Estimation Method of Greenhouse Gas Reduction for Electrical Energy Storage Based on Load-Leveling Application

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Abstract: In recent years, there have been several types of energy storage technologies adopted in many different areas, such as peak shaving, frequency regulation, and renewable stabilization applications. Moreover, technologies of high energy and power density are useful for load leveling, power smoothing for renewable energy systems (RESs), and peak shaving for demand management. Under these circumstances, an estimation technique for assessing environmental issues applied to electrical energy storage (EES) systems is essential in order to promote commercialization of EES systems. Therefore, this paper proposes an estimation method for CO$_2$ emission in cases where EES systems are introduced and not introduced. It is essential to evaluate environmental issues in EES systems at operation stages of their life cycle and make an effective contribution to environmental improvement and reduce potential adverse environmental impacts. Thus, this paper deals with an evaluation method for CO$_2$ emission based on an optimal algorithm including a successive approximation method for the best-mix solution of power sources, etc. From the simulation result based on the proposed evaluation algorithm, it is found that the output power of a coal power plant (high CO$_2$ emission) is replaced by the output powers of the EES systems and the nuclear generator (low CO$_2$ emission).

Keywords: EES systems; greenhouse gas; load leveling; best-mix solution; optimal operation algorithm

1. Introduction

Recently, the operation of electric power systems has become more difficult because the peak load demand is increasing continuously and also the daily and annual load factors are worsening [1,2]. Furthermore, the global environmental issues need to be considered in electric power systems [3–6]. One countermeasure to overcome these problems is a study of the operation method of electric power systems, including novel energy storage systems such as secondary batteries, superconducting magnets (SMES), and flywheels, which have made astonishing improvements lately [7–9].

In general, the cost of power generation can be reduced if the energy storage system is charged during the off-peak time interval and discharged during the peak time interval [10]. In addition, the benefit of storing electricity is increasing along with the increase in the difference in demand between off-peak and peak time intervals. The result is load leveling by time shifting [11]. Furthermore, the output of power generation can be flatter if the difference in demand between daytime and nighttime is reduced using an EES system; as a result, the operation efficiency can be improved and the fuel cost can be reduced [12]. For these reasons, a lot of utilities have built pumped hydro-generators and have started installing large-scaled batteries in substations recently [13].

In contrast, it is expected that an RES such as a PV system or a wind power system will be widely installed and operated in order to overcome global environmental issues [14]. However, operation problems such as output fluctuation or unpredictability may occur
if the RES is integrated with the power grid [15]. When the total volume of renewable energies connected to the grid exceeds a certain level, such problems will appear and countermeasures will be needed. To stabilize the fluctuation and to control load management, an EES system is essential for the introduction of large amounts of renewable energy [16–19].

Under these circumstances, an estimation technique for assessing environmental issues applied to EES systems is essential in order to promote commercialization of EES systems. However, when introducing an EES system, many previous studies have been conducted on their merits in terms of economic dispatch, but studies evaluating the environmental merits are insufficient [20,21]. Therefore, this paper presents a concept of an estimation method for CO\textsubscript{2} emission in cases where EES systems are introduced and not introduced. It is based on the idea that can reduce CO\textsubscript{2} emission by existing generator units by operation of EES systems.

It is essential to evaluate environmental issues in EES systems at operation stages of their life cycle and make an effective contribution to environmental improvement and reduce potential adverse environmental impacts. Thus, this paper proposes an evaluation method of CO\textsubscript{2} emission based on an optimal algorithm including a successive approximation method for the best-mix solution of power sources, etc. From the result of calculating the CO\textsubscript{2} emission using the proposed evaluation algorithm of GHG reduction, it is found that the output power of the coal power plant (high CO\textsubscript{2} emission) is replaced by the output powers of the EES systems and the nuclear generator (low CO\textsubscript{2} emission).

2. Formulation of Load-Leveling Application Using EES Systems

2.1. Concepts of Load Leveling

To estimate GHG reduction for EES systems, the proper charging and discharging amounts of EES systems must be obtained in advance, in other words, the composition ratio of generation units at a peak load demand for the two cases where EES systems are introduced and not introduced. This means that the output power of the oil-power plant (high CO\textsubscript{2} emission) is replaced by the output powers of the EES systems and the nuclear power plant (low CO\textsubscript{2} emission). With the allocation of EES systems to distribution systems in Figure 1, the simultaneous load leveling of both the total power system and the distribution systems increases the utilization rates of less expensive generator units, and the benefit of the reduction in the total power operation cost is expected, as shown in Figure 2. In other words, the operation problem of load leveling is to obtain the most appropriate type and number of generators, called an optimal generation mix, in cases where EES systems are introduced and allocated to distribution systems [22–25].

Figure 1. Power system with EES systems.
2.2. Problem Formulation of Load-Leveling Application

As mentioned above, the fundamental problem of load leveling is to decide the most appropriate type and number of generators, called an optimal generation mix, in cases where EES systems are operated in the power (distribution) systems [26]. The optimal generation mix with EES systems is a static problem against the time period and in which the objective is to determine the process in such a manner as to minimize the total cost for load demands provided for a target year [27]. Both both the generation mix (nonlinear integer programming problem) and the operating mode of EES systems (nonlinear programming problem) must be optimized. The problem can be thus formulated and solved as a nonlinear mixed integer programming problem. However, the nonlinear mixed integer programming problem will be complicated, along with the dimension of the problem. The optimal generation mix problem considering EES systems, whose objective is to determine the generation mix that minimizes the total cost for a target year, can be formulated as a nonlinear mixed integer programming problem as follows [28]:

\[
\text{Min} F_n(x, v) = \sum_{i=1}^{n} [a_i x_i + b_i Q_i(X_{i-1}, X_{i-1} + x_i, v)] + a_s x_s
\]  

Subject to \[
\sum_{i=1}^{n} x_i + x_s \geq P_D + P_R
\]  

\[
x_{i\text{min}} \leq x_i \leq x_{i\text{max}}, \quad i = 1, \ldots, n
\]  

\[
v_{i\text{min}} \leq v_{ik} \leq v_{i\text{max}}, \quad i = 1, \ldots, n, \quad k = 1, \ldots, T
\]  

\[
Q_i(X_{i-1}, X_{i-1} + x_i, v) = \sum_{k=1}^{K} \left[ z_k \int_{X_{i-1}}^{X_{i-1}+X_i} L_k(u, v_{ik}) du \right], \quad i = 1, \ldots, n
\]  

\[
X_i = \sum_{k=1}^{i} x_{ik}, \quad i = 1, \ldots, n, \quad X_0 = 0
\]

where \( F_n \) is the total cost for the target year; \( n \) is the number of generation types; \( a_i, b_i \) are the fixed and variable costs, respectively, of generation type \( i \); \( x_i, x_s \) are the capacity of generation type \( i \) and EES systems, respectively; \( a_s \) is the fixed cost of EES systems; \( v_{ik} \) is the output power of EES systems at a daily load curve \( i \) and time period \( k \); \( Q_i \) is the annual energy production for generation type \( i \); \( X_i \) is the cumulative capacity up to generation type \( i \); \( X_0 \) is the cumulative capacity up to generation type \( 0 \); \( L_k(u) \) is the time fraction that demand is more than load level \( u \) at duration curve \( k \); \( z_k \) is the number of days that provide \( L_k(u) \); \( P_D, P_R \) are the peak demand and spinning

Figure 2. Concept of load duration curve with EES systems.
reserve, respectively; \( K \) is the number of patterns of the daily load duration curve; and \( T \) is the number of time intervals for the daily load duration curve.

### 3. Evaluation Algorithm of GHG Reduction Based on Load Leveling

#### 3.1. Optimal Operation Algorithm for Load Leveling

The problem, as formulated above, which is composed of two kinds of variables, such as generation mix (\( x \)) and operating mode of EES systems (\( v \)), is a nonlinear mixed integer programming problem. From a theoretical perspective, the problem can be solved by evaluating the objective function for the generation mix (\( x \)) under the constraint conditions. However, this method will be complicated with the increase in the system size. Therefore, this paper adapts a successive approximation method considering the parameters of the fixed cost and capacity of EES systems, as shown in Figure 3. The optimization procedure can be illustrated as follows:

<Step1> Assumes system parameters. Put \( K_0 = 0 \) (fixed cost of EES systems) and \( X_0 = 0 \) (initial capacity of EES systems).

<Step2> Determines the optimal generation mix for existing generators (\( x \)) while fixing the output power of EES systems to zero (\( v = 0 \)). Assume \( F_0 \) as the total cost of this solution.

<Step3> Determines the optimal operating mode of EES systems (\( v \)) while fixing the generation mix (\( x \)). Calculate the optimal generation mix with EES systems, \( F_s \).

<Step4> If \( F_s \leq F_0 \), add the unit size of EES systems \( \Delta X \) and go to <Step3>. Otherwise, go to the next step.

<Step5> If the introduction capacity of EES systems is zero (\( X = 0 \)), the algorithm terminates. The generation mix (\( x \)) and the capacity and fixed cost of EES systems are the optimal solution. Otherwise, increase the unit fixed cost of EES systems \( \Delta K \) and go to <Step3>.

![Figure 3. Evaluation algorithm of EES systems.](image-url)
3.2. Operation Algorithm of EES Systems

This section describes the algorithm to determine the optimal operation of an EES system while fixing the generation mix of existing generators. This paper adopts a gradient method to decide the optimal operation mode of the EES system. Therefore, the economic operating conditions for the EES system are obtained by Equation (7).

\[ \eta > \frac{\lambda_{\text{charge}}}{\lambda_{\text{discharge}}} \]  

where \( \eta \) is the round-trip efficiency of the EES system, \( \lambda_{\text{charge}} \) is the incremental cost in the charging period, and \( \lambda_{\text{discharge}} \) shows the incremental cost in the discharging period of the EES system.

The minimization for the objective function \( F_n \), while fixing the generation mix, is obtained by load leveling in order to satisfy the economic operating conditions. The constraint of power (kW) and capacity (kWh) in EES systems must be also satisfied in this procedure. The optimal operation mode of EES systems over the target year is decided, along with all daily duration curves. The procedure is as follows:

**<Step1>** Decide the lowest and highest load demand periods in the daily load duration curve. Compute the incremental costs \( \lambda_{\text{charge}} \) and \( \lambda_{\text{discharge}} \).

**<Step2>** If \( \eta < \frac{\lambda_{\text{charge}}}{\lambda_{\text{discharge}}} \), the algorithm terminates. Otherwise, charge a small amount of power \( \Delta P_S \) in the lowest period, and discharge the power \( \eta \Delta P_S \) in the highest period.

**<Step3>** If the maximum storage capacity (kWh) constraint is reached, the algorithm terminates.

**<Step4>** If the maximum output (kW) constraint for EES systems is reached, eliminate the period from consideration. Go to **<Step1>**.

In addition, the allocation site of each small amount for EES systems is decided by selecting the lowest \( F_n(x) \) for the cases where the above algorithm is applied at all allocation sites that are the distribution substations.

3.3. Estimation Algorithm of GHG Reduction Based on Load Leveling

With the allocation of EES systems to distribution systems, the benefit of the reduction of the total power operation cost is expected because the simultaneous load leveling of both the total power system and the power distribution systems increases the utilization rates of less expensive generator units such as nuclear and coal power plants. In addition, if EES systems are replaced with existing generators in the peak (discharging) time interval, the benefit of CO\(_2\) emission reduction can be expected, where it is ideally assumed that EES systems are charged by the nuclear power unit in the off-peak time and discharged by oil or gas generator units in the peak time. Therefore, the estimation of GHG reduction during a year can be quantified using the formula given below:

\[
GHG(y) = \sum_{i=1}^{n} \sum_{t=1}^{T} P_{DE,i}(t) \times \alpha_i - \sum_{i=1}^{n} \sum_{t=1}^{T} P_{CE,i}(t) \times \beta_i - \sum_{t=1}^{T} P_{\text{ESS}}(t) \times \eta_{\text{in, out}} \times \gamma_{\text{ESS}} \]  

Subj. to \( \sum_{i=1}^{n} P_{DE,i} - P_{\text{ESS}} = 0 \)  

where \( GHG(y) \) is the GHG reduction amount a year (kt), \( P_{DE,i}(t) \) is the output (kW) of the existing generators in peak (discharging) times, \( P_{CE,i}(t) \) is the output (kW) of the existing generators in off-peak (charging) times, \( P_{\text{ESS}}(t) \) is the charging and discharging output (kW) of the EES systems, \( \alpha_i \) is the CO\(_2\) emission coefficient of the generator units in the discharging time, \( \beta_i \) is the CO\(_2\) emission coefficient of the generator units in the charging time, \( \gamma_{\text{ESS}} \) is the CO\(_2\) emission coefficient of the EES systems, \( i \) is the generator type, \( t \) is
the time interval, \( T \) is the target year, and \( \eta_{\text{in, out}} \) is the charging and discharging efficiency (loss) of the EES systems.

To find out \( GHG(y) \), the proper charging and discharging amounts of EES systems must be obtained in advance, in other words, the composition ratio of generation units at a peak load demand for the two cases where EES systems are introduced and not introduced. This means that the output power of the oil power plant (high \( \text{CO}_2 \) emission) is replaced by the output powers of the EES systems and the nuclear power plant (low \( \text{CO}_2 \) emission).

4. Case Studies
4.1. Simulation Conditions

To validate the proposed method, this paper carried out simulations using the model systems and parameters shown in Figure 4 and Table 1. The table is the data of the statistical materials of the Korea Electric Power Cooperation in the fiscal year of 2018. The four load patterns for distribution substations (A, B, C, D) and the peak demand of 10 million kW in Figure 5 were considered. This figure is the typical load pattern in summer, and the load patterns of other seasons were assumed by the same pattern and the size of 70%, 80%, and 90% based on the typical load pattern. In addition, the round-trip efficiencies for EES systems of 70% and 80% were assumed. Furthermore, the \( \text{CO}_2 \) emission amount of the generator type was assumed, as shown in Table 2, in order to find out GHG reduction based on load leveling in EES systems. This paper found out the optimal generation mix, considering the operation of EES systems under the following assumptions [29–32]:

1. The total cost of generators is calculated by the sum of the variable and fixed costs, and the total cost of EES systems is only the fixed cost.
2. The maintenance cost of generators is ignored.
3. Unit sizes for the existing generators are previously provided, and unit sizes for new generators are not fixed.

| Type      | Variable Cost (won/kWh) | Fixed Cost (1000 won/kW) | Rating (MW) | Failure Rate (%) |
|-----------|------------------------|--------------------------|-------------|------------------|
| Nuclear   | 39.7                   | 2385                     | -           | 6.5              |
| Coal      | 60.9                   | 1399                     | 1000        | 7.0              |
| LNG       | 147.2                  | 576                      | 1000        | 6.0              |
| Oil       | 184.7                  | 576                      | -           | 6.0              |
| EES systems | -                  | Ca                      | 20 (8 h)    | -                |

Figure 4. Model power systems.
Table 2. CO₂ emission amount of generator type [33,34]. (Reprint with permission [33,34]; September 2008 and 14 June 2012, ISO and IEC).

| Type of Generator                        | Amount of CO₂ Emission (kT/TWh) |
|-----------------------------------------|---------------------------------|
| Run-of-the-river hydropower             | 1                               |
| Wind (without back-up production)       | 9                               |
| Solar photovoltaic                      | 13                              |
| Hydropower with a reservoir             | 15                              |
| Nuclear                                 | 15                              |
| Biomass (plantation)                    | 120                             |
| Biomass (forestry waste)                | 120                             |
| Natural gas combined cycle              | 510                             |
| Fuel cell (H₂ from CH₄ reforming)       | 665                             |
| Heavy oil steam boiler                  | 780                             |
| Diesel                                  | 780                             |
| Coal steam plant with SO₂ removal       | 975                             |

Figure 5. Yearly load patterns of distribution substations.

4.2. Operation Characteristics of Load Leveling

By comparing the total operation cost $F_0(x)$ and $F_0(x)$ for the two cases where EES systems are introduced and not introduced, with the increase in a small unit of the capacity and fixed cost of EES systems, the optimal capacity and fixed cost of EES systems are obtained, as shown in Figure 6. Because of the computation time for parameter analysis, a small unit of the capacity of EES systems is considered as 20 MW (160 MWh, 8 h) and the fixed cost of EES systems is considered as 1000 won. Figure 6 shows that the benefits of the load leveling of EES systems in the distribution substations, which is the fixed cost ($C_a$), becomes 75,000~94,000 won/kW.

As shown in Figure 6, the marginal and saturated fixed costs are also obtained. The marginal cost, in which the composition ratio of EES systems is zero, represents the economical point for EES systems. In addition, the saturated fixed cost keeps a constant value, although the fixed cost changes, because complete load leveling is accomplished at each fixed cost. Table 3 shows the comparison results for the composition ratios of generation units and the total cost at a fixed cost of 75,000 won for the two cases where EES systems are introduced and not introduced.
Figure 6. Optimal capacity and fixed cost of EES systems (1 unit: 20 MW, 180 MWh).

Table 3. Composition ratio of generation units (at a fixed cost of 75,000 won).

| Type            | Output Power without EES Systems (MW) | Output Power with EES Systems (MW) |
|-----------------|---------------------------------------|-----------------------------------|
| Nuclear         | 2899.3                                | 3766.3                            |
| Coal            | 1000.0                                | 1000.0                            |
| LNG             | 1000.0                                | 1000.0                            |
| Oil             | 5100.0                                | 2399.1                            |
| EES systems     | 0.0                                   | 2600.0                            |
| Total cost      | 1920.1 million won                    | 1914.1 million won                |

4.3. Estimation of GHG Reduction Based on Load Leveling

Based on the operation characteristics of EES systems, as shown in Section 4.2, this paper ideally assumes that EES systems are charged by the nuclear power unit in off-peak time and discharged by the oil power plant (heavy oil steam boiler) units in peak time. Here, the total capacity of EES systems is calculated as 2600 MW (8 h, 20,800 MWh) and the efficiency of EES systems is also assumed as 80% for load leveling. Therefore, the amount of GHG reduction during a year can be obtained as 5785.14 kt based on the following procedure:

- \( P_{DE}(t)/\text{year} = 365 \text{ days} \times 20,800 \text{ MWh} \times 780 \text{ kt/MWh} \times 10^{-6} = 5921.8 \text{ kt} \)
- \( P_{CE}(t)/\text{year} = 365 \text{ days} \times 20,800 \text{ MWh} \times 15 \text{ kt/MWh} \times 10^{-6} = 113.88 \text{ kt} \)
- \( P_{ESS}(t) \eta_{in,out}/\text{year} = 365 \text{ days} \times 20,800 \text{ MWh} \times (1 - 0.8) \times 15 \text{ kt/MWh} \times 10^{-6} = 22.78 \text{ kt} \)
- \( \text{GHG}(y) = 5785.14 \text{ kt} \)

However, if the coal power plant is replaced with an oil power plant (heavy oil steam boiler) for discharging EES systems during the peak-time interval, the annual amount of GHG reduction can be calculated as 7265.54 kt based on the following procedure:

- \( P_{DE}(t)/\text{year} = 365 \text{ days} \times 20,800 \text{ MWh} \times 975 \text{ kt/MWh} \times 10^{-6} = 7402.2 \text{ kt} \)
- \( P_{CE}(t)/\text{year} = 365 \text{ days} \times 20,800 \text{ MWh} \times 15 \text{ kt/MWh} \times 10^{-6} = 113.88 \text{ kt} \)
- \( P_{ESS}(t) \eta_{in,out}/\text{year} = 365 \text{ days} \times 20,800 \text{ MWh} \times (1 - 0.8) \times 15 \text{ kt/MWh} \times 10^{-6} = 22.78 \text{ kt} \)
- \( \text{GHG}(y) = 7265.54 \text{ kt} \)
Therefore, it is clear that the amount of GHG reduction with a coal power plant is more effective than with an oil power plant for charging the power of EES systems.

5. Conclusions

This paper presents a concept of an estimation method for CO$_2$ emission based on an optimal algorithm including a successive approximation method for the best-mix solution of power sources, etc. The main results are summarized as follows.

1. From the optimal operation algorithm of EES systems, the benefits of load leveling in the distribution substations is calculated as 75,000–94,000 won/kW for 1 year.
2. The total amount of GHG reduction with an oil power plant is calculated as 5785.14 kt, and that with a coal power plant is obtained as 7265.54 kt. The amount of GHG reduction with a coal power plant is more effective than with an oil power plant for charging the power of EES systems.
3. It is confirmed that the output of the coal power plant with high CO$_2$ emission is replaced by the EES systems and the nuclear power plant with low CO$_2$ emission. Therefore, it is confirmed that EES systems affect the environment at operation stages of their life cycle and contribute to environmental improvement and reduction in potential adverse environmental impacts.

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