

Data Driven Energy Economy Prediction for Electric City Buses Using Dynamic Network based Machine Learning

Jegadeesan R¹, Midhun chakkaravarthy², Mudassir Khan³

¹Research Scholar, Lincoln University College Malaysia, ²CSE-Lincoln University College Malaysia,

³Computer Science and Engineering-King Khalid University, Saudi Arabia

Email ID: ramjaganjagan@gmail.com , midhun@lincoln.edu.my, mkmiyob@kku.edu.sa

Abstract: It looks like using electricity to power transportation systems, especially city buses, has a lot of potential. For fleet management and vehicle design to work well, they need to fully understand real-time driving data. Alternative power trains need to be carefully looked at from many technical points of view before they can work at their best. As the need for energy can change quickly, a careful design is used to keep costs low and energy waste to a minimum. Because the problem is so complicated and has many sides, businesses and researchers have not tried to find analytical solutions. If you do operational optimization right, accurate predictions of energy demand will save you a lot of money. In the energy sector of battery electric buses (BEBs), the goal of this study is to make things more open and clear. New machine learning methods were used to get a better understanding of speed profiles by using different sets of variables that could provide different explanations. Five different methods were used to do a full evaluation of the predictions' overall usefulness, reliability, and accuracy. The models used made predictions that were more than 94% accurate, which is great and in line with a careful process for choosing features. The suggested method gives manufacturers, fleet operators, and cities environmentally friendly public transportation options. It will also make people's travel habits very different.

Keywords: Battery electric buses, Machine learning, Electric-vehicle, Battery electric vehicles, Particle-based Dynamic Bayesian Networks.

Introduction

About 25% of all greenhouse gas (GHG) emissions in Europe come from cars and commercial vehicles [1]. At this rate, progress is being made. Getting all transportation powered by electricity would be a great way to fight climate change and promote sustainability [2, 3]. In cities, electric buses are the most environmentally friendly way to get around. Due to their amazing efficiency improvement of up to 77% [5], electric vehicles have much lower overall and operational costs than vehicles powered by internal combustion engines. So, the cost of electricity at first isn't very important because it quickly pays for itself. Even so, an electric bus (BEb) can cost twice as much as a diesel bus [4]. Putting electricity into the power train has many benefits, such as lowering noise and pollution. While diesel bus batteries can be charged much faster, electric bus batteries can only be charged much slower. However, this statement is wrong when you look at their sizes. However, more research is needed to find the best way to incorporate electricity into the transportation industry, despite the many problems that exist. However, it is a very effective way to clean up the environment and lower greenhouse gas emissions. This project was made because people in Seville were worried about the people who run the public buses. They wanted to switch from using diesel cars to using electric cars as their main goal. First, they had to find

the best charging stations in cities and the batteries with the most power. Because of this, it is necessary to use computers to estimate consumption for each route [12]. At this time, this restriction applies to models that have a complex physical structure and need a lot of time to compute, or models that are based on data and have a lot of mechanical, driving, and road measurements. Nevertheless, these data-driven models need a lot less computing power after they have been trained (as explained in Section I-A). This project's main goal is to do the following: Estimates of how much energy the new electric buses use are made using data-driven and physics-based models from the operator's database. Compared to other data-driven methods, our method is easier to understand because it uses machine learning to accurately predict consumption. You know enough about the bus's speed and number of passengers to get by.

Literature Review

[1]The global electric-vehicle market is amped up and on the rise: With the sale of ten million self-driving cars in 2017, the market is likely to see a big jump. In 2021 and 2022, about 14% of recently bought cars had batteries that could be charged. This is a big number because the usage rate is only expected to be 9% by 2100 and even less than 5% in 2019 and 2020. Global shopping was dominated by a number of markets. With more than 60% of the global market share, Asia stayed the biggest market for self-driving cars. Asia brings in more than half of the electric cars that are on the road today. Already, more power-generating cars have been sold in the area than were expected for the next year. Italy has the world's second-largest real economy. Between 2021 and 2022, its income from gasoline-powered cars grew by almost 15%, showing how important that industry is to the country's economy. After that, electric cars made up almost twenty percent of all vehicle sales. The third-most expensive fully electric car cost 55% more in 2017 than it did in 2016. At the moment, we have an 8% revenue share.

[2]**Gasoline compression ignition approach to efficient, clean and affordable future engines:** While the demand for alternative fuels is growing quickly, gasoline will still meet a large part of the world's transportation fuel needs (about 90%) once it is available. The main reason for the rise in fuel use during travel is the use of oil in commercial cars, which is similar to how low-octane cigarettes can also lead to too much. Modern diesel vehicles aren't too expensive, but they are hard to handle because they need to be maintained and regulated. The main reason for this is that hydrogen and other harmful chemicals are harder to control when diesel engines are running because their ignition temperatures are higher. Power tools that make small amounts of nitrogen oxides (NOx) and particulate matter (soot) over time can run on gasoline. About two-thirds of the best gasoline needs to be of a higher octane type in order for fuel compression ignition to work. Without meaning to, we ate the extra low-octane ingredients. It may also have a higher boiling point than the current octane increase.

[3]**Financial analysis of battery electric transit buses,” National Renewable Energy Laboratory:** A standard van ship and an electric battery carriage asset type were used in the first display of standard establishment. Finally, after a lot of thought, fully electric frigates (BEB) were put into service. Based on a first look at the numbers, the ships might be paid for in full after 3.3 months, given that they are worth about \$785 each. There was a $\pm 50\%$ difference in the outcomes of the 33 basic factors when looking at how they affected winning or losing. The uncertainty of the variable is found by dividing the data by the standard value. Before fleet owners decide if electric cars (EVs) are a good investment option compared to other options, they should carefully look at the important and changing factors.

System Analysis

Existing System

A lot of research has been done on electric hybrid buses (BEBs) and battery electric vehicles (BEVs). As shown in [13], electric bus routes are more resistant to changes in objective characteristics than gasoline

bus routes. They also perform better than cars, which makes sense. This robot manipulator's road conditions and switching frequency are very different from those of a person whose attention shifts more often, because it cares more about seat load than about timelines, paths, and social connections with other robot manipulators. Most of these studies, like earlier ones, are based on complex thermodynamic models and structures. Each person has their own focus and goal. Pages 14–21 are what the book is about. The authors of reference [14] look at how much energy battery electric vehicles (BEVs) use and how engine and transmission efficiency, emergency power, and friction force affect this. Since most public buses go slower than 50 km/h, they depend on an emergency power source a lot. On the other hand, how well the power train works and how well it can handle wind are important parts of how a car normally works. To make accurate predictions about energy costs, you need to know a lot about auxiliary power. Considering that the investigation is about a cow... Use information approaches and an automotive industry modeler to look into and rate the connection between a vehicle's power needs and its motion parameters [15]. Temperature of the air and distance are often used to measure motion. The length of time that the movement lasts has not been thought about, though.

Proposed System

Particle-based Dynamic Bayesian Networks (DBMs) are used for this by people in the shipping business. To guess how much energy the next cars will use, we look at the upcoming hybrid cars using advanced knowledge models. The material of the previous post gives a different point of view on the list of observable elements, among other things. Through the use of computer vision, it also shows how to accurately predict how it will be used along a certain route. To put it simply, one only needs to know how many people the vehicle can carry and how fast it goes. There are three different parts that make up efficiency. We figure out how much power the car needs for each road by using a well-known thermodynamic framework. Also, the density estimate isn't very accurate because it depends on things like the bus's total weight and what's inside it. The factors start in a database owned by one of the owners.

Accelerate Notification has a wide range of gadgets that can be used for both time and frequency-based tasks. Van product indicators will be measured and computer vision models will be trained to correctly estimate energy consumption in order to find the most important parts. Finding out that self-motivation is the most important thing for everyone is interesting. It has become less important to study frequency, electron density, and kinetic energy in this area of study over the past few years.

System Design System Architecture

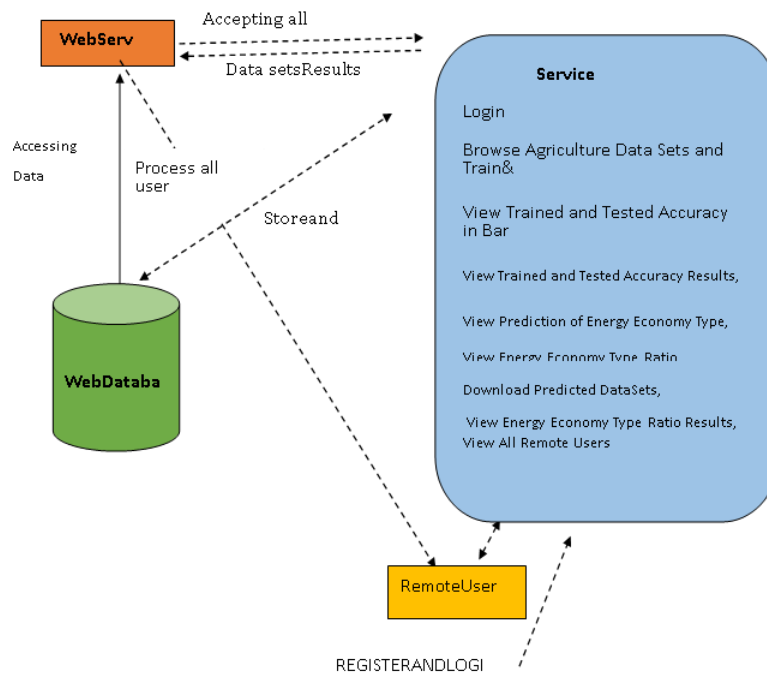


Fig 1. Data Flow Diagram

Data Flow Diagrams : The main word they use is "data flow diagram," which is also written as "DFD." The way that data is organized in a database shows how data moves through a system or process. Also, it is common for independent contractors to use the same tools and follow the same rules. There are no channels, sequences, or decision trees in the data flow diagram. The main goal of this graphic tool is to make it easier for managers, employees, and customers to talk to each other and work together. This method makes it easier to evaluate current methods and system designs.

CLASS DIAGRAM: The picture shows a "class diagram" with a single class name, different combinations, and different levels of abstraction in how they work together. A lot of the time, these charts are built into diagrams and act as three-dimensional structures. A schematic map and a categorical diagram are often used to show how each image fits into the bigger picture. It gives a detailed plan for how to use each method. The definition of a group is a collection of things that have similar traits and goals and talk to each other regularly using established phonetic rules. It can only quickly put things into groups, but not give a clear definition of an object. Interactive classes can help people make connections at first, but they can also get in the way of further growth. Some college courses, on the other hand, are made to appeal to both intellectual and romantic interests. It shows how much the project manager loves talking about value systems in college settings and how good she is at making lasting connections with first responders through the design. List all the alliances that existed before and after the term of the alliance in question. With this complete guide, you'll be able to show how different objects can be used. Make a list of the things that need to be done right away after this group project is over. Conducting a more thorough analysis of their respective duties and obligations is the most effective way to establish a link between the two courses.

Implementation

Modules Description

Service Provider

The person in charge of the network has to enter the user's ID and password for this device. Within a week of getting an active user account, John showed how well he could do important tasks like analyzing data points. Besides that, he made a framework for a subway system and gathered a lot of data. He used perspective-based analysis to accurately interpret charts by using both facts and his own opinion. He also learned more about the structure as a whole and made models that could predict what would happen with different parts of the electrical grid. He also used commonly available data and a perspective-based approach to look at energy market trends and predict what would happen in the future. Besides that, he gave each user a certain attitude.

Remote User

This feature is used by a lot of people. It is important to finish the registration process before starting any transaction. The first set of user data will be added to the dataset. In less than a week after the license was activated, William had to use someone else's login and password. When users create an account, they have to enter their password and learn how to use the electric grid prediction form before they can use it for the first time.

A Python web framework called Django aims to speed up and simplify the process of designing and building websites. This software, which was made by experienced developers, takes care of many difficult parts of web development quickly and easily, so you can focus on the design of the app instead of its initial development. It's free and easy to change to fit your needs.

The main goal of Django is to make it easier to make complex websites that use databases. Django puts a lot of weight on the idea of "pluggability," which means being able to reuse parts and make development go faster. It also tries to keep useless tasks from happening again for no reason. The Python programming language is used to build the whole system, including the data models and configuration files. Django also has an optional feature called an administrative interface that is created dynamically using introspection and managed by administrative models. Making objects, reading them, updating them, and deleting them is easier with this interface.

Results

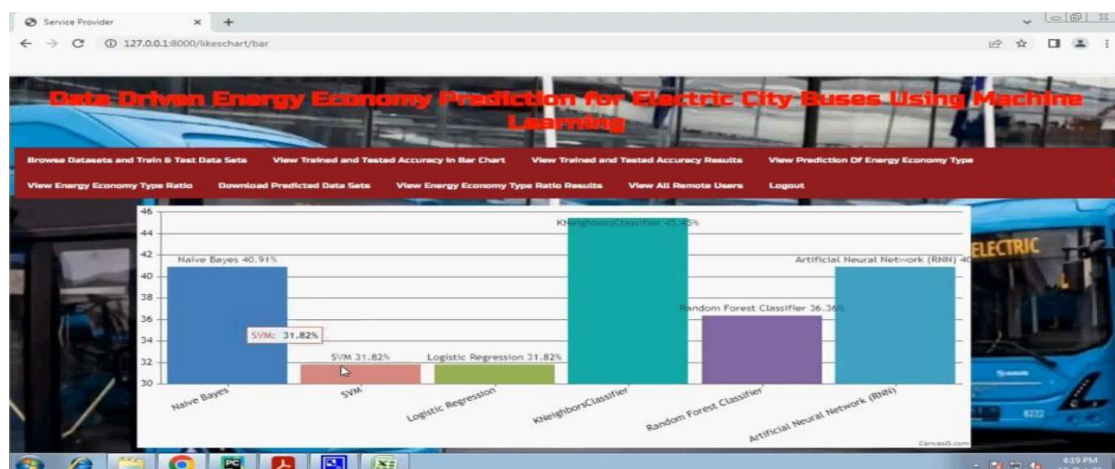


Fig 4.1 Trained and Tested accuracy in bar chart

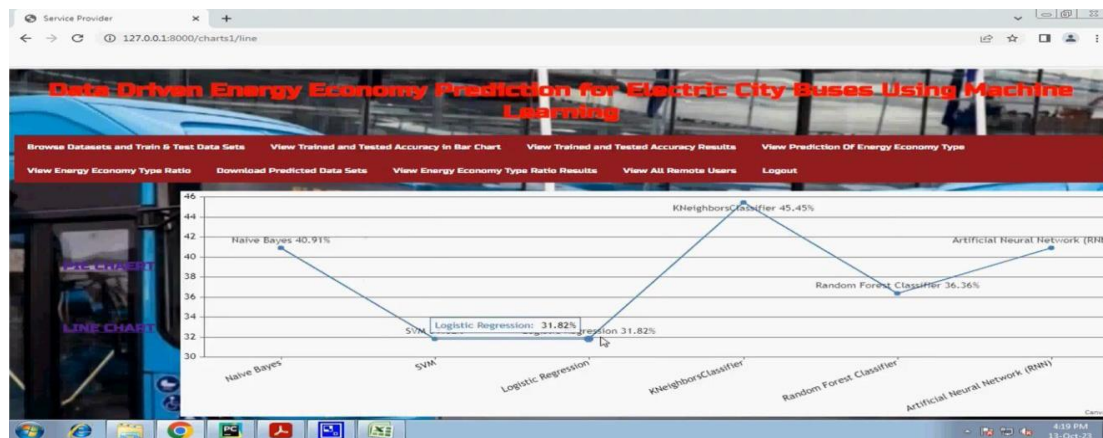


Fig 4.2 Trained and Tested accuracy in line chart

Conclusion

Using both real-world and virtual data, this project uses a data-driven approach to plan and carry out the electrification of public transportation. The results show how well the features picked from regression analysis and feature selection show how much energy Battery Electric Vehicles (BEVs) use in different real-life driving situations. This method has a lot of potential for bus companies that want to replace or improve their diesel buses with electric ones and set up the charging stations they need. The "Vehicle Routing Problem" from papers [59] and [60] has been studied and written about a lot. After a thorough analysis, we found that knowing how much energy each route needs is important for finding the best charging strategies. This includes choosing between opportunity charging and regular charging, picking the right bus operating modes (allelectric or hybrid electric), and figuring out how much space is needed for the batteries. The worst situation, which is usually called the constraint, needs the most effort to get through. The goal of this information is to help fleet managers avoid problems, come up with proactive plans to deal with big operational problems, and build trust in new technologies. The main goal is to provide a service that is always dependable and has a fair price for all routes.

Future Scope

The future of data-driven energy economy prediction for electric city buses using machine learning is bright, with the potential to significantly improve efficiency, reduce costs, enhance passenger experience, and leads to sustainable urban transportation.

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