

Prosumer of Power Grid as Convex Optimization Problem

Aggelos Aggelidakis
Charilaos Akasiadis

1. INTRODUCTION

The depletion of fossil fuel resources, as well as the global warming due to the “greenhouse effect”, have led electricity suppliers to turn to renewable energy generation. The use of such sources can help reduce harmful gases emissions and promise a more sustainable generation process [7]. Due to the fact that renewable energy plants are not centralized and may cover broad geographical regions, researchers focus has turned to distributed generation, where electric power generation sources are connected directly to the distribution network or on the customer side of the meter [14]. The installation of additional sensors and controllers, on every level of the power network, that are able to communicate with each other, makes it possible to effectively manage the flow of energy, as if we had a large and centralized power plant, that sells energy to neighboring plants when it generates surplus, or buys from them when there is shortage. Such infrastructures are known as Virtual Power Plants or Microgrids [1].

The aggregation of the aforementioned energy systems is called the *Smart Grid*. More specifically, the Smart Grid aspires to use secure, two-way communication technologies to exploit valuable information across every stage, from generators and distribution networks to final consumption. This way we can achieve clean production, safety, security, reliability, resilience, efficiency and sustainability in the electricity grid [4]. In fact, the Smart Grid operation requires complex control and decision making, so it is common to represent each entity by *autonomous agents*. As Smart Grid entities we consider consumers, producers and *prosumers* of electricity.

A prosumer is an entity that both consumes and produces energy [11]. Residential prosumers in particular, are expected to be of much importance for the smartgrid era, as they possess increased awareness and the ability to adjust their behavior according to dynamic indicators [8]. There also exist commercial prosumers, industrial wind farms or electric car fleets for example, whose contribution in the electricity market also has a large impact. Due to the economic nature of a prosumer, it is desired to implement selfish and economically thinking autonomous agents that make effective decisions, regarding when to use which source, so that to meet demand at a minimum cost, or even make a profit by selling the electricity surplus to other entities that need it [15]. In this work, we present an algorithm that could be used by prosumers to efficiently manage the operation of their generating and consuming devices, so as to maximize their own profits while contributing to the stability of the overall electricity grid.

The proposed algorithm takes into account the cost of electricity generation by each available device or the charging of a utility company, as well as the predicted consumption and production levels of the facility, and generates an operation plan that results to the maximum profitability of the prosumer. As we will be explaining, the electricity generation cost function can be considered to be convex and so are the constraints that need to be met. Based on this feature, we develop a Primal-Dual interior point algorithm and find the optimal point, that is the *exact amount of energy that each controllable generating device must produce* in order to achieve the largest monetary profit.

The proposed algorithm is evaluated on three prosumer profiles that have different generation and storage characteristics. In our simulations we have also considered a variable pricing scheme for the electricity sold or bought by the utility company, as well as a scenario where the prosumers are not allowed to sell to the utility company. Results show that our method can be used by Smart Grid prosumers to manage their generation process effectively.

This report is structured as follows. Section 2 discusses some of the recent related work. In Section 3 a description of the problem is presented. In Section 4 we provide the prosumer profiles which we chose to evaluate. Section 5 presents the results from the experimental evaluation. Finally, in Section 6 we conclude.

2. RELATED WORK

The implementation and operation of the Smart Grid is a large and new venture. From the modernization of the infrastructures, to the redesign of operational strategies, there are a lot to be done in order to end up with stable, secure and sustainable systems. The work of [13] discusses the need of intelligent online algorithms for the management of grids operation. They propose two methods for tackling uncertainties regarding the future, followed by an analysis based on regret. In our approach, we assume that there exist mechanisms that are able to precisely predict unknown or uncertain indicators e.g., the weather, the output of wind turbines and PV panels, as well as the price rates that a utility company charges electricity during a day.

As argued by authors of [8], “key benefit for the residential prosumer would be the

possibility of optimizing his energy usage by adjusting his behaviour to dynamic market prices as well as have some financial benefits for the electricity he produces (and does not use)". Our method takes into consideration the aforementioned indicators and generates hourly prosumer operation plans that are cost-efficient.

In [12] authors aim to minimize grid's operational and generation costs, by solving a mixed-integer programming problem. We are also concerned about the same costs, but use convex optimization to obtain solutions.

The use of convex optimization methods in Smart Grid applications is not new. In [10], authors use convex optimization to select the best among 64 electricity storage portfolios, by minimizing a combined cost function. The portfolios, i.e. set-ups of different storage devices, are characterized by their capital cost and their average operating cost. Then, Pareto optimal portfolios are found, with respect to these two cost types. The difference with our approach is that we consider a single portfolio that can both store and generate electricity, and we minimize a single cost function, that indicates the capital cost of the system's operation, while meeting electricity demand. Our problem formulation is generic and can be configured to describe any portfolio, though.

The work of [6] describes a distributed convex optimization framework to manage energy flows between islanded microgrids. A scenario of two microgrids that exchange energy is presented. Electricity is transmitted and generated in an optimal manner, by minimizing the respective cost functions. Although authors include a distributed approach, we choose to use a multi-agent formulation, because it is more natural to use when modeling trades and markets. In our work we are concerned about the electricity generation costs of an economically minded prosumer only and seek a method that can be adopted by every prosumer in the Grid.

The most related work to ours is that of [16], where authors present VERA, that is a software system for energy management of contemporary microgrids. They present a cost and constraint formulation that is very close to that adopted in our work. Then, the minimization problem is solved using sequential quadratic programming. Similarly, we implement a Primal-Dual interior point method to solve the minimization problem which we prove that is convex.

We now proceed to a detailed description of our case.

3. PROBLEM FORMULATION

We begin by defining the operating costs of every type of electricity generating or storage devices that are probable to find in a prosumer facility. The sum of these costs is the objective function which we want to minimize. In our model, we consider plans on a daily basis, that is for 24 time intervals, representing one hour of the day each.

3.1. ELECTRICITY COSTS

A prosumer can employ any type of renewable energy generation system, such as wind turbines and solar panels. These generators generate power at each time interval t , and their output values are noted as P_t^w and $P_t^{pv} \in \mathbf{R}^+$, representing the KWh produced.

Electricity generated by such devices is considered to have zero cost. The amount of energy can be accurately estimated, though it can not be controlled. There are techniques to obtain estimates about renewable production size and we use them in this work. In more detail, the units, for both production and consumption, are modelled based on the work of [9]. Authors provide the methodology for optimal sizing of stand-alone photovoltaic/wind-generator systems using genetic algorithms, and also model the 24-hour consumption curve well enough 3.1. Note that there exist two peaks during the day, the first one at midday and the second one early at night. Furthermore, the work [5] presents a model of photovoltaic arrays based on the circuits that are contained in their cells.

The power output of a PV module depends on the number of cells in the module, the type of cells, and the total surface area of the cells. All modules are rated by manufacturers in terms of their peak power (W_p) under standard test conditions: ie. 1000 W/m^2 of sunlight ('peak sun'), $T = 25 \text{ C}$, and air mass of 1.5. The type of the cell used in this study is Crystalline Silicon. Open circuit voltage is $V_g = 1.12 \text{ eV}$ and short circuit current is $I_{sc} = 5.75 \text{ A}$.

The second renewable energy unit that our system supports, wind-turbine, are based on the model that is created by [3]. The most important feature of the model is the swept area and the rotor diameter that characterize each wind-turbine.

The prosumer unit combines the above units in order to create a more realistic environment and simulates the behaviour of different units of the prosumer during the optimal management of the energy system. The modelling of the consumption based on real facts aims to bring the simulation closer to reality.

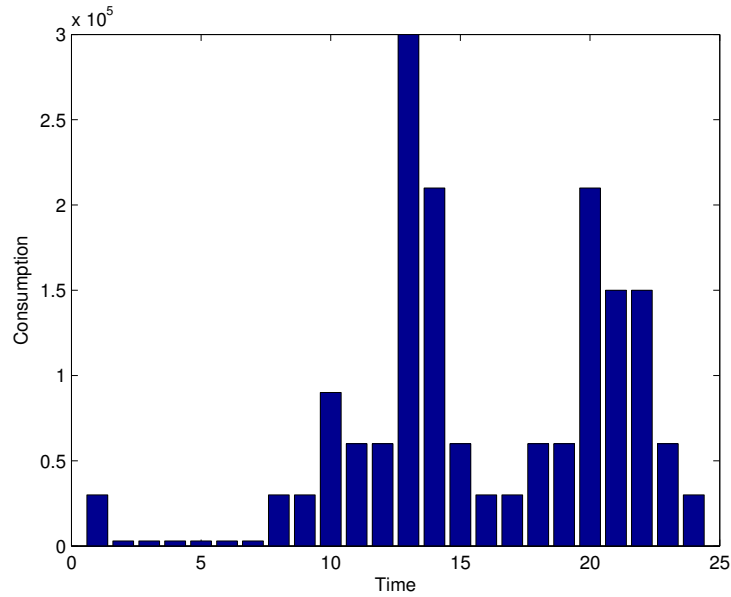


Figure 3.1: 24-hour Consumption Curve

The energy that the prosumer requests to consume at each t is noted as P_t^{cons} .

3.1.1. UTILITY COMPANY

Prosumers can buy from or sell to a utility company that may exist externally. The pricing of the utility company might be either *flat* or *variable among time intervals* and is noted as c_t^{uc} . The amount of energy exchanged between the prosumer and the utility company at each time interval t is noted as $P_t^{uc} \in \mathbf{R}$ and it is measured in KWh. The product $c_t^{uc} \cdot P_t^{uc}$ is the total cost of the utility company services at a time interval t , and when its value is negative it denotes the profit of the prosumer for selling excessive energy.

3.1.2. MICRO TURBINES

There may also be micro turbines (MTs) available, that run on a variety of fuels, like biogas, diesel, or natural gas. Each MT has a maximum generation capability noted as P_{\max}^{mt} . For each time interval t , the amount of the energy that we need the microturbine to produce is P_t^{mt} , where $0 \leq P_t^{mt} \leq P_{\max}^{mt}$. The cost for producing energy using the MT is noted as c_t^{mt} , and is given by the sum of the fuel cost of the generator and the start-up cost of the generator.

The fuel cost of a typical diesel generator at the time interval t is usually expressed as [17]:

$$\hat{c}_t^{mt}(P_t^{mt}) = a \cdot (P_t^{mt})^2 + b \cdot P_t^{mt} + c$$

where, P_t^{mt} is the output power of the generator during t ; a , b and c are constants determined by the type of generator, here $a = 0.0547$, $b = 1.7362$, $c = 3.2456$. A binary control signal $X_t \in \{0, 1\}$ is also defined, enabling the prosumer to decide when to turn the MT on or off. We must note that the generator has a startup cost SC_t^{mt} that is added to the overall generation cost c_t^{mt} at times when the generator needs to be turned on. Thus, each time interval t the MT generates P_t^{mt} KWh, the associated cost is $c_t^{mt}(P_t^{mt}) = X_t \cdot (\hat{c}_t^{mt}(P_t^{MT}) + SC_t^{mt})$.

3.1.3. ELECTRICITY STORAGE

Apart from generators, a prosumer can possess electricity storage systems, such as batteries. In the stationary battery operation, the amount of charged power corresponds to the surplus renewable energy sources output over the threshold for charge. The amount of discharged power corresponds to the surplus load power over the threshold for discharge. Hence, the amount of discharged power is restricted so that no reverse power flow is generated by the stationary batteries.

Each battery has a certain state of charge at every t that must be within certain levels:

$$SOC_t^{min} \leq SOC_t \leq SOC_t^{max}$$

The output power of the battery at t , P_t^b , is also subject to constraints:

$$P_{ch_max}^b \leq P_t^b \leq P_{disch_max}^b$$

where $P_{ch_max}^b$ is the maximum charging capability of the battery and $P_{disch_max}^b$ its maximum discharge capability. Cap_{max} denotes the battery capacity. The state of charge for the next time interval depends on P_t^b :

$$SOC_t = SOC_{t-1} - \frac{P_t^b}{Cap_{max}}$$

3.1.4. COST FUNCTION

The operating cost at a time interval for producing $\mathbf{x} = [P^{uc}, P^{mt}, P^b]$ amount of energy, is given by:

$$Cost = c^{uc} \cdot P^{uc} + \sum_{j=1}^n c_j^{mt} (P_j^{mt})$$

where n is the number of micro turbines.

3.1.5. CONVEX MINIMIZATION PROBLEM

Our main concern in this work is to keep the generation costs of $\mathbf{x} = [P^{uc}, P^{mt}, P^b]$ at a minimum, while satisfying specific constraints.

First of all, demand must be met at all times, i.e.:

$$P^{uc} + P^{mt} + P^b = P^{cons} - P^w - P^{pv}$$

Next, the battery needs to operate at certain levels. The constraints are the following:

$$\begin{aligned} -P^b - P_{ch_max}^b &\leq 0 \\ -P^b - (1 - SOC_{t-1}) \cdot Cap_{max} &\leq 0 \\ P^b - (SOC_{t-1} - 0.2) \cdot Cap_{max} &\leq 0 \\ P^b - SOC_{t-1} \cdot Cap_{max} &\leq 0 \\ P^b - P_{disch_max}^b &\leq 0 \end{aligned}$$

that is, we can charge no more than the amount of load stored in the battery, charged during a time interval has a maximum value that depends on the battery, state of charge must not drop below 0.2, we can discharge no more than the amount of load stored in the battery, and the electricity discharged during a time interval has a maximum value that depends on the battery. The above inequalities are all affine.

Consider that there are constraints regarding the MT operation too. More specifically,

$$\begin{aligned} P^{mt} &\geq 0 \\ P^{mt} - X \cdot P_{max}^{mt} &\leq 0 \end{aligned}$$

that is, we can only produce energy and not store, and there is a limit in the amount of production governed by the specifications of each MT. Once again the inequalities are affine.

The amount of energy exchanged with the grid is unlimited in the general case, but in the experimental section we test a scenario where P^{uc} is strictly positive, and another one that is strictly negative.

Finally, we examine the convexity of the cost function. To begin, the domain of the cost function is \mathbf{R} , that is convex. The first term of the sum, is an affine combination of P^{uc} and can be considered convex. The second term is quadratic functions of P^{mt} . The second derivative of the MT term is 2α , $\alpha > 0$, that is positive, thus the function is convex. The sum of convex functions is also convex, concluding with the convexity of the cost function.

Summarizing, to find the minimum cost, we must obtain the appropriate values for

$$\mathbf{x} = [P^{uc}, P^{mt}]$$

namely:

$$\begin{aligned} \min_{\mathbf{x}} \quad & f(\mathbf{x}) = c^{uc} \cdot P^{uc} + \sum_{j=1}^n c_j^{mt} (P_j^{mt}) \\ \text{subject to} \quad & P^{uc} + \sum_{j=1}^n (P_j^{mt}) + \sum_{j=1}^m (P_j^b) + P^w + P^{pv} - P^{cons} = 0 \\ & -P_j^b - P_{ch_max,j}^b \leq 0, \forall j \in M \\ & -P_j^b - (1 - SOC_{t-1,j}) \cdot Cap_{max,j} \leq 0, \forall j \in M \\ & P_j^b - (SOC_{t-1,j} - 0.2) \cdot Cap_{max,j} \leq 0, \forall j \in M \\ & P_j^b - SOC_{t-1,j} \cdot Cap_{max,j} \leq 0, \forall j \in M \\ & P_j^b - P_{disch_max,j}^b \leq 0, \forall j \in M \\ & -P_j^{mt} \leq 0, \forall j \in N \\ & P_j^{mt} - X \cdot P_{max,j}^{mt} \leq 0, \forall j \in N \end{aligned} \quad (3.1)$$

For ease of presentation, from now on we assume that there exist one MT and one battery array.

3.2. PRIMAL-DUAL METHOD

For the minimization procedure we use the Primal-Dual interior point method [2]. The linear system that we solve to obtain the search direction Δx_{pd} , in order to minimize the problem (3.3) is given by:

$$\begin{bmatrix} \nabla^2 f_0(x) + \sum \lambda_i \nabla^2 f_i(x) & Df(x)^T & A^T \\ -diag(\lambda) Df(x) & -diag(-f(x)) & 0 \\ A & 0 & 0 \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \\ \Delta v \end{bmatrix} = - \begin{bmatrix} \nabla f_0(x) + Df(x)^T \lambda + A^T v \\ -diag(\lambda) f(x) - \frac{1}{t} \mathbf{1} \\ Ax - b \end{bmatrix} \quad (3.2)$$

PROSUMER IS PERMITTED TO BUY & SELL TO THE UTILITY

We have two different cases, one for MT control signal $X = 0$ and another for $X = 1$. When the MT is turned on ($X = 1$):

$$\begin{aligned} \nabla^2 f_0(x) &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2\alpha & 0 \\ 0 & 0 & 0 \end{bmatrix}, \\ \nabla^2 f_i(x) &= \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \forall i \\ f(x) &= \begin{bmatrix} P^{mt} - P_{\max} \\ -P^{mt} \\ -P^b - P_{ch_max}^b \\ -P^b - (1 - SOC_{t-1}) \cdot Cap_{max} \\ P^b - (SOC_{t-1} - 0.2) \cdot Cap_{max} \\ P^b - SOC_{t-1} \cdot Cap_{max} \\ P^b - P_{disch_max}^b \end{bmatrix} \\ Df(x) &= \begin{bmatrix} 0 & 1 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \\ 0 & 0 & -1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix} \\ \lambda &= -\frac{1}{f(x)} \\ A &= \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \\ b &= P^{cons} - P^w - P^{pv} \end{aligned}$$

In the case it is turned off ($X = 0$):

$$\begin{aligned} \nabla^2 f_0(x) &= \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \\ \nabla^2 f_i(x) &= \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \\ f(x) &= \begin{bmatrix} -P^b - P_{ch_max}^b \\ -P^b - (1 - SOC_{t-1}) \cdot Cap_{max} \\ P^b - (SOC_{t-1} - 0.2) \cdot Cap_{max} \\ P^b - SOC_{t-1} \cdot Cap_{max} \\ P^b - P_{disch_max}^b \end{bmatrix} \end{aligned}$$

$$Df(x) = \begin{bmatrix} 0 & -1 \\ 0 & -1 \\ 0 & 1 \\ 0 & 1 \\ 0 & 1 \end{bmatrix}$$

$$\lambda = -\frac{1}{f(x)}$$

$$A = \begin{bmatrix} 1 & 1 \end{bmatrix}$$

$$b = P^{cons} - P^w - P^{pv}$$

PROSUMER IS NOT PERMITTED TO SELL TO THE UTILITY

This problem is slightly different from the previous one. When the renewable energy sources produce more power than the power that consumers need, surplus is created. However the prosumer is not permitted to sell that amount of power to the utility. Hence, "Dump Load" is created. As a result of that, we have to create a new decision variable maps to the amount of power of the "Dump Load". Consider that amount of power as P^{dump} . Let formulate the problem again:

$$\begin{aligned} \min_{\mathbf{x}} \quad & f(\mathbf{x}) = c^{uc} \cdot P^{uc} + \sum_{j=1}^n c_j^{mt} (P_j^{mt}) \\ \text{subject to} \quad & P^{uc} + \sum_{j=1}^n (P_j^{mt}) + \sum_{j=1}^m (P_j^b) + P^w + P^{pv} - P^{cons} - P^{dump} = 0 \\ & -P_j^b - P_{ch_max,j}^b \leq 0, \forall j \in M \\ & -P_j^b - (1 - SOC_{t-1,j}) \cdot Cap_{max,j} \leq 0, \forall j \in M \\ & P_j^b - (SOC_{t-1,j} - 0.2) \cdot Cap_{max,j} \leq 0, \forall j \in M \\ & P_j^b - SOC_{t-1,j} \cdot Cap_{max,j} \leq 0, \forall j \in M \\ & P_j^b - P_{disch_max,j}^b \leq 0, \forall j \in M \\ & -P_j^{mt} \leq 0, \forall j \in N \\ & P_j^{mt} - X \cdot P_{max,j}^{mt} \leq 0, \forall j \in N \\ & -P_j^{dump} \leq 0, \forall j \in N \end{aligned} \tag{3.3}$$

We have two different cases, one for MT control signal $X = 0$ and another for $X = 1$. When the MT is turned on ($X = 1$):

$$\nabla^2 f_0(x) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 2\alpha & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix},$$

$$\nabla^2 f_i(x) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \forall i$$

$$f(x) = \begin{bmatrix} P^{mt} - P_{\max} \\ -P^{mt} \\ -P^b - P_{ch_max}^b \\ -P^b - (1 - SOC_{t-1}) \cdot Cap_{max} \\ P^b - (SOC_{t-1} - 0.2) \cdot Cap_{max} \\ P^b - SOC_{t-1} \cdot Cap_{max} \\ P^b - P_{disch_max}^b \\ -P^{dump} \end{bmatrix}$$

$$Df(x) = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$$

$$\lambda = -\frac{1}{f(x)}$$

$$A = \begin{bmatrix} 1 & 1 & 1 & -1 \end{bmatrix}$$

$$b = P^{cons} - P^w - P^{pv}$$

In the case it is turned off ($X = 0$):

$$\nabla^2 f_0(x) = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix},$$

$$\nabla^2 f_i(x) = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$f(x) = \begin{bmatrix} -P^b - P_{ch_max}^b \\ -P^b - (1 - SOC_{t-1}) \cdot Cap_{max} \\ P^b - (SOC_{t-1} - 0.2) \cdot Cap_{max} \\ P^b - SOC_{t-1} \cdot Cap_{max} \\ P^b - P_{disch_max}^b \\ -P^{dump} \end{bmatrix}$$

$$Df(x) = \begin{bmatrix} 0 & -1 & 0 \\ 0 & -1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$

$$\lambda = -\frac{1}{f(x)}$$

$$A = \begin{bmatrix} 1 & 1 & -1 \end{bmatrix}$$

$$b = P^{cons} - P^w - P^{pv}$$

The Primal dual algorithm is given as:

Algorithm 1 Primal Dual

Input: \mathbf{x} satisfying $f_1(\mathbf{x}) < 0, \dots, f_n(\mathbf{x}) < 0, \lambda > 0, \mu > 1, \epsilon > 0$
while TRUE **do**
 Determine $t, t := n\mu/\hat{\eta}$
 Compute primal-dual search direction Δy_{pd}
 Line search and update
 Quit if $\|r_{pri}\|_2 \leq \epsilon, \|r_{dual}\|_2 \leq \epsilon, \hat{\eta} \leq \epsilon$
end while

4. PROSUMER PROFILES

The unit of the prosumer includes photovoltaic arrays, wind turbines, storage unit (batteries) and micro-turbines (in our scenario micro-turbines are diesel generators) to produce power. In our setting we consider three different prosumers, each possessing electricity generation units of different capabilities.

More specifically, the capabilities of each prosumer are summarized in Table 4.1.

Prosumer Id	Battery Capacity	MT Capability
a	80000	200000
b	100000	100000
c	250000	200000

Table 4.1: Prosumer equipment.

The price of the utility company for each time interval is a uniformly distributed random variable normalized in the range (0.0785 – 0.09) Euros/KWh, modeling the recent trend on variable pricing.

We now proceed with the experimental evaluation of our setting.

5. EXPERIMENTAL RESULTS

In this section we test the performance of the Primal-Dual algorithm we presented, when adopted by the prosumers we discussed above. Each experiment simulates the energy flow management for 24-hours. We consider two scenarios:

1. Prosumers can both buy and sell electricity to the utility company.
2. Prosumers only buys (and not sells) electricity from the utility company.

5.1. SCENARIO 1

The typical costs of the prosumer profiles for a single day, according to the first scenario where prosumers fully interact with the utility company, are presented in Table 5.1.

Prosumer Id	Daily profit
a	39.20 Euros
b	57.63 Euros
c	43.74 Euros

Table 5.1: Generation costs when the prosumers fully interact with the utility company (buy and sell).

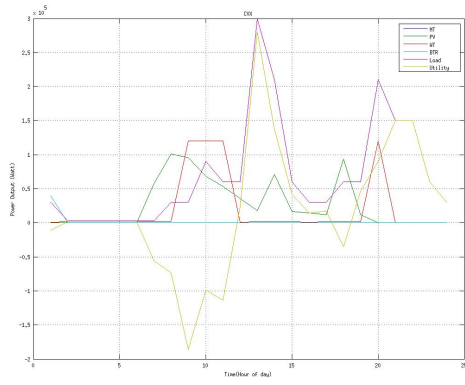
The algorithm "weighs" the generation costs of each available device and chooses a discharging profile for each of them that generates the requested amount of energy at a minimum cost. The minimum cost operation profiles are shown in Figure 5.1.

5.2. SCENARIO 2

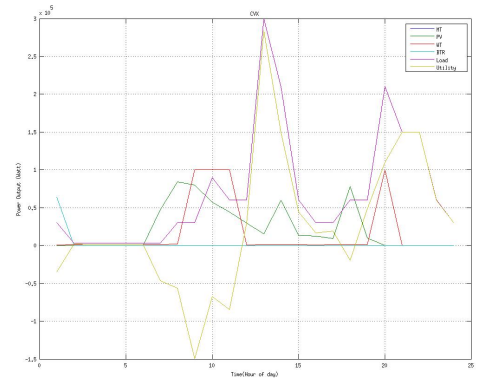
In this case, prosumers are not able to sell their energy surplus to the utility company, and thus any excess is considered dump load. Here, the algorithm actually tries to minimize the amount of the dump load, as the prosumer loses its produced energy. It is natural that the prosumers in this scenario have increased costs, because the utility company does not pay them. The prosumer costs for a typical day are shown in 5.2. The minimum cost operation profiles are shown in Figure 5.2.

Prosumer Id	Daily profit
a	82.96 Euros
b	68.21 Euros
c	86.08 Euros

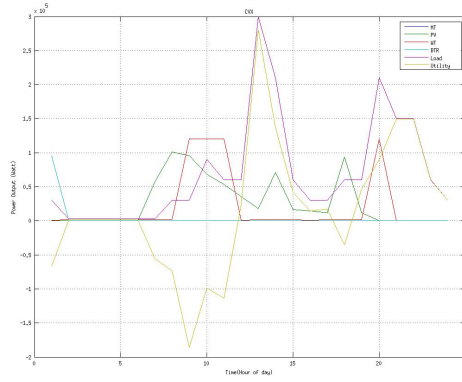
Table 5.2: Generation costs when the prosumers only buy electricity from the utility company.



(a) Prosumer a

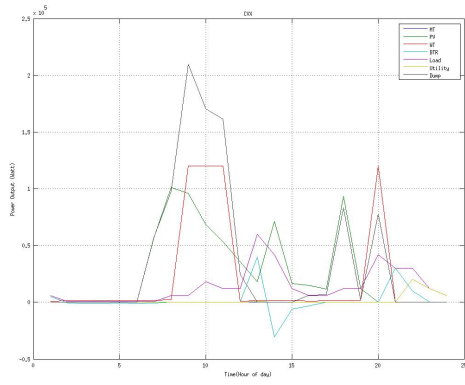


(b) Prosumer b

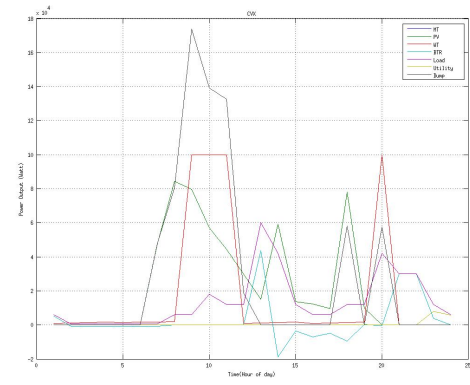


(c) Prosumer c

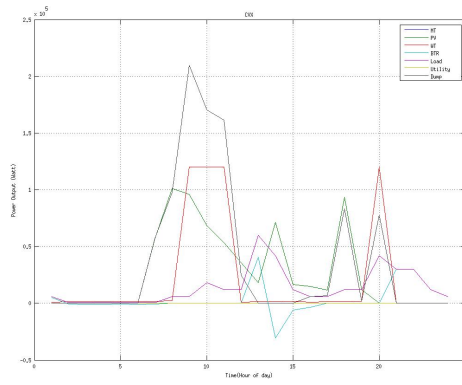
Figure 5.1: Operation levels of each source during the day (scenario 1).



(a) Prosumer a



(b) Prosumer b



(c) Prosumer c

Figure 5.2: Operation levels of each source during the day (scenario 2).

6. CONCLUSIONS AND FUTURE WORK

In this work, we presented a method for minimizing the operating costs of prosumer entities in the Smart Electricity Grid. We studied the application of convex optimization techniques in Smart Grid environments that have been applied in past work. A problem formulation is provided, which we prove that is convex. For the solution of the minimization problem we constructed a Primal-Dual algorithm and effectively applied it in our simulations of realistic prosumer profiles.

There are many additions that can be researched in the near future. Firstly, it will be interesting to explore the behavior of a larger number of prosumers, who are trying to exchange energy with a retailer instead of the utility directly, while the retailer might have contradicting interests. Secondly, the use of different type of micro-turbines, as well as, different type of batteries, will make the simulation even more realistic. Furthermore, the possibilities of non-convex scenarios is quite interesting, specifically the ones where we can convert them to convex ones, and thus efficiently solve them.

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A. MATLAB SCRIPTS

A.1. FIRST PROSUMER'S PROFILE

B. ENERGY MANAGEMENT SYSTEM – CVX – PRIMAL DUAL INTERIOR POINT METHOD

```
1 clear all; close all; clc;
2
3 %% READ CSV FILE
4 row=1;
5
6 numm=[ 120000 -100000 120000 120000 40000
7         80000 -75000 80000 80000 750000
8         50000 -45000 120000 120000 80000 ];
9
10 for row=1:3
11 %% READ DATA
12 fprintf('Simulation Started\n\n') ;
13
14 num = csvread('Data.csv');
15
16 BATTpmax = num(row,1);
17 BATTpmin = num(row,2);
18 PVpmax = num(row,3);
19 WTpmax = num(row,4);
20 MTpmax = num(row,5);
21
22 %% TIME STEPS, AC BUS
23 time_step = 24;
24 totalUnits = 5;
25 p_bus = zeros(totalUnits,time_step);
26 index_GenMT = 1; index_GenPV = 2; index_GenWT = 3; index_GenBTR
    = 4; index_Load = 5; index_Util = 6;
27
28 % MICROURBINE
29 mut = 6000;% minimum up time (sec)
30 mdt = 200;% minimum down time (sec)
31 ton=0;% counting the times turned on
32 toff=0;% counting the times turned off
33 statusMT=ones(1,time_step);
34
35 %% WIND TURBINE
```

```

36 statusWT=ones(1,time_step);
37
38 %% BATTERY
39 Vb=12;% initial voltage
40 pbold = 0;
41 [soc_cvx, qq1, qq2,num_of_batt] = battery( 0, pbold, BATTpmax);
42
43 %% UTILITY
44 vwind = [10.12 11.654 12.43 12.7654 12.123 12.5643 12.626
13.845 15.113 16.322 17.11 10.12 11.654 12.43 12.7654 10.12
11.654 12.43 12.7654 24.6363 25.346 26.7562 27.23 28.88];
45 irrads = [0.0075 0.3948 0.674 0.789 0.519 0.555 0.493 0.935
0.891 0.583 0.514 0.35 0.169 0.6472 0.1508 0.1324 0.10385
0.8206 0.1069 0.3603 0.7578 0.8679 0.0767 0.5666];
46 cutil = 10(-3)*[0.0787 0.0785 0.0785 0.0789 0.088 0.088 0.089
0.09 0.0885 0.0805 0.0815 0.0825 0.079 0.078 0.089 0.0866
0.0799 0.082 0.08 0.085 0.09 0.0788 0.0788 0.085];
47 tempr = [33.57 66.38 54.61 65.75 42.07 70.09 30.81 54.57
57.26 28.02 65.66 72.92 62.93 49.59 54.64 43.63 32.40
39.59 56.84 47.46 68.34 59.68 55.18 44.52];
48 stp=3600;
49 %% EVALUATION
50 for i=1:time_step
51     %% CONSTRAINTS MT
52     if( ton >= mut/stp )
53         statusMT(i) = 0;
54     end
55
56     if( toff >= mdt/stp )
57         statusMT(i) = 1;
58     end
59
60     if( statusMT(i) == 0 )
61         ton=0;
62         toff=toff+1;
63         if(i<length(statusMT))
64             statusMT(i+1)=0;
65         end
66     else
67         toff = 0;
68         ton = ton+1;
69         if(i<length(statusMT))
70             statusMT(i+1)=1;
71         end

```

```

72     end
73
74     %% AC BUS
75     p_bus(index_GenWT,i) = windturbine(vwind(i),statusWT(i),
        WTpmax);
76     p_bus(index_Load,i) = power_bus_per_dayhour(time_step,i);
77     if(mod(i,time_step)<=19 && mod(i,time_step)>=7)
78         p_bus(index_GenPV,i) = pv_array(irrad(i),tempr(i),
        PVpmax);
79     end
80
81     if soc_cvx<=0.201
82         soc_cvx=0.201;
83     end
84     %% CVX
85     %         n=3;
86     %         cvx_begin
87     %             variable x(n,1)
88     %             minimize objFun(x(1), x(2), statusMT, toff, cutil
        (i), i);
89     %             subject to
90     %                 x(1)+x(2)+x(3)*Vb==p_bus(index_Load,i)-(p_bus(
        index_GenWT,i)+p_bus(index_GenPV,i));
91     %                 %-x(3)<=0;
92     %                 %#####
93     %                 -x(3)*Vb-abs(BATTpmin)<=0;
94     %                 -x(3)*Vb-(1-soc_cvx)*BATTpmax<=0;
95     %                 %#####
96     %                 x(3)*Vb-(soc_cvx-0.2)*BATTpmax<=0;
97     %                 x(3)*Vb-soc_cvx*BATTpmax<=0;
98     %                 x(3)*Vb-abs(BATTpmin)<=0;
99     %                 x(2)-MTpmax*statusMT(i)<=0;
100    %                 -x(2)<=0;
101    %         cvx_end
102    %         x_cvx_x = cvx_optpnt.x;
103    %         fprintf('\nCVX finished. ');
104
105    %% PRIMAL DUAL INTERIOR POINT METHOD
106    clear r_t_d r_t_c r_t_p r_t
107    clear array_mtr
108    clear Dy Dx_pd Dlamda_pd Dv_pd
109    clear x lamda v
110    clear eta tau f Df
111

```

```

112     p=1;
113     mu=10;
114     alpha = 1e-2;
115     beta = 5e-1;
116     epsilon_feas = 1e-6;
117     epsilon = 1e-6;
118
119     b = p_bus(index_Load , i)-p_bus(index_GenWT , i)-p_bus(
        index_GenPV , i );
120
121     v = rand(p,1) ;
122
123     if statusMT(i)==1
124         m=7;
125         n=3;
126         A = [1 1 Vb];
127         x = rand(n,1) ;
128
129         f = [ x(2)-MTpmax;
130             -x(2) ;
131             -x(3)*Vb-abs(BATTpmin) ;
132             -x(3)*Vb-(1-soc_cvx)*BATTpmax;
133             x(3)*Vb-(soc_cvx-0.2)*BATTpmax;
134             x(3)*Vb-soc_cvx*BATTpmax;
135             x(3)*Vb-abs(BATTpmin) ] ;
136
137         Df = [0 1 0;
138             0 -1 0;
139             0 0 -Vb;
140             0 0 -Vb;
141             0 0 Vb;
142             0 0 Vb;
143             0 0 Vb];
144     else
145         m=5;
146         n=2;
147         A = [1 Vb];
148         x = rand(n,1) ;
149
150         f = [ -x(2)*Vb-abs(BATTpmin) ;
151             -x(2)*Vb-(1-soc_cvx)*BATTpmax;
152             x(2)*Vb-(soc_cvx-0.2)*BATTpmax;
153             x(2)*Vb-soc_cvx*BATTpmax;
154             x(2)*Vb-abs(BATTpmin) ] ;

```

```

155
156         Df = [0 -Vb;
157               0 -Vb;
158               0 Vb;
159               0 Vb;
160               0 Vb];
161     end
162
163     Hessian = zeros(m,n);
164     Hessian_f0 = D2_f_obj(statusMT(i));
165
166     lamda = -1./f;
167
168     k = 1;
169
170     %% Repeat
171     while (1)
172
173         %% DETERMINE tau
174         if statusMT(i)==1
175             f = [ x(2,k)-MTpmax;
176                 -x(2,k);
177                 -x(3,k)*Vb-abs(BATTpmin);
178                 -x(3,k)*Vb-(1-soc_cvx)*BATTpmax;
179                 x(3,k)*Vb-(soc_cvx-0.2)*BATTpmax;
180                 x(3,k)*Vb-soc_cvx*BATTpmax;
181                 x(3,k)*Vb-abs(BATTpmin) ];
182
183         else
184             f = [ -x(2,k)*Vb-abs(BATTpmin);
185                 -x(2,k)*Vb-(1-soc_cvx)*BATTpmax;
186                 x(2,k)*Vb-(soc_cvx-0.2)*BATTpmax;
187                 x(2,k)*Vb-soc_cvx*BATTpmax;
188                 x(2,k)*Vb-abs(BATTpmin) ];
189         end
190
191         eta(k) = -(f)' * lamda(:,k);
192         tau = (mu * n)/eta(k);
193         t_max = 1;
194         t = 0.99*t_max;
195         %% UPDATE residuals
196         if statusMT(i)
197             r_t_d(:,k) = D_f_obj(x(2,k), statusMT(i), cutil(i))
198                             + Df'*lamda(:,k) + A'*v(:,k);

```

```

198         r_t_c(:,k) = -diag(lamda(:,k))*f - (1/tau)*ones(m
199             ,1);
200         r_t_p(:,k) = A*x(:,k)-b;
201         r_t(:,k) = [r_t_d(:,k); r_t_c(:,k); r_t_p(:,k)];
202     else
203         r_t_d(:,k) = D_f_obj(0, statusMT(i), cutil(i)) + Df
204             '*lamda(:,k) + A'*v(:,k);
205         r_t_c(:,k) = -diag(lamda(:,k))*f - (1/tau)*ones(m
206             ,1);
207         r_t_p(:,k) = A*x(:,k)-b;
208         r_t(:,k) = [r_t_d(:,k); r_t_c(:,k); r_t_p(:,k)];
209     end
210
211 %% UPDATE the invertible matrix for  $\hat{\mathbf{T}}\mathbf{y}$ 
212 array_mtr = [Hessian_f0 Df' A';
213     -diag(lamda(:,k))*Df -diag(f) zeros(m,p);
214     A zeros(p,m) zeros(p,p)];
215
216 %% UPDATE  $\hat{\mathbf{T}}\mathbf{y}$ 
217 Dy = -inv(array_mtr)*r_t(:,k);
218 Dx_pd = Dy(1:n);
219 Dlamda_pd = Dy(n+1:n+m);
220 Dv_pd = Dy(n+m+1:size(Dy,1));
221
222 %%  $\hat{\mathbf{I}}\dot{\mathbf{z}} + \mathbf{D}\hat{\mathbf{I}}\dot{\mathbf{z}} > 0$ 
223 while ~(sum((lamda(:,k)+t*Dlamda_pd) > 0) == m)
224     t=beta*t;
225 end
226
227 %%  $f(\mathbf{x}+) < 0$ 
228 if statusMT(i)==1
229     f = [ (x(2,k)+t*Dx_pd(2))-MTpmax;
230         -(x(2,k)+t*Dx_pd(2));
231         -(x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin);
232         -(x(3,k)+t*Dx_pd(3))*Vb-(1-soc_cvx)*BATTpmax;
233         (x(3,k)+t*Dx_pd(3))*Vb-(soc_cvx-0.2)*BATTpmax;
234         (x(3,k)+t*Dx_pd(3))*Vb-soc_cvx*BATTpmax;
235         (x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin)];
236 else
237     f = [ -(x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin);
238         -(x(2,k)+t*Dx_pd(2))*Vb-(1-soc_cvx)*BATTpmax;
239         (x(2,k)+t*Dx_pd(2))*Vb-(soc_cvx-0.2)*BATTpmax;
240         (x(2,k)+t*Dx_pd(2))*Vb-soc_cvx*BATTpmax;

```

```

238         (x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin) ] ;
239     end
240
241     while ~( sum ( f < 0 ) == m )
242         t=beta*t ;
243
244         if statusMT(i)==1
245             f = [ (x(2,k)+t*Dx_pd(2))-MTpmax;
246                 -(x(2,k)+t*Dx_pd(2)) ;
247                 -(x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin) ;
248                 -(x(3,k)+t*Dx_pd(3))*Vb-(1-soc_cvx)*BATTpmax;
249                 (x(3,k)+t*Dx_pd(3))*Vb-(soc_cvx-0.2)*BATTpmax;
250                 (x(3,k)+t*Dx_pd(3))*Vb-soc_cvx*BATTpmax;
251                 (x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin) ] ;
252         else
253             f = [ -(x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin) ;
254                 -(x(2,k)+t*Dx_pd(2))*Vb-(1-soc_cvx)*BATTpmax;
255                 (x(2,k)+t*Dx_pd(2))*Vb-(soc_cvx-0.2)*
256                     BATTpmax;
257                 (x(2,k)+t*Dx_pd(2))*Vb-soc_cvx*BATTpmax;
258                 (x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin) ] ;
259         end
260     end
261
262     %% BACKTRACKING
263     while norm( func_res( x(:,k) + t*Dx_pd, lamda(:,k) + t*
264         Dlamda_pd, v(:,k) + t*Dv_pd, A, b, m, tau, statusMT(
265             i), Vb, MTpmax, cutil(i), BATTpmax,BATTpmin,soc_cvx
266             ) ) >...
267         (1-alpha*t) * norm(r_t(:,k))
268         t = beta*t ;
269     end
270
271     %% UPDATE x,  $\hat{I}z$ , v
272     x(:,k+1) = x(:,k) + t*Dx_pd ;
273     lamda(:,k+1) = lamda(:,k) + t*Dlamda_pd;
274     v(:,k+1) = v(:,k) + t*Dv_pd;
275
276     %% TERMINATION CONDITION
277     if norm(r_t_p(:,k))<=epsilon_feas && norm(r_t_d(:,k))<=
278         epsilon_feas && eta(k)<=epsilon
279         break ;

```

```

276         end
277         k = k+1;
278     end
279
280     if statusMT(i)==1
281         [optimum_primal_dual_interior_point , Futility_prim ,
          Fgen_prim , suc_prim] = objFun(x(1,k) , x(2,k) ,
          statusMT , toff , cutil(i) , i);
282     else
283         [optimum_primal_dual_interior_point , Futility_prim ,
          Fgen_prim , suc_prim] = objFun(x(1,k) , 0 , statusMT ,
          toff , cutil(i) , i);
284     end
285
286     fprintf( '\n\nPrimal Dual interior point method finished. ')
287     ;
288     fprintf( '\n\nOptimum Primal Dual Interior Point: %d' ,
289             optimum_primal_dual_interior_point);
290 %     fprintf( '\n\nOptimal Points from CVX: [ %d %d %d ]' ,
291 %     x_cvx_x(1) , x_cvx_x(2) , x_cvx_x(3) );
292
293     if statusMT(i)==1
294         fprintf( '\n\nOptimal Points from Primal Dual Interior
295             Point: [ %d %d %d ]' , x(1,k) , x(2,k) , x(3,k) );
296     else
297         fprintf( '\n\nOptimal Points from Primal Dual Interior
298             Point: [ %d 0 %d ]' , x(1,k) , x(2,k) );
299     end
300 %% WRITE TO BUS
301 %     p_bus(index_Util , i) = x_cvx_x(1);
302 %     p_bus(index_GenMT , i) = x_cvx_x(2);
303 %     p_bus(index_GenBTR , i) = x_cvx_x(3)*Vb;
304 %     val_obj(i) = cvx_optval;
305
306     p_bus(index_Util , i) = x(1,k);
307     if statusMT(i)==1
308         p_bus(index_GenMT , i) = x(2,k);
309         p_bus(index_GenBTR , i) = x(3,k)*Vb;
310     else
311         p_bus(index_GenMT , i) = 0;
312         p_bus(index_GenBTR , i) = x(2,k)*Vb;
313     end

```



```

311         val_obj(i) = optimum_primal_dual_interior_point;
312     %% UPDATE BATTERY
313     pbold = [pbold p_bus(index_GenBTR,i)];
314     [soc_cvx, pout_btr, qq3, num_of_batt] = battery( p_bus(
        index_GenBTR,i), pbold(1:end-1), BATTpmax);
315 end
316
317 loss_24_hour = sum(val_obj)
318 num(row,5+i+1) = loss_24_hour;
319 num(row,6:5+i) = val_obj;
320 csvwrite('Data.csv',num)
321 FileName = sprintf('%d_%d_%d_%d_DATA.csv', BATTpmax, PVpmax,
        WTpmax, MTpmax);
322 csvwrite(FileName,p_bus)
323 plot_from_excel(FileName);
324 fprintf('Simulation Ended\n\n') ;
325 end

```

C. OBJECTIVE FUNCTION

```

1 function [ObjFun,Futility,Fgen,suc] = objFun(putil, pmt,
        status_mt, toff, cutil, index)
2
3     if(index==1)
4         statusMT = status_mt(1);
5         statusMT_prev = 0;
6     else
7         statusMT = status_mt(index);
8         statusMT_prev = status_mt(index-1);
9     end
10
11     %% UTILITY
12     Futility = cutil .* putil;
13
14     %% GENERATOR MAINTAINANCE AND START-UP
15     stp = 3600; %
16     hot_startup = 30; %sec
17     cold_startup = 200; %sec
18
19     cooling_time = 520; %sec
20
21     if statusMT_prev==0 && statusMT==1
22         suc = ( ( hot_startup / stp ) + ( cold_startup / stp )
            * ( 1 - exp ( - toff / (cooling_time/stp) ) ) ); %

```

```

                                start-up cost
23     else
24         suc=0;
25     end
26
27     alpha=0.0074; beta=0.2333; cu=0.4333;
28     Fgen = ( alpha * (pmt*10^(-3)).^2 + beta * pmt*10^(-3) + cu
                + suc) .* statusMT;
29
30     %% OBJECTIVE FUNCTION
31     ObjFun = sum( Futility + Fgen); % $/Wh
32 end

```

D. GRADIENT OF OBJECTIVE

```

1 function [gradient] = D_f_obj(pmt, statusMT, cutil)
2     alpha=0.0074*10^(-3); beta=0.2333*10^(-3);
3
4     if statusMT==1
5         gradient = [ cutil ( 2*alpha*pmt + beta ) 0]';
6     else
7         gradient = [ cutil 0]';
8     end
9 end

```

E. HESSIAN OF OBJECTIVE

```

1 function [Hess] = D2_f_obj(statusMT)
2     alpha=0.0074*10^(-3);
3     if statusMT==1
4         Hess = [0 0 0; 0 2*alpha 0; 0 0 0];
5     else
6         Hess = [0 0; 0 0];
7     end
8 end

```

F. CALCULATION OF RESIDUALS

```

1 function r_t = func_res(x, lamda, v, A, b, m, tau, statusMT, Vb, MTpmax,
    cutil, BATTpmax, BATTpmin, soc)
2
3     if statusMT==1
4         f = [ x(2)-MTpmax;
5             -x(2);

```

```

6         -x(3)*Vb-abs(BATTpmin);
7         -x(3)*Vb-(1-soc)*BATTpmax;
8         x(3)*Vb-(soc-0.2)*BATTpmax;
9         x(3)*Vb-soc*BATTpmax;
10        x(3)*Vb-abs(BATTpmin)];
11
12        Df = [0 1 0;
13              0 -1 0;
14              0 0 -Vb;
15              0 0 -Vb;
16              0 0 Vb;
17              0 0 Vb;
18              0 0 Vb];
19
20        r_t_d = D_f_obj(x(2), statusMT, cutil) + Df'*lamda + A
21                '*v;
22
23        else
24            f = [ -x(2)*Vb-abs(BATTpmin);
25                  -x(2)*Vb-(1-soc)*BATTpmax;
26                  x(2)*Vb-(soc-0.2)*BATTpmax;
27                  x(2)*Vb-soc*BATTpmax;
28                  x(2)*Vb-abs(BATTpmin)];
29
30            Df = [0 -Vb;
31                  0 -Vb;
32                  0 Vb;
33                  0 Vb;
34                  0 Vb];
35
36            r_t_d = D_f_obj(0, statusMT, cutil) + Df'*lamda + A'*v;
37
38
39        r_t_c = -diag(lamda)*f - (1/tau)*ones(m,1);
40        r_t_p = A*x-b;
41        r_t = [r_t_d; r_t_c; r_t_p];
42    end

```

G. DC AC INVERTER

```

1    function Pout = dc_ac_inverter(P)
2        Pout=0.97*P;
3    end

```

H. BATTERY UNIT

```

1 function [soc , pout , Ah_cur_con , K] = battery( pbnew , pbold ,
   pbmax)
2
3 %      nb=0.9;      %efficiency
4      nb=1;      %efficiency
5      dT=1;      %1 hour
6      Vb=12;      %inital Voltage
7      Ahinit=25; %inital Ampere-hours
8
9      K = pbmax/(Vb*Ahinit*dT/nb);% total batteries
10
11      Ahinit= Ahinit*K;
12
13      Ah_cur_con = (pbnew/pbmax)*Ahinit;
14
15      Ah_con = (sum(pbold)/pbmax)*Ahinit;
16
17      Ah_remain = (Ahinit)-Ah_con;
18
19      soc = (Ah_remain-Ah_cur_con)/(Ahinit);
20
21      pout(soc>=0) = (Vb*(Ah_cur_con));
22
23      pout(soc>1) = -(Vb*(Ah_con));
24
25      pout(soc<0) = 0;
26
27      soc(soc<0) = (Ah_remain)/(Ahinit);
28
29      soc(soc>1) = 1;
30 end

```

I. MICRO-TURBINE

```

1 function [pout , emission] = microturbine(p,status ,prated)
2      emissionSO2 = 0.206*10^(-6); % kg/Wh
3      emissionCO2 = 649*10^(-6);
4      emissionNOx = 9.89*10^(-6);
5
6      if(status==1)
7          pout(p>prated) = prated;
8          pout(p<=prated && p>=0)=p;

```

```

9         else
10             pout = 0;
11         end
12
13         emission = [emissionSO2*pout emissionCO2*pout emissionNOx*
14                     pout];
15     end

```

J. PLOT FIGURES FROM CSV FILE

```

1 function plot_from_excel(name)
2     Dat=csvread(name,0,0);
3     h=figure;
4     set(gcf,'Visible','off');
5     t=1:24;
6     plot(t,Dat(1,1:24),t,Dat(2,1:24),t,Dat(3,1:24),t,Dat
7           (4,1:24),t,Dat(5,1:24),t,Dat(6,1:24));
8     xlabel('Time(Hour of day)')
9     ylabel('Power Output (Watt)')
10    grid on;
11    title('CVX')
12    legend('MT','PV','WT','BTR','Load','Utility');
13    baseFileName = sprintf('%s.jpg', name);
14    saveas(h, baseFileName);
15 end

```

K. POWER BUS PER DAYHOUR

```

1 function [p_bus] = power_bus_per_dayhour(pertime,stp_time)
2
3     if(stp_time<=pertime)
4         if (stp_time==1 || stp_time==8 || stp_time==9 ||
5             stp_time==16 || stp_time==17 || stp_time==24)
6             p_bus = 30000;
7         elseif (stp_time==11 || stp_time==12 || stp_time==15 ||
8             stp_time==18 || stp_time==19 || stp_time==23)
9             p_bus = 60000;
10        elseif (stp_time==10)
11            p_bus = 90000;
12        elseif (stp_time==21 || stp_time==22)
13            p_bus = 150000;
14        elseif (stp_time==14 || stp_time==20)
15            p_bus = 210000;
16        elseif (stp_time==13)
17            p_bus = 30000;
18        else
19            p_bus = 0;
20        end
21    end
22 end

```

```

15         p_bus = 300000;
16     else
17         p_bus = 3000;
18     end
19 else
20     if (mod(stp_time,pertime)==1 || mod(stp_time,pertime)
        ==8 || mod(stp_time,pertime)==9 || mod(stp_time,
        pertime)==16 || mod(stp_time,pertime)==17 || mod(
        stp_time,pertime)==24)
21         p_bus = 30000;
22     elseif (mod(stp_time,pertime)==11 || mod(stp_time,
        pertime)==12 || mod(stp_time,pertime)==15 || mod(
        stp_time,pertime)==18 || mod(stp_time,pertime)==19
        || mod(stp_time,pertime)==23)
23         p_bus = 60000;
24     elseif (mod(stp_time,pertime)==10)
25         p_bus = 90000;
26     elseif (mod(stp_time,pertime)==21 || mod(stp_time,
        pertime)==22)
27         p_bus = 150000;
28     elseif (mod(stp_time,pertime)==14 || mod(stp_time,
        pertime)==20)
29         p_bus = 210000;
30     elseif (mod(stp_time,pertime)==13)
31         p_bus = 300000;
32     else
33         p_bus = 3000;
34     end
35 end
36
37 end

```

L. PHOTOVOLTAIC ARRAY

```

1 function p = pv_array(Irrad, Tempr, prated)
2     %% Initialize
3
4     p=0;
5     totalPanels = (40*prated)/20000;
6     totalCells = 172;
7     Vcell = 0:0.01:0.665; % voltage vector??
8     Ppv = zeros(totalPanels, length(Vcell));
9     Tempr(1:totalPanels) = Tempr;
10    Irrad(1:totalPanels) = Irrad;

```

```

11
12 %% Panels Computation
13 for i=1:totalPanels
14     Pcell(i,:) = Vcell .* solarcell( Vcell , Irrad(i) ,
15         Tempr(i) );
16
17     Ppv(i,:) = totalCells * Pcell(i,:);
18
19     pmax(i) = max(Ppv(i,:));
20
21     p = p + dc_ac_inverter( pmax(i) );
22 end
end

```

M. SOLAR CELL OF PHOTOVOLTAIC ARRAY

```

1 function Ia = solarcell(Va,Suns,TaC)
2     k = 1.38e-23; % Boltzmann constant
3     q = 1.60e-19; % Electron
4     n = 1.2; % Quality factor for the diode. n=1.2 for
5         crystalline
6     Vg = 1.12; % [eV]
7     T1 = 273 + 25; % Kelvin
8
9     Voc_T1 = 0.665; % Open-current voltage at T1 [V]
10    Isc_T1 = 5.75; % Short-circuit current at T1 [A]
11
12    K0 = 3.5/1000; % Current/Temperature coefficient [A/K]
13
14    dVdI_Voc = -0.00985; % dV/dI coefficient at Voc [A/V]
15
16    TaK = 273 + TaC; % Convert cell's temperature from Celsius
17        to Kelvin [K]
18
19    IL_T1 = Isc_T1 * Suns; % Compute IL depending the suns at
20        T1
21    IL = IL_T1 + K0 * (TaK - T1); % Apply the temperature
22        effect
23
24    I0_T1 = Isc_T1 / ( exp( q * Voc_T1 / ( n * k * T1 ) ) - 1
25        );
26    I0 = I0_T1 * ( TaK / T1 ) .^ (3/n) .* exp( -q * Vg / ( n *
27        k ) .* ( (1./TaK) - (1/T1) ) );
28
29

```

```

23     Xv = I0_T1 * q / ( n * k * T1 ) * exp( q * Voc_T1 / ( n * k
        * T1 ) );
24     Rs = - dVdI_Voc - 1/Xv; %Compute Rs Resistance
25
26     Vt_Ta = n * k * TaK / q;
27     Ia = zeros(size(Va)); %Initialize Ia vector
28
29     % Compute Ia with Newton method
30     for j=1:5;
31         Ia = Ia - (IL - Ia - I0.*( exp( (Va + Ia .* Rs) ./
            Vt_Ta ) - 1 ) ) ./ (-1 - ( I0.*( exp( (Va+Ia.*Rs) ./
            Vt_Ta ) -1 ) ) .* Rs ./ Vt_Ta );
32     end
33 end

```

N. WINDTURBINE

```

1 function p = windturbine(v,status,prated)
2     % wind.m
3     % v
        =[10,11,12,12,12,12,12,13,15,16,17,17,18,19,19,20,21,22,23,24,25,26,27
4
5     von = 3;           % cut-on speed (m/s)
6     vc = 14;           % corner speed (m/s)
7     vout = 25;         % cut-out speed (m/s)
8     diameter = 10;
9     rho = 0.1;
10    capacity_factor=0.2;
11    %p = zeros( size(v) );
12
13    if(status==1)
14        swept_area = pi * (diameter/2)^2;
15        % Below cut-on
16        p(v < von) = 0;
17        % Ramp up (use model)
18        I = (v >= von & v < vc);
19        p(I) = 0.5 * swept_area * (v(I).^3) * rho *
            capacity_factor; % P = 1/2 * v^3 * A * p * Cp
20        % At rated power
21        p(v >= vc & v <= vout) = prated;
22        % Above cut-out
23        p(v > vout) = 0;
24

```



```

25 %           figure(1);%Plot the I-V characteristic curve
26 %           axis([0,30,0,800])
27 %           plot(v,p);
28 %           grid;
29     end
30 end

```

N.1. SECOND PROSUMER'S PROFILE (NOT PERMITTED TO SELL TO UTILITY)

O. ENERGY MANAGEMENT SYSTEM – CVX – PRIMAL DUAL INTERIOR POINT METHOD

```

1  clear all; close all; clc;
2
3  %% READ CSV FILE
4  row=1;
5
6  numm=[ 120000  -100000  120000  120000  40000
7         80000   -75000  100000  100000  750000
8         50000   -45000  120000  120000  80000  ];
9
10 for row=1:3
11   %% READ DATA
12   fprintf('Simulation Started\n\n') ;
13
14   num = csvread('Data.csv');
15
16   BATTpmax = num(row,1);
17   BATTpmin = num(row,2);
18   PVpmax = num(row,3);
19   WTpmax = num(row,4);
20   MTpmax = num(row,5);
21
22   %% INIALIZATION
23
24   %TIME STEPS, AC BUS
25   time_step = 24;
26   totalUnits = 5;
27   p_bus = zeros(totalUnits,time_step);
28   index_GenMT = 1; index_GenPV = 2; index_GenWT = 3;
29       index_GenBTR = 4; index_Load = 5; index_Util = 6;
30       index_Dump = 7;
31
32   % MICROURBINE

```

```

31     mut = 6000;% minimum up time (sec)
32     mdt = 200;% minimum down time (sec)
33     ton=0;% counting the times turned on
34     toff=0;% counting the times turned off
35     statusMT=ones(1,time_step);
36
37     % WIND TURBINE
38     statusWT=ones(1,time_step);
39
40     % BATTERY
41     Vb=12;% initial voltage
42     pbold = 0;
43     [soc_cvx, qq1, qq2, num_of_batt] = battery( 0, pbold,
44         BATTpmax);
45
46     % UTILITY
47     vwind = [10.12 11.654 12.43 12.7654 12.123 12.5643 12.626
48         13.845 15.113 16.322 17.11 10.12 11.654 12.43 12.7654
49         10.12 11.654 12.43 12.7654 24.6363 25.346 26.7562 27.23
50         28.88];
51     irrad = [0.0075 0.3948 0.674 0.789 0.519 0.555 0.493
52         0.935 0.891 0.583 0.514 0.35 0.169 0.6472 0.1508
53         0.1324 0.10385 0.8206 0.1069 0.3603 0.7578 0.8679 0.0767
54         0.5666];
55     cutil = 10(-3)*[0.0787 0.0785 0.0785 0.0789 0.088 0.088
56         0.089 0.09 0.0885 0.0805 0.0815 0.0825 0.079 0.078 0.089
57         0.0866 0.0799 0.082 0.08 0.085 0.09 0.0788 0.0788
58         0.085];
59     tempr = [33.57 66.38 54.61 65.75 42.07 70.09 30.81
60         54.57 57.26 28.02 65.66 72.92 62.93 49.59 54.64
61         43.63 32.40 39.59 56.84 47.46 68.34 59.68 55.18
62         44.52];
63
64     stp=3600;
65     %% EVALUATION
66     for i=1:time_step
67         %% CONSTRAINTS MT
68         if( ton >= mut/stp )
69             statusMT(i) = 0;
70         end
71
72         if( toff >= mdt/stp )
73             statusMT(i) = 1;
74         end
75     end

```

```

62
63         if( statusMT(i) == 0 )
64             ton=0;
65             toff=toff+1;
66             if(i<length(statusMT))
67                 statusMT(i+1)=0;
68         end
69     else
70         toff = 0;
71         ton = ton+1;
72         if(i<length(statusMT))
73             statusMT(i+1)=1;
74         end
75     end
76
77     %% AC BUS
78     p_bus(index_GenWT,i) = windturbine(vwind(i),statusWT(i)
79         ,WTpmax);
80     p_bus(index_Load,i) = power_bus_per_dayhour(time_step,i
81         );
82     if(mod(i,time_step)<=19 && mod(i,time_step)>=7)
83         p_bus(index_GenPV,i) = pv_array(irrad(i),tempr(i),
84             PVpmax);
85     end
86
87     %% CVX
88     n=4;
89     cvx_begin
90         variable x(n,1)
91         minimize objFun(x(1), x(2), statusMT, toff, cutil(
92             i), i, x(4));
93         subject to
94             x(1)+x(2)+x(3)*Vb-x(4)==p_bus(index_Load,i)-(
95             p_bus(index_GenWT,i)+p_bus(index_GenPV,i)); %%% x(4) —>
96             DUMP LOAD
97             %-x(3)<=0;
98             %#####
99             -x(3)*Vb-abs(BATTpmin)<=0;
100             -x(3)*Vb-(1-soc_cvx)*BATTpmax<=0;
101             %#####
102             x(3)*Vb-(soc_cvx-0.2)*BATTpmax<=0;
103             x(3)*Vb-soc_cvx*BATTpmax<=0;
104             x(3)*Vb-abs(BATTpmin)<=0;
105             x(2)-MTpmax*statusMT(i)<=0;
106             -x(2)<=0;

```

```

100 %           -x(1) <= 0;
101 %           -x(4) <= 0;
102 %   cvx_end
103 %
104 %   x_cvx_x = cvx_optpnt.x;
105 %
106 %   fprintf('\nCVX finished. ');
107
108   if soc_cvx < 0.201;
109       soc_cvx = 0.201;
110   end
111 %% PRIMAL DUAL
112 clear r_t_d r_t_c r_t_p r_t
113 clear array_mtr
114 clear Dy Dx_pd Dlamda_pd Dv_pd
115 clear x lamda v
116 clear eta tau f Df
117
118 p=1;
119 mu=10;
120 alpha = 1e-2;
121 beta = 5e-1;
122 epsilon_feas = 1e-6;
123 epsilon = 1e-6;
124
125 b = p_bus(index_Load, i) - p_bus(index_GenWT, i) - p_bus(
    index_GenPV, i);
126
127 v = rand(p, 1);
128
129 if statusMT(i) == 1
130     m=9;
131     n=4;
132     A = [1 1 Vb -1];
133     x = rand(n, 1);
134
135     f = [ x(2) - MTpmax;
136          -x(2);
137          -x(3)*Vb - abs(BATTpmin);
138          -x(3)*Vb - (1 - soc_cvx)*BATTpmax;
139          x(3)*Vb - (soc_cvx - 0.2)*BATTpmax;
140          x(3)*Vb - soc_cvx*BATTpmax;
141          x(3)*Vb - abs(BATTpmin);
142          -x(1);

```

```

143         -x(4) ] ;
144
145     Df = [0  1  0  0;
146           0 -1  0  0;
147           0  0 -Vb  0;
148           0  0 -Vb  0;
149           0  0  Vb  0;
150           0  0  Vb  0;
151           0  0  Vb  0;
152           -1  0  0  0;
153           0  0  0  -1];
154 else
155     m=7;
156     n=3;
157     A = [1  Vb  -1];
158     x = rand(n,1) ;
159
160     f = [ -x(2)*Vb-abs(BATTpmin) ;
161          -x(2)*Vb-(1-soc_cvx)*BATTpmax;
162          x(2)*Vb-(soc_cvx-0.2)*BATTpmax;
163          x(2)*Vb-soc_cvx*BATTpmax;
164          x(2)*Vb-abs(BATTpmin) ;
165          -x(1) ;
166          -x(3) ] ;
167
168     Df = [0 -Vb  0;
169           0 -Vb  0;
170           0  Vb  0;
171           0  Vb  0;
172           0  Vb  0;
173           -1  0  0;
174           0  0  -1];
175 end
176
177     Hessian = zeros(m,n) ;
178     Hessian_f0 = D2_f_obj(statusMT(i)) ;
179
180     lamda = -1./f;
181
182     k = 1;
183
184     %% Repeat
185     while (1)
186         %% DETERMINE tau

```

```

187         if statusMT(i)==1
188             f = [ x(2,k)-MTpmax;
189                 -x(2,k);
190                 -x(3,k)*Vb-abs(BATTpmin);
191             -x(3,k)*Vb-(1-soc_cvx)*BATTpmax;
192                 x(3,k)*Vb-(soc_cvx-0.2)*BATTpmax;
193                 x(3,k)*Vb-soc_cvx*BATTpmax;
194                 x(3,k)*Vb-abs(BATTpmin);
195                 -x(1,k);
196                 -x(4,k) ];
197         else
198             f = [ -x(2,k)*Vb-abs(BATTpmin);
199             -x(2,k)*Vb-(1-soc_cvx)*BATTpmax;
200                 x(2,k)*Vb-(soc_cvx-0.2)*BATTpmax;
201                 x(2,k)*Vb-soc_cvx*BATTpmax;
202                 x(2,k)*Vb-abs(BATTpmin);
203                 -x(1,k);
204                 -x(3,k) ];
205         end
206
207         eta(k) = -(f)' * lamda(:,k);
208         tau = (mu * n)/eta(k);
209         t_max = 1;
210         t = 0.99*t_max;
211         %% UPDATE residuals
212         if statusMT(i)
213             r_t_d(:,k) = D_f_obj(x(2,k), statusMT(i), cutil
214                 (i)) + Df'*lamda(:,k) + A'*v(:,k);
215             r_t_c(:,k) = -diag(lamda(:,k))*f - (1/tau)*ones
216                 (m,1);
217             r_t_p(:,k) = A*x(:,k)-b;
218             r_t(:,k) = [r_t_d(:,k); r_t_c(:,k); r_t_p(:,k)
219                 ];
220         else
221             r_t_d(:,k) = D_f_obj(0, statusMT(i), cutil(i))
222                 + Df'*lamda(:,k) + A'*v(:,k);
223             r_t_c(:,k) = -diag(lamda(:,k))*f - (1/tau)*ones
224                 (m,1);
225             r_t_p(:,k) = A*x(:,k)-b;
226             r_t(:,k) = [r_t_d(:,k); r_t_c(:,k); r_t_p(:,k)
227                 ];
228         end

```

```

224 %% UPDATE the invertible matrix for  $\hat{\mathbf{T}}\mathbf{y}$ 
225 array_mtr = [ Hessian_f0 Df' A'; ...
226 -diag(lamda(:,k))*Df -diag(f) zeros(m,p); ...
227 A zeros(p,m) zeros(p,p) ];
228
229 %% UPDATE  $\hat{\mathbf{T}}\mathbf{y}$ 
230 Dy = -inv(array_mtr)*r_t(:,k);
231 Dx_pd = Dy(1:n);
232 Dlamda_pd = Dy(n+1:n+m);
233 Dv_pd = Dy(n+m+1:size(Dy,1));
234
235 %%  $\hat{\mathbf{I}}\dot{\mathbf{z}} + \mathbf{D}\hat{\mathbf{I}}\dot{\mathbf{z}} > 0$ 
236 while ~(sum((lamda(:,k)+t*Dlamda_pd) > 0) == m)
237     t=beta*t;
238 end
239
240 %%  $f(\mathbf{x}) < 0$ 
241 if statusMT(i)==1
242     f = [ (x(2,k)+t*Dx_pd(2))-MTpmax;
243           -(x(2,k)+t*Dx_pd(2));
244           -(x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin);
245           -(x(3,k)+t*Dx_pd(3))*Vb-(1-soc_cvx)*BATTpmax;
246           (x(3,k)+t*Dx_pd(3))*Vb-(soc_cvx-0.2)*BATTpmax;
247           (x(3,k)+t*Dx_pd(3))*Vb-soc_cvx*BATTpmax;
248           (x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin);
249           -(x(1,k)+t*Dx_pd(1));
250           -(x(4,k)+t*Dx_pd(4)) ];
251 else
252     f = [ -(x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin);
253           -(x(2,k)+t*Dx_pd(2))*Vb-(1-soc_cvx)*BATTpmax;
254           (x(2,k)+t*Dx_pd(2))*Vb-(soc_cvx-0.2)*BATTpmax;
255           (x(2,k)+t*Dx_pd(2))*Vb-soc_cvx*BATTpmax;
256           (x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin);
257           -(x(1,k)+t*Dx_pd(1));
258           -(x(3,k)+t*Dx_pd(3)) ];
259 end
260
261 while ~(sum(f < 0) == m)
262     t=beta*t;
263
264     if statusMT(i)==1
265         f = [ (x(2,k)+t*Dx_pd(2))-MTpmax;
266               -(x(2,k)+t*Dx_pd(2));

```

```

267         -(x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin);
268     -(x(3,k)+t*Dx_pd(3))*Vb-(1-soc_cvx)*BATTpmax;
269     (x(3,k)+t*Dx_pd(3))*Vb-(soc_cvx-0.2)*
        BATTpmax;
270     (x(3,k)+t*Dx_pd(3))*Vb-soc_cvx*BATTpmax;
271     (x(3,k)+t*Dx_pd(3))*Vb-abs(BATTpmin);
272     -(x(1,k)+t*Dx_pd(1));
273     -(x(4,k)+t*Dx_pd(4));
274     else
275         f = [ -(x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin);
276             -(x(2,k)+t*Dx_pd(2))*Vb-(1-soc_cvx)*BATTpmax;
277             (x(2,k)+t*Dx_pd(2))*Vb-(soc_cvx-0.2)*
                BATTpmax;
278             (x(2,k)+t*Dx_pd(2))*Vb-soc_cvx*BATTpmax;
279             (x(2,k)+t*Dx_pd(2))*Vb-abs(BATTpmin);
280             -(x(1,k)+t*Dx_pd(1));
281             -(x(3,k)+t*Dx_pd(3)) ];
282     end
283
284 end
285
286 %% BACKTRACKING
287 while norm( func_res( x(:,k) + t*Dx_pd, lamda(:,k)
    + t*Dlamda_pd, v(:,k) + t*Dv_pd, A, b, m, tau,
    statusMT(i), Vb, MTpmax, soc_cvx, cutil(i),
    BATTpmax,BATTpmin) ) >...
    (1-alpha*t) * norm(r_t(:,k))
288     t = beta*t;
289 end
290
291 %% UPDATE x, \dot{z}, v
292 x(:,k+1) = x(:,k) + t*Dx_pd;
293 lamda(:,k+1) = lamda(:,k) + t*Dlamda_pd;
294 v(:,k+1) = v(:,k) + t*Dv_pd;
295
296 %% TERMINATION CONDITION
297 if norm(r_t_p(:,k))<=epsilon_feas && norm(r_t_d(:,k)
    )<=epsilon_feas && eta(k)<=epsilon
298     break;
299 end
300 k = k+1;
301 end
302
303

```



```

304         if statusMT(i)==1
305             [optimum_primal_dual_interior_point , Futility_prim
               , Fgen_prim, suc_prim] = objFun(x(1,k), x(2,k),
               statusMT, toff, cutil(i),i,x(4,k));
306         else
307             [optimum_primal_dual_interior_point , Futility_prim
               , Fgen_prim, suc_prim] = objFun(x(1,k), 0,
               statusMT, toff, cutil(i),i,x(3,k));
308         end
309
310         optimum_primal_dual_interior_point
311
312         fprintf('\nPrimal Dual interior point method finished.
               ');
313
314         %% WRITE TO BUS
315         %       p_bus(index_Util,i) = x_cvx_x(1);
316         %       p_bus(index_GenMT,i) = x_cvx_x(2);
317         %       p_bus(index_GenBTR,i) = x_cvx_x(3)*Vb;
318         %       p_bus(index_Dump,i) = x_cvx_x(4);
319         %       val_obj(i) = cvx_optval;
320
321         p_bus(index_Util,i) = x(1,k);
322         if statusMT(i)==1
323             p_bus(index_GenMT,i) = x(2,k);
324             p_bus(index_GenBTR,i) = x(3,k)*Vb;
325             p_bus(index_Dump,i) = x(4,k);
326         else
327             p_bus(index_GenBTR,i) = x(2,k)*Vb;
328             p_bus(index_Dump,i) = x(3,k);
329         end
330         val_obj(i) = optimum_primal_dual_interior_point;
331         %% UPDATE BATTERY
332         pbold = [pbold p_bus(index_GenBTR,i)];
333         [soc_cvx, pout_btr, qq1,num_of_batt] = battery( p_bus(
               index_GenBTR,i), pbold(1:end-1), BATTpmax);
334     end
335
336     loss_24_hour = sum(val_obj)
337     num(row,6+i+1) = loss_24_hour;
338     num(row,7:6+i) = val_obj;
339     csvwrite('Data.csv',num)
340     FileName = sprintf('%d_%d_%d_%d_DATA.csv', BATTpmax,
               PVpmax, WTpmax, MTpmax);

```

```

341     csvwrite(FileName,p_bus)
342     plot_from_excel(FileName);
343     fprintf( 'Simulation Ended\n\n' ) ;
344 end

```

P. OBJECTIVE FUNCTION

```

1  function [ObjFun,Futility,Fgen,suc] = objFun(putil , pmt ,
    status_mt , toff ,cutil ,index , pdump)
2
3      if(index==1)
4          statusMT = status_mt(1);
5          statusMT_prev = 0;
6      else
7          statusMT = status_mt(index);
8          statusMT_prev = status_mt(index-1);
9      end
10
11     %% UTILITY
12     Futility = cutil * putil;
13
14     %% GENERATOR MAINTAINANCE AND START-UP
15     stp = 3600; %
16     hot_startup = 30; %sec
17     cold_startup = 200; %sec
18
19     cooling_time = 520; %sec
20
21     if statusMT_prev==0 && statusMT==1
22         suc = ( ( hot_startup / stp ) + ( cold_startup / stp )
                * (1 - exp ( - toff / (cooling_time/stp) )) ) * (1 -
                statusMT_prev); % start-up cost
23     else
24         suc=0;
25     end
26
27     alpha=0.0074; beta=0.2333; cu=0.4333;
28     Fgen = ( alpha * (pmt*10^(-3)).^2 + beta * pmt*10^(-3) + cu
              + suc) .* statusMT;
29
30     %% DUMP LOAD
31     Fdump = cutil * pdump;
32     %% OBJECTIVE FUNCTION
33     ObjFun = sum( Futility + Fgen + Fdump); % $/Wh

```

34 end

Q. GRADIENT OF OBJECTIVE

```
1 function [gradient] = D_f_obj(pmt, statusMT, cutil)
2     alpha=0.0074*10^(-3); beta=0.2333*10^(-3);
3     if statusMT==1
4         gradient = [cutil ( 2*alpha*pmt + beta ) 0 cutil]';
5     else
6         gradient = [cutil 0 cutil]';
7     end
8 end
```

R. HESSIAN OF OBJECTIVE

```
1 function [Hess] = D2_f_obj(statusMT)
2     alpha=0.0074*10^(-3);
3     if statusMT==1
4         Hess = [0 0 0 0; 0 2*alpha 0 0; 0 0 0 0; 0 0 0 0];
5     else
6         Hess = [0 0 0; 0 0 0; 0 0 0];
7     end
8 end
```

S. CALCULATION OF RESIDUALS

```
1 function r_t = func_res(x, lamda, v, A, b, m, tau, statusMT, Vb, MTpmax,
2 soc, cutil, BATTpmax, BATTpmin)
3     if statusMT==1
4         f = [ x(2)-MTpmax;
5               -x(2);
6               -x(3)*Vb-abs(BATTpmin);
7               -x(3)*Vb-(1-soc)*BATTpmax;
8               x(3)*Vb-(soc-0.2)*BATTpmax;
9               x(3)*Vb-soc*BATTpmax;
10              x(3)*Vb-abs(BATTpmin);
11              -x(1);
12              -x(4)];
13
14         Df = [0 1 0 0;
15               0 -1 0 0;
16               0 0 -Vb 0;
17               0 0 -Vb 0;
```

```

18         0 0 Vb 0;
19         0 0 Vb 0;
20         0 0 Vb 0;
21         -1 0 0 0;
22         0 0 0 -1];
23
24     r_t_d = D_f_obj(x(2), statusMT, cutil) + Df'*lamda + A
           '*v;
25
26     else
27         f = [ -x(2)*Vb-abs(BATTpmin);
28               -x(2)*Vb-(1-soc)*BATTpmax;
29               x(2)*Vb-(soc-0.2)*BATTpmax;
30               x(2)*Vb-soc*BATTpmax;
31               x(2)*Vb-abs(BATTpmin);
32               -x(1);
33               -x(3) ];
34
35         Df = [0 -Vb 0;
36               0 -Vb 0;
37               0 Vb 0;
38               0 Vb 0;
39               0 Vb 0;
40               -1 0 0;
41               0 0 -1];
42
43         r_t_d = D_f_obj(0, statusMT, cutil) + Df'*lamda + A'*v;
44     end
45
46     r_t_c = -diag(lamda)*f - (1/tau)*ones(m,1);
47     r_t_p = A*x-b;
48     r_t = [r_t_d; r_t_c; r_t_p];
49 end

```