



Enabling large-scale integration of electric bus fleets in harsh environments: Possibilities, potentials, and challenges

Salman Harasis^{a,*}, Irfan Khan^b, Ahmed Massoud^a

^a Department of Electrical Engineering, Qatar University, Doha, Qatar

^b Clean and Resilient Energy Systems Lab, Texas A&M University at Galveston, Galveston, TX, USA

ARTICLE INFO

Handling editor: X Ou

Keywords:

E-Bus
Fast charging
Grid impact
Hot temperature
Li-ion battery
Optimization
Thermal management

ABSTRACT

This paper presents a comprehensive and detailed investigation of electric bus transit systems, focusing on their feasibility in harsh environmental areas. The aspects that combine economics, operation, long-term performance, and reliability are all addressed and discussed in both short and long-term horizons. This work covers most of the “important to consider” criteria, compares the strategies applied in E-bus energy management, and recommends the best practices to follow in the presence of many system variables. Due to the significant impact of the ambient temperature, characterizing the battery performance under hot ambient temperatures is performed considering the existing Li-ion battery technologies. Moreover, the grid impact of large-scale charging systems is identified and classified by identifying the factors affecting the grid hosting capacity and permissible interaction levels. This work includes developing a comprehensive picture of the state of the art, emphasizing the gaps found in the literature, and proposing new venues for future research that help address the identified issues to enable reliable E-transportation plans.

1. Introduction

Globally, many projects are replacing fuel-based transportation systems with electric-based systems. Their environmental needs and technology readiness increase their potential. Transitioning public bus fleets into electric is a great step toward a clean energy transportation system. Electric buses (E-buses) have the potential to minimize carbon emissions and enhance air quality, leading to a better eco-friendly future. However, it has been discovered that optimal placement and capacity sizing of charging systems in large-scale electrified public transportation is still in the early stages [1]. This, in turn, requires more attention for future research to develop advanced grid planning and operation systems. Transportation electrification has become a requirement in some countries and will be in the rest of the globe as some environmental mandates should be met.

It is important to mention the hazard associated with fuel-based mobility where carbon dioxide (CO₂) emission from fuel transit buses is about 10 kg for a 100 km trip [1]. This is a major source of pollution and global warming dilemma. Fig. 1 compares the most recent sources of CO₂ emissions where the conventional transportation system is the second major contributor to the overall emissions.

Medium- and heavy-duty vehicles are categorized according to their weight, with medium-duty vehicles typically falling within the weight range of 5–12 tons [2]. Examples of medium-duty vehicles include trucks and shuttle buses. In contrast, heavy-duty vehicles exceed 12 tons in weight, and city transit buses are classified within the heavy-duty category [2]. It has been calculated that a heavy-duty E-bus bus consumes 40 % less energy compared to a conventional diesel bus.

It is anticipated that the driving cycles for transit vehicles will lead to more significant battery discharges in comparison to Light-Duty electric vehicles (EVs), which can

potentially reduce the overall lifespan of the batteries. Furthermore, the faster charging rates for these batteries also have an impact on their cycle life. Recharging E-buses demands significant power compared to the current fast charging of small EVs. Fig. 2 depicts the yielded consumption profiles of both types [2].

The 2023 global EV outlook of the International Energy Agency (IEA) exposed that around 27 countries committed to selling only zero-emission vehicle buses and trucks by 2040.

US and EU have both proposed stricter emission regulations for heavy-duty vehicles [3]. Moreover, the study showed that more than sixty thousand E-buses were sold worldwide in 2022, reaching about 5

* Corresponding author.

E-mail address: salman.harasis@qu.edu.qa (S. Harasis).

<https://doi.org/10.1016/j.energy.2024.131564>

Received 12 December 2023; Received in revised form 31 March 2024; Accepted 4 May 2024

Available online 10 May 2024

0360-5442/© 2024 The Authors. Published by Elsevier Ltd. This is an open access article under the CC BY license (<http://creativecommons.org/licenses/by/4.0/>).

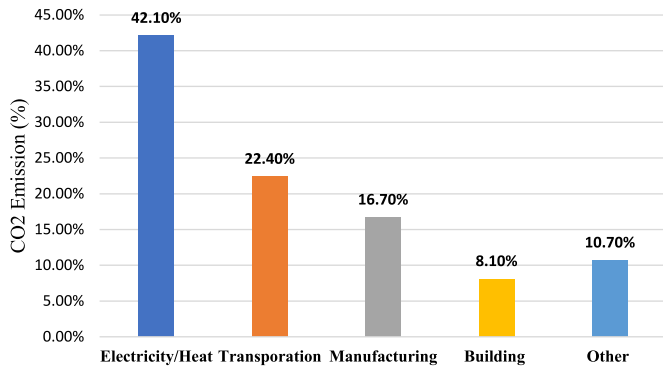


Fig. 1. Sector-wise global CO₂ emissions in 2023 [1].

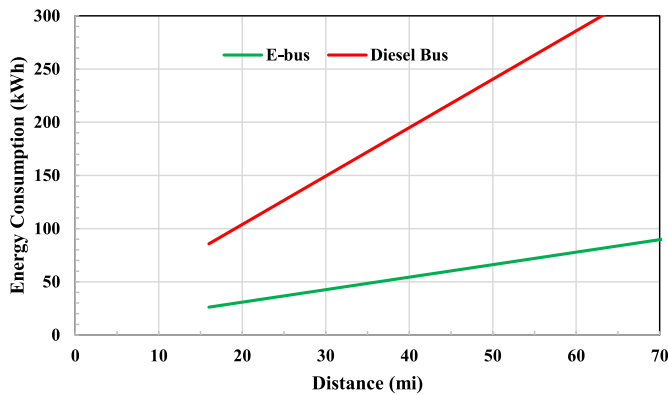


Fig. 2. Energy consumption for E-bus compared to fuel bus [2].

% of bus sales [3,4]. The size of the E-bus market size in USD with future projections until 2030 is depicted in Fig. 3 [5]. The E-bus market distribution in recent years is shown in Fig. 4 with the Chinese biggest market share [6].

Despite the widespread use of this technology, an electrified transportation system poses several challenges, particularly when integrating E-buses as heavy-duty vehicles into the existing power grid. Consequently, power networks would not be able to support the globally accelerated electrification processes without appropriate planning, regulatory reforms, and dedicating massive investments to upgrade the power networks. Additionally, with the widespread of E-bus transit systems, fast charging technology is likely to be the dominant EV charging method. In this regard, this work primarily aims to emphasize the use of this technology as a viable solution for next-generation electric transit systems (ETS), enhancing the transformation by enabling fast charging technologies.

The incremental power demand and the significant grid capacity upgrade are both among the major challenges encountered in electrification. On top of that, the economic and environmental needs to integrate renewables would increase the complexity and constraints on the

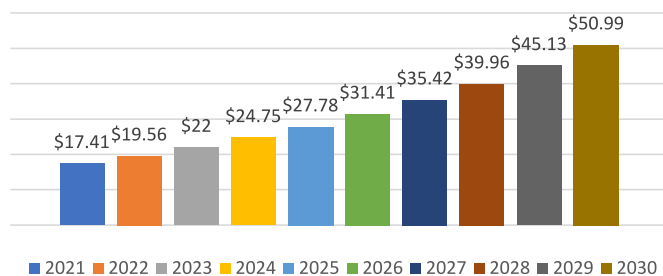


Fig. 3. E-bus market size 2021–2030 (\$ Billion) [5].

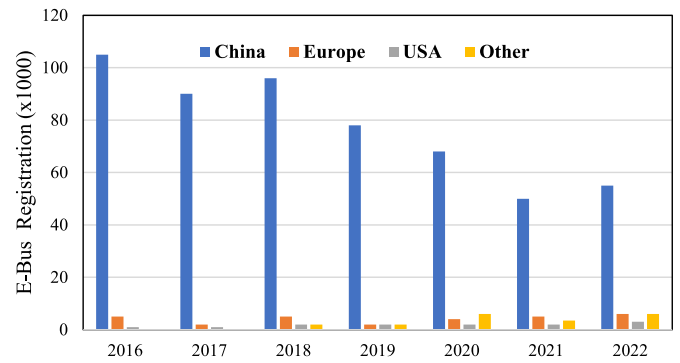


Fig. 4. E-bus registrations and sales by region, 2016–2022 [6].

required transformation [7–10]. By employing appropriate solutions, such as grid upgrades, smart charging, and demand response, seamless integration of E-buses can be successfully realized. Therefore, optimal planning, deploying, and operating of public E-bus fleets is of interest and must be carefully investigated.

Among the technical aspects that justify the urge implementation of fast charging E-bus systems include; (i) fast charging, unlike slow in-depot charging, allows for charging at any time during the day. (ii) When connected to a renewable source like a photovoltaic (PV) system, the daytime energy generation minimizes the fast-charging high-grid imported power, thereby maintaining power system adequacy. In essence, fast charging takes advantage of efficient PV self-consumption, which primarily happens during the day hours. (iii) Fast charging schedules are less constrained and offer more flexibility compared to slow charging. (iv) Furthermore, transit E-bus, compared to light-duty EVs, enforces the need for extra fast charging rates to meet the larger battery size and driving cycle. Thus, this research focuses on facilitating fast charging technology to distribute and reduce the impact of aggregated charging loads on the utility grid.

2. Problem statement, contributions, and article structure

Over the past decade, there have been many pieces of research concerning transportation electrification including the impact of EVs on utility grids. These studies have primarily focused on aspects such as trip ranges, battery size, grid compatibility, and line congestion, where the studies were primarily on light-duty EVs. Furthermore, the majority of the existing literature assumes that EVs operate optimally under normal driving conditions. However, when operating E-buses under abnormal temperatures, (i.e., less than 8 °C and greater than 35 °C), their performance experiences significant degradation. This degradation is primarily caused by additional energy requirements for thermal management purposes, besides, temperature-caused capacity loss. Consequently, the battery capacity reduces under such ambient temperatures.

The following contributions provide valuable insights into the complex issues affecting the transportation electrification pathway in adverse weather conditions.

- **Investigating E-bus grid integration:** This paper offers an extensive examination of the integration of E-bus into the grid, with a specific focus on how they perform in harsh weather conditions and their impact on power grids.
- **Characterizing the impact on distribution network hosting capacity:** The investigations include potential challenges to grid stability, shifts in demand patterns, and implications for the electrical infrastructure. A study of existing literature concerning the broader effects of ETS on the power grid is conducted. Moreover, definitions and standards related to EV charging, power networks, and power quality are also discussed.

- **Investigating the weather-related consequences on E-bus.** The consequences are categorized into:
 - **Battery Performance:** The study explores how harsh environments affect the functionality of batteries. This includes an in-depth review of Li-ion batteries, covering their operational principles, electro-chemical properties, battery health considerations, and performance characteristics in different ambient temperatures. Moreover, real and lab test pieces of evidence are reported and analyzed.
 - **Trip range:** The effects of unusual weather conditions on the overall travel distance and efficiency are evaluated. This work particularly examines how high temperatures affect various vehicle-level factors, including the reduction in travel distance, the energy needed for cooling, and charging duration.
- **Investigating battery thermal management systems for harsh environments:** detailed examinations of useful mechanisms used to alleviate the difficulties associated with batteries in harsh environments are performed.

Fig. 5 shows the structure of the article according to the addressed topics.

3. The challenges facing the electrification plan and charging infrastructure

Public transit buses are a class of heavy-duty vehicles that require significant power consumption rates to electrify [11]. Therefore, the major challenges encountered E-mobility transition include; (i) increased unpredictable electricity demand. The increased electricity demand is hazardous and requires upgrades to the utility grid infrastructure to handle the extra-large loadings added to the grid [12]. Adding the charging load of E-buses to the existing utility grid results in grid congestion, risking grid reliability. (ii) Establishment of new charging infrastructure. (iii) Grid stability and power quality; the integration of large-scale charging stations into the grid leads to fluctuations

in power supply, impacting grid voltage and frequency profiles [13,14]. (iv) In big cities, where the E-bus is economically feasible, the increased power demand and the difficulty of incorporating renewables, amplifies the encountered challenges. (v) The lack of infrastructure and the required makeover in many urban cities. (vi) The battery size of a typical E-bus ranges from about 5 to 10 times the sedan EV battery size [15]. (vii) IEEE-1547 and IEC-61000-3-2:2018 standards have established strict limitations on the allowable harmonic content and DC offset currents into the grid, and EV chargers must adhere to these standards in their design and operation [16].

Due to the aforementioned reasons, it is very crucial to coordinate and optimize all the parameters involved in the transformation plan. Therefore, alleviating the challenges and maximizing the benefits can be reasonably achieved.

4. Optimal planning strategies of E-bus public fleets

Many frameworks and strategies have been proposed for optimally deploying charging infrastructure while accounting for grid stability, achieving economic benefits, and reducing carbon emissions. In Ref. [9], the study examined the most suitable charging system to minimize both grid impact and overall cost. It was revealed that truncated fast charging followed by cooldown intervals is feasible and helps mitigate battery heat generation caused by high charging rates. In Ref. [10], a mathematical model was presented to improve the reliability of a power grid with huge E-bus charging loads. The focus was to minimize voltage drop by placing the charging infrastructure with a PV source installed.

In [17,18], the authors reviewed different approaches proposed for optimal EV charging planning and their impact on power grids. In Refs. [19,20,21, and 21]] the work focused on the thermal management system as an integrated parameter in the optimal charging of Li-ion batteries. Different schemes were studied to obtain their effect on the optimal driving range of EVs. The authors in Ref. [22] developed a dual-level optimization framework to manage an EV fleet in a power system with multiple types of energy sources. The paper focused on scheduling the EV fleet according to variable energy costs. Therefore, the optimization was tuned according to the selected route, time, and energy cost.

In [23], the researchers considered the IEEE 39-bus system with high EV charging demand distributed all over the network. The work was directed toward the feasible operation of the E-bus fleet in the open wholesale electricity market where the V2G cost-minimization approach was investigated. The researchers implemented the distributional robust optimization method (DRO) to account for the high variability and uncertainty level of the input variables in the framework. In Ref. [24], a comprehensive analysis of the optimal deployment of fast-charging infrastructure was made. The work discussed the urban and rural deployment considering renewable integration to minimize the grid impact. Moreover, the work discussed the business model and the payback period of different infrastructure models when additional storage and PV systems are attached. Globally, Shenzhen city in China is the first complete E-bus city.

The authors in Ref. [25] considered this city to design an optimal approach to meet the charging demand and the grid capability. A planning strategy was made to deploy the charging stations in a way to serve the buses and distribute the loadings among the grid nodes. The study made in Ref. [26] showed that the feasibility study of electrifying public bus fleets should consider the power of the charging, consumed energy, and lifespan prices. Simulations were established to assess E-buses under different operating conditions, considering bus configuration, charging methods, and operating routes. In Ref. [27], a scheduling optimization was proposed to minimize the number of E-buses in a heterogeneous fleet by determining the optimum number of bus types and sizes.

An optimal selection of charging technology for urban transport was made in Ref. [28]. The work revealed that overnight charging provides

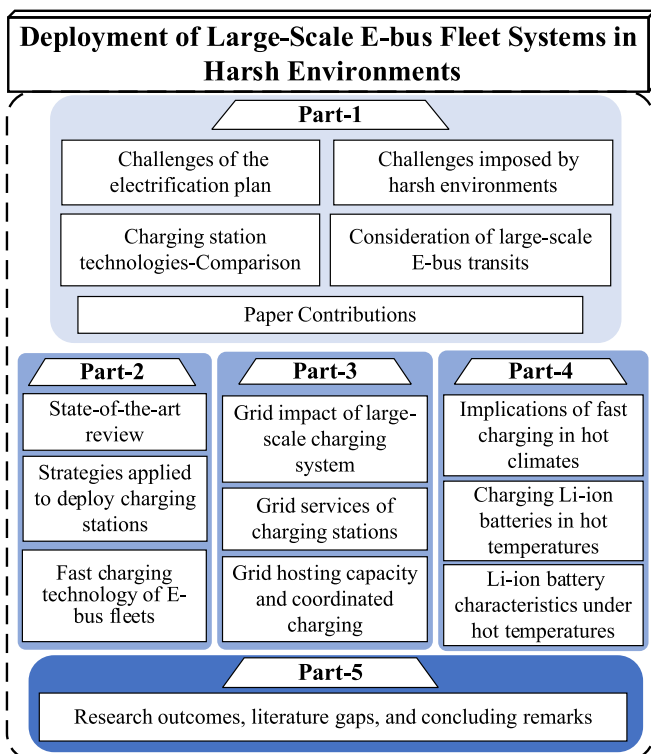


Fig. 5. The structure of the article.

acceptable performance with no significant impact on the grid. The shorter charging time that can be achieved using fast charging would impact the service quality. In Ref. [29], a configuration was proposed to determine the number of E-buses and battery capacities to meet the E-bus predefined schedule with opportunity and depot chargers. The design included charger and battery capacities. The results showed that large fleets should be heterogeneous, where different battery sizes lead to better short and long-term benefits. Compared to the work performed in Ref. [29], an additional variable was incorporated in optimizing the charging infrastructure considered in Ref. [30] where the impact of different Li-ion battery technologies on designing the E-bus fleet was conducted. The results revealed that coordination between Li-ion battery chemistry, charging technology, and battery capacity leads to a more optimized system. In particular, it was shown that according to the different characteristics of the chemistries, opportunity fast charging using Lithium Nickel Cobalt Manganese Oxide (NMC) and Lithium Titanate (LTO) chemistries, and slow in-depot charging with Lithium Iron Phosphate (LFP) chemistry yield feasible charging schemes. The findings indicated that achieving optimal system performance depends on coordination among Li-ion chemistries, charging technology, and battery capacity.

Depot and opportunity charging schemes of the city E-bus transit system were investigated in Ref. [31] to determine the optimal strategy. The study found that the regenerative braking recaptures 15–40 % of the energy. Moreover, it was found the 18 m E-bus consumes less energy which is about 33 % more than the 8.9 m E-bus. Table 1 summarizes the study findings and Fig. 6 compares the consumption rates of the E-buses used in public transportation. It can be seen that the energy efficiency of the 18 m E-bus is better than the other model as it only consumes 33.4 % energy more than the 8.9 m E-bus with double passenger capacity.

In [32], the authors investigated the impact of charging infrastructure versatility and interoperability on cost reduction. The investigations revealed that charging stations can be used by different E-bus brands to optimize the cost. However, the existence of different standards and communication protocols impedes interoperability. The work performed in Ref. [33] considered establishing a power reserve to minimize the cost and optimize the charging. The charging location was also variable and can be determined according to the grid status. Another optimization was proposed in Ref. [34] where a two-stage problem was formulated to first minimize the trip time of the E-bus and second to reduce the charging cost. In Ref. [35], the work introduced an approach for addressing the challenging task of locating charging stations within a transportation network. Optimizing charging station placement was performed by utilizing self-avoiding random steps and a probabilistic rule to achieve comprehensive network coverage with the fewest number of charging stations.

In [36], the research utilized a multi-objective optimization, which incorporated a "brute force" Monte Carlo simulation. The objective was to find the most efficient combination of chargers and charging power that minimizes the total cost of fleet operation while minimizing grid impact. Similar to the work done in Ref. [24], the study in Ref. [37] aimed to find the optimal energy and power capacity for battery storage to support charging scenarios for E-buses. It used a Mixed Integer Linear Program (MILP) to determine the storage battery power and energy ratings for a configuration involving depot and en-route charging stations. It was found that the storage battery was able to reduce the peak

Table 1

The electrical specifications of the most deployed public E-buses.

| Parameter | 8.9 m E-Bus | 12 m E-Bus | 18 m E-bus |
|--|-------------|------------|------------|
| Battery (kWh) | 160 | 160 | 240 |
| HVAC consumption (kW) | 24 | 24 | 48 |
| Electric boiler (kW) | 25 | 25 | 37.5 |
| Energy Consumption of the drivetrain (%) | 100 % | 101.7 % | 133.4 % |

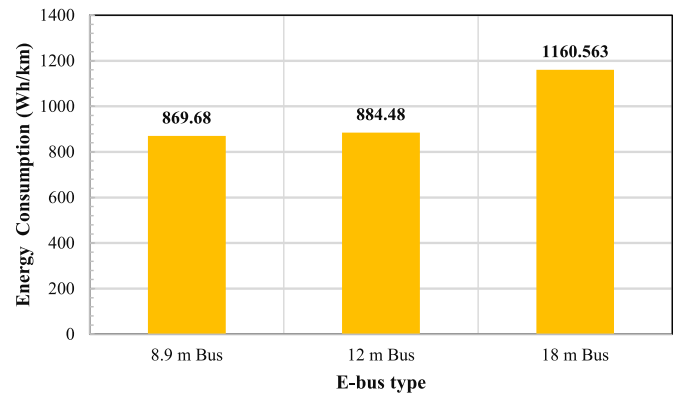


Fig. 6. Energy consumption of different E-bus models [31].

demand and cost by minimizing operational expenses. In Ref. [38], the problem of EV charging was addressed using an optimal power flow (OPF) to account for network constraints at various charging locations on the IEEE 14 bus system. The study employed a nested optimization approach based on valley filling and peak shaving of the charging demand.

The battery-swapping technique was proposed in Refs. [39,40,41]. The authors introduced an optimization model for efficient scheduling of battery swapping stations in an E-bus transit system. The swapping aimed to minimize operational costs by charging during low electricity price periods and providing grid ancillary services. Even though the technique provides a certain paid grid service, large additional investment cost was incurred which minimized the feasibility of the proposed technique.

The study conducted in Ref. [42] optimized charging infrastructure and battery capacity for E-bus public transit using a MILP. The article presented in Ref. [43] proposed a method to minimize the total cost of ownership (TCO) in an E-bus fleet. The target was to optimize battery size, the amount of chargers, and power ratings based on the prepared schedule. Battery customization was able to reduce the cost; however, it adds extra cost to manufacturing. To enhance EV resilience [44], discussed the optimization of EV battery size to reduce peak load and supply power during grid outages by imposing a penalty during peak intervals.

The framework presented in Ref. [45] demonstrated the potential of EVs in both G2V and V2G modes. It was revealed that the existing power network can accommodate a certain number of EVs. However, the fleet size handling capability increases when EVs participate in a V2G market. This suggests that the integration of V2G ancillary service could effectively enhance the capacity of current power networks to handle larger numbers of EVs. In Ref. [46], storage batteries were involved in optimizing the operation of E-bus charging stations, similar to the work presented in Refs. [24,37]. The research incorporated factors like battery price, capacity charge, and electricity price arbitrage to reduce peak charging loads and electricity purchase costs.

Fig. 7 depicts a mesh diagram illustrating the major challenges facing the deployment of E-bus fleet systems and the relations among the variables associated with the bus electrification plan. The mesh aims to link the factors influencing the operation of E-transportation systems. The diagram provides a comprehensive view of the complexities involved in managing a battery-based E-bus fleet system, highlighting the need for a balanced approach to address concerns while optimizing variables and efforts. These variables and efforts are interconnected and can influence each other. They all contribute to the overall cost, which is a crucial aspect to consider in the implementation and operation of E-bus fleet systems.

Table 2 is introduced to summarize the findings related to review articles on public E-transportation systems. This Table serves to describe the extent of these reviews and offers insight into the subjects covered.

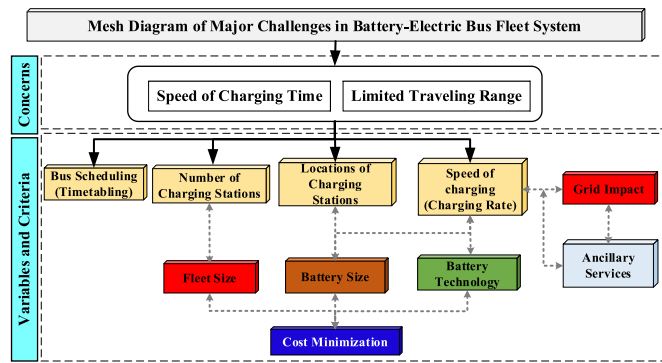


Fig. 7. Mesh diagram showing the relation between bus electrification plan associated variables.

Furthermore, Table 3 is designed to present a comprehensive overview of global initiatives, along with reports from completed projects and devised strategic roadmaps in the field.

Table 2
Summary and comparison of state-of-the-art review papers on E-bus Fleet systems.

| [Ref] Year | Review Scope | EV Type | Grid Impact | Renewable Energy Source | Comparing Different Charging Technologies | Impact of Harsh Environment | Remark (Paper Highlight and Topics) |
|--------------------|--|--------------------|----------------|-------------------------------|---|--------------------------------|---|
| [11] Year:2022 | Waiting Time, bus scheduling, placements of fast chargers | E-Bus | ✓ | × | ✓ | × | The focus is evenly distributed on the criteria considered with literature comparisons |
| [16] Year:2023 | Charging methodologies and topologies, control algorithms, Grid impact of the topologies | Light Duty & E-Bus | ✓ | × | Fast Charging Only | × | Focused on the power converter topologies |
| [47] Year: 2022 | EV charging planning considering distribution networks | Light Duty Vehicle | ✓ | × | ✓ | × | Focused on the integrated planning strategies with the distribution networks, positioning the charging stations according to the technology |
| [48] Year: 2021 | Sizing of energy storage, energy management and modeling, and charging schedules. | E-Bus | ✓ | × | × | × | Focused on powertrain and motor drive |
| [49] Year: 2013 | Operations of E-bus. Study the critical aspects (trip range and charging time) | E-Bus | × | × | ✓ | × | Focused on lessons learned from previously deployed systems worldwide |
| [50] Year: 2016 | Comparative review of hybrid, fuel cell, and battery-based bus | E-Bus | × | × | × | × | Focused on the comparison includes economic, operating, energy, and environmental aspects |
| [51] Year: 2022 | Planning and scheduling of E-bus | E-Bus | ✓ | × | ✓ | × | Focused on investment in E-bus, charging infrastructure, placement of chargers, and scheduling |
| [52] Year: 2022 | Mapping of the methods, trends, and gaps of the field E-bus | E-Bus | ✓ | × | ✓ | × | – |
| [53] Year: 2021 | EV powertrain modeling and general charging methods | Light Duty Vehicle | ✓ | ✓ | × | × | Focused on EV and Hybrid EV models and operation |
| [54] Year: 2021 | E-Bus battery, Power Management, and Charging Scheduling | E-Bus | × | × | × | × | Sizing of battery, types of powertrains and motors, and control schemes |
| [55] Year:2023 | EV charging stations with PV Considering techno-economic Implications | Light Duty Vehicle | ✓ | ✓ | ✓ | × | The EV market, technical requirements, and infrastructure. Control strategies. |
| [56] Year:2021 | Converters architecture and International standards for EV charging stations | Light Duty Vehicle | × | × | ✓ | × | Review of connection standards. Detailed analysis and comparison of the power converters used in different charging technologies |
| Proposed | Public E-bus fast charging infrastructure in harsh environments | E-Bus | ✓ | ✓ | ✓ | ✓ | Investigation of energy management aspects, grid impact, renewable integration, Li-ion battery performance under hot climates. |

✓: Considered/Included in the paper, ×: not considered.

5. The considerations of large-scale E-bus fleet systems

This section aims to assess the potential of E-transit systems and sheds light on the viability of electrifying public mobility systems. It investigates numerous technical criteria and aspects, including: (i) charging technologies and infrastructure, (ii) battery performance under extreme ambient temperature conditions, and (iii) considerations of battery capacity, type, and weight. The discussion explores cutting-edge strategies that have implemented various energy management schemes for E-bus fleets. It also emphasizes the research gaps by outlining the specific aspects considered in system development and realization.

Public transportation buses are ideal for electrification due to their predefined routes and the cost-effectiveness of their operations compared to diesel buses. Moreover, Public buses are ideal for electrification due to their certain trip, fixed routes, and cost-effective operation compared to light-duty private vehicles. However, on large-scale fleets, bus depots require substantial electricity to charge the fleet, which often occurs overnight, potentially drawing 25–100 kW per bus, totalling 200–400 kWh per day [61]. A depot serving 100 buses might need up to 10 MW of service for charging [60,62].

One key parameter in the optimal integration of E-Bus fleet systems is the charging strategy that directly affects the system’s performance

Table 3
International studies and funded projects on E-bus transportation systems.

| [Ref] Year | Project Name | Country | Deliverables and Scope |
|-----------------|--|---|---|
| [2] 2020 | “Medium- and Heavy-Duty Vehicle Electrification, An Assessment of Technology and Knowledge Gaps” | USA | <ul style="list-style-type: none"> - Leveraging medium- and heavy-duty EV information to evaluate the commercialization of EVs. - Assessment of current E-bus electrification architectures. - Identification of E-bus technologies. |
| [3] Year:2023 | “Global EV Outlook 2023, Catching up with climate ambitions” | International (International Energy Agency) | <ul style="list-style-type: none"> - Integration of historical analysis and projections to 2030, covering EV deployment, charging infrastructure, battery demand, emissions reduction, and policy advancements. - Guide policymakers and stakeholders in strategies for EV adoption. - Examines the financial performance of EV-related investments. |
| [24] Year: 2017 | “Considerations for Corridor and Community DC Fast Charging Complex System Design” | USA | <ul style="list-style-type: none"> - Focus on DCFC chargers for convenient charging. - Lesson learned from past DCFC and data collection from many EV brands. - Considerations for designing and enhancing DCFC infrastructure. - Cost estimation of high-power DCFC. |
| [57] Year:2023 | “FLOW- Flexible energy systems Leveraging the Optimal integration of EVs deployment Wave” | EU | <ul style="list-style-type: none"> - Investigation of V2G definitions and terminologies for both EVs and electric vehicle supply equipment (EVSE) |
| [58] Year:2022 | “U.S. Virgin Islands transportation Electrification roadmap” | USA | <ul style="list-style-type: none"> - Fostering the expedited proliferation of EVs across the Virgin Islands to enhance EV dependability, - Improve air quality, and enable a more resilient power grid. |
| [59] Year:2018 | “Electric Buses in Cities, driving towards cleaner air and lower CO ₂ ” | EU | <ul style="list-style-type: none"> - Investigation of E-bus barriers, financing, and TCO, - Review of Li-ion battery market |
| [60] Year:2021 | “Electrification of Public transport, A Case Study of the Shenzhen Bus Group” | Shenzhen, China | <ul style="list-style-type: none"> - Analysis of the Policies and Infrastructure of Shenzhen E-Bus Group. - Analysis of the business model. - Assessment of the TCO and the overall benefits |
| [61] Year:2020 | “Electric Vehicle Charging Infrastructure” | India | <ul style="list-style-type: none"> - EV growth scenarios - EV charging infrastructure technologies |

Table 3 (continued)

| [Ref] Year | Project Name | Country | Deliverables and Scope |
|----------------|---|---------|---|
| [62] Year:2023 | “Decarbonising Transport: What Does It Mean for India?” | India | <ul style="list-style-type: none"> - Key considerations for Public transit buses - Review of existing efforts in rolling out public charging stations - Investigation of India’s transport sector emissions - Investigation of the potential of EV technology pathway |

and cost. In large-scale systems, the interaction with the utility grid is critical when the energy transacted from the grid induces frequency instability and voltage regulation issues at the distribution system level. Therefore, many approaches were proposed to establish feasible planning frameworks. A reliability-based deployment for large-scale charging stations is a considered solution for safe grid and charging system operation. In this approach, power system reliability indices can be employed for the optimal allocation of charging stations [63].

The charging technologies are mainly featured by their charging rates and voltage levels. These technologies are categorized and classified according to the power levels as presented in Table 4, and according to voltage levels as presented in Table 5. The classification made in Table 4 shows the impact of each charging technology on the battery charging time and lifecycle, whereas Table 5 presents the technology’s standardized level according to the utility voltage and power ratings. Slow charging is typically related to a more cost-effective charging infrastructure and larger batteries.

On the other hand, opportunity charging using a Pantograph has a higher initial infrastructure cost but smaller battery packs with overall feasible operation in a fully electric fleet system [65]. A practical challenge that faces the deployment of fast chargers lies in the need for a much larger power transformer rating, where it has been determined that opportunity chargers require very large power transformers [65].

Typical operational requirements for E-buses involve covering an average distance of 150 km over 15 h, with 6 h dedicated to depot charging at a rate of 100 kW. Currently, most commercially available DC fast charging (DCFC) stations typically provide a power capacity in the range of 300–600 kW. However, the trend of increasing charging power ratings is significant. This development aligns with the growing demand for faster charging times and the need to support higher-capacity batteries.

6. Strategies applied to deploy charging stations

Optimizing E-bus charging systems is important to enable sustainable and cost-effective solutions. At the same time, the interaction level

Table 4
State-of-the-art charging technologies of battery E-bus [11,31,32].

| Charging Strategy | Features | Comment |
|------------------------------------|--|---|
| Fast Charging (Opportunity/ Flash) | Fast charger with a large intermittent and variable charging profile, its efficiency depends on a high C-rate which in turn, induces a rough grid interface. | Short average battery lifetime (Around 4–5 years) |
| Slow Charging (In-depot) | Parking charger/slow, flat charging profile, seamless interaction with the grid | Long average lifetime (Around 12–14 years) |
| Swapping Charging | Vary in size, the charging profile is controllable | Expensive capital cost |
| Wireless Charging | Easy and safe charging-contactless | Expensive capital cost |

Table 5
Charging power levels in compliance with IEC-61851 standard [64].

| Technology Level | Supply Voltage (V) | Current (A) | Charging Power (kW) | Charging time of 50 kWh battery (hr) |
|-------------------------|--------------------|-------------|---------------------|--------------------------------------|
| Level-1 (AC) | 230 – 1 ϕ | 16 | 3.7 | 13.6 |
| | 230 – 1 ϕ | 32 | 7.4 | 6.8 |
| | 400 – 3 ϕ | 16 | 11 | 4.5 |
| | 400 – 3 ϕ | 32 | 22 | 2.3 |
| Level-2 (AC) | 400 – 3 ϕ | 63 | 44 | 1.1 |
| Level-3 (DC) | 400 | 125 | 50 | 1.0 |
| Level-3 (DC-Fast) | 400 | 375 | 150 | 0.3 (20 min) |
| Level-3 (DC-Ultra Fast) | 800 | <437 | 350 | 1.1 (9 min) |

with the utility grid should be established such that the interaction is risk-free. In typical EV-charging stations, an agent-based locating strategy can be effectively used to determine the location of the charging station that considers the scattered patterns of charging behaviors and can meet the convenience level of the agents [66]. However, in E-bus charging stations, the locating process is more restricted to transportation networks and power distribution networks. In this regard, intersection-based and traffic-based strategies can be adopted [67]. In a junction-based locating strategy, the charging station can be deployed at bus stops (route-end) that intersect with other routes [67].

This should maximize the served buses where this type is favorable in city (urban) fleets. On a traffic-based locating strategy, the aim is to locate the chargers at the routes with higher traffic. This plan is considered to maximize the utilization level of the chargers and to serve the fleets with long trips at divergent locations. This strategy, indeed, distributes the demand seen by the power distribution network (PDN) and vividly serves the inter-city fleets [68].

Many studies revealed that charging slowly (at depots) necessitates a larger onboard battery. On the other hand, fast charging en-route (at stops and terminals) can reduce the required battery capacity by approximately two-thirds [66–68]. Furthermore, the energy consumption of the bus was decreased to 90 % due to a reduction in bus weight [67,69]. However, the strategy of fast opportunist charging for large-scale fleets can lead to risky interactions with the grid. As a result, implementing coordinated fast charging strategies (non-opportunist) and utilizing renewable energy sources seems to be an optimal solution to the charging problem. This approach not only addresses the issue of charging speed but also promotes the use of renewable energy, making it a comprehensive solution.

The comprehensive literature survey conducted in this study encompasses state-of-the-art strategies and optimizations, providing a thorough overview of the field. In addition to surveying the review papers presented in Table 2, an analytical Table (i.e., Table 6) has been created to present the merits of the reviews and to display the criteria considered in the proposed frameworks. This allows for a deeper understanding of the advancements in the field and helps identify existing technological gaps.

Table 6 illustrates the strategies adopted in planning and optimizing E-bus transit systems all over the world. The Table highlights the main objectives and the mathematical tools employed to conclude the outcomes. It's important to note that a significant percentage of the literature has conducted research under the assumption of normal ambient temperatures. However, the presence of severe ambient temperatures can have a substantial effect on optimization processes. This highlights the need for future studies to consider the influence of a wider range of ambient temperatures to ensure the robustness and reliability of their findings.

While the basic principles of electrification apply to both heavy-duty and light-duty EVs, the differences in scale, energy demands, and operational profiles necessitate tailored solutions and considerations for each category. When considering the E-bus transportation systems

compared to light-duty EVs, several distinctions arise. E-buses require significantly larger and more powerful batteries to accommodate the higher energy demands of their larger size and heavier loads. Light-duty EVs have smaller batteries tailored to meet the energy needs of individual consumers. The widespread adoption of E-bus systems and EVs affects the electrical grid and charging infrastructure. The impact and requirements may lead to compulsory grid upgrades. To ensure consistent conclusions about the technical aspects of charging technologies when applied to large-scale systems, Table 7 is constructed. This Table serves to illustrate and contrast the features of each technology from a business model perspective.

According to the primary features of each charging technology and considering the associated advantages and disadvantages discussed earlier and outlined in previous Tables, Pugh matrix is constructed in Table 8. This matrix offers a fair methodological way of comparing the technologies, supported by recent facts and features. It facilitates the analysis of different options to identify the optimal technology for a specific application. The results of the Table reveal that even though fast charging technology scores the minimum regarding grid impact that has the largest weight (i.e., 15 points), it scores the highest with 82 points among other technologies. The battery swapping technique scores the second highest with 78 points which gives it great potential as battery storage system price decreases.

Results Discussion: The requirement for slow charging necessitates a larger battery size capable of supporting multiple trips. This, in turn, results in a battery that is not only large but also heavy. Consequently, the bus carrying this battery must also be larger and heavier. In conclusion, considering future transportation, which is likely to be dominated by EVs, there is a demanding need to advance fast charging rates and expand the network of charging stations while adhering to grid quality constraints [85]. This approach aims to reduce charging time and battery capacity. It is a strategic move that will facilitate the fast transition to complete E-transportation systems. Key takeaways include the requirement for fast chargers that prioritize charging speed without being opportunistic, with the ability to operate during off-peak hours. Additionally, fast-distributed chargers provide potential solutions that are not only efficient but also have a minimal impact on the grid, have lower power ratings, and are reliable.

7. The impact of transportation electrification on grid hosting capacity

The accelerated transition toward E-mobility systems is a demanding issue that risks the adequacy of power networks to meet the incremental demand. In recent surveys conducted by European Commission agencies, it was shown that the growth rate of battery E-buses increased by 50 % on average in most of the union countries in the last two years [86].

Moreover, the trend of this transition is expected to accelerate more with an additional 20 % growth by 2027 [11,86]. This, in turn, adds hundreds of TWh demand on power networks. At the same time, the level of grid flexibility to accommodate the unpredictable power production coming from renewable sources and heavy-consuming loads (like ETS) is another concern to be considered.

Grid-supporting services were proposed in different pieces of literature. Renewable energy and storage systems were proposed to shave the peak load caused by fast charging facilities [87–89]. Integrating an energy storage system was proposed in Ref. [87] to minimize the grid impact and meet the high charging rates in a large-scale E-bus transit system. In Ref. [65], it was shown that the load factor (LF) of the fast chargers is very low compared to the depot charger. From a power system perspective, this value represents a poor load profile where the ratio between the average and maximum load is small. However, from an optimal operation perspective, this means that fast charger utilization of the transformer is small and can be easily shaped by shifting and coordinating the charging process of the whole fleet of buses. Table 9

Table 6
Comparison of fast charging schemes dedicated for battery E-bus fleet systems (from 2015 till 2023).

| [Ref] Year: | Developed Framework | Objective Function | Problem Formulation | (Region) Climate/ Temperature |
|-------------------|---|--|--|--|
| [39] Year:2023 | Optimization of MW-scale battery storage swapping stations | Minimizing the running costs of the BSS by exploiting the low electricity prices, and utilizing the BSS in the provision of grid reserve. | Saving cost index (SCI) based Pyomo abstract model | (Ontario, CA) Min = -11 °C Max = 27 °C |
| [42] 2017 | Optimization of the trade-off between charging stations and battery size | Cost-minimizing by reducing the number of chargers and battery sizing | Deterministic optimization/MILP | (Germany) Min = -15 °C Max = 30 °C |
| [70] Year:2018 | Determination of charging station locations considering battery degradation | Cost-minimizing including battery aging function | Deterministic optimization/MILP with several periods | (German city) Min = -1 °C Max = 26 °C |
| [71] Year:2018 | Locating charging stations considering energy consumption uncertainty using affinely adjustable robust counterpart approach | Cost-minimizing with set-based RO function | Deterministic optimization/robust MILP | (USA/Utah) Min = -3 °C Max = 32 °C |
| [72] Year:2019 | Locating charging stations considering high electricity demand charges | Minimizing the total cost of EV batteries and electricity cost | MILP | (USA/Utah) Min = -3 °C Max = 32 °C |
| [73] Year:2019 | Locating charging stations based on a multi-objective techno-economic optimization framework | Minimizing the total cost of ownership | Matlab-based iterative approach | (Spain) Min = 7 °C Max = 24 °C |
| [74] Year:2019 | Locating large-scale charging stations based on multistage planning considering transportation and power networks | A spatial-temporal model that finds the locations and capacities of E-bus charging stations, and strategies for multiple stages of infrastructure planning | MI second-order cone programming | (China) Min = 11 °C Max = 32 °C |
| [75] Year:2020 | A bi-objective optimization framework for efficient scheduling considering the limited driving range with chargers installed at terminal stations | Minimize the total number of EVs and chargers | Hybrid Lexicographic and max-flow solution methodology | (Singapore) Min = 20 °C Max = 33 °C |
| [76] Year:2020 | Two incorporated stages of optimization: (1st): estimation of energy consumption (2nd): optimizing the placement of the chargers | Optimal sites for charging infrastructure based on routes, energy consumption profiles, and running costs | Enhanced heuristic descent gradient and Voronoi diagram framework (graphical representation) | (Toronto, CA)/ Cold Min = -7 °C Max = 27 °C |
| [77] Year:2021 | A bi-level programming framework: (1st): find the charging station location (2nd): calculate user cost | To optimize the costs of the users and operators | Modified genetic algorithm | (Zurich, Switzerland) Min = -1 °C Max = 25 °C |
| [78] Year:2021 | Location planning model including bus operation and distribution networks | To minimize the total cost of the construction, operation and maintenance, travel, and the power loss | Affinity Propagation method and Binary Particle Swarm Optimization a | (Yangjiang, China) Min = 12 °C Max = 32 °C |
| [79] Year:2022 | Locating opportunity charging infrastructure, sensitivity of E-bus fleets under both optimization of charging stations and trip schedules | Optimizes the charging stations and E-bus trip schedules | variable neighborhood search metaheuristic approach suited for large-scale problem | (Germany) NA |
| [80] Year:2022 | Locating and scheduling chargers at bus stops and considering variable electricity prices | Optimizes the total cost of charging, chargers price, battery price, and penalty cost due to waiting time and trip delay. | Deterministic model linearized to be solved using MILP | (Sydney, Australia) Min = 8 °C Max = 26 °C |
| [81] Year:2022 | Optimization of charger placements and fleet scheduling for E-buses with time-varying ridership, dwelling time, and travel time with opportunity charging | Optimizes the battery capacity, bus fleet size, and charger locations at stops and terminals to minimize the total costs. | MI-Nonlinear Programming | (Oslo, Norway) Min = -5 °C Max = 23 °C |
| [82] Year:2023 | Cooperative optimization of E-bus trips and charging timetabling including sensitivity analysis of the impact of charging technology on cost | Minimize the charging cost | Integrated adaptive large neighborhood searching and branch and bound (ALNS-BB) | (Shenzhen, China) Min = 11 °C Max = 32 °C |
| [83] Year:2023 | Optimal placement and sizing that maximize the returns of a swap station: (1st): E-Bus consumption model formulation. (2nd): battery model for profit maximization. | Maximizing income from swapping, minimizing charging costs, and regulating service | Constraint-based iterative solution approach | (Berlin, Germany) Min = -2 °C Max = 24 °C |

presents a comparison between the positive and negative impacts of deploying large-scale E-bus charging systems on the utility grid.

When the charging station is connected to distribution transformers under hot temperatures, their performance becomes critical due to the excessive heating of the internal windings caused by the injected harmonics. Under heavy charging loadings, a critical winding temperature of more than 200°C can be reached [90]. This leads to quick transformer aging and reduced grid reliability. In this regard, the work presented in Refs. [90,91] showed that applying controllable-coordinated charging scenarios can significantly reduce the transformer rating requirements and reduce the thermal stresses. A mixture of overnight (depot) and opportunity (fast) charging is also proposed to ensure that buses start

their operations fully charged, allowing for top-ups. For overnight charging, medium-voltage charging using standard slow charging was employed, achieving power levels similar to light EVs. However, the charging process for E-buses takes longer due to their larger battery capacity, which is approximately 10 times that of light-duty EVs.

Since depot charging occurs overnight, it has reduced grid impact. This is a conservative assumption as the actual grid impact depends on the number of E-buses that are charging simultaneously and harmonic distortion levels. Fast charging is more expensive compared to depot charging. Nevertheless [92], indicated that deploying a fully electric fleet can lead to smaller battery sizes, resulting in a lower TCO, which includes battery replacement and retirement costs. Fig. 8 summarizes

Table 7
Existing E-bus charging strategies [84].

| Charging Strategy | Battery Size | Capital Cost | Overall System Cost | Applicability in Large-Scales |
|----------------------------|---|--------------|---|--|
| Slow (Depot) | Big For overnight charging, larger capacities are required for batteries. | Low | Average Battery prices are expensive. However, if reduced night tariffs are applied, charging will be economic. | Average (i) With large-scale systems, overnight incurs serious problems in bus queuing, number of chargers, and grid impact. (ii) Large batteries and heavy-weight issues |
| Fast (Opportunity) | Small Buses can top up at terminals in a relatively fast manner, battery packs are small. | Medium | Average The higher cost of the fast charging system is met by a small battery size. | High (i) With large-scale systems, it is applicable in terms of charging time, queuing, and battery size. (ii) Indirectly, distributes the charging demand. |
| Fast (wireless/Pantograph) | Small There is no need for a big battery (en-route charging). | High | Expensive (i) Wireless charging is expensive. (ii) Dedicated to a single route (limited feasibility). | High (i) Pantograph charging is being implemented worldwide. (ii) System cost reduction is proportional to the increased number of buses |

Table 8
Pugh matrix of large-scale deployment of E-bus transit systems.

| Category | Maximum Weight | Fast Charging | Slow charging | Battery Swapping | Wireless |
|--|----------------|-----------------|---------------|------------------|----------------|
| 1. Capital Investment | (15) | 12 ^a | 14 | 7 | 8 |
| 2. Grid Impact | (15) | 10 | 15 | 15 | 12 |
| 3. Charging Time | (10) | 10 | 5 | 9 | 6 |
| 4. Efficiency | (8) | 7 | 4 | 6 | 5 ^b |
| 5. Restrictions on Station Location | (7) | 5 | 3 | 6 | 2 |
| 6. Ease of deployment | (6) | 4 | 4 | 4 | 3 |
| 7. Compatibility with other EV charging | (5) | 4 | 4 | 0 | 0 ^c |
| 8. Quality of service (queue, availability) | (6) | 6 | 4 | 6 | 4 |
| 9. Timetabling and Bus Scheduling | (6) | 6 | 4 | 5 | 4 |
| 10. Grid Service Capability | (6) | 5 | 4 | 6 | 3 |
| 11. Charging System Reliability | (6) | 5 | 5 | 6 | 3 |
| 12. Compliance with grid codes/standards | (6) | 4 | 6 | 6 | 5 |
| 13. Requirement of upgrading grid infrastructure | (4) | 4 | 2 | 2 | 2 |
| Total (100) | | 82 | 70 | 78 | 57 |

^a Expensive charging infrastructure with cheaper and small-size batteries.
^b Poor Power coupling and transfer.
^c Dedicated for certain bus charger configurations.

Table 9
The grid impact aspects of large-scale transportation electrification.

| Positive Grid Impact <i>Ancillary services (Bus-to-Grid)</i> | Negative Grid Impact |
|--|--|
| Power Reserve | Violation of Voltage Drop Limit |
| Peak Shaving | Violation of Frequency Drop |
| Electricity Trade | Power Quality Deterioration |
| Power Quality (Harmonic Compensation) | Power Imbalance |
| Transmission Line Congestion Relief | Increased Power Loss |
| Voltage Regulation | Transformer and Lines Overloading |
| Frequency Regulation | Forming Peak Load Demand |
| Load Factor Enhancement of lightly loaded lines (Improves the utilization factor-valley filling) | Overheating Distribution transformers (Harmonic currents interactions) |

the essential criteria necessary for optimal charging scheduling. It highlights the necessity of holistically considering technical, economic, and environmental factors, particularly when scheduling for fast-charging where the grid impact is significant. It is crucial to emphasize that the level and nature of grid interaction are influenced not only by techno-economic factors but also by environmental considerations. It indirectly facilitates the integration of larger renewable systems with the grid.

8. Ancillary services and grid codes

The ancillary services are vital for ensuring the stability and

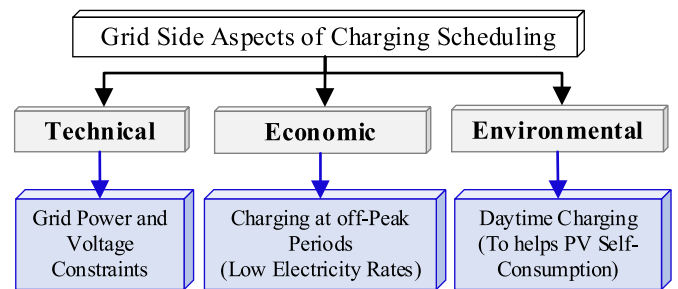


Fig. 8. The aspects associated with optimal charging scheduling.

functionality of the power grid. Ancillary services include services outside of primary power generation and are subject to specific requirements [93]. Most grid codes necessitate the capability to provide reactive power and power reserve as major ancillary services [93]. The requirements aim to guarantee that generating units can provide these services promptly upon request, particularly during emergencies. According to the latest grid code revisions, EV charging stations are now obligated to meet specific requirements. These requirements include power quality, voltage support, demand response capabilities, anti-islanding measures, and the ability to withstand grid conditions, including fault ride-through (FRT). Furthermore, advanced network codes differentiate EVs based on their roles as either demand (V1G) or generation (V2G) entities. The specific ancillary services offered by

large-scale charging stations depend on the technology and capabilities integrated into the charging infrastructure. Moreover, advancements in smart charging and V2G technologies continue to expand the potential contributions of charging infrastructure to the overall stability and efficiency of the power system [93,94].

Large-scale E-bus charging stations can contribute to grid ancillary services. Charging stations can contribute according to demand response agreements. By adjusting the charging tariffs according to grid conditions, they can help balance electricity supply and demand, especially during peak periods. Moreover, they can provide voltage support and reactive power control to help maintain power quality on the grid. This is particularly relevant as charging can lead to voltage drop, and the charging stations can manage the charging to compensate for the resultant drop. If the charging infrastructure supports V2G technology, E-bus can feed surplus energy back into the grid during high-demand periods [95]. This bidirectional flow of electricity can be a valuable asset for grid balancing. E-bus charging stations can align their charging schedules with renewable energy generation patterns, helping to absorb excess renewable energy during peak production times and providing flexibility to the grid.

Unlike medium or high-voltage grids, the low-voltage parameters are characterized by less strict criteria, and there is currently no obligation to provide voltage support [93]. Table 10 compares several grid aspects of fast charging systems that have been discussed in the literature.

9. The coordination between E-bus transit systems and distribution grid

As EVs and their charging stations proliferate, power quality becomes increasingly vital in electric networks due to their non-linear voltage and current profiles. Numerous studies investigate the effects of transportation electrification on the utility grid, especially on the distribution networks. This includes aspects such as substation overloading, harmonics, voltage drops, poor load factor, and unbalanced loadings [109]. Therefore, comprehending the impact of charging stations on power quality is of utmost importance [110]. Other studies focused on the impacts on power grid reliability and stability with the augmented charging demand that maximizes power network loadings. Moreover, grid stability is majorly concerned with the on/off switching of large charging infrastructures and their impacts on system frequency [111,112].

Ensuring adequate power quality is one of the utility's important responsibilities and is typically governed by grid codes (i.e., IEEE 519, IEC 61000). The power quality issues in the presence of power electronics chargers depend on charger type, converter topology, and charger capacity. The hosting capacity (HC) of a given network establishes the maximum load that can be connected while maintaining power quality within acceptable limits. Due to the nonlinear charging power, utility companies are concerned about exceeding the HC of power networks. This concern arises because a poor power quality profile degrades the HC of the networks besides overloading the power transformers and lines [113,114]. Consequently, deploying more charging infrastructures leads to power quality issues. Fig. 9 shows a block diagram of implications associated with the large-scale E-transportation on power grid hosting capacity.

On a large scale, charging stations have a significant impact on power quality where the large impact of harmonic distortion is generated by the multiple chargers, accumulatively. Harmonic distortion can pose issues when the cumulative harmonic signals surpass predefined thresholds, transformers and lines heating losses, and instrumentation malfunction. Several factors determine the impact of harmonics. These factors include the penetration level of chargers and converter topology [115–117]. A significant degradation in grid power quality was observed when a heavy fast charging occurred due to the presence of 3rd and 11th order harmonic currents, and a large drop in voltage and frequency [118,119]. The studies conducted in Refs. [120,121] revealed

that the main concern for fast charging stations is the high THD_i, rather than the overloading of power transformers. The research demonstrated that the number of chargers that can be installed without exceeding the harmonic limits is restricted by the grid code. Deploying the charging stations in higher voltage substations would help compliance with the grid code [122].

When considering a network with both charging stations and inverter-based DG (I-DG), both can contribute common harmonics to the grid. However, if I-DG and charging stations are appropriately sized and located, they can help reduce harmonics by canceling out each other's harmonics [123]. Idaho National Lab (INL) researched EV compatibility with the grid and has published the results of testing the grid impact of level-2 charging [124]. The results show that low-order harmonics contaminate the grid at low-charging power, as shown in Fig. 10(a)-(b). Also, the results show that charging efficiency is enhanced at higher power rates, as shown in Fig. 10(c).

The studies in Refs. [113,125,126, and 126] revealed that at extremely low temperatures (i.e., $\leq 0^{\circ}\text{C}$), THD_i exceeds the acceptable limits. This is attributed to the fact that charging at low temperatures is a low-power process due to the Li-ion electrode plating phenomenon, resulting in a minimal fundamental current flow. As a result, total demand distortion (TDD) can be used to compare charging profiles at various temperatures, as per the IEEE Std 519–2022, which states that the maximum TDD should be 5% [122]. In Ref. [127], it was observed that charging a cold battery took at least two times the room temperature charging. Interestingly, the study showed that the TDD was within the permissible range throughout all charging processes.

Charging infrastructure must be designed to accommodate the specific requirements of battery chemistry used for each charging technology. Compared to nickel-metal hydride (NiMH) and solid-state batteries, the most commercial battery type used in the EV industry is Li-ion batteries [128]. Besides the cost of the battery, important criteria affect the adoption of the charging technology.

These criteria mainly depend on the battery's chemistry, including battery lifespan and number of cycles, maximum charging rate that is affected by the thermal sensitivity of the battery, and cell voltage requirements [128,129]. The literature agrees that Li-ion batteries support faster charging compared to other battery types. Solid-state batteries have the potential to offer even faster charging speeds due to their unique characteristics. However, no broad commercial production exists so far. Furthermore, Li-ion batteries have a lower voltage range and can handle high currents, which is suitable for fast charging technology compared to NiMH, which has a poor thermal coefficient. In Ref. [129], the authors compared three Li-ion chemistries (namely; LFP, NMC, and LTO) used in E-bus regarding the impact of fast charging rate, battery size, and lifespan. The results showed that LTO exhibits the minimum impact on battery lifespan at higher charging rates with a 1.5% yearly degradation. On the other side, the NMC battery loses more than 20% of its capacity with fast charging schemes. Furthermore, it was revealed that all chemistries benefit from the increased capacity where the degradation rate decreases proportionally. The results inferred a stable operation of the LFP battery at different charging rates.

10. The battery performance in harsh climates

As of 2022, the prevailing battery chemistry in the market was lithium nickel manganese cobalt oxide (NMC), constituting a significant 60% market share [3]. Following closely behind was lithium iron phosphate (LFP), with a market share just shy of 30%, and nickel cobalt aluminum oxide (NCA) with a market share of approximately 8% [130]. As the volumetric (Wh/L) and gravimetric (Wh/kg) energy densities of Li-ion are much higher than other technologies, they become more mature and widespread [131]. Table 11 presents the specifications of several Lithium-based batteries and their thermal characteristics for industrial applications.

Similar to cold weather operation conditions of Li-ion batteries, hot

Table 10

Comparison between state-of-the-art studies on grid impact of large-scale fast-charging infrastructures.

| [Ref] | Developed Scheme-Problem | Grid impact Aspect | System Size (kW) |
|-------|--|---|---|
| [96] | A joint optimization of transit network and power grid using a spatial-temporal model to locate sites and sizes of e-bus charging stations- | Power loss and voltage drop minimization | 110 kV lines 30 E-bus 250 km routes |
| [97] | 2020-Warsaw-Poland Pantograph exemplary model of a bus station with charging infrastructure | Power quality (THDv) | 1st: Sparta' nsk-200 kW 2nd: Wilanów-2x400kW |
| [98] | maximizing profit bus of depot operator and grid stability enhancement with PV and energy storage (Depot) | Peak power limits | 50 kW, 100 kW PV 500 kWh ESS |
| [99] | Grid code compliance and grid impact details of Fast charging stations considering the power quality | Voltage fluctuation Harmonic resonant Harmonic distortion | 50 kW DCFC |
| [69] | investigates the performance of different locations and sizes of chargers | estimation of the benefits of PV to reduce impacts on the utility | 100 kW–1200 kW |
| [100] | Bi-level Distribution locational marginal price-based optimization algorithm for E-bus grid interaction | Managing load congestion as a demand response service | 51.5 kW 112 kWh (charge-G2V) |
| [101] | Optimize the charging schedule to minimize the charging costs and the grid impact | Dynamic charging according to variable grid tariff | 47 batteries of 240 kWh each (47 chargers with a charging power of 50 kW) |
| [92] | Impact of fast-charging solution of battery E-bus on distribution networks of two solutions: 1. Pantograph, 2. Slow-charging | Power quality/Harmonic distortion in current and voltages | Fast-chargers (350 kW) using pantograph and slow-chargers (50 kW) |
| [65] | Technical-based feasibility study of Battery Electric Buses with utility impact measures | Distribution transformers lifetime, and grid overloading. voltage regulation | 250 kW-fast 200 kW-depot From 300 kW to 600 kW |
| [102] | grid impact evaluation of 3 charging methods: - Every Station Charging: more intermittent charging power and transformer tapping, - End Station Charging: Less intermittent charging power and transformer tapping - Overnight Charging: continuous charging power and less transformer tapping considering PV and LV grid. | - Power/current Overloading - Voltage drops (The overnight charging is grid-protecting and predictable. However, requires a very large battery (700 kWh)). | |
| [103] | Replacement of the passive diode rectifier with an active front-end converter in the charging station | Power quality/harmonic distortion (THDv), Flickering | 120 kW |
| [104] | Interaction between an bus aggregator and grid operator | Power/current Overloading | – |
| [90] | Study on the determination of the safe level of EV penetration according to transformer loadings | Impact of higher order harmonics on distribution transformers' temperatures | 160 kW |
| [91] | Strategic placement of charging stations considering network capacity and future forecast. | Transformer Overloading and load shifting | 600 kW |
| [85] | Study of the Impact of EV Harmonic Currents on Transformers in the presence of PV system. | The harmonic currents of PV to cancel the injected charging station harmonics | – |
| [105] | Practices used to minimize congestion impact of fast charging grid. | Congestion and power quality | 600 kW (ultra-fast) |
| [106] | Role of E-bus charging in regulating renewable source production to smooth the impact on power demand. A V2G E-bus smart charging is compared to opportunity charging | Smart charging with PV increases the self-consumption of renewable systems by using E-transportation infrastructure as a buffer stage | 80-kWh buses. (Daily total net bus charging MWh = 188.36) |
| [107] | PSO optimization considers the charging model including optimal power flow, degree of satisfaction, and the power grid cost. | Voltage adjustment of 10 kV On-load-tap-changer (OLTC) | – |
| [108] | To study the effect of harmonics injected by EV charging stations with a high level of PV penetration | Host capacity degradation as a result of low-order harmonics | – |

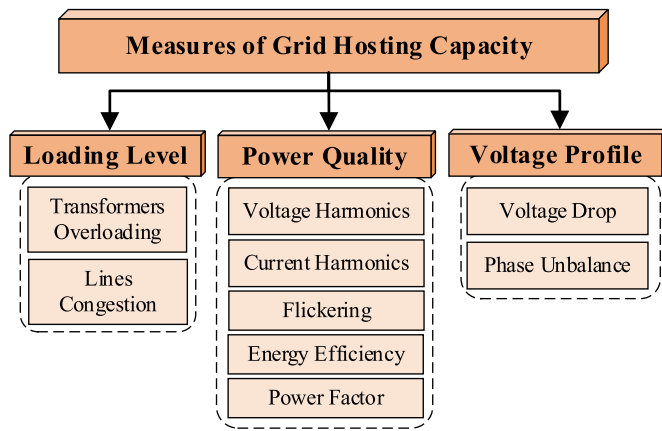


Fig. 9. Criteria affect grid hosting capacity.

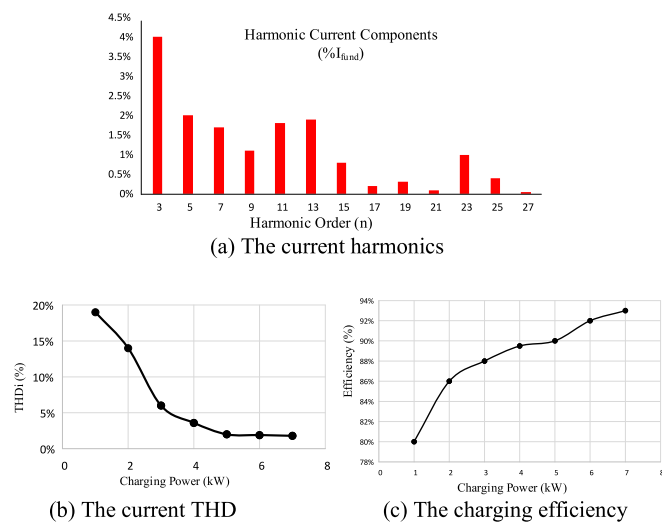


Fig. 10. EV charging and grid compatibility test [124].

ambient temperatures significantly impact Li-ion batteries. Hot temperatures cause unwanted internal reactions that work to deplete and break down the composite bonds. Hot temperatures also cause the capacity to fade, reducing lifespan and safety [132]. Therefore, temperature effects must be considered in Li-ion battery design and operation for EVs. It's important to note that while high temperatures can initially boost the performance of Li-ion batteries by quickly facilitating the reactions, the long-term effects are harmful, leading to a reduced lifespan and potential safety risks. Therefore, it's crucial to manage the temperature conditions of Li-ion batteries effectively [132].

Roundtrip efficiency of the battery operating under a high temperature for a long time will be significantly reduced. Prolonged exposure to elevated temperatures accelerates the degradation of the battery, leading to a permanent loss of capacity over time. This can result in reduced driving range. This is because the internal temperature is an important aging accelerator that influences the battery capacity and power characteristics. Moreover, optimal battery capacity is typically achieved around 21 °C, but it significantly decreases when temperatures exceed 35 °C [132,133]. Consequently, running Li-ion EVs outside the normal range incurs more charging cycles to maintain similar trip distances.

11. Implications of fast charging in hot climates

Due to changes in the electro-chemical properties of Li-ion batteries under different temperatures, non-linear response yields to represent the

Table 11

Thermal specifications of most common Lithium batteries.

| Battery Chemistry | Heat Tolerance Capability (Overheating)- Thermal Stability | Main Features | Safety Under Fast Charging | Generalized Evaluation ^a |
|--|---|---|-------------------------------|-------------------------------------|
| 1. Lithium Iron Phosphate (LFP) | Good thermal stability | Lower Energy Density, (larger and heavier for the same energy capacity) | Most Safe | 83 % |
| 2. Lithium Titanate (LTO) | Exceptional thermal stability | Lower Energy Density (Long cycle life) | Safe | 75 % |
| 3. Lithium Nickel Cobalt Manganese Oxide (NMC) | Moderate thermal stability, degrades in high-temperatures | Good balance of energy and power | Used in most of hybrid EV | 79 % |
| 4. Lithium Cobalt Oxide (LCO) | Sensitive to high temperatures, which can lead to capacity loss | High Energy Density | Safe | 66 % |
| 5. Lithium Nickel Cobalt Oxide (NCA) | Sensitive to high temperatures | High Energy Density | Safe- used in cell phones | 71 % |
| 6. Lithium Manganese Oxide (LMO) | Excellent thermal stability | Moderate Energy Density | less prone to thermal runaway | 66 % |

^a The evaluation is made by evenly scoring the main features:(Cost, Lifespan, Performance, Thermal Stability, Specific Power, Specific Energy).

battery model. This non-linearity adds a layer of complexity to the battery mathematical modeling and the estimation of SOH and state-of-health (SOH) [134]. SOH is essentially a measure of the current energy level in a battery as a percentage of its maximum chargeable energy. SOH, on the other hand, is a long-term indicator of the battery's aging process and represents the ratio between the maximum chargeable energy and its full capacity. While the battery operating temperature increases, its non-linear behavior becomes more pronounced [135]. This increased non-linearity at high temperatures complicates battery modeling and analysis, leading to more complicated management systems for E-bus systems [136].

Long-term exposure to high temperatures can lead to significant damage to cells. As depicted in Fig. 11, an increase in temperature results in a greater degree of capacity fading. High temperatures cause the electrolyte layer to gradually deteriorate and dissolve into the electrolyte. This exposes part of the anode's active material to the electrolyte again, triggering side reactions. Consequently, intercalation at the anode

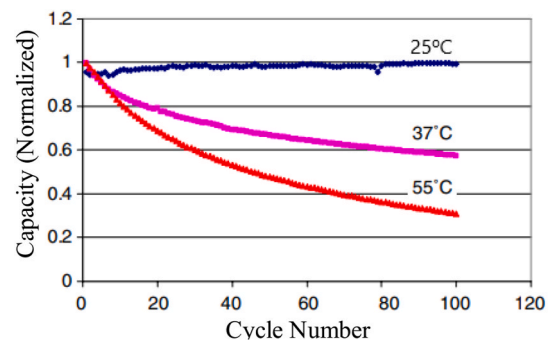


Fig. 11. The accelerated capacity fading due to high temperatures [137].

becomes more challenging, and ionic conductivity decreases [136]. Fig. 12 shows the impact of the fast charging rates (i.e., C-rate>1) on increasing the internal temperature of the battery where a significant increase occurs. This increase would impede the fast charging rates at high ambient temperatures if no efficient thermal extraction mechanism is employed. Detailed comparisons between the impact of different charging rates (C-rate) are performed in Ref. [135].

According to many experimental validations, it has been proven that the optimal operating temperature of an EV Li-ion battery pack is in the range of 15 °C to 35 °C [138–140]. Hence, the degradation is influenced by abnormal temperatures. Higher temperatures cause electrolytes to break down and increase the self-discharge rate of the battery leading to a shorter lifespan [141,142]. To comprehend the degradation behavior of batteries under different temperatures, it's essential to devise modeling methods that accurately represent the electrochemistry degradation [143,144]. Ensuring the safe operation of Li-ion batteries is a significant concern due to the large energy density and inflammable electrolytes.

It is crucial to implement safety measures to prevent thermal runaway as Li-ion batteries become unsafe at high temperatures, leading to uncontrolled electro-chemical reactions and thermal runaway. This can result in battery destruction, increasing the risk of fire and explosion [138]. Table 12 summarizes the major attributes associated with untypical operating temperatures of Li-ion batteries.

Most of Lithium battery types used in E-transportation have been surveyed to address the temperature effect on battery performance. The experimental research performed in Ref. [145] showed that a higher C-rate generates additional heat due to the increased cathode resistance. Therefore, fast charging augments the harsh environmental conditions. In Ref. [146], the authors experimented with the battery lifecycle under 30, 40, and 50°C. It was shown that working with 50°C shrinks the lifespan by about 3 times. In the experiments, the results showed that lifecycle decreases dramatically when the ambient temperature goes above 45°C.

The authors in Ref. [147] performed a practical comparison between EV Li-ion batteries based on their state of function (SOF), considering the degraded lifespan caused by high-temperature operating conditions. The SOF includes a comprehensive evaluation that incorporates various factors influencing the performance of a battery. These factors include the weight of the SOC, charging/discharging rates, environmental temperatures, and other degradation-influencing variables. The essence of SOF lies in its ability to articulate how well the battery performance aligns with the actual power demands during its service life [147]. A direct study on EV consumption and efficiency based on city local temperature was conducted in Ref. [148]. The study considered all US states where the temperature varies from -20 °C to 44 °C according to the state. It was shown the energy consumption of the EV increased significantly when the ambient temperature exceeded 30 °C.

In [149], a review article focused on the barriers faced by the adoption of EVs in Arab Gulf countries, including the high-temperature conditions. In Ref. [150], the authors experimentally investigated the

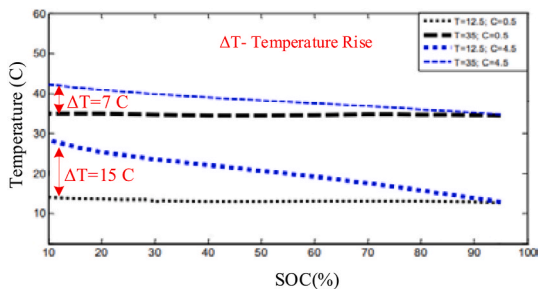


Fig. 12. The impact of fast charging rates on the internal temperature of the battery [136].

Table 12

The impacts of abnormal temperatures on Li-ion batteries.

| Attribute | Hot Temperature Impacts | Cold Temperature Impacts |
|-------------------------|--|--|
| Temperature Range | $T \geq 35\text{ }^{\circ}\text{C}$ | $T \leq 8\text{ }^{\circ}\text{C}$ |
| Root Causes | Electrolyte breakdown Binder decomposition/ breakdown | Increased lithium plating Increased internal resistance |
| Chemical Side Effects | Decreased SOC | Decreased SOC |
| Electric Side Effects | ^a Decreased equivalent resistance (short-term). ^a Increased equivalent resistance (long-term). | Reduced conductivity. Increased equivalent resistance |
| Consequences on Battery | Quick capacity Fading. Complex thermal management system. | Higher Power loss |
| Consequences on EV | Higher cost, less age, more weight, less trip range, low charging/discharging rates, more charging stations, more power quality issues | |
| Long Term Impact | More | Less |

^a note: the internal resistance decreases with the temperature increase; however, with cycling increase, the resistance increases significantly at higher temperatures.

impact of summer ambient temperature in Kuwait on the EV trip range. It has been discovered that more than a 30 % reduction in the estimated travel range occurs due to high temperatures, and the temperature at the start of the trip should be as low as possible to reduce EV energy consumption. Matlab simulations performed in Ref. [151] revealed that the harsh ambient temperatures in some Moroccan cities increase the required number of charging stations. Moreover, it has been depicted that a 20 % increase in charging stations is required when the ambient temperature increases from 29 °C to 45 °C due to the reduction in the actual battery capacity. In Ref. [152], the authors proposed an adaptive thermal management system to cool the battery in harsh temperatures with minimal energy consumption.

The experimental results carried out in Ref. [153] found that fast charging rates of more than 2.0C worsen the degradation (i.e., degradation >0.015 % per cycle) of the battery, considering the cycle duration. Further, as the temperature increases, the degradation increases. In Ref. [154], the authors showed that V2G services can be restricted in high-temperature areas as the demand increases in proportion to ambient temperature. This would increase the cost of the service as it harms the long-term battery performance. Fig. 13 depicts the challenges that face the deployment of E-mobility in hot regions. The root causes and the consequences are highlighted to address the issue and enable

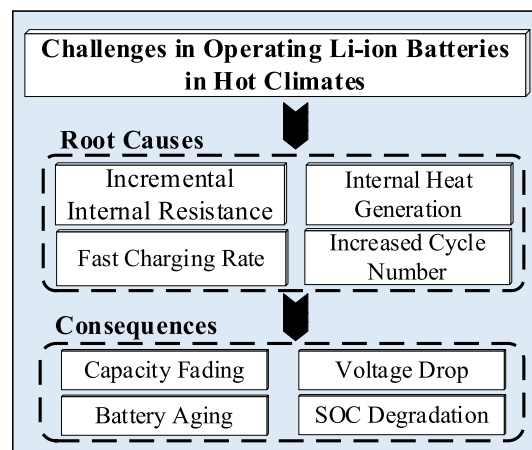


Fig. 13. Challenges imposed by hot climates on the performance of Li-ion batteries.

E-mobility in hot climate regions.

12. Thermal management systems of EV battery

The majority of operational difficulties encountered by operating heavy-duty EVs in high-temperature conditions are associated with the thermal impacts on the Li-ion batteries. As a result, precise representations of the batteries' thermal characteristics are necessary for the design of efficient battery thermal management systems. Active/passive air, fluid coolants, and phase-change materials (PCMs) are frequently used methods for battery cooling in different applications [147,152]. The initial method of heat management considered for high temperatures is air cooling. A battery pack is typically constructed with several parallel cells. This configuration makes air cooling systems a popular choice for heat management. Directing an airflow at the pack can effectively achieve the cooling effect.

In [155], the authors showed that thermal comfort inside an E-bus around 21°C in harsh conditions may increase heating, ventilation, and air conditioning (HVAC) energy consumption by 50 %. The comparison made in Ref. [19] affirmed that excessive capacity fading of Li-ion batteries occurs when the temperature exceeds 35 °C. The authors determined that LFP chemistry has the best evaluation among other chemistries, in addition to being the safest type. The work also discussed the thermal management systems used to cool battery packs. Another piece of research was reported in Ref. [20], where a detailed comparison between the thermal energy management of EV Li-ion batteries was made. The impact of cold and hot temperatures on capacity loss and battery size has been discussed and compared, where the review showed that the effect of hot temperatures (binder decomposition and electrolyte breakdown) is more dangerous than cold temperatures due to the unrecoverable characteristics and thermal explosion possibility. Due to the wide range of operating temperatures, heat pipe (HP) technology-based cooling systems were found to be more efficient than other water and PCM management systems. However, both reviews [19, 20] indicated the necessity for more investigations into heat extraction mechanisms for high charging rates at abnormal operating temperatures, including hybrid management systems for battery pack levels and real applications.

Increased temperatures can speed up reactions and boost charging/discharging power, but they also raise the risk of overheating. If the generated heat inside the battery is not removed fast enough, it can lead to a continuous temperature increase. Therefore, causing a thermal runaway. This is a situation where rising temperatures trigger conditions that lead to even more heat, which may result in battery destruction.

The thermal management system (TMS) of Li-ion batteries consists of active components, which include external and internal cooling sources, and passive systems. One of the key strategies to enhance the operation and lifespan of Li-ion batteries in E-bus under high temperatures is to reduce the internal heat generation within the batteries. The liquid (working fluid) cooling method is more prevalent due to its ease of implementation and lower technical complexity. However, PCM demonstrates superior performance in terms of the speed which can be utilized to cool down the battery [138].

The impact of high temperatures is more complicated compared to low temperatures due to the strong relation between heat generation, internal resistance, charging rate, and cycling [156]. Air cooling presents challenges due to air's low conductivity, making this method suitable for moderate temperatures and charging rates [138]. Both liquid and air-cooling systems are costly due to the need for a pump and ventilation system. As a result, PCMs have been introduced as advanced methods for EVs in low-temperature environments [157,158]. PCMs are substances that can store and release a significant amount of heat which provides an efficient heat transfer mechanism [140]. During the transformation of solid-liquid material, heat is either absorbed or released.

As shown in Table 13 where the cooling systems are compared, it's evident that each mechanism has its unique strengths and weaknesses,

which are influenced by several factors. These include the specific chemistry of the battery, the temperature, and the driving cycle of the bus [159–162]. Generally, active cooling strategies are deemed more efficient in dissipating heat generated within the battery [163]. The efficiency of a specific TMS should be assessed based on an exact EV location, and its impact on battery long-term performance. Conversely, alternative strategies focus on developing battery components like the electrode and electrolyte to withstand broader ranges of temperatures [164]. According to the literature discussed before, it has been inferred that PCM provides an optimal thermal manager due to its satisfactory performance in addition to the major criteria like weight, energy consumption, and size.

In real-world tests, the researchers in Ref. [124] measured the SOC of different EV brands. The tests were conducted under freezing, typical, and hot ambient temperatures to capture the impact of the thermal management system on the battery performance. The results are given in Table 14 where it can be seen how installing an effective system is important in maintaining the battery capacity.

For a detailed analysis of the ambient temperature effect on EVs, Fig. 14 (a) and (b) show how severe ambient temperatures degrade the performance of the battery and the kilometers driven by an EV. This degradation occurs in many EV brands with some differences that exist according to the used TMS.

14. The research essential outcomes

It is not possible to optimally plan for large-scale E-transport networks without collaborating and preplanning with utility grid operators. The optimal solutions require coordinated charging systems whether including unidirectional or bidirectional power flow (i.e., G2B, B2G). The following points are highlighted in the study.

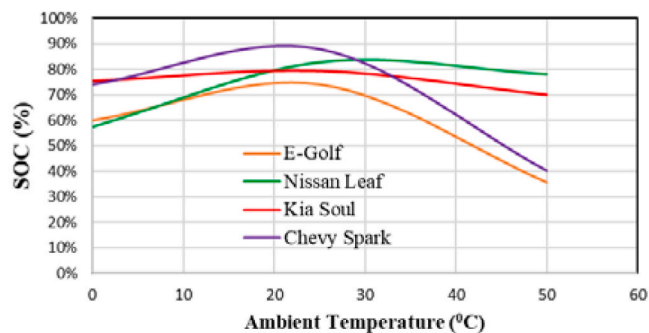
1. The paper conducted a detailed literature survey of the most relatively experimental-based articles that discuss the impact of harsh environmental conditions on battery performance, charging rates, and ancillary services. It has been shown that much of the existing literature has assumed that EVs operate under normal temperatures. However, EVs experience a range of temperatures in countries with abnormal climates, where transportation electrification is on the rise.
2. Operating E-bus in cold temperatures has been focused on in most of the literature as EVs are heavily integrated in many European and Canadian cities that have low temperatures. Therefore, the hot environment is focused on in this work to facilitate the deployment of E-bus in hot climate countries.
3. It is important to mention that many pieces of research found a significant discrepancy between the laboratory and field results of Li-ion battery characterization under different temperatures. Moreover,

Table 13
Main features of main thermal management strategies for EV batteries.

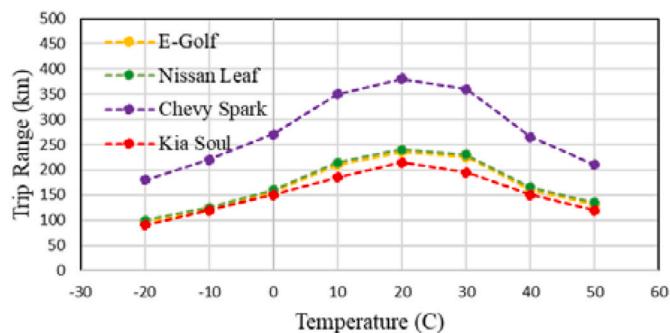
| Cooling System | Advantages | Disadvantages |
|--|--|--|
| Air Cooling System (Passive, active) | Easy to implement, less technical complexity | Low heat transfer efficiency |
| Coolant Cooling System (Passive, active) | High heat transfer efficiency, effective in reducing temperature influence | High energy consumption, complexity, high heat transfer efficiency |
| Refrigerant Cooling System | Effective cooling power, good for high temperatures | Environmental impact, safety concerns |
| PCM-based Cooling System | High heat storage capacity, good for low temperatures | Less mature technology, expensive |
| Heat Pipe | High cooling power, efficient heat transfer, moderate complex design | High cost |
| Thermoelectric Cooling System | Direct conversion of temperature differences to electric voltage | Low efficiency at small temperature differences |

Table 14
EV battery SOC measured values after 30 min at different ambient temperatures [124].

| Make/Model | SOC at 0 °C | SOC at 25 °C | SOC at 50 °C | Cooling Mechanism |
|------------------------|-------------|--------------|--------------|--------------------------------|
| 2015 E-Golf | 60 % | 74.2 % | 35.6 % | Passive- Air (Radiator) |
| 2013 Nissan Leaf | 57.5 % | 82.8 % | 78.1 % | Passive- Air (Radiator) |
| 2015 Mitsubishi i-MiEV | 31.8 % | 43.5 % | 32 % | Active -Air (Air conditioning) |
| 2015 Kia soul | 75.4 % | 79.3 % | 70 % | Active-Air (Air conditioning) |
| 2015 Chevy Spark | 74.2 % | 88.1 % | 40.1 % | Active-Cooling liquid |



(a) The temperature impact on the SOC



(b) the temperature impact on the EV trip range

Fig. 14. The behavior of SOC profiling and the ranges of different EVs at different temperatures [124].

It has been investigated that even though high temperature decreases the internal resistance of Li-ion batteries which, in turn, increases its Ah-capacity, binder decomposition and electrolyte breakdown cause fast aging. This would converse with the discussion made in Ref. [113] regarding hot temperature. In conclusion, three points should be carefully addressed to avoid the conflict:

- (i) An increase in temperature correlates with a decrease in battery internal resistance,
 - (ii) the internal resistance tends to rise with an increased number of battery cycles, and
 - (iii) fast charging increases the internal temperature of the battery.
4. It is shown in this paper that the optimal solution that ensures reliable integration of fast-charging infrastructure can be achieved when the following aspects are holistically considered.
- Proper coordination with grid operator including Bus-to-Grid service.
 - Enable short-term load forecasting, load shifting, and peak shaving.
 - Enabling demand response by establishing an incentive-based electricity market.

5. As discussed throughout the paper, there is no simple answer to optimal size, location, or charging system type. The work highlights all the factors that affect the reliability and feasibility of large-scale fast-charging systems.
6. Besides addressing the harmonic effect on the grid hosting capacity, research should look into the energy efficiency aspects of the harmonics. This is because the time horizon of the charging profile may conflict with the consequences of different THD levels. In other words, a long-term, low THD charging profile is worse than a short-term, High THD in terms of energy efficiency.
7. After analyzing several tests conducted by Idaho National Lab [124] on Level-2 (AC) and level-3 (DC) fast charging it has been concluded that the hot climate areas (40+ °C) have more potential to have feasible E-transportation systems compared to cold areas (0 °C) in many aspects. These aspects include the charging time, the round-trip efficiency, and capacity fading. However, the successful operation of fast charging E-bus systems in hot temperatures depends on adaptively coordinating an advanced thermal management system (as discussed in Refs. [19,20]) with adaptive charging/discharging rates and customized cooling mechanisms.
8. It has been concluded that fast charging rates increase the internal temperature of the battery. Therefore, deploying fast-charging infrastructure in hot ambient temperatures requires careful attention including shading and ventilation of the charging stations.

13. Li-ion battery cycling under high temperatures

High temperatures can have damaging effects on both the charging and discharging processes of Li-ion batteries. After a certain threshold, they can cause capacity degradation, thermal-safety issues, reduced efficiency, and overall diminished battery performance. Hence, it is crucial to assess the thermal efficacy of Li-ion batteries following a sufficient number of tests, as the impacts of many factors appear after a specific number of cycles and a considerable service duration. Table 15 compares the consequences between the two modes and their effects on the battery performance.

To further get insights into the effect of temperature on Li-ion batteries, Fig. 15 shows the testing results of an EV battery SOC profile and the corresponding battery temperature at different operating temperatures. It can be seen how the high temperature affects the charging time (i.e., 15 min more) and degrades the battery voltage which impedes the battery from reaching a higher SOC level (i.e., 27 % less).

In [165], the effectiveness of the power conversion of a fast charging system showed a slight reduction when the charging temperature was shifted from 25°C to 40°C where the efficiency decreased from 93 % to around 90 %. In Phoenix, AZ, E-bus route testing where summer temperatures touch the upper 40s (i.e., $T > 40^{\circ}\text{C}$), it was revealed that hot climates extend the charging time of the E-buses due to uncontrolled heat generation. In addition, it was recommended to shade and air condition the charging stations to satisfy charging time [166].

This city is among very few hot climate cities that integrate E-bus services and shows the importance of shading and air conditioning the charging infrastructure itself which affects the economic aspects of such projects.

In [167], the authors have introduced statistical models for predicting battery charging durations based on ambient temperature. Using regression analysis, they have developed a formula that represents the relationship between the final SOC, charging duration, ambient temperature, and initial SOC. Notably, their empirical model demonstrates that as ambient temperature increases, the time required to achieve a high SOC decreases. However, the validation has only been conducted at temperatures of 0 °C and 25 °C which may not be valid for higher temperatures and limits the correctness of the results.

Given the importance of documenting the results of practical experiments related to the automotive industry, the charging profiles of two common EVs are depicted in Figs. 16 and 17. The figures

Table 15

The major aspects associated with the cycling of Li-ion batteries at high temperatures.

| Charging Mode | | Discharging Mode | |
|-----------------------------------|--|-----------------------------------|--|
| Challenge | Comment | Challenge | Comment |
| Increased Reaction Rate | Hot temperatures augment the reaction rate, resulting in an augmented power output. But, this increases heat dissipation which adds extra unwanted heat. | Increased Power Output | High temperatures can increase the power output of the battery. |
| Slow Heat Dissipation | If heat isn't dissipated faster than it's generated, the temperature will continue to rise, potentially leading to a thermal runaway situation. | Electrolyte Transfer Speed | As the temperature increases faster, the transfer rate of the electrolyte decreases. |
| Temperature Limits | The permissible charge temperature is between 0 °C and 45 °C. Good charge performance is ensured at high temperatures but it leads to shorter life. | Performance Impact | Lithium-ion batteries can handle an elevation in temperatures. However, keeping the battery discharging for long periods at higher temperatures may lead to gas generation. |
| Charging Restrictions | As the temperature goes high, the chemical balance can be destroyed and side reactions start. | Voltage Drop | The battery's internal resistance can cause a significant voltage drop during discharge, reducing the available power output. |
| Safety Concerns | Charging Li-ion batteries in high temperatures leads to partial breakdown and initiates thermal runaway | Reduced Energy Efficiency | High-temperature discharging is less energy-efficient. More of the energy is dissipated as heat due to increased internal resistance. |
| Reduced Capacity | The reactions that occur during charging are less efficient at elevated temperatures, resulting in a lower amount of charge stored. | Shortened Lifespan | Similar to charging, discharging a Li-ion battery at high temperatures can also negatively affect its lifespan. The accelerated chemical reactions at elevated temperatures can lead to quicker degradation of the battery's components. |
| Shortened Battery Lifespan | Prolonged charging at high temperatures can cause accelerated degradation of the battery. This can result in a shorter overall lifespan. | | |

demonstrate the effects of ambient temperatures on charging duration. For the temperatures considered in the test, the charger maintains a steady power output during the constant current phase until reaching a specific SOC. After that, constant voltage mode starts where the charging power begins to decrease gradually at higher SOCs. Several minutes longer charging time of high-temperature charging can be observed. Comparing the three charging profiles, it can be seen that the 35 °C profile takes a longer time to finish the charging cycle and the battery consumes more power required by the battery management

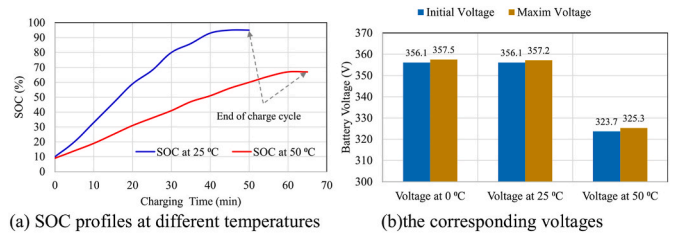


Fig. 15. Complete charging cycle measurement of Li-ion 2015 E-Golf EV and the battery voltage.

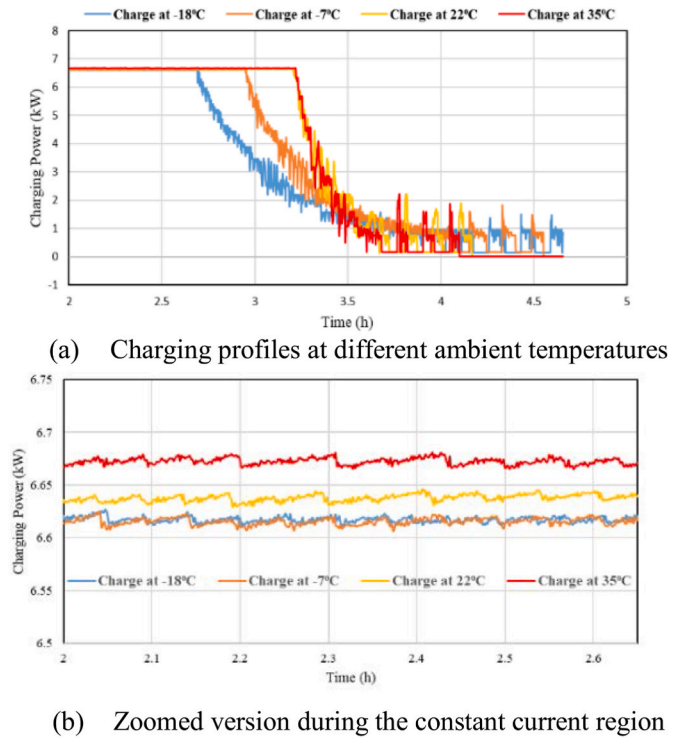
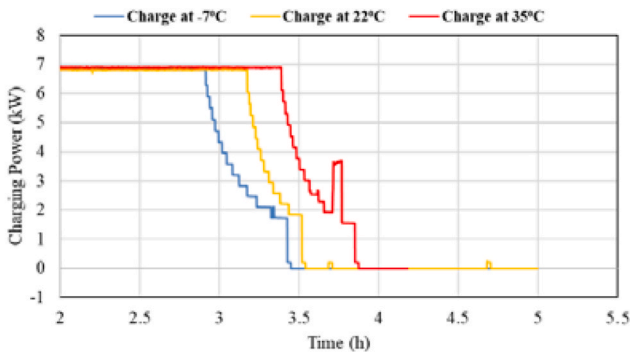


Fig. 16. Level- 2 charging profiles of Li-ion Nissan Leaf battery under different ambient temperatures [168].

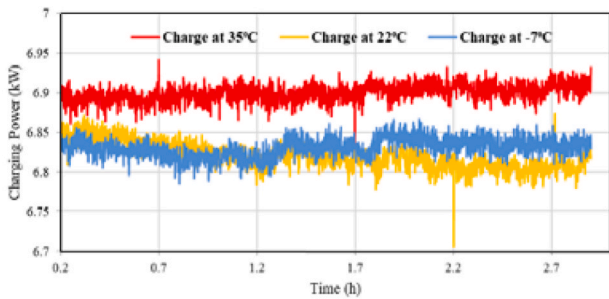
system.

The report made by the American Automobile Association (AAA) [169] showed that several tests have been made to show the impact of cold (-7 °C) and hot (35 °C) temperatures on the energy consumption of different EV brands. For the high-temperature operation, the comparison compared the normal 22 °C temperature to 35 °C where variations in ambient temperatures led to reductions in trip range. Moreover, the utilization of heating, ventilation, and air conditioning (HVAC) systems leads to substantial reductions in trip ranges and reduced fuel economy. These reductions are influenced by temperature-dependent factors and battery capacity. Normally, when the temperature reached 35 °C, it caused a 4 % reduction in driving range and a 5 % reduction in equivalent fuel efficiency referring to 24 °C test conditions, considering that HVAC was not running. After turning on the HVAC, substantial declines in trip range and fuel efficiency were observed. It is important to mention that the target of this paper is to address the consequences of temperatures higher than 35 °C, whose impacts are more significant and pronounced. However, to the best of the authors' knowledge, limited real automotive-related experimentations and resources are reported in the literature.

To quantify the impact of the HVAC on the trip range, Fig. 18 (a) and (b) present a bar graph of fuel consumption in Mi/Gallon values of different EVs before and after operating the HVAC system at three



(a) Charging profiles at different ambient temperatures



(b) Zoomed version during the constant current region

Fig. 17. Level-2 charging profiles of Li-ion Ford Focus battery under different ambient temperatures [168].

different ambient temperatures. The values show the significant impact of the HVAC on the fuel economy of EVs. It can be concluded from Fig. 18 that both hot and cold ambient temperatures, caused a tangible decrease in trip range and efficiency. At low temperatures (-7°C), a 12 % decrease in trip range occurs. At high temperatures (35°C), around a 5 % decrease in the trip range compared to the 24°C operating climate. When HVAC is ON, -7°C ambient temperature decreases the trip range by about 40 % whereas 35°C driving conditions decrease the range by almost 17 % compared to 24°C driving temperature [169]. Higher temperatures would lead to increased HVAC energy consumption.

A nonlinear degradation model presented in Eq. (1), can be used to accommodate the battery performance model in a more accurate way compared with many linear models. The model illustrates a specific case involving a temperature effect where the model parameters are denoted as β_0 , β_1 , and ρ . The average performance is characterized by a degradation model, $\mu(T, t)$. It's important to note that this model is semi-empirical in nature. Accordingly, the change in the degradation parameter is constrained by the availability of reactive materials. The final model can vary significantly according to the calculated parameters. Fig. 19 shows the resulting nonlinear model fit. The adverse effect

of high temperature on increasing the resistance and thereby decreasing the capacity can be observed.

$$\mu(T, t) = [1 + e^{t(\beta_0 + \beta_1/T)}]^{\rho} \quad (1)$$

The employed degradation model can be utilized to assess the average lifespan of a cell under specific ambient temperature conditions and predefined end-of-life (EOL) criteria. The determination of EOL criteria is contingent upon various factors encompassed within the life model. In instances where a solitary stress factor, such as temperature, and a performance metric of relative resistance are considered, the EOL criterion may be specified as a 30 % augmentation in degradation at a given temperature, denoted as $\mu(T_0, t) = 1.3$ at EOL. The relative resistance (i.e., battery resistance after a certain number of cycles divided by the initial resistance) is used as a performance measure to quantify the battery degradation. The measure is used as the internal impedance of the battery increases over time irrespective of ambient temperature.

Regarding the nonlinear degradation model delineated in Eq. (1), the resultant estimated lifetime, (\hat{t}_{EOL}), is expressed in Eq. (2), where parameters marked with the hat symbol represent values derived from nonlinear regression-based estimations. The resultant degradation model can predict the average lifespan of a cell under defined temperature conditions and EOL criteria. By employing the nonlinear degradation model presented in Eq. (2), the estimated lifetime can be computed using parameters acquired through nonlinear regression. Assuming an EOL criterion of 30 % rise in relative resistance at $T_0 = 30^{\circ}\text{C}$, substituting the relevant parameters into Eq. (2) results in a calculated lifetime estimation of 12.0 years.

$$\hat{t}_{EOL} = \frac{\mu(T_0, t)^{\frac{1}{\rho}} - 1}{e^{\left(\beta_0 + \frac{\beta_1}{T_0}\right)}} \quad (2)$$

The authors in Ref. [170] conducted a test on EVs to characterize the impact of hot climates under different charging modes. The concluded

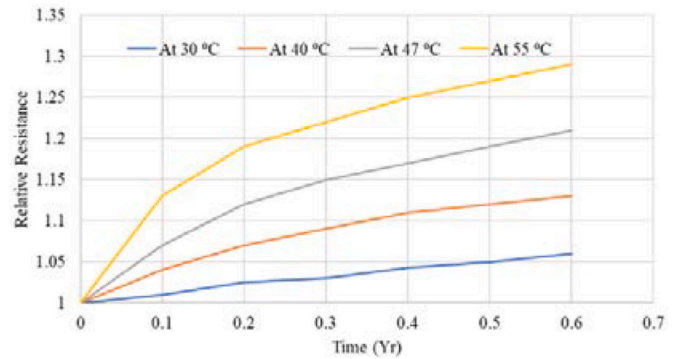


Fig. 19. Temperature-dependent fitted nonlinear model.

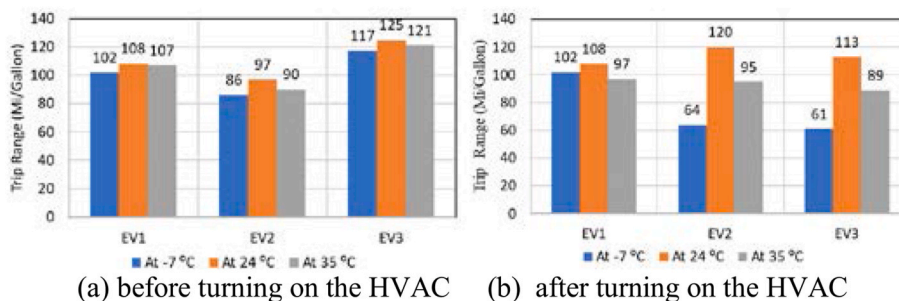


Fig. 18. The ambient temperature impact on Li-ion EV with and without turning on the HVAC.

results are depicted in Fig. 20 where it can be seen that the fast charging causes additional heating to the battery. On top of that, the results show that there was a slightly greater reduction in battery capacity when being fast-charged. It's worth mentioning that higher ambient temperatures seemed to accelerate the capacity loss in all the E-buses in the study. Additionally, the same study revealed that fast charging, on average, raised the battery temperature by approximately 8°C. Fig. 21 shows the measured value of the internal resistance and its relation to the charging speed. This factor is very important to explain the decaying performance of EV batteries under both high temperatures and fast charging.

15. Research gaps

- Considering the grid impact when planning for fast charging systems is crucial and includes several technical aspects. Fast charging incurs extra risk and costs on the investment when charging during peak hours and high electricity rates. However, the majority of the conducted research focused on cost minimization without identifying realistic grid impacts.
- One of the research gaps found in the literature is the missing mathematical model and circuit model of Li-Ion batteries at high temperatures. After obtaining the experimental models, the existing models have been identified through several parameter identification techniques as performed in many pieces of research.
- Some discrepancies were found in the literature regarding the exact effect of high temperatures on the performance of Li-ion batteries. Therefore, concise and detailed analyses should be performed considering the application, size, charging speed, and actual conditions when comparing the results.
- Another research gap found in the literature is the very limited studies on E-bus systems performance in hot climates.

16. Conclusion

This paper investigates the optimal deployment of E-bus charging infrastructures in harsh climate cities. The investigations consider the impact of high temperatures, Li-ion battery performance and characteristics, fast charging rates, advantages and disadvantages of grid interaction, types of charging stations, and optimal approaches in planning, designing, and operating the charging infrastructure.

The substantial influence of high temperatures on Li-ion batteries is highlighted and discussed. The aspects include long-term performance, capacity deterioration, electro-chemical reactions, thermal management schemes, and safety. The effect of higher temperatures has been specified and compared to help alleviate the temperature effect on reliable deployment of E-bus transit systems in hot climate cities. The discussion highlighted a significant reduction in the driving range of EVs, with the decrease being further exacerbated by the increased energy consumption for HVAC. It negatively affects the performance of Li-ion batteries, leading to a reduced E-bus trip. High temperatures accelerate degradation, reducing the overall lifespan of the battery. This, in turn, can increase the cost of battery replacement or maintenance. High temperatures can lead to thermal runaway and potentially hazardous situations. Implementing efficient thermal management systems, utilizing cutting-edge technologies, and using standby and pre-conditioning techniques can all help regulate and maintain the battery's temperature within an ideal range in order to lessen the negative effects of temperature on Li-ion batteries. Solutions proposed in the literature focused on advanced cooling, battery oversizing, regulated charging, and deploying more charging stations.

The operation of Li-ion batteries in hot temperatures has great potential compared to cold weather operations with the existence of an efficient thermal management system. This would emphasize the high possibility of deploying E-buses in hot climate countries while ensuring a feasible, safe, and environmentally friendly transportation system.

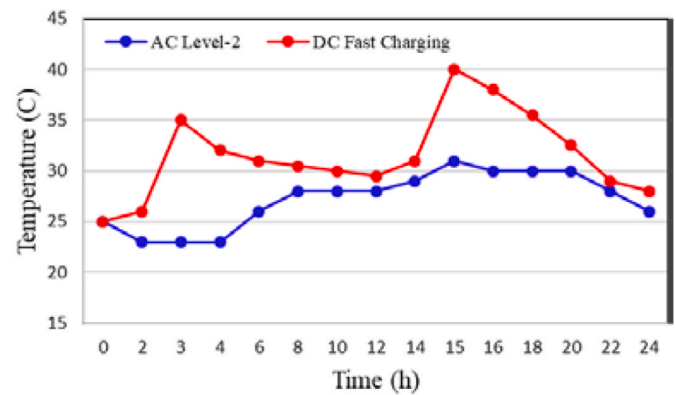


Fig. 20. Charging mode effect on battery temperature [170].

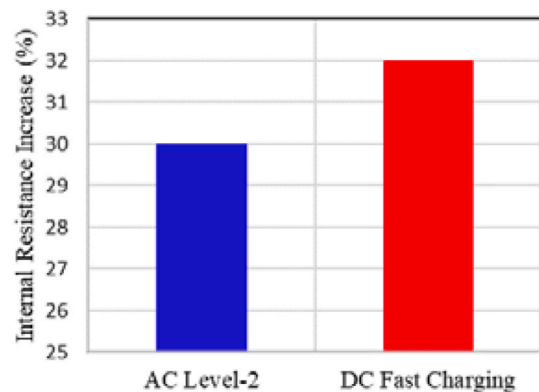


Fig. 21. Percentage increase of the battery internal resistance measured after 80, 000 km [170].

CRedit authorship contribution statement

Salman Harasis: Conceptualization, Data curation, Formal analysis, Writing – original draft, Writing – review & editing. **Irfan Khan:** Supervision, Validation, Writing – review & editing. **Ahmed Massoud:** Funding acquisition, Methodology, Project administration, Resources, Supervision, Validation, Writing – original draft, Writing – review & editing.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Data availability

No data was used for the research described in the article.

Acknowledgment

This work was supported by Qatar University under the Post-Doc Initiative Program-Cycle 1 (Grant Number: POSTDOC-CENG-2022-554). The findings achieved herein are solely the responsibility of the authors.

References

- [1] Keller V, Lyseng B, Wade C, Scholtysik S, Fowler M, Donald J, Palmer-Wilson K, Robertson B, Wild P, Rowe A. Electricity system and emission impact of direct

- and indirect electrification of heavy-duty transportation. *Energy* 2019;172:740–51.
- [2] Fusco Rovai F, Regina da Cal Seixas S, Keutenedjian Mady CE. Regional energy policies for electrifying car fleets. *Energy* 2023;278:127908.
 - [3] <https://IEA.BLOB.CORE.WINDOWS.NET/assets/dacf14d2-eabc-498a-8263-9f97fd5dc327/GEVO2023.pdf>. [Accessed 18 September 2023].
 - [4] Prohaska R, et al. Field evaluation of medium duty Plug-in electric Delivery trucks. National Renewable Energy Laboratory; December 2016. NREL/TP 5400-66382.
 - [5] Du J, Li F, Li J, Wu X, Song Z, Zou Y, Ouyang M. Evaluating the technological evolution of battery electric buses: China as a case. *Energy* 2019;176:309–19.
 - [6] IEA, Electric bus registrations and sales share by region, 2015–2022, IEA, Paris, <https://www.iea.org/data-and-statistics/charts/electric-bus-registrations-and-sales-share-by-region-2015-2022..>
 - [7] Foda A, Abdelaty H, Mohamed M, El-Saadany E. A generic cost-utility-emission optimization for electric bus transit infrastructure planning and charging scheduling. *Energy* 2023;277:127592.
 - [8] Zafar MH, Mansoor M, Abou Houran M, Khan NM, Khan K, Raza Moosavi SK, Sanfilippo F. Hybrid deep learning model for efficient state of charge estimation of Li-ion batteries in electric vehicles. *Energy* 2023;282:128317.
 - [9] Babaeiyazdi I, Rezaei-Zare A, Shokrzadeh S. State of charge prediction of EV Li-ion batteries using EIS: a machine learning approach. *Energy* 2021;223:120116.
 - [10] Zhao L, Ke H, Li Y, Chen Y. Research on personalized charging strategy of electric bus under time-varying constraints. *Energy* 2023;276:127584.
 - [11] Alamatsaz K, Hussain S, Lai C, Eicker U. Electric bus scheduling and Timetabling, fast charging infrastructure planning, and their impact on the grid: a review. *Energies* Oct. 2022;15(21):7919.
 - [12] Zhang C, Wang H, Wu L. Life prediction model for lithium-ion battery considering fast-charging protocol. *Energy* 2023;263:126109.
 - [13] Kong W, Luo Y, Feng G, Li K, Peng H. Optimal location planning method of fast charging station for electric vehicles considering operators, drivers, vehicles, traffic flow and power grid. *Energy* 2019;186:115826.
 - [14] Hartvigsson E, Taljegard M, Odenberger M, Chen P. A large-scale high-resolution geographic analysis of impacts of electric vehicle charging on low-voltage grids. *Energy* 2022;261:125180.
 - [15] El-Taweel NA, Zidan A, Farag HEZ. Novel electric bus energy consumption model based on probabilistic Synthetic speed profile integrated with HVAC. In: *IEEE Transactions on Intelligent transportation systems*, vol. 22; March 2021. p. 1517–31. 3.
 - [16] Cui D, Wang Z, Liu P, Wang S, Dorrell DG, Li X, Zhan W. Operation optimization approaches of electric vehicle battery swapping and charging station: a literature review. *Energy* 2023;263:126095.
 - [17] Rajani B, Kommula BN. An optimal energy management among the electric vehicle charging stations and electricity distribution system using GPC-RERNN approach. *Energy* 2022;245:123180.
 - [18] Jiang Z, Han J, Li Y, Chen X, Peng T, Xiong J, Shu Z. Charging station layout planning for electric vehicles based on power system flexibility requirements. *Energy* 2023;283:128983.
 - [19] Liu H, Wei Z, He W, Zhao J. Thermal issues about Li-ion batteries and recent progress in battery thermal management systems: a review. *Energy Convers Manag* 2017;150:304–30. ISSN 0196-8904.
 - [20] Kumar R, Goel V. A study on thermal management system of lithium-ion batteries for electrical vehicles: a critical review. *J Energy Storage* 2023;71:108025. ISSN 2352-152X.
 - [21] Behi H, Karimi D, Jaguemont J, Gandoman FH, Kalogiannis T, Berecibar M, Van Mierlo J. Novel thermal management methods to improve the performance of the Li-ion batteries in high discharge current applications. *Energy* 2021;224:120165.
 - [22] Subramanian V, Feijoo F, Sankaranarayanan S, Melendez K, Das TK. A bilevel conic optimization model for routing and charging of EV fleets serving long distance delivery networks. *Energy* 2022;251:123808.
 - [23] Nasiri N, Zeynali S, Ravadanegh SN. Tactical participation of a Multienergy system in distributionally robust wholesale electricity market under high penetration of electric vehicles. *IEEE Syst J* March 2023;17(1):1465–76. <https://doi.org/10.1109/JSYST.2022.3221490>.
 - [24] Francfort J, Salisbury S, Smart J, Garetson T, Karner D. " considerations for corridor and community DC fast charging complex system design.", U.S. Department of Energy Vehicle Technologies Office 2017 Annual Merit Review And Peer Evaluation Meeting 2017.
 - [25] Lin Y, Zhang K, Shen Z–JM, Miao. Charging network planning for electric bus cities: a case study of Shenzhen, China. *Sustainability* 2019;11:4713. <https://doi.org/10.3390/su11174713>.
 - [26] Lajunen A. Lifecycle costs and charging requirements of electric buses with different charging methods. *J Clean Prod* 2018;172:56–67. <https://doi.org/10.1016/j.jclepro.2017.10.066>. ISSN 0959-6526.
 - [27] Jahic M, Plenz M, Eskander M, Schulz D. Route scheduling for centralized electric bus depots. *IEEE Open Journal of Intelligent Transportation Systems* 2021;2:149–59. <https://doi.org/10.1109/OJITS.2021.3096115>.
 - [28] Sasiidharan C, Ray A, Das S. Selection of charging technology for electric bus fleets in intra-city public transport in India. In: 2019 global conference for advancement in technology (GCAT); 2019. Bangalore, India.
 - [29] El-Taweel NA, Farag HEZ, Barai G, Zeineldin H, Al-Durra A, El-Saadany EF. A systematic approach for design and analysis of electrified public bus transit fleets. *IEEE Syst J* June 2022;16(2):2989–3000. <https://doi.org/10.1109/JSYST.2021.3114271>.
 - [30] Göhlich D, Fay T, Park S. Conceptual design of urban E-bus systems with special focus on battery technology. In: *Proceedings of the design society: international conference on engineering design*, vol. 1; 2019. p. 2823–32. <https://doi.org/10.1017/dsi.2019.289>. 1.
 - [31] Al-Saadi M, Patkowski B, Zaremba M, Karwat A, Pol M, Chelchowski L, Mierlo JV, Berecibar M. Slow and fast charging solutions for Li-ion batteries of electric heavy-duty vehicles with fleet management strategies. *Sustainability* 2021;13:10639. <https://doi.org/10.3390/su131910639>.
 - [32] Al-Saadi M, Mathes M, Käsgen J, Robert K, Mayrock M, Mierlo JV, Berecibar M. Optimization and analysis of electric vehicle operation with fast-charging technologies. *World Electr. Veh. J.* 2022;13.
 - [33] Kong L, Zhang H, Li W, Bai H, Dai N. Spatial–Temporal scheduling of electric bus fleet in power-transportation coupled network. *IEEE Transactions on Transportation Electrification* June 2023;9(2):2969–82.
 - [34] González S, Feijoo F, Basso F, Subramanian V, Sankaranarayanan S, Das TK. Routing and charging facility location for EVs under nodal pricing of electricity: a bilevel model solved using special ordered set. *IEEE Trans Smart Grid* July 2022;13(4):3059–68.
 - [35] Fredriksson H, Dahl M, Holmgren J. Optimal placement of charging stations for electric vehicles in large-scale transportation networks. *Procedia Comput Sci* 2019;160:77–84.
 - [36] Hasan MM, Avramis N, Ranta M, Saez-de-Ibarra A, El Baghdadi M, Hegazy O. Multi-objective energy management and charging strategy for electric bus fleets in cities using various ECO strategies. *Sustainability* 2021;13(14).
 - [37] Gowda SN, Ahmadian A, Anantharaman V, Chu C-C, Gadh R. Power management via integration of battery energy storage systems with electric bus charging. In: 2022 IEEE power & energy society innovative smart grid technologies conference (ISGT); 2022. New Orleans, LA, USA.
 - [38] Chen N, Tan CW, Quek TQS. Electric vehicle charging in smart grid: optimality and valley-filling algorithms. *IEEE Journal of Selected Topics in Signal Processing* 2014;8(6).
 - [39] Zhang M, Li W, Yu SS, Wen K, Zhou C, Shi P. A unified configurational optimization framework for battery swapping and charging stations considering electric vehicle uncertainty. *Energy* 2021;218:119536.
 - [40] Zhan W, Wang Z, Zhang L, Liu P, Cui D, Dorrell DG. A review of siting, sizing, optimal scheduling, and cost-benefit analysis for battery swapping stations. *Energy* 2022;258:124723.
 - [41] Liang Y, Cai H, Zou G. Configuration and system operation for battery swapping stations in Beijing. *Energy* 2020;214:118883.
 - [42] Kunith A, Mendelevitch R, Goehlich D. Electrification of a city bus network—an optimization model for cost-effective placing of charging infrastructure and battery sizing of fast-charging electric bus systems. *Int. J. Sustain. Transp.* 2017;11:707–20.
 - [43] Teichert O, Chang F, Ongel A, Lienkamp M. Joint optimization of vehicle battery pack capacity and charging infrastructure for electrified public bus systems. *IEEE Trans. Transp. Electrific. Sep.* 2019;5(3):672–82.
 - [44] Hussain A, Bui V-H, Kim H-M. Optimal sizing of battery energy storage system in a fast EV charging station considering power outages. *IEEE Transactions on Transportation Electrification* June 2020;6(2):453–63. <https://doi.org/10.1109/TTE.2020.2980744>.
 - [45] Ahmad MS, Sivasubramani S. Optimal number of electric vehicles for existing networks considering economic and emission dispatch. *IEEE Trans Ind Inf* April 2019;15(4):1926–35. <https://doi.org/10.1109/TII.2018.2861409>.
 - [46] Wang J, Kang L, Liu Y. Optimal scheduling for electric bus fleets based on dynamic programming approach by considering battery capacity fade. *Renewable Sustainable Energy Rev* 2020;130:13.
 - [47] Manzolli JA, Tróvão JPF, Henggeler Antunes C. Electric bus coordinated charging strategy considering V2G and battery degradation. *Energy* 2022;254:124252.
 - [48] Lim LK, Muis ZA, Ho WS, Hashim H. Review of the energy forecasting and scheduling model for electric buses. *Energy* 2023;263:125773.
 - [49] Li JQ. Battery-electric transit bus developments and operations: a review. *Int. J. Sustain. Transp.* 2016;10:157–69.
 - [50] Mahmoud M, Garnett R, Ferguson M, Kanaroglou P. Electric buses: a review of alternative powertrains. *Renew Sustain Energy Rev* 2016;62:673–84.
 - [51] Perumal SS, Lusby RM, Larsen J. Electric bus planning & scheduling: a review of related problems and methodologies. *Eur J Oper Res* 2021;301:395–413.
 - [52] Manzolli JA, Tróvão JP, Antunes CH. A review of electric bus vehicles research topics—Methods and trends. *Renew Sustain Energy Rev* 2022;159:112211.
 - [53] Arif SM, Lie TT, Seet BC, Ayyadi S, Jensen K. Review of electric vehicle technologies, charging methods, standards and optimization techniques. *Electronics* 2021;10:1910. <https://doi.org/10.3390/electronics10161910>.
 - [54] Deng R, Liu Y, Chen W, Liang H. A survey on electric buses energy storage, power management, and charging scheduling. *IEEE Trans Intell Transport Syst* Jan. 2021;22(1):9–22.
 - [55] Alrubaie AJ, Salem M, Yahya K, Mohamed M, Kamarol M. A comprehensive review of electric vehicle charging stations with solar photovoltaic system considering market, technical requirements, network implications, and future challenges. *Sustainability* 2023;15.
 - [56] Rajendran G, Vaithilingam CA, Mison N, Naidu K, Ahmed MR. A comprehensive review on system architecture and international standards for electric vehicle charging stations. *J Energy Storage* 2021;42.
 - [57] Secchi M, Andersen PB. Flexible energy systems leveraging the optimal integration of EVs deployment Wave- Deliverable 1.2 [Online]. Available: <https://www.theflowproject.eu/>; 2022.
 - [58] U.S. Virgin Islands transportation electrification roadmap. 2022 [Online]. Available: <https://electrificationcoalition.org/roadmaps>.

- [59] Electric buses in cities: driving towards cleaner air and lower CO2 [Online]. Available: <https://about.bnef.com/blog/electric-buses-cities-driving-towards-cleaner-air-lower-co2>; 2018.
- [60] The World Bank. Electrification of public transport: a case study of Shenzhen bus Group. Washington, DC: World Bank; 2021. License: Creative Commons Attribution CC BY 3.0 IGO.
- [61] RMI India. Electric vehicle charging infrastructure: a Guide for Discom readiness. 2019.
- [62] Roychowdhury A. Decarbonising transport: what Does it mean for India? New Delhi: Centre for Science and Environment; 2023.
- [63] Archana N, Rajeev T. A novel reliability Index based approach for EV charging station allocation in distribution system. *IEEE Trans Ind Appl Nov.-Dec. 2021*;57(6):6385–94.
- [64] Secchi M, Andersen PB. FLOW's D1.2 - Baseline: scenarios and definitions. Flow; 2023.
- [65] Mohamed M, Farag H, El-Taweel N, Ferguson M. Simulation of electric buses on a full transit network: operational feasibility and grid impact analysis. *Elec Power Syst Res 2017*;142:163–75.
- [66] Yi Z, Chen B, Liu XC, Wei R, Chen J, Chen Z. An agent-based modeling approach for public charging demand estimation and charging station location optimization at urban scale. *Comput Environ Urban Syst 2023*;101:101949. <https://doi.org/10.1016/j.compenvurbsys.2023>. ISSN 0198-9715.
- [67] Csiszár C, Csonka B, Földes D, Wirth E, Lovas T. Urban public charging station locating method for electric vehicles based on land use approach. *J Transport Geogr 2019*;74:173–80. <https://doi.org/10.1016/j.jtrangeo.2018.11.016>. ISSN 0966-6923.
- [68] Csiszár C, Csonka B, Földes D, Wirth E, Lovas T. Location optimisation method for fast-charging stations along national roads. *J Transport Geogr 2020*;88:102833. <https://doi.org/10.1016/j.jtrangeo.2020.102833>. ISSN 0966-6923.
- [69] Dougier N, Celik B, Chabi-Sika S-K, Sechilariu M, Locment F, Emery J. Modelling of electric bus operation and charging process: potential contribution of local photovoltaic production. *Appl Sci 2023*;13:4372.
- [70] Rohrbeck B, Berthold K, Hettich F. Location planning of charging stations for electric city buses considering battery ageing effects. In: *Operations research Proceedings 2017*. Berlin/Heidelberg, Germany: Springer; 2018. p. 701–7.
- [71] Liu Z, Song Z, He Y. Planning of fast-charging stations for a battery electric bus system under energy consumption uncertainty. *Transport Res Rec 2018*;2672:96–107.
- [72] He Y, Song Z, Liu Z. Fast-charging station deployment for battery electric bus systems considering electricity demand charges. *Sustain Cities Soc 2019*;48:101530.
- [73] Olmos J, López JA, Gaztañaga H, Herrera VI. Analysis of optimal charging points location and storage capacity for hybrid and full electric buses. In: *Proceedings of the 2019 Fourteenth International Conference on Ecological vehicles and renewable Energies (EVER)*, Monte-Carlo, Monaco, 8–10 may; 2019. p. 1–7.
- [74] Lin Y, Zhang K, Shen ZJM, Ye B, Miao L. Multistage large-scale charging station planning for electric buses considering transportation network and power grid. *Transport Res C Emerg Technol 2019*;107:423–43.
- [75] Liu T, Ceder AA. Battery-electric transit vehicle scheduling with optimal number of stationary chargers. *Transport Res C Emerg Technol 2020*;114:118–39.
- [76] Othman AM, Gabbar HA, Pino F, Repetto M. Optimal electrical fast charging stations by enhanced descent gradient and Voronoi diagram. *Comput Electr Eng 2020*;83:106574.
- [77] Zhang W, Zhao H, Song Z. Integrating transit route network design and fast charging station planning for battery electric buses. *IEEE Access 2021*;9:51604–17.
- [78] Wu X, Feng Q, Bai C, Lai CS, Jia Y, Lai LL. A novel fast-charging stations locational planning model for electric bus transit system. *Energy 2021*;224:120106.
- [79] Stumpe M, Rößler D, Schryen G, Kliever N. Study on sensitivity of electric bus systems under simultaneous optimization of charging infrastructure and vehicle schedules. *EURO J. Transp. Logist. 2021*;10:100049.
- [80] Hu H, Du B, Liu W, Perez P. A joint optimization model for charger locating and electric bus charging scheduling considering opportunity fast charging and uncertainties. *Transport Res C Emerg Technol 2022*;141:103732.
- [81] Wang Y, Liao F, Lu C. Integrated optimization of charger deployment and fleet scheduling for battery electric buses. *Transport Res Transport Environ 2022*;109:103382.
- [82] Li P, Jiang M, Zhang Y, Zhang Y. Cooperative optimization of bus service and charging schedules for a fast-charging battery electric bus network. *IEEE Trans Intell Transport Syst May 2023*;24(5):5362–75. <https://doi.org/10.1109/TITS.2023.3243105>.
- [83] Kocer MC, Onen A, Jung J, Gultekin H, Albayrak S. Optimal location and sizing of electric bus battery swapping station in microgrid systems by considering Revenue maximization. *IEEE Access 2023*;11:41084–95.
- [84] International Energy Agency (IEA). "Trucks and buses". Available online: <https://www.iea.org/reports/trucks-and-buses>. [Accessed 2 October 2023].
- [85] Alame D, Azzouz M, Kar N. Assessing and mitigating impacts of electric vehicle harmonic currents on distribution systems. *Energies 2020*;13:3257. <https://doi.org/10.3390/en13123257>.
- [86] Adequacy & flexibility study for Belgium (2024-2034)," available online: <https://www.elia.be/en/electricity-market-and-system/adequacy/adequacy-studies>.
- [87] Chen H, Hu Z, Zhang H, Luo H. Coordinated charging and discharging strategies for plug-in electric bus fast charging station with energy storage system. *IET Gener Transm Distrib May 2018*;12(9):2019–28.
- [88] Cheng Y, Wang W, Ding Z, He Z. Electric bus fast charging station resource planning considering load aggregation and renewable integration. *IET Renew Power Gener 2019*;13(7):1132–41.
- [89] Alwesabi Y, Liu Z, Kwon S, Wang Y. A novel integration of scheduling and dynamic wireless charging planning models of battery electric buses. *Energy 2021*;230:120806.
- [90] Visakh A, Parvathy Selvan M. Analysis and mitigation of the impact of electric vehicle charging on service disruption of distribution transformers. *Sustainable Energy, Grids and Networks 2023*;35:101096. ISSN 2352-4677.
- [91] van den Berg MA, Lampropoulos I, AlSkaif TA. Impact of electric vehicles charging demand on distribution transformers in an office area and determination of flexibility potential. *Sustainable Energy, Grids and Networks 2021*;26:100452. <https://doi.org/10.1016/j.segan.2021.100452>. ISSN 2352-4677.
- [92] Al-Saadi M, Bhattacharyya S, Tichelen PV, Mathes M, Käsge J, Van Mierlo J, Berecibar M. Impact on the power grid caused via Ultra-fast charging technologies of the electric buses fleet. *Energies 2022*;15:1424. <https://doi.org/10.3390/en15041424>.
- [93] IRENA. "Grid codes for renewable powered systems. International Renewable Energy Agency," 2022. Abu Dhabi, ISBN: 978-92-9260-427-1.
- [94] Harighi T, Bayindir R, Hossain E. Overview of quality of service evaluation of a charging station for electric vehicle. In: *Renewable energy research and applications (ICRERA)*, 2017 IEEE 6th International Conference on; 2017. San Diego, CA, USA.
- [95] Shi J, Zeng T, Moura S. Electric fleet charging management considering battery degradation and nonlinear charging profile. *Energy 2023*;283:129094.
- [96] Davidov S, Pantoš M. Planning of electric vehicle infrastructure based on charging reliability and quality of service. *Energy 2016*;118:1156–67.
- [97] Zagrajek K, Paska J, Klos M, Pawlak K, Marchel P, Bartecka M, Michalski Ł, Terlikowski P. Impact of electric bus charging on distribution substation and local grid in Warsaw. *Energies 2020*;13:1210.
- [98] Arif SM, Lie TT, Seet BC, Ahsan SM, Khan HA. Plug-in electric bus depot charging with PV and ESS and their impact on LV feeder. *Energies 2020*;13:2139.
- [99] García-Triviño P, Torreglosa JP, Fernández-Ramírez LM, Jurado F. Control and operation of power sources in a medium-voltage direct-current microgrid for an electric vehicle fast charging station with a photovoltaic and a battery energy storage system. *Energy 2016*;115:38–48.
- [100] Wu Z, Guo F, Polak J, Strbac G. Evaluating grid-interactive electric bus operation and demand response with load management tariff. *Appl Energy 2019*;255:113798. ISSN 0306-2619.
- [101] Abdelwahed A, van den Berg PL, Brandt T, Collins J, Ketter W. Evaluating and optimizing opportunity fast-charging schedules in transit battery electric bus networks. *Transport Sci 2020*;54(6):1601–15.
- [102] Stahleder D, Reihls D, Ledinger S, Lehfuß F. Impact assessment of high power electric bus charging on urban distribution grids. In: *45th Annual Conference of the IEEE industrial electronics Society*. Lisbon, Portugal: IECON 2019; 2019. p. 4304–9. <https://doi.org/10.1109/IECON.2019.8927526>.
- [103] Pirouzi S, Aghaei J, Niknam T, Farahmand H, Korpås M. Exploring prospective benefits of electric vehicles for optimal energy conditioning in distribution networks. *Energy 2018*;157:679–89.
- [104] Clairand JM, González-Rodríguez M, Terán PG, Cedenño I, Escrivá-Escrivá G. The impact of charging electric buses on the power grid. In: *2020 IEEE power & energy Society general meeting (PESGM)*; 2020. p. 1–5. <https://doi.org/10.1109/PESGM41954.2020.9282014>. Montreal, QC, Canada.
- [105] Oualmakran Y, De Vroey L, Novak-Zdravkovi A, Lehfuß F, Hegazy O. "Mitigating the impact of high power charging of electric buses: perspective of European distribution grid operators,". In: *33rd electric vehicle Symposium (EVS33)*; June 14 - 17, 2020. Portland, Oregon.
- [106] Wellik T, Griffin J, Kockelman K, Mohamed M. Utility-transit nexus: Leveraging intelligently charged electrified transit to support a renewable energy grid. *Renew Sustain Energy Rev 2021*;139:110657.
- [107] Yang J, He L, Fu S. An improved PSO-based charging strategy of electric vehicles in electrical distribution grid. *Appl Energy 2014*;128:82–92.
- [108] Dávila-Sacoto M, González LG, Hernández-Callejo L, Duque-Pérez Ó, Zorita-Lamadrid ÁL, Alonso-Gómez V, Espinoza JL. Harmonic distortion and hosting capacity in electrical distribution systems with high photovoltaic penetration: the impact of electric vehicles. *Electronics 2023*;12(11):2415.
- [109] Khalid MR, Alam MS, Sarwar A, Asghar MSJ. "A comprehensive review on electric vehicles charging infrastructures and their impacts on power-quality of the utility grid,". vol. 1. *eTransportation*; Aug. 2019, 100006.
- [110] Terkes M, Arikani O, Gokalp E. The effect of electric vehicle charging demand variability on optimal hybrid power systems with second-life lithium-ion or fresh Na-S batteries considering power quality. *Energy 2024*;288:129760.
- [111] Taljegard M, Göransson L, Odenberger M, Johnsson F. Impacts of electric vehicles on the electricity generation portfolio—a scandinavian/German case study. *Appl Energy Feb. 2019*;235:1637–50.
- [112] Wang Y, Su H, Wang W, Zhu Y. The impact of electric vehicle charging on grid reliability. *IOP Conf Ser Earth Environ Sci 2018*;199(5):052033.
- [113] Kong W, Luo Y, Feng G, Li K, Peng H. Optimal location planning method of fast charging station for electric vehicles considering operators, drivers, vehicles, traffic flow and power grid. *Energy 2019*;186:115826.
- [114] Borghetti F, Colombo CG, Longo M, Mazzoncini R, Panarese A, Somaschini C. Vehicle-to-grid: a case study of ATM E-bus depots in the city of Milan in Italy. In: *2021 AET International Annual Conference (AETI)*; 2021. Milan, Italy.
- [115] Lucas A, Bonavitacola F, Kotsakis E, Fulli G. Grid harmonic impact of multiple electric vehicle fast charging. *Elec Power Syst Res Oct. 2015*;127:13–21.

- [116] Dharmakeerthi CH, Mithulananthan N, Saha TK. Impact of electric vehicle fast charging on power system voltage stability. *Int J Electr Power Energy Syst May* 2014;57:241–9.
- [117] Torres S, Durán I, Marulanda A, Pavas A, Quirós-Tortós J. Electric vehicles and power quality in low voltage networks: real data analysis and modeling. *Appl Energy Jan.* 2022;305:117718.
- [118] Lai CS, Chen D, Zhang J, Zhang X, Xu X, Taylor GA, Lai LL. Profit maximization for large-scale energy storage systems to enable fast EV charging infrastructure in distribution networks. *Energy* 2022;259:124852.
- [119] Shafiei M, Ghasemi-Marzbali A. Electric vehicle fast charging station design by considering probabilistic model of renewable energy source and demand response. *Energy* 2023;267:126545.
- [120] Bosak A, Kulakovskiy L, Oboronov T. Impact of EV chargers on total harmonic distortion in the distribution system network. In: 2019 IEEE 6th International Conference on energy smart systems (ESS); 2019. Kyiv.
- [121] Beaudé O, Lasaulce S, Hennebel M, Mohand-Kaci I. Reducing the impact of EV charging operations on the distribution network. *IEEE Trans Smart Grid* 2016;7(6):2666–79.
- [122] IEEE standard for harmonic control in electric power systems. In: IEEE Std 519-2022 (revision of IEEE Std 519-2014); 5 Aug. 2022. p. 1–31. <https://doi.org/10.1109/IEEESTD.2022.9848440>.
- [123] Misra R, Paudyal S. Analysis and reduction of total harmonic distortions in distribution system with electric vehicles and wind generators. *Proc. IEEE Power Energy Soc. Gen. Meeting; Jul.* 2015. p. 1–5.
- [124] Idaho National Laboratory, “Charging Device Testing,” Accessed on 10 October. 2023, Available: <https://avt.inl.gov/eve-button/abb.html>.
- [125] Bordes A, et al. A holistic contribution to fast innovation in electric vehicles: an overview of the DEMOBASE research project. *eTransportation* 2022;11:100144. <https://doi.org/10.1016/j.etrans.2021.100144>.
- [126] Awasthi A, Venkiteswarlu B, Padmanaban S, Selvamuthukumar R, Blaabjerg F, Singh AK. Optimal planning of electric vehicle charging station at the distribution system using hybrid optimization algorithm. *Energy* 2017;133:70–8.
- [127] Andersson D, Carlsson D. Measurement of ABB’s prototype fast charging station for electric vehicles. In: M.S. Thesis, Dept. Energy Environ. Division Electr. Power Eng. Gothenburg, Sweden: Chalmers Univ. Technol.; 2012.
- [128] Zhang X, Li Z, Luo L, Fan Y, Du Z. A review on thermal management of lithium-ion batteries for electric vehicles. *Energy* 2021;238:121652.
- [129] Al-Saadi M, Olmos J, Saez-de-Ibarra A, Van Mierlo J, Berecibar M. Fast charging impact on the lithium-ion batteries’ lifetime and cost-effective battery sizing in heavy-duty electric vehicles applications. *Energies* 2022;15:1278. <https://doi.org/10.3390/en15041278>.
- [130] <https://data.bloomberglp.com/professional/sites/24/2018/05/Electric-Buses-in-Cities-Report-BNEF-C40-Citi.pdf>. [Accessed 19 October 2023].
- [131] <https://assets.bhub.io/professional/sites/24/2018/05/Electric-Buses-in-Cities-Report-BNEF-C40-Citi.pdf>. [Accessed 20 October 2023].
- [132] Saghir MZ, Bicer Y. Thermohydraulic performance of ammonia, isopropanol, water and nanofluids as cooling fluid for lithium-ion 1C and 3C rating batteries. *International Journal of Thermofluids* 2023;20:100433. <https://doi.org/10.1016/j.ijft.2023.100433>. ISSN 2666-2027.
- [133] Saghir MZ, Rahman MM, Bicer Y. Investigation of channel materials toward better cooling lithium-ion batteries in the presence of nanofluid and pin-fins. *International Journal of Thermofluids* 2023;18:100349. <https://doi.org/10.1016/j.ijft.2023.100349>. ISSN 2666-2027.
- [134] Zha Y, He S, Meng X, Zuo H, Zhao X. Heat dissipation performance research between drop contact and immersion contact of lithium-ion battery cooling. *Energy* 2023;279:128126.
- [135] E S, Liu Y, Cui Y, Wu A, Yin H. Effects of composite cooling strategy including phase change material and cooling air on the heat dissipation performance improvement of lithium-ion power batteries pack in hot climate and its catastrophe evaluation. *Energy* 2023;283:129074.
- [136] Gao Y, Zhu C, Zhang X, Guo B. Implementation and evaluation of a practical electrochemical-thermal model of lithium-ion batteries for EV battery management system. *Energy* 2021;221:119688.
- [137] Amine K, Liu J, Belharouk I. High-temperature storage and cycling of C-LiFePO₄/graphite Li-ion cells. *Electrochem Commun* 2005;7(7):669–73. ISSN 1388-2481.
- [138] Chavan S, Venkateswarlu B, Prabhakaran R, Salman M, Joo SW, Choi GS, Kim SC. Thermal runaway and mitigation strategies for electric vehicle lithium-ion batteries using battery cooling approach: a review of the current status and challenges. *J Energy Storage* 2023;72(Part D):108569. <https://doi.org/10.1016/j.est.2023.108569>. ISSN 2352-152X.
- [139] An Z, Zhang C, Luo Y, Zhang J. Cooling and preheating behavior of compact power Lithium-ion battery thermal management system. *Appl Therm Eng* 2023; 226:120238. <https://doi.org/10.1016/j.applthermaleng.2023.120238>. ISSN 1359-4311.
- [140] Liu Z, Wang B, Tan Y, Li P. Thermal management of lithium-ion battery pack under demanding conditions and long operating cycles using fin-enhanced PCMs/water hybrid cooling system. *Appl Therm Eng* 2023;233:121214. ISSN 1359-4311.
- [141] Kucinskis G, Bozorgchenani M, Feinauer M, Kasper M, Wohlfahrt-Mehrens M, Waldmann T. Arrhenius plots for Li-ion battery ageing as a function of temperature, C-rate, and ageing state – an experimental study. *J Power Sources* 2022;549:232129. ISSN 0378-7753.
- [142] Yang Y, Hu X, Qing D, Chen F. Arrhenius Equation-based cell-health assessment: application to thermal energy management design of a HEV NiMH battery pack. *Energies* 2013;6:2709–25. <https://doi.org/10.3390/en6052709>.
- [143] Martyushev NV, Malozyomov BV, Sorokova SN, Efremkov EA, Qi M. Mathematical modeling of the state of the battery of cargo electric vehicles. *Mathematics* 2023;11:536. <https://doi.org/10.3390/math11030536>.
- [144] Huang L, Piontek U, Chen M, Zheng R, Zhuang L, Zou D. Thermal performance of cold plate based on phase change emulsion for Li-ion battery. *Energy* 2023;282:128743.
- [145] Ouyang D, Hu J, Chen M, Weng J, Huang Q, Liu J, Wang J. Effects of abusive temperature environment and cycle rate on the homogeneity of lithium-ion battery. *Thermochim Acta* 2019;676:241–8. ISSN 0040-6031.
- [146] Soltsev A, Asoyan A, Nikitin D, Bagrin V, Fediushkina O, Evtykov S, Marusin A. Influence of temperature on the performance and life cycle of storage batteries. *Transport Res Procedia* 2021;57:652–9.
- [147] Rezvanzaniani SM, Liu Z, Chen Y, Lee J. Review and recent advances in battery health monitoring and prognostics technologies for electric vehicle (EV) safety and mobility. *J Power Sources* 2014;256:110–24. ISSN 0378-7753.
- [148] Yuksel T, Michalek JJ. Effects of regional temperature on electric vehicle efficiency, range, and emissions in the United States. *Environ Sci Technol* 2015;49(6):3974–80. <https://doi.org/10.1021/es505621s>.
- [149] Alotaibi S, Omer S, Su Y. Identification of potential barriers to electric vehicle adoption in oil-producing nations—the case of Saudi Arabia. *Electricity* 2022;3: 365–95. <https://doi.org/10.3390/electricity3030020>.
- [150] Hamwi H, Rushby T, Mahdy M, Bahaj AS. Effects of high ambient temperature on electric vehicle efficiency and range: case study of Kuwait. *Energies* 2022;15: 3178. <https://doi.org/10.3390/en15093178>.
- [151] El Hafdaoui H, El Alaoui H, Mahidat S, El Harmouzi Z, Khallaayoun A. Impact of hot Arid climate on optimal placement of electric vehicle charging stations. *Energies* 2023;16:753. <https://doi.org/10.3390/en16020753>.
- [152] Singirikonda S, Obulesu YP. Adaptive secondary loop liquid cooling with refrigerant cabin active thermal management system for electric vehicle. *J Energy Storage* 2022;50:104624. ISSN 2352-152X.
- [153] Ouyang D, Weng J, Chen M, Wang J. Impact of high-temperature environment on the optimal cycle rate of lithium-ion battery. *J Energy Storage* 2020;28:101242. ISSN 2352-152X.
- [154] Wen S, Lin N, Huang S, Wang Z, Zhang Z. Lithium battery health state assessment based on vehicle-to-grid (V2G) real-world data and natural gradient boosting model. *Energy* 2023;284:129246.
- [155] Cigarini F, Schminkel P, Sonnekalm M, Best P, Göhlich D. Determination of improved climatic conditions for thermal comfort and energy efficiency in electric buses. *Appl Ergon* 2022;105:103856. ISSN 0003-6870.
- [156] Ma S, Jiang M, Tao P, Song C, Wu J, Wang J, Deng T, Shang W. Temperature effect and thermal impact in lithium-ion batteries: a review. *Prog Nat Sci: Mater Int* 2018;28(6):653–66. ISSN 1002-0071.
- [157] Chang TB, Xiao YZ, Liu YF. Development of energy saving battery Pre-Cooling system for electric vehicles. *Sustainability* 2023;15:13182. <https://doi.org/10.3390/su151713182>.
- [158] Li J, Zhu Z. “Battery thermal management systems of electric vehicles,” Master’s Thesis in automotive Engineering. Göteborg, Sweden: Chalmers University of Technology; 2014.
- [159] Ayetor G, Mbonigaba I, Sunnu AK, Nyantekyi-Kwakye B. Impact of replacing ICE bus fleet with electric bus fleet in Africa: a lifetime assessment. *Energy* 2021;221: 119852.
- [160] Perugu H, Collier S, Tan Y, Yoon S, Herner J. Characterization of battery electric transit bus energy consumption by temporal and speed variation. *Energy* 2023; 263:125914.
- [161] Basma H, Mansour C, Haddad M, Nemer M, Stabat P. Comprehensive energy modeling methodology for battery electric buses. *Energy* 2020;207:118241.
- [162] Zhou B, Wu Y, Zhou B, Wang R, Ke W, Zhang S, Hao J. Real-world performance of battery electric buses and their life-cycle benefits with respect to energy consumption and carbon dioxide emissions. *Energy* 2016;96:603–13.
- [163] Yang Y, Yang L, Du X, Yang Y. Pre-cooling of air by water spray evaporation to improve thermal performance of lithium battery pack. *Appl Therm Eng* 2019;163: 114401. <https://doi.org/10.1016/j.applthermaleng.2019.114401>. ISSN 1359-4311.
- [164] Li N, Liu X, Yu B, Li L, Xu J, Tan Q. Study on the environmental adaptability of lithium-ion battery powered UAV under extreme temperature conditions. *Energy* 2021;219:119481.
- [165] Trentadue G, Lucas A, Otura M, Pliakostathis K, Zanni M, Scholz H. Evaluation of fast charging efficiency under extreme temperatures. *Energies* Jul. 2018;11(8): 1937.
- [166] McGuffie M. Driving the shift to electric buses in hot climates- battery electric buses in the Phoenix Metropolitan area. https://www.azta.org/images/uploads/e-vent-files/Driving_the_Shift_to_Electric_Buses_in_Hot_Weather_Climates_VM.PDF.pdf; 2021.
- [167] Motoaki Y, Yi W, Salisbury S. Empirical analysis of electric vehicle fast charging under cold temperatures. *Energy Pol* 2018;122:162–8.
- [168] Idaho National Laboratory. Charging power profiles at varying temperatures. Available: <https://avt.inl.gov/project-type/data.html>. [Accessed 3 September 2023].
- [169] American Automobile Association (AAA) Electric Vehicle Range Testing. Amer. Automobile Assoc., Heathrow. USA: FL; 2019.
- [170] Burke A, Miller M, Zhao H. Fast charging tests (up to 6C) of lithium Titanate cells and Modules: electrical and thermal response. UC Davis: Institute of Transportation Studies; 2012.