

A Novel Central Voltage-Control Strategy for Smart LV Distribution Networks

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Abstract. With the inclusion of Information and Communication Technology (ICT) components into the low-voltage (LV) distribution grid, some measurement data from smart meters are available for the control of the distribution networks with high penetration of photovoltaic (PV). This paper undertakes a central voltage-control strategy for smart LV distribution networks, by using a novel optimal power flow (OPF) methodology in combination with the information collected from smart meters for the power flow calculation. The proposed strategy can simultaneously mitigate the PV reactive power fluctuations, as well as minimize the voltage rise and power losses. The results are very promising, as voltage control is achieved fast and accurately, the reactive power is smoothed in reference to the typical optimization techniques and the local control strategies as validated with a real-time simulator.

Keywords: Smart grid · Renewable energy integration · PV power forecasting · Smart meter · Power system state estimation · LV distribution networks · ICT

1 Introduction

As the penetration of residential and commercial PV increases into electrical distribution systems, an actual trend in the south of Germany, problems such as voltage rise, overloading of network equipment, harmonic current emissions, network resonance, false islanding detection, and dc current injections are becoming more of an issue to be addressed carefully [1]. Concrete solutions for secure and reliable integration of distributed PV generation into distribution systems are a fundamental concern for both academia [4, 6, 7] and industry [8, 19, 32].

A wide range of research projects have focused their efforts particularly on studying how to develop reverse power flow or voltage rise regulation methods to allow this integration. Conventional control mechanisms for distribution voltage regulation are: Voltage control on the feeder by using on-load tap-changing transformers (OLTC) [28, 33, 34], and fixed or switched capacitors to offset the reactive power demand from the load and thus reduce the current flow through the feeder and the related voltage drop [5, 19, 35]. The problem with OLTC transformers or voltage regulators is that, the amount of permissible voltage increase is limited if there is a load near the voltage regulator, a common case in LV distribution networks, so additional voltage regulators along the feeder may be necessary.

On the other hand, it is also questionable whether the capacitor bank technology is sufficient to answer these challenges, because it may require faster and more flexible control systems than the achievable with capacitor banks [35].

Another potential solution is the use of PV inverters' reactive power as a promising inexpensive concept to resolve the problems caused by PV penetration. Its development and realization attract research efforts in a fairly large number of issues ranging from modeling [3, 8] to implementation [32, 36].

Originally, researchers focused their attention on local or decentralized voltage-control approaches. Nonetheless, in the last few years, optimization techniques to support central control strategies have been proposed, using deterministic optimization methods [13, 24, 25, 27]; non-deterministic optimization methods [20, 23]; and hybrid methods [26, 31].

Central control strategies are demonstrated to be able to resolve voltage violation in LV distribution systems. However, it has repeatedly been shown that these methods may produce unwanted reactive power fluctuations [10, 25, 27, 28]. If the PV penetration is large and widespread, this may also affect subtransmission and transmission systems. This can have important economic impacts and technical implications for distribution substations and transmission lines, such as increasing losses and line loading and so on [9].

At this aim, in this paper a novel central voltage-control strategy is proposed. It is based on optimal reactive power control of smart three-phase solar inverters and the analysis of the data received from the smart meter to solve the OPF. The proposed optimal formulation, which simultaneously minimizes the magnitude of the voltage rise and reduces the power losses, includes a function to smooth the reactive power output of the inverters to improve the power quality in LV distribution network. The remainder of this paper is organized as follows. Section 2 presents the structure of the PV control strategy. The experimental setup and simulation results are illustrated in Sect. 3. Section 4 describes the conclusions drawn from the study carried out in this paper and suggests some guidelines for the future work.

2 PV Control Strategy

The main strategies to control PV systems can be classified as: Local, decentralized and central control strategies [29].

- Local control schemes (also known as droop-based regulation strategies) make autonomous control of the reactive power supply via characteristic curves.
- Decentralized control is based on the control of the reactive power of PV and the interaction with the OLTC transformer in the substation. In this case, some local communication is necessary to enable the interaction between the inverters and the decentralized methodology.
- The central control scheme can be described as a communication based control methodology that allows optimizing the LV distribution grid operation not only locally but also regionally with a common beneficial level for producers and consumers.

2.1 Local Voltage-Control Strategies

The local voltage-control strategies analyzed in this paper correspond to the proposed by German code of practice GC VDE-ARN 4105:

- Power factor characteristic: $\cos\phi(P)$ method .
- Reactive power/voltage characteristic: $Q(U)$ method.

The $\cos\phi(P)$ and $Q(U)$ scheme are based on droop characteristic, as shown in Fig. 1. In Germany, the $\cos\phi(P)$ method proposes a procedure for the calculation of reactive power as a function of the active power generated by the solar systems. With a low active power of PV, the risk of voltage overshoot is low. The reactive power is then adjusted to zero. Once the real power increases to the half nominal power of the solar system, the reactive power increases linearly until the power factor of 0.95 for the case of PV with reactive power level between 3.68 kVA and 13.8 kVA, the typical residential case. For PV with power delivery > 30 kVA the suggested power factor is 0.9. Equations (1) express the $\cos\phi(P)$ curve, as shown in Fig. 1a.

$$\cos \varphi (P) = \begin{cases} \cos \varphi_1, & P < P_1 \\ \cos \varphi_1 + (1 - \cos \varphi_1) (P_1 - P/P_1 - P_2), & P_1 < P \leq P_2 \\ -1 + (1 + \cos \varphi_2) (P_2 - P/P_2 - P_3), & P_2 < P \leq P_3 \\ \cos \varphi_2, & P \geq P_3 \end{cases} \quad (1)$$

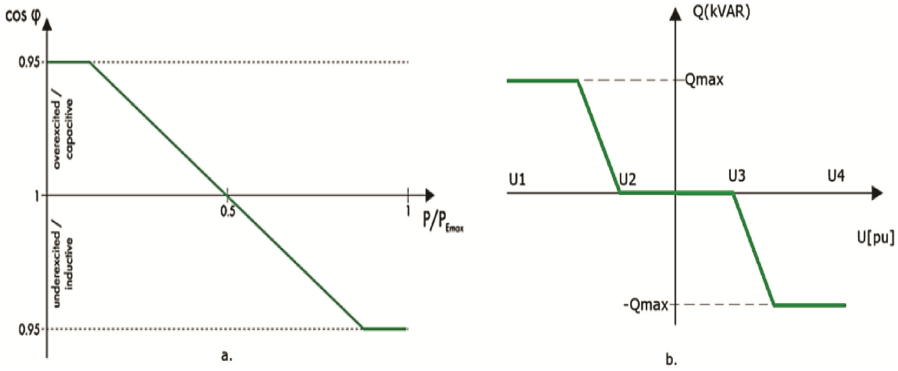


Fig. 1. Characteristic reactive power curves. **a.** $\cos\phi(P)$ curve. **b.** $Q(U)$ curve.

On the other hand, in the $Q(U)$ method the reactive power of the inverter is regulated as a function of the voltage at the coupling point, as shown in Fig. 1b. It is worth noting that two droop ratios are available when the voltage is higher than the normal range. Besides, achieving better voltage control functions can differentiate the voltage responses of the inverters near LV transformer from the rest along the feeder, so the reactive power contributions from all inverters along the feeder can be more equally

distributed as in the case of the cosphi(P) method [10]. As stated in the German GC, the droop curve for the Q(U) method is provided by the network operator.

The algorithm for Q(U) method can be summarized by (2).

$$Q(U) = \begin{cases} Q_{\max}, U < U_{\min} \\ (U - U_1/U_{\min} - U_1) \cdot Q_{\max}, U_{\min} \leq U < U_1 \\ 0, U_1 \leq U \leq U_2 \\ -(U - U_2/U_{\max} - U_2) \cdot Q_{\max}, U_2 < U \leq U_{\max} \\ -Q_{\max}, U > U_{\max} \end{cases} \quad (2)$$

2.2 Central Voltage-Control Strategies

The technical effectiveness of local voltage-control strategies has been very well studied. In particular, study [2] finds that many inverters have the capability of providing reactive power to the grid in order to reduce the voltage rise. Using a similar method, study [3, 6] find that a decentralized voltage-control strategy works just as well, via special located measurement systems, distribution OLTC transformers and controllable PV inverters.

In contrast, central control strategy aims for coordinated control of the complete LV distribution system by using the static and dynamic system information. The target of this strategy is to find the OPF of the LV distribution systems with high penetration of PV. OPF has been the predominant method for such analysis since its introduction by Carpentier (1962) [12].

OPF seeks to optimize a given cost, planning, or reliability objective by controlling the power within an electrical network without violating network power constraints or system and equipment operating limits. Such as conventional power analysis, OPF determines voltage, current, and injected power throughout an electrical power system, that is, the system's state of operation. The general OPF problem is a nonlinear, non-convex, large-scale optimization problem which may contain both continuous and discrete control variables [11].

The most common OPF objective function for the case of PV integration are: Power loss minimization (PLM) [14–16], Voltage rise minimization (VRM) [17–19], PV generation cost minimization (GCM) [20–22], and their combination as multi-objective OPF problems [10, 13, 24, 25, 27].

In the case of VRM, it can be implemented with local control approaches, but the line impedances data is required for the calculation. As when it is implemented in a central control scheme, this approach becomes a similar optimization problem as the PLM strategy [13]. So, this formulation is used as reference to compare the improvements of the proposed central voltage-control strategy, as follows.

2.3 Proposed Central Voltage-Control Strategy

It is well known that central control strategies require the information of the grid topology and the characteristics of the distribution systems, as well as the current status of the buses in terms of voltage, reactive and active power. As the dynamic information is usually only available for a few locations in the system, some distributed state estimation algorithms are required to guarantee the power flow calculation [29, 31]. However, few studies have integrated the actual ICT components of the smart LV distribution networks into the central control strategy [13, 27]. Smart meter is one of these ICT components, which is possible to avoid the necessity of complex state estimation algorithms.

The analysis of the information received from the smart meters allows the power flow calculation of the smart LV distribution system, as presented by the authors in [30] for the implementation of OPF for the central voltage-control methodology. To do so, the proposed control system computes the reactive reference values for the controllable PV inverters every 10 s by minimizing a multi-objective problem using a sequential quadratic programming algorithm developed in Matlab®.

The multi-objective function consists in three optimization objectives, as follows.

Minimization of the power losses:

$$F_1 = \sum_{i=1}^{n-1} \sum_{j=1}^{n-1} R_{ij} S_{ij}^2 / U_{ij}^2 \quad (3)$$

where $i, j = 1, 2, \dots, n$ is the bus number, R_{ij} , S_{ij} and U_{ij} are the resistance of the branch between nodes, the power and voltage obtained from the power flow calculation respectively.

Minimization of the amount of power provided by the PV systems:

$$F_2 = \sum_{i \in \delta} S^2 \quad (4)$$

where δ denotes the set of buses with PV installation [27].

Minimization of the violations to a dead filter-based band to smooth the reactive fluctuations:

$$F_3 = \sum_{i \in \delta} (Q_i - Qf_i)^2 \quad (5)$$

where Q_i is the reactive power for the PV inverter in the bus number, and Qf_i is the output of a Parzen window filter designed to smooth the reactive power fluctuations.

This FIR filter is based on a buffer with the last smoothed reactive power outputs, on which the buffer size corresponds to the size of the average window.

Finally the proposed OPF formulation is as follows:

$$\min \sum_{o=1}^3 W_o F_o(X)$$

where W_o are the weighting factors for each objective function;

subject to:

$$h = \begin{cases} (P_i^2 + Q_i^2) \leq S_i^2 \\ U_{min} \leq U_i \leq U_{max} \\ |Q_i| \leq Q_{i,max} \end{cases}$$

where U_{min} and U_{max} denote voltage boundaries and $Q_{i,max}$ corresponds to the maximal reactive power provided by the PV inverter. A basic System schematic of a Smart LV distribution network with the proposed central voltage-control strategy is presented in Fig. 2.

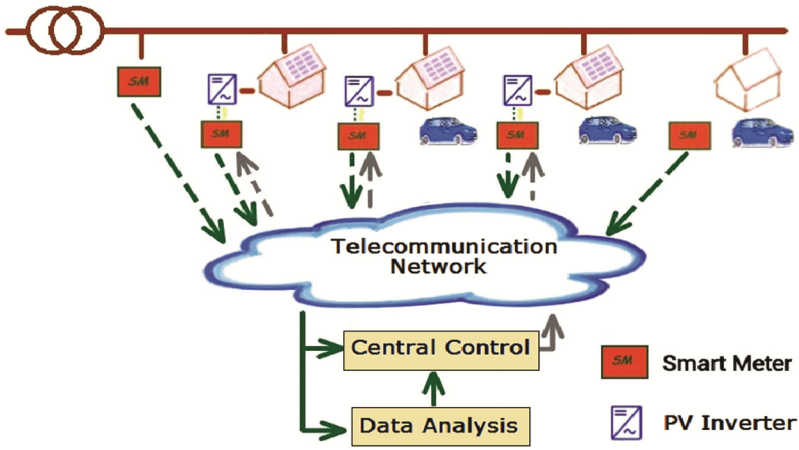


Fig. 2. Smart LV distribution network with central voltage-control strategy.

2.4 Implementation of the Central Voltage-Control Strategy

Figure 3a outlines the architecture of the central strategy developed. For each sample time a power flow analysis is done with the flat start conditions for voltage $U_i = 230 + j0V$, the set active power generation P_i and the reactive power for the PV inverters as $Q_i = 0VAR$. Then, the OPF algorithm takes as reference the output values of the power flow algorithm to find a feasible solution for the defined objective function and the respective restrictions.

Afterwards, the output values of the OPC algorithm are given as reference to the inverters and other load flow analysis is done to check the status of the system and giving a feedback to the OPF algorithms for the smoothing function.

Figure 3b shows the flow chart of the proposed formulation for power flow calculation based on the analysis of the information provided from the smart meters. More detailed description of the power flow calculation with the smart meter information can be found in [30].

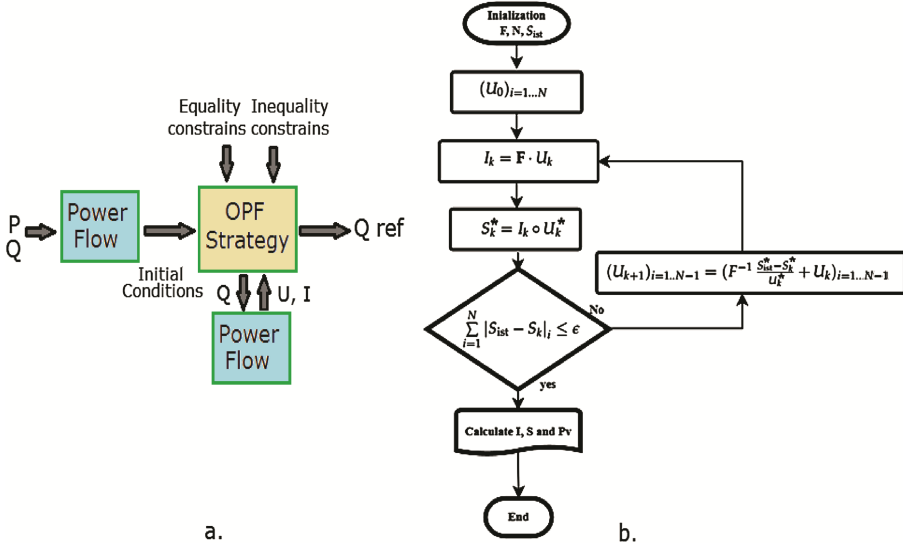


Fig. 3. Flow chart with the implementation of the proposed OPF strategy.

3 Simulation Results

3.1 LV Distribution Grid Model

The representative LV distribution network configuration of the German power system used for the simulations and analyses comprises 30 smart homes equipped with a scalable PV system. All main feeder cables are of type NYY $4 \times 25 \text{ mm}^2$. The households have a three-phase connection with a nominal line-to-neutral voltage of 230 V.

The voltage at the secondary side of the transformer is considered to be 235 V during no load, which can be considered as a typical LV transformer tap to avoid low voltages at the end of the feeder. The solar installations are defined at maximum 5.5 kVA including a 10 % overrating, to support a reactive power compensation until 46 %, even when operating with full power generation. The loads are defined as smart homes instead of passive loads to improve the simulation scenarios. The German smart home load profile used in this paper is very well defined in [32].

3.2 Voltage Rise and Fluctuation

The voltage magnitude along feeders at the end of the simulation for the different reactive power control strategies are illustrated in Fig. 4. Due to the insertion of PV and the fact that there is no control law in the reference case, overvoltage occurs over the limit of 10 % as the worst case scenario, as shown in Fig. 4a.

Despite that all the inverters under $\cos\phi(P)$ strategy contribute equally to the grid voltage support, the overvoltage limit is reached in the same way as the reference case

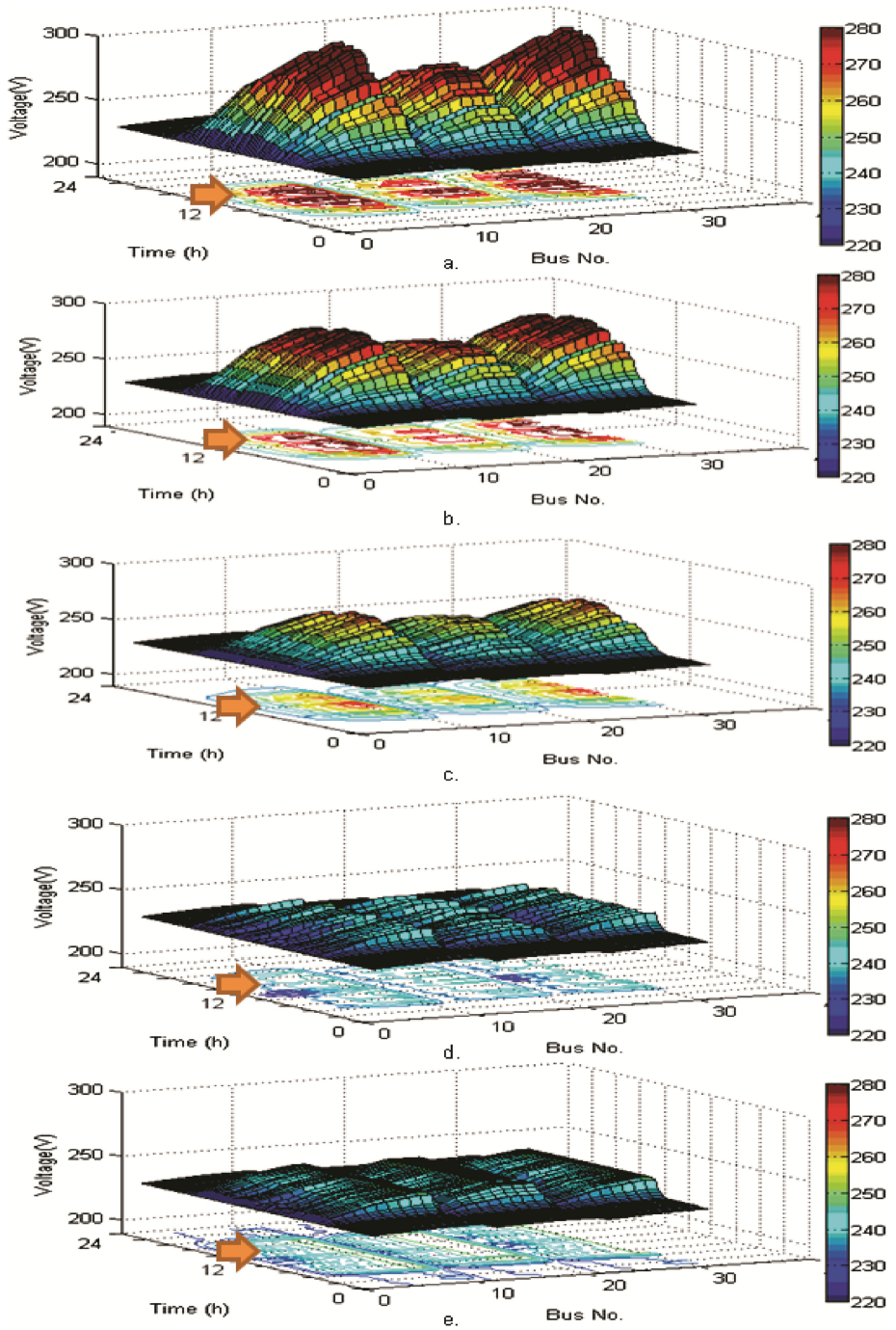


Fig. 4. Comparison of voltage magnitude for the different control strategies.

with a decrease only of 8 %, as shown in Fig. 4b. In the case of Q(U) method, the overvoltage is less in comparison with the cosphi(P) case, but some overvoltage is evident as seen in Fig. 4c.

As could be expected, the bus voltages in the optimization cases remain closer to the nominal value in both cases. It can be observed that the typical optimization strategy allows a several reduction in the voltage level around the nominal value, but the voltage magnitude presents a higher fluctuation, as shown in Fig. 4d. In the case of the proposed strategy, the voltage profiles are more homogeneous close to the references, as seen in Fig. 4e.

3.3 Network Power Loss

Figure 5 provides a comparison of the power losses for the different strategies. The disparity in the power losses is gross as the PV active power injection increases. From these results, it can be seen that Q(U) strategy reaches lower power loss in comparison with the reference case and the cosphi(P) strategy. The proposed OPF strategy has the lowest power loss, with a reduction of 67 % with respect to the reference case. In the case of the Q(U) method and reference optimization strategy, the losses are almost 20 % and 30 % higher than the proposed strategy respectively. The reason for this is because both strategies try to maintain the voltage profile closer to the nominal voltage magnitude value by injecting extra reactive power, as can be observed in the following subsection.

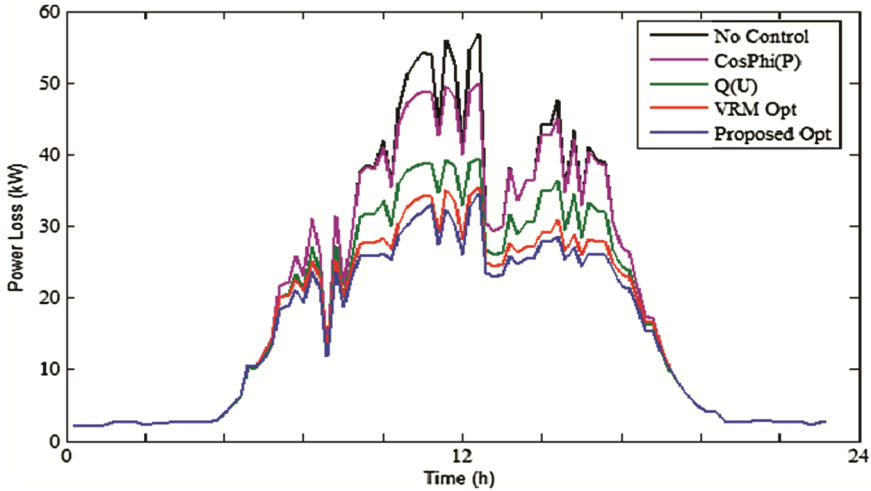


Fig. 5. Comparison of the power losses for the different strategies.

3.4 Reactive Power Fluctuations

A comparison of the cumulative reactive power in the critical feeders is presented in Fig. 6. As expected, in the case of cosphi(P) method, the injected reactive power is

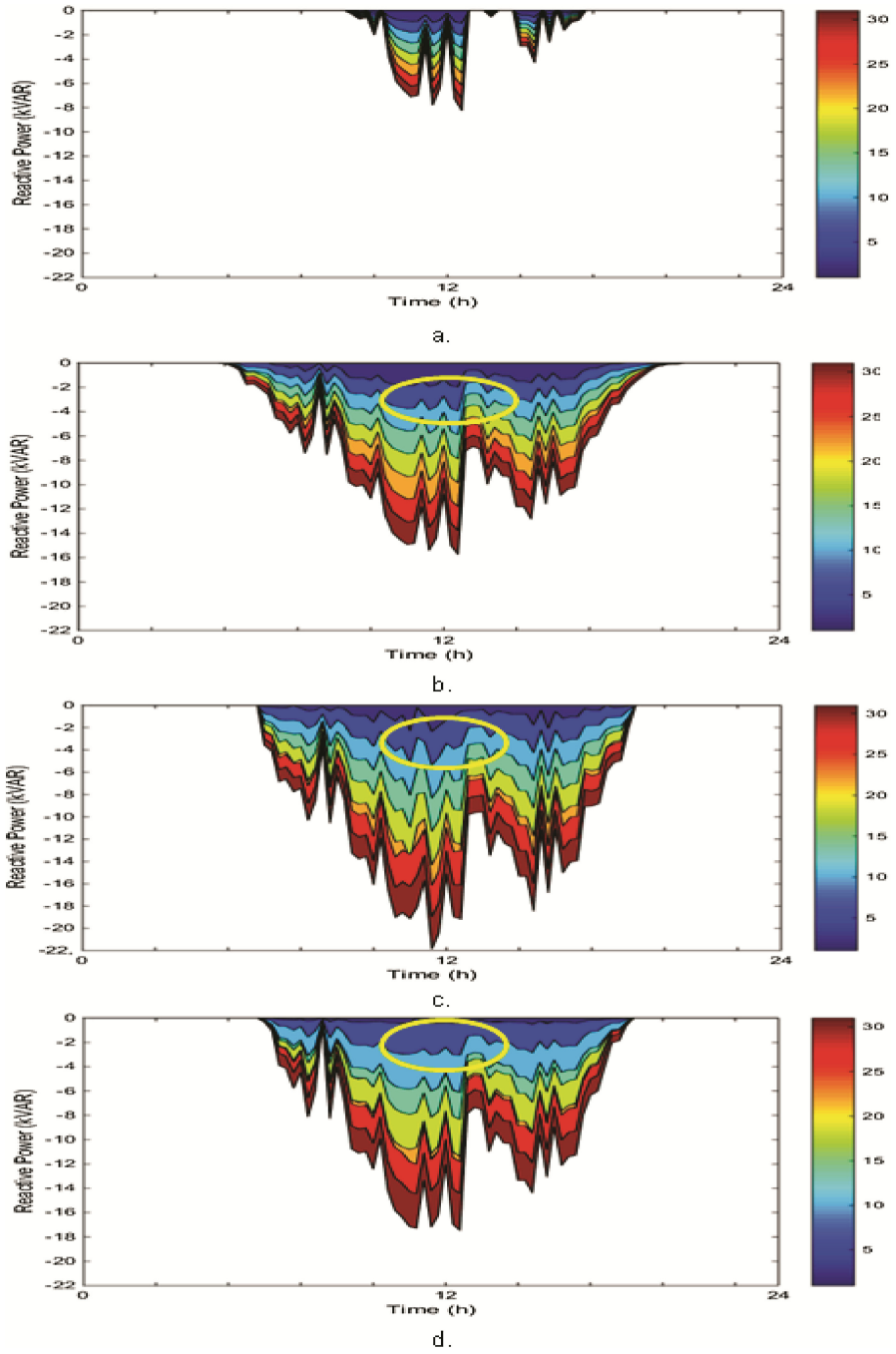


Fig. 6. Comparison of the cumulative reactive power in the critical feeders.

relative low and has the same performance for all the analyzed households, as shown in Fig. 6a. In the case of Q(U) method, as the reactive power depends on the voltage magnitude measured at its corresponding coupling point, the cumulative reactive power increases proportionally at the increasing of the PV active power injection, as shown in Fig. 6b.

Typically, the reactive power fluctuations of PV systems for the OPF case reach or exceed the performance of the PV active power injection, due to the influence of the minimization voltage rise function, which attends to deal with the deviation of the voltage magnitude even for negatives values. Figure 6c and d show the cumulative reactive power of the reference OPC strategy and the proposed strategy respectively.

While the reference OPC strategy and Q(U) method have similar behavior, with high reactive output power fluctuations, the proposed method significantly outperforms the two. This observation is confirmed by Fig. 7, where the reactive power outputs of two PV inverters are shown for the proposed strategy and the reference optimization technique.

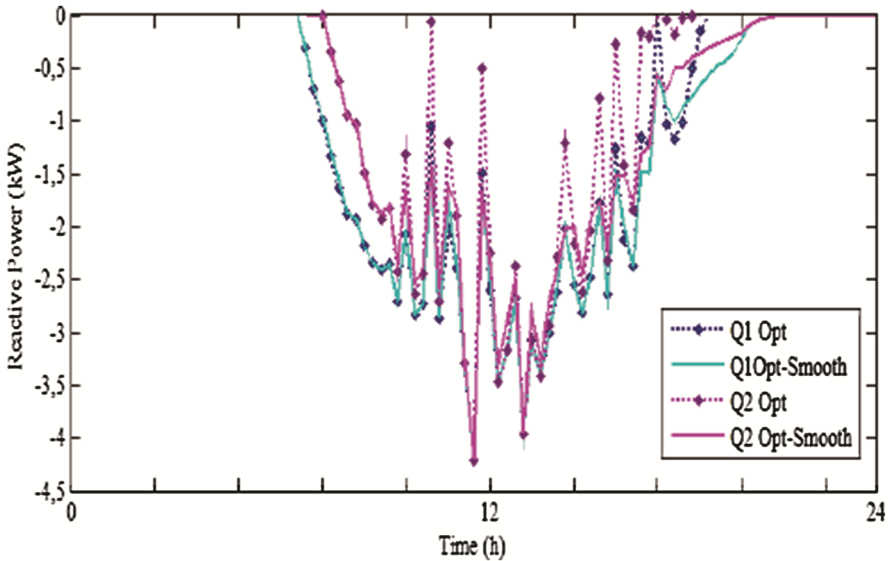


Fig. 7. Reactive power fluctuations.

3.5 Real-Time Digital Simulation Testing

To validate the theory which has been discussed before, a real-time simulation is performed, as shown in Fig. 8. The Simulink model which compares the OPF central control strategy and the Q(U) method with respect to the reference scenario is built in a host PC.

At that point, the code is compiled to the target computer to run the simulation of the smart LV distribution network in real time. Subsequently, the target computer runs the real-time simulation and sends the data, which represents the smart meter's information, to the central controller where the reactive power for the PV inverters is calculated and sends back to the real time simulator.

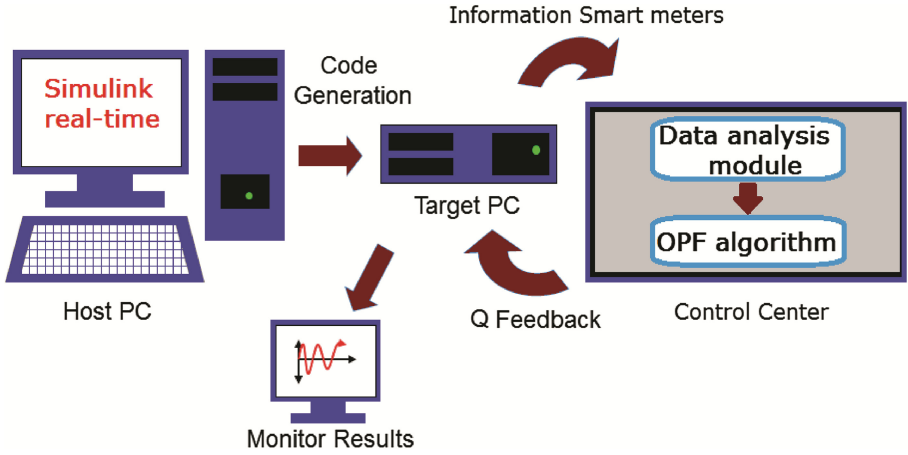


Fig. 8. Real-time simulation scheme based on Simulink Real-Time™.

Figure 9 shows the results obtained from the real-time simulator. Scope 1 shows the PV solar profile simulated for 5 days. Scope 2 and 3 correspond to the RMS voltage response of the household without control nearest and farthest to the transformer respectively.

The results obtained for the smart homes with Q(U) local control strategy are showed in the scopes 4–6, and the smart homes with OPF central control are illustrated in scope 7–9. Scope 4 and 7 compare the reactive power performance of the PV inverter of the household nearest and farthest to the transformer for local and central control respectively.

As can be seen from above real-time simulation results, when the solar power goes higher, the voltage from the household without control exceeds the up limit. The results obtained from the local control improves, but still exceeds the up limit during some peak solar power time period. However, on the other hand, the optimization central control not only regulates the voltage into the limit band for all solar scenarios, but also makes the voltage and reactive power profile smoother in accordance with which has been proposed in this paper.

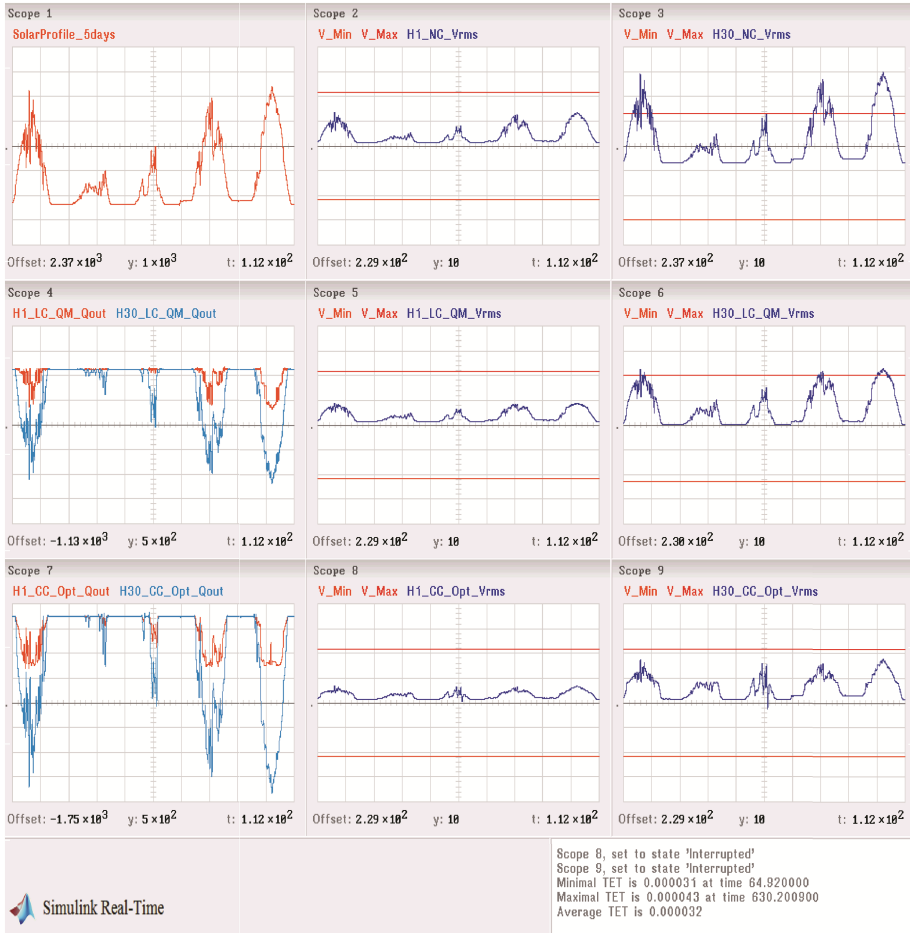


Fig. 9. Real-time digital simulation results.

4 Conclusion

This paper has presented a novel central voltage-control strategy for smart LV distribution networks. The proposed strategy is designed to reduce the reactive power fluctuation of the PV inverters to improve the power quality and reduce the power losses of the system. This is one of the important features of the proposed OPF method, because it improves damping and stability of the reactive power significantly in reference with the typical optimization techniques and the local control strategies.

The results are very promising, as the central voltage-control strategy is capable to mitigate the voltage rise with a better minimization of the power loss. Moreover, a new formulation for the power flow analysis based on the information from the smart meters was included and validated with several simulations included a test with a real-time simulator.

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