



# “Geosense”: Smart Iot Solutions For Landslide Risk Mitigation

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**Abstract:** The paper presents a study on "Landslide Detection Using IoT Technology" focusing on the design and implementation of an innovative monitoring system in landslide-prone areas. Landslides pose significant risks to life, infrastructure, and the environment, particularly in mountainous areas with steep slopes and heavy rainfall. Traditional monitoring techniques often fall short in providing timely alerts, highlighting the need for advanced technological solutions. The proposed IoT-based system relies on a network of interconnected sensors to continuously monitor critical parameters such as soil moisture, ground movement, and environmental conditions. By collecting and analyzing real-time data, the system can detect unusual patterns indicative of potential landslides, enabling early warning and evacuation measures. Its low-cost sensors and wireless communication capabilities make it scalable and suitable for remote areas where conventional systems are challenging to deploy. The project involved studying various soil parameters affecting landslide occurrences, analyzing soil stability, and building a prototype for early detection. The findings demonstrate IoT technology's potential in providing continuous monitoring and predictive capabilities, strengthening community resilience against natural disasters. The report advocates further research to enhance IoT-based landslide detection systems, supporting the advancement of disaster preparedness strategies in civil engineering and ensuring safer environments in vulnerable regions.

**Index Terms** - IoT Technology, Monitoring System, Landslide-Prone Areas, Real-Time Data, Soil Moisture, Ground Movement, Early Warning, Low-Cost Sensors, Wireless Communication.

## I. INTRODUCTION

Landslide detection through IoT technology is a landmark improvement in landslide hazard management and mitigation. Landslides are among the most destructive natural disasters, causing great loss of life, destruction of infrastructure, and degradation of the environment, especially in mountainous and hilly areas with steep slopes and heavy rainfall. Early detection and real-time monitoring are important steps toward reducing the impacts of landslides. The best times for evacuation or taking preventative measures can be anticipated during early detection. This aspect is usually beyond the limits of traditional monitoring, where only manual observation or the usage of basic sensors might exist. This is where IoT technology comes in, which proves a powerful solution in landslide monitoring systems. This would create an IoT network of interlinked devices, such as sensors, cameras, and communication modules that work together to collect data in real-time and continuously transmit the same. Strategically placed on slopes and other high-risk landslide-prone areas, these devices could monitor parameters that indicate a landslide has begun. These include soil moisture content, ground movement, vibration, and weather-related phenomena like rainfall and temperature. It is then transmitted to a central server or cloud platform for processing and analysis to identify potential landslide triggers. The landslide detection via IoT provides one of the critical benefits, which are real-time monitoring. Sensors placed at high-risk areas will, therefore continue measuring changes in soil moisture, ground tilt, among other relevant factors. For instance, when the rainfall sensor shows high levels of rainfalls,

this can translate to a danger of a landslide, particularly if information is combined with what the sensors indicate about saturated grounds by soil moisture sensors. Applying algorithms to these data points, the system can detect an abnormal pattern or breach in a threshold that might have signaled an impending landslide. On detection of such levels of risk, it automatically issues an alert to the local authorities and even communities at risk so they may be evacuated before landslide impact. Additionally, the systems using IoT for landslide detection are cheaper and more scalable. The cost-effective sensors and wireless communication technologies make the deployment of IoT networks even possible over wide areas even in remote places. These systems employ sun-powered sensors and extended-distance wireless modules to ensure their proper operation even in remote areas where conventional sources of power or communication equipment do not exist. There are also other advantages with regards to the integration of machine learning or data analytics on an IoT platform because, learning is done for a machine learning model to be able to distinguish patterns of landslides analysing historical as well as real-time data gathered from various sensors within the IoT ecosystem. They will evolve over time into more accurate systems and ensure that their predictions become very reliable while the false alarm level is very low. Secondly, by being able to monitor and control an IoT network from a remote area, it increases flexibility and responsiveness for landslide management systems. Thus, summarily, the landslide detection scenario has improved with IoT, offering it continuous and real-time collection, cost-effective methods, and predictive approaches. This technological integration would then offer a proactive approach to landslide risk management, thus improving the safety of the community while lowering losses from disaster situations. As the technology develops, landslide detection applications are likely to get more sophisticated, landslide-prone areas becoming safer and better prepared to respond to natural disasters.

## II. METHODOLOGY

### 2.1 AREA OF STUDY

Preliminary studies were conducted to identify a suitable location having slope stability issues from the interaction with the residence in the Idukki region, based on the response the study location was selected. The chosen location is Anachal - Vellathooval road near Chenkulam reservoir, Anachal, Idukki, Kerala (Latitude: 10°00'47.6"North, Longitude: 77°02'00.7"East). Many series of Landslides were reported at our chosen location. In 2018, the landslide occurred severely. In 2018, the worstly affected regions were Idukki, Kottayam thus we had choose the particular location.

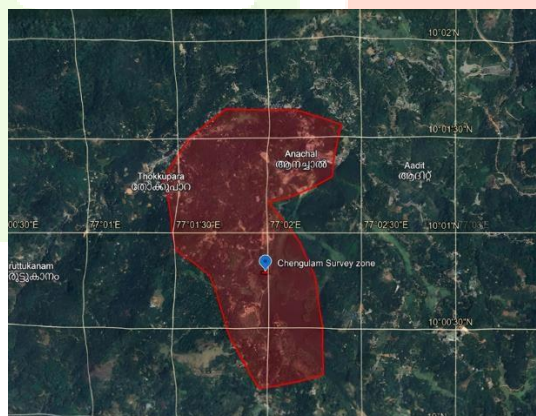


Figure1. Landslide Prone Area Idukki

### 2.2 SOIL PROFILE SURVEY

A comprehensive slope profile survey was conducted near the Chengulam Reservoir, located at latitude 10.013221 and longitude 77.033518, using a Total Station, a highly accurate surveying instrument. The survey aimed to capture precise coordinates at various elevations across the site. These detailed data points were essential in accurately defining the topography, contour, and gradient of the terrain, which are crucial for understanding the area's geomorphology. By mapping the exact shape and slope of the land, the survey provided foundational information for subsequent geotechnical analysis. This data plays a pivotal role in assessing slope stability, enabling engineers to evaluate potential risks of landslides or other forms of slope failure. The results from this survey are vital for planning and designing effective stabilization measures, ensuring the safety and resilience of the structures and communities in the vicinity of the reservoir.

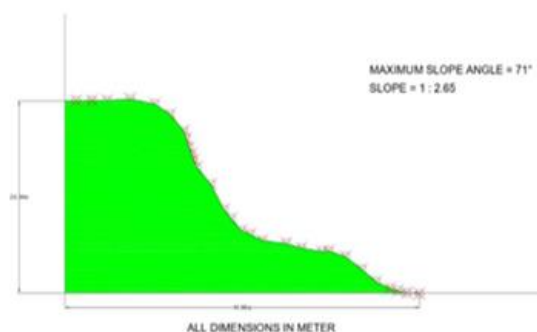


Figure 2. Soil Profile

### 2.3 SAMPLE COLLECTION

Undisturbed soil samples were collected from the site near Chengulam Reservoir for laboratory testing, ensuring the preservation of the soil's natural structure and moisture content. These samples will undergo tests to assess key properties such as shear strength and moisture content, which are essential for evaluating slope stability and informing mitigation measures. The results will enhance the understanding of the site's geotechnical profile, supporting effective infrastructure and slope management decisions.

### 2.4 RESULTS OF LABORATORY TESTS

The laboratory tests conducted on the collected soil samples provided key geotechnical parameters essential for further analysis. The Uniformity Coefficient ( $C_u$ ) of the soil was determined to be 7.33, indicating a well-graded soil composition. The Coefficient of Curvature ( $C_c$ ) was found to be 0.169, which suggests a poorly graded soil. The moisture content of the soil sample was measured as 18.5%, reflecting its water holding capacity. The soil density was recorded at  $1.69 \text{ g/cm}^3$ , which influences the compaction characteristics and bearing capacity. The cohesion ( $C$ ) of the soil was found to be 0.145 KPa, signifying a relatively low binding strength among particles. Additionally, the angle of internal friction was observed to be  $28^\circ$ , indicating moderate shear strength. These parameters are crucial for assessing the soil's stability and suitability for construction or slope stability analysis.

SL NO	PARAMETERS	VALUE
1.	Uniformity Coefficient, $C_u$	7.33
2.	Coefficient of Curvature, $C_c$	0.169
3.	Moisture Content	18.5%
4.	Soil density	$1.69 \text{ g/cm}^3$
5.	Cohesion, $C$	0.145 KPa
6.	Angle of Internal Friction	$28^\circ$

Table 1. Laboratory Test Results

### 2.5. SOFTWARE ANALYSIS ON SOIL

Slope stability analysis involves complex calculations that account for various factors such as soil properties, loading conditions, and environmental influences. Software tools can handle these calculations more accurately than manual methods, reducing the risk of errors and increasing reliability. Modern software allows for detailed modeling of soil behavior and slope conditions, incorporating advanced techniques like finite element analysis (FEA) and limit equilibrium methods that provide a more precise assessment of stability. Software can integrate and process data from various sources, such as soil tests, site surveys, and environmental conditions, facilitating comprehensive analysis. Tools offer advanced visualization capabilities, including 2D and 3D graphical representations of slopes and stress distributions, which enhance the understanding of complex stability issues.

### 2.5.1.GEOSTUDIO

Slope stability analysis using GeoStudio, specifically with the SLOPE/W module, is a critical process in geotechnical engineering to assess the safety of natural and man-made slopes. The software allows users to model different slope geometries, soil properties, and loading conditions to determine the factor of safety against slope failure. SLOPE/W uses various methods like the Limit Equilibrium Method (LEM) to evaluate the potential slip surfaces and calculate the factor of safety. To perform the analysis, users define the slope geometry, input soil properties such as unit weight, cohesion, and angle of internal friction, and apply boundary conditions like pore-water pressure or external loads. SLOPE/W then calculates the factor of safety for the slope under various conditions, including static and seismic loads.

The **Limit State Method** in GeoStudio is a widely used approach in geotechnical engineering for evaluating slope stability. This method assesses the equilibrium between the resisting forces, which prevent slope failure, and the driving forces, which contribute to potential failure. The result is the **factor of safety** (Fs), defined as the ratio of resisting to driving forces. A factor of safety greater than one indicates that the slope is stable, while a factor less than one suggests that failure is likely, requiring intervention to improve stability. GeoStudio's SLOPE/W module is particularly effective in applying the Limit State Method to a variety of slope stability problems.

Among the several techniques under the Limit State Method, **Bishop's Simplified Method** is used here due to its balance of computational efficiency and accuracy. This method is specifically tailored for circular slip surfaces, which are common in natural and engineered slopes. The simplification in Bishop's method comes from the assumption that inter-slice shear forces are negligible, which simplifies the calculations without significantly compromising the accuracy for many practical scenarios.

### 2.5.2 SLOPE STABILITY ANALYSIS RESULTS

The slope stability analysis conducted using GeoStudio identified a critical factor of safety (FOS) of 1.018 from a total of 124 slip surfaces, of which 48 were considered valid based on the software's analysis criteria. The critical slip surface, corresponding to the lowest FOS, indicates a marginally stable slope, as an FOS close to 1 suggests that the slope is on the verge of failure under current conditions. The presence of multiple valid slip surfaces further highlights the potential for instability, requiring careful evaluation of soil properties, pore water pressure effects, and external loading conditions. To enhance stability, remedial measures such as drainage improvements, slope reinforcement, or modification of slope geometry may be necessary to increase the factor of safety above the generally acceptable threshold of 1.3–1.5 for long-term stability.

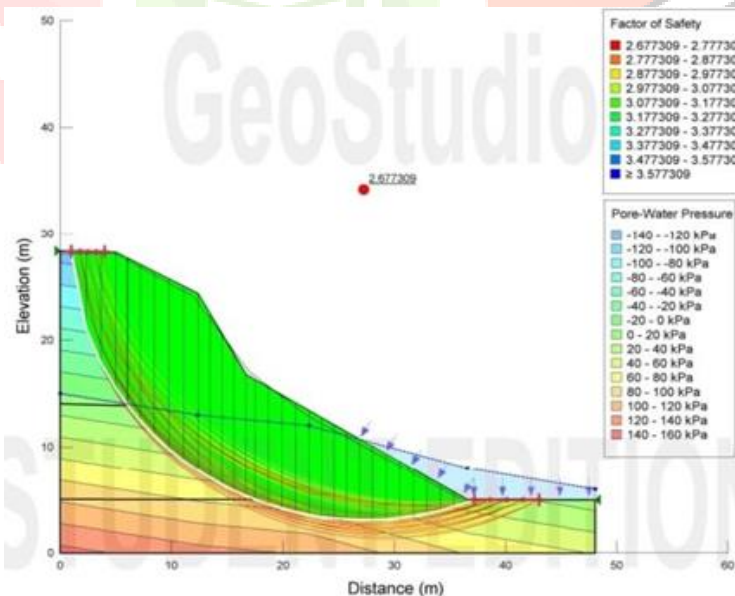


Figure 3. Example of GeoStudio Analysis

The critical factor of safety (FOS) of 1.018 obtained from the slope stability analysis indicates a slope that is highly susceptible to failure, as an FOS close to 1.0 suggests that the resisting forces barely exceed the driving forces. This marginal stability requires immediate intervention to prevent potential landslides.

## 2.6 DESIGN OF THE LANDSLIDE DETECTION PROTOTYPE USING IOT

Designing an IoT-based landslide detection prototype involves creating a system that can monitor ground conditions and provide early warnings of potential landslides.

### 2.6.1 THE MAIN COMPONENTS TYPICALLY INCLUDE:

#### 1.SENSORS:

- ❖ Soil Moisture Sensor: Measures the soil's water content, as high moisture can weaken soil stability.
- ❖ Vibration Sensor: Detects ground tremors or minor seismic activity that might indicate soil movement or destabilization.
- ❖ Gyroscope (e.g., MPU6050 or MPU9250): Monitors tilting or shifting of the soil, which is often a precursor to landslides.
- ❖ Temperature and Humidity Sensor (e.g., DHT22): Helps track environmental conditions, as temperature and humidity can influence soil moisture and stability.

2.MICROCONTROLLER (e.g., ESP32): Serves as the control unit, handling the sensor inputs, processing data, and transmitting it to the cloud. The ESP32 is ideal because it has built-in Wi-Fi, enabling IoT connectivity for remote monitoring.

3.COMMUNICATION MODULE (if using Arduino): If the ESP32 isn't used, an ESP8266 Wi-Fi module or GSM module (like SIM800) can enable cloud data transmission from an Arduino.

4.POWER SUPPLY: A rechargeable battery or solar setup powers the system, especially for outdoor applications.

5.CLOUD PLATFORM: Stores, analyzes, and sends alerts via email, SMS, or a mobile app when landslide risks are detected based on sensor readings.

6.LOCAL ALARM (OPTIONAL): A buzzer or LED that triggers when any sensor value crosses a critical threshold.

### 2.6.2 CIRCUIT DESIGN AND CONNECTIONS:

#### Components and Connections

#### ESP32 Microcontroller :

- Gyroscope (MPU6050): Connect the SCL pin to D22, SDA pin to D21 on the ESP32.
- Soil Moisture Sensor: Connect the analog output to an analog pin (e.g., D34).
- Vibration Sensor: Connect the digital output pin of the vibration sensor to a digital pin (e.g., D32).
- Temperature and Humidity Sensor (DHT22): Connect the data pin to a digital pin (e.g., D33) with a pull-up resistor if required.
- Power Supply Connect the ESP32's Vin or 3.3V pin to a battery or solar-powered setup.
- Wi-Fi or GSM Module (if using Arduino) If using Arduino, connect an ESP8266 Wi-Fi module or GSM module for cloud connectivity.
- Alarm System (Optional) Connect a buzzer or LED to a digital pin on the ESP32 to provide local alerts.

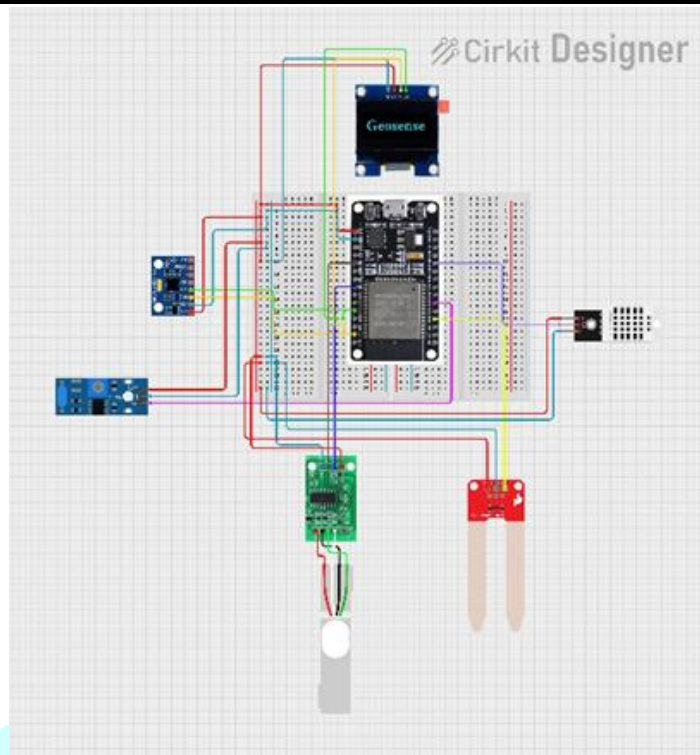


Figure 4. Circuit Design

### 2.6.3 OPERATIONS

**Data Collection:** The soil moisture sensor, vibration sensor, gyroscope, and temperature/humidity sensor continuously gather data on soil and environmental conditions.

**Data Processing:** The ESP32 reads and processes sensor values, comparing them against preset thresholds for landslide risk indicators, such as high soil moisture, significant tilt, or ground vibrations.

**Cloud Communication:** If any threshold is crossed, the ESP32 sends real-time data to a cloud platform over Wi-Fi. The cloud platform can then analyze trends and trigger alerts if landslide conditions seem likely.

**Local Alarm (Optional):** A buzzer or LED provides immediate on-site alerts if a high-risk condition is detected, notifying those nearby.

**Programming code for the device:** The ESP32 microcontroller in this project was programmed using Arduino IDE, a widely used open-source platform that simplifies coding and uploading firmware to the device. The Arduino C/C++ programming language was used to interface with various sensors, including the moisture sensor, MPU6050 (gyroscope), rain sensor, and load cell (soil pressure sensor), enabling real-time data acquisition. The ESP32 Wi-Fi module was configured to transmit sensor readings to cloud platforms like ThingSpeak, ensuring remote monitoring through the "Geosense" mobile application. The code includes libraries for sensor integration, data filtering, and communication protocols such as MQTT and HTTP, allowing seamless connectivity and real-time updates. Using Arduino IDE made development efficient by providing an easy-to-use environment with built-in debugging tools, serial monitoring, and extensive library support, ensuring reliable performance of the IoT-based landslide detection system.

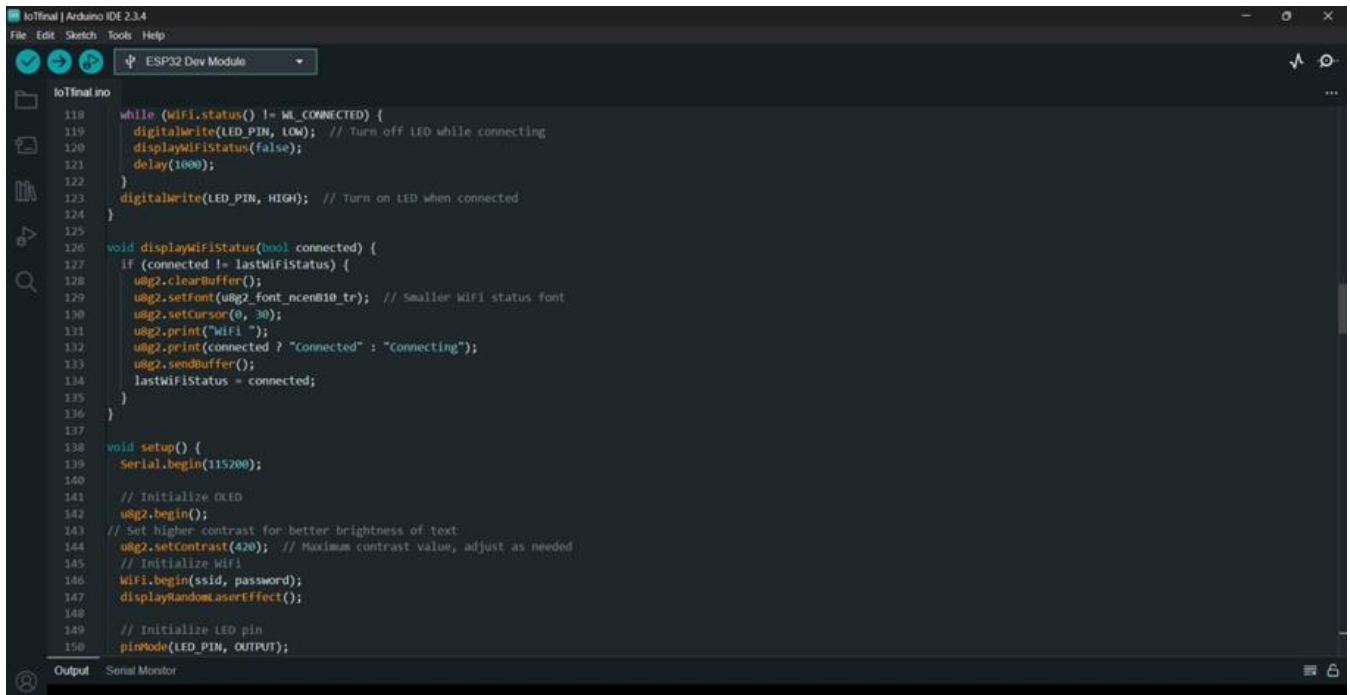


Figure 5. Arduino IDE

This IoT landslide detection prototype continuously monitors ground moisture, vibrations, tilt, and environmental factors like temperature and humidity, providing early warnings when landslide conditions are likely. Using the ESP32 with built-in Wi-Fi enables remote data monitoring, making it an effective tool for areas at risk of landslides.

## 2.7 MODEL IMPLEMENTATION AND TESTING

The model implementation for the IoT-based landslide detection prototype can be broken down into three primary stages: hardware setup, software development, and cloud integration. Each stage involves specific tasks to ensure the system is operational and capable of real-time monitoring.

### HARDWARE SETUP:

- **Component Assembly:** The ESP32 microcontroller, sensors (soil moisture, vibration, gyroscope, temperature, and humidity), and optional buzzer are assembled as per the circuit design. Proper connections are ensured between the sensors and the ESP32 to enable smooth data acquisition.
- **Power Configuration:** A battery or solar panel setup powers the ESP32, ensuring the device remains functional in remote areas.
- **Sensor Calibration:** Each sensor is calibrated to measure environmental and soil conditions accurately. For example:
  - ❖ The soil moisture sensor is calibrated to distinguish between dry, damp, and saturated soil.
  - ❖ The gyroscope is calibrated to detect tilt accurately, and its sensitivity is adjusted to detect slight shifts in soil movement.
  - ❖ The vibration sensor's threshold is adjusted to detect relevant seismic activities or ground shifts.
  - ❖ The DHT22 sensor is calibrated to report temperature and humidity with accuracy.

### 2.7.1 FIELD TESTING

The field testing and calibration of sensors play a crucial role in ensuring the accuracy and reliability of the IoT-based landslide monitoring prototype. Once the **ESP32 microcontroller** and all integrated sensors were assembled, the system was deployed in a landslide-prone region for real-world testing. The **soil moisture sensor** was carefully calibrated by testing it in soil samples with varying moisture levels, ensuring that the device could accurately distinguish between dry, damp, and saturated soil conditions. This calibration was essential for monitoring water infiltration, a key factor influencing slope stability. Similarly, the **gyroscope sensor** was fine-tuned to detect even the slightest tilts and angular displacements in the ground. By adjusting its sensitivity, the

sensor was optimized to identify minor shifts in soil movement that could indicate potential slope failure. The **vibration sensor** underwent extensive threshold adjustments to differentiate between normal ground vibrations and significant tremors associated with landslide-triggering events, thereby reducing the occurrence of false alarms. In parallel, the **DHT22 sensor**, responsible for measuring temperature and humidity, was cross-verified with standard meteorological instruments to ensure precise readings, as fluctuations in temperature and humidity can impact soil conditions. Additionally, a **load cell sensor** was integrated into the system to act as a **soil pressure sensor**, detecting changes in pressure distribution within the slope. It was calibrated by applying controlled loads, ensuring accurate measurement of variations in soil pressure caused by ground shifts, water saturation, or external loading conditions. The entire system was subjected to **prolonged field testing**, where sensor readings were continuously monitored and compared with expected values under different environmental conditions. This rigorous testing process ensured that the sensors functioned optimally, providing reliable real-time data transmission to the cloud for further analysis. The final adjustments involved fine-tuning the sensor thresholds and response times to enhance the **overall efficiency and accuracy of the landslide detection model**, ensuring its effectiveness in providing early warnings in high-risk areas.



Figure 6. Field Testing



Figure 7. Model Designed

## SOFTWARE DEVELOPMENT:

**Sensor Data Acquisition:** Code is written in the Arduino IDE to enable the ESP32 to read data from each sensor. Data readings are taken at regular intervals, processed, and stored temporarily.

**Data Processing and Threshold Analysis:** Using the ESP32, each sensor reading is compared against predefined threshold values:

- If soil moisture readings reach a level indicating potential soil saturation, an alert is triggered.
- Tilt readings from the gyroscope exceeding a safe range imply soil movement, while vibration sensors detecting unusual ground activity signify a warning.
- High temperature and humidity readings correlate with rain and soil moisture changes.

**Communication Module:** The ESP32's built-in Wi-Fi allows it to connect to a cloud server or IoT platform. The microcontroller is programmed to send data packets or alerts when critical thresholds are reached.

**Alert System:** A local alert is programmed to activate a buzzer or LED indicator on-site if thresholds are breached, providing immediate warning to nearby personnel.

## CLOUD INTEGRATION:

**Cloud Platform Setup:** A cloud service (e.g., ThingSpeak, AWS IoT, or Blynk) is configured to store and visualize data from the ESP32. It enables remote monitoring, so authorized users can view real-time data and historical trends from the sensors.

**Data Transmission and Logging:** The ESP32 is programmed to transmit data in regular intervals to the cloud. Alerts are also sent immediately if thresholds are breached, creating records of critical incidents.

**Alert Notifications:** The cloud platform is configured to trigger notifications (e.g., SMS, email, or app alerts) when a high-risk condition is detected. This real-time alert mechanism provides an early warning system for landslide risks.



Figure 8. ThingSpeak Public View

## DEVELOPING MOBILE APPLICATION FOR MONITORING

A mobile application named "**Geosense**" was designed and developed using **Android Studio** to enable real-time monitoring of the IoT-based landslide detection device. The app serves as a user-friendly interface for remotely accessing sensor data, including soil moisture, vibration, gyroscopic movement, rainfall, and soil pressure. It connects with the device via **Wi-Fi (ESP32)** or cloud platforms like **ThingSpeak**, allowing continuous data transmission and visualization. The application features real-time alerts and graphical data representation, enabling quick assessment of slope stability and potential landslide risks. By integrating **push notifications** and **threshold-based warnings**, "**Geosense**" enhances disaster preparedness, providing users with timely updates and proactive monitoring of vulnerable areas.



Figure 9. Geosense Mobile Application

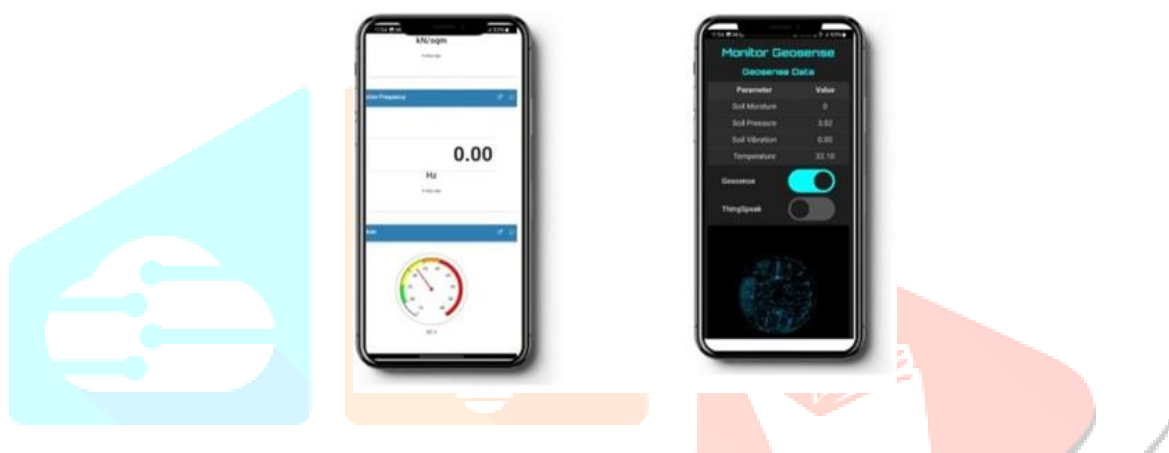


Figure 10. Application Interface

**Accessibility:** To enhance accessibility and ease of use, a **website** was developed to facilitate the installation of the **Geosense** mobile application and provide a **web-based monitoring platform**. The website, hosted at [syamgvnd.wixsite.com/geosense](https://syamgvnd.wixsite.com/geosense), serves as a central hub where users can download the mobile app, access project details, and monitor real-time sensor data from the IoT-based landslide detection system. The platform integrates cloud-based data visualization, displaying live updates on **soil moisture, vibration, gyroscopic movement, rainfall, and soil pressure**. Users can log in to track environmental conditions, receive alerts, and analyze historical trends for better disaster preparedness. By offering both **mobile and web-based monitoring**, the system ensures comprehensive accessibility, enabling proactive risk assessment and early warnings in landslide-prone areas.



Figure 11. Website Interface

## 2.7.2 TESTING AND EVALUTION

To ensure the system's effectiveness, the following testing steps are conducted:

### 1. FUNCTIONAL TESTING

- **Individual Sensor Testing:** Each sensor is tested individually to ensure accurate readings. For example, the soil moisture sensor is tested with different soil moisture levels to confirm it can reliably distinguish between safe and risk- indicating moisture content.
- **Threshold Testing:** Simulated conditions (e.g., tilting the gyroscope or shaking the vibration sensor) are used to test the response of each sensor to ensure threshold settings trigger alerts correctly.

### 2. SYSTEM INTEGRATION TESTING

- **Data Acquisition and Transmission:** After assembling the complete circuit, the system is tested to confirm that data from all sensors are collected accurately and sent to the cloud without transmission loss.
- **Real-Time Alert Testing:** Testing includes triggering threshold conditions (e.g., high soil moisture or vibrations) to confirm that alerts are sent to the cloud and to the local buzzer or LED.

### 3. FIELD TESTING

- **Outdoor Environment Testing:** The prototype is tested in an outdoor environment with varying soil conditions, such as damp and dry soil. The temperature and humidity sensor are exposed to changing weather conditions to test real-world data collection.
- **Simulated Landslide Conditions:** Conditions that could lead to a landslide (such as saturating soil or creating small tremors) are simulated to observe if the system accurately detects potential landslide indicators.

### 4. RELIABILITY AND BATTERY LIFE TESTING

- **Power Consumption Testing:** The power consumption is measured over time to assess battery life, ensuring that the prototype can sustain prolonged operation in remote areas.
- **Connectivity Testing:** Testing the stability and range of the Wi-Fi or GSM module to ensure that data can be sent reliably even with intermittent connectivity.

## 2.7.3 DATA ANALYSIS AND EVALUATION

After testing, data is analysed to evaluate the system's accuracy and response to potential landslide conditions. By examining historical sensor data in the cloud platform, patterns or thresholds indicating landslide conditions can be refined. The following metrics are assessed:

- **Detection Accuracy:** Comparison of sensor data to actual soil conditions to ensure correct threshold settings.
- **Response Time:** Measurement of the time taken for alerts to reach the cloud and notify users.

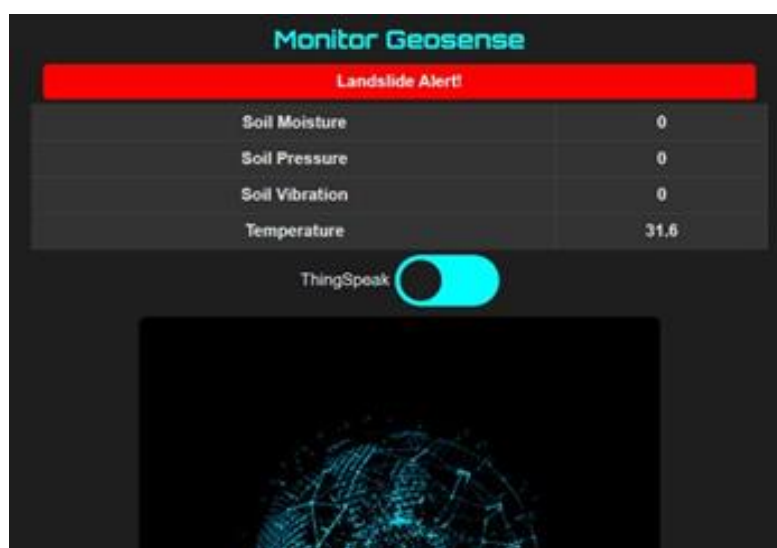


Figure 12. Alert Notification

- **Battery Performance:** Evaluation of battery life under continuous operation to determine whether power efficiency improvements are needed.

Based on testing results, the IoT landslide detection prototype's performance is reviewed to identify any necessary adjustments in sensor calibration, threshold settings, or communication reliability. Recommendations may include optimizing power usage, enhancing alert mechanisms, or improving sensor accuracy for a more effective landslide early warning system.

## 2.8 DATA ANALYSIS AND INTERPRETATION

Data analysis and interpretation are essential to evaluate the performance and reliability of the IoT-based landslide detection prototype. This stage involves examining the data collected by the sensors, identifying patterns, and making informed conclusions about the prototype's accuracy in detecting potential landslide conditions.

### 1. DATA COLLECTION AND PREPROCESSING

- **Sensor Data Logging:** Sensor readings are collected at regular intervals and sent to a cloud platform, where data is stored in a structured format, often as time-stamped entries. Each sensor's data soil moisture, vibration, tilt, temperature, and humidity are tracked over time.
- **Data Cleaning:** Any anomalies, such as missing or outlier values from sensor readings (due to connectivity issues or power interruptions), are addressed. Outliers are evaluated to confirm if they represent actual threshold breaches or erroneous data.

### 2. DATA ANALYSIS TECHNIQUES

- **Trend Analysis:** Data from each sensor is analyzed for trends that indicate landslide risks. For instance, increasing soil moisture and temperature changes combined with tilt variations can suggest worsening soil conditions.
- **Threshold-Based Analysis:** The analysis involves checking if sensor readings exceed critical thresholds. Each time a threshold (e.g., high soil moisture or significant tilt) is breached, it's logged and examined to understand the conditions that may have led to a potential landslide trigger.
- **Correlation Analysis:** Relationships between different sensors (e.g., the correlation between soil moisture and temperature) are analyzed to identify environmental conditions that heighten landslide risk. Strong correlations can help refine thresholds and improve predictive accuracy.
- **Event Frequency Analysis:** The frequency of threshold breaches is analyzed to assess the prototype's sensitivity. Frequent alerts might indicate a need to adjust thresholds, while infrequent alerts can suggest appropriate sensitivity.

### 3. INTERPRETATION OF RESULTS

- **Identifying High-Risk Patterns:** By interpreting patterns such as sudden spikes in moisture combined with tilt and vibration, potential landslide conditions are confirmed, and thresholds are refined for future alerts.
- **Evaluating System Accuracy:** Comparing alert instances with actual environmental conditions helps determine the accuracy of the system. If alerts accurately coincide with risky conditions, it validates the model; discrepancies suggest areas for improvement.
- **Predictive Insights:** Over time, data analysis may reveal recurring conditions that precede landslide events, enabling more proactive alerts. For instance, continuous high moisture with increased soil tilt could reliably indicate imminent instability.

### 4. INSIGHTS AND RECOMMENDATIONS

- **Optimizing Thresholds:** Based on data interpretation, thresholds can be fine-tuned to improve detection accuracy, reducing false positives and enhancing reliability.
- **Improving Model Performance:** By analyzing patterns and correlations, the system can be modified to better respond to complex environmental conditions, creating a more robust landslide detection solution.

Overall, data analysis and interpretation ensure that the landslide detection prototype provides timely and accurate warnings, improving its practical value as an early warning system.

### III. CONCLUSION

Through this project, the factors influencing soil instability at a specific location were identified and thoroughly analyzed. Soil samples were collected from the identified site, and these samples were subjected to a series of laboratory tests to determine their index properties. Based on the results from these tests, the soil was classified as sandy silt, which is a type of soil that can be prone to instability under certain conditions.

Key factors such as the height of the slope, slope inclination, and the water table level were selected for analysis. These parameters play a critical role in determining the stability of the slope. Using the GeoStudio software, a detailed analysis was conducted to evaluate the slope's behavior under various combinations of these parameters. The software generated the factor of safety (FOS) for different critical slope configurations, helping to identify which parameter combinations would lead to the most unstable conditions.

To enhance real-time monitoring and improve landslide detection capabilities, an IoT-based landslide detection prototype was developed and implemented. The system utilizes various sensors, including soil moisture, vibration, gyroscope, and temperature sensors, connected to an ESP32 microcontroller for data acquisition and processing. Real-time sensor data is transmitted to a cloud platform, where it is stored, analyzed, and visualized. This data-driven approach allows for early detection of potential landslides, enabling timely alerts and improved disaster preparedness.

The field testing of the IoT prototype validated the system's effectiveness in detecting abnormal soil conditions and triggering alerts based on preset thresholds. Additionally, a mobile application named "Geosense" and a web-based platform were developed to enhance accessibility, allowing users to monitor real-time sensor data and receive early warning notifications remotely.

Overall, this project demonstrates the effectiveness of integrating geotechnical analysis with IoT technology for landslide risk mitigation. By providing continuous, real-time monitoring and early warning capabilities, the system enhances the safety and resilience of communities living in landslide-prone areas. Further improvements, such as refining sensor calibration, optimizing power usage, and incorporating predictive algorithms, could enhance the prototype's accuracy and scalability, making it an essential tool in disaster risk reduction efforts.

### IV. ACKNOWLEDGEMENT

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