

Predictive Modeling of Battery Powered Vehicles (BPVs) Charging Load on the Power Grid: A Machine Learning Approach

Thesis Proposal submission info

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1. Motivation, Existing Systems & Research Gaps

1.1 Motivation

- Battery Powered Vehicles (BPVs) are becoming a central component of global efforts toward sustainable transportation.
- In Bangladesh, particularly in expanding cities, the adoption rate of BPVs is rising rapidly.
- This growing adoption is contributing to increased electricity demand, especially from BPV charging.
- Accurate forecasting of BPV charging load is essential for:
 - Ensuring grid stability
 - Avoiding overloading or underutilization of power supply
 - Preventing power imbalances and system stress
 - Promoting efficient energy planning

1.1 Motivation

- Without predictive models, utilities may misjudge supply needs, leading to:
 - Inefficiencies in generation and distribution
 - Higher operational costs
 - This research proposes a machine learning-based model to forecast BPV charging load at the city level.
 - The objective is to provide decision-makers with data-driven tools for smart power grid management.

1.2 Existing Systems & Research Gaps

Reference	Journal Metrics	Existing Systems/Methods	Research Gaps	Key Contributions
Vulfovich et al. (2021)	Journal: Simulation Modelling Practice and Theory Quartile: Q1 Impact Factor: 4.719 (2022)	<ul style="list-style-type: none">- Modified First Harmonic Approximation (MFHA) modeling- SN-compensated inductive power transfer links- Load-independent-voltage-output frequency systems	<ul style="list-style-type: none">- Focused on component-level analysis rather than system-level- Limited scalability for grid integration- Insufficient consideration of multiple charging scenarios	<ul style="list-style-type: none">- Advanced modeling of power transfer links- Load-independent voltage output techniques- Simulation frameworks for inductive charging
Schreiber & Ulbig (2023)	Journal: Energy and AI Quartile: Q1 Impact Factor: 7.762 (2022)	<ul style="list-style-type: none">- Model selection, adaptation, and combination- Transfer learning for renewable forecasts- Wind and PV power prediction	<ul style="list-style-type: none">- Focus on generation rather than load prediction- Limited application to BPV charging load profiles- Insufficient integration of charging behavior patterns	<ul style="list-style-type: none">- Transfer learning techniques for power systems- Model adaptation methodologies- Renewable energy integration framework

1.2 Existing Systems & Research Gaps

Reference	Journal Metrics	Existing Systems/Methods	Research Gaps	Key Contributions
Yaprakdal (2023)	Journal: Energies Quartile: Q1 Impact Factor: 3.252 (2022)	<ul style="list-style-type: none">- Ensemble deep learning for hour-ahead load forecasting- Feature selection approach- Comparative analysis with state-of-the-art methods	<ul style="list-style-type: none">- Not specifically focused on BPV charging loads- Limited temporal resolution (hour-ahead only)- Lack of spatial distribution consideration in grid impact	<ul style="list-style-type: none">- Novel ensemble deep learning architecture- Effective feature selection methodology- Comparative benchmarking of forecasting models
Klausmann et al. (2023)	Journal: Energy and AI Quartile: Q1 Impact Factor: 3.252 (2022)	<ul style="list-style-type: none">- Ensemble deep learning for hour-ahead load forecasting- Feature selection approach- Comparative analysis with state-of-the-art methods	<ul style="list-style-type: none">- Not specifically focused on BPV charging loads- Limited temporal resolution (hour-ahead only)- Lack of spatial distribution consideration in grid impact	<ul style="list-style-type: none">- Novel ensemble deep learning architecture- Effective feature selection methodology- Comparative benchmarking of forecasting models etc.
Crespo & Aliberti (2024)	Journal: Engineering Applications of Artificial Intelligence Quartile: Q1 Impact Factor: 8.036 (2022)	<ul style="list-style-type: none">- Comparative analysis of ML techniques- Short-term grid power forecasting- Uncertainty analysis for wave energy converters	<ul style="list-style-type: none">- Focus on wave energy rather than BPV charging- Limited consideration of charging behavior variability- Insufficient integration with charging infrastructure	<ul style="list-style-type: none">- ML comparison framework for power forecasting- Uncertainty quantification techniques- Short-term grid power prediction methods

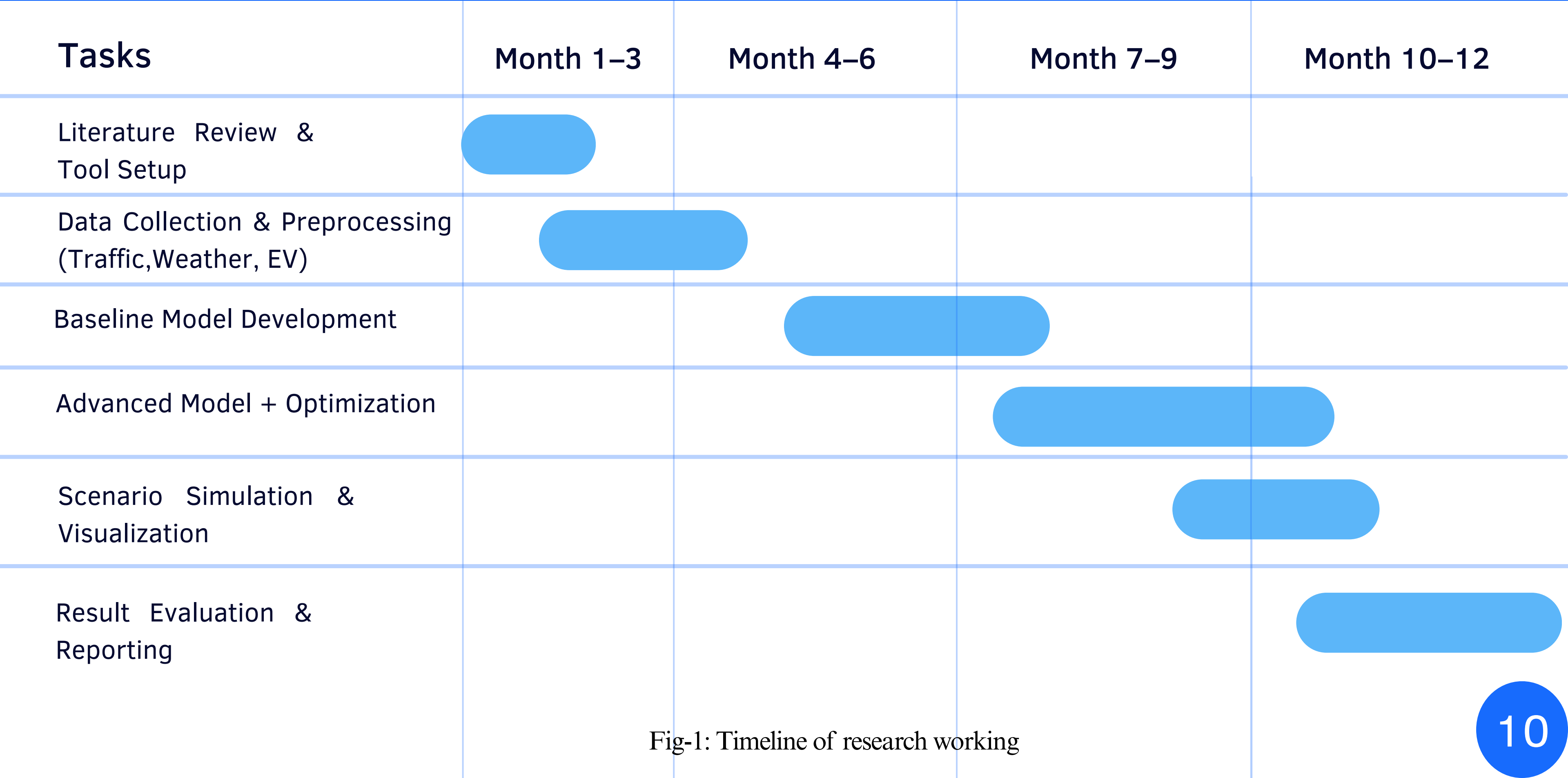
2. Objectives

- Design a machine learning model to forecast BPV charging load in Bangladeshi cities (Specially in Khulna, Jashore etc).
- Use a combination of synthetic datasets (traffic, temperature, trip patterns).
- Compare performance of various models: Linear Regression, Random Forest, XGBoost, LSTM, Quantile LSTM . . . etc .
- Simulate different demand scenarios (rush hours, weekends, seasons).
- (Optional) Visualize predictions and usage patterns .

3. Technology / Tools

- **Languages/Platforms:** Python, Jupyter Notebook, Google Colab
- **Visualization:** Matplotlib, Seaborn, OpenStreetMap (OSM)
- **Traffic Simulation:** SUMO (Simulation of Urban Mobility), TraCI framework
- **Communication Modeling:** OMNeT++ (for BPV-grid data exchange simulation)
- **Big Data Processing:** KNIME (optional, for visual workflow and preprocessing)

4.Time Line



5. References (Selected Q1/Q2 Journals & Papers)

1.Klausmann et al. (2023) - Energies (Q1)

Adaptive control for battery storage with peak load limitation. [Citations: 11]

2.Yaprakdal (2023) - Energies (Q1)

Ensemble deep learning for hour-ahead load forecasting with feature selection. [Citations: 19]

3.Vulfovich et al. (2021) - Simulation Modelling Practice and Theory (Q1)

Modified harmonic approximation modeling for inductive power transfer links. [Citations: 14]

4.Schreiber & Ulbig (2023) - Energy and AI (Q1)

Transfer learning for renewable energy forecasting; a focus on generation. [Citations: 23]

5.Crespo & Aliberti (2024) - Engineering Applications of Artificial Intelligence (Q1)

A comparative study of machine learning for short-term grid forecasting with uncertainty analysis. [Citations: 9]

6.Suggestions / Any question

If you have any question ,
you may ask .

Thank You