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Impacts of supercapacitors on battery lifetime in hybrid energy storage system in building integrated photovoltaic DC micro-grid

M. Gaetani-Liseo, C. Alonso and B. Jammes

LAAS-CNRS

University Paul SABATIER - TOULOUSE 3

Toulouse, FRANCE

mgaetani@laas.fr, alonsoc@laas.fr, jammes@laas.fr

Abstract—This paper focuses on a study of an hybrid energy storage system constituted by lead acid batteries and supercapacitors and designed to work in a low voltage DC micro-grid supplying by photovoltaic sources and dedicated to building application. We proposed a simulation methodology in order to evaluate supercapacitors performances and behaviors for long term analysis. We quantify their impacts on battery lifetime and on the levelized cost of electricity of our storage system. This analysis has been validated on the micro-grid developed in LAAS-CNRS in ADREAM building integrated photovoltaic. The power profiles used for our work correspond to 1 years production and consumption data from 100m² floor of the building. Results of comparative studies with different load profiles, SC costs and methods to estimated battery lifetime are presented and discussed.

Index Terms—LVDC-MG; hybrid ESS; Supercapacitor; Lead-acid Battery; LCOE; aging model; PV building

I. INTRODUCTION

Ecological issues as global warming, fossil energy depletion with increasing electrical demand is nowadays a well know challenge. One of the key to reduce our ecological footprint are initiatives as Zero Emission Building (ZEB) and autonomous buildings. To tackle this issue distributed generators as decentralized Low Voltage DC Micro-Grid (LVDC-MG) with PhotoVoltaic sources (PV) integrated in Building (BiPV) is a widely used solution for urban area [1]–[3]. Nevertheless the intermittent nature of the PV production, and the time shift between consumption and production requires Energy Storage Systems (ESS) with associated strategies [4], [5].

Within this context Hybrid Energy Storage System (HESS) with batteries and SuperCapacitors (SC) association have been attracted a considerable attention in order to improve efficiency and promote sustainable electrical network. Indeed this topology, frequently proposed in electric vehicle applications as in [6] and [7], seems appropriate to improve batteries lifetime because of the complementary performances of SC and based on the assumption that the SC lifetime isn't impacted by micro cycles. Therefore, they can avoid small batteries discharging-charging cycles by delivering (or absorbing) the power during the irradiance (or load) intermitences [8]–[10]. The aim of this paper is to analyze the impact of the SC on the battery lifetime

and on the HESS life cycle cost in building application. This analyze is based on real operations thanks to data recorded every minute in ADREAM BiPV Data Base (DB) [11]. Two different load profiles: the lighting network and the electrical outlets power profiles, two different battery aging estimation method and best and worst SC costs are used in order to compared results and achieve a complete analysis and discussion.

In II we present the existing LVDC-MG developed in the LAAS-CNRS BiPV.

Afterwards, in section III we details the ESS models and the LVDC-MG simulation algorithm, developed on Matlab ©.

In section IV we define the two battery aging estimation methods used and the criterion used to analyze the HESS cost.

Finally we present, compared and discussed our results for different load power profiles and SC cost in section V.

II. THE LVDC MG

The LVDC-MG developed in LAAS-CNRS was described in [12]. Its architecture is presented in fig 1. The HESS associates two ESS connected in parallel to the DC bus through bidirectional DC/DC converters. ESS A is SC pack and ESS B is OPzV Lead-acid (Ld) batteries pack. These two technologies have been combined because of their different and complementary characteristics: Ld batteries are known as a mature and cheap electrochemical storage technology with a high energy density, while SC support a high number of cycles and present a high power density, but it is a very expensive technology [13]. We developed a specific power management algorithm explained in [12] and derived from DC bus signaling control [14]. This control algorithm gives the priority to SC charge and discharge, in order to avoid micro-cycles in the batteries.

ESS A is made up of Maxwell Technologies SC with maximum voltage V_{ESSA} of 48V and a capacity C_{SC} equal to 165 F [15]. ESS B is constituted by 6 Hoppecke OPzV 6V batteries connected in series. The nominal capacity Cap_{nom} of each battery is equal to 250Ah (@C10) [16]. The maximum voltage of EES B is 42.3V. The characteristics of the two ESS are summarized in table I.

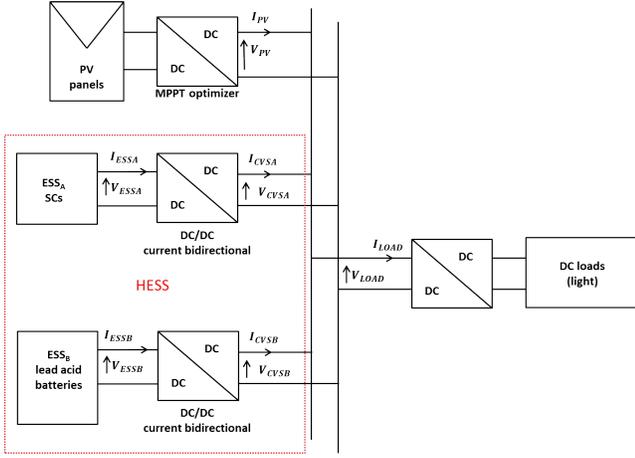


Fig. 1. Electric synoptic of the LVDC-MG developed in the LAAS-CNRS

TABLE I
MAIN CHARACTERISTICS OF ESS IN LVDC-MG

ESSi	A : SC pack	B : OPzV Ld battery pack
Number of element in serial connection (n_s)	1	6
Maximum Voltage [V]	51	42.3
Nominal Voltage [V]	48	36
Current range I_{ESSi} [A]	-	[-25,25]
Power ranges P_{ESSi} [kW]	-	[-1,1]
Nominal Energy [Wh]	53	9000 (@C10)
State of charge range (SoC) [%]	[0,100]	[10,90]
Investment cost [€/kWh]	24000	300

Currents and powers have been chosen positive when EES are discharging. According to the data-sheet the maximal value of the battery current corresponds to the nominal C-rate (0.1C). To reach different values of energy installed with the technologies A or B, we connected in parallel n_s element in each ESS_i . For each ESS, the investment cost proposed is an average value of installation costs given by distributors. These values are coherent to the ones publish in [17]–[19] for OPzV Ld batteries and in [20] for SC. However, SC installation cost varies a lot in the literature and few data only are available owing it is a developing technology. Usually, the cost used in papers is lower than the one we give in table I. A sensitive analysis of this parameter will be done in section V.

III. SIMULATION METHODOLOGY AND ESS MODELS

A. PV and Load power profile data

For our study, we selected two one year power data set from the ADREAM BiPV DB (time step = 1 min) which are

representative of the consumption associated to tertiary buildings and relevant to be supplying by LVDC MG [21]–[23]: lighting network and electrical outlets. The PV production profile corresponds to 8 panels (2 kWp) of PV plant installed on the ADREAM building terrace. The lighting and electrical outlets power demand are both corresponding to 100 m² of the second floor of the ADREAM building occupied by offices. Within this context we calculated the global capacity of the HESS in order to be autonomous.

These power profiles allowed us to determine the power profile input (P_{BAL}) of our LVDC-MG model according to equation 1.

$$P_{BAL} = P_{PV} - P_{LOAD_p} \quad (1)$$

where p index can be "lighting" or "outlets" consumption power profile.

B. ESS models

1) *SC model*: SC are modelled by their equivalent electric Thevenin circuit (Eq. (2)), where C_{SC} is the capacity in Farad and R_{SC} the equivalent serial resistor given by manufacturer.

$$V_{ESSA}(t) = \frac{1}{C_{SC}} * I_{ESSA}(t-1) * \Delta t - R_{ESSA} * \Delta I_{ESSA}(t-1) + V_{ESSA}(t-1) \quad (2)$$

2) *Battery model*: The battery model is based on the Shepherd [24] and Tremblay's [25] models, which are commonly used for PV and MG applications [26]. According to the operating range of the battery (10% < SoC < 90%), we don't need to consider end of charge and end of discharge phenomena. In these conditions, the accuracy of a linear model is sufficient [27], [28]. Moreover, according to Tremblay's assumptions in [25], the battery temperature is considered constant, and self-discharged and Peukert effect aren't taken into account. The discrete version of the model is given by (3). V_{OC} is the open circuit voltage depending on the battery SoC and determined by (4) and (5) (Ah counting), Cap_{nom} is the nominal capacity in Ah, and R_{ESSB} is the battery internal resistance.

$$V_{ESSB}(t) = V_{OC}(t) - R_{ESSB} * \Delta I_{ESSB}(t-1) + V_{ESSB}(t-1) \quad (3)$$

$$V_{OC}(t) = K * SoC(t-1) + E_o \quad (4)$$

$$SoC(t) = SoC(t-1) - \frac{I_{ESSB}(t) * \Delta t}{Cap_{nom}} \quad (5)$$

The parameters K, E_o and R_{ESSB} have been identified thanks to several charging and discharging tests at constant current. Two different sets of parameters are defined for charging and discharging operations. The model has been validated with dynamic current profiles (one minute steps). In these conditions the maximum relative error is 6%, which is accurate enough for our application.

C. Overall simulation algorithm

The overall flow chart diagram in fig 2 represents the HESS simulation process. P_{ESS}^* is the power exchanged between the battery and the bus after SoC and C-rate limitation.

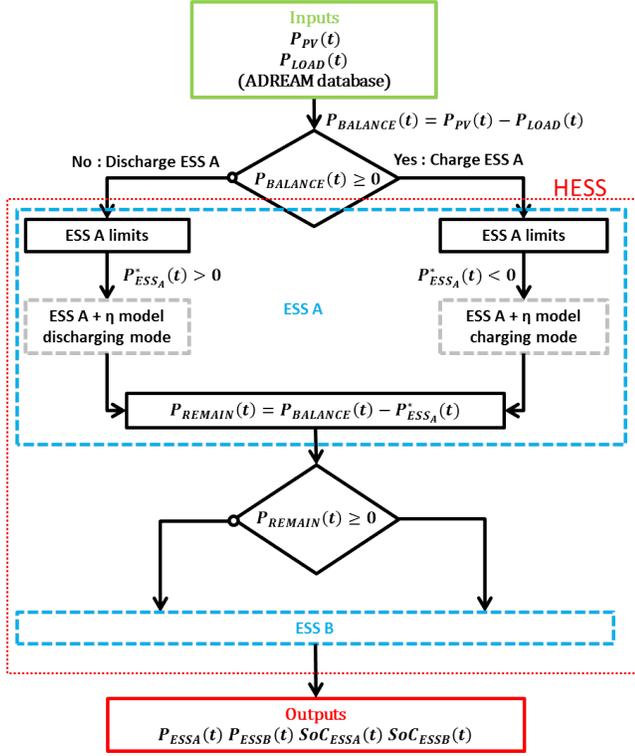


Fig. 2. Flow chart diagram of HESS simulation algorithm in LVDC-MG

The simulation process take into account the bidirectional DC/DC converter efficiency, noted $\eta(t)$. This efficiency which depends on the voltage and the power delivered or absorbed by ESS_i , where i is ESS A (SC) or ESS B (OPzV Ld batteries). The expression of $\eta(t)$ is given in 6, where f_k is a parametric quadratic function depending on V_{ESS_i} the ESS_i voltage. Its parameters are identified thanks to experimental tests.

$$\eta(t) = f_1(V_{ESS_i}(t)) * exp^{f_2(V_{ESS_i}(t)) * |P_{ESS_i}(t)|} + f_3(V_{ESS_i}(t)) * exp^{f_4(V_{ESS_i}(t)) * |P_{ESS}(t)|} \quad (6)$$

At each time t of the simulation, we solve (7), according to equations (2), (3) and (6).

$$\begin{cases} |P_{ESS_i}(t)| * \eta(t) - |P_{ESS_i}(t)| = 0, \text{charging} \\ |P_{ESS_i}(t)| * \eta(t) - |P_{ESS_i}(t)| = 0, \text{discharging} \end{cases} \quad (7)$$

Therefore, the simulation of power exchanges between the HESS and the MG takes into account the battery voltage variation, the power lost in the ESS_i and in the associated DC/DC converter.

IV. CRITERION

A. Lifetime and aging battery model

The battery life time L_B , expressed in years, can be estimated with different methods [29]. In this paper we propose to use two of them : the widely used equivalent full cycles to failure method (Ah throughput method) inspired by [30], and the Rainflow counting method [31].

1) *Ah Throughput*: The easy implementation of Ah throughput method make it the usually used in the literature to estimated batteries lifetimes. This methods proposes to compare the energy discharged (E_{disch}) during one year with the total energy that the battery can deliver according to its cycle lifetime at nominal deep of discharge (DoD_{nom}). The number of cycles to failure (N_{CF}) at different DoD are given in data-sheets, and the expression of the battery lifetime in years is given by (8). In this equation, E_{inst} represents the battery installed energy .

$$L_B = \frac{E_{inst} * DoD_{nom} * N_{CF}(DoD_{nom})}{E_{disch}} \quad (8)$$

One inconvenient of the Ah throughput method is that the lifetime estimation is based on the number of cycles to failure at fixed value of DoD. Therefore, this method doesn't take into account the nonlinearity of the N_{CF} vs DoD curve, as well as the battery SoC at each cycle.

2) *"Rainflow" counting*: This method allows to take into account the real impact of the DoD of each cycle on the lifetime estimation [31], [32]. The DoD range is divided in N intervals DoD_k ($k=1$ to N). For each cycle of the battery inside the DoD_k interval, the number of cycles N_{cycles_k} is incremented. At the end of the year, N_{cycles_k} values is compared with the number of cycles to failure, N_{CF} at DoD_k . Thus, the Cumulative Damage (CD) of the batteries is defined by (9), and the lifetime L_B in year can be estimated by (10).

$$CD = \sum_{k=1}^N \frac{N_{cycles_k}}{N_{CF}(DoD_k)} \quad (9)$$

$$L_B = \frac{1}{CD} \quad (10)$$

B. LCOE

We propose to analyze the economical impact of the quantity of energy made by SC in the HESS according to the Levelized Cost Of Electricity (LCOE), in €/kWh. The LCOE is the ratio between lifecycle cost and lifecycle energy production, and allows to asses and compare electricity generators. This criteria estimates the electrical energy cost including capital expenditure, reinvestment, operational and maintenance costs (electronic interface, etc...) among the system lifetime [5], [33]–[35]. In [19], [36]–[39] the authors proposed a metric derived from the LCOE and dedicated to ESS applications. The ESS LCOE, also called Levelized Cost of Storage (LCOS), represents the cost of the electricity

delivered by an ESS according to its lifetime estimation. The criteria can be evaluated by including different parameters as in [18], [36], where the authors propose to include in the LCOE the cost of the energy needed to charge the ESS. Moreover, in [38] the authors show that the LCOE clearly depends on the ESS applications. In our case we only focus on the ESS cost, without taking into account the PV LCOE when battery charging and the storage application is the integration of RES and to promote autonomy building.

In our study, we only consider that the batteries have to be replaced every L_B years and the associated reinvestment cost. The LCOE of the HESS is given by (11). It is the sum of the $LCOE_i$ of each ESS, where $Cost_i(y)$ represents the global cost, in euros, of ESS_i for the year y , which depends on L_B .

$$LCOE = \sum_{i=ESS_A}^B LCOE_i \quad (11)$$

$$LCOE_i = \frac{\sum_{y=0}^H \frac{Cost_i(y)}{(1+r)^y}}{\frac{E_{disch_i}(y)}{(1+r)^y}}$$

H is the time horizon for the study, in years, r is the discount rate and E_{disch_i} is the energy delivered on a year by ESS_i , in kWh.

The time horizon chosen for this study is 25 years, considering that corresponds to an average value of SC calendar lifetime, according to data given in [5], [13], [20]. The discount rate can vary from 3.5% to 8% regarding the values proposed in the literature, we selected the worst value of 3.5% as proposed by [18].

V. RESULTS AND DISCUSSION

In order to analyze the interest of SC in building applications, we run the HESS simulation for different values of the energy installed thanks to SC at a fixed value of the HESS energy. We propose to analyze 26 configurations of the HESS, by varying the energy stores in the SC from 0 to 2.5% of global installed energy, by step of 0.1%. Fig 3 presents the power exchanged between the 3 electrical sources (DC Bus, batteries, SC), for 2 HESS configurations (without and with SC for 2.5% of the installed energy) in case of a particularly intermittent PV production day. The batteries SOC for the 2 HESS configurations are presented on this graph, in order to show the benefits of the SC on the batteries operating. This analysis is proposed for the 2 load profiles : lighting and electrical outlets.

The following subsections present the influence of the lifetime estimation method, the load profile and the SC cost on the criterion estimation, in order to validate our chosen methodology before to conclude about SC impacts.

A. Influence of the battery aging model

Figure 4 shows the battery lifetime estimation obtained for the 26 HESS configurations, for the electrical lighting network and electrical outlets power profiles. We can see that SC

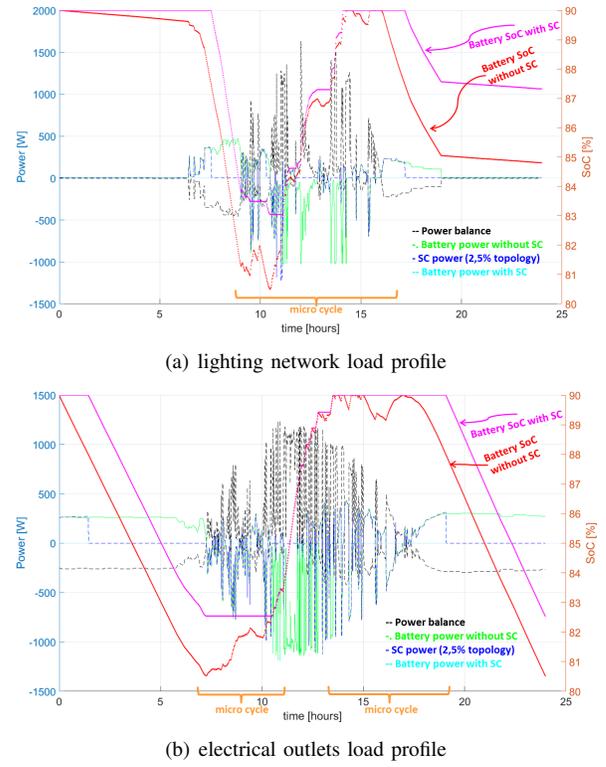


Fig. 3. Power profiles and batteries SoC for lighting network and electrical outlets demands, with and without SC

improves the battery lifetime, regardless the aging method used. The batteries lifetime estimated by the Ah throughput method is higher than the one returned by the rainflow counting. Indeed, as the Ah throughput method doesn't take into account the DoD of each cycle, it overestimates the number of cycles to failure at small DoD. For this reason, we select the rainflow counting method for the rest of the paper.

This results allow us to already say that SC impact study have to consider battery recycling. As a matter of fact, improving battery lifetime allows to increase the sustainability of our system by decreasing the replacement rate.

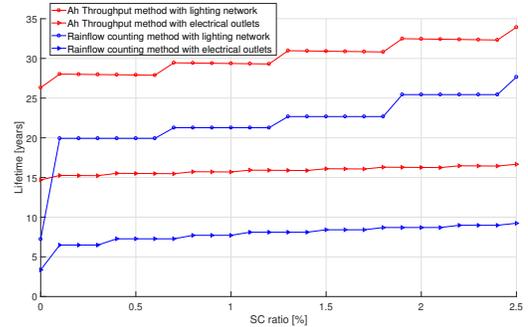


Fig. 4. Battery aging estimation calculated by "Rainflow Counting" and "Ah Throughput" method

In future works it would be interesting to improve the rainflow counting method by taking into account the SoC at

each cycle, but this analyze needs more information from the manufacturers or long experiments.

B. Load profile impact

Figure 5 shows the LCOE of the HESS normalized by the LCOE of configuration with only OPzV battery for different load profiles.

In spite of the SC significantly increases the lifetime of the HESS when the DC bus supplies the lighting network (fig. 4), the LCOE of the HESS never decreases when we increase the percentage of the installed energy store in the SC. The price of the SC installed in our LVDC-MG is the reason of this (table I), and we will show in the next section that this trend could be reversed by using cheaper SC.

When the load of the LVDC-MG are the electrical outlets, the effect of the ESS_A size is more significant. Indeed, when 0.1% of the installed energy is constituted by SC, we note a small decrease of the LCOE. If we look at the P_{bal} profile, we could see more frequent change between charge and discharge compare to the lighting network, which generates more micro-cycles in the HESS. According to the curves presented in figure 4, the SC seems mainly useful in the case of the outlets load power profile.

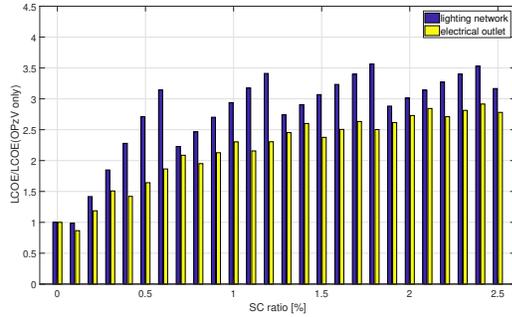


Fig. 5. The LCOE of the HESS normalized by the LCOE of OPzV battery, for the both load power profiles

C. SC cost

With the cost of the SC chosen for our analysis, we can obtain a small decrease of the LCOE when considering the electrical outlets power profile, but in the literature the SC cost varies from 255 to 30000 €/kWh [13], [19], [20]. The figure 6 presents the battery lifetime and the HESS LCOE normalized for three different SC costs : the lowest values (255 €/kWh), the average values of the value found in the literature (17200 €/kWh) and the cost of the SC installed in our LVDC-MG (24000 €/kWh).

We can see that for a price of 255 €/kWh the SC reduce the HESS LCOE. Anyway, adding a small amount of SC to the HESS reduces the global LCOE in case of electrical outlets load profiles. Furthermore, if we consider the ecologic replacement cost of the battery and the development of SC in the future, the proposed HESS could become even more attractive.

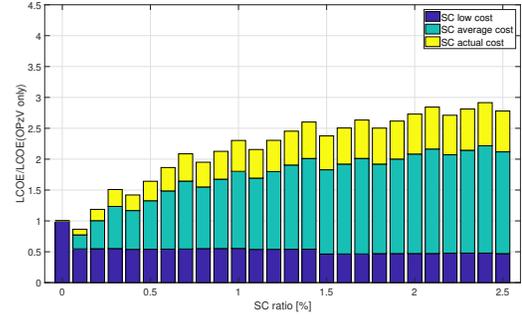


Fig. 6. LCOE for different SC costs in case of the electrical outlets power profile

VI. CONCLUSION

In this paper we proposed a model of LVDC-MG to analyze the impacts of SC combined with OPzV Ld batteries in an HESS. The target application of this work is electrical autonomous BiPV. This model was simulated for different types of LVDC-MG load (lighting and electrical outlets). The power load profiles extracted from ADREAM BiPV DB allow to analyze the impact of SC on batteries lifetime in real operating conditions. This analysis shows the SC have a positive impact on battery lifetime, by avoiding micro-cycles to the batteries. Before to conclude on the benefit of the SC, we analyzed how the battery aging estimation method and the cost of the SC influence the HESS global LCOE. At the end of this study, and according to the model used to simulated our LVDC-MG, we can conclude that adding a small amount of SC to Ld battery could reduces the LCOE of the storage system for some power load profiles. Furthermore, if we consider the ecologic replacement cost of the battery, the supposed that HESS with SC dedicated to stationary storage for autonomous BiPV application becomes even more attractive.

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