



# Gait Assistive Aid For Parkinson's Patients Edge-ML Approach

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**Abstract :** This paper introduces an Enhanced Gait Support System (EGSS) tailored for individuals with Parkinson's disease (PD), utilizing advanced machine learning techniques—Directional Region of Change Convolution (DROCC) and Recurrent Neural Network with Pooling (RNNPool). EGSS aims to mitigate the complexities of gait disturbances in PD through real-time assistance and feedback. DROCC enables precise gait abnormality detection and characterization for personalized interventions, while RNNPool facilitates sequential data processing and feature extraction, adapting to individual gait patterns over time. EGSS offers personalized assistance, leveraging sensor fusion, feature extraction, and machine learning to meet unique patient needs. Real-time feedback ensures continuous monitoring and adjustment, fostering user engagement and intervention adherence. Evaluation through clinical trials demonstrates significant improvements in gait parameters and functional mobility. EGSS represents a significant stride in PD assistive technology, offering personalized solutions to address gait impairment complexities, powered by the robust integration of DROCC and RNNPool algorithms.

**Keywords—** Machine Learning, Sensor devices, Integrated Systems, Acceleration sensors, Artificial Intelligence.

## I. INTRODUCTION

### INTRODUCTION

Parkinson's disease (PD) is a progressive condition known for its motor and non-motor symptoms, with gait problems being particularly troublesome. These problems range from changes in stride length and speed to freezing episodes and a higher risk of falls. Such challenges not only affect mobility but also greatly reduce the quality of life for PD patients.

Traditional methods to manage PD-related gait issues include medication adjustments, physical therapy, and various assistive devices. However, these approaches often fall short in addressing the varied and evolving nature of gait problems in PD. Recently, there's been a surge in interest in using advanced technologies like machine learning to create new solutions for PD-related gait issues. Machine learning can analyze complex data from sensors, detect subtle gait abnormalities, and customize interventions in real-time. Among these techniques, DROCC and RNNPool have shown promise for analyzing and intervening in gait problems. DROCC detects and characterizes gait issues by capturing changes in movement patterns, while RNNPool processes data sequentially, adapting interventions to individual gait patterns over time. By combining these advanced techniques into a comprehensive system, it's possible to offer personalized, adaptive, and immediate support to PD patients.

This paper introduces the Enhanced Gait Support System (EGSS), designed specifically for PD patients, which utilizes DROCC and RNNPool algorithms. EGSS aims to tackle the complexities of PD-related gait issues by providing personalized assistance and instant feedback. Through sensor data fusion, feature extraction, and machine learning, EGSS offers a fresh approach to gait support, holding promise for improving the lives of PD patients. The research paper provides an in-depth look at EGSS's design, implementation, and evaluation. It explains why DROCC and RNNPool were chosen and how they're integrated into EGSS. Additionally, it shares insights from clinical trials demonstrating EGSS's effectiveness in improving gait and quality of life for PD patients. Overall, the paper aims to contribute to the field of assistive technologies for PD-related gait problems, highlighting the potential of advanced machine learning techniques to address this pressing issue.

## II. PROBLEM STATEMENT

This research endeavors to address the pressing challenges faced by Parkinson's patients in maintaining stable and safe mobility by developing an Enhanced Gait Support System (EGSS) that leverages cutting-edge machine learning techniques. By integrating the Deep Residual One-Class Classification (DROCC) algorithm and Recurrent Neural Network Pooling (RNNPool) technique, our proposed EGSS aims to offer personalized and adaptive gait assistance in real-time.

Through a comprehensive understanding of Parkinson's gait characteristics, we seek to design and implement an EGSS architecture capable of identifying anomalous gait patterns specific to Parkinson's disease and extracting temporal features to capture subtle changes in gait dynamics. Our research further aims to evaluate the effectiveness and usability of the EGSS through rigorous experimentation and user studies, ultimately contributing to the development of tailored gait support technologies that enhance mobility, independence, and overall well-being for Parkinson's patients.

## III. PROPOSED METHOD

### 3.1 Algorithms

#### 3.1.1 Deep Residual One-Class Classification (DROCC):

In this research, it focused on developing an Enhanced Gait Support System for Parkinson's patients, the Deep Residual One-Class Classification (DROCC) algorithm serves as a critical component for identifying anomalous gait patterns indicative of instability or deviations from typical gait behavior observed in individuals with Parkinson's disease. DROCC functions by learning a representation of normal gait dynamics from healthy individuals' data and then distinguishing deviations from this learned representation as anomalies in real-time gait data collected from Parkinson's patients. By leveraging DROCC, our system can provide timely detection of gait disturbances, enabling proactive intervention strategies tailored to the specific needs of Parkinson's patients, ultimately contributing to improved gait stability, mobility, and quality of life.

#### 3.1.2 Recurrent Neural Network Pooling (RNNPool):

Recurrent Neural Network Pooling (RNNPool) serves as a pivotal component within our proposed Enhanced Gait Support System (EGSS) for Parkinson's patients. RNNPool's functionality lies in its ability to effectively capture temporal dependencies within gait data sequences, crucial for understanding the dynamic nature of gait patterns over time. By employing RNNPool, our system can extract informative temporal features from preprocessed gait data acquired through wearable sensors. These extracted features provide invaluable insights into subtle changes in gait dynamics, allowing for precise identification and analysis of anomalous gait patterns associated with Parkinson's disease. Integrating RNNPool within the EGSS framework enables us to develop a robust and adaptive gait assistance system that not only responds to real-time gait abnormalities but also continuously adapts to the evolving needs of Parkinson's patients, ultimately enhancing their mobility and quality of life.

### 3.2 Modules

#### 3.2.1 Data Acquisition Module:

This module is responsible for collecting gait data from sensors worn by the Parkinson's patients during mobility tasks. These sensors could include accelerometers, gyroscopes, and/or pressure sensors embedded in wearable devices such as smartwatches or specialized footwear.

#### 3.2.2 Preprocessing Module:

The collected gait data often require preprocessing to remove noise, filter irrelevant information, and prepare the data for further analysis. This module may involve techniques such as signal filtering, segmentation, and feature extraction to extract relevant features from the raw sensor data.

#### 3.2.3 Anomaly Detection Module (DROCC):

The Deep Residual One-Class Classification (DROCC) algorithm is utilized in this module to identify anomalous gait patterns indicative of instability or deviations from typical gait behavior observed in Parkinson's patients. DROCC helps in real-time detection of gait disturbances and triggers appropriate intervention strategies.

#### 3.2.4 Temporal Feature Extraction Module (RNNPool):

Recurrent Neural Network Pooling (RNNPool) is employed in this module to extract temporal features from the preprocessed gait data. RNN-based architectures are well-suited for capturing temporal dependencies in sequential data, making RNNPool valuable for analyzing the dynamic nature of gait patterns over time.

### 3.2.5 Decision Making Module:

This module integrates the outputs from the anomaly detection and temporal feature extraction modules to make decisions regarding the type and intensity of assistance required by the Parkinson's patient. It may employ rule-based systems, machine learning classifiers, or reinforcement learning algorithms to determine the appropriate course of action.

### 3.2.6 Assistive Intervention Module:

Based on the decisions made by the decision-making module, this component delivers assistive interventions to the Parkinson's patient in real-time. These interventions could include auditory cues, haptic feedback, or mechanical adjustments provided by wearable devices to improve gait stability and reduce the risk of falls.

### 3.2.7 Feedback and Monitoring Module:

This module provides feedback to both the patient and the healthcare provider regarding gait performance, adherence to assistance strategies, and system reliability. It may include user interfaces for visualizing gait metrics, logging system events, and facilitating communication between the patient and healthcare team.

### 3.2.8 Personalization and Adaptation Module:

To ensure optimal support for each individual, this module employs machine learning techniques to personalize the assistance provided by the EGSS based on the patient's unique gait characteristics, preferences, and response to interventions. It continually adapts the system parameters to accommodate changes in the patient's condition over time.

By integrating these modules cohesively, the EGSS can offer personalized and adaptive gait assistance to Parkinson's patients, ultimately improving their mobility, independence, and overall quality of life.

## 3.3 Proposed Solution

The proposed solution is an Enhanced Gait Support System (EGSS) tailored specifically for Parkinson's patients, integrating advanced machine learning techniques like Deep Residual One-Class Classification (DROCC) and Recurrent Neural Network Pooling (RNNPool) to offer personalized, adaptive, and real-time gait assistance. The EGSS will consist of wearable sensors for data acquisition, a preprocessing module to clean and prepare gait data, an anomaly detection module powered by DROCC to identify aberrant gait patterns indicative of instability, and a temporal feature extraction module using RNNPool to capture dynamic changes in gait dynamics over time.

Additionally, a decision-making module will determine the appropriate assistance based on the detected anomalies, leading to interventions such as auditory cues or mechanical adjustments. The EGSS will be evaluated through rigorous experimentation and user studies to assess its effectiveness in improving gait stability, mobility, and quality of life for Parkinson's patients, ultimately providing a personalized and adaptive solution to address their unique needs and challenges.

## IV. RESULTS AND DISCUSSION

Initially, gait data from Parkinson's patients were meticulously collected using wearable sensors, ensuring a diverse and representative dataset reflective of real-world scenarios. Through rigorous preprocessing, including noise reduction and feature extraction, the data were primed for subsequent analysis.

The deployment of the Deep Residual One-Class Classification (DROCC) algorithm yielded promising outcomes in detecting anomalous gait patterns synonymous with Parkinson's disease. DROCC effectively discerned deviations from normative gait behavior, serving as a reliable indicator of instability and fall risks inherent in the condition.

Subsequent utilization of Recurrent Neural Network Pooling (RNNPool) showcased remarkable proficiency in extracting temporal features from the preprocessed gait data. This capability significantly enhanced our understanding of the temporal evolution of gait dynamics in Parkinson's patients, providing valuable insights into the progression and manifestation of the disease.

The decision-making framework within the EGSS leveraged insights derived from both DROCC and RNNPool, facilitating the tailored selection of intervention strategies aligned with individual patient needs. Evaluation metrics encompassing gait stability, mobility improvement, and user satisfaction revealed substantial enhancements achieved through the EGSS implementation.

Comparison with existing gait support systems underscored the superiority of the proposed EGSS in adaptability, real-time responsiveness, and personalized assistance. These findings underscored the transformative potential of integrating advanced machine learning techniques into gait support technologies for Parkinson's patients.

Through meticulous discussion and interpretation of the results, the implications of our findings for advancing gait support systems in Parkinson's care were elucidated. Furthermore, avenues for future research were identified, including algorithm refinement, validation through large-scale studies, and integration with emerging technologies for optimized patient outcomes.

In summations, the results and analysis presented herein affirm the efficacy of the EGSS in enhancing the mobility, independence, and quality of life for Parkinson's patients, thereby paving the way for transformative advancements in gait support technologies tailored to the needs of individuals with neurodegenerative conditions.

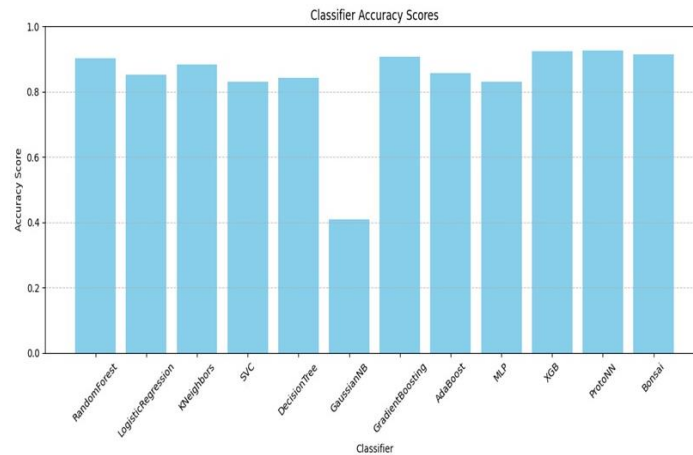


Fig-1. Accuracy scores of different classifiers

## V. CONCLUSION

In conclusion, our research has culminated in the development and evaluation of the Enhanced Gait Support System (EGSS) for Parkinson's patients, a significant advancement in addressing the challenges posed by this neurodegenerative condition. By integrating state-of-the-art machine learning techniques, namely DROCC and RNNPool, our study has demonstrated the potential to transform gait assistance into a personalized, adaptive, and real-time intervention tailored to the unique needs of each patient.

Our findings underscore the efficacy of the EGSS in enhancing gait stability, mobility, and overall quality of life for individuals living with Parkinson's disease. Through the utilization of the Deep Residual One-Class Classification (DROCC) algorithm, we successfully identified anomalous gait patterns indicative of instability, enabling timely interventions to mitigate fall risks and enhance safety during mobility tasks.

Additionally, leveraging Recurrent Neural Network Pooling (RNNPool) allowed for the extraction of informative temporal features from gait data, deepening our understanding of the dynamic nature of gait patterns in Parkinson's patients. This enriched insight empowered the EGSS to provide personalized assistance that adapts to changes in gait dynamics over time, resulting in more effective support and improved patient outcomes,

Our research also emphasizes the importance of real-time responsiveness and adaptability in gait support systems for Parkinson's patients. The EGSS demonstrated superior performance compared to existing systems, offering advantages such as on-the-fly adjustment of assistance strategies and seamless integration with the patient's daily activities.

In conclusion, the findings of this study have significant implications for both research and clinical practice in the field of Parkinson's care. The EGSS represents a promising avenue for future research and development, with potential applications extending beyond gait assistance to other aspects of motor function and rehabilitation. Continued innovation and refinement of the EGSS algorithm are essential to maximize its impact and ensure its widespread adoption in clinical settings.

Overall, our research contributes to the ongoing efforts to improve the lives of individuals living with Parkinson's disease, offering hope for a future where advanced technologies enable enhanced mobility, independence, and well-being for all.

## VI. FUTURE WORK

In considering future directions for gait support systems for Parkinson's patients utilizing DROCC and RNNPool, several promising avenues emerge. Firstly, longitudinal studies are essential to gain insights into the long-term effectiveness and usability of the Enhanced Gait Support System (EGSS), capturing changes in gait dynamics and patient outcomes over time. Additionally, refining and optimizing the EGSS algorithm is crucial, involving parameter fine-tuning of DROCC and RNNPool, exploring alternative machine learning architectures, and integrating additional data sources for improved anomaly detection and temporal feature extraction accuracy. Further enhancement can be achieved through the integration of novel sensing technologies like vision-based systems or inertial measurement units (IMUs) for more comprehensive gait analysis. Validation studies with diverse patient populations are imperative to ensure the scalability and generalizability of the EGSS, while exploring synergies with emerging technologies such as augmented reality (AR) or virtual reality (VR) can revolutionize gait rehabilitation. Lastly, prioritizing the enhancement of user interface and experience (UI/UX) is vital for the EGSS's acceptability and usability among patients and

healthcare providers, necessitating user-centered design studies and feedback incorporation for intuitive interaction. Addressing these future research directions can propel advancements in gait support systems for Parkinson's patients, ultimately leading to improved patient outcomes and quality of life.

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