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Load prioritization technique to guarantee the continuous electric supply for essential loads in rural microgrids

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Abstract: Microgrid (MG) is one of the practical and best concepts to provide energy access to rural communities, where electric grid extension is not techno-economically feasible. Since the trend of load consumption is not uniform with low load factor in the rural area, the required rating of the system becomes very high. Similarly, the generation is fixed for these MGs, whereas the load increases continuously over time. Such a system faces supply deficit issues triggering a high number of interruptions that may cause frequent blackouts. Hence, rolling blackout and load clipping techniques are preferred during the peak load period in most of the rural MGs. These issues lead to an unreliable power supply and low satisfaction level in rural communities. This paper presents the load prioritization technique to guarantee the continuous supply for the essential loads within the community. A day-ahead energy allocation technique is mathematically formulated and optimized to maximize the total hours of energy served. This technique maximized the hours of energy served to the load with higher priority followed by the load with lower priorities. From this study, it is found that the proposed strategy helps to improve the hours of energy served in the overall system, by improving the state of charge (SoC) level of the battery system. Besides, the user satisfaction level has been improved by 5% through 100% of continuity for the essential loads.

Keywords: Demand Side Management; Microgrid; Rural Electrification; Renewable Energy

1. INTRODUCTION

With the increases in energy demand and fixed generation capacity, most of the countries have started to manage the loads using various Demand Side Management (DSM) strategies, where the growing demand has pushed the generation to its limits [1]. DSM technique has been considered as one of the ways to stretch the limits of the system a bit further. Moreover, it helps in reducing the stress on the existing power grid by balancing the demand and available energy resources locally. This technique of managing demand directly from the load side has already attracted a great deal of attention from research as well as the general communities [2-6]. With an increase in demand, the systems designed for specific load handling capacity are certain to face the issues of voltage and frequency fluctuation that may lead to unreliable energy supply, and even system blackout in some cases [7]. Especially in developing countries like Nepal, the standalone MGs have played an important role in providing electricity in most of the rural areas [8, 9]. However, the generation capacity of Renewable Energy Resources (RES) unit is limited [10, 11]. Taking the case of Nepal, more than 55 MW of electricity has been produced from RES like mini/micro-hydropower plants (MHP), solar and wind energy, which fulfils the electrical demand of 3.6 million households [12]. However, different MGs in Nepal face the issues on availability, as the average availability of electricity varies from eight hours in a day in minimum for the solar/wind MGs, and up to 23 hours of supply for the Micro Hydropower Plant (MHP) [13]. This indicates a deficit of supply in these rural MGs. Pieces of literature show that researches have been conducted for design of the reliable system and to reduce the cost through various means like interconnection of MGs [14], optimization of generation technology [7, 15-17], finding the optimal way to share surplus energy between gridlines [7], load management [6, 13] etc. However, these researches only focus on managing energy from the generation side. Furthermore, the supply of these isolated MGs relies on RESs, where generation itself is on the question of a reliable source.

Generation of RESs like solar and wind has highly fluctuated in nature. Its generation depends upon various factors like weather, temperature and swing based on climatic change. On the other side, demand also has its own nature that may vary on hourly, daily basis and/ or seasonal basis. The demand also based on some occasional events like community and social norm programs. Basically, the load hinge upon the type of consumers in the region. MGs in developing countries have a majority of the residential load, and the nature of load pattern in these MGs are highly tilted toward residential load consumption. Each year the connection to a new household is guaranteed that leads to an increase in demand. Likewise, consumption in a household in Nepal has increased by 38% for the last years [18]. Similarly, the gap between demand and generation has been increasing at a rate of 13.3% annually [19]. Due to a significant increase in demand in recent years, the current supply infrastructures are being inadequate, resulting in load shedding during peak hours and eventually the generation systems are suffering from various problems [20]. As adding generation to the system could be a solution but this with further increase the cost of energy. As these systems are already facing problem on revenue generation due to high operation and maintenance cost [20], with development of new technologies, the current energy market could introduce a smart concept that provides the opportunities to understand the load consumption in user perspective. In the current world, the energy systems are getting smarter, which can help in the optimizing uses of energy concerning the system requirement, without affecting the user comfort and stratification. This implements the system management easier allowing system conserve energy and keep continuity in the supply. Concept of smart load control to shed or/ and reduce the consumption of the load appliance, can reduce the peak demand by 30%, and its Implementation can help to reduce the grid's operation costs and increase the system reliability [2].

DSM strategy for managing the loads include everything that can be done in the demand side of an energy system [1]. This strategy sorts from replacing the high demand load equipment with a lightweight load to installing complex load management controller. The strategies can be categorized based on its impact and time-based solution that it can provide on the electrical system. With the implementation of a smart system, commonly used techniques focus on altering the load profile to keeps the balance between load and supply. With the customers' permission, the DSM can change the load shape of the load by reducing the total demand of the user, during peak hours though shifting or shedding. Six broadly discussed and implemented technique on DSM includes; peak clipping, valley filling, load shifting, strategic conservation and strategic load growth [5]. These strategies can be implemented by analyzing the system's load profiles of the user to manage peak demand, reduce the cost of energy, conserve energy for strategic load managing and in some case even to increases demand [3]. To shift, shed or adjust the consumption, understanding of these loads become very important. Studies tend to perform the load classification based on user preference, energy demand and flexibility [6, 13]. As loads are the appliances being used in each household, researchers classified the appliances as static and dynamic [21], shed-able and unshed-able [6], critical and uncritical [22], programmable and dimmable load [2] etc., based on their power consumption and behaviour of their use. Performance analysis of these load of individual home-based-appliances helps to achieve the good demand response of the energy management system [3]. Through the control of appliances, the demand curve can be shaped by reducing the total load on the distribution system during peak periods, and shift the load in a better time. The direct control of load has an effect on user comfort and might question the satisfaction level that the system can provide. Hence proper prioritization of appliances in either in terms of time-based and device-based strategies can help to achieve maximum user comfort in the defined budget limit [21]. In [22], the classified load are further divided into priorities given to each appliance where the system shed and shift load to satisfy the demand. This load control is defined by the kVA rating of each equipment, and dividing the loads into base and shed-able load, with control of combined heat and power (CHP) equipment as the shed-able load can keep the demand below the targeted load [6].

Considering the demand study conducted in Nepal [13], a framework on prioritization of load has been presented, which is based on the user's preference and storability of the load. This classification of load helps to improve user satisfaction [21]. With classification on each load, identifying control mechanism and the proper load shedding/ shift time are also equally important. A question on what basis the control action must be performed should be answered, e.g. various strategies are implemented such as: to reduce the cost of energy for individual consumer [5], to perform peak shaving, or to minimize carbon emission [23], to maintain battery life [24] etc. Optimization process such as Heuristic Optimization [5], Genetic Algorithm Optimization, Hybrid Bacterial Foraging [25], Whale Optimization [26] and Fuzzy Logics [27] etc. are implemented to find the optimal point for timely control action while maintaining user satisfaction at the same time. However, these techniques require high computational speed and programming skill, which is a lot to ask for as simple energy management system such as found in rural locations. Controlling each appliance is not an easy task, as every equipment needs to be connected to the controller through cable or communication, especially in rural area. A control system implemented in Bhutan uses the light indicator to indicate the health of the grid and prevent the use of the heavy appliance in households during peak hours [28]. Similarly, a management strategy has been implemented on a solar MGs in Baidi, Nepal, which adopted package-based system that provide three different levels of packages based on user demand and affordability [29]. With the limited generation capacity of isolated MGs, the priority-based load control strategy comes in popularity in such remote locality.

Following the complexity in controlling the load, this research put up an idea to conserve the energy for loads that are prioritized by the consumers by shedding, and can help to increase the reliability of the system via automatic control with a simpler optimization algorithm that allocates optimum energy for different levels of loads. For a healthy operation, the proposed method traces the working pattern of each appliance in each household based on the three different levels of priority. The individual smart controller is adjusted to follow the consumption characteristics of the appliances in each level avoiding the requirement for a system to control and communicate with each appliance. The technique follows an exhaustive search algorithm allowing the central controller to identify the optimum point of shedding of lower priority load for better user satisfaction. This research article is organized as follow: Introduction and previous studies have been discussed in Section 1. Section 2 describes the demand and generation study in case of MGs in Nepal followed by mathematical model development of this study. The outputs of this study are presented in Section 3, and the conclusions have been discussed in Section 4.

2. METHOD AND MATERIALS

2.1. Case scenarios

Fluctuation in generation and demand has a huge impact on system stability as well as reliability. Lack in the generation or higher electricity demand can lead to supply deficit, reducing the availability of system supply for hours. In case of a system relying on RES like solar and wind energy, the generation faces major fluctuation than load. The change in climate and surrounding temperature could reduce the generation significantly. Isolated MGs in Nepal has major two sources of generation (i.e. solar and MHPs). Generation in MHPs have a variation in a long-time frame as variation in the generation is affected based on location and season change. Whereas the solar has a high fluctuating nature. Uncertainty of the solar generation might occur at multiple periods, ranging from a few minutes to a few hours and even in some cases to days ahead [30]. A one-year graph of solar generation (i.e. generation by 800 W solar PV) for a random location of Nepal can be seen in Figure 1. The variation can be observed in the Figure, as the generation of a system can decrease to 35 %, and can last up to two days straight. From the one-year span, the seven days sample has been highlighted to shows the generation pattern over one week. Likewise, to maintain the continuity in the supply solar-based MGs have Energy Storage System (ESS) as a backup.

Figure 1: One-year PV generation (7 days low generation study)

On the other side, the potential appliances that can be connected to a solar-based MGs were identified via a survey study for the Nepalese MGs, and presented in Table 1 [13]. The study considered 13 major appliances that were in use in the isolated solar MGs. In this study, the load consumption has been scaled for one particular household with a maximum demand of 725W and average energy demand of 4.5 kWh per day. All of the appliances are categorized into 3 different levels based on their priority in the daily life of consumers. The highest priority of load consists of the appliances, which are required on a daily basis, whose importance cannot be neglected, as their service of hours should always be maximum. Similarly, the second or medium priority load consists of the load appliances used for entertainment and information purpose, like a TV set, Radio, Laptops and has lower energy consumption. Finally, the lowest priority is given to the equipment whose demand is high and are used to improver life standards like woofers, VCRs and Oven, Fridges etc. In Figure 2 (a), the hourly uses of these appliances within a household can be easily observed. Similarly, Figure 2 (b) presents the same consumption based on different priority levels. It can be observed that the appliance-based study can be further simplified when classified. The residential demand for the simulation has an average demand of 4.8kWh per day. Here, the highest priority load covers 32.36%, second or medium covers 35.28% and lowest priority covers 32.36 % of the total load respectively. From the Figures 2 (b), it can be observed that the first priory load has a continuous demand throughout the day, whereas the second priority load has at least 12 hours demand period, and load with least priority has an average 6 hours demand period per day.

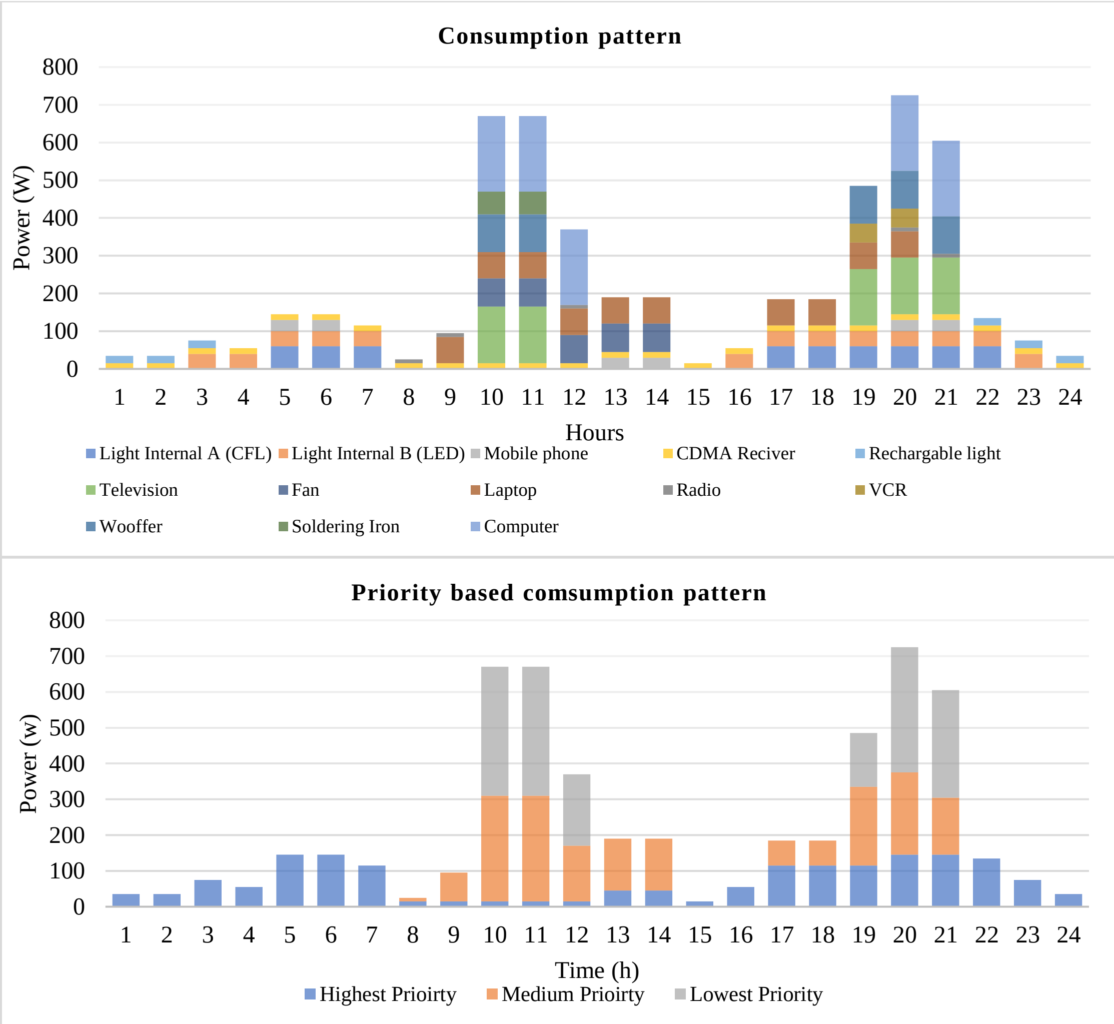

Figure 2: (a) Consumption pattern on appliance level, (b) Consumption pattern based on priority

Table 1: Prioritization of load demand

Appliances	Power (W)	Quantity (nos.)	Total Power (W)	ToU of appliances per day	Priority	Time of Operation of Each priority
Light Internal A (CFL)	15	4	60	9	Higher Priority (Level 3)	24
Light Internal B (LED)	10	4	40	13		
Mobile phone	15	2	30	6		
CDMA Receiver	15	1	15	24		
Rechargeable light	10	2	20	6		
Television	150	1	150	5	Medium Priority (Level 2)	12
Fan	75	1	75	5		
Laptop	70	1	70	10		
Radio	10	1	10	5		
VCR	50	1	50	3	Lower Priority (Level 1)	6
Woofer	100	1	100	3		
Exhust fan	60	1	60	3		
Computer	200	1	200	3		

2.2. Problem Formulation

Generally, the prioritization techniques help to improve the user comfort and satisfaction of the services [21]. This allows the user to run the selected appliances for higher periods as a trade-off with giving up lower priority appliances to avoid blackout on the system. The proposed algorithm sheds the connection of supply based on load priority in a way to maximize the hours of load-served for the load with the highest priority. Load shedding technique for such applications can be tested through a model-based simulation to observe the hours of load demand served. Mathematically, the ratio of the hour for services can be calculated with the help of Equation 1. Here, $P_{Served\ hours}$ is the total hours of power served, and $P_{Demand\ hours}$ is the value for hours for power demanded obtain through simulation of solar MG model. *N* defines the total number of equipment as *n* shows the appliances in function. During the priority set, *n* defines all of the appliances considered within the specific priority. Since the hours of power served for all of the appliances that must be maximized, each load priority may change from user to user. By observing the total hourly demand of each priority level, the values of *x*, *y* and *z* are set, which defines the priority of each appliance. Considering the three different appliances Load A, Load B and Load C with hourly consumption per day as *t1*, *t2* and *t3*, the priorities are generally set by using Equation 2.

$$\sum_{n=1}^{N} \frac{P\ Serverd\ hours(n)}{P\ Demand\ hours(n)} \tag{1}$$

$$x = \frac{t1}{t1+t2+t3}, y = \frac{t2}{t1+t2+t3}, z = \frac{t3}{t1+t2+t3} \tag{2}$$

Here, *x* defined the priority set of Load A, *y* for Load B and *z* for Load C. The user's satisfaction is calculated based on hours of energy-served and total energy-served on each priority set. Maximum user satisfaction is obtained when the sum of *x*, *y* and *z* from Equation 3 tends to 1 (i.e. when all of the demand is served). However, as the available energy decreases, this value will be hard to obtain. So, the system must identify the optimal point where the least priority load can be shaded without affecting the hours of energy-served on other loads.

$$x\frac{Energy_{Served}(n-1)}{Energy_{Demand}(n-1)} + y\frac{Energy_{Served}(n)}{Energy_{Demand}(n)} + z\frac{Energy_{Served}(n+1)}{Energy_{Demand}(n+1)} = Users\ satisfaction \tag{3}$$

$$User\ Satisfaction \leq 1 \tag{4}$$

2.3. Proposed Architecture and Strategy

To address the discussed problem in MGs, an optimim approach could be implemented. In this study, a DSM based strategy has been proposed and implemented in simulation environment with two modes of control: a central controller at the generation point, and a remote load-controlling agent (smart meter) for each consumer. The remote controller monitors the characteristics of the load, which contains both way communication channel that send the dynamic data. Similarly, it receives the cut-off points as the controlling command from the central controlling unit. The remote controllers contain three supply points as shown in Figure 3, which is designed to adjust or shed the load at those points. The loads are scheduled to be cut when the battery level reaches a certain value. The optimal points for load cut-off are identified using an exhaustive search algorithm. The data of load consumption and cut-off signal are handled though the communication setup in the system. GSM, CDMA, IOT are the common mode of the communication protocol for the smart metering purpose in most of the developing countries [31]. Whereas, reliability and availability of the communication do not fall under the scope of the research. Hence, as shown in Figure 3, the system architecture depends on two-way communication between the smart meter and central controller to implement the proposed DSM strategy. Generally, the signal is sent from the central controller, which define the level of shedding to be implemented, and a signal from the remote controller contains the priority-based load data for monitoring and control application. Each smarter meter receives the shedding signal based on which it shed the load on every household.

This paper presents a generalized day ahead DSM strategy which can be used in MG relying on an ESS for uninterrupted power supply. It uses the load shedding techniques and controls the appliances through the smart meter on the household level. The proposed optimization algorithm aims to identify the optimum point of SoC level to shed the load on the required level such that the higher priority load gets an uninterrupted supply. With the classified load, the shedding can be performed to increase the availability of the stored energy for the higher-prioritized load. For example, considering three different loads like lights, TV and fan as observed from the customer point of view, where light has high priority followed by TV and then fan. For the case when available energy is lower such that serving all three loads, the system only lasts for a few hours. To maximize the hours of load served to light, the proposed algorithm identifies the optimum battery level for TV and Fan. Hence, the proposed algorithm will maximize the hours of energy-served for the load with the highest priority (i.e. ratio of demand energy to served energy will be increased). According to the proposed architecture, the DSM receives the SoC level of ESS, predicates generation and demand,

and identifies an optimum SoC through the search algorithm in order to fulfil the required load consumption of highest priority load. The DSM is performed at the beginning of a defined control period for the assigned period. In this study, the control action is simulated for a dynamic model based solar MGs in the MATLAB environment. When SoC goes below a defined limit set by the search algorithm, the central controller sent a cut of signal to the smart meter to shed the loads. As discussed previously, the shedding is performed in three levels, where the last level (i.e. highest priority) is triggered when the battery reached a minimum SoC. Similarly, the connection is again restored when the battery capacity restored the SoC of 5%; the defined threshold point to avoid flickering. Figure 4 presents the flowchart of how the shedding of the load is achieved in the proposed architecture. Here in the Figure, the F_Set1 defines the level 1 shedding point, F_Set2 defines the level 2 shedding point, and finally, DoD is the Depth of discharge of the battery.

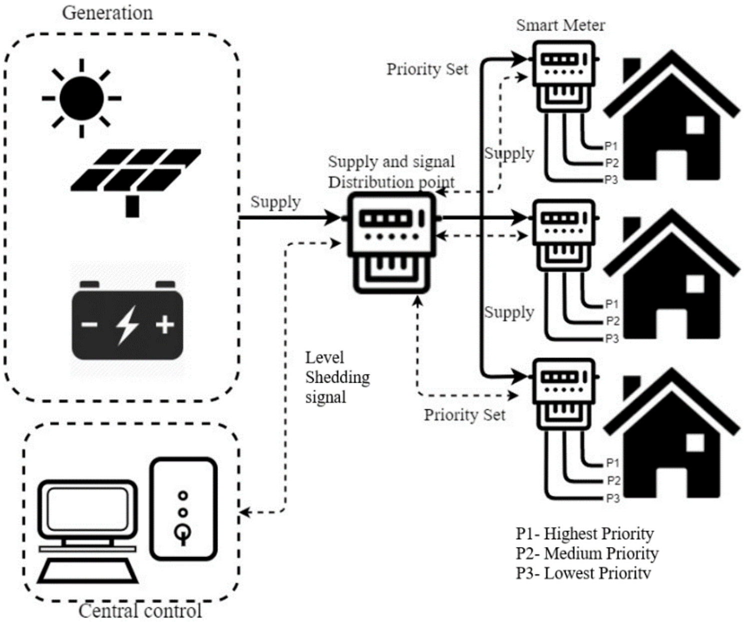

Figure 3: System architecture of proposed model

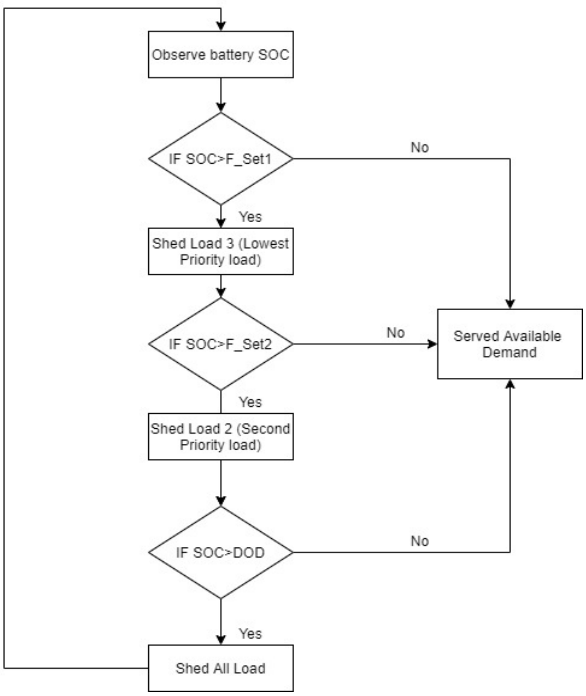

Figure 4:Demand side load shedding Flowchart

The proposed DSM strategy requires the search of an optimum point though model-based technique. Furthermore, the search evaluates the point to find the best user satisfaction which depends upon the available energy in the storage. As the SoC of ESS is dynamic though out the day, a model-based analysis must be performed. A various heuristic-based searching algorithm like genetic, particle swarm, whale optimization is commonly used to find the best point or position in this research sectors. However, these algorithms are suitable to be used for a large search area, whereas in the proposed case area, an exhaustive search shows a higher advantage over other algorithms as our search portion is smaller. The area of search has been reduced by categorizing the load in three levels. Exhaustive search becomes the best option to fine two optimum points on a SoC of ESS, which ranges from SOC_{min} to SOC_{max}. The proposed algorithm checks every possible solution that should be considered as an effective and efficient output. Also, the algorithm calculates the user satisfaction as the objective function for a day ahead for each possible case.

To improve user satisfaction, the proposed algorithm adjusted the shedding point of the specified load that maximize the hours of the energy-served. The algorithm uses 48 hours (considering 2 days autonomy), and predicates the data to identify the maximum hours that load with the highest priority can be served with available energy by changing the threshold point from SOC_{min} to SOC_{max}. Once the hours of energy served for load with the highest priority is maximized, the same approach will be followed for others. The Set1 and Set2 are the thresholds for battery level that is defined for level 1 and level 2 shedding. With the optimal threshold points for potential load shedding, the system then observes the battery level on a real-time basis to shed the respective priorities. F_H1, F_H2, F_H3 indicates the final hours of energy served for load with priority 1, 2 and 3. Similarly, the T_H1, T_H2, T_H3 suggest the measurement of test value obtained for hours of energy-served. The steps of the proposed algorithm are given below. The significant advantage of the purposed algorithm is it improves the availability of energy from the load with the highest priority. The flexibility of the algorithm allows the system to maintain the load profile pattern by adjusting the hourly use lower priority load.

Algorithm

```
Define Threshold values for each priority
FF_H1 = 0, FF_H2 =0, FF_H3 = 0
For Threshold1 < 100 %Priority 3 threshold value
For Threshold2 < 100 %Priority 2 threshold value
if Threshold1 > Threshold2 %priority 3 should be shedded before Priority 2 cannot be shedded together
Model Simulation
Obtain the value T_H1, T_H2, T_H3 though modeling simulation
if T_H1 > FF_H1              %maximize Hours for Load 1
FF_H1 = Present_H1;
Update F_Set1, F_Set2 with Threshold1, Threshold2
elseif T_H1 == FF_H1
if T_H2>FF_H2              %maximize Hours for Load
FF_H2=Present_H2;
Update F_Set1, F_Set2 with Threshold1, Threshold2
elseif Present_H2==DayFinal_H2
if Present_H3>DayFinal_H3   %maximize Hours for Load 3
FF_H3=Present_H3;
elseif Present_H3==DayFinal_H3
maximize energy saving  %Select higher value of threshold
Calculate User Satisfaction;
Update F_Set1, F_Set2 with Threshold1, Threshold2
End all
```

For the identification of optimized thresholds, an exhaustive search optimization has been performed based on user satisfaction indices as shown in Table 2. The threshold point is checked for every 10% change in SoC with limitation on energy allocated for higher priority must be greater than the lower. The user satisfaction ranges from 0 to 1, as zero defines for 0% of user satisfaction and 1 for 100% of user satisfaction. The algorithm searches the presented threshold point for each Setpointas given in Table 2. For Set 1, 0.7 show the lowest priority that will be shedded when SoC reaches 70%, and for Set 2 level at 0.6 shows that the medium priority load will be shedded when SoC reaches or goes below 60%. Among multiple choices, the algorithm compares the threshold point with the highest energy-saved for the system. For the example, the choice for Set 1 ranges from 40% to 70%, and that for Set 2 ranges from 30% to 60%, since the maximum energy is stored when the load is shed at 70% and 60% SoC. The Set points are selected as shown in Table 2 so that the maximum energy is allocated to the load with the highest priority.

Table 1:Optimum threshold identification using exhaustive search

Set 2 \ Set1 (User Satisfaction)	Threshold	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
Set 2	0.1	-	-	-	-	-	-	-	-	-
	0.2	-	-	-	-	-		-	-	-
	0.3	-	-	-	1	1	1	1	0.97	0.86
	0.4	-	-	-	-	1	1	1	0.97	0.86
	0.5	-	-	-	-	-	1	1	0.97	0.86
	0.6	-	-	-	-	-	-	1	0.97	0.86
	0.7	-	-	-	-	-	-	-	0.96	0.86
	0.8	-	-	-	-	-	-	-	-	0.85
	0.9	-	-	-	-	-	-	-	-	-

2.4. Modelling of PV generation and Storage

For the system to be stable, the demand must match the supply. Condition for an independent energy system, the relation can be expressed as in Equations 5 and 6. The technique of load management is tested for an independent PV system with the ESS. So, the major components of this system are the PV solar cell, batteries, user loads, power electronic converters (PEC), battery charging unit etc., whereas the detailed description on the modelling of the generation and storage are presented here in this study.

$$P_{load} + P_{Loss} = P_{supply} \quad (5)$$

$$P_{supply} = P_{Generation} + P_{storage} \quad (6)$$

2.4.1. Generation Modeling

This study is focused on the PV based MGs, so the PV generation has been modelled based on linear relation to ambient temperature and the radiation of the PV panel. The parameters are obtained from information available from the manufactures datasheet for Nominal Operating Cell Temperature (NOCT) and Standard Test Condition (STC). To introduce more realistic output power from the system, a proper cell temperature at the defined radiation has been considered, for which the NOCT has been used to obtained generation of the PV panel. As panel operates at a higher temperature than that of air temperature (T_{air}), so, Equations 7 is used to calculate the operating temperature of the panel. Similarly, Equations 8 and 9 are used to determine the output power and energy from the PV panel. Here, P_{pvmax} is the maximum output power at the presented temperature (T_{air}) and radiation (GT). The $P_{pvmax,STC}$ is the maximum power of the PV cell at standard test condition, and γ is the temperature coefficient of power usually given by the manufacturer. The T_{panel}, STC indicates panel temperature at the standard test condition in the equation.

$$T_{panel} = T_{air} + \frac{G_T}{800} * (T_{NOCT} - 20) \quad (7)$$

$$P_{pvmax} = [P_{PVmax,STC} * G_T * [1 - \gamma * (T_{panel} - T_{panel,STC})]] \quad (8)$$

$$E_{pv} = P_{pvmax}.\Delta t \quad (9)$$

2.4.2. Storage Modeling

The actual demand of the system depends upon the efficiency of PECs and the operated appliances. However, considering the efficiency of each appliance is not possible as we have classified the appliance into three priority levels. Therefore, the proposed method only considers the efficiency of PECs. The power demand of various appliance considered at the instant time "t" is given by Equation 10. Here $E_{demand}(t)$ is the energy demand, $Power_{Appliance}(i)$ is the power demand of i^{th} appliance and n define the number of appliances. The limit to the energy demand is defined by the capacity of the inverter. Consider $P_{inverter}^{max}$ to be the maximum power following through the inverter, the energy flow through at a time interval of Δt can be calculated by using Equation 11, which is the maximum energy that can flow through the inverter for Δt. For this system, the demand and generation define the charging and discharging of the system, where $E_{PV}(t)$ is the energy generated by PV in the time range (*t*), $\eta_{Inverter}$ defined the energy efficiency of the inverter.

$$E_{demand}(t) = \sum_{i=1}^{n} Power_{Appliance}(i).\Delta t \quad (10)$$

$$E_{inveter}^{max} = P_{inverter}^{max}.\Delta t \quad (11)$$

$$E_{demand}(t) = \begin{Bmatrix} \Delta E_{inveter}^{max} \; if E_{demand}(t) \geq E_{inveter}^{max} \\ E_{demand} \; if \;\; E_{demand} < E_{inveter}^{max} \end{Bmatrix} \quad (12)$$

$$\Delta E(t) = E_{PV}(t) - \frac{E_{demand}(t)}{\eta_{Inverter}} \quad (13)$$

Similarly, Equation 13 identifies the limitation of the battery system, where the SoC indicates the State of charge of the battery, as the modelled battery's SoC should be in between the maximum and minimum limits. However, the battery's SoC can reach its limit within a few hours or even in minutes if the discharging or charging energy is very high. The system considers the inverter capacity and charger controller capacity that limit the energy flowing thought the battery. To limit the energy flow within the battery, two conditions are considered for the charging and discharging modes. The rate of charging and discharging energy can be calculated by using Equations 15, 16, 17 and 18. Where $\Delta E_{Charging}^{max}(t)$ is the maximum charging energy, $BESS_{size}$ is the battery size and *THS* is the total sun hours depending upon the location of the system. Similarly, Δt is the defined period for which the energy calculation is conducted. In some case, the *TSH* can be an unidentified term where charger capacity is defined. Similarly, for the discharging mode of the battery, the maximum power flow of the battery is defined by the inverter capacity. During the night time and/ or cloudy day, the whole system depends upon the battery, which is considered by Equation 12 through limiting the demand.

$$SoC^{min} \leq SoC(t) \leq SoC^{max} \quad (14)$$

$$\Delta E_{Charging}^{max}(t) = \frac{BESS_{size}}{TSH}.\Delta t \quad (15)$$

$$TSH = \frac{BESS_{size}}{\Delta E_{Charging}^{max}(t)}.\Delta t \quad (16)$$

Where,

$$E_{Charging}^{max}(t) = P_{charger}^{max}.\Delta t \quad (17)$$

$$\Delta E = \begin{Bmatrix} \Delta E_{Charging}^{max}(t) \; if \; \Delta E \geq E_{Charging}^{max}(t) \\ \Delta E(t) \qquad if \; \Delta E \leq \Delta E_{Charging}^{max}(t) \end{Bmatrix} \quad (18)$$

Further, the energy flow takes place through the charge controller, depending upon the available storage in the battery. Hence, the protection of the battery system from overcharging and over-discharging is necessary. As the battery SoC cannot be reached to 200% of its limit, the stored-energy within a defined period is limited by the charge controller. Here $\Delta AE(t)$ is the available energy stored in the battery system. The SoC_{max} may change in longer duration, but the simulation is done for a shorter duration, so the change in SoC_{max} is neglected and $SoC_{max}$is consider as constant.

$$\Delta AE(t) = SoC_{max} - SoC\,(t) \quad (19)$$

$$\Delta E = \begin{Bmatrix} \Delta E(t) \quad if \; \Delta E(t) \leq \Delta AE(t) \\ \Delta AE(t) \quad if \; \Delta E(t) > \Delta AE(t) \end{Bmatrix} \quad (20)$$

Similarly, in the case when the ΔE is negative (i.e. discharging mode), the battery is protected by the conditions as provided by Equations 21 and 22.

$$\Delta AE(t) = SOC(t) - SOC_{min} \quad (21)$$

$$\Delta E(t) = \begin{cases} \Delta AE(t) & if\ \Delta E(t) \geq \Delta AE(t) \\ \Delta E(t) & if\ \Delta E(t) < \Delta AE(t) \end{cases} \quad (22)$$

3. RESULT AND DISCUSSION

To show the effectiveness of the proposed technique, the demand-side strategy is implemented in the three different scenario considering days lower generations as discussed in Section 2.1. For each priority, the maximum hours of demand are 24 hours, 12 hours and 6 hours as observed from Table 2. With the proposed algorithm, the central controller verifies the best point to shed a higher priority level. Figure 5 shows the total battery capacity allocated for each priority selected by the optimization algorithm for 7 days. In day 1, the algorithm allocates 30% of the battery till the least priority load is shedded (i.e. all three-priority load can be served until the battery reaches 70% of its total capacity), so the level one shedding is done when SoC reaches 70% and level 2 when SoC reaches 60%. High variation of energy allocation can be observed at the 5th day of the simulation as with lower generation, the least energy is allocated for load with lower priority as level 1 shedding is triggered at 90% SoC followed by the shedding level 2, when SoC reaches 50%. Comparision of the day to day user satisfaction obtained for each day is shown in Figure 6. As when DSM has not implemented the system retains 55% user satisfaction. However, the satisfaction comes at a price as the system face blackout up to 11hours on the sixth day. Along with 21 hours of total interruption period in supply comprising of 4 hours in day 5 and 6 hours in day 7. With the implementation of the proposed DSM strategy, the user satisfaction of energy supply is maintained at 70% even during the day with solar generation going below 35%. As the load with higher priority served throughout 7 days, with proper prioritization of load, the approach can maintain continuity in the supply while maintaining user satisfaction of the customer.

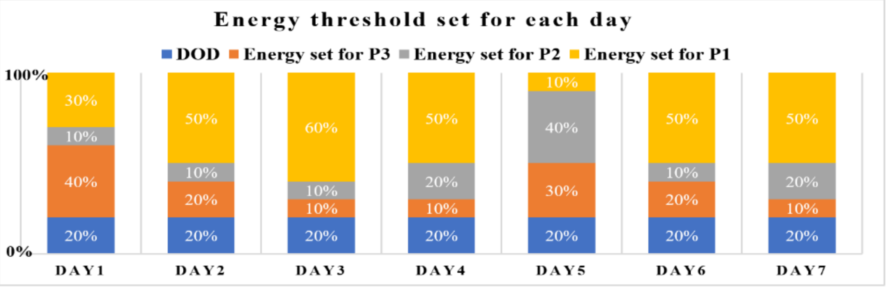

Figure 5:Energy Allocation for each priority level on each day

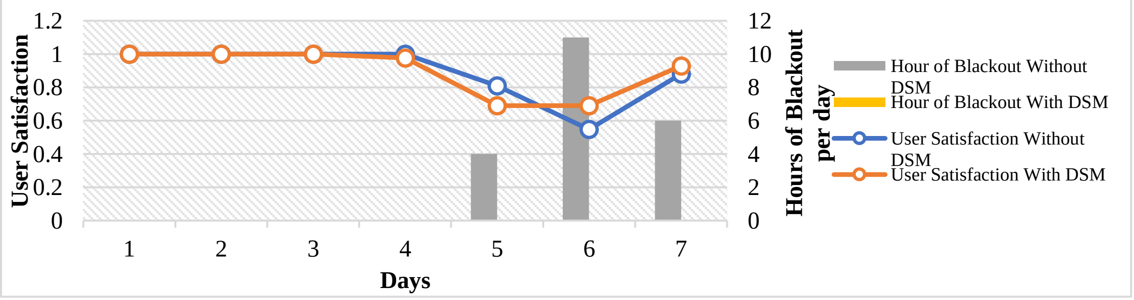

Figure 6:User satisfaction and Blackout hours (With and Without DSM)

Comparing the system load-served by the system with and without DSM strategy, the system faces multiple backouts as seen in Figure 7(a). Likewise, Figure 7(b) shows the load served for each priority by the system with the implementation of the proposed strategy. From the Figures, it is seen that the load with the highest priority is being served though out the service period without any interruption. With the daily load demand, the SoC can be observed to be decreased with time. Even during the day with less generation, to avoid the system blackout the algorithm priorities the load with the highest priority, and sheds the loads with lower priority. Whereas Figure 7 (c) shows the system demand when the proposed algorithm is not implemented. In this case, the system faces blackout up to 21 hours in the total period. Table 3 shows that the total backout of the system can be reduced especially in the residential area of rural communities as well as of urban area. The implementation of the proposed strategy can benefit both the consumers and the utility companies by avoiding the blackouts. In addition, the system can maintain user satisfaction and improve battery life. Comparing the system with and without the DSM strategy, it observes the random distribution of energy for the system without the DSM. When load management is not used, a rule of first come first serve is implied where the priority or appliances turned on first consumes the available energy until it is turned back off or battery reaches its limit. In term of user satisfaction, the system obtained 89% total user satisfaction as shown in Table 3. The algorithm reduces the total backout time of the system from 21 hours to zero hours for the highest priority load, and increase user satisfaction by 5%.

Table 2:Data analysis for overall user satisfaction

Priority level	Total demand hours	Priority percentage based on hours (x, y, z)	Energy demand kWh	Energy served with DSM	User satisfaction per priority (with DSM)	Energy served without DSM	User satisfaction per priority (without DSM)
1	168	0.57	12.13	12.13	0.57	9.82	0.46
2	84	0.29	13.11	10.87	0.24	11.98	0.26
3	42	0.14	12.05	7.14	0.08	9.84	0.12
User Satisfaction		**100%**			**89%**		**84%**

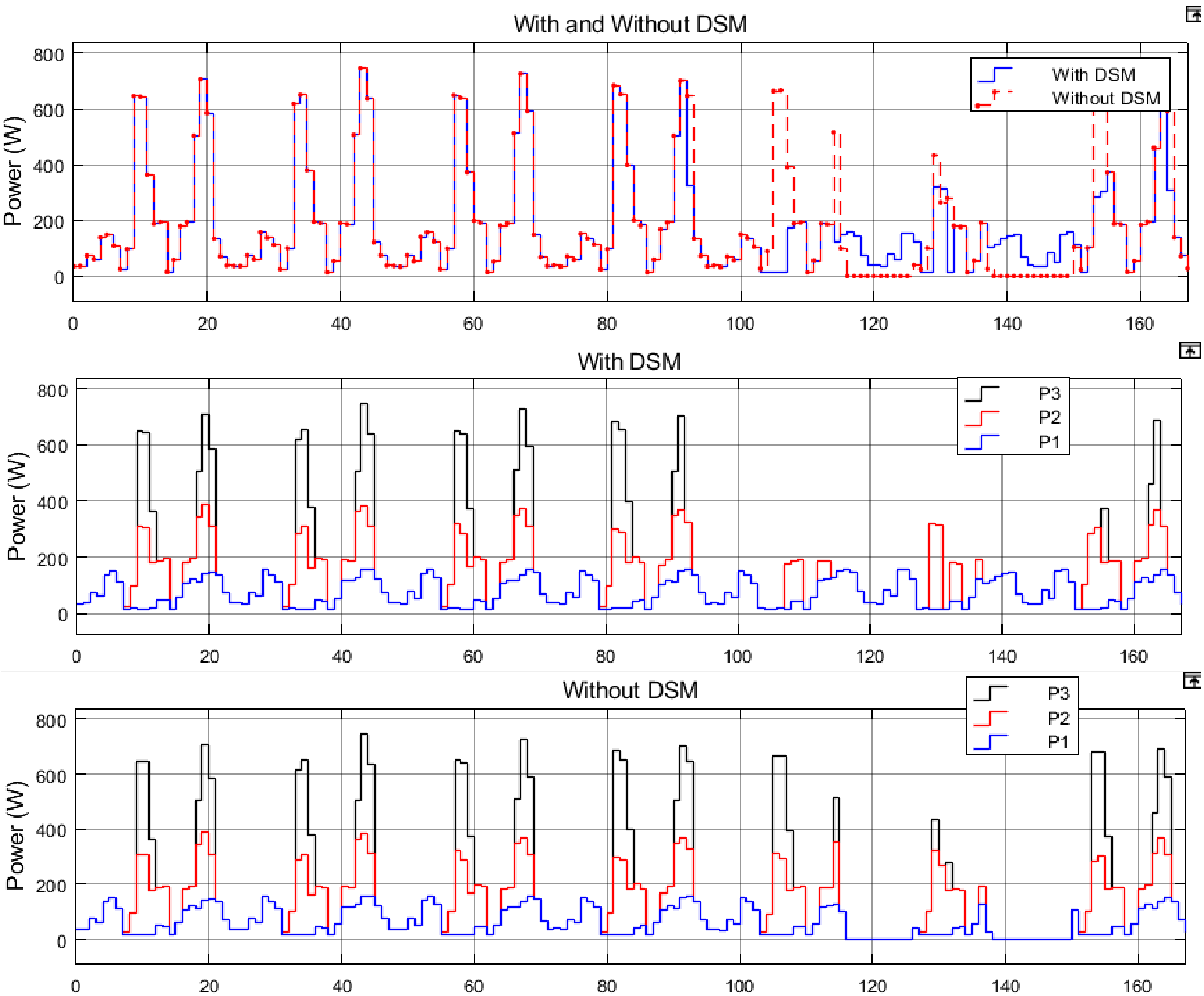

Figure 7:(a) With and Without DSM, (b) Priority based curve with DSM strategy implemented, (c) Demand curve without DSM strategy implemented

As we can see in Figure 8, with the implementation of the load management technique, the system allocates enough energy to serve load with higher priority as demand on load 1 is completely served followed by load 2 and finally load 3 with least priority. Comparison between these two cases shows that the load with lower priority has higher energy served when DSM is not implemented as a higher priority load is not served which affect user satisfaction.

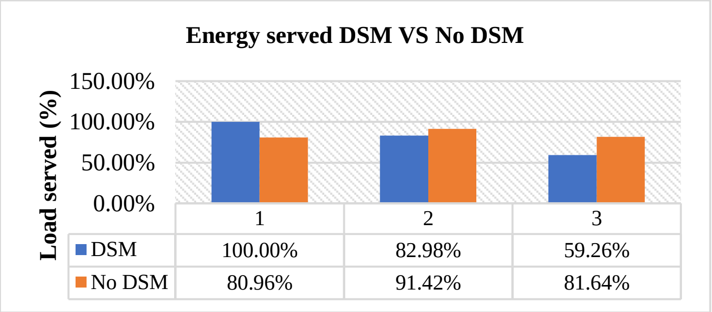

Figure 8: Energy served on each priority

4. CONCLUSION

DSM technique can improve the quality of energy supply in many ways, especially in residential locations. This paper presents a priority-based technique that can be employed in the MGs, where energy crisis occurs due to high demand. Considering the rural MGs with the limited generation, this strategy presents the priorities of the appliance for curtailment to minimize the hours of blackout. Further, to improve the system limits and the calculation complexity, the appliance on residential household are classified into a three-priority level. The simulation shows the effectiveness of the proposed prioritization technique, and the total hours of load serve have been improved for the appliance with higher priority. The system has improved the user satisfaction by 5% via reducing the blackout of essential load, faced by the system from 21 hours to zero hours in the taken period of seven days.

The implementation of the priority concept on load can be further furnished with proper study of load profile on appliances level. With accurate and detailed data on appliance-based load consumption, the priority implementation can be done with more accuracy. In addition, in the presented method, priority classifications is done on three levels to avoid complexity. Therefore, considering the system capability, the priority level can be increased or decrease or even modified to the shifting of loads to better service hours based on capacity of the energy system to remove the blackout or/ and interruption to/ from the system.

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