

Online Optimal Variable Charge-Rate Coordination of Plug-In Electric Vehicles to Maximize Customer Satisfaction and Improve Grid Performance

Somayeh Hajforoosh, Mohammad A.S. Masoum and Syed M. Islam

Department of Electrical and Computer Engineering, Curtin University of Technology, Perth, WA, Australia.

Somayeh.hajforoosh@postgrad.curtin.edu.au, m.masoum@curtin.edu.au and s.islam@curtin.edu.au

Abstract- Participation of plug-in electric vehicles (PEVs) is expected to grow in emerging smart grids. A strategy to overcome potential grid overloading caused by large penetrations of PEVs is to optimize their battery charge-rates to fully explore grid capacity and maximize the customer satisfaction for all PEV owners. This paper proposes an online dynamically optimized algorithm for optimal variable charge-rate scheduling of PEVs based on coordinated aggregated particle swarm optimization (CAPSO). The online algorithm is updated at regular intervals of $\Delta t=5\text{min}$ to maximize the customers' satisfactions for all PEV owners based on their requested plug-out times, requested battery state of charges (SOC_{Req}) and willingness to pay the higher charging energy prices. The algorithm also ensures that the distribution transformer is not overloaded while grid losses and node voltage deviations are minimized. Simulation results for uncoordinated PEV charging as well as CAPSO with fixed charge-rate coordination (FCC) and variable charge-rate coordination (VCC) strategies are compared for a 449-node network with different levels of PEV penetrations. The key contributions are optimal VCC of PEVs considering battery modeling, chargers' efficiencies and customer satisfaction based on requested plug-out times, driving pattern, desired final SOCs and their interest to pay for energy at a higher rate.

Index Terms- Electric vehicle charging coordination, customer satisfaction, variable charging, coordinated aggregated PSO, smart grid.

NOMENCLATURE

Index:

i, j	Counters
m	Node number
n	Total number of nodes

Parameters:

$SOC_{initial}(i)$	State of charge of the i^{th} PEV at plug-in time (%)
$SOC_{Req}(i)$	Requested SOC of the i^{th} PEV (%)
$N_{PEV}(\Delta t_k)$	Number of available PEVs for current time slot
$T_{Req}(i)$	Plug-out time of the i^{th} PEV (hour)
$Bid(\Delta t_k, i)$	The price that the i^{th} PEV owner is willing to pay at current time slot (\$/kWh)
$Bid_{Max}(\Delta t_k)$	Maximum offered bid by all existing PEVs at current time slot (\$/kWh)
$R_{m,m+1}$	Resistance of the line segment between nodes m and $m+1$ (ohm)
$Y_{m,m+1}$	Admittance of the line segment between nodes m and $m+1$ (ohm)
$V_{oc,i}$	Open circuit voltage for i^{th} node (V)
R_i	Battery equivalent internal resistance for the i^{th} node (ohm)
I_i^{Rated}	Rated charger current for the i^{th} PEV (A)
Q_i	Rated battery ampere hour for the i^{th} PEV (Ah)
CR_i^{max}	Maximum charging rate for the i^{th} PEV (A)
$\eta_{ch}(CR_i^{best}(\Delta t_k))$	Charger efficiency for the i^{th} PEV at the best charge-rate (%)
V_{min} and V_{max}	Lower and upper node voltage limits (per unit; p.u.)
$D_{max}(\Delta t_k)$	Maximum demand level that would normally occur without any PEVs during a day where selected to be 0.84 MW corresponding to the maximum load for the selected DLC (MW)
C	Ratio of charging or discharging current in A to the capacity of battery in Ah
L_j	Trip path for j^{th} PEV (km)
L_i^{max}	Rated length path that each type of PEVs can trip (km)
α_D, α_{V1} and α_{V2}	Coefficients used to adjust the slopes of the penalty functions
k_1, k_2, k_3	Coefficients used to adjust the objective function based on the priority

Variables:

$FV(\Delta t_k)$	Penalty function for node voltage at current time slot
$FD(\Delta t_k)$	Penalty function for demand (distribution transformer loading) at current time slot
$SOC(\Delta t_k, i)$	State of charge of the i^{th} PEV at k^{th} time slot (%)
$T_{Remain}(\Delta t_k, i)$	The remaining available time for charging the i^{th} PEV at current time slot (hour)
$C_S(\Delta t_k, i)$	Customer satisfaction level for current time slot at i^{th} node (%)
$C_S(\Delta t_{k+1}, i)$	Customer satisfaction level for next time slot at i^{th} node (%)
V_i	Terminal voltage for i^{th} node (per unit; p.u.)
$I(\Delta t_k, i)$	Charging current for the i^{th} PEV at current time slot (A)
$CR_i^{best}(\Delta t_k)$	Optimized charging rate for the i^{th} PEV at current time slot (A)
$Dt(\Delta t_k)$	Total load at current time slot (MW)
$DL(\Delta t_k)$	Daily load at current time slot (MW)
$PLoad(\Delta t_k)$	Base-load power at current time slot (MW)
$P_{PEV,i}(\Delta t_k)$	Consumed power for the i^{th} PEV (KW)

1. INTRODUCTION

High-Tech developments in the automotive technology, growing environmental concerns in oil prices have triggered the advent of plug in electrical vehicles (PEVs). However, large fleets of PEV charging will require additional electric power demand that may lead to undesirable peaks in power consumption, transformer overloading, and interruptions. A potential solution is using online and/or offline PEV charging coordination strategies [1-4]. Ref. [3] proposes real-time PEV coordinated charging in residential distribution systems to reduce costs of power generation and losses. Ref. [4] presents real-time PEV charging/discharging coordination without considering customer preferences and variable charge-rates.

Ref. [5] proposes an online auction protocol such that vehicle owners use agents to bid for the charging opportunities. However, all PEVs have the same fixed charge-rate which is not usually the case in practical applications as vehicles have different battery and charger types, and ratings. Ref. [6] presents online coordination of PEV charging and discharging in a small geographic area based on the unrealistic assumption that no PEVs will arrive when a charging schedule is made. Ref. [7] analyzes the performance of optimal PEV charging coordination including customer satisfaction without considering variable charge-rates. Refs. [8-9] focus on maximizing aggregator revenue without carefully addressing customers' preferences and may not necessarily lead to maximum benefit for customers. Alonso et al. [10] designed the PEV scheduling to fill the valleys of the residential load profile during periods of lower load demands to avoid vehicle charging during peak load hours using a genetic algorithm. In addition, Nguyen and Le [11] presented an optimization problem that aims to minimize the total cost of energy of each PEV user. This work considers time-varying electricity prices and performs daily scheduling. Also, a real-time scheduling method of PEV charging loads is proposed in [12] to increase voltage security margin in a low-voltage distribution system. A strategy is proposed in [13] to mitigate the adverse impacts that uncontrolled charging of the PEVs impose on the host power system. However, [10-13] don't include variable charging rates and ignore battery and charger efficiencies. Ref. [14] assumes that electric vehicles drivers are insensitive to charging costs and discharging benefits. In addition, in [15], the PEV charging and wind power scheduling were integrated.

In [26], a cost minimizing strategy benefiting is proposed, but does not consider fairness in charging for all PEVs. A real-time charging coordination of PEVs based on hybrid fuzzy discrete particle swarm optimization (PSO) was presented in [27]. In addition, in [28] a multi-agent system that coordinates EV charging in distribution networks has been proposed using a distributed control method. A multi-objective scheduling strategy is formulated to charge a number of PEVs while a fuzzy solution is proposed to achieve the best compromise between the two objective functions in [29]. Moreover, Ref [30] used the population-based metaheuristics approach to solve the optimization problems. Another study [31] also shows that optimizing the charging schedule can reduce grid voltage drops and power losses as well as optimizing the load profiles.

In performing PEVs charging coordination considering customer satisfaction, some vehicles can submit requested plug-out times along with the associated requested state of charges (SOC_{Req}). Meeting these requirements is not a big problem when all the vehicles plug-out at their requested departure times. However, when unexpected departures of PEVs occur, the conventional schemes such as those proposed in [3-6], may not be able to provide acceptable levels of satisfaction fairness

among the users. Moreover, some vehicles may not be fully charged at the end of charging horizon. The problem can be resolved by using variable charging rates as a strategy to adapt the power drawn by the charger from the grid to the load, in order to fully exploit grid capability and provide a high degree of user satisfaction.

While the objective functions of [14-15] are optimization of aggregators' income and the cost of energy without addressing customers' satisfactions. In [16], the variable-based charging of PEVs is investigated; however, requested plug-out times and customers' preferences are not considered.

The main objective of this paper is to perform optimal PEV charging coordination to maximize all customers' satisfactions without exceeding grid constraints. This is done by *i*) allowing customers to specify their own charging demands including requested plug-out times, desired departure SOC and the higher electricity prices they are willing to pay, *ii*) developing an optimization problem where the decision variables are the charging rates updated at time slots of $\Delta t=5\text{min}$, and *iii*) solving the problem using coordinated aggregated particle swarm optimization (CAPSO). We rely on the quality and speed of the CAPSO solution for accurate and quick online PEV charging [24], [34-35]. Among the artificial intelligent based algorithms, the CAPSO is known to achieve near optimal solutions with better convergence characteristics. Simulation results for uncoordinated PEV charging, as well as CAPSO with fixed charge-rate coordination (FCC) and variable charge-rate coordination (VCC) are compared for a 449-node network. The proposed algorithm takes into consideration random plug-in times, initial SOC, requested plug-out times, requested final SOC and maximum charging rates of PEV batteries.

2. MODELING OF BATTERY AND CHARGER FOR PEVS

Coordination of PEVs in smart grid requires accurate modeling of its battery profile and charging characteristics. Many modern battery chargers are capable of achieving high efficiency values; however, their charging efficiencies indicate significant dependency on the charging rate due to the internal battery resistance [19-21]. This is particularly important in calculating the actual stored energies and SOC at different times during the charging period. Fig.1 (a) shows a sample experimental data of the average charging efficiency as a function of the charging rate [17]. In this paper, vehicle batteries are modelled in the steady state mode. The details of the selected model (designed by Idaho National Laboratory) are presented in [32-33].

Only a few studies have considered the power losses in the vehicle battery charging procedure by assuming a constant efficiency for the energy transfer from the grid to the battery [2, 9, 18]. In this study, for accurate implementation and evaluation of PEV variable charging coordination, battery model and dependency of charging efficiency on the charging rate are taken into account. In addition, the equivalent circuit of Fig.1 (b) is used to include the impact of battery internal resistance in SOC calculation of Section III.

3. PROBLEM FORMULATION

The objective of this study is focused on a scenario with multiple PEV owners that have different preferences and will schedule their charging profiles over time slots of $\Delta t=5\text{min}$ to maximize the customer satisfaction for all PEVs at the next

time slot while avoiding grid constrains. The proposed charging approach will ensure fairness in the SOC distribution at each time slot for all PEVs. Furthermore, if a PEV owner decides to leave prior to his/her initially requested departure time, the vehicle will receive a reasonable level of SOC. This is an improvement compared to the fix charging based methods [2-7] where PEVs may not receive any charging services if they are plugged-out before the designated times. Therefore, the comprehensive nonlinear objective function of Eq.1 is defined to maximize the total customer satisfaction by optimizing the PEVs' charging rates at each time slot:

$$\text{Max} \left(F(t) = \sum_{i=1}^{N_{PEV}} w_i(\Delta t_k) (C_s(\Delta t_{k+1}, i) - C_s(\Delta t_k, i)) \right) \quad (1)$$

for $i = 1, \dots, N_{PEV}(\Delta t_k)$, $t_k = \Delta t_k, 2\Delta t_k, \dots, 24 \text{ hours}$

$$\text{where } w_i(\Delta t_k) = k_1 \left[1 - \frac{SOC(\Delta t_k, i)}{SOC_{Req}(i)} \right] + k_2 \left[1 - \frac{T_{Remain}(\Delta t_k, i)}{T_{Req}(i)} \right] + k_3 \frac{Bid(\Delta t_k, i)}{Bid_{Max}(\Delta t_k)} \quad (2)$$

$$C_s(\Delta t_k, i) = \frac{SOC(\Delta t_k, i) - SOC_{initial}(i)}{SOC_{Req}(i) - SOC_{initial}(i)} \quad (3)$$

In Eq.2, $w_i(\Delta t_k)$ is the weighting factor that includes the customers' preferences and their enthusiasm to pay higher energy prices at each time slot. For example, if there is a vehicle with lower initial SOC and less remaining charging time, but the PEV owner prefers to pay a price higher than others, then the PEV will receive more power at that time slot.

To calculate SOC for the next time slot $SOC(\Delta t_{k+1}, i)$ there are different techniques in [21,37-40], this paper adopts the battery equivalent circuit model of [21] consisting of a constant voltage source in series with a constant resistance as shown in Fig.1b. This model is represented as:

$$V_i(\Delta t_k) = V_{oc,i} + R_i I(\Delta t_k, i) \quad (4)$$

$$I(\Delta t_k, i) = CR_i^{best}(\Delta t_k, i) \times I^{Rated}(i) \quad (5)$$

The $SOC(\Delta t_{k+1}, i)$ can be formulated based on the charging current as follows:

$$SOC(\Delta t_{k+1}, i) = SOC(\Delta t_k, i) + \left(\frac{\Delta t}{Q_i} I(\Delta t_k, i) \times \Upsilon \right) \times 100 \quad (6)$$

where, Υ is the status of each PEV where digits "1" and "0" correspond to the PEV being connected or not connected. The power delivered to PEV during the charging process is:

$$\begin{aligned} P_{PEV}^{Delivered}(\Delta t_k, i) &= V_{oc}(i) \times I(\Delta t_k, i) + R_i I^2(\Delta t_k, i) \\ &= V_{oc}(i) \times CR_i^{best}(\Delta t_k, i) \times I^{Rated}(i) + R_i (CR_i^{best}(\Delta t_k, i) \times I^{Rated}(i))^2 \end{aligned} \quad (7A)$$

and the power consumed by the i^{th} PEV from grid considering the impact of charger's efficiency is:

$$P_{PEV}^{Consumed}(\Delta t_k, i) = P_{PEV}^{Delivered}(\Delta t_k, i) / \eta_{ch}(CR_i(\Delta t_k)) \quad (7B)$$

$$P_{PEV}^{Consumed}(\Delta t_k, i) \times \eta_{ch}(CR_i(\Delta t_k)) = V_{oc}(i) \times CR_i^{best}(\Delta t_k, i) \times I^{Rated}(i) + R_i (CR_i^{best}(\Delta t_k, i) \times I^{Rated}(i))^2 \quad (7C)$$

The charging current can be calculated from (7A) as follows:

$$I(\Delta t_k, i) = \frac{\sqrt{4 \times \Delta t \times R_i \times P_{PEV}^{Consumed}(\Delta t_k, i) \times \eta_{ch}(CR_i^{best}(\Delta t_k)) + V_{oc,i}^2 - V_{oc,i}}}{2R_i \times CR_i^{best}(\Delta t_k, i)} \quad (8)$$

Substituting Eq.8 into Eq.6 yields:

$$SOC(\Delta t_{k+1}, i) = SOC(\Delta t_k, i) + \left(\frac{\Delta t \times \sqrt{4 \times \Delta t \times R_i \times P_{PEV}^{Consumed}(\Delta t_k, i) \times \eta_{ch}(CR_i^{best}(\Delta t_k)) + V_{oc,i}^2 - V_{oc,i}}}{2R_i \times CR_i^{best}(\Delta t_k, i) \times Qi} \times \Upsilon \right) \times 100 \quad (9)$$

A numerical example for the calculation of SOC (based on Eqs. 6-9) is provided in the Appendix.

3.1 ASSUMPTIONS AND DEFINITIONS

- PEVs can be connected/disconnected at any time according to the customer's needs. Customers will input their requested plug-out times and requested final SOCs at the time of plug-in. They are willing to pay higher fees compared with the short term market energy price (MEP, Fig.2(a)) for their requested special charging arrangements.
- Each hour is divided into 12 time slots of $\Delta t=5$ minutes.
- The aggregator is assumed to know the available charging power during each time slot. Each PEV can be charged after plug-in with a variable charge-rate at each time slot, and expects to reach a desired SOC_{Req} by requested plug-out time.
- The aggregator has access to PEV information using smart metering technology including their locations, charger types, battery sizes, and plug-in time.
- At each time slot, the status of each PEV will be updated. This is not a given parameter and each PEV will send a plug-in signal when it's being randomly connected to the grid.
- Fig.2(b) shows the spectrums of the random plug-in times and requested plug-out times of the PEVs.
- A PEV-Queue Table will be generated to keep track of vehicles' status including their plugged-in times; requested and actual plugged-out times; initial SOCs; requested and actual SOCs; charger type and battery sizes. As a result, after plugging a new PEV at Δt_k , the Table will be updated and the implemented CAPSO coordination algorithm will be executed to obtain a new optimal online charging schedule.
- PEV chargers are controllable and have variable charging functions. During the charging process, each PEV is assumed as a variable active load.
- The requested time $T_{Req}(i)$ for each PEV must be greater than the minimum charging time $T_{min}(i)$ required to charge the battery which depends on the maximum allowed charge-rate.

$$\text{where, } T_{min}(i) = \frac{SOC_{Req}(i) - SOC_{initial}(i)}{Charger_{Size}(i)} \quad (10)$$

- The proposed coordination process is updated when a new vehicle is plugged-in or an existing one plugs out, or a time slot has passed periodically.
- In order to make the system more robust and improve customer satisfaction, PEVs are allowed to be disconnected before their requested plug-out times. This will considerably complicate the coordination algorithm.

3.2 Constraints

In this paper the objective function (Eq.1) is subjected to the following constraints at each time slot to preserve power quality of the grid while supplying base and PEV loads:

$$V_{\min} \leq V_j(\Delta t_k) \leq V_{\max}, \text{ for } j = 1, \dots, N_{node} \quad (11)$$

$$D_i(\Delta t_k) = \sum_{k=1}^n P_k(\Delta t_k) = \sum_{k=1}^n (P_{load_k}(\Delta t_k) + P_{PEV,i}(\Delta t_k)) \leq D_{\max}(\Delta t_k) \quad (12)$$

$$D_{\max}(\Delta t_k) = \text{Max}\{DL(\Delta t_1), DL(\Delta t_2), \dots, DL(\Delta t_k)\}, \kappa = 1, \dots, 288 \quad (13)$$

To sustain battery health, its SOC level should be kept within a certain range recommended by the manufacturer. Therefore, the following SOC constraint is included:

$$SOC_{initial}(i) \leq SOC(\Delta t_k, i) \leq SOC_{Req}(i), \text{ for } i = 1, \dots, N_{PEV}(\Delta t_k) \quad (14)$$

Once $SOC(\Delta t_k, i)$ reaches $SOC_{Req}(i)$, the i^{th} battery charger will be switched to a standby mode.

The charge and discharge rates are often represented as C or C -rate, which is a measure of the rate at which a battery is charged or discharged relative to the total capacity of the battery. The C -rate is given by the numerical value of the ratio of the charging or discharging current in A to the total capacity of the battery in Ah. In this paper, the variable charging rates are considered to be from $0C$ to $CR_i^{\max} = 1C$ while 10% of vehicles are assumed to have fast charging facilities with charge rates of up to $CR_i^{\max} = 2C$. A random generator is used to generate the charge rates. The PEV charge-rates will be limited as follows:

$$0 \leq CR_i \leq CR_i^{\max} \quad CR_i \in \mathfrak{R} \quad i = 1, \dots, N_{PEV}(\Delta t_k) \quad (15)$$

3.3 Simulated input data

The continuous uniform random number generator is used to simulate the random plug-in times and expected plug-out times (Fig. 2(b)), as well as the requested SOCs. The initial state of charge $SOC_{initial}$ (%) for each PEV is calculated based on its trip length, as follows (Eq.16) [22]:

$$SOC_{initial} = \begin{cases} \alpha_i - (\alpha_i - \beta_i) \times \frac{L_j}{L_i^{\max}} & \text{for } L_j \leq L_i^{\max} \\ \beta_i & \text{otherwise} \end{cases} \quad (16)$$

for $i = 1, \dots, N_{\text{type of PEVs}}, j = 1, \dots, N_{PEV}(\Delta t_k)$

where i indicates the type of PEVs, j is the number of PEVs. In addition, three types of PEVs including Volkswagen e-golf (Type 1), Honda Fit (Type 2) and Ford C-Max (Type 3) with chargers' rates of 7.2, 6.6, and 3.3 kW that correspond to battery sizes of 24, 20, and 7.6 kWh are considered [23]. The selected values for parameters α_1, α_2 and α_3 , are 0.85, 0.8 and 0.75; β_1, β_2 and β_3 are 0.15, 0.2 and 0.25; and L_1, L_2 and L_3 are 40, 50 and 60 miles, respectively.

4. PROPOSED ONLINE HEURISTIC BASED COORDINATION ALGORITHM FOR PEV CHARGING

PSO algorithms differ in the way the swarm is updated in the feasible search space. The CAPSO approach has been applied to solve many steady state optimization problems related to power network [24]. In CAPSO, each particle updates its position by only considering the positions of particles with better achievements. Thus, this paper applies a coordinated aggregation-based PSO (CAPSO) algorithm to capture the best solutions for the PEV coordination problem (Eqs.1-15). The developed CAPSO algorithm is similar to the algorithm developed in [25] for economic load dispatching with the main difference of a proposed updating approach for the velocity vector.

4.1 Proposed Initial Population and Structure of Particles:

The selected particles for variable charge-rate online PEV coordination contain the charging rates $CR_i(\Delta t_k)$ for each PEV at Δt_k that are limited between 0 and $2C$, as follows (Eq.17):

$$\text{Number of Population} \left\{ \begin{array}{c} \overbrace{\left(\begin{array}{cccc} CR_1(\Delta t_k) & CR_2(\Delta t_k) & \dots & CR_{N_{PEV}}(\Delta t_k) \\ CR_1(\Delta t_k) & CR_2(\Delta t_k) & \dots & CR_{N_{PEV}}(\Delta t_k) \\ \vdots & \vdots & \vdots & \vdots \\ CR_1(\Delta t_k) & CR_2(\Delta t_k) & \dots & CR_{N_{PEV}}(\Delta t_k) \end{array} \right)}^{\text{Number of PEVs}} \end{array} \right. \quad (17)$$

4.2 CAPSO Fitness Function

To improve quality of CAPSO solutions fitness functions are used for the objective and constraints (Eqs.1-15). The inverse algebraic products (Eq.18) of the proposed penalty functions for voltage (Eqs.19-20) and demand (Eq.21) are used as the fitness function to combine the PEV coordination objective function (Eq.1) and constraints (Eqs.11-15):

$$F_{fitness}(\Delta t_k) = \frac{F(\Delta t_k)}{F_V(\Delta t_k) \times F_D(\Delta t_k)} \quad (18)$$

$$F_V(\Delta t_k) = \prod_{k=1}^n F_{V,k}(\Delta t_k) \quad (19)$$

$$F_{V,k}(t) = \begin{cases} e^{\alpha_{V1}(1-V_k(t))} & V_k(t) \leq V_{\min} \\ 1 & V_{\min} \leq V_k(t) \leq V_{\max} \\ e^{\alpha_{V2}(V_k(t)-1)} & V_k(t) \geq V_{\max} \end{cases} \quad (20)$$

$$F_D(\Delta t_k) = \begin{cases} 1, & D_t(\Delta t_k) \leq D_{\max}(\Delta t_k) \\ e^{\alpha_D(D_t(\Delta t_k)-D_{\max}(\Delta t_k))}, & D_t(\Delta t_k) \geq D_{\max}(\Delta t_k) \end{cases} \quad (21)$$

4.3 Simulation Results

To present the effectiveness of the proposed algorithm and the impacts of considering variable charging, simulations are performed on the SG of Fig. 2(c) for the three cases of uncoordinated and CAPSO-based coordinated PEV charging with FCC and VCC using a time interval of $\Delta t=5$ minutes; PEV penetration levels of 0% (no PEVs), 16%, 32%, 47% and 63% (Fig. 2(d)) considering $N_{bus}=449$, $N_{line}=448$, $W=0.73$, $C1=2.05$, $C2=2.05$, and $N_{pop} = 100$, $\alpha_{V1}=\alpha_{V2}=0.3$ and $\alpha_D=0.5$ (Eqs.19-

20). Simulation results are presented in Figs. 3-6 and Tables I-II. It should be noted that the network representation in Fig. 2(c) is at the low-voltage side of each distribution transformer (DT).

To compare the performance of FCC and VCC schemes, detailed simulations will also be presented for the three selected feeders in Fig. 2(c) with the best (DT-20), moderate (DT-12) and worst (DT-14) performances. The first feeder receive 100% of customer satisfaction with both FCC and VCC, the second feeder receives 100% of customer satisfaction with VCC and the third feeder does not receive 100% customer satisfaction regardless of the selected (FCC or VCC) coordination approach. It should be noted that DT-20 and DT-12 are not the only feeders that receive 100% of customer satisfaction with VCC. The complete lists of best feeders are DT-9, DT-13, DT-20 and DT-22 whereas the complete lists of the moderate feeders are DT-3, DT-4, DT-5, DT-7, DT-11, DT-12, DT-18 and DT-21.

In this paper, the backward-forward sweep method is used to calculate power (load) flows and bus voltages. It is assumed that the generation capacity is large enough to supply both the base and the PEV charging loads in all timeslots.

At each timeslot ($\Delta t=5$ minutes), the weakest bus is defined to be the bus with the lowest voltage magnitude. The locations and voltage magnitudes of the weak buses will change within the 24 hours depending on the system base load and system configuration and the PEV loadings (numbers, locations, random plug-in times and charging rates of the activated PEVs). To identify the weakest bus at each timeslot, the optimal PEV coordination is performed, the selected PEVs are activated, power flow calculation is performed, nodes are sorted based on their voltage magnitudes, and the node with lowest voltage value is selected as the weakest bus. This process is repeated for 24-hour to generate the weak voltage profiles of Figs. 3c, 4c and 5c.

Case A: Uncoordinated PEV Charging

The impact of uncoordinated PEV charging is investigated by starting the charging process as soon as vehicles are randomly plugged in. Simulation results are presented in Table I (rows 4-8) and Fig.3. As expected, the SG is facing overloading, voltage regulation and efficiency problems. For example, for 63% PEV penetration, maximum power consumption has increased by about 45% (Fig.3(b)) compared to the nominal operation with no vehicles. In addition, the minimum voltage for 63% of PEVs penetration has decreased by 30% compared to its nominal value as shown in Fig.3(c). Moreover, in this case there is about 10% voltage violation for 32% PEV penetration. Furthermore, it can be seen that the voltage drops to 0.7P p.u. (Fig.3(c)), where in reality it may cause system collapse and should be limited by system operator. To overcome problems associated with uncoordinated PEV charging, the CAPSO algorithm of Section IV is adopted.

Case B: Coordinated PEV Charging using CAPSO with FCC

In this approach, the charging process of each PEV is realized at a fixed rate, corresponding to the nominal charging rate of its charger. In details, the charging process of each user starts by receiving a charging signal from charging center and will be connected till receive its required state of charge. The implemented CAPSO algorithm is used for optimal PEV charging coordination with a fix nominal charging rate. While PEV will be automatically disconnected when reaching their requested SOC levels, the consumers can also disconnect their vehicles prior to the requested plug-out times. Simulation results for

FCC are presented in Fig.4 and Table I (rows 9-13). Compared to Case A, FCC is offering further improvements in substation transformer loading, loss power and weak bus voltages. Note that, there is still overloading of the main substation transformer.

Case C: Coordinated PEV Charging using CAPSO with VCC

Simulation results including system power consumptions and loss using CAPSO algorithm with variable charge-rate strategy are presented in Fig.5 and Table I (rows 14-18). Compared to Case B, VCC is more strictly preventing system (transformer) overloading (Fig.5(b)). It can be observed that for 63% PEV penetration, VCC has the advantage of completing the charging process of all vehicles sooner than the FCC. However for lower PEV penetration levels of 47%, 32% and 16%, the two charging strategies have similar characteristics. In addition, there is no overload in system power consumption with VCC as it is limited to the designated 0.84MW while there is about 2.72% overloading in substation transformer using FCC (Fig. 4(b)). Moreover, the voltages are in their permissible limit and there is no problem with the voltage profiles (Fig.5(c)).

5. DISCUSSION AND ANALYSIS

The SOC variations within the 24 hours are presented in Figs. 6(a-d) for the best (DT-20) and the worst (DT-14) feeders:

- According to Fig.6(c) with FCC, at the worst feeder there are 4 out of 12 PEVs that are not charged at all and their initial SOCs are not changed. Therefore, the customer satisfaction at these nodes is zero which has reduced the overall satisfaction level at this feeder (DT-14).
- It is depicted that with FCC the first and the last PEV start charging times in feeder DT-14 are at 19:00pm and 4:00am, respectively; while with VCC (Fig.6(d)) some of the PEVs are started charging as early as 17:50pm and the last vehicle is being activated at 19:20pm on node “s”.
- The VCC strategy is capable of fully or partially charging the PEVs that were not scheduled with FCC (e.g., were not allowed to start charging). For example with the FCC, the PEVs located on nodes b, d, f and h of feeder DT-14 have zero customer satisfaction (Fig.6(f)) while with the VCC their customer satisfaction rates are improved to 100%, 23.85%, 35.94% and 23.70%, respectively.
- Comparison between Figs.6(a, b) shows that all PEVs in the best feeder (DT-20) reach their requested SOCs before their requested plug-out times using VCC and all vehicles are charged by 3:40am.
- With FCC the PEV on node “l” is not activated and has not received charging service until 3:35am, while with VCC strategy the same vehicle at the same time will be fully satisfied and receives 100% of its requested SOC.

The bar charts of Fig.6 (e, f) and Table II show the amount of customer satisfaction for the best and worst feeders:

- Using FCC, the customer satisfactions for four PEVs are zero, while with VCC all vehicles are receiving full or partial charging; Table II (rows 9-10, 18-19, 27-28). Figs.6 (g, h) show the customer satisfaction profiles for all feeders (DT-1 to DT-22) using VCC and FCC strategies:

- The numbers of feeders reaching 100% customer satisfaction with VCC and FCC approaches are 12 and 4, while according to Table II (rows 9-10) the minimum levels of customer satisfactions in the worst feeder (DT-14) are 78% and 64%, respectively.

- The feeders have reached their requested SOC at the same time with VCC (Fig. 6(h)), while with FCC there is a significant time difference in obtaining the requested SOC even for the fully satisfied customers (Fig.6(g)). For instance, feeder DT-12 has reached 100% customer satisfaction at time slot 110, while feeders DT-8 and DT-9 will received 100% of customer satisfaction at time slot 130.

Table III presents a detailed comparison between the proposed PEV coordination strategy and the three recently implemented methods ([26]-[28]). Based on rows 8-10 of Table III, the main aim and contribution of this paper which is customer satisfaction with different customer preferences using variable charging rates has not been addressed in [26-27]. In addition, this paper also presents detailed work on PEV charge optimization and analysis based on the customers' requirements that includes requested SOC_s, bids and plug-out times. Furthermore, this paper is mainly focused on the variable charging strategies and uses CAPSO to achieve optimal charge rates for each time slot whereas a fix charge is considered in most of the previous studies. Finally, the impacts of charger efficiency and battery modeling have also been included in the paper which are not considered in Refs. [26] to [27].

6. CONCLUSION

This paper has implemented an optimal, fast and effective online variable charge-rate PEV coordination strategy using CAPSO to maximize the total customer satisfaction for all PEV owners. The proposed VCC approach will also minimize the grid losses without exceeding grid constraints based on costumers requested plug-out times, requested battery state of charges (SOC_{Req}) and their interests to pay for higher charging energy prices at time slots of $\Delta t=5$ min.

Detailed simulations results for a 449-node SG network are presented, compared and analyzed for uncoordinated PEV charging and coordinated PEV charging with FCC and VCC strategies. Main conclusions are:

- The proposed coordinated charging algorithm takes into consideration random plug-in times, initial SOC_s, requested plug-out times, and requested final SOC_s, as well as the maximum battery charging rates, battery and charger efficiencies.
- With VCC, customers received higher levels of satisfaction, while with FCC some vehicles may not even start charging before their requested plug-out times.
- The substation transformer is not overloaded with the VCC option, while overloading conditions are noticed with FCC even at low levels of PEV penetrations of 16% and 32%.
- The VCC strategy is capable of fully or partially charging the PEVs that were not scheduled with FCC.

- The proposed charging approach (VCC) will ensure fairness in the SOC distribution at each time slot for all PEVs. Then, if a PEV owner decides to leave prior to his/her initially requested departure time, the vehicle will receive a reasonable level of SOC.
- There is no overload in system power consumption with VCC as it is limited to the designated 0.84MW while there is about 2.72% overloading in substation transformer using FCC .

APPENDIX A: NUMERICAL EXAMPLE FOR CALCULATION OF SOC

This appendix presents a numerical example for the calculation of SOC based on equations 6 to 9. Assuming the battery bank voltage is 400V, the battery capacity is 10kWh, and the nominal cell voltage for lithium ion batteries is 3.2V then the number of cells for the whole battery bank will be 125. It is also considered that the internal resistance for each cell is 2m Ω and then the total battery bank resistance $R_b = 2\text{m}\Omega \times 125 = 250\text{m}\Omega$. In this paper, each time slot is assumed to be 5 minutes; therefore, $\Delta t = 1/12 = 0.0833$ hours. To calculate Q_i , the battery capacity (10KW) should be divided by its open circuit voltage (400V); therefore, $Q_i = 10,000/400 = 25\text{Ah}$. Detailed calculations of SOC using Q_i are presented in Table A1.

In Coulomb counting technique, the charges flowing into and out of the battery are integrated to get an accurate estimate of the remaining capacity and calculation of SOC [36]. This technique uses a shunt to measure battery current, and a coulomb counting circuit which is effectively a very accurate current-integrating ADC (analog to digital) technique. Then, the measured battery voltage and current are sent to a microprocessor where the microprocessor contains battery chemistry specific information, such as cell impedance in its memory. To communicate with the rest of the system a standard protocol such as I²C communication can be used. Then, the SOC is calculated using Eqs. 4-9 (presented in section 3 and Table A1). Then, the SOC will be transmitted through Wi-Fi system to the central PEV charging coordination.

Practically, for real-time calculation of battery SOC at each time slot ($\Delta t = 5$ minutes), the following steps should be taken (see Table A1 for more details).

- 1- The battery ampere hour rate is calculated as $Q_i = (\text{Battery Capacity})/V_{oc}$.
- 2- Measuring open circuit voltages of the PEV batteries (V_{oc}).
- 3- Measuring the battery current using a shunt.
- 4- Sent the measured battery voltage and current to the microprocessor using a standard protocol such as I²C communication.
- 5- Using Equation 7-9 to calculate SOC.
- 6- Transmitting the SOC to the central PEV charging coordination center through Wi-Fi system.

Note that if the battery is connected to a charger with CR=0.3, then the required time (without considering losses) to fully charge the battery is $1/0.3 = 3.333$ hours. If we consider the impacts of the losses, then the battery will only be charged to 86% of its rated capacity. Therefore, more time will be required to fully charge the battery if the losses are included.

APPENDIX B: DETAILS OF THE THREE SIMULATED CASES STUDIES

The selected input parameters for the simulation cases with 63% PEV penetration are provided in Table B1. This table presents the selected battery and charger types for each bus with PEV as well as the selected random values for initial SOCs, requested SOCs, plug-in times and plug-out times. The selected charger and battery types are:

- Charger Type: A = 3.3kW, B= 7.2kW and C=6.6kW
- Battery type: D=6kWh, E= 19.2kWh and F= 16kWh

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TABLE CAPTION

Table I: Impact of PEV charging on the SG of Fig. 2(c).

Table II: Detailed simulation results for coordinated (CAPSO) PEV charging of Fig. 2(c) for worst, moderate and best feeders using FCC and VCC.

Table III: Detailed comparison of the proposed CAPSO based PEV coordination approach with the implemented strategies in references 26-28.

Table A1: Calculation of SOC based on Eqs. 6-9.

Table B1: Selected input parameters for the three simulated cases studies

TABLE I

PEV [%]	ΔV^* [%]	IMAX [%]**	Customer Satisfaction	Computing time*** (Sec)
Nominal Case: With no PEV				
0	7.63	0	NA	NA
Case A: Uncoordinated PEV Charging; Fig. 3				
16	10.08	18.42	NA	NA
32	12.60	19.58	NA	NA
47	25.10	37.62	NA	NA
63	31.00	45.27	NA	NA
Online PEV Coordination (DPSO) with FCC (Ref. [27])				
16	9.32	0.82	NA	0.026
32	9.38	0.93	NA	0.028
47	9.90	0.99	NA	0.031
63	9.90	2.72	NA	0.032
Case B: Online PEV Coordination (CAPSO) with FCC; Fig.4				
16	9.42	1.39	98.37	0.027
32	9.53	2.03	95.57	0.028
47	9.91	2.09	90.54	0.032
63	9.94	2.72	88.38	0.034
Case C: Online PEV Coordination (CAPSO) with VCC; Fig.5				
16	9.06	0.00	99.09	0.028
32	9.16	0.00	96.91	0.029
47	9.58	0.00	94.47	0.034
63	9.73	0.00	93.89	0.035

*) Average voltage deviation over 24 hours.

**) Increase in transformer current compared with nominal case (no PEVs).

***) Intel Core i5-3570 3.40 GHz processor, 8 GB RAM, using MATLAB ver. 8

TABLE III

		Proposed in [26]	Proposed in [27]	Proposed in [28]	Proposed in This Paper
Method		Fuzzy	DPSO/GA Fuzzy	Multi Agent Based	CAPSO
Charge Type		FCC	FCC	FCC	VCC, FCC
Battery Type		One Type	Variety of Battery Types	One Type	Variety of Battery Types
Battery Modelling		NA	NA	✓	✓
Charger Efficiency Effect		NA	NA	NA	✓
Customer Preference	$SOC_{Req}(i)$	NA	NA	✓	✓
	$Bid(\Delta t_k, i)$	NA	NA	NA	✓
	$T_{Req}(i)$	NA	NA	NA	✓
Customer Satisfaction Analysis		NA	NA	NA	✓
Driving Pattern		NA	✓	NA	✓

TABLE A1

Input Data to Calculate SOC	
V_{oc} for the battery bank (V)	400
R_i for the battery bank (ohm)	$125 \times 0.002 = 0.25$
Charger efficiency at CR^{Best} based on Fig 2.a	0.93
CR^{Best} which is a sample of results for i^{th} PEV achieved by CAPSO	0.30
Nominal charger capacity $P^{consumed}$ (W)	10000
Δt	0.083
$SOC_{Initial}$ (%)	0
Rated battery ampere hour Q_i (Ah)	$10000 \div 400 = 25$
Calculations	
I^{Rated} (considering losses) using Eq. 8 (A)	6.450
I^{Rated} (not considering losses) using Eq. 5 (A)	7.5
Increment of SOC for the Next 5 Minutes Using Eq.9	
Increment of SOC for the next 5 minutes (considering losses) (%)	2.150
Increment of SOC for the next 5 minutes (not considering losses) (%)	2.5
SOC After the Nominal Charge Time of $1/0.3=3.333$ Hours with $CR=0.3$	
SOC after 3.333 hours (considering losses) (%)	86.0
SOC after 3.333 hours (not considering losses) (%)	100

TABLE B1

Bus #	SOC Initial	SOC Req	Battery Type	Charger Type	Plug-in Time 63	Plug-out Time	Bus #	SOC Initial	SOC Req	Battery Type	Charger Type	Plug-in Time	Plug-out Time
2	18	89	E	B	22	135	222	26	84	F	C	73	198
4	14	90	F	C	24	154	224	7	59	D	A	83	205
6	25	53	F	C	18	77	226	12	73	D	A	46	184
7	2	82	E	B	22	144	227	23	94	E	B	42	181
8	20	92	E	B	44	182	228	18	97	F	C	55	189
10	27	63	F	C	15	60	230	28	64	F	C	54	188
12	4	56	E	B	92	211	232	4	93	E	B	63	192
13	14	71	E	B	62	192	233	28	65	F	C	42	181
15	10	73	F	C	6	52	235	7	68	D	A	61	192
17	28	82	F	C	24	149	237	17	66	E	B	68	194
18	0	96	E	B	45	183	238	27	53	F	C	51	187
19	11	72	D	A	13	53	239	20	57	D	A	50	186
22	24	69	F	C	38	176	242	10	52	F	C	110	216
24	24	83	F	C	69	196	244	2	62	F	C	68	194
26	23	92	F	C	44	182	246	2	89	F	C	68	195
27	23	51	F	C	46	184	247	8	85	F	C	42	181
28	8	92	E	B	27	166	248	13	96	E	B	33	174
30	29	97	E	B	36	176	250	16	65	E	B	75	200
32	10	55	E	B	20	98	252	21	53	D	A	55	189
33	28	55	E	B	54	189	253	10	59	E	B	72	197
35	23	52	E	B	38	177	255	8	67	F	C	42	181
37	11	53	F	C	63	192	257	4	61	D	A	77	201
38	22	88	F	C	72	197	258	10	82	F	C	23	144
39	10	82	E	B	69	196	259	24	81	E	B	90	210
42	21	97	E	B	51	188	262	9	62	D	A	21	103
44	28	67	D	A	72	197	264	12	92	F	C	14	58
46	1	78	D	A	64	193	266	5	99	E	B	20	84
47	27	81	D	A	55	189	267	8	89	F	C	15	62
48	1	69	F	C	58	191	268	2	82	E	B	19	77
50	4	85	E	B	68	195	270	19	72	E	B	24	151
52	8	54	D	A	50	186	272	11	70	E	B	25	157
53	19	74	E	B	38	178	273	9	83	D	A	21	104
55	2	88	F	C	56	190	275	19	61	D	A	21	112
57	15	71	F	C	62	192	277	5	68	F	C	23	145
58	21	71	D	A	69	196	278	22	80	F	C	28	167
59	2	57	D	A	58	191	279	20	67	E	B	36	175
62	24	70	E	B	50	187	282	24	96	F	C	29	169
64	22	83	E	B	71	197	284	18	57	F	C	20	91
66	23	62	D	A	57	190	286	20	63	E	B	29	170
67	3	76	F	C	75	200	287	19	87	E	B	21	118
68	29	99	F	C	57	190	288	21	94	F	C	5	48
70	11	90	E	B	59	191	290	1	72	E	B	0	26
72	27	56	F	C	57	191	292	22	93	F	C	6	53
73	1	75	D	A	41	180	293	10	94	F	C	21	122
75	11	62	F	C	40	179	295	24	55	D	A	17	71
77	28	76	D	A	40	180	297	16	86	D	A	16	62
78	23	63	D	A	95	212	298	6	89	D	A	23	146
79	6	64	E	B	55	190	299	11	77	D	A	16	66
82	16	58	D	A	78	201	302	22	77	D	A	17	71
84	28	54	E	B	66	194	304	19	53	D	A	29	171
86	5	80	E	B	96	212	306	7	86	D	A	31	173
87	22	83	E	B	78	202	307	18	74	E	B	47	184
88	7	89	E	B	34	175	308	1	51	D	A	26	163
90	29	55	F	C	52	188	310	7	97	F	C	24	154
92	23	72	F	C	77	201	312	1	88	F	C	17	72
93	9	92	F	C	65	193	313	6	65	D	A	21	125
95	27	58	F	C	104	215	315	5	100	F	C	22	143
97	22	91	E	B	58	191	317	23	89	F	C	5	50
98	15	68	E	B	28	168	318	6	89	E	B	20	91
99	0	53	D	A	68	195	319	4	60	F	C	30	172
102	3	56	D	A	41	180	322	11	81	F	C	17	76
104	14	66	E	B	60	191	324	4	83	E	B	28	168
106	15	98	D	A	81	204	326	27	87	F	C	33	175
107	10	80	D	A	65	193	327	20	60	F	C	34	175
108	16	65	D	A	65	193	328	19	97	E	B	31	173
110	24	89	F	C	55	189	330	13	87	E	B	0	35
112	16	78	E	B	56	190	332	20	90	D	A	17	77
113	29	75	E	B	57	190	333	7	55	D	A	32	173
115	25	51	F	C	3	46	335	24	74	D	A	9	53
117	15	53	E	B	67	194	337	1	81	E	B	60	191
118	25	80	F	C	79	202	338	9	100	D	A	62	192
119	11	72	E	B	33	174	339	9	95	D	A	51	187
122	14	81	E	B	29	168	342	28	61	F	C	32	174
124	3	54	E	B	71	197	344	21	94	F	C	47	185
126	6	71	F	C	29	169	346	29	99	F	C	49	185
127	15	67	D	A	25	155	347	5	81	D	A	66	193
128	2	89	F	C	75	199	348	16	59	F	C	60	191
130	23	87	E	B	46	183	350	19	57	F	C	40	179
132	14	60	E	B	84	205	352	9	80	E	B	79	203
133	3	63	F	C	33	174	353	11	60	F	C	46	184
135	15	95	E	B	47	184	355	5	97	F	C	31	173
137	21	57	D	A	64	193	357	3	97	F	C	38	177
138	5	83	F	C	70	196	358	23	91	F	C	64	193
139	9	89	E	B	68	194	359	12	69	E	B	51	187

142	15	68	D	A	63	192	362	21	67	F	C	65	193
144	8	55	F	C	54	188	364	19	54	F	C	48	185
146	16	59	F	C	47	184	366	13	73	D	A	97	213
147	16	80	E	B	69	195	367	27	71	E	B	14	60
148	23	99	D	A	51	187	368	1	51	E	B	39	179
150	19	80	E	B	67	194	370	8	72	D	A	45	183
152	15	79	E	B	81	204	372	23	62	E	B	49	185
153	15	74	F	C	14	55	373	0	73	D	A	76	200
155	23	55	E	B	95	211	375	17	80	D	A	51	188
157	6	53	F	C	46	184	377	19	74	F	C	45	183
158	14	88	F	C	85	206	378	18	85	F	C	48	185
159	19	92	D	A	90	208	379	23	90	F	C	62	192
162	14	67	D	A	30	171	382	9	58	D	A	76	201
164	8	81	E	B	63	192	384	24	84	E	B	61	192
166	8	63	E	B	128	217	386	7	74	D	A	36	176
167	23	77	D	A	55	189	387	3	76	E	B	51	188
168	13	66	F	C	69	195	388	28	76	E	B	64	193
170	25	93	D	A	66	193	390	26	57	D	A	83	205
172	1	72	E	B	84	206	392	16	55	D	A	38	178
173	10	61	E	B	81	203	393	12	60	F	C	85	207
175	19	81	D	A	40	179	395	8	52	F	C	114	216
177	29	65	F	C	50	186	397	21	74	D	A	49	185
178	26	77	E	B	66	193	398	21	58	D	A	68	195
179	25	100	F	C	58	191	399	28	56	D	A	46	184
182	26	97	E	B	33	174	402	28	57	F	C	25	159
184	16	57	E	B	56	190	404	17	81	F	C	88	207
186	15	72	F	C	16	62	406	5	67	E	B	73	198
187	29	53	D	A	45	183	407	15	89	D	A	49	186
188	14	81	D	A	28	167	408	22	62	D	A	61	192
190	27	64	E	B	34	175	410	1	87	F	C	54	189
192	21	83	F	C	53	188	412	6	56	D	A	30	172
193	25	72	F	C	24	149	413	7	96	D	A	1	39
195	11	64	E	B	27	165	415	13	64	F	C	58	191
197	8	82	D	A	17	69	417	20	75	E	B	50	186
198	11	71	D	A	2	44	418	14	93	F	C	47	185
199	8	58	D	A	44	182	419	23	92	F	C	42	182
202	10	98	E	B	70	196	422	14	52	E	B	23	148
204	7	57	D	A	43	182	424	10	79	F	C	49	186
206	6	70	D	A	42	180	426	24	85	E	B	62	192
207	0	74	D	A	26	161	427	25	92	E	B	38	178
208	21	58	F	C	79	202	428	11	99	E	B	68	195
210	26	62	E	B	42	180	430	22	69	D	A	54	189
212	15	99	F	C	73	198	432	17	83	E	B	57	190
213	12	77	F	C	54	188	433	10	73	D	A	30	172
215	14	73	E	B	42	181	435	25	56	D	A	44	182
217	25	58	D	A	69	196	437	4	79	E	B	67	194
218	22	63	F	C	67	194	438	13	81	D	A	43	182
219	21	64	E	B	45	183	439	18	69	F	C	62	192

FIGURE CAPTION

Fig. 1. Modeling of PEV battery [17]; (a) typical charger efficiency (CR corresponds to charging efficiencies), (b) equivalent circuit [21].

Fig. 2. System characteristics; (a) daily residential load curve (DLC) and short term market energy price (MEP) [3], (b) spectrums of the random plug-in times and the requested plug-out times of the simulated PEVs, (c) the 449 node SG consisting of IEEE 31-node 23kV system and 22 low voltage 19-node 415V residential feeders populated with PEVs [3], (d) detailed diagram of one residential feeder in Fig. 2(c) with 16%, 32%, 47% and 63% PEV penetrations.

Fig. 3. Simulation results for Case A with 16, 32, 47 and 63 percent of PEV penetrations; (a) loss power, (b) system power consumption, (c) weak bus voltage.

Fig. 4. Simulation results for Case B with 16, 32, 47 and 63 percent of PEV penetrations; (a) loss power, (b) system power consumption, (c) weak bus voltage.

Fig. 5. Simulation results for Case C with 16, 32, 47 and 63 percent of PEV penetrations; (a) loss power, (b) system power consumption, (c) weak bus voltage.

Fig.6. (a) Sample battery SOC's of feeders in Fig.2(c) for best feeder (DT-20) using FCC,

Fig.6. (b) Sample battery SOC's of feeders in Fig.2(c) for best feeder (DT-20) using VCC,

Fig.6. (c) Sample battery SOC's of feeders in Fig.2(c) for worst feeder (DT-14) using FCC,

Fig.6. (d) Sample battery SOC's of feeders in Fig.2(c) for worst feeder (DT-14) using VCC,

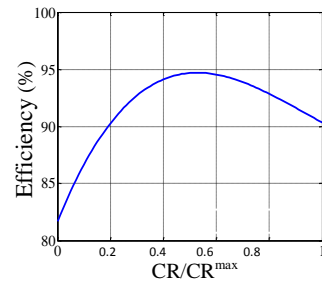
Fig.6. (e) Customer satisfaction of few feeders in Fig.2(c) using FCC and VCC for best feeder (DT-20),

Fig.6. (f) Customer satisfaction of few feeders in Fig.2(c) using FCC and VCC for worst feeder (DT-14),

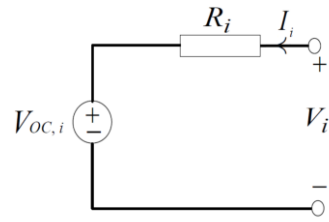
Fig.6. (g) Customer satisfaction profiles for all feeders (DT-1 to DT-22) in Fig.2(c) using FCC,

Fig.6. (h) Customer satisfaction profiles for all feeders (DT-1 to DT-22) in Fig.2(c) using VCC.

Figure 1

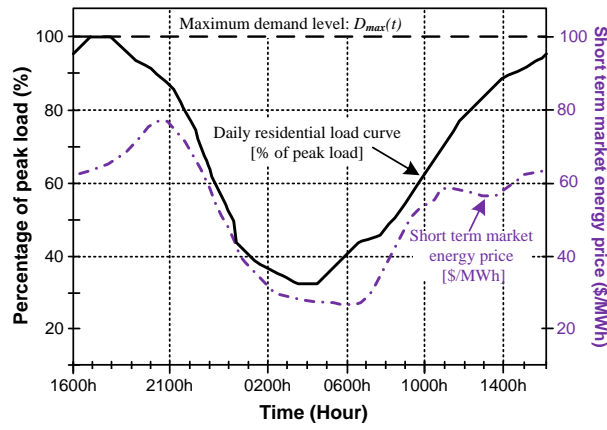


(a)

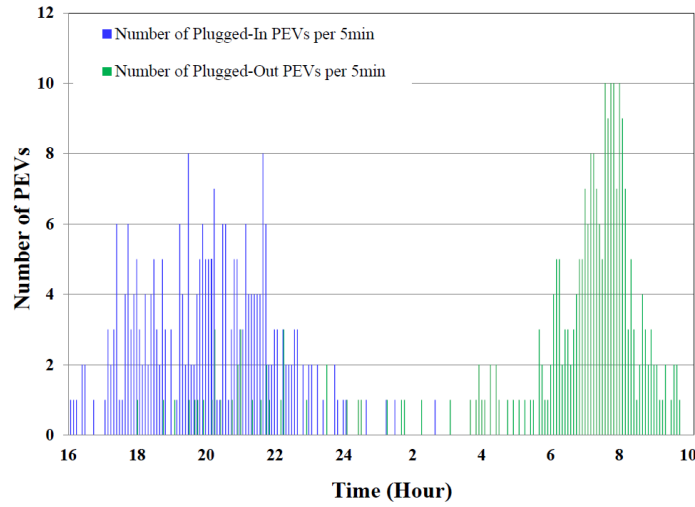


(b)

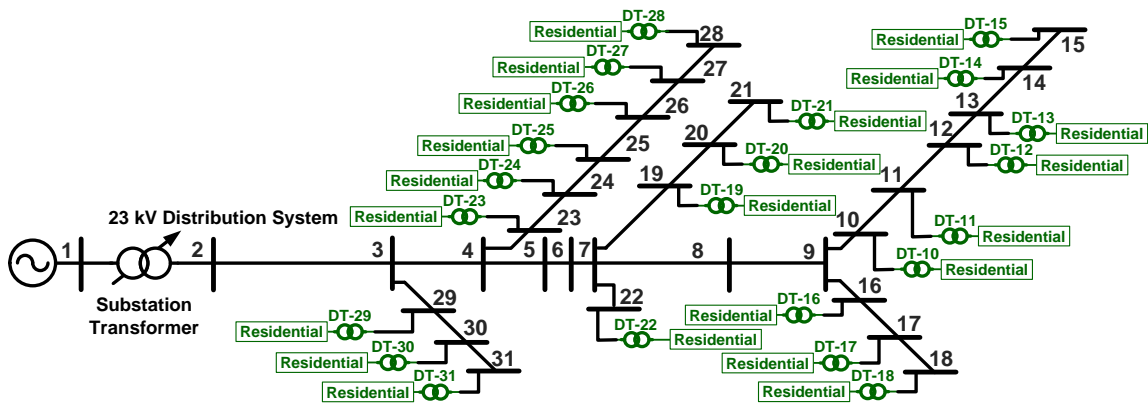
Figure 2



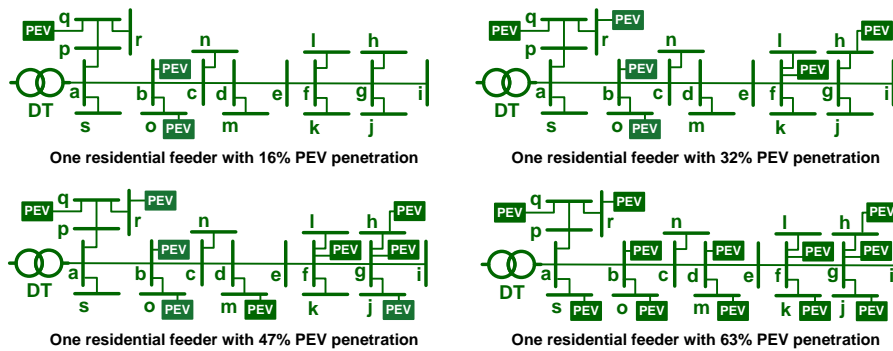
(a)



(b)

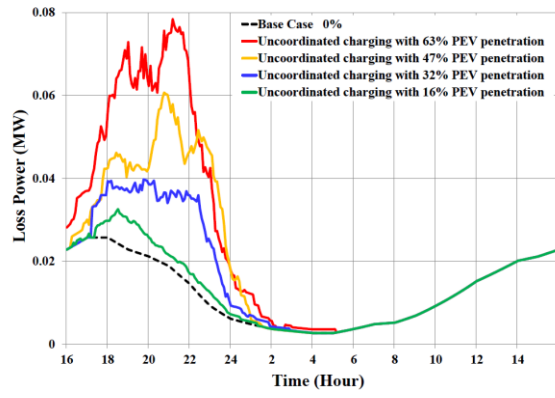


(c)

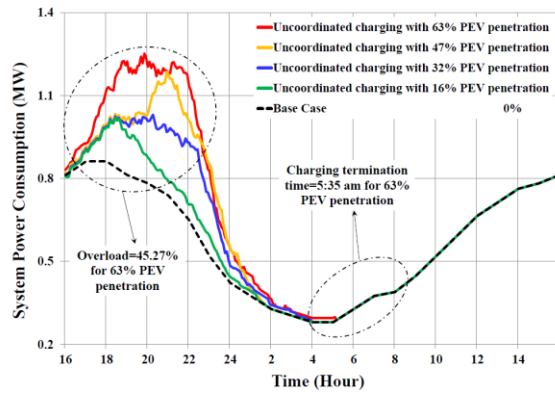


(d)

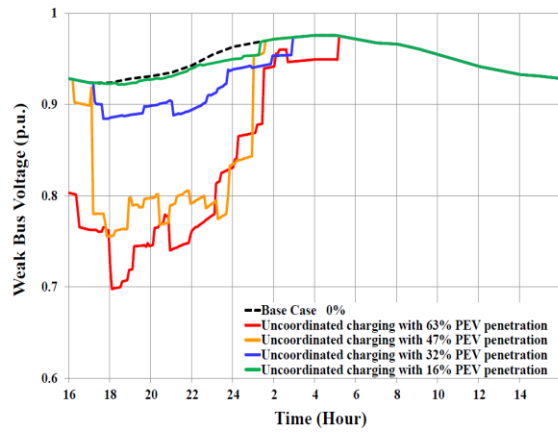
Figure 3



(a)

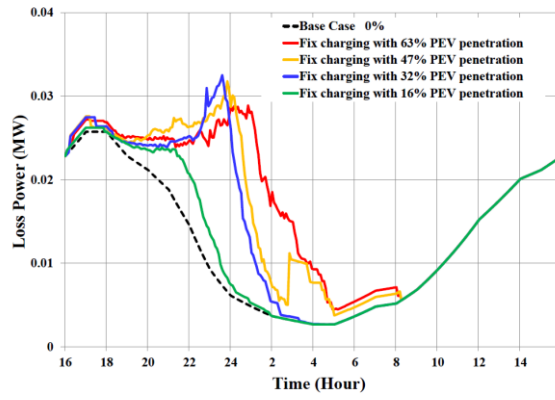


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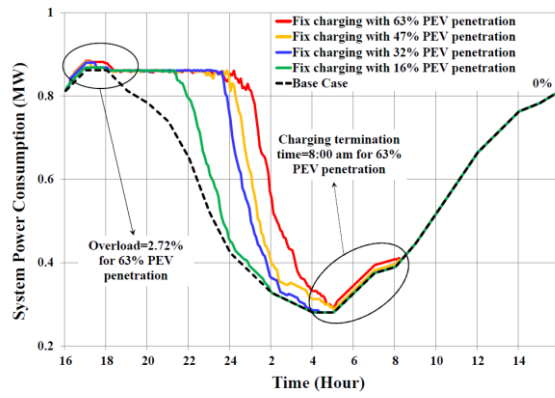


(c)

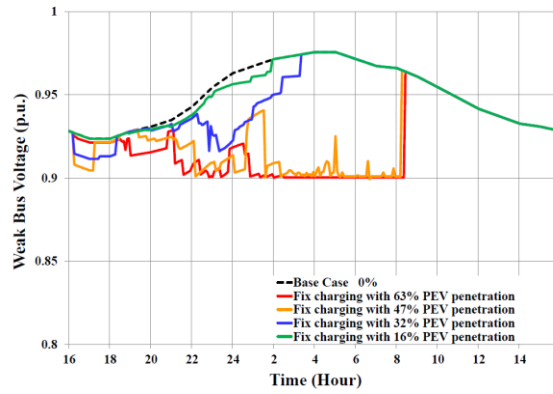
Figure 4



(a)

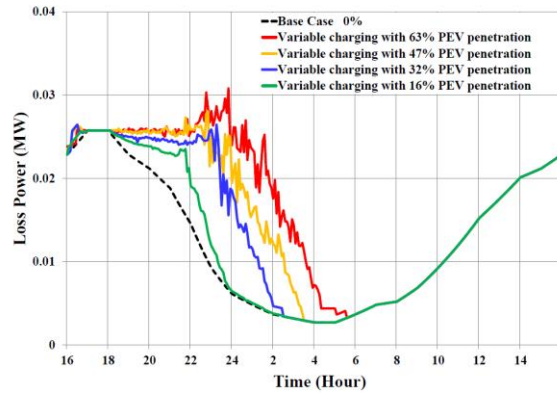


(b)

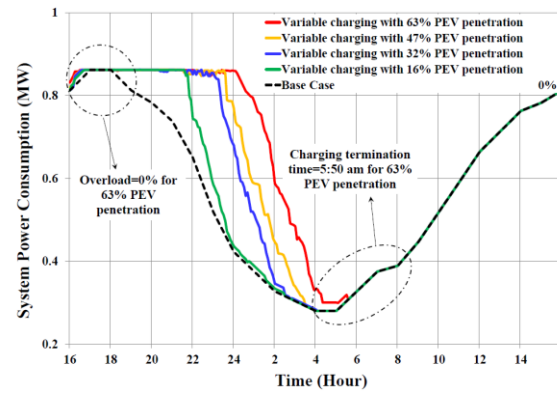


(c)

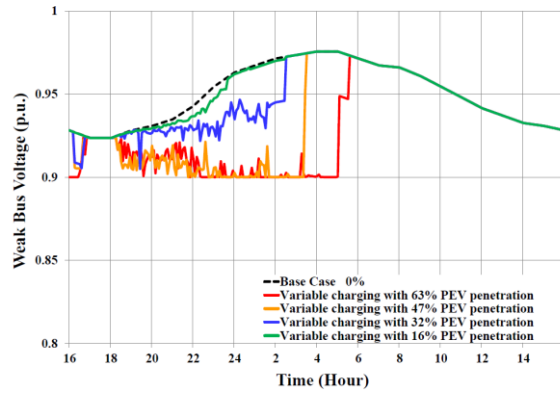
Figure 5



(a)

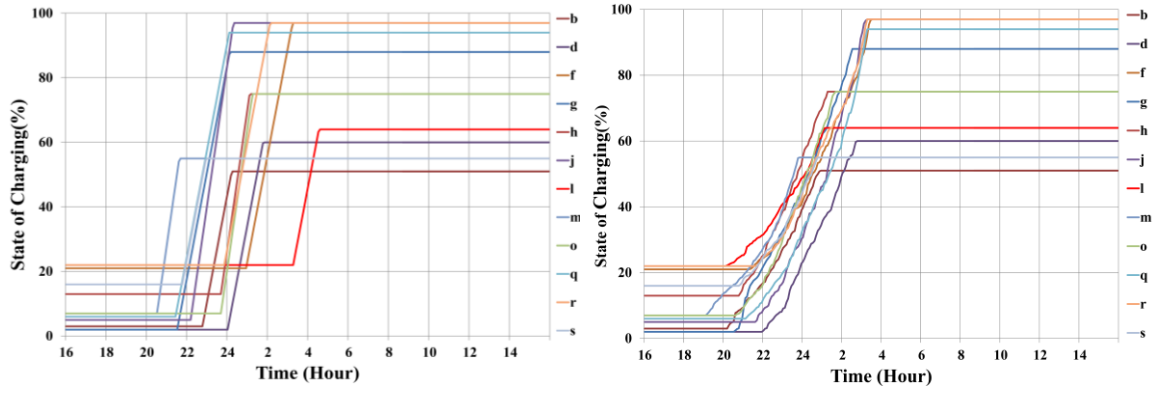


(b)



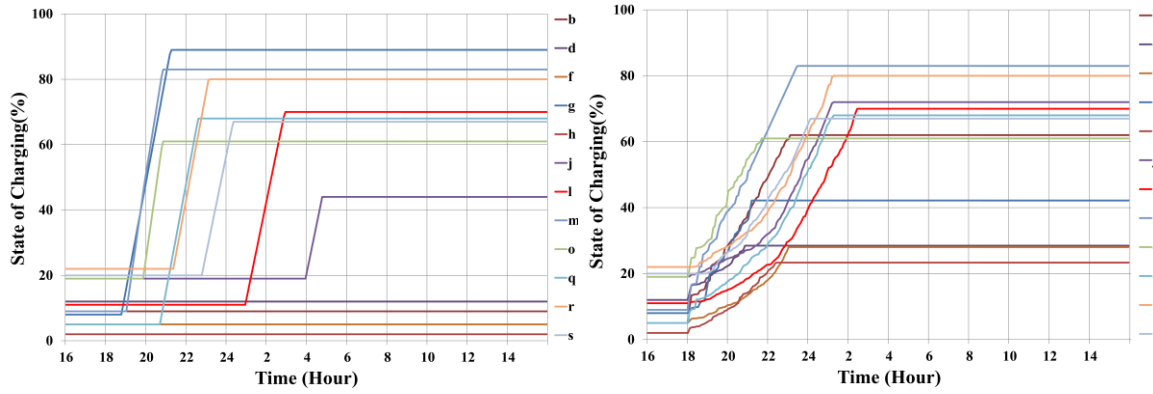
(c)

Figure 6



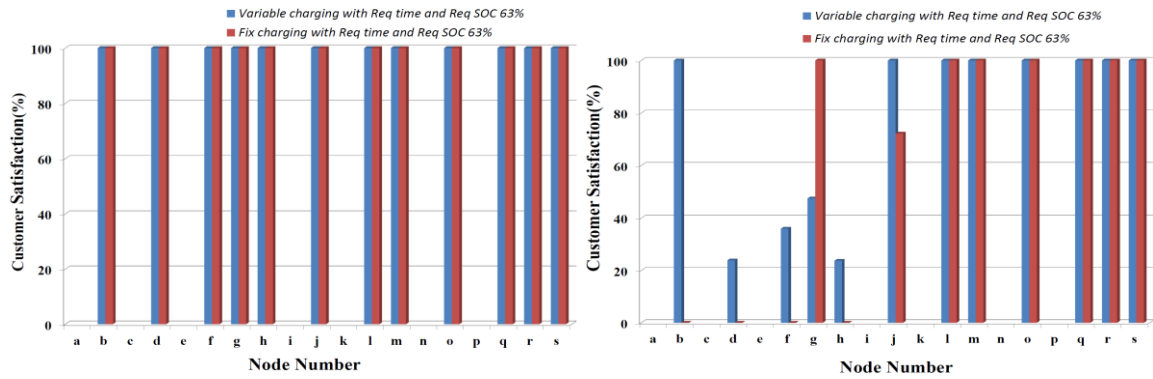
(a)

(b)



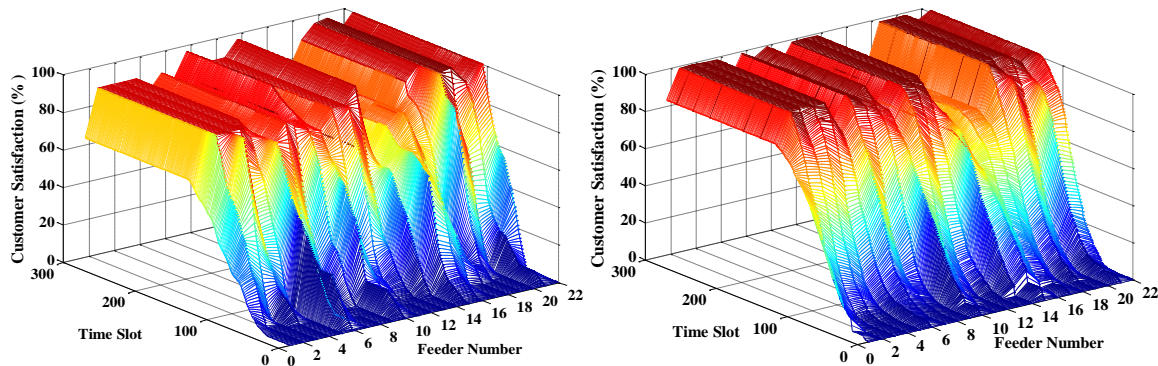
(c)

(d)



(e)

(f)



(g)

(h)

*Highlights (for review)

- Proposing a CAPSO based algorithm for variable online PEV charging coordination.
- Precise modeling of battery and charger for each PEV considering their efficiency curves.
- Formulating the variable PEVs charging coordination problem to optimize customer satisfaction index.
- Modeling of customer behavior and their driving patterns as well as required SOC in a desired charging time.
- Comparing the performance of variable and fix charging coordination.