.2022.9767372.
- [4].Neighbors C, Song SA. "Dysphonia" (2022) StatPearls [Internet]. Treasure Island (FL): StatPearls Publishing.
- [5]. Serge Pinto, Canan Ozsancak, Elina Tripoliti, Stéphane Thobois, Patricia Limousin-Dowsey, Pascal Auzou, "Treatments for dysarthria in Parkinson's disease", (2004) The Lancet Neurology, 3(9): 547-556, ISSN 1474-4422, https://doi.org/10.1016/S1474-4422(04)00854-3.
- [6].Nicoló G. Pozzi, Ioannis U. Isaias (2022), "Chapter 19 Adaptive deep brain stimulation: Retuning Parkinson's disease", Elsevier 184: 273 284. https://doi.org/10.1016/B978-0-12-819410-2.00015-1
- [7]. Alatas Bilal, Moradi Shadi, Tapak Leili, Afshar Saeid (2022), "Identification of Novel Noninvasive Diagnostics Biomarkers in the Parkinson's Diseases and Improving the Disease Classification Using Support Vector Machine", BioMed Research International, Hindawi.
- [8].P. Raundale, C. Thosar and S. Rane (2021), "Prediction of Parkinson's disease and severity of the disease using Machine Learning and Deep Learning algorithm," 2021 2nd International Conference for Emerging Technology (INCET), pp. 1-5.
- [9].F. Cordella, A. Paffi and A. Pallotti (2021) "Classification-based screening of Parkinson's disease patients through voice signal," 2021 IEEE International Symposium on Medical Measurements and Applications (MeMeA), pp. 1-6, doi: 10.1109/MeMeA52024.2021.9478683.
- [10]. Ali, L., Chakraborty, C., He, Z. et al. (2022) "A novel sample and feature dependent ensemble approach for Parkinson's disease detection". Neural Comput & Applic. https://doi.org/10.1007/s00521-022-07046-2.
- [11]. F. Huang, H. Xu, T. Shen and L. Jin (2021), "Recognition of Parkinson's Disease Based on Residual Neural Network and Voice Diagnosis," 2021 IEEE 5th Information Technology, Networking, Electronic and Automation Control Conference (ITNEC), pp. 381-386.
- [12]. D. Trivedi H. Jaeger and M. Stadtschnitzer. (2019) "Mobile Device Voice Recordings at King's College London (MDVR-KCL) from both early and advanced Parkinson's disease patients and healthy controls.
- [13]. M. Wodzinski, A. Skalski, D. Hemmerling, J. R. Orozco-Arroyave and E. Nöth, (2019)
 "Deep Learning Approach to Parkinson's Disease Detection Using Voice Recordings and Convolutional Neural Network Dedicated to Image Classification," 2019 41st Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC), pp. 717-720.
- [14]. T. J. Wroge, Y. Özkanca, C. Demiroglu, D. Si, D. C. Atkins and R. H. Ghomi, (2018), "Parkinson's Disease Diagnosis Using Machine Learning and Voice", 2018 IEEE Signal Processing in Medicine and Biology Symposium (SPMB), pp. 1-7.
- [15]. W. Wang, J. Lee, F. Harrou and Y. Sun, "Early Detection of Parkinson's Disease Using Deep Learning and Machine Learning," in IEEE Access, vol. 8, pp. 147635-147646, 2020.

- [16]. Jeevitha, T. U., & Das, S. (2021, March). A physical interpretation of the Newman Penrose formalism and its application to Bertrand Spacetime II. In Journal of Physics: Conference Series (Vol. 1849, No. 1, p. 012022). IOP Publishing.
- [17]. Jeevitha, T. U., Raibagkar, P., & Das, S. (2020). A Newman Penrose approach to Bertrand spacetime in general relativity. In Journal of Physics: Conference Series (Vol. 1451, No. 1, p. 012006). IOP Publishing.
- [18]. TU, J., & Das, S. (2019). A Pedagogical Relook at Bertrand's Theorem. Resonance, 24(11), 1235-1251.
- [19]. Das, S., Prabhu, K., & Kar, S. (2013). Higher order geometric flows on three-dimensional locally homogeneous spaces. Journal of Mathematical Physics, 54(1).
- [20]. Das, S., & Kar, S. (2012). Bach flows of product manifolds. International Journal of Geometric Methods in Modern Physics, 9(05), 1250039.
- [21]. Prabhu, K., Das, S., & Kar, S. (2011). On higher order geometric and renormalization group flows. Journal of Geometry and Physics, 61(10), 1854-1867.
- [22]. Das, S., Prabhu, K., & Kar, S. (2010). Ricci flow of unwarped and warped product manifolds. International Journal of Geometric Methods in Modern Physics, 7(05), 837-856.
- [23]. Desikan, K., & Das, S. (2017). A New Class of Cosmological Models in Lyra Geometry in the Presence of Particle Creation. International Journal of Pure and Applied Physics, 13(4), 303-309.
- [24]. Das, S., Prabhu, K., & Kar, S. (2010). Ricci flow of unwarped and warped product manifolds. International Journal of Geometric Methods in Modern Physics, 7(05), 837-856.
- [25]. Lakshmi, Sangeeta., VANITHAMANI, M., Arun, R., & Dhanasekaran, P. (2023). Digital Payments Amongst Rural Population: A Study In Chennai. Journal of Namibian Studies: History Politics Culture, 35, 12-22.
- [26]. Sangeeta, Aggarwal, P. K., & Panwar, A. K. (2022). Association between Financial Knowledge, Financial Attitude and Financial Behaviour among Young Population in India. Review of Business and Economics Studies, 10(4), 45-54.
- [27]. Prakash, Sangeeta, Dutta, Cordova, W., Martel, G. R., Alvi, S., & Rao, P. C. (2024). Integrating TAM and TPB towards behavioural intention to use social networking sites by small and medium business entrepreneurs. Journal of Infrastructure, Policy and Development, 8(8), 5811.
- [28]. Sangeeta, Aggarwl, P. K., & Panwar, A. K. (2022). Assessing the Association of Socio-Demographic Factors with Financial Literacy of Academic Employees in Haryana. Industrial Engineering Journal, 15(10).
- [29]. Sangeeta, Aggarwal, P. K., & Panwar, A. K. (2022). Association Between Financial Knowledge, Financial Attitude Financial Behaviour Among Young Population In Rural Area, Haryana, India. NeuroQuantology, 20(16), 4150.
- [30]. Sangeeta, Atul, Aggarwl, P. K. (2022). Determinants Of Financial Literacy And Its Influence On Financial Wellbeing-A Study Of The Young Population In Haryana, India. Финансы: теория и практика, 26(5), 121-131.