#### MULTI-PERIOD OPTIMAL POWER FLOW

#### Ву

#### ARYAN RITWAJEET JHA

A dissertation submitted in partial fulfillment of the requirements for the degree of

#### DOCTOR OF PHILOSOPHY

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| То | the | Faculty | of | Washington | State | University | : |
|----|-----|---------|----|------------|-------|------------|---|
|    |     |         |    |            |       |            |   |

The members of the Committee appointed to examine the dissertation of ARYAN RIT-WAJEET JHA find it satisfactory and recommend that it be accepted.

| Anamika Dubey, Ph.D., Chair |
|-----------------------------|
|                             |
|                             |
|                             |
| Committee Member A, Ph.D.   |
|                             |
|                             |
|                             |
| Committee Member C, Ph.D.   |

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TBA

### MULTI-PERIOD OPTIMAL POWER FLOW

### Abstract

by Aryan Ritwajeet Jha, Ph.D. Washington State University May 2023

: Anamika Dubey

TBA

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### Dedication

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# Chapter One

## SOME FORMATTING EXAMPLES

1.1 Chapter one tittle section

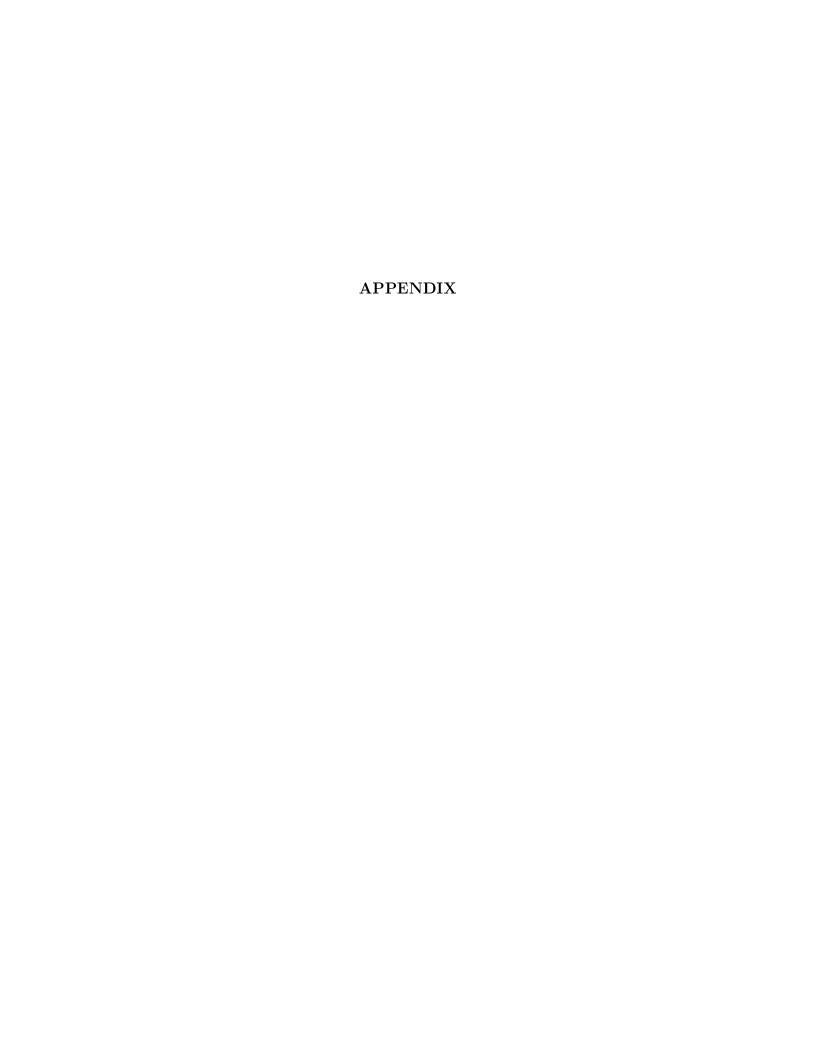
# Chapter Two

## FIGURES AND TABLES

2.1 Examples of a figure

### REFERENCES

- [1] Aayushya Agarwal and Larry Pileggi. "Large Scale Multi-Period Optimal Power Flow With Energy Storage Systems Using Differential Dynamic Programming". In: *IEEE Trans. Power Syst.* 37.3 (Sept. 2021), pp. 1750–1759. ISSN: 1558-0679. DOI: 10.1109/TPWRS.2021.3115636.
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## Appendix A

### Branch Flow Model: Relaxations and

### Convexification

In [2] the authors came up the Relaxed Branch Flow Model, and showed that in the case of Tree/Radial networks, the Relaxed Model can solve for the unique optimal solution, including the bus angles, and in the case of weakly meshed networks, there is a mechanism for extracting the bus angles from the relaxed solution, to find out its unique solution, if it exists.

#### Legend for Table A.2:

Table A.1 Table describing the variables involved in the Branch Flow Model equations.

| Symbol           | Meaning   |
|------------------|---|
| $p_j, q_j$       | Real, Reactive Power flowing from bus $j$ into the network.     |
| $P_{ij}, Q_{ij}$ | Real, Reactive Power flowing in branch $i \to j$ (sending-end). |
| $I_{ij}, l_{ij}$ | Complex Current flowing in branch                               |

Table A.2 Table describing the Branch Flow Model equations.

| Equation $\#$ | Equation   | Unknowns          | Knowns                           | No. of Equations |
|---------------|--|-------------------|----------------------------------|------------------|
|               | $p_j = \sum P_{jk} + \sum (P_{ij} - r_{ij}l_{ij}) + g_j v_j$               | $1 \times p_0$    | $n \times p_j$                   | (n+1)            |
| 13            |  | $m \times P_{ij}$ | $m \times r_{ij}$                |                  |
| 10            |  | $m \times l_{ij}$ | $(n+1) \times g_j$               |                  |
|               |  | $n \times v_j$    | $1 \times v_0$                   |                  |
|               | $q_j = \Sigma Q_{jk} + \Sigma (Q_{ij} - x_{ij}l_{ij}) + b_j v_j$           | $1 \times q_0$    | $n \times q_j$                   | (n+1)            |
| 14            |  | $m \times Q_{ij}$ | $m \times x_{ij}$                |                  |
| 14            |  | $m \times l_{ij}$ | $(n+1) \times b_j$               |                  |
|               |  | $n \times v_j$    | $1 \times v_0$                   |                  |
|               |  | $m \times P_{ij}$ | ,                                |                  |
| 1.5           |  | $m \times Q_{ij}$ | $b \times r_{ij}$                |                  |
| 15            | $v_j = v_i + (r_{ij}^2 + x_{ij}^2)l_{ij} - 2(r_{ij}P_{ij} + x_{ij}Q_{ij})$ | $m \times l_{ij}$ | $m \times x_{ij}$ $1 \times v_0$ | m                |
|               |  | $n \times v_j$    |                                  |                  |
|               | $l_{ij} = \frac{P_{ij}^2 + Q_{ij}^2}{v_j}$                                 | $m \times P_{ij}$ | $1 \times v_0$                   | m                |
| 16            |  | $m \times Q_{ij}$ |                                  |                  |
| 10            |  | $m \times l_{ij}$ |                                  |                  |
|               |  | $n \times v_j$    |                                  |                  |
|               |  | 1                 | $n \times p_j$                   |                  |
|               |  | $1 \times p_0$    | $n \times q_j$                   |                  |
|               |  | $1 \times q_0$    | $m \times r_{ij}$                |                  |
| 13 to 16      |  | $m \times P_{ij}$ | $m \times x_{ij}$                | 2(n+1+m)         |
|               |  | $m \times Q_{ij}$ | $(n+1) \times g_j$               |                  |
|               |  | $m \times l_{ij}$ | $(n+1) \times b_j$               |                  |
|               |  | $n \times v_j$    | $1 \times v_0$                   |                  |
|               |  | 2(n+1+m)          | 4n + 2m + 3                      | 2(n+1+m)         |

### Appendix B

Abstracts: Optimization-based Methods

for solving MP-OPF

In [4], the authors use a two-step paradigm for solving the MP-OPF problem, by first solving for a more relaxed SOCP problem for the all of the time-steps in a horizon, and using the SOC values from its solution, solve for the NLP OPF problem for every time-step in parallel.

In [3], they prove that for 'realistic' systems, appending an additional 'complementarity' cost function to the original objective function, Simultaneous Charging and Discharging (SCD) in the optimal solution is avoided, and that the Mixed-Integer SOCP problem of AC-OPF with energy storage can be relaxed into a regular SOCP problem without violating the battery physics in the optimal solution.

# Appendix C

**Abstracts: Dynamic Programming** 

Methods for solving MP-OPF

In [1], the authors use a Differential Dynamic Programming approach, which involved usage of Forward and Backward passes made over a sequence of time-steps, doing a back-and-forth between computation for one time-step, say t, by making some assumptions on any variables required from the next time-step t+1, and then updating the assumed values at t, once new values for the t+1 time-step have been made.