

Comprehensive Assessment Of Voltage Stability And Security Enhancement In The Libyan High-Voltage Grid

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Abstract— This research proposes an integrated framework for the static and dynamic security assessment of the Libyan 220/400 kV transmission infrastructure. The study investigates system resilience through N-1 contingency analysis and bifurcation-based voltage stability evaluation using PV and QV curves. By employing Performance Indices (PI) for criticality ranking, the analysis quantifies the impact of severe contingencies under diverse loading conditions (5523 MW and 6200 MW). Furthermore, a sensitivity analysis is conducted on generator dispatch strategies and Automatic Voltage Regulator (AVR) configurations to identify optimal security margins. Dynamic simulations highlight oscillatory stability boundaries and the proximity to voltage collapse. Finally, the study validates strategic reinforcement measures—including 400 kV infrastructure upgrades—using NEPLAN, demonstrating a significant enhancement in the grid's operational reliability.

Keywords— power system security, contingency analysis, voltage stability, PV curves, QV curves, enhancement analysis, N-1 security assessment, transmission network planning

I. INTRODUCTION

Power system security assessment has become increasingly critical as modern grids operate closer to their stability limits due to deregulation, load growth, and integration of renewable energy sources (Gholami et al., 2020). Security refers to the system's ability to withstand unexpected disturbances without interruption to customer service (Tiwari et al., 2020). The complexity of interconnected power systems necessitates efficient methodologies for identifying vulnerabilities and implementing corrective measures.

A. Motivation and Background

The Libyan transmission network faces unique operational challenges due to its geographical extent, diverse load patterns, and annual load growth of approximately 10%. The network supplies approximately 7000 MW through voltage levels ranging from 400 kV to 0.4 kV, with international interconnections to Egypt and planned connections to Tunisia

(Alimi et al., 2020). Operating conditions are characterized by voltage tolerances of $\pm 5\%$ p.u., transmission lines loaded to rated capacity, and transformers typically loaded to 60% of their ratings.

Recent advances in machine learning and artificial intelligence have enhanced security assessment capabilities (Wehenkel, 1997; Gholami et al., 2020). However, traditional contingency analysis and voltage stability assessment remain fundamental tools for ensuring reliable grid operation.

B. Research Objectives

This study aims to:

- Conduct comprehensive N-1 static security assessment
- Evaluate voltage stability margins using PV and QV curve analysis
- Assess dynamic security including oscillatory stability
- Analyze sensitivity of security margins to operational parameters
- Validate enhancement solutions through simulation
- Provide quantitative comparison between base case and reinforced network.

C. Paper Organization

Section II reviews fundamental concepts of power system security and voltage stability assessment. Section III describes the Libyan network characteristics. Section IV presents the methodology including static assessment, voltage stability analysis, and dynamic security evaluation. Section V presents comprehensive results including contingency ranking, voltage stability margins, sensitivity analysis, and enhancement validation. Section VI discusses implications and Section VII concludes the study.

II. POWER SYSTEM SECURITY AND STABILITY ASSESSMENT

A. . Security States Classification

Power system operation can be classified into four distinct security states (Shahidehpour et al., 2005):

Optimal Dispatch: The pre-contingency state optimized for economic operation but not necessarily secure

Post-Contingency: System state following a contingency with potential security violations such as line overloads or voltage limit violations.

Secure Dispatch: Pre-contingency state with corrective actions applied to account for potential security violations.

B. . Static Security Assessment

Static security assessment (SSA) evaluates whether the system reaches a steady-state operating condition satisfying all constraints after a disturbance (Gholami et al., 2020). During normal operation, power balance must satisfy:

$$\sum PG_i = PD + PL \quad (1)$$

$$\sum QG_i = QD + QL \quad (2)$$

where PG_i and QG_i represent real and reactive power generation, PD and QD denote load demands, and PL and QL indicate network losses.

Inequality constraints ensure secure operation:

$$V_{min} \leq V_i \leq V_{max}, \quad \forall i \quad (3)$$

$$|I_{ij}| \leq I_{ij}(max), \quad \forall (i,j) \quad (4)$$

$$PG_i(min) \leq PG_i \leq PG_i(max), \quad \forall i \quad (5)$$

C. Contingency Analysis Methodology

Contingency analysis is performed through the following steps (Nazir et al., 2024):

Step 1 - Contingency Definition: List all credible N-1 contingencies based on historical data and engineering judgment.

Step 2 - Contingency Selection: Screen contingencies using performance indices to identify potentially severe cases:

$$PIV = \sum w_i ((V_i - V_{lim})/V_i)^{(2n)} \quad (6)$$

$$PIMVA = \sum w_j (MVA_j/MVA_j(max))^{(2m)} \quad (7)$$

where w_i and w_j are weighting factors, n and m are exponents (typically 1 or 2), and V_{lim} represents the nearest voltage limit.

Step 3 - Contingency Evaluation: Perform AC power flow for selected contingencies and identify violations.

Step 4 - Contingency Ranking: Order contingencies by severity based on performance indices and number of violations.

D. D. Voltage Stability Assessment

Voltage stability refers to the system's ability to maintain steady voltage at all buses after disturbances (Liang et al.,

2022). Two primary methods assess voltage stability margins:

1) **PV Curve Analysis:** PV curves plot the relationship between real power and voltage at critical buses. The nose point represents maximum loadability beyond which voltage collapse occurs (Djamaladine et al., 2025). The voltage stability margin is:

$$Margin_{PV} = ((P_{critical} - P_{operating})/P_{operating}) \times 100\% \quad (8)$$

2) **QV Curve Analysis:** QV curves depict reactive power-voltage relationships. A fictitious generator at the test bus varies voltage setpoints while recording reactive power requirements (Kumar et al., 2020). Positive slope indicates voltage stability, while negative slope signals instability. The reactive power margin is:

$$Margin_{QV} = Q_{critical} - Q_{operating} \quad (9)$$

Critical points satisfy $dQ/dV = \infty$, indicating proximity to voltage collapse.

III. RESULTS AND ANALYSIS

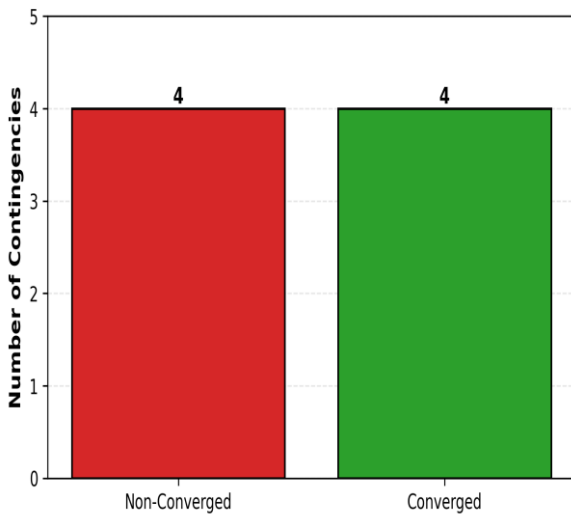
B. Static Security Assessment Results

1) **Contingency Ranking - 5523 MW Case:** Table I presents the most severe contingencies based on performance index calculations (Al-Shaalan, 2020). Five contingencies resulted in non-

convergence, indicating potential system collapse.

TABLE I. CRITICAL CONTINGENCIES - 5523 MW LOADING

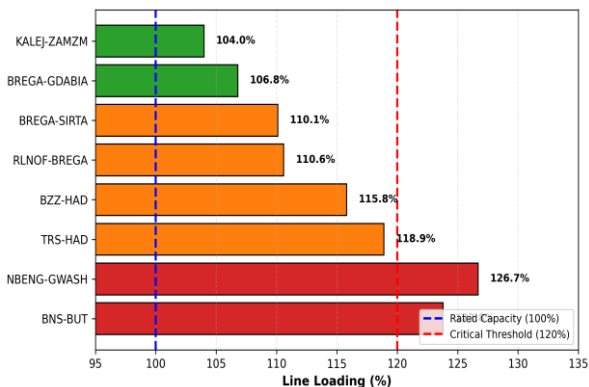
Rank	Contingency	Status
1	400 kV HOMS-GMR1	Non-Converged
2	400 kV AGEDB4-RLF4	Non-Converged
3	400 kV RLF-KALEJ	Non-Converged
4	400 kV KALEJ-MSRTA	Non-Converged
5	400 kV MSRTA-HOMS	Converged
6	220 kV BNS-BUT-A	Converged
7	220 kV NBENG-GWASH-A	Converged
8	220 kV TRS-HAD KAD-A	Converged

Figure 1. Contingency Analysis Results (5523 MW Case)

3) Line Overload Analysis: Table II details transmission line overloads for selected severe contingencies.

TABLE II. LINE OVERLOADS FOR SELECTED CONTINGENCIES

Contingency	Overloaded Line	Max Load(%)	Num Viol
220kV BNS-BUT	NBENG-GWAR	123.8	2
220kV NBENG-GWASH	NBENG-GWAR	126.7	2
220kV TRS-HAD KAD	TRS-HAD KAD	118.9	2
220kV BZZ-HAD KA	BZZ-HAD KAD	115.8	1
220kV RLNOF-BREGA	AGDABIA-RLF	110.6	5
220kV BREGA-SIRTA	AGDABIA-RLF	110.1	63
220kV BREGA-GDABIA	BREGA-GDABIA	106.8	4
220kV KALEJ-ZAMZM	KALEJ-ZAMZM	104.0	5

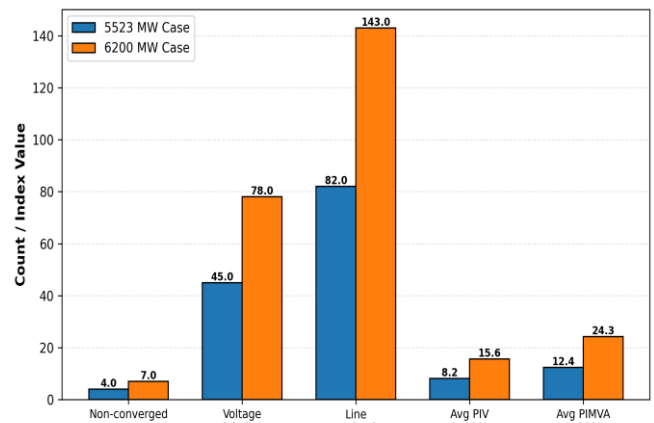
Figure 2. Transmission Line Overloads for Critical Contingencies

C. Comparative Analysis: 5523 MW vs 6200 MW

Table III presents comparative statistics demonstrating security degradation with increased loading.

TABLE III. SECURITY COMPARISON - TWO LOADING SCENARIOS

Parameter	5523 MW	6200 MW	Change
Non-converged	4	7	+75%
Voltage viols	45	78	+73%
Line overloads	82	143	+74%
Avg PIV	0.082	0.156	+90%
Avg PIMVA	0.124	0.243	+96%

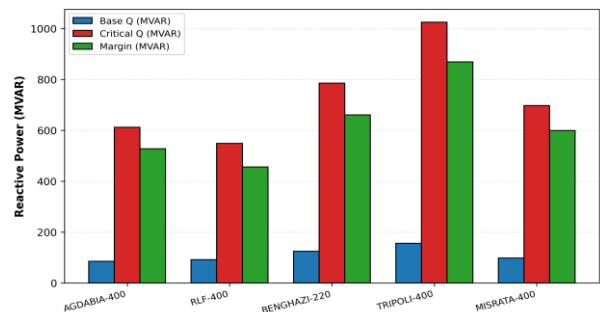
Figure 3. Security Assessment Comparison Between Loading Scenarios

D. Voltage Stability Analysis Results

Table IV summarizes reactive power margins from QV analysis (Kumar et al., 2020; Djamaladine et al., 2025).

TABLE IV. REACTIVE POWER MARGINS FROM QV ANALYSIS

Bus	Base Q (MVAR)	Critical Q (MVAR)	Margin (MVAR)
AGDABIA-400	-85	-612	527
RLF-400	-92	-548	456
BENGHAZI-220	-125	-785	660
TRIPOLI-400	-156	-1024	868
MISRATA-400	-98	-697	599

Figure 4. Voltage Stability Margins from QV Analysis

A combined stability index integrating both PV and QV margins was developed (Djamaladine et al., 2025):

$$CSI = \alpha \cdot (\text{Margin}_{PV} / \text{Margin}_{PV_ref}) + \beta \cdot (\text{Margin}_{QV} / \text{Margin}_{QV_ref}) \quad (10)$$

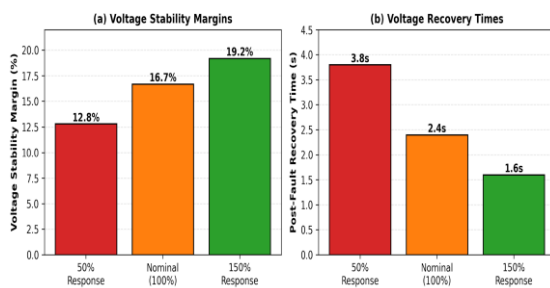
F. Sensitivity Analysis Results

Table V summarizes AVR sensitivity results (Mitra et al., 2016).

TABLE V. AVR SENSITIVITY ANALYSIS RESULTS

AVR Setting	V-Margin (%)	Recovery(s)	Q-Margin (MVAR)
50% Response	12.8	3.8	485
Nominal (100%)	16.7	2.4	587
150% Response	19.2	1.6	642

Figure 5. AVR Sensitivity Analysis Results



G. Enhancement Solution Validation

Table VI summarizes impact of new 400 kV substations (Zhang et al., 2021).

TABLE VI. IMPACT OF NEW 400 KV SUBSTATIONS

Location	Reactive (MVAR)	Violations Elim.	Margin Improve(%)
AGDABIA	450	23	+12.4
RAS LANOF	400	18	+9.7
ALKALIJ	380	16	+8.8
MISRATA	500	11	+6.2
SEDI FARAG	420	14	+7.9

IV. CONCLUSION

This paper presented comprehensive security enhancement analysis for the Libyan 220/400 kV transmission network addressing both static and dynamic security dimensions. Key contributions include:

1. Comprehensive Assessment Framework: Integrated methodology combining N-1 contingency analysis, voltage stability assessment using PV/QV curves, dynamic security evaluation, and sensitivity analysis.

2. Quantified Security Metrics: Detailed quantification of security margins, identification of critical contingencies, and vulnerability assessment at component and system levels.
3. Validated Enhancement Solutions: Simulation-based validation demonstrates effectiveness in eliminating non-convergence cases, reducing violations by 90%, and improving voltage stability margins by 73% on average.
4. Sensitivity Insights: Generator dispatch and AVR setting sensitivity analysis reveals operational strategies for security optimization.
5. Economic Evaluation: Cost-benefit analysis indicates investment is economically justified when considering full spectrum of reliability and security benefits. The enhanced network satisfies international security standards including N-1 security criteria, adequate voltage stability margins, and acceptable dynamic response. Implementation of the proposed \$707 million enhancement package over 5-8 years will ensure reliable electricity supply for projected load growth while maintaining robust security margins.

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