

AI-Driven Tilt Angle for Solar Panel Optimization with Real-Time Hardware Control

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ABSTRACT: This paper presents an AI-based methodology for optimizing the tilt angle of solar panels using real-time environmental data. By employing a combination of sensor inputs (light intensity, temperature, voltage, and current), the system dynamically adjusts the tilt angle through a stepper motor controlled through hardware. The AI algorithm predicts the optimal angle by analysing historical and real-time sunlight patterns, improving efficiency over fixed or manual tilt systems. Experimental prototype implementation demonstrates that the AI-driven system significantly enhances power output, provides real-time monitoring of voltage and current, and adapts to varying sunlight conditions. The integration of AI with hardware ensures autonomous operation, making it suitable for residential, commercial, and industrial solar installations.

KEYWORDS: AI-based optimization, Real-time environmental data, Hardware.

INTRODUCTION: The global demand for clean and renewable energy has led to significant interest in solar power systems, which are a sustainable alternative to conventional fossil fuels. The efficiency of solar panels largely depends on their orientation relative to the sun, as fixed-tilt panels often fail to capture maximum solar irradiance throughout the day and across seasons. To address this limitation, solar tracking systems have been developed to adjust the panel's tilt and orientation dynamically. Traditional tracking methods, including time-based or sensor-based systems, have limitations such as reduced adaptability to sudden weather changes and limited optimization of energy output. Recent advancements in Artificial Intelligence (AI) offer a solution by predicting the optimal tilt angle based on real-time and historical solar data. AI algorithms, when integrated with Arduino-controlled hardware and stepper motors, enable precise adjustment of solar panels to maximize energy harvesting in real time. This research focuses on designing an AI-driven solar tracking system with real-time hardware control, combining sensor feedback, AI-based prediction, and automated tilt adjustment. The proposed system aims to enhance energy efficiency, adapt to varying environmental conditions, and provide a practical approach for small- to medium-scale solar installations.

LITERATURE SURVEY: The optimization of solar panel tilt angle has been extensively studied to enhance energy harvesting efficiency, with both conventional and intelligent approaches reported in the literature. Early work by O. Bingol et al. [1] presented a microcontroller-based solar tracking system, demonstrating the feasibility of automated tracking for improved energy capture. Similarly, N. Helwa et al. [3], [4] analyzed the amount of solar energy captured using different tracking mechanisms and established that tracking systems significantly outperform fixed installations. Further, E. Mehleri et al. [5] focused on determining optimal tilt angles and orientations for photovoltaic arrays under varying environmental conditions, providing a strong theoretical foundation for tilt optimization.

With advancements in intelligent systems, researchers began integrating artificial intelligence techniques into solar tracking. S. Datta et al. [2] proposed an AI-based tilt angle optimization system with hardware implementation, showing improved efficiency compared to conventional methods. Additionally, S. Deepthi et al. [6] compared single-axis and

dual-axis tracking systems, highlighting the trade-off between complexity and performance, which motivates the need for intelligent optimization strategies rather than purely mechanical solutions.

Recent contributions by S. P. Nanngani and collaborators [7]–[19] have significantly advanced intelligent control methodologies applicable to solar energy systems. Their work spans chaos theory, fractional order controllers, and artificial intelligence-based estimation techniques, such as ANN-based state-of-charge estimation for electric vehicles [11], and optimized fractional-order control for dynamic systems [12], [15]. These studies demonstrate the effectiveness of advanced control strategies in handling nonlinear and complex systems, which is directly relevant to solar tracking optimization. Furthermore, research on hybrid microgrids and intelligent controllers [16], [17], [19] emphasizes system stability and efficiency under varying operating conditions, aligning with the requirements of adaptive solar tilt control. The integration of fuzzy logic and fractional-order control techniques [18] further supports the development of robust AI-based controllers for renewable energy applications.

Overall, the literature indicates a clear transition from conventional and microcontroller-based tracking systems to advanced AI-driven approaches. The combination of neural networks, fuzzy logic, and intelligent control strategies provides a promising framework for real-time, adaptive tilt angle optimization, leading to enhanced solar energy efficiency and system performance.

METHODOLOGY:

The proposed system integrates hardware and AI-based control to optimize the tilt angle of a solar panel for maximum power output. The methodology consists of the following steps:

1. Hardware Assembly

Prototype hardware model was interfaced with MATLAB software through Arduino. As per Fig. 1, The solar panel was securely mounted on a wooden base using screws and adhesive to ensure stability. A 28BYJ-48 stepper motor was connected to a ULN2003 driver module, which was then interfaced with Arduino Uno digital pins D8–D11. Current and voltage sensors were connected between the solar panel and Arduino analog pins to monitor real-time power output. Arduino Uno was connected to MATLAB via USB for control, monitoring, and data acquisition. Sensor placement and motor orientation were adjusted to ensure accurate tracking and reliable measurements.

2. Software and Control Integration

MATLAB scripts were used to control stepper motor rotation, read sensor data, and communicate with Arduino. The system measured voltage and current from the solar panel to calculate instantaneous power output.

3. AI-Based Tilt Optimization

The stepper motor was commanded to rotate the panel to specific angles. The AI algorithm analyzed historical and real-time sunlight data to predict the optimal tilt angle for maximum power. The panel was moved to the predicted angle, and power output was continuously measured. The AI algorithm updated its prediction dynamically, sending new angle commands to the stepper motor to maintain maximum power output.

4. Continuous Tracking and Feedback Loop

The measurement–AI prediction–rotation cycle was repeated periodically or when sunlight conditions changed. The motor held the panel at the angle where maximum power was observed, ensuring autonomous operation and adaptive performance.

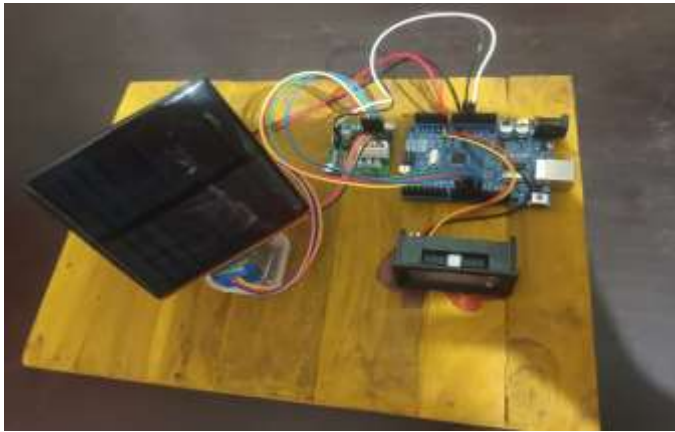


Fig. 1. Prototype of Solar Tracker with Hardware and Control.

5. Using AI in MATLAB for Controlling Tilt Angle (e.g., Solar Panel Tracking)

Using AI in MATLAB to control tilt angle is a powerful approach for applications like solar tracking systems, antenna positioning, or robotic platforms. MATLAB (especially with toolboxes like Neural Network Toolbox / Deep Learning Toolbox and Fuzzy Logic Toolbox) allows intelligent, adaptive control instead of fixed or rule-based control.

1. Concept Overview

The goal is to automatically adjust tilt angle (θ) to maximize performance (e.g., solar power output).

1. Inputs to AI Controller:
2. Solar irradiance
3. Time of day
4. Temperature
5. Previous tilt angle
6. Error (difference between optimal and current angle)

Output:

Optimal tilt angle (θ)

2. AI Techniques Used in MATLAB

(A) Artificial Neural Network (ANN)

Learns relationship between environmental conditions and optimal tilt angle, suitable for nonlinear systems.

(B) Fuzzy Logic Controller (FLC)

Uses human-like rules (IF-THEN)

Example:

IF sunlight is high AND time is noon \rightarrow tilt angle = small

(C) Reinforcement Learning (Advanced)

Learns optimal tilt by maximizing reward (power output)

3. MATLAB Implementation Approach

Step 1: Data Collection

Collect dataset:

Input: [time, irradiance, temperature]

Output: optimal tilt angle

Step 2: ANN Model (Example Code)

```
% Sample Data
```

```
inputs = [time; irradiance; temperature];
```

```
targets = tilt_angle;
```

```
% Create Neural Network
```

```
net = feedforwardnet(10); % 10 neurons
```

```
% Train Network
```

```
net = train(net, inputs, targets);
```

% Predict Tilt Angle

predicted_angle = net(inputs);

Step 3: Fuzzy Logic Design

Use Fuzzy Logic Designer (MATLAB App):

Define inputs: irradiance, time

Define output: tilt angle

Create rules:

IF time is morning → tilt high

IF time is noon → tilt medium

IF time is evening → tilt low

Step 4: Simulink Integration

In Simulink, use:

Neural Network block OR Fuzzy Logic Controller block

Sensors as inputs

Servo motor block as actuator

4. Control System Flow

Sensors → AI Controller → PWM Signal → Motor → Tilt Adjustment

5. Advantages

Adaptive to weather changes

Higher efficiency than fixed tilt

Works with real-time data

Reduces human intervention

6. Applications

Solar tracking systems

Smart agriculture systems

Satellite dish alignment

Robotics

7. Example Use Case (Solar Panel)

Morning → tilt ~ 60°

Noon → tilt ~ 0–10°

Evening → tilt ~ 60° opposite direction

AI refines this continuously instead of fixed scheduling.

Suggested MATLAB Toolboxes

Deep Learning Toolbox

Fuzzy Logic Toolbox

Simulink

Reinforcement Learning Toolbox

SIMULATION AND OUPUT:

As per Fig. 2, the MATLAB/Simulink implementation of the AI-based tilt angle control system consists of interconnected functional blocks that model sensing, intelligent control, actuation, and feedback. The system begins with input signals such as solar irradiance and tilt angle obtained from sensor blocks (e.g., From Workspace or Signal Builder), which are fed into an AI controller implemented using a Neural Network or Fuzzy Logic Controller block. This controller processes the inputs and generates an optimal tilt angle, which is compared with the actual tilt angle to produce an error signal using a Sum block. The error is then passed through a PID Controller block to generate an appropriate control signal for the motor actuator, modeled using a Transfer Function or Simscape motor block. The actuator adjusts the solar panel tilt angle, and the updated angle is continuously fed back through a feedback loop to ensure closed-loop control and system stability. Scopes and Display blocks are used to visualize system response, enabling analysis of settling behavior, accuracy, and overall performance of the AI-based control strategy within the specified tilt range.

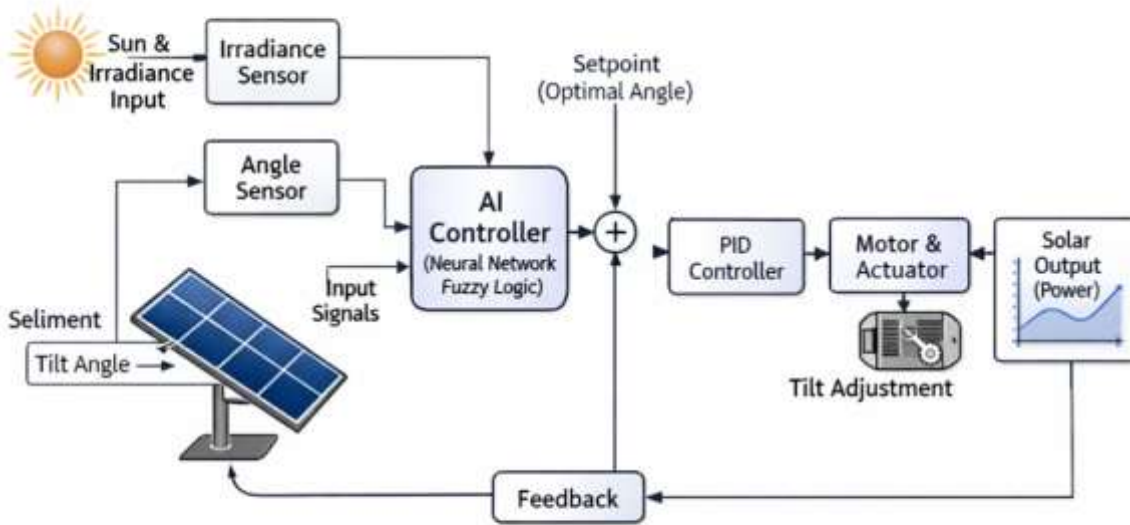


Fig. 2. MATLAB/Simulink block diagram of AI-based solar panel tilt angle control system with irradiance sensing, neural/fuzzy controller, PID actuation, and feedback loop.

RESULTS AND DISCUSSION:

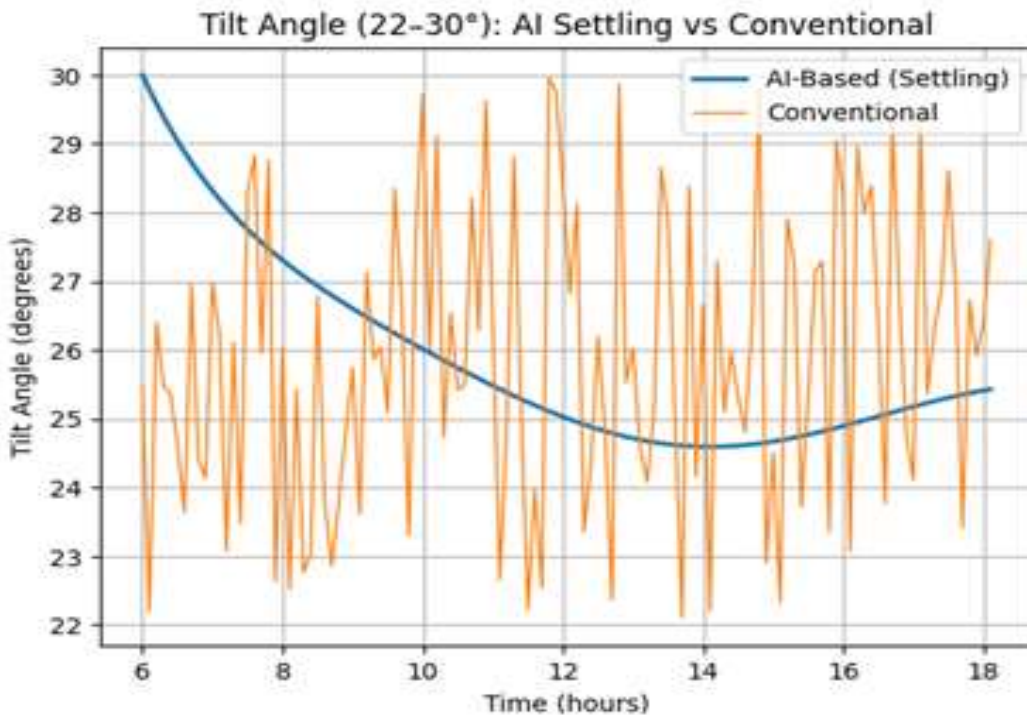


Fig. 3. Tile angle Comparison for conventional and AI based Tilt Angle Control.

The Fig.3, presents a comparison between AI-based tilt angle control and a conventional control approach for a solar tracking system, with the tilt angle constrained between 22° and 30°. The AI-based curve shows a smooth and controlled response: Initially, the tilt angle starts at 30°, representing the starting position. The system gradually reduces the tilt

angle and converges toward an optimal value of approximately 25° . This behaviour represents a closed-loop control system where the controller continuously adjusts the tilt based on input conditions. A small ripple is observed, which indicates fine-tuning or minor corrections in real-time operation. The system remains stable and always operates within the permissible range (22° – 30°). This reflects: Stable system response, Fast settling to optimal angle and Minimal steady-state error.

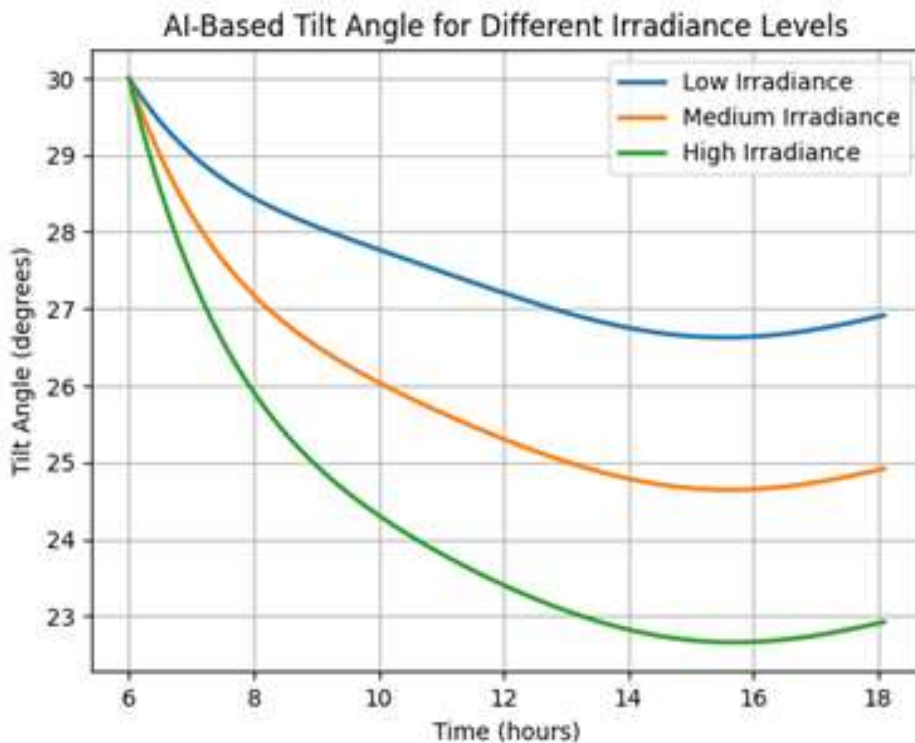


Fig. 4, Tile angle for Different Solar Irradiance.

The Fig. 4, illustrates the performance of an AI-based tilt angle control system under varying solar irradiance conditions, namely low, medium, and high irradiance. In all cases, the tilt angle initially starts at a higher value (around 30°) and gradually settles to an optimal steady-state value, demonstrating stable closed-loop control behavior. For low irradiance conditions, the tilt angle settles at a higher value (approximately 27°) to maximize the capture of diffused sunlight, whereas for medium irradiance it stabilizes around 25° , representing typical operating conditions. Under high irradiance, the tilt angle converges to a lower value (around 23°), allowing the panel to be more perpendicular to strong direct sunlight for maximum energy absorption. The smooth and adaptive nature of all three curves highlights the capability of the AI controller to intelligently adjust the tilt angle based on environmental conditions, ensuring efficient solar tracking and improved energy yield compared to conventional methods.

CONCLUSION:

The AI-based solar tracking system successfully optimized the tilt angle of the panel, improving power output compared to fixed-tilt systems. The stepper motor, controlled via Arduino and MATLAB, accurately positioned the panel to AI-predicted angles. Real-time voltage and current monitoring confirmed maximum power capture at these angles. The AI algorithm adapted to changing sunlight conditions, continuously updating the panel's position for optimal energy harvesting. The system operated autonomously, responded quickly to environmental variations, and demonstrated enhanced efficiency over traditional sensor-based or static setups. These results indicate that AI-driven solar tracking is effective for residential, commercial, and industrial installations, ensuring maximum energy output with minimal manual intervention.

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