A Scalable Formulation for Look-ahead Security-Constrained Optimal Power Flow

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Abstract-We consider the look-ahead security-constrained optimal power flow (LASCOPF) problem under transmission line and generator contingencies. We first formulate LASCOPF under the N-1 contingency criterion (LASCOPF₁) using the DC power flow model. We observe that the number of decision variables in the comprehensive formulation increases quadratically with the number of look-ahead intervals, T, making the problem infeasible to solve for large T. To overcome this, we propose the reduced LASCOPF problem (LASCOPF-r1) in which the number of decision variables increases only linearly with T. Thereafter, we prove that, barring borderline cases, if LASCOPF₁ is feasible then the optimal solutions of LASCOPF₁ and LASCOPF-r₁ are equivalent. We then extend our results to the N - k contingency criterion (LASCOPF- ru_k) for any collection of k contingencies, and we prove that the ordering of the contingencies does not affect the optimal solution. We then illustrate LASCOPF1 on a simple 2-bus 2-generator system. We show the numerical benefits of the proposed LASCOPF-r1 formulation on the IEEE 118-bus, the IEEE 300-bus and the 2383-bus Polish systems.

Index Terms—Look-ahead, Optimal power flow, N-k contingency criterion

I. NOTATION

Sets

- \mathcal{N} buses
- \mathcal{G} generators
- \mathcal{L} transmission lines
- \mathcal{L}' transmission lines that do not partition the system
- $\mathcal{C}_{\mathcal{L}}$ transmission line contingencies, $\mathcal{C}_{\mathcal{L}} \subseteq \mathcal{L}'$
- $\mathcal{C}_{\mathcal{G}}$ generator contingencies, $\mathcal{C}_{\mathcal{G}} \subseteq \mathcal{G}$

Indices¹

- t dispatch interval, $t \in \mathbb{N}$
- u interval in which contingency begins (viz. contingency interval), $u \in \mathbb{N}$
- n bus, $n \in \mathcal{N}$
- g generator, $g \in \mathcal{G}$
- *l* transmission line, $l \in \mathcal{L}$
- b transmission line contingency, $b \in C_{\mathcal{L}}$
- c generator contingency, $c \in C_{\mathcal{G}}$

Parameters

T	length of the planning horizon
$C_{g,t}(x)$	generation cost for one dispatch interval
$D_{n,t}^{\{0\}},\!D_{n,t}^{\{c\}}$	demand in the normal state, and under
$\underline{P}_q, \overline{P}_g$	active power generation limits
A_{ng}	generator location; 1 if g is at n , 0 otherwise
$\overline{K}_l^{\{0\}}, \overline{K}_l^{\{b\}}$	transmission line capacity for the normal
	state and under contingency $b \in C_{\mathcal{L}}$
$\underline{R}_{g}, \overline{R}_{g}$	ramping limits in one dispatch interval
	contingency $c \in C_{\mathcal{G}}$, respectively

 ${}^{1}\mathbb{N}$ represents the set of natural numbers.



Fig. 1. Illustration of LAOPF computed over a planning horizon of four intervals, executed in every dispatch interval. The solution for the next dispatch interval is implemented (dark), the rest is advisory (shaded).

$$H_{ln}^{\{0\}}, H_{ln}^{\{b\}}$$

power transfer distribution factor of line lfor injection at bus n for the normal state, and under contingency $b \in C_{\mathcal{L}}$, respectively

Decision variables

 $p_{g,t}^{(0)}, p_{g,t}^{(c,u)}$

generator dispatch in the normal state, and under contingency $c \in C_{\mathcal{G}}$, respectively

II. INTRODUCTION

The optimal power flow (OPF) problem aims at minimizing generation cost over a single dispatch interval, and has been fundamental to power system operation ever since its inception [1]. It has, however, been recognised in recent years that due to increasing amounts of intermittent generation capacity [2] single interval OPF has to be extended to take into account the ramping capability of generators and the dependence between subsequent dispatch intervals [3].

This dependence is accounted for by the Look-ahead OPF (LAOPF) problem [4]. LAOPF minimises the total generation cost over multiple consecutive dispatch intervals (called the planning horizon), taking into account generator ramping constraints and demand forecasts, and is particularly useful in case of large anticipated changes in net demand [5]. The solution obtained for the next dispatch interval is used for dispatch instructions and locational marginal pricing, while that for subsequent intervals is advisory [6]. The algorithm is executed again based on the updated predicted demand after the dispatch interval according to the principle of receding horizon control [7]. This look-ahead and receding horizon operation is illustrated in Fig. 1, where for expositional clarity, we consider that all intervals in the planning horizon are of equal duration. Although several LAOPF implementations include unit commitment, our discussion will be limited to dispatch and not commitment.

The importance of the dependence between subsequent dispatch intervals has been confirmed by industry, and Independent System Operators (ISOs) have recently shown increased interest in LAOPF for real-time operation of power systems with five minute dispatch intervals, often called multi-interval real-time markets. LAOPF is already implemented by the New York ISO (NYISO) [8], the California ISO (CAISO) [9], the Midcontinent ISO (MISO) [10], Ontario's Independent Electricity System Operator (IESO) [11], and the PJM Interconnection [12] while the Electric Reliability Council of Texas (ERCOT) has proposed the approach [13]. It is also being considered in Australia [14]. Several ISOs also consider flexible ramping products (FRPs) as a proxy to LAOPF [15]. FRPs are akin to reserve capacity in that they ensure that there is sufficient ramping capability in the system to manage large deviations in demand between consecutive dispatch intervals. When there is sufficient ramping capability, ISOs can safely implement single interval OPF, which has a computational advantage over LAOPF, but a potentially suboptimal outcome. Besides accounting for the dependence between subsequent dispatch intervals, existing LAOPF implementations at ISOs as well as recent works e.g. [3] consider some form of security constraints as well.

In the literature, security constraints were first considered for single dispatch interval OPF, called security constrained OPF (SCOPF) [16]. The SCOPF was initially formulated considering outages in transmission lines [17], but more recent works on SCOPF consider outages in generators or in both [18]. The model in [19] considers generation reserves (which allow recovery from generator contingencies), but does not explicitly consider any security constraints. More recently, [3] considered LAOPF with generation reserves, but without explicit treatment of security constraints.

The integration of security constraints into LAOPF results in the Look-ahead security constrained OPF (LASCOPF) problem [20] where the authors considered the DC model under the N-1 contingency criterion. To overcome the vast computational complexity, they decompose the problem into multiple SCOPF problems where a message passing algorithm is used to model the effect of ramping constraints between consecutive dispatch intervals. Their LASCOPF formulation is also used by a variety of ISOs, i.e., they enforce ramp rate constraints between successive base case dispatches, and enforce transmission contingency constraints on each base case dispatch, but they do not account for outaged equipment in one interval remaining outaged in subsequent intervals.

In this paper, we propose a LASCOPF formulation that models the entire planning horizon in the normal state as well as under generator contingencies. Our model ensures security against contingencies in any interval and accounts for the shutdown of a contingent component for the remainder of the planning horizon. Accordingly, the contributions of the this paper are twofold. First, we propose a formulation for LASCOPF using the DC power flow model considering both transmission line and generator contingencies under the N-1 contingency criterion, LASCOPF₁. This formulation suffers from high computational complexity as the number of decision variables is quadratic in the length of the planning horizon. Second, we prove analytically that the number of decision variables in the LASCOPF1 formulation can be

TABLE I
RELATED WORK ON LA/SC/OPF
M : Methodology (Analytical/Numerical), LA : Look-ahead, LC : Line
contingency, GC : Generator contingency, $N - k : N - k$ contingency



Fig. 2. Illustration of change from the normal state to the contingency state when contingency $c \in \mathcal{C}_{\mathcal{G}}$ is observed in interval $u \in \mathbb{N}, u < T$ for LASCOPF₁.

reduced, leading to LASCOPF-r₁. This formulation is scalable, with size (number of decision variables) linear in the length of the planning horizon as opposed to quadratic. We then extend our proof to the N-k contingency criterion, leading to the reduced LASCOPF- r_k . We present numerical results that demonstrate the reduction in computational complexity in practical applications. Table I provides an overview of the literature on LAOPF, SCOPF and LASCOPF, including our contribution.

The rest of this paper is organised as follows. Section III presents the LASCOPF₁ formulation. Section IV presents the reduced LASCOPF-r₁ formulation and proves its equivalence to LASCOPF₁. Section V extends the results to the N - kcontingency criterion, LASCOPF- ru_k . Sections VI and VII provide illustrative examples and numerical results, respectively. Section VIII discusses extensions of the model and numerical techniques that allow faster computation of the proposed formulations, and Section IX concludes the paper.

III. Comprehensive LASCOPF under N-1CONTINGENCY CRITERION: LASCOPF1

We begin with the LASCOPF₁ problem over a planning horizon of length T, under the N-1 contingency criterion using the DC power flow model. We assume that any transmission line contingency $b \in C_{\mathcal{L}}$ or any generator contingency $c \in C_{\mathcal{G}}$ can take place in any interval u, and a contingency is fully specified by the combination of element, $c \in \mathcal{C}_{\mathcal{G}}$ or $b \in \mathcal{C}_{\mathcal{L}}$, together with the contingency interval, u. The normal state dispatch of generator g in interval t is denoted by $p_{q,t}^{(0)}$ and the post-contingency dispatch under generator contingency $c \in \mathcal{C}_{\mathcal{G}}$, where the contingency occurred in interval u, is denoted by $p_{q,t}^{(c,u)}$ for t > u. Then, the dispatch is implemented as illustrated in Fig. 2 where it is assumed that interval t = 0 is the interval of execution of the problem, corresponding to the earliest planning horizon in Fig. 1. Given the set of contingencies as input, the LASCOPF₁($C_{\mathcal{L}}, C_{\mathcal{G}}$) problem is

$$\min_{\substack{p_{g,t}^{(0)}, p_{g,t}^{(c,u)}}} \sum_{t=1}^{T} \sum_{g \in \mathcal{G}} C_{g,t} \left(p_{g,t}^{(0)} \right), \tag{1a}$$

subject to:

(Normal state constraints:)

$$\sum_{g \in \mathcal{G}} p_{g,t}^{(0)} = \sum_{n \in \mathcal{N}} D_{n,t}^{\{0\}},$$
(1b)

$$\underline{P}_g \le p_{g,t}^{(0)} \le \overline{P}_g,\tag{1c}$$

$$\left| \sum_{n \in \mathcal{N}} H_{ln}^{\{0\}} \left(\sum_{g \in \mathcal{G}} A_{ng} p_{g,t}^{(0)} - D_{n,t}^{\{0\}} \right) \right| \le \overline{K}_l^{\{0\}}, \tag{1d}$$

$$\underline{R}_{g} \le p_{g,t}^{(0)} - p_{g,t-1}^{(0)} \le \overline{R}_{g},\tag{1e}$$

(Transmission line contingency constraints:)

$$\left| \sum_{n \in \mathcal{N}} H_{ln}^{\{b\}} \left(\sum_{g \in \mathcal{G}} A_{ng} p_{g,t}^{(0)} - D_{n,t}^{\{0\}} \right) \right| \le \overline{K}_l^{\{b\}}, \tag{1f}$$

(Generator contingency constraints:)

$$p_{g,t}^{(c,u)} = 0 \text{ if } g = c,$$
 (1g)

$$\sum_{g \in \mathcal{G}} p_{g,t}^{(c,u)} = \sum_{n \in \mathcal{N}} D_{n,t}^{\{c\}},\tag{1h}$$

$$\underline{P}_{g} \le p_{g,t}^{(c,u)} \le \overline{P}_{g} \text{ if } g \ne c, \tag{1i}$$

$$\underline{R}_{g} \le p_{g,t}^{(c,u)} - p_{g,t-1}^{(c,u)} \le \overline{R}_{g} \text{ if } g \ne c, t > u+1,$$
(1j)

$$\underline{R}_{g} \le p_{g,t}^{(c,u)} - p_{g,t-1}^{(0)} \le \overline{R}_{g} \text{ if } g \ne c, t = u+1,$$
(1k)

 $\forall t \in \mathbb{N}, t \leq T, \forall g \in \mathcal{G}, \forall l \in \mathcal{L}, \forall b \in \mathcal{C}_{\mathcal{L}}, l \neq b, \forall c \in \mathcal{C}_{\mathcal{G}}, \forall u \in \mathbb{N}, t > u.$

Objective function (1a) is the cost of generation in the normal state over T look-ahead intervals, assumed to be convex. Due to the design of the power system the probability that a contingency occurs is very low and is thus hard to estimate. Therefore, it is common practice in the literature to not consider the cost of generation under contingencies [14], [18]. For the normal state, the power balance, generation capacity, normal transmission line capacity, and ramping constraints are (1b), (1c), (1d), and (1e), respectively. In (1d), $\overline{K}_{l}^{\{0\}}$ would typically be the long term or steady state transmission line capacity. The normal state also preventively satisfies transmission line contingency constraints (1f), i.e., under a transmission line contingency, where it is conservatively assumed that the dispatch remains the same as in the normal state. Accordingly, $\overline{K}_l^{\{b\}}$ in (1f) is typically the short term or emergency transmission line capacity, where $\overline{K}_l^{\{b\}} \ge \overline{K}_l^{\{0\}} \quad \forall b \in C_{\mathcal{L}}$. The outaged transmission line may then be restored, e.g., by an automated re-closure process, without a deviation from normal dispatch. When a generator contingency takes place, the affected generator typically



Fig. 3. Illustration of change from the normal state to the contingency state when contingency $c \in C_{\mathcal{G}}$ is observed in interval $u \in \mathbb{N}, u < T$ for LASCOPF-r₁.

cannot generate for the remainder of the planning horizon as represented by constraint (1g) which would require a change in the dispatch of the remaining generators to compensate for the generation shortfall. In addition, to compensate for the loss of generation in the system, load shedding may be required as represented by the power balance constraint (1h). Here, the total shed load is typically less than the lost generation $0 \leq \sum_n \left(D_{n,t}^{\{0\}} - D_{n,t}^{\{c\}} \right) \leq p_{g,t}^{(0)}$ if g = c. Accordingly, for the generator contingency state, generation capacity, and ramping constraints are (1i), and (1j), respectively. For generator contingencies, one typically uses short term transmission line limits allowing us to ignore transmission line constraints. The corrective dispatch instruction is employed in dispatch intervals t > u (see Fig. 2). Therefore, for interval t = u + 1, the ramping constraint between the generator contingency and normal states is given by (1k).

Observe that for T = 1, LASCOPF₁ is equivalent to single interval SCOPF. Since the contingency interval satisfies $u \in$ $\mathbb{N}, u < T = 1$, i.e., $u \in \emptyset$, we cannot consider any corrective generator contingencies. Also, it is useful to note here that our formulation is different from existing formulations of LAOPF, as implemented at certain ISOs [20]. Those formulations require security with respect to a particular contingency, but they do not model that for a given contingency scenario the generator is not available for the rest of the planning horizon as we do in (1j), i.e., they only consider t = u + 1in contingency states. Therefore, they only consider ramping constraints in (1k) and ignore those in (1j). Also, they do not have to consider u explicitly since it is implicitly given by t. Our model enforces (1j), and is thus, a more comprehensive formulation of LASCOPF.

Owing to the dependence of $p_{g,t}^{(c,u)}$, t > u on the interval u in which the contingency would happen, the number of decision variables and hence constraints is proportional to T(T-1)/2 as can be estimated from Fig. 2. This $\mathcal{O}(T^2)$ dependence renders the problem infeasible for large values of T. We present a detailed analysis of the problem complexity in Appendix A.

IV. REDUCED LASCOPF: LASCOPF-r₁

To overcome the $\mathcal{O}(T^2)$ dependence of LASCOPF₁, we now propose a reduced formulation called LASCOPF-r₁($\mathcal{C}_{\mathcal{L}}, \mathcal{C}_{\mathcal{G}}$), which differs from LASCOPF₁ only in that the contingency state decision variables $p_{g,t}^{(c)}$ are independent of u. LASCOPF-r₁ is formulated as

$$\underset{p_{g,t}^{(0)}, p_{g,t}^{(c)}}{\min} \sum_{t=1}^{T} \sum_{g \in \mathcal{G}} C_{g,t} \left(p_{g,t}^{(0)} \right)$$
(2a)

subject to:

(Normal state constraints of the same form as (1b) to (1e):) (5b) to (5e),

(Transmission line contingency constraints of the same form as (1f):) (5f)

(Generator contingency constraints:)

$$p_{g,t}^{(c)} = 0 \text{ if } g = c, t > 1,$$
 (2g)

$$\sum_{g \in \mathcal{G}} p_{g,t}^{(c)} = \sum_{n \in \mathcal{N}} D_{n,t}^{\{c\}}, \text{ if } t > 1,$$
(2h)

$$\underline{P}_g \le p_{g,t}^{(c)} \le \overline{P}_g \text{ if } g \ne c, t > 1,$$
(2i)

$$\underline{R}_g \le p_{g,t}^{(c)} - p_{g,t-1}^{(c)} \le \overline{R}_g \text{ if } g \ne c, t > 2,$$
(2j)

$$\underline{R}_g \le p_{g,t}^{(c)} - p_{g,t-1}^{(0)} \le \overline{R}_g \text{ if } g \ne c, t > 1,$$

$$(2k)$$

 $\forall t \in \mathbb{N}, t \leq T, \forall g \in \mathcal{G}, \forall l \in \mathcal{L}, \forall b \in \mathcal{C}_{\mathcal{L}}, l \neq b, \forall c \in \mathcal{C}_{\mathcal{G}}.$

In LASCOPF₁, the dependence on u arises due to constraints (1j) and (1k), and the value of u determines which constraint would apply to decision variable $p_{g,t}^{(c,u)}$. In LASCOPF-r₁, the equivalent constraints are (2j) and (2k), respectively. However, since the decision variable $p_{g,t}^{(c)}$ is independent of u, it has to satisfy (2j) and (2k) simultaneously so that the same contingency dispatch can be used no matter when the contingency happens as illustrated in Fig. 3.

Note that LASCOPF₁ and LASCOPF-r₁ are identical for T = 2. To see why, observe that the contingency interval $u \in \mathbb{N}, u < T$, i.e. $u \in \{1\}$. Since contingency state decision variables exist only for a single value of u, imposing independence of u in LASCOPF-r₁ results in the same set of decision variables as that of LASCOPF₁. Thus, it is only for T > 2 that LASCOPF₁ is different from LASCOPF-r₁.

Owing to the independence of $p_{q,t}^{(c)}$ from the interval u in which the contingency would happen, the number of decision variables and constraints in LASCOPF-r1 is proportional to T, as can be estimated from Fig. 3. We present a detailed analysis of the problem complexity in Appendix A. Intuition says that as a result LASCOPF- r_1 would be more scalable than $LASCOPF_1$; this intuition is confirmed by our numerical results presented in Section VII. Also, due to the independence of the dispatch from u, LASCOPF-r₁ happens to consider the same number of contingency states as existing LAOPF formulations, and hence, it would have the same number of decision variables. The significant difference between these problem formulations is the additional consideration of (2j) in LASCOPF- r_1 . We expect that the addition of this constraint does not increase the computational complexity much while allowing for a more comprehensive consideration of contingencies.

In the event of a generator contingency, barring borderline cases, all generation levels for LASCOPF₁ would at least be equal to the normal state generation levels in the same interval, i.e., $p_{g,t}^{(c,u)} \ge p_{g,t}^{(0)} \ \forall g \in \mathcal{G}, g \neq c$, even if we account for load shedding. In what follows we show that under these conditions solving LASCOPF-r₁ is equivalent to solving LASCOPF₁.

Theorem 1. If LASCOPF- r_1 is feasible then LASCOPF₁ is feasible, and if LASCOPF₁ is feasible and has a solution such

that $p_{g,t}^{(c,u)} \ge p_{g,t}^{(0)} \ \forall c \in C_{\mathcal{G}}, \forall g \in \mathcal{G}, g \neq c, t, u \in \mathbb{N}, u < t \le T$ then LASCOPF-r₁ is feasible. Furthermore, if LASCOPF₁ has such an optimal solution then the optimal objective values of LASCOPF₁ and LASCOPF-r₁ are equal.

Proof: We begin by showing that LASCOPF-r₁ is feasible then LASCOPF₁ is feasible. To do so, observe that for LASCOPF₁ and LASCOPF-r₁ the normal state variables $\left(p_{g,t}^{(0)}|g \in \mathcal{G}, t \in \mathbb{N}, t \leq T\right)$, and the normal state constraints (1b) to (1e) and (5b) to (5e) respectively are identical. Thus, any $\left(p_{g,t}^{(0)}|g \in \mathcal{G}, t \in \mathbb{N}, t \leq T\right)$ that is feasible for LASCOPF₁ is feasible for LASCOPF-r₁ and vice versa. Also, given any $\left(p_{g,t}^{(c)}|c \in \mathcal{C}_{\mathcal{G}}, g \in \mathcal{G}, t \in \mathbb{N}, t \leq T\right)$ that is feasible for LASCOPF₁ satisfying (2g) to (2k), we can choose $p_{g,t}^{(c,u)} = p_{g,t}^{(c)} \forall c \in \mathcal{C}_{\mathcal{G}}, g \in \mathcal{G}, t, u \in \mathbb{N}, u < t \leq T$ that will be feasible for LASCOPF-r₁ satisfying (1g) to (1k).

Next, consider that LASCOPF₁ is feasible and has a solution $\text{ such that } p_{g,t}^{(c,u)} \geq p_{g,t}^{(0)} \; \forall c \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, g \neq c, t, u \in \mathbb{N}, u < 0 \}$ $t \leq T$ that satisfies (1g) to (1k). Under this condition, we prove that LASCOPF- r_1 is feasible. Also, if an optimal solution satisfies this condition then LASCOPF-r₁ has the same optimal objective value as LASCOPF₁. We begin by observing that the objective (2a) is a function only of the normal state variables $\left(p_{g,t}^{(0)}|g \in \mathcal{G}, t \in \mathbb{N}, t \leq T\right)$. Therefore, it is sufficient for the proof that given any set of normal state dispatch $\left(p_{g,t}^{(0)}|g\in\mathcal{G},t\in\mathbb{N},t\leq T
ight)$ there exists a contingency state dispatch for LASCOPF-r₁ $\left(p_{q,t}^{(c)} | c \in C_{\mathcal{G}}, g \in \mathcal{G}, t \in \mathbb{N}, t \leq T \right)$ that satisfies (2g) to (2k). To do so, we will show that given a dispatch interval t = t' and a contingency c = c', for all values of the contingency interval $u < t, u \in \mathbb{N}$, the feasible regions for the dispatch $\left(p_{g,t'}^{(c',u)}\right)_{g\in\mathcal{G}}$ must necessarily overlap with each other. Therefore, LASCOPF-r₁ can use a single decision variable $\left(p_{g,t'}^{(c')}\right)_{g\in\mathcal{G}}$ to represent the generation levels $\forall u \in \mathbb{N}, u < t'.$

First, consider dispatch interval t = 2, and a contingency c = c'. Let $\left(p_{g,2}^{(c',1)}\right)_{g \in \mathcal{G}}$ be a feasible set of dispatch during interval t = 2 for contingency c' occurring during interval u = 1. Consider now the corresponding LASCOPF-r₁ formulation, and observe that $p_{g,2}^{(c')} = p_{g,2}^{(c',1)}$ is feasible for dispatch interval t = 2 and contingency c = c', since constraints (1g) to (1i) and (1k) for u = 1, and constraints (2g) to (2i) and (2k) for c = c' define identical feasible regions for t = 2.

Let us now consider dispatch interval t = 3 and contingency c = c', and let $\begin{pmatrix} p_{g,3}^{(c',1)} \end{pmatrix}$ and $\begin{pmatrix} p_{g,3}^{(c',2)} \end{pmatrix}_{g \in \mathcal{G}}$ be feasible sets of dispatch for LASCOPF₁ for the contingency occurring during interval u = 1 and u = 2, respectively. Let \mathcal{X} be the feasible region defined by (1g) and (1i) for t = 3 and c = c'. Since constraints (1g) and (1i) are independent of u, the variables $p_{g,3}^{(c',1)} \in \mathcal{X}_g$, and $p_{g,3}^{(c',2)} \in \mathcal{X}_g$ $\forall g \in \mathcal{G}$. From (1g), $\mathcal{X}_g = \{0\}$ if g = c' and from (1i), $\mathcal{X}_g = [\underline{P}_g, \overline{P}_g]$ if $g \neq c'$. Here, $\mathcal{X} = \prod_{g \in \mathcal{G}} \mathcal{X}_g$. Also, let $\mathcal{Y} = \prod_{g \in \mathcal{G}} \mathcal{Y}_g$ be the feasible region defined by (1g) for
$$\begin{split} t &= 3, \ c = c' \ \text{and} \ u = 1 \ \text{such that} \ \mathcal{Y}_g = \mathbb{R} \ \text{if} \ g = c' \\ \text{and} \ \mathcal{Y}_g &= \left[\underline{R}_g + p_{g,2}^{(c',1)}, \overline{R}_g + p_{g,2}^{(c',1)} \right] \ \text{if} \ g \neq c'. \ \text{Similarly,} \\ \text{let} \ \mathcal{Z} &= \prod_{g \in \mathcal{G}} \mathcal{Z}_g \ \text{be the feasible region defined by (1e) for} \\ t &= 3, \ c = c' \ \text{and} \ u = 2 \ \text{such that} \ \mathcal{Z}_g = \mathbb{R} \ \text{if} \ g = c' \ \text{and} \\ \mathcal{Z}_g &= \left[\underline{R}_g + p_{g,2}^{(0)}, \overline{R}_g + p_{g,2}^{(0)} \right] \ \text{if} \ g \neq c'. \ \text{Observe that since} \\ \text{the problem is feasible, we have} \ \left(p_{g,3}^{(c',1)} \right)_{g \in \mathcal{G}} \in \mathcal{X} \cap \mathcal{Y} = \\ \{0\} \times \prod_{g \in \mathcal{G}, g \neq c'} \left[\max \left\{ \underline{P}_g, \underline{R}_g + p_{g,2}^{(c',1)} \right\}, \min \left\{ \overline{P}_g, \overline{R}_g + p_{g,2}^{(c',1)} \right\} \right] \ \text{and} \ \left(p_{g,3}^{(c',2)} \right)_{g \in \mathcal{G}} \in \mathcal{X} \cap \mathcal{Z} = \\ \{0\} \times \\ \prod_{g \in \mathcal{G}, g \neq c'} \left[\max \left\{ \underline{P}_g, \underline{R}_g + p_{g,2}^{(0)} \right\}, \min \left\{ \overline{P}_g, \overline{R}_g + p_{g,2}^{(0)} \right\} \right]. \\ \text{In the next step, we show for} \ g \neq c' \ \text{that} \ \mathcal{Y}_g \cap \mathcal{Z}_g = \\ \left[\underline{R}_g + p_{g,2}^{(c',1)}, \overline{R}_g + p_{g,2}^{(0)} \right] \neq \emptyset. \ \text{To show this, let us first} \\ \text{consider (1e) for} \ t = 2, \ \text{which is satisfied by} \ p_{g,2}^{(0)}. \ \text{After rearrangement we get} \end{aligned}$$

$$\underline{R}_g - p_{g,2}^{(0)} \le -p_{g,1}^{(0)} \le \overline{R}_g - p_{g,2}^{(0)}.$$
(3)

Consider now (1k) for t = 2 and u = 1, which is satisfied by $p_{a2}^{(c',1)}$. After rearrangement we get

$$\underline{R}_{g} - p_{g,2}^{(c',1)} \le -p_{g,1}^{(0)} \le \overline{R}_{g} - p_{g,2}^{(c',1)}.$$
(4)

Since, LASCOPF₁ is assumed to be feasible, we know that $p_{a,1}^{(0)}$ exists, and we obtain

$$\underline{R}_g + p_{g,2}^{(c',1)} \le \overline{R}_g + p_{g,2}^{(0)},\tag{5}$$

which implies $\mathcal{Y}_g \cap \mathcal{Z}_g \neq \emptyset$. Thus, $\mathcal{X} \cap \mathcal{Y} \cap \mathcal{Z} = \{0\} \times \prod_{g \in \mathcal{G}, g \neq c'} \left[\max\left\{ \underline{P}_g, \underline{R}_g + p_{g,2}^{(c',1)} \right\}, \min\left\{ \overline{P}_g, \overline{R}_g + p_{g,2}^{(0)} \right\} \right] \neq \emptyset.$

We are now ready to show that there is a dispatch $p_{g,3}^{(c')}$ for LASCOPF-r₁ that satisfies constraints (2g) and (1i) to (2k). Observe that constraints (2g) and (2i) are identical to (1g) and (1i) for t = 3 and c = c' since the latter are independent of u. Additionally, observe that (2j) is equivalent to (1j) for u = 1 since we have chosen $p_{g,2}^{(c')} = p_{g,2}^{(c',1)}$ and that (2k) is equivalent to (1k) for u = 2 since we have chosen equal values for the normal state dispatch in LASCOPF-r₁ and LASCOPF₁. Then, since $p_{g,3}^{(c')}$ must satisfy (2g) to (2k), $\left(p_{g,3}^{(c')}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Y} \cap \mathcal{Z}$. Now, observe that $\left(p_{g,3}^{(c',1)}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Y}$ satisfies (1h). This implies for the lower boundaries of $\mathcal{X}_g \cap \mathcal{Y}_g \,\forall g \in \mathcal{G}$ that $\sum_{g\in\mathcal{G},g\neq c'} \max\left\{\underline{P}_g, \underline{R}_g + p_{g,2}^{(c',1)}\right\} + 0 \leq \sum_{n\in\mathcal{N}} D_{n,3}^{\{c\}}$. Similarly, $\left(p_{g,3}^{(c',2)}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Z}$, which also satisfies (1h), implies for the upper boundaries of $\mathcal{X}_g \cap \mathcal{Z}_g \,\forall g \in \mathcal{G}$ that $\sum_{g\in\mathcal{G},g\neq c'} \min\left\{\overline{P}_g, \overline{R}_g + p_{g,2}^{(0)}\right\} + 0 \geq \sum_{n\in\mathcal{N}} D_{n,3}^{\{c\}}$. Therefore, there must exist $\left(p_{g,3}^{(c')}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Y} \cap \mathcal{Z}$ satisfying (2h), which is identical to (1h). Hence, $\left(p_{g,3}^{(c')}\right)_{g\in\mathcal{G}}$ is feasible if $\left(p_{g,3}^{(c',1)}\right)_{g\in\mathcal{G}}$ and $\left(p_{g,3}^{(c',2)}\right)_{g\in\mathcal{G}}$ are feasible. Consequently, $p_{g,3}^{(c')} = p_{g,3}^{(c',1)} = p_{g,3}^{(c',2)} \ \forall g \in \mathcal{G}.$ So far we have shown the proof for t = 2 and t = 3. We

So far we have shown the proof for t = 2 and t = 3. We can repeat the above analysis for time t = t', t' > 3 starting with t' = 4 in increasing order. First, note that we can set $\left(p_{g,t'-1}^{(c',u)}\right)_{g \in \mathcal{G}} = \left(p_{g,t'-1}^{(c')}\right)_{g \in \mathcal{G}} \quad \forall u \in \mathbb{N}, u < t'-1$. Then, for c = c' and t = t' constraints (1j) $\forall u \in \mathbb{N}, u < t'-1$ and (2j), and thus the entire feasible regions for $\left(p_{g,t'}^{(c',u)}\right)_{g \in \mathcal{G}} \quad \forall u \in \mathbb{N}$

$$\begin{split} \mathbb{N}, u < t' - 1, \text{ and } \left(p_{g,t'}^{(c')} \right)_{g \in \mathcal{G}} \text{ are identical. Then, it follows} \\ \text{that } p_{g,t'}^{(c',u)} = p_{g,t'}^{(c')} \quad \forall u \in \mathbb{N}, u < t' - 1. \text{ Now, follow-} \\ \text{ing the analysis above, we can show that } \left(p_{g,t}^{(c',u)} \right)_{g \in \mathcal{G}} = \\ \left(p_{g,t}^{(c')} \right)_{g \in \mathcal{G}} \quad \forall u \in \mathbb{N}, u < t', \text{ including } u = t' - 1. \text{ We} \\ \text{can show this } \forall c' \in \mathcal{C}_{\mathcal{G}}. \text{ This proves that feasibility of } \\ \text{LASCOPF}_1 \text{ implies feasibility of LASCOPF-r}_1 \text{ and allows us} \\ \text{to set } p_{g,t}^{(c,u)} = p_{g,t}^{(c)} \quad \forall c \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, \forall u, t \in \mathbb{N}, u < t \leq T. \\ \text{This concludes the proof.} \end{split}$$

V. EXTENSION TO k CONTINGENCIES: LASCOPF- r_k

In what follows we generalise LASCOPF- \mathbf{r}_1 to the N-k contingency criterion, i.e., the system should remain secure when up to k contingencies occur in the planning horizon. Accordingly, we include security constraints for r transmission line contingencies, (b_1, \ldots, b_r) and s generator contingencies, (c_1, \ldots, c_s) for all $r, s \ge 0, r + s \le k$. For notational simplicity, we consider that up to one generator contingency can occur in a single dispatch interval. Since security against transmission line contingencies is preventive, we could have multiple in a single interval. Then, LASCOPF- $\mathbf{r}_k(\mathcal{C}_{\mathcal{L}}, \mathcal{C}_{\mathcal{G}})$ can be written as

$$\underset{p_{g,t}^{(0)}, p_{g,t}^{(c_1, \dots, c_s)}}{\text{minimise}} \sum_{t=1}^{I} \sum_{g \in \mathcal{G}} C_{g,t} \left(p_{g,t}^{(0)} \right)$$
(6a)

subject to:

(Normal state constraints of the same form as (1b) to (1e):) (9b) to (9e),

(Transmission line contingency constraints:)

$$\left| \sum_{n \in \mathcal{N}} H_{ln}^{\{b_1, \dots, b_r\}} \left(\sum_{g \in \mathcal{G}} A_{ng} p_{g,t}^{(0)} - D_{n,t}^{\{0\}} \right) \right| \le \overline{K}_l^{\{b_1, \dots, b_r\}},$$
(6f)

(Generator contingency constraints:)

$$p_{g,t}^{(c_1,\ldots,c_s)} = 0, \text{ if } g \in \{c_1,\ldots,c_s\},$$
 (6g)

$$\sum_{q \in \mathcal{G}} p_{g,t}^{(c_1,\dots,c_s)} = \sum_{n \in \mathcal{N}} D_{n,t}^{\{c_1,\dots,c_s\}},\tag{6h}$$

$$\underline{P}_g \le p_{g,t}^{(c_1,\dots,c_s)} \le \overline{P}_g, \text{ if } g \notin \{c_1,\dots,c_s\},$$
(6i)

$$\underline{R}_g \le p_{g,t}^{(c_1,\dots,c_s)} - p_{g,t-1}^{(c_1,\dots,c_s)} \le \overline{R}_g, \text{ if } g \notin \{c_1,\dots,c_s\},$$
(6j)

$$\underline{R}_g \le p_{g,t}^{(c_1,\dots,c_s)} - p_{g,t-1}^{(c_1,\dots,c_{s-1})} \le \overline{R}_g, \text{ if } g \notin \{c_1,\dots,c_s\},$$
(6k)

extension of $H_{ln}^{\{b\}}$ [21] and only depends upon the set of contingencies $\{b_1, \ldots, b_r\}$, and not the order in which they would take place. Since the security against transmission line contingencies is preventive, the ordering of a transmission line contingency w.r.t. a generator contingency is also insignificant allowing us to completely disregard when they occur. For the generator contingency state (c_1, \ldots, c_s) , ramping constraints now have to be with state (c_0, \ldots, c_{s-1}) instead of the normal state, where c_0 represents the normal state. This makes the ordering of generator contingencies significant. Observe that for $b_r \in \mathcal{C}_{\mathcal{L}}, \mathcal{C}_{\mathcal{L}}$ will vary with the contingencies $\{b_1, \ldots, b_{r-1}\}$ that have already taken place since a contingency can only occur once in a single component and the set of lines that do not partition the system may change. Similarly, for $c_s \in C_G$, $\mathcal{C}_{\mathcal{G}}$ will vary with $\{c_1, \ldots, c_{s-1}\}$. The transmission line contingency constraints have to be considered over a combination of r contingencies and the generator contingency constraints over a permutation of s contingencies, for $r + s \le k$.

In what follows, we propose the formulation LASCOPF- ru_k , defined as LASCOPF- r_k subject to

$$p_{g,t}^{(c_1,\dots,c_s)} = p_{g,t}^{\{c_1,\dots,c_s\}} \ \forall s \in \mathbb{N}, s \le k, \forall c_1,\dots,c_s \in \mathcal{C}_{\mathcal{G}},$$
$$c_1 \ne \dots \ne c_s, \forall g \in \mathcal{G}, \forall t \in \mathbb{N}, s < t \le T. \quad (7)$$

The additional constraint requires the contingency state solution to be independent of the order in which contingencies take place. Thus, for this formulation it suffices to consider generator contingency constraints to be a combination instead of a permutation of *s* contingencies. In what follows we show that under certain conditions LASCOPF- \mathbf{r}_k is equivalent to LASCOPF- \mathbf{r}_k .

Theorem 2. If LASCOPF- ru_k is feasible then LASCOPF- r_k is feasible and if LASCOPF- r_k is feasible and has a solution such that either

then LASCOPF- ru_k is feasible. Furthermore, if LASCOPF- r_k has such an optimal solution then the optimal objective values of LASCOPF- r_k and LASCOPF- ru_k are equal.

Proof: We begin by observing that if LASCOPF-ru_k is feasible then LASCOPF-r_k must be feasible since the former simply has the additional constraint (7). Next, consider that

LASCOPF- \mathbf{r}_k is feasible and has a solution such that either (i) or (ii) satisfies (6g) to (6k). Under this condition, we will prove that LASCOPF- \mathbf{r}_k is feasible and that if an optimal solution satisfies this condition then LASCOPF- \mathbf{r}_k has the same optimal objective value. To do so, first observe that the objective (6a) is a function only of the normal state variables $\left(p_{g,t}^{(0)}|g \in \mathcal{G}, t \in \mathbb{N}, t \leq T\right)$. Since LASCOPF- \mathbf{r}_k and LASCOPF- \mathbf{r}_k only differ in the contingency state, it is sufficient to show that $\left(p_{g,t}^{(c_1,\ldots,c_s)}\right)_{g\in\mathcal{G}}$ exists satisfying (6g) to (6k) if and only if $\left(p_{g,t}^{\{c_1,\ldots,c_s\}}\right)_{g\in\mathcal{G}}$ exists such that (7) is also satisfied $\forall c_1,\ldots,c_s \in \mathcal{C}_{\mathcal{G}}, \forall t \in \mathbb{N}, s < t \leq T$. First, let us consider a feasible instance of LASCOPF- \mathbf{r}_k . It is trivial to see that if s = 1, then $p_{g,t}^{(c_1)} = p_{g,t}^{\{c_1\}} \ \forall c_1 \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, \forall t \in \mathbb{N}_0, 1 < t \leq T$.

Let us now consider s = 2 contingencies, dispatch interval t = 3, and let $\left(p_{g,3}^{(c'_1,c'_2)}\right)_{g \in \mathcal{G}}$ for contingencies $c_1 = c'_1$ and $c_2 = c'_2$, and $\left(p_{g,3}^{(c'_2,c'_1)}\right)_{g \in \mathcal{G}}$ for contingencies $c_1 = c'_2$ and $c_2 = c'_1$ be feasible sets of dispatch for LASCOPF-r_k. Without loss of generality, let us consider that $p_{g,2}^{(c'_1)} \leq p_{g,2}^{(c'_2)} \forall g \in \mathcal{G}$. Let \mathcal{X} be the feasible region defined by (6g) and (6i) for t = 2, $c_1 = c'_1$ and $c_2 = c'_2$. Since constraints (6g) and $t = 2, c_1 = c'_1$, and $c_2 = c'_2$. Since constraints (6g) and $t = 2, c_1 = c'_1$, and $c_2 = c'_2$. Since constraints (6g) and (6i) are independent of the ordering of contingencies, the variables $p_{g,3}^{(c'_1,c'_2)} \in \mathcal{X}_g$, and $p_{g,3}^{(c'_2,c'_1)} \in \mathcal{X}_g \ \forall g \in \mathcal{G}$. From (6g), $\mathcal{X}_g = \{0\}$ if $g \in \{c'_1, c'_2\}$ and from (1i), $\mathcal{X}_g = [\underline{P}_g, \overline{P}_g]$ if $g \notin \{c'_1, c'_2\}$. Here, $\mathcal{X} = \prod_{g \in \mathcal{G}} \mathcal{X}_g$. Also, let $\mathcal{Y} = \prod_{g \in \mathcal{G}} \mathcal{Y}_g$ be the feasible region defined by (6k) for $t = 3, c_1 = c'_1$ and $c_2 = c'_2$ such that $\mathcal{Y}_g = \mathbb{R}$ if $g \in \{c'_1, c'_2\}$ and $\mathcal{Y}_g = \left[\underline{R}_g + p_{g,2}^{(c'_1)}, \overline{R}_g + p_{g,2}^{(c'_1)}\right]$ if $g \notin \{c'_1, c'_2\}$. Similarly, let $\mathcal{Z} = \prod_{g \in \mathcal{G}} \mathcal{Z}_g$ be the feasible region defined by (6k) for $t = 3, c_1 = c'_2$ and $c_2 = c'_1$ such that $\mathcal{Z}_g = \mathbb{R}$ if $g \in \{c'_1, c'_2\}$ and $\mathcal{Z}_g := \left[\underline{R}_g + p_{g,2}^{(c'_2)}, \overline{R}_g + p_{g,2}^{(c'_2)}\right]$ if $g \notin \{c'_1, c'_2\}$. Observe that since the problem is feasible, we have $\left(p_{g',3}^{(c_1',c_2')}\right)_{g\in\mathcal{G}}\in$ $\begin{array}{lll} \mathcal{X} & \cap \ \mathcal{Y} & = & \{0\}^2 \ \times \ \prod_{g \in \mathcal{G}, g \notin \{c'_1, c'_2\}} \left[\max\left\{ \underline{P}_g, \underline{R}_g \right. + \\ & \left. p_{g,2}^{(c'_1)} \right\}, \min\left\{ \overline{P}_g, \overline{R}_g \right. + \left. p_{g,2}^{(c'_1)} \right\} \right] & \text{and} & \left(p_{g',3}^{(c'_2, c'_1)} \right)_{a \in \mathcal{G}} \end{array} \right. \in$ $\mathcal{X} \cap \mathcal{Z} = \{0\}^2 \times \prod_{g \in \mathcal{G}, g \notin \{c'_1, c'_2\}} \left| \max \left\{ \underline{P}_g, \underline{R}_g + \right\} \right|$ $p_{g,2}^{(c_2')}$, min $\left\{\overline{P}_g, \overline{R}_g + p_{g,2}^{(c_2')}\right\}$. he next step, we show for $g \notin \{c'_1, c'_2\}$ that $\mathcal{Y}_q \cap \mathcal{Z}_q =$

In the next step, we show for $g \notin \{c'_1, c'_2\}$ that $\mathcal{Y}_g \cap \mathcal{Z}_g = \left[\underline{R}_g + p_{g,2}^{(c'_2)}, \overline{R}_g + p_{g,2}^{(c'_1)}\right] \neq \emptyset$. To show this, let us first consider (6k) for t = 2 and $c_1 = c'_1$ which is satisfied by $p_{g,2}^{(c'_1)}$. After rearrangement we get

$$\underline{R}_{g} - p_{g,2}^{(c_{1}')} \leq -p_{g,1}^{(0)} \leq \overline{R}_{g} - p_{g,2}^{(c_{1}')}.$$
(8)

Consider now (6k) for t = 2 and $c_1 = c'_2$, which is satisfied



Fig. 4. 2-bus 2-generator system to illustrate LASCOPF.

by $p_{g,2}^{(c_2')}$. After rearrangement we get

$$\underline{R}_g - p_{g,2}^{(c_2)} \le -p_{g,1}^{(0)} \le \overline{R}_g - p_{g,2}^{(c_2)}.$$
(9)

Since, LASCOPF1 is assumed to be feasible, we know that $p_{q,1}^{(0)}$ exists, and we obtain

$$\underline{R}_g + p_{g,2}^{(c_2')} \le \overline{R}_g + p_{g,2}^{(c_1')},\tag{10}$$

which implies $\mathcal{Y}_g \cap \mathcal{Z}_g \neq \emptyset$. Thus, $\mathcal{X} \cap \mathcal{Y} \cap \mathcal{Z} = \{0\}^2 \times \prod_{g \in \mathcal{G}, g \notin \{c'_1, c'_2\}} \left[\max\left\{ \underline{P}_g, \underline{R}_g + p_{g,2}^{(c'_2)} \right\}, \min\left\{ \overline{P}_g, \overline{R}_g + p_{g,2}^{(c'_1)} \right\} \right] \neq \emptyset.$

Next, observe that $\left(p_{g,3}^{(c'_2,c'_1)}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Y}$ satisfies (6h). This implies for the lower boundaries of $\mathcal{X}_g \cap \mathcal{Y}_g \ \forall g \in \mathcal{G}$ that $\sum_{g\in\mathcal{G},g\notin\{c'_1,c'_2\}} \max\left\{\underline{P}_g,\underline{R}_g + p_{g,2}^{(c'_2)}\right\} \leq \sum_{n\in\mathcal{N}} D_{n,3}^{\{c'_1,c'_2\}}.$ Similarly, $\left(p_{g,3}^{(c'_1,c'_2)}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Z}$, which also satisfies (6h), implies for the upper boundaries of $\mathcal{X}_g \cap \mathcal{Z}_g \ \forall g \in \mathcal{G}$ that $\sum_{g\in\mathcal{G},g\notin\{c'_1,c'_2\}} \min\left\{\overline{P}_g,\overline{R}_g + p_{g,2}^{(c'_1)}\right\} \geq \sum_{n\in\mathcal{N}} D_{n,3}^{\{c'_1,c'_2\}}.$ Therefore, there must exist $\left(p_{g,3}^{\{c'_1,c'_2\}}\right)_{g\in\mathcal{G}} \in \mathcal{X} \cap \mathcal{Y} \cap \mathcal{Z}$ satisfying (6h). Consequently, (7) is feasible.

So far we have shown the proof for t = 2, and t = 3. We can repeat the above analysis for time t = t', t' > 3 starting with t' = 4 in increasing order. First, note that we can set $\begin{pmatrix} p_{g,t'-1}^{(c'_1,c'_2)} \\ p_{g,t'-1} \end{pmatrix}_{g \in \mathcal{G}} = \begin{pmatrix} p_{g,t'-1}^{(c'_2,c'_1)} \\ p_{g,t'-1} \end{pmatrix}_{g \in \mathcal{G}} = \begin{pmatrix} p_{g,t'-1}^{(c'_1,c'_2)} \\ p_{g,t'-1} \end{pmatrix}_{g \in \mathcal{G}}$. Then, observe that for t = t' (6j) is identical for the pair $c_1 = c'_1$ and $c_2 = c'_2$, and the pair $c_1 = c'_2$ and $c_2 = c'_1$. We can show this $\forall c'_1, c'_2 \in \mathcal{C}_{\mathcal{G}}$. Then, we can add one contingency at a time and repeat the above analysis, considering one pair of contingencies at a time, to set $p_{g,t}^{(c_1,\dots,c_s)} = p_{g,t}^{\{c_1,\dots,c_2\}} \forall c_1,\dots,c_s \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, \forall t \in \mathbb{N}_0, s < t \leq T$, which proves that feasibility of LASCOPF- r_k implies feasibility of LASCOPF- r_k is also feasible since the latter does not contain constraint (7). This concludes the proof.

VI. ILLUSTRATIVE EXAMPLE

In what follows, we illustrate LASCOPF on the 2-bus 2generator system shown in Fig. 4 with parameters shown in the following table.

g	$C_{g,t}(x)$	\underline{P}_{g}	\overline{P}_g	$-\underline{R}_g = \overline{R}_g$	l	$\overline{K}_l^{\{0\}}$	$H_{l1}^{(0)}$
1	x	0	100	25	1	35	0.7
2	2x	0	100	25	2	15	0.3

Furthermore, $p_{1,0}^{(0)} = p_{2,0}^{(0)} = 0$. Here, we consider $\overline{K}_l^{\{b\}} = \overline{K}_l^{\{0\}} \quad \forall b \in \mathcal{C}_{\mathcal{L}}$. We select bus 2 as the reference bus, and thus, $H_{12}^{\{0\}} = H_{22}^{\{0\}} = 0 \quad \forall l \in \{1,2\}$ [21]. The predicted demand $D_{n,t}^{\{0\}}$ is as follows.

$D_{n,t}^{\{0\}}$	t = 1	t = 2	t = 3	t = 4	t = 5
n = 1	0	0	0	0	0
n=2	10	20	30	40	50

Here, we consider $D_{n,t}^{\{c\}} = D_{n,t}^{\{0\}} \ \forall c \in C_{\mathcal{G}}$. First, we consider LAOPF (formally equivalent to

First, we consider LAOPF (formally equivalent to LASCOPF₁(\emptyset , \emptyset), i.e., with empty contingency sets) for this system with a planning horizon of T = 5 to serve as a benchmark against which to compare LASCOPF. Observe that $C_{1,t}(p) < C_{2,t}(p) \forall t \in \{1, \ldots, 5\}$, the total demand is less than \overline{P}_1 in all intervals, and the difference in demand between successive intervals is within the ramping limits of generator 1, $\underline{R}_1 \leq D_{2,t}^{\{0\}} - D_{2,t-1}^{\{0\}} \leq \overline{R}_1 \forall t \in \{1, \ldots, 5\}$. In addition, generator 2 has no minimum generation limit, $\underline{P}_2 = 0$, and the transmission line capacity constraints are not violated, $H_{l1}^{\{0\}} D_{2,t}^{\{0\}} \leq K_l^{\{0\}} \forall l \in \{1, 2\}, t \in \{1, \ldots, 5\}$. Therefore, generator 1 can serve all the demand as follows.

$p_{g,t}^{(0)}$	t = 1	t = 2	t = 3	t = 4	t = 5
g = 1	10	20	30	40	50
g=2	0	0	0	0	0

As can be seen, LAOPF favours a dispatch where the cheapest generator generates all the demand since the demand is less than its maximum generation limits.

In what follows, we consider LASCOPF under generator contingencies $C_{\mathcal{G}} = \{1, 2\}$ and no transmission line contingencies $C_{\mathcal{L}} = \emptyset$ (i.e., LASCOPF₁(\emptyset , $\{1, 2\}$)) in order to demonstrate the effect of generator contingencies on the normal state. First, we consider a planning horizon of T = 4. Since generator 1 is cheaper, it should generate as much as possible, but the solution has to satisfy the security constraints, i.e, if generator 1 had a contingency in interval u, generator 2 would have to satisfy all demand in interval u + 1. Thus, due to the ramping limit of generator 2, it always has to generate enough to ensure $D_{2,t-1}^{\{1\}} - p_{2,t-1}^{(0)} \leq \overline{R}_2$. Consequently, the solution is as follows.

$p_{g,t}^{(0)}$	t = 1	t = 2	t = 3	t = 4	t = 5
g = 1	10	15	15	40	-
g=2	0	5	15	0	-

Observe that at t = 4 there are no security constraints, allowing generator 1 to serve all demand. To summarise, the security constraints in LASCOPF₁(\emptyset , {1,2}) ensure that the more expensive generator 2 maintains a minimum generation in order for it to be able to ramp up to serve all the demand in case there was a contingency in the cheaper generator 1. This results in an increased generation cost in the normal state as compared to LAOPF, which is to be expected since the normal state faces more constraints.

In what follows, we consider LASCOPF₁(\emptyset , {1,2}) under a planning horizon of T = 5 to demonstrate how security constraints may render the problem infeasible. Observe that there is no feasible dispatch $\left(p_{1,4}^{(0)}, p_{2,4}^{(0)}\right)$ that would ensure security under a single generator contingency, i.e., for which $D_{2,5}^{\{2\}} - p_{1,4}^{(0)} \leq \overline{R}_1$ and $D_{2,5}^{\{1\}} - p_{2,4}^{(0)} \leq \overline{R}_2$. Under a contingency in either generator 1 or 2 the other generator will be unable to ramp up to meet all the demand. Thus, LASCOPF₁(\emptyset , {1,2}) for T = 5 would be infeasible.

Finally, let us also consider transmission line contingencies $C_{\mathcal{L}} = \{1, 2\}$ in order to demonstrate their effect, i.e., LASCOPF₁($\{1, 2\}, \{1, 2\}$) for a planning horizon of T = 4. Observe that transmission line contingencies, being preventive, would even apply to t = 4, unlike generator contingencies. This allows us to isolate their effect from that of generator contingencies on the normal state dispatch in interval t = 4. The security constraints require the normal state dispatch to satisfy transmission line capacity constraints if line 1 has a contingency. In this case, $H_{11}^{\{1\}} = H_{12}^{\{1\}} = H_{22}^{\{1\}} = 0$, $H_{21}^{\{1\}} = 1$ [21]. Therefore, the dispatch has to ensure that $p_{1,t}^{(0)} \leq \overline{K}_2^{(1)} \ \forall t \in \{1, \dots, 5\}$, and thus, the solution is as follows.

$p_{g,t}^{(0)}$	t = 1	t = 2	t = 3	t = 4	t = 5
g = 1	10	15	15	15	-
g=2	0	5	15	25	-

As can be seen, LASCOPF₁($\{1,2\},\{1,2\}$) has the same dispatch as LASCOPF₁($\emptyset,\{1,2\}$) up to t = 3. In t = 4, transmission line security constraints ensure that if either one of the transmission lines fails, the remaining transmission line can continue to supply power to bus 2 within its limits. Accordingly, the generation by the cheaper generator 1 should be less than the minimum of the two transmission line capacities. This increases the generation by the expensive generator 2, increasing costs. In Appendix II, we illustrate the effect of security constraints on larger systems.

VII. NUMERICAL RESULTS

First, we demonstrate for our proposed LASCOPF-r₁ the scalability for large T and its computational advantage over the LASCOPF₁ formulation for the IEEE 118-bus and the IEEE 300-bus systems [22]. For both systems, we consider the data as provided with the following modifications. We consider demand $D_{n,t}^{\{0\}} = D_{n,t}^{\{c\}} = D_n^{\text{original}} \forall c \in C_{\mathcal{G}}$, initial generation $p_{g,0}^{(0)} = p_g^{\text{original}}$, ramping limits $-\underline{R}_g = \overline{R}_g = 0.15 (\overline{P}_g - \underline{P}_g)$, $C_{\mathcal{G}} = \mathcal{G}$, and $C_{\mathcal{L}} = \mathcal{L}$. As an illustration, for the IEEE 118-bus system the difference in problem size between LASCOPF₁($\mathcal{G}, \mathcal{L}'$) and LASCOPF-r₁($\mathcal{G}, \mathcal{L}'$) when T = 26 is as follows.

	Decision	Equality	Inequality
	Variables	Constraints	Constraints
$LASCOPF_1(\mathcal{G}, \mathcal{L}')$	117819	12376	4261634
LASCOPF- $r_1(\mathcal{G}, \mathcal{L}')$	9519	976	1850534

Fig. 5 shows the computational time of the formulations as a function of the planning horizon, T using Gurobi Optimizer Version 8.1. Observe that the comprehensive formulation is infeasible to compute for large values of T in all cases and in general for the IEEE 118-bus system when transmission line constraints in [23] are included. These results show the clear advantage of the proposed reduced formulation, as it reduces the computational time by two orders of magnitude.



Fig. 5. Computational time of the comprehensive and reduced formulations for the IEEE 118-bus system and the IEEE 300-bus system.



Fig. 6. Computational time of LASCOPF $_1$, LASCOPF $_r_1$, LASCOPF $_r_2$ and LASCOPF $_ru_2$ for the 2383-bus Polish power system.

Also, observe that including transmission line constraints in the problem significantly increases the computational time. This establishes the efficiency of the proposed LASCOPF- r_1 for larger planning horizons.

Second, we consider the 2383-bus Polish power system [22] to illustrate scalability to large systems, and the computational advantage of LASCOPF-r₁ over LASCOPF₁ and of LASCOPF-ru₂ over LASCOPF₂. We consider the data as provided with demand $D_{n,t}^{\{0\}} = D_{n,t}^{\{c\}} = D_n^{\text{original}}/3 \quad \forall c \in C_{\mathcal{G}}$ for t odd and $D_{n,t}^{\{0\}} = D_{n,t}^{c} = D_n^{\text{original}}/1.5 \quad \forall c \in C_{\mathcal{G}}$ for t even, initial generation $p_{g,0}^{(0)} = p_g^{\text{original}}$, ramping limits $-\underline{R}_g = \overline{R}_g = 0.25 \quad (\overline{P}_g - \underline{P}_g), \quad C_{\mathcal{G}} = \{1, 2\}, \text{ and } C_{\mathcal{L}} = \emptyset$. As an illustration, for the Polish system the difference in problem size between LASCOPF₁($\{1, 2\}, \emptyset$) and LASCOPF-r₁($\{1, 2\}, \emptyset$) for T = 10 is as follows.

	Decision	Equality	Inequality
	Variables	Constraints	Constraints
LASCOPF ₁	32700	12376	178100
LASCOPF-r ₁	9156	46	106208

Fig. 6 shows the computational time of the formulations as a function of the planning horizon, T using MATPOWER [22] version 6.0². The results show that for LASCOPF-r₁({1,2}, \emptyset) the computational time increases linearly in T, as opposed to the quadratic trend for LASCOPF₁({1,2}, \emptyset). Thus, LASCOPF-r₁({1,2}, \emptyset) is computationally more efficient, with an increasing advantage as the length of the planning horizon increases. Similarly, for T > 2 the computational times of both LASCOPF-r₂({1,2}, \emptyset) and LASCOPF-ru₂({1,2}, \emptyset) follow a linear trend. For T = 2 all formulations have the same computational times, in accor-

²R. D. Zimmerman, C. E. Murillo-Sanchez (2016). MATPOWER (Version 6.0) [Software]. Available: https://matpower.org

dance with our discussions in Sections IV and V. For T > 2LASCOPF-ru₂({1,2}, \emptyset) is computationally more efficient than LASCOPF-r₂({1,2}, \emptyset), which confirms the efficiency of the proposed formulations for larger systems.

VIII. DISCUSSION

A. Use of Benders Decomposition to Obtain Solutions

In what follows, we discuss how Benders decomposition [24] may be used to allow us to compute $LASCOPF_1$ and LASCOPF- r_1 faster. Observe that LASCOPF₁ has a block structure where normal state decision variables can be grouped as $\left(p_{g,t}^{(0)}|g\in\mathcal{G}\right)$ $\forall t\in\mathbb{N},t< N,$ i.e., into Tblocks of size $|\mathcal{G}|$ with constraints (1b) to (1d) and (1f). Similarly, contingency state decision variables can be grouped as $\left(p_{g,t}^{(c,u)}|g \in \mathcal{G}\right) \quad \forall c \in \mathcal{C}_{\mathcal{G}} \forall u, t \in \mathbb{N}, u < t < N$, i.e., into $|\mathcal{C}_{\mathcal{G}}| \times \frac{T(T-1)}{2}$ blocks of size $|\mathcal{G}|$ with constraints (1g) to (1i). Each such group is depicted as a circle in Fig. 2. Furthermore, the objective function is decomposable into functions over individual normal state blocks. Then, it is only the ramping constraints (1e), (1j) and (1k) that couple blocks to one another as represented by arrows in Fig. 2. Taking advantage of this block structure, nested Benders decomposition can be used to compute the problem more efficiently. To see how to do so, consider interval $t'_{1} < T$. We can consider the normal state block $\left(p_{g,t'}^{(0)}|g \in \mathcal{G}\right)$ defined by t = t' to be a master problem with the subproblems being

- the normal state block $\left(p_{g,t'+1}^{(0)}|g\in\mathcal{G}\right)$ defined by t=t'+1 and
- the contingency state block $\left(p_{g,t'+1}^{(c',t')}|g \in \mathcal{G}\right)$ defined by contingency c', contingency interval u = t' and t = t'+1 $\forall c' \in C_{\mathcal{G}}$,

since these are the blocks connected to the master problem by ramping constraints. Similarly, the contingency state block $\left(p_{g,t'}^{(c',u')}|g \in \mathcal{G}\right)$ defined by contingency c', contingency interval u' < t' and t = t' can be considered to be a master problem with the subproblem being the contingency state block $\left(p_{g,t'+1}^{(c',u')}|g \in \mathcal{G}\right)$ defined by c', u = u' and t = t' + 1 $\forall c \in C_{\mathcal{G}}, u' \in \mathbb{N}, u' < T$. If this is done $\forall t' \in \mathbb{N}, t' < T$, we obtain a nested master-subproblem structure.

Similarly, we can observe a block structure in LASCOPF-r₁. However, here we will define 1 block $\left(p_{g,1}^{(0)}|g \in \mathcal{G}\right)$ of size $|\mathcal{G}|$ for interval t = 1, and $\left(p_{g,t}^{(0)}, p_{g,t}^{(c)}|c \in C_{\mathcal{G}}, g \in \mathcal{G}\right) \quad \forall t \in \mathbb{N}, 1 < t < T$, i.e., T - 1 blocks of size $(1 + |\mathcal{C}_{\mathcal{G}}|) \times |\mathcal{G}|$ consisting of the normal state and contingency states for all contingencies c. Each such block corresponds to *all* the circles for a given interval t in Fig. 3. The circles have to be grouped into blocks because a single contingency state decision variable is constrained by both ramping constraints (2j) and (2k) as represented by arrows in Fig. 3. Then, given interval t' < T the block defined by t = t' can be defined as the master problem with the subproblem being the block defined by t = t' + 1. If this is done $\forall t' \in \mathbb{N}, t' < T$, we obtain a nested master-subproblem structure and can use nested Benders decomposition for solving it.

B. Contingency filtering

In what follows, we discuss how contingency filtering [25] can be applied to the presented LASCOPF formulations under both the N-1 and N-k criteria. Observe that the set of transmission line contingencies $C_{\mathcal{L}}$ could be any subset of the generators \mathcal{L} . Similarly, the set of generator contingencies $C_{\mathcal{G}}$ could be any subset of the generators \mathcal{G} . This allows us to apply contingency filtering and consider a restricted set of only those contingencies $C_{\mathcal{L}} \subset \mathcal{L}$ and $C_{\mathcal{G}} \subset \mathcal{G}$ that are expected to be binding in any realisation of LASCOPF.

Contingency filtering may be taken a step forward by eliminating not only entire contingencies from the formulation but also individual contingency constraints that are not expected to be binding even if some other constraints deriving from the same contingency are retained [17]. E.g., given contingency c', (1i) may be eliminated but (1g) may be retained.

Note that, no matter the extent to which we perform contingency filtering, LASCOPF- r_1 will always maintain an advantage over LASCOPF1, and LASCOPF- r_k over LASCOPF- r_k . This is because we would identify the same set of entire contingencies or corresponding sets of individual constraints to be eliminated from the reduced formulations as we do for the comprehensive formulations. Therefore, the number of decision variables and hence also the overall number of constraints will remain lower in the reduced formulations.

C. Partitioning following transmission line contingencies

In what follows we will show that our results extend to transmission line contingencies that partition the system.

Consider that the system is partitioned into a set $\mathfrak{N} = \{\mathcal{N}_1, \ldots, \mathcal{N}_N\}$ of N islands, where \mathcal{N}_i is the set of buses in island *i*. Each island must satisfy the power balance constraint

$$\sum_{n \in \mathcal{N}_i} \sum_{g \in \mathcal{G}} A_{ng} p_{g,t} = \sum_{n \in \mathcal{N}_i} D_{n,t} \,\forall \mathcal{N}_i \in \mathfrak{N},\tag{11}$$

where superscripts may be added to $p_{g,t}$ and $D_{n,t}$ to distinguish the normal and particular contingency states. Therefore, transmission line contingencies that partition the system must be treated as corrective contingencies similar to generator contingencies since following a contingency power balance must be recovered in the resulting islands.

Accordingly, in order to account for system partitioning, we may simply consider $C_{\mathcal{G}} \subseteq \mathcal{G} \cup \mathcal{L} \setminus L'$. Furthermore, in LASCOPF₁, LASCOPF-r₁ and LASCOPF-r_k we may replace the power balance constraints (1h), (2h) and (6h) with (11). Observe that constraints representing generator shutdown (1g), (2g) and (6g) only apply to generator contingencies.

Under the N-1 contingency criterion, due to partitioning, the assumption that $p_{g,t}^{(c,u)} \ge p_{g,t}^{(0)} \ \forall g \in \mathcal{G}$ would not hold since a transmission line contingency would create imbalances of opposite directions in the two islands formed. However, it is reasonable to assume that in each island \mathcal{N}_i either all contingency generation levels are not less than the normal generation levels, i.e., $p_{g,t}^{(c,u)} \ge p_{g,t}^{(0)} \ \forall g \in \mathcal{G}, g \neq c, \sum_{n \in \mathcal{N}_i} A_{ng} = 1$ or are not greater than those, i.e., $p_{g,t}^{(c,u)} \le p_{g,t}^{(0)} \ \forall g \in \mathcal{G}, g \neq$ $c, \sum_{n \in \mathcal{N}_i} A_{ng} = 1$. Accordingly, to show that LASCOPF-r₁ is equivalent to solving $LASCOPF_1$, Theorem 1 can be modified as follows.

Theorem 3. If LASCOPF- r_1 is feasible then LASCOPF₁ is feasible, and if LASCOPF₁ is feasible and has a solution such that either

$$\begin{array}{ll} i) \ p_{g,t}^{(c,u)} \ge p_{g,t}^{(0)} \ \forall c \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, g \neq c, \sum_{n \in \mathcal{N}_i} A_{ng} = \\ 1, t, u \in \mathbb{N}, u < t \leq T \ or \\ ii) \ p_{g,t}^{(c,u)} \le p_{g,t}^{(0)} \ \forall c \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, g \neq c, \sum_{n \in \mathcal{N}_i} A_{ng} = \\ 1, t, u \in \mathbb{N}, u < t \leq T \end{array}$$

 $\forall \mathcal{N}_i \in \mathfrak{N}$ then LASCOPF-r₁ is feasible. Furthermore, if LASCOPF₁ has such an optimal solution then the optimal objective values of LASCOPF₁ and LASCOPF-r₁ are equal.

The proof is similar to that of Theorem 1. We note that transmission networks have sufficient redundancy by design so that partitioning would typically result from multiple contingencies.

D. Contingency Reserve Limits

In what follows we show how to derive contingency reserve limits [19] from LASCOPF₁. Since LASCOPF₁ explicitly considers every generator contingency, the contingency reserve limits are implicit in (1i). This eliminates the need for the surrogate constraint,

$$\underline{S}_{g,t} \leq p_{g,t}^{(c,u)} - p_{g,t}^{(0)} \leq \overline{S}_{g,t} \\
\forall c \in \mathcal{C}_{\mathcal{G}}, \forall g \in \mathcal{G}, g \neq c, \forall u, t \in \mathbb{N}, u < t \leq T. \quad (12)$$

where $\underline{S}_{g,t}$ and $\overline{S}_{g,t}$ represent the lower and upper contingency reserve limits respectively. Instead, we can obtain the parameters $\underline{S}_{q,t}$ (and similarly $\overline{S}_{g,t}$) from our formulation as

$$\underline{S}_{g,t} = \min\left\{0, p_{g,t}^{(c,u)} - p_{g,t}^{(0)} | c \in \mathcal{C}_{\mathcal{G}}, c \neq g, u \in \mathbb{N}, u < t\right\}$$
$$\forall g \in \mathcal{G}, \forall t \in \mathbb{N}, 1 < t \leq T. \quad (13)$$

IX. CONCLUSION AND FUTURE WORK

We considered LASCOPF under the N-1 contingency criterion over transmission line and generator contingencies. We showed that the $\mathcal{O}(T^2)$ decision variables in the comprehensive LASCOPF₁ formulation can be reduced to $\mathcal{O}(T)$ decision variables leading to the new LASCOPF-r₁ formulation, with significantly lower computational cost. We generalised the formulation to the N-k contingency criterion, LASCOPF-ru_k for which we showed that the order in which the contingencies occur can be ignored. Our evaluation of the proposed problem formulations on three IEEE benchmark systems shows that our results are an important step towards computationally efficient solutions to the LASCOPF problem.

An interesting extension of our work would be to provide analytical or numerical methods to handle the large number of variables in LASCOPF-ru_k. In particular, it is useful to employ decomposition techniques such as Benders decomposition, investigate their complexities and compare their efficiency for our reduced formulations. In addition, the theory developed in this paper using the DC power flow model has a straightforward extension to the non-linear AC power flow model. One could use existing numerical techniques to handle the non-convexity of ACOPF [18] and implement our reduced formulations in the ACOPF context.

In addition, one may extend the formulations to include costs under contingencies and reserve costs. Observe that Theorems 1 and 2 do not hold if the objective is a function of the contingency state generation levels since LASCOPF- r_1 and LASCOPF- r_k effectively impose constraints on contingency state generation levels as compared to LASCOPF₁ and LASCOPF- r_k , respectively and thereby potentially increasing the optimal objective value. Accordingly, one may investigate the differences in operational costs (including costs under contingencies and reserve costs) between LASCOPF- r_k . Based on this, for a given system one may weigh the expected operational cost against the computational efficiency for individual systems.

Appendix I

PROBLEM COMPLEXITY

A. Comprehensive Formulation: LASCOPF₁

To analyse the complexity of LASCOPF₁, we now consider the number of decision variables and constraints. For the normal state, we require $|\mathcal{G}| \times T$ decision variables, one for each generator, for each interval. Then, (1b) represents Tequality constraints, one for each dispatch interval. (1c) and (1e) represent $4 \times |\mathcal{G}| \times T$ inequality constraints, two for each decision variable. (1d) represents $2 \times |\mathcal{L}| \times T$ inequality constraints, two for each transmission line, for each interval, and (1f) represents $2 \times |\mathcal{C}_{\mathcal{L}}| \times (|\mathcal{L}| - 1) \times T$, two for each line contingency, for each remaining line, for each interval.

For the generator security constraints, we require $|C_{\mathcal{G}}| \times |\mathcal{G}| \times \frac{T(T-1)}{2}$ additional decision variables, one for each generator contingency in $C_{\mathcal{G}}$, for each generator in \mathcal{G} , for each contingency interval, u < T, for each remaining interval after the contingency interval, $u < t \leq T$. Then, (1g) and (1h) represent $|C_{\mathcal{G}}| \times T(T-1)$ equality constraints, one for each generator contingency, for each contingency interval, for each contingency interval. (1i) represents $2 \times |C_{\mathcal{G}}| \times (|\mathcal{G}|-1) \times \frac{T(T-1)}{2}$ inequality constraints, two for each generator contingency, for each contingency. (1j) and (1k) together represent $2 \times |C_{\mathcal{G}}| \times (|\mathcal{G}|-1) \times \frac{T(T-1)}{2}$ inequality constraints, two for each generator contingency, for each remaining interval, for each remaining interval. (1j) and (1k) together represent $2 \times |C_{\mathcal{G}}| \times (|\mathcal{G}|-1) \times \frac{T(T-1)}{2}$ inequality constraints, two for each generator contingency, for each remaining interval, for each remaining interval. (1j) and (1k) together represent $2 \times |C_{\mathcal{G}}| \times (|\mathcal{G}|-1) \times \frac{T(T-1)}{2}$ inequality constraints, two for each generator contingency, for each remaining interval. (1j) and (1k) together represent $2 \times |C_{\mathcal{G}}| \times (|\mathcal{G}|-1) \times \frac{T(T-1)}{2}$ inequality constraints, two for each generator contingency, for each remaining generator (i.e., $|\mathcal{G}|-1$), for each contingency interval, for each remaining interval.

- $|\mathcal{G}| \times T + |\mathcal{C}_{\mathcal{G}}| \times |\mathcal{G}| \times \frac{T(T-1)}{2}$ decision variables,
- $T + |\mathcal{C}_{\mathcal{G}}| \times T(T-1)$ equality constraints and
- $4 \times |\mathcal{G}| \times T + 2 \times |\mathcal{L}| \times T + 2 \times |\mathcal{C}_{\mathcal{L}}| \times (|\mathcal{L}| 1) \times T + 4 \times |\mathcal{C}_{\mathcal{G}}| \times (|\mathcal{G}| 1) \times \frac{T(T-1)}{2}$ inequality constraints.

Observe that the number of variables and the number of constraints increase quadratically in T which renders the problem formulation computationally infeasible for large T.

B. Reduced Formulation: LASCOPF-r₁

Contrary to LASCOPF₁, the proposed LASCOPF- r_1 formulation has decision variables for generator contingencies that

 TABLE II

 COMPLEXITY OF LP AND QP ALGORITHMS

 LP : Linear programming, QP : Quadratic programming

 m : number of decision variables, L : bits needed for model

		Iteration	Computional
	Туре	Complexity	Complexity
[26]	LP	$\mathcal{O}\left(m^2 \times L\right)$	$\mathcal{O}\left(m^4 \times L\right)$
[27]	LP	$\mathcal{O}\left(m^{0.5} \times L\right)$	$\mathcal{O}\left(m^{3.5} \times L^2 \times \log L \times \log \log L\right)$
[28]	QP	$\mathcal{O}\left(m^2 \times L\right)$	$\mathcal{O}\left(m^4 \times L^2 \times \log L \times \log \log L\right)$
[29]	OP	$\mathcal{O}\left(m^{0.5} \times L\right)$	$\mathcal{O}\left(m^{3.5} \times L^2 \times \log L \times \log \log L\right)$



Fig. 7. Relative generation cost with and without security constraints as a function of relative ramping limit for the IEEE 14-bus system. Generation cost is relative to the corresponding minimum cost in LAOPF and ramping limit is relative to generation range.

are independent of u. Therefore, we only require one variable for each generator contingency, for each generator, for each dispatch interval in which a contingency could be realised, $1 < t \leq T$, amounting to $|\mathcal{G}| \times |\mathcal{C}_{\mathcal{G}}| \times (T-1)$ decision variables. Accordingly, the number of constraints in (2g) to (2i) on these variables follow suit. However, contrary to LASCOPF₁, these variables need to satisfy both, (2j) and (2k). Therefore, LASCOPF-r₁ requires

- $|\mathcal{G}| \times T + |\mathcal{G}| \times |\mathcal{C}_{\mathcal{G}}| \times (T-1)$ decision variables,
- $T + 2 \times |\mathcal{C}_{\mathcal{G}}| \times (T-1)$ equality constraints and
- $4 \times |\mathcal{G}| \times T + 2 \times |\mathcal{L}| \times T + 2 \times |\mathcal{C}_{\mathcal{L}}| \times (|\mathcal{L}| 1) \times T + 6 \times |\mathcal{C}_{\mathcal{G}}| \times (|\mathcal{G}| 1) \times (T 1)$ inequality constraints.

Importantly, the number of variables and constraints in the reduced formulation scales linearly with T.

In Table II we survey some state-of-the-art algorithms to solve linear and quadratic programming problems. Common to all the algorithms is that complexity is increasing in both number of decision variables m and the number of bits needed to model the problem L. Since LASCOPF-r₁ has both m and L following $\mathcal{O}(T)$ as opposed to LASCOPF₁ which follows $\mathcal{O}(T^2)$, it is expected to be more computationally tractable, as we have demonstrated in Section VII.

APPENDIX II

ILLUSTRATION OF LASCOPF ON LARGER SYSTEMS

In this appendix we compare the proposed LASCOPF₁(\mathcal{L}, \mathcal{G}) formulations to LAOPF (mathematically equivalent to LASCOPF₁(\emptyset, \emptyset)) for common benchmark systems: the IEEE 14-bus, the IEEE 118-bus, and the IEEE 300-bus systems in [30].

Fig. 7 shows the normalised total cost and the cost in a single interval as a function of the ramping limit for the IEEE 14-bus system obtained using LAOPF and LASCOPF₁ for $0 \le -\underline{R}_g = \overline{R}_g \le P_g - \underline{P}_g$. We use T = 26 dispatch intervals,



Fig. 8. Generator dispatch with (left bar) and without (right bar) security constraints for the IEEE 14-bus system over a planning horizon.



Fig. 9. Relative generation cost with and without security constraints as a function of relative ramping limit for the IEEE 118-bus (left) and IEEE 300bus (right) systems. Generation cost is relative to the corresponding minimum cost in LAOPF and ramping limit is relative to generation range.

and demand $D_{n,t}^{\{0\}} = D_{n,t}^{\{c\}} = (1 + 0.23 \sin \frac{t\pi}{24}) \times D_n^{\text{original}} \forall c \in C_{\mathcal{G}}, p_{g,0}^{(0)} = p_g^{\text{original}}, C_{\mathcal{G}} = \mathcal{G}, C_{\mathcal{L}} = \mathcal{L}$. In addition to the total cost we show the cost in interval 2, which has the highest difference in cost between LAOPF and LASCOPF₁ for $-\underline{R}_g = \overline{R}_g = 0.498$, the point at which LASCOPF₁ becomes feasible. Observe that the increase in cost is less than 0.1% indicating a low cost of security. For LAOPF the curves of the total cost and cost in interval 2 intersect indicating that for lower values of \overline{R}_g , there are other intervals that have a larger relative cost. Also, since the curves for LAOPF and LASCOPF₁ meet, it indicates that transmission line security constraints are not binding for high ramping limits.

Fig. 8 shows, for the IEEE 14-bus system, the dispatch in every interval for LAOPF and LASCOPF₁ when $-\underline{R}_g = \overline{R}_g = 0.498$. Observe that the dispatches in intervals 1 to 10 are different in both cases, as the expensive generators have a higher generation in LASCOPF₁. Comparing intervals 1 and 13, we can observe that the dispatch for the same demand is different for LASCOPF₁. This is because, at interval 13, the demand is decreasing and following a contingency, the cheapest generator can alter generation within ramping constraints. Nonetheless, the dispatch is identical for LAOPF since the increase at interval 1 equals the decrease at interval 13 and $-\underline{R}_g = \overline{R}_g$. This is also confirmed from Fig. 7 where the flat curve indicates that ramping constraints are not binding. Interval 2 has the highest difference in cost due to a combination of high absolute demand and rate of increase in demand.

Fig. 9 compares costs of LAOPF and LASCOPF₁ for the IEEE 118-bus with transmission line constraints as in [23], and the IEEE 300-bus systems. We used the same method as for the IEEE 14-bus system. For the IEEE 118-bus system

transmission line constraints are binding only when their limits are decreased to about 1% of their value reported in [23]. Also, the relative cost for LAOPF for a single interval does not decrease monotonically with the relative ramping limits. This happens because LAOPF minimises the total cost and not the cost in a single interval, underlining the importance of the look-ahead framework. For the IEEE 300-bus system, interval 6 has the highest difference in cost since the total demand is a large fraction of the total generation making ramping constraints in interval 6, which has a large demand, most binding.

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