



# **Smart Energy consumption using Big Data**

## **Minor Project Presentation (IA-1)**

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# Introduction

- Growing urbanization → Rising energy demand
- Traditional monitoring = limited & inefficient
- Big Data + AI = Real-time monitoring & prediction
- Tools: Apache Spark, Hadoop, Kafka, Cloud, ML/DL
- Benefits: Cost savings, grid reliability, carbon footprint reduction

# Problem Statement

- Huge & complex energy consumption patterns
- Traditional forecasting fails at:
  - Handling large-scale smart meter data
  - Real-time demand prediction
  - Anomaly/cyber-attack detection
  - Renewable energy integration
- Consequences → Grid instability, high costs, energy wastage

# Objectives

- **Monitor & process** smart meter/IoT energy data in real time
- **Predict future consumption** using ML/DL models
- **Detect anomalies** to ensure grid reliability & security
- **Enable renewable integration** and track carbon footprint
- **Visualize insights** through dashboards (Grafana / Power BI / Tableau)

# Related Works (Literature Survey)

- **Big Data Pipelines**

- Real-time processing of massive smart meter & IoT data enables fast **forecasting & anomaly detection** → we use similar pipelines for energy data handling.

- **Early Energy Analytics** → Showed how raw consumption data can be turned into **actionable insights** → forms the base of our system design.

- **ML for Forecasting & Fault Detection** → Proved effectiveness of ML in **predicting demand, detecting faults, and optimizing control** → directly applied in our consumption prediction module.

- **Interpretable ML** → Highlighted the need for **trustworthy, explainable predictions** → we ensure transparency in our model outputs for decision-making.

- **Carbon Tracking** → Demonstrated near real-time carbon intensity & emission monitoring → we extend this by integrating **energy + carbon analytics** in one framework.

# Proposed Technology/Methodology (Planned)

- Smart Meters + Kafka:**

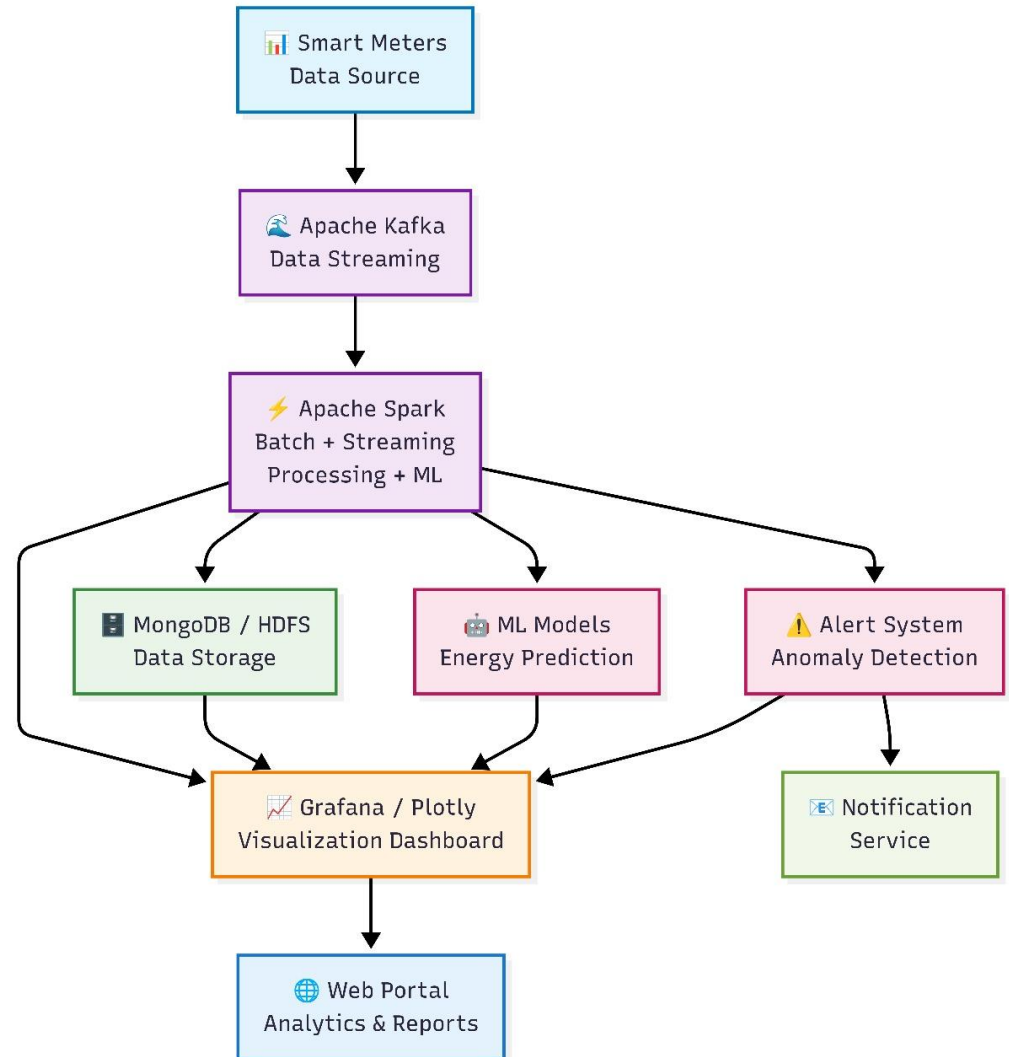
Real-time energy data streaming.

- Spark ML Processing:** Batch + streaming with predictive models.

- Data Storage:** MongoDB / HDFS for scalable storage.

- Dashboards:** Grafana/Plotly for visualization & anomaly alerts.

- Web Portal:** Analytics, reporting, and notification services.



# Progress (Till IA – 1)

- **Scope Finalization** – Studied the real-world problem and defined objectives clearly.
- **Literature Survey** – Reviewed existing research papers on smart energy systems, ML for efficiency (15–30%), and real-time carbon tracking frameworks.
- **Data Understanding** – Identified potential datasets (building energy, HVAC, grid data, renewable integration cases).
- **Blueprint / Dummy Model** – Built an initial dummy model to test interdependencies among parameters (kept as a baseline reference).
- **Planning** – Divided responsibilities (data collection, modeling, system framework, reporting).
- **Documentation Work** – Prepared initial draft with problem statement, objectives, and literature review findings.

# Conclusion

Significant progress has been made:

from literature study → dataset collection → cleaning → development of a basic linear regression model.

**Next Steps:** Implement advanced ML models, integrate optimization techniques, and design the framework for renewable energy and real-time carbon tracking.

**Expected Outcome:** An efficient, ML-driven energy management system capable of delivering **15–30% energy savings** while supporting sustainability through automated optimization and carbon monitoring.



# Roadmap for IA-2 Evaluation

**Week 5** – Feature engineering & exploratory analysis

**Week 6** – Model planning with evaluation metrics

**Week 7** – Advanced ML model implementation (baseline forecasting)

**Week 8 (IA 2)** – Framework design & preliminary results

**Week 9** – Model refinement & hyperparameter tuning

**Week 10** – Optimization/control integration (MPC/RL baseline)

**Week 11** – Testing & validation on complete dataset

**Week 12 (Final)** – Final report & presentation preparation

# References

- Mujeeb, S., et al. *Review of Big Data Analytics for Smart Electrical Energy Systems*. Energies, 2023.
- Amasyali, K., & El-Gohary, N. *Machine Learning for Smart and Energy-Efficient Buildings: A Review*. Environmental Data Science, 2023.
- Afram, A., & Janabi-Sharifi, F. *Artificial Intelligence Approaches to Energy Management in HVAC Systems – A Review*. Buildings, 2025.
- Ma, Y., et al. *Field Demonstration of Model Predictive Control in Building HVAC Systems*. Applied Energy, 2022.
- Killian, M., et al. *Lessons Learned from Field Demonstrations of RL/MPC in Buildings*. arXiv preprint, 2025.
- Zhang, X., et al. *Near Real-Time Carbon Accounting Framework for Power Systems*. Electric Power Systems Research, 2024.
- Zhao, Y., et al. *Real-Time Industrial Carbon Emission Estimation with Deep Learning*. Engineering Applications of Artificial Intelligence, 2024.