



DEDAN KIMATHI UNIVERSITY OF TECHNOLOGY

DEPARTMENT OF ELECTRICAL AND ELECTRONICS ENGINEERING

FINAL YEAR PROJECT REPORT

PROJECT TITLE:

LIVESTOCK HEALTH AND ACTIVITY MONITORING USING IoT- BASED SENSOR SYSTEM

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19TH DECEMBER, 2024

DECLARATION

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ACKNOWLEDGEMENT

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List of Abbreviations

AWS - Amazon Web Services

DDA - Digital Differential Analyzer

ESP-NOW - Enhanced Serial Protocol - Networking Over Wi-Fi

ESP32 - A microcontroller with integrated Wi-Fi and Bluetooth

IMU - Inertial Measurement Unit

IoT - Internet of Things

IRT - Infrared Thermography

ML - Machine Learning

MQTT - Message Queuing Telemetry Transport

MPU6050 - Microprocessor Unit (Accelerometer and Gyroscope Sensor)

PCB - Printed Circuit Board

RFID - Radio Frequency Identification

SVM - Support Vector Machine

Wi-Fi - Wireless Fidelity

ABSTRACT

The health and productivity of livestock, are essential to the agricultural industry, impacting food security, economic stability, and animal welfare. Traditional livestock health monitoring methods are often labour intensive, time- consuming, and prone to human error, necessitating the adoption of modern technological solutions. This project developed an advanced livestock activity and health monitoring system using sensors and Internet of Things (IoT) technologies. The developed system comprised of a cattle-wearable IoT device comprising of MPU6050 and ESP 32. The MPU6050 was used to measure the activity magnitude, and it comprised of a built-in temperature sensor. Data collected by the sensors was transmitted to another ESP 32 gateway via ESP-NOW protocol, where it was processed using Random Forests Machine Learning algorithms on the gateway to display real-time animal health status. The trained machine learning model analysed the data to identify patterns and deviations indicative of potential health issues such as lameness, fever, or abnormal activity. Results from the model were transmitted to the cloud via Wi-Fi for storage. Also, a visualization dashboard was created as an interface to where the farmers could view real-time cattle metrics. The dashboard was able to fetch this data from the AWS Cloud platform. Overall, this project enhanced cattle health and productivity by continuous and accurate temperature and activity monitoring.

CHAPTER 1: INTRODUCTION

1.0 INTRODUCTION

Livestock farming is a key economic activity in Kenya, providing a livelihood for up to 90% of the population and contributing approximately 12% to the national GDP and 42% to the agricultural GDP. In arid and semi-arid regions, livestock accounts for nearly 95% of family income, making it a critical component of food and nutrition security. Precision livestock farming, which involves applying advanced technologies to cattle management, is increasingly popular in Kenya. Improving this sector offers immense potential to boost food security, economic stability, and overall agricultural productivity [1], [2].

Kenya's livestock sector, however, faces persistent health challenges. Diseases such as foot-and-mouth disease, East Coast fever, and malignant catarrhal fever significantly reduce productivity, leading to economic losses and affecting food supply chains. These diseases often exhibit seasonal patterns, further complicating their management. Traditional health monitoring methods, including visual observations and irregular veterinary visits, are labor-intensive, time-consuming, and prone to human error. These methods often fail to detect early signs of diseases, resulting in delayed interventions, increased mortality, and higher costs for farmers [3], [4].

Advanced livestock health monitoring systems are necessary to address these challenges. Real-time monitoring technologies can track key health metrics such as body temperature and activity levels, providing early warnings of potential health issues. By allowing timely interventions, these systems reduce the severity of diseases and improve livestock welfare and productivity. This project aims to use these technologies to provide a scalable and efficient solution tailored to the needs of Kenyan farmers, ultimately improving economic outcomes and supporting sustainable agriculture [5].

1.1 PROBLEM STATEMENT

Traditional methods for monitoring the health and productivity of livestock, particularly cattle, often require intensive labor, are time-consuming, and susceptible to human error. This inefficiency poses significant challenges to the agricultural industry, affecting food security, economic stability, and animal welfare. Farmers struggle to detect early signs of health issues such as lameness, fever, or abnormal activity, leading to delayed interventions and worsening health problems. The lack of real-time, accurate monitoring systems hampers timely decision-making and effective health management, reducing productivity and increasing costs. There is,

therefore, a critical need for an advanced solution that can provide continuous and precise monitoring of cattle health to enhance the efficiency and accuracy of livestock management, ensuring better animal welfare, economic savings, and higher-quality livestock products.

1.2 OBJECTIVES

1.2.0 Main objective

To develop a system that monitors the health and activity of livestock and provides real time health status of the Cattle.

1.2.1 Specific objectives

To design attachable IoT device capable of collecting real-time data.

To ensure reliable data transmission and processing.

To build and deploy machine learning model on the gateway.

To develop an interactive dashboard for real-time visualization.

1.3 JUSTIFICATION

Effective health and activity monitoring of cattle requires developing an advanced system to address the above-mentioned prevalent issues, particularly the labor-intensive and error-prone nature of traditional livestock health monitoring. By utilizing sensor technology prediction models, the system offers a modern solution to enhance the efficiency and accuracy of cattle health management. Real-time monitoring and anomaly detection techniques enable immediate interventions, reducing the risk of severe health issues and improving animal welfare. This system promises economic savings through better health management and increased productivity and enhances the quality of livestock products. Real-time monitoring also includes timely alerts to farmers to enable swift decision-making and response to affected cattle, thus improving treatment and recovery times. Overall, this system leads to better livestock well-being while increasing farmers' and the economy's yields.

1.4 SCOPE OF THE STUDY

This project focuses on developing a livestock health and activity monitoring system using IoT-based wearable devices and machine learning for cattle reared through paddocking. The system collects real-time data on cattle activity and body temperature using MPU6050 sensors integrated with ESP32 microcontrollers, transmitting the data wirelessly via ESP-NOW and

Wi-Fi. Machine learning models are deployed on the gateway to analyse the data and identify health anomalies, providing timely alerts to farmers when cattle require attention, without diagnosing specific diseases. A dashboard fetches and displays this data from the cloud, offering a user interface for farmers to monitor livestock health remotely. The prototype prioritizes functionality, anomaly detection, and health status prediction, using dummy data for training and testing. While energy efficiency and disease-specific diagnostics were not a focus at this stage, the system demonstrates significant potential for scalable, reliable, and cost-effective health monitoring solutions in diverse farming environments.

CHAPTER 2: LITERATURE REVIEW

2.0 Background

Efficient livestock management is critical to the agricultural sector, impacting productivity, animal health, and overall farm sustainability. Traditional livestock monitoring methods often rely on manual observations and periodic veterinary checks, which can be labour-intensive and may fail to provide timely insights into the health and activity of animals. The advent of sensor technology and data analytics offers an innovative approach to livestock monitoring, enabling continuous data collection and real-time analysis. This project uses machine learning algorithms to identify unitary behaviours and movements of cattle recorded by IMU sensors [6]. It will also investigate the time window on the performance of unitary behaviour classification and discuss the necessity of movement analysis. These advancements are capacitated by the developing a cattle-wearable device that monitors health and activity, a data transmission technique to send the data to the gateway, utilizing machine-learning for anomaly detection, and a way to visualize the health stats of the cattle at real-time.

Temperature and Activity monitoring can help the farmers predict significant events such as such as calving, estruses, lameness, and disease [7]. Examples of diseases characterized by reduced movement and activity include bovine tuberculosis, foot and mouth, east coast fever, and Nagana. Diseases characterized by a sudden increase in movement and activity include malignant catarrhal fever. Diseases characterized by fever include; brucellosis, East Coast Fever, Anaplasmosis, Babesiosis. Foot and Mouth Diseases, and Trypanosomiasis [8,9].

The worn position of the sensor is closely related to the behaviour that can be monitored, with sensors fixed on cows' legs being best for monitoring walking, lying, and standing behaviours [10]. The axillary position would be the ideal position to place the wearable device as it allows simultaneous monitoring of core body temperature, activity, and localization measurements, ensuring comprehensive data collection with a single device. Axillary body temperature would be ideal for continuously monitoring core body temperature over time. This will also ensure that only one device is attached to a cow, as activity and localization measurements also depend on the limb movement. Average axillary temperature ranges from (35.90 C to 36.70C) [11]. Figure 2.1 below describes various positions were, the wearable device could be attached.

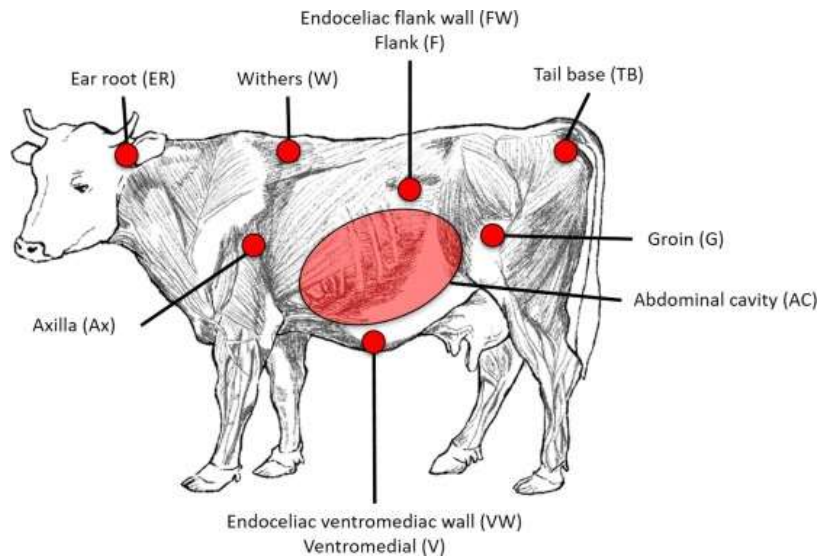


Figure 0.1 Temperature Measuring Sites for Cattle. Adapted from [10]

2.1 Current Trends

The future of animal farming will be guided by the principles of precision, sustainability, and intelligence, which have already begun globally. Accurate cattle production can only be attained with the rapid spread of intelligent technology for early warning of illnesses, feeding precision, and remote diagnosis [12]. In recent years, the cattle farming sector has developed and deployed several technologies to monitor various behavioural and physiological indicators [13, 14] automatically. Collecting large amounts of data is made possible by using sensors and technology, and these data must be analysed with sophisticated statistical methods before any conclusions can be drawn about the animals' behaviour, health, or welfare. Innovations and information technologies (ITs) are essential for achieving sustainable operations because they enable early and rapid disease detection, help measure environmental emissions, and optimize production. Below are some examples of various technologies already being applied in cattle farming.

2.1.0 Wearable sensors

Various sensors have been adopted to track different metrics such as, feeding behaviour, rumination behaviour, rumen pH, temperature, body temperature, laying behaviour, animal activity, and animal location or placement. [15]. Figure 2.2 describes different sensor placement positions currently in use, and the aspects that each sensor measures.

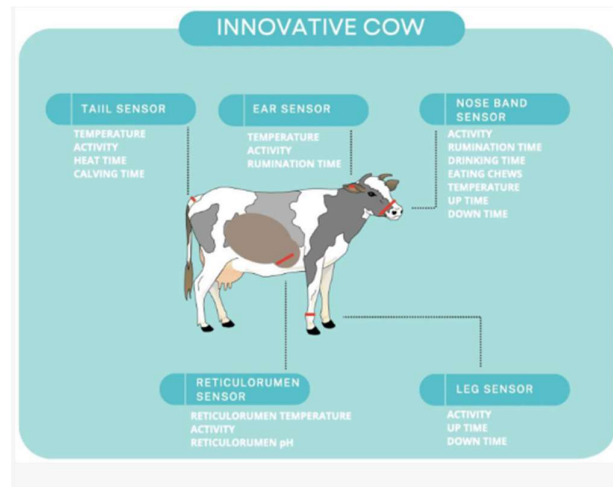


Figure 0.2 Current Sensor Technologies applied on cattle. Adapted from [15]

2.1.1 Surveillance through video and imaging

Livestock management can greatly benefit from automated tracking systems, which can be manual, wearable, or computerized. Computer vision technology offers a non-contact, low-cost method for tracking cattle. Researchers are working on autonomous monitoring, but accurate image identification and recognition through machine learning are still challenging [16].

2.1.2 Infrared thermography

Infrared thermography (IRT) is a method for diagnosing and assessing pain since it indicates physiological changes [17]. Thermography can identify and determine thermal abnormalities in animals by placing a rise or decrease in the skin's surface temperature. IR thermography is a non-invasive method that monitors emitted infrared radiation and displays the data as a thermogram, a visual representation of an object's surface temperature. Each pixel in the thermogram represents the recorded surface temperature of an object. The data can be displayed in grayscale or colour. The warmest areas are displayed in white or red on a colour scale, while the coldest areas are shown in black or blue. When an animal is stressed, the hypothalamic-pituitary-adrenocortical axis is engaged, and heat is created as a result of increased catecholamine and cortisol concentrations, as well as blood flow reactions, resulting in changes in heat generation and heat loss from the animal. As a result, this technique may be beneficial as a general stress indicator [18]. However, it is expensive and prone to unrealistic values resulting from several environmental factors affecting the detected thermal radiations.

2.2 Case studies

A study by Nasirahmadi involved a sensor-based system developed to monitor a pig's health and behavior using accelerometers and image processing techniques. The collected data was analyzed to identify specific behaviors such as eating, drinking, resting, and locomotion. By correlating movement data with visual data, the system aimed to achieve high accuracy in detecting and categorizing different behaviors [19].

Maltz analyzed the application of IoT in monitoring dairy cows, which was explored through the implementation of RFID technology and wireless communication systems. The system deployed RFID tags attached to the cows' ears, which emitted signals picked up by strategically placed RFID readers around the farm [20]. Signals tracked the cows' movements, providing real-time data on their location and activity levels.

Guo investigated using accelerometers and machine learning to detect lameness in cows. The researchers attached accelerometers to the legs of dairy cows to collect data on their movement patterns. The data included step count, stride length, and limb motion. Using this data, machine learning algorithms were trained to recognize normal versus abnormal movement patterns indicative of lameness. Various algorithms, including decision trees, support vector machines (SVM), and neural networks, were tested to determine the most effective approach [21].

2.3 Limitations and gaps to be solved

The project was built on findings from other researchers by integrating advanced real-time analytics and ensuring large-farm scalability. Its infrastructure was simple hence cost friendly. Integration of activity and temperature sensors helped in detecting a broader range of health anomalies. Visualization of real-time cattle metrics via a dashboard gave room for timely actions in case of anomalies. The all-weather system could withstand any weather and environmental condition. The system was also scalable and open to future expansions.

CHAPTER 3: METHODOLOGY

This chapter gives a vivid description of the actual steps followed during the implementation of this project. In line with the specific objectives aforementioned, this process entailed four major steps that upon success would achieve the specific objective it was derived from. Each step was carefully thought through in order to achieve the overall objective of livestock health and activity monitoring through the use of an IoT-based sensor system. The inclusion of machine learning models was crucial in the detection of anomalies within the processed data acquired from the sensor system. The figure 3.1 is a block diagram representation of the steps taken to complete the project.



Figure 0.1 Block diagram

3.1 To Design and Develop Wearable IoT Devices

3.1.1 Circuit and PCB Design

The schematic for integrating the MPU6050 sensor and ESP32 microcontroller was developed using KiCad software. I2C communication lines were established between the MPU6050 and ESP32, ensuring seamless data transfer. A 3.7V LiPo battery provided the power supply through a regulated circuit to ensure stable operation. The PCB layout was optimized for compactness, minimizing trace lengths and potential noise interference. Additionally, breakout pins were included in the design to facilitate flexibility in testing and troubleshooting. Figures 3.2 and 3.3 are the diagrams designed in the KiCAD to show the expected outlook of the IoT device.

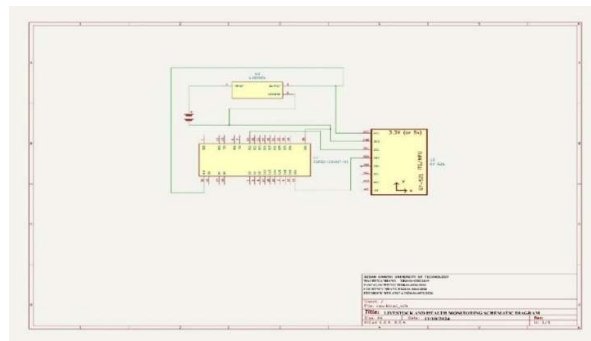


Figure 0.2 KiCad Schematic diagram

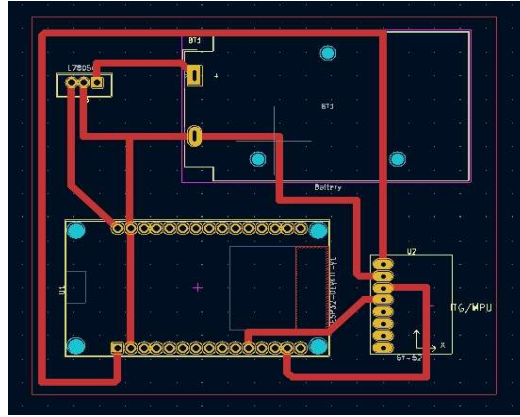


Figure 0.3 PCB Layout

3.1.2 PCB Fabrication and Assembly

The PCB design was fabricated using standard photolithography techniques in the university lab. After the etching process, the copper traces were cleaned to ensure smooth soldering. Components, including the MPU6050 sensor, ESP32 microcontroller, and battery connectors, were carefully mounted and soldered onto the PCB. Quality control measures were implemented during assembly to verify the reliability of connections and overall functionality.

3.1.3 Device Testing

Device testing involved validating each component's performance. The temperature sensor in the MPU6050 was tested by comparing its readings against a calibrated thermometer, confirming its accuracy. The motion detection capabilities of the device were evaluated by subjecting it to known motion patterns, which validated the gyroscope and accelerometer data. Additionally, the LiPo battery was tested for voltage stability and operational duration under standard load conditions.

3.2 Data Transmission and Processing

In order to ensure seamless transmission and processing of data from the individual cattle nodes all the way to the cloud database, this process was divided into four critical steps:

- Transmission of data from all nodes to the central gateway.
- Transmission of processed data and results from the gateway to the cloud-based database.
- Successful fetching of the classified results from the cloud to the mobile application.

3.2.1 Node to gateway transmission

The microcontrollers used for this project were the ESP32 microcontrollers as both the data collection nodes and the receiving central gateway. The ESP32 devices have an in-built communication protocol known as ESP-NOW which can be used when there is a need for them to communicate. However, this protocol only works when the distance between the microcontrollers is not more than 50 meters. It also requires that the ESP32 devices sending the data to the central gateway have to include the gateway's MAC address in the code used to program them.

ESP32 microcontrollers are compatible with the Arduino IDE and therefore made it easier to program them once the required libraries were installed. The libraries indicated in fig 3.4 required for successful data transmission between the nodes and the gateway are the ESP-NOW and Wi-Fi libraries.

```
1  #include <Adafruit MPU6050.h>
2  #include <Adafruit_Sensor.h>
3  #include <esp_now.h>
4  #include <WiFi.h>
5
6  // Initialize MPU6050
7  Adafruit_MPU6050 mpu;
8
9  // Define the MAC address of the gateway
10 uint8_t gatewayAddress[] = {0xEC, 0x64, 0xC9, 0x85, 0x61, 0x98}; // Update with your gateway's MAC address
11
```

Figure 0.4 Code snippet showing libraries

The sending nodes also needed to have a structure in the code to receive and hold data from the MPU6050 sensor. The figure 3.5 shows the code snippets of the sending ESP32 node with the necessary libraries and structures to ensure successful transmission.

```
// Structure to hold sensor data
typedef struct struct_message {
    uint8_t nodeID;      // Unique ID for each node
    float temperature;    // Temperature from MPU6050
    float accelX;
    float accelY;
    float accelZ;
} struct_message;

struct_message sensorData;
```

Figure 0.5 Code snippets for the sensor data holding structure

3.2.2 Gateway to Cloud Transmission

The central gateway was used to receive incoming data from the cattle nodes, hold the machine learning model used to process the data and transmit the results to the cloud. In order for the

incoming data to be received and processed appropriately, the structure used to hold the incoming data had to be similar to the one used when sending it. Once received, the data was pre-processed using helper functions before feeding the data to the anomaly detection model already uploaded onto the gateway.

The anomaly detection model was utilized to process and classify the incoming data from the nodes according to the features used in training the model as shown in the figure below. This was a continuous real-time process and the results of from the model were programmed to display on the serial monitor as long as data was being received from the nodes.

The results would then be ready for transmission to the AWS IoT Core platform via Wi-Fi. An IoT thing was created for the gateway on the AWS IoT Core platform in order for it to send and receive data. A JSON payload was employed to transmit the data in format suitable for transmission and reception by the cloud platform. To establish a connection between the ESP32 gateway and AWS, the PubSubClient library, a file containing the name of the thing sending the data together with the AWS root and device certificates and the private key were required. A stable internet connection was also key. The figures below show the relevant code snippets used to achieve the above.

3.2.3 Cloud to mobile application transmission

Successfully transmitted cattle data and health prediction required database storage which was made possible using DynamoDB that is also hosted on the AWS platform. A rule was created to write all incoming data from the gateway into the arbitrary database. The figure below shows the parameters used to create the rule.

3.3 To Build and Deploy Machine Learning Models

The development and deployment of machine learning models aimed to detect health anomalies in livestock using data collected from IoT-based sensors. This process involved systematic data collection, preparation, model training, and integration into the IoT ecosystem.

3.3.1 Data Collection and Preparation

Data Collection

The data used in this project was generated through a prototype system that represented livestock movement and temperature variations. The system incorporated MPU6050 motion sensors integrated with temperature sensor, ensuring real-world relevance. Data captured

included acceleration metrics ('AccelX', 'AccelY', 'AccelZ') and temperature readings. The collected data covered various health conditions to ensure a robust dataset.

Data Labeling

Threshold-based labeling was employed to categorize data into "Normal" and "Attention required" states. Elevated temperature readings or significant deviations in motion patterns were flagged as indicators of health anomalies.

Data Preparation

The collected dataset was cleaned and split into training (80%) and testing (20%) subsets for evaluation. Features representing health indicators, such as temperature and activity metrics, were extracted and normalized to ensure consistency.

3.3.2 Model Training and Evaluation

Algorithm Selection

The Random Forest Classifier algorithm was utilized for its high accuracy and ability to handle diverse data types.

Training Process

Model training was conducted using Python's Scikit-learn library. Features such as Temperature and ActivityMagnitude (derived from motion sensor data) were selected for training. The model was tuned to balance precision and recall, minimizing false positives while maintaining accurate anomaly detection.

Evaluation Metrics

Evaluation metrics included accuracy which was measured by reliable classification of data into health states, as well as precision and recall metrics which demonstrated balanced performance, ensuring anomalies were correctly identified without excessive false positives.

3.3.3 Model Deployment

Model Conversion

The trained Random Forest Classifier model was converted into C code using the micromlgen library, ensuring compatibility with the ESP32 microcontroller.

Integration with ESP32

The ESP32 gateway, programmed using Arduino IDE, incorporated the converted model. This allowed real-time processing of sensor data to detect anomalies.

Real-Time Processing

The ESP32 executed the model effectively, identifying anomalies with minimal latency (under 50 milliseconds). Data was analyzed in real time, and health statuses were classified as either "Normal" or "Attention required."

3.4 Mobile Application development

A mobile application was developed to display the animal metrics with real-time insights about the animal health. The app was developed using Java programming language, and Android Studio API. AWS DynamoDB was the sole cloud service where the app fetched the metrics. AWS was compatible with android studio enabling this service. While the animals were added, subtracted or modified in the machine-learning model, the app listed them in order as indicated in the Amazon DynamoDB. The Java codes in android studio included Amazon regions, client identity pool and DynamoDB name to fetch the exact metrics. The app also included firebase console to save and authenticate its account users, the farmers. Security rules were applied to ensure that each farmer could only access their animals.

3.5 Flow Chart

The project involved the development of an IoT-based system for livestock health monitoring. It began with the design of an IoT device equipped with sensors to collect data such as temperature and activity levels from the animals. The collected data was transmitted wirelessly to a centralized system for processing. During data processing, machine learning models were used to analyze the data, identifying patterns or anomalies indicative of potential health issues. Finally, the results were visualized through a mobile phone application interface, enabling real-time monitoring and decision-making for improved livestock health management. Figure 3.7 shows a summary of the step-to-step methodology.

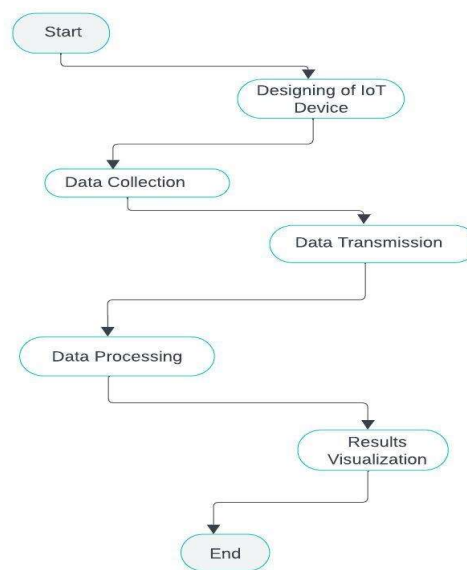


Figure 0.6 Summary of the methodology

CHAPTER 4: RESULTS AND DISCUSSION

This chapter discusses and showcases the results of this project upon successful implementation of the methodology. Each of the steps followed were accurately followed to achieve the overall objective of monitoring cow health using temperature and activity data. These results have been organized to showcase the performance of the data collection nodes, gateway communication, anomaly detection in sensor data and cloud integration. More accurately, the findings included real-time data transmission, anomaly detection outcomes, and system accuracy in identifying cow health statuses. The reliability and efficiency of the developed system is also evaluated under different conditions and variations.

4.1 To Design and Develop Wearable IoT Devices

4.1.1 Circuit and PCB Design

The PCB design successfully achieved the objectives of compactness and reliability. The use of KiCad software facilitated accurate placement of components and optimized trace routing. The final design minimized noise and ensured clear communication between the MPU6050 and ESP32, meeting the project requirements.

4.1.2 PCB Fabrication and Assembly

During PCB fabrication, challenges such as ensuring precise copper trace formation were encountered. These challenges were addressed through iterative quality checks. The final assembly was robust, and soldered connections were thoroughly tested for continuity, confirming electrical stability. The front and the back view of the hardware device were captured in figures 4.1 and 4.2 below.

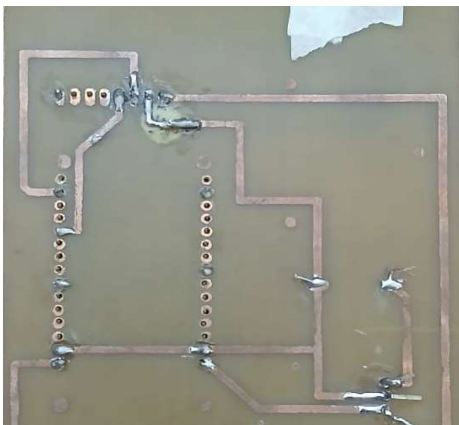


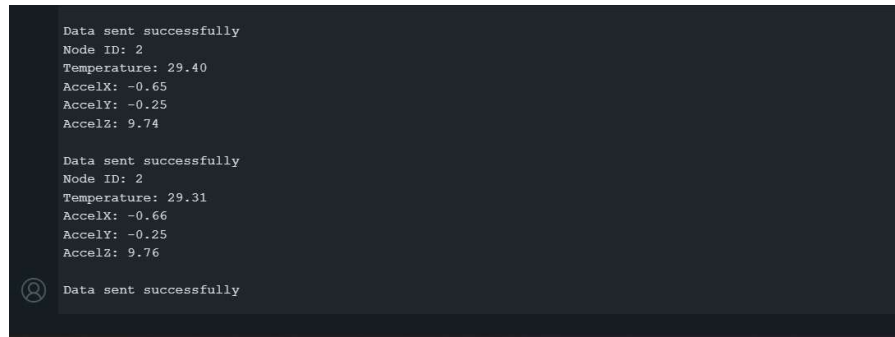
Figure 0.1 Etched PCB showing copper traces and assembly



Figure 0.2 Assembly

4.1.3 Device Testing

The device demonstrated excellent performance during testing. The temperature sensor exhibited high accuracy, with deviations of less than $\pm 0.2^{\circ}\text{C}$ compared to reference devices. Motion tracking tests confirmed a 95% accuracy in detecting predefined tilt angles and movement patterns. The LiPo battery exceeded expectations, powering the device for sixteen hours under continuous operation, surpassing the design target of ten hours. Figure 4.3 was a snippet showing the sensor data acquired, as it appeared in the Arduino environment.



```
Data sent successfully
Node ID: 2
Temperature: 29.40
AccelX: -0.65
AccelY: -0.25
AccelZ: 9.74

Data sent successfully
Node ID: 2
Temperature: 29.31
AccelX: -0.66
AccelY: -0.25
AccelZ: 9.76

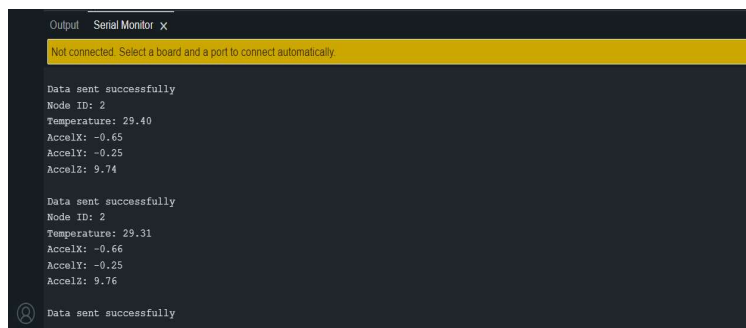
Data sent successfully
```

Figure 0.3 Sensor data acquired

4.2 Transmission and Processing of cattle data

4.2.1 Node to gateway transmission

Figure 4:4 shows the successful transmission of sensor data from the individual cattle nodes to the gateway. This process was achieved through the ESP-NOW protocol that enabled all the ESP32 microcontrollers that were used to send and receive data. The pre-analyzed data received in the ESP 32 gateway could be visualized in the Arduino environment as seen in figure 4.4.



```
Output Serial Monitor x
Not connected Select a board and a port to connect automatically.

Data sent successfully
Node ID: 2
Temperature: 29.40
AccelX: -0.65
AccelY: -0.25
AccelZ: 9.74

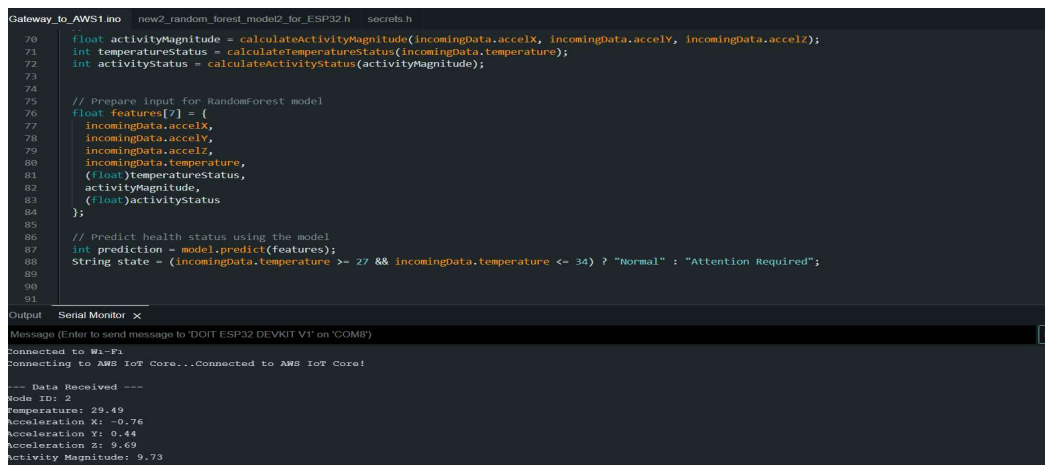
Data sent successfully
Node ID: 2
Temperature: 29.31
AccelX: -0.66
AccelY: -0.25
AccelZ: 9.76

Data sent successfully
```

Figure 0.4 Successful node data transmission

4.2.2 Gateway to cloud transmission

The gateway was programmed to first print the data it received from the nodes in real-time and show the accurate prediction of health status before sending it to the cloud. Transmission of processed data was successfully sent to the cloud via WiFi as confirmed by various debugging statements used in the code. The figures below show results of successful receipt of sensor data, health status prediction as well as successfully published results on AWS IoT Core. Figure 4.5 showed the post-analyze data received from the nodes, as it appeared in the gateway. The results were then sent to the AWS cloud where they would later be fetched as shown in figure 4.6. The gateway-cloud connection was successful and the results were displayed in AWS cloud as per figure 4.7 below.



```
Gateway_to_AWS1.ino  new2_random_forest_model2_for_ESP32.h  secrets.h
70 float activityMagnitude = calculateActivityMagnitude(incomingData.accelX, incomingData.accelY, incomingData.accelZ);
71 int temperatureStatus = calculateTemperatureStatus(incomingData.temperature);
72 int activityStatus = calculateActivityStatus(activityMagnitude);
73
74
75 // Prepare input for RandomForest model
76 float features[7] = {
77     incomingData.accelX,
78     incomingData.accelY,
79     incomingData.accelZ,
80     incomingData.temperature,
81     (float)temperatureStatus,
82     activityMagnitude,
83     (float)activityStatus
84 };
85
86 // Predict health status using the model
87 int prediction = model.predict(features);
88 String state = (incomingData.temperature >= 27 && incomingData.temperature <= 34) ? "Normal" : "Attention Required";
89
90
91
Output  Serial Monitor  X
Message (Enter to send message to 'DOIT ESP32 DEVKIT V1' on 'COM8')
Connected to Wi-Fi
Connecting to AWS IoT Core...Connected to AWS IoT Core!
--- Data Received ---
Node ID: 2
Temperature: 29.49
Acceleration X: -0.76
Acceleration Y: 0.44
Acceleration Z: 9.69
Activity Magnitude: 9.73
```

Figure 0.5 Data successfully received from the nodes



```
137 while (!client.connected()) {
138     Serial.print("Connecting to AWS IoT...");
139     if (client.connect("ESP32Client")) { // Client ID can be any unique string
140         Serial.println("connected!");
141         client.subscribe(AWS_IOT_SUBSCRIBE_TOPIC);
142     } else {
143         Serial.print("failed, rc=");
144         Serial.print(client.state());
145         Serial.println(" try again in 5 seconds");
146         delay(5000);
147     }
148 }
149
150 void loop() {
151     if (!client.connected()) {
152         setupAWSconnection();
153     }
154     client.loop();
155 }
156
Output  Serial Monitor  X
Message (Enter to send message to 'DOIT ESP32 DEVKIT V1' on 'COM8')
Connecting to WiFi...
Connecting to WiFi...
Connecting to WiFi...
Connected to WiFi
Connecting to AWS IoT...connected!
Message arrived on topic: esp32/sub. Message: {
  "message": "Hello from AWS IoT console"
}
```

Figure 0.6 Connection to AWS IoT Core



Figure 0.7 Processed data published to AWS IoT Core

4.2.3 Fetching of data from cloud database to mobile application

The processed data was received on the cloud platform as previously seen. However, the MQTT Test Client only holds data temporarily thus necessitating the use of a database to hold all the incoming data as per their node IDs. The rule that was created was able to save the data as shown in the figure below in DynamoDB which was then accessible by the mobile app. The figure 4.8 was a representation of how the data appeared in the DynamoDB.

The screenshot shows the AWS DynamoDB console interface for a table named 'CowData'. It includes a 'Scan or query items' button and a 'View table details' link. Below, it shows 'Items returned (2)' with a table of data. The table has columns for CowID, HealthStatus, ActivityMagnitude, Temperature, and timestamp.

	CowID (Number)	HealthStatus (String)	ActivityMagnitude	Temperature	timestamp
<input type="checkbox"/>	2	Attention Required	10.75084	-11.56118	173263140
<input type="checkbox"/>	2	Normal	10.09993	29.61823	173263192

Figure 0.8 Data saved in DynamoDB

4.3 To Build and Deploy Machine Learning Models

The machine learning model demonstrated success in detecting anomalies and supporting real-time livestock health monitoring.

4.3.1 Data Collection and Preparation

The automated data collection system reliably captured motion and temperature data under diverse conditions. Threshold-based labeling effectively distinguished between normal and anomalous states, providing a solid foundation for model training. Table 4.1 and 4.2 below show the data collected. This data was used to train and test the working of the machine learning model.

Table 0:1 Sample Data

accel_x	accel_y	accel_z	temperature
10.19	0.31	0.57	25.66
10.18	0.34	0.58	25.62
10.16	0.32	0.55	25.61
9.71	0.2	0.73	25.59
10.19	0.32	0.57	25.55
10.18	0.32	0.53	25.56
10.17	0.32	0.56	25.55
10.17	0.3	0.55	25.52
10.18	0.32	0.58	25.52
10.19	0.31	0.59	25.51

Table 0:2 Labelled data

ModelD	AcclX	AcclY	AcclZ	Temperature	TemperatureStatus	ActivityMagnitude	ActivityStatus	HealthStatus
1	-2.34	5.21	7.83	28.03	Normal	9.691676841	Normal	Healthy Cow
2	6.34	-1.19	7.98	23.52	Normal	10.26119389	High	Healthy Cow
1	-2.29	5.18	7.78	27.93	Normal	9.623143977	Normal	Healthy Cow
2	1.52	-3.56	10.63	23.59	Normal	11.31286436	High	Healthy Cow
1	-1.18	4.58	8.57	27.86	Normal	9.788447272	Normal	Healthy Cow
2	0.27	-3.09	9.93	31.91	High	10.40316779	High	Sick (Attention Required)
1	1.72	-2.17	9.66	31.96	High	10.04902483	High	Sick (Attention Required)
2	-0.19	-2.79	9.62	31.95	High	10.01821341	High	Sick (Attention Required)
1	2.15	-2.45	9.53	31.82	High	10.07203554	High	Sick (Attention Required)
2	0.87	-4.02	10.34	31.68	High	11.12802318	High	Sick (Attention Required)
1	5.22	-8.94	0.44	31.41	High	10.36173731	High	Sick (Attention Required)

4.3.2 Model Training Results

The Random Forest Classifier model achieved the following a model accuracy of 92% on the test set, confirming its robustness in identifying health anomalies. Figure 4.10 showed the machine learning results with the predicted clusters.

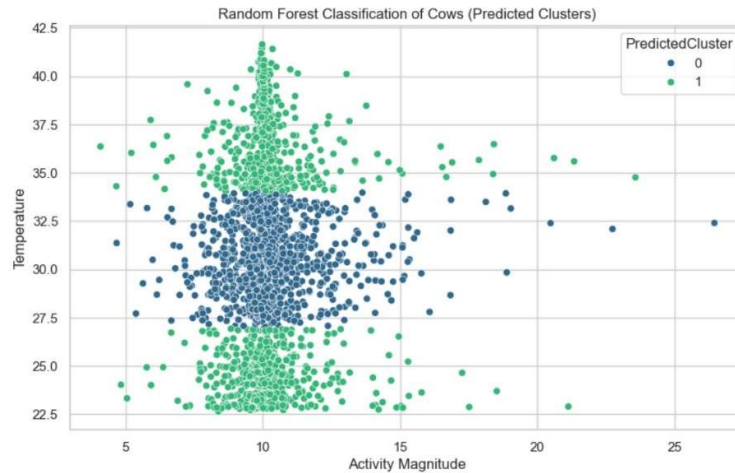


Figure 0.9 Machine learning model results

Feature Importance

Temperature and Activity Magnitude were identified as key features contributing to accurate anomaly detection. They were represented by the figure 4.11 below. The figures 4.12, 4.13, 4.14 and 4.15 showed the activity magnitude and temperature of the two cows as displayed by the machine learning model.

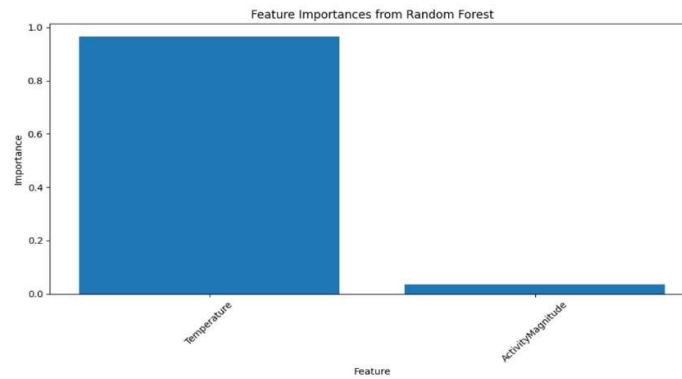


Figure 0.10 Feature importance from Random Forest

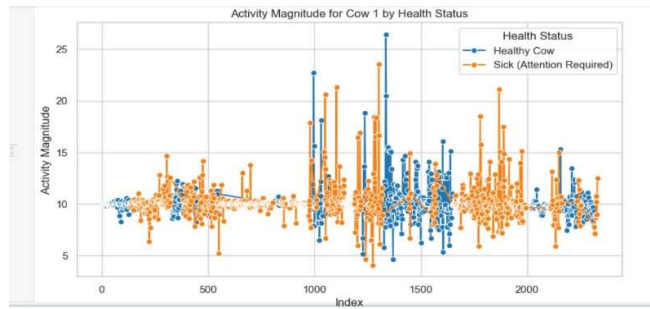


Figure 0.11 Activity magnitude for Cow 1

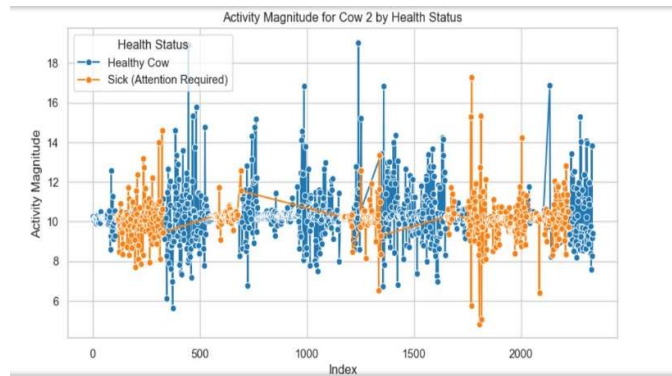


Figure 0.12 Activity magnitude for Cow 2

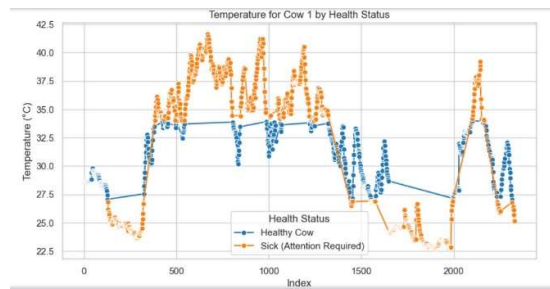


Figure 0.13 Temperature for Cow 1

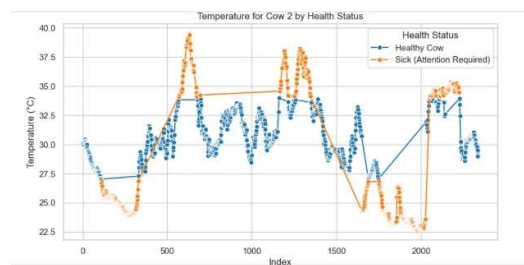


Figure 0.14 Temperature for Cow 2

4.3.3 Deployment and Real-Time Analysis

The deployed model enabled real-time anomaly detection at the ESP32 gateway. Key observations included effective anomaly identification, ensuring timely alerts for farmers, minimal false positives and enhancing the reliability of the notification system.

Figure 4.16 was a section of the code responsible for the machine learning model deployment on the gateway. Figures 4.17 and 4.18 were a representation of analyzed data results for a healthy and a sick cow respectively.

```
5 #include <ArduinoJson.h>
6 #include "secrets.h" // Include your secrets file
7 #include "new2_random_forest_model2_for_ESP32.h" // Include your generated
8
9 #define AWS_IOT_PUBLISH_TOPIC "esp32/pub"
10 #define AWS_IOT_SUBSCRIBE_TOPIC "esp32/sub"
11
12 // Instantiate the RandomForest model globally
13 Eloquent::ML::Port::RandomForest model;
```

Figure 0.15 ML Model deployment on gateway

```
{
  "CowID": 2,
  "Temperature": 29.43,
  "ActivityMagnitude": 9.73,
  "HealthStatus": "Normal"
}
```

Figure 0.16 Analyzed data results for healthy cow

```
{
  "CowID": 2,
  "Temperature": 35.25647,
  "ActivityMagnitude": 9.978549,
  "HealthStatus": "Attention Required"
}
```

Figure 0.17 Analyzed data results for abnormal cow

4.4 Mobile Application Development

The developed mobile application displayed real-time animal metrics. Farmers could create or modify their profiles, while maintaining the privacy of only viewing their animals. The dashboard was easy to use, and gave room for future expansion. Figure 4.18 captured the log in page, where the farmer used their email and password to log in. New users could also create new accounts through the create account tab.

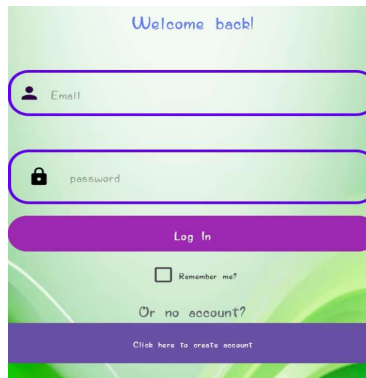


Figure 0.18 Mobile app log in page

Figure 4.19 was the page where new farmers created new accounts.

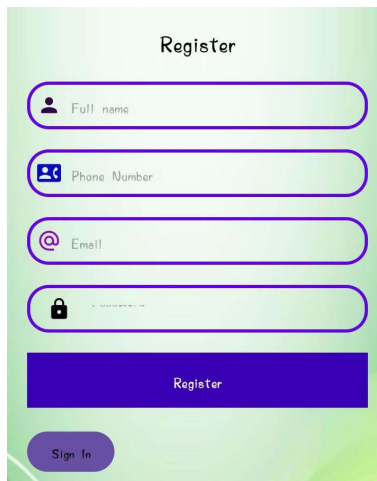


Figure 0.19 Create Account Page

Figure 4.20 showed the dashboard page, which would redirect the farmers to the animals tab. It also prompted the farmer to log out or adjust in-app settings.



Figure 0.20 Dashboard Page

Figure 4.21 fetched the animals by the names displayed from the machine learning model. They appeared by the order of names as listed in the machine learning model.

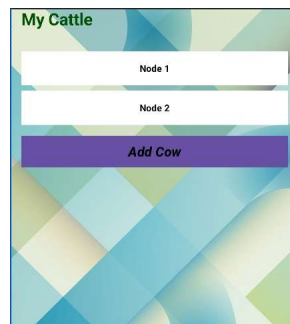


Figure 0.21 The animals page as displayed in the dynamoDB

Figure 4.21 was used to fetch animals from the model. It separated the animals, giving room for future expansions.

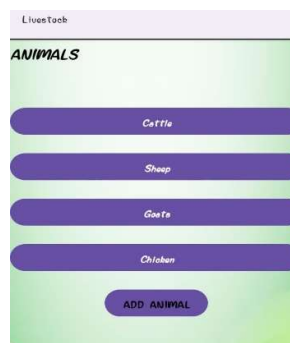


Figure 0.22 Page showing different animals for future expansion

Figure 4.23 and figure 4.24 were the cows' specific metrics, as fetched from the dynamoDB. They represented real-time data, as it keyed-in the dynamoDB.

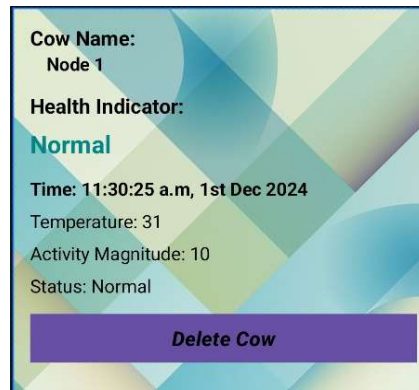


Figure 0.23 Real-time metrics for Cow 1

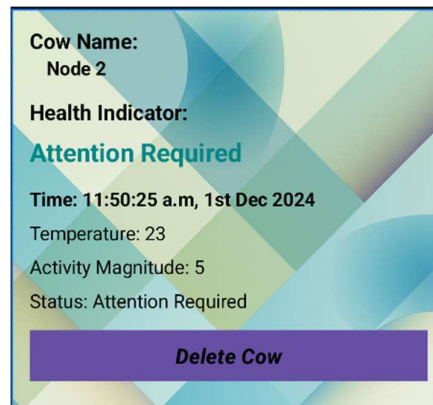


Figure 0.24 Real-time metrics for cow 2

CHAPTER 5: CONCLUSION AND RECOMMENDATION

CONCLUSION

The Livestock Health and Activity Monitoring System using IoT-based sensor technology successfully met the outlined objectives and achieved the project's aim. The design and development of wearable IoT devices were accomplished with compact and reliable devices that integrated MPU6050 (IMU) and temperature sensors. Reliable data transmission and processing were achieved through seamless communication from sensor nodes to the central gateway and AWS cloud, where real-time data analysis was performed. The Random Forest Classifier machine learning model was developed and deployed with an impressive accuracy of 92%, successfully detecting health anomalies in livestock based on movement and temperature data. The project also delivered a user-friendly mobile application, which provided farmers with real-time insights into their livestock's health, enabling them to respond promptly to any detected issues. The system achieved significant improvements in accuracy, with motion tracking at 95% and temperature deviations of only $\pm 0.2^{\circ}\text{C}$, compared to traditional monitoring methods. Furthermore, efficiency was enhanced through real-time anomaly detection, allowing for timely interventions and reducing veterinary costs. Overall, the system provides a scalable, reliable, and cost-effective solution for improving livestock health and productivity.

RECOMMENDATION

While the project achieved its goals, future improvements can further enhance its performance. Sensor placement optimization should be explored to determine the ideal positioning for accurately monitoring cattle behaviors such as lying or rumination. Integrating advanced sensors to monitor additional physiological parameters, such as heart rate and respiration, would allow for the detection of a broader range of health conditions. Battery life can be extended by incorporating energy-efficient microcontrollers or solar-powered IoT nodes, ensuring prolonged operation in remote field conditions. To improve scalability, the system can adopt advanced multi-node communication protocols to handle larger cattle herds effectively. Future researchers could also consider using a long-range transmission protocol to handle hers in a large geographical area. Additionally, future research could focus on enhancing the machine learning models by incorporating larger and more diverse datasets, as well as including deep learning techniques to identify complex health patterns. By addressing these areas, the system can be improved to provide an even more comprehensive, efficient, and intelligent livestock health monitoring solution.

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APPENDIX

Appendix 1: Budget

Table 6:1 Project Budget

ITEM	DESCRIPTION	PRICE PER ITEM	QUANTITY	TOTAL
DEVKITV1 WROOM	Micro-controller IC by ESP32	1000	3	3000
PCB BOARD	Etched PCB with schematic design	100	1	100
MPU 6050	IMU sensor with accelerometer, gyroscope and temperature sensor	500	2	1000
3.7v Lithium battery	Rechargeable Batteries	200	2	400
Miscellaneous	Connectors, power regulator	500	1	500
Grand total				5000

Appendix 2: Code used in the project

Node

```

#include <Adafruit_MPU6050.h>           float accelZ;           if (esp_now_add_peer(&peerInfo) !=
#include <Adafruit_Sensor.h>           } struct_message;           ESP_OK) {
#include <esp_now.h>                     struct_message sensorData;           Serial.println("Failed to add peer");
#include <WiFi.h>                         void setup() {           return;
// Initialize MPU6050                   Serial.begin(115200);           }}
Adafruit_MPU6050 mpu;                   // Initialize MPU6050           void loop() {
// Define the MAC address of the        if (!mpu.begin()) {           // Read accelerometer and temperature
gateway                                  Serial.println("Failed to find      data from MPU6050
uint8_t gatewayAddress[] = {0xEC,      MPU6050 chip");           sensors_event_t a, g, temp;
0x64, 0xC9, 0x85, 0x61, 0x98}; //    while (1) {           mpu.getEvent(&a, &g, &temp);
Update with your gateway's MAC        delay(10);           sensorData.temperature =
address                                } }           temp.temperature; // Get temperature
// Define a unique node ID for this    from MPU6050
sensor node
const uint8_t NODE_ID = 1;
#include <Adafruit_MPU6050.h>           mpu.setAccelerometerRange(MPU6050
0_RANGE_2_G);
#include <Adafruit_Sensor.h>
#include <esp_now.h>
#include <WiFi.h>
// Initialize MPU6050
Adafruit_MPU6050 mpu;
// Define the MAC address of the        mpu.setGyroRange(MPU6050_RANGE
E_250_DEG);
gateway
uint8_t gatewayAddress[] = {0xEC,      mpu.setFilterBandwidth(MPU6050_B
AND_21_HZ);
0x64, 0xC9, 0x85, 0x61, 0x98}; //    Serial.println("MPU6050
initialized");
Update with your gateway's MAC        // Initialize WiFi in station mode
address
// Define a unique node ID for this    WiFi.mode(WIFI_STA);
sensor node
const uint8_t NODE_ID = 1;
// Structure to hold sensor data
typedef struct struct_message {
uint8_t nodeId; // Unique ID for
each node
float temperature; // Temperature
from MPU6050
float accelX;
float accelY;
float accelZ;
}
// Register the gateway's MAC address
esp_now_peer_info_t peerInfo;
memcpy(peerInfo.peer_addr,
gatewayAddress, 6);
peerInfo.channel = 0;
peerInfo.encrypt = false;
if (esp_now_init() != ESP_OK) {
Serial.println("Error initializing
ESP-NOW");
return;
}
Serial.println(sensorData.temperature);
Serial.print("AccelX: ");
Serial.println(sensorData.accelX);
Serial.print("AccelY: ");
Serial.println(sensorData.accelY);
Serial.print("AccelZ: ");
Serial.println(sensorData.accelZ);
Serial.println(); // Print a blank line
between each data set
// Send data via ESP-NOW

```



```

    esp_err_t result =
    esp_now_send(gatewayAddress,
    (uint8_t *)&sensorData,
    sizeof(sensorData));

```

```

    if (result == ESP_OK) {

        Serial.println("Data sent
        successfully");

    } else {

        Serial.println("Error sending data");
    }

    delay(5000); // Delay between sends
}

```

Gateway

```

#include <esp_now.h>

#include <esp_wifi.h>

#include <WiFi.h>

#include <WiFiClientSecure.h>

#include <PubSubClient.h>

#include <ArduinoJson.h>

#include "secrets.h" // Include your
secrets file

#include
"new2_random_forest_model2_for_ES
P32.h" // Include your generated C
model

#define ESP_MAC_WIFI_STA 0

#define AWS_IOT_PUBLISH_TOPIC
"esp32/pub"

#define
AWS_IOT_SUBSCRIBE_TOPIC
"esp32/sub"

// Wi-Fi credentials

#define WIFI_SSID "QUOVADIS
YOUTH HUB"

#define WIFI_PASSWORD
"Quo@2023"

// Instantiate the RandomForest model
globally

Eloquent::ML::Port::RandomForest
model;

// MQTT client and secure Wi-Fi client

WiFiClientSecure net;

```

```

PubSubClient client(net);

// Structure to hold incoming sensor
data

```

```

typedef struct SensorData {

```

```

    uint8_t nodeID;

    float temperature;

    float accelX;

    float accelY;

    float accelZ;
} SensorData;

```

```

// Helper functions

```

```

int calculateTemperatureStatus(float
temperature) {

    if (temperature > 34) return 1; //
Elevated temperature

    else if (temperature < 27) return -1; //
Low temperature

    return 0; // Normal temperature
}

```

```

float calculateActivityMagnitude(float
accelX, float accelY, float accelZ) {

    return sqrt(accelX * accelX + accelY *
accelY + accelZ * accelZ);
}

```

```

int calculateActivityStatus(float
activityMagnitude) {

    if (activityMagnitude > 10) return 1; //
High activity

    else if (activityMagnitude < 7) return -
1; // Low activity

    return 0; // Normal activity
}

```

```

// Callback function for receiving data
via ESP-NOW

```

```

void onDataReceive(const
esp_now_recv_info_t *info, const
uint8_t *data, int len) {

```

```

    Serial.println("onDataReceive
    triggered");

```

```

    Serial.print("Data length received: ");

```

```

Serial.println(len);

```

```

Serial.print("Expected length: ");

```

```

Serial.println(sizeof(SensorData));

```

```

// Ensure we don't read more data than
expected

```

```

//if (len != sizeof(SensorData)) {

```

```

    //Serial.println("Received incorrect
    data length");
}

```

```

//return;

```

```

//}

```

```

// Copy the incoming data into the
structure

```

```

SensorData incomingData;

```

```

memcpy(&incomingData, data,
sizeof(incomingData));

```

```

// Calculate additional metrics

```

```

float activityMagnitude =
calculateActivityMagnitude(incoming
Data.accelX, incomingData.accelY,
incomingData.accelZ);

```

```

int temperatureStatus =
calculateTemperatureStatus(incoming
Data.temperature);

```

```

int activityStatus =
calculateActivityStatus(activityMagnit
ude);

```

```

// Prepare input for RandomForest
model

```

```

float features[7] = {

```

```

    incomingData.accelX,

```

```

    incomingData.accelY,

```

```

    incomingData.accelZ,

```

```

    incomingData.temperature,

```

```

    (float)temperatureStatus,

```

```

    activityMagnitude,

```

```

    (float)activityStatus
};

```

```

// Predict health status using the model

```

```

int prediction =
model.predict(features);

```

```

String      state      =
(incomingData.temperature >= 27 &&
incomingData.temperature <= 34) ?
"Normal" : "Attention Required";

//Prepare JSON payload for AWS IoT
Core

DynamicJsonDocument doc(256);

doc["CowID"]      =
incomingData.nodeID;

doc["Temperature"]      =
incomingData.temperature;

doc["ActivityMagnitude"]      =
activityMagnitude;

doc["HealthStatus"] = state;

String json;

serializeJson(doc, json);

// Publish to AWS IoT Core

if
(!client.publish(AWS_IOT_PUBLISH
_TOPIC, json.c_str())) {

    Serial.println("Failed to publish
message");

} else {

    Serial.println("Message successfully
published to AWS IoT Core");

}}

void setup() {

    Serial.begin(115200);

    WiFi.begin(WIFI_SSID,
WIFI_PASSWORD);

    while      (WiFi.status()      !=
WL_CONNECTED) {

        delay(1000);

        Serial.println("Connecting      to
WiFi...");

    }

    Serial.println("Connected to WiFi");

    // Print the MAC address

    Serial.print("MAC Address: ");

    Serial.println(WiFi.macAddress());

    setupAWSConnection();

    // Step 2: Check ESP-NOW
initialization

    if(esp_now_init() == ESP_OK) {

        Serial.println("ESP-NOW Initialized
successfully");

    } else {

        Serial.println("ESP-NOW
Initialization failed");

        return; // Stop setup if ESP-NOW
initialization failed

    }

    // Step 3: Register receive callback

    if
(esp_now_register_recv_cb(onDataRec
eive) == ESP_OK) {

        Serial.println("Callback registered
successfully");

    } else {

        Serial.println("Failed to register
callback");

        return; // Stop setup if callback
registration failed

    }

    // Step 4: Print MAC address to ensure
pairing is correct

    //uint8_t mac[6];

    //esp_read_mac(mac,
ESP_MAC_WIFI_STA);

    //Serial.printf("MAC      Address:
%02X:%02X:%02X:%02X:%02X:%0
2X\n", mac[0], mac[1], mac[2], mac[3],
mac[4], mac[5]);

    //Serial.println("Setup complete");

}

// Helper functions (same as in your
code)

void mqttCallback(char* topic, byte*
payload, unsigned int length) {

    Serial.print("Message arrived on
topic: ");

    Serial.print(topic);

    Serial.print(". Message: ");

    for (int i = 0; i < length; i++) {

        Serial.print((char)payload[i]);

    }

    Serial.println();

}

void setupAWSConnection() {

    net.setCACert(AWS_ROOT_CA);

    net.setCertificate(CERTIFICATE);

    net.setPrivateKey(PRIVATE_KEY);

    client.setServer(AWS_IOT_ENDPOIN
T, 8883);

    client.setCallback(mqttCallback);

    while (!client.connected()) {

        Serial.print("Connecting to AWS
IoT...");

        if (client.connect("ESP32Client")) {
// Client ID can be any unique string

            Serial.println("connected!");

            client.subscribe(AWS_IOT_SUBSCRI
BE_TOPIC);

        } else {

            Serial.print("failed, rc=");

            Serial.print(client.state());

            Serial.println(" try again in 5
seconds");

            delay(5000); }}}

void loop() {

    SensorData incomingData;

    float      activityMagnitude      =
calculateActivityMagnitude(incoming
Data.accelX, incomingData.accelY,
incomingData.accelZ);

    String      state      =
(incomingData.temperature >= 27 &&
incomingData.temperature <= 34) ?
"Normal" : "Attention Required";

    if      (WiFi.status()      ==
WL_CONNECTED) {

```

```

// Prepare JSON payload

String jsonData = "{\"CowID\": 1,
\"Temperature\": 29.5,
\"ActivityMagnitude\": 0.75,
\"HealthStatus\": \"Normal\"}";

// Print the received data to the Serial
Monitor

Serial.println("\n--- Data Received ---");

Serial.print("Node ID: ");

Serial.println(incomingData.nodeID);

Serial.print("Temperature: ");

Serial.println(incomingData.temperature);

Serial.print("Acceleration X: ");

Serial.println(incomingData.accelX);

Serial.print("Acceleration Y: ");

Serial.println(incomingData.accelY);

Serial.print("Acceleration Z: ");

Serial.println(incomingData.accelZ);

Serial.print("Activity Magnitude: ");

Serial.println(activityMagnitude);

// Print health status

Serial.print("Health Status
(Prediction): ");

Serial.println(state);

}

if (!client.connected()) {

    setupAWSConnection();

}

client.loop();

delay(10000); // Wait 10 seconds
before sending next data

}

```

Data Saving Automation Script

```
import serial
```

```

import csv

import time

# Initialize serial connection (update
COM port and baud rate as needed)

ser = serial.Serial('COM8', 115200) #
Replace 'COM8' with your port (e.g.,
'/dev/ttyUSB0' on Linux/Mac)

# Open a CSV file to store data

with open('livestock_data.csv', 'w',
newline="", encoding="utf-8") as file:

    writer = csv.writer(file)

    # Write the header

    writer.writerow(["Timestamp",
"accelX", "accelY", "accelZ",
"Temperature"])

    while True:

        try:

            # Read a line from the serial port

            line = ser.readline().decode('utf-
8', errors='replace').strip()

            data = line.split('\t') # Assuming
data is tab-separated

            # Get the current timestamp

            timestamp = time.strftime("%Y-
%m-%d %H:%M:%S")

            # Write the data to the CSV file

            writer.writerow([timestamp] +
data)

            file.flush() # Ensure data is
written to disk after each row

            print(f"Data logged:
{timestamp}, {data}")

        except KeyboardInterrupt:

            print("Data collection
stopped.")

            break

Random Forest Classifier

import pandas as pd

import numpy as np

from sklearn.preprocessing import
StandardScaler, MinMaxScaler

```

```

from sklearn.ensemble import
RandomForestClassifier

from sklearn.svm import SVC

import joblib

import matplotlib.pyplot as plt

import seaborn as sns

from sklearn.decomposition import
PCA

import os

# Read the CSV File

file_path =
r'C:\Users\Courtney\Desktop\5.2\senso
r_data.csv' # Prefix with 'r'

data = pd.read_csv(file_path)

# Define the correct file paths

input_file_path =
r'C:\Users\Courtney\Desktop\5.2\senso
r_data.csv'

output_file_path =
r'C:\Users\Courtney\Desktop\5.2\labele
dsensor_data_file.csv'

# Function to classify temperature

def classify_temperature(temp):

    if temp < 27:

        return 'Low'

    elif temp > 34:

        return 'High'

    else:

        return 'Normal'

# Calculate the magnitude of activity
from acceleration data

def calculate_activity_magnitude(row):

    return np.sqrt(row['AccelX']**2 +
row['AccelY']**2 + row['AccelZ']**2)

# Define thresholds for activity

def
categorize_activity(activity_magnitude
):

    if activity_magnitude < 7:

        return 'Low'

```

```

elif activity_magnitude > 10:

    return 'High'

else:

    return 'Normal'

# Simplified function to label health
status with default state as "Sick
(Attention Required)"

def label_health_status(row):

    temp_status =
classify_temperature(row['Temperature
'])

    activity_magnitude =
calculate_activity_magnitude(row)

    activity_status =
categorize_activity(activity_magnitude
)

    if temp_status == 'Normal' and
activity_status in
['Low','Normal','High']:

        return 'Healthy Cow'

    else:

        return 'Sick (Attention Required)'

# Check if the input file exists

if os.path.isfile(input_file_path):

    try:

        # Load the CSV file into a
DataFrame

        df = pd.read_csv(input_file_path)

        # Apply the functions to classify
and label

        df['TemperatureStatus'] =
df['Temperature'].apply(classify_tempe
rature)

        df['ActivityMagnitude'] =
df.apply(calculate_activity_magnitude,
axis=1)

        df['ActivityStatus'] =
df['ActivityMagnitude'].apply(categori
ze_activity)

        df['HealthStatus'] =
df.apply(label_health_status, axis=1)

        # Save the updated DataFrame to a
new CSV file

        df.to_csv(output_file_path,
index=False)

        # Verify if the file was created

        if os.path.isfile(output_file_path):

            print(f"File successfully saved
to: {output_file_path}")

            else:

                print(f"File not found at the
expected path: {output_file_path}")

            except pd.errors.EmptyDataError:

                print("The file is empty.")

            except pd.errors.ParserError:

                print("Error parsing the file. It may
not be a valid CSV.")

            except Exception as e:

                print(f"An unexpected error
occurred: {e}")

            else:

                print(f"Input file not found at the
path: {input_file_path}")

import pandas as pd

from sklearn.model_selection import
train_test_split

from sklearn.preprocessing import
LabelEncoder

from sklearn.ensemble import
RandomForestClassifier

import joblib

# Load the dataset (update with the path
to your CSV file)

file_path =
r'C:\Users\Courtney\Desktop\5.2\labe
lsensor_data_file.csv'

data = pd.read_csv(file_path)

# Convert categorical columns to
numerical using Label Encoding

le = LabelEncoder()

# Encode the 'HealthStatus' column

data['HealthStatus'] =
le.fit_transform(data['HealthStatus']) #
Target variable

# Select features (input variables) and
the target variable

# Only include features used in the
Arduino code

X = data[['Temperature',
'ActivityMagnitude']]

y = data['HealthStatus']

# Split the dataset into training and
testing sets (80% train, 20% test)

X_train, X_test, y_train, y_test =
train_test_split(X, y, test_size=0.2,
random_state=100)

# Output the shapes of training and
testing data

print(f"Training data shape:
{X_train.shape}, Test data shape:
{X_test.shape}")

print(f"Training labels shape:
{y_train.shape}, Test labels shape:
{y_test.shape}")

# Train the RandomForest model

model = RandomForestClassifier()

model.fit(X_train, y_train)

# Save the trained model

joblib.dump(model,
'random_forest_model.pkl')

from sklearn.preprocessing import
LabelEncoder

import pandas as pd

from sklearn.model_selection import
train_test_split, cross_val_score

from sklearn.preprocessing import
StandardScaler

from sklearn.ensemble import
RandomForestClassifier

from sklearn.metrics import
accuracy_score, precision_score,
recall_score, f1_score, make_scorer

from sklearn.compose import
ColumnTransformer

from sklearn.pipeline import Pipeline

import joblib

from micromlgen import port

```

```

# Load the dataset (update with the path
to your CSV file)

file_path =
r'C:\Users\Courtney\Desktop\5.2\labeledsensor_data_file.csv'

data = pd.read_csv(file_path)

# Encode the target variable

le = LabelEncoder()

data['HealthStatus'] =
le.fit_transform(data['HealthStatus']) #
Converts 'Healthy Cow' and 'Sick
(Attention Required)' to numeric

# Select only the features you are using
in the ESP32 code: Temperature and
ActivityMagnitude

X = data[['Temperature',
'ActivityMagnitude']]

y = data['HealthStatus']

# Define preprocessor to scale
numerical features (in this case, only 2
features)

preprocessor = ColumnTransformer(

    transformers=[

        ('num', StandardScaler(),
['Temperature', 'ActivityMagnitude'])

    ])

# Create a pipeline that preprocesses the
data and then applies the Random
Forest model

pipeline = Pipeline(steps=[

    ('preprocessor', preprocessor),

    ('classifier',
RandomForestClassifier(n_estimators=
100, random_state=100))

])

# Split the data into training and testing
sets

X_train, X_test, y_train, y_test =
train_test_split(X, y, test_size=0.2,
random_state=100)

# Train the model

pipeline.fit(X_train, y_train)

# Make predictions on the test data

y_pred = pipeline.predict(X_test)

# Define the positive label after
encoding (assuming 'Sick (Attention
Required)' was encoded as 1)

positive_label = 1

# Evaluate the model's performance on
the test data

accuracy = accuracy_score(y_test,
y_pred)

precision = precision_score(y_test,
y_pred, pos_label=positive_label)

recall = recall_score(y_test, y_pred,
pos_label=positive_label)

f1 = f1_score(y_test, y_pred,
pos_label=positive_label)

# Print the results

print("Test Data Performance:")

print(f"Accuracy: {accuracy *
100:.2f}%")

print(f"Precision: {precision:.2f}")

print(f"Recall: {recall:.2f}")

print(f"F1-Score: {f1:.2f}")

# Define custom scorers with the
appropriate positive label

precision_scorer =
make_scorer(precision_score,
pos_label=positive_label)

recall_scorer =
make_scorer(recall_score,
pos_label=positive_label)

f1_scorer = make_scorer(f1_score,
pos_label=positive_label)

# Perform cross-validation with 5 folds

cv_accuracy =
cross_val_score(pipeline, X, y, cv=5,
scoring='accuracy')

cv_precision =
cross_val_score(pipeline, X, y, cv=5,
scoring=precision_scorer)

cv_recall = cross_val_score(pipeline,
X, y, cv=5, scoring=recall_scorer)

cv_f1 = cross_val_score(pipeline, X, y,
cv=5, scoring=f1_scorer)

# Print the average cross-validation
results

print("\nCross-Validation
Performance:")

print(f"Cross-Validation Accuracy:
{cv_accuracy.mean() * 100:.2f}%")

print(f"Cross-Validation Precision:
{cv_precision.mean():.2f}")

print(f"Cross-Validation Recall:
{cv_recall.mean():.2f}")

print(f"Cross-Validation F1-Score:
{cv_f1.mean():.2f}")

# After training the RandomForest
model, save the model

joblib.dump(pipeline,
'new2_random_forest_model.pkl')

# Convert the model to C code using
micromlgen for deployment on ESP32

c_code =
port(pipeline.named_steps['classifier'])

# Save the generated C code to a file

with
open('new2_random_forest_model2_fo
r_ESP32.c', 'w') as f:

    f.write(c_code)

import matplotlib.pyplot as plt

import numpy as np

# Extract feature importances from the
trained RandomForest model

importances =
pipeline.named_steps['classifier'].featu
re_importances_

# Updated features list based on your
current model setup

features = ['Temperature',
'ActivityMagnitude']

# Display the feature importances
alongside their corresponding feature
names

for feature, importance in zip(features,
importances):

    print(f"Feature: {feature},
Importance: {importance:.4f}")

# Sort the feature importances in
descending order

indices = np.argsort(importances)[::-1]

# Plot the feature importances

```

```

plt.figure(figsize=(10, 6))

plt.title("Feature Importances from
Random Forest")

plt.bar(range(len(importances)),
importances[indices], align="center")

plt.xticks(range(len(importances)),
[features[i] for i in indices],
rotation=45)

plt.ylabel("Importance")

plt.xlabel("Feature")

plt.tight_layout()

# Display the plot

plt.show()

import matplotlib.pyplot as plt

import seaborn as sns

from sklearn.ensemble import
RandomForestClassifier

from sklearn.model_selection import
train_test_split

# Assuming 'data' DataFrame contains
ActivityMagnitude, Temperature,
AccelX, AccelY, AccelZ, and
'HealthStatus'

# Define the feature columns

features = ['ActivityMagnitude',
'Temperature', 'AccelX', 'AccelY', 'AccelZ']

# Split the data into features (X) and
target (y)

X = data[features]

y = data['HealthStatus'] # Assuming
this column represents the health status
(e.g., 'Healthy' or 'Sick')

# Split into training and test sets
(optional, could use all data for
visualization)

X_train, X_test, y_train, y_test =
train_test_split(X, y, test_size=0.2,
random_state=42)

# Train the Random Forest Classifier

rf_classifier =
RandomForestClassifier(n_estimators=
100, random_state=42)

rf_classifier.fit(X_train, y_train)

# Predict the health status (or cluster)
for the entire dataset

data['PredictedCluster'] =
rf_classifier.predict(X)

# Plot the scatterplot for
'ActivityMagnitude' vs 'Temperature',
colored by the predicted cluster

plt.figure(figsize=(10, 6))

sns.scatterplot(data=data,
x='ActivityMagnitude',
y='Temperature',
hue='PredictedCluster',
palette='viridis')

plt.title("Random Forest Classification
of Cows (Predicted Clusters)")

plt.xlabel("Activity Magnitude")

plt.ylabel("Temperature")

plt.show()

import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

# Load your data

data =
pd.read_csv(r'C:\Users\Courtney\Desktop\5.2\labeledsensor_data_file.csv')

# Filter out 'Unknown Status' from the
dataset

data =
data[data['HealthStatus'].isin(['Healthy
Cow', 'Sick (Attention Required)'])]

# Set up the plotting style

sns.set(style="whitegrid")

# Separate the data by NodeID (Cows 1
and 2)

data_cow_1 = data[data['NodeID'] ==
1]

data_cow_2 = data[data['NodeID'] ==
2]

# Function to plot Activity Magnitude
for each cow and compare health

def plot_activity_magnitude(data,
cow_id):

plt.figure(figsize=(10, 5))

sns.lineplot(data=data, x=data.index,
y='ActivityMagnitude',
hue='HealthStatus', marker='o',
palette='tab10')

plt.title(f'Activity Magnitude for
Cow {cow_id} by Health Status')

plt.xlabel('Index')

plt.ylabel('Activity Magnitude')

plt.legend(title='Health Status')

plt.grid(True)

plt.show()

# Plot Activity Magnitude for Cow 1

plot_activity_magnitude(data_cow_1,
cow_id=1)

# Plot Activity Magnitude for Cow 2

plot_activity_magnitude(data_cow_2,
cow_id=2)

# Function to plot other parameters for
comparison between cows

def plot_parameter(data, parameter,
title, ylabel, cow_id):

plt.figure(figsize=(10, 5))

sns.lineplot(data=data, x=data.index,
y=parameter, hue='HealthStatus',
marker='o', palette='tab10')

plt.title(f'{title} for Cow {cow_id}
by Health Status')

plt.xlabel('Index')

plt.ylabel(ylabel)

plt.legend(title='Health Status')

plt.grid(True)

plt.show()

# Plot Temperature for Cow 1

plot_parameter(data_cow_1,
'Temperature', 'Temperature',
'Temperature (°C)', cow_id=1)

# Plot Temperature for Cow 2

plot_parameter(data_cow_2,
'Temperature', 'Temperature',
'Temperature (°C)', cow_id=2)

# Plot Acceleration X for Cow 1

```

```
plot_parameter(data_cow_1, 'AcclX',
'Acceleration X', 'Acceleration X',
cow_id=1)
```

```
# Plot Acceleration X for Cow 2
```

```
plot_parameter(data_cow_2, 'AcclX',
'Acceleration X', 'Acceleration X',
cow_id=2)
```

```
# Plot Acceleration Y for Cow 1
```

```
plot_parameter(data_cow_1, 'AcclY',
'Acceleration Y', 'Acceleration Y',
cow_id=1)
```

```
# Plot Acceleration Y for Cow 2
```

```
plot_parameter(data_cow_2, 'AcclY',
'Acceleration Y', 'Acceleration Y',
cow_id=2)
```

```
# Plot Acceleration Z for Cow 1
```

```
plot_parameter(data_cow_1, 'AcclZ',
'Acceleration Z', 'Acceleration Z',
cow_id=1)
```

```
# Plot Acceleration Z for Cow 2
```

```
plot_parameter(data_cow_2, 'AcclZ',
'Acceleration Z', 'Acceleration Z',
cow_id=2)
```

```
import pandas as pd
```

```
import seaborn as sns
```

```
import matplotlib.pyplot as plt
```

```
# Load your data
```

```
data =
pd.read_csv(r'C:\Users\Courtney\Desktop\5.2\labeleddata_data_file.csv')
```

```
# Filter out 'Unknown Status' from the
dataset
```

```
data =
data[data['HealthStatus'].isin(['Healthy
Cow', 'Sick (Attention Required)'])]
```

```
# Select only numeric columns for
correlation matrix
```

```
numeric_data =
data.select_dtypes(include=[float, int])
```

```
# Calculate the correlation matrix
```

```
correlation_matrix =
numeric_data.corr()
```

```
# Plot the correlation matrix
```

```
plt.figure(figsize=(10, 8))
```

```
sns.heatmap(correlation_matrix,
annot=True, cmap='coolwarm',
fmt='.2f')
```

```
plt.title("Correlation Matrix")
```

```
plt.show()
```

```
from sklearn.ensemble import
IsolationForest
```

```
# Apply Isolation Forest for anomaly
detection
```

```
model =
IsolationForest(contamination=0.05,
random_state=42)
```

```
data['Anomaly'] =
model.fit_predict(data[['ActivityMagni
tude', 'Temperature', 'AcclX', 'AcclY',
'AcclZ']])
```

```
# Visualize anomalies
```

```
sns.scatterplot(data=data,
x='ActivityMagnitude',
y='Temperature', hue='Anomaly',
palette='coolwarm')
```

```
plt.title("Anomaly Detection in Cows")
```

```
plt.show()
```

```
App development code
```

```
Main Activity Java Code
```

```
package com.example.animalwashuka;
```

```
import android.content.Intent;
```

```
import android.os.Bundle;
```

```
import android.text.TextUtils;
```

```
import android.view.View;
```

```
import android.widget.Button;
```

```
import android.widget.CheckBox;
```

```
import android.widget.EditText;
```

```
import android.widget.Toast;
```

```
import androidx.annotation.NonNull;
```

```
import
androidx.appcompat.app.AppCompatActivity;
```

```
import
com.google.android.gms.tasks.OnCom
pleteListener;
```

```
import
com.google.android.gms.tasks.Task;
```

```
import
com.google.firebase.auth.AuthResult;
```

```
import
com.google.firebase.auth.FirebaseAuth
;
```

```
public class MainActivity extends
AppCompatActivity {
```

```
private EditText emailEditText;
```

```
private EditText passwordEditText;
```

```
private Button loginButton;
```

```
private CheckBox
rememberMeCheckBox;
```

```
private Button createAccount;
```

```
private FirebaseAuth mAuth;
```

```
@Override
```

```
protected void onCreate(Bundle
savedInstanceState) {
```

```
super.onCreate(savedInstanceState);
```

```
setContentView(R.layout.activity_mai
n);
```

```
// Initialize FirebaseAuth
```

```
mAuth =
FirebaseAuth.getInstance();
```

```
// Initialize views
```

```
emailEditText =
findViewById(R.id.username);
```

```
passwordEditText =
findViewById(R.id.password_toggle);
```

```
loginButton =
findViewById(R.id.loginButton);
```

```
createAccount =
findViewById(R.id.createAccount);
```

```
// Set onClick listener for Login
button
```

```
loginButton.setOnClickListener(new
View.OnClickListener() {
```

```
@Override
```

```
public void onClick(View v) {
```

```
loginUser();
```

```
}
```

```
});
```

```
// Set onClick listener for Create
Account button
```

```
createAccount.setOnClickListener(new
View.OnClickListener() {
```

```
@Override
```

```
public void onClick(View v) {
```

```

        Intent intent = new
Intent(MainActivity.this,
register.class);

        startActivity(intent);

        finish();

    }

});

}

private void loginUser() {

    String email =
emailEditText.getText().toString().trim
();

    String password =
passwordEditText.getText().toString().t
rim();

    if (TextUtils.isEmpty(email)) {

        Toast.makeText(MainActivity.this,
"Email is required",
Toast.LENGTH_SHORT).show();

        return;

    }

    if (TextUtils.isEmpty(password))
    {

        Toast.makeText(MainActivity.this,
"Password is required",
Toast.LENGTH_SHORT).show();

        return;

    }

    // Sign in with email and
password without email verification
check

    mAuth.signInWithEmailAndPassword(
email, password)

    .addOnCompleteListener(this, new
OnCompleteListener<AuthResult>() {

        @Override

        public void
onComplete(@NonNull
Task<AuthResult> task) {

            if (task.isSuccessful()) {

                // Sign in successful,
no email verification required

                Toast.makeText(MainActivity.this,
"Log in successful.",
Toast.LENGTH_SHORT).show();

                // Redirect to
DashboardActivity

```

```

        Intent intent = new
Intent(MainActivity.this,
Dashboard.class);

        startActivity(intent);

        finish();

    } else {

        // If sign in fails,
display a message to the user

        Toast.makeText(MainActivity.this,
"Log in failed. Please check your
credentials and try again.",
Toast.LENGTH_SHORT).show();

    } });

}

MainActivity XML

<?xml version="1.0" encoding="utf-
8"?>

<LinearLayout
xmlns:android="http://schemas.android
.com/apk/res/android"

xmlns:app="http://schemas.android.co
m/apk/res-auto"

xmlns:tools="http://schemas.android.c
om/tools"

android:layout_width="match_parent"

android:layout_height="match_parent"

android:background="@drawable/pay
ment"

android:gravity="center"

android:orientation="vertical"

tools:context=".MainActivity">

<TextView

    android:id="@+id/LogInText"

    android:layout_width="match_parent"

    android:layout_height="wrap_content"

    android:text="Welcome back!"

    android:textAlignment="center"

    android:layout_marginTop="10dp"

    android:textColor="#0329FF"

    android:textSize="30dp"

    android:alpha="0.6"

```

```

        android:textStyle="normal" />

<EditText

    android:id="@+id/username"

    android:layout_width="match_parent"

    android:layout_height="55dp"

    android:layout_marginTop="40dp"

    android:background="@drawable/cust
om_edittext"

    android:drawableLeft="@drawable/bas
eline_person_24"

    android:drawablePadding="10dp"

    android:hint="Email"

    android:padding="10dp"

    android:textColor="@color/black"

    android:textColorHighlight="@color/b
lack"

    android:textSize="20dp" />

<EditText

    android:id="@+id/password_toggle"

    android:layout_width="match_parent"

    android:layout_height="65dp"

    android:layout_marginTop="40dp"

    android:background="@drawable/cust
om_edittext"

    android:drawableLeft="@drawable/bas
eline_lock_24"

    android:drawablePadding="20dp"

    android:hint="password"

    android:inputType="textPassword"

    android:padding="20dp"

    android:textColor="@color/black"

    android:textColorHighlight="#5814D1
"

    android:textSize="20dp" />

<Button

    android:id="@+id/loginButton"

```


<pre> android:layout_width="match_parent" android:layout_height="50dp" android:layout_marginTop="10dp" android:backgroundTint="#9C27B0" android:onClick="launchDashboard" android:text="Log In" android:textSize="20dp" app:cornerRadius="20dp" tools:visibility="visible" /> <CheckBox android:layout_width="wrap_content" android:layout_height="wrap_content" android:text="Remember me?" android:layout_centerVertical="true"/> <TextView android:layout_width="wrap_content" android:layout_height="wrap_content" android:text="Or no account?" android:textSize="25dp" android:alpha="0.8" android:layout_centerVertical="true"/> <LinearLayout android:layout_width="wrap_content" android:layout_height="wrap_content" > </LinearLayout> <Button android:layout_width="match_parent" android:layout_height="48dp" android:background="@color/black" android:id="@+id/createAccount" android:height="50dp" </pre>	<pre> android:layout_gravity="center bottom" " android:text="Click here to create account" android:textColor="@color/white" app:cornerRadius="50dp" tools:visibility="visible" /> </LinearLayout> Dashboard Java code package com.example.animalwashuka; import android.content.Intent; import android.os.Bundle; import android.view.View; import android.widget.Button; import androidx.appcompat.app.AppCompatActivity; public class Dashboard extends AppCompatActivity { private Button animalsButton; private Button settingsButton; private Button logoutButton; @Override protected void onCreate(Bundle savedInstanceState) { super.onCreate(savedInstanceState); setContentView(R.layout.activity_dash board); // Initialize buttons animalsButton = findViewById(R.id.animalsID); settingsButton = findViewById(R.id.settings); logoutButton = findViewById(R.id.Logout); // Set click listeners animalsButton.setOnClickListener(ne w View.OnClickListener() { @Override public void onClick(View v) { // Open AnimalsActivity Intent intent = new Intent(Dashboard.this, Animals.class); </pre>	<pre> startActivity(intent); } }); settingsButton.setOnClickListener(new View.OnClickListener() { @Override public void onClick(View v) { // Open SettingsActivity Intent intent = new Intent(Dashboard.this, Settings.class); startActivity(intent); } }); logoutButton.setOnClickListener(new View.OnClickListener() { @Override public void onClick(View v) { // Open MainActivity and clear activity stack Intent intent = new Intent(Dashboard.this, MainActivity.class); intent.addFlags(Intent.FLAG_ACTIVI TY_CLEAR_TOP Intent.FLAG_ACTIVITY_NEW_TAS K Intent.FLAG_ACTIVITY_CLEAR_T ASK); startActivity(intent); finish(); // Optional: Finish this activity } }); }} Dashbord XML <?xml version="1.0" encoding="utf- 8"> <RelativeLayout xmlns:android="http://schemas.android .com/apk/res/android" xmlns:app="http://schemas.android.co m/apk/res-auto" xmlns:tools="http://schemas.android.c om/tools" android:layout_width="match_parent" android:layout_height="match_parent" </pre>
---	---	--

<pre> android:background="@drawable/hd_ geometric_light_lines_pastel_sh_stripe s_warm" android:layout_gravity="center" android:orientation="vertical" tools:context=".Dashboard"> <TextView android:layout_width="match_parent" android:layout_height="50dp" android:text="My Dashboard" android:textSize="30dp" android:textStyle="bold" android:id="@+id/Dashboardtop" android:textColor="#025104" android:layout_alignParentTop="true" android:layout_marginTop="1dp"> </TextView> <ImageView android:layout_width="50dp" android:layout_height="50dp" android:background="@drawable/base line_home_24" android:layout_alignParentLeft="true" android:layout_below="@id/Dashboar dtop" android:layout_marginTop="50dp"> </ImageView> <Button android:id="@+id/animalsID" android:layout_width="match_parent" android:layout_height="wrap_content" android:text="ANIMALS" android:textSize="25dp" android:textColor="@color/black" android:textStyle="italic bold" </pre>	<pre> android:background="@color/teal_200 " android:layout_below="@id/Dashboar dtop" android:layout_marginRight="20dp" android:layout_marginTop="50dp" android:layout_marginLeft="45dp"> </Button> <ImageView android:layout_width="50dp" android:layout_height="50dp" android:background="@drawable/base line_settings_24" android:layout_alignParentLeft="true" android:layout_marginTop="30dp" android:layout_below="@id/animalsI D"> </ImageView> <Button android:layout_width="match_parent" android:layout_height="wrap_content" android:id="@+id/settings" android:text="SETTINGS" android:background="@color/white" android:textColor="@color/black" android:textSize="25dp" android:textStyle="bold italic" android:layout_below="@id/animalsI D" android:layout_marginTop="30dp" android:layout_marginLeft="45dp" android:layout_marginRight="20dp"> </Button> <ImageView </pre>	<pre> android:layout_width="50dp" android:layout_height="50dp" android:layout_below="@id/settings" android:layout_marginTop="30dp" android:layout_alignParentLeft="true" android:background="@drawable/base line_lock_24" > </ImageView> <Button android:layout_width="match_parent" android:layout_height="wrap_content" android:id="@+id/Logout" android:background="@color/teal_200 " android:text="LOG OUT" android:textStyle="italic bold" android:textColor="@color/black" android:textSize="25dp" android:layout_below="@id/settings" android:layout_marginTop="30dp" android:layout_marginLeft="45dp" android:layout_marginRight="20dp"> </Button> </RelativeLayout> Animal Java code package com.example.animalwashuka; import android.content.Intent; import android.os.Bundle; import android.view.View; import android.widget.Button; </pre>
--	--	---

<pre> import androidx.appcompat.app.AppCompatActivity; public class Animals extends AppCompatActivity { private Button cattleButton, sheepButton, goatsButton, chickenButton, addAnimalButton; @Override protected void onCreate(Bundle savedInstanceState) { super.onCreate(savedInstanceState); setContentView(R.layout.activity_ani mals); // Initialize buttons cattleButton = findViewById(R.id.cattle); sheepButton = findViewById(R.id.sheep); goatsButton = findViewById(R.id.goats); chickenButton = findViewById(R.id.chicken); addAnimalButton = findViewById(R.id.addAnimal); // Set onClick listeners for each button cattleButton.setOnClickListener(new View.OnClickListener() { @Override public void onClick(View v) { // Navigate to the Cattle activity Intent intent = new Intent(Animals.this, Cattle.class); startActivity(intent); } }); // Implement sheep, goats, and chicken buttons similarly if needed sheepButton.setOnClickListener(new View.OnClickListener() { @Override </pre>	<pre> public void onClick(View v) { // Code to navigate to Sheep activity Intent intent = new Intent(Animals.this, Sheep.class); startActivity(intent); } }); goatsButton.setOnClickListener(new View.OnClickListener() { @Override public void onClick(View v) { // Code to navigate to Goats activity Intent intent = new Intent(Animals.this, Goats.class); startActivity(intent); } }); chickenButton.setOnClickListener(ne w View.OnClickListener() { @Override public void onClick(View v) { // Code to navigate to Chicken activity Intent intent = new Intent(Animals.this, Chicken.class); startActivity(intent); } }); addAnimalButton.setOnClickListener(new View.OnClickListener() { @Override public void onClick(View v) { // Code to navigate to Add Animal activity Intent intent = new Intent(Animals.this, AddAnimal.class); startActivity(intent); } }); }} </pre>	<pre> Animals XML <?xml version="1.0" encoding="utf- 8"?> <RelativeLayout xmlns:android="http://schemas.android .com/apk/res/android" xmlns:app="http://schemas.android.co m/apk/res-auto" xmlns:tools="http://schemas.android.c om/tools" android:layout_width="match_parent" android:layout_height="match_parent" android:orientation="vertical" android:background="@drawable/pay ment" tools:context=".Animals"> <TextView android:layout_width="match_parent" android:layout_height="50dp" android:text="ANIMALS" android:id="@+id/animaltypes" android:textStyle="italic bold" android:textColor="@color/black" android:textSize="30dp" > </TextView> <Button android:layout_width="match_parent" android:layout_height="wrap_content" android:id="@+id/cattle" android:text="Cattle" android:textSize="20dp" android:textStyle="italic bold" android:textColor="@color/white" android:layout_marginTop="50dp" android:layout_below="@id/animaltyp es" </pre>
---	---	---

>	android:textSize="20dp"	import
</Button>	android:id="@+id/chicken">	com.amazonaws.services.dynamodbv2.
<Button	</Button>	document.DynamoDB;
android:layout_width="match_parent"	<Button	import
android:layout_height="wrap_content"	android:layout_width="wrap_content"	com.amazonaws.services.dynamodbv2.
android:id="@+id/sheep"	android:layout_height="wrap_content"	document.Item;
android:layout_below="@id/cattle"	android:layout_below="@id/chicken"	import
android:layout_marginTop="20dp"	android:text="ADD ANIMAL"	com.amazonaws.services.dynamodbv2.
android:text="Sheep"	android:textColor="@color/black"	document.Spec.ScanSpec;
android:textSize="20dp"	android:textStyle="normal bold"	import java.util.Iterator;
android:textColor="@color/white"	android:textSize="20dp"	public class Cattle extends
android:textStyle="italic bold"	android:layout_marginTop="20dp"	AppCompatActivity {
>	android:id="@+id/addAnimal"	private LinearLayout
</Button>	android:layout_centerHorizontal="true"	cowsContainer;
<Button	</Button>	private Button addCattleButton;
android:layout_width="match_parent"	</RelativeLayout>	private DynamoDB dynamoDB;
android:layout_height="wrap_content"		@Override
android:id="@+id/goats"		protected void onCreate(Bundle
android:layout_below="@id/sheep"		savedInstanceState) {
android:layout_marginTop="20dp"		super.onCreate(savedInstanceState);
android:text="Goats"		
android:textColor="@color/white"		setContentView(R.layout.activity_cattl
android:textStyle="italic bold"		e);
android:textSize="20dp">		cowsContainer =
</Button>		findViewById(R.id.cowsContainer);
<Button		addCattleButton =
android:layout_width="match_parent"		findViewById(R.id.AddCattle);
android:layout_height="wrap_content"		// Initialize AWS Mobile Client
android:layout_below="@id/goats"		
android:layout_marginTop="20dp"		AWSMobileClient.getInstance().initiali
android:text="Chicken"		ze(this).execute();
android:textColor="@color/white"		// Configure Cognito credentials
android:textStyle="italic bold"		provider with region
		CognitoCachingCredentialsProvider
		credentialsProvider = new
		CognitoCachingCredentialsProvider(
		getApplicationContext(),
		"YOUR_COGNITO_IDENTITY_PO
		OL_ID", // Replace with your Cognito
		Identity Pool ID
		Regions.US_EAST_1 //
		Replace with your DynamoDB region,
		e.g., US_EAST_1
);
		// Create a DynamoDB client with
		region configuration

```

        AmazonDynamoDBClient
        dbClient = new
        AmazonDynamoDBClient(credentials
        Provider);

        dbClient.setRegion(com.amazonaws.re
        gions.Region.getRegion(Regions.US_
        EAST_1)); // Set region explicitly

        dynamoDB = new
        DynamoDB(dbClient);

        // Retrieve and display cows from
        DynamoDB

        displayCows();

        // Add Cattle Button functionality

        addCattleButton.setOnClickListener(n
        ew View.OnClickListener() {

            @Override

            public void onClick(View v) {

                Intent intent = new
                Intent(Cattle.this, plus.class);

                startActivity(intent);

            } });

        private void displayCows() {

            // Reference the DynamoDB table

            Table table =
            dynamoDB.getTable("CowData");

            // Scan the table to fetch all cow
            records

            new Thread() -> {

                try {

                    ScanSpec scanSpec = new
                    ScanSpec();

                    Iterator<Item> iterator =
                    table.scan(scanSpec).iterator();

                    runOnUiThread() -> {

                        cowsContainer.removeAllViews(); //
                        Clear existing views

                        while (iterator.hasNext())

                        {

                            Item item =
                            iterator.next();

                            // Get cow details and
                            create a Cow object

                            Cow cow = new Cow(

                                item.getString("CowID"),

                                item.getString("Name"),

```

```

                                item.getInt("Age"),

                                item.getString("Breed")

                                );

                                // Create a button for
                                each cow

                                Button cowButton =
                                new Button(Cattle.this);

                                cowButton.setText(cow.getName());

                                cowButton.setTextSize(20);

                                cowButton.setPadding(10, 10, 10, 10);

                                cowButton.setOnClickListener(v -> {

                                    // Start activity for
                                    detailed cow view (SpecificCow)

                                    Intent intent = new
                                    Intent(Cattle.this, SpecificCow.class);

                                    intent.putExtra("cowId",
                                    cow.getCowID());

                                    startActivity(intent);

                                });

                                // Add button to the
                                container

                                cowsContainer.addView(cowButton);

                                }

                                }); catch (Exception e) {

                                    e.printStackTrace();

                                }

                                }).start();

                                }}

            Cattle XML

            <?xml version="1.0" encoding="utf-
            8"?>

            <ScrollView
            xmlns:android="http://schemas.android
            .com/apk/res/android"

            xmlns:app="http://schemas.android.co
            m/apk/res-auto"

            xmlns:tools="http://schemas.android.c
            om/tools"

```

```

            android:layout_width="match_parent"

            android:layout_height="match_parent"

            android:background="@drawable/hd_
            geometric_light_lines_pastel_sh_stripe
            s_warm"

            tools:context=".Cattle">

            <LinearLayout

            android:layout_width="match_parent"

            android:layout_height="wrap_content"

            android:orientation="vertical"

            android:padding="16dp">

            <TextView

            android:id="@+id/cattleHeader"

            android:layout_width="match_parent"

            android:layout_height="wrap_content"

            android:text="My Cattle"

            android:textSize="25sp"

            android:textStyle="bold"

            android:textColor="#025104"

            android:layout_marginTop="10dp" />

            <LinearLayout

            android:id="@+id/cowsContainer"

            android:layout_width="match_parent"

            android:layout_height="wrap_content"

            android:orientation="vertical"

            android:layout_marginTop="16dp" />

            <!-- Add Cow Button -->

            <Button

            android:id="@+id/AddCattle"

            android:layout_width="match_parent"

            android:layout_height="wrap_content"

```

```

android:background="@color/teal_200"
"

    android:text="Add Cow"

android:textColor="@color/black"

    android:textSize="20sp"

    android:textStyle="italic|bold"

android:layout_marginTop="16dp" />

</LinearLayout>

</ScrollView>

Specific Cow Java Code

package com.example.animalwashuka;

import android.os.Bundle;

import android.widget.Button;

import android.widget.TextView;

import android.widget.Toast;

import
androidx.appcompat.app.AppCompatActivity;

import
com.amazonaws.mobile.client.AWSMobileClient;

import
com.amazonaws.auth.CognitoCaching
CredentialsProvider;

import
com.amazonaws.regions.Regions;

import
com.amazonaws.services.dynamodbv2.
AmazonDynamoDBClient;

//import
com.amazonaws.services.dynamodb.
R;

import
com.amazonaws.services.dynamodb2.
document.DynamoDB;

//import
com.amazonaws.services.dynamodbv2.
document.Table;

import
com.amazonaws.services.dynamodbv2.
document.spec.GetItemSpec;

import
com.amazonaws.services.dynamodbv2.
document.util.ValueMap;

import
com.amazonaws.services.dynamodbv2.
model.GetItemRequest;

import
com.amazonaws.services.dynamodbv2.
model.GetItemResult;

```

```

import java.text.SimpleDateFormat;

import java.util.Date;

import java.util.Locale;

public class SpecificCow extends
AppCompatActivity {

    private TextView cowName,
healthIndicator, timestamp,
temperature, activityMagnitude, status;

    private Button deleteCowButton;

    @Override

    protected void onCreate(Bundle
savedInstanceState) {

        super.onCreate(savedInstanceState);

        setContentView(R.layout.activity_spec
ific_cow);

        // Initialize UI components

        cowName =
findViewById(R.id.cowName);

        healthIndicator =
findViewById(R.id.healthIndicator);

        timestamp =
findViewById(R.id.timestamp);

        temperature =
findViewById(R.id.temperature);

        activityMagnitude =
findViewById(R.id.activityMagnitude)
;

        status =
findViewById(R.id.status);

        deleteCowButton =
findViewById(R.id.deleteCowButton);

        // Fetch data from DynamoDB
(assuming cowId is passed in an Intent)

        String cowId =
getIntent().getStringExtra("COW_ID")
;

        // Fetch details from DynamoDB

        fetchCowDetails(cowId);

        // Set timestamp to current time
from phone

        setTimestamp();

        // Handle delete cow button click
(optional)

        deleteCowButton.setOnClickListener(
view -> {

```

```

// Logic to delete the cow from
DynamoDB

        deleteCow(cowId);

    });

}

    private void fetchCowDetails(String
cowId) {

        // Initialize AWS DynamoDB
client

        DynamoDB dynamoDB = new
DynamoDB(AWSMobileClient.getInst
ance());

        Table table =
dynamoDB.getTable("Cattle");

        // Query the DynamoDB table for
the specific cow using its cowId

        GetItemSpec spec = new
GetItemSpec().withPrimaryKey("cowI
D", cowId);

        try {

            // Fetch the cow data from
DynamoDB

            com.amazonaws.services.dynamodbv2.
document.Item item =
table.getItem(spec);

            if (item != null) {

                // Populate UI elements with
the retrieved data

                cowName.setText(item.getString("cow
Name"));

                healthIndicator.setText(item.getString(
"healthStatus"));

                temperature.setText("Temperature: " +
item.getString("temperature"));

                activityMagnitude.setText("Activity
Magnitude: " +
item.getString("activityMagnitude"));

                status.setText("Status: " +
item.getString("status"));

            } else {

                // Show a message if the cow
data is not found

                Toast.makeText(this, "Cow
data not found",
Toast.LENGTH_SHORT).show();

            }

        } catch (Exception e) {

            // Handle any errors

```

```

        Toast.makeText(this, "Error
        fetching data: " + e.getMessage(),
        Toast.LENGTH_SHORT).show();

    }

}

private void setTimestamp() {

    // Set the timestamp to the current
    date and time of the phone

    SimpleDateFormat sdf = new
    SimpleDateFormat("hh:mm a, dd
    MMM yyyy", Locale.getDefault());

    String currentTime =
    sdf.format(new Date());

    timestamp.setText(currentTime);

}

private void deleteCow(String
cowId) {

    // Logic to delete the cow from
    DynamoDB

    try {

        // Assume that the delete
        functionality is implemented

        // For example, using a
        DynamoDB delete request

        DynamoDB dynamoDB = new
        DynamoDB(AWSMobileClient.getInst
        ance());

        Table table =
        dynamoDB.getTable("Cattle");

        table.deleteItem("cowID",
        cowId);

        Toast.makeText(this, "Cow
        deleted successfully",
        Toast.LENGTH_SHORT).show();

    } catch (Exception e) {

        // Handle any errors during
        deletion

        Toast.makeText(this, "Error
        deleting cow: " + e.getMessage(),
        Toast.LENGTH_SHORT).show();

    } }
}

```

SpecificCow XML

```

<?xml version="1.0" encoding="utf-
8"?>

<ScrollView
xmlns:android="http://schemas.android
.com/apk/res/android"

xmlns:tools="http://schemas.android.c
om/tools"

android:layout_width="match_parent"

```

```

android:layout_height="match_parent"

android:background="@drawable/hd_
geometric_light_lines_pastel_sh_stripe
s_warm"

tools:context=".SpecificCow">

<LinearLayout

android:layout_width="match_parent"

android:layout_height="wrap_content"

    android:orientation="vertical"

    android:padding="16dp">

    <!-- Cow Name -->

    <TextView

        android:id="@+id/cowName"

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:text=""

        android:textSize="22sp"

        android:textStyle="bold"

        android:textColor="#000000"

        android:layout_marginBottom="8dp"
    />

    <!-- Health Indicator -->

    <TextView

        android:id="@+id/healthIndicator"

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:text=""

        android:textSize="20sp"

        android:textStyle="bold"

        android:textColor="#4CAF50"

        android:layout_marginBottom="16dp"
    />

    <!-- Timestamp -->

    <TextView

        android:id="@+id/timestamp"

```

```

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:text=""

        android:textSize="18sp"

        android:textColor="#000000"

        android:layout_marginBottom="8dp"
    />

    <!-- Temperature -->

    <TextView

        android:id="@+id/temperature"

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:text=""

        android:textSize="18sp"

        android:textColor="#000000"

        android:layout_marginBottom="8dp"
    />

    <!-- Activity Magnitude -->

    <TextView

        android:id="@+id/activityMagnitude"

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:text=""

        android:textSize="18sp"

        android:textColor="#000000"

        android:layout_marginBottom="8dp"
    />

    <!-- Status -->

    <TextView

        android:id="@+id/status"

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:text=""

        android:textSize="18sp"

        android:textColor="#000000"

```


<pre> emailEditText.requestFocus(); return; } if (TextUtils.isEmpty(password) password.length() < 6) { passwordEditText.setError("Password must be at least 6 characters"); passwordEditText.requestFocus(); return; } // Show progress bar progressBar.setVisibility(View.VISIBL E); // Check if the email is already in use mAuth.fetchSignInMethodsForEmail(e mail).addOnCompleteListener(new OnCompleteListener<com.google.fireb ase.auth.SignInMethodQueryResult>() { @Override public void onComplete(@NonNull Task<com.google.firebase.auth.SignIn MethodQueryResult> task) { if (task.isSuccessful()) { boolean emailExists = !task.getResult().getSignInMethods().i sEmpty(); if (emailExists) { // Email already in use progressBar.setVisibility(View.GONE) ; emailEditText.setError("This email is already registered"); emailEditText.requestFocus(); } else { // Proceed with registration registerWithEmailAndPassword(email, password); } } else { // Handle error </pre>	<pre> progressBar.setVisibility(View.GONE) ; Toast.makeText(register.this, "Error checking email: " + task.getException().getMessage(), Toast.LENGTH_SHORT).show(); } } }); } private void registerWithEmailAndPassword(String email, String password) { mAuth.createUserWithEmailAndPass word(email, password) .addOnCompleteListener(register.this, new OnCompleteListener<AuthResult>() { @Override public void onComplete(@NonNull Task<AuthResult> task) { progressBar.setVisibility(View.GONE) ; if (task.isSuccessful()) { // Send verification email FirebaseUser user = mAuth.getCurrentUser(); if (user != null) { user.sendEmailVerification() .addOnCompleteListener(new OnCompleteListener<Void>() { @Override public void onComplete(@NonNull Task<Void> task) { if (task.isSuccessful()) { Toast.makeText(register.this, "Registration successful. Please check your email to verify your account.", Toast.LENGTH_LONG).show(); // Redirect to Dashboard Activity Intent intent = new Intent(register.this, Dashboard.class); startActivity(intent); </pre>	<pre> finish(); } else { Toast.makeText(register.this, "Failed to send verification email: " + task.getException().getMessage(), Toast.LENGTH_SHORT).show(); } } }); } } else { // Registration failed Toast.makeText(register.this, "Registration failed: " + task.getException().getMessage(), Toast.LENGTH_SHORT).show(); } } }); } } Register XML <ScrollView xmlns:android="http://schemas.android .com/apk/res/android" xmlns:app="http://schemas.android.co m/apk/res-auto" xmlns:tools="http://schemas.android.c om/tools" android:layout_width="match_parent" android:layout_height="match_parent" android:background="@drawable/pay ment" tools:context=".register"> <LinearLayout android:layout_width="match_parent" android:layout_height="wrap_content" android:orientation="vertical" android:padding="20dp"> <!-- ProgressBar for registration -- > <ProgressBar </pre>
---	---	---

android:id="@+id/progressBar"	android:id="@+id/phonenumber"	android:id="@+id/password_toggle"
android:layout_width="wrap_content"	android:layout_width="match_parent"	android:layout_width="match_parent"
android:layout_height="wrap_content"	android:layout_height="48dp"	android:layout_height="48dp"
android:layout_gravity="center"	android:layout_marginTop="20dp"	android:layout_marginTop="20dp"
android:visibility="gone"	android:background="@drawable/custom_edittext"	android:background="@drawable/custom_edittext"
android:indeterminate="true"		
</>	android:drawableLeft="@drawable/baseline_contact_phone_24"	android:drawableLeft="@drawable/baseline_lock_24"
<TextView		
android:layout_width="match_parent"	android:drawablePadding="10dp"	android:drawablePadding="20dp"
android:layout_height="wrap_content"	android:hint="Phone Number"	android:hint="Password"
android:text="Register"	android:padding="10dp"	android:inputType="textPassword"
android:textAlignment="center"	android:textColor="@color/black"	android:padding="20dp"
android:textSize="32sp"	android:textColorHighlight="@color/black"	android:textColor="@color/black"
android:textColor="@color/black" />	android:textSize="20sp" />	android:textColorHighlight="#5814D1"
<!-- Full Name Input field -->	<!-- Email Input Field -->	
<EditText	<EditText	android:textSize="20sp" />
android:id="@+id/fullname"	android:id="@+id/emailRegistration"	<!-- Register Button -->
android:layout_width="match_parent"	android:layout_width="match_parent"	<Button
android:layout_height="48dp"	android:layout_height="48dp"	android:id="@+id/registerBtn"
android:layout_marginTop="20dp"	android:layout_marginTop="20dp"	android:layout_width="match_parent"
android:background="@drawable/custom_edittext"	android:background="@drawable/custom_edittext"	android:layout_height="wrap_content"
android:drawableLeft="@drawable/baseline_person_24"	android:drawableLeft="@drawable/baseline_alternate_email_24"	android:layout_marginTop="20dp"
android:drawablePadding="10dp"	android:drawablePadding="10dp"	android:background="@color/purple_700"
android:hint="Full name"	android:hint="Email"	android:padding="10dp"
android:padding="10dp"	android:padding="10dp"	android:text="Register"
android:textColor="@color/black"	android:textColor="@color/black"	android:textColor="@color/white"
android:textColorHighlight="@color/black"	android:textColorHighlight="@color/black"	android:textSize="20sp"
android:textSize="20sp" />	android:textSize="20sp" />	app:backgroundTint="@null"
<!-- Phone Number Input Field -->	<!-- Password Input Field -->	<!-- Sign In Button -->
<EditText	<EditText	<Button
		android:id="@+id/signInButton"
		android:layout_width="wrap_content"

```

        android:layout_height="wrap_content"

        android:layout_marginTop="20dp"
            android:text="Sign In"

        android:textColor="@color/black"
            android:textSize="18sp" />
    </LinearLayout>
</ScrollView>

```

Contact support Java Code

```

package com.example.animalwashuka;

import android.content.Intent;

import android.net.Uri;

import android.os.Bundle;

import android.widget.Button;

import android.widget.Toast;

import
androidx.appcompat.app.AppCompatActivity;

public class ContactSupport extends
AppCompatActivity {

    private static final String EMAIL =
"riniwachuka2002@gmail.com";

    private static final String
PHONE_NUMBER =
"+254759787739";

    @Override

    protected void onCreate(Bundle
savedInstanceState) {

super.onCreate(savedInstanceState);

setContentView(R.layout.activity_cont
act_support);

    Button emailSupportButton =
findViewById(R.id.emailSupport);

    Button callSupportButton =
findViewById(R.id.callSupport);

    // Email support button click
listener

    emailSupportButton.setOnClickListener(
r(v -> sendEmail());

    // Call support button click
listener

    callSupportButton.setOnClickListener(
v -> makePhoneCall());

```

```

    }

    private void sendEmail() {

        Intent emailIntent = new
Intent(Intent.ACTION_SENDTO);

        emailIntent.setData(Uri.parse("mailto:"
+ EMAIL)); // Only email apps should
handle this

        emailIntent.putExtra(Intent.EXTRA_S
UBJECT, "Support Request");

        emailIntent.putExtra(Intent.EXTRA_T
EXT, "Please describe your issue:");

        // Check if there's an app that can
handle this intent

        if
(emailIntent.resolveActivity(getPackag
eManager()) != null) {

            startActivity(emailIntent);

        } else {

            Toast.makeText(this, "No email
app found",
Toast.LENGTH_SHORT).show();

        } }

    private void makePhoneCall() {

        Intent callIntent = new
Intent(Intent.ACTION_DIAL);

        callIntent.setData(Uri.parse("tel:"
+ PHONE_NUMBER)); // The tel URI
scheme

        startActivity(callIntent);

    } }

```

Contact Support XML

```

<ScrollView
xmlns:android="http://schemas.android
.com/apk/res/android"

    android:layout_width="match_parent"

    android:layout_height="match_parent"

    android:padding="16dp"

    android:background="@drawable/hd_
geometric_light_lines_pastel_sh_stripe
s_warm">

    <LinearLayout

        android:layout_width="match_parent"

        android:layout_height="wrap_content"

        android:orientation="vertical">

```

```

        <TextView

            android:layout_width="wrap_content"

            android:layout_height="wrap_content"

            android:text="Contact Support"

            android:textSize="24sp"

            android:textStyle="bold"

            android:paddingBottom="16dp"/>

        <TextView

            android:layout_width="wrap_content"

            android:layout_height="wrap_content"

            android:text="If you need
assistance, you can reach out to us via
email or call:"

            android:textSize="16sp"

            android:alpha="0.7"

            android:paddingBottom="16dp"/>

        <!-- Email Option -->

        <Button

            android:id="@+id/emailSupport"

            android:layout_width="match_parent"

            android:layout_height="50dp"

            android:text="Email Support"

            android:backgroundTint="@color/mat
erial_dynamic_neutral30"

            android:textColor="@color/black"

            android:layout_marginLeft="10dp"

            android:layout_marginRight="10dp"

            android:background="@color/purple_7
00"

            android:textSize="13dp"

            android:padding="12dp"/>

        <!-- Phone Call Option -->

        <Button

            android:id="@+id/callSupport"

            android:layout_width="match_parent"

```

```

        android:layout_height="50dp"

        android:layout_marginLeft="10dp"

        android:layout_marginRight="10dp"

        android:text="Call Support"

        android:backgroundTint="@color/material_dynamic_neutral30"

        android:background="@color/purple_700"

        android:textSize="13dp"

        android:textColor="@color/black"

        android:padding="12dp"

        android:layout_marginTop="16dp"/>
    </LinearLayout>
</ScrollView>

```

Settings Java Code

```

package com.example.animalwashuka;

import android.content.Intent;
import android.os.Bundle;
import android.util.Log;
import android.widget.TextView;

import androidx.appcompat.app.AppCompatActivity;

public class Settings extends AppCompatActivity {

    private static final String TAG = "SettingsActivity";

    @Override
    protected void onCreate(Bundle savedInstanceState) {

        super.onCreate(savedInstanceState);

        setContentView(R.layout.activity_settings);

        try {

            // User Account Settings

            TextView changeEmail = findViewById(R.id.changeEmail);

            changeEmail.setOnClickListener(v -> {

```

```

                startActivity(new Intent(Settings.this, ChangeEmail.class));

            });

            TextView changePassword = findViewById(R.id.changePassword);

            changePassword.setOnClickListener(v -> {

                startActivity(new Intent(Settings.this, ChangePassword.class));

            });

            TextView updateProfile = findViewById(R.id.updateProfile);

            updateProfile.setOnClickListener(v -> {

                startActivity(new Intent(Settings.this, UpdateProfile.class));

            });

            // Notifications Settings

            TextView enableNotifications = findViewById(R.id.enableNotifications);

            enableNotifications.setOnClickListener(v -> {

                startActivity(new Intent(Settings.this, EnableNotifications.class));

            });

            // Data Settings

            TextView dataSyncFrequency = findViewById(R.id.dataSyncFrequency);

            dataSyncFrequency.setOnClickListener(v -> {

                showDataSyncOptions();

            });

            // Support & Feedback

            TextView contactSupport = findViewById(R.id.contactSupport);

            contactSupport.setOnClickListener(v -> {

                startActivity(new Intent(Settings.this, ContactSupport.class));

```

```

            });

            } catch (Exception e) {

                Log.e(TAG, "Error initializing settings activity", e);

            }

        }

        private void showDataSyncOptions() {

            String[] options = {"Real-time", "Every 1 second"};

            new android.app.AlertDialog.Builder(this)

                .setTitle("Select Data Sync Frequency")

                .setItems(options, (dialog, which) -> {

                    if (which == 0) {

                        setDataSyncFrequency(0); // Real-time

                    } else if (which == 1) {

                        setDataSyncFrequency(1000); // Every 1 second

                    }

                })

                .show();

        }

        private void setDataSyncFrequency(int intervalMs) {

            android.content.SharedPreferences preferences = getSharedPreferences("settings", MODE_PRIVATE);

            preferences.edit().putInt("dataSyncFrequency", intervalMs).apply();

            // Apply the sync frequency to your data sync service

        }

    }

```

Settings XML

```

<ScrollView
    xmlns:android="http://schemas.android.com/apk/res/android"

    android:layout_width="match_parent"

    android:layout_height="match_parent"

    android:padding="16dp"

```

<pre> android:background="@drawable/hd_ geometric_light_lines_pastel_sh_stripe s_warm"> <LinearLayout android:layout_width="match_parent" android:layout_height="wrap_content" android:orientation="vertical"> <!-- User Account Settings --> <TextView android:layout_width="wrap_content" android:layout_height="48dp" android:text="User Account Settings" android:textSize="18sp" android:textStyle="bold" android:paddingTop="16dp" android:paddingBottom="8dp"/> <View android:layout_width="match_parent" android:layout_height="1dp" /> <TextView android:id="@+id/changeEmail" android:layout_width="match_parent" android:layout_height="48dp" android:text="Change Email" android:paddingVertical="12dp" android:clickable="true" android:focusable="true" android:background="?attr/selectableIt emBackground"/> <TextView android:id="@+id/changePassword" android:layout_width="match_parent" android:layout_height="48dp" </pre>	<pre> android:text="Change Password" android:paddingVertical="12dp" android:clickable="true" android:focusable="true" android:background="?attr/selectableIt emBackground"/> <TextView android:id="@+id/updateProfile" android:layout_width="match_parent" android:layout_height="48dp" android:text="Update Profile" android:paddingVertical="12dp" android:clickable="true" android:focusable="true" android:background="?attr/selectableIt emBackground"/> <!-- Notifications Settings --> <TextView android:layout_width="wrap_content" android:layout_height="wrap_content" android:text="Notifications Settings" android:textSize="18sp" android:textStyle="bold" android:paddingTop="16dp" android:paddingBottom="8dp"/> <View android:layout_width="match_parent" android:layout_height="1dp" <TextView android:background="@color/material _dynamic_neutral30"/> <TextView android:id="@+id/enableNotifications " android:layout_width="match_parent" </pre>	<pre> android:layout_height="48dp" android:text="Enable/Disable Notifications" android:paddingVertical="12dp" android:clickable="true" android:focusable="true" android:background="?attr/selectableIt emBackground"/> <!-- Data Settings --> <TextView android:layout_width="wrap_content" android:layout_height="wrap_content" android:text="Data Settings" android:textSize="18sp" android:textStyle="bold" android:paddingTop="16dp" android:paddingBottom="8dp"/> <View android:layout_width="match_parent" android:layout_height="1dp" /> <TextView android:id="@+id/dataSyncFrequency " android:layout_width="match_parent" android:layout_height="48dp" android:text="Data Sync Frequency" android:paddingVertical="12dp" android:clickable="true" android:focusable="true" android:background="?attr/selectableIt emBackground"/> <!-- Support & Feedback --> <TextView android:layout_width="wrap_content" android:layout_height="48dp" </pre>
--	--	---

```

        android:text="Support and
Feedback"
        android:textSize="18sp"
        android:textStyle="bold"
        android:paddingTop="16dp"

        android:paddingBottom="8dp"/>
        <View
        android:layout_width="match_parent"
        android:layout_height="1dp"

```

```

        android:background="@color/material
_dynamic_neutral30"/>
        <TextView
        android:id="@+id/contactSupport"

        android:layout_width="match_parent"
        android:layout_height="48dp"
        android:text="Contact Support"

        android:paddingVertical="12dp"

```

```

        android:clickable="true"
        android:focusable="true"

        android:background="?attr/selectableIt
emBackground"/>
        </LinearLayout>
    </ScrollView>

```