

Operation of the Multiple Energy System with Optimal Coordination of the Consumers in Energy Market

Ngakan Ketut Acwin DWIJENDRA¹, I Gusti Ngurah Kerta ARSANA², Sulieman Ibraheem Shelash Al-HAWARY³, A. S PRAKAASH⁴, Rosario Mireya ROMERO PARRA^{5*}, Abduladheem TURKI JALIL⁶, Ali Thaeer HAMMID⁷

^{1,2}Faculty of Engineering, Udayana University, Bali, Indonesia

³Department of Business Administration, Business School, Al al-Bayt University, Mafrq, Jordan

⁴Department of Mathematics, Panimalar Engineering College Poonamalle, Chennai Tamilnadu, India

⁵Universidad Continental, Lima, Perú

⁶Medical Laboratories Techniques Department, Al-Mustaqbal University College, Babylon, Hilla, Iraq

⁷Bilad Alrafidain University College, Diyala, Iraq

Received 21.05.2022; accepted 11.11.2022

Abstract – In this paper, optimal coordination of the demand side under uncertainty of the energy price in energy market is studied. The consumers by demand response programs (DRPs) have optimal role in minimization of the energy generation costs in multiple energy system. The consumers can participate via local generation strategy (LGS) and demand curtailment strategy (DCS). The optimal coordination is considered as two stage optimization, in which minimization of the consumers' bills is done in first stage. In following, the minimization of the generation costs is performed in second stage optimization. The LGS is taken into account through optimal discharging of plug electric vehicles (PEVs). Finally, numerical simulation is implemented to show superiority of the proposed approach to minimization of the energy generation costs.

Keywords – Demand curtailment strategy (DCS); demand response programs (DRPs); multiple energy system; onsite generation strategy (LGS); optimal coordination.

Nomenclature

t, T	Time index	hour
s, S	Scenario index	–
n, NC	Consumer index	–
EC, NGC	Electrical company (EC) and natural gas company (NGC)	–
D_E^n	Total electrical demand at n^{th}	kWh
$D^{n_{\text{NCL}}}, D^{n_{\text{CL}}}$	Non-curtable demand, curtable demand, respectively	kWh
$P_{\text{PEV}}^{\text{ch}}, P_{\text{PEV}}^{\text{dch}}$	Power charge and discharge of the PEV, respectively	kWh
P^{PEV}	Power rate of the PEV	kWh
$P_{\text{EC}}, P_{\text{CHP}}$	Power generated by EC and power of CHP, respectively	kWh

* Corresponding author.

E-mail address: Parra.romero.ac@gmail.com

P_{GAS}	Gas generated by NGC	m^3
$H_{\text{CHP}}, H_{\text{BO}}$	Heat generated by CHP and boiler, respectively	kWh
$C_{\text{EP}}, C_{\text{GP}}$	Electrical price and gas price, respectively	\$
$C_{\text{EC}}, C_{\text{NGC}}$	Generation costs of the EC and NGC	\$
$\eta^{\text{ch}}, \eta^{\text{dch}}$	PEV efficiency in charge and discharge modes, respectively	%
u_{PEV}	Binary variable of PEV (1 = discharge mode and 0 = otherwise)	–

1. INTRODUCTION

Integration of the smart grids technology in energy systems have provided new revolution in energy systems' infrastructures [1]–[5]. In these infrastructures, demand side have relevance with generation side at any times [6]–[10]. As well, energy companies using these infrastructures can be managed self-grids to decrease energy losses and costs [11]. On the other side, developing urbanization in many countries and consumption of the fossil fuels to energy generation are increasing in the power plants [12]–[14]. Hence, participation of the demand side in optimal energy consumption has direct effects on economic and technical indices. This participation can do by energy price signals and demand response programs (DRPs) in the energy markets [15]–[22]. The utilization of these strategies are various subject to energy system topology [23]–[25]. For instance, multiple energies like natural gas and electrical energy in smart grid technology with energy storage systems (ESSs) technology are effective strategies to optimal energy consumption [26]–[28]. In such energy systems, consumers can meet self-demand using multi-parallel energy resources [29]–[33]. Also, consumers by self-energy resource can have optimal role to the meet self-demand in high energy price [34]–[38]. The proposed topology of the smart multiple energy system in this paper is shown in Fig. 1. The proposed energy system including participants as follows:

- 1) Energy companies: The energy companies are electrical company (EC) and natural gas company (NGC). These companies have various prices at each hours in energy markets [39]–[41],
- 2) Distributed generators (DGs): The DGs are combined heat and power (CHP) units and boiler units. The DGs are fed by natural gas to energy generation [42]–[45],
- 3) Operator: This participant is main coordinator between generation side and demand side. The operator can provide optimal status of the system via informing energy prices to consumers at operation time [46], [47].

The operation of the various energy systems is studied by many researchers. Authors in [48] optimal power management of the electrical system considering uncertainty of the renewable energy systems is studied. In [49], the demand management by load shifting strategy in smart buildings to reducing the energy costs is proposed. The energy planning in the hybrid energy system based on optimal siting and sizing of the DGs is studied in [50]. The scheduling of the energy hub system in smart buildings without consideration of the DRPs is proposed in [51]. In [52], multi-objectives optimization of the multiple energy system is analyzed with stochastic modeling of the electrical price in energy market. In [53], the economic and environmental modelling of the electrical energy systems under risk assessment for electrical price is proposed. The assessment of the reliability index in electrical grids with attention to consumers' satisfaction level and minimization of the blackouts is studied in [54]. The co-optimization modelling is presented in [55] to energy-saving in electrical microgrids

via demand shifting strategy. In [56], optimal load control is implemented in multiple energy systems via uncoordinated and coordinated modelling of the DGs.

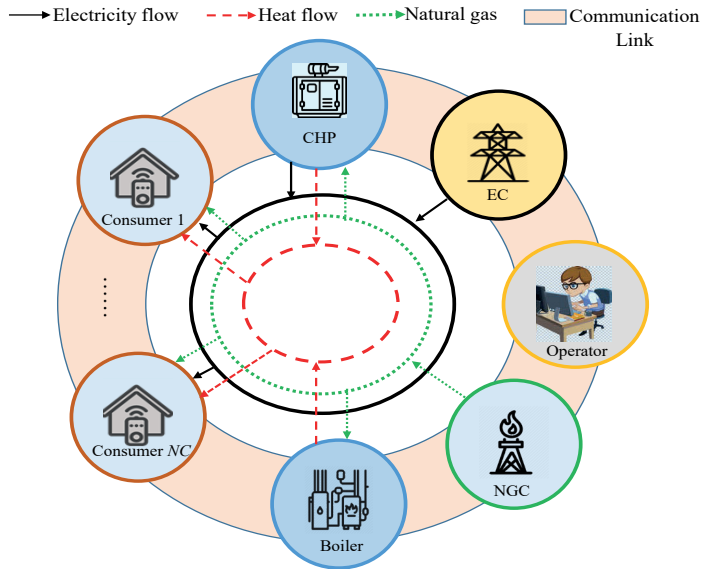


Fig. 1. Smart multiple energy system topology.

This paper presents two-stage energy optimization of the smart multiple energy system with DRPs and uncertainty of the electricity price (EP) and gas price (GP) in the energy market. The consumption costs of the consumers at first stage are optimized via demand curtailment strategy (DCS). As well, local generation strategy (LGS) is implemented by plug electric vehicles (PEVs) in second stage. The optimized load demand at first stage is considered in second stage alongside LGS to minimizing the generation costs. Thus, contributions and novelties of this work can be summarized as follow:

1. A modelling two-stage energy optimization is proposed in smart multiple energy system.
2. The DCS and LGS of the DRPs are considered in first and second stages to minimizing generation costs.
3. The PEVs are proposed to meet demand in peak time via LGS and optimal participation in second stage.
4. The energy prices including EP and GP are modelled under uncertainty approach.

2. UNCERTAINTY MODELLING

The uncertainty of the energy prices including EP and GP in energy market is modelled by lognormal probability density function Eq. (1) as follow [49]:

$$f(p) = \frac{1}{p\sigma\sqrt{2\pi}} e^{-\left(\frac{(\ln(p)-\mu)^2}{2\sigma^2}\right)}, \tag{1}$$

where p , μ and σ are distribution function parameter, mean value and standard deviation, respectively.

By Monte Carlo technique, scenarios or random variables for distribution function parameter (p) are generated at day-ahead. In this modelling, EP and GP are distribution function parameters (p). On the other side, probability in each scenario can be modelled by Eq. (2) [56]:

$$\pi_s = \pi_s^{\text{EP}} \times \pi_s^{\text{GP}}, \quad (2)$$

where π_s , π_s^{EP} and π_s^{GP} are probability of scenario s , probability of EP and GP at scenario s , respectively.

3. PEV MODELLING

The PEVs can be used by consumers as energy resource to feed self-demand. The PEV is taken into account as LGS in second stage optimization. The LGS modelling by PEVs is as follow [50]:

$$0 \leq P_{\text{PEV}}^{\text{ch}}(s,t) \leq [1 - u_{\text{PEV}}(s,t)] \times P_{\text{PEV}}^r \quad \forall s,t, \quad (3)$$

$$0 \leq P_{\text{PEV}}^{\text{dch}}(s,t) \leq u_{\text{PEV}}(s,t) \times P_{\text{PEV}}^r \quad \forall s,t, \quad (4)$$

$$\left[\sum_{t \in T} P_{\text{PEV}}^{\text{dch}}(s,t, \text{lg}) \times \frac{1}{\eta^{\text{dch}}} \right] - \left[\sum_{t \in T} P_{\text{PEV}}^{\text{ch}}(s,t) \times \eta^{\text{ch}} \right] = 0 \quad \forall s,t, \quad (5)$$

where charging power and discharging power of the PEV are modelled by Eq. (3) and Eq. (4), respectively. LGS by PEV are modelled by Eq. (5).

4. OPTIMIZATION APPROACH MODELLING

The two-stage optimization problem of the proposed approach is modelled in this section. The mathematical modeling for proposed approach is as follow.

4.1. First stage

The DCS in first stage optimization is modelled. In this strategy, consumers' bill subject to EP is minimized. Hence, electrical demand using DCS can be optimized. The objective function of the DCS is modelled by Eq. (6):

$$\min f_{fs} = \sum_{s=1}^S \pi_s \sum_{t=1}^T \sum_{n=1}^{NC} \{ C_{\text{EP}}(s,t) \times D_n^{\text{E}}(t) \} \quad \forall s,t,n, \quad (6)$$

subject to:

$$P_{\text{EC}}(s,t) = D_n^{\text{E}}(s,t) \quad \forall s,t,n, \quad (7)$$

$$D_n^{\text{E}}(s,t) = D_{\text{NCL}}^{\text{E}}(s,t) - D_{\text{CL}}^{\text{E}}(s,t) \quad \forall s,t,n, \quad (8)$$

$$0 \leq D_{\text{CL}}^{\text{E}}(s,t) \leq D_{\text{CL}}^{\text{E,max}} \quad \forall s,t,n, \quad (9)$$

where Eqs. (7)–(9) are power balance, electrical demand modelling and bound of the curtailable demand in DCS, respectively.

4.2. Second stage

The minimization of the generation costs as objective function in second stage optimization is considered. The modelling generation costs is formulated by Eq. (10) [2]:

$$\min f_{ss} = \sum_{s=1}^S \pi_s \sum_{t=1}^T \left\{ C_{EC}(s,t) + C_{GAS}(s,t) + \sum_{chp=1}^{CHP} C_{CHP}(s,t,CHP) + \sum_{bo=1}^{BO} C_{BO}(s,t,BO) \right\}, \quad (10)$$

where:

$$C_{EC}(t) = C_{EP}(s,t) \times P_{EC}(s,t) \quad \forall s,t, \quad (11)$$

$$C_{NGC}(s,t) = C_{GP}(s,t) \times P_{GAS}(s,t) \quad \forall s,t, \quad (12)$$

$$C_{CHP}(s,t,CHP) = \{C_{GP}(s,t) \times (H_{CHP}(s,t,CHP) + P_{CHP}(s,t,CHP))\} \quad \forall s,t,CHP, \quad (13)$$

$$C_{BO}(s,t,BO) = \{C_{GP}(s,t) \times (H_{BO}(s,t,BO))\} \quad \forall s,t,BO. \quad (14)$$

Here, Eqs. (11)–(14) are generation costs of the EC, NGC, CHP units and boilers units, respectively [2]. It should be mentioned, we assumed that efficiency of DGs are equal to 100 %.

4.2.1. Constraints

In second stage optimization, implementation of some constraints is necessary. These constraints are modelled as follow.

$$P_{GAS}(s,t) - \sum_{chp=1}^{CHP} P_{CHP}(s,t,CHP) - \quad (15)$$

$$\sum_{chp=1}^{CHP} H_{CHP}(s,t,CHP) - \sum_{bo=1}^{BO} H_{BO}(s,t,BO) = D_n^{GAS}(s,t) \quad \forall s,t$$

$$\sum_{chp=1}^{CHP} H_{CHP}(s,t,CHP) + \sum_{bo=1}^{BO} H_{BO}(s,t,BO) = D_n^H(s,t) \quad \forall s,t \quad (16)$$

$$P_{EC}(s,t) + \sum_{chp=1}^{CHP} P_{CHP}(s,t,CHP) + \sum_{pev=1}^{PEV} P_{PEV}^{dch}(s,t) = \sum_{pev=1}^{PEV} P_{PEV}^{ch}(s,t) + D_n^E(s,t) \quad \forall s,t \quad (17)$$

$$P_{CHP}^{\min} \leq P_{CHP}(s,t,CHP) \leq P_{CHP}^{\max} \quad \forall s,t,CHP \quad (18)$$

$$H_{CHP}^{\min} \leq H_{CHP}(s,t,CHP) \leq H_{CHP}^{\max} \quad \forall s,t,CHP \quad (19)$$

$$H_{BO}^{\min} \leq H_{BO}(s,t,BO) \leq H_{BO}^{\max} \quad \forall s,t,BO \quad (20)$$

The gas energy balance, heat energy balance and electrical energy balance are constrained by Eqs. (15)–(17), respectively. Constraints (18)–(20) are electrical and heat generation by DGs.

5. NUMERICAL SIMULATION AND CASE STUDIES

To validation and confirmation of the proposed approach, numerical simulation based on two case studies are done. The case studies are as follows: 1) Optimization of the proposed approach without DCS and LGS; 2) Optimization of the proposed approach with DCS and LGS.

The 15-node test system as proposed energy grid is depicted in Fig. 2. In Fig. 3, flowchart of the optimization approach is shown. In Fig. 4, the gas price and electrical price at 5 scenarios are simulated by Monte Carlo technique. In order to reduction of the computations time and computational burden; the optimization approach is solved at fourth scenario. The DGs data are given in Table 1 [48]–[51]. It should be mentioned that all DGs are feed by natural gas. The PEV data is provided in Table 2 [52]–[55]. As well, energy demand of the consumers is shown in Fig. 5. The maximum curtailable demand for implementation of the DCS at each node is 25 kWh [5], [56]–[69]. The GAMS software is employed to solving numerical simulation.

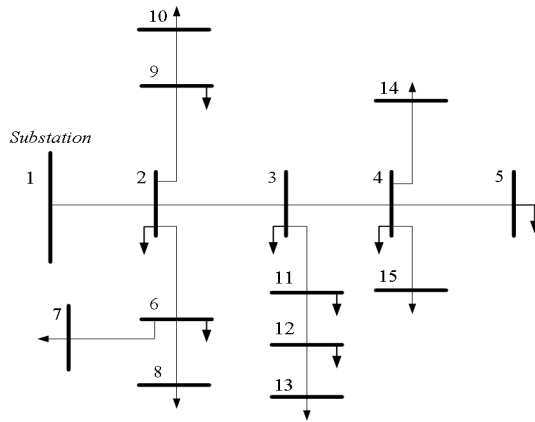


Fig. 2. 15-node test system.

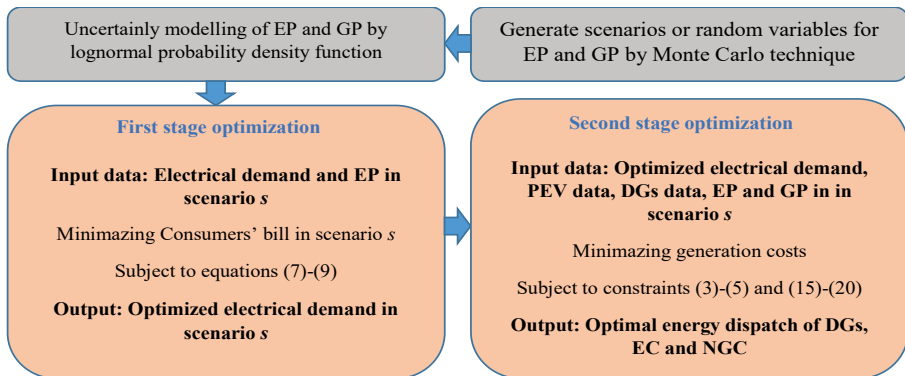


Fig. 3. Flowchart of the proposed approach.

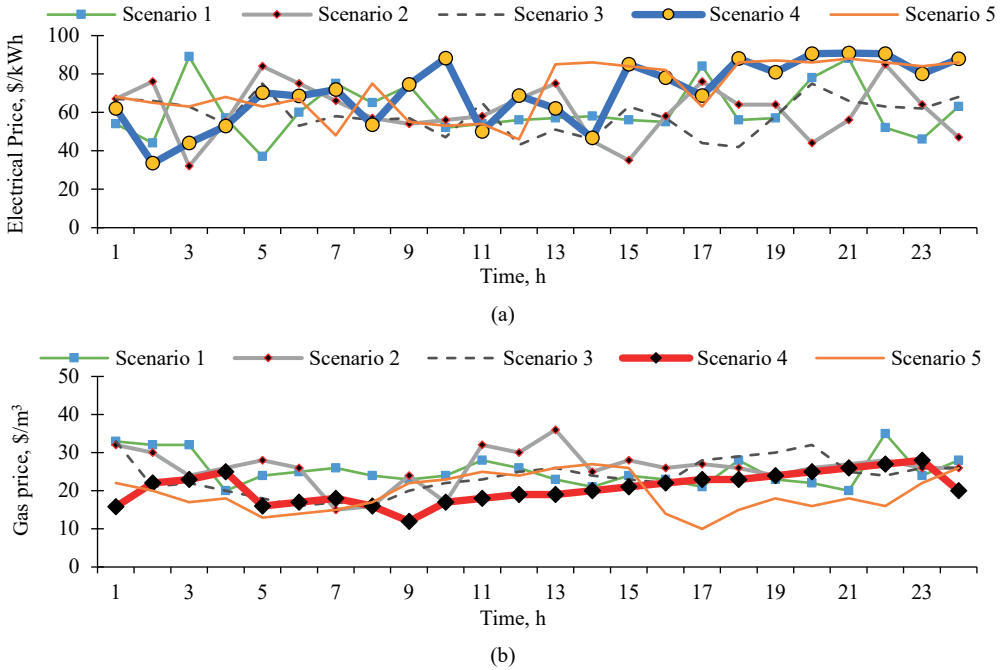


Fig. 4. Energy price in energy market: (a) Electrical price, and (b) Gas price.

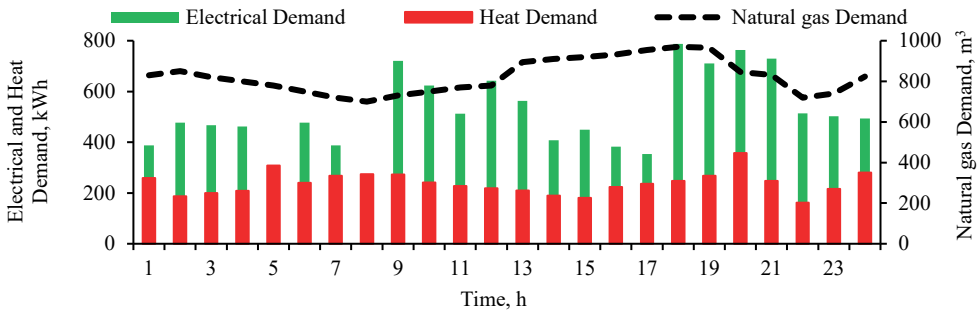


Fig. 5. Energy demand.

TABLE 1. DGs DATA

Parameters Units	p^{\min}	p^{\max}	H^{\min}	H^{\max}	Location, Node
	kWh	kWh	kWh	kWh	
Boiler 1	–	–	0	100	3
Boiler 2	–	–	0	120	8
CHP 1	0	125	0	100	10
CHP 2	0	120	0	110	12

TABLE 2. PEV DATA

Parameters	Value
η^{ch}	90 %
η^{dis}	95 %
P^{r}_{PEV}	50 kWh
Location (node)	6

5.1. Discussion and results analysis

The results analysis of the mentioned case studies are discussed in this subsection. As well, results are compared than each other for showing superiority of the DCS and LGS. As mentioned before, optimization is done in fourth scenario and results are analysed in this scenario.

In Fig. 6, electrical demand in first stage is optimized by DCS. In this figure, optimized electrical demand is curtailed at high EP. The total consumers' bill without DCS and with DCS is equal to \$ 894 182.6 and \$ 790 441.2, respectively. Also, total demand curtailment is 710 kWh. In Fig. 7, electrical generation in Case 1 are depicted.

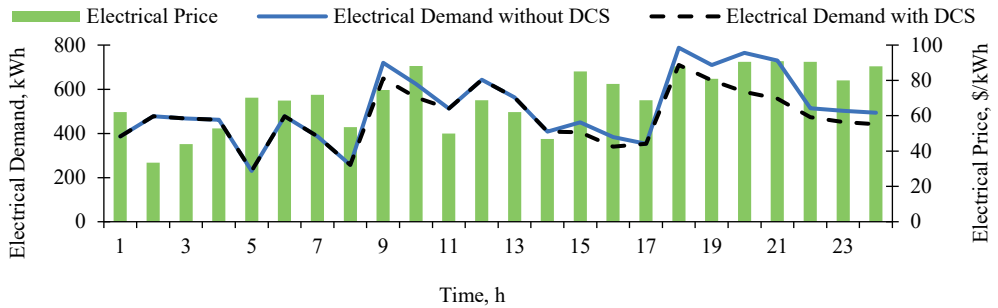


Fig. 6. Electrical demand with DCS and without DCS.

In Fig. 7(a), electrical energy generation without DCS and LGS is shown. As shown, EC has more participation in meet demand at than DGs. The generation cost of the NGC, EC and DGs in Case 1 are equal to \$ 357 795.6, \$ 467 163.4 and \$ 147 220.1, respectively. It's visible, EC in Case 1 has most generation cost in comparison with NGC and DGs. The maximum electrical generation by EC in Case 1 at high EP and peak demand is done. In Fig. 7(b), power generation in Case 2 with implementing LGS and DCS is shown. In Case 2, cost of the EC is reduced by 12.3 % in comparison to Case 1. The power generation of the EC in Fig. 7(b) at peak demand is less than Fig. 7(a). Also, electrical demand is meet at hours 10 and 18 with high EP by PEV. The PEV is feed at low EP, and power of the PEV is used to meet demand at peak. The total discharging power and total charging power of PEV are equal to 98 kWh and 100 kWh in total operation time, respectively.

In Fig. 8, heat generation by CHPs and boilers in Cases 1 and 2 are operated. The heat generation by DGs at all times is done and generation cost to heat generation in both case are almost same.

The results obtained in case studies are listed in Table 3. As shown, generation cost in Cases 1 and 2 are equal to \$ 967 465.2 and \$ 879 323.4, respectively. In Case 2, minimum

generation cost in energy system are provided, due to implementing DCS and LGS. The reduction value of the generation cost in Case 2 than Case 1 is equal to \$ 88 141.8.

TABLE 3. RESULTS OF THE CASE STUDIES

Case study	Case 1	Case 2
Generation cost, \$	972 179.1	879 323.4

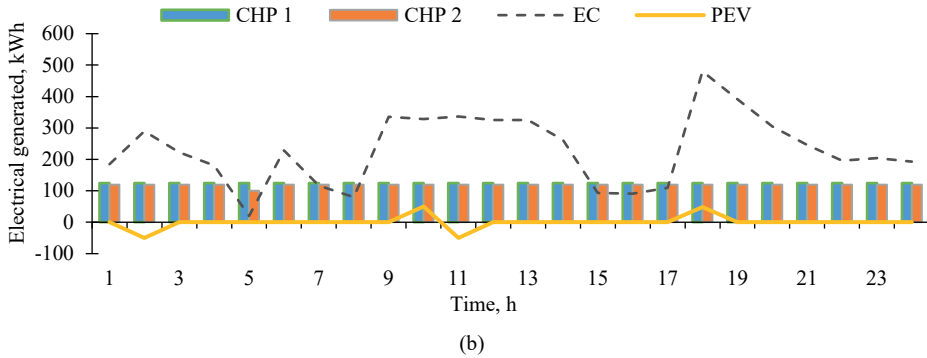
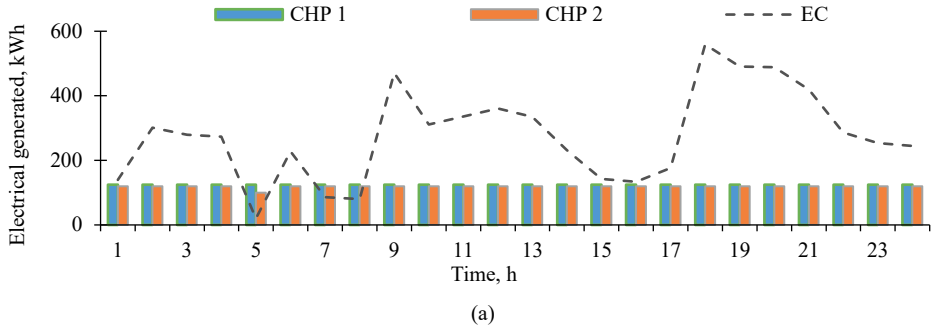
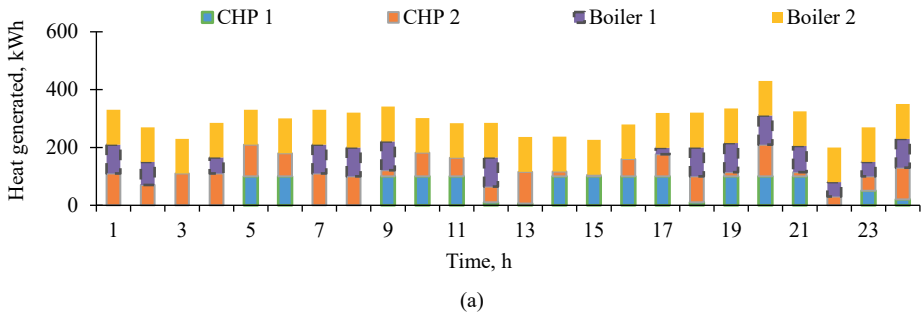


Fig. 7. Electrical generated: (a) Case 1, and (b) Case 2.



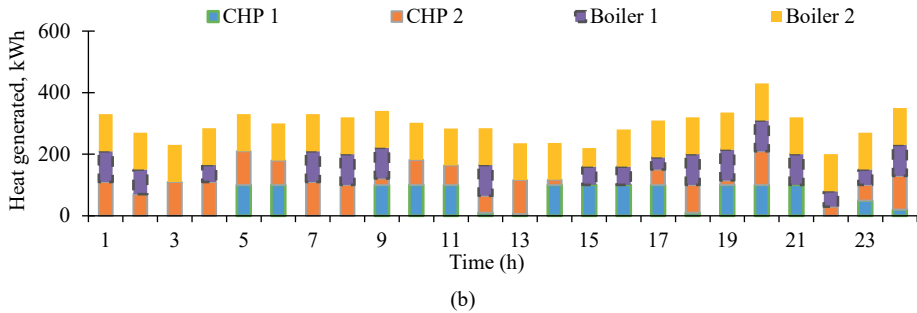


Fig. 8. Heat generated: (a) Case 1, and (b) Case 2.

6. CONCLUSION

In this paper, optimal operation of the multiple energy system is studied based on uncertainty of the EP and GP in energy market. The LGS and DCS are utilized as optimal solution to consumers' participation in energy market. The optimization is modelled as two stage problem in proposed approach. The consumers' bill is minimized in first stage by DCS, whereby energy demand is optimized. Thus, optimized energy demand is taken into account in second stage to minimizing generation cost. The obtained results of the numerical simulation in two case studies are expressed as follow:

Case 1) In this case, LGS and DCS are not taken into account. The generation cost is equal to \$ 972 179.1.

Case 2) The DCS and LGS are implemented in Case 2. The reduction rate of the generation cost in this case than Case 1 is equal to 9.55 %.

With attention to obtained results, participation of the consumers in energy market leads to decrease generation cost, and economic status of the system is provided.

ACKNOWLEDGEMENT

The authors would like to acknowledge Ali Thaeer Hammid in the Bilad Alrafidain University College in Diyala in Iraq, for major contribution in this study.

REFERENCES

- [1] Rashidi Zadeh D. *et al.* An economic and environmental optimization model in a micro grid with demand response. *Environmental and Climate Technologies* 2022;26:730–741. <https://doi.org/10.2478/rtuct-2022-0056>
- [2] Kacare M. *et al.* Impact assesment of the renewable energy policy scenario- a case study of Latvia. *Environmental and Climate Technologies* 2022;26:998–1019. <https://doi.org/10.2478/rtuct-2022-0075>
- [3] Chamandoust H. *et al.* Tri-objective optimal scheduling of smart energy hub system with schedulable loads. *Journal of Cleaner Production* 2019;236:117584. <https://doi.org/10.1016/j.jclepro.2019.07.059>
- [4] Chamandoust H. *et al.* Multi-objective performance of smart hybrid energy system with Multi-optimal participation of customers in day-ahead energy market. *Energy & Buildings* 2020;216:109964. <https://doi.org/10.1016/j.enbuild.2020.109964>
- [5] Chamandoust H. *et al.* Multi-objectives Optimal Scheduling in Smart Energy Hub System with Electrical and Thermal Responsive Loads. *Environmental and Climate Technologies* 2020;24:209–232. <https://doi.org/10.2478/rtuct-2020-0013>
- [6] Chamandoust H. *et al.* Tri-objective scheduling of residential smart electrical distribution grids with optimal joint of responsive loads with renewable energy sources. *Journal of Energy Storage* 2020;27:101112. <https://doi.org/10.1016/j.est.2019.101112>

- [7] Chamandoust H. *et al.* Multi-objective operation of smart stand-alone microgrid with the optimal performance of customers to improve economic and technical indices. *Journal of Energy Storage* 2020;31:101738. <https://doi.org/10.1016/j.est.2020.101738>
- [8] Chamandoust H. *et al.* Energy management of a smart autonomous electrical grid with a hydrogen storage system. *International journal of Hydrogen Energy* 2021;46(34):17608–17626. <https://doi.org/10.1016/j.ijhydene.2021.02.174>
- [9] Chamandoust H. *et al.* Day-ahead scheduling problem of smart micro-grid with high penetration of wind energy and demand side management strategies. *Sustainable Energy Technologies and Assessments* 2020;40:100747. <https://doi.org/10.1016/j.seta.2020.100747>
- [10] Chamandoust H. *et al.* Optimal hybrid participation of customers in a smart micro grid based on day ahead electrical market. *Artificial Intelligence Review* 2022;55:5891–5915. <https://doi.org/10.1007/s10462-022-10154-z>
- [11] Chamandoust H. *et al.* Energy Economic Management of Hybrid Energy System Based on Short-term Generation and Demand Response. *Environmental and Climate Technologies* 2020;24:653–668. <https://doi.org/10.2478/rtuect-2020-0040>
- [12] Chamandoust H., Hashemi A., Derakhshan G., Abdi B. Optimal Hybrid System Design Based on Renewable Energy Resources. *IEEE Smart Grid Conference (SGC)*, Dec 2017. <https://doi.org/10.1109/SGC.2017.8308878>
- [13] Chamandoust H. *et al.* Scheduling of Smart Micro Grid Considering Reserve and Demand Side Management. *IEEE Smart Grid Conference (SGC)*, 2018. <https://doi.org/10.1109/SGC.2018.8777926>
- [14] Osborne G. J., Wood A., Ishak S. M. Knowledge management processes in south australian infrastructure projects: aligning key stakeholders expectations and practices. *The Journal of Modern Project Management* 2022;10(1):126–139.
- [15] Sina M. A., Adeel M. A. Assessment of stand-alone photovoltaic system and mini-grid solar system as solutions to electrification of remote villages in Afghanistan. *International Journal of Innovative Research and Scientific Studies* 2021;4(2):92–99. <https://doi.org/10.53894/ijirss.v4i2.62>
- [16] Dwijendra N. K. A. *et al.* A Multi-Objective Optimization Approach of Smart Autonomous Electrical Grid with Active Consumers and Hydrogen Storage System. *Environmental and Climate Technologies* 2022;26(1):1067–1079. <https://doi.org/10.2478/rtuect-2022-0080>
- [17] Watanakul S., Henry S., Reeveerakul N., Ouzrout Y. A Port Digital Twin Model for Operational Uncertainty Management. *The Journal of Modern Project Management* 2022;9(3).
- [18] Wood A. J. *et al.* Selection and Engagement of Professional Consulting Services: Decision-Making Processes used by Project Management Offices in South Australia. *The Journal of Modern Project Management* 2022;9(3).
- [19] Venkatesh N. *et al.* Design Of Environmental Monitoring System in Farm House Based on Zigbee. *International Journal of Communication and Computer Technologies* 2022;10(2):1–4.
- [20] Richardson T. M., Marion J. W., Anantatmula V. S., Gibson J. R. Insights from the Field: Project Execution Success and Failure. *The Journal of Modern Project Management* 2022;9(2).
- [21] Masrouf A. Global Alliances to Accelerate Innovation at Plug and Play Technology Center. *Journal of Commercial Biotechnology* 2021;26(1):102–103. <https://doi.org/10.5912/jcb976>
- [22] Van Hoa N., *et al.* Impact of Trained Human Resources, Adoption of Technology and International Standards on the Improvement of Accounting and Auditing Activities in the Agricultural Sector in Viet Nam. *AgBioForum* 2022;24(1):59–71.
- [23] Athiyaman A., Magapa T. Market Intelligence From The Internet: An Illustration Using The Biomass Heating Industry. *International Journal Of Economics And Finance Studies* 2019;11(1):1–6. <https://doi.org/10.34109/ijefs.201911101>
- [24] Sibuea M. B., Sibuea S. R., Pratama I. The Impact of Renewable Energy and Economic Development on Environmental Quality of ASEAN Countries. *AgBioForum* 2021;23(1):12–21.
- [25] Jermsittiparsert K. Examining the sustainable energy and carbon emission on the economy: panel evidence from asean. *International Journal of Economics and Finance Studies* 2021;13(1):405–426. <https://doi.org/10.34109/ijefs.202112239>
- [26] Tamoor M., ZakaUllah P., Mobeen, M., Zaka M. A. Solar Powered Automated Irrigation System in Rural Area and their Socio Economic and Environmental Impact. *International Journal of Sustainable Energy and Environmental Research* 2021;10(1):17–28. <https://doi.org/10.18488/journal.13.2021.101.17.28>
- [27] Molajou A., Afshar A., Khosravi M., Soleimanian E., Vahabzadeh M., Varihani H. A. A new paradigm of water, food, and energy nexus. *Environmental Science and Pollution Research* 2021. <https://doi.org/10.1007/s11356-021-13034-1>
- [28] Nourani V., Rouzegari N., Molajou A., Baghanam A. H. An integrated simulation-optimization framework to optimize the reservoir operation adapted to climate change scenarios. *Journal of Hydrology* 2020;587:125018. <https://doi.org/10.1016/j.jhydrol.2020.125018>
- [29] Lozić J. Application of Data Envelopment Analysis (DEA) in Information and Communication Technologies. *Tehnički glasnik*, 2022;16(1):129–134. <https://doi.org/10.31803/tg-20210906103816>
- [30] Wang Y., Teng H. Extension of Intersection Method for Multi-Objective Optimization in Case of Interval Number and its Application. *Tehnički glasnik* 2022;16(1):135–138. <https://doi.org/10.31803/tg-20211006122700>
- [31] Ghazvini, M., Pourkiaei, S. M., & Pourfayaz, F. Thermo-economic assessment and optimization of actual heat engine performance by implementation of NSGA II. *Renewable Energy Research and Applications*, 2020;1(2):235–245. <https://doi.org/10.22044/rera.2020.9677.1034>

- [32] Bariss U., Bazbauers G., Blumberga A., Blumberga D. System Dynamics Modeling of Households' Electricity Consumption and Cost-Income Ratio: A Case Study of Latvia. *Environmental and Climate Technologies* 2017;20(1):36–50. <https://doi.org/10.1515/rtuect-2017-0009>
- [33] Ngakan K. A. D. From Tradition to Modernization in Morphological Process of Indigenous Settlement Patterns in Bali, Indonesia. *International Journal of Advanced Science and Technology* 2020;29(8s):172–184.
- [34] Sifat W. O., et al. Innovative work behaviors in pharmacies of Indonesia: Role of employee voice, generational diversity management and employee engagement. *Systematic Reviews in Pharmacy* 2020;11(2):725–734.
- [35] Dwijendra N. K. A., Akhmadeev R., Tumanov D., Kosov M., Shoar S., Banaitis A. Modeling Social Impacts of High-Rise Residential Buildings during the Post-Occupancy Phase Using DEMATEL Method: A Case Study. *Buildings* 2021;11(11):504. <https://doi.org/10.3390/buildings11110504>
- [36] Ngakan K. A. D., Wiriantari F., Widiyani D. M. S., Yulianasari A. A. A. S. R. Transformation of Catuspatha in Bali Indonesia: Alteration Ideas from Empty Space to Aesthetic Element of City. *Rupkatha Journal on Interdisciplinary Studies in Humanities* 2020;12(6). <https://doi.org/10.21659/rupkatha.v12n6.15>
- [37] Primadewi S. P. N., Sueca N. P., Dwijendra N. K. A., Siwalatri N. K. A. Emerging Architect's Design Method in Designing Tourist Accommodation Case Study: Tourist Accommodation in Ubud, Bali. *Civil Engineering and Architecture* 2021;9(2):271–280. <https://doi.org/10.13189/cea.2021.090201>
- [38] Manakkadu S., Joshi S. P., Halverson T., Dutta S. Top-k User-Based Collaborative Recommendation System Using MapReduce. In *IEEE International Conference on Big Data (Big Data)*. IEEE, 2021. <https://doi.org/10.1109/BigData52589.2021.9671395>
- [39] Casti J. L. From Social Mood to Collective Events: Measuring the Path by Sociometers. *The Beacon: Journal for Studying Ideologies and Mental Dimensions* 2021;4(2):020110153. <https://doi.org/10.55269/thebeacon.4.020110153>
- [40] Babu A. R. V., Dheer D. K., Tagore Y. R., Kumar T. M. S., Shaik S., Rao G. S. A review on the progress of intermetallic solid-state hydrogen storage material for fuel cell vehicles. *European Chemical Bulletin* 2022;11(1):17–29. <https://doi.org/10.31838/ecb/2022.11.01.005>
- [41] Srinivasareddy S., Narayana Y. V., Krishna D. Sector Beam Synthesis in Linear Antenna Arrays using Social Group Optimization Algorithm. *National Journal Of Antennas And Propagation* 2021;3(2):6–9.
- [42] Battula B., Lakshmi P. V., Sri L. S. N., Karpurapu S., Sravya S. D. S. Design a Low Power and High-Speed Parity Checker using Exclusive-or Gates. *Journal Of VLSI Circuits And Systems* 2021;3(2):48–53. <https://doi.org/10.31838/jvcs/03.02.06>
- [43] Saberi K., et al. Optimal performance of CCHP based microgrid considering environmental issue in the presence of real time demand response. *Sustainable Cities and Society* 2019;45:596–606. <https://doi.org/10.1016/j.scs.2018.12.023>
- [44] Elfaki A. O., Abouabdalla O. A., Fong S. L., Johar G. M., Aik K. L., Bachok R. Review and future directions of the automated validation in software product line engineering. *Journal of Theoretical and Applied Information Technology* 2012;42(1):75–93.
- [45] Al-Shuaili S., Ali M., Jaharadak A. A., Al-Sheky M. An Investigate on the Critical Factors that can Affect the Implementation of E-government in Oman. *IEEE 15th International Colloquium on Signal Processing & Its Applications (CSPA)*. IEEE, 2019. <https://doi.org/10.1109/CSPA.2019.8695988>
- [46] Omar H. A. M. B. B., Ali M., Jaharadak A. Green supply chain integrations and corporate sustainability. *Uncertain Supply Chain Management* 2019;7(4):713–726. <https://doi.org/10.5267/j.uscm.2019.3.001>
- [47] Mosbah A., Ali M. A., Aljubari I. H., Sherief S. R. Migrants in the High-Tech and engineering sectors: an emerging research area. *IEEE Conference on Systems, Process and Control (ICSPC)*. IEEE, 2018. <https://doi.org/10.1109/SPC.2018.8704139>
- [48] MGM J., Mohd Shukri Ab Yajid M., Khatibi A. Data Mining Technology and its Applications for Sales Productivity Analysis. *Systematic Review in Pharmacy* 2020;11(1):626–632. <https://doi.org/10.5530/srp.2020.1.79>
- [49] Pambreni Y., Khatibi A., Azam S., Tham J. J. M. S. L. The influence of total quality management toward organization performance. *Management Science Letters* 2019;9:1397–1406. <https://doi.org/10.5267/j.msl.2019.5.011>
- [50] Seyedhoseini S. M., Esfahani M. J., Ghaffari M. A novel hybrid algorithm based on a harmony search and artificial bee colony for solving a portfolio optimization problem using a mean-semi variance approach. *J. Cent. South Univ.* 2016;23:181–188. <https://doi.org/10.1007/s11771-016-3061-9>
- [51] Ahmadi M. H., Baghban A., Sadeghzadeh M., Zamen M., Mosavi A., Shamshirband S., Kumar R., Mohammadi-Khanaposhtani M. Evaluation of electrical efficiency of photovoltaic thermal solar collector. *Engineering Applications of Computational Fluid Mechanics* 2020;14(1):545–565. <https://doi.org/10.1080/19942060.2020.1734094>
- [52] Shamshirband S., Joloudari J. H., GhasemiGol M., Saadatfar H., Mosavi A., Nabipour N. FCS-MBFLEACH: Designing an energy-aware fault detection system for mobile wireless sensor networks. *Mathematics* 2019;8(1):28. <https://doi.org/10.3390/math8010028>
- [53] Torabi M., Hashemi S., Saybani M. R., Shamshirband S., Mosavi A. A Hybrid clustering and classification technique for forecasting short-term energy consumption. *Environmental Progress & Sustainable Energy* 2018;38(1):66–76. <https://doi.org/10.1002/ep.12934>

- [54] Sudari S., Tarofder A., Khatibi A., Tham J. Measuring the critical effect of marketing mix on customer loyalty through customer satisfaction in food and beverage products. *Management Science Letters* 2019;9:1385–1396. <https://doi.org/10.5267/j.msl.2019.5.012>
- [55] Chen C., Sun H., Shen X., Guo Y., Guo Q., Xia T. Two-stage robust planning-operation co-optimization of energy hub considering precise energy storage economic model. *Applied Energy* 2019;252:113372. <https://doi.org/10.1016/j.apenergy.2019.113372>
- [56] Blumberga A., Timma L., Blumberga D. System Dynamic Model for the Accumulation of Renewable Electricity using Power-to-Gas and Power-to-Liquid Concepts. *Environmental and Climate Technologies* 2016;16:54–68. <https://doi.org/10.1515/rtuct-2015-0012>
- [57] Nguyen H. N., Tham J., Khatibi A., Azam S. M. F. Conceptualizing the effects of transfer pricing law on transfer pricing decision making of FDI enterprises in Vietnam. *International Journal of Data and Network Science* 2020;4(2):187–198. <https://doi.org/10.5267/j.ijdns.2020.1.002>
- [58] Nguyen H., Tham J., Khatibi A., Azam S. Enhancing the capacity of tax authorities and its impact on transfer pricing activities of FDI enterprises in Ha Noi, Ho Chi Minh, Dong Nai, and Binh Duong province of Vietnam. *Management Science Letters* 2019;8:1299–1310. <https://doi.org/10.5267/j.msl.2019.4.011>
- [59] Božić D. Applying Simulation Modelling in Quantifying Optimization Results. *Tehnički glasnik* 2021;15(4):518–523. <https://doi.org/10.31803/tg-20210326111551>
- [60] Gurrbach P. The Metaphorical Culturalistic Approach to Technology Assessment. *Tehnički glasnik* 2021;15(4):554–561. <https://doi.org/10.31803/tg-20210628194103>
- [61] Dwijendra N. K. A. Meru as a Hindu Sacred Building Architecture with a High Roof and Resistant to Earthquakes in Bali, Indonesia. *Civil Engineering and Architecture* 2020;8(3):350–358. <https://doi.org/10.13189/cea.2020.080319>
- [62] Dwijendra N. K. A., Jalil A. T., Abed A. M., Bashar B. S., Al-Nussairi A. K. J., Hammid A. T., Shamel A., Uktamov K. F. Improving the transition capability of the low-voltage wind turbine in the sub-synchronous state using a fuzzy controller. *Clean Energy* 2022;4:682–692. <https://doi.org/10.1093/ce/zkac033>
- [63] Dwijendra N. K. A., Vaslavskaya I., Skvortsova N. V., Rakhlis T. P., Rahardja U., Ali M. H. Application of Experimental Design in Optimizing Fuel Station Queuing System. *Industrial Engineering & Management Systems* 2022;21(2):381–389. <https://doi.org/10.7232/iems.2022.21.2.381>
- [64] Dwijendra N. K. A., et al. Economic Performance of a Hybrid Renewable Energy System with Optimal Design of Resources. *Environmental and Climate Technologies* 2022;26(1):441–453. <https://doi.org/10.2478/rtuct-2022-0034>
- [65] Dwijendra N. K. A., et al. “The effect of various irrigation technologies and strategies on water resources management. *Journal of Water and Land Development* 2022;53(IV–VI):143–147. <https://doi.org/10.24425/jwld.2022.140790>
- [66] Ahmed A. A. A., Dwijendra N. K. A., Bynagari N. B., Modenov A. K., Kavitha M. Multi Project Scheduling and Material Planning Using Lagrangian Relaxation Algorithm. *Industrial Engineering & Management Systems* 2021;20(4):580–587. <https://doi.org/10.7232/iems.2021.20.4.580>
- [67] Dwijendra N. K. A., Wiriantari F., Widiyani D. M. S., Yulianasari A. A. S. R., Wijaatmaja A. B. M. Transformation of Catuspatha in Bali Indonesia: Alteration Ideas from Empty Space to Aesthetic Element of City. *Rupkatha Journal on Interdisciplinary Studies in Humanities* 2020;12(6). <https://doi.org/10.21659/rupkatha.v12n6.15>
- [68] Nnamchi S. N., Jagun Z. O., Ijomah M. A., Nnamchi O. A., Busingye J. D. Time Series Analysis of Global Energy Indices: Logarithmic and Normalized Techniques for Developmental Studies. *Energy Economics Letters* 2022;9(1):1–19. <https://doi.org/10.18488/5049.v9i1.4396>
- [69] Ferdous, R., Ahmed M. T. Is Environmental Kuznets Hypothesis Vice-Versa for Bangladesh Especially in the Times of Global Climate Change? – An ARDL Econometric Modeling Approach. *Energy Economics Letters* 2022;9(1):55–66. <https://doi.org/10.55493/5049.v9i1.4571>