

Scheduled Charging and Discharging of Electric Vehicles for Charging Cost Optimization

Naveen^{1,*}, Puneet Pahuja²

Abstract

The growing number of Electric Vehicles (EVs) will have remarkable impacts on the power grid as it leads to the increase in total energy consumption. This increases the overall burden on the grid which would require new power plants. However, it is possible to minimize the impact by proper scheduling for charging and discharging of electric vehicles. In this paper, three optimal schemes have been compared i.e. global optimal scheduling scheme and local optimal scheduling scheme and equal allocation scheme for EV charging and discharging. The first objective is to flatten the load profile. The EVs can be charged during off peak hours when the demand is less and can be discharged during peak hours when the demand is high. It will result in the constant power demand from the grid. The second objective is to minimize the total charging cost through scheduled charging and discharging. The globally optimal scheduling scheme provides the globally minimal total cost but this scheme is impractical as it requires information about future base loads and arrival time of each EV and charging periods of EV's that will arrive in future. While local optimal scheduling scheme is quite practical as it considers only current ongoing EV sets. This scheme is scalable and resilient to dynamic EV arrival. In equal allocation scheme; the charging power of an EV is allocated based on the electricity price on the previous day in an interval. Secondly, the charging power is equal in each interval. The total cost for each scheme is calculated. Moreover, through the software tool, we demonstrate that the locally optimal scheduling scheme can achieve a close performance compared to the globally optimal scheduling scheme.

Keywords: Electrical vehicles, charging, load profile, local optimal scheduling, global optimal scheduling equal allocation

INTRODUCTION

Due to growing concern on environmental issues such as carbon dioxide emission and global warming, the electric vehicles (EVs) have come out to as a better alternative.

Moreover, it also reduces dependency on non-renewable sources such as oil. EVs have been

accepted by public in past years. In 2016, more than 1, 30,000 electric vehicles were sold in U.S. only and sale of EVs grew by 37.5% to 22,000 units in India during last year. However some schemes only dealt with charging even without vehicle to grid while other schemes.

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Electric Vehicle (EV) technology is more efficient than gasoline and diesel powered vehicles due to its less emission and simple drive train mechanism. The first EV invented in the year 1834 with a non-rechargeable battery [1]. The first rechargeable EV is built with the lead-acid battery in 1874. In the year 1894, Pedro G. Salom and

Henry G. Morris developed an electric automobile named 'Electroboat' [2]. After 1894 different countries like England, the United States and France involved in developing electric automobiles [3].

After the invention of a rechargeable battery-based storage system, EVs had two significant developments. The first one was a reduction in the price of EV. Henry Ford did it in the year 1925 and named as Ford Model T by increasing the volume of the production even though the EVs are costlier than Internal Combustion Engine (ICE) based cars [4].

The second development was in the year 1950 by the California Air Resources Board (CARB), and the development was commercializing the EV usage, which is reducing the emission by 2% [5]. The General Motors developed three experimental EVs named Electrovair, Electrován and Electrovette in the years between 1966 to 1979 [6]. At the same time, the National Aeronautics and Space Administration (NASA) developed a lunar rover in 1971 that runs with the electricity [7]. This concept helped to raise the usage of the electric vehicle. The Nissan initialized the research and development of battery storage based on lithium-ion battery technology. However, Nissan released the first lithium-ion battery based EV by the name

Prairie Joy EV [8]. Besides, Nissan developed another EV named Altera, which is driven by the permanent magnet synchronous motor drive. This popular product developed as Nissan Leaf in 2010. The Tesla Roadster was more popular in the year 2008 because it covers 200 miles at one charge. The different companies followed Tesla's model and developed their own hybrid and pure electric vehicles. The Tesla developed EVs and named as Model S, Model 3, Model X and Model Y with the different driving range like 630 km, 519 km, 565 km, and 509 km with different specifications. The development stages of EVs are shown in Figure 1.

In recent years, EVs are very popular because it leads the conventional road transportation. The prevalent model everyone knows is Tesla and Nissan. The company Tesla developed and launched different electrical vehicles. For example, it is ranging from initial Model Roadster to recent Model S. This contemporary model gives the range of 370 km in standard level and long-distance range as 507 km with Environmental Protection Agency (EPA rated). Similarly, different companies produced EVs like BMW i3, Chevy's. But these are Plug-in Hybrid Electric Vehicle (PHEV).

TYPES OF EV CHARGER

The EV chargers are classified into two types based on the location of the charger: on-board and off-board charger as shown in Figure 2 (a) [9]. The EV can be charged in three ways: 1) Conductive coupled charging, 2) wireless charging and 3) battery swapping [9, 10].

The conductive coupled charging is simple, which has a conductive cable between the charging port and EV. Here, an electrical outlet with the plug-type connector is used to charge the EVs. The wireless charging is done by inductive and capacitive coupling.

PROPOSED GLOBAL OPTIMIZATION SCHEDULING

In this scheme, set of EVs are denoted by 'M' is considered which consist of two types of EVs. One is charging only set MCHG which includes EVs which only charge their battery. Other types of EVs are V2G, MV2G EVs which include EVs that perform charging as well as discharging. The total number of intervals is 'N' and the length of the interval is 'I'. Let x_{mi} is the charging power of an EV m at interval i . For the charging power greater than zero, EV charges its battery [11]. If the charging power is less than zero, EV discharging its battery. The EVs in the charging only set have power greater than or equal to zero while the EVs in V2G set may have positive, zero or negative power in interval i as they have bidirectional energy flow between power and grid. The charging period (T_m) must be fall between the arrival time and departure time of the EVs [12] (Figure 3).

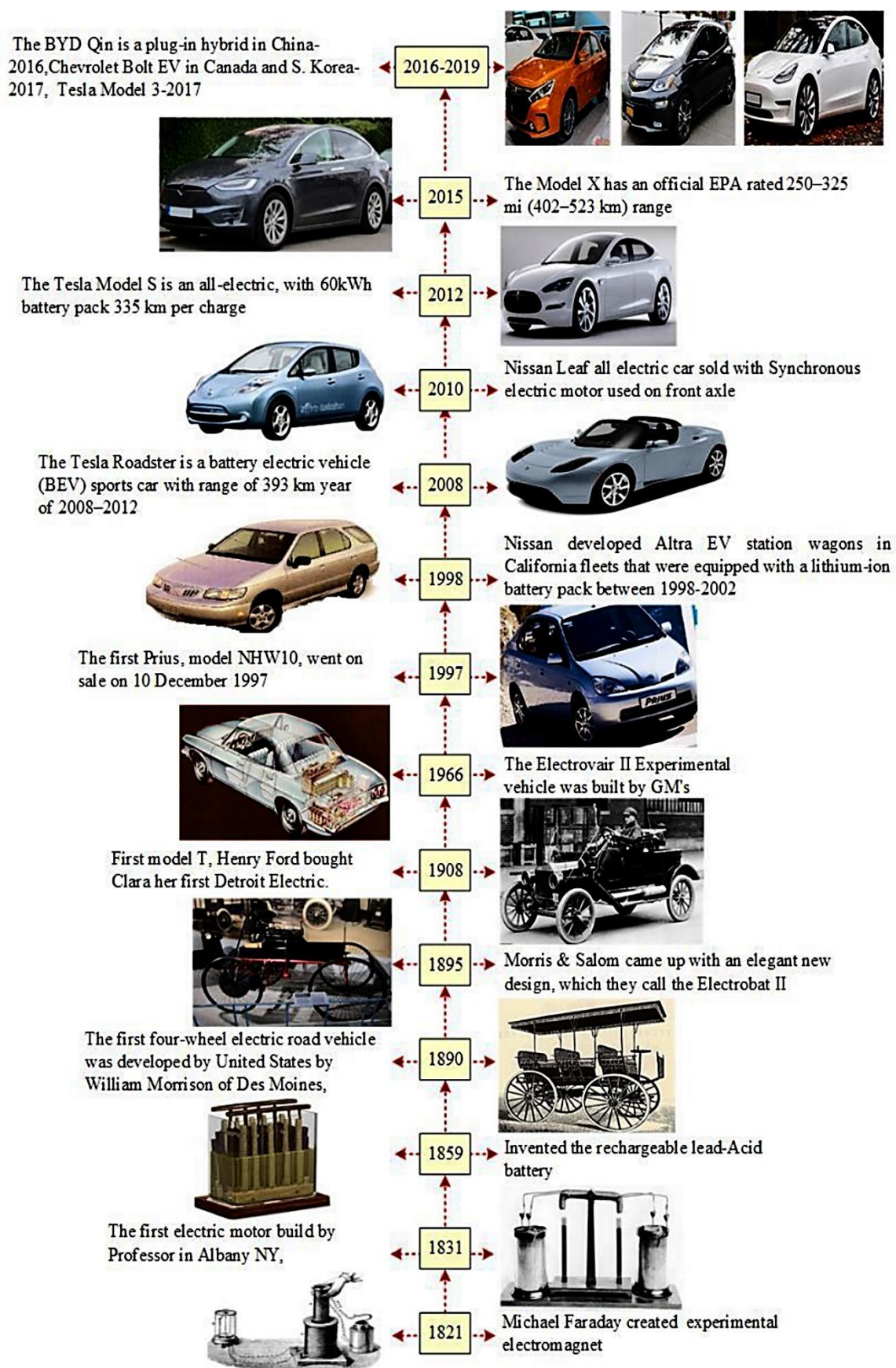


Figure 1. History of Electric Vehicles [1].

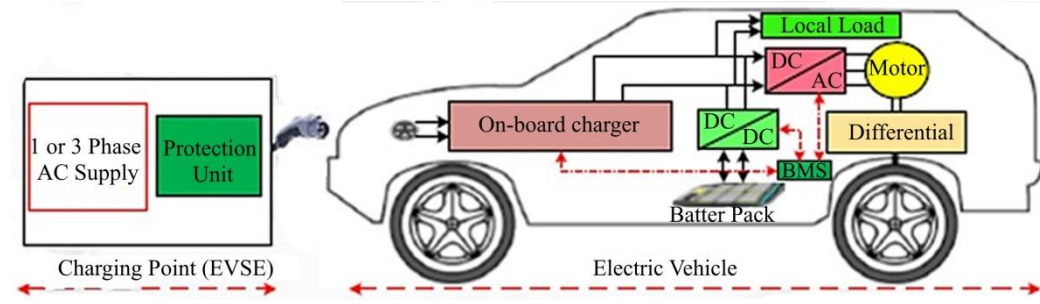


Figure 2. (a) On-board charger [9].

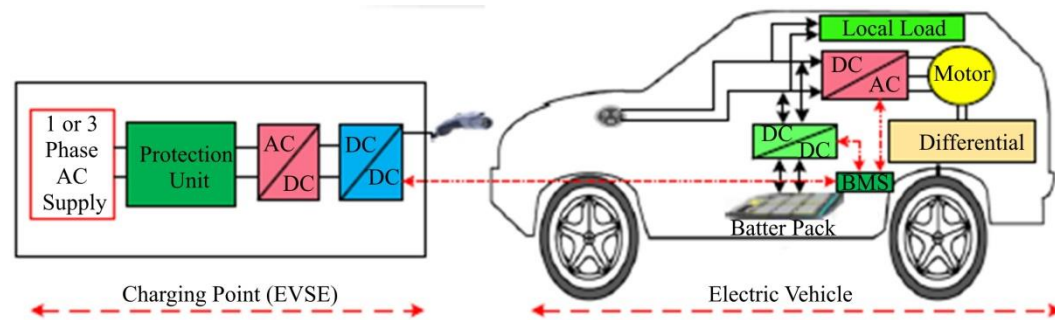


Figure 2. (b) Off-board Charger [10].

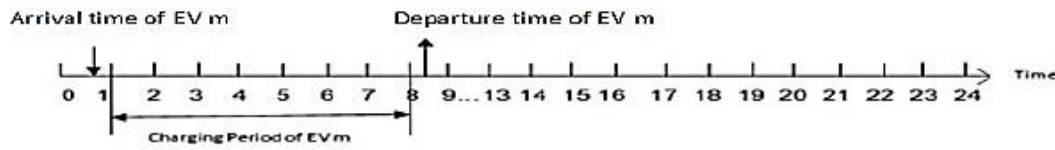


Figure 3. Charging period of EV m.

The battery capacity of EV m is denoted by E_m^{cap} . And the final energy of EV m is the energy of EV at departure time. And the final energy should be less than battery capacity. In the real pricing model, two assumptions have been made [13]. The first assumption is that the losses between nodes are small and they can be neglected and the second assumption is that there is no congestion in transmission. By these assumptions variation of electricity price can be neglected and the electricity price at a time instant is the same regardless of the charging location. The optimizations of EV charging based on only temporal variation but not spatial variation of the price have been seen in [14, 15]. The electricity price can be taken as a linear function of the instant load [15], which is given as follows:

$$g(Z_t) = k_0 + k_1 Z_t$$

Where k_0 is intercept and k_1 is slope and Z_t is the total load at time t . The total load consists of two parts; base load and charging load. The load of all electricity consumptions in an interval except EV charging is considered as a base load [16]. The charging load represents the load of EV charging. Base load is assumed to be constant in an interval. If the load from the grid to the batteries of EVs is greater than the batteries of the EVs to the grid in that interval, the charging load is positive. Otherwise, the charging load is negative. The total load in interval is given by the sum of base load and charging load [17]. Since both the base load and the charging power remain constant in interval; the total load is constant in interval. Based on the pricing model, the charging cost in interval i is given by the charging cost can be positive or negative, if the charging load is positive, charging cost is positive or vice versa.

$$\begin{aligned} C_i &= \int_{L_i^b}^{Z_i} (k_0 + k_1 Z_t) \\ &= (k_0 + k_1/2 Z_i^2) - (k_0 L_i^b + k_1/2 (L_i^b)^2) \end{aligned}$$

To find a globally optimal scheduling scheme for the EVs that perform charging and discharging during the day, it is assumed that arrival time and the departure time of each EV in the EV set are known, the initial energy and the final energy of the battery for each EV in the EV set are known and the base load in each interval of the day is known [18, 19]. A central controller is installed which collects all the information of load and EVs and then performs the scheduling optimization. The total cost is the sum of the charging costs of all the EVs over the interval set N. The total cost is then given by [20, 21].

$$C_1 = (k_0 + k_1/2z_i^2) - (k_0L_i^b + k_1/2(L_i^b)^2)$$

The main objective of the scheme is to minimize the total cost. This objective function is subjected to many constraints which are following:

$$\text{Objective : Minimize } (k_0 + k_1/2z_i^2) - (k_0L_i^b + k_1/2(L_i^b)^2)$$

Subjected to

$$\begin{aligned} z_i &= L_i^b + \sum_{m \in M} x_{mi} f_{mi}, \forall i \in N \\ 0 &\leq E_m^{ini} + \sum_{k \in i} x_{mk} f_{mk} l \leq E_m^{cap}, \forall m \in M, \forall i \in N \\ E_m^{ini} + \sum_{i \in N} x_{mi} f_{mi} l &\geq \mu_m E_m^{cap}, \forall m \in M \\ 0 &\leq x_{mi} \leq P_{max}, \forall i \in N, \forall m \in M^{CHG} \\ -P_{max} &\leq x_{mi} \leq P_{max}, \forall i \in N, \forall m \in M^{V2G} \end{aligned}$$

Here, L_i^b is base load and f_{mi} is charging interval matrix and x_{mi} is charging power at interval i of EV m . E_m^{ini} is power of EV at arrival time and E_m^{cap} is battery capacity of EV m . μ_m is the final energy ratio. P_{max} is the maximum charging power of EV Figure 4.

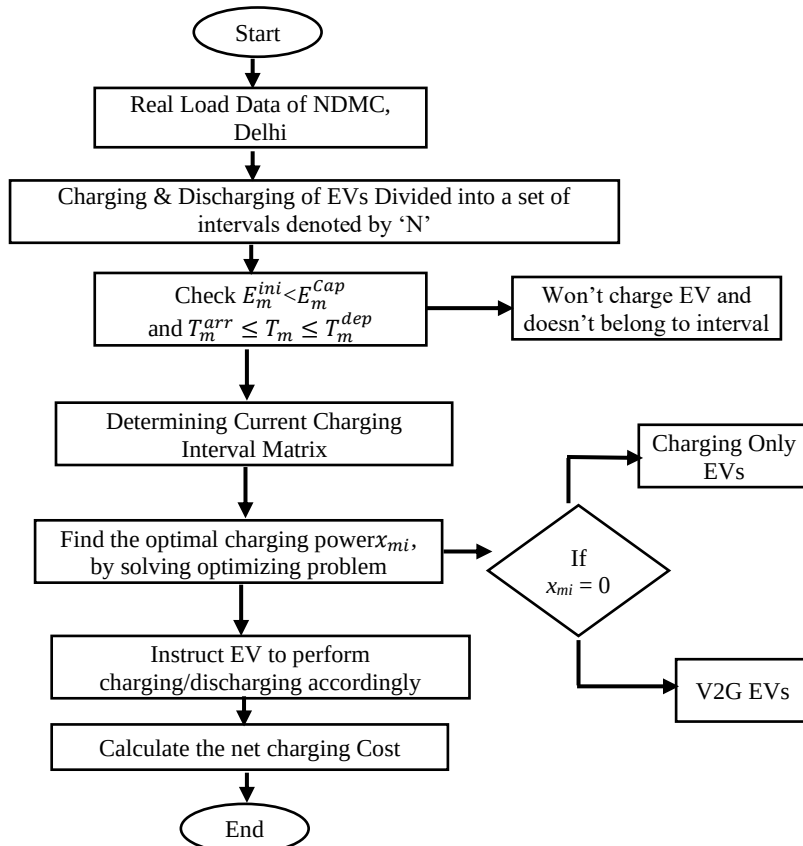


Figure 4. Flow chart for globally optimization scheme.

ESULT ANALYSIS

The real base load is used in proposed globally optimal-scheduling-scheme. This scheme is not practical as real base loads. In respect of this same local optimal scheduling is used for practical as forecasted base load is used.

The comparative analysis forecasted & real base load is represented by Figure 5. The scaling factor of real base load is 1/1500.00 & the NDMC, Delhi real base load data collected on July 2021.

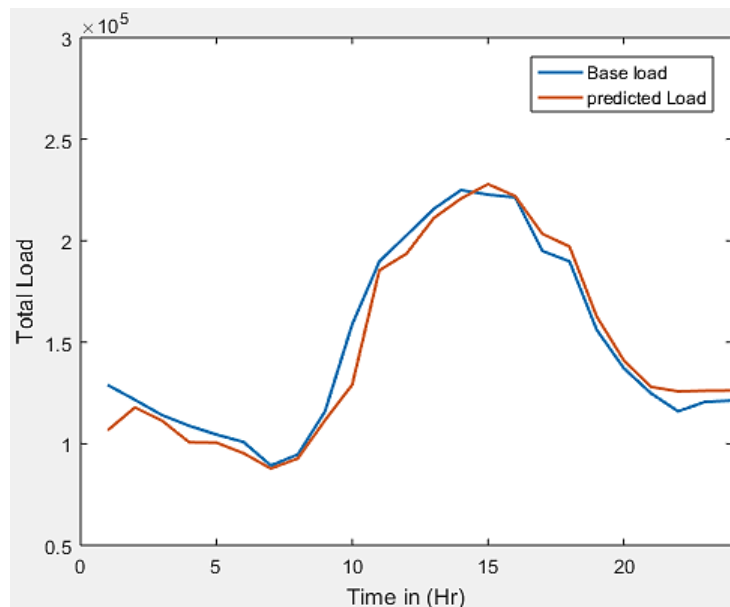


Figure 5. Comparison between base load and forecasted load.

Figure 6 represents the variation of charging load w.r.t time. This graph represent the different EVs are charged during peak-off hours using locally scheduling scheme & globally optimal scheduling scheme and whenever the demand is low, all EVs are discharge during peak hours.

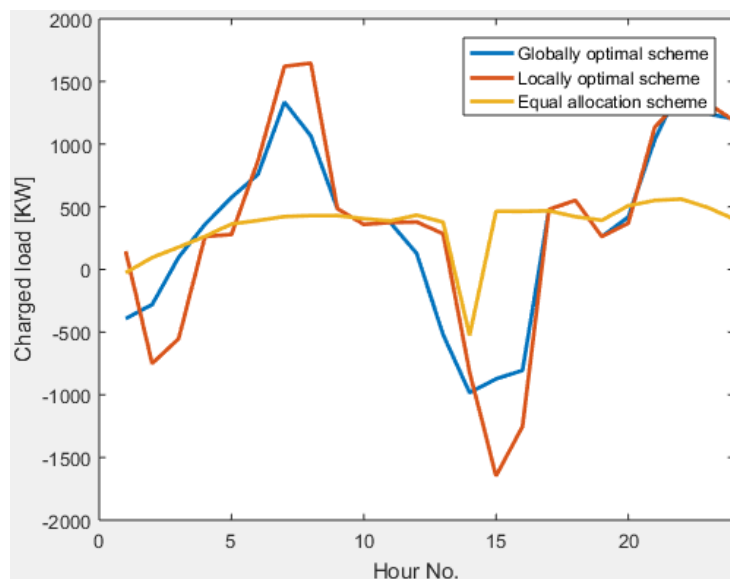


Figure 6. Variation of charging load with time.

The globally optimal scheduling scheme flattens the load profile in interval 1–7, 11–17 and 20–24 to minimize the total cost as shown in Figure 7.

The charging power of randomly chosen EV is shown in Figure 8. The charging period of EV is from 15 to 24. At interval 15, the equal allocation scheme discharge the battery without scheduling based on price of previous day. And then charging interval is from 16–24.

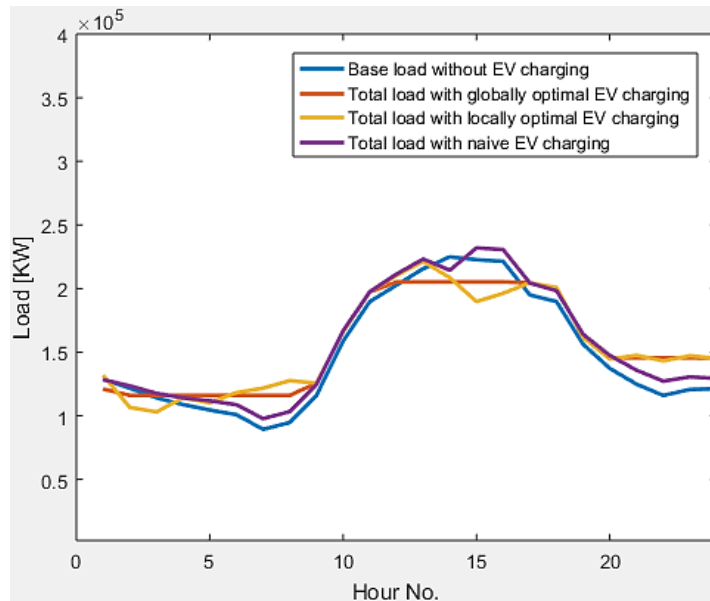


Figure 7. Variation of total load with time.

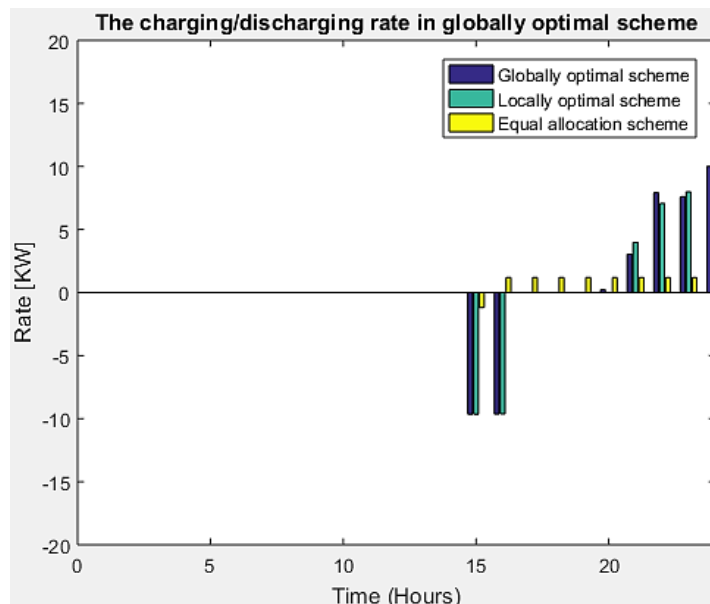


Figure 8. Variation of charging power with time.

The globally optimal scheduling scheme and locally optimal scheduling scheme calculate the charging power and discharging power by solving optimization problems. All the three schemes enable EV to reach the same final energy as shown in Figure 9.

The number of EVs is set to 200. All the EVs are V2G; can perform charging as well as discharging. The total cost of three schemes is calculated. The total cost in the globally optimal scheduling, locally optimal scheduling and equal allocation schemes are ₹14506, ₹15043, and ₹20896 respectively. The globally optimal scheme and locally optimal scheduling reduce the cost compared to equal allocation scheme (Table 1).

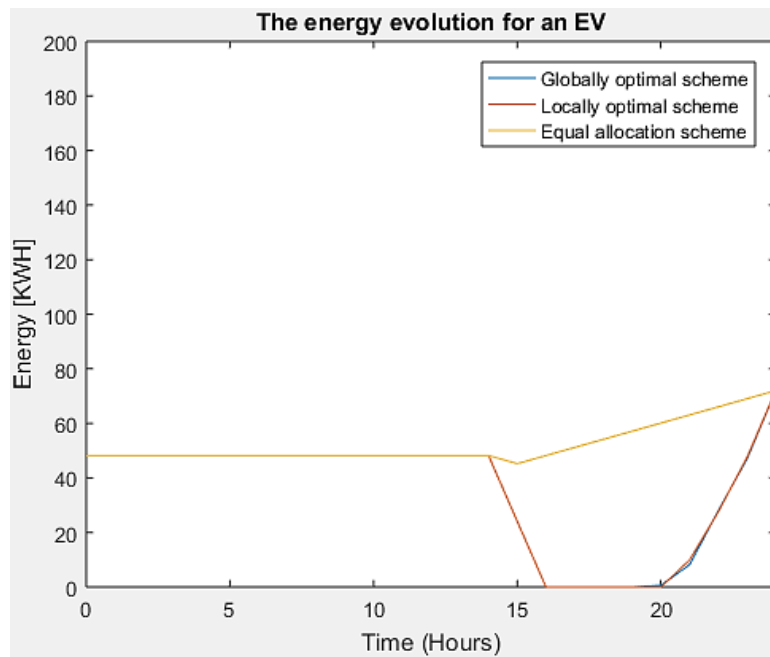


Figure 9. Variation of Energy of EV with time.

Table 1. Charging cost comparison of different schemes.

Scheme Implemented	Total Cost
Globally Optimization scheduling Scheme	₹ 14506
Locally optimization scheduling scheme	₹ 15043
Equal allocation scheme	₹ 20896

Each EV will itself decide whether it is willing to discharge battery to the grid or not before starting charging. EVs are classified into either charging only EVs or V2G EVs.

Charging ratio is defined as the ratio between the number of EVs in the charging only set and the total number of EVs. Figure 10 shows the impact of charging only ratio on total cost.

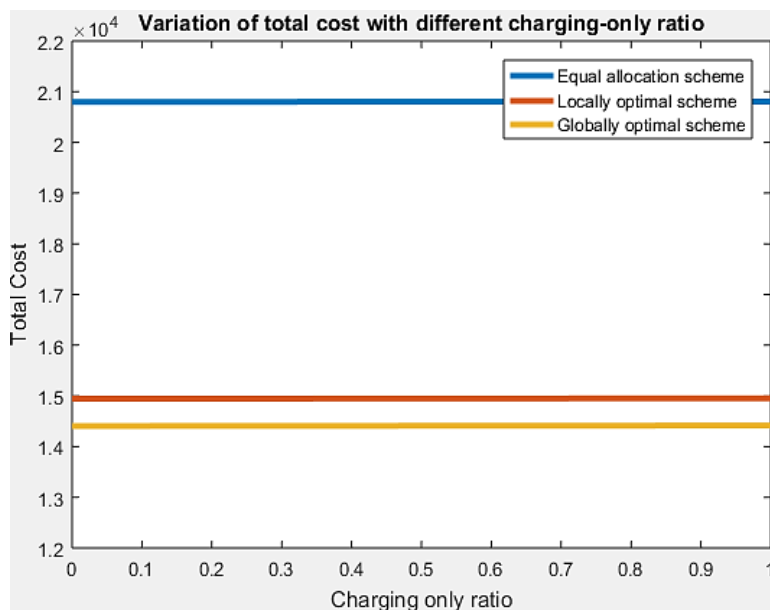


Figure 10. Variation of total cost with different charging ratio.

CONCLUSION

In this paper, we study the scheduling optimization problem for EV charging and discharging. It is important to schedule charging and discharging of EV battery as unscheduled charging may increase overall base load. The problem is formulated as a linear programming problem with a minimizing convex objective function of cost subjected to various linear constraints which have already discussed. Therefore optimization problem is convex optimization problem. The optimal charging/discharging power and minimum cost can be obtained by two schemes i.e. globally scheduling and locally scheduling optimization schemes. The result of these two schemes are compared along with equal allocation scheme and it is seen that solution of globally optimal scheduling scheme provides the minimum charging cost but this scheme is not practical as it require future details such as base load and EVs charging load which cannot determine earlier so locally scheduling scheme is preferred as the performance to this scheme is very close to globally optimal scheduling scheme.

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