BATTERY MANAGMENT SYSTEM

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Abstract - The design teams achieved this by breaking down the major problem into subtasks and devising solutions for each. In this way, the design teams chose an appropriate generator and mounted it inconspicuously in the vehicle; they designed a battery heating scheme using heating pads; and adapted existing vehicle circuitry to accommodate the new battery charging system. The design teams programmed a microcontroller to activate each component as needed with transistor-controlled current relays. Therefore, the electric vehicle is capable of activating its generator through the microcontroller, and the microcontroller further decides whether it should activate the heating pads in case the system temperature falls below an acceptable boundary or whether to activate the charger in case the battery state of charge falls below an acceptable boundary. Based on what the temperature sensor connected to batteries, a signal from a voltage divider tied to the same batteries, as well as one from a current sensor in series to the battery charger and the pack of batteries says, the systems to be used are determined through the microcontroller. The design teams rewired the vehicle circuitry so that when the generator activates, it always assumes responsibility for driving the vehicle, with the batteries becoming completely disconnected from the motor to minimize the change of their becoming damaged. As a side effect of the small generator size necessary to fit in the vehicle, this means that when the generator is active, the vehicle is limited to only approximately 6mph.

Towards the end of the project, it was realized by the design teams that their selected method of reading the battery of state of charge could not function as implemented.

The time was too short to repair. The sponsor, in the stopgap measure, suggested a hand switch for the activation and deactivation of the generator. That way, it would ensure that the design was not actually automatic but a feature that could easily be remedied by the design teams in the future. The design teams are glad of their work, since all other systems have already been tested and tried to be confirmed working, the vehicle can indeed run from its batteries or its generator.

Keyword: Design teams, electric vehicle, generator, battery heating scheme, heating pads, vehicle circuitry, microcontroller, current relays, temperature sensor, voltage divider, current sensor, battery state of charge, system activation, rewiring circuitry, battery charging system, generator-driven motor, speed limitation (6 mph), manual switch, automatic systems, system troubleshooting, future improvements, stopgap measure, system testing, battery protection, design challenges.

1.INTRODUCTION

A BMS (BATTERY MANAGMENT SYSTEM) is of paramount importance in most of the applications, which involve the use of energy storage in most applications currently. In general, it is supposed to monitor and control batteries by ensuring the most optimal usage along with safety, efficiency, and long lifespan for batteries. Thus, it empowers the battery, which essentially takes the task of making uninterruptible energy available in return while maintaining sound management systems.

The key roles of a BMS include state-of-charge estimation, state-of-health estimation, and the presence of fault detection capabilities that help avoid such dangers as thermal runaway or overcharging. Technological progress in BMS over the past few years has been more oriented toward better monitoring accuracy, the development of energy-efficient balancing techniques, and predictive analytics integration for lifecycle management.

Although much has been achieved, the scale-up of such systems, their flexibility to adapt to different battery chemistries, and improvement in fault detection precision remain some of the major challenges. Issues with the "Battery Management System" project, such as looking at the development of a scalable and smart BMS solution that supports various applications, face it.

The innovative features of the proposed BMS, namely active cell balancing, AI-driven anomaly detection, and adaptive thermal management, are explained in this report along with their design, implementation, and testing, showing how those improvements fill current research gaps. Further work is discussed to maintain BMS up-to-date in the rapidly changing energy ecosystem.

2. RESEARCH ELABORATION

The core research area of BMS deals with the monitoring, control, and safety features of predictive analytics. The improvement in these areas greatly enhances the performance and reliability of a battery. This is from simple coulomb counting of the first-generation monitoring system to complex Kalman filters and then machine learning algorithms toward its higher accuracy even in dynamic conditions. This covers all nonlinear properties of the battery and enables very reliable estimation both in terms of state of charge as well as a state of health for every application, from automotive to renewable-energy systems.

One very critical application of cell balancing is to prolong the life and reliable performance of the battery. Simple passive balancing, however, wastes the surplus energy as heat. Active balancing is complex and costly but would, without a doubt, promise a lot for even more efficient transfer since it will redistribute the energy among the cells. This concept is picking up really fast these days with applications in EVs and large-scale energy storage applications.

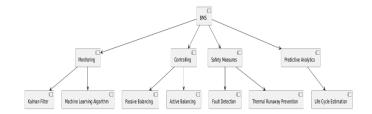
The safety features of BMS are having real-time detection, overcharge prevention, and thermal management. Except for thermal management, NTC resistors are some of the technologies that find their application within the BMS including fault predictions through AI algorithms. Safety operation needs to include overcharge prevention and inability to generate thermal runaway even in changed circumstances.

More BMS development, including neural network and hybrid model predictive analytics applied for the prediction of degradation and lifetimes of the battery. The models improve the usage of energy and assure high lifetime with safe operation. With the introduction of the BMS in renewable energy and EVs comes more significant challenges due to adaptation of loads, rapid charge protocols, and compatibility of new or upcoming battery technologies including solid-state, sodium-ion-based batteries.

The future challenges, therefore, in BMS concern scalability for the grid-level application, real-time AI integration with the evolving technology, and further compatibility. This is a key area of overall progress in terms of safety and efficiency in terms of scalability by BMS for energy storage as well as for electric mobility.

3. SYSTEM ANALYSIS

A BMS is the core function through which safe and efficient battery performance is maintained by the control of parameters such as voltage, current, and temperature. The specification of the desired application would thus form the starting point for system analysis, which can be an EV, renewable energy systems, or portable devices. This would be a rather detailed analysis to know the chemistry involved, capacity, chargedischarge cycles, and the thermal behavior of the battery. For instance, when it comes to lithium-ion batteries, they have to be kept under very close monitoring since they are sensitive to overcharging and high temperatures. Most BMS architecture, therefore, incorporates sensors for real-time data acquisition, control units for decision-making, and algorithms for predictive analytics. Analysis encompasses fault scenarios like short circuits or thermal runaway and includes safety measures such as automated disconnects and thermal management systems. CAN bus further allows monitoring and logging in real-time, thus increasing further transparency and reliability of data exchange. Moreover, scalability studies of BMS are further analyzed taking small-sized battery packs into large energy storage systems configurations. System analysis provides the foundation path towards the development of an adaptive BMS, thereby answering today's demands as well as further evolutions.



4. REQUIREMENT ANALYSIS

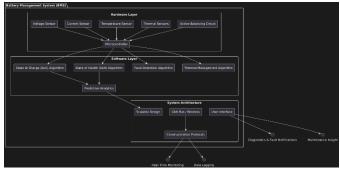
Requirements analysis is very important in developing an efficient and reliable BMS. It starts first by identifying the operating parameters of the battery, such as the range of voltage, capacity, and rates of charging/discharging. The sensors must be highly accurate while monitoring the operational parameters so as to produce data that would be adequate to help make informed decisions. SoC and SoH estimation algorithms are needed; at least it is assumed that its accuracy should be around $\pm 1.5\%$ for SoC predictions and 90% for SoH. Safety mechanisms like overvoltage and overcurrent protection are of prime importance in order to avoid critical failures. Active balancing circuits are preferred in order to ensure uniform energy distribution as they are more efficient than the passive balancing methods. Such protocols as CAN bus may be needed when the system aims to send its data in a real-time as well as have the ability for monitoring. Scalability and the ability for the BMS to support many configurations from tiny portable devices with a small quantity of cells for energy storage on a large, large scale. Last but not least, diagnostic and maintenance interfaces must be friendly enough to report performance information and fault conditions.

5. SYSTEM DESIGN

The design of BMS can be defined by hardware components, algorithms of software, and architecture of the system. It involves sensors on a hardware level measuring the voltage, current, and temperature. Further, there are controllers within it, which work for processing the information and providing real-time decisions. There is a brain of the system, acting as a microcontroller, operating on algorithms to make estimations of SoC and SoH and to detect faults and manage thermals. Energy is distributed in active balancing circuits among cells for equal charge levels. Safety features include thermal sensors and disconnect systems to mitigate the risks associated with critical failures. The software design focuses on predictive analytics with the help of machine learning algorithms that identify degradation patterns and predict battery life. The architecture developed is scalable in nature, allowing larger battery configurations and chemistries as well. Communication protocols in the system give real-time monitoring and data logging facilities according to the requirements established with CAN bus or any wireless system. An intuitive user interface for diagnostics and fault notifications along with insight into maintenance activities has also been provided for this system. The approach adopted here is to get it as modular as possible to allow up-gradation along with future

advancements.

maintenance.



1. Hardware Layer: Sensor- voltage, current, temperature, and thermal sensors.

Energy will be re-distributed amongst the cells having active balancing circuitry.

Sensors data received through microcontrollers are analyzed over there to make a decision

2. Software Layer: Estimation and detection of SOC, SOH along with algorithms related to faults

Predictive analytics with patterns of degradation so, life prediction could be done from the battery.

3. System Architecture: It is designed to scale up to large configurations and support any battery chemistry

Communication protocols: it can have CAN Bus or wireless systems with real-time monitoring and data logging User interface with feedbacks from diagnostic, fault, and

C BMS voltageSensor : Sensor currentSensor : Sensor ler · Microcont balancingCircuit : BalancingCircuit measureTemperature() : floatestimateSoC() : float estimateSoH(): float detectFaults(): boolear distributeEnergy(): voidlogData(): void stics(): void C Sensor (C) Protoco (C) UI (C) BalancingCircuit o type : Strina type : String displayFaults(): void activeBalancing(): void o communicate(): void o measure(): floa

6. IMPLEMENTATION AND RESULTS

Advanced BMS design methodology will focus on enhanced safety, efficiency, and durability for electric vehicles, renewable energy storage, and portable devices. Real-time data acquisition will get close to high precision using sensors and hybrid algorithms and distributed thermal sensors toward enhancing monitoring and estimation techniques. It would be supported by adaptive balancing mechanism, much stronger fault detection, and safety mechanisms, with applications of neural networks and data-driven models to back the support from predictive analytics. It would allow for support on multiple chemistries of batteries, large packs, even large packs for grid storage applications. Adaptive algorithms balancing the speed against the safety, in addition to balancing the longterm health of batteries, will ensure fast charging is optimized. Finally, the decision would be provided by artificial intelligence after real-time data processing, and the design of an embedded system would also be carried out to efficiently process the data and its control algorithms. These systems thus

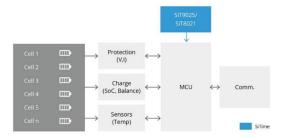
developed will be tested in a real-world setting to ascertain performance, safety, and scalability issues.

The Block diagrams representing the system are given below:

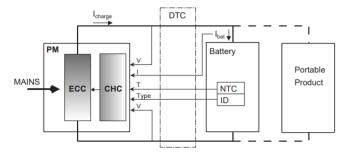
7. SYSTEM STUDY AND TESTING

Testing and Checking: the entire BMS should be thoroughly tested on performance, safety as well as the reliability point. Testing should happen under a series of different operational conditions, amongst which there includes high discharge rate, extreme temperature, dynamic profile load and such others. Functional tests ensure that the algorithms SoC and SoH are within an acceptable error threshold, and this means the accuracy for SoC estimation should not be greater than $\pm 1.5\%$, while the same goes for SoH with an accuracy of 90%. Fault detection mechanisms fall under the umbrella of safety tests, where reactions to overcharging, overheating, and even short circuits should be evident. This test verifies the thermal management system's ability to keep within safe temperatures if heavy loads are put upon it. The testing for scalability is done on the battery packs in different configurations, from the smallest packs up to more extensive systems up to 48V to test reliability of application. Communication protocol accuracy and latency will be tested for a 99.9% reliability in data transfer. The user interface will ensure intuitive operation with real-time insight into the performance of the battery and the fault conditions that exist. In this testing phase, it reveals where improvement should be made-for instance, mechanisms to enhance cooling of extreme ambient temperatures and advanced wireless protocols for remote monitoring. All-in-all, through this system study and testing stage, the efficiency and flexibility of the BMS are confirmed.

BMS Block Diagram



This is a block diagram of a power management and control system. In this system, the central unit is the microcontroller, which coordinates signals and power distribution. The system comprises a generator and generator battery, which supply power regulated by power converters. Other components controlled by the microcontroller include a temperature sensor for monitoring, a battery charger for recharging batteries, and a heating pad for temperature regulation. Motor switches control mechanical tasks to work with the motors. There are three categories for connections, such as control signals (red), power signals (green), and physical connections (black) which allow communication effectively throughout the system.



The image represents a battery management system integrated into a Power Module (PM). It efficiently monitors the charging process, hence permitting the proper and safe usage of the battery.

Major Components:

1. Power Module (PM):

It consists of two major components:

ECC (Energy Conversion Circuit): This circuit converts the input electrical power into the desired form to charge.

CHC (Charge Handling Circuit): This circuit manages and monitors the charging process.

- 2. Battery:
 - It is provided with: NTC is Negative-Temperature Coefficient resistor, which measure the temperature in the battery through detection of difference in resistance which causes due to difference in the temperature.
- 3. ID stands for the Identification system: it is considered to be a method of identifying a type of battery or in chemistry. For this, one would include a resistor within the pack of the battery

This is the device that is going to be charged by this battery, an example of smartphone, laptop or another portable device which has been designed electronically

How It Works

- The PM will relate to the battery at the time of charging process; hence it is able to continuously monitor some vital parameters
- A. Temperature: Because of NTC. For not being overheated by monitoring.
- B. Voltage: It is constantly monitored because it remains attached to the battery for decades
- C. Present: Measured by measuring the voltage fall across a low ohmic resistor placed in parallel to the battery.

According to the measurements, CHC uses charging algorithms to ensure that the right amount of current and voltage is going into the battery in order for the proper charging. The ID system implemented ensures compatibility for different types of batteries so right charging parameters are enforced in each type. This configuration provides safety to the battery. Its life cycle is extended as it optimizes charging on portable products.

8. CONCLUSION

This thesis simulates a battery management system set against the emphasis of models for the improvement of the BMS. As such, it incorporates new faster means of evaluation that can be applied to portable devices based on these models, understanding better the nature of battery behaviors and

optimizing functionalities of a BMS. Simulations open more avenues for exploratory scenarios for efficient experiments that help quickly reach the optimal condition. The thesis presents a physical system dynamics-based approach for modeling batteries, by which models for different types of batteries can be created using a unified set of components. Such models, based on electrochemical principles, are transparent and allow simulation of battery behavior in both electrical and thermal contexts.

This again shows the practicality of a complete network model in simulations like charging of Li-ion batteries and talk time simulations for mobile devices. These models allow manufacturers of batteries to design without extensive prototyping and enable them to get insights into the parameters of the design and the electrode potentials. Good quantitative agreement between simulations and measurements, like charging and discharge curves, give support to the reliability of the model. New charging algorithms such as thermostatic charging have been covered in order to mitigate capacity loss due to charging and temperature. An improved SoC indication method has also been described that promises accuracy even at low discharge rates. Using the battery models, improvements in talk time for mobile devices are predicted to be up to 24% improvement in CDMA phones.

9. FUTURE ENHANCEMENT

- 1. In the future of BMS, there will be improvement made in order to enhance its precision, scalability, and effectiveness in the system.
- 2. Estimation of SoC and SoH: The machine learning models accompanied by adaptive filtering techniques would prove beneficial in elevating the accuracy of estimating SoC and SoH. The former would look at looking up deviations arising due to dynamic load, and the latter would examine early phase degradation phases. Real-time learning will also increase the prediction accuracy.
- 3. Thermal Management: It should be equipped with sophisticated cooling systems such as liquid cooling systems or phase-change materials to exhibit enhanced performance in the thermal properties at extreme conditions. This would raise the system's immunity toward the extreme high-temperature environment as well as boost overall efficiency. Further, more advanced in technology, thermal sensors would also provide a better sense of temperature differences across the cells.
- 4. Active cell balancing: this can reach as high as 1% current energy overhead with active cell balancing using optimized algorithms. It further means that wireless balancing exploration will provide much greater freedom and scalability.
- 5. Fault Detection and Safety: The fault detection modes of over-voltage, over-current, and thermal runaway are very nice, but it could have been even better if a few other fault modes like short circuit or deep discharge were there; the algorithms can also get better, and the speed can improve to let a response come in no time.

6. Improvements can provide more than one battery configuration of over 48V. The wireless comms will be upgraded with less latency. However, 5G or LoRa could be used to enable a big system or even a remote system to surpass the connectivity problems.

Improvement aspects ensure that better performance along with a safe system is assured for applications regardless of any environment setup.

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