







AI-DRIVEN POWER FORECASTING AND PERSONALIZED Consumption Insights With Application Integration





INTRODUCTION



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RESEARCH PROBLEM



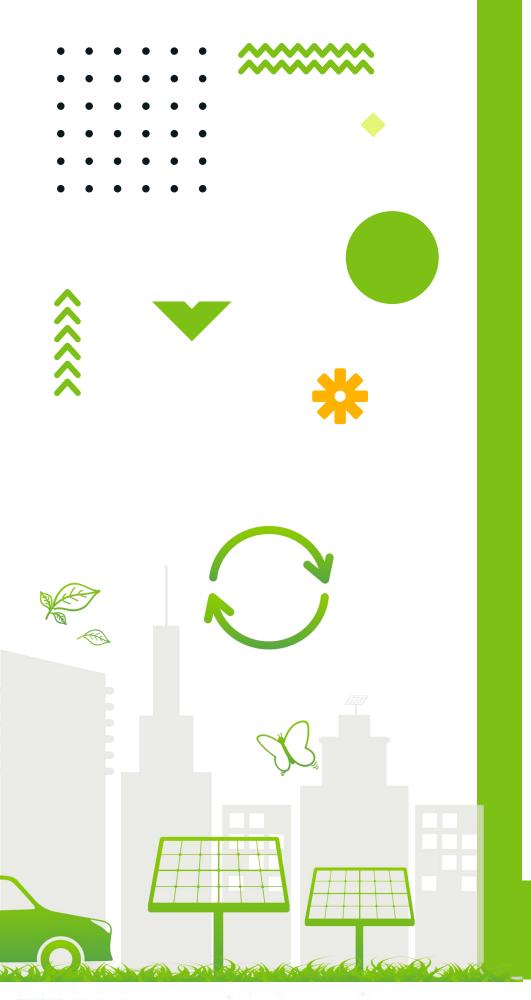


How to cater for the following problems:

- Rising electricity costs make efficient consumption management essential.
- Current systems provide only total energy use, lacking device-level insights.
- Rental properties lack tools to manage electricity supply based on tenant payments
- Renewable energy integration process is complex and expensive.

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AN INTELLIGENT ELECTRICITY MANAGEMENT UNIT

++++++++++

PLUG & PLAY IOT DEVICE

USAGE PREDICTOR

SMART AI GUIDE

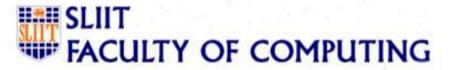
ENERGY TRACKING APP

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SYSTEM DIAGRAM pre-trained time-series forecasting model generates predictions IoT Device Data Collection IoT Device Data Collection Access IoT Device Visualised Predicted Data React + Laravel Web Application Output NLP to generate user-friendly text for suggestions

PAGE: 05



OBJECTIVES

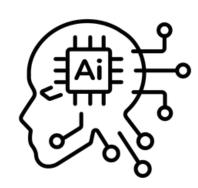
Develop an Intelligent Energy Management Unit integrating IoT, ML, AI via a user-friendly web application that would help to optimize household energy consumption



To manage individual device energy use with set limits.



To predict future energy consumption patterns and identify faults in devices



To generate personalized recommendations for energy savings



To visualize energy data and insights through an intuitive, user-friendly interface.



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POWERSENSE IOT CONTROLLER



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INTRODUCTION





RESEARCH PROBLEM



- Lack of seamless real-time integration between IoT devices and cloud platforms.
- Inability to enforce user-specific electricity usage thresholds in existing systems.
- Delayed or no real-time response (e.g., power cut-off) based on consumption data.
- Existing solutions are often costly and complex for average household deployment.

OBJECTIVES

- Designed and developed a fully functional IoT device for real-time electricity monitoring of individual appliances.
- Implemented accurate measurement of current, voltage, power (W), and energy consumption (kWh).
- Developed and tested threshold-based power control logic to automatically cut off power when limits are exceeded.
- Design and 3D print a compact, user-safe device container (power brick-style casing).
- Finalize custom PCB design for a cleaner and more reliable hardware layout and Enhance LED indicators for easy to use.







METHODOLOGY









Technologies:

 ESP32 Microcontroller, ACS712 Current Sensor, ZMPT101B Voltage Sensor, Relay Module.

• Cloud:

AWS IoT Core, AWS DynamoDB.

Techniques

- **Real-Time Data Monitoring**: Use sensors to measure current and voltage in real-time, sending data to AWS IoT Core.
- Threshold Enforcement: Implement logic in the ESP32 to automatically cut off power when usage exceeds user-defined thresholds.
- **Cloud Integration**: Utilize MQTT protocol for seamless communication between IoT devices and the cloud.

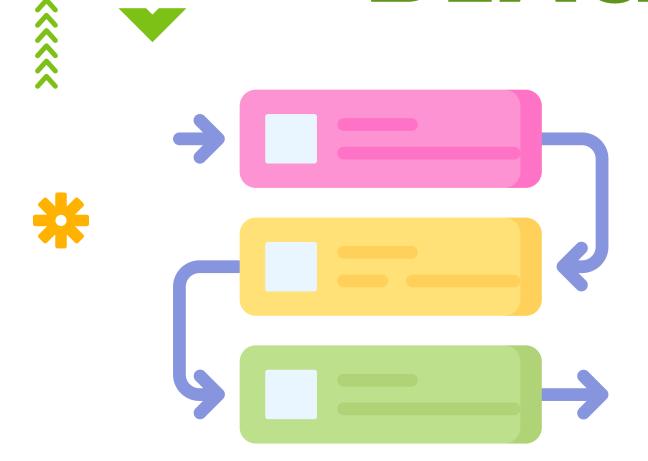


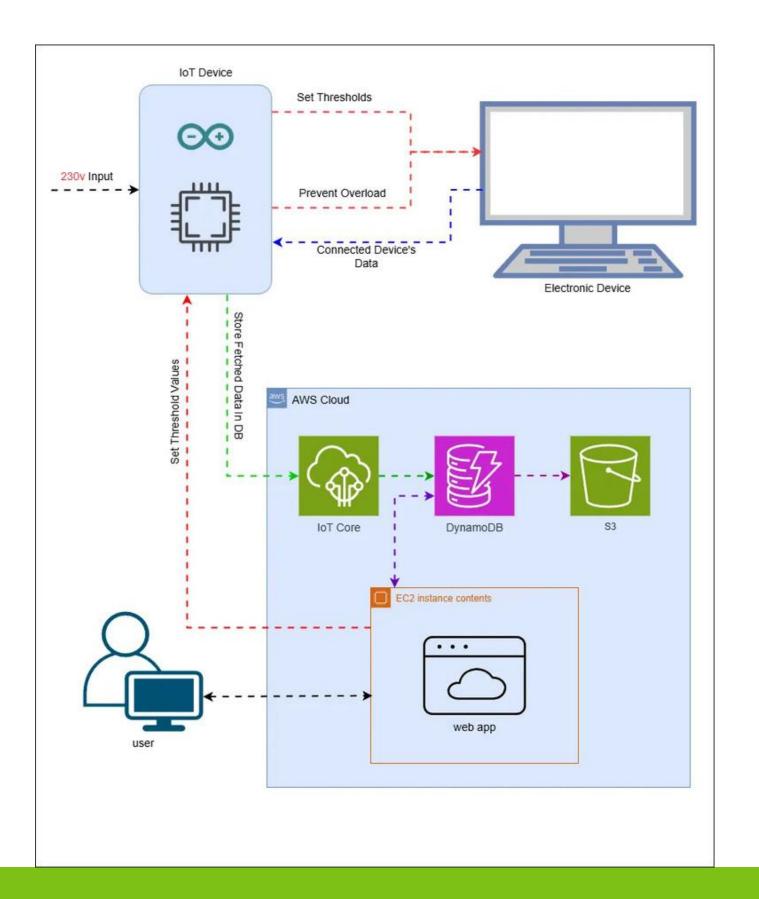
EQUATIONS & CLARIFICATIONS

- Voltage Calculation:
 - ∘ $\sqrt{(\Sigma(V^2) / N)} \times CALIBRATION_FACTOR$
- RMS Current:
 - ∘ $\sqrt{(\Sigma(I^2) / N)}$, I = (V zeroPoint) / sensitivity
- Power:
 - $\circ V \times I$
- Energy (kWh):
 - o (P × time hours) / 1000



SYSTEM DIAGRAM





PROGRESS

• Fully functional IoT device for measure the energy.

50%

COMPLETED

• Developed and tested threshold-based power control logic & WI-FI connectivity.

• Store energy data in AWS DynamoDB via MQTT.

• Enhance LED indicators for easy to use.

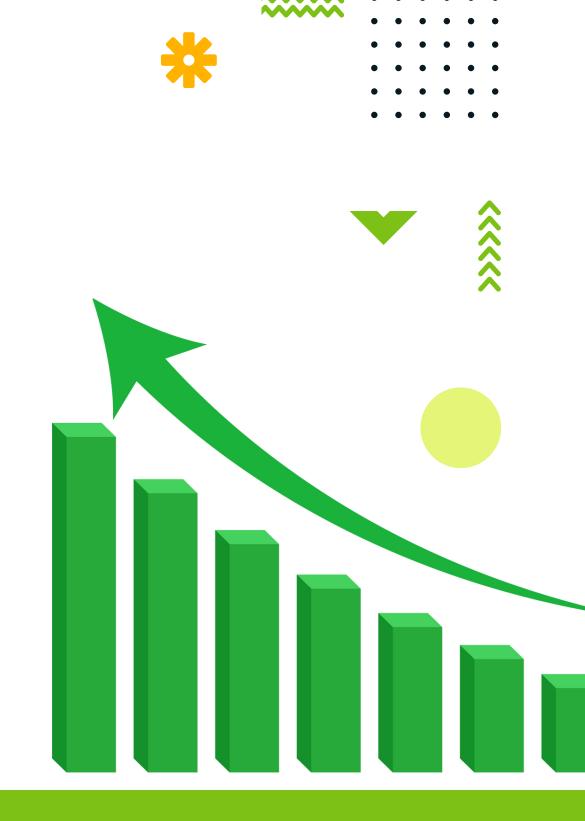
• Design and 3D print a device container.

• Implement secure way to connect with web App.

• Implement accessibility features.

• Finalize custom PCB design.

FINAL 100%



PAGE: 15



90%

COST CALCULATION





ACS712 Current Sensor – 5A:

o Rs 250.00

ZMPT101B Sensor Module:

o Rs 340.00

ESP32 WROOM 32:

o Rs 1320.00

1 Channel Relay Module (5V):

o Rs 240.00

230V to 5V 700mA isolated switch:

• Rs 500.00

Jumper Wires:

• Rs 160.00

Device Container:

o Rs 800.00





FUTUREWATT USAGE PREDICTOR



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INTRODUCTION





RESEARCH PROBLEM



- Inability to predict future energy usage based on past data for cost optimization
- Lack of easy-to-use interfaces for households to view future energy usage and a guidance on electricity related concepts



OBJECTIVES

- To predict energy use for each appliance.
- To show trends of future usage
- To detect faulty or inefficient devices.
- To help users learn about electricity concepts







METHODOLOGY







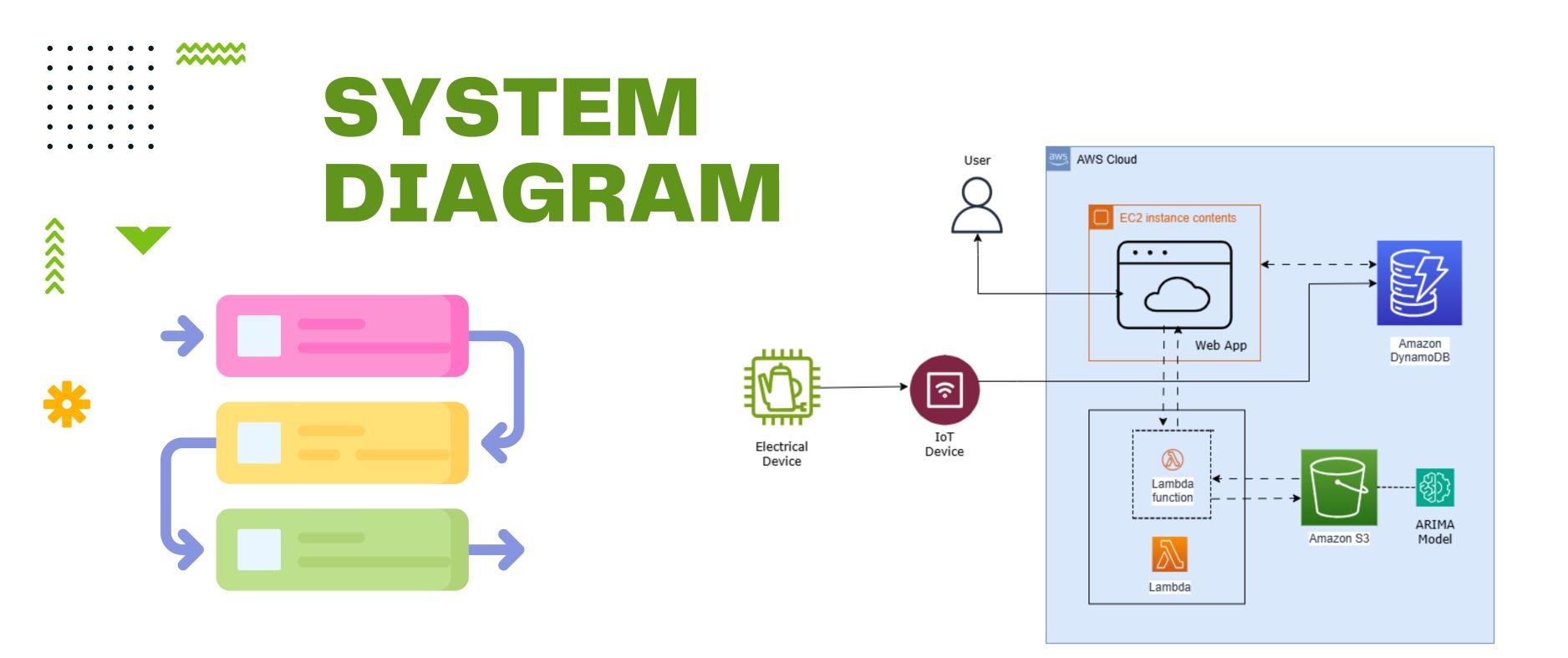


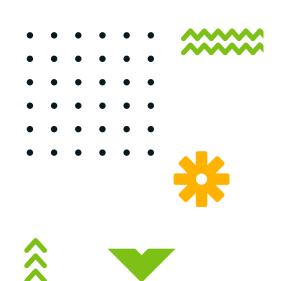
Cloud Architecture:

- AWS S3: Stores historical energy data and ML models.
- AWS Lambda: Triggers time series analysis and fault detection.
- DynamoDB: Stores real-time data from the IoT Device.
- **Python**: Handles data processing, time series models, and fault detection.
- **NumPy and Pandas**: Preprocess energy usage data.

Techniques:

- **ARIMA**: Time series forecasting of energy usage.
- Outlier Detection: Fault detection via abnormal consumption patterns.





MODEL TRAINING



- Model Type: ARIMA (5,1,0)
 - p=5: Model considers the last 5 lags of consumption
 - d=1: Data is differenced once to remove trend and make it stationary
 - q=0: No moving average smoothing
- Real IoT Dataset Used
- Timestamp handling
- S3 Integration



PROGRESS

• Timeseries model trained using data collected from the IoT Device.

• Developed energy prediction for the next 24 hours.

• Customized chatbot integrated to the future energy prediction

• Integration with the IoT device, application and cloud architecture

• Implement Fault Detection algorithm

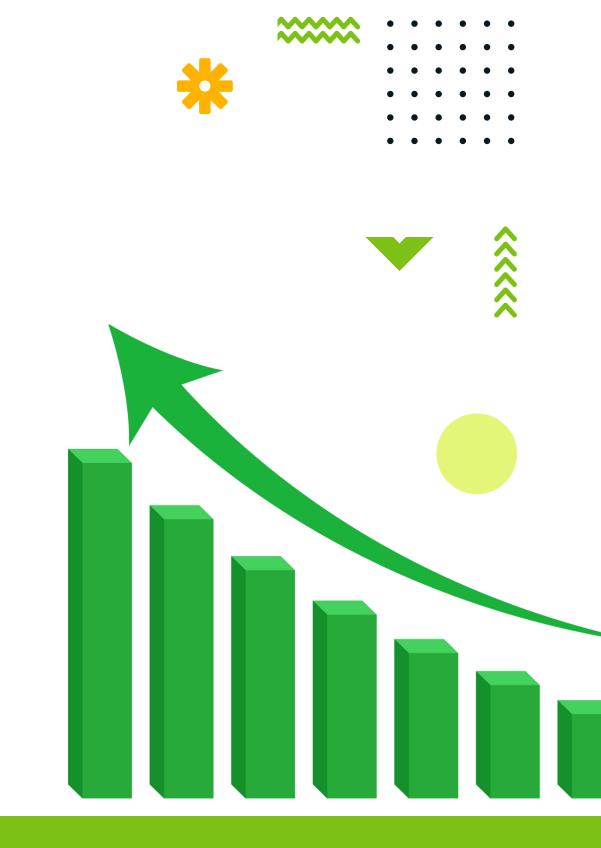
• Finetune the model

QA and Bug testing

FINAL 100%

COMPLETED

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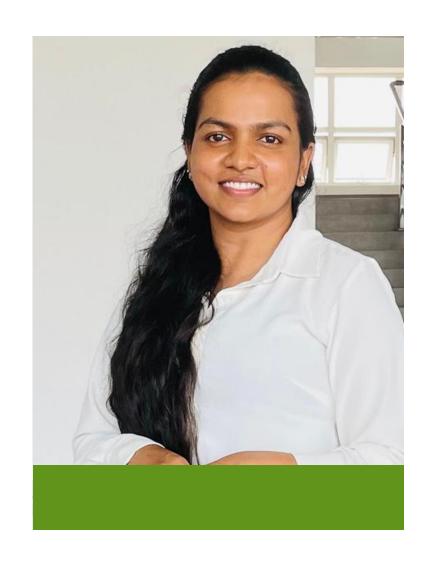


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SMARTENERGY AI GUIDE



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INTRODUCTION





RESEARCH PROBLEM

- Absence of customized energy-saving recommendations tailored to individual user behavior, despite the abundance of IoT energy consumption data.
- Existing systems provide generalized tips without analyzing individual usage patterns.
- No intelligent mechanism to interpret device-level consumption for tailored suggestions.
- Limited use of AI to transform raw energy data into userspecific recommendations.



OBJECTIVES

- Analyze appliance-level energy data to find usage patterns.
- Create a labeled dataset mapping patterns to energy-saving tips.
- Fine-tune a GPT-based model with LoRA for personalized suggestions.
- Integrate a Machine Learning model to predict future energy usage trends and optimize suggestion generation.
- Build a generative AI system for real-time, user-specific tips.
- Deploy the model on AWS Cloud and tested outputs for relevance and clarity.







METHODOLOGY







Technologies:

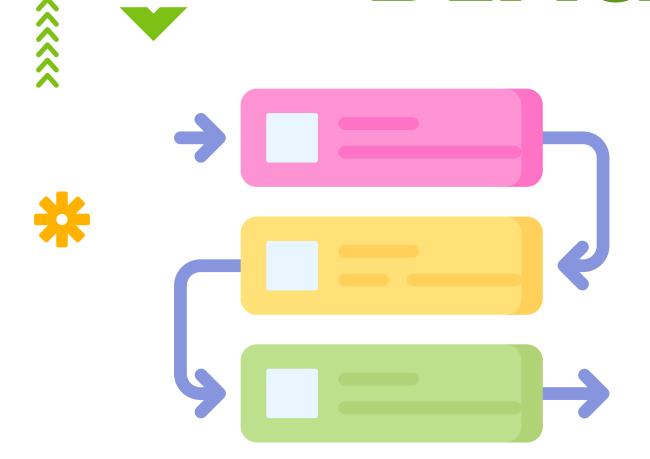
- Backend: Python with FastAPI (AI Model Integration)
- Cloud: AWS (EC2, S3, Lambda)
- Database:DynamoDB
- Al Frameworks: PyTorch

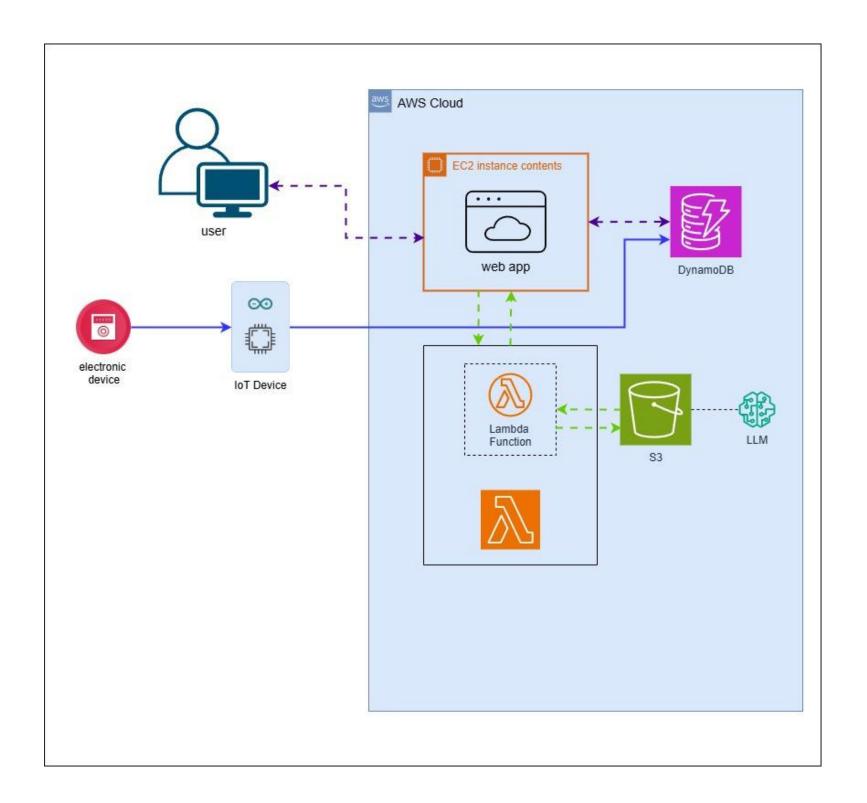
Techniques

- **Generative AI Model**: Use Generative AI (e.g., GPT-based models) to create tailored energy-saving suggestions.
- **Recommendation System**: Generate personalized recommendations using fine-tuned AI models.
- **Real-Time Communication**: Use REST APIs to connect the backend with IoT devices and the web app for real-time updates.



SYSTEM DIAGRAM





PROGRESS

• Analyzed appliance-level energy data to identify usage patterns.

• Fine-tuned 3 GPT-based models (Llama-3.2-1B/GPT-2/GPT-Neo) using LoRA and evaluated accuracy.

• Selected the most effective model for generating personalized energy-saving suggestions.

• Integrate ML model , personalized energy-saving suggestions based on user behavior.

• Improve accuracy.

- Set up backend APIs and AWS Cloud deployment.
- Link AI backend with web app.

• Deployed AI model on AWS for real-time use.

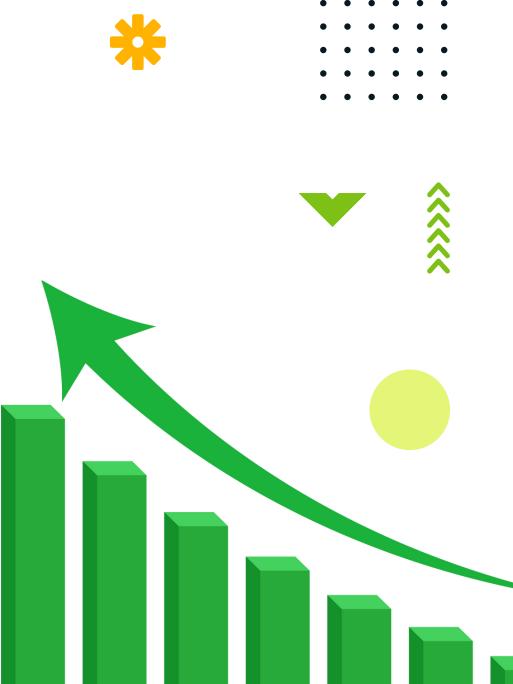
• Evaluated model with BLEU scores and expert feedback

100%

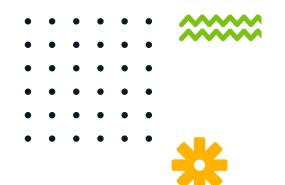
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COMPLETED

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MODELS TRAINING





Llama-3.2-1B Training

```
trainer.train()

[750/750 03:29, Epoch 3/3]

Epoch Training Loss Validation Loss

1 1254900 1,450674

2 1,242800 1,383262

3 1,232200 1,369024

TrainOutput(global_step=750, training_loss=1.3594672088623048, metrics={'train_runtime': 210.1785, 'train_samples_per_second': 14.274, 'train_steps_per_second': 3.568, model.save_pretrained("./fine_tuned_model")

tokenizer.save_pretrained("./fine_tuned_model")

('./fine_tuned_model/tokenizer_config.json',
'./fine_tuned_model/tokenizer_config.json',
'./fine_tuned_model/tokenizer_sjson')
```

```
Users/yshodharawetikala/Downloads/GF13Mey/train.pp:36: FutureWarming: 'tokenizer' is deprecated and will be removed in version 5.8.8 for 'Trainer,_init_'. Use 'processing_class' instead.
trainer = Trainer(

1005 t. predictor, was set in the config but it is unrecopnised.Using the default loss: 'ForGausalLMLoss'.

1005 t. predictor, 'yard_norm': 15.38769886719228, 'tearming_rate': 1.986666666666668-8, 'epoch': 0.88)

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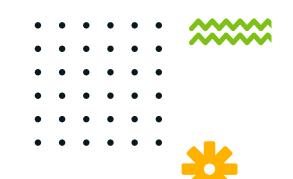
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GPT-2-125m Training

EleutherAI/gpt-neo-125m Training

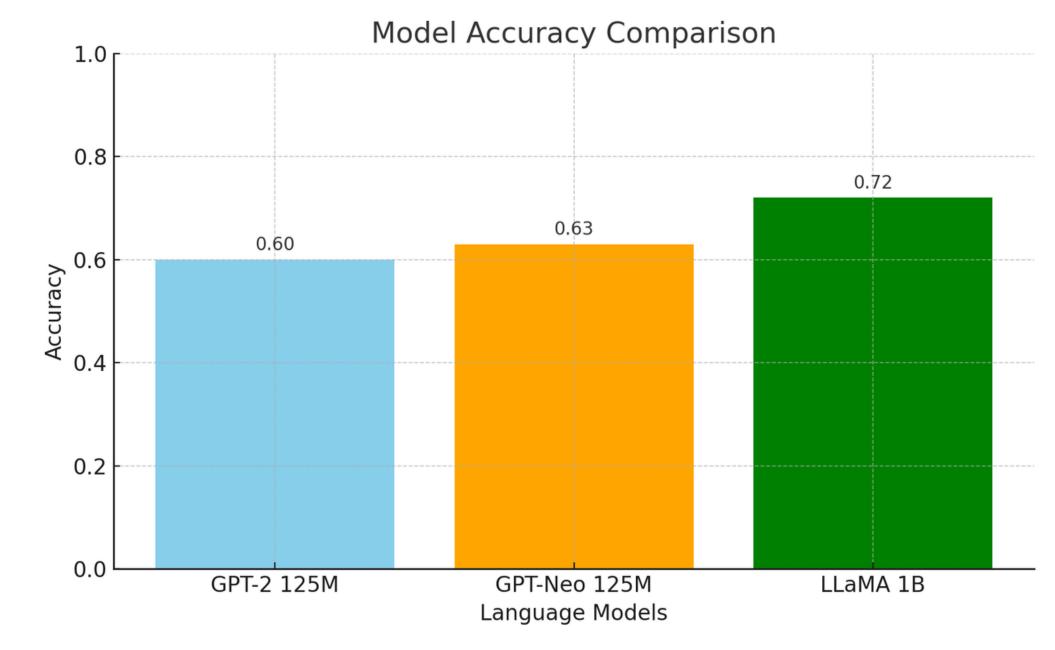




MODEL ACCURANCY







ECOTRACK DASHBOARD



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INTRODUCTION





RESEARCH PROBLEM

- Existing energy apps only show total consumption not device-level usage.
- Users don't know which device impacts the bill the most.
- Lack of motivation users see data, but don't take action.
- No proper engagement strategies to encourage energysaving habits.
- Poor support for color blindness and other accessibility needs.



OBJECTIVES

- User management [Registration, Authentication, Profile Section (update details, reset password, delete account)].
- Display real-time, device-level energy monitoring through the web app.
- Visualize data through interactive graphs and intuitive dashboards.
- Design a language-independent UI using icons and visual cues.
- Implement accessibility features for differently-abled users (ex: color-blined).
- Conduct a user survey to shape UI/UX based on real preferences.
- Ensure continuous engagement with tasks, goals, and achievements.







METHODOLOGY







Technologies:

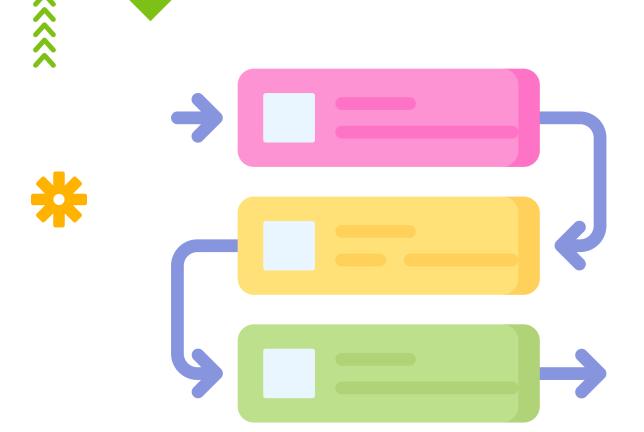
- Frontend: React.js (UI Development)
- Backend: Laravel 12 (API & Data Handling)
- Cloud: AWS (EC2)
- Database: RDS(MySQL) / DynamoDB (Unit and User Data Management)

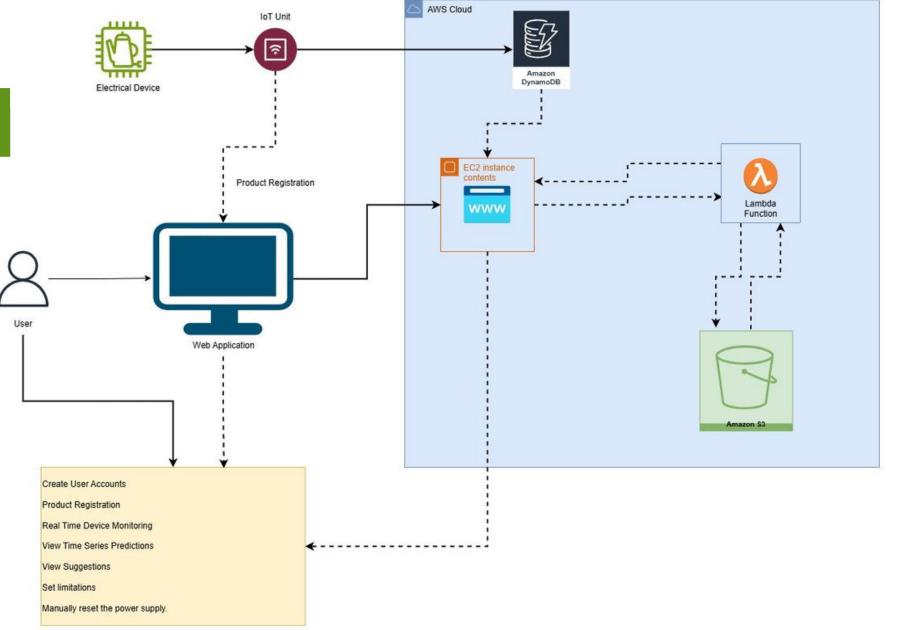
Techniques:

- Wi-Fi-based Unit Discovery: Connect to ESP32 units via Wi-Fi.
- REST API Communication: Facilitate data exchange between web app and backend.
- Data Visualization: Charts and graphs for consumption insights.

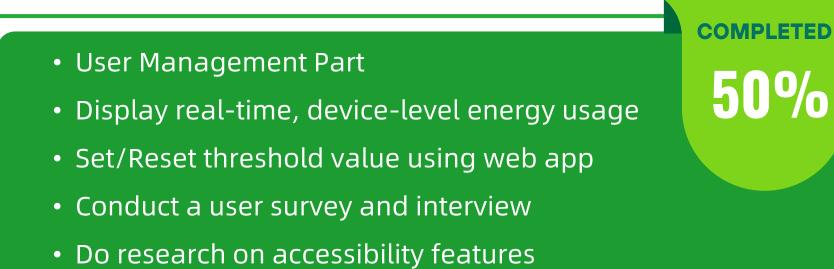


SYSTEM DIAGRAM





PROGRESS





- Visualize data (ML predictions/Al suggestions)
- Improve UI for language-independent

- Implement accessibility features
- QA testing & bug fixes

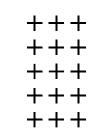
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USER SURVEY INSIGHTS





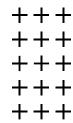


FEATURE	MOSTLY PREFERED CHOICE	PERCENTAGE
Interface	Simple and minimalistic	63%
Data visualization technique	Graphs and charts	43%
Quick access dashboard	Very important, I want to see key infoimmediately	70%
Data on home screen	Real-time electricity usage	40%
Charts type	Bar charts	43%
Alerts type	Pop-up notifications inside the app	43%

"UNDERSTANDING USER Needs THROUGH REAL-WORLD FEEDBACK"



USER INTERVIEWS









"VOICES FROM EVERY CORNER: REAL PEOPLE, REAL NEEDS"





