Supplementary Material of "An Evolutionary Optimization-Learning Hybrid Algorithm for Energy Resource Management"

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I. THE MATHEMATICAL MODEL OF RISK-BASED ENERGY			$C_{e,t}^{ESS}$	Cost of the eth ESS unit discharge at t (m.u./MWh).
RESOURCE MANAGEMENT			$C_{v,t}^{EV}$	Cost of the v th EV unit discharge at t (m.u./MWh).
Risk-based energy resource management (ERM) is modeled as a mixed-integer nonlinear programming model, using a multi-period optimization approach [1], [2]. The mathematical			$C_{m,s,t}^{EM}$	Price of the m th electricity market at period t
				in scenario s (m.u./MWh).
model for risk-based ERM is represented by the following			$UB_{d,s,t}$	Upper bound active power generation of the dth unit
notations and formulas, as detailed in [3].				at period t in scenario s (MW).
Indexs:			$LB_{d,s,t}$	Lower bound active power generation of the dth unit
		s Index of scenario.		at period t in scenario s (MW).
		t Index of period.	$P_{k,s,t}^{Load}$	Forecasted active power of the k th load at period t
Sets:				in scenario s (MW).
Ω_{I}	DG	Set of distributed generation (DG) units.	$P_{i,s,t}^{nd}$	Active power generation of the <i>i</i> th non-dispatchable
Ω_{L}^{d}		Subset of dispatchable DG units.		DG unit at period t in scenario s (MW).
L	JG	(nonrenewable energy generation units).	$LB_{i,t}^{DG_d}$	Minimum active power of the <i>i</i> th dispatchable DG unit
Ω_I^n	nd	Subset of non-dispatchable DG units	,	at period t (MW).
L	JG.	(renewable energy generation units).	$UB_{i,t}^{DG_d}$	Maximum active power of the <i>i</i> th dispatchable DG unit
		,	- 7-	at period t (MW).
Paramet	ters:		$LB_{o,t}^{EXT}$	Minimum active power of the ith EXT unit
				at period t (MW).
N_s	Total number of periods. Total number of DG units.		$UB_{o,t}^{EXT}$ $LB_{e,t}^{ESS}$	Maximum active power of the ith EXT unit
$N_t \ N_i$				at period t (MW).
N_k				Minimum active discharging power of the eth ESSs
N_o	Total number of demand response (DR) units. Total number of external supplier (EXT) units.		- /-	at period t (MW).
N_v	Total number of electric vehicles (EVs) units.		$UB_{e,t}^{ESS}$	Maximum active charging power of the eth ESSs
N_e	Total number of energy storage systems (ESSs) units.		,	at period t (MW).
N_m	Total number of electricity markets (EM) units.		$\eta_e^{ESS_c}$	Charging efficiency of the eth ESS.
N_r	Total number of non-supplied demand (NSD) units.		$\eta_e^{ESS_d}$	Discharging efficiency of the eth ESS.
$ ho_s$	Scenario probability.		$E_{e,s,t}^{ESS}$	Stored energy in the eth ESS at period t (MWh).
α	Confidence level.		$E_e^{ESS_{min}}$	Minimum stored energy in the eth ESS (MWh).
ξ	Penalty coefficient.		$E_e^{ESS_{max}}$	Maximum stored energy in the eth ESS (MWh).
		of the i th DG unit generation at t (m.u./MWh).	$LB_{v,t}^{EV}$	Minimum active discharging power of the vth EV
$C_{k,t}^{DR}$	Cost	of the k th DR unit generation at t (m.u./MWh).		at period t (MW).
$C_{i,t}^{DG}$ $C_{k,t}^{DR}$ $C_{o,t}^{EXT}$			$UB_{v,t}^{EV}$	Maximum active charging power of the vth EV
$C_{i,t}^{EAP}$	Cost	of the ith non-dispatchable DG unit excess		at period t (MW).
- ,-	avail	able power (EAP) at t (m.u./MWh).	$\eta_e^{EV_c}$	Charging efficiency of the eth EV.
$C_{r,t}^{NSD}$	Cost	of the rth NSD unit at t (m.u./MWh).	$\eta_e^{EV_d}$	Discharging efficiency of the eth EV.

 $E_{v,s,t}^{EV}$ Stored energy in the vth EV at period tin scenario s (MW).

 $E_{\nu}^{EV_{min}}$ Minimum stored energy in the vth EV (MWh).

 $E_{v}^{EV_{max}}$ Maximum stored energy in the vth EV (MWh).

 $UB_{k,t}^{DR}$ Maximum active load reduction power in the kth DR at period t (MW).

Maximum active power sell from the eth EM at period t (MW).

 $UB_{e,t}^{EM}$ Maximum active power buy from the eth EM at period t (MW).

 $E_{e,s,t}^{EV}$ Stored energy in the eth EV at period t in scenario s (MWh)

Binary variable of the ith DG unit in period t. $x_{o,t}^{EXT}$ Binary variable of the oth EXT unit in period t.

Model

$$\min F = f^{Ex} + \beta \cdot CVaR_{\alpha}(\Gamma) \tag{1}$$

s.t.

Variables:

Operational cost in scenario s (m.u.).

Operational income in scenario s (m.u.).

Penalty for bound violation in scenario s (m.u.).

 f_s Net operational cost in scenario s (m.u.).

 f^{Ex} Expected cost (m.u.).

 $P_{i,t}^{DG_d}$ Active power generation of the ith dispatchable DG unit at period t in scenario s (MW).

 $P_{i,s,t}^{DG_{nd}}$ Active power generation of the *i*th non-dispatchable DG unit at period t in scenario s (MW).

 $P_{k,t}^{DR}$ Active power generation of the kth DR unit at period t (MW).

Active power generation of the kth EXT unit at period t (MW).

Active power generation of the vth EV unit at period t (MW).

 $P_{i,s,t}^{EAP}$ Active power generation of the *i*th EAP unit at period t in scenario s (MW).

 $P_{r,s,t}^{NSD}$ Active power generation of the rth NSD unit at period t in scenario s (MW).

Active power generation of the eth ESS unit at period t (MW).

 $ESS_{e,t}^{cost}$ Discharging cost of the eth ESS unit at period t (m.u.).

 $EV_{v,t}^{cost}$ Discharging cost of the vth ESS unit at period t (m.u.).

 $P_{m,t}^{EM}$ Active power traded in the mth EM unit at period t (MW).

 $P_{d,s,t}$ Active power generation of the dth unit at period tin scenario s (MW).

 $E_{e,s,t}^{ESS}$ Stored energy in the eth ESS at period t in scenario s (MWh)

$$f_s = f_s^{Co} - f_s^{In} + f_s^{Pe} (2)$$

$$\Gamma = \bigcup_{s=1}^{N_s} f_s \tag{3}$$

$$f^{Ex} = \sum_{s=1}^{N_s} (\rho_s \cdot f_s) \tag{4}$$

$$VaR_{\alpha}(\Gamma) = \sigma(\Gamma) \cdot z\text{-}score(\alpha)$$
 (5)

$$\varphi_s = \begin{cases} f_s - f^{Ex} - VaR_{\alpha}(\Gamma) & , if \ f_s \ge f^{Ex} + VaR_{\alpha}(\Gamma) \\ 0 & , otherwise \end{cases}$$
 (6)

$$CVaR_{\alpha}(\Gamma) = VaR_{\alpha}(\Gamma) + \frac{1}{1-\alpha} \sum_{s=1}^{N_s} \rho_s \cdot \varphi_s$$
 (7)

$$f_{s}^{Co} = \sum_{t=1}^{N_{t}} \begin{bmatrix} \sum_{i \in \Omega_{DG}^{d}} P_{i,t}^{DG_{d}} \cdot C_{i,t}^{DG} + \sum_{i \in \Omega_{DG}^{d}} P_{i,s,t}^{DG_{nd}} \cdot C_{i,t}^{DG} + \\ \sum_{i \in \Omega_{DG}^{d}} P_{k,t}^{DR} \cdot C_{k,t}^{DR} + \sum_{o=1}^{N_{o}} P_{o,t}^{EXT} \cdot C_{o,t}^{EXT} + \\ \sum_{k=1}^{N_{e}} ESS_{e,t}^{cost} + \sum_{v=1}^{N_{v}} EV_{v,s,t}^{cost} + \\ \sum_{i \in \Omega_{DG}^{nd}} P_{i,s,t}^{EAP} \cdot C_{i,t}^{EAP} + \sum_{r=1}^{N_{r}} P_{r,s,t}^{NSD} \cdot C_{r,t}^{NSD} \end{bmatrix}, \forall s \quad (8)$$

$$ESS_{e,t}^{cost} = \begin{cases} \sum_{e=1}^{N_e} P_{e,t}^{ESS} \cdot C_{e,t}^{ESS} & , if P_{e,t}^{ESS} \le 0\\ 0 & , otherwise \end{cases}$$
(9)

$$EV_{v,t}^{cost} = \begin{cases} \sum_{v=1}^{N_v} P_{v,t}^{EV} \cdot C_{v,t}^{EV} &, if P_{v,t}^{EV} \le 0\\ 0 &, otherwise \end{cases}$$
 (10)

$$f_s^{In} = \sum_{t=1}^{N_t} \sum_{m=1}^{N_m} P_{m,s,t}^{EM} \cdot C_{m,s,t}^{EM}, \forall s$$
 (11)

$$f_s^{Pe} = \xi \cdot \sum_{t=1}^{N_t} \sum_{d=1}^{N_D} \left(\max(0, P_{d,s,t} - UB_{d,s,t}) + \max(0, LB_{d,s,t} - P_{d,s,t}) \right), \forall s$$
(12)

$$N_D = N_i + N_k + N_o + N_v + N_e + N_m + N_r \tag{13}$$

$$LB_{i,t}^{DG_d} \cdot x_{i,t}^{DG} \le P_{i,t}^{DG_d} \le UB_{i,t}^{DG_d} \cdot x_{i,t}^{DG}, \forall i \in \Omega_{DG}^d, t$$
(14)

$$\begin{split} P_{i,s,t}^{DG_{nd}} &= P_{i,s,t}^{nd} \cdot x_{i,t}^{DG}, \forall i \in \Omega_{DG}^{nd}, t, s \\ LB_{o,t}^{EXT} \cdot x_{o,t}^{EXT} &\leq P_{o,t}^{EXT} \leq UB_{o,t}^{EXT} \cdot x_{o,t}^{EXT}, \forall o, t \end{split} \tag{15}$$

$$LB_{o,t}^{EXT} \cdot x_{o,t}^{EXT} < P_{o,t}^{EXT} < UB_{o,t}^{EXT} \cdot x_{o,t}^{EXT}, \forall o, t$$
 (16)

$$\begin{split} - LB_{e,t}^{ESS} &\leq P_{e,t}^{ESS} \leq UB_{e,t}^{ESS}, \forall e, t \\ E_{e,s,t}^{ESS} &= E_{e,s,t-1}^{ESS} + \eta_e^{ESS_c} \cdot P_{e,t}^{ESS} - \frac{1}{n^{ESS_d}} \cdot P_{e,t}^{ESS}, \forall e, t, s \end{split} \tag{17}$$

$$E_{e,s,t}^{ESS} = E_{e,s,t-1}^{ESS} + \eta_e^{ESS_c} \cdot P_{e,t}^{ESS} - \frac{1}{n_e^{ESS_d}} \cdot P_{e,t}^{ESS}, \forall e, t, s$$
 (18)

$$\begin{bmatrix}
\sum_{i \in \Omega_{DG}^{d}} P_{i,t}^{DG_{d}} + \sum_{o=1}^{N_{o}} P_{o,t}^{EXT} + \\
\sum_{i \in \Omega_{DG}^{nd}} (P_{i,s,t}^{DG_{nd}} - P_{i,s,t}^{EAP}) + \\
\sum_{i \in \Omega_{DG}^{nd}} (P_{k,t}^{DR} - P_{k,s,t}^{Load}) + \\
\sum_{k=1}^{N_{e}} (P_{k,t}^{ESS} + \sum_{v=1}^{N_{v}} P_{v,t}^{EV} + \\
\sum_{e=1}^{N_{e}} P_{r,s,t}^{ESS} + \sum_{m=1}^{N_{m}} P_{m,t}^{EM}
\end{bmatrix} = 0, \forall s \tag{19}$$

$$E_{e,t}^{ESS_{min}} \leq E_{e,s,t}^{ESS} \leq E_{e}^{ESS_{max}}, \forall e, t, s$$

$$-LB_{v,t}^{EV} \leq P_{v,t}^{EV} \leq UB_{v,t}^{EV}, \forall v, t$$
(20)

$$-LB_{v,t}^{EV} \le P_{v,t}^{EV} \le UB_{v,t}^{EV}, \forall v, t \tag{21}$$

$$E_v^{EV_{min}} \le E_{v,s,t}^{EV} \le E_v^{EV_{max}}, \forall v, t, s$$

$$(21)$$

$$E_{v,s,t}^{EV} = E_{v,s,t-1}^{EV} + \eta_v^{EV_c} \cdot P_{v,t}^{EV} - \frac{1}{\eta_v^{EV_d}} \cdot P_{v,t}^{EV}, \forall v, t, s \quad (23)$$

$$P_{k,t}^{DR} \le UB_{k,t}^{DR}, \forall k, t \tag{24}$$

$$-LB_{m,t}^{EM} \le P_{m,t}^{EM} \le UB_{m,t}^{EM}, \forall m, t$$
 (25)

In the risk-based ERM formulation, the task is to determine the optimal values of N_i for DG, N_k for DR, N_o for EXT, N_v for EVs, N_e for ESS, N_m for EM, and N_r NSD in each period, with the aim of minimizing the evaluation metrics Fin (1). This metric consists of two components: the expected operation costs f^{Ex} and $\beta \cdot CVaR_{\alpha}(\Gamma)$. For each scenario s, a cost-benefit analysis is executed. The difference between the operational costs f_s^{Co} and the income f_s^{In} , plus the penalty for bound violations f_s^{Pe} constitutes amended operation costs f_s . Then, the set Γ is constructed by (3). In (4), the expected operation cost f^{Ex} is calculated, taking into account the probability of various scenarios. In (5), VaR_{α} at a $\alpha = 95\%$ is calculated by the standard deviation of the amended operation costs $\sigma(\Gamma)$ and z-score obtained from MATLAB's norminv() function. (6) shows that if f_s exceeds $f^{Ex} + VaR_{\alpha}(\Gamma)$, φ indicates an extreme scenario (with a probability of $1 - \alpha$); otherwise, $\varphi = 0$. In (7), $CVaR_{\alpha}$ is calculated by adding VaR_{α} and an extra penalty to the extreme scenarios.

The formulation incorporates several key constraints. The generation limits of dispatchable DG are captured by (14). (15) quantifies the non-dispatchable generation, encompassing resources such as wind and photovoltaic power, under varying scenarios. External supply resources are constrained by (16), while (17) shows discharge limits on energy storage systems. The battery balance constraint, which governs the remaining electricity level, is represented by (18). (19) showcases the active power balance constraint, ensuring the equilibrium between generation and demand. The storage system's battery capacity is restricted by (20), (21)-(22) delineates the battery balance and capacity limits for EVs. Demand response programs are accounted for through the maximum load reduction constraint in (24). Finally, (25) limits the amount of electricity that can be traded with the EM, ensuring that buy or sell adheres to specified bounds.

REFERENCES

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