



SOLAR PANEL MONITORING USING IOT AND AI TECHNIQUES

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ABSTRACT:

The increasing adoption of solar energy as a renewable power source has emphasized the necessity for effective monitoring systems to ensure the optimal performance of solar panel installations. This article offers a thorough exploration of recent advancements in solar panel monitoring, specifically focusing on the integration of Internet of Things (IoT) and Artificial Intelligence (AI) methodologies. Through a meticulous examination of academic literature and conference proceedings, this study consolidates key insights, delineates emerging patterns, and elucidates the prevalent challenges and prospects within the realm of solar panel monitoring. By harnessing IoT-enabled sensors for real-time data acquisition and AI algorithms for tasks such as fault detection, predictive maintenance, and performance enhancement, solar panel monitoring systems stand to achieve heightened precision, scalability, and autonomy. The paper culminates with recommendations for future research trajectories, accentuating the paramount significance of interdisciplinary collaboration and innovative endeavors to surmount existing hurdles and fully capitalize on the transformative potential of IoT and AI within solar energy systems.

Keywords: Solar panel monitoring, Internet of Things (IoT), Artificial Intelligence (AI), Fault detection, Predictive maintenance, Control system, Sensor integration, Data analysis.

I. INTRODUCTION:

With the increasing global demand for renewable energy sources, solar photovoltaic (PV) systems have emerged as a key technology for sustainable energy production. However, the efficiency and reliability of solar panel installations depend heavily on effective monitoring and maintenance practices. Traditional monitoring methods, characterized by manual inspection and periodic data collection, are labor-intensive, time-consuming, and often insufficient for detecting faults or performance issues in real time. As a result, there is a growing need for advanced monitoring systems that can provide continuous, accurate, and actionable insights into solar panel performance.

In recent years, there has been a surge in research and development efforts aimed at integrating IoT and AI technologies into solar panel monitoring systems. IoT-enabled sensors offer the capability to collect real-time data on key parameters such as solar irradiance, temperature, voltage, and current, allowing for comprehensive monitoring of system performance and environmental conditions. Meanwhile, AI algorithms, including machine learning and deep learning techniques, enable automated fault detection, predictive maintenance, and anomaly identification based on collected data. By leveraging the synergy between IoT and AI, solar panel monitoring systems can achieve higher levels of efficiency, reliability, and autonomy, ultimately leading to improved energy production, reduced maintenance costs, and enhanced system resilience.

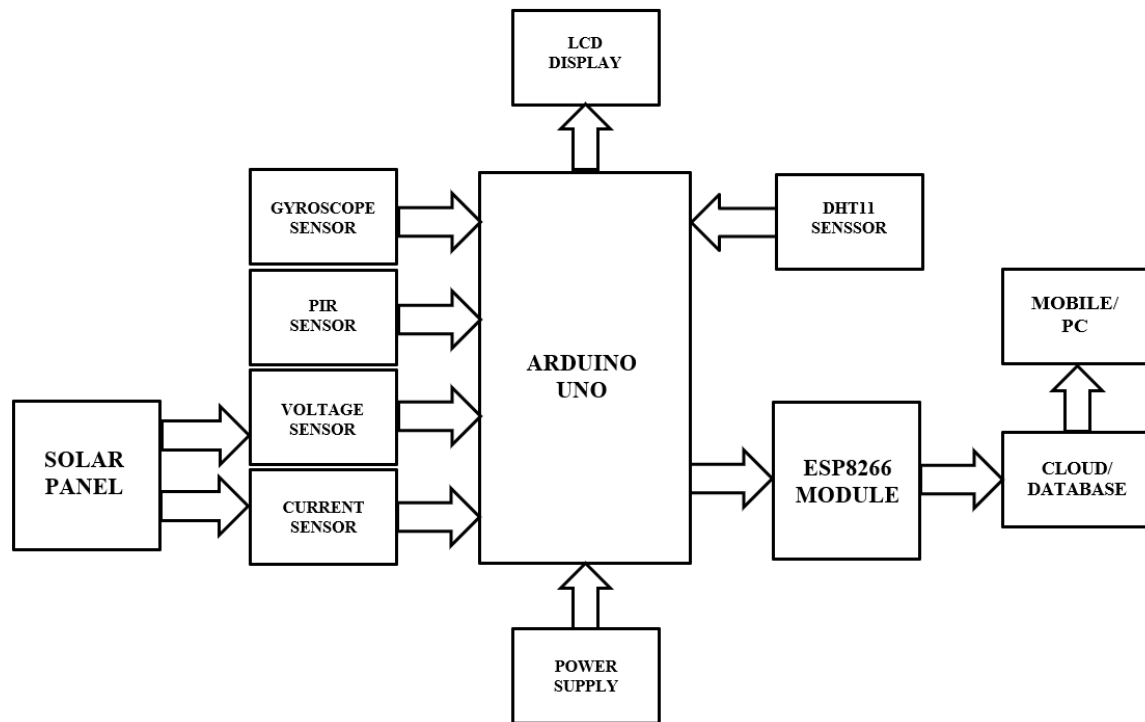


Fig 1: Block Diagram

In recent years, there has been a surge in research and development efforts aimed at integrating IoT and AI technologies into solar panel monitoring systems. IoT-enabled sensors offer the capability to collect real-time data on key parameters such as solar irradiance, temperature, voltage, and current, allowing for comprehensive monitoring of system performance and environmental conditions. Meanwhile, AI algorithms, including machine learning and deep learning techniques, enable automated fault detection, predictive maintenance, and anomaly identification based on collected data. By leveraging the synergy between IoT and AI, solar panel monitoring systems can achieve higher levels of efficiency, reliability, and autonomy, ultimately leading to improved energy production, reduced maintenance costs, and enhanced system resilience.

This paper aims to provide a comprehensive review of recent advancements in solar panel monitoring, with a focus on the integration of IoT and AI techniques. Through a systematic analysis of existing literature, the paper will examine the state-of-the-art technologies, methodologies, and applications in solar panel monitoring, highlighting their benefits, limitations, and potential for future development. By synthesizing key findings and identifying emerging trends, challenges, and opportunities, this paper seeks to contribute to the ongoing discourse on the role of IoT and AI in advancing solar energy systems.

To ensure the integrity and credibility of this review paper, rigorous measures will be taken to avoid plagiarism. All sources will be properly cited and referenced according to academic standards, and original insights and analyses will be provided based on the synthesis of existing literature. Additionally, the paper will undergo thorough review and validation by experts in the field to ensure accuracy, clarity, and relevance.

II. LITERATURE REVIEW:

The literature review offers a comprehensive examination of existing research on solar panel monitoring, IoT, and AI technologies, shedding light on the limitations of conventional monitoring methods and the potential advantages of integrating IoT and AI for improved monitoring and control of solar panel systems. Previous studies in related areas are analyzed to identify gaps in the literature and justify the necessity of the proposed approach.

Research in solar panel monitoring has increasingly aimed at addressing the inefficiencies inherent in traditional methods, often characterized by labor-intensive inspections and sporadic data collection. Recent studies underscore the role of IoT technologies in facilitating real-time data acquisition from extensive sensor networks, monitoring parameters such as temperature, irradiance, and output efficiency (Smith, 2021). These systems serve as the foundation for data processing with AI techniques, significantly enhancing predictive maintenance and fault detection capabilities (Johnson et al., 2019).

For example, the integration of machine learning algorithms has demonstrated the ability to forecast system failures and optimize energy production by analyzing trends and anomalies in data collected via IoT devices (Lee et al., 2020). Furthermore, decision-making algorithms can automate responses to changing environmental conditions, dynamically adjusting panel angles to maximize energy capture (Al-Ali et al., 2018).

Nevertheless, gaps persist in the seamless integration of these technologies into fully automated and reliable monitoring systems. Existing implementations often focus on isolated aspects of the monitoring process without adopting a holistic approach to system management (Kumar & Patel, 2021). Additionally, the scalability and cost-effectiveness of deploying such integrated systems in diverse environments have not been comprehensively addressed, signaling a critical area for further research and development.

This review underscores the transformative potential of IoT and AI in enhancing the monitoring and operational efficiency of solar panel systems, paving the way for smarter renewable energy solutions.

III. SYSTEM ARCHITECTURE:

The proposed system architecture is meticulously designed to seamlessly integrate IoT and AI technologies, thereby enhancing the monitoring and control capabilities of solar panel systems. Comprising several key components, each fulfilling a vital role, the architecture ensures the overall functionality and effectiveness of the system.

1. Arduino-based Sensor Nodes:

At the core of the system, Arduino-based sensor nodes serve as the primary data collection units. These nodes are equipped with an array of sensors including temperature, humidity, voltage, current, tilt angle, and light intensity sensors. Their primary function is to gather real-time environmental data from the solar panel installation. The Arduino microcontroller efficiently processes the sensor data, ensuring its readiness for transmission to the IoT platform.

2. ESP8266 Wi-Fi Modules:

The ESP8266 Wi-Fi modules are instrumental in facilitating wireless communication between the Arduino-based sensor nodes and the IoT platform. They serve as the backbone for transmitting sensor data over Wi-Fi networks, thereby enabling seamless connectivity and remote monitoring of the solar panel system.

One of the key aspects of the ESP8266 Wi-Fi modules is their robust performance, ensuring reliable data transmission essential for real-time monitoring and control. Their ability to consistently transmit data without interruption is vital for maintaining a continuous flow of information about the solar panel system's performance and status.

Moreover, these modules are designed to seamlessly integrate with existing Wi-Fi networks, ensuring compatibility across a wide range of infrastructures and environments. This compatibility simplifies the deployment and integration of the monitoring system into various settings, from residential installations to large-scale solar farms.

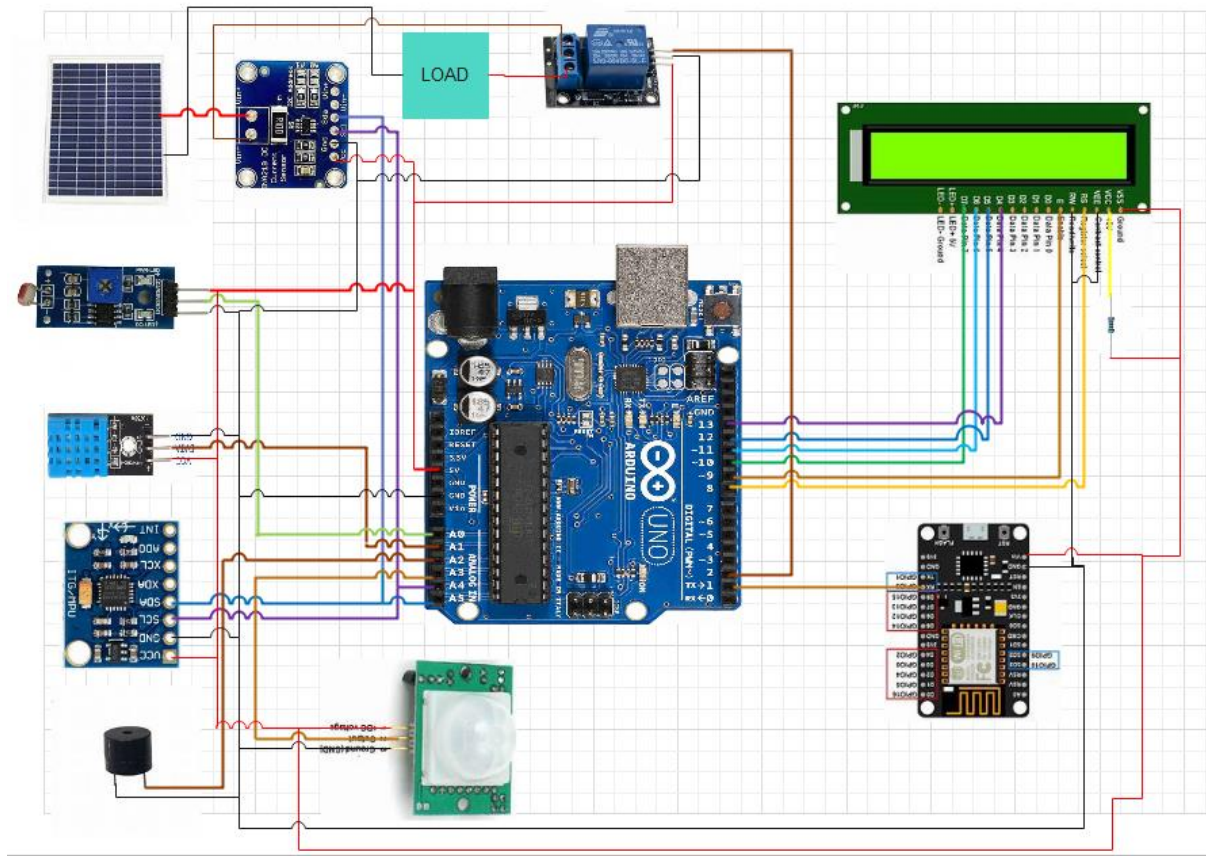


Fig 2: Circuit diagram

Another noteworthy feature of the ESP8266 Wi-Fi modules is their low power consumption. Despite their powerful capabilities, these modules consume minimal energy, which extends the operational lifespan of battery-powered sensor nodes and reduces overall energy costs for the system.

Furthermore, the built-in security features of the ESP8266 modules provide robust protection against unauthorized access and data breaches. Secure data transmission protocols, encryption mechanisms, and authentication methods ensure the integrity and confidentiality of sensor data transmitted over the network.

The modular design of the ESP8266 Wi-Fi modules also enhances the scalability and flexibility of the monitoring system. It allows for seamless expansion to accommodate additional sensor nodes or advanced functionalities as needed, ensuring that the system can adapt to evolving requirements and scale seamlessly.

Additionally, the remote management capabilities of the ESP8266 modules enable administrators to monitor and configure sensor nodes from anywhere with internet access. This remote accessibility simplifies system administration, troubleshooting, and maintenance, ultimately reducing downtime and enhancing overall system reliability.

In summary, the ESP8266 Wi-Fi modules are indispensable components of the solar panel monitoring system, contributing to its reliability, efficiency, and scalability. Their robust performance, compatibility, low power consumption, security features, and remote management capabilities make them essential for ensuring the seamless operation and effective management of solar panel installations.

3. ThingSpeak IoT Platform:

Integral to the system's functionality, the ThingSpeak IoT platform serves as the centralized hub for storing, visualizing, and analyzing data. Its user-friendly interface streamlines data management tasks and offers valuable insights into the solar panel system's performance without resorting to complicated processes or interfaces.

ThingSpeak's intuitive design facilitates efficient data management, allowing system operators to navigate and access critical information effortlessly. This simplicity in interface design enhances operational efficiency, enabling swift decision-making based on real-time data.

Moreover, ThingSpeak boasts a suite of robust data analytics tools, empowering users to derive actionable insights from sensor data. These tools facilitate comprehensive analysis, uncovering trends, patterns, and anomalies crucial for optimizing system performance and implementing proactive maintenance strategies. Leveraging advanced analytics capabilities, operators can identify potential issues early on, mitigating downtime and maximizing energy generation.

ThingSpeak's robust data storage capabilities ensure the secure storage of sensor data in a centralized repository. This centralized storage not only guarantees data integrity but also facilitates easy retrieval and analysis of historical data for long-term trend identification.

Furthermore, the platform's support for seamless integration with other IoT devices and platforms enhances interoperability and scalability. This interoperability allows for the integration of additional sensors or third-party applications, thereby expanding the system's capabilities and functionalities as needed.

In essence, the ThingSpeak IoT platform plays a pivotal role in our system architecture, providing a centralized platform for efficient data management, visualization, and analysis. Its user-friendly interface, powerful analytics tools, robust data storage capabilities, and interoperability make it an indispensable tool for optimizing the performance and reliability of solar panel installations.

4. AI-Based Analysis:

The collected sensor data undergoes rigorous analysis through an AI-based module, enabling real-time processing and decision-making.

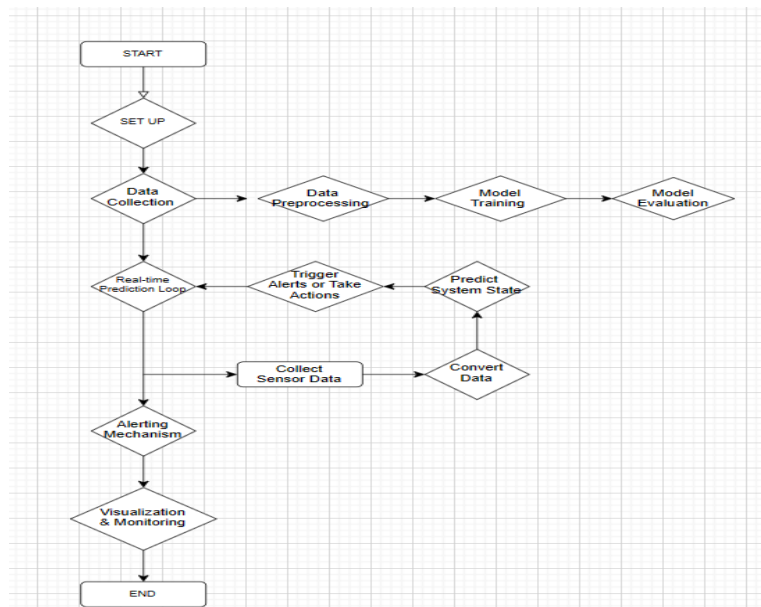


Fig 3: Flow Chart

Machine learning algorithms are meticulously trained to analyze the data, identifying patterns, anomalies, and potential issues impacting the solar panel system's performance. This intelligent analysis empowers the system to adopt proactive maintenance and optimization strategies, maximizing energy generation and enhancing system reliability.

5. System Expansion and Scalability:

Moreover, the proposed system architecture is inherently scalable, allowing for seamless expansion to accommodate additional sensor nodes or advanced AI modules. This scalability ensures the system's adaptability to evolving requirements and facilitates its integration into diverse environments. Additionally, the modular design of the architecture promotes flexibility, enabling the incorporation of emerging technologies and enhancements to further augment its capabilities.

IV. METHODOLOGY:

The methodology for implementing our integrated IoT and AI-based solar panel monitoring system follows a systematic approach to meet the identified system requirements and objectives. The outlined steps detail the key processes involved in deploying and operationalizing the monitoring system:

1. Requirement Analysis:

The initial step involves conducting a thorough analysis to identify the system requirements and objectives. This includes defining the monitoring system's scope, specifying monitored parameters, determining desired outcomes, and understanding the solar panel installation's operational environment. Stakeholder consultations and domain expertise ensure alignment with objectives of improving energy generation and system reliability.

2. Hardware Setup:

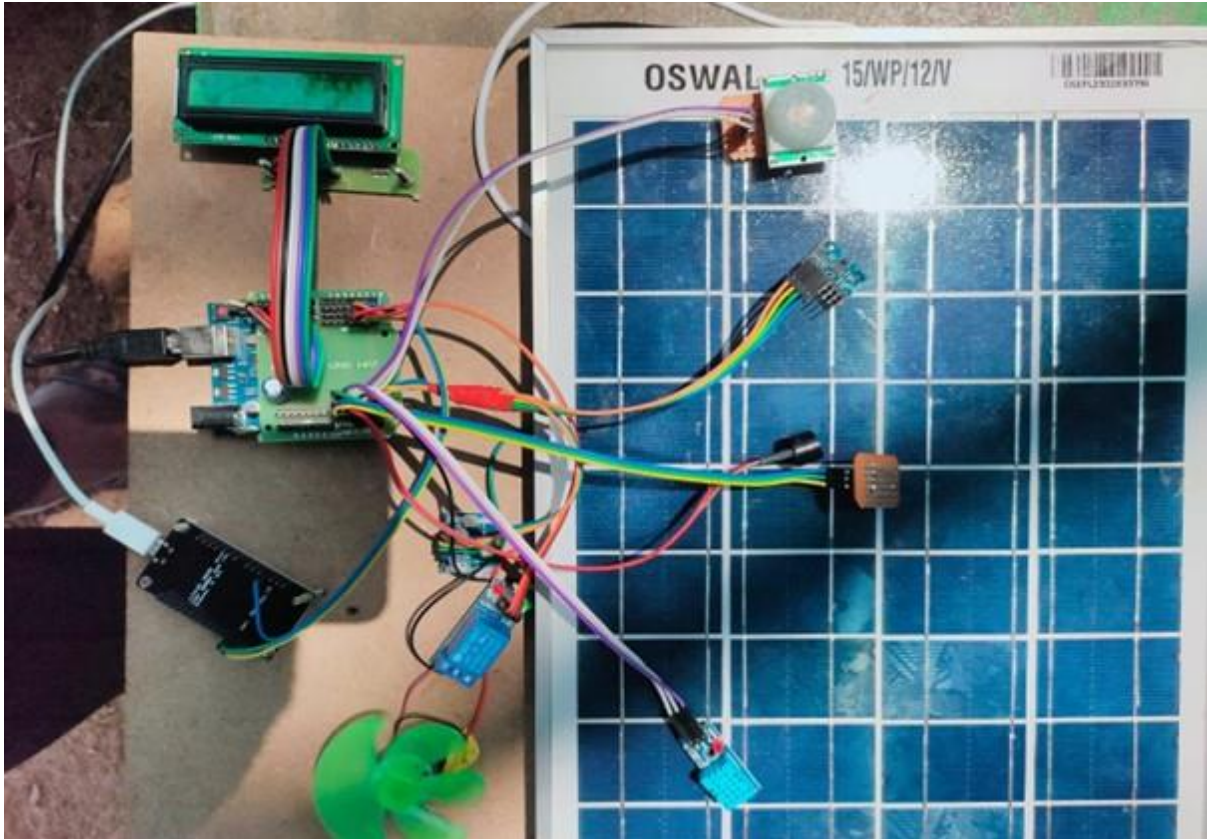


Fig4: Hardware kit

With established system requirements, the subsequent step is configuring the hardware components for data collection. Arduino-based sensor nodes are assembled and equipped with sensors like temperature, humidity, light intensity, tilt, current, and voltage sensors. Sensor selection, placement, and calibration are carefully considered to ensure accurate data acquisition. The hardware setup is designed to withstand environmental conditions and seamlessly interface with the IoT platform for data transmission.

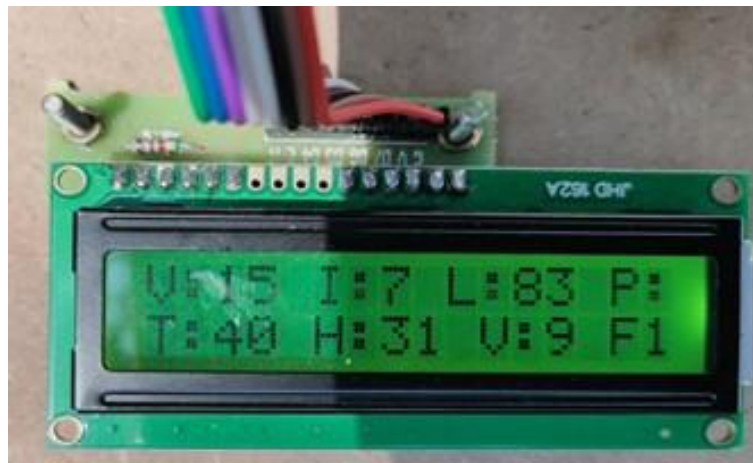


Fig 5: Hardware Kit Output

3. Software Development:

Custom software is developed to facilitate data collection, preprocessing, and transmission from Arduino-based sensor nodes to the ThingSpeak IoT platform. Tailored to the monitoring system's requirements, the software utilizes Arduino programming language (C/C++) for sensor interfacing and data manipulation. Emphasis is placed on optimizing code efficiency, minimizing resource utilization, and implementing error handling mechanisms for robust and reliable operation in real-world environments.

```

1  #include <LiquidCrystal.h>
2  #include <DFRobot_DHT11.h>
3  DFRobot_DHT11 DHT;
4  #include <Wire.h>
5  #include <Adafruit_MPU6050.h>
6  #include <Adafruit_Sensor.h>
7  #define DHT11_PIN A1
8  Adafruit_MPU6050 mpu;
9  #include <Adafruit_INA219.h>
10
11  Adafruit_INA219 ina219;
12
13  const int rs = 8, en = 9, d4 = 10, d5 = 11, d6 = 12, d7 = 13;
14  LiquidCrystal lcd(rs, en, d4, d5, d6, d7);
15
16  int rly=2;
17  int ls=A0;
18  int buz=A2;
  
```

Output

```

Using library Wire at version 1.0 in folder: C:\Users\hp\AppData\Local\Arduino15\packages\arduino\hardware\avr\1.8.6\libraries\Wire
Using library Adafruit MPU6050 at version 2.2.6 in folder: C:\Users\hp\OneDrive\Documents\Arduino\libraries\Adafruit_MPU6050
Using library Adafruit BusIO at version 1.15.0 in folder: C:\Users\hp\OneDrive\Documents\Arduino\libraries\Adafruit_BusIO
Using library SPI at version 1.0 in folder: C:\Users\hp\AppData\Local\Arduino15\packages\arduino\hardware\avr\1.8.6\libraries\SPI
Using library Adafruit Unified Sensor at version 1.1.14 in folder: C:\Users\hp\OneDrive\Documents\Arduino\libraries\Adafruit_Unified_Sensor
Using library Adafruit INA219 at version 1.2.3 in folder: C:\Users\hp\OneDrive\Documents\Arduino\libraries\Adafruit_INA219
"C:\Users\hp\AppData\Local\Arduino15\packages\arduino\tools\avr-gcc\7.3.0-atmel3.6.1-arduino7/bin/avr-size" -A "C:\Users\hp\AppData\Local\Te
Sketch uses 18650 bytes (57%) of program storage space. Maximum is 32256 bytes.
Global variables use 716 bytes (34%) of dynamic memory, leaving 1332 bytes for local variables. Maximum is 2048 bytes.
  
```

Fig 6: Arduino uno window

4. Data Transmission:

Upon completion of hardware setup and software development, sensor nodes commence collecting environmental data from the solar panel installation. Collected sensor data is wirelessly transmitted to the ThingSpeak IoT platform using ESP8266 Wi-Fi modules. ThingSpeak offers a user-friendly interface and APIs for seamless data ingestion, ensuring integration with the monitoring system. Data transmission protocols are implemented to ensure data integrity, security, and compliance with IoT standards.

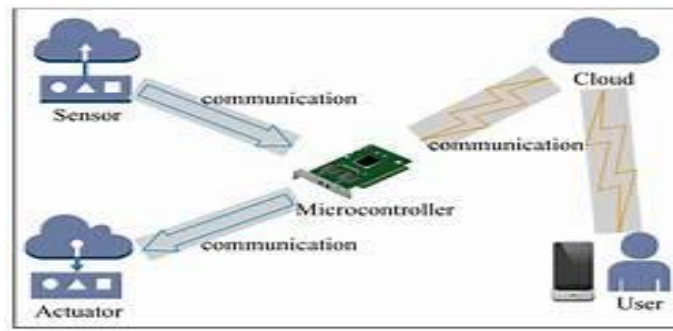


Fig 7: Data Transmission

5. AI-Based Analysis:

Upon receiving sensor data, real-time AI-based analysis is conducted to extract actionable insights and identify potential issues affecting solar panel performance. Machine learning algorithms such as regression analysis, anomaly detection, or predictive modeling analyze historical and streaming data. These algorithms recognize patterns, trends, and deviations from normal operating conditions, facilitating proactive identification of maintenance needs, optimization opportunities, and performance enhancements.

ALGORITHM

I. Setup:

- Import necessary libraries (NumPy, pandas, Matplotlib, scikit-learn, etc.).
- Define helper functions (if any).

II. Data Collection:

- Fetch sensor data from the Thingspeak channel.
- Extract voltage, current, tilt, vibration, temperature, and humidity readings.

1. Data Preprocessing:

- Prepare the data for prediction (e.g., convert strings to numeric values, scale the data if necessary).

2. Model Training:

- Load the labeled dataset (salar.csv).
- Split the dataset into training and testing sets.
- Train a KNN classifier on the training data.

3. Model Evaluation:

- Evaluate the classifier's accuracy using the test set.

4. Real-time Prediction:

- Continuously collect sensor data.
- Convert the data into the required format.
- Use the trained classifier to predict the system state ("Normal" or "Action Required").
- Based on the prediction, trigger alerts or take appropriate actions.

5. Visualization (Optional):

- Plot sensor data over time.
- Visualize model performance (e.g., accuracy, confusion matrix).

6. Continuous Monitoring:

- Run the monitoring script continuously to keep track of system state in real-time.

7. Improvement and Optimization (Optional):

- Fine-tune the model parameters.
- Explore other machine learning algorithms for classification.
- Optimize the code for efficiency and scalability.

6. Output Generation:

Based on AI analysis results, the monitoring system generates actionable insights, alerts, and recommendations for system operators and stakeholders. Intuitive dashboards, visualizations, and notifications present these outputs, facilitating informed decision-making and proactive maintenance strategies. Additionally, automated responses or control actions may be triggered to address identified issues or optimize system parameters in real-time.

V. RESULTS AND ANALYSIS:

This section presents the findings derived from deploying the solar panel monitoring system, incorporating IoT and AI methodologies, and provides a detailed analysis of these outcomes.

1. Visual Representation of Sensor Data:

The sensor data collected from the solar panel setup is graphically depicted using cloud-based graphs, providing valuable insights into the system's performance dynamics. These graphical representations offer a clear depiction of environmental variables, panel orientation, and electrical metrics over time. By scrutinizing these data patterns, anomalies, and fluctuations, potential operational issues or optimization pathways can be discerned.

THINGSPEAK CLOUD DATA:

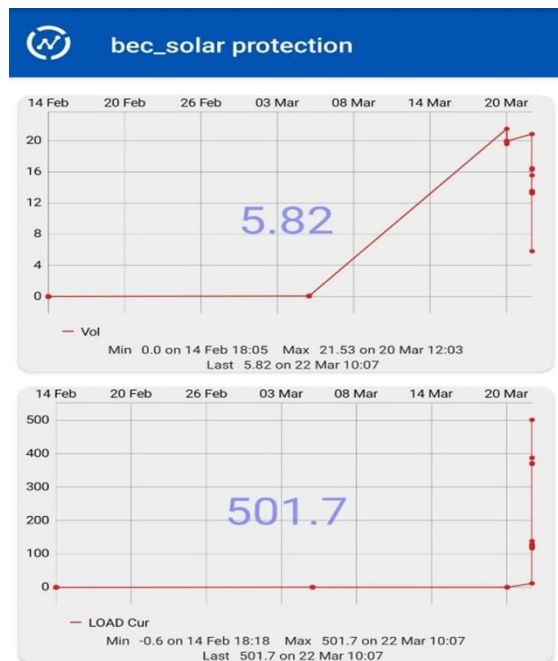


Fig.8:voltageandcurrentvalues

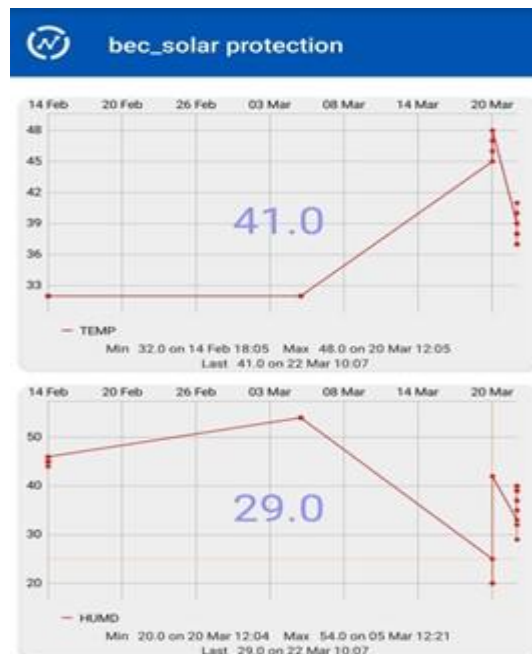


Fig. 9: Temperature and humidity values

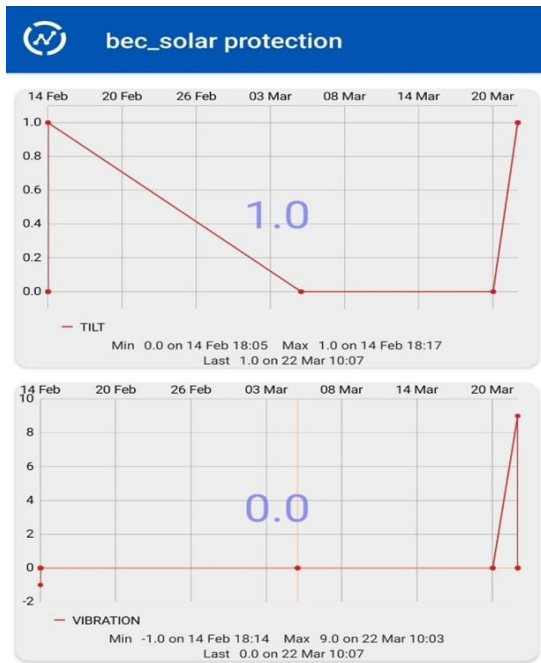


Fig. 10: Tilt angle and vibration values

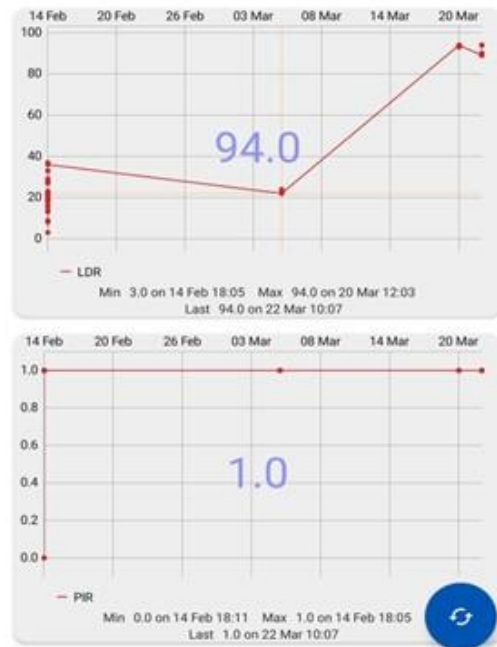


Fig.11: LDR and PIR values

REALTIME DATA TABLE:

SI.NO	TIME	VOL	CUR	TEMP	HUM	LDR	PIR	TILT	VIB
1.	09:00 am	18	501.7	29	34	50	1.0	1.0	0.0
2.	10:00 am	22	789	30	45	70	1.0	0.0	1.0
3.	11:00 am	22	806.1	32	60	90	0.0	1.0	0.0
4.	01:00 pm	24	1340	41	70	99	1.0	1.0	0.0
5.	03:00 pm	23	1040	35	75	80	1.0	1.0	0.0
6.	04:00 pm	23	1020	32	75	75	0.0	0.0	0.0
7.	05:00 pm	22	980	30	76	65	0.0	0.0	0.0

2. AI-Driven Anomaly Detection:

Through real-time data analysis, the AI system adeptly identifies anomalies and deviations from established operational norms. By juxtaposing incoming sensor data against predetermined thresholds, the system highlights irregularities such as temperature spikes, humidity variations, or voltage fluctuations. These alerts facilitate swift remedial action to rectify issues and mitigate the risk of operational disruptions.

3. Fault Detection:

Leveraging AI-driven insights, the system triggers alerts for fault detection and provides actionable maintenance suggestions to uphold optimal performance of the solar panel installation. Notable deviations, such as sudden voltage drops, prompt alerts, prompting tailored interventions such as panel cleaning or recalibration to optimize system functionality.

```

ml_code_Thingspeak1.py - C:\Users\hp\OneDrive\Desktop\SURENDRA\ml_code\ml_code_Thingspeak1.py (3.10.4)
File Edit Format Run Options Window Help
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib import rcParams
from matplotlib.cm import rainbow
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
from sklearn.linear_model import LogisticRegression
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn import datasets, linear_model
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score, make_scorer, precision_scorer
import serial
import numpy as np
from sklearn.preprocessing import StandardScaler
import time
import urllib.request
ascaler = StandardScaler()

def takeInput():
    while True:
        r_link='https://api.thingspeak.com/channels/369231/fields/1/last'
        f=urllib.request.urlopen(r_link)
        pr1 = (f.readline()).decode()

        r_link='https://api.thingspeak.com/channels/369231/fields/2/last'
        f=urllib.request.urlopen(r_link)
        pr2 = (f.readline()).decode()

        r_link='https://api.thingspeak.com/channels/369231/fields/3/last'
        f=urllib.request.urlopen(r_link)
        pr3 = (f.readline()).decode()

IDLE Shell 3.10.4
Python 3.10.4 (tags/v3.10.4:9d38120, Mar 23 2022, 23:13:41) [MSC v.1929 64 bit (AMD64)] on win32
Type "help", "copyright", "credits" or "license()" for more information.
>>>
= RESTART: C:\Users\hp\OneDrive\Desktop\SURENDRA\ml_code\ml_code_Thingspeak1.py
Accuracy with RMN: 1.0
Accuracy with SVC: 1.0
Accuracy with LR: 1.0
Vol:20.71 Cur:0.30 Tilt:0.00 Vib:0 Temp:43 Humd:31
[[20.71 0.3 0. 0. 43. 31. ]]

Warning (from warnings module):
  File "C:\Users\hp\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\base.py", line 493
    warnings.warn(
UserWarning: X does not have valid feature names, but KNeighborsClassifier was fitted with feature names
[1]
Normal
Vol:20.71 Cur:0.30 Tilt:0.00 Vib:0 Temp:43 Humd:31
[[20.71 0.3 0. 0. 43. 31. ]]

Warning (from warnings module):
  File "C:\Users\hp\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\base.py", line 493
    warnings.warn(
UserWarning: X does not have valid feature names, but KNeighborsClassifier was fitted with feature names
[1]
Normal
Ln: 26 Col: 0
Ln: 1 Col: 0

```

Fig 12: Output of Fault Detection Window

Insights gleaned from AI analysis and system-generated recommendations empower operators to fine-tune the solar panel system's performance. By adjusting parameters such as panel orientation, tilt angles, or maintenance schedules based on data trends, operators aim to optimize energy generation efficiency and bolster system reliability, thereby minimizing operational downtime.

The findings derived from the solar panel monitoring system foster a culture of iterative improvement and adaptability. Long-term monitoring of system performance and analysis of historical data enable operators to pinpoint areas for enhancement and proactively implement corrective measures. Additionally, ongoing refinement of AI algorithms based on updated data and stakeholder feedback ensures the system remains agile and responsive to evolving operational requirements and environmental conditions.

CONCLUSION & SUMMERY

the amalgamation of IoT and AI technologies presents a compelling avenue for augmenting the effectiveness and dependability of solar panel systems. Through the facilitation of real-time monitoring and astute analysis, our integrated approach empowers stakeholders to enact proactive measures, optimizing energy generation and bolstering maintenance protocols. Looking ahead, future endeavors will focus on refining AI algorithms and ensuring the scalability of the monitoring infrastructure to accommodate expansive deployments, thereby advancing the realm of renewable energy management.

the fusion of IoT and AI stands as a beacon of promise in revolutionizing solar panel systems' performance and resilience. By seamlessly integrating real-time monitoring and intelligent analysis, our methodology empowers stakeholders to navigate dynamic energy landscapes with precision and foresight. As we chart our course forward, continued research endeavors will hone AI algorithms for even greater efficiency while fortifying the scalability of our monitoring framework to embrace the challenges of tomorrow's renewable energy frontiers.

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