



Novel Economic Analysis to Design the Energy Storage Control System of a Remote Islanded Microgrid

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Abstract—This paper presents a novel power flow control strategy, combined with an economic analysis, for an energy management system (EMS) involving a hybrid energy storage. The EMS operates a remote microgrid and directs the power flow to either batteries or supercapacitors to increase the life of the batteries. This paper demonstrates how the use of supercapacitors increases the lifetime of the batteries and ultimately how it affects the economics of the system. The proposed EMS controller also compensates for the 120-Hz ripple on the dc bus. Modeling, simulations, and experimental verification are presented together with the procedure to perform the assessment of the battery lifetime, according to the tuning parameters of the controller.

Index Terms—Battery lifetime, energy management, hybrid energy storage, power converters, supercapacitors (SCs).

I. INTRODUCTION

MICROGRID technology has been developed and closely investigated as one of the solutions to increase energy security. Solid-state power converters are instrumental to microgrid operations [1]. Power-electronics-based energy management systems (EMS) have been recently explored (see [2]–[8]) to control loads and distributed energy resources (DERs), to detect grid failure, and to enable the microgrid islanding mode of operation. Although an EMS is sometimes used to define the software or controller that manages the energy in a power system or microgrid, in this paper, it is used to define a system, which includes one or more power converters that interface to

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a microgrid and different DERs. In addition to the hardware, the EMS includes several layers of control to manage currents and voltages, as well as loads and sources. An EMS has been recently proposed to optimize operations in remote military microgrids, where continuous service to critical power loads is essential [5], [8]. In this paper, we focus on the EMS ability to control the power flow when a hybrid energy storage system (HESS) is added to the architecture because the real load profile, showing sudden peaks, is considered [9]. The goal of the proposed HESS is to divert the 120-Hz ripple and the peak current ripple away from the batteries by adding supercapacitors (SCs) controlled by a buck–boost converter, thus increasing the lifetime expectation of the batteries available in the microgrid. A novel study to tune the controllers parameters is carried out for a remote military microgrid model. Recent publications [10]–[19] have addressed hybrid storage systems with batteries and SCs in different configurations. Some HESS configurations have used different power converter topologies and controllers [10], [11], [15], or they do not show a thorough analysis of the control system [6]. In [20], the focus is exclusively the energy management of a light rail vehicle; therefore, the power electronics controller is not addressed.

Other HESS do not control the 120-Hz ripple on the dc bus either because they are applied to three-phase systems [16] or because they service loads that are not single-phase ac [12], [21]. Papers [17] and [18] present HESS controllers that are very similar to the one proposed in this paper, but they do not include the battery lifetime analysis nor the application to the economics of a microgrid as they are presented in this paper. In [19], although a similar HESS controller was used, the power management strategy used has a different scope than the one proposed in this paper; losses and state of charge (SoC) of the SCs are weighted to be optimized, but the authors do not quantify how this procedure affects the expectation of battery lifetime extension. In [22], an interesting application of optimal power flow problem with the HESS is considered, but time steps are bigger (30 s versus our 0.1 s), and no consideration on investments is included: the economic is based on the cost of the saved energy, thus neglecting the role of sizing, which we show is also a key issue when dealing with the HESS. Here, we evaluate how the different controller strategies, implemented with a proper number of SCs, can increase the battery lifetime.

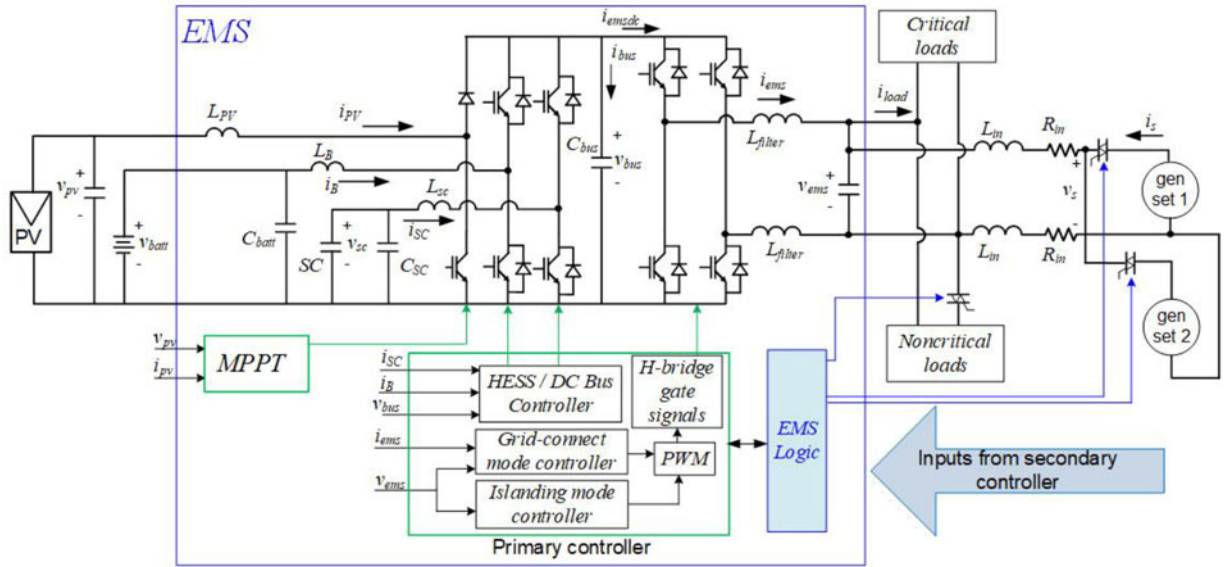


Fig. 1. EMS architecture and its connection to a remote military microgrid with two diesel generators, hybrid energy storage, and PV panels.

75 To the knowledge of these authors, the proposed combination
 76 of a buck–boost converter, a control architecture, and a tight
 77 link with the battery lifetime presented in this paper has not
 78 been previously presented. Furthermore, the application to a re-
 79 mote microgrid introduces peculiarities in the economics of the
 80 considered case study. In this paper, the EMS architecture is pre-
 81 sented in Section II. The proposed HESS controller is described
 82 in Section III. In Section IV, the procedure, involving the bat-
 83 tery lifetime and its link with a specific controller parameter, is
 84 presented for a typical power profile of a remote military micro-
 85 grid. Experimental measurements and conclusions are reported in
 86 Sections V and VI, respectively.

87 II. EMS FUNCTIONALITY AND MICROGRID SETUP

88 A schematic of the EMS' architecture is provided in Fig. 1
 89 together with the remote military microgrid power system. Also
 90 shown in Fig. 1 are critical and noncritical loads and two diesel
 91 generator sets (gensets). Critical loads are those loads, including
 92 computers, radars, and some air conditioning systems, which are
 93 critical to the military operations success and must be powered
 94 at all times. Thus, they are hard wired to the ac power bus,
 95 while noncritical loads are connected to the ac bus through a
 96 solid-state switch controlled by the EMS to enable shedding
 97 when necessary. In this setup, the noncritical loads are grouped
 98 together for ease of laboratory demonstration; however, an EMS
 99 can control multiple noncritical load switches.

100 The EMS consists of five inverter legs, a field-programmable
 101 gate array (FPGA)-based control system, photovoltaic (PV) pan-
 102 els, battery pack, and SCs. Lead–acid batteries are used for the
 103 work presented in this paper; however, any other type of bat-
 104 tery could be used, and this is true as far as the methodology
 105 concerns. However, lead–acid batteries are presently the tech-
 106 nology of choice in remote military camps because they fail
 107 without catching fire, unlike Li-ion batteries. The FPGA-based
 108 controller includes the dc-bus voltage controller, the ac-bus

voltage control during islanding operations, and the EMS ac
 current in the grid-connected mode. This paper focuses on a
 new dc-bus voltage controller, while the ac current and voltage
 control systems are the same as presented in [7]. The FPGA
 also houses the controls for the HESS and the energy manage-
 ment logic such as load scheduling and grid connect/disconnect.
 Two legs of the power module are employed as a single-phase
 bidirectional H-bridge converter, which can be controlled as a
 current source to inject power from the battery pack to the mi-
 crogrid or as a voltage source when the gensets are OFF. The
 third leg of the power module is operated as a bidirectional
 buck–boost converter to either charge the battery bank or draw
 energy from it. The fourth leg of the power module is operated
 as a bidirectional buck–boost converter to either charge the SCs
 or draw energy from them. Batteries and SCs form the HESS,
 which is controlled by the EMS. A fifth inverter leg is used as
 a boost converter that is the interface to the PV panels.

126 III. HESS CONTROL SYSTEM

127 In this section, the HESS control system architecture is pre-
 128 sented, and its functionality is demonstrated with analysis in the
 129 frequency-domain as well as time-domain simulations.

130 A. Controller Architecture and Functionality

131 The dc-bus voltage is held constant by the HESS controller
 132 shown in Fig. 2. In addition to regulating the dc bus, the goal
 133 of this controller is to distribute the load current between the
 134 battery and the SCs. Specifically, the load current is the current
 135 that the EMS injects into the ac bus to supplement the power
 136 provided by the generators. The peak current demanded by the
 137 loads is provided by the SCs instead of the battery to reduce the
 138 ac current stress on the battery. The low-pass filter commands
 139 the battery current to be absent of abrupt changes. The bandpass
 140 filter (BPF) is added to extract the 120-Hz signal, which a

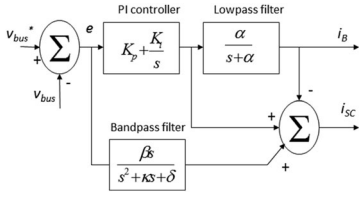


Fig. 2. HESS control system.

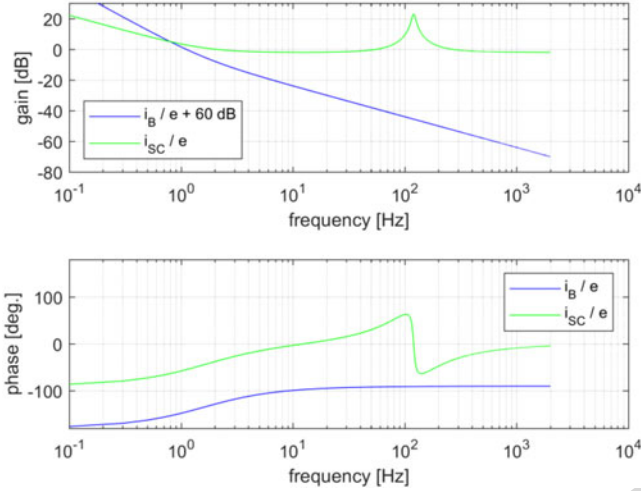


Fig. 3. Bode plots of battery current and SC current transfer functions.

141 frequency equal to twice the 60-Hz output frequency. The goal
142 of the BPF is to reduce the second harmonic voltage ripple on
143 the dc bus and ac current in the battery. The BPF is analyzed in
144 the next section.

145 From the control diagram in Fig. 2, the transfer function of
146 the battery current over the dc voltage error can be derived as

$$\frac{i_B}{e} = \frac{\alpha(sK_p + K_i)}{s(s + \alpha)}. \quad (1)$$

147 The dc voltage error leads to an SC current that is

$$\frac{i_{SC}}{e} = \frac{(sK_p + K_i)}{s} - \frac{\alpha(sK_p + K_i)}{s(s + \alpha)} + \frac{\beta s}{s^2 + \kappa s + \delta}. \quad (2)$$

148 The Bode plots of the transfer functions (1) and (2) are shown
149 in Fig. 3. The battery current i_B contains only low-frequency
150 components, while the SCs provide the current for any ac distur-
151 bances in the dc bus, especially at 120 Hz. In fact, the top
152 plot in Fig. 3 shows that the gain of the transfer function (2) is
153 high at 120 Hz. The parameters used for this analysis are shown
154 in Table I.

155 B. Contribution and Analysis of the BPF

156 It is well known that a 120-Hz component is drawn from the
157 dc bus by a single-phase inverter delivering 60-Hz ac power,
158 so that pulsating power flows from the dc bus in addition to dc
159 power. The role of the BPF in Fig. 3 is to ensure that the 120-Hz
160 current is drawn from the SCs, not the batteries. The following
161 analysis clarifies the contribution of the BPF.

162 The equivalent circuit in Fig. 4 represents the currents that
163 are flowing to/from the EMS dc bus, which are the SC current

TABLE I
PARAMETERS OF THE CONTROLLER

Description	Symbol	Value
DC bus capacitor	C_{bus}	100 μ F
Proportional gain	K_p	0.8
Integral gain	K_i	8
BPF coefficient	β	754
BPF coefficient	κ	57
BPF coefficient	δ	568,000
Low-pass filter coefficient	α	0.005 rad/s

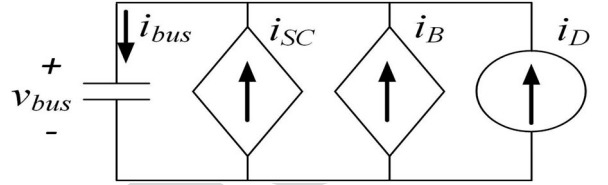


Fig. 4. EMS equivalent circuit for analysis and simulations.

i_{SC} , the battery current i_B , and a disturbance current i_D , which
is equal to i_{emsc} in Fig. 1.

For the equivalent circuit in Fig. 4, the following equations
can be written using basic circuit analysis:

$$i_{SC} + i_B + i_D = sCv_{bus}. \quad (3)$$

First, let us analyze the HESS controller, when there is no
BPF. This is accomplished by removing the BPF block from
Fig. 3 and then rewriting (3) by substituting i_{SC}

$$(v_{bus}^* - v_{bus}) = \left(K_p + \frac{K_i}{s}\right) - i_B + i_B + i_D. \quad (4)$$

From (4), when the disturbance i_D is zero, the dc voltage
transfer function can be derived

$$\frac{v_{bus}}{v_{bus}^*} \Big|_{i_D=0} = \frac{sK_p + K_i}{s^2C + sK_p + K_i}. \quad (5)$$

The dc voltage due to the disturbance current, when the refer-
ence dc-bus voltage is zero, is

$$\frac{v_{bus}}{i_D} \Big|_{v_{bus}^*=0} = \frac{s}{s^2C + sK_p + K_i}. \quad (6)$$

The dc-bus voltage transfer function v_{bus}/v_{bus}^* and the dc-bus
voltage transfer function with the disturbance current as the
input v_{bus}/i_D are plotted in Fig. 5.

The gain for v_{bus}/i_D is greater than 1, which will cause a
lot of ripple if there is any ac current present in the disturbance
current. The second harmonic current flowing from the dc bus to
the ac load is significant and causes dc-bus ripple. This problem
can be mitigated by adding a BPF to identify and close a control
loop on the second harmonic voltage error.

In contrast to the previous analysis, let us analyze the com-
plete HESS shown in the block diagram of Fig. 3. With the BPF,
the dc voltage transfer function becomes

$$\frac{v_{bus}}{v_{bus}^*} \Big|_{i_D=0} = \frac{s^3K_p + s^2K_2 + sK_1 + \delta K_i}{s^4C + s^3K_3 + s^2(K_2 + \delta C) + sK_1 + \delta K_i} \quad (7)$$

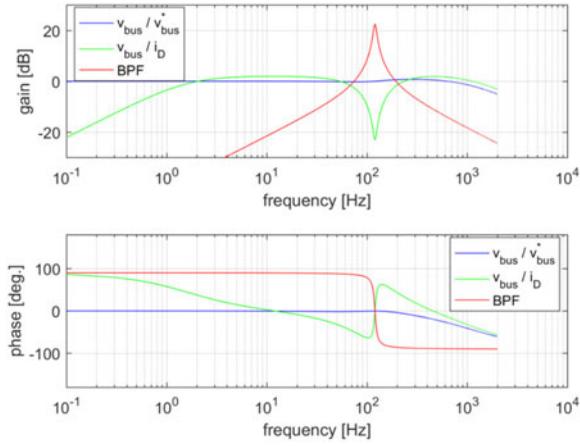


Fig. 5. Bode plots of the transfer functions (5) and (6) without the BPF.

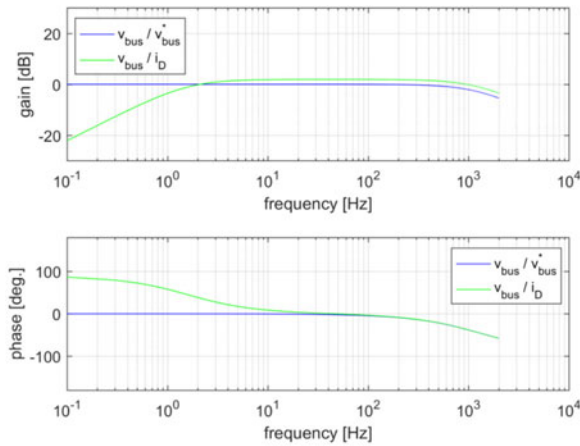


Fig. 6. Bode plots of the transfer functions (7) and (8) with the BPF.

187 and the dc voltage due to the disturbance current is

$$\frac{v_{bus}}{i_D} \Big|_{v_{bus}^* = 0} = \frac{s^3 + \kappa s^2 + \delta s}{s^4 C + s^3 K_3 + s^2 (K_2 + \delta C) + s K_1 + \delta K_i} \quad (8)$$

188 The coefficients for the transfer functions (7) and (8) are

$$K_1 = \delta K_p + \delta K_i \quad (9)$$

$$K_2 = \kappa K_p + \delta K_i + \beta \quad (10)$$

$$K_3 = \delta K_p + \kappa C. \quad (11)$$

189 Fig. 6 shows the Bode plots of the transfer functions (7) and
 190 (8), where the BPF was added to the control architecture. In
 191 contrast with Fig. 5, it can be observed that the addition of the
 192 BPF reduces the second harmonic voltage ripple in the dc bus.
 193 The BPF has very high gain at 120 Hz, and it reduces the transfer
 194 function v_{bus}/i_D significantly at 120 Hz. The BPF has a minimal
 195 effect on the transfer function v_{bus}/v_{bus}^* .

196 A time-domain simulation of the system represented in Fig. 1
 197 is shown in Fig. 7. The 120-Hz component of the dc-bus ripple is
 198 reduced by the BPF. Also, as shown in Fig. 7, the step response
 199 to an increase in the dc load current is reduced. The disturbance
 200 current used in the simulation of Fig. 7 is a 10-A sinewave at

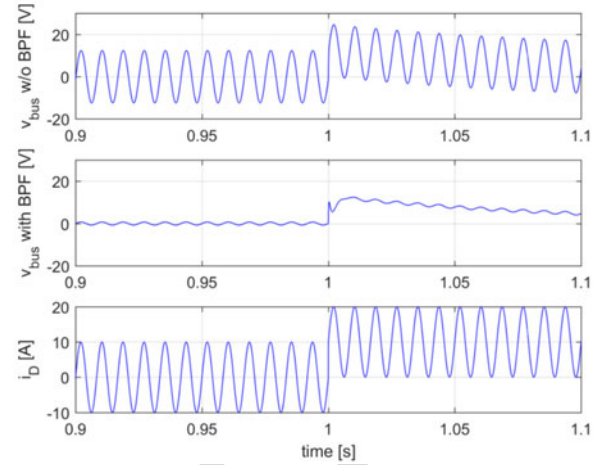


Fig. 7. Time-domain behavior of the HESS controller with and without the BPF for an injected disturbance current i_D (v_{bus} error is shown here).

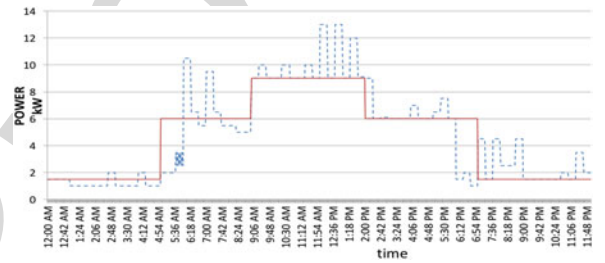


Fig. 8. Load power consumption profile of a remote microgrid, real (DASHED) versus linearized (SOLID) (2-min resolution).

120 Hz plus a 10-A step change in the load current at $t = 1$ s 201

$$i_D(t) = 10u(t-1) + 10\sin(2\pi 60t). \quad (12)$$

IV. LIFETIME EXTENSION AND ECONOMICS OF AN OPTIMIZED HESS 202

203
 204 In this section, the proposed HESS is used in a remote military
 205 microgrid to demonstrate how the above control increases the
 206 battery lifetime compared to the same microgrid, where only
 207 batteries are used for energy storage. A HESS shows its potential
 208 when sudden spikes, not negligible because of the same order of
 209 magnitude than the base load, occur. The analysis in this section
 210 proves that, when the HESS draws the load transient currents
 211 from the SCs, the batteries will last longer. The battery lifetime
 212 extension is quantified for different values of the low-pass filter
 213 coefficient, and the overall microgrid economics is analyzed.

214 The power profile of Fig. 8 (dashed line) represents the typical
 215 daily consumption in a remote military microgrid, where sudden
 216 peaks occur and seriously affect the lifetime of the batteries.
 217 This profile is used for the following analysis and case study
 218 in contrast to the simplified profile (solid) also plotted in Fig. 8
 219 and used for the study reported in [8].

A. A Few Considerations on the Role of the Optimization 220

221 Fuel consumption is one of the parameters that are worth
 222 minimizing in a remote military microgrid because fuel

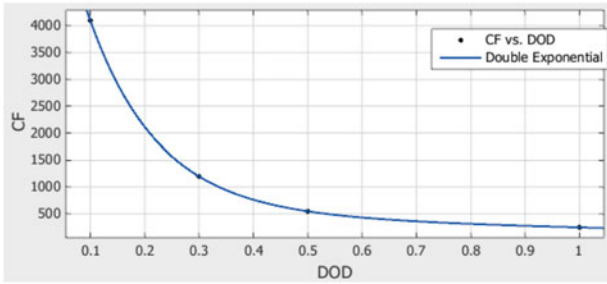


Fig. 9. Reference fitting curve for the lead-acid batteries of the experimental test (CF versus DoD) [26].

transportation to remote sites can result in casualties. In a previous study, the optimization model and its constraints were thoroughly discussed [23]. The results of that optimization are based on 2-min intervals, and it provides the rules for the power sharing among the various sources, taking into account how fast the response from SCs can arrive. These sources include two diesel generators (5 and 15 kW), the PV source (3 kW_P, which is deterministic in the proposed example), and the HESS. In addition, the optimization algorithm ensures that the batteries operate within safe SoC limits, and the generators operate within their range of operation and efficiency. The power associated with HESS, P_{HESS} , is thus obtained and is being used in this novel analysis, where the focus is the evaluation of the lifetime of the batteries and the economics of the system, when the controller parameter α varies.

B. Link Between the Controller and the Battery Lifetime

Different battery and SC currents can be obtained by changing the low-pass filter coefficient α , still keeping P_{HESS} constant. With these currents as inputs, we can evaluate the SoC for both devices and find which is the best SoC* to support the optimized rules

$$\text{SoC}_*(t) = \text{SoC}_*(t-1) - \frac{P_*(t)\Delta t}{\text{ASE}_*}. \quad (13)$$

With SoC*, P_* , and ASE*, we identify the SoC, the active power (positive when storage is feeding the load, and negative when is charging), and the storage capacity of each specific device (either the battery or the SCs)

$$P_*(t) = P_{\text{BAT}}(t) = v_{\text{bus}}(t) \cdot i_B(t) \text{ but also}$$

$$P_*(t) = P_{\text{SC}}(t) = v_{\text{bus}}(t) \cdot i_{\text{SC}}(t) \quad (14)$$

$$P_{\text{HESS}}(t) = P_{\text{BAT}}(t) + P_{\text{SC}}(t). \quad (15)$$

The currents i_B and i_{SC} must have the same sign, or being 0, meaning that when one device is charging or discharging, the other must act accordingly or it must be OFF.

The battery lifetime is thus assessed by using the Rainflow counting method [24], [25], which needs the results of the SoC over time to provide the number and typology of cycles characterizing the charge and discharge of the battery over a typical horizon. Each kind of battery shows its own cycle to failure (CF) versus depth of discharge (DoD). In Fig. 9, such data for the lead-acid batteries used in the laboratory prototype are

reported. We recall that the use of lead-acid batteries is due to safety reasons. Nevertheless, this methodology applies to any kind of battery technology, as long as the CF versus DoD curve can be obtained.

The Rainflow counting algorithm provides information on amplitude (related to the DoD) and frequency of cycles presenting the same amplitude on a set time horizon. The life expectancy of the battery is related to the CF, with $1/\text{CF}$ being the life fraction. We can assess D , the inverse of the lifetime, as

$$D = \sum_{i=1}^m \frac{N_i}{\text{CF}_i} \quad (16)$$

where m is the number of different DoD_{*i*}, occurring in the set horizon, N_i is the frequency associated with DoD_{*i*}, and CF_{*i*} is the corresponding number of cycles at DoD_{*i*}. For a fully functional battery, D has to be less than 1. When $D = 1$, the battery is considered dead; its unit measure depends on how the number of cycles N_i is counted: if N_i are counted over a day, then the lifetime of the battery (inverse of D) counts the days to failure (DF).

An exemplification: if in a typical day, a battery experiences 10 cycles/day (N), where DoD (the amplitude of the equivalent charge/discharge cycle) is equal to 0.2, then that battery can ideally survive for up to 200 equivalent days, before being considered dead. In fact, $\text{CF}_{\text{DoD}=0.2}$ is 2000; hence, the lifetime in days $1/D = \text{CF}/N = 200$. When multiple cycles occur, the lifetime is the composition of each single assessment. D depends on N , which relates to DoD; DoD depends on SoC_{BAT} and SoC depends on the low-pass filter coefficient α ; thus, D depends on α .

To sum up the analysis: the higher α , the less current on SC; thus, the lower the lifetime of the battery. To achieve a certain lifetime, we tune the α value, accordingly.

The overall implemented procedure ensures minimum fuel consumption, while suitably tuning the battery lifetime, at the same time. This last objective is achieved by tuning the HESS controller. In the following subsection, we will show how α will also affect the HESS investments and its economics.

C. Case Study Results

In Fig. 10, input and output data, from the optimization procedure described in [23], are reported for the case study: a remote military microgrid. Case A is the reference case when storage is made up only by batteries (no SCs), while case B represents the case when the HESS is present (with SCs). The needed data for both cases, regarding the features of the optimized considered system, deal with the battery, the gensets, and the load profile on a set horizon. In particular, for the battery, the parameters are the following:

- 1) SoC_{min} and SoC_{max};
- 2) charging/discharging efficiency η ;
- 3) rated power P_{max} ;
- 4) discharging/charging time at P_{max} ;
- 5) available capacity.

For gensets 1 and 2, the parameters are the rated powers P_{n1} and P_{n2} and the related relationships between the load factor

	Description	Case B	Case A (Reference)	unit measure	note or symbol
battery	state of charge: range	0.5-1	0.2-1	p.u	SoC min-max values
		SoC(t=0)=SoC(t=end of the day)			typical day assumption
	charge/discharge efficiency	0.9/0.9	0.9/0.9	p.u.	
	maximum available constant battery power (rated)	3	3	kW	Pbat max
	duration time @ Pbat max	6	6	hours	T
	Battery capacity	18	18	kWh	ASE
genset	rated power	5-15	5-15	kW	P1n, P2n
	a total of 720 t per typical day	2	2	minutes	t step
general parameters	Load in a typical day	Fig.8	Fig. 8	W	PL
	power from/to the storage	Phess = 200*(iB+ iSC)	Pbat= 200*iB	W	
results	power shares among sources	Fig.14	Fig.11		

Fig. 10. Matrix visualization of all the most important input and output information coming from the previous optimization.

310 and the fuel consumption [8]. Furthermore, the time step t and
311 the load profile P_L shall be established.

312 Also, Fig. 10 reports in the last row the indication to the result-
313 ing output, the balancing of powers to feed the load, meaning the
314 sequence of powers from gensets and to/from the storage unit.

315 Additionally, for case B, in Table II, technical data of the used
316 basic SC module are reported.

317 When the DASHED load profile of Fig. 8 is considered for a
318 typical day, the optimized procedure identifies the best $P_{HESS}(t)$
319 (or P_{BAT} if no SCs are present) for each time step of the day and
320 for the given conditions.

321 For case B, we consider three scenarios, cases B1, B2, and B3,
322 identified by different values of the low-pass filter coefficient α ,
323 equal to 0.005 (B1), 0.003 (B2), and 0.001 (B3). Once the opti-
324 mization has produced the power share among the gensets and
325 the storage, then different alphas determine a different sequence
326 for $i_B^{B1, B2, B3}(t)$ and $i_{SC}^{B1, B2, B3}(t)$ and thus $SoC_B^{B1, B2, B3}(t)$: su-
327 perscripts identify the respective scenarios. Similarly, when no
328 SCs are considered, then we will have $i_B^A(t)$ and $SoC^A(t)$.

TABLE II
SC MAXWELL DATA [27]

Product name	BMOD0130P056 B03
Rated Capacity (F)	130
Rated Voltage (V)	56
ESR (mΩ)	8.1
Leakage current (mA)	120
Absolute maximum current (A)	1,900
max continuous current (A)	61 ÷ 99
Weight (kg)	18
Stored Energy (Wh)	56.6
$P_{SC_{max}}$ (kW)	96.79
$t @ P_{SC_{max}}$ (s)	2.11
cost of single unit (\$)	1,300

TABLE III
CASE STUDY: MAIN RESULTS FOR THE THREE SCENARIOS (B1, B2, AND B3) AND THE REFERENCE CASE A

Description ↓	$\alpha \rightarrow$	0.005	0.003	0.001	Ref.
		B1	B2	B3	case A
SCs capacitance	F	650	910	1300	no SC
set SoC_{min}		0.5	0.5	0.5	0.2
The lowest SoC	%	69.9	67.53	67.06	30
Cycles		20	18	16	51
Lifetime	days	274	282	363	122
Invest. on SC	k\$	6.5	9.1	13	N.S.
Invest. on BAT.	k\$	10.8	10.8	9.	25.2
Tot. Inv. (5y)	k\$	17.3	19.9	22.	25.2

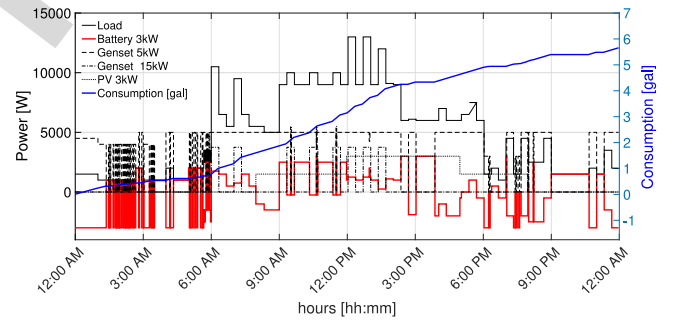


Fig. 11. Case A (only with batteries): from the optimization [23]: power profile and consumption (resolution step 2 min; $SoC_{min} = 0.2$).

329 For each of the four SoC sequences, a new series of
330 $DoD^{B1, B2, B3, A}$ is derived, and different lifetimes are expected.

331 In Table III, the main results are reported for the three scenar-
332 ios [increasing SCs number from 5 (B1) to 10 (B3)], after the
333 optimization and the tuning of α , as well as for the Reference
334 case A, optimized but without SCs.

335 The investment (INV^i) in each i th scenario/case is thus eval-
336 uated as in the following, depending on the DF of the batteries,
337 which ultimately depends on alpha:

$$INV^i(DF(\alpha)) = INV_{SC}^i + \sum_{DF} INV_B^i(DF(\alpha)). \quad (17)$$

338 The change in i_B and i_{SC} sequences can be visualized when
339 simulating the battery current with and without SCs. In Figs. 11

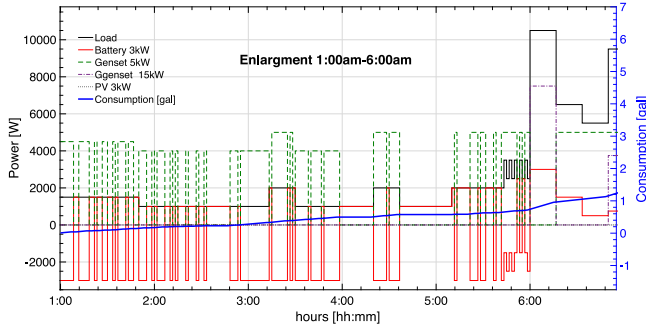


Fig. 12. Case A: Enlargement between 1:00 A.M. and 6:00 A.M. (resolution step 2 min; $SoC_{min} = 0.2$).

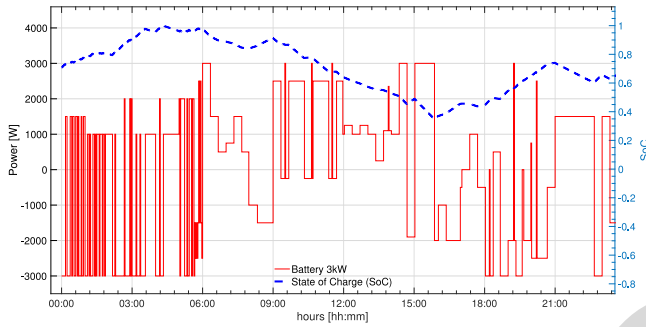


Fig. 13. Case A: Optimal P_{BAT} and consequent SoC on batteries (resolution step 2 min; $SoC_{min} = 0.2$).

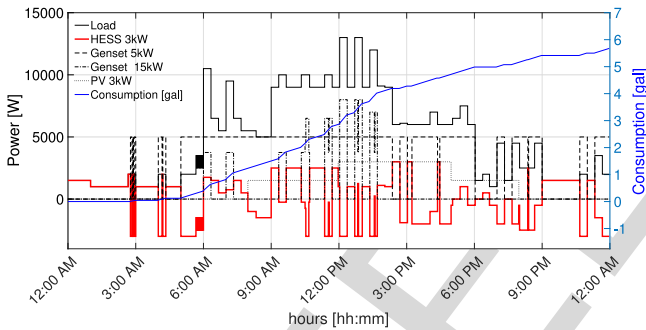


Fig. 14. Case B (with the HESS): from the optimization [23]: power profile and consumption (resolution step 2 min; $SoC_{min} = 0.5$).

340 and 14, the optimization results are reported for case A (Ref-
 341 erence) and for case B: what is referred as HESS profile is the
 342 active power associated with what is coming from/to the storage
 343 unit (no transient considered, resolution step 2 min).

344 For a better understanding of the battery dynamic, in terms of
 345 charging/discharging cycles and consequent SoC, Figs. 12 and
 346 13 for case A and Figs. 15 and 16 for case B are reported.

347 They show an enlargement of Figs. 11 and 14, respectively,
 348 when SCs are not included (case A) and when they are included
 349 (case B).

350 A Simulink model produced the simulated plots in Figs. 17–
 351 19. Omitting the switching behavior of the EMS power
 352 converters lead to shorter simulation times for the battery
 353 current over a 24-h period (resolution 0.1 s). In Fig 17, the
 354 battery current is plotted (upper), when yet the SCs are to be

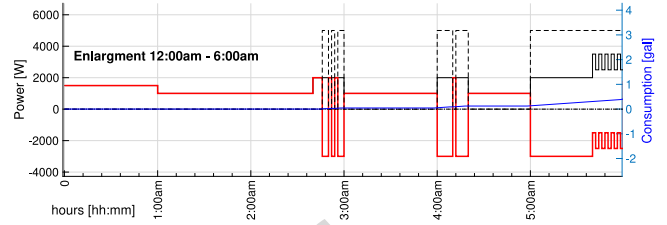


Fig. 15. Case B: Enlargement between 1:00 A.M. and 6:00 A.M. (resolution step 2 min; $SoC_{min} = 0.5$).

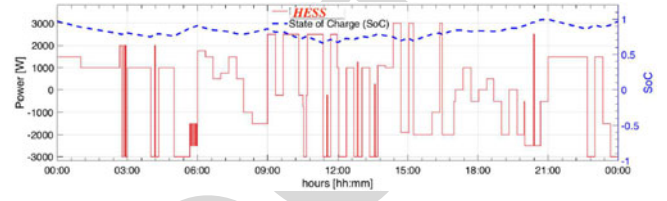


Fig. 16. Case B: Optimal P_{HESS} and consequent SoC on batteries (resolution step 2 min; $SoC_{min} = 0.5$).

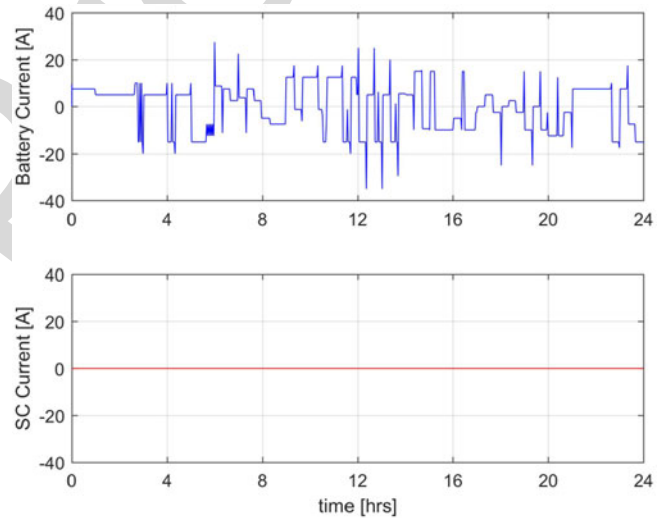


Fig. 17. Case B: battery current when the HESS control system is in place but is disabled (with transient, resolution step 0.1 s). SC current is zero.

355 turned ON, the transient is considered, and the resolution step
 356 is 0.1 s. In Figs. 18 and 19, battery and SC currents are shown,
 357 respectively, when the HESS controller is operational with the
 358 low-pass filter coefficient $\alpha = 0.005$ (B1) and $\alpha = 0.001$ (B3).
 359 It can be observed that the battery current is much smoother
 360 when the HESS controller is used to redirect the peak currents
 361 to the SCs, and we can also notice how the α value affects the
 362 i_B profile (upper graph of Figs. 17–19).

363 In Fig. 20 (Scenario B1) and Fig. 21 (Scenario B3), the battery
 364 cycles are reported for α equal to 0.005 (smaller SCs) and 0.001
 365 (bigger SCs). The main results are reported in Table III, where
 366 the increase in DF (122 estimated days with no SCs, 274 for
 367 $\alpha = 0.005$ up to 363 for $\alpha = 0.001$), the assessment of the lowest
 368 SoC, and investments are assessed with respect to the illustrated
 369 procedure.

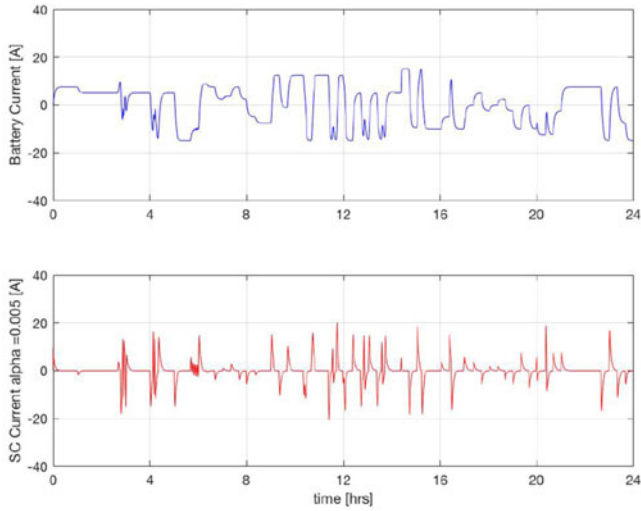


Fig. 18. Scenario B1: battery and SC current when the HESS control system is enable. SCs takes the peaks of the load current ($\alpha = .005$, resolution step 0.1 s).

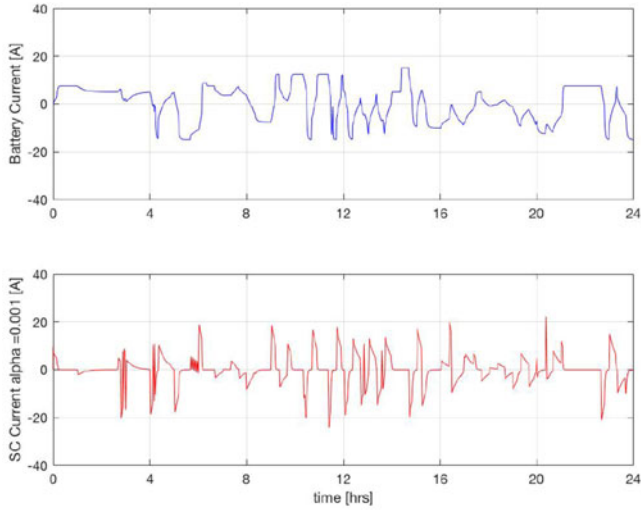


Fig. 19. Scenario B3: battery and SC current with the HESS control system. SCs take the peaks of the load current ($\alpha = 0.001$, resolution step 0.1 s).

370 The plots in Fig. 22 and the results in Table III demonstrate
 371 how the battery lifetime is extended when the HESS controller
 372 is used, realizing the least investment over five years, when α
 373 is lower, thus finding the suitable tradeoff between increasing
 374 SCs size and the battery wearing. We can also notice that
 375 below ~ 900 operating days, even only five SCs modules are
 376 not convenient against batteries, but above 900 days, the HESS
 377 becomes cost effective. Over ~ 1500 days, every investigated
 378 HESS is more cost effective than batteries alone.

379 Depending on the size of the SC and batteries, thus on the
 380 deriving cycles to failure, we can infer that the daily power
 381 consumption is a key parameter for the economic evaluation.
 382 Therefore, careful microgrid load analysis should be done to
 383 create a reliable load profile. A sensitivity analysis can also be
 384 performed to identify not only the actual optimum, but also the
 385 proper range of validity for the current assessment and link it
 386 to the controller parameters. This will be illustrated in a future
 387 work.

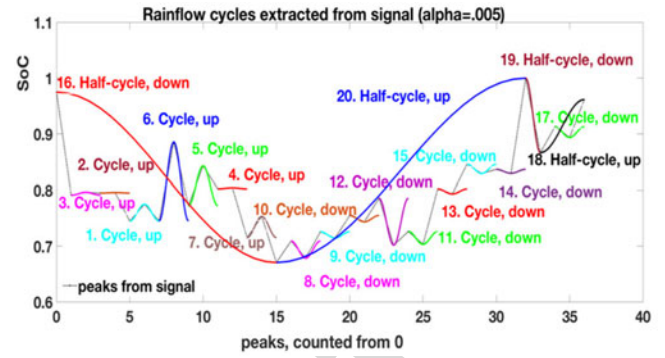


Fig. 20. Scenario B1: Counting cycles at different DoD ($\alpha = 0.005$).

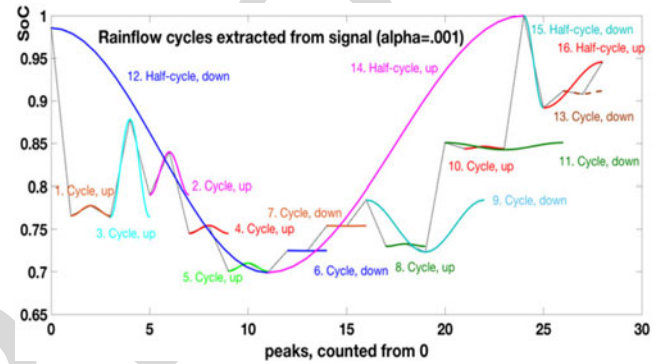


Fig. 21. Scenario B3: Counting cycles at different DoD ($\alpha = 0.001$).

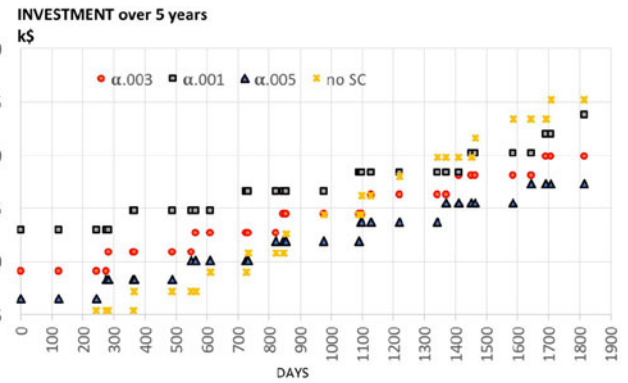


Fig. 22. Investment over five years ($\alpha = 0.005$ blue triangle; $\alpha = 0.003$ orange circle; $\alpha = 0.001$ gray square; yellow star = Reference—NO SCs).

388 Our methodology makes easily evident how those battery
 389 technologies with higher CF versus DoD (for instance, the
 390 lithium ones) can positively affect the lifetime assessment be-
 391 cause higher CF values directly influence (16). On the other
 392 hand, they cost more; thus, again, another sensitivity analysis,
 393 focusing on prices, can help in investigating how far our con-
 394 siderations can be stretched.

V. EXPERIMENTAL MEASUREMENTS

395
 396 To verify the functionality of the proposed HESS controller, a
 397 laboratory experiment was conducted with a scaled-down EMS
 398 prototype. The laboratory setup is represented in Fig. 23, where
 399 the EMS is included inside the blue box. Note that instead of

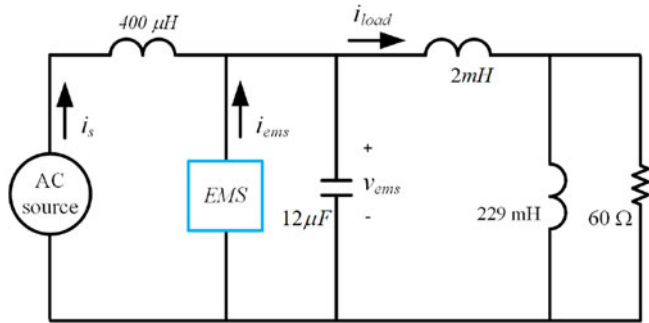


Fig. 23. Schematic of the laboratory setup for the experimental validation.

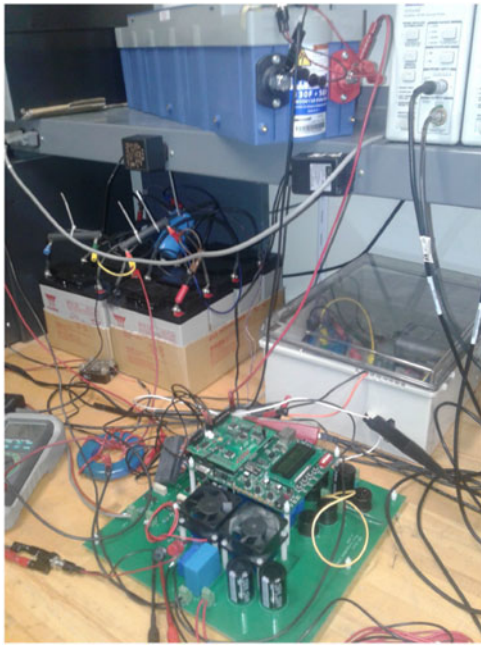


Fig. 24. Photograph of the laboratory setup.

400 a diesel generator, the ac 120-V 60-Hz power available in the
401 laboratory was used.

402 A photograph of the prototype on the laboratory bench is
403 shown in Fig. 24. A 130-F Maxwell SC [27] and six Genesis
404 12-V lead-acid batteries [26] connected in series are visible
405 in the photograph, together with the EMS hardware, which in-
406 cludes several printed circuit boards (PCBs) and external pas-
407 sive components. The EMS controller is embedded on an FPGA,
408 which is part of a Xilinx developed board [28]. The other PCBs
409 are custom made, with the bottom one comprising the power
410 electronics and passive components and the top PCB includ-
411 ing A/D converters, USB interface to communicate with a per-
412 sonal computer, and other electronic components that interface
413 with the FPGA board. The code for the FPGA is developed
414 in Simulink with the additional Xilinx System Generator [29]
415 blocks. Further details of the EMS hardware and FPGA software
416 implementation are available in [7] and [9].

417 The first set of experiments produced the steady-state plots in
418 Figs. 25 and 26 with an without the proposed HESS controller,

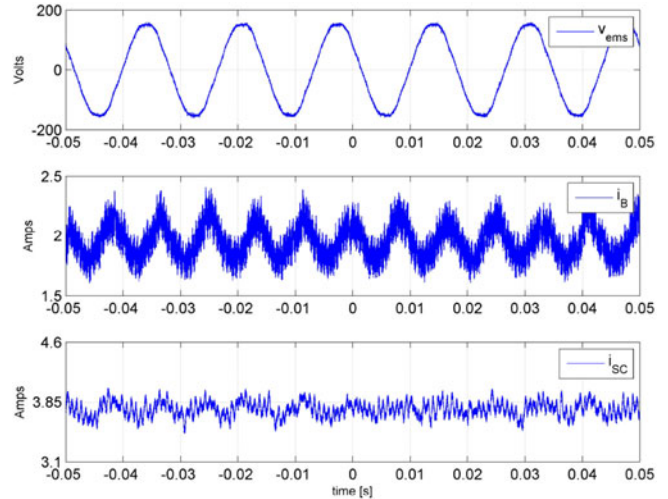


Fig. 25. Case B without HESS controller steady-state experimental waveforms. From the top: ac voltage, battery current, and SC current.

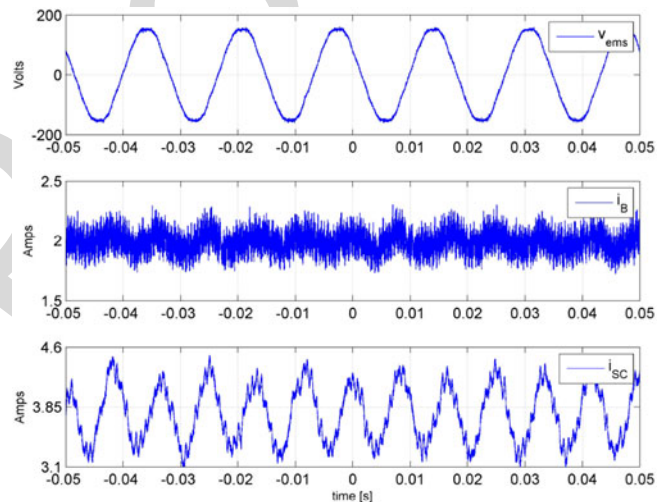


Fig. 26. Case B with HESS controller steady-state experimental waveforms. From the top: ac voltage, battery current, and SC current.

419 respectively. The two figures include, from the top, the ac source
420 voltage, the battery current, and the SC current. The contrast
421 between the battery current in Fig. 25 and the battery current
422 in Fig. 26 validates the effectiveness of the HESS controller in
423 removing the 120-Hz frequency component from the battery and
424 sending it to the SC. Harmonic analysis of the battery current
425 from Fig. 25 shows that the amplitude of the 120-Hz harmonic
426 is 158 mA. In contrast, the 120-Hz harmonic of i_B in Fig. 26 is
427 45.5 mA, a substantial reduction.

428 A second set of experiments is shown in Figs. 27 and 28,
429 where the dynamic performance of the system is contrasted
430 without and with the HESS controller, respectively. As discussed
431 in previous sections, in order to reduce the charge and discharge
432 cycles on the batteries, the SC is commanded to absorb or deliver
433 currents that are suddenly needed by the microgrid. One example
434 is just before 3:00 P.M. (or 15:00 hours; see Fig. 14); when a
435 large amount of energy is being sent to the HESS and, as shown
436 in Fig. 18, the SC absorbs the initial peak. Fig. 27 demonstrates

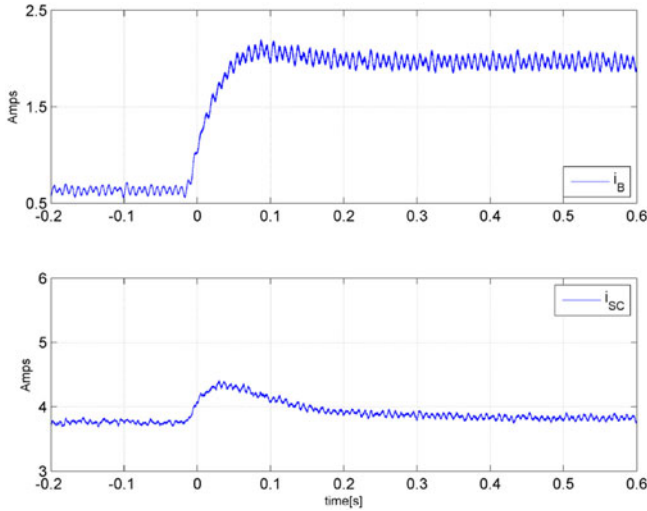


Fig. 27. Battery charge current increase from 1 to 2 A without the HESS controller. Upper i_B and lower i_{SC} .

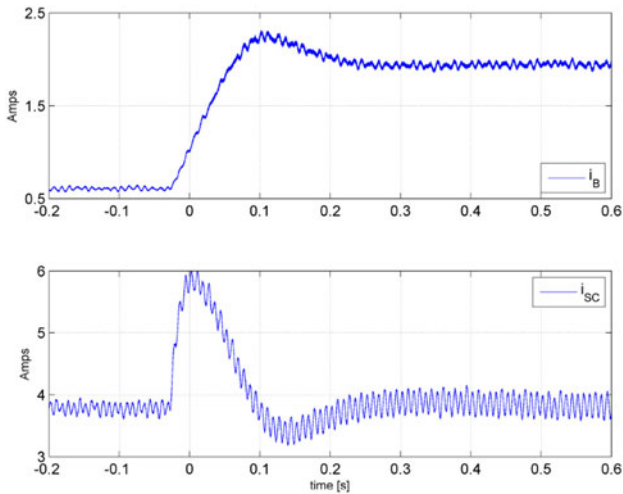


Fig. 28. Battery charge current increase from 1 to 2 A with the HESS controller. Upper i_B and lower i_{SC} .

437 what happens when the current sent to the battery is doubled
 438 from 1 to 2 A: the surge is evident in the battery current, i_B , as
 439 as well as the 120-Hz ripple. In contrast, the di/dt on the battery
 440 current i_B is reduced when the HESS controller is turned ON,
 441 and also, its ripple is noticeably reduced. Note that $\alpha = 5$ Hz
 442 for this experiment.

443

VI. CONCLUSION

444 This paper presents a novel HESS controller focused on
 445 increasing the lifetime of the batteries by using SCs with a
 446 buck-boost converter to control their charge and discharge, thus
 447 maximizing their utilization. A realistic load profile is used, and
 448 several scenarios are compared to link the controller parameter
 449 α with the battery lifetime extension and to the economic eval-
 450 uations. Therefore, the economic evaluation is performed on a
 451 five-year period, time needed to show when the HESS may be-
 452 come cost effective for the case study. The SCs are sized to take
 453 the stress off the load power transients from the battery pack,

so that the batteries only “see” an idealized load profile and can
 perform at better conditions.

Experimental measurements demonstrate the ability of the
 proposed HESS controller to suppress the 120-Hz ripple from
 the battery as well as reduce the di/dt when higher currents
 are commanded. This result proves that the HESS controller
 redirects higher frequency currents to the SC and leave for the
 batteries only slow current changes in order to increase the
 battery lifetime.

Future work will focus on optimizing the number of SCs in
 order to reduce their economic impact on the microgrid.

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645

Novel Economic Analysis to Design the Energy Storage Control System of a Remote Islanded Microgrid

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Abstract—This paper presents a novel power flow control strategy, combined with an economic analysis, for an energy management system (EMS) involving a hybrid energy storage. The EMS operates a remote microgrid and directs the power flow to either batteries or supercapacitors to increase the life of the batteries. This paper demonstrates how the use of supercapacitors increases the lifetime of the batteries and ultimately how it affects the economics of the system. The proposed EMS controller also compensates for the 120-Hz ripple on the dc bus. Modeling, simulations, and experimental verification are presented together with the procedure to perform the assessment of the battery lifetime, according to the tuning parameters of the controller.

Index Terms—Battery lifetime, energy management, hybrid energy storage, power converters, supercapacitors (SCs).

I. INTRODUCTION

MICROGRID technology has been developed and closely investigated as one of the solutions to increase energy security. Solid-state power converters are instrumental to microgrid operations [1]. Power-electronics-based energy management systems (EMS) have been recently explored (see [2]–[8]) to control loads and distributed energy resources (DERs), to detect grid failure, and to enable the microgrid islanding mode of operation. Although an EMS is sometimes used to define the software or controller that manages the energy in a power system or microgrid, in this paper, it is used to define a system, which includes one or more power converters that interface to

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a microgrid and different DERs. In addition to the hardware, the EMS includes several layers of control to manage currents and voltages, as well as loads and sources. An EMS has been recently proposed to optimize operations in remote military microgrids, where continuous service to critical power loads is essential [5], [8]. In this paper, we focus on the EMS ability to control the power flow when a hybrid energy storage system (HESS) is added to the architecture because the real load profile, showing sudden peaks, is considered [9]. The goal of the proposed HESS is to divert the 120-Hz ripple and the peak current ripple away from the batteries by adding supercapacitors (SCs) controlled by a buck–boost converter, thus increasing the lifetime expectation of the batteries available in the microgrid. A novel study to tune the controllers parameters is carried out for a remote military microgrid model. Recent publications [10]–[19] have addressed hybrid storage systems with batteries and SCs in different configurations. Some HESS configurations have used different power converter topologies and controllers [10], [11], [15], or they do not show a thorough analysis of the control system [6]. In [20], the focus is exclusively the energy management of a light rail vehicle; therefore, the power electronics controller is not addressed.

Other HESS do not control the 120-Hz ripple on the dc bus either because they are applied to three-phase systems [16] or because they service loads that are not single-phase ac [12], [21]. Papers [17] and [18] present HESS controllers that are very similar to the one proposed in this paper, but they do not include the battery lifetime analysis nor the application to the economics of a microgrid as they are presented in this paper. In [19], although a similar HESS controller was used, the power management strategy used has a different scope than the one proposed in this paper; losses and state of charge (SoC) of the SCs are weighted to be optimized, but the authors do not quantify how this procedure affects the expectation of battery lifetime extension. In [22], an interesting application of optimal power flow problem with the HESS is considered, but time steps are bigger (30 s versus our 0.1 s), and no consideration on investments is included: the economic is based on the cost of the saved energy, thus neglecting the role of sizing, which we show is also a key issue when dealing with the HESS. Here, we evaluate how the different controller strategies, implemented with a proper number of SCs, can increase the battery lifetime.

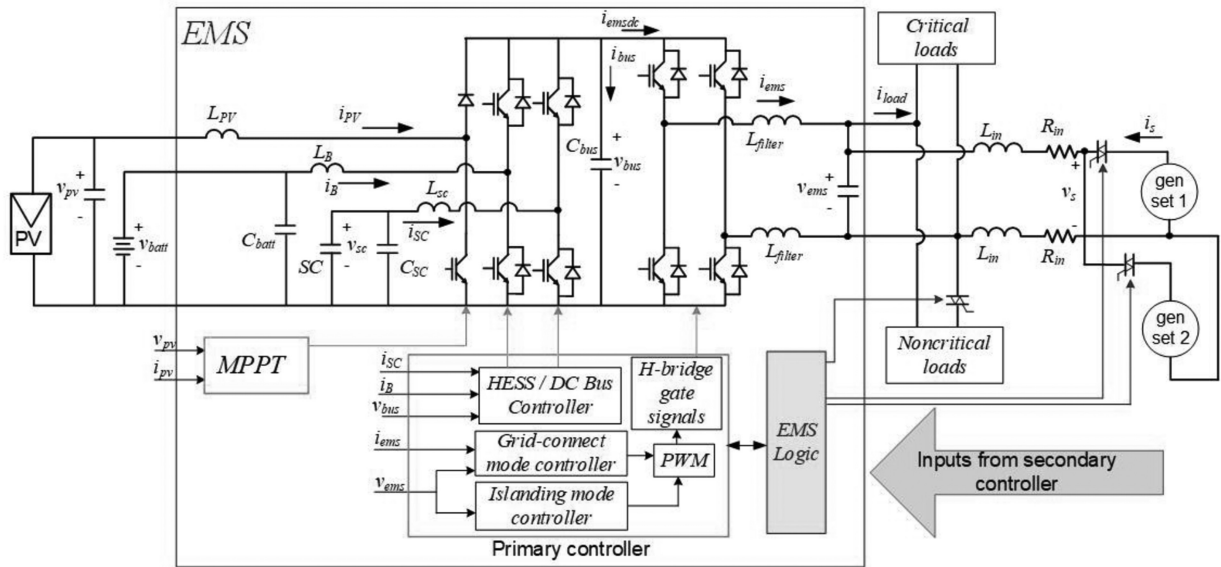


Fig. 1. EMS architecture and its connection to a remote military microgrid with two diesel generators, hybrid energy storage, and PV panels.

To the knowledge of these authors, the proposed combination of a buck–boost converter, a control architecture, and a tight link with the battery lifetime presented in this paper has not been previously presented. Furthermore, the application to a remote microgrid introduces peculiarities in the economics of the considered case study. In this paper, the EMS architecture is presented in Section II. The proposed HESS controller is described in Section III. In Section IV, the procedure, involving the battery lifetime and its link with a specific controller parameter, is presented for a typical power profile of a remote military microgrid. Experimental measurements and conclusions are reported in Sections V and VI, respectively.

II. EMS FUNCTIONALITY AND MICROGRID SETUP

A schematic of the EMS' architecture is provided in Fig. 1 together with the remote military microgrid power system. Also shown in Fig. 1 are critical and noncritical loads and two diesel generator sets (gensets). Critical loads are those loads, including computers, radars, and some air conditioning systems, which are critical to the military operations success and must be powered at all times. Thus, they are hard wired to the ac power bus, while noncritical loads are connected to the ac bus through a solid-state switch controlled by the EMS to enable shedding when necessary. In this setup, the noncritical loads are grouped together for ease of laboratory demonstration; however, an EMS can control multiple noncritical load switches.

The EMS consists of five inverter legs, a field-programmable gate array (FPGA)-based control system, photovoltaic (PV) panels, battery pack, and SCs. Lead–acid batteries are used for the work presented in this paper; however, any other type of battery could be used, and this is true as far as the methodology concerns. However, lead–acid batteries are presently the technology of choice in remote military camps because they fail without catching fire, unlike Li-ion batteries. The FPGA-based controller includes the dc-bus voltage controller, the ac-bus

voltage control during islanding operations, and the EMS ac current in the grid-connected mode. This paper focuses on a new dc-bus voltage controller, while the ac current and voltage control systems are the same as presented in [7]. The FPGA also houses the controls for the HESS and the energy management logic such as load scheduling and grid connect/disconnect. Two legs of the power module are employed as a single-phase bidirectional H-bridge converter, which can be controlled as a current source to inject power from the battery pack to the microgrid or as a voltage source when the gensets are OFF. The third leg of the power module is operated as a bidirectional buck–boost converter to either charge the battery bank or draw energy from it. The fourth leg of the power module is operated as a bidirectional buck–boost converter to either charge the SCs or draw energy from them. Batteries and SCs form the HESS, which is controlled by the EMS. A fifth inverter leg is used as a boost converter that is the interface to the PV panels.

III. HESS CONTROL SYSTEM

In this section, the HESS control system architecture is presented, and its functionality is demonstrated with analysis in the frequency-domain as well as time-domain simulations.

A. Controller Architecture and Functionality

The dc-bus voltage is held constant by the HESS controller shown in Fig. 2. In addition to regulating the dc bus, the goal of this controller is to distribute the load current between the battery and the SCs. Specifically, the load current is the current that the EMS injects into the ac bus to supplement the power provided by the generators. The peak current demanded by the loads is provided by the SCs instead of the battery to reduce the ac current stress on the battery. The low-pass filter commands the battery current to be absent of abrupt changes. The bandpass filter (BPF) is added to extract the 120-Hz signal, which is a

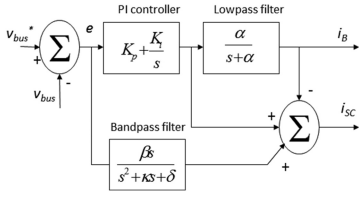


Fig. 2. HESS control system.

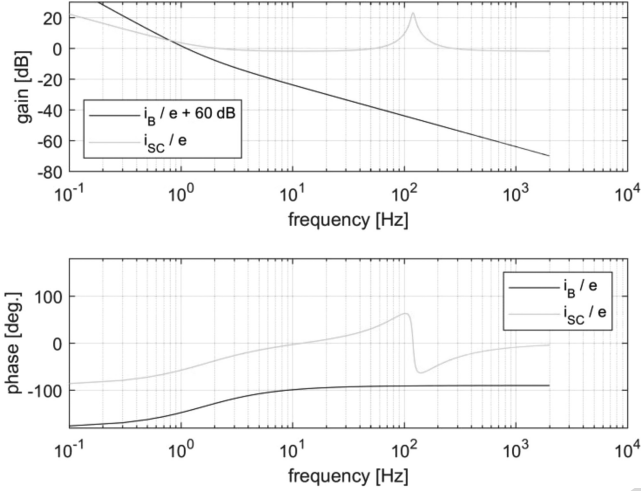


Fig. 3. Bode plots of battery current and SC current transfer functions.

141 frequency equal to twice the 60-Hz output frequency. The goal
142 of the BPF is to reduce the second harmonic voltage ripple on
143 the dc bus and ac current in the battery. The BPF is analyzed in
144 the next section.

145 From the control diagram in Fig. 2, the transfer function of
146 the battery current over the dc voltage error can be derived as

$$\frac{i_B}{e} = \frac{\alpha(sK_p + K_i)}{s(s + \alpha)}. \quad (1)$$

147 The dc voltage error leads to an SC current that is

$$\frac{i_{SC}}{e} = \frac{(sK_p + K_i)}{s} - \frac{\alpha(sK_p + K_i)}{s(s + \alpha)} + \frac{\beta s}{s^2 + \kappa s + \delta}. \quad (2)$$

148 The Bode plots of the transfer functions (1) and (2) are shown
149 in Fig. 3. The battery current i_B contains only low-frequency
150 components, while the SCs provide the current for any ac distur-
151 bances in the dc bus, especially at 120 Hz. In fact, the top
152 plot in Fig. 3 shows that the gain of the transfer function (2) is
153 high at 120 Hz. The parameters used for this analysis are shown
154 in Table I.

155 B. Contribution and Analysis of the BPF

156 It is well known that a 120-Hz component is drawn from the
157 dc bus by a single-phase inverter delivering 60-Hz ac power,
158 so that pulsating power flows from the dc bus in addition to dc
159 power. The role of the BPF in Fig. 3 is to ensure that the 120-Hz
160 current is drawn from the SCs, not the batteries. The following
161 analysis clarifies the contribution of the BPF.

162 The equivalent circuit in Fig. 4 represents the currents that
163 are flowing to/from the EMS dc bus, which are the SC current

TABLE I
PARAMETERS OF THE CONTROLLER

Description	Symbol	Value
DC bus capacitor	C_{bus}	100 μF
Proportional gain	K_p	0.8
Integral gain	K_i	8
BPF coefficient	β	754
BPF coefficient	κ	57
BPF coefficient	δ	568,000
Low-pass filter coefficient	α	0.005 rad/s

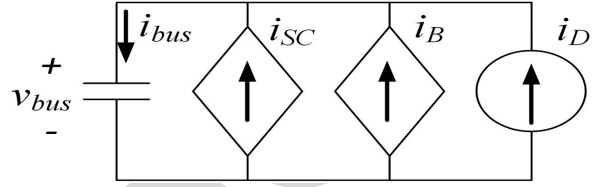


Fig. 4. EMS equivalent circuit for analysis and simulations.

i_{SC} , the battery current i_B , and a disturbance current i_D , which
is equal to i_{emsc} in Fig. 1.

For the equivalent circuit in Fig. 4, the following equations
can be written using basic circuit analysis:

$$i_{SC} + i_B + i_D = sCv_{bus}. \quad (3)$$

First, let us analyze the HESS controller, when there is no
BPF. This is accomplished by removing the BPF block from
Fig. 3 and then rewriting (3) by substituting i_{SC}

$$(v_{bus}^* - v_{bus}) = \left(K_p + \frac{K_i}{s}\right) - i_B + i_B + i_D. \quad (4)$$

From (4), when the disturbance i_D is zero, the dc voltage
transfer function can be derived

$$\frac{v_{bus}}{v_{bus}^*} \Big|_{i_D=0} = \frac{sK_p + K_i}{s^2C + sK_p + K_i}. \quad (5)$$

The dc voltage due to the disturbance current, when the refer-
ence dc-bus voltage is zero, is

$$\frac{v_{bus}}{i_D} \Big|_{v_{bus}^*=0} = \frac{s}{s^2C + sK_p + K_i}. \quad (6)$$

The dc-bus voltage transfer function v_{bus}/v_{bus}^* and the dc-bus
voltage transfer function with the disturbance current as the
input v_{bus}/i_D are plotted in Fig. 5.

The gain for v_{bus}/i_D is greater than 1, which will cause a
lot of ripple if there is any ac current present in the disturbance
current. The second harmonic current flowing from the dc bus to
the ac load is significant and causes dc-bus ripple. This problem
can be mitigated by adding a BPF to identify and close a control
loop on the second harmonic voltage error.

In contrast to the previous analysis, let us analyze the com-
plete HESS shown in the block diagram of Fig. 3. With the BPF,
the dc voltage transfer function becomes

$$\frac{v_{bus}}{v_{bus}^*} \Big|_{i_D=0} = \frac{s^3K_p + s^2K_2 + sK_1 + \delta K_i}{s^4C + s^3K_3 + s^2(K_2 + \delta C) + sK_1 + \delta K_i} \quad (7)$$

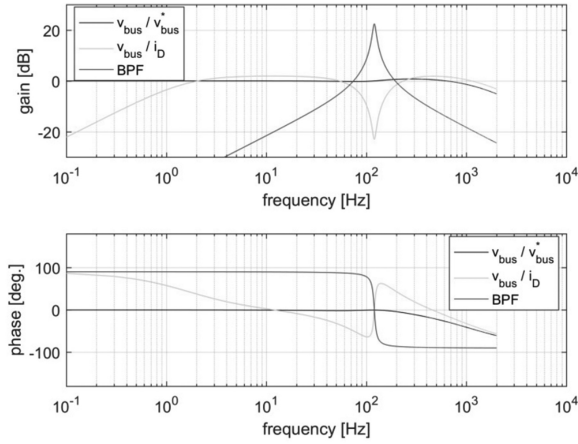


Fig. 5. Bode plots of the transfer functions (5) and (6) without the BPF.

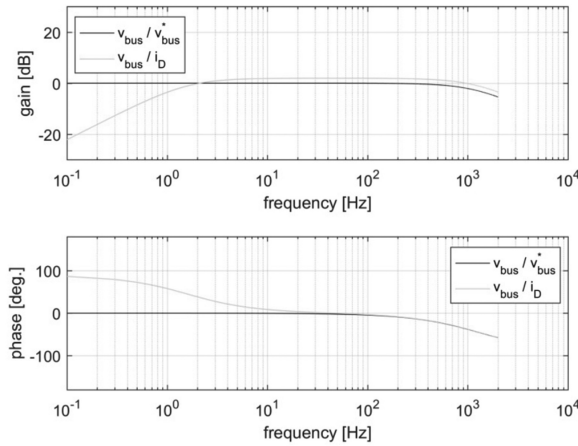


Fig. 6. Bode plots of the transfer functions (7) and (8) with the BPF.

187 and the dc voltage due to the disturbance current is

$$\frac{v_{bus}}{i_D} \Big|_{v_{bus}^* = 0} = \frac{s^3 + \kappa s^2 + \delta s}{s^4 C + s^3 K_3 + s^2 (K_2 + \delta C) + s K_1 + \delta K_i} \quad (8)$$

188 The coefficients for the transfer functions (7) and (8) are

$$K_1 = \delta K_p + \delta K_i \quad (9)$$

$$K_2 = \kappa K_p + \delta K_i + \beta \quad (10)$$

$$K_3 = \delta K_p + \kappa C. \quad (11)$$

189 Fig. 6 shows the Bode plots of the transfer functions (7) and
 190 (8), where the BPF was added to the control architecture. In
 191 contrast with Fig. 5, it can be observed that the addition of the
 192 BPF reduces the second harmonic voltage ripple in the dc bus.
 193 The BPF has very high gain at 120 Hz, and it reduces the transfer
 194 function v_{bus}/i_D significantly at 120 Hz. The BPF has a minimal
 195 effect on the transfer function v_{bus}/v_{bus}^* .

196 A time-domain simulation of the system represented in Fig. 1
 197 is shown in Fig. 7. The 120-Hz component of the dc-bus ripple is
 198 reduced by the BPF. Also, as shown in Fig. 7, the step response
 199 to an increase in the dc load current is reduced. The disturbance
 200 current used in the simulation of Fig. 7 is a 10-A sinewave at

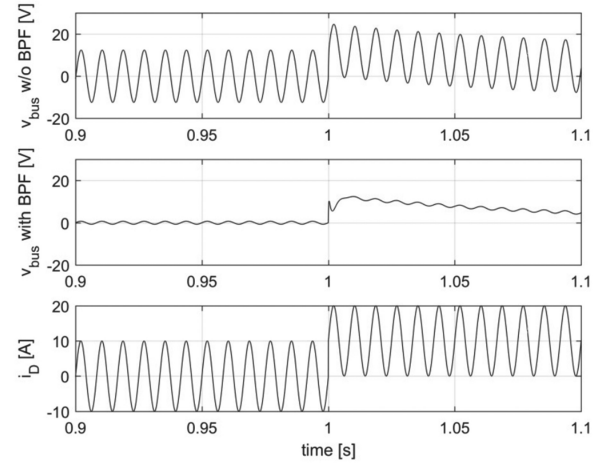
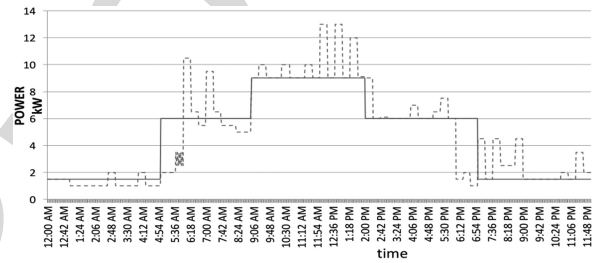
Fig. 7. Time-domain behavior of the HESS controller with and without the BPF for an injected disturbance current i_D (v_{bus} error is shown here).

Fig. 8. Load power consumption profile of a remote microgrid, real (DASHED) versus linearized (SOLID) (2-min resolution).

120 Hz plus a 10-A step change in the load current at $t = 1$ s 201

$$i_D(t) = 10u(t-1) + 10\sin(2\pi 60t). \quad (12)$$

IV. LIFETIME EXTENSION AND ECONOMICS OF AN OPTIMIZED HESS 202

203
 204 In this section, the proposed HESS is used in a remote military
 205 microgrid to demonstrate how the above control increases the
 206 battery lifetime compared to the same microgrid, where only
 207 batteries are used for energy storage. A HESS shows its potential
 208 when sudden spikes, not negligible because of the same order of
 209 magnitude than the base load, occur. The analysis in this section
 210 proves that, when the HESS draws the load transient currents
 211 from the SCs, the batteries will last longer. The battery lifetime
 212 extension is quantified for different values of the low-pass filter
 213 coefficient, and the overall microgrid economics is analyzed.

214 The power profile of Fig. 8 (dashed line) represents the typical
 215 daily consumption in a remote military microgrid, where sudden
 216 peaks occur and seriously affect the lifetime of the batteries.
 217 This profile is used for the following analysis and case study
 218 in contrast to the simplified profile (solid) also plotted in Fig. 8
 219 and used for the study reported in [8].

A. A Few Considerations on the Role of the Optimization 220

221 Fuel consumption is one of the parameters that are worth
 222 minimizing in a remote military microgrid because fuel

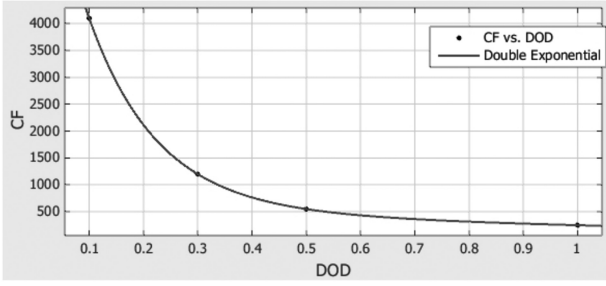


Fig. 9. Reference fitting curve for the lead-acid batteries of the experimental test (CF versus DoD) [26].

transportation to remote sites can result in casualties. In a previous study, the optimization model and its constraints were thoroughly discussed [23]. The results of that optimization are based on 2-min intervals, and it provides the rules for the power sharing among the various sources, taking into account how fast the response from SCs can arrive. These sources include two diesel generators (5 and 15 kW), the PV source (3 kW_P, which is deterministic in the proposed example), and the HESS. In addition, the optimization algorithm ensures that the batteries operate within safe SoC limits, and the generators operate within their range of operation and efficiency. The power associated with HESS, P_{HESS} , is thus obtained and is being used in this novel analysis, where the focus is the evaluation of the lifetime of the batteries and the economics of the system, when the controller parameter α varies.

B. Link Between the Controller and the Battery Lifetime

Different battery and SC currents can be obtained by changing the low-pass filter coefficient α , still keeping P_{HESS} constant. With these currents as inputs, we can evaluate the SoC for both devices and find which is the best SoC* to support the optimized rules

$$\text{SoC}_*(t) = \text{SoC}_*(t-1) - \frac{P_*(t)\Delta t}{\text{ASE}_*}. \quad (13)$$

With SoC_* , P_* , and ASE_* , we identify the SoC, the active power (positive when storage is feeding the load, and negative when is charging), and the storage capacity of each specific device (either the battery or the SCs)

$$P_*(t) = P_{\text{BAT}}(t) = v_{\text{bus}}(t) \cdot i_B(t) \text{ but also}$$

$$P_*(t) = P_{\text{SC}}(t) = v_{\text{bus}}(t) \cdot i_{\text{SC}}(t) \quad (14)$$

$$P_{\text{HESS}}(t) = P_{\text{BAT}}(t) + P_{\text{SC}}(t). \quad (15)$$

The currents i_B and i_{SC} must have the same sign, or being 0, meaning that when one device is charging or discharging, the other must act accordingly or it must be OFF.

The battery lifetime is thus assessed by using the Rainflow counting method [24], [25], which needs the results of the SoC over time to provide the number and typology of cycles characterizing the charge and discharge of the battery over a typical horizon. Each kind of battery shows its own cycle to failure (CF) versus depth of discharge (DoD). In Fig. 9, such data for the lead-acid batteries used in the laboratory prototype are

reported. We recall that the use of lead-acid batteries is due to safety reasons. Nevertheless, this methodology applies to any kind of battery technology, as long as the CF versus DoD curve can be obtained.

The Rainflow counting algorithm provides information on amplitude (related to the DoD) and frequency of cycles presenting the same amplitude on a set time horizon. The life expectancy of the battery is related to the CF, with $1/\text{CF}$ being the life fraction. We can assess D , the inverse of the lifetime, as

$$D = \sum_{i=1}^m \frac{N_i}{\text{CF}_i} \quad (16)$$

where m is the number of different DoD_{*i*}, occurring in the set horizon, N_i is the frequency associated with DoD_{*i*}, and CF_i is the corresponding number of cycles at DoD_{*i*}. For a fully functional battery, D has to be less than 1. When $D = 1$, the battery is considered dead; its unit measure depends on how the number of cycles N_i is counted: if N_i are counted over a day, then the lifetime of the battery (inverse of D) counts the days to failure (DF).

An exemplification: if in a typical day, a battery experiences 10 cycles/day (N), where DoD (the amplitude of the equivalent charge/discharge cycle) is equal to 0.2, then that battery can ideally survive for up to 200 equivalent days, before being considered dead. In fact, $\text{CF}_{\text{DoD}=0.2}$ is 2000; hence, the lifetime in days $1/D = \text{CF}/N = 200$. When multiple cycles occur, the lifetime is the composition of each single assessment. D depends on N , which relates to DoD; DoD depends on SoC_{BAT} and SoC depends on the low-pass filter coefficient α ; thus, D depends on α .

To sum up the analysis: the higher α , the less current on SC; thus, the lower the lifetime of the battery. To achieve a certain lifetime, we tune the α value, accordingly.

The overall implemented procedure ensures minimum fuel consumption, while suitably tuning the battery lifetime, at the same time. This last objective is achieved by tuning the HESS controller. In the following subsection, we will show how α will also affect the HESS investments and its economics.

C. Case Study Results

In Fig. 10, input and output data, from the optimization procedure described in [23], are reported for the case study: a remote military microgrid. Case A is the reference case when storage is made up only by batteries (no SCs), while case B represents the case when the HESS is present (with SCs). The needed data for both cases, regarding the features of the optimized considered system, deal with the battery, the gensets, and the load profile on a set horizon. In particular, for the battery, the parameters are the following:

- 1) SoC_{min} and SoC_{max} ;
- 2) charging/discharging efficiency η ;
- 3) rated power P_{max} ;
- 4) discharging/charging time at P_{max} ;
- 5) available capacity.

For gensets 1 and 2, the parameters are the rated powers P_{n1} and P_{n2} and the related relationships between the load factor

	Description	Case B	Case A (Reference)	unit measure	note or symbol
battery	state of charge: range	0.5-1	0.2-1	p.u	SoC min-max values
		SoC(t=0)=SoC(t=end of the day)			typical day assumption
	charge/discharge efficiency	0.9/0.9	0.9/0.9	p.u.	
	maximum available constant battery power (rated)	3	3	kW	Pbat max
	duration time @ Pbat max	6	6	hours	T
	Battery capacity	18	18	kWh	ASE
genset	rated power	5-15	5-15	kW	P1n, P2n
general parameters	a total of 720 t per typical day	2	2	minutes	t step
	Load in a typical day	Fig.8	Fig. 8	W	PL
results	power from/to the storage	$P_{\text{hess}} = 200 \cdot (i_B + i_{\text{SC}})$	$P_{\text{bat}} = 200 \cdot i_B$	W	
	power shares among sources	Fig.14	Fig.11		

Fig. 10. Matrix visualization of all the most important input and output information coming from the previous optimization.

310 and the fuel consumption [8]. Furthermore, the time step t and
311 the load profile P_L shall be established.

312 Also, Fig. 10 reports in the last row the indication to the result-
313 ing output, the balancing of powers to feed the load, meaning the
314 sequence of powers from gensets and to/from the storage unit.

315 Additionally, for case B, in Table II, technical data of the used
316 basic SC module are reported.

317 When the DASHED load profile of Fig. 8 is considered for a
318 typical day, the optimized procedure identifies the best $P_{\text{HESS}}(t)$
319 (or P_{BAT} if no SCs are present) for each time step of the day and
320 for the given conditions.

321 For case B, we consider three scenarios, cases B1, B2, and B3,
322 identified by different values of the low-pass filter coefficient α ,
323 equal to 0.005 (B1), 0.003 (B2), and 0.001 (B3). Once the opti-
324 mization has produced the power share among the gensets and
325 the storage, then different alphas determine a different sequence
326 for $i_B^{B1, B2, B3}(t)$ and $i_{\text{SC}}^{B1, B2, B3}(t)$ and thus $\text{SoC}_B^{B1, B2, B3}(t)$: su-
327 perscripts identify the respective scenarios. Similarly, when no
328 SCs are considered, then we will have $i_B^A(t)$ and $\text{SoC}^A(t)$.

TABLE II
SC MAXWELL DATA [27]

Product name	BMOD0130P056 B03
Rated Capacity (F)	130
Rated Voltage (V)	56
ESR (mΩ)	8.1
Leakage current (mA)	120
Absolute maximum current (A)	1,900
max continuous current (A)	61 ÷ 99
Weight (kg)	18
Stored Energy (Wh)	56.6
$P_{\text{SC}_{\text{max}}}$ (kW)	96.79
$t @ P_{\text{SC}_{\text{max}}}$ (s)	2.11
cost of single unit (\$)	1,300

TABLE III
CASE STUDY: MAIN RESULTS FOR THE THREE SCENARIOS (B1, B2, AND B3) AND THE REFERENCE CASE A

Description ↓	$\alpha \rightarrow$	0.005	0.003	0.001	Ref.
		B1	B2	B3	case A
SCs capacitance	F	650	910	1300	no SC
set SoC_{min}		0.5	0.5	0.5	0.2
The lowest SoC	%	69.9	67.53	67.06	30
Cycles		20	18	16	51
Lifetime	days	274	282	363	122
Invest. on SC	k\$	6.5	9.1	13	N.S.
Invest. on BAT.	k\$	10.8	10.8	9.	25.2
Tot. Inv. (5y)	k\$	17.3	19.9	22.	25.2

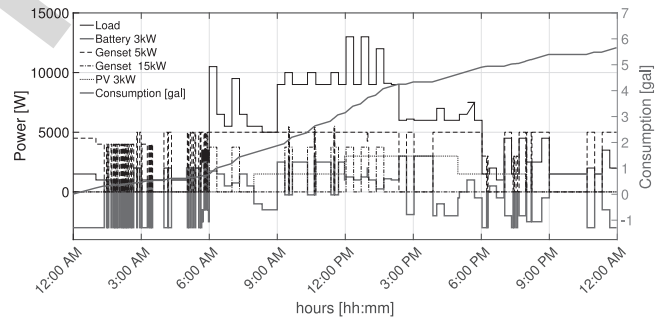


Fig. 11. Case A (only with batteries): from the optimization [23]: power profile and consumption (resolution step 2 min; $\text{SoC}_{\text{min}} = 0.2$).

329 For each of the four SoC sequences, a new series of
330 $\text{DoD}^{B1, B2, B3, A}$ is derived, and different lifetimes are expected.

331 In Table III, the main results are reported for the three scenar-
332 ios [increasing SCs number from 5 (B1) to 10 (B3)], after the
333 optimization and the tuning of α , as well as for the Reference
334 case A, optimized but without SCs.

335 The investment (INV^i) in each i th scenario/case is thus eval-
336 uated as in the following, depending on the DF of the batteries,
337 which ultimately depends on alpha:

$$\text{INV}^i(\text{DF}(\alpha)) = \text{INV}_{\text{SC}}^i + \sum_{\text{DF}} \text{INV}_B^i(\text{DF}(\alpha)). \quad (17)$$

338 The change in i_B and i_{SC} sequences can be visualized when
339 simulating the battery current with and without SCs. In Figs. 11

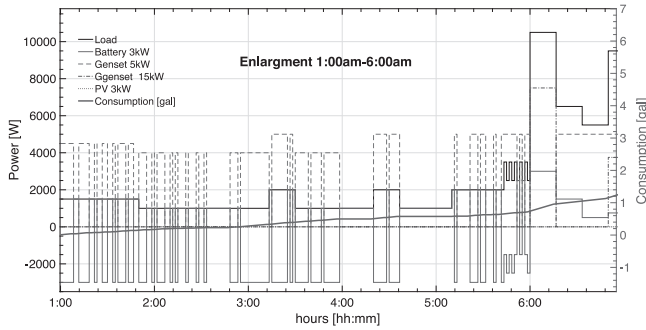


Fig. 12. Case A: Enlargement between 1:00 A.M. and 6:00 A.M. (resolution step 2 min; $SoC_{min} = 0.2$).

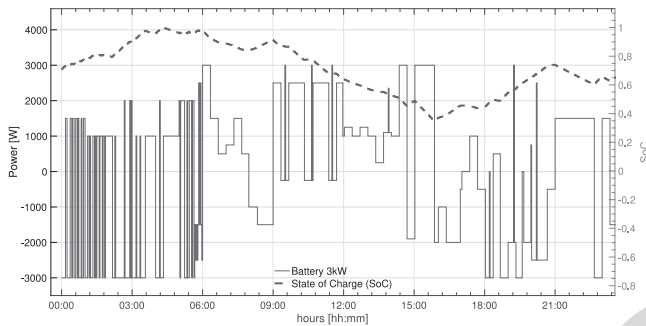


Fig. 13. Case A: Optimal P_{BAT} and consequent SoC on batteries (resolution step 2 min; $SoC_{min} = 0.2$).

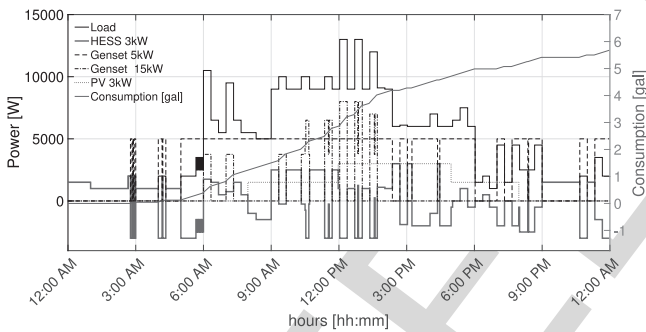


Fig. 14. Case B (with the HESS): from the optimization [23]: power profile and consumption (resolution step 2 min; $SoC_{min} = 0.5$).

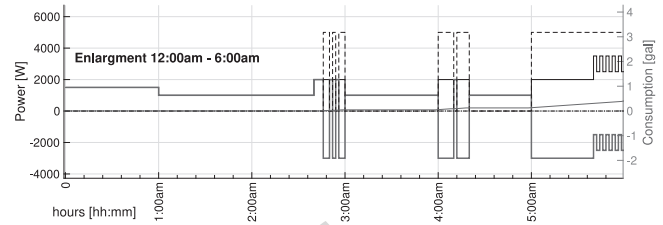


Fig. 15. Case B: Enlargement between 1:00 A.M. and 6:00 A.M. (resolution step 2 min; $SoC_{min} = 0.5$).

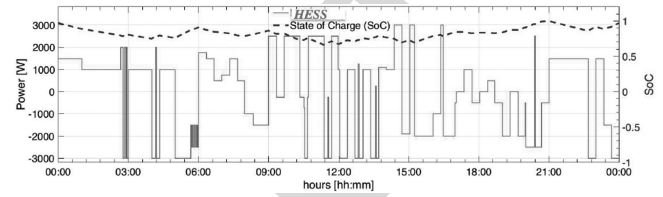


Fig. 16. Case B: Optimal P_{HESS} and consequent SoC on batteries (resolution step 2 min; $SoC_{min} = 0.5$).

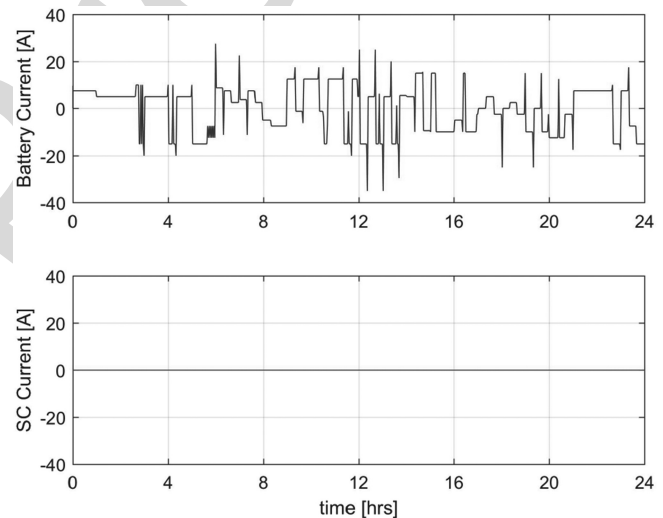


Fig. 17. Case B: battery current when the HESS control system is in place but is disabled (with transient, resolution step 0.1 s). SC current is zero.

340 and 14, the optimization results are reported for case A (Ref-
 341 erence) and for case B: what is referred as HESS profile is the
 342 active power associated with what is coming from/to the storage
 343 unit (no transient considered, resolution step 2 min).

344 For a better understanding of the battery dynamic, in terms of
 345 charging/discharging cycles and consequent SoC, Figs. 12 and
 346 13 for case A and Figs. 15 and 16 for case B are reported.

347 They show an enlargement of Figs. 11 and 14, respectively,
 348 when SCs are not included (case A) and when they are included
 349 (case B).

350 A Simulink model produced the simulated plots in Figs. 17–
 351 19. Omitting the switching behavior of the EMS power
 352 converters lead to shorter simulation times for the battery
 353 current over a 24-h period (resolution 0.1 s). In Fig 17, the
 354 battery current is plotted (upper), when yet the SCs are to be

turned ON, the transient is considered, and the resolution step
 is 0.1 s. In Figs. 18 and 19, battery and SC currents are shown,
 respectively, when the HESS controller is operational with the
 low-pass filter coefficient $\alpha = 0.005$ (B1) and $\alpha = 0.001$ (B3).
 It can be observed that the battery current is much smoother
 when the HESS controller is used to redirect the peak currents
 to the SCs, and we can also notice how the α value affects the
 i_B profile (upper graph of Figs. 17–19).

In Fig. 20 (Scenario B1) and Fig. 21 (Scenario B3), the battery
 cycles are reported for α equal to 0.005 (smaller SCs) and 0.001
 (bigger SCs). The main results are reported in Table III, where
 the increase in DF (122 estimated days with no SCs, 274 for
 $\alpha = 0.005$ up to 363 for $\alpha = 0.001$), the assessment of the lowest
 SoC, and investments are assessed with respect to the illustrated
 procedure.

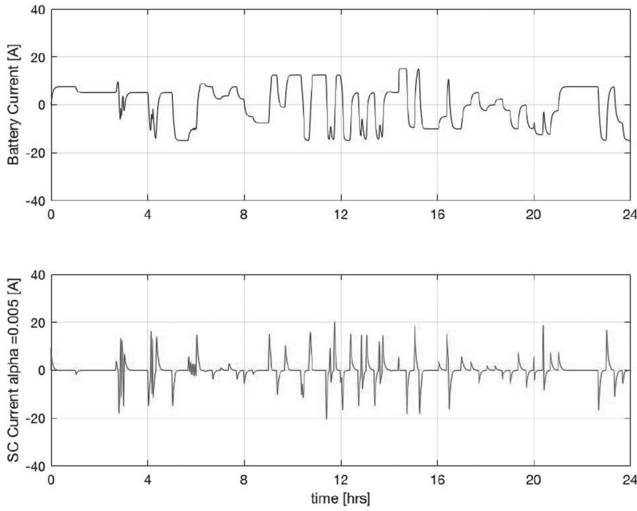


Fig. 18. Scenario B1: battery and SC current when the HESS control system is enable. SCs takes the peaks of the load current ($\alpha = .005$, resolution step 0.1 s).

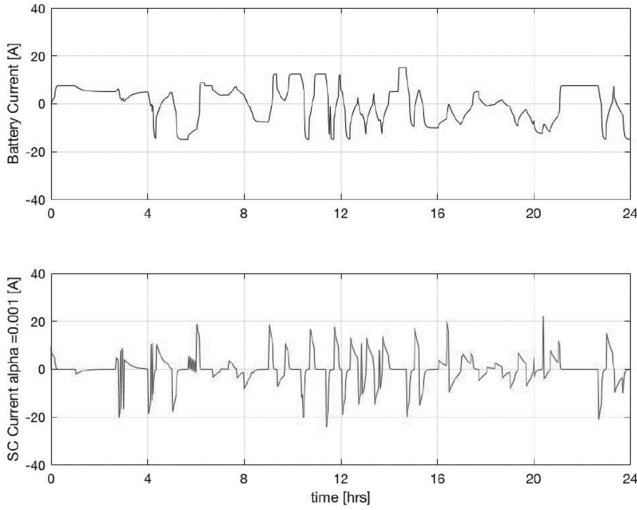


Fig. 19. Scenario B3: battery and SC current with the HESS control system. SCs take the peaks of the load current ($\alpha = 0.001$, resolution step 0.1 s).

370 The plots in Fig. 22 and the results in Table III demonstrate
 371 how the battery lifetime is extended when the HESS controller
 372 is used, realizing the least investment over five years, when α
 373 is lower, thus finding the suitable tradeoff between increasing
 374 SCs size and the battery wearing. We can also notice that
 375 below ~ 900 operating days, even only five SCs modules are
 376 not convenient against batteries, but above 900 days, the HESS
 377 becomes cost effective. Over ~ 1500 days, every investigated
 378 HESS is more cost effective than batteries alone.

379 Depending on the size of the SC and batteries, thus on the
 380 deriving cycles to failure, we can infer that the daily power
 381 consumption is a key parameter for the economic evaluation.
 382 Therefore, careful microgrid load analysis should be done to
 383 create a reliable load profile. A sensitivity analysis can also be
 384 performed to identify not only the actual optimum, but also the
 385 proper range of validity for the current assessment and link it
 386 to the controller parameters. This will be illustrated in a future
 387 work.

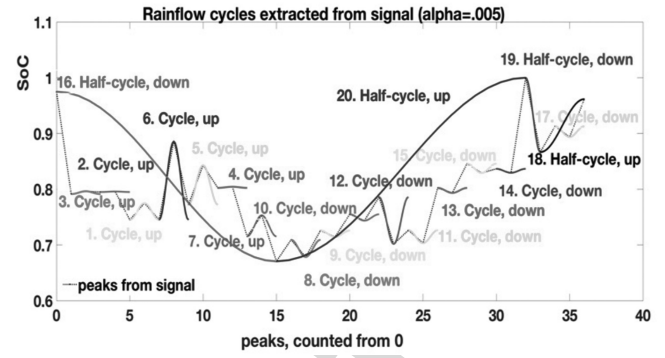


Fig. 20. Scenario B1: Counting cycles at different DoD ($\alpha = 0.005$).

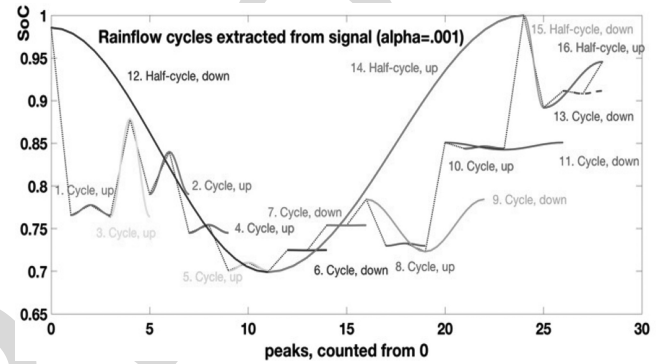


Fig. 21. Scenario B3: Counting cycles at different DoD ($\alpha = 0.001$).

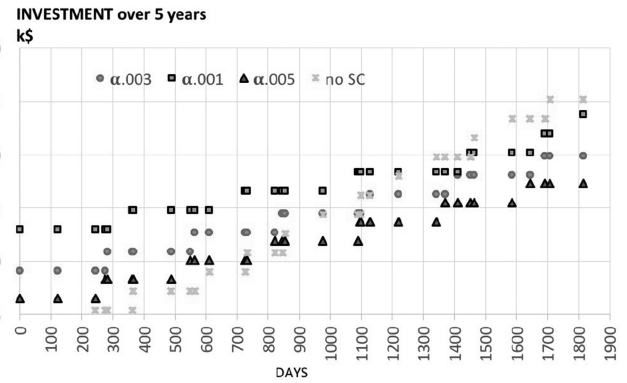


Fig. 22. Investment over five years ($\alpha = 0.005$ blue triangle; $\alpha = 0.003$ orange circle; $\alpha = 0.001$ gray square; yellow star = Reference—NO SCs).

388 Our methodology makes easily evident how those battery
 389 technologies with higher CF versus DoD (for instance, the
 390 lithium ones) can positively affect the lifetime assessment be-
 391 cause higher CF values directly influence (16). On the other
 392 hand, they cost more; thus, again, another sensitivity analysis,
 393 focusing on prices, can help in investigating how far our con-
 394 siderations can be stretched.

V. EXPERIMENTAL MEASUREMENTS

395
 396 To verify the functionality of the proposed HESS controller, a
 397 laboratory experiment was conducted with a scaled-down EMS
 398 prototype. The laboratory setup is represented in Fig. 23, where
 399 the EMS is included inside the blue box. Note that instead of

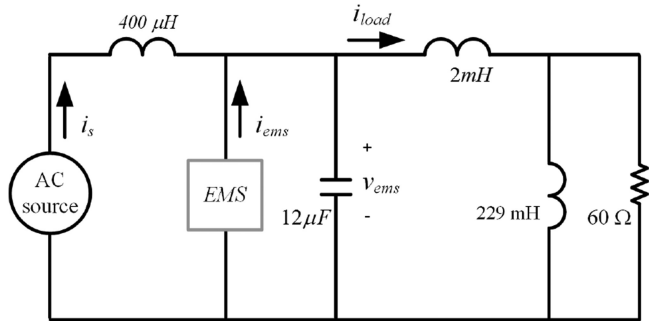


Fig. 23. Schematic of the laboratory setup for the experimental validation.

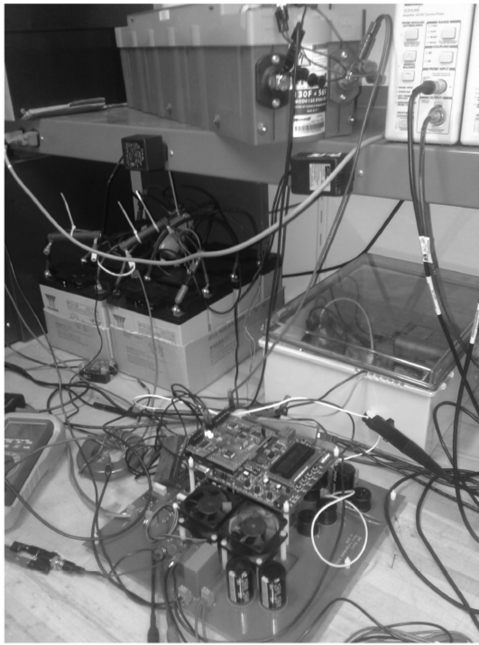


Fig. 24. Photograph of the laboratory setup.

400 a diesel generator, the ac 120-V 60-Hz power available in the
401 laboratory was used.

402 A photograph of the prototype on the laboratory bench is
403 shown in Fig. 24. A 130-F Maxwell SC [27] and six Genesis
404 12-V lead–acid batteries [26] connected in series are visible
405 in the photograph, together with the EMS hardware, which in-
406 cludes several printed circuit boards (PCBs) and external pas-
407 sive components. The EMS controller is embedded on an FPGA,
408 which is part of a Xilinx developed board [28]. The other PCBs
409 are custom made, with the bottom one comprising the power
410 electronics and passive components and the top PCB includ-
411 ing A/D converters, USB interface to communicate with a per-
412 sonal computer, and other electronic components that interface
413 with the FPGA board. The code for the FPGA is developed
414 in Simulink with the additional Xilinx System Generator [29]
415 blocks. Further details of the EMS hardware and FPGA software
416 implementation are available in [7] and [9].

417 The first set of experiments produced the steady-state plots in
418 Figs. 25 and 26 with an without the proposed HESS controller,

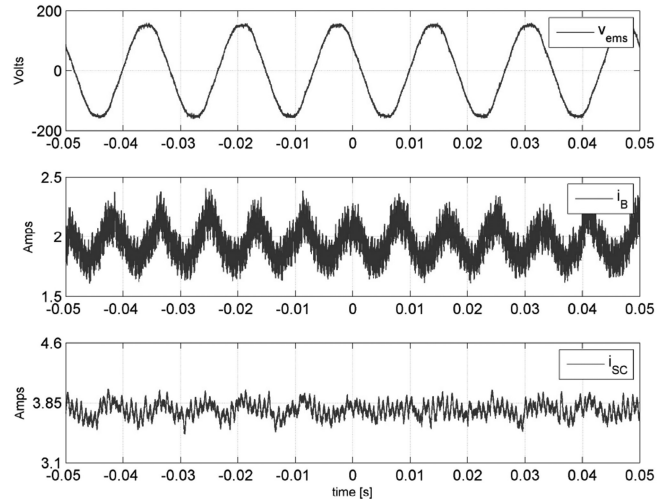


Fig. 25. Case B without HESS controller steady-state experimental waveforms. From the top: ac voltage, battery current, and SC current.

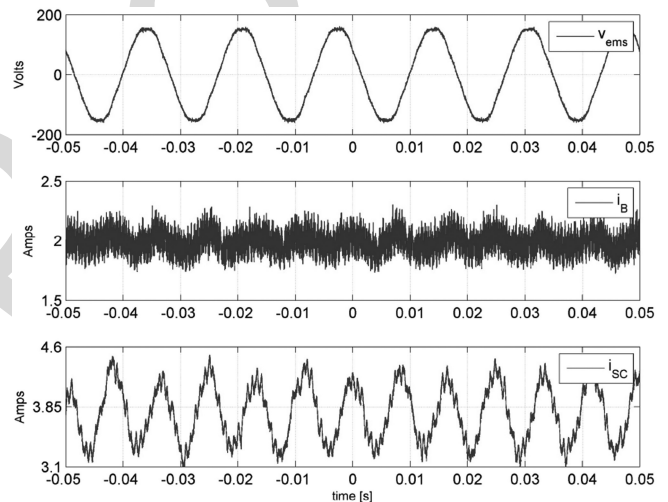


Fig. 26. Case B with HESS controller steady-state experimental waveforms. From the top: ac voltage, battery current, and SC current.

419 respectively. The two figures include, from the top, the ac source
420 voltage, the battery current, and the SC current. The contrast
421 between the battery current in Fig. 25 and the battery current
422 in Fig. 26 validates the effectiveness of the HESS controller in
423 removing the 120-Hz frequency component from the battery and
424 sending it to the SC. Harmonic analysis of the battery current
425 from Fig. 25 shows that the amplitude of the 120-Hz harmonic
426 is 158 mA. In contrast, the 120-Hz harmonic of i_B in Fig. 26 is
427 45.5 mA, a substantial reduction.

428 A second set of experiments is shown in Figs. 27 and 28,
429 where the dynamic performance of the system is contrasted
430 without and with the HESS controller, respectively. As discussed
431 in previous sections, in order to reduce the charge and discharge
432 cycles on the batteries, the SC is commanded to absorb or deliver
433 currents that are suddenly needed by the microgrid. One example
434 is just before 3:00 P.M. (or 15:00 hours; see Fig. 14); when a
435 large amount of energy is being sent to the HESS and, as shown
436 in Fig. 18, the SC absorbs the initial peak. Fig. 27 demonstrates

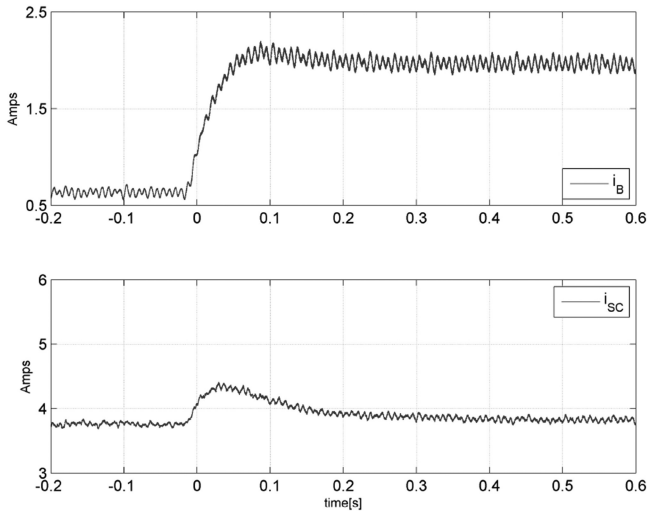


Fig. 27. Battery charge current increase from 1 to 2 A without the HESS controller. Upper i_B and lower i_{SC} .

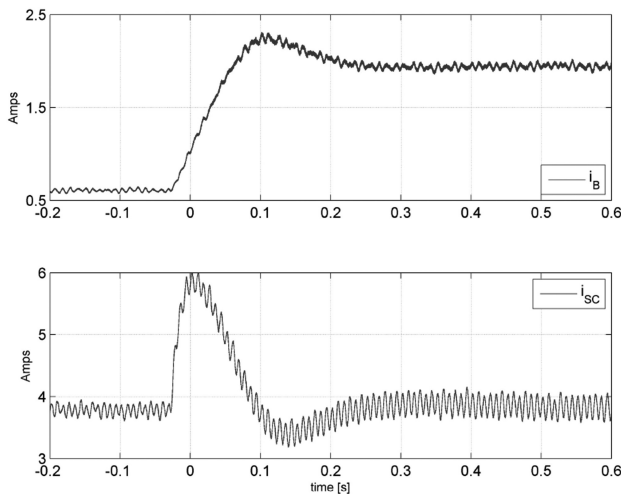


Fig. 28. Battery charge current increase from 1 to 2 A with the HESS controller. Upper i_B and lower i_{SC} .

437 what happens when the current sent to the battery is doubled
 438 from 1 to 2 A: the surge is evident in the battery current, i_B , as
 439 as well as the 120-Hz ripple. In contrast, the di/dt on the battery
 440 current i_B is reduced when the HESS controller is turned ON,
 441 and also, its ripple is noticeably reduced. Note that $\alpha = 5$ Hz
 442 for this experiment.

443

VI. CONCLUSION

444 This paper presents a novel HESS controller focused on
 445 increasing the lifetime of the batteries by using SCs with a
 446 buck-boost converter to control their charge and discharge, thus
 447 maximizing their utilization. A realistic load profile is used, and
 448 several scenarios are compared to link the controller parameter
 449 α with the battery lifetime extension and to the economic eval-
 450 uations. Therefore, the economic evaluation is performed on a
 451 five-year period, time needed to show when the HESS may be-
 452 come cost effective for the case study. The SCs are sized to take
 453 the stress off the load power transients from the battery pack,

so that the batteries only “see” an idealized load profile and can
 perform at better conditions.

Experimental measurements demonstrate the ability of the
 proposed HESS controller to suppress the 120-Hz ripple from
 the battery as well as reduce the di/dt when higher currents
 are commanded. This result proves that the HESS controller
 redirects higher frequency currents to the SC and leave for the
 batteries only slow current changes in order to increase the
 battery lifetime.

Future work will focus on optimizing the number of SCs in
 order to reduce their economic impact on the microgrid.

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