



Article

Flexibility-Based Evaluation of Variable Generation Acceptability in Korean Power System

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Abstract: This study proposes an evaluation method for variable generation (VG) acceptability with an adequate level of power system flexibility. In this method, a risk index referred to as the ramping capability shortage expectation (*RSE*) is used to quantify flexibility. The *RSE* value of the current power system is selected as the adequate level of flexibility (i.e., *RSE* criterion). VG acceptability is represented by the VG penetration level for the *RSE* criterion. The proposed evaluation method was applied to the generation expansion plan in Korea for 2029 in order to examine the validity of the existing plan for VG penetration. Sensitivity analysis was also performed to analyze the effects of changes in system uncertainty on VG acceptability. The results show that the planned VG penetration level for 2029 can improve by approximately 12% while securing flexibility.

Keywords: flexibility; ramping capability shortage expectation; sensitivity analysis; variable generation (VG) acceptability; system uncertainty

1. Introduction

For the next few decades, the national renewable energy plan of Korea has been outlined in the second basic energy plan (every five years), fourth new and renewable energy basic plan (every five years), and seventh basic plan for long-term electricity supply and demand [1–3]. These plans indicate that integrating renewable energy is essential to improve energy security and cope with the post-2020 climate change regime. The expansion of photovoltaic (PV) and wind power systems is considered as a core strategy to implement these plans. Their installed capacity is expected to account for approximately 75% of total installed capacity in 2029 [3]. The annual average growth of PV and wind power systems between 2017 and 2029 will be 1012 MW and 528 MW, respectively; these figures correspond to the largest and second largest increase in renewable energy resources.

To cope with the challenges of supplying sustainable energy, renewable energy implementation plans in Korea emphasize variable generation (VG) through PV and wind power systems, although other renewable resources will be included. In the process of introducing VG into the power grid, many studies have demonstrated the lack of flexibility in the operation and planning of the system because unexpected variations in VG would lead to a power mismatch [4–8]. Flexibility is generally defined as the ability to respond to changes in net load (i.e., load minus VG). Power systems may not support flexibility because of uncertainties such as failure of power plants (FOPP) and load forecast errors [8]. Increasing VG has made it harder to secure flexibility because of increased system uncertainty. Accordingly, the flexibility issue has recently surfaced in the operation and planning of power systems.

The flexibility issue is not limited to Korea. Most system operators and planners have encountered this problem and have made efforts to secure flexibility [9,10]. Hence, the conventional generator has received attention because it can provide ramping capability, which can be considered as the most

effective means to respond to variations in net load [11]. The greater the ramping capability of the power system, the larger the grid acceptability for VG. However, as mentioned previously, system uncertainty is an inevitable problem. Therefore, controlling the available ramping capability is of vital importance to integrate more VG output, while maintaining flexibility.

Studies have been conducted to assess VG acceptability based on power system flexibility, which can be categorized as deterministic and probabilistic approaches [12–15]. Ramping capability was considered in the former approach, but related uncertainties in the system were neglected. The load shedding experience of the Electric Reliability Council of Texas (Austin, TX, USA) is a prime example of the necessity of reflecting system uncertainty [16], which was determined in [14,15]. The grid-acceptable wind power capacity was calculated using the unit commitment (UC) and economic dispatch model [14]. A chronological method was used to generate the scenario. However, the FOPP was neglected, which influences reliability. The wind curtailment ratio was also used as a risk criterion. However, wind curtailment risk is not as serious as load curtailment risk, whose effects on reliability are more severe. In [15], the wind power acceptability in China was computed. Various risks were considered through a systematic approach, particularly transmission and storage capacity risks, but FOPP was also neglected.

This study proposes a flexibility-based evaluation method for VG acceptability. In this method, major system uncertainties, such as FOPP and VG forecast error, are considered in the flexibility calculation. Flexibility is quantified using a risk index referred to as the ramping capability shortage expectation (*RSE*), which represents the possibility of a ramping capability shortage due to major system uncertainties in a particular period [8]. The *RSE* is used as a criterion in the evaluation of VG acceptability. Korea's VG penetration plan for 2029 does not provide a reasonable basis for the required flexibility [3]. Thus, we examine the validity of the existing plan for VG penetration and evaluate VG acceptability with an adequate level of flexibility using the proposed evaluation method. Moreover, sensitivity analysis is performed to analyze the effects of system uncertainties on flexibility.

The rest of this paper is organized as follows. Section 2 explains the flexibility index *RSE*, which is used as a criterion to evaluate the VG acceptability. In Section 3, the evaluation method for VG acceptability is described. In Section 4, VG acceptability for the generation expansion plan in Korea is evaluated. Sensitivity analysis is performed to confirm the effects of changes in uncertainty parameters on VG acceptability. The conclusions and future prospects are discussed in Section 5.

2. Flexibility Index: Ramping Capability Shortage Expectation

When considering power balance, the concept of flexibility can be incorporated in power system reliability, which addresses the risks faced by power systems. The risk considered is simply a power mismatch due to system uncertainties, such as FOPP and net load forecast error (*NLFE*), which are sources of risk. In other words, a causal relationship exists between risk and system uncertainty. Increasing the VG of PV and wind power systems can increase risk. Studies conducted to evaluate risk have failed to demonstrate this causal relationship [17–19]. The *RSE* index in our study was proposed to overcome this limitation and explicitly evaluate power system flexibility [8].

2.1. RSE Definition

The ramping capability of a generator is defined as the ability to change its output during a specific period. The system ramping capability (*SRC*) can be computed by adding the ramping capabilities of every generator as follows:

$$SRC_t = \sum_{i \in I} A_{i,t-\Delta t} O_{i,t-\Delta t} \min(P_{max,i} - P_{i,t-\Delta t}, rr_i \Delta t)$$
(1)

In the *SRC*, the uncertainty of the generator is taken into account using availability $A_{i,t-\Delta t}$, which is calculated using the Markov chain-based capacity state model [8]. This value varies with time $O_{i,t-\Delta t}$ and indicates whether generator i is online at time $t-\Delta t$. If the generator is online, the value is one;

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otherwise, it is zero. $P_{max,i}$ and $P_{i,t-\Delta t}$ are the maximum output and scheduled output of generator i, respectively. rr_i is the ramp rate of generator i. Δt is the minimum time interval considered in the ramping capability calculation. The SRC should then be compared with the RC requirement (RCR) in the net load side, which is given as follows:

$$RCR_t = NLFE_t + FNL_t - \sum_{i \in I} A_{i,t-\Delta t} O_{i,t-\Delta t} P_{i,t-\Delta t}$$
(2)

where $NLFE_t$ and FNL_t are the net load forecast error and forecast net load at time t, respectively. $A_{i,t-\Delta t}$ is included in this equation, indicating that RCR_t is affected by FOPP. If the loading generator unexpectedly fails, additional ramping capability is required, which is reasonable.

If RCR_t is not covered by SRC_t , then the power is mismatched. This situation can be called a ramping capability shortage if load shedding is required. Therefore, the risk index RSE is defined as the sum of the probabilities, wherein the RCR will not be satisfied by the SRC for the entire period, and is given as follows:

$$RSE = \sum_{t} RSP_{t} = \sum_{t} \sum_{e \in E_{t-\Delta t}} Prob(e) \left[\sum_{c \in C_{t-\Delta t}} Prob_{c}[RCR_{t} > SRC_{t}] \right]$$
(3)

such that:

$$RSE = \sum_{t} \sum_{e \in E_{t-\Delta t}} Prob(e) \left[\sum_{c \in C_{t-\Delta t}} Prob_{c}[FNL_{t} + NLFE_{t}] \right]$$

$$> \sum_{i \in I} A_{i,t-\Delta t} O_{i,t-\Delta t} \{ P_{i,t-\Delta t} + \min(P_{max,i} - P_{i,t-\Delta t}, rr_{i}\Delta t) \}$$
(4)

where $C_{i,t-\Delta t}$ and $E_{i,t-\Delta t}$ are the possible uncertainty scenarios set for the FOPP and *NLFE* at time $t-\Delta t$, respectively. The *RSE* has the following characteristics: (1) only the up-RC shortage risk (related to increasing net load) is considered so that more critical problems can be incorporated in terms of reliability; and (2) the calculation of the *RSE* is applied to the worst case of all possible FOPP scenarios. The second characteristic intends to reduce the computational burden in generating scenarios. For more details, refer to [8].

2.2. Comparison of RSE with Reliability Indices

The *RSE*-based flexibility evaluation is different from conventional evaluations based on reliability indices, such as loss of load probability and loss of load expectation [20,21]. First, the reliability indices-based evaluation is combined with the solution process of the stochastic UC problem. The indices are included in the objective function of the problem; thus, an independent evaluation is impossible. However, calculation of the *RSE* is independent from the solution process of the UC problem. Second, in the reliability indices-based evaluation, generating uncertainty scenarios such as FOPP and *NLFE*, as well as establishing simulation procedures are time-consuming. It is impossible to consider all the possible scenarios for system uncertainty. In most related studies, reduction techniques were used to generate uncertainty scenarios. However, scenarios for system severity were not generated. System severity is defined as the extent to which the system is in danger of losing the power balance. Severity should be considered in the flexibility study because the nature of the risks observed in flexibility and reliability studies are the same. The *RSE* can take into account severity by selecting the worst case for the FOPP scenario, thereby reducing the computational burden that occurs as the size of the system increases.

3. Evaluation Procedure of the Variable Generation Acceptability

VG acceptability was evaluated to determine the maximum acceptable level of VG penetration with an adequate level of power system flexibility (i.e., RSE criterion). This evaluation procedure is based on the relationship between VG penetration level and flexibility. The RSE helps quantify

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the effects of VG penetration level on flexibility. As shown in Equation (4), the terms FNL_t , $NLFE_t$, $P_{i,t-\Delta t}$, Prob(e), and $Prob_c$ are associated with VG penetration level and their changes directly affect the RSE. The RSE may decrease as VG penetration level increases, although the extent to which it decreases depends on the characteristics of the system. If a power system has many fast-ramping units, then the RSE is more likely to decrease gradually compared to a system with many slow-ramping units. However, system uncertainties, such as the failure rate of generators, can affect the RSE, thereby decreasing its value more abruptly.

After obtaining the relationship between the penetration level and *RSE*, the VG acceptability under the *RSE* criterion can be easily determined. In this study, the *RSE* value of the current power system was used as the *RSE* criterion in order to maintain the same flexibility level as that of the current system. This approach is reliable because the flexibility of the current system is empirically verified. *RSE* values above the *RSE* criterion and their corresponding VG penetration levels are feasible. The maximum value among the feasible VG penetration levels indicates the VG acceptability. Sensitivity analysis of the uncertainty parameters is then performed at this value to examine the effects of uncertainty parameters on VG acceptability.

Figure 1 shows the flexibility-based evaluation procedure. The basic information for calculating the *RSE* includes the forecasted net load profile for the worst case scenario, VG-related information, load distribution information, VG forecast error, and the technical information of generators. The forecasted net load profile can be obtained by combining the forecasted load profile and VG-related information. The distribution information of the load and VG forecast errors are factored into the uncertainty scenarios for *NLFE* (i.e., $E_{i,t-\Delta t}$ in Equation (4)). The technical information of the generators, such as failure rate, repair rate, and maximum generation output, are used to calculate the generator states and their probabilities, which are then used to obtain uncertainty scenarios for FOPP (i.e., $C_{i,t-\Delta t}$ in Equation (4)).

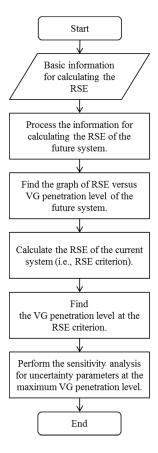


Figure 1. Flexibility-based evaluation process of variable generation (VG) acceptability.

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4. Case Study

The main objective of our simulation is not only to evaluate the maximum acceptable level of VG penetration in Korea for 2029 but also to examine the influence of uncertainty parameters on power system flexibility. The year of interest is 2029, which is the last forecasted year in the latest generation expansion plan [3]. In recent years, the annual peak load tends to occur in the winter season (i.e., December to February) in Korea. This is attributed to the use of equipment with temperature-sensitive loads, such as electric heaters. In addition, 16 December 2016 was chosen as one of the peak days [22]. Figure 2 shows the load profiles of the peak day; the peak load was 75,500 MW at around 10:00 a.m.

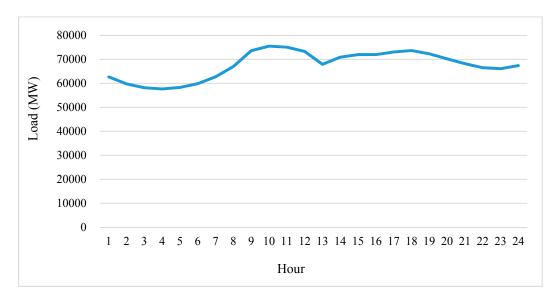


Figure 2. Load profiles of peak day in 2016.

The generators in Korea can be classified into dispatchable and non-dispatchable units. The power system operator has no control over non-dispatchable units, which include not only VG systems but also non-VG systems such as units with power below 20 MW, units owned by community energy service providers [23], and new-energy units (i.e., units using energy converted from fossil fuels or obtained through the chemical reaction between oxygen and hydrogen—for example, fuel cells, integrated gasification combined cycle, hydrogen energy, etc.). We assumed that on the peak day in 2029, the output patterns of VG and non-VG systems are the same as that in 2016. However, to represent the changes in VG penetration level, the proportions of the PV and wind power systems were adjusted to their expected generation capacity in 2029. This assumption is based on the fact that the number of PV and wind power systems will proportionally increase from their current installation rates by region, which may be significant because of the smoothing effect. This means that fluctuations in total VG outputs are smoothened by the correlations between the VG outputs of each region [24]. Figure 3 shows the profiles of VG and non-VG outputs. The increase in VG output during hours 8–13 is because of PV power. With exception of PV power, most of the power of VG comes from wind systems. However, the output is flattened by the smoothing effect. Figure 4 shows the net load profiles. By employing an upscaling technique, the net load profile of the peak day in 2029 can be represented, with a peak load of 101,154 MW that occurs during hour 18.

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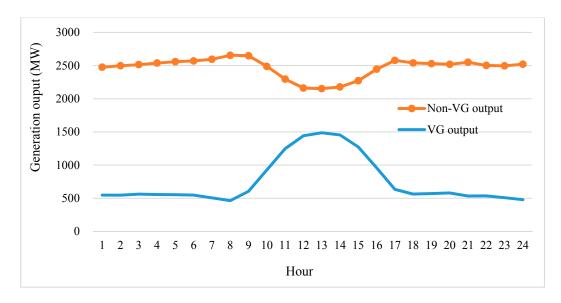


Figure 3. VG/Non-VG output profiles of peak day in 2016.

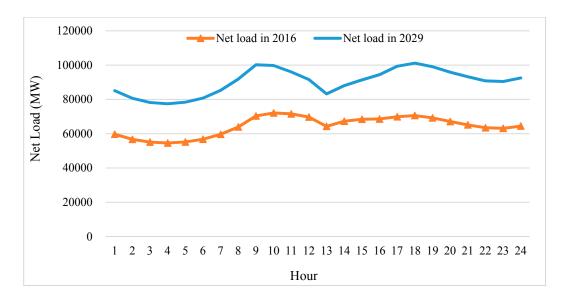


Figure 4. Net load profiles of peak day.

On the peak day in 2016, the number of available generators was 202, excluding non-dispatchable units, units scheduled for maintenance, and failed units; the available capacity was 87,183 MW. The dispatchable units can be categorized into several types: combined-cycle (using both gas and steam turbines) fossil-fuel units, steam turbine fossil-fuel units, cogeneration units, hydraulic/pumped hydraulic units, and nuclear power units. In 2016, the ratios of non-available units to total units, with respect to the aforementioned types, were 1.3%, 10%, 4.8%, 45.5% and 21.7%, respectively. The same ratios were applied to each type of generator in 2029. Accordingly, the number of available dispatchable units is 296, with available capacity of 116,712 MW (as of 16 December 2029). Information on the type and size of generators in 2029 is listed in [3]. Technical information on newly installed generators from 2017 to 2029 is assumed to be the same as that of the latest generators (see Appendix A for the details of newly installed generators). Information on failure and repair rates can be seen in [25]. The *NLFE* is assumed to have a normal distribution with a standard deviation of 5%.

The generation schedule was determined using M-CoreS (version: 2.10.161101, Master's Space, Seoul, Korea), a commercial software designed to simulate the short-term electricity market in

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Korea [26]. The Korean electricity market currently operates under the cost-based pool market rule; the latest market rule was applied for this simulation. The detailed market rule is given in [27]. A MATLAB program (R2013b version) (MathWorks, Natick, MA, USA) was used to compute the *RSE* for the resulting power generation schedule. Using a PC with a 3.7 GHz Intel Core i3-6100 CPU (Santa Clara, CA, USA) and 16 GB RAM, the computational time for the market simulation and *RSE* calculation were 5 and 13 min, respectively.

Figure 5 shows the generation schedule for the power system in 2029; "Hyd/pumped" is an acronym of "hydraulic/pumped hydraulic unit". For reference, the results were obtained for the net load in 2029. The combined-cycle units followed the hourly variations in net load, and the hydraulic/pumped hydraulic units offset the sharp increase in net load. The negative values of the Hyd/pumped unit appeared during hours 4 and 13 because of the action of the pumped hydraulic units; these values, however, were too small to be seen in the graph.

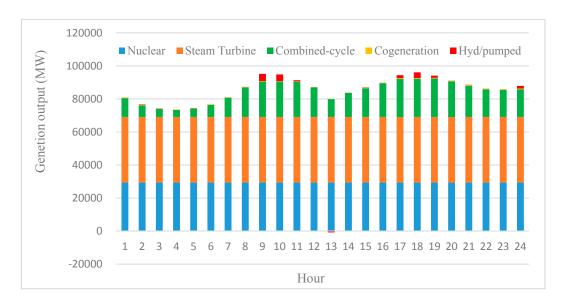


Figure 5. Generation schedule on the peak day in 2029.

Figure 6 shows the graph of *RSE* versus VG penetration level, with a step size of 1%, where the *RSE* criterion (i.e., 10.6720 h/day) is indicated with a blue dotted line. The *RSE* values (i.e., risk level) decreased in some sections as VG penetration level increased, although the sum of *RCR* during the entire period increased (that is, more RC was required) with the increase in VG penetration level, as shown in Figure 7. This may be because the system ramping capability is determined by the generation schedule. The market rule for the generation schedule requires the system operator to secure operating reserves. These are classified into three types: frequency regulation reserves (for governor free and automatic generation control services), standby reserves, and replacement reserves. The second and third type of reserves can either be spinning or non-spinning. The requirement for frequency regulation reserves is at least 1500 MW, while the minimum requirements for spinning and non-spinning reserves are 1500 MW and 1000 MW, respectively. The *SRC* can vary with this constraint and net load situations. For all VG penetration levels, the highest ramping up in net load occurred at hour 8. When comparing generation schedules based on VG penetration levels at hour 8, the result with larger *RSE* had greater *SRC* (i.e., more operating reserves).

The planned VG penetration level in 2029 is 22%, and the corresponding RSE is $10.4513 \, h/day$, which is less than the RSE criterion. The RSE criterion intersects the RSE graph at two points, and the VG penetration levels at each point are 19.2% and 35.8%, respectively. The feasible VG penetration levels are between these two points. The larger one (i.e., 35.8%) corresponds to the maximum acceptable level of VG penetration. It should be noted that this value is not only 12% larger than the planned VG

penetration level (i.e., 22%) but is also helpful in maintaining the same flexibility as that of the current system (as of 16 December 2016). According to the generation schedule, some *RSE* values may fall below the *RSE* criterion in some sections after a VG penetration of 37%. However, those values would be unstable because some variations in the values can lead to a violation of the *RSE* criterion.

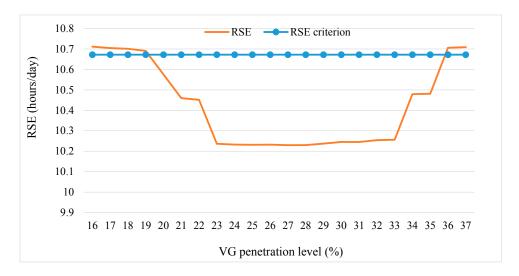


Figure 6. RSE versus VG penetration level.

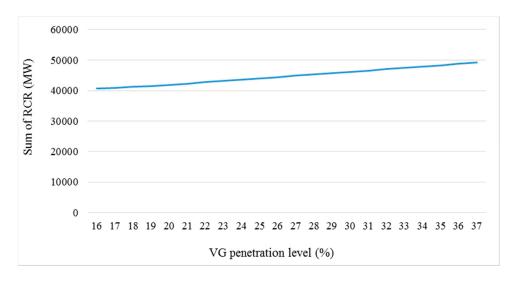


Figure 7. Sum of RC requirement (*RCR*) versus VG penetration level.

In order to examine the impact of changes in system uncertainty on VG penetration level, sensitivity analysis was performed with the following uncertainty parameters: the standard deviation of *NLFE* and generator failure rates. The standard deviations of *NLFE* and generator failure rates were chosen as representative uncertainty parameters for the load and generation sides, respectively. Figure 8 shows the results of the sensitivity analysis. In Figure 8a, the graph was obtained by increasing/decreasing the failure rates of the generators from -50% to +50%; in the horizontal axis, the value of 100 indicates the original failure rates. It should be noted that the variations in *RSE* values did not exceed the *RSE* criterion (i.e., 10.6720 h/day), indicating that the power system can withstand, to some extent, future changes in failure rates. Moreover, the *RSE* changed linearly with the failure rates of the generators, as expected from Equation (4).

The impact of the standard deviation of *NLFE* on the *RSE* was also examined by increasing/decreasing its value from 1% to 9%, with a step size of 1%, as shown in Figure 8b.

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For reference, the default value was set to 5%. The graph shows the increase or decrease for each section, although it tends to increase with respect to standard deviation in general. The standard deviation can be used to interpret the accuracy of the forecasting technique. The results show that the higher the accuracy, the lower the risk (i.e., *RSE*). This result will vary depending on the type of system. The relationship between the *RSE* and uncertainty parameters enables the system operator to effectively manage system risks.

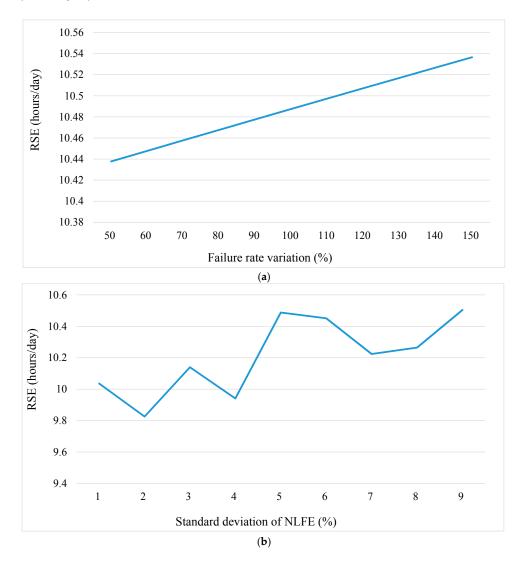


Figure 8. Results of sensitivity analysis: (a) *RSE* and failure rates of generators; (b) *RSE* and standard deviation of net load forecast error (*NLFE*).

5. Conclusions

This study proposed a flexibility-based evaluation method for VG acceptability. We explicitly quantified power system flexibility using a risk index called the *RSE*. In this method, the *RSE* of the current power system was used as the acceptable level of flexibility (i.e., *RSE* criterion). The VG acceptability of the generation expansion plan in Korea for 2029 was evaluated using the proposed method. The results show that the VG penetration level could improve by approximately 12% compared to the planned VG penetration level. The results of sensitivity analysis also show the extent to which system uncertainty affected VG acceptability. As part of future work, it would be interesting to investigate the impact of flexible demand-side resources on the *RSE*. We also plan to examine the time complementarity between VG resources.

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Author Contributions: Chang-Gi Min carried out the main body of research and Mun-Kyeom Kim reviewed the work continuously.

Conflicts of Interest: The authors declare no conflict of interest.

Appendix A. Newly Installed Generators from 2017 to 2029 (★ ST: Steam Turbine; CC: Combined Cycle; Nucl.: Nuclear)

Table A1. Generation output limits and coefficient of cost function.

Generator	Maximum Output (MW)	Minimum Output (MW)	Coefficient of Cost Function			
Name			2nd-Order (Gcal/MW ²)	1st-Order (Gcal/MW)	Constant (Gcal)	
CC1-GT	140	48	0.005454	2.489639	18.7936	
CC2-CC	200	82	0.001115	1.842653	18.01031	
CC3-GT	274	184	0.00003	1.871501	157.2697	
CC4-GT	274	184	0.00003	1.871501	157.2697	
CC3-CC	400	281	0.000386	1.138112	125.5207	
CC4-CC	400	281	0.000386	1.138112	125.5207	
CC5-GT	572.6	330	0.000643	1.19617	472.335	
CC5-CC	1000	542	0.00007	1.501309	100.2745	
CC6-GT	571.6	188	0.000111	2.111027	65.73368	
CC6-CC	950	439	0.000003	1.494521	70.40799	
CC7-GT	572.6	330	0.000643	1.19617	472.335	
CC8-GT	572.6	330	0.000643	1.19617	472.335	
CC7-CC	900	542	0.000045	1.501309	100.2745	
CC8-CC	900	542	0.000007	1.501309	100.2745	
CC9-GT	306.2	116	0.000007	2.461418	67.76106	
CC9-G1	470	175	0.000077	1.551216	40.38037	
	572.6	330				
CC10-GT			0.000643	1.19617	472.335	
CC10-CC	960	542	0.000007	1.501309	100.2745	
CC11-GT	572.6	330	0.000643	1.19617	472.335	
CC11-CC	1000	542	0.000007	1.501309	100.2745	
CC12-GT	572.6	330	0.000643	1.19617	472.335	
CC12-CC	920	542	0.000007	1.501309	100.2745	
ST1	595	260	0.000136	1.863031	107.3797	
ST2	595	260	0.000136	1.863031	107.3797	
ST3	1022	609	0.000098	1.765468	199.9551	
ST4	1022	609	0.000098	1.765468	199.9551	
ST5	1000	609	0.000098	1.765468	199.9551	
ST6	1000	609	0.000098	1.765468	199.9551	
ST7	1000	609	0.000098	1.765468	199.9551	
ST8	1040	609	0.000098	1.765468	199.9551	
ST9	1040	609	0.000098	1.765468	199.9551	
ST10	580	260	0.000136	1.863031	107.3797	
ST11	580	260	0.000136	1.863031	107.3797	
ST12	1040	609	0.000098	1.765468	199.9551	
ST13	1040	609	0.000098	1.765468	199.9551	
ST14	1050	609	0.000098	1.765468	199.9551	
ST15	1050	609	0.000098	1.765468	199.9551	
ST16	1050	635	0.000025	1.899346	165.6219	
Nucl.1	1400	795	0.000105	2.043505	267.8154	
Nucl.2	1400	795	0.000105	2.043505	267.8154	
Nucl.3	1400	795	0.000105	2.043505	267.8154	
Nucl.4	1400	795	0.000105	2.043505	267.8154	
Nucl.5	1500	795	0.000105	2.043505	267.8154	
Nucl.6	1500	795	0.000105	2.043505	267.8154	
Nucl.7	1500	795 795	0.000105	2.043505	267.8154	
Nucl.8	1500	795 795	0.000105	2.043505	267.8154	
Nucl.8 Nucl.9	1400	795 795	0.000105	2.043505	267.8154 267.8154	
Nucl.10	1400	795 705	0.000105	2.043505	267.8154	
Nucl.11	1400	795	0.000105	2.043505	267.8154	
Nucl.12	1400	795	0.000105	2.043505	267.8154	

Table A2. Ramp and on/off constraints.

Generator Name	Ramp-Up Rate (MW/min)	Minimum Up Time (h)	Minimum Down Time (h)	Ramp-Up Rate for Start Up (MW/min)	Ramp-Down Rate for Shut Down (MW/min)
CC1-GT	7	1	1	7	7
CC2-CC	8.7	4	3	8.7	8.7
CC3-GT	13.1	1.3	3.8	13.1	13.1
CC4-GT	13.1	1.3	3.8	13.1	13.1
CC3-CC	20	4	3.8	20	20
CC4-CC	20	4	3.8	20	20
CC5-GT	22.5	3.1	4.8	22.5	22.5
CC5-CC	36	4.2	5.3	48	48
CC6-GT	22.5	1	2	22.5	22.5
CC6-CC	34.5	4	3	34.5	34.5
CC7-GT	22.5	3.1	4.8	22.5	22.5
CC8-GT	22.5	3.1	4.8	22.5	22.5
CC7-CC	36	4.2	5.3	48	48
CC8-CC	36	4.2	5.3	48	48
CC9-GT	11.5	4.2	3	11.5	11.5
CC9-G1 CC9-CC	18.55	4	3	18.6	18.6
CC10-GT	22.5	3.1	4.8	22.5	22.5
CC10-CC	36	4.2	5.3	48	48
CC11-GT	22.5	3.1	4.8	22.5	22.5
CC11-CC	36	4.2	5.3	48	48
CC12-GT	22.5	3.1	4.8	22.5	22.5
CC12-CC	36	4.2	5.3	48	48
ST1	15	6	12	15	15
ST2	15	6	12	15	15
ST3	12.5	8	18	30.6	30.6
ST4	12.5	8	18	30.6	30.6
ST5	12.5	8	18	30.6	30.6
ST6	12.5	8	18	30.6	30.6
ST7	12.5	8	18	30.6	30.6
ST8	12.5	8	18	30.6	30.6
ST9	12.5	8	18	30.6	30.6
ST10	15	6	12	15	15
ST11	15	6	12	15	15
ST12	12.5	8	18	30.6	30.6
ST13	12.5	8	18	30.6	30.6
ST14	12.5	8	18	30.6	30.6
ST15	12.5	8	18	30.6	30.6
ST16	10.5	7.7	20.8	1	1
Nucl.1	0.97	8	12	0.5	1.6
Nucl.2	0.97	8	12	0.5	1.6
Nucl.3	0.97	8	12	0.5	1.6
Nucl.4	0.97	8	12	0.5	1.6
Nucl.5	0.97	8	12	0.5	1.6
Nucl.6	0.97	8	12	0.5	1.6
Nucl.7	0.97	8	12	0.5	1.6
Nucl.8	0.97	8	12	0.5	1.6
Nucl.9	0.97	8	12	0.5	1.6
Nucl.10	0.97	8	12	0.5	1.6
Nucl.11	0.97	8	12	0.5	1.6
Nucl.12	0.97	8	12	0.5	1.6

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