

An enterprise control assessment case study of the energy–water nexus for the ISO New England system

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ABSTRACT

The generation mix of Independent System Operator in New England (ISO-NE) is fundamentally changing. Nuclear, coal, and oil generation facilities are retiring and are replaced with natural gas, solar, and wind generation. Variable renewable energy resources (VREs) such as solar and wind present multiple operational challenges that require new and innovative ways to manage and control the grid. This paper studies how water supply systems (water and wastewater treatment), and water-dependent electricity generating resources (hydro, and thermal power plants) can be operated flexibly to enhance the reliability of the grid. The study's methodology employs the novel Electric Power Enterprise Control System (EPECS) simulator to study power systems operation, and the System-Level Generic Model (SGEM) to study water consumption and withdrawals. This work considers six potential 2040 scenarios for the ISO-NE energy–water nexus (EWN). It presents a holistic analysis that quantifies grid imbalances, normal operating reserves, energy market production costs, and water withdrawals and consumption. For scenarios with high amounts of VREs, the study shows great potential of water resources to enhance grid flexibility through improvements in load-following (up to 12.66%), and ramping (up to 18.35%) reserves. Flexible operation also results in up to 10.90% reduction in the total time VREs are curtailed. Additionally, flexible operation reduces water withdrawals by up to 25.58%, water consumption by up to 5.30%, and carbon dioxide emissions by up to 3.46%. In general, this work provides significant insights into how to jointly control the water and energy supply systems to aid in their synergistic integration.

1. Introduction

The bulk electric power system of New England is fundamentally changing to include more solar and wind generation resources. This evolving resource mix has triggered changes to how the electricity grid is managed and controlled. The bulk of these changes have been in capacity and transmission expansion. However, with the growing uncertainty and variability introduced by variable renewable energy, there is an even greater need for increased amounts of operational flexibility [1,2]. ISO-NE is the independent system operator for the states of Connecticut, Maine, Massachusetts, New Hampshire, Rhode Island and Vermont. It is tasked with performing three critical roles namely; (1) coordinating and running the electricity grid for the region, (2) designing, managing and running the region's deregulated wholesale electricity market based on minimum generation costs, and (3) planning the system such that it continues to meet the region's electricity needs over the next 10 years. Water plays a fundamental

role in the ISO New England (ISO-NE) system. Conventional and run-of-river hydro make up over 9% of the overall electricity generated in the six New England states [3]. An additional 1% of electricity generation comes from the two main pumped energy storage facilities, Bears Swamp and Northfield [3]. In the meantime, over 83% of the current ISO-NE electricity generation fleet comes from thermal generation facilities which withdraw and consume large quantities of water for cooling purposes [3]. In spite of the changing resource mix, recent studies predict that thermal generation facilities will still account for a significant percentage of future generation facilities in 2040 [4,5]. Fig. 1 illustrates the extent of the coupling between the water and electricity generation resources in New England. From Fig. 1, it is clear that most generating facilities are located near a water source and rely on adequate water supply to perform their function. These factors not only indicate significant coupling between the water and electricity supply systems but they also emphasize the need for more coordination

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Nomenclature

Acronyms

EIA	Energy Information Agency
EPECS	Electric Power Enterprise Control System
EWN	Energy–water nexus
FCA	Forward Capacity Auctions
ISO	Independent System Operator
ISO-NE	Independent System Operator New England
NGCC	Natural gas combined-cycle
NICR	net Installed Capacity Requirement
NREL	National Renewable Energy Laboratory
PHEV	plug-in hybrid electric vehicles
ReEDS	Regional Energy Deployment System
RPS	Renewable Portfolio Standards
RTUC	Real-Time unit commitment
SCADA	Supervisory control and data acquisition
SCED	Security-constrained economic dispatch
SCUC	Security-constrained unit commitment
SGEM	System-Level Generic Model
SOARES	System Operational Analysis and Renewable Energy Integration Study
UCED	Unit-commitment-economic-dispatch
VREs	Variable Renewable Energy Resources
CEII	Critical Energy/Electric Infrastructure Information

between the two systems. Specifically, the potential synergies between the two systems cannot be ignored especially as the electricity grid undergoes its sustainable energy transition.

Concern about water security is growing especially with climate change affecting hydrology patterns and the decline of freshwater resources [6–8]. At the same time, significant attention has gone into the integration of variable renewable energy into the electricity grid as a means of decarbonizing the electricity supply system. As discussed in the prequel to this paper [9], the challenges of renewable energy integration and energy–water-nexus are very much related. In addition to presenting low CO₂ emissions, VREs have very low life-cycle water intensities [10] hence reducing the overall water intensity of electricity generating systems. On the other hand, water is easily stored and therefore, has the potential to serve as a flexible energy–water resource on both the supply-side as well as the demand-side so as to support the integration of VREs into electricity operations [11].

The growing penetration of solar and wind poses several challenges to maintaining the reliability of the electricity grid. In addition to being highly variable, these resources also lower the overall marginal costs of electricity forcing thermal units into early retirement [2]. In the absence of established market rules for VREs participation, curtailment is widely applied as a way to balance power systems with high penetrations of VREs. While curtailment serves to balance the grid, it raises the overall production costs as well as emissions. Given these challenges, independent system operators and utilities are largely constrained with respect to maintaining the reliable performance of the grid [2]. Therefore, alternative techniques for managing VREs such as allowing these resources to provide active power support and operating reserves could greatly improve the operating flexibility of the grid [2,12]. Furthermore, engaging active demand-side participation in the provision of ancillary services such as reserves, and active power support through load-shedding or load-shifting would go a long way to improve the flexible performance of the electricity grid [12]. Water and wastewater treatment systems are already equipped with

the necessary monitoring technologies such as supervisory control and data acquisition (SCADA) systems to provide ancillary services, and in turn improve their profits and also achieve a more robust operation of their systems. In order to better leverage the potential synergies in real-time operation of water and power supply systems, the methodologies of energy–water-nexus and renewable energy integration studies must converge.

1.1. Literature review

Despite the benefits of joint operation, renewable energy integration and EWN studies have not yet converged to realize benefits. While some EWN studies have quantified the withdrawals by thermal power plants, these studies have largely been conducted in isolation of actual operation of the electricity generation industry [13–15]. Thus, the full impact on either infrastructure is not assessed. For example, [16] quantifies water withdrawal and consumption coefficients primarily based on literature sources. Other EWN works have focused solely on optimizing the operations of water systems such as in the optimal operation and scheduling of water pumps to minimize electricity usage [17,18] and water pumping costs [19]. These include the optimal scheduling of water systems [20,21] and flexible operation of water systems for electricity demand response [22,23] and other ancillary services so as to maximize returns for water systems [24]. Finally, a small subset of EWN studies have presented mostly single-layer approaches that co-optimize the water and electricity networks. Examples of such works include the optimal network flow in [25], the economic dispatch in [26,27], and the unit commitment problem in [28] for a combined water, power, and co-production facilities. A majority of EWN studies however, still focus on specific case study geographies such as the Middle East [29,30], California [31], or North Africa [32]. Despite the large body of work and research on the energy–water nexus, there is still a lack of a generic, case and geography-independent methodologies that encompass all flows within, and between the water and energy systems [33,34]. In fact, a recent review [35] of EWN studies shows that these studies require integral methodologies that capture the overall complexity of the nexus.

In the meantime, renewable energy integration studies have often been case and geography specific and have mostly utilized unit-commitment-economic-dispatch (UCED) models of power system control to study the operation of electricity markets with large penetrations of VREs [36–38]. A significant percentage of these studies have taken statistical approaches to determine the impact of wind and solar forecast errors on dispatch decisions. A subset of renewable energy integration studies have recognized the vital role of reserves in the balancing performance of systems with high VRE penetration and have thus, focused on the acquisition of normal operating reserves such as load-following, regulation, and ramping reserves [39–42].

However, a recent review of renewable energy integration studies shows major methodological limitations [43]. Firstly, while some studies focus on reserve acquisition, the required quantity of reserves is usually based on the experiences of grid operators which no longer applies to systems with high penetrations of VREs [44,45]. Secondly, most studies only consider either the net load variability or the forecast error in determining the amount of reserves despite evidence that shows that both of these variables contribute towards normal operating reserve requirements [44,46]. Lastly, although studies have shown that VREs possess dynamics that span multiple timescales of power system operation [47,48], most renewable energy integration studies have largely neglected the effect of timescales on the various types of operating reserve quantities [43]. Farid et al. [43] proposed a holistic approach based on enterprise control to study the full impact of VREs on power system balancing performance and reserve requirements while considering the multi-timescale dynamics of VREs. *Enterprise control* is an integrated and holistic approach that allows operators to study and improve the technical performance of the grid while realizing

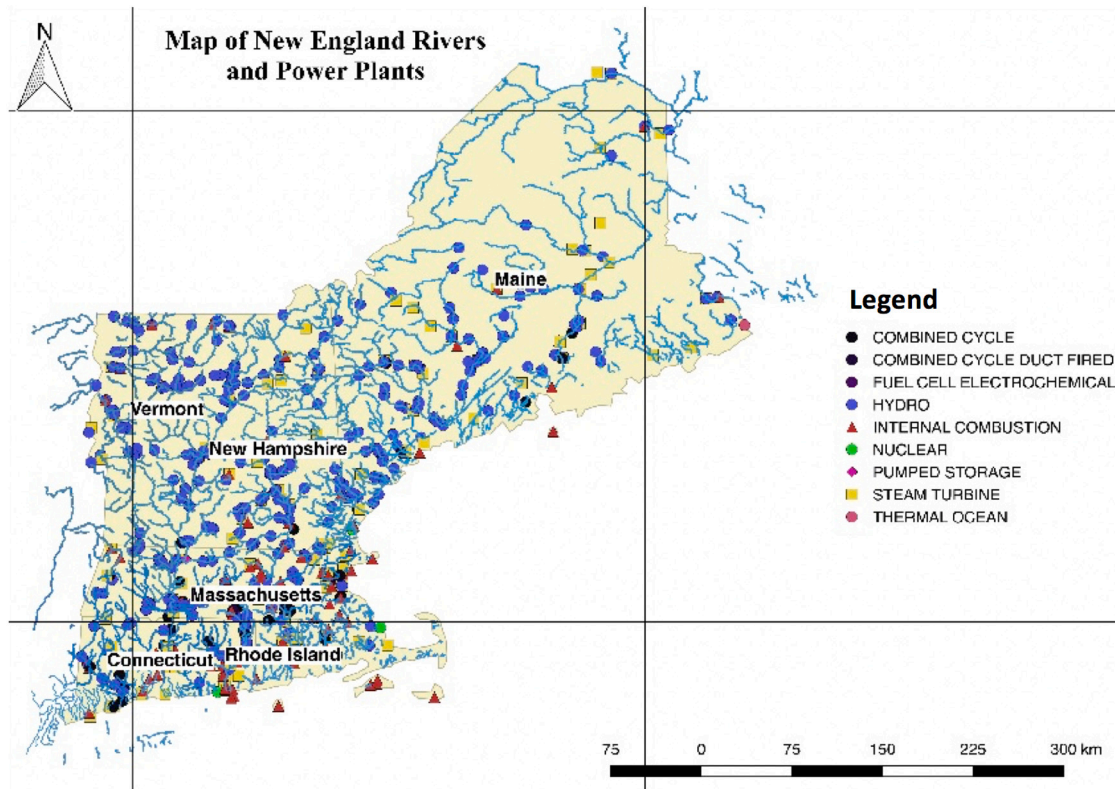


Fig. 1. A map of New England's electric power generation units and rivers.

cost savings [43]. An application of enterprise control in the form of the Electric Power Enterprise Control System (EPECS) simulator has been proposed in literature [43,49–52] and tested on various case studies including the ISO New England system [53]. In [53], the EPECS simulator is used to study the performance of the ISO-NE system on 12 scenarios with varying penetrations of VREs. This study highlighted the key role of curtailment and normal operating reserves on the balancing performance of the ISO-NE system. This paper extends the work in [53] and [9] to quantify the flexibility afforded the ISO-NE system through flexible operation of energy–water resources. For the purposes of this study, the term “energy–water resources” collectively refers to water and wastewater treatment systems (which are assumed to only consume electricity in this study), run-of-river and conventional hydro (which generate electricity), thermal power plants (which consume water for cooling and generate electricity), and finally, pumped energy storage (which consumes and generates electricity).

1.2. Original contribution

The main contribution of this paper is a case study of the energy–water nexus in the New England region. It utilizes the methodology presented in the prequel [9] and extends the results of renewable energy integration study found in [53] to specifically include several environmental performance and economic performance measures. This techno-economic study of the EWN in New England addresses twelve predefined 2040 scenarios; 6 with a “flexible” operation of energy–water resources and 6 “conventional” (i.e. inflexible) operation of energy–water resources. This case study takes the yellow rectangle of Fig. 2 as its system boundary and consequently is able to quantify the mass and energy flows in and out of the defined yellow system boundary regardless of the test case or geographical region. Additionally, this paper provides insight into some of the operational challenges presented by high penetrations of VREs and assesses the flexibility value of flexible energy–water resources by quantifying the amounts

of normal operating reserves for the ISO-NE system for each scenario. Given that the methodology presented in the prequel [9] is generic and modular, the EPECS simulator is modified to reflect the ISO-NE operations as fully outlined in [53]. Each simulation scenario runs for a full year with one minute time step. In this study, the following operational parameters are quantified: (1) load-following, ramping, and regulation reserves, (2) the ability of water and wastewater treatment facilities to shift their electricity demand in response to changes in electricity supply, (3) the fuel flows of thermal units and their carbon emissions, (4) water withdrawals and consumption by thermal power plants, and (5) the overall effect of flexible operation of energy–water resources on the production cost of operation for the New England electricity grid.

1.3. Outline

The rest of the paper is structured as follows: Section 2 presents the methodology for the ISO New England EWN study. Section 3 gives a detailed description of the case study data. Section 4 presents the results of the study within the context of the key performance characteristics of the power grid. Finally, the paper is concluded in Section 5.

2. Methodology

As shown in Fig. 3, the methodology of the ISO New England EWN study is best viewed in two parts: planning and operations. Section 2.1 describes how the National Renewable Energy Laboratory's (NREL) Regional Energy Deployment System (ReEDS) was used to evolve the 2030 ISO New England electric power generation capacity mixes to six distinct 2040 capacity mix scenarios. From there, the remainder of the section describes the Electric Power Enterprise Control System (EPECS) simulator as customized for ISO New England's operation [9,53]. Typically, it includes simulation functionality for two energy market layers: the Security Constrained Unit Commitment (SCUC) and the Security

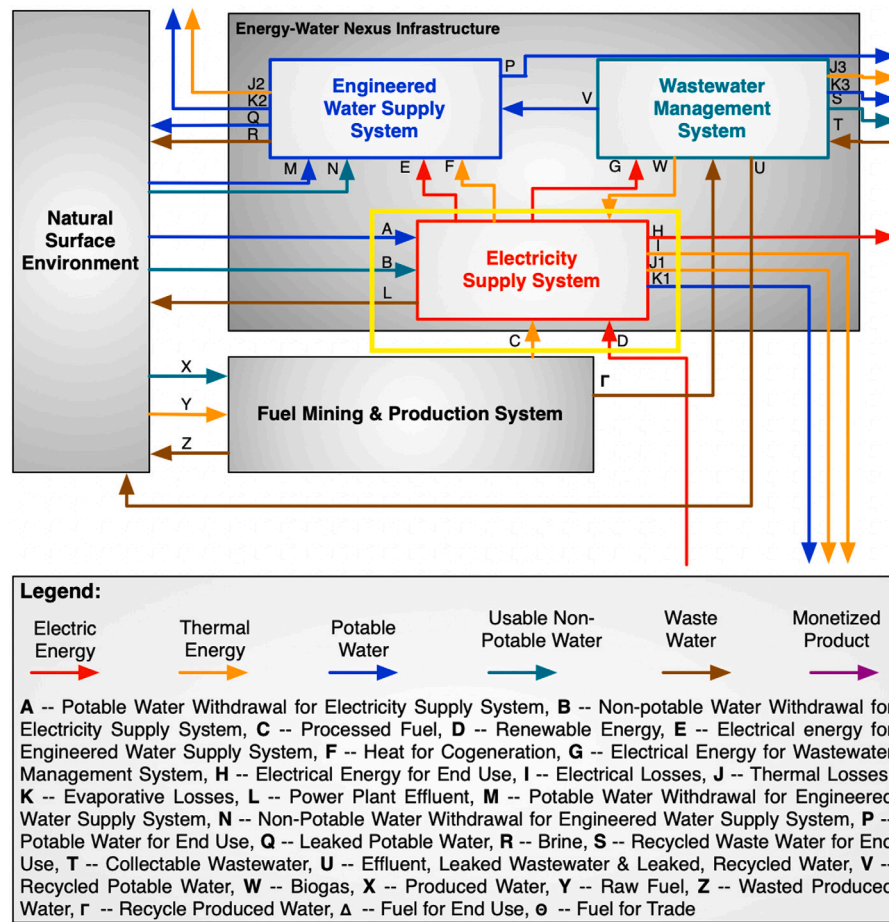


Fig. 2. A diagram of the physical flows between the physical infrastructures (water supply system, wastewater management system, and electricity supply system) and the natural surface environment that were quantified in this study [54].

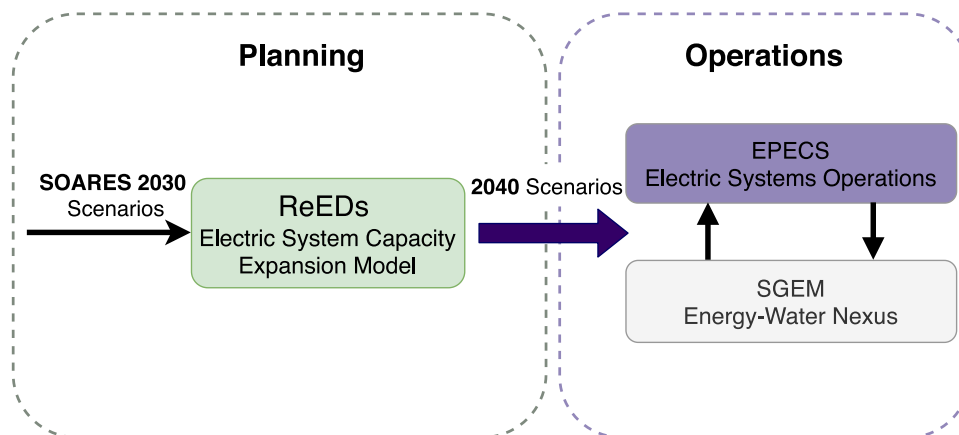


Fig. 3. Diagram of the Simulators Used for the ISO New England EWN Study.

Constrained Economic Dispatch (SCED), power system regulation and a physical model of the power grid itself (i.e. power flow analysis). For this study, the simulator has been customized for ISO-NE operations to include the Real-Time Unit Commitment (RTUC) as shown in Fig. 4. Furthermore, the SGEM model [9,55] is used to capture the essential physics of cooling processes for thermal power plants and in turn compute the water withdrawals and consumption for each power plant.

2.1. Regional energy deployment system (ReEDS) for capacity planning

ReEDS is a capacity planning tool that was developed by NREL starting in 2003. ReEDS is a tool that identifies the long-term evolution of the electric power grid for various regions in the United States [56–58]. At its core ReEDS is an optimization tool that identifies the cost-optimal mix of generation technologies subject to reliability, generation resource, and regulatory constraints [56–58]. The optimization has a

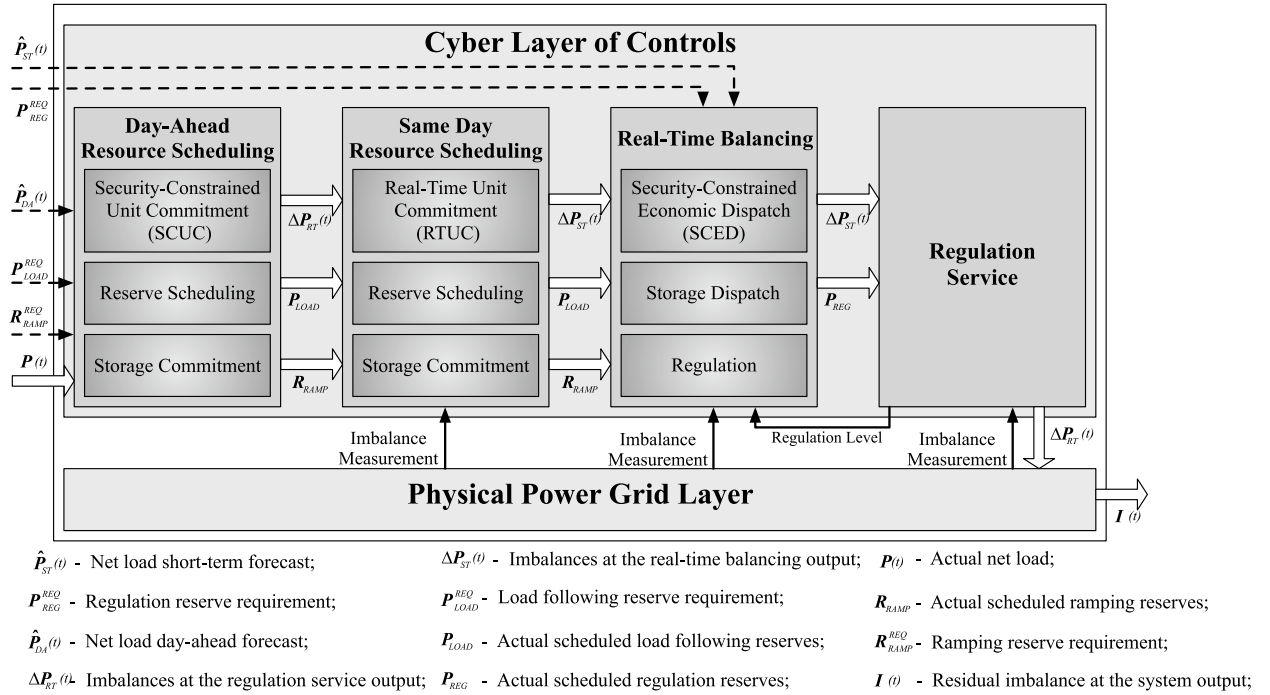


Fig. 4. Architecture of the Electric Power Enterprise Control System (EPECS) simulator customized for ISO New England operations [53].

two-year time step for a total of 42 years ending in 2050[56–58]. The final output of the simulation is generation capacity by technology, storage capacity, electricity costs among others [56–58]. This optimization tool was used to determine the evolution of the ISO-NE system from the 2030 scenarios to the 2040 scenarios. The model input assumptions were selected from configurations defined by the 2018 Standard Scenarios [59] (see Table 1) to align with the 2030 capacity mixes described in Section 3.1. Details on added capacities for each scenario can be found in Section 3.

2.2. The physical power grid

The physical power grid layer of Fig. 4 is represented by the zonal network shown in Fig. 5. The system data is in turn consolidated into the zonal network model of Fig. 5. This zonal network captures the power flows between pre-defined electricity load zones (i.e. “bubbles”) along abstracted “pipes”; thus eliminating the need for Critical Energy/Electric Infrastructure Information (CEII) clearance. The EPECS simulator implements a lossless DC Power Flow Analysis to determine these flows as described in [9,53]. The high-level interface flow limits between the various bubbles are indicative of the line congestion often experienced in the ISO New England territory.

2.3. The security constrained unit commitment (SCUC)

The power system balancing operation commences with the day-ahead resource scheduling Fig. 4 in form of the SCUC. It is performed the day before to determine the best set of generators that can meet the hourly demand at a minimum cost. The time step for the SCUC is 1-hour and it determines the optimal set of generators for the next 24-hours. A simplified version of this program is presented in [9] and the full version customized for ISO-NE operations is presented in [53]. Note that the SCUC formulation used for this study extends the methodology in [53] to also include ramping constraints for wind, solar, and hydro resources [9]. Ramping constraints define the limits to how fast an energy resource can increase or decrease its output per unit time. When variable resources such as solar and wind become semi-dispatchable through curtailment, it means that these resources

must ramp between two consecutive curtailment values (in time). This study assumes these variable energy resources can ramp between their maximum and minimum capacities within a single SCED time step of five minutes as defined in Ref. [9]. Conventional generation resources have ramp rates as well.

2.4. Real-Time Unit Commitment (RTUC)

The same day resource scheduling of Fig. 4 is conducted every hour through the RTUC. It uses an optimization program that is quite similar to that of SCUC but only commits and de-commits *fast-start* units. Fast-start units are defined by their ability to go online and produce at full capacity within 15–30 min. The RTUC runs every hour with a 15-minute time step and a 4-hour look-ahead. The complete mathematics for the RTUC can be found in [53] with slight modifications to include ramping constraints for wind, solar, and hydro resources as presented in [9].

2.5. The Security Constrained Economic Dispatch (SCED)

The real-time balancing operation of Fig. 4 is implemented through the SCED which is run every 10-minutes. The role of the SCED is to move available generator outputs to new set points in a cost-effective way. The SCED does not bring online any units but rather ramps up or down the available online units. The SCED methodology is presented in [9,53] and similar to SCUC and RTUC, it has been extended to allow for the ramping of wind, solar, and hydro resources [9]. A more comprehensive description of the EPECS methodology and mathematical formulations for each control layer can be found in [9,53]. This methodology has been analyzed and validated by ISO-NE.

2.6. Regulation

A pseudo-steady-state approximation of the regulation service model that ties directly to a power flow model of the physical power grid is also used in this study. Normally, imbalances at the output of the regulation service would be represented in the form of frequency changes [60]. However, for steady-state simulations with 1-minute

Table 1

This table maps the SOARES 2030 scenarios to the ReEDS 2018 standard scenarios [59] that were used to evolve the SOARES 2030 scenario data into the 2040 scenarios used for this study.

Scenario Name	SOARES 2030 Scenarios	ReEDS Scenarios
Scenario 2040-1	RPSs + Gas	High RE Cost
Scenario 2040-2	ISO Queue	Accelerated Nuclear Retirements
Scenario 2040-3	Renewables Plus	Low RE Cost
Scenario 2040-4	No Retirements beyond Forward Capacity Auctions (FCA) #10	Low Wind Cost
Scenario 2040-5	ACPs + Gas	Extended Cost Recovery
Scenario 2040-6	Renewable Portfolio Standards (RPSs) + Geodiverse Renewables	Low Natural Gas Prices

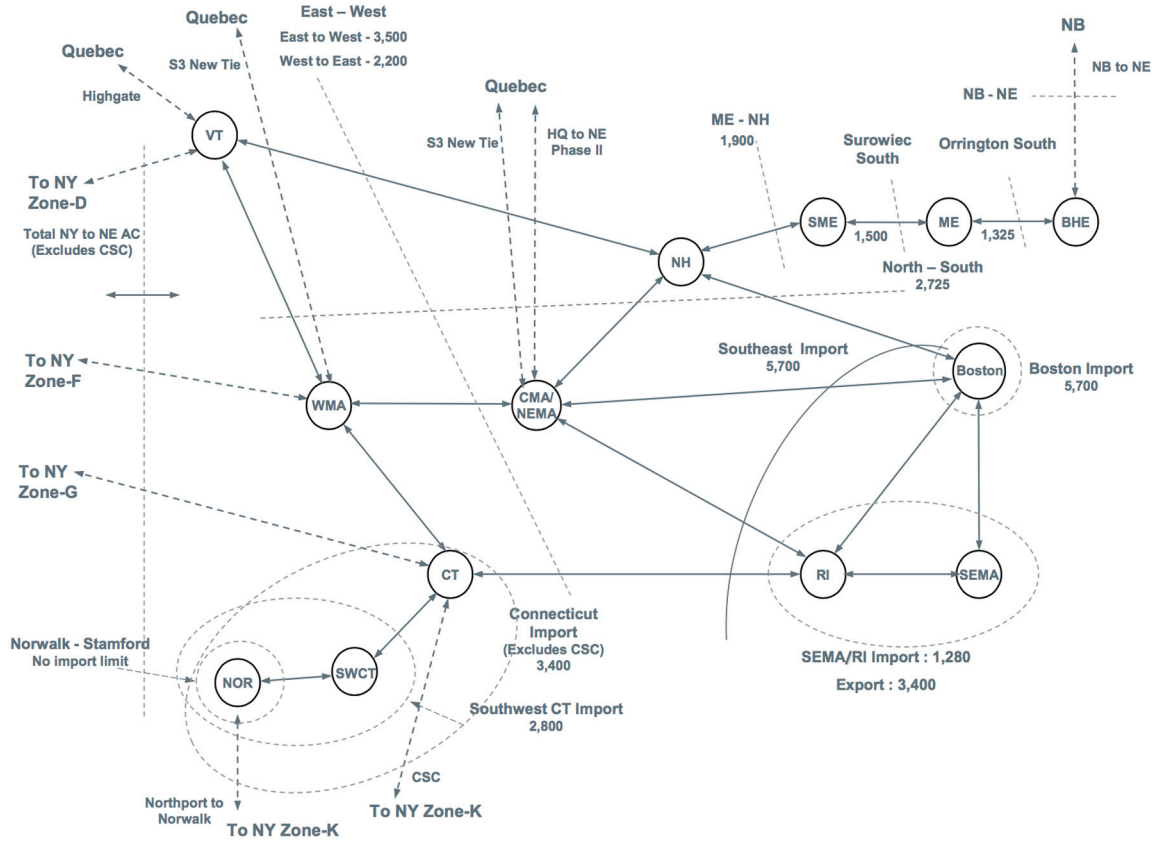


Fig. 5. The ISO-NE zonal network model represented as “pipes” and “bubbles”[53].

time step, the concept of frequency is not applicable. Instead, a designated *virtual* swing bus consumes the mismatches between generation and electricity load to make the steady state power flow equations solvable [53].

2.7. Variable renewable energy

Variable renewable energy resources in the EPECS simulator are studied as time-dependent, spatially distributed exogenous quantities that contribute directly to the electricity net load. Where the term *net load* here is defined as the difference between the aggregated electricity system load and the total generation produced by VREs, tieline profiles and any transmission losses [53].

As previously defined in [9], the EPECS simulator differentiates energy resources into several classes:

Definition 2.7.1 (Variable Renewable Energy Resources (VREs)). Generation resources with a stochastic and intermittent power output. Wind, solar, run-of-river hydro, and tie-lines are assumed to be VREs.

Definition 2.7.2 (Semi-Dispatchable Resources). Energy resources that can be dispatched downwards (i.e curtailed) from their uncurtailed power injection value. When curtailment is allowed for VREs, they become semi-dispatchable. In this study, wind, solar and tie-lines are treated as semi-dispatchable resources. Note that for the purposes of this study, electricity generated by run-of-river and conventional hydro resources can be curtailed and, therefore, these resources are treated as semi-dispatchable in the “flexible case” mentioned in Section 3.1. However, in the “conventional case”, the electricity output of run-of-river and conventional hydro resources is *not* semi-dispatchable but rather variable. Similarly, water and wastewater treatment facilities have the ability to shed their electricity consumption in the “flexible case” and are inflexible or variable in the “conventional case”.

Definition 2.7.3 (Must-Run Resources). Generation resources that must run at their maximum output at all times. In this study, nuclear generation units are assumed to be *must run* resources.

Definition 2.7.4 (Dispatchable Resources). Energy resources that can be dispatched up and down from their current value of power injection. In this study, all other resources are assumed to be dispatchable.

The EPECS simulator employs the operating reserve concepts described in [61,62] with only a few changes. This study focuses on the normal operating reserves that are able to respond to real-time changes in wind and solar generation. Specifically, how much of these reserve quantities comes from electricity generated by water resources such as conventional hydro and run-of-river hydro power plants, and the electricity load-shedding potential of electricity consumed by water and waste-water treatment facilities. Normal operating reserves are classified as load following, ramping, and regulation reserves based on the mechanisms upon which they are acquired and activated. For the purposes of this study, the curtailment of VREs was assumed to provide both load-following and ramping reserves in an upward direction to their forecasted value and in a downward direction to their minimum operating capacity limit.

These three types of operating reserves work together to respond to real-time forecast errors and variability in the electricity *net load* during normal system operation. Note that the actual quantities of these reserves are physical properties of the power system and exist regardless of whether they are monetized or not. The EPECS simulator provides as output the following quantities: system imbalances, operating reserves (load-following, ramping and regulation), generator set points, curtailed generation and line flows for every minute.

2.8. System-level Generic Model (SGEM)

The SGEM was developed to study the water use of fossil fuel, nuclear, geothermal and solar thermal power plants using either steam or combined cycle technologies [63]. This model is also geography and case-independent; making it ideal for application to the ISO-NE system. Three main cooling processes are applied in this paper: once-through cooling, wet tower cooling and dry-air cooling. Majority of the older generation power plants used once-through cooling technology while the newer power plants were either recirculating or dry-cooling. The formulae for computing water withdrawals and consumption are presented in [9].

With this information, the energy–water flows through the yellow system boundary of Fig. 2 can be easily quantified (as detailed in [9]) to determine, water withdrawals and consumption by thermal power plants, as well as other aspects such as fuel consumption and CO₂ emissions. As illustrated in Fig. 2, it is important to capture all the physical flows between the three physical infrastructures (water supply system, wastewater management system, and electricity supply system) and the natural surface environment. In this study, however, each water resource fits within an electric power system load area (or “bubble” as they commonly called within the New England Power Pool). Therefore, full hydraulic modeling does not provide additional insight in the provision of flexibility services to the electric power grid. The approach presented here is sufficient to capture all the interfaces between the water supply system and the electricity supply system and impose aggregate energy constraints as necessary.

2.9. Assessing the flexibility of the system

The term power system flexibility is quantified by assessing the availability of several different types of normal operating reserves namely; load-following, ramping, and regulation reserves. Together, these reserves determine how well the system can respond to real-time variability in the electricity system net load. The formulae for these reserves are established in the following Refs. [61,62,64]. Therefore, a system with abundant amounts of operating reserves is well-equipped to respond to real-time variability in electricity net load and is thus, considered to be more flexible.

3. Case study scenarios and data

3.1. Study scenarios

The case study scenarios presented in this work are best understood in the context of the twelve scenarios that were studied in the 2017 System Operational Analysis and Renewable Energy Integration Study (SOARES) that was commissioned by ISO-NE. These 12 scenarios distinguished between the amount and diversity of dispatchable generation resources, electricity load profiles, and the penetration of VREs [53]. Of these scenarios, six were meant to describe the year 2025 while the other six were meant to describe the year 2030. Both the 2025 and 2030 scenarios used in the SOARES were defined by ISO New England and its respective stakeholders. The ReEDS capacity expansion model was used to evolve the 2030 SOARES scenarios to the 2040 scenarios used in this work. To achieve this, the ReEDS capacity planning tool was first calibrated to reach the SOARES 2030 energy mixes from a 2015 base year. From there, the ReEDS model was extended along these six distinct trends (as outlined in Table 1) another 10 years into the future to 2040 to reach the energy mixes presented here. The final capacity mixes of the six 2040 scenarios are summarized in Fig. 6 and are described further below. Note that these scenarios are by no means a prediction of ISO New England’s future energy mixes. They are simply indicative of the trends demonstrated by the SOARES 2030 scenarios if they were to continue another 10 years to 2040.

In order to assess the value of uncoordinated vs coordinated EWN operation, each of these six scenarios were simulated twice; once with energy–water resources as variable resources and another as semi-dispatchable resources. These scenario variants are respectively referred to as the “conventional” operating mode (as a control case) and the “flexible” operating mode (as the experimental case).

3.1.1. Scenario 2040-1: RPSs + gas

In this scenario, the oldest oil and coal generation units are retired by 2030 and the retired units are replaced by natural gas combined-cycle (NGCC) units at the same locations. Furthermore, the ReEDS model adds 50 MW of biomass, 233 MW of solar, 75MW of hydro and 6351 MW of natural gas (NG) to this scenario. It also retires 870 MW of nuclear, 667 MW of NG and 1127 MW of oil generation.

3.1.2. Scenario 2040-2: ISO queue

In this scenario, the retired oil and coal units from Scenario 1 are replaced by renewable energy resources instead of NGCC. The locations of the renewable energy resources are determined according to the ISO-NE Interconnection Queue. The ReEDS model resulted in the addition of 2498 MW of solar, 9.77 MW of hydro, and 5831.75 MW of NG (mostly in New Hampshire). In addition, 2471 MW of nuclear, 668 MW of natural gas and 25 MW of coal generation units were retired.

3.1.3. Scenario 2040-3: Renewables plus

In this scenario, more renewable energy resources are used to replace the retiring units. Additionally, battery energy systems, energy efficiency and plug-in hybrid electric vehicles (PHEV) are added to the system. Moreover, two new tie lines are added to increase the amounts of hydroelectricity imports. The ReEDS model results in the following modifications to this scenario: (1) addition of 2760 MW of solar, 9 MW of hydro, 2765 MW of NG, and (2) the retirement of 378 MW of coal, 870 MW nuclear, 667 MW of NG and 1127 MW of oil.

3.1.4. Scenario 2040-4: No retirements beyond Forward Capacity Auctions (FCA) #10

In contrast to other scenarios, no generation units are retired beyond the known FCA resources. The FCA resources are replaced by NGCC located at the Hub. This scenario is the *business-as-usual* scenario. The ReEDS model results in the following modifications to this scenario: (1) addition of 989 MW of solar, 4.2 MW of hydro, and 3987 MW of NG, and (2) the retirement of 383 MW of coal, 870 MW nuclear, 667 MW of NG and 1127 MW of oil.

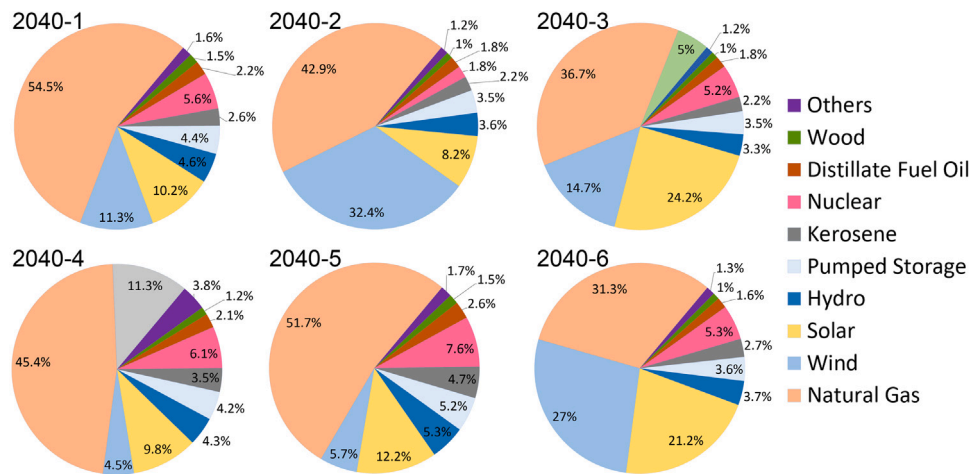


Fig. 6. Summary of available generation capacity as a percentage of total available capacity by fuel type for all six 2040 scenarios.

3.1.5. Scenario 2040-5: ACPs + gas

In this scenario, the oldest oil and coal generation units are retired by 2030 and these units are replaced by new NGCC units to meet the net Installed Capacity Requirement (NICR). The ReEDS model results in the following modifications to this scenario: (1) addition of 3089 MW of solar, 11.1 MW of hydro, and 2496 MW of NG, and (2) the retirement of 253 MW of coal, 870 MW nuclear, 667 MW of NG and 1127 MW of oil.

3.1.6. Scenario 2040-6: Renewable Portfolio Standards (RPSs) + geodiverse renewables

This scenario is similar to Scenario 5 but instead of replacing the retired units with NGCC units, additional renewable energy generation is used to meet the RPSs and the NICR. However, the solar PV and offshore wind units are located closer to the main electricity load centers while the onshore wind is located in a remote area in Maine. The ReEDS model results in the following modifications to this scenario: (1) addition of 3011 MW of solar, 6.2 MW of hydro, and 2430 MW of NG, and (2) the retirement of 870 MW nuclear, 667 MW of NG and 1127 MW of oil.

In addition to the changes in capacity mixes implemented in ReEDS, interface limits shown in Fig. 5 were raised to reflect the likely situation that New England would work to resolve line congestion found in the 2025 and 2030 scenarios in the SOARES scenarios [53]. Finally, in addition to the electric data, data on power consumption by water and waste-water treatment facilities as well as the cooling mechanisms of thermal generators were used to determine their share of the peak electricity load. The cooling data for thermal power plants was further enhanced by data from the Energy Information Agency's (EIA) databases [65–67].

3.2. Electricity net load profiles

The electricity net load profile comprised of the system electricity load profile minus the electricity generation from wind, solar, run-of-river and pond-hydro power plants, as well as tie-line flows between New England and other regions. Fig. 7 contrasts the electricity net load profile of Scenario 2040-4 as a “business-as-usual” case to that of Scenario 2040-3 as a high VRE case. The latter exhibits significant negative net load especially during low electricity load periods such as the Spring and Fall seasons. Fig. 8 summarizes the statistics of the electricity net load profiles for all six scenarios. The system electricity peak load for Scenarios 2040-1/2/4/5/6 was 28594 MW while that of Scenario 2040-3 was 22103 MW due to a higher penetration of energy efficiency measures. All scenarios had the same profile for electricity demand by water and wastewater treatment facilities. Run-of-river and pond-hydro

generation profiles were curtailable at a price of \$4.5/MWh similar to the 2017 ISO-NE SOARES. In this study, electricity consumed by water and wastewater treatment plants is treated as flexible in that it has a load-shedding rather than load-shifting capability and is assumed to contribute towards operating reserves. The 709 GWh of available pumped energy storage capacity is treated as dispatchable for all six scenarios throughout the study. Table 2 summarizes the capacity data for these flexible energy–water resources. Again, in order to assess the “flexibility value” of these energy–water resources, each of the six scenarios is simulated in a “conventional or uncoordinated” mode of operation and a “flexible or coordinated” mode of operation.

4. Results & discussion

Given the aforementioned scenarios, the value of flexible energy–water resources is assessed from reliability, economic, and environmental perspectives. From a reliability perspective, Section 4.1 presents the relative improvements in the system's balancing performance as quantified by the available quantities of operating reserves (i.e. load-following, ramping, and regulation reserves), curtailment, and the magnitude of system imbalances. From an environmental perspective, Section 4.2 quantifies the improvements in the quantities of water withdrawn and consumed as well as CO₂ emitted. Here, *water withdrawn* refers to the volumetric flow rate of water withdrawn from the natural surface environment and *water consumption* refers to the amount of water not returned to its original point of withdrawal (due to evaporative losses). Finally, Section 4.3 quantifies the associated production costs in the day-ahead and real-time energy markets.

4.1. Balancing performance of coordinated energy–water operation

As mentioned above, this section presents the system balancing performance improvements as result of coordinated energy–water operation in terms of: the available quantities of operating reserves (i.e. load-following, ramping, and regulation reserves), curtailment, and the magnitude of system imbalances.

4.1.1. Load-following reserves

In day-to-day operation, the upward and downward load-following reserves are used in time to allow the system to respond to variability and uncertainty in the electricity net load. In the traditional operation of the electricity grid, having sufficient load-following reserves is a primary concern especially in systems with high penetrations of renewables. Both upward and downward load-following reserves are equally important in ensuring system reliability. As upward load following

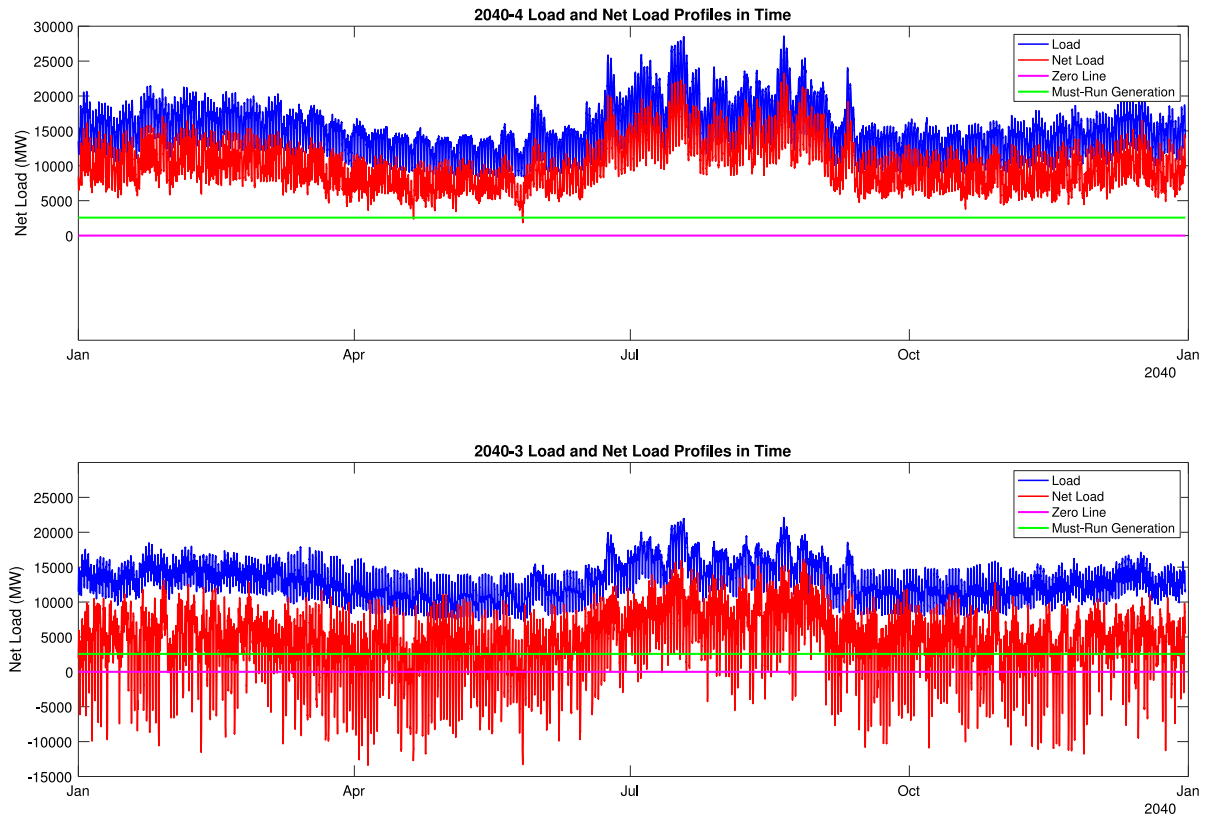


Fig. 7. The load and net load profiles from Scenario 2040-4 (top) and 2040-3 (bottom).

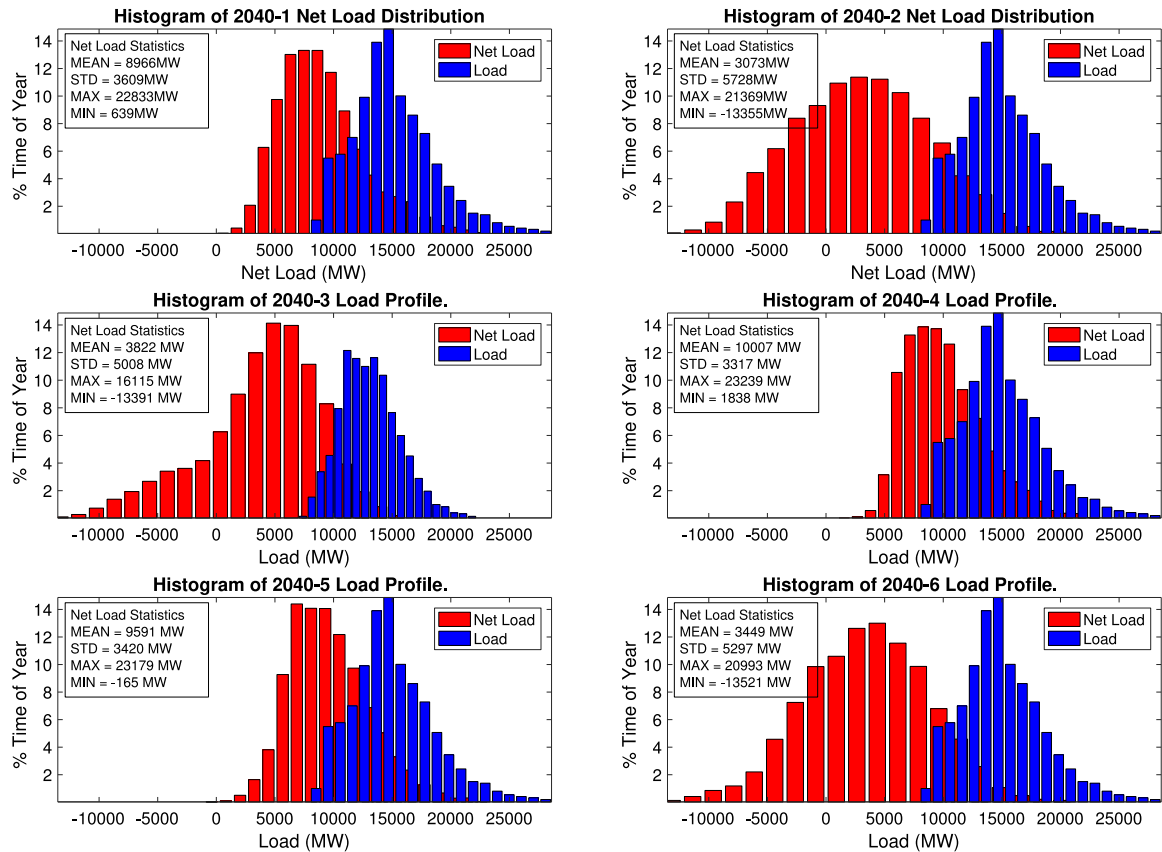
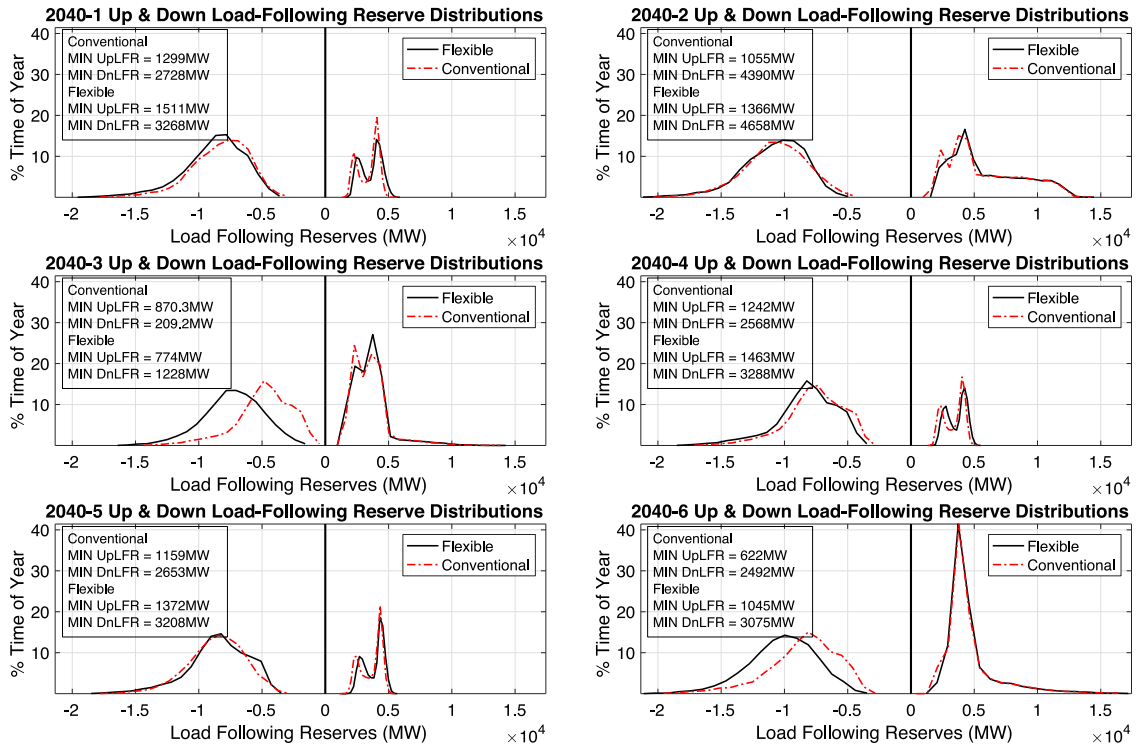


Fig. 8. A comparison of load and net load distributions for all six 2040 scenarios.

Table 2

A summary of available flexible water resources in the system as percentage of the peak load.

	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Hydro Run-of-River & Pond	1854MW (6.21%)	1788MW (5.99%)	1646MW (7.10%)	1782MW (5.97%)	1798MW (5.99%)	1784MW (5.97%)
Pumped Storage	1758MW (6.15%)	1758MW (6.15%)	1758MW (6.15%)	1758MW (6.15%)	1758MW (6.15%)	1758MW (6.15%)
Water Load	565MW (1.89%)	565MW (1.89%)	565MW (2.44%)	565MW (1.89%)	565MW (1.89%)	565MW (1.89%)
System Peak Load	28594 MW	28594 MW	22103MW	28594MW	28594 MW	28594 MW

**Fig. 9.** Distributions of the available upward and downward load following reserves for all six 2040 scenarios in both the conventional and flexible operating modes.**Table 3**

Change in downward and upward load-following reserves statistics (flexible minus conventional) for 2040 scenarios.

Δ LFR (MW)	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Up Mean	208.1 (5.77%)	171.7 (2.86%)	65.6 (1.83%)	207.1 (5.78%)	194.2 (5.08%)	57.7 (1.24%)
Up STD	8.4 (1.00%)	-55.6 (-1.94%)	-17.3 (-1.22%)	-42.1 (-5.32%)	-67.6 (-8.36%)	-36.09 (-1.74%)
Up Max	178.3 (3.07%)	228.3 (1.56%)	335.3 (2.32%)	242.5 (4.37%)	107.9 (1.92%)	686.8 (3.94%)
Up Min	211.9 (14.03%)	311.1 (22.77%)	-96.3 (-12.45%)	221.2 (15.12%)	212.6 (15.50%)	422.6 (40.46%)
Up 95 percentile ¹	241.1 (10.51%)	282.7 (11.59%)	6.0 (0.31%)	288.9 (12.35%)	294.6 (11.83%)	244.5 (9.15%)
Down Mean	743.8 (8.48%)	801.6 (7.41%)	925.5 (12.66%)	647.2 (7.83%)	744.0 (8.77%)	984.1 (9.68%)
Down STD	8.75 (0.36%)	16.29 (0.66%)	36.01 (1.52%)	2.98 (0.12%)	9.50 (0.39%)	67.97 (2.55%)
Down Max	1177.0 (6.11%)	932.5 (4.37%)	1678.0 (10.27%)	961.1 (5.22%)	1086.0 (5.79%)	1424.0 (6.77%)
Down Min	540.3 (16.53%)	267.9 (5.75%)	1019.0 (82.96%)	720.5 (21.91%)	554.9 (17.30%)	583.2 (18.97%)
Down 95 percentile	749.0 (13.96%)	790.6 (10.79%)	1026.0 (28.55%)	717.7 (14.73%)	750.7 (14.99%)	876.3 (14.43%)

Table 4Change in downward and upward ramping reserves statistics (flexible *minus* conventional) for all six 2040 scenarios.

Δ RampR Stats (MW/min)	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Up Mean	334.9 (11.83%)	259.4 (5.28%)	291.3 (8.26%)	308.7 (13.78%)	325.3 (14.31%)	287.7 (6.16%)
Up STD	14.8 (2.86%)	27.9 (5.42%)	3.5 (0.31%)	16.3 (3.40%)	11.6 (2.55%)	15.8 (1.48%)
Up Max	430.7 (10.40%)	354.7 (5.65%)	271.0 (4.83%)	361.5 (10.43%)	372.9 (10.58%)	331.1 (4.79%)
Up Min	-59.3 (-3.89%)	69.7 (3.07%)	410.6 (47.32%)	-4.4 (-0.40%)	-5.6 (-0.49%)	305.1 (15.21%)
Up 95 percentile	310.6 (14.77%)	195.5 (4.68%)	314.9 (14.11%)	300.0 (18.78%)	318.0 (19.19%)	42.5 (1.28%)
Down Mean	339.7 (14.81%)	261.8 (5.86%)	292.3 (8.70%)	317.3 (18.35%)	325.8 (17.88%)	288.9 (6.50%)
Down STD	16.4 (3.69%)	21.4 (4.81%)	1.5 (0.13%)	16.1 (3.67%)	12.7 (2.94%)	12.4 (1.20%)
Down Min	294.2 (22.51%)	22.1 (1.06%)	199.7 (31.65%)	-15.1 (-1.92%)	-6.7 (-0.76%)	293.9 (18.44%)
Down Max	417.3 (15.37%)	354.3 (7.06%)	275.9 (5.64%)	385.1 (17.38%)	345.1 (14.42%)	320.7 (5.40%)
Down 95 percentile	344.3 (19.12%)	208.5 (5.31%)	308.0 (13.94%)	328.3 (26.15%)	337.4 (24.92%)	42.1 (1.32%)

Table 5Change in the curtailment statistics (flexible *minus* conventional) for all six 2040 scenarios.

	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Tot. Semi-Disp. Res. (GWh)	0.00	0.00	0.00	0.00	0.00	0.00
Tot. Curtailed Semi-Disp. Energy (GWh)	17.71	-1.95	60.86	23.44	20.57	-6.18
% Semi-Disp. Energy Curtailed	0.03	-0.00	0.07	0.05	0.04	-0.01
% Time Curtailed	-10.42	-2.67	-5.97	-10.90	-10.74	-3.08
Max Curtailment Level (MW)	1.82	2.68	330.16	-63.03	-1.81	397.67

reserves are exhausted (approach zero), the ability of the system to respond to fluctuation in the electricity net load is constrained.

Therefore, an enhanced balancing performance with respect to load following reserves would show a significant trough around the zero LFR-axis in the distributions of load following reserves shown in Fig. 9. The larger the trough is, the more the system is not using its load following reserves to balance the system. Fig. 9 shows that the flexible use of energy–water resources (in black) widens the trough of load-following reserves around the zero line relative to conventional operation (in red). These graphical results are confirmed numerically in Table 3. Flexible operation enhances the mean values of the upward and downward load following reserves (treated as separate distributions) by 1.24%–12.66% across all six scenarios. Furthermore, the minimum upward and downward load following reserves are improved by flexible operation by 5.75%–82.96% across all but one of the six scenarios. The minimum statistic is particularly important because it defines a type of worst case “safety margin” that the system will always have available to ensure its security. Similarly the 95 percentile statistic gives a measure of how much this minimum level increases when 5% of the distribution is treated as abnormal outlier behavior. The simulations show improvements in the 95 percentile statistic of 0.13–28.55% across all six scenarios; thus demonstrating its robustness to not just the minimum worst-case point but also the distribution tail that represents challenging periods of operation. The maximum and standard deviation statistics are provided for completeness.

4.1.2. Ramping reserves

Ramping reserves describe the total amount of power that the system can respond up or down within a minute. Traditionally, only

Table 6Change in regulation reserves statistics (flexible *minus* conventional) for all six 2040 scenarios.

	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
% Time Reg. Res Exhausted	0.001	0.001	0.000	0.000	0.001	0.001
Reg. Res. Mileage (GWh)	1.800	0.354	0.788	1.014	1.190	0.468
% Reg. Res. Mileage	1.349	0.251	0.638	0.777	0.909	0.326

dispatchable resources are assumed to contribute towards ramping reserves. In this study, renewable energy resources are semi-dispatchable by virtue of curtailment. Consequently, they are assumed to not just be able to ramp down or up to their minimum or maximum values but also do so within five minutes given their power-electronics based control. Five minutes, in this case, coincide with the minimum time-step used in the real-time market. Similar to load-following reserves, ramping reserves are key to ensuring that the system can respond in time to fluctuations in the electricity net load. Having sufficient amounts of both upward and downward ramping reserves is equally important to ensuring reliable performance. As the amount of ramping reserves approaches zero, the ability of the system to respond to net load variability is significantly diminished.

Similar to load-following reserves, both upward and downward ramping reserves are enhanced through the flexible operation of energy–water resources. Fig. 10 illustrates a widened trough in the

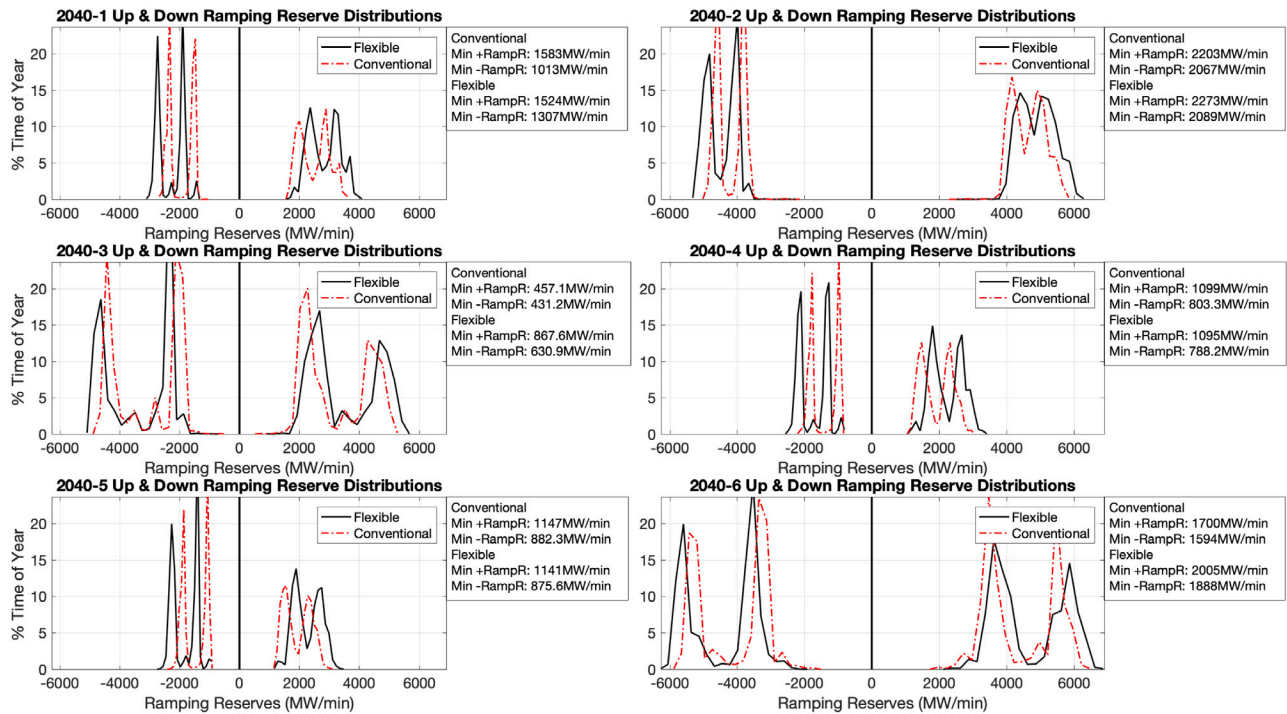


Fig. 10. Distributions of the available upward and downward ramping reserves for all six 2040 scenarios in both the flexible and conventional operating modes.

flexible operating mode relative to the conventional mode. This observation is supported by the statistics in Table 4. The mean value for the upward ramping reserves is improved across all scenarios by up to 14.31%. Likewise, the mean downward ramping reserves are improved by up to 18.35%. Another key measure of sufficient ramping reserves is the minimum level. As illustrated in Table 4, flexible operation enhances the minimum downward ramping reserves by 31.65% and the minimum upward ramping reserves by a maximum of 47.32%. However, in cases with a lower penetration of VREs such as scenarios 2040-1/4/5, the minimum levels are slightly worse in the flexible case than in the conventional case. Despite these anomalies, flexible operation improved 95% percentile levels of upward and downward ramping reserves in all cases (by 1.28%–26.15%). These results show that the curtailment of VREs increases the flexibility to the system if they are used to provide ramping reserves. A complete summary of ramping reserves statistics for all six scenarios is found in Table 4.

4.1.3. Curtailment

By definition, flexible energy–water resources increase the amount of generation available for curtailment. Recall that by Definition 2.7.2, run-of-river and conventional hydro-pond resources are semi-dispatchable resources that can be curtailed in a flexible operating mode. As illustrated in Fig. 11, scenarios with a lower penetration of VREs such as scenario 2040-1/4/5 curtail infrequently and the amount of megawatt curtailed is generally zero. For scenarios 2040-2/3/6, curtailment is used at least 40% of the time. Although, the two cases appear to have similar curtailment levels, a closer look at Table 5 shows that the flexible case curtails for a smaller percentage of the year (2.67%–10.9%) less than the conventional case. Furthermore, the two operating modes show nearly identical levels of total curtailed energy. In the absence of sufficient load-following and ramping reserves, curtailment serves a key role in ensuring system balance. This role is particularly crucial for VREs located in remote areas (e.g. Maine) where it serves as the only control option given topological constraints and distance from electricity load areas.

4.1.4. Regulation service

The regulation service is used to correct system imbalances in real-time. This control lever is used to meet any left-over imbalances after curtailment, load-following and ramping reserves have been used up during market operation. In both cases, all scenarios appear to use their regulation effectively as shown in Fig. 12. This is indicative of a system that has sufficient regulation to mitigate real-time imbalances and maintain balancing performance. A closer inspection of Table 6 illustrates that flexible operation marginally increases the reliance on regulation (as shown by the excess mileage) and exhausts its regulation (albeit for a small fraction of the year 0.001) for all but scenarios 2040-3 and 2040-4.

4.1.5. System imbalances

Balancing performance indicates the residual imbalances after the regulation service has been deployed. Given that the regulation service was barely saturated, the amount of imbalances are expected to be minimal. As shown in Fig. 13, flexible energy–water resources had a small impact on the range of final imbalances of the system. Both systems appear to perform similarly with all cases maintaining a standard deviation of less than 16MW across all six scenarios. Table 7 illustrates that the flexible operating mode performs slightly better than the conventional with up to a 6.48% improvement in standard deviation. The minimum imbalances are lower in all cases except for Scenarios 2040-1 and 2040-2. Similarly, the maximum imbalances are lower for the flexible operating mode except for Scenarios 2040-2 and 2040-6 which represent scenarios with high VREs.

4.2. Environmental performance of coordinated energy–water operation

As mentioned before, the environmental performance of coordinated energy–water operation is assessed through overall reductions in water withdrawals, consumption and CO₂ emissions.

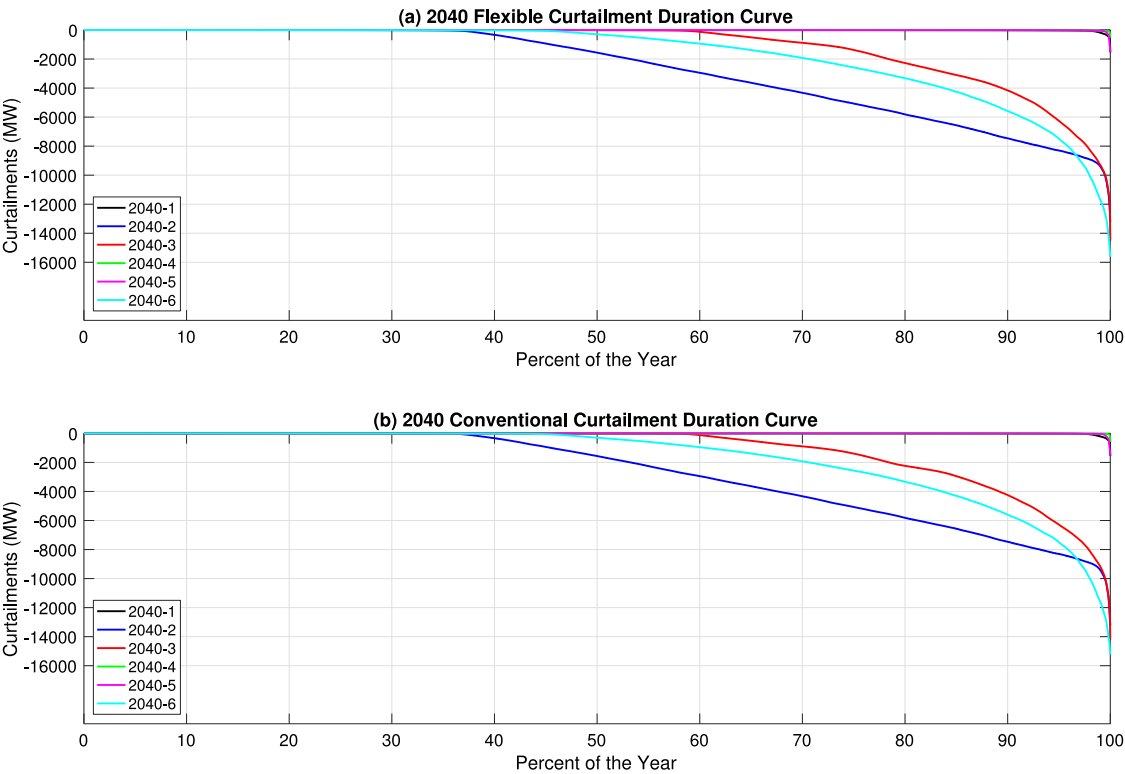


Fig. 11. Curtailment duration curves for all six 2040 scenarios in both the flexible (above) and conventional (below) operating modes.

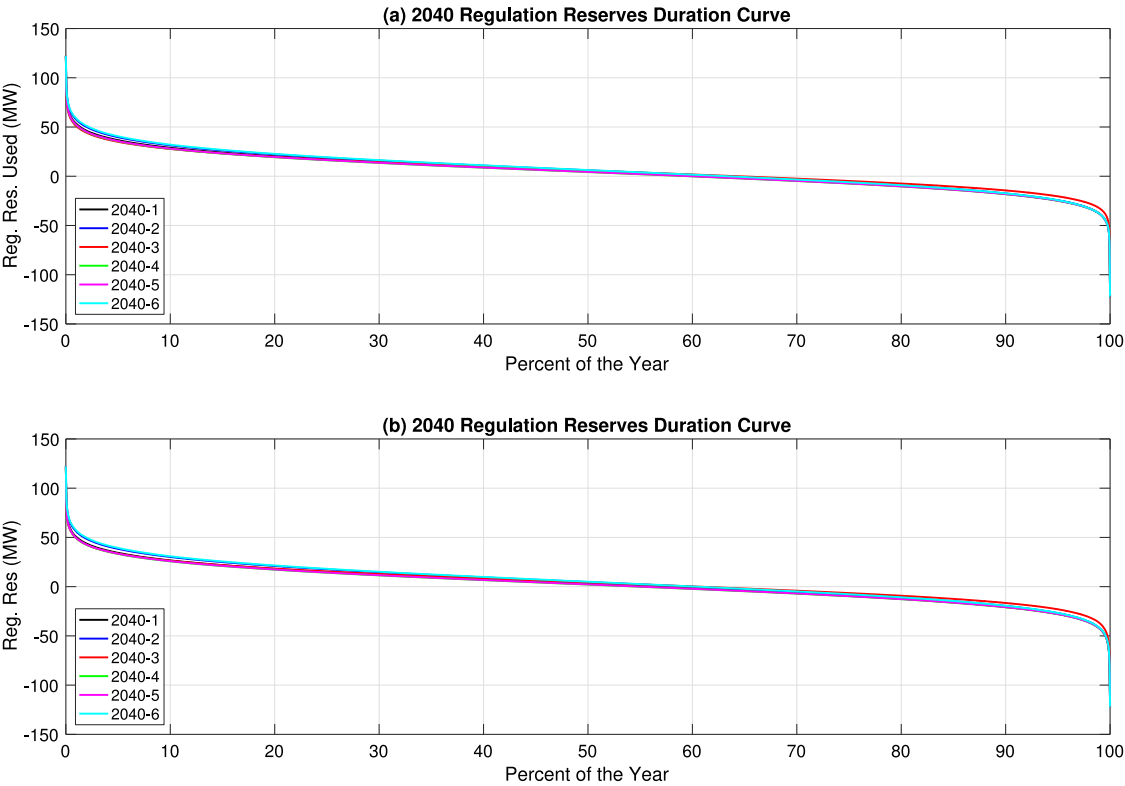


Fig. 12. Regulation duration curves for all six 2040 scenarios in both the flexible (above) and conventional (below) operating modes.

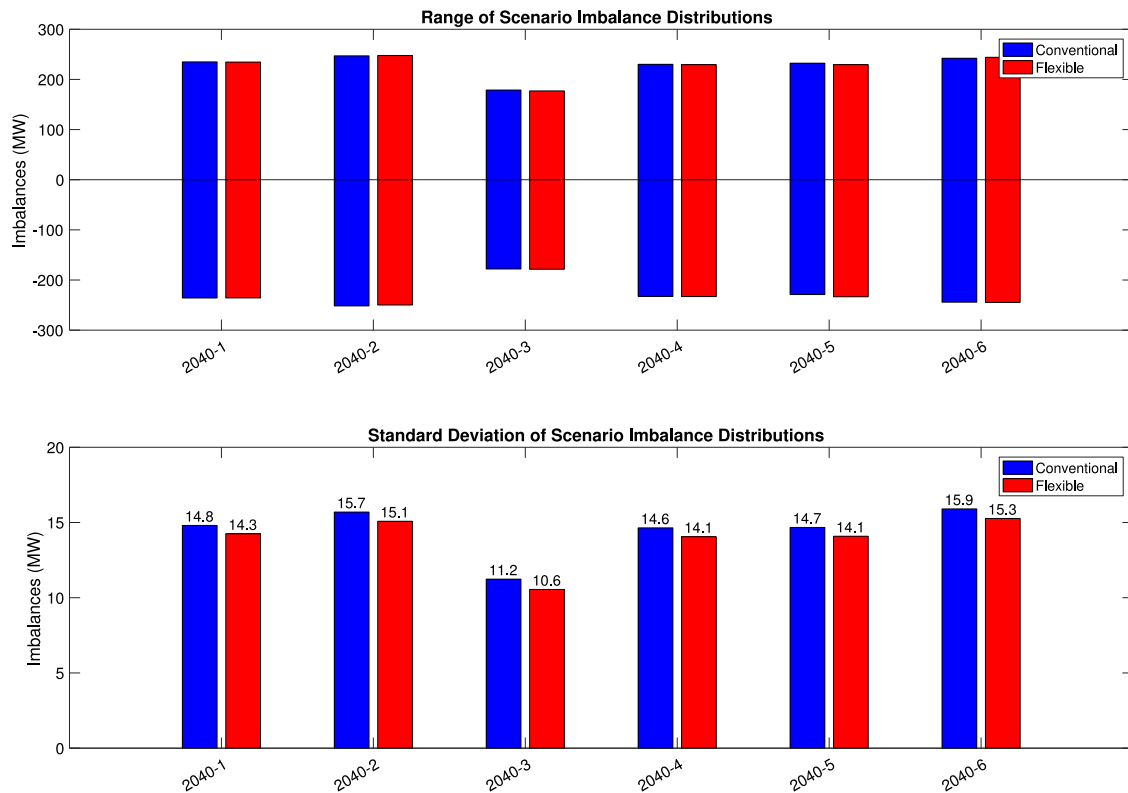


Fig. 13. Range (above) and standard deviation (below) statistics for all six 2050 scenarios in both the flexible (red) and conventional (blue) operation modes.

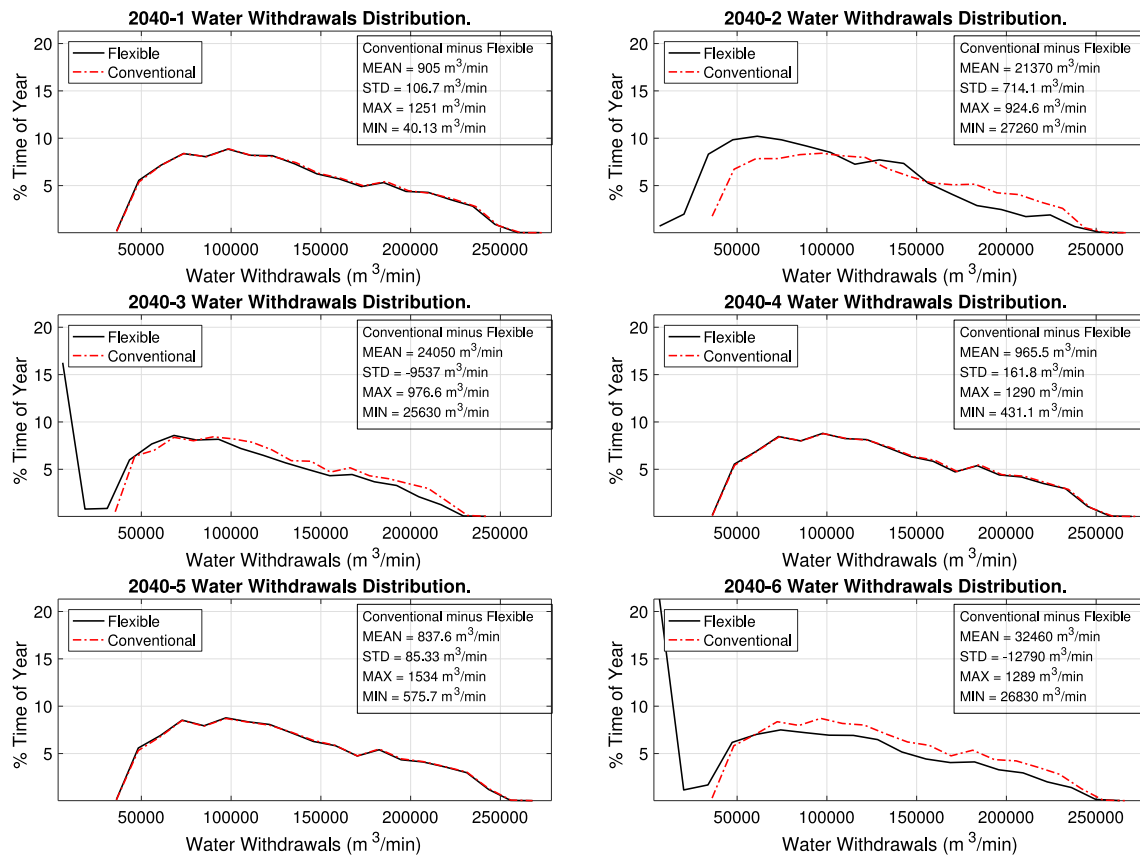


Fig. 14. Distributions of water withdrawals for all six 2040 scenarios in both the flexible and conventional operating modes.

Table 7

Change in range and standard deviations of imbalances (flexible minus conventional) for all six 2040 scenarios.

Change in Imbalance	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Max (MW)	-0.384	0.597	-1.767	-0.682	-2.911	1.902
% Max	-0.164	0.241	-0.998	-0.297	-1.269	0.779
Min (MW)	0.118	1.831	-0.598	-0.363	-4.405	-0.462
% Min	-0.050	-0.733	0.335	0.156	1.887	0.189
Std. (MW)	-0.552	-0.611	-0.684	-0.589	-0.584	-0.634
% Std.	-3.874	-4.052	-6.484	-4.188	-4.147	-4.155

4.2.1. Water withdrawals

Fig. 14 shows the water withdrawal distributions for the flexible and conventional operating modes. Flexible operation results in significantly lower withdrawals compared to conventional operation because the flexible energy–water resources are able to offset the use of thermo-electric power plants in favor of VREs. This phenomena is seen in how the flexible withdrawal distributions are shifted left towards zero. The associated water withdrawal statistics are summarized in Table 8 indicating improvements in mean withdrawals of up to 25.58%. These improvements are most pronounced in Scenarios 2040-2/3/6 with high penetrations of VREs. Indeed, the integration of several percent (on a capacity basis) of flexible energy–water resources as shown in Table 2, serves to reduce water withdrawals by many multiples of that percentage. Such a phenomena can potentially appear in any scenario where VRE curtailment serves as a major lever of balancing control. Nevertheless, the flexible operation of energy–water resources reduces water withdrawals across all six scenarios.

4.2.2. Water consumption

Electric power system water consumption occurs through the evaporative losses from cooling towers in recirculating cooling systems. Fig. 15 shows the water consumption distribution for both the conventional and flexible operating modes. While the effect is not large, the flexible mode of operation shifts the distribution slightly towards the zero mark. Specifically, flexible operation consumes 1.07–4.51% less water than the conventional operation across all six scenarios, as shown in Table 9. This relatively small percentage nevertheless accounts for $258 \times 10^3 \text{ m}^3$ of water saved every year. Scenarios 2040-3 and 2040-6 have the least savings. Due to high penetrations of VREs, these scenarios require faster ramping generation which mostly comes from fast-ramping natural gas units with recirculating cooling systems. In short, the water saving effect of integrating VREs is a diminished to a certain extent by the need for operating reserves from water-consuming but flexible NGCC plants. If demand side resources (from electricity consumed by water and wastewater treatment plants or otherwise) played a large balancing role, then the water saving role of integrating VREs would be more pronounced.

Table 8

Change in water withdrawals statistics (conventional minus flexible) for all six 2040 scenarios.

$\Delta \text{H}_2\text{O}$ Withdrawals	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Mean (m^3/min)	905.0 (0.70%)	21370.0 (17.29%)	24050.0 (20.59%)	965.5 (0.74%)	837.6 (0.65%)	32460.0 (25.58%)
STD (m^3/min)	106.7 (0.20%)	714.1 (1.35%)	-9537.0 (-19.92%)	161.8 (0.31%)	85.3 (0.16%)	-12790.0 (-24.40%)
Max (m^3/min)	1251.0 (0.45%)	924.6 (0.34%)	976.6 (0.39%)	1290.0 (0.47%)	1534.0 (0.56%)	1289.0 (0.47%)
Min (m^3/min)	40.1 (0.11%)	27260.0 (88.22%)	25630.0 (75.82%)	431.1 (1.17%)	575.7 (1.54%)	26830.0 (75.99%)
Total ($\text{m}^3/\text{min} \times 10^6$)	475.7	11230.0	12640.0	507.5	440.2	18090.0
Percent change (%)	0.70	17.29	20.59	0.74	0.65	25.58

4.2.3. CO₂ emissions

Finally, as shown in Fig. 16, the overall CO₂ emissions are significantly reduced through flexible operation. It reduces the overall CO₂ emissions by 2.10%–3.46%, as shown in Table 10. The mean, max, and standard deviation of emissions are all improved. This CO₂ emissions reduction occurs because flexible energy–water resources 1.) eliminate the need for some generation through reduced electricity consumption, 2.) enable greater VRE generation through a reduction in curtailment and 3.) displace fossil-fueled conventional generation.

4.3. Economic performance of coordinated energy–water operation

The economic performance of coordinated energy–water operation is assessed in terms of the day-ahead and real-time production costs.

4.3.1. Day-ahead energy market production costs

Fig. 17 shows flexible operation reduced the total production cost in the day-ahead energy market for all 2040 scenarios. Table 11 summarizes the associated statistics. Flexible operation reduced total production costs by 29.3–68.09M\$ or between 1.22–1.76%. As illustrated in Fig. 17, Scenarios 2040-2/3/6 have much lower day-ahead production costs due to a high penetration of VREs. In contrast, scenarios 2040-1/4/5 have significantly higher costs as they are forced to commit expensive thermal power plants. In short, the day-ahead energy market production costs are lower because the flexible mode of operation represents an optimization program that is less constrained than the program associated with the conventional mode of operation.

4.3.2. Real-time energy market production costs

Fig. 18 illustrates the total real-time energy market production cost for all six scenarios. Similar to the day-ahead energy market, Scenarios 2040-1/4/5 have significantly higher production costs as they are forced to dispatch more expensive thermal power plants. Meanwhile, Scenarios 2040-2/3/6 have lower real-time energy market production costs due to a greater utilization of renewable energy. As detailed in Table 12, flexible operation reduces the average real-time market production costs by 2.46%–3.70% (or 19.58–70.83M\$) across all six scenarios.

4.4. Discussion

This study provides results for six 2040 scenarios for the New England energy–water nexus. It illustrates significant improvements in balancing performance of the electricity system that arise from two key methodological differences from [53] namely; (1) treating energy–water resources as flexible, and (2) allowing solar and wind to provide load-following, and ramping reserves. These two changes in how resources are treated within electricity markets amount to significant improvements in overall minimum load-following and ramping reserves that ensure the system is able to better respond to variability in the net load. Compared to the renewable energy integration study in [53], the approach in this work results in overall lower curtailment

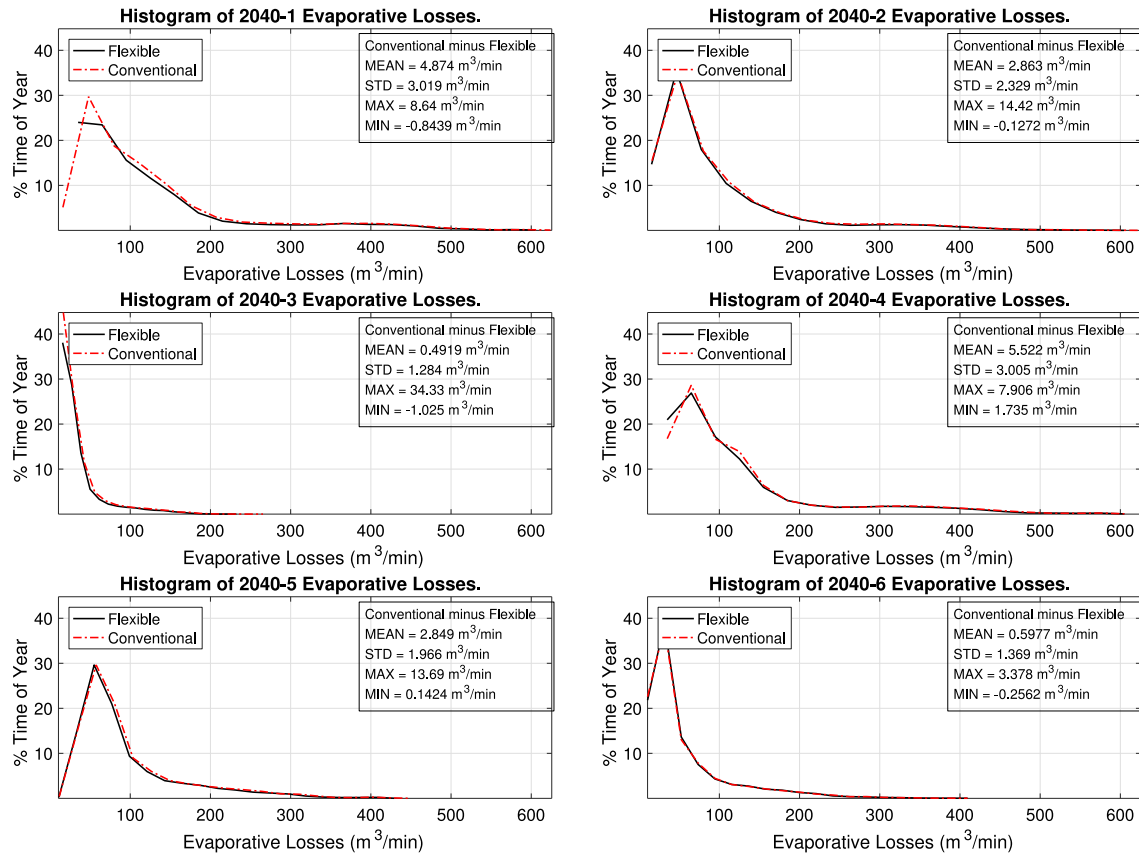


Fig. 15. Distributions of water consumption for all six 2040 scenarios in both the flexible and conventional operating modes.

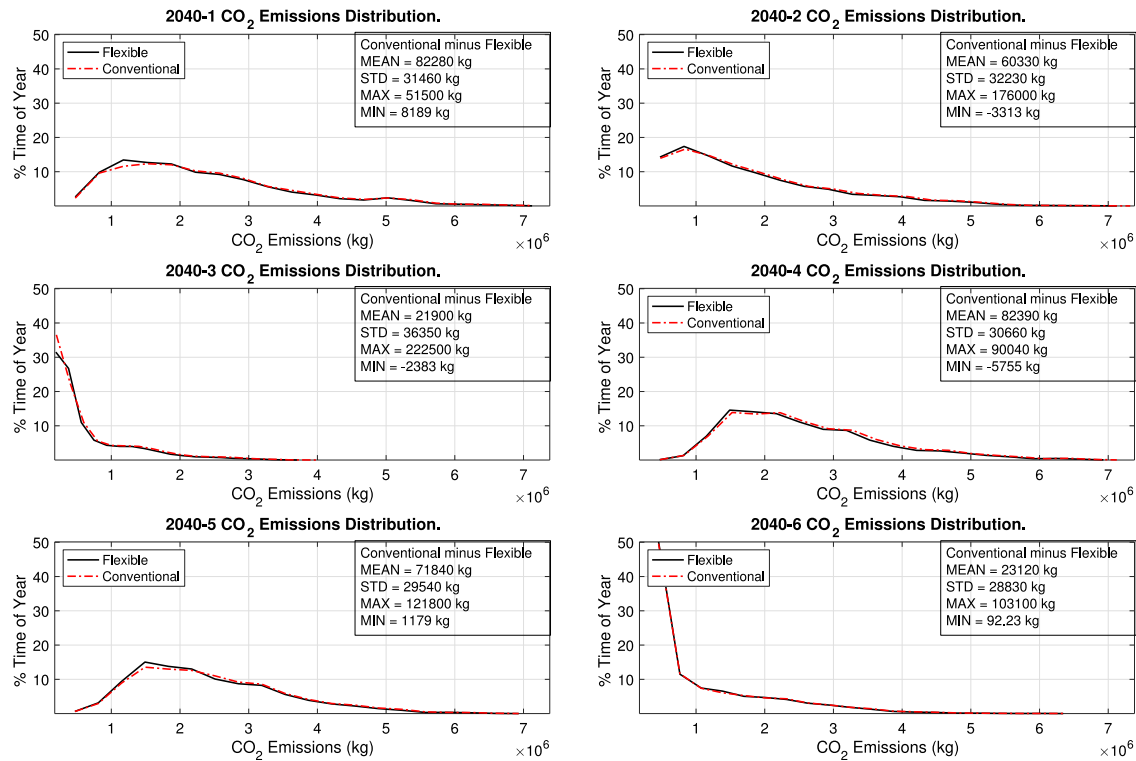


Fig. 16. Distributions of CO₂ emissions for all six 2040 scenarios in both the flexible and conventional operating modes.

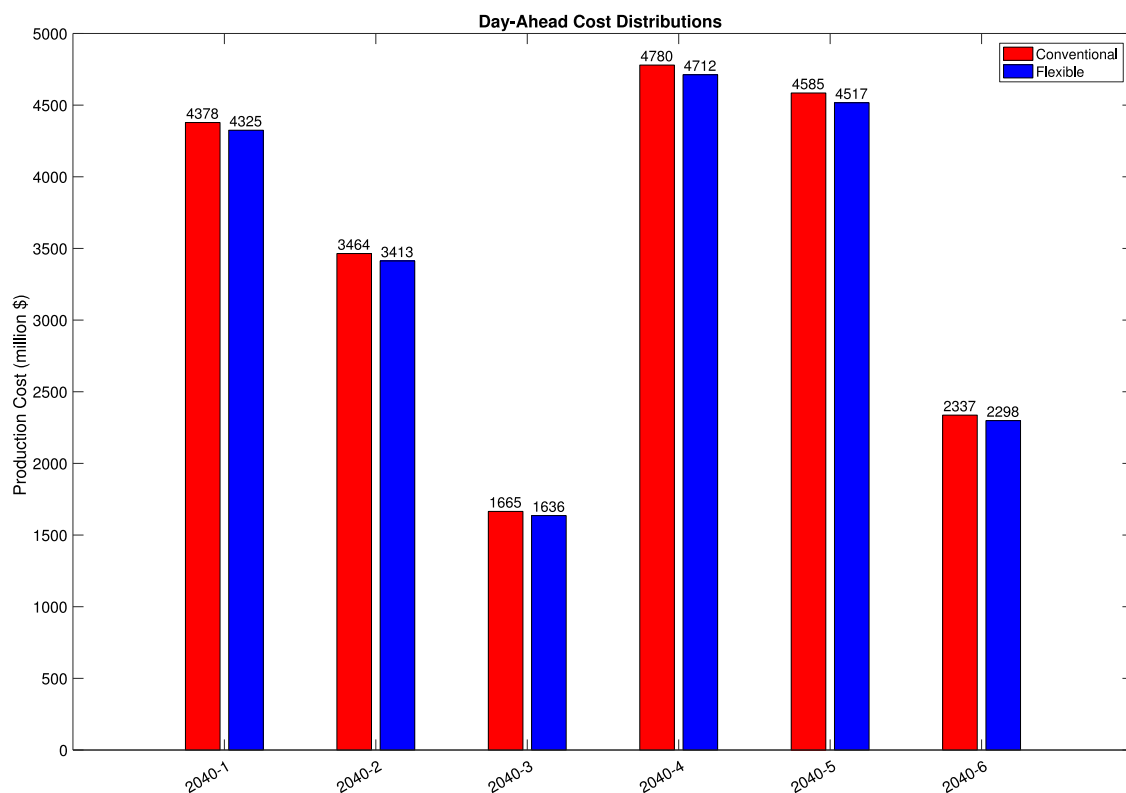


Fig. 17. Total production cost in the day-ahead energy market for all 2040 scenarios in both the flexible and conventional operating modes.

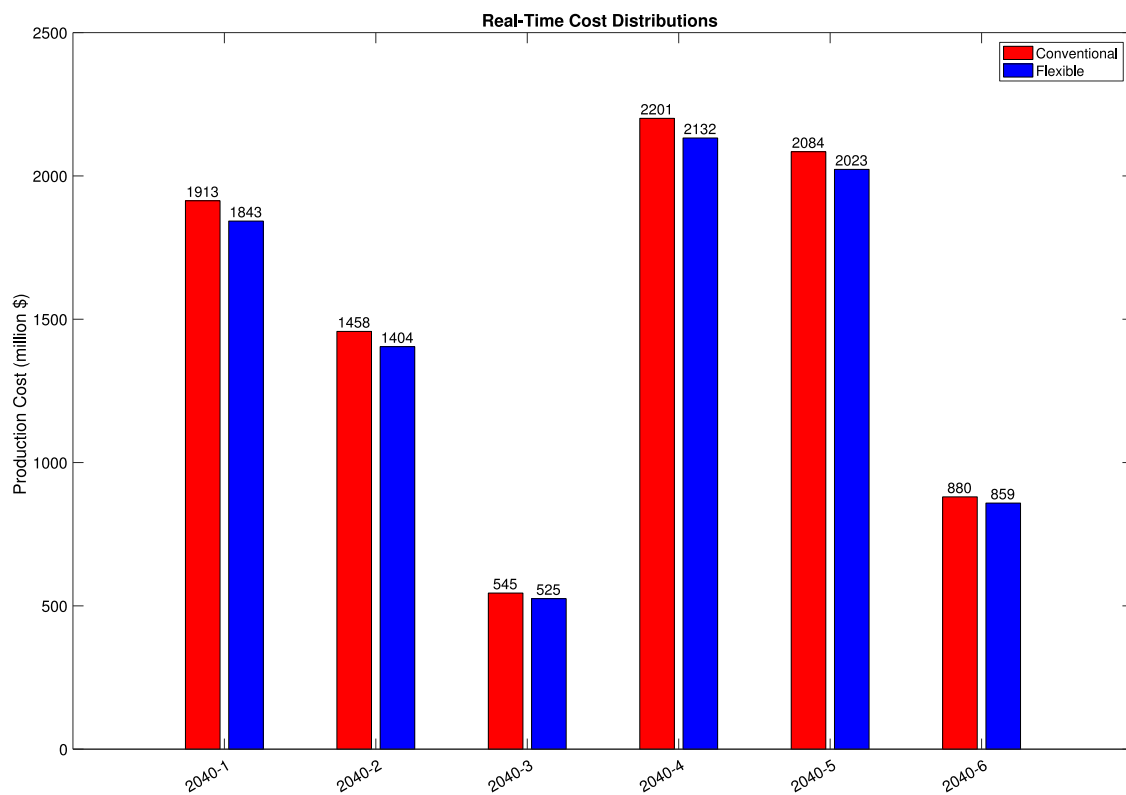


Fig. 18. A comparison of the real-time production costs for flexible and conventional operation.

Table 9Change in evaporative loss statistics (conventional *minus* flexible) for all six 2040 scenarios.

Δ Evap Losses	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Mean (m ³ /min)	2.67 (3.96%)	1.63 (3.11%)	0.30 (1.44%)	3.37 (5.03%)	1.51 (2.84%)	0.31 (1.03%)
STD (m ³ /min)	1.10 (2.77%)	1.05 (2.97%)	0.74 (5.58%)	1.23 (3.33%)	0.61 (2.61%)	0.68 (3.05%)
Max (m ³ /min)	5.71 (2.45%)	3.42 (1.44%)	6.40 (6.02%)	−0.00 (−0.00%)	1.80 (1.11%)	0.07 (0.04%)
Min (m ³ /min)	−0.62 (−3.50%)	−0.00 (−0.00%)	−0.13 (−1.65%)	0.47 (2.56%)	−0.12 (−0.83%)	−0.06 (−0.52%)
Total (m ³ × 10 ³)	1402	859	158	1769	794	165
Percent change (%)	4.12	3.21	1.46	5.30	2.92	1.03

Table 10Change in CO₂ emissions statistics (flexible *minus* conventional) for all six 2040 scenarios.

Δ CO ₂ Emissions	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Mean (kg)	82 280 (3.46%)	60 330 (3.28%)	21 900 (3.17%)	82 390 (3.11%)	71 840 (2.90%)	23 120 (2.10%)
STD (kg)	31460.0 (2.44%)	32230.0 (2.66%)	36350.0 (5.75%)	30660.0 (2.69%)	29540.0 (2.71%)	28 830 (2.96%)
Max (kg)	51 500 (0.71%)	176 000 (2.38%)	222 500 (5.54%)	90 040 (1.26%)	121 800 (1.72%)	103 100 (1.59%)
Min (kg)	8189.00 (2.07%)	−3313.00 (−1.08%)	−2383.00 (−1.35%)	−5755.00 (−1.14%)	1179.00 (0.31%)	92.23 (0.03%)
Total (kg × 10 ⁶)	43 240	31 710	11 510	43 300	37 760	12 150
Percent change (%)	3.46	3.28	3.17	3.11	2.90	2.10

Table 11Change in day-ahead energy market production cost statistics (flexible *minus* conventional) for all six 2040 scenarios.

Δ Day-Ahead Costs	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Mean (\$/h)	6115.1 (1.22%)	5909.4 (1.49%)	3345.2 (1.76%)	7712.7 (1.41%)	7773.1 (1.49%)	4388.1 (1.64%)
STD (\$/h)	4859.0 (2.09%)	4355.7 (1.89%)	5336.3 (3.89%)	5327.3 (2.62%)	6160.9 (3.05%)	6095.2 (3.02%)
Max (\$/h)	−16071.5 (−0.95%)	38820.1 (2.65%)	66093.4 (5.44%)	−76701.8 (−4.56%)	15683.0 (0.83%)	476535.0 (23.20%)
Min (\$/h)	19290.1 (18.95%)	−2738.0 (−3.14%)	15922.7 (19.18%)	−706.4 (−0.45%)	−419.0 (−0.36%)	−10860.0 (−12.17%)
Total (million \$)	53.57	51.77	29.30	67.56	68.09	38.44
% Reduction	1.22	1.49	1.76	1.41	1.49	1.64

Table 12A summary of the real-time production cost statistics (flexible *minus* conventional).

Δ Real-Time Cost	2040-1	2040-2	2040-3	2040-4	2040-5	2040-6
Mean (\$/min)	1347.5 (3.70%)	1013.5 (3.65%)	372.5 (3.59%)	1304.9 (3.12%)	1173.1 (2.96%)	412.5 (2.46%)
STD (\$/min)	493.5 (2.31%)	533.2 (2.62%)	553.8 (5.21%)	497.8 (2.58%)	545.8 (2.90%)	536.9 (3.30%)
Max (\$/min)	895.8 (0.58%)	3976.9 (2.69%)	385.2 (0.36%)	3163.4 (2.02%)	−5845.8 (−3.41%)	40662.3 (23.52%)
Min (\$/min)	88.4 (2.76%)	75.5 (3.45%)	−0.0 (−0.00%)	65.3 (0.98%)	−0.0 (−0.00%)	157.3 (3.78%)
Total (million \$)	70.83	53.27	19.58	68.58	61.66	21.7
% Reduction	3.70	3.65	3.59	3.12	2.96	2.46

levels and therefore, greater utilization of VREs. While these two studies cannot be compared one-to-one given that they used different data sets, the greater utilization of renewables in this work shows the significant value of flexible energy–water resources. The simulation results also show that flexible operation improves environmental performance of the electricity grid by reducing water withdrawals and consumption, and total CO₂ emissions of the system. Greater utilization of VREs in

turn decreases the day-ahead and real-time market production costs. These results indicate that the study of renewable energy integration must leverage the value of demand-side resources (such as demand-side energy–water resources) in order to sustain higher penetrations of VREs. Furthermore, it shows that there is significant economic, environmental as well as reliability value in jointly studying/operating

Table 13

Balanced Sustainability Scorecard: The range of *improvements* caused by coordinated flexible operation of the EWN.

Balancing Performance	
Average Load Following Reserves	1.24–12.66%
Average Ramping Reserves	5.28–18.35%
Percent Time Curtailed	2.67–10.90%
Percent Time Exhausted Regulation Reserves	0%
Std. Dev. of Imbalances	3.874–6.484%
Environmental Performance	
Total Water Withdrawals	0.65–25.58%
Total Water Consumption	1.03–5.30%
Total CO ₂ Emissions	2.10–3.46%
Economic Performance	
Total Day-Ahead Energy Market Production Cost	29.30–68.09M\$
Total Real-Time Energy Market Production Cost	19.58–70.83M\$

interdependent infrastructures such as the energy and water supply systems.

5. Conclusion

This work has used a novel enterprise control assessment methodology to study the EWN for the ISO New England System. Six scenarios were studied representing plausible electric power capacity mixes in 2040. The study specifically sought to understand the impact of flexible coordinated operation of energy–water resources on the holistic behavior of these six scenarios. In short, the flexible operation energy–water resources demonstrated truly “sustainable synergies” with respect to balancing, environmental, and economic performance. Table 13 summarizes the most important results of the study in a balanced sustainability scorecard and highlights the synergistic improvements caused by flexible coordinated operation of the EWN. Flexible operation of energy–water resources results in up to 12.66% improvements in load-following reserves, up to 18.35% increase in available ramping reserves and up to 10.90% reduction in the total time that curtailment of VREs occurs. Additionally, the environmental performance of the system is significantly improved with flexible operation resulting in up to 25.58% reductions in water withdrawals, 5.30% reductions in water consumption and up to 3.46% reductions in carbon dioxide emissions. These results show that as VRE resources become an ever-important part of the electric power system landscape, so too must the electric power system evolve to engage energy–water resources as control levers. In some cases, such resources – like hydro-power plants – are mainstays of traditional operation. In other cases, particularly water utility electric loads, these resources will have to evolve their operation to become true electric power grid participants.

CRedit authorship contribution statement

Steffi Olesi Muhanji: Data curation, Writing - original draft, Software, Validation, Visualization, Investigation, Formal analysis. **Clayton Barrows:** Funding acquisition, Data curation, Writing - review & editing. **Jordan Macknick:** Funding acquisition, Data curation, Writing - review & editing. **Amro M. Farid:** Funding acquisition, Project administration, Conceptualization, Methodology, Supervision, Writing - review & editing.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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