



## WEARABLE ACCELEROMETER AND GYROSCOPE SENSORS FOR ESTIMATING THE SEVERITY OF ESSENTIAL TREMOR

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### ABSTRACT :

Essential Tremor (ET) is a neurological disorder characterized by involuntary, rhythmic shaking, primarily affecting the hands and arms, and in severe cases, impacting daily tasks such as eating, writing, or holding objects. The severity of ET can vary significantly from mild, manageable tremors to intense shaking that disrupts quality of life. This project aims to provide continuous, real-time monitoring of ET severity through a wearable device equipped with accelerometer, gyroscope, pulse oximeter, temperature, and pressure sensors. The accelerometer and gyroscope sensors measure tremor intensity and frequency, capturing subtle to severe movements associated with ET. The pulse oximeter monitors heart rate and blood oxygen levels, as tremor severity may increase with changes in heart rate. A temperature sensor helps detect fluctuations in body temperature, which can affect tremor intensity, while a pressure sensor assesses the force applied during actions like drawing, providing insights into fine motor control. When sensor data deviates from normal ranges, a buzzer is triggered, and alert messages are sent to a designated health protector, enabling timely intervention. All sensor data is continuously displayed on an LCD screen and the Blynk Android app through IoT, offering an accessible platform for caregivers and health professionals to track ET progression. This system aids in understanding the fluctuations in ET severity, helping patients and caregivers manage the disorder more effectively and potentially improving the quality of life for ET patients.

**Index Terms—Tremor, Components, Sensor, Wearable.**

### I. INTRODUCTION

Essential tremor (ET) is a neurological disorder that leads to involuntary and rhythmic shaking, most commonly affecting the hands but may also involve the head, voice, legs, or trunk. It is one of the most prevalent movement disorders globally and often worsens over time, impacting the individual's ability to perform daily tasks such as writing, eating, or holding objects. Unlike other tremor-related conditions like Parkinson's disease, essential tremor is typically an action tremor, meaning it occurs during voluntary movements rather than at rest. The exact cause of essential tremor is not fully understood, though it is often hereditary and linked to abnormal brain activity in areas controlling movement, such as the cerebellum.

Traditional methods of assessing tremor severity involve clinical evaluations that are typically limited to brief observations during hospital visits. These assessments often rely on subjective scoring systems, such as the Fahn–Tolosa–Marin Tremor Rating Scale, which depend heavily on the clinician's interpretation. This approach poses several challenges, such as inconsistency between evaluations, limited data on daily tremor fluctuations, and the inability to track long-term progress effectively. In addition, patients may experience varying tremor intensity depending on stress, fatigue, medication timing, or environmental factors—none of which are captured adequately in short, clinical visits. As a result, there is a growing need for objective, continuous, and real-world tremor monitoring solutions that can support more informed clinical decision-making and improve patient quality of life.

Recent technological advancements have opened new possibilities in the field of biomedical engineering and wearable health monitoring. In particular, motion sensors such as accelerometers and gyroscopes have emerged as promising tools for capturing detailed movement data. Accelerometers measure linear acceleration in multiple directions, while gyroscopes detect angular velocity and rotational movement. When used together, they can provide comprehensive data on the amplitude, frequency, and patterns of tremor, offering a more accurate and quantifiable representation of tremor severity. These sensors are compact, affordable, and easily integrated into wearable devices, making them suitable for long-term monitoring outside clinical environments.

This project focuses on the development of a wearable system that utilizes these motion sensors to monitor and estimate the severity of essential tremor in real-time. By embedding an MPU6050 sensor—which combines both a 3-axis accelerometer and a 3-axis gyroscope—onto a wearable device connected

to a microcontroller with Wi-Fi capabilities, such as the ESP32, the collected data can be transmitted wirelessly. The integration of Internet of Things (IoT) technology allows this data to be sent to the cloud, where it can be visualized and analyzed using an Android-based mobile application. The Blynk app is used to display real-time tremor information through user-friendly interfaces such as graphs and gauges, enabling both patients and healthcare providers to monitor changes continuously.

The significance of this system lies in its ability to offer a low-cost, user-friendly, and non-invasive solution for continuous tremor tracking. Unlike traditional systems that require complex setups and clinical supervision, this wearable device can be used by patients at home or in daily settings. It not only empowers patients by providing real-time feedback on their condition but also helps doctors make better-informed treatment decisions based on actual tremor data over time. Moreover, by collecting large volumes of longitudinal tremor data, the system lays the groundwork for future integration of advanced analytics and machine learning algorithms to detect patterns, predict tremor episodes, and recommend personalized treatment plans.

In conclusion, the proposed wearable tremor monitoring system represents a step forward in the management of essential tremor. By combining sensor technology, IoT, and mobile health applications, it bridges the gap between clinical observation and real-life symptom tracking. It enables both proactive care and remote monitoring, making healthcare more accessible and tailored to individual needs. This innovative approach has the potential to significantly enhance the diagnosis, treatment assessment, and daily management of individuals living with essential tremor.

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## II. LITERATURE REVIEW

The study and monitoring of essential tremor (ET) have been areas of growing interest in biomedical engineering and clinical neurology. Over the years, researchers and clinicians have made various attempts to develop accurate, reliable, and patient-friendly methods to measure tremor severity. Traditional assessment methods mostly relied on visual observation and subjective scoring, which limited their accuracy and made them unsuitable for long-term monitoring. With the rise of wearable technology and embedded systems, motion sensors such as accelerometers and gyroscopes have become essential tools for tracking tremor characteristics with high precision and reliability.

Accelerometers have been widely used in tremor studies because they are compact, lightweight, cost-effective, and capable of measuring linear acceleration across three axes. These sensors detect the motion of body parts in straight-line directions, such as vertical, horizontal, and depth-based movements. This makes them ideal for capturing tremors involving up-and-down or side-to-side shaking. Many earlier research projects have implemented accelerometer-based systems to quantify tremor amplitude and frequency. The collected data helped in distinguishing tremors from voluntary motion and analyzing their intensity over time. However, while accelerometers are useful, they provide an incomplete picture of tremor activity. One of their major limitations is the inability to capture rotational or angular motion, which is also a key characteristic of essential tremor, especially in hand and wrist movements.

To overcome this limitation, gyroscope sensors were introduced as a complementary technology. A gyroscope measures angular velocity—essentially how fast an object is rotating around a particular axis. By detecting this rotational movement, gyroscopes help fill the gap left by accelerometers. When accelerometer and gyroscope data are combined, the system is capable of capturing both linear and angular motion, resulting in a more comprehensive understanding of tremor patterns. This combined sensing approach enables better classification of tremor episodes, improved analysis of severity, and a richer dataset for tracking disease progression.

One of the most popular and widely adopted sensor modules for such applications is the MPU6050, which combines a 3-axis accelerometer and a 3-axis gyroscope in a single chip. It is known for its high accuracy, low power consumption, and ease of integration with microcontrollers. The MPU6050 provides six degrees of freedom (DoF) in motion tracking, which allows for highly detailed and responsive tremor measurement. Many academic projects and medical device prototypes have utilized this sensor for applications in Parkinson's disease, essential tremor, and rehabilitation monitoring. However, even with improved sensor capabilities, several limitations remained in earlier research systems.

Many of the early tremor monitoring solutions were focused purely on data collection. These systems often required manual retrieval of data through physical connections such as USB or memory cards. As a result, real-time monitoring was not possible, and patients had to revisit clinical centers to transfer and interpret the data. This made the systems impractical for continuous daily use and limited their usefulness in understanding how tremors behave throughout the day in different scenarios. Additionally, most systems lacked the ability to alert users or healthcare providers immediately in response to changes in tremor severity. Without live feedback, patients could not take proactive measures, and clinicians lacked timely information to adjust treatments.

Moreover, many users—especially older adults—found earlier systems difficult to operate due to technical complexity or lack of user-friendly interfaces. These barriers reduced the potential adoption of wearable monitoring systems in real-life scenarios. A critical challenge was ensuring that tremor monitoring tools were not just technically accurate but also convenient, accessible, and integrated into everyday life. This is where the role of Internet of Things (IoT) technology became transformative.

Our project addresses these limitations by leveraging IoT to create a real-time, user-friendly tremor monitoring solution. We use an ESP32 microcontroller, which comes with built-in Wi-Fi, allowing seamless wireless transmission of sensor data. By connecting the ESP32 to the MPU6050 module, we create a compact wearable system capable of detecting and processing tremor data continuously. This data is then transmitted to the cloud

via Wi-Fi, eliminating the need for physical connections or manual downloads. The entire system becomes autonomous, requiring minimal user input once set up.

The innovation extends further with the integration of a mobile application based on the Blynk platform. Blynk is a customizable app development environment that allows real-time data visualization through widgets like graphs, gauges, and indicators. Using this platform, we have created a smartphone app that displays tremor intensity in real time. This app not only makes the data accessible to users but also allows caregivers and doctors to monitor a patient's condition remotely. Users can carry the device as a wristband or wearable module and view tremor behavior anytime on their phone.

This approach greatly enhances the practicality and usefulness of tremor monitoring systems. By eliminating manual data collection and providing live feedback, the device becomes a powerful tool for continuous health management. It allows patients to recognize their condition more clearly and adopt appropriate responses, such as taking medication or resting during high-tremor periods. For clinicians, it provides a constant stream of objective data, helping them make data-driven treatment decisions. Furthermore, since the data is stored in the cloud, long-term trends and changes can be analyzed over weeks or months, leading to more personalized care strategies.

In summary, the combination of accelerometers, gyroscopes, and IoT integration represents a significant advancement in the field of tremor monitoring. It addresses the shortcomings of earlier systems by enabling accurate, real-time, and user-accessible monitoring. Our project builds on previous research by creating a wearable, internet-connected solution that is both effective and easy to use. This advancement not only supports better clinical outcomes but also improves the daily lives of individuals living with essential tremor.

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### III. PROPOSED SYSTEM

The proposed system is designed to provide a modern, wearable solution for the continuous monitoring and classification of essential tremor severity. It uses a combination of motion sensors, wireless data transmission, cloud-based storage, and a mobile application interface to offer real-time feedback to both users and healthcare professionals. The key concept is to replace traditional, subjective tremor assessments with an objective and automated system that can operate in real-life conditions. By integrating both accelerometer and gyroscope technologies in a wearable module and using the Internet of Things (IoT) to deliver live data, this project aims to fill the gap between clinical monitoring and everyday health tracking.

The system begins with a sensor module that includes an MPU6050, a well-known integrated circuit that combines a 3-axis accelerometer and a 3-axis gyroscope. This sensor is capable of capturing the full range of motion caused by tremors, including linear acceleration and rotational velocity. These two types of data are essential in understanding the characteristics of essential tremor, which often include both shaking and twisting movements of the hand. The MPU6050 is chosen for its compact size, low power consumption, and ability to interface easily with microcontrollers. It is mounted in such a way that it can be worn on the wrist, which is the most commonly affected area in individuals with essential tremor.

The sensor readings are sent to an ESP32 microcontroller, which plays a central role in data processing and transmission. The ESP32 is selected due to its dual-core processing power, built-in Wi-Fi capability, and compatibility with various sensor interfaces. It receives analog and digital signals from the MPU6050, converts them into usable data values, and processes them to determine the intensity and frequency of tremors. These values are used to compute a real-time tremor severity index, which can help in classifying the tremor as mild, moderate, or severe. The ESP32 then establishes a Wi-Fi connection to transmit the processed data wirelessly to the cloud.

To make the system user-accessible, the cloud data is displayed through a smartphone application built on the Blynk platform. Blynk is an IoT app development tool that supports easy integration with microcontrollers like ESP32 and provides ready-to-use widgets for visualizing live data. In this system, the app is configured to show real-time tremor metrics in the form of graphs, value displays, and alerts. The interface is designed to be simple and intuitive so that even users with minimal technical knowledge can understand the output. The application continuously fetches data from the cloud and refreshes the interface, allowing the user to observe how tremors change throughout the day.

One of the major benefits of this proposed system is its capability for long-term, non-invasive monitoring. Unlike traditional methods that only measure tremors during clinical visits, this wearable system collects data continuously. This provides a more accurate representation of how tremors behave in natural environments, influenced by variables like emotional stress, medication adherence, fatigue, and activity levels. As a result, both patients and healthcare providers gain a richer and more reliable dataset to base treatment decisions on. For example, a neurologist could review weekly tremor trends and adjust medication doses accordingly, or a patient could identify specific triggers for tremor spikes in daily life.

The wearable system also enhances convenience and accessibility. It is lightweight, portable, and easy to wear throughout the day, similar to a fitness tracker or smartwatch. There is no need for bulky equipment or complicated setup procedures. The entire system can be operated with a smartphone and a USB-charged device. Since the data is stored on the cloud, it can also be accessed remotely, making it especially useful for telemedicine and rural healthcare settings. Doctors can monitor their patients' conditions without requiring frequent in-person appointments, which saves time and resources for both sides.

Another important advantage of the proposed system is the ability to track tremor progression over time. The Blynk platform enables historical data logging, so users can observe weekly or monthly trends in tremor intensity. This long-term data can help identify patterns in symptom progression, such

as worsening tremors during certain times of day or after physical activity. Over time, the system could be further enhanced with machine learning algorithms that learn from the data and provide predictive insights or automated severity classification.

Additionally, the system is scalable and customizable. The same framework can be extended to monitor tremors in other parts of the body, such as the head or legs, by simply adjusting sensor placement. Multiple sensors can be synchronized and their data merged for more comprehensive analysis. The code running on the ESP32 is open-source and can be modified to include additional features like GPS location tagging, fall detection, or medication reminders. This flexibility makes the system highly adaptable for future development and research.

In conclusion, the proposed system represents an innovative and practical solution for monitoring essential tremor using a wearable, IoT-integrated device. It leverages modern sensor technology to collect detailed motion data, a powerful microcontroller for processing and transmission, and a user-friendly mobile application for visualization and interaction. The system improves upon previous designs by enabling real-time monitoring, long-term data tracking, and remote accessibility. Its simplicity and effectiveness make it a promising tool for both clinical and personal use, contributing to better management and understanding of essential tremor.

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#### IV. Components used

The proposed wearable tremor monitoring system relies on several key components to ensure its functionality, accuracy, and user-friendliness. These components work in conjunction to measure tremor movements, process the data, and deliver real-time feedback to the user. Each component was chosen based on its specific performance characteristics, compatibility with the overall system, and ability to meet the project's goals.

The core component of the system is the MPU6050 sensor, which integrates both an accelerometer and a gyroscope into a single package. The MPU6050 sensor has six degrees of freedom, capable of measuring linear acceleration along the X, Y, and Z axes using its accelerometer, as well as detecting rotational motion around the same three axes using its gyroscope. This combination of accelerometer and gyroscope makes the MPU6050 ideal for capturing both the translational and rotational components of tremor motion, which are critical for accurately characterizing essential tremors. The sensor operates on a low voltage and is relatively power-efficient, making it suitable for battery-operated wearable devices.

The ESP32 microcontroller is the heart of the system, serving as the interface between the sensor and the mobile application. It is a dual-core processor with built-in Wi-Fi and Bluetooth capabilities, which makes it well-suited for wireless data transmission. The ESP32 processes the raw data from the MPU6050, filters it to remove noise, and computes relevant metrics such as tremor frequency and amplitude. The microcontroller then uploads this processed data to the cloud via its Wi-Fi connection, where it can be accessed by the mobile application. In addition to its processing power and wireless connectivity, the ESP32 is relatively compact and low-cost, making it an ideal choice for a small, wearable device.

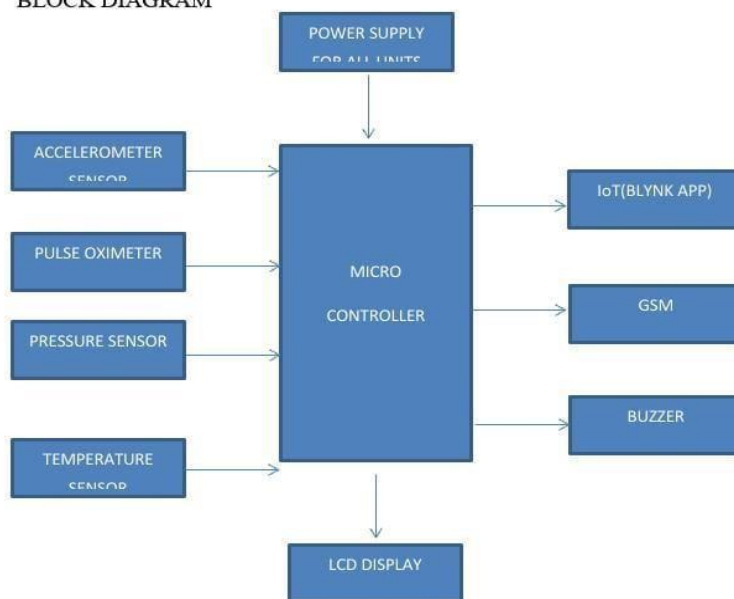
The Blynk mobile app serves as the user interface for the system. It is an IoT app development platform that allows users to interact with the device and view real-time tremor data through customizable widgets. The app communicates with the ESP32 through the cloud, where it receives the tremor data and presents it to the user in an easily interpretable format. The Blynk app provides features such as graphical displays of tremor intensity, real-time alerts when tremor severity crosses a certain threshold, and historical data visualization. These features empower both users and healthcare providers to track tremor progression over time and make informed decisions based on the data.

To facilitate continuous operation, the system relies on a battery pack to power the sensor and microcontroller. Typically, a rechargeable lithium-ion or lithium-polymer battery is used to provide sufficient power for the wearable device. Depending on the power consumption of the individual components, the battery is expected to last for several hours of continuous operation, with the possibility of extending battery life through low-power modes or sleep cycles in the ESP32 when it is not transmitting data. The battery pack is designed to be small and lightweight, ensuring that it does not add significant bulk or discomfort to the wearable device.

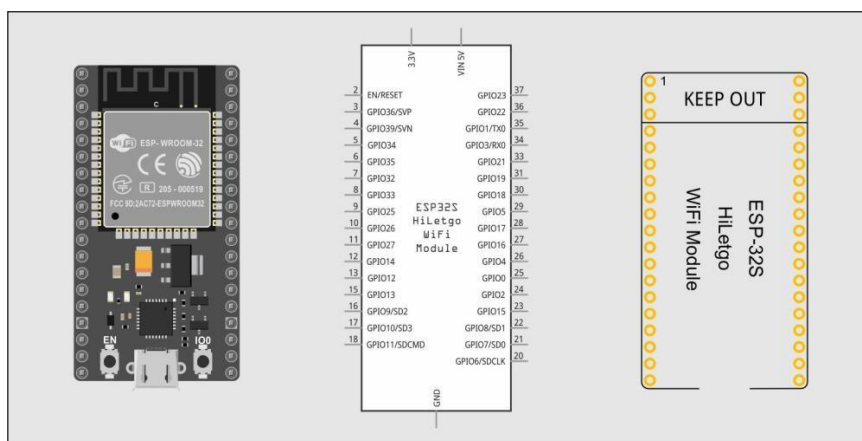
In terms of physical form, the system is designed to be worn comfortably on the user's wrist. The accelerometer and gyroscope are embedded in a small, lightweight enclosure, typically made of durable plastic or other ergonomic materials, which ensures that the device can be worn for long periods without causing discomfort. The device is designed to resemble a wristband or smartwatch, making it discreet and suitable for daily use. The modular design also allows for easy adjustments to sensor placement, enabling the system to monitor tremors in different body parts, such as the hand, wrist, or even the head, depending on the needs of the user.

The data from the wearable device is uploaded to a cloud-based platform, which acts as a central repository for the tremor data. The cloud platform enables real-time data sharing and accessibility, allowing users and healthcare providers to monitor tremor intensity continuously. This cloud-based approach offers several advantages, including remote access to data, the ability to track long-term trends, and the flexibility to update the system's features or functionality through software updates. The cloud platform integrates seamlessly with the Blynk mobile app, providing users with up-to-date tremor information at their fingertips.

In conclusion, the system relies on several components, each chosen for its unique capabilities and role in creating an effective tremor monitoring solution. The combination of the MPU6050 sensor, ESP32 microcontroller, Blynk mobile app, and cloud platform enables the system to provide real-time, accurate, and user-friendly monitoring of essential tremor severity. The system's compact, wearable design, combined with its wireless connectivity and cloud-based features, makes it a highly effective tool for tracking tremor movements and improving the management of essential tremor over time.

**BLOCK DIAGRAM****MICROCONTROLLER:**

The microcontroller serves as the central processing unit of the wearable tremor monitoring system. It is responsible for reading the data from the various sensors (accelerometer, gyroscope, pulse oximeter, etc.), processing that data, and controlling the outputs such as sending alerts or displaying information. In this system, the ESP32 microcontroller is commonly used due to its integrated Wi-Fi and Bluetooth capabilities, which enable real-time data transmission to the cloud and mobile devices (via the Blynk app). The microcontroller acts as the bridge between the sensors and the IoT interface, ensuring smooth communication and monitoring of tremor severity.

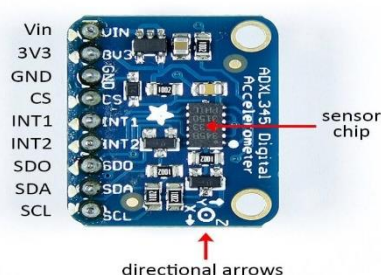
**FORCE SENSOR FOR DETECTING TREMORS:**

A force sensor is used to detect changes in force or pressure exerted during actions like writing, drawing, or grasping objects. In the context of Essential Tremor, these sensors can track the fine motor movements of the hands, capturing the subtle force variations caused by tremors. This sensor provides valuable insight into the user's ability to perform daily tasks, as tremors often impact motor control during tasks requiring precision. The force sensor can measure how much force is applied when interacting with objects, helping to assess the severity of tremor-related motor impairments.



### ACCELEROMETER:

The accelerometer is a key component for detecting tremors. It measures the linear acceleration of the device along three axes (X, Y, and Z), which helps detect up-and-down or side-to-side shaking movements. Accelerometers capture subtle to severe tremor movements, providing accurate data on tremor frequency and amplitude. By using this data, the wearable device can assess the intensity and frequency of the tremors, which can vary throughout the day or in different situations. This data is crucial for quantifying the severity of Essential Tremor (ET) and tracking its progression over time.



### GSM (Global System for Mobile Communications):

The GSM module in the wearable system is used for sending alerts when abnormal sensor readings are detected. For example, if tremor intensity reaches a dangerous level or if vital signs like heart rate or oxygen saturation fall outside of normal ranges, the GSM module can send an SMS alert to caregivers or health professionals. This allows for timely intervention and ensures that the patient receives immediate attention if needed. The GSM module enhances the system's functionality by enabling remote monitoring and communication between the device and the healthcare network.



***BUZZER:***

. The buzzer acts as an audible alert system in the wearable device. When the tremor intensity or other sensor readings deviate from normal ranges (such as an abnormal heart rate or significant tremor spikes), the buzzer emits a sound to notify the user of the issue. This real-time feedback is crucial for the user or caregiver to take immediate action. The buzzer ensures that the patient is alerted to significant changes in their condition, even if they are not actively monitoring the device at that moment.

***HEARTBEAT SENSOR:***

The heartbeat sensor (often based on photoplethysmography, or PPG) measures the heart rate of the user. Since tremors can be associated with changes in heart rate, such as during episodes of anxiety or stress, this sensor provides important data to help monitor the patient's overall health. By tracking the heartbeat alongside tremor data, the system can correlate changes in heart rate with



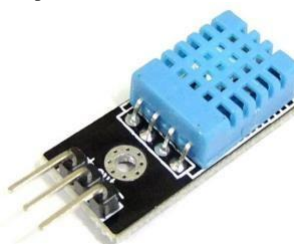
tremor severity, offering a deeper understanding of the patient's condition. The heartbeat sensor can also serve as an early warning indicator if abnormal fluctuations in heart rate are detected, prompting further investigation.

***SPO2 SENSOR (PULSE OXIMETER):***

. The SPO2 sensor monitors the oxygen saturation level of the blood, which can be a vital indicator of overall health. When integrated with the tremor monitoring system, the SPO2 sensor provides additional context, as tremor severity may increase with respiratory distress or changes in oxygen levels. By monitoring both heart rate and oxygen levels, this sensor helps ensure that the patient's body is receiving adequate oxygen, which is important for those who may have respiratory difficulties alongside tremors. It also assists in tracking the patient's recovery or stability during tremor episodes.

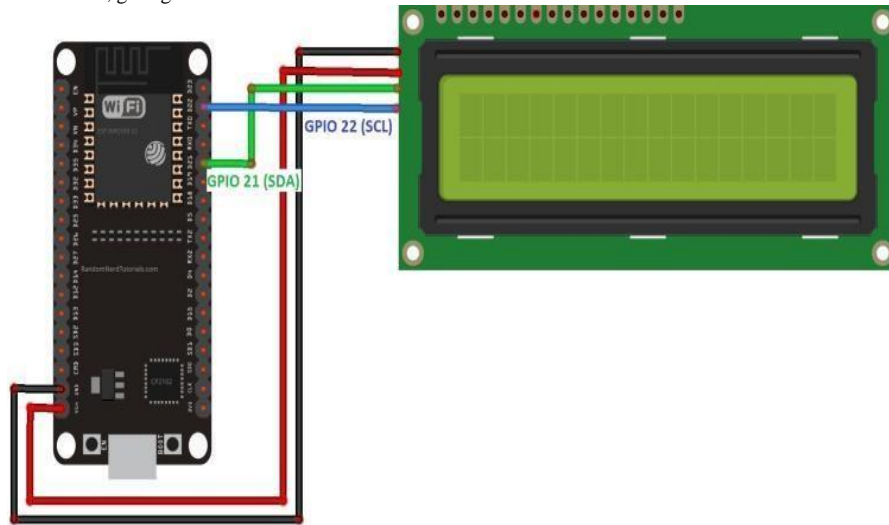
***TEMPERATURE SENSOR:***

The temperature sensor tracks fluctuations in the user's body temperature, which can influence tremor intensity. Changes in body temperature, such as during a fever or stress response, can cause tremors to become more pronounced. The temperature sensor helps provide a more comprehensive analysis of the patient's condition by linking thermal changes with tremor patterns. For instance, a sudden rise in temperature could indicate an underlying issue such as an infection, prompting the system to notify caregivers of a potential health concern.



**LCD DISPLAY:**

The LCD display provides a visual output for the user, displaying key information in real-time. This includes tremor severity, heart rate, oxygen saturation, body temperature, and any other relevant metrics. The LCD serves as an interface for the user to quickly check their health data, ensuring that they can monitor their condition independently without needing to access the mobile app at all times. It also provides real-time alerts in the form of text or icon-based notifications, giving immediate feedback to the user about their health status.

**MAX30100 (Heart Rate and SPO2 Sensor):**

The MAX30100 is an integrated sensor that measures both heart rate and blood oxygen levels (SPO2) using photoplethysmography (PPG). It emits light through the skin to detect the amount of light reflected back by oxygenated and deoxygenated blood, helping calculate heart rate and SPO2 levels. In the wearable system, the MAX30100 offers accurate, realtime monitoring of the user's cardiovascular and respiratory health, making it an important component for understanding the impact of Essential Tremor on the user's overall well-being. By combining heart rate and oxygen saturation monitoring, this sensor adds a level of sophistication to the tremor monitoring system, enhancing its ability to detect and respond to changes in the user's health status.

**V. Internet of things (IOT)**

The internet of Things (IoT) refers to a network of physical devices embedded with sensors, software, and other technologies that can collect and exchange data over the internet. In this project, IoT enables real-time communication between the wearable device and a mobile or cloud platform. The sensors (accelerometer, gyroscope, pulse oximeter, temperature, and pressure sensors) continuously monitor the patient's physiological parameters and transmit the data via a microcontroller (such as ESP32) to the cloud or mobile app.

By using IoT, the system allows remote monitoring of Essential Tremor (ET) symptoms, ensuring that data is accessible from anywhere. It helps caregivers and healthcare providers track patient status, even from a distance, and take timely action in emergencies.

**Features of IoT**

One of the primary features of IoT is real-time data acquisition, which allows the system to continuously collect and transmit data from sensors like accelerometers, gyroscopes, pulse oximeters, temperature, and pressure sensors. This enables the monitoring of tremor severity and physiological changes as they occur, providing immediate insight into the patient's condition.



Remote accessibility is another significant advantage. With IoT, caregivers, doctors, and even family members can access sensor data anytime, anywhere, through internet-connected devices like smartphones or tablets. This is especially important for Essential Tremor (ET) patients who require frequent monitoring but may not always be able to visit healthcare facilities.

IoT systems also offer automated alerting capabilities. If any parameter crosses a critical threshold—such as a sudden spike in tremor intensity or a drop in blood oxygen level—the system automatically activates a buzzer and sends alert notifications to assigned contacts or health professionals. This ensures timely medical intervention and enhances patient safety.

Another useful feature is data logging and cloud storage, which allows for long-term tracking of ET symptoms. Historical data can be reviewed to analyze tremor patterns, frequency, and severity over time, helping doctors to adjust treatment plans effectively.

Inter-device communication is a core part of IoT. Sensors, microcontrollers, and mobile applications work together seamlessly, exchanging data without manual input. This smooth communication improves the reliability and responsiveness of the overall system.

Lastly, energy efficiency and portability make IoT-based systems ideal for wearable health monitoring. Devices are designed to run on low power and be lightweight, enabling continuous usage without causing discomfort to the patient.

### ***Applications of IoT***

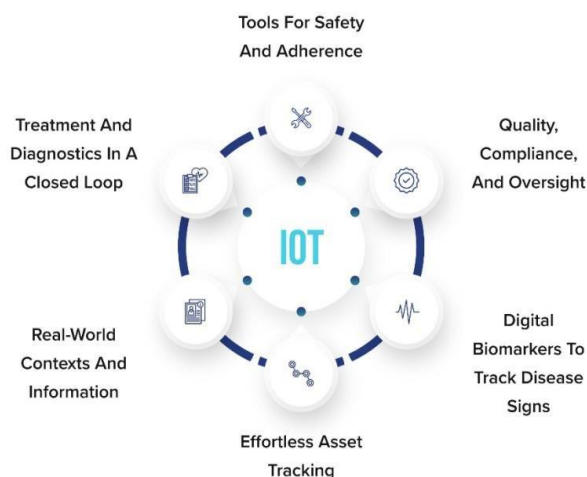
The Internet of Things (IoT) has revolutionized how patient care is delivered and managed. One of the most powerful applications is remote patient monitoring, where wearable devices equipped with sensors continuously track vital health parameters and physiological signals. In the case of Essential Tremor (ET), this technology allows patients to be monitored from the comfort of their home, reducing the frequency of hospital visits while ensuring that their condition is under observation 24/7.

IoT facilitates real-time emergency detection. By analyzing data like tremor intensity, heart rate, oxygen saturation, body temperature, and muscle pressure, the system can identify unusual or dangerous patterns. For instance, a sudden increase in tremor frequency combined with a spike in heart rate might signal a serious episode, prompting the system to instantly alert caregivers or medical personnel via app notifications or SMS messages. This kind of rapid response capability is vital for preventing complications and ensuring patient safety.

Another major application is long-term data storage and trend analysis. With continuous data collection uploaded to the cloud, doctors can observe patterns in a patient's tremor activity over days, weeks, or months. This helps in evaluating how the condition is progressing and whether current treatments are effective. Based on this data, physicians can adjust medication dosages, modify therapy routines, or recommend new interventions that are more suited to the individual patient's needs.

IoT also supports personalized healthcare by adapting to each user's unique tremor patterns and medical history. The system can learn from the collected data and provide custom alerts, health reports, or suggestions tailored to the individual, making the entire healthcare experience more efficient and targeted.

Finally, IoT enhances patient engagement and empowerment. By allowing patients to view their own health data through a mobile app like Blynk, they become more aware of their condition and more involved in its management. This not only improves compliance with treatment but also encourages healthier lifestyle habits, contributing to a better quality of life.



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## VI. Blynk android app

The Blynk Android app plays a central role in transforming sensor data into a visual, interactive, and accessible format for both patients and caregivers. It acts as the user-friendly front-end of the IoT system, allowing real-time monitoring and control of all connected devices and sensors involved in the Essential Tremor (ET) monitoring project.

Using the Blynk Cloud, the app connects seamlessly with the ESP32 microcontroller, which collects data from various onboard sensors including the accelerometer, gyroscope, pulse oximeter (MAX30100), temperature sensor, and force sensor. This data is transmitted via Wi-Fi and instantly reflected on the app interface.

Blynk provides a wide range of customizable widgets such as LCD displays, value displays, graphs, and gauges. These widgets can be configured to show tremor intensity, frequency, heart rate, SpO2 levels, body temperature, and applied pressure. The use of live graphs is particularly useful in visually analyzing tremor patterns over time, helping users and healthcare providers identify fluctuations and possible triggers.

One key advantage of the Blynk app is its support for push notifications and threshold alerts. If the system detects abnormal sensor readings—like dangerously high tremor activity or irregular heart rate—it automatically sends an alert message to a caregiver or medical professional. This ensures faster response and intervention, which can be crucial in preventing health risks.

Another benefit is flexibility and ease of use. Blynk is built as a low-code platform, which means that adding new sensors or adjusting the dashboard does not require deep programming knowledge. This makes it ideal for scalable healthcare projects where sensor configurations may evolve. You can simply drag and drop widgets, assign virtual pins, and instantly visualize new parameters.

Moreover, the Blynk app is cross-platform and cloud-based, meaning data can be accessed anytime from any device with internet connectivity. This not only improves monitoring efficiency but also provides peace of mind for both patients and their families.

In essence, Blynk transforms your wearable ET monitoring system into an intelligent, interactive, and highly accessible healthcare solution, bridging the gap between complex sensor technology and day-to-day usability.



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## VII. Conclusion

Essential Tremor (ET) is a common but often misunderstood neurological disorder that can significantly impair daily functioning for many individuals. Its primary symptom—uncontrolled, rhythmic shaking—can interfere with fundamental tasks such as writing, eating, or holding objects. Accurate, continuous monitoring of tremor severity plays a vital role in diagnosis, treatment evaluation, and daily management. Traditional monitoring methods, such as periodic clinical visits or subjective self-assessment, lack consistency and fail to offer real-time insights. This project presents a meaningful advancement by proposing a wearable, sensor-based solution that combines modern technologies like accelerometers, gyroscopes, and Internet of Things (IoT) integration to bridge this gap.

The wearable system developed in this project uses an accelerometer and gyroscope to capture both linear and angular motion, which together provide a comprehensive analysis of tremor activity. Accelerometers alone can detect motion along straight paths, but they cannot capture rotational movements that are also typical in ET. By adding gyroscopic data, we are able to more accurately measure tremor frequency, amplitude, and pattern. This hybrid sensor approach allows for real-time tracking of tremors ranging from subtle hand shakes to more severe, erratic movements, making the system valuable for diverse ET cases.

Beyond motion tracking, the system incorporates additional biomedical sensors, such as a pulse oximeter (MAX30100), temperature sensor, and pressure sensor. These provide insight into the user's physiological state, including heart rate, SpO2 (blood oxygen saturation), body temperature, and

force applied during movement. These parameters can fluctuate with emotional state, medication effects, or environmental factors—all of which may influence tremor severity. The system's ability to correlate these factors with tremor activity gives a more holistic view of the patient's condition, opening the door for personalized care and more effective treatment planning.

A key feature of this system is its IoT capability, allowing data to be wirelessly transmitted from the ESP32 microcontroller to the Blynk Android application via Wi-Fi. The mobile app serves as a live dashboard, displaying graphical and numerical representations of the sensor data in real-time. This ensures that patients, caregivers, and medical professionals can monitor the condition anytime, anywhere. Additionally, the system is equipped with an alert mechanism—if any sensor reading crosses predefined safe limits, a buzzer is triggered and a warning message is sent to a designated contact. This adds an essential layer of safety, particularly in home-care or unsupervised settings.

The flexibility and scalability of this system are also worth highlighting. The use of modular components and the low-code environment of the Blynk app make it easy to upgrade or expand the device. New sensors can be integrated with minimal changes to the overall architecture. Moreover, the collected data can be stored or uploaded to a cloud database for long-term trend analysis, offering clinicians the ability to monitor disease progression and assess how well a patient is responding to therapy.

From a usability standpoint, the design focuses on being lightweight, wearable, and user-friendly. Patients can wear the device on their wrist or arm, where tremors are most pronounced. The interface of the Blynk app is simple enough for non-technical users, while still offering enough depth for clinical interpretation. This ensures accessibility across a wide demographic, including elderly users or those unfamiliar with digital health tools.

In conclusion, this project presents a robust, real-time, and intelligent wearable monitoring system for estimating the severity of Essential Tremor. By integrating accelerometer and gyroscope data with physiological sensors and IoT connectivity, the system delivers precise tremor analysis, health monitoring, and timely alerts, all through a mobile interface. This makes the system not only practical for everyday use but also a valuable tool in clinical settings. It empowers patients with more control over their condition, supports caregivers with instant data access, and aids doctors with accurate insights, all of which contribute to better quality of life and improved medical outcomes for ET patients.

As future enhancements, this system could integrate AI-based algorithms to automatically classify tremor severity levels or even predict tremor episodes before they occur. Additional cloud-based storage and analytics could further support long-term care strategies and remote diagnostics. With continued development and refinement, such systems have the potential to become standard tools in the management of movement disorders like Essential Tremor.



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## VII. Conclusion:

I would like to express my sincere gratitude to all those who supported and guided me throughout the successful completion of this project, “Wearable Accelerometer and Gyroscope Sensors for Estimating the Severity of Essential Tremor.” This work would not have been possible without the constant encouragement, valuable suggestions, and assistance I received from many individuals.

First, I am deeply thankful to my project guide, Mr.Thennarasu.N, for their continuous support, insightful feedback, and expert guidance throughout the project. Their encouragement and mentorship played a vital role in shaping the technical and analytical aspects of my work.

I would also like to extend my appreciation to the Head of the Department, Mr.Prabu.R, and all faculty members of the [Department Name], for providing the necessary resources, facilities, and an encouraging environment to carry out this project effectively.

I am grateful to my peers and friends for their support, encouragement, and helpful suggestions during various stages of the project development. Their motivation helped me stay focused and confident.

Lastly, I express my heartfelt thanks to my family for their unconditional love, patience, and moral support. Their belief in my abilities gave me the strength to complete this project with dedication and enthusiasm.

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