


Article

Exploration of Alternatives to Reduce the Gap in Access to Electricity in Rural Communities—Las Nubes Village Case (Barranquilla, Colombia)

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Abstract: The global crisis associated with COVID-19 and the Russia-Ukraine conflict has affected progress towards the Sustainable Development Goals (SDGs). Projections for SDG7 (Ensure access to affordable, reliable, sustainable, and modern energy for all) indicate a slowdown in the pace of electrification. Thus, the problem of poverty will persist in many regions as long as access to electricity remains difficult. This work analyzes some solutions to the lack of electricity supply in a rural community using organic waste from its economic activity and the integration of other available renewable sources to make electricity affordable and reliable. A model that minimizes the levelized cost of energy and restricts the proportion of annual energy not supplied to less than 5% of the community's annual demand optimizes the performance of off-grid and on-grid systems. These systems have in common the production of electricity from biogas produced from swine manure, supplemented with wind and solar generation. Batteries and diesel generators support the operation of off-grid systems. As expected, the grid-connected system presented the best performance; however, the result reaffirms the need for governments to ensure the policy and infrastructure conditions that facilitate the grid connection of vulnerable communities to achieve SDG7.

Keywords: affordable electricity; reliable electricity; renewable energy; SDG7; hybrid renewable energy systems (HRES)



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1. Introduction

The recent global crises associated with COVID-19 and the Russia-Ukraine conflict have affected progress toward achieving the UN Sustainable Development Goals (SDGs) [1]. Specifically, the level of poverty increased by 5 percentage points due to COVID-19, from 6.7% to 7.2% globally, undermining progress for SDG1 [2]. Regarding UN SDG7 (Ensure access to affordable, reliable, sustainable, and modern energy for all), projections indicate a slowdown in the pace of electrification in the 2020–2030 period compared to the 2010–2020 period [3].

Thus, the problem of poverty will persist in many regions as long as access to electricity remains difficult due to a lack of infrastructure and disinterest in investing for this purpose. There is a direct relationship between negative social effects and lack of electricity [4]. Rural areas are the most affected by the lack of access to electricity because most of them are in remote or isolated locations where it is very difficult and/or very costly to transport energy using the conventional networks of the electricity system [5]. Many of these communities subsist by raising animal or vegetable species for self-consumption or income generation, and their waste generates pollution problems. However, some of these communities live in regions with great potential to take advantage of renewable energy sources such as solar, wind, and biomass that, if well utilized, can meet their current and future electricity demand and create new sources of income generation. Renewable energies could enable these communities to increase their quality of life, improve their economic and social

situation, have efficient educational and health services, and develop easier and more effective work techniques, among other things.

Renewable energies can play an important role in reducing the energy access gap due to their efficiency and effectiveness, thus resolving the asymmetries caused by traditional electricity systems [6]. Some authors have proposed hybrid micro-grids composed of renewable energies to reduce the electricity access gap in remote communities [4,7]. Other authors insist that, in order to achieve SDG7, the use of renewable energy must be promoted through policies, because reducing consumption alone will not work in the long term [8].

Energy access has been analyzed from different perspectives in the literature. Some authors have focused on the use of available renewable resources [9,10] or on the analysis of regulatory incentives [11]. Other authors have explored the effects of changes in renewable energy regulatory policies on the attractiveness of using micro-grids [12]. In [13], the authors break down SDG 7 into a series of indicators that facilitate the definition of actions to achieve it. For example, affordability is achieved using low-cost electricity, lower capital investment, and a lower net present cost. Those authors explain that reliable energy consists of a continuous supply of energy and the minimization of energy shortages; sustainable energy is achieved with a higher share of renewable energy and minimal emissions. Finally, the authors associate modern energy with micro-grids, hybrid energy systems, and community micro-grids. Recent research emphasizes the need to approach the modeling of hybrid renewable energy systems (HRES) from the perspective of energy sustainability, energy security, and energy affordability [4].

HRES integrate different renewable resources to generate electricity, offering a diversified strategy that provides multiple advantages [14]. In recent years, the development of HRES implementation projects in rural areas with no or poor electricity supply is increasing because they generate multiple benefits for the community. The main one is to provide clean and affordable energy to millions of people, which contributes to a better standard of living through the creation and/or restructuring of hospitals, educational institutions, and recreation centers, offering new and varied social services, local job opportunities, and more technical, agricultural, and livestock activities. All this generates the social, economic, and technological development of the communities. The World Bank reports the implementation of such systems in rural areas of countries in North America, Latin America, Europe, Asia, and Africa (<https://www.worldbank.org/en/search?q=Renewable+Energy+to+Rural+Development¤tTab=2>, accessed on 29 August 2023).

Table 1 shows some works, identifying the location of the case study, methods, tools, software, criteria, backup systems, and a summary of the study.

Table 1. Studies in hybrid solar PV–wind–biomass.

Reference	Location	Methods/Software Tool	Evaluation Criteria	Additional Components	Summary
[14]	India	DE-MATLAB	NPC	MHP BAT	The authors dimensioned an isolated system based on renewable resources in the area. Comparison with PSO and GA. Sensitivity analysis with costs and input parameters.
[15]	India	HOMER	NPC	DG	Analysis of several configurations; the best was PV/BM/DG. The authors sought to reduce pollution and minimize NPCs.
[16]	Iran	HOMER	NPC, COE, CEQ	BAT	Analysis of seven systems; the best one was PV/BM/DG.
[17]	Saudi Arabia	MDVP	NPC	BAT DG	Analysis of four configurations; the best one was PV/BM. They compared the algorithms AEFA, HHO, GWO, and MDVP. IN conclusion, MDVP was the best.

Table 1. Cont.

Reference	Location	Methods/Software Tool	Evaluation Criteria	Additional Components	Summary
[18]	Iran	GA, PSO	NPC		The authors considered the rare PV/W/BM. The PV/biomass system was the most cost-effective. They compared GA and PSO.
[19]	Afghanistan	HOMER	NPC	DG	The most optimal solution for 300 families was PV/BM/DG.
[20]	Egypt	QRUN	COE		They analyzed the superiority of the QRUN algorithm with respect to other solution algorithms.
[21]	Egypt	SMA, SOA, GWO, SCA	COE	BAT	Analysis of three configurations; the best was W/BM/BAT. Comparison of the algorithms; the best was SMA.
[22]	Egypt	MOGA	TC	BAT	They used Monte Carlo simulations to simulate solar radiation and wind speed. They considered six systems; the best was solar PV. They used rice straw as the biomass.
[23]	India	PMBSA	TC, LPSP	BAT DG	Weibull distribution models for wind and solar radiation. Normal distribution models for load demand. They analyzed two scenarios: with and without biomass. The first was better.
[24]	Saudi-Arabia	WOA, FF, PSO	COE	PHS BAT	Two configurations were considered. The best was W/PV/BM/PHS.
[25]	India	PSO, GWO	COE, DPSP		The performance of the algorithm was compared with others under different scenarios in three places.
[26]	Iran	HOMER	COE	BAT	The economic system was BM/PV/BAT. CO ₂ decreased significantly.
[27]	China	HOMER	TC	BAT	The sensitivity analysis indicated the PV/wind/BDG/battery model was more economically viable than grid extension.
[28]	India	HOMER	ACS	BAT	Different operational strategies were compared. PV/W/BM/was the best configuration.
[29]		PSO, HS, Jaya	LPSP, EFF	BAT	Different configurations were modeled, simulated, and optimized. The best system was PV/W/BM/BAT.
[30]	Pakistan	HOMER	NPC		Four configurations were modeled; the best was PV/BM.
[31]	Pakistan	HOMER	NPC		Wind power was modeled with Weibull distribution. The biomass was dung cow manure. The grid-connected hybrid system proved to be more cost-efficient.
[32]	Pakistan		LCOE, NPV		The authors calculated the potential of solar, wind, and biomass energy to build a portfolio that supplied 10% of Karachi's peak demand.

The above studies show different methodologies for sizing a hybrid renewable energy system considering biomass. These works use classical optimization methods, genetic algorithms, artificial intelligence, and hybrid methods. As for specialized software, HOMER is the most widely used because it combines several technologies and metrics. All of them have advantages and disadvantages and it is not possible to say which one is the best. It is necessary to analyze the desired system and propose solutions according to its own characteristics (large- or small-scale system, available renewable sources, existence of real values of input variables, real or estimated load information, decision-making indicators, etc.) [33].

Although these works do not make explicit their relationship with the SDG7 indicators [13], they do contribute methodologically to their achievement. This work, unlike the previous ones, does establish the relationship and this implies the inclusion of on-grid HRES within the assessment, considering that the community on which the analysis was performed has the potential to be connected to the electricity grid.

Therefore, this article analyzes localized HRES [34] considering two of the perspectives indicated in the SDG7: affordability and reliability, via the levelized cost of energy (LCOE) and the probability of loss of power supply (LPSP). The electricity demand of a rural community located on the outskirts of the city of Barranquilla that is not connected to the grid served as the basis for sizing hybrid on-grid and off-grid HRES. The biogas produced by the community's pig farming was the main energy source. This source was complemented by varying the participation of solar and wind energy to obtain the most suitable system for the community. The heuristics used to solve the problem was based on Sparse Search (SS). The SS maintains a set of solutions taking into account a selection criterion [35] and combines them in a way that minimizes the levelized cost of energy (LCOE).

The article presents three cases. Case 1 considers an isolated HRES supported by a battery bank to supplement any energy deficit. Case 2 shows the use of diesel generation to supplement any energy deficit instead of batteries, and the third case assumes an on-grid HRES without batteries or a diesel generator to make easy the purchase and sale of energy between the prosumer and the grid.

Hence, the main contributions of this work are:

1. It establishes an explicit connection between the metrics commonly used in HRES modeling with SDG 7 indicators.
2. It takes into account the current economic activity of the community as the main source for energy generation complemented by other renewable sources.

The article is organized as follows: Section 2 develops the methodology implemented. Section 3 presents the results and discussion for the applied case. Finally, Section 4 presents the conclusions.

2. Materials and Methods

The available sources are solar, wind, and biomass, the latter obtained from pig manure. A battery bank (BAT) and diesel generation (DG) served as backup systems for the off-grid systems. Three cases were considered: Case 1. Standalone system SPV/W/BM/BAT. Case 2. Standalone system SPV/W/BM/DG. Case 3. Grid-connected system SPV/W/BM. In all three cases, the steps shown in Figure 1 were followed. The algorithm was coded in R, using the libraries cited in the references [36–40].

1. Data upload
2. Calculate the energy produced by biomass
3. Obtaining the annual unsatisfied electricity demand by simulation
4. Determination of solar wind hybrid pairs
5. Simulation of hybrid systems

Figure 1. Methodology.

The steps in Figure 1 are explained below:

1. The first step was to load the data for a Typical Meteorological Year (TMY) for the study location as well as the minimum and maximum hourly values of daily electricity demand. The TMY data were downloaded from PVWatts Calculator® <https://pvwatts.nrel.gov/pvwatts.php> (accessed on 19 July 2023). Particularly, wind speed at 10 m (corrected to 50 m) and beam irradiance were useful for this research. The demand data correspond with a low social class extracted from a characterization of the load profile according to socioeconomic stratification in the city of Barranquilla [41] but adapted to the locality of interest according to a set of assumptions described in the results section.
2. The second step was to calculate the energy produced by the biomass. We assumed a constant supply of electricity from biomass every hour. The procedure for estimating the production of biogas from the volatile solid content was as follows:
 - The animal mass M was determined, given by Equation (1), where N_i is the number of pigs of small size ($i = 1$), medium size ($i = 2$), and large size ($i = 3$). $P_{prom,i}$ is the average weight per each size of pig.

$$M = \sum_{i=1}^n N_i P_{prom,i} \quad (1)$$

- The number of average animal units UA was calculated considering that, for pigs, 1 UA = 113.7 kg. In other words,

$$UA = \frac{M}{113.7 \text{ kg}} \quad (2)$$

- Each animal unit (UA) generates approximately 19 tons of swine manure per year so the amount of total E swine manure was:

$$E = 19 \text{ UA} \quad (3)$$

- Swine manure has approximately 6% dry matter (MS) and between 70% and 80% volatile solids (SV). Hence, the total of volatile solids in the swine manure was:

$$SV = \%SV \times \%MS \times E \quad (4)$$

- Each kilogram of volatile solids generates up to 300 L of biogas, with 60% methane content (CH_4) so the amount of biogas B that can be obtained from SV is:

$$B = 300SV \quad (5)$$

- The amount of energy produced from biogas E_{biogas} , considering its minimum energy content of $MEC_{biogas} = 6 \frac{\text{kWh}}{\text{m}^3}$, was:

$$E_{biogas} = MEC_{biogas} B \quad (6)$$

- Equation (7) gave the capacity of the electricity-generating unit GUC for a generating unit efficiency η_{elec} of 30% and an annual operating time $t = 8000$ h:

$$GUC = \frac{E_{biogas} \eta_{elec}}{t} \quad (7)$$

3. The third step consisted of building an annual unsatisfied electricity demand (the demand not covered by biomass) curve using simulation. In each iteration, random values of hourly demand were taken between the minimum and maximum values of each hour of the daily demand curve. The procedure was performed a thousand times,

resulting in a normal probability distribution for the annual demand. The biomass power generation (of assumed constant capacity) was subtracted from the mean of the distribution, keeping the same variance. Thus, the demand curve not covered by biomass energy production was obtained.

- The fourth step consisted of determining the combinations of solar panels and wind turbines needed to cover the unsatisfied electricity demand. If f is the fraction to be covered by panels, and $1 - f$ the fraction to be covered by turbines, with $0 \leq f \leq 1$, the number of panels n_{pv} and the number of turbines n_w needed to cover the deficit were calculated according to the following equations:

$$n_{pv} = \frac{f \times Demand}{\eta \times P_{pv} \times AFS} \quad (8)$$

$$n_w = \frac{(1 - f) \times Demand}{E_w} \quad (9)$$

$Demand$ is the 95th percentile of the annual electricity demand curve; η and P_{pv} are the efficiency and power of the panels, respectively; and AFS (5.88 kWh/m²/d from PVWatts[®] is the daily beam irradiance for the place analyzed; AFS is 5.88×365) is the annual full solar hours. E_w is the amount of annual energy produced by one wind turbine, estimated by using Equation (10), with the turbine technical characteristics and wind speed from the TMY data. In Equation (10), v_c , v_r , v_f , and P_{nom} are the cut-in wind speed, the rated wind speed, the cut-out wind speed, and the rated power of wind turbine, respectively [42]. For values of f incremented by 10%, the pairs (n_{pv}, n_w) were found.

$$P_w = \begin{cases} P_{nom} \times \left(\frac{v^3 - v_c^3}{v_r^3 - v_c^3} \right) & v_c \leq v \leq v_r \\ P_{nom} & v_r \leq v \leq v_f \\ 0 & \text{otherwise} \end{cases} \quad (10)$$

- The fifth step was the simulation of each hybrid system. The TMY data contain the annual hourly estimate of irradiance and wind speed for the location of interest. These hourly data, grouped by month, were taken as initial samples to generate random data for these variables using bootstrapping. The amount of hourly energy produced by each hybrid system L_h was estimated using the equation:

$$L_h = n_{pv} P_{pv,h} \eta I_c / 1000 + n_w P_{w,h} + P_{bio} - D_h \quad h = 1, \dots, 8760 \quad (11)$$

Values of $L_h \geq 0$ indicate energy surplus (set S), and $L_h \leq 0$ indicate energy deficits (set D). I_c is the hourly irradiance [kW/m²] given by TMY corrected by the standard irradiance of 1 kW/m². The $I_c/1000$ quotient is actually an empirical approximation that, when added on a daily basis, approximates the daily full sun hours [h/d] of the analyzed area.

The minimum hourly energy deficit, L_i , at each iteration was calculated using Equation (12). The set L_i of all iterations produced a probability distribution of minimum deficits, which was used in cases 1 and 2 with isolated systems.

$$L_i = \min_h \{L_h\}, \quad \forall L_h \in D \quad (12)$$

System Optimization Model

The system was determined using the following cost minimization model.

$$\min LCOE$$

Subject to:

$$LPSP \leq 0.05$$

Equation (13) calculated the levelized cost of energy $LCOE$, and it represents the cost per unit of energy produced by HRES. $LPSP$ is the loss of power supply probability and represents the proportion of energy not supplied by the system in a year.

$$LCOE = \frac{ACS}{E_t} \quad (13)$$

The values of ACS , E_t , $LPSP$, and $LCOE$ were calculated considering the following three cases.

Case 1. Isolated hybrid system supported by a battery bank to supply the energy deficit.

The absolute value associated with a 5% probability of non-coverage of the deficit, $|L_{5\%}|$, was extracted from the probability distribution of minimum deficits mentioned above. This value was used to calculate the number of batteries needed to cover that energy deficit. The following equation shows the required variables:

$$N_{bat-max} = \frac{|L_{5\%}|}{V_{bat}CAP_{bat}DOD} \quad (14)$$

The ACS was calculated using Equation (15).

$$ACS = C_{inv} \left[\frac{r(1+r)^T}{(1+r)^T - 1} \right] + C_{O\&M} + C_{rep, bat} + C_{rep, inv} \quad (15)$$

$$C_{inv} = n_{pv}CI_{pv} + n_wCI_w + n_{bat}CI_{bat} + N_{inv}CI_{inv} + CI_{biomass} \quad (16)$$

$$C_{O\&M} = n_{pv}COM_{pv} + n_wCOM_w + n_{bat}COM_{bat} + N_{inv}COM_{inv} + COM_{biomass} \quad (17)$$

where COM is the annual operation and maintenance cost for each unit. CI is the investment cost for each unit (panel, wind turbine, battery, inverter, and biodigester). C_{rep} is the replacement cost of those units that have a useful life shorter than the operating time of the project, in this case, the batteries and inverters. UL_{bat} and UL_{inv} are the useful life for the batteries and inverters, respectively.

$$C_{rep, bat} = \left[n_{bat}N_{Rbat}CI_{bat} / (1+r)^{UL_{bat}} \right] * \left[r * (1+r)^T / [(1+r)^T - 1] \right] \quad (18)$$

$$C_{rep, inv} = \left[N_{inv}N_{Rinv}CI_{inv} / (1+r)^{UL_{inv}} \right] * \left[r * (1+r)^T / [(1+r)^T - 1] \right] \quad (19)$$

The costs associated with the $C_{biomass}$ biomass system comprise the biodigester, the power generation unit, and all additional implements and processes associated with its implementation. $LPSP$ was calculated as follow:

$$LPSP = \frac{\sum_{j=1}^{8760} L2_j}{E_t} \quad (20)$$

All values for the number of batteries N_{bat} such that $N_{bat} \leq N_{bat-max}$ were considered. For all terms (n_{pv}, n_w, n_{bat}) , the $LCOE$ was evaluated, and that term which made it as low as possible was chosen.

Case 2. Isolated hybrid system with a diesel generator to supply the energy deficit.

For each pair (n_{pv}, n_w) , $|L_{5\%}|$ was evaluated to determine the capacity of the diesel generator to supply this load. The next equations estimated the investment, operation, and maintenance costs:

$$C_{inv} = n_{pv}CI_{pv} + n_wCI_w + N_{inv}CI_{inv} + C_{biomass} + CI_{diesel} \quad (21)$$

$$C_{O\&M} = n_{pv}COM_{pv} + n_wCOM_w + N_{inv}COM_{inv} + COM_{biomass} + C_{fuel} \quad (22)$$

C_{fuel} is the annual fuel cost. CI_{diesel} is the investment cost of the diesel generator. The replacement cost C_{rep} is associated with the inverters if their useful life is less than the operating time of the project.

$$C_{rep, inv} = \left[N_{inv}N_{Rinv}CI_{inv} / (1+r)^{UL_{inv}} \right] * \left[r * (1+r)^T / \left[(1+r)^T - 1 \right] \right] \quad (23)$$

The E_t energy produced by the system was:

$$E_t = \sum_{h=1}^{8760} (P_{bio,j} + n_{pv}P_{pv,h} + n_wP_{w,h} + P_{diesel,h}) \quad (24)$$

For this case, the *LPSP* constraint was not considered, because the diesel generator supplemented the entire energy deficit.

Case 3. A grid-connected hybrid system.

The solar photovoltaic–wind–biomass hybrid system produced energy excess in some hours so users were able to sell energy to the grid, but in other hours, the production was insufficient, and users bought energy from the grid. Equation (11) gives information about when the user imports (L_h is negative) and exports (L_h is positive). As a result, it is possible to produce an annual profit from the sum of the monthly profits. The following equation allowed for the estimation of monthly profit:

$$Profit_t = |Imp_t - Dem_t| * CU - Exp1_t * Cv + Exp2_t * PB \quad (25)$$

Imp_t : It is the sum of the hourly energy import of the AGPE in the period t .

Dem_t : It is the monthly demand in the period t .

$Exp1$: The sum of the hourly energy export of the AGPE in the period t , $0 \leq Exp1 \leq Imp$.

CU : It is the unit cost of service provision, in \$/kWh of the serving supplier.

Cv : The commercialization cost, in \$/kWh.

$Exp2$: The amount of hourly energy exported of the AGPE in the period t , which exceeds Imp_t .

PB : Hourly exchange price of exported energy, in \$/kWh.

Therefore, *ACS* is given by:

$$ACS = \left| Profit_{t_{annual}} - \left\{ C_{inv} \left[\frac{r(1+r)^N}{(1+r)^N - 1} \right] + C_{O\&M} + C_{rep} \left[\frac{r}{(1+r)^N - 1} \right] \right\} \right| \quad (26)$$

VE_{annual} was calculated by adding the *VE* values corresponding to each month. In this case, the constraint for *LPSP* was not considered because the grid covers the energy deficit.

3. Results

The proposed methodology was applied in a village with 25 families living in plots along a dirt road, located in the city of Barranquilla, Colombia, 3 km from the nearest highway. The village was created approximately 40 years ago by a group of peasant families, who, since then, have struggled against the adversities and lack of opportunities in the area. They raise animals such as goats, chickens, and, to a greater extent, pigs, or, in some cases, grow vegetables, legumes, and tubers in the fertile areas of the site. Other families work in recycling and others have informal jobs. There is a basic education school in the area with three classrooms where general preschool, elementary, and high-school classes are taught, with approximately 80 enrolled children.

The community does not have access to electricity from the distribution networks, which causes economic, social, and health problems, among others. The little infrastructure

that the settlement has in terms of electricity service is the result of strategies generated by its inhabitants. For this reason, the service is discontinuous, unsafe, and restricted to a few.

The village has 24 houses grouped into 10 plots, a school, and a church. The number of inhabitants per plot is shown in Table 2.

Table 2. Inhabitants per plot (Information collected on site).

Plot	1	2	3	4	5	6	7	8	9	10	Total
Number of houses	4	2	2	2	2	2	1	4	2	3	24
Number of inhabitants	12	12	6	5	6	3	13	12	10	18	97

Some houses have appliances because they have diesel generators to supply power, but others do not. The school has a photovoltaic system that supplies electricity for educational activities. For this reason, the load profile of each household was assumed to be the consumption of a family of four with a set of basic household appliances (TV, cell phone, blender, refrigerator, iron, washing machine, computer, fan, and stereo). Figure 2 shows the minimum and maximum values of hourly demand per day of the assumed load profile.

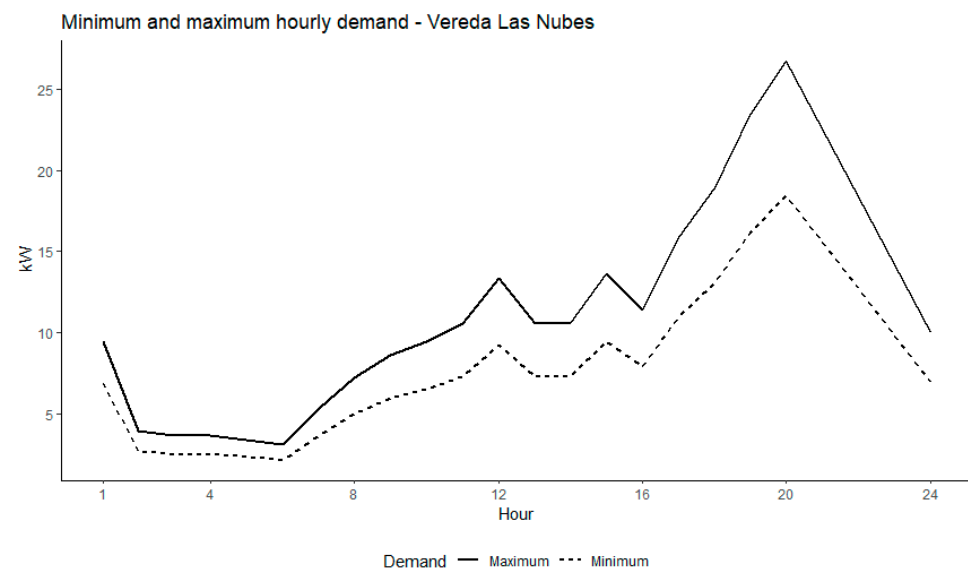


Figure 2. Load profile “Vereda Las Nubes”.

The biogas energy produced by the pig excreta was used to cover the load profile described in Figure 2. The values used to estimate this energy are shown in Table 3.

Table 3. Classification of pigs by size (Information collected on site).

Pig Size	Plot Number of Pigs							N_i	P_{prom_i} , kg	Average Excreta, kg	Total Excreta, kg
	1	2	3	4	5	6	7				
Large	8	10	2	6	2	21	2	51	60	3	153
Medium	42	27	0	10	0	80	36	195	25	2.5	487.5
Small	10	43	14	10	5	26	10	118	15	1.5	177
Total								364			817.5

The first column of Table 3 shows a size classification of the pigs. The second column, subdivided into seven sections, shows the number of pigs by size in seven of the plots. The third column shows the total number of pigs by size and their sum total. The fourth column shows the average weight in kg per pig size. The fifth and sixth columns show the mean

weight and total weight of the excreta, respectively. These values allow the calculation of biogas using Equations (1)–(7).

$M = 9705$ kg, $UA = 85.4$ UA, $E = 1621.77$ ton/year, $SV = 68.11$ ton/year, $B = 55.98$ m³/day, and $E_{biogas} = 122,605.65$ kWh/year. The capacity of the electricity generating unit *GUC* with an efficiency of 30% and an annual operating time $t = 8000$ h was $GUC = 4.6$ kW. According to these results, a 5 kW generating unit with a single-phase alternating current output was chosen. Table 4 shows the generating unit characteristics.

Table 4. Biogas generator characteristics.

Parameter	Value
Maximum power (biogas)	5000 W
Frequency	50 Hk/60 Hz
Nominal voltage	400 V/300 V
Nominal current [A]	8.3
Velocity	1500 rpm/1800 rpm
Engine oil capacity [l]	0.55
Gas consumption [m ³ /h]	1.4
Useful life [years]	20

The energy provided by biogas was not enough to cover the demand and therefore we analyzed the annual unmet demand to be met with the other sources available in the area, i.e., solar and wind energy. Figure 3 shows the probability distribution of the remaining annual demand after supplying the energy produced by the biodigester. To size the complementary system, the value of 37,020 kWh was taken as a reference. This implies that 5% of the annual demand may not be covered by the entire hybrid system. The curve of Figure 3 was obtained using simulation.

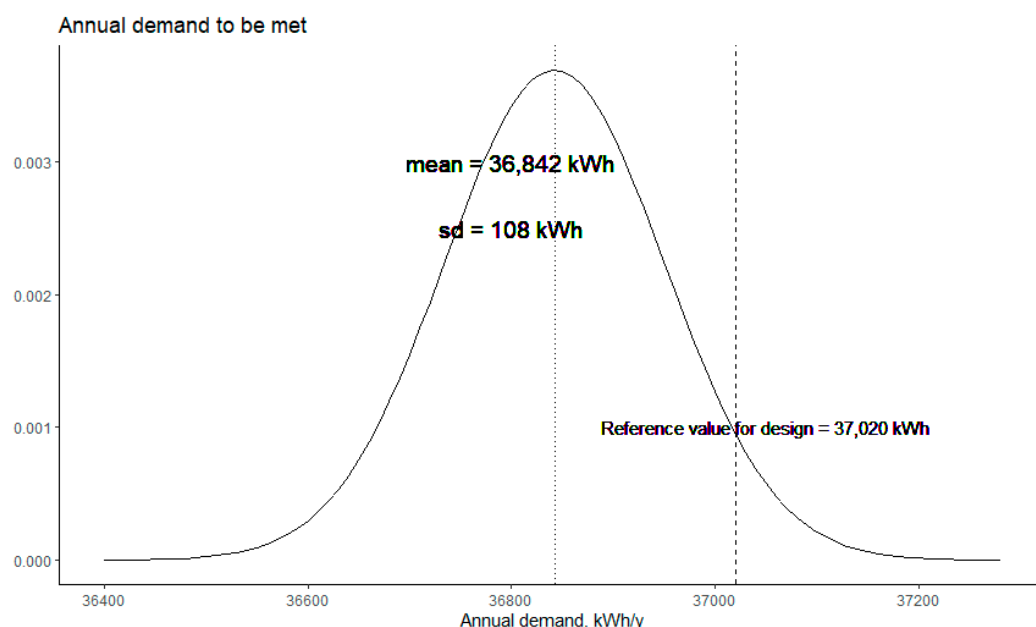


Figure 3. Remaining demand to be covered using solar panels and wind turbines.

The complementary system consisted of a combination of panels and wind turbines in different quantities to obtain the design capacity shown in Figure 3. To calculate these combinations, it was necessary to determine the energy provided by one panel and one wind turbine. The unit contributions of these technologies are shown in Figure 4.

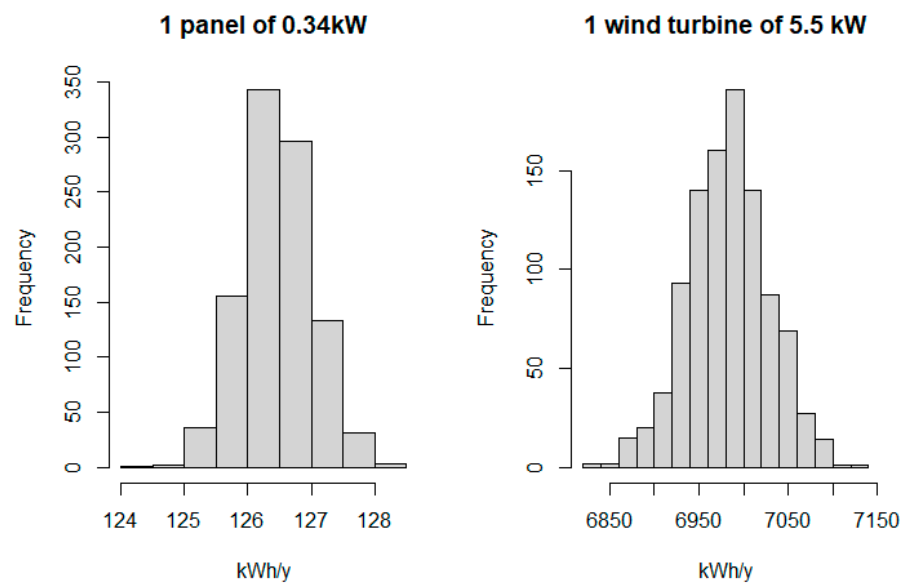


Figure 4. Distribution of annual electricity produced by one panel and one wind turbine.

Figure 4 shows the annual electricity produced by a single photovoltaic panel (see characteristics in Table 5) and a wind turbine from the TMY data ($10^{\circ}59'58.1''$ north latitude $74^{\circ}51'21.4''$ west longitude).

Table 5. Panel specification <https://www.erasolar.com.cn/product/1011589699020197888.html> (accessed on 16 August 2023).

Parameter	Value
Maximum power [W]	340
Maximum voltage [V]	38.5
Maximum current [A]	8.84
Open circuit voltage [V]	46.4
Short circuit current [A]	9.45
Temperature coefficient of open circuit voltage [%/°C]	−0.29506
Temperature coefficient of short circuit current [%/°C]	0.08558
Normal operating temperature [°C]	40
Efficiency [%]	17.5
Useful life [year]	20
Area [m ²]	1.94

Hybrid systems with different combinations of these components were simulated to meet the annual community demand not covered by biomass generation (Figure 3). These combinations are shown in Table 6.

Table 6. Combinations of 0.34 kW panels and 5.5 kW wind turbines to meet unsatisfied demand.

Number of PV panels (PV system)	0	7	13	20	26	32	39	45	51	58	64
Number of wind turbines (Wind system)	6	5	5	4	4	3	3	2	2	1	0

Each of the hybrid systems defined in Table 6 can cover, together with the biomass generation, the annual energy needs defined in Figure 3. This means that on an hourly basis, they produce energy for self-consumption, but they can also produce surpluses if the energy produced by the system exceeds the hourly demand. Additional energy may be

required because the hybrid system does not produce enough to meet demand. However, the annual demand is expected to be met.

The hourly turbine power P_w was estimated by using Equation (9). An Enair (<https://www.enair.es/es/aerogeneradores/e70pro>, accessed on 16 August 2023) E70Pro with an output power of 5500 W was used. Table 7 shows the specifications.

Table 7. Wind turbine specifications.

Parameter	Value
Nominal power [W]	5500
Sweep area [m ²]	145
Cut-in wind speed [m/s], v_c	2
Nominal speed [m/s], v_r	11
Cut-out wind speed [m/s], v_f	12
Useful life [years]	20

Surpluses occur when the energy produced by the hybrid system exceeds the community's hourly electricity demand. If the community does not have a storage system or is not connected to the grid, these surpluses are lost, and with them, the opportunity to generate income or save costs; in the case of being connected to the conventional electricity grid, these surpluses are considered exports of electricity to the grid. In some electricity systems, there is the possibility to offset energy imports with surpluses; the net amount may then represent a value to be paid because the energy imports exceed exports or a credit in favor because more energy is sent to the grid than is imported.

Next, the optimal economic performance of HRES is analyzed by considering three cases:

1. The community uses an isolated hybrid system, that is, there is no connection to the grid, and it has batteries to store any surplus.
2. The community is not connected to the grid and the deficit is covered using a diesel-based generation system.
3. Finally, the community is connected to the grid and is allowed to send its surpluses to the grid.

The three cases have common costs: investment, operation and maintenance, and the replacement of the solar, wind, and biomass systems. Case 1 adds the cost of the batteries while Case 2 adds the costs of the diesel generator instead. These costs are in Table 8.

Table 8. Unit costs.

Costs	CI (\$)	COM (\$/year)	C_{rep} (\$)
Solar panel	103.1	4.84	0
Wind turbine	7777.8	427.8	0
Biodigester and components	400	16	0
Generating unit for biogas	1000	40	0
Inverter	3013.3	30	136.4
Battery	179.7	8.9	8.1
Diesel generator	5555	650	0
Diesel [\$/L] 0.5			

Case 1. Isolated hybrid system supported by a battery bank to supply the energy deficit.

To size the number of batteries required for each combination given in Table 5 of the hybrid system, the monthly hourly wind speed and irradiance were simulated using the corresponding values contained in the TMY data; from these values, the hourly electricity

production of the number of panels and wind turbines given by each combination in Table 5 was estimated. These were added to the hourly electricity production of the biomass-based system and the total was subtracted from the simulated hourly electricity demand (Equation (11)); negative values result in hourly deficits, indicating that the demand is greater than the supply, while positive values indicate hourly surpluses. For each month, the minimum of the deficits was identified (Equation (12)), and of these, the minimum per iteration. This procedure made it possible to construct a probability distribution of the hourly minimums for each hybrid system and thus define a battery sizing that at least 95% of the time was capable of supplying the hourly deficit presented. This implies a probability of not covering the hourly demand of a maximum of 5%. Figure 5 shows the distribution of the extreme hourly deficits for a system with 26 panels and 4 turbines, where the value used for sizing is marked. The maximum number of batteries to cover the 21 kW requirement is 11. This value was the same for all the hybrid systems.

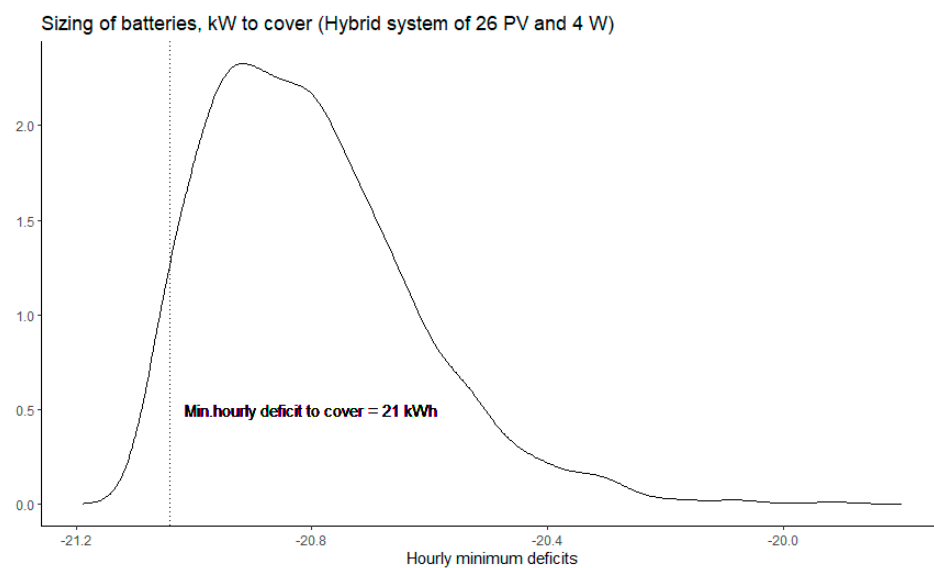


Figure 5. Curve of minimum hourly deficits for (26.4) system.

The next step was estimating *LPSP*. For each pair (n_{pv}, n_w) , we found all possible battery values n_{bat} such that $LPSP \leq 0.05$. For ternaries (n_{pv}, n_w, n_{bat}) , we recorded the 1st percentile of the *LPSP* distribution (99% of the values are greater) and *LCOE*. Obviously, a greater number of batteries generated a more reliable system. It was found that a hybrid system with 64 panels, no turbine, the biodigester, and 11 batteries produced the lowest *LCOE* among all the hybrid systems analyzed, being $LCOE = 0.09$ \$/kWh.

Case 2. Isolated hybrid system with a diesel generator to supply the energy deficit.

In this case, a 20 kW diesel generator was selected to operate for 6 h a day, when there was no solar or wind resource. For a year, that would be 2190 h of operation. If the fuel consumption of the generator is 5 L/h, the annual consumption will be 10,950 L. The hybrid system with 64 panels and no turbine (pair (64 – 0) in Table 9) presented the minimum *LCOE* with a value of 0.054 \$/kWh. Table 9 shows the *LCOE* values for the panel and turbine combinations considered. All systems overcome the constraint imposed by *LPSP*.

Case 3. Grid-connected hybrid system.

In this case, electricity exchanges between the prosumer and the grid were allowed, which were valued according to the prices shown in Table 10 and Equation (25). These average prices were taken from the Colombian market assuming that they were constant throughout the year. This assumption does not affect the overall results obtained.

Table 9. LCOE values for feasible systems including diesel generation.

$(n_{pv} - n_w)$	LCOE_mean
(0 – 6)	0.110
(7 – 5)	0.103
(13 – 5)	0.104
(20 – 4)	0.096
(26 – 4)	0.097
(32 – 3)	0.088
(39 – 3)	0.089
(45 – 2)	0.079
(51 – 2)	0.079
(58 – 1)	0.068
(64 – 0)	0.054

Table 10. Prices used to value energy exchanges.

Agent	Cost [\$]
Unit cost of service provision, \$/kWh	$C_U : 0.109$
Commercialization cost, \$/kWh	$C_v : 0.015$
Power exchange price, \$/kWh	$P_B : 0.04$

The annual performance indicates that energy exports to the grid were higher than imports, resulting in energy credits (see Figure 6). Figure 6 shows box-and-whisker plots of the profits for each pair of the hybrid system, keeping in mind that these were simulation results. Note that the interquartile range was narrow, indicating that there was little variation between quartiles. The credits favor systems where there are more wind turbines. This is explained because, although all the hybrid systems were sized to supply the same demand (37,020 kWh), when panels are used, there are more energy imports from the grid (which reduced the utility) because the electricity demand (Figure 2) is higher during non-sunny hours (at this location, the sun is available from 6 a.m. to 6 p.m.).

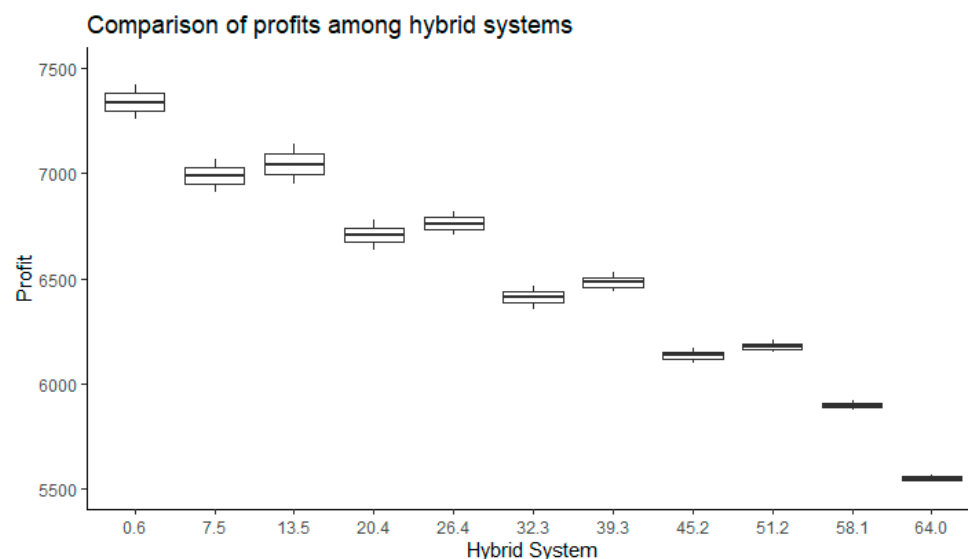


Figure 6. Comparison of profits produced by hybrid systems. X-axis is $(n_{pv}.n_w)$.

However, when the ratio of profit per energy produced was used, the result favored more the use of solar panels (Figure 7). The systems with a higher share of wind turbines produced more energy than the solar panels (the annual average produced by the system without panels and with six turbines produced 91,870 kWh/year, while the system with 64 panels and no turbines produced an average of 58,970 kWh/year). Note that the indicator

practically does not change (narrow interquartile ranges in the box-and-whisker plots) despite being the result of a simulation.

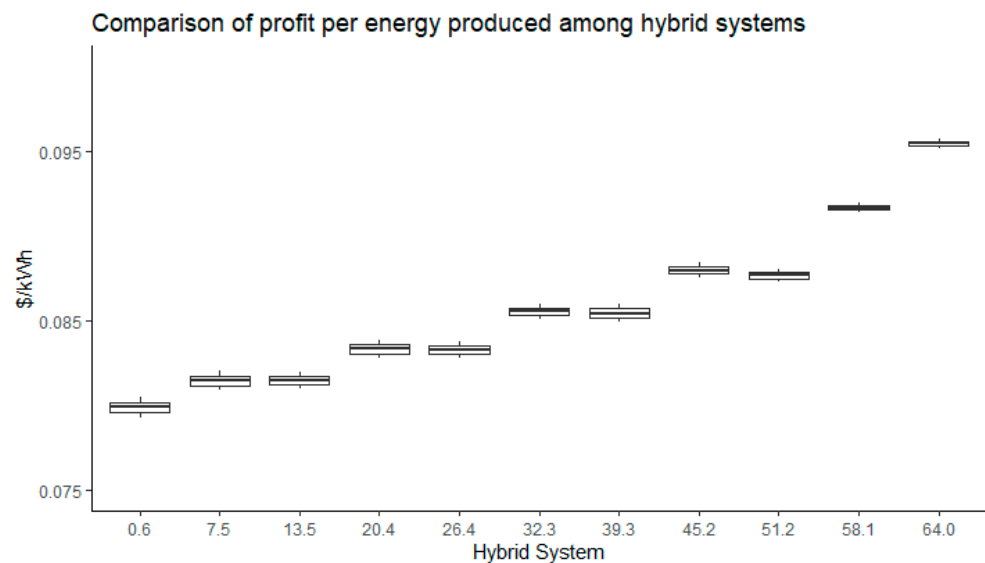


Figure 7. Comparison of profits per kWh produced by hybrid systems. X-axis is (n_{pv}, n_w) .

On the other hand, when reviewing the *LCOE*, as wind turbines were replaced with panels for electricity production, the costs involved were increasingly lower, which favors the economic performance of solar panel systems (Figure 8). The *LCOE* calculation for the best system was virtually unchanged in the simulations.

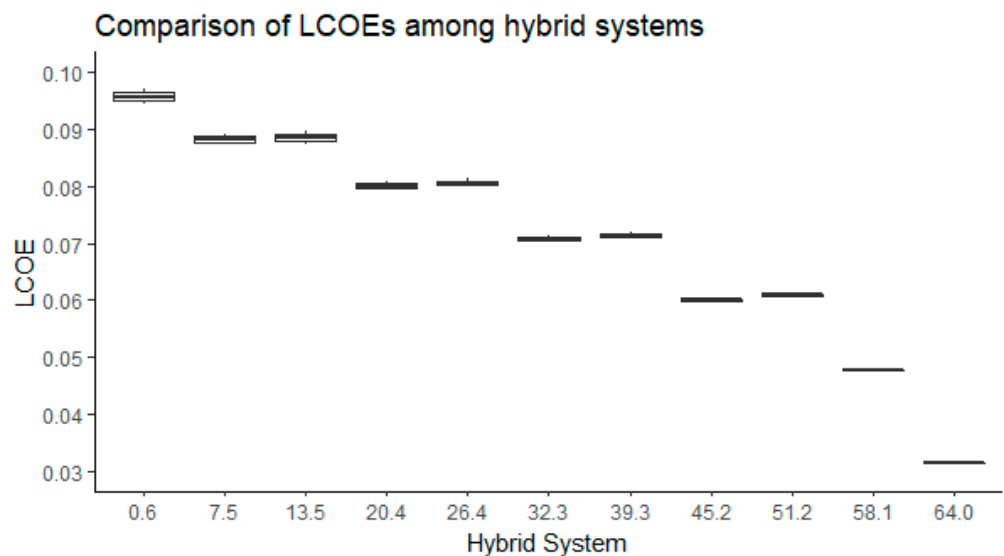


Figure 8. Comparison of the LCOE of hybrid systems.

All three systems had in common that the lowest LCOE value was obtained using 64 panels and no turbine; this is due to the location of the site, which allows high solar radiation throughout the year, and the fact that winds are strong only during some months of the year. However, the on-grid systems were the best performers. They can produce benefits and have the lowest LCOE.

4. Discussion

There are communities living in remote locations where access to electricity has not been easy because conventional electricity systems that use transmission and distribu-

tion networks to transport energy have not been able to reach them. In these cases, the technological development that has enabled the use of renewable energies has played an important role in achieving SDG7. Much of the existing scientific literature has focused on the use of isolated hybrid systems based on renewables that use battery backups or diesel generators to reduce this gap.

However, there are other cases of communities that have not connected to the conventional power grid because of a purchasing power problem, but not because their connection is hindered by physical or geographic barriers.

For these vulnerable communities, renewable energy alone is not the solution due to the high costs involved. This paper demonstrated that the LCOE of on-grid systems is lower than that of off-grid systems. The community alone cannot solve this problem, so the government must finance the solution; the self-sustainability of the operation in the long-term increases if conditions and incentives are also provided for self-generation and energy exchanges with the power grid [11], to which the community should be connected if possible. This coincides with the conclusions obtained in other studies [43,44]. Policies can also be implemented to train community members in the operation and maintenance of the systems, thus generating new sources of employment and improving their social, economic, and health conditions, among others.

For future work, the results obtained in this work can be compared with the results that can be obtained using the hybrid methods that combine heuristics, algorithms, and characteristics of the problem that are considered highly effective.

5. Conclusions

This article analyzed how to supply the electricity needs of a vulnerable community without access to electricity in the framework of SDG7. Hybrid systems that integrate renewable sources such as solar and wind with biomass generated by the community's economic activity were sized. Off-grid hybrid systems supported by batteries or diesel generation were reviewed considering that the community is not connected to the city distribution grid. On-grid systems were also analyzed under the assumption of grid connection. The results indicate that the community's electricity demand can be supplied using energy from the pig biomass currently available to the community, integrated with solar panels and grid connection. Energy from wind turbines makes supply more expensive. Similarly, working with off-grid systems produce shortfalls that must be covered using backup systems such as batteries or diesel generation, which makes electricity supply more expensive. This conclusion is supported by the leveled cost of energy of the different systems evaluated, which shows that hybrid systems with biomass and solar energy presented the best performance when connected to the grid. Grid connection is important in contexts where regulation allows energy exchanges between prosumers and the grid.

According to these results, to make progress toward achieving SDG7, communities without access to energy should not work with off-grid systems as long as they can be connected to the grid. This implies that governments should support this by creating the regulatory and infrastructure conditions for connection to distribution networks and provide support for the integration of adequate renewable sources, seeking to turn vulnerable communities into prosumers.

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Nomenclature

List of Abbreviations

ACS: annualized cost [\$/year]	LCOE: levelized cost of energy [\$/kWh]
AEFA: artificial electric field algorithm	LPSP: loss of power supply probability
BAT: battery	MDVP: movable damped wave algorithm
BM: biomass	MHP: micro-hydro power
CEQ: carbon emission quantity	MOGA: multi-objective genetic algorithm
COE: cost of the energy	NPC: Net present cost
DE: differential evolution algorithm	NSGA II: non-dominated sorting genetic algorithm II
DG: diesel generator	PHS: pumped-hydro storage
DPSP: deficiency of power supply probability	PSO: particle swarm optimization
EFF: excess energy fraction	PMBSA: Pareto multi-objective backtrack search algorithm
FF: firefly algorithm	SOA: seagull optimization algorithm
GA: genetic algorithm	SDG: UN Sustainable Development Goals
GUC: capacity of the electricity generating unit	SMA: slime mould algorithm (SMA)
GWO: gray wolf algorithm	SPV: solar photovoltaic system
HHO: Harris hawks optimization	QRUN: quantum Runge–Kutta
HOMER: Hybrid Optimization of Multiple Energy Resources	TC: total cost
HS: harmony search	TMY: Typical Meteorological Year
HRES: hybrid renewable energy systems	W: wind
	WOA: whale optimization algorithm

List of Symbols

AFS: annual full solar hours	L_h : hourly energy produced by a hybrid system [kWh]
B : amount of biogas [m ³ /day]	M : livestock mass [kg]
CAP_{bat} : battery nominal capacity [Ah]	MEC_{biogas} : minimum energy content on biogas [kWh/m ³]
$C_{biomass}$: total costs associated with biomass [\$]	T : project lifetime [year]
CI: unit investment costs [\$]	n_{bat} : numbers of batteries
C_{inv} : total investment cost [\$]	n_{pv} : numbers of panels
C_{fuel} : the annual fuel cost [\$]	n_w : numbers of wind turbines
$C_{O\&M}$: total annual operation and maintenance cost [\$/year]	N_j : number of pigs
COM: unit annual operation and maintenance costs [\$/year]	N_{inv} : numbers of inverters
C_{rep} : unit replacement cost [\$]	P_{bio} : monthly energy generated from biomass [kWh]
DOD: depth of discharge of the batteries [%]	P_{prom} : average weight of pig [kg]
E : amount of swine manure [ton/year]	P_{pv} : panel unit power [W]
E_{biogas} : biogas annual energy [kWh/year]	P_w : wind turbine unit power [W]
E_t : annual energy production [kWh/year]	r : interest rate
E_w : unit annual energy of wind turbine [kWh/year]	SV : volatile solids [ton/year]
f : energy fraction covered by panels	UA: average animal units
GUC: capacity bigas generating unit [kW]	V_{bat} : batteries operating voltage [V]
I_c : hourly irradiance [kW/m ²]	η : panel efficiency
L_i : monthly energy deficit in month i [kWh]	

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