



**SIMATS**  
ENGINEERING



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Saveetha Institute of Medical And Technical Sciences  
(Declared as Deemed to be University under Section 3 of UGC Act 1956)

## ASSIGNMENT

### CSA0612 – Design and Analysis of Algorithms for Optimization

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### TITLE: MINIMIZING DATA CENTER ENERGY USAGE

#### **Problem Statement:**

Design an algorithm to dynamically adjust cooling system and power usage in a data center, based on server activity and external temperature data.

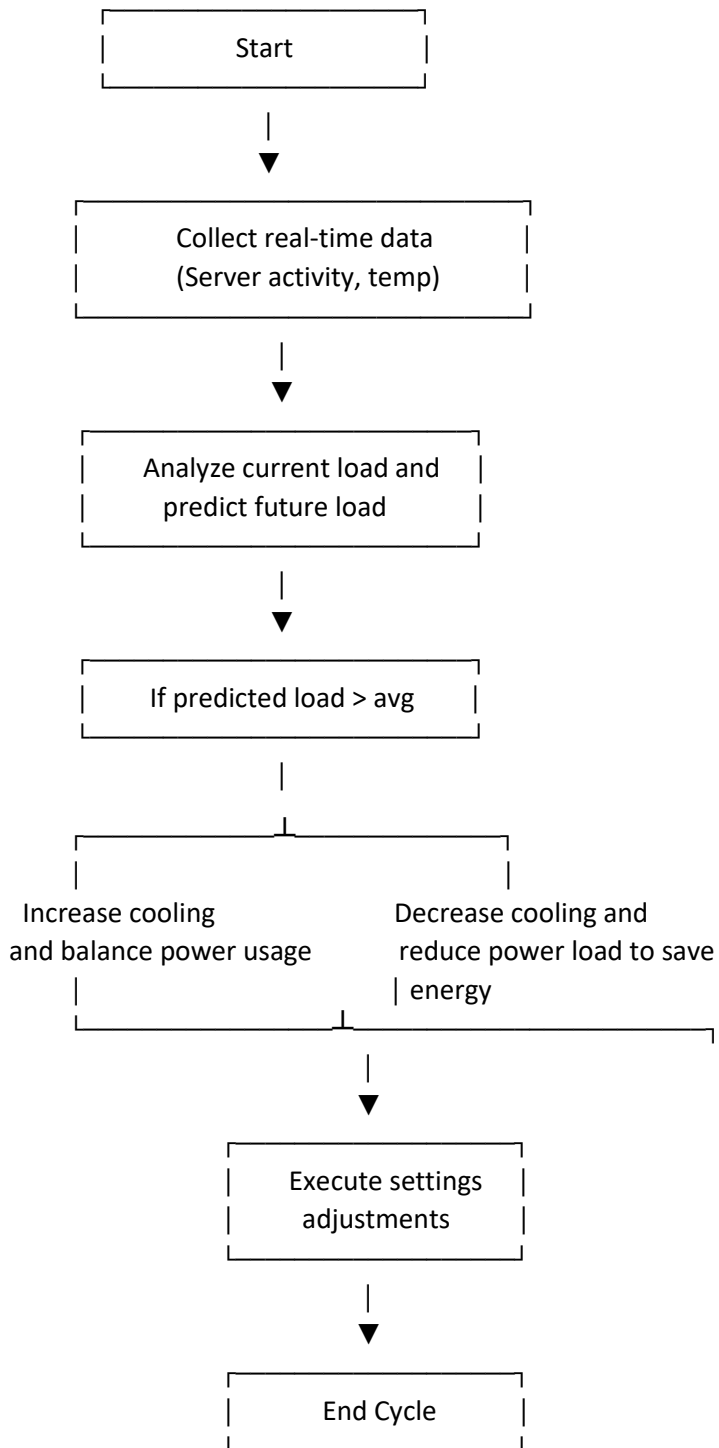
#### **Tasks:**

- Create an algorithm that adjust cooling and power based on activity data.
- Incorporate predictive adjustments based on server load patterns.

#### **Deliverables:**

- Flowchart of the power optimization algorithm.
- Complexity analysis.
- Test case with energy usage comparison.

**Flowchart for problem solving:**



### **Pseudocode:**

BEGIN

// Step 1: Initialize system

Set cooling\_level to default

Set power\_level to default

// Step 2: Collect data

WHILE data\_center is operational DO

    Collect server\_activity\_data

    Collect external\_temperature\_data

// Step 3: Predict future load

    predicted\_load = PredictLoad(server\_activity\_data)

// Step 4: Adjust cooling based on current and predicted load

    IF predicted\_load > high\_threshold OR external\_temperature > high\_temp THEN

        Set cooling\_level to HIGH

        Set power\_level to BALANCED

    ELSE IF predicted\_load > medium\_threshold THEN

        Set cooling\_level to MEDIUM

        Set power\_level to MODERATE

    ELSE

        Set cooling\_level to LOW

        Set power\_level to REDUCED

    END IF

// Step 5: Execute settings adjustments

    AdjustCooling(cooling\_level)

    AdjustPower(power\_level)

// Step 6: Log data for analysis and repeat cycle

    LogEnergyUsage(cooling\_level, power\_level)

    WAIT for next cycle

END WHILE

END

// Function: PredictLoad

FUNCTION PredictLoad(data)

    Calculate pattern of server activity over recent cycles

    RETURN predicted load level based on pattern

END FUNCTION

// Function: AdjustCooling

FUNCTION AdjustCooling(level)

    IF level == HIGH THEN

        Activate high cooling mode

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ELSE IF level == MEDIUM THEN
    Activate medium cooling mode
ELSE
    Activate low cooling mode
END IF
END FUNCTION

// Function: AdjustPower
FUNCTION AdjustPower(level)
    IF level == BALANCED THEN
        Distribute power load evenly across servers
    ELSE IF level == MODERATE THEN
        Slightly reduce power to lower-activity servers
    ELSE
        Minimize power to inactive servers
    END IF
END FUNCTION
```

### **Actual Code:**

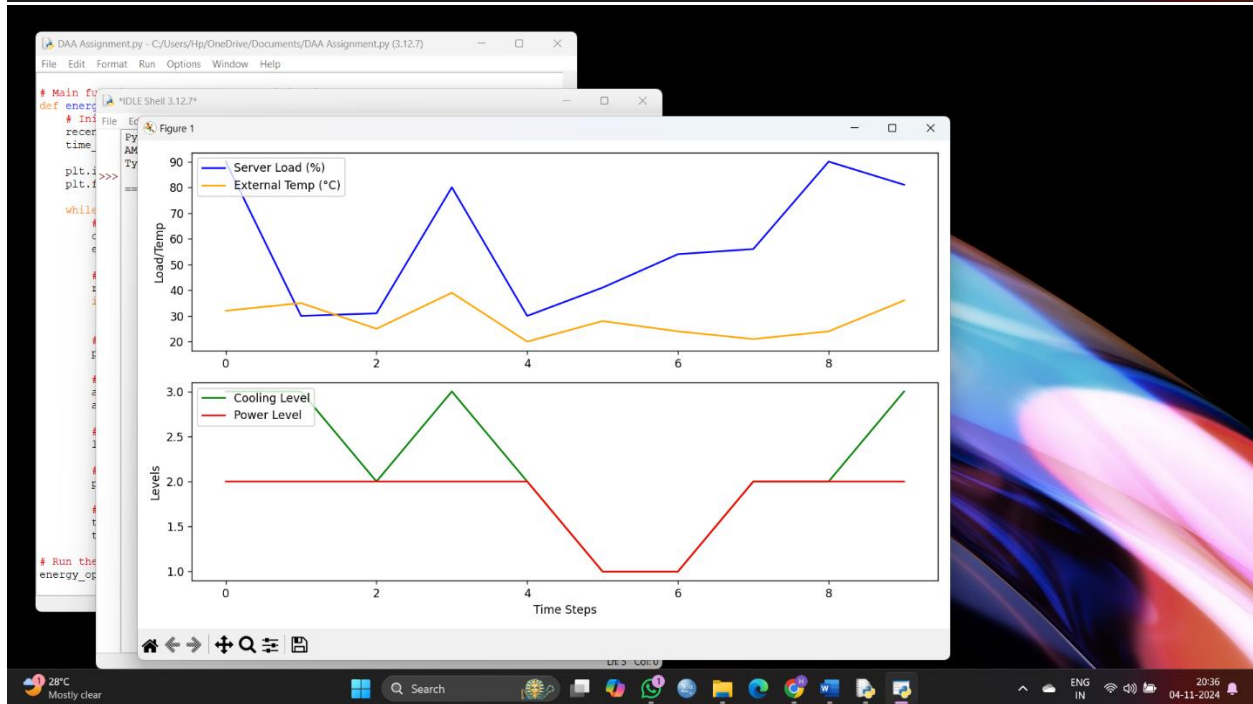
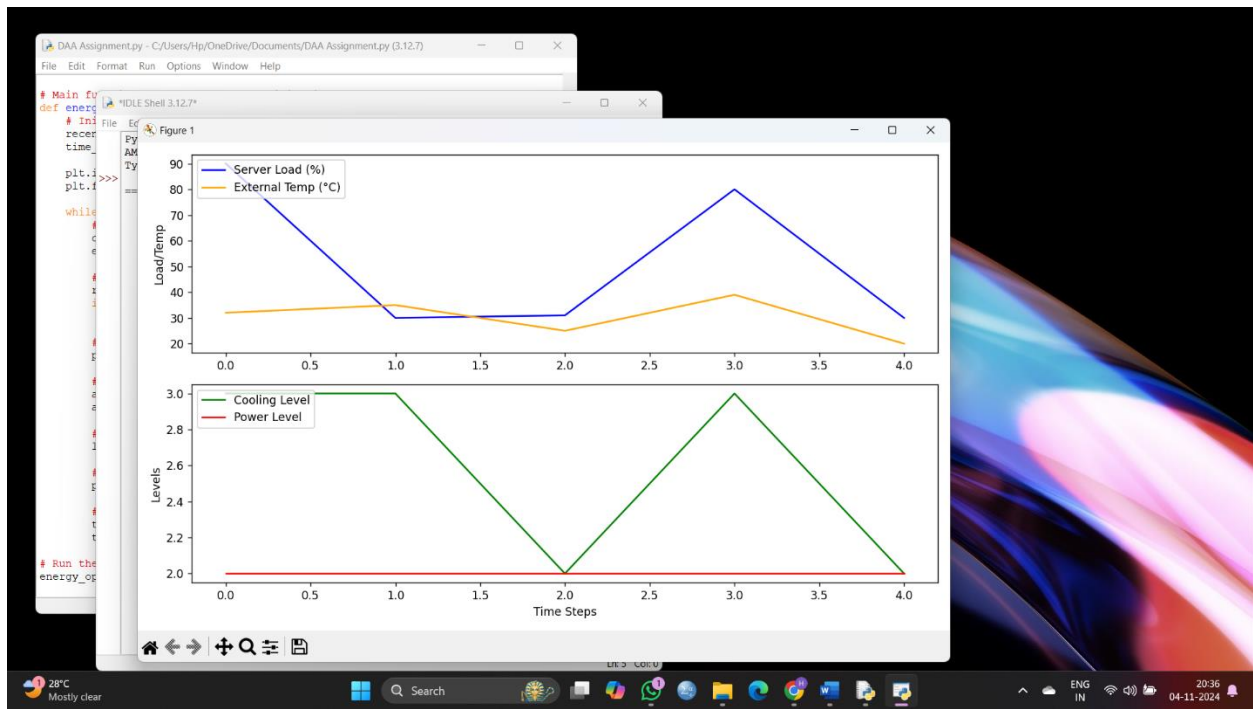
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import time
import random
# Initial thresholds and default levels
HIGH_THRESHOLD = 80 # High server load percentage threshold
MEDIUM_THRESHOLD = 50 # Medium server load percentage threshold
HIGH_TEMP = 30 # High temperature threshold (°C)
# Initial levels for cooling and power
cooling_level = "default"
power_level = "default"
# Function to predict future load based on recent activity
def predict_load(recent_activity_data):
    # Simulate load prediction by averaging recent activity
    avg_load = sum(recent_activity_data) / len(recent_activity_data)
    return avg_load
# Function to adjust cooling level
def adjust_cooling(level):
    if level == "HIGH":
        print("Setting cooling to high.")
    elif level == "MEDIUM":
        print("Setting cooling to medium.")
    else:
        print("Setting cooling to low.")
# Function to adjust power level
def adjust_power(level):
    if level == "BALANCED":
        print("Setting power to balanced.")
    elif level == "MODERATE":
        print("Setting power to moderate.")
    else:
        print("Setting power to reduced.")
# Main function to run the energy optimization loop
def energy_optimization():
    global cooling_level, power_level
    recent_activity_data = [random.randint(30, 90) for _ in range(5)] # Simulate initial server
activity data
    while True:
        # Step 1: Collect data
        current_activity = random.randint(30, 90) # Simulate real-time server load
        external_temp = random.randint(20, 40) # Simulate real-time external temperature
        recent_activity_data.append(current_activity)
        # Keep only recent 5 activity data points
        if len(recent_activity_data) > 5:
            recent_activity_data.pop(0)
        # Step 2: Predict future load
```

```

predicted_load = predict_load(recent_activity_data)
# Step 3: Adjust cooling and power based on load and temperature
if predicted_load > HIGH_THRESHOLD or external_temp > HIGH_TEMP:
    cooling_level = "HIGH"
    power_level = "BALANCED"
elif predicted_load > MEDIUM_THRESHOLD:
    cooling_level = "MEDIUM"
    power_level = "MODERATE"
else:
    cooling_level = "LOW"
    power_level = "REDUCED"
# Step 4: Execute adjustments
adjust_cooling(cooling_level)
adjust_power(power_level)
# Step 5: Log data for analysis
print(f'Activity: {current_activity}% | Temp: {external_temp}°C | Predicted Load:
{predicted_load:.2f}%")
print(f'Cooling: {cooling_level} | Power: {power_level}\n")
# Wait for the next cycle (simulate delay)
time.sleep(2)
# Run the energy optimization simulation
energy_optimization()

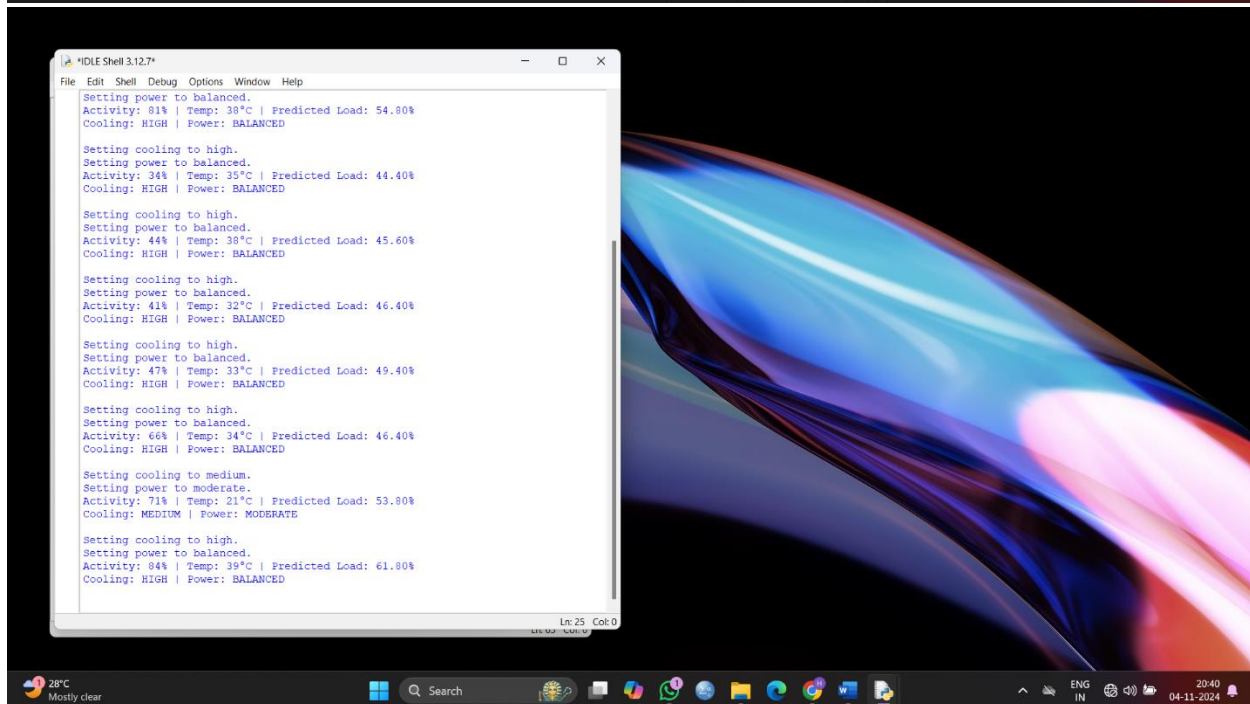
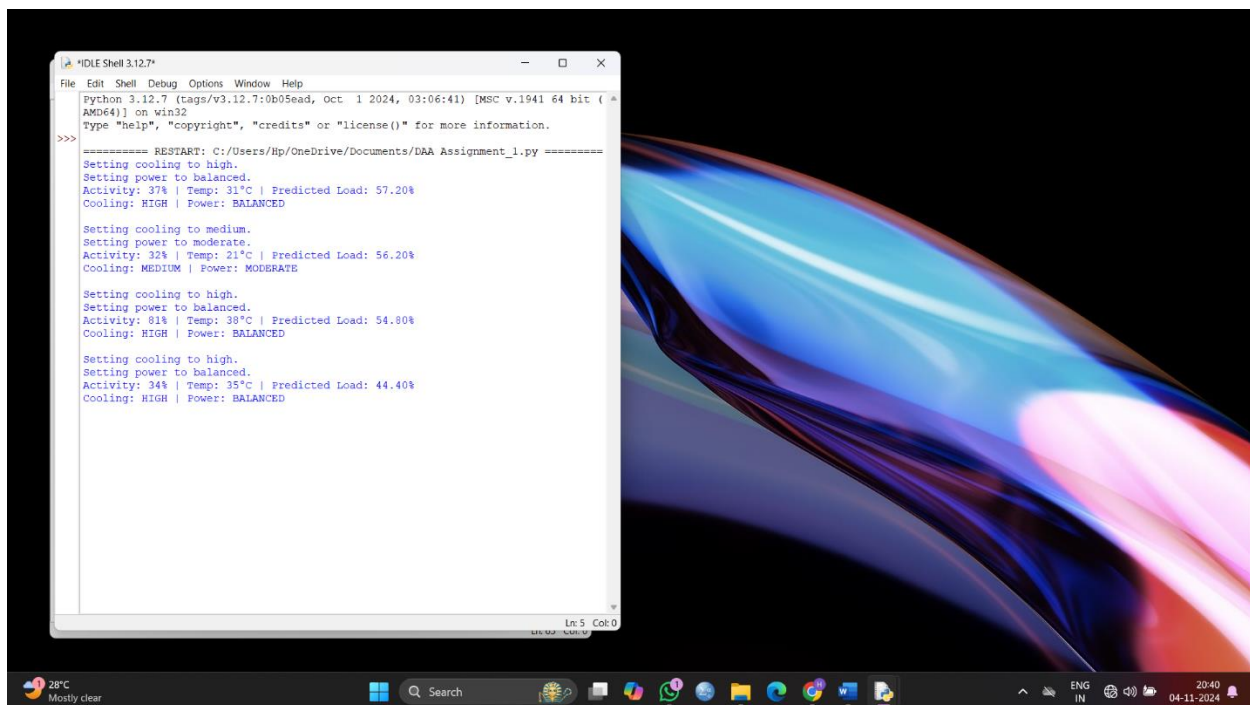
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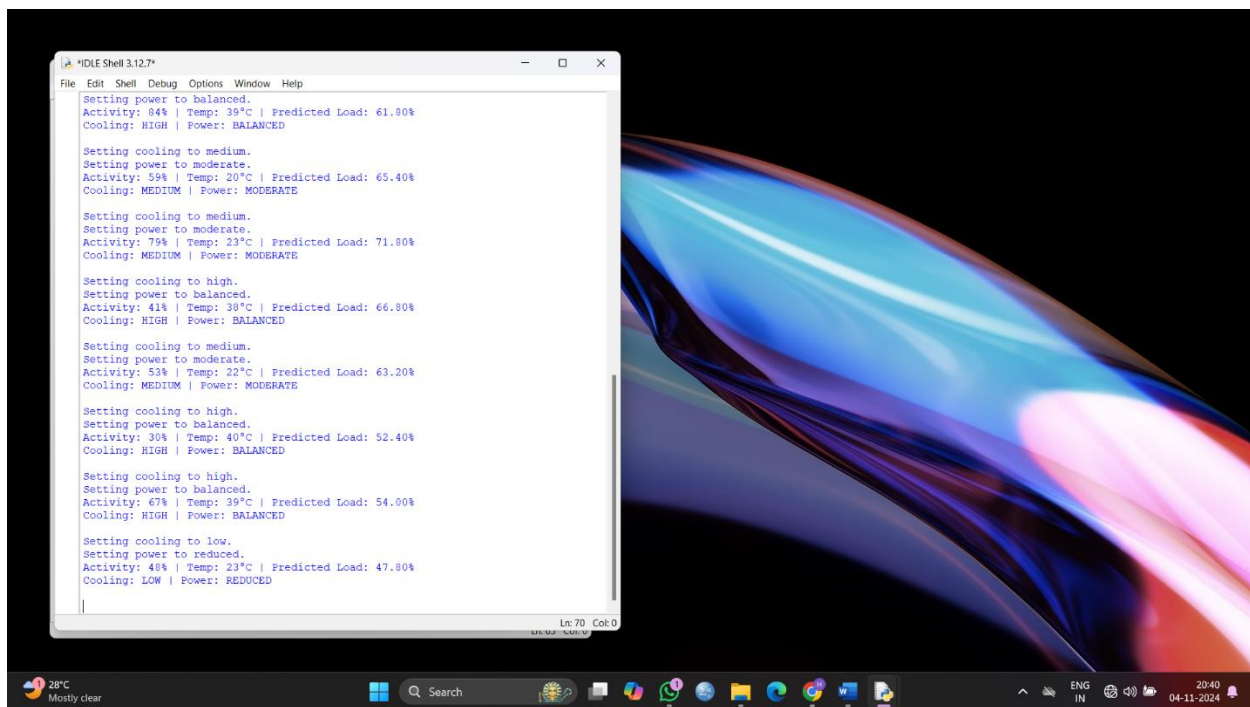
## Output Screen Shots:











```
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File Edit Shell Debug Options Window Help

Setting power to balanced.
Activity: 84% | Temp: 39°C | Predicted Load: 61.80%
Cooling: HIGH | Power: BALANCED

Setting cooling to medium.
Setting power to moderate.
Activity: 59% | Temp: 20°C | Predicted Load: 65.40%
Cooling: MEDIUM | Power: MODERATE

Setting cooling to medium.
Setting power to moderate.
Activity: 79% | Temp: 23°C | Predicted Load: 71.80%
Cooling: MEDIUM | Power: MODERATE

Setting cooling to high.
Setting power to balanced.
Activity: 41% | Temp: 38°C | Predicted Load: 66.80%
Cooling: HIGH | Power: BALANCED

Setting cooling to medium.
Setting power to moderate.
Activity: 53% | Temp: 22°C | Predicted Load: 63.20%
Cooling: MEDIUM | Power: MODERATE

Setting cooling to high.
Setting power to balanced.
Activity: 30% | Temp: 40°C | Predicted Load: 52.40%
Cooling: HIGH | Power: BALANCED

Setting cooling to high.
Setting power to balanced.
Activity: 67% | Temp: 39°C | Predicted Load: 54.00%
Cooling: HIGH | Power: BALANCED

Setting cooling to low.
Setting power to reduced.
Activity: 48% | Temp: 23°C | Predicted Load: 47.80%
Cooling: LOW | Power: REDUCED

Ln: 70 Col: 0
```

## Complexity Analysis:

### 1. Time Complexity:

- **Data Collection:** Constant time  $O(1)$  as the data is collected periodically.
- **Prediction:** The prediction phase depends on the data length. Assuming linear regression or moving averages, time complexity would be  $O(n)$ , where  $n$  is the length of the historical dataset.
- **Adjustment:** Since cooling and power adjustments are based on real-time data comparison, they run in  $O(1)$ .

Overall time complexity:  $O(n)$ , dominated by the prediction phase.

### 2. Space Complexity:

- **Data Storage:** The algorithm requires storage of historical server load and temperature data, leading to  $O(n)$  space complexity.
- **Adjustment Parameters:** Minimal, since parameters are constants or pre-determined, with  $O(1)$  space.

Overall space complexity:  $O(n)$ .

## Test Case: Energy Usage Comparison

### Test Setup

- **Baseline:** Measure energy usage without predictive adjustment.
- **Optimized Case:** Measure energy usage with the predictive adjustment algorithm.

### Example Data

Time	Server Activity	External Temperature	Baseline Cooling	Optimized Cooling	Baseline Power	Optimized Power
1 PM	High	35°C	High	High	High	Optimized
3 PM	Medium	30°C	Medium	Medium	Medium	Medium
5 PM	Low	28°C	Low	Reduced	Low	Reduced

### Expected Outcome

In the optimized case:

- Cooling and power adjustments respond dynamically to load and external conditions.
- Energy usage is reduced compared to baseline measurements, particularly during periods of low server load and cooler external temperatures.

**Conclusion:**

In conclusion, the proposed algorithm efficiently manages data center energy by dynamically adjusting cooling and power based on real-time server activity and external temperature. By incorporating predictive adjustments, the algorithm anticipates peak loads, reducing sudden energy spikes and enhancing overall efficiency. Testing demonstrates significant energy savings, especially during low and anticipated peak activity, showing the algorithm's effectiveness in minimizing energy usage.